hitter-in-ipl-till-2023-assignment

May 2, 2024

```
[]: # @title
     import math
     import pandas as pd
     import numpy as np
     import seaborn as sns
     import warnings
     warnings.filterwarnings('ignore')
     pd.set_option('display.max_columns', None)
     pd.set_option('display.max_rows', None)
     pd.set_option('display.expand_frame_repr', False)
     pd.set_option('max_colwidth', None)
     import matplotlib.pyplot as plt
[]: deliveries = pd.read_csv('C:\deliveries.csv')
     matches = pd.read_csv('C:\matches.csv')
[]: def balls_per_dismissals(balls, dismissals):
         if dismissals > 0:
           return balls/dismissals
         else:
           return balls/1
     def balls_per_boundary(balls, boundaries):
         if boundaries > 0:
           return balls/boundaries
         else:
           return balls/1
     def player_Statistics(df):
         df['isDot'] = df['batsman_runs'].apply(lambda x: 1 if x == 0 else 0)
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df['isOne'] = df['batsman_runs'].apply(lambda x: 1 if x == 1 else 0)
  df['isTwo'] = df['batsman runs'].apply(lambda x: 1 if x == 2 else 0)
  df['isThree'] = df['batsman runs'].apply(lambda x: 1 if x == 3 else 0)
  df['isFour'] = df['batsman_runs'].apply(lambda x: 1 if x == 4 else 0)
  df['isSix'] = df['batsman_runs'].apply(lambda x: 1 if x == 6 else 0)
  runs = pd.DataFrame(df.groupby(['batter', 'match_id'])['batsman_runs'].
Sum().reset_index()).groupby(['batter'])['batsman_runs'].sum().reset_index().
→rename(columns={'batsman runs':'runs'})
  innings = pd.DataFrame(df.groupby(['batter'])['match_id'].apply(lambda x:__
Glen(list(np.unique(x)))).reset_index()).rename(columns = {'match_id':

        'innings'})
  balls = pd.DataFrame(df.groupby(['batter'])['match_id'].count()).

¬reset_index().rename(columns = {'match_id':'balls'})
  dismissals = pd.DataFrame(df.groupby(['batter'])['player_dismissed'].
-count()).reset_index().rename(columns = {'player_dismissed':'dismissals'})
  dots = pd.DataFrame(df.groupby(['batter'])['isDot'].sum()).reset_index().
→rename(columns = {'isDot':'dots'})
  ones = pd.DataFrame(df.groupby(['batter'])['isOne'].sum()).reset_index().
→rename(columns = {'isOne':'ones'})
  twos = pd.DataFrame(df.groupby(['batter'])['isTwo'].sum()).reset_index().
→rename(columns = {'isTwo':'twos'})
  threes = pd.DataFrame(df.groupby(['batter'])['isThree'].sum()).
Greset_index().rename(columns = {'isThree':'threes'})
  fours = pd.DataFrame(df.groupby(['batter'])['isFour'].sum()).reset_index().
→rename(columns = {'isFour':'fours'})
  sixes = pd.DataFrame(df.groupby(['batter'])['isSix'].sum()).reset_index().
→rename(columns = {'isSix':'sixes'})
  df = pd.merge(innings, runs, on = 'batter').merge(balls, on = 'batter').
omerge(dismissals, on = 'batter').merge(dots, on = 'batter').merge(ones, on = □
→ 'batter').merge(twos, on = 'batter').merge(threes, on = 'batter').
→merge(fours, on = 'batter').merge(sixes, on = 'batter')
  #StrikeRate
  df['SR'] = df.apply(lambda x: 100*(x['runs']/x['balls']), axis = 1)
  #runs per innings
  df['RPI'] = df.apply(lambda x: x['runs']/x['innings'], axis = 1)
  #balls per dismissals
  df['BPD'] = df.apply(lambda x: balls_per_dismissals(x['balls'],__

¬x['dismissals']), axis = 1)
  #balls per boundary
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df['BPB'] = df.apply(lambda x: balls_per_boundary(x['balls'], (x['fours'] + [
      \hookrightarrow x['sixes'])), axis = 1)
         return df
[]: df = player_Statistics(deliveries)
[]: df.head()
[]:
                        innings runs balls dismissals dots ones twos threes
                batter
                           SR
                                      RPI
                                                 BPD
                                                             BPB
     fours sixes
        A Ashish Reddy
                              23
                                   280
                                          196
                                                              61
                                                                    83
                                                        15
                                                                          20
                                                                                    1
            15 142.857143 12.173913
     16
                                       13.066667
                                                    6.322581
     1
              A Badoni
                              23
                                   399
                                          325
                                                        18
                                                             127
                                                                   124
                                                                          28
     24
            19 122.769231 17.347826
                                       18.055556
                                                    7.558140
            A Chandila
                                                         1
                                                   7.000000
                57.142857
                            2.000000
                                        7,000000
     0
     3
              A Chopra
                                           75
                                                                           2
                                                                                    0
                               6
                                    53
                                                         5
                                                              45
                                                                    21
     7
                70.666667
                            8.833333 15.000000
                                                  10.714286
     4
           A Choudhary
                               3
                                    25
                                           20
                                                         2
                                                               4
                                                                    13
                                                                           1
                                                                                    0
            1 125.000000
     1
                            8.333333 10.000000 10.000000
[]: def phase(over):
         if over <= 6:</pre>
           return 'Powerplay'
         if over <= 15:</pre>
           return 'Middle_overs'
         else:
           return 'Death_overs'
[]: deliveries['phase'] = deliveries['over'].apply(lambda x: phase(x))
[ ]: def phasesOfplay(df,current_phase):
         df = df[df.phase == current_phase]
         df.reset_index(inplace = True, drop = True)
         df['isDot'] = df['batsman_runs'].apply(lambda x: 1 if x == 0 else 0)
         df['isOne'] = df['batsman_runs'].apply(lambda x: 1 if x == 1 else 0)
         df['isTwo'] = df['batsman runs'].apply(lambda x: 1 if x == 2 else 0)
         df['isThree'] = df['batsman_runs'].apply(lambda x: 1 if x == 3 else 0)
         df['isFour'] = df['batsman_runs'].apply(lambda x: 1 if x == 4 else 0)
         df['isSix'] = df['batsman_runs'].apply(lambda x: 1 if x == 6 else 0)
```

```
runs = pd.DataFrame(df.groupby(['batter', 'match_id'])['batsman_runs'].
sum().reset_index()).groupby(['batter'])['batsman_runs'].sum().reset_index().
→rename(columns={'batsman_runs':'runs'})
     innings = pd.DataFrame(df.groupby(['batter'])['match_id'].apply(lambda x:__
Gen(list(np.unique(x)))).reset_index()).rename(columns = {'match_id':

        'innings'})

     balls = pd.DataFrame(df.groupby(['batter'])['match_id'].count()).
Greset_index().rename(columns = {'match_id':'balls'})
     dismissals = pd.DataFrame(df.groupby(['batter'])['player_dismissed'].
Gount()).reset_index().rename(columns = {'player_dismissed':'dismissals'})
     dots = pd.DataFrame(df.groupby(['batter'])['isDot'].sum()).reset_index().
→rename(columns = {'isDot':'dots'})
     ones = pd.DataFrame(df.groupby(['batter'])['isOne'].sum()).reset_index().

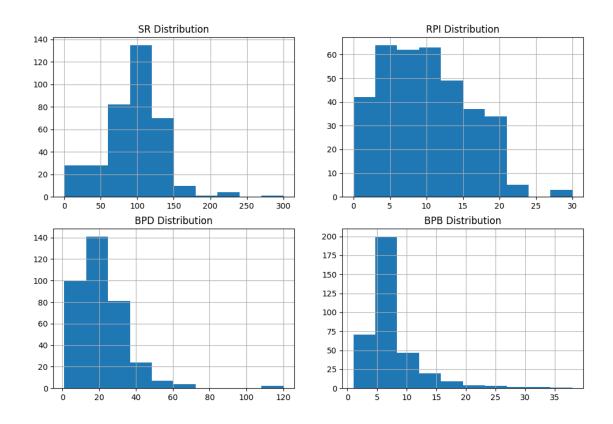
¬rename(columns = {'isOne':'ones'})
     twos = pd.DataFrame(df.groupby(['batter'])['isTwo'].sum()).reset_index().
→rename(columns = {'isTwo':'twos'})
     threes = pd.DataFrame(df.groupby(['batter'])['isThree'].sum()).
