

```
In [1]: from sklearn import datasets
        from sklearn import preprocessing
        from sklearn.decomposition import PCA
        import pandas as pd
        import numpy as np
        from sklearn.preprocessing import StandardScaler
        from matplotlib import pyplot as plt
```

```
In [2]: df = pd.read_csv('churn_clean.csv')
```

```
In [3]: df.info
```

```
Out[3]: <bound method DataFrame.info of          CaseOrder Customer_id
Interaction \
0          1      K409198 aa90260b-4141-4a24-8e36-b04ce1f4f77b
1          2      S120509 fb76459f-c047-4a9d-8af9-e0f7d4ac2524
2          3      K191035 344d114c-3736-4be5-98f7-c72c281e2d35
3          4      D90850  abfa2b40-2d43-4994-b15a-989b8c79e311
4          5      K662701 68a861fd-0d20-4e51-a587-8a90407ee574
...      ...      ...      ...
9995      9996      M324793 45deb5a2-ae04-4518-bf0b-c82db8dbe4a4
9996      9997      D861732 6e96b921-0c09-4993-bbda-a1ac6411061a
9997      9998      I243405 e8307ddf-9a01-4fff-bc59-4742e03fd24f
9998      9999      I641617 3775ccfc-0052-4107-81ae-9657f81ecdff
9999      10000      T38070 9de5fb6e-bd33-4995-aec8-f01d0172a499
```

```

                                UID          City State \
0      e885b299883d4f9fb18e39c75155d990 Point Baker AK
1      f2de8bef964785f41a2959829830fb8a West Branch MI
2      f1784cfa9f6d92ae816197eb175d3c71  Yamhill OR
3      dc8a365077241bb5cd5ccd305136b05e Del Mar CA
4      aabb64a116e83fdc4befc1fbab1663f9  Needville TX
...      ...      ...      ...
9995      9499fb4de537af195d16d046b79fd20a Mount Holly VT
9996      c09a841117fa81b5c8e19afec2760104 Clarksville TN
9997      9c41f212d1e04dca84445019bbc9b41c Mobeetie TX
9998      3e1f269b40c235a1038863ecf6b7a0df Carrollton GA
9999      0ea683a03acd544aefe8388aab16176 Clarkesville GA
```

```

                                County Zip Lat Lng ... MonthlyCharge \
0      Prince of Wales-Hyder 99927 56.25100 -133.37571 ... 172.455519
1      Ogemaw 48661 44.32893 -84.24080 ... 242.632554
2      Yamhill 97148 45.35589 -123.24657 ... 159.947583
3      San Diego 92014 32.96687 -117.24798 ... 119.956840
4      Fort Bend 77461 29.38012 -95.80673 ... 149.948316
...      ...      ...      ...      ...      ...
9995      Rutland 5758 43.43391 -72.78734 ... 159.979400
9996      Montgomery 37042 36.56907 -87.41694 ... 207.481100
9997      Wheeler 79061 35.52039 -100.44180 ... 169.974100
9998      Carroll 30117 33.58016 -85.13241 ... 252.624000
9999      Habersham 30523 34.70783 -83.53648 ... 217.484000
```

```

Bandwidth_GB_Year Item1 Item2 Item3 Item4 Item5 Item6 Item7 Item8
0      904.536110      5      5      5      3      4      4      3      4
1      800.982766      3      4      3      3      4      3      4      4
2      2054.706961      4      4      2      4      4      3      3      3
3      2164.579412      4      4      4      2      5      4      3      3
4      271.493436      4      4      4      3      4      4      4      5
...      ...      ...      ...      ...      ...      ...      ...
9995      6511.252601      3      2      3      3      4      3      2      3
9996      5695.951810      4      5      5      4      4      5      2      5
9997      4159.305799      4      4      4      4      4      4      4      5
9998      6468.456752      4      4      6      4      3      3      5      4
9999      5857.586167      2      2      3      3      3      3      4      1
```

```
[10000 rows x 50 columns]>
```

```
In [4]: #df.rename dfr = dfr.rename(columns={'Item1': 'Timely response'})
```

```
In [5]: dfpca = df[['Outage_sec_perweek', 'Tenure', 'MonthlyCharge', 'Bandwidth_GB_Year', 'Item1', 'Item2', 'Item3', 'Item4', 'Item5', 'Item6', 'Item7', 'Item8']]
dfpcanormalized=(dfpca-dfpca.mean())/dfpca.std()
```

```
pca = PCA(n_components=dfpca.shape[1])
pca.fit(dfpcanormalized)
PCA(n_components=12)
dfpca2 = pd.DataFrame(pca.transform(dfpcanormalized), columns=['PC1', 'PC2', 'PC3', 'PC4',
```

```
In [6]: pd.DataFrame(dfpcanormalized).to_csv("churn_clean_normalized.csv")
```

```
In [7]: loadings=pd.DataFrame(pca.components_.T,
columns=['PC1', 'PC2', 'PC3', 'PC4', 'PC5', 'PC6', 'PC7', 'PC8', 'PC9', 'PC10', 'PC11', 'PC12'],
index=dfpcanormalized.columns)
loadings
```

```
Out[7]:
```

