Project2

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20/03/2022

1 Introduction

The numbers of the missing values in each column:

```
## country_of_origin aroma flavor
## 0 0 0
## acidity category_two_defects altitude_mean_meters
## 0 0 162
## harvested Qualityclass
## 55 0
```

The data after we remove the missing values:

```
## Rows: 858
## Columns: 8
                          <chr> "Guatemala", "China", "Colombia", "Guatemala", "C~
## $ country_of_origin
## $ aroma
                          <dbl> 7.92, 7.67, 7.75, 7.83, 7.67, 8.17, 7.83, 7.67, 7~
## $ flavor
                          <dbl> 7.67, 7.67, 7.50, 7.67, 7.42, 8.00, 7.50, 7.75, 7~
## $ acidity
                          <dbl> 7.75, 7.67, 7.50, 7.33, 7.33, 7.17, 7.42, 7.67, 7~
## $ category_two_defects <int> 3, 3, 0, 1, 5, 0, 2, 1, 4, 0, 10, 0, 4, 4, 2, 4, ~
## $ altitude_mean_meters <dbl> 1650.00, 1600.00, 1750.00, 1310.64, 1600.00, 1750~
                          <int> 2015, 2015, 2013, 2013, 2011, 2014, 2013, 2015, 2~
## $ harvested
## $ Qualityclass
                         <chr> "Good", "Good", "Poor", "Poor", "Good", "~
```

The number of unique values in country of origin:

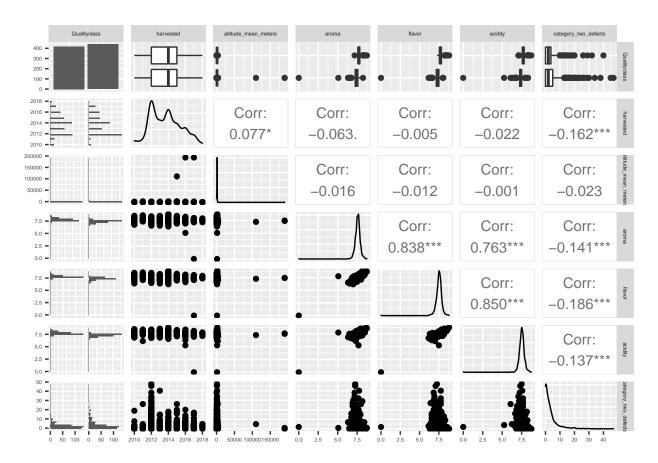
```
## [1] 34
```

The number of unique values in harvest year:

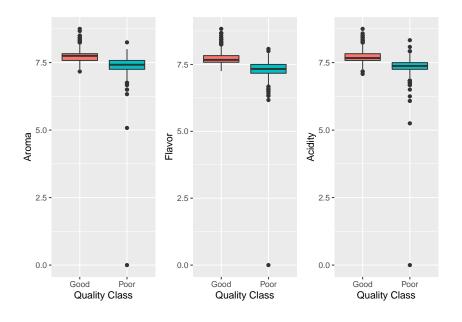
```
## [1] 9
```

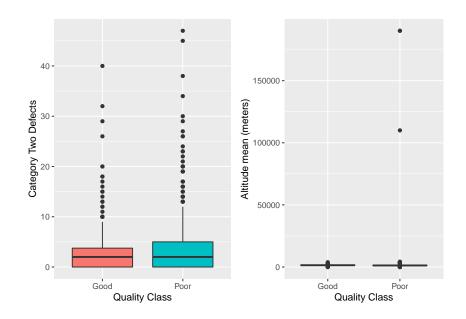
2 Explantory Analysis

The correlation between the quantitative variables:

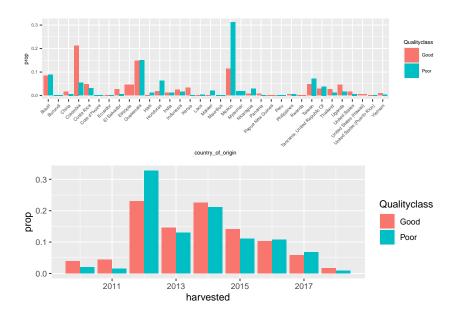


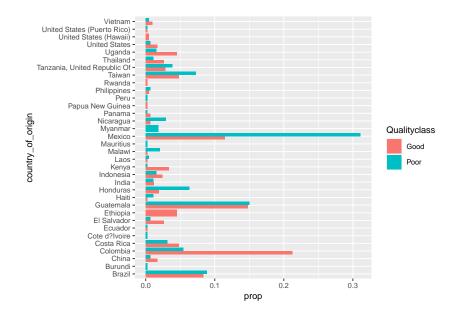
Box plots showing the distribution of the quantitative variables

