

# SELF-ORGANIZING MAP (SOM)

Kohonen Network

# Kohonen Network

- Its un-supervised neural network (**clustering**)
- Un-supervised NN are trained by letting the network **continually adjust itself to new inputs**. They find relationships within data and can automatically define classification schemes.
- Self -organizing maps (SOM) is a type of artificial neural network which can represent a one-dimensional input space in a two-dimensional map

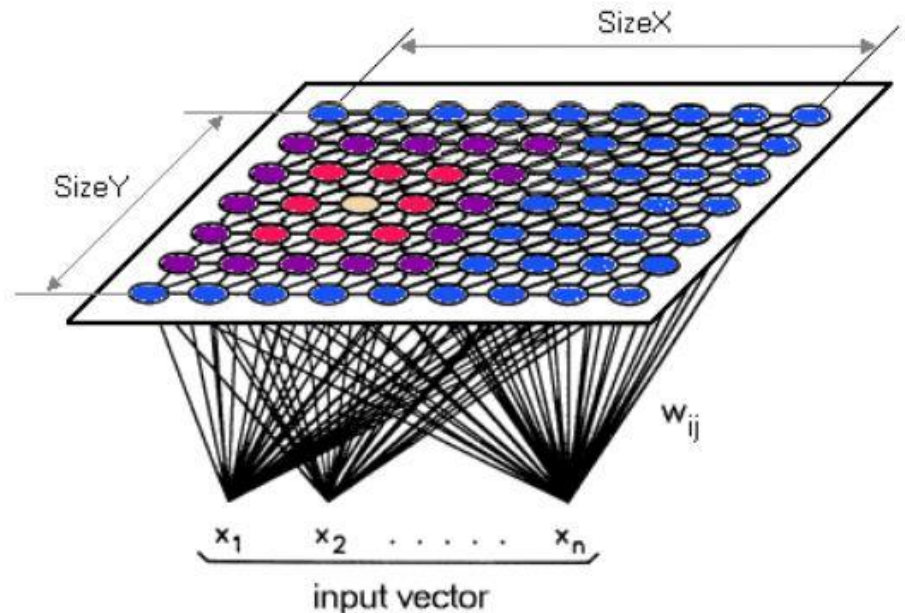
# Network Structure

- **It has two layer:**

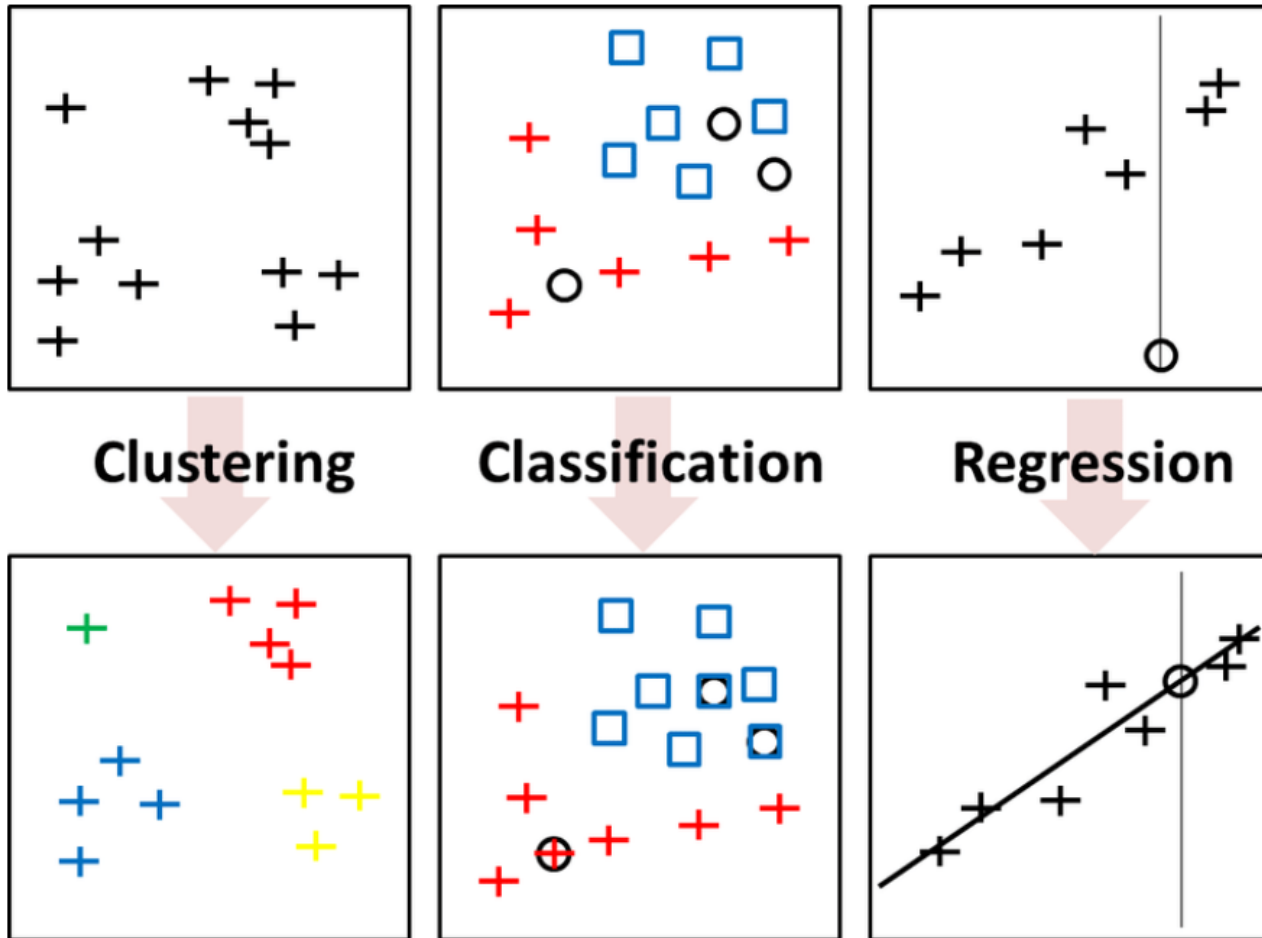
- Input layer

- Computational layer (Grid)

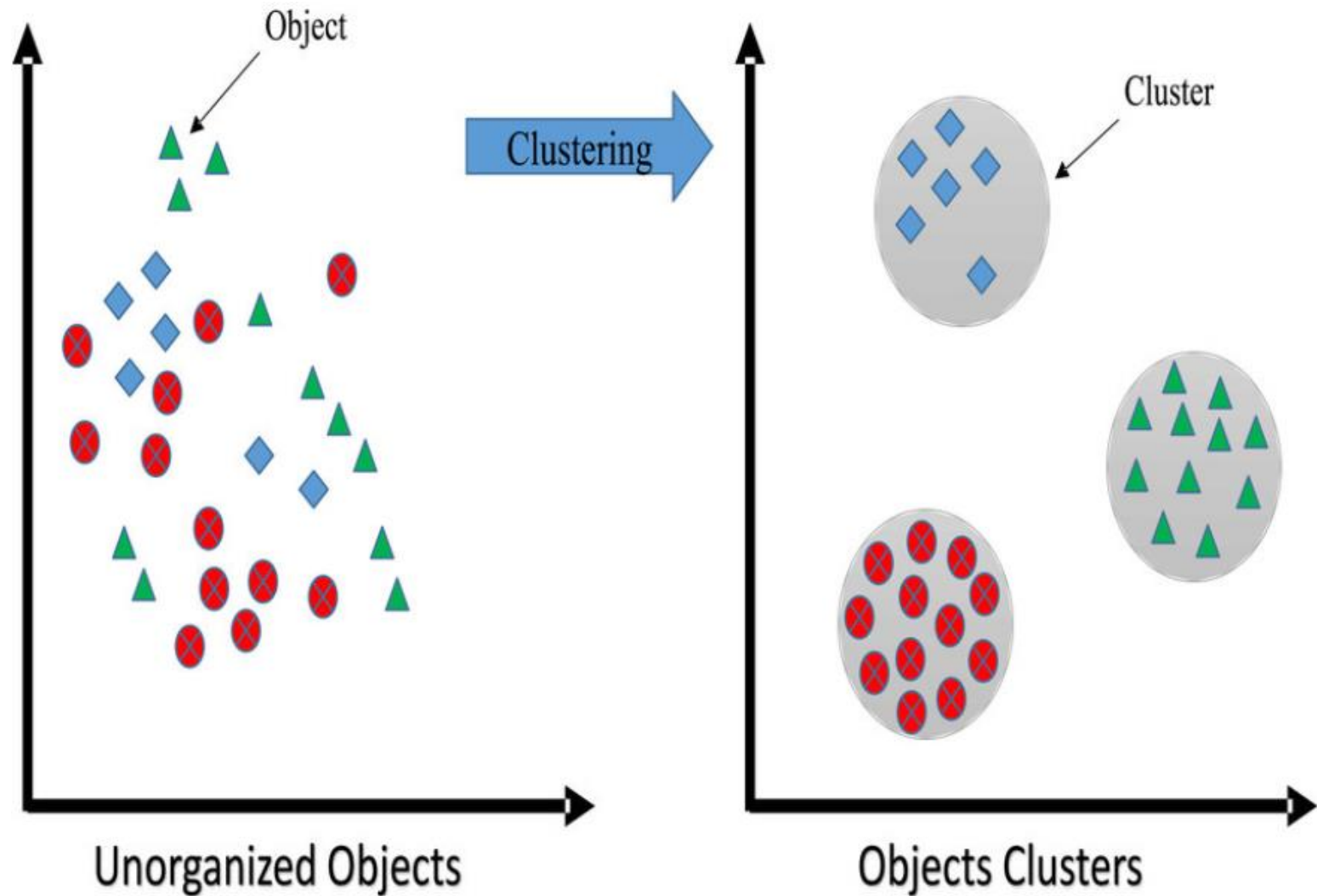
- Each neuron is fully connected to all the source nodes in the input layer



# Classification vs Regression vs Clustering



# Clustering



# How the algorithm learn?

- It learn to classify input vectors according to similarity. They are used for classification and pattern recognition tasks.
- The stages of the SOM algorithm can be summarized as follows:
  1. **Sampling:** get a sample training input vector from the input space.
  2. **Initialization:** build a grid and initialize each node with random features
  3. **Matching:**
    - Compare each input with all nodes in the grid
    - Select the best matching unit (BMU)
  4. **Updating:** update each node in the range of the BMU
  5. **Continuation:** keep returning to step 3 until the feature map stops changing

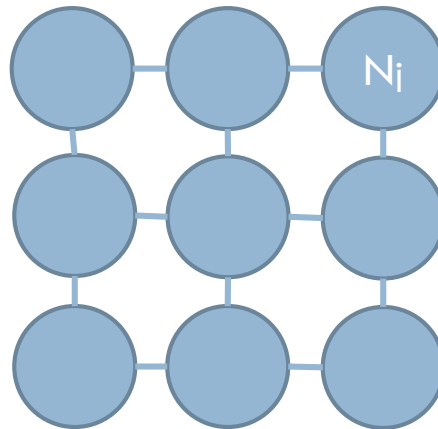
# How the algorithm learn?

**1) Sampling**: get a sample training input vector from the input space.

$X=[X_0, X_1, X_2, \dots, X_n]$  each input is vector

$X_0=[X_{00}, X_{01}, \dots, X_{0m}]$

**2) Initialization**: build a grid and initialize each node with random features

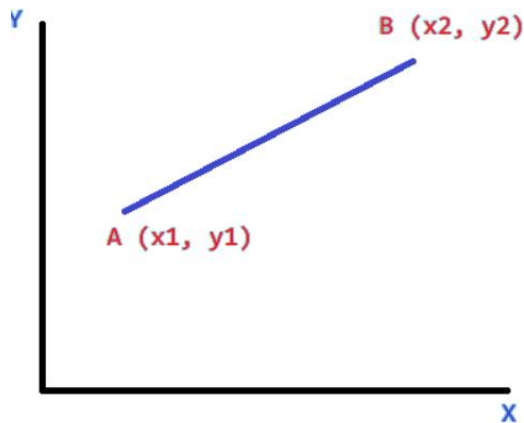


$N_j=[N_{j0}, N_{j1}, \dots, N_{jm}]$

# How the algorithm learn?

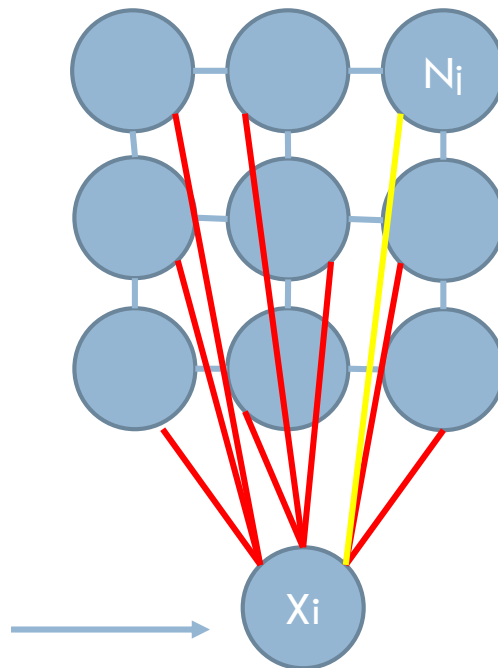
## 3) Matching:

- Compare each input with all nodes in the grid
- Select the best matching unit (BMU)



$$AB = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

$X_0, X_1, X_2, \dots, X_n$



$N_j = [N_{j0}, N_{j1}, \dots, N_{jm}]$

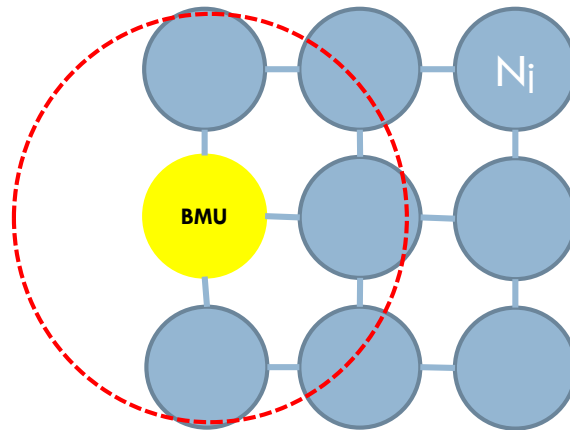
$$D_{ij} = \sqrt{\sum_{k=1}^m (x_{ik} - N_{jk})^2}$$

$X_i = [X_{i0}, X_{i1}, \dots, X_{im}]$



# How the algorithm learn?

**4) Updating:** update each node in the range of the BMU



$$N_j = N_j + \eta w_j (x_i - N_j)$$

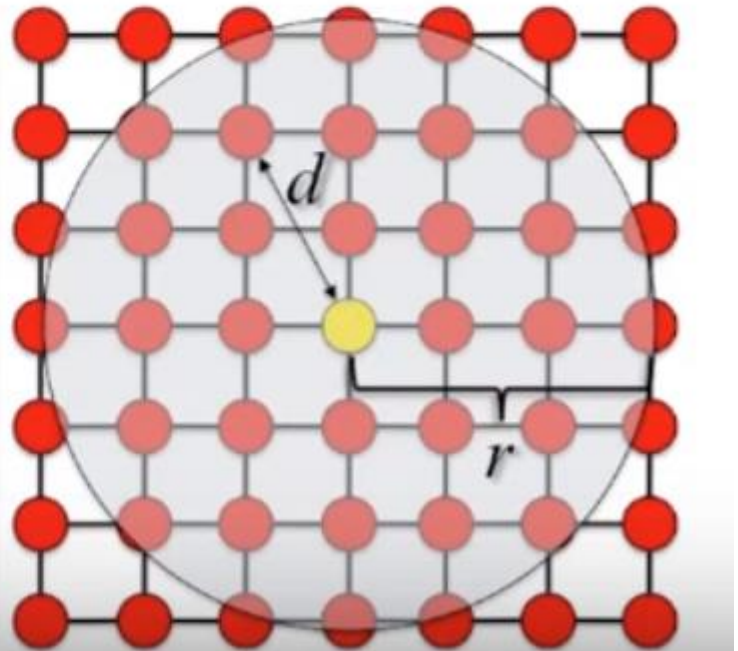
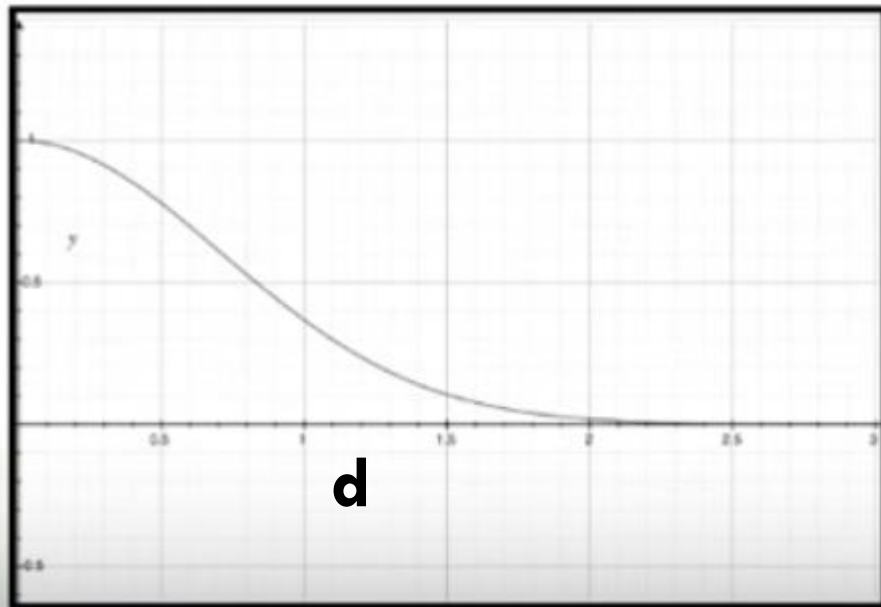
Where  $w_j$  is a weight parameter that depends on the distance between the node and the BMU and  $\eta$  is the learning rate

**5) Continuation:** keep returning to step 3 until the feature map stops changing

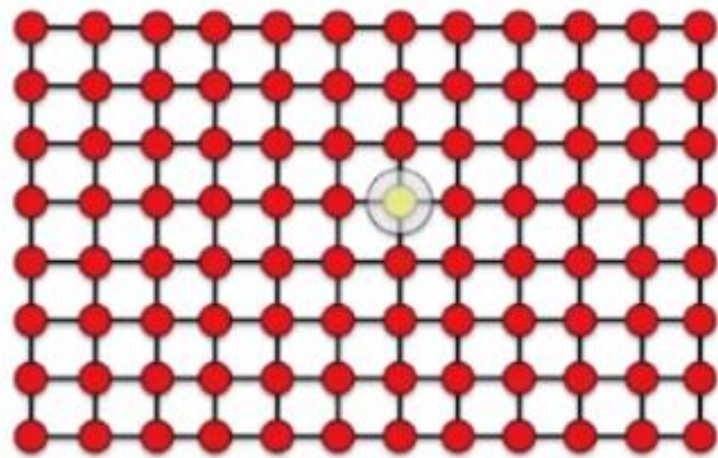
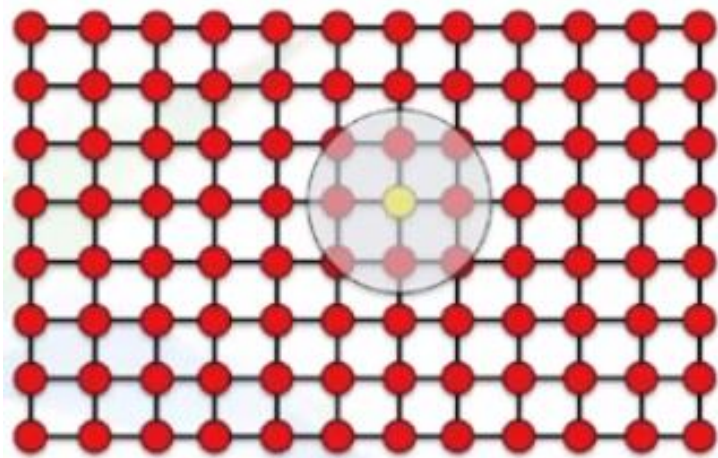
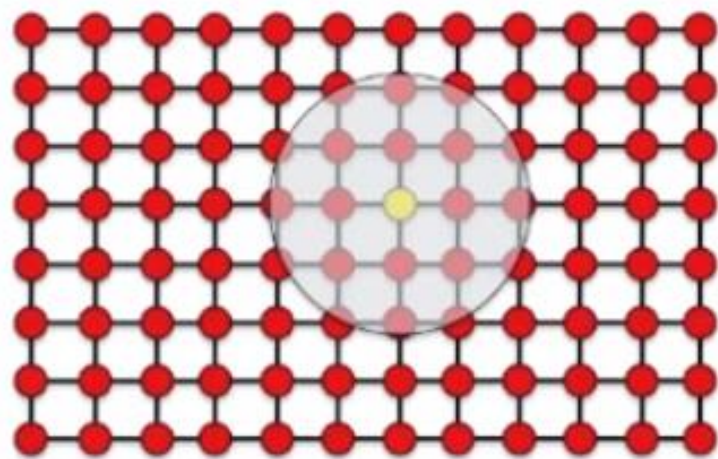
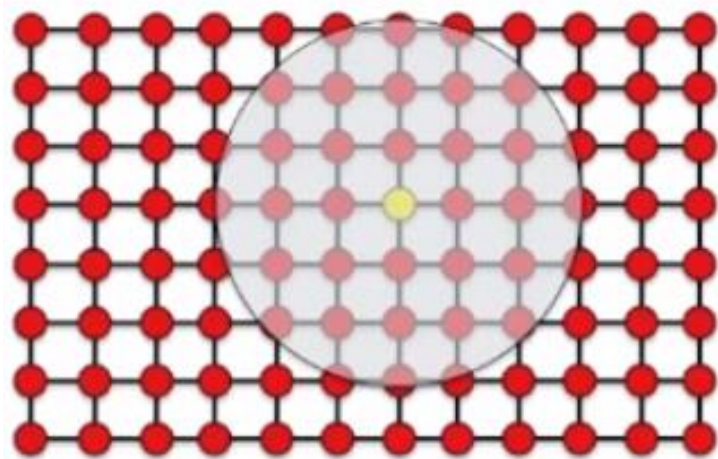
- Repeat with different input
  - Repeat with smaller radius (Go again with the input but with smaller radius)

$$w_j = e^{-\frac{d_j^2}{2r^2}}$$

**w**



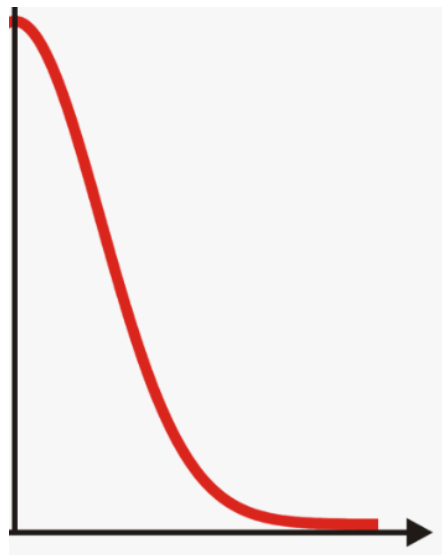
Repeat with smaller radius



# How to calculate radius

## □ Radius

$$r(T) = r(0) * e^{\left(-t \left(\frac{\ln(\text{Grid size})}{\text{Max } t}\right)\right)}$$



**Grid Size**

**Initial**  
 $r = 50 * \exp(0) = 50$

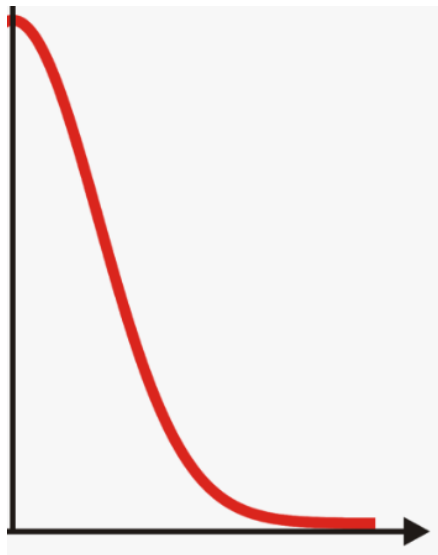
**Last iteration**  
 $r = 50 * 1 / 50 = 1$

46.237102  
44.463219  
42.757392  
41.117008  
39.539557  
38.022626  
36.563891  
35.161120  
33.812167  
32.514966  
31.267532  
30.067956  
28.914401  
27.805103  
26.738362  
25.712547  
24.726088  
23.777473  
22.865253  
21.988029  
21.144460  
20.333255  
19.553171  
18.803015  
18.081639  
17.387939  
16.720852  
16.079358  
15.462475

# How to calculate learning rate

## □ Learning Rate

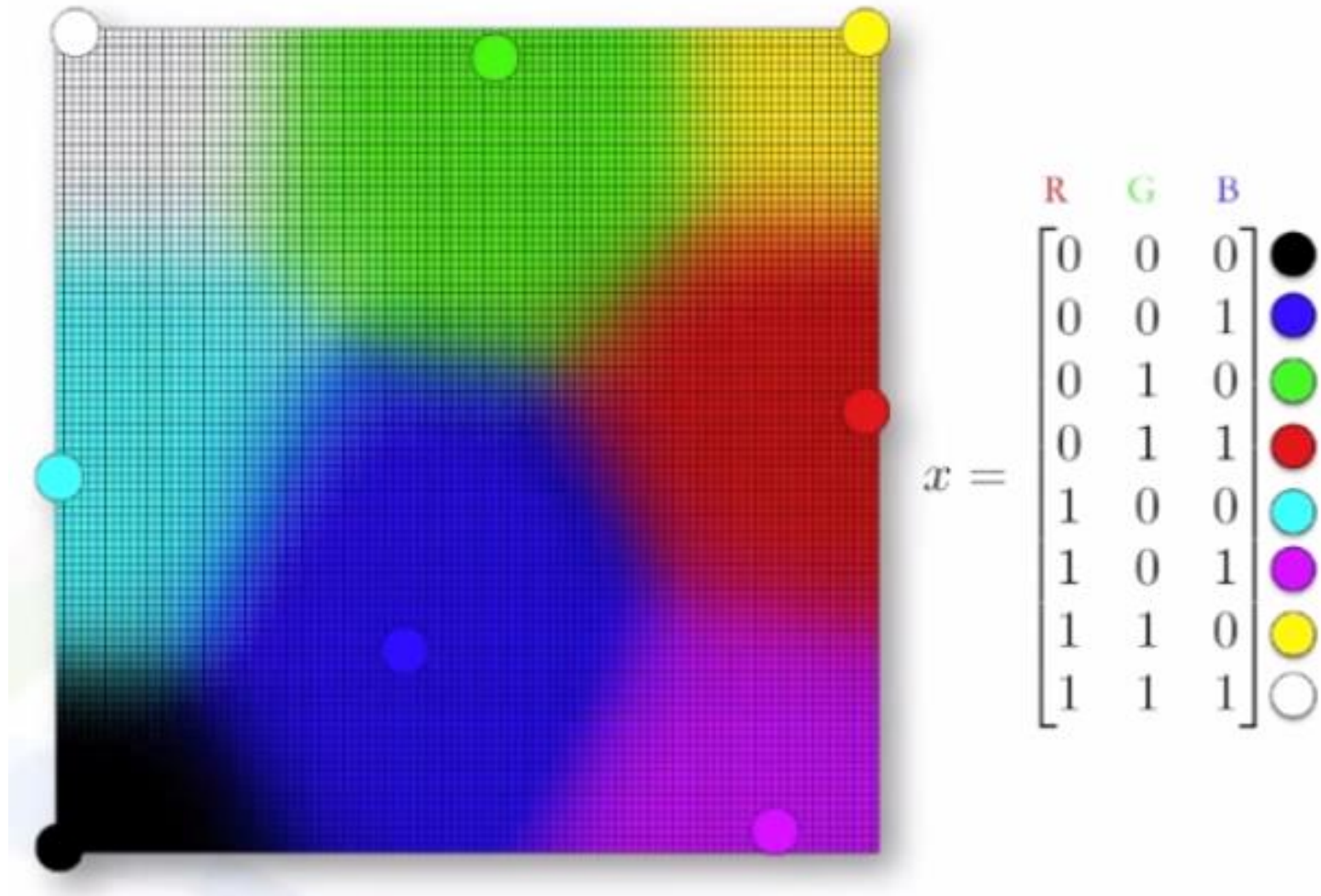
$$\alpha(T) = \alpha(0) * e^{\left(-t \left(\frac{\ln(\text{Grid size})}{\text{Max } t}\right)\right)}$$



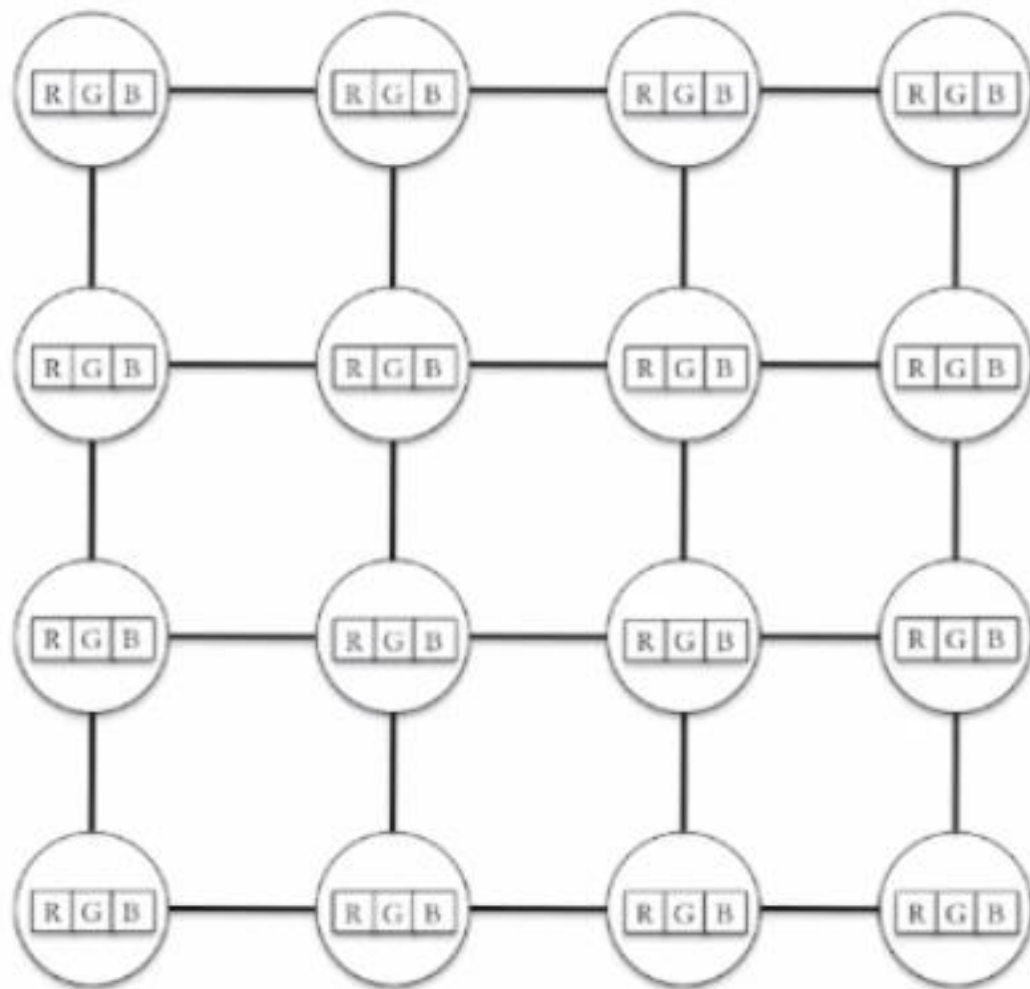
```
1.000000  
0.961635  
0.924742  
0.889264  
0.855148  
0.822340  
0.790791  
0.760453  
0.731278  
0.703222  
0.676243  
0.650299  
0.625351  
0.601359  
0.578288  
0.556102  
0.534767  
0.514251  
0.494522  
0.475549  
0.457305  
0.439761  
0.422800
```



# Colour Example

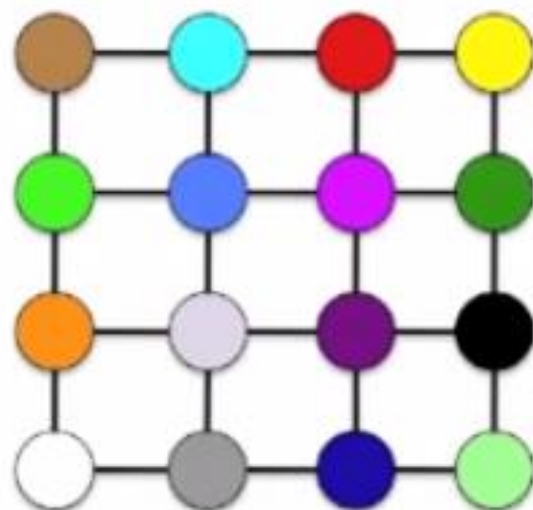


# Grid



# Initialize Randomly

- Initialize each node with random features (colors)

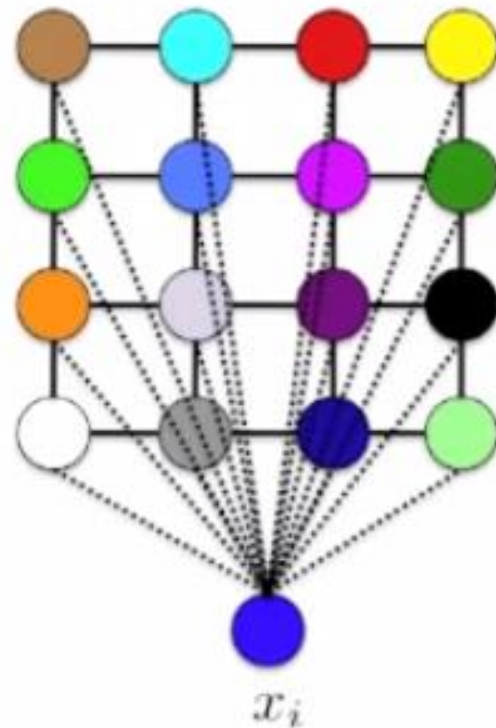




# Compare with Input

- Compare each input with all nodes in the grid

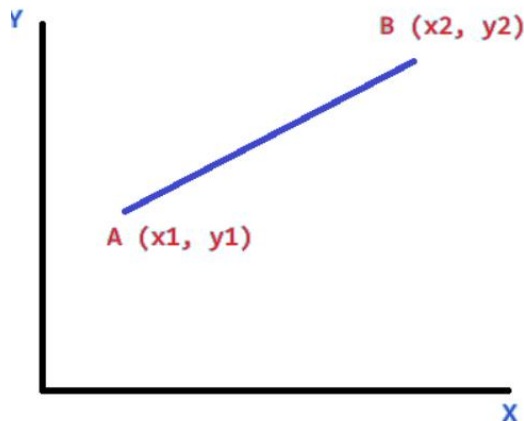
$$D_{ij} = \sqrt{\sum_{k=1}^m (x_{ik} - N_{jk})^2}$$



