



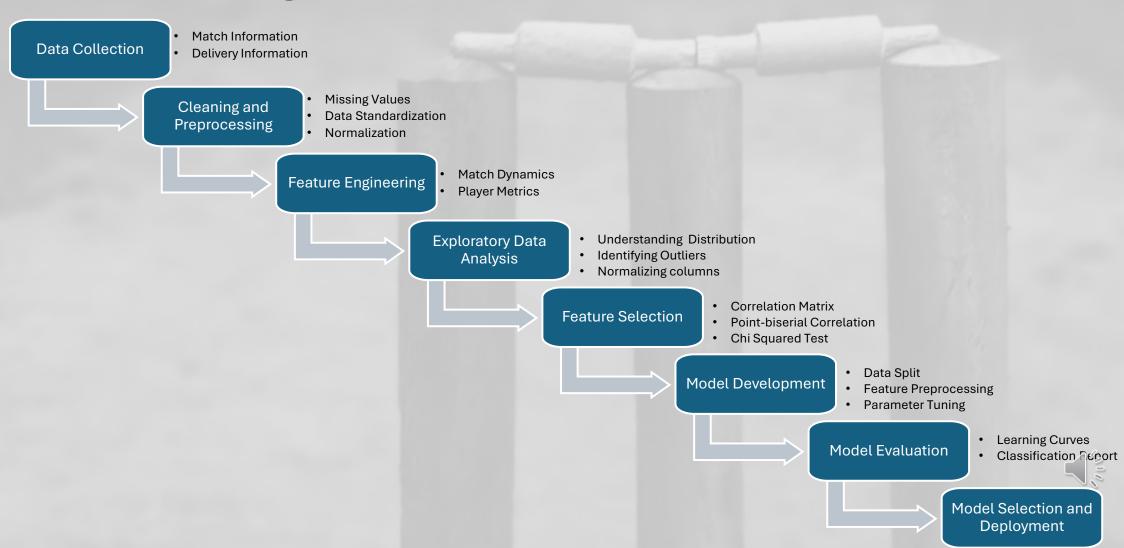


Objectives

- Predicting winner of an IPL match at a given point of time in real-time using machine learning based on historical data.
- To analyse how much impact various factors have on the match outcomes.



Methodology



- Match Information
- Delivery Information

Cleaning and Preprocessing

- Missing Values
- Data Standardization
- Normalization

Feature Engineering

- Match Dynamics
- Player Metrics

Automated Data Extraction

- Toss Information, Location, Team and Player Information
- Ball-By-Ball Information of runs, dismissals, deliveries

Exploratory Data Analysis

- Understanding Distribution
- Identifying Outliers
- Normalizing columns

eature Selection

- Correlation Matrix
- Point-biserial Correlation
- Chi Squared Test

Model Development

- Data Split
- Feature Preprocessing
- Parameter Tuning

Model Evaluatior

- Learning Curves
- Classification Cort

Model Selection and Deployment

Cleaning and Preprocessing

- Missing Values
- Data Standardization
- Normalization

Match Dynamics

Player Metrics

Preprocessing

- Handled missing values and inconsistencies in data
- Filtering data of matches that had no result or impacted by the weather
- Normalising team names and excluding the inactive teams

Exploratory Data Analysis

- Understanding Distribution
- Identifying Outliers
- Normalizing columns

eature Selection

- Correlation Matrix
- Point-biserial Correlation
- Chi Squared Test

Model Development

- Data Split
- Feature Preprocessing
- Parameter Tuning

Model Evaluatior

- Learning Curves
- Classification Cort

Model Selection and Deployment

- Match Information
- Delivery Information

Cleaning and Preprocessing

Feature Engineering

- Match Dynamics
- Player Metrics

Exploratory Data

Feature Engineering

- Home Ground Advantage to Team Mapping & Toss Impact
- Match Characteristics such as runs, balls, run rates, wickets, legal deliveries, etc.
- Strike Rate, Economy, and Batting and Bowling Averages
- Momentum Features like runs or wickets in Last-N-Balls
- Understanding Distribution
- Identifying Outliers
- Normalizing columns

eature Selection

- Correlation Matrix
- Point-biserial Correlation
- Chi Squared Test

Model Development

- Data Split
- Feature Preprocessing
- Parameter Tuning

1odel Evaluation

- Learning Curves
- Classification Cort

Model Selection and Deployment

 Match Information Delivery Information Data Standardization current run rate Understanding Distribution Identifying Outliers Normalizing columns target

75 100 125 150 175 200 225 250

EDA

- Excluded matches with rare Targets
- Normalizing Team, Venue, City and Player Names
- Capped Values for Outliers

Exploratory Data
Analysis

Correlation Matrix
Point-biserial Correlation
Chi Squared Test

Model Development

Data Split
Feature Preprocessing
Parameter Tuning

Model Evaluation

Learning Curves
Classification Foot

Match Environment Features such as Venue, Batting Team, Toss Winner Match Information Delivery Information Match characteristics and Momentum **Features** Data Standardization Batsman's match and all-time, and bowler's match statistics Match Dynamics Player Metrics Шининин Feature Selection Categorical Features vs. Win/Loss (Chi-squared Correlation Matrix Data Split Point-biserial Correlation Feature Preprocessing Parameter Tuning Chi Squared Test Learning Curves Classification Cont

Feature Selection

- Match Information
- Delivery Information

Cleaning and Preprocessing

- Missing Values
- Data Standardization
- Normalization

Feature Engineering

- Match Dynamics
- Player Metrics

Data Modelling

- Used 'TimeSeriesSplit' for splitting the dataset for training and testing
- Feature Encoding and Scaling
- Model Pipeline and parameter grid construction

Exploratory Data Analysis

- Understanding Distribution
- Identifying Outliers
- Normalizing columns

Feature Selectic

Model Development

- Data Split
- Feature Preprocessing
- Parameter Tuning

Model Evaluation

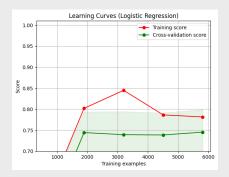
- Match Information
- Delivery Information

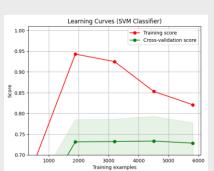
Cleaning and Preprocessing

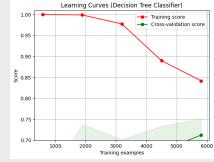
- Missing Values
- Data Standardization
- Normalization

Feature Engineering

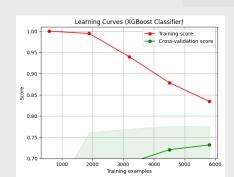
- Match Dynamics
- Player Metrics











- Exploratory Data
 Analysis
- Understanding Distribution
- Identifying Outliers
- Normalizing columns

Feature Selection

- Correlation Matrix
- Point-biserial Correlation

Evaluation Techniques

data and spot overfitting

evaluation of all accuracies

Plotted Learning Curves to understand

Classification Report to have thorough

how well models learn with the amount of

Chi Squared Test

1odel Developme

Model Evaluation

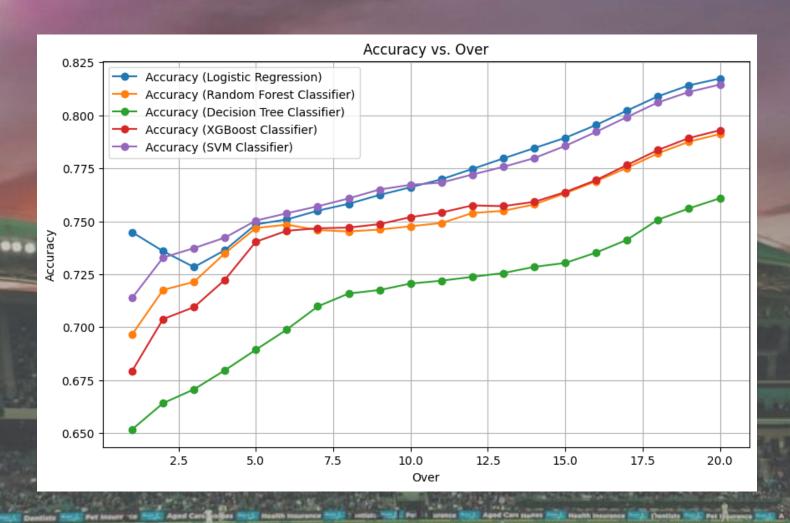
- Learning Curves
- Classification Report

Mode D Evaluation of the state of the

Microscopic view to accuracy and f1 scores during the match Match Information Delivery Information Streamlined the Model using its pickle to create a user-friendly interface for real- Data Standardization time inputs for predictions Match Dynamics Player Metrics Accuracy vs. Over Accuracy (Logistic Regression) Understanding Distribution Accuracy (Random Forest Classifier Accuracy (XGBoost Classifier) Identifying Outliers Normalizing columns Correlation Matrix Point-biserial Correlation 0.700 -F1 Score vs. Over · Chi Squared Test 0.675 Data Split Feature Preprocessing Parameter Tuning Match Progression an Evaluation Streamlit for webapplication

Model Selection & Deployment

Results



- We can see the importance of factors such as venue, toss, etc.
- Provided cricket fans with insights
- Successfully demonstrated how machine learning can be used to predict match outcomes

Future Scope

- Web-scrapping or use of APIs to collect live information and integrating other datasets
- Analysing over various types of dataset, such as image, player performance, etc.
- Predicting various different aspects such as injuries, score, player performance, etc.



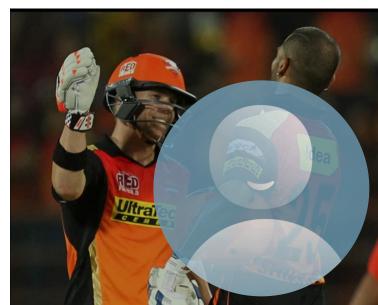
Moments











Special Thanks To

Dr. Chengjia Wang for supervising this project

