

Correlations between User Variables and Attitude Ratings

Nele Albers

06 February, 2023

Contents

Introduction	1
Setup	1
Data	1
Bayesian Analysis of Correlations	2
Summary	11
References	12

Introduction

The purpose of this file is to you to reproduce our results of the Bayesian analyses of the correlations between user variables and attitude ratings. These values are presented in the “Results”-section of the paper.

Authored by Nele Albers, Mark A. Neerincx, Nadyne L. Aretz, Mahira Ali, Arsen Ekinici, and Willem-Paul Brinkman.

Setup

First, we load the packages that we need.

```
library(BayesianFirstAid) # For computing correlations
library(formatR) # For formatting
library(ggplot2) # For plots
library(pander) # For table
```

Data

We load the pre-processed data.

```
df = read.csv(file = "preprocessed_data.csv")
```

Next, we need to compute indices for the ease of and motivation to do the preparatory activities. For each of these, participants rated two items on scales from -5 to 5.

```
# Multiply the difficulty rating by -1 so that we have an overall rating of ease
df$Ease = (df$MB_Easy - df$MB_Difficult)/2
```

```
# And similarly we multiply the unmotivated rating by -1
df$Motivation = (df$MB_Motivated - df$MB_Unmotivated)/2
```

Bayesian Analysis of Correlations

We will now conduct Bayesian analyses of the Pearson correlations between user variables and attitude ratings.

First we compute the correlation between the ease of doing the activities and the rating for “willingness to continue.”

```
set.seed(18) # For reproducibility

fit = bayes.cor.test(df$Ease, df$A3R_1)

print(summary(fit))
```

