

Bayesian Analysis of Correlations with Stage of Change

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This file is meant to reproduce our results on correlations between the Transtheoretical Model (TTM)-stage for becoming physically active on the one hand and the change in self-efficacy between the pre- and post-measurement, motivational impact ratings for the generic and the personalized examples, and the acceptance on the other hand. These results are displayed in Table 7 in the Appendix.

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Required files

The following files are required: Data/df_se.csv, Data/df_ratings.csv, and Data/df_acceptance.csv.

Load packages

First, we load the packages that we need.

```
library(BayesianFirstAid) # For Bayesian correlations
library(formatR)         # To wrap lines
library(pander)          # For table
```

Load data

Now we load the pre-processed data.

```
# self-efficacy data
df_se <- read.csv(file = 'Data/df_se.csv')
# motivation ratings data
df_ratings <- read.csv(file = 'Data/df_ratings.csv')
# acceptance questions data
df_acceptance <- read.csv(file = 'Data/df_acceptance.csv')
```

And we reverse the scale for the TTM-stage such that a higher value is associated with a higher stage of change.

```
df_se$ttm_pa = abs(df_se$ttm_pa - 6)
df_ratings$ttm_pa = abs(df_ratings$ttm_pa - 6)
df_acceptance$ttm_pa = abs(df_acceptance$ttm_pa - 6)
```

Correlations

First we compute the correlation between the change in self-efficacy between the pre- and post-measurement and the TTM-stage for becoming physically active.

```
set.seed(18) # For reproducibility
fit = bayes.cor.test(df_se$diff, df_se$ttm_pa)

print(summary(fit))
```

```
## Data
## df_se$diff and df_se$ttm_pa, n = 39
##
## Model parameters
## rho: the correlation between df_se$diff and df_se$ttm_pa
## mu[1]: the mean of df_se$diff
## sigma[1]: the scale of df_se$diff , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df_se$ttm_pa
## sigma[2]: the scale of df_se$ttm_pa
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df_se$diff
## xy_pred[2]: the posterior predictive distribution of df_se$ttm_pa
##
## Measures
##
```

	mean	sd	HDILo	HDIup	%<comp	%>comp
## rho	-0.002	0.161	-0.308	0.318	0.506	0.494
## mu[1]	-12.416	4.240	-20.853	-4.099	0.998	0.002
## mu[2]	3.256	0.237	2.796	3.739	0.000	1.000
## sigma[1]	25.589	3.285	19.749	32.416	0.000	1.000
## sigma[2]	1.412	0.175	1.091	1.755	0.000	1.000
## nu	48.324	32.747	5.272	112.373	0.000	1.000
## xy_pred[1]	-12.695	26.866	-67.378	37.970	0.692	0.308
## xy_pred[2]	3.255	1.468	0.408	6.199	0.015	0.985

```
##
## '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
##
```

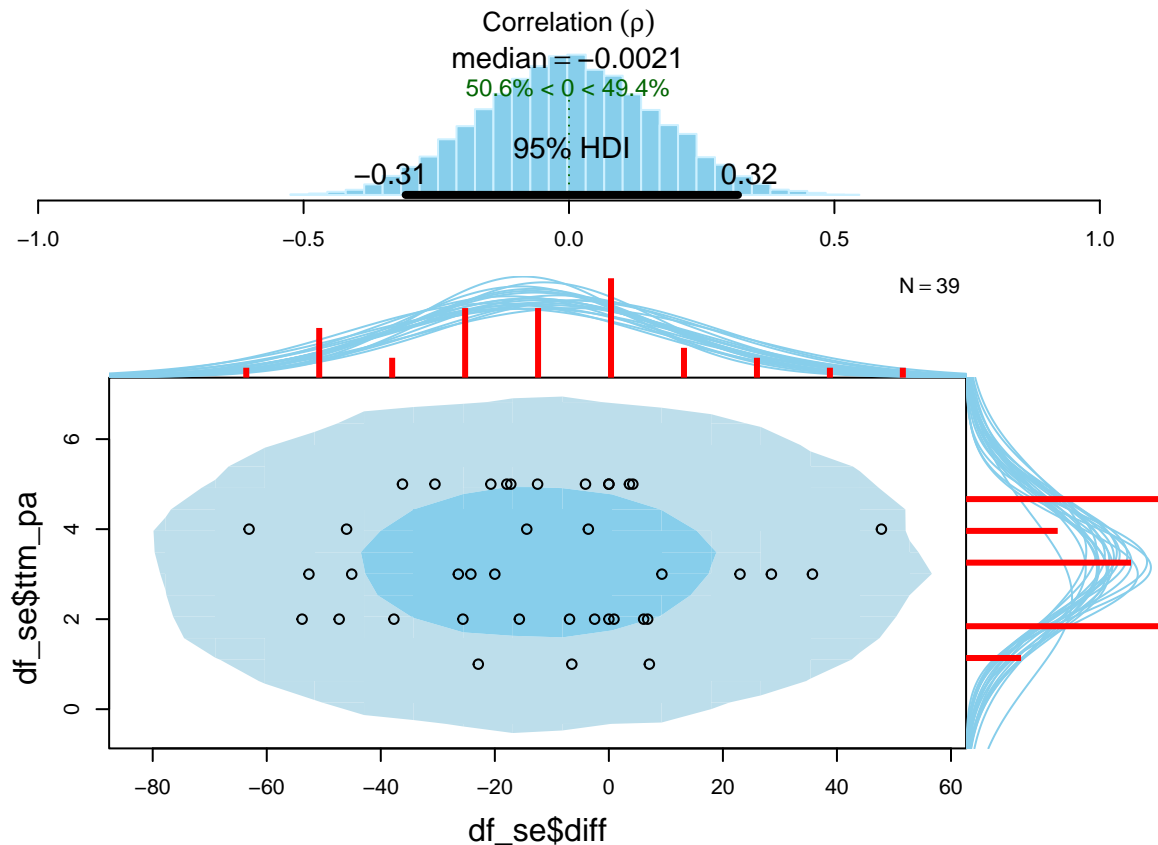
	q2.5%	q25%	median	q75%	q97.5%
## rho	-0.315	-0.112	-0.002	0.109	0.312
## mu[1]	-20.854	-15.126	-12.408	-9.669	-4.101
## mu[2]	2.780	3.100	3.256	3.414	3.727
## sigma[1]	20.051	23.310	25.253	27.526	32.981
## sigma[2]	1.117	1.289	1.396	1.516	1.801
## nu	10.298	24.779	39.855	63.228	133.140

```

## xy_pred[1] -65.371 -29.938 -12.831 4.727 40.073
## xy_pred[2] 0.348 2.287 3.252 4.213 6.147
##          mean          sd HDI% comp          HDIlo          HDIup
## rho          -0.001835462 0.1612468 95 0 -0.3077267 0.3183315
## mu[1]         -12.416033420 4.2401962 95 0 -20.8531963 -4.0989168
## mu[2]          3.255542150 0.2371679 95 0 2.7963492 3.7391802
## sigma[1]       25.589027399 3.2854456 95 0 19.7493544 32.4158332
## sigma[2]       1.412252075 0.1749537 95 0 1.0914515 1.7553266
## nu            48.324182843 32.7465029 95 0 5.2715194 112.3727505
## xy_pred[1]    -12.694537044 26.8655011 95 0 -67.3784003 37.9700436
## xy_pred[2]     3.254960391 1.4684840 95 0 0.4081745 6.1987592
##          %>comp      %<comp      q2.5%      q25%      median
## rho          0.494467404 5.055326e-01 -0.3152487 -0.1124296 -0.002107981
## mu[1]         0.002133049 9.978670e-01 -20.8536149 -15.1259552 -12.408261231
## mu[2]         0.999933342 6.665778e-05 2.7798535 3.1000518 3.256049172
## sigma[1]      0.999933342 6.665778e-05 20.0506560 23.3097309 25.253192086
## sigma[2]      0.999933342 6.665778e-05 1.1168288 1.2894648 1.396118656
## nu            0.999933342 6.665778e-05 10.2981097 24.7787421 39.854658990
## xy_pred[1]    0.308425543 6.915745e-01 -65.3710014 -29.9376203 -12.831018958
## xy_pred[2]    0.985268631 1.473137e-02 0.3479952 2.2873882 3.251785237
##          q75%      q97.5%      mcmc_se      Rhat n_eff
## rho          0.1087044 0.3124496 0.001668934 1.0003439 9406
## mu[1]        -9.6685499 -4.1011464 0.044965153 0.9999956 8949
## mu[2]         3.4136618 3.7269721 0.002551661 1.0022543 8645
## sigma[1]     27.5261616 32.9805501 0.038654618 1.0019553 7386
## sigma[2]     1.5158370 1.8006102 0.002047686 1.0000710 7304
## nu           63.2283333 133.1397685 0.535405569 0.9999234 3764
## xy_pred[1]    4.7271656 40.0725202 0.221532527 1.0002670 14712
## xy_pred[2]    4.2127889 6.1471187 0.012147135 1.0000706 14634

```

```
plot(fit)
```



```
# Mean, SD, CI
se_mean <- fit$stats[1, 1]
se_SD <- fit$stats[1, 2]
se_ci_low <- fit$stats[1, 5]
se_ci_high <- fit$stats[1, 6]

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

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

```
## [1] "Posterior probability that the correlation is greater than 0: 0.49"
```

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_se < 0.0005){
  evaluation_se = "Nearing certainty against"
}else if (post_prob_se < 0.005){
  evaluation_se = "Very strong bet against"
}else if (post_prob_se < 0.01){
  evaluation_se = "Strong bet against - irresponsible to avoid"
}else if (post_prob_se < 0.1){
  evaluation_se = "A promising but risky bet against"
}else if (post_prob_se < 0.25){
  evaluation_se = "Only a casual bet against"
}else if (post_prob_se < 0.5){
  evaluation_se = "Not worth betting against"
```

```

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

evaluation_se

```

```
## [1] "Not worth betting against"
```

Now we calculate the correlation between the motivational impact rating for the generic examples and the TTM-stage for becoming physically active.

```

set.seed(18) # For reproducibility
fit = bayes.cor.test(df_ratings$general, df_ratings$ttm_pa)

print(summary(fit))

```

```

## Data
## df_ratings$general and df_ratings$ttm_pa, n = 39
##
## Model parameters
## rho: the correlation between df_ratings$general and df_ratings$ttm_pa
## mu[1]: the mean of df_ratings$general
## sigma[1]: the scale of df_ratings$general , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df_ratings$ttm_pa
## sigma[2]: the scale of df_ratings$ttm_pa
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df_ratings$general
## xy_pred[2]: the posterior predictive distribution of df_ratings$ttm_pa
##
## Measures
##          mean      sd HDIlo  HDIup %<comp %>comp
## rho        0.043  0.165 -0.277  0.360  0.399  0.601
## mu[1]       1.170  0.250  0.668  1.649  0.000  1.000
## mu[2]       3.258  0.233  2.819  3.730  0.000  1.000
## sigma[1]    1.474  0.194  1.109  1.852  0.000  1.000
## sigma[2]    1.400  0.180  1.072  1.758  0.000  1.000
## nu         40.503 30.760  3.980 103.814  0.000  1.000
## xy_pred[1]  1.164  1.584 -2.011  4.290  0.223  0.777
## xy_pred[2]  3.247  1.505  0.256  6.213  0.018  0.982
##

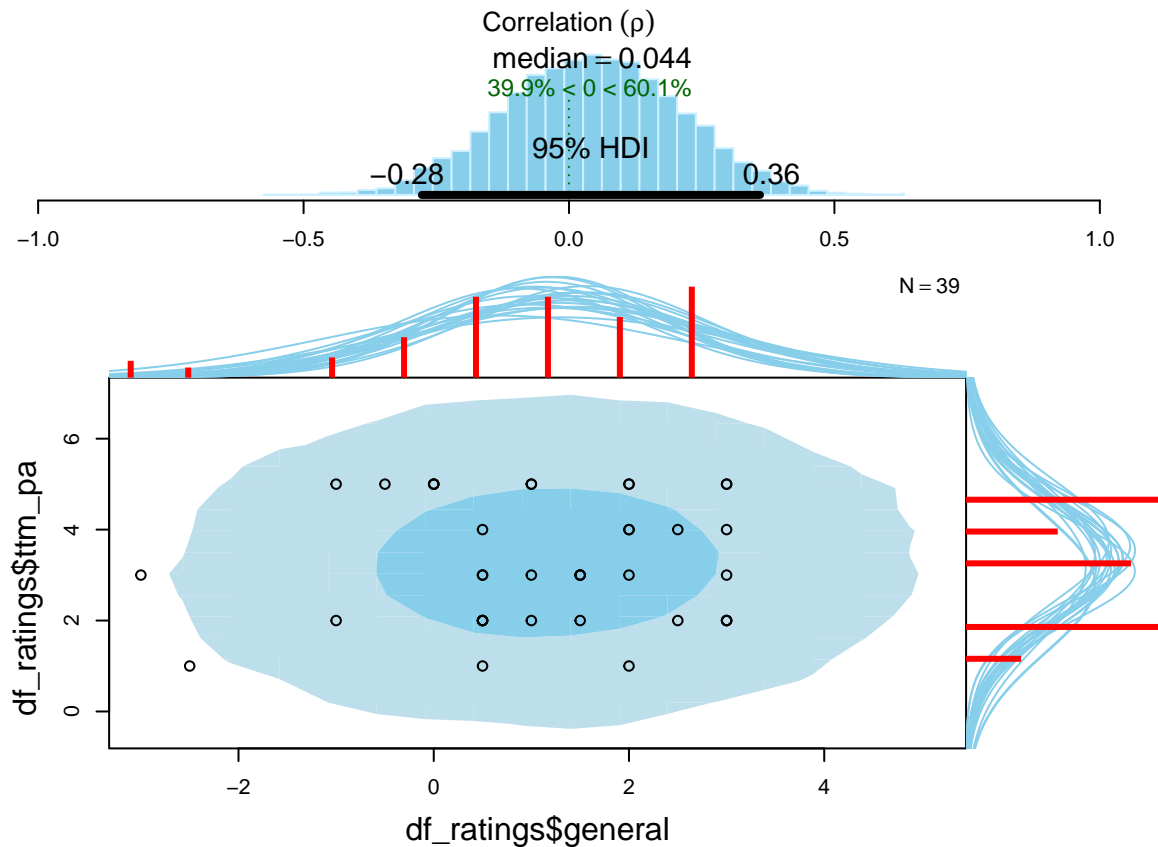
```

```

## '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          -0.276 -0.070  0.044  0.156  0.362
## mu[1]         0.669  1.005  1.171  1.337  1.653
## mu[2]         2.809  3.101  3.257  3.411  3.726
## sigma[1]      1.144  1.337  1.457  1.591  1.900
## sigma[2]      1.100  1.274  1.381  1.506  1.796
## nu            6.774 18.408 31.805 53.563 121.913
## xy_pred[1]    -1.969  0.154  1.171  2.164  4.333
## xy_pred[2]     0.268  2.293  3.253  4.201  6.230
##
##           mean          sd HDI% comp      HDIlo      HDIup      %>comp
## rho          0.04333615  0.1647501  95    0 -0.2772335  0.3598297  0.6010532
## mu[1]         1.16965837  0.2495627  95    0  0.6678476  1.6494806  0.9999333
## mu[2]         3.25807642  0.2330602  95    0  2.8185499  3.7302383  0.9999333
## sigma[1]      1.47386096  0.1935823  95    0  1.1089421  1.8522526  0.9999333
## sigma[2]      1.39961065  0.1797884  95    0  1.0715070  1.7583037  0.9999333
## nu           40.50257539 30.7604779  95    0  3.9800393 103.8137059  0.9999333
## xy_pred[1]     1.16356770  1.5842565  95    0 -2.0110351  4.2897457  0.7765631
## xy_pred[2]     3.24743215  1.5048192  95    0  0.2559503  6.2133701  0.9820024
##
##           %<comp      q2.5%      q25%      median      q75%
## rho          3.989468e-01 -0.2757545 -0.07040037  0.0437215  0.1562794
## mu[1]         6.665778e-05  0.6689865  1.00476100  1.1708738  1.3372788
## mu[2]         6.665778e-05  2.8093185  3.10065330  3.2570778  3.4110518
## sigma[1]      6.665778e-05  1.1436720  1.33673895  1.4571341  1.5913656
## sigma[2]      6.665778e-05  1.0997664  1.27396644  1.3810024  1.5055549
## nu           6.665778e-05  6.7744150 18.40811272 31.8052140 53.5631829
## xy_pred[1]     2.234369e-01 -1.9687267  0.15371196  1.1713598  2.1637887
## xy_pred[2]     1.799760e-02  0.2677949  2.29274316  3.2526863  4.2011555
##
##           q97.5%      mcmc_se      Rhat n_eff
## rho          0.3615855  0.001774651  1.000128  8656
## mu[1]         1.6527325  0.002759052  1.000924  8182
## mu[2]         3.7256385  0.002598019  1.001541  8051
## sigma[1]      1.9000160  0.002404633  1.000539  6526
## sigma[2]      1.7963724  0.002194896  1.001712  6732
## nu           121.9125109  0.517838478  1.001270  3528
## xy_pred[1]     4.3333047  0.013209776  1.000348 14413
## xy_pred[2]     6.2304142  0.012448669  1.000207 14648

```

```
plot(fit)
```



```
# Mean, SD, CI
rat_gen_mean <- fit$stats[1, 1]
rat_gen_SD <- fit$stats[1, 2]
rat_gen_ci_low <- fit$stats[1, 5]
rat_gen_ci_high <- fit$stats[1, 6]

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

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

## [1] "Posterior probability that the correlation is greater than 0: 0.6"
```

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_rat_gen < 0.0005){
  evaluation_rat_gen = "Nearing certainty against"
}else if (post_prob_rat_gen < 0.005){
  evaluation_rat_gen = "Very strong bet against"
}else if (post_prob_rat_gen < 0.01){
  evaluation_rat_gen = "Strong bet against - irresponsible to avoid"
}else if (post_prob_rat_gen < 0.1){
  evaluation_rat_gen = "A promising but risky bet against"
}else if (post_prob_rat_gen < 0.25){
  evaluation_rat_gen = "Only a casual bet against"
}else if (post_prob_rat_gen < 0.5){
  evaluation_rat_gen = "Not worth betting against"
}
```

```

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

evaluation_rat_gen

```

```
## [1] "Not worth betting on"
```

Next we calculate the correlation between the motivational impact rating for the personalized examples and the TTM-stage for becoming physically active.

```

set.seed(18) # For reproducibility
fit = bayes.cor.test(df_ratings$personalized, df_ratings$ttm_pa)

print(summary(fit))

```

```

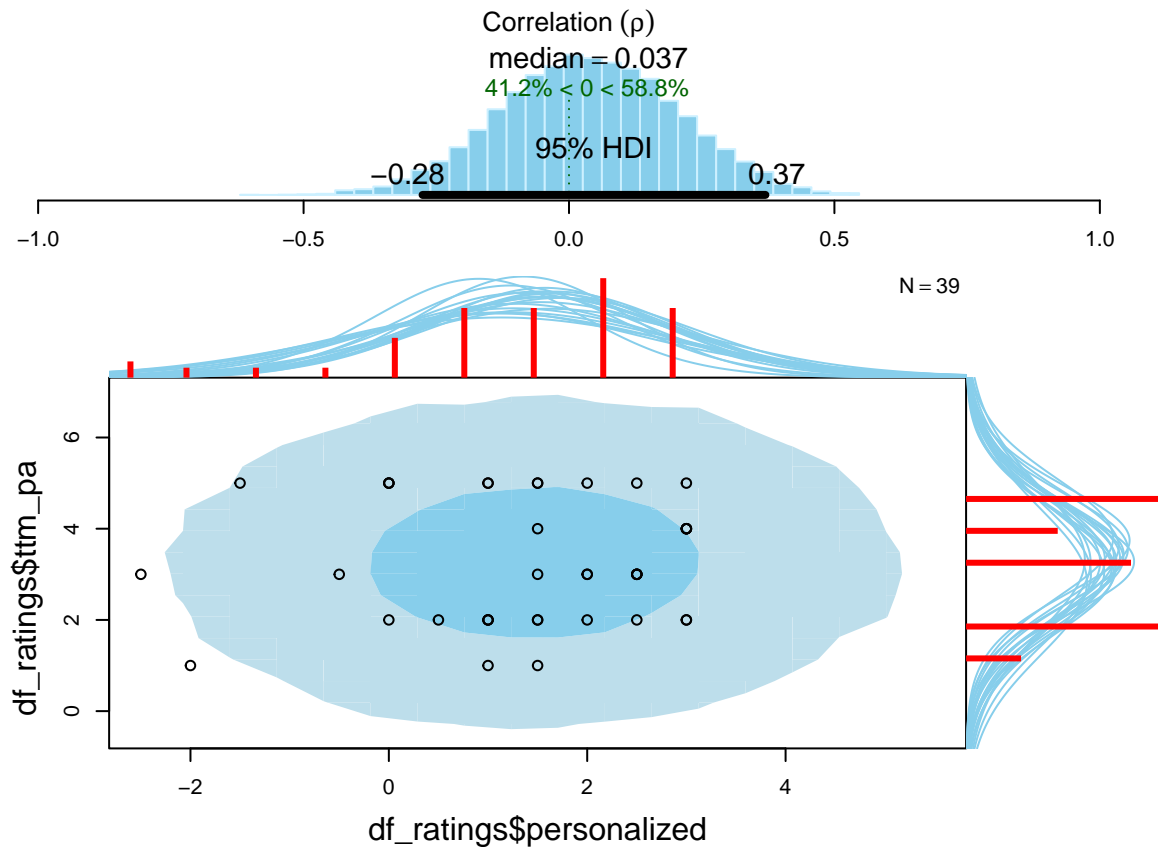
## Data
## df_ratings$personalized and df_ratings$ttm_pa, n = 39
##
## Model parameters
## rho: the correlation between df_ratings$personalized and df_ratings$ttm_pa
## mu[1]: the mean of df_ratings$personalized
## sigma[1]: the scale of df_ratings$personalized , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df_ratings$ttm_pa
## sigma[2]: the scale of df_ratings$ttm_pa
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df_ratings$personalized
## xy_pred[2]: the posterior predictive distribution of df_ratings$ttm_pa
##
## Measures
##          mean      sd HDIlo HDIup %<comp %>comp
## rho          0.038  0.166 -0.276  0.370  0.412  0.588
## mu[1]         1.461  0.238  1.008  1.953  0.000  1.000
## mu[2]         3.254  0.234  2.809  3.746  0.000  1.000
## sigma[1]       1.401  0.185  1.055  1.774  0.000  1.000
## sigma[2]       1.399  0.179  1.060  1.757  0.000  1.000
## nu           38.757 29.206  3.230 96.733  0.000  1.000
## xy_pred[1]     1.481  1.504 -1.570  4.387  0.156  0.844
## xy_pred[2]     3.251  1.496  0.239  6.114  0.017  0.983
##

```

```

## '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          -0.288 -0.077  0.037  0.152   0.359
## mu[1]         0.988  1.305  1.461  1.618   1.936
## mu[2]         2.788  3.098  3.255  3.407   3.726
## sigma[1]      1.077  1.273  1.387  1.512   1.807
## sigma[2]      1.093  1.274  1.381  1.508   1.800
## nu            6.677 18.081 30.777 51.103 116.473
## xy_pred[1]    -1.517  0.524  1.470  2.444   4.465
## xy_pred[2]     0.281  2.303  3.243  4.216   6.174
##           mean          sd HDI% comp      HDIlo      HDIup      %>comp
## rho          0.0375357  0.1658904  95    0 -0.2758743  0.3700189  0.5878550
## mu[1]         1.4613122  0.2382760  95    0  1.0081916  1.9526140  0.9999333
## mu[2]         3.2542234  0.2342290  95    0  2.8087074  3.7457039  0.9999333
## sigma[1]      1.4007176  0.1854746  95    0  1.0552129  1.7743443  0.9999333
## sigma[2]      1.3987058  0.1792815  95    0  1.0601064  1.7565116  0.9999333
## nu            38.7572451 29.2063312  95    0  3.2298356 96.7332709  0.9999333
## xy_pred[1]     1.4808496  1.5041713  95    0 -1.5698369  4.3870784  0.8438208
## xy_pred[2]     3.2511127  1.4962439  95    0  0.2391788  6.1135252  0.9828690
##           %<comp      q2.5%      q25%      median      q75%      q97.5%
## rho          4.121450e-01 -0.2882551 -0.076516  0.0365048  0.1523033  0.3585485
## mu[1]         6.665778e-05  0.9879020  1.305063  1.4607357  1.6175040  1.9355964
## mu[2]         6.665778e-05  2.7879091  3.098214  3.2545570  3.4074934  3.7260810
## sigma[1]      6.665778e-05  1.0767309  1.272922  1.3870796  1.5120997  1.8072638
## sigma[2]      6.665778e-05  1.0931504  1.273646  1.3810272  1.5078599  1.7996117
## nu            6.665778e-05  6.6769314 18.081185 30.7766655 51.1032186 116.4726793
## xy_pred[1]     1.561792e-01 -1.5174963  0.523788  1.4702864  2.4439678  4.4645395
## xy_pred[2]     1.713105e-02  0.2814674  2.303293  3.2431481  4.2160270  6.1742120
##           mcmc_se      Rhat n_eff
## rho          0.001727009 1.0005793  9312
## mu[1]         0.002711806 1.0007556  7731
## mu[2]         0.002551339 1.0000716  8432
## sigma[1]      0.002312666 1.0003558  6448
## sigma[2]      0.002128627 1.0000397  7093
## nu            0.550290549 1.0016025  2997
## xy_pred[1]     0.012395052 1.0001222 14732
## xy_pred[2]     0.012217011 0.9999368 15000
plot(fit)

```



```
# Mean, SD, CI
rat_pers_mean <- fit$stats[1, 1]
rat_pers_SD <- fit$stats[1, 2]
rat_pers_ci_low <- fit$stats[1, 5]
rat_pers_ci_high <- fit$stats[1, 6]

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

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

## [1] "Posterior probability that the correlation is greater than 0: 0.59"
```

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_rat_pers < 0.0005){
  evaluation_rat_pers = "Nearing certainty against"
}else if (post_prob_rat_pers < 0.005){
  evaluation_rat_pers = "Very strong bet against"
}else if (post_prob_rat_pers < 0.01){
  evaluation_rat_pers = "Strong bet against - irresponsible to avoid"
}else if (post_prob_rat_pers < 0.1){
  evaluation_rat_pers = "A promising but risky bet against"
}else if (post_prob_rat_pers < 0.25){
  evaluation_rat_pers = "Only a casual bet against"
}else if (post_prob_rat_pers < 0.5){
  evaluation_rat_pers = "Not worth betting against"
```

```

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

evaluation_rat_pers

```

```
## [1] "Not worth betting on"
```

Finally, we calculate the correlation between the mean acceptance rating and the TTM-stage for becoming physically active.

```

# Compute mean acceptance rating
df_acceptance$mean = (df_acceptance$acceptance_1 + df_acceptance$acceptance_2 + df_acceptance$acceptance_3) / 3

set.seed(18) # For reproducibility
fit = bayes.cor.test(df_acceptance$mean, df_acceptance$ttm_pa)

print(summary(fit))

```

```

## Data
## df_acceptance$mean and df_acceptance$ttm_pa, n = 39
##
## Model parameters
## rho: the correlation between df_acceptance$mean and df_acceptance$ttm_pa
## mu[1]: the mean of df_acceptance$mean
## sigma[1]: the scale of df_acceptance$mean , a consistent
## estimate of SD when nu is large.
## mu[2]: the mean of df_acceptance$ttm_pa
## sigma[2]: the scale of df_acceptance$ttm_pa
## nu: the degrees-of-freedom for the bivariate t distribution
## xy_pred[1]: the posterior predictive distribution of df_acceptance$mean
## xy_pred[2]: the posterior predictive distribution of df_acceptance$ttm_pa
##
## Measures
##
```

