

PCA

September 30, 2020

```
[2]: import pandas as pd
from sklearn.preprocessing import StandardScaler
import numpy as np
from mpl_toolkits.mplot3d import Axes3D
import matplotlib.pyplot as plt
import matplotlib.tri as mtri
from sklearn.decomposition import PCA
import seaborn as sns
sns.set()
df=pd.read_csv("https://archive.ics.uci.edu/ml/machine-learning-databases/iris/
→iris.data", names=['sepal length','sepal width','petal length','petal_
→width','target'])
print(df)
```

	sepal length	sepal width	petal length	petal width	target
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa
..
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

[150 rows x 5 columns]

```
[3]: features = ['sepal length', 'sepal width', 'petal length', 'petal width']
# Separating out the features
x = df.loc[:, features].values
# Separating out the target
y = df.loc[:,['target']].values
# Standardizing the features
x = StandardScaler().fit_transform(x)
print(x)
```

```

[-9.00681170e-01  1.03205722e+00 -1.34127240e+00 -1.31297673e+00]
[-1.14301691e+00 -1.24957601e-01 -1.34127240e+00 -1.31297673e+00]
[-1.38535265e+00  3.37848329e-01 -1.39813811e+00 -1.31297673e+00]
[-1.50652052e+00  1.06445364e-01 -1.28440670e+00 -1.31297673e+00]
[-1.02184904e+00  1.26346019e+00 -1.34127240e+00 -1.31297673e+00]
[-5.37177559e-01  1.95766909e+00 -1.17067529e+00 -1.05003079e+00]
[-1.50652052e+00  8.00654259e-01 -1.34127240e+00 -1.18150376e+00]
[-1.02184904e+00  8.00654259e-01 -1.28440670e+00 -1.31297673e+00]
[-1.74885626e+00 -3.56360566e-01 -1.34127240e+00 -1.31297673e+00]
[-1.14301691e+00  1.06445364e-01 -1.28440670e+00 -1.44444970e+00]
[-5.37177559e-01  1.49486315e+00 -1.28440670e+00 -1.31297673e+00]
[-1.26418478e+00  8.00654259e-01 -1.22754100e+00 -1.31297673e+00]
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[-1.38535265e+00  3.37848329e-01 -1.22754100e+00 -1.31297673e+00]
[-1.26418478e+00  1.06445364e-01 -1.22754100e+00 -1.31297673e+00]
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[-1.02184904e+00  3.37848329e-01 -1.45500381e+00 -1.31297673e+00]
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[-1.74885626e+00  3.37848329e-01 -1.39813811e+00 -1.31297673e+00]
[-1.02184904e+00  1.03205722e+00 -1.22754100e+00 -7.87084847e-01]
[-9.00681170e-01  1.72626612e+00 -1.05694388e+00 -1.05003079e+00]
[-1.26418478e+00 -1.24957601e-01 -1.34127240e+00 -1.18150376e+00]
[-9.00681170e-01  1.72626612e+00 -1.22754100e+00 -1.31297673e+00]
[-1.50652052e+00  3.37848329e-01 -1.34127240e+00 -1.31297673e+00]

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[-1.02184904e+00  5.69251294e-01 -1.34127240e+00 -1.31297673e+00]
[ 1.40150837e+00  3.37848329e-01  5.35295827e-01  2.64698913e-01]
[ 6.74501145e-01  3.37848329e-01  4.21564419e-01  3.96171883e-01]
[ 1.28034050e+00  1.06445364e-01  6.49027235e-01  3.96171883e-01]
[-4.16009689e-01 -1.74477836e+00  1.37235899e-01  1.33225943e-01]
[ 7.95669016e-01 -5.87763531e-01  4.78430123e-01  3.96171883e-01]
[-1.73673948e-01 -5.87763531e-01  4.21564419e-01  1.33225943e-01]
[ 5.53333275e-01  5.69251294e-01  5.35295827e-01  5.27644853e-01]
[-1.14301691e+00 -1.51337539e+00 -2.60824029e-01 -2.61192967e-01]
[ 9.16836886e-01 -3.56360566e-01  4.78430123e-01  1.33225943e-01]
[-7.79513300e-01 -8.19166497e-01  8.03701950e-02  2.64698913e-01]
[-1.02184904e+00 -2.43898725e+00 -1.47092621e-01 -2.61192967e-01]
[ 6.86617933e-02 -1.24957601e-01  2.50967307e-01  3.96171883e-01]
[ 1.89829664e-01 -1.97618132e+00  1.37235899e-01 -2.61192967e-01]
[ 3.10997534e-01 -3.56360566e-01  5.35295827e-01  2.64698913e-01]
[-2.94841818e-01 -3.56360566e-01 -9.02269170e-02  1.33225943e-01]
[ 1.03800476e+00  1.06445364e-01  3.64698715e-01  2.64698913e-01]
[-2.94841818e-01 -1.24957601e-01  4.21564419e-01  3.96171883e-01]
[-5.25060772e-02 -8.19166497e-01  1.94101603e-01 -2.61192967e-01]
[ 4.32165405e-01 -1.97618132e+00  4.21564419e-01  3.96171883e-01]
[-2.94841818e-01 -1.28197243e+00  8.03701950e-02 -1.29719997e-01]
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[ 3.10997534e-01 -5.87763531e-01  1.37235899e-01  1.33225943e-01]
[ 5.53333275e-01 -1.28197243e+00  6.49027235e-01  3.96171883e-01]
[ 3.10997534e-01 -5.87763531e-01  5.35295827e-01  1.75297293e-03]
[ 6.74501145e-01 -3.56360566e-01  3.07833011e-01  1.33225943e-01]
[ 9.16836886e-01 -1.24957601e-01  3.64698715e-01  2.64698913e-01]
[ 1.15917263e+00 -5.87763531e-01  5.92161531e-01  2.64698913e-01]
[ 1.03800476e+00 -1.24957601e-01  7.05892939e-01  6.59117823e-01]
[ 1.89829664e-01 -3.56360566e-01  4.21564419e-01  3.96171883e-01]
[-1.73673948e-01 -1.05056946e+00 -1.47092621e-01 -2.61192967e-01]
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[ 1.89829664e-01 -8.19166497e-01  7.62758643e-01  5.27644853e-01]
[-5.37177559e-01 -1.24957601e-01  4.21564419e-01  3.96171883e-01]
[ 1.89829664e-01  8.00654259e-01  4.21564419e-01  5.27644853e-01]
[ 1.03800476e+00  1.06445364e-01  5.35295827e-01  3.96171883e-01]
[ 5.53333275e-01 -1.74477836e+00  3.64698715e-01  1.33225943e-01]
[-2.94841818e-01 -1.24957601e-01  1.94101603e-01  1.33225943e-01]
[-4.16009689e-01 -1.28197243e+00  1.37235899e-01  1.33225943e-01]
[-4.16009689e-01 -1.05056946e+00  3.64698715e-01  1.75297293e-03]
[ 3.10997534e-01 -1.24957601e-01  4.78430123e-01  2.64698913e-01]
[-5.25060772e-02 -1.05056946e+00  1.37235899e-01  1.75297293e-03]
[-1.02184904e+00 -1.74477836e+00 -2.60824029e-01 -2.61192967e-01]
[-2.94841818e-01 -8.19166497e-01  2.50967307e-01  1.33225943e-01]
[-1.73673948e-01 -1.24957601e-01  2.50967307e-01  1.75297293e-03]

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[-1.73673948e-01	-3.56360566e-01	2.50967307e-01	1.33225943e-01]
[4.32165405e-01	-3.56360566e-01	3.07833011e-01	1.33225943e-01]
[-9.00681170e-01	-1.28197243e+00	-4.31421141e-01	-1.29719997e-01]
[-1.73673948e-01	-5.87763531e-01	1.94101603e-01	1.33225943e-01]
[5.53333275e-01	5.69251294e-01	1.27454998e+00	1.71090158e+00]
[-5.25060772e-02	-8.19166497e-01	7.62758643e-01	9.22063763e-01]
[1.52267624e+00	-1.24957601e-01	1.21768427e+00	1.18500970e+00]
[5.53333275e-01	-3.56360566e-01	1.04708716e+00	7.90590793e-01]
[7.95669016e-01	-1.24957601e-01	1.16081857e+00	1.31648267e+00]
[2.12851559e+00	-1.24957601e-01	1.61574420e+00	1.18500970e+00]
[-1.14301691e+00	-1.28197243e+00	4.21564419e-01	6.59117823e-01]
[1.76501198e+00	-3.56360566e-01	1.44514709e+00	7.90590793e-01]
[1.03800476e+00	-1.28197243e+00	1.16081857e+00	7.90590793e-01]
[1.64384411e+00	1.26346019e+00	1.33141568e+00	1.71090158e+00]
[7.95669016e-01	3.37848329e-01	7.62758643e-01	1.05353673e+00]
[6.74501145e-01	-8.19166497e-01	8.76490051e-01	9.22063763e-01]
[1.15917263e+00	-1.24957601e-01	9.90221459e-01	1.18500970e+00]
[-1.73673948e-01	-1.28197243e+00	7.05892939e-01	1.05353673e+00]
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[7.95669016e-01	-1.24957601e-01	9.90221459e-01	7.90590793e-01]
[2.24968346e+00	1.72626612e+00	1.67260991e+00	1.31648267e+00]
[2.24968346e+00	-1.05056946e+00	1.78634131e+00	1.44795564e+00]
[1.89829664e-01	-1.97618132e+00	7.05892939e-01	3.96171883e-01]
[1.28034050e+00	3.37848329e-01	1.10395287e+00	1.44795564e+00]
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[6.74501145e-01	-5.87763531e-01	1.04708716e+00	1.18500970e+00]
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[1.88617985e+00	-5.87763531e-01	1.33141568e+00	9.22063763e-01]
[2.49201920e+00	1.72626612e+00	1.50201279e+00	1.05353673e+00]
[6.74501145e-01	-5.87763531e-01	1.04708716e+00	1.31648267e+00]
[5.53333275e-01	-5.87763531e-01	7.62758643e-01	3.96171883e-01]
[3.10997534e-01	-1.05056946e+00	1.04708716e+00	2.64698913e-01]
[2.24968346e+00	-1.24957601e-01	1.33141568e+00	1.44795564e+00]
[5.53333275e-01	8.00654259e-01	1.04708716e+00	1.57942861e+00]
[6.74501145e-01	1.06445364e-01	9.90221459e-01	7.90590793e-01]
[1.89829664e-01	-1.24957601e-01	5.92161531e-01	7.90590793e-01]
[1.28034050e+00	1.06445364e-01	9.33355755e-01	1.18500970e+00]
[1.03800476e+00	1.06445364e-01	1.04708716e+00	1.57942861e+00]
[1.28034050e+00	1.06445364e-01	7.62758643e-01	1.44795564e+00]
[-5.25060772e-02	-8.19166497e-01	7.62758643e-01	9.22063763e-01]
[1.15917263e+00	3.37848329e-01	1.21768427e+00	1.44795564e+00]

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[ 1.03800476e+00  5.69251294e-01  1.10395287e+00  1.71090158e+00]
[ 1.03800476e+00 -1.24957601e-01  8.19624347e-01  1.44795564e+00]
[ 5.53333275e-01 -1.28197243e+00  7.05892939e-01  9.22063763e-01]
[ 7.95669016e-01 -1.24957601e-01  8.19624347e-01  1.05353673e+00]
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[ 6.86617933e-02 -1.24957601e-01  7.62758643e-01  7.90590793e-01]]
```

```
[4]: pca = PCA(n_components=4)
principalComponents = pca.fit_transform(x)
principalDf = pd.DataFrame(data = principalComponents
                           , columns = ['component1', 'component2', 'com3', 'comp4'])
```

```
[5]: df.describe()
```

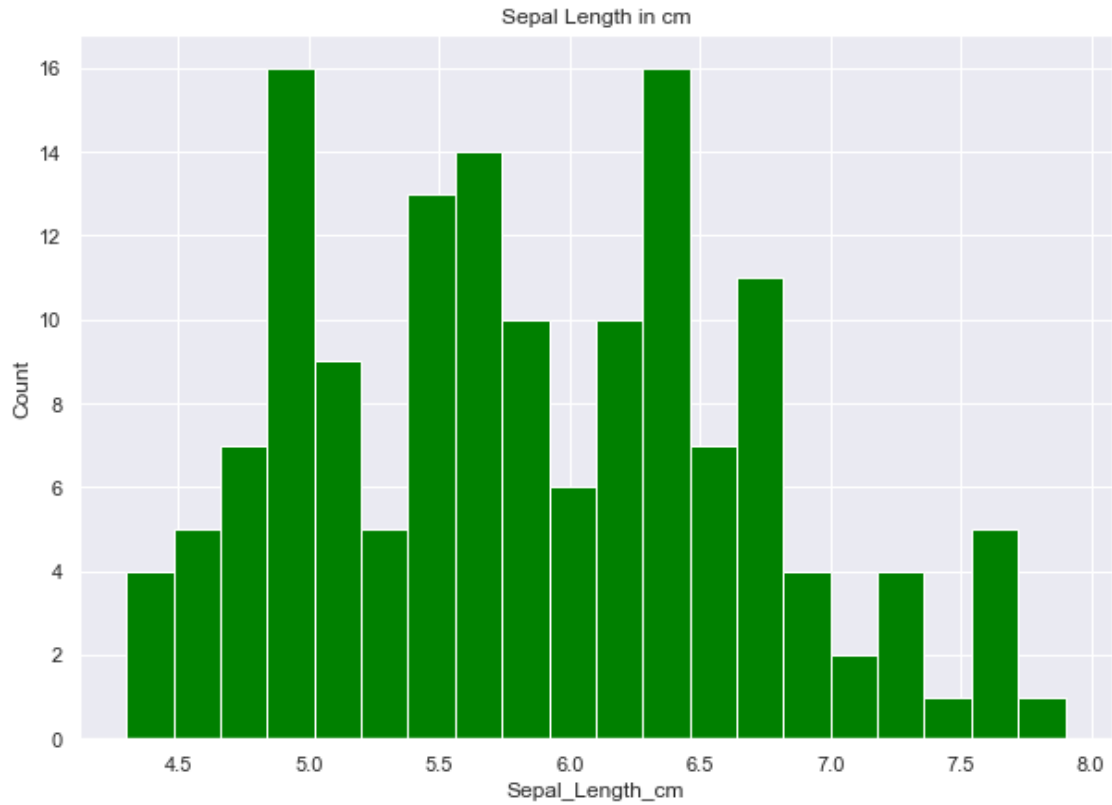
```
[5]:
```

	sepal length	sepal width	petal length	petal width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
[6]: plt.figure(figsize = (10, 7))
a = df["sepal length"]