	PC1	PC2	PC3	PC4	PC5	PC6	PC7	
Outage_sec_perweek	-0.017428	0.004010	-0.014271	0.707528	-0.700306	-0.085860	0.017491	-0.009
Tenure	-0.016148	0.702882	-0.061382	-0.044803	-0.040867	0.007139	-0.004091	0.006
MonthlyCharge	0.000901	0.040213	-0.008127	0.700704	0.710047	-0.006815	0.014660	-0.015
Bandwidth_GB_Year	-0.016646	0.704158	-0.061496	0.000301	0.004703	0.009129	-0.002994	0.005
Item1	0.458823	0.030630	0.281317	0.011999	0.010224	-0.070293	-0.119133	-0.045
Item2	0.433927	0.038135	0.282943	0.020531	-0.010460	-0.110173	-0.168688	-0.066
Item3	0.400496	0.035273	0.281071	-0.015650	-0.007066	-0.174392	-0.254671	-0.147
Item4	0.145814	-0.038722	-0.568735	-0.020574	0.012388	-0.170738	-0.483552	-0.442
Item5	-0.175661	0.055808	0.588414	-0.003980	-0.004810	0.137971	0.058711	-0.208
Item6	0.405127	-0.006603	-0.183885	0.000066	0.005222	-0.061882	0.064847	0.758
Item7	0.358286	0.002014	-0.181958	-0.038178	0.005552	-0.177723	0.806397	-0.379
Item8	0.308760	-0.013666	-0.131801	0.060848	-0.056694	0.928091	-0.014057	-0.112

```
In [8]: print(loadings)
```

	PC1	PC2	PC3	PC4	PC5	\
Outage_sec_perweek	-0.017428	0.004010	-0.014271	0.707528	-0.700306	
Tenure	-0.016148	0.702882	-0.061382	-0.044803	-0.040867	
MonthlyCharge	0.000901	0.040213	-0.008127	0.700704	0.710047	
Bandwidth_GB_Year	-0.016646	0.704158	-0.061496	0.000301	0.004703	
Item1	0.458823	0.030630	0.281317	0.011999	0.010224	
Item2	0.433927	0.038135	0.282943	0.020531	-0.010460	
Item3	0.400496	0.035273	0.281071	-0.015650	-0.007066	
Item4	0.145814	-0.038722	-0.568735	-0.020574	0.012388	
Item5	-0.175661	0.055808	0.588414	-0.003980	-0.004810	
Item6	0.405127	-0.006603	-0.183885	0.000066	0.005222	
Item7	0.358286	0.002014	-0.181958	-0.038178	0.005552	
Item8	0.308760	-0.013666	-0.131801	0.060848	-0.056694	

	PC6	PC7	PC8	PC9	PC10	\
Outage_sec_perweek	-0.085860	0.017491	-0.009299	0.012963	0.018676	
Tenure	0.007139	-0.004091	0.006191	-0.007068	-0.003703	
MonthlyCharge	-0.006815	0.014660	-0.015930	-0.000782	0.021294	
Bandwidth_GB_Year	0.009129	-0.002994	0.005920	-0.006864	-0.002301	
Item1	-0.070293	-0.119133	-0.045767	0.026016	-0.241054	
Item2	-0.110173	-0.168688	-0.066098	0.076924	-0.590979	
Item3	-0.174392	-0.254671	-0.147274	-0.398166	0.672440	
Item4	-0.170738	-0.483552	-0.442260	0.431741	0.088821	
Item5	0.137971	0.058711	-0.208105	0.693275	0.267613	
Item6	-0.061882	0.064847	0.758796	0.400822	0.231202	
Item7	-0.177723	0.806397	-0.379584	0.069242	0.068276	
Item8	0.928091	-0.014057	-0.112587	-0.046051	0.044924	

