

ipl-score

August 1, 2024

```
[57]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from seaborn import heatmap
```

```
[58]: df=pd.read_csv('ipl_data.csv')
```

```
[59]: df.head(10)
```

```
[59]:   mid      date      venue      bat_team \
0    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
1    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
2    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
3    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
4    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
5    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
6    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
7    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
8    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
9    1  2008-04-18  M Chinnaswamy Stadium  Kolkata Knight Riders
```

			bowl_team	batsman	bowler	runs	wickets	overs	\
0	Royal Challengers Bangalore	SC Ganguly	P Kumar	1	0	0.1			
1	Royal Challengers Bangalore	BB McCullum	P Kumar	1	0	0.2			
2	Royal Challengers Bangalore	BB McCullum	P Kumar	2	0	0.2			
3	Royal Challengers Bangalore	BB McCullum	P Kumar	2	0	0.3			
4	Royal Challengers Bangalore	BB McCullum	P Kumar	2	0	0.4			
5	Royal Challengers Bangalore	BB McCullum	P Kumar	2	0	0.5			
6	Royal Challengers Bangalore	BB McCullum	P Kumar	3	0	0.6			
7	Royal Challengers Bangalore	BB McCullum	Z Khan	3	0	1.1			
8	Royal Challengers Bangalore	BB McCullum	Z Khan	7	0	1.2			
9	Royal Challengers Bangalore	BB McCullum	Z Khan	11	0	1.3			

	runs_last_5	wickets_last_5	striker	non-striker	total
0	1	0	0	0	222
1	1	0	0	0	222

2	2	0	0	0	222
3	2	0	0	0	222
4	2	0	0	0	222
5	2	0	0	0	222
6	3	0	0	0	222
7	3	0	0	0	222
8	7	0	4	0	222
9	11	0	8	0	222

```
[60]: df['bat_team'].unique()
```

```
[60]: array(['Kolkata Knight Riders', 'Chennai Super Kings', 'Rajasthan Royals',
        'Mumbai Indians', 'Deccan Chargers', 'Kings XI Punjab',
        'Royal Challengers Bangalore', 'Delhi Daredevils',
        'Kochi Tuskers Kerala', 'Pune Warriors', 'Sunrisers Hyderabad',
        'Rising Pune Supergiants', 'Gujarat Lions',
        'Rising Pune Supergiant'], dtype=object)
```

```
[61]: df.isnull().sum()
```

```
[61]: mid                0
      date                0
      venue              0
      bat_team           0
      bowl_team          0
      batsman            0
      bowler             0
      runs               0
      wickets            0
      overs              0
      runs_last_5        0
      wickets_last_5     0
      striker            0
      non-striker        0
      total              0
      dtype: int64
```

```
[62]: df.duplicated().sum()
```

```
[62]: 0
```

```
[63]: df.describe()
```

```
[63]:
```

	mid	runs	wickets	overs	runs_last_5 \
count	76014.000000	76014.000000	76014.000000	76014.000000	76014.000000
mean	308.627740	74.889349	2.415844	9.783068	33.216434
std	178.156878	48.823327	2.015207	5.772587	14.914174

min	1.000000	0.000000	0.000000	0.000000	0.000000
25%	154.000000	34.000000	1.000000	4.600000	24.000000
50%	308.000000	70.000000	2.000000	9.600000	34.000000
75%	463.000000	111.000000	4.000000	14.600000	43.000000
max	617.000000	263.000000	10.000000	19.600000	113.000000

	wickets_last_5	striker	non-striker	total
count	76014.000000	76014.000000	76014.000000	76014.000000
mean	1.120307	24.962283	8.869287	160.901452
std	1.053343	20.079752	10.795742	29.246231
min	0.000000	0.000000	0.000000	67.000000
25%	0.000000	10.000000	1.000000	142.000000
50%	1.000000	20.000000	5.000000	162.000000
75%	2.000000	35.000000	13.000000	181.000000
max	7.000000	175.000000	109.000000	263.000000

```
[64]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 76014 entries, 0 to 76013
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   mid                    76014 non-null  int64
1   date                   76014 non-null  object
2   venue                  76014 non-null  object
3   bat_team               76014 non-null  object
4   bowl_team              76014 non-null  object
5   batsman                76014 non-null  object
6   bowler                 76014 non-null  object
7   runs                   76014 non-null  int64
8   wickets                76014 non-null  int64
9   overs                  76014 non-null  float64
10  runs_last_5            76014 non-null  int64
11  wickets_last_5         76014 non-null  int64
12  striker                76014 non-null  int64
13  non-striker            76014 non-null  int64
14  total                  76014 non-null  int64
dtypes: float64(1), int64(8), object(6)
memory usage: 8.7+ MB
```

```
[65]: df.shape
```

```
[65]: (76014, 15)
```

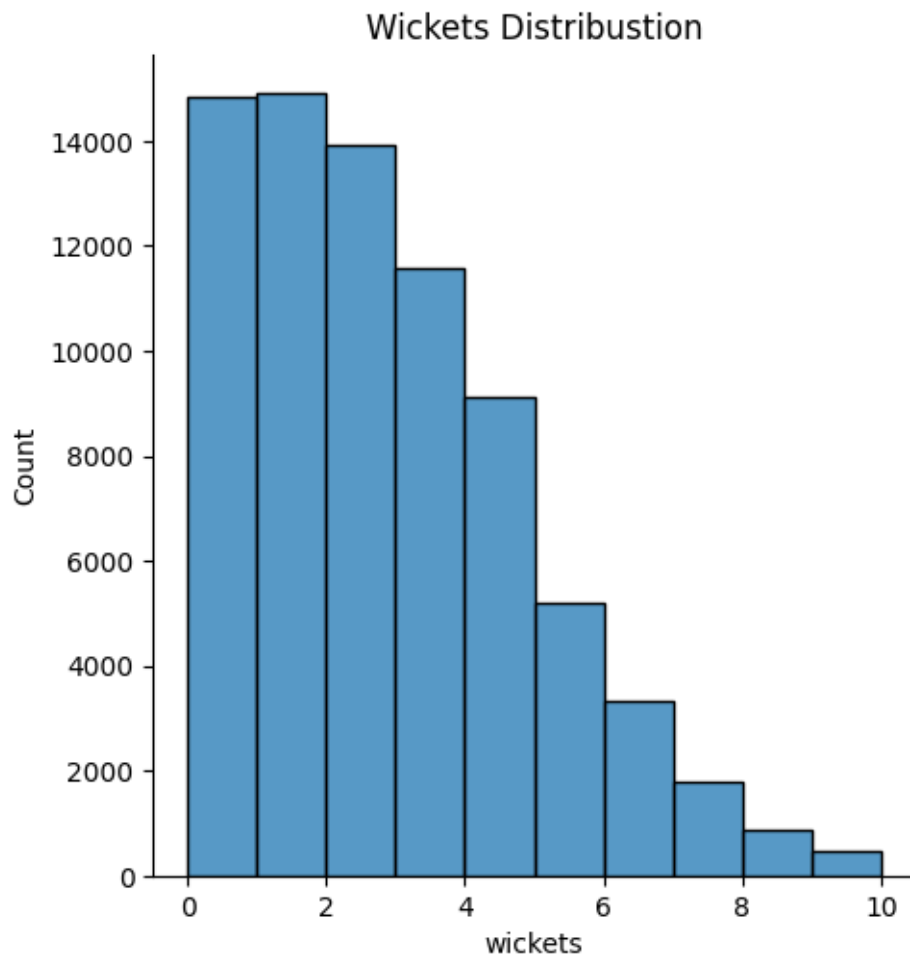
```
[66]: df.nunique()
```

```
[66]: mid          617
      date         442
      venue         35
      bat_team      14
      bowl_team     14
      batsman       411
      bowler        329
      runs          252
      wickets        11
      overs         140
      runs_last_5    102
      wickets_last_5 8
      striker       155
      non-striker     88
      total         138
      dtype: int64
```

wicket distribustribustion countd of run per wickets

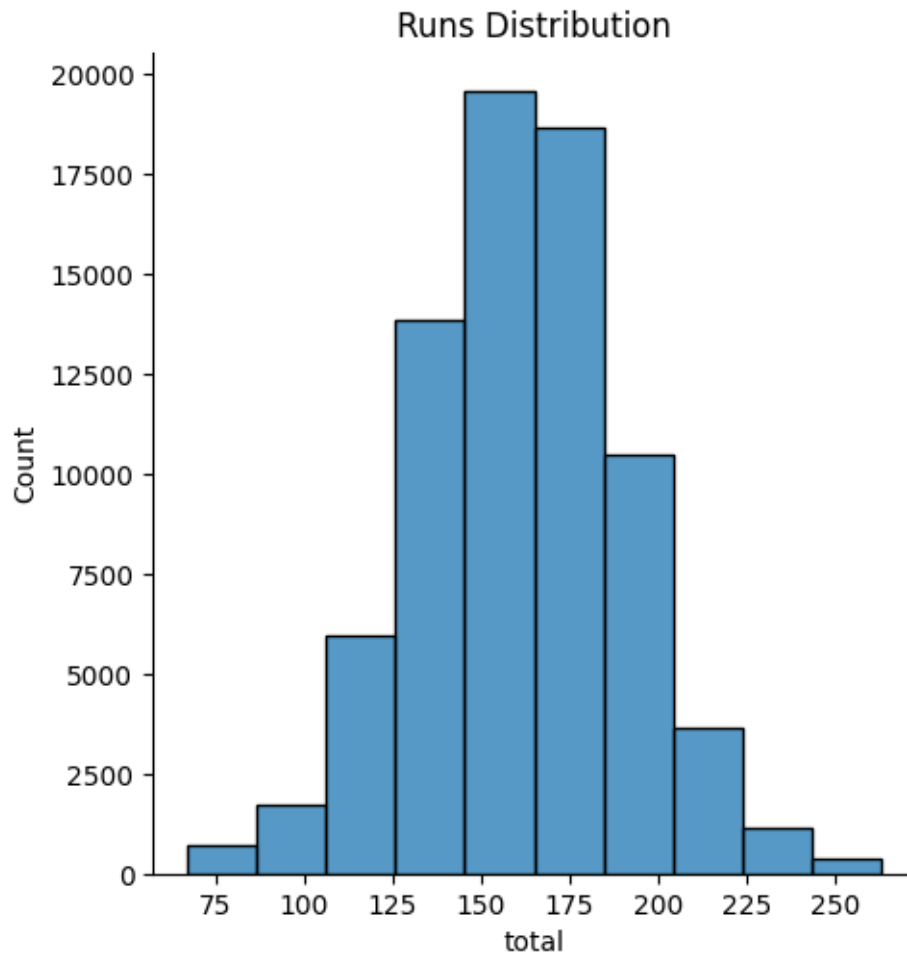
```
[67]: sns.displot(df['wickets'],kde=False,bins=10)
      plt.title('Wickets Distribustion')
```

```
[67]: Text(0.5, 1.0, 'Wickets Distribustion')
```



```
[68]: sns.displot(df['total'],kde=False,bins=10)  
      plt.title('Runs Distribution')
```

```
[68]: Text(0.5, 1.0, 'Runs Distribution')
```



```
[69]: df.columns
```

```
[69]: Index(['mid', 'date', 'venue', 'bat_team', 'bowl_team', 'batsman', 'bowler',  
        'runs', 'wickets', 'overs', 'runs_last_5', 'wickets_last_5', 'striker',  
        'non-striker', 'total'],  
        dtype='object')
```

```
[70]: irrelevant=['mid', 'date', 'venue', 'batsman', 'bowler', 'striker',  
        'non-striker']  
print("Before removing col", df.shape)  
  
df=df.drop(irrelevant,axis=1)  
print("After removing col", df.shape)
```

