import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

data=pd.read\_csv("/content/fake\_news\_dataset.csv")

data.head()

<b>→</b>		id	title	author	text	state	date_published	source	category	sen
	0	1	Breaking News 1	Jane Smith	This is the content of article 1. It contains	Tennessee	30-11-2021	The Onion	Entertainment	
	1	2	Breaking News 2	Emily Davis	This is the content of article 2. It contains	Wisconsin	02-09-2021	The Guardian	Technology	
	2	3	Breaking News 3	John Doe	This is the content of article 3. It contains	Missouri	13-04-2021	New York Times	Sports	
	3	4	Breaking News 4	Alex Johnson	This is the content of article 4. It contains	North Carolina	08-03-2020	CNN	Sports	
	4	5	Breaking News 5	Emily Davis	This is the content of article 5. It contains	California	23-03-2022	Daily Mail	Technology	

5 rows × 24 columns

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	id	title	author	text	state	date_published	source	categor
0	1	Breaking News 1	Jane Smith	This is the content of article 1. It contains	Tennessee	30-11-2021	The Onion	Entertainmer
1	2	Breaking News 2	Emily Davis	This is the content of article 2. It contains	Wisconsin	02-09-2021	The Guardian	Technolog
2	3	Breaking News 3	John Doe	This is the content of article 3. It contains	Missouri	13-04-2021	New York Times	Sport
3	4	Breaking News 4	Alex Johnson	This is the content of article 4. It contains	North Carolina	08-03-2020	CNN	Sport
4	5	Breaking News 5	Emily Davis	This is the content of article 5. It contains	California	23-03-2022	Daily Mail	Technolog
•••								
3995	3996	Breaking News 3996	John Doe	This is the content of article 3996. It contai	Ohio	25-04-2020	InfoWars	Technolog
3996	3997	Breaking News 3997	Alex Johnson	This is the content of article 3997. It contai This is	Washington	09-01-2022	CNN	Sport

3997	3998	Breaking News 3998	Alex Johnson	content of article 3998. It contai	California	03-03-2023	Breitbart	Entertainmer
3998	3999	Breaking News 3999	John Doe	This is the content of article 3999. It contai	Illinois	13-04-2021	New York Times	Healt
3999	4000	Breaking News 4000	John Doe	This is the content of article 4000. It contai	Texas	20-12-2023	The Guardian	Healt

df.drop\_duplicates()

data.columns

data.info()

<<class 'pandas.core.frame.DataFrame'>
 RangeIndex: 4000 entries, 0 to 3999
 Data columns (total 24 columns):

#	Column	Non-Null Count	Dtype
0	id	4000 non-null	int64
1	title	4000 non-null	object
2	author	4000 non-null	object
3	text	4000 non-null	object
4	state	4000 non-null	object
5	date_published	4000 non-null	object
6	source	4000 non-null	object
7	category	4000 non-null	object
8	sentiment_score	4000 non-null	float64
9	word_count	4000 non-null	int64
10	char_count	4000 non-null	int64
11	has_images	4000 non-null	int64
12	has_videos	4000 non-null	int64
13	readability_score	4000 non-null	float64
14	num_shares	4000 non-null	int64
15	num_comments	4000 non-null	int64
16	political_bias	4000 non-null	object
17	fact_check_rating	4000 non-null	object
18	is_satirical	4000 non-null	int64

19	trust_score	4000	non-null	int64
20	source_reputation	4000	non-null	int64
21	clickbait_score	4000	non-null	float64
22	plagiarism_score	4000	non-null	float64
23	label	4000	non-null	object
				_

dtypes: float64(4), int64(10), object(10)
memory usage: 750.1+ KB

## data.isnull().sum()

<b>→</b>		0
	id	0
	title	0
	author	0
	text	0
	state	0
	date_published	0
	source	0
	category	0
	sentiment_score	0
	word_count	0
	char_count	0
	has_images	0
	has_videos	0
	readability_score	0
	num_shares	0
	num_comments	0
	political_bias	0
	fact_check_rating	0
	is_satirical	0
	trust_score	0
	source_reputation	0
	clickbait_score	0
	plagiarism_score	0
	label	0

dtype: int64

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<b>→</b>		id	title	author	text	state	date_published	source	categor
	0	1	Breaking News 1	Jane Smith	This is the content of article 1. It contains	Tennessee	30-11-2021	The Onion	Entertainmer
	1	2	Breaking News 2	Emily Davis	This is the content of article 2. It contains	Wisconsin	02-09-2021	The Guardian	Technolog
	2	3	Breaking News 3	John Doe	This is the content of article 3. It contains	Missouri	13-04-2021	New York Times	Sport
	3	4	Breaking News 4	Alex Johnson	This is the content of article 4. It contains	North Carolina	08-03-2020	CNN	Sport
	4	5	Breaking News 5	Emily Davis	This is the content of article 5. It contains	California	23-03-2022	Daily Mail	Technolog
	•••								
	3995	3996	Breaking News 3996	John Doe	This is the content of article 3996. It contai	Ohio	25-04-2020	InfoWars	Technolog
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0	1	Breaking News 1	Jane Smith	This is the content of article 1. It contains	Tennessee	30-11-2021	The Onion	Entertainmer
				3998 ils contahe content				
1	2	Breaking News 2	Emily Davis	This ผู≰ arti <b>l</b> elle	Wisconsin	02-09-2021	The Guardian	Technolog
3999	4000	Breaking News 4000	John Doe	content contains article	Texas	20-12-2023	The Guardian	Healt
2	3	Breaking News 3	John Doe	4000 It This is the content of article 3. It contains	Missouri	13-04-2021	New York Times	Sport
3	4	Breaking News 4	Alex Johnson	This is the content of article 4. It contains	North Carolina	08-03-2020	CNN	Sport
4	5	Breaking News 5	Emily Davis	This is the content of article 5. It contains	California	23-03-2022	Daily Mail	Technolog
3995	3996	Breaking News 3996	John Doe	This is the content of article 3996. It contai	Ohio	25-04-2020	InfoWars	Technolog
3996	3997	Breaking News 3997	Alex Johnson	This is the content of article 3997. It contai This is	Washington	09-01-2022	CNN	Sport

3997	3998	Breaking News 3998	Alex Johnson	content of article 3998. It contai	California	03-03-2023	Breitbart	Entertainmer
3998	3999	Breaking News 3999	John Doe	This is the content of article 3999. It contai	Illinois	13-04-2021	New York Times	Healt
3999	4000	Breaking News 4000	John Doe	This is the content of article 4000. It contai	Texas	20-12-2023	The Guardian	Healt

from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
data\_scaled=data.copy()
data\_scaled[["clickbait\_score","plagiarism\_score"]]=scaler.fit\_transform(data[["clickbait\_score","
data\_scaled

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	id	title	author	text	state	date_published	source	categor
0	1	Breaking News 1	Jane Smith	This is the content of article 1. It contains	Tennessee	30-11-2021	The Onion	Entertainmer
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4	5	Breaking News 5	Emily Davis	This is the content of article 5. It contains	California	23-03-2022	Daily Mail	Technolog
3995	3996	Breaking News 3996	John Doe	This is the content of article 3996. It contai	Ohio	25-04-2020	InfoWars	Technolog
3996	3997	Breaking News 3997	Alex Johnson	This is the content of article 3997. It contai This is	Washington	09-01-2022	CNN	Sport

3997	3998	Breaking News 3998	Alex Johnson	content of article 3998. It contai	California	03-03-2023	Breitbart	Entertainmer
3998	3999	Breaking News 3999	John Doe	This is the content of article 3999. It contai	Illinois	13-04-2021	New York Times	Healt
3999	4000	Breaking News 4000	John Doe	This is the content of article 4000. It contai	Texas	20-12-2023	The Guardian	Healt

from sklearn.preprocessing import MinMaxScaler
Scaler=MinMaxScaler()

data\_scaled[["clickbait\_score","plagiarism\_score"]]=Scaler.fit\_transform(data[["clickbait\_score","
data\_scaled

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	id	title	author	text	state	date_published	source	categor
0	1	Breaking News 1	Jane Smith	This is the content of article 1. It contains	Tennessee	30-11-2021	The Onion	Entertainmer
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3	4	Breaking News 4	Alex Johnson	This is the content of article 4. It contains	North Carolina	08-03-2020	CNN	Sport
4	5	Breaking News 5	Emily Davis	This is the content of article 5. It contains	California	23-03-2022	Daily Mail	Technolog
3995	3996	Breaking News 3996	John Doe	This is the content of article 3996. It contai	Ohio	25-04-2020	InfoWars	Technolog
3996	3997	Breaking News 3997	Alex Johnson	This is the content of article 3997. It contai This is	Washington	09-01-2022	CNN	Sport

3997	3998	Breaking News 3998	Alex Johnson	content of article 3998. It contai	California	03-03-2023	Breitbart	Entertainmer
3998	3999	Breaking News 3999	John Doe	This is the content of article 3999. It contai	Illinois	13-04-2021	New York Times	Healt
3999	4000	Breaking News 4000	John Doe	This is the content of article 4000. It contai	Texas	20-12-2023	The Guardian	Healt

data\_encoded=pd.get\_dummies(data,columns=["label"],drop\_first=True)
print(data\_encoded)

<b>→</b>	3996 3997 3998 3999	This is the con This is the con	ntent of article ntent of article ntent of article ntent of article	3998. It conta 3999. It conta	i California i Illinois
	0	date_published	source The Opies	category	sentiment_score \
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	1	02-09-2021	The Guardian	Technology	0.92
	2	13-04-2021	New York Times	Sports	0.25
	3	08-03-2020	CNN	Sports	0.94
	4	23-03-2022	Daily Mail	Technology	-0.01
	3995	25-04-2020	InfoWars	Technology	0.91
	3996	09-01-2022	CNN	Sports	-0.57

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39 39 39 39 6 6 6 7 8 8 8 8 8 8 9 8 9 9 9 9 9 9 9 9 9 9 9	clickbait_score	g import La	53.35 28.28 0.38 32.20 77.70  95.46 16.54 28.51 71.16 27.65		
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def performance_category(clickbait_score):
  if clickbait_score>=0.80:
    return "High"
  elif clickbait score>=0.50:
    return "Medium"
  else:
    return "Low"
data["performance"]=data["clickbait_score"].apply(performance_category)
print (data)
₹
             id
                               title
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     0
              1
                     Breaking News 1
                                         Jane Smith
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              2
                    Breaking News 2
                                        Emily Davis
              3
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                     Breaking News 3
                                           John Doe
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              4
                     Breaking News 4
                                       Alex Johnson
              5
     4
                     Breaking News 5
                                        Emily Davis
           3996
                 Breaking News 3996
                                           John Doe
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                 Breaking News 3997
                                       Alex Johnson
                 Breaking News 3998
                                       Alex Johnson
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           4000
                 Breaking News 4000
                                           John Doe
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                                                                       Tennessee
           This is the content of article 2. It contains ...
     1
                                                                       Wisconsin
           This is the content of article 3. It contains ...
     2
                                                                        Missouri
           This is the content of article 4. It contains ...
     3
                                                                 North Carolina
     4
           This is the content of article 5. It contains ...
                                                                      California
          This is the content of article 3996. It contai...
     3995
                                                                            Ohio
           This is the content of article 3997. It contai...
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                                                                      Washington
     3997
           This is the content of article 3998. It contai...
                                                                      California
           This is the content of article 3999. It contai...
     3998
                                                                        Illinois
     3999
           This is the content of article 4000. It contai...
                                                                           Texas
          date_published
                                                  category sentiment_score \
                                    source
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              30-11-2021
                                The Onion
                                            Entertainment
     1
              02-09-2021
                             The Guardian
                                                Technology
                                                                        0.92
```

```
2
          13-04-2021
                        New York Times
                                                                        0.25
                                                  Sports
3
          08-03-2020
                                     CNN
                                                                        0.94
                                                  Sports
4
          23-03-2022
                            Daily Mail
                                              Technology
                                                                       -0.01
                                                                         . . .
3995
          25-04-2020
                               InfoWars
                                                                        0.91
                                              Technology
3996
          09-01-2022
                                     CNN
                                                  Sports
                                                                       -0.57
3997
          03-03-2023
                              Breitbart
                                          Entertainment
                                                                       -0.17
3998
          13-04-2021
                        New York Times
                                                  Health
                                                                       -0.88
3999
          20-12-2023
                          The Guardian
                                                  Health
                                                                       -0.95
                          num_comments
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                                          political_bias
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               465
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                                                     Right
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                      trust_score
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                                 ^{\circ}
```

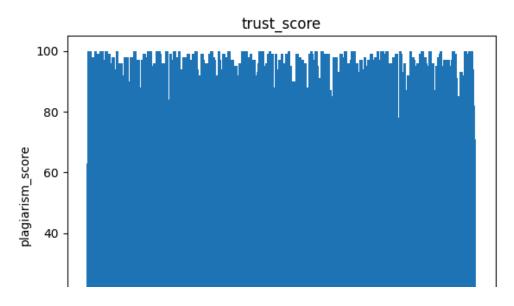
plt.bar(data["plagiarism\_score"],data["trust\_score"])
plt.xlabel("clickbait\_score")

plt.ylabel("plagiarism\_score")

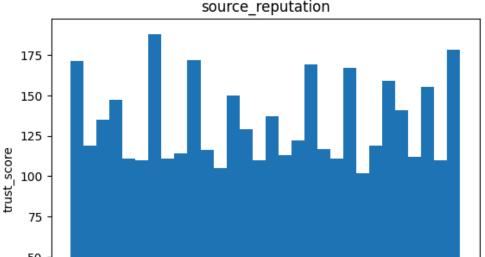
plt.title("trust\_score")

plt.show()

**₹** 

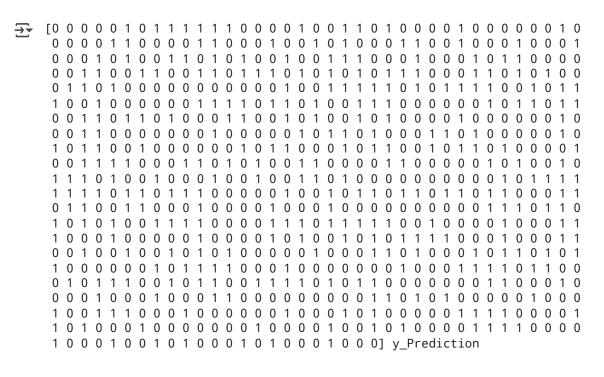


```
plt.hist(data["trust_score"], bins=30)
plt.xlabel("clickbait_score")
plt.ylabel("trust_score")
plt.title("source_reputation")
plt.show()
```

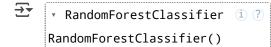


```
50
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
import seaborn as sns
# Assuming 'data' is your processed DataFrame
X = data.drop('label', axis=1) # Features
y = data['label'] # Target variable
# Convert non-numeric columns to numerical using one-hot encoding if needed
X = pd.get_dummies(X, drop_first=True)
# Split data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Initialize and train a Logistic Regression model (you can choose other models)
model = LogisticRegression()
model.fit(X_train, y_train)
    /usr/local/lib/python3.11/dist-packages/sklearn/linear_model/_logistic.py:465: C
    STOP: TOTAL NO. OF ITERATIONS REACHED LIMIT.
    Increase the number of iterations (max_iter) or scale the data as shown in:
        https://scikit-learn.org/stable/modules/preprocessing.html
    Please also refer to the documentation for alternative solver options:
        https://scikit-learn.org/stable/modules/linear_model.html#logistic-regressio
      n_iter_i = _check_optimize_result(
     ▼ LogisticRegression ① ?
     LogisticRegression()
```

