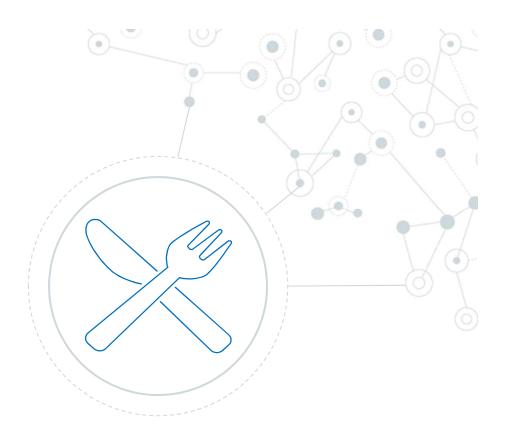


# The Question

Can a restaurant inspection's grade be predicted using the restaurant's identifying information and inspection violations?







147,623
Restaurant Inspections Since 2012

428,405
Total Restaurant Code Violations



## Data Example: The Golden Unicorn

Type: Chinese

Borough: Manhattan

ZIP Code: 10002

Violation Code: 06A

Description: "Personal

cleanliness

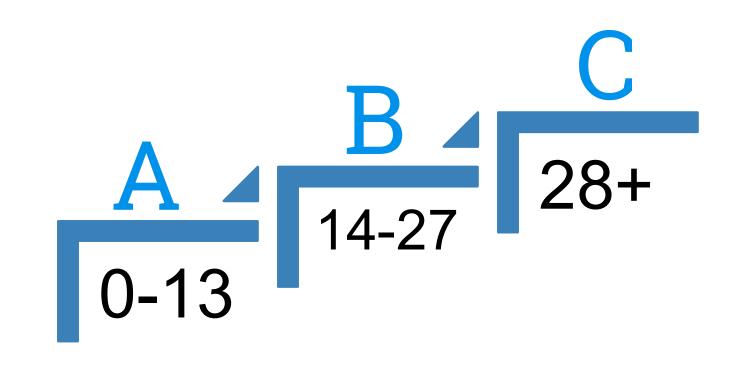
inadequate."

Flag: Critical Violation

Score: 12

Grade: A

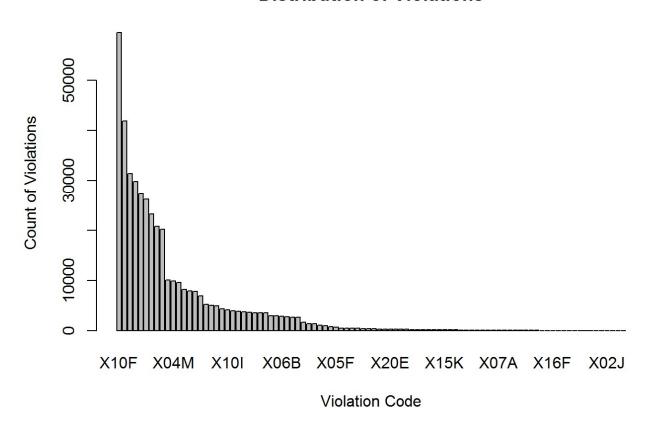






#### Not all violations are created equal

#### **Distribution of Violations**





#### The four most frequent violations

#### **10F**

#### 59,491 occurrences

Non-food contact surface improperly constructed. Unacceptable material used.



**08A** 

#### 41,905 occurrences

Facility not vermin-proof. Harborage or conditions conducive to attracting vermin to the premises and/or allowing vermin to exist.



02**G** 

#### 31,317 occurrences

Cold food item held above 41° F (smoked fish and reduced oxygen packaged foods above 38° F) except during necessary preparation.



04L

#### 29,735 occurrences

Evidence of mice or live mice present in faculty's food and/or non-food areas



### We regressed inspection score on inspection violations to see which violations were more severe

```
## Coefficients: (1 not defined because of singularities)
                Estimate Std. Error t value
##
                                            Pr(>|t|)
                -0.17500 0.02324 -7.530
                                            5.10e-14
## (Intercept)
## X02A
                10.35111 0.21174 48.885
                                            < 2e-16
                7.81572 0.02951 264.892
## X02B
                                            < 2e-16
## X22F
                0.79420 0.25346 3.133
                                            0.001728
## X22G
                -0.79789 0.65068 -1.226
                                            0.220111
## ---
## Residual standard error: 3.903 on 142941 degrees of freedom
## (4588 observations deleted due to missingness)
## Multiple R-squared: 0.8529, Adjusted R-squared: 0.8528
## F-statistic: 8914 on 93 and 142941 DF, p-value: < 2.2e-16
```



#### **Selected severe violations**

#### O7A Associated with score increase of 29.46

Duties of an officer of the Department interfered with or obstructed.



06H 29.29

Records and logs not maintained to demonstrate that HACCP plan has been properly implemented.



05E 23.21

Toilet facility not provided for employees or for patrons when required



04F 21.54

Food, food preparation area, food storage area, area used by employees and patrons, contaminated by sewage or liquid waste.



#### The Modeling

#### Four models used

- ANOVA/Linear Regression
- Naïve Bayes
- Classification Tree
- Random Forest

#### **Predictors**

- Violations
- Restaurant Type
- Zipcode
- Borough

Linear Regression used numeric scores and converted to grade classifications; others used categorical grade classifications from the get-go.

#### Method: ANOVA/Linear Regression

- Best at predicting B and C scores
- Zipcode was not a statistically significant variable, so it was removed

	Actual	Α	В	С
Predicted	Α	7423	156	44
	В	1568	3287	250
	С	1	343	1058

#### Method: ANOVA/Linear Regression

- All restaurant classifications had negative coefficients, with Polynesian restaurants having the smallest coefficient of -3
- Only minor effects associated with borough

cdescripCajun cdescripPolynesian
-1.82287276 -2.91906602
cdescripCalifornian cdescripPortuguese
-1.35129517 -1.12886537
cdescripCaribbean cdescripRussian
-1.33100462 -1.13244733

#### Method: ANOVA/Linear Regression

- All restaurant classifications had negative coefficients, with Polynesian restaurants having the smallest coefficient of -3
- Only minor effects associated with borough

boroBROOKLYN boroQUEENS
-0.18990319 -0.18785666
boroMANHATTAN boroSTATEN ISLAND
-0.14908169 -0.14752918

#### **Method: Naïve Bayes**

- Strong tendency to predict A scores
- Possible model improvements with different classification percentage

	Actual	Α	В	С
Predicted	Α	8459	2504	102
	В	432	1160	926
	С	101	122	324

#### **Method: Naïve Bayes**

- Most probability estimates were very small
- However, "severe" violations from before continued to have the most signal

#### **Method: Classification Tree**

- Strong performance at predicting A, B scores
- Borough, Zipcode, Restaurant Type always ended up being the last splits

	Actual	Α	В	С
Predicted	Α	8858	427	89
	В	117	3207	465
	С	15	152	798



#### **Method: Classification Tree**

#### **05F**

Insufficient or no refrigerated or hot holding equipment to keep potentially hazardous foods at required temperatures.



#### **08A**

Facility not vermin-proof. Harborage or conditions conducive to attracting vermin to the premises and/or allowing vermin to exist.



#### **02G**

Cold food item held above 41 F (smoked fish and reduced oxygen packaged foods above 38 F) except during necessary preparation



#### **Method: Random Forest**

- Performed worse than the classification tree
- The model was quite difficult to read

	Actual	Α	В	С
Predicted	Α	8850	795	141
	В	122	2919	714
	С	18	72	497

#### What to do next

Incorporate date and previous inspection grades in the model



Pursue a more thorough model selection process



Look into using different classification thresholds



# Thanks! Any Questions?

