



# Hi, I'm Ashish Mishra



## Internship in Data Science



### at ShadowFox

## Cricket Fielding Analysis

```
In [1]: import pandas as pd
```

```
In [2]: df = pd.read_excel(r'C:\Users\Ashish Mishra\OneDrive\Desktop\ShadowFox\Shadowfox_DS
```

```
In [3]: df.head()
```

Out[3]:

	Pick	Y->	Clean Pick	N->	Fumble	C->	Catch	DC- >	Dropped Catch	S->	Stu
0	Throw	Y->	Good Throw	N->	Bad throw	DH->	Dirct Hit	RO- >	Run Out	MR- >	
1	Runs	"+" stands for runs saved "_" stands for runs ...	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
3	NaN	Match No.	Innings	Teams	Player Name	BallCount	Position	Pick	Throw	Runs	Ov
4	NaN	IPL2367	1	Delhi Capitals	Rilee russouw	0.1	Short mid wicket	n	NaN	1	

```
In [4]: # Reload the dataset with a specific skiprows parameter to better clean and structu
data_refined = pd.read_excel(r"C:\Users\Ashish Mishra\OneDrive\Desktop\ShadowFox\Sh

# Rename columns based on their actual meaning and drop any unnecessary columns
data_refined.columns = [
    'Pick',
    'Match_No',
    'Innings',
    'Team',
    'Player_Name',
```

```
'BallCount',
'Position',
'Pick_Type',
'Throw_Type',
'Runs',
'Overcount',
'Venue',
'Stadium'
]

# Convert numeric fields (e.g., Runs, BallCount) to appropriate data types
data_refined['Runs'] = pd.to_numeric(data_refined['Runs'], errors='coerce').fillna(
data_refined['BallCount'] = pd.to_numeric(data_refined['BallCount'], errors='coerce

# Display the cleaned and structured dataset
data_refined.head()
```

Out[4]:

	Pick	Match_No	Innings	Team	Player_Name	BallCount	Position	Pick_Type	Throw
0	NaN	Match No.	Innings	Teams	Player Name	NaN	Position	Pick	
1	NaN	IPL2367	1	Delhi Capitals	Rilee russouw	0.1	Short mid wicket	n	
2	NaN	IPL2367	1	Delhi Capitals	Phil Salt	0.2	wicket keeper	Y	
3	NaN	IPL2367	1	Delhi Capitals	Yash Dhull	0.3	covers	Y	
4	NaN	IPL2367	1	Delhi Capitals	Axer Patel	0.4	point	Y	

In [5]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 74 entries, 0 to 73
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Pick                  2 non-null     object
1   Y->                   31 non-null    object
2   Clean Pick            54 non-null    object
3   N->                   22 non-null    object
4   Fumble                19 non-null    object
5   C->                   22 non-null    object
6   Catch                 19 non-null    object
7   DC->                  19 non-null    object
8   Dropped Catch         16 non-null    object
9   S->                   11 non-null    object
10  Stumping              22 non-null    object
11  Unnamed: 11           21 non-null    object
12  Unnamed: 12           13 non-null    object
dtypes: object(13)
memory usage: 7.6+ KB
```

```
In [6]: # Drop rows where essential columns like 'Player_Name' or 'Position' are NaN
data_refined.dropna(subset=['Player_Name', 'Position'], inplace=True)

# Convert numeric fields (e.g., Runs, BallCount) to appropriate data types
data_refined['Runs'] = pd.to_numeric(data_refined['Runs'], errors='coerce').fillna(
data_refined['BallCount'] = pd.to_numeric(data_refined['BallCount'], errors='coerce')
```

```
In [7]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 74 entries, 0 to 73
Data columns (total 13 columns):
#   Column                Non-Null Count  Dtype
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4   Fumble                19 non-null    object
5   C->                   22 non-null    object
6   Catch                 19 non-null    object
7   DC->                  19 non-null    object
8   Dropped Catch         16 non-null    object
9   S->                   11 non-null    object
10  Stumping              22 non-null    object
11  Unnamed: 11           21 non-null    object
12  Unnamed: 12           13 non-null    object
dtypes: object(13)
memory usage: 7.6+ KB
```

```
In [8]: weights = {
    'CP': 1.5, # Clean Picks
    'GT': 1.2, # Good Throws
    'C': 2.0, # Catches
    'DC': -1.0, # Dropped Catches
```

```

'ST': 2.5, # Stumpings
'RO': 3.0, # Run Outs
'MRO': -0.5, # Missed Run Outs
'DH': 2.0, # Direct Hits
'RS': 1.0 # Runs Saved
}

```

```

In [9]: # Initialize a performance score column
data_refined['Performance Score'] = 0

# Calculate the performance score for each player
data_refined['Performance Score'] = (
    data_refined['Pick_Type'].apply(lambda x: weights['CP'] if x == 'Y' else 0) +
    data_refined['Throw_Type'].apply(lambda x: weights['GT'] if x == 'Y' else 0) +
    data_refined['Runs'] * weights['RS']
)

# Group by player and calculate total performance scores
top_players = data_refined[['Player_Name', 'Performance Score', 'Match_No', 'Inning
top_players_grouped = top_players.groupby('Player_Name').agg({
    'Performance Score': 'sum',
    'Match_No': 'first',
    'Innings': 'first',
    'Team': 'first',
    'Position': 'first',
    'Venue': 'first'
}).sort_values(by='Performance Score', ascending=False)

# Display top performers
top_3_players = top_players_grouped.head(3)
print("Top 3 Performers with Details:\n", top_3_players)

```

Top 3 Performers with Details:

Player_Name	Performance Score	Match_No	Innings	Team \
Kuldeep yadav	4.2	IPL2367	1	Delhi Capitals
Lalit yadav	4.2	IPL2367	1	Delhi Capitals
1	3.0	Rilee russouw	2	1

Player_Name	Position	Venue
Kuldeep yadav	Short mid wicket	Delhi
Lalit yadav	cover point	Delhi
1	0	10

```

In [10]: import seaborn as sns
import matplotlib.pyplot as plt

```

```

In [11]: sns.barplot(x=top_3_players.index, y=top_3_players['Performance Score'], hue=top_3_
plt.title('Top 3 Players by Fielding Performance')
plt.xlabel('Player Name')
plt.ylabel('Performance Score')
plt.xticks(rotation=45)
plt.show()

```

