

Diabetes Prediction

August 29, 2021

```
[1]: import pandas as pd
import numpy as np
```

```
[18]: data=pd.read_csv("Diabetes Dataset.csv")
data.head(50)
```

```
[18]: Unnamed: 0  Pregnancies  Glucose  blood pressure  skin thickness  Insulin  \
0          NaN          6.0    148.0           72.0           35.0         0.0
1          NaN          1.0     85.0           66.0           29.0         0.0
2          NaN          8.0    183.0           64.0            0.0         0.0
3          NaN          1.0     89.0           66.0           23.0        94.0
4          NaN          0.0    137.0           40.0           35.0       168.0
5          NaN          5.0    116.0           74.0            0.0         0.0
6          NaN          3.0     78.0           50.0           32.0        88.0
7          NaN         10.0    115.0            0.0            0.0         0.0
8          NaN          2.0    197.0           70.0           45.0       543.0
9          NaN          8.0    125.0           96.0            0.0         0.0
10         NaN          4.0    110.0           92.0            0.0         0.0
11         NaN         10.0    168.0           74.0            0.0         0.0
12         NaN         10.0    139.0           80.0            0.0         0.0
13         NaN          1.0    189.0           60.0           23.0       846.0
14         NaN          5.0    166.0           72.0           19.0       175.0
15         NaN          7.0    100.0            0.0            0.0         0.0
16         NaN          0.0    118.0           84.0           47.0       230.0
17         NaN          7.0    107.0           74.0            0.0         0.0
18         NaN          1.0    103.0           30.0           38.0        83.0
19         NaN          1.0    115.0           70.0           30.0        96.0
20         NaN          3.0    126.0           88.0           41.0       235.0
21         NaN          8.0     99.0           84.0            0.0         0.0
22         NaN          7.0    196.0           90.0            0.0         0.0
23         NaN          9.0    119.0           80.0           35.0         0.0
24         NaN         11.0    143.0           94.0           33.0       146.0
25         NaN         10.0    125.0           70.0           26.0       115.0
26         NaN          7.0    147.0           76.0            0.0         0.0
27         NaN          1.0     97.0           66.0           15.0       140.0
28         NaN         13.0    145.0           82.0           19.0       110.0
29         NaN          5.0    117.0           92.0            0.0         0.0
30         NaN          5.0    109.0           75.0           26.0         0.0
```

31	NaN	3.0	158.0	76.0	36.0	245.0
32	NaN	3.0	88.0	58.0	11.0	54.0
33	NaN	6.0	92.0	92.0	0.0	0.0
34	NaN	10.0	122.0	78.0	31.0	0.0
35	NaN	4.0	103.0	60.0	33.0	192.0
36	NaN	11.0	138.0	76.0	0.0	0.0
37	NaN	9.0	102.0	76.0	37.0	0.0
38	NaN	2.0	90.0	68.0	42.0	0.0
39	NaN	4.0	111.0	72.0	47.0	207.0
40	NaN	3.0	180.0	64.0	25.0	70.0
41	NaN	7.0	133.0	84.0	0.0	0.0
42	NaN	7.0	106.0	92.0	18.0	0.0
43	NaN	9.0	171.0	110.0	24.0	240.0
44	NaN	7.0	159.0	64.0	0.0	0.0
45	NaN	0.0	180.0	66.0	39.0	0.0
46	NaN	1.0	146.0	56.0	0.0	0.0
47	NaN	2.0	71.0	70.0	27.0	0.0
48	NaN	7.0	103.0	66.0	32.0	0.0
49	NaN	7.0	105.0	0.0	0.0	0.0

	BMI	DiabetesPedigreeFunction	Age	Outcome
0	33.6	0.627	50.0	1.0
1	26.6	0.351	31.0	0.0
2	23.3	0.672	32.0	1.0
3	28.1	0.167	21.0	0.0
4	43.1	2.288	33.0	1.0
5	25.6	0.201	30.0	0.0
6	31.0	0.248	26.0	1.0
7	35.3	0.134	29.0	0.0
8	30.5	0.158	53.0	1.0
9	0.0	0.232	54.0	1.0
10	37.6	0.191	30.0	0.0
11	38.0	0.537	34.0	1.0
12	27.1	1.441	57.0	0.0
13	30.1	0.398	59.0	1.0
14	25.8	0.587	51.0	1.0
15	30.0	0.484	32.0	1.0
16	45.8	0.551	31.0	1.0
17	29.6	0.254	31.0	1.0
18	43.3	0.183	33.0	0.0
19	34.6	0.529	32.0	1.0
20	39.3	0.704	27.0	0.0
21	35.4	0.388	50.0	0.0
22	39.8	0.451	41.0	1.0
23	29.0	0.263	29.0	1.0
24	36.6	0.254	51.0	1.0
25	31.1	0.205	41.0	1.0

26	39.4	0.257	43.0	1.0
27	23.2	0.487	22.0	0.0
28	22.2	0.245	57.0	0.0
29	34.1	0.337	38.0	0.0
30	36.0	0.546	60.0	0.0
31	31.6	0.851	28.0	1.0
32	24.8	0.267	22.0	0.0
33	19.9	0.188	28.0	0.0
34	27.6	0.512	45.0	0.0
35	24.0	0.966	33.0	0.0
36	33.2	0.420	35.0	0.0
37	32.9	0.665	46.0	1.0
38	38.2	0.503	27.0	1.0
39	37.1	1.390	56.0	1.0
40	34.0	0.271	26.0	0.0
41	40.2	0.696	37.0	0.0
42	22.7	0.235	48.0	0.0
43	45.4	0.721	54.0	1.0
44	27.4	0.294	40.0	0.0
45	42.0	1.893	25.0	1.0
46	29.7	0.564	29.0	0.0
47	28.0	0.586	22.0	0.0
48	39.1	0.344	31.0	1.0
49	0.0	0.305	24.0	0.0

```
[21]: data=data.drop(columns=['Unnamed: 0'])
```

```
[23]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 769 entries, 0 to 768
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Pregnancies            768 non-null    float64
1   Glucose                 768 non-null    float64
2   blood pressure         768 non-null    float64
3   skin thickness         768 non-null    float64
4   Insulin                768 non-null    float64
5   BMI                    768 non-null    float64
6   DiabetesPedigreeFunction 768 non-null    float64
7   Age                    768 non-null    float64
8   Outcome                768 non-null    float64
dtypes: float64(9)
memory usage: 54.2 KB
```

```
[25]: data.isnull().sum()
```

```
[25]: Pregnancies      1
      Glucose          1
      blood pressure    1
      skin thickness    1
      Insulin          1
      BMI              1
      DiabetesPedigreeFunction  1
      Age              1
      Outcome          1
      dtype: int64
```

```
[29]: data=data.drop([768])
```

```
[30]: data.isnull().sum()
```

```
[30]: Pregnancies      0
      Glucose          0
      blood pressure    0
      skin thickness    0
      Insulin          0
      BMI              0
      DiabetesPedigreeFunction  0
      Age              0
      Outcome          0
      dtype: int64
```

```
[34]: x=data.iloc[:, :-1]
      y=data.iloc[:, -1]
      print(y)
```

```
0      1.0
1      0.0
2      1.0
3      0.0
4      1.0
...
763    0.0
764    0.0
765    0.0
766    1.0
767    0.0
Name: Outcome, Length: 768, dtype: float64
```

```
[46]: from sklearn.model_selection import train_test_split
      x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=50,random_state=0)
```

```
[47]: from sklearn.ensemble import RandomForestClassifier
classifier = RandomForestClassifier(n_estimators=6, criterion='entropy',
    ↪random_state=0)
classifier.fit(x_train, y_train)
y_pred = classifier.predict(x_test)
```

```
[48]: from sklearn.metrics import accuracy_score
accuracy=round(accuracy_score(y_pred, y_test),2)*100
print("Random Forest Algorithm Accuracy is: ",accuracy)
```

Random Forest Algorithm Accuracy is: 78.0

```
[49]: from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score,r2_score,classification_report
lr=LogisticRegression(solver='lbfgs',max_iter=1000)
lr.fit(x_train,y_train)
y_pred=lr.predict(x_test)
accuracy1=round(accuracy_score(y_pred, y_test),2)*100
print("Logistic Regression Algorithm Accuracy is: ",accuracy1)
```

Logistic Regression Algorithm Accuracy is: 86.0

```
[52]: from sklearn.neighbors import KNeighborsClassifier
knn=KNeighborsClassifier(n_neighbors=2)
knn.fit(x_train,y_train)
y_pred=knn.predict(x_test)
accuracy2=round(accuracy_score(y_pred, y_test),2)*100
print("KNN Algorithm Accuracy is: ",accuracy2)
```

KNN Algorithm Accuracy is: 76.0

```
[ ]:
```