plt.hist(a, bins = 20, color = "green")
plt.title("Sepal Length in cm")
plt.xlabel("Sepal_Length_cm")
plt.ylabel("Count")
```

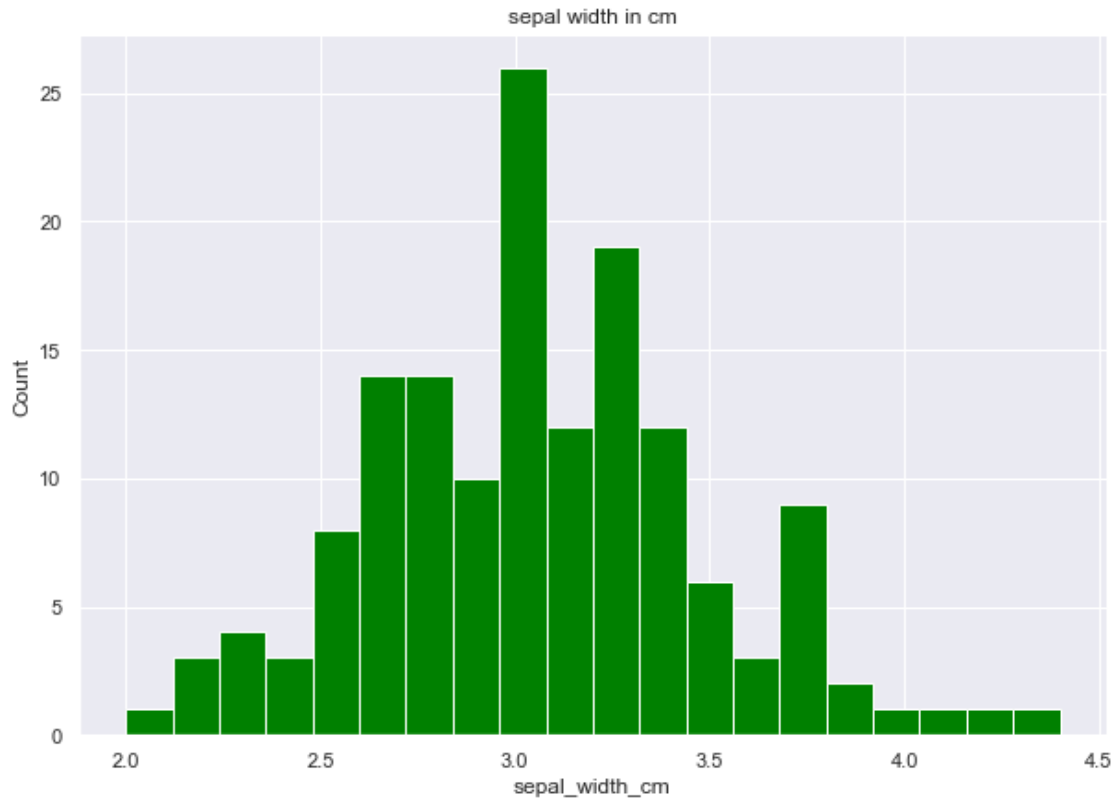
```
[6]: Text(0, 0.5, 'Count')
```



```
[7]: plt.figure(figsize = (10, 7))
      b = df["sepal width"]

      plt.hist(b, bins = 20, color = "green")
      plt.title("sepal width in cm")
      plt.xlabel("sepal_width_cm")
      plt.ylabel("Count")
```

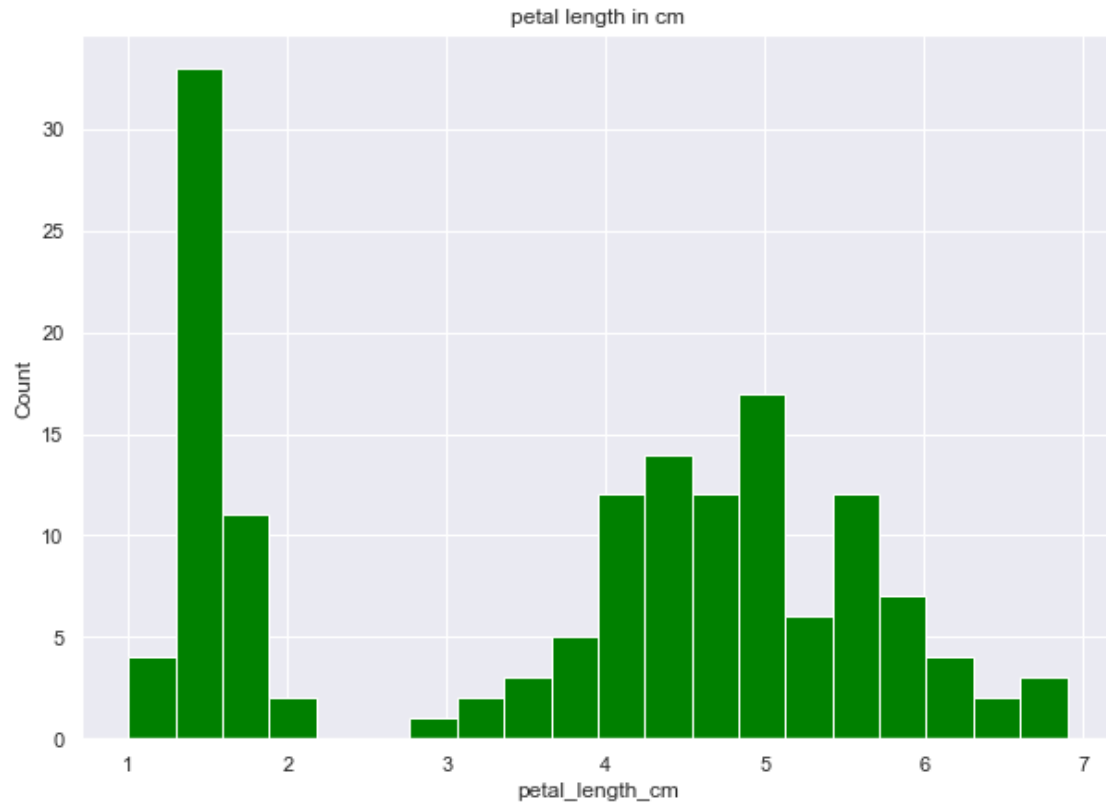
```
[7]: Text(0, 0.5, 'Count')
```



```
[8]: plt.figure(figsize = (10, 7))
c = df["petal length"]

plt.hist(c, bins = 20, color = "green")
plt.title("petal length in cm")
plt.xlabel("petal_length_cm")
plt.ylabel("Count")
```

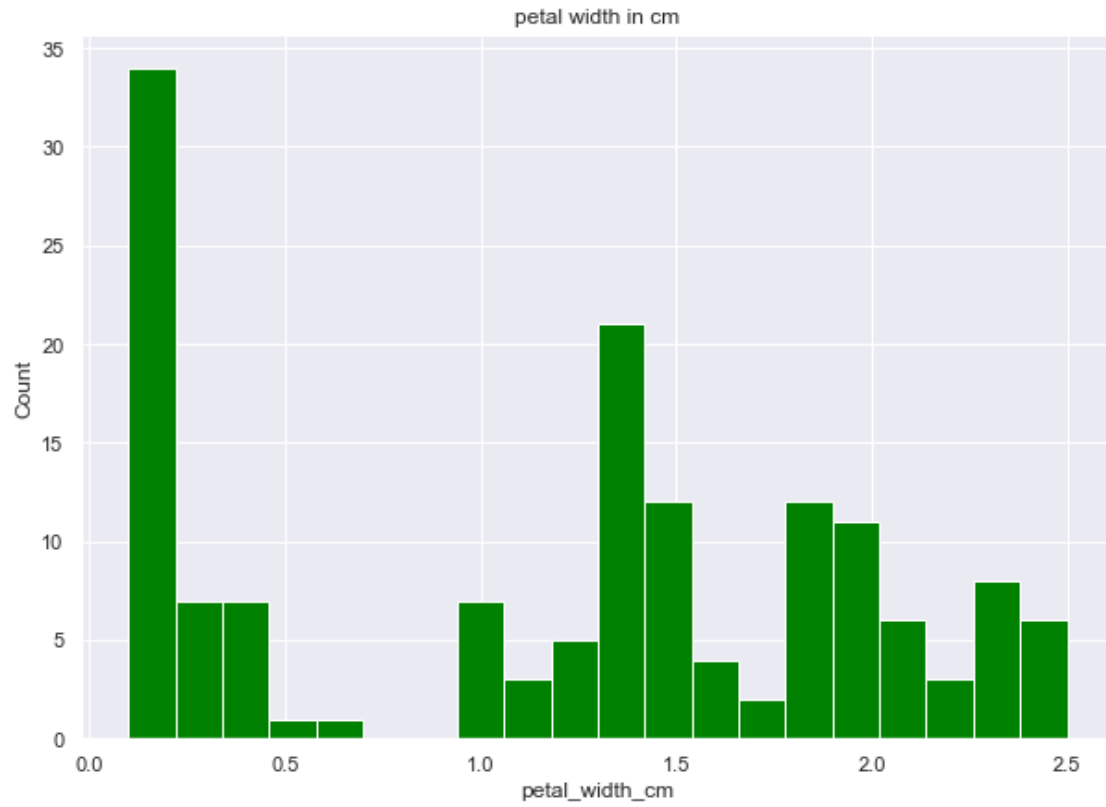
```
[8]: Text(0, 0.5, 'Count')
```



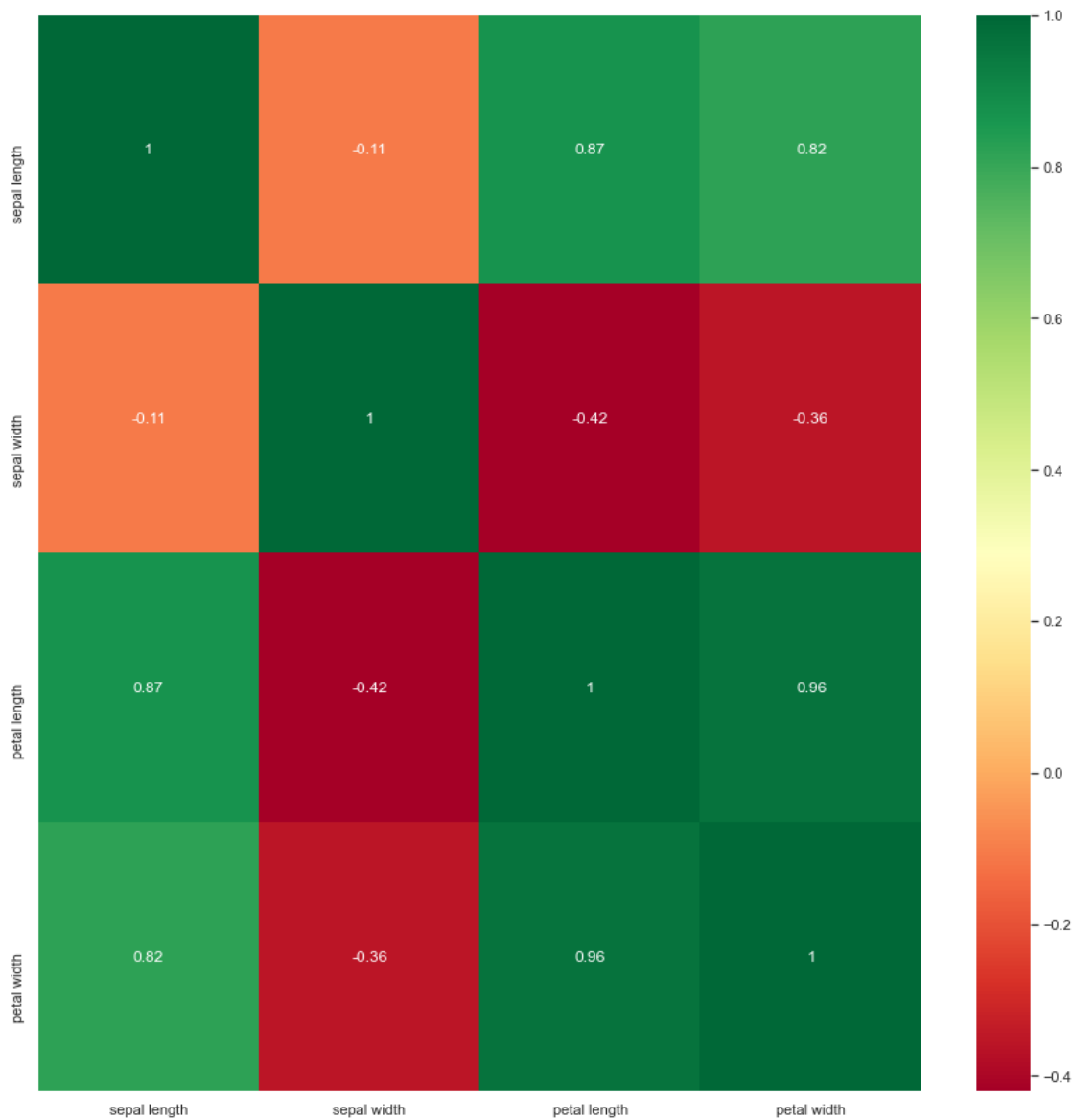
```
[14]: plt.figure(figsize = (10, 7))
      d = df["petal width"]

      plt.hist(d, bins = 20, color = "green")
      plt.title("petal width in cm")
      plt.xlabel("petal_width_cm")
      plt.ylabel("Count")
```

```
[14]: Text(0, 0.5, 'Count')
```

```
[15]: plt.figure(figsize=(15,15))  
p=sns.heatmap(df.corr(), annot=True,cmap='RdYlGn')
```



```
[16]: print(principalDf)
```

	component1	component2	com3	comp4
0	-2.264542	0.505704	-0.121943	-0.023073
1	-2.086426	-0.655405	-0.227251	-0.103208
2	-2.367950	-0.318477	0.051480	-0.027825
3	-2.304197	-0.575368	0.098860	0.066311
4	-2.388777	0.674767	0.021428	0.037397
..
145	1.870522	0.382822	0.254532	-0.388890
146	1.558492	-0.905314	-0.025382	-0.221322
147	1.520845	0.266795	0.179277	-0.118903

```

148    1.376391    1.016362  0.931405 -0.024146
149    0.959299   -0.022284  0.528794  0.163676

```

[150 rows x 4 columns]

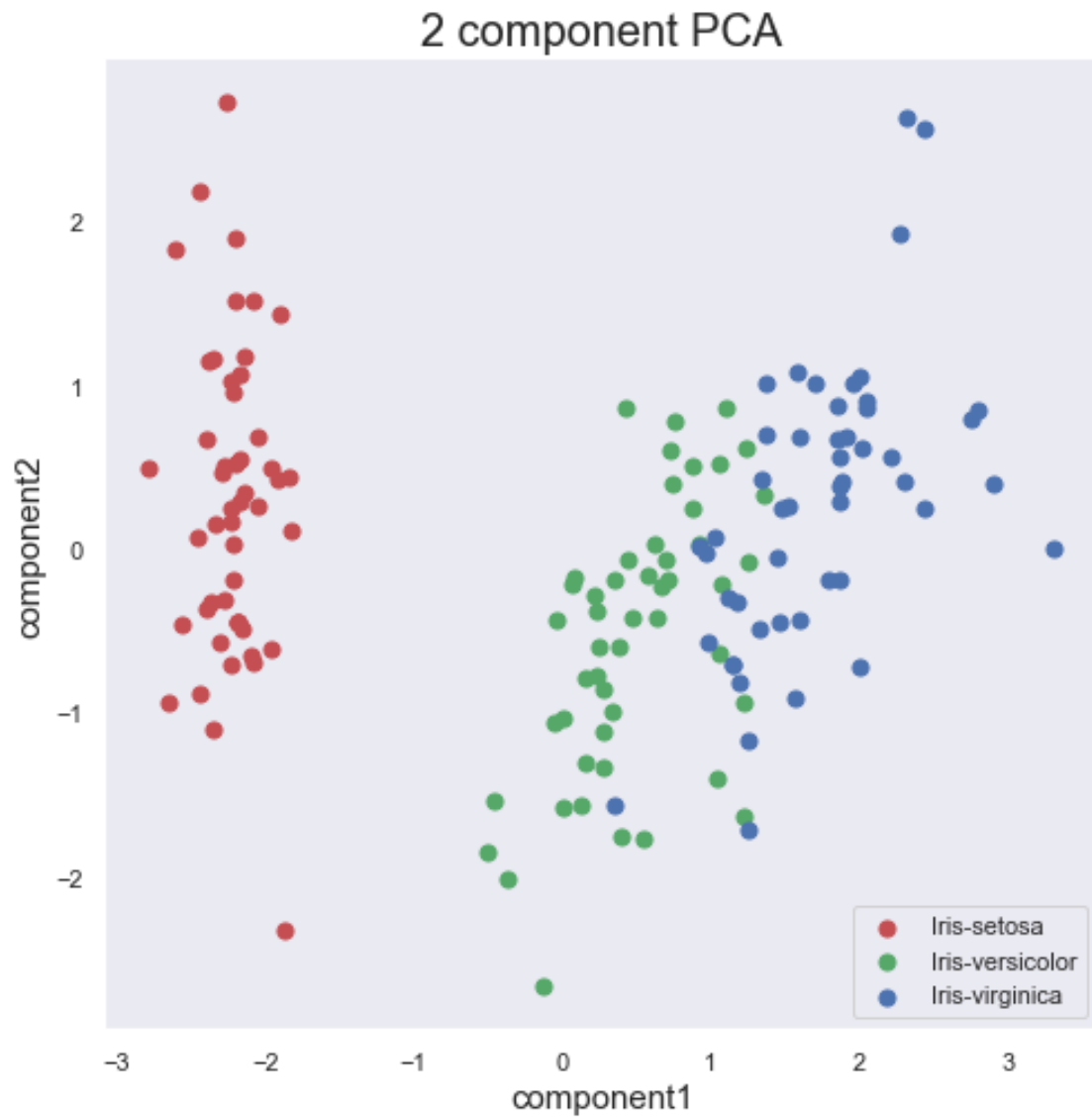
```
[17]: finalDf = pd.concat([principalDf, df[['target']], axis = 1)
```

```
[18]: print(finalDf)
```

	component1	component2	com3	comp4	target
0	-2.264542	0.505704	-0.121943	-0.023073	Iris-setosa
1	-2.086426	-0.655405	-0.227251	-0.103208	Iris-setosa
2	-2.367950	-0.318477	0.051480	-0.027825	Iris-setosa
3	-2.304197	-0.575368	0.098860	0.066311	Iris-setosa
4	-2.388777	0.674767	0.021428	0.037397	Iris-setosa
..
145	1.870522	0.382822	0.254532	-0.388890	Iris-virginica
146	1.558492	-0.905314	-0.025382	-0.221322	Iris-virginica
147	1.520845	0.266795	0.179277	-0.118903	Iris-virginica
148	1.376391	1.016362	0.931405	-0.024146	Iris-virginica
149	0.959299	-0.022284	0.528794	0.163676	Iris-virginica

[150 rows x 5 columns]

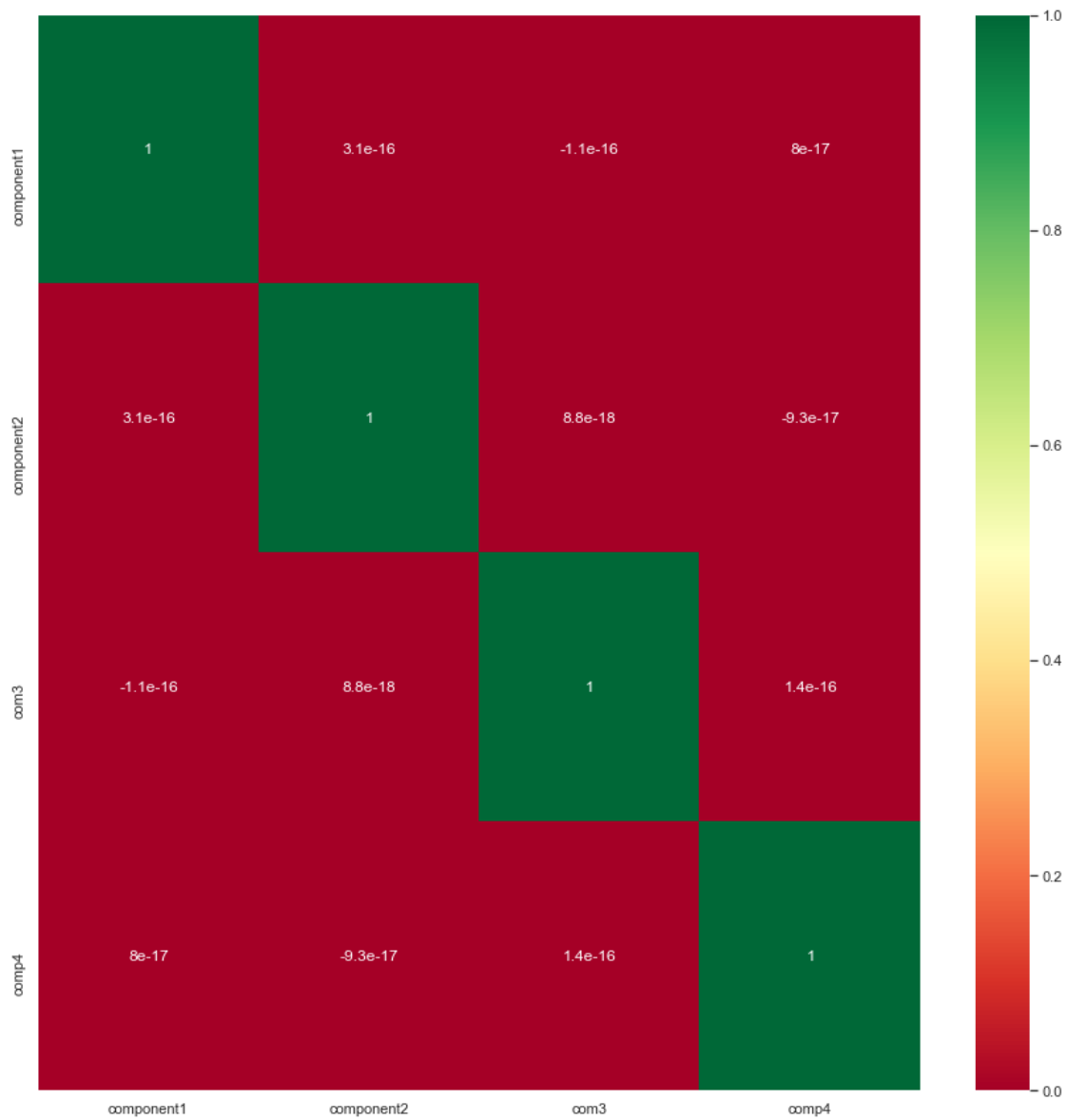
```
[19]: fig = plt.figure(figsize = (8,8))
ax = fig.add_subplot(1,1,1)
ax.set_xlabel('component1', fontsize = 15)
ax.set_ylabel('component2', fontsize = 15)
ax.set_title('2 component PCA', fontsize = 20)
targets = ['Iris-setosa', 'Iris-versicolor', 'Iris-virginica']
colors = ['r', 'g', 'b']
for target, color in zip(targets, colors):
    indicesToKeep = finalDf['target'] == target
    ax.scatter(finalDf.loc[indicesToKeep, 'component1']
               , finalDf.loc[indicesToKeep, 'component2']
               , c = color
               , s = 50)
ax.legend(targets)
ax.grid()
```



```
[20]: pca.explained_variance_ratio_
```

```
[20]: array([0.72770452, 0.23030523, 0.03683832, 0.00515193])
```

```
[21]: plt.figure(figsize=(15,15))  
p=sns.heatmap(finalDf.corr(), annot=True,cmap='RdYlGn')
```



[]: