

Amazon Reviews

Dr. Jarad Niemi

STAT 544 - Iowa State University

February 23, 2017

Amazon Reviews - Upright, bagless, cyclonic vacuum cleaners

	Number of ratings							
product_id	n1	n2	n3	n4	n5	n_total	mean	sd
B000REMVGK	21	17	2	8	7	55	2.33	1.44
B001EFMD8W	40	34	28	77	347	526	4.25	1.26
B001PB51GQ	14	12	13	31	69	139	3.93	1.36
B002DGSJVG	22	8	3	6	10	49	2.47	1.63
B002G9UQZC	8	0	1	1	1	11	1.82	1.47
B002GHBRX4	18	8	9	14	27	76	3.32	1.61
B002HF66BI	9	5	2	2	3	21	2.29	1.49
B003OA77MC	15	7	8	24	42	96	3.74	1.47
B003OAD24Y	7	7	4	9	19	46	3.57	1.53
B003Y3AA3C	20	3	1	2	2	28	1.68	1.28
B0043EW354	40	25	25	60	163	313	3.90	1.44
B00440EO8G	2	1	1	1	7	12	3.83	1.64
B004R9197I	9	1	1	9	26	46	3.91	1.58
B008L5F4H0	3	1	2	12	7	25	3.76	1.27

Model for Amazon Reviews

Let y_{ij} be the j th review for the i th product. Assume

$$y_{ij} \stackrel{\text{ind}}{\sim} N(\theta_i, \sigma^2)$$

and

$$\theta_i \stackrel{\text{ind}}{\sim} N(\mu, \tau^2)$$

and

$$p(\mu, \tau, \sigma) \propto Ca^+(\sigma; 0, 1)Ca^+(\tau; 0, 1)$$

Normal hierarchical model in Stan

```
normal_model = "  
data {  
  int <lower=1> n;  
  int <lower=1> n_products;  
  int <lower=1,upper=5> stars[n];  
  int <lower=1,upper=n_products> id[n];  
}  
  
parameters {  
  real mu;                // implied uniform prior  
  real<lower=0> sigma;  
  real<lower=0> tau;  
  real theta[n_products];  
}  
  
model {  
  // Prior  
  sigma ~ cauchy(0,1);  
  tau ~ cauchy(0,1);  
  
  // Hierarchical model  
  theta ~ normal(mu,tau);  
  
  // Data model  
  for (i in 1:n) stars[i] ~ normal(theta[id[i]], sigma);  
}  
"
```

Fit model

```
m = stan_model(model_code = normal_model)
```

```
In file included from C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/config.hpp:39:0,
                 from C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/math/tools/config.hpp:13,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core/var.hpp:13,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core/gevv_vv.hpp:13,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core.hpp:12,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/mat.hpp:4,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math.hpp:4,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/src/stan/model/model_header.hpp:4,
                 from file34dc40857830.cpp:8:
```

```
C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/config/compiler/gcc.hpp:186:0: warning: "BOOST_NO_CXX11_RVALUE_REFERENCES"
# define BOOST_NO_CXX11_RVALUE_REFERENCES
^
```

```
<command-line>:0:0: note: this is the location of the previous definition
cc1plus.exe: warning: unrecognized command line option "-Wno-ignored-attributes"
```

```
dat = list(n = nrow(d),
           n_products = nlevels(d$product_id),
           stars = d$stars,
           id = as.numeric(d$product_id))
r = sampling(m, dat)
```

```
SAMPLING FOR MODEL '5edd831c9bc5b4f7faa8979cea0d3b99' NOW (CHAIN 1).
```

```
Chain 1, Iteration: 1 / 2000 [ 0%] (Warmup)
```

```
Chain 1, Iteration: 200 / 2000 [ 10%] (Warmup)
```

```
Chain 1, Iteration: 400 / 2000 [ 20%] (Warmup)
```

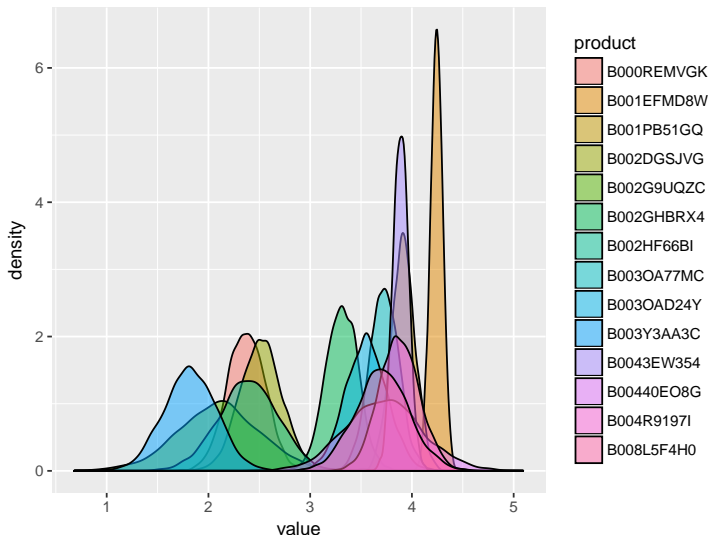
Tabular summary

Inference for Stan model: 5edd831c9bc5b4f7faa8979cea0d3b99.
 4 chains, each with iter=2000; warmup=1000; thin=1;
 post-warmup draws per chain=1000, total post-warmup draws=4000.

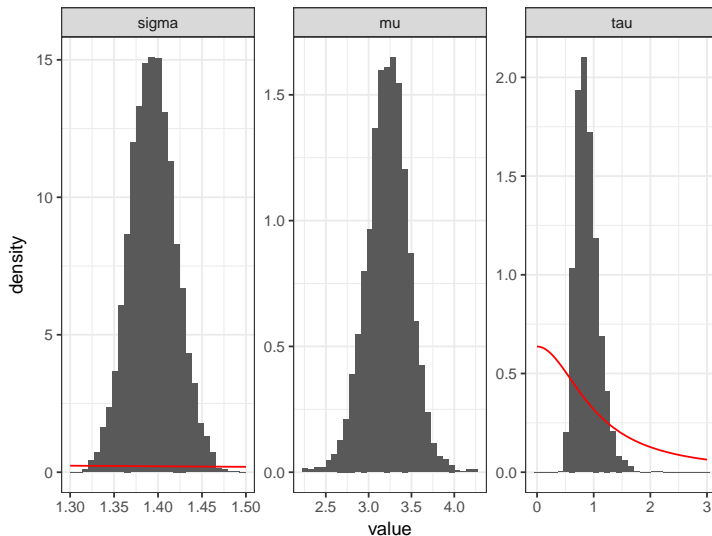
	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	Rhat
mu	3.22	0.00	0.25	2.73	3.07	3.23	3.38	3.70	4000	1
sigma	1.39	0.00	0.03	1.34	1.38	1.39	1.41	1.45	4000	1
tau	0.89	0.00	0.20	0.58	0.74	0.86	1.00	1.36	4000	1
theta[1]	2.37	0.00	0.19	2.00	2.24	2.37	2.50	2.73	4000	1
theta[2]	4.24	0.00	0.06	4.12	4.20	4.24	4.28	4.36	4000	1
theta[3]	3.91	0.00	0.12	3.68	3.84	3.91	3.99	4.14	4000	1
theta[4]	2.51	0.00	0.20	2.12	2.38	2.51	2.65	2.90	4000	1
theta[5]	2.10	0.01	0.41	1.27	1.83	2.10	2.37	2.88	4000	1
theta[6]	3.31	0.00	0.16	3.00	3.20	3.31	3.42	3.62	4000	1
theta[7]	2.39	0.00	0.29	1.80	2.20	2.39	2.59	2.95	4000	1
theta[8]	3.73	0.00	0.14	3.44	3.63	3.73	3.82	4.02	4000	1
theta[9]	3.55	0.00	0.20	3.15	3.41	3.55	3.69	3.95	4000	1
theta[10]	1.82	0.00	0.26	1.31	1.65	1.82	1.99	2.32	4000	1
theta[11]	3.89	0.00	0.08	3.74	3.84	3.89	3.94	4.04	4000	1
theta[12]	3.72	0.01	0.38	3.00	3.47	3.72	3.97	4.48	4000	1
theta[13]	3.88	0.00	0.20	3.50	3.74	3.87	4.01	4.27	4000	1
theta[14]	3.71	0.00	0.26	3.20	3.54	3.71	3.89	4.24	4000	1
lp__	-1207.46	0.08	2.99	-1214.35	-1209.24	-1207.14	-1205.34	-1202.46	1572	1

