

The two datasets are related to red and white variants of the Portuguese "Vinho Verde" wine. Due to privacy and logistic issues, only physicochemical (inputs) and sensory (the output) variables are available (e.g. there is no data about grape types, wine brand, wine selling price, etc.). Два набора данных относятся к красному и белому вариантам португальского вина «Виньо Верде». Из соображений конфиденциальности и логистики доступны только физико-химические (входные) и сенсорные (выходные) переменные (например, нет данных о сортах винограда, марке вина, отпускной цене вина и т. д.).

Input variables (based on physicochemical tests):

- 1 - fixed acidity (фиксированная кислотность)
- 2 - volatile acidity (летучая кислотность)
- 3 - citric acid (лимонная кислота)
- 4 - residual sugar (остаточный сахар)
- 5 - chlorides (хлориды)
- 6 - free sulfur dioxide (свободный диоксид серы)
- 7 - total sulfur dioxide (общий диоксид серы)
- 8 - density (плотность)
- 9 - pH (водородный показатель)
- 10 - sulphates (сульфаты)
- 11 - alcohol (алкоголь)

Output variable (based on sensory data):

- 12 - quality (score between 0 and 10) качество (оценка от 0 до 10)

```
] import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
```

```
] dataset = pd.read_csv('/kaggle/input/red-wine-quality-cortez-et-al-2009/winequality-red.csv')
```

```
] dataset.head()
```

```
]
fixed acidity  volatile acidity  citric acid  residual sugar  chlorides  free sulfur dioxide  total sulfur dioxide  density  pH  sulphates  alcohol  quality
0           7.4             0.70         0.00           1.9      0.076             11.0             34.0  0.9978  3.51         0.56         9.4         5
1           7.8             0.88         0.00           2.6      0.098             25.0             67.0  0.9968  3.20         0.68         9.8         5
2           7.8             0.76         0.04           2.3      0.092             15.0             54.0  0.9970  3.26         0.65         9.8         5
3          11.2             0.28         0.56           1.9      0.075             17.0             60.0  0.9980  3.16         0.58         9.8         6
4           7.4             0.70         0.00           1.9      0.076             11.0             34.0  0.9978  3.51         0.56         9.4         5
```

```
] dataset.shape
```

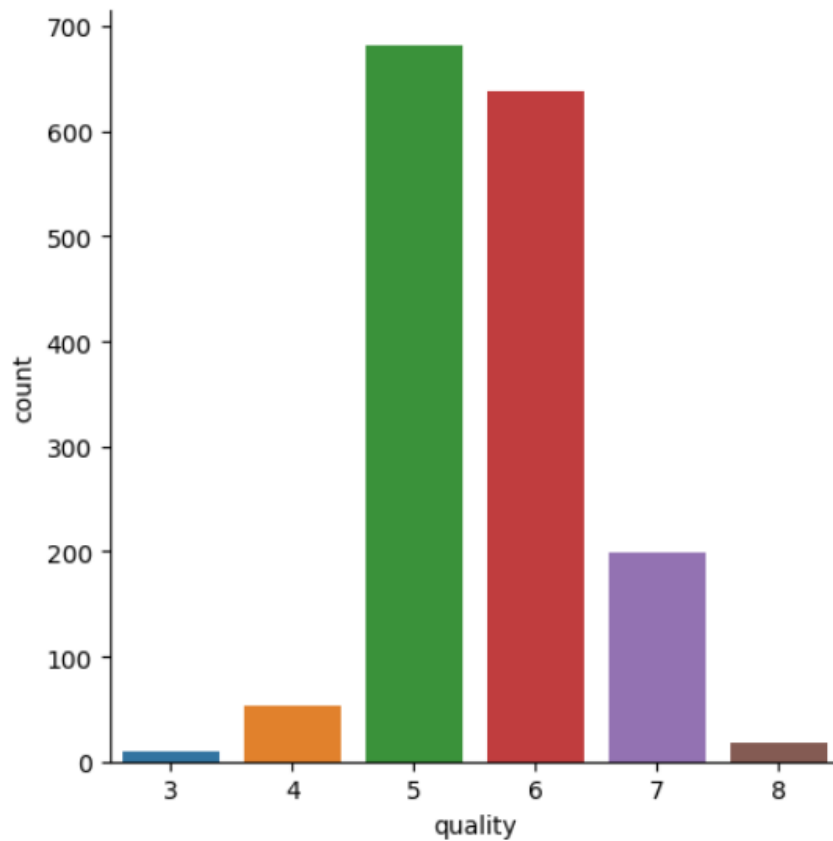
```
] (1599, 12)
```

```
[5]: dataset.isnull().sum()
```

```
[5]: fixed acidity      0  
volatile acidity      0  
citric acid           0  
residual sugar        0  
chlorides             0  
free sulfur dioxide   0  
total sulfur dioxide  0  
density               0  
pH                   0  
sulphates             0  
alcohol               0  
quality               0  
dtype: int64
```

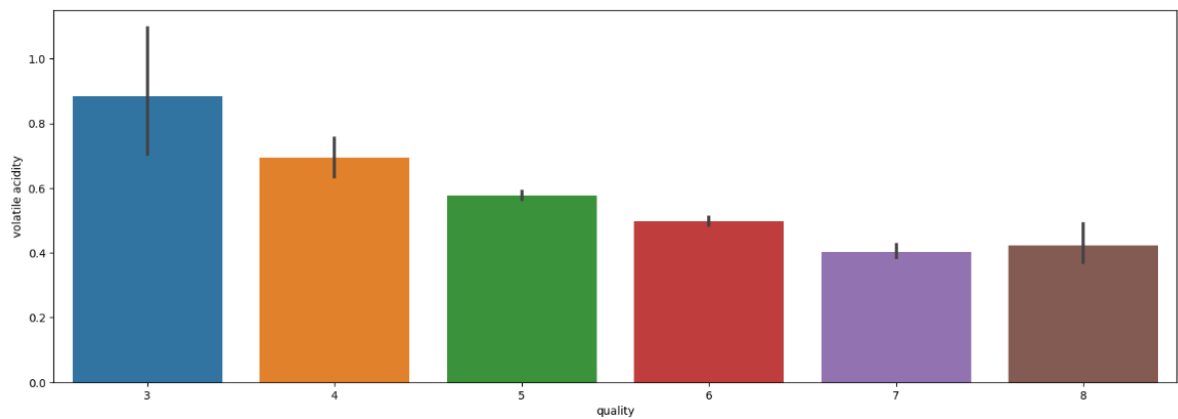
```
[6]: sns.catplot(x='quality', data = dataset, kind = 'count')
```

```
[6]: <seaborn.axisgrid.FacetGrid at 0x7ea366d11b10>
```



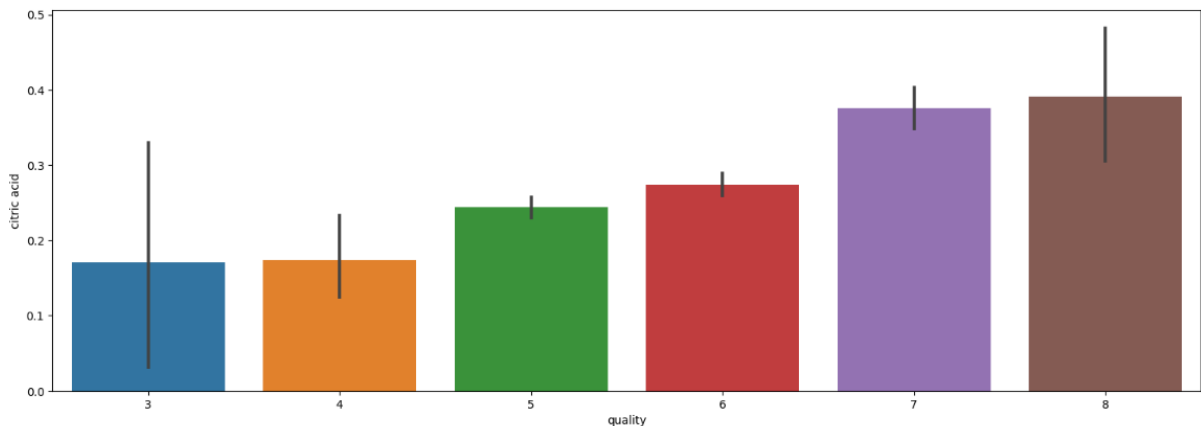
```
[7]: # volatile acidity vs Quality  
plot = plt.figure(figsize=(18,6))  
sns.barplot(x='quality', y = 'volatile acidity', data = dataset)
```

```
[7]: <Axes: xlabel='quality', ylabel='volatile acidity'>
```



```
[8]: # citric acid vs Quality
plot = plt.figure(figsize=(18,6))
sns.barplot(x='quality', y = 'citric acid', data = dataset)
```

```
[8]: <Axes: xlabel='quality', ylabel='citric acid'>
```



```
[9]: dataset.describe()
```

| | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | pH | sulphates | alcohol | |
|-------|---------------|------------------|-------------|----------------|-------------|---------------------|----------------------|-------------|-------------|-------------|-------------|------|
| count | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599.000000 | 1599 |
| mean | 8.319637 | 0.527821 | 0.270976 | 2.538806 | 0.087467 | 15.874922 | 46.467792 | 0.996747 | 3.311113 | 0.658149 | 10.422983 | 5 |
| std | 1.741096 | 0.179060 | 0.194801 | 1.409928 | 0.047065 | 10.460157 | 32.895324 | 0.001887 | 0.154386 | 0.169507 | 1.065668 | 0 |
| min | 4.600000 | 0.120000 | 0.000000 | 0.900000 | 0.012000 | 1.000000 | 6.000000 | 0.990070 | 2.740000 | 0.330000 | 8.400000 | 3 |
| 25% | 7.100000 | 0.390000 | 0.090000 | 1.900000 | 0.070000 | 7.000000 | 22.000000 | 0.995600 | 3.210000 | 0.550000 | 9.500000 | 5 |
| 50% | 7.900000 | 0.520000 | 0.260000 | 2.200000 | 0.079000 | 14.000000 | 38.000000 | 0.996750 | 3.310000 | 0.620000 | 10.200000 | 6 |
| 75% | 9.200000 | 0.640000 | 0.420000 | 2.600000 | 0.090000 | 21.000000 | 62.000000 | 0.997835 | 3.400000 | 0.730000 | 11.100000 | 6 |
| max | 15.900000 | 1.580000 | 1.000000 | 15.500000 | 0.611000 | 72.000000 | 289.000000 | 1.003690 | 4.010000 | 2.000000 | 14.900000 | 8 |

```
[10]: corr = dataset.corr()
corr
```

| | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | pH | sulphates | alcohol | quality |
|----------------------|---------------|------------------|-------------|----------------|-----------|---------------------|----------------------|-----------|-----------|-----------|-----------|-----------|
| fixed acidity | 1.000000 | -0.256131 | 0.671703 | 0.114777 | 0.093705 | -0.153794 | -0.113181 | 0.668047 | -0.682978 | 0.183006 | -0.061668 | 0.124052 |
| volatile acidity | -0.256131 | 1.000000 | -0.552496 | 0.001918 | 0.061298 | -0.010504 | 0.076470 | 0.022026 | 0.234937 | -0.260987 | -0.202288 | -0.390558 |
| citric acid | 0.671703 | -0.552496 | 1.000000 | 0.143577 | 0.203823 | -0.060978 | 0.035533 | 0.364947 | -0.541904 | 0.312770 | 0.109903 | 0.226373 |
| residual sugar | 0.114777 | 0.001918 | 0.143577 | 1.000000 | 0.055610 | 0.187049 | 0.203028 | 0.355283 | -0.085652 | 0.005527 | 0.042075 | 0.013732 |
| chlorides | 0.093705 | 0.061298 | 0.203823 | 0.055610 | 1.000000 | 0.005562 | 0.047400 | 0.200632 | -0.265026 | 0.371260 | -0.221141 | -0.128907 |
| free sulfur dioxide | -0.153794 | -0.010504 | -0.060978 | 0.187049 | 0.005562 | 1.000000 | 0.667666 | -0.021946 | 0.070377 | 0.051658 | -0.069408 | -0.050656 |
| total sulfur dioxide | -0.113181 | 0.076470 | 0.035533 | 0.203028 | 0.047400 | 0.667666 | 1.000000 | 0.071269 | -0.066495 | 0.042947 | -0.205654 | -0.185100 |
| density | 0.668047 | 0.022026 | 0.364947 | 0.355283 | 0.200632 | -0.021946 | 0.071269 | 1.000000 | -0.341699 | 0.148506 | -0.496180 | -0.174919 |
| pH | -0.682978 | 0.234937 | -0.541904 | -0.085652 | -0.265026 | 0.070377 | -0.066495 | -0.341699 | 1.000000 | -0.196648 | 0.205633 | -0.057731 |
| sulphates | 0.183006 | -0.260987 | 0.312770 | 0.005527 | 0.371260 | 0.051658 | 0.042947 | 0.148506 | -0.196648 | 1.000000 | 0.093595 | 0.251397 |
| alcohol | -0.061668 | -0.202288 | 0.109903 | 0.042075 | -0.221141 | -0.069408 | -0.205654 | -0.496180 | 0.205633 | 0.093595 | 1.000000 | 0.476166 |
| quality | 0.124052 | -0.390558 | 0.226373 | 0.013732 | -0.128907 | -0.050656 | -0.185100 | -0.174919 | -0.057731 | 0.251397 | 0.476166 | 1.000000 |

```
[ ]: plt.figure(figsize=(18,6))
      sns.heatmap(corr, cbar=True, square=True, fmt = '.1f', annot = True, annot_kws={'size':8}, cmap = 'Blues')
```

[11]: <Axes: >

