

Effects of individual actions on confidence

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09 November, 2023

This file contains the code for the analysis of the effects of individual actions on confidence. Note that, due to the nature of the Bayesian t -tests, results will not always be the same for the means and standard deviations. From the output of the `print` of the result, we use the probability of the mean difference being larger and smaller than 0. From the output of the `summary` of the result, we use the mean and standard deviation values, and their 95% CIs.

```
## Loading required package: rjags
```

```
## Loading required package: coda
```

```
## Linked to JAGS 4.3.0
```

```
## Loaded modules: basemod,bugs
```

Run a Bayesian t -test on the raw data before and after the first change to the plan made, and calculate Cohen's d for these values.

```
# get the values for confidence before making the first change to the plan as a list
c_changes_1_b = as.numeric(t(read.csv(
  "~/analysis/data/c_changes_to_plan_1_before.csv", header = FALSE)))
```

```
# get the values for confidence after making the first change to the plan as a list
c_changes_1_a = as.numeric(t(read.csv(
  "~/analysis/data/c_changes_to_plan_1_after.csv", header = FALSE)))
```

```
# run the Bayesian t-test
```

```
c_changes_to_plan_1 = bayes.t.test(c_changes_1_b, c_changes_1_a, paired = TRUE)
```

```
# print the results of the test, and the summary
```

```
print(c_changes_to_plan_1)
```

```
##
```

```
## Bayesian estimation supersedes the t test (BEST) - paired samples
```

```
##
```

```
## data: c_changes_1_b and c_changes_1_a, n = 114
```

```
##
```

```
## Estimates [95% credible interval]
```

```
## mean paired difference: -4e-06 [-0.00071, 0.00069]
```

```
## sd of the paired differences: 0.0020 [0.0014, 0.0077]
```

```
##
```

```
## The mean difference is more than 0 by a probability of 0.492
```

```
## and less than 0 by a probability of 0.508
```

```
summary(c_changes_to_plan_1)
```

```
## Data
```

```

## c_changes_1_b, n = 114
## c_changes_1_a, n = 114
##
## Model parameters and generated quantities
## mu_diff: the mean pairwise difference between c_changes_1_b and c_changes_1_a
## sigma_diff: the scale of the pairwise difference, a consistent
## estimate of SD when nu is large.
## nu: the degrees-of-freedom for the t distribution fitted to the pairwise difference
## eff_size: the effect size calculated as (mu_diff - 0) / sigma_diff
## diff_pred: predicted distribution for a new datapoint generated
## as the pairwise difference between c_changes_1_b and c_changes_1_a
##
## Measures
##      mean      sd  HDIlo HDIup %<comp %>comp
## mu_diff   -0.002 0.029 -0.001 0.001  0.508  0.492
## sigma_diff 0.007 0.049  0.001 0.008  0.000  1.000
## nu         1.012 0.131  1.000 1.011  0.000  1.000
## eff_size   -0.005 0.103 -0.191 0.186  0.508  0.492
## diff_pred  0.006 0.868 -0.042 0.044  0.501  0.499
##
## '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%
## mu_diff   -0.001  0.000  0.000  0.000  0.001
## sigma_diff 0.001  0.002  0.002  0.003  0.012
## nu         1.000  1.001  1.002  1.005  1.015
## eff_size   -0.195 -0.067 -0.002  0.061  0.182
## diff_pred  -0.045 -0.002  0.000  0.002  0.041

```

```

# calculate Cohen's d
cohen.d(c_changes_1_a, c_changes_1_b, paired=TRUE)

```

```

##
## Cohen's d
##
## d estimate: 0.3242204 (small)
## 95 percent confidence interval:
##      lower      upper
## 0.2111578 0.4372830

```

Run a Bayesian *t*-test on the raw data before and after the second change to the plan made, and calculate Cohen's *d* for these values.

```

c_changes_2_b = as.numeric(t(read.csv(
  "~/analysis/data/c_changes_to_plan_2_before.csv", header = FALSE)))
c_changes_2_a = as.numeric(t(read.csv(
  "~/analysis/data/c_changes_to_plan_2_after.csv", header = FALSE)))
c_changes_to_plan_2 = bayes.t.test(c_changes_2_b, c_changes_2_a, paired = TRUE)

print(c_changes_to_plan_2)

```

```

##
## Bayesian estimation supersedes the t test (BEST) - paired samples

