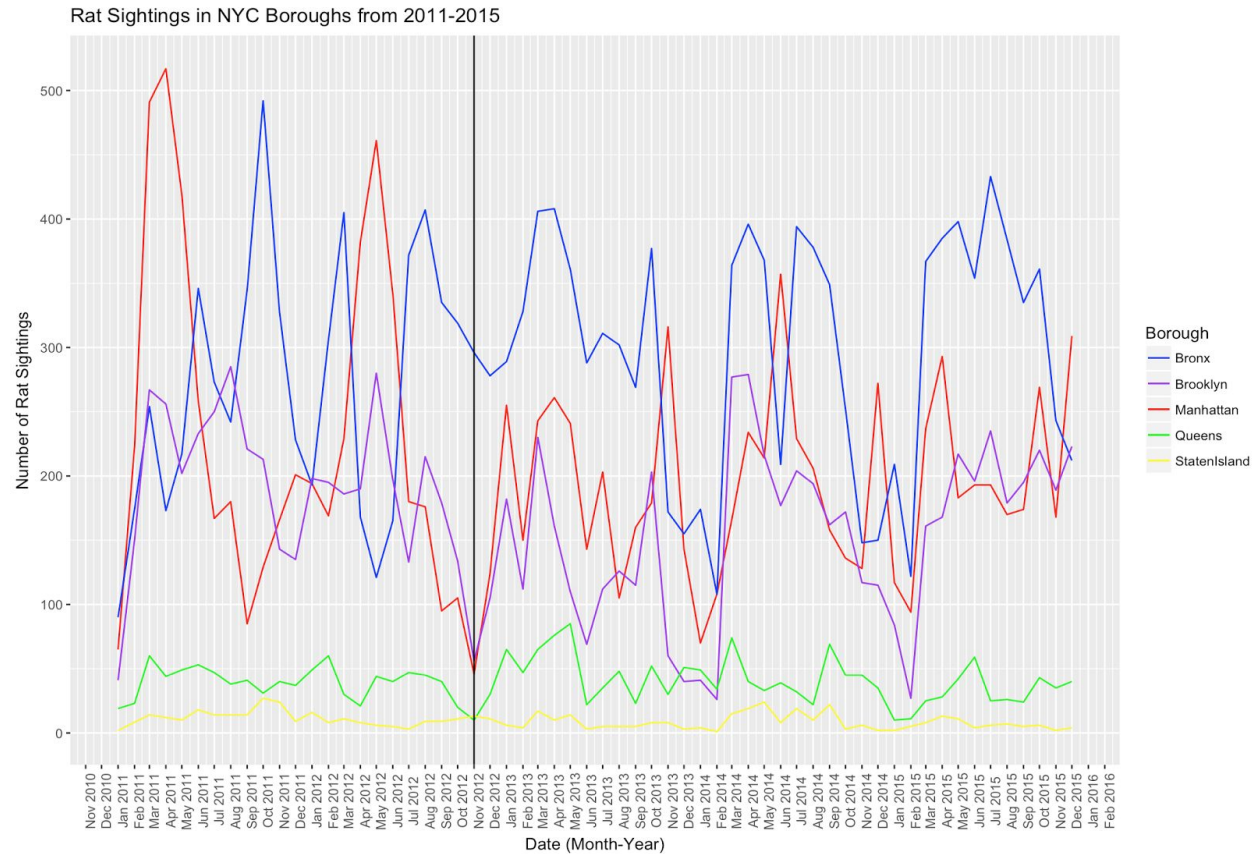


Descriptive Statistics and Figures

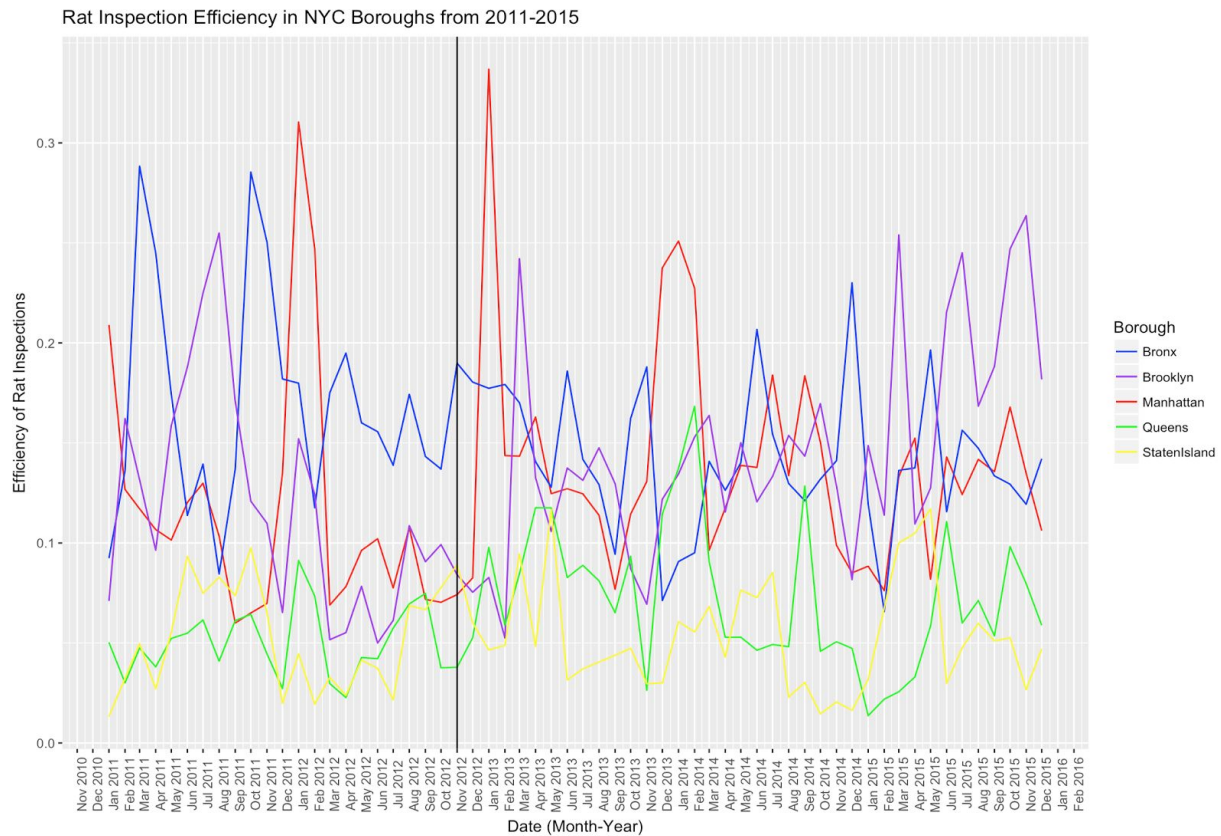
Part 1A.



From the chart, it appears that there are extreme and periodic fluctuations in rat sightings in Manhattan, The Bronx, and Brooklyn over the five year period between 2011 and 2015. In contrast, rat sightings have been relatively infrequent and steady in Queens and Staten Island. Overall, there is also a general downward trend in rat sightings in Brooklyn and Manhattan from 2011 to 2015.

The chart shows a meaningful increase in rat sightings after Hurricane Sandy (Oct-Nov 2012), with the largest spikes occurring in The Bronx, Manhattan, and Brooklyn, and a smaller spike in Queens.

Part 1B.



At a glance, it is clear there was a significant upwards spike in Rat Inspection Efficiency in Manhattan following Hurricane Sandy (Oct-Nov 2012).

Part 1C.

Top 10 “Hot Spot” Zip Codes for Rat Sightings

```
> print(zipRat[1:10,])
Source: local data frame [10 x 3]
Groups: ZIP_CODE [10]
```

	ZIP_CODE	count	BOROUGH
	<dbl>	<dbl>	<fctr>
1	10457	2745	Bronx
2	10458	2595	Bronx
3	10456	2081	Bronx
4	10468	1719	Bronx
5	10453	1503	Bronx
6	11221	1494	Brooklyn
7	10452	1157	Bronx
8	10467	1102	Bronx
9	11237	1070	Brooklyn
10	11206	935	Brooklyn

Most rat sightings are in The Bronx, followed by Brooklyn.

Geographic Patterns

Part 2A.

Top 20 Zip Codes ...

Before 2012

```
> print(topBefore[1:20,])
Source: local data frame [20 x 3]
Groups: ZIP_CODE [20]
```

	ZIP_CODE	count	BOROUGH
	<dbl>	<dbl>	<fctr>
1	10457	978	Bronx
2	10456	890	Bronx
3	10458	816	Bronx
4	11221	646	Brooklyn
5	10453	551	Bronx
6	10468	534	Bronx
7	10031	506	Manhattan
8	11206	480	Brooklyn
9	10032	477	Manhattan
10	10009	455	Manhattan
11	11237	434	Brooklyn
12	10002	424	Manhattan
13	10472	400	Bronx
14	11238	394	Brooklyn
15	11216	391	Brooklyn
16	11211	390	Brooklyn
17	10029	388	Manhattan
18	11217	385	Brooklyn
19	10013	378	Manhattan
20	10459	362	Bronx

```
> |
```

During Hurricane Sandy

```
> print(topSandy[1:20,])
Source: local data frame [20 x 3]
Groups: Incident.Zip [20]
```

	Incident.Zip	count	Borough
	<fctr>	<dbl>	<fctr>
1	10025	14	MANHATTAN
2	10456	11	BRONX
3	11237	10	BROOKLYN
4	11207	9	BROOKLYN
5	11208	9	BROOKLYN
6	10027	8	MANHATTAN
7	10458	8	BRONX
8	10016	7	MANHATTAN
9	10467	7	BRONX
10	11226	7	BROOKLYN
11	10452	6	BRONX
12	11216	6	BROOKLYN
13	11222	6	BROOKLYN
14	10013	5	MANHATTAN
15	10024	5	MANHATTAN
16	10032	5	MANHATTAN
17	10453	5	BRONX
18	11212	5	BROOKLYN
19	11221	5	BROOKLYN
20	11235	5	BROOKLYN

```
> |
```

After 2012

```
> print(topAfter[1:20,])
Source: local data frame [20 x 3]
Groups: ZIP_CODE [20]
```

	ZIP_CODE	count	BOROUGH
	<dbl>	<dbl>	<fctr>
1	10457	1800	Bronx
2	10458	1757	Bronx
3	10456	1225	Bronx
4	10468	1196	Bronx
5	10453	997	Bronx
6	10452	963	Bronx
7	11221	876	Brooklyn
8	10467	764	Bronx
9	11237	738	Brooklyn
10	10009	637	Manhattan
11	10460	636	Bronx
12	11206	627	Brooklyn
13	10029	556	Manhattan
14	10002	534	Manhattan
15	11216	534	Brooklyn
16	10025	528	Manhattan
17	10033	475	Manhattan
18	10027	457	Manhattan
19	10032	439	Manhattan
20	11385	404	Queens

```
>
```

Part 2B.

Based on these results, Hurricane Sandy appears to have displaced rats to Manhattan and Brooklyn during the hurricane. In the aftermath, there is a greater number of rat sightings across all boroughs, with the majority of the rat population still residing in The Bronx both before and after 2012. There appears to be significant displacement of rats moving away from Brooklyn, and towards The Bronx and Manhattan.

Rodents and Restaurant Inspection

Part 3.

```
> set.seed(4698)
> fit = glm(ratViolation ~ activeRatSightings + month + year, data=restaurant, family=binomial)
> summary(fit)

Call:
glm(formula = ratViolation ~ activeRatSightings + month + year,
    family = binomial, data = restaurant)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-0.6512  -0.6315  -0.6069  -0.5824   1.9532

Coefficients:
            Estimate Std. Error z value Pr(>|z|)
(Intercept)  -1.501429   0.022375  -67.103  < 2e-16 ***
activeRatSightings -0.007975   0.020816   -0.383  0.701639
month02      -0.024215   0.021037   -1.151  0.249702
month03      -0.095263   0.020418   -4.666  3.08e-06 ***
month04      -0.143535   0.020148   -7.124  1.05e-12 ***
month05      -0.222871   0.020390  -10.930  < 2e-16 ***
month06      -0.202039   0.020461   -9.875  < 2e-16 ***
month07      -0.172278   0.021081   -8.172  3.03e-16 ***
month08      -0.104337   0.020711   -5.038  4.71e-07 ***
month09      -0.076216   0.020428   -3.731  0.000191 ***
month10      -0.037760   0.020016   -1.886  0.059234 .
month11       0.009284   0.020915    0.444  0.657128
month12       0.004926   0.020104    0.245  0.806452
year2013     -0.014568   0.017582   -0.829  0.407340
year2014      0.048892   0.017331    2.821  0.004786 **
year2015      0.045450   0.017344    2.620  0.008781 **
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

    Null deviance: 405664  on 440981  degrees of freedom
Residual deviance: 405218  on 440966  degrees of freedom
(591 observations deleted due to missingness)
AIC: 405250

Number of Fisher Scoring iterations: 4
```

Note: Data from years 2012-2015.

a) Looking at the logistic regression model summary, the predictor activeRatSightings is associated with a p-value ($p=0.7016$) greater than the threshold $p=0.05$. Thus, the number of Rat Sightings **does not have a statistically significant relationship** with the variable ratViolation (which represents whether a restaurant violation in the same zip code, year, and month is due to rat and mouse problems).

b) A statistically insignificant relationship between active rat sightings and restaurant violations suggests that **neither argument can be supported** by the regression results. From this, we can gather that the

rodent inspection data is **not a useful source** for predicting whether restaurants located in a same zip code are likely to have violations related to rat sightings. I would recommend to the city of New York to use other datasets that may be more useful in predicting restaurant violations, such as customer transactional data for rat exterminator companies, or text mining consumer reviews from restaurant-specific sources like Yelp and Open Table.