

#Importing Libraries

```
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn import svm
from sklearn.metrics import accuracy_score
```

Parkinson Disease Prediction Using Support Vector Machine Model

#Reading the CSV File

```
data = pd.read_csv('Parkinsons_Dataset.csv')
data
```

	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				
..
...				
190	phon_R01_S50_2	174.188	230.978	94.261
0.00459				
191	phon_R01_S50_3	209.516	253.017	89.488
0.00564				
192	phon_R01_S50_4	174.688	240.005	74.287
0.01360				
193	phon_R01_S50_5	198.764	396.961	74.904
0.00740				
194	phon_R01_S50_6	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	...					

	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						
..
.						
190	0.07008	0.02764	19.517	0	0.448439	0.657899 -
6.538586						
191	0.04812	0.01810	19.147	0	0.431674	0.683244 -
6.195325						
192	0.03804	0.10715	17.883	0	0.407567	0.655683 -
6.787197						
193	0.03794	0.07223	19.020	0	0.451221	0.643956 -
6.744577						
194	0.03078	0.04398	21.209	0	0.462803	0.664357 -
5.724056						

	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
..
190	0.121952	2.657476	0.133050

```

191  0.129303  2.784312  0.168895
192  0.158453  2.679772  0.131728
193  0.207454  2.138608  0.123306
194  0.190667  2.555477  0.148569

```

```
[195 rows x 24 columns]
```

```
#Checking for null values
```

```
data.isnull().sum()
```

```

name                                0
MDVP:Fo(Hz)                        0
MDVP:Fhi(Hz)                       0
MDVP:Flo(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

```

```
#Removing different statistic values
```

```
data.describe()
```

	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(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]

#1 -> Parkinson Positive

#0 -> Healthy

data['status'].value_counts()

status

1 147

0 48

Name: count, dtype: int64

#Data Pre-processing

x = data.drop(columns=['name', 'status'], axis=1)

y = data['status']

print(x)

print(y)

	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	

RPDE \	MDVP:Shimmer(dB)	...	MDVP:APQ	Shimmer:DDA	NHR	HNR
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]

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 into training data and testing data
```

```
x_train,x_test,y_train,y_test =
train_test_split(x,y,test_size=0.2,random_state=2)
print(x_train)
print(x_test)
```

	MDVP:F0(Hz)	MDVP:F1(Hz)	MDVP:F2(Hz)	MDVP:Jitter(%)	\
123	182.018	197.173	79.187	0.00842	
160	114.238	124.393	77.022	0.00581	
94	157.821	172.975	68.401	0.00358	
57	117.274	129.916	110.402	0.00752	
41	184.055	196.537	166.977	0.00258	
..	
43	241.404	248.834	232.483	0.00281	
22	167.930	193.221	79.068	0.00442	
72	120.080	139.710	111.208	0.00405	
15	142.167	217.455	83.159	0.00369	
168	197.569	217.627	90.794	0.00803	

	MDVP:Jitter(Abs)	MDVP:RAP	MDVP:PPQ	Jitter:DDP	MDVP:Shimmer	\
123	0.00005	0.00506	0.00449	0.01517	0.02503	
160	0.00005	0.00299	0.00316	0.00896	0.04009	
94	0.00002	0.00196	0.00196	0.00587	0.03716	
57	0.00006	0.00299	0.00469	0.00898	0.02293	
41	0.00001	0.00134	0.00147	0.00403	0.01463	
..	
43	0.00001	0.00157	0.00173	0.00470	0.01760	
22	0.00003	0.00220	0.00247	0.00661	0.04351	
72	0.00003	0.00180	0.00220	0.00540	0.01706	
15	0.00003	0.00157	0.00203	0.00471	0.01503	
168	0.00004	0.00490	0.00448	0.01470	0.02177	

	MDVP:Shimmer(dB)	...	MDVP:APQ	Shimmer:DDA	NHR	HNR
RPDE \						
123	0.231	...	0.01931	0.04115	0.01813	18.784
0.589956						
160	0.406	...	0.04114	0.04736	0.02073	20.437
0.653139						
94	0.307	...	0.02764	0.06185	0.00850	22.219
0.502380						
57	0.221	...	0.01948	0.03568	0.00681	22.817
0.530529						
41	0.132	...	0.01234	0.02226	0.00257	26.453
0.306443						
..
...						

43	0.154	...	0.01251	0.03017	0.00675	23.145
0.457702						
22	0.377	...	0.04246	0.06685	0.01280	22.468
0.619060						
72	0.152	...	0.01345	0.02921	0.00442	25.742
0.495954						
15	0.126	...	0.01359	0.02316	0.00839	25.175
0.565924						
168	0.189	...	0.01439	0.03836	0.01337	19.269
0.372222						

	DFA	spread1	spread2	D2	PPE
123	0.732903	-5.445140	0.142466	2.174306	0.215558
160	0.694571	-5.185987	0.259229	2.151121	0.244948
94	0.712170	-6.251425	0.188056	2.143851	0.160812
57	0.817756	-4.608260	0.290024	2.021591	0.314464
41	0.759203	-7.044105	0.063412	2.361532	0.115730
..
43	0.634267	-6.793547	0.158266	2.256699	0.117399
22	0.679834	-4.330956	0.262384	2.916777	0.285695
72	0.762959	-5.791820	0.329066	2.205024	0.188180
15	0.658245	-5.340115	0.210185	2.205546	0.234589
168	0.725216	-5.736781	0.164529	2.882450	0.202879

[156 rows x 22 columns]

	MDVP:Fo(Hz)	MDVP:Fhi(Hz)	MDVP:Flo(Hz)	MDVP:Jitter(%)	\
10	88.333	112.240	84.072	0.00505	
79	100.770	115.697	85.545	0.01038	
164	102.273	142.830	85.902	0.00907	
142	198.458	219.290	148.691	0.00376	
186	116.556	592.030	86.228	0.00496	
133	118.747	123.723	109.836	0.00331	
35	201.464	210.565	195.708	0.00198	
137	113.166	130.270	100.673	0.00502	
25	104.400	206.002	77.968	0.00633	
2	116.682	131.111	111.555	0.01050	
12	136.926	159.866	131.276	0.00293	
128	119.031	127.533	109.216	0.00440	
144	202.544	241.350	164.168	0.00254	
3	116.676	137.871	111.366	0.00997	
48	122.188	128.611	115.765	0.00524	
29	162.568	198.346	77.630	0.00502	
14	152.845	163.305	75.836	0.00294	
119	217.116	233.481	93.978	0.00404	
6	120.267	137.244	114.820	0.00333	
23	173.917	192.735	86.180	0.00476	
108	151.989	157.339	132.857	0.00174	
143	202.805	231.508	86.232	0.00370	
129	120.078	126.632	105.667	0.00270	

174	117.004	144.466	99.923	0.00353
45	242.852	255.034	227.911	0.00225
120	128.940	479.697	88.251	0.00581
173	113.715	116.443	96.913	0.00349
125	145.174	198.109	80.637	0.00733
9	95.056	120.103	91.226	0.00532
163	112.150	131.669	97.527	0.00519
54	108.807	134.656	102.874	0.00761
13	139.173	179.139	76.556	0.00390
109	193.030	208.900	80.297	0.00766
194	214.289	260.277	77.973	0.00567
78	95.385	102.145	90.264	0.00608
114	206.327	226.355	92.020	0.00495
44	243.439	250.912	232.435	0.00210
82	100.960	110.019	95.628	0.00606
158	126.144	154.284	97.543	0.00975

	MDVP:Jitter(Abs)	MDVP:RAP	MDVP:PPQ	Jitter:DDP	MDVP:Shimmer	\
10	0.000060	0.00254	0.00330	0.00763	0.02143	
79	0.000100	0.00622	0.00576	0.01865	0.03121	
164	0.000090	0.00493	0.00461	0.01480	0.02814	
142	0.000020	0.00182	0.00215	0.00546	0.03527	
186	0.000040	0.00254	0.00263	0.00762	0.01660	
133	0.000030	0.00168	0.00171	0.00504	0.01043	
35	0.000010	0.00105	0.00115	0.00314	0.01194	
137	0.000040	0.00257	0.00312	0.00772	0.05279	
25	0.000060	0.00316	0.00375	0.00948	0.03767	
2	0.000090	0.00544	0.00781	0.01633	0.05233	
12	0.000020	0.00118	0.00153	0.00355	0.01259	
128	0.000040	0.00214	0.00192	0.00641	0.01033	
144	0.000010	0.00100	0.00133	0.00301	0.02662	
3	0.000090	0.00502	0.00698	0.01505	0.05492	
48	0.000040	0.00169	0.00203	0.00507	0.01613	
29	0.000030	0.00280	0.00253	0.00841	0.01791	
14	0.000020	0.00121	0.00149	0.00364	0.01828	
119	0.000020	0.00127	0.00128	0.00381	0.01299	
6	0.000030	0.00155	0.00202	0.00466	0.01608	
23	0.000030	0.00221	0.00258	0.00663	0.04192	
108	0.000010	0.00075	0.00096	0.00225	0.01024	
143	0.000020	0.00189	0.00211	0.00568	0.01997	
129	0.000020	0.00116	0.00135	0.00349	0.01022	
174	0.000030	0.00176	0.00218	0.00528	0.01657	
45	0.000009	0.00117	0.00139	0.00350	0.01494	
120	0.000050	0.00241	0.00314	0.00723	0.02008	
173	0.000030	0.00171	0.00203	0.00514	0.01472	
125	0.000050	0.00414	0.00422	0.01242	0.02362	
9	0.000060	0.00268	0.00332	0.00803	0.02838	
163	0.000050	0.00291	0.00284	0.00873	0.01756	
54	0.000070	0.00349	0.00486	0.01046	0.02719	

13	0.000030	0.00165	0.00208	0.00496	0.01642
109	0.000040	0.00450	0.00389	0.01351	0.03044
194	0.000030	0.00295	0.00317	0.00885	0.01884
78	0.000060	0.00331	0.00332	0.00994	0.03202
114	0.000020	0.00302	0.00246	0.00905	0.02105
44	0.000009	0.00109	0.00137	0.00327	0.01419
82	0.000060	0.00351	0.00348	0.01053	0.02427
158	0.000080	0.00593	0.00454	0.01778	0.02852

	MDVP:Shimmer(dB)	...	MDVP:APQ	Shimmer:DDA	NHR	HNR
RPDE \						
10	0.197	...	0.01892	0.03237	0.01166	21.118
0.611137						
79	0.361	...	0.02139	0.05320	0.02220	19.200
0.594387						
164	0.272	...	0.02073	0.04736	0.03882	18.447
0.671378						
142	0.297	...	0.02530	0.06165	0.01728	18.702
0.606273						
186	0.154	...	0.01491	0.02460	0.01397	23.958
0.566424						
133	0.099	...	0.00903	0.01471	0.00504	25.619
0.482296						
35	0.107	...	0.00957	0.01758	0.00135	31.732
0.344252						
137	0.476	...	0.04134	0.08689	0.03690	16.747
0.625362						
25	0.381	...	0.03780	0.05197	0.02887	22.066
0.522746						
2	0.482	...	0.03590	0.08270	0.01309	20.651
0.429895						
12	0.112	...	0.01140	0.01968	0.00581	25.703
0.460600						
128	0.098	...	0.00811	0.01614	0.01724	26.842
0.457541						
144	0.228	...	0.02006	0.04426	0.01049	20.680
0.497480						
3	0.517	...	0.03772	0.08771	0.01353	20.644
0.434969						
48	0.143	...	0.01433	0.02566	0.00839	23.162
0.579597						
29	0.168	...	0.01799	0.02380	0.01170	25.678
0.427785						
14	0.158	...	0.01246	0.03191	0.00609	24.922
0.474791						
119	0.124	...	0.01075	0.02038	0.00681	24.581
0.462516						
6	0.140	...	0.01351	0.02337	0.00607	24.886
0.596040						

23	0.364	...	0.03772	0.06562	0.01840	20.422
0.537264						
108	0.093	...	0.00993	0.01364	0.00238	29.928
0.311369						
143	0.180	...	0.01506	0.03350	0.02010	18.687
0.536102						
129	0.090	...	0.00903	0.01428	0.00487	26.369
0.491345						
174	0.145	...	0.01318	0.02643	0.00590	25.445
0.387482						
45	0.134	...	0.01014	0.02542	0.00476	25.032
0.431285						
120	0.221	...	0.01734	0.02548	0.02350	24.743
0.487756						
173	0.133	...	0.01148	0.02245	0.00478	26.547
0.380253						
125	0.233	...	0.01944	0.03706	0.01874	18.857
0.637518						
9	0.255	...	0.02444	0.04324	0.01022	21.862
0.547037						
163	0.155	...	0.01363	0.02902	0.01435	21.219
0.557045						
54	0.255	...	0.02067	0.04450	0.01036	21.028
0.536009						
13	0.154	...	0.01797	0.02184	0.01041	24.889
0.430166						
109	0.275	...	0.02084	0.05312	0.00947	21.934
0.497554						
194	0.190	...	0.01373	0.03078	0.04398	21.209
0.462803						
78	0.263	...	0.02455	0.05408	0.01062	21.875
0.644954						
114	0.209	...	0.01604	0.03320	0.01048	21.864
0.331508						
44	0.126	...	0.01033	0.02330	0.00454	25.368
0.438296						
82	0.216	...	0.01751	0.04114	0.01237	20.536
0.554610						
158	0.266	...	0.02157	0.04499	0.03828	21.534
0.635015						
	DFA	spread1	spread2	D2	PPE	
10	0.776156	-5.249770	0.391002	2.407313	0.249740	
79	0.790117	-4.913885	0.265699	2.398422	0.277948	
164	0.674562	-2.929379	0.396746	2.560422	0.367233	
142	0.661735	-5.585259	0.310746	2.465528	0.209863	
186	0.667654	-6.431119	0.153310	2.161936	0.120605	
133	0.723096	-6.448134	0.178713	2.034827	0.141422	
35	0.742737	-7.777685	0.170183	2.447064	0.057610	

137	0.708617	-4.654894	0.304107	2.672362	0.274387
25	0.737948	-5.571843	0.236853	2.846369	0.219514
2	0.825288	-4.443179	0.311173	2.342259	0.332634
12	0.646846	-6.547148	0.152813	2.041277	0.138512
128	0.699787	-6.890021	0.152941	2.328513	0.112856
144	0.630409	-6.132663	0.220617	2.576563	0.159777
3	0.819235	-4.117501	0.334147	2.405554	0.368975
48	0.733659	-6.439398	0.266392	2.079922	0.133867
29	0.723797	-6.635729	0.209866	1.957961	0.135242
14	0.654027	-6.105098	0.203653	2.125618	0.170100
119	0.582710	-5.517173	0.389295	2.925862	0.220657
6	0.764112	-5.634322	0.257682	1.854785	0.211756
23	0.686894	-5.248776	0.210279	2.547508	0.253556
108	0.676066	-6.739151	0.160686	2.296873	0.115130
143	0.632631	-5.898673	0.213353	2.470746	0.189032
129	0.718839	-5.892061	0.195976	2.108873	0.183572
174	0.756482	-6.012559	0.229298	1.872946	0.163118
45	0.638928	-6.995820	0.102083	2.365800	0.102706
120	0.684130	-6.186128	0.279933	2.686240	0.152428
173	0.766700	-5.943501	0.192150	1.852542	0.179677
125	0.735546	-5.594275	0.127950	1.765957	0.222716
9	0.798463	-5.011879	0.325996	2.432792	0.271362
163	0.673086	-5.617124	0.184896	1.871871	0.212386
54	0.819032	-4.649573	0.205558	1.986899	0.316700
13	0.665833	-5.660217	0.254989	2.519422	0.199889
109	0.740539	-5.845099	0.278679	2.608749	0.185668
194	0.664357	-5.724056	0.190667	2.555477	0.148569
78	0.779612	-5.115212	0.249494	2.017753	0.260015
114	0.715121	-6.729713	0.181701	2.938114	0.115515
44	0.635285	-7.057869	0.091608	2.330716	0.091470
82	0.787896	-5.022288	0.146948	2.428306	0.264666
158	0.627337	-5.070096	0.280091	2.892300	0.249703

[39 rows x 22 columns]

```
scalar = StandardScaler()  
scalar.fit(x_train)
```

```
StandardScaler()
```

```
x_train = scalar.transform(x_train)  
x_test = scalar.transform(x_test)
```

#Model Training

```
model = svm.SVC(kernel='linear')  
model.fit(x_train,y_train)
```

```
SVC(kernel='linear')
```

#Model Evaluation

```
x_train_prediction = model.predict(x_train)
```

```
x_train_data_accuracy = accuracy_score(y_train,x_train_prediction)
print("Accuracy Score : ", x_train_data_accuracy)
```

Accuracy Score : 0.8846153846153846

```
x_test_prediction = model.predict(x_test)
x_test_data_accuracy = accuracy_score(y_test,x_test_prediction)
print("Accuracy Score : ",x_test_data_accuracy)
```

Accuracy Score : 0.8717948717948718

#Predicting from a User Data

```
input_data
=(95.05600,120.10300,91.22600,0.00532,0.00006,0.00268,0.00332,0.00803,
0.02838,0.25500,0.01441,0.01725,0.02444,0.04324,0.01022,21.86200,0.547
037,0.798463,-5.011879,0.325996,2.432792,0.271362)
input = np.asarray(input_data)
input_reshaped = input.reshape(1,-1)
standard_data = scalar.transform(input_reshaped)
prediction = model.predict(standard_data)
print(prediction)
if(prediction[0] == 1):
    print("Person has Parkinson")
else:
    print("Person does not have Parkinson")
```

[1]
Person has Parkinson

c:\Users\RUPESH SHAH\AppData\Local\Programs\Python\Python312\Lib\site-packages\sklearn\base.py:493: UserWarning: X does not have valid feature names, but StandardScaler was fitted with feature names
warnings.warn(

#Checking for Person with Not Parkinson Positive

```
input_data
=(114.56300,119.16700,86.64700,0.00327,0.00003,0.00146,0.00184,0.00439
,0.01185,0.10600,0.00557,0.00721,0.01095,0.01672,0.00703,24.77500,0.55
5303,0.659132,-6.710219,0.149694,1.913990,0.121777)
input = np.asarray(input_data)
input_reshaped = input.reshape(1,-1)
standard_data = scalar.transform(input_reshaped)
prediction = model.predict(standard_data)
print(prediction)
if(prediction[0] == 1):
    print("Person has Parkinson")
else:
    print("Person does not have Parkinson")
```

```
[0]
```

```
Person does not have Parkinson
```

```
c:\Users\RUPESH SHAH\AppData\Local\Programs\Python\Python312\Lib\site-  
packages\sklearn\base.py:493: UserWarning: X does not have valid  
feature names, but StandardScaler was fitted with feature names  
warnings.warn(
```