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1. (1,0) Leia a base de dados de nome cidade_br.csv e exiba os dados brutos no Jupyter Notebook.

```
In [69]: import pandas
```

```
In [70]: dados = pandas.read_csv('A:/DP-FACULDADE/AC03/cidades_br.csv')
```

```
In [71]: print(dados)
```

	Código IBGE	Nome do Município	Código UF	UF	Estado \
0	5200050	Abadia de Goiás	52	GO	Goiás
1	3100104	Abadia dos Dourados	31	MG	Minas Gerais
2	5200100	Abadiânia	52	GO	Goiás
3	3100203	Abaeté	31	MG	Minas Gerais
4	1500107	Abaetetuba	15	PA	Pará
...
5565	4314548	Pinto Bandeira	43	RS	Rio Grande do Sul
5566	4220000	Balneário Rincão	42	SC	Santa Catarina
5567	4212650	Pescaria Brava	42	SC	Santa Catarina
5568	1504752	Mojú dos Campos	15	PA	Pará
5569	5006275	Paraíso das Águas	50	MS	Mato Grosso do Sul

	Latitude	Longitude
0	-16.75730	-49.4412
1	-18.48310	-47.3916
2	-16.19700	-48.7057
3	-19.15510	-45.4444
4	-1.72183	-48.8788
...
5565	-29.09750	-51.4503
5566	-28.83140	-49.2352
5567	-28.39660	-48.8864
5568	-2.68220	-54.6425
5569	-19.02160	-53.0116

[5570 rows x 7 columns]

2. (1,0) Atribua x para Latitude e y para Longitude e visualize as coordenadas das cidades no gráfico tipo scatter

```
In [72]: from pandas import DataFrame
```

```
In [73]: import matplotlib.pyplot as plt
```

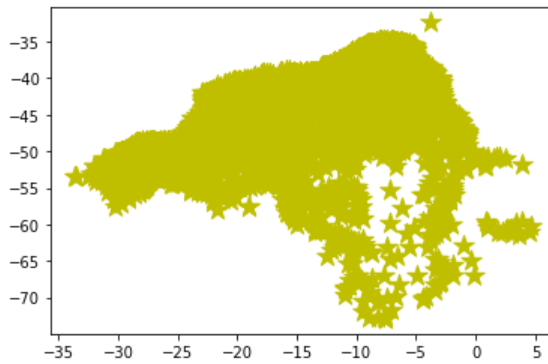
```
In [74]: from sklearn.cluster import KMeans
```

```
In [75]: Cidade = {'x': dados.Latitude,
                  'y': dados.Longitude
                  }
```

```
In [76]: df = DataFrame(Cidade, columns=['x', 'y'])
```

```
In [77]: plt.scatter(Cidade['x'],Cidade['y'], label = 'Pontos', color = 'y', marker = '*', s = 200)
```

<matplotlib.collections.PathCollection at 0x1d11b9eb148>



3. Aplique o algoritmo K-Médias para agrupar a base:

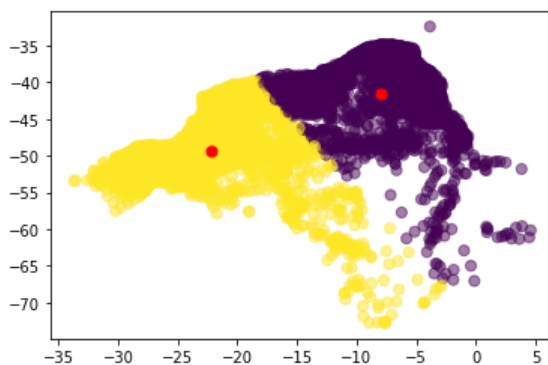
(1,0) Para dois clusters. Apresente os gráficos correspondentes com os pontos de cada cluster em uma cor diferente

```
In [78]: kmeans = KMeans(n_clusters=2).fit(df)
centroids = kmeans.cluster_centers_
print(centroids)
```

```
[[ -8.00238897 -41.55229953]
 [-22.23791819 -49.43743607]]
```

```
In [79]: plt.scatter(df['x'], df['y'], c= kmeans.labels_.astype(float), s=50, alpha=0.5)
plt.scatter(centroids[:, 0], centroids[:, 1], c='r', s=50)
```

<matplotlib.collections.PathCollection at 0x1d11a881088>



(1,0) Para o valor de clusters n a sua escolha.

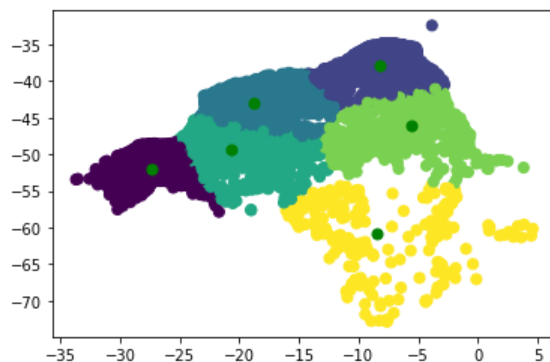
Apresente os gráficos correspondentes com os pontos de cada cluster em uma cor diferente

```
In [80]: kmeans = KMeans(n_clusters=6).fit(df)
centroids = kmeans.cluster_centers_
print(centroids)
```

```
[[-27.37052782 -52.02494798]
 [ -8.16573994 -37.96833149]
 [-18.7982683  -42.95060045]
 [-20.70610984 -49.37727555]
 [ -5.52076186 -46.1457416 ]
 [ -8.45392943 -60.91168438]]
```

```
In [81]: plt.scatter(df['x'], df['y'], c= kmeans.labels_.astype(float), s=50, alpha=1)
plt.scatter(centroids[:, 0], centroids[:, 1], c='g', s=50)
```

<matplotlib.collections.PathCollection at 0x1d11a7f7ac8>



4. Aplique o algoritmo DBSCAN para agrupar a base:

(1,0) Para os parâmetros $\text{eps}=0.3$, $\text{min_samples}=2$. Apresente os gráficos correspondentes.

```
In [82]: from sklearn.cluster import KMeans
from sklearn.cluster import DBSCAN
import matplotlib.pyplot as plt
import numpy as np
```

```
In [83]: Brasil = dados[['Latitude', 'Longitude']].where(dados['UF'].isin(['RO', 'AC', 'AM', 'RR', 'PA', 'AP',
'TO', 'MA', 'PI', 'CE', 'RN', 'PB', 'PE', 'AL', 'SE', 'BA', 'MG', 'ES', 'RJ', 'SP', 'PR', 'SC', 'RS', 'MS', 'MT',
'GO', 'DF'])).dropna()
x = Brasil.values
```

```
In [84]: print(Brasil)
kmeans = KMeans(n_clusters=2)
kmeans.fit(x)

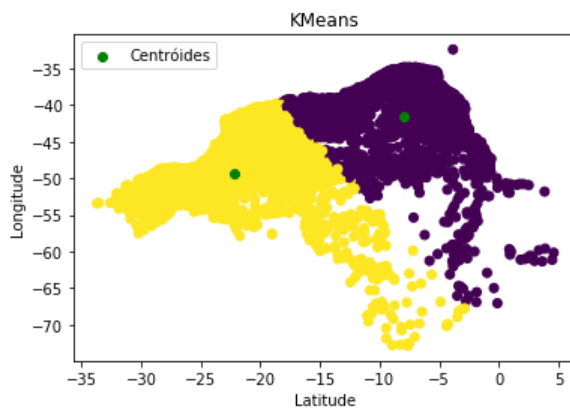
   Latitude Longitude
0   -16.75730   -49.4412
1   -18.48310   -47.3916
2   -16.19700   -48.7057
3   -19.15510   -45.4444
4    -1.72183   -48.8788
...      ...      ...
5565 -29.09750   -51.4503
5566 -28.83140   -49.2352
5567 -28.39660   -48.8864
5568 -2.68220   -54.6425
5569 -19.02160   -53.0116

[5570 rows x 2 columns]

KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300,
       n_clusters=2, n_init=10, n_jobs=None, precompute_distances='auto',
       random_state=None, tol=0.0001, verbose=0)
```

