Exercise 13: Different linkage, different hierarchical clustering!

In the video, you saw a hierarchical clustering of the voting countries at the Eurovision song contest using 'complete' linkage. Now, perform a hierarchical clustering of the voting countries with 'single' linkage, and compare the resulting dendrogram with the one in the video. Different linkage, different hierarchical clustering!

First, we need to do a little pre-processing to account for one of the Eurovision rules: countries are not allowed to vote for themselves.

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Step 1: Load the DataFrame (written for you)

In [2]: scores_df.head()

```
In [1]: import pandas as pd
        scores_df = pd.read_csv('../datasets/eurovision-2016-televoting.csv', index_col=0)
        country_names = list(scores_df.index)
```

Step 2: Display the DataFrame, and have a look. Each row represents a country that voted, while each column represents a country that performed.

Notice the NaN ("not-a-number") values. These correspond to missing scores in the original CSV file. These scores are missing because countries that performed were not allowed to vote for themselves.

```
Armenia Australia Austria Azerbaijan Belgium Bulgaria \
Out [2]:
```

From cou	ntry							
Albania	2.0	12	.0 0	.0	0.0	0.0	8.	0
Armenia	NaN	0	.0 4	.0	0.0	0.0	0.	0
Australi	a 0.0	Na	aN 3	.0	0.0	12.0	10.	0
Austria	0.0	3	.0 N	aN	0.0	0.0	5.	0
Azerbaij	an 0.0	2	.0 0	.0	NaN	0.0	8.	0
	Croatia	Cyprus	Czech Re	public	France	• • •		\
From cou	ntry						•	
Albania	0.0	0.0		0.0	0.0		•	
Armenia	0.0	6.0		0.0	7.0			
Australi	a 0.0	0.0		0.0	7.0			
Austria	0.0	0.0		0.0	1.0			
Azerbaij	an 0.0	0.0		0.0	4.0	• • •		
	Lithuania	a Malta	Poland	Russia	Serbia	Spain	Sweden	\
From cou	ntry							
Albania	4.	0.0	5.0	7.0	0.0	0.0	3.0	

Armenia	0.0	5.0	1.0	12.0	0.0	0.0	2.0
Australia	1.0	6.0	0.0	5.0	0.0	2.0	0.0
Austria	0.0	0.0	12.0	8.0	4.0	0.0	7.0
Azerbaijan	0.0	5.0	3.0	12.0	0.0	0.0	0.0
	The Netherlands		Ukraine	United	Kingdom		
From country							
Albania		0.0	6.0		0.0		
Armenia		0.0	10.0		0.0		
Australia		0.0	8.0		4.0		
Austria		6.0	10.0		0.0		
Azerbaijan		0.0	10.0		0.0		

[5 rows x 26 columns]

Step 3: Fill in the NaNs with the highest possible score (12) - we are assuming that countries would vote for themselves, if they had been allowed to do so. (*This bit written for you*).

```
In [3]: scores_df = scores_df.fillna(12)
```

Step 4: Import the normalize function from sklearn.preprocessing.

```
In [4]: from sklearn.preprocessing import normalize
```

Step 5: Apply the normalize function to scores_df.values, assigning the result to samples.

(Why do we need to normalize? Because now that the missing values are filled with 12 points, some countries (those that performed) given a greater total number of points when voting. The normalize function corrects for this.)

```
In [5]: samples = normalize(scores_df.values)
```

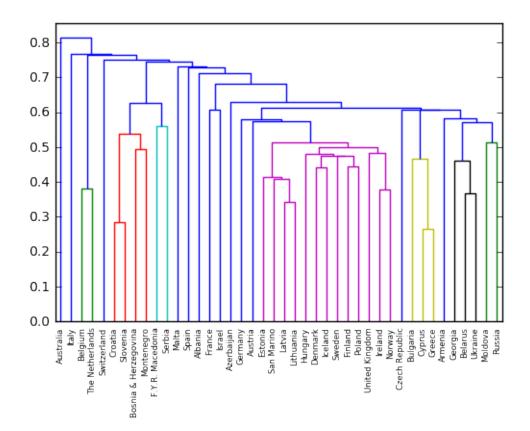
Step 6: Import: + linkage and dendrogram from scipy.cluster.hierarchy. + matplotlib.pyplot as plt.

```
In [6]: from scipy.cluster.hierarchy import linkage, dendrogram
    import matplotlib.pyplot as plt
```

Step 7: Perform hierarchical clustering on samples using the linkage() function with the method='single' keyword argument. Assign the result to mergings.

```
In [7]: mergings = linkage(samples, method='single')
```

Step 8: Plot a dendrogram of the hierarchical clustering, using the list country_names as the labels. In addition, specify the leaf_rotation=90, and leaf_font_size=6 keyword arguments as you have done earlier.



Step 9: Compare your dendrogram above to the one in the slides and notice that different linkage functions give different hierarchical clusterings.

Both the linkage functions we've considered, "complete" and "single", have advantages and disadvantages. In practice, just try both out, and see which dendrogram seems more sensible.

In []: