

Artificial Intelligence Methods for Social Good

MI-4 [Optimization]: Influence Maximization

08-537 (9-unit) and 08-737 (12-unit)

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Outline

- ▶ Propagation Process
- ▶ Influence Propagation Models
 - ▶ Independent Cascade Model
 - ▶ Linear Threshold Model
- ▶ Influence Maximization Problem

Learning Objectives

- ▶ Understand the concept of
 - ▶ Submodular function
- ▶ Describe
 - ▶ Independent Cascade Model
 - ▶ Linear Threshold Model
 - ▶ Influence Maximization Problem
 - ▶ Greedy Algorithm for Influence Maximization Problem

Propagation Process

- ▶ **Viral propagation**
 - ▶ Virus/Rumors
 - ▶ Get infected immediately and spread automatically
 - ▶ Individual agent does not make decisions
- ▶ **Decision based models**
 - ▶ Individual agent makes decisions
 - ▶ Influence and adoption

Influence Response Function

- ▶ Discuss: when would you adopt a recommendation from your friends?

Influence Response Function

▶ Influence Response Function

▶ Independent Draws

- ▶ n friends recommend it to me
- ▶ $P(n) = 1 - (1 - p)^n$
- ▶ Diminishing return (concave function)

▶ Linear Threshold

- ▶ b percentage of my friends bought the item
- ▶ $P(b) = \delta(b > b_0)$
- ▶ Critical Mass

Influence Propagation Models

- ▶ Independent Cascade Model (Goldenberg, 2001)
 - ▶ Model I
 - ▶ Initial set of active nodes
 - ▶ Discrete time steps
 - ▶ On every step, an active node can activate connected neighbor with a probability $p_{v,w}$ (single chance, if failed, no more trial on this edge)
 - ▶ If v succeeds, w becomes active on the next time step
 - ▶ Process runs until no more activations possible

Influence Propagation Models

- ▶ Independent Cascade Model (Goldenberg, 2001)
 - ▶ Exp I

Quiz I

- ▶ How many time steps are needed to achieve global cascade in Exp I?
 - ▶ 2
 - ▶ 3
 - ▶ 4
 - ▶ 5

Influence Propagation Models

- ▶ Linear Threshold Model (M. Granovetter, 1978, T. Schelling, 1970, 1978)
 - ▶ Each node i has a threshold θ_i
 - ▶ Each edge has a weight w_{ij} indicating the influence of node i to node j
 - ▶ Activated if total weight of active neighbors exceeds threshold
 - ▶ Given initial set of active nodes, proceed iteratively with discrete time steps
 - ▶ Once activated, keep active
 - ▶ Model 2

Quiz 2

- ▶ Let $\theta_0 =$ common threshold, $N_0 =$ common number of neighbors. $w_{ij} = \frac{1}{N_0}$. Consider the following three scenarios
 - ▶ S1: $\theta_0 = a, N_0 = b$
 - ▶ S2: $\theta_0 = a + 0.1, N_0 = b$
 - ▶ S3: $\theta_0 = a, N_0 = b + 1$
- ▶ When $b > 1$, what is ordering of the probability of getting global cascade following the LTM model under this three scenarios?
 - ▶ A: $S1 \geq S2 \geq S3$
 - ▶ B: $S3 \geq S2 \geq S1$
 - ▶ C: $S2 \geq S1, S3 \geq S1$, relationship between $S2, S3$ is unknown

Influence Maximization Problem

- ▶ How to select initial nodes A_0 to maximize influence $\sigma(A_0)$, under the constraint that A_0 has no more than k nodes
 - ▶ Problem I
- ▶ NP-Hard (reduction from *Set Cover*, Kempe, Kleinberg & Tardos, 2003, 2005)

Greedy Algorithm

- ▶ Submodular Functions
 - ▶ Def I
 - ▶ Diminishing return (similar to concave function)
 - ▶ Exp: Team of defensive resources
- ▶ Greedy algorithm leads to $1 - \frac{1}{e}$ approximation for submodular monotone function
 - ▶ Exp: Maximum Coverage problem
- ▶ Theorem: In both LTM and ICM, $\sigma(A_0)$ is a submodular function (Kempe, Kleinberg & Tardos, 2003)
- ▶ Alg I

Extensions

- ▶ Further propagation
 - ▶ If I bought the product, then I need to decide whether or not to recommend to others
 - ▶ May choose the level of advocating effort based on my satisfaction, e.g., tweet about it, talk about it to my friend etc
- ▶ Compete with other sources of influence
 - ▶ Quit drinking/unhealthy behavior

Summary

- ▶ Propagation Process
- ▶ Influence Propagation Models
 - ▶ Independent Cascade Model
 - ▶ Linear Threshold Model
- ▶ Influence Maximization Problem

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- ▶ The slides are prepared based on lecture slides of Leonid Zhukov.