```
In [1]: import pandas as pd
          import copy
In [2]: xls = pd.ExcelFile('Cricket Data Set.xlsx')
Out[2]: <pandas.io.excel._base.ExcelFile at 0x29082be1ac0>
In [3]: print(len(xls.sheet_names))
          xls.sheet_names
         12
Out[3]: ['Player Details',
           'Parameters',
           'Control Sheet',
           'IND Vs SA',
           'IND Vs PAK',
           'IND Vs AUS',
           'IND Vs AFG',
           'IND Vs WI',
           'IND Vs ENG',
           'IND Vs BAN',
           'IND Vs SL',
           'Analysis parameters']
          df_names=[] for sheet_names in xls.sheet_names: df_name = " for name in sheet_names.split(): df_name += f'{name.lower()}_' df_name
          += 'df' df_names.append(df_name)
          print(len(df_names))
         df names
In [4]: df_names = ['_'.join(sheet_names.lower().split()) + '_df' for sheet_names in xls.sheet_names]
          print(len(df_names))
          df_names
         12
Out[4]: ['player_details_df',
           'parameters_df',
           'control_sheet_df',
           'ind_vs_sa_df'
           'ind_vs_sa_df',
'ind_vs_pak_df',
           'ind_vs_aus_df',
           'ind_vs_afg_df',
           'ind_vs_wi_df',
           'ind_vs_eng_df',
           'ind_vs_ban_df',
           'ind_vs_sl_df',
           'analysis_parameters_df']
          sheet index = 0
         for df_name in df_names: exec(f'{df_name} = pd.read_excel(xls,sheet_index)') sheet_index +=1
In [5]: df_dict = {df_names[sheet_index] : pd.read_excel(xls,sheet_index) for sheet_index in range(len(df_names))}
          df_dict.keys()
Out[5]: dict_keys(['player_details_df', 'parameters_df', 'control_sheet_df', 'ind_vs_sa_df', 'ind_vs_pak_df', 'ind_vs_ aus_df', 'ind_vs_afg_df', 'ind_vs_wi_df', 'ind_vs_eng_df', 'ind_vs_ban_df', 'ind_vs_sl_df', 'analysis_paramete
          rs_df'])
In [6]: pd.set_option('max_columns', None)
```

In [7]: for df_name in df_names:
 print(df_name)
 display(df_dict[df_name].head(5))

player_details_df

	S0No	Player ID	Player Name	Type of Player	Jersey Number	Country
0	1.0	INDO175	Virat Kohli (c)	Batsmen	18.0	India
1	2.0	INDO168	Rohit Sharma (vc)	Batsmen	45.0	India
2	3.0	INDO230	Mayank Agarwal	Batsmen	16.0	India
3	4.0	INDO210	Jasprit Bumrah	Bowler	93.0	India
4	5.0	INDO211	Yuzvendra Chahal	Bowler	3.0	India

parameters_df

	Ball Type	ball symbol	Shot Type	shot symbol	Dismissal Kind	dismisal symbol	Pitch type
0	Full Toss	FT	Defend	D	Catch	Ct	Green tops

In [8]: | df_dict['control_sheet_df'].head(30)

Out[8]:

	Unnamed: 0	Unnamed: 1	F	Correct execution of shot according to the ball	Effectiveness of the shot execution	Υ	Correct execution of shot according to the ball.1	Effectiveness of the shot execution.1	FT	Correct execution of shot according to the ball.2	Effectiveness of the shot execution.2	G	Correct execution of shot according to the ball.3
0	D	1.0	0.0	1.0	1.0	0.0	1.0	1.0	0.0	1.0	1.0	0.0	1.0
1	NaN	1.0	1.0	1.0	0.8	1.0	1.0	0.8	1.0	1.0	0.8	1.0	1.0
2	NaN	1.0	2.0	1.0	0.6	2.0	1.0	0.6	2.0	1.0	0.6	2.0	1.0
3	NaN	1.0	3.0	1.0	0.4	3.0	1.0	0.4	3.0	1.0	0.4	3.0	1.0
4	NaN	1.0	4.0	1.0	0.2	4.0	1.0	0.2	4.0	1.0	0.2	4.0	1.0
5	NaN	1.0	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0
6	NaN	1.0	6.0	1.0	0.2	6.0	1.0	0.2	6.0	1.0	0.2	6.0	1.0
7	NaN	NaN	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0
8	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
9	Dr	1.0	0.0	1.0	0.2	0.0	1.0	0.2	0.0	1.0	0.2	0.0	1.0
10	NaN	1.0	1.0	1.0	0.4	1.0	1.0	0.4	1.0	1.0	0.4	1.0	1.0
11	NaN	1.0	2.0	1.0	0.5	2.0	1.0	0.5	2.0	1.0	0.5	2.0	1.0
12	NaN	1.0	3.0	1.0	0.6	3.0	1.0	0.6	3.0	1.0	0.6	3.0	1.0
13	NaN	1.0	4.0	1.0	1.0	4.0	1.0	1.0	4.0	1.0	1.0	4.0	1.0
14	NaN	1.0	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0
15	NaN	1.0	6.0	1.0	1.0	6.0	1.0	1.0	6.0	1.0	1.0	6.0	1.0
16	NaN	NaN	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0
17	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
18	С	1.0	0.0	0.0	0.2	0.0	0.0	0.2	0.0	1.0	0.2	0.0	1.0
19	NaN	1.0	1.0	0.0	0.4	1.0	0.0	0.2	1.0	1.0	0.4	1.0	1.0
20	NaN	1.0	2.0	0.0	0.4	2.0	0.0	0.2	2.0	1.0	0.4	2.0	1.0
21	NaN	1.0	3.0	0.0	0.4	3.0	0.0	0.2	3.0	1.0	0.4	3.0	1.0
22	NaN	1.0	4.0	0.0	0.5	4.0	0.0	0.2	4.0	1.0	0.5	4.0	1.0
23	NaN	1.0	5.0	0.0	0.2	5.0	0.0	0.2	5.0	1.0	0.2	5.0	1.0
24	NaN	1.0	6.0	0.0	0.5	6.0	0.0	0.2	6.0	1.0	0.5	6.0	1.0
25	NaN	NaN	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0
26	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
27	Р	1.0	0.0	0.0	0.2	0.0	0.0	0.2	0.0	0.0	0.2	0.0	1.0
28	NaN	1.0	1.0	0.0	0.2	1.0	0.0	0.2	1.0	0.0	0.2	1.0	1.0
29	NaN	1.0	2.0	0.0	0.2	2.0	0.0	0.2	2.0	0.0	0.2	2.0	1.0
4													

```
In [9]: df_dict['control_sheet_df'].rename({'Unnamed: 0': 'Shot Type'}, axis=1, inplace=True)
df_dict['control_sheet_df'].drop('Unnamed: 1',axis = 1, inplace = True)
```

In [10]: df_dict['control_sheet_df'].head(30)

Out[10]:

	Shot Type	F	Correct execution of shot according to the ball	Effectiveness of the shot execution	Y	Correct execution of shot according to the ball.1	Effectiveness of the shot execution.1	FT	Correct execution of shot according to the ball.2	Effectiveness of the shot execution.2	G	Correct execution of shot according to the ball.3	Effectiveness of the shot execution.3	:
0	D	0.0	1.0	1.0	0.0	1.0	1.0	0.0	1.0	1.0	0.0	1.0	1.0	
1	NaN	1.0	1.0	0.8	1.0	1.0	0.8	1.0	1.0	0.8	1.0	1.0	0.8	
2	NaN	2.0	1.0	0.6	2.0	1.0	0.6	2.0	1.0	0.6	2.0	1.0	.0 0.6	
3	NaN	3.0	1.0	0.4	3.0	1.0	0.4	3.0	1.0	0.4	3.0	1.0	0.4	
4	NaN	4.0	1.0	0.2	4.0	1.0	0.2	4.0	1.0	0.2	4.0	1.0	0.2	
5	NaN	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	
6	NaN	6.0	1.0	0.2	6.0	1.0	0.2	0.2 6.0 1.0 0.2 6.0		1.0	0.2			
7	NaN	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	1
8	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Ν
9	Dr	0.0	1.0	0.2	0.0	1.0	0.2	0.0	1.0	0.2	0.0	1.0	0.2	
10	NaN	1.0	1.0	0.4	1.0	1.0	0.4	1.0	1.0	0.4	1.0	1.0	0.4	
11	NaN	2.0	1.0	0.5	2.0	1.0	0.5	2.0	1.0	0.5	2.0	1.0	0.5	
12	NaN	3.0	1.0	0.6	3.0	1.0	0.6	3.0	1.0	0.6	3.0	1.0	0.6	
13	NaN	4.0	1.0	1.0	4.0	1.0	1.0	4.0	1.0	1.0	4.0	1.0	1.0	
14	NaN	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	
15	NaN	6.0	1.0	1.0	6.0	1.0	1.0	6.0	1.0	1.0	6.0	1.0	1.0	
16	NaN	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	1
17	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Ν
18	С	0.0	0.0	0.2	0.0	0.0	0.2	0.0	1.0	0.2	0.0	1.0	0.2	
19	NaN	1.0	0.0	0.4	1.0	0.0	0.2	1.0	1.0	0.4	1.0	1.0	0.4	
20	NaN	2.0	0.0	0.4	2.0	0.0	0.2	2.0	1.0	0.4	2.0	1.0	0.5	
21	NaN	3.0	0.0	0.4	3.0	0.0	0.2	3.0	1.0	0.4	3.0	1.0	0.6	
22	NaN	4.0	0.0	0.5	4.0	0.0	0.2	4.0	1.0	0.5	4.0	1.0	1.0	
23	NaN	5.0	0.0	0.2	5.0	0.0	0.2	5.0	1.0	0.2	5.0	1.0	0.2	
24	NaN	6.0	0.0	0.5	6.0	0.0	0.2	6.0	1.0	0.5	6.0	1.0	1.0	
25	NaN	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	Out	0.0	-1.0	1
26	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Ν
27	Р	0.0	0.0	0.2	0.0	0.0	0.2	0.0	0.0	0.2	0.0	1.0	0.2	
28	NaN	1.0	0.0	0.2	1.0	0.0	0.2	1.0	0.0	0.2	1.0	1.0	0.4	
29	NaN	2.0	0.0	0.2	2.0	0.0	0.2	2.0	0.0	0.2	2.0	1.0	0.5	

In [11]: df_dict['control_sheet_df'].shape

Out[11]: (107, 28)

```
In [12]: for row_index in range(8,df_dict['control_sheet_df'].shape[0],9):
             print(row_index)
         8
         17
         26
         35
         44
         53
         62
         71
         89
         98
In [13]: df_dict['control_sheet_df'].isna().sum()
Out[13]: Shot Type
                                                                95
                                                                11
         Correct execution of shot according to the ball
                                                                11
         Effectiveness of the shot execution
                                                                11
                                                                11
         Correct execution of shot according to the ball.1
                                                                11
         Effectiveness of the shot execution.1
                                                                11
         FT
                                                                11
         Correct execution of shot according to the ball.2
                                                                11
         Effectiveness of the shot execution.2
                                                               11
                                                               11
         Correct execution of shot according to the ball.3
                                                                11
         Effectiveness of the shot execution.3
                                                                11
                                                                11
         Correct execution of shot according to the ball.4
                                                                11
         Effectiveness of the shot execution.4
                                                               11
         S1
                                                               11
         Correct execution of shot according to the ball.5
                                                                11
         Effectiveness of the shot execution.5
                                                                11
                                                                11
         Correct execution of shot according to the ball.6
                                                                11
         Effectiveness of the shot execution.6
                                                               11
                                                               11
         Correct execution of shot according to the ball.7
                                                               11
         Effectiveness of the shot execution.7
                                                               11
         ΙB
                                                               11
         Correct execution of shot according to the ball.8
                                                                11
         Effectiveness of the shot execution.8
                                                               11
         dtype: int64
In [14]: df_dict['control_sheet_df'] = df_dict['control_sheet_df'].dropna(how='all')
         df dict['control sheet df'].isna().sum()
Out[14]: Shot Type
                                                                84
                                                                а
         Correct execution of shot according to the ball
                                                                0
         Effectiveness of the shot execution
                                                                0
                                                                a
         Correct execution of shot according to the ball.1
                                                                0
         Effectiveness of the shot execution.1
                                                                0
         Correct execution of shot according to the ball.2
                                                                0
         Effectiveness of the shot execution.2
                                                                0
         G
                                                                a
         Correct execution of shot according to the ball.3
                                                                0
         Effectiveness of the shot execution.3
                                                                0
                                                                0
         Correct execution of shot according to the ball.4
                                                                0
         Effectiveness of the shot execution.4
                                                                0
                                                                a
         Correct execution of shot according to the ball.5
                                                                0
         Effectiveness of the shot execution.5
                                                                0
         Correct execution of shot according to the ball.6
                                                                0
         Effectiveness of the shot execution.6
                                                                a
         Correct execution of shot according to the ball.7
                                                                0
         Effectiveness of the shot execution.7
                                                                0
         LB
         Correct execution of shot according to the ball.8
                                                                0
         Effectiveness of the shot execution.8
                                                                0
          dtype: int64
```

```
In [15]: df_dict['parameters_df']
               Ball Type ball symbol
                                       Shot Type shot symbol
                                                               Dismissal Kind dismisal symbol
                                                                                                  Pitch type
            0
               Full Toss
                               FT
                                          Defend
                                                          D
                                                                       Catch
                                                                                         Ct
                                                                                                  Green tops
                                Υ
                                      OFF DRIVE
                                                                                         В
            1
                                                         Dr
                                                                     Bowled
                 Yorker
                                                                                             Hard and bouncy
            2
                   Full
                                F
                                             Cut
                                                          С
                                                                     Run Out
                                                                                        RO
                                                                                               Slow and dusty
                                                          Р
            3
                                G
                                            Pull
                                                             Leg Before Wicket
                                                                                       LBW
                                                                                            Dead and flat pitch
                  Good
                                S
                                                                                         St
                  Short
                                          Sweep
                                                         Sw
                                                                    Stumped
                                                                                                       NaN
            5
                 slower
                                SL
                                   Reverse sweep
                                                         RS
                                                                    Hit Wicket
                                                                                         Н
                                                                                                       NaN
                  length
                                 L
                                           Leave
                                                          L
                                                                        NaN
                                                                                       NaN
                                                                                                       NaN
                               GΥ
                                       ONDRIVE
                                                        ODR
                                                                        NaN
                                                                                       NaN
                                                                                                       NaN
                 googly
                                          FLICK
                                                          F
                               LB
                                                                        NaN
                                                                                       NaN
                                                                                                       NaN
               leg break
                              NaN
                                          RAMP
                                                          R
                                                                        NaN
                                                                                       NaN
                                                                                                       NaN
                   NaN
           10
                                                         ΙG
                                                                                                       NaN
                   NaN
                              NaN
                                      Lea Glance
                                                                        NaN
                                                                                       NaN
           11
                   NaN
                              NaN
                                                         SC
                                                                        NaN
                                                                                       NaN
                                                                                                       NaN
                                           scoop
In [16]: shot_type = df_dict['control_sheet_df'].pop('Shot Type')
In [17]: shot_type
Out[17]: 0
                    D
                  NaN
          2
                  NaN
          3
                  NaN
          4
                  NaN
          102
                  NaN
          103
                  NaN
          104
                  NaN
          105
                  NaN
          106
                  NaN
          Name: Shot Type, Length: 96, dtype: object
In [18]: shot_type.value_counts()
Out[18]: D
                  1
          Dr
                  1
          C
                  1
          Ρ
                  1
          Sw
                  1
          RS
                  1
          ODR
                  1
          F
          R
                  1
          LG
                  1
          SC
                  1
          1
                  1
          Name: Shot Type, dtype: int64
In [19]: shot_type.isna().sum()
Out[19]: 84
In [20]: shot_type_temp = copy.deepcopy(shot_type)
          shot_type_temp
Out[20]: 0
                    D
          1
                  NaN
          2
                  NaN
          3
                  NaN
          4
                  NaN
          102
                  NaN
          103
                  NaN
          104
                  NaN
          105
                  NaN
          106
                  NaN
          Name: Shot Type, Length: 96, dtype: object
```

```
In [21]: for index,value in shot_type_temp.items():
              if type(value) != str:
                   shot_type_temp[index] = shot_type_temp[index-1]
          shot_type_temp
Out[21]: 0
                  D
          1
          2
                  D
          3
                  D
          4
                  D
          102
          103
                  L
          104
          105
                  L
          106
          Name: Shot Type, Length: 96, dtype: object
In [22]: shot_type_temp.value_counts()
Out[22]: D
                  8
          Dr
                  8
          C
                  8
          Р
                  8
          Sw
                  8
          RS
                  8
          F
                  8
          R
                  8
          LG
                  8
          SC
                  8
          L
                  8
          Name: Shot Type, dtype: int64
In [23]: shot_type = shot_type_temp
          shot_type.value_counts()
Out[23]: D
                  8
          Dr
                  8
          C
                  8
          Р
                  8
          Sw
                  8
          RS
                  8
          ODR
                  8
          F
                  8
          R
                  8
          LG
                  8
          SC
                  8
          L
                  8
          Name: Shot Type, dtype: int64
In [24]: df_dict['control_sheet_df'].shape
Out[24]: (96, 27)
In [25]: ball_type_df = df_dict['parameters_df'].loc[:,['Ball Type','ball symbol']].dropna()
          ball_type_dict = dict(zip(ball_type_df['ball symbol'],ball_type_df['Ball Type']))
          ball_type_dict
Out[25]: {'FT': 'Full Toss', 'Y': 'Yorker',
            'F': 'Full',
            'G': 'Good',
           'S': 'Short',
'SL': 'slower',
'L': 'length',
'GY': 'googly',
'LB': 'leg break'}
```

```
In [26]: |print(df_dict['control_sheet_df'].columns)
          df_dict['control_sheet_df'].rename(columns = {'Sl':'SL'}, inplace = True)
          df_dict['control_sheet_df'].columns
          Index(['F', 'Correct execution of shot according to the ball',
                  'Effectiveness of the shot execution', 'Y',
                  'Correct execution of shot according to the ball.1',
                  'Effectiveness of the shot execution.1', 'FT',
                  'Correct execution of shot according to the ball.2',
                  'Effectiveness of the shot execution.2', 'G',
                  'Correct execution of shot according to the ball.3',
                 'Effectiveness of the shot execution.3', 'S',
                 'Correct execution of shot according to the ball.4',
                  'Effectiveness of the shot execution.4', 'S1',
                  'Correct execution of shot according to the ball.5',
                  'Effectiveness of the shot execution.5', 'L',
                  'Correct execution of shot according to the ball.6',
                 'Effectiveness of the shot execution.6', 'GY',
                 'Correct execution of shot according to the ball.7',
                  'Effectiveness of the shot execution.7', 'LB',
                  'Correct execution of shot according to the ball.8',
                 'Effectiveness of the shot execution.8'],
                dtype='object')
Out[26]: Index(['F', 'Correct execution of shot according to the ball',
                  'Effectiveness of the shot execution', 'Y',
                  'Correct execution of shot according to the ball.1',
                  'Effectiveness of the shot execution.1', 'FT',
                  'Correct execution of shot according to the ball.2',
                  'Effectiveness of the shot execution.2', 'G',
                  'Correct execution of shot according to the ball.3',
                 'Effectiveness of the shot execution.3', 'S',
                  'Correct execution of shot according to the ball.4',
                  'Effectiveness of the shot execution.4', 'SL',
                  'Correct execution of shot according to the ball.5',
                  'Effectiveness of the shot execution.5', 'L',
                  'Correct execution of shot according to the ball.6',
                 'Effectiveness of the shot execution.6', 'GY',
                  'Correct execution of shot according to the ball.7',
                  'Effectiveness of the shot execution.7', 'LB',
                  'Correct execution of shot according to the ball.8',
                  'Effectiveness of the shot execution.8'],
                dtype='object')
In [27]: def generate_ballwise_df(df = df_dict['control_sheet_df']):
              for col_index in range(0,df.shape[1],3):
                  ball_symbol = df.iloc[:,col_index:col_index+3].columns[0]
                  ball_type = ball_type_dict[ball_symbol]
                  df_dict[f'{ball_type}_shots_df'] = df.iloc[:,col_index:col_index+3]
                  df_dict[f'{ball_type}_shots_df'].columns = ['Runs Scored', 'Correct execution of shot according to the
df_dict[f'{ball_type}_shots_df'].insert(0, 'Ball Type', ball_symbol)
df_dict[f'{ball_type}_shots_df'].insert(1, 'Shot Type', shot_type)
In [28]: generate_ballwise_df()
```

