



National University
of computer and emerging sciences

NATIONAL UNIVERSITY OF COMPUTER & EMERGING SCIENCES

KARACHI CAMPUS

Browser tabs: (no subject) - k228732@nu.edu, Home Page - Select or create, ML_PROJECT (T20I CRICKET PR...

Address bar: localhost:8888/notebooks/ML_PROJECT%20(T20I%20CRICKET%20PREDICTION%20MODEL).ipynb

UPDATE: Read the [migration plan](#) to Notebook 7 to learn about the new features and the actions to take if you are using extensions - Please note that updating to Notebook 7 might break some of your extensions. Don't show anymore

Jupyter ML_PROJECT (T20I CRICKET PREDICTION MODEL) Last Checkpoint: 8 hours ago (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3 (pykernel)

CL1002 - Machine Learning - LAB
PROJECT : "CRICKET PREDICTION MODEL"

Group Members

Name	Student ID	Section
Mudasir	22K - 8732	BAI - 5A
Nihal Ali	22K - 4054	BAI - 5A
Irteza	22K - 8731	BAI - 5A

In [1]: # Importing All necessary Libraries
import pandas as pd

Windows taskbar: Search the web and Windows, 2:46 AM 14/12/2024

CLASS/SECTION: BAI-5A

DATE: 14/DEC/2024

INSTRUCTOR: SIR USAMA BIN UMAR

T20I CRICKET PREDICTION MODEL

This project aims to predict the winner of T20 International cricket matches based on historical data and statistical analysis. The model uses a **Random Forest Classifier**, which outperformed other algorithms, achieving an accuracy of 0.95 or 95%. Various machine learning models were evaluated, and an interactive GUI was developed to enhance user engagement.

DATASET OVERVIEW

The dataset is structured on a ball-by-ball basis, later aggregated by match date to eliminate duplicate rows. Key attributes include:

- Match ID, Date, Venue
- Batting and bowling statistics (runs, wickets, overs, etc.)
- Match outcomes like Winner and Toss Decision (Bat First/Second).

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TrustedPython 3 (ipykernel)

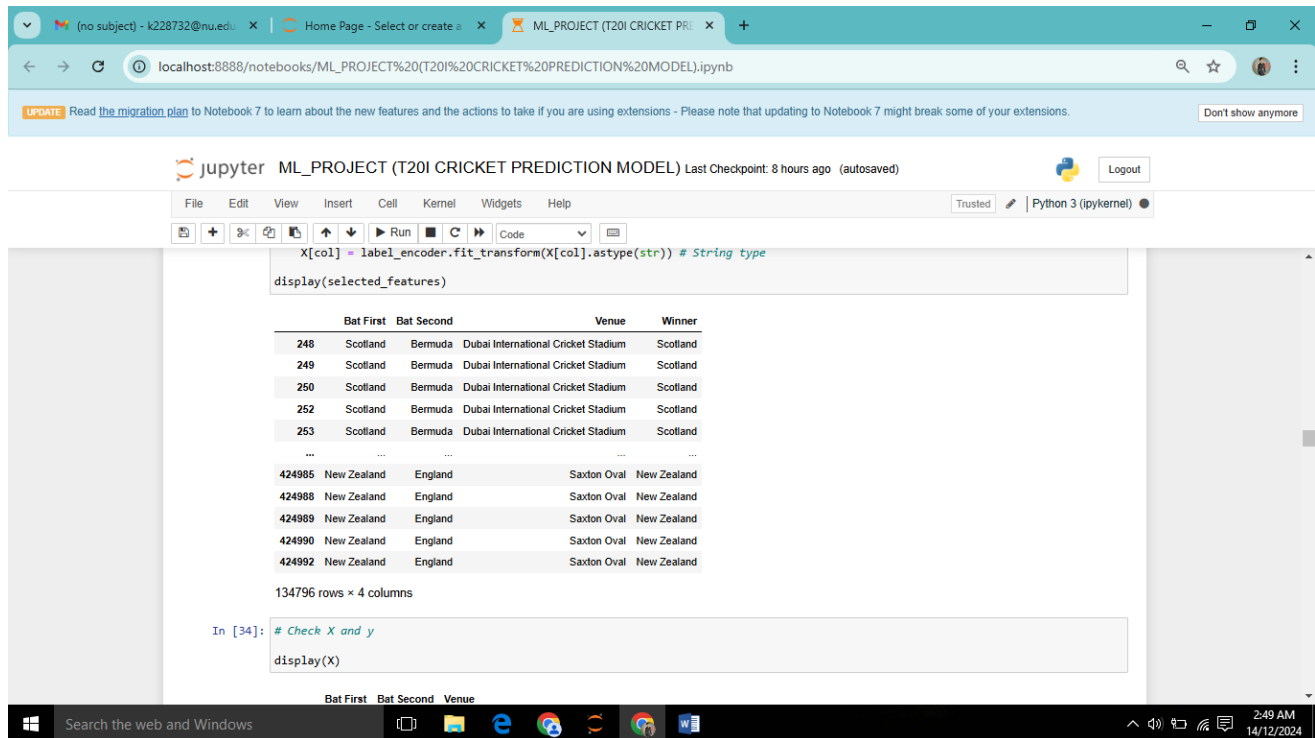
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Selected Features for the Model:

- Bat First (toss winner)
- Bat Second
- Venue
- Winner (label/target column)



The screenshot shows a Jupyter Notebook titled "ML_PROJECT (T20I CRICKET PREDICTION MODEL)". The code cell contains the following Python code:

```
X[col] = label_encoder.fit_transform(X[col].astype(str)) # String type
display(selected_features)
```

The output of the code is a table with 4 columns: Bat First, Bat Second, Venue, and Winner. The table contains 134796 rows of data. The first few rows are:

	Bat First	Bat Second	Venue	Winner
248	Scotland	Bermuda	Dubai International Cricket Stadium	Scotland
249	Scotland	Bermuda	Dubai International Cricket Stadium	Scotland
250	Scotland	Bermuda	Dubai International Cricket Stadium	Scotland
252	Scotland	Bermuda	Dubai International Cricket Stadium	Scotland
253	Scotland	Bermuda	Dubai International Cricket Stadium	Scotland
...
424985	New Zealand	England	Saxton Oval	New Zealand
424988	New Zealand	England	Saxton Oval	New Zealand
424989	New Zealand	England	Saxton Oval	New Zealand
424990	New Zealand	England	Saxton Oval	New Zealand
424992	New Zealand	England	Saxton Oval	New Zealand

Below the table, it says "134796 rows x 4 columns". The code cell also includes a comment "# Check X and y" and a call to "display(X)".

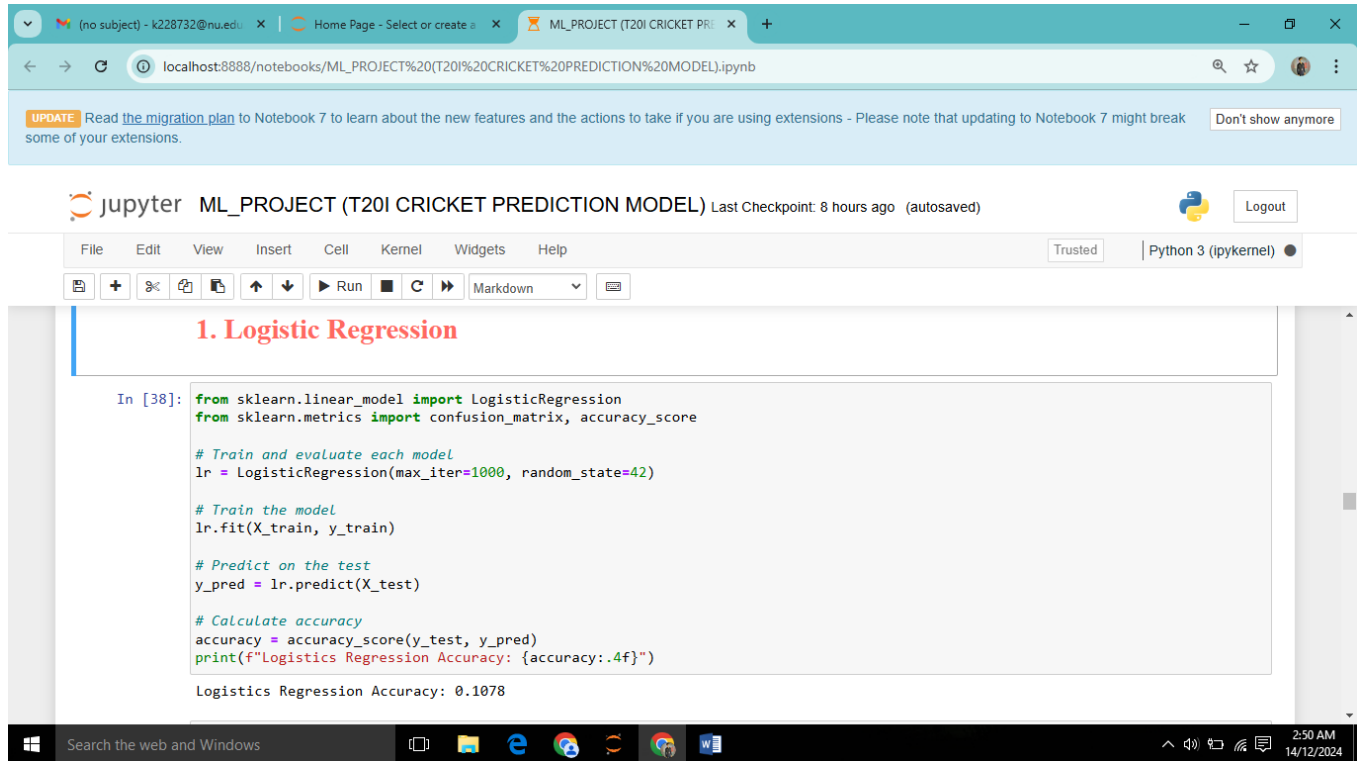
The dataset considers the toss-winning team as the team batting first.

MACHINE LEARNING MODELS EVALUATED

1. LOGISTIC REGRESSION

- Testing Accuracy: 0.1078 or 10.78%
- Reason for Low Accuracy

Logistic Regression is primarily suited for binary classification. Since the target variable includes multiple teams as possible winners, the algorithm fails to generalize.



```
In [38]: from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix, accuracy_score

# Train and evaluate each model
lr = LogisticRegression(max_iter=1000, random_state=42)

# Train the model
lr.fit(X_train, y_train)

# Predict on the test
y_pred = lr.predict(X_test)

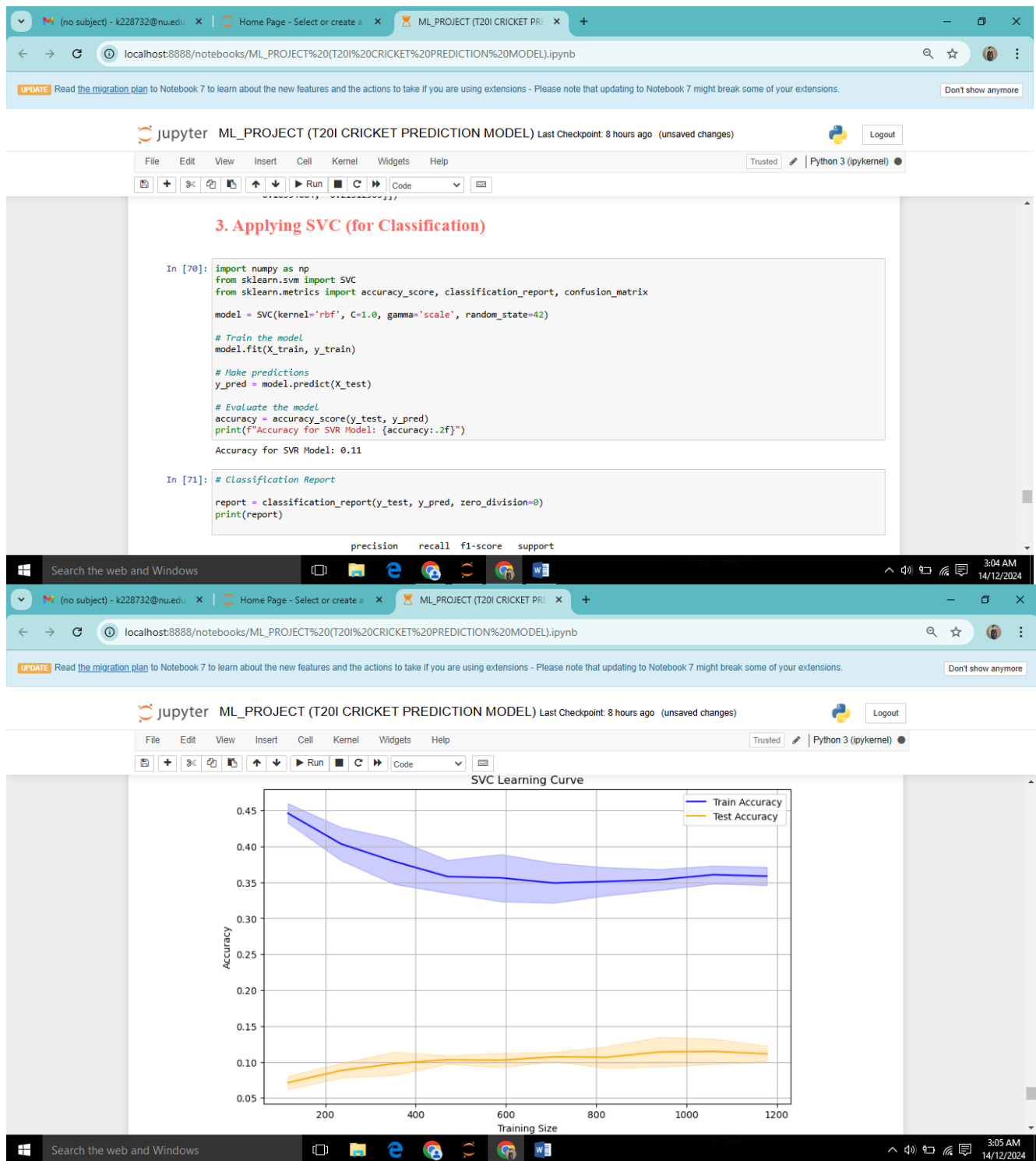
# Calculate accuracy
accuracy = accuracy_score(y_test, y_pred)
print(f"Logistics Regression Accuracy: {accuracy:.4f}")

Logistics Regression Accuracy: 0.1078
```

2. SUPPORT VECTOR CLASSIFIER (SVC)

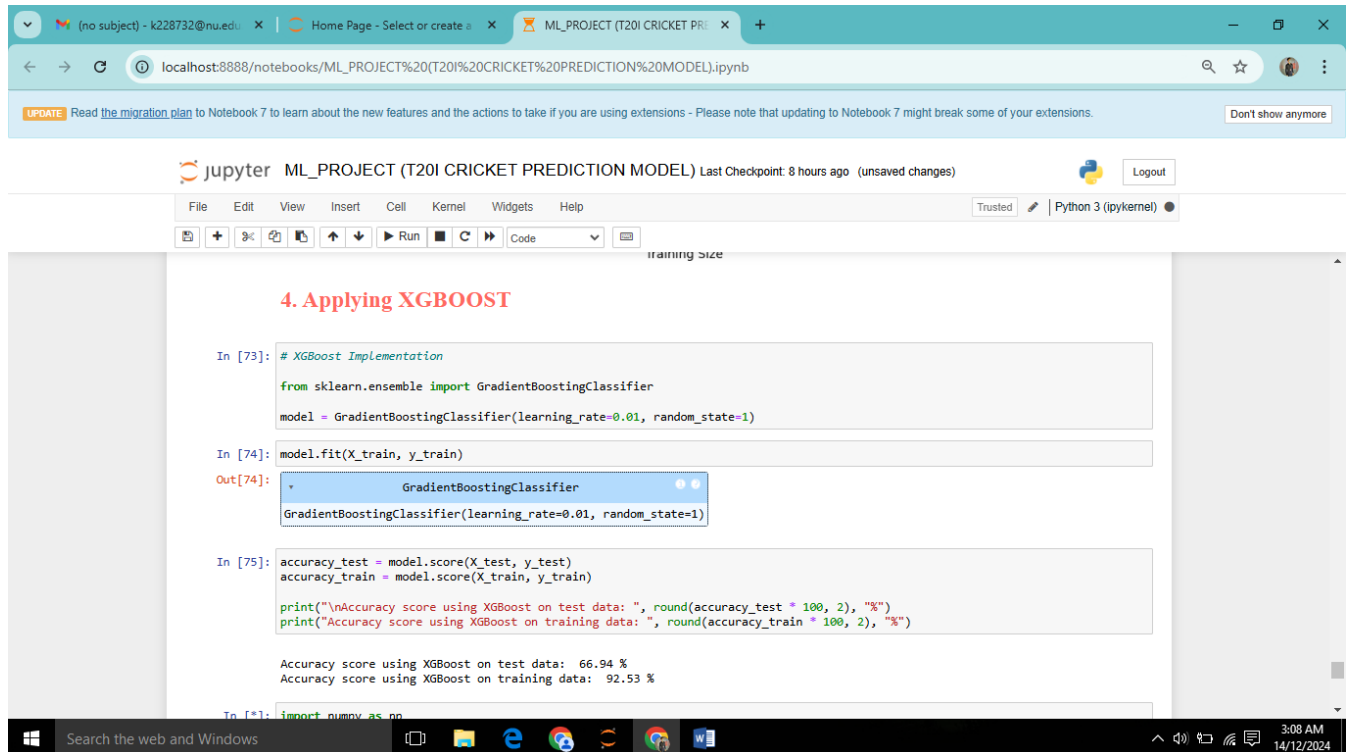
- Testing Accuracy: 0.11 or 11%
- Reason for Low Performance
SVC struggles with multi-class classification when the dataset is large and complex. The need for significant parameter tuning also impacted its performance.

Machine Learning LAB (AL3002) - Project Report



3. XGBOOST CLASSIFIER

- Testing Accuracy: 0.6694 or 66.94%
- Reason for Partial Success:
XGBoost handles classification tasks better but requires fine-tuned hyperparameters and specific handling of categorical features for optimal performance.



The screenshot shows a Jupyter Notebook interface with the following content:

