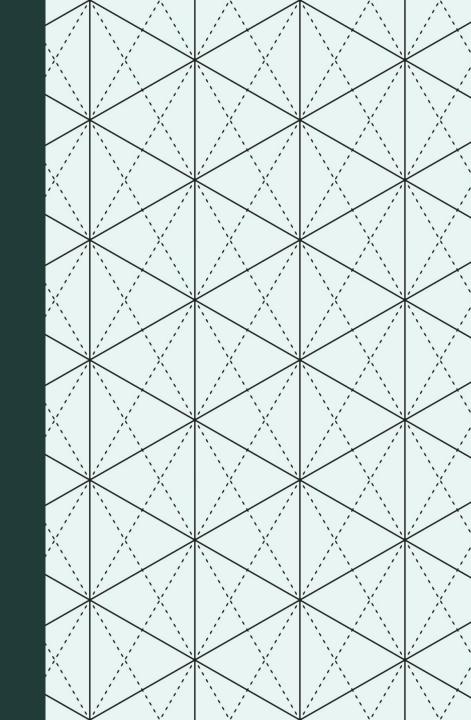
WELCOME TO PSY 653 LAB!

MODULE 11: MINIMUM EFFECTS TESTING & BAYESIAN ANALYSES



*Thanks to Gemma Wallace for her help with these slides

OBJECTIVES

- 1) PRACTICE IDENTIFYING MINIMUM EFFECTS FOR REJECTING SPECIFIC HYPOTHESES
- 2) CONDUCT REGULAR AND BAYESIAN ANOVAS TO DETERMINE WHETHER GENDER AND OCCUPATION ARE RELATED TO MENTAL HEALTH
 - a) Interpret and compare results from the two analytic approaches

PART 1: MINIMUM EFFECTS PRACTICE

MINIMUM EFFECTS TESTING (MET)

 It tests the hypothesis that the effect of treatments falls somewhere in an interval between zero and some number

 Rather than testing if an effect is precisely zero, we can test if it falls above a range of values (The minimum effect you are testing) THE AUTHORS USED RIGHT WING AUTHORITARIANISM (RWA) TO PREDICT DIFFERENCES IN RESPONSE TIME TO IN-GROUP AND OUT-GROUP FACES, AND FOUND A SQUARED CORRELATION OF .07, WHICH WAS SIGNIFICANT, WITH F(1,161) = 4.81

IDENTIFYING MINIMUM EFFECTS: KEVIN'S MINIMUM EFFECTS CODE

```
# 1a Applied to Bret et al. (Right Wing Authoritarianism)
```{r}
dfhyp=1
dferr=161
alpha=.05
effect=.01 #enter the minimum effect you are testing
sse=100
mse=(((1/effect)-1)*sse)/dferr
noncen=sse/mse
qf((1-alpha),dfhyp,dferr,noncen)
```
```

Γ17 8.683388

```
F(1,161) = 4.81,
\alpha = .05
Minimum effect = 1%
Don't change
```

DETECTING MINIMUM EFFECTS: KEVIN'S MINIMUM EFFECTS CODE

```
# 1a Applied to Bret et al. (Right Wing Authoritarianism)
```{r}
dfhyp=1
dferr=161
alpha=.05
effect=.01 #enter the minimum effect you are testing
 F(1,161) = 4.81
sse=100
mse=(((1/effect)-1)*sse)/dferr
noncen=sse/mse
qf((1-alpha),dfhyp,dferr,noncen)
 [1] 8.683388
```

Compare the resulting minimum F-value needed to test an effect of 1% or more to the obtained F-value. In this case we did NOT reach the threshold F-value to have a significant effect at 1%. The authors need an F-value of 8.68 to obtain a significant effect at a minimum effect of 1%.

# IDENTIFYING MINIMUM EFFECTS VIA MURPHY, MYORS & WOLOCH (2014) "ONE STOP" F-TABLE (APPENDIX B)

## MINIMUM F REQUIRED (APPENDIX B)

With a  $DF_{hyp} = 1$  and  $DF_{err} = 161$ , the Authors need an F-value of 3.89 or more to obtain a significant effect.

Obtained F = 4.81

2.20 2.11 2.04 1.98 1.94 1.86 1.78 2.84 2.72 1.21 1.10 1.01 0.94 1.86 2.00 2.38 2.26 3.01 1.47 1.33 1.21 2.52 1.95 3.80 2.78 2.41 2.14 1.92 1.82 1.55 3.70 3.30 2.99 2.10 2.03 2.82 1.20 1.10 1.01 0.93 0.88 2.37 1.99 1.84 1.72 1.56 1.38 3.02 2.87 2.75 1.49 1.34 1.22 1.12 1.04 1.95 3.21 3.83 3.00 2.60 4.56 4.18 3.88 3.65 3.29 2.87 2.49 2.29 2.06 1.87 1.58 2.79 1.25 1.09 1.00 0.92 0.87 0.74 0.64 0.54 1.97 1.82 1.70 3.24 3.04 2.88 2.76 1.36 1.52 3.70 3.41 3.17 2.98 4.80 4.38 4.06 3.80 3.41 6.12 5.35 3.02 3.15 2.73 2.41 2.25 2.04 1.72 8.83 6.78 5.51 4.67 4.06 3.61 3.30 3.00 2.57 2.16 1.71 1.29 1.10 1.08 0.99 0.92 0.86 0.73 2.09 1.68 1.52 2.32 1.94 1.80 3.31 3.09 2.93 2.79 2.58 1.78 1.56 1.40 1.27 1.15 4.03 3.69 3.41 3.20 2.88 2.57 5.78 S 10 4.69 4.33 4.04 3.60 5.19 4.19 3.52 3.04 2.67 2.48 2.25 pow .80 28.49 14.57 9.81 7.46 6.05 5.10 4,43 3.92 3.56 3.24 2.76 2.31 1.81 1.35 1.09 0.85 0.58

One Stop F Table

## MINIMUM F REQUIRED (APPENDIX B)

### COMPARISON TO A NIL EFFECT

With a  $DF_{hyp} = 1$  and  $DF_{err} = 161$ , the Authors need an F-value of 3.89 or more to obtain a significant effect.

Obtained F = 4.81

We have a significant effect

	One Stop F Table																
.~	1	2	3	4	5	6	7	dfH 8	9 9	10	12	15	20	30	40	60	120
90 nil .05 nil .01 pow .50 pow .80 1% .05 1% .01 pow .50 pow .80 5% .05 5% .01 pow .50	6.92 3.86 7.97 6.97 11.29 6.86 12.12 15.17 21.57 14.87	4.85 2.53 4.95 4.37 6.64 3.74 6.62 8.31 1.59 7.58	1.93 3.75 3.48 5.06 2.66 4.74 6.02 8.25 5.14	3.53 1.66 3.12 3.00 4.26 2.18 3.19 4.88 6.59 3.92	3.23 1.43 2.69 2.71 3.77 1.83 3.19 4.15 5.58 3.29	3.01 1.26 2.38 2.52 3.43 1.58 2.78 3.70 4.92 2.78	2.84 1.21 2.18 2.38 3.19 1.47 2.52 3.37 4.44 2.41	2.72 1.10 2.00 2.26 3.01 1.33 2.28 3.12 4.08 2.14	2.61 1.01 1.86 2.18 2.86 1.21 2.10 2.93 3.80 1.92	2.52 0.94 1.74 2.11 2.74 1.12 1.95 2.77 3.57 1.82	2.39 0.89 1.58 2.00 2.56 1.03 1.75 2.54 3.24 1.55	2.24 0.81 1.40 1.88 2.38 0.87 1.51 2.30 2.90 1.34	2.09 0.70 1.20 1.76 2.18 0.79 1.29 2.07 2.55 1.09	2.21 0.85	1.82 0.54 0.86 1.56 1.86 0.55 0.89 1.70 2.03 0.69	1.72 0.46 0.73 1.48 1.74 0.49 0.76 1.58 1.85 0.58	1.60 0.40 0.59 1.40 1.61 0.40 0.60 1.44 1.66 0.45