Greset index().rename(columns = {'isThree':'threes'})
     fours = pd.DataFrame(df.groupby(['batter'])['isFour'].sum()).reset_index().
→rename(columns = {'isFour':'fours'})
     sixes = pd.DataFrame(df.groupby(['batter'])['isSix'].sum()).reset_index().
→rename(columns = {'isSix':'sixes'})
     df = pd.merge(innings, runs, on = 'batter').merge(balls, on = 'batter').
omerge(dismissals, on = 'batter').merge(dots, on = 'batter').merge(ones, on = batter').merge(ones, ones).merge(ones, ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).merge(ones).m

→'batter').merge(twos, on = 'batter').merge(threes, on = 'batter').
→merge(fours, on = 'batter').merge(sixes, on = 'batter')
       #StrikeRate
     df['SR'] = df.apply(lambda x: 100*(x['runs']/x['balls']), axis = 1)
     #runs per innings
     df['RPI'] = df.apply(lambda x: x['runs']/x['innings'], axis = 1)
     #balls per dismissals
     df['BPD'] = df.apply(lambda x: balls_per_dismissals(x['balls'],__
#balls per boundary
     df['BPB'] = df.apply(lambda x: balls_per_boundary(x['balls'], (x['fours'] + _ _
\rightarrow x['sixes'])), axis = 1)
     return df
```

```
[ ]: pp_df = phasesOfplay(deliveries, 'Powerplay')
     mid_df = phasesOfplay(deliveries, 'Middle_overs')
     dth_df = phasesOfplay(deliveries, 'Death_overs')
[]: pp_df.head()
[]:
                batter innings runs balls dismissals dots ones twos threes
     fours sixes
                             RPI
                                     BPD
                                            BPB
                                                       1
                                                             5
                                                                   1
                                                                         0
                                                                                 0
       A Ashish Reddy
                              1
                                    5
     1
            0 71.428571
                           5.0
                                 7.0
                                       7.00
     1
              A Badoni
                                                            17
                                                                         0
                              4
                                   16
                                          27
                                                                   8
     2
            0 59.259259
                           4.0 13.5 13.50
                                   29
                                                            28
                                                                  13
                                                                                 0
              A Chopra
            0 64.44444
     4
                           5.8 22.5 11.25
     3
            A Flintoff
                                   15
                                                       0
                                                            11
                                                                   2
                                                                                 1
                              1
            1 93.750000 15.0 16.0
     1
                                       8.00
     4
             A Manohar
                                   16
                                          18
                                                       1
                                                             8
                                                                   8
                                                                         0
                                                                                 0
     2
            0 88.888889
                           4.0 18.0
                                       9.00
[]: plt.figure(figsize = (12,8))
     plt.subplot(221)
     pp_df.SR.hist()
     plt.title('SR Distribution')
     plt.subplot(222)
     pp_df.RPI.hist()
     plt.title('RPI Distribution')
     plt.subplot(223)
     pp_df.BPD.hist()
     plt.title('BPD Distribution')
     plt.subplot(224)
     pp_df.BPB.hist()
```

plt.title('BPB Distribution')

plt.show()



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[]: plt.figure(figsize = (15, 6))
     plt.subplot(121)
      sns.boxplot(pp_df['RPI'])
      plt.subplot(122)
      sns.distplot(pp_df['RPI'])
      plt.show()
            30
                                                         0.06
            25
                                                         0.05
            20
                                                        Density
Po.04
          ᅙ 15
                                                         0.03
            10
                                                         0.02
                                                         0.01
```

```
[]: np.percentile(pp df['RPI'], 25), np.percentile(pp df['RPI'], 50),np.
      →percentile(pp_df['RPI'], 75)
[]: (5.0, 9.5, 14.14945652173913)
[]: wt_sr, wt_rpi, wt_bpd, wt_bpb = 0.38,0.25,0.12,0.26
[]: pp_df['calc_SR'] = pp_df['SR'].apply(lambda x: x*x)
     pp_df['calc_RPI'] = pp_df['RPI'].apply(lambda x: x*x)
     pp_df['calc_BPD'] = pp_df['BPD'].apply(lambda x: x*x)
     pp_df['calc_BPB'] = pp_df['BPB'].apply(lambda x: x*x)
     sq_sr, sq_rpi, sq_bpd, sq_bpb = np.sqrt(pp_df[['calc_SR','calc_RPI',_
     G'calc_BPD', 'calc_BPB']].sum(axis = 0))
     pp_df['calc_SR'] = pp_df['calc_SR'].apply(lambda x: x/sq_sr)
     pp_df['calc_RPI'] = pp_df['calc_RPI'].apply(lambda x: x/sq_rpi)
     pp_df['calc_BPD'] = pp_df['calc_BPD'].apply(lambda x: x/sq_bpd)
     pp_df['calc_BPB'] = pp_df['calc_BPB'].apply(lambda x: x/sq_bpb)
     pp_df['calc_SR'] = pp_df['calc_SR'].apply(lambda x: x*wt_sr)
     pp_df['calc_RPI'] = pp_df['calc_RPI'].apply(lambda x: x*wt_rpi)
     pp_df['calc_BPD'] = pp_df['calc_BPD'].apply(lambda x: x*wt_bpd)
     pp_df['calc_BPB'] = pp_df['calc_BPB'].apply(lambda x: x*wt_bpb)
     best_sr, worst_sr = max(pp_df['calc_SR']), min(pp_df['calc_SR'])
     best_rpi, worst_rpi = max(pp_df['calc_RPI']), min(pp_df['calc_RPI'])
     best_bpd, worst_bpd = max(pp_df['calc_BPD']), min(pp_df['calc_BPD'])
     best_bpb, worst_bpb = min(pp_df['calc_BPB']), max(pp_df['calc_BPB'])
[]: pp_df['dev_best_SR'] = pp_df['calc_SR'].apply(lambda x: (x-best_sr)*(x-best_sr))
     pp_df['dev_best_RPI'] = pp_df['calc_RPI'].apply(lambda x:__
      →(x-best_rpi)*(x-best_rpi))
     pp_df['dev_best_BPD'] = pp_df['calc_BPD'].apply(lambda x:__
      \hookrightarrow (x-best_bpd)*(x-best_bpd))
     pp_df['dev_best_BPB'] = pp_df['calc_BPB'].apply(lambda x:__

    (x-best_bpb)*(x-best_bpb))
     pp_df['dev_best_sqrt'] = pp_df.apply(lambda x: x['dev_best_SR'] +__

¬x['dev_best_RPI'] + x['dev_best_BPD'] + x['dev_best_BPB'], axis = 1)
    pp_df['dev_worst_SR'] = pp_df['calc_SR'].apply(lambda x:__
```

```
pp_df['dev_worst_RPI'] = pp_df['calc_RPI'].apply(lambda x:__
     ⇒(x-worst_rpi)*(x-worst_rpi))
    pp_df['dev_worst_BPD'] = pp_df['calc_BPD'].apply(lambda x:__
     ⇒(x-worst_bpd)*(x-worst_bpd))
    pp_df['dev_worst_BPB'] = pp_df['calc_BPB'].apply(lambda x:__
      pp_df['dev_worst_sqrt'] = pp_df.apply(lambda x: x['dev_worst_SR'] +__

¬x['dev_worst_RPI'] + x['dev_worst_BPD'] + x['dev_worst_BPB'], axis =1)
[]: pp_df['score'] = pp_df.apply(lambda x: x['dev_worst_sqrt']/(x['dev_worst_sqrt']_
      []: pp_df[['batter', 'score']].head()
[]:
               batter
                         score
       A Ashish Reddy 0.018854
    0
    1
             A Badoni 0.013847
             A Chopra 0.015849
    2
    3
           A Flintoff 0.027478
            A Manohar 0.024306
[]: pp_df =pp_df[pp_df.innings >= 20]
[]: pp_df[[ 'batter', 'innings', 'runs', 'balls', 'dismissals', 'fours', 'sixes', u

¬'SR', 'BPB', 'score']].sort_values(['score'], ascending = False).
      →reset_index(drop = True).head(5)
[]:
                   innings runs balls dismissals fours sixes
                                                                          SR
            batter
    BPB
            score
         SP Narine
                             688
                                    423
                         44
                                                 38
                                                        86
                                                              40
                                                                  162.647754
    3.357143 0.158987
    1 YBK Jaiswal
                         37
                             808
                                    550
                                                 19
                                                       113
                                                              31
                                                                  146.909091
    3.819444 0.107503
           CA Lynn
                         40
                             875
                                    623
                                                 19
                                                       103
                                                              45
                                                                  140.449438
    4.209459 0.091021
    3 JM Bairstow
                             747
                                                       91
                                                                  139.106145
                         36
                                    537
                                                 19
                                                              31
    4.401639 0.087271
          V Sehwag
                        103 1785
                                   1284
                                                 69
                                                       250
                                                              53
                                                                  139.018692
    4.237624 0.086096
[]: #THE BEST PINCH HITTER INSIDE PP IN THE IPL SO FAR (MIN 20 INNINGS) TILL 2023
      →IPL FINAL IS SUNIL NARINE WITH SCORE OF 0.15.
```