Dimensionality_reduction_visualisation

January 21, 2020

1 Dimensionality reduciton on the iris data set

```
[1]: from sklearn import datasets
from sklearn.utils import shuffle
import pandas as pd
import numpy as np
#Load data in matrix X for easier reference
iris = datasets.load_iris()
X = iris.data
#Normalize Data to be in range [0,1]
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
scaler.fit(X)
X = scaler.transform(X)
#Create Pandas Dataframe for table display
df = pd.DataFrame(X[:,0:3])
df.columns = iris.feature_names[0:3]
df['class'] = iris.target
#label class names
df['class_name'] = [iris.target_names[i] for i in df['class']]
#shuffle data
df = shuffle(df)
X = np.array(df.loc[:,'sepal length (cm)':'petal length (cm)'])
#Visualize data in table
df
```

```
[1]:
      sepal length (cm) sepal width (cm) petal length (cm)
                                                              class class_name
 108
               0.666667
                                 0.208333
                                                    0.813559
                                                                       virginica
 25
               0.194444
                                 0.416667
                                                    0.101695
                                                                   0
                                                                          setosa
 147
               0.611111
                                 0.416667
                                                    0.711864
                                                                       virginica
```

16	0.305556	0.791667	0.050847	0	setosa
69	0.361111	0.208333	0.491525	1	versicolor
77	0.666667	0.416667	0.677966	1	versicolor
42	0.027778	0.500000	0.050847	0	setosa
14	0.416667	0.833333	0.033898	0	setosa
12	0.138889	0.416667	0.067797	0	setosa
148	0.527778	0.583333	0.745763	2	virginica
115	0.583333	0.500000	0.728814	2	virginica
59	0.250000	0.291667	0.491525	1	versicolor
83	0.472222	0.291667	0.694915	1	versicolor
34	0.166667	0.458333	0.084746	0	setosa
41	0.055556	0.125000	0.050847	0	setosa
120	0.722222	0.500000	0.796610	2	virginica
17	0.222222	0.625000	0.067797	0	setosa
79	0.388889	0.250000	0.423729	1	versicolor
125	0.805556	0.500000	0.847458	2	virginica
131	1.000000	0.750000	0.915254	2	virginica
4	0.194444	0.666667	0.067797	0	setosa
63	0.500000	0.375000	0.627119	1	versicolor
1	0.166667	0.416667	0.067797	0	setosa
55	0.388889	0.333333	0.593220	1	versicolor
140	0.666667	0.458333	0.779661	2	virginica
126	0.527778	0.333333	0.644068	2	virginica
92	0.416667	0.250000	0.508475	1	versicolor
37	0.166667	0.666667	0.067797	0	setosa
75	0.638889	0.416667	0.576271	1	versicolor
110	0.611111	0.500000	0.694915	2	virginica
		• • •			
88	0.361111	0.416667	0.525424	1	versicolor
67	0.416667	0.291667	0.525424	1	versicolor
124	0.666667	0.541667	0.796610	2	virginica
38	0.027778	0.416667	0.050847	0	setosa
135	0.944444	0.416667	0.864407	2	virginica
43	0.194444	0.625000	0.101695	0	setosa
31	0.305556	0.583333	0.084746	0	setosa
10	0.305556	0.708333	0.084746	0	setosa
89	0.333333	0.208333	0.508475	1	versicolor
61	0.44444	0.416667	0.542373	1	versicolor
33	0.333333	0.916667	0.067797	0	setosa
80	0.333333	0.166667	0.474576	1	versicolor
137	0.583333	0.458333	0.762712	2	virginica
78	0.472222	0.375000	0.593220	1	versicolor
85	0.472222	0.583333	0.593220	1	versicolor
45	0.138889	0.416667	0.067797	0	setosa
73	0.500000	0.333333	0.627119	1	versicolor
76	0.694444	0.333333	0.644068	1	versicolor
111	0.583333	0.291667	0.728814	2	virginica
					-

93	0.194444	0.125000	0.389831	1	versicolor
23	0.222222	0.541667	0.118644	0	setosa
142	0.416667	0.291667	0.694915	2	virginica
8	0.027778	0.375000	0.067797	0	setosa
57	0.166667	0.166667	0.389831	1	versicolor
109	0.805556	0.666667	0.864407	2	virginica
18	0.388889	0.750000	0.118644	0	setosa
82	0.416667	0.291667	0.491525	1	versicolor
48	0.277778	0.708333	0.084746	0	setosa
72	0.555556	0.208333	0.661017	1	versicolor
122	0.944444	0.333333	0.966102	2	virginica

[150 rows x 5 columns]

```
[3]: from mpl_toolkits import mplot3d
import matplotlib.pyplot as plt
%matplotlib notebook
#Iniatiate figure
plt.figure(figsize = (10,10))
ax = plt.axes(projection = '3d')
#Filter dataframe for different classes
versicolor = df[df['class name'].str.match('versicolor')]
virginica = df[df['class_name'].str.match('virginica')]
setosa = df[df['class_name'].str.match('setosa')]
#Plot different classes
ax.scatter3D(versicolor['sepal length (cm)'],versicolor['sepal width_
 →(cm)'],versicolor['petal length (cm)'],color = 'red', marker = '^')
ax.scatter3D(virginica['sepal length (cm)'],virginica['sepal widthu
 →(cm)'],virginica['petal length (cm)'],color = 'blue', marker = 'o')
ax.scatter3D(setosa['sepal length (cm)'],setosa['sepal width_
 →(cm)'],setosa['petal length (cm)'],color = 'green', marker = '*')
ax.set_title('Three dimensional plot of reduced data set')
ax.set_xlabel('sepal length [cm]')
ax.set_ylabel('sepal width [cm]')
ax.set_zlabel('petal length [cm]')
plt.show()
```

<IPython.core.display.Javascript object>

<IPython.core.display.HTML object>

[]: