Causes and Consequences of Recombination Rate Variation

by

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Dissertation submitted in partial fulfillment of the requirements for the degree of Doctor of Philosophy in the Department of Biology in the Graduate School of Duke University

2014
ABSTRACT

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Abstract

Recombination is the process in which genetic material is exchanged between one's homologous chromosome pairs during egg or sperm development (meiosis). Recombination is necessary for proper segregation of chromosomes during meiosis, and also plays a role in purging deleterious mutations, accelerating adaptation, and influencing the distribution of genomic features over evolutionary time. While recombination is clearly an important process, recombination rate is known to vary within and between individuals, populations, and species. Furthermore, what causes this variation remains relatively unknown. Using empirical and sequenced based estimates of recombination rate for the closely related species *Drosophila pseudoobscura* and *Drosophila miranda*, I seek to understand where recombination happens across the genome, to what extent recombination changes between species, and what genomic features are responsible for these changes. These data will deepen our understanding of mechanisms determining the recombination landscape, and shed light on generalized patterns and exceptions of recombination rate variation across the tree of life.
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1. Recombination rate variation in closely related species

Caiti Smukowski and Mohamed Noor


Introduction

One of the main goals of evolutionary genetics is to understand how variation is created and maintained within and between species. Homologous meiotic recombination affects variation in the genome: differences in the rate of recombination determine the amount of genetic variation within populations and the rate at which new combinations of alleles are introduced into populations (Brooks et al. 1986; Coop et al. 2007). Indeed, in flies and humans, variability in recombination rate explains more than 50% of variation in nucleotide heterozygosity across the genome (Nachman 2002), and recombination may shape features of the genomic landscape such as codon bias, base composition, and the distribution of repetitive elements and polymorphisms ((Charlesworth et al. 1994; Comeron et al. 1999; Duret et al. 2008), see also section “Molecular Evolutionary Consequences of Recombination Rate Variation”).

Recombination rates are also expected to mediate the effectiveness of natural and sexual selection on genome evolution because the breakdown of linkage between nucleotide sites allows sites to behave independently, permitting selection to act efficiently,
eliminate mutations, reduce genetic hitchhiking, and facilitate adaptive evolution by reducing interference between sites (Hill et al. 1966).

Examining variation in recombination rates among closely related species may provide clues as to the evolutionary forces affecting recombination, and therefore offer insight into the forces shaping the genome over time. Studies across virtually all eukaryotic kingdoms have determined that the distribution of recombination events is non-uniform. Indeed, there is variation in recombination rate across the genome; extreme rates are known as “hotspots” and “coldspots” in yeast and mammals, and reflect more of a quantitative change in other organisms. Hotspots are conventionally defined as a significant increase in recombination rate from the background recombination rate, usually ranging on orders of magnitude, that takes place in a small percentage of the genome. In humans this translates to about 80% of recombination taking place in less than 15% of the sequence (Myers et al. 2006). Although other organisms have not been assayed at the same resolution as yeast, mouse, and humans, many studies observe regions with a several fold increase above the background recombination rate. For example, in Drosophila miranda, several regions show recombination rates between 25-30 cM/Mb, several times the chromosomal average of ~5 cM/Mb (Cirulli et al. 2007; Stevison et al. 2010; McGaugh et al. 2012).

Recombination rates are variable between individuals, populations, and species; however, the causative factors underlying this variation are largely unknown. Through
the comparison of fine scale genetic maps of closely related species, it may be possible to
identify features influencing fine scale and broad scale recombination patterns, as well
as features that predict shifts in recombination landscapes between species. Such linkage
map comparisons have the potential to answer questions such as how fast
recombination rates change and how changes in recombination impact selection and
evolution in natural populations. Theoretical work on the evolution of recombination
and recombination modifiers has greatly contributed to these efforts, but this review will
primarily focus on empirical work (For review of theoretical considerations, see
(Charlesworth 1990; Barton 1995; Feldman et al. 1996; Lenormand et al. 2000; Martin et
al. 2006; Barton 2010)).

The past few years have seen remarkable progress in the development of fine
scale maps and in revealing novel modifiers of recombination rate. However, some
comparisons of recombination maps show seemingly contradictory results, particularly
in relation to conservation and divergence of recombination rates. Therefore, in this
review, we discuss conservation and divergence in homologous meiotic recombination
rate between closely related species. We begin by briefly considering sequence and
epigenetic features known to affect recombination and the molecular evolutionary
consequences of recombination rate variation. We proceed to discuss several issues
surrounding the generation and analysis of recombination maps to understand why we
would see conservation or divergence between some species and not others. Similarly,
we evaluate how constraints and regulating features could produce conflicting patterns of conservation and divergence when surveyed at different scales. Overall, we hope to highlight important questions pertaining to how recombination shapes genome evolution, and how studying closely related species can help answer them. Because the recombination literature is skewed towards humans, mice, and yeast, this review places an emphasis on these taxa, but attempts to incorporate other organisms whenever possible.

**Determinants and Correlates**

It is difficult to rigorously evaluate studies of recombination without considering various factors that may influence recombination rate. Control of recombination rate appears to be multifaceted, with molecular, environmental, and demographic factors all playing a role (for example, see (Wilfert et al. 2007), see also “Why would we see conservation of recombination rates between some species and not others?” below). Attempts to elucidate the determinants of recombination have identified numerous factors of influence, many of which appear to be conserved across eukaryotes (Lichten et al. 1995; Roeder 1997; Hassold et al. 2000; Keeney 2001; Petes 2001; Page et al. 2003). New hypotheses have emerged, for example, that variation in epigenetic features could explain variation in rates of recombination between closely related species (Myers et al. 2005; Ptak et al. 2005; Winckler et al. 2005; Myers et al. 2008). The continued study of
recombination between closely related species has the potential to illuminate more
decisive determinants and how they change over time. Here, we focus on molecular
patterns and features that appear to be associated with or impact recombination.

At the chromosome level, the prevailing pattern across organisms is that a
minimum number of crossovers must be achieved for proper segregation. In humans,
the number of crossovers is strongly correlated with the number of chromosomes, where
one crossover per chromosome ensures proper segregation at meiosis (Fledel-Alon et al.
2009). In other organisms, crossover rates range from one crossover per chromosome
arm to as many as five per chromosome (Beye et al. 2006). Animals with numerous
shorter chromosomes, like the “microchromosomes” found in many birds, tend to have
higher recombination rates, again, likely as a result of ensuring proper disjunction
(Groenen et al. 2009). Along chromosomes in many organisms, rates of recombination
tend to be higher toward the distal portions of the chromosome and low around the
centromere, perhaps because repression of meiotic recombination by centromeric
heterochromatin is also critical in proper segregation (Ellermeier et al. 2010).

At the genomic level, high recombination rates are positively and nearly
ubiquitously associated with GC content, gene density, simple repeats, transposable
elements, and a number of different sequence motifs (Thuriaux 1977; Gerton et al. 2000;
Marais et al. 2001; Marais 2003; Jensen-Seaman et al. 2004; Meunier et al. 2004; Myers et
al. 2005; Groenen et al. 2009; Wong et al. 2010). In particular, it appears that a 13-mer
A degenerate motif may be responsible for recruiting recombination events in at least 40% of human hotspots (Myers et al. 2008). This motif binds the zinc finger protein PRDM9 in humans, and allelic variation controls hotspot activity in both humans and mice (Baudat et al. 2010; Berg et al. 2010). Relatedly, the Drosophila zinc finger protein, Trade Embargo (*trem*), initiates double strand breaks and is necessary for localization of the protein Mei-P22 to discrete foci on meiotic chromosomes, some or all of which are thought to mark sites for future double strand breaks (Lake et al. 2011). Thus, zinc finger proteins and sequence motifs may be major determinants of high recombination rate locations and recombination rate intensities at these locations.

Observations of divergent hotspot locations and usage among human individuals and between humans and chimpanzees has sparked a rigorous analysis of how epigenetics is involved in meiotic hotspot determination (Myers et al. 2005; Ptak et al. 2005; Winckler et al. 2005; Neumann et al. 2006). Subsequent studies show correlations between recombination hotspots and open chromatin, numerous histone modification patterns, and DNA methylation in yeast, mice, and humans (Berchowitz et al. 2009; Buard et al. 2009; Sigurdsson et al. 2009). Of particular note is the presence of a SET-methyltransferase domain in the Prdm9 gene, which is responsible for the common chromatin feature trimethylation of lysine 4 of histone H3, or H3K4me3 (Baudat et al. 2010). H3K4me3 in yeast seems to be a prominent and pre-existing mark of active
recombination sites (Borde et al. 2009), potentially creating a link between sequence and epigenetic features affecting recombination.

Continued analysis between individuals and species will surely lead to a greater understanding of existing features and the discovery of novel ones. For example, the analysis of Prdm9 across species has already produced fascinating results. Chimpanzee PRDM9 has dramatically different predicted binding sequence than human PRDM9 and seems to be the most divergent of all orthologous zinc finger proteins (Myers et al. 2010). Furthermore, Prdm9 in other mammals shows rapid evolution with variation in zinc finger number and patterns of substitution suggestive of complex repeat shuffling (Oliver et al. 2009; Myers et al. 2010). Although there is no direct evidence that these changes have generated recombination rate differences, it is surely an intriguing area to be researched. If proven, this may provide an explanation as to how recombination hotspots are created and how they change over time (see section “Why would we see recombination rate conservation at some scales and not others?” below). Closely related species present the unique opportunity to study the evolution of features regulating and influencing recombination rate and should be central in future studies of this basic biological process.
Molecular Evolutionary Consequences of Recombination Rate Variation

Finally, recombination rate variation within and between closely related species allows evolutionary biologists to make conclusions as to whether selective or neutral forces are governing genomic landscapes. First postulated in a groundbreaking study, Begun and Aquadro (1992) found that recombination rate was positively correlated with nucleotide diversity in *Drosophila melanogaster*, but did not observe an association between recombination and *D. melanogaster*- *D. simulans* divergence. This pattern is interpreted to mean that natural selection, in particular selective sweeps and/or background selection, eliminates nucleotide variability in regions of low recombination (Smith et al. 1974; Charlesworth et al. 1993), and is supported by studies in several organisms (see Table 1). However, a similar association between recombination rate and nucleotide diversity may be predicted if recombination is mutagenic, but fewer studies have detected a correlation between recombination and nucleotide divergence between species and so have not met the prediction of the mutagenic hypothesis (but, see empirical studies (Brown et al. 1987; Brown et al. 1989; Strathern et al. 1995; Papavasiliou et al. 2000)).

Additionally, the interpretations are complicated by conflicting results in several organisms surveyed (Payseur et al. 2000; Baudry et al. 2001; Nachman 2002; Yi et al. 2004; Yi et al. 2005; Noor 2008; Tsai et al. 2010). Confounding factors that may lead to conflicting results are listed in Table 1, but there are several we would like to highlight...
here. First, conflicting results may simply reflect taxon specific mutagenicity, but this hypothesis requires more empirical work. Second, correlations of diversity or divergence to recombination rate may change according to the scale with which recombination is assayed (Bussell et al. 2006; Spencer et al. 2006; Kulathinal et al. 2008; Noor 2008; Stevison et al. 2010), making it a priority to assess these measures using fine scale recombination over varying magnitudes. Third, and most relevant to the primary topic of this review, many studies up to this point have only assayed recombination in one species of interest, assuming recombination rates are conserved. It remains unresolved as to whether the selection or mutagenic hypothesis primarily accounts for the observed pattern, but perhaps with increasing amounts of recombination and sequence data, we will be able to make firmer conclusions.

Conservation and Divergence of Recombination

Following the progressive discoveries in diverse species that recombination events are non-random across the genome, one of the most exciting and surprising findings has been the realization that recombination rates sometimes change, even within species or between closely related species. The fact that recombination rates are variable and heritable implies that recombination itself can evolve in response to natural selection (Chinnici 1971; Charlesworth et al. 1985; Otto et al. 1998). Furthermore, evidence from human recombination hotspots seems to show that this change can occur
quickly on an evolutionary timescale, with hotspots emerging and disappearing in as little as 120,000 years and certainly within the six million years human divergence from chimpanzee (Ptak et al. 2005; Winckler et al. 2005; Jeffreys et al. 2009). However, conservation of recombination between closely related species has also been detected at varying scales (see Table 2), raising many questions. (1) How does the methodology by which recombination is measured affect estimates of recombination rate? (2) Why would we see conservation of recombination rates between some species and not others? (3) Why would we see conservation at some scales and not others? (4) Finally, and perhaps most fundamentally, should we expect to see conservation between closely related species? Here we comprehensively review empirical studies that compare recombination rates between closely related species, and speculate on the answers to these questions.

How does the methodology by which recombination is measured affect estimates of recombination rate?

The construction of a recombination map can dramatically affect the estimate of recombination rate depending on the methodology employed (and associated biases). Three methods are commonly used for estimating recombination rate: Linkage Disequilibrium (LD) mapping, sperm-typing, and direct mapping using polymorphic markers (see Table 2 for examples of recombination maps made with these approaches, Table 3 for potential strengths and weaknesses). The first two methods are used
primarily with human data (but see (Guillon et al. 2002; Ptak et al. 2004; Kim et al. 2007; Arguello et al. 2010)), with the labor and resource intensive direct mapping applied more in other model organisms. Fundamentally, the major differences between these measures are (1) whether the recombination rates measured are current vs. historical, and (2) whether the recombination rates measured reflect a population average or focus on a particular individual or set of individuals.

There has been some doubt as to whether LD consistently and accurately predicts hotspots (Jeffreys et al. 2005; Reed et al. 2006). For instance, Coop et al. (2008) estimated that 40% of crossovers occurred outside of LD-predicted hotspots, but Khil et al. (2010) suspect this may be an overestimate due to the way hotspots were measured using particular populations. Khil et al. also found that 26-32% of crossovers happened outside of European population LD-predicted hotspots, however this discrepancy disappeared when hotspot locations of other populations were taken into account. There are also specific examples of LD-predicted hotspots being absent when checked with sperm typing (Kauppi et al. 2005), although such inconsistencies appear to be the exception rather than the rule (Jeffreys et al. 2000; Jeffreys et al. 2001; Yauk et al. 2003).

In pedigrees and controlled crosses, increasing sample size and marker coverage changes the way we measure and perceive recombination, similar to the “Beavis effect” for mapping (Beavis et al. 1994; Beavis 1996). For example, the original honeybee linkage map used 94-142 individuals and 365 RAPD markers for a total map length of 3450 cM
(Hunt et al. 1995); the newer map used 541 markers and 283 individuals for a total length of 4061.2 cM (Solignac et al. 2004). With more accurate technologies, the honeybee genome increased in length from 178 Mb in 1974 (Jordan et al. 1974) to 262 Mb in 2006 (The Honeybee Genome Consortium), thereby decreasing the average recombination rate from 19.38 cM/Mb to 16 cM/Mb today. Similarly, in the chicken, the current map used 9268 markers for a total length of 3228 cM (Groenen et al. 2009), substantially smaller than the 4200 cM previously estimated with 2261 markers (Schmid et al. 2005). However, obtaining enough markers to detect fine scale recombination is resource intensive, thereby producing maps that range in scale from kilobases (hereafter referred to as “fine scale”) to hundreds of kilobases (hereafter referred to as “intermediate scale”) to tens of megabases (hereafter referred to as “broad scale”) to whole genomes. This is relevant, as recombination rate conservation and divergence between species is scale dependent (see below).

Other indirect quantitative approaches also exist, such as immunostaining as used by Dumont and Payseur (2011) in Murid rodents and Double Strand Break (DSB) mapping, most commonly used in yeast (Gerton et al. 2000; Buhler et al. 2007; Mancera et al. 2008; Dumont et al. 2011). Of course, choosing an approach is constrained by the organism and resources available, and researchers must be aware of limitations when making generalizations and conclusions.
Why would we see conservation of recombination rates between some species and not others?

There are several pertinent issues to consider when comparing recombination rates between closely related species. First, differential action of selection, or selection in changing environments, could give rise to differences between species (Chinnici 1971; Charlesworth et al. 1985; True et al. 1996). For example, artificial selection may have increased recombination rates in the domesticated species of chicken, honeybee, and many plants (Rees et al. 1974; Burt et al. 1987; Otto et al. 2001; Ross-Ibarra 2004; Wilfert et al. 2007; Groenen et al. 2009) in comparison to their wild progenitors.

Second, imprecise measures of physical distance can lead to inferred differences in recombination rate. Large insertions, deletions, or inversions can affect the recombination fraction between two points in the genome because single crossover events may often lead to aneuploidy. Even if such changes do not change the actual recombination “rate” (e.g., centiMorgans per megabase), if a large insertion is present in one taxon but not another, it may appear that recombination rates have diverged between species, when in fact recombination is being measured over intervals of unequal size. Moreover, chromosomal rearrangements, such as inversions, reduce the observed amount of recombination in heterokaryotypes (Hartl and Jones 2004; pp. 319-324). This repression of recombination associated with inversions extends outside the inversion several megabases (Mb), thereby producing broader scale changes in recombination rate (Kulathinal et al. 2009). Furthermore, inversion heterozygotes exhibit
increased recombination further outside the inverted region, known as the “interchromosomal effect” (Schultz et al. 1951). Large insertions and deletions have the potential to produce comparable outcomes to inversions. In comparing the same intervals between two closely related species in the absence of a genome sequence, one runs the risk of concluding increased divergence between species when in actuality, an inversion, insertion, or deletion segregating in only one species is obscuring their comparable recombination rates.

Third, recombination is variable within individuals and populations (Brooks et al. 1986; True et al. 1996; Carrington et al. 2004; Neumann et al. 2006; Graffelman et al. 2007; Coop et al. 2008; Paigen et al. 2008; Cheng et al. 2009; Dumont et al. 2009; Kong et al. 2010). This variation may stem from actual heritable variation in recombination rates among individuals, variation within an individual among regions of its genome (as discussed above), or from environmental variation. Because of potentially dramatic within-population variation, one must use caution in making generalizations about species as a whole or presumptions that a single linkage map describes the species. Without first surveying recombination within the populations, subspecies, or species of interest, it will be impossible to conclude true conservation or divergence between groups. These potential challenges in estimating recombination rate can translate into real or perceived differences between closely related species.
Why would we see recombination rate conservation at some scales and not others?

Maps produced at varying scales provide an unexpected and seemingly contradictory view of conservation and divergence between species (see Table 2). In mice and humans, conservation of recombination rate was identified at a broad scale, but divergence of recombination hotspots was identified at a fine scale, suggesting that there may be differential regulation among the scales (Myers et al. 2005; Coop et al. 2007), or perhaps, simply that we do not understand the relationship between fine scale and broad scale recombination patterns.

Divergence of fine scale recombination rates observed in yeast and humans is likely caused by the rapid turnover of hotspots due to biased gene conversion and meiotic drive. In the process of double stranded break repair, one pathway leads to crossing over and exchanging of content between homologous chromosomes; the other pathway leads to gene conversion (Boulton et al. 1997; Marais 2003). Biased gene conversion favors one allele over the other, in which case the initiating hotspot is replaced by a copy of its homolog, effectively suppressing subsequent recombination. Simulations and empirical evidence have demonstrated the self-destructive nature of hotspots through overtransmission of recombination suppressing alleles (meiotic drive), creating what is known as “the hotspot paradox” (Boulton et al. 1997; Jeffreys et al. 2002).
Recent evidence might provide the elusive answer as to how new hotspots are created and regulated to counteract losses due to the hotspot paradox. The zinc finger protein PRDM9, confirmed to play a significant role in recombination in human and mouse, contains a zinc finger encoding region with a minisatellite structure (Baudat et al. 2010). This particular structure may confer a strong potential to generate variability by recombination or replication slippage within the array. Indeed, studies documented variability in the contact residues predicting DNA binding between human populations (Baudat et al. 2010; Berg et al. 2010; Parvanov et al. 2010), and the number of zinc fingers and their contact residues vary significantly across rodents, primates, and other Metazoans (Oliver et al. 2009). The changing of contact residues could create a new family of hotspots by the binding of the protein to new sequence motifs, thereby counteracting the loss of hotspots due to biased gene conversion. Regardless of the mechanism, it is clear by the constant flux of hotspots that the precise locations seem to be unconstrained, allowing for divergence at the hotspot level.

It is believed that broad scale recombination is controlled in a different, but potentially non-mutually exclusive manner. Above all, the broad scale rate is defined by the necessity of one crossover per chromosome to ensure proper disjunction (Hassold et al. 2004; Fledel-Alon et al. 2009). Other chromosomal properties, including size and number, are correlated with this trend. This selective constraint is likely a large influence on the conservation of broad scale rates between closely related species. At the
intermediate scale, conclusions are more vague. Regional recombination is decidedly influenced by crossover interference, where a crossover in one location prevents another crossover from occurring close by (Foss et al. 1993; Hillers 2004; Stahl et al. 2004; Copenhaver 2005). Regional properties of chromosomes have an impact as well, clearly shown by the lack of crossovers in the centromeric region and typically high number of crossovers near the telomeres. Overall, perhaps these broader scale processes are more likely to be conserved, and hence preserve recombination at this scale as a byproduct.

The connection between the different scales is also unclear. In humans, there are an estimated 60-80,000 hotspots across the genome (Khil et al. 2010), occurring in clusters every 60-90 kb with individual hotspots separated by 1-7 kb within each cluster (Jeffreys et al. 2001; Kelmenson et al. 2005) and no region greater than 200kb in which recombination is absent (Myers et al. 2006). It appears that non-hotspot recombination is relatively rare, with sperm genotyping studies showing very low levels of background, non-hotspot recombination, most likely in weaker and polymorphic hotspots (Jeffreys et al. 1998; Jeffreys et al. 2001; Jeffreys et al. 2002; Jeffreys et al. 2005; Jeffreys et al. 2005). It is proposed that larger scale variation may be a product of the varying density or intensity of these hotspots in different regions of the genome (Nachman 2002; Myers et al. 2006).

There are several hypotheses as to which selective pressures influence the overall process and distribution. Hey (2004) proposed that individual hotspots may arise as a
byproduct of linkage disequilibrium between genes that are being selected (Hey 2004). If selection favors a recombinant haplotype, then it will favor chromosomes with high recombination between the two genes, thereby producing variation in location and intensity of local recombination rates that fluctuate depending on the genes under selection and the linkage disequilibrium patterns. Kauppi et al. (2004) put forward three hypotheses; one, that the distribution of recombination is governed by a balance between the need for recombination to ensure proper segregation during meiosis and the need to minimize the breakdown of favorable haplotypes. Two and three, more mechanistically, that restrictions on crossover position potentially facilitate the optimal mechanical/biochemical function of chiasmata in chromosome segregation or restrictions on position confine double strand breaks to regions that are most conducive to efficient assembly of machinery and repair.

The answer is plausibly a combination of all of these propositions. The continued analysis of recombination rates assayed at different scales, combined with theory that could test selective constraints at these scales, will help to determine why we would see conservation in recombination rate at some scales and not others.

Should we expect to see conservation between closely related species?

Just as we should expect to see conservation in sequence between closely related species, we should expect to see conservation in recombination (Dumont et al. 2008).
Presumably, with sequence similarity comes, on average, greater sharing of features that influence recombination. Divergence should be seen as a departure from the null hypothesis, but only if there is a significant increase in divergence relative to variability of recombination within species, something rarely studied.

Organisms with resolution of recombination rate at the fine scale (e.g. human, mouse, and yeast) show that hotspots do indeed diverge over time. Other organisms with maps at an intermediate to broad scale generally portray a trend of conservation (see Table 2). From the evidence we have, we can determine that hotspots are not conserved due to their transient nature. Over long periods of time, enough changes occur on the fine scale to produce broad scale change. Additionally, on a broad scale, shared constraints will hold the rate more constant, so that rates are more conserved between closely related species. These processes are compounded by changes in the genome such as inversions, translocations, insertions, and deletions that lead to altered recombination rates over time.

Future studies should strive to confirm both these trends, and the mechanisms behind them, to better understand the impact recombination has on the genome, and which features of the genome regulate recombination. A combination of theoretical and empirical work will be necessary. Theoretically, models and simulations can inform role of selective and neutral processes and different population parameters that could lead to conservation or divergence of recombination rates. Empirically, in chimpanzees,
comparisons of human-chimpanzee recombination rates should be extended beyond isolated regions of the genome. Recombination should also be surveyed in a broader set of species, with a particular attempt to obtain a resolution of less than 10 kb in order to observe the presence or absence of hotspots, albeit this is a major challenge in non-model systems. The ability to make cross species generalizations about recombination rates will have an impact on our understanding of genome evolution, therefore implicating diverse topics such as human health, selection and neutrality in the genome, and speciation and mapping studies.

Conclusions

Several patterns emerge in reviewing data from these recombination maps. First, the labels such as conservation and divergence are somewhat misleading. No pair of species studied to date exhibits complete divergence or conservation of all studied hotspots or regions of high recombination, and furthermore, many studies report only a few intervals across the genome that can be categorized in these terms. Therefore, it is necessary to attempt to incorporate a standard, or at least explicitly defined, set of parameters when discussing divergence. Of course, the vast range of scales at which recombination has been, and continues to be, surveyed makes this difficult. Because recombination is considered a quantitative genetic trait displaying variation and heritability that can be acted upon by natural selection, it shares features with other
phenotypes and can be discussed in similar terms. A standard approach is to discuss sequence divergence as a percentage, and here we recommend recombination rate divergence to be treated in a similar manner. It is appropriate to express intervals conserved or diverged out of the total number surveyed, also citing the percentage of the total of the physical and recombinational maps that was surveyed.

With more genome sequences becoming available, the decreasing cost of genotyping, and sophisticated software and technology at our fingertips, detection of recombination can be completed with greater feasibility. These projects should be undertaken with several parameters in mind. First, recombination variation should be assayed within and across populations within species before making conclusions about differences between species. This is necessary to make conclusions regarding conservation or divergence between species, but will also provide a perspective on the speed at which recombination rate is changing. Second, genome sequences should be available for the populations or species involved, and particularly for the strains being surveyed. This data will ensure that no chromosomal rearrangements or insertions/deletions exist, and will facilitate analysis of features such as motifs and measures of diversity and divergence. Surveying diversity and divergence will also provide a better understanding of forces at work in the genome. Third, examining recombination at multiple scales will aid in the interpretation of different constraints influencing the distribution of recombination events, especially when compared
between populations and species. Finally, in interpreting results, population history and the methodology used to infer recombination should be taken into account to avoid biases and complications.

Fine scale recombination maps can help us address essential questions, such as how variation is created and maintained within and between species. Recombination shapes features of the genome and creates new allelic combinations that allow increased adaptability in all sexual organisms. However, many researchers have assumed that recombination rate is invariable among individuals and between species, which we now know is inherently false. Indeed, knowledge of fine scale variation in crossover rate is essential in modeling genome evolution, population genetics studies, genome wide association studies, and inferring evolutionary processes. Thus, results indicating how recombination rate is distributed in the genome will have implications in human health, molecular evolution, and the way we study genetics. The role of recombination in genome evolution is a fundamental issue in understanding basic biological processes, and while much progress has been made, many questions remain to be answered.
Table 1: Relationships between recombination rate and measures of diversity and divergence

<table>
<thead>
<tr>
<th>Species 1</th>
<th>Species 2</th>
<th>Time since divergence (My)</th>
<th>Relationship between Diversity and Recombination</th>
<th>Relationship between Divergence and Recombination</th>
<th>Confounding factors</th>
<th>Claims Support for Natural Selection or Mutagenic Hypothesis?</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Human</td>
<td>Chimpanzee</td>
<td>5-6</td>
<td>Positive, p = 0.06, r = 0.678</td>
<td>Positive, p = 0.006, r = 0.866</td>
<td>Only surveyed PAR1; recombination rate only surveyed in humans</td>
<td>Claims support for mutagenic hypothesis for recombination</td>
<td>(Buswell et al. 2006)</td>
</tr>
<tr>
<td>Human</td>
<td>Chimpanzee</td>
<td>5-6</td>
<td>Positive, p = 0.01, r = 0.127 (Stephens 2001), p = 0.037, r = 0.257 (SeattleSNPs), p = 0.032, r = 0.249 (NIEHS SNPs)</td>
<td>Positive, p = 0.037, r = 0.259 (Seattle), p = 10^-9, r = 0.214 (chimp shotgun), p = 4x10^-4, r = 0.289 (chimp BAC), p = 0.204, r = 0.264 (Baboon BAC)</td>
<td>Recombination rate only surveyed in humans</td>
<td>Claims support for mutagenic hypothesis for recombination</td>
<td>(Hellmann et al. 2003)</td>
</tr>
<tr>
<td>Human</td>
<td>Chimpanzee</td>
<td>5-6</td>
<td>Positive, p &lt; 0.001, r = 0.735</td>
<td>Uncorrelated, p = 0.57, r = 0.141</td>
<td>Used recombination estimates from numerous studies, which used various methods to estimate recombination and may be imprecise; used SNPs in close proximity to exons, which may be influenced by other forces; recombination rate only surveyed in humans</td>
<td>Claims support for natural selection hypothesis for recombination</td>
<td>(Nachman et al. 1998)</td>
</tr>
<tr>
<td>Human</td>
<td>Chimpanzee</td>
<td>5-6</td>
<td>Positive, p = 0.041, r = 0.726</td>
<td>Uncorrelated, p = 0.726, r = 0.726</td>
<td>Recombination rate only surveyed in humans;</td>
<td>Claims support for natural selection</td>
<td>(Nachman et al. 1998)</td>
</tr>
<tr>
<td>Species 1</td>
<td>Species 2</td>
<td>r</td>
<td>p-value 1</td>
<td>p-value 2</td>
<td>Notes</td>
<td>References</td>
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<tr>
<td><em>Mus musculus</em></td>
<td><em>Mus spicilegus</em></td>
<td>0.775</td>
<td>0.06</td>
<td>0.46</td>
<td>Uncorrelated, based on small sample size (seven data points) hypothesis for recombination in <em>M. m. domesticus</em>; used sequence data near functional genes; small sample size</td>
<td>(Takahashi et al. 2004)</td>
<td></td>
</tr>
<tr>
<td><em>Mus musculus</em></td>
<td><em>Mus caroli</em></td>
<td>0.164</td>
<td>0.65</td>
<td>0.46</td>
<td>Uncorrelated, based on small sample size (seven data points) hypothesis for recombination in <em>M. m. domesticus</em>; used sequence data near functional genes; small sample size</td>
<td>(Nachman 1997)</td>
<td></td>
</tr>
<tr>
<td><em>Saccharomyces cerevisiae</em></td>
<td><em>Saccharomyces paradoxus</em></td>
<td>0.06</td>
<td>0.5663</td>
<td>0.434</td>
<td>Uncorrelated, based on small sample size; only surveyed at four loci hypothesis for recombination; small sample size</td>
<td>(Tsai et al. 2010)</td>
<td></td>
</tr>
<tr>
<td><em>Drosophila melanogaster</em></td>
<td><em>Drosophila simulans</em></td>
<td>0.0007</td>
<td>0.983</td>
<td>2.5-3.4</td>
<td>Positive, recombination only surveyed in <em>D. melanogaster</em>; low quality estimates of recombination; only surveyed at 20 gene regions</td>
<td>(Begun et al. 1992)</td>
<td></td>
</tr>
<tr>
<td><em>Drosophila simulans</em></td>
<td><em>Drosophila melanogaster</em></td>
<td>5-20</td>
<td>NA</td>
<td>0.03</td>
<td>Positive, recombination only surveyed in <em>D. melanogaster</em>; low quality map for <em>D. melanogaster</em>; only surveyed X</td>
<td>(Begun et al. 2007)</td>
<td></td>
</tr>
<tr>
<td>Species</td>
<td>Chr.</td>
<td>Window Size</td>
<td>Evidence</td>
<td>Recombination Surveyed In</td>
<td>Hypothesis Support</td>
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<tr>
<td>D. pseudoobscura, D. persimilis</td>
<td>0.5-1</td>
<td>≤500 kb window: Intronic p = 0.0156, r = 0.488, Intergenic p = 0.0020, r = 0.588 2 Mb window: Intronic p = 0.8322, r = 0.045, Intergenic p = 0.3260, r = 0.205</td>
<td>Positive</td>
<td>Shared ancestral polymorphism; hybridization; Recombination only surveyed in D. pseudoobscura</td>
<td>Claims support for mutagenic hypothesis for recombination (Kulathinal et al. 2008)</td>
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</tr>
<tr>
<td>D. pseudoobscura, D. miranda</td>
<td>2-3</td>
<td>Intronic p = 0.004, r = 0.333, Intergenic p = 0.258, r = 0.133</td>
<td>Negative</td>
<td>Recombination only surveyed in D. pseudoobscura</td>
<td>Claims support for natural selection hypothesis for recombination (Stevision and Noor 2010)</td>
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<tr>
<td>Beta maritima (sea beet, wild relative of the sugar beet)</td>
<td>NA</td>
<td>Positive, p = 0.007, r = 0.226 (based on number of distinct alleles) p = 0.084, r = 0.117 (based on heterozygosity) *calculated using Θ</td>
<td>Negative, Not significant, r = -0.02</td>
<td>Recombination estimates for Beta vulgaris maritima come from B. vulgaris vulgaris, which may not be an accurate assumption; recombination estimates are coarse</td>
<td>Claims support for natural selection hypothesis for recombination (Kraft et al. 1998)</td>
<td></td>
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<tr>
<td>L. chilense</td>
<td>5.8-18.6</td>
<td>Positive, p = 0.089, r = 0.471 *calculated using Θ</td>
<td>Uncorrelated, p = 0.640, r = 0.159</td>
<td>Demographic processes and life history traits may confound diversity and recombination estimates; genomic recombination rate used is based on a map from L. esculentum x L. pennellii, which may not be an accurate assumption</td>
<td>Demographic processes have a strong influence on shaping patterns in the genome in combination with mutation rate and/or selective constraint (Roselius et al. 2005) see also (Stephan et al. 1998)</td>
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<tr>
<td>L. peruvianum</td>
<td>Uncorrelated, p = 0.498, r = 0.198 *calculated using Θ</td>
<td>Uncorrelated, p = 0.852, r = 0.064</td>
<td>Demographic processes and life history traits may confound diversity and recombination estimates; genomic recombination rate used is based on a map from L. esculentum x L. pennellii, which may not be an accurate assumption</td>
<td>Demographic processes have a strong influence on shaping patterns in the genome in combination with mutation rate and/or selective constraint (Roselius et al. 2005) see also (Stephan et al. 1998)</td>
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</tr>
<tr>
<td>L. chilense</td>
<td>Positive, Uncorrelated,</td>
<td>Uncorrelated,</td>
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<td></td>
</tr>
<tr>
<td>L. peruvianum</td>
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<tr>
<td><strong>hirsutum</strong></td>
<td></td>
<td></td>
<td><strong>Zea mays</strong></td>
<td><strong>Zea mays parviglumis</strong></td>
<td>7,500-9,000 years</td>
<td>Positive, ( p = 0.007, \ r = 0.65 ) (using ( 4Nc_{\text{huds}} )); Uncorrelated when recombination measured with physical measure ( R ) or ( 4Nc_{\text{huds}} );</td>
<td>Uncorrelated, NA</td>
</tr>
</tbody>
</table>
Table 2: Relationships between recombination rate and measures of diversity and divergence

<table>
<thead>
<tr>
<th>Species 1</th>
<th>Species 2</th>
<th>Time since divergence (My); Approximate Sequence Divergence (%)</th>
<th>Region of genome surveyed</th>
<th>Approximate percentage of physical genome surveyed in between species comparison (%)</th>
<th>Methodology used to measure recombination rate</th>
<th>Number of markers</th>
<th>Marker distance</th>
<th>Size of intervals compared between species</th>
<th>Sample size</th>
<th>Conservation or divergence?</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Human</td>
<td>Chimpanzee</td>
<td>5-6; 1</td>
<td>Several known hotspots; 3 500kb regions located on 4q26, 7q21, 7q31</td>
<td>&lt;1</td>
<td>Linkage disequilibrium (LD)</td>
<td>Hotspot specific</td>
<td>Hotspot specific</td>
<td>Hotspot specific</td>
<td>15/18</td>
<td>10kb windows; Divergent in 10 kb windows; Divergent in 14 additional 160 kb windows</td>
<td>(Winckler et al. 2005)</td>
</tr>
<tr>
<td>Human</td>
<td>Chimpanzee</td>
<td>5-6; 1</td>
<td>2 regions totaling 14 Mb on Chr. 21</td>
<td>&lt;1</td>
<td>Linkage disequilibrium (LD)</td>
<td>30,611 (11,642)</td>
<td>440 bp (1.2 kb)</td>
<td>20 kb</td>
<td>8 (71)</td>
<td>36/39 hotspots are divergent; Significantly, but weakly correlated at 50 kb (r = 0.276); Total recombination rates significantly, but weakly correlated (r =</td>
<td>(Ptak et al. 2005)</td>
</tr>
<tr>
<td>Species 1</td>
<td>Species 2</td>
<td>Chr.</td>
<td>SNP genotyping</td>
<td>Recombination data from</td>
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<tr>
<td>C57BL/6J x CAST/EiJ</td>
<td>8 Heterogeneous stock mice (A/J, AKR/J, BALB/cJ, DBA/2J, C57BL/6J, LP/J, I, and RIIIS/J)</td>
<td>0.2375-0.475; &lt;1</td>
<td>7</td>
<td>Genotyping of controlled crosses</td>
<td>Significant, but weak regional correlation (r = .38) (Paigen et al. 2008)</td>
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<tr>
<td>Mus musculus domesticus x Mus musculus castaneus</td>
<td>0.2375-0.475; &lt;1</td>
<td>Genome wide</td>
<td>84.8 (84.0)</td>
<td>SNP genotyping of controlled crosses</td>
<td>1.2-65.6 Mb (mean 15.8 Mb)</td>
<td>31/131 intervals divergent (Dumont et al. 2010)</td>
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<tr>
<td>Gallus gallus (chicken)</td>
<td>Taeniopygia guttata (zebra finch)</td>
<td>100; NA</td>
<td>Genome wide</td>
<td>92</td>
<td>Pedigree SNP genotyping</td>
<td>800 kb (100kb)</td>
<td>1079 (235)</td>
<td>Significantly correlated when comparing 275 intervals (r = 0.5) (Backstrom et al. 2010) (recombination data for chicken from Groene et al. 2009)</td>
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<tr>
<td>Gallus gallus (chicken)</td>
<td>Acrocephalus arundinaceus (great reed warbler)</td>
<td>80-100; NA</td>
<td>Regions of 9 chromosome(s) (ranging from 5%-66% coverage)</td>
<td>26% of chicken genome</td>
<td>Pedigree SNP genotyping</td>
<td>NA</td>
<td>NA</td>
<td>Great reed warbler has substantially shorter map than chicken (Dawson et al. 2007) (recombination data for chicken from Groene et al. 2009)</td>
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</tr>
<tr>
<td>Species</td>
<td>Chromosomes</td>
<td>Chr.</td>
<td>SNP genotyping of controlled crosses</td>
<td>SNP genotyping</td>
<td>Linkage disequilibrium (LD)</td>
<td>Recombination</td>
<td>References</td>
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<tr>
<td><em>Drosophila pseudoobscura</em></td>
<td>0.5-1; NA</td>
<td>Chr. 2</td>
<td>23</td>
<td>130 (50)</td>
<td>466 kb</td>
<td>500 kb</td>
<td>1294</td>
<td>Divergent in 5/38 intervals (Stevison and Noor 2010) (recombination data for <em>D. pseudoobscura</em> from Kulathinal et al. 2008)</td>
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<tr>
<td><em>Drosophila persimilis</em></td>
<td>NA</td>
<td>Chr. 3</td>
<td>12.9</td>
<td>500 kb</td>
<td>780 kb</td>
<td>120 (92-187)</td>
<td>19/19 intervals (Mezner et al. 2010) (recombination data for <em>A. mellifera</em> from Solignac et al. 2007)</td>
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<tr>
<td><em>Apis florea</em> (honey bee)</td>
<td>8-10; NA</td>
<td>Chr. 3 and Chr. 12</td>
<td>Linkage disequilibrium (LD)</td>
<td>464 European, 232 Far East</td>
<td>600 bp European, 1200 bp Far</td>
<td>20 strains (51 meioses)</td>
<td>6/10 hotspots are divergent (Tsai et al. 2010) (recombination data for <em>S. cerevisiae</em> from Solignac et al. 2007)</td>
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<tr>
<td><em>Saccharomyces paradoxus</em></td>
<td>5-20; 13</td>
<td>Chr. 3</td>
<td>3</td>
<td>Hotspot specific</td>
<td>20 strains (51 meioses)</td>
<td>6/10 hotspots are divergent (Tsai et al. 2010) (recombination data for <em>S. cerevisiae</em> from Solignac et al. 2007)</td>
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<td></td>
<td></td>
<td>East</td>
<td>; 204 spores</td>
<td>data for S. cerevisiae from Buhler et al. 2007, Mancer a et al. 2008</td>
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<td>(52,000 genome wide)</td>
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</table>
Table 3: A comparison of recombination rates between species pairs

<table>
<thead>
<tr>
<th>Approach</th>
<th>Linkage Disequilibrium</th>
<th>Sperm-typing</th>
<th>Pedigrees and crosses</th>
</tr>
</thead>
<tbody>
<tr>
<td>Description</td>
<td>LD-based genetic maps use statistics to estimate historical recombination indirectly from patterns of allelic associations in samples from natural populations</td>
<td>Individual or pooled sperm is analyzed for linkage disequilibrium blocks using allele-specific PCR directed to heterozygous SNP sites</td>
<td>Genotype markers in pedigrees or controlled crosses</td>
</tr>
<tr>
<td>Current vs. historical recombination</td>
<td>Historical</td>
<td>Current</td>
<td>Current</td>
</tr>
<tr>
<td>Application</td>
<td>Used in humans, chimpanzees; Can be applied to other organisms depending on population history</td>
<td>Used in humans, mice</td>
<td>Used in mammals, birds, yeast, plants, insects</td>
</tr>
<tr>
<td>Potential Strengths</td>
<td>Relatively quick and inexpensive way to assay whole genome</td>
<td>Direct assay of current recombination; Looking at variation within and between individuals; Looking at specific hotspots; Detecting gene conversion events</td>
<td>Direct assay of current recombination; Can apply across whole genome; Can get sex-specific rates for both sexes;</td>
</tr>
<tr>
<td>Potential Weaknesses</td>
<td>Breaks in LD are not always hotspots; Cannot necessarily detect recent recombination events; Sex-averaged; Made with heterogeneous populations; Known to be influenced by</td>
<td>Male specific (although this can also be a strength as it gives sex-specific information that LD cannot); Labor intensive; Can’t assay whole genome or population</td>
<td>Sensitive to sample size, number of markers, and unknown variation in genome size, structure, and individual variation in recombination rate;</td>
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<tr>
<td>genetic drift, demographic factors, natural selection, variable mutation rates, and gene conversion</td>
<td></td>
<td></td>
<td>Difficult to achieve fine scale; Labor and resource intensive</td>
</tr>
</tbody>
</table>
2. Recombination modulates how selection affects linked sites in Drosophila

Suzanne E. McGaugh, Caiti S. S. Heil, Brenda Manzano-Winkler, Laurence Loewe, Steve Goldstein, Tiffany L. Himmel, Mohamed A. F. Noor


Smukowski Heil Contribution

I was responsible for all Drosophila miranda data in this manuscript, which includes conducting the experimental cross, designing SNP markers, and analyzing all genotypic results. I collaborated with a statistician, Tiffany Himmel, to statistically analyze conservation and divergence of recombination rates, and did the analyses for the global recombination modifier in D. miranda. I furthermore looked at various genomic correlates with recombination. Finally, I assisted in the writing and editing of the manuscript.

Introduction

Homologous meiotic recombination has an important role in molecular evolution. Sufficient recombination uncouples the evolution of different sites on the same chromosome allowing positive or negative selection at one site to act independently from selection at another site. If there is less than effectively free recombination between two selected sites, then linkage results in selection at one site interfering with selection at another site. This has been termed “Hill–Roberson
interference” (Hill et al. 1966; McVean et al. 2000; Comeron et al. 2002; Charlesworth et al. 2009; Loewe et al. 2010; Charlesworth 2012). Hill–Robertson interference increases the probability of fixation of deleterious mutations, decreases the probability of fixation of advantageous mutations, and reduces overall DNA sequence diversity. Thus, the breakdown of linkage disequilibrium between loci experiencing Hill–Robertson interference allows selection to act more efficiently, purging deleterious mutations and accelerating adaptation (Hill et al. 1966; McVean et al. 2000; Comeron et al. 2002; Charlesworth et al. 2009; Loewe et al. 2010; Charlesworth 2012).

Such indirect effects of recombination on the genome (Webster et al. 2012) result in a positive association between the rate of recombination and adaptive evolution (Colegrave 2002; Bourguet et al. 2003; Goddard et al. 2005). For example, recombination rate is positively associated with codon usage bias, whereby those codons coded by the most abundant tRNAs are “preferred” and used more often (Hey et al. 2002; Haddrill et al. 2007). Recombination has direct effects on a genome sequence as well, because recombination influences base composition through biased gene conversion and the distribution of repetitive elements, hotspot sequences, and indels (Begun et al. 1992; Charlesworth et al. 1994; Comeron et al. 1999; Duret et al. 2008; Myers et al. 2010; Webster et al. 2012). Understanding the magnitude of indirect effects in light of these direct effects has proved challenging (Webster et al. 2012).
One striking association is a positive relationship of local recombination rate and nucleotide diversity (Aguade et al. 1989; Stephan et al. 1989; Begun et al. 1992). Originally described in *Drosophila melanogaster* (Begun et al. 1992), the positive relationship between recombination rate and nucleotide diversity has been demonstrated in a wide range of taxa, including humans, mice, yeast, maize, and tomatoes (reviewed in (Smukowski et al. 2011)). It is not fully understood how much of this relationship results from recombination’s indirect versus direct effects on the genome. For instance, mutations created during crossing over or double-strand break repair may generate new polymorphisms and hence increase diversity (Esposito et al. 1993; Strathern et al. 1995; Rattray et al. 2001; Lercher et al. 2002; Filatov et al. 2003; Kulathinal et al. 2008; Hicks et al. 2010). Alternatively, recombination may indirectly influence genetic diversity by mitigating the genomic footprint of selective sweeps and background selection (Smith et al. 1974; Charlesworth et al. 1993; Nordborg et al. 1996).

One way to distinguish between these general explanations is to evaluate the relationship of between-species nucleotide divergence at neutral sites and local recombination rate, because truly neutral mutations are substituted at the same average rate between species as they appear between generations, even if linked to sites under selection (Kimura 1983; Birky et al. 1988). This allows us to predict that both within-species nucleotide diversity and between-species nucleotide divergence would have a positive relationship with local recombination rate (Begun et al. 1992), if the
recombination–diversity association was purely caused by mutation. In contrast, selective sweeps and background selection will cause an association between recombination and within-species nucleotide diversity, but not a relationship between recombination and between-species nucleotide divergence (Smith et al. 1974; Birky et al. 1988). The absence of an association of between-species nucleotide divergence and local recombination rate suggests that variation in recombination rate translates to variation in the efficiency of selection (Begun et al. 1992). Past work relating nucleotide divergence to recombination rate found no relationship between these two variables in several species of *Drosophila*, mouse, beet, yeast, and other species (Begun et al. 1992; Kraft et al. 1998; Takahashi et al. 2004; Noor 2008; Stevison et al. 2010; Smukowski et al. 2011; Mackay et al. 2012). Furthermore, in several species, evidence indicates that segregating ancestral polymorphisms may be responsible for correlations between divergence and recombination rate ((Begun et al. 2007; Cutter et al. 2010; Lohmueller et al. 2011), also suggested by (Kulathinal et al. 2008; Noor 2008)).

The test above, however, implicitly assumes that local recombination rates are conserved between the two species used to generate the nucleotide divergence measure. If recombination rate has diverged between the two species, no relationship between local recombination rate and nucleotide divergence may be detected even when recombination is mutagenic (see Figure 1). Recombination rates, especially at fine scales, are often not conserved among closely related species, as is the case between humans
and chimpanzees (Ptak et al. 2004; Ptak et al. 2005; Winckler et al. 2005); thus, the assumption of conservation of recombination rates may be violated in previous studies, and a more definitive understanding of the diversity–recombination association awaits estimates that are free from this assumption.

Figure 1: Expectation of the relationship between divergence and recombination rate if the recombination–diversity positive correlation is the result of recombination being mostly mutagenic or the result of recombination's effect on selection at linked sites.
Though there are theoretical expectations concerning how recombination rate should affect selection efficiency (Kim 2004; Loewe et al. 2007), it is unclear empirically whether variation in local recombination rates translates into significant variation in the efficiency of selection (Webster et al. 2012). Several empirical studies have tackled this problem (Pal et al. 2001; Betancourt et al. 2002; Presgraves 2005; Haddrill et al. 2007; Bullaughey et al. 2008; Larracuente et al. 2008; Weber et al. 2009; Cutter et al. 2010), and many findings suggest that recombination rate influences the efficiency of positive or negative selection in regions of moderate or high recombination. Still, various confounding factors (e.g., biased gene conversion, gene density) may produce spurious correlations between both recombination and substitution rate, and some authors suggest that there is no strong empirical evidence for recombination affecting the efficiency of selection (apart from reduced selection in regions with essentially no recombination).

The *Drosophila pseudoobscura* system is ideal for pursuing questions about recombination rate variation and its molecular evolutionary consequences. The average crossover rate of *D. pseudoobscura* (about 7 cM/Mb in females) is over twice that of *D. melanogaster* (Ortiz-Barrientos et al. 2006). There is also considerable fine-scale (<200 kb windows) variation in the local recombination rate within the genome of *D. pseudoobscura* and within the genome of its sister species, *D. persimilis* (Cirulli et al. 2007; Kulathinal et al. 2008; Stevison et al. 2010). While some recombination data are available...
for *D. pseudoobscura* and *D. persimilis*, these sister taxa interbreed in the wild (Dobzhansky 1973; Powell 1983; Machado et al. 2002) and are, therefore, not ideal for examining the divergence–recombination association. For example, shared polymorphism due to hybridization and recent speciation may be responsible for the positive divergence–recombination association found in a previous study (Kulathinal et al. 2008) see also (Cutter et al. 2010; Lohmueller et al. 2011). Fortunately, a third species exists (*D. miranda*) that is phylogenetically close to *D. pseudoobscura* but does not interbreed with *D. pseudoobscura*. Since there is still some residual shared ancestral polymorphism (Nowell et al. 2011), we also obtained the genome sequence for a slightly more distantly related outgroup species, *D. lowei* (Figure 2). Sequence from *D. lowei* is useful for generating a proxy for neutral mutation rate across the genome.

**Figure 2: Relationships of study species**

In this work, we generate and compare two fine-scale recombination maps for *D. pseudoobscura*, which each cover approximately 43% of the *D. pseudoobscura* physical genome and one fine-scale recombination map that covers approximately 31% of the *D. pseudoobscura* genome.
miranda physical genome. In order to circumvent the assumption of classic studies, we analyze the relationship of local recombination rate to nucleotide diversity and divergence in regions with very similar recombination rates between the two species. By employing a linear model framework to account for multiple covariates, we conclude that the contribution of recombination to diversity is significant and positive, but recombination contributes little to divergence. This indicates that recombination is likely to modulate the footprint of selection in the genome. Next, we tested the impact of recombination rate on the efficiency of selection. We examined whether recombination rate (1) affects the distribution of nonsynonymous substitutions across the genome and (2) affects the pattern of diversity around nonsynonymous and synonymous substitutions. In particular, we use a generalized linear model to test how recombination modulates the magnitude and physical extent of the loss of diversity surrounding substitutions. Our analysis of these putative selective sweeps should be less sensitive to common confounding factors such as gene expression and GC content than previous measures. In total, this work allowed us to determine that recombination rate has an important impact on how selection shapes diversity across the genome of Drosophila pseudoobscura and its close relatives.

**Materials and Methods**

**Fine-Crossover Maps: Crosses and Technical Details**
Using a backcross design, we developed two recombination maps for \textit{D. pseudoobscura} (Flagstaff and Pikes Peak) and one recombination map for \textit{D. miranda}. For each cross, Duke’s Genomic Analysis Facility genotyped 1,440 individual backcrossed flies for 384 line-specific SNP markers using the Illumina BeadArray platform (Illumina, San Diego, CA) (Fan et al. 2003).

**Fine-Crossover Maps: Recombination Map Construction**

Recombination events were scored when an individual fly’s genotype changed from heterozygous to homozygous (for the parent in the backcross) or vice versa for autosomes and when the fly’s genotype changed between the possible allele combinations for the sex chromosome arms XL and XR. Double crossovers were defined as adjacent intervals with different genotypes on both sides (for instance, a single homozygote genotype call nested in a tract of heterozygote genotype calls). We deemed these as genotyping errors as crossover interference is high within 2 Mb (Fitzpatrick 2009) and removed the single inconsistent genotype, scoring it as missing data.

CentiMorgans were defined as the number of recombination events over the total number of individuals examined for each recombination interval, and we scaled this raw measure with a correction for recombination interference (Kosambi 1943). Throughout the article, recombination rates are given in Kosambi centiMorgans (Kosambi 1943) per Megabase (cM/Mb).
Approximately 1,400 backcross progeny were scored for the Pikes Peak *D. pseudoobscura* map, approximately 1,250 backcross progeny were scored for the Flagstaff *D. pseudoobscura* map, and approximately 1,170 backcross progeny were scored for the *D. miranda* map. Physical genomic distances used to calculate centiMorgans per Megabase (cM/Mb) per interval were based on the *D. pseudoobscura* reference genome v2.6 (Flagstaff) and v2.9 (Pikes Peak, *D. miranda*). Marker order was confirmed by the R (The R Foundation for Statistical Computing 2010) package OneMap (Margarido et al. 2007) using the algorithms Recombination Counting and Ordering (Van Os et al. 2005) and Unidirectional Growth (Tan et al. 2006). OneMap does not accommodate backcrossed designs for sex chromosomes; therefore, we specified an F2 intercross design in these cases. We found one small inversion in *D. miranda* relative to *D. pseudoobscura* on chromosome 2. We estimated the left breakpoint was between the markers at 10,491,527 and 10,660,216 bp, and the right breakpoint was between the markers at 13,318,705 bp and 14,068,383 bp from the telomeric end of chromosome 2. This inversion corresponds to one previously documented between *D. miranda* and *D. pseudoobscura* between markers *rosy* and *nop56* (Bartolome et al. 2006). Recombination rate differences are probably not due to differences in gene order; thus, we used the *D. pseudoobscura* orientation for this inversion when comparing recombination between maps and excluded intervals that included the breakpoints. Confidence intervals (95%) for cM/Mb for each recombination interval were calculated by permutation (Cirulli et al.
2007; Stevison et al. 2010). Confidence intervals for those intervals where we did not find a single recombinant individual were estimated from a binomial distribution—simply, we solved the equation \((1-x)^N = 0.05\), where \(x\) is the 95% upper bound of recombination frequency, and \(N\) is the number of individuals surveyed.

**Fine-Scale Recombination Maps: Defining Intervals With Conserved and Divergent Recombination**

The rationale for regressing out the effect of species (when identifying conserved intervals) was to account for the globally higher recombination rate in *D. miranda* relative to *D. pseudoobscura* and to identify regions where the recombination profile overlapped (e.g., where peaks and troughs can be overlaid). To delimit conserved regions using data that have not been corrected for elevated recombination rate of *D. miranda*, one might identify a region with very similar recombination rates between *D. miranda* and *D. pseudoobscura*, but this region may be a trough in recombination rate for *D. miranda* and a peak in recombination rate for *D. pseudoobscura*. Not correcting for the global elevation of *D. miranda* may lead to falsely concluding that a region has a conserved recombination profile between two maps. Thus, we used a rare events logistic regression (Zelig package in R) between each set of condensed fine-scale recombination maps to identify regions of conserved recombination after accounting for map identity (Flagstaff–Pikes Peak, Flagstaff–*D. miranda*, Pikes Peak–*D. miranda*). The package Zelig
uses the same model as a logistic regression, but it corrects for a bias that is introduced when the sample contains many more of one of the dichotomous outcomes than the other. Recombination events conditioned on the total number of observations was the response variable, and species, interval, and species-by-interval were included as factors in the model. We defined “divergent” intervals as those where tests in each interval between the species from the rare events logistic regression had a q-value of <0.05 after correction for multiple tests (Benjamini 1995). “Conserved intervals” were those intervals that displayed a nonsignificant difference across all three maps when analyzed with a rare events logistic regression and had an odds ratio between 0.62 and 1.615, after accounting for a species effect. We did not correct for multiple tests in defining conserved intervals.

In this way, only intervals that were conserved within and between species were delineated as conserved intervals. The final dataset used to differentiate between the mutagenic and selection hypotheses contained 27 conserved intervals on chromosome 2. We did not use the XR to differentiate between the mutagenic and selection hypotheses—of the 44 intervals condensed across three XR maps, only seven were conserved within and between species. We chose not to combine data from chromosome 2 and XR, as there is some evidence for different evolutionary patterns between autosomal and sex chromosomes in Drosophila (Vicoso et al. 2006).
**Fine-Scale Recombination Maps: Recombination, Diversity and Divergence**

We analyzed the effect of recombination on diversity and divergence by applying a quasibinomial GLM as the data were overdispersed, which has several statistical properties favorable to analyzing proportions such as pairwise diversity (Wilson 2002; Warton et al. 2011). Diversity or divergence was used as a response variable by binding the number of SNP bases to the number of non-SNP, eligible bases with cbind in R. We included recombination rate, proportion of G or C bases within the recombination interval, gene density (measured as a proportion of nucleotides within the recombination interval that are coding), a proxy for neutral mutation rate, and interaction terms as factors in the model.

For these models, the analysis presented is restricted to those conserved, condensed intervals with highly similar recombination rates between all three maps, unless otherwise noted. This restriction removes a classic bias by requiring that the intervals have similar recombination rates between the two species compared for the divergence measures (Figure 1). Similar linear models were also analyzed using the uncondensed intervals for each of the three maps individually. All statistics were performed in R version 2.12.1 (The R Foundation for Statistical Computing 2010) unless otherwise noted.

**Ultrafine Crossover Maps: Recombination Map Construction and Analysis**
Using Flagstaff 16 and Flagstaff 14, we followed the same backcross scheme described in the section “Fine-Crossover Maps: Crosses and Technical Details.” Over 10,000 progeny from this backcross were stored in 96-well plates, frozen at −20°C and amplified for markers over these three regions. PCR products were visualized on a polyacrylamide gel using LICOR 4300.

**Recombination and Nonsynonymous Substitutions**

The number of nonsynonymous substitutions, specific to the *D. pseudoobscura*+*D. persimilis* lineage, were calculated for each gene using PAML using the resequenced genomic and reference genomic data (one *D. lowei*, three *D. miranda*, three *D. persimilis*, two *D. pseudoobscura bogotana*, and 11 *D. pseudoobscura* genomes, filtered for quality as described above). We used a tree rooted with *D. lowei* and considered the branches leading to [D. *persimilis* (D. *pseudoobscura*, D. *pseudoobscura bogotana*)] to be the foreground branches. We included *D. persimilis* a part of the foreground branch because relatively extensive interbreeding occurs between *D. pseudoobscura* and *D. persimilis* across much of the genome, aside from a few inverted regions (Machado et al. 2007; Noor et al. 2007; McGaugh et al. 2012).

Following (Bullaughey et al. 2008), we used a GLMM with Poisson distribution to examine the potential for recombination rate to shape the distribution of nonsynonymous substitutions along the *D. pseudoobscura*+*D. persimilis* lineage. The
model contained the following main effects: the number of silent segregating sites in each gene, GC content in each gene within Flagstaff 16, the proportion of coding bases 50 kb on either side of the gene’s midpoint, weakly selected average pairwise divergence within the gene between *D. persimilis* and *D. lowei* at 4-fold degenerate sites of unpreferred codons (a proxy for neutral mutation rate), recombination rate observed for the interval containing the gene, and a random variable included to account for pseudoreplication of multiple genes per interval. The response variable was the number of nonsynonymous substitutions observed in each gene. This model construction allowed the inclusion of genes whose synonymous substitution count was zero (*sensu* (Bullaughey et al. 2008)). The GC content from Flagstaff16 was used as this was the line used for backcrossing in the crossing scheme, and the Flagstaff map (*D. pseudoobscura*) was used in this analysis.

**Recombination and Reduction in Diversity Around Nonsynonymous Substitutions**

We used 4-fold degenerate sites of unpreferred codons to measure the average levels of diversity as a function of distance from amino acid substitutions along the *D. pseudoobscura*+*D. persimilis* lineage (as identified by PAML, see above). Generalized linear mixed models with a Poisson distribution were used to compare the diversity around nonsynonymous substitutions along the *D. pseudoobscura*+*D. persimilis* lineage in relation to distance from the site and recombination rates measured
in the Flagstaff cross. Measures of diversity at 4-fold degenerate sites were taken 60 kb
(sensu (Sattath et al. 2011)) from the site in either direction (120 kb total) with
nonoverlapping bins of 1,000 bp. The random effects of identities of each substitution
were estimated. We included as covariates (1) divergence between D. persimilis and D.
lowei at 4-fold degenerate sites (a proxy for neutral mutation rate), (2) proportion of
bases that were either G or C in Flagstaff 16 within the 1,000 bp window, (3) proportion
of codons that were nonsynonymous substitutions within the 1,000 bp window, and (4)
proportion of bases that were coding over each 1,000 bp window. The absolute value of
the distance from the site and local recombination rate (at the particular
nonsynonymous substitution) were included in the model as well as the interaction
between distance and recombination rate. All effects in the model were standardized to
mean zero and unit standard deviation. As a control, similar analyses were performed
using synonymous substitutions along the D. pseudoobscura+D. persimilis lineage.
Synonymous substitutions should evolve in a more neutral fashion; thus, less of an
interaction between distance and recombination rate is expected. Any 1,000 bp window
with less than 75 eligible, 4-fold degenerate sites was excluded from the analysis. Any
nonsynonymous or nonsynonymous changes with less than 10 windows were excluded
from the analysis. For the 60 kb analysis, after all filtering steps, our data consisted of
4,338 nonsynonymous and 8,670 synonymous substitutions along the D.
pseudoobscura+D. persimilis lineage on chromosome 2. Four-fold degenerate sites were
used here, rather than 4-fold degenerate sites at unpreferred codons, because too little data were available in each 1,000 bp nonoverlapping window.

**Results**

We first discuss general features of the recombination landscapes we observed in *Drosophila pseudoobscura* and *D. miranda* before we address the implications of these observations for understanding diversity, divergence, and the nature of selection in the genomes we sequenced.

**General Summary of Recombination Data: Fine-Scale Maps**

We generated linkage maps for chromosome 2 and parts of the X chromosome for *D. pseudoobscura* and *D. miranda*. Using a backcross design and inbred lines, we developed two replicate recombination maps (referred to here as “Flagstaff” and “Pikes Peak”) for *D. pseudoobscura* and one recombination map for *D. miranda* using the Illumina BeadArray platform to distinguish heterozygotes from homozygotes of the inbred lines used in the backcross design. These maps measure recombination rate across <200 kb windows, and we refer to these as “fine-scale” maps.

Recombination was surveyed across approximately 43% of the *D. pseudoobscura* physical genome and about 31% of the *D. miranda* physical genome. For each of the three maps, nearly the entire assembled region of chromosome 2 (97.8%–99.4%), the majority of the XR chromosome arm (70.8%–89.4%), and part of the XL...
chromosome arm (~22%–23%) were surveyed. After removal of likely erroneous putative double recombinants, ambiguous genotypes, and markers that did not work or gave inconsistent genotypes, recombination was measured for three different crosses for 1,158–1,404 individuals per map. Excluding larger intervals at the telomeres and centromeres, intervals between markers had a median size across the three maps of 141–148 kb for chromosome 2 and 146–160 kb for the XR chromosome arm.

For chromosome 2, recombination rates ranged from 0–30.8 cM/Mb in *D. pseudoobscura* and 0–24.0 cM/Mb in *D. miranda*. The number of individuals surveyed is often slightly different per interval; therefore, for all intervals where no recombination was detected, we report 0 cM/Mb. The recombination rate for those intervals with “0 cM” should be interpreted as <1 recombination event per total number of individuals surveyed for each interval. Recombination near the telomere and centromere was measured at a broader scale than the remainder of chromosome 2 because we expected these regions to have lower crossover rates than the center of the chromosome (chromosome 2 is telocentric). Because of this limitation, comparisons of recombination rates between the ends of the chromosome and the center are more tentative. Nonetheless, examining recombination across roughly 3 Mb of sequence at the telomeric end and 3 Mb at the centromeric end, we found up to an 8.9-fold difference between the recombination rates at the middle of chromosome 2 relative to the centromeric end. The Pikes Peak *D. pseudoobscura* map exhibited the largest reduction of recombination at the
telomeric or centromeric ends relative to the center of the chromosome for all three maps, though in the Flagstaff *D. pseudoobscura* map and the *D. miranda* map, recombination rates were reduced by at least 2.6-fold in the centromere and telomere relative to the center of the chromosome.

For the XR chromosome arm, recombination rates ranged from 0–25.2 cM/Mb in *D. pseudoobscura* and 0–32.3 cM/Mb in *D. miranda*. The number of crossovers per individual for both chromosome 2 and the XR arm was close to 1 (1.01–1.06) for *D. pseudoobscura* and was 1.40–1.54 for *D. miranda*, illustrating that a greater overall recombination rate in *D. miranda* relative to *D. pseudoobscura* is observed in both an autosome and a sex chromosome.

The XL chromosome arm was not surveyed as intensively (~22%–23% of the XL arm in Pikes Peak and *D. miranda* and ~60% of the XL arm in Flagstaff). The number of crossovers per individual appears consistent with ~1 crossover per chromosome arm, as in *D. pseudoobscura* XR and chromosome 2, but the average number of crossovers per individual on the XL reflects how much of the arm was surveyed. For example, when ~22%–23% of the arm was surveyed, crossovers per individual ranged from 0.23–0.26. A binomial Generalized Linear Model (GLM) with size of the interval as a covariate and interval identity as a factor in the model indicated significant heterogeneity in recombination rate among intervals for chromosome 2, XR, and XL (each tested separately) for each of the three maps (each tested separately, interval
identity \( p < 0.00001, \chi^2 \geq 64.67, df \geq 3, \) in all cases). Furthermore, 95% confidence intervals (generated via the same method in (Cirulli et al. 2007)) do not overlap in many cases between different intervals (shown in Figure 3). Overall, we observe heterogeneity in fine-scale recombination rates within each of the three maps (see Figure 3 with 95% confidence intervals plotted), and we note a reduction in recombination rate around the telomeric and centromeric ends consistent with other studies in Drosophila (Stevison et al. 2010).

Figure 3: Fine-scale recombination rates on chromosome 2
General Summary of Recombination Data: Ultrafine-Scale Maps

Our three fine-scale crossover maps utilized markers on average 141–160 kb apart (median interval size for each of the three maps, with the exception of XL where the median distance between markers was 200–1,775 kb for the three crosses). We additionally examined three regions on chromosome 2 in more detail. Each of these regions spanned a total of 99–125 kb, and we placed markers every ~20 kb within the region (16 total intervals). These regions were originally picked because previous data (Kulathinal et al. 2008; Stevison et al. 2010) indicated that recombination rates for each of these regions differed (regions are referred to as 6 Mb, 17 Mb, and 21 Mb, which indicate approximate location on chromosome 2). We refer to these as “ultrafine-scale” maps. For these ultrafine maps, we followed the same backcross scheme as above, and we scored approximately 10,000 individuals for each marker. For the 16 ultrafine intervals, each interval was on average 20.61 kb long (range 12.6–27.4 kb).

Recombination rates range from 1.6–21.2 cM/Mb for these ~20 kb intervals (Figure 4). The ultrafine-scale map uncovered variation in recombination rates that was not apparent with the fine-scale maps. For example, for the 17 Mb ultrafine-scale region on chromosome 2, the recombination rates for the two fine-scale intervals spanning this region (17.5–17.7 Mb) are 5.6 and 4.4 cM/Mb. The ultrafine-scale recombination rates, in contrast, ranged from 3.5–21.2 cM/Mb (markers spanning 17.5–17.7 Mb). This heterogeneity in recombination rates within the ultrafine regions was statistically
significant (binomial GLM similar to that described in fine-scale section above: \( p = 0.0011, df = 14, \chi^2 = 35.91 \)) and highlights the fact that “broader” scale measures of recombination rates (such as the fine-scale measures here) are averages of true variation in recombination rate.

![Figure 4: Ultrafine recombination rates](image)

**Recombination Rate Comparison between Maps**

For comparisons of recombination rates between fine-scale maps, we restricted our analysis to intervals that were condensed to have nearly identical physical marker placement between the three fine-scale maps (Figure 5). Recombination was estimated as detailed above, using the number of crossovers spanning the newly defined physical intervals. After condensing across all three maps, 97 intervals remained for chromosome 2 and 44 intervals for XR (see Table 4). The XL chromosome arm was not included in the analysis that used condensed intervals across maps because too few intervals overlapped between all three maps. When comparing two maps, intervals were condensed between those two maps only.
Local Recombination Rates between Two D. pseudoobscura Maps Are Similar

Recombination rates did not differ significantly between the two D. pseudoobscura maps for either the XR or chromosome 2 for the two-map comparisons (each chromosome analyzed separately, rare events logistic regression, absolute value of $z > 0.3901$, $p > 0.236$, in both cases). For chromosome 2, one interval was significantly different in recombination rate after correcting for multiple tests (Benjamini 1995). For the XR, no intervals between the two D. pseudoobscura maps were significantly different.
in recombination rate after correcting for multiple tests. The 95% confidence intervals for the odds ratio of the difference between maps were narrow and located around zero, indicating that the maps are likely very similar (chromosome 2, 0.87–1.10; XR, 0.94, 1.28; within-species two map comparison). It is unlikely that the single significant difference observed within the same species is because of slight differences in marker placement between the two maps. The marker placement for this interval was nearly identical between the two maps (left marker, 102 nucleotides different between maps; right marker, 17 nucleotides).

**Globally Higher Recombination Rate in D. miranda Relative to D. pseudoobscura**

For both chromosome 2 and the XR chromosome arm, *Drosophila miranda* had significantly higher recombination rates than both *D. pseudoobscura* maps (Figure 5, Table 4). A rare events logistic regression of two-map comparisons indicated that the recombination rate of the *D. pseudoobscura* crosses we surveyed is about 76%–78% of the *D. miranda* recombination rate we observed on chromosome 2 (absolute z value>4.5374, p<0.001 for *D. miranda* relative to either *D. pseudoobscura* map, Table 4). The recombination rate of *D. pseudoobscura* is about 68%–71% of the *D. miranda* recombination rate on the XR chromosome arm (rare events logistic regression absolute z value>5.101, p<0.001 for *D. miranda* relative to either *D. pseudoobscura* map, Table 4).
<table>
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<th>Ch2 (N = 97)</th>
<th>XR (N = 44)</th>
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<td>different</td>
<td>Odds ratio</td>
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<td><em>D. pseudoobscura</em> Pikes Peak - <em>D. miranda</em></td>
<td>3/48</td>
<td>0.7629 (0.6780, 0.8584)</td>
</tr>
</tbody>
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**Limited Local Recombination Rate Divergence between *D. pseudoobscura* and *D. miranda***

After the global difference between *D. miranda* and *D. pseudoobscura* is accounted for by the rare events logistic regression, recombination rates within and between species appear very similar for chromosome 2 (Figure 5). None of the intervals for the two-map comparison between *D. miranda* and *D. pseudoobscura*–Flagstaff were significantly different after correction for multiple tests, though power to detect significant differences on a per interval basis was likely weak. For example, 15 of the 115 intervals on chromosome 2 showed at least a 3-fold difference in recombination rate between maps, though this magnitude of difference was not significant in our rare events logistic regression after correcting for multiple tests. Likewise, only one of the
intervals for the two-map comparison between *D. miranda* and *D. pseudoobscura*–Pikes Peak was significantly different after correction for multiple tests, but 19 of the 123 intervals exhibited at least a 3-fold difference in recombination rate between maps for chromosome 2.

The XR chromosome exhibited a qualitatively larger difference in recombination rate between species than chromosome 2. After the global difference between *D. miranda* and *D. pseudoobscura* is accounted for by a rare events logistic regression, two of the intervals between *D. miranda* and *D. pseudoobscura*–Flagstaff for the two-map comparison and seven of the intervals between the *D. miranda* and *D. pseudoobscura*–Pikes Peak two-map comparison were significantly different after correction for multiple tests. Six of the 72 intervals between *D. miranda* and *D. pseudoobscura*–Flagstaff two-map comparison exhibited at least a 3-fold difference, and 12 of 102 intervals between *D. miranda* and *D. pseudoobscura*–Pikes Peak exhibited at least a 3-fold difference.

Twenty-seven of 97 condensed intervals (three-map comparison, condensed between all three maps) for chromosome 2 were considered to be “conserved” within and between species. This means that they displayed a nonsignificant difference across all three maps when analyzed with a rare events logistic regression and had an odds ratio between 0.62 and 1.615 after the effect of map identity was taken into account. These “conserved” intervals were used for further downstream analyses (see “Diversity, Divergence, and Recombination”). For the XR, seven of 44 intervals condensed between
all three maps were conserved within and between species according to the criteria outlined above.

In sum, we observe strong conservation in recombination rates within a single species, while between species, we see globally elevated recombination rates in *D. miranda*. Once the global difference is accounted for, there are few intervals with significant differences in recombination rate within and between species. Thus, it is possible and parsimonious that recombination rate is generally conserved at the scale examined here (~180 kb) over moderate evolutionary timescales (2–2.5 my).

**Diversity, Divergence, and Recombination**

We used various Illumina platforms to resequence genomic DNA from 10 *D. pseudoobscura* lines using virgin females from lines that were inbred for five or more generations with full-sibling single-pair mating. *Drosophila pseudoobscura* populations across North America display very little differentiation, as indicated by low *F*<sub>st</sub> values (always<0.10, often<0.05 for loci located outside of the inversion polymorphisms of the third chromosome) (Schaeffer et al. 1992; Noor et al. 2000). Therefore, the choice of strains sequenced for estimating diversity covered much of the species range but was fairly random. We also sequenced two lines of *D. persimilis* (one of these was provided by S. Nuzhdin), two lines of *D. pseudoobscura bogotana* (one of these was provided by S. Nuzhdin), one line of *D. lowei*, and three lines of *D. miranda* (two provided by D. Bachtrog, Short Read Archive accession numbers SRA044960.1, SRA044955.2, and
SRA044956.1; see also http://pseudobase.biology.duke.edu/). The divergence between \textit{D. persimilis} and \textit{D. lowei} was used to generate measures of a proxy for neutral mutation rate across the genome. In all diversity and divergence calculations, the reference sequences for the \textit{D. pseudoobscura} and \textit{D. persimilis} genomes were both included (Richards et al. 2005; Clark et al. 2007). Briefly, average pairwise diversity and divergence was calculated for 4-fold degenerate sites, focusing exclusively on unpreferred codons (Vicario et al. 2007), though we obtained very similar results when using all 4-fold degenerate sites. Overall, recombination is significantly and positively associated with average pairwise diversity but not average pairwise divergence at 4-fold degenerate sites of unpreferred codons. We examined this relationship in several ways.

\textit{Diversity, Not Divergence, Is Positively Associated with Recombination in All Intervals}

We analyzed each chromosome for each uncondensed recombination map independently using a generalized linear model for diversity and a separate model for divergence. After accounting for multiple covariates, diversity at 4-fold degenerate sites of unpreferred codons shows a significant, positive relationship with recombination, while divergence at 4-fold degenerate sites of unpreferred codons does not. This result is consistent for each of the three recombination maps (\textit{D. pseudoobscura}–Flagstaff, \textit{D. pseudoobscura}–Pikes Peak, and \textit{D. miranda}) for both chromosome 2 and the XR
chromosome arm. The XL chromosome arm contained too few intervals for analysis for D. pseudoobscura–Flagstaff. For D. pseudoobscura–Pikes Peak and D. miranda, diversity showed a significant, or nearly significant, positive relationship with recombination, while divergence did not.

Table 5: Factors affecting diversity within species at four-fold degenerate sites for unpreferred codons using intervals with conserved recombination rate

<table>
<thead>
<tr>
<th></th>
<th>Df</th>
<th>Deviance</th>
<th>Residual Df</th>
<th>Residual Dev.</th>
<th>F</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Null</td>
<td>26</td>
<td>57.009</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Gene Density</td>
<td>1</td>
<td>2.3190</td>
<td>25</td>
<td>54.690</td>
<td>2.2948</td>
<td>0.147171</td>
</tr>
<tr>
<td>Mutation</td>
<td>1</td>
<td>12.7343</td>
<td>24</td>
<td>41.955</td>
<td>12.6013</td>
<td>0.002289*</td>
</tr>
<tr>
<td>Recombination</td>
<td>1</td>
<td>6.1877</td>
<td>23</td>
<td>35.768</td>
<td>6.1231</td>
<td>0.023521*</td>
</tr>
<tr>
<td>GC</td>
<td>1</td>
<td>11.1854</td>
<td>22</td>
<td>24.582</td>
<td>11.0685</td>
<td>0.003751*</td>
</tr>
<tr>
<td>Gene Density*Mutation</td>
<td>1</td>
<td>2.0720</td>
<td>21</td>
<td>22.510</td>
<td>2.0504</td>
<td>0.169304</td>
</tr>
<tr>
<td>Gene Density*Recombination</td>
<td>1</td>
<td>2.8041</td>
<td>20</td>
<td>19.706</td>
<td>2.7748</td>
<td>0.113065</td>
</tr>
<tr>
<td>Mutation* GC</td>
<td>1</td>
<td>0.5488</td>
<td>19</td>
<td>19.157</td>
<td>0.5430</td>
<td>0.470669</td>
</tr>
<tr>
<td>Recombination* GC</td>
<td>1</td>
<td>0.0007</td>
<td>18</td>
<td>19.156</td>
<td>0.0007</td>
<td>0.978599</td>
</tr>
</tbody>
</table>

Table 6: Factors affecting divergence between species at four-fold degenerate sites for unpreferred codons using intervals with conserved recombination rate

<table>
<thead>
<tr>
<th></th>
<th>Df</th>
<th>Deviance</th>
<th>Residual Df</th>
<th>Residual Dev.</th>
<th>F</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Null</td>
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<td>53.578</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Gene Density</td>
<td>1</td>
<td>2.1647</td>
<td>25</td>
<td>51.414</td>
<td>1.1784</td>
<td>0.29201</td>
</tr>
<tr>
<td>Mutation</td>
<td>1</td>
<td>4.8404</td>
<td>24</td>
<td>46.573</td>
<td>2.6349</td>
<td>0.12192</td>
</tr>
<tr>
<td>Recombination</td>
<td>1</td>
<td>0.2540</td>
<td>23</td>
<td>46.319</td>
<td>0.1382</td>
<td>0.71437</td>
</tr>
<tr>
<td>GC</td>
<td>1</td>
<td>7.3218</td>
<td>22</td>
<td>38.997</td>
<td>3.9857</td>
<td>0.06124</td>
</tr>
<tr>
<td>Gene Density*Mutation</td>
<td>1</td>
<td>0.5094</td>
<td>21</td>
<td>38.488</td>
<td>0.2773</td>
<td>0.60492</td>
</tr>
<tr>
<td>Gene Density*Recombination</td>
<td>1</td>
<td>1.9069</td>
<td>20</td>
<td>36.581</td>
<td>1.0380</td>
<td>0.32178</td>
</tr>
<tr>
<td>Mutation* GC</td>
<td>1</td>
<td>0.0309</td>
<td>19</td>
<td>36.550</td>
<td>0.0168</td>
<td>0.89827</td>
</tr>
<tr>
<td>Recombination* GC</td>
<td>1</td>
<td>0.2399</td>
<td>18</td>
<td>36.310</td>
<td>0.1306</td>
<td>0.72202</td>
</tr>
</tbody>
</table>

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Diversity, Not Divergence, Is Positively Associated with Recombination in Conserved Intervals

The analysis above suggests that the recombination–diversity relationship is probably the result of the effect of recombination on selection at linked sites; however, inadvertently including regions with discordant recombination rates between species in the analysis above could result in a pattern that supports this hypothesis—even when recombination is predominantly mutagenic (Figure 1). To resolve this potential bias, we restricted analysis to only regions that exhibited conserved recombination rates between all three chromosome 2 maps (N = 27 intervals; described above) and examined recombination in association with average pairwise \textit{D. pseudoobscura} diversity at 4-fold degenerate sites of unpreferred codons (Table 5; Figure 6) and average pairwise \textit{D. pseudoobscura–D. miranda} divergence at 4-fold degenerate sites of unpreferred codons (Table 6; Figure 6). The effect of recombination on diversity was significant when the analysis was restricted to only those regions with the most conserved recombination rates (quasibinomial GLM, $F = 6.123$, $p$ value = 0.024), and the effect of recombination on divergence remained nonsignificant (quasibinomial GLM, $F = 0.138$, $p$ value = 0.714). These regions contained only one interval within 4 Mb of the telomeric end and no intervals within 4 Mb of the centromeric end of the chromosome; thus, these results are not a function of broad-scale regional recombination rate differences across the chromosome. These results support the hypothesis that recombination affects diversity
through the effect of selection on linked sites. We did not perform an analysis on conserved windows for the X chromosome, as only seven intervals were conserved within and between species.

Figure 6: Relationship of recombination rate to diversity (filled circles, solid line) and divergence (open circles, dotted line) for fine-scale regions with conserved recombination between D. pseudoobscura–D. miranda.

Recombination and Selection

To determine the impact of recombination rate on selection at linked sites in the genome, we used two generalized linear models to analyze the relationship of recombination rate and several measures that may be indicative of the efficiency of selection: (1) abundance of nonsynonymous substitutions and (2) average pairwise
nucleotide diversity at 4-fold degenerate sites around nonsynonymous substitutions. We analyzed the association of recombination rate with these two measures in a generalized linear model framework to account for covariates such as gene density, GC content, and a proxy for neutral mutation rate. Biased gene conversion may influence substitution rates; thus, we controlled for GC content in all of the analyses below (Marais et al. 2001; Marais 2003; Duret et al. 2008; Webster et al. 2012). We did not consider gene expression as a covariate, though some studies point to a negative relationship with recombination rate (Haddrill et al. 2008).

No Correlation of Recombination With Nonsynonymous Substitution Abundance

The relationship of recombination rate to nonsynonymous substitution abundance was examined with the *D. pseudoobscura* Flagstaff fine-scale recombination maps. Nonsynonymous substitution abundance was measured as the nonsynonymous substitutions on the branch leading to *D. pseudoobscura*+*D. persimilis* as identified with PAML. The response variable was the number of nonsynonymous substitutions in each gene, and the covariates of the linear model included (1) the number of synonymous substitutions in the gene in question allowing for inclusion of genes where \( K_s = 0 \), (2), GC content of the gene, (3) gene density of 50 kb on either side of the midpoint of the gene, and (4) average pairwise divergence at 4-fold degenerate sites of unpreferred codons between *D. persimilis* and *D. lowei* as a proxy for neutral mutation rate within the gene.
We found no relationship (Table 7) between recombination and nonsynonymous substitution abundance with the fine-scale data (generalized linear model with Poisson distribution, $z = -0.614$, $p = 0.539$).

Table 7: Test for relationship between recombination rate and number of nonsynonymous substitutions

<table>
<thead>
<tr>
<th>Response: Nonsynonymous substitutions along the D. pseudoobscura+ D. persimilis lineage</th>
<th>Model</th>
<th>Factor tested</th>
<th>Estimate</th>
<th>Std. error</th>
<th>z-value</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Fine-scale</td>
<td>(Intercept)</td>
<td>2.574891</td>
<td>0.207963</td>
<td>12.38</td>
<td>&lt;0.0001*</td>
<td></td>
</tr>
<tr>
<td>Synonymous</td>
<td>0.053427</td>
<td>0.001557</td>
<td>34.31</td>
<td>&lt;0.0001*</td>
<td></td>
<td></td>
</tr>
<tr>
<td>GC content</td>
<td>-4.892668</td>
<td>0.339146</td>
<td>-14.43</td>
<td>&lt;0.0001*</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Gene density</td>
<td>0.158809</td>
<td>0.196072</td>
<td>0.81</td>
<td>0.418</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Neutral mutation rate</td>
<td>0.470959</td>
<td>3.417997</td>
<td>0.14</td>
<td>0.890</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Recombination</td>
<td>-0.015829</td>
<td>0.019014</td>
<td>0.83</td>
<td>0.405</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Footprints from Putative Hitchhiking May Be Slightly Larger in Low Recombination

Regions

In response to selective sweeps, a trough in diversity should be visible around selected variants (Kaplan et al. 1989; Stephan 1992; Kim et al. 2002; Stephan 2010; Sattath et al. 2011). We analyzed diversity surrounding the nonsynonymous substitutions along the lineage leading to D. pseudoobscura+ D. persimilis identified by PAML. We compared the average pairwise diversity patterns at 4-fold degenerate sites surrounding these substitutions in relation to the Flagstaff recombination rate and distance in basepairs from the substitution. In regions with high recombination rates, the footprints of selection are thought to be narrower than in regions with low recombination rates, where strong linkage between sites will create a stronger signature of sweeps (Kaplan et
al. 1989; Kim et al. 2002; JH 2004; Lohmueller et al. 2011). As a control, similar analyses were performed using synonymous substitutions along the *D. pseudoobscura*+*D. persimilis* lineage following (Sattath et al. 2011). Synonymous substitutions, in many cases, evolve in a more neutral fashion than nonsynonymous substitutions ((Sattath et al. 2011), but see (Kern et al. 2002; Bartolome et al. 2005)). In a recent genome-scale analysis conducted with data similar to what are presented here, little reduction in diversity was seen around synonymous substitutions (Sattath et al. 2011); this study instead saw an increase in diversity, which disappeared after correction for local mutation rates.

We considered 60 kb on either side of the substitution along the *D. pseudoobscura* lineage divided into 1,000 bp nonoverlapping windows (*sensu* (Sattath et al. 2011)). For each 1,000 bp window, the response variable was the number of polymorphic 4-fold degenerate sites. The generalized linear model included the following covariates: (1) total 4-fold degenerate sites, (2) GC content, (3) proportion of coding bases, (4) divergence of *D. lowei*–*D. persimilis* at 4-fold degenerate sites as a proxy for neutral mutation rate, and (5) proportion of bases that were nonsynonymous substitutions. The identities of each nonsynonymous substitution were included as random effects. This generalized linear mixed model with Poisson distribution included the following factors: absolute physical distance from the substitution, fine-scale-derived estimates of recombination rate, and the interaction between these two factors. A
negative interaction term means that short distances from a substitution and high recombination rates have similar effects on diversity as large distances and low recombination rates. We expect the interaction term for distance and recombination rate to be much reduced in magnitude for synonymous substitutions in comparison to the nonsynonymous analysis.

We found a small but significant negative interaction term of physical distance from the nonsynonymous site and recombination rate on nucleotide diversity around nonsynonymous substitutions (Poisson GLMM, $z = -7.52, p<0.001$; **Figure 7 and 8**). In other words, higher rates of recombination allow for recovery of diversity at shorter physical distances from the nonsynonymous site than lower recombination rates (**Figure 8**). In contrast, a weaker interaction was detected for the interaction of distance and recombination rate on diversity around synonymous substitutions along the *D. pseudoobscura* lineage (Poisson GLMM, $z = -2.43, p = 0.015$; **Figure 7 and 8**). GLM plots for the very low recombination rates of <0.5 cM/Mb show wider dips in diversity (and more associated noise; **Figure 8**) than plots for recombination rates of >0.5 cM/Mb (**Figure 8**).
Distance from a substitution had a positive, significant effect on diversity as expected if linked selection of substitutions generates a dip in diversity. Recombination rate also had a positive, significant effect on diversity as expected, if either recombination was mutagenic or if positive/negative selection was operating on the chromosome. The proportion of nonsynonymous substitutions around a substitution had a negative significant effect on diversity surrounding a nonsynonymous site as expected if many of these substitutions combine forces to generate stronger selective sweeps. The interaction term pointing to deeper dips in diversity for lower recombination rates is no longer significant when examining only 5 kb or 15 kb on either side of the focal substitution (it is negative for nonsynonymous substitutions and
positive for synonymous substitutions), but it is conceivable that this lack of significance represents an issue with window size or sampling.

![Figure 8: Footprints in diversity around substitutions](image)

**Discussion**

Overall, our study identified both global and local differences in recombination rate between two closely related species of *Drosophila*. Aside from regions with exceptionally low recombination rates (Haddrill et al. 2007; Betancourt et al. 2009), variation in local recombination rates between species must be accounted for prior to concluding that the association between recombination rate and diversity is probably caused by recombination modulating the effects of selection at linked sites (Charlesworth 2010). By restricting our analysis in the *Drosophila pseudoobscura* system to only those regions with conserved recombination rates within...
and between species, we rejected the hypothesis that recombination rate (at the scale tested) significantly affects divergence at 4-fold degenerate sites for unpreferred codons. These results support the conclusion that recombination has a substantial impact on how selection affects diversity in the genome. Furthermore, additional analyses suggest that recombination rate variation affects the impact of Hill–Robertson effects like selective sweeps and background selection in this system.

**Ultrafine and Fine-Scale Variation in Crossover Rate in Drosophila**

Here and in other recent work (Cirulli et al. 2007), we demonstrate that ultrafine-scale patterns of crossover rate (intervals spanning 20 kb) are also significantly heterogeneous in *D. pseudoobscura*. In each ultrafine region on chromosome 2, recombination rates varied by up to 6-fold (17 Mb region) over only approximately 120 kb (6 Mb region variation is 3.6-fold, and 21 Mb region variation is 5.1-fold), and ultrafine-scale maps reveal variation not detected in the fine-scale maps. This was especially apparent for the 17 Mb region, where ultrafine-scale recombination rates ranged from 3.5 to 21.2 cM/Mb, and fine-scale recombination rates in the same area ranged only from 4.4 to 5.6 cM/Mb. This heterogeneity suggests that our fine-scale measures (intervals spanning <200 kb) are averages of actual variation in recombination rate.
In humans, broad-scale variation averages over the density and intensity of ~2 kb hotspots that occur in clusters every 60–90 kb (McVean et al. 2004; Myers et al. 2005). The majority of recombination occurs at these hotspots, and the majority of recombination is governed by the DNA binding protein PRDM9 and its recognition motifs in humans (Baudat et al. 2010; Berg et al. 2010; Hochwagen et al. 2010; Parvanov et al. 2010; Hinch et al. 2011). Interestingly, several studies in different regions of the D. melanogaster genome indicate that linkage disequilibrium decays rapidly (Langley et al. 2000; Schaeffer et al. 2001; Palsson et al. 2004; Mackay et al. 2012), suggesting that the heterogeneity we observed in ultrafine-scale maps may not be governed by clustered hotspots similar to those in humans, or at least that a nontrivial amount of recombination may occur outside such “hotspots.”

To assess whether “hotspots” of some sort exist in D. pseudoobscura, genome-wide patterns of linkage decay need to be investigated or incredibly fine-scale maps (interval size <5 kb) need to be made. Such a line of inquiry would help address basic questions about the requirements for functional recombination across various taxa. For example, there are several notable differences regarding the formation and function of the synaptonemal complex and the role of double-strand breaks across taxa (Hawley et al. 1992; Roeder 1997; Keeney 2001; Hawley et al. 2002; McKim et al. 2002; Page et al. 2003). Furthermore, the Drosophila lineage completely lacks several proteins essential for generating crossovers and double-strand break repair in other organisms (McKim et al. 2002).
It is likely that understanding particular sequence features associated with recombination on a kilobase scale in *Drosophila* will uncover more details about the mechanistic underpinnings of meiosis that differentiate these species and the distribution of crossovers across the genome.

**Drosophila miranda Has Elevated Global Recombination Rate Relative to D. pseudoobscura**

Recombination rates at broad scales are conserved between populations and species (Brooks et al. 1986; True et al. 1996; Backstrom et al. 2010; Beukeboom et al. 2010; Meznar et al. 2010; Dumont et al. 2011; Smukowski et al. 2011). Our fine-scale data are generally consistent with these findings except that *D. pseudoobscura* has about three-fourths the rate of recombination, on average, as *D. miranda* for chromosome 2 and about three-fifths the rate of recombination of *D. miranda* on the XR chromosome arm.

Notably, *D. melanogaster* has one of the lowest recombination rates in the genus, as evidence indicates that *D. mauritiana, D. simulans, D. virilis, D. pseudoobscura, D. miranda,* and *D. persimilis* all exhibit higher rates of recombination (True et al. 1996; Ortiz-Barrientos et al. 2006; Stevison et al. 2010); this should be considered when interpreting hitchhiking and linkage data from *D. melanogaster* to patterns of recombination in *Drosophila* in general.
Recombination Prevents Diversity Erosion During Selection

Our results indicate that recombination affects diversity through mediating selection in the genome. While accounting for multiple covariates, we found no association between recombination and average pairwise divergence at 4-fold degenerate sites of unpreferred codons, and a significant, positive association of recombination with average pairwise diversity at 4-fold degenerate sites of unpreferred codons. Using data from our fine-scale maps, we ensured that recombination rates are nearly identical between the species used to generate divergence estimates; thus, we absolved a key assumption made in previous studies (see Figure 1). Data from *Drosophila* suggest both positive and negative selection are markedly less efficient in nearly nonrecombining regions of the genome (Bachtrog et al. 2002; Betancourt et al. 2002; Haddrill et al. 2007; Betancourt et al. 2009; Campos et al. 2012), and a relationship of diversity but not divergence to recombination is apparent for other species of *Drosophila* (Begun et al. 1992; Presgraves 2005; Begun et al. 2007; Stevison et al. 2010), mouse (Takahashi et al. 2004), beet (Kraft et al. 1998), tomato (Stephan et al. 1998; Roselius et al. 2005), *Caenorhabditis* (Cutter et al. 2010), and yeast (Cutter et al. 2011). This last example is especially interesting because recombination is known to be mutagenic in yeast (Strathern et al. 1995; Rattray et al. 2001), but there is a negative or absent divergence–recombination correlation (Noor 2008; Cutter et al. 2011); thus, it may be that recombination is somewhat mutagenic in many organisms, but the power of
recombination to modulate the diversity eroding effects of selection likely has a much greater impact on the genome.

In other systems, the divergence–recombination association is positive, which may be interpreted as evidence that recombination is predominately mutagenic. A positive divergence–recombination association is apparent for humans (Hellmann et al. 2003; Spencer et al. 2006), maize (Tenaillon et al. 2004), and in an inverted region between *D. pseudoobscura* and *D. persimilis* (Kulathinal et al. 2008). This association may be attributable to mutation (Lercher et al. 2002), but unmeasured variables or segregating ancestral polymorphism could predispose a system to exhibiting a positive divergence–recombination relationship (Begun et al. 2007; Noor 2008; Noor 2008; Cutter et al. 2010; Lohmueller et al. 2011). For instance, in *C. briggsae*, segregating ancestral polymorphism leads to the signature of recombination-associated mutation (i.e., a positive divergence–recombination association), but further examination shows the majority of polymorphism heterogeneity is caused by recombination affecting the impact of selection at linked sites (Cutter et al. 2010).

**Recombination Rate and Abundance of Nonsynonymous Substitutions**

Since recombination probably mediates the effects of hitchhiking in our system, we sought to understand whether this hitchhiking is primarily positive or negative (background, purifying) selection and if recombination rate variation has a significant
impact on the potential efficacy of selection. Evidence is emerging that in many organisms, especially those with large population sizes, selection may play a substantial role in shaping the genome (Hahn 2008). For partial selfers, it seems that background selection substantially affects the genome (Cutter et al. 2003; Nordborg et al. 2005; Rockman et al. 2010; Slotte et al. 2011), while in outcrossing species Drosophila, mice, and Capsella grandiflora a large fraction of the genome may be influenced by positive selection (Sella et al. 2009; Halligan et al. 2010; Slotte et al. 2011). The majority of studies find strong support that recombination can shape adaptive evolution when comparing regions of no recombination to regions with some or abundant recombination. However, after accounting for multiple covariates in regions with detectable recombination rates, there is often very little relationship between recombination rate and the efficacy of selection (Marais et al. 2001; Haddrill et al. 2007; Webster et al. 2012).

Across chromosome 2, we found no relationship between the number of nonsynonymous substitutions and the recombination rate as measured with our fine-scale Flagstaff map. Reanalysis of the fine-scale data after removal of the first and last 3 Mb of the chromosome did not change the relationship of fine-scale recombination rate to nonsynonymous substitutions.

Recombination Rate and Diversity Around Nonsynonymous Substitutions (GLM)
Our observation of a reduction of average pairwise diversity at 4-fold degenerate sites around nonsynonymous substitutions (Figure 8) is consistent with the idea that positive selection may have fixed many nonsynonymous substitutions along the ancestral lineage leading to *D. pseudoobscura*+*D. persimilis*, as has been argued elsewhere for other *Drosophila* species (Haddrill et al. 2010; Sattath et al. 2011). While potentially less common, dips in diversity could also be caused by deleterious mutations that can get fixed by chance if deleterious selection coefficients are small enough—a situation we call “loser's luck” (but see (Loewe et al. 2006; Haddrill et al. 2010)), and theoretical investigations of entirely neutral substitutions showed that their quick fixation can also lead to dips in diversity (Tajima 1990). Thus, while many of the dips in diversity we see may be caused by positive selection, both loser's luck and fixation of neutral substitutions may also contribute.

Diversity may be recovered slightly farther from a nonsynonymous substitution in areas of low recombination than in areas of high recombination, and such a relationship is not as pronounced for synonymous substitutions fixed along the lineage leading from the common ancestor of *D. pseudoobscura* and *D. persimilis* (Figure 8). Similarly, in *Arabidopsis*, haplotype blocks around nonsynonymous SNPs are larger than around synonymous SNPs (Kim et al. 2007). Our data agree with theoretical expectations (Kaplan et al. 1989; Kim et al. 2002) and past studies that show negative correlations of polymorphisms and nonsynonymous substitutions in
Drosophila (Andolfatto 2007; Begun et al. 2007; Macpherson et al. 2007; Sattath et al. 2011); indeed, our data also show a significant negative relationship for nonsynonymous substitutions and within-species polymorphisms, generally. Yet the negative interaction term between recombination rate and distance from focal substitutions we observed is dependent on window size and distance from the substitution examined.

Conclusions

Our study documented global and local differences in recombination rate between two closely related species, and these data indicate that recombination probably modulates Hill–Robertson effects in the genome, causing a positive association of diversity with recombination. While we found no overall association of recombination rate with the number of nonsynonymous substitutions at the fine scale, we found evidence for dips in diversity around nonsynonymous substitutions that are dependent on the distance from the substitution, local recombination rate, and a number of other factors. In total, our study adds to the growing literature that indicates that selection must be a ubiquitously important factor for shaping diversity across much of the genome (Smith et al. 1974; Kaplan et al. 1989; Kim et al. 2002).
3. Zinc finger binding motifs do not explain recombination variation within or between species of Drosophila

Caiti S.S. Heil and Mohamed Noor

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Introduction

Meiotic recombination is an essential process both mechanistically and evolutionarily, and thus should experience strong selective pressures. However, identifying how selection affects the locations of recombination events is more complex than was once assumed. Recombination rate is variable within and among genomes, displaying significant heterogeneity across most living organisms and evolving rapidly, with recombination “hotspot” turnover in as short as 120,000 years (Jeffreys et al. 2009). While years of research have determined some elements associated with recombination rate variation, such as temperature, GC content, repeats, SNP density, chromatin state, and histone modifications, the specific effects of DNA sequence “motifs” have attracted much attention.

The 13-base degenerate motif CCNCCNTNNCCNC recruits recombination events in approximately 40% of European human hotspots (Myers et al. 2008; Baudat et al. 2010). This motif binds the Cys2His2 zinc finger protein PRDM9 in humans, and allelic variation at Prdm9 modifies hotspot activity within both humans and mice (Baudat et al.
The *Prdm9* gene also contains a SET-methyltransferase domain, which is responsible for the common chromatin feature trimethylation of lysine 4 of histone H3, or H3K4me3. H3K4me3 in yeast and mouse seems to be a prominent and pre-existing mark of active recombination sites, creating a link between sequence and epigenetic features affecting recombination (Borde et al. 2009). This link inspired a proposed model in humans involving the recognition of a DNA sequence motif by PRDM9 and the modification of adjacent nucleosomes by the SET domain (Barthes 2011). Proteins with an affinity to the modification H3K4me3 are recruited and may modify the chromatin or nucleosomes further. The conserved topoisomerase II-like protein SPO11 subsequently recognizes one or several of these signals, binds to the DNA at that location, and initiates recombination by a double-strand-break (DSB).

Cys:His zinc fingers are among the most common DNA-binding motifs found in eukaryotic transcription factors. These zinc finger proteins usually contain multiple “fingers”, all of which have a conserved ββα structure with amino acids in the α-helix
contacting DNA in the major groove of the double helix (Wolfe et al. 2000) (**Figure 9**).

**Figure 9: A model of Cys2His2 zinc finger binding**

Zinc finger proteins function chiefly in protein-DNA binding, but also may be involved in protein-RNA binding and protein-protein binding, making them key elements in transcriptional regulation and many other processes. While transcription factors have long been recognized for their required role in yeast α recombination hotspots (Petes 2001), the discovery of *Prdm9* is the first implication of zinc finger
proteins and their predicted binding sequence motifs as major determinants of recombination hotspot location and usage in multi-cellular organisms (Segurel et al. 2011).

Yeast and mammals share many conserved meiotic proteins and processes. However, the Drosophila meiotic recombination process differs from yeast and mammals in several key components. First, homologous chromosome pairing and synapsis proceed normally in the absence of double strand breaks in Drosophila (Hawley et al. 1992; Roeder 1997; Keeney 2001; Hawley et al. 2002; McKim et al. 2002; Page et al. 2003). Indeed, Drosophila males undergo meiosis without any homologous recombination, a phenomenon rarely seen in other animals (Morgan 1910; Rasmussen 1977). Furthermore, the synaptonemal complex, a proteinaceous structure that binds homologs together during meiotic prophase, is conserved in structure but has diverged in function in Drosophila. The Drosophila synaptonemal complex does not require SPO11 to form, and functions in the initiation of recombination and the facilitation of DSBs. Finally, Drosophila lack some genes known to be crucial in recombination in other organisms, like Msh4 and Msh5, and use a smaller subset of proteins in DNA repair (McKim et al. 2002; Schurko et al. 2008). With known differences in meiotic proteins and some apparent differences in the initiation of recombination, it is unclear if Drosophila would possess a recombination initiation process involving a protein like PRDM9.
In 2011, Lake et al. demonstrated that the Cys:His:zinc finger protein *trade embargo* is required for meiotic DSB initiation in *Drosophila melanogaster* and suggested its essential role for processing DSBs into crossovers (Lake et al. 2011). However, while *Prdm9* binds to discrete sites across the genome, *trade embargo* appears to bind the entire length of the chromatin, casting doubt on the similarity between the two proteins. Nonetheless, the discovery of *trade embargo*’s role in DSB initiation and resolution implicates a general role for zinc finger proteins in the distribution of recombination.

Here, we explore the importance of Cys:His:zinc finger genes in the initiation of Drosophila recombination and whether the abundance of predicted binding sites of such genes may correlate with recombination variation within and among species. First, we attempt to identify a *Prdm9* homolog in Drosophila, and confirm that *Prdm9* is indeed missing in this lineage. We then attempt to characterize any other zinc finger proteins involved in recombination by analyzing associations between predicted DNA sequence motifs and our empirically derived broad-and fine-scale measures of recombination rate in *D. pseudoobscura* and its close relative. As a validation of our approach, we apply the same procedure to the *Prdm9*-predicted motif to comparably scaled measures of recombination rate in humans. Our results suggest that Drosophila possess a recombination initiation mechanism disparate from human *Prdm9*. 
Materials and Methods

System and Datasets

The species Drosophila pseudoobscura was selected due to the availability of high resolution recombination data not yet available in Drosophila melanogaster. Additionally, the availability of recombination data in closely related species D. miranda facilitates inter-specific comparisons. D. pseudoobscura inhabits the western coast of North America and diverged from D. melanogaster approximately 55 mya (Tamura et al. 2004). The recombination datasets for D. pseudoobscura consist of two recombination maps from the Flagstaff population (collected Flagstaff, AZ 1997), for more information about the recombination maps, see McGaugh et al. (2012)(McGaugh et al. 2012). The “superfine” recombination map consists of three ~100kb regions on Chromosome 2 with markers spaced within these regions every 20kb (219 crossovers observed). Briefly, the map was constructed by genotyping over 10,000 F2 backcross progeny by PCR at 19 markers. Their coordinates on chromosome 2 are: 6.003 Mb - 6.108 Mb (6 markers, 5 intervals, average interval 20.280 Kbp), 17.534 MB - 17.660 MB (7 markers, 6 intervals, average interval Kbp 20.878), 21.438 Mb -21.537 Mb (6 markers, 5 intervals, average interval 19.870 Kbp). The “broad-scale” recombination map spans the majority of Chromosome 2 with markers approximately every 180 kb, resulting in 140 intervals (1344 crossovers observed) (McGaugh et al. 2012). The broad-scale map was constructed by genotyping
1440 individual backcrossed flies for 384 line-specific SNP markers using the Illumina BeadArray platform (Fan et al. 2003) (Illumina, San Diego, California, United States). For comparisons across species, a “broad-scale” D. miranda recombination map of Chromosome 2 was used. The D. miranda broad-scale map was constructed using the same method as the D. pseudoobscura broad-scale map, and SNP markers were designed at comparable physical coordinates. For the between-species comparison, both the D. pseudoobscura and D. miranda recombination maps were condensed to comparable interval sizes, yielding 97 windows of about 320kb (McGaugh et al. 2012). Chromosome 2 is 30 Mb and makes up 23% of the physical genome.

DNA sequence for the strains corresponding to the recombination maps was also obtained from McGaugh et al. (2012) (McGaugh et al. 2012). We obtained the amino acid sequence for Cys2His2 zinc finger proteins for D. melanogaster and D. pseudoobscura from FlyBase (Lyne et al. 2007; McQuilton et al. 2012), for D. persimilis from FlyMine (Lyne et al. 2007), and for D. miranda from our own sequence data (McGaugh et al. 2012).

Identification of a Prdm9 homolog using BLAST

We used NCBI BLAST protein tools blastp and PSI-BLAST and the nucleotide tool blastn with default parameters, specifying the organism as Drosophila (Altschul et al. 1990). For the input query, we examined all genes and proteins annotated as Prdm9, selecting
Homo sapiens, Strongylocentrotus purpatus, and Mus musculus PRDM9 proteins and Prdm9 sequence for input queries.

Motif Prediction

We used custom Perl and Unix scripts to extract the zinc finger domains from each D. pseudoobscura protein using the canonical Cys:His: binding pattern CX(2-6)CX(11-13)HX(2-6)H. Each Cys:His: protein contains a number of zinc fingers ranging from one to 21, with an average of five. We used a protein only if it had more than one zinc finger, obtaining binding sequences longer than 3 base pairs. This procedure resulted in an amino acid dataset of 186 unique D. pseudoobscura proteins. For each zinc finger, we recorded the amino acid residues at positions -1, 3, and 6 in relation to the start of the alpha helix, which are responsible for predicting DNA binding specificity (Kaplan et al. 2005). To generate the DNA sequence that these amino acids are predicted to bind to, we used two approaches.

To examine candidate proteins containing a SET domain, those identified in our BLAST searches, or ontologically identified as functioning in meiosis, we used the rigorous approach of Baudat et al. (2010) to generate the sequence motif for Prdm9. Briefly, we used the Zinc Finger Consortium database to obtain a matrix of binding residues, positions, and empirically determined binding sequence (Fu et al. 2009), then input this data into WebLogo to generate the sequence motif (Crooks et al. 2004). To take
into account that not all zinc fingers may be important in binding, we used a 3 letter sliding window for the DNA sequence motifs, looking at the whole motif and all possible contiguous 9 bp motifs from the whole motif. This approach was applied to zinc finger proteins GA18168 (trade embargo), GA23469 (Blimp-1), GA25755 (hamlet), GA26409 (CG9817), GA25849 (crooked legs), GA26228 (CG5245), GA26117, GA21024 (combgap), GA21437 (teflon), and GA17308 (grauzone) (Table 8).

For all other zinc finger proteins, we used the more scalable program enoLOGOS, with default parameters (Workman et al. 2005). The input for this program simply requires the amino acid contact residues for each zinc finger. The output is a normalized sequence logo of nucleotides, with the information content of each nucleotide position measured in bits (ranging from zero to two). Again, we used a 3 letter sliding window for the DNA sequence motifs, looking at the whole motif and all possible contiguous 9 bp motifs from the whole motif.

Motif occurrence

DNA sequence for D. pseudoobscura Flagstaff was split into intervals of known recombination. This resulted in 140 windows of average size 180kb for the D. pseudoobscura Flagstaff broad-scale dataset and 16 windows of average size 20kb for the D. pseudoobscura Flagstaff superfine-scale dataset. To identify the frequency of occurrence of all D. pseudoobscura zinc finger motifs, we used the EMBOSS command
“dreg” (Rice et al. 2000). The command “dreg” searches one or more sequences with the supplied regular expression and writes a report file with the matches. The frequency of motifs in a given interval for forward and reverse strands was combined and corrected for interval size, then regressed with recombination rate using custom Perl and R scripts. p-values were adjusted for multiple comparisons using a sequential Bonferroni correction (Rice 1989). For proteins that were significantly associated with recombination after correction for multiple comparisons, we ran a multiple regression accounting for GC content (JMP Version 9.0. SAS Institute Inc., Cary, NC).

**Amino acid differences between species**

To identify changes in Cys2His2 zinc fingers that alter DNA binding, we compared number of fingers and amino acids at positions -1, 3, and 6 for each protein in *D. melanogaster*, *D. miranda*, and *D. persimilis* to *D. pseudoobscura* using a custom Perl script. After identifying proteins that had differences in their zinc fingers between *D. pseudoobscura* and *D. miranda*, we followed the protocol outlined in the Motif Occurrence section above, but using the condensed *D. miranda* recombination data and sequence with this subset of proteins. The frequency of predicted motifs for this subset of *D. miranda* zinc finger proteins was identified using *D. miranda* recombination and sequence and *D. pseudoobscura* recombination and sequence, and then the correlation

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coefficients compared. The same was done for predicted motifs for the same subset of D. pseudoobscura zinc finger proteins.

**Candidate motif analysis**

To identify any overrepresented sequence motifs not *a priori* associated with zinc finger binding, we used the EMBOSS command “wordcount,” which counts and extracts all possible unique sequence words of a specified size in one or more DNA sequences. This analysis was done using a word size of six with the superfine-scale and broad-scale recombination datasets. To identify associations with recombination rate, the forward and reverse complement motif counts were combined and the motifs with the highest frequency difference between the highest and lowest recombination intervals were noted. Following Cirulli *et al.* (2007), the two windows (six windows for the broad-scale) used were excluded and the frequency of the subset of motifs was regressed using the remaining windows. Results were corrected for multiple comparisons using a sequential Bonferroni correction. Additionally, we analyzed the human motif CCNCCNTNNCCNC (Myers et al. 2008; Stevison et al. 2010) and the *D. melanogaster* motif GTGGAAA (Miller et al. 2012) using the approach described above in the Motif Occurrence section above.

**Human comparison**
We obtained human recombination data from Kong et al. (2002) and genome sequence from a Finnish population, a part of the 1000 genomes project (Altshuler et al. 2010). As above, the sequence was partitioned into intervals of known recombination across human chromosome 1 (used because of its large size). Using the same EMBOSS script “dreg,” motif frequency of the 13-mer degenerate motif CCNCCNTNNCCNC (Myers et al. 2008; Baudat et al. 2010) was tallied and a regression looking at motif frequency corrected for interval size and recombination rate was performed. Recombination intervals used for the regression were restricted to the same number of windows and similar recombination range of our *D. pseudoobscura* recombination data (Number of intervals = 140 for both datasets; *D. pseudoobscura* cM range: 0.079-3.97, mean: 0.765, median: 0.487; Human cM range: 0.142-3.11, mean: 0.693, median: 0.505).

**Results**

**Prdm9 homology**

Oliver *et al.* (2009) suggested that, although *Prdm9* is essential for fertility in mice, it appears to be absent in *Drosophila melanogaster* and its function in meiosis may be lineage or even species-specific. Previous studies of the PRDM protein family support this conclusion (Fumasoni et al. 2007; Fog et al. 2012). To confirm that this protein is
Indeed missing in the Drosophila genus, we BLASTed Prdm9 and PRDM9 against Drosophila species. Using human, sea urchin, and mouse protein input queries with the BLAST tools blastp and PSI-BLAST, we identified the genes GA26117, CG5245, crooked legs, meics, combgap, CG9817, Blimp-1, and trithorax-related. All Drosophila proteins identified using BLAST contained zinc finger domains, and CG9817, Blimp-1, and trithorax-related contained SET domains. The maximum amino acid sequence identity ranged between 49% and 38% and the part of the query sequence that was covered ranged between 98% and 73%. Nucleotide input queries using blastn yielded results with a maximum identity between 84% to 97%, but the query only covered between 1% to 13% of the nucleotide sequences of the surveyed genes. These BLAST results, combined with previous data, suggest there is not a Prdm9 homolog detectable in Drosophila. However, genes identified in this manner, which are proteins that possess SET domains and/or zinc fingers, are candidates that may function similarly to PRDM9.

<table>
<thead>
<tr>
<th>Gene name (D. melanogaster homolog)</th>
<th>Sequence Motif</th>
<th>Protein Domains or notes</th>
</tr>
</thead>
<tbody>
<tr>
<td>GA23469 (Blimp-1)</td>
<td>TGA[TG]ANGGA[GT]AA</td>
<td>SET domain, 4 zinc fingers</td>
</tr>
<tr>
<td>GA25755 (hamlet)</td>
<td>GAAGATGAGGAAANN[TG][CT]NNC</td>
<td>SET domain, 7 zinc fingers</td>
</tr>
<tr>
<td>GA26409 (CG9817)</td>
<td>NCTTA[AT]NGAGAN[TG]N[TC]</td>
<td>SET domain, 5 zinc fingers</td>
</tr>
<tr>
<td>GA25849 (crooked legs)</td>
<td>GAC[TG]GNNA[TC]GGGGGGGGGGGGGGG</td>
<td>15 zinc fingers</td>
</tr>
</tbody>
</table>
GA26228 (CG5245) 
\[[\text{GT}]][\text{TC}]\text{CGNGGGGTNCTNC}\]
6 zinc fingers

GA26117 
\[\text{A}[\text{TG}][\text{CT}]\text{GNNTC[CT][GC][CT]}][\text{GT}][\text{GC}]\text{ATTNNTNCAN[TC][TG]}\text{GANG[TC]GA[TC]}\]
11 zinc fingers

GA21024 (combgap) 
\[\text{NN}[\text{CT}][\text{TG}][\text{TC}]\text{NN[CT][TNACGNGNG}]\text{A}[\text{TG}]\text{G[TC][TG]}\text{G[TC][TG]}\text{N}[\text{TC}[\text{TG}][\text{G[TC]}\]
10 zinc fingers

GA18168 (trade embargo) 
\[\text{TGGNAANGCGCCG[CG]}\text{ACNT}\]
5 zinc fingers; meiotic protein

GA21437 (teflon) 
\[\text{GNNGNNG[TC][TC]}\]
3 zinc fingers; meiotic protein

GA17308 (grauzone) 
\[\text{NANGNN[TC][TC]NNACG[TC][C[TG]}[\text{TC}]\text{GN[TC]NGNC}\]
8 zinc fingers; meiotic protein

**Candidate protein analysis**

To identify if any proteins function in Drosophila recombination in a similar manner to PRDM9 in humans, we selected a subset of *D. pseudoobscura* Cys$^2$His$^2$ zinc finger proteins as candidates. In addition to those Cys$^2$His$^2$ zinc finger proteins identified using BLAST above, annotated SET domain proteins and proteins involved in meiotic recombination in Drosophila were targeted (Lyne et al. 2007) (Table 8). Cys$^2$His$^2$ zinc finger DNA binding residues are determined by amino acids at positions -1, 3, and 6 in relation to the start of the alpha helix (Kaplan et al. 2005). We recorded the amino acid binding residues for all BLAST, SET domain, and meiotic protein candidates, and obtained the predicted nucleotide targets using the approach taken to identify the
binding preferences of PRDM9 (Baudat et al. 2010). Once a consensus motif was established for each protein, a 3 base pair sliding window of 9 base pairs was used for each motif, as the binding length for a protein with N fingers is 3N and not all zinc fingers may be used in binding. Motif occurrence was then analyzed using two *D. pseudoobscura* sequence and recombination datasets: a “superfine-scale” recombination map and a “broad-scale” recombination map. The superfine-scale dataset surveys recombination in 16 intervals of approximately 20 kb in size over 3 Mb of chromosome 2. This dataset was constructed using over 10,000 individuals and contains 219 observed crossover events. The broad-scale dataset estimates recombination in 140 intervals of approximately 180 kb in size across all of chromosome 2 (30 Mb) (McGaugh et al. 2012). This dataset was constructed by genotyping approximately 1400 individuals at 384 markers across the genome and captured 1344 crossover events.

Motif frequency was regressed with recombination rate, and after correcting for multiple comparisons, no motifs were significantly associated with recombination at the superfine scale, and three sequence motifs were significantly associated with recombination at the broad scale. These proteins are *crooked legs* (p=0.003, r=0.331), which functions in lateral inhibition, cell adhesion, and negative regulation of transcription; *GA26117* (p=0.019, r=-0.296), of unknown biological function; and *combgap* (p=0.0253, r=-0.290), which functions in imaginal disc-derived wing morphogenesis. These results, while significant, are not particularly compelling due to the high repeat content in the
crooked legs motif, and the highly degenerate nature of the other two motifs (Table 9). Of note, trade embargo was not significantly associated with recombination at either scale, providing support that it may not bind to discrete foci (Lake et al. 2011).
### Table 9: Proteins with motifs significantly associated with recombination

<table>
<thead>
<tr>
<th>Gene (D. melanogaster homolog)</th>
<th>Sequence Motif</th>
<th>p, r (GC content corrected p, r)</th>
<th>Differences between zinc fingers in comparison to D. pseudoobscura</th>
</tr>
</thead>
<tbody>
<tr>
<td>GA25849 (crooked legs)</td>
<td>GGGGGGGGGG</td>
<td>Broad: 0.0034, 0.3307 (0.0003, 0.3321) Superfine: 0.9106, -0.0306 (0.8514, 0.1564)</td>
<td>No differences in mel, per, mir</td>
</tr>
<tr>
<td>GA26117</td>
<td>TNNTNCAN[TC]</td>
<td>Broad: 0.0192, -0.2963 (0.0015, 0.3013) Superfine: 0.3510, -0.2497 (0.3050, 0.4086)</td>
<td>NA</td>
</tr>
<tr>
<td>GA21024 (combgap)</td>
<td>[TG][TC]NN[CT]TNA C</td>
<td>Broad: 0.0253, -0.2902 (0.0017, 0.2979) Superfine: 0.3241, 0.2635 (0.5318, 0.3043)</td>
<td>No differences in mel, mir, or per</td>
</tr>
<tr>
<td>GA14083 (CG1663)</td>
<td>NTTACTNT[CT]</td>
<td>Broad: 0.3370, -0.2893 (0.0025, 0.2894) Superfine: 0.005, 0.6644 (0.0227, 0.6644)</td>
<td>7 mismatches in mel, no differences in per or mir</td>
</tr>
<tr>
<td>GA11272 (CG11906)</td>
<td>NTTGNG[TG]NTATT</td>
<td>Broad: &gt;1.0, -0.1061 (0.3842, 0.1178) Superfine: 0.0080, 0.6362 (0.0343, 0.6362)</td>
<td>3 mismatches in mel, no differences in mir, 1 extra finger in per</td>
</tr>
<tr>
<td>GA18222 (CG4496)</td>
<td>GAT[TG]CNGAT</td>
<td>Broad: &gt;1.0, 0.0389 (0.7787, 0.0604) Superfine: 0.0136, 0.6019 (0.0530, 0.6030)</td>
<td>2 mismatches in mel, no differences in per or mir</td>
</tr>
<tr>
<td>GA11481 (pebbled)</td>
<td>GTTGCTNGN</td>
<td>Broad: &gt;1.0, 0.0248 (0.8067, 0.0560) Superfine: 0.0165, 0.5885 (0.0573, 0.5965)</td>
<td>Unknown in mel, no differences in mir, 6 fewer fingers in per</td>
</tr>
<tr>
<td>GA18373 (CG4707)</td>
<td>[TG]ATNCGGTT</td>
<td>Broad: &gt;1.0, -0.1866 (0.0571, 0.2023) Superfine: 0.0259, 0.5542 (0.0918, 0.5545)</td>
<td>2 mismatches in mel, no differences in per or mir</td>
</tr>
</tbody>
</table>
| GA15299 (CG2202) | N[G]GGGGGGG | Broad: <0.0001, 0.4723 (<0.0001, 0.4702)  
Superfine: 0.8288, -0.0588 (0.8399, 0.1627) | 1 mismatch in mel, 3 fewer fingers in per, no differences in mir |
| GA21173 (su(Hw)) | [CA][CT]TNAG[GC]T | Broad: <0.0001, -0.4444 (<0.0001, 0.4488)  
Superfine: 0.2679, 0.2946 (0.8471, 0.1588) | 1 mismatch in mel, 1 fewer finger in per, missing sequence in mir |
| GA12131 (zfh1) | GTTANNNTN | Broad: 0.0050, -0.3676 (<0.0001, 0.3678)  
Superfine: 0.7078, 0.1017 (0.8518, 0.1561) | 1 extra finger in mel no differences in per, missing sequence in mir |
| GA22134 (CG9932) | NNTANN[GC][TC]N | Broad: 0.0083, -0.3592 (<0.0001, 0.3621)  
Superfine: 0.4855, -0.1881 (0.5491, 0.2968) | 3 mismatches in mel, no differences in per or mir |
| GA14502 (Oaz) | [GC]TTANNGNN | Broad: 0.0166, -0.3474 (0.0001, 0.3535)  
Superfine: 0.1056, -0.4197 (0.3145, 0.4037) | Unknown in mel, no differences in per or mir |
| | TNTT[CA][GA]G | Broad: 0.0234, -0.3413 (0.0002, 0.3413)  
Superfine: 0.5369, -0.1668 (0.7461, 0.2099) |
| GA20521 (CG7691) | NACNTN | Broad: 0.0219, -0.3424 (0.0001, 0.3481)  
Superfine: 0.1618, -0.3672 (0.5137, 0.3121) | No differences in mel, per, or mir |
| GA11205 (charlatan) | NNTN[TG]GG[AT]C | Broad: 0.0328, -0.3351 (0.0002, 0.3401)  
Superfine: 0.4588, -0.1995 (0.8200, 0.1734) | 1 mismatch in mel, no differences in per or mir |
| GA11270 (CG11902) | [CA]ATN[TG][GC][A][CT] | Broad: 0.0389, -0.332 (0.0003, 0.3374)  
Superfine: 0.1401, 0.3857 (0.7821, 0.1926) | 6 extra fingers in mel and 5 mismatches, no differences in per or mir |
| GA15842 (CG30431) | NNTATT[GC]NG | Broad: 0.042, -0.3305 (0.0003, 0.3352)  
Superfine: 0.9141, 0.0293 (0.7495, 0.2083) | No differences in mel, per, or mir |
A comprehensive search for a zinc finger binding sequence motif

To determine if any *D. pseudoobscura* Cys$_2$His$_2$ zinc finger proteins are associated with recombination, the amino acid binding residues of all *D. pseudoobscura* zinc finger proteins were recorded, and predicted nucleotide targets were generated using the program enoLOGOS (Workman et al. 2005). Sequence motifs for 186 *D. pseudoobscura* proteins were identified. Again, a 3 base pair sliding window of 9 base pairs was used for each motif, and motif occurrence was determined and analyzed using both the superfine- and broad-scale *D. pseudoobscura* sequence and recombination datasets (McGaugh et al. 2012). The superfine scale recombination dataset was analyzed using all nucleotide sequence and strictly intergenic sequence, while the broad-scale dataset was analyzed using all nucleotide sequence. Intergenic sequence was used for the superfine scale recombination dataset because recombination is known to commonly initiate in intergenic regions in yeast (Petes 2001; Pan et al. 2011).

After correcting for multiple comparisons, no Cys$_2$His$_2$ zinc finger protein analyzed at the superfine scale was significant. This is complicated by the fact that there are 640 comparisons and only 16 windows of known recombination, so power to detect significance is low. At the broad scale, ten proteins were significantly associated with recombination after correction for multiple comparisons (Table 9). Only one protein, GA15299 (CG2202), was positively associated with recombination. The remaining nine proteins were negatively associated with recombination, which may be expected for a
protein like suppressor of Hairy wing (su(Hw)) that functions in the negative regulation of transcription and negative regulation of chromatin silencing, but contradicts the expectation from PRDM9. Furthermore, there was no overlap in zinc finger motifs between the fine scale and broad scale analyses, which casts doubt on the detected associations.

**Differences in motif occurrences do not account for changes in recombination landscapes between closely related species**

PRDM9 is known to be undergoing rapid positive selection, changing both the number of zinc fingers present and the DNA-binding amino acid residues at positions -1, 3, and 6 (Oliver et al. 2009; Ponting 2011). To determine Cys\(^2\)His\(^2\) zinc finger proteins changing rapidly across the Drosophila lineage, we compared number of zinc fingers present and number of changes in binding residues for each Cys\(^2\)His\(^2\) protein in *D. melanogaster* (55 mya), *D. miranda* (3 mya), and *D. persimilis* (0.5-1 mya) to *D. pseudoobscura*.

Between *D. pseudoobscura* and *D. miranda*, a large majority of Cys\(^2\)His\(^2\) zinc finger protein binding residues are conserved. To identify any changes in recombination rate associated with change in binding of Cys\(^2\)His\(^2\) zinc finger proteins, we generated new sequence motif predictions for proteins with mismatches between *D. miranda* and *D. pseudoobscura*. Utilizing a *D. miranda* broad-scale recombination map directly comparable to the broad-scale *D. pseudoobscura* recombination map, we then compared the
association between \textit{D. miranda} binding motifs and \textit{D. miranda} recombination to \textit{D. miranda} binding motifs and \textit{D. pseudoobscura} recombination and then repeated with comparing \textit{D. pseudoobscura} motifs to \textit{D. pseudoobscura} and \textit{D. miranda} recombination. If a protein is involved in recombination, we expect to see a stronger correlation between binding motif and recombination within species than between species (Figure 10).

Comparing all associations with an un-corrected p-value <0.05, no protein was consistently more strongly correlated with recombination within species than between species. Therefore, we conclude that no differences associated with zinc finger binding are responsible for recombination rate changes between these species of \textit{Drosophila}.

<table>
<thead>
<tr>
<th></th>
<th>\textit{D. miranda} Recombination &amp; Sequence</th>
<th>\textit{D. pseudoobscura} Recombination &amp; Sequence</th>
</tr>
</thead>
<tbody>
<tr>
<td>\textit{D. miranda} Predicted motif</td>
<td>correlation coefficient (r) &gt; correlation coefficient (r)</td>
<td></td>
</tr>
<tr>
<td>\textit{D. pseudoobscura} Predicted motif</td>
<td>correlation coefficient (r) &lt; correlation coefficient (r)</td>
<td></td>
</tr>
</tbody>
</table>

Figure 10: Predictions to test if changes in DNA binding motifs between species account for variation in recombination rate between species
**Sequence motifs**

To identify overrepresented sequence motifs without an identified association with zinc fingers, all possible 6 base pair motifs were analyzed for frequency using the superfine and broad scale recombination datasets. Ten motifs with the greatest frequency difference between regions of high and low recombination intervals were selected and regressed with recombination rate. At the broad scale, the motifs AATAAA (p=0.0397, r=-0.178) and CTGCTG (p=0.0539, r=-0.1669) were weakly, negatively associated with recombination and the motifs CTCTCT (p=0.0115, r=0.0115) and TCTCTC (p=0.0126, r=0.2149) were weakly, positively associated with recombination. At the superfine scale, the motifs AAATTT (p=0.0717, r=0.4954) and ACAAT (p=0.0594, r=0.5151) were weakly, positively associated with recombination.

Previous studies in Drosophila found associations between local recombination rates and the human Prdm9 motif CCNCCNTNNCCNC (Stevison et al. 2010), and the D. melanogaster motif GTGGAAA (Miller et al. 2012). In this study, neither of these previously described motifs were significantly associated with recombination rate variation in D. pseudoobscura at either scale, although this lack of association is not unexpected as these motif associations were detected previously in different species.
A validation of our approach using human recombination data

To assess if one can detect an association between a sequence motif and recombination rate using relatively coarse recombination rate estimates, we utilized recombination data from an Icelandic population that empirically surveyed genome wide recombination in 869 individuals (average window size: 650kb). We restricted the dataset to a subset of chromosome one with a comparable recombination range to *D. pseudoobscura*. A regression between the frequency of the human Prdm9 motif CCNCCNTNNCCNC and recombination rate was positive and statistically significant (p=0.0004, r=0.3), thus demonstrating sequence motif signals can be detected in humans with broad-scale recombination data comparable to that used in the Drosophila studies.

Discussion

Our attempts to identify a PRDM9-like protein involved in meiotic recombination initiation in Drosophila yielded negative results. Generating predicted nucleotide sequence motifs from Cys2His2 zinc finger proteins and regressing their frequency with estimated recombination in *D. pseudoobscura* produced a handful of recombination associated sequence motifs, but the biological relevance of these associations remains uncertain. Furthermore, changes in the binding motifs between species do not appear to account for variation in the recombination landscape. Our
results could be complicated by the approach taken, or alternatively, we suggest these findings could be explained by the existence of a different recombination initiation system in Drosophila.

**Approach**

Our results could be due in part to the scale at which recombination was estimated in *D. pseudoobscura*. While the superfine (20kb) and broad scale (180kb) recombination datasets used represent one of the most comprehensive recombination maps outside human, mouse, and yeast, the datasets might still lack the resolution needed to determine sequence motifs associated with recombination. Successful work with sequence motifs in yeast and human recombination has been analyzed at a scale <1-2 kb (Myers et al. 2005; Myers et al. 2008; Steiner et al. 2009; Pan et al. 2011; Steiner et al. 2011), although we were able to detect a strong association here between the frequency of the human *Prdm9* motif and human recombination rate using broader (~650kb intervals), comparable in recombination to what we studied, thereby validating our approach. Furthermore, recombination associated motifs have been identified at scales ranging from 220kb to 5 Mb in other organisms (Shifman et al. 2006; Groenen et al. 2009; Backstrom et al. 2010; Stevison et al. 2010; Wong et al. 2010).

Additionally, there is an inherent limitation in one of the bioinformatic approaches utilized here, in the ability of currently developed programs to accurately
identify DNA binding motifs of zinc fingers. While algorithms have improved over the years, it is impossible to be certain that identified motifs are “correct.” Hence some motifs predicted in this manner could be biologically irrelevant. We attempted to address this problem in two ways. First, for BLAST, SET domain, and meiotic candidate zinc finger proteins, we followed a proven motif prediction protocol: the methods utilized to identify the binding nucleotides for PRDM9 (Baudat et al. 2010). Because of its success in determining the PRDM9 associated sequence motif, we can be somewhat more confident in concluding that our motif predictions for these proteins are correct, and therefore, that no Drosophila candidate proteins we tested are associated with recombination. Second, we used an unbiased approach to identify all motifs of six base pairs in length and to test their association with recombination, although even this approach is imperfect since it is not possible to search for degenerate motifs of all possible lengths. Despite these accommodations, it remains possible that Cys-His: zinc finger protein associated sequence motifs do play a role in Drosophila meiotic recombination, but that it is beyond the scope of the technology to detect them at this point in time.

*A different recombination initiation system in Drosophila?*

Alternatively, it is possible and likely that other factors play a major role in the determination of recombination in Drosophila. Historically, it has been thought that
Drosophila do not have the 1-2 kb hotspots characteristic of yeast, humans, and mice (Aquadro et al. 2001; Nachman 2002; Hey 2004; Coop et al. 2007). This is supported by the lack of apparent hotspots of intragenic recombination in *rosy* (Clark et al. 1988; Hilliker et al. 1991; Radford et al. 2007) and in *white-echinus* (Singh et al. 2009), and the lack of linkage disequilibrium among nearby nucleotides as compared to humans (Aquadro et al. 2001; Ardlie et al. 2002; Nachman 2002). The human *Prdm9* recombination initiation model is based on the specific targeting and binding of the PRDM9 protein to a sequence motif, enriched in recombination hotspots (although this model may need some refining, see below). If Drosophila lack such recombination hotspots, this evidence supports Drosophila lacking a recombination initiation system that functions in a sequence specific binding function like *Prdm9* in humans, although obviously cannot rule out a sequence binding independent function.

Additionally, Drosophila recombination is known to differ from other organisms (Hawley et al. 1992; Roeder 1997; Keeney 2001; Hawley et al. 2002; McKim et al. 2002; Page et al. 2003). First, homologous chromosome pairing and synapsis proceed normally in the absence of double strand breaks. Second, the synaptonemal complex does not require SPO11 to form and functions in the initiation of recombination and the facilitation of DSBs. Third, Drosophila are missing some genes known to be crucial in recombination in other organisms (McKim et al. 2002). With these known differences in meiotic proteins, and apparent differences in the initiation of recombination, this
evidence is supportive of Drosophila possessing a different recombination initiation process than humans.

Furthermore, Prdm9 is missing or altered in many organisms (Fumasoni et al. 2007; Oliver et al. 2009; Ponting 2011), necessitating the existence of alternative recombination initiation systems. The PRDM family is absent in plants and fungi, and is quite small in other taxa, with only two genes in nematodes and three genes in arthropods (D. melanogaster: CG5249 – PRDM1, CG9817 – PRDM5, and hamlet). While PRDM9 functions in meiotic recombination in both mouse and human, it seems as if this function is lineage specific. Prdm9 is non-functional in canines (Munoz-Fuentes et al. 2011; Axelsson et al. 2012), and is missing all zinc fingers in the marsupial Monodelphis domestica (Ponting 2011), so even amongst mammals, recombination initiation may vary.

Finally, the PRDM9 story is made more complex by a general lack of understanding of the in vivo function of PRDM9. In humans, although the PRDM9 motif is only detected in a proportion of hotspots, data suggest that PRDM9 influences hotspot activity even at hotspots in which the motif is absent (Berg et al. 2010; Berg et al. 2011; Segurel et al. 2011). While there is in vitro evidence that the zinc fingers of PRDM9 do bind to the motif, this suggests PRDM9 interacts with hotspots genome wide in a more complex and subtle way than first expected. Furthermore, the predominant human sequence motif is neither necessary nor sufficient to drive hotspot activity in humans, with the motif represented approximately 290,000 times in the genome and only about
50,000 detectable hotspots (Segurel et al. 2011). In chimpanzees, there is extensive variation in the PRDM9 protein, and little evidence of any sequence motifs enriched in hotspots (Auton et al. 2012). Researchers suggest the most plausible explanation for this observation is that PRDM9 may still play the same roles in chimpanzee as it does in mouse and human, but the PRDM9 alleles may bind to a much greater variety of sequence motifs than seen in human. This implies that other factors, like chromatin state, play a more dominant role in the hotspot localization. Taken in this context, our data could suggest that there is a PRDM9-like protein in Drosophila, but it either binds a wide repertoire of sequence motifs, or functions in a sequence-specific-binding independent manner.

Regardless of the model, given recent observations that PRDM9 influences more human recombination hotspots than previously thought, and possibly all hotspots (Segurel et al. 2011), it is quite remarkable that a single protein rapidly evolved to become responsible for all recombination in the human lineage. Recombination is an essential mechanistic and evolutionary process, so Prdm9 poses an intriguing step in the evolution of meiosis. However, Prdm9 appears to be only a piece of the puzzle when looking at recombination across taxa. Evidence from Drosophila and other organisms suggests that Prdm9 is not the quintessential element defining meiotic recombination; instead, there remain many mysteries to explore.
4. No detectable effect of DNA methyltransferase DNMT2 on meiotic recombination

Caiti S. Smukowski Heil

Introduction

Epigenetics has long been predicted to play a role in the initiation of meiotic recombination. Observations of variation in recombination rate within and between individuals, sexes, populations, and across the genome (such as euchromatin versus heterochromatin) suggest a role beyond DNA sequence in determining locations of recombination events (Lichten et al. 1995; Barthes 2011). An association between open chromatin formation and double strand breaks, the first step in the initiation of recombination, has been identified in yeast, dog, and several plants (Berchowitz et al. 2009; Pan et al. 2011; Auton et al. 2013; Choi et al. 2013; Hellsten et al. 2013), and the histone H3K4 methyltransferase PRDM9 influences the distribution of recombination sites in human and mouse (Buard et al. 2009; Baudat et al. 2010; Berg et al. 2011; Grey et al. 2011; Brick et al. 2012; Acquaviva et al. 2013; Sommermeyer et al. 2013). However, the possible roles of specific epigenetic marks apart from H3K4me3 are less understood. For example, the relation between recombination initiation and DNA methylation, the best characterized epigenetic factor, appears complex and remains relatively unexplored.

DNA methylation describes the transferring of a methyl group (CH$_3$) to the 5th position of a cytosine residue, typically at CpG sites and repeat elements (Robertson et
al. 2000). The reaction is catalyzed by a family of conserved proteins known as DNA methyltransferases: DNMT1, the maintenance methyltransferase, ensures proper inheritance of methylation patterning after replication in somatic cells; DNMT3s (3A, 3B, 3L), the de novo methyltransferases, establish DNA methylation patterns during embryogenesis; and DNMT2, an enigmatic methyltransferase with conserved catalytic motifs, has a historically disputed function (Yoder et al. 1997; Dong et al. 2001; Schaefer et al. 2010; Barthes 2011; Krauss et al. 2011). These genes function within an ancient regulatory mechanism shared by animals, plants, and fungi, serving in diverse roles often related to repression of gene expression (Feng et al. 2010; Zemach et al. 2010; Zemach et al. 2010; Jurkowski et al. 2011; Nanty et al. 2011).

A link between DNA methylation and recombination was first hypothesized by Rossignol and Fagueron (Rossignol et al. 1994) and Yoder et al. (Yoder et al. 1997), in which DNA methylation promotes genome integrity through the suppression of recombination between dispersed repetitive sequence. Such ectopic recombination events result in gross chromosomal rearrangements of deleterious consequence (Lichten et al. 1987; Rouyer et al. 1987; Montgomery et al. 1991; Small et al. 1997; Li et al. 2012), and because the prevalence of repetitive elements in a genome is so high, a system must exist to limit recombination between repeats. Indeed, early evidence showed that repeats were enriched for methylation in plants, fungi, and mammals (Whitkus et al. 1992; Bennetzen et al. 1994; Yoder et al. 1997; Maloisel et al. 1998), and the loss of DNMT3A or
DNMT3L resulted in reactivation of retrotransposons, impaired synapsis of homologous chromosomes, failure to progress through meiotic prophase, and sterility (Bourc’his et al. 2001; Bourc’his et al. 2004; Kaneda et al. 2004).

Direct evidence of a functional link between DNA methylation and recombination is somewhat limited, but a study in the fungus *Ascobolus immersus* showed crossover formation was reduced several hundred fold in an *in vivo* methylated hotspot compared to an unmethylated hotspot (Maloisel et al. 1998). In more recent years, several reports in Arabidopsis paint a more nuanced picture. For example, in the absence of MET1 (the DNMT1 homolog), researchers independently observed a pattern of increased recombination in euchromatin, and decreased recombination in (typically hypermethylated, transposon rich) heterochromatin (Melamed-Bessudo et al. 2012; Mirouze et al. 2012; Yelina et al. 2012). Each study found that total number of crossover events was not different between *met1* mutants and wild type, indicating that the loss of DNA methylation affects the distribution of crossovers, but not their overall number. Looking specifically at wild type Arabidopsis transcription start (TSS) and termination sites (TTS), DNA methylation was decreased in recombination hotspots relative to TSS or TTSs in which recombination was absent (Choi et al. 2013), supporting data from *A. immersus*. In contrast, some indirect evidence in humans found a positive association between recombination rate and DNA methylation (Sigurdsson et al. 2009).
In this study, I explore the complex and perhaps contradictory role of DNA methylation in the determination of recombination events in the model system *Drosophila melanogaster*. *D. melanogaster* possesses the DNA methyltransferase DNMT2 and a methyl binding domain protein, MBD2/3, which typically binds to methylated DNA and recruits chromatin remodeling complexes (Tweedie et al. 1999; Roder et al. 2000; Ballestar et al. 2001; Marhold et al. 2004). Transcripts of *Dnmt2* were particularly enriched during early stages of embryonic development; expression in adult flies was limited to female ovaries and there was no activity in male testes (Lyko et al. 2000), perhaps indicating a role in recombination, which is female-specific in Drosophila. Furthermore, like MBD2 knockout mice, Drosophila null mutants of MBD2/3 were viable and fertile, but revealed chromosome segregation defects (Lyko et al. 2006).

DNA methylation was purportedly detected at low levels, ranging from 0.21-0.4% across the Drosophila genus (Tweedie et al. 1999; Gowher et al. 2000; Lyko et al. 2003; Marhold et al. 2004; Salzberg et al. 2004; D’Avila et al. 2010). One of its functions in Drosophila was thought to be retrotransposon silencing and stabilization of repeats, similar to its role in vertebrates and plants (Salzberg et al. 2004; Phalke et al. 2009; Krauss et al. 2011). Although methylation levels may be low (or maybe even virtually absent, see Raddatz et al. 2013), it appears that DNMT2 is functional in a capacity similar to its role in other taxa. For example, in somatic cells, the loss of *Dnmt2* eliminates H4K20 trimethylation at retrotransposons and impairs maintenance of retrotransposon
silencing (Phalke et al. 2009), and Su(var)3-9 (H3-K9 specific histone methyltransferase) mutants, normally associated with inactive regions in the genome such as constitutive and facultative heterochromatin and genome stability, showed dramatic reduction or complete loss of DNA methylation (Lyko et al. 2003).

Therefore, to determine if there is a detectable effect of DNA methylation and/or the DNA methyltransferase Dnmt2 on the distribution and frequency of recombination rate in Drosophila, I assayed recombination at two euchromatic regions on the X chromosome in Dnmt2 -/- and control D. melanogaster. I did not detect any change in recombination rate or distribution in the absence of Dnmt2, and conclude that other factors are determining sites of recombination events in Drosophila.

**Materials & Methods**

**Stocks and crossing scheme**

For all crosses, virgin flies were collected, separated by sex, and aged for 7 days. The crossing scheme (Figure 11) consisted of: (1) crossing a D. melanogaster Dnmt2 p-element excision line Dnmt299(Schaefer et al. 2010) to wild type D. melanogaster Zim29 to generate variability to score recombination events. (2) F1 females were crossed to a D. melanogaster chromosome 2L deficiency line over a balancer, Df(2L)BSC826/SM6a (#27900 Bloomington Stock Center, Bloomington, IN) . F2 females were collected, and females carrying the SM6a balancer were identified by the Cy phenotype and discarded.
(3) The remaining F² females (bearing 0-1 functional copies of *Dnmt2*) were crossed to wild type males (Zim29) in single pair crosses, allowed to lay eggs, and subsequently genotyped after larvae appeared. Undesired genotypes (see below for genotyping methods) were discarded and F³ progeny were collected from remaining vials. Recombination was assayed in these individuals.

**Figure 11: DNA methylation crossing scheme**

**Reverse transcription**

To ensure that individuals of Dnmt2⁹⁹/ Df(2L)BSC826 genotype were *Dnmt2*-/-, reverse transcription PCR was completed. For each line (Dnmt2⁹⁹, Df(2L)BSC826/SM6a, zim29, Dnmt2⁹⁹/ Df(2L)BSC826), I prepped RNA from approximately 25 pooled flies of varying life stages using the Qiagen RNeasy kit, Qiagen QIAShredder kit, and Qiagen
DNase kit. For each line, we used genomic DNA, H₂O, and a negative control reverse transcription product as controls. For the reverse transcription reaction, the following recipe was used: 2µl 10X PCR buffer (15mM MgCl2), 1.5µl 50mM MgCl2, 0.8µl 25mM dNTPs, 1µl 50µM experimental R primer, 1µl 50µM control intron R primer, 0.5µl 40U/µl RNase inhibitor, 0.1µl 200U/µl MMLV reverse transcriptase, 9.1µl DEPC H₂O, and 4µl RNA mix + H₂O to get to 500ng RNA. The reverse transcription program consisted of: 15 minutes at 42°C followed by five minutes at 99°C. Then a PCR was completed with the following recipe: 2.5µl 10X PCR buffer (15mM MgCl2), 1.5µl 2mM dNTPs, 1.25µl, 10µM experimental F primer, 1.25µl 10µM experimental R primer, 0.3µl DNA polymerase, 16.2µl H₂O, and 2µl reverse transcription reaction product. The PCR program consisted of an initial denaturing step of 95°C (60 sec), three touch-down cycles of 94°C (30 sec)- 56°C (30 sec)- 72°C (45 sec) each, three touch-down cycles of 94°C (30 sec)- 53°C (30 sec)- 72°C (45 sec) each, followed by 33 main cycles of 94°C (30 sec)- 50°C (30 sec)- 72°C (45 sec) each. We used the trade embargo (TREM) gene as a control with the following primers: Forward: CAGTAAGTGTGAATCCTGCTTGGTTTGC; Reverse: GCATGTCCATAATGTGCTGATGGGATC. The primers used for flanking the Dnmt2 gen intron were: Forward: GGTCCTAGAATCTATTTAGTGGCATGGGGC; Reverse: TAATTGTGCACAAACCCGATTTGGC.

**Scoring recombination**
Flies were collected in 96 well plates and frozen at -20°C. DNA was extracted following the protocol of Gloor and Engels (Gloor 1992), using 49.5ul squish buffer (10mM Tris-HCl (pH 8.2), 1mM EDTA, 25mM NaCl) + 0.5ul proteinase-K. A zirconium bead was placed in each well, and plates were then shaken using a Qiagen TissueLyser II for 45 seconds. The DNA preps were then incubated at 37°C for 30 minutes and 95°C for 2 minutes in a thermal cycler. The PCR recipe consisted of 0.5 uL of forward primer +M13 tag (CACGACGTTGTAAAACGAC added to 5’ end of forward primer) , 0.5 uL of reverse primer, 0.4 uL of 700IRD or 800IRD-labeled M13 tag, 1.3 uL of MgCl2, 10 X buffer, 1 uL 2 mM dNTPs, and 0.2 ul of Taq polymerase in a 10 uL reaction volume. The same PCR program was used as above. Products were visualized on a polyacrylamide gel using a LICOR 4300.

To identify crosses with desired genotypes, F2 females were genotyped at three loci on the X chromosome and two loci on Chromosome 3 (Table 10) to ensure heterozygosity (Df(2L)BSC826/zim29) across the region of interest, and at one locus on chromosome 2L (Table 10) to identify if the genotype was Dnmt2+/ Df(2L)BSC826 (Dnmt2/-) or Df(2L)BSC826/zim29 (control). Vials from parents of the desired genotypes were kept and the F3 progeny were collected, all other vials and their progeny were discarded.

Recombination was scored in F3 progeny by genotyping at the same three markers on the X chromosome and two markers on the 3rd chromosome (Table 10). The
markers on the X chromosome delineate 2 regions, one of lower recombination (~1.8 cM/Mb) and one of higher recombination (~3.7 cM/Mb). A recombinant was called when an individual fly’s genotype changed from heterozygous to homozygous or vice versa for females, and when the fly’s genotype changed between the possible allele combinations for the males. In total, 1,536 F3 control progeny and 1,177 F3 experimental progeny were scored for the euchromatic regions.

Table 10: Markers used to assay recombination

<table>
<thead>
<tr>
<th>Marker</th>
<th>Primer name</th>
<th>Physical distance/genetic distance between markers</th>
<th>Sequence (F/R)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Marker X1</td>
<td>DMELX_494471F DMELX_494621R</td>
<td>Marker X1-Marker X2: 2.75 Mb/5 cM</td>
<td>CGACGGCCTGTCTATTGCGTTC TCAATTCAATTCGATTTGGAGTCGGC</td>
</tr>
<tr>
<td>Marker X2</td>
<td>DMELX_3240050F DMELX_3240200R</td>
<td>Marker X2-Marker X3: 2.68 Mb/10 cM</td>
<td>GGAAACAGTGTATTGCCTACACATGG AAC CTTGCGCAAGTTCACATGAGATAC</td>
</tr>
<tr>
<td>Marker X3</td>
<td>DMELX_5922532F DMELX_5922673R</td>
<td>Marker X1-X3: 5.43 Mb/15 cM</td>
<td>GGATCGTTGCAGATCGGATAGAACTC CCGTCTCAATTGATGGACGCCTAT</td>
</tr>
<tr>
<td>Marker Chr2</td>
<td>DMEL2L_12024260F DMEL2L_12024434R</td>
<td>NA</td>
<td>CGTCACATTCTTCAATTGGAACGACTTTCCG CAAAACGCTGCTCAAACGTCCTG</td>
</tr>
</tbody>
</table>

Statistics

Recombination fractions between experimental and control individuals were compared using an unpaired t-test (GraphPad Software, Inc. La Jolla, CA). A power
Results

Recombination across a euchromatic region

To identify an effect of the DNA methyltransferase gene Dnmt2 on meiotic recombination, we created a variable Dnmt2 null fly stock and identified recombinants in two adjacent euchromatin regions on the X chromosome. Region 1 spans assembly positions 494,471 to 3,240,200 bp (Adams et al. 2000) and represents a region of relatively low recombination (2.75 Mb, approximately 5 cM), while Region 2 spans assembly positions 3,240,200 to 5,922,673 bp and represents a region of relatively high recombination (2.68 Mb, 10 cM). We chose this delineation to account for possible changes in the distribution of recombination events in euchromatin in the absence of Dnmt2.

In total, 1536 control (Dnmt2+/-) individuals and 1177 Dnmt2-/- individuals were scored. In Region 1, we identified no significant difference between Dnmt2 -/- and control individuals (p=0.86; Control: 3.02% recombinant, 1.07 cM/Mb; Experimental: 3.16% recombinant, 1.11 cM/Mb). Results were similar for Region 2 (p=0.84 ;Control: 10.58% recombinant, 3.57 cM/Mb; Experimental: 8.98% recombinant, 3.07 cM/Mb). A
power analysis showed with our sample sizes, we could detect an effect size of $d=0.1$ (power=0.8, significance level 0.05).

**Discussion**

Although DNA methylation influences the recombination landscape in fungus *A. immersus* and Arabidopsis, we detected no effect of knocking out the DNA methyltransferase *Dnmt2* on recombination in Drosophila. Our crossover analyses were limited to two genomic regions, but these regions captured both low- and high-recombination areas of euchromatin. As such, DNMT2 appears to have no major effect on rates of recombination in specific regions of the genome, and DNA methylation more generally may also not affect recombination in Drosophila.

The presence of DNA methylation in Drosophila and other *Dnmt2*-only systems is a long debated issue (Schaefer et al. 2010; Raddatz et al. 2013). While *Dnmt2* shows strong sequence and structural conservation to established methyltransferases, the enzymatic activity was found to be much weaker (Okano et al. 1998; Dong et al. 2001). Various studies reporting DNA methylation in Drosophila (Tweedie et al. 1999; Gowher et al. 2000; Lyko et al. 2000; Lyko 2001; Lyko et al. 2003; Lyko et al. 2003; Marhold et al. 2004; Marhold et al. 2004; Salzberg et al. 2004; Phalke et al. 2009; Krauss et al. 2011) may have been confounded by contaminations from other organisms, detection limits, low antibody specificity, and/or false positives (Zemach et al. 2010; Raddatz et al. 2013).
More advanced bisulfite sequencing with thorough controls revealed no detectable levels of DNA methylation in Drosophila (Raddatz et al. 2013). Our finding that there is no effect of the gene \textit{Dnmt2} on meiotic recombination supports this recent data, and points to other epigenetic mechanisms directing recombination in Drosophila. Indeed, there is some evidence that histone modifications in Drosophila may mimic the role of DNA methylation in transcriptional processes in other invertebrates (Cedar et al. 2009; Chodavarapu et al. 2010; Nanty et al. 2011; Hunt et al. 2013). Additionally, although DNA methylation was reportedly involved in specific transposons in Drosophila (Phalke et al. 2009), the small RNA Piwi-piRNA pathway is known to be the main genome defense system against repetitive elements in the germline (Aravin et al. 2007; Brennecke et al. 2007; Blumenstiel 2011).

Irrespective of its lack of clear role in recombination, new data show that \textit{Dnmt2} may still play a role in the control of RNA viruses in an unanticipated way: a necessity of \textit{Dnmt2} for acute immune responses after RNA virus infection and the presence of RNA virus genomes in \textit{Dnmt2} protein complexes points to a role in innate immune response, perhaps via RNA methylation (Durdevic et al. 2013; Durdevic et al. 2013; Durdevic et al. 2013). Goll \textit{et al.} (2006) first showed that \textit{Dnmt2} methylates cytosine 38 of tRNA\textsuperscript{Asp}, and while at that time it was thought to also methylate DNA, RNA may perhaps be its only genuine substrate (with possible “star activity:” low enzymatic activity with relaxed substrate specificity on DNA substrates). It’s likely that the tRNA
methylation mechanism is conserved between DNA methyltransferases, and phylogenetic analyses support the idea that DNMT2 is a non-canonical member of the DNA methyltransferase family, not an RNA methyltransferase (Jeltsch et al. 2006; Jurkowski et al. 2008; Jurkowski et al. 2011). While Dnmt2’s biological function is still being explored, observations in Drosophila and other organisms indicate that Dnmt2 mediates assorted forms of “RNA stress,” triggered by events such as an excess of retrotransposons or RNA viruses (Lin et al. 2005; Phalke et al. 2009; Schaefer et al. 2010; Becker et al. 2012; Tuorto et al. 2012; Durdevic et al. 2013; Durdevic et al. 2013). Finally, new evidence in mice and Drosophila has shown that Dnmt2 is required for transgenerational RNA-mediated epigenetic heredity via paramutation, maintaining Dnmt2’s function as an epigenetic factor (Kiani et al. 2013).

Clearly Dnmt2 research has experienced a tumultuous twenty years, with some long debated questions answered (e.g. (Raddatz et al. 2013)), and many new ones posed (e.g. (Durdevic et al. 2013; Kiani et al. 2013)). Whether DNA methylation influences the recombination landscape in organisms besides the fungus A. immersus and Arabidopsis is one question that remains to be elucidated, but I conclude based on the available results that it has no role in Drosophila. While both recombination and the Dnmt2 phenotype are known to be affected by environment and stress (Stevison 2009; Durdevic et al. 2013), and though it remains possible that Dnmt2 methylates DNA in stress conditions or via unknown environmental cues, this seems unlikely (Krauss et al. 2011;
Raddatz et al. 2013). Future investigation of such a role should be pursued in organisms possessing DNMT1 and DNMT3.
5. The fine scale recombination landscape in two species of Drosophila

Caiti S. Smukowski Heil, Edwin Iversen, and Mohamed Noor

Introduction

Recombination is a key evolutionary process, affecting adaptation, speciation, sex chromosome evolution, and the distribution of genomic features. Sexual reproduction is thought to persist in part because homologous meiotic recombination breaks down linkage and permits sites to behave independently, theoretically enhancing the efficacy of selection, allowing the purging of deleterious variants, and accelerating adaptation (Smith et al. 1974; Kondrashov 1988; Otto et al. 2001; Otto et al. 2002; Haddrill et al. 2007; Barton 2010). Theoretical predictions are supported by observations of sex chromosomes, where deleterious mutations accumulate and cause the genome to deteriorate in regions of no recombination (Charlesworth et al. 2000; Bachtrog et al. 2002; Bachtrog 2003), and experimental studies, in which organisms with recombination experience greater rates of adaptation when selection is strong (Colegrave 2002; Goddard et al. 2005; Morran et al. 2009; Webster et al. 2012).

Recombination both influences the efficacy of selection, and can be selected upon (Chinnici 1971; Charlesworth et al. 1985). Which selective pressures act on recombination is dependent upon the genomic scale examined. At broad scales, one of the most striking general patterns of recombination is the necessity for at least one
crossover per chromosome to ensure proper segregation of chromosomes during meiosis (Roeder 1997; Sapienza 2001; Dumont et al. 2008; Fledel-Alon et al. 2009). The physical link created during a crossover event provides tension that is needed for the spindle to pull the correct chromatids to the poles (Roeder 1997); breakdown of this process can cause aneuploidy and cell death. This mandates a minimum level of recombination, as a crossover rate that is too low is likely to be highly deleterious.

Contrarily, too much recombination breaks apart advantageous allele combinations and threatens genomic integrity (Kauppi et al. 2004; Coop et al. 2007). Ectopic recombination, or recombination events that occur between non-homologous regions, such as repetitive sequence, can result in gross, deleterious chromosomal rearrangements and disease (Inoue et al. 2002). Therefore, on a broad scale, recombination is likely constrained between a lower bound dictated by chromosome segregation, and an upper bound determined by maintaining genomic integrity. Studies of recombination rate variation across species support this broad level of constraint: recombination rates on the order of tens of kilobases up to genome wide genetic maps show conservation between closely related species (Ptak et al. 2005; Dumont et al. 2008; Paigen et al. 2008; Backstrom et al. 2010; Meznar et al. 2010; Stevison et al. 2010; Smukowski et al. 2011; McGaugh et al. 2012).

Within these bounds, where a crossover occurs on a chromosome is non-random. Observations across many species, such as yeast, Drosophila, mice, and human reveal
chromosomal domains of high and low recombination, commonly with reduced recombination around centromeres. Evidence in humans suggests that recombination too close to the centromere or telomere elevates rates of non-disjunction (Hassold et al. 2001). Apart from chromosomal domains, recombination is also influenced at intermediate scales by crossover interference. Crossover interference describes the procedure in which a crossover event reduces the probability of a second such event in its proximity, which leads to a more equal distribution of crossovers throughout the genome and likely prevents non-disjunction as well (Chua et al. 1997; Kaback et al. 1999; Broman et al. 2000; Broman et al. 2002; Bishop et al. 2004).

Finally, in most organisms observed, recombination takes place in discrete genomic regions called “hotspots,” typically 1-2 kilobases in length in which recombination increases a magnitude fold above the background rate. At fine scales, the recombination landscape is shaped by a phenomenon known as the “hotspot paradox” (Boulton et al. 1997; Jeffreys et al. 2002; Jeffreys et al. 2005). The hotspot paradox describes the persistence of hotspots despite a mechanistic process that results in their loss. More specifically, when a double strand break occurs to initiate a recombination event, the initiating allele is repaired using the non-initiating allele as a template, thereby converting active alleles to inactive alleles, which should lead to the disappearance of hotspots over time. A method to counteract the loss of hotspots was unknown until the recent discovery of the zinc finger histone-methyltransferase, Prdm9,
which controls the distribution of recombination in mice and humans (Baudat et al. 2010; Parvanov et al. 2010). The amino acids that determine DNA-binding of Prdm9 are under rapid, positive selection, and minisatellite shuffling has added, subtracted and rearranged the order of its’ zinc fingers, thereby allowing the creation of new populations of hotspots (Oliver et al. 2009; Baudat et al. 2010; Berg et al. 2010; Myers et al. 2010; Berg et al. 2011; Brick et al. 2012).

Organisms possessing Prdm9 should see a rapid turnover of hotspots, and indeed, differences in DNA-binding appear to have created fine scale recombination rate divergence between individuals, populations, and species (Ptak et al. 2004; Ptak et al. 2005; Winckler et al. 2005; Jeffreys et al. 2009; Berg et al. 2010; Myers et al. 2010; Berg et al. 2011; Hinch et al. 2011; Smagulova et al. 2011; Auton et al. 2012; Brick et al. 2012). However, many organisms, including dogs, reptiles, birds, plants, yeast, nematodes, and Drosophila do not possess a copy of Prdm9 (Birtle et al. 2006; Oliver et al. 2009; Ponting 2011; Heil et al. 2012). How are sites of recombination selected in these species? Does fine scale recombination rate turnover persist in the absence of Prdm9? Do fine scale recombination rates diverge even in species lacking hotspots?

Drosophila, for example, have an atypical recombination landscape in that they seem to lack hotspots, the only organism like this besides C. elegans (Hey 2004; Singh et al. 2009; Miller et al. 2012; Manzano-Winkler et al. 2013; Singh et al. 2013). Although they are missing this extreme fine scale variation in recombination rate, recombination rate is
known to vary over different scales and between populations and species (Brooks et al. 1986; True et al. 1996; Cirulli et al. 2007; Kulathinal et al. 2008; Stevison et al. 2010; Chan et al. 2012; Comeron et al. 2012; McGaugh et al. 2012). The unique recombination landscape, combined with a hundred years of recombination research, makes Drosophila an intriguing target to further explore.

In this study, we infer historic recombination rates from linkage disequilibrium present in modern-day DNA sequences using the procedure implemented in the program LDhelmet (Chan et al. 2012) to estimate fine scale recombination rates in Drosophila pseudoobscura and its closely related relative Drosophila miranda (~2-3 my divergence). We leverage empirical maps in both these species to validate our computational approach. We identify patterns of where recombination is happening in relation to genomic features. Finally, we compare recombination rates between D. pseudoobscura and D. miranda at multiple scales and deduce that recombination rates are more divergent at fine scales and conserved at broad scales, as seen in human-chimpanzee.

**Materials and Methods**

**Population sequencing and variant calling**

We used 11 whole genome sequences of *D. pseudoobscura* (Lines: MV2-25 (reference), Mather32, MSH24, MSH9, TL, PP1134, PP1137, AFC12, FLG14, FLG16,
FLG18) from McGaugh et al. (McGaugh et al. 2012) and 23 whole genome sequences of *D. miranda* (Lines: 0101.5, MA28, MA32, MAO101.4, MAO101.7, MAO3.1, MAO3.3, MAO3.4, MAO3.5, MAO3.6, ML1, ML14, ML16, ML22, ML26, ML4, ML51, ML6f, ML71, MSH22, MSH38, SP138, and SP235), provided by Doris Bachtrog’s lab. All sequences were previously aligned to the *D. pseudoobscura* reference sequence v2.9.

The following steps were taken to call variants for the *D. miranda* genome sequences. (1) We used the Genome Analysis Tool Kit (GATK) v2.7-2 “RealignerTargetCreator” and “IndelRealigner” tools to locally realign reads and minimize mismatch calls due to the presence of insertions or deletions (McKenna et al. 2010; DePristo et al. 2011). (2) samtools v0.1.18 and bcftools v0.1.17 were used to call variants (Li et al. 2009). We filtered out unmapped reads and used a mapping quality score of 15. (3) We used a set of custom Perl scripts to convert the variant output files into individual chromosome aligned FASTA files for input to LDhelmet. It is imperative that each FASTA sequence in the aligned file is the same length.

**Recombination estimates**

Empirical estimates of recombination for *D. pseudoobscura* and *D. miranda* were obtained from McGaugh et al. (McGaugh et al. 2012). Population sequencing based estimates of recombination were determined using the LDhelmet program (Chan et al. 2012). LDhelmet is modeled after LDhat (McVean et al. 2002; McVean et al. 2004), but
specifically tailored to address issues in Drosophila, such as a magnitude fold higher background recombination rate, higher SNP density, and a large portion of the genome influenced by positive selection (Sella et al. 2009; Sattath et al. 2011). The program was run individually for each chromosome using default/suggested parameters, with the exception that we estimated theta from our dataset and created our own mutation transition matrix. A block penalty of 50 was used unless otherwise noted.

To compare empirical recombination estimates to LD-based recombination estimates the recombination estimate from LDhelmet was corrected for distance and averaged over a given interval. The empirical chromosomal recombination average (cM/Mb) was divided by the total average LD-based recombination rate for a chromosome to get a conversion factor. Each interval’s average LD-based estimate was multiplied by the conversion factor to get an approximation of recombination rates in the units centiMorgan per Megabase (cM/Mb) per Chan et al (2012). Because we were interested in comparing recombination rates between species, and empirical data suggests that D. miranda has a higher recombination rate than D. pseudoobscura (McGaugh et al. 2012), we sought an additional method of converting LD-based recombination rates without assuming knowledge of empirical recombination rates.

Analyses of genomic correlates
Transcription start sites (TSS): Locations of *D. pseudoobscura* transcription start sites (TSS) were obtained from Main *et al.* (Main *et al.* 2013). The recombination estimate from LDhelmet was corrected for distance and averaged over 5000bp intervals for the 200kb regions flanking a TSS. This data was aggregated together to create chromosome specific TSS figures. Introns and Exons: Locations of exons and introns and relative positions in a gene were extracted from *D. pseudoobscura* v2.9 annotations from FlyBase (St Pierre *et al.* 2014). The recombination estimate from LDhelmet was corrected for distance and averaged over the given interval, then aggregated to give chromosome and genome wide totals for each exon and intron position within a gene. Nucleotide diversity: For our analyses of nucleotide diversity, FlyBase annotations of the *D. pseudoobscura* genome v.2.9 and Perl scripts were used to identify four-fold degenerate coding positions. We created Perl scripts to calculate pairwise nucleotide diversity (π) using four-fold degenerate sites by making all pairwise comparisons for each given interval (e.g. for *D. pseudoobscura* diversity all pairwise comparisons between the 11 lines were made and for *D. miranda*, all pairwise comparisons between the 23 lines were made), excluding sites where an insertion or deletion was found in any line. The number of SNP bases was averaged across each recombination interval for all pairwise comparisons.
Results and Discussion

LD-based recombination maps are good estimates of empirical recombination maps

Historically researchers have used a number of methods to estimate recombination rate including sperm-typing (Jeffreys et al. 1998; Jeffreys et al. 2001), genotyping pedigrees or controlled crosses (Kong et al. 2010), using immunocytological markers (Dumont et al. 2011), ChIP-seq of proteins associated with double strand break or crossover intermediates (Pan et al. 2011; Brick et al. 2012), and employing statistical methods to population sequencing data (Myers et al. 2005). It has been particularly difficult to achieve fine scale resolution in organisms besides human, mouse, and yeast. The latter approach was believed to be inappropriate in Drosophila, where background recombination rates are much higher than humans, for which it was created. A slew of other problems including higher SNP density and known genome wide effects of positive selection (Sella et al. 2009; Sattath et al. 2011) added to the complexity.

Therefore, to create fine scale recombination maps in D. pseudoobscura and D. miranda, we utilized a new program: LDhelmet (Chan et al. 2012), a statistical approach designed for Drosophila that estimates a population recombination rate, $\rho$, from population sequencing data as in (McVean et al. 2002; McVean et al. 2004). The program calculates $\rho = 4N_e r$ (where $N_e$ is the effective population size and $r$ is the recombination rate per generation) to estimate the amount of recombination needed in the population to produce the observed levels of linkage disequilibrium under a given model. Linkage
disequilibrium can be influenced by processes besides recombination including mutation, genetic drift, selection, demography, migration, and gene conversion (Slatkin 2008), and some studies have identified discrepancies between empirical and LD-based recombination maps (Jeffreys et al. 2005; Kauppi et al. 2005; Reed et al. 2006).

**Figure 12: Comparison of D. pseudoobscura empirical and LDhelmet recombination rates on chromosome 2**

To identify how well our LD-based maps replicates empirical recombination rates, we used empirical recombination rates estimated from two populations of *D. pseudoobscura* and one population of *D. miranda* (McGaugh et al. 2012). Empirical and LD-based recombination estimates are positively and significantly correlated at broad scales (**Figure 12**: ~150kb intervals; Chr 2: p<0.0001, r=0.6266; X chr: p<0.0001, r=0.7114) and fine scales (~20 kb intervals, p=0.0084, r=0.6336) across the genome in *D. pseudoobscura*. The broad-scale correlation is similar to that of the two empirical *D.
*pseudoobscura* population maps \((r=0.7183)\), suggesting that we are capturing the full amount of species variation in our LD-based map. In *D. miranda*, the effect is less pronounced, although still significant \((X \text{ chr: } p=0.0090, r=0.3989)\). It is unclear why the correlation between the empirical and LD-based estimates is weaker in *D. miranda*. One possibility is that there is greater population variation in *D. miranda*, as seen in *D. melanogaster* (Comeron et al. 2012), although without another empirical map to compare, it’s not possible to make conclusions.

**Between species recombination rate comparison**

Previous work has shown that recombination rates are conserved at broad scales between closely related species of mammals, birds, and insects (Smukowski et al. 2011). In contrast, it’s clear that fine scale recombination rates diverge rapidly in humans (Berg et al. 2010; Myers et al. 2010; Berg et al. 2011; Hinch et al. 2011; Auton et al. 2012), but little data exists outside primates and it remains uncertain if this phenomenon applies across taxa. In yeast, hotspot turnover is indeed less rapid. Hotspots show relative conservation (about 50%) between *Saccharomyces cerevisiae* and *Saccharomyces paradoxus* despite 13% sequence divergence, although this is likely attributable to the low frequency of sex and outcrossing in these species (Tsai et al. 2010). In dogs, which have a non-functional copy of *Prdm9*, conclusions are mixed. Inferring recombination hotspot history from GC-biased gene conversion, researchers have come to differing opinions.
regarding the evolutionary stability of hotspots in this species (Axelsson et al. 2012; Auton et al. 2013). Regardless, no true comparison has been made in dogs to closely related species, so this remains an intriguing question.

Recombination rate is known to be conserved at broad scales (~200kb) between *D. pseudoobscura* and *D. miranda*, although rates in *D. miranda* are significantly elevated across all regions surveyed (McGaugh et al. 2012). With our LD-based recombination maps, we have the opportunity to identify if the putative global recombination modifier identified in empirical work persists species-wide in *D. miranda*. Indeed, we recapitulate results of McGaugh *et al.*, detecting broad scale conservation of recombination rates between species, with *D. miranda* recombination rates remaining elevated (X chr: 1 Mb: \( r=0.7586; 100\text{kb}: p<0.0001, r=0.6391 \)). Moreover, *D. miranda* recombination rates are higher than detected in empirical work, with average rho values of approximately three times that of *D. pseudoobscura* on the X chromosome. This difference persists despite a difference in effective population sizes, *D. pseudoobscura* \( N_e \) is thought to be several times that of *D. miranda* (Bachtrog et al. 2006; Loewe et al. 2006; Jensen et al. 2011). We can only speculate as to why recombination rates would increase in *D. miranda* (or decrease in *D. pseudoobscura*) since their divergence. Theoretical work shows selection for or against new haplotypes can indirectly select for alleles that modify recombination rates (Otto et al. 2002), observations in domesticated crops and several direct selection experiments in
Drosophila support this (Flexon et al. 1982; Burt et al. 1987; Korol et al. 1994; Otto et al. 2001; Ross-Ibarra 2004; Coop et al. 2007).

Analysis of fine scale recombination in Drosophila has been limited to isolated genomic regions in *D. melanogaster* (Schweitzer 1935; Singh et al. 2009; Miller et al. 2012; Singh et al. 2013) and *D. pseudoobscura* (Cirulli et al. 2007; McGaugh et al. 2012; Manzano-Winkler et al. 2013). With LD-based maps, we are able to characterize the genome wide fine scale recombination landscape in *D. pseudoobscura* and *D. miranda* for the first time, and to test whether turnover in fine scale recombination rates occurs outside of mammals. Recombination rates averaged over 10kb windows show heterogeneity in both species, and again, *D. miranda* has higher recombination than *D. pseudoobscura*. However, we detect less variation in *D. miranda* than *D. pseudoobscura* (Figure 13). Whether this is an artifact of lower nucleotide diversity in *D. miranda* is unclear. We do detect greater divergence at fine scales (X chr: 10kb: p<0.0001, r=0.4604), consistent with less constraint on recombination rates or greater error in estimation at this scale. It’s suggestive that Drosophila too face the hotspot paradox, and as they don’t have a *Prdm9*-like protein, have come up with an alternate solution to counteract the loss.
Distribution of recombination rate across the genome

Transcription Start Sites and coding regions

Hotspots in yeast, plants, and dog are primarily found overlapping promoters, presumably due to the open chromatin structure at promoters allowing recombination machinery access to the DNA (Petes 2001; Lichten 2008; Mancera et al. 2008; Pan et al. 2011; Choi et al. 2013; Hellsten et al. 2013). In contrast, hotspots primarily occur outside of promoters in mouse and human (Myers et al. 2005; Smagulova et al. 2011), and there is evidence that Prdm9 may even be directing recombination machinery away from promoters and other functional genomic elements (Brick et al. 2012). In these species, interactions with histones, particularly a meiosis specific H3K4me3 modification, govern hotspot distribution. These observations set up a model in which organisms
without Prdm9 (yeast, plants, dog) are opportunistic: they form recombination hotspots in “windows of opportunity,” most often nucleosome depleted regions around promoters. On the other hand, in organisms with Prdm9 (humans, mouse), Prdm9 most likely functions to introduce meiotic specific H3K4me3 modifications and directs recombination machinery to initiate at these locations away from genomic elements (Hayashi et al. 2005; Grey et al. 2011). Although factors that determine hotspots in these organisms differ, the overall patterns and intensities of hotspots are remarkably similar between organisms.

![Figure 14: Recombination rate around transcription start sites](image_url)

In D. pseudoobscura, recombination is reduced around transcription start sites and adjacent regions up to 35 kb away (Figure 14). This is supported by findings in D.
melanogaster as well (Chan et al. 2012), although the reduction in rates appears to persist for greater distances in D. pseudoobscura. Similarly, we see reduction in recombination rates at the 5' end of genes, with greater rates in introns than exons, which both increase with distance from the start of the coding region (Figure 15). This reveals a contradictory pattern from the monkeyflower, Mimulus guttatus (Hellsten et al. 2013), and speaks to Drosophila's unusual recombination landscape. It remains unclear what genomic features influence the distribution of recombination events in Drosophila; future studies looking at histone modifications and nucleosome occupancy will hopefully shed light on this enigma.

![Figure 15: Recombination rate within genes](cumulative_recombination_map.png)

**Figure 15: Recombination rate within genes**

Cumulative recombination map
Looking at the proportion of recombination that happens in a proportion of the sequence reveals the degree to which recombination events cluster across the genome. In humans, recombination is highly punctate, with about 80% of recombination happening in less than 20% of the sequence (Myers et al. 2006), or a Gini coefficient of approximately 0.8 (Kong et al. 2010; Kaur et al. 2014). As previously discussed, Drosophila lack hotspots characteristic of other organisms (Singh et al. 2009; Manzano-Winkler et al. 2013; Singh et al. 2013), although fine scale variation has been identified (see above), and Chan et al. even detected a handful of more typical hotspots (Chan et al. 2012). This is reflected in D. pseudoobscura’s intermediate Gini coefficient of approximately 0.50, or about 80% of recombination occurring in 50% of the sequence (Figure 16). This is similar to the Gini coefficient estimated for D. melanoagaster (0.47), falling between C. elegans (0.28) and S. cerevisae (0.64) (Kaur et al. 2014).

Figure 16: Proportion recombination occurring in proportion of sequence
Conclusions

Here we employ a computational approach to present the first genome wide fine scale recombination maps for *D. pseudoobscura* and *D. miranda*. Utilizing existing empirical recombination maps in these species, we were able to validate our results and conclude the program LDhelmet is a valuable resource for the Drosophila and recombination communities. Furthermore, we reveal an unprecedented view of the distribution of recombination in these species, illustrating new patterns of where recombination is taking place. Finally, we show broad scale conservation of recombination rates and fine scale divergence between closely related species, one of the only studies to show this phenomenon exists outside of mouse and human. This raises interesting implications about the evolution of meiosis and selection pressures on recombination that persist across distantly related taxa with widely different recombination landscapes.

Future studies can apply these types of fine scale recombination maps to identify determinants of recombination, improve theory and modeling, and to better understand molecular evolutionary processes and genome evolution.
Works Cited


Barthes, P., J. Buard, B. de Massy (2011). Epigenetic Factors and Regulation of Meiotic Recombination in Mammals Springer.


Biography

Caitlin Smukowski Heil was born in Bellevue, Washington on October 24, 1986 to David and Patty Smukowski. After attending Seattle Preparatory High School, she graduated cum laude with a Bachelor of Arts in Biology from the University of Pennsylvania in 2009. Caiti became interested in evolutionary genetics in her senior year, completing an honors thesis entitled “An analysis of candidate interaction genes with couch potato, a gene for reproductive diapauses in Drosophila melanogaster” under the guidance of Paul Schmidt. In the fall of 2009, Caiti joined Mohamed Noor’s lab at Duke University, where she has published five peer-reviewed articles (“Zinc finger binding motifs do not explain recombination variation within or between species of Drosophila”; “Recombination modulates how selection affects linked sites in Drosophila”; “Recombination rate variation in closely related species”; “Witnessing Evolution First-hand: A K-12 Laboratory Exercise in Genetics and Evolution Using Drosophila”; and “Witnessing phenotypic and molecular evolution in the fruit fly”), and two non-peer reviewed papers (“Studying recombination with high-throughput sequencing: An educational primer for use with ‘Fine-scale heterogeneity in crossover rate in the garnet-scalloped region of the Drosophila melanogaster X chromosome’”; and “Mentor vs. Monolith: Finding and being a good graduate advisor”). Caiti is a recipient of the James B. Duke fellowship and the National Science Foundation Graduate Fellowship, and has received grants from Sigma Xi, Society for the Study of Evolution, and the National Science
Foundation (DDIG). She is continuing on to a Postdoctoral position in Maitreya Dunham’s lab at the University of Washington in 2014, where she will focus on genome evolution. She lives with her husband, Lou, and two cats, Monkey and Business, and enjoys reading, cooking, running, and hiking.