

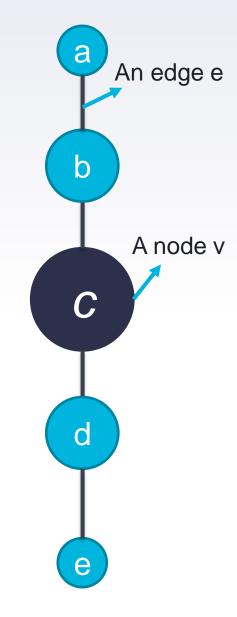
### Betweenness Centrality in Dynamic Graphs

Class: Presenter: COMP 5704 Nathan Bowness

## Betweenness Centrality (BC)

BC for a node v

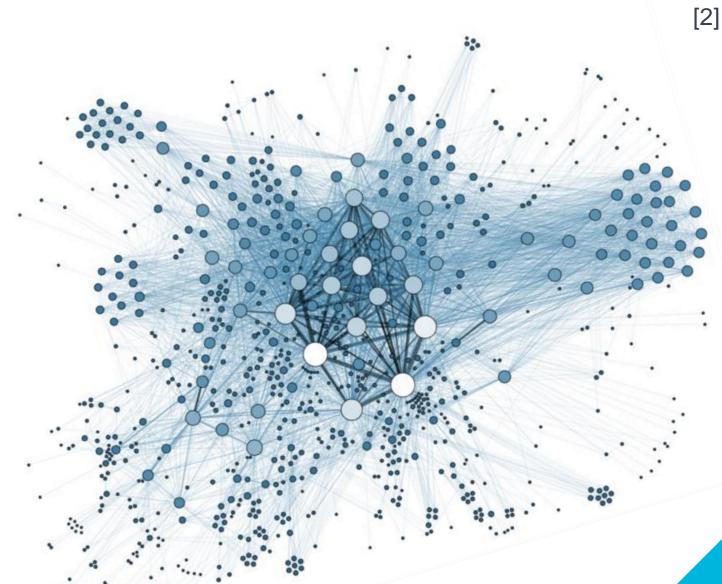
Is the fraction of the shortest paths between all pairs of nodes that passthrough *v* [1]



BC of **c** > BC of **b**, **d** > BC of **a**, **e** 

### Why Use Betweenness Centrality?

- Social Networks
- Transportation Networks
- Road Networks



### Calculating Betweenness Centrality in Static Graphs

#### Formula:

$$BC_G[v] = \sum_{\substack{s,t \in V, \\ s \neq t \neq v}} \frac{\sigma_{st}(v)}{\sigma_{st}} = \sum_{\substack{s,t \in V, \\ s \neq t \neq v}}$$

# shortest paths from s to t that include v
# shortest paths from s to t

(1)

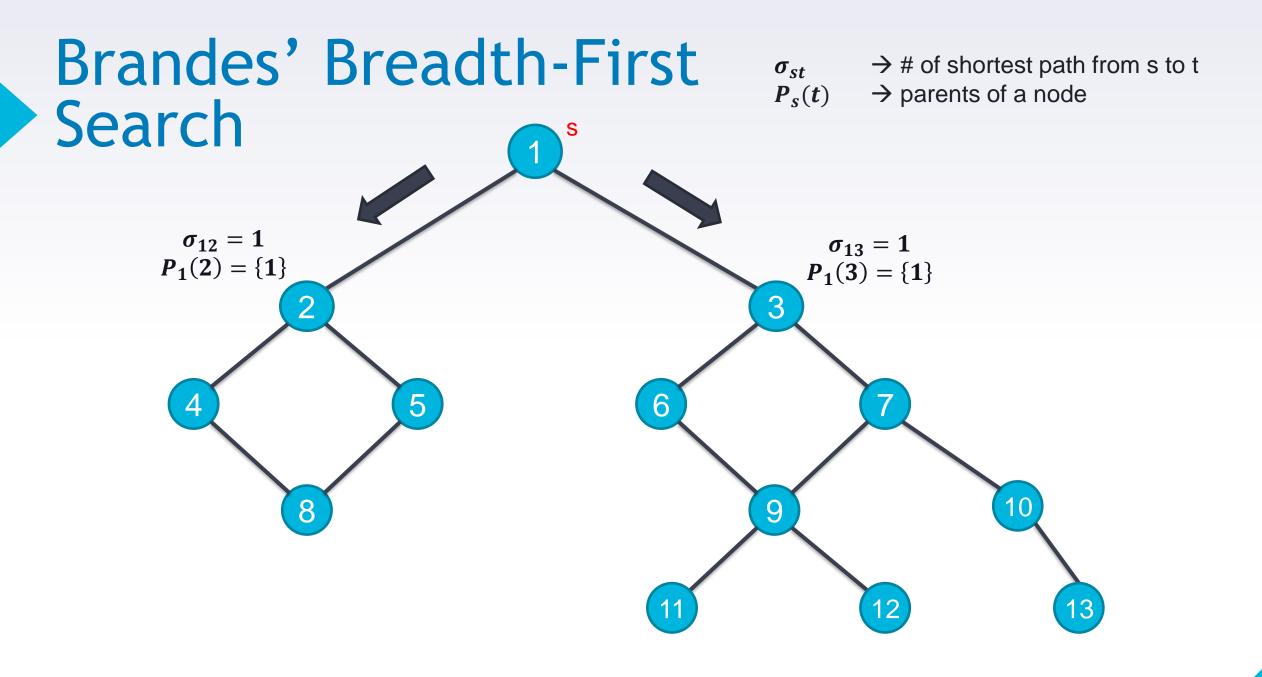
#### Brandes Algorithm:

