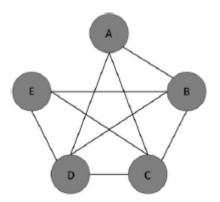
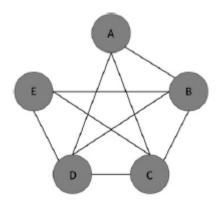
Entity Resolution and Link Prediction

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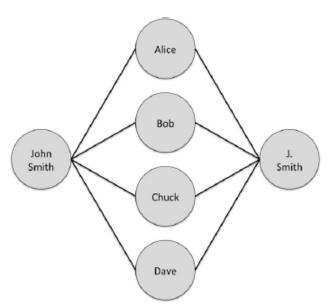


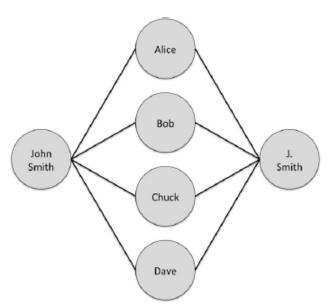
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 Link Prediction is method of analysis that detects where missing links should be present





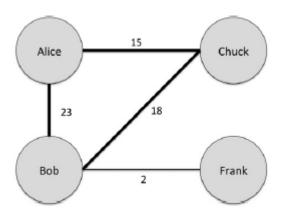


• Analyze the network at a set point of time

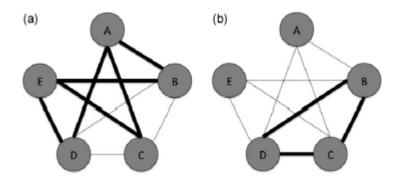
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- useful in data cleaning for "missing links"

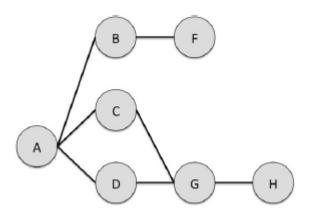
Link Prediction example



Missing Link



score(A,B)



$$\mathsf{score}(\mathsf{A},\mathsf{B}) = \mathsf{-ShortestPath}(\mathsf{A},\mathsf{B})$$

 $\mathsf{score}(\mathsf{A},\mathsf{B}) = \mathsf{Neighbours}(\mathsf{A}) \cap \mathsf{Neighbours}(\mathsf{B})$

$$score(A,B) = \frac{|\textit{Neighbours}(A) \cap \textit{Neighours}(B)|}{|\textit{Neighbours}(A) \cup \textit{Neighours}(B)|}$$

This is known as Jaccard Index

Case Study: Jaccard Index

Say there are four nodes, Alice, Bob, Chuck and Dave Let Alice and Bob be celebrities, each with 1 million friends Chuck and Dave are average users with 100 friends each. Say Alice and Bob have 2000 friends in common while Chuck and Dave have only 20 friends in common.

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$$score(Alice, Bob) = \frac{2000}{1000000 + 1000000 - 2000} = 0.001$$

 $score(Chuck, Dave) = \frac{20}{100 + 100 - 20} = 0.11$

Adamic and Adar, 2003

$$score(A, B) = \sum_{x \in Neighbours(A) \cap Neighbours(B)} \frac{1}{log(|Neighbours(x)|)}$$

preferential attachment

$$score(A, B) = |Neighbours(A)| * |Neighbors(B)| = degree(A) * degree(B)$$

Advanced Link Prediction Techniques

- We can take the average ranking of each node pair from each measure and rank by that value
- The result would be a ranking that considers all the factors
- probabilistic models(Markov Networks)
- some approaches consider node's attributes in addition to network structure
- weighted and directed graphs
- machine learning