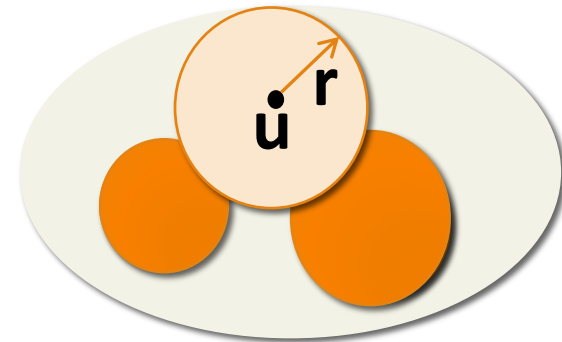




# + Session 6. SpiderMine: Mining Top-K Large Structural Patterns in a Single Network

# SpiderMine: Mining Top-K Large Structural Patterns in a Massive Network

- ❑ Large patterns are informative to characterize a large network (e.g., social network, web, or bio-network)
- ❑ Similar to pattern fusion, mining large pattern should not aim for completeness but for representativeness of the target results
- ❑ Spider-Mine (F. Zhu, et al., VLDB'11): Mine top- $K$  largest frequent substructure patterns whose diameter is bounded by  $D_{\max}$  with a probability at least  $1-\epsilon$
- ❑ General idea: Large patterns are composed of a number of small components (“spiders”) which will eventually connect together after some rounds of pattern growth
- ❑ **r-Spider:** An  $r$ -spider is a frequent graph pattern  $P$  such that there exists a vertex  $u$  of  $P$ , and all other vertices of  $P$  are within distance  $r$  from  $u$



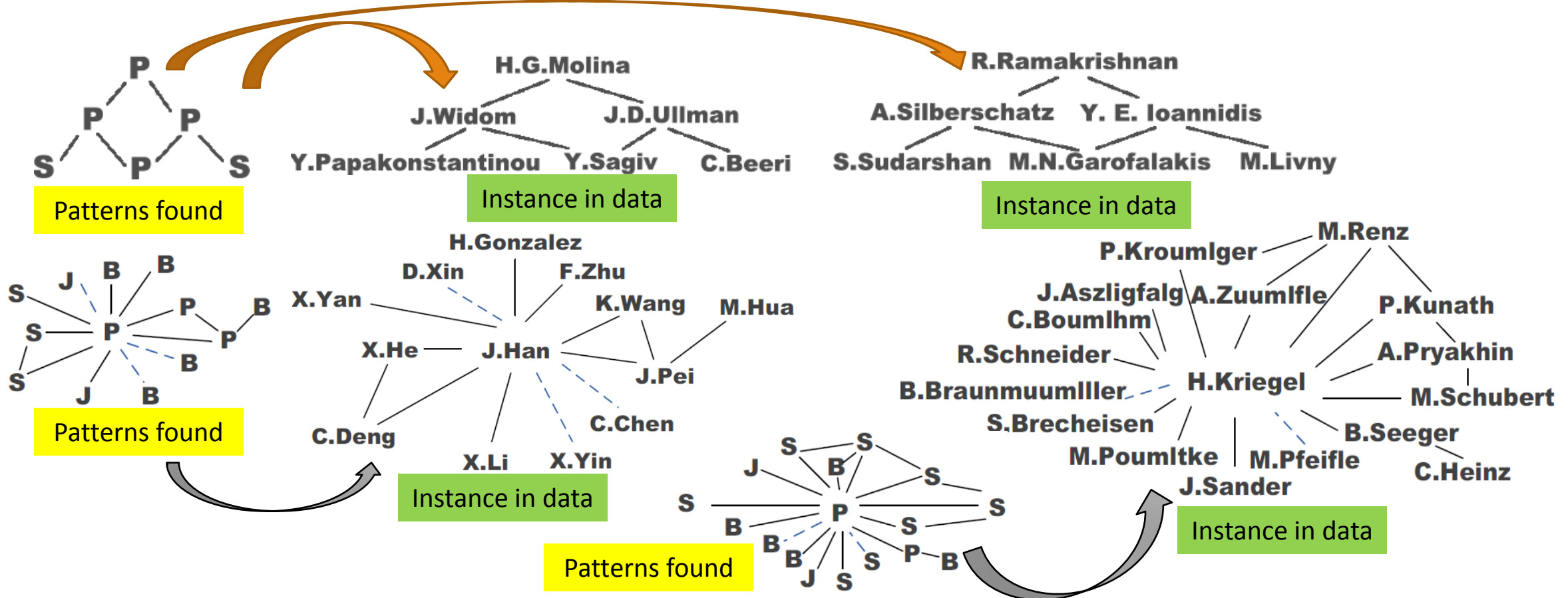
# Why Is SpiderMine Good for Mining Large Patterns

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- ❑ The SpiderMine Algorithm
  - ❑ Mine the set  $S$  of all the  $r$ -spiders
  - ❑ Randomly draw  $M$   $r$ -spiders
  - ❑ Grow these  $M$   $r$ -spiders for  $t = D_{\max}/2$  iterations, and merge two patterns whenever possible
  - ❑ Discard unmerged patterns
  - ❑ Continue to grow the remaining ones to maximum size
  - ❑ Return the top- $K$  largest ones in the result
- ❑ Why is SpiderMine likely to retain large patterns and prune small ones?
  - ❑ Small patterns are much less likely to be hit in the random draw
  - ❑ Even if a small pattern is hit, it is even less likely to be hit multiple times
  - ❑ The larger the pattern, the greater the chance it is hit and saved

# Mining Collaboration Patterns in DBLP Networks

- ❑ Data description: 600 top confs, 9 major CS areas, 15071 authors in DB/DM
- ❑ Author labeled by # of papers published in DB/DM
  - ❑ Prolific (P):  $\geq 50$ , Senior (S): 20~49, Junior (J): 10~19, Beginner(B): 5~9



# Summary

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- ❑ Graph pattern mining: Basic concepts
- ❑ Apriori-based graph pattern mining methods
- ❑ gSpan: A pattern-growth-based method
- ❑ CloseGraph: Mining closed graph patterns
- ❑ Graph Indexing: A graph pattern mining application example
- ❑ SpiderMine: Mining top-k large structural patterns in a large network



# Recommended Readings

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- ❑ C. Borgelt and M. R. Berthold, “Mining molecular fragments: Finding relevant substructures of molecules”, ICDM'02
- ❑ J. Huan, W. Wang, and J. Prins. “Efficient mining of frequent subgraph in the presence of isomorphism”, ICDM'03
- ❑ A. Inokuchi, T. Washio, and H. Motoda. “An apriori-based algorithm for mining frequent substructures from graph data”, PKDD'00
- ❑ M. Kuramochi and G. Karypis. “Frequent subgraph discovery”, ICDM'01
- ❑ S. Nijssen and J. Kok. A quickstart in frequent structure mining can make a difference. KDD'04
- ❑ N. Vanetik, E. Gudes, and S. E. Shimony. “Computing frequent graph patterns from semistructured data”, ICDM'02
- ❑ X. Yan and J. Han, “gSpan: Graph-Based Substructure Pattern Mining”, ICDM'02
- ❑ X. Yan and J. Han, “CloseGraph: Mining Closed Frequent Graph Patterns”, KDD'03
- ❑ X. Yan, P. S. Yu, and J. Han, “Graph Indexing: A Frequent Structure-based Approach”, SIGMOD'04
- ❑ F. Zhu, Q. Qu, D. Lo, X. Yan, J. Han, and P. S. Yu, "Mining Top-K Large Structural Patterns in a Massive Network", VLDB'11