- Use source dependencies to perform calculation
  - Breadth-first search (BFS)
  - Reverse breadth-first search (R-BFS)

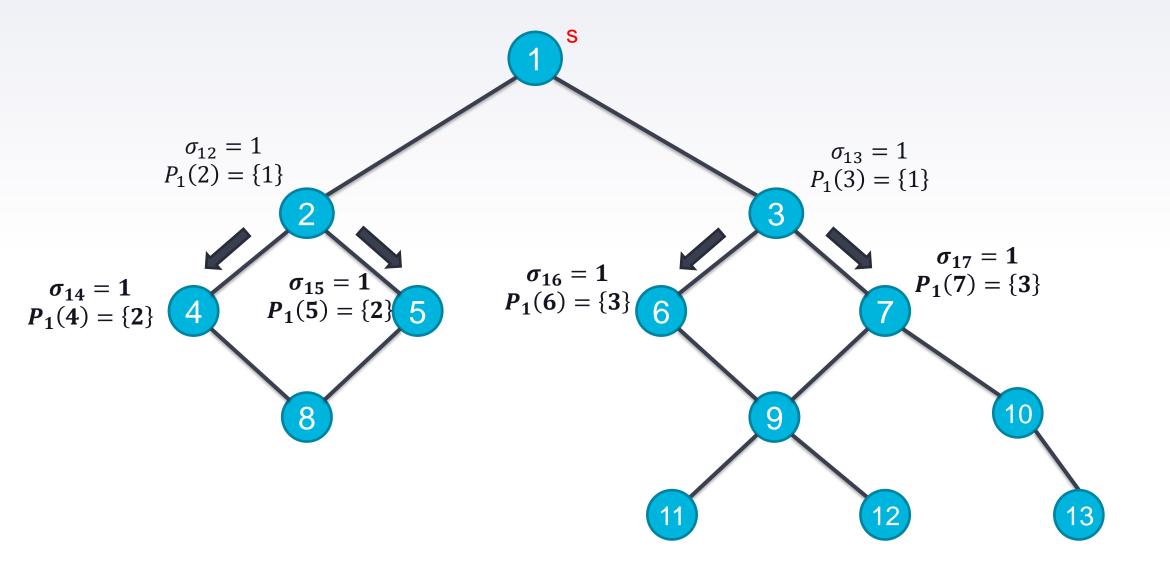
$$BC_G[v] = \sum_{s \in V, s \neq v} \delta_{s \bullet}(v) \qquad (2)$$

where, 
$$\delta_{s\bullet}(v) = \sum_{v \in P_s(w)} \frac{\sigma_{sv}}{\sigma_{sw}} \cdot (1 + \delta_{s\bullet}(w))$$
 (3) [3]

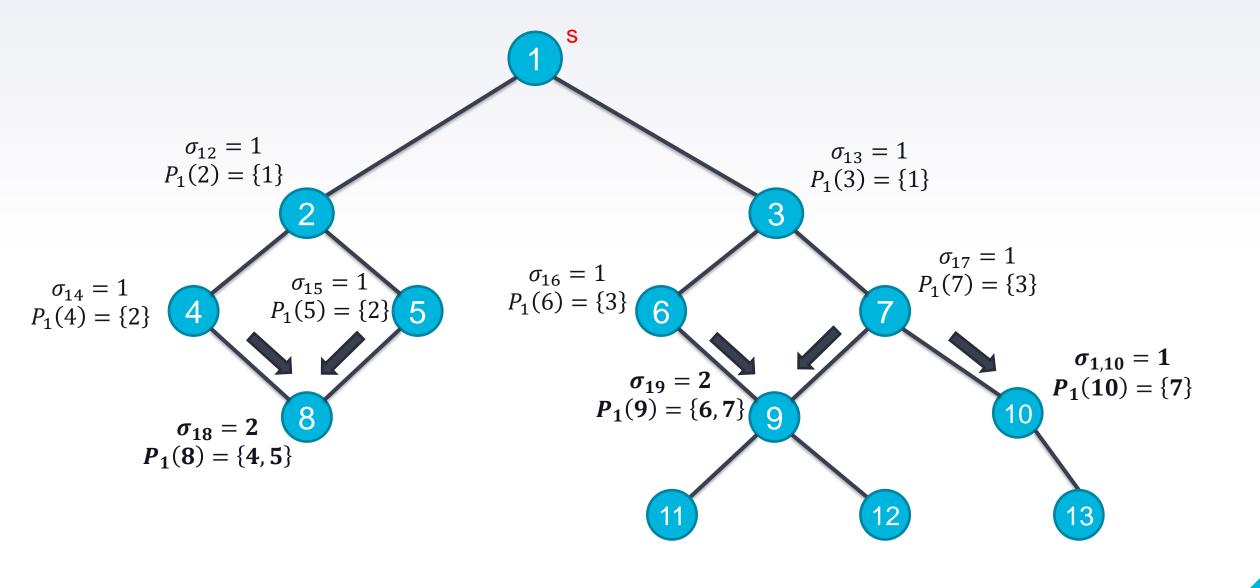
source dependencies



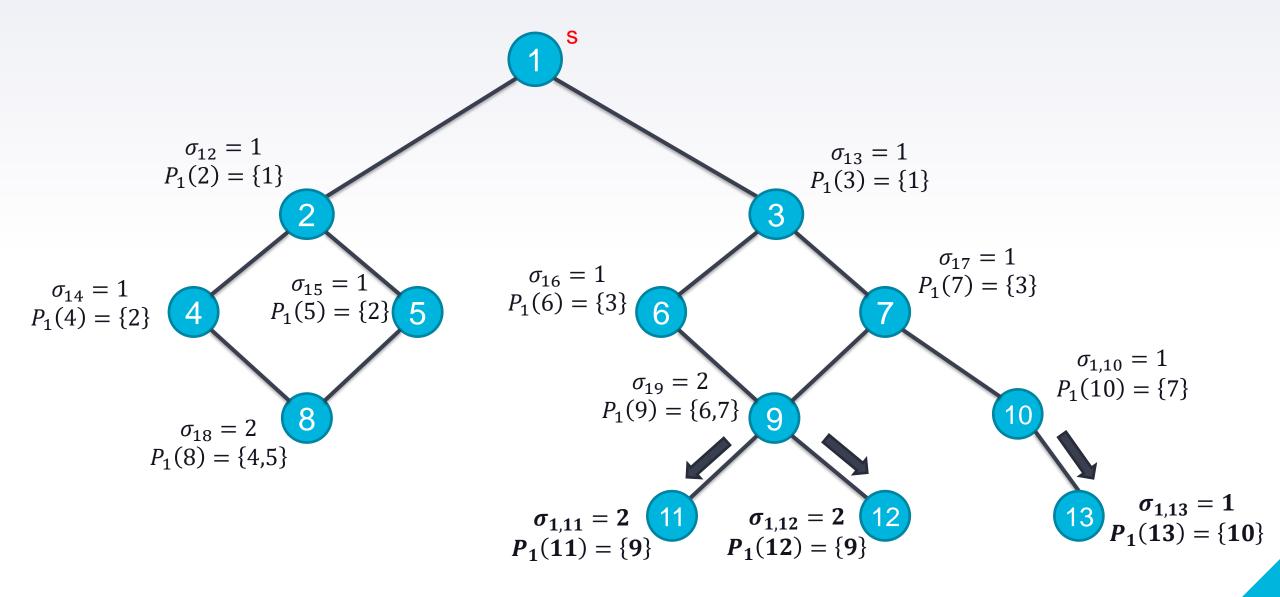
#### Brandes' BFS (Cont'd)

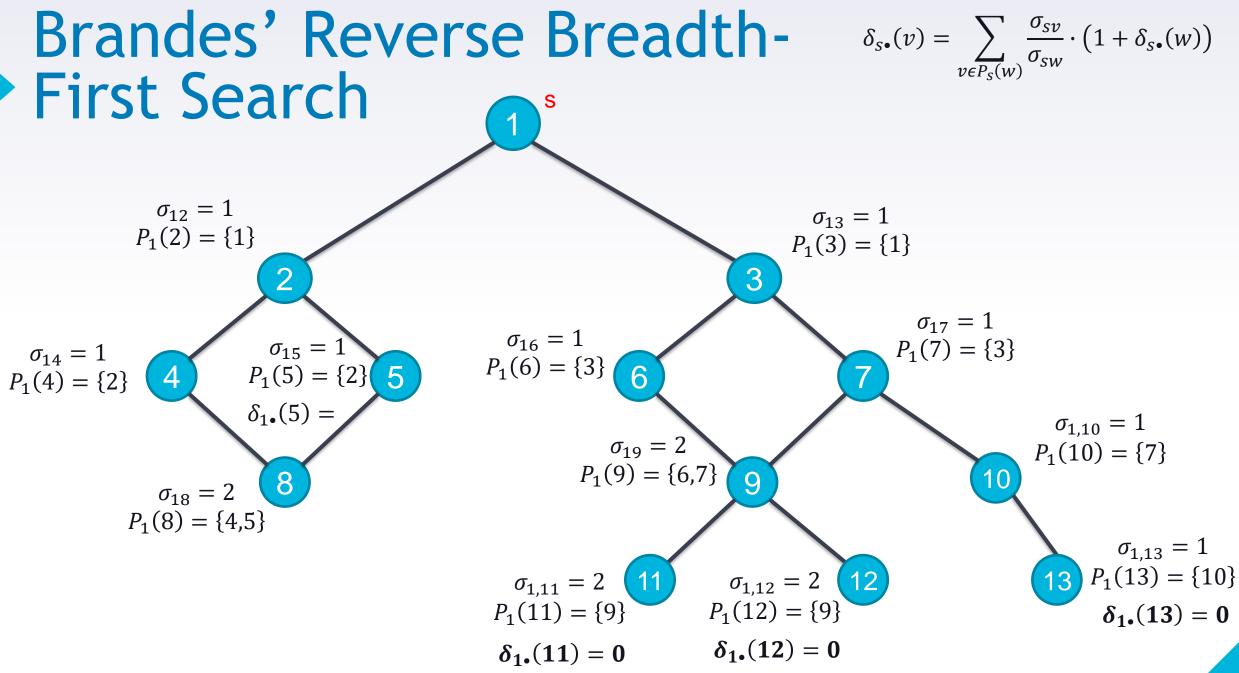


#### Brandes' BFS (Cont'd)

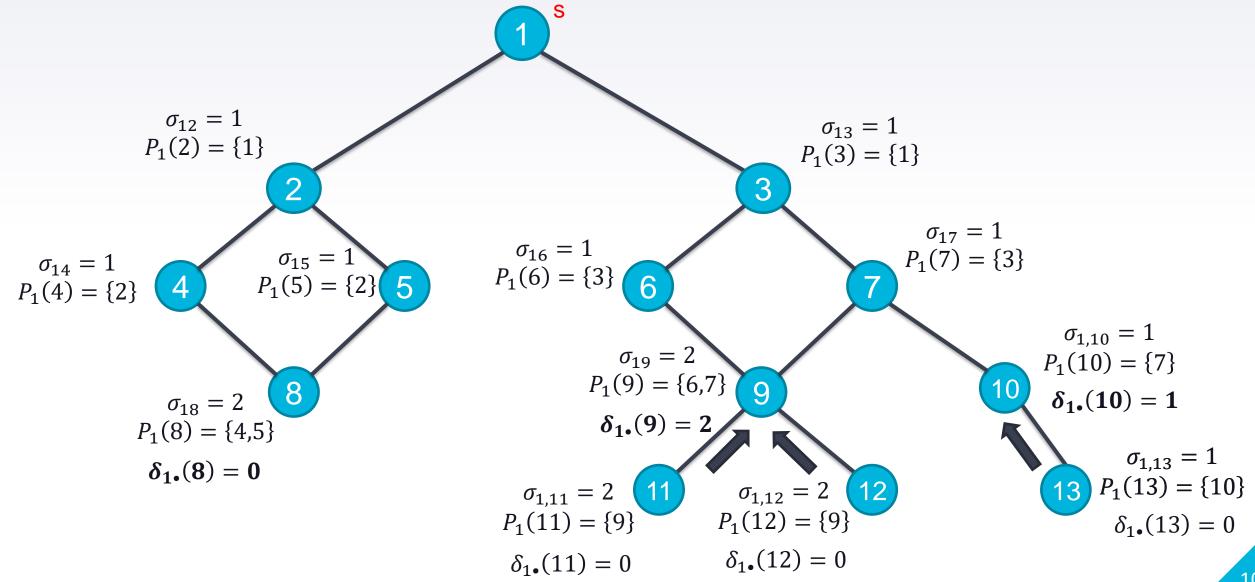


#### Brandes' BFS (Cont'd)

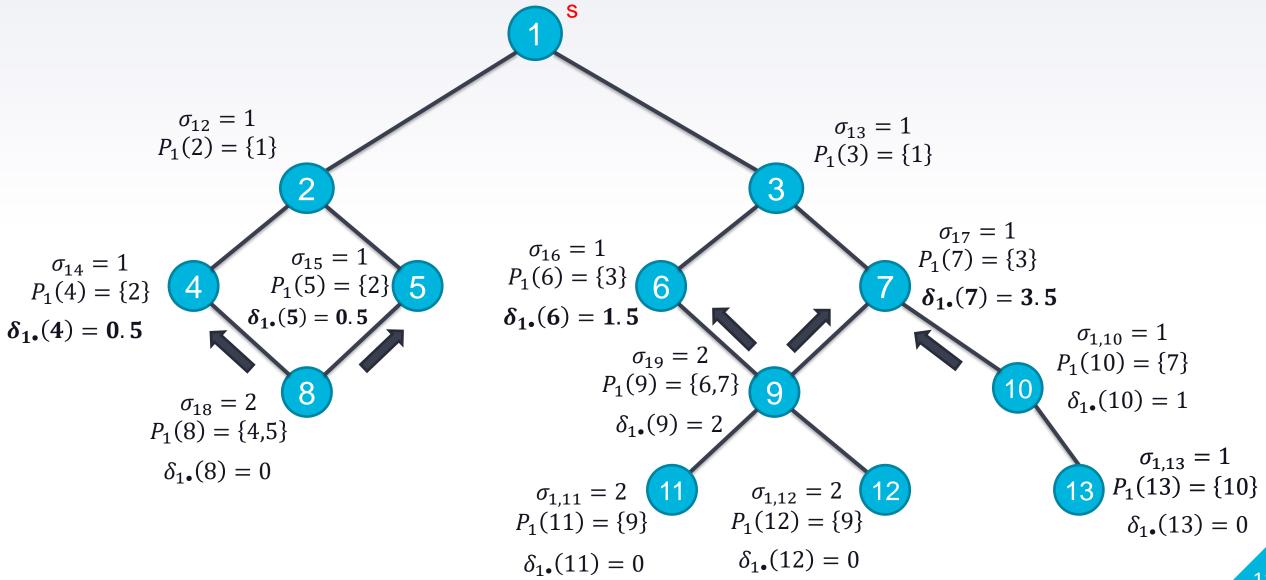


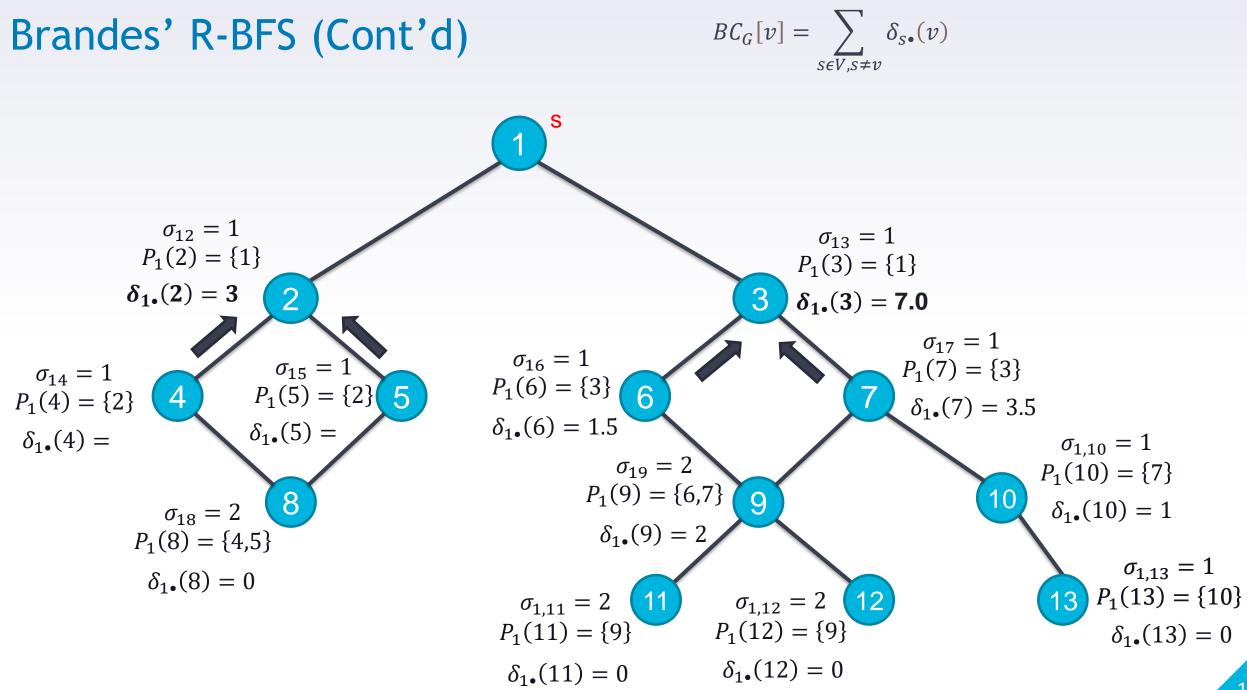


#### Brandes' R-BFS (Cont'd)



#### Brandes' R-BFS (Cont'd)





### **Calculating Betweenness Centrality**

Betweenness Centrality  $\rightarrow BC_G[v] = \sum_{s \in V.s \neq v} \delta_{s \bullet}(v)$ 

- **Completed**  $\rightarrow$  s = 1
- **Remaining** → s = 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13
- Once finished, we keep certain data to help out with dynamic algorithms
  - $\circ$   $\sigma_{st}$
  - $\circ$   $P_S$
  - $\circ$  S<sub>S</sub>
  - $\circ$  D<sub>S</sub>

### Betweenness Centrality in Dynamic Graphs

- Social Networks
- Transportation Networks
- Road Networks

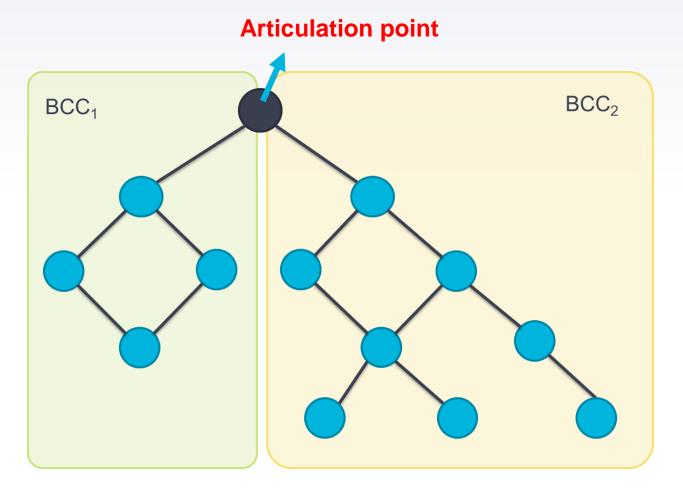
Networks are always changing

### Calculating BC in Dynamic Graphs - 2020

- Recent algorithm in 2020  $\rightarrow$  **Batch iCentral** by Shukla et Al. [4],[5]
- Key Concepts:
  - Avoid BFS
  - Recompute betweenness centrality after a batch of updates
  - Leverage previous stored data from Brandes'

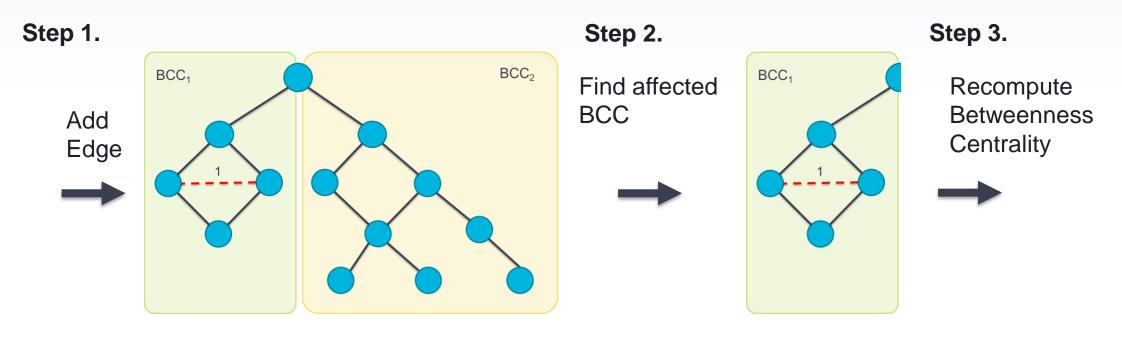
### Avoid BFS $\rightarrow$ Biconnected Components (BCC)

- Biconnected Components are a "maximal biconnected subgraph" [5]
- Connected only by **articulation points**
- Allow graph to be split into subsections



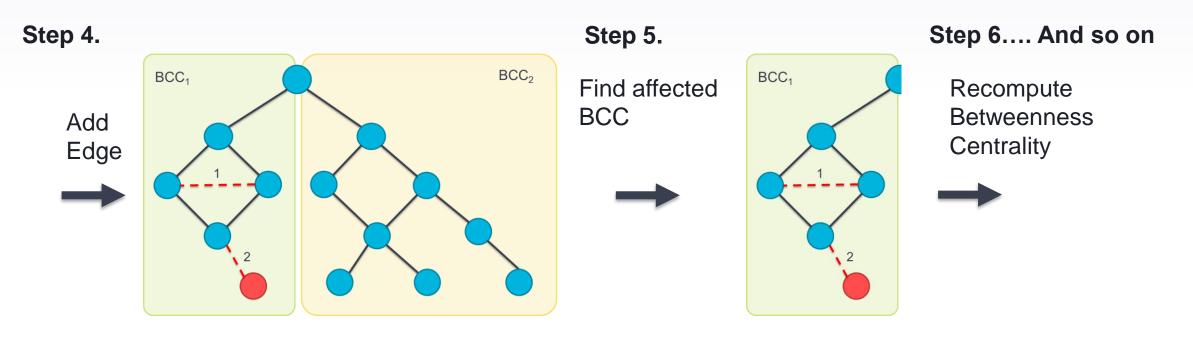
## Updating Edges in Sequence

- Previous algorithms applied edges 1-by-1 for Dynamic Graphs
- Regular iCentral does this



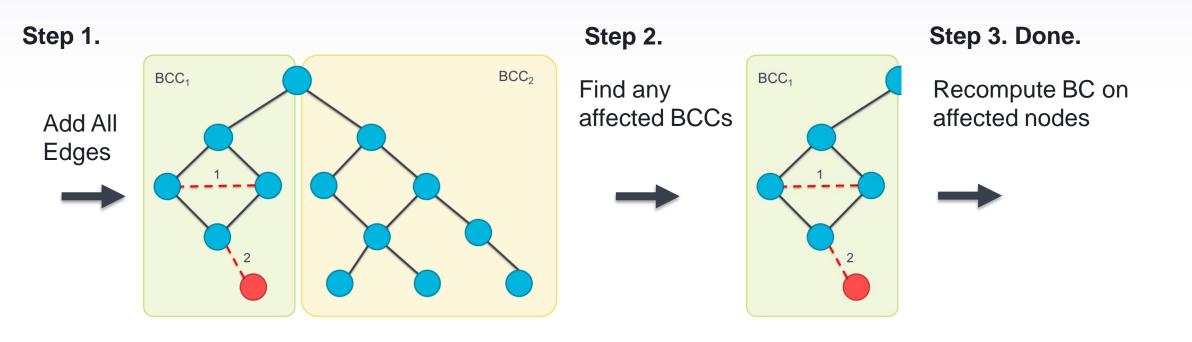
## Updating Edges in Sequence

- Previous algorithms applied edges 1-by-1 for Dynamic Graphs
- Regular iCentral does this



## Updating Edges in Batches

• Newer algorithms apply all edges to start, then recompute BC

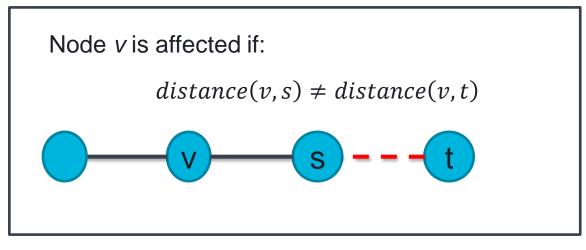


## Parallelizing BC Calculation

- Few ways to parallelize
  - On affected Biconnected Components
  - On affected nodes
  - (Rare) On Graphs

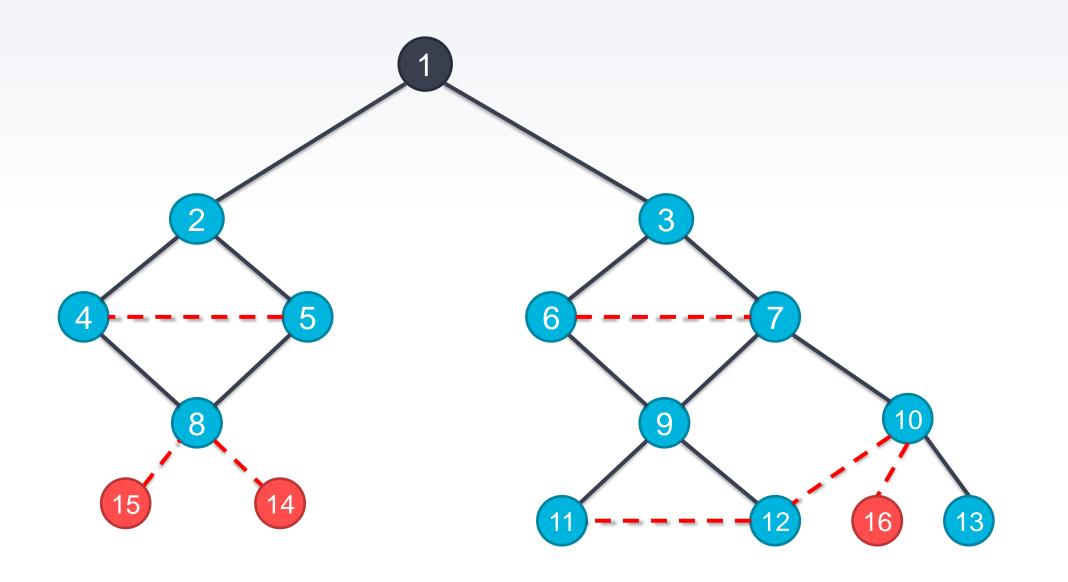
## Parallelizing BC Calculation

- Few ways to parallelize
  - On affected Biconnected Components
  - On affected nodes  $\leftarrow$
  - (Rare) On Graphs

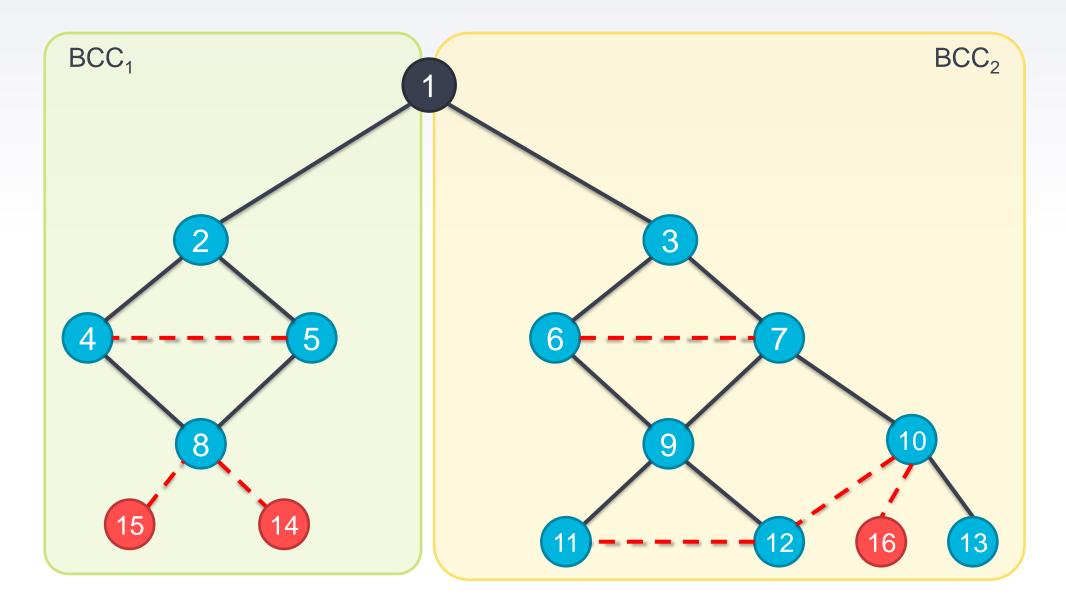


### Parallelizing BC: Example

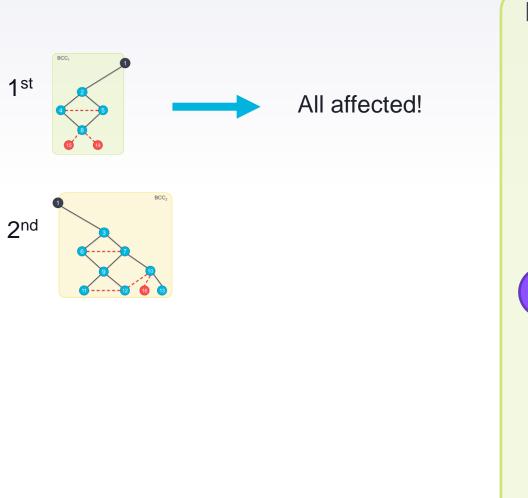
• Add 7 edges to the graph

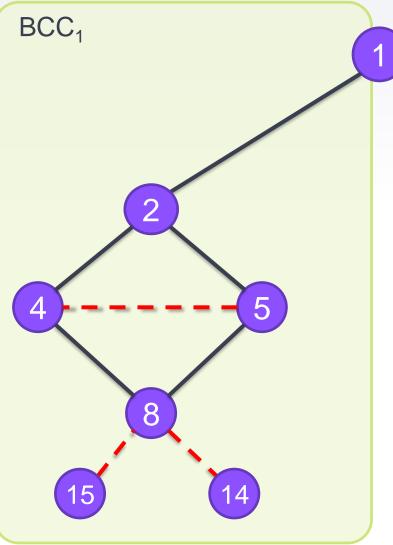


Identify affected Biconnected Components

