Wine White Quanlity Analysis by Zhenning Tan 4/6/2016

Basic statistics of the dataset

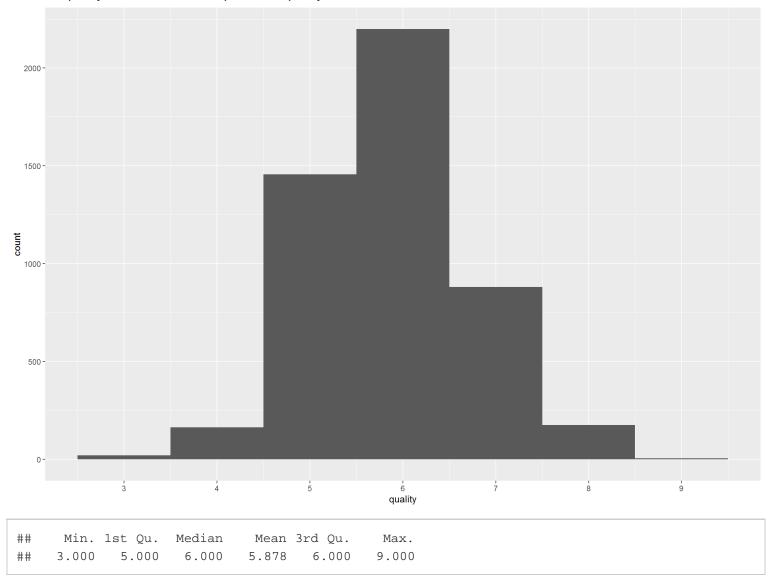
```
fixed.acidity
                               volatile.acidity citric.acid
  Min. : 1 Min. : 3.800 Min. :0.0800 Min. :0.0000
##
  1st Qu.:1225    1st Qu.: 6.300
                               1st Qu.:0.2100 1st Qu.:0.2700
  Median : 2450 Median : 6.800 Median : 0.2600 Median : 0.3200
  Mean :2450 Mean : 6.855
                             Mean :0.2782 Mean :0.3342
  3rd Qu.:3674 3rd Qu.: 7.300
                               3rd Qu.:0.3200 3rd Qu.:0.3900
  Max. :4898 Max. :14.200 Max. :1.1000 Max. :1.6600
  residual.sugar chlorides
                                free.sulfur.dioxide
  Min. : 0.600 Min. : 0.00900 Min. : 2.00
  1st Qu.: 1.700
                1st Qu.:0.03600 1st Qu.: 23.00
##
## Median : 5.200 Median : 0.04300 Median : 34.00
                Mean
  Mean : 6.391
                       :0.04577 Mean : 35.31
  3rd Qu.: 9.900 3rd Qu.:0.05000 3rd Qu.: 46.00
##
## Max. :65.800 Max. :0.34600 Max. :289.00
  total.sulfur.dioxide density
                                         рН
                                                  sulphates
  Min. : 9.0
                    Min. :0.9871 Min. :2.720 Min. :0.2200
                    1st Qu.:0.9917 1st Qu.:3.090
  1st Qu.:108.0
                                                  1st Qu.:0.4100
## Median :134.0
                    Median :0.9937 Median :3.180 Median :0.4700
                     Mean :0.9940 Mean :3.188
  Mean :138.4
                                                 Mean :0.4898
  3rd Qu.:167.0
                    3rd Qu.:0.9961 3rd Qu.:3.280 3rd Qu.:0.5500
##
  Max. :440.0
                    Max. :1.0390 Max. :3.820
                                                  Max. :1.0800
    alcohol
##
                  quality
  Min. : 8.00 Min. :3.000
##
  1st Qu.: 9.50
               1st Qu.:5.000
  Median :10.40 Median :6.000
  Mean :10.51 Mean :5.878
  3rd Qu.:11.40 3rd Qu.:6.000
## Max. :14.20
               Max. :9.000
```

```
## 'data.frame':
                   4898 obs. of 13 variables:
## $ X
                        : int 1 2 3 4 5 6 7 8 9 10 ...
                        : num 7 6.3 8.1 7.2 7.2 8.1 6.2 7 6.3 8.1 ...
## $ fixed.acidity
## $ volatile.acidity
                       : num 0.27 0.3 0.28 0.23 0.28 0.32 0.27 0.3 0.22 ...
  $ citric.acid
                       : num 0.36 0.34 0.4 0.32 0.32 0.4 0.16 0.36 0.34 0.43 ...
##
##
  $ residual.sugar
                       : num 20.7 1.6 6.9 8.5 8.5 6.9 7 20.7 1.6 1.5 ...
  $ chlorides
                         : num 0.045 0.049 0.05 0.058 0.058 0.05 0.045 0.045 0.049 0.044 .
##
  $ free.sulfur.dioxide : num 45 14 30 47 47 30 30 45 14 28 ...
##
## $ total.sulfur.dioxide: num 170 132 97 186 186 97 136 170 132 129 ...
  $ density
                       : num 1.001 0.994 0.995 0.996 0.996 ...
                         : num 3 3.3 3.26 3.19 3.19 3.26 3.18 3 3.3 3.22 ...
   Hq $
   $ sulphates
                       : num 0.45 0.49 0.44 0.4 0.4 0.44 0.47 0.45 0.49 0.45 ...
##
                         : num 8.8 9.5 10.1 9.9 9.9 10.1 9.6 8.8 9.5 11 ...
   $ alcohol
```

This dataset contains 4898 observations of 13 variables. The quality variable is integer. All other factors of the wine chemical properties are numeric type.