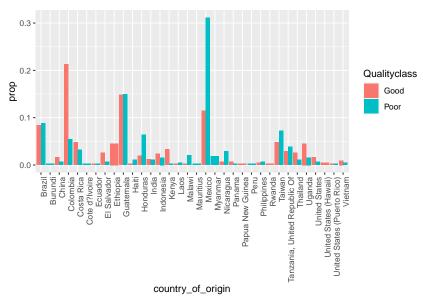




2.1 bar charts:







The percentages:

Table showing the percentage of the quality classes for each country

Table 1: The Proportion of Quality Classs in Different Country

country_of_origin	Good	Poor
Brazil	47.3% (35)	52.7% (39)
Burundi	0.0% (0)	100.0% (1)
China	70.0% (7)	30.0% (3)
Colombia	78.8% (89)	21.2% (24)
Costa Rica	58.8% (20)	41.2% (14)
Cote d?Ivoire	0.0% (0)	100.0% (1)
Ecuador	50.0% (1)	50.0% (1)
El Salvador	78.6% (11)	21.4% (3)
Ethiopia	100.0% (19)	0.0% (0)
Guatemala	48.4% (62)	51.6% (66)
Haiti	16.7% (1)	83.3% (5)
Honduras	22.2% (8)	77.8% (28)
India	50.0% (5)	50.0% (5)
Indonesia	58.8% (10)	41.2% (7)
Kenya	93.3% (14)	6.7% (1)
Laos	33.3% (1)	66.7% (2)
Malawi	10.0% (1)	90.0% (9)
Mauritius	0.0% (0)	100.0% (1)
Mexico	25.9% (48)	74.1% (137)
Myanmar	0.0% (0)	100.0% (8)
Nicaragua	18.8% (3)	81.2% (13)
Panama	75.0% (3)	25.0% (1)
Papua New Guinea	100.0% (1)	0.0% (0)
Peru	0.0% (0)	100.0% (1)
Philippines	40.0% (2)	60.0% (3)
Rwanda	100.0% (1)	0.0% (0)
Taiwan	38.5% (20)	61.5% (32)
Tanzania, United Republic Of	41.4% (12)	58.6% (17)
Thailand	68.8% (11)	31.2% (5)
Uganda	73.1% (19)	26.9% (7)
United States	70.0% (7)	30.0% (3)
United States (Hawaii)	100.0% (2)	0.0% (0)
United States (Puerto Rico)	50.0% (1)	50.0% (1)
Vietnam	66.7% (4)	33.3% (2)

Table showing the percentage of the quality classes for each harvest year:

Table 2: The Proportion of Quality Classs in Different Harvested Year

harvested	Good	Poor
2010	64.0% (16)	36.0% (9)
2011	72.0% (18)	28.0% (7)
2012	40.0% (96)	60.0% (144)
2013	51.7% (61)	48.3% (57)
2014	50.3% (94)	49.7% (93)
2015	54.6% (59)	45.4% (49)
2016	47.8% (43)	52.2% (47)
2017	44.4% (24)	55.6% (30)
2018	63.6% (7)	36.4% (4)

3 Formal Analsis

Model 1:

$$ln\left(\frac{p_{Poor}}{1-p_{Poor}}\right) = \alpha + \beta_1 \cdot \text{Country} + \beta_2 \cdot \text{Aroma} + \beta_3 \cdot \text{Flavor} + \beta_4 \cdot \text{Acidity} + \beta_5 \cdot \text{Category Two Defects} + \beta_6 \cdot \text{Harvested} + \beta_7 \cdot \text{Altitude}$$

Model 2:

$$ln\left(\frac{p_{Poor}}{1-p_{Poor}}\right) = \alpha + \beta_1 \cdot \text{Country} + \beta_2 \cdot \text{Aroma} + \beta_3 \cdot \text{Flavor} + \beta_4 \cdot \text{Acidity} + \beta_5 \cdot \text{Category Two Defects} + \beta_6 \cdot \text{Harvested}$$

Model 3:

$$ln\left(\frac{p_{poor}}{1-p_{poor}}\right) = \alpha + \beta_1 \cdot \text{Country of origin} + \beta_2 \cdot \text{aroma} + \beta_3 \cdot \text{flavor} + \beta_4 \cdot \text{acidity} + \beta_5 \cdot \text{category two defects}$$

Model 4:

$$ln\left(\frac{p_{Poor}}{1-p_{Poor}}\right) = \alpha + \beta_1 \cdot \text{Aroma} + \beta_2 \cdot \text{Flavor} + \beta_3 \cdot \text{Acidity} + \beta_4 \cdot \text{Category Two Defects}$$

Model 5:

$$ln\left(\frac{p_{Poor}}{1 - p_{Poor}}\right) = \alpha + \beta_1 \cdot \text{Aroma} + \beta_2 \cdot \text{Flavor} + \beta_3 \cdot \text{Acidity}$$