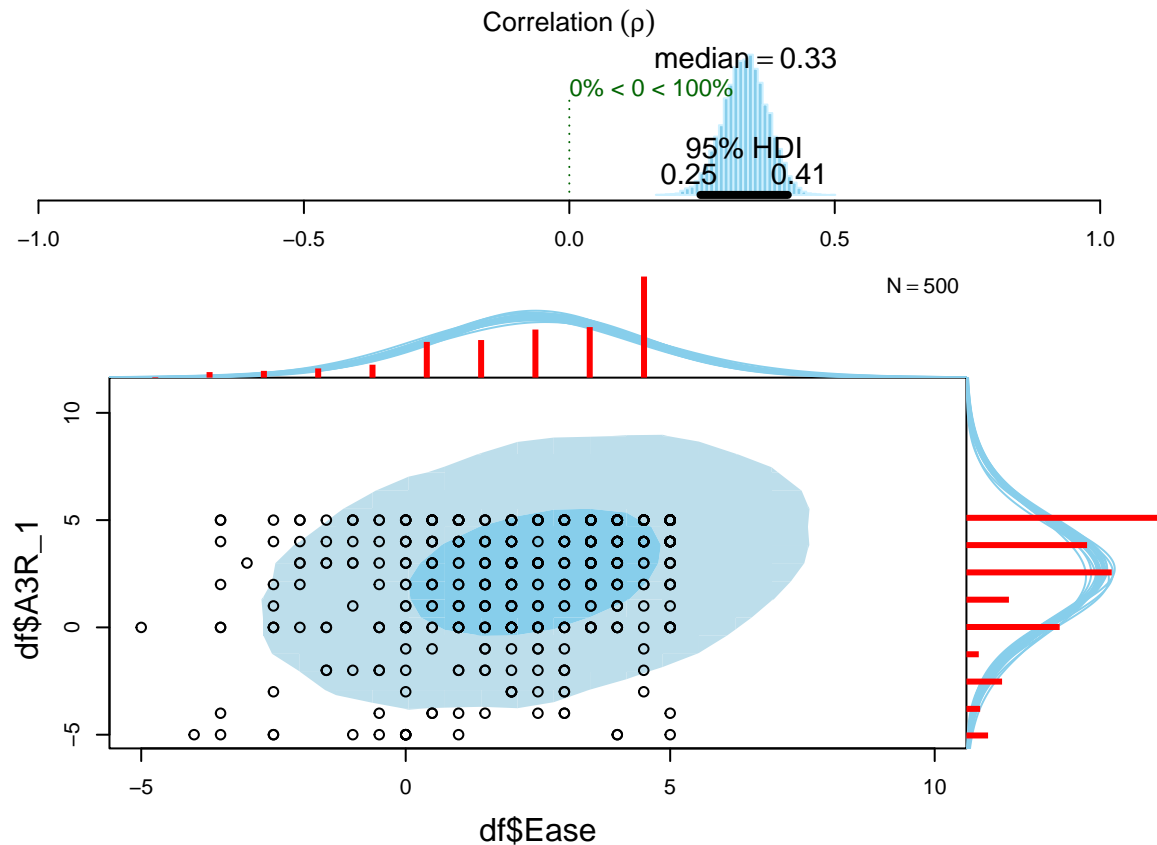
```
## Data
## df$Ease and df$A3R_1, n = 500
##
## Model parameters
## rho: the correlation between df$Ease and df$A3R_1
## mu[1]: the mean of df$Ease
## sigma[1]: the scale of df$Ease , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df$A3R_1
## sigma[2]: the scale of df$A3R_1
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df$Ease
## xy_pred[2]: the posterior predictive distribution of df$A3R_1
##
## Measures
##      mean      sd  HDIlo  HDIup %<comp %>comp
## rho      0.332  0.042  0.247  0.412  0.000  1.000
## mu[1]     2.452  0.109  2.253  2.677  0.000  1.000
## mu[2]     2.563  0.134  2.306  2.832  0.000  1.000
## sigma[1]   2.053  0.083  1.884  2.211  0.000  1.000
## sigma[2]   2.544  0.108  2.333  2.753  0.000  1.000
## nu      22.875 16.305  6.570 54.749  0.000  1.000
## xy_pred[1] 2.442  2.191 -1.794  6.798  0.126  0.874
## xy_pred[2] 2.544  2.697 -2.819  7.842  0.163  0.837
##
## 'HDIlo' and 'HDIup' are the limits of a 95% HDI credible interval.
## '%<comp' and '%>comp' are the probabilities of the respective parameter being
## smaller or larger than 0.
##
## Quantiles
##      q2.5%  q25% median  q75% q97.5%
## rho      0.247  0.304  0.332  0.360  0.412
## mu[1]     2.240  2.377  2.451  2.526  2.665
## mu[2]     2.301  2.474  2.562  2.653  2.828
## sigma[1]   1.891  1.997  2.052  2.110  2.218
## sigma[2]   2.335  2.469  2.544  2.618  2.756
## nu      8.602 13.214 18.033 26.402 68.899
```

```

## xy_pred[1] -1.820  1.010  2.434  3.864  6.784
## xy_pred[2] -2.727  0.816  2.543  4.273  7.940
##              mean          sd  HDI% comp      HDIlo      HDIup      %>comp
## rho          0.3316769  0.04209706   95    0  0.2470129  0.411584  0.9999333
## mu[1]         2.4522104  0.10891462   95    0  2.2531371  2.676688  0.9999333
## mu[2]         2.5632065  0.13397144   95    0  2.3055800  2.831898  0.9999333
## sigma[1]       2.0532468  0.08340933   95    0  1.8844121  2.210913  0.9999333
## sigma[2]       2.5438342  0.10826206   95    0  2.3327280  2.752610  0.9999333
## nu            22.8746499 16.30537570   95    0  6.5702072 54.748779  0.9999333
## xy_pred[1]     2.4424190  2.19084019   95    0 -1.7935893  6.798239  0.8739501
## xy_pred[2]     2.5435337  2.69653223   95    0 -2.8190203  7.841633  0.8373550
##              %<comp      q2.5%      q25%      median      q75%      q97.5%
## rho          6.665778e-05  0.2473846  0.3038744  0.3319636  0.3603242  0.4121233
## mu[1]         6.665778e-05  2.2400861  2.3768778  2.4513930  2.5263412  2.6648612
## mu[2]         6.665778e-05  2.3006755  2.4739998  2.5620664  2.6530763  2.8276789
## sigma[1]       6.665778e-05  1.8907453  1.9969214  2.0518253  2.1098168  2.2177119
## sigma[2]       6.665778e-05  2.3350486  2.4689657  2.5435291  2.6175131  2.7562226
## nu            6.665778e-05  8.6015471 13.2144792 18.0325505 26.4017991 68.8993540
## xy_pred[1]     1.260499e-01 -1.8200863  1.0099904  2.4338761  3.8644576  6.7843620
## xy_pred[2]     1.626450e-01 -2.7266882  0.8159219  2.5428100  4.2732223  7.9395604
##              mcmc_se      Rhat  n_eff
## rho          0.0004750926 1.0008833  7884
## mu[1]         0.0018229827 1.0000910  3567
## mu[2]         0.0023678478 0.9998613  3260
## sigma[1]       0.0015922248 1.0002772  2772
## sigma[2]       0.0020668418 1.0002884  2769
## nu            0.5149029217 1.0158083  1051
## xy_pred[1]     0.0179092480 1.0001033 14964
## xy_pred[2]     0.0216449088 1.0000771 15551

```

```
plot(fit)
```



```
# Mean, SD, HDI
ease_mean <- fit$stats[1, 1]
ease_SD <- fit$stats[1, 2]
ease_hdi_low <- fit$stats[1, 5]
ease_hdi_high <- fit$stats[1, 6]

# Posterior probability that correlation is
# greater than 0
post_prob_ease <- fit$stats[1, 7]

print(paste("Posterior probability that the correlation is greater than 0:",
  round(post_prob_ease, 5)))
```

```
## [1] "Posterior probability that the correlation is greater than 0: 0.99993"
```

This posterior probability can be evaluated based on the guidelines from (Chechile (2020)) and their extension to posterior probabilities below 0.5 by (Andraszewicz et al. (2015)).

```
if (post_prob_ease < 0.0005){
  evaluation_ease = "Nearing certainty against"
}else if (post_prob_ease < 0.005){
  evaluation_ease = "Very strong bet against"
}else if (post_prob_ease < 0.01){
  evaluation_ease = "Strong bet against - irresponsible to avoid"
}else if (post_prob_ease < 0.1){
  evaluation_ease = "A promising but risky bet against"
}else if (post_prob_ease < 0.25){
  evaluation_ease = "Only a casual bet against"
```

```

}else if (post_prob_ease < 0.5){
  evaluation_ease = "Not worth betting against"
}else if (post_prob_ease < 0.75){
  evaluation_ease = "Not worth betting on"
}else if (post_prob_ease < 0.9){
  evaluation_ease = "Only a casual bet"
}else if (post_prob_ease < 0.95){
  evaluation_ease = "A promising but risky bet"
}else if (post_prob_ease < 0.99){
  evaluation_ease = "Good bet - too good to disregard"
}else if (post_prob_ease < 0.995){
  evaluation_ease = "Strong bet - irresponsible to avoid"
}else if (post_prob_ease < 0.9995){
  evaluation_ease = "Very strong bet"
}else if (post_prob_ease < 0.99995){
  evaluation_ease = "Nearing certainty"
}else{
  evaluation_ease = "Virtually certain"
}

```

```
evaluation_ease
```

```
## [1] "Nearing certainty"
```

Next we compute the correlation between the motivation to do the activities and the rating for “willingness to continue.”

```
set.seed(18) # For reproducibility
```

```
fit = bayes.cor.test(df$Motivation, df$A3R_1)
```

```
print(summary(fit))
```

```

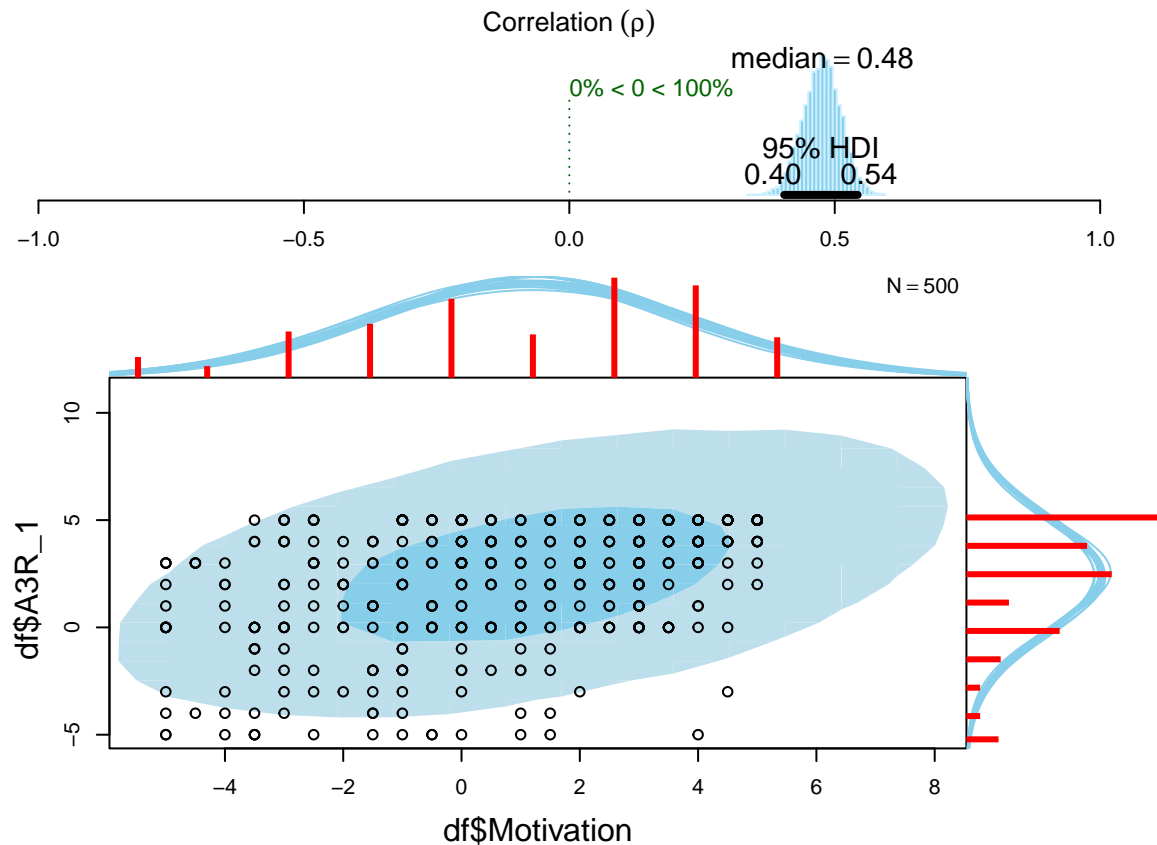
## Data
## df$Motivation and df$A3R_1, n = 500
##
## Model parameters
## rho: the correlation between df$Motivation and df$A3R_1
## mu[1]: the mean of df$Motivation
## sigma[1]: the scale of df$Motivation , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df$A3R_1
## sigma[2]: the scale of df$A3R_1
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df$Motivation
## xy_pred[2]: the posterior predictive distribution of df$A3R_1
##
## Measures
##
```

