

E3 rate

R. Cassano-Coleman

2025-10-02

This notebook takes response rates by subject, log-transforms them, and runs a Bayesian version of a mixed effects model.

```
set.seed(15000)
```

Load the data.

```
data <- read_csv('../data/E3/response_rate_by_sub.csv', show_col_types = FALSE)
```

Change musician and scramble into a factor.

```
data %<>% mutate(Musician = factor(Musician, levels = c('Yes', 'No')),  
                 scramble = factor(scramble, levels = c('Intact', '8B', '2B', '1B')))
```

Set Intact as reference level.

```
contrasts(data$scramble) <- contr.treatment(4)
```

Set the musician/non-musician contrast.

```
contrasts(data$Musician) <- c(-1,1)  
print(contrasts(data$Musician))
```

```
##      [,1]  
## Yes   -1  
## No     1
```

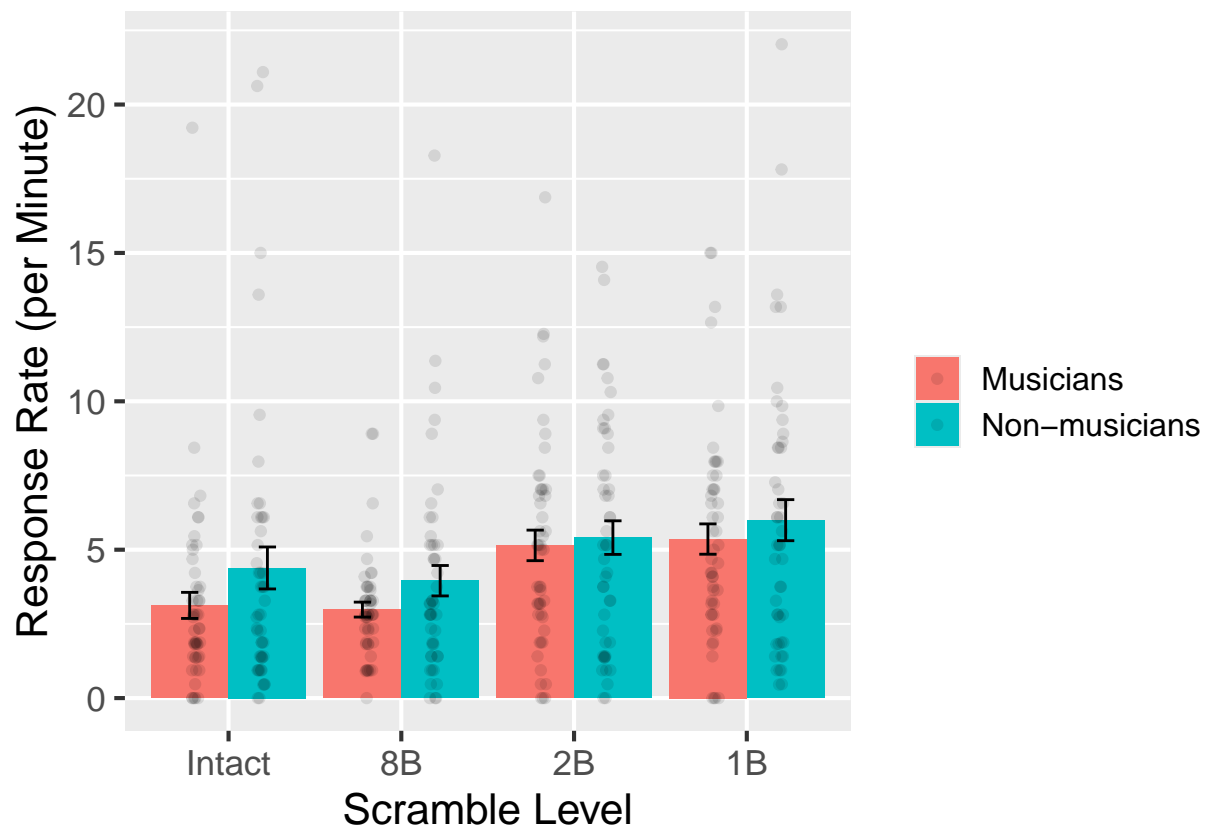
Check normality of the true response rate data. (Ignoring stimulus set.)

```
data %>%
  group_by(Musician, scramble) %>%
  shapiro_test(mean_response_rate)
```

```
## # A tibble: 8 x 5
##   Musician scramble variable      statistic      p
##   <fct>    <fct>    <chr>          <dbl>    <dbl>
## 1 Yes      Intact    mean_response_rate  0.724 0.0000000283
## 2 Yes      8B        mean_response_rate  0.863 0.0000429
## 3 Yes      2B        mean_response_rate  0.941 0.0162
## 4 Yes      1B        mean_response_rate  0.929 0.00575
## 5 No       Intact    mean_response_rate  0.744 0.000000135
## 6 No       8B        mean_response_rate  0.839 0.0000157
## 7 No       2B        mean_response_rate  0.948 0.0401
## 8 No       1B        mean_response_rate  0.894 0.000530
```

```
data %>%
  ggplot(aes(x = scramble, y = mean_response_rate, fill = Musician)) +
  geom_bar(position = "dodge", stat = "summary", fun = mean) +
  geom_errorbar(position = position_dodge(width = 0.9), width = 0.2, stat = "summary") +
  geom_point(position = position_jitterdodge(jitter.width = 0.1), alpha = 0.1) +
  theme_gray(base_size = 16) +
  xlab('Scramble Level') +
  ylab('Response Rate (per Minute)') +
  scale_fill_discrete(name="", labels=c('Musicians', 'Non-musicians')) +
  theme(legend.text = element_text(size = 12))
```

```
## No summary function supplied, defaulting to `mean_se()`
```



Log-transform the rates and check for normality.

```
data %<>% mutate(log_rate = log(1 + mean_response_rate))  
# add 1 so rates that are zero transform to 0 (rather than negative infinity)
```

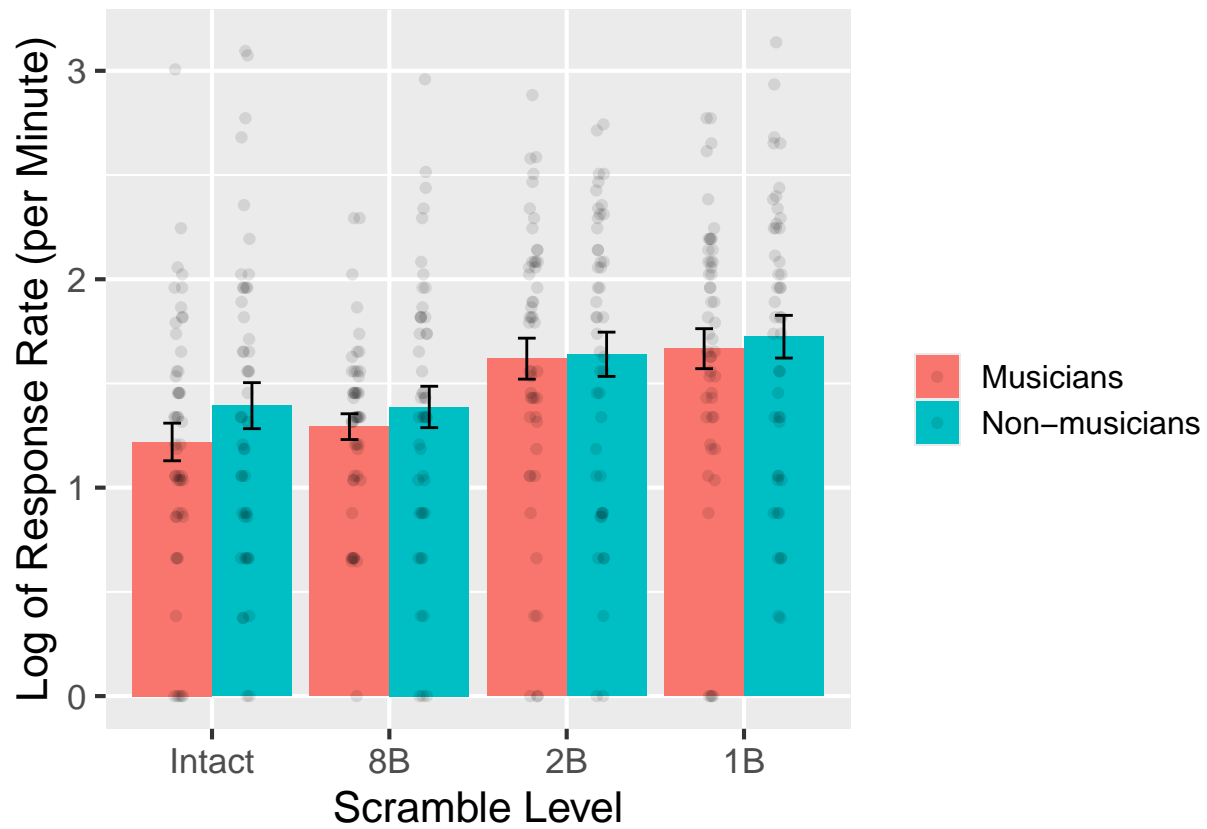
```
data %>%  
  group_by(Musician, scramble) %>%  
  shapiro_test(log_rate)
```

```
## # A tibble: 8 x 5  
##   Musician scramble variable statistic      p  
##   <fct>    <fct>    <chr>      <dbl>  <dbl>  
## 1 Yes      Intact    log_rate    0.963 0.130  
## 2 Yes      8B        log_rate    0.947 0.0284  
## 3 Yes      2B        log_rate    0.943 0.0193  
## 4 Yes      1B        log_rate    0.913 0.00153  
## 5 No      Intact    log_rate    0.976 0.457  
## 6 No      8B        log_rate    0.982 0.674  
## 7 No      2B        log_rate    0.946 0.0319  
## 8 No      1B        log_rate    0.973 0.358
```

Some of these are worse than others. Visualize:

```
data %>%  
  ggplot(aes(x = scramble, y = log_rate, fill = Musician)) +  
  geom_bar(position = "dodge", stat = "summary", fun = mean) +  
  geom_errorbar(position = position_dodge(width = 0.9), width = 0.2, stat = "summary") +  
  geom_point(position = position_jitterdodge(jitter.width = 0.1), alpha = 0.1) +  
  theme_gray(base_size = 16) +  
  xlab('Scramble Level') +  
  ylab('Log of Response Rate (per Minute)') +  
  scale_fill_discrete(name="", labels=c('Musicians', 'Non-musicians')) +  
  theme(legend.text = element_text(size = 12))
```

```
## No summary function supplied, defaulting to `mean_se()`
```



```
#ggsave('log_rate.png', width = 7, height = 5)
```

It seems like this lack of normality is driven by the zero rates.

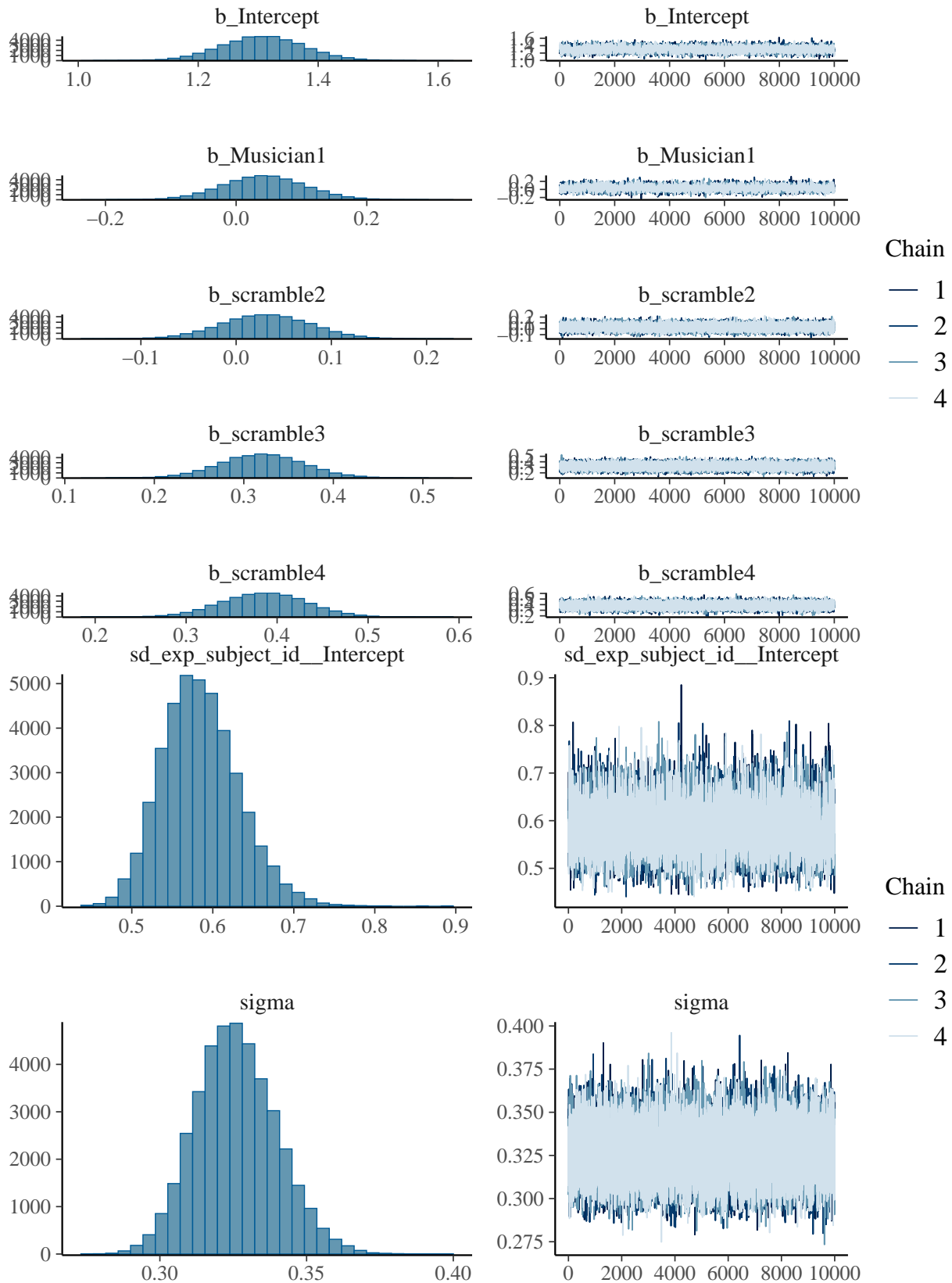
```
get_prior(log_rate ~ Musician + scramble + (1|exp_subject_id), data = data)
```

```
##           prior      class      coef      group resp dpar nlpar lb
##           (flat)         b
##           (flat)         b Musician1
##           (flat)         b scramble2
##           (flat)         b scramble3
##           (flat)         b scramble4
## student_t(3, 1.5, 2.5) Intercept
## student_t(3, 0, 2.5)      sd
## student_t(3, 0, 2.5)      sd      exp_subject_id
## student_t(3, 0, 2.5)      sd Intercept exp_subject_id
## student_t(3, 0, 2.5)      sigma
## ub      source
##      default
## (vectorized)
## (vectorized)
## (vectorized)
## (vectorized)
##      default
##      default
## (vectorized)
## (vectorized)
##      default
```

```
these_priors <- c(
  set_prior('normal(0, 0.5)', coef = "Musician1"), # don't necessarily expect a difference between groups
  set_prior('normal(0, 0.5)', coef = "scramble2"), # intact vs 8B
  set_prior('normal(0, 0.5)', coef = "scramble3"), # intact vs 2B
  set_prior('normal(0, 0.5)', coef = "scramble4") # intact vs 1B
)
```

```
brm_log_rate <- brm(log_rate ~ Musician + scramble + (1|exp_subject_id), data = data,
  prior = these_priors,
  save_pars = save_pars(all = TRUE),
  iter = 20000, refresh = 0,
  file = 'models/E3_log_rate')
```

```
plot(brm_log_rate)
```



```

print(summary(brm_log_rate, robust = TRUE), digits = 4)

## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: log_rate ~ Musician + scramble + (1 | exp_subject_id)
## Data: data (Number of observations: 380)
## Draws: 4 chains, each with iter = 20000; warmup = 10000; thin = 1;
## total post-warmup draws = 40000
##
## Multilevel Hyperparameters:
## ~exp_subject_id (Number of levels: 95)
##      Estimate Est.Error l-95% CI u-95% CI   Rhat Bulk_ESS Tail_ESS
## sd(Intercept)   0.5821    0.0467   0.5005   0.6861 1.0005     5899    11676
##
## Regression Coefficients:
##      Estimate Est.Error l-95% CI u-95% CI   Rhat Bulk_ESS Tail_ESS
## Intercept    1.3075    0.0681   1.1729   1.4420 1.0014     4194     8832
## Musician1     0.0427    0.0624  -0.0799   0.1664 1.0012     3209     6536
## scramble2     0.0312    0.0470  -0.0606   0.1228 1.0000    28358    29264
## scramble3     0.3208    0.0465   0.2289   0.4132 1.0001    28158    29971
## scramble4     0.3863    0.0470   0.2935   0.4771 1.0000    28591    28847
##
## Further Distributional Parameters:
##      Estimate Est.Error l-95% CI u-95% CI   Rhat Bulk_ESS Tail_ESS
## sigma   0.3254    0.0138   0.3004   0.3545 1.0000     31495     30490
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).

```



```
emm_log_rate <- emmeans(brm_log_rate, specs = c("scramble", "Musician"))
summary(emm_log_rate)
```

```
##  scramble Musician emmean lower.HPD upper.HPD
##  Intact    Yes      1.26      1.09      1.45
##  8B        Yes      1.30      1.12      1.47
##  2B        Yes      1.59      1.41      1.76
##  1B        Yes      1.65      1.47      1.82
##  Intact    No       1.35      1.17      1.54
##  8B        No       1.38      1.19      1.57
##  2B        No       1.67      1.49      1.86
##  1B        No       1.74      1.55      1.92
```

```
##
## Point estimate displayed: median
## HPD interval probability: 0.95
```

```
emm_log_rate_s <- emmeans(brm_log_rate, specs = "scramble")
summary(emm_log_rate_s)
```

```
##  scramble emmean lower.HPD upper.HPD
##  Intact    1.31      1.17      1.44
##  8B        1.34      1.20      1.47
##  2B        1.63      1.49      1.76
##  1B        1.69      1.56      1.83
```

```
##
## Results are averaged over the levels of: Musician
## Point estimate displayed: median
## HPD interval probability: 0.95
```

```
contrast(emm_log_rate_s, method = "pairwise")
```

```
##  contrast      estimate lower.HPD upper.HPD
##  Intact - 8B  -0.0312    -0.121    0.0624
##  Intact - 2B  -0.3208    -0.412   -0.2284
##  Intact - 1B  -0.3863    -0.476   -0.2927
##  8B - 2B      -0.2894    -0.382   -0.1968
##  8B - 1B      -0.3547    -0.449   -0.2641
##  2B - 1B      -0.0652    -0.158    0.0274
```

```
##
## Results are averaged over the levels of: Musician
## Point estimate displayed: median
## HPD interval probability: 0.95
```

```
log_rate_BF <- describe_posterior(brm_log_rate,
                                  estimate = "median", dispersion = TRUE,
                                  ci = .95, ci_method = "HDI",
                                  test = c("bayes_factor"))
print(log_rate_BF, digits = 4)
```

Summary of Posterior Distribution

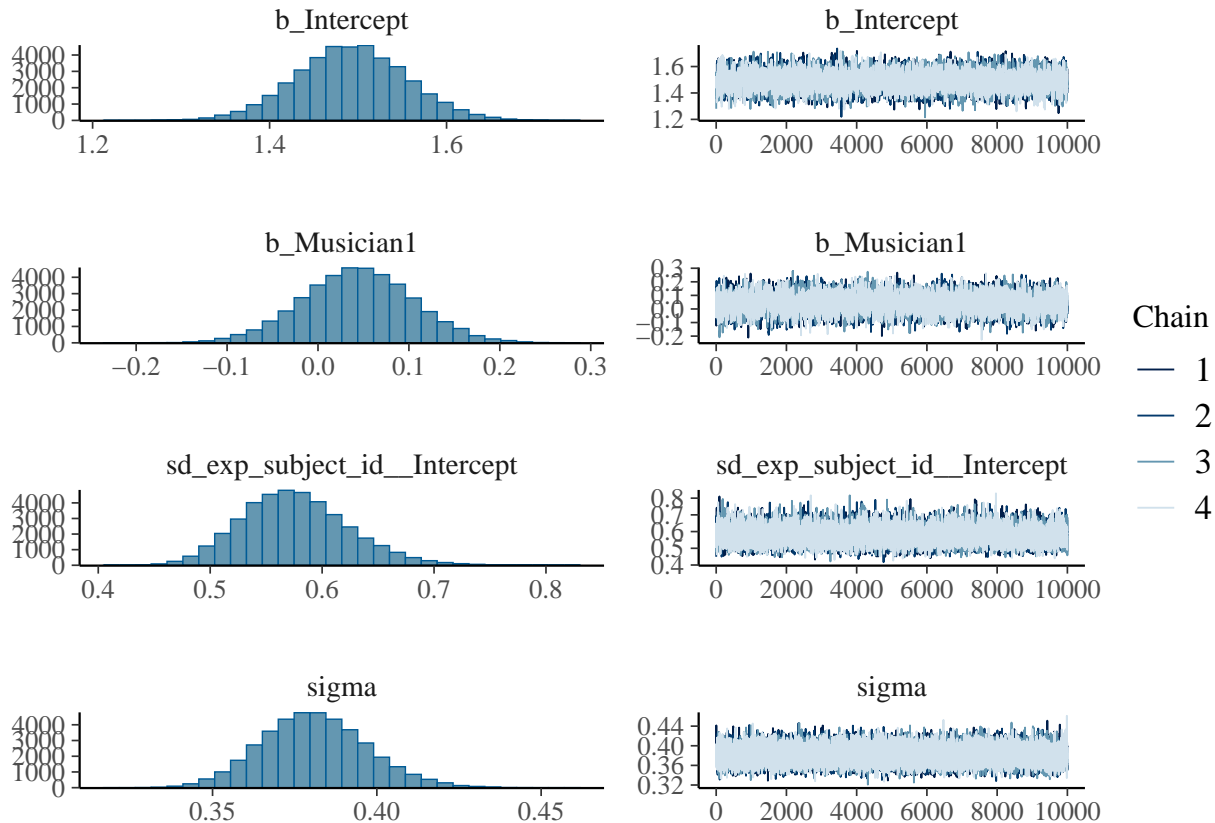
##

## Parameter	Median	MAD	95% CI	BF	Rhat	ESS
##	-----	-----	-----	-----	-----	-----
## (Intercept)	1.3075	0.0681	[1.17, 1.44]	3.57e+19	1.001	4186.0000
## Musician1	0.0427	0.0624	[-0.08, 0.16]	0.154	1.001	3199.0000
## scramble2	0.0312	0.0470	[-0.06, 0.12]	0.117	1.000	28320.0000
## scramble3	0.3208	0.0465	[0.23, 0.41]	5.45e+06	1.000	28127.0000
## scramble4	0.3863	0.0470	[0.29, 0.48]	5.05e+07	1.000	28556.0000

Compare the full model to a model without scramble condition.

```
brm_log_rate_null <- brm(log_rate ~ Musician + (1|exp_subject_id), data = data,
  prior = set_prior('normal(0, 0.5)', class = 'b'),
  save_pars = save_pars(all = TRUE),
  iter = 20000, refresh = 0,
  file = 'models/E3_log_rate_null')
```

```
plot(brm_log_rate_null)
```



```
print(summary(brm_log_rate_null, robust = TRUE), digits = 4)
```

```
## Family: gaussian
## Links: mu = identity; sigma = identity
## Formula: log_rate ~ Musician + (1 | exp_subject_id)
## Data: data (Number of observations: 380)
## Draws: 4 chains, each with iter = 20000; warmup = 10000; thin = 1;
## total post-warmup draws = 40000
##
## Multilevel Hyperparameters:
## ~exp_subject_id (Number of levels: 95)
## Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## sd(Intercept) 0.5734 0.0468 0.4909 0.6796 1.0004 8462 14860
##
## Regression Coefficients:
## Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS
## Intercept 1.4923 0.0617 1.3702 1.6157 1.0002 5632 11148
## Musician1 0.0437 0.0608 -0.0784 0.1637 1.0005 6158 11569
##
```

```

## Further Distributional Parameters:
##      Estimate Est.Error l-95% CI u-95% CI   Rhat Bulk_ESS Tail_ESS
## sigma   0.3806    0.0160   0.3510   0.4143 1.0002   38097   30501
##
## Draws were sampled using sampling(NUTS). For each parameter, Bulk_ESS
## and Tail_ESS are effective sample size measures, and Rhat is the potential
## scale reduction factor on split chains (at convergence, Rhat = 1).
BF_log_rate <- bayes_factor(brm_log_rate, brm_log_rate_null)

## Iteration: 1
## Iteration: 2
## Iteration: 3
## Iteration: 4
## Iteration: 5
## Iteration: 1
## Iteration: 2
## Iteration: 3
## Iteration: 4
## Iteration: 5
print(BF_log_rate)

## Estimated Bayes factor in favor of brm_log_rate over brm_log_rate_null: 25486133621873056.00000

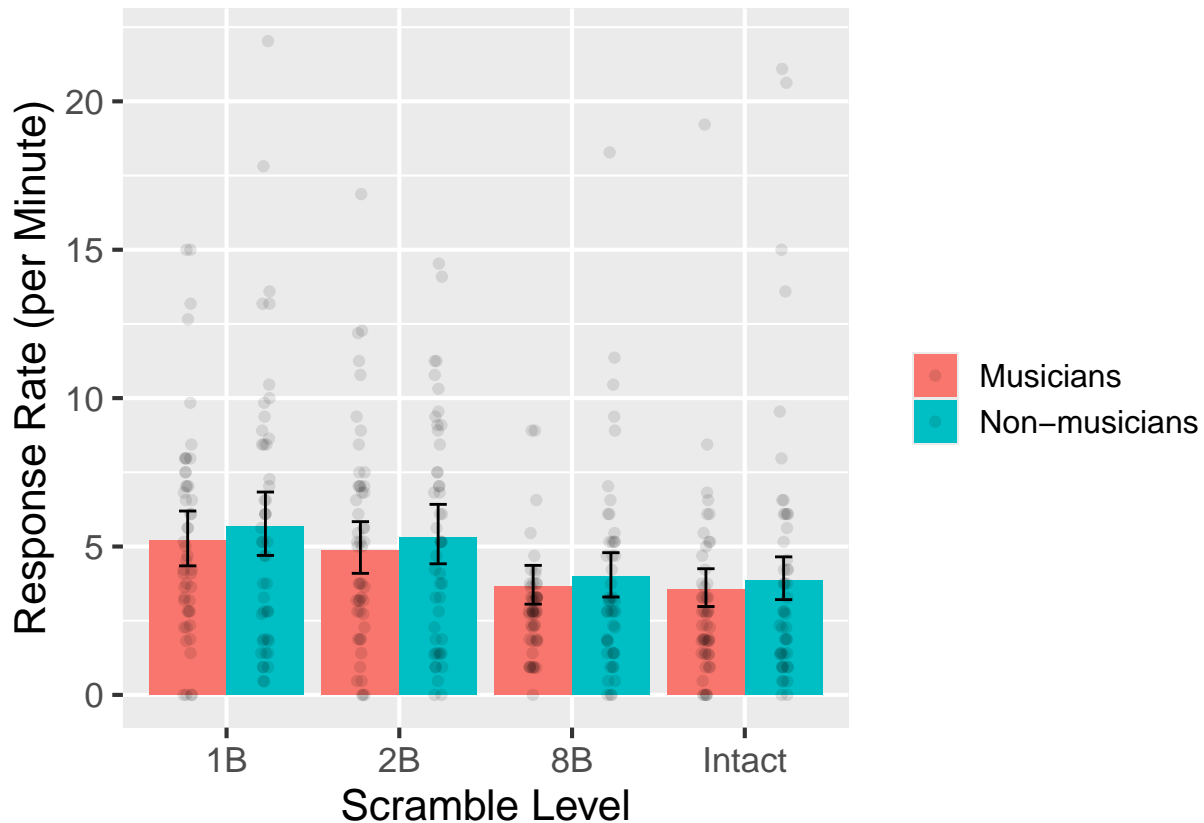
```

Visualize with posterior estimates and 95% CrI

```
posterior_est <- as.data.frame(emm_log_rate)
```

```
ggplot() +  
  geom_col(aes(x = scramble, y = exp(emmean), fill = Musician), data = posterior_est,  
            position = "dodge") +  
  geom_errorbar(aes(x = scramble, ymin = exp(lower.HPD), ymax = exp(upper.HPD), fill = Musician),  
                data = posterior_est, position = position_dodge(width = 0.9), width = 0.2) +  
  geom_point(aes(x = scramble, y = mean_response_rate, fill = Musician), data = data,  
             position = position_jitterdodge(dodge.width = 0.9, jitter.width = 0.1), alpha = 0.1) +  
  theme_gray(base_size = 16) +  
  scale_x_discrete(limits = rev) +  
  xlab('Scramble Level') +  
  ylab('Response Rate (per Minute)') +  
  scale_fill_discrete(name="", labels=c('Musicians', 'Non-musicians')) +  
  theme(legend.text = element_text(size = 12))
```

```
## Warning in geom_errorbar(aes(x = scramble, ymin = exp(lower.HPD), ymax =  
## exp(upper.HPD), : Ignoring unknown aesthetics: fill
```



```
ggsave('../figures/Fig3_rate.png', width = 7, height = 5)
```