

	User ID	Gender	Age	Estimated Salary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0

Write a python code to implement Random Forest algorithm.

Random forest algorithm is a powerful tree learning technique in Machine Learning, which works by creating a no. of decision trees during the training phase.

Each tree is constructed using a random subset of dataset to measure a random subset of feature in each partition.

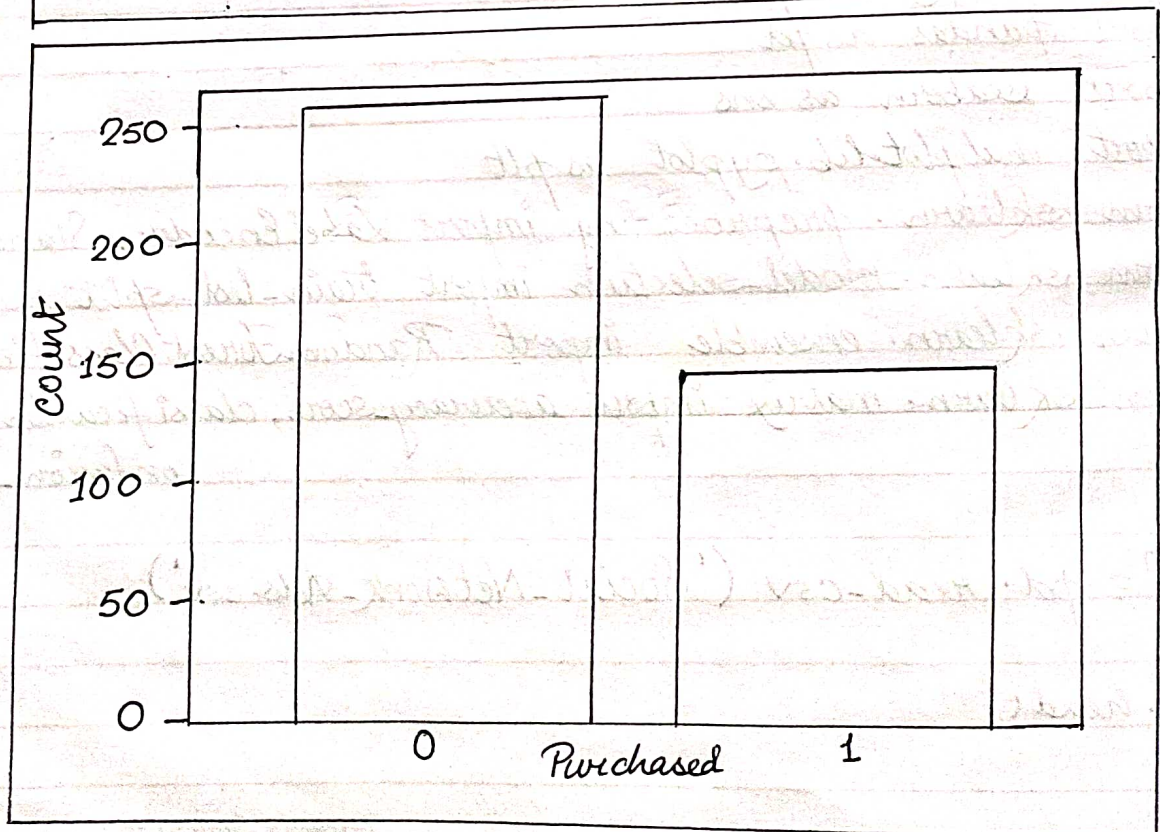
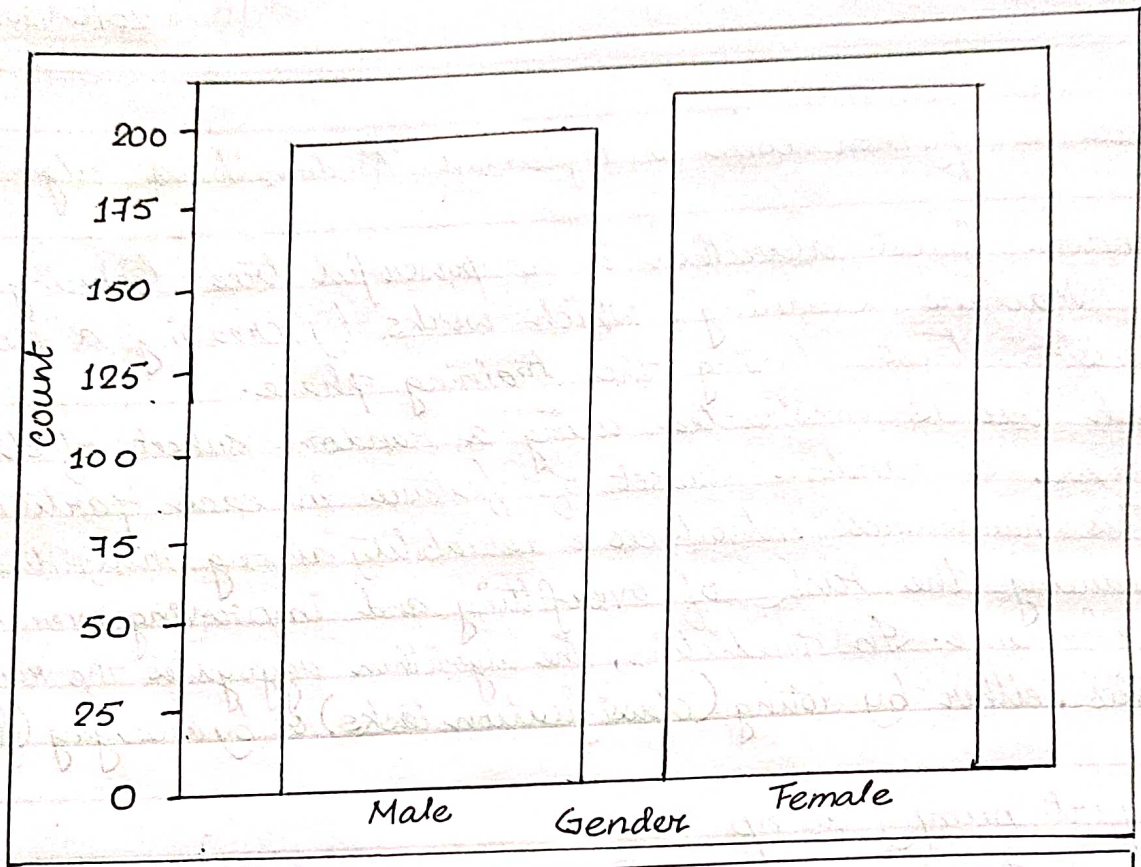
This randomness introduces a variability among individual trees, reducing the risk of overfitting and improving overall prediction performance. In prediction, the algorithm aggregates the results of all trees, either by voting (classification tasks) or by averaging (for regression tasks).

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
```

```
df = pd.read_csv('Social-Network-Ads.csv')
```

```
df.head()
```

Teacher's Signature :.....



```
ax = plt.subplots(figsize=(4,4))
```

```
ax = sns.countplot(x=df['Gender'])
```

```
plt.show()
```

```
ax = plt.subplots(figsize=(4,4))
```

```
ax = sns.countplot(x=df['Purchased'])
```

```
plt.show()
```

```
X = df.iloc[:, [1, 2, 3]].values
```

```
y = df.iloc[:, 4].values
```

```
label_encoder = LabelEncoder()
```

```
X[:, 0] = label_encoder.fit_transform(X[:, 0])
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,  
                                                    random_state=42)
```

Teacher's Signature :.....

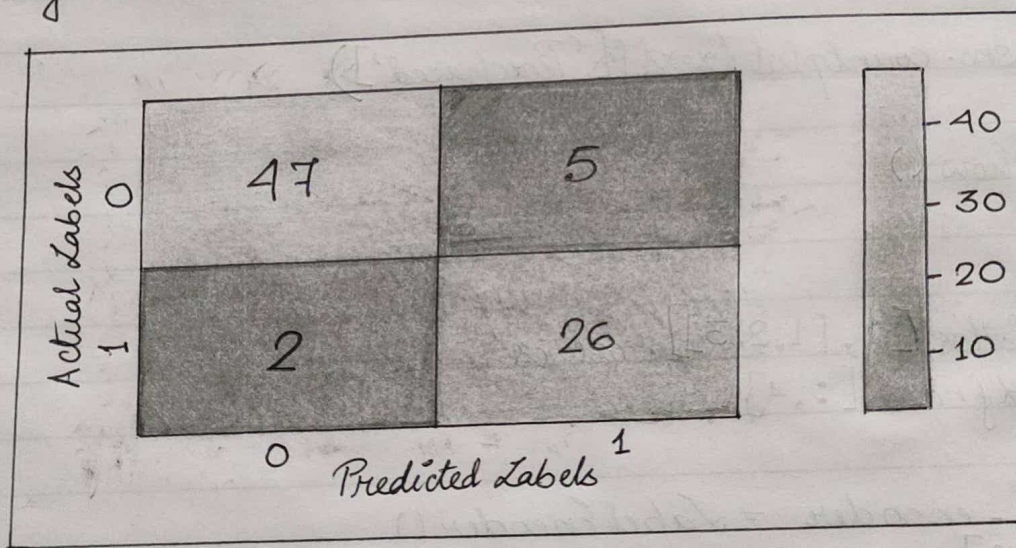
RF Accuracy : 0.9125

RF Confusion Matrix :

$\begin{bmatrix} 47 & 5 \\ 2 & 26 \end{bmatrix}$

RF Classification Report :

	precision	recall	f1-score	support
0	0.96	0.90	0.93	52
1	0.84	0.93	0.88	28
			0.91	80
accuracy			0.91	80
macro avg	0.90	0.92	0.91	80
weighted avg	0.92	0.91	0.91	80



The targeted audience is predicted not to purchase the product.


```
rf_classifier = RandomForestClassifier(n_estimators = 50)
```

```
rf_classifier.fit(X_train, y_train)
```

```
y_pred = rf_classifier.predict(X_test)
```

```
accuracy = accuracy_score(y_test, y_pred)
```

```
conf_matrix = confusion_matrix(y_test, y_pred)
```

```
classification_report_str = classification_report(y_test, y_pred)
```

```
print(f'RF Accuracy: {accuracy}')
```

```
print(f'RF Confusion Matrix: \n {conf_matrix}')
```

```
sns.set(rc={'figure.figsize': (6, 3)})
```

```
sns.heatmap(confusion_matrix(y_test, y_pred), annot=True, fmt='d')
```

```
plt.xlabel('Predicted Labels')
```

```
plt.ylabel('Actual Labels')
```

```
print(f'RF Classification Report: \n {classification_report_str}')
```

```
new_data = np.array([[0, 30, 50000]])
```

```
predicted_purchase = rf_classifier.predict(new_data)
```

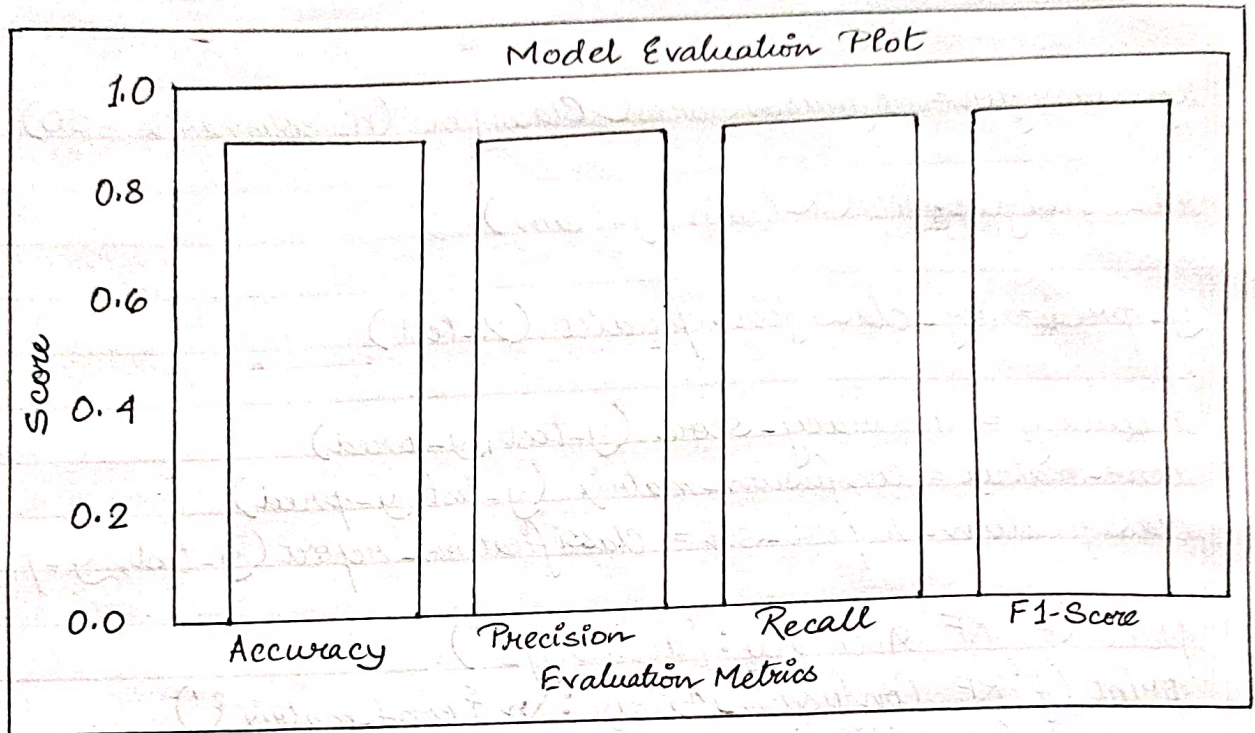
```
if predicted_purchase[0] == 1:
```

```
    print("The targeted audience is predicted to purchase the product.")
```

```
else:
```

```
    print("The targeted audience is predicted not to purchase the product.")
```

Teacher's Signature :



Cross-Validation Results:

Individual Accuracies: [0.875 0.9 0.975 0.95 0.9 0.8 0.8 0.9 0.9 0.875]

Average Accuracy: 0.8875

accuracy = 0.9125

precision = 0.91

recall = 0.91

f1-score = 0.91

metrics_names = ['Accuracy', 'Precision', 'Recall', 'F1-Score']

metrics_values = [accuracy, precision, recall, f1-score]

plt.bar(metrics_names, metrics_values, color=['blue', 'green', 'orange', 'red'])

plt.ylim([0, 1])

plt.title('Model Evaluation plot')

plt.xlabel('Evaluation Metrics')

plt.ylabel('Score')

plt.show()

from sklearn.model_selection import cross_val_score, StratifiedKFold
from sklearn.metrics import make_scorer, roc_curve, auc

cross_val = StratifiedKFold(n_splits=10, shuffle=True, random_state=42)

scoring = make_scorer(accuracy_score)

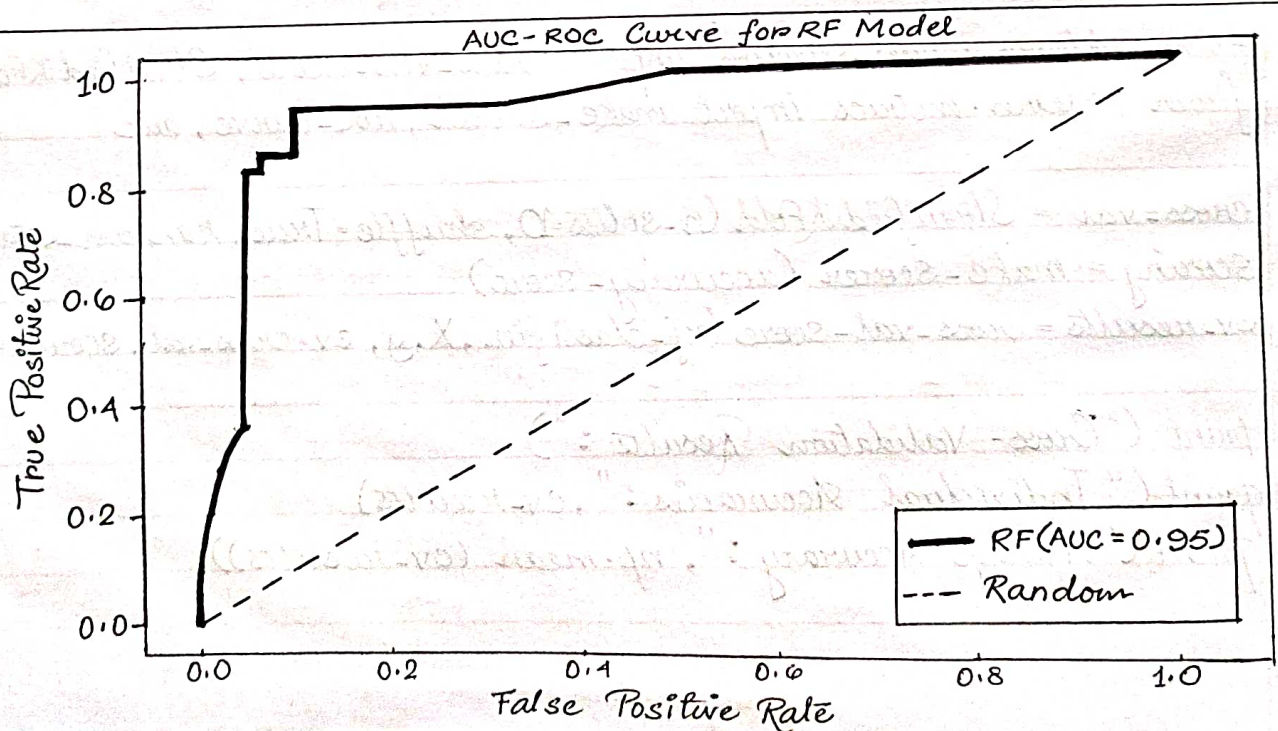
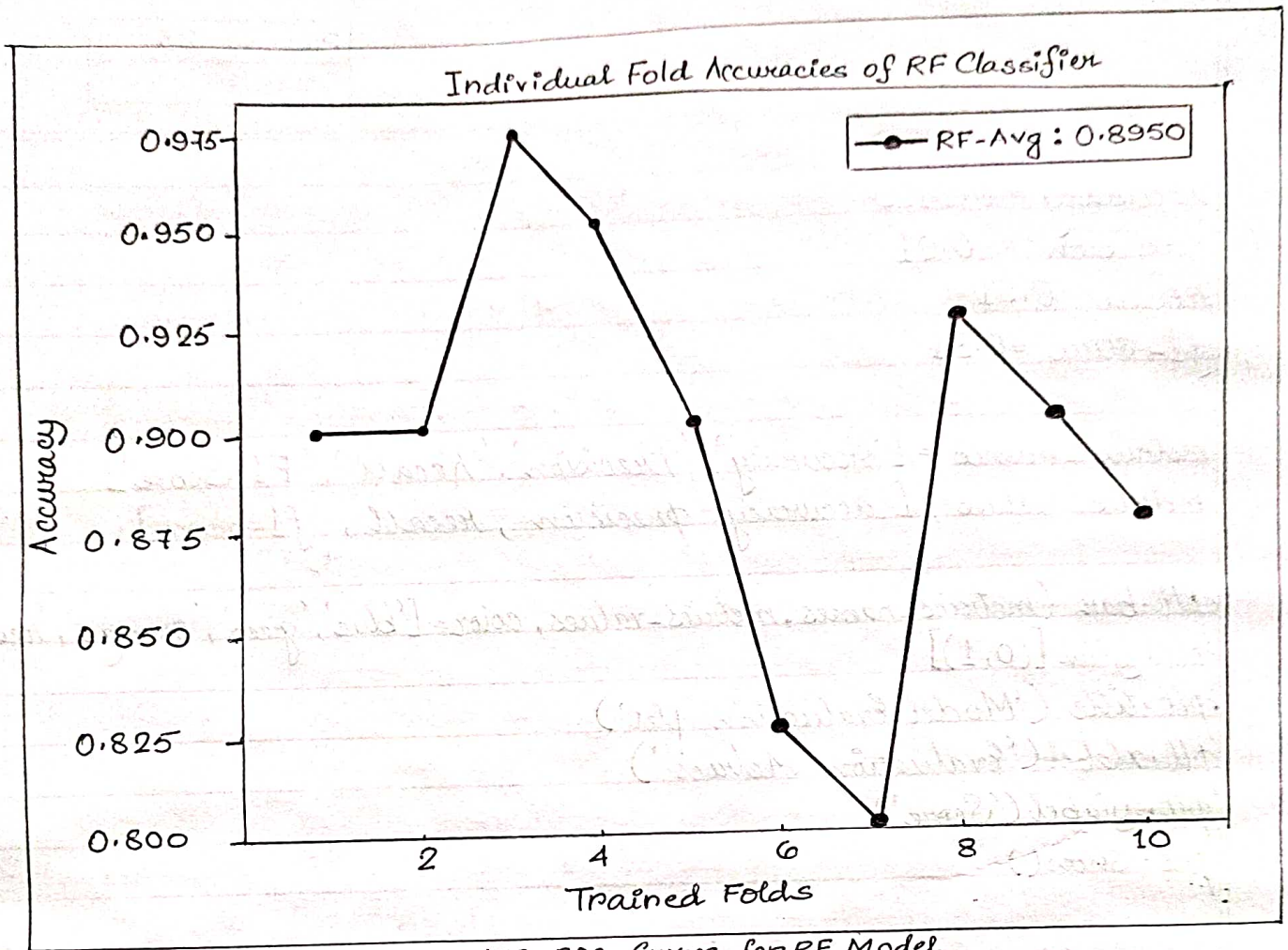
cv_results = cross_val_score(rf_classifier, X, y, cv=cross_val, scoring=scoring)

print("Cross-Validation Results :")

print("Individual Accuracies :", cv_results)

print("Average Accuracy :", np.mean(cv_results))

Teacher's Signature :.....



```

model = ['RF']
accuracies = {'RF': [0.9, 0.9, 0.975, 0.95, 0.9, 0.825, 0.8, 0.925, 0.9, 0.875], }

plt.figure(figsize=(8,4))
for model in model:
    plt.plot(range(1,11), accuracies[model], marker='o', label=f'{model} - Avg : {sum(accuracies[model])/10:.4f}')

plt.title('Individual Fold Accuracies of RF Classifier')
plt.xlabel('Trained Folds')
plt.ylabel('Accuracy')
plt.legend(bbox_to_anchor=(1.05,1), loc='upper left')
plt.show()

label_encoder = LabelEncoder()
y_test_binary = label_encoder.fit_transform(y_test)
kf_predicted_scores = kf_classifier.predict_proba(X_test)[:,-1]
kf_fpr, kf_tpr, _ = roc_curve(y_test_binary, kf_predicted_scores)
kf_auc = auc(kf_fpr, kf_tpr)

plt.figure(figsize=(8,6))
sns.set(style='darkgrid')
plt.plot(kf_fpr, kf_tpr, color='purple', lw=2, label=f'RF (AUC = {kf_auc:.2f})')
plt.plot([0,1], [0,1], linestyle='--', color='gray', label='Random')
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('AUC-ROC Curve for RF Model')
plt.legend(loc='lower right')
plt.show()

```

Teacher's Signature :