Artificial Intelligence Methods for Social Good

M1-4 [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 1
  - 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 1
Quiz 1

- How many time steps are needed to achieve global cascade in Exp 1?
  - 2
  - 3
  - 4
  - 5
Influence Propagation Models

  - Each node $i$ has a threshold $\theta_i$
  - 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 1
- NP-Hard (reduction from Set Cover, Kempe, Kleinberg & Tardos, 2003, 2005)
Greedy Algorithm

- Submodular Functions
  - Def 1
  - 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 1
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., twit 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
Acknowledgment

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