Samples were drawn using NUTS(diag_e) at Thu Feb 23 06:24:00 2017.
 For each parameter, n_eff is a crude measure of effective sample size,
 and Rhat is the potential scale reduction factor on split chains (at
 convergence, Rhat=1).

Vacuum cleaner mean posteriors (θ_i)



Other parameter posteriors



A quick rating

Suppose a new vacuum cleaner comes on the market and there are two Amazon reviews both with 5 stars. What do you think the averaging star rating will be (in the future) for this new product?

Let n^* be the number of new ratings and \bar{y}^* be the average of those ratings, then

$$\begin{aligned} E[\theta^* | \bar{y}^*, n^*, \sigma, \mu, \tau] &= \frac{\frac{n^*}{\sigma^2}}{\frac{n^*}{\sigma^2} + \frac{1}{\tau^2}} \bar{y}^* + \frac{\frac{1}{\tau^2}}{\frac{n^*}{\sigma^2} + \frac{1}{\tau^2}} \mu \\ &= \frac{n^*}{n^* + \frac{\sigma^2}{\tau^2}} \bar{y}^* + \frac{\frac{\sigma^2}{\tau^2}}{n^* + \frac{\sigma^2}{\tau^2}} \mu \\ &= \frac{n^*}{n^* + m} \bar{y}^* + \frac{m}{n^* + m} \mu \end{aligned}$$

where $m = \sigma^2/\tau^2$ is a measure of how many *prior* samples there are.

IMDB rating

From <http://www.imdb.com/chart/top.html>:

$$\text{weighted rating (WR)} = (v / (v+m)) R + (m / (v+m)) C$$

Where:

R = average for the movie (mean) = (Rating)

v = number of votes for the movie = (votes)

m = minimum votes required to be listed in the Top 250
(currently 25000)

C = the mean vote across the whole report (currently 7.1)

Thus IMDB uses a Bayesian estimate for the rating for each movie where $m = \sigma^2 / \tau^2 = 25,000$. IMDB has enough data that the uncertainty in $\mu(C)$, σ^2 , and τ^2 is pretty minimal.

Clearly incorrect model

We assumed

$$y_{ij} \stackrel{\text{ind}}{\sim} N(\theta_i, \sigma^2)$$

for the j th star rating of product i . Clearly this model is incorrect since $y_{ij} \in \{1, 2, 3, 4, 5\}$.

An alternative model is

$$z_{ij} \stackrel{\text{ind}}{\sim} \text{Bin}(4, \theta_i)$$

where $z_{ij} = y_{ij} - 1$ is the j th star rating minus 1 of product i and

$$\theta_i \sim \text{Be}(\alpha, \beta) \quad \text{and} \quad p(\alpha, \beta) \propto (\alpha + \beta)^{-5/2}.$$

The idea behind this model would be that product i has an independent, but common probability (θ_i) of earning each star.

Binomial hierarchical model in Stan

```
binomial_model = "  
data {  
  int <lower=1> n;  
  int <lower=1> n_products;  
  int <lower=0,upper=4> z[n];  
  int <lower=1,upper=n_products> id[n];  
}  
  
parameters {  
  real<lower=0> alpha;  
  real<lower=0> beta;  
  real<lower=0,upper=1> theta[n_products];  
}  
  
model {  
  // Prior  
  target += -5*log(alpha+beta)/2; // improper prior  
  
  // Hierarchical model  
  theta ~ beta(alpha,beta);  
  
  // Data model  
  for (i in 1:n) z[i] ~ binomial(4, theta[id[i]]);  
}  
"
```