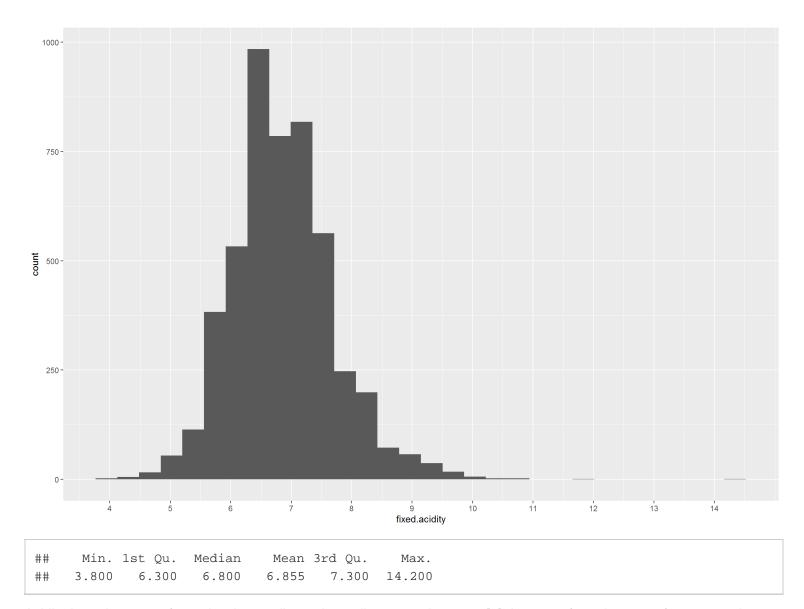
Univariate Plots Section

Wine quality should be the dependent variable in this dataset. My goal is to investigate how each chemicals in wine affect its quality. First, I'd like to explore the quality distribution.

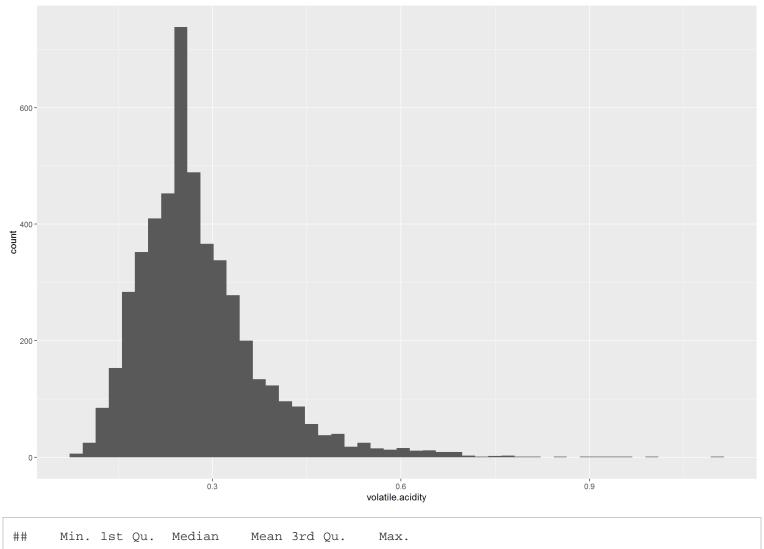


The quality distribution appears normal. The median quality is 6. The mean quality is 5.878.

Next, I'd like to explore the distribution of all the chemicals in white wine



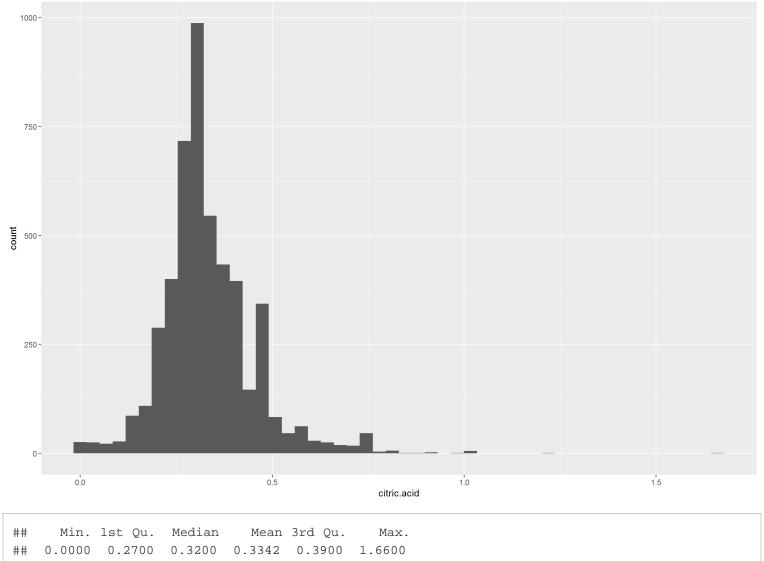
Acidity is an important factor in wine quality and contributes to wine taste [1]. It comes from the type of grapes and fermentation process. However, there's no direct correlation of fixed acidity to wine quality. In this plot, fixed acidity appears normal distribution. The median is 6.8. The mean is 6.855. However, there are some outliers with fixed acidity more than 10.



Min. 1st Qu. Median Mean 3rd Qu. Max.

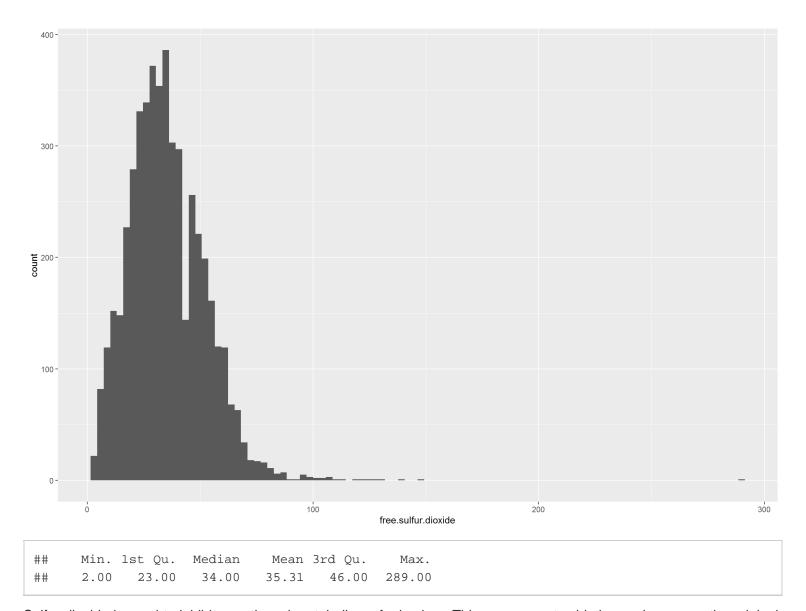
0.0800 0.2100 0.2600 0.2782 0.3200 1.1000

The volatitle acidity comes from acetic acid in wine, a byproduct of bacterial metabolism. The U.S. legal limits of volatile acidity for white table wine is 1.1 g/L [2]. In our dataset, most wine have a volatile acidity between 0.21 and 0.32 g/L. Low volatile acidity can reduce formation of other concomitant, sometimes unpleasant, aroma compounds.

