MINISTRY OF EDUCATION AND SCIENCE OF UKRAINE

NATIONAL TECHNICAL UNIVERSITY OF UKRAINE

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Software Security

Lab Work 2

Machine Learning Heart Disease Prediction

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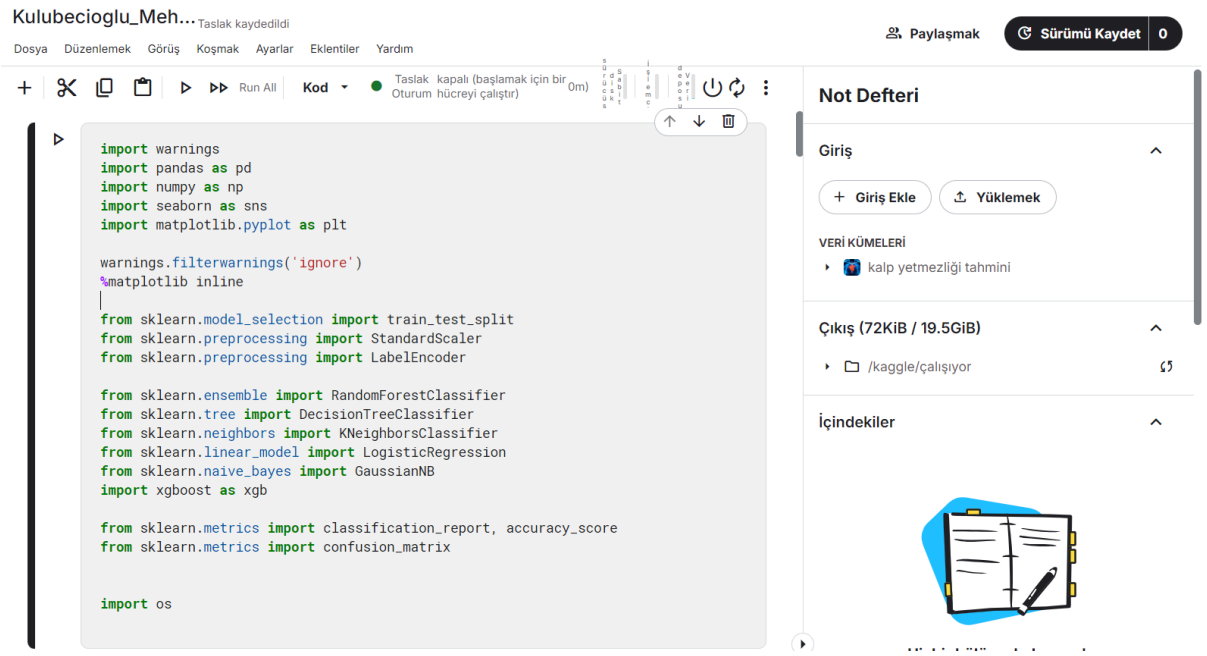
IHORY SIKORSKY KYIV POLYTECHNIC INSTITUTE

2024

Machine Learning Heart Disease Prediction Report

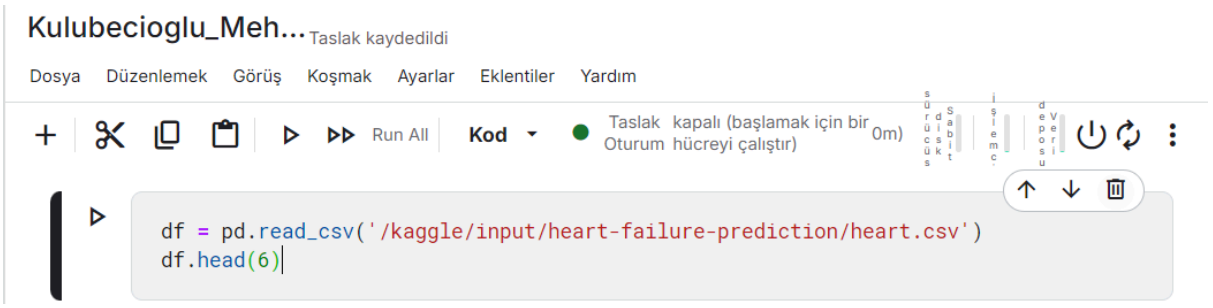
1. Importing Libraries



Explanation:

In this section, I imported all the necessary libraries for data analysis, preprocessing, and machine learning. The libraries include `pandas` for data manipulation, `seaborn` and `matplotlib` for data visualization, and several classification algorithms from `sklearn`. Additionally, I imported metrics like `accuracy_score` and `confusion_matrix` to evaluate the models.

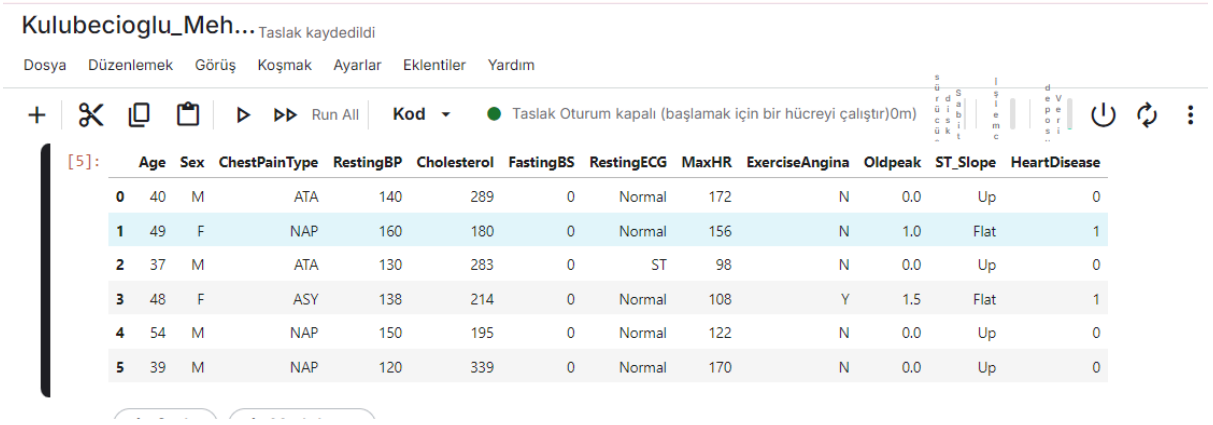
2. Reading the Dataset



Explanation:

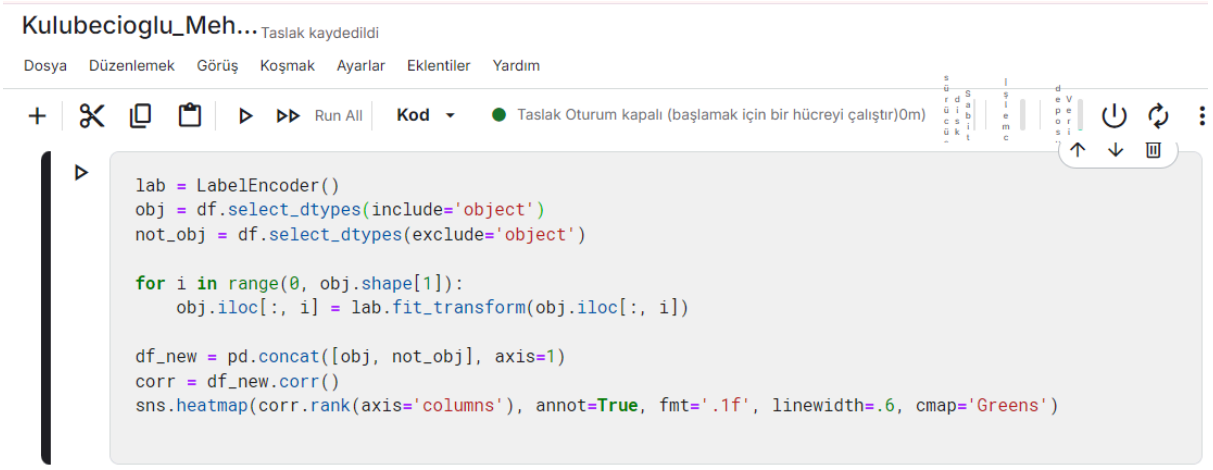
In this step, I loaded the heart disease dataset using `pandas`. The `df.head(6)` function displays the first 6 rows of the dataset, which allows me to inspect the data.

Output:



	Age	Sex	ChestPainType	RestingBP	Cholesterol	FastingBS	RestingECG	MaxHR	ExerciseAngina	Oldpeak	ST_Slope	HeartDisease
0	40	M	ATA	140	289	0	Normal	172	N	0.0	Up	0
1	49	F	NAP	160	180	0	Normal	156	N	1.0	Flat	1
2	37	M	ATA	130	283	0	ST	98	N	0.0	Up	0
3	48	F	ASY	138	214	0	Normal	108	Y	1.5	Flat	1
4	54	M	NAP	150	195	0	Normal	122	N	0.0	Up	0
5	39	M	NAP	120	339	0	Normal	170	N	0.0	Up	0

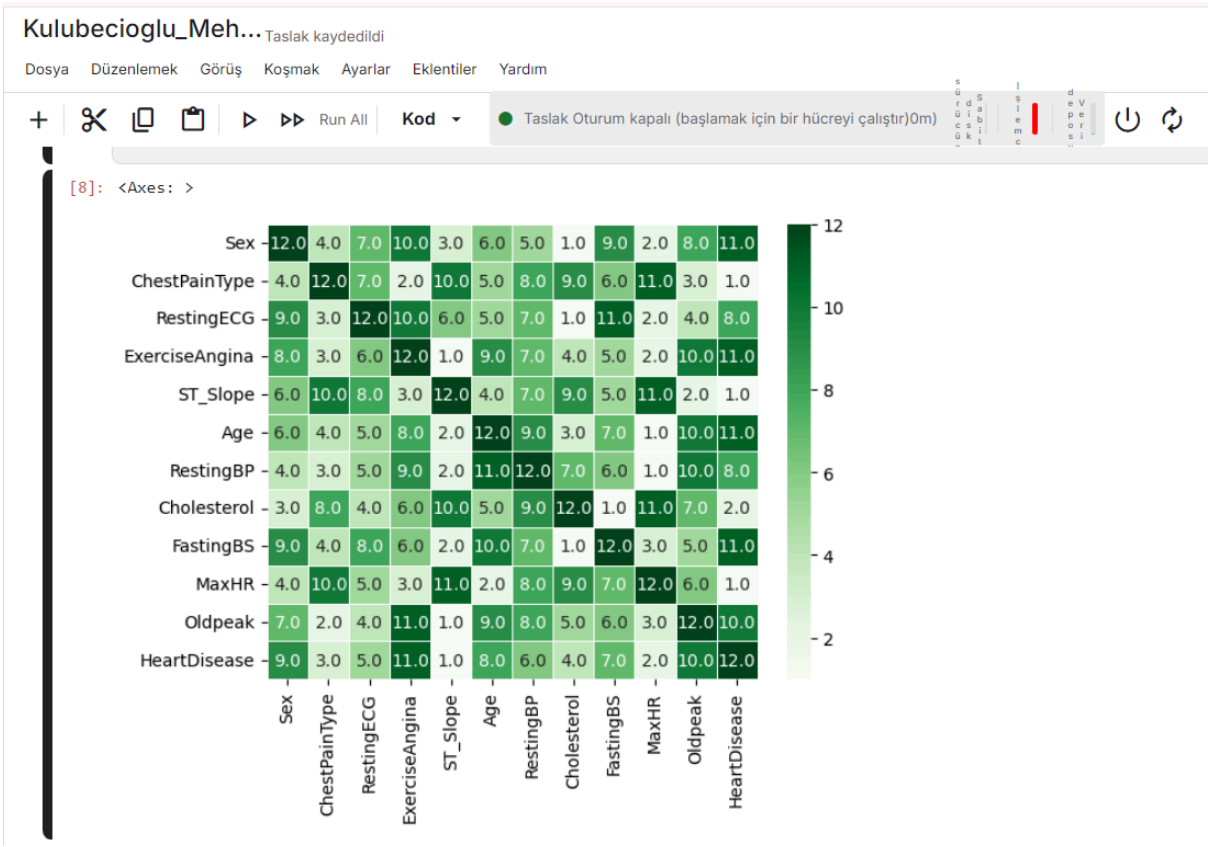
3. Label Encoding and Correlation Analysis



Explanation:

In this part, I used `LabelEncoder` to convert categorical variables into numeric form. After that, I concatenated the encoded categorical data with the numerical columns to create a new dataset (`df_new`). Then, I visualized the correlation matrix of the dataset using a heatmap to understand the relationships between the features.

Output:

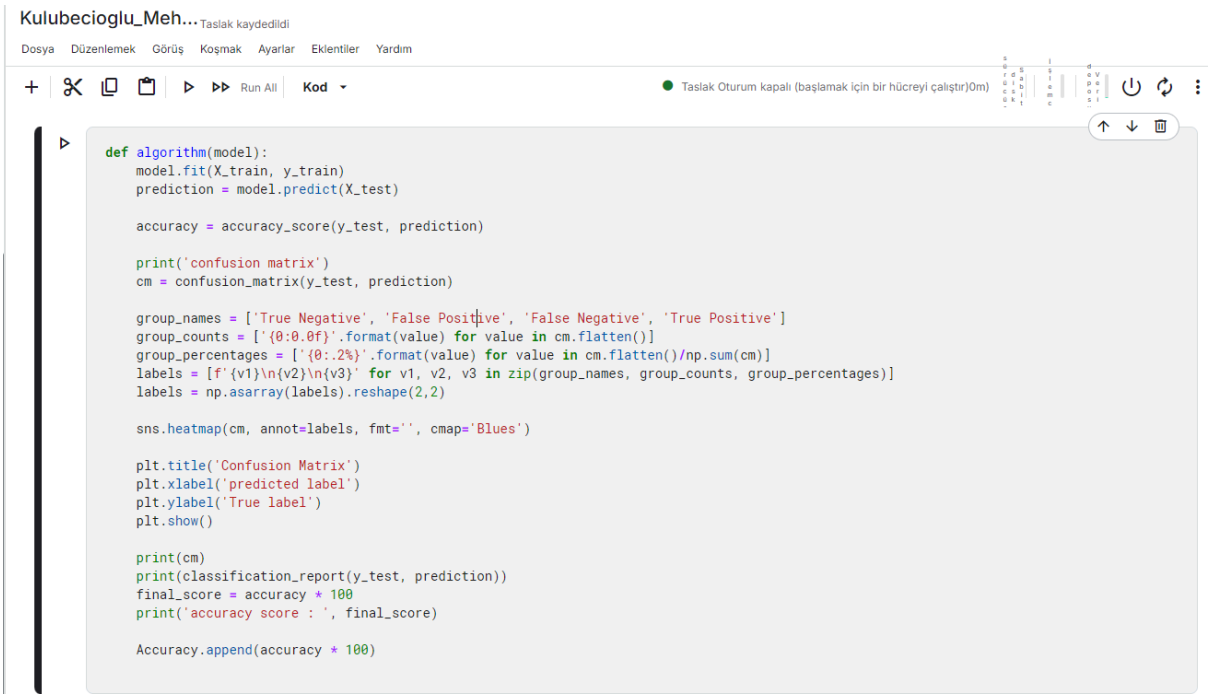


4. Splitting Data into Train and Test Sets

Explanation:

Here, I split the dataset into features (`X`) and target variable (`y`). The dataset is then split into training and testing sets with an 80-20 ratio using `train_test_split`. I also initialized an empty list (`Accuracy`) to store the accuracy of each model.

5. Function for Model Training and Evaluation



Explanation:

This function, `algorithm()`, is designed to train a given model, make predictions on the test data, and evaluate the model using a confusion matrix and a classification report. The function also plots the confusion matrix and calculates the accuracy score.

6-10. Training and Evaluating Models

Model1:

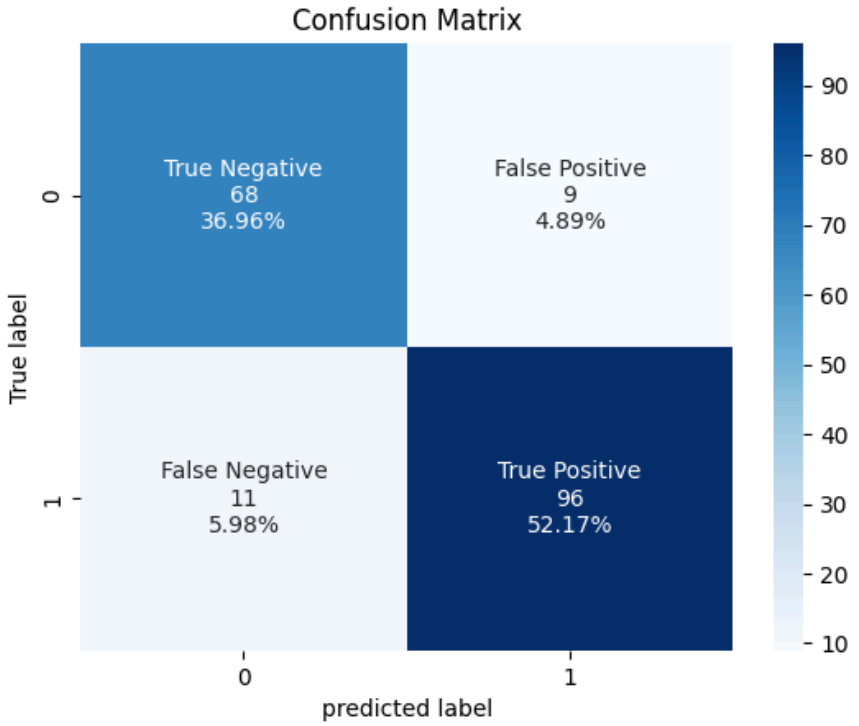
```
model_1 = RandomForestClassifier(n_estimators=100)
algorithm(model_1)
```

Kulubecioglu_Meh... Taslak kaydedildi

Dosya Düzenlemek Görüş Koşmak Ayarlar Eklentiler Yardım

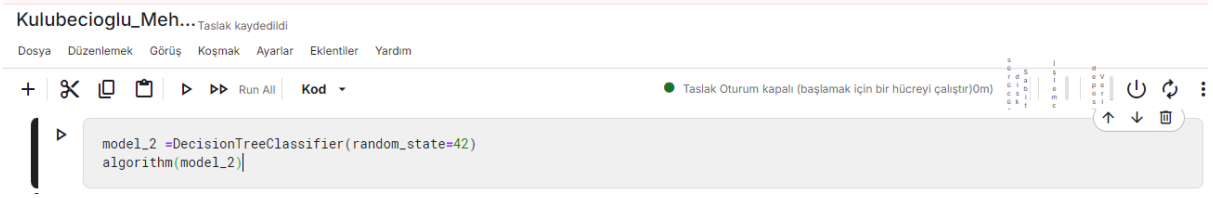
Run All Kod

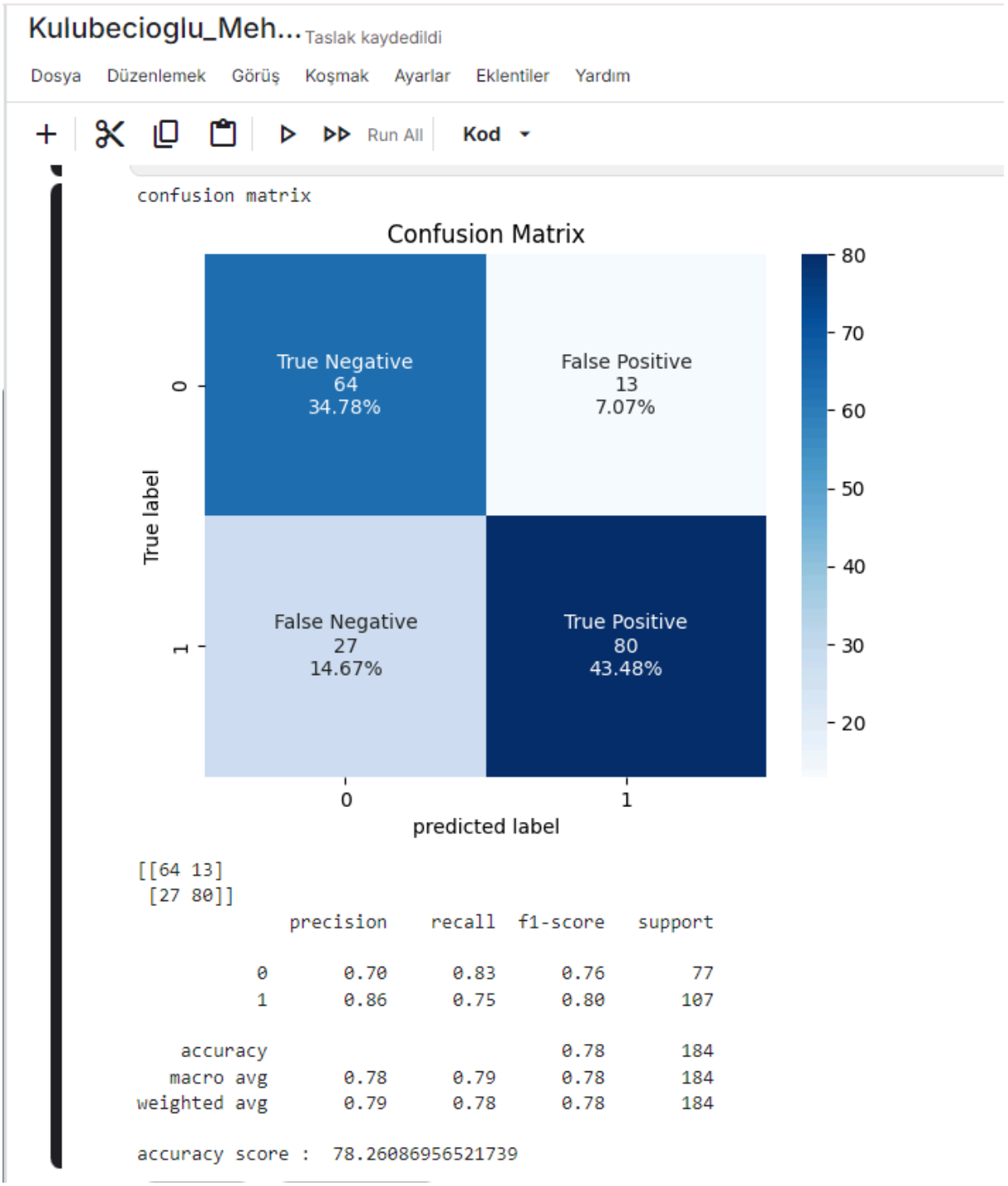
```
confusion matrix
```




```
[[68  9]
 [11 96]]
              precision    recall  f1-score   support

           0       0.86      0.88      0.87        77
           1       0.91      0.90      0.91       107

    accuracy                           0.89       184
   macro avg       0.89      0.89      0.89       184
weighted avg       0.89      0.89      0.89       184

accuracy score :  89.13043478260869
```