Fit model

```
m = stan_model(model_code = binomial_model)
```

```
In file included from C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/config.hpp:39:0,
      from C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/math/tools/config.hpp:13,
      from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core/var.hpp:13,
      from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core/gevv_vv.hpp:13,
      from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core.hpp:12,
      from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/mat.hpp:4,
      from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math.hpp:4,
      from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/src/stan/model/model_header.hpp:4,
      from file34dc508c208b.cpp:8:
```

```
C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/config/compiler/gcc.hpp:186:0: warning: "BOOST_NO_CXX11_RVALUE_REFERENCES"
# define BOOST_NO_CXX11_RVALUE_REFERENCES
^
```

```
<command-line>:0:0: note: this is the location of the previous definition
cc1plus.exe: warning: unrecognized command line option "-Wno-ignored-attributes"
```

```
dat = list(n = nrow(d),
           n_products = nlevels(d$product_id),
           z = d$stars-1,
           id = as.numeric(d$product_id))
r = sampling(m, dat)
```

```
SAMPLING FOR MODEL 'c562d81df0c52a21eec024bb5bedd1d6' NOW (CHAIN 1).
```

```
Chain 1, Iteration: 1 / 2000 [ 0%] (Warmup)
```

```
Chain 1, Iteration: 200 / 2000 [ 10%] (Warmup)
```

```
Chain 1, Iteration: 400 / 2000 [ 20%] (Warmup)
```

Tabular summary

Inference for Stan model: c562d81df0c52a21eec024bb5bedd1d6.

4 chains, each with iter=2000; warmup=1000; thin=1;

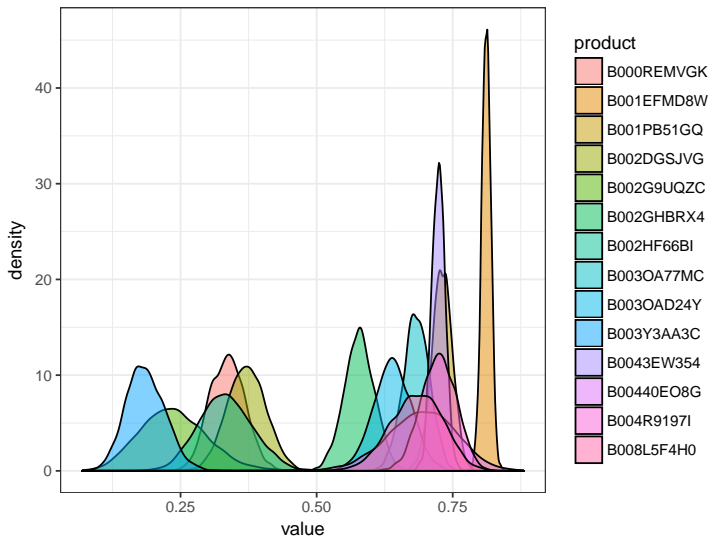
post-warmup draws per chain=1000, total post-warmup draws=4000.

	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	Rhat
alpha	2.67	0.02	1.06	1.06	1.90	2.52	3.27	5.20	2797	1
beta	2.24	0.02	0.85	0.94	1.63	2.11	2.70	4.28	2663	1
theta[1]	0.34	0.00	0.03	0.28	0.31	0.34	0.36	0.40	4000	1
theta[2]	0.81	0.00	0.01	0.79	0.81	0.81	0.82	0.83	4000	1
theta[3]	0.73	0.00	0.02	0.69	0.72	0.73	0.74	0.77	4000	1
theta[4]	0.37	0.00	0.04	0.30	0.35	0.37	0.40	0.44	4000	1
theta[5]	0.24	0.00	0.06	0.13	0.19	0.23	0.28	0.37	4000	1
theta[6]	0.58	0.00	0.03	0.52	0.56	0.58	0.60	0.63	4000	1
theta[7]	0.33	0.00	0.05	0.24	0.30	0.33	0.37	0.43	4000	1
theta[8]	0.68	0.00	0.02	0.63	0.67	0.68	0.70	0.73	4000	1
theta[9]	0.64	0.00	0.03	0.57	0.62	0.64	0.66	0.71	4000	1
theta[10]	0.19	0.00	0.04	0.12	0.16	0.18	0.21	0.26	4000	1
theta[11]	0.72	0.00	0.01	0.70	0.72	0.72	0.73	0.75	4000	1
theta[12]	0.69	0.00	0.06	0.56	0.65	0.69	0.74	0.81	4000	1
theta[13]	0.72	0.00	0.03	0.66	0.70	0.72	0.75	0.78	4000	1
theta[14]	0.68	0.00	0.05	0.59	0.65	0.69	0.72	0.77	4000	1
lp__	-3265.24	0.07	2.90	-3271.82	-3267.00	-3264.87	-3263.17	-3260.59	1928	1

Samples were drawn using NUTS(diag_e) at Thu Feb 23 06:24:52 2017.

