

# preliminary\_clean

December 14, 2022

This is the first jupyter notebook where I'd be making preliminary cleaning of the datasets so as to use for analysis in a later notebook. After the cleaning is done here, I'll then export it to MySQL to make use of JOINS to merge different datasets and columns to another.

## 0.1 Match

```
[160]: import pandas as pd
match = pd.read_csv('Match.csv')
match.head()
```

```
[160]:   id  country_id  league_id   season  stage          date \
0    1           1           1  2008/2009      1  2008-08-17 00:00:00
1    2           1           1  2008/2009      1  2008-08-16 00:00:00
2    3           1           1  2008/2009      1  2008-08-16 00:00:00
3    4           1           1  2008/2009      1  2008-08-17 00:00:00
4    5           1           1  2008/2009      1  2008-08-16 00:00:00

   match_api_id  home_team_api_id  away_team_api_id  home_team_goal  ... \
0         492473             9987             9993              1  ...
1         492474            10000             9994              0  ...
2         492475             9984             8635              0  ...
3         492476             9991             9998              5  ...
4         492477             7947             9985              1  ...

   SJA   VCH   VCD   VCA   GBH   GBD   GBA   BSH   BSD   BSA
0  4.00  1.65  3.40  4.50  1.78  3.25  4.00  1.73  3.40  4.20
1  3.80  2.00  3.25  3.25  1.85  3.25  3.75  1.91  3.25  3.60
2  2.50  2.35  3.25  2.65  2.50  3.20  2.50  2.30  3.20  2.75
3  7.50  1.45  3.75  6.50  1.50  3.75  5.50  1.44  3.75  6.50
4  1.73  4.50  3.40  1.65  4.50  3.50  1.65  4.75  3.30  1.67
```

[5 rows x 115 columns]

```
[161]: match.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 25979 entries, 0 to 25978
Columns: 115 entries, id to BSA
```

```
dtypes: float64(96), int64(9), object(10)
memory usage: 22.8+ MB
```

```
[162]: match.dtypes
```

```
[162]: id                int64
      country_id        int64
      league_id         int64
      season            object
      stage             int64
      ...
      GBD               float64
      GBA               float64
      BSH               float64
      BSD               float64
      BSA               float64
      Length: 115, dtype: object
```

Change date datatype

```
[163]: match['date'] = pd.to_datetime(match['date']).dt.date
```

```
[164]: match['date'] = pd.to_datetime(match['date'])
```

```
[165]: match.columns
```

```
[165]: Index(['id', 'country_id', 'league_id', 'season', 'stage', 'date',
          'match_api_id', 'home_team_api_id', 'away_team_api_id',
          'home_team_goal',
          ...
          'SJA', 'VCH', 'VCD', 'VCA', 'GBH', 'GBD', 'GBA', 'BSH', 'BSD', 'BSA'],
          dtype='object', length=115)
```

```
[166]: print(match.columns.tolist())
```

```
['id', 'country_id', 'league_id', 'season', 'stage', 'date', 'match_api_id',
'home_team_api_id', 'away_team_api_id', 'home_team_goal', 'away_team_goal',
'home_player_X1', 'home_player_X2', 'home_player_X3', 'home_player_X4',
'home_player_X5', 'home_player_X6', 'home_player_X7', 'home_player_X8',
'home_player_X9', 'home_player_X10', 'home_player_X11', 'away_player_X1',
'away_player_X2', 'away_player_X3', 'away_player_X4', 'away_player_X5',
'away_player_X6', 'away_player_X7', 'away_player_X8', 'away_player_X9',
'away_player_X10', 'away_player_X11', 'home_player_Y1', 'home_player_Y2',
'home_player_Y3', 'home_player_Y4', 'home_player_Y5', 'home_player_Y6',
'home_player_Y7', 'home_player_Y8', 'home_player_Y9', 'home_player_Y10',
'home_player_Y11', 'away_player_Y1', 'away_player_Y2', 'away_player_Y3',
'away_player_Y4', 'away_player_Y5', 'away_player_Y6', 'away_player_Y7',
'away_player_Y8', 'away_player_Y9', 'away_player_Y10', 'away_player_Y11',
'home_player_1', 'home_player_2', 'home_player_3', 'home_player_4',
```

```
'home_player_5', 'home_player_6', 'home_player_7', 'home_player_8',
'home_player_9', 'home_player_10', 'home_player_11', 'away_player_1',
'away_player_2', 'away_player_3', 'away_player_4', 'away_player_5',
'away_player_6', 'away_player_7', 'away_player_8', 'away_player_9',
'away_player_10', 'away_player_11', 'goal', 'shoton', 'shotoff', 'foulcommit',
'card', 'cross', 'corner', 'possession', 'B365H', 'B365D', 'B365A', 'BWH',
'BWD', 'BWA', 'IWH', 'IWD', 'IWA', 'LBH', 'LBD', 'LBA', 'PSH', 'PSD', 'PSA',
'WHH', 'WHD', 'WHA', 'SJH', 'SJD', 'SJA', 'VCH', 'VCD', 'VCA', 'GBH', 'GBD',
'GBA', 'BSH', 'BSD', 'BSA']
```

from the match dataset, the data below is what I'd be needing for this analysis.

```
[167]: match = match[['id', 'country_id', 'league_id', 'date', 'season', 'stage',
↳ 'home_team_api_id', 'away_team_api_id', 'home_team_goal', 'away_team_goal']]
```

```
[168]: match.isnull().sum()
```

```
[168]: id                0
country_id             0
league_id              0
date                  0
season                0
stage                 0
home_team_api_id       0
away_team_api_id       0
home_team_goal         0
away_team_goal         0
dtype: int64
```

```
[171]: match.duplicated().sum()
```

```
[171]: 0
```

```
[172]: match.shape
```

```
[172]: (25979, 10)
```

```
[173]: # drop columns whose name contains specific string from dataframe
# match[match.columns.drop(list(match.filter(regex='away_')))]
```

```
[174]: match.to_csv('Match.csv', index=False)
```

## 0.2 Player attributes

```
[70]: play_att = pd.read_csv('Player_Attributes.csv')
play_att.head()
```

```
[70]:
```

	id	player_fifa_api_id	player_api_id	date	overall_rating	\
0	1	218353	505942	2016-02-18 00:00:00	67.0	
1	2	218353	505942	2015-11-19 00:00:00	67.0	
2	3	218353	505942	2015-09-21 00:00:00	62.0	
3	4	218353	505942	2015-03-20 00:00:00	61.0	
4	5	218353	505942	2007-02-22 00:00:00	61.0	

  

	potential	preferred_foot	attacking_work_rate	defensive_work_rate	crossing	\
0	71.0	right	medium	medium	49.0	
1	71.0	right	medium	medium	49.0	
2	66.0	right	medium	medium	49.0	
3	65.0	right	medium	medium	48.0	
4	65.0	right	medium	medium	48.0	

  

	...	vision	penalties	marking	standing_tackle	sliding_tackle	\
0	...	54.0	48.0	65.0	69.0	69.0	
1	...	54.0	48.0	65.0	69.0	69.0	
2	...	54.0	48.0	65.0	66.0	69.0	
3	...	53.0	47.0	62.0	63.0	66.0	
4	...	53.0	47.0	62.0	63.0	66.0	

  

	gk_diving	gk_handling	gk_kicking	gk_positioning	gk_reflexes
0	6.0	11.0	10.0	8.0	8.0
1	6.0	11.0	10.0	8.0	8.0
2	6.0	11.0	10.0	8.0	8.0
3	5.0	10.0	9.0	7.0	7.0
4	5.0	10.0	9.0	7.0	7.0

[5 rows x 42 columns]

```
[71]: play_att.shape
```

```
[71]: (183978, 42)
```

```
[72]: play_att.isnull().sum().sum()
```

```
[72]: 47301
```

```
[73]: play_att.duplicated().sum()
```

```
[73]: 0
```

```
[74]: play_att.dropna(inplace=True)
```

```
[76]: play_att.isnull().sum().sum()
```

```
[76]: 0
```

```
[77]: play_att.shape
```

```
[77]: (180354, 42)
```

```
[78]: play_att.columns
```

```
[78]: Index(['id', 'player_fifa_api_id', 'player_api_id', 'date', 'overall_rating',  
        'potential', 'preferred_foot', 'attacking_work_rate',  
        'defensive_work_rate', 'crossing', 'finishing', 'heading_accuracy',  
        'short_passing', 'volleys', 'dribbling', 'curve', 'free_kick_accuracy',  
        'long_passing', 'ball_control', 'acceleration', 'sprint_speed',  
        'agility', 'reactions', 'balance', 'shot_power', 'jumping', 'stamina',  
        'strength', 'long_shots', 'aggression', 'interceptions', 'positioning',  
        'vision', 'penalties', 'marking', 'standing_tackle', 'sliding_tackle',  
        'gk_diving', 'gk_handling', 'gk_kicking', 'gk_positioning',  
        'gk_reflexes'],  
        dtype='object')
```

```
[79]: play_att.drop(['volleys', 'curve', 'vision', 'standing_tackle', 'gk_reflexes'],  
                  ↪axis=1, inplace=True)
```

Convert date datatype

```
[81]: play_att['date'] = pd.to_datetime(play_att['date']).dt.date
```

```
[82]: play_att['date']= pd.to_datetime(play_att['date'])
```

```
[84]: print(play_att.dtypes)
```

id	int64
player_fifa_api_id	int64
player_api_id	int64
date	datetime64[ns]
overall_rating	float64
potential	float64
preferred_foot	object
attacking_work_rate	object
defensive_work_rate	object
crossing	float64
finishing	float64
heading_accuracy	float64
short_passing	float64
dribbling	float64
free_kick_accuracy	float64
long_passing	float64
ball_control	float64
acceleration	float64
sprint_speed	float64
agility	float64

```

reactions                float64
balance                  float64
shot_power                float64
jumping                  float64
stamina                  float64
strength                  float64
long_shots                float64
aggression                float64
interceptions            float64
positioning              float64
penalties                float64
marking                  float64
sliding_tackle           float64
gk_diving                float64
gk_handling              float64
gk_kicking               float64
gk_positioning           float64
dtype: object

```

```
[85]: play_att.to_csv('Player_Attributes.csv', index=False)
```

### 0.3 Player

```
[57]: play = pd.read_csv('Player.csv')
play.head()
```

```
[57]:
```

	id	player_api_id	player_name	player_fifa_api_id	\
0	1	505942	Aaron Appindangoye	218353	
1	2	155782	Aaron Cresswell	189615	
2	3	162549	Aaron Doran	186170	
3	4	30572	Aaron Galindo	140161	
4	5	23780	Aaron Hughes	17725	

		birthday	height	weight
0	1992-02-29 00:00:00	182.88	187	
1	1989-12-15 00:00:00	170.18	146	
2	1991-05-13 00:00:00	170.18	163	
3	1982-05-08 00:00:00	182.88	198	
4	1979-11-08 00:00:00	182.88	154	

Convert birthday from datetime to date

```
[59]: play['birthday'] = pd.to_datetime(play['birthday']).dt.date
```

```
[62]: play['birthday'] = pd.to_datetime(play['birthday'])
```

```
[63]: play.dtypes
```

```
[63]: id                int64
      player_api_id     int64
      player_name       object
      player_fifa_api_id int64
      birthday          datetime64[ns]
      height            float64
      weight            int64
      dtype: object
```

```
[65]: play.isnull().sum()
```

```
[65]: id                0
      player_api_id     0
      player_name       0
      player_fifa_api_id int64
      birthday          0
      height            0
      weight            0
      dtype: int64
```

```
[67]: play.duplicated().sum()
```

```
[67]: 0
```

```
[68]: play.shape
```

```
[68]: (11060, 7)
```

```
[69]: play.to_csv('Player.csv', index=False)
```

## 0.4 Team Attributes

```
[86]: team_att = pd.read_csv('Team_Attributes.csv')
      team_att.head()
```

```
[86]:
```

	id	team_fifa_api_id	team_api_id	date	buildUpPlaySpeed	\
0	1	434	9930	2010-02-22 00:00:00	60	
1	2	434	9930	2014-09-19 00:00:00	52	
2	3	434	9930	2015-09-10 00:00:00	47	
3	4	77	8485	2010-02-22 00:00:00	70	
4	5	77	8485	2011-02-22 00:00:00	47	

  

	buildUpPlaySpeedClass	buildUpPlayDribbling	buildUpPlayDribblingClass	\
0	Balanced	NaN	Little	
1	Balanced	48.0	Normal	
2	Balanced	41.0	Normal	
3	Fast	NaN	Little	
4	Balanced	NaN	Little	

	buildUpPlayPassing	buildUpPlayPassingClass	...	chanceCreationShooting	\
0	50	Mixed	...	55	
1	56	Mixed	...	64	
2	54	Mixed	...	64	
3	70	Long	...	70	
4	52	Mixed	...	52	

	chanceCreationShootingClass	chanceCreationPositioningClass	\
0	Normal	Organised	
1	Normal	Organised	
2	Normal	Organised	
3	Lots	Organised	
4	Normal	Organised	

	defencePressure	defencePressureClass	defenceAggression	\
0	50	Medium	55	
1	47	Medium	44	
2	47	Medium	44	
3	60	Medium	70	
4	47	Medium	47	

	defenceAggressionClass	defenceTeamWidth	defenceTeamWidthClass	\
0	Press	45	Normal	
1	Press	54	Normal	
2	Press	54	Normal	
3	Double	70	Wide	
4	Press	52	Normal	

	defenceDefenderLineClass
0	Cover
1	Cover
2	Cover
3	Cover
4	Cover

[5 rows x 25 columns]

Change date datatype

```
[87]: team_att['date'] = pd.to_datetime(team_att['date']).dt.date
```

```
[88]: team_att['date'] = pd.to_datetime(team_att['date'])
```

```
[89]: team_att.dtypes
```

```
[89]: id                int64
      team_fifa_api_id  int64
```



```

team_api_id                int64
date                      datetime64[ns]
buildUpPlaySpeed           int64
buildUpPlaySpeedClass      object
buildUpPlayDribbling       float64
buildUpPlayDribblingClass  object
buildUpPlayPassing         int64
buildUpPlayPassingClass    object
buildUpPlayPositioningClass object
chanceCreationPassing      int64
chanceCreationPassingClass object
chanceCreationCrossing     int64
chanceCreationCrossingClass object
chanceCreationShooting     int64
chanceCreationShootingClass object
chanceCreationPositioningClass object
defencePressure            int64
defencePressureClass       object
defenceAggression         int64
defenceAggressionClass     object
defenceTeamWidth          int64
defenceTeamWidthClass      object
defenceDefenderLineClass  object
dtype: object

```

```
[90]: team_att.isnull().sum().sum()
```

```
[90]: 969
```

```
[91]: team_att.duplicated().sum()
```

```
[91]: 0
```

```
[92]: team_att.dropna(inplace=True)
```

```
[93]: team_att.shape
```

```
[93]: (489, 25)
```

```
[94]: team_att.to_csv('Team_Attributes.csv', index=False)
```

## 0.5 Team

```
[95]: team = pd.read_csv('Team.csv')
team.head()
```

```
[95]:   id  team_api_id  team_fifa_api_id  team_long_name  team_short_name
0    1         9987          673.0         KRC Genk             GEN
```

1	2	9993	675.0	Beerschot AC	BAC
2	3	10000	15005.0	SV Zulte-Waregem	ZUL
3	4	9994	2007.0	Sporting Lokeren	LOK
4	5	9984	1750.0	KSV Cercle Brugge	CEB

```
[96]: team.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 299 entries, 0 to 298
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    299 non-null   int64
1   team_api_id           299 non-null   int64
2   team_fifa_api_id      288 non-null   float64
3   team_long_name        299 non-null   object
4   team_short_name       299 non-null   object
dtypes: float64(1), int64(2), object(2)
memory usage: 11.8+ KB
```

```
[97]: team.isnull().sum().sum()
```

```
[97]: 11
```

```
[98]: team.duplicated().sum()
```

```
[98]: 0
```

```
[99]: team.dropna(inplace=True)
```

```
[100]: team.to_csv('Team.csv', index=False)
```

```
[ ]:
```