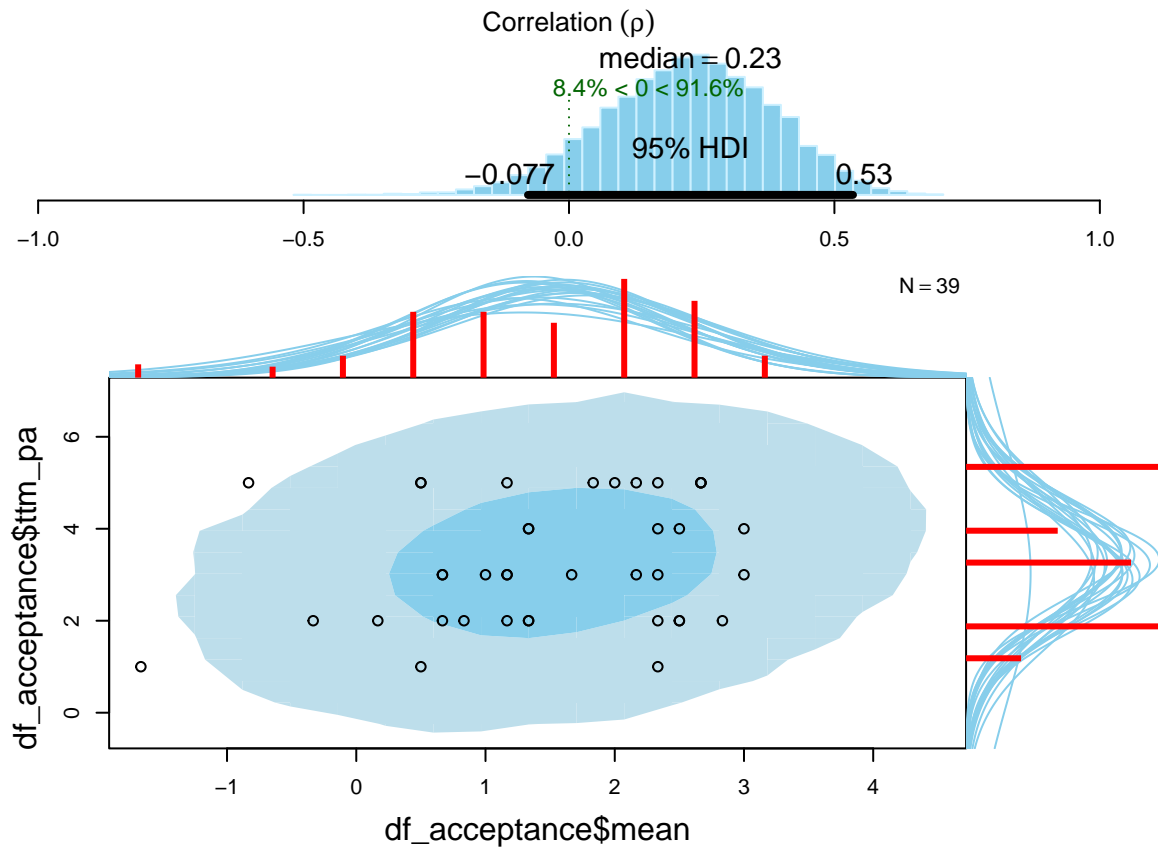
	mean	sd	HDIlo	HDIup	%<comp	%>comp
## rho	0.223	0.158	-0.077	0.535	0.084	0.916
## mu[1]	1.529	0.183	1.181	1.898	0.000	1.000
## mu[2]	3.264	0.232	2.801	3.709	0.000	1.000
## sigma[1]	1.088	0.142	0.826	1.378	0.000	1.000
## sigma[2]	1.387	0.178	1.075	1.743	0.000	1.000
## nu	38.899	29.566	3.433	97.383	0.000	1.000

```

## xy_pred[1] 1.538 1.159 -0.741 3.881 0.086 0.914
## xy_pred[2] 3.256 1.496 0.395 6.258 0.017 0.983
##
## '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      -0.100 0.118 0.229 0.334 0.516
## mu[1]     1.164 1.410 1.529 1.650 1.887
## mu[2]     2.809 3.110 3.266 3.420 3.718
## sigma[1]  0.844 0.990 1.077 1.174 1.405
## sigma[2]  1.089 1.264 1.370 1.495 1.766
## nu        6.626 17.699 30.554 51.243 115.901
## xy_pred[1] -0.758 0.795 1.538 2.279 3.873
## xy_pred[2] 0.333 2.303 3.244 4.215 6.204
##      mean      sd HDI% comp      HDIlo      HDIup      %>comp
## rho      0.2232205 0.1580734 95 0 -0.07722339 0.5349316 0.9156779
## mu[1]     1.5290689 0.1826225 95 0 1.18059062 1.8976009 0.9999333
## mu[2]     3.2643080 0.2318841 95 0 2.80143179 3.7090112 0.9999333
## sigma[1]  1.0882272 0.1422360 95 0 0.82559764 1.3780535 0.9999333
## sigma[2]  1.3871585 0.1778934 95 0 1.07455134 1.7428477 0.9999333
## nu       38.8994140 29.5664223 95 0 3.43304156 97.3832492 0.9999333
## xy_pred[1] 1.5381323 1.1588837 95 0 -0.74097966 3.8806061 0.9136782
## xy_pred[2] 3.2560230 1.4961742 95 0 0.39482592 6.2583476 0.9834022
##      %<comp      q2.5%      q25%      median      q75%      q97.5%
## rho      8.432209e-02 -0.1000032 0.1184219 0.2287848 0.3343109 0.5156122
## mu[1]     6.665778e-05 1.1642693 1.4097444 1.5289843 1.6500413 1.8868859
## mu[2]     6.665778e-05 2.8088995 3.1103198 3.2657175 3.4198292 3.7177158
## sigma[1]  6.665778e-05 0.8438772 0.9900189 1.0765948 1.1736230 1.4047606
## sigma[2]  6.665778e-05 1.0890326 1.2637805 1.3695946 1.4946546 1.7660591
## nu       6.665778e-05 6.6257110 17.6988018 30.5536666 51.2428280 115.9009873
## xy_pred[1] 8.632182e-02 -0.7575072 0.7945306 1.5380671 2.2785431 3.8731838
## xy_pred[2] 1.659779e-02 0.3328805 2.3028570 3.2441583 4.2152614 6.2036044
##      mcmc_se      Rhat n_eff
## rho      0.001689726 1.0006546 8782
## mu[1]     0.002059826 1.0002671 7873
## mu[2]     0.002510608 1.0000598 8566
## sigma[1]  0.001766301 1.0000143 6499
## sigma[2]  0.002422104 1.0003611 6107
## nu       0.521889591 1.0049685 3263
## xy_pred[1] 0.009603114 1.0006149 14650
## xy_pred[2] 0.012409244 0.9999461 14543