For each parameter, `n_eff` is a crude measure of effective sample size, and `Rhat` is the potential scale reduction factor on split chains (at convergence, `Rhat`=1).

Review mean posteriors (θ_i)



Other parameter posteriors

Recall that

- α is the prior success
- β is the prior failures

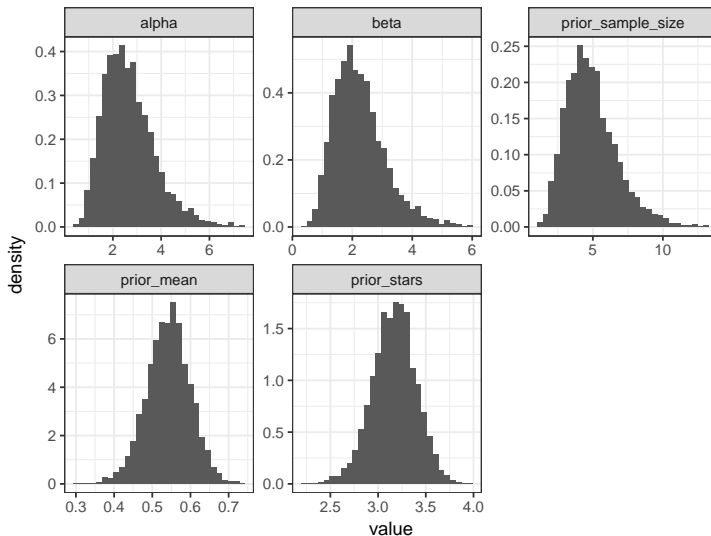
So

- $\alpha + \beta$ is the prior sample size
- $E[\theta_i | \alpha, \beta] = \frac{\alpha}{\alpha + \beta}$ is the prior expectation for the probability

But we might want to show results on the original scale (stars), so the expected number of stars for a new product is

$$\begin{aligned} E[\text{stars}_{*j} | \alpha, \beta] &= E[y_{*j} + 1 | \alpha, \beta] = E[y_{*j} | \alpha, \beta] + 1 \\ &= E[E[y_{*j} | \theta^*] | \alpha, \beta] + 1 = E[4\theta^* | \alpha, \beta] + 1 \\ &= 4 \frac{\alpha}{\alpha + \beta} + 1 \end{aligned}$$

Other parameter posteriors



Uniform use of star ratings

This binomial model has the proper support $\{0, 1, 2, 3, 4\}$ for stars minus 1, but does it have the correct proportion of observations in each star category?

As an example, $\hat{\theta}_2 = 0.81$ (and a 95% CI is $(0.79, 0.83)$). Thus, we would expect

stars	theoretical	observed
1	0.001	0.076
2	0.022	0.065
3	0.142	0.053
4	0.404	0.146
5	0.430	0.660

But this ignores the uncertainty in θ_2 , so perhaps this difference is due to this uncertainty.

Posterior predictive pvalue

To assess this model fit, we will simulate posterior predictive star ratings for product 2 and compare to the observed ratings:

product_id	n1	n2	n3	n4	n5	n_total
B001EFMD8W	40	34	28	77	347	526

Let \tilde{y}_2 be all the predictive data for product 2, i.e. $\tilde{y}_2 = (\tilde{y}_{21}, \dots, \tilde{y}_{2J})$ with $J = 526$ where \tilde{y}_{2j} is the j th predictive star rating minus 1 for review j of product 2. Then

$$p(\tilde{y}_2|y) = \int \left[\prod_{j=1}^J p(\tilde{y}_{2j}|\theta_2) \right] p(\theta_2|y) d\theta_2$$

