Part I

Project 1: Wine quality

Introduction

In the paragraphs to come we will discuss different approaches and models to be used in the dataset used in [1].

The following dataset consist of different a sample of wines with different characteristics and their relevant quality. The box below has an overview description of the dataset we will analyse. A further description of each variable can be found in Table 1.

```
>>> wine_data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1599 entries, 0 to 1598
Data columns (total 12 columns):
     Column
                            Non-Null Count
                                            Dtype
     ____
                            _____
     fixed acidity
                            1599 non-null
                                            float64
 0
 1
     volatile acidity
                           1599 non-null
                                            float64
 2
                                            float64
     citric acid
                           1599 non-null
 3
                                            float64
     residual sugar
                           1599 non-null
                                            float64
 4
     chlorides
                            1599 non-null
     free sulfur dioxide
 5
                                            float64
                           1599 non-null
                                            float64
 6
     total sulfur dioxide 1599 non-null
 7
                            1599 non-null
                                            float64
     density
 8
     рΗ
                            1599 non-null
                                            float64
 9
                            1599 non-null
                                            float64
     sulphates
 10
     alcohol
                            1599 non-null
                                            float64
11
     quality
                            1599 non-null
                                            int64
dtypes: float64(11), int64(1)
memory usage: 150.0 KB
>>> wine_data.describe().round(decimals=2).transpose()
                                        std
                                                             50%
                                                                    75%
                        count
                                mean
                                              min
                                                      25%
                                                                             max
fixed acidity
                      1599.0
                                8.32
                                       1.74 4.60
                                                     7.10
                                                            7.90
                                                                    9.20
                                                                           15.90
                                                            0.52
                                                                           1.58
volatile acidity
                                0.53
                                       0.18 0.12
                                                     0.39
                                                                    0.64
                      1599.0
citric acid
                                0.27
                                       0.19
                                             0.00
                                                            0.26
                                                                    0.42
                                                                            1.00
                      1599.0
                                                     0.09
residual sugar
                      1599.0
                                2.54
                                       1.41
                                             0.90
                                                     1.90
                                                            2.20
                                                                    2.60
                                                                           15.50
chlorides
                                                                            0.61
                      1599.0
                                0.09
                                       0.05
                                             0.01
                                                     0.07
                                                            0.08
                                                                   0.09
free sulfur dioxide
                      1599.0
                               15.87
                                      10.46
                                             1.00
                                                     7.00
                                                           14.00
                                                                  21.00
                                                                          72.00
total sulfur dioxide
                               46.47
                                      32.90
                                             6.00
                                                    22.00
                                                           38.00
                                                                  62.00
                                                                          289.00
                      1599.0
density
                      1599.0
                                1.00
                                       0.00 0.99
                                                     1.00
                                                            1.00
                                                                    1.00
                                                                            1.00
Нф
                      1599.0
                                3.31
                                       0.15 2.74
                                                     3.21
                                                            3.31
                                                                   3.40
                                                                            4.01
sulphates
                      1599.0
                                0.66
                                       0.17
                                             0.33
                                                     0.55
                                                            0.62
                                                                   0.73
                                                                            2.00
alcohol
                      1599.0
                              10.42
                                       1.07
                                             8.40
                                                     9.50
                                                           10.20
                                                                  11.10
                                                                           14.90
quality
                       1599.0
                                5.64
                                       0.81
                                             3.00
                                                     5.00
                                                            6.00
                                                                    6.00
                                                                            8.00
```

Exploratory analysis

To avoid unnecessary analysis, first we will perform a few checks to get a deeper understanding of the data we are using. To that end, we first check that the correlation structure amongst the variables (see fig. 1), including the quality of the wine. This should give us a general idea of how and if the variables are related to one another.

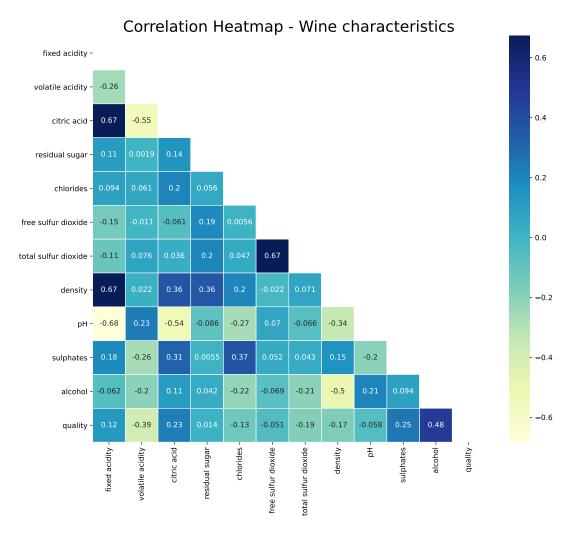


Figure 1: Correlation structure of wine characteristics.

