# **Clustering Continued**

•••

By Jake Sauter

## Kohonen Self Organizing Feature Maps (SOFM)

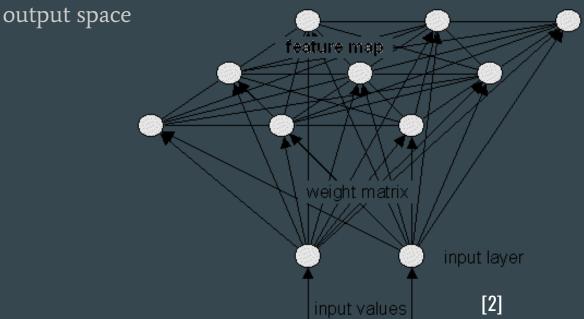
- SOFM is a type of clustering
- Novelty in that it the relationship between the clustered patterns actually contains information about the relationships and reciprocal positions of the patterns in the input space
- SOFM is designed to plot similar patterns next to one another, creating a **feature** map
- A feature map has the property that the distances and relationships measured on the feature map are proportional of distances and relationships between patterns according to the similarity metric chosen

#### SOFMs

- The SOFM is actually a **neural network** technique
- Neural networks are graphs connecting simple processing nodes (**neurons**), with each connection possessing a specific connection strength (**weight**)
- Common SOFM architectures implement one dimensional and two dimensional networks

#### SOFMs

• The **input layer** of the network represents the dimensionality of the input space, and the **output layer** of the network represents the dimensionality of the desired



## **SOFM Training**

- In order to form the output space, the SOFM must be **trained** with respect to a **training rule**
- The **training rule** in combination with a **learning rate** will inform us on how to update the weights of the network as training occurs

## **SOFM Training**

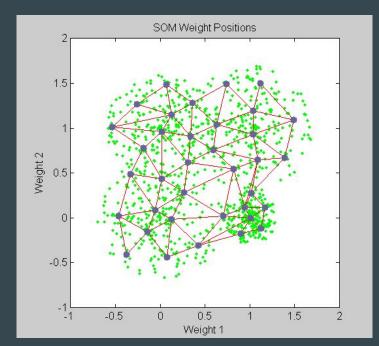
- The process of training occurs the following way:
  - An input is presented to the input layer of the neural network
  - Each neuron calculates the distance from the input to its weight vector
  - The neuron that possesses the smallest distance will be updated such that the weights are made more similar to the input profile
    - Depending on the training rule, weight of surrounding neurons may be updated as well
  - Each input is presented to the network in this way
  - After each input is presented, the learning rate and/or neighborhood distance is decreased and this process is repeated

## **SOFM Output**

- The SOFM provides three benefits
  - Each neuron of the SOFM will represent the set of common features extracted from the input patterns by the neuron
  - The SOFM yields a set of clusters, with all inputs activating the same unit being clustered together
  - The relationship between the neurons activated by specific genes will be closely related to the relationship between the genes

#### SOFM Use

- SOFMs can be used as a clustering or visualization tool
- One could simply plot the clusters generated from the SOFM
- The prototypes (weights of the units) could be plotted in the input space, with the topology of the network showing the links between the units



## K Means Results

- The K-Means clustering algorithm was implemented from scratch to allow for different distance metrics to be used
- A function to produce meaningful output for supervised learning sets, with any amount of clusters, was crafted
  - This technique could be applied to determine how well classification can even be implemented

### K Means Results

## **Custom Implementation**

#### between ss / total ss = 62.23644 %Euc.

Cluster Results: ALL AML

[1,] 1 0.09090909

0 0.90909091

### Man.

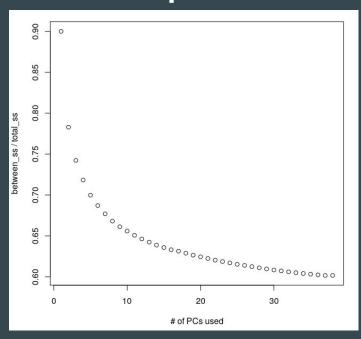
```
between ss / total ss = 60.78344 \%
Cluster Results:
           ALL AML
[1,] 0.4074074
[2,] 0.5925926
```

## **Library Function**

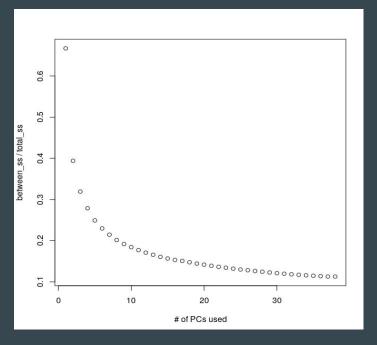
```
between ss / total ss = 11.30524 \%
Cluster Results:
           ALL
                     AML
[1,] 0.8518519 0.2727273
[2,] 0.1481481 0.7272727
```

# K Means of Principle Components (Euc)

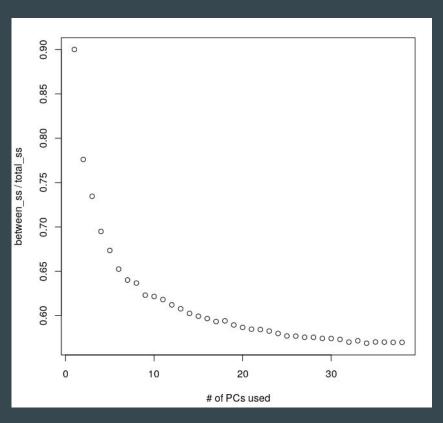
## **Custom Implementation**



## **Library Function**



# K Means of Principle Components (Man)



#### K Medoids Results

#### Euclidean

```
Cluster Results:
ALL AML
[1,] 0.88888889 0.1818182
[2,] 0.1111111 0.8181818
```

#### Manhattan

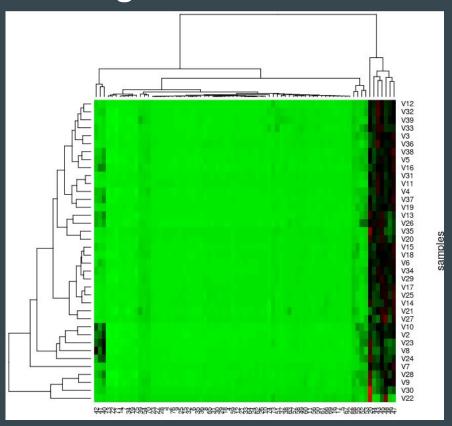
```
Cluster Results:

ALL AML

[1,] 0.92592593 0.1818182

[2<u>,</u>] 0.07407407 0.8181818
```

# **Hierarchical Clustering**



#### Rerenerences

[1] Draghici Sorin. Statistics and Data Analysis for Microarrays: Using R and Bioconductor. Chapman and Hall, 2012.

[2] "Kohonen Feature Map." Kohonen Feature Map - Neural Networks with Java, www.nnwj.de/kohonen-feature-map.html.

[3] "Cluster with Self-Organizing Map Neural Network." Cluster with Self-Organizing Map Neural Network - MATLAB & Simulink, www.mathworks.com/help/deeplearning/ug/cluster-with-self-organizing-map-neural-network.html;jsessionid=38782e43df7fa2b45736dd152319.