Thus the following procedure will simulation from the joint distribution for the predictive ratings:

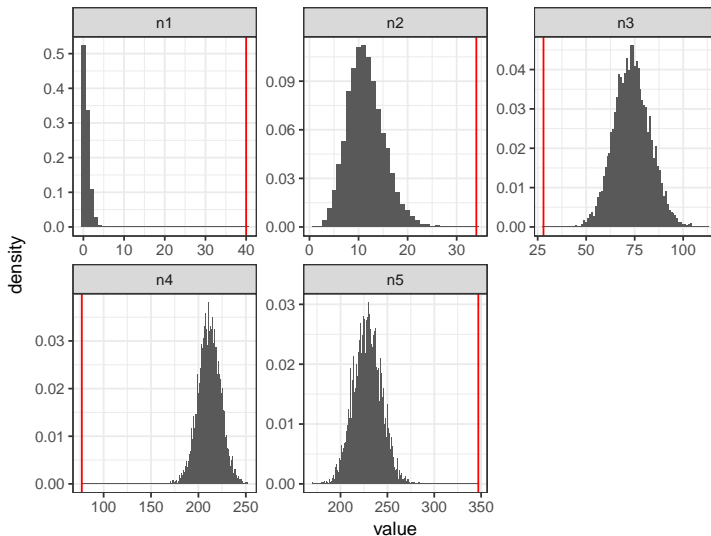
1. $\theta_2 \sim p(\theta_2|y)$,
2. For $j = 1, \dots, 526$, $y_{2j} \stackrel{ind}{\sim} \text{Bin}(4, \theta_2)$, and
3. $\text{star}_{2j} = y_{2j} + 1$.

Posterior predictive distribution in R

```
theta2 = as.numeric(draws$theta[,2])
ytilde2 = adply(theta2, 1, function(x) {
  ytilde = rbinom(526, 4, x) + 1
  data.frame(n1 = sum(ytilde==1),
             n2 = sum(ytilde==2),
             n3 = sum(ytilde==3),
             n4 = sum(ytilde==4),
             n5 = sum(ytilde==5))
})
head(ytilde2)
```

	X1	n1	n2	n3	n4	n5
1	1	0	14	63	208	241
2	2	0	11	72	208	235
3	3	0	11	68	204	243
4	4	0	5	72	182	267
5	5	0	8	89	221	208
6	6	0	8	60	217	241

Posterior predictive distribution in R



Ordinal data model

Let $y_i = (y_{i1}, \dots, y_{i5})$ be the vector of 1-star to 5-star ratings, assume

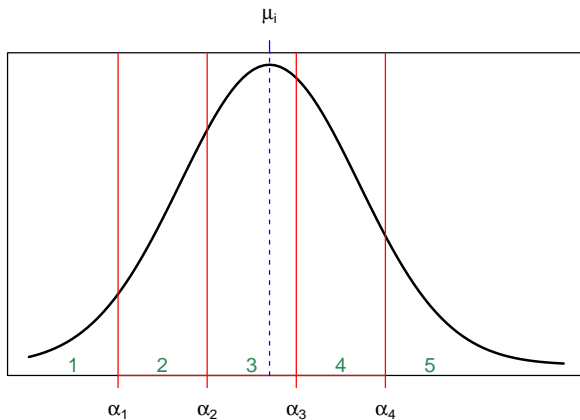
$$Y_i \stackrel{\text{ind}}{\sim} \text{Mult}(n_i, \theta_i)$$

where θ_i is a probability vector

$$\theta_{ik} = \int_{\alpha_{k-1}}^{\alpha_k} N(x|\mu_i, 1)dx = \Phi(\alpha_k - \mu_i) - \Phi(\alpha_{k-1} - \mu_i)$$

where $\alpha_0 = -\infty$, $\alpha_1 = 0$, and $\alpha_5 = \infty$, and Φ is the standard normal cumulative distribution function (cdf).

Visualizing the model



Hierarchical model

So each product has its own mean μ_i . The larger μ_i is the more 5-star ratings the product will receive and the fewer 1-star ratings the product will review.

