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In [1]: import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
```

```
In [3]: df = pd.read_csv("datasets/diabetes.csv")
df.head()
```

```
Out[3]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	Pedigree	Age	Outcome
0	6	148	72	35	0	33.6	0.627	50	1
1	1	85	66	29	0	26.6	0.351	31	0
2	8	183	64	0	0	23.3	0.672	32	1
3	1	89	66	23	94	28.1	0.167	21	0
4	0	137	40	35	168	43.1	2.288	33	1

```
In [4]: df.shape
```

```
Out[4]: (768, 9)
```

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In [5]: df.columns
```

```
Out[5]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
              'BMI', 'Pedigree', 'Age', 'Outcome'],
              dtype='object')
```

```
In [6]: df.isna().sum()
```

```
Out[6]: Pregnancies      0
Glucose      0
BloodPressure  0
SkinThickness  0
Insulin      0
BMI          0
Pedigree     0
Age          0
Outcome      0
dtype: int64
```

```
In [7]: X = df.drop(["Outcome"], axis=1)
y = df["Outcome"]
```

```
In [8]: X.shape
```

```
Out[8]: (768, 8)
```

```
In [9]: y.shape
```

```
Out[9]: (768,)
```

```

In [10]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_stat

In [31]: knn = KNeighborsClassifier(n_neighbors=3)

In [32]: knn.fit(X_train, y_train)

Out[32]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                             metric_params=None, n_jobs=None, n_neighbors=3, p=2,
                             weights='uniform')

In [33]: y_pred = knn.predict(X_test)

In [34]: # accuracy score
         metrics.accuracy_score(y_test, y_pred)

Out[34]: 0.7012987012987013

```

Confusion Matrix

	Actually Positive (1)	Actually Negative (0)
Predicted Positive (1)	True Positives (TPs)	False Positives (FPs)
Predicted Negative (0)	False Negatives (FNs)	True Negatives (TNs)

```

In [35]: from sklearn.metrics import confusion_matrix

         #extracting true_positives, false_positives, true_negatives, false_negatives
         print(confusion_matrix(y_test, y_pred))
         tn, fp, fn, tp = confusion_matrix(y_test, y_pred).ravel()
         print("True Negatives: ",tn)
         print("False Positives: ",fp)
         print("False Negatives: ",fn)
         print("True Positives: ",tp)

[[83 21]
 [25 25]]
True Negatives:  83
False Positives:  21
False Negatives:  25
True Positives:  25

In [36]: #Accuracy

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```
Accuracy = (tn+tp)*100/(tp+tn+fp+fn)
print("Accuracy {:.2f}%".format(Accuracy))
```

Accuracy 70.13%:

In [37]:

```
#Precision
Precision = tp/(tp+fp)
print("Precision {:.2f}".format(Precision))
```

Precision 0.54

In [38]:

```
#Recall
Recall = tp/(tp+fn)
print("Recall {:.2f}".format(Recall))
```

Recall 0.50

In [39]:

```
#Error rate
err = (fp + fn)/(tp + tn + fn + fp)
print("Error rate {:.2f}".format(err))
```

Error rate 0.30