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- Yes there are issues in using ML, but we can still use it to help solve societal problems
- Look at Amulya's research to solve problems within low resource communities
 - Access to public health
 - Access to healthy food
 - Access to rehab facilities
 - Access to good governance
- How can we use AI to solve the above issues?

Example 1: HIV among homeless Youth

- Interventions among homeless youth
 - Try to raise awareness about HIV prevention
 - Urge them to adopt safer behaviors
 - Encourage them to spread the message among social circles
- Homeless shelters: Resource Constraints
 - Cannot intervene with every homeless youth
 - Rely on word-of-mouth effects to inform about HIV
- Standard influence maximization
 - Inputs:
 - Social network G
 - Influence Model I
 - Number of Nodes :
 - Outputs:
 - Optimal set of K nodes to maximize expected influence spread
- State of the art in influence maximization
 - Algorithms with theoretical guarantees
 - Greedy
 - Celf
 - TIM
 - Efficient heuristics
 - IRIE
 - Sketch based heuristics
 - But do these algorithms suffice
- uncertainties: what do they mean?
 - In social network structure, there are inconsistencies. For example, homeless people are not the most reliable so we assume that they are not 100% correct
 - Uncertainties about state of influence of nodes - uncertainty about which nodes influence each other
 - Critical challenges in the real world: multi-stage selection

- Problem - dynamic influence maximization under certainty

Pilot Study

- 3 pilot studies done with 173 homeless youth in LA
 - 2 different homeless shelters
 - Raise awareness
 - Head to head comparison of algorithms
- HEALER and DOSIM in the field,
 - Test healer and dosim
 - Verify need for using AI algorithms
- 1st pilot study
 - HEALER - 62 youth
- 2nd pilot study
 - DOSIM - 56 youth
- 3rd pilot study - around
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Possible Issues with pilot studies

- Different case study groups
 - Healer (62 youth)
 - DOSIM (56 youth)
 - Degree centrality (55 youth)
 - Groups from similar locations
 - Small size

Standard influence maximization

- Input
 - Social network G
 - Influence model I
 - Number of nodes to choose K
- Output
 - Optimal set of K nodes to maximize expected influence spread

Maximizing the spread of influence through a social network

Problem Setting

- Given
 - A limited budget B for initial advertising (give away free sample of product)
 - Estimates for influence between individuals
- Goal
 - Trigger a large cascade of influence (further adoptions of a product)
- Field of maximization is governed by S-curve of adoption
- Question:
 - Which set of individuals should B target at?
- Application besides product marketing
 - Spread an innovation
 - Detect stories in blogs

What we need

- Form models of influence in social networks
- Obtain data about particular network
- Devise algorithm to maximize spread of influence

Outline

- Models of influence
 - Linear threshold
 - Independent

General operational view

- A social network is represented as a directed graph, with each person (customer) as a node
- Nodes start either active or inactive
- An active node may trigger activation of neighboring nodes
- Monotonicity assumption: active nodes never deactivate

Linear Threshold Model

- A node v , has random threshold
 - Threshold between $0,1$
- A node v is influenced by each neighbor w according to a weight b such that
- A node v becomes active when at least fractions of its neighbors are active

Independent Cascade

- When node v becomes active

Critical challenge in the real world

- Strength of friendship can be in ranges
- Emerging trends in ML - blackbox, morality, ethics
(all negative)

- AI for social good
 - Despite issues with ML, AI has many positive impacts
 - How can AI machine learning solve societal problems
- How can we use AI to solve low resource communities problem?
 - No access to public health facilities
 - Healthy food
 - Access to rehab facilities
 - Access to good governance (farming communities)
- Many of the problems of low resource communities can be solved by increasing social networks (not necessarily online)
- Current problems: Typically NGO's who help these communities don't have enough money, time and volunteers
- How can AI be used to raise awareness of HIV prevention to homeless youth?
 - Homeless use 10x more likely to get affected by HIV

- Traditional method: NGO's conduct "interventions"
 - peer leader is given information and taught how to spread this information to their peers.
 - Urge them to adopt safer behaviours
 - Encourage to spread message
 - Diffusion of information
 - Resource constraints, cannot intervene on every homeless youth themselves.
 - Rely on word of mouth
 - GOAL: maximize number of youth who is informed about HIV
- Friend network, which people should we target to maximize spread of information (similar to viral marketing)
- Standard Influence maximization
 - Inputs:
 - Social network (G)
 - Influence model (I): description of how influence spreads between people
 - Number of Nodes to choose (K): depends on number of people in homeless shelter
 - Output
 - Optimal set of K nodes to maximize expected influence spread

- INFLUENCE maximization algorithms are not great with real problems
- Critical challenges in the real world: uncertainties
 - Uncertainty in social network structure
 - Homeless don't use social media (hard to gather data)
 - Uncertain about existence of
 - Uncertainty about state of influence of nodes
 - Unable to sample every node
 - Unknown which nodes are influenced and which are not
 - Multi-stage selection
 - Does not work for influence maximization
- given a social network $G = (V,E)$
 - Picking M subsets of nodes (M interventions organized by shelter) 3
 - Size of each subset is K (Max capacity of shelter) 4 leaders
- 2 algorithms to solve this problem, HEALER and DOSIM
 - 3 studies with 173 homeless youth in LA
 - 2 shelters (my friends place and Safe place for youth)
 - Raise awareness about HIV

- Goal: verify need for using AI algorithms, verify usability
- HEALER/DOSIM/ Degree Centrality - select peer leaders
 - Network application: interacts with homeless youth
 - Algorithm: determines which people should be recommended for the intervention
 - Shelter official collects information and refines algorithm (are people really friends or not)
 - Looks at social network data and selects a set of nodes for first intervention
 - What percentage of non peer leaders were informed about HIV after the interventions 70% healer and dosim
 - What percentage of informed, non peer leader, untested people made a proactive move afterwards and got an HIV test
 - Improvement of 150% over degree centrality
- Degree Centrality (selects most popular people)
 - Creates too many redundant edges
 - Influencing people who are already influenced
 - Cross community edges are poor

- Ignores cross community connections (only need to influence 1 person in a small group)
- What is wrong with the study
 - 3 different populations
 - Sample size too small
 - Randomness
 - Only ran the algorithm once
 - Can't run twice
 - Cant determine statistical significance
 - Different messages spread differently

11/14/19

1. Network application (connects with youth) Algorithm (Feedback)
2. # of non leaders informed: Healer and Dosim 70%
DC 27%
3. Maximizing the spread of influence through social network
 - a. How to mitigate/maximize spread of influence
4. Influence: can cover anything
(awareness/product/propaganda)

5. Scenario Given: a limited budget B for advertising, estimate **influence between individuals?**
 - a. Goal: trigger a large cascade of influence
 - b. Question : which set of individuals should B target
 - c. What you need:**
 - i. Form models of influence in social networks
 - ii. Obtain data about particular network (estimate inter personal influence)
 - iii. Devise algorithm to maximize spread of influence)
6. Viral marketing: goal - ensure further adoptions of a product
7. **S- curve of adoption: the spread of innovations curve**
 - a. Y: number of adopters
 - b. X: time S - top part: sold too much
8. Applications besides product marketing
 - a. Spread innovation
 - b. Detect stories in blogs
9. Models of influence
 - a. Two basic classes of diffusion models: threshold and cascade
 - b. Operational View

- i. A social network is represented as a directed graph and each node is a person/customer
 - ii. Nodes start either active or inactive (influenced or not)
 - iii. An active node may trigger activation of neighboring nodes
 - iv. Monotonicity assumption: active nodes never deactivate
 - v. Directed vs undirected graph (directional edges)
 - vi. Bidirectional normal friendship
10. Linear model threshold
 - a. A node v has random threshold (0.7, will not become influenced until 70% of people in social network become influenced)
 - b. Every relationship as different influential threshold number on a person (person around them .3 .5 .2 etc..)
11. Independent Cascade
 - a. Probability that if you get influenced the other person will get influenced
 - b. Person you rarely see, low percentage -- good friend, high
 - c. Strength of friendships

12. How to solve influence maximization problem
 - a. Set of K-nodes (initial adopters)
 - b. Greedy algorithm
 - i. Repeat procedure for K timesteps
 - ii. For each timestep: add a single node to the result set
 - iii. Results - 4 nodes
 - c. Also works with any algorithm that has diminishing returns property (return on investment) math name - submodularity
 - d. Influence maximization problem with diminishing returns property means always use the greedy algorithm
13. Real world algorithms: just give ranges of nodes from each other
 - a. Rank people based off of the discussion of their friends, how good of friends is the other person
 - b. Use ranges (be robust)
 - c. Assume there is
14. 3 ways games can be solved
 - a. Nash Equilibrium - what we are using
 - b. Dominance
 - c. Something else