In order to borrow information across different products, we might assume a hierarchical model for the μ_i , e.g.

$$\mu_i \stackrel{ind}{\sim} N(\eta, \tau^2)$$

with a prior

$$p(\eta, \tau) \propto Ca(\tau; 0, 1).$$


```

ordinal_model = "
data {
  int <lower=1> n_products;
  int <lower=0> y[n_products,5]; // summarized count by product
}

parameters {
  real<lower=0> alpha_diff[3];
  real mu[n_products];
  real eta;
  real<lower=0> tau;
}

transformed parameters {
  ordered[4] alpha;           // cut points
  simplex[5] theta[n_products]; // each theta vector sums to 1

  alpha[1] = 0; for (i in 1:3) alpha[i+1] = alpha[i] + alpha_diff[i];

  for (p in 1:n_products) {
    theta[p,1] = Phi(-mu[p]);
    for (j in 2:4)
      theta[p,j] = Phi(alpha[j]-mu[p]) - Phi(alpha[j-1]-mu[p]);
    theta[p,5] = 1-Phi(alpha[4]-mu[p]);
  }
}

model {
  tau ~ cauchy(0,1);
  mu ~ normal(eta, tau);
  for (p in 1:n_products) y[p] ~ multinomial(theta[p]); // n_reviews[p] is implicit
}
"
```

Fit model

```
m = stan_model(model_code = ordinal_model)
```

```
In file included from C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/config.hpp:39:0,
                 from C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/math/tools/config.hpp:13,
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                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core/geom_vv.hpp:13,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/core/geom.hpp:12,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math/rev/mat.hpp:4,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/stan/math.hpp:4,
                 from C:/Users/jarad/Documents/R/win-library/3.3/StanHeaders/include/src/stan/model/model_header.hpp:4,
                 from file263439e75e6f.cpp:8:
```

```
C:/Users/jarad/Documents/R/win-library/3.3/BH/include/boost/config/compiler/gcc.hpp:186:0: warning: "BOOST_NO_CXX11_RVALUE_REFERENCES"
# define BOOST_NO_CXX11_RVALUE_REFERENCES
^
```

```
<command-line>:0:0: note: this is the location of the previous definition
cc1plus.exe: warning: unrecognized command line option "-Wno-ignored-attributes"
```

```
dat = list(n_products = nrow(for_table),
           y = as.matrix(for_table[,2:6]))
r = sampling(m, dat, pars = c("alpha", "eta", "tau", "mu"))
```

```
SAMPLING FOR MODEL 'b056f9f334941d8210fb48d1614408e7' NOW (CHAIN 1).
```

```
Chain 1, Iteration: 1 / 2000 [ 0%] (Warmup)
Chain 1, Iteration: 200 / 2000 [ 10%] (Warmup)
Chain 1, Iteration: 400 / 2000 [ 20%] (Warmup)
Chain 1, Iteration: 600 / 2000 [ 30%] (Warmup)
Chain 1, Iteration: 800 / 2000 [ 40%] (Warmup)
```

Fit model

r

Inference for Stan model: b056f9f334941d8210fb48d1614408e7.

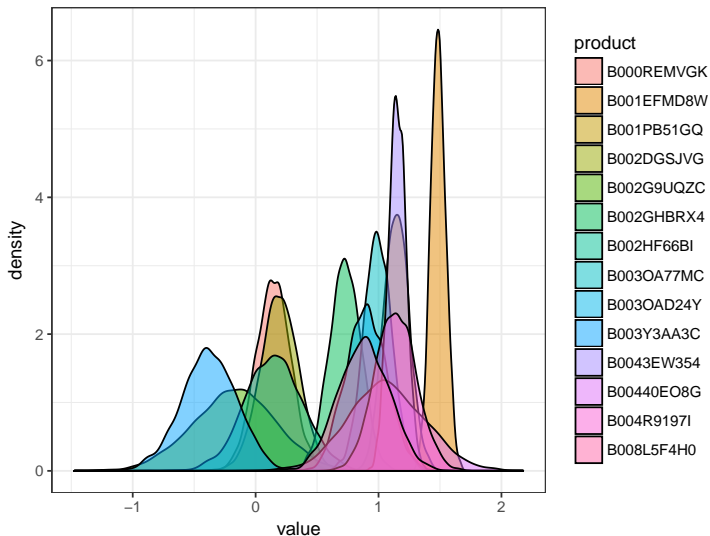
4 chains, each with iter=2000; warmup=1000; thin=1;

post-warmup draws per chain=1000, total post-warmup draws=4000.

