oliveOilAnalysis-python

February 6, 2024

0.1 Required Packages

/var/folders/49/qhkpplds6xxghpsg8cyk9crc0000gn/T/ipykernel_17677/243197114.py:2: DeprecationWarning:

Pyarrow will become a required dependency of pandas in the next major release of pandas (pandas 3.0),

(to allow more performant data types, such as the Arrow string type, and better interoperability with other libraries)

but was not found to be installed on your system.

If this would cause problems for you,

please provide us feedback at https://github.com/pandas-dev/pandas/issues/54466

import pandas as pd

0.2 Load the Data

```
[]: # Loading into Pandas DataFrame
    oo_sensory = pd.read_csv("olive_oil_sensory.csv")

[]: # Loading into Pandas DataFrame
    oo_chemical = pd.read_csv("olive_oil_chemical.csv")
```

0.3 Combining Data

```
[]: combined = pd.merge(oo_sensory, oo_chemical, how='left', on='region') combined.head(16)
```

```
[]: region s_yellow s_green s_brown s_glossy s_transp s_syrup c_Acidity \ 0 G1 21.4 73.4 10.1 79.7 75.2 50.3 0.73
```

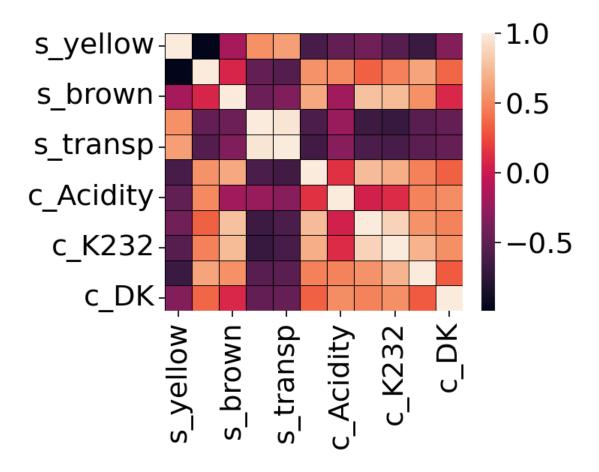
```
9.8
                                                77.8
                                                                                  0.19
1
       G2
                23.4
                          66.3
                                                           68.7
                                                                     51.7
2
       GЗ
                32.7
                          53.5
                                      8.7
                                                82.3
                                                           83.2
                                                                     45.4
                                                                                  0.26
3
                                                                     47.8
       G4
                30.2
                          58.3
                                     12.2
                                                81.1
                                                           77.1
                                                                                  0.67
                                                           65.3
4
       G5
                51.8
                          32.5
                                      8.0
                                                72.4
                                                                     46.5
                                                                                  0.52
5
       Ι1
                40.7
                          42.9
                                     20.1
                                                67.7
                                                           63.5
                                                                     52.2
                                                                                  0.26
6
       12
                          30.4
                                     11.5
                                                77.8
                                                           77.3
                                                                     45.2
                                                                                  0.24
                53.8
7
       Ι3
                26.4
                          66.5
                                     14.2
                                                78.7
                                                           74.6
                                                                     51.8
                                                                                  0.30
8
       14
                                     10.3
                                                81.6
                                                           79.6
                                                                     48.3
                                                                                  0.35
                65.7
                          12.1
9
                                                                     52.8
       Ι5
                45.0
                          31.9
                                     28.4
                                                75.7
                                                           72.9
                                                                                  0.19
10
       S1
                70.9
                          12.2
                                     10.8
                                                87.7
                                                           88.1
                                                                     44.5
                                                                                  0.15
                                      8.3
                                                           89.7
                                                                     42.3
11
       S2
                73.5
                            9.7
                                                89.9
                                                                                  0.16
12
       S3
                68.1
                          12.0
                                     10.8
                                                78.4
                                                           75.1
                                                                     46.4
                                                                                  0.27
13
       S4
                67.6
                          13.9
                                     11.9
                                                84.6
                                                           83.8
                                                                     48.5
                                                                                  0.16
                                                88.1
                                                           88.5
                                                                     46.7
                                                                                  0.24
14
       S5
                71.4
                          10.6
                                     10.8
15
       S6
                71.4
                          10.0
                                     11.4
                                                89.5
                                                           88.5
                                                                     47.2
                                                                                  0.30
```

```
c_K232
   c_Peroxide
                        c_K270
                                 c_DK
0
         12.70
                 1.900
                       0.1390 0.003
1
         12.30
                 1.678 0.1160 -0.004
2
         10.30
                 1.629
                       0.1160 -0.005
3
         13.70
                 1.701
                       0.1680 -0.002
                       0.1190 -0.001
4
         11.20
                 1.539
5
         18.70
                 2.117 0.1420 0.001
6
         15.30
                 1.891
                       0.1160 0.000
7
         18.50
                 1.908 0.1250 0.001
8
         15.60
                 1.824 0.1040 0.000
                 2.222 0.1580 -0.003
9
         19.40
10
         10.50
                 1.522 0.1160 -0.004
                 1.527
                        0.1063 -0.002
11
          8.14
12
         12.50
                 1.555 0.0930 -0.002
13
         11.00
                 1.573
                       0.0940 -0.003
14
         10.80
                 1.331
                        0.0850 -0.003
                 1.415
15
         11.40
                       0.0930 -0.004
```

0.4 Delete the first Column

```
[]: numeric = combined.drop('region', axis=1)
    cMatrix = numeric.corr()
    sb.heatmap(cMatrix, linewidths=0.5, linecolor="black")
```

[]: <Axes: >



0.5 Describe

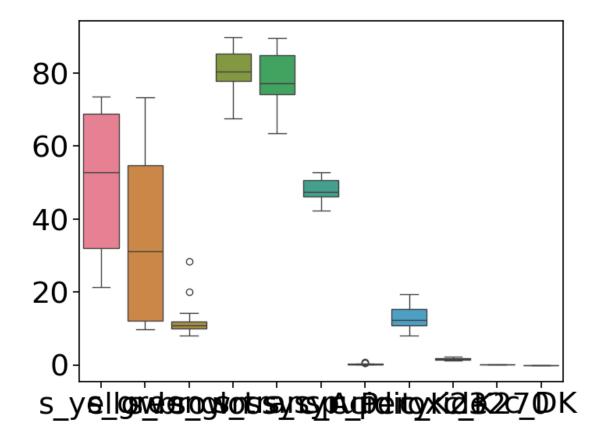
]: numeric.des	describe().T							
]:	count	mean	std	min	25%	50%	75%	\
s_yellow	16.0	50.875000	19.458623	21.400	32.07500	52.8000	68.80000	
s_green	16.0	33.512500	23.486986	9.700	12.07500	31.1500	54.70000	
s_brown	16.0	12.331250	5.128706	8.000	10.02500	10.8000	11.97500	
s_glossy	16.0	80.812500	6.188040	67.700	77.80000	80.4000	85.37500	
s_transp	16.0	78.193750	8.307384	63.500	74.17500	77.2000	84.87500	
s_syrup	16.0	47.975000	3.065398	42.300	46.15000	47.5000	50.65000	
$c_Acidity$	16.0	0.311875	0.176568	0.150	0.19000	0.2600	0.31250	
<pre>c_Peroxide</pre>	16.0	13.252500	3.345141	8.140	10.95000	12.4000	15.37500	
c_K232	16.0	1.708250	0.248731	1.331	1.53600	1.6535	1.89325	
c_K270	16.0	0.118144	0.023707	0.085	0.10150	0.1160	0.12850	
c_DK	16.0	-0.001750	0.002236	-0.005	-0.00325	-0.0020	0.00000	
	max							
s_yellow	73.500							

```
s_green
            73.400
s_brown
            28.400
            89.900
s_glossy
            89.700
s_transp
s_syrup
            52.800
c_Acidity
             0.730
c_Peroxide
            19.400
c_K232
             2.222
c_K270
             0.168
c_DK
             0.003
```

0.6 Boxplot

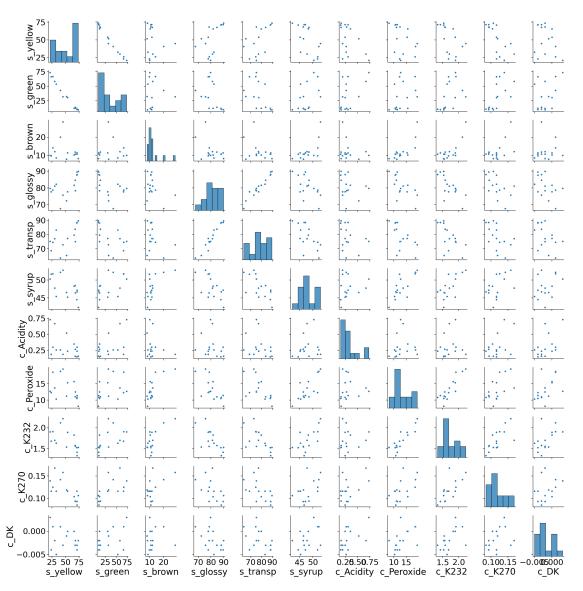
```
[]: plt.figure(figsize=(8, 6)) sb.boxplot(data=numeric)
```

[]: <Axes: >



0.7 Pairplot

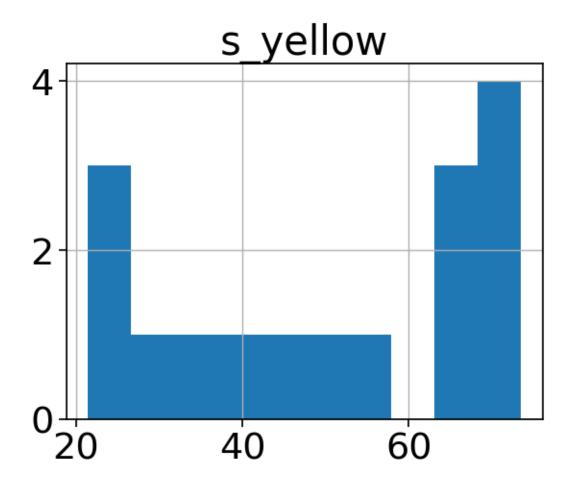
- []: sb.pairplot(numeric)
- []: <seaborn.axisgrid.PairGrid at 0x13c2826f0>



0.8 Histogram

```
[]: numeric.hist("s_yellow")
```

[]: array([[<Axes: title={'center': 's_yellow'}>]], dtype=object)



0.9 Mean, Median, Diff

```
[]: means = numeric.mean()
    print(means)
    s_yellow
                  50.875000
    s_green
                  33.512500
    s_brown
                   12.331250
    s_glossy
                  80.812500
    s\_transp
                  78.193750
                  47.975000
    s_syrup
    c_Acidity
                    0.311875
    c\_Peroxide
                   13.252500
    c_K232
                    1.708250
    c_K270
                    0.118144
    c_DK
                   -0.001750
    dtype: float64
```

```
[]: median = numeric.median()
    print(median)
    s_yellow
                  52.8000
    s_green
                  31.1500
    s_brown
                  10.8000
    s_glossy
                  80.4000
    s_{transp}
                  77.2000
                  47.5000
    s_syrup
    c_Acidity
                   0.2600
    c_Peroxide
                  12.4000
    c_K232
                   1.6535
    c_K270
                   0.1160
                  -0.0020
    c_DK
    dtype: float64
[]: for i in range(11):
         print(means.iloc[i] - median.iloc[i])
```

- -1.924999999999972
- 2.3625000000000043
- 1.53125
- 0.4124999999999943
- 0.9937500000000057
- 0.4750000000000014
- 0.051875000000000004
- 0.8525000000000009
- 0.054750000000000076
- 0.002143749999999986
- 0.00025