cosijoeza

November 23, 2024

[212]:
```
import pandas as pd
import seaborn as sns
import numpy as np
from matplotlib import pyplot as plt
from statsmodels.graphics.gofplots import qqplot
```

1 Hipertensión Arterial en México

1.1 Lectura y despliegue de información de la base de datos

[213]:
```
dataHipertention = pd.read_csv('hipertension-arterial-mexico.csv')
```

[214]:
```
dataHipertention.head()
```

[214]:

	FOLIO_I	sexo	edad	concentracion_hemoglobina	temperatura_ambiente	\
0	2022_01001004	2	41	14.2	22	
1	2022_01001009	2	65	14.1	9	
2	2022_01001012	2	68	14.2	22	
3	2022_01001013	1	35	15.7	11	
4	2022_01001015	2	65	12.7	7	

	valor_acido_urico	valor_albumina	valor_colesterol_hdl	\
0	4.8	4.0	34	
1	4.4	3.8	73	
2	4.8	4.0	34	
3	6.5	4.1	49	
4	4.2	4.2	41	

	valor_colesterol_ldl	valor_colesterol_total	…	segundamedicion_peso	\
0	86.0	139	…	64.70	
1	130.0	252	…	96.75	
2	86.0	139	…	68.70	
3	107.0	203	…	64.70	
4	76.0	145	…	97.15	

	segundamedicion_estatura	distancia_rodilla_talon	\
0	154.0	48.5	

```
1                    152.2                    44.5
2                    144.8                    42.3
3                    154.0                    48.5
4                    161.3                    49.6

   circunferencia_de_la_pantorrilla  segundamedicion_cintura  \
0                              33.5                      0.0
1                              41.1                    113.7
2                              37.8                    103.7
3                              33.5                      0.0
4                              42.0                    118.9

   tension_arterial  sueno_horas  masa_corporal  actividad_total  \
0               107            4      32.889389              120
1               104            2       1.000000              240
2               105            1       1.000000              480
3               117            5      26.265339              275
4               123            2       1.000000              255

   riesgo_hipertension
0                    1
1                    0
2                    0
3                    1
4                    0

[5 rows x 36 columns]
```

[215]:
```
dataHipertention.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4363 entries, 0 to 4362
Data columns (total 36 columns):
 #   Column                           Non-Null Count  Dtype
---  ------                           --------------  -----
 0   FOLIO_I                          4363 non-null   object
 1   sexo                             4363 non-null   int64
 2   edad                             4363 non-null   int64
 3   concentracion_hemoglobina        4363 non-null   float64
 4   temperatura_ambiente             4363 non-null   int64
 5   valor_acido_urico                4363 non-null   float64
 6   valor_albumina                   4363 non-null   float64
 7   valor_colesterol_hdl             4363 non-null   int64
 8   valor_colesterol_ldl             4363 non-null   float64
 9   valor_colesterol_total           4363 non-null   int64
 10  valor_creatina                   4363 non-null   float64
 11  resultado_glucosa                4363 non-null   float64
```

```
 12  valor_insulina                    4363 non-null   float64
 13  valor_trigliceridos               4363 non-null   int64
 14  resultado_glucosa_promedio        4363 non-null   int64
 15  valor_hemoglobina_glucosilada     4363 non-null   float64
 16  valor_ferritina                   4363 non-null   float64
 17  valor_folato                      4363 non-null   float64
 18  valor_homocisteina                4363 non-null   float64
 19  valor_proteinac_reactiva          4363 non-null   float64
 20  valor_transferrina                4363 non-null   float64
 21  valor_vitamina_bdoce              4363 non-null   float64
 22  valor_vitamina_d                  4363 non-null   float64
 23  peso                              4363 non-null   float64
 24  estatura                          4363 non-null   float64
 25  medida_cintura                    4363 non-null   float64
 26  segundamedicion_peso              4363 non-null   float64
 27  segundamedicion_estatura          4363 non-null   float64
 28  distancia_rodilla_talon           4363 non-null   float64
 29  circunferencia_de_la_pantorrilla  4363 non-null   float64
 30  segundamedicion_cintura           4363 non-null   float64
 31  tension_arterial                  4363 non-null   int64
 32  sueno_horas                       4363 non-null   int64
 33  masa_corporal                     4363 non-null   float64
 34  actividad_total                   4363 non-null   int64
 35  riesgo_hipertension               4363 non-null   int64
dtypes: float64(24), int64(11), object(1)
memory usage: 1.2+ MB
```

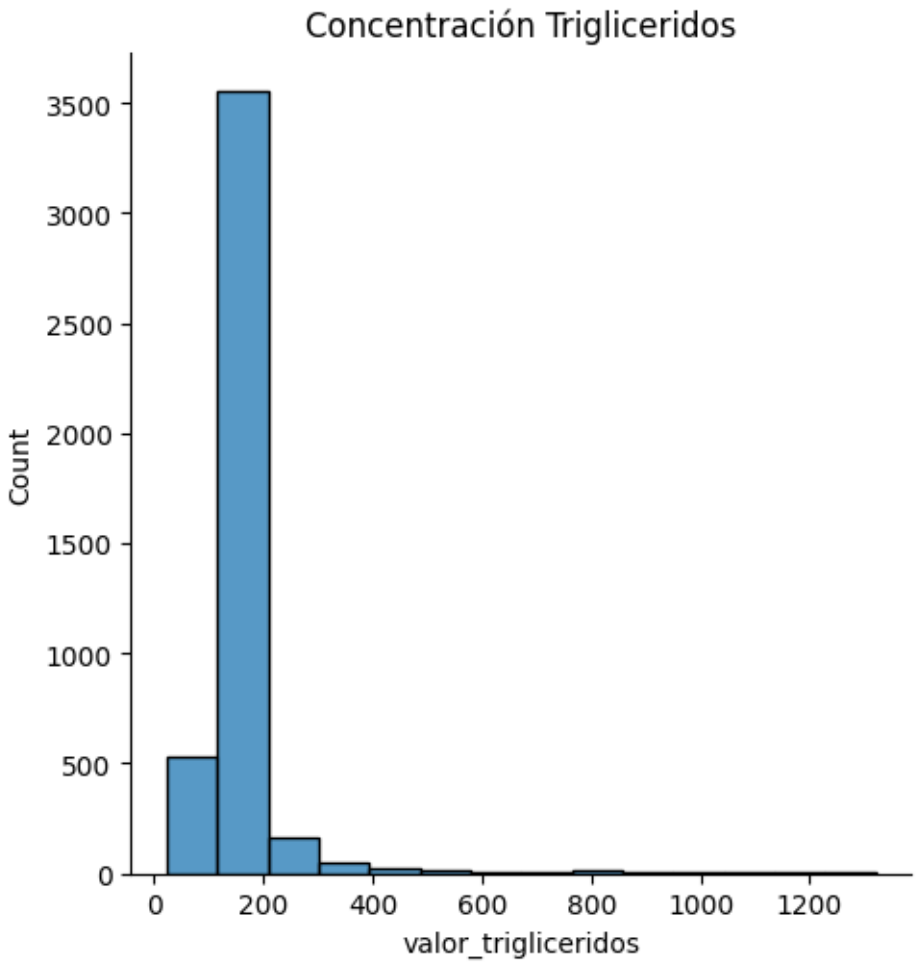
1.2 Grafica Concentración de Trigliceridos

Esta representación nos muestra que los valores de los trigliceridos tiene más valores entre 0 y 200.

[216]:
```
df = dataHipertention['valor_trigliceridos']
```

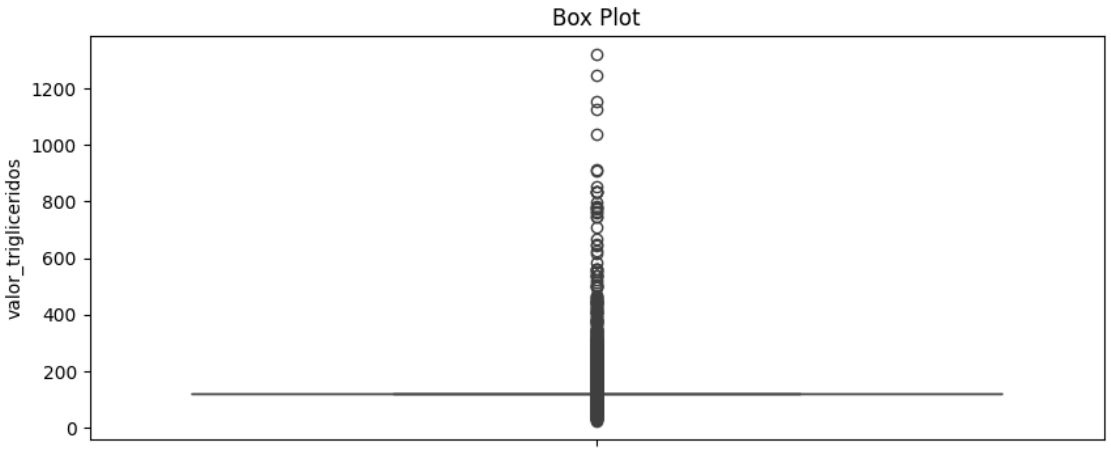
[217]:
```
plt.figure(figsize=(10,4))
sns.displot(df)
plt.title("Concentración Trigliceridos")
sns.despine()
plt.show()
```

```
<Figure size 1000x400 with 0 Axes>
```



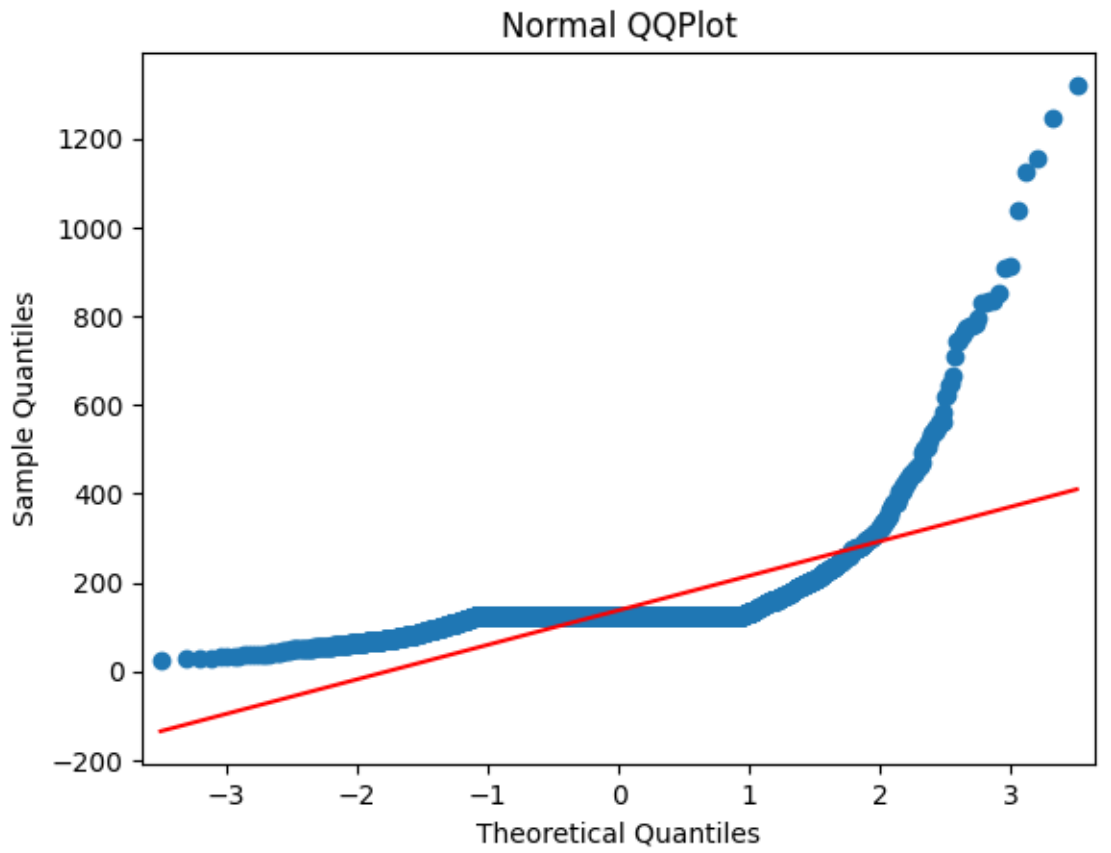
1.3 Busqueda de Outliers en los Trigliceridos

[218]:
```
plt.figure(figsize=(10,4))
plt.title("Box Plot")
sns.boxplot(df)
plt.show()
```



[219]:
```
plt.figure(figsize=(10,4))
qqplot(df,line='s')
plt.title("Normal QQPlot")
plt.show()
```

```
<Figure size 1000x400 with 0 Axes>
```



1.4 ¿Cuáles son los outliers y cuántos son?

[220]:
```
out = []
def Zscore_outlier(df,umbral):
  mean = np.mean(df)
  standarDesviation = np.std(df)
  for i in df:
      z = (i - mean) / standarDesviation
      if np.abs(z) > umbral:
        out.append(i)
  print("Outliers: ",out)
  return out
outliers = Zscore_outlier(df,umbral=3)
```

```
Outliers:  [376, 379, 452, 382, 507, 443, 438, 563, 773, 563, 408, 445, 423,
431, 835, 563, 440, 1040, 390, 832, 453, 456, 393, 500, 797, 446, 777, 582, 550,
432, 408, 382, 767, 373, 550, 1245, 758, 624, 445, 832, 499, 524, 540, 423, 375,
744, 835, 409, 470, 649, 745, 516, 417, 463, 461, 784, 618, 1320, 408, 852, 778,
1124, 910, 441, 501, 540, 418, 494, 407, 1154, 382, 400, 466, 667, 710, 536,
644, 381, 914, 537]
```

[221]:
```
print(len(outliers))
```

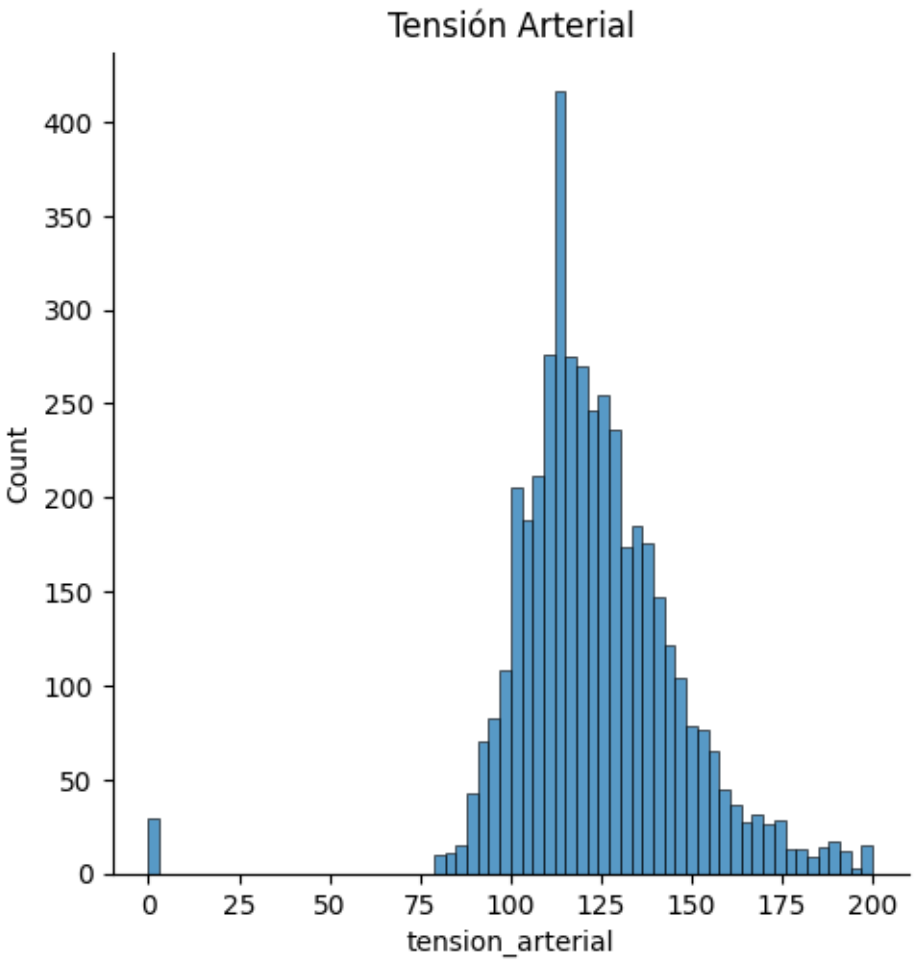
```
80
```

1.5 Gráfica Concentración de Tensión Arterial

[222]:
```
df = dataHipertention['tension_arterial']
```

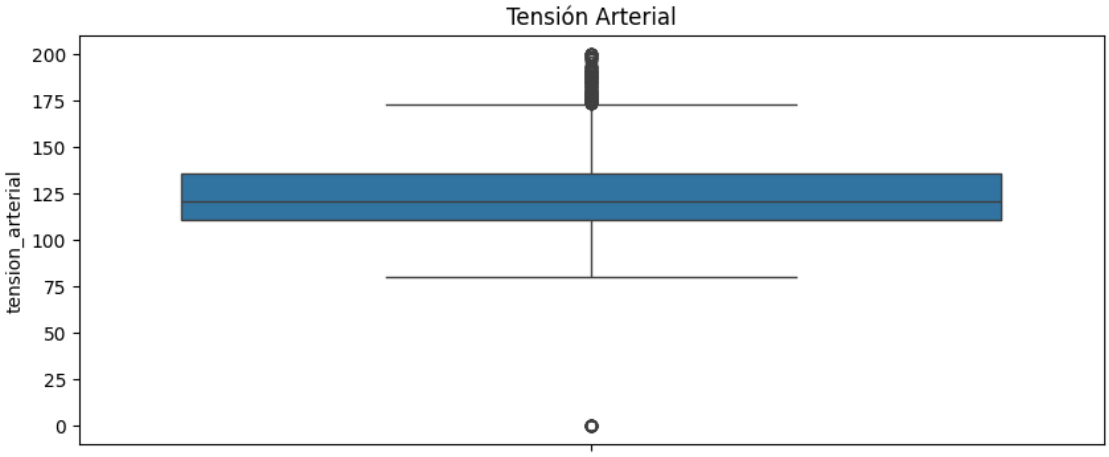
[223]:
```
plt.figure(figsize=(10,4))
sns.displot(df)
plt.title("Tensión Arterial")
sns.despine()
plt.show()
```

```
<Figure size 1000x400 with 0 Axes>
```



1.6 Busqueda de Outliers en la Tensión Arterial

[224]:
```
plt.figure(figsize=(10,4))
plt.title("Tensión Arterial")
sns.boxplot(df)
plt.show()
```



[225]:
```
plt.figure(figsize=(10,4))
sns.distplot(df)
plt.title("Tensión Arterial")
sns.despine()
plt.show()
```

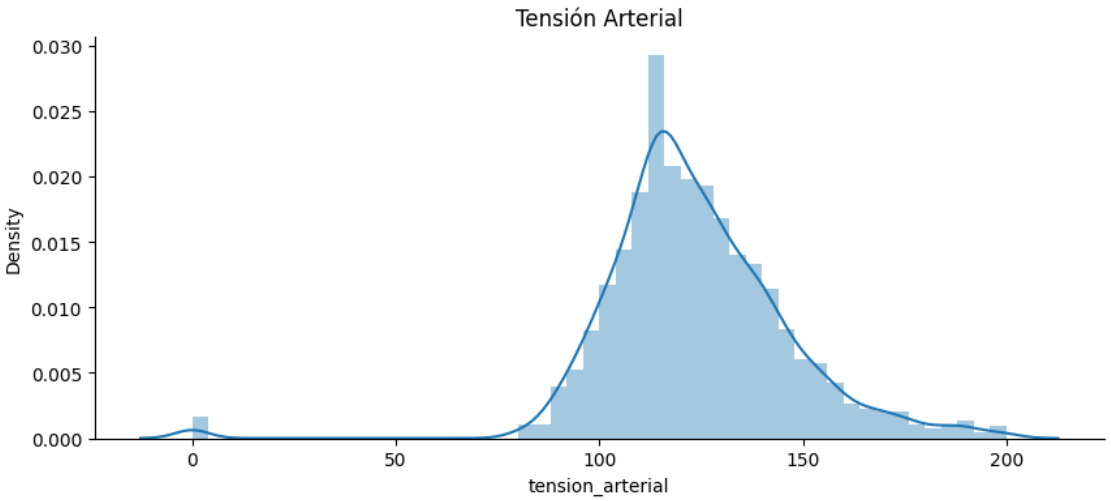
```
<ipython-input-225-fbbd1529f507>:2: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with
similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

  sns.distplot(df)
```



1.7 ¿Cuáles son los outliers y cuántos son?

[226]:
```
outliers = Zscore_outlier(df,umbral=3)
```

```
Outliers:  [376, 379, 452, 382, 507, 443, 438, 563, 773, 563, 408, 445, 423,
431, 835, 563, 440, 1040, 390, 832, 453, 456, 393, 500, 797, 446, 777, 582, 550,
432, 408, 382, 767, 373, 550, 1245, 758, 624, 445, 832, 499, 524, 540, 423, 375,
744, 835, 409, 470, 649, 745, 516, 417, 463, 461, 784, 618, 1320, 408, 852, 778,
1124, 910, 441, 501, 540, 418, 494, 407, 1154, 382, 400, 466, 667, 710, 536,
644, 381, 914, 537, 0, 0, 0, 193, 198, 0, 0, 0, 200, 193, 193, 200, 200, 0, 0,
0, 198, 199, 194, 0, 200, 0, 0, 192, 0, 0, 0, 199, 0, 197, 194, 0, 0, 0, 0, 0,
198, 192, 0, 200, 0, 197, 0, 200, 200, 0, 196, 0, 0, 200, 0, 0, 193]
```

[227]:
```
outliers.sort()
print(outliers)
```

```
[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 192, 192, 193, 193, 193, 193, 194, 194, 196, 197, 197, 198, 198, 198,
199, 199, 200, 200, 200, 200, 200, 200, 200, 200, 373, 375, 376, 379, 381, 382,
382, 382, 390, 393, 400, 407, 408, 408, 408, 409, 417, 418, 423, 423, 431, 432,
438, 440, 441, 443, 445, 445, 446, 452, 453, 456, 461, 463, 466, 470, 494, 499,
500, 501, 507, 516, 524, 536, 537, 540, 540, 550, 550, 563, 563, 563, 582, 618,
624, 644, 649, 667, 710, 744, 745, 758, 767, 773, 777, 778, 784, 797, 832, 832,
835, 835, 852, 910, 914, 1040, 1124, 1154, 1245, 1320]
```

2 COVID-19 en Brazil

2.1 Lectura y despliegue de información de la base de datos

[228]:
```
brazilCovid = pd.read_csv('brazil-covid19.csv')
```

[236]:
```
brazilCovid.head()
```

[236]:

	date	hour	state	suspects	refuses	cases	deaths
0	2020-01-30	16:00	Minas Gerais	1	0	0	0
1	2020-01-30	16:00	Rio de Janeiro	1	0	0	0
2	2020-01-30	16:00	Santa Catarina	0	2	0	0
3	2020-01-30	16:00	São Paulo	3	1	0	0
4	2020-01-30	16:00	Rio Grande do Sul	2	2	0	0

[230]:
```
brazilCovid.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1008 entries, 0 to 1007
Data columns (total 7 columns):
 #   Column    Non-Null Count  Dtype 
---  ------    --------------  ----- 
 0   date      1008 non-null   object
 1   hour      684 non-null    object
 2   state     1008 non-null   object
 3   suspects  1008 non-null   int64 
 4   refuses   1008 non-null   int64 
 5   cases     1008 non-null   object
 6   deaths    1008 non-null   int64 
dtypes: int64(3), object(4)
memory usage: 55.2+ KB
```

3 Terrorismo Global

3.1 Lectura y despliegue de información de la base de datos

[232]:
```
terrorismoGlobal = pd.read_csv('global-terrorism.csv',encoding='latin-1')
```

```
<ipython-input-232-de8b786cd3cf>:1: DtypeWarning: Columns
(4,31,33,62,76,79,94,96,121) have mixed types. Specify dtype option on import or
set low_memory=False.
  terrorismoGlobal = pd.read_csv('global-terrorism.csv',encoding='latin-1')
```

[233]:
```
terrorismoGlobal.head()
```

[233]:

	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	\
0	197000000001	1970	7	2	NaN	0	NaN	58	
1	197000000002	1970	0	0	NaN	0	NaN	130	

```
2  197001000001   1970       1     0        NaN         0        NaN      160
3  197001000002   1970       1     0        NaN         0        NaN       78
4  197001000003   1970       1     0        NaN         0        NaN      101

          country_txt  region  … addnotes scite1 scite2  scite3  dbsource  \
0  Dominican Republic       2  …      NaN    NaN    NaN     NaN      PGIS
1              Mexico       1  …      NaN    NaN    NaN     NaN      PGIS
2         Philippines       5  …      NaN    NaN    NaN     NaN      PGIS
3              Greece       8  …      NaN    NaN    NaN     NaN      PGIS
4               Japan       4  …      NaN    NaN    NaN     NaN      PGIS

   INT_LOG  INT_IDEO INT_MISC INT_ANY  related
0      0.0       0.0      0.0     0.0      NaN
1      0.0       1.0      1.0     1.0      NaN
2     -9.0      -9.0      1.0     1.0      NaN
3     -9.0      -9.0      1.0     1.0      NaN
4     -9.0      -9.0      1.0     1.0      NaN

[5 rows x 135 columns]
```

[239]:
```
terrorismoGlobal.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18870 entries, 0 to 18869
Columns: 135 entries, eventid to related
dtypes: float64(57), int64(22), object(56)
memory usage: 19.4+ MB
```

[ ]:
```

```