```

```
plot(fit)
```



```
# Mean, SD, CI
acc_mean <- fit$stats[1, 1]
acc_SD <- fit$stats[1, 2]
acc_ci_low <- fit$stats[1, 5]
acc_ci_high <- fit$stats[1, 6]

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

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

## [1] "Posterior probability that the correlation is greater than 0: 0.92"
```

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_acc < 0.0005){
  evaluation_acc = "Nearing certainty against"
}else if (post_prob_acc < 0.005){
  evaluation_acc = "Very strong bet against"
}else if (post_prob_acc < 0.01){
  evaluation_acc = "Strong bet against - irresponsible to avoid"
}else if (post_prob_acc < 0.1){
  evaluation_acc = "A promising but risky bet against"
}else if (post_prob_acc < 0.25){
  evaluation_acc = "Only a casual bet against"
}else if (post_prob_acc < 0.5){
  evaluation_acc = "Not worth betting against"
```

```

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

evaluation_acc

```

```
## [1] "A promising but risky bet"
```

Summary

Below we print a summary, which reproduces Table 7 from the Appendix of the paper.

```

tab <- rbind(c("Change in self-efficacy", paste(round(se_mean,
2), "(", round(se_SD, 2), ")"), paste("[", round(se_ci_low,
2), ",", round(se_ci_high, 2), "]"), round(post_prob_se,
2), evaluation_se))

tab <- rbind(tab, c("Motivational impact generic examples", paste(round(rat_gen_mean,
2), "(", round(rat_gen_SD, 2), ")"), paste("[", round(rat_gen_ci_low,
2), ",", round(rat_gen_ci_high, 2), "]"), round(post_prob_rat_gen,
2), evaluation_rat_gen))

tab <- rbind(tab, c("Motivational impact personalized examples",
paste(round(rat_pers_mean, 2), "(", round(rat_pers_SD, 2),
"),"), paste("[", round(rat_pers_ci_low, 2), ",", round(rat_pers_ci_high,
2), "]"), round(post_prob_rat_pers, 2), evaluation_rat_pers))

tab <- rbind(tab, c("Acceptance", paste(round(acc_mean, 2), "(",
round(acc_SD, 2), ")"), paste("[", round(acc_ci_low, 2),
",", round(acc_ci_high, 2), "]"), round(post_prob_acc, 2),
evaluation_acc))

colnames(tab) = c("Outcome Measure", "Mean (SD)", "95% CI", "Post > 0",
"Evaluation")

pander(tab, caption = "Results of Bayesian analyses of Pearson correlations between the TTM-stage for b

```

Table 1: Results of Bayesian analyses of Pearson correlations between the TTM-stage for becoming physically active on the one hand and the outcome measures of change in self-efficacy between the pre- and post-measurement, motivational impact ratings for the two types of examples, and the acceptance on the other hand. (continued below)

Outcome Measure	Mean (SD)	95% CI	Post > 0
Change in self-efficacy	0 (0.16)	[-0.31 , 0.32]	0.49
Motivational impact generic examples	0.04 (0.16)	[-0.28 , 0.36]	0.6
Motivational impact personalized examples	0.04 (0.17)	[-0.28 , 0.37]	0.59
Acceptance	0.22 (0.16)	[-0.08 , 0.53]	0.92

Evaluation
Not worth betting against
Not worth betting on
Not worth betting on
A promising but risky bet

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.

Checkile, Richard A. 2020. *Bayesian Statistics for Experimental Scientists: A General Introduction Using Distribution-Free Methods*. MIT Press.