4. Applying XGBOOST

```
In [73]: # XGBoost Implementation
from sklearn.ensemble import GradientBoostingClassifier
model = GradientBoostingClassifier(learning_rate=0.01, random_state=1)

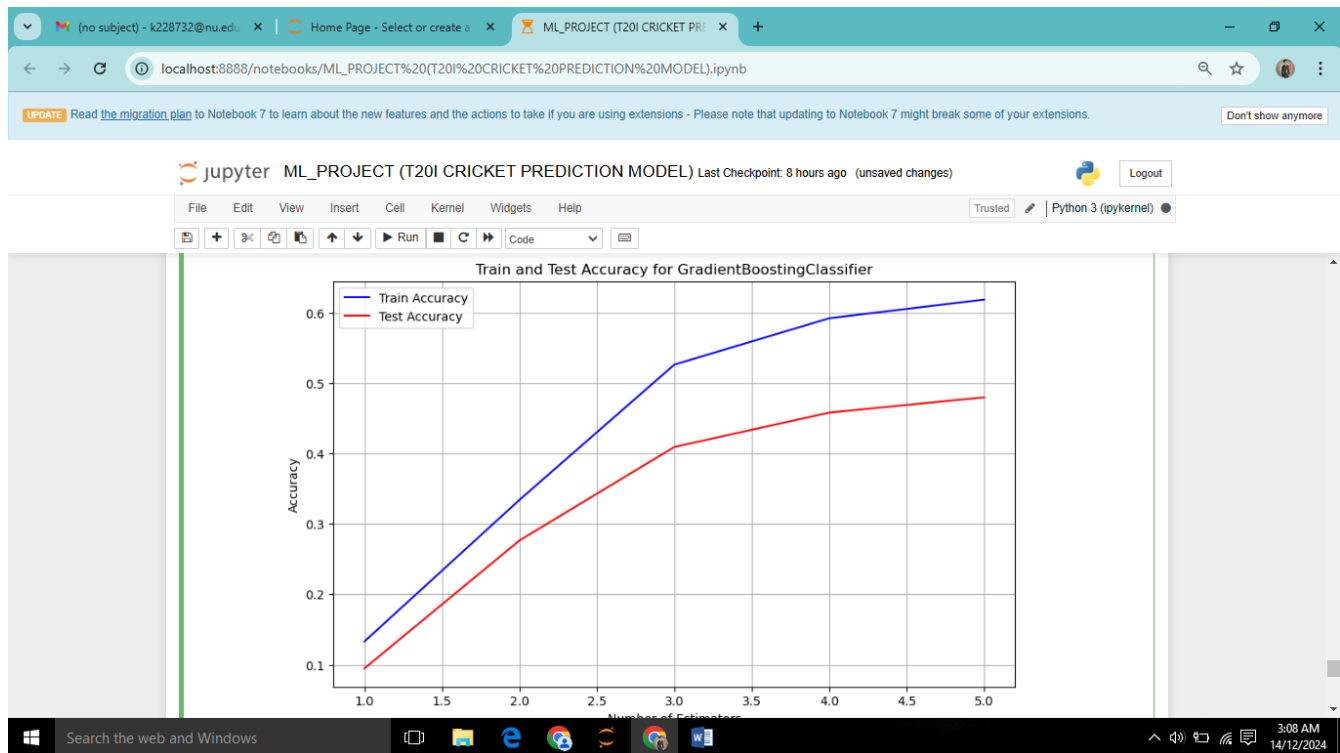
In [74]: model.fit(X_train, y_train)

Out[74]: GradientBoostingClassifier
GradientBoostingClassifier(learning_rate=0.01, random_state=1)

In [75]: accuracy_test = model.score(X_test, y_test)
accuracy_train = model.score(X_train, y_train)

print("\nAccuracy score using XGBoost on test data: ", round(accuracy_test * 100, 2), "%")
print("Accuracy score using XGBoost on training data: ", round(accuracy_train * 100, 2), "%")

Accuracy score using XGBoost on test data: 66.94 %
Accuracy score using XGBoost on training data: 92.53 %
```



4. ADABOOST CLASSIFIER

- Testing Accuracy: 0.0921 or 9.21%
- Reason for Lower Performance:
AdaBoost relies heavily on weak learners and performs well for simpler datasets. The complexity and scale of the dataset limited its effectiveness.

Machine Learning LAB (AL3002) - Project Report

localhost:8888/notebooks/ML_PROJECT%20(T20I%20CRICKET%20PREDICTION%20MODEL).ipynb

UPDATE Read the [migration plan](#) to Notebook 7 to learn about the new features and the actions to take if you are using extensions - Please note that updating to Notebook 7 might break some of your extensions. Don't show anymore

Jupyter ML_PROJECT (T20I CRICKET PREDICTION MODEL) Last Checkpoint: 8 hours ago (unsaved changes) Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3 (ipykernel)

5. Applying AdaBoost

```
In [78]: # AdaBoost
from sklearn.ensemble import AdaBoostClassifier
model = AdaBoostClassifier(n_estimators=n_estimators, algorithm='SAMME', random_state=42)
model.fit(X_train, y_train)

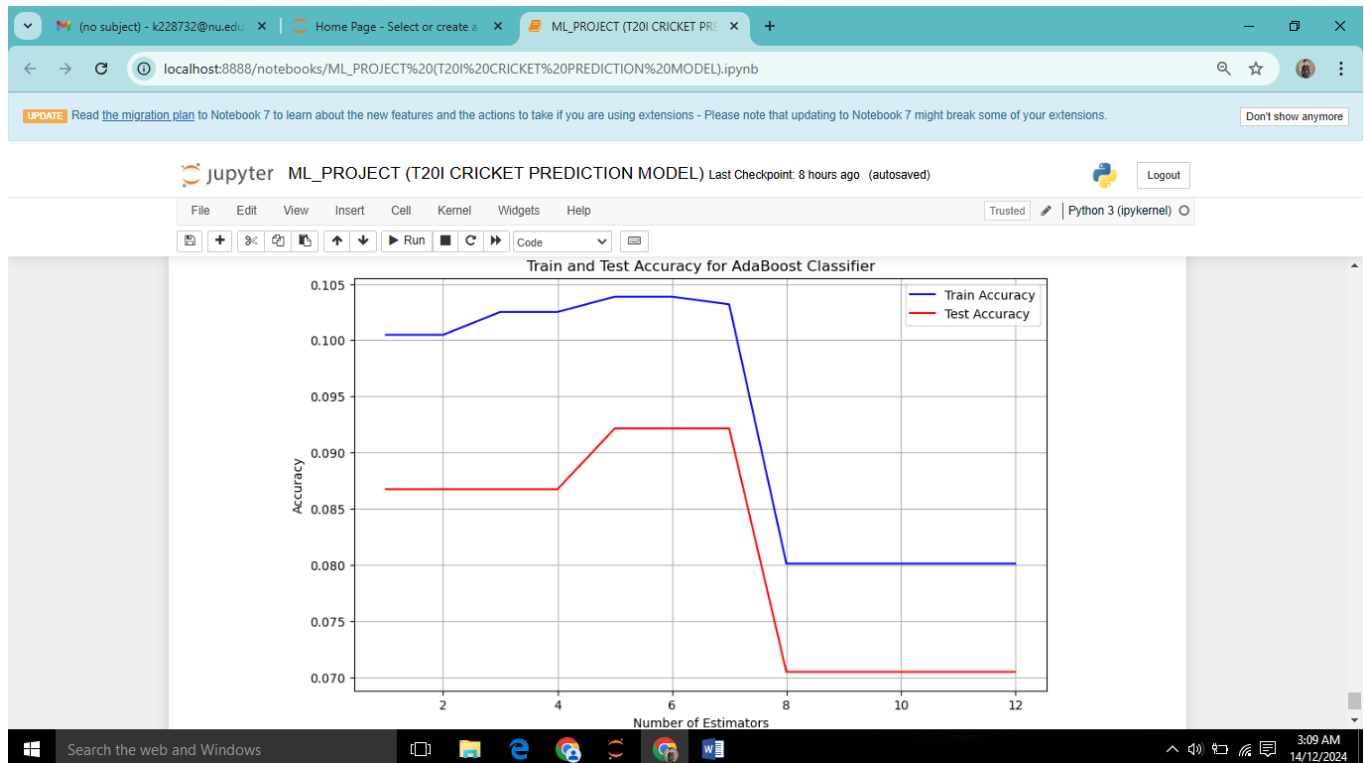
Out[78]:
AdaBoostClassifier
AdaBoostClassifier(algorithm='SAMME', n_estimators=9, random_state=42)

In [79]: accuracy_test = model.score(X_test, y_test)
accuracy_train = model.score(X_train, y_train)

print("\nAccuracy score using AdaBoost on test data: ", round(accuracy_test * 100, 2), "%")
print("Accuracy score using AdaBoost on training data: ", round(accuracy_train * 100, 2), "%")

Accuracy score using AdaBoost on test data: 9.21 %
Accuracy score using AdaBoost on training data: 10.39 %

In [80]: import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
```

