



# TUTORIAL OF WEEK 11

INT301 Bio-computation, Week 11, 2025



# Example of Simple Competitive Learning

- We are given 6 data samples:

$$\begin{array}{c} [0 \ 1 \ 1] \\ [0.5 \ 0.5 \ 0.5] \end{array} \quad \begin{array}{c} [1 \ 1 \ 0.5] \\ [0.4 \ 0.6 \ 0.5] \end{array} \quad \begin{array}{c} [0.2 \ 0.2 \ 0.2] \\ [0 \ 0 \ 0] \end{array}$$

- 3 output units each associated with a prototype vector.
- Learning rate: 0.5
- You are required to update the prototype vectors using Simple Competitive Learning (SCL) according to such updating rule:

$$\Delta w_{ji} = \eta y_j (x_i - w_{ji})$$
$$\begin{cases} y_{j^*} = 1 \\ y_j = 0 \quad \text{if} \quad j \neq j^* \end{cases}$$

# Example of Simple Competitive Learning

- Question 1: The 3 prototype vectors are randomly initialized as:

$$[0.14 \ 0.75 \ 0.71] \ [0.99 \ 0.51 \ 0.37] \ [0.73 \ 0.81 \ 0.87]$$

and the winner units are determined by minimizing the Euclidean distance. Please update the vectors one epoch and list the clustering result.

$$\Delta w_{ji} = \eta y_j (x_i - w_{ji})$$

Training on Input Vectors

Input vector # 1:  $[0.00 \ 1.00 \ 1.00]$  Computing the distance between Input vector to each prototype vector:  
[0.14 0.75 0.71]  $[0.99 \ 0.51 \ 0.37]$   $[0.73 \ 0.81 \ 0.87]$   $j \neq j^*$

$$\begin{array}{ccc} \downarrow & \downarrow & \downarrow \\ 0.1662 & 1.6171 & 0.5859 \end{array}$$

Winning weight vector # 1: [ 0.14 0.75 0.71 ]

Updated winning weight vector:

$$[0.14 \ 0.75 \ 0.71] + 0.5 * ([0.00 \ 1.00 \ 1.00] - [0.14 \ 0.75 \ 0.71])$$
$$[0.07 \ 0.87 \ 0.85] \ [0.99 \ 0.51 \ 0.37] \ [0.73 \ 0.81 \ 0.87]$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 2: [ 1.00 1.00 0.5 ]

Computing the distance between Input vector to each prototype vector:

[0.07 0.87 0.85]	[ 0.99 0.51 0.37 ]	[ 0.73 0.81 0.87 ]
↓	↓	↓
1.0043	0.2571	0.2459

Winning weight vector # 3: [ 0.73 0.81 0.87 ]

Updated winning weight vector:

$$[ 0.73 0.81 0.87 ] + 0.5 * ([ 1.00 1.00 0.5 ] - [ 0.73 0.81 0.87 ])$$
$$[ 0.07 0.87 0.85 ] [ 0.99 0.51 0.37 ] [ \mathbf{0.87} \mathbf{0.90} \mathbf{0.69} ]$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 3: [ 0.2 0.2 0.2 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{matrix} [0.07 & 0.87 & 0.85] & [0.99 & 0.51 & 0.37] & [0.87 & 0.90 & 0.69] \\ \downarrow & \downarrow & \downarrow \\ 0.8883 & 0.7491 & 1.1790 \end{matrix}$$

Winning weight vector # 2: [ 0.99 0.51 0.37 ]

Updated winning weight vector:

$$\begin{matrix} [0.99 & 0.51 & 0.37] + 0.5 * ([0.2 & 0.2 & 0.2] - [0.99 & 0.51 & 0.37]) \\ [0.07 & 0.87 & 0.85] & [0.59 & 0.36 & 0.29] & [0.87 & 0.90 & 0.69] \end{matrix}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 4: [ 0.5 0.5 0.5 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{matrix} [0.07 & 0.87 & 0.85] & [0.59 & 0.36 & 0.29] & [0.87 & 0.90 & 0.69] \\ \downarrow & \downarrow & \downarrow \\ 0.4443 & 0.0718 & 0.3330 \end{matrix}$$