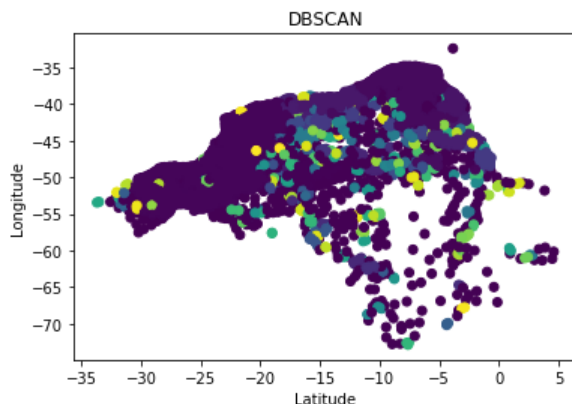
```
In [85]: Brasil['clusters'] = kmeans.predict(x)
centros = kmeans.cluster_centers_
```

```
In [86]: plt.title('KMeans')
plt.xlabel('Latitude')
plt.ylabel('Longitude')
plt.scatter(Brasil['Latitude'].values, Brasil['Longitude'].values, c = Brasil['clusters'])
plt.scatter(centros[:,0], centros[:,1], c = 'g', label='Centróides')
plt.legend()
plt.show()
```



```
In [87]: dbscan = DBSCAN(eps=.3,min_samples=2)
z = dbscan.fit_predict(x)
```

```
In [88]: plt.title('DBSCAN')
plt.xlabel('Latitude')
plt.ylabel('Longitude')
plt.scatter(Brasil['Latitude'].values, Brasil['Longitude'].values, c = z)
plt.show()
```



5.(2,0) Aplique os algoritmos K-Médias e DBSCAN para agrupar as cidades da região Sudeste.

Ajuste os parâmetros para obtermos 5 clusters com ambos os algoritmos.

Apresente os gráficos correspondentes.

```
In [89]: Sudeste = dados[['Latitude', 'Longitude']].where(dados['UF'].isin(['SP', 'sp', 'MG', 'mg', 'ES', 'es',
, 'RJ', 'rj'])).dropna()
x = Sudeste.values
```

```
In [90]: print(Sudeste)
kmeans = KMeans(n_clusters=5)
kmeans.fit(x)
```

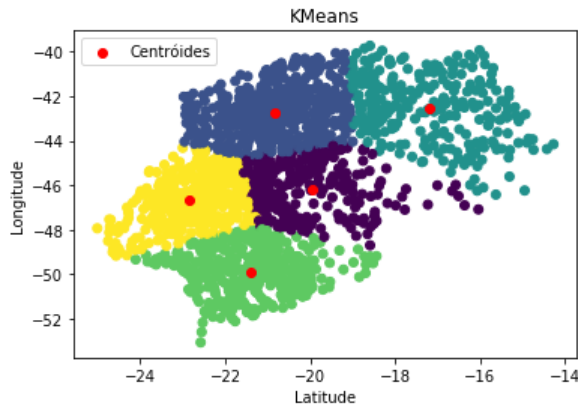
```
Latitude Longitude
1 -18.483100 -47.391600
3 -19.155100 -45.444400
12 -20.299600 -42.474300
15 -20.359000 -43.143900
28 -19.067100 -42.541900
...
5541 -23.443890 -46.917778
5542 -23.221994 -45.310883
5543 -23.541056 -46.370972
5544 -21.549999 -47.780083
5545 -22.868000 -50.681374
```

```
[1668 rows x 2 columns]
```

```
KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300,
n_clusters=5, n_init=10, n_jobs=None, precompute_distances='auto',
random_state=None, tol=0.0001, verbose=0)
```

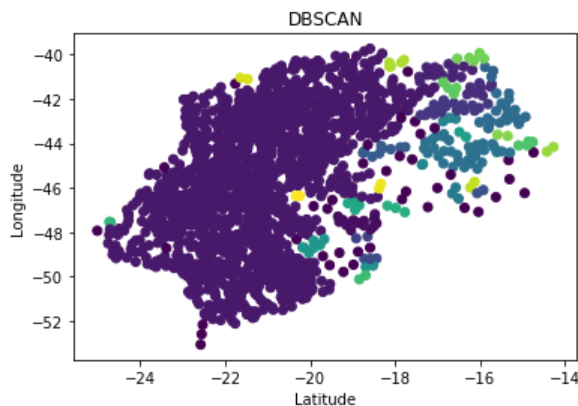
```
In [91]: Sudeste['clusters'] = kmeans.predict(x)
centros = kmeans.cluster_centers_
```

```
In [92]: plt.title('KMeans')
plt.xlabel('Latitude')
plt.ylabel('Longitude')
plt.scatter(Sudeste['Latitude'].values, Sudeste['Longitude'].values, c = Sudeste['clusters'])
plt.scatter(centros[:,0], centros[:,1], c = 'r', label='Centróides')
plt.legend()
plt.show()
```



```
In [93]: dbscan = DBSCAN(eps=.3,min_samples=2)
z = dbscan.fit_predict(x)
```

```
In [94]: plt.title('DBSCAN')
plt.xlabel('Latitude')
plt.ylabel('Longitude')
plt.scatter(Sudeste['Latitude'].values, Sudeste['Longitude'].values, c = z)
plt.show()
```



6.(2,0) Aplique os algoritmos K-Médias e DBSCAN para agrupar as cidades do estado do Amazonas.

Ajuste os parâmetros para obtermos 5 clusters com ambos os algoritmos. Apresente os gráficos correspondentes.

```
In [95]: Amazonas = dados[['Latitude', 'Longitude']].where(dados['UF'].isin(['AM', 'am'])).dropna()
x = Amazonas.values
```

```
In [96]: print(Amazonas)
```

```

      Latitude Longitude
167  -3.227270 -64.800700
192  -3.374550 -68.200500
213  -3.566970 -61.396300
249  -3.746030 -61.657500
283  -7.194090 -59.896000
...
5249 -2.996090 -65.113300
5304 -2.529360 -57.753800
5308 -3.128410 -58.149600
5473 -3.314000 -59.555737
5474 -6.438519 -68.243736

```

```
[62 rows x 2 columns]
```

```
In [97]: kmeans = KMeans(n_clusters=5)
kmeans.fit(x)
```

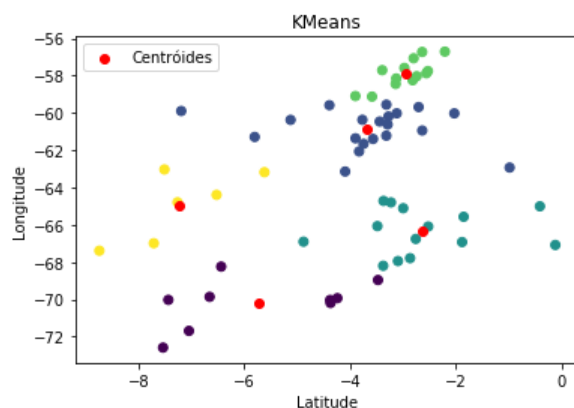
```

KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300,
       n_clusters=5, n_init=10, n_jobs=None, precompute_distances='auto',
       random_state=None, tol=0.0001, verbose=0)

```

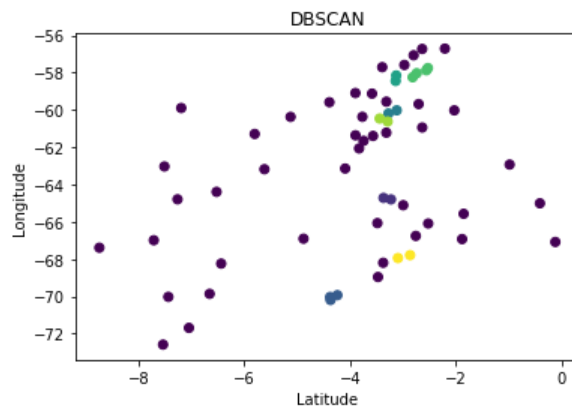
```
In [98]: Amazonas['clusters'] = kmeans.predict(x)
centros = kmeans.cluster_centers_
```

```
In [99]: plt.title('KMeans')
plt.xlabel('Latitude')
plt.ylabel('Longitude')
plt.scatter(Amazonas['Latitude'].values, Amazonas['Longitude'].values, c = Amazonas['clusters']
])
plt.scatter(centros[:,0], centros[:,1], c = 'r', label='Centróides')
plt.legend()
plt.show()
```



```
In [100]: dbSCAN = DBSCAN(eps=.3,min_samples=2)
z = dbSCAN.fit_predict(x)
```

```
In [101]: plt.title('DBSCAN')  
plt.xlabel('Latitude')  
plt.ylabel('Longitude')  
plt.scatter(Amazonas['Latitude'].values, Amazonas['Longitude'].values, c = z)  
plt.show()
```



```
In [ ]:
```