

# parkinsons

December 1, 2023

Importing the Dependencies

```
[ ]: import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn import svm
from sklearn.metrics import accuracy_score
```

Data Collection & Analysis

```
[ ]: # loading the data from csv file to a Pandas DataFrame
parkinsons_data = pd.read_csv('/content/parkinsons.csv')
```

```
[ ]: # printing the first 5 rows of the dataframe
parkinsons_data.head()
```

```
[ ]:
      name  MDVP:Fo(Hz)  MDVP:Fhi(Hz)  MDVP:Flo(Hz)  MDVP:Jitter(%) \
0  phon_R01_S01_1      119.992      157.302        74.997      0.00784
1  phon_R01_S01_2      122.400      148.650      113.819      0.00968
2  phon_R01_S01_3      116.682      131.111      111.555      0.01050
3  phon_R01_S01_4      116.676      137.871      111.366      0.00997
4  phon_R01_S01_5      116.014      141.781      110.655      0.01284
```

```
      MDVP:Jitter(Abs)  MDVP:RAP  MDVP:PPQ  Jitter:DDP  MDVP:Shimmer  ... \
0          0.00007      0.00370      0.00554      0.01109      0.04374  ...
1          0.00008      0.00465      0.00696      0.01394      0.06134  ...
2          0.00009      0.00544      0.00781      0.01633      0.05233  ...
3          0.00009      0.00502      0.00698      0.01505      0.05492  ...
4          0.00011      0.00655      0.00908      0.01966      0.06425  ...
```

```
      Shimmer:DDA      NHR      HNR  status      RPDE      DFA  spread1 \
0      0.06545  0.02211  21.033      1  0.414783  0.815285 -4.813031
1      0.09403  0.01929  19.085      1  0.458359  0.819521 -4.075192
2      0.08270  0.01309  20.651      1  0.429895  0.825288 -4.443179
3      0.08771  0.01353  20.644      1  0.434969  0.819235 -4.117501
4      0.10470  0.01767  19.649      1  0.417356  0.823484 -3.747787
```

```
      spread2      D2      PPE
```

```

0  0.266482  2.301442  0.284654
1  0.335590  2.486855  0.368674
2  0.311173  2.342259  0.332634
3  0.334147  2.405554  0.368975
4  0.234513  2.332180  0.410335

```

[5 rows x 24 columns]

```
[ ]: # number of rows and columns in the dataframe
parkinsons_data.shape
```

```
[ ]: (195, 24)
```

```
[ ]: # getting more information about the dataset
parkinsons_data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 195 entries, 0 to 194
Data columns (total 24 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   name                  195 non-null    object
 1   MDVP:Fo(Hz)           195 non-null    float64
 2   MDVP:Fhi(Hz)          195 non-null    float64
 3   MDVP:Flo(Hz)          195 non-null    float64
 4   MDVP:Jitter(%)        195 non-null    float64
 5   MDVP:Jitter(Abs)      195 non-null    float64
 6   MDVP:RAP               195 non-null    float64
 7   MDVP:PPQ              195 non-null    float64
 8   Jitter:DDP            195 non-null    float64
 9   MDVP:Shimmer          195 non-null    float64
10  MDVP:Shimmer(dB)      195 non-null    float64
11  Shimmer:APQ3           195 non-null    float64
12  Shimmer:APQ5           195 non-null    float64
13  MDVP:APQ              195 non-null    float64
14  Shimmer:DDA            195 non-null    float64
15  NHR                    195 non-null    float64
16  HNR                    195 non-null    float64
17  status                 195 non-null    int64
18  RPDE                   195 non-null    float64
19  DFA                    195 non-null    float64
20  spread1                195 non-null    float64
21  spread2                195 non-null    float64
22  D2                     195 non-null    float64
23  PPE                    195 non-null    float64
dtypes: float64(22), int64(1), object(1)
memory usage: 36.7+ KB

```

```
[ ]: # checking for missing values in each column
parkinsons_data.isnull().sum()
```

```
[ ]: name                0
MDVP:F0(Hz)             0
MDVP:F1(Hz)             0
MDVP:F0(Hz)             0
MDVP:Jitter(%)          0
MDVP:Jitter(Abs)        0
MDVP:RAP                 0
MDVP:PPQ                 0
Jitter:DDP              0
MDVP:Shimmer             0
MDVP:Shimmer(dB)        0
Shimmer:APQ3             0
Shimmer:APQ5             0
MDVP:APQ                 0
Shimmer:DDA              0
NHR                      0
HNR                      0
status                   0
RPDE                     0
DFA                      0
spread1                  0
spread2                  0
D2                       0
PPE                      0
dtype: int64
```

```
[ ]: # getting some statistical measures about the data
parkinsons_data.describe()
```

```
[ ]:      MDVP:F0(Hz)  MDVP:F1(Hz)  MDVP:F0(Hz)  MDVP:Jitter(%)  \
count    195.000000    195.000000    195.000000    195.000000
mean     154.228641    197.104918    116.324631     0.006220
std       41.390065     91.491548     43.521413     0.004848
min       88.333000    102.145000     65.476000     0.001680
25%      117.572000    134.862500     84.291000     0.003460
50%      148.790000    175.829000    104.315000     0.004940
75%      182.769000    224.205500    140.018500     0.007365
max       260.105000    592.030000    239.170000     0.033160

      MDVP:Jitter(Abs)  MDVP:RAP  MDVP:PPQ  Jitter:DDP  MDVP:Shimmer  \
count    195.000000    195.000000    195.000000    195.000000    195.000000
mean         0.000044     0.003306     0.003446     0.009920     0.029709
std         0.000035     0.002968     0.002759     0.008903     0.018857
min         0.000007     0.000680     0.000920     0.002040     0.009540
```

25%	0.000020	0.001660	0.001860	0.004985	0.016505
50%	0.000030	0.002500	0.002690	0.007490	0.022970
75%	0.000060	0.003835	0.003955	0.011505	0.037885
max	0.000260	0.021440	0.019580	0.064330	0.119080

	MDVP:Shimmer(dB)	...	Shimmer:DDA	NHR	HNR	status \
count	195.000000	...	195.000000	195.000000	195.000000	195.000000
mean	0.282251	...	0.046993	0.024847	21.885974	0.753846
std	0.194877	...	0.030459	0.040418	4.425764	0.431878
min	0.085000	...	0.013640	0.000650	8.441000	0.000000
25%	0.148500	...	0.024735	0.005925	19.198000	1.000000
50%	0.221000	...	0.038360	0.011660	22.085000	1.000000
75%	0.350000	...	0.060795	0.025640	25.075500	1.000000
max	1.302000	...	0.169420	0.314820	33.047000	1.000000

	RPDE	DFA	spread1	spread2	D2	PPE
count	195.000000	195.000000	195.000000	195.000000	195.000000	195.000000
mean	0.498536	0.718099	-5.684397	0.226510	2.381826	0.206552
std	0.103942	0.055336	1.090208	0.083406	0.382799	0.090119
min	0.256570	0.574282	-7.964984	0.006274	1.423287	0.044539
25%	0.421306	0.674758	-6.450096	0.174351	2.099125	0.137451
50%	0.495954	0.722254	-5.720868	0.218885	2.361532	0.194052
75%	0.587562	0.761881	-5.046192	0.279234	2.636456	0.252980
max	0.685151	0.825288	-2.434031	0.450493	3.671155	0.527367

[8 rows x 23 columns]

```
[ ]: # distribution of target Variable
parkinsons_data['status'].value_counts()
```

```
[ ]: 1    147
      0    48
      Name: status, dtype: int64
```

1 -> Parkinson's Positive

0 -> Healthy

```
[ ]: import matplotlib.pyplot as plt
import seaborn as sns
corrMatrix=parkinsons_data.corr()
corrMatrix
```

<ipython-input-26-b9b1b4466795>:3: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
corrMatrix=parkinsons_data.corr()
```

[ ]:	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%)	\
MDVP:Fo(Hz)	1.000000	0.400985	0.596546	-0.118003	
MDVP:Fhi(Hz)	0.400985	1.000000	0.084951	0.102086	
MDVP:Flo(Hz)	0.596546	0.084951	1.000000	-0.139919	
MDVP:Jitter(%)	-0.118003	0.102086	-0.139919	1.000000	
MDVP:Jitter(Abs)	-0.382027	-0.029198	-0.277815	0.935714	
MDVP:RAP	-0.076194	0.097177	-0.100519	0.990276	
MDVP:PPQ	-0.112165	0.091126	-0.095828	0.974256	
Jitter:DDP	-0.076213	0.097150	-0.100488	0.990276	
MDVP:Shimmer	-0.098374	0.002281	-0.144543	0.769063	
MDVP:Shimmer(dB)	-0.073742	0.043465	-0.119089	0.804289	
Shimmer:APQ3	-0.094717	-0.003743	-0.150747	0.746625	
Shimmer:APQ5	-0.070682	-0.009997	-0.101095	0.725561	
MDVP:APQ	-0.077774	0.004937	-0.107293	0.758255	
Shimmer:DDA	-0.094732	-0.003733	-0.150737	0.746635	
NHR	-0.021981	0.163766	-0.108670	0.906959	
HNR	0.059144	-0.024893	0.210851	-0.728165	
status	-0.383535	-0.166136	-0.380200	0.278220	
RPDE	-0.383894	-0.112404	-0.400143	0.360673	
DFA	-0.446013	-0.343097	-0.050406	0.098572	
spread1	-0.413738	-0.076658	-0.394857	0.693577	
spread2	-0.249450	-0.002954	-0.243829	0.385123	
D2	0.177980	0.176323	-0.100629	0.433434	
PPE	-0.372356	-0.069543	-0.340071	0.721543	

	MDVP:Jitter(Abs)	MDVP:RAP	MDVP:PPQ	Jitter:DDP	\
MDVP:Fo(Hz)	-0.382027	-0.076194	-0.112165	-0.076213	
MDVP:Fhi(Hz)	-0.029198	0.097177	0.091126	0.097150	
MDVP:Flo(Hz)	-0.277815	-0.100519	-0.095828	-0.100488	
MDVP:Jitter(%)	0.935714	0.990276	0.974256	0.990276	
MDVP:Jitter(Abs)	1.000000	0.922911	0.897778	0.922913	
MDVP:RAP	0.922911	1.000000	0.957317	1.000000	
MDVP:PPQ	0.897778	0.957317	1.000000	0.957319	
Jitter:DDP	0.922913	1.000000	0.957319	1.000000	
MDVP:Shimmer	0.703322	0.759581	0.797826	0.759555	
MDVP:Shimmer(dB)	0.716601	0.790652	0.839239	0.790621	
Shimmer:APQ3	0.697153	0.744912	0.763580	0.744894	
Shimmer:APQ5	0.648961	0.709927	0.786780	0.709907	
MDVP:APQ	0.648793	0.737455	0.804139	0.737439	
Shimmer:DDA	0.697170	0.744919	0.763592	0.744901	
NHR	0.834972	0.919521	0.844604	0.919548	
HNR	-0.656810	-0.721543	-0.731510	-0.721494	
status	0.338653	0.266668	0.288698	0.266646	
RPDE	0.441839	0.342140	0.333274	0.342079	
DFA	0.175036	0.064083	0.196301	0.064026	
spread1	0.735779	0.648328	0.716489	0.648328	
spread2	0.388543	0.324407	0.407605	0.324377	

D2	0.310694	0.426605	0.412524	0.426556
PPE	0.748162	0.670999	0.769647	0.671005

	MDVP:Shimmer	MDVP:Shimmer(dB)	...	Shimmer:DDA	NHR \
MDVP:Fo(Hz)	-0.098374	-0.073742	...	-0.094732	-0.021981
MDVP:Fhi(Hz)	0.002281	0.043465	...	-0.003733	0.163766
MDVP:Flo(Hz)	-0.144543	-0.119089	...	-0.150737	-0.108670
MDVP:Jitter(%)	0.769063	0.804289	...	0.746635	0.906959
MDVP:Jitter(Abs)	0.703322	0.716601	...	0.697170	0.834972
MDVP:RAP	0.759581	0.790652	...	0.744919	0.919521
MDVP:PPQ	0.797826	0.839239	...	0.763592	0.844604
Jitter:DDP	0.759555	0.790621	...	0.744901	0.919548
MDVP:Shimmer	1.000000	0.987258	...	0.987626	0.722194
MDVP:Shimmer(dB)	0.987258	1.000000	...	0.963202	0.744477
Shimmer:APQ3	0.987625	0.963198	...	1.000000	0.716207
Shimmer:APQ5	0.982835	0.973751	...	0.960072	0.658080
MDVP:APQ	0.950083	0.960977	...	0.896647	0.694019
Shimmer:DDA	0.987626	0.963202	...	1.000000	0.716215
NHR	0.722194	0.744477	...	0.716215	1.000000
HNR	-0.835271	-0.827805	...	-0.827130	-0.714072
status	0.367430	0.350697	...	0.347608	0.189429
RPDE	0.447424	0.410684	...	0.435237	0.370890
DFA	0.159954	0.165157	...	0.151132	-0.131882
spread1	0.654734	0.652547	...	0.610971	0.540865
spread2	0.452025	0.454314	...	0.402223	0.318099
D2	0.507088	0.512233	...	0.467261	0.470949
PPE	0.693771	0.695058	...	0.645389	0.552591

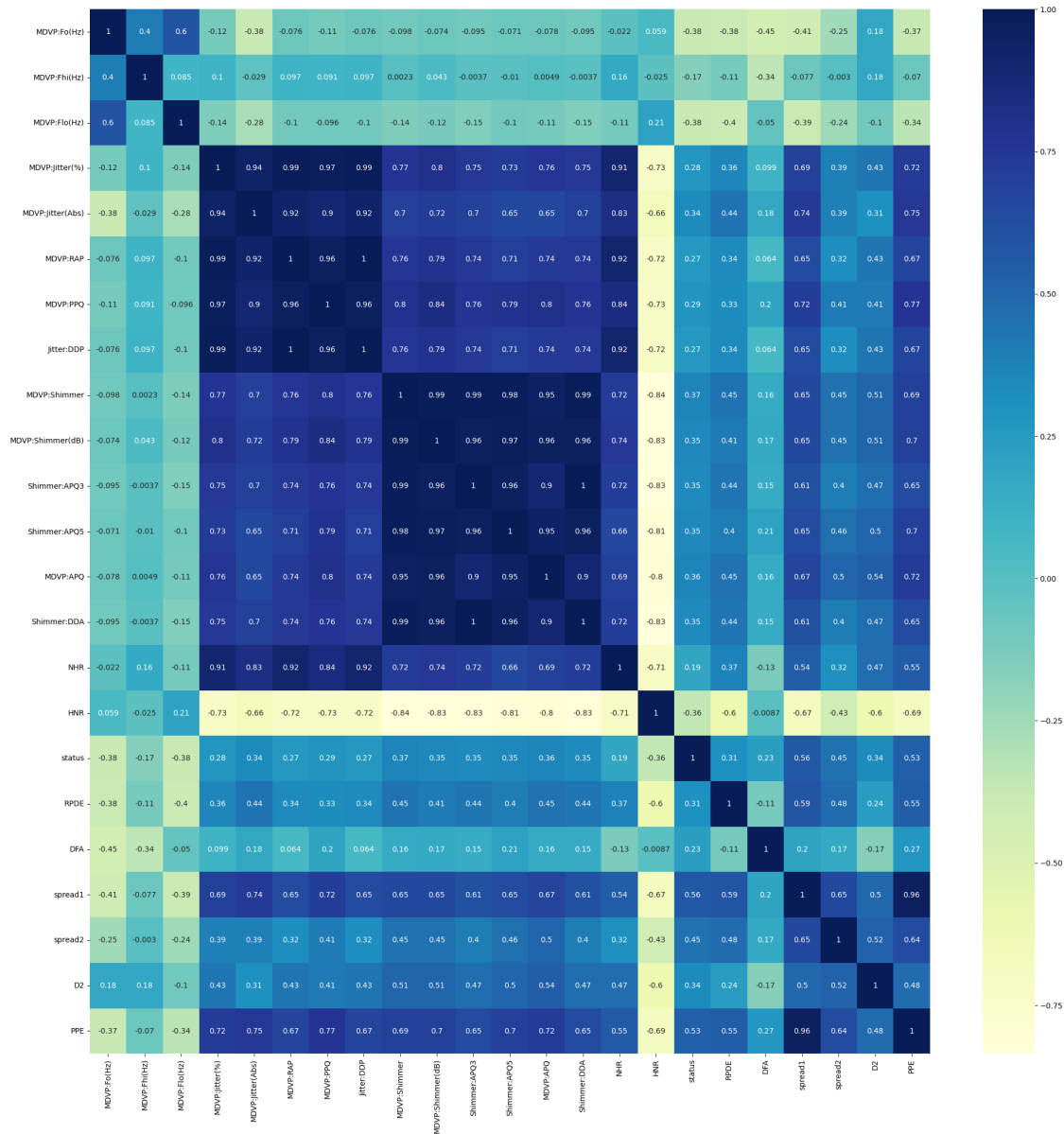
	HNR	status	RPDE	DFA	spread1	spread2 \
MDVP:Fo(Hz)	0.059144	-0.383535	-0.383894	-0.446013	-0.413738	-0.249450
MDVP:Fhi(Hz)	-0.024893	-0.166136	-0.112404	-0.343097	-0.076658	-0.002954
MDVP:Flo(Hz)	0.210851	-0.380200	-0.400143	-0.050406	-0.394857	-0.243829
MDVP:Jitter(%)	-0.728165	0.278220	0.360673	0.098572	0.693577	0.385123
MDVP:Jitter(Abs)	-0.656810	0.338653	0.441839	0.175036	0.735779	0.388543
MDVP:RAP	-0.721543	0.266668	0.342140	0.064083	0.648328	0.324407
MDVP:PPQ	-0.731510	0.288698	0.333274	0.196301	0.716489	0.407605
Jitter:DDP	-0.721494	0.266646	0.342079	0.064026	0.648328	0.324377
MDVP:Shimmer	-0.835271	0.367430	0.447424	0.159954	0.654734	0.452025
MDVP:Shimmer(dB)	-0.827805	0.350697	0.410684	0.165157	0.652547	0.454314
Shimmer:APQ3	-0.827123	0.347617	0.435242	0.151124	0.610967	0.402243
Shimmer:APQ5	-0.813753	0.351148	0.399903	0.213873	0.646809	0.457195
MDVP:APQ	-0.800407	0.364316	0.451379	0.157276	0.673158	0.502188
Shimmer:DDA	-0.827130	0.347608	0.435237	0.151132	0.610971	0.402223
NHR	-0.714072	0.189429	0.370890	-0.131882	0.540865	0.318099
HNR	1.000000	-0.361515	-0.598736	-0.008665	-0.673210	-0.431564
status	-0.361515	1.000000	0.308567	0.231739	0.564838	0.454842
RPDE	-0.598736	0.308567	1.000000	-0.110950	0.591117	0.479905

DFA	-0.008665	0.231739	-0.110950	1.000000	0.195668	0.166548
spread1	-0.673210	0.564838	0.591117	0.195668	1.000000	0.652358
spread2	-0.431564	0.454842	0.479905	0.166548	0.652358	1.000000
D2	-0.601401	0.340232	0.236931	-0.165381	0.495123	0.523532
PPE	-0.692876	0.531039	0.545886	0.270445	0.962435	0.644711

	D2	PPE
MDVP:Fo(Hz)	0.177980	-0.372356
MDVP:Fhi(Hz)	0.176323	-0.069543
MDVP:Flo(Hz)	-0.100629	-0.340071
MDVP:Jitter(%)	0.433434	0.721543
MDVP:Jitter(Abs)	0.310694	0.748162
MDVP:RAP	0.426605	0.670999
MDVP:PPQ	0.412524	0.769647
Jitter:DDP	0.426556	0.671005
MDVP:Shimmer	0.507088	0.693771
MDVP:Shimmer(dB)	0.512233	0.695058
Shimmer:APQ3	0.467265	0.645377
Shimmer:APQ5	0.502174	0.702456
MDVP:APQ	0.536869	0.721694
Shimmer:DDA	0.467261	0.645389
NHR	0.470949	0.552591
HNR	-0.601401	-0.692876
status	0.340232	0.531039
RPDE	0.236931	0.545886
DFA	-0.165381	0.270445
spread1	0.495123	0.962435
spread2	0.523532	0.644711
D2	1.000000	0.480585
PPE	0.480585	1.000000

[23 rows x 23 columns]

```
[ ]: import seaborn as sns
sns.heatmap(corrMatrix,cmap="YlGnBu",annot=True)
plt.gcf().set_size_inches(25, 25)
```



```
[ ]: # grouping the data bas3ed on the target variable
parkinsons_data.groupby('status').mean()
```

<ipython-input-9-fe279e55666c>:2: FutureWarning: The default value of numeric\_only in DataFrameGroupBy.mean is deprecated. In a future version, numeric\_only will default to False. Either specify numeric\_only or select only columns which should be valid for the function.

```
parkinsons_data.groupby('status').mean()
```

```
[ ]:      MDVP:Fo(Hz)  MDVP:Fhi(Hz)  MDVP:Flo(Hz)  MDVP:Jitter(%)  \
status
```



0	181.937771	223.636750	145.207292	0.003866
1	145.180762	188.441463	106.893558	0.006989

	MDVP:Jitter(Abs)	MDVP:RAP	MDVP:PPQ	Jitter:DDP	MDVP:Shimmer \
status					
0	0.000023	0.001925	0.002056	0.005776	0.017615
1	0.000051	0.003757	0.003900	0.011273	0.033658

	MDVP:Shimmer(dB)	...	MDVP:APQ	Shimmer:DDA	NHR	HNR \
status		...				
0	0.162958	...	0.013305	0.028511	0.011483	24.678750
1	0.321204	...	0.027600	0.053027	0.029211	20.974048

	RPDE	DFA	spread1	spread2	D2	PPE
status						
0	0.442552	0.695716	-6.759264	0.160292	2.154491	0.123017
1	0.516816	0.725408	-5.333420	0.248133	2.456058	0.233828

[2 rows x 22 columns]

Data Pre-Processing

Separating the features & Target

```
[ ]: X = parkinsons_data.drop(columns=['name', 'status'], axis=1)
      Y = parkinsons_data['status']
```

```
[ ]: print(X)
```

	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%) \
0	119.992	157.302	74.997	0.00784
1	122.400	148.650	113.819	0.00968
2	116.682	131.111	111.555	0.01050
3	116.676	137.871	111.366	0.00997
4	116.014	141.781	110.655	0.01284
..	...	...	...	...
190	174.188	230.978	94.261	0.00459
191	209.516	253.017	89.488	0.00564
192	174.688	240.005	74.287	0.01360
193	198.764	396.961	74.904	0.00740
194	214.289	260.277	77.973	0.00567

	MDVP:Jitter(Abs)	MDVP:RAP	MDVP:PPQ	Jitter:DDP	MDVP:Shimmer \
0	0.00007	0.00370	0.00554	0.01109	0.04374
1	0.00008	0.00465	0.00696	0.01394	0.06134
2	0.00009	0.00544	0.00781	0.01633	0.05233
3	0.00009	0.00502	0.00698	0.01505	0.05492
4	0.00011	0.00655	0.00908	0.01966	0.06425

```

..
190      0.00003  0.00263  0.00259  0.00790  0.04087
191      0.00003  0.00331  0.00292  0.00994  0.02751
192      0.00008  0.00624  0.00564  0.01873  0.02308
193      0.00004  0.00370  0.00390  0.01109  0.02296
194      0.00003  0.00295  0.00317  0.00885  0.01884

      MDVP:Shimmer(dB) ... MDVP:APQ  Shimmer:DDA      NHR      HNR      RPDE  \
0      0.426 ...  0.02971      0.06545  0.02211  21.033  0.414783
1      0.626 ...  0.04368      0.09403  0.01929  19.085  0.458359
2      0.482 ...  0.03590      0.08270  0.01309  20.651  0.429895
3      0.517 ...  0.03772      0.08771  0.01353  20.644  0.434969
4      0.584 ...  0.04465      0.10470  0.01767  19.649  0.417356
..
190      0.405 ...  0.02745      0.07008  0.02764  19.517  0.448439
191      0.263 ...  0.01879      0.04812  0.01810  19.147  0.431674
192      0.256 ...  0.01667      0.03804  0.10715  17.883  0.407567
193      0.241 ...  0.01588      0.03794  0.07223  19.020  0.451221
194      0.190 ...  0.01373      0.03078  0.04398  21.209  0.462803

      DFA      spread1      spread2      D2      PPE
0      0.815285 -4.813031  0.266482  2.301442  0.284654
1      0.819521 -4.075192  0.335590  2.486855  0.368674
2      0.825288 -4.443179  0.311173  2.342259  0.332634
3      0.819235 -4.117501  0.334147  2.405554  0.368975
4      0.823484 -3.747787  0.234513  2.332180  0.410335
..
190      0.657899 -6.538586  0.121952  2.657476  0.133050
191      0.683244 -6.195325  0.129303  2.784312  0.168895
192      0.655683 -6.787197  0.158453  2.679772  0.131728
193      0.643956 -6.744577  0.207454  2.138608  0.123306
194      0.664357 -5.724056  0.190667  2.555477  0.148569

```

[195 rows x 22 columns]

```
[ ]: print(Y)
```

```

0      1
1      1
2      1
3      1
4      1
..
190    0
191    0
192    0
193    0
194    0

```

Name: status, Length: 195, dtype: int64

Splitting the data to training data & Test data

```
[ ]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2,  
↳ random_state=2)
```

```
[ ]: print(X.shape, X_train.shape, X_test.shape)
```

(195, 22) (156, 22) (39, 22)

Model Training

Support Vector Machine Model

```
[ ]: model = svm.SVC(kernel='linear')
```

```
[ ]: # training the SVM model with training data  
model.fit(X_train, Y_train)
```

```
[ ]: SVC(kernel='linear')
```

Model Evaluation

Accuracy Score

```
[ ]: # accuracy score on training data  
X_train_prediction = model.predict(X_train)  
training_data_accuracy = accuracy_score(Y_train, X_train_prediction)
```

```
[ ]: print('Accuracy score of training data : ', training_data_accuracy)
```

Accuracy score of training data : 0.8717948717948718

```
[ ]: # accuracy score on training data  
X_test_prediction = model.predict(X_test)  
test_data_accuracy = accuracy_score(Y_test, X_test_prediction)
```

```
[ ]: print('Accuracy score of test data : ', test_data_accuracy)
```

Accuracy score of test data : 0.8717948717948718

Building a Predictive System

```
[ ]: input_data = (197.07600,206.89600,192.05500,0.00289,0.00001,0.00166,0.00168,0.  
↳ 00498,0.01098,0.09700,0.00563,0.00680,0.00802,0.01689,0.00339,26.77500,0.  
↳ 422229,0.741367,-7.348300,0.177551,1.743867,0.085569)  
  
# changing input data to a numpy array  
input_data_as_numpy_array = np.asarray(input_data)
```

```

# reshape the numpy array
input_data_resaped = input_data_as_numpy_array.reshape(1,-1)

prediction = model.predict(input_data_resaped)
print(prediction)

if (prediction[0] == 0):
    print("The Person does not have Parkinsons Disease")
else:
    print("The Person has Parkinsons")

```

[0]

The Person does not have Parkinsons Disease

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but SVC was fitted with feature names

warnings.warn(

Saving the trained model

```
[ ]: import pickle
```

```
[ ]: filename = 'parkinsons_model.sav'
pickle.dump(model, open(filename, 'wb'))
```

```
[ ]: # loading the saved model
loaded_model = pickle.load(open('parkinsons_model.sav', 'rb'))
```

```
[ ]: for column in X.columns:
    print(column)
```

MDVP:Fo(Hz)  
MDVP:Fhi(Hz)  
MDVP:Flo(Hz)  
MDVP:Jitter(%)  
MDVP:Jitter(Abs)  
MDVP:RAP  
MDVP:PPQ  
Jitter:DDP  
MDVP:Shimmer  
MDVP:Shimmer(dB)  
Shimmer:APQ3  
Shimmer:APQ5  
MDVP:APQ  
Shimmer:DDA  
NHR  
HNR

RPDE  
DFA  
spread1  
spread2  
D2  
PPE

[ ]: