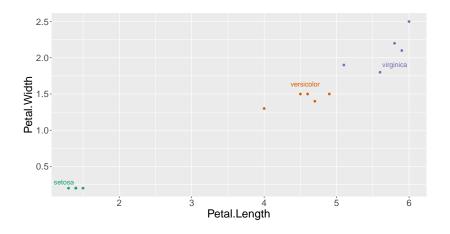
## Hierarchical Clustering

Toby Dylan Hocking

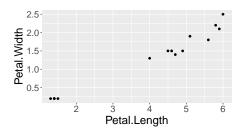
### Visualize iris data with labels



#### Visualize iris data without labels

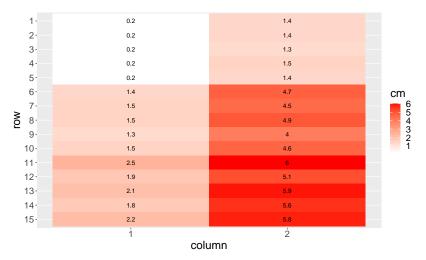
- Let  $X = [x_1 \cdots x_n]^{\mathsf{T}} \in \mathbb{R}^{n \times p}$  be the data matrix (input for clustering), where  $x_i \in \mathbb{R}^p$  is the input vector for observation i.
- **Example** iris n = 150 observations, p = 2 dimensions.

##		Petal.Width	Petal.Length
##	[1,]	0.2	1.4
##	[2,]	0.2	1.4
##	[3,]	0.2	1.3
##	[4,]	0.2	1.5



## Which pair of rows is most similar?

This is a visualization of 15 rows and two columns from the iris data.



# Hyper-parameter choices (must be fixed prior to learning)

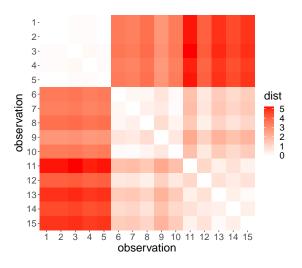
How to compute similarity/distance between rows?

- Let  $x, x' \in \mathbb{R}^p$  be two feature vectors (rows of data matrix).
- ▶ L1/manhattan distance:  $||x x'||_1 = \sum_{j=1}^p |x_j x_j'|$ .
- ▶ L2/euclidean distance:  $||x x'||_2 = \sqrt{\sum_{j=1}^p (x_j x_j')^2}$ .

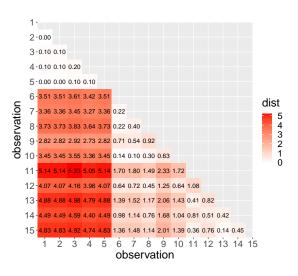
How to compute distance with a group/cluster? There are several rules, or agglomeration methods:

- single: min distance from any point,
- complete: max distance from any point,
- average: mean distance over all points,
- there are others.

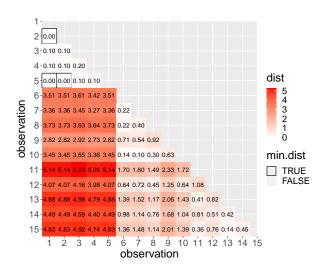
## Hierarchical clustering inputs a pairwise distance matrix



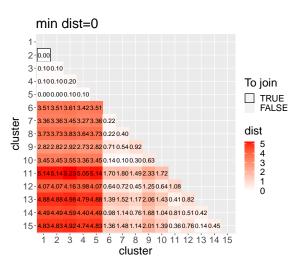
## Only need lower triangle (symmetry)

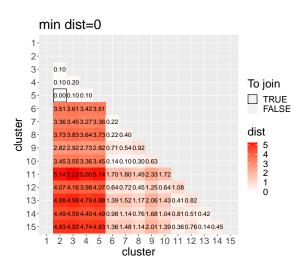


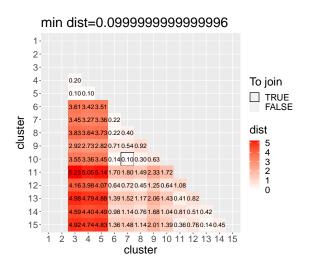
### Find the closest pairs

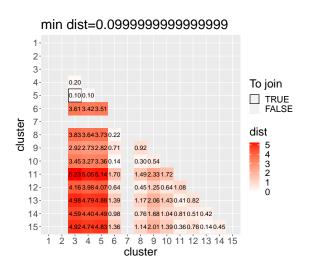


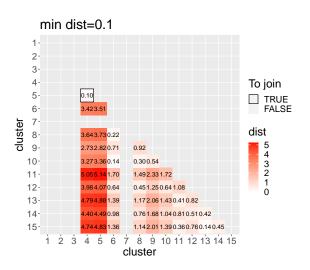
# Join one of the closest pairs (iteration 1)

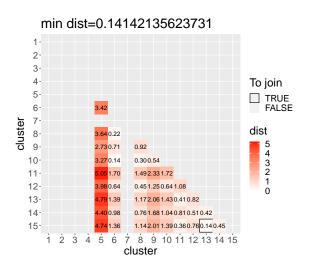


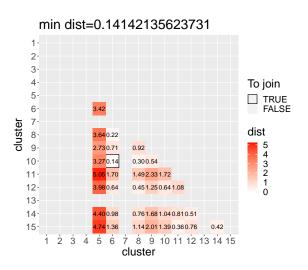


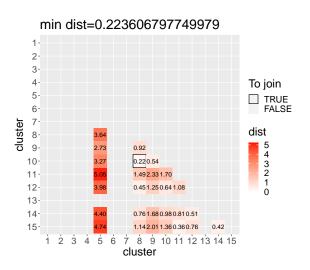


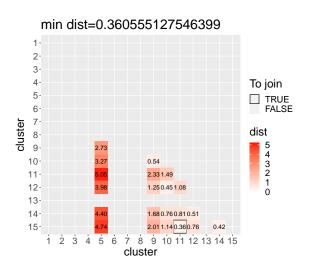


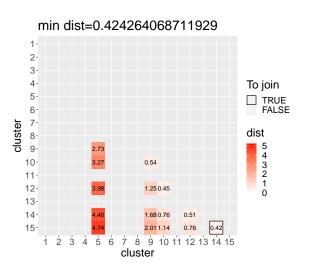


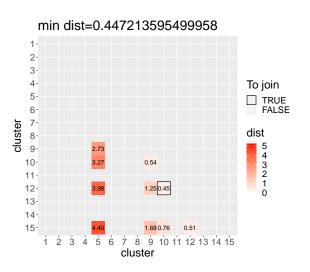


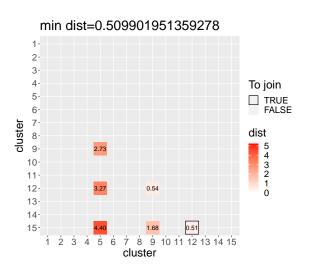


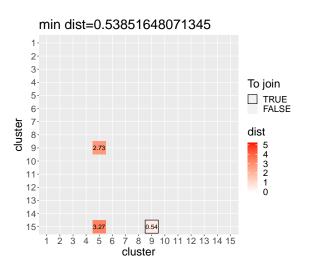


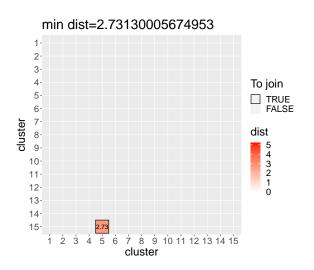




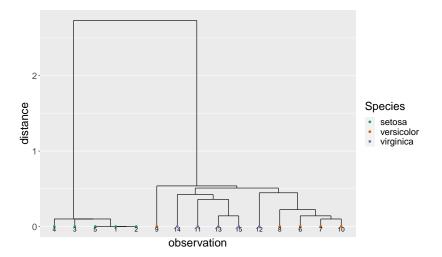




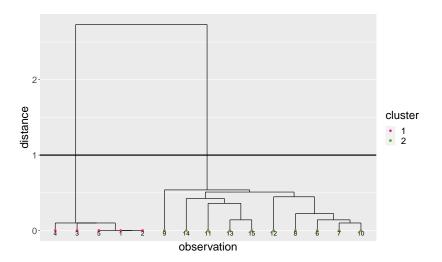




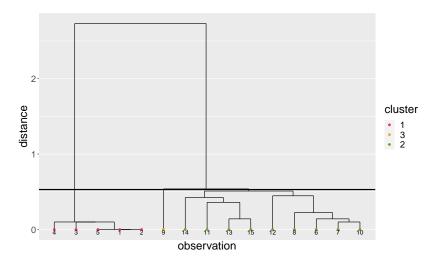
# Visualization of dendrogram (tree diagram)



