

# Lecture 3: ML Terminology

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Over the last two lectures, we reviewed probability and linear algebra. Now, we introduce some basic ML terminology.

## 1 An Example

This example is in continuation from Lecture 2. We have a graph  $G = (V, E)$ , where  $V = \{1, \dots, n\}$  is the set of vertices, and  $E$  is the set of edges. Define the value  $x_i(t)$  to be the ‘opinion’ of node  $i$  at time  $t$ .  $x_1(0), \dots, x_n(0)$  are given. For  $t \geq 0$ ,

$$x_i(t+1) = \frac{\sum_{j \in \mathcal{N}(i)} x_j(t)}{|\mathcal{N}(i)|} \quad (1)$$

where  $\mathcal{N}(i)$  denotes the set of neighbors of node  $i$ . That is, the opinion of  $i$  at time  $t+1$  is the average of the opinions of its neighbors at time  $t$ .

Consider  $\mathbf{x}(t) = [x_1(t) \cdots x_n(t)]^T$ . Define  $\mathbf{A} \in \mathbb{R}^{n \times n}$  such that  $\mathbf{A}\mathbf{x}(t) = \mathbf{x}(t+1)$ . From (1),  $\mathbf{A}$  is a doubly-stochastic matrix, i.e., its entries are non-negative, and its rows and columns add up to 1.

We state the following result from [1]. A stochastic matrix  $\mathbf{M}$  is called semi-positive if all entries of some power  $\mathbf{M}^\alpha$  are positive.

**Theorem 1.1.** *If  $\mathbf{M} \in \mathbb{R}^{n \times n}$  is a semi-positive doubly stochastic matrix, then  $\lim_{t \rightarrow \infty} \mathbf{M}^t = \frac{1}{n} \mathbf{J}$ , where  $\mathbf{J} \in \mathbb{R}^{n \times n}$  with all entries 1.*

Thus, if  $\mathbf{A}$  is semi-positive and doubly stochastic, then for all nodes  $i$ ,  $\lim_{t \rightarrow \infty} x_i(t) = \frac{\sum_{j=1}^n x_j(0)}{n}$

## 2 Image Classification Problem

## 3 Training and Validation Sets

## References

- [1] S. Baik and K. Bang. Limit theorem of the doubly stochastic matrices. *Kangweon-Kyungki Mathematics Journal*, 11(2):155–160, 2003.