BT3041 - Analysis and Interpretation of Biological Data

Assignment 2

Report

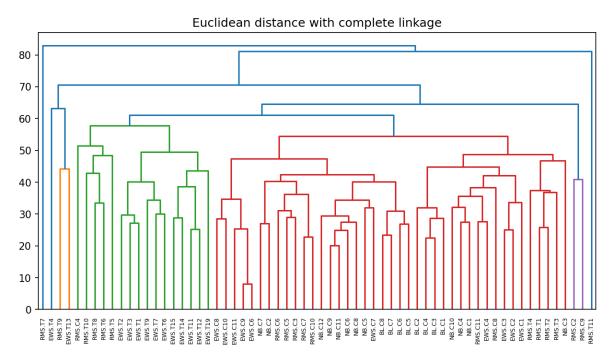
The link to the Python notebook containing the code to generate the dendrograms in this report can be found here —

https://drive.google.com/file/d/1Vroh_pUv_ftylSOv-IzBkGVTXKUoyJUr/view?usp=sharing.

- The goal of this analysis is to perform agglomerative clustering with different distance metrics and linkage methods on the given dataset.
- We hope to uncover the different cancer types by clustering them based on their gene expression patterns.
- As the expression levels have the same order of magnitude across genes, we do NOT scale the given data.

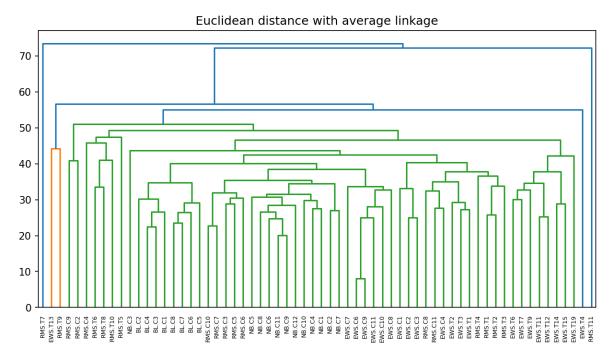
1

When Euclidean distance and complete linkage are used, the dendrogram looks like:



The clustering algorithm identifies 5 clusters. The **green** cluster contains only rhabdomyosarcomas (RMS) and Ewing tumours (EWS). However, the **red** cluster contains all cancer types. This is a decent clustering.

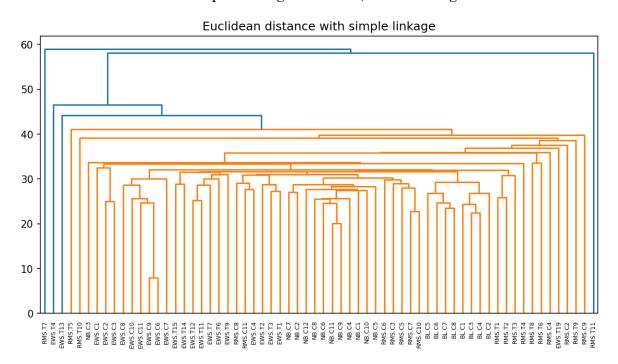
When Euclidean distance and average linkage are used, the dendrogram looks like:



The clustering algorithm identifies 3 clusters. The **green** cluster contains almost all the data points. This is not a good clustering.

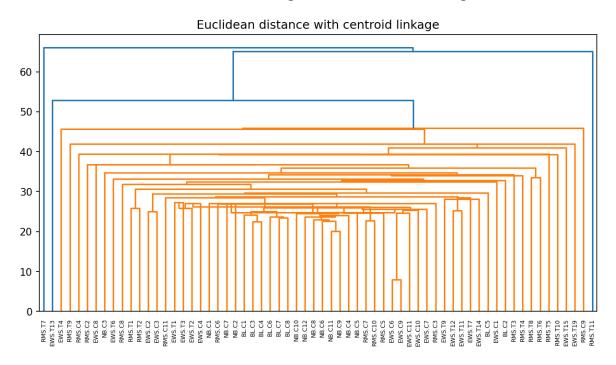
3

When Euclidean distance and simple linkage are used, the dendrogram looks like:



The clustering algorithm identifies 2 clusters. The **orange** cluster contains almost all the data points. This is not a good clustering.

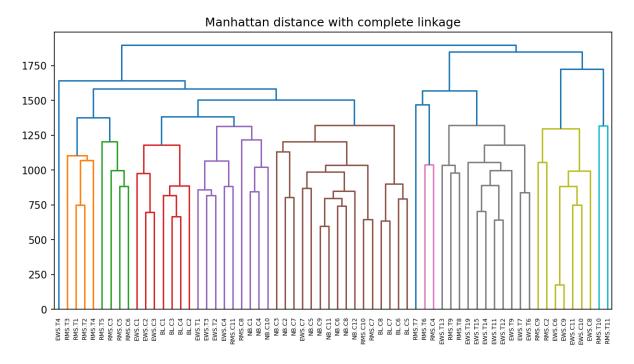
When Euclidean distance and centroid linkage are used, the dendrogram looks like:



The clustering algorithm identifies 2 clusters. The **orange** cluster contains almost all the data points. There are also a lot of overlaps in the dendrogram. This is not a good clustering.

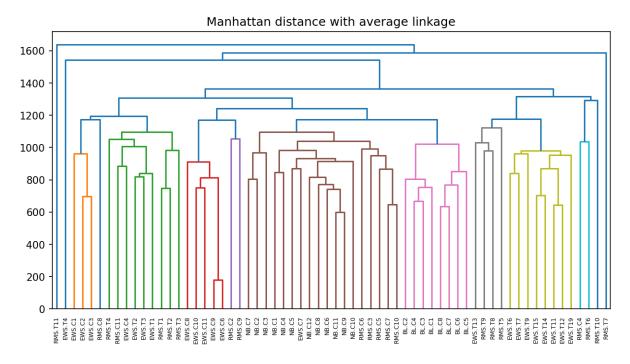
5

When Manhattan distance and complete linkage are used, the dendrogram looks like:



The algorithm identifies 10 clusters. However, most of these are composed of multiple cancer types. RMS and EWS are usually clustered together. This is a decent clustering.

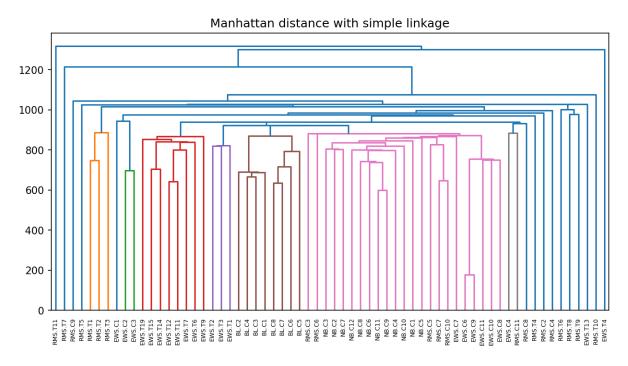
When Manhattan distance and average linkage are used, the dendrogram looks like:



The algorithm identifies 10 clusters. However, most of these are composed of multiple cancer types. The **pink** cluster contains Burkitt lymphomas (BL) exclusively and the **lime** cluster contains EWS exclusively. All the neuroblastomas (NB) are concentrated in the **brown** cluster. This is better than the previous clustering with Manhattan distance and complete linkage.

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When Manhattan distance and simple linkage are used, the dendrogram looks like:



The algorithm identifies 8 clusters. However, most of these are composed of multiple cancer types, with RMS and EWS clustered together frequently. This is a slightly messy clustering.

When Manhattan distance and centroid linkage are used, a dendrogram cannot be produced. Centroid linkage requires that Euclidean distance be used as the metric.

```
ValueError
                                          Traceback (most recent call last)
Cell In[17], line 1
----> 1 Z = linkage(
           y = df
     2
     3
           method='centroid',
     4
           metric='cityblock',
           optimal_ordering = True
     5
     6)
     8 plt.figure(figsize=(10, 5),dpi=100)
     9 dendrogram(Z,labels=df.index)
File ~\anaconda3\Lib\site-packages\scipy\cluster\hierarchy.py:1052, in linkage(y, method, metric, optimal ordering)
   1050 elif y.ndim == 2:
         if method in _EUCLIDEAN_METHODS and metric != 'euclidean':
               raise ValueError("Method '{}' requires the distance metric "
-> 1052
                                 "to be Euclidean".format(method))
   1053
          if y.shape[0] == y.shape[1] and np.allclose(np.diag(y), 0):
   1054
   1055
                if np.all(y >= 0) and np.allclose(y, y.T):
ValueError: Method 'centroid' requires the distance metric to be Euclidean
```

Conclusions

- Manhattan distance appears to be a better metric than Euclidean distance for this particular dataset.
- Manhattan distance with average linkage is able to cluster the cancer types better than the other combinations.
- Clustering with simple linkage or centroid linkage is quite messy with this dataset.
- Euclidean distance works well when complete linkage is used.