

Importing Libraries:

```
In [67]: import pandas as pd  
import numpy as np
```

Importing Data:

```
In [68]: data_fifa = pd.read_csv("C:/Users/amade/OneDrive/Área de Trabalho/FELIPE(NÃO MEXER)/Projects Data Análisys Portfolio/Data Cleaning and Transformation Pr
display(data_fifa)
```

	photoUrl	LongName	playerUrl	Nationality	Positions	Name	Age	IOVA	POT	Team & Contract	...	A/W
0	https://cdn.sofifa.com/players/158/023/21_60.png	Lionel Messi	http://sofifa.com/player/158023/lionel-messi/2...	Argentina	RW ST CF	L. Messi	33	93	93	\n\n\n\nFC Barcelona\n2004 ~ 2021\n\n	...	Medium
1	https://cdn.sofifa.com/players/020/801/21_60.png	C. Ronaldo dos Santos Aveiro	http://sofifa.com/player/20801/c-ronaldo-dos-s...	Portugal	ST LW	Cristiano Ronaldo	35	92	92	\n\n\n\nJuventus\n2018 ~ 2022\n\n	...	High
2	https://cdn.sofifa.com/players/200/389/21_60.png	Jan Oblak	http://sofifa.com/player/200389/jan-oblak/210005/	Slovenia	GK	J. Oblak	27	91	93	\n\n\n\nAtlético Madrid\n2014 ~ 2023\n\n	...	Medium
3	https://cdn.sofifa.com/players/192/985/21_60.png	Kevin De Bruyne	http://sofifa.com/player/192985/kevin-de-bruyn...	Belgium	CAM CM	K. De Bruyne	29	91	91	\n\n\n\nManchester City\n2015 ~ 2023\n\n	...	High
4	https://cdn.sofifa.com/players/190/871/21_60.png	Neymar da Silva Santos Jr.	http://sofifa.com/player/190871/neymar-da-silv...	Brazil	LW CAM	Neymar Jr	28	91	91	\n\n\n\nParis Saint-Germain\n2017 ~ 2022\n\n	...	High
...
18974	https://cdn.sofifa.com/players/257/710/21_60.png	Mengxuan Zhang	http://sofifa.com/player/257710/mengxuan-zhang...	China PR	CB	Zhang Mengxuan	21	47	52	\n\n\n\nChongqing Dangdai Lifan FC SWM Team\n2...	...	Low
18975	https://cdn.sofifa.com/players/258/736/21_60.png	Vani Da Silva	http://sofifa.com/player/258736/vani-da-silva/...	England	ST	V. Da Silva	17	47	67	\n\n\n\nOldham Athletic\n2020 ~ 2021\n\n	...	Medium
18976	https://cdn.sofifa.com/players/247/223/21_60.png	Ao Xia	http://sofifa.com/player/247223/ao-xia/210005/	China PR	CB	Xia Ao	21	47	55	\n\n\n\nWuhan Zall\n2018 ~ 2022\n\n	...	Medium
18977	https://cdn.sofifa.com/players/258/760/21_60.png	Ben Hough	http://sofifa.com/player/258760/ben-hough/210005/	England	CM	B. Hough	17	47	67	\n\n\n\nOldham Athletic\n2020 ~ 2021\n\n	...	Medium
18978	https://cdn.sofifa.com/players/255/958/21_60.png	Mateo Flores	http://sofifa.com/player/255958/mateo-flores/2...	Bolivia	CDM	M. Flores	19	47	63	\n\n\n\nClub Bolívar\n2020 ~ 2024\n\n	...	Medium

18979 rows × 77 columns

Analysing Dataset:

In [69]: `# ANALYSING COLUMNS:`

```
data_fifa.columns
```

Out[69]: Index(['photoUrl', 'LongName', 'playerUrl', 'Nationality', 'Positions', 'Name', 'Age', 'JOVA', 'POT', 'Team & Contract', 'ID', 'Height', 'Weight', 'foot', 'BOV', 'BP', 'Growth', 'Joined', 'Loan Date End', 'Value', 'Wage', 'Release Clause', 'Attacking', 'Crossing', 'Finishing', 'Heading Accuracy', 'Short Passing', 'Volleys', 'Skill', 'Dribbling', 'Curve', 'FK Accuracy', 'Long Passing', 'Ball Control', 'Movement', 'Acceleration', 'Sprint Speed', 'Agility', 'Reactions', 'Balance', 'Power', 'Shot Power', 'Jumping', 'Stamina', 'Strength', 'Long Shots', 'Mentality', 'Aggression', 'Interceptions', 'Positioning', 'Vision', 'Penalties', 'Composure', 'Defending', 'Marking', 'Standing Tackle', 'Sliding Tackle', 'Goalkeeping', 'GK Diving', 'GK Handling', 'GK Kicking', 'GK Positioning', 'GK Reflexes', 'Total Stats', 'Base Stats', 'W/F', 'SM', 'A/W', 'D/W', 'IR', 'PAC', 'SHO', 'PAS', 'DRI', 'DEF', 'PHY', 'Hits'], dtype='object')

In [70]: *# ANALYSING DATA INFO:*

```
data_fifa.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18979 entries, 0 to 18978
Data columns (total 77 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   photoUrl              18979 non-null  object
 1   LongName              18979 non-null  object
 2   playerUrl            18979 non-null  object
 3   Nationality          18979 non-null  object
 4   Positions            18979 non-null  object
 5   Name                 18979 non-null  object
 6   Age                  18979 non-null  int64
 7   ↓OVA                 18979 non-null  int64
 8   POT                  18979 non-null  int64
 9   Team & Contract      18979 non-null  object
10   ID                   18979 non-null  int64
11   Height               18979 non-null  object
12   Weight               18979 non-null  object
13   foot                 18979 non-null  object
14   BOV                  18979 non-null  int64
15   BP                   18979 non-null  object
16   Growth               18979 non-null  int64
17   Joined               18979 non-null  object
18   Loan Date End        1013 non-null   object
19   Value                18979 non-null  object
20   Wage                 18979 non-null  object
21   Release Clause       18979 non-null  object
22   Attacking            18979 non-null  int64
23   Crossing             18979 non-null  int64
24   Finishing            18979 non-null  int64
25   Heading Accuracy     18979 non-null  int64
26   Short Passing        18979 non-null  int64
27   Volleys              18979 non-null  int64
28   Skill                18979 non-null  int64
29   Dribbling            18979 non-null  int64
30   Curve                18979 non-null  int64
31   FK Accuracy          18979 non-null  int64
32   Long Passing         18979 non-null  int64
33   Ball Control         18979 non-null  int64
34   Movement             18979 non-null  int64
35   Acceleration         18979 non-null  int64
36   Sprint Speed         18979 non-null  int64
37   Agility              18979 non-null  int64
38   Reactions            18979 non-null  int64
39   Balance              18979 non-null  int64
40   Power                18979 non-null  int64
41   Shot Power           18979 non-null  int64
42   Jumping              18979 non-null  int64
43   Stamina              18979 non-null  int64
44   Strength             18979 non-null  int64
45   Long Shots           18979 non-null  int64
46   Mentality            18979 non-null  int64

```

```
47 Aggression      18979 non-null int64
48 Interceptions   18979 non-null int64
49 Positioning     18979 non-null int64
50 Vision          18979 non-null int64
51 Penalties       18979 non-null int64
52 Composure       18979 non-null int64
53 Defending       18979 non-null int64
54 Marking         18979 non-null int64
55 Standing Tackle 18979 non-null int64
56 Sliding Tackle  18979 non-null int64
57 Goalkeeping     18979 non-null int64
58 GK Diving       18979 non-null int64
59 GK Handling      18979 non-null int64
60 GK Kicking      18979 non-null int64
61 GK Positioning  18979 non-null int64
62 GK Reflexes     18979 non-null int64
63 Total Stats     18979 non-null int64
64 Base Stats      18979 non-null int64
65 W/F             18979 non-null object
66 SM              18979 non-null object
67 A/W             18979 non-null object
68 D/W             18979 non-null object
69 IR              18979 non-null object
70 PAC             18979 non-null int64
71 SHO             18979 non-null int64
72 PAS             18979 non-null int64
73 DRI             18979 non-null int64
74 DEF             18979 non-null int64
75 PHY             18979 non-null int64
76 Hits           18979 non-null object
dtypes: int64(55), object(22)
memory usage: 11.1+ MB
```

```
In [71]: # Columns we have to change/drop: Height, Weight, Team & Contract, Joined, Loan Date End, Value, Wage, Release Clause.
```

In [72]: `data_fifa[["Height", "Weight", "Team & Contract", "Joined", "Loan Date End", "Value", "Wage", "Release Clause", "Hits"]]`

Out[72]:

	Height	Weight	Team & Contract	Joined	Loan Date End	Value	Wage	Release Clause	Hits
0	5'7"	159lbs	\n\n\n\nFC Barcelona\n2004 ~ 2021\n\n	Jul 1, 2004	NaN	€67.5M	€560K	€138.4M	\n372
1	6'2"	183lbs	\n\n\n\nJuventus\n2018 ~ 2022\n\n	Jul 10, 2018	NaN	€46M	€220K	€75.9M	\n344
2	6'2"	192lbs	\n\n\n\nAtlético Madrid\n2014 ~ 2023\n\n	Jul 16, 2014	NaN	€75M	€125K	€159.4M	\n86
3	5'11"	154lbs	\n\n\n\nManchester City\n2015 ~ 2023\n\n	Aug 30, 2015	NaN	€87M	€370K	€161M	\n163
4	5'9"	150lbs	\n\n\n\nParis Saint-Germain\n2017 ~ 2022\n\n	Aug 3, 2017	NaN	€90M	€270K	€166.5M	\n273
...
18974	5'10"	154lbs	\n\n\n\nChongqing Dangdai Lifan FC SWM Team\n2...	Aug 1, 2020	NaN	€35K	€1K	€57K	\n2
18975	5'7"	128lbs	\n\n\n\nOldham Athletic\n2020 ~ 2021\n\n	Aug 1, 2020	NaN	€60K	€500	€165K	\n3
18976	5'10"	146lbs	\n\n\n\nWuhan Zall\n2018 ~ 2022\n\n	Jul 13, 2018	NaN	€40K	€1K	€70K	\n3
18977	5'9"	143lbs	\n\n\n\nOldham Athletic\n2020 ~ 2021\n\n	Aug 1, 2020	NaN	€60K	€500	€165K	\n5
18978	5'9"	150lbs	\n\n\n\nClub Bolívar\n2020 ~ 2024\n\n	Jan 1, 2020	NaN	€60K	€500	€167K	\n2

18979 rows × 9 columns

In [73]: `# (DATA CLEANING AND TRANSFORMATION PROCESS):`

```
# 1º Drop columns(photoUrl, playerId and Loan Date End).
# 2º Rename colum Long Name to Full Name, OVA to Overall and POT to Potential.
# 3º Trasform Height from feet to cm + convert dtype to int.
# 4º Transform Weight from pounds to kg + convert dtype to int.
# 5º Transform Joined column from : Month day, year, to : month-day-year.
# 6º Split Team and contract columns and remove the \n's: \n\n\n\nFC Barcelona\n2004 ~ 2021\n\n ,to : FC Barcelona / 2004 ~ 2021.
# 7º Create a column years with club (end contract - start contract).
# 8º Transform Value,Wage and Release Clause collumns into normal extendend numbers without currency symbols + convert dtype to int.
# 9º Remove the \n's from hits and convert dtype to int.
# 10º Remove the star symbol(★) from the columns that have it.
```

Transforming Data:

In [74]: `# (DROPPING COLUMNS PHOTOURL, PLAYERURL AND LOAN DATE END):`

```
data_fifa = data_fifa.drop(["photoUrl", "playerUrl", "Loan Date End"], axis = 1)
data_fifa.shape
```

Out[74]: (18979, 74)

In [75]: # (RENAMING COLUMN LONG NAME TO FULL NAME, OVA TO OVERALL AND POT TO POTENTIAL):

```
data_fifa = data_fifa.rename(columns={'LongName': 'Full Name', '↓OVA': 'Overall', 'POT': 'Potential'})
display(data_fifa[['Full Name', 'Overall', 'Potential']])
```

	Full Name	Overall	Potential
0	Lionel Messi	93	93
1	C. Ronaldo dos Santos Aveiro	92	92
2	Jan Oblak	91	93
3	Kevin De Bruyne	91	91
4	Neymar da Silva Santos Jr.	91	91
...
18974	Mengxuan Zhang	47	52
18975	Vani Da Silva	47	67
18976	Ao Xia	47	55
18977	Ben Hough	47	67
18978	Mateo Flores	47	63

18979 rows × 3 columns

In [76]: # (1ª TRASFORMING HEIGHT FROM FEET TO CM + CONVERTING DTYPE TO INT):

```
data_fifa["Height"] = data_fifa["Height"].str.replace("'", ".") # replacing ' for .
```

In [77]: data_fifa["Height"] = data_fifa["Height"].apply(lambda x: x.replace("'", '')) # replacing " for empty
data_fifa["Height"]

```
Out[77]: 0      5.7
1      6.2
2      6.2
3      5.11
4      5.9
...
18974  5.10
18975  5.7
18976  5.10
18977  5.9
18978  5.9
Name: Height, Length: 18979, dtype: object
```



```
In [78]: data_fifa["Height"] = data_fifa["Height"].astype(str).astype(float) # Transforming from Object to float, to make possible the
# multiplication from feet to cm with a higher precision.
```

```
In [79]: data_fifa["Height"] = data_fifa["Height"].apply(lambda x: x * 30.48) # Transforming height from feet to cm

data_fifa["Height"] = data_fifa["Height"].astype(int) # Transforming column from float to int.
data_fifa["Height"]
```

```
Out[79]: 0      173
1      188
2      188
3      155
4      179
...
18974   155
18975   173
18976   155
18977   179
18978   179
Name: Height, Length: 18979, dtype: int32
```

```
In [80]: # (2º TRANSFORMING WEIGHT FROM POUNDS TO KG + CONVERTING DTYPE TO INT):
```

```
In [81]: data_fifa["Weight"] = data_fifa["Weight"].apply(lambda x: x.replace("lbs", "")) # Removing the lbs from the weight value.
data_fifa["Weight"]
```

```
Out[81]: 0      159
1      183
2      192
3      154
4      150
...
18974   154
18975   128
18976   146
18977   143
18978   150
Name: Weight, Length: 18979, dtype: object
```

```
In [82]: data_fifa["Weight"] = data_fifa["Weight"].astype(str).astype(float) # transforming column weight from object type to float to
# make possible the multiplication from pounds to kg with a higher precision.
data_fifa["Weight"]
```

```
Out[82]: 0      159.0
1      183.0
2      192.0
3      154.0
4      150.0
...
18974   154.0
18975   128.0
18976   146.0
18977   143.0
18978   150.0
Name: Weight, Length: 18979, dtype: float64
```

```
In [83]: data_fifa["Weight"] = data_fifa["Weight"].apply(lambda x: x * 0.453) # Transforming weight from pounds to Kg
data_fifa["Weight"]
```

```
Out[83]: 0      72.027
1      82.899
2      86.976
3      69.762
4      67.950
...
18974   69.762
18975   57.984
18976   66.138
18977   64.779
18978   67.950
Name: Weight, Length: 18979, dtype: float64
```

```
In [84]: data_fifa["Weight"] = data_fifa["Weight"].astype(int) # Transforming column from float to int.
data_fifa["Weight"]
```

```
Out[84]: 0      72
1      82
2      86
3      69
4      67
..
18974   69
18975   57
18976   66
18977   64
18978   67
Name: Weight, Length: 18979, dtype: int32
```

```
In [85]: # (3º JOINED FROM : MONTH DAY, YEAR, TO : MONTH-DAY-YEAR):
```

```
In [86]: data_fifa["Joined"]
```

```
Out[86]: 0      Jul 1, 2004
          1      Jul 10, 2018
          2      Jul 16, 2014
          3      Aug 30, 2015
          4      Aug 3, 2017
          ...
          18974    Aug 1, 2020
          18975    Aug 1, 2020
          18976    Jul 13, 2018
          18977    Aug 1, 2020
          18978    Jan 1, 2020
          Name: Joined, Length: 18979, dtype: object
```

```
In [87]: def transform_date(df, column_name): # Function that transform Jul 1, 2004 to 07-01-2004.
          # Convert the column to datetime format
          data_fifa["Joined"] = pd.to_datetime(data_fifa["Joined"], format="%b %d, %Y")

          # Transform the date format to MONTH-DAY-YEAR
          data_fifa["Joined"] = data_fifa["Joined"].dt.strftime("%m-%d-%Y")

          return df

          data_fifa = transform_date(data_fifa, 'Joined')
```

```
In [88]: data_fifa["Joined"]
```

```
Out[88]: 0      07-01-2004
          1      07-10-2018
          2      07-16-2014
          3      08-30-2015
          4      08-03-2017
          ...
          18974    08-01-2020
          18975    08-01-2020
          18976    07-13-2018
          18977    08-01-2020
          18978    01-01-2020
          Name: Joined, Length: 18979, dtype: object
```

```
In [89]: # (4º Transforming Weight from pounds to kg + convert dtype to int):
```

```
In [90]: data_fifa['Team & Contract'] = data_fifa['Team & Contract'].astype('str') # Creating a function that splits Team from Contract.
data_fifa['Team & Contract'].replace('\n', '', regex=True, inplace=True)
test = data_fifa['Team & Contract'][0]
Team = []
Contract_Duration = []
for x in range(len(data_fifa['Team & Contract'])):
    Team.append(str(data_fifa['Team & Contract'][x][::-11]))
    c = str(data_fifa['Team & Contract'][x][-11:])
    if c.startswith("2") == True:
        Contract_Duration.append(c)
    else:
        Contract_Duration.append("0")
data_fifa = data_fifa.drop(columns = ['Team & Contract'])
data_fifa['Team'] = Team
data_fifa['Contract Duration'] = Contract_Duration
```

```
In [91]: data_fifa[["Team", "Contract Duration"]]
```

Out[91]:

	Team	Contract Duration
0	FC Barcelona	2004 ~ 2021
1	Juventus	2018 ~ 2022
2	Atlético Madrid	2014 ~ 2023
3	Manchester City	2015 ~ 2023
4	Paris Saint-Germain	2017 ~ 2022
...
18974	Chongqing Dangdai Lifan FC SWM Team	2020 ~ 2020
18975	Oldham Athletic	2020 ~ 2021
18976	Wuhan Zall	2018 ~ 2022
18977	Oldham Athletic	2020 ~ 2021
18978	Club Bolívar	2020 ~ 2024

18979 rows × 2 columns

```
In [92]: # (5º CREATING A COLUMN YEARS WITH CLUB):
```

```
In [93]: def subtract_values(column): # Function that subtracts end of contract from start of contract.
# Split the values in the column based on "~"
values = column.str.split("~")

# Apply the subtraction operation and set the result to 0 if it's already 0
subtracted_values = values.apply(lambda x: float(x[1]) - float(x[0]) if x[0] != '0' else 0)

return subtracted_values
```

```
In [94]: data_fifa['Years_in_Club'] = subtract_values(data_fifa["Contract Duration"]) # Creating new column Years_in_Club.
```

```
In [95]: data_fifa['Years_in_Club'] = data_fifa['Years_in_Club'].astype(int)
```

```
In [96]: # (6º SPLIT TEAM AND CONTRACT COLUMNS AND REMOVE THE \n's: \n\n\nFC Barcelona\n2004 ~ 2021\n\n ,to : FC Barcelona / 2004 ~ 2021)
```

```
In [97]: data_fifa[["Value", "Wage", "Release Clause"]]
```

```
Out[97]:
```

	Value	Wage	Release Clause
0	€67.5M	€560K	€138.4M
1	€46M	€220K	€75.9M
2	€75M	€125K	€159.4M
3	€87M	€370K	€161M
4	€90M	€270K	€166.5M
...
18974	€35K	€1K	€57K
18975	€60K	€500	€165K
18976	€40K	€1K	€70K
18977	€60K	€500	€165K
18978	€60K	€500	€167K

18979 rows × 3 columns

```
In [98]: data_fifa["Value"] = data_fifa["Value"].apply(lambda x: x.replace("€", "")) # Removing the euro sign(€) from the values
data_fifa["Wage"] = data_fifa["Wage"].apply(lambda x: x.replace("€", ""))
data_fifa["Release Clause"] = data_fifa["Release Clause"].apply(lambda x: x.replace("€", ""))
```

```
In [99]: # Function to convert the Letter to a multiplier
def get_multiplier(letter):
    if letter == 'M':
        return 1000000
    elif letter == 'K':
        return 1000
    else:
        return 1

# Iterate over each value in the column and update it
for index, value in data_fifa["Value"].items():
    if value[-1].isdigit():
        # No Letter present, so no multiplication needed
        continue
    multiplier = get_multiplier(value[-1])
    number = float(value[:-1])
    data_fifa.at[index, "Value"] = number * multiplier
```

```
In [100]: # Iterate over each value in the column and update it
for index, value in data_fifa["Release Clause"].items():
    if value[-1].isdigit():
        # No Letter present, so no multiplication needed
        continue
    multiplier = get_multiplier(value[-1])
    number = float(value[:-1])
    data_fifa.at[index, "Release Clause"] = number * multiplier
```

```
In [101]: # Iterate over each value in the column and update it
for index, value in data_fifa["Wage"].items():
    if value[-1].isdigit():
        # No Letter present, so no multiplication needed
        continue
    multiplier = get_multiplier(value[-1])
    number = float(value[:-1])
    data_fifa.at[index, "Wage"] = number * multiplier
```

```
In [102]: data_fifa[["Value", "Wage", "Release Clause"]]
```

```
Out[102]:
```

	Value	Wage	Release Clause
0	67500000.0	560000.0	138400000.0
1	46000000.0	220000.0	75900000.0
2	75000000.0	125000.0	159400000.0
3	87000000.0	370000.0	161000000.0
4	90000000.0	270000.0	166500000.0
...
18974	35000.0	1000.0	57000.0
18975	60000.0	500	165000.0
18976	40000.0	1000.0	70000.0
18977	60000.0	500	165000.0
18978	60000.0	500	167000.0

18979 rows × 3 columns

```
In [103]: # Transforming dtype from object to float !!
```

```
data_fifa[["Value", "Wage", "Release Clause"]] = data_fifa[["Value", "Wage", "Release Clause"]].astype(str).astype(float)
```

```
In [104]: data_fifa[["Value", "Wage", "Release Clause"]].info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18979 entries, 0 to 18978
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Value           18979 non-null float64
1   Wage            18979 non-null float64
2   Release Clause  18979 non-null float64
dtypes: float64(3)
memory usage: 444.9 KB
```

```
In [105]: # (7° REMOVING THE \n'S FROM HITS + CONVERT DTYPE TO INT):
```

```
In [106]: # Visualizing data before transformation:  
data_fifa["Hits"].info()
```

```
<class 'pandas.core.series.Series'>  
RangeIndex: 18979 entries, 0 to 18978  
Series name: Hits  
Non-Null Count  Dtype  
-----  
18979 non-null  object  
dtypes: object(1)  
memory usage: 148.4+ KB
```

```
In [107]: data_fifa["Hits"].describe()
```

```
Out[107]: count      18979  
unique        374  
top           \n1  
freq          4321  
Name: Hits, dtype: object
```

```
In [108]: set(data_fifa["Hits"])
```

```
Out[108]: {'\n1',  
          '\n1.1K',  
          '\n1.2K',  
          '\n1.3K',  
          '\n1.5K',  
          '\n1.7K',  
          '\n1.8K',  
          '\n1.9K',  
          '\n10',  
          '\n100',  
          '\n101',  
          '\n102',  
          '\n103',  
          '\n104',  
          '\n105',  
          '\n106',  
          '\n107',  
          '\n108',  
          '\n109',  
          '\n11'}
```

```
In [109]: data_fifa["Hits"] = data_fifa["Hits"].replace(r'\n', '', regex=True) # Removing the \n's from the column.
```



```
In [110]: # Function to convert the letter to a multiplier
def get_multiplier(letter):
    if letter == 'M':
        return 1000000
    elif letter == 'K':
        return 1000
    else:
        return 1

# Iterate over each value in the column and update it
for index, value in data_fifa["Hits"].items():
    if value[-1].isdigit():
        # No letter present, so no multiplication needed
        continue
    multiplier = get_multiplier(value[-1])
    number = float(value[:-1])
    data_fifa.at[index, "Hits"] = number * multiplier
```

```
In [111]: data_fifa["Hits"] = data_fifa["Hits"].astype(int) # Converting data from object to int.
```

```
In [112]: # Visualizing data after transformation:
```

```
data_fifa[["Hits", "Full Name"]]
```

```
Out[112]:
```

	Hits	Full Name
0	372	Lionel Messi
1	344	C. Ronaldo dos Santos Aveiro
2	86	Jan Oblak
3	163	Kevin De Bruyne
4	273	Neymar da Silva Santos Jr.
...
18974	2	Mengxuan Zhang
18975	3	Vani Da Silva
18976	3	Ao Xia
18977	5	Ben Hough
18978	2	Mateo Flores

18979 rows × 2 columns

```
In [113]: set(data_fifa["Hits"])
```

```
Out[113]: {1,  
2,  
3,  
4,  
5,  
6,  
7,  
8,  
9,  
10,  
11,  
12,  
13,  
14,  
15,  
16,  
17,  
18,  
19,  
~~
```

```
In [114]: data_fifa["Hits"].info()
```

```
<class 'pandas.core.series.Series'>  
RangeIndex: 18979 entries, 0 to 18978  
Series name: Hits  
Non-Null Count  Dtype  
-----  
18979 non-null  int32  
dtypes: int32(1)  
memory usage: 74.3 KB
```

```
In [115]: data_fifa["Hits"].describe()
```

```
Out[115]: count    18979.000000  
mean         16.934085  
std          73.203131  
min           1.000000  
25%           2.000000  
50%           3.000000  
75%          10.000000  
max         4500.000000  
Name: Hits, dtype: float64
```

```
In [116]: # (8° Remove the star symbol(★) from the columns W/F, SM and IR):
```

```
In [117]: # Search for columns that have the star symbol.

# Specify the character to search for
character_to_find = '★'

# Iterate over the columns and check for the character
columns_with_character = []
for column in data_fifa.columns:
    if any(data_fifa[column].astype(str).str.contains(character_to_find)):
        columns_with_character.append(column)

print(f"The columns that contain the character '{character_to_find}' are: {columns_with_character}")
```

The columns that contain the character '★' are: ['W/F', 'SM', 'IR']

```
In [118]: data_fifa[["W/F", "SM", "IR"]]
```

```
Out[118]:
```

	W/F	SM	IR
0	4★	4★	5★
1	4★	5★	5★
2	3★	1★	3★
3	5★	4★	4★
4	5★	5★	5★
...
18974	2★	2★	1★
18975	2★	2★	1★
18976	2★	2★	1★
18977	2★	2★	1★
18978	3★	2★	1★

18979 rows × 3 columns

```
In [119]: data_fifa[["W/F", "SM", "IR"]] = data_fifa[["W/F", "SM", "IR"]].apply(lambda x: x.str.replace('★', '')) # removing star from columns that have it !!!.
```

```
In [120]: data_fifa[["W/F", "SM", "IR"]]
```

```
Out[120]:
```

	W/F	SM	IR
0	4	4	5
1	4	5	5
2	3	1	3
3	5	4	4
4	5	5	5
...
18974	2	2	1
18975	2	2	1
18976	2	2	1
18977	2	2	1
18978	3	2	1

18979 rows × 3 columns

```
In [ ]:
```