Greedily maximizing social influence

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In this document we present the results of some experimentation with the algorithm presented in the paper "Maximizing the Spread of Influence through a Social Network" by Kempe / Kleinberg / Tardos.

1 Implementation

Let G = (V, E) be the graph we want to analyze. Let $A \subseteq V$ be the output of the algorithm, the subset of nodes we will target. Let k := |A| be the size of the desired initial active node set. Let $\sigma(A)$ be the expected number of nodes that are activated after a simulation with the cascade / linear threshold model, given the initial set A.

The algorithm starts with an empty set, and adds at each iteration a node to the set A: it choses greedily the node that, if added to A, increments the most our objective function $\sigma(A)$ in this step. Because we cannot evaluate the expected value analytically, we simulate the process num_sim times, and approximate the expected value with the average value.

The pseudocode of the algorithm is the following:

Algorithm 1 Greedy Algorithm

```
A := \emptyset
\mathbf{while} \ |A| < k \ \mathbf{do}
\mathbf{for} \ \mathrm{all} \ \mathrm{nodes} \ v \in V \setminus A \ \mathbf{do}
\mathbf{for} \ num\_sim \ \mathrm{iterations} \ \mathbf{do}
\mathrm{Simulate} \ \mathrm{IC} \ / \ \mathrm{LT} \ \mathrm{with} \ \mathrm{initial} \ \mathrm{nodes} \ A \cup v
\mathbf{end} \ \mathbf{for}
\mathrm{Calculate} \ \mathrm{an} \ \mathrm{average} \ \mathrm{activation} \ \mathrm{score} \ \mathrm{for} \ v
\mathbf{end} \ \mathbf{for}
\mathrm{Determine} \ \mathrm{the} \ \mathrm{node} \ v \ \mathrm{with} \ \mathrm{the} \ \mathrm{highest} \ \mathrm{average} \ \mathrm{score}
\mathrm{Set} \ A := A \cup v
\mathbf{end} \ \mathbf{while}
\mathbf{return} \ A
```

Obviously, the algorithm can produce different results whether used with the linear threshold model or with the independent cascade model.

2 Graph Used

We used an undirected graph of an onymized facebook data, with |V|=52 and |E|=292.

3 Results

We first observe how $\sigma(A)$ increases in both models when we allow for a bigger size k of the initial set. We use $num_sim = 100000$ as in the paper.

INSERT GRAPH

We then want to analyze how imprecise the approximated value of $\sigma(A)$ is when num_sim is low. We find that on our graph, even with only 10 simulations, the estimation is quite near the value for 100000 simulations.

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