	mean	sd	HDIlo	HDIup	%<comp	%>comp
## rho	0.476	0.035	0.405	0.543	0.000	1.000
## mu[1]	1.207	0.130	0.962	1.466	0.000	1.000
## mu[2]	2.479	0.124	2.236	2.720	0.000	1.000
## sigma[1]	2.754	0.092	2.577	2.940	0.000	1.000
## sigma[2]	2.644	0.092	2.468	2.825	0.000	1.000
## nu	60.250	33.119	13.106	126.245	0.000	1.000

```

## xy_pred[1] 1.200 2.862 -4.571 6.659 0.333 0.667
## xy_pred[2] 2.493 2.722 -2.891 7.980 0.173 0.827
##
## 'HDIlo' and 'HDIup' are the limits of a 95% HDI credible interval.
## '%<comp' and '%>comp' are the probabilities of the respective parameter being
## smaller or larger than 0.
##
## Quantiles
##      q2.5%  q25% median  q75%  q97.5%
## rho      0.406  0.452  0.477  0.500  0.545
## mu[1]    0.954  1.120  1.206  1.294  1.460
## mu[2]    2.238  2.396  2.479  2.561  2.722
## sigma[1] 2.579  2.691  2.751  2.815  2.943
## sigma[2] 2.468  2.582  2.644  2.706  2.826
## nu      18.928 36.321 52.477 75.934 145.514
## xy_pred[1] -4.474 -0.678 1.208 3.132 6.764
## xy_pred[2] -2.962 0.729 2.487 4.255 7.910
##      mean      sd HDI% comp      HDIlo      HDIup      %>comp
## rho      0.4760307 0.03548331 95 0 0.4049008 0.5431083 0.9999333
## mu[1]    1.2068797 0.12966276 95 0 0.9619609 1.4656785 0.9999333
## mu[2]    2.4788748 0.12380552 95 0 2.2363070 2.7203090 0.9999333
## sigma[1] 2.7536671 0.09249617 95 0 2.5768140 2.9396572 0.9999333
## sigma[2] 2.6443602 0.09166592 95 0 2.4677732 2.8254241 0.9999333
## nu      60.2495164 33.11936342 95 0 13.1061732 126.2450739 0.9999333
## xy_pred[1] 1.1997940 2.86230125 95 0 -4.5710943 6.6589270 0.6672444
## xy_pred[2] 2.4928115 2.72217398 95 0 -2.8911189 7.9800782 0.8272897
##      %<comp      q2.5%      q25%      median      q75%      q97.5%
## rho      6.665778e-05 0.4060514 0.4523706 0.4768253 0.4999526 0.544663
## mu[1]    6.665778e-05 0.9539587 1.1200080 1.2058986 1.2939497 1.459935
## mu[2]    6.665778e-05 2.2380227 2.3959264 2.4791775 2.5613239 2.722232
## sigma[1] 6.665778e-05 2.5786035 2.6907845 2.7509644 2.8145938 2.942519
## sigma[2] 6.665778e-05 2.4681866 2.5818573 2.6442309 2.7055190 2.826010
## nu      6.665778e-05 18.9281475 36.3206773 52.4771810 75.9339953 145.514154
## xy_pred[1] 3.327556e-01 -4.4740988 -0.6777344 1.2084371 3.1324716 6.764372
## xy_pred[2] 1.727103e-01 -2.9618956 0.7288366 2.4869985 4.2547478 7.910244
##      mcmc_se      Rhat n_eff
## rho      0.0004369683 1.0009404 6687
## mu[1]    0.0018703304 1.0007332 4806
## mu[2]    0.0017282920 1.0008454 5153
## sigma[1] 0.0012060670 1.0006297 6001
## sigma[2] 0.0012075829 1.0003648 5782
## nu      0.5226830032 1.0062635 4041
## xy_pred[1] 0.0237671016 1.0002600 14523
## xy_pred[2] 0.0219271988 0.9999298 15417
plot(fit)

```



```
# Mean, SD, HDI
motiv_mean <- fit$stats[1, 1]
motiv_SD <- fit$stats[1, 2]
motiv_hdi_low <- fit$stats[1, 5]
motiv_hdi_high <- fit$stats[1, 6]

# Posterior probability that correlation is
# greater than 0
post_prob_motiv <- fit$stats[1, 7]

print(paste("Posterior probability that the correlation is greater than 0:",
  round(post_prob_motiv, 5)))
```

```
## [1] "Posterior probability that the correlation is greater than 0: 0.99993"
```

This posterior probability can be evaluated based on the guidelines from (Chechile (2020)) and their extension to posterior probabilities below 0.5 by (Andrzejewicz et al. (2015)).

```
if (post_prob_motiv < 0.0005){
  evaluation_motiv = "Nearing certainty against"
}else if (post_prob_motiv < 0.005){
  evaluation_motiv = "Very strong bet against"
}else if (post_prob_motiv < 0.01){
  evaluation_motiv = "Strong bet against - irresponsible to avoid"
}else if (post_prob_motiv < 0.1){
  evaluation_motiv = "A promising but risky bet against"
}else if (post_prob_motiv < 0.25){
  evaluation_motiv = "Only a casual bet against"
}
```

```

}else if (post_prob_motiv < 0.5){
  evaluation_motiv = "Not worth betting against"
}else if (post_prob_motiv < 0.75){
  evaluation_motiv = "Not worth betting on"
}else if (post_prob_motiv < 0.9){
  evaluation_motiv = "Only a casual bet"
}else if (post_prob_motiv < 0.95){
  evaluation_motiv = "A promising but risky bet"
}else if (post_prob_motiv < 0.99){
  evaluation_motiv = "Good bet - too good to disregard"
}else if (post_prob_motiv < 0.995){
  evaluation_motiv = "Strong bet - irresponsible to avoid"
}else if (post_prob_motiv < 0.9995){
  evaluation_motiv = "Very strong bet"
}else if (post_prob_motiv < 0.99995){
  evaluation_motiv = "Nearing certainty"
}else{
  evaluation_motiv = "Virtually certain"
}