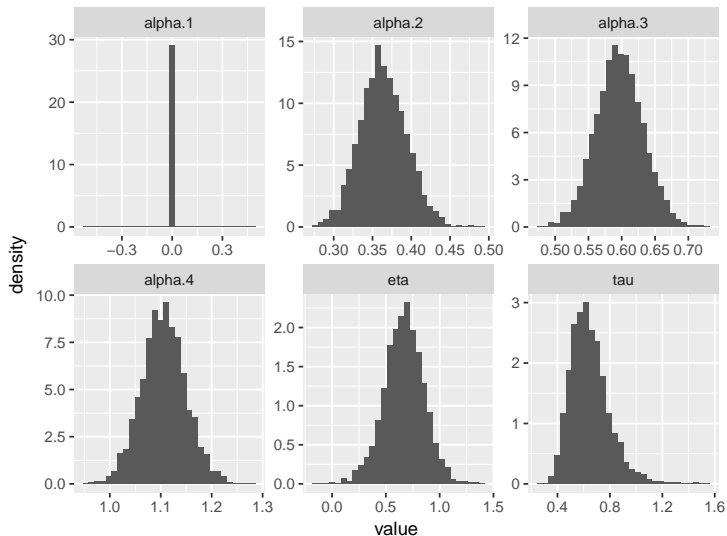
	mean	se_mean	sd	2.5%	25%	50%	75%	97.5%	n_eff	Rhat
alpha[1]	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	4000	NaN
alpha[2]	0.36	0.00	0.03	0.31	0.34	0.36	0.38	0.42	4000	1
alpha[3]	0.60	0.00	0.04	0.53	0.57	0.60	0.62	0.67	4000	1
alpha[4]	1.11	0.00	0.04	1.02	1.08	1.11	1.13	1.19	4000	1
eta	0.67	0.00	0.18	0.29	0.55	0.67	0.79	1.03	4000	1
tau	0.65	0.00	0.15	0.42	0.54	0.63	0.73	1.01	4000	1
mu[1]	0.14	0.00	0.14	-0.14	0.05	0.14	0.24	0.42	4000	1
mu[2]	1.49	0.00	0.06	1.37	1.44	1.48	1.53	1.61	4000	1
mu[3]	1.15	0.00	0.10	0.95	1.08	1.15	1.21	1.35	4000	1
mu[4]	0.19	0.00	0.15	-0.11	0.09	0.19	0.30	0.49	4000	1
mu[5]	-0.18	0.01	0.34	-0.84	-0.40	-0.17	0.05	0.46	4000	1
mu[6]	0.73	0.00	0.13	0.48	0.64	0.73	0.81	0.97	4000	1
mu[7]	0.15	0.00	0.23	-0.32	-0.01	0.15	0.30	0.61	4000	1
mu[8]	0.99	0.00	0.12	0.76	0.92	0.99	1.07	1.22	4000	1
mu[9]	0.90	0.00	0.16	0.59	0.79	0.90	1.01	1.22	4000	1
mu[10]	-0.38	0.00	0.22	-0.84	-0.53	-0.38	-0.23	0.05	4000	1
mu[11]	1.15	0.00	0.07	1.02	1.10	1.15	1.20	1.29	4000	1
mu[12]	1.07	0.00	0.31	0.47	0.86	1.06	1.27	1.69	4000	1
mu[13]	1.14	0.00	0.17	0.82	1.03	1.14	1.25	1.47	4000	1
mu[14]	0.88	0.00	0.21	0.48	0.74	0.88	1.03	1.27	4000	1
lp__	-1835.78	0.08	3.17	-1843.41	-1837.58	-1835.43	-1833.50	-1830.83	1688	1

Samples were drawn using NUTS(diag_e) at Thu Feb 23 06:27:45 2017.

Review mean posteriors (θ_i)



Other parameter posteriors



Visualizing the model

