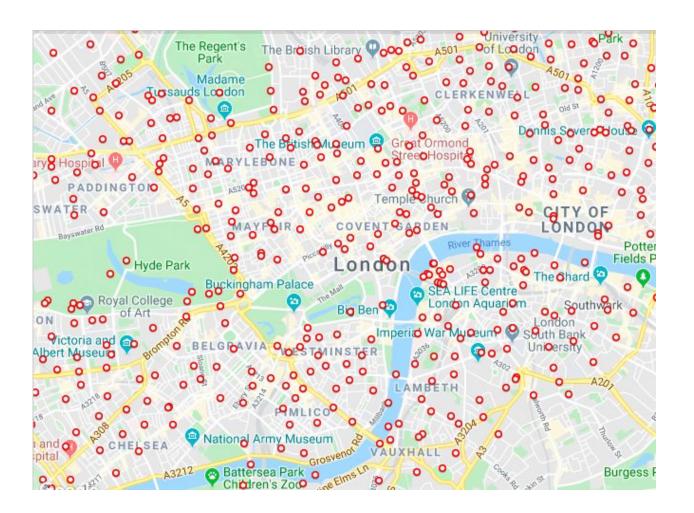
### **Mod 3 Project**

Joe Read and Mohammed Hannan



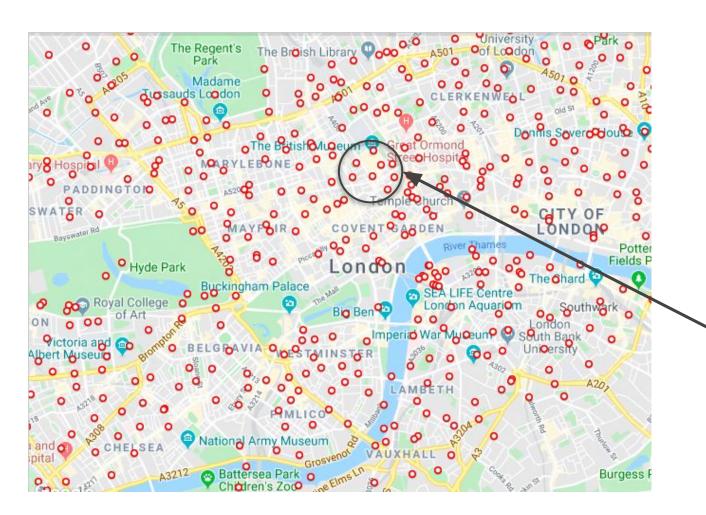
## TFL Bikepoints

#### **Hypotheses**

- 1. A faulty bikepoints will be less likely to have other faulty bikepoints near it
- 2. Bikepoints at high elevation have fewer docks
- 3. Bikepoints in high populated areas are more likely to be faulty
- 4. High populated areas have larger bikepoints

#### Methodology

- Our hypotheses allowed the data to be split into boolean catorageries:
  faulty / non-faulty, hill / no-hill, high pop / non-high pop
- We suspected a small effect size so decided to sample 198 times
- Each sample was formed of 50 data points drawn at random
- The resulting t distribution was formed from the means of each sample



# Local Features

To check local features we looked at all bikepoints within a certain radius

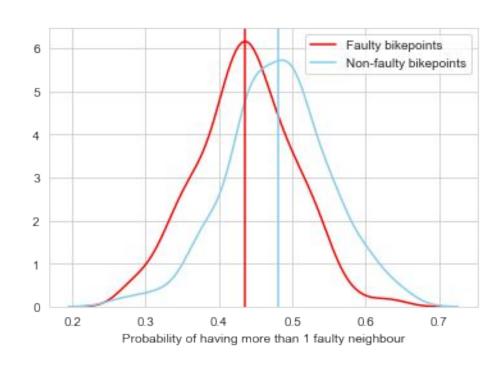
H0 = Faulty bikepoints have the same number of faulty bikepoints nearby as general bikepoints

H1 = Faulty bikepoints have more likely to have more than 1 faulty bikepoints near them than general bikepoints

Statistic: -6.41

Pvalue: 4.22 x 10^-10

Cohen's d: -0.646



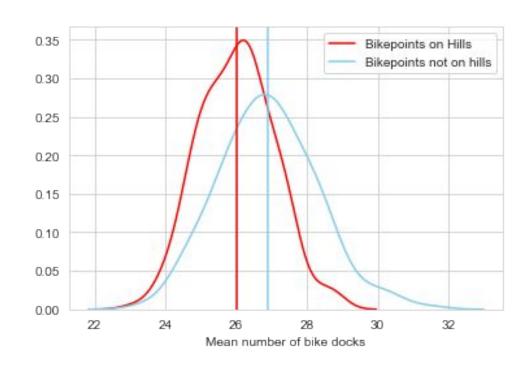
H0 = Bikepoints on hills have the same number of docks as general bikepoints

H1= Bikepoints on hills have fewer docks than general bikepoints

Statistic: -7.15

Pvalue: 4.66 x10^-12

Cohen's d: -0.721



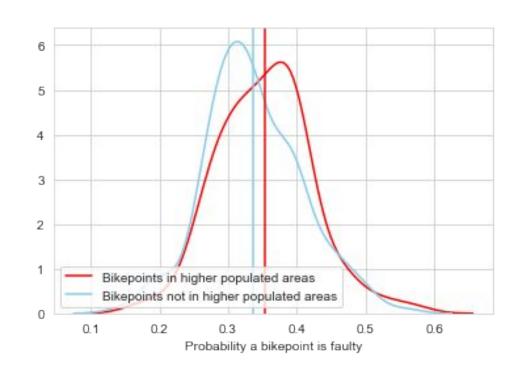
H0 = Bikepoints in highly populated areas have the same probability of having a faulty dock as general bikepoints

H1 = Bikepoints in highly populated areas have a higher probability of having a faulty dock

Statistic: 2.31

Pvalue: 2.14 x10^-2

Cohen's d: 0.233



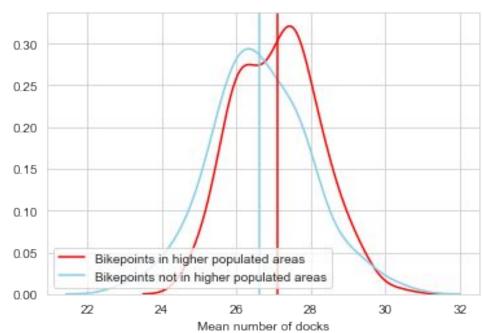
H0 = Bikepoints in highly populated areas have the same number of docks as general bikepoints

H1 = Bikepoints in highly populated areas have more docks than general bikepoints

Statistic: 3.735

Pvalue: 2.16 x 10^-4

Cohen's d: 0.376



# Thanks for listening

Any Questions?