100 nil .05 nil .01 pow .50 pow .80 1% .05 1% .01 pow .50 pow .80 5% .05 5% .01 pow .50 pow .80	6.89 3.85 7.95 7.24 11.60 7.11 12.45 16.18 22.59 15.62	4.82 2.52 4.94 4.49 6.76 3.84 6.76 8.81 2.08 7.93	3.98 1.92 3.73 3.55 5.13 2.71 4.82 6.27 8.57 5.51	3.51 1.66 3.10 3.04 4.30 2.22 3.83 5.05 6.82 4.19	3.21 1.43 2.67 2.74 3.80 1.85 3.22 4.32 5.76 3.40	2.99 1.26 2.37 2.54 3.45 1.59 2.80 3.83 5.06 2.87	2.82 1.20 2.17 2.39 3.21 1.49 2.53 3.49 4.56 2.49	2.69 1.10 1.99 2.28 3.02 1.34 2.29 3.21 4.18 2.29	2.59 1.01 1.84 2.19 2.87 1.22 2.11 3.00 3.88 2.06	2.50 0.93 1.72 2.11 2.75 1.12 1.95 2.84 3.65 1.87	2.37 0.88 1.56 2.00 2.56 1.04 1.75 2.60 3.29 1.58	2.22 0.80 1.38 1.88 2.37 0.87 1.50 2.34 2.94 1.36	2.07 0.70 1.18 1.76 2.17 0.74 1.26 2.09 2.58 1.11	0.97 1.62 1.96 0.61 1.01	1.80 0.50 0.83 1.55 1.84 0.55 0.88 1.71 2.03 0.70	1.69 0.46 0.71 1.47 1.72 0.49 0.74 1.57 1.84 0.59	1.57 0.39 0.57 1.38 1.58 0.39 0.58 1.43 1.64 0.44
120 nil .05 nil .01 pow .50 pow .80 1% .05 1% .01 pow .50 pow .80 5% .05 5% .01 pow .50 pow .50	6.85 3.84 7.93 7.76 12.20 7.58 13.10 17.88 24.59 17.37	4.79 2.51 4.91 4.74 7.02 4.04 7.05 9.64 3.05 8.79	5.28 2.92 4.98 6.89 9.20 5.92	3.48 1.56 3.05 3.13 4.40 2.29 3.93 5.45 7.28 4.63	3.17 1.42 2.65 2.81 3.86 1.90 3.29 4.64 6.12 3.74	2.96 1.25 2.34 2.59 3.50 1.63 2.85 4.09 5.35 3.15	2.79 1.11 2.12 2.43 3.24 1.52 2.56 3.70 4.80 2.73	2.66 1.09 1.97 2.31 3.04 1.36 2.32 3.41 4.38 2.41	2.56 1.00 1.82 2.21 2.88 1.24	2.47 0.92 1.70 2.13 2.76 1.13 1.97 2.98 3.80 2.04	2.34 0.87 1.54 2.01 2.56 1.05 1.75 2.71 3.41 1.72	2.19 0.74 1.34 1.89 2.36 0.87 1.50 2.43 3.02 1.47	2.03 0.64 1.14 1.75 2.15 0.74 1.25 2.15 2.63	1.55 1.86 0.54 0.92 1.61 1.93 0.61 1.00 1.87 2.23 0.87 1.29	1.76 0.50 0.81 1.54 1.81 0.54 0.86 1.72 2.03 0.74	1.65 0.43 0.68 1.45 1.69 0.46 0.71 1.57 1.83 0.59	1.53 0.36 0.53 1.36 1.55 0.37 0.55 1.42 1.61
1% .01 pow .50 pow .80 5% .05 5% .01 pow .50	6.80 3.83	2.50 4.R9 5.01 7.40 4.42 7.43 0.86 4.46 0.24	1.90 3.09 3.86 5.51 3.09 5.21 7.64 10.12 6.86	3.45 1.55 3.02 3.28 4.56 2.40 4.09 6.06 7.95 5.19	1.41 2.63 2.92 3.98 1.98 3.40 5.11 6.65 4.19	2.92 1.24 2.32 2.66 3.59 1.78 2.96 4.48 5.78 3.52	2.76 1.10 2.09 2.49 3.31 1.56 2.62 4.03 \$ 10 3.04	2.63 1.08 1.94 2.36 3.09 1.40 2.37 3.69 4.69 2.67	0.99 1.80 2.25 2.93 1.27 2.16 3.41 4.33 2.48	2.44 0.92 1.68 2.17 2.79 1.15 2.00 3.20 4.04 2.25	2.30 0.86 1.52 2.03 2.58 1.06 1.77 2.88 3.60 1.90	2.16 0.73 1.31 1.90 2.37 0.88 1.51 2.57 3.17 1.61	2.00 0.63 1.11 1.76 2.15 0.74 1.25 2.24 2.73 1.23	1.83 0.54 0.90 1.61 1.92 0.61 0.98 1.92 2.28 0.93	0.45 0.77 1.53 1.79 0.51 0.83 1.75 2.06 0.75	0.64 1.44 1.66 0.43 0.68 1.59 1.83 0.59	0.49 1.34 1.51 0.34 0.51 1.41 1.59 0.41

## MINIMUM F REQUIRED (APPENDIX B)

### COMPARISON TO A 1% EFFECT

With a  $DF_{hyp} = 1$  and  $DF_{err} = 161$ , the Authors need an F-value of 8.61 or more to obtain a significant effect.

Obtained F = 4.81

We do NOT have a significant effect

								One	Stop F									
dfErr		1	2	3	4	5	6	7	dfH 8		10	12	15	20	30	40	60	120
90 nil nil pow pow 1% 1%	.01 .50 .80 .05 .01 .50 .80 .05	3.86 7.97 6.97 11.29 6.86 12.12	4.85 2.53 4.95 4.37 6.64 3.74 6.62 8.31 1.59 7.58	3.75 3.48 5.06 2.66 4.74 6.02 8.25 5.14	3.53 1.66 3.12 3.00 4.26 2.18 3.19 4.88 6.59 3.92	3.23 1.43 2.69 2.71 3.77 1.83 3.19 4.15 5.58 3.29	1.26 2.38 2.52 3.43 1.58 2.78 3.70 4.92 2.78	2.84 1.21 2.18 2.38 3.19 1.47 2.52 3.37 4.44 2.41	2.72 1.10 2.00 2.26 3.01 1.33 2.28 3.12 4.08 2.14	2.61 1.01 1.86 2.18 2.86 1.21 2.10 2.93 3.80 1.92	2.52 0.94 1.74 2.11 2.74 1.12 1.95 2.77 3.57 1.82	2.39 0.89 1.58 2.00 2.56 1.03 1.75 2.54 3.24 1.55	2.24 0.81 1.40 1.88 2.38 0.87 1.51 2.30 2.90 1.34	0.70 1.20 1.76 2.18 0.79 1.29 2.07 2.55 1.09	1.91 0.59 0.98 1.63 1.97 0.65 1.04 1.83 2.21 0.85	0.54 0.86 1.56 1.86 0.55 0.89 1.70 2.03 0.69	0.73 1.48 1.74 0.49 0.76 1.58 1.85 0.58	1.60 0.40 0.59 1.40 1.61 0.40 0.60 1.44 1.66 0.45
100 nil nil pow pow 1% 1% pow pow 5% 5% pow	.01 .50 .80 .05 .01 .50 .80 .05	6.89 3.85 7.95 7.24 11.60 7.11	4.82 2.52 4.94 4.49 6.76 3.84 6.76 8.81 2.08 7.93	6.27 8.57 5.51	3.51 1.66 3.10 3.04 4.30 2.22 3.83 5.05 6.82 4.19	3.21 1.43 2.67 2.74 3.80 1.85 3.22 4.32 5.76 3.40	2.99 1.26 2.37 2.54 3.45 1.59 2.80 3.83 5.06 2.87	2.82 1.20 2.17 2.39 3.21 1.49 2.53 3.49 4.56 2.49	2.69 1.10 1.99 2.28 3.02 1.34 2.29 3.21 4.18 2.29	2.59 1.01 1.84 2.19 2.87 1.22 2.11 3.00 3.88 2.06	2.50 0.93 1.72 2.11 2.75 1.12 1.95 2.84 3.65 1.87	2.37 0.88 1.56 2.00 2.56 1.04 1.75 2.60 3.29 1.58	2.22 0.80 1.38 1.88 2.37 0.87 1.50 2.34 2.94 1.36	2.07 0.70 1.18 1.76 2.17 0.74 1.26 2.09 2.58 1.11	1.89 0.59 0.97 1.62 1.96 0.61 1.01 1.84 2.21 0.86	1.80 0.50 0.83 1.55 1.84 0.55 0.88 1.71 2.03 0.70	1.69 0.46 0.71 1.47 1.72 0.49 0.74 1.57 1.84 0.59	1.57 0.39 0.57 1.38 1.58 0.39 0.58 1.43 1.64 0.44
5%	.01 .50 .80 .05 .01 .50 .80 .05	7.93 7.76 12.20 7.58 13.10 17.88	4.79 2.51 4.91 4.74 7.02 4.04 7.05 9.64 3.05 8.79	3.95 1.91 3.71 3.66 5.28 2.92 4.98 6.89 9.20 5.92	3.48 1.56 3.05 3.13 4.40 2.29 3.93 5.45 7.28 4.63	3.17 1.42 2.65 2.81 3.86 1.90 3.29 4.64 6.12 3.74	2.96 1.25 2.34 2.59 3.50 1.63 2.85 4.09 5.35 3.15	2.79 1.11 2.12 2.43 3.24 1.52 2.56 3.70 4.80 2.73	2.66 1.09 1.97 2.31 3.04 1.36 2.32 3.41 4.38 2.41	2.56 1.00 1.82 2.21 2.88 1.24 2.12 3.17 4.06 2.25	2.47 0.92 1.70 2.13 2.76 1.13 1.97 2.98 3.80 2.04	2.34 0.87 1.54 2.01 2.56 1.05 1.75 2.71 3.41 1.72	2.19 0.74 1.34 1.89 2.36 0.87 1.50 2.43 3.02 1.47	2.03 0.64 1.14 1.75 2.15 0.74 1.25 2.15 2.63 1.13	1.86 0.54 0.92 1.61 1.93 0.61 1.00 1.87 2.23 0.87	1.76 0.50 0.81 1.54 1.81 0.54 0.86 1.72 2.03 0.74	1.65 0.43 0.68 1.45 1.69 0.46 0.71 1.57 1.83 0.59	1.53 0.36 0.53 1.36 1.55 0.37 0.55 1.42 1.61 0.42
150 nil nil pow	.01 .50	6.80 3.83 7.00	4.75 2.50 4.80	1.90	3.45 1.55 3.02	3.14 1.41 2.63	2.92 1.24 2.32	2.76 1.10 2.00	2.63 1.08	2.53 0.99	2.44 0.92	2.30 0.86	2.16 0.73	2.00 0.63	1.83 0.54	1.73 0.45 0.77	1.62 0.40 0.64	1.49 0.33 0.40
1%		8.61																
pow pow 5% 5% pow	.50 .80 .05 .01	8.26 14.11 20.52 27.47 19.73 28.49	4.42 7.43 0.86 4.46 0.24	3.09 5.21 7.64 10.12 6.86	2.40 4.09 6.06 7.95 5.19	1.98 3.40 5.11 6.65 4.19	1.78 2.96 4.48 5.78 3.52	1.56 2.62 4.03 S 10 3.04	1.40 2.37 3.69 4.69 2.67	1.27 2.16 3.41 4.33 2.48	1.15 2.00 3.20 4.04 2.25	1.06 1.77 2.88 3.60 1.90	0.88 1.51 2.57 3.17 1.61	0.74 1.25 2.24 2.73 1.23	0.61 0.98 1.92 2.28 0.93	0.51 0.83 1.75 2.06 0.75	0.43 0.68 1.59 1.83 0.59	0.34 0.51 1.41 1.59 0.41

PART 2: BAYESIAN ANALYSES

#### A QUICK INTRODUCTION TO BAYESIAN STATISTICS

- Increasingly popular in psychology
- Models account for background knowledge (not discussed in detail in this lab)
- May allow us to overcome several limitations of NHSTs (e.g., the reproducibility crisis in psychology and reliance on large sample sizes)
- All common statistical analyses can be conducted in a bayesian framework (ANOVAs, regression, correlation, factor analysis, etc.)
- Interpretations are often more concrete than significance tests (clear probability statements)
- Today we will be calculating a statistic called a "Bayes factor"

Note: While bayesian analyses are on the rise and may overcome several limitations of frequentist methods, use of classical statistics has been argued for as well (e.g., this NY Times article from 2014: https://www.nytimes.com/2014/09/30/science/the-odds-continually-updated.html?\_r=1)

**Table 1.** Evidence Categories for *p* Values (adapted from Wasserman, 2004, p. 157), for Effect Sizes (as proposed by Cohen, 1988), and for Bayes Factor BF<sub>A0</sub> (Jeffreys, 1961)

Statistic	Interpretation
p value	
<.001	Decisive evidence against $H_0$
.00101	Substantive evidence against H <sub>0</sub>
.0105	Positive evidence against $H_0$
>.05	No evidence against H <sub>0</sub>
Effect size	_
<0.2	Small effect size
0.2-0.5	Small to medium effect size
0.5–0.8	Medium to large effect size
0.8	Large to very large effect size
Bayes factor	
>100	Decisive evidence for $H_A$
30-100	Very strong evidence for $H_A$
10–30	Strong evidence for $H_A$
3–10	Substantial evidence for $H_A$
I <i>-</i> 3	Anecdotal evidence for $H_A$
1	No evidence
1/3–1	Anecdotal evidence for $H_0$
1/10–1/3	Substantial evidence for $H_0$
1/30–1/10	Strong evidence for $H_0$
1/100-1/30	Very strong evidence for $H_0$
<1/100	Decisive evidence for $H_0$

Note: For the Bayes factor categories, we replaced the label "worth no more than a bare mention" with "anecdotal." Also, in contrast to p values, the Bayes factor can quantify evidence in favor of the null hypothesis.

## RULES OF THUMB FOR BAYES FACTOR INTERPRETATIONS

Bayes factors are indices of *relative* evidence of one model (or hypothesis) over another

Evidence for alternative hypothesis (compared to null hypothesis)

Evidence for null hypothesis (compared to alternative hypothesis)

Wetzels et al., 2011

https://journals.sagepub.com/doi/abs/10.1177/1745691611406923

#### DATASET DESCRIPTION

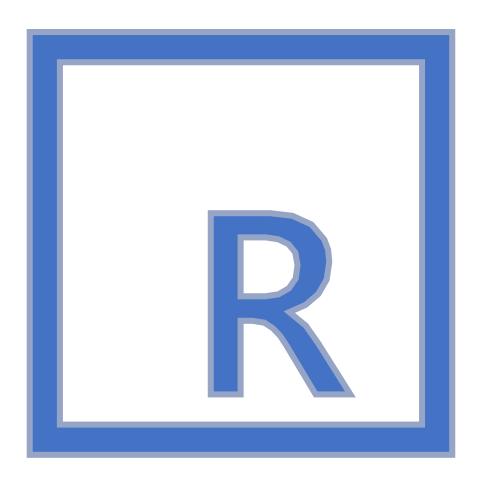
A team of researchers analyzed if People's mental health after retirement was better or worse based on their sex and previous occupation.

- $\times$  **sex:** participant sex. 1 = Female, 2 = Male
- x occupation: Participant occupation prior to retirement. 1 = Professional, 2 = Manager, 3 = nonmanual worker, 4 = Skilled worker, 5 = semi-skilled worker, 6 = unskilled worker.
- **mental:** The participants score on a mental health measure. Higher scores indicate better mental health.

## LET'S CODE!

LAST TUTORIAL OF THE SEMESTER! ©

YOU ALL HAVE BEEN GREAT!



## CREATE A NEW R-PROJECT AND R-NOTEBOOK!

Download the "retirement.csv" file from Canvas and save it into your R-project file

#### LOAD LIBRARIES

```
Load libraries
```{r}
library(tidyverse)
library(psych)
library(BayesFactor)
```

READ IN THE retirement.csv

```
# read in retirement.csv
```{r}
retirement <- read_csv("retirement.csv")</pre>
```

```
-- Column specification -----
cols(
 occupation = col_double(),
 sex = col_double(),
 mental = col_double()
```

#### FACTOR CATEGORICAL VARIABLES

#### **DESCRIBE DATA**

#### # describe data ```{r}

describe(retirement)



Description: df[,13] [5 x 13]

	vars <dbl></dbl>	<b>n</b> <dbl></dbl>	mean <dbl></dbl>	<b>sd</b> <dbl></dbl>	median <dbl></dbl>	trimmed <dbl></dbl>	mad <dbl></dbl>	min <dbl></dbl>	max <dbl> ▶</dbl>
occupation	1	1910	3.15	1.25	3	3.09	1.48	1	6
sex	2	1910	1.61	0.49	2	1.64	0.00	1	2
mental	3	1910	3.79	0.91	4	3.84	1.48	1	5
sex.f*	4	1910	1.61	0.49	2	1.64	0.00	1	2
occupation.f*	5	1910	3.15	1.25	3	3.09	1.48	1	6

5 rows | 1-10 of 13 columns

#### RUN A NORMAL ANOVA

```
Normal ANOVA
```{r}
anova(lm(mental \sim sex.f*occupation.f, data = retirement))
Analysis of Variance Table
Response: mental
                   Df Sum Sq Mean Sq F value Pr(>F)
sex.f
                    1 0.21 0.2062 0.2558 0.6131
             5 45.93 9.1865 11.3962 7.439e-11 ***
occupation.f
sex.f:occupation.f 5 6.01 1.2021 1.4912 0.1894
Residuals 1898 1529.99 0.8061
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
```

CALCULATE ETA² EFFECT SIZE

Eta² = SSeffect / SStotal

- [1] 0.0001327316
- [1] 0.0290303
- [1] 0.003798652

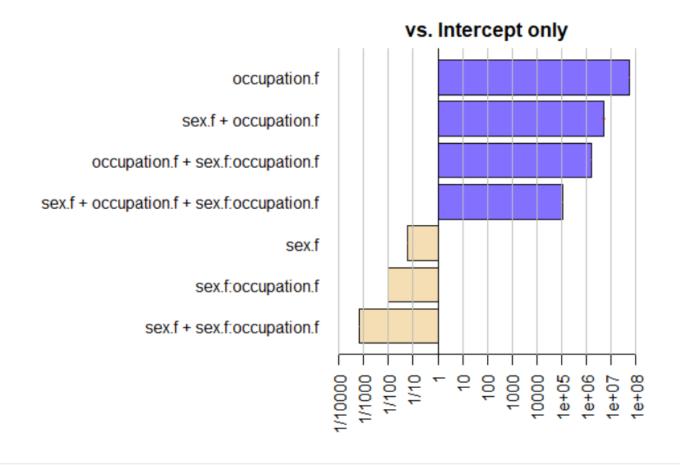
RUN BAYES FACTOR ANOVA, which Models = "all"

Bayes factor type: BFlinearModel, JZS

```
# Bayes ANOVA: "all"
```{r}
m1 <- anovaBF(mental~ sex.f*occupation.f, data = retirement, whichModels = "all")
m1
Bayes factor analysis
 Shows Bayes factor for all
 [1] sex.f
 : 0.05966566
 +0%
 possible models.
 [2] occupation.f
 : 54417793
 +0.01%
 [3] sex.f:occupation.f
 \pm 0.1\%
 : 0.009737589
 [4] sex.f + occupation.f
 : 5049013
 \pm 7.05\%
 [5] sex.f + sex.f:occupation.f
 : 0.0006661466 ±2.5%
 As can be seen: Any model
 [6] occupation.f + sex.f:occupation.f
 : 1620368
 \pm 0.69\%
 with the inclusion of
 [7] sex.f + occupation.f + sex.f:occupation.f : 112545.7
 \pm 1.14\%
 occupation.f main effect results
Against denominator:
 in an extremely large Bayes
 Intercept only
 factor!
```

#### PLOT BAYES FACTOR ANOVA, which Models = "all"

```{r} plot(m1)



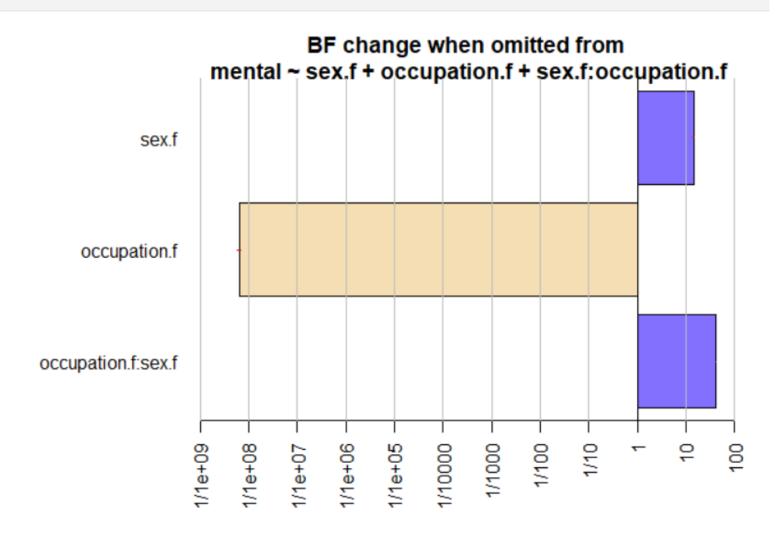
RUN BAYES FACTOR ANOVA, which Models = "top"

```
#Bayes ANOVA: "top"
 ``{r}
m2 <- anovaBF(mental~ sex.f*occupation.f, data = retirement, whichModels = "top")
 Bayes factor top-down analysis
                                                                               Shows how the omission of a
When effect is omitted from sex.f + occupation.f + sex.f:occupation.f , BF is...
                                                                               variable will affect the Bayes
 [1] Omit occupation.f:sex.f: 41.26221
                                        +1.68%
    Omit occupation.f
                       : 6.41187e-09 ±8.88%
                                                                               factor value
 [3] Omit sex.f
                  : 14.37825
                                        +1.62%
Against denominator:
                                                                               As can be seen: Any model
  mental ~ sex.f + occupation.f + sex.f:occupation.f
                                                                               with the omission of
Bayes factor type: BFlinearModel, JZS
                                                                               occupation.f main effect
                                                                               results in an extremely small
```

Bayes factor!

PLOT BAYES FACTOR ANOVA, which Models = "top"

```{r} plot(m2)

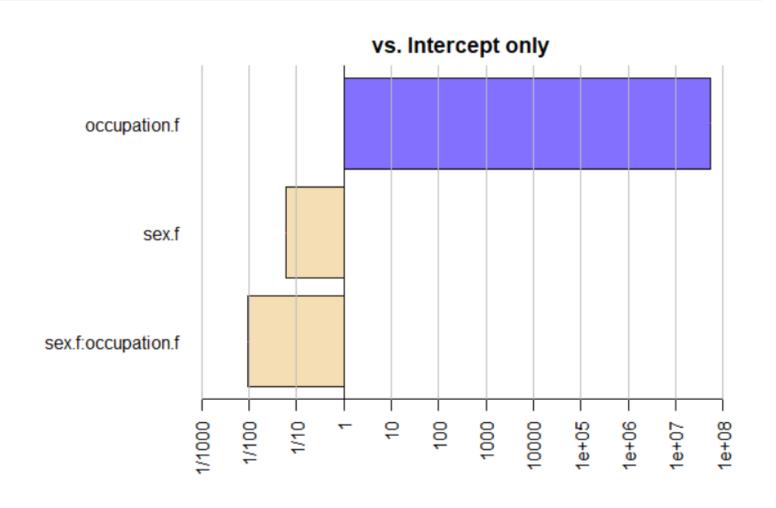


#### RUN BAYES FACTOR ANOVA, which Models = "bottom"

```
#Bayes ANOVA: "bottom"
 `{r}
m3 <- anovaBF(mental~ sex.f*occupation.f, data = retirement, whichModels = "bottom")
Bayes factor analysis
 Shows Bayes factor for
 [1] sex.f
 : 0.05966566 ±0%
 adding in a variable
 [2] occupation.f : 54417793
 \pm 0.01\%
 [3] sex.f:occupation.f: 0.009737589 ±0.1%
 As can be seen: Any
Against denominator:
 Intercept only
 model with the addition of
 occupation.f main effect
Bayes factor type: BFlinearModel, JZS
 results in an extremely
 large Bayes factor!
```

#### PLOT BAYES FACTOR ANOVA, which Models = "bottom"

\`\{r} plot(m3)



#### ADDITIONAL RESOURCES ON BAYESIAN APPROACHES

An article by Etz & Vandekerckhove (2018) about basic bayesian inferences. It opens with a quote by Dumbledore, so you know you want to read it! <a href="https://link.springer.com/article/10.3758/s13423-017-1262-3">https://link.springer.com/article/10.3758/s13423-017-1262-3</a>

Helpful tutorials for learning bayesian analyses using the BayesFactor package: <a href="https://richarddmorey.github.io/BayesFactor/#fixed">https://richarddmorey.github.io/BayesFactor/#fixed</a>

More great tutorials for getting started with bayesian analyses, this time from the BayestestR package: <a href="https://cran.r-project.org/web/packages/bayestestR/vignettes/bayestestR.html">https://cran.r-project.org/web/packages/bayestestR/vignettes/bayestestR.html</a>

The accompanying citation for the BayestestR package can be found at: <a href="https://www.theoj.org/joss-papers/joss.01541/10.21105.joss.01541.pdf">https://www.theoj.org/joss-papers/joss.01541/10.21105.joss.01541.pdf</a>

An example of using top down and bottom-up approaches with bayesian analyses: <a href="https://datascienceplus.com/bayesian-statistics-analysis-of-health-data/">https://datascienceplus.com/bayesian-statistics-analysis-of-health-data/</a>

An article by Krypotos et al. (2017) that calls for increased use of Bayesian approaches (and less NHST) in experimental psychology: <a href="https://journals.sagepub.com/doi/10.5127/jep.057316">https://journals.sagepub.com/doi/10.5127/jep.057316</a>

#### THANKS FOR A GREAT SEMESTER!

- × Once a student, always a student!
- × Never hesitate to reach out in the future! ndyetz@gmail.com
- × I will keep all the YouTube videos up for your future reference

#### THANK YOU FOR BEING SUCH A WONDERFUL CLASS!

I enjoyed every minute of it!