

Negative Affect and SM Use - SMASH Study

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Descriptive Statistics

Age

```
mean(data$Age, na.rm=TRUE)
```

```
## [1] 15.82034
```

```
sd(data$Age, na.rm=TRUE)
```

```
## [1] 0.9989791
```

Race

```
table(data$Race_012, data$pid)
```

```
##
##      1002 1004 1005 1006 1007 1008 1009 1011 1013 1014 1021 1022 1023 1024 1025
## 0   724   739    0   678   432   652   695    0    0   621   337   675   698    0   817
## 1    0    0   677    0    0    0    0   989    0    0    0    0    0    0    0
## 2    0    0    0    0    0    0    0    0   834    0    0    0    0   672    0
##
##      1026 1027 1029 1030
## 0   815   704   669   602
## 1    0    0    0    0
## 2    0    0    0    0
```

```
table(data$Gender, data$pid)
```

```
##
##      1002 1004 1005 1006 1007 1008 1009 1011 1013 1014 1021 1022 1023 1024 1025
## 0    0    0   677    0    0    0    0   989   834    0    0    0    0   672   817
## 1   724    0    0   678   432   652   695    0    0   621   337   675   698    0    0
## 2    0   739    0    0    0    0    0    0    0    0    0    0    0    0
##
##      1026 1027 1029 1030
## 0    0   704    0   602
## 1   815    0   669    0
## 2    0    0    0    0
```

```
## Days in Study

# summarize max days in study
Max_days <- data %>%
  group_by(pid) %>%
  summarise(Max_day = max(day_in_study, na.rm=TRUE))

# get mean/sd day in study
mean(Max_days$Max_day, na.rm=TRUE)
```

```
## [1] 30.57895
```

```
sd(Max_days$Max_day, na.rm=TRUE)
```

```
## [1] 5.620555
```

```
## Get Means/SDs of SM time spent

sm_summary <- day %>%
  group_by %>%
  summarise(sm_time = (mean(sum_sm, na.rm=TRUE) * 60), sm_checks = mean(count_sm, na.rm=TRUE))
```

Negative Mood - Bayesian Framework

```
## Negative mood - sumduration
NA_sm_sum_bayes <- brm(NAf_pm_p ~ sum_sm_p + NAf_am_p + sum_sm_p_c + day_in_study + (1 | pid), prior =
  family = "gaussian", data = day, warmup = 2.5e3, iter = 1.5e4, thin = 1,
  chains = 4, cores = 4, seed = "123", control = list(adapt_delta = 0.999, max_treedepth = 5))

model_parameters(NA_sm_sum_bayes, centrality = "mean")
```

```
## # Fixed effects
```

```
##
```

Parameter	Mean	95% CI	pd	% in ROPE	Rhat	ESS
(Intercept)	3.16	[-119.74, 127.59]	52.05%	61.13%	1.000	70526.00
sum_sm_p	0.19	[-0.40, 0.77]	73.76%	100%	1.000	79259.00
NAf_am_p	2.15	[-0.49, 4.78]	94.42%	100%	1.000	70054.00
sum_sm_p_c	-0.19	[-1.04, 0.66]	66.96%	100%	1.000	62673.00
day_in_study	1.19	[-4.47, 6.92]	65.79%	100%	1.000	70156.00

```
##
```

```
## # Fixed effects sigma
```

```
##
```

Parameter	Mean	95% CI	pd	% in ROPE	Rhat	ESS
sigma	506.33	[470.27, 545.79]	100%	0%	1.000	76395.00

```
standard_error(NA_sm_sum_bayes)
```

```
##      Parameter      SE
## 1    b_Intercept 62.8516695
## 2    b_sum_sm_p  0.2950649
## 3    b_NAf_am_p  1.3409750
## 4    b_sum_sm_p_c 0.4319137
## 5 b_day_in_study 2.9119264
## 6          sigma 19.2549247
```

```
## Negative mood - counts
```

```
NA_sm_count_bayes <- brm(NAf_pm_p ~ count_sm_p + NAf_am_p + count_sm_p_c + day_in_study + (1 | pid), p
                        family = "gaussian", data = day, warmup = 2.5e3, iter = 1.5e4, thin = 1,
                        chains = 4, cores = 4, seed = "123", control = list(adapt_delta = 0.999, max_treedep
model_parameters(NA_sm_count_bayes, centrality = "mean")
```

```
## # Fixed effects
```

```
##
## Parameter | Mean | 95% CI | pd | % in ROPE | Rhat | ESS
## -----
## (Intercept) | 20.62 | [-98.18, 142.02] | 63.00% | 59.64% | 1.000 | 67682.00
## count_sm_p | 0.90 | [ 0.11, 1.68] | 98.75% | 100% | 1.000 | 67157.00
## NAf_am_p | 2.17 | [-0.43, 4.77] | 94.99% | 100% | 1.000 | 65563.00
## count_sm_p_c | -0.35 | [-0.90, 0.20] | 89.08% | 100% | 1.000 | 65397.00
## day_in_study | 1.73 | [-3.90, 7.33] | 72.69% | 100% | 1.000 | 68991.00
```

```
##
```

```
## # Fixed effects sigma
```

```
##
## Parameter | Mean | 95% CI | pd | % in ROPE | Rhat | ESS
## -----
## sigma | 502.10 | [466.45, 541.03] | 100% | 0% | 1.000 | 72613.00
```

```
standard_error(NA_sm_count_bayes)
```

```
##      Parameter      SE
## 1    b_Intercept 61.0762471
## 2    b_count_sm_p 0.4000597
## 3    b_NAf_am_p  1.3194582
## 4    b_count_sm_p_c 0.2816419
## 5    b_day_in_study 2.8707431
## 6          sigma 19.0431873
```

Positive Affect on SM - Within-Day Models Bayesian

```
#-----Pos affect & same day SM-----
```

```
## Positive affect & minutes of SM
```

```
PA_on_SM_day_bayes <- brm(sum_sm_p ~ SM_Pos_p + SM_Pos_p_c + day_in_study + (1 | pid), prior = prior1,
  family = "gaussian", data = day, warmup = 2.5e3, iter = 1.5e4, thin = 1,
  chains = 4, cores = 4, seed = "123", control = list(adapt_delta = 0.999, max_treedepth = 5),
  model_parameters(PA_on_SM_day_bayes, centrality = "mean")
```

```
## # Fixed effects
##
## Parameter      | Mean |          95% CI |      pd | % in ROPE | Rhat |      ESS
## -----
## (Intercept)    | 2.28 | [-25.85, 30.56] | 56.22% |    53.18% | 1.000 | 80331.00
## SM_Pos_p        | 0.24 | [ -0.31,  0.77] | 80.21% |    100% | 1.000 | 83385.00
## SM_Pos_p_c      | 0.07 | [ -0.32,  0.46] | 64.20% |    100% | 1.000 | 79000.00
## day_in_study    | -0.18 | [ -1.25,  0.90] | 62.34% |    100% | 1.000 | 82131.00
##
## # Fixed effects sigma
##
## Parameter      | Mean |          95% CI |      pd | % in ROPE | Rhat |      ESS
## -----
## sigma          | 98.95 | [92.40, 106.11] | 100% |      0% | 1.000 | 80474.00
```

```
standard_error(PA_on_SM_day_bayes)
```

```
##      Parameter      SE
## 1      b_Intercept 14.4081824
## 2      b_SM_Pos_p  0.2767788
## 3      b_SM_Pos_p_c 0.1993240
## 4 b_day_in_study  0.5510825
## 5          sigma   3.4919061
```

Positive affect & SM checks

```
PA_on_SM_count_day_bayes <- brm(count_sm_p ~ SM_Pos_p + SM_Pos_p_c + day_in_study + (1 | pid), prior =
  family = "gaussian", data = day, warmup = 2.5e3, iter = 1.5e4, thin = 1,
  chains = 4, cores = 4, seed = "123", control = list(adapt_delta = 0.999, max_treedepth = 5),
  model_parameters(PA_on_SM_count_day_bayes, centrality = "mean")
```

```
## # Fixed effects
##
## Parameter      | Mean |          95% CI |      pd | % in ROPE | Rhat |      ESS
## -----
## (Intercept)    | 2.69 | [-16.54, 21.97] | 60.86% |    52.20% | 1.000 | 77444.00
## SM_Pos_p        | 0.75 | [  0.39,  1.12] | 100.00% |    100% | 1.000 | 79619.00
## SM_Pos_p_c      | -0.04 | [ -0.30,  0.23] | 60.84% |    100% | 1.000 | 75156.00
## day_in_study    | 0.10 | [ -0.62,  0.83] | 60.95% |    100% | 1.000 | 77993.00
##
## # Fixed effects sigma
##
## Parameter      | Mean |          95% CI |      pd | % in ROPE | Rhat |      ESS
## -----
## sigma          | 66.91 | [62.53, 71.71] | 100% |      0% | 1.000 | 88627.00
```

```
standard_error(PA_on_SM_count_day_bayes)
```

```
##           Parameter           SE
## 1    b_Intercept 9.8469461
## 2    b_SM_Pos_p 0.1876314
## 3    b_SM_Pos_p_c 0.1348922
## 4 b_day_in_study 0.3715356
## 5           sigma 2.3562323
```

Negative Affect on SM

```
#-----Pos affect & same day SM-----

## Negative affect & minutes of SM

NA_on_SM_day_bayes <- brm(sum_sm_p ~ SM_Neg_p + SM_Neg_p_c + day_in_study + (1 | pid), prior = prior1,
  family = "gaussian", data = day, warmup = 2.5e3, iter = 1.5e4, thin = 1,
  chains = 4, cores = 4, seed = "123", control = list(adapt_delta = 0.999, max_treedepth = 5),
  save_pars = list(save_prior = FALSE, save_data = FALSE))

model_parameters(NA_on_SM_day_bayes, centrality = "mean")
```

```
## # Fixed effects
##
## Parameter      | Mean |          95% CI |    pd | % in ROPE | Rhat |      ESS
## -----
## (Intercept)    | 14.74 | [-4.46, 34.73] | 93.38% |    20.11% | 1.000 | 41717.00
## SM_Neg_p        |  0.24 | [-0.28,  0.77] | 81.37% |    100% | 1.000 | 67586.00
## SM_Neg_p_c      | -0.29 | [-1.13,  0.48] | 77.07% |    100% | 1.000 | 29861.00
## day_in_study    | -0.66 | [-1.57,  0.26] | 91.97% |    100% | 1.000 | 53227.00
##
## # Fixed effects sigma
##
## Parameter      | Mean |          95% CI |    pd | % in ROPE | Rhat |      ESS
## -----
## sigma          | 69.38 | [63.97, 75.42] | 100% |      0% | 1.000 | 60783.00
```

```
standard_error(NA_on_SM_day_bayes)
```

```
##           Parameter           SE
## 1    b_Intercept 9.9603126
## 2    b_SM_Neg_p 0.2687266
## 3    b_SM_Neg_p_c 0.4081199
## 4 b_day_in_study 0.4675308
## 5           sigma 2.9305403
```

Negative affect & SM checks

```
NA_on_SM_count_day_bayes <- brm(count_sm_p ~ SM_Neg_p + SM_Neg_p_c + day_in_study + (1 | pid), prior =
  family = "gaussian", data = day, warmup = 2.5e3, iter = 1.5e4, thin = 1,
  chains = 4, cores = 4, seed = "123", control = list(adapt_delta = 0.999, max_treedepth = 5),
  save_pars = list(save_prior = FALSE, save_data = FALSE))
```

```
model_parameters(NA_on_SM_count_day_bayes, centrality = "mean")
```

```
## # Fixed effects
##
## Parameter      | Mean |          95% CI |      pd | % in ROPE | Rhat |      ESS
## -----
## (Intercept)    | 12.67 | [-4.60, 30.09] | 92.61% |    23.63% | 1.000 | 69702.00
## SM_Neg_p       |  0.49 | [-0.03,  1.00] | 96.73% |    100% | 1.000 | 64540.00
## SM_Neg_p_c     |  0.09 | [-0.56,  0.73] | 61.42% |    100% | 1.000 | 55622.00
## day_in_study   | -0.62 | [-1.48,  0.24] | 91.90% |    100% | 1.000 | 63927.00
##
## # Fixed effects sigma
##
## Parameter      | Mean |          95% CI |      pd | % in ROPE | Rhat |      ESS
## -----
## sigma          | 67.64 | [62.40, 73.37] | 100% |      0% | 1.000 | 71599.00
```

```
standard_error(NA_on_SM_count_day_bayes)
```

```
##      Parameter      SE
## 1    b_Intercept 8.8099235
## 2    b_SM_Neg_p 0.2613456
## 3    b_SM_Neg_p_c 0.3279461
## 4 b_day_in_study 0.4425889
## 5          sigma 2.8079436
```