# Compare with Input

- Compare each input with all nodes in the grid

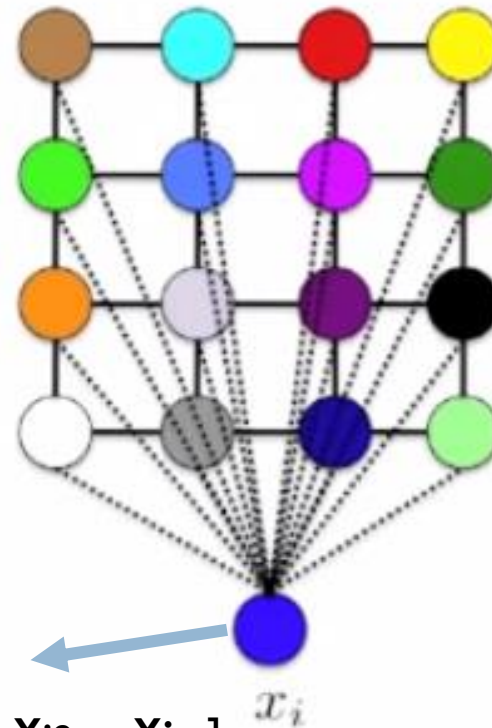
$$D_{ij} = \sqrt{\sum_{k=1}^m (x_{ik} - N_{jk})^2}$$



$$AB = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

Vector

$N_j = [N_{j0}, N_{j1}, N_{j2}, \dots, N_{jm}]$

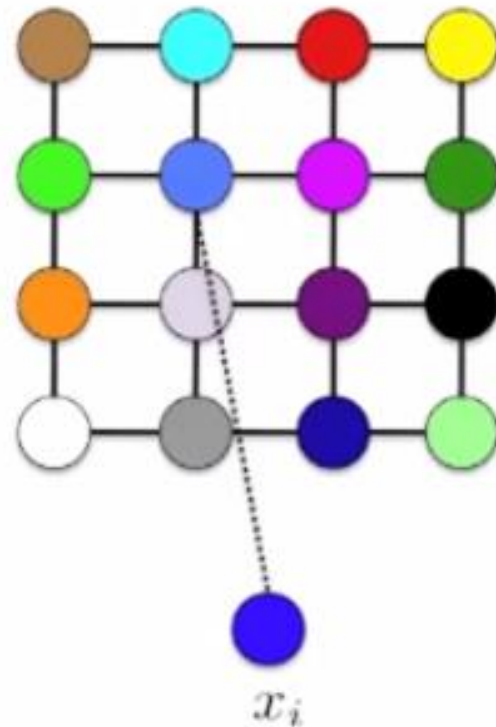


Vector

$X_i = [X_{i0}, X_{i1}, X_{i2}, \dots, X_{im}]$

# Best Matching Unit

- Select the node with minimum distance or the best matching unit (BMU)

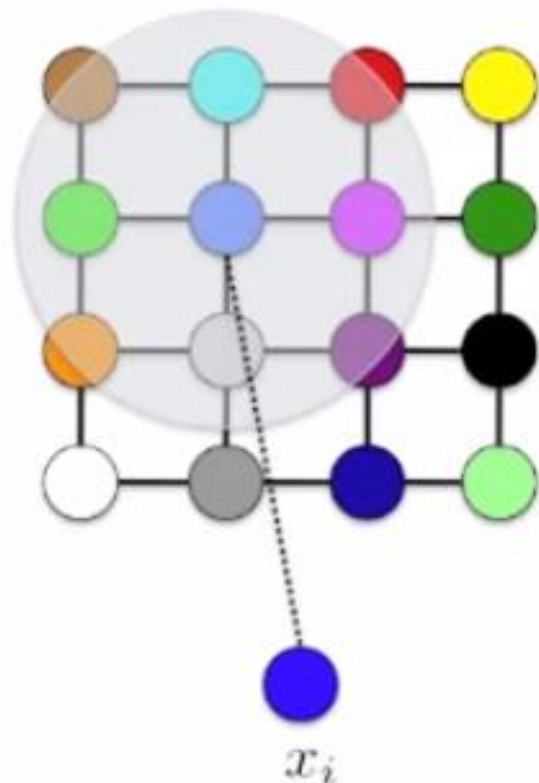


# Update

- For each node in the range of the BMU

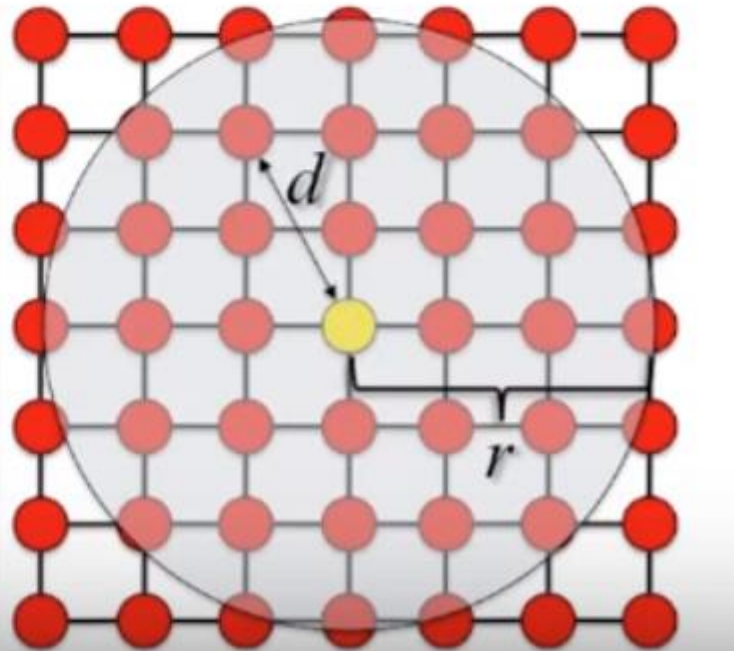
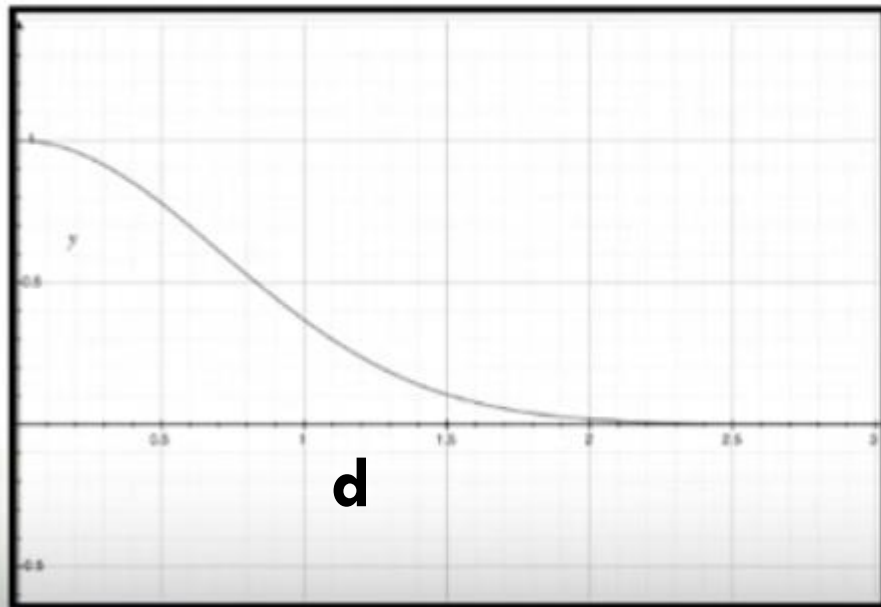
$$N_j = N_j + \eta w_j (x_i - N_j)$$

- Where  $w_j$  is a weight parameter that depends on the distance between the node and the BMU and  $\eta$  is the learning rate



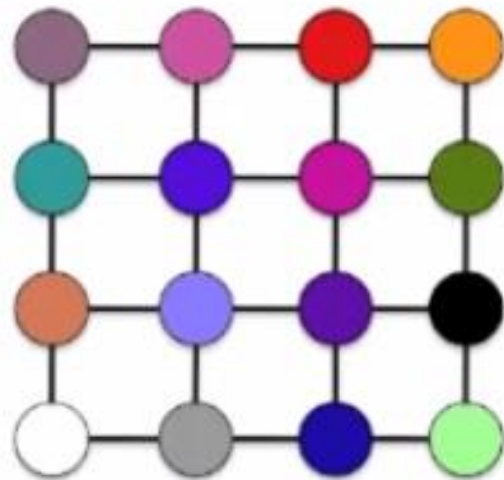
$$w_j = e^{-\frac{d_j^2}{2r^2}}$$

**w**



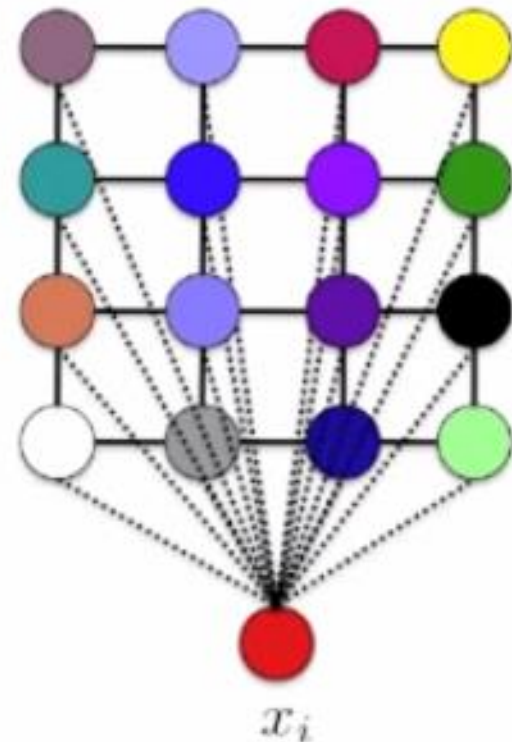
# Compare with Input

- Repeat the process with different input



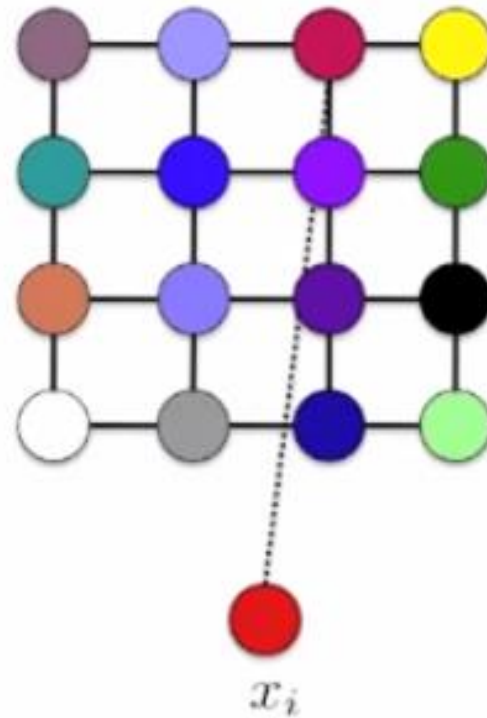
# Compare with Input

- Repeat the process with different input



# Compare with Input

- Repeat the process with different input





Repeat with smaller radius

