```
# Carque de Librerías básicas
1
2
    import pandas as pd
3
    import matplotlib.pyplot as plt
4
    import seaborn as sns
5
6
    # Importar tensorflow
7
    import tensorflow as tf
    print("TF version : ", tf.__version__)
8
9
10
    # Necesitaremos GPU
    print("GPU available: ", tf.config.list_physical_devices('GPU'))
11
12
13
    # keras version is 2.11.0
14
    import keras
15
    print("Keras version : ", keras.__version__)
16
17
```

→ TF version : 2.15.0 GPU available: []

Keras version : 2.15.0

```
#-----#
debido a que estoy usando COLAB #
#-----#

from google.colab import drive
drive.mount('/content/drive') #/content/drive/MyDrive/pec2/data/xl.pickle
print("GPU available: ", tf.config.list_physical_devices('GPU'))
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call GPU available: []

```
import pandas as pd

home = '/content/drive/MyDrive/TFM/'

file_path = home + "2017_2023DSTrabajo.xlsx"

dsXls = pd.read_excel(file_path)
dsXls.head(5)
dsXls.info()
```

```
10
11
   12
   # LIMPIEZA DE DATOS
13
   14
   #1. validar duplicados
15
   dsXls.nunique()
16
   #2. validar nulos, rellenar valores faltantes con la mediana
17
18
   #dsXls.isnull().sum()
   dsXls['Dist'].fillna(dsXls['Dist'].median(), inplace=True)
19
   dsXls['Attendance'].fillna(dsXls['Attendance'].median(), inplace=True)
20
21
   dsXls.isnull().sum()
22
23
24
   25
   # ESTADISTICAS
26
   27
   #dsXls.describe().T
28
   dsXls.iloc[:,1:].describe()
```



<class 'pandas.core.frame.DataFrame'> RangeIndex: 4092 entries, 0 to 4091 Data columns (total 21 columns):

| # | Column | Non-Null Count | Dtype | |
|-----|------------|----------------|---------|--|
| | | | | |
| 0 | Date | 4092 non-null | | |
| 1 | Round | 4092 non-null | object | |
| 2 | Day | 4092 non-null | object | |
| 3 | Venue | 4092 non-null | object | |
| 4 | Result | 4092 non-null | object | |
| 5 | GF | 4092 non-null | float64 | |
| 6 | GA | 4092 non-null | float64 | |
| 7 | Opponent | 4092 non-null | object | |
| 8 | xG | 4092 non-null | float64 | |
| 9 | xGA | 4092 non-null | float64 | |
| 10 | Poss | 4092 non-null | float64 | |
| 11 | Attendance | 3212 non-null | float64 | |
| 12 | Season | 4092 non-null | int64 | |
| 13 | Team | 4092 non-null | object | |
| 14 | Sh | 4092 non-null | float64 | |
| 15 | SoT | 4092 non-null | float64 | |
| 16 | Dist | 4089 non-null | float64 | |
| 17 | SCA | 4092 non-null | float64 | |
| 18 | KP | 4092 non-null | float64 | |
| 19 | PPA | 4092 non-null | float64 | |
| 20 | CrsPA | 4092 non-null | float64 | |
| • . | _ | | _ | |

dtypes: datetime64[ns](1), float64(13), int64(1), object(6)

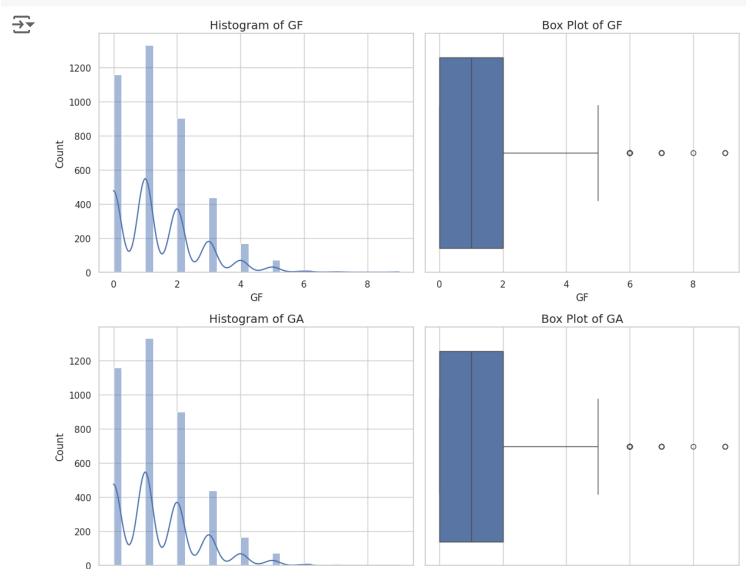
memory usage: 671.5+ KB

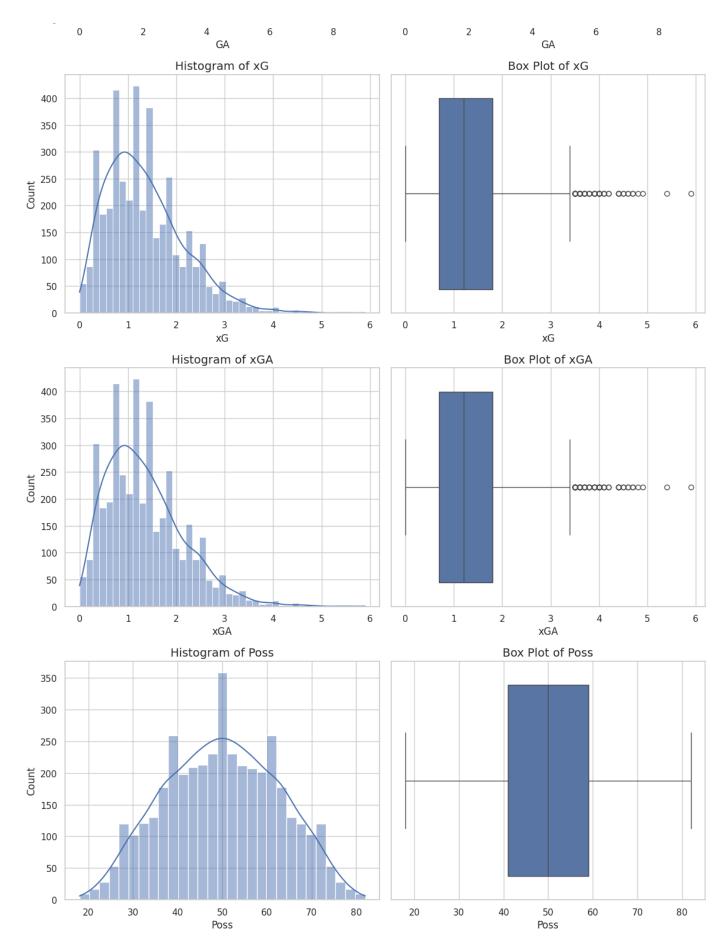
| | GF | GA | хG | xGA | Poss | Attendance | |
|-------|-------------|-------------|-------------|-------------|-------------|--------------|----|
| count | 4092.000000 | 4092.000000 | 4092.000000 | 4092.000000 | 4092.000000 | 4092.000000 | 4(|
| mean | 1.377810 | 1.377810 | 1.346163 | 1.346163 | 50.001222 | 36912.650049 | 20 |
| std | 1.277631 | 1.277631 | 0.796551 | 0.796551 | 12.726702 | 15301.262664 | |
| min | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 18.000000 | 2000.000000 | 20 |
| 25% | 0.000000 | 0.000000 | 0.700000 | 0.700000 | 41.000000 | 29296.000000 | 20 |
| 50% | 1.000000 | 1.000000 | 1.200000 | 1.200000 | 50.000000 | 32092.500000 | 2(|
| 75% | 2.000000 | 2.000000 | 1.800000 | 1.800000 | 59.000000 | 51237.000000 | 20 |
| max | 9.000000 | 9.000000 | 5.900000 | 5.900000 | 82.000000 | 83222.000000 | 20 |

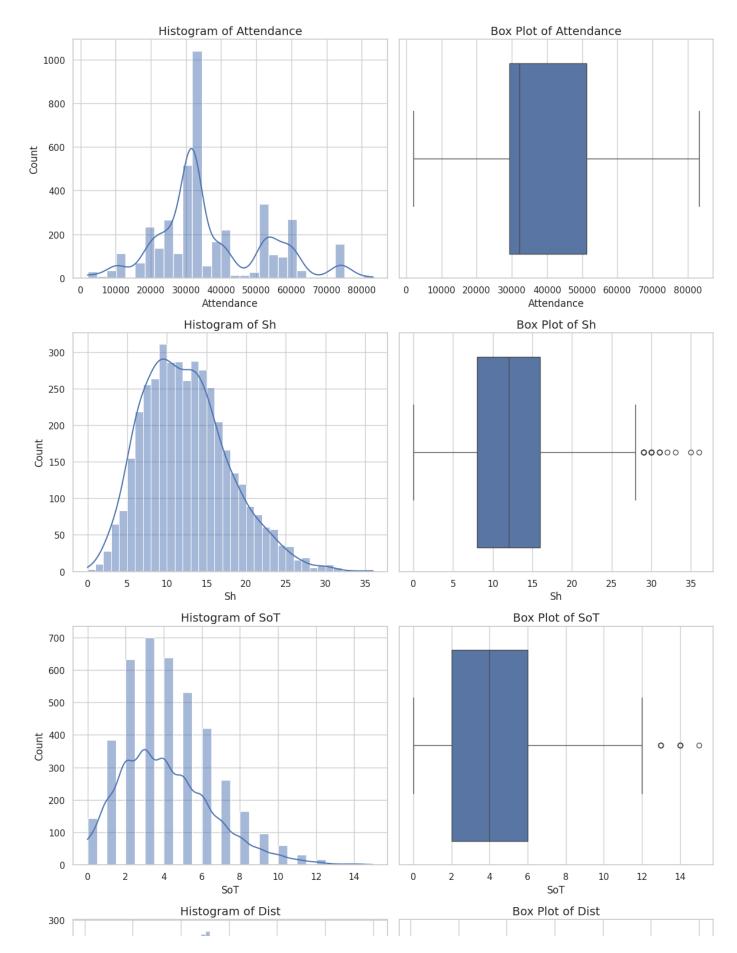
¹ import matplotlib.pyplot as plt

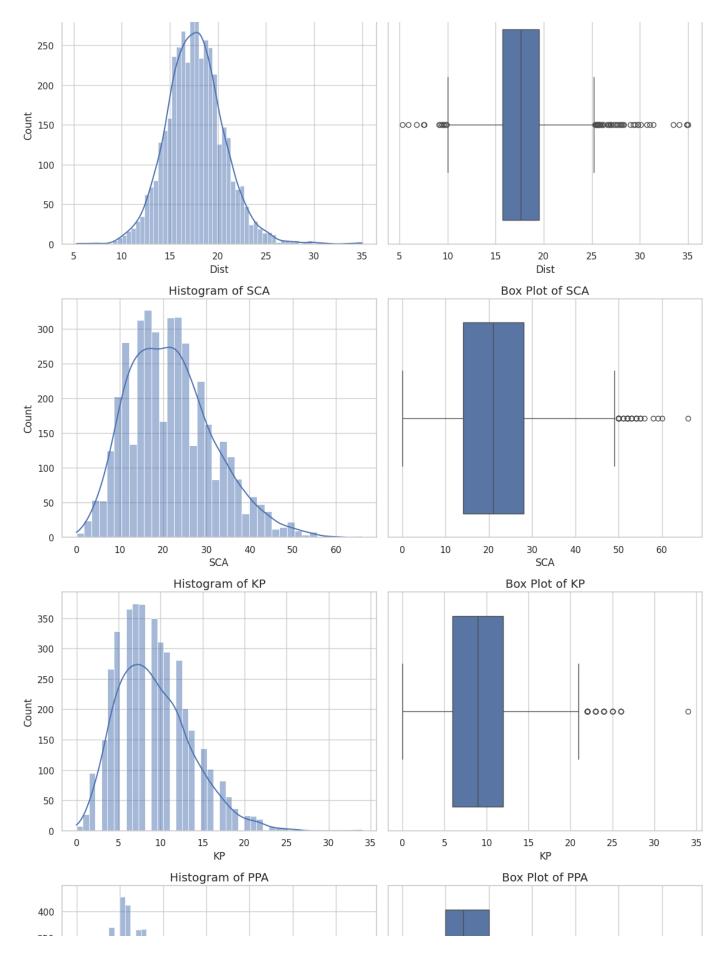
² import seaborn as sns

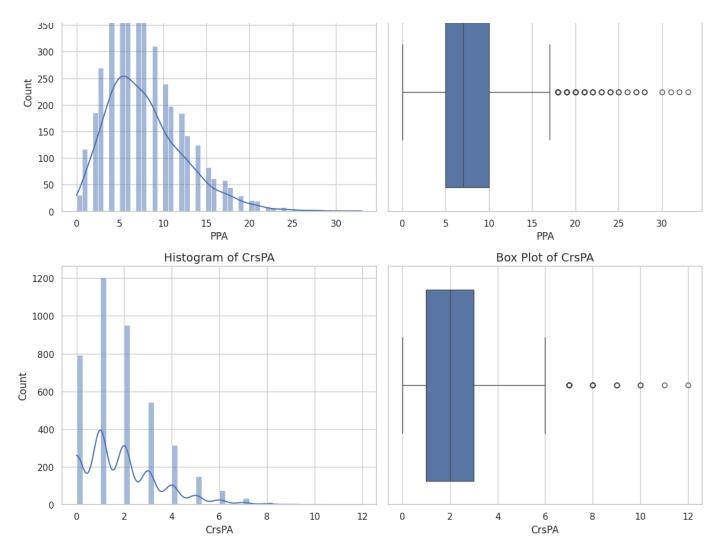
```
3
 4 # Configuración de los estilos de los gráficos
 5 sns.set(style="whitegrid")
 7 # Lista de columnas numéricas para visualizar
8 numeric_columns = ['GF', 'GA', 'xG', 'xGA', 'Poss', 'Attendance', 'Sh', 'SoT',
10 # Creación de histogramas y diagramas de caja para cada variable numérica
11 fig, axes = plt.subplots(len(numeric_columns), 2, figsize=(12, 5 * len(numeric_
12 for i, col in enumerate(numeric_columns):
      sns.histplot(dsXls[col], kde=True, ax=axes[i, 0])
13
14
      axes[i, 0].set_title(f'Histogram of {col}', fontsize=14)
15
      sns.boxplot(x=dsXls[col], ax=axes[i, 1])
      axes[i, 1].set_title(f'Box Plot of {col}', fontsize=14)
16
17
18 plt.tight_layout()
19 plt.show()
```











```
1
    #analisis de goles a favor
2
    import pandas as pd
3
    import matplotlib.pyplot as plt
4
    import seaborn as sns
5
6
    # Suponiendo que 'data' es tu DataFrame
7
    # Estadísticas descriptivas
    ######dsXlsNumeric = dsXls.loc[:, ['GF', 'GA', 'xG', 'xGA', 'Poss', 'Attendanc
8
9
    print(dsXlsNumeric.describe())
    print(dsXlsNumeric.skew()) # Asimetría
10
11
    print(dsXlsNumeric.kurt()) # Curtosis
12
13
    # Histograma para la variable 'GF' (Goles a favor)
14
    plt.figure(figsize=(10, 6))
15
    sns.histplot(dsXlsNumeric['GF'], kde=True)
16
    plt.title('Distribución de Goles a Favor')
17
    plt.xlabel('Goles a Favor')
18
    plt.ylabel('Frecuencia')
19
    plt.show()
20
21
    # Diagrama de caja para 'GF'
22
    plt.figure(figsize=(10, 6))
23
    sns.boxplot(x=dsXlsNumeric['GF'])
24
    plt.title('Box Plot de Goles a Favor')
25
    plt.xlabel('Goles a Favor')
    plt.show()
26
27
\rightarrow
                     GF
                                                                                \
                                  GA
                                                хG
                                                             xGA
                                                                         Poss
```

4092.000000

1.346163

0.796551

0.000000

0.700000

1.200000

1.800000

5.900000

4092.000000

1.346163

0.796551

0.000000

0.700000

1.200000

1.800000

5.900000

4092.000000

50.001222

12.726702

18.000000

41.000000 50.000000

59.000000

82.000000

4092.000000

1.377810

1.277631

0.000000

0.000000

1.000000

2.000000

9.000000

4092.000000

1.377810

1.277631

0.000000

0.00000

1.000000

2.000000

9.000000

count

std

min

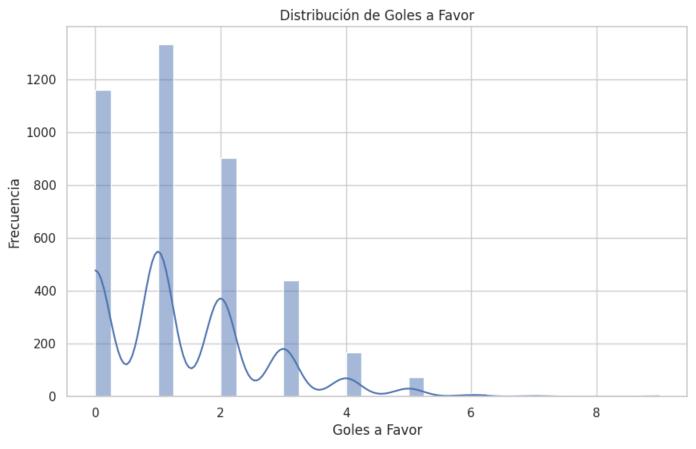
25%

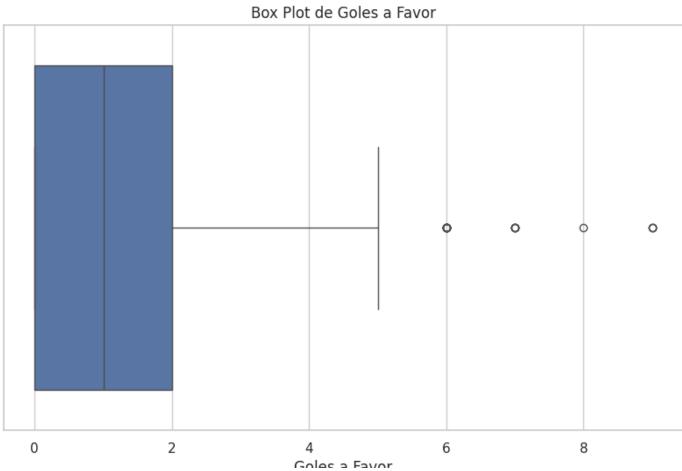
50%

75%

| | | | | 2.,,,,,,, | 02.00000 | |
|--------|--------------|-------------|-------------|-------------|-------------|---|
| | | _ | | | | , |
| | Attendance | Season | Sh | SoT | Dist | \ |
| count | 4092.000000 | 4092.000000 | 4092.000000 | 4092.000000 | 4092.000000 | |
| mean | 36912.650049 | 2019.214076 | 12.317693 | 4.102884 | 17.675318 | |
| std | 15301.262664 | 1.566615 | 5.427259 | 2.437269 | 3.038064 | |
| min | 2000.000000 | 2017.000000 | 0.000000 | 0.00000 | 5.300000 | |
| 25% | 29296.000000 | 2018.000000 | 8.000000 | 2.000000 | 15.700000 | |
| 50% | 32092.500000 | 2019.000000 | 12.000000 | 4.000000 | 17.600000 | |
| 75% | 51237.000000 | 2021.000000 | 16.000000 | 6.000000 | 19.500000 | |
| max | 83222.000000 | 2022.000000 | 36.000000 | 15.000000 | 35.000000 | |
| | | | | | | |
| | SCA | KP | PPA | CrsPA | | |
| count | 4092.000000 | 4092.000000 | 4092.000000 | 4092.000000 | | |
| mean | 21.840176 | 9.215054 | 7.901760 | 1.869501 | | |
| std | 9.897661 | 4.401972 | 4.553291 | 1.613642 | | |
| min | 0.000000 | 0.00000 | 0.00000 | 0.00000 | | |
| 25% | 14.000000 | 6.000000 | 5.000000 | 1.000000 | | |
| 50% | 21.000000 | 9.000000 | 7.000000 | 2.000000 | | |
| 75% | 28.000000 | 12.000000 | 10.000000 | 3.000000 | | |
| max | 66.000000 | 34.000000 | 33.000000 | 12.000000 | | |
| GF | 1.099 | | | | | |
| GA | 1.099 | | | | | |
| xG | 0.919 | | | | | |
| xGA | 0.919 | | | | | |
| Poss | -0.000 | | | | | |
| Attend | | | | | | |
| Season | | | | | | |
| Sh | 0.577 | | | | | |
| SoT | 0.716 | | | | | |
| Dist | 0.457 | | | | | |
| SCA | 0.586 | | | | | |
| KP | 0.666 | | | | | |
| PPA | 0.998 | | | | | |
| CrsPA | 1.215 | 871 | | | | |
| | float64 | | | | | |
| GF | 1.712 | | | | | |
| GA | 1.712 | | | | | |
| xG | 1.066 | | | | | |
| xGA | 1.066 | | | | | |
| Poss | -0.670 | | | | | |
| Attend | | | | | | |
| Season | | | | | | |
| Sh | 0.228 | | | | | |
| SoT | 0.497 | | | | | |
| Dist | 1.744 | | | | | |
| SCA | 0.206 | | | | | |
| KP | 0.448 | | | | | |
| PPA | 1.463 | | | | | |
| CrsPA | 2.104 | 929 | | | | |

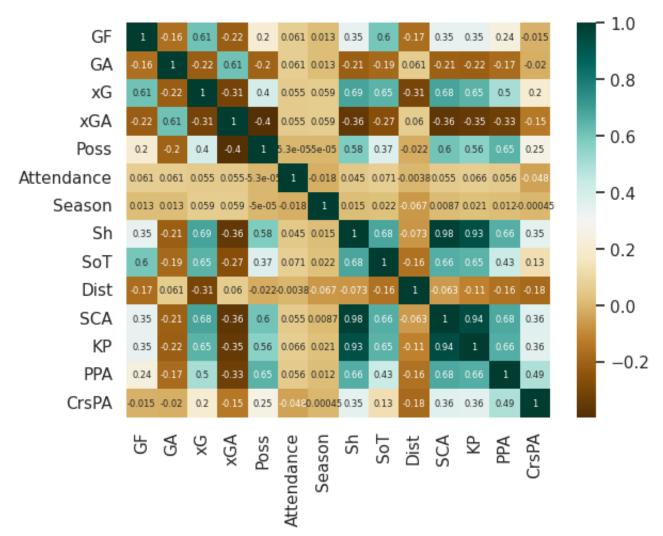
dtype: float64





OUICS a Lavoi

```
1 #CORRELACIÓN ENTRE VARIABLES
2 dsXlsNumeric = dsXls[['GF','GA','xG','xGA','Poss','Attendance','Season','Sh',
3 c= dsXlsNumeric.corr()
4
5 sns.heatmap(c,cmap="BrBG", annot=True, annot_kws={"size": 6})
6
7 plt.figure(figsize=(19, 9))
8
9 plt rcParame("font size") = "2"
```



<Figure size 1900x900 with 0 Axes>