	Ball Type	ball symbol	Shot Type	shot symbol	Dismissal Kind	dismisal symbol	Pitch type
0	Full Toss	FT	Defend	D	Catch	Ct	Green tops
1	Yorker	Υ	OFF DRIVE	Dr	Bowled	В	Hard and bouncy
2	Full	F	Cut	С	Run Out	RO	Slow and dusty
3	Good	G	Pull	Р	Leg Before Wicket	LBW	Dead and flat pitch
4	Short	S	Sweep	Sw	Stumped	St	NaN
5	slower	SL	Reverse sweep	RS	Hit Wicket	Н	NaN
6	length	L	Leave	L	NaN	NaN	NaN
7	googly	GY	ONDRIVE	ODR	NaN	NaN	NaN
8	leg break	LB	FLICK	F	NaN	NaN	NaN
9	NaN	NaN	RAMP	R	NaN	NaN	NaN
					_		

```
In [30]: df_dict['analysis_parameters_df'] = df_dict['analysis_parameters_df'].dropna(how = 'all')
df_dict['analysis_parameters_df'].reset_index(drop = True, inplace = True)
```

Out[31]:

	Analysis Parameters	Formulae	Parameters	Unnamed: 3	Unnamed: 4
0	Best off side player	Highest average of effectiveness of shot of all off side shots	Dr,C and RS	The highest percentage of shots in Dr, C and RS	O= ((no of Dr+C+RS)/ total number of balls faced) * ((sum of effectiveness of Dr +C+RS)/no of Dr+C+RS)
1	Best Leg side player	Highest average of effectiveness of shot of all Leg side shots	P, ODR, LG, F and S	The highest percentage of shots in P, ODR, LG, F and S	O= ((no of P+ODR+LG+F+S)/ total number of balls faced) * ((sum of effectiveness of P+ODR+LG+F+S)/no of P+ODR+LG+F+S)
2	Best player for bouncy tracks	Highest average of effectiveness of shot of all Bouncy pitch shots	D, P, C and L	The highest percentage of shots in D, P, C and L	O= ((no of D+P+C+L)/ total number of balls faced) * ((sum of effectiveness of D+P+C+L)/no of D+P+C+L)
3	Best player for high scoring matches	Highest average of effectiveness of shot of all agressive shots	D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	The highest percentage of shots in D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	O= ((no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/ total number of balls faced) * ((sum of effectiveness of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)
4	Best defencive player	NaN	D and L	The highest percentage of shots in D and L	O= ((no of D+L)/ total number of balls faced) * ((sum of effectiveness of D+L)/no of D+L)
5	Best spin ball player	Highest average of effectiveness of shot of all spin ball shots	D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	The highest percentage of shots in D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	O= ((no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/ total number of balls faced) * ((sum of effectiveness of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)
6	The player with the best control	Highest control of shots	D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	The highest percentage of shots in D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	O= ((no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/ total number of balls faced) * ((sum of effectiveness of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)

```
In [32]: def unique_values(df):
    unique_vals = [(col,list(df[col].unique())) for col in df.columns if col!='Sno']
    return unique_vals
```

```
In [33]: for df_name in df_dict.keys():
                 if 'vs' not in df_name:
                       continue
                  print(df name)
                  for col,vals in unique_values(df_dict[df_name]):
                       print(f'{col} : {vals}')
                  print()
            ind_vs_sa_df
            Player Name : ['Hardik Pandya', 'Rohit Sharma', 'MS Dhoni', 'KL Rahul', 'Virat Kohli', 'Shikhar Dhawan'] Player ID : ['INDO215', 'INDO168', 'INDO157', 'INDO213', 'INDO175', 'INDO188']
            Match ID : ['WC19M08']
            Pitch Type : [nan]
            Bowler : ['Andile Phehlukwayo', 'Chris Morris', 'Kagiso Rabada', 'Imran Tahir', 'Tabraiz Shamsi']
            Player ID.1 : ['SA0118', 'SA0110', 'SA0114', 'SA0102', 'SA0116']
            Over No : [47.0, 46.0, 45.0, 44.0, 43.0, 42.0, 41.0, 40.0, 39.0, 38.0, 37.0, 36.0, 35.0, 34.0, 33.0, 32.0,
             31.0, 30.0, 29.0, 28.0, 27.0, 26.0, 25.0, 24.0, 23.0, 22.0, 21.0, 20.0, 19.0, 18.0, 17.0, 16.0, 15.0, 14.
            0, 13.0, 12.0, 11.0, 10.0, 9.0, 8.0, 7.0, 6.0, 5.0, 4.0, 3.0, 2.0, 1.0, 0.0]
            Ball No : [3.0, 2.0, 1.0, 6.0, 5.0, 4.0]
Ball type : ['F', 'S', 'L', 'FT', 'Y', 'FL']
Type of Shot : ['C', 'L', 'DR', 'P', 'SW', 'ODR', 'D', 'S', 'LG', 'RS']
Wide : [nan, 'Yes', 'Wide']
            No ball : [nan]
            Hit the Bat : [1.0, nan, 0.0]
            Control: [0.0, 1.0, -1.0]
            X : [0.5, 0.0, 1.0, 0.4, 0.2, -1.0, 0.6, 0.8]
            Runs: [4.0, '1w', 0.0, 2.0, 1.0, 'W', 3.0, '1lb', 6.0]
In [34]: for df_name in df_dict.keys():
                if 'vs' in df_name:
                       continue
                  print(df_name)
                  for col,vals in unique_values(df_dict[df_name]):
                      print(f'{col} : {vals}')
                  print()
            n', 'Liam Dawson', 'Liam Plunkett', 'Adil Rashid', 'Joe Root', 'Jason Roy', 'Ben Stokes', 'James Vince',
            'Chris Woakes', 'Mark Wood', 'Joe Denly', 'Alex Hales', 'David Willey', 'Mashrafe Mortaza (c)', 'Abu Jaye d', 'Litton Das †', 'Mahmudullah', 'Mehidy Hasan Miraz', 'Mohammad Mithun', 'Mohammad Saifuddin', 'Mosadde k Hossain', 'Mushfiqur Rahim †', 'Mustafizur Rahman', 'Rubel Hossain', 'Sabbir Rahman', 'Shakib Al Hasan (vc)', 'Soumya Sarkar', 'Tamim Iqbal', 'Dimuth Karunaratne (c)', 'Dhananjaya de Silva', 'Avishka Fernand o', 'Suranga Lakmal', 'Lasith Malinga', 'Angelo Mathews', 'Kusal Mendis †', 'Jeevan Mendis', 'Kusal Perera †', 'Thisara Perera', 'Milinda Siriwardana', 'Lahiru Thirimanne', 'Isuru Udana', 'Jeffrey Vandersay', 'Nuw
             an Pradeep', 'Kasun Rajitha']
             Type of Player : ['Batsmen', 'Bowler', 'Wicket Keeper', 'All Rounder', 'Batting All Rounder', 'Bowling All
            Rounder', 'Batter']
             Jersey Number: [18.0, 45.0, 16.0, 93.0, 3.0, 7.0, 8.0, 21.0, 79.0, 23.0, 15.0, 11.0, 17.0, 33.0, 1.0, 42.
            0, 59.0, 90.0, 12.0, 14.0, 99.0, 4.0, 10.0, 2.0, 22.0, 29.0, 25.0, 26.0, 72.0, 20.0, 5.0, 65.0, 6.0, 30.0,
            54.0, 67.0, 32.0, 47.0, 49.0, 56.0, 13.0, 31.0, 63.0, 9.0, 60.0, 39.0, 89.0, 87.0, 40.0, 41.0, 83.0, 44.0,
            66.0, 50.0, nan, 88.0, 19.0, 55.0, 77.0, 97.0, 46.0, 85.0, 24.0, 51.0, 95.0, 53.0, 74.0, 34.0, 75.0, 28.0,
            82.0, 69.0, 86.0, 57.01
            Country : ['India', 'South Africa', 'Australia', 'Pakistan', 'Afghanistan', 'West Indies', 'England', 'Ban
            gladesh', 'Srilanka']
            parameters_df
```

Ball Type: ['Full Toss', 'Yorker', 'Full', 'Good', 'Short', 'slower', 'length', 'googly', 'leg break', na

```
In [35]: for df_name in df_dict.keys():
             if 'vs' not in df_name:
                 continue
             print(df name)
             print(df_dict[df_name][df_dict[df_name]['Dismissal kind'].notnull()])
         \verb"ind_vs_sa_df"
                Sno
                        Player Name Player ID Match ID Pitch Type
                                                                                Bowler
               10.0
                           MS Dhoni
                                    INDO157 WC19M08
                                                               NaN
                                                                          Chris Morris
         98
               99.0
                           KI Rahul
                                      TND0213 WC19M08
                                                               NaN
                                                                         Kagiso Rabada
         195 196.0
                        Virat Kohli
                                      IND0175 WC19M08
                                                               NaN
                                                                   Andile Phehlukwayo
         258 259.0 Shikhar Dhawan
                                      IND0188 WC19M08
                                                               NaN
                                                                         Kagiso Rabada
             Player ID.1 Over No Ball No Ball type Type of Shot Wide
                                                                        No ball
                  SA0110
                             46.0
                                       1.0
                                                   S
                                                               DR NaN
                                                                            NaN
                                                               DR NaN
         98
                  SA0114
                             31.0
                                       3.0
                                                   L
                                                                            NaN
         195
                  SA0118
                             15.0
                                       3.0
                                                   S
                                                               DR
                                                                   NaN
                                                                            NaN
         258
                  SA0114
                              5.0
                                       1.0
                                                   S
                                                                C
                                                                   NaN
                                                                            NaN
              Hit the Bat
                           Control
                                      X Runs Wicket Dismissal kind Caught/Bowled By
         9
                               0.0 -1.0
                                           W Wicket
                      NaN
                                                                  C
                                                                  C
         98
                      NaN
                               1.0 -1.0
                                           W Wicket
                                                                              SA0101
                               1.0 -1.0
         195
                      NaN
                                           W
                                              Wicket
                                                                  C
                                                                              SA0101
                               0.0 -1.0
                                           W Wicket
                                                                  C
                                                                              SA0105
         258
                      NaN
         ind_vs_pak_df
                                          ------
In [36]: |df_dict['ind_vs_afg_df']['Dismissal kind'] = df_dict['ind_vs_afg_df']['Dismissal kind'].replace('Leg Stump', 'B
         df_dict['ind_vs_afg_df']['Dismissal kind'] = df_dict['ind_vs_afg_df']['Dismissal kind'].replace('Stump Out',
In [37]: | df_dict['ind_vs_afg_df'].loc[df_dict['ind_vs_afg_df']['Dismissal kind'].notnull() & df_dict['ind_vs_afg_df']['W
In [38]: def clean_missing_values(df):
             cols_nan = [col_name for col_name, nan_count in df.isna().sum().items() if nan_count!=0 ]
             unary_cols = ['Pitch Type', 'Match ID']
             binary_cols = ['Wide', 'No ball', 'Wicket', 'Hit the Bat']
             for col in cols_nan:
                 if col in unary_cols:
                     df.drop(col, axis = 1, inplace = True)
                     continue
                 elif col in binary_cols:
                     df[col] = df[col].notnull().astype('float')
                     continue
                 df[col].fillna(value = 0, inplace = True)
             return df
In [39]: def drop_incorrect_data(df,column_name,child_values,parent_values):
             if all(value in parent_values for x in child_values):
                 return
             incorrect_values = [val for val in child_values if val not in parent_values]
             df.drop(df[df[column_name].isin(incorrect_values)].index, inplace=True)
```

```
In [40]: def validate_column_values(df):
             ball_symbol_list = df_dict['parameters_df']['ball symbol'].dropna().tolist()
             ball_symbol_list = list(map(lambda x: x.upper(),ball_symbol_list))
             shot_symbol_list = df_dict['parameters_df']['shot symbol'].dropna().tolist()
             shot_symbol_list = list(map(lambda x: x.upper(),shot_symbol_list))
             dismissal_symbol_list = df_dict['parameters_df']['dismisal_symbol'].dropna().tolist()
             dismissal_symbol_list = list(map(lambda x: x.upper(),dismissal_symbol_list))
             dismissal_dict = dict(zip(df_dict['parameters_df']['Dismissal Kind'].dropna(),df_dict['parameters_df']['dis
             control_values_list = [0.0,1.0]
             effectiveness_values_list = [x/10.0 \text{ for } x \text{ in } range(0,11)] + [-1.0]
             for col_name, col_values in unique_values(df):
                 if 'shot' in col_name.lower():
                     col_values = list(map(lambda x: x.upper(), col_values))
                     drop_incorrect_data(df,col_name,col_values,shot_symbol_list)
                 elif col_name.lower() == 'ball type':
                     col_values = list(map(lambda x: x.upper(), col_values))
                     drop incorrect data(df,col name,col values,ball symbol list)
                 elif 'control' in col_name.lower():
                     \label{eq:dfcol_name} $$ df[col_name].replace(-1.0,0.0) if -1.0 in col_values else df[col_name] $$
                     col_values.remove(-1.0) if -1.0 in col_values else col_values
                     drop_incorrect_data(df,col_name,col_values,control_values_list)
                  elif 'effectiveness' in col_name.lower():
                     drop_incorrect_data(df,col_name,col_values,effectiveness_values_list)
                  elif 'dismissal' in col_name.lower():
                     col values.remove(0)
                     df[col_name] = df[col_name].replace('C','Ct') if 'C' in col_values else df[col_name]
                     col_values = ['Ct' if col_value == 'C' else col_value for col_value in col_values]
                     for col value in col values:
                         df[col_name] = df[col_name].replace(col_value,dismissal_dict[col_value]) if len(col_value) > 3
                     col_values = [dismissal_dict[col_value] if len(col_value) > 3 else col_value for col_value in col_v
                     col_values = [col_value.upper() for col_value in col_values]
                     drop_incorrect_data(df,col_name,col_values,dismissal_symbol_list)
             return df
In [41]: def clean_df(df):
             df.drop('Sno', axis = 1, inplace = True)
             df = df.dropna(how = 'all')
             df.reset_index(drop = True, inplace = True)
             df.rename(columns = {'Player Name': 'Batsman Name', 'Player ID': 'Batsman Player ID', 'Bowler': 'Bowler Name',
             non_nan_df = clean_missing_values(df)
             validated df = validate column values(non nan df)
             return validated_df
In [42]: length = 0
         for df_name in df_dict.keys():
             if 'vs' not in df_name:
                 continue
             length += df_dict[df_name].shape[0]
         print(length)
         2397
In [43]: matches_df = pd.DataFrame()
         for df_name in df_dict.keys():
             if 'vs' not in df_name:
                 continue
             cleaned_df = clean_df(df_dict[df_name])
             matches_df = matches_df.append(cleaned_df, ignore_index = True)
         matches_df.shape
Out[43]: (2387, 18)
```

```
In [44]: for df_name in df_dict.keys():
                  if 'vs' not in df_name:
                       continue
                  print(df_name)
                  for col_name, col_values in unique_values(df_dict[df_name]):
                       print(col_name, col_values)
                  print()
            Nums [ w , 1.0, 0.0, 4.0, 2.0, 410 , 3.0, 1w , 110 , 0.0, 3ml ] Wicket ['Wicket', nan]
             Dismissal kind ['Bowled', 'Run Out', nan, 'Catch']
            Caught/Bowled By ['BAN0118', 'BAN081', nan, 'BAN082', 'BAN0119', 'BAN0115', 'BAN095', 'BAN0117']
            ind_vs_sl_df
            Player Name ['Hardik Pandya', 'Virat Kohli', 'Rishabh Pant', 'KL Rahul', 'Rohit Sharma'] Player ID ['INDO215', 'INDO175', 'INDO224', 'INDO213', 'INDO168']
            Match ID ['WC19M44']
            Pitch Type [nan]
            Bowler ['Isuru Udana', 'Lasith Malinga', 'Kusal Perera', 'Dhananjaya de Silva', 'Kasun Rajitha']
Player ID.1 ['SRI0152', 'SRI0123', 'SRI0155', 'SRI0169', 'SRI0146']
Over No [43.0, 42.0, 41.0, 40.0, 39.0, 38.0, 37.0, 36.0, 35.0, 34.0, 33.0, 32.0, 31.0, 30.0, 29.0, 28.0, 2
            7.0, 26.0, 25.0, 24.0, 23.0, 22.0, 21.0, 20.0, 19.0, 18.0, 17.0, 16.0, 15.0, 14.0, 13.0, 12.0, 11.0, 10.0,
            9.0, 8.0, 7.0, 6.0, 5.0, 4.0, 3.0, 2.0, 1.0, 0.0]
            Ball No [3.0, 2.0, 1.0, 6.0, 5.0, 4.0]
Ball type ['SL', 'G', 'L', 'FT', 'S', 'Y', 'F']
Type of Shot ['DR', 'P', 'ODR', 'LG', 'L', 'SW', 'C', 'F', 'D', '`F']
            Wide [nan, 'Yes']
            No ball [nan]
```

In [45]: matches_df

Out[45]:

	Batsman Name	Batsman Player ID	Match ID	Bowler Name	Bowler Player ID	Over No		Ball type	Type of Shot	Wide	No ball	Hit the Bat	Control	Effectiveness	Runs	Wicke
0	Hardik Pandya	INDO215	WC19M08	Andile Phehlukwayo	SAO118	47.0	3.0	F	С	0.0	0.0	1.0	0.0	0.5	4.0	0.
1	Hardik Pandya	INDO215	WC19M08	Andile Phehlukwayo	SAO118	47.0	3.0	S	L	1.0	0.0	0.0	1.0	0.0	1w	0.
2	Hardik Pandya	INDO215	WC19M08	Andile Phehlukwayo	SAO118	47.0	2.0	S	L	0.0	0.0	1.0	1.0	1.0	0.0	0.
3	Hardik Pandya	INDO215	WC19M08	Andile Phehlukwayo	SAO118	47.0	1.0	S	С	0.0	0.0	1.0	1.0	0.5	2.0	0.
4	Hardik Pandya	INDO215	WC19M08	Chris Morris	SAO110	46.0	6.0	S	DR	0.0	0.0	1.0	1.0	0.4	1.0	0.

2382	KL Rahul	INDO213	WC19M44	Lasith Malinga	SRIO123	0.0	5.0	L	L	0.0	0.0	1.0	1.0	1.0	0.0	0.
2383	KL Rahul	INDO213	WC19M44	Lasith Malinga	SRIO123	0.0	4.0	L	DR	0.0	0.0	1.0	1.0	0.2	0.0	0.
2384	KL Rahul	INDO213	WC19M44	Lasith Malinga	SRIO123	0.0	3.0	F	DR	0.0	0.0	1.0	1.0	0.2	0.0	0.
2385	Rohit Sharma	INDO168	WC19M44	Lasith Malinga	SRIO123	0.0	2.0	G	ODR	0.0	0.0	1.0	1.0	0.4	1.0	0.
2386	KL Rahul	INDO213	WC19M44	Lasith Malinga	SRIO123	0.0	1.0	L	ODR	0.0	0.0	1.0	1.0	0.4	1.0	0.