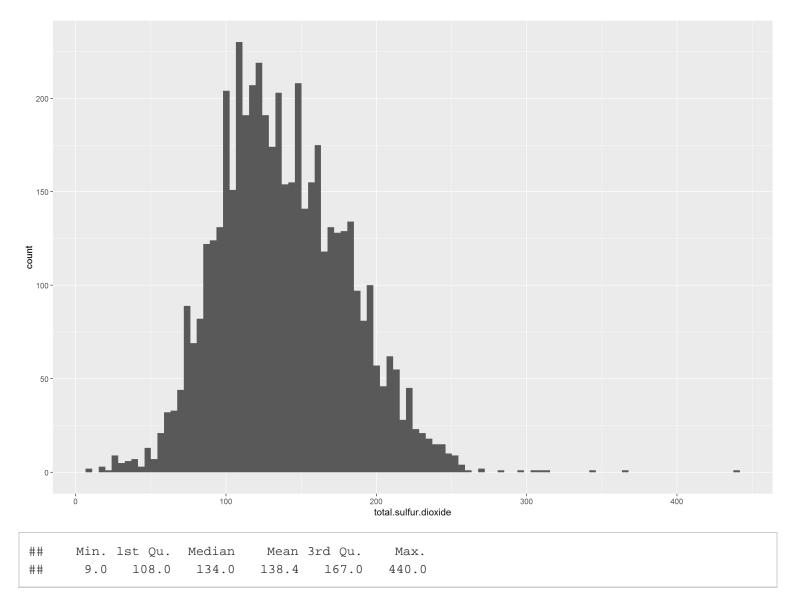


```
##
           TRUE
## FALSE
##
    4876
             22
```

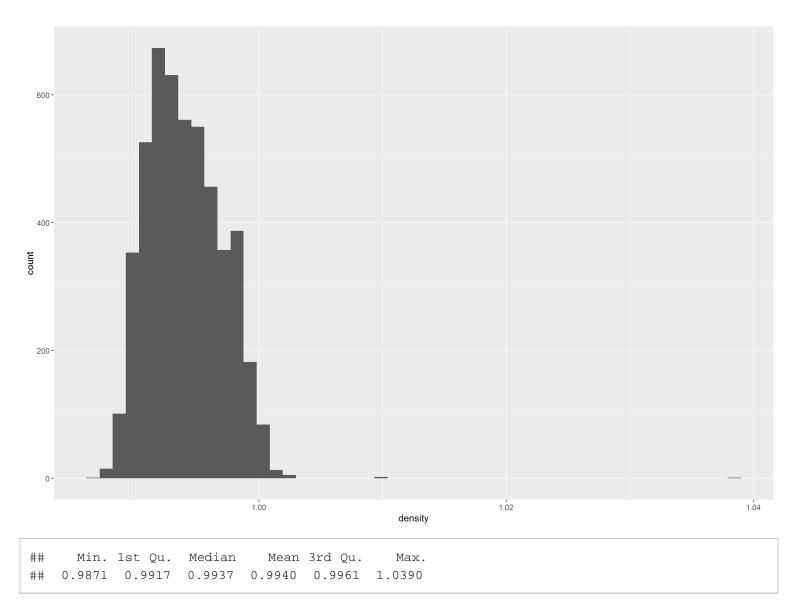
Citric acid contributes to the acidity of wine, adding "freshness" taste to wine. However, as a molecule in energy production reaction, citric acid can also lead to growth of microbes [3]. In our dataset, most wine have a citric acid amount between 0.27 and 0.39. There are 22 outliers with citric acid above 0.75 g/L.



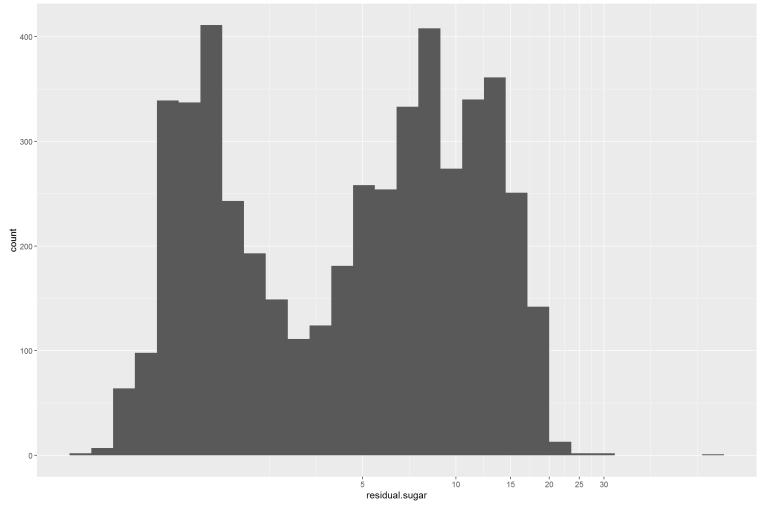
Sulfur dioxide is used to inhibit growth and metabolism of microbes. This can prevent oxidation and preserve the original fruity flavor and freshness taste [4]. In this dataset, most of the free sulfur dioxide is between 23 and 46 ppm. However, there are outliers with



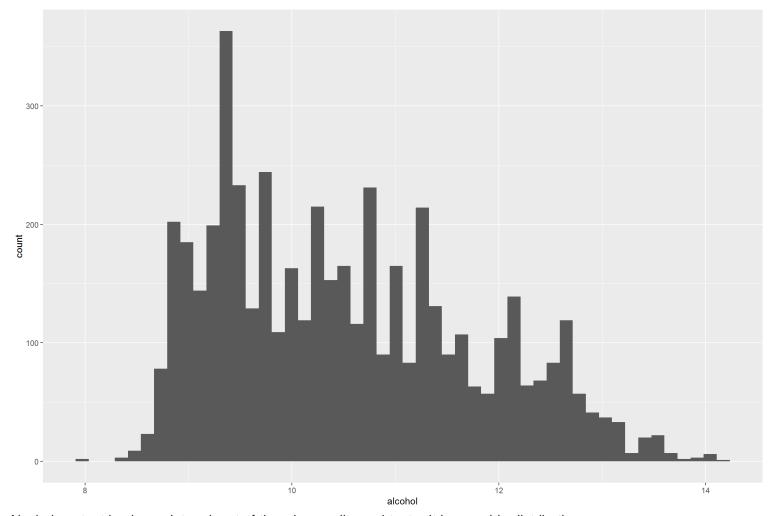
Most of the total sulfur dioxide is between 108 and 167 ppm. However, there are outliers with extreme total sulfur dioxide.



