

**ΟΙΚΟΝΟΜΙΚΟ
ΠΑΝΕΠΙΣΤΗΜΙΟ
ΑΘΗΝΩΝ**



ATHENS UNIVERSITY
OF ECONOMICS
AND BUSINESS

Social Networks Characteristics & Link Prediction

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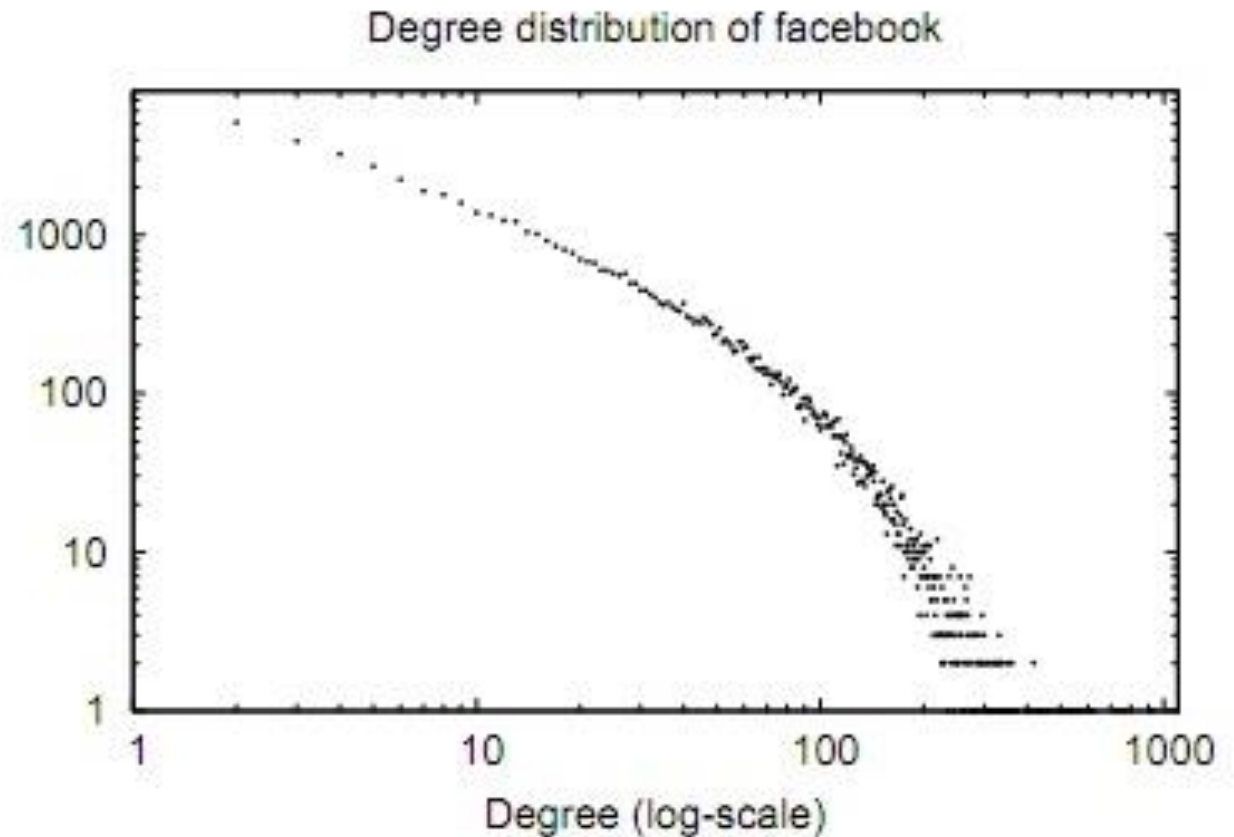
The small world effect

- The average distance in the topology is very small compared to the size of the network.
 - Facebook avg distance is ~ 5



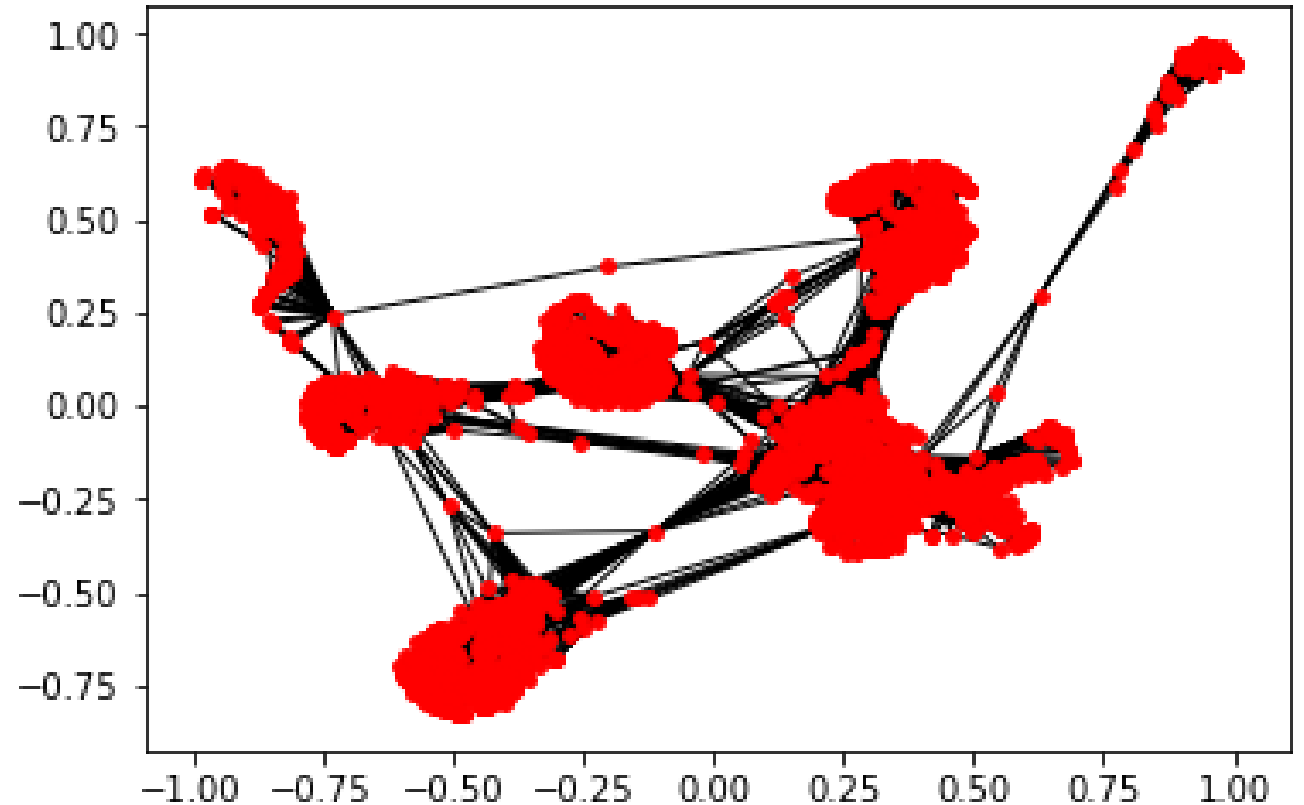
Power law degree distribution

- Only a few nodes have lots of links. Most nodes' links are very small in the network.
- See: M. Faloutsos, P. Faloutsos, C. Faloutsos: *On power-law relationships of the Internet topology*.



Clustering effect

- There are many small groups in a social network where each member of the group knows each other.
 - Results in many fully connected subgraphs.

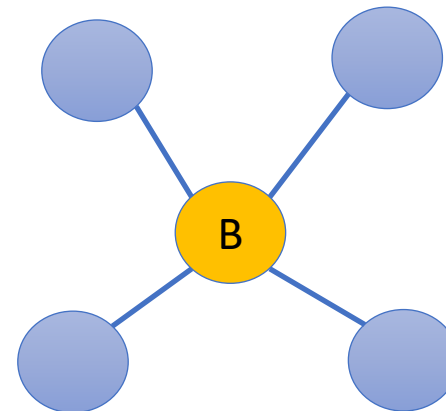
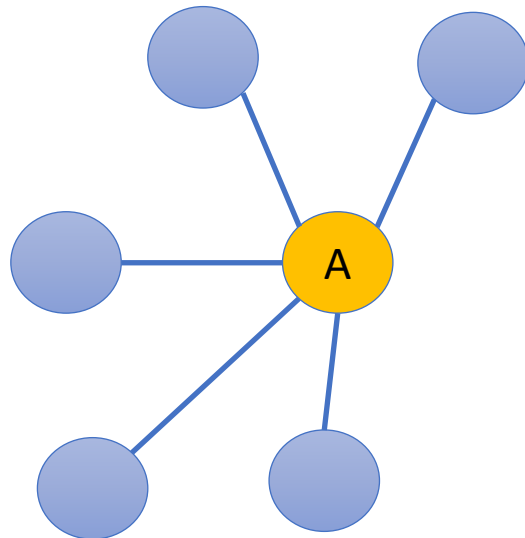


Link prediction

- Predict the **likelihood** of a future association between two nodes
- This is typically achieved by utilizing some notion of **similarity** between the two nodes

Preferential Attachment

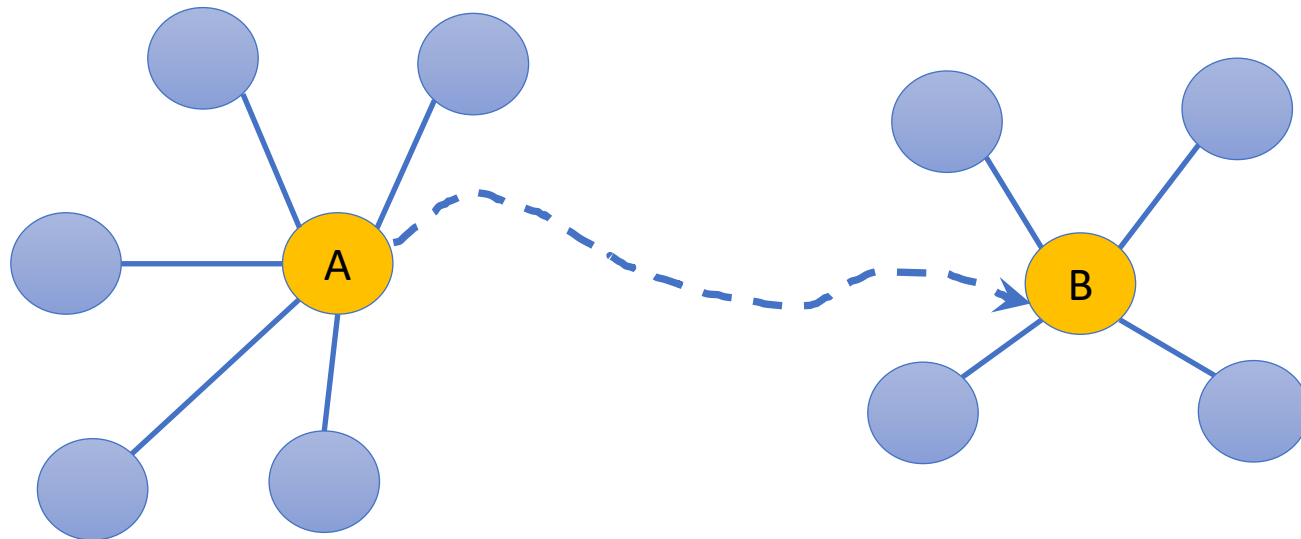
- **Intuition:** users with many friends tend to create more connections in the future
 - Lets $N(u)$ denote the set of friends of node u
 - Then, use $|N(u)| * |N(v)|$ as a measure for scoring the likelihood of a link between node u and node v



Caveat?

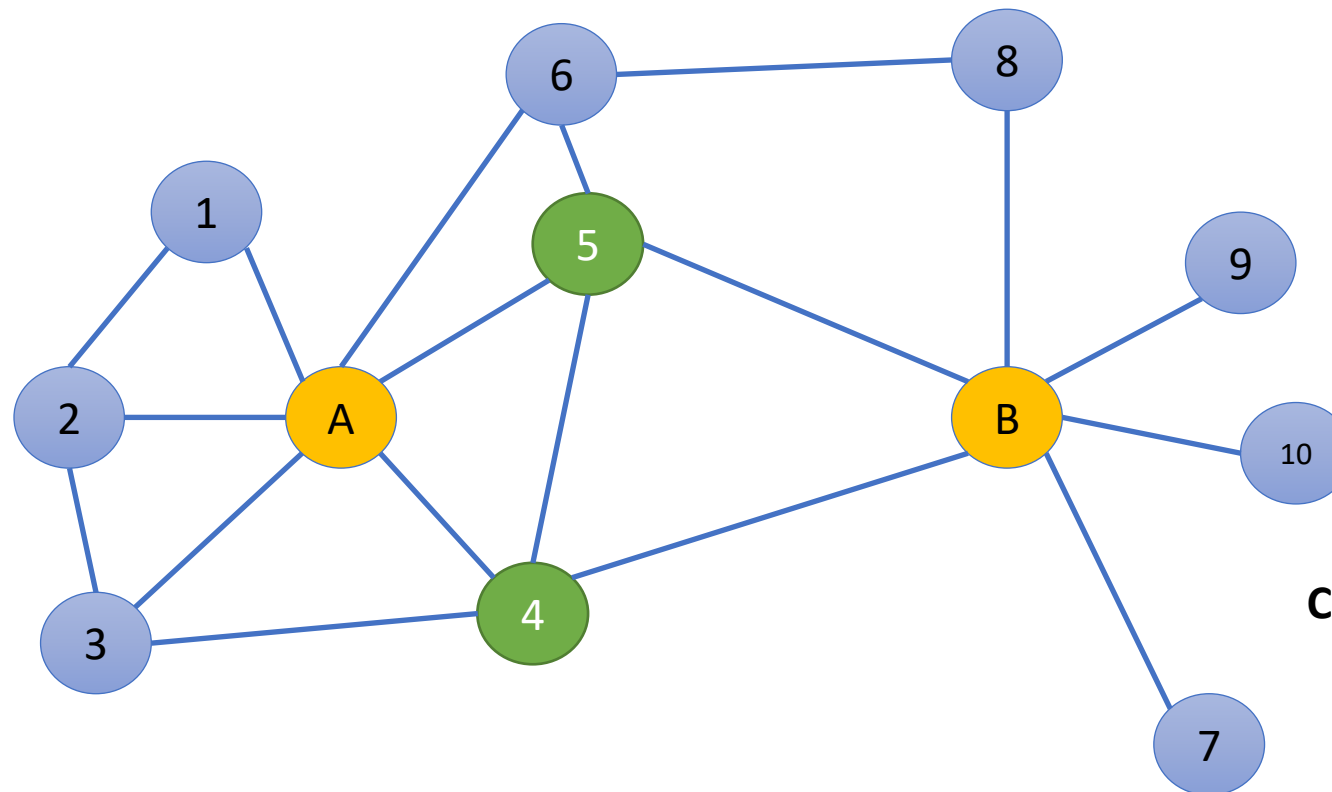
Graph Distance

- Define similarity as the negative length of the shortest path between nodes A and B
 - Recall that due to the small-world effect this distance is typically small
 - Thus, it may be hard to differentiate pairs using shortest-paths



Count Common Neighbours

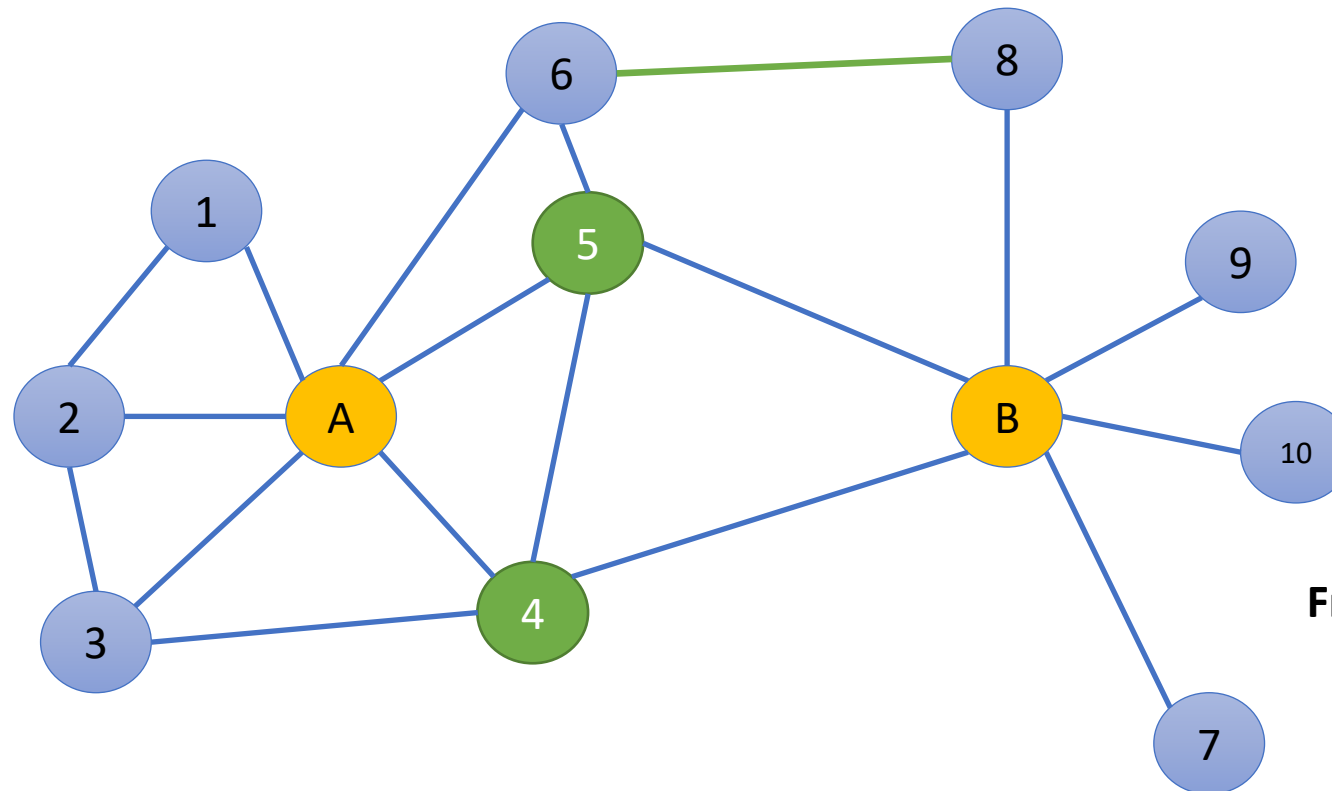
- **Intuition:** two strangers who have a common acquaintance may be introduced by that friend.



Common-Neighbours(A,B)=2

Friends-measure

- Count #common-friends + #connections_between_neighbors

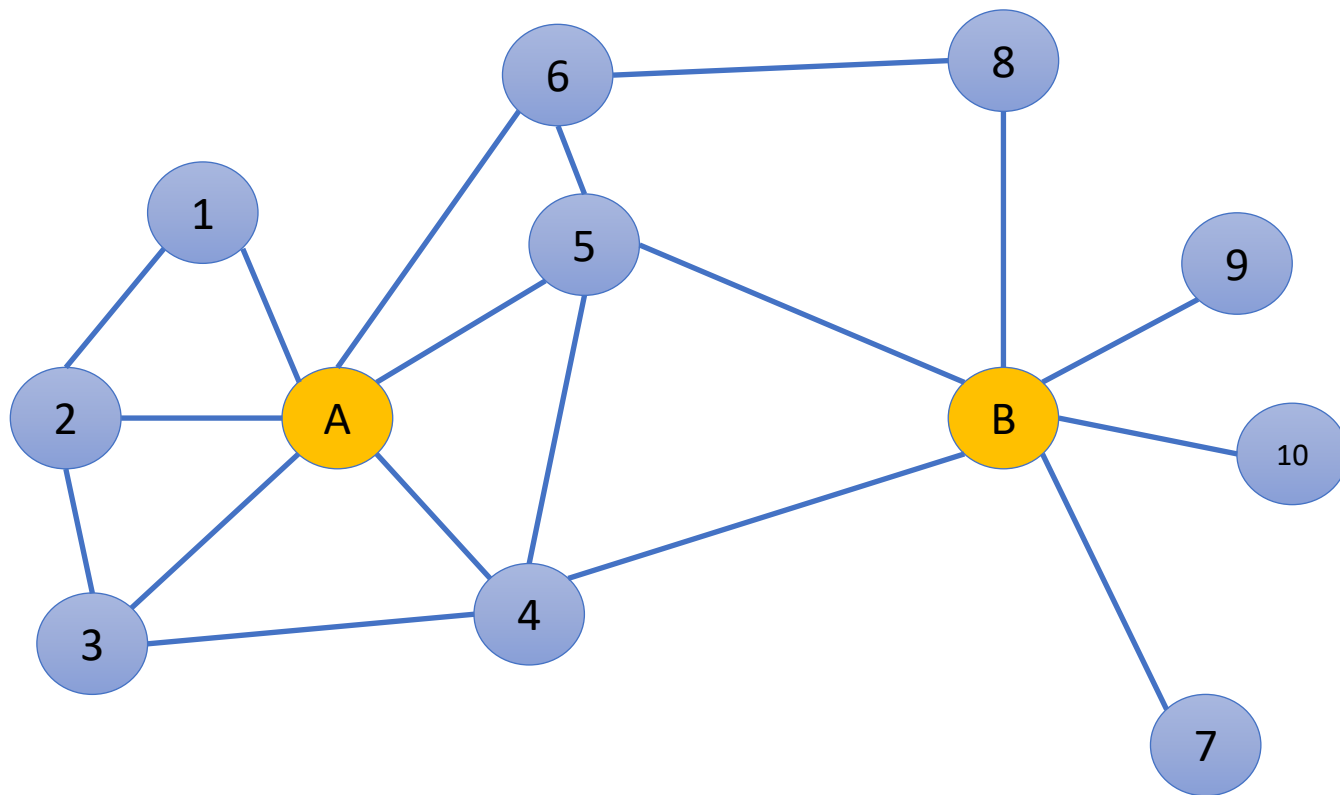


Friends-measure(A,B)=3

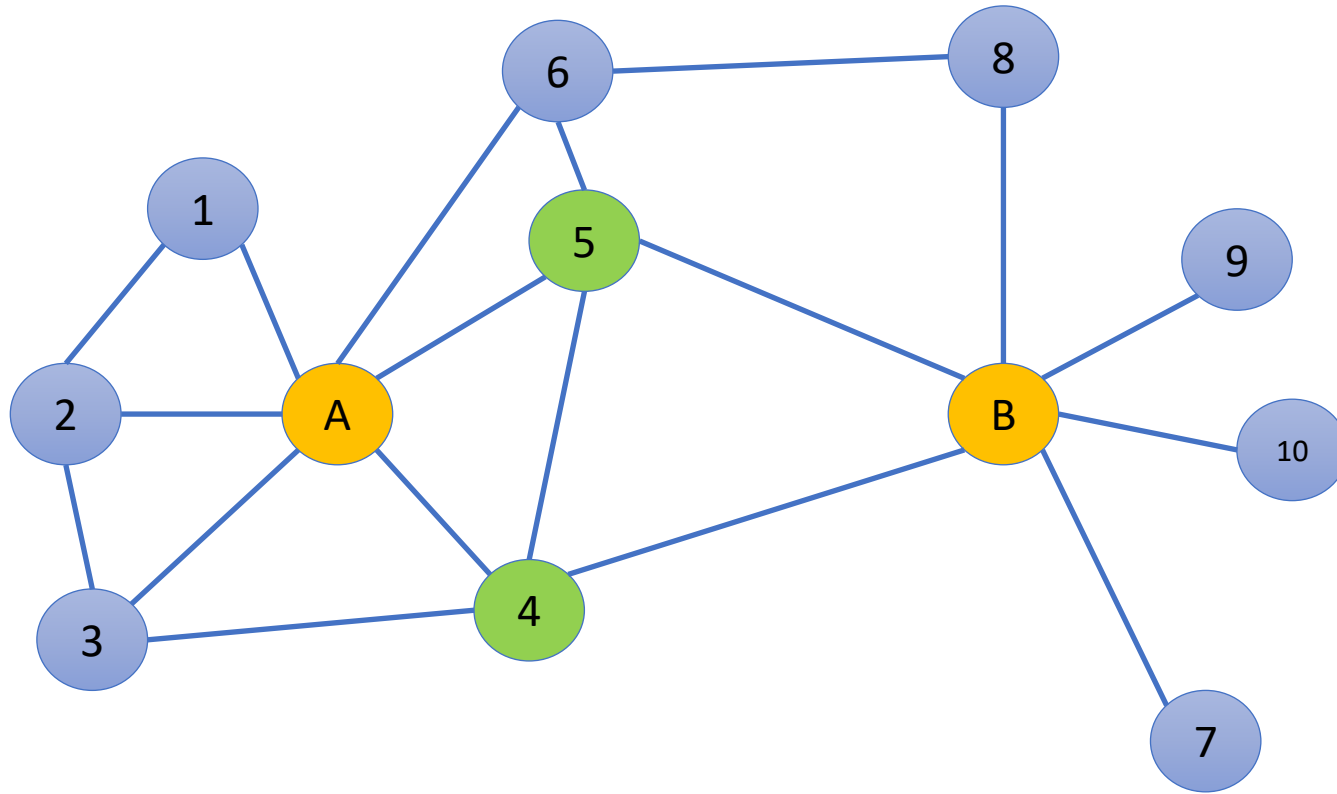
Jaccard Coefficient

- Use proportion of the common friends as a similarity metric
- $\text{Jaccard}(N(A), N(B)) = \frac{|N(A) \cap N(B)|}{|N(A) \cup N(B)|}$

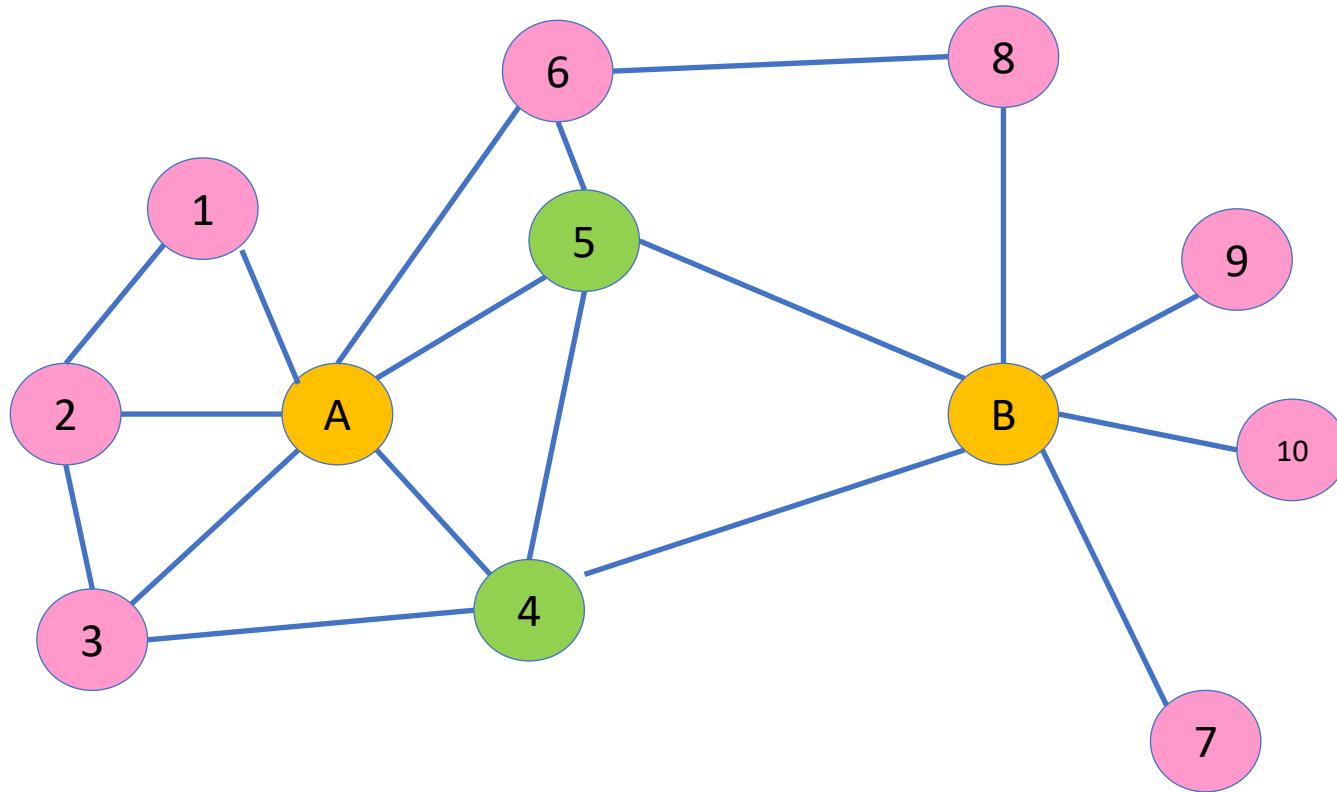
Example



Consider neighbors in-common

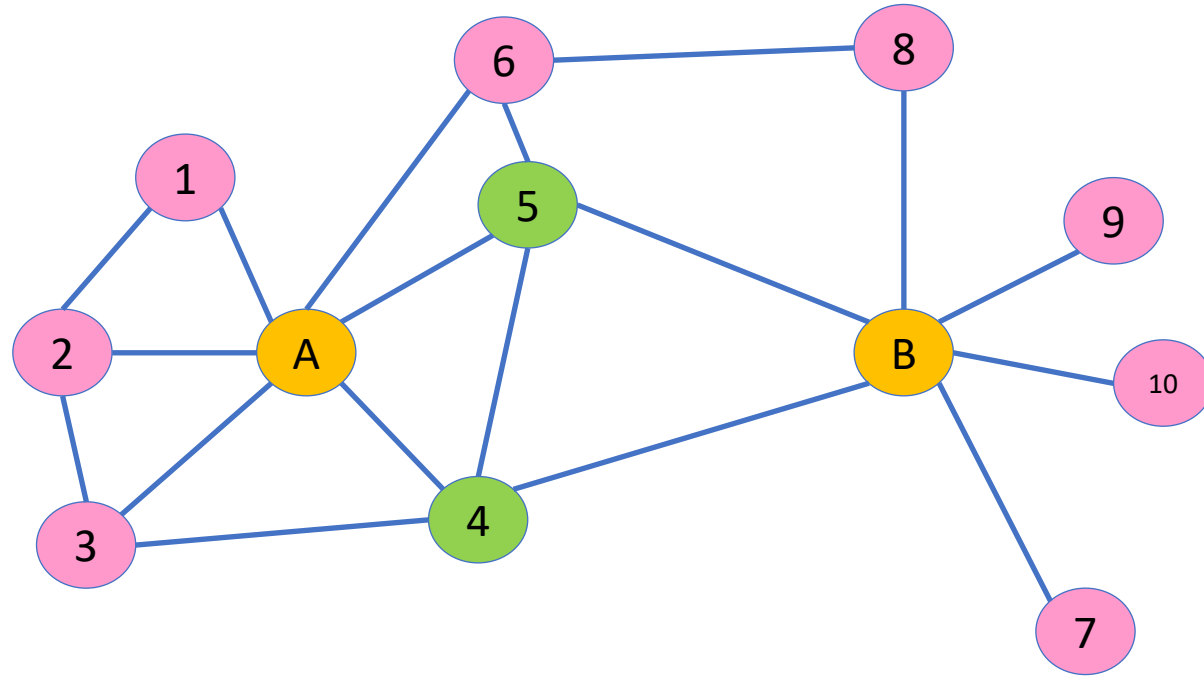


Consider neighbors not in-common

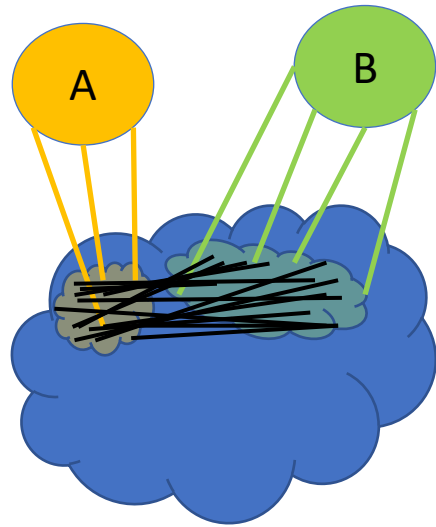


Jaccard Coefficient

- $\text{Jaccard}(A,B) = 2/(2+8) = 20\%$

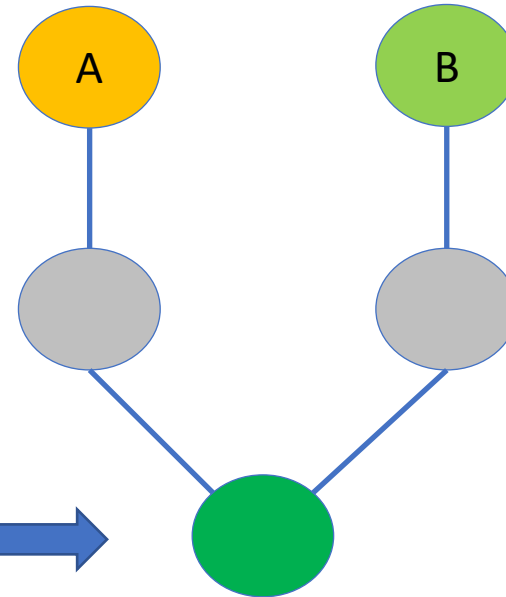


Caveat – 1: Common friends beyond 1-hop



$\text{sim}(A,B)=0$

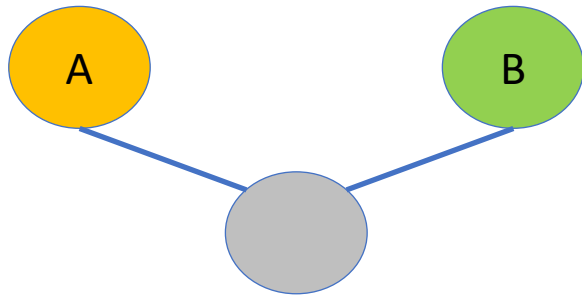
Simpler case:
common friend-of-friend



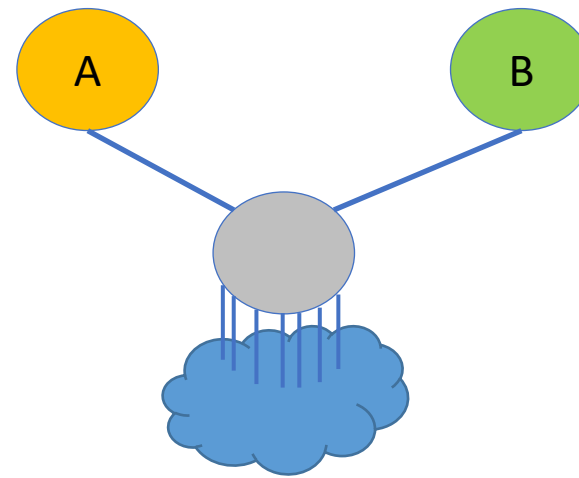
Should we increase $\text{sim}(A,B)$
because of this node?



Caveat – 2: Popularity of common neighbor



VS



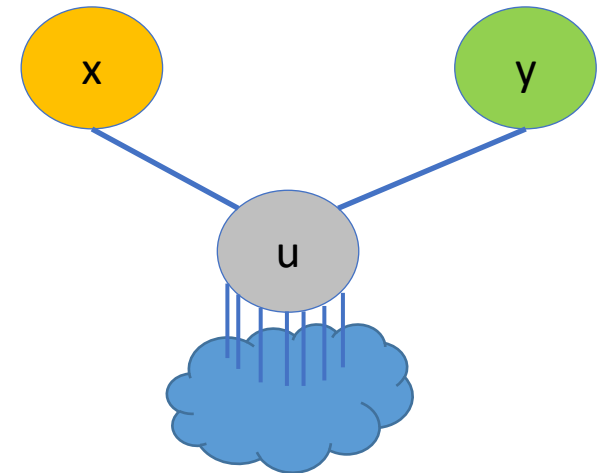
Adamic/Adar index

- The **Adamic/Adar index** computes the sum of the inverse logarithmic degree centrality of the neighbours shared by the two nodes

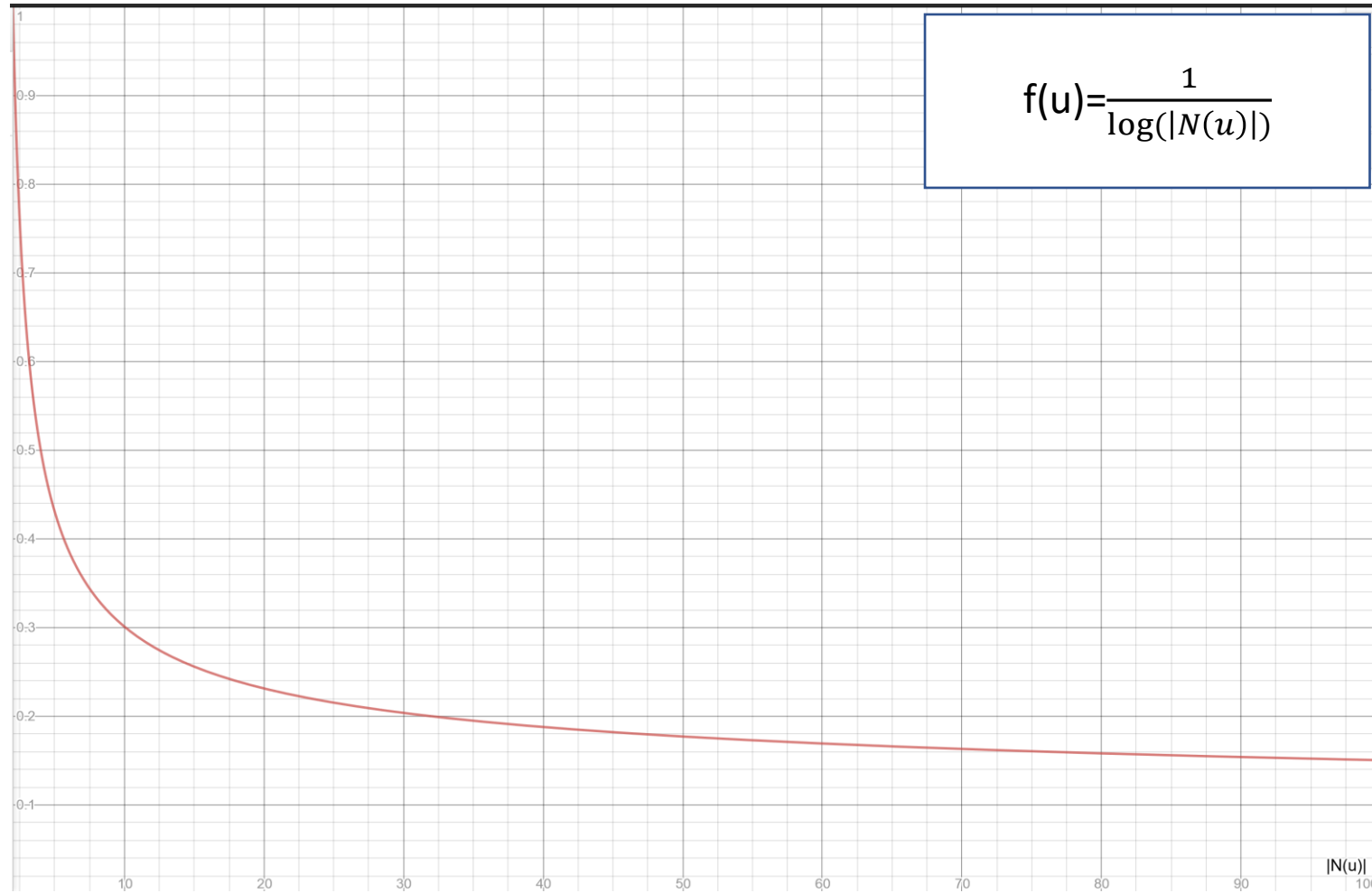
$$A(x, y) = \sum_{u \in N(x) \cap N(y)} \frac{1}{\log |N(u)|}$$

$N(x)$: set of neighbours of node x

For $u \in N(x) \cap N(y)$: $|N(u)| \geq 2$ (why?)



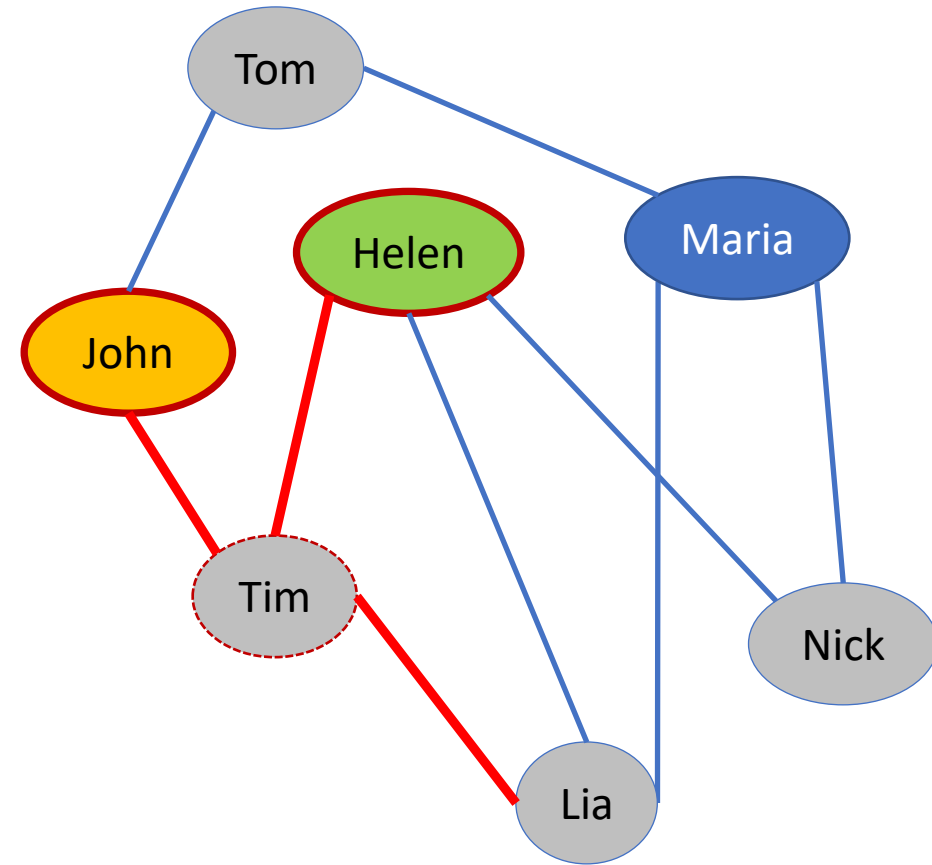
Reduce weight of common neighbor (u) with many connections ($|N(u)|$)



Adamic/Adar Example

$$A(x, y) = \sum_{u \in N(x) \cap N(y)} \frac{1}{\log |N(u)|}$$

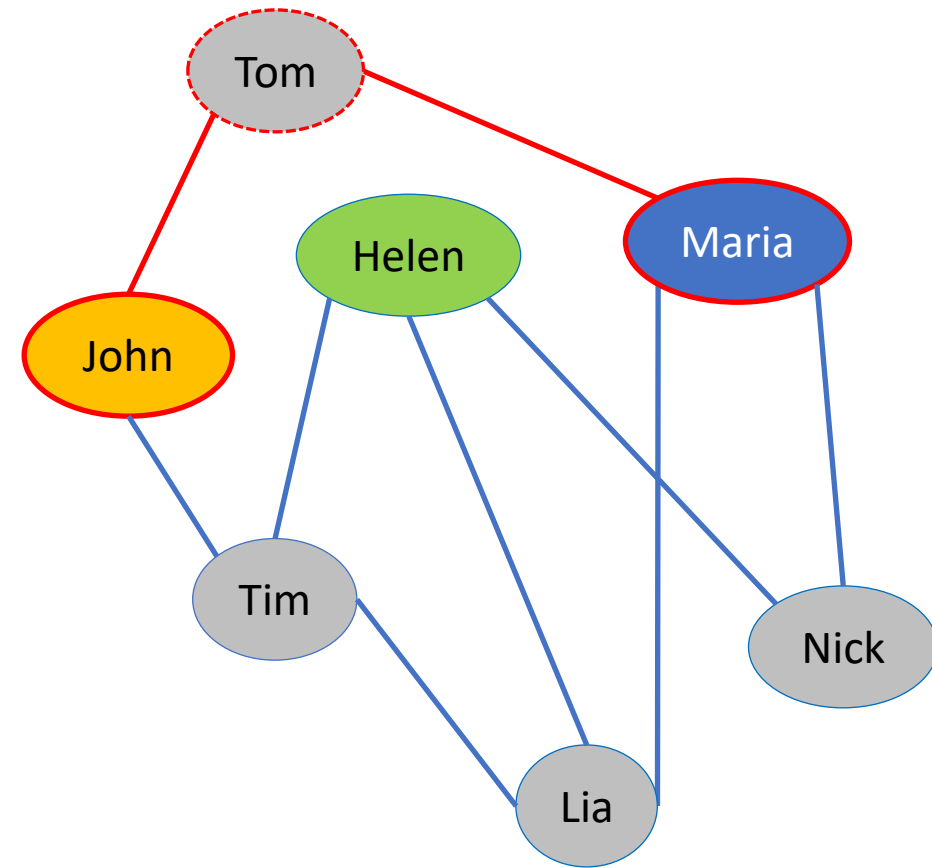
$$A(\text{John}, \text{Helen}) = 1/\log(3) = 0.63$$



Adamic/Adar Example

$$A(x, y) = \sum_{u \in N(x) \cap N(y)} \frac{1}{\log |N(u)|}$$

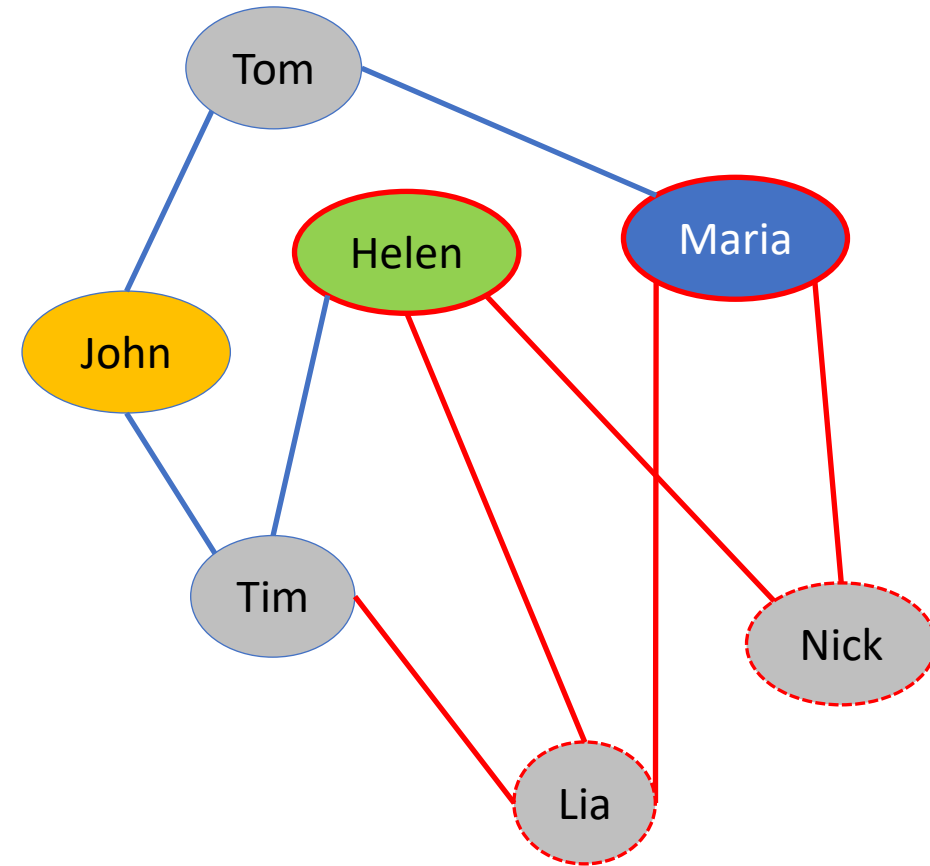
$$A(\text{John}, \text{Maria}) = 1/\log(2) = 1$$



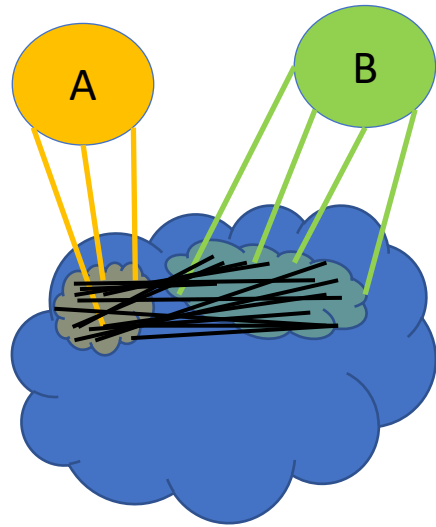
Adamic/Adar Example

$$A(x, y) = \sum_{u \in N(x) \cap N(y)} \frac{1}{\log |N(u)|}$$

$$A(\text{Helen}, \text{Maria}) = 1/\log(3) + 1/\log(2) = 1.63$$

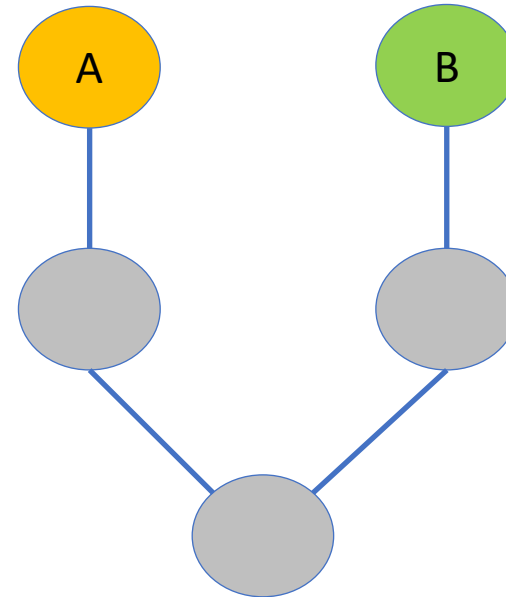


Caveat – 1: Need to look beyond 1-hop



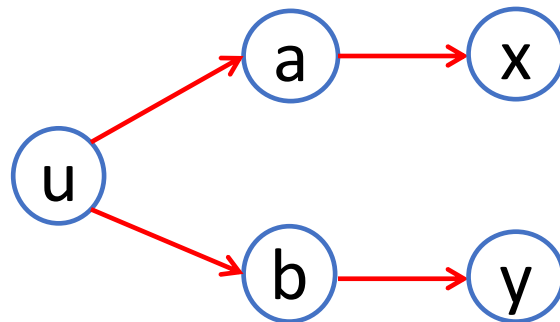
$$\text{sim}(A,B)=0$$

Simpler case:
common friend-of-friend



SimRank: two nodes are similar if they are referenced by similar nodes

- Recursive calculation in the same spirit as pageRank
- In the example bellow u references both a,b and this contributes to their similarity



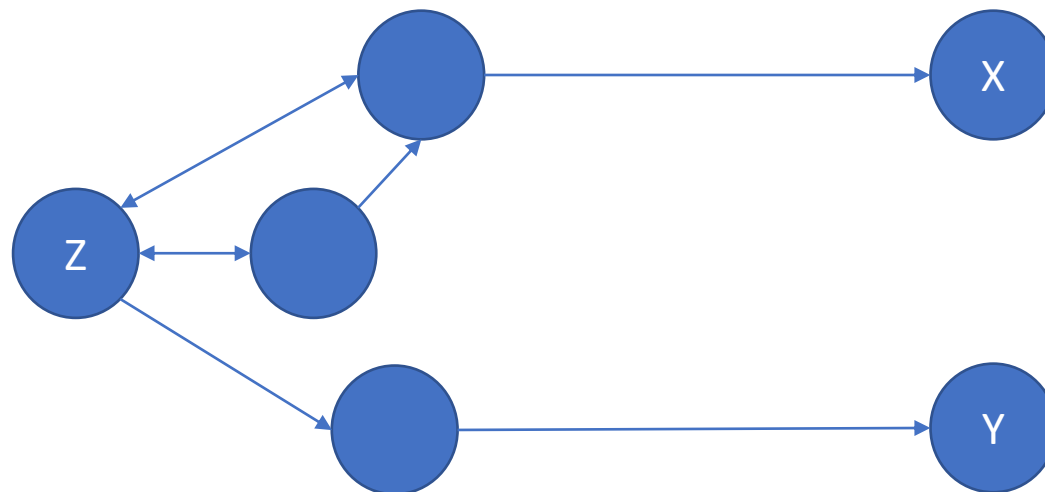
- Continuing this example, x and y are similar because they are referenced by a,b (respectively) that are similar
 - But a is not similar to x, nor b to y!

Motivation

- A similarity measure that exploits the object-to-object relationships found in many domains of interest
 - Web page X “points to” Web page Y
 - Customer “buys” product
 - Customer X transfers money to Customer Y
- May be used to cluster objects, such as for collaborative filtering in a recommender system

Intuition

- Concentrate on *structural* content
 - Can be combined with other similarity metrics that consider content similarity
- Two nodes are similar if they are referenced by similar nodes
 - Accounts X,Y are similar if they both receive money from some account Z (directly or in-directly)



SimRank Recursive Computation

- Initialize:

- $s(a,b) = \begin{cases} 1, & \text{if } a=b \\ 0, & \text{otherwise} \end{cases}$

- Iteratively compute ($a \neq b$):

$$s(a, b) = \frac{C}{|I(a)||I(b)|} \sum_{i=1}^{|I(a)|} \sum_{j=1}^{|I(b)|} s(I_i(a), I_j(b))$$

- Where

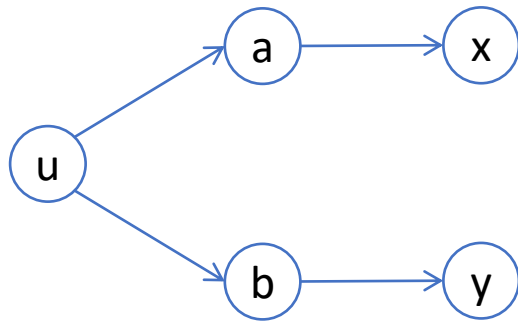
- $I(x)$ = set of in-neighbors of x
 - $I_i(x)$ = i^{th} in-neighbor of x and $C < 1$ (decay factor)

Explanation

$$s(a, b) = \frac{C}{|I(a)||I(b)|} \sum_{i=1}^{|I(a)|} \sum_{j=1}^{|I(b)|} s(I_i(a), I_j(b))$$

- Nodes receive the average similarity of their in-neighbors multiplied by the **decay factor C**
- Special case: $s(a,b) = 0$ if $|I(a)| = 0$ or $|I(b)| = 0$
 - i.e. nodes have no in-neighbors

Example



Initialization

$$s(u,u)=1$$

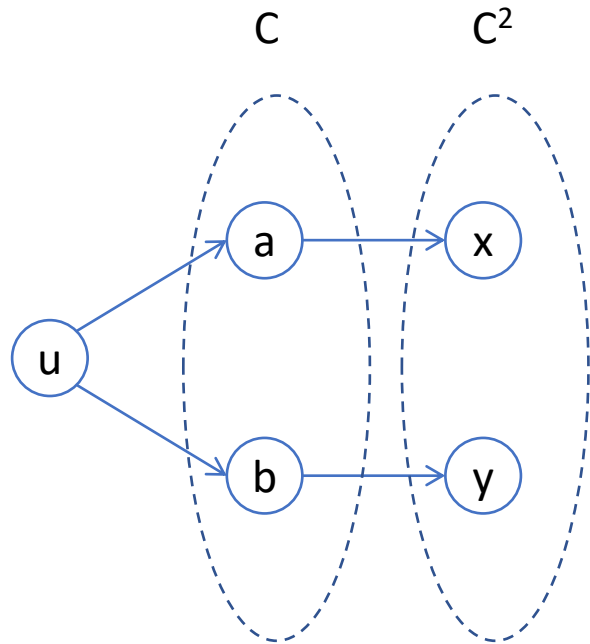
$$s(a,b)=0$$

$$s(a,x)=0$$

$$s(x,y)=0$$

Assume $C=0.8$

Iterate



Updated SimRank

$$s(u,u)=1$$

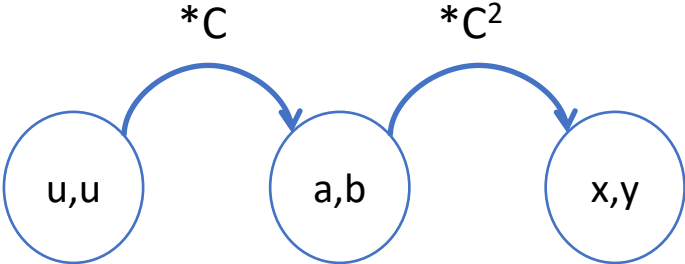
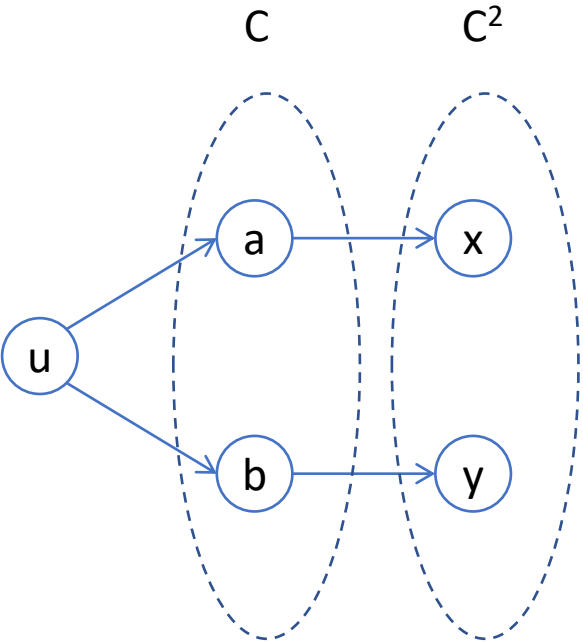
$$s(a,b)=0.8*s(u,u)=0.8$$

$$s(a,x)=0.8*s(u,a)=0$$

$$s(x,y)=0,8*s(a,b)=0,8*0,8=0,64$$

Assume $C=0.8$

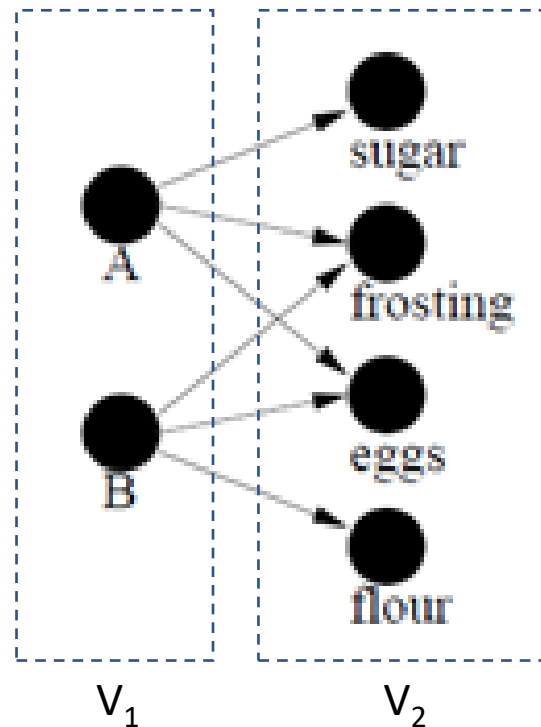
SimRank propagation



Assume $C=0.8$

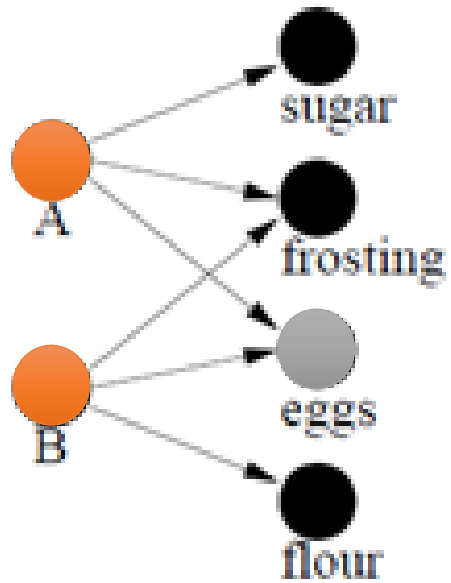
SimRank in bipartite graphs

- Bipartite graph: two disjoint classes of nodes V_1 , V_2
 - e.g. V_1 ={customers}, V_2 ={items}
 - Edges only between nodes in V_1 to nodes in V_2



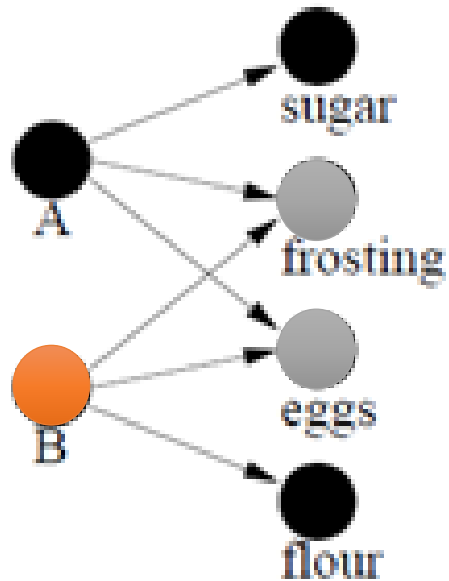
Intuition-1

- People are similar if they purchase similar objects



Intuition-2

- Items are similar if they are purchased by similar people



Bipartite SimRank

- SimRank between persons A and B, (A≠B)

$$s(A, B) = \frac{C_1}{|O(A)||O(B)|} \sum_{i=1}^{|O(A)|} \sum_{j=1}^{|O(B)|} s(O_i(A), O_j(B))$$

- SimRank between items x and y, (x≠y)

$$s(x, y) = \frac{C_2}{|I(x)||I(y)|} \sum_{i=1}^{|I(x)|} \sum_{j=1}^{|I(y)|} s(I_i(x), I_j(y))$$

- The similarity between persons A and B is the average similarity between the items they purchased
 - $O(A)$ are the out-neighbors (items) for person A
- The similarity between items x and y is the average similarity between the people who purchased them

Modified SimRank in bipartite graphs