```

```
evaluation_motiv
```

```
## [1] "Nearing certainty"
```

Now we calculate the correlation between quitter self-identity and the rating for “willingness to continue.”

```
set.seed(18) # For reproducibility
```

```
fit = bayes.cor.test(df$Quitter_Self_Identity, df$A3R_1)
```

```
print(summary(fit))
```

```

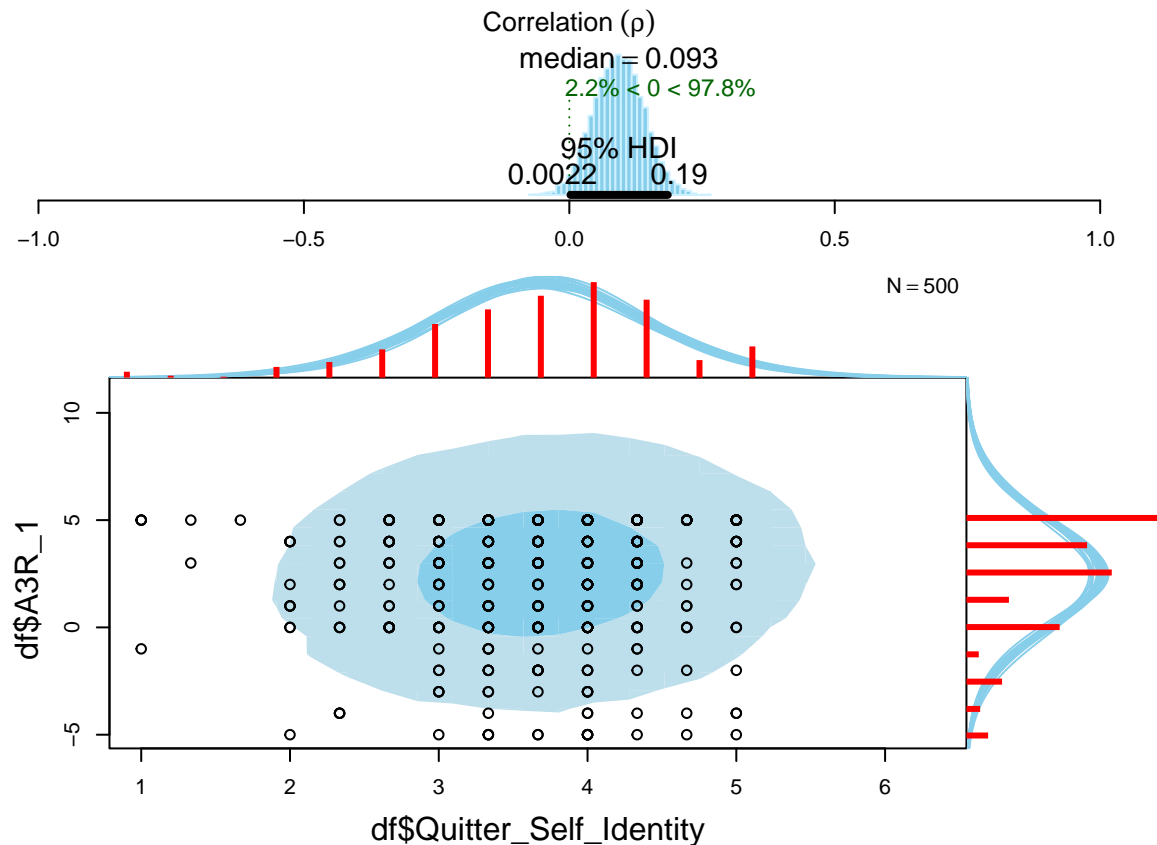
## Data
## df$Quitter_Self_Identity and df$A3R_1, n = 500
##
## Model parameters
## rho: the correlation between df$Quitter_Self_Identity and df$A3R_1
## mu[1]: the mean of df$Quitter_Self_Identity
## sigma[1]: the scale of df$Quitter_Self_Identity , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df$A3R_1
## sigma[2]: the scale of df$A3R_1
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df$Quitter_Self_Identity
## xy_pred[2]: the posterior predictive distribution of df$A3R_1
##
## Measures
##
```

	mean	sd	HDIlo	HDIup	%<comp	%>comp
## rho	0.093	0.047	0.002	0.185	0.022	0.978
## mu[1]	3.686	0.033	3.623	3.752	0.000	1.000
## mu[2]	2.558	0.131	2.306	2.813	0.000	1.000
## sigma[1]	0.711	0.029	0.654	0.767	0.000	1.000
## sigma[2]	2.543	0.107	2.326	2.748	0.000	1.000
## nu	20.400	12.185	6.854	42.981	0.000	1.000
## xy_pred[1]	3.676	0.754	2.183	5.147	0.000	1.000


```

## xy_pred[2] 2.555 2.706 -2.783 7.878 0.166 0.834
##
## 'HDIlo' and 'HDIup' are the limits of a 95% HDI credible interval.
## '%<comp' and '%>comp' are the probabilities of the respective parameter being
## smaller or larger than 0.
##
## Quantiles
##          q2.5%   q25% median   q75% q97.5%
## rho          0.003 0.061 0.093 0.125 0.186
## mu[1]         3.621 3.663 3.686 3.709 3.751
## mu[2]         2.303 2.470 2.558 2.646 2.811
## sigma[1]      0.656 0.690 0.710 0.731 0.770
## sigma[2]      2.330 2.473 2.544 2.615 2.753
## nu            8.656 12.914 16.909 23.800 50.924
## xy_pred[1]    2.185 3.187 3.677 4.169 5.152
## xy_pred[2]   -2.803 0.809 2.562 4.274 7.873
##          mean          sd HDI% comp          HDIlo          HDIup          %>comp
## rho          0.0934430 0.04701486 95 0 0.002189648 0.1854029 0.9778030
## mu[1]         3.6862472 0.03347483 95 0 3.622921593 3.7522262 0.9999333
## mu[2]         2.5581636 0.13054866 95 0 2.305807593 2.8130587 0.9999333
## sigma[1]      0.7108535 0.02923348 95 0 0.654173382 0.7673809 0.9999333
## sigma[2]      2.5430111 0.10693696 95 0 2.326481303 2.7479707 0.9999333
## nu            20.4003752 12.18492038 95 0 6.853844955 42.9810161 0.9999333
## xy_pred[1]    3.6755762 0.75423254 95 0 2.182526451 5.1465316 0.9999333
## xy_pred[2]    2.5545915 2.70599257 95 0 -2.782894789 7.8776205 0.8338222
##          %<comp          q2.5%          q25%          median          q75%
## rho          2.219704e-02 0.002538719 0.06118693 0.09337381 0.1253303
## mu[1]         6.665778e-05 3.621327874 3.66349191 3.68617529 3.7085740
## mu[2]         6.665778e-05 2.302611557 2.46950662 2.55779586 2.6461468
## sigma[1]      6.665778e-05 0.656185328 0.69037160 0.71024090 0.7309898
## sigma[2]      6.665778e-05 2.329814306 2.47268945 2.54371816 2.6149587
## nu            6.665778e-05 8.655873325 12.91378312 16.90897553 23.7995788
## xy_pred[1]    6.665778e-05 2.185140823 3.18720456 3.67701773 4.1692892
## xy_pred[2]    1.661778e-01 -2.803142623 0.80926032 2.56214711 4.2736655
##          q97.5%          mcmc_se          Rhat n_eff
## rho          0.1860244 0.0005083177 1.000184 8598
## mu[1]         3.7509872 0.0003739524 1.000789 8046
## mu[2]         2.8112726 0.0018448916 1.002481 5005
## sigma[1]      0.7698032 0.0004845776 1.000126 3672
## sigma[2]      2.7526654 0.0017830102 1.000130 3603
## nu            50.9239292 0.3438086725 1.016007 1406
## xy_pred[1]    5.1521319 0.0062595509 1.000308 14555
## xy_pred[2]    7.8727133 0.0225261584 1.000177 14461
plot(fit)

