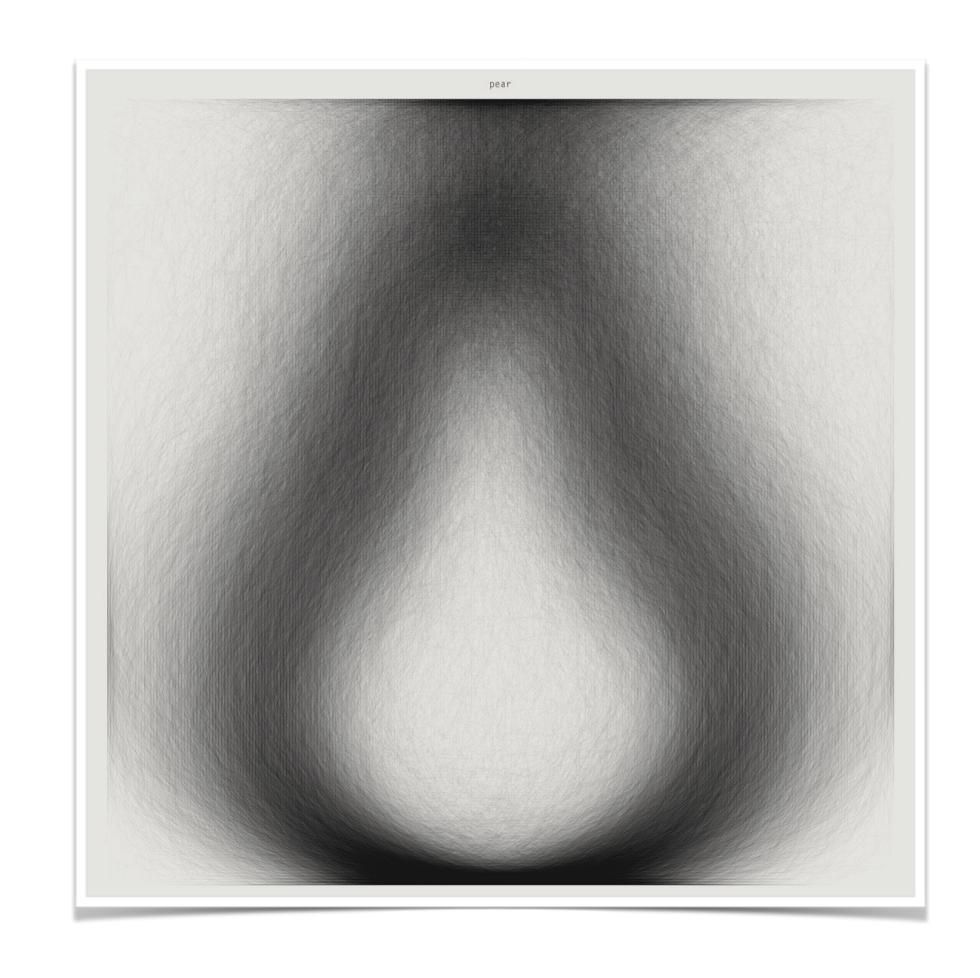
IT UNIVERSITY OF COPENHAGEN

Reflections in Data Science BSREDAS1KU-20201



2020-03-03

Pear,
Overlap of drawings from people all around the world
formafluens.io

Exercise 4 - Plan

Hack your way to scientific glory (10 minutes)

•P-Hacking. (20 minutes)

Multiple Comparisons Bias and Bonferroni's correction. (20 minutes)

Hack your way to scientific glory

TORTURE DATA
LONG ENOUGH
THEY WILL
CONFESS.

-RONALD COASE

https://projects.fivethirtyeight.com/p-hacking/

- Please record the configurations you found more interesting.

Hack your way to scientific glory

- •The data can be narrowed and expanded (p-hacked) to make either hypothesis appear correct.
- •Answering even a simple scientific question requires lots of choices that can shape the results.
- •This doesn't mean that science is unreliable. It just means that it's more challenging than we sometimes give it credit for.

P-hacking



- •A study made a hypothesis that green dice are loaded and that they will yield a six outcome more often than usual dice would.
- They took a green die, and rolled it 1000 times.
- •A six appeared 188 times, which is rather high. One would expect that the number of a six outcome to be closer to 1,000/6 = 167.

P-hacking

•It seems that the hypothesis about green dice may have some truth in it.

•To prove it the scientists ran an experiment and computed the p-value.

•You are asked to evaluate their experiment and explain if you agree with their analyses and why.

https://github.com/tocunha/reflectionsdatascience/blob/master/exercise4-02-03-2020/Green-dice-p-value-hacking.ipynb

Multi comparison bias

- •Effectively it's just the notion that the more tests you run, the more likely you are to get false positives
- •Things that look like they confirm your hypothesis, but are really just random chance.

•If you don't correct for this at some point you're very likely to accept hypotheses that aren't based on any real relationships.

•If you look in your data for too many things at the same time, you will see things that look interesting.

•You can expect events to occur, even if the data is completely random.

•The number of occurrences of these events will grow as the size of the data grows.

•Calculate the expected number of occurrences of the events you are looking for, on the assumption that data is random.

•If this number is significantly larger than the number of real instances you hope to find, then you must expect almost anything you find to be bogus.

The original p value.

Bonferroni-corrected p value =
$$\frac{\alpha}{n}$$

The number of tests performed -

Multi comparison bias

https://github.com/tocunha/reflectionsdatascience/blob/master/exercise4-02-03-2020/multiple_comparisons.ipynb