```

```
##
## data: c_changes_2_b and c_changes_2_a, n = 59
##
## Estimates [95% credible interval]
## mean paired difference: -5.6e-07 [-0.00014, 0.00014]
## sd of the paired differences: 0.00064 [0.00062, 7e-04]
##
## The mean difference is more than 0 by a probability of 0.497
## and less than 0 by a probability of 0.503
summary(c_changes_to_plan_2)

## Data
## c_changes_2_b, n = 59
## c_changes_2_a, n = 59
##
## Model parameters and generated quantities
## mu_diff: the mean pairwise difference between c_changes_2_b and c_changes_2_a
## sigma_diff: the scale of the pairwise difference, a consistent
## estimate of SD when nu is large.
## nu: the degrees-of-freedom for the t distribution fitted to the pairwise difference
## eff_size: the effect size calculated as (mu_diff - 0) / sigma_diff
## diff_pred: predicted distribution for a new datapoint generated
## as the pairwise difference between c_changes_2_b and c_changes_2_a
##
## Measures
##      mean      sd  HDIlo HDIup %<comp %>comp
## mu_diff    0.000 0.000  0.000 0.000  0.503  0.497
## sigma_diff  0.001 0.000  0.001 0.001  0.000  1.000
## nu         1.011 0.010  1.000 1.031  0.000  1.000
## eff_size   -0.001 0.111 -0.220 0.214  0.503  0.497
## diff_pred   0.001 0.196 -0.008 0.008  0.498  0.502
##
## '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%
## mu_diff    0.000  0.000  0.000 0.000  0.000
## sigma_diff  0.001  0.001  0.001 0.001  0.001
## nu         1.000  1.003  1.007 1.015  1.038
## eff_size   -0.220 -0.076 -0.001 0.074  0.216
## diff_pred  -0.008 -0.001  0.000 0.001  0.009
cohen.d(c_changes_2_a, c_changes_2_b, paired=TRUE)

##
## Cohen's d
##
## d estimate: 0.1084785 (negligible)
## 95 percent confidence interval:
##      lower      upper
## 0.02982207 0.18713500
```

Run a Bayesian *t*-test on the raw data before and after explaining the importance of planning, and calculate

Cohen's d for these values.

```
c_planning_b = as.numeric(t(read.csv(
  "~/analysis/data/c_explain_planning_before.csv", header = FALSE)))
c_planning_a = as.numeric(t(read.csv(
  "~/analysis/data/c_explain_planning_after.csv", header = FALSE)))
c_explain_planning = bayes.t.test(c_planning_b, c_planning_a, paired = TRUE)

print(c_explain_planning)

##
## Bayesian estimation supersedes the t test (BEST) - paired samples
##
## data: c_planning_b and c_planning_a, n = 86
##
## Estimates [95% credible interval]
## mean paired difference: 1.1e-07 [-0.00016, 0.00016]
## sd of the paired differences: 9e-04 [0.00089, 0.00097]
##
## The mean difference is more than 0 by a probability of 0.501
## and less than 0 by a probability of 0.499
summary(c_explain_planning)

## Data
## c_planning_b, n = 86
## c_planning_a, n = 86
##
## Model parameters and generated quantities
## mu_diff: the mean pairwise difference between c_planning_b and c_planning_a
## sigma_diff: the scale of the pairwise difference, a consistent
## estimate of SD when nu is large.
## nu: the degrees-of-freedom for the t distribution fitted to the pairwise difference
## eff_size: the effect size calculated as (mu_diff - 0) / sigma_diff
## diff_pred: predicted distribution for a new datapoint generated
## as the pairwise difference between c_planning_b and c_planning_a
##
## Measures
##      mean      sd  HDIlo HDIup %<comp %>comp
## mu_diff  0.000 0.000  0.000 0.000  0.499  0.501
## sigma_diff 0.001 0.000  0.001 0.001  0.000  1.000
## nu        1.007 0.007  1.000 1.022  0.000  1.000
## eff_size   0.000 0.091 -0.180 0.173  0.499  0.501
## diff_pred  0.001 0.187 -0.011 0.011  0.496  0.504
##
## '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%
## mu_diff  0.000 0.000  0.000 0.000  0.000
## sigma_diff 0.001 0.001  0.001 0.001  0.001
## nu        1.000 1.002  1.005 1.010  1.027
## eff_size  -0.177 -0.061  0.000 0.061  0.176
```

```
## diff_pred -0.011 -0.001 0.000 0.001 0.011
cohen.d(c_planning_a, c_planning_b, paired=TRUE)
```

```
##
## Cohen's d
##
## d estimate: 0.1481568 (negligible)
## 95 percent confidence interval:
##      lower      upper
## 0.06176261 0.23455108
```

Run a Bayesian t -test on the raw data before and after identifying barriers, and calculate Cohen's d for these values.

```
c_i_barriers_b = as.numeric(t(read.csv(
  "~/analysis/data/c_identify_barriers_before.csv", header = FALSE)))
c_i_barriers_a = as.numeric(t(read.csv(
  "~/analysis/data/c_identify_barriers_after.csv", header = FALSE)))
c_identify_barriers = bayes.t.test(c_i_barriers_b, c_i_barriers_a, paired = TRUE)

print(c_identify_barriers)
```

```
##
## Bayesian estimation supersedes the t test (BEST) - paired samples
##
## data: c_i_barriers_b and c_i_barriers_a, n = 85
##
## Estimates [95% credible interval]
## mean paired difference: 1.3e-06 [-0.00019, 0.00018]
## sd of the paired differences: 0.00095 [0.00091, 0.0011]
##
## The mean difference is more than 0 by a probability of 0.506
## and less than 0 by a probability of 0.494
```

```
summary(c_identify_barriers)
```

```
## Data
## c_i_barriers_b, n = 85
## c_i_barriers_a, n = 85
##
## Model parameters and generated quantities
## mu_diff: the mean pairwise difference between c_i_barriers_b and c_i_barriers_a
## sigma_diff: the scale of the pairwise difference, a consistent
## estimate of SD when nu is large.
## nu: the degrees-of-freedom for the t distribution fitted to the pairwise difference
## eff_size: the effect size calculated as (mu_diff - 0) / sigma_diff
## diff_pred: predicted distribution for a new datapoint generated
## as the pairwise difference between c_i_barriers_b and c_i_barriers_a
##
## Measures
##      mean      sd HDIlo HDIup %<comp %>comp
## mu_diff 0.000 0.000 0.000 0.000 0.494 0.506
## sigma_diff 0.001 0.000 0.001 0.001 0.000 1.000
## nu      1.006 0.006 1.000 1.017 0.000 1.000
## eff_size 0.001 0.098 -0.191 0.190 0.494 0.506
```

```
## diff_pred 0.002 0.217 -0.012 0.012 0.498 0.502
##
## '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%
## mu_diff    0.000 0.000 0.000 0.000 0.000
## sigma_diff 0.001 0.001 0.001 0.001 0.001
## nu         1.000 1.002 1.004 1.008 1.021
## eff_size   -0.190 -0.064 0.001 0.067 0.192
## diff_pred  -0.012 -0.001 0.000 0.001 0.012
cohen.d(c_i_barriers_a, c_i_barriers_b, paired=TRUE)
```

```
##
## Cohen's d
##
## d estimate: -0.06919324 (negligible)
## 95 percent confidence interval:
##      lower      upper
## -0.16532241 0.02693592
```

Run a Bayesian *t*-test on the raw data before and after dealing with barriers, and calculate Cohen's *d* for these values.

```
c_d_barriers_b = as.numeric(t(read.csv(
  "~/analysis/data/c_deal_with_barriers_before.csv", header = FALSE)))
c_d_barriers_a = as.numeric(t(read.csv(
  "~/analysis/data/c_deal_with_barriers_after.csv", header = FALSE)))
c_deal_with_barriers = bayes.t.test(c_d_barriers_b, c_d_barriers_a, paired = TRUE)
print(c_deal_with_barriers)
```

```
##
## Bayesian estimation supersedes the t test (BEST) - paired samples
##
## data: c_d_barriers_b and c_d_barriers_a, n = 85
##
## Estimates [95% credible interval]
## mean paired difference: 2e-06 [-0.00024, 0.00024]
## sd of the paired differences: 0.0013 [0.0012, 0.0014]
##
## The mean difference is more than 0 by a probability of 0.507
## and less than 0 by a probability of 0.493
summary(c_deal_with_barriers)
```

```
## Data
## c_d_barriers_b, n = 85
## c_d_barriers_a, n = 85
##
## Model parameters and generated quantities
## mu_diff: the mean pairwise difference between c_d_barriers_b and c_d_barriers_a
## sigma_diff: the scale of the pairwise difference, a consistent
## estimate of SD when nu is large.
```

```

## nu: the degrees-of-freedom for the t distribution fitted to the pairwise difference
## eff_size: the effect size calculated as (mu_diff - 0) / sigma_diff
## diff_pred: predicted distribution for a new datapoint generated
## as the pairwise difference between c_d_barriers_b and c_d_barriers_a
##
## Measures
##      mean      sd  HDIlo HDIup %<comp %>comp
## mu_diff    0.000 0.000  0.000 0.000  0.493  0.507
## sigma_diff 0.001 0.000  0.001 0.001  0.000  1.000
## nu         1.006 0.006  1.000 1.019  0.000  1.000
## eff_size    0.001 0.095 -0.187 0.187  0.493  0.507
## diff_pred -0.001 0.287 -0.017 0.015  0.496  0.504
##
## '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%
## mu_diff    0.000 0.000  0.000 0.000  0.000
## sigma_diff 0.001 0.001  0.001 0.001  0.001
## nu         1.000 1.002  1.004 1.009  1.023
## eff_size   -0.187 -0.063  0.002 0.065  0.187
## diff_pred -0.016 -0.001  0.000 0.001  0.016

```

```
cohen.d(c_d_barriers_a, c_d_barriers_b, paired=TRUE)
```

```

##
## Cohen's d
##
## d estimate: 0.2011579 (small)
## 95 percent confidence interval:
##      lower      upper
## 0.0814886 0.3208271

```

Run a Bayesian *t*-test on the raw data before and after showing testimonials from other people who created plans, and calculate Cohen's *d* for these values.

```

c_testimonials_b = as.numeric(t(read.csv(
  "~/analysis/data/c_show_testimonials_before.csv", header = FALSE)))
c_testimonials_a = as.numeric(t(read.csv(
  "~/analysis/data/c_show_testimonials_after.csv", header = FALSE)))
c_show_testimonials = bayes.t.test(c_testimonials_b, c_testimonials_a, paired = TRUE)

print(c_show_testimonials)

```

```

##
## Bayesian estimation supersedes the t test (BEST) - paired samples
##
## data: c_testimonials_b and c_testimonials_a, n = 90
##
## Estimates [95% credible interval]
## mean paired difference: -2.8e-06 [-0.00026, 0.00026]
## sd of the paired differences: 0.0012 [0.0010, 0.0018]
##
## The mean difference is more than 0 by a probability of 0.491

```

```
## and less than 0 by a probability of 0.509
summary(c_show_testimonials)

## Data
## c_testimonials_b, n = 90
## c_testimonials_a, n = 90
##
## Model parameters and generated quantities
## mu_diff: the mean pairwise difference between c_testimonials_b and c_testimonials_a
## sigma_diff: the scale of the pairwise difference, a consistent
## estimate of SD when nu is large.
## nu: the degrees-of-freedom for the t distribution fitted to the pairwise difference
## eff_size: the effect size calculated as (mu_diff - 0) / sigma_diff
## diff_pred: predicted distribution for a new datapoint generated
## as the pairwise difference between c_testimonials_b and c_testimonials_a
##
## Measures
##      mean      sd  HDIlo HDIup %<comp %>comp
## mu_diff    0.000 0.000  0.000 0.000  0.509  0.491
## sigma_diff  0.001 0.000  0.001 0.002  0.000  1.000
## nu          1.004 0.004  1.000 1.013  0.000  1.000
## eff_size   -0.002 0.103 -0.205 0.199  0.509  0.491
## diff_pred  -0.032 5.459 -0.016 0.016  0.499  0.501
##
## '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%
## mu_diff    0.000  0.000  0.000 0.000  0.000
## sigma_diff  0.001  0.001  0.001 0.001  0.002
## nu          1.000  1.001  1.003 1.006  1.016
## eff_size   -0.203 -0.071 -0.002 0.067  0.201
## diff_pred  -0.016 -0.001  0.000 0.001  0.017

cohen.d(c_testimonials_a, c_testimonials_b, paired=TRUE)

##
## Cohen's d
##
## d estimate: 0.2148119 (small)
## 95 percent confidence interval:
##      lower      upper
## 0.1140171 0.3156066
```