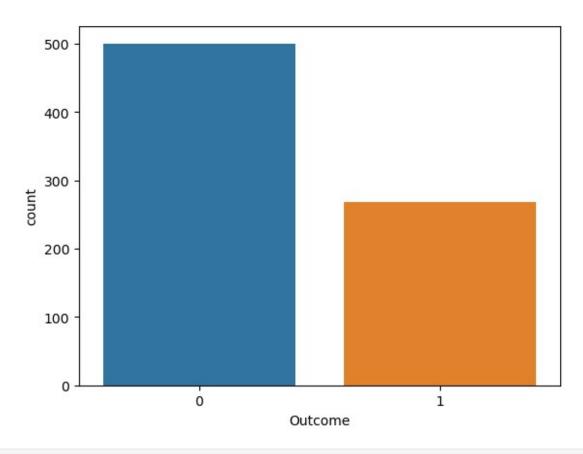
```
import pandas as pd
import numpy as np
import plotly.express as px
from sklearn.preprocessing import StandardScaler, MinMaxScaler
from sklearn.utils import resample
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
from sklearn.metrics import confusion matrix, ConfusionMatrixDisplay
from tgdm.notebook import tgdm
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")
df = pd.read csv("diabetes.csv")
df
     Pregnancies Glucose BloodPressure SkinThickness Insulin
                                                                       BMI
0
                6
                       148
                                        72
                                                        35
                                                                   0
                                                                      33.6
1
                1
                        85
                                        66
                                                        29
                                                                   0
                                                                      26.6
2
                8
                       183
                                        64
                                                         0
                                                                   0
                                                                      23.3
                        89
                                                        23
                                                                      28.1
                                        66
                                                                  94
                                        40
                0
                       137
                                                        35
                                                                 168
                                                                      43.1
763
                       101
                                        76
                                                        48
               10
                                                                 180
                                                                      32.9
                                        70
764
                2
                       122
                                                        27
                                                                      36.8
                5
765
                       121
                                        72
                                                        23
                                                                 112
                                                                      26.2
                       126
                                        60
766
                1
                                                         0
                                                                   0
                                                                      30.1
767
                        93
                                        70
                1
                                                        31
                                                                      30.4
     Pedigree
               Age
                    Outcome
0
        0.627
                 50
                           1
1
        0.351
                 31
                           0
2
                           1
        0.672
                 32
3
        0.167
                 21
                           0
4
                           1
        2.288
                 33
           . . .
                          . . .
                . . .
```

```
763
        0.171
                63
                           0
764
        0.340
                27
                           0
765
        0.245
                30
                           0
766
        0.349
                47
                           1
767
        0.315
                23
                           0
[768 rows x 9 columns]
df.describe().T
                                                   min
                                                              25%
                                           std
               count
                             mean
50% \
Pregnancies
               768.0
                         3.845052
                                     3.369578
                                                 0.000
                                                         1.00000
3.0000
Glucose
               768.0
                      120.894531
                                    31.972618
                                                 0.000
                                                        99.00000
117.0000
BloodPressure 768.0
                        69.105469
                                    19.355807
                                                 0.000
                                                        62.00000
72.0000
SkinThickness 768.0
                       20.536458
                                    15.952218
                                                 0.000
                                                         0.00000
23,0000
               768.0
                        79.799479
                                   115.244002
                                                 0.000
Insulin
                                                         0.00000
30.5000
BMI
               768.0
                        31.992578
                                     7.884160
                                                 0.000
                                                        27.30000
32,0000
               768.0
                         0.471876
                                     0.331329
                                                 0.078
Pedigree
                                                         0.24375
0.3725
Age
               768.0
                        33.240885
                                    11.760232
                                                21.000 24.00000
29.0000
Outcome
               768.0
                         0.348958
                                     0.476951
                                                 0.000
                                                         0.00000
0.0000
                      75%
                              max
Pregnancies
                 6.00000
                            17.00
                           199.00
Glucose
               140.25000
BloodPressure
                80.00000
                           122.00
                32.00000
                            99.00
SkinThickness
Insulin
               127.25000
                           846.00
                            67.10
BMI
                36.60000
Pedigree
                  0.62625
                             2.42
                41.00000
                            81.00
Age
Outcome
                 1.00000
                           1.00
df["Outcome"].value counts()
0
     500
1
     268
```

Name: Outcome, dtype: int64

plt.show()

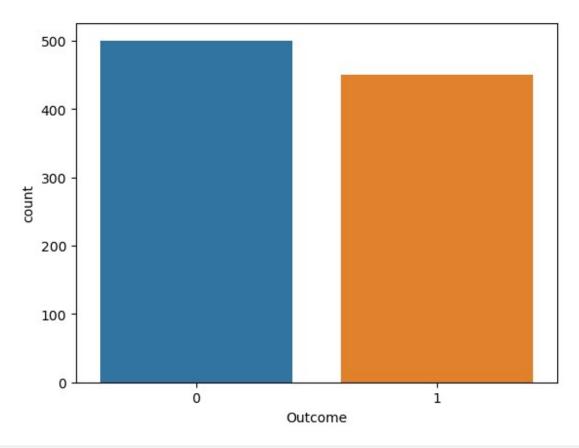
sns.countplot(data=df, x=df["Outcome"])



```
negative_data = df[df["Outcome"] == 0]
positive_data = df[df["Outcome"] == 1]

positive_upsample = resample(
    positive_data,
    replace=True,
    n_samples=int(0.9 * len(negative_data)),
    random_state=42,
)

new_df = negative_data
new_df = pd.concat([new_df, positive_upsample], ignore_index=True)
new_df.shape
(950, 9)
new_df = new_df.sample(frac=1)
sns.countplot(data=new_df, x=new_df["Outcome"])
plt.show()
```



```
x = new_df.drop("Outcome", axis=1)
y = new df[["Outcome"]]
scaler = MinMaxScaler()
scaled values = scaler.fit transform(x)
x_train, x_test, y_train, y_test = train_test_split(
    scaled_values, y, test_size=0.2, random_state=42
k_values = [1, 3, 5, 7, 9, 11, 13, 15, 17, 19, 21, 23, 25, 27, 29,
\overline{31}, 33, 35, 37, 39, 41, 43, 45, 47, 49]
accuracy values = []
for i in tqdm(range(len(k values))):
    model = KNeighborsClassifier(n_neighbors=k_values[i])
    model.fit(x train, y train)
    y_pred = model.predict(x_test)
    accuracy = metrics.accuracy_score(y_test, y_pred)
    accuracy values.append(accuracy)
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ion_minor":0}
px.line(x=k_values, y=accuracy_values)
```

```
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optimal k = -1
optimal accuracy = -1
for i in list(zip(k values, accuracy values)):
    if i[1] > optimal accuracy:
```

```
optimal k = i[0]
        optimal accuracy = i[1]
knn model = KNeighborsClassifier(n neighbors=optimal k)
knn model.fit(x train, y train)
KNeighborsClassifier(n neighbors=1)
y pred = knn model.predict(x test)
print(metrics.classification report(y test, y pred))
                           recall f1-score
              precision
                                               support
           0
                   0.90
                             0.86
                                        0.88
                                                   105
           1
                   0.83
                             0.88
                                        0.86
                                                    85
                                        0.87
                                                   190
    accuracy
                   0.87
                             0.87
                                        0.87
                                                   190
   macro avg
weighted avg
                   0.87
                             0.87
                                       0.87
                                                   190
cm = metrics.confusion matrix(y test, y pred)
plot confusion matrix(cm)
plt.show()
                                          Traceback (most recent call
NameError
last)
Cell In[26], line 2
      1 cm = metrics.confusion matrix(y test, y pred)
----> 2 plot confusion matrix(cm)
      3 plt.show()
NameError: name 'plot confusion matrix' is not defined
y score = model.predict proba(x test)[:, 1]
false positive rate, true positive rate, threshold =
metrics.roc_curve(y_test, y_score)
print("roc auc score for DecisionTree: ",
metrics.roc auc score(y test, y score))
roc auc score for DecisionTree: 0.8794397759103642
plt.subplots(1, figsize=(10, 7))
plt.title("Receiver Operating Characteristic - KNN")
plt.plot(false positive rate, true positive rate)
plt.plot([0, 1], ls="--")
plt.plot([0, 0], [1, 0], c=".7"), plt.plot([1, 1], c=".7")
```

```
plt.ylabel("True Positive Rate")
plt.xlabel("False Positive Rate")
plt.show()
```

