Game-Theoretic Models of Information Overload in Social **Networks**

A Presentation for CS886

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Background

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- Some surveys claim the average person has five social media accounts and spends 1hr 40 mins on them every day [2].
- Increasing irrelevant updates on social media newsfeeds, or information overload.

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- Symmetric: requires consent from both sides to maintain tie eg., Facebook.
- Asymmetric: requires consent from only one side to maintain tie eg., Twitter.
- Authors mainly look at asymmetric social networks.

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- Makeup of newsfeed becomes very important to user.
- Mix of newsfeed is determined by the activity level of user's friends.
- How much one hears from one particular friend is not in user's control.

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 - Updates from friends are useful, but excessive updates have diminishing value.
 - Users can be partitioned as producers and consumers of information (80 20 rule).

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- Engagement: Users get frustrated by high update rate of followees and leave the social network.

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 - q_{ij} denotes utility consumer j derives from producer i's updates.
- Producer i updates at a frequency (rate) of r_i.
- Payoff for producer i is r_i times the number of followers he/she has.

Nash equilibrium

 In Nash equilibrium, each player knows the strategy of other players, and no player has anything to gain by changing their strategy.

Figure: Utility of Consumer for a Specific Producer



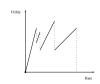
Figure : Utility of a Producer



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• Utility for consumer j is $U_j = \sum_{i \in C_j} r_i q_{ij} - \lambda (\sum_{i \in C_j} r_i)^2$. C_j being set of producers consumer j follows.

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- Utility of consumer has an inverse U-shape, which was predicted by literature.
- For producers, discontinuities at points where consumers stopped following.

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- For a consumer j, let x_i represent the indicator variable of whether j follows producer i or not (1 if follows, 0 otherwise).
- $U_j = \sum_i x_i r_i q_{ij} \lambda(\sum_i x_i r_i)$
- Very hard to solve if $x_i \in \{0, 1\}$, so simplify it to $x_i \in [0, 1]$.

Greedy Users

• If consumer j follows producer i fractionally ($x_{ij} < 1$) then assume that consumer j does not follow producer i ($x_{ij} = 0$).

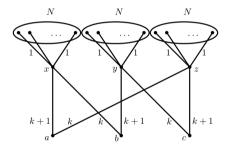
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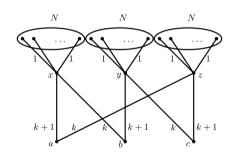
Definition

Consider consumer j and let $q_1 \ge ... \ge q_n$ be the sorted order of $q_{1j}, ..., q_{nj}$, and k be the largest index such that $\sum_{i=1}^k r_i \le q_k$. Under the greedy model, consumer j follows the k producers for who he has the highest quality and no one else.

Example for Nash Equilibrium not existing

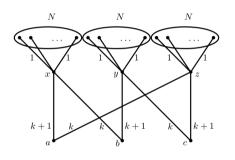


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 - Similar argument as previously nodes are topologically ordered and same induction argument.

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- Find a matching from all subsets of producers to consumers that are critical for them.
- Check matching to see if it is a possible Nash equilibrium.

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- C_i be the set of producers that consumer i follows.
- Let S be a function such that, $S(\sum_{i \in C_j} r_i)$ is the probability that consumer j stays in the social network.
- Expected utility of producer i is $U_i(r) = r_i(\sum_{j \in F_i} S(\sum_{i' \in C_i} r_{i'}))$.

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Cons

- In my opinion, not enough motivation as to why the particular models of social networks has been chosen.
- Matching Characterization not elaborated upon, perhaps had to be shortened?
- Understandable why fractional flow is considered in Followership model, but may not be realistic.
- Could look at Mixed State Nash equilibrium?

Pros

- Takes into account rate of information flow in social networks.
- Characterizations of Nash equilibrium shown (in Followership model).
- Empirical evidence to support rate of updates as a parameter.

References

Borgs, C., Chayes, J., Karrer, B., Meeder, B., Ravi, R., Reagans, R., & Sayedi, A. (2010). Game-theoretic models of information overload in social networks. In Algorithms and Models for the Web-Graph (pp. 146-161). Springer Berlin Heidelberg.