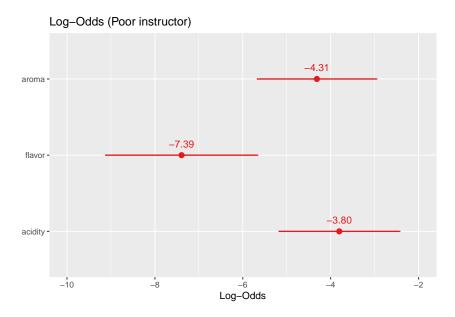
[1] "Good" "Poor"

3.1 Models comparison and Selection:

Table 3: The Result of Model comparison

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	Formula									
$Quality class \sim country_of_origin + aroma + flavor + acidity + category_two_defects + altitude_mean_meters + harvested$										
	Qualityclass ~ country_of_origin + aroma + flavor + acidity + category_two_defects + harvested									
	$Quality class \sim country_of_origin + aroma + flavor + acidity + category_two_defects$									
	$Quality class \sim aroma + flavor + acidity + category_two_defects$									
				Qua	alityclass	~ aroma + fla	avor + acidity			
	Rank	Df.res	AIC	AICc	BIC	McFadden	Cox.and.Snell	Nagelkerke	p.value	
	40	818	508.0	512.2	702.9	0.642	0.589	0.785	0	
	39	819	506.4	510.4	696.6	0.641	0.589	0.785	0	
	38	820	506.5	510.3	691.9	0.640	0.588	0.784	0	
	5	853	529.5	529.6	558.0	0.565	0.543	0.724	0	
	4	854	527.7	527.7	551.4	0.565	0.543	0.724	0	

$3.2 \log Odds$:



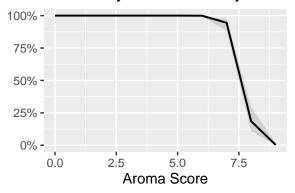
3.3 Confidence Intervals:

Table 4: Confidence Intervals for log odds in Model 5

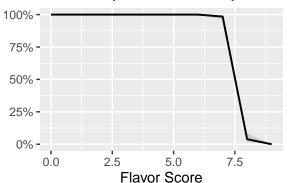
	2.5 %	97.5 %
(Intercept)	101.235732	134.953218
aroma	-5.720601	-2.988415
flavor	-9.197355	-5.721335
acidity	-5.211874	-2.449520

3.4 The Probability Plot:

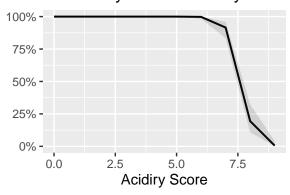
Probability of Poor Quality Class



Probability of Poor Quality Class



Probability of Poor Quality Class



4 Extend Analysis – Prediction Assessment.

4.1 Confusion Matrix

```
##
## Call:
  glm(formula = Qualityclass ~ aroma + flavor + acidity, family = binomial(link = "logit"),
##
      data = train_data)
##
## Deviance Residuals:
##
                 1Q
                      Median
## -3.2555
                      0.0017
                               0.3521
                                        3.9062
           -0.4244
##
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
##
## (Intercept) 115.7000
                            9.5036 12.174 < 2e-16 ***
                -3.8906
                            0.7683 -5.064 4.11e-07 ***
## aroma
## flavor
                -7.8288
                            1.0373 -7.547 4.44e-14 ***
                -3.5932
                            0.7963 -4.512 6.41e-06 ***
## acidity
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
```

```
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 950.62 on 685 degrees of freedom
## Residual deviance: 408.69 on 682 degrees of freedom
## AIC: 416.69
##
## Number of Fisher Scoring iterations: 7
## [1] "0" "1"
```

Table 5: Accuracy of Prediction.

	Value
Accuracy	0.8430233
Kappa	0.6868087
AccuracyLower	0.7798858
AccuracyUpper	0.8939306
AccuracyNull	0.5174419
AccuracyPValue	0.0000000
McnemarPValue	0.1236577

Table 6: The Resule of Sensitivity and Specificity of Prediction.

	Value
Sensitivity	0.7977528
Specificity	0.8915663
Pos Pred Value	0.8875000
Neg Pred Value	0.8043478
Precision	0.8875000
Recall	0.7977528
F1	0.8402367
Prevalence	0.5174419
Detection Rate	0.4127907
Detection Prevalence	0.4651163
Balanced Accuracy	0.8446595

Table 7: Confuse table.

	Actual Good	Actual Bad
0	74	18
1	9	71

4.2 ROC Curve

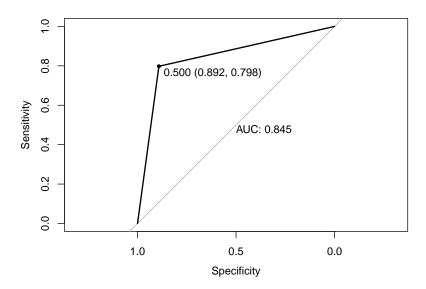


Figure 1: ROC cureve for model predicton

5 Conclusion