



# 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:

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

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

$\downarrow$	$\downarrow$	$\downarrow$
0.1662	1.6171	0.5859

Winning weight vector # 1:  $[0.14 \ 0.75 \ 0.71]$

Updated winning weight vector:

$[0.14 \ 0.75 \ 0.71] + 0.5 * ([0.00 \ 1.00 \ 1.00] - [0.14 \ 0.75 \ 0.71])$

# 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:

$$\begin{matrix} [0.07 & 0.87 & 0.85] & [0.99 & 0.51 & 0.37] & [0.73 & 0.81 & 0.87] \\ \downarrow & \downarrow & \downarrow \\ 1.0043 & 0.2571 & 0.2459 \end{matrix}$$

Winning weight vector # 3: [ 0.73 0.81 0.87 ]

Updated winning weight vector:

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

# 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:

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

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:

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

Winning weight vector # 2: [ 0.59 0.36 0.29 ]

Updated winning weight vector:

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

# 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 ]+[ 0.00 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:

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

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:

[0.14 0.75 0.71]	[ 0.99 0.51 0.37 ]	[0.69 0.96 0.72 ]
↓	↓	↓
0.3200	0.3740	0.4740

Winning weight vector # 3: [0.69 0.96 0.72 ]

Updated winning weight vector:

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

# 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:

[0.14 0.75 0.71]	[ 0.71 0.51 0.44 ]	[0.45 0.58 0.46 ]
↓	↓	↓
0.8610	0.8100	0.7580

Winning weight vector # 1: [ 0.14 0.75 0.71 ]

Updated winning weight vector:

$$[ 0.14 0.75 0.71 ] + 0.5 * ([ 0.4 0.6 0.5 ] - [ 0.14 0.75 0.71 ]) \\ [ 0.27 \mathbf{0.68} \mathbf{0.61} ] [ 0.71 0.51 0.44 ] [ 0.45 0.58 0.46 ]$$

# 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:

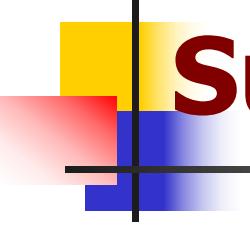
$$[ 0.27 0.68 0.61 ] + 0.5 * ([ 0.0 0.0 0.0 ] - [ 0.27 0.68 0.61 ]) \\ [ \mathbf{0.14} \mathbf{0.34} \mathbf{0.31} ] [ 0.71 0.51 0.44 ] [ 0.45 0.58 0.46 ]$$

## Finish of Epoch 1

# Example of Simple Competitive Learning

Clusters after epoch 1:

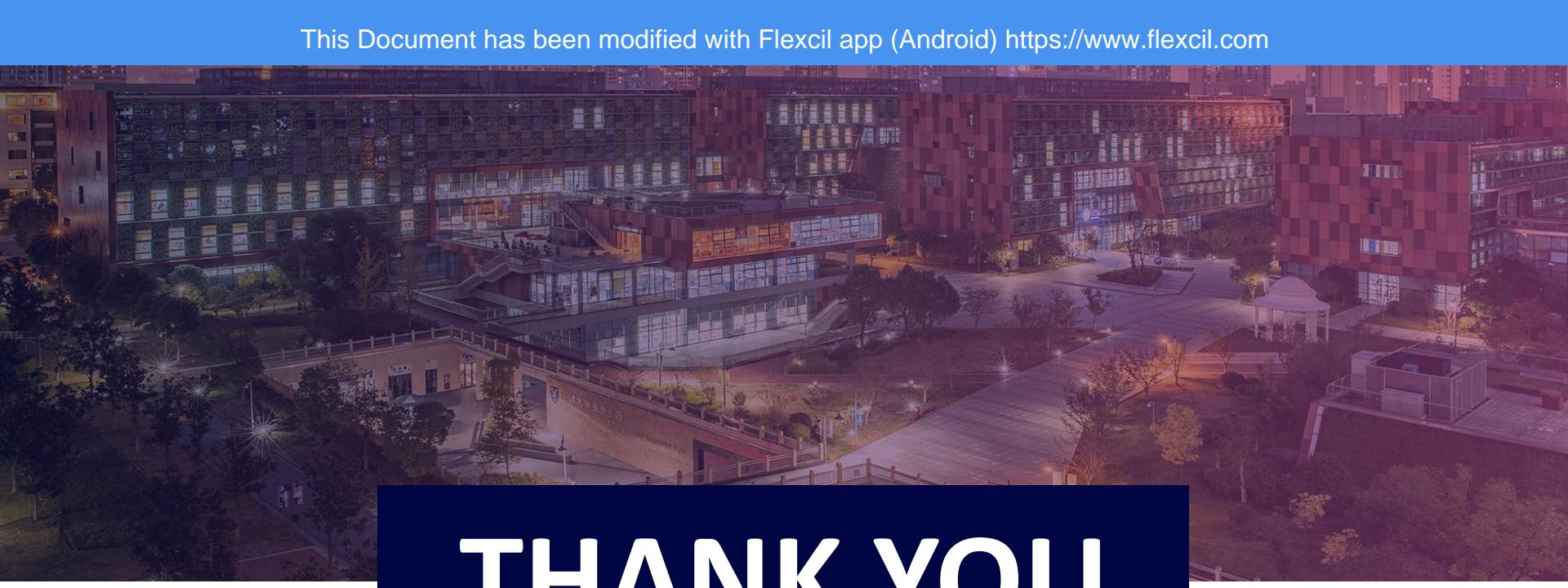
Weight vector # 1:	[ 0.14    0.34    0.31 ]
Input vector # 5:	[ 0.40    0.60    0.50 ]
Input vector # 6:	[ 0.00    0.00    0.00 ]
Weight vector # 2:	[ 0.71    0.51    0.44 ]
Input vector # 4:	[ 0.50    0.50    0.50 ]
Weight vector # 3:	[ 0.45    0.58    0.46 ]
Input vector # 1:	[ 0.00    1.00    1.00 ]
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? 会的，通常收敛到局部极小解。



THANK YOU



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