

D206

August 19, 2021

```
[2]: import pandas as pd
import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
from sklearn.decomposition import PCA
```

```
[3]: pd.options.mode.chained_assignment = None # default='warn' ---- ignores false_
      ↳ warning for database writing
```

```
[4]: # write csv into datafile
df = pd.read_csv('medical_raw_data.csv')
print(df.head())
```

	Unnamed: 0	CaseOrder	Customer_id	Interaction	\
0	1	1	C412403	8cd49b13-f45a-4b47-a2bd-173ffa932c2f	
1	2	2	Z919181	d2450b70-0337-4406-bdbb-bc1037f1734c	
2	3	3	F995323	a2057123-abf5-4a2c-abad-8ffe33512562	
3	4	4	A879973	1dec528d-eb34-4079-adce-0d7a40e82205	
4	5	5	C544523	5885f56b-d6da-43a3-8760-83583af94266	

	UID	City	State	County	Zip	\
0	3a83ddb66e2ae73798bdf1d705dc0932	Eva	AL	Morgan	35621	
1	176354c5eef714957d486009feabf195	Marianna	FL	Jackson	32446	
2	e19a0fa00aeda885b8a436757e889bc9	Sioux Falls	SD	Minnehaha	57110	
3	cd17d7b6d152cb6f23957346d11c3f07	New Richland	MN	Waseca	56072	
4	d2f0425877b10ed6bb381f3e2579424a	West Point	VA	King William	23181	

	Lat	...	TotalCharge	Additional_charges	Item1	Item2	Item3	Item4	\
0	34.34960	...	3191.048774	17939.403420	3	3	2	2	
1	30.84513	...	4214.905346	17612.998120	3	4	3	4	
2	43.54321	...	2177.586768	17505.192460	2	4	4	4	
3	43.89744	...	2465.118965	12993.437350	3	5	5	3	
4	37.59894	...	1885.655137	3716.525786	2	1	3	3	

	Item5	Item6	Item7	Item8
0	4	3	3	4
1	4	4	3	3
2	3	4	3	3

```
3      4      5      5      5
4      5      3      4      3
```

[5 rows x 53 columns]

```
[5]: # checks for duplicates
data = df.loc[df.duplicated()]
print(data)
```

Empty DataFrame

Columns: [Unnamed: 0, CaseOrder, Customer_id, Interaction, UID, City, State, County, Zip, Lat, Lng, Population, Area, Timezone, Job, Children, Age, Education, Employment, Income, Marital, Gender, ReAdmis, VitD_levels, Doc_visits, Full_meals_eaten, VitD_supp, Soft_drink, Initial_admin, HighBlood, Stroke, Complication_risk, Overweight, Arthritis, Diabetes, Hyperlipidemia, BackPain, Anxiety, Allergic_rhinitis, Reflux_esophagitis, Asthma, Services, Initial_days, TotalCharge, Additional_charges, Item1, Item2, Item3, Item4, Item5, Item6, Item7, Item8]
Index: []

[0 rows x 53 columns]

```
[6]: newdf = df[['ReAdmis', 'HighBlood', 'Stroke', 'Overweight', 'Arthritis', 'Diabetes', 'Hyperlipidemia', 'BackPain', 'Anxiety', 'Allergic_rhinitis', 'Reflux_esophagitis', 'Asthma']].copy()
print(newdf.head())
```

	ReAdmis	HighBlood	Stroke	Overweight	Arthritis	Diabetes	Hyperlipidemia	\
0	No	Yes	No	0.0	Yes	Yes	No	
1	No	Yes	No	1.0	No	No	No	
2	No	Yes	No	1.0	No	Yes	No	
3	No	No	Yes	0.0	Yes	No	No	
4	No	No	No	0.0	No	No	Yes	

	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	Asthma
0	Yes	1.0	Yes	No	Yes
1	No	NaN	No	Yes	No
2	No	NaN	No	No	No
3	No	NaN	No	Yes	Yes
4	No	0.0	Yes	No	No

```
[7]: # checks the number of null values.
print(newdf.isnull().sum())
```

```
ReAdmis      0
HighBlood    0
```

```

Stroke          0
Overweight      982
Arthritis       0
Diabetes        0
Hyperlipidemia  0
BackPain        0
Anxiety         984
Allergic_rhinitis  0
Reflux_esophagitis  0
Asthma          0
dtype: int64

```

```
[8]: newdfNoNull = newdf.dropna() #create datafile with no null values
newdfNoNull.isnull().sum()
```

```
[8]: ReAdmis          0
HighBlood           0
Stroke              0
Overweight          0
Arthritis           0
Diabetes            0
Hyperlipidemia      0
BackPain            0
Anxiety             0
Allergic_rhinitis   0
Reflux_esophagitis  0
Asthma              0
dtype: int64

```

```
[9]: newdfNoNull.shape
```

```
[9]: (8126, 12)
```

```
[10]: # create a database with all values being 1 or 0 instead of Yes/No
di = {'Yes': 1, 'No': 0}
noNullwith1_0 = newdfNoNull.replace({'ReAdmis': di, 'HighBlood': di, 'Stroke': di,
    ↳di, 'Overweight': di, 'Arthritis': di, 'Diabetes': di, 'Hyperlipidemia': di,
    ↳di, 'BackPain': di, 'Anxiety': di, 'Allergic_rhinitis': di, 'Reflux_esophagitis': di,
    ↳di, 'Asthma': di})
noNullwith1_0.head()
```

```
[10]:   ReAdmis  HighBlood  Stroke  Overweight  Arthritis  Diabetes  \
0         0          1        0          0.0          1          1
4         0          0        0          0.0          0          0
5         0          0        0          1.0          1          1
6         0          1        0          1.0          1          1
7         0          0        0          1.0          0          0

```

	Hyperlipidemia	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	\
0	0	1	1.0	1	0	
4	1	0	0.0	1	0	
5	0	1	0.0	1	0	
6	1	1	1.0	0	1	
7	0	0	0.0	0	0	

	Asthma
0	1
4	0
5	0
6	0
7	0

```
[11]: noNullwith1_0.dtypes
```

```
[11]: ReAdmis          int64
      HighBlood       int64
      Stroke          int64
      Overweight      float64
      Arthritis        int64
      Diabetes         int64
      Hyperlipidemia   int64
      BackPain         int64
      Anxiety          float64
      Allergic_rhinitis int64
      Reflux_esophagitis int64
      Asthma           int64
      dtype: object
```

```
[13]: # create 2 databases. 1 is patients that have been readmitted, 0 are first time
      ↪admitted patients
      readmissionDF_0 = noNullwith1_0.groupby(noNullwith1_0['ReAdmis']).get_group(0)
      readmissionDF_0.head()
```

	ReAdmis	HighBlood	Stroke	Overweight	Arthritis	Diabetes	\
0	0	1	0	0.0	1	1	
4	0	0	0	0.0	0	0	
5	0	0	0	1.0	1	1	
6	0	1	0	1.0	1	1	
7	0	0	0	1.0	0	0	

	Hyperlipidemia	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	\
0	0	1	1.0	1	0	
4	1	0	0.0	1	0	
5	0	1	0.0	1	0	

6	1	1	1.0	0	1
7	0	0	0.0	0	0

	Asthma
0	1
4	0
5	0
6	0
7	0

```
[14]: readmissionDF_1 = noNullwith1_0.groupby(noNullwith1_0['ReAdmis']).get_group(1)
      readmissionDF_1.head()
```

```
[14]:
```

	ReAdmis	HighBlood	Stroke	Overweight	Arthritis	Diabetes	\
5000	1	1	0	0.0	0	0	
5005	1	0	0	1.0	0	0	
5008	1	1	0	0.0	1	0	
5009	1	0	0	0.0	0	0	
5010	1	0	0	1.0	1	0	

	Hyperlipidemia	BackPain	Anxiety	Allergic_rhinitis	\
5000	1	0	0.0	0	
5005	1	0	0.0	0	
5008	0	0	0.0	1	
5009	0	0	1.0	0	
5010	0	0	1.0	0	

	Reflux_esophagitis	Asthma
5000	0	0
5005	1	0
5008	1	0
5009	0	0
5010	0	0

```
[15]: # readmissionDF_1 is database of patients readmitted
      # readmissionDF_0 is database of patients for the first time
      readmissionDF_0 = readmissionDF_0.drop(['ReAdmis'], axis=1)
      readmissionDF_0['sum'] = readmissionDF_0.sum(axis=1)
      readmissionDF_1 = readmissionDF_1.drop(['ReAdmis'], axis=1)
      readmissionDF_1['sum'] = readmissionDF_1.sum(axis=1)

      median0 = readmissionDF_0['sum'].median()
      mean0 = readmissionDF_0['sum'].mean()
      mode0 = readmissionDF_0['sum'].mode().iloc[0]

      median1 = readmissionDF_1['sum'].median()
      mean1 = readmissionDF_1['sum'].mean()
```

```
mode1 = readmissionDF_1['sum'].mode().iloc[0]
```

```
[16]: print(readmissionDF_0.head())
print('median0: ', median0)
print('mean0: ', mean0)
print('mode0: ', mode0)

print(readmissionDF_1.head())
print('median1: ', median1)
print('mean1: ', mean1)
print('mode1: ', mode1)
```

	HighBlood	Stroke	Overweight	Arthritis	Diabetes	Hyperlipidemia	\
0	1	0	0.0	1	1	0	
4	0	0	0.0	0	0	1	
5	0	0	1.0	1	1	0	
6	1	0	1.0	1	1	1	
7	0	0	1.0	0	0	0	

	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	Asthma	sum
0	1	1.0		1	0	1 7.0
4	0	0.0		1	0	0 2.0
5	1	0.0		1	0	0 5.0
6	1	1.0		0	1	0 8.0
7	0	0.0		0	0	0 1.0

median0: 4.0

mean0: 4.116048903551329

mode0: 4.0

	HighBlood	Stroke	Overweight	Arthritis	Diabetes	Hyperlipidemia	\
5000	1	0	0.0	0	0	1	
5005	0	0	1.0	0	0	1	
5008	1	0	0.0	1	0	0	
5009	0	0	0.0	0	0	0	
5010	0	0	1.0	1	0	0	

	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	Asthma	sum
5000	0	0.0		0	0	0 2.0
5005	0	0.0		0	1	0 3.0
5008	0	0.0		1	1	0 4.0
5009	0	1.0		0	0	0 1.0
5010	0	1.0		0	0	0 3.0

median1: 4.0

mean1: 4.13757147662294

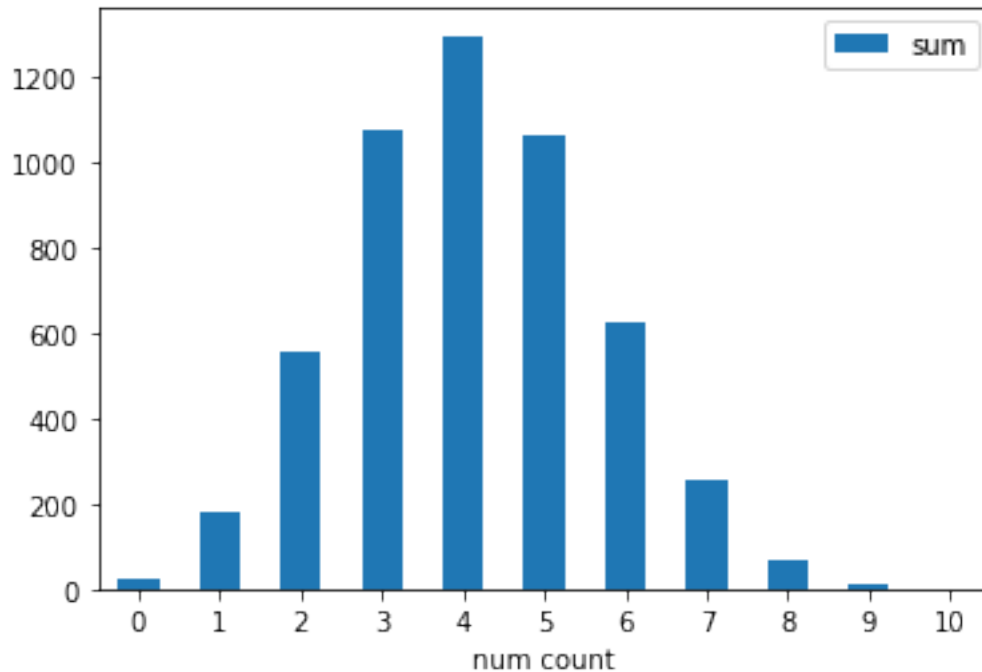
mode1: 4.0

```
[17]: # creates bar chart of patients for FIRST TIME in hospital
```

```

counts0 = readmissionDF_0['sum'].value_counts() # gets counts of how many
→patients have how many conditions
counts0 = pd.DataFrame(counts0) # converts data back into dataframe
counts0['num count'] = [4,3,5,6,2,7,1,8,0,9,10] #order for num. better to make
→bell curve.
counts0 = counts0.sort_values(by=['num count']) # sort values by count to make
→an easier to understand bar graph
ax = counts0.plot.bar(x='num count', y='sum', rot=0) # creates the bar chart

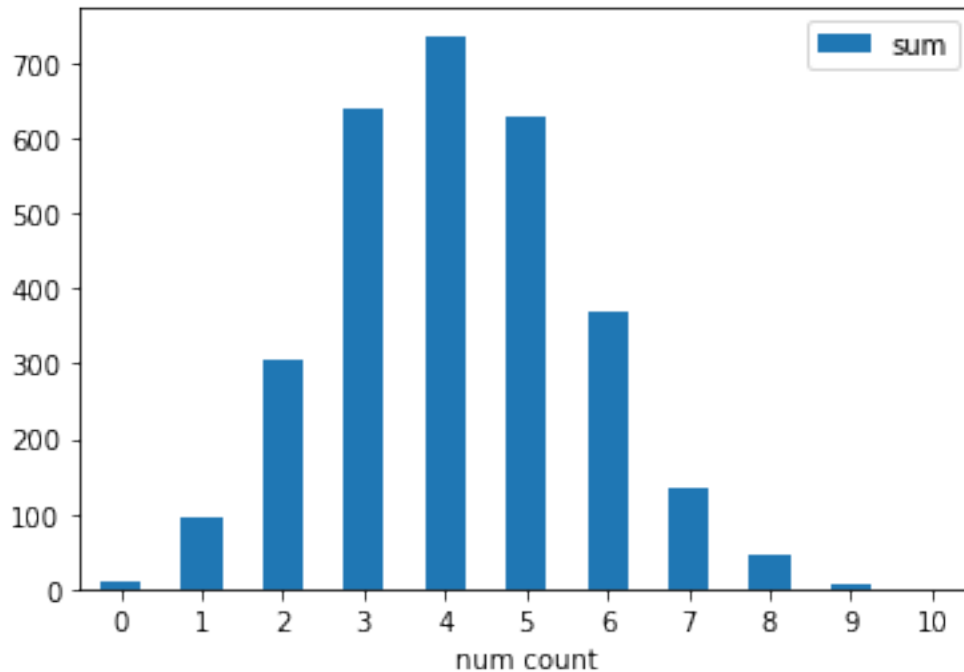
```



```

[18]: # creates bar chart of patients for READMISSION in hospital
counts1 = readmissionDF_1['sum'].value_counts() # gets counts of how many
→patients have how many conditions
counts1 = pd.DataFrame(counts1) # converts data back into dataframe
counts1['num count'] = [4,3,5,6,2,7,1,8,0,9,10] #order for num. better to make
→bell curve.
counts1 = counts1.sort_values(by=['num count']) # sort values by count to make
→an easier to understand bar graph
ax = counts1.plot.bar(x='num count', y='sum', rot=0) # creates the bar chart

```



```
[19]: # PCA analysis start for patients on first visit
dataNumeric_0 = readmissionDF_0[['HighBlood', 'Stroke', 'Overweight',
    ↳ 'Arthritis', 'Diabetes', 'Hyperlipidemia', 'BackPain', 'Anxiety',
    ↳ 'Allergic_rhinitis', 'Reflux_esophagitis', 'Asthma']]
pcsNames_0 = []
for i, col in enumerate(dataNumeric_0.columns):
    pcsNames_0.append('PC'+str(i+1))
print(pcsNames_0)
normal0 = (dataNumeric_0 - dataNumeric_0.mean())/(dataNumeric_0.std())
print(normal0)
pca = PCA(n_components=normal0.shape[1])
pca.fit(normal0)
data_pca = pd.DataFrame(pca.transform(dataNumeric_0), columns = pcsNames_0)
print(data_pca)
```

```
['PC1', 'PC2', 'PC3', 'PC4', 'PC5', 'PC6', 'PC7', 'PC8', 'PC9', 'PC10', 'PC11']
      HighBlood  Stroke  Overweight  Arthritis  Diabetes  Hyperlipidemia \
0      1.200791 -0.496432 -1.561873   1.347974   1.621223    -0.709612
4     -0.832622 -0.496432 -1.561873  -0.741710  -0.616699     1.408947
5     -0.832622 -0.496432   0.640133   1.347974   1.621223    -0.709612
6      1.200791 -0.496432   0.640133   1.347974   1.621223     1.408947
7     -0.832622 -0.496432   0.640133  -0.741710  -0.616699    -0.709612
...      ...      ...      ...      ...      ...      ...
9981  -0.832622  2.013984   0.640133  -0.741710  -0.616699     1.408947
9985  -0.832622  2.013984   0.640133  -0.741710  -0.616699     1.408947
```


9986	1.200791	-0.496432	-1.561873	-0.741710	-0.616699	1.408947
9993	1.200791	-0.496432	0.640133	-0.741710	-0.616699	1.408947
9994	-0.832622	-0.496432	-1.561873	1.347974	-0.616699	-0.709612

	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	Asthma
0	1.211457	1.451096	1.227701	-0.844370	1.557474
4	-0.825292	-0.689001	1.227701	-0.844370	-0.641941
5	1.211457	-0.689001	1.227701	-0.844370	-0.641941
6	1.211457	1.451096	-0.814372	1.184086	-0.641941
7	-0.825292	-0.689001	-0.814372	-0.844370	-0.641941
...
9981	1.211457	1.451096	1.227701	-0.844370	-0.641941
9985	-0.825292	-0.689001	-0.814372	-0.844370	-0.641941
9986	1.211457	-0.689001	-0.814372	-0.844370	1.557474
9993	-0.825292	-0.689001	-0.814372	-0.844370	-0.641941
9994	-0.825292	-0.689001	-0.814372	1.184086	1.557474

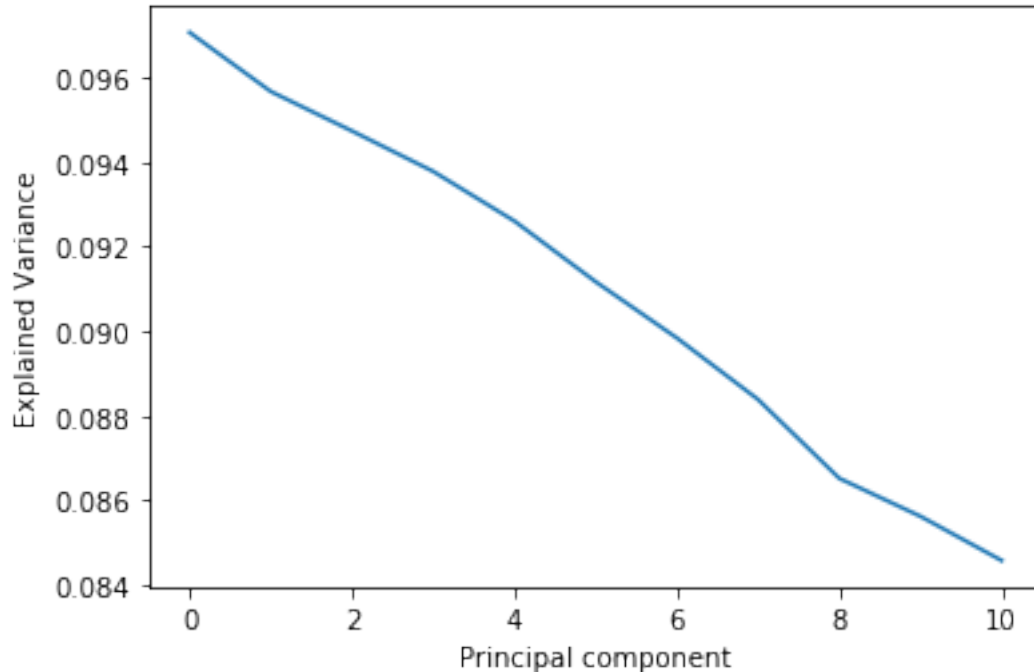
[5153 rows x 11 columns]

	PC1	PC2	PC3	PC4	PC5	PC6	PC7	\
0	0.196327	-0.410238	1.383062	-1.166821	0.081773	1.399893	0.805733	
1	0.885717	0.072989	0.302157	0.195194	-0.139855	0.665885	-0.424035	
2	0.142324	-0.084869	0.967830	-0.652045	0.428960	1.157124	-0.335043	
3	-0.250186	-0.040082	1.225882	-0.933982	0.031410	1.849890	0.696928	
4	-0.134692	0.173090	0.433226	-0.315596	0.331322	-0.213724	-0.328156	
...	
5148	0.647965	0.804272	1.287304	-0.047820	-0.227835	0.628963	0.439061	
5149	0.151117	0.666257	0.377895	0.257562	0.642372	0.251795	-0.003250	
5150	0.348480	0.936948	0.720583	-0.641555	0.011431	0.945741	0.269631	
5151	0.289876	0.273262	0.977830	-0.152919	0.487021	0.402835	-0.234152	
5152	-0.998474	-0.151875	-0.148067	-0.761090	-0.055331	0.897511	-0.196845	

	PC8	PC9	PC10	PC11
0	0.723223	-0.076267	0.507427	-0.341883
1	0.116285	-0.071702	0.358126	0.539531
2	-0.162546	-1.108161	0.490695	-0.693010
3	-1.204631	-0.267982	0.118820	0.338873
4	-0.503723	-0.335406	0.111907	0.150721
...
5148	-0.466781	-0.833051	1.198036	0.535182
5149	-0.822471	-0.162355	0.842160	0.660316
5150	0.044584	1.033357	0.176231	0.041574
5151	-0.907958	0.517633	-0.005269	0.561348
5152	0.364037	0.192672	0.468315	0.377495

[5153 rows x 11 columns]

```
[20]: sns.lineplot(y = pca.explained_variance_ratio_, x=range(0,11))
plt.ylabel('Explained Variance')
plt.xlabel('Principal component')
plt.show()
```



```
[22]: for pc, var in zip(pcsNames_0, np.cumsum(pca.explained_variance_ratio_)):
        print(pc,var)
rotation_0 = pd.DataFrame(pca.components_.T, columns = pcsNames_0, index =
    ↳dataNumeric_0.columns)
print(rotation_0)
dataReduced_0 = data_pca.iloc[:,0:11]
print(dataReduced_0)
```

```
PC1 0.09706998511449275
PC2 0.19274243800950047
PC3 0.2874806766911285
PC4 0.38126295759967577
PC5 0.47387133432324124
PC6 0.5650443257884363
PC7 0.6548873137785118
PC8 0.743275560967274
PC9 0.8297980863917527
PC10 0.9154216572356548
PC11 0.9999999999999999
```

```
PC1      PC2      PC3      PC4      PC5 \
```

HighBlood	-0.099398	-0.106521	0.674979	0.112587	0.188947
Stroke	-0.238156	0.286473	0.075044	0.523068	0.344299
Overweight	-0.134692	0.173090	0.433226	-0.315596	0.331322
Arthritis	-0.331963	-0.527839	0.006855	-0.107834	-0.014810
Diabetes	0.227078	-0.173542	-0.144470	-0.217839	0.570582
Hyperlipidemia	0.523966	0.206694	-0.130374	0.050090	-0.033249
BackPain	0.020150	0.577127	0.239689	-0.155880	-0.351528
Anxiety	0.114947	-0.305407	0.237190	-0.294607	-0.412073
Allergic_rhinitis	0.361751	-0.133705	0.432531	0.145104	-0.106606
Reflux_esophagitis	-0.570273	0.116316	-0.091211	-0.004903	-0.247782
Asthma	-0.096238	0.259648	-0.063710	-0.648352	0.207261

	PC6	PC7	PC8	PC9	PC10	PC11
HighBlood	0.169390	0.169206	0.008102	0.551583	-0.335402	0.025932
Stroke	0.018350	0.400108	0.093589	-0.128405	0.512027	0.124900
Overweight	-0.213724	-0.328156	-0.503723	-0.335406	0.111907	0.150721
Arthritis	0.431857	-0.170463	-0.148489	0.100834	0.505685	-0.309516
Diabetes	0.406285	0.350731	0.010071	-0.369369	-0.289468	-0.126582
Hyperlipidemia	0.447169	-0.075202	-0.412337	0.301456	0.218226	0.384694
BackPain	0.313990	0.161678	-0.049029	-0.131063	0.022671	-0.562470
Anxiety	-0.155537	0.629466	-0.123904	-0.166476	0.193305	0.282499
Allergic_rhinitis	0.218716	-0.348833	0.528623	-0.373158	0.139900	0.154836
Reflux_esophagitis	0.450461	-0.040331	0.014678	-0.219542	-0.308105	0.493594
Asthma	0.015193	0.013949	0.497849	0.311380	0.270735	0.193418

	PC1	PC2	PC3	PC4	PC5	PC6	PC7 \
0	0.196327	-0.410238	1.383062	-1.166821	0.081773	1.399893	0.805733
1	0.885717	0.072989	0.302157	0.195194	-0.139855	0.665885	-0.424035
2	0.142324	-0.084869	0.967830	-0.652045	0.428960	1.157124	-0.335043
3	-0.250186	-0.040082	1.225882	-0.933982	0.031410	1.849890	0.696928
4	-0.134692	0.173090	0.433226	-0.315596	0.331322	-0.213724	-0.328156
...
5148	0.647965	0.804272	1.287304	-0.047820	-0.227835	0.628963	0.439061
5149	0.151117	0.666257	0.377895	0.257562	0.642372	0.251795	-0.003250
5150	0.348480	0.936948	0.720583	-0.641555	0.011431	0.945741	0.269631
5151	0.289876	0.273262	0.977830	-0.152919	0.487021	0.402835	-0.234152
5152	-0.998474	-0.151875	-0.148067	-0.761090	-0.055331	0.897511	-0.196845

	PC8	PC9	PC10	PC11
0	0.723223	-0.076267	0.507427	-0.341883
1	0.116285	-0.071702	0.358126	0.539531
2	-0.162546	-1.108161	0.490695	-0.693010
3	-1.204631	-0.267982	0.118820	0.338873
4	-0.503723	-0.335406	0.111907	0.150721
...
5148	-0.466781	-0.833051	1.198036	0.535182
5149	-0.822471	-0.162355	0.842160	0.660316
5150	0.044584	1.033357	0.176231	0.041574
5151	-0.907958	0.517633	-0.005269	0.561348

5152 0.364037 0.192672 0.468315 0.377495

[5153 rows x 11 columns]

```
[23]: # PCA analysis start for patients readmission
dataNumeric_1 = readmissionDF_1[['HighBlood', 'Stroke', 'Overweight',
    ↳'Arthritis', 'Diabetes', 'Hyperlipidemia', 'BackPain', 'Anxiety',
    ↳'Allergic_rhinitis', 'Reflux_esophagitis', 'Asthma']]
pcsNames_1 = []
for i, col in enumerate(dataNumeric_1.columns):
    pcsNames_1.append('PC'+str(i+1))
print(pcsNames_1)
normal1 = (dataNumeric_1 - dataNumeric_1.mean())/(dataNumeric_1.std())
print(normal1)
pca = PCA(n_components=normal1.shape[1])
pca.fit(normal1)
data_pca = pd.DataFrame(pca.transform(dataNumeric_1), columns = pcsNames_1)
print(data_pca)
```

['PC1', 'PC2', 'PC3', 'PC4', 'PC5', 'PC6', 'PC7', 'PC8', 'PC9', 'PC10', 'PC11']

	HighBlood	Stroke	Overweight	Arthritis	Diabetes	Hyperlipidemia	\
5000	1.182812	-0.504326	-1.546942	-0.748626	-0.611326	1.388674	
5005	-0.845158	-0.504326	0.646219	-0.748626	-0.611326	1.388674	
5008	1.182812	-0.504326	-1.546942	1.335331	-0.611326	-0.719869	
5009	-0.845158	-0.504326	-1.546942	-0.748626	-0.611326	-0.719869	
5010	-0.845158	-0.504326	0.646219	1.335331	-0.611326	-0.719869	
...	
9992	1.182812	-0.504326	-1.546942	1.335331	1.635239	1.388674	
9996	1.182812	-0.504326	0.646219	1.335331	1.635239	-0.719869	
9997	1.182812	-0.504326	0.646219	-0.748626	-0.611326	-0.719869	
9998	-0.845158	-0.504326	0.646219	-0.748626	-0.611326	-0.719869	
9999	-0.845158	-0.504326	0.646219	1.335331	-0.611326	1.388674	

	BackPain	Anxiety	Allergic_rhinitis	Reflux_esophagitis	Asthma
5000	-0.855139	-0.692057	-0.792988	-0.856318	-0.630042
5005	-0.855139	-0.692057	-0.792988	1.167397	-0.630042
5008	-0.855139	-0.692057	1.260629	1.167397	-0.630042
5009	-0.855139	1.444481	-0.792988	-0.856318	-0.630042
5010	-0.855139	1.444481	-0.792988	-0.856318	-0.630042
...
9992	-0.855139	-0.692057	1.260629	1.167397	1.586662
9996	-0.855139	-0.692057	-0.792988	-0.856318	1.586662
9997	-0.855139	1.444481	1.260629	-0.856318	-0.630042
9998	1.169007	-0.692057	-0.792988	-0.856318	-0.630042
9999	-0.855139	-0.692057	1.260629	-0.856318	-0.630042

[2973 rows x 11 columns]

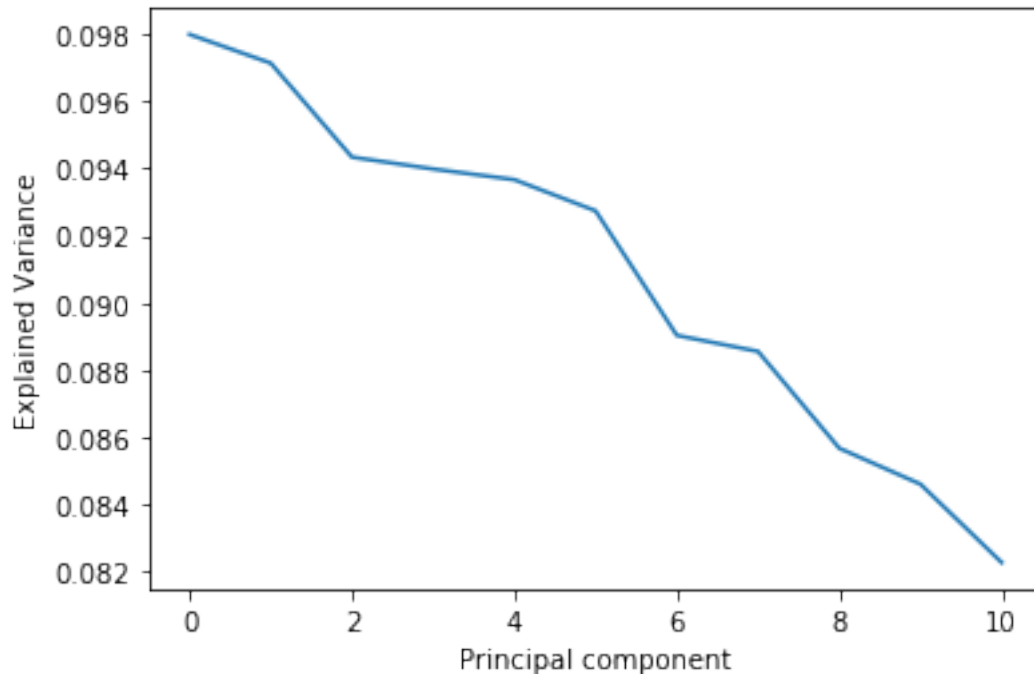
PC1 PC2 PC3 PC4 PC5 PC6 PC7 \

0	-0.057924	-0.536595	-0.160879	-0.192670	-0.311028	0.853253	0.026772
1	0.330288	-0.978442	0.151272	0.022804	-0.293706	0.237988	-0.141919
2	-0.309435	-1.064048	-0.240404	0.389100	0.870112	0.248398	-0.882834
3	-0.095368	0.465004	-0.382503	-0.310448	0.189645	0.185412	-0.424486
4	-0.434822	-0.224128	-0.174833	-1.086455	0.421219	-0.416535	-0.686364
...
2968	0.008291	-1.462838	0.050304	0.002264	1.557662	1.137754	0.038704
2969	-0.864654	-1.065429	0.040172	-0.901392	1.205513	0.352058	0.263750
2970	-1.197910	-0.310507	-0.097749	-0.022320	0.375478	0.240810	-0.956485
2971	-0.525658	-0.313854	-0.287255	-0.000009	-0.093694	-0.738721	0.426110
2972	-0.200811	-1.120639	0.705484	-0.540291	0.280253	-0.294110	-0.309866

	PC8	PC9	PC10	PC11
0	0.397912	0.379004	0.537984	0.478336
1	0.760394	0.824330	-0.095291	-0.693047
2	0.070953	0.684389	-0.157267	0.683017
3	0.004348	0.300422	0.196920	-0.390397
4	0.334234	0.626948	-0.044357	-0.470085
...
2968	0.260376	0.959070	0.222987	0.312844
2969	0.805522	-0.043838	-0.069991	0.025584
2970	0.523069	0.557283	0.798303	-0.351077
2971	0.651737	0.458581	0.171004	-0.377235
2972	0.068328	1.157999	0.552872	-0.043188

[2973 rows x 11 columns]

```
[24]: sns.lineplot(y = pca.explained_variance_ratio_, x=range(0,11))
plt.ylabel('Explained Variance')
plt.xlabel('Principal component')
plt.show()
```



```
[25]: for pc, var in zip(pcsNames_1, np.cumsum(pca.explained_variance_ratio_)):
      print(pc,var)

rotation_1 = pd.DataFrame(pca.components_.T, columns = pcsNames_1, index =
↳dataNumeric_1.columns)
print(rotation_1)
dataReduced_1 = data_pca.iloc[:,0:11]
print(dataReduced_1)
```

```
PC1 0.097998633008865
PC2 0.19514090128863842
PC3 0.28948311426047973
PC4 0.38347117397745756
PC5 0.4771416609844364
PC6 0.5698834012351238
PC7 0.6589150354964439
PC8 0.7474688772361852
PC9 0.8331380215196443
PC10 0.9177363790545597
PC11 1.0
```

	PC1	PC2	PC3	PC4	PC5 \
HighBlood	-0.450425	-0.257051	-0.309543	0.034390	-0.012320
Stroke	0.044978	0.502939	0.438284	0.066247	0.137785
Overweight	-0.398258	-0.366498	0.245146	-0.209038	-0.149234
Arthritis	0.058804	-0.322634	-0.037476	-0.566969	0.380808

Diabetes	0.332395	-0.108234	0.008130	-0.071723	0.658177
Hyperlipidemia	0.392501	-0.279544	0.148663	-0.227060	-0.298709
BackPain	-0.127401	0.052644	-0.532401	0.209029	0.055540
Anxiety	-0.095368	0.465004	-0.382503	-0.310448	0.189645
Allergic_rhinitis	-0.253859	-0.151963	0.349151	0.462776	0.347387
Reflux_esophagitis	0.336045	-0.332400	-0.242537	0.458903	0.154236
Asthma	-0.407170	-0.011012	0.133915	-0.088053	0.328082

	PC6	PC7	PC8	PC9	PC10	PC11
HighBlood	0.458951	-0.184569	0.342062	-0.133021	0.164508	0.476891
Stroke	0.011828	-0.013496	0.548261	0.291886	-0.183095	0.331556
Overweight	-0.317087	-0.088101	0.494066	0.070435	0.016202	-0.472626
Arthritis	-0.284859	-0.173777	-0.164180	0.256092	-0.257479	0.392937
Diabetes	0.042533	0.198934	0.321128	-0.378257	0.362865	-0.141899
Hyperlipidemia	0.394302	0.211341	0.055850	0.512025	0.373476	0.001445
BackPain	-0.421634	0.514211	0.157671	0.388147	0.154802	0.095391
Anxiety	0.185412	-0.424486	0.004348	0.300422	0.196920	-0.390397
Allergic_rhinitis	-0.086466	-0.259329	-0.317408	0.319448	0.420673	0.035055
Reflux_esophagitis	0.160773	-0.265160	0.210478	0.241871	-0.484969	-0.221866
Asthma	0.452521	0.511263	-0.187554	0.140914	-0.356087	-0.229718

	PC1	PC2	PC3	PC4	PC5	PC6	PC7 \
0	-0.057924	-0.536595	-0.160879	-0.192670	-0.311028	0.853253	0.026772
1	0.330288	-0.978442	0.151272	0.022804	-0.293706	0.237988	-0.141919
2	-0.309435	-1.064048	-0.240404	0.389100	0.870112	0.248398	-0.882834
3	-0.095368	0.465004	-0.382503	-0.310448	0.189645	0.185412	-0.424486
4	-0.434822	-0.224128	-0.174833	-1.086455	0.421219	-0.416535	-0.686364
...
2968	0.008291	-1.462838	0.050304	0.002264	1.557662	1.137754	0.038704
2969	-0.864654	-1.065429	0.040172	-0.901392	1.205513	0.352058	0.263750
2970	-1.197910	-0.310507	-0.097749	-0.022320	0.375478	0.240810	-0.956485
2971	-0.525658	-0.313854	-0.287255	-0.000009	-0.093694	-0.738721	0.426110
2972	-0.200811	-1.120639	0.705484	-0.540291	0.280253	-0.294110	-0.309866

	PC8	PC9	PC10	PC11
0	0.397912	0.379004	0.537984	0.478336
1	0.760394	0.824330	-0.095291	-0.693047
2	0.070953	0.684389	-0.157267	0.683017
3	0.004348	0.300422	0.196920	-0.390397
4	0.334234	0.626948	-0.044357	-0.470085
...
2968	0.260376	0.959070	0.222987	0.312844
2969	0.805522	-0.043838	-0.069991	0.025584
2970	0.523069	0.557283	0.798303	-0.351077
2971	0.651737	0.458581	0.171004	-0.377235
2972	0.068328	1.157999	0.552872	-0.043188

[2973 rows x 11 columns]

```
[26]: readmissionDF_0.to_csv('readmissionDF_0_clean.csv') # clean dataset for first_
      ↪time admittance
      readmissionDF_1.to_csv('readmissionDF_1_clean.csv') # clena dataset for_
      ↪readmittence
```

```
[ ]:
```