Final Project

Tutorial section 0212, group 212-3

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Introduction

Riipen is a service where students, professors, and companies can collaborate and complete tasks for each other through "requests". The main dataset we will be working with:

Objectives

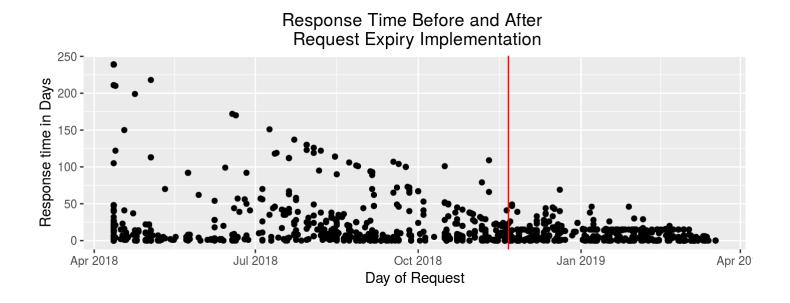
On November 21st, 2018, Riipen introduced a "request expiry" feature where users must respond to requests within 14 days or they expire.

- Did the implementation of this feature affect volume of requests?
- How did the acceptance rate of requests change after the new feature was introduced?
- What about the response time?

Data Summary: Cleaning the Data

- Created two new variables: response_time and featureimplemented
- Filtered out currently pending requests
- Filtered out August 30th

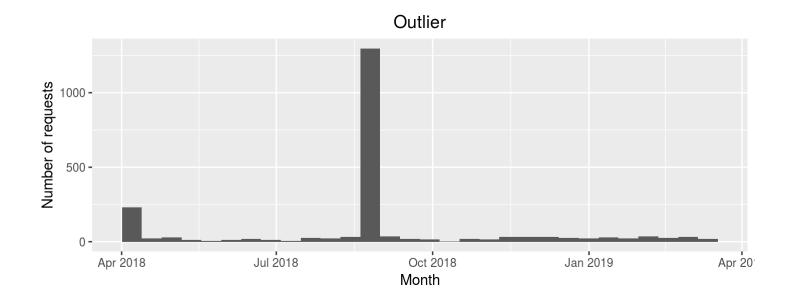
Overview of Data



Data Summary: Why Did We Remove August 30th?

- More than half the observations in the dataset were on august 30th
- All of them were responded to in 0 days
- None of them had ids
- We concluded that a server glitch must have occured on that day

Why we Removed August 30th



How August 30th Skewed the Dataset

Overall Acceptance rate, Before and After New Feature

- Overall: 43.6%
- Before implementation of request-expiry: 38%
- · After implementation: 52%

Mean and Median Request Response Time: Before and After Feature

Before Feature

- Mean response time of 25 days
- Median response time of 20 days

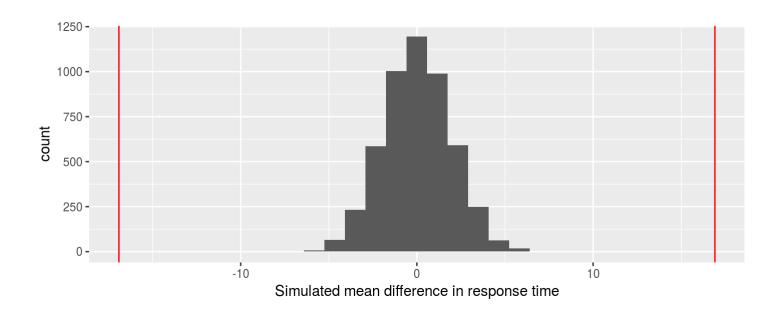
After Feature

- Mean response time of 8 days
- Median response time of 5 days

Statistical Methods

- Significance testing on response time before and after feature
 - H_0 : Response time after response time before = 0
 - H_A : Response time after response time before $\neq 0$
- Significance testing on request acceptance rate before and after feature
 - H_0 : Acceptance rate after Acceptance rate before = 0
 - H_A : Acceptance rate after Acceptance rate before $\neq 0$

Mean Response Time Significance Test

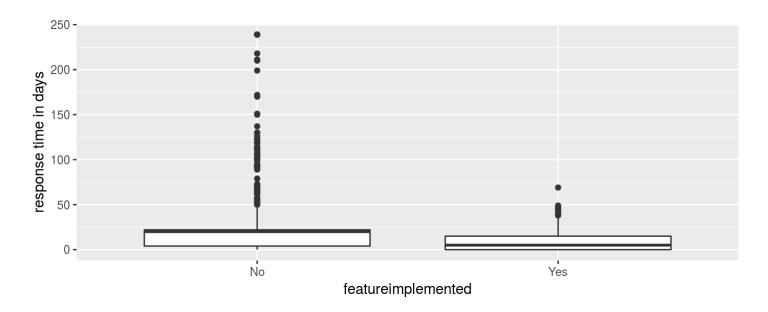


This is a histogram of all the simulated statistics, where our test statistic is represented by the red lines. The p-value is 0.00.

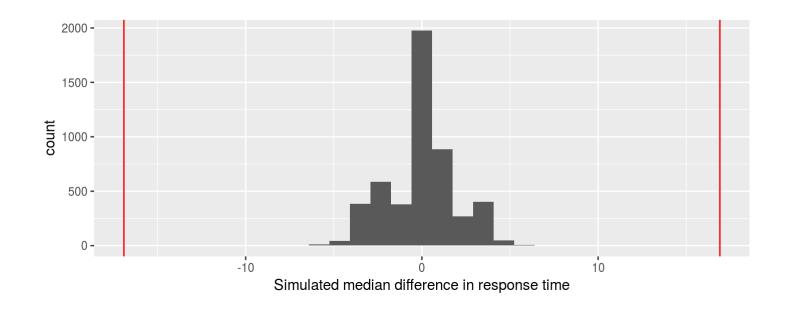
Boxplot of Response Times

```
requests_grouped %>%

ggplot(mapping = aes(x = featureimplemented, y = response_time)) + geom_boxplot()
```

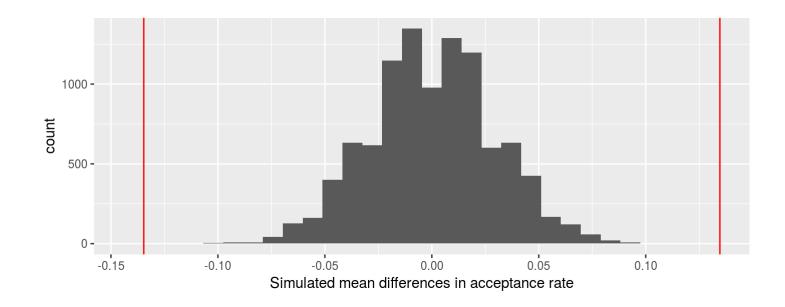


Significance Testing for Median Response Time



When using median instead of mean, the p-value is also 0.00.

Significance Testing for Mean Request Acceptance Rate



We obtained a p-value of 0.00.

Conclusion

 There is a strong evidence against the response time and request acceptance rate being the same after the implementation of the request expiry feature

Type I/II Errors

- Since we rejected the null-hypothesis, there is a possiblity of type I errors, I.E rejecting the null-hypothesis when we shouldn't have
- · However, our p-values are so low that this is not really a concern
- No possibility for Type II errors because we did not uphold any nullhypotheses

Why Did We Obtain a P-value of 0?

- · Theorectically, it is impossible to have p-value of 0
- Recall that we filtered out requests that was created on August 30th 2018, and only used 5000 simulations
- This could be the reason

Limitations

- No data for August 30th 2018
- 5000 simulations may or may not be enough

Acknowledgement

-We would like to thank Emily Masching, the Product Owner of Riipen who provided us to work with the most up-to-date data source.

Thank you for providing us this opportunity to learn more about statistics.