Most wines have density between 0.9917 and 0.9961. However, there are outliers with extreme density.



Residual sugar affects the sweetness of wine. Wines have different sweetness depending on the residual sugar. In the plot, I see that the residual sugar appears as bimodal distribution. This two peaks probably correponds to dry wine and sweet wine.



Alcohol content is also a determinant of the wine quality and taste. It has a wide distribution.

Univariate Analysis

What is the structure of your dataset?

This white wine quality dataset contains 4898 observations of 13 variables, which describes chemical properties of the wine, such as acidity, sugar, density, pH, etc.

What is/are the main feature(s) of interest in your dataset?

Most of features of the wine in the dataset have similar normal distribution as quality, except residual sugar and alcohol content. I need to run bivariate analysis to determine which features are related to wine quality.

What other features in the dataset do you think will help support your investigation into your feature(s) of interest?

At this step, I cannot determine the features that will contribute to the wine quality. Given the small number of variates, I will perform bivariate analysis on all the features and work on the pairs of features with high correlation coefficient.

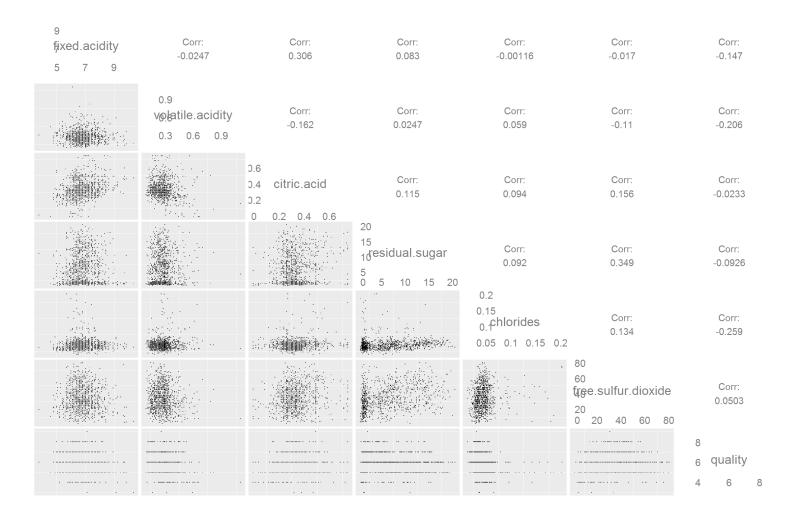
Did you create any new variables from existing variables in the dataset?

Later in the analysis, I transformed the numeric data type of quality to categorical type.

Of the features you investigated, were there any unusual distributions? Did you perform any operations on the data to tidy, adjust, or change the form of the data? If so, why did you do this?

Since the wine quality appears as normal distribution, I would guess that the factors that determines of wine quality will have similar normal distribution. I noticed that residual sugar distribution has two peaks. This is interesting and I will pay more attention in later analysis.

Bivariate Plots Section

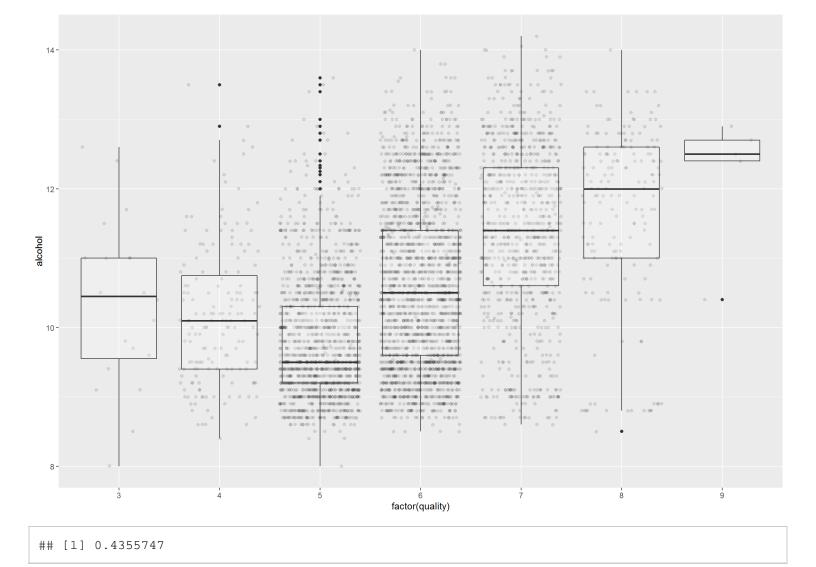


300					
200tal.sulfur.dioxide	Corr:	Corr:	Corr:	Corr:	Corr:
100	0.546	0.00647	0.136	-0.46	-0.167
0 100 200 300					
	1 1.996 density 1.992 1.988 0.992 0.996 1	Corr: -0.13	Corr: 0.0675	Corr: -0.807	Corr: -0.338
		3.6 3.3 pH 3 2.7 3 3.3 3.6	Corr: 0.182	Corr: 0.165	Corr: 0.129
			1 0.8 sulphates 0.6 0.4 0.6 0.8 1	Corr: -0.0117	Corr: 0.0748
				14 ¹² alcohol 10 12 14	Corr: 0.498
·					8
					₆ quality
					4 6 8

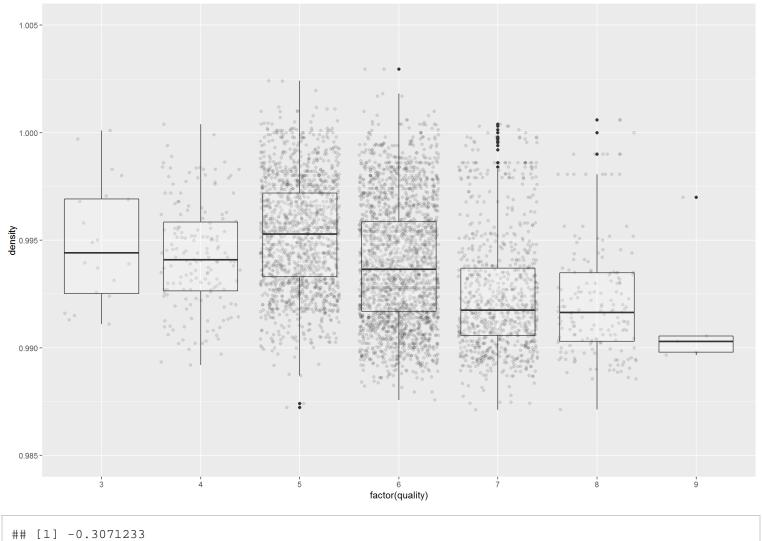
The above ggpair plot give clear visulation of labels by separating the features into two parts. However, in this case, I may miss interesting relationships between features. So I did the ggpair analysis on the whole dataset again to explore possible correlations between features.

9 ixed.acidity 5 7 9	Corr: -0.0261	Corr: 0.307	Corr: 0.0858	Corr: -0.00338	Corr: -0.0205	Corr: 0.0831	Corr: 0.278	Corr: -0.458	Corr: -0.0191	Corr: -0.165	Corr: -0.145
	0.9 ol at ile.acidi 0.3 0.6 0.9	Corr: -0.162	Corr: 0.0254	Corr: 0.0586	Corr: -0.11	Corr: 0.0952	Corr: -0.0282	Corr: -0.0261	Corr: -0.013	Corr: 0.0796	Corr: -0.205
).6). ø itric.acid).2 0 0 20 40 6	Corr: 0.116	Corr: 0.0937	Corr: 0.156	Corr: 0.153	Corr: 0.166	Corr: -0.222	Corr: 0.0277	Corr: -0.109	Corr: -0.0229
Angel Chinasa		24244	20 15 sigual.sug ð 5 10 15 20	0.0931	Corr: 0.349	Corr: 0.408	Corr: 0.827	Corr: -0.213	Corr: -0.0319	Corr: -0.452	Corr: -0.0938
		a seguina de la		0.2 0.15 6hlorides 0.050.10.150.2	Corr: 0.134	Corr: 0.244	Corr: 0.281	Corr: -0.0855	Corr: 0.014	Corr: -0.401	Corr: -0.258
				200	80 60 45ulfur.diox 3020 40 60 80	Corr: 0.616	Corr: 0.332	Corr: -0.0483	Corr: 0.0299	Corr: -0.244	Corr: 0.0511
					11800 °	250 200 1 50 llfur.diox 100 5010 0 520250	0 554	Corr: 0.00521	Corr: 0.13	Corr: -0.462	Corr: -0.155
					170		1 99 density 992 9 88 9 9 29961	Corr: -0.124	Corr: 0.0703	Corr: -0.81	Corr: -0.337
1006		W						3.6 3.3 pH 3.73 3.33.6	Corr: 0.183	Corr: 0.159	Corr: 0.128
									1 Supplates 0.40.60.8 1	Corr: -0.0116	Corr: 0.0749
						Signation of the second				14 åfcohol 10 12 14	Corr: 0.497
											8 c quality 4 6 8

Based on the bivariate plots, several pairs of factors are with relatively high correlation coefficient. Next, I examined each pair of features with correlation coefficients (above 0.3)

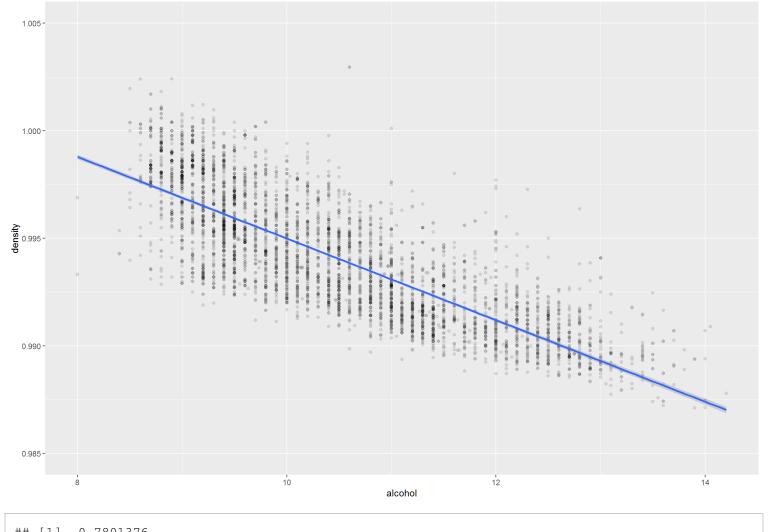


The overlay of boxplot and scatter plot clearly demonstrate the alcohol content for each quality. Generally higher quality wine has higher alcohol content.



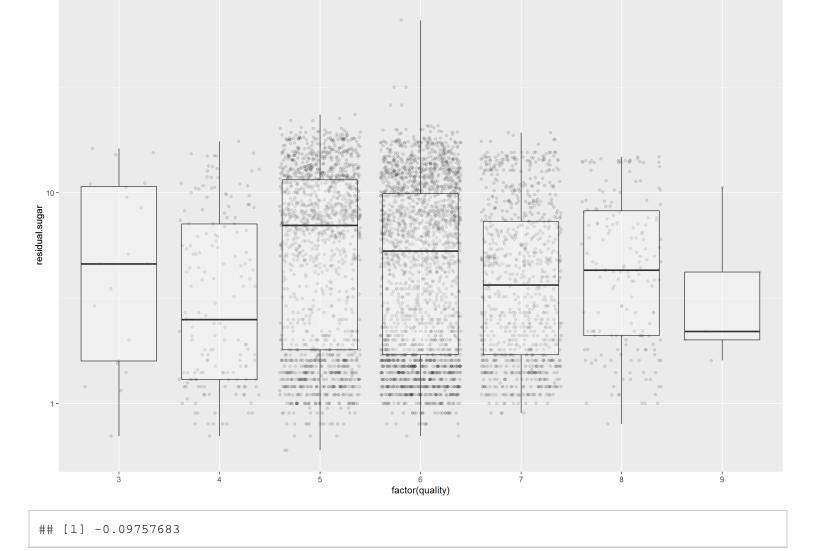
[1] -0.3071233

The overlay of boxplot and scatter plot show that wine with higher quality usually has lower density. This is consistent with alcohol content since alcohol density is lower than water.

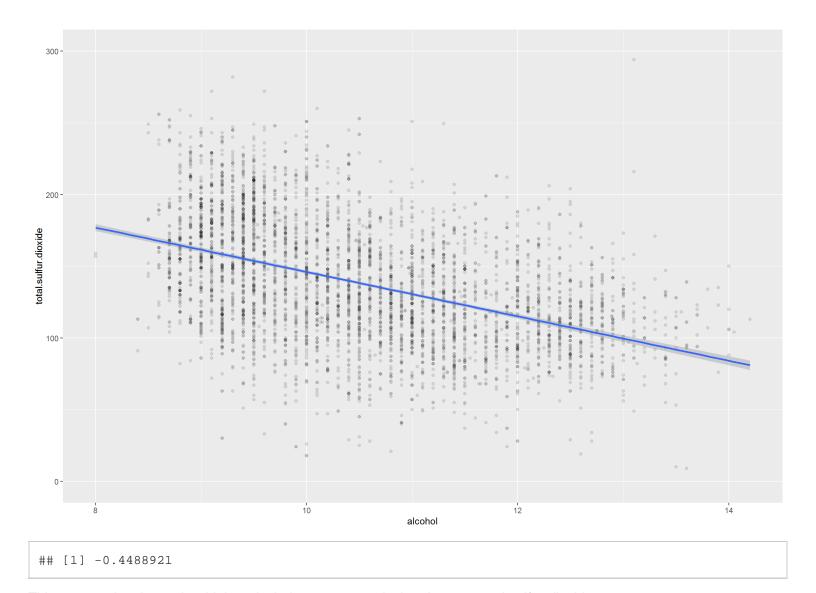


[1] -0.7801376

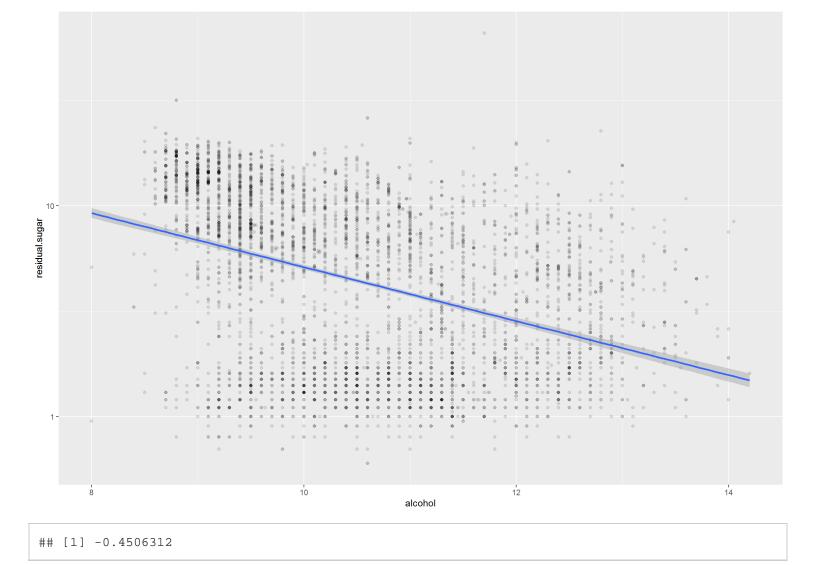
The above scatter plot shows the correlation between alcohol content and density. The trend line shows that with increasing alcohol content, wine density decreases.



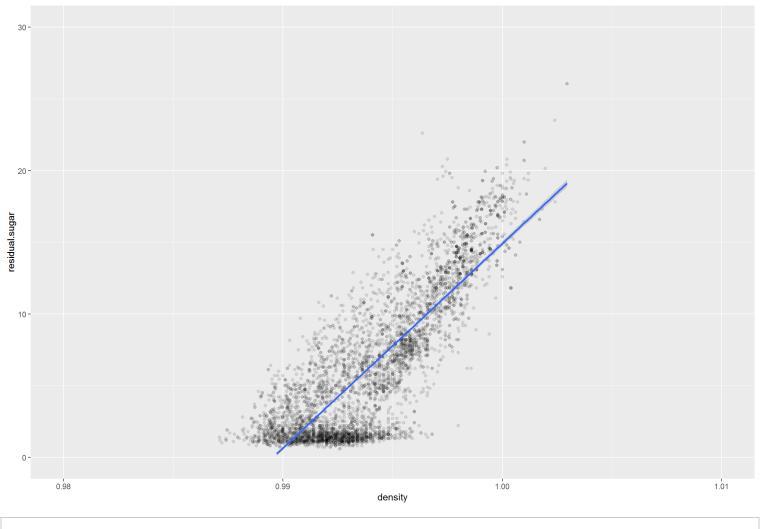
This scatter plot shows that for wine with quality 4 to 8, residual sugar appears to have bimodal distribution. For wine quality 3 and 9, the data points are too few to see the bimodal distribution. There's no strong correlation between wine quality and residual sugar



This scatter plot shows that higher alcohol content negatively relates to total sulfur dioxide.



This graph shows the negative correlation between alcohol and residual sugar. With increasing alcohol content, residual sugar decreases.



[1] 0.8389665

This graph shows strong correlation between density and residual sugar. Higher density wine has higher residual sugar amount.

Bivariate Analysis

Talk about some of the relationships you observed in this part of the investigation. How did the feature(s) of interest vary with other features in the dataset?

- 1. The alcohol content positively relates to wine quality, with a correlation coefficient of 0.436.
- 2. Wine quality is negatively correlated with density.
- 3. The above relationships are also reflected in the correlation between density and alcohol content.

Did you observe any interesting relationships between the other features (not the main feature(s) of interest)?

1. The white density and residual sugar amount has an interesting relationship. At density above 0.995, there's a strong positive correlation between density and residual sugar amount. However, for density below 0.995, the majority of wines have a similar level of residual sugar.

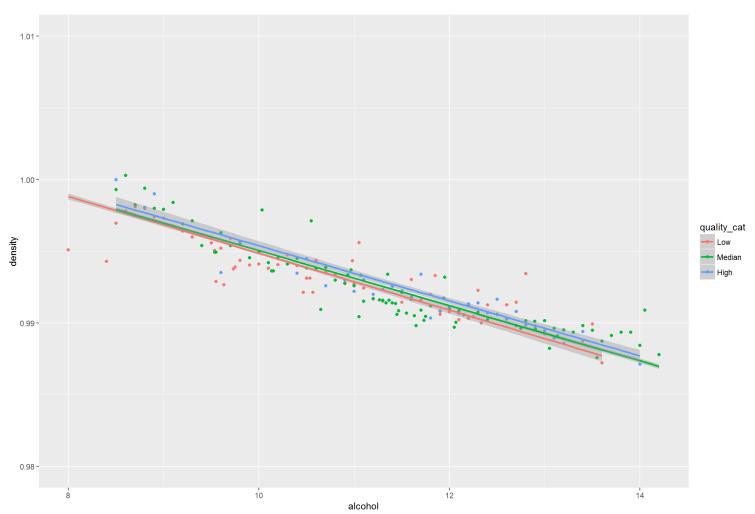
2. Alcohol and density is negatively correlated. This is due to the physical property of alcohol.

What was the strongest relationship you found?

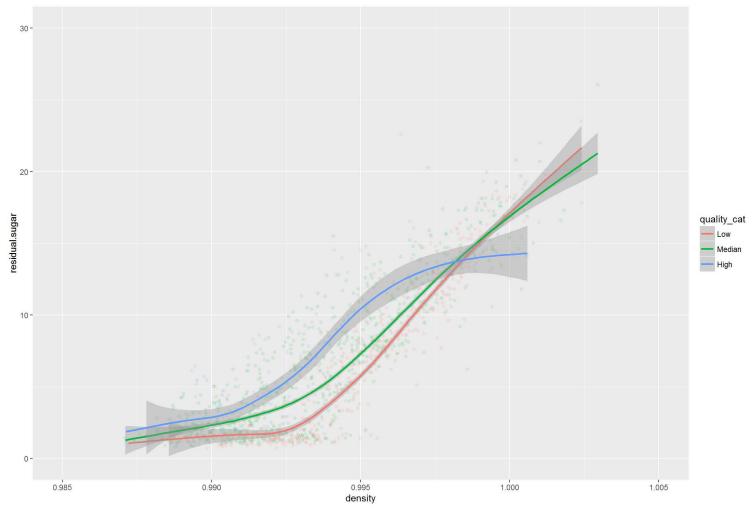
The strongest relationship is the negative correlation between alcohol and density.

Multivariate Plots Section

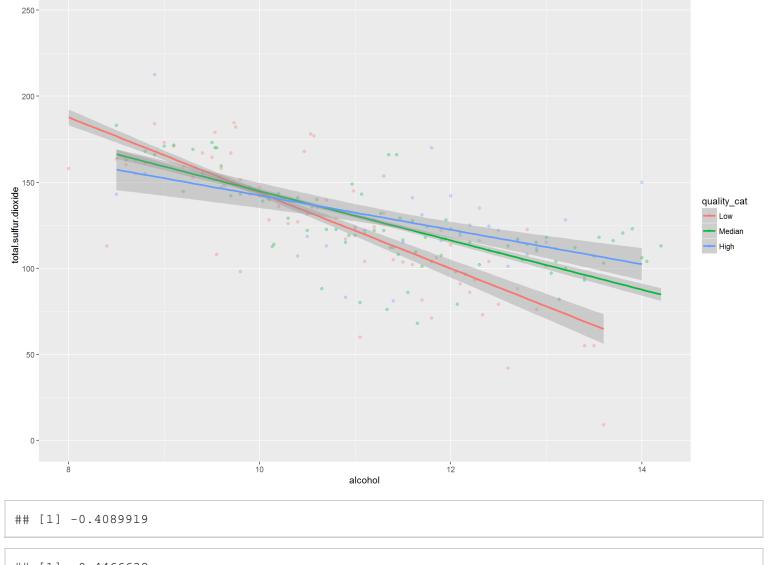
```
##
## Low Median High
## 1640 3078 180
```



The above graph shows that median desnity for different alcohol content for each quality of wine. Wine density shows a negative correlation with alcohol content, indepdent of wine quality.



The above graphs shows the trend of residual sugar vs density for different quality of wines. Overall, residual sugar increases with increasing of wine density. The residual sugar of high quality wines reaches plateau eventually. However, low and median quality wines keep having higher residual sugar with increasing of density.



[1] -0.4466632

[1] -0.3893316

The above graph shows the linear negative correlation of alcohol content and total sulfur dioxide. For all quality of wines, higher alcohol content correlates to lower total sulfur dioxide. Based on statistical analysis, median quality wines have the highest correlation coefficient.

Multivariate Analysis

Talk about some of the relationships you observed in this part of the investigation. Were there features that strengthened each other in terms of looking at your feature(s) of interest?

As discussed above, I observed that with increasing content of alcohol, the density decreases. This trend is independent of wine quality.

Different quality of wines have different relationship of residual sugar vs density.

Were there any interesting or surprising interactions between features?

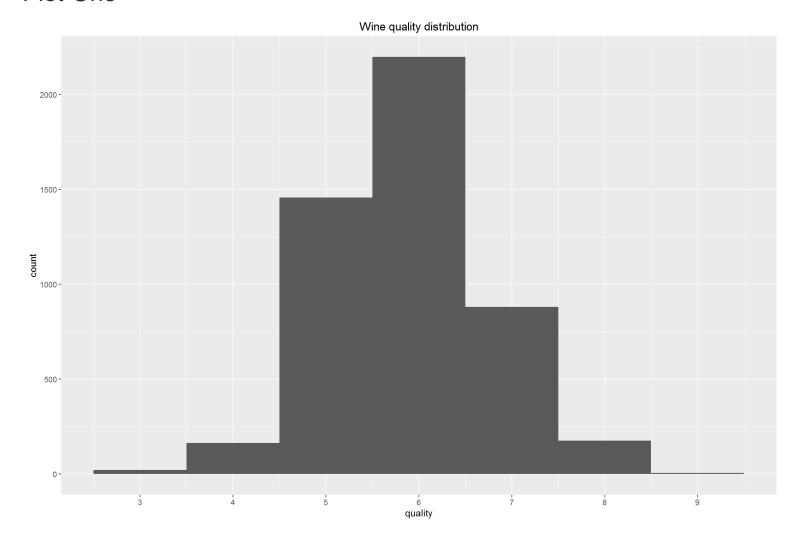
Due to the weak correlations between most of the features, I did not find strong interesting correlations between multiple variates. However, the different trend of residual sugar vs density for different quality of wines are interesting.

OPTIONAL: Did you create any models with your dataset? Discuss the strengths and limitations of your model.

I did not create any model.

Final Plots and Summary

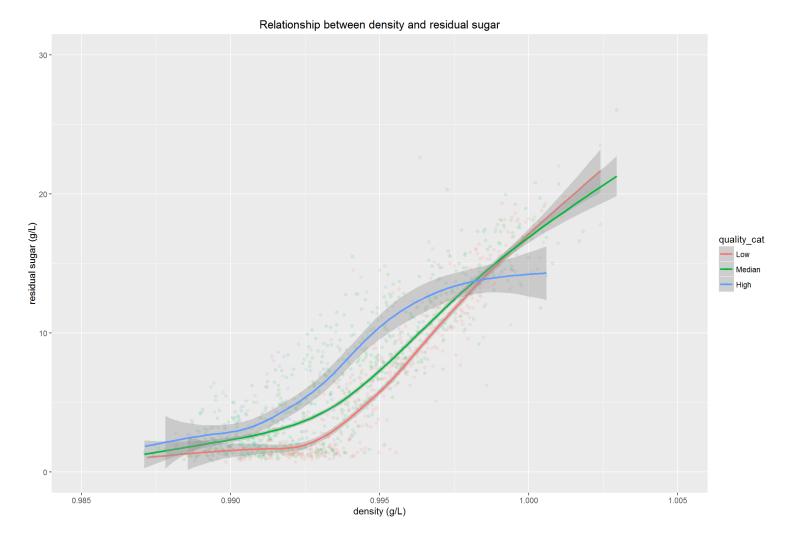
Plot One



Description One

Wine qualit appears to be normal distribution. Based on statistic analysis, the mean quality is 5.878. The median quality is 6.

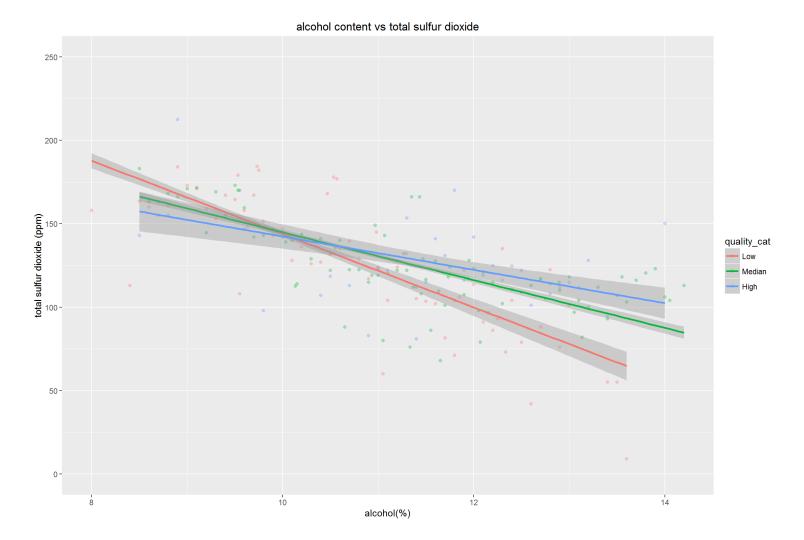
Plot Two



Description Two

The plot of residual sugar vs density shows different trend for different quality of wine. For low (quality between 3 and 5) and median (quality between 6 and 7) quality wines, residual sugar started to increase with density for density above 0.992. For high (quality between 8 and 9) quality wines, residual sugar started to increase at even lower density. However, it reaches plateau around density 0.998.

Plot Three



Description Three

There is a negative correlation between alcohol content and total sulfur dioxide. Different quality wines show slight different correlation coefficient. Based on statistialy analysis shown in the "multivariate plot section", median quality wine has the strongest correlation, while high quality wine shows a relative weaker correlation. Low quality wine has a steeper slope, indicating that change of alcohol content corresponds to a bigger change of total sulfur dioxide compared to median and high quality wines.

Reflection

This white wine dataset is relatively small, with only 15 variables. Initially, I wanted to explore features that affect wine quality. First, I looked at each individual feature distribution using histograms. Then I performed bivariate analysis to search for features that coorelate with wine quality. I found that acohol content and wine density have the highest correlation coefficient. In addition, I also analyzed other correlated features, including residual sugar content, density and pH. At last, I transformed the quality variable from integer to categorical variable to perform multi-variate analysis.

One of the struggles I had is how to find features that contribute to wine quality. Unlike the diamond dataset, after running the bivariate analysis, many features correlate with diamond price with correlation coefficient higher than 0.8. In this wine dataset, the correlation coefficient between chemical properties of wine and wine quality is small, less than 0.4. This indicates that it is hard to predict wine quality with current features accurately. It may require feature engineering to discover the deep relationship between features and wine quality. Given my limited knowledge of wine, currently I am not able to perform complicated feature engineering to come up a reliable model for prediction of wine quality based on

chemical properties.

Overall, the features in the dataset did not have strong correlation with wine quality. It is difficult to build up a prediction model with current features. In the future, it would be interesting to know more features about the wine to get better idea of wine quality, such as grape type, incubation time, manufacture and production location.