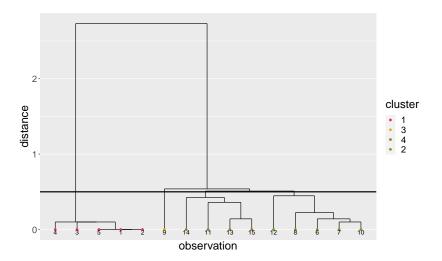
### Cutting the tree to get two clusters



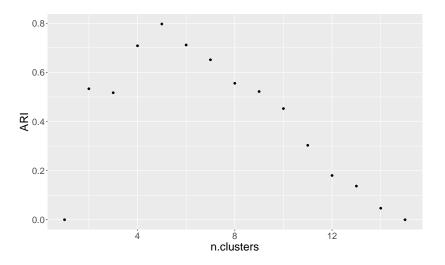
## Cutting the tree to get three clusters



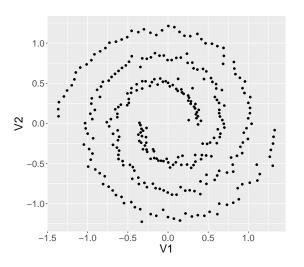
# Cutting the tree to get four clusters



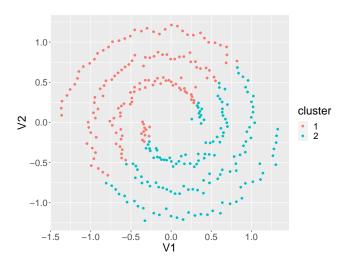
# ARI computation



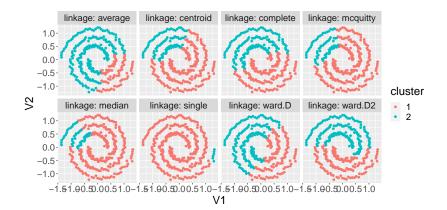
# Another 2d data set



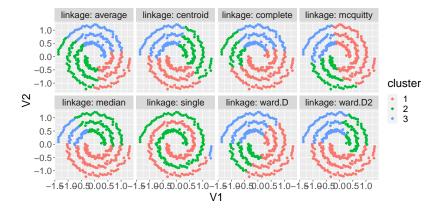
## K-means fails



#### Hierarchical with two clusters fails

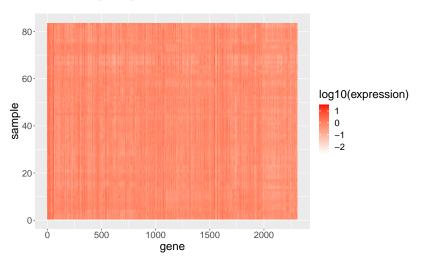


## Hierarchical with single linkage and 3 clusters better

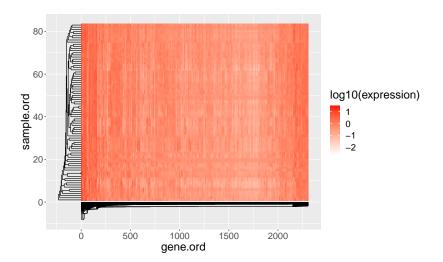


## Gene expression clustering (cancer classes)

Small Round Blue Cell Tumors (SRBCT) of childhood cancer study of Khan et al. (2001).



## Clustering samples



### Possible Exam Questions

What is the big O notation asymptotic time complexity of the following algorithms in terms of N (number of data observations/rows), P (number of data features/columns), and K (number of clusters).

- K-means.
- Gaussian mixture model with diagonal covariance matrix.
- Gaussian mixture model with unconstrained covariance matrix.
- Hierarchical clustering with single linkage.

### Possible Exam Questions 2

- ▶ What are the two hyper-parameters that must be chosen before running the hierarchical clustering algorithm?
- ► For a data set with N=200 observations/rows, how large is the pairwise distance matrix? How many iterations of the cluster joining occur?