Before removing col (76014, 15)

After removing col (76014, 8)

```
[71]: df.columns
```

```
[71]: Index(['bat_team', 'bowl_team', 'runs', 'wickets', 'overs', 'runs_last_5',  
        'wickets_last_5', 'total'],  
        dtype='object')
```

```
[ ]:
```

```
[72]: # keeping only consistent team
```

```
const_teams=['Kolkata Knight Riders', 'Chennai Super Kings', 'Rajasthan Royals',  
             'Mumbai Indians', 'Kings XI Punjab',  
             'Royal Challengers Bangalore', 'Delhi Daredevils', 'Sunrisers Hyderabad']
```

```
[73]: df.head(20)
```

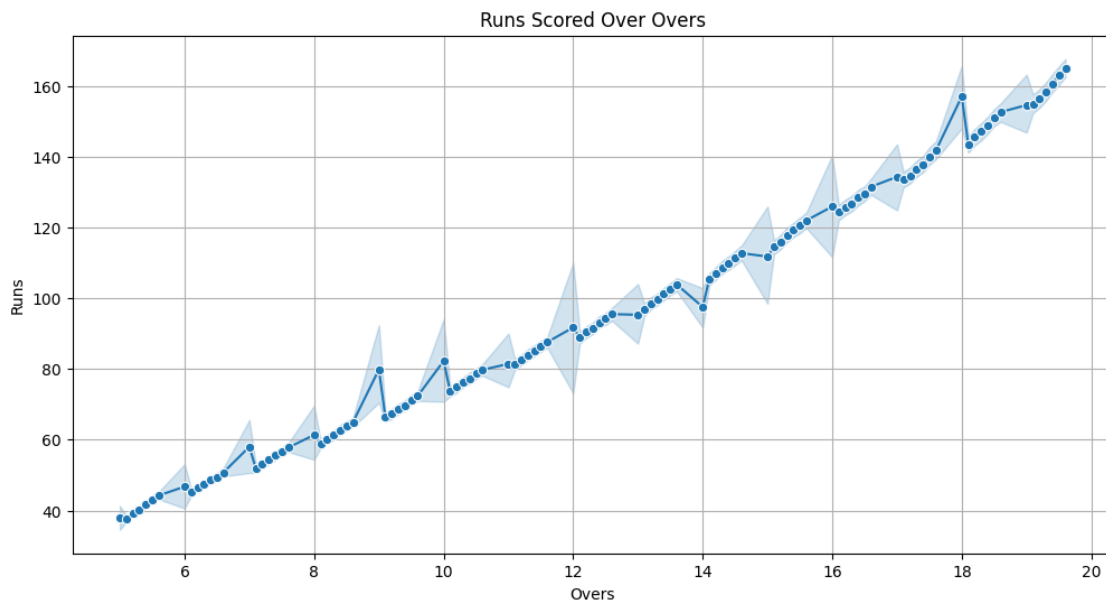
```
[73]:
```

	bat_team	bowl_team	runs	wickets	overs	\
0	Kolkata Knight Riders	Royal Challengers Bangalore	1	0	0.1	
1	Kolkata Knight Riders	Royal Challengers Bangalore	1	0	0.2	
2	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.2	
3	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.3	
4	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.4	
5	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.5	
6	Kolkata Knight Riders	Royal Challengers Bangalore	3	0	0.6	
7	Kolkata Knight Riders	Royal Challengers Bangalore	3	0	1.1	
8	Kolkata Knight Riders	Royal Challengers Bangalore	7	0	1.2	
9	Kolkata Knight Riders	Royal Challengers Bangalore	11	0	1.3	
10	Kolkata Knight Riders	Royal Challengers Bangalore	17	0	1.4	
11	Kolkata Knight Riders	Royal Challengers Bangalore	21	0	1.5	
12	Kolkata Knight Riders	Royal Challengers Bangalore	21	0	1.6	
13	Kolkata Knight Riders	Royal Challengers Bangalore	21	0	2.1	
14	Kolkata Knight Riders	Royal Challengers Bangalore	21	0	2.2	
15	Kolkata Knight Riders	Royal Challengers Bangalore	22	0	2.3	
16	Kolkata Knight Riders	Royal Challengers Bangalore	26	0	2.4	
17	Kolkata Knight Riders	Royal Challengers Bangalore	27	0	2.5	
18	Kolkata Knight Riders	Royal Challengers Bangalore	27	0	2.6	
19	Kolkata Knight Riders	Royal Challengers Bangalore	32	0	3.0	

	runs_last_5	wickets_last_5	total
0	1	0	222
1	1	0	222
2	2	0	222
3	2	0	222
4	2	0	222
5	2	0	222
6	3	0	222
7	3	0	222

8	7	0	222
9	11	0	222
10	17	0	222
11	21	0	222
12	21	0	222
13	21	0	222
14	21	0	222
15	22	0	222
16	26	0	222
17	27	0	222
18	27	0	222
19	32	0	222

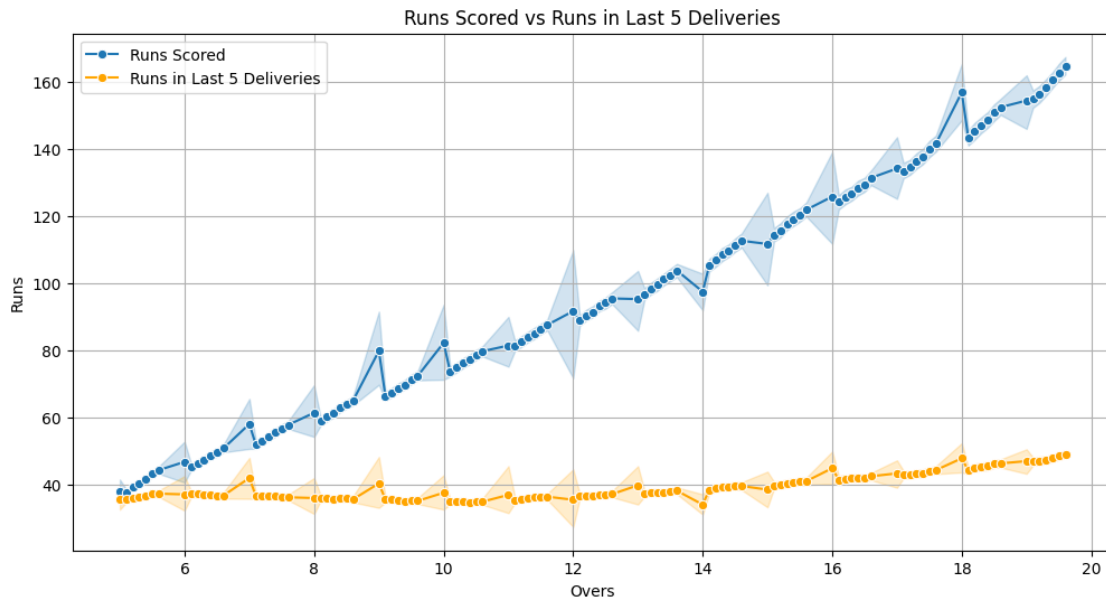
```
[113]: plt.figure(figsize=(12, 6))
sns.lineplot(x='overs', y='runs', data=df, marker='o')
plt.title('Runs Scored Over Overs')
plt.xlabel('Overs')
plt.ylabel('Runs')
plt.grid(True)
plt.show()
```



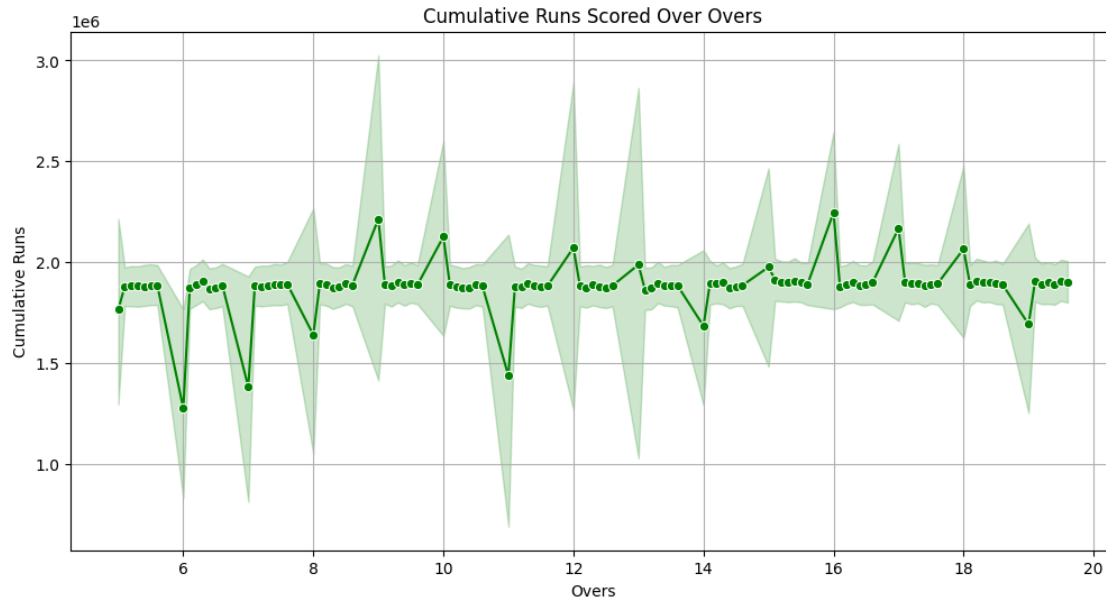
```
[112]: plt.figure(figsize=(12, 6))
sns.lineplot(x='overs', y='runs', data=df, marker='o', label='Runs Scored')
sns.lineplot(x='overs', y='runs_last_5', data=df, marker='o', color='orange',
             label='Runs in Last 5 Deliveries')
plt.title('Runs Scored vs Runs in Last 5 Deliveries')
plt.xlabel('Overs')
```



```
plt.ylabel('Runs')
plt.legend()
plt.grid(True)
plt.show()
```

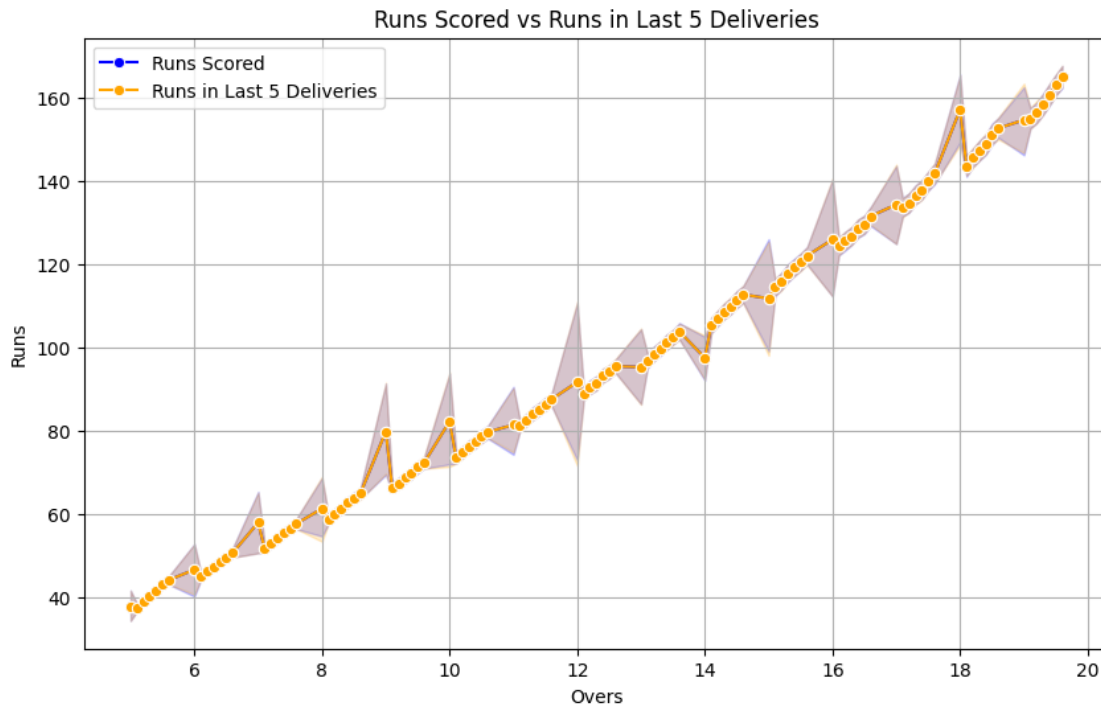


```
[114]: df['cumulative_runs'] = df['runs'].cumsum()
plt.figure(figsize=(12, 6))
sns.lineplot(x='overs', y='cumulative_runs', data=df, marker='o', color='green')
plt.title('Cumulative Runs Scored Over Overs')
plt.xlabel('Overs')
plt.ylabel('Cumulative Runs')
plt.grid(True)
plt.show()
```



```
[115]: # Adding runs_last_5 for demonstration
df['runs_last_5'] = df['runs'] # Assuming same for simplicity

# Plot: Runs Scored vs Runs in Last 5 Deliveries
plt.figure(figsize=(10, 6))
sns.lineplot(x='overs', y='runs', data=df, marker='o', label='Runs Scored',
             color='blue')
sns.lineplot(x='overs', y='runs_last_5', data=df, marker='o', label='Runs in
             Last 5 Deliveries', color='orange')
plt.title('Runs Scored vs Runs in Last 5 Deliveries')
plt.xlabel('Overs')
plt.ylabel('Runs')
plt.legend()
plt.grid(True)
plt.show()
```



```
[74]: print("Before removing inconsistent team : ",df.shape)

df=df[(df['bat_team'].isin(const_teams)) & (df['bowl_team'].isin(const_teams))]

print("after removing inconsistent team : ",df.shape)
```

Before removing inconsistent team : (76014, 8)
after removing inconsistent team : (53811, 8)

```
[75]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 53811 entries, 0 to 75888
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  -
0   bat_team        53811 non-null  object
1   bowl_team       53811 non-null  object
2   runs            53811 non-null  int64
3   wickets        53811 non-null  int64
4   overs          53811 non-null  float64
5   runs_last_5     53811 non-null  int64
6   wickets_last_5  53811 non-null  int64
7   total          53811 non-null  int64
```

```
dtypes: float64(1), int64(5), object(2)
memory usage: 3.7+ MB
```

```
[76]: df.head()
```

```
[76]:
```

	bat_team	bowl_team	runs	wickets	overs	\
0	Kolkata Knight Riders	Royal Challengers Bangalore	1	0	0.1	
1	Kolkata Knight Riders	Royal Challengers Bangalore	1	0	0.2	
2	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.2	
3	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.3	
4	Kolkata Knight Riders	Royal Challengers Bangalore	2	0	0.4	

	runs_last_5	wickets_last_5	total
0	1	0	222
1	1	0	222
2	2	0	222
3	2	0	222
4	2	0	222

```
[77]: # Remove first 5 over

print('Before removing over',df.shape)

df=df[df['overs']>=5.0]

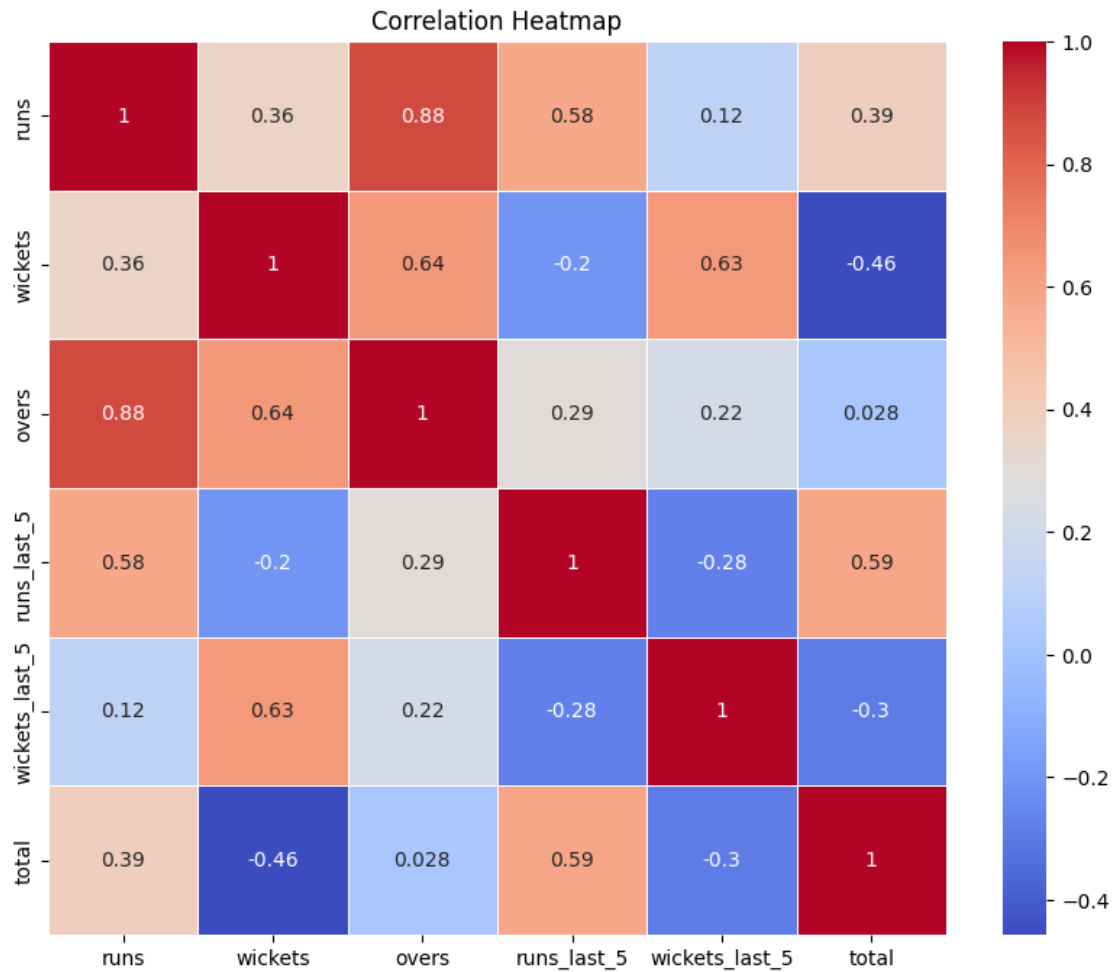
print('After removing over',df.shape)
```

```
Before removing over (53811, 8)
After removing over (40108, 8)
```

```
[78]: num_col = df.select_dtypes(include=['float64', 'int64'])

# Compute correlation matrix
corr_matrix = num_col.corr()

# Plot heatmap
plt.figure(figsize=(10, 8))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', linewidths=0.5)
plt.title('Correlation Heatmap')
plt.show()
```



Perform Label Encoding

```
[79]: from sklearn.preprocessing import LabelEncoder, OneHotEncoder
```

```
[80]: le=LabelEncoder()
```

```
[81]: for col in ['bat_team', 'bowl_team']:
      df[col]=le.fit_transform(df[col])
```

```
[82]: df.head()
```

```
[82]:   bat_team  bowl_team  runs  wickets  overs  runs_last_5  wickets_last_5  \
32      3         6    61      0    5.1         59         0
33      3         6    61      1    5.2         59         1
34      3         6    61      1    5.3         59         1
35      3         6    61      1    5.4         59         1
36      3         6    61      1    5.5         58         1
```

```

        total
32      222
33      222
34      222
35      222
36      222

```

```
[83]: from sklearn.compose import ColumnTransformer
```

```
[84]: columnTransformer = ColumnTransformer([('encoder',
                                             OneHotEncoder(),
                                             [0, 1])],
                                             remainder='passthrough')
```

```
[85]: ipl_df = np.array(columnTransformer.fit_transform(df))
```

```
[86]: cols = ['batting_team_Chennai Super Kings', 'batting_team_Delhi Daredevils',
             ↪ 'batting_team_Kings XI Punjab',
             ↪ 'batting_team_Kolkata Knight Riders', 'batting_team_Mumbai
             ↪ Indians', 'batting_team_Rajasthan Royals',
             ↪ 'batting_team_Royal Challengers Bangalore',
             ↪ 'batting_team_Sunrisers Hyderabad',
             ↪ 'bowling_team_Chennai Super Kings', 'bowling_team_Delhi
             ↪ Daredevils', 'bowling_team_Kings XI Punjab',
             ↪ 'bowling_team_Kolkata Knight Riders', 'bowling_team_Mumbai
             ↪ Indians', 'bowling_team_Rajasthan Royals',
             ↪ 'bowling_team_Royal Challengers Bangalore',
             ↪ 'bowling_team_Sunrisers Hyderabad', 'runs', 'wickets', 'overs',
             ↪ 'runs_last_5', 'wickets_last_5', 'total']
df = pd.DataFrame(ipl_df, columns=cols)
```

```
[87]: print(df.shape)
```

```
(40108, 22)
```

```
[88]: df = pd.DataFrame(df, columns=cols)
```

```
[89]: x=df.drop('total',axis=1)
      y=df['total']
```

```
[90]: from sklearn.model_selection import train_test_split
```

```
[91]: X_TRAIN, X_TEST, Y_TRAIN, Y_TEST=train_test_split(x,y,test_size=0.
             ↪ 20,random_state=42)
```

```
[92]: print(X_TRAIN.shape)
      print(X_TEST.shape)
      print(Y_TRAIN.shape)
      print(Y_TEST.shape)
```

```
(32086, 21)
(8022, 21)
(32086,)
(8022,)
```

```
[93]: from sklearn.metrics import accuracy_score
```

```
[94]: from sklearn.linear_model import LinearRegression
      from sklearn.tree import DecisionTreeRegressor
      from sklearn.ensemble import RandomForestRegressor
      from sklearn.svm import SVR
      from xgboost import XGBRegressor
```

```
[95]: from sklearn.metrics import mean_squared_error as mse, r2_score as r2, \
      ↪mean_absolute_error as mae
```

```
[96]: LR_model=LinearRegression()
      DST_model=DecisionTreeRegressor()
      RF_model=RandomForestRegressor()
      SVR_model=SVR()
      SG_model=XGBRegressor()
```

Linear Regrssion

```
[97]: LR_model.fit(X_TRAIN,Y_TRAIN)
```

```
[97]: LinearRegression()
```

```
[98]: Y_Pred_LR=LR_model.predict(X_TEST)
```

```
[99]: LR_rmse=np.sqrt(mse(Y_TEST,Y_Pred_LR))
      LR_r2=r2(Y_TEST,Y_Pred_LR)
```

```
[100]: y_train_pred=LR_model.predict(X_TRAIN)
       y_test_pred=LR_model.predict(X_TEST)
```

```
[101]: train_r2_percentage = r2(Y_TRAIN, y_train_pred) * 100
       test_r2_percentage = r2(Y_TEST, y_test_pred) * 100
```

```
[102]: print('Training score',train_r2_percentage)
       print('Testing score',test_r2_percentage)
```

Trainning score 65.69923425260747
Testing score 66.78598203684052

```
[103]: print("LinearRegression RMSE:", LR_rmse)
       print("LinearRegression r2:", LR_r2)
```

LinearRegression RMSE: 17.23030069769983
LinearRegression r2: 0.6678598203684052

Decision Tree

```
[104]: DST_model.fit(X_TRAIN,Y_TRAIN)
```

```
[104]: DecisionTreeRegressor()
```

```
[105]: Y_Pred_DST=DST_model.predict(X_TEST)
```

```
[106]: train_score_tree = str(DST_model.score(X_TRAIN, Y_TRAIN) * 100)
       test_score_tree = str(DST_model.score(X_TEST, Y_TEST) * 100)
       print(f'Train Score : {train_score_tree[:5]}%\nTest Score : {test_score_tree[:5]}%')
```

Train Score : 99.98%
Test Score : 85.52%

Random Forest

```
[107]: RF_model.fit(X_TRAIN,Y_TRAIN)
```

```
[107]: RandomForestRegressor()
```

```
[108]: Y_Pred_RF=RF_model.predict(X_TEST)
```

```
[109]: train_score_RF = str(RF_model.score(X_TRAIN, Y_TRAIN) * 100)
       test_score_RF = str(RF_model.score(X_TEST, Y_TEST) * 100)
       print('Train Score :',train_score_RF)
       print('Test Score',test_score_RF)
```

Train Score : 99.04072899010858
Test Score 93.56599512396242

```
[110]: from sklearn.metrics import classification_report
```

```
[111]: print(classification_report(Y_TEST,Y_Pred_RF))
```

```
-----
ValueError                                Traceback (most recent call last)
Cell In[111], line 1
----> 1 print(classification_report(Y_TEST,Y_Pred_RF))
```



```

File_
↳~\AppData\Roaming\Python\Python312\site-packages\sklearn\utils\_param_validation.py:213, in validate_params.<locals>.decorator.<locals>.wrapper(*args, **kwargs)
    207 try:
    208     with config_context(
    209         skip_parameter_validation=(
    210             prefer_skip_nested_validation or global_skip_validation
    211         )
    212     ):
--> 213         return func(*args, **kwargs)
    214 except InvalidParameterError as e:
    215     # When the function is just a wrapper around an estimator, we allow
    216     # the function to delegate validation to the estimator, but we
↳replace
    217     # the name of the estimator by the name of the function in the error
    218     # message to avoid confusion.
    219     msg = re.sub(
    220         r"parameter of \w+ must be",
    221         f"parameter of {func.__qualname__} must be",
    222         str(e),
    223     )

```

```

File_
↳~\AppData\Roaming\Python\Python312\site-packages\sklearn\metrics\_classification.py:2612, in classification_report(y_true, y_pred, labels, target_names,
↳sample_weight, digits, output_dict, zero_division)
    2477 @validate_params(
    2478     {
    2479         "y_true": ["array-like", "sparse matrix"],
    (...
    2503     zero_division="warn",
    2504 ):
    2505     """Build a text report showing the main classification metrics.
    2506
    2507     Read more in the :ref:`User Guide <classification_report>`.
    (...
    2609     <BLANKLINE>
    2610     """
-> 2612     y_type, y_true, y_pred = _check_targets(y_true, y_pred)
    2614     if labels is None:
    2615         labels = unique_labels(y_true, y_pred)

```

```

File_
↳~\AppData\Roaming\Python\Python312\site-packages\sklearn\metrics\_classification.py:108, in _check_targets(y_true, y_pred)
    105     y_type = {"multiclass"}
    107 if len(y_type) > 1:
--> 108     raise ValueError(

```

```

109         "Classification metrics can't handle a mix of {0} and {1}␣
↪targets".format(
110             type_true, type_pred
111         )
112     )
114     # We can't have more than one value on y_type => The set is no more␣
↪needed
115     y_type = y_type.pop()

```

ValueError: Classification metrics can't handle a mix of multiclass and␣
↪continuous targets

```

[ ]: import numpy as np

def score_predict(batting_team, bowling_team, runs, wickets, overs,␣
↪runs_last_5, wickets_last_5, model):
    prediction_array = []

    # One-hot encoding for batting team
    teams = ['Chennai Super Kings', 'Delhi Daredevils', 'Kings XI Punjab',␣
↪'Kolkata Knight Riders',
            'Mumbai Indians', 'Rajasthan Royals', 'Royal Challengers␣
↪Bangalore', 'Sunrisers Hyderabad']

    for team in teams:
        if batting_team == team:
            prediction_array.append(1)
        else:
            prediction_array.append(0)

    # One-hot encoding for bowling team
    for team in teams:
        if bowling_team == team:
            prediction_array.append(1)
        else:
            prediction_array.append(0)

    # Append match-specific features
    prediction_array += [runs, wickets, overs, runs_last_5, wickets_last_5]

    # Convert to numpy array and reshape for prediction
    prediction_array = np.array([prediction_array])

    # Make the prediction
    pred = RF_model.predict(prediction_array)

```

```
return int(round(pred[0]))
```

```
[ ]: batting_team = 'Delhi Daredevils'
      bowling_team = 'Chennai Super Kings'
      score = score_predict(batting_team, bowling_team, overs=10.2, runs=68,
      ↪wickets=3, runs_last_5=29, wickets_last_5=1, model=RF_model)
      print(f'Predicted Score : {score} || Actual Score : 147')
```

Cell In[1], line 4

```
    print(f'Predicted Score : {score})
```

SyntaxError: unterminated f-string literal (detected at line 4)

```
[ ]:
```