

# Web-based Supporting Materials for “Robust Bayesian hierarchical model using normal/independent distributions” by Geng Chen and Sheng Luo

## BUGS code for fitting model Indep-CN

```
model
{
  for (i in 1:obs) # obs: number of total observations
  {
    # Construct contaminated Normal distribution
    u.contam2[i] ~ dbern(nu2)
    w2[i] <- gamma2*u.contam2[i] + (1-u.contam2[i])
    Y.conti[i] ~ dnorm(mu.conti[i], tau2[i]) # continuous outcome
    tau2[i] <- w2[i]*tau.conti
    # K.ordi: number of ordinal variables, K=2 in DATATOP study
    for (k in 1:K.ordi) { Y.ordi[i, k] ~ dcat(prob.y[i, k, 1:n[k]]) }
  }

  # construct mean of the continuous variable
  for (i in 1:obs)
  {
    mu.conti[i] <- a.conti + b.conti * theta[i]
  }

  # Construct the probability vector for the ordinal variables
  for (i in 1:obs)
  {
    for (k in 1:K.ordi)
    {
      for (l in 1:(n[k]-1)) { logit(psi[i, k, l]) <- a.ordi[k,l]
        - b.ordi[k]*theta[i] }

      prob.y[i, k, 1] <- psi[i, k, 1]
      for (l in 2:(n[k]-1)) { prob.y[i, k, l] <- psi[i, k, l]
        - psi[i, k, l-1] }

      prob.y[i, k, n[k]] <- 1 - psi[i, k, (n[k]-1)]
    }
  }

  # construct random effects
  for (i in 1:N)
  {
    u[i, 1:2] ~ dmnorm(zero[], precision[i, ,])
    u.contam1[i] ~ dbern(nu1)
    w1[i] <- gamma1*u.contam1[i] + (1-u.contam1[i]) # Contaminated Normal
  }
}
```

```

# construct the variance-covariance matrix for random effects
precision[i,1:2,1:2] <- inverse(Sigma[i,,])
Sigma[i,1,1] <- 1/w1[i]
Sigma[i,1,2] <- rho*sig/w1[i]
Sigma[i,2,1] <- Sigma[i,1,2]
Sigma[i,2,2] <- sig*sig/w1[i]
}

# construct theta, the latent variable of subject i at time j
for (i in 1:obs)
{
  theta[i] <- u[subject[i], 1] + (beta[1] + beta[2]*treat[i]
    + u[subject[i], 2])*time[i]
}

# prior for regression coefficients
for (i in 1:2) { beta[i] ~ dnorm(0, 0.01) }

# specify prior distributions
rho ~ dunif(-1, 1)
sig ~ dgamma(0.01, 0.01)

# prior for continuous variable's parameters
b.conti ~ dgamma(0.001,0.001)
a.conti ~ dnorm(0, 0.0005)
tau.conti ~ dgamma(0.001,0.001)
sd.conti <- 1/sqrt(tau.conti)

# prior for ordinal variables' parameters
for (k in 1:K.ordi)
{
  b.ordi[k] ~ dgamma(0.001,0.001)
  a.ordi[k, 1] ~ dnorm(0,0.001)
  for (l in 2:(n[k]-1)) { a.ordi[k, l] <- a.ordi[k, l-1] + delta[k, l-1] }
  for (i in 1:(n[k]-2)) {delta[k, i] ~ dnorm(0,0.01)I(0,)}
}

# prior for contaminated normal tuning parameters
nu1 ~ dbeta(1,1)
gamma1 ~ dbeta(1,1)
nu2 ~ dbeta(1,1)
gamma2 ~ dbeta(1,1)
}

```

Table 1: Simulation results from models Indep-T, Indep-SL, and Indep-CN to recover their true parameters.

	Indep-T					Indep-SL					Indep-CN				
	True	Bias	SD	SE	CP	True	Bias	SD	SE	CP	True	Bias	SD	SE	CP
$a_1$	25.00	0.12	0.55	0.60	0.93	25.00	-0.01	0.71	0.62	0.91	25.00	-0.04	0.59	0.59	0.98
$b_1$	10.00	-0.11	0.58	0.55	0.93	10.00	-0.03	0.70	0.62	0.88	10.00	-0.06	0.49	0.48	0.95
$a_2$	80.00	0.25	1.06	1.14	0.94	80.00	-0.08	1.40	1.18	0.91	80.00	-0.07	1.16	1.12	0.97
$b_2$	18.00	-0.20	1.07	1.00	0.91	18.00	-0.01	1.27	1.13	0.89	18.00	-0.16	0.87	0.88	0.97
$a_{31}$	-2.70	-0.04	0.16	0.15	0.91	-2.70	-0.01	0.17	0.16	0.95	-2.70	0.00	0.15	0.15	0.92
$a_{32}$	-0.60	-0.03	0.13	0.14	0.93	-0.60	0.01	0.15	0.14	0.93	-0.60	-0.00	0.12	0.13	0.97
$a_{33}$	2.00	-0.03	0.14	0.14	0.95	2.00	0.02	0.17	0.15	0.93	2.00	-0.01	0.13	0.14	0.99
$a_{34}$	2.80	-0.03	0.15	0.15	0.98	2.80	0.01	0.20	0.16	0.89	2.80	-0.01	0.15	0.15	0.98
$a_{35}$	5.00	-0.01	0.20	0.20	0.95	5.00	0.02	0.28	0.20	0.86	5.00	0.01	0.19	0.20	0.97
$a_{36}$	6.00	-0.00	0.20	0.22	0.99	6.00	0.02	0.27	0.22	0.91	6.00	0.01	0.22	0.22	0.95
$b_3$	2.00	-0.01	0.13	0.12	0.95	2.00	0.00	0.16	0.14	0.93	2.00	-0.01	0.10	0.11	0.97
$a_{41}$	-0.10	-0.01	0.05	0.05	0.94	-0.10	0.00	0.06	0.05	0.95	-0.10	-0.00	0.05	0.05	0.95
$a_{42}$	1.00	-0.01	0.06	0.06	0.91	1.00	0.02	0.06	0.06	0.91	1.00	-0.00	0.06	0.06	0.93
$a_{43}$	1.80	-0.01	0.07	0.07	0.98	1.80	0.01	0.07	0.07	0.95	1.80	0.00	0.07	0.07	0.92
$a_{44}$	2.60	-0.01	0.09	0.09	0.95	2.60	0.02	0.08	0.09	0.95	2.60	0.01	0.09	0.09	0.99
$a_{45}$	3.30	-0.00	0.11	0.11	0.94	3.30	0.02	0.10	0.11	0.95	3.30	0.01	0.12	0.11	0.91
$a_{46}$	4.00	0.02	0.15	0.14	0.93	4.00	0.02	0.14	0.14	0.95	4.00	0.03	0.13	0.14	0.97
$b_4$	0.40	-0.00	0.03	0.03	0.93	0.40	0.00	0.03	0.03	0.89	0.40	-0.00	0.03	0.03	0.95
$\beta_{10}$	0.40	0.03	0.11	0.11	0.95	0.40	0.01	0.12	0.11	0.91	0.40	0.01	0.11	0.10	0.92
$\beta_{11}$	-0.50	-0.04	0.15	0.15	0.91	-0.50	-0.03	0.17	0.15	0.91	-0.50	-0.03	0.14	0.14	0.93
$\nu_\sigma$	2.00	0.12	0.40	0.38	0.94	3.00	1.12	2.07	2.78	0.88	—	—	—	—	—
$\nu_y$	3.00	0.26	0.69	0.70	0.93	4.00	1.42	2.60	3.17	0.88	—	—	—	—	—
$\gamma_1$	—	—	—	—	—	—	—	—	—	—	0.10	0.01	0.02	0.02	0.97
$\gamma_2$	—	—	—	—	—	—	—	—	—	—	0.12	0.01	0.02	0.02	0.88
$\nu_1$	—	—	—	—	—	—	—	—	—	—	0.10	0.02	0.03	0.03	0.91
$\nu_2$	—	—	—	—	—	—	—	—	—	—	0.15	0.02	0.03	0.03	1.00
$\rho$	0.40	-0.01	0.05	0.05	0.94	0.40	-0.01	0.04	0.05	0.96	0.40	-0.00	0.05	0.05	0.93
$\sigma_u$	1.30	0.02	0.08	0.07	0.89	1.30	0.02	0.07	0.07	0.93	1.30	0.01	0.07	0.07	0.93

Table 2: Simulation results from models Dep-T, Dep-SL, and Dep-CN to recover their true parameters.

	Dep-T					Dep-SL					Dep-CN				
	True	Bias	SD	SE	CP	True	Bias	SD	SE	CP	True	Bias	SD	SE	CP
$a_1$	25.00	0.00	0.54	0.57	0.95	25.00	-0.02	0.60	0.61	0.97	25.00	-0.08	0.57	0.55	0.91
$b_1$	10.00	-0.05	0.47	0.47	0.95	10.00	-0.02	0.49	0.47	0.92	10.00	-0.00	0.45	0.38	0.91
$a_2$	80.00	0.00	1.10	1.11	0.96	80.00	-0.09	1.16	1.16	0.92	80.00	-0.20	1.10	1.06	0.90
$b_2$	18.00	-0.10	0.84	0.86	0.96	18.00	-0.01	0.89	0.87	0.94	18.00	-0.02	0.80	0.70	0.88
$a_{31}$	-2.70	-0.00	0.17	0.15	0.92	-2.70	-0.00	0.15	0.15	0.93	-2.70	0.03	0.15	0.14	0.92
$a_{32}$	-0.60	-0.00	0.13	0.13	0.95	-0.60	-0.01	0.14	0.14	0.94	-0.60	0.02	0.13	0.13	0.95
$a_{33}$	2.00	0.00	0.14	0.14	0.94	2.00	0.01	0.14	0.15	0.93	2.00	0.02	0.15	0.14	0.96
$a_{34}$	2.80	-0.00	0.15	0.15	0.95	2.80	0.02	0.14	0.16	0.97	2.80	0.02	0.15	0.15	0.96
$a_{35}$	5.00	0.01	0.20	0.20	0.94	5.00	0.03	0.19	0.20	0.97	5.00	-0.00	0.20	0.19	0.96
$a_{36}$	6.00	0.01	0.22	0.22	0.94	6.00	0.04	0.22	0.22	0.94	6.00	0.04	0.23	0.22	0.96
$b_3$	2.00	-0.00	0.12	0.11	0.91	2.00	0.00	0.11	0.11	0.98	2.00	0.00	0.10	0.10	0.94
$a_{41}$	-0.10	0.00	0.05	0.05	0.98	-0.10	0.00	0.05	0.05	0.92	-0.10	0.00	0.05	0.05	0.90
$a_{42}$	1.00	-0.00	0.05	0.06	0.97	1.00	0.00	0.06	0.06	0.91	1.00	-0.00	0.06	0.06	0.96
$a_{43}$	1.80	-0.00	0.07	0.07	0.96	1.80	0.00	0.07	0.07	0.94	1.80	0.01	0.07	0.07	0.97
$a_{44}$	2.60	0.01	0.09	0.09	0.94	2.60	0.01	0.10	0.09	0.92	2.60	0.00	0.07	0.09	0.96
$a_{45}$	3.30	0.01	0.12	0.11	0.94	3.30	0.01	0.12	0.11	0.98	3.30	0.00	0.11	0.11	0.94
$a_{46}$	4.00	0.05	0.16	0.14	0.91	4.00	0.01	0.14	0.14	0.96	4.00	0.02	0.15	0.14	0.95
$b_4$	0.40	-0.00	0.02	0.03	0.97	0.40	-0.00	0.03	0.03	0.90	0.40	0.00	0.02	0.02	0.96
$\beta_{10}$	0.40	0.00	0.11	0.10	0.97	0.40	0.02	0.11	0.11	0.92	0.40	0.02	0.11	0.10	0.94
$\beta_{11}$	-0.50	0.01	0.14	0.14	0.94	-0.50	-0.01	0.16	0.15	0.97	-0.50	-0.01	0.14	0.13	0.92
$\nu$	4.00	0.13	0.47	0.44	0.91	3.00	0.20	0.58	0.94	0.97	—	—	—	—	—
$\gamma_1$	—	—	—	—	—	—	—	—	—	—	0.10	0.00	0.01	0.01	1.00
$\nu_1$	—	—	—	—	—	—	—	—	—	—	0.10	0.01	0.01	0.02	0.96
$\rho$	0.40	0.00	0.04	0.05	0.95	0.40	0.00	0.04	0.05	0.99	0.40	-0.00	0.05	0.05	0.95
$\sigma_u$	1.30	0.03	0.09	0.07	0.86	1.30	0.01	0.07	0.07	0.93	1.30	0.01	0.07	0.06	0.94

Table 3: Simulation results from models Indep-CN, Dep-CN and Indep-N using half normal distribution priors in setting II in which there were 5% outliers in both the continuous outcome and random effects.

	True	Indep-N				Dep-CN				Indep-CN			
		Bias	SD	SE	CP	Bias	SD	SE	CP	Bias	SD	SE	CP
$a_1$	25.000	1.300	0.639	0.888	0.761	0.172	0.607	0.531	0.908	0.147	0.574	0.530	0.930
$b_1$	10.000	6.442	0.767	0.642	0.000	-0.123	0.416	0.383	0.882	0.040	0.393	0.383	0.924
$a_2$	80.000	3.771	1.095	1.587	0.283	0.033	1.140	1.037	0.941	0.069	1.081	1.034	0.936
$b_2$	18.000	12.655	0.936	1.181	0.000	-0.180	0.756	0.707	0.921	0.056	0.713	0.697	0.930
$a_{31}$	-2.700	-0.438	0.147	0.198	0.333	-0.025	0.144	0.142	0.941	-0.028	0.148	0.142	0.962
$a_{32}$	-0.600	-0.408	0.131	0.180	0.302	-0.006	0.134	0.124	0.914	-0.011	0.123	0.124	0.962
$a_{33}$	2.000	-0.368	0.146	0.184	0.465	0.007	0.155	0.135	0.901	0.006	0.148	0.135	0.911
$a_{34}$	2.800	-0.354	0.147	0.191	0.509	0.018	0.160	0.144	0.947	0.012	0.157	0.144	0.930
$a_{35}$	5.000	-0.366	0.189	0.229	0.660	0.014	0.197	0.191	0.928	0.023	0.195	0.191	0.949
$a_{36}$	6.000	-0.356	0.228	0.254	0.736	0.029	0.233	0.219	0.947	0.040	0.222	0.219	0.962
$b_3$	2.000	1.341	0.157	0.167	0.000	-0.022	0.106	0.096	0.901	0.019	0.106	0.096	0.892
$a_{41}$	-0.100	-0.080	0.051	0.059	0.792	0.003	0.052	0.053	0.941	0.000	0.054	0.053	0.943
$a_{42}$	1.000	-0.079	0.052	0.063	0.805	0.001	0.051	0.058	0.967	0.002	0.055	0.058	0.968
$a_{43}$	1.800	-0.075	0.060	0.073	0.874	0.010	0.061	0.069	0.974	0.008	0.065	0.069	0.968
$a_{44}$	2.600	-0.077	0.087	0.090	0.855	0.015	0.089	0.088	0.934	0.011	0.091	0.087	0.924
$a_{45}$	3.300	-0.073	0.105	0.113	0.893	0.024	0.110	0.112	0.954	0.021	0.113	0.111	0.955
$a_{46}$	4.000	-0.075	0.143	0.144	0.918	0.030	0.151	0.144	0.941	0.028	0.161	0.144	0.917
$b_4$	0.400	0.278	0.038	0.042	0.000	-0.002	0.025	0.025	0.954	0.005	0.025	0.025	0.962
$\beta_{10}$	0.400	0.019	0.063	0.074	0.975	-0.003	0.105	0.096	0.908	0.006	0.097	0.092	0.930
$\beta_{11}$	-0.500	0.113	0.082	0.088	0.780	0.017	0.131	0.126	0.947	-0.001	0.124	0.125	0.943
$\rho$	0.400	0.354	0.021	0.025	0.000	0.041	0.044	0.045	0.862	0.031	0.041	0.046	0.911
$\sigma_u$	1.300	-0.109	0.056	0.048	0.346	-0.004	0.066	0.063	0.921	-0.001	0.065	0.064	0.949