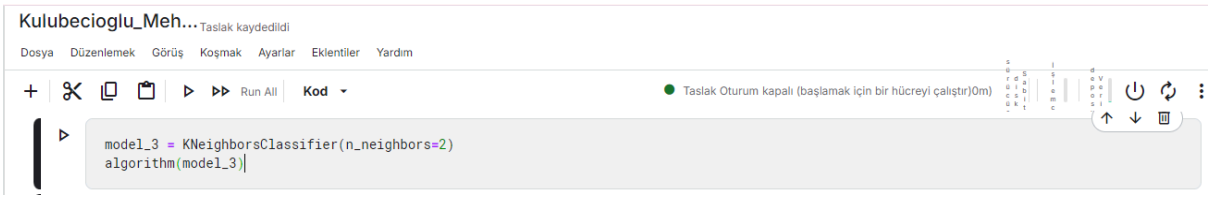
Model2:

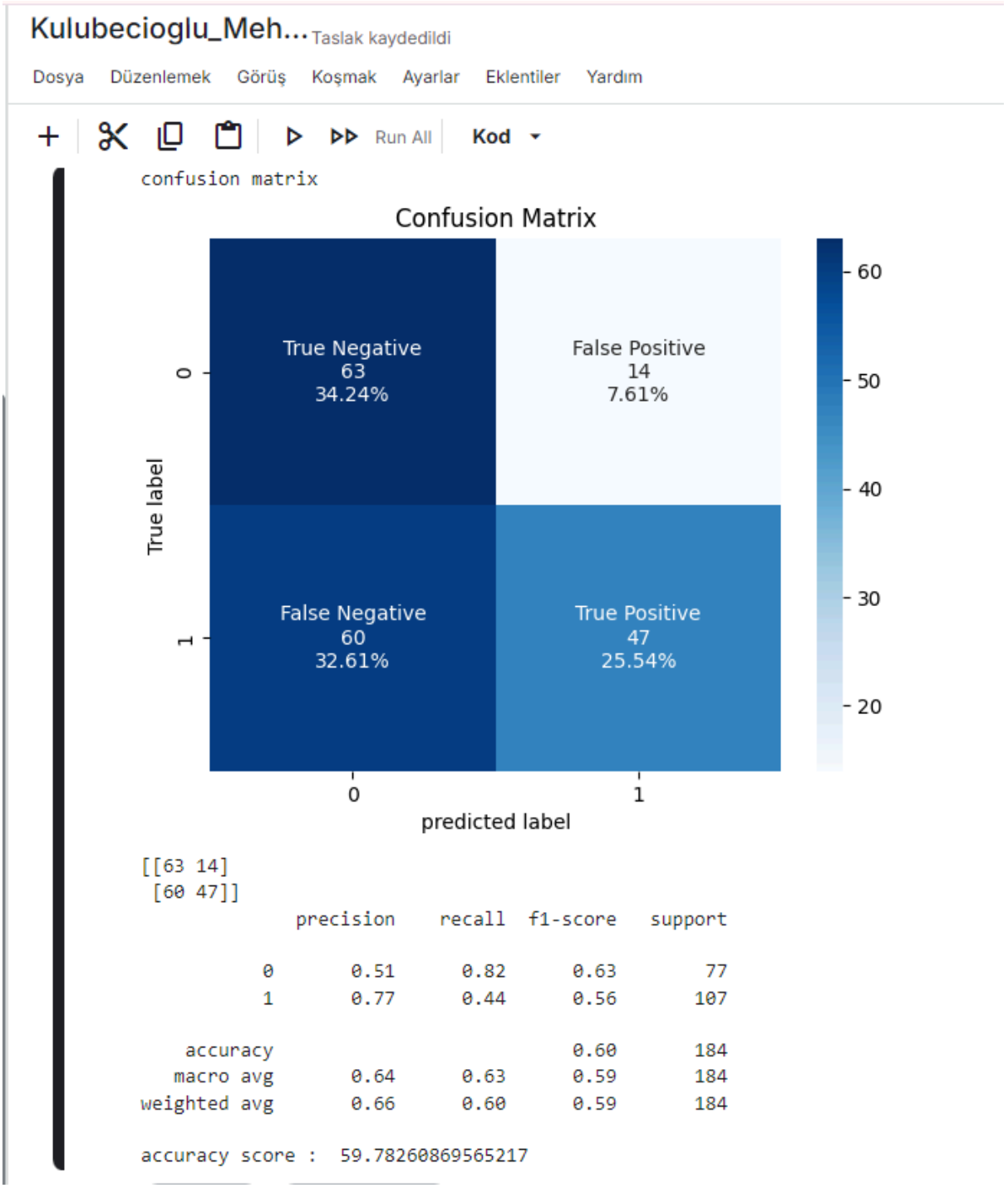





```
[[64 13]
 [27 80]]
              precision    recall  f1-score   support

           0       0.70      0.83      0.76        77
           1       0.86      0.75      0.80       107

    accuracy                           0.78       184
   macro avg       0.78      0.79      0.78       184
weighted avg       0.79      0.78      0.78       184

accuracy score :  78.26086956521739
```

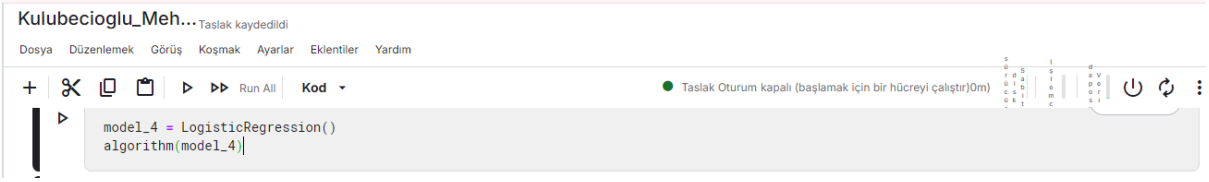
Model3:

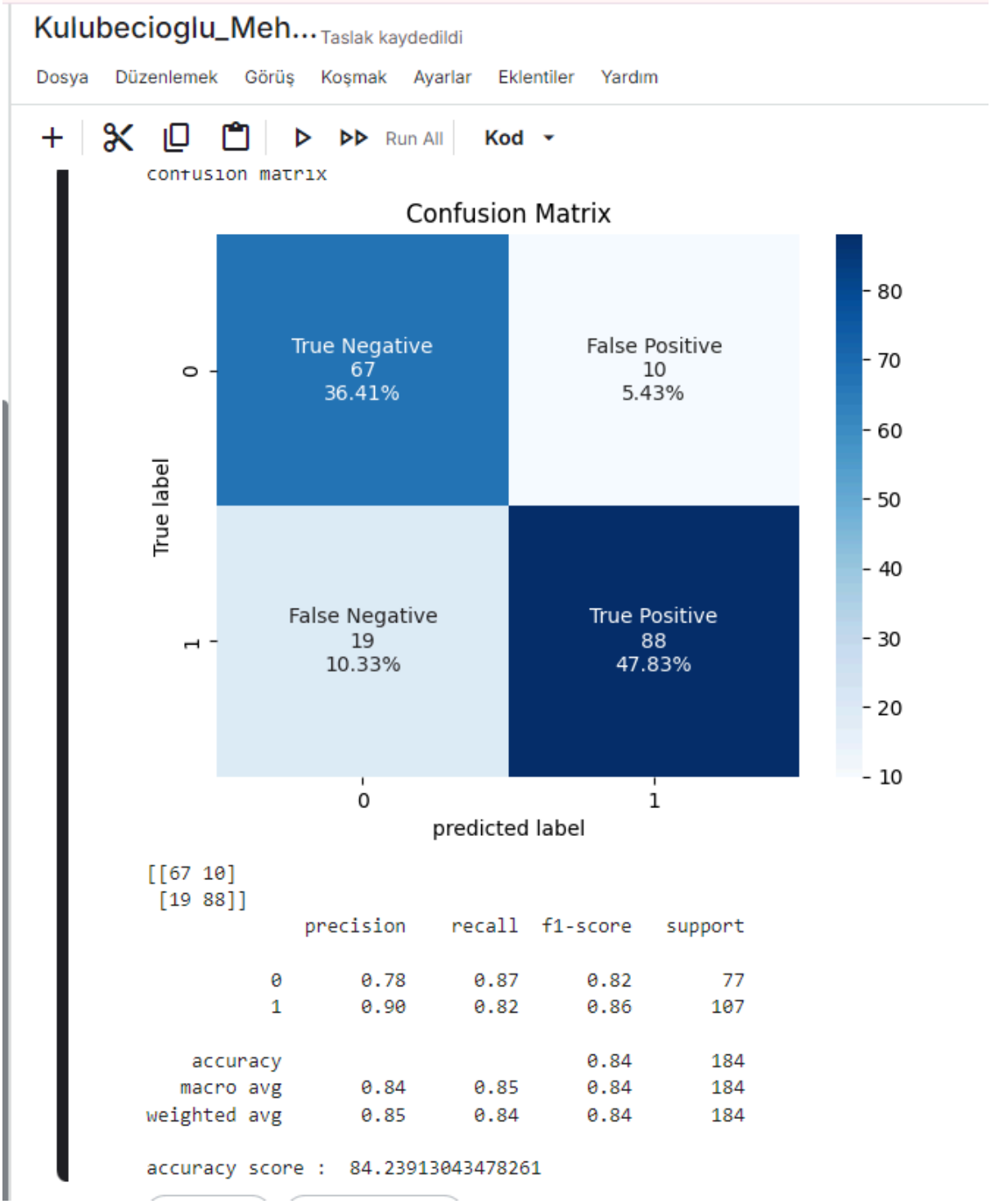





```
[[63 14]
 [60 47]]
              precision    recall  f1-score   support

           0       0.51      0.82      0.63        77
           1       0.77      0.44      0.56       107

    accuracy                           0.60       184
   macro avg       0.64      0.63      0.59       184
weighted avg       0.66      0.60      0.59       184

accuracy score :  59.78260869565217
```

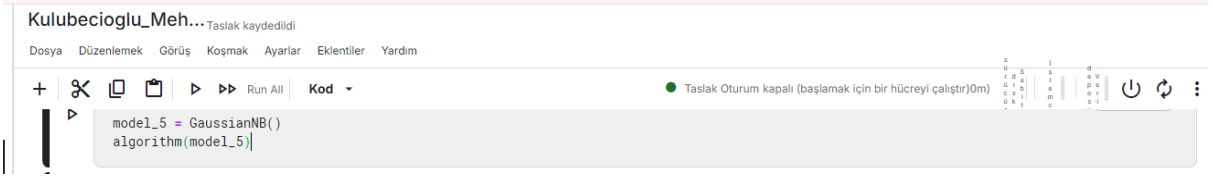
Model4:

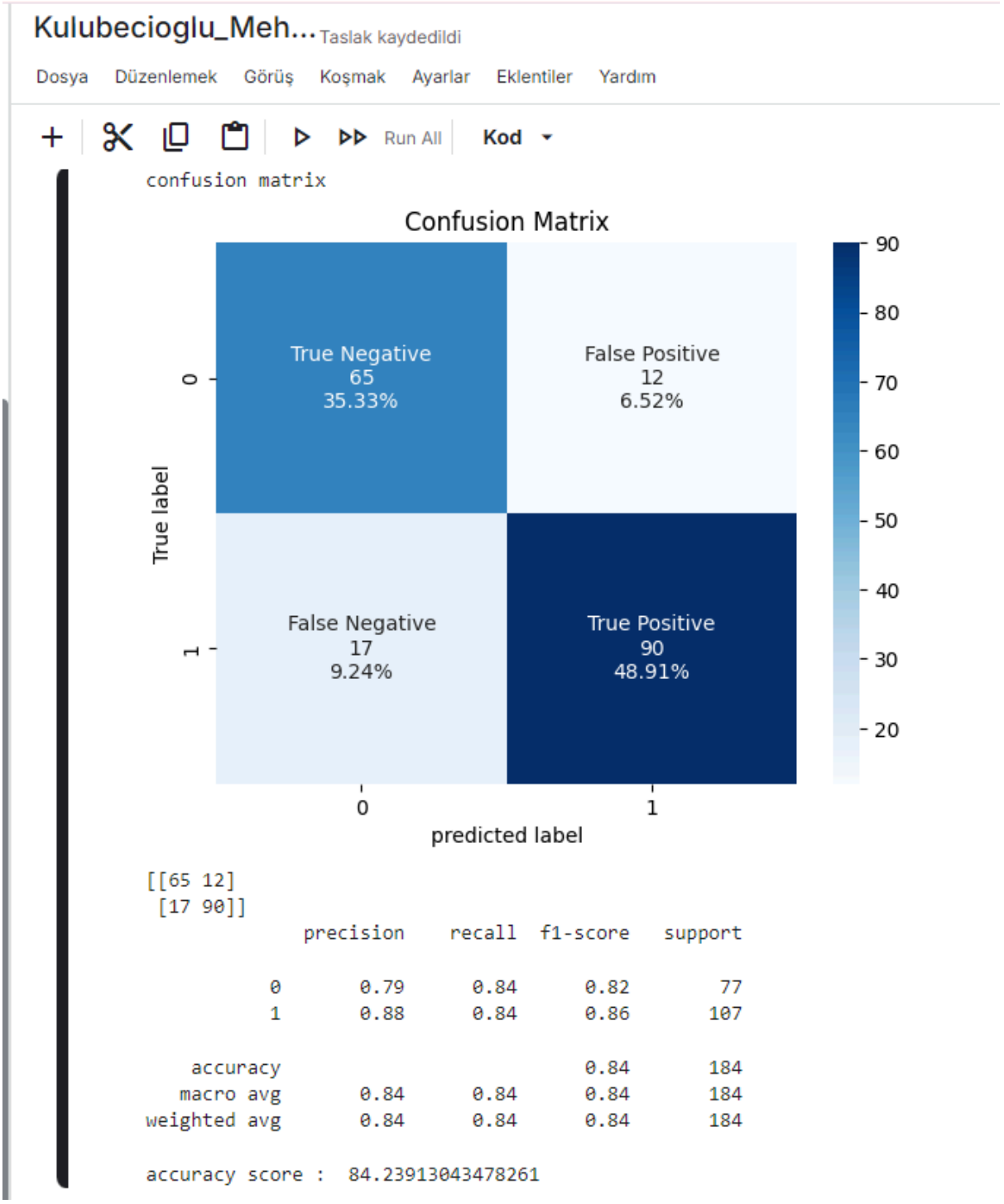





```
[[67 10]
 [19 88]]
              precision    recall  f1-score   support

           0       0.78      0.87      0.82        77
           1       0.90      0.82      0.86       107

    accuracy                           0.84       184
   macro avg       0.84      0.85      0.84       184
weighted avg       0.85      0.84      0.84       184

accuracy score :  84.23913043478261
```

Model5:



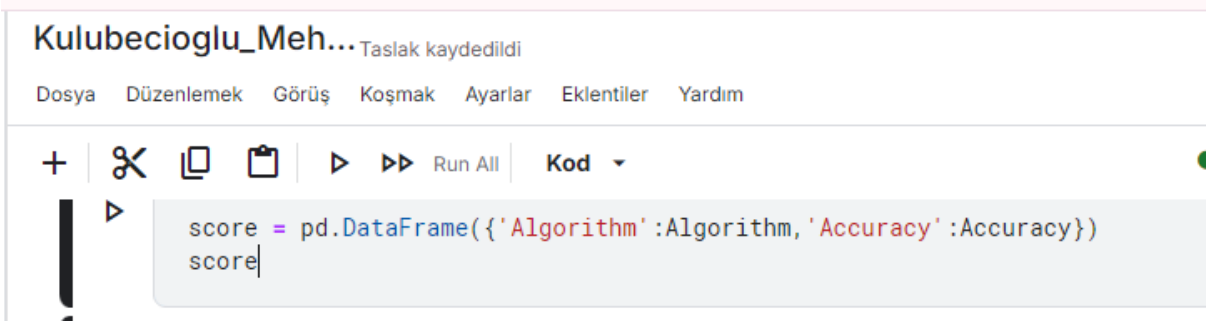



```
[[65 12]
 [17 90]]
              precision    recall  f1-score   support

           0       0.79      0.84      0.82        77
           1       0.88      0.84      0.86       107

    accuracy                           0.84       184
   macro avg       0.84      0.84      0.84       184
weighted avg       0.84      0.84      0.84       184

accuracy score :  84.23913043478261
```

Explanation:

In these steps, I applied the `algorithm()` function to five different models: `RandomForestClassifier`, `DecisionTreeClassifier`, `KNeighborsClassifier`, `LogisticRegression`, and `GaussianNB`. For each model, the function trained the model, evaluated it, and appended the accuracy to the `Accuracy` list.

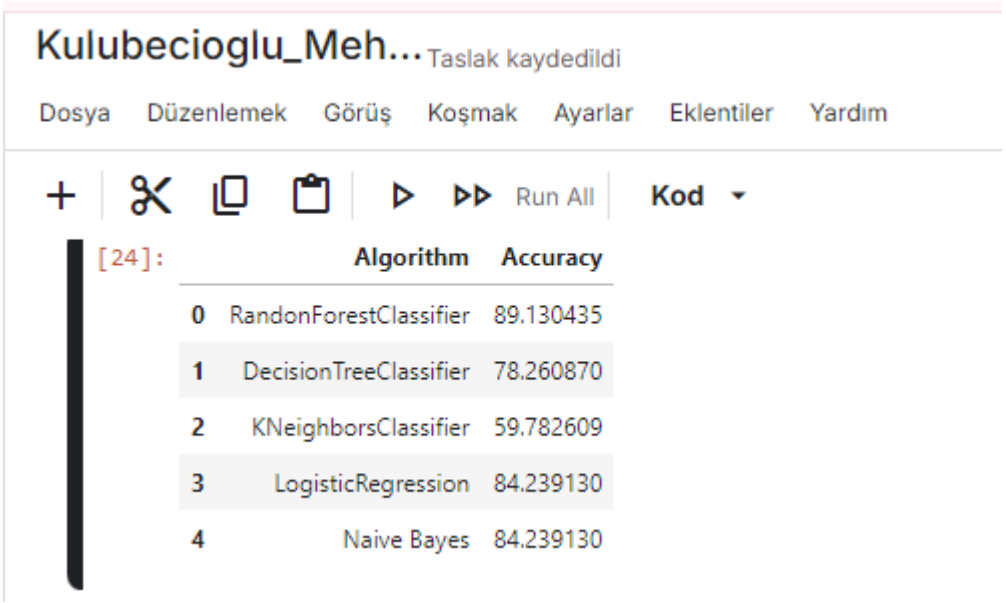
11. Creating a DataFrame for Accuracy



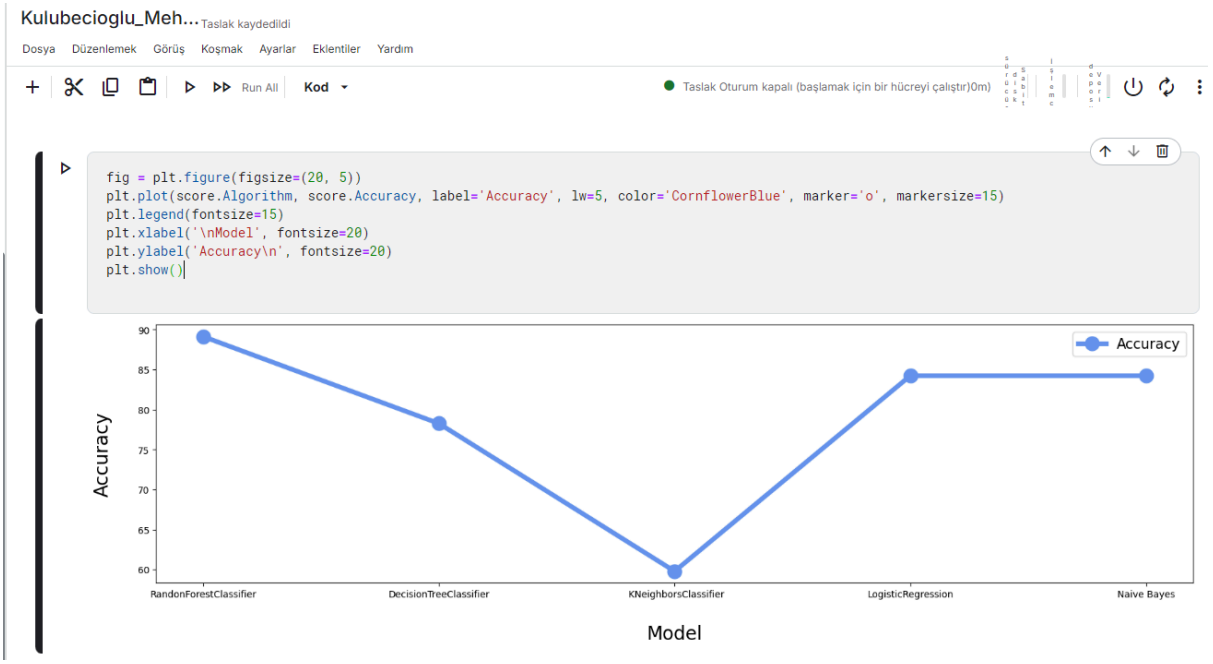
Explanation:

In this step, I created a pandas DataFrame called `score` that contains the names of the algorithms and their respective accuracies.

Output:



12. Plotting Model Accuracy



Explanation:

Finally, I plotted the accuracies of all models using a line plot. This visualization helps to compare the performance of different algorithms in terms of accuracy.