	PC11	PC12
Outage_sec_perweek	0.013044	0.000114
Tenure	0.006657	-0.705717
MonthlyCharge	-0.011640	-0.045372
Bandwidth_GB_Year	0.002403	0.707032
Item1	0.793182	0.002251
Item2	-0.574045	-0.001056
Item3	-0.177177	0.000132
Item4	0.018565	0.000734
Item5	-0.041482	-0.000553
Item6	-0.063788	-0.000334
Item7	-0.040348	0.000502
Item8	-0.042713	-0.001487

```
In [9]: print("Variance explained by all PC =", sum (pca.explained_variance_ratio_ * 100))
```

Variance explained by all PC = 100.0

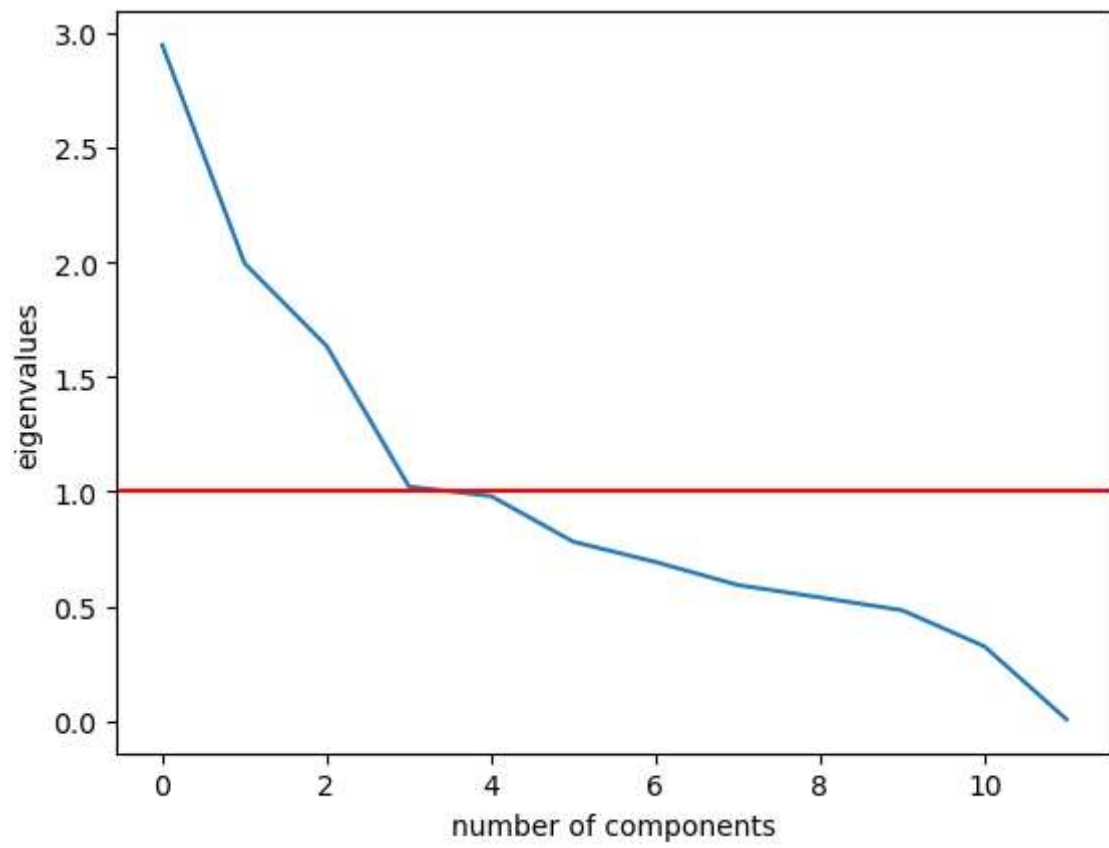
```
In [10]: varex = pca.explained_variance_ratio_ * 100
varidf = pd.DataFrame(varex.round(2), columns= ["Variance per PC"], index = ['PC1','PC2',
varidf
```

Out[10]: **Variance per PC**

PC1	24.57
PC2	16.63
PC3	13.62
PC4	8.52
PC5	8.17
PC6	6.51
PC7	5.77
PC8	4.94
PC9	4.49
PC10	4.02
PC11	2.71
PC12	0.05

```
In [11]: print("Variance sum of PC1, PC2 and PC3: ", np.cumsum(pca.explained_variance_ratio_ *  
Variance sum of PC1, PC2 and PC3: 54.82299383765756
```

```
In [12]: cov_matrix = np.dot(dfpcanormalized.T, dfpcanormalized) / dfpca.shape[0]  
eigenvalues = [np.dot(eigenvector.T, np.dot(cov_matrix, eigenvector)) for eigenvector  
plt.plot(eigenvalues)  
plt.xlabel('number of components')  
plt.ylabel('eigenvalues')  
plt.axhline(y=1, color='red')  
plt.show()
```



In []: