In [1]: import numpy as np
 import pandas as pd
 import seaborn as sns
 import matplotlib.pyplot as plt

In [6]: df = pd.read\_csv(r"C:\Users\admin\Desktop\TE IT\archive (23)\Admission\_Pred:

In [7]: df.head()

Out[7]:

	Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
0	1	337	118	4	4.5	4.5	9.65	1	0.92
1	2	324	107	4	4.0	4.5	8.87	1	0.76
2	3	316	104	3	3.0	3.5	8.00	1	0.72
3	4	322	110	3	3.5	2.5	8.67	1	0.80
4	5	314	103	2	2.0	3.0	8.21	0	0.65

In [8]: df.shape

Out[8]: (500, 9)

In [9]: df = df.drop('Serial No.',axis=1)

In [10]: df.shape

Out[10]: (500, 8)

In [11]: df.head()

Out[11]:

	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
0	337	118	4	4.5	4.5	9.65	1	0.92
1	324	107	4	4.0	4.5	8.87	1	0.76
2	316	104	3	3.0	3.5	8.00	1	0.72
3	322	110	3	3.5	2.5	8.67	1	0.80
4	314	103	2	2.0	3.0	8.21	0	0.65

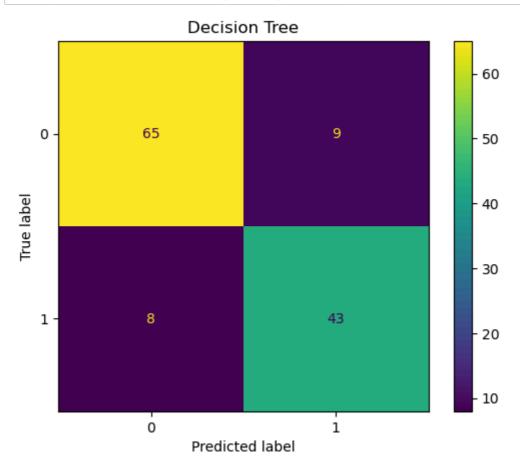
In [12]: df['Chance of Admit '] = [1 if each > 0.75 else 0 for each in df['Chance of
df.head()

Out[12]:

:		GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
	0	337	118	4	4.5	4.5	9.65	1	1
	1	324	107	4	4.0	4.5	8.87	1	1
	2	316	104	3	3.0	3.5	8.00	1	0
	3	322	110	3	3.5	2.5	8.67	1	1
	4	314	103	2	2.0	3.0	8.21	0	0

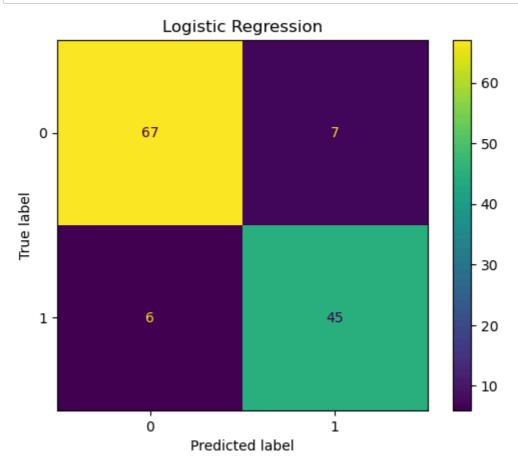
```
In [13]: x = df[['GRE Score', 'TOEFL Score', 'University Rating', 'SOP', 'LOR ', 'CGF
                 'Research']]
         y = df['Chance of Admit ']
In [14]: from sklearn.model_selection import train_test_split
In [15]: |x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.25,rande
In [16]: print(f"Size of splitted data")
         print(f"x_train {x_train.shape}")
         print(f"y_train {y_train.shape}")
         print(f"y_train {x_test.shape}")
         print(f"y_test {y_test.shape}")
         Size of splitted data
         x_train (375, 7)
         y_train (375,)
         y_train (125, 7)
         y_test (125,)
In [18]: | from sklearn.tree import DecisionTreeRegressor
         from sklearn.ensemble import RandomForestRegressor
         from sklearn.linear_model import LogisticRegression
In [19]: model_dt = DecisionTreeRegressor(random_state=1)
         model_rf = RandomForestRegressor(random_state=1)
         model_lr = LogisticRegression(random_state=1,solver='lbfgs',max_iter=1000)
In [20]: model_dt.fit(x_train,y_train)
Out[20]: DecisionTreeRegressor(random_state=1)
In [21]: model_rf.fit(x_train,y_train)
Out[21]: RandomForestRegressor(random_state=1)
In [22]: model_lr.fit(x_train,y_train)
Out[22]: LogisticRegression(max_iter=1000, random_state=1)
In [23]: |y_pred_dt = model_dt.predict(x_test) #int
         y pred rf = model rf.predict(x test) #float
         y_pred_lr = model_lr.predict(x_test) #
In [24]: y_pred_rf=[1 if each >0.75 else 0 for each in y_pred_rf]
In [25]: from sklearn.metrics import ConfusionMatrixDisplay, accuracy score
         from sklearn.metrics import classification report
```

In [27]: ConfusionMatrixDisplay.from\_predictions(y\_test,y\_pred\_dt)
 plt.title('Decision Tree')
 plt.show()
 print(f" Accuracy is {accuracy\_score(y\_test,y\_pred\_dt)}")
 print(classification\_report(y\_test,y\_pred\_dt))



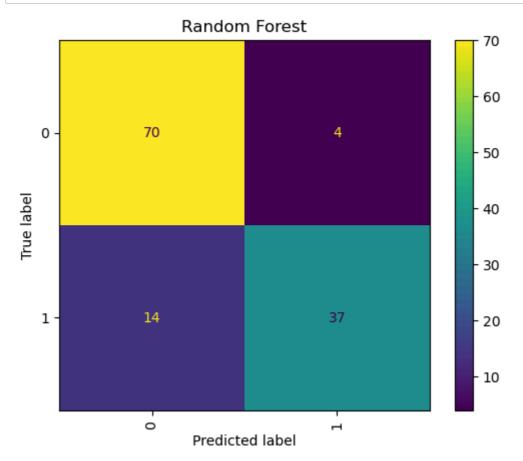
Accuracy is	0.864			
	precision	recall	f1-score	support
	-			
0	0.89	0.88	0.88	74
1	0.83	0.84	0.83	51
accuracy			0.86	125
macro avg	0.86	0.86	0.86	125
weighted avg	0.86	0.86	0.86	125

In [28]: ConfusionMatrixDisplay.from\_predictions(y\_test,y\_pred\_lr)
 plt.title('Logistic Regression')
 plt.show()
 print(f" Accuracy is {accuracy\_score(y\_test,y\_pred\_lr)}")
 print(classification\_report(y\_test,y\_pred\_lr))



Accuracy is	0.896			
	precision	recall	f1-score	support
	•			
0	0.92	0.91	0.91	74
1	0.87	0.88	0.87	51
accuracy			0.90	125
macro avg	0.89	0.89	0.89	125
weighted avg	0.90	0.90	0.90	125

```
In [29]: ConfusionMatrixDisplay.from_predictions(y_test,y_pred_rf,xticks_rotation='veplt.title('Random Forest')
    plt.show()
    print(f" Accuracy is {accuracy_score(y_test,y_pred_rf)}")
    print(classification_report(y_test,y_pred_rf))
```



Accuracy is	0.856			
	precision	recall	f1-score	support
	•			
0	0.83	0.95	0.89	74
1	0.90	0.73	0.80	51
accuracy			0.86	125
macro avg	0.87	0.84	0.85	125
weighted avg	0.86	0.86	0.85	125

In [ ]: