
Maximizing the Spread of Influence through a Social Network

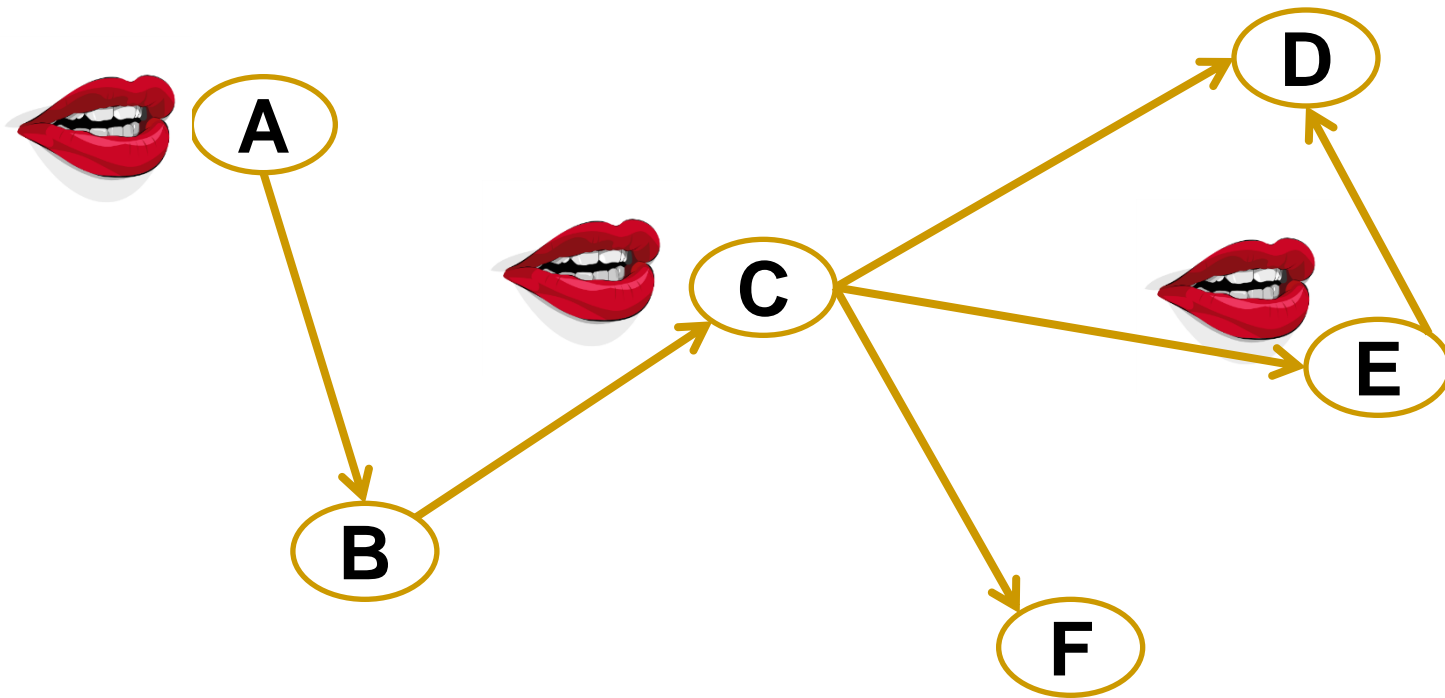
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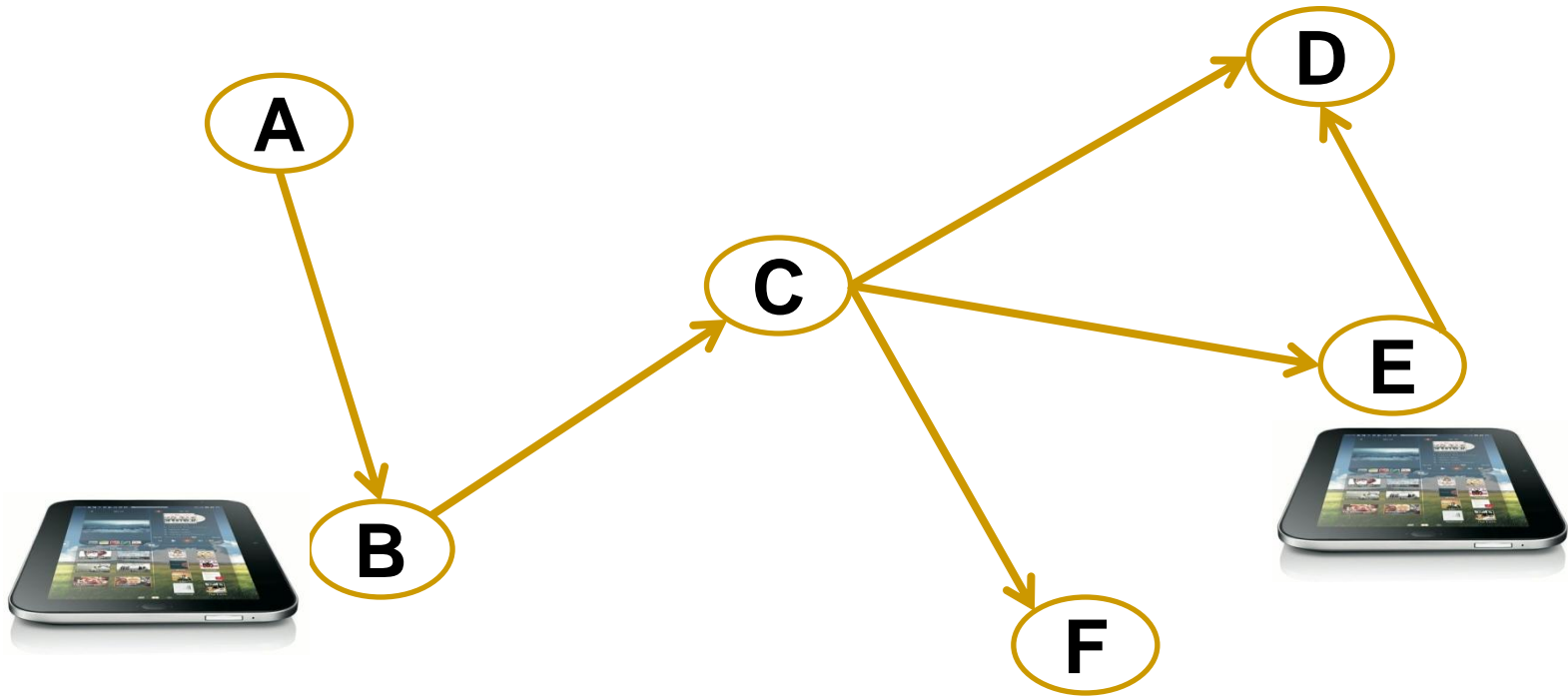
Problem Example 1: Spread of Rumor

- 2012 = end!



Problem Example 2: Viral Marketing

- ezPad 1 beats iPad 3



Problem Definition

- G : a social network (n nodes)
- Model: spread process
- S : initially active subset (k **seeds**)
- $\sigma(S)$: #final active nodes (**achievement**)

■ Task: Choose S^*

■ Goal: ~~$\sigma(S^*) = \max \sigma(S)$~~ NP-Hard

Realistic Goal:

Approximate the maximum with a guarantee

Choose S : $\sigma(S) \geq r \cdot \sigma(S^*)$

Contents in This Talk

- G : a social network (n nodes)
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Prove:

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Model 1:

Independent Cascade Model

Model 1: Cascade Model

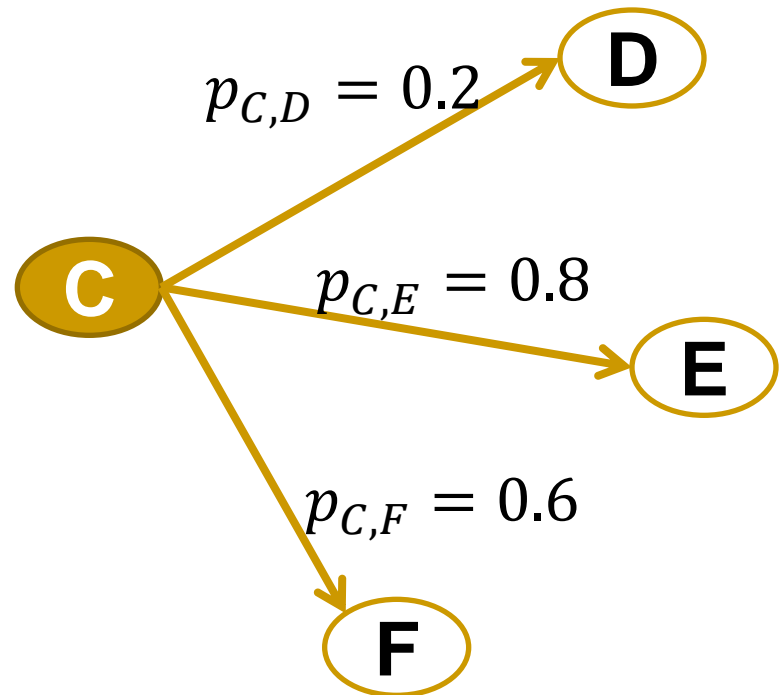
- Each active node try to activate his neighbors



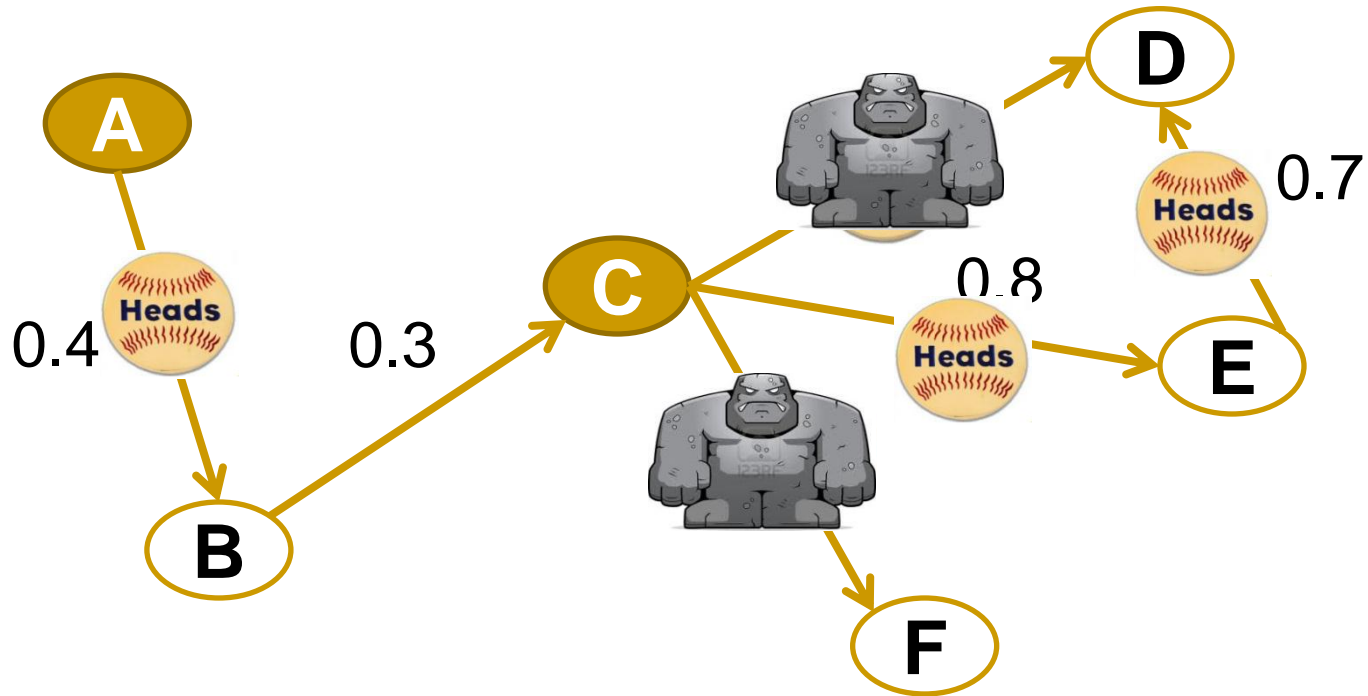
□ $p_{u,v}$

$1 - p_{u,v}$

- Only a single chance

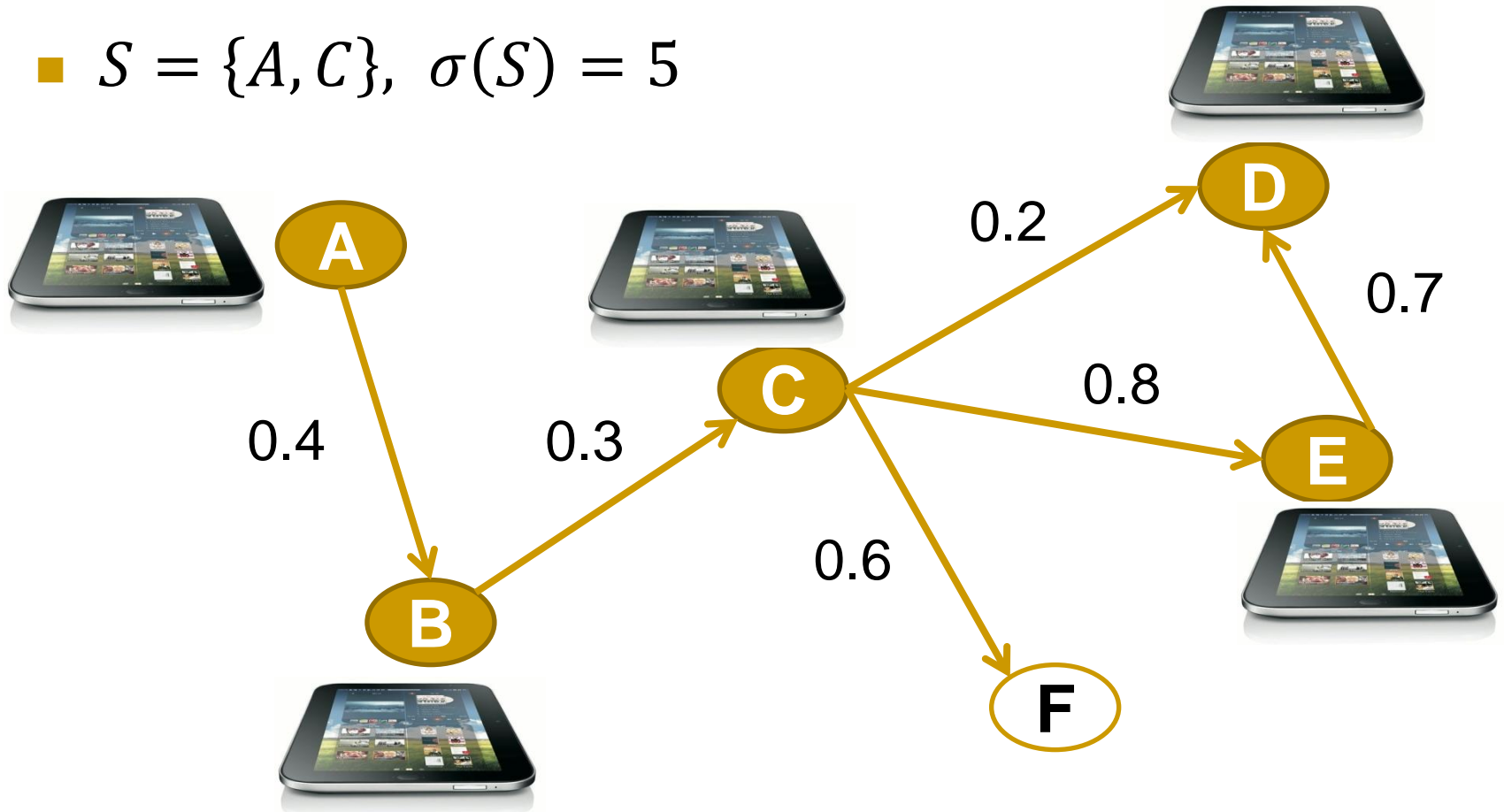


Model 1: Cascade Model



Model 1: Cascade Model

- $S = \{A, C\}$, $\sigma(S) = 5$

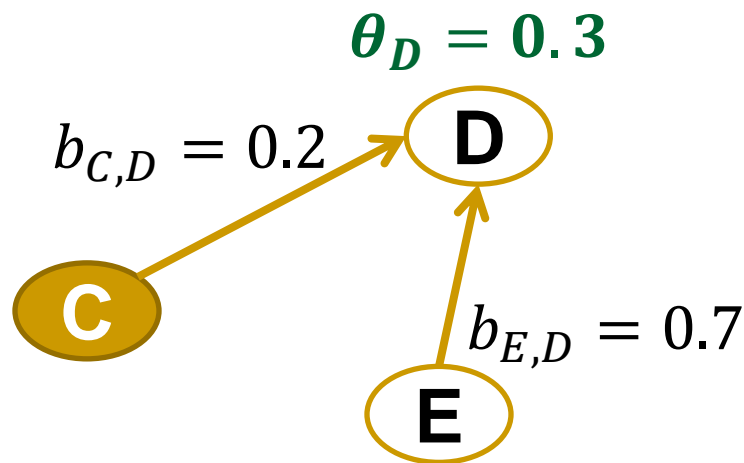


Model 2:

Linear Threshold Model

Model 2: Threshold Model

- Each inactive node picks a random $\theta_v \in [0,1]$
 - Active condition: $\sum_{u: \text{active neighbor of } v} b_{u,v} \geq \theta_v$



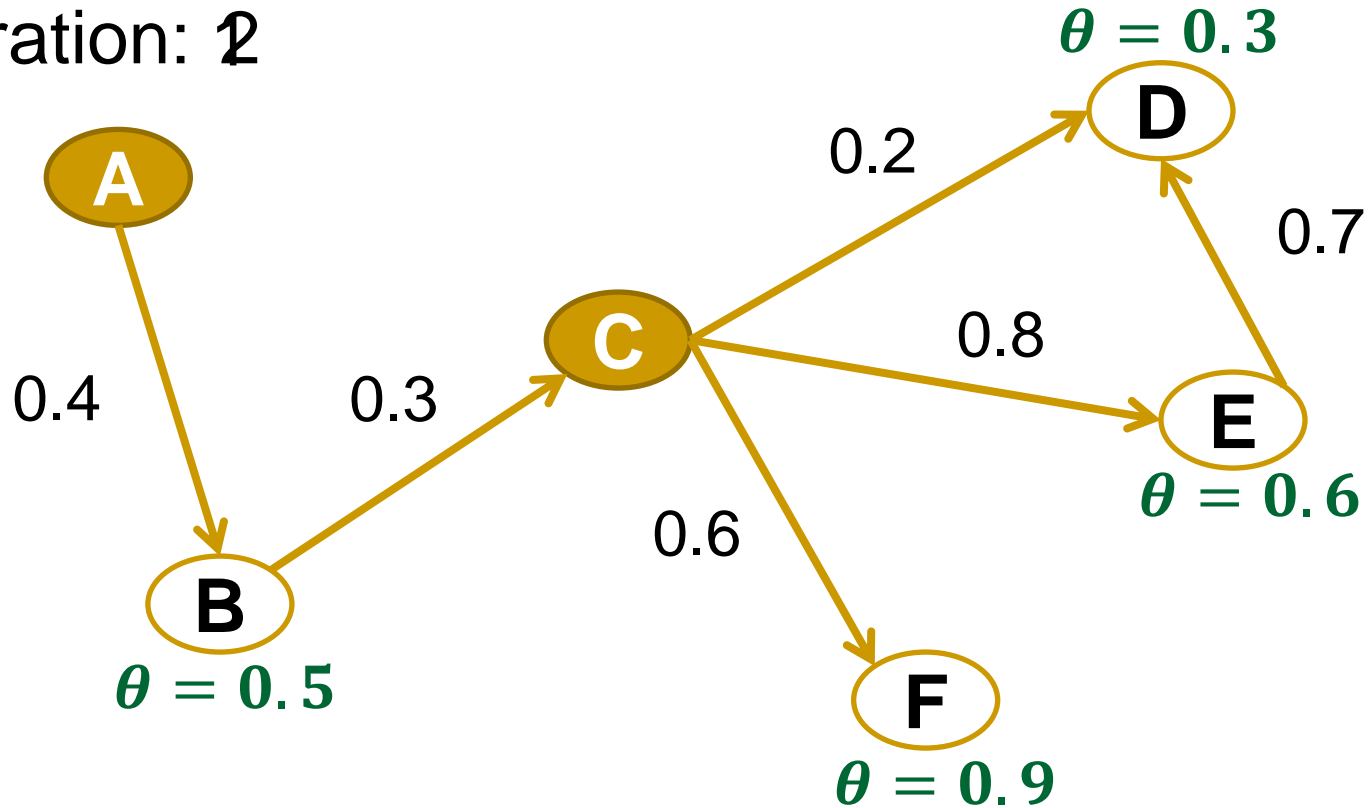
Iteration 2: $0.2 < 0.3$

Iteration 4: E \rightarrow active

Iteration 5: $0.2 + 0.7 > 0.3$
D \rightarrow active

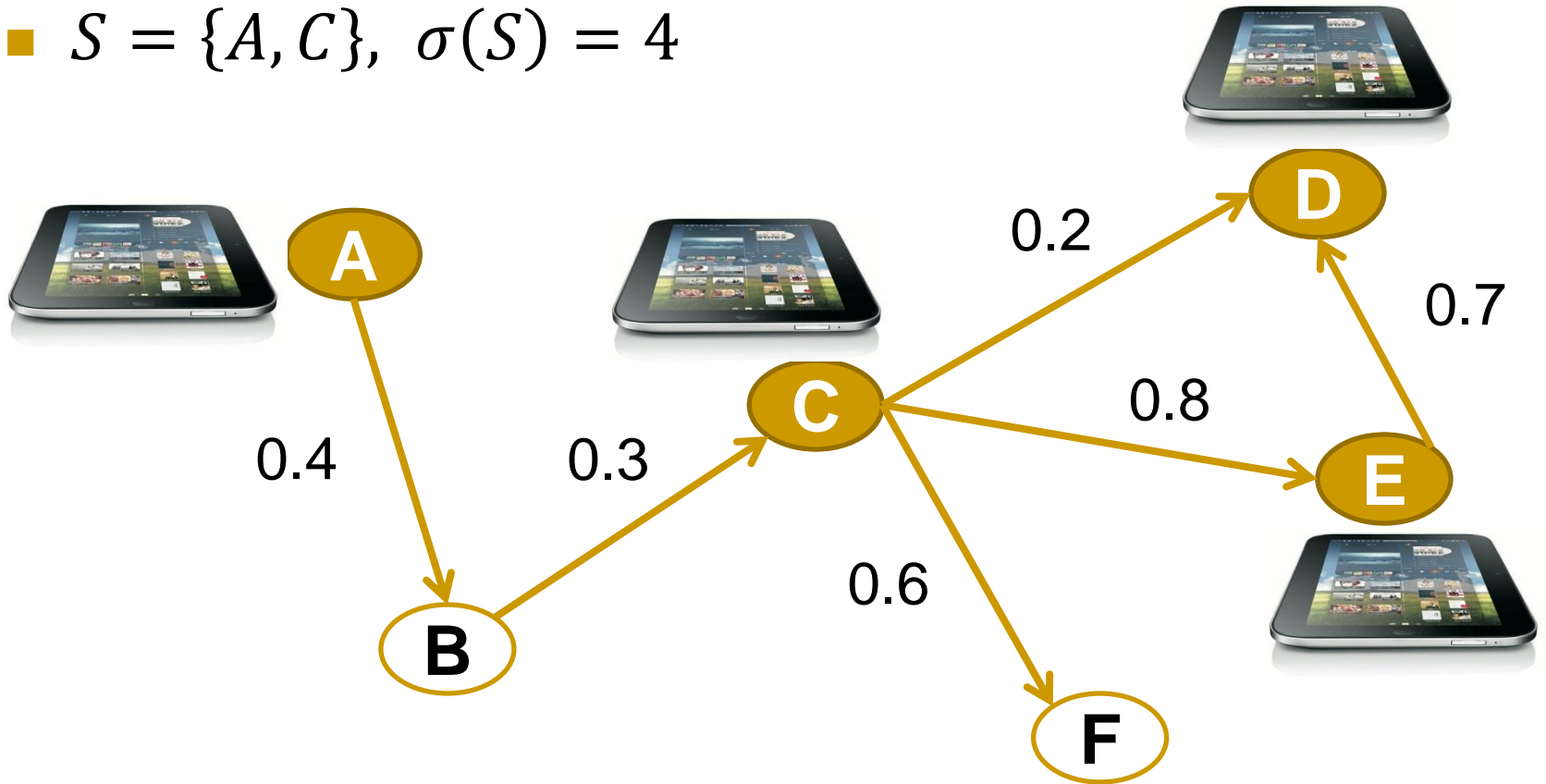
Model 2: Threshold Model

Iteration: 2



Model 2: Threshold Model

- $S = \{A, C\}$, $\sigma(S) = 4$



How to Prove the Guarantee?

Given a
spread model

???

find S , s.t.
 $\sigma(S) \geq r \cdot \sigma(S^*)$

find S , s.t.
 $f(S) \geq \left(1 - \frac{1}{e}\right) \cdot f(S^*)$

Nemhauser

$f(S)$:
Non-negative
monotone
submodular

Submodularity

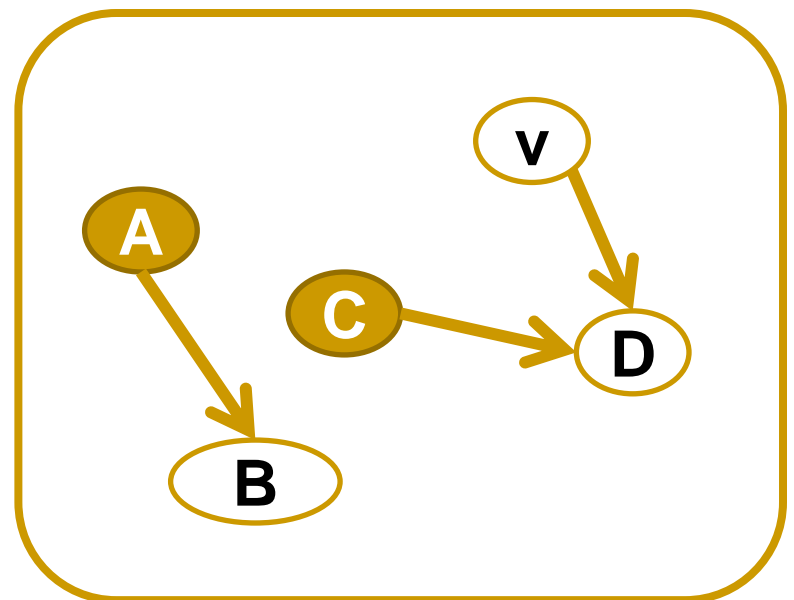
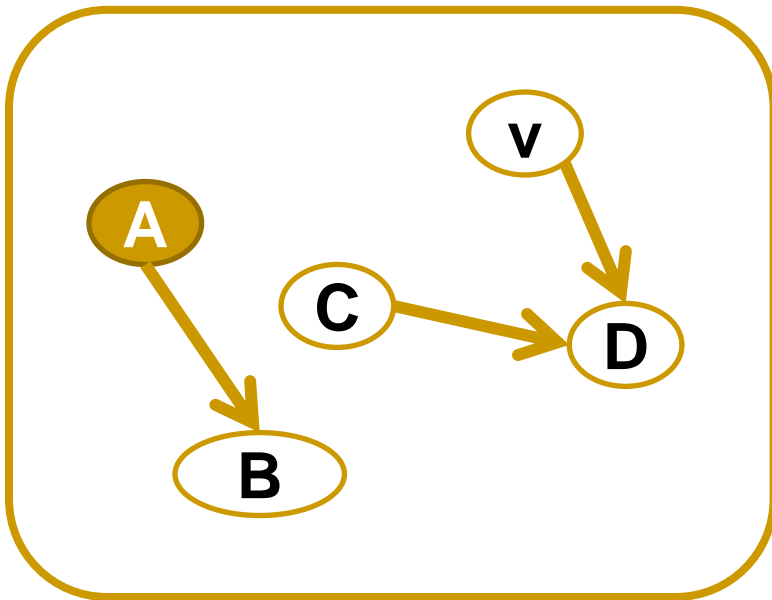
- U : a finite ground set
- $P(U)$: power set of U
- $f(\cdot): P(U) \rightarrow R^*$

- Submodularity: $\forall \text{ node } v, \forall S \subseteq T$

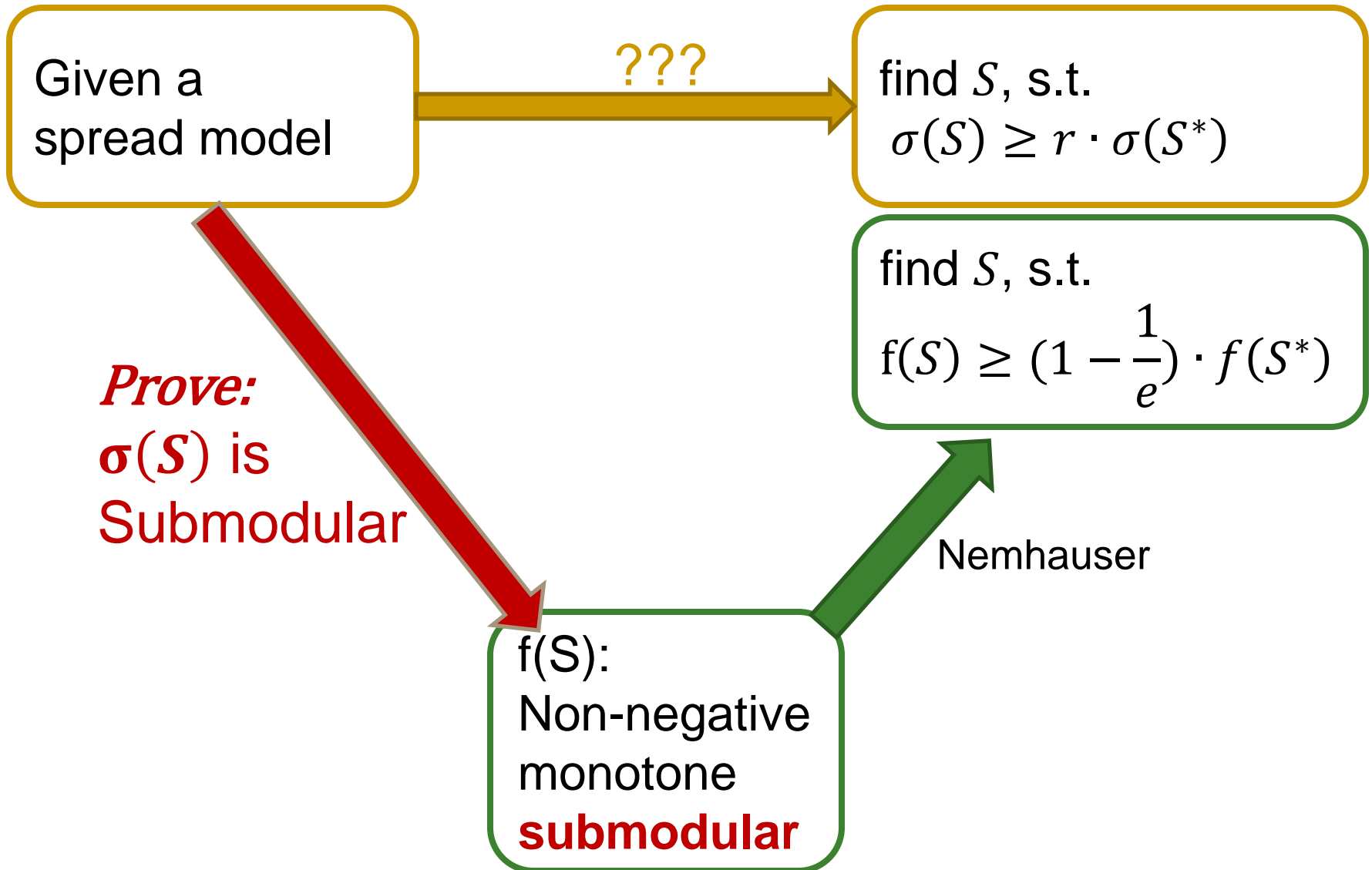
$$f(S \cup \{v\}) - f(S) \geq f(T \cup \{v\}) - f(T)$$

Example: Submodularity





- $f(S)$: number of vertexes reachable from vertexes in S



How to Prove the Guarantee?



We Want to Prove...

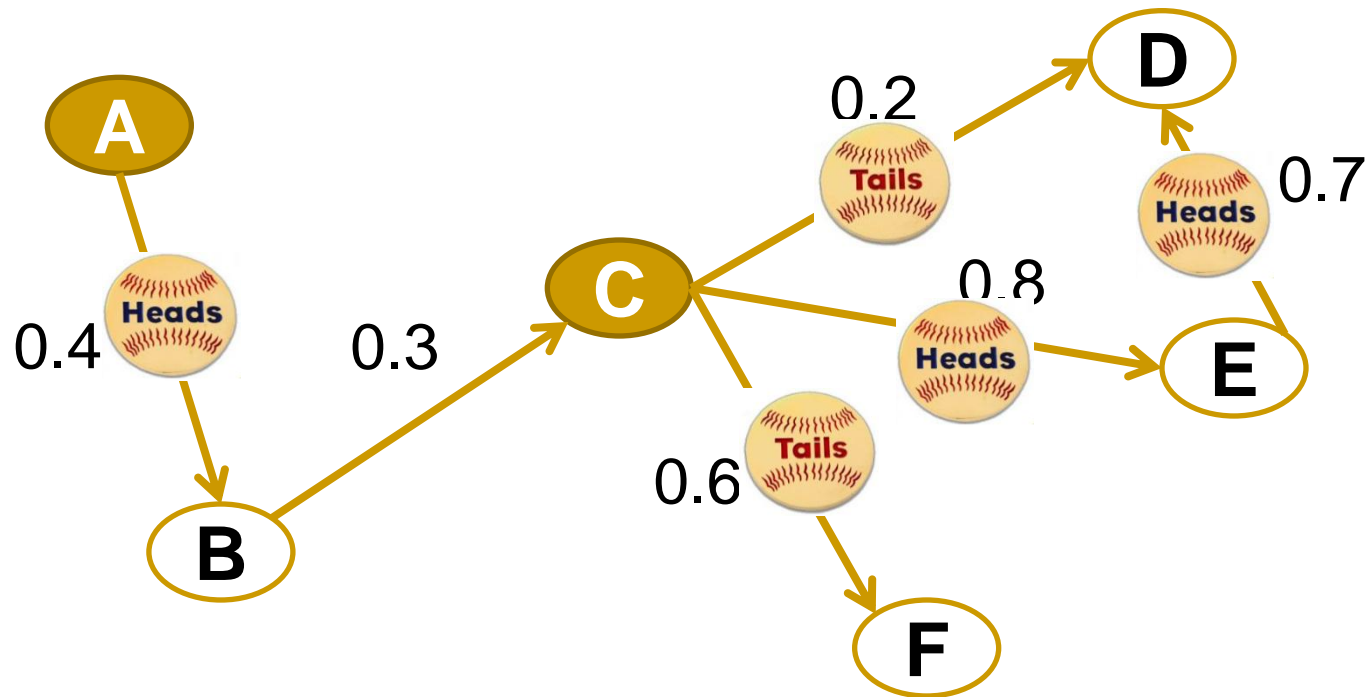
Model	$\sigma(S)$ is Submodular	NP-hard
Independent Cascade		
Linear Threshold		

Prove:
Submodularity

Cascade Model

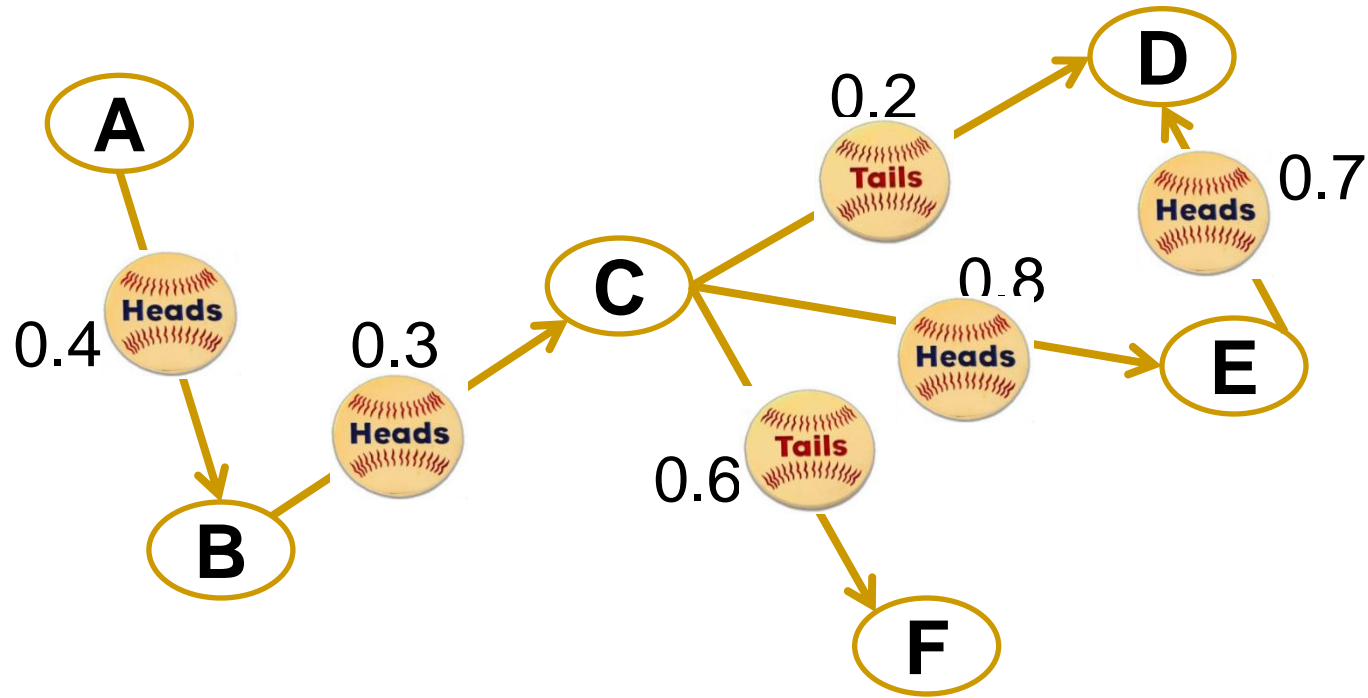
Submodularity (Cascade Model)

- Recall: flip coin



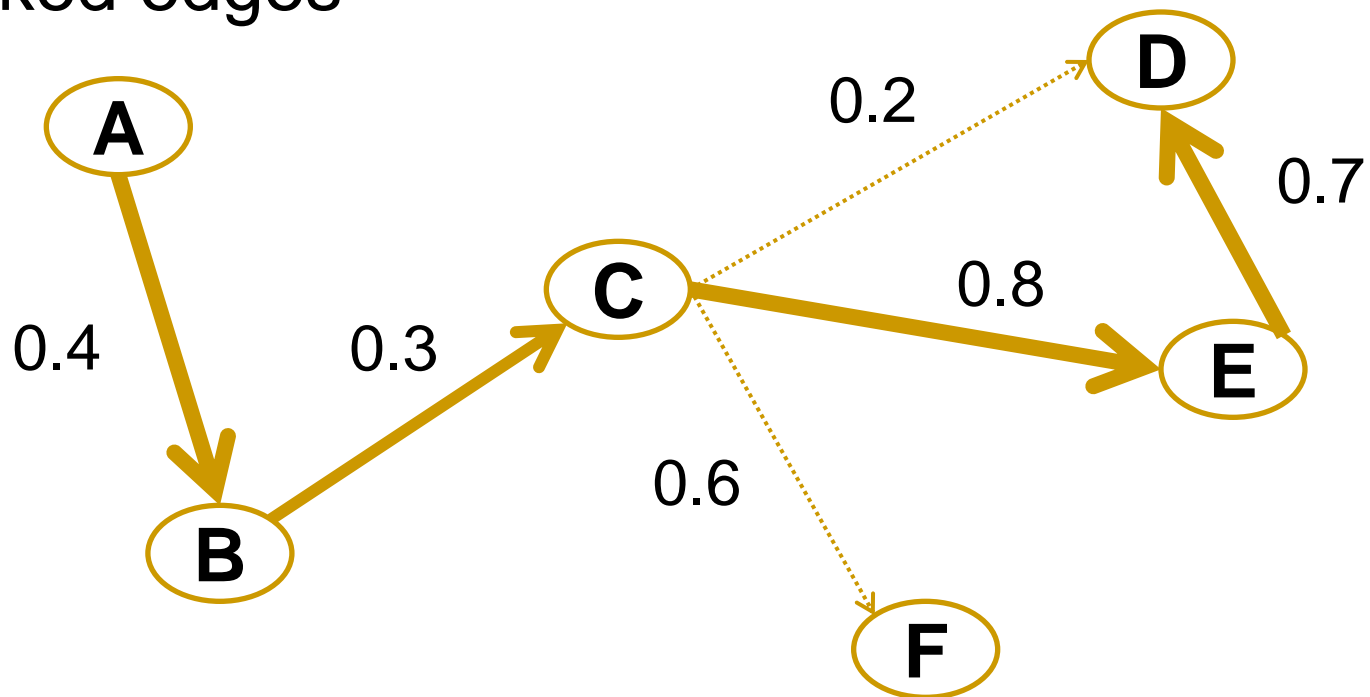
Submodularity (Cascade Model)

- Why not flip all the coins in the beginning?



Submodularity (Cascade Model)

- Live edges → live paths
- blocked edges



Simplify Cascade Model

Node v ends up active



A live path: some seed $\rightarrow v$

Achievement (Simplified Model)

- X : coin flipping outcome

- e.g. X_1, X_2

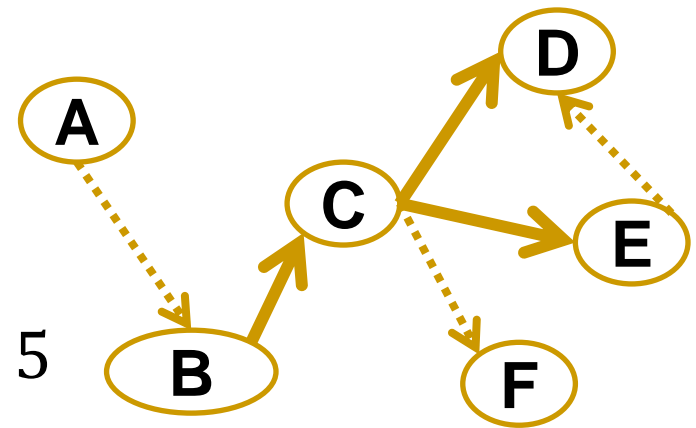
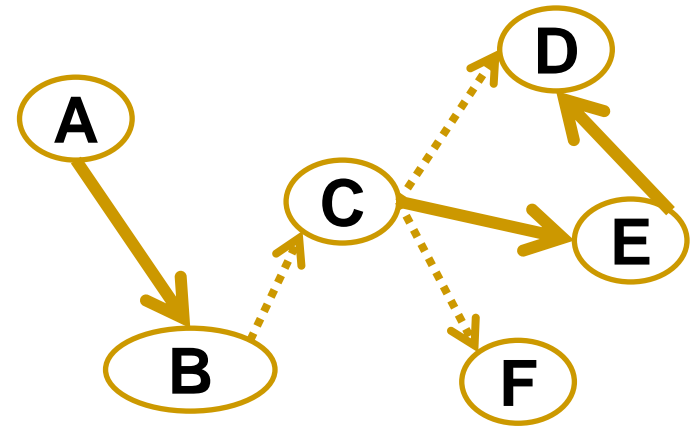
- $R_X(v)$

- $R_{X_1}(A) = \{A, B\}$


- $R_{X_1}(C) = \{C, D, E\}$

- $\sigma_X(S) = |\cup_{v \in S} R_X(v)|$

- $\sigma_{X_1}(\{A, C\}) = |\{A, B, C, D, E\}| = 5$



Submodularity (Cascade Model)

- Fix x , $\sigma_x(S)$ is submodular 
- Linear combination of submodular functions is still submodular

$$\sigma(S) = \sum_x \text{Prob}[X] \cdot \sigma_x(S)$$

Summary of the proof

Active = Has a live path

$\sigma_X(S)$ is submodular

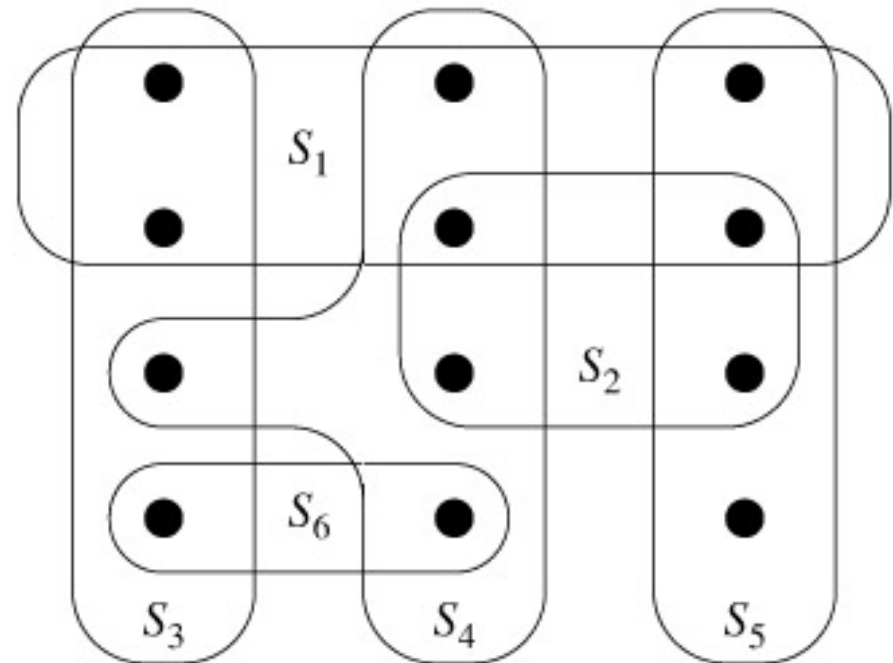
$\sigma(S)$ is submodular

Prove:
NP-hard

Simplified Cascade Model

NP-Hard (Cascade Model)

- Set Cover Problem: k subsets cover all?
- $K=1$: No
- $K=2$: No
- $K=3$: Yes
- $K=4$: ...



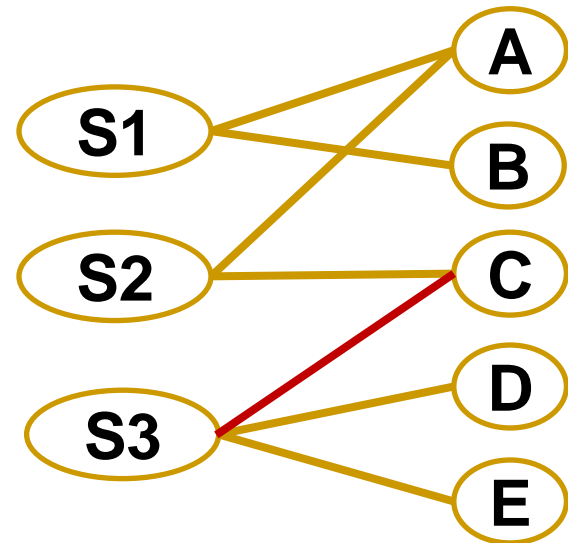
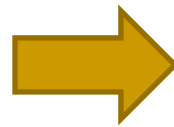
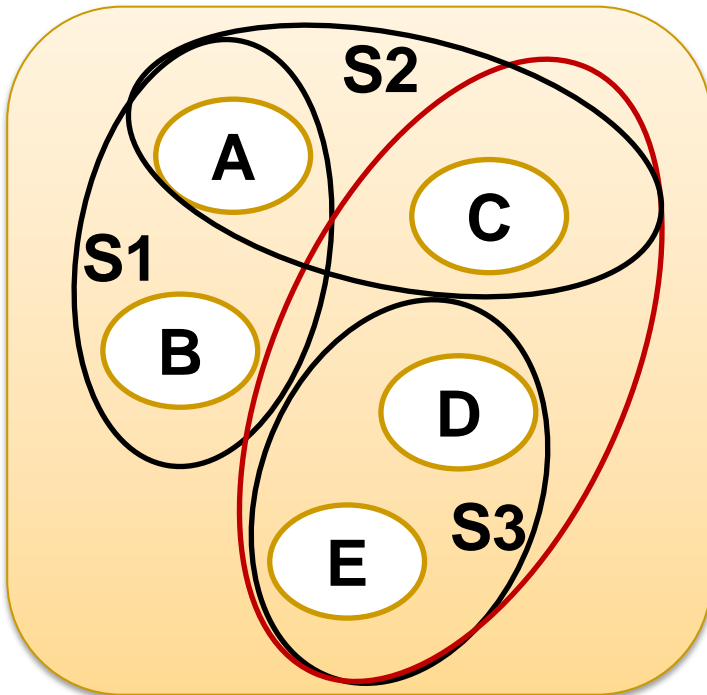
NP-Hard (Cascade Model)

■ Solve Set Cover

Q: 2 subsets cover all ?

■ Influence maximization

Q: $|S| = 2, \sigma(S) \geq 2 + 5?$



NP-Hard (Cascade Model)

Influence Maximization Problem

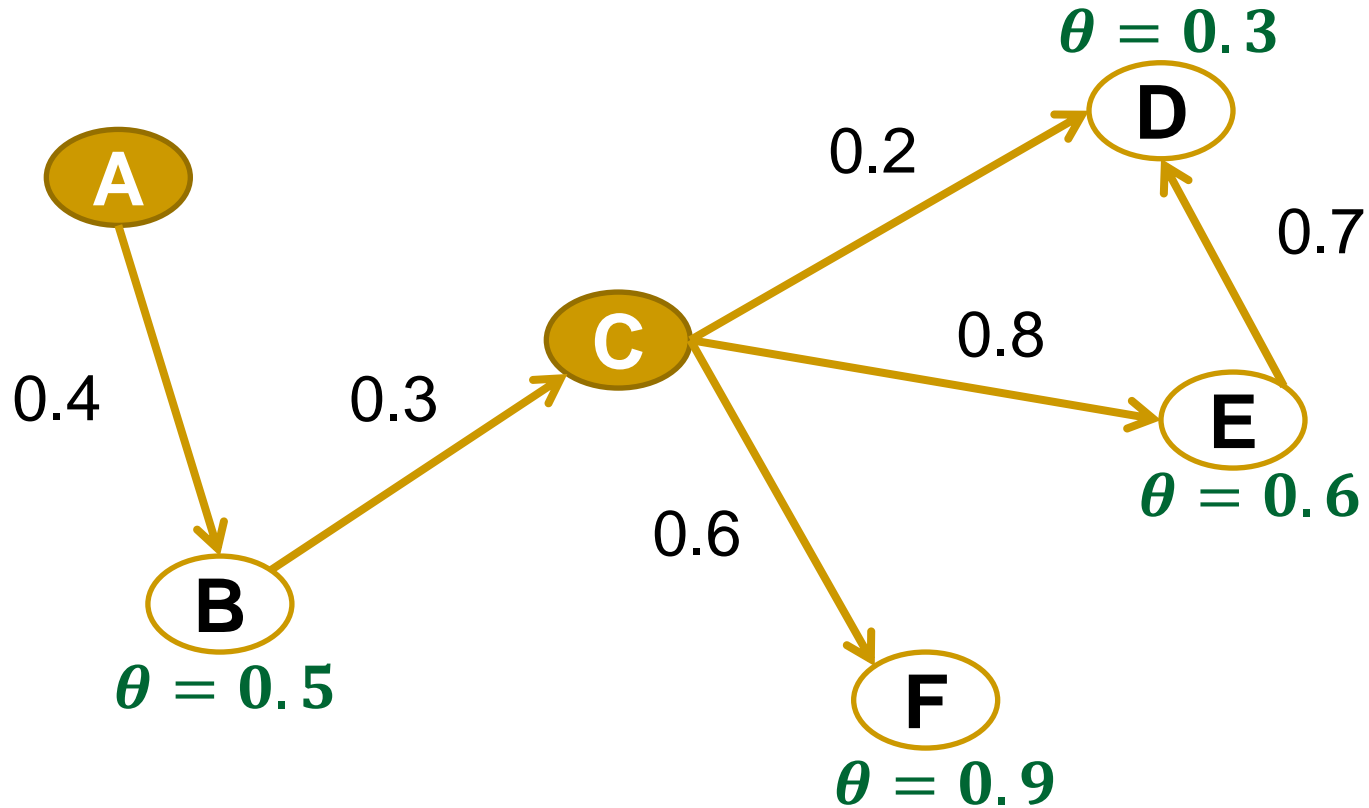
is at least as difficult as

Set Cover Problem

Prove:
Submodularity

Linear Threshold Model

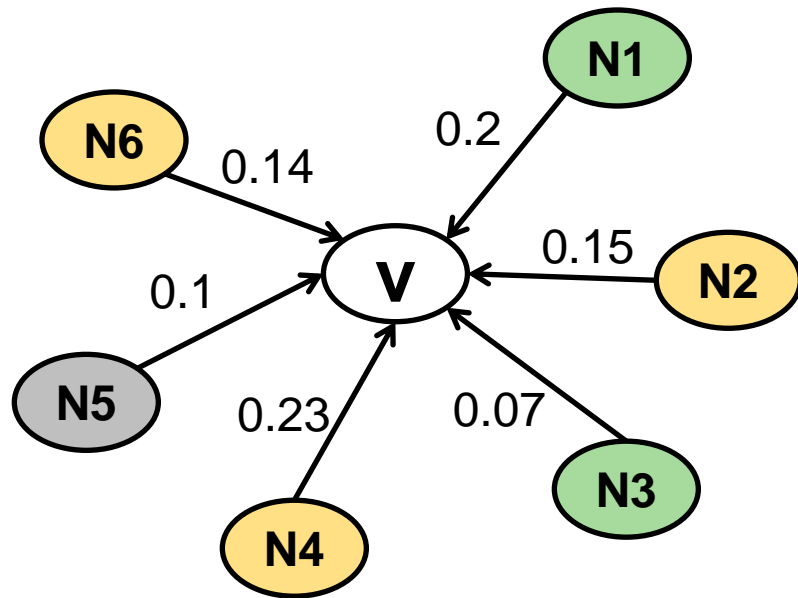
Recall: Threshold Model



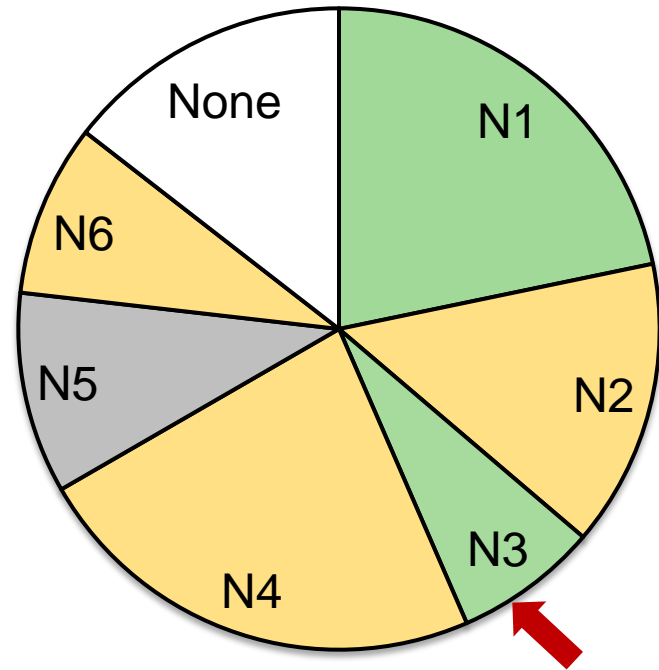
Gamble: Roulette



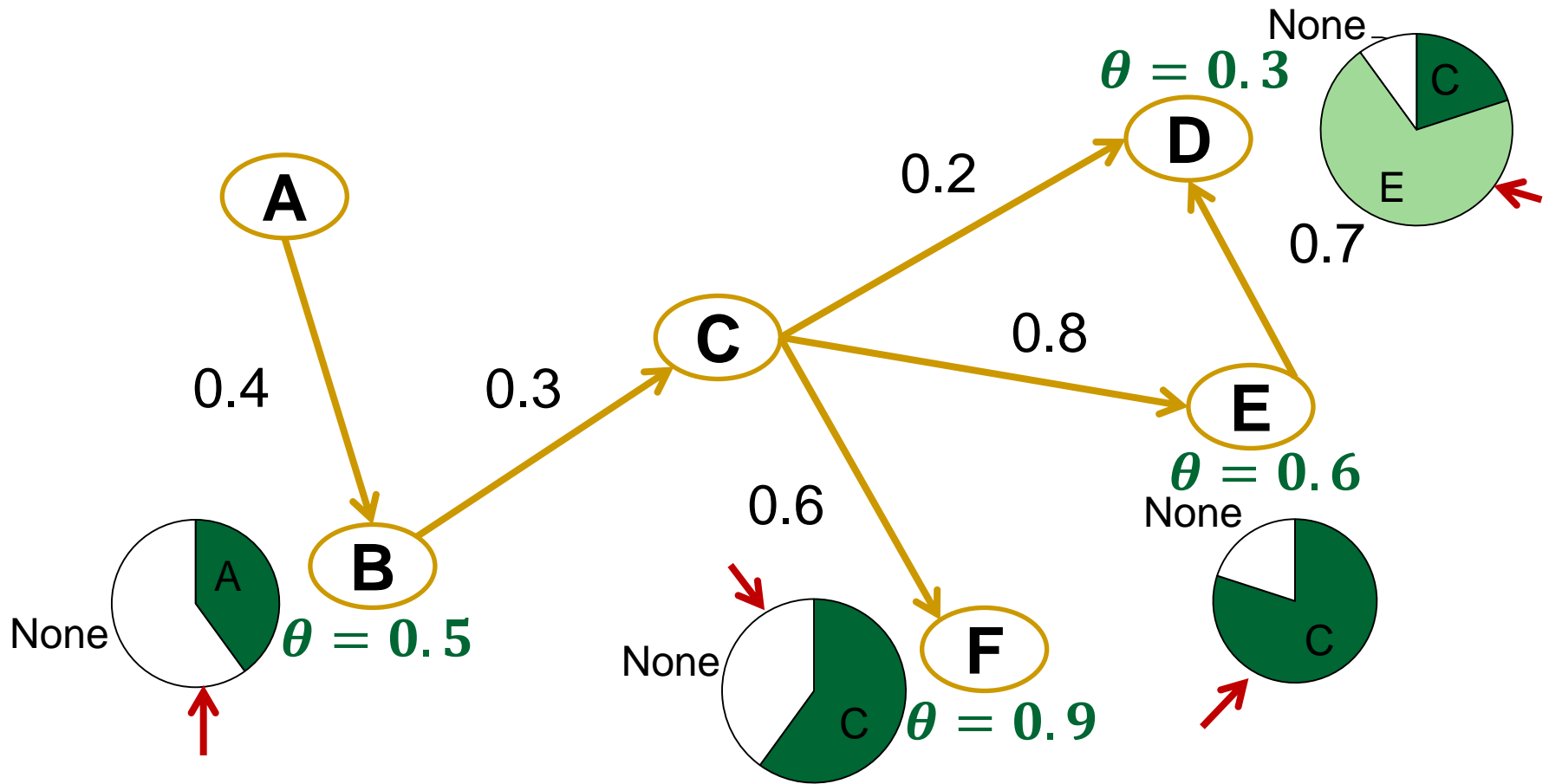
Gamble: Roulette



$$\theta = 0.4$$

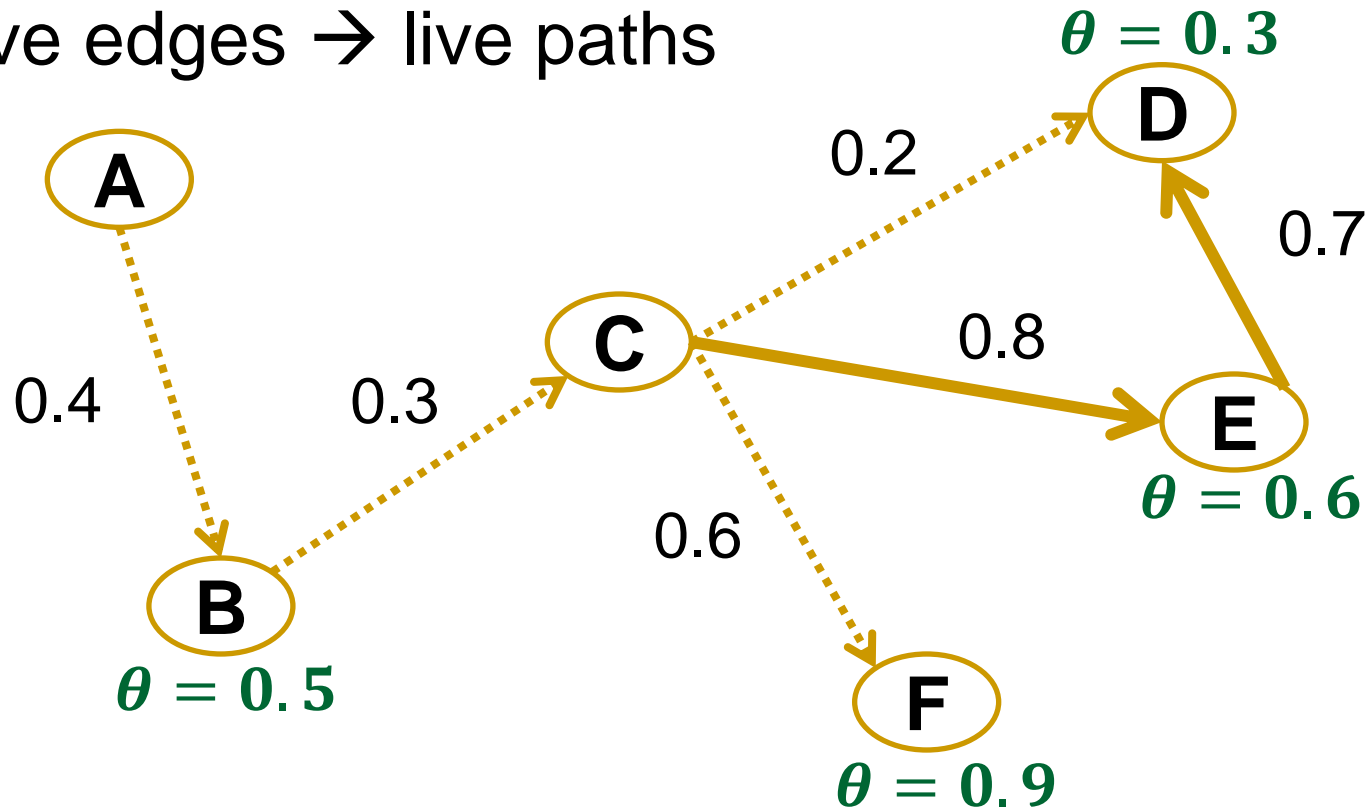


Submodularity (Threshold Model)



Submodularity (Threshold Model)

- Live edges \rightarrow live paths



Correctness of Simplification

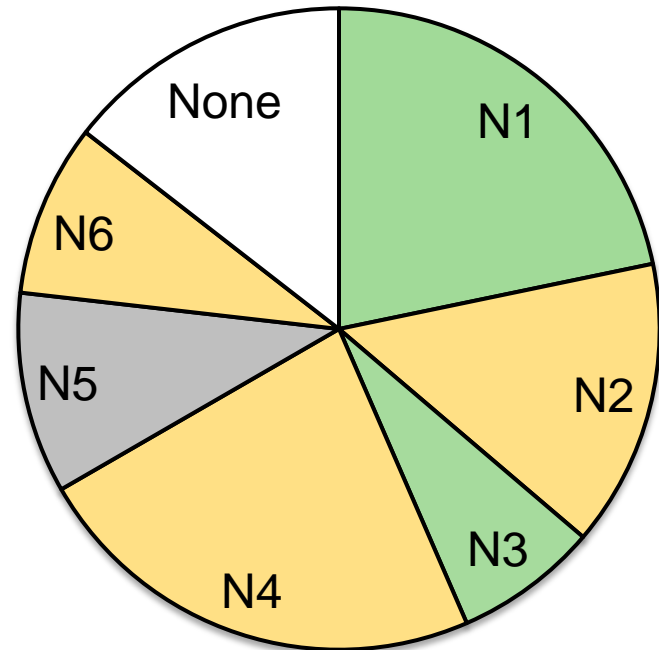
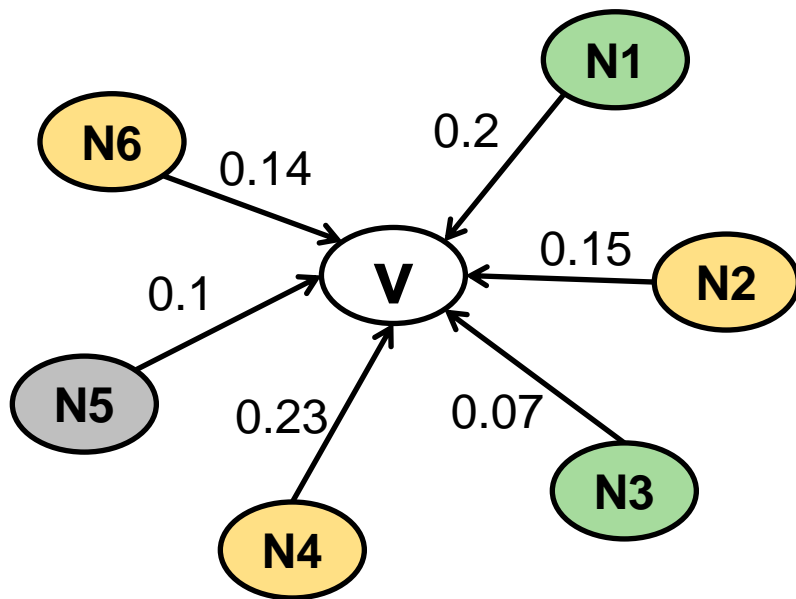
For node v :

$P(\textit{active in Iteration } t + 1 \mid \textit{inactive in Iterations } \leq t)$

$$= \frac{P(\textit{active in Iteration } t + 1)}{P(\textit{inactive in Iterations } \leq t)}$$

Simplified Model

- Active before iteration 5
- becomes active in iteration 5

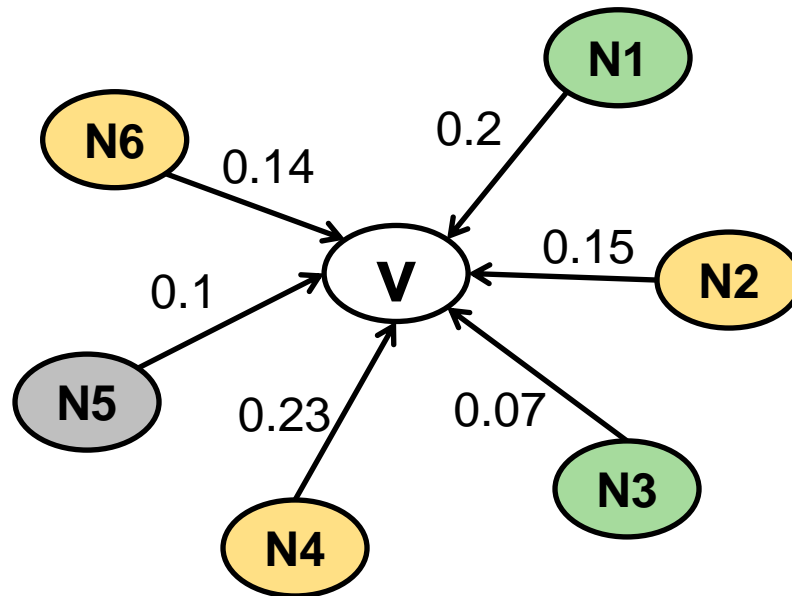


Simplified Model

A_t : Nodes becoming active in iteration t

$$\frac{\sum_{u \in A_t} b_{u,v}}{1 - \sum_{u \in A_1 \cup A_2 \cup \dots \cup A_{t-1}} b_{u,v}}$$

Original Model



Original Model

A_t : Nodes becoming active in iteration t

$$\frac{\sum_{u \in A_t} b_{u,v}}{1 - \sum_{u \in A_1 \cup A_2 \cup \dots \cup A_{t-1}} b_{u,v}}$$

Simplify Threshold Model

Node v ends up active



A live path: some seed $\rightarrow v$

Similarly, we have...

Active = Has a live path

$\sigma_X(S)$ is submodular

$\sigma(S)$ is submodular

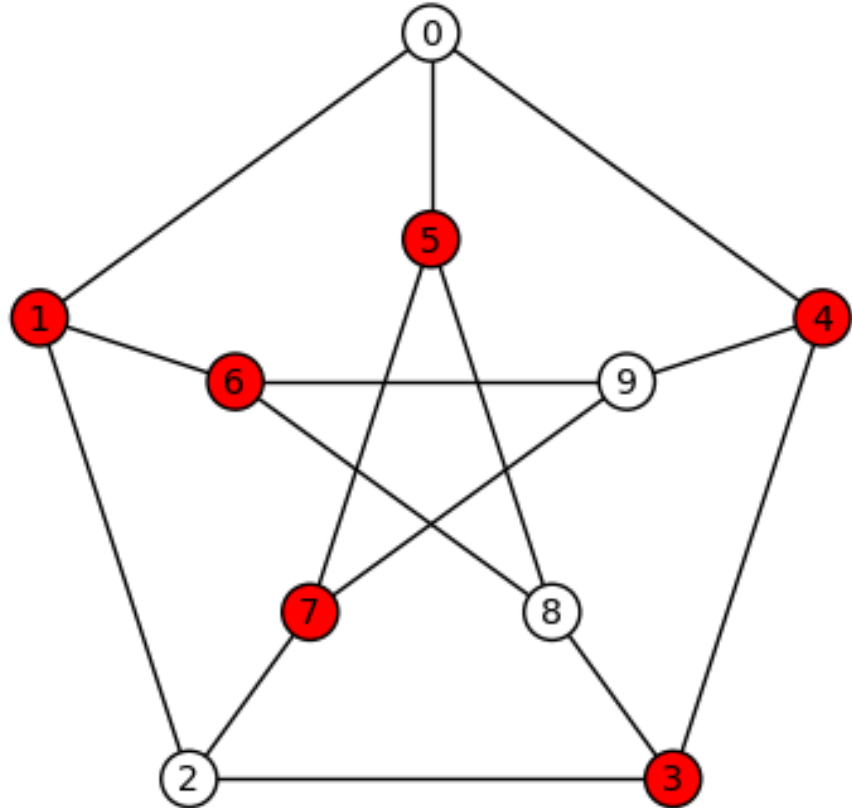
Prove:
NP-hard

Linear Threshold Model

NP-Hard (Threshold Model)

- Vertex Cover Problem
 - k vertexes (S)

each edge
is incident to
at least one vertex in S



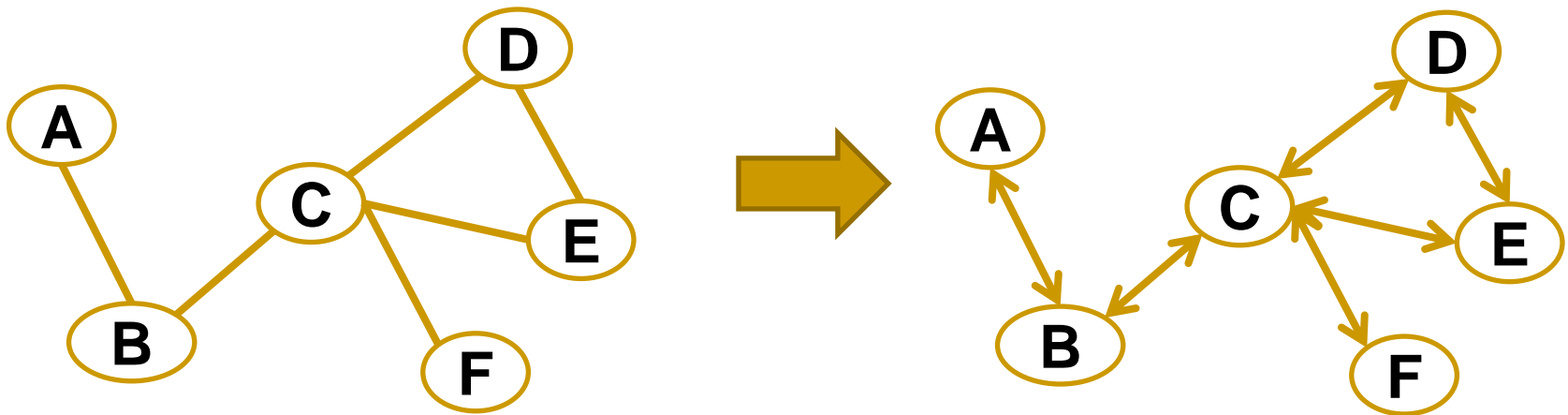
NP-Hard (Threshold Model)

■ Vertex Set Cover

Q: 3 vertexes cover all ?

■ Influence maximization

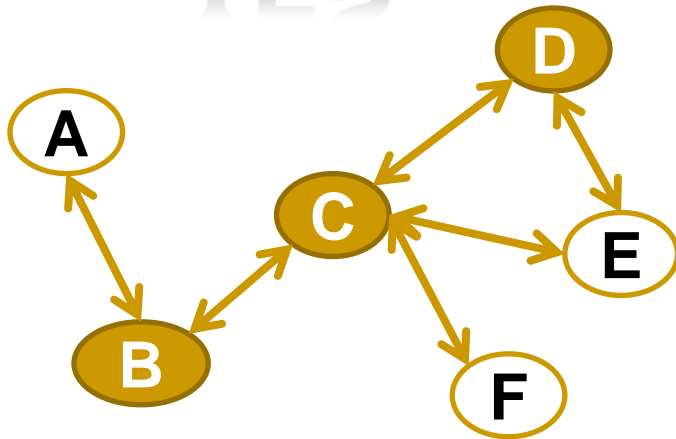
Q: $|S| = 3, \sigma(S) = 6?$



Influence Maximization

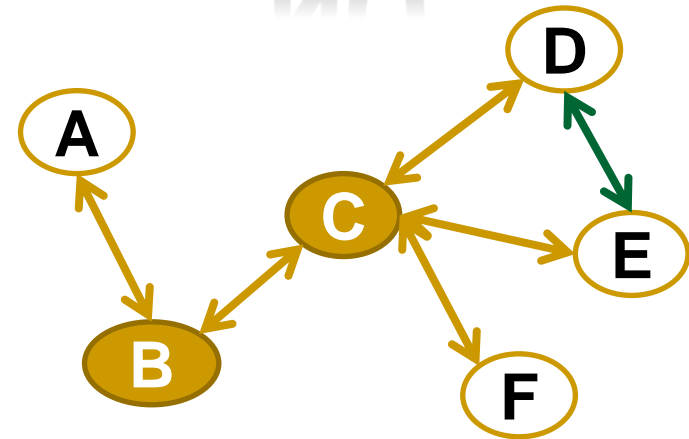
Q: $|S| = 3, \sigma(S) = 6$?

YES



Q: $|S| = 2, \sigma(S) = 6$?

NO



NP-Hard (Threshold Model)





Influence Maximization Problem

is at least as difficult as

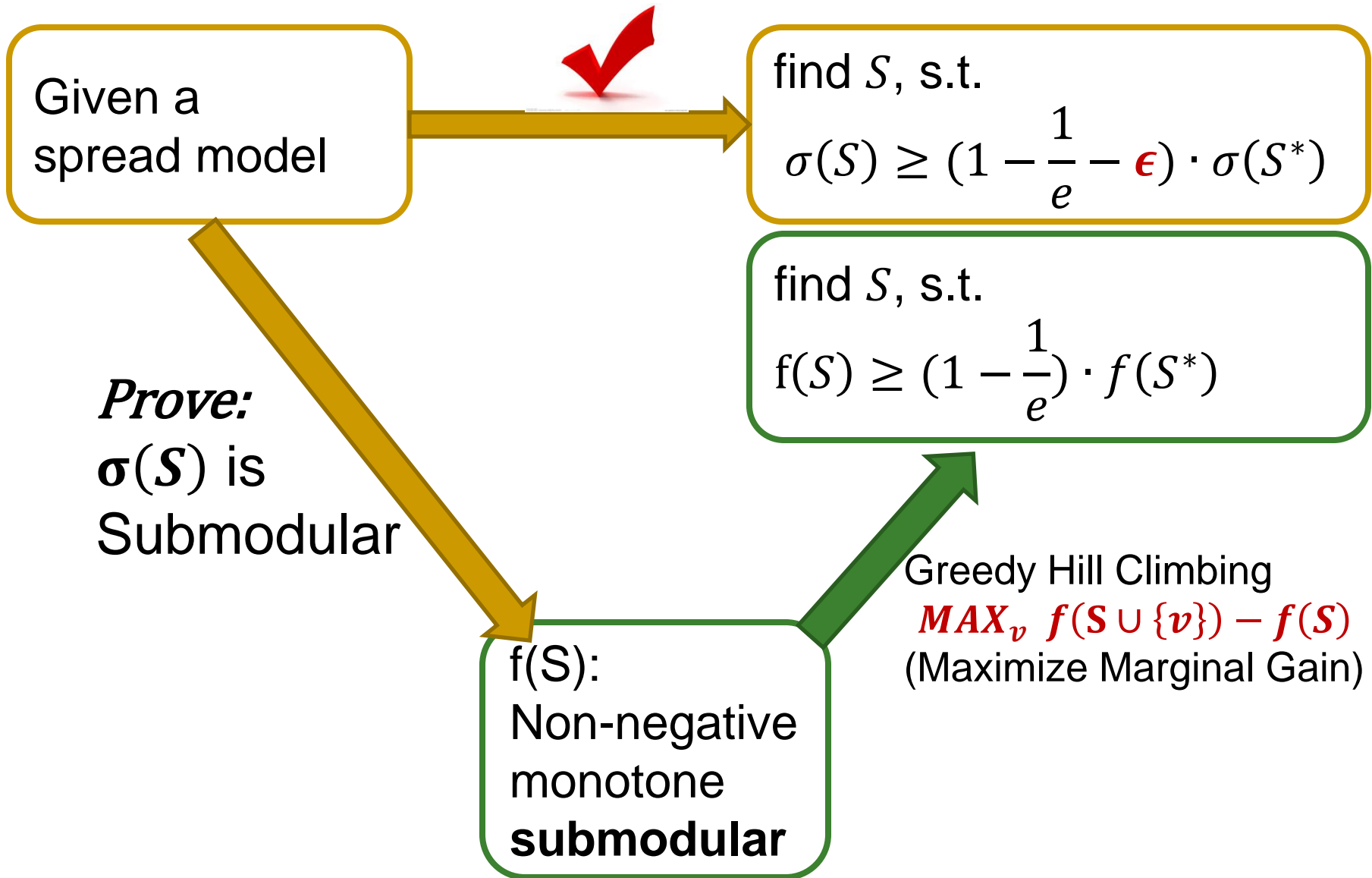
Vertex Cover Problem

End of Proofs

□ Influence Maximization Problem

Model	$\sigma(S)$ is Submodular	NP-hard
Independent Cascade		
Linear Threshold		

Initial Problem



Summary

- Problem Description
- Two Models
 - Independent Cascade Model
 - Linear Threshold Model
- Submodular Functions
- Proof of Approximation Guarantee
- Proof of NP-Hardness



Q&A