2387 rows × 18 columns

```
In [46]: for col_name, col_values in unique_values(matches_df):
                               print(col_name,col_values)
                     Batsman Name ['Hardik Pandya', 'Rohit Sharma', 'MS Dhoni', 'KL Rahul', 'Virat Kohli', 'Shikhar Dhawan', 'Kedar Jadhav', 'Vijay Shankar', 'Shikar Dhawan', 'Jasprit Bumrah', 'Kuldeep Yadav', 'Mohammed Shami', 'Kedhar Jadha v', 'Rishabh Pant', 'Bhuvneshwar Kumar', 'Dinesh Karthik']
Batsman Player ID ['IND0215', 'IND0168', 'IND0157', 'IND0213', 'IND0175', 'IND0188', 'IND0205', 'IND0226', 'IND0210', 'IND0219', 'IND0219', 'IND0194', 'IND0156']
Match ID ['WC19M08', 'WC19M22', 'WC19M14', 'WC19M28', 'WC19M34', 'WC19M38', 'WC19M40', 'WC19M44']
                     Bowler Name ['Andile Phehlukwayo', 'Chris Morris', 'Kagiso Rabada', 'Imran Tahir', 'Tabraiz Shamsi', 'Mohammad Amir', 'Wahab Riaz', 'Hasan Ali', 'Shadab Khan', 'Imad Wasim', 'Mohammad Hafeez', 'Shoaib Malik', 'Marcus Stoi
                      nis', 'Mitchell Starc', 'Pat Cummins', 'Adam Zampa', 'Nathan Coulter-Nile', 'Glenn Maxwell', 'Gulbadin Naib',
                     'Aftab Alam', 'Rashid Khan', 'Mujeeb Ur Rahman', 'Mohammad Nabi', 'Rahmat Shah', 'Oshane Thomas', 'Sheldon Cot trell', 'Carlos Brathwaite', 'Jason Holder', 'Fabian Allen', 'Kemar Roach', 'Chris Woakes', 'Jofra Archer', 'M ark Wood', 'Liam Plunkett', 'Adil Rashid', 'Ben Stokes', 'Mustafizur Rahman', 'Mohammad Saifuddin', 'Shakib Al Hasan', 'Rubel Hossain', 'Soumya Sarkar', 'Mosaddek Hossain', 'Mashrafe Mortaza', 'Isuru Udana', 'Lasith Malin ga', 'Kusal Perera', 'Dhananjaya de Silva', 'Kasun Rajitha']
                     ga', 'Kusal Perera', 'Dhananjaya de Silva', 'Kasun Rajitha']
Bowler Player ID ['SA0118', 'SA0110', 'SA0114', 'SA0192', 'SA0116', 'PAK0173', 'PAK0168', 'PAK0209', 'PAK021
1', 'PAK0204', 'PAK0144', 'PAK0128', 'AUS0209', 'AUS0185', 'AUS0189', 'AUS0212', 'AUS0204', 'AUS0196', 'AFG02
4', 'AFG017', 'AFG036', 'AFG043', 'AFG07', 'AFG029', 'WI0186', 'WI0169', 'WI0161', 'WI0166', 'WI0188', 'WI014
4', 'ENG0217', 'ENG0252', 'ENG0241', 'ENG0190', 'ENG0210', 'ENG0221', 'BAN0118', 'BAN0125', 'BAN082', 'BAN09
5', 'BAN0115', 'BAN0119', 'BAN054', 'SRI0152', 'SRI0123', 'SRI0155', 'SRI0169', 'SRI0146']
Over No [47.0, 46.0, 45.0, 44.0, 43.0, 42.0, 41.0, 40.0, 39.0, 38.0, 37.0, 36.0, 35.0, 34.0, 33.0, 32.0, 31.0,
                      30.0, 29.0, 28.0, 27.0, 26.0, 25.0, 24.0, 23.0, 22.0, 21.0, 20.0, 19.0, 18.0, 17.0, 16.0, 15.0, 14.0, 13.0, 1
                      2.0, 11.0, 10.0, 9.0, 8.0, 7.0, 6.0, 5.0, 4.0, 3.0, 2.0, 1.0, 0.0, 49.0, 48.0]
                      Ball No [3.0, 2.0, 1.0, 6.0, 5.0, 4.0]
                     Ball type ['F', 'S', 'L', 'FT', 'Y', 'GY', 'LB', 'G', 'SL']
Type of Shot ['C', 'L', 'DR', 'P', 'SW', 'ODR', 'D', 'LG', 'RS', 'F', 'SC', 'R']
                      Wide [0.0, 1.0]
                      No ball [0.0, 1.0]
                      Hit the Bat [1.0, 0.0]
                      Control [0.0, 1.0]
                      Effectiveness [0.5, 0.0, 1.0, 0.4, 0.2, -1.0, 0.6, 0.8, 0.3]
                      Runs [4.0, '1w', 0.0, 2.0, 1.0, 'W', 3.0, '1lb', 6.0, '1b', '4lb', '5nb']
                      Wicket [0.0, 1.0]
                     Dismissal kind [0, 'Ct', 'B', 'St', 'LBW', 'RO']
Caught/Bowled By [0, 'SA0110', 'SA0101', 'SA0105', 'PAK0159', 'PAK0203', 'PAK0168', 'AUS0189', 'AUS0209', 'AUS
0197', 'AUS0194', 'AUS0223', 'AFG09', 'AFG024', 'AFG045', 'AFG029', 'AFG034', 'AFG043', 'WI0175', 'WI0188', 'W
I0146', 'WI0166', 'ENG0239', 'ENG0217', 'ENG0226', 'BAN0118', 'BAN081', 'BAN082', 'BAN0119', 'BAN0115', 'BAN09
                      5', 'BAN0117', 'SRI0152', 'SRI0155', 'SRI0137']
In [47]: matches_df.nunique()
                                                                      14
                     Batsman Player ID
                      Match ID
                                                                       8
```

```
Out[47]: Batsman Name
         Bowler Name
                               48
         Bowler Player ID
                               48
         Over No
                               50
         Ball No
                               6
         Ball type
                               9
         Type of Shot
                               12
         Wide
         No ball
                               2
         Hit the Bat
                               2
         Control
                               2
         Effectiveness
                                9
         Runs
                               12
         Wicket
                               2
         Dismissal kind
                               6
         Caught/Bowled By
                               35
         dtype: int64
```

```
In [48]: player_info = dict(zip(matches_df['Batsman Name'],matches_df['Batsman Player ID']))
            player_info
'MS Dhoni': 'INDO157',
             'KL Rahul': 'INDO213'
             'Virat Kohli': 'INDO175'
             'Shikhar Dhawan': 'IND0188',
             'Kedar Jadhav': 'INDO205',
'Vijay Shankar': 'INDO226',
'Shikar Dhawan': 'IND0188',
             'Jasprit Bumrah': 'INDO210',
'Kuldeep Yadav': 'INDO217',
'Mohammed Shami': 'INDO195',
             'Kedhar Jadhav': 'INDO205', 'Rishabh Pant': 'INDO224',
             'Bhuvneshwar Kumar': 'INDO194',
             'Dinesh Karthik': 'INDO156'}
In [49]: len(player_info)
Out[49]: 16
In [50]: matches_df['Batsman_name'] = matches_df['Batsman Name'].replace('Kedar Jadhav','Kedhar Jadhav', inplace = True)
matches_df['Batsman_name'] = matches_df['Batsman Name'].replace('Shikar Dhawan','Shikhar Dhawan', inplace = Tru
In [51]: matches_df.nunique()
Out[51]: Batsman Name
                                     14
            Batsman Player ID
                                     14
            Match ID
                                      8
            Bowler Name
                                      48
            Bowler Player ID
                                     48
            Over No
                                      50
            Ball No
                                      6
            Ball type
                                      9
            Type of Shot
                                     12
            Wide
                                       2
            No ball
            Hit the Bat
                                       2
            Control
                                       2
            Effectiveness
                                       9
            Runs
                                     12
            Wicket
                                       2
            Dismissal kind
                                       6
            Caught/Bowled By
                                     35
            Batsman_name
                                       0
            dtype: int64
```

Out[52]:

	Analysis Parameters	Formulae	Parameters	Unnamed: 3	Unnamed: 4
0	Best off side player	Highest average of effectiveness of shot of all off side shots	Dr,C and RS	The highest percentage of shots in Dr, C and RS	O= ((no of Dr+C+RS)/ total number of balls faced) * ((sum of effectiveness of Dr +C+RS)/no of Dr+C+RS)
1	Best Leg side player	Highest average of effectiveness of shot of all Leg side shots	P, ODR, LG, F and S	The highest percentage of shots in P, ODR, LG, F and S	O= ((no of P+ODR+LG+F+S)/ total number of balls faced) * ((sum of effectiveness of P+ODR+LG+F+S)/no of P+ODR+LG+F+S)
2	Best player for bouncy tracks	Highest average of effectiveness of shot of all Bouncy pitch shots	D, P, C and L	The highest percentage of shots in D, P, C and L	O= ((no of D+P+C+L)/ total number of balls faced) * ((sum of effectiveness of D+P+C+L)/no of D+P+C+L)
3	Best player for high scoring matches	Highest average of effectiveness of shot of all agressive shots	D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	The highest percentage of shots in D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	O= ((no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/ total number of balls faced) * ((sum of effectiveness of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)
4	Best defencive player	NaN	D and L	The highest percentage of shots in D and L	O= ((no of D+L)/ total number of balls faced) * ((sum of effectiveness of D+L)/no of D+L)
5	Best spin ball player	Highest average of effectiveness of shot of all spin ball shots	D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	The highest percentage of shots in D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	O= ((no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/ total number of balls faced) * ((sum of effectiveness of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)
6	The player with the best control	Highest control of shots	D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	The highest percentage of shots in D, Dr, C, P, Sw, RS, L, ODR, F, R, LG and SC	O= ((no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/ total number of balls faced) * ((sum of effectiveness of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)/no of D+Dr+ C+P+Sw+RS+L+ODR+F+ R,+LG +SC)

```
In [53]: def off_side_score(df = matches_df):
             off_side_shots = ['DR','C','RS']
             off_side_shots_count = df[df['Type of Shot'].isin(off_side_shots)].groupby('Batsman Name').agg(off_side_sho
             grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))
             off_side_score_df = off_side_shots_count.merge(grouped_df,on = 'Batsman Name')
             off_side_score_df = off_side_score_df.reset_index()
             off_side_score_df['Highest percentage of off side shots'] = (off_side_score_df['off_side_shots_count']/off_
             off_side_score_df['Highest effectiveness average of off side shots'] = off_side_score_df['sum_effectiveness
             off_side_score_df['off_side_score'] = (off_side_score_df['Highest percentage of off side shots']/100) * off
             off_side_score_df = off_side_score_df.sort_values(by=['off_side_score'], ascending=False)
             return off_side_score_df.loc[:,['Batsman Name','Highest percentage of off side shots','Highest effectivenes
```

In [54]: off_side_score()

Out[54]:

	Batsman Name	Highest percentage of off side shots	Highest effectiveness average of off side shots	off_side_score
0	Bhuvneshwar Kumar	50.000000	0.400000	0.200000
2	Hardik Pandya	42.465753	0.422581	0.179452
8	Shikhar Dhawan	43.089431	0.383019	0.165041
7	Rohit Sharma	42.057489	0.378417	0.159153
10	Virat Kohli	41.880342	0.320408	0.134188
6	Rishabh Pant	31.168831	0.404167	0.125974
3	KL Rahul	40.086207	0.280108	0.112284
5	MS Dhoni	35.245902	0.306977	0.108197
1	Dinesh Karthik	44.444444	0.200000	0.088889
4	Kedhar Jadhav	26.923077	0.321429	0.086538
9	Vijay Shankar	30.769231	0.266667	0.082051

```
In [55]: def leg_side_score(df = matches_df):
                          leg_side_shots = ['P','ODR','LG','F','S']
                          leg_side_shots_count = df[df['Type of Shot'].isin(leg_side_shots)].groupby('Batsman Name').agg(leg_side_shots)
                          grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))
                          leg_side_score_df = leg_side_shots_count.merge(grouped_df,on = 'Batsman Name')
                          leg_side_score_df = leg_side_score_df.reset_index()
                          leg\_side\_score\_df['Highest\ percentage\ of\ leg\ side\ shots']\ =\ (leg\_side\_score\_df['leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_side\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg_shots\_count']/leg\_shots\_count']/leg\_shots\_count']/leg\_shots\_cou
                          leg_side_score_df['Highest effectiveness average of leg side shots'] = leg_side_score_df['sum_effectiveness
                          leg_side_score_df['leg_side_score'] = (leg_side_score_df['Highest percentage of leg side shots']/100) * leg
                          leg_side_score_df = leg_side_score_df.sort_values(by=['leg_side_score'], ascending=False)
                          return leg_side_score_df.loc[:,['Batsman Name','Highest percentage of leg side shots','Highest effectivenes
In [56]: leg_side_score()
Out[56]:
                                Batsman Name Highest percentage of leg side shots Highest effectiveness average of leg side shots leg_side_score
                     3
                                  Jasprit Bumrah
                                                                                               100.000000
                                                                                                                                                                         0.400000
                                                                                                                                                                                                 0.400000
                                                                                               100.000000
                                                                                                                                                                         0.300000
                                                                                                                                                                                                 0.300000
                     6
                                  Kuldeep Yadav
                                                                                                                                                                         0.400000
                                    Rishabh Pant
                                                                                                55.844156
                                                                                                                                                                                                 0.223377
                                         MS Dhoni
                                                                                                43.852459
                                                                                                                                                                         0.412150
                                                                                                                                                                                                 0.180738
                                                                                                44.520548
                                                                                                                                                                         0.372308
                                                                                                                                                                                                 0.165753
                                   Hardik Pandva
                     5
                                                                                                49.038462
                                                                                                                                                                         0.323529
                                                                                                                                                                                                 0.158654
                                  Kedhar Jadhav
                                                                                                                                                                         0.428571
                    12
                                    Viiav Shankar
                                                                                                35.897436
                                                                                                                                                                                                 0.153846
                                                                                                39.743590
                    13
                                        Virat Kohli
                                                                                                                                                                        0.371505
                                                                                                                                                                                                 0.147650
                                          KL Rahul
                                                                                                35.775862
                                                                                                                                                                         0.398193
                                                                                                                                                                                                 0.142457
                    10
                                    Rohit Sharma
                                                                                                32.829047
                                                                                                                                                                         0.393548
                                                                                                                                                                                                 0.129198
                                   Dinesh Karthik
                                                                                                55.55556
                                                                                                                                                                         0.200000
                                                                                                                                                                                                  0.111111
                                Shikhar Dhawan
                                                                                                25.203252
                                                                                                                                                                         0.332258
                                                                                                                                                                                                 0.083740
                             Mohammed Shami
                                                                                                83.333333
                                                                                                                                                                        -0.200000
                                                                                                                                                                                                -0.166667
                     0 Bhuvneshwar Kumar
                                                                                                25.000000
                                                                                                                                                                        -1.000000
                                                                                                                                                                                                -0.250000
In [57]: def bouncy track score(df = matches df):
                          bouncy_track_shots = ['D','P','C','L']
                          bouncy_track_shots_count = df[df['Type of Shot'].isin(bouncy_track_shots)].groupby('Batsman Name').agg(boun
                          grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))
                          bouncy_track_score_df = bouncy_track_shots_count.merge(grouped_df,on = 'Batsman Name')
                          bouncy_track_score_df = bouncy_track_score_df.reset_index()
                          bouncy_track_score_df['Highest percentage of bouncy track shots'] = (bouncy_track_score_df['bouncy_track_sh
                          bouncy_track_score_df['Highest effectiveness average of bouncy track shots'] = bouncy_track_score_df['sum_e'
                          bouncy_track_score_df['bouncy_track_score'] = (bouncy_track_score_df['Highest percentage of bouncy track sh
                          bouncy_track_score_df = bouncy_track_score_df.sort_values(by=['bouncy_track_score'], ascending=False)
                          return bouncy_track_score_df.loc[:,['Batsman Name','Highest percentage of bouncy track shots','Highest effe
```

```
In [58]: bouncy_track_score()
Out[58]:
                                       Batsman Name Highest percentage of bouncy track shots Highest effectiveness average of bouncy track shots bouncy_track_score
                          9
                                       Shikhar Dhawan
                                                                                                                             50.406504
                                                                                                                                                                                                                                                                      0.313821
                                                                                                                                                                                                                               0.622581
                          8
                                           Rohit Sharma
                                                                                                                             46.898638
                                                                                                                                                                                                                               0.658065
                                                                                                                                                                                                                                                                      0.308623
                        10
                                            Vijay Shankar
                                                                                                                             42.307692
                                                                                                                                                                                                                               0.715152
                                                                                                                                                                                                                                                                      0.302564
                          3
                                                   KL Rahul
                                                                                                                             40 732759
                                                                                                                                                                                                                               0.669841
                                                                                                                                                                                                                                                                      0.272845
                          5
                                                  MS Dhoni
                                                                                                                             41.803279
                                                                                                                                                                                                                               0.592157
                                                                                                                                                                                                                                                                      0.247541
                                         Kedhar Jadhav
                                                                                                                             39.423077
                                                                                                                                                                                                                               0.600000
                                                                                                                                                                                                                                                                      0.236538
                                                 Virat Kohli
                                                                                                                             37.820513
                                                                                                                                                                                                                               0.546893
                                                                                                                                                                                                                                                                      0.206838
                          2
                                          Hardik Pandya
                                                                                                                             36.986301
                                                                                                                                                                                                                               0.490741
                                                                                                                                                                                                                                                                      0.181507
                                            Rishabh Pant
                                                                                                                             25.974026
                                                                                                                                                                                                                               0.350000
                                                                                                                                                                                                                                                                      0.090909
                               Bhuvneshwar Kumar
                                                                                                                             50 000000
                                                                                                                                                                                                                               0.000000
                                                                                                                                                                                                                                                                      0.000000
                          0
                                   Mohammed Shami
                                                                                                                             50 000000
                                                                                                                                                                                                                              -0.066667
                                                                                                                                                                                                                                                                     -0.0333333
                                          Dinesh Karthik
                                                                                                                             33 333333
                                                                                                                                                                                                                                                                     -0.066667
                                                                                                                                                                                                                              -0.200000
In [59]: def aggressive_shots_score(df = matches_df):
                               aggressive_shots = ['Dr', 'C', 'P', 'Sw', 'RS', 'ODR', 'F', 'R', 'LG', 'SC']
                               aggressive\_shots\_count = df[df['Type \ of \ Shot'].isin(aggressive\_shots)].groupby('Batsman \ Name').agg(aggressive\_shots)].groupby('Batsman \ Name').agg(aggressive\_shots)].groupby('
                               grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))
                               aggressive_score_df = aggressive_shots_count.merge(grouped_df,on = 'Batsman Name')
                               aggressive_score_df = aggressive_score_df.reset_index()
                               aggressive_score_df['Highest percentage of aggressive shots'] = (aggressive_score_df['aggressive_shots_coun
                               aggressive_score_df['Highest effectiveness average of aggressive shots'] = aggressive_score_df['sum_effecti
                               aggressive\_score\_df['aggressive\_score'] = (aggressive\_score\_df['Highest percentage \ of \ aggressive \ shots']/10
                               aggressive_score_df = aggressive_score_df.sort_values(by=['aggressive_score'], ascending=False)
                               return aggressive_score_df.loc[:,['Batsman Name','Highest percentage of aggressive shots','Highest effectiv
In [60]: aggressive_shots_score()
Out[60]:
```