Winning weight vector # 2: [ 0.59 0.36 0.29 ]

Updated winning weight vector:

$$\begin{matrix} [0.59 & 0.36 & 0.29] + 0.5 * ([0.5 & 0.5 & 0.5] - [0.59 & 0.36 & 0.29]) \\ [0.07 & 0.87 & 0.85] & [0.55 & 0.43 & 0.39] & [0.87 & 0.90 & 0.69] \end{matrix}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 5: [ 0.4 0.6 0.5 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{array}{ccc} [0.07 \ 0.87 \ 0.85] & [0.55 \ 0.43 \ 0.39] & [0.87 \ 0.90 \ 0.69] \\ \downarrow & \downarrow & \downarrow \\ 0.3043 & 0.0635 & 0.3470 \end{array}$$

Winning weight vector # 2: [ 0.59 0.36 0.29 ]

Updated winning weight vector:

$$\begin{aligned} & [0.55 \ 0.43 \ 0.39] + 0.5 * ([0.4 \ 0.6 \ 0.5] - [0.55 \ 0.43 \ 0.39]) \\ & [0.07 \ 0.87 \ 0.85] [0.47 \ 0.51 \ 0.45] [0.87 \ 0.90 \ 0.69] \end{aligned}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 6: [ 0.0 0.0 0.0 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{matrix} [0.07 & 0.87 & 0.85] & [0.47 & 0.51 & 0.45] & [0.87 & 0.90 & 0.69] \\ \downarrow & \downarrow & \downarrow \\ 1.4843 & 0.6835 & 2.0430 \end{matrix}$$

Winning weight vector # 2: [ 0.47 0.51 0.45 ]

Updated winning weight vector:

$$\begin{matrix} [0.47 & 0.51 & 0.45] + 0.5 * ([0.0 & 0.0 & 0.0] - [0.47 & 0.51 & 0.45]) \\ [0.07 & 0.87 & 0.85] & [0.24 & 0.26 & 0.22] & [0.87 & 0.90 & 0.69] \end{matrix}$$

Finish of Epoch 1

# Example of Simple Competitive Learning

Clusters after epoch 1:

Weight vector # 1:	[ 0.07    0.87    0.85 ]
Input vector # 1:	[ 0.00    1.00    1.00 ]
Weight vector # 2:	[ 0.24    0.26    0.22 ]
Input vector # 3:	[ 0.20    0.20    0.20 ]
Input vector # 4:	[ 0.50    0.50    0.50 ]
Input vector # 5:	[ 0.40    0.60    0.50 ]
Input vector # 6:	[ 0.00    0.00    0.00 ]
Weight vector # 3:	[ 0.87    0.90    0.69 ]
Input vector # 2:	[ 1.00    1.00    0.50 ]

# Example of Simple Competitive Learning

- Question 2: Based on the 3 prototype vectors initialized in question 1 and the winner units are determined by maximizing inner product. Please update the vectors one epoch and list the clustering result again.

Training on Input Vectors

Input vector # 1: [ 0.00 1.00 1.00 ]

Computing the distance between Input vector to each prototype vector:

[0.14 0.75 0.71]	[ 0.99 0.51 0.37 ]	[ 0.73 0.81 0.87 ]
↓	↓	↓
1.4600	0.8800	1.6800

Winning weight vector # 3: [ 0.73 0.81 0.87 ]

Updated winning weight vector:

$$[ 0.73 0.81 0.87 ] + 0.5 * ([ 0.00 1.00 1.00 ] - [ 0.73 0.81 0.87 ]) \\ [ 0.14 0.75 0.71 ] [ 0.99 0.51 0.37 ] [ 0.37 0.91 0.94 ]$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 2: [ 1.00 1.00 0.5 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{matrix} [0.14 & 0.75 & 0.71] & [0.99 & 0.51 & 0.37] & [0.37 & 0.91 & 0.94] \\ \downarrow & \downarrow & \downarrow \\ 1.2450 & 1.6850 & 1.7500 \end{matrix}$$

Winning weight vector # 3: [ 0.37 0.91 0.94 ]

Updated winning weight vector:

$$\begin{matrix} [0.37 & 0.91 & 0.94] + 0.5 * ([1.00 & 1.00 & 0.5] - [0.37 & 0.91 & 0.94]) \\ [0.14 & 0.75 & 0.71] [0.99 & 0.51 & 0.37] [0.69 & 0.96 & 0.72] \end{matrix}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 3: [ 0.2 0.2 0.2 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{matrix} [0.14 & 0.75 & 0.71] & [0.99 & 0.51 & 0.37] & [0.69 & 0.96 & 0.72] \\ \downarrow & \downarrow & \downarrow \\ 0.3200 & 0.3740 & 0.4740 \end{matrix}$$

Winning weight vector # 3: [0.69 0.96 0.72 ]

Updated winning weight vector:

$$\begin{matrix} [0.69 & 0.96 & 0.72] + 0.5 * ([0.2 & 0.2 & 0.2] - [0.69 & 0.96 & 0.72]) \\ [0.14 & 0.75 & 0.71] [0.99 & 0.51 & 0.37] [\mathbf{0.45} & \mathbf{0.58} & \mathbf{0.46}] \end{matrix}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 4: [ 0.5 0.5 0.5 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{matrix} [0.14 & 0.75 & 0.71] & [0.99 & 0.51 & 0.37] & [0.45 & 0.58 & 0.46] \\ \downarrow & \downarrow & \downarrow \\ 0.8000 & 0.9350 & 0.7450 \end{matrix}$$

Winning weight vector # 2: [ 0.99 0.51 0.37 ]

Updated winning weight vector:

$$\begin{matrix} [0.99 & 0.51 & 0.37] + 0.5 * ([0.5 & 0.5 & 0.5] - [0.99 & 0.51 & 0.37]) \\ [0.14 & 0.75 & 0.71] & [0.71 & 0.51 & 0.44] & [0.45 & 0.58 & 0.46] \end{matrix}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 5: [ 0.4 0.6 0.5 ]

Computing the distance between Input vector to each prototype vector:

$$\begin{array}{ccc} [0.14 \quad 0.75 \quad 0.71] & [0.71 \quad 0.51 \quad 0.44] & [0.45 \quad 0.58 \quad 0.46] \\ \downarrow & \downarrow & \downarrow \\ 0.8610 & 0.8100 & 0.7580 \end{array}$$

Winning weight vector # 1: [ 0.14 0.75 0.71 ]

Updated winning weight vector:

$$\begin{aligned} & [0.14 \quad 0.75 \quad 0.71] + 0.5 * ([0.4 \quad 0.6 \quad 0.5] - [0.14 \quad 0.75 \quad 0.71]) \\ & [0.27 \quad 0.68 \quad 0.61] \end{aligned}$$

# Example of Simple Competitive Learning

Training on Input Vectors

Input vector # 6: [ 0.0 0.0 0.0 ]

Computing the distance between Input vector to each prototype vector:

[0.27 0.68 0.61]	[ 0.71 0.51 0.44 ]	[0.45 0.58 0.46 ]
↓	↓	↓
0.0000	0.0000	0.0000

Winning weight vector # 1: [ 0.27 0.61 0.61 ]

Updated winning weight vector:

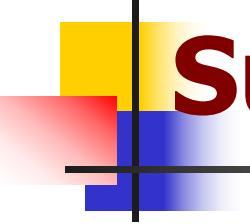
$$[ 0.27 0.68 0.61 ] + 0.5 * ([ 0.0 0.0 0.0 ] - [ 0.27 0.68 0.61 ]) \\ [ \mathbf{0.14} \mathbf{0.34} \mathbf{0.31} ] [ 0.71 0.51 0.44 ] [ 0.45 0.58 0.46 ]$$

Finish of Epoch 1

# Example of Simple Competitive Learning

Clusters after epoch 1:

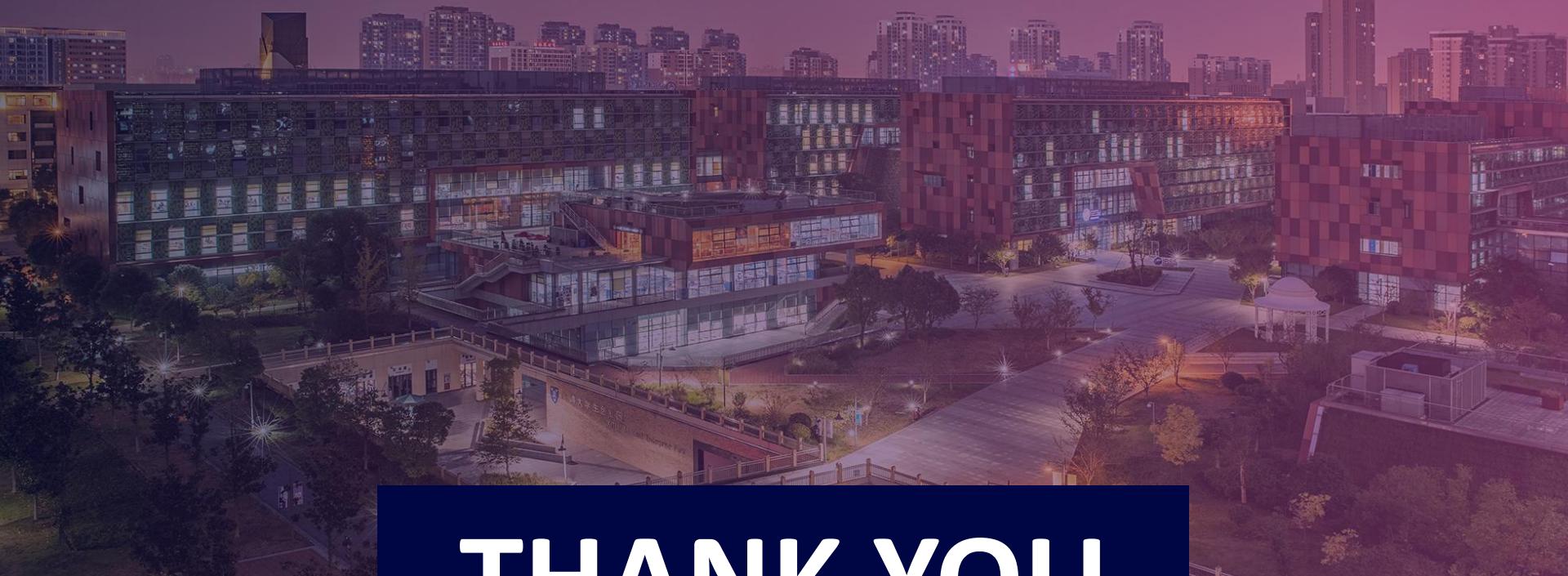
Weight vector # 1:	[ 0.14    0.34    0.31 ]
Input vector # 5:	[ 0.40    0.60    0.50 ]
Input vector # 6:	[ 0.00    0.00    0.00 ]
Weight vector # 2:	[ 0.71    0.51    0.44 ]
Input vector # 4:	[ 0.50    0.50    0.50 ]
Weight vector # 3:	[ 0.45    0.58    0.46 ]
Input vector # 1:	[ 0.00    1.00    1.00 ]
Input vector # 2:	[ 1.00    1.00    0.50 ]
Input vector # 3:	[ 0.20    0.20    0.20 ]



# Summary

- SCL is sensitive to the choice of winning unit criterion
- Different similarity measures lead to different final prototypes

Will altering the presentation order of the training samples produce different update trajectories and hence different converged vectors?



# THANK YOU



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