Exercise 4: How many clusters of grain?

In the video, you learnt how to choose a good number of clusters for a dataset using the k-means inertia graph. You are given a dataset of the measurements of samples of grain. What's a good number of clusters in this case?

This dataset was obtained from the UCI.

From the course *Transition to Data Science*. Buy the entire course for just \$10 for many more exercises and helpful video lectures.

Step 1: Load the dataset (written for you).

In [7]: seeds_df

```
In [6]: import pandas as pd

seeds_df = pd.read_csv('datasets/seeds.csv')
    # forget about the grain variety for the moment - we'll use this later
    del seeds_df['grain_variety']
```

Step 2: Display the DataFrame to inspect the data. Notice that there are 7 columns - so each grain sample (row) is a point in 7D space! Scatter plots can't help us here.

```
Out [7]:
              area perimeter compactness length width
                                                             asymmetry_coefficient
        0
             15.26
                        14.84
                                              5.763
                                                     3.312
                                     0.8710
                                                                             2.2210
        1
             14.88
                        14.57
                                     0.8811
                                              5.554 3.333
                                                                             1.0180
        2
             14.29
                        14.09
                                     0.9050
                                              5.291 3.337
                                                                             2.6990
        3
             13.84
                        13.94
                                     0.8955
                                              5.324 3.379
                                                                             2.2590
             16.14
                        14.99
                                     0.9034
                                              5.658 3.562
                                                                            1.3550
        4
        5
             14.38
                        14.21
                                     0.8951
                                              5.386 3.312
                                                                             2.4620
        6
             14.69
                        14.49
                                     0.8799
                                              5.563 3.259
                                                                             3.5860
        7
             14.11
                        14.10
                                     0.8911
                                              5.420 3.302
                                                                             2.7000
        8
             16.63
                        15.46
                                     0.8747
                                              6.053 3.465
                                                                             2.0400
        9
             16.44
                        15.25
                                     0.8880
                                              5.884 3.505
                                                                            1.9690
             15.26
        10
                        14.85
                                     0.8696
                                              5.714 3.242
                                                                             4.5430
        11
             14.03
                        14.16
                                     0.8796
                                              5.438 3.201
                                                                             1.7170
        12
             13.89
                        14.02
                                     0.8880
                                              5.439 3.199
                                                                             3.9860
        13
             13.78
                        14.06
                                     0.8759
                                              5.479 3.156
                                                                             3.1360
             13.74
                        14.05
                                     0.8744
                                              5.482 3.114
        14
                                                                             2.9320
        15
             14.59
                        14.28
                                     0.8993
                                              5.351 3.333
                                                                             4.1850
             13.99
        16
                        13.83
                                     0.9183
                                              5.119 3.383
                                                                             5.2340
        17
             15.69
                        14.75
                                     0.9058
                                              5.527 3.514
                                                                             1.5990
        18
             14.70
                        14.21
                                     0.9153
                                              5.205 3.466
                                                                             1.7670
```

0.8686

0.8584

5.226 3.049

5.658 3.129

12.72

14.16

13.57

14.40

19

20

4.1020

3.0720

21	14.11	14.26	0.8722	5.520	3.168	2.6880
22	15.88	14.90	0.8988	5.618	3.507	0.7651
23	12.08	13.23	0.8664	5.099	2.936	1.4150
24	15.01	14.76	0.8657	5.789	3.245	1.7910
25	16.19	15.16	0.8849	5.833	3.421	0.9030
26	13.02	13.76	0.8641	5.395	3.026	3.3730
27	12.74	13.67	0.8564	5.395	2.956	2.5040
28	14.11	14.18	0.8820	5.541	3.221	2.7540
29	13.45	14.02	0.8604	5.516	3.065	3.5310
		• • •			• • •	
180	11.41		0.8560	5.090	2.775	4.9570
181	12.46	13.41	0.8706	5.236	3.017	4.9870
182	12.19	13.36	0.8579	5.240	2.909	4.8570
183	11.65	13.07	0.8575	5.108	2.850	5.2090
184	12.89	13.77	0.8541	5.495	3.026	6.1850
185	11.56	13.31	0.8198	5.363	2.683	4.0620
186	11.81	13.45	0.8198	5.413	2.716	4.8980
187	10.91	12.80	0.8372	5.088	2.675	4.1790
188	11.23	12.82	0.8594	5.089	2.821	7.5240
189	10.59	12.41	0.8648	4.899	2.787	4.9750
190	10.93	12.80	0.8390	5.046	2.717	5.3980
191	11.27	12.86	0.8563	5.091	2.804	3.9850
192	11.87	13.02	0.8795	5.132	2.953	3.5970
193	10.82	12.83	0.8256	5.180	2.630	4.8530
194	12.11	13.27	0.8639	5.236	2.975	4.1320
195	12.80	13.47	0.8860	5.160	3.126	4.8730
196	12.79	13.53	0.8786	5.224	3.054	5.4830
197	13.37	13.78	0.8849	5.320	3.128	4.6700
198	12.62	13.67	0.8481	5.410	2.911	3.3060
199	12.76	13.38	0.8964	5.073	3.155	2.8280
200	12.38	13.44	0.8609	5.219	2.989	5.4720
201	12.67	13.32	0.8977	4.984	3.135	2.3000
202	11.18	12.72	0.8680	5.009	2.810	4.0510
203	12.70	13.41	0.8874	5.183	3.091	8.4560
204	12.37	13.47	0.8567	5.204	2.960	3.9190
205	12.19	13.20	0.8783	5.137	2.981	3.6310
206	11.23	12.88	0.8511	5.140	2.795	4.3250
207	13.20	13.66	0.8883	5.236	3.232	8.3150
208	11.84	13.21	0.8521	5.175	2.836	3.5980
209	12.30	13.34	0.8684	5.243	2.974	5.6370

groove_length
5.220

0

1	4.956
2	4.825
3	4.805
4	5.175
5	4.956
6	5.219
7	5.000
8	5.877
9	5.533
10	5.314
11	5.001
12	4.738
13	4.872
14	4.825
15	4.781
16	4.781
17	5.046
18	4.649
19	4.914
20	5.176
21	5.219
22	5.091
23	4.961
24	5.001
25	5.307
26	4.825
27	4.869
28	5.038
29	5.097
• •	
180	4.825
181	5.147
182	5.158
183	5.135
184	5.316
185	5.182
186	5.352
187	4.956
188	4.957
189	4.794
190	5.045
191	5.001
192	5.132

```
193
              5.089
194
              5.012
195
              4.914
              4.958
196
197
              5.091
198
              5.231
              4.830
199
              5.045
200
201
              4.745
202
              4.828
203
              5.000
204
              5.001
              4.870
205
206
              5.003
207
              5.056
208
              5.044
209
              5.063
[210 rows x 7 columns]
```

Step 3: Extract the measurements from the DataFrame using its .values attribute:

```
In [8]: samples = seeds_df.values
```

Step 4: (Written for you). Measure the quality of clusterings with different numbers of clusters using the inertia. For each of the given values of k, perform the following steps:

- Create a KMeans instance called model with k clusters.
- Fit the model to the grain data samples.
- Append the value of the inertia_ attribute of model to the list inertias.

In [9]: from sklearn.cluster import KMeans

```
ks = range(1, 6)
inertias = []

for k in ks:
    # Create a KMeans instance with k clusters: model
    model = KMeans(n_clusters=k)

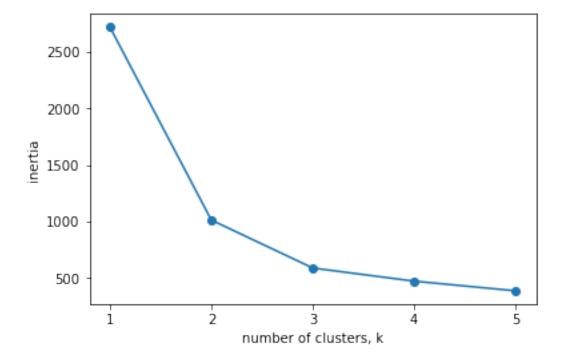
# Fit model to samples
    model.fit(samples)

# Append the inertia to the list of inertias
inertias.append(model.inertia_)
```

Step 5: Plot the inertia to see which number of clusters is best. Remember: lower numbers are better!

In [10]: import matplotlib.pyplot as plt

```
# Plot ks vs inertias
plt.plot(ks, inertias, '-o')
plt.xlabel('number of clusters, k')
plt.ylabel('inertia')
plt.xticks(ks)
plt.show()
```



Excellent work! You can see from the graph that 3 is a good number of clusters, since these are points where the inertia begins to decrease more slowly.

In []: