

```
In [2]: #iberias generales
import numpy as np
import pandas as pd

#librerias graficas
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.graph_objects as go

#modelo
from sklearn.cluster import KMeans, AgglomerativeClustering
from sklearn.mixture import GaussianMixture

#
from yellowbrick.cluster import KElbowVisualizer
from scipy.cluster.hierarchy import dendrogram, linkage

#preprocesamiento
from sklearn.preprocessing import StandardScaler

#otros
import warnings
warnings.filterwarnings('ignore')
```

## Carga de datos

```
In [2]: #Carga de dataset
df = pd.read_csv('https://raw.githubusercontent.com/SaulSebastian/dataset/main')
```

## Analisis

```
In [3]: #Muestreo aleatorio
df.sample(5)
```

```
Out[3]:
```

	cases	deaths	countriesAndTerritories	continentExp	Cumulative_number_for_19_
<b>345</b>	0	0	Brazil	America	
<b>33578</b>	15	0	Malta	Europe	
<b>2531</b>	0	0	Philippines	Asia	
<b>51280</b>	0	0	Bermuda	America	
<b>34112</b>	0	0	Brunei Darussalam	Asia	

```
In [4]: #Visualización de los tipos de datos
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 61900 entries, 0 to 61899
Data columns (total 7 columns):
#   Column                                     Non-Null Co
unt  Dtype                                     unt
---  ---
0    cases                                     61900 non-n
ull  int64
1    deaths                                   61900 non-n
ull  int64
2    countriesAndTerritories                 61900 non-n
ull  object
3    continentExp                           61900 non-n
ull  object
4    Cumulative_number_for_14_days_of_COVID-19_cases_per_100000 59021 non-n
ull  float64
5    PORC_SOBREPESO_F                       61900 non-n
ull  float64
6    PORC_SOBREPESO_M                       61900 non-n
ull  float64
dtypes: float64(3), int64(2), object(2)
memory usage: 3.3+ MB

```

```

In [5]: #Verificar la existencia y cantidad de null's
df.isna().sum()

```

```

Out[5]: cases                                     0
deaths                                           0
countriesAndTerritories                         0
continentExp                                    0
Cumulative_number_for_14_days_of_COVID-19_cases_per_100000 2879
PORC_SOBREPESO_F                                0
PORC_SOBREPESO_M                                0
dtype: int64

```

```

In [6]: #Extraer variables numericas
var_num = [var for var in df.columns if df[var].dtypes != 'O']
var_cat = [var for var in df.columns if df[var].dtypes == 'O']

```

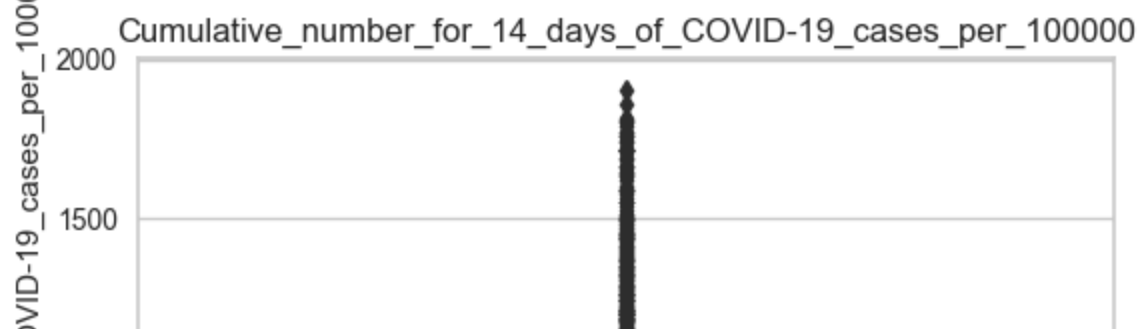
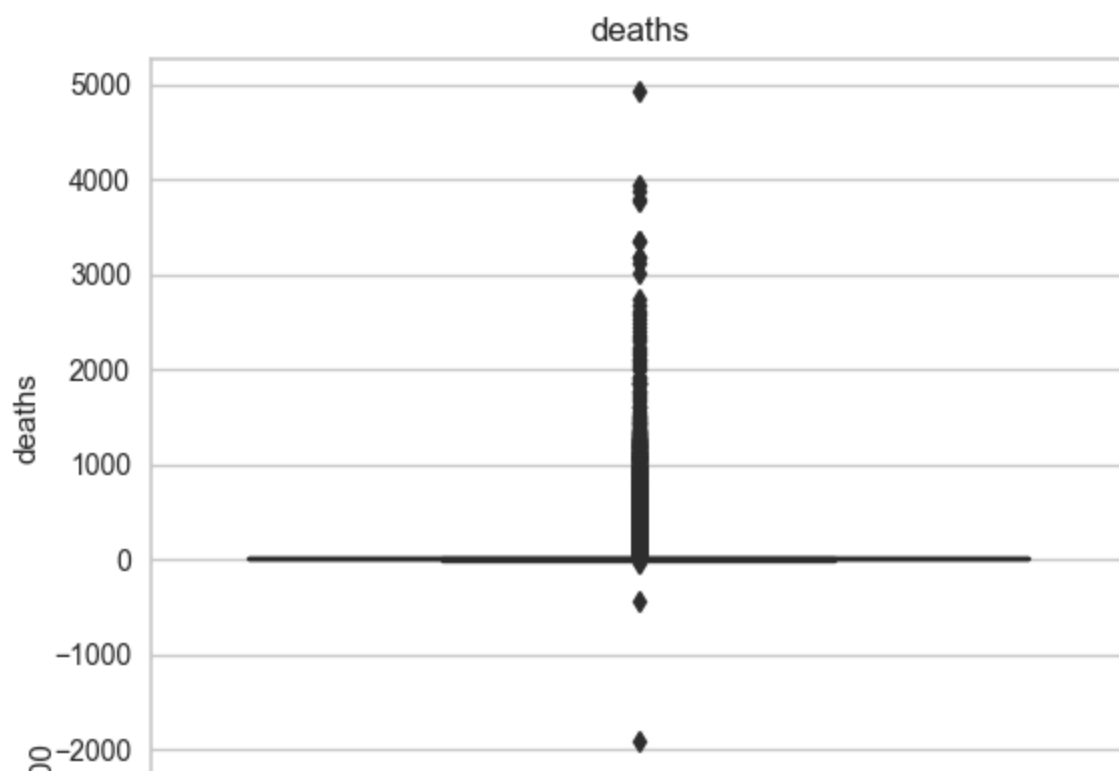
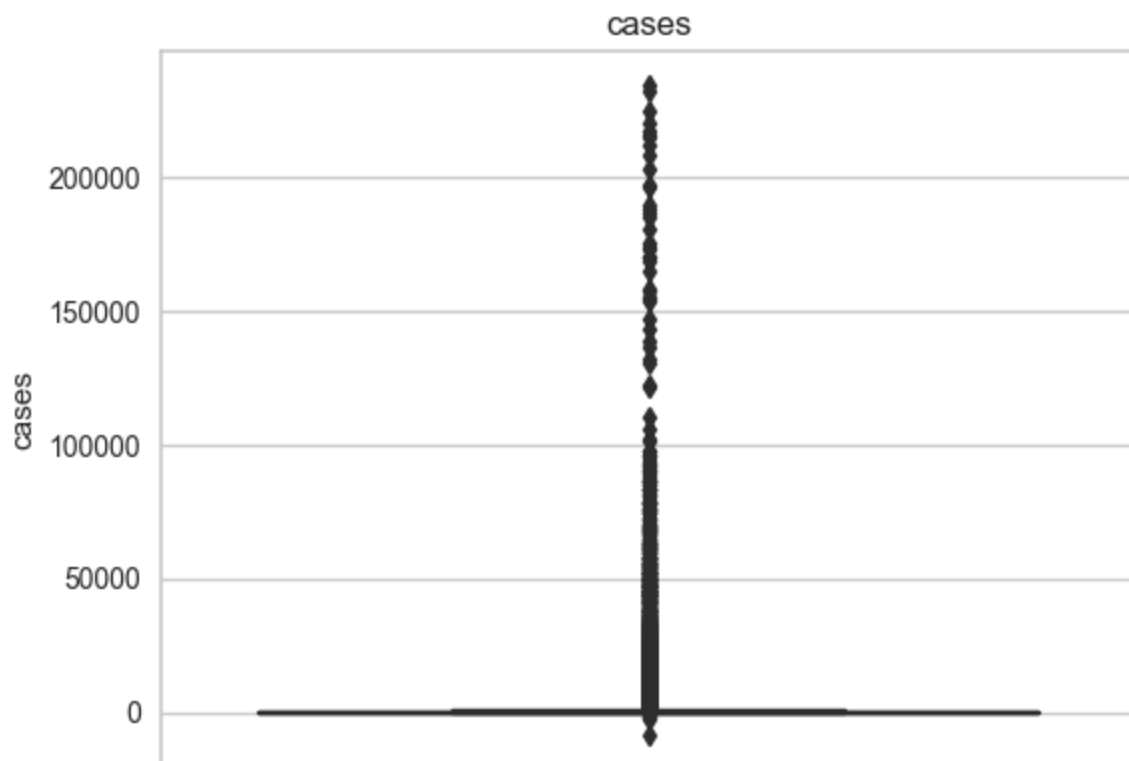
```

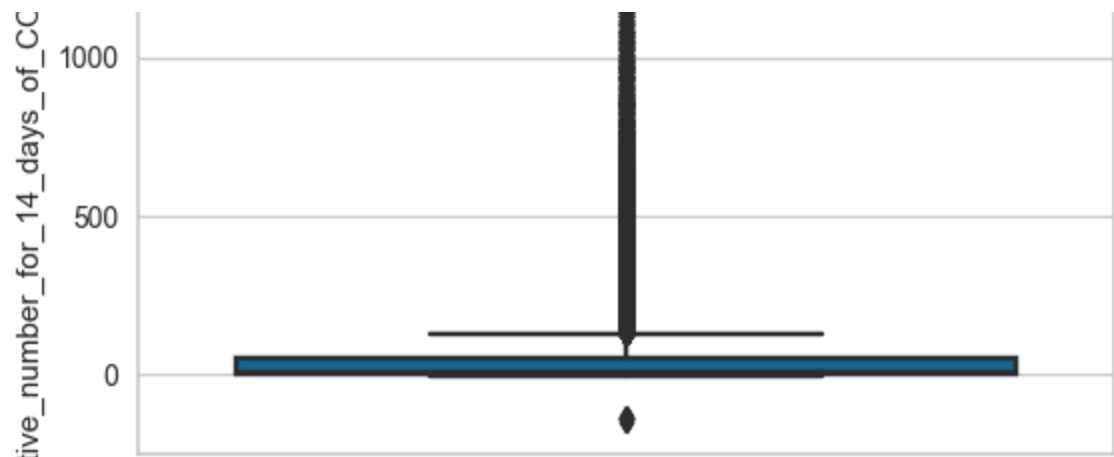
In [7]: # Crear la figura con subplots
fig, axs = plt.subplots(len(var_num), 1, figsize=(6, 4*len(var_num)))
axs = axs.flatten()

# Crear los gráficos de caja para las variables numéricas
for i, col in enumerate(var_num):
    sns.boxplot(y=col, data=df, ax=axs[i])
    axs[i].set_title(f"{col}")

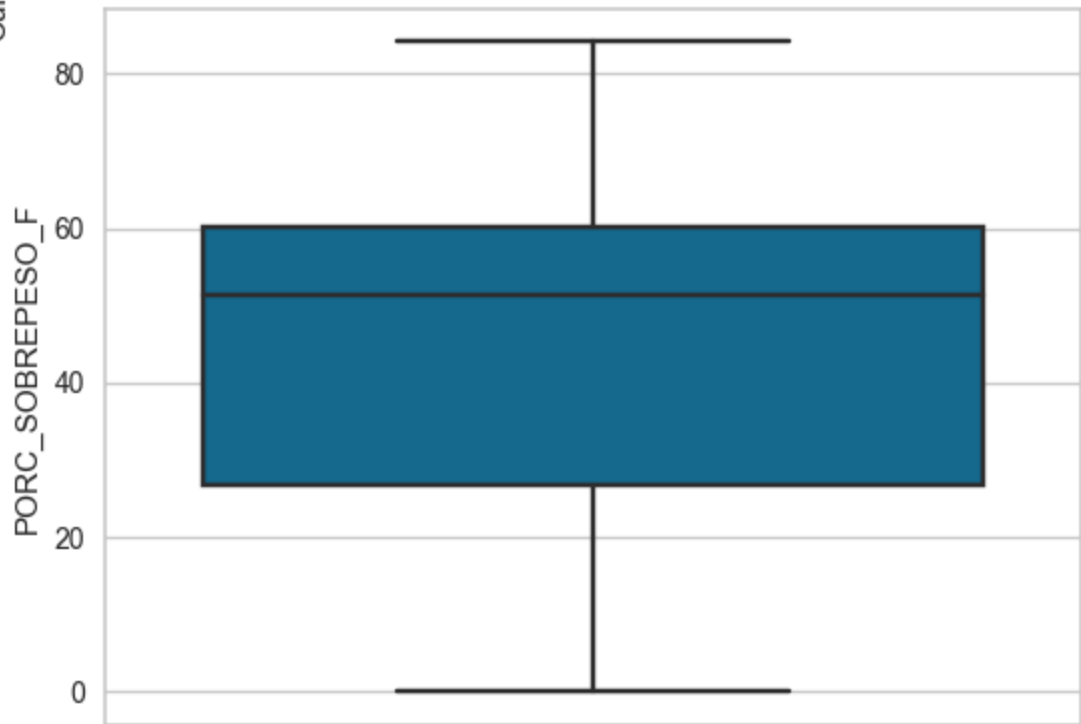
# Ajustar los subplots y mostrar la figura
fig.tight_layout()
plt.show()

```

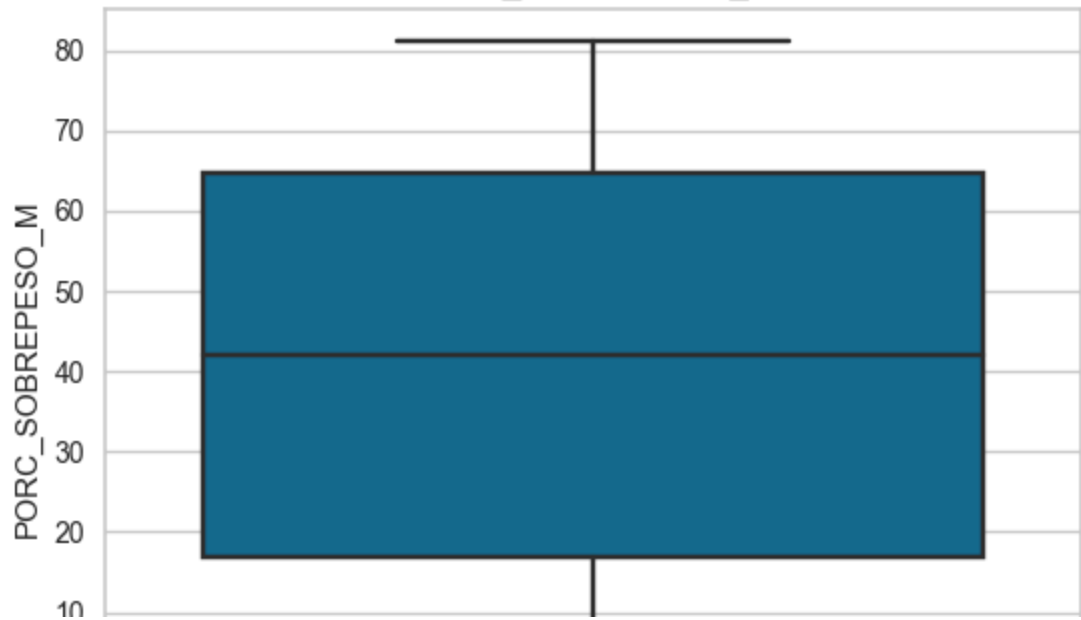




PORC\_SOBREPESO\_F



PORC\_SOBREPESO\_M

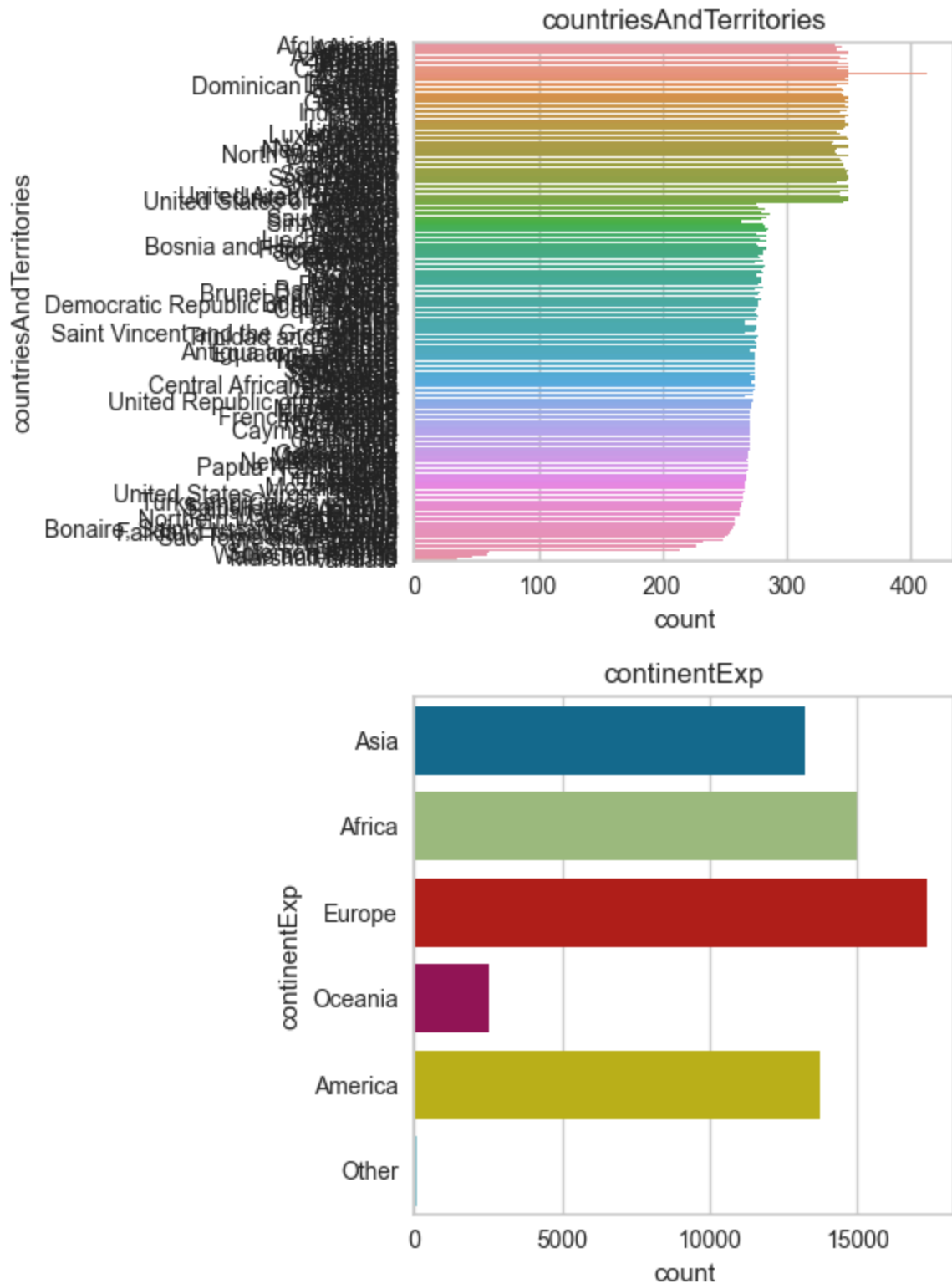




```
In [8]: # Crear la figura con subplots
fig, axs = plt.subplots(len(var_cat), 1, figsize=(6, 4*len(var_cat)))
axs = axs.flatten()

# Crear los gráficos de caja para las variables numéricas
for i, col in enumerate(var_cat):
    sns.countplot(y=col, data=df, ax=axs[i])
    axs[i].set_title(f"{col}")

# Ajustar los subplots y mostrar la figura
fig.tight_layout()
plt.show()
```



## Modificación y Transformación

- en base al análisis anterior, se observan valores null's y outliers en las variables de casos y muertes

```
In [9]: df_0 = df.dropna().reset_index().copy()
```

```
In [10]: df_1 = df_0[['cases',
                    , 'deaths',
                    , 'Cumulative_number_for_14_days_of_COVID-19_cases_per_100000',
                    , 'PORC_SOBREPESO_F',
                    , 'PORC_SOBREPESO_M']].copy()
```

```
In [11]: df_1.sample(5)
```

```
Out[11]:
```

	cases	deaths	Cumulative_number_for_14_days_of_COVID-19_cases_per_100000	PORC_SOBREPESO_F
44127	33	0	2.185363	0.0
16710	0	0	0.000000	56.3
46558	15186	222	312.392809	61.0
10730	76	0	82.989878	67.2
56651	2043	30	354.339601	53.3

```
In [12]: # escalado
escalado = StandardScaler()
df_scaler = escalado.fit_transform(df_1)
```

## Funciones adicionales

```
In [13]: # Crea la función para calcular el centroide de los clusters
def calcular_centroide(x, y):
    return (np.sum(x)/len(x), np.sum(y)/len(y))# Crea la función para calcul
```

```
In [14]: # Crear la función para Graficar las agrupaciones
def plot_graficos_clusters(atributos, predicciones, nombre_algoritmo):

    colors = ['#1f77b4', '#ff7f0e', '#2ca02c', '#d62728', '#9467bd', '#8c564b']
    markers = ['*', '^', 'X', '+', 'D', 'H', 'o', '+', 's', 'v']

    plt.figure(figsize=(12,6))
    plt.title('Pronósticos de algoritmos' + nombre_algoritmo)

    for i in np.unique(predicciones):
        x = atributos[predicciones == i, 0]
        y = atributos[predicciones == i, 1]
        centroide = calcular_centroide(x,y)

        if i < 0:
            nombre = 'Sin Cluster ' + str(i)
            plt.scatter(x, y, s = 100, c = 'gray', label = nombre)
            plt.scatter(centroide[0], centroide[1], marker = marker)
        else:
            nombre = 'Cluster ' + str(i+1)
```

```

plt.scatter(x, y, s = 100, c = colors[i], label = nombr
plt.scatter(centroide[0], centroide[1], marker = marker

plt.xlabel('Limite 1')
plt.ylabel('Limite 2')
plt.legend()
plt.show()

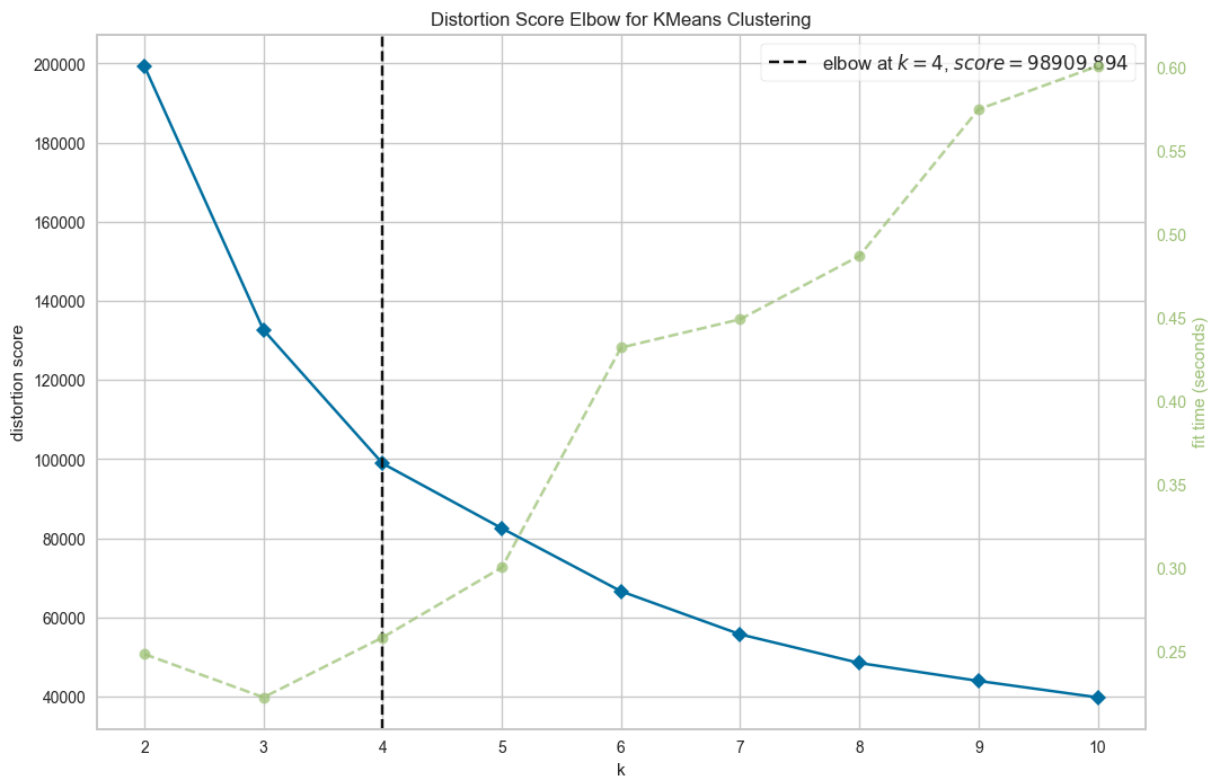
```

## Modelo KNN

```

In [15]: plt.figure(figsize=(12, 8))
g = KElbowVisualizer(KMeans(random_state=111), k=10)
g.fit(df_scaler)
g.show()

```



```

Out[15]: <Axes: title={'center': 'Distortion Score Elbow for KMeans Clustering'}, xlabel='k', ylabel='distortion score'>

```

```

In [16]: #Aplicamos el resultado de la tecnica del codo
grupos = KMeans(n_clusters=4, random_state= 111)

```

```

In [17]: #Entrenamos el modelo
grupos.fit(df_scaler)

```

```

Out[17]:
▼ KMeans
KMeans(n_clusters=4, random_state=111)

```

```

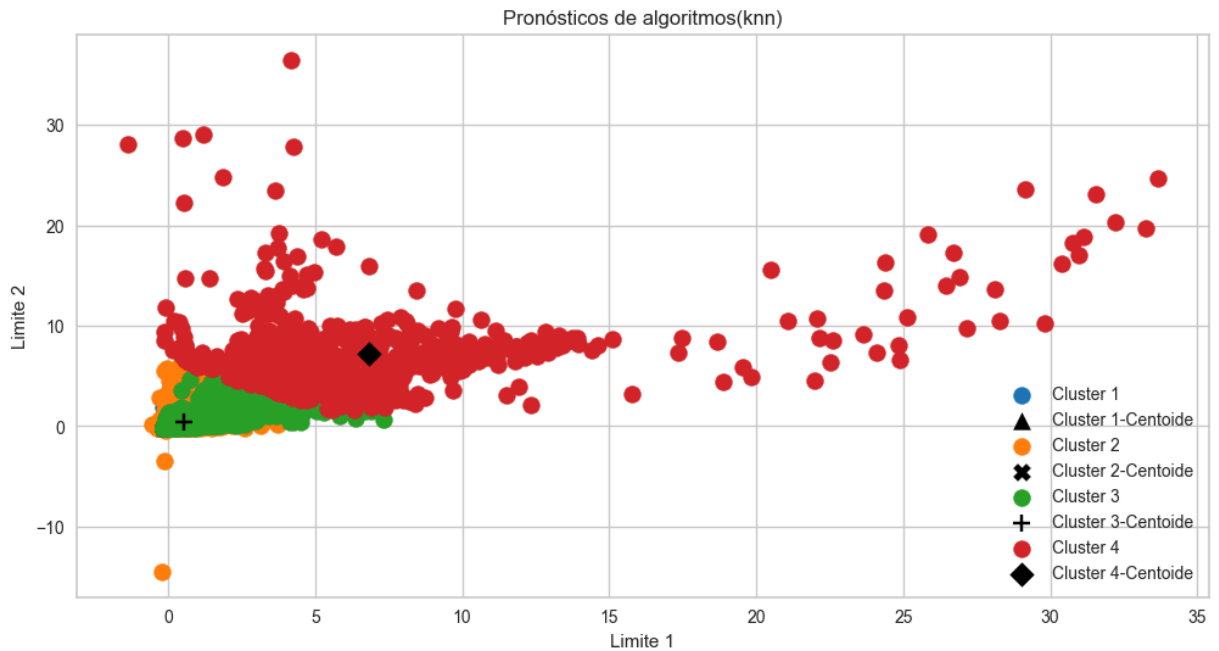
In [18]: #Obtenemos las etiquetas de grupos

```



```
pred_knn = grupos.predict(df_scaler)
```

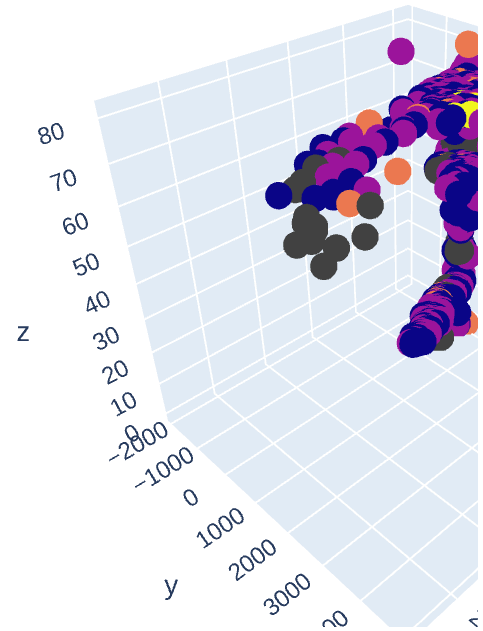
```
In [19]: plot_graficos_clusters(df_scaler ,pred_knn, '(knn)')
```



```
In [20]: fig = go.Figure()
fig.add_trace(go.Scatter(x=df['cases']
                        , y = df['deaths']
                        , mode = 'markers'
                        , marker = dict(color = pred_knn.astype(np.float64)
                        , text = pred_knn
                        )
fig.update_layout(title = 'casos X muertes')
fig.update_xaxes(title = 'casos')
fig.update_yaxes(title = 'muertes')
fig.show()
```







## GaussianMixture

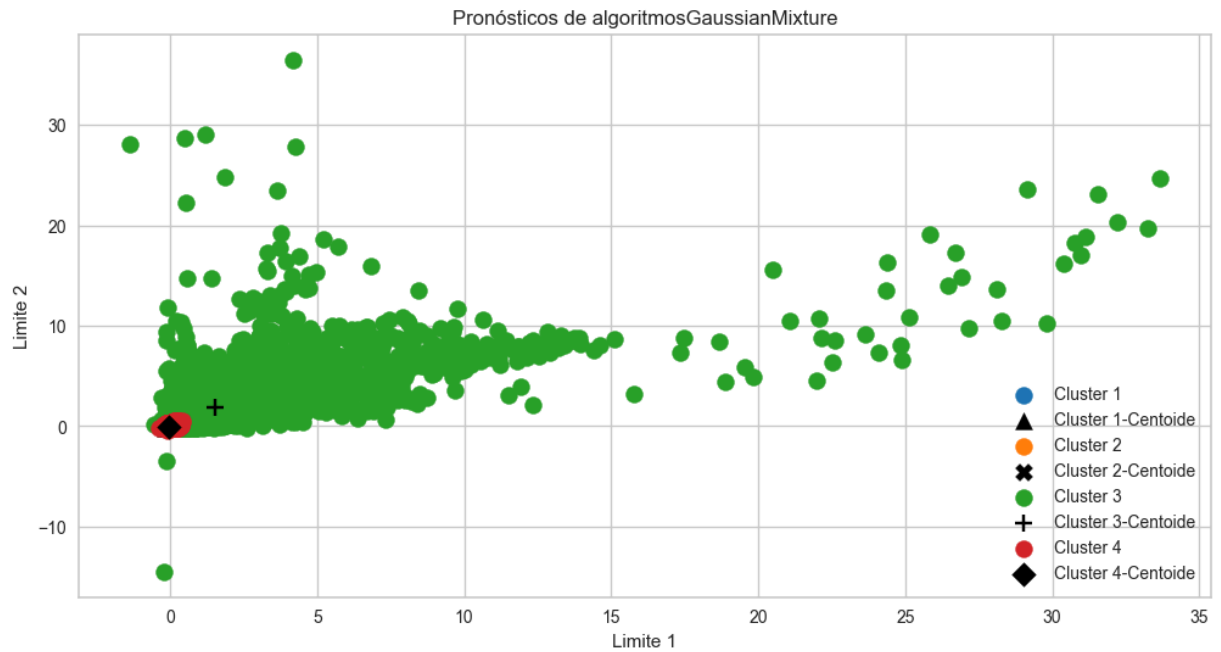
```
In [23]: # Crear una instancia de Gaussian Mixture Model
gmm = GaussianMixture(n_components= 4, random_state= 111)
```

```
In [24]: # Ajustar el modelo a los datos
gmm.fit(df_scaler)
```

```
Out[24]: ▼ GaussianMixture
GaussianMixture(n_components=4, random_state=111)
```

```
In [25]: # Obtener las etiquetas de los clusters
pred_gmm = gmm.predict(df_scaler)
```

```
In [26]: plot_graficos_clusters(df_scaler ,pred_gmm, 'GaussianMixture')
```



```
In [27]: fig = go.Figure()
fig.add_trace(go.Scatter(x=df['cases'], y = df['deaths'],
                        mode = 'markers',
                        marker = dict(color
                        text = pred_gmm))

fig.update_layout(title = 'casos X muertes')
fig.update_xaxes(title = 'casos')
fig.update_yaxes(title = 'muertes')
fig.show()
```

