



Introduction

Clean and Process Data

Explore the Data

Transfer Strategy

Data Visualisation

Finalise Team & Budget

Conclusion

CHELSEA '24

40.9%

10TH

WINNING PERCENT

RANK IN LEAGUE

\$ 1 BN+

MONEY SPENT



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PRE-PROCESSING

- → Fetch data from understat API for different leagues
- → Combine relevant data
- → Fix header formats and issues
- → Remove columns and attributes of no use
- **→** Ensure no missing values
- → Calculated field from other data tables and included in main dataset



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EXPLORE THE DATA

- → Figured out variables we would need for statistics
- → df.info()

→ df.describe()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2946 entries, 0 to 2945
Data columns (total 18 columns):
Column Non-Null Count Dty

Data	columns (total	l 18 columns):	
#	Column	Non-Null Count	Dtype
0	id	2946 non-null	int64
1	player_name	2946 non-null	object
2	games	2946 non-null	int64
	time	2946 non-null	int64
4	goals	2946 non-null	int64
	xG	2946 non-null	float64
6	assists	2946 non-null	int64
7	xA	2946 non-null	float64
8	shots	2946 non-null	int64
9	key_passes	2946 non-null	int64
10	yellow_cards	2946 non-null	int64
11	red_cards	2946 non-null	int64
12	position	2946 non-null	object
13	team_title	2946 non-null	object
14	npg	2946 non-null	int64
15	npxG	2946 non-null	float64
16	xGChain	2946 non-null	float64
17	xGBuildup	2946 non-null	float64

<u> </u>		id	games	time	goals	хG
	count	2946.000000	2946.000000	2946.000000	2946.000000	2946.000000
	mean	7453.357434	12.124576	778.911745	1.069246	1.188806
	std	3726.036579	6.558999	585.076435	2.037332	1.939957
	min	3.000000	1.000000	1.000000	0.000000	0.000000
	25%	5058.750000	6.000000	214.250000	0.000000	0.061358
	50%	8080.500000	13.000000	747.000000	0.000000	0.443733
	75%	10769.500000	18.000000	1269.000000	1.000000	1.445090
	max	12413.000000	23.000000	2070.000000	24.000000	18.847552
		assists	хA	shots	key_passes	yellow_cards
	count	2946.000000	2946.000000	2946.000000	2946.000000	2946.000000
	mean	0.753564	0.830263	10.114732	7.495927	1.657162
	std	1.320221	1.178907	12.190358	9.332392	1.781411
	min	0.000000	0.000000	0.000000	0.000000	0.000000
	25%	0.000000	0.039970	1.000000	1.000000	0.000000
	50%	0.000000	0.379686	6.000000	4.000000	1.000000
	75%	1.000000	1.130833	14.000000	11.000000	3.000000
	max	11.000000	10.750426	94.000000	66.000000	13.000000
		red cards	npg	npxG	xGChain	xGBuildup
	count	2946.000000	2946.000000	2946.000000	2946.000000	2946.000000
	mean	0.078072	0.963001	1.086612	3.357657	2.051938
	std	0.287866	1.805760	1.723559	3.505017	2.307226
	min	0.000000	0.000000	0.000000	0.000000	0.000000
	25%	0.000000	0.000000	0.060115	0.705534	0.372605
	50%	0.000000	0.000000	0.433047	2.366352	1.380206
	75%	0.000000	1.000000	1.352715	4.786649	2.887566
	max	2.000000	21.000000	16.574221	25.598131	17.123714



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TRANSFER STRATEGY

- 1) Chelsea vs Table Toppers Comparision
- 2) Chelsea Past vs Present Comparision

Identify Chelsea's Weakness

- 3) Comparison of Chelsea's Players vs Other team's players
- 4) Comparison of Chelsea's Players within themselves

Decide Players to SELL

3) Optimise best players wrt budget

Decide Players to BUY







SOLD



INJURED



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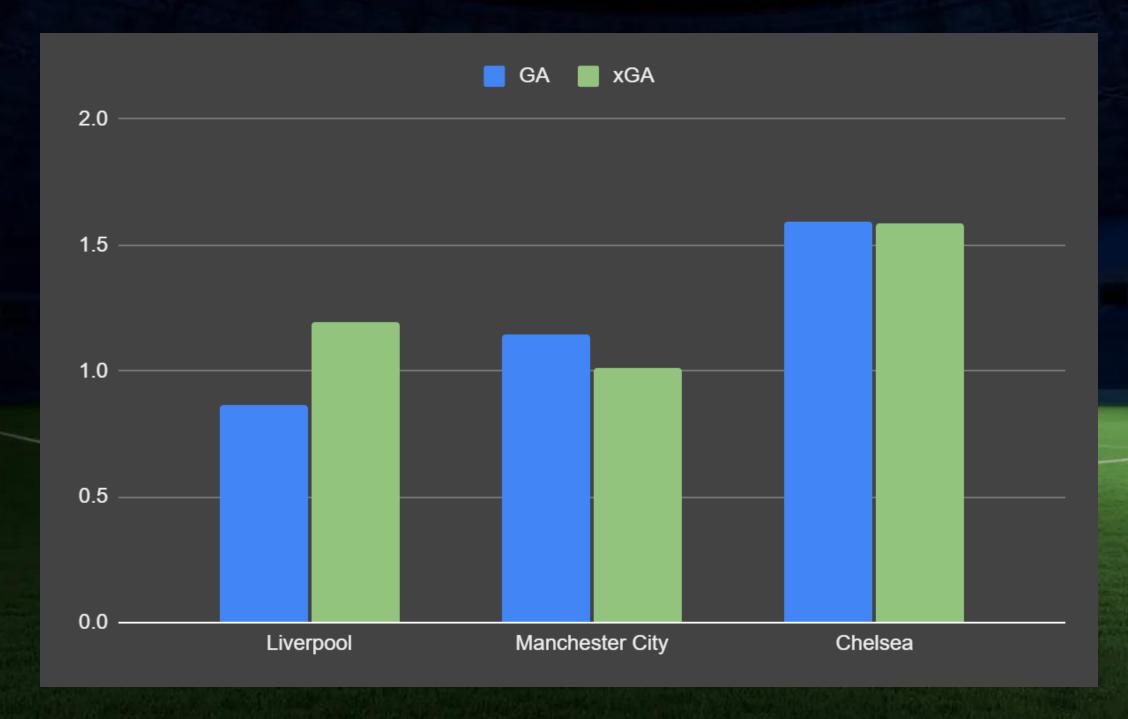
Data Visualisation

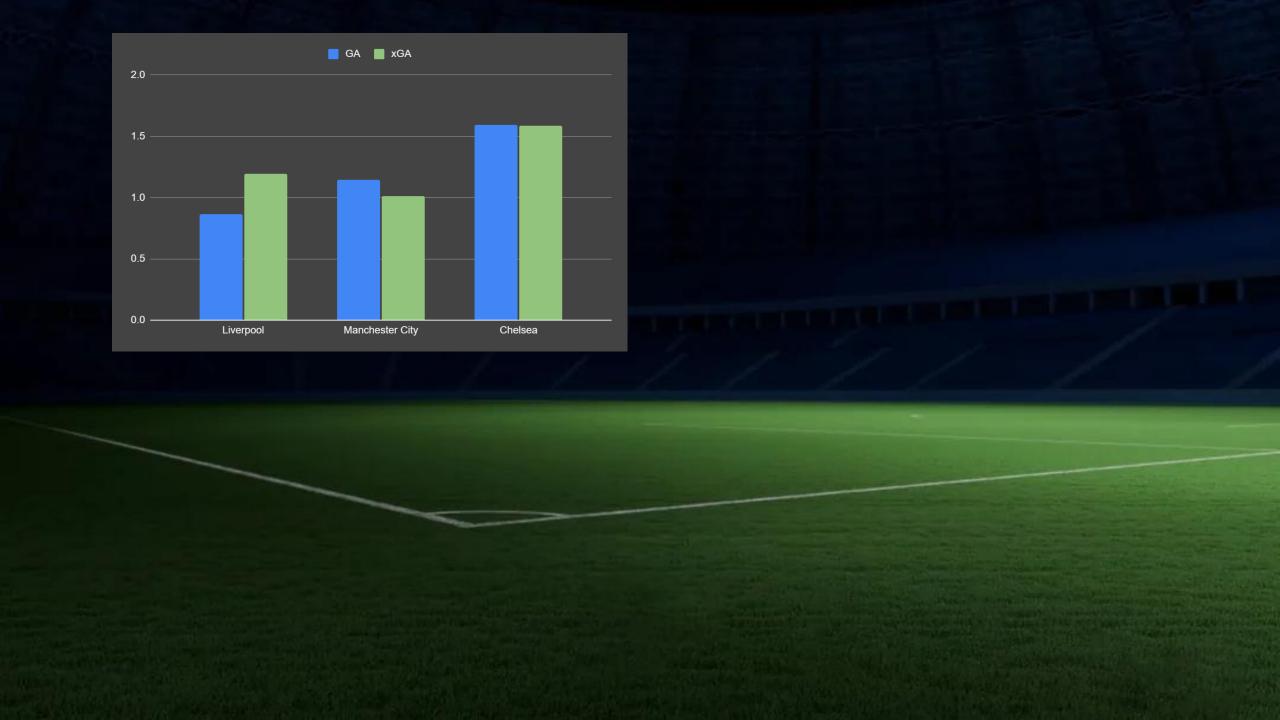
Finalise Team & Budget

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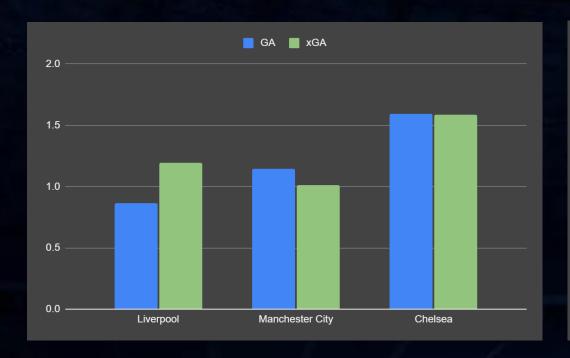
DATA VISUALISATION















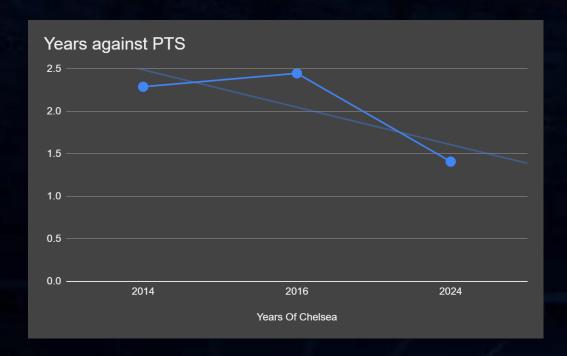


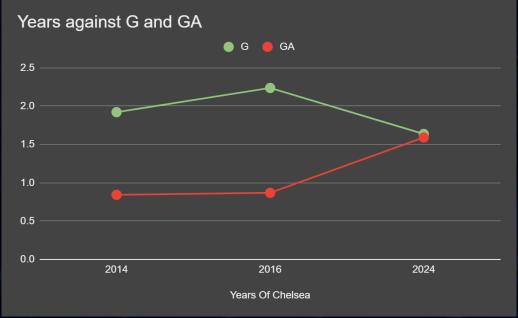


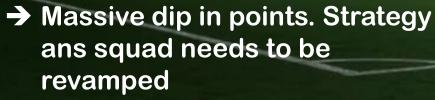
- → An attacker outperforming their XG is needed.
- → GA and xGA extremely high in comparison. Defense upgrade massively needed. A major shot stopper or blocker, a GK or a DEF.
- → Lesser Shots taken by the team.

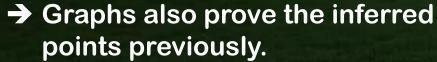
 Creative mid with more key passes required.
- → NPXg<<Xg. This means penalties are getting converted and a penalty specialist is not a necessity

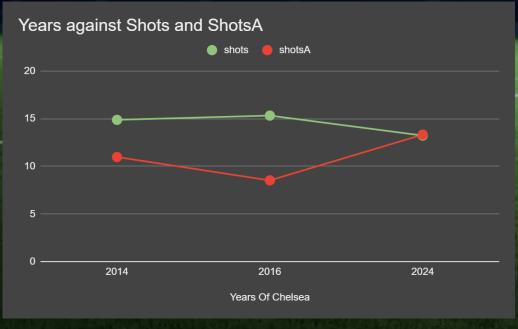












	В	С	D	Е	F	G	Н
1	player_nam∈ =	games =	time =	goals =	xG −	assists =	xA −
2	Thiago Silva	21	1829	1	0.2096096575	0	0.7779160039
3	Conor Gallagher	21	1758	0	1.670500018	4	3.112052305
4	Axel Disasi	20	1780	1	0.9299943522	0	0.1573617086
5	Levi Colwill	20	1616	1	0.7760256305	1	1.237992246
6	Raheem Sterling	20	1613	5	5.865220726	3	3.692453085
7	Moisés Caicedo	19	1526	0	0.1482277075	0	0.8831624333
8	Enzo Fernández	19	1466	2	4.832361309	0	2.14346958
9	Nicolas Jackson	19	1458	7	11.18073603	2	2.957573416
10	Mykhailo Mudrył	18	826	3	2.585933138	1	1.404977664
11	Cole Palmer	18	1320	9	7.731454914	4	5.087213239
12	Robert Sánchez	16	1438	0	0	0	0
13	Malo Gusto	15	921	0	0.6059665196	3	1.985742092
14	Armando Broja	13	448	1	2.489345549	0	0.1379258521
15	Ian Maatsen	12	169	0	0.6834889837	0	0.05732445326
16	Noni Madueke	11	286	2	1.264032314	0	0.3727686107
17	Chimuanya Ugo	10	266	0	0	0	0.05015848577
18	Marc Cucurella	9	708	0	0.9692549165	0	0.2287536319
19	Ben Chilwell	8	411	0	0.959005028	1	1.341160949
20	Reece James	8	404	0	0.3389957156	1	2.19567249
21	Benoit Badiashile	7	549	0	0.3124654777	1	0.5357111124
22	Djordje Petrovic	7	542	0	0	0	0
23	Christopher Nku	4	167	2	0.9201845936	0	0.07201976329

THROUGH COLAB

```
print("top5 strikers as replacement for Jackson:\n",top 5 players)
    loop = asyncio.get event loop()
    loop.run until complete(main())
top5 strikers as replacement for Jackson:
            id
                         player_name games time goals
                                                               xG assists \
          647
                         Harry Kane
                                       20 1780
                                                    24 18.847552
          3423 Kylian Mbappe-Lottin
                                       19 1596
                                                    20 17.824278
    1864 7006
                   Lautaro Martínez
                                       19 1611
                                                    19 13.635117
    1406 3738
                    Serhou Guirassy
                                       14 1056
                                                   17 12.764916
          9988
                                       16 1316
                                                    14 8.516328
                    Mateo Cassierra
               xA shots key passes
                                     yellow cards red cards position \
    1405 5.078877
                                  21
          3.949947
                                                               F M S
    1864 1.521982
                                 16
                                                                FS
         3.850828
                                 24
                                                                 FS
          1.276945
                                                               FMS
                   team title npg
                                        npxG
                                                xGChain xGBuildup
    1405
                Bayern Munich
                              21 16.574221 23.444284
                                                         6.288854
           Paris Saint Germain 15 12.503612 22.080693
    400
                                                         7.934952
    1864
                              17 12.112519 18.748646
                                                         7.593619
                        Inter
    1406
                VfB Stuttgart 14 10.491586 16.754187
                                                         4.090644
          Zenit St. Petersburg 14 8.516328 11.992199
                                                         2.883037
```

top 5 players = filtered df.sort values(by='goals', ascending=False).head(5)

filtered df = df[df['goals'] > df['xG']]

```
filtered_df2 = df[df['position']=='M S']
        # Sorting the filtered DataFrame by 'key passes' column in descending order and selecting 1
        top 5 players = filtered df2.sort values(by='key passes', ascending=False).head(5)
        print("Most key passes Midfielder:\n", top 5 players)
loop = asyncio.get_event_loop()
loop.run_until_complete(main())
Most key passes Midfielder:
                player name games time goals
                                                      xG assists
                                                                         xA \
      314 Ilkay Gündogan
                               22 1849
                                             4 4.988644
                                                               4 6.249550
      922 Andreas Pereira
1035
                               23 1658
                                             1 1.993093
                                                               5 4.626337
2477 2094
                               18 1438
                                             2 1.080336
              Ivan Rakitic
                                                               1 3.312995
     8397
            Florian Wirtz
                                  1577
                                             5 4.982234
                                                               8 6.899908
2531 2369
                Toni Kroos
                               21 1386
                                            1 1.199653
                                                               7 4.584406
      shots key passes yellow cards red cards position
                                                               team title \
        29
                                                    MS
                                                                Barcelona
2433
                                                    MS
                                                                   Fulham
1035
2477
                    49
                                                    MS
                                                                  Sevilla
        43
1432
                    48
                                                         Bayer Leverkusen
2531
        26
                    43
                                                    MS
                                                              Real Madrid
                      xGChain xGBuildup
2433
           4.245366
                    21.189583
                     8.797632
                               5.037407
          1.080336
                     4.866999
                               3.656939
```

Defense

id	player_name	games	time	goals	xG	assists	хA	shots	key_passes
10891	Carlos Augusto	21	777	0	0.6726682186126709	1	1.9542889073491096	14	10
557	Matteo Darmian	20	1543	0	0.36853274889290333	0	0.7621378395706415	4	14
1463	Francesco Acerbi	18	1481	1	0.8611980956047773	1	1.5077504962682724	11	5
9655	Pablo Rosario	17	1006	0	0.5473238825798035	0	0.7476404532790184	9	8
5613	Gabriel	20	1634	2	1.903535034507513	0	0.8747859355062246	15	3
64	Filip Kostic	20	1353	0	0.4250156534835696	4	5.691956587135792	12	34
6888	William Saliba	22	1980	1	1.6387791624292731	1	0.15471498668193817	6	4
987	Joseph Gomez	19	1071	0	0.5445053642615676	1	1.034731399267912	12	14
6960	Bremer	22	1975	2	1.9601727705448866	0	0.05695765279233456	9	2
148	Dante	19	1710	0	0.4793338105082512	1	0.6169781722128391	8	9

Mid Field

id	player_name	games	time	goals	хG	assists	хA	shots	key_passes
337	Leroy Sané	20	1757	8	9.621190639212728	11	10.750425968319178	61	63
11866	Álex Grimaldo	2)	1796	7	3.0829383581876755	9	6.583418718539178	40	43
8397	Florian Wirtz	20	1577	5	4.982234204187989	8	6.8999078422784805	43	48
224	Thomas Müller	18	674	1	2.5389295406639576	7	6.67482359521091	14	33
6055	Phil Foden	21	1690	5	5.10170772485435	7	5.194825515151024	51	43
320	Jonas Hofmann	20	1580	5	3.7150106858462095	6	6.014208421111107	37	48
10846	Julián Álvarez	21	1800	8	8.56253857910633	5	4.240109716542065	58	47
3635	Bernardo Silva	17	1433	6	2.3876431211829185	4	4.979102602228522	19	33
8327	Bruno Guimarães	22	1961	1	2.5503133088350296	4	4.392050586640835	25	32
226	Kingsley Coman	15	1081	3	2.9748985283076763	3	3.4939965773373842	33	27

GK

id	player_name	games	time	goals	хG	assists	хA	shots	key_passes	yellow_cards	red_cards	position	team_title	npg	npxG
199	Yann Sommer	21	1890	0	0.0	0	0.0	0	0	1	0	GK	Inter	0	0.0
1244	Wojciech Szczesny	20	1800	0	0.0	0	0.07034061849117279	0	1	0	0	GK	Juventus	0	0.0
1257	Alisson	20	1800	0	0.0	0	0.0	0	0	1	0	GK	Liverpool	0	0.0
7772	Marcin Bulka	19	1710	0	0.0	0	0.0	0	0	1	0	GK	Nice	0	0.0
9676	David Raya	17	1530	0	0.0	0	0.0	0	0	0	0	GK	Arsenal	0	0.0
5603	Aaron Ramsdale	5	450	0	0.0	0	0.0	0	0	0	0	GK	Arsenal	0	0.0
1093	Mattia Perin	2	180	0	0.0	0	0.0	0	0	0	0	GK	Juventus	0	0.0
7904	Caoimhin Kelleher	2	180	0	0.0	0	0.0	0	0	0	0	GK	Liverpool	0	0.0



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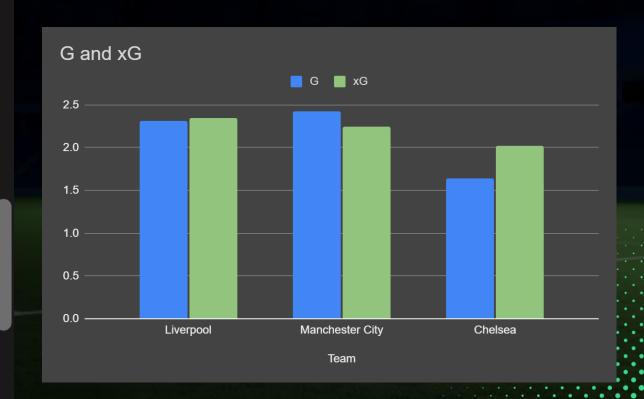
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FINAL TEAM

\$ 150 M

CURRENT BUDGET



2-WAY STRATEGY

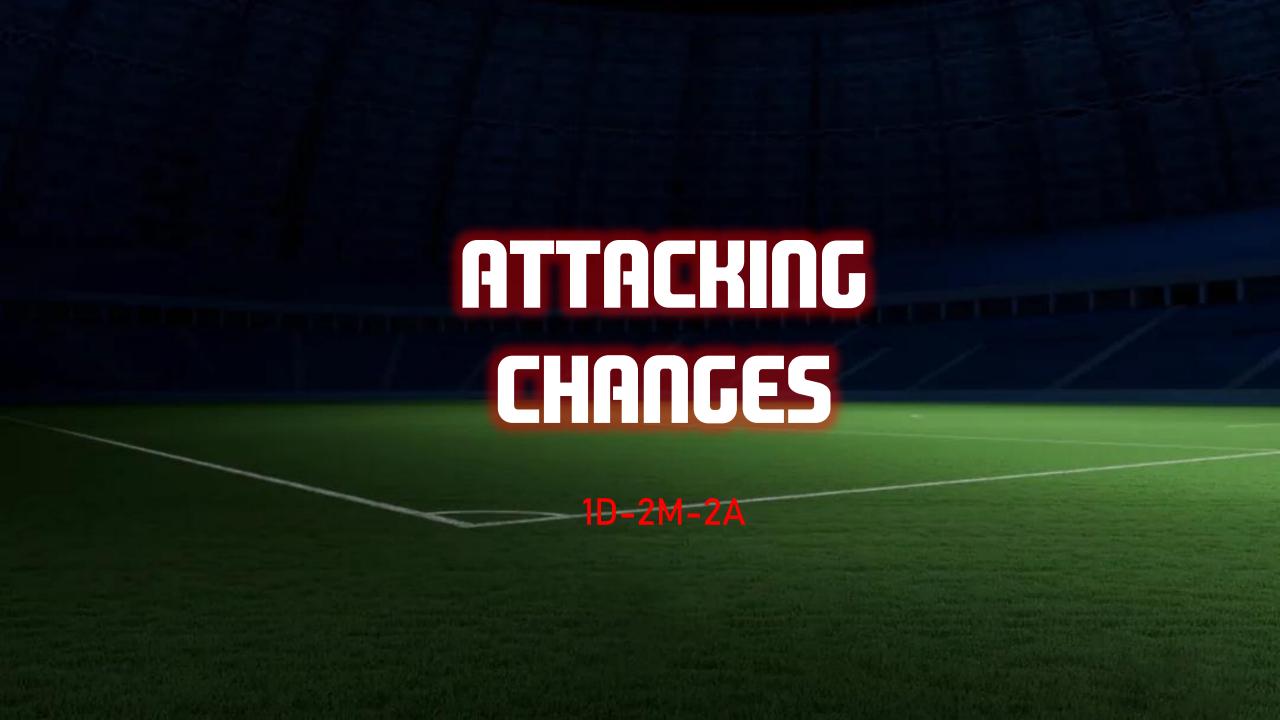
ATTACKING CHANGES

1D-2M-2A



DEFENSIVE CHANGES

1A-2M-2D



+ \$ 25 M







-\$ 40 M

ATTACKING CHANGES

BOUGHT

SERHOU GUIRASSY

WILLIAM SALIBA

ÁLEX GRIMALDO

MATEO CASSIERRA

FLORIAN WIRTZ (*MIDFIELDER)

SOLD

NICOLAS JACKSON

BENOIT BADIASHILE MUKINAYI

MARC CUCURELLA

RAHEEM STERLING

DEFENSIVE CHANGES 1A-2M-2D

+ \$ 25 M







-\$ 110 M

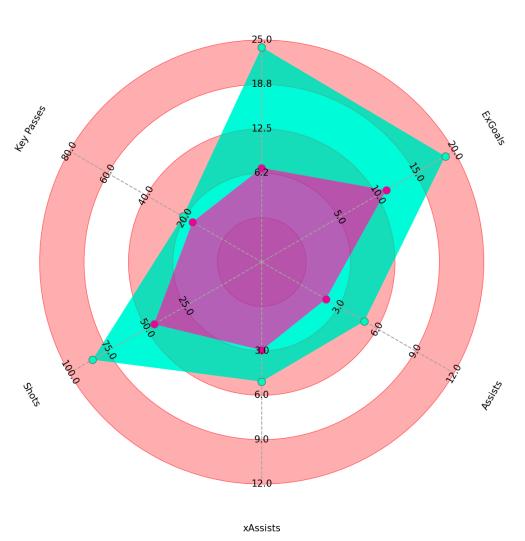


Nicolas Jackson Chelsea



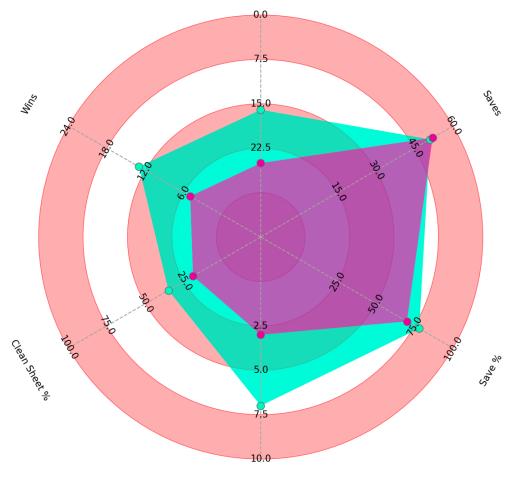
Robert Sanchez Chelsea





Goals

Goals Against



Clean Sheet

DEFENSIVE CHANGES

BOUGHT

HARRY KANE

ALISSON BECKER

JOSEPH GOMEZ

PABLO ROSARIO

FLORIAN WIRTZ (*MIDFIELDER)

SOLD

NICOLAS JACKSON

ROBERT SÁNCHEZ

MARC CUCURELLA

RAHEEM STERLING



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CURRENT BUDGET

ATTACKING



DEFENSIVE



BUDGET SHEET

	Att	acking Team	1		Defensive Team				
Transfer OUT	Transfer IN	OUT Value	IN Value	150 mil	Transfer OUT	Transfer IN	OUT Value	IN Value	150 mil
-	Wirtz	-	100 mil.	- 100 mil	-	Wirtz	-	100 mil	- 100 mil
Badiashile	Saliba	35 mil	75 mil	- 40 mil	Cucurella	Gomez	25 mil	22 mil	3 mil
Cucurella	Grimaldo	25 mil	35 mil	- 10 mil	Sterling	Rosario	45 mil	7 mil	38 mil
Jackson	Guirassy	25 mil	40 mil	- 15 mil	Sanchez	Alisson	30 mil	32 mil	-2 mil
Sterling	Cassiera	45 mil	9 mil	36 mil	Jackson	Kane	25 mil	110 mil	-85 mil
			Net Remains:	21 mil				Net Remains:	4 mil