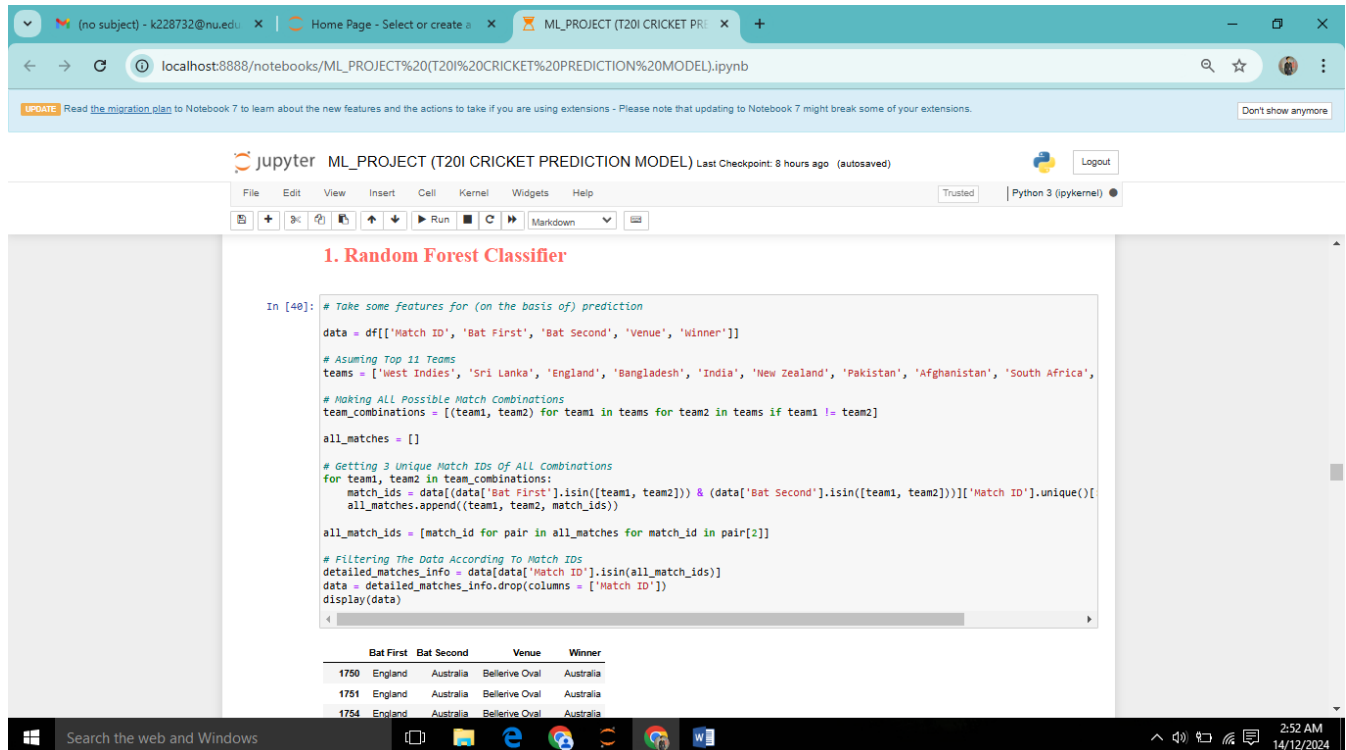


5. RANDOM FOREST CLASSIFIER

➤ Testing Accuracy: 0.95 or 95%

➤ Why It Succeeded

Random Forest effectively handles categorical variables, captures non-linear relationships, and prevents overfitting with its ensemble approach.



```
In [40]: # Take some features for (on the basis of) prediction
data = df[['Match ID', 'Bat First', 'Bat Second', 'Venue', 'Winner']]

# Assuming Top 11 Teams
teams = ['West Indies', 'Sri Lanka', 'England', 'Bangladesh', 'India', 'New Zealand', 'Pakistan', 'Afghanistan', 'South Africa',
         'Australia', 'Zimbabwe']

# Making All Possible Match Combinations
team_combinations = [(team1, team2) for team1 in teams for team2 in teams if team1 != team2]

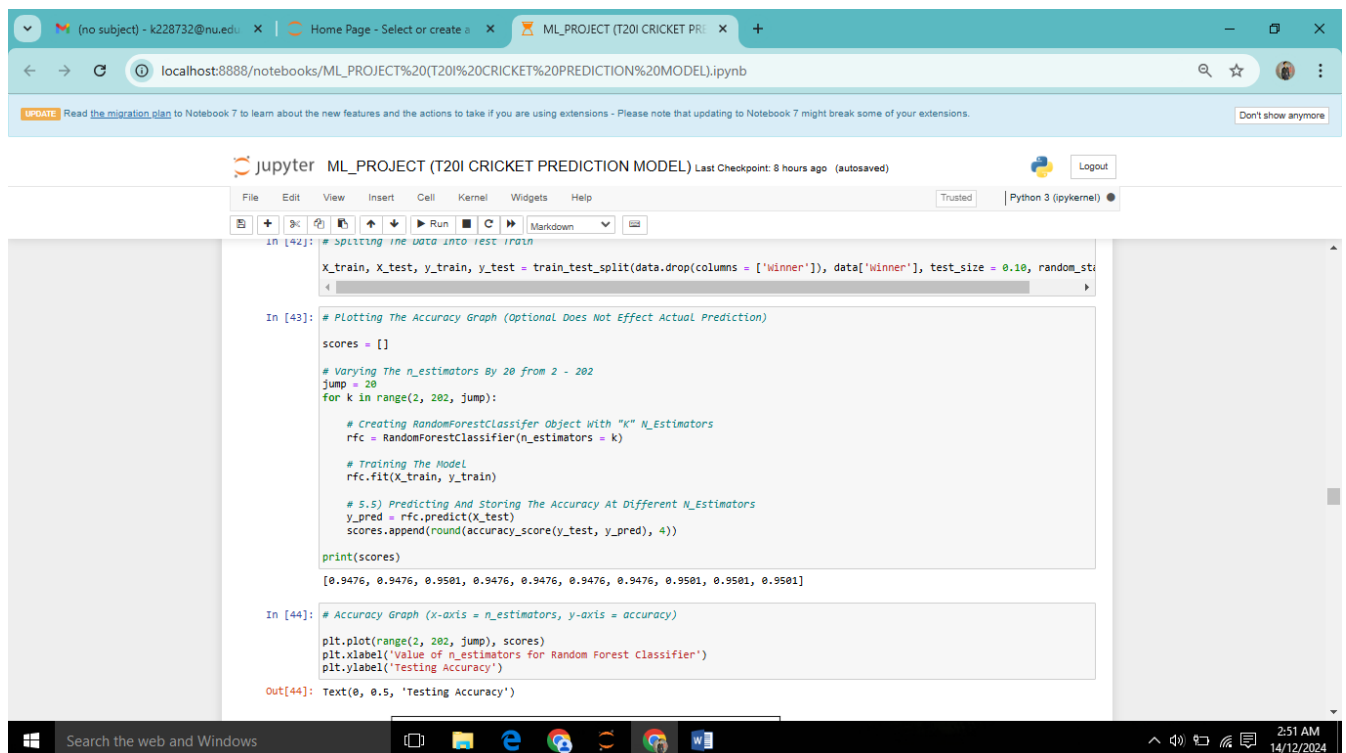
all_matches = []

# Getting 3 Unique Match IDs Of All Combinations
for team1, team2 in team_combinations:
    match_ids = data[(data['Bat First'].isin([team1, team2]) & (data['Bat Second'].isin([team1, team2])))]['Match ID'].unique()
    all_matches.append((team1, team2, match_ids))

all_match_ids = [match_id for pair in all_matches for match_id in pair[2]]

# Filtering The Data According To Match IDs
detailed_matches_info = data[data['Match ID'].isin(all_match_ids)]
data = detailed_matches_info.drop(columns = ['Match ID'])
display(data)
```

	Bat First	Bat Second	Venue	Winner
1750	England	Australia	Bellerive Oval	Australia
1751	England	Australia	Bellerive Oval	Australia
1754	England	Australia	Bellerive Oval	Australia



```
In [42]: # Splitting the data into test/train
X_train, X_test, y_train, y_test = train_test_split(data.drop(columns = ['winner']), data['winner'], test_size = 0.10, random_state = 42)

In [43]: # Plotting The Accuracy Graph (Optional Does Not Effect Actual Prediction)
scores = []

# Varying The n_estimators By 20 from 2 - 202
jump = 20
for k in range(2, 202, jump):

    # Creating RandomForestClassifier Object With "K" N_Estimators
    rfc = RandomForestClassifier(n_estimators = k)

    # Training The Model
    rfc.fit(X_train, y_train)

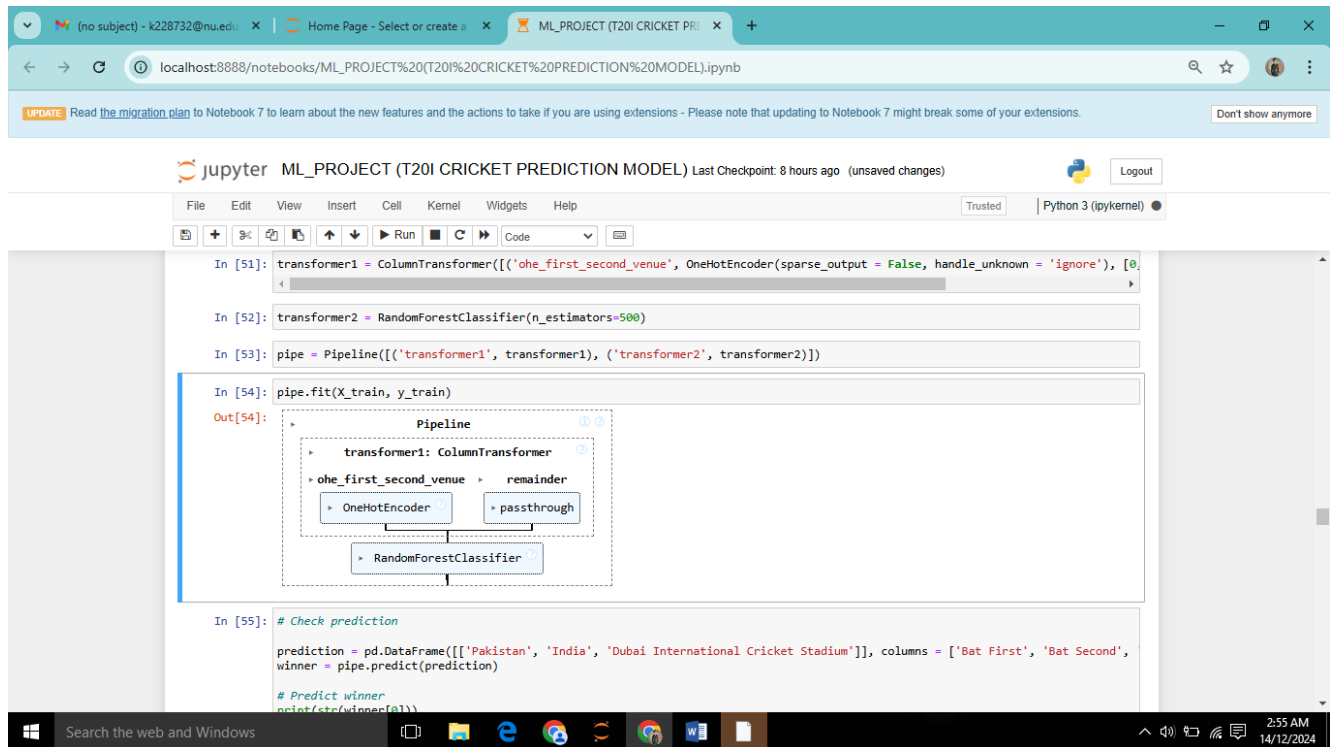
    # 5.5) Predicting And Storing The Accuracy At Different N_Estimators
    y_pred = rfc.predict(X_test)
    scores.append(round(accuracy_score(y_test, y_pred), 4))

print(scores)

[0.9476, 0.9476, 0.9501, 0.9476, 0.9476, 0.9476, 0.9476, 0.9501, 0.9501, 0.9501]

In [44]: # Accuracy Graph (x-axis = n_estimators, y-axis = accuracy)
plt.plot(range(2, 202, jump), scores)
plt.xlabel('Value of n_estimators for Random Forest Classifier')
plt.ylabel('Testing Accuracy')

Out[44]: Text(0, 0.5, 'Testing Accuracy')
```



The screenshot displays a Jupyter Notebook titled "ML_PROJECT (T20I CRICKET PREDICTION MODEL)". The notebook contains the following code cells:

```
In [51]: transformer1 = ColumnTransformer([('ohe_first_second_venue', OneHotEncoder(sparse_output = False, handle_unknown = 'ignore')), [0, 1]], remainder='passthrough')

In [52]: transformer2 = RandomForestClassifier(n_estimators=500)

In [53]: pipe = Pipeline([('transformer1', transformer1), ('transformer2', transformer2)])

In [54]: pipe.fit(X_train, y_train)

Out[54]: Pipeline
  transformer1: ColumnTransformer
    ohe_first_second_venue: OneHotEncoder
    remainder: passthrough
  transformer2: RandomForestClassifier

In [55]: # Check prediction
prediction = pd.DataFrame([['Pakistan', 'India', 'Dubai International Cricket Stadium']], columns = ['Bat First', 'Bat Second', 'Venue'])
winner = pipe.predict(prediction)

# Predict winner
print(winner[0])
```

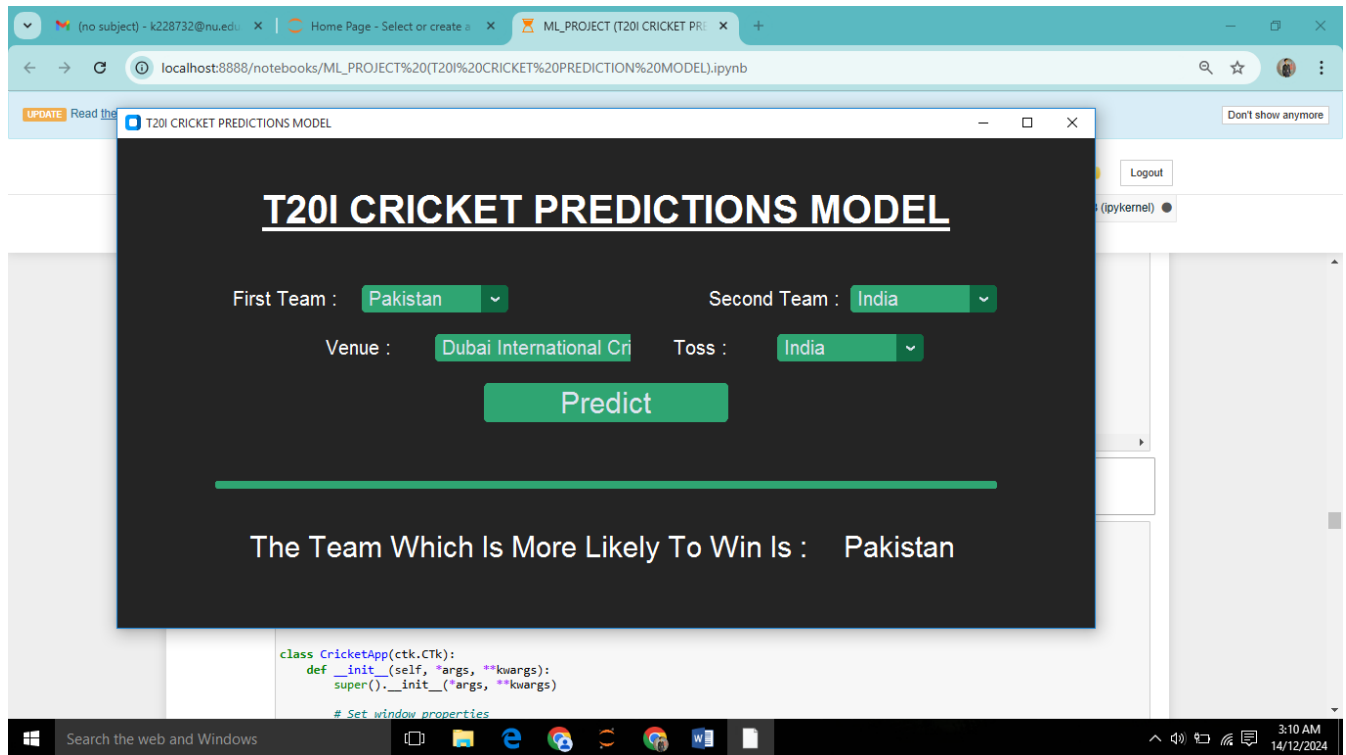
The output of cell [54] shows a Pipeline object with two transformers: a ColumnTransformer and a RandomForestClassifier. The ColumnTransformer has two outputs: 'ohe_first_second_venue' (OneHotEncoder) and 'remainder' (passthrough). The RandomForestClassifier has 500 estimators.

CREATING GUI FOR PREDICTION AND STATISTICS OF MATCHES

1. MATCH OUTCOME PREDICTION

- Functionality: Users can select:
 - Team 1
 - Team 2
 - Venue
 - Toss Prediction (the team bat first)

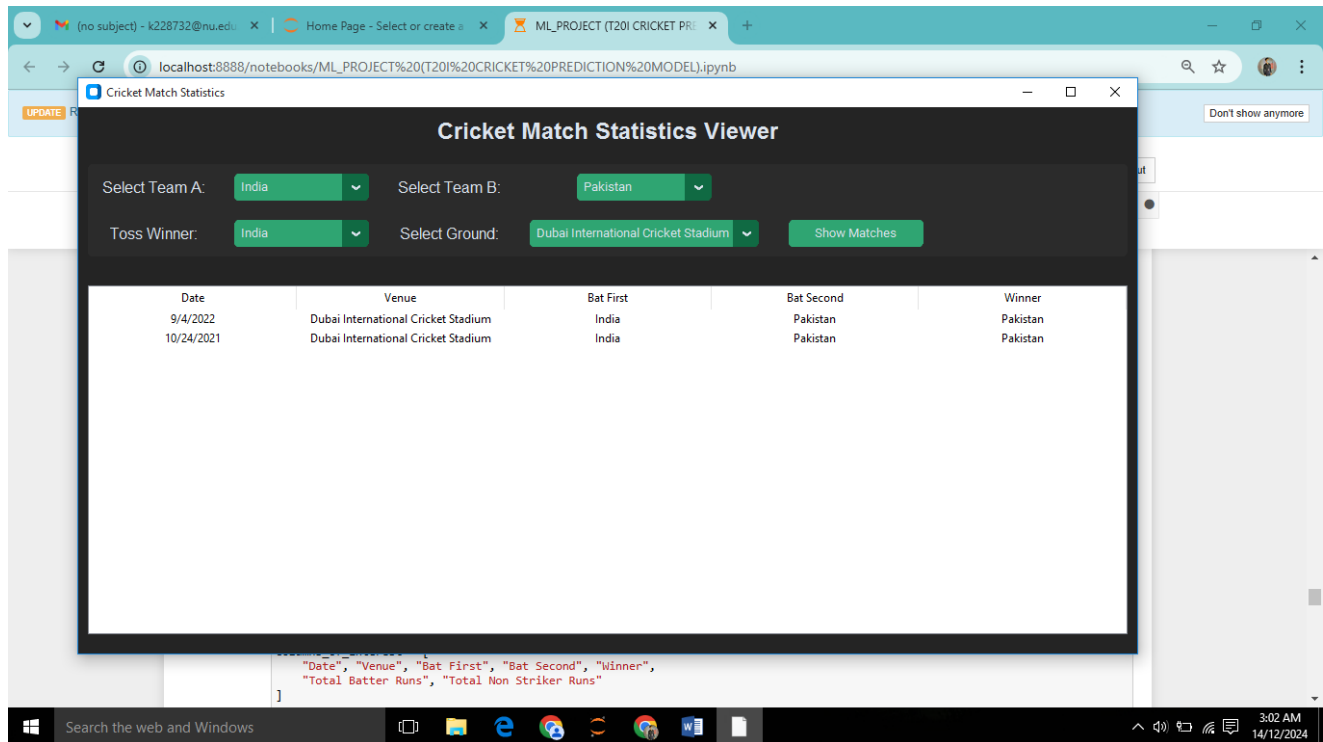
Result: The model predicts the winning team based on past match winning statistics on this specific venue.



2. CHECKING OUR PREDICTION BY PAST MATCHES STATISTICS OF THIS SPECIFIC SELECTED FIELDS (GUI)

Feature: Displays all historical match data for the selected teams.

Purpose: Users can verify the accuracy of model predictions by comparing them with actual outcomes.



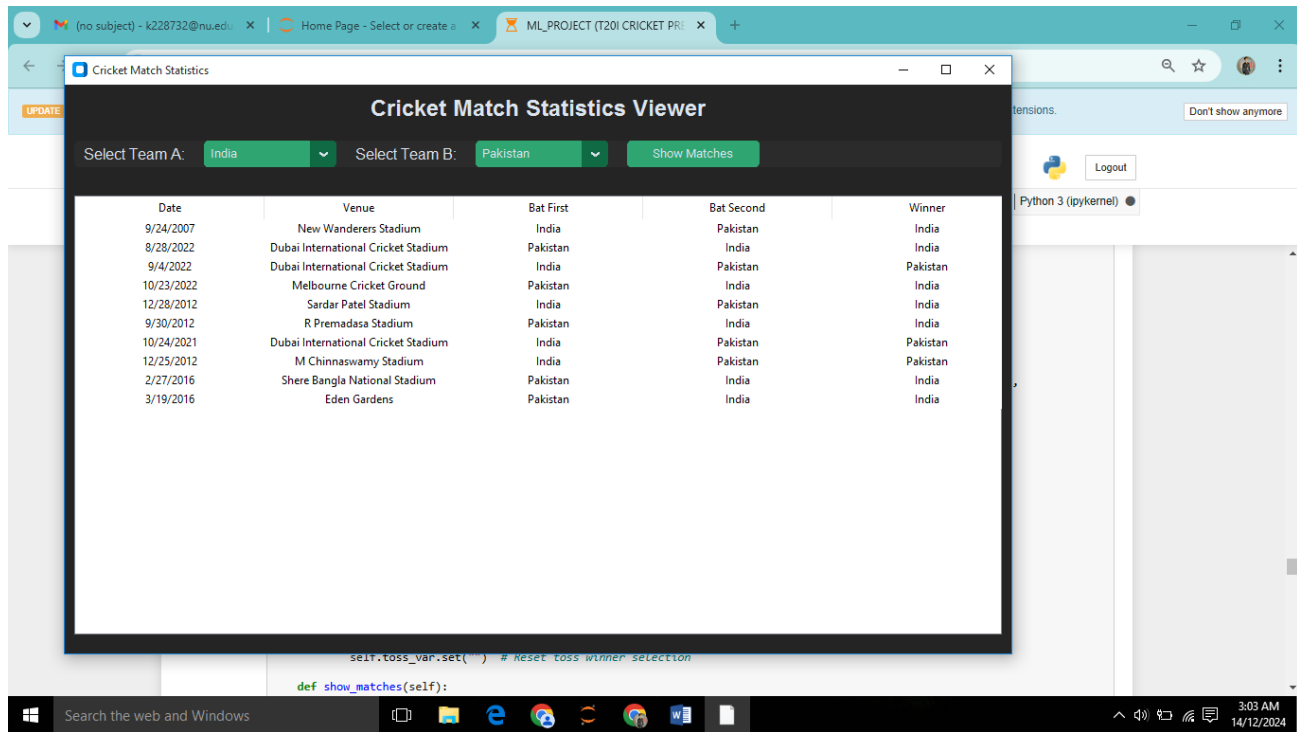
Testing Prediction:

- ✓ 2 matches played in the past between India and Pakistan at Dubai International Stadium where India was bat first in both games and Pakistan were won the both matches. Now when I am predicting for future match with same venue and other statistics. Our model is predicting Winner as "Pakistan".

3. GROUND SPECIFIC STATISTICS (GUI)

Feature: Users select Team 1, Team 2, and a specific venue.

Output: Displays past match statistics specific to the selected ground to help users understand performance trends.



The screenshot shows a web browser window with a tab titled 'ML_PROJECT (T20I CRICKET PR...'. The browser address bar shows 'no subject - k228732@nu.edu...'. The web application is titled 'Cricket Match Statistics Viewer'. It has two dropdown menus for 'Select Team A:' (set to 'India') and 'Select Team B:' (set to 'Pakistan'), and a 'Show Matches' button. Below these is a table of cricket matches. The table has columns: Date, Venue, Bat First, Bat Second, and Winner. The matches listed are between India and Pakistan from 2007 to 2016. The table is displayed within a Jupyter Notebook interface, with a 'Python 3 (ipykernel)' kernel selected. The Windows taskbar at the bottom shows the date as 14/12/2024 and time as 3:03 AM.

Date	Venue	Bat First	Bat Second	Winner
9/24/2007	New Wanderers Stadium	India	Pakistan	India
8/28/2022	Dubai International Cricket Stadium	Pakistan	India	India
9/4/2022	Dubai International Cricket Stadium	India	Pakistan	Pakistan
10/23/2022	Melbourne Cricket Ground	Pakistan	India	India
12/28/2012	Sardar Patel Stadium	India	Pakistan	India
9/30/2012	R Premadasa Stadium	Pakistan	India	India
10/24/2021	Dubai International Cricket Stadium	India	Pakistan	Pakistan
12/25/2012	M Chinnaswamy Stadium	India	Pakistan	Pakistan
2/27/2016	Shere Bangla National Stadium	Pakistan	India	India
3/19/2016	Eden Gardens	Pakistan	India	India

MODEL DEVELOPMENT AND ANALYSIS

1. DATASET PREPROCESSING (PERFORMED EDA)

- Aggregated the dataset by date to ensure each match is represented as a single entry.
- Eliminated duplicate rows and retained only relevant features.

2. FEATURE ENGINEERING AND SCALING

- Encoded categorical variables such as teams and venues using one-hot encoding and label encoding.p
- Implement Standard Scaler to make sure the data is in small range.

3. SPLITTING IN TRAINING AND TESTING DATA:

- Split the dataset into training (80%) and testing (20%) sets.

RESULTS AND CONCLUSION

- The [Random Forest Classifier](#) emerged as the best model, achieving a 95% testing accuracy.
- GUI enhancements provide a user-friendly interface for exploring predictions and historical data.