	Batsman Name	Highest percentage of aggressive shots	Highest effectiveness average of aggressive shots	aggressive_score
3	Jasprit Bumrah	100.000000	0.400000	0.400000
6	Kuldeep Yadav	100.000000	0.300000	0.300000
9	Rishabh Pant	59.740260	0.395652	0.236364
2	Hardik Pandya	57.534247	0.388095	0.223288
7	MS Dhoni	51.229508	0.385600	0.197541
10	Rohit Sharma	47.049924	0.398071	0.187292
5	Kedhar Jadhav	54.807692	0.328070	0.179808
13	Virat Kohli	50.427350	0.346610	0.174786
4	KL Rahul	44.612069	0.363768	0.162284
12	Vijay Shankar	43.589744	0.358824	0.156410
1	Dinesh Karthik	77.777778	0.200000	0.155556
11	Shikhar Dhawan	46.341463	0.315789	0.146341
0	Bhuvneshwar Kumar	25.000000	-1.000000	-0.250000
8	Mohammed Shami	100.000000	-0.333333	-0.333333

```
In [61]: def defense_score(df = matches_df):
              defense_shots = ['D','L']
              defense_shots_count = df[df['Type of Shot'].isin(defense_shots)].groupby('Batsman Name').agg(defense_shots_
              grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))
              defense_score_df = defense_shots_count.merge(grouped_df,on = 'Batsman Name')
              defense_score_df = defense_score_df.reset_index()
              defense_score_df['Highest percentage of defense shots'] = (defense_score_df['defense_shots_count']/defense_
              defense_score_df['Highest effectiveness average of defense shots'] = defense_score_df['sum_effectiveness']/
              defense_score_df['defense_score'] = (defense_score_df['Highest percentage of defense shots']/100) * defense
              defense_score_df = defense_score_df.sort_values(by=['defense_score'], ascending=False)
              return defense_score_df.loc[:,['Batsman Name','Highest percentage of defense shots','Highest effectiveness
In [62]: defense_score()
Out[62]:
                 Batsman Name Highest percentage of defense shots Highest effectiveness average of defense shots defense score
                   Vijay Shankar
                                                    30.769231
                                                                                             0.891667
                                                                                                           0.274359
           0
             Bhuvneshwar Kumar
                                                    25.000000
                                                                                             1.000000
                                                                                                           0.250000
                                                     23.577236
                                                                                             0.924138
                                                                                                           0.217886
                 Shikhar Dhawan
           2
                      KL Rahul
                                                    21.982759
                                                                                             0.935294
                                                                                                           0.205603
           6
                   Rohit Sharma
                                                    20.877458
                                                                                             0.931884
                                                                                                           0.194554
           3
                  Kedhar Jadhav
                                                    20.192308
                                                                                             0.952381
                                                                                                           0.192308
                      MS Dhoni
                                                                                             0.782222
                                                                                                           0.144262
                                                     18.442623
           9
                     Virat Kohli
                                                    16.239316
                                                                                             0.886842
                                                                                                          0.144017
           1
                  Hardik Pandya
                                                     8.904110
                                                                                             0.707692
                                                                                                           0.063014
           5
                   Rishabh Pant
                                                     2.597403
                                                                                             0.000000
                                                                                                           0.000000
In [63]: def spin_ball_score(df = matches_df):
              spin_ball_shots = ['D', 'Dr', 'C', 'P', 'Sw', 'RS', 'ODR', 'F', 'R', 'LG', 'SC']
              spin_ball = ['GY', 'LB']
              spin_ball_shots_count = df[df['Type of Shot'].isin(spin_ball_shots) & df['Ball type'].isin(spin_ball)].grou
              grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))
              spin ball score df = spin ball shots count.merge(grouped df,on = 'Batsman Name')
              spin_ball_score_df = spin_ball_score_df.reset_index()
              spin_ball_score_df['Highest percentage of spin ball shots'] = (spin_ball_score_df['spin_ball_shots_count']/
              spin_ball_score_df['Highest effectiveness average of spin ball shots'] = spin_ball_score_df['sum_effectiven
              spin_ball_score_df['spin_ball_score'] = (spin_ball_score_df['Highest percentage of spin ball shots']/100) *
              spin_ball_score_df = spin_ball_score_df.sort_values(by=['spin_ball_score'], ascending=False)
              return spin_ball_score_df.loc[:,['Batsman Name','Highest percentage of spin ball shots','Highest effectiven
In [64]: | spin_ball_score()
Out[64]:
             Batsman Name Highest percentage of spin ball shots Highest effectiveness average of spin ball shots spin ball score
               Vijay Shankar
                                                  10.256410
                                                                                          0.762500
                                                                                                         0.078205
              Kedhar Jadhav
                                                  14.423077
                                                                                          0.520000
                                                                                                         0.075000
                  MS Dhoni
                                                  4.508197
                                                                                          0.345455
                                                                                                         0.015574
           2
                  Virat Kohli
                                                   1.068376
                                                                                          0.840000
                                                                                                         0.008974
               Hardik Pandya
                                                  0.684932
                                                                                          1.000000
                                                                                                         0.006849
                                                  0.151286
                                                                                          0.400000
                                                                                                         0.000605
               Rohit Sharma
```

```
In [65]:
    def control_score(df = matches_df):
        shots = ['D', 'Dr', 'C', 'P', 'Sw', 'RS', 'ODR', 'F', 'R', 'LG', 'SC', 'L']

        shot_count_df = df[df['Control'] == 1.0].groupby('Batsman Name').agg(controlled_shots_count = ('Type of Shot grouped_df = df.groupby('Batsman Name').agg(total_no_of_balls_faced = ('Type of Shot', 'count'))

        control_score_df = shot_count_df.merge(grouped_df, on ='Batsman Name')
        control_score_df = control_score_df.reset_index()

        control_score_df['Highest percentage of controlled shots'] = (control_score_df['controlled_shots_count']/control_score_df['Highest effectiveness average of controlled shots'] = control_score_df['sum_effectiveness control_score_df['control_score_df['Highest percentage of controlled shots']/100) * control_score_df = control_score_df.sort_values(by=['control_score'], ascending=False)

        return control_score_df.loc[:,['Batsman Name','Highest percentage of controlled shots','Highest effectivene
```

In [66]: control_score()

Out[66]:

	Batsman Name	Highest percentage of controlled shots	Highest effectiveness average of controlled shots	control_score
10	Rohit Sharma	94.553707	0.506560	0.478971
11	Shikhar Dhawan	86.991870	0.539252	0.469106
4	KL Rahul	90.086207	0.464115	0.418103
13	Virat Kohli	92.307692	0.445602	0.411325
3	Jasprit Bumrah	100.000000	0.400000	0.400000
2	Hardik Pandya	89.726027	0.438168	0.393151
7	MS Dhoni	81.967213	0.460500	0.377459
12	Vijay Shankar	71.794872	0.485714	0.348718
9	Rishabh Pant	90.909091	0.370000	0.336364
6	Kuldeep Yadav	100.000000	0.300000	0.300000
5	Kedhar Jadhav	65.384615	0.398529	0.260577