# **FACES**

### phil newsome (12.5.2024)

```
here() starts at /Users/philnewsome/Downloads/usc/coursework/Bayesian_Stats/final project
This is cmdstanr version 0.8.1.9000
- CmdStanR documentation and vignettes: mc-stan.org/cmdstanr
- CmdStan path: /Users/philnewsome/.cmdstan/cmdstan-2.35.0
- CmdStan version: 2.35.0
A newer version of CmdStan is available. See ?install_cmdstan() to install it.
To disable this check set option or environment variable cmdstanr_no_ver_check=TRUE.
Attaching package: 'dplyr'
The following objects are masked from 'package:stats':
    filter, lag
The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
Warning: package 'brms' was built under R version 4.3.3
Loading required package: Rcpp
```

```
Attaching package: 'brms'
The following object is masked from 'package:stats':
    ar
Warning: package 'modelsummary' was built under R version 4.3.3
`modelsummary` 2.0.0 now uses `tinytable` as its default table-drawing
  backend. Learn more at: https://vincentarelbundock.github.io/tinytable/
Revert to `kableExtra` for one session:
  options(modelsummary_factory_default = 'kableExtra')
  options(modelsummary_factory_latex = 'kableExtra')
  options(modelsummary_factory_html = 'kableExtra')
Silence this message forever:
  config_modelsummary(startup_message = FALSE)
Warning: package 'posterior' was built under R version 4.3.3
This is posterior version 1.6.0
Attaching package: 'posterior'
The following objects are masked from 'package:stats':
   mad, sd, var
The following objects are masked from 'package:base':
```

Loading 'brms' package (version 2.22.0). Useful instructions

can be found by typing help('brms'). A more detailed introduction to the package is available through vignette('brms\_overview').

%in%, match

```
Warning: package 'bayesplot' was built under R version 4.3.2
This is bayesplot version 1.11.1
- Online documentation and vignettes at mc-stan.org/bayesplot
- bayesplot theme set to bayesplot::theme_default()
   * Does _not_ affect other ggplot2 plots
   * See ?bayesplot_theme_set for details on theme setting
Attaching package: 'bayesplot'
The following object is masked from 'package:posterior':
    rhat
The following object is masked from 'package:brms':
    rhat
Warning: package 'ggplot2' was built under R version 4.3.2
Warning: package 'patchwork' was built under R version 4.3.3
```

### **Research Questions**

- (1) Does machiavellianism predict fear recognition accuracy at the prenatal timepoint? (linear regression model)
- (2) Does machiavellianism and fear accuracy predict change in amygdala response to fear faces across the transition to fatherhood? (multiple regression model)

	N	Mean	SD	Min	Max	Histogram
hit_rate_pre	34	0.62	0.27	0.18	1.07	
$hit\_rate\_post$	22	0.72	0.29	0.21	1.33	
raw_accuracy	22	0.78	0.18	0.40	1.00	
me_pre	30	43.17	7.62	32.00	65.00	
$lamyg\_mean\_prenat$	34	0.07	0.15	-0.24	0.34	
$ramyg\_mean\_prenat$	34	0.08	0.14	-0.38	0.32	
$lamyg\_mean\_postpart$	22	0.16	0.16	-0.10	0.46	
$ramyg\_mean\_postpart$	22	0.14	0.15	-0.09	0.43	
lamyg_change	22	0.06	0.24	-0.23	0.60	
ramyg_change	22	0.04	0.20	-0.24	0.40	

#### **Variables**

- 'hit\_rate\_pre' : prenatal unbiased hit-rate (raw accuracy and differential accuracy combined)
- 'raw accuracy': fear recognition accuracy
- 'lamyg\_mean\_change' : change in left amygdala activation to fear>rest contrast
- 'ramyg\_mean\_change' : change in right amygdala activation to fear>rest contrast
- 'me\_pre': prenatal machiavellianism

#### More Variables

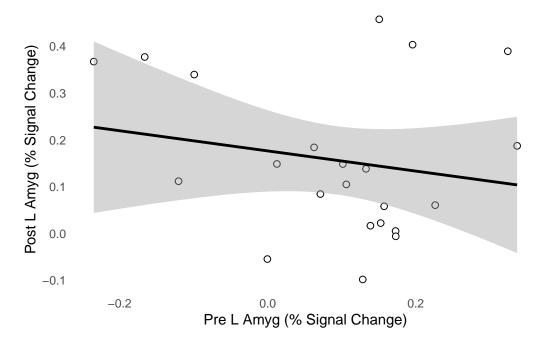
- 'lamyg\_mean\_change' : change in left amygdala activation to fear>rest contrast
- 'ramyg mean change': change in right amygdala activation to fear>rest contrast

Table @tbl-summ-var shows summary statistics

```
`geom_smooth()` using formula = 'y ~ x'
```

Warning: Removed 12 rows containing non-finite outside the scale range (`stat\_smooth()`).

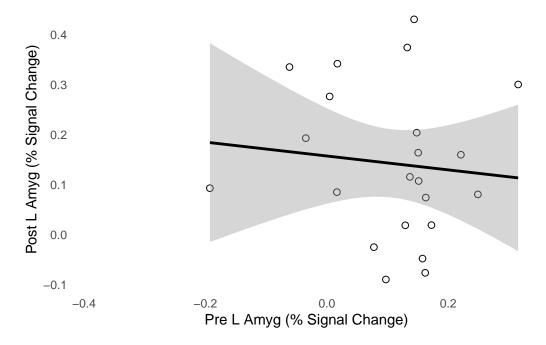
Warning: Removed 12 rows containing missing values or values outside the scale range (`geom\_point()`).



`geom\_smooth()` using formula = 'y ~ x'

Warning: Removed 12 rows containing non-finite outside the scale range (`stat\_smooth()`).

Removed 12 rows containing missing values or values outside the scale range (`geom\_point()`).



### Model 1

Let  $RAW = \text{raw\_accuracy}$ ,  $ME = \text{me\_pre}$ 

$$\begin{split} RAW_i \sim N(\mu_i, \sigma) \\ \mu_i = \beta_0 + \beta_1 M E_i \end{split}$$

priors:

$$\begin{split} \beta_0 &\sim N(0,2) \\ \beta_1 &\sim N(0,10) \\ \sigma &\sim \text{Student-t}(4,0,1) \end{split}$$

# Model 2

Let  $RAW = \text{raw\_accuracy}$ ,  $ME = \text{me\_pre}$ ,  $\Delta LAMY = \text{lamyg\_change}$ 

$$\Delta LAMY \sim N(\mu_i, \sigma)$$
 
$$\mu_i = \beta_0 + \beta_1 M E_i + \beta_2 RAW_i$$
 
$$\beta_0 \sim N(0, 2)$$

priors:

$$\beta_1 \sim N(0, 1)$$

$$\beta_2 \sim N(0, 10)$$

$$\sigma \sim \text{Student-t}(4, 0, 1)$$

# Model 3

Let  $RAW = \text{raw\_accuracy}$ ,  $ME = \text{me\_pre}$ ,  $\Delta RAMY = \text{ramyg\_change}$ 

$$\Delta RAMY \sim N(\mu_i, \sigma)$$
 
$$\mu_i = \beta_0 + \beta_1 M E_i + \beta_2 RAW_i$$

priors:

$$\begin{split} \beta_0 &\sim N(0,2) \\ \beta_1 &\sim N(0,1) \\ \beta_2 &\sim N(0,10) \\ \sigma &\sim \text{Student-t}(4,0,1) \end{split}$$

### Model 1 Analysis

#### Results

The chains mixed well.

```
Family: gaussian
  Links: mu = identity; sigma = identity
Formula: raw_accuracy ~ me_pre
   Data: faces (Number of observations: 21)
Draws: 4 chains, each with iter = 4000; warmup = 2000; thin = 1;
   total post-warmup draws = 8000
```

#### Regression Coefficients:

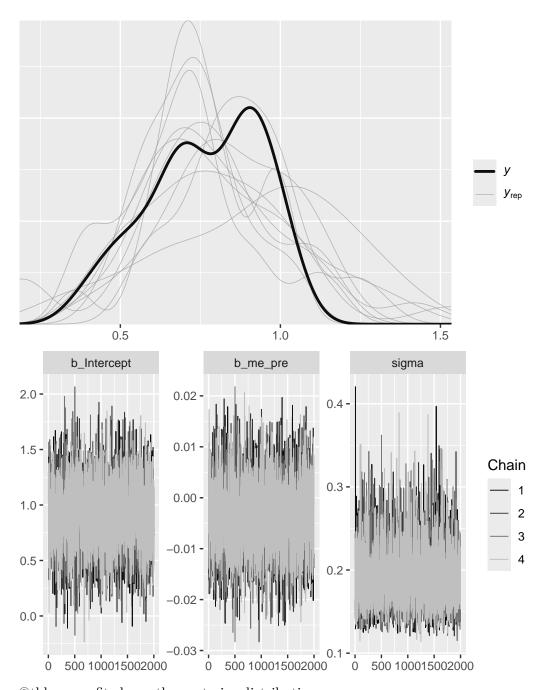
```
Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS Intercept 0.92 0.29 0.35 1.48 1.00 7283 5347 me_pre -0.00 0.01 -0.02 0.01 1.00 7324 4975
```

#### Further Distributional Parameters:

```
Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS sigma 0.19 0.03 0.14 0.27 1.00 5561 4812
```

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).

Using 10 posterior draws for ppc type 'dens\_overlay' by default.



@tbl-summ-fit shows the posterior distributions

Table 1: Posterior summary of the model parameters.

variable	mean	median	$\operatorname{sd}$	mad	q5	q95	rhat	ess_bulk	ess_tail
b_Intercept	0.92	0.92	0.29	0.27	0.45	1.39	1	7282.56	5347.31
$b\_me\_pre$	0.00	0.00	0.01	0.01	-0.01	0.01	1	7324.46	4975.09
sigma	0.19	0.19	0.03	0.03	0.15	0.26	1	5561.20	4811.62

Using all posterior draws for ppc type 'intervals' by default.

Warning: The following aesthetics were dropped during statistical transformation: ymin and ymax.

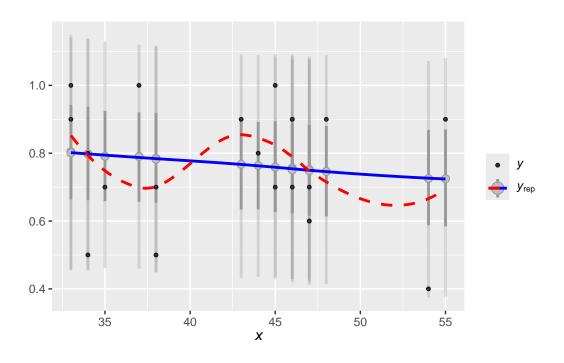
- i This can happen when ggplot fails to infer the correct grouping structure in the data.
- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?

Warning: The following aesthetics were dropped during statistical transformation: ymin and ymax.

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<sup>&#</sup>x27;geom\_smooth()' using method = 'loess' and formula = 'y ~ x'

<sup>&#</sup>x27;geom\_smooth()' using method = 'loess' and formula = 'y ~ x'



# Model 2 Analysis

#### Results

The chains mixed well.

Family: gaussian

Links: mu = identity; sigma = identity

Formula: lamyg\_change ~ raw\_accuracy + me\_pre  $\dot{}$ 

Data: faces (Number of observations: 21)

Draws: 4 chains, each with iter = 4000; warmup = 2000; thin = 1;

total post-warmup draws = 8000

#### Regression Coefficients:

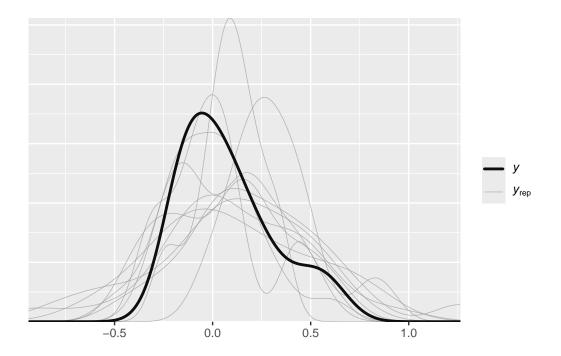
	Estimate	Est.Error	1-95% CI	u-95% CI	Rhat	Bulk_ESS	Tail_ESS
Intercept	0.09	0.47	-0.85	1.00	1.00	8081	5531
raw_accuracy	-0.29	0.31	-0.90	0.35	1.00	8078	5219
me pre	0.00	0.01	-0.01	0.02	1.00	8071	5151

#### Further Distributional Parameters:

Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk\_ESS Tail\_ESS sigma 0.25 0.05 0.18 0.36 1.00 6118 5270

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).

Using 10 posterior draws for ppc type 'dens\_overlay' by default.



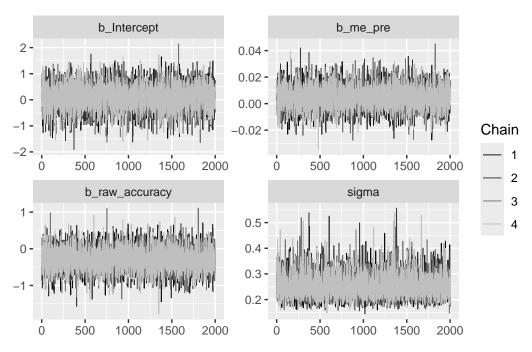


Table 2: Posterior summary of the model parameters.

variable	mean	median	$\operatorname{sd}$	mad	q5	q95	rhat	ess_bulk	ess_tail
b_Intercept	0.09	0.09	0.47	0.46	-0.67	0.85	1	8080.76	5530.93
$b\_me\_pre$	0.00	0.00	0.01	0.01	-0.01	0.02	1	8071.11	5151.43
b_raw_accuracy	-0.29	-0.29	0.31	0.31	-0.79	0.23	1	8078.40	5218.91
sigma	0.25	0.25	0.05	0.04	0.19	0.34	1	6118.33	5269.50

Using all posterior draws for ppc type 'intervals' by default. Using all posterior draws for ppc type 'intervals' by default.

Warning: The following aesthetics were dropped during statistical transformation: ymin and ymax.

- i This can happen when ggplot fails to infer the correct grouping structure in the data.
- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?

<sup>&#</sup>x27;geom\_smooth()' using method = 'loess' and formula = 'y ~ x'

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- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?

'geom\_smooth()' using method = 'loess' and formula = 'y ~ x'

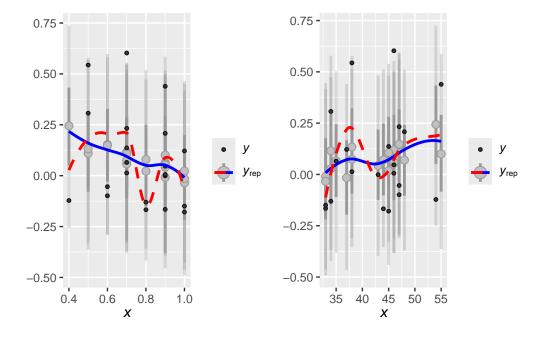
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Warning: The following aesthetics were dropped during statistical transformation: ymin and ymax.

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- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?



# Model 3 Analysis

#### Results

The chains mixed well.

```
Family: gaussian
  Links: mu = identity; sigma = identity
Formula: ramyg_change ~ raw_accuracy + me_pre
  Data: faces (Number of observations: 21)
  Draws: 4 chains, each with iter = 4000; warmup = 2000; thin = 1;
      total post-warmup draws = 8000
```

#### Regression Coefficients:

	Estimate	Est.Error	1-95% CI	u-95% CI	Rhat	Bulk_ESS	Tail_ESS
Intercept	0.02	0.41	-0.79	0.82	1.00	7710	5330
raw_accuracy	-0.07	0.27	-0.59	0.48	1.00	8552	5746
me_pre	0.00	0.01	-0.01	0.02	1.00	7666	5570

#### Further Distributional Parameters:

```
Estimate Est.Error 1-95% CI u-95% CI Rhat Bulk_ESS Tail_ESS sigma 0.22 0.04 0.15 0.31 1.00 6603 5846
```

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).

Using 10 posterior draws for ppc type 'dens\_overlay' by default.

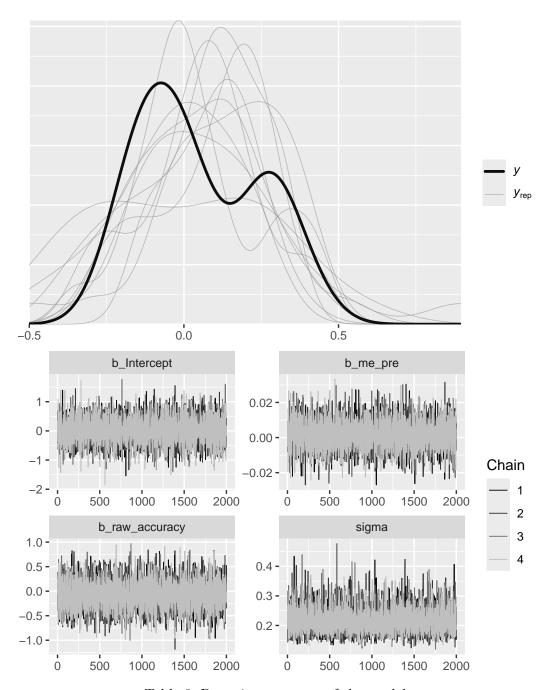


Table 3: Posterior summary of the model parameters.

variable	mean	median	$\operatorname{sd}$	mad	q5	q95	rhat	ess_bulk	ess_tail
b_Intercept	0.02	0.03	0.41	0.38	-0.64	0.69	1	7709.71	5330.03
b_me_pre	0.00	0.00	0.01	0.01	-0.01	0.01	1	7665.95	5569.52

variable	mean	median	$\operatorname{sd}$	mad	q5	q95	rhat	ess_bulk	ess_tail
b_raw_accuracy	v -0.07	-0.07	0.27	0.26	-0.51	0.37	1	8551.90	5746.41
sigma	0.22	0.21	0.04	0.04	0.16	0.29	1	6602.67	5845.85

Using all posterior draws for ppc type 'intervals' by default. Using all posterior draws for ppc type 'intervals' by default.

'geom\_smooth()' using method = 'loess' and formula = 'y ~ x'

Warning: The following aesthetics were dropped during statistical transformation: ymin and ymax.

- i This can happen when ggplot fails to infer the correct grouping structure in the data.
- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?

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- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?

'geom\_smooth()' using method = 'loess' and formula = 'y ~ x'

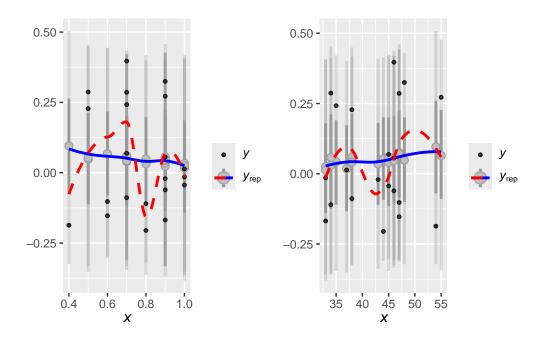
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`geom\_smooth()` using method = 'loess' and formula = 'y ~ x'

Warning: The following aesthetics were dropped during statistical transformation: ymin and ymax.

- i This can happen when ggplot fails to infer the correct grouping structure in the data.
- i Did you forget to specify a `group` aesthetic or to convert a numerical variable into a factor?



### save plots

save\_plot <- function(plot, filename, width = 10, height = 6) { ggsave(filename = filename, plot = plot, width = width, height = height, dpi = 300) }

# Save plots from different models

# Model 1 plots

ggsave("m1\_pp\_check.png", pp\_check(m1), width = 10, height = 6, dpi = 300) ggsave("m1\_marginal\_plot.png", pp\_check(m1, type = "intervals", x = "me\_pre") + geom\_smooth(se = FALSE, col = "blue") + geom\_smooth(aes( $y = y_obs$ ), se = FALSE, col = "red", linetype = "dashed"), width = 10, height = 6, dpi = 300)

 $png("m1\_trace\_plot.png", width = 10, height = 6, units = "in", res = 300) mcmc\_trace(draws\_df\_m1, pars = c("b\_Intercept", "b\_me\_pre", "sigma")) dev.off()$ 

### Model 2 plots

```
ggsave("m2_pp_check.png", pp_check(m2), width = 10, height = 6, dpi = 300) ggsave("m2_combined_plot.png", m2_combined, width = 16, height = 6, dpi = 300) png("m2_trace_plot.png", width = 10, height = 6, units = "in", res = 300) mcmc_trace(draws_df_m2, pars = c("b_Intercept", "b_me_pre", "b_lamyg_mean_prenat", "sigma")) dev.off()
```

### Model 3 plots

```
\begin{split} & ggsave(\text{``m3\_pp\_check.png''}, \ pp\_check(\text{m3}), \ width = 10, \ height = 6, \ dpi = 300) \\ & ggsave(\text{``m3\_combined\_plot.png''}, \ m3\_combined, \ width = 16, \ height = 6, \ dpi = 300) \\ & png(\text{``m3\_trace\_plot.png''}, \ width = 10, \ height = 6, \ units = \text{``in''}, \ res = 300) \ mcmc\_trace(draws\_df\_m3, \ pars = c(\text{``b\_Intercept''}, \ \text{``b\_me\_pre''}, \ \text{``b\_ramyg\_mean\_prenat''}, \ \text{``sigma''})) \ dev.off() \end{split}
```