```



```
# Mean, SD, HDI
qi_mean <- fit$stats[1, 1]
qi_SD <- fit$stats[1, 2]
qi_hdi_low <- fit$stats[1, 5]
qi_hdi_high <- fit$stats[1, 6]

# Posterior probability that correlation is
# greater than 0
post_prob_qi <- fit$stats[1, 7]

print(paste("Posterior probability that the correlation is greater than 0:",
  round(post_prob_qi, 2)))
```

```
## [1] "Posterior probability that the correlation is greater than 0: 0.98"
```

This posterior probability can be evaluated based on the guidelines from (Chechile (2020)) and their extension to posterior probabilities below 0.5 by (Andrzejewicz et al. (2015)).

```
if (post_prob_qi < 0.0005){
  evaluation_qi = "Nearing certainty against"
}else if (post_prob_qi < 0.005){
  evaluation_qi = "Very strong bet against"
}else if (post_prob_qi < 0.01){
  evaluation_qi = "Strong bet against - irresponsible to avoid"
}else if (post_prob_qi < 0.1){
  evaluation_qi = "A promising but risky bet against"
}else if (post_prob_qi < 0.25){
  evaluation_qi = "Only a casual bet against"
```

```

}else if (post_prob_qi < 0.5){
  evaluation_qi = "Not worth betting against"
}else if (post_prob_qi < 0.75){
  evaluation_qi = "Not worth betting on"
}else if (post_prob_qi < 0.9){
  evaluation_qi = "Only a casual bet"
}else if (post_prob_qi < 0.95){
  evaluation_qi = "A promising but risky bet"
}else if (post_prob_qi < 0.99){
  evaluation_qi = "Good bet - too good to disregard"
}else if (post_prob_qi < 0.995){
  evaluation_qi = "Strong bet - irresponsible to avoid"
}else if (post_prob_qi < 0.9995){
  evaluation_qi = "Very strong bet"
}else if (post_prob_qi < 0.99995){
  evaluation_qi = "Nearing certainty"
}else{
  evaluation_qi = "Virtually certain"
}

evaluation_qi

```

```
## [1] "Good bet - too good to disregard"
```

Summary

Below we print a summary of the correlations.

```

tab <- rbind(c("Ease of doing activities", "Willingness to continue",
  paste(round(ease_mean, 2), "(", round(ease_SD, 2), ")"),
  paste("[", round(ease_hdi_low, 2), ",", round(ease_hdi_high,
    2), "]" ), round(post_prob_ease, 5), evaluation_ease))

tab <- rbind(tab, c("Motivation to do activities", "Willingness to continue",
  paste(round(motiv_mean, 2), "(", round(motiv_SD, 2), ")"),
  paste("[", round(motiv_hdi_low, 2), ",", round(motiv_hdi_high,
    2), "]" ), round(post_prob_motiv, 5), evaluation_motiv))

tab <- rbind(tab, c("Quitter self-identity", "Willingness to continue",
  paste(round(qi_mean, 2), "(", round(qi_SD, 2), ")"), paste("[",
    round(qi_hdi_low, 2), ",", round(qi_hdi_high, 2), "]" ),
  round(post_prob_qi, 2), evaluation_qi))

colnames(tab) = c("User variable", "Attitude question", "Mean (SD)",
  "95% CI", "Post > 0", "Evaluation")

pander(tab, caption = "Results of Bayesian analyses of Pearson correlations between user variables and a

```

Table 1: Results of Bayesian analyses of Pearson correlations between user variables and attitude ratings. (continued below)

User variable	Attitude question	Mean (SD)
Ease of doing activities	Willingness to continue	0.33 (0.04)

User variable	Attitude question	Mean (SD)
Motivation to do activities	Willingness to continue	0.48 (0.04)
Quitter self-identity	Willingness to continue	0.09 (0.05)

95% CI	Post > 0	Evaluation
[0.25 , 0.41]	0.99993	Nearing certainty
[0.4 , 0.54]	0.99993	Nearing certainty
[0 , 0.19]	0.98	Good bet - too good to disregard

References

- Andraszewicz, Sandra, Benjamin Scheibehenne, Jörg Rieskamp, Raoul Grasman, Josine Verhagen, and Eric-Jan Wagenmakers. 2015. "An Introduction to Bayesian Hypothesis Testing for Management Research." *Journal of Management* 41 (2): 521–43.
- Chechile, Richard A. 2020. *Bayesian Statistics for Experimental Scientists: A General Introduction Using Distribution-Free Methods*. MIT Press.