There are a few things we could check whether the data in the data to assess whether what we are looking relates somehow to our prior knowledge about the topic. This prior knowledge might help us identify certain links, that might not be obvious if the data were not labeled.

At first glance, one can note that fixed acidity and citric acidity are strongly correlated (negatively) to the pH, although their correlation is not -1. This should relate to our prior knowledge, given that pH is directly related to the acidity of a solution. Additionally, we can see that alcohol correlates negatively with the density of the wine. This makes sense, given that the wine is a solution of different solutes, those in all likelihood are "heavier" than the alcohol solute. Hence, the more the alcohol content increases in the wine, the less dense it becomes.

Another interesting characteristic of the wine that is highly correlated to its quality is the alcohol content.

Another visual analysis we could perform is to look at the KDEs (Kernel Density Estimates). These we could imagine as a cross-sectional cut in a bi-variate probability distribution. Figure 2 shows us that even though the wines in the sample range from quality 3 to 8 (see subplot in row 3, column 4), it would seem that there are mostly 4 important groups. The earlier stated fact, ad priori, gives us an interesting thing we should have in mind when trying to fit any kind of model. I.e., the tails of the quality (grades 3 and 8) will be underrepresented, then most model we could think of fitting will have trouble predicting a grade close to 3 and 8 and beyond (to each direction).

KDE plots - Quality against each feature

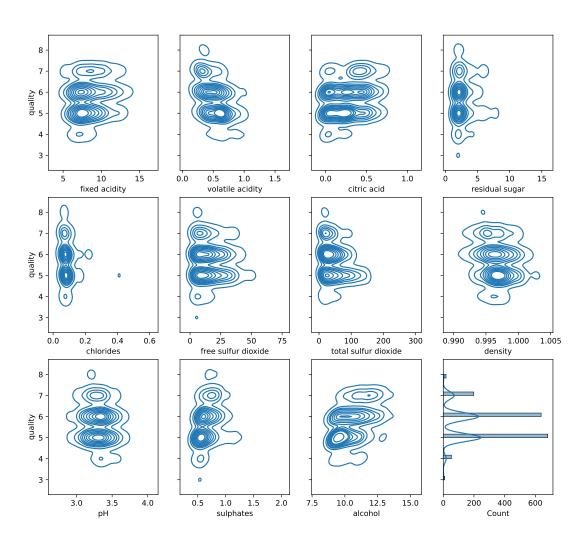


Figure 2: KDEs for wine features.

Conclusion

[1] All these characteristics are important in a wine. We used all these statistical models to try to understand to which degree they are important to determine a wine's quality. These analyses might be really important for a winemaker to know. Given that by affecting the inherent characteristics of the wine will most likely have an impact on the quality of that wine.

Part II Project 2: Food Preferences

Part III

Project 3: Store Sales

References

[1] P. Cortez, A. Cerdeira, F. Almeida, T. Matos and J. Reis. Modeling wine preferences by data mining from physicochemical properties. In Decision Support Systems, Elsevier, 47(4):547-553, 2009.

Table 1: Description of wine characteristics.

| Characteristic Description | Description |
|----------------------------|---|
| fixed acidity | most acids involved with wine or fixed or nonvolatile (do not evaporate readily). |
| volatile acidity | the amount of acetic acid in wine, which at too high of levels can lead to an unpleasant, vinegar taste. |
| citric acidity | found in small quantities, citric acid can add "freshness" and flavor to wines. |
| residual sugar | the amount of sugar remaining after fermentation stops, it's rare to find wines with less than 1 gram/liter and wines with greater than 45 grams/liter are considered sweet. |
| chlorides | the amount of salt in the wine. |
| free sulfur dioxide | free sulfur dioxide the free form of SO2 exists in equilibrium between molecular SO2 (as a dissolved gas) and bi-sulfate ion; it prevents microbial growth and the oxidation of wine. |
| total sulfur dioxide | total sulfur dioxide amount of free and bound forms of S02; in low concentrations, SO2 is mostly undetectable in wine, but at free SO2 concentrations over 50 ppm, SO2 becomes evident in the nose and taste of wine. |
| density | the density of water is close to that of water depending on the percent alcohol and sugar content. |
| $_{ m Hd}$ | describes how acidic or basic a wine is on a scale from 0 (very acidic) to 14 (very basic); most wines are between 3-4 on the pH scale. |
| sulphates | a wine additive which can contribute to sulfur dioxide gas (802) levels, which acts as an antimicrobial and antioxidant. |
| alcohol | the percent alcohol content of the wine. |
| quality | output variable (based on sensory data, score between 0 and 10). |