



TUTORIAL OF WEEK 11

INT301 Bio-computation, Week 11, 2025



Example of Simple Competitive Learning

- We are given 6 data samples:

$$\begin{array}{ccc} [0 \ 1 \ 1] & [1 \ 1 \ 0.5] & [0.2 \ 0.2 \ 0.2] \\ [0.5 \ 0.5 \ 0.5] & [0.4 \ 0.6 \ 0.5] & [0 \ 0 \ 0] \end{array}$$

- 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:

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

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] if $j \neq j^*$

↓

0.1662

↓

1.16171

↓

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:

[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] [**0.87 0.90 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:

[0.07 0.87 0.85]	[0.99 0.51 0.37]	[0.87 0.90 0.69]
↓	↓	↓
0.8883	0.7491	1.1790

Winning weight vector # 2: [0.99 0.51 0.37]

Updated winning weight vector:

[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]

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:

[0.07 0.87 0.85]	[0.59 0.36 0.29]	[0.87 0.90 0.69]
↓	↓	↓
0.4443	0.0718	0.3330

Winning weight vector # 2: [0.59 0.36 0.29]

Updated winning weight vector:

[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]

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:

[0.07 0.87 0.85]	[0.55 0.43 0.39]	[0.87 0.90 0.69]
↓	↓	↓
0.3043	0.0635	0.3470

Winning weight vector # 2: [0.59 0.36 0.29]

Updated winning weight vector:

[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]

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:

[0.07	0.87	0.85]	[0.47	0.51	0.45]	[0.87	0.90	0.69]
↓			↓			↓		
1.4843			0.6835			2.0430		

Winning weight vector # 2: [0.47 0.51 0.45]

Updated winning weight vector:

[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]

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:

[0.14	0.75	0.71]	[0.99	0.51	0.37]	[0.37 0.91 0.94]
↓			↓		↓	
1.2450			1.6850		1.7500	

Winning weight vector # 3: [0.37 0.91 0.94]

Updated winning weight vector:

[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**]

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] [**0.45 0.58 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:

[0.14 0.75 0.71]	[0.99 0.51 0.37]	[0.45 0.58 0.46]
↓	↓	↓
0.8000	0.9350	0.7450

Winning weight vector # 2: [0.99 0.51 0.37]

Updated winning weight vector:

[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]

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 0.68 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])
[**0.14** **0.34** **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, 通常收敛到局部最优解



THANK YOU



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