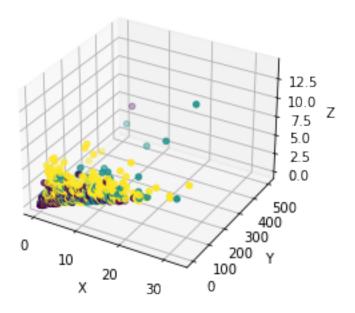
### terrain agrupamiento

### November 14, 2021

```
[]: import numpy as np
    import matplotlib.pyplot as plt
    from sklearn.cluster import KMeans
    from sklearn.cluster import AgglomerativeClustering
    from sklearn.cluster import SpectralClustering
    from sklearn.cluster import OPTICS
    from sklearn.cluster import DBSCAN
    from scipy.cluster.hierarchy import dendrogram, linkage
    from sklearn.metrics import pairwise_distances
    def plot_d(points, labels, title):
        fig = plt.figure()
        if points.shape[1] > 2:
             ax = fig.add_subplot(projection='3d')
            ax.scatter(points[:,0], points[:,1], points[:,2], c=labels,_
      ax.set_xlabel('X')
            ax.set_ylabel('Y')
            ax.set_zlabel('Z')
            ax.set_title(title)
        else:
            plt.scatter(points[:,0], points[:,1], c=labels, cmap='viridis')
            plt.xlabel('X')
            plt.ylabel('Y')
            plt.title(title)
        plt.show()
[]:  # Data
    terrains = ('Plano', 'DM', 'Elevaciones')
    x = np.load('processed_terrain_data.npy')
    y = np.load('terrain_data_labels.npy')
[]: plot_d(x, y, 'Map Features')
```

# Map Features

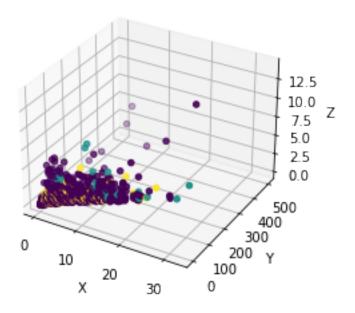


## 1 K-Means

```
[]: kmeans = KMeans(n_clusters=len(terrains)).fit(x)
clustering_labels = kmeans.labels_
centers = kmeans.cluster_centers_

plot_d(x, clustering_labels, 'K-Means')
```

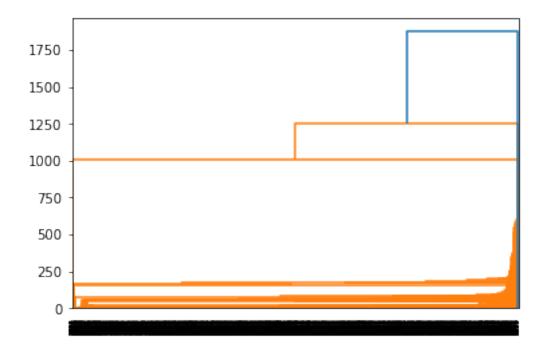
### K-Means



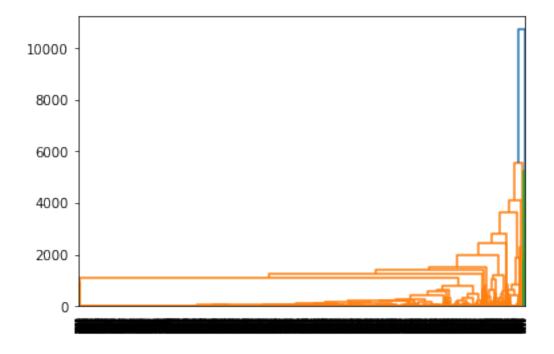
# 2 Dendogram

Sabe que rollo

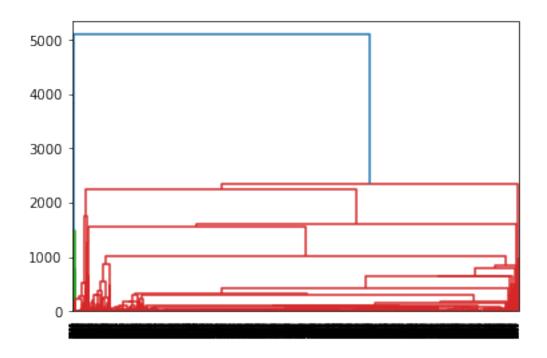
---- Dendrogram plot ----



---- Dendrogram plot ----

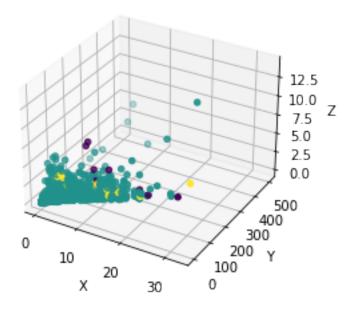


---- Dendrogram plot ----



# 3 Aglomerative

## Aglomerative clustering



```
[]: # Aglomerative clustering with manhattan distance and mean linkage

agl = AgglomerativeClustering(n_clusters=len(terrains), linkage = 'complete',

⇒affinity = 'manhattan').fit(x)

clustering_labels = agl.labels_

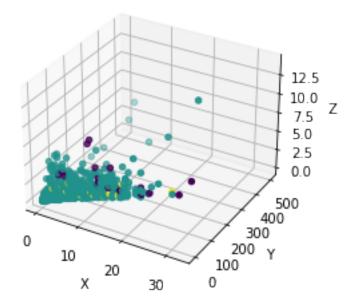
print('Labels: ', clustering_labels)

plot_d(x, clustering_labels, 'Aglomerative clustering (Manhattan affinity &∟

⇒Mean Linkage)')
```

Labels: [1 1 1 ... 1 1 1]

## Aglomerative clustering (Manhattan affinity & Mean Linkage)



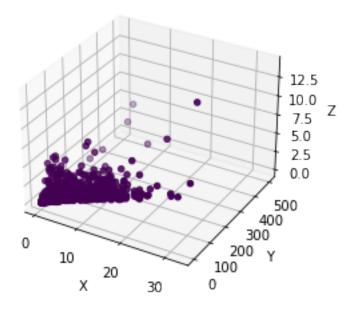
```
[]: ###### Spectral clustering ######
print('---- Spectral clustering -----')
spectral = SpectralClustering(n_clusters=len(terrains)).fit(x)
clustering_labels = spectral.labels_
print('Labels: ', clustering_labels)
plot_d(x, clustering_labels, 'Spectral clustering')
```

---- Spectral clustering ----

/home/hivini/anaconda3/envs/tf-gpu/lib/python3.9/sitepackages/sklearn/manifold/\_spectral\_embedding.py:236: UserWarning: Graph is not fully connected, spectral embedding may not work as expected. warnings.warn("Graph is not fully connected, spectral embedding"

Labels: [0 0 0 ... 0 0 0]

# Spectral clustering

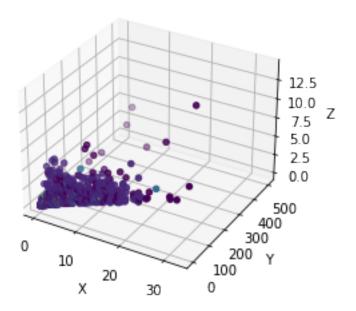


```
[]: print('---- DBSCAN -----')
     dbscan = DBSCAN(eps=20).fit(x)
     clustering_labels = dbscan.labels_
    print('Labels: ', clustering_labels)
    plot_d(x, clustering_labels, 'DBSCAN')
```

---- DBSCAN ----

Labels: [0 0 0 ... 0 8 0]

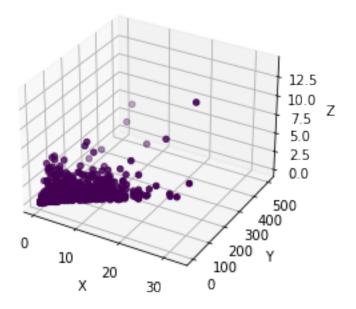
### DBSCAN



/home/hivini/anaconda3/envs/tf-gpu/lib/python3.9/sitepackages/sklearn/manifold/\_spectral\_embedding.py:236: UserWarning: Graph is not fully connected, spectral embedding may not work as expected. warnings.warn("Graph is not fully connected, spectral embedding"

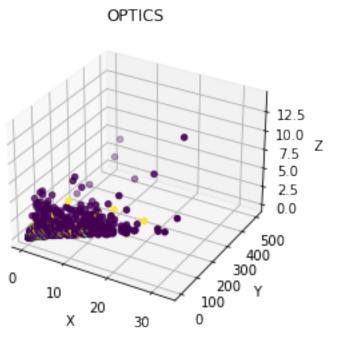
Labels: [0 0 0 ... 0 0 0]

# Spectral clustering



```
[]: ###### OPTICS ######
print('---- OPTICS ----')
optics = OPTICS(min_samples = 20).fit(x)
clustering_labels = optics.labels_
print('Labels: ', clustering_labels)
plot_d(x, clustering_labels, 'OPTICS')
---- OPTICS -----
```

Labels: [-1 -1 0 ... -1 -1 4]



Χ