```
# Make predictions on the test set
y_pred = model.predict(X_test)
print(y_pred,"y_Prediction")
```



#random forest
from sklearn.ensemble import RandomForestClassifier
model=RandomForestClassifier()
model.fit(X\_train,y\_train)



#evaluate
y\_pred=model.predict(X\_test)
print(y\_pred)

1 1 0 0 0 0 0 0 1 1 1 0 1 1 1 1 1 0 0 1 1 0 1 1 1 0 0 1 0 1 0 1 0 0 1 1 1 1 1 0 0 1 1 0 1 1 1  $0\;1\;0\;1\;1\;1\;1\;0\;1\;0\;0\;1\;1\;1\;0\;1\;1\;0\;0\;0\;1\;0\;0\;0\;1\;1\;1\;1\;0\;1\;0\;1\;0\;1\;1\;1\;0$ 1 1 1 1 1 1 0 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0]

y\_random\_model=model.predict(X\_test)
print(y random model)

 $0\;1\;0\;1\;1\;1\;1\;0\;1\;0\;0\;1\;1\;1\;0\;1\;0\;0\;1\;0\;0\;0\;1\;1\;1\;1\;0\;1\;0\;1\;0\;1\;1\;1\;0$ 1 1 1 1 1 1 0 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0]

#accuracy score, classification report, classifier matrix
from sklearn.metrics import accuracy\_score,classification\_report,confusion\_matrix
print("Accuracy Score",accuracy\_score(y\_test,y\_pred))
print("Classification Report",classification\_report(y\_test,y\_pred))
print("Confusion Matrix",confusion\_matrix(y\_test,y\_pred))

<del></del>	Accuracy Score Classification			precision	recall	f1-score	support
	0	0.53	0.56	0.54	411		
	1	0.50	0.47	0.49	389		
	accuracy			0.52	800		
	macro avg	0.52	0.52	0.51	800		
	weighted avg	0.52	0.52	0.52	800		

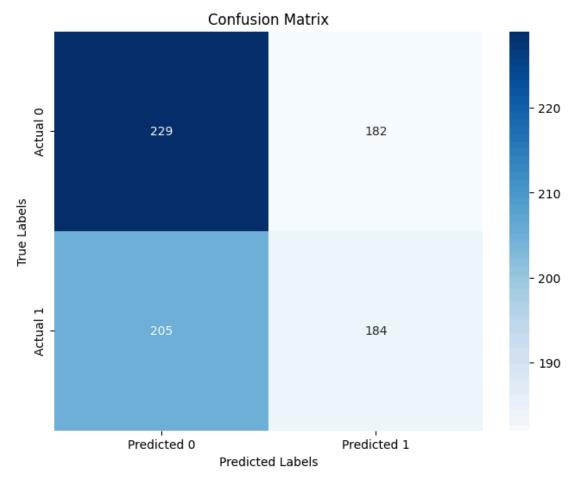
Confusion Matrix [[229 182] [205 184]]

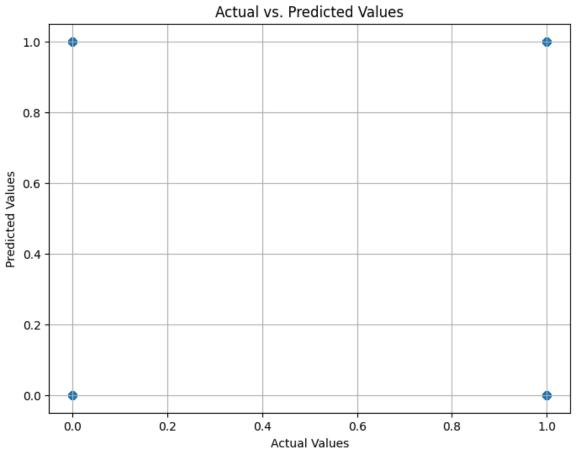
#accuracy score, classification report, classifier matrix
from sklearn.metrics import accuracy\_score,classification\_report,confusion\_matrix
print("Accuracy Score",accuracy\_score(y\_test,y\_random\_model))
print("Classification Report",classification\_report(y\_test,y\_random\_model))
print("Confusion Matrix",confusion\_matrix(y\_test,y\_random\_model))

<b>→</b>	Accuracy Classific				precision	recall	f1-score	support
		0 1	0.53 0.50	0.56 0.47	0.54 0.49	411 389		
	accur macro weighted	avg	0.52 0.52	0.52 0.52	0.52 0.51 0.52	800 800 800		

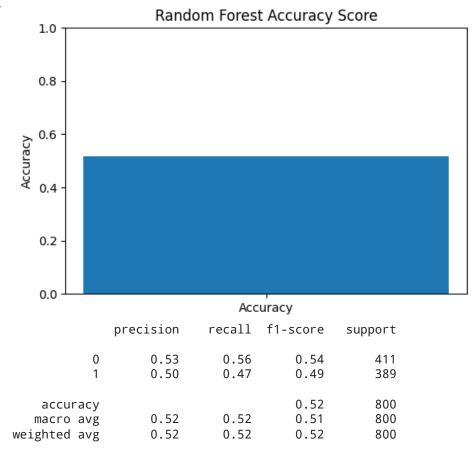
```
Confusion Matrix [[229 182]
     [205 184]]
plt.figure(figsize=(8, 6))
plt.scatter(y_test, y_pred)
plt.xlabel("Actual Values")
plt.ylabel("Predicted Values")
plt.title("Actual vs. Predicted Values")
plt.show()
₹
                                   Actual vs. Predicted Values
        1.0
        0.8
     Predicted Values
        0.6
import matplotlib.pyplot as plt
from sklearn.metrics import confusion_matrix
import seaborn as sns
# Assuming y_test and y_pred are already defined from your model evaluation
# Confusion Matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues",
            xticklabels=["Predicted 0", "Predicted 1"],
            yticklabels=["Actual 0", "Actual 1"])
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
# Actual vs Predicted Values chart
plt.figure(figsize=(8, 6))
plt.scatter(y_test, y_pred, alpha=0.5) # Added alpha for better visibility with overlapping point
plt.xlabel("Actual Values")
plt.ylabel("Predicted Values")
plt.title("Actual vs. Predicted Values")
plt.grid(True) # Added grid for better readability
```

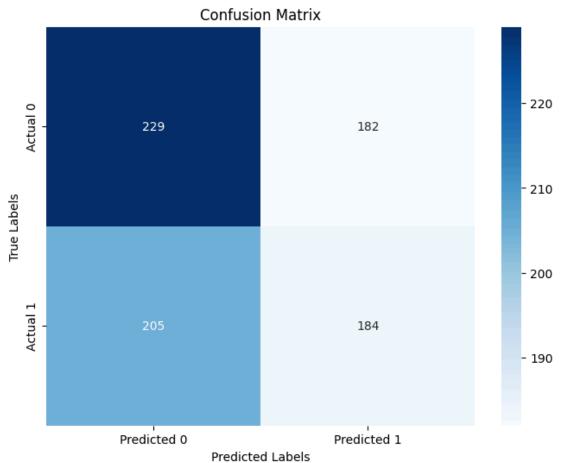
plt.show()





```
# Accuracy Score
accuracy = accuracy_score(y_test, y_random_model)
plt.figure(figsize=(6, 4))
plt.bar(['Accuracy'], [accuracy])
plt.ylim(0, 1) # Set y-axis limits for better visualization
plt.ylabel('Accuracy')
plt.title('Random Forest Accuracy Score')
plt.show()
# Classification Report
print(classification_report(y_test, y_random_model)) #printing classification report
# Confusion Matrix
cm = confusion_matrix(y_test, y_random_model)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues",
           xticklabels=["Predicted 0", "Predicted 1"],
           yticklabels=["Actual 0", "Actual 1"])
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Confusion Matrix")
plt.show()
```





```
# Assuming 'y_test', 'y_pred' (from Logistic Regression), and 'y_random_model' (from RandomForest)
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
import matplotlib.pyplot as plt
import seaborn as sns
# --- Logistic Regression Evaluation ---
print("Logistic Regression Evaluation:")
print("Accuracy Score:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test, y_pred))
cm_logistic = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(6, 4))
sns.heatmap(cm_logistic, annot=True, fmt="d", cmap="Blues",
            xticklabels=["Predicted 0", "Predicted 1"],
            yticklabels=["Actual 0", "Actual 1"])
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Logistic Regression Confusion Matrix")
plt.show()
# --- Random Forest Evaluation ---
print("\nRandom Forest Evaluation:")
print("Accuracy Score:", accuracy_score(y_test, y_random_model))
print("Classification Report:\n", classification_report(y_test, y_random_model))
cm_random_forest = confusion_matrix(y_test, y_random_model)
plt.figure(figsize=(6, 4))
sns.heatmap(cm_random_forest, annot=True, fmt="d", cmap="Blues",
            xticklabels=["Predicted 0", "Predicted 1"],
            yticklabels=["Actual 0", "Actual 1"])
plt.xlabel("Predicted Labels")
plt.ylabel("True Labels")
plt.title("Random Forest Confusion Matrix")
plt.show()
# --- Comparison ---
print("\nModel Comparison:")
print(f"Logistic Regression Accuracy: {accuracy_score(y_test, y_pred):.4f}")
print(f"Random Forest Accuracy: {accuracy_score(y_test, y_random_model):.4f}")
```