

Date: 10th December, 2020

## XMAS INDIVIDUAL PROJECT

The project is about hypothesis: "The winners of the best running races in the over the world has been won by African athletes".

It's going to show with two datasets: first one is about the winners in 120 years of Olympic Games in the Sport's History, and the second one is about the six best world marathon majors.

### legend file:

- blue box markdown: alert info about the file's content.
- green text: Comments about results.

### ALERT INFO (STEPS)

First of all, It worked with 2 files about tests and tests2 in notebook/ folder where there are different operations, correct and fail code. After, when the code returned correctly, it was transferred and copied in each module appropriate with its function in utils/ folders. In Addition, sometimes it used excel (not much because it has limits) to confirm the results. Finally, they were imported every function with the operations in main.ipynb file in src/ folder.

In [1]:

```
import pandas as pd
import numpy as np
import seaborn as sns #visualisation
import matplotlib.pyplot as plt #visualisation
```

In [2]:

```
import os.path
print(os.path)
#/Users/ariadnapuigventos/Documents/CURSOS/BRIDGE/DS_Ejercicios_Python/BootCamp_TheBridge
/Proyecto_Navidad_Ariadna/src/utils/folders_tb.py
```

```
<module 'posixpath' from '/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9
/posixpath.py'>
```

### Explain the code organization of this file:

It's going to tell about one of two datasets and show the collecting data to understand our hypothesis. Below all these lines, it will show the second datasets with the best insights of Olympic Games Athletes. Finally, It's going to create a new dataframe to try show some similarities to confirm or not hypothesis.

In [3]:

```
from utils.folders_tb import readcsv
#This is one of two dataframes about Best Marathon Majors in all Sport History.
readcsv()
```

	year	winner	gender	country	time	marathon
0	2014	Dennis Kimetto	Male	Kenya	02:02:57	Berlin
1	2011	Geoffrey Mutai	Male	Kenya	02:03:02	Boston
2	2016	Kenenisa Bekele	Male	Ethiopia	02:03:03	Berlin
3	2016	Eliud Kipchoge	Male	Kenya	02:03:05	London
4	2013	Wilson Kipsang	Male	Kenya	02:03:23	Berlin
..	...	...	...	...	...	...
531	1966	Bobbi Gibb	Female	United States	03:21:40	Boston
532	1974	Jutta von Haase	Female	Germany	03:22:01	Berlin
533	1969	Sara Mae Berman	Female	United States	03:22:46	Boston
534	1967	Bobbi Gibb	Female	United States	03:27:17	Boston
535	1968	Bobbi Gibb	Female	United States	03:30:00	Boston

[536 rows x 6 columns]

In [4]:

```
from utils.mining_data_tb import topandtail, dimention

topandtail()
```

	year	winner	gender	country	time	marathon
0	2014	Dennis Kimetto	Male	Kenya	02:02:57	Berlin
1	2011	Geoffrey Mutai	Male	Kenya	02:03:02	Boston
2	2016	Kenenisa Bekele	Male	Ethiopia	02:03:03	Berlin
3	2016	Eliud Kipchoge	Male	Kenya	02:03:05	London
4	2013	Wilson Kipsang	Male	Kenya	02:03:23	Berlin
5	2017	Eliud Kipchoge	Male	Kenya	02:03:32	Berlin
6	2011	Patrick Musyoki	Male	Kenya	02:03:38	Berlin
7	2013	Dennis Kimetto	Male	Kenya	02:03:45	Chicago
8	2017	Wilson Kipsang	Male	Kenya	02:03:58	Tokyo
9	2008	Haile Gebrselassie	Male	Ethiopia	02:03:59	Berlin
.....						
	year	winner	gender	country	time	marathon
531	1966	Bobbi Gibb	Female	United States	03:21:40	Boston
532	1974	Jutta von Haase	Female	Germany	03:22:01	Berlin
533	1969	Sara Mae Berman	Female	United States	03:22:46	Boston
534	1967	Bobbi Gibb	Female	United States	03:27:17	Boston
535	1968	Bobbi Gibb	Female	United States	03:30:00	Boston

In [5]:

```
dimention()
```

```
(536, 6)
number of duplicate rows: Empty DataFrame
Columns: [year, winner, gender, country, time, marathon]
Index: []
```

	year	winner	gender	country	time	marathon
0	2014	Dennis Kimetto	Male	Kenya	02:02:57	Berlin
1	2011	Geoffrey Mutai	Male	Kenya	02:03:02	Boston
2	2016	Kenenisa Bekele	Male	Ethiopia	02:03:03	Berlin
3	2016	Eliud Kipchoge	Male	Kenya	02:03:05	London
4	2013	Wilson Kipsang	Male	Kenya	02:03:23	Berlin
..	...	...	...	...	...	...
531	1966	Bobbi Gibb	Female	United States	03:21:40	Boston
532	1974	Jutta von Haase	Female	Germany	03:22:01	Berlin
533	1969	Sara Mae Berman	Female	United States	03:22:46	Boston
534	1967	Bobbi Gibb	Female	United States	03:27:17	Boston
535	1968	Bobbi Gibb	Female	United States	03:30:00	Boston

```
[536 rows x 6 columns]
(536, 6)
```

## ALERT INFO (STEPS)

The Dataframe has not any duplicates but there are some values equality. It needs to check what it means because it's possible some majors who has already won more than one marathons, that's why it's going to show using the method `values_counts` by country and winner.

In [6]:

```
from utils.mining_data_tb import repite_pais, repetidores

# Effectively, the most country that's repeat is Kenya on the top, below United States and Ethiopia.
# It's considering that Kenya started to compete in 1960 and United States since 1896.

repite_pais()
```

Kenya	136
United States	104
Ethiopia	51
Germany	36
United Kingdom	35

Japan	22
Norway	20
Canada	17
Portugal	11
Mexico	10
Finland	10
Russia	8
Poland	8
Brazil	7
Italy	6

Name: country, dtype: int64

In [7]:

```
from utils.visualization_tb import piechart_repitpais
```

*#Thanks to this pie chart graphic it's showing that Kenya is the country winner with 136 marathons, it's 25,4% of the total of the competition. In addition, the third country is Ethiopia with aprox 10%, so if it's talking about African Athletes are winners of the competition for a aprox. 35% of the total pie chart. For curiosity, only there was 1 Spanish athlete who won 2 World Marathons: Berlin 1996 and London 1998 with the best time 2:09:15 and 2:07:57, respectively.*

```
piechart_repitpais()
```

```
0      0 days 02:02:57
1      0 days 02:03:02
2      0 days 02:03:03
3      0 days 02:03:05
4      0 days 02:03:23
...
531    0 days 03:21:40
532    0 days 03:22:01
533    0 days 03:22:46
534    0 days 03:27:17
535    0 days 03:30:00
Name: time, Length: 536, dtype: timedelta64[ns]
[136, 104, 51, 36, 35, 22, 20, 17, 11, 10, 10, 8, 8, 7, 6, 5, 5, 5, 5, 4, 3, 3, 3, 2, 2, 2, 2, 2, 2, 2, 2, 2, 1, 1, 1, 1, 1, 1]
```

In [8]:

```
repetidores()
```

Grete Waitz	11
Ingrid Kristiansen	8
Bill Rodgers	8
Paula Radcliffe	7
Clarence DeMar	7
Uta Pippig	7
Mary Keitany	6
Eliud Kipchoge	6
Catherine Ndereba	6
Rosa Mota	6
Joyce Chepchumba	5
Khalid Khannouchi	5
Martin Lel	5
Wilson Kipsang	5
Gérard Côté	4

Name: winner, dtype: int64

In [9]:

```
#AQUÍ VA GENDER DATA CLASIFICATION!!!!
```

In [10]:

```
from utils.mining_data_tb import checkingdata
```

*#It wants to know how are the values because it has seen that there is one about time.*

```
checkingdata()
```

```
year          int64
winner        object
gender        object
country       object
time          object
marathon      object
dtype: object
```

### ALERT INFO (STEPS)

It needs to change some data rows after to see time column in dataframe is an object. It will be necessary to change from a object to pd.to\_timedelta and after from timedelta to float64 with method "timedelta64[s]" for detecting some outliers and for doing to histogram bins=5.

In [11]:

```
from utils.mining_data_tb import changetype
```

*#With this fuction it changed from object time column with seconds to use it in boxplot f or detecting outliers.*

```
changetype()
```

```
0      0 days 02:02:57
1      0 days 02:03:02
2      0 days 02:03:03
3      0 days 02:03:05
4      0 days 02:03:23
...
531    0 days 03:21:40
532    0 days 03:22:01
533    0 days 03:22:46
534    0 days 03:27:17
535    0 days 03:30:00
Name: time, Length: 536, dtype: timedelta64[ns]
0          7377.0
1          7382.0
2          7383.0
3          7385.0
4          7403.0
...
531        12100.0
532        12121.0
533        12166.0
534        12437.0
535        12600.0
Name: time, Length: 536, dtype: float64
```

In [12]:

```
from utils.visualization_tb import detect_outliers
```

*#2 extreams: the first time was 2:02:57 by Kenian Athlete in Berlin Marathon in 2014; and the last time was 3:30:00 by United States Athlete in Boston Marathon in 1968. Althought, 25% Marathon majors got a median around 2 hours and 16 minuts and the most majors with 75 % got 2 hours and 46 minuts.*

```
detect_outliers()
```

```
AxesSubplot(0.125,0.125;0.775x0.755)
7783.0
8856.25
1073.25
```

### ALERT INFO (STEPS)

It's showing the histogram of each column. In this case, every columns fo the World Marathon Majors Dataframe, except Year, one hand, has been changed by astype "Category" because they were object types;

and the other hand, Time Column has been changed from `pd.to_timedelta` to `timedelta64[s]` because It needs in seconds to showing in histogram. It shows how the ranges are different between them.

In [13]:

```
from utils.visualization_tb import histogram_time, histogram_year_time

#There are more participation years later than the begginers of competition when only Can
ada and United States were winners for a long time ago consecutively.

histogram_time()
histogram_year_time()

[<AxesSubplot:title={'center':'year'}>
 <AxesSubplot:title={'center':'time'}>]]
```

In [14]:

```
from utils.visualization_tb import histogram_gender

#Only there is a different of 13% (70 World Marathons) more won them by Male athletes wit
h 56,5% (303 WM) than Female with 43,4% (233 WM). It's a great appreciation because it no
tices that the first woman athlete who could compete was in 1967 (71 years after than men
).

histogram_gender()

AxesSubplot(0.125,0.125;0.775x0.755)
```

In [15]:

```
from utils.visualization_tb import histogram_country

#It is not showing the real situation with bins=5, below these lines it's changed to a hi
stogram with bins=37 (total countries).

histogram_country()

AxesSubplot(0.125,0.125;0.775x0.755)
Get better another argument to see almost it
```

In [16]:

```
from utils.visualization_tb import histogram_countryby37bins

#This graphic is showing how there are 2 countries stand out (Kenia and Ethiopia) versus
of the rest countries.

histogram_countryby37bins()

AxesSubplot(0.125,0.125;0.775x0.755)
```

