

# Bivariate analysis

[vanderbi.lt/r](http://vanderbi.lt/r)

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# Before we start...

- need to install MVN library

- test by:

```
library(MVN)
```

- install by:

```
install.packages("MVN")
```

- or use the package installer on the Packages tab of the lower right pane of RStudio

# Categories of analysis for today:

- two continuous variables: **correlation**
- one continuous, one discontinuous variable: **t-test of means**
- two discontinuous variables: **chi-squared contingency test** (a.k.a. *test of association*). I think this is also known as "*cross tabulation*". NOT to be confused with *chi-squared goodness of fit test*.

# Recap maternal\_closeness

```
mutate(mc_H1PF1 = ifelse(!H1PF1==1 &
!is.na(H1PF1), 0, ifelse(H1PF1==1, 1,
NA)))
```

- `!H1PF1==1 & !is.na(H1PF1)` not a 1 and not a NA... : 0 if true, otherwise...
- `H1PF1==1` if a 1: 1 if true, otherwise NA
- Final test requires all five responses to be 1, otherwise none to be NA

# What is P ?

- What is P?
  - P is the probability that we would see results like this if nothing interesting were going on (variation is random).
  - $P = 0.6$  (could be like this 60% of the time if random; **likely to be random**)
  - $P = 0.001$  (could be like this 0.1% of the time if random; **not likely to be random**)
- If it's really unlikely that our results would occur when only random things are happening, we think something interesting is going on.

# Why do we like it when $P < 0.05$ ?

- Hypotheses:
  - things are different (alternate hypothesis)
  - things are the same (null hypothesis)
- Strategy:
  - show that the null hypothesis is wrong
- If  $P < 0.05$ , then we assume the null hypothesis is wrong because it's so unlikely.
- If  $P > 0.05$ , then either the null hypothesis is wrong **or our experiment SUCKS !**
- We probably know what's going on if  $P < 0.05$  **but not if  $P > 0.05$**

# Statistical power

- **Power** is the ability to show that different things are different ( $P < 0.05$ )
- We get more statistical power if there's **less variation** in the data or a **larger sample size**.
- We may be able to control variation by experimental conditions
- We should be able to increase the sample size (if we have time and money).
- If **not different**, increasing power **won't reduce P**.
- If **different**, increasing power will **make P get smaller**.

# Power tradeoff

- Too little statistical power:
  - can't show that different things are different
  - Unable to get  $P > 0.05$  when there are differences.
- It seems like more statistical power would always be a good thing, but...
- Too much statistical power
  - tiny unimportant things are shown to be different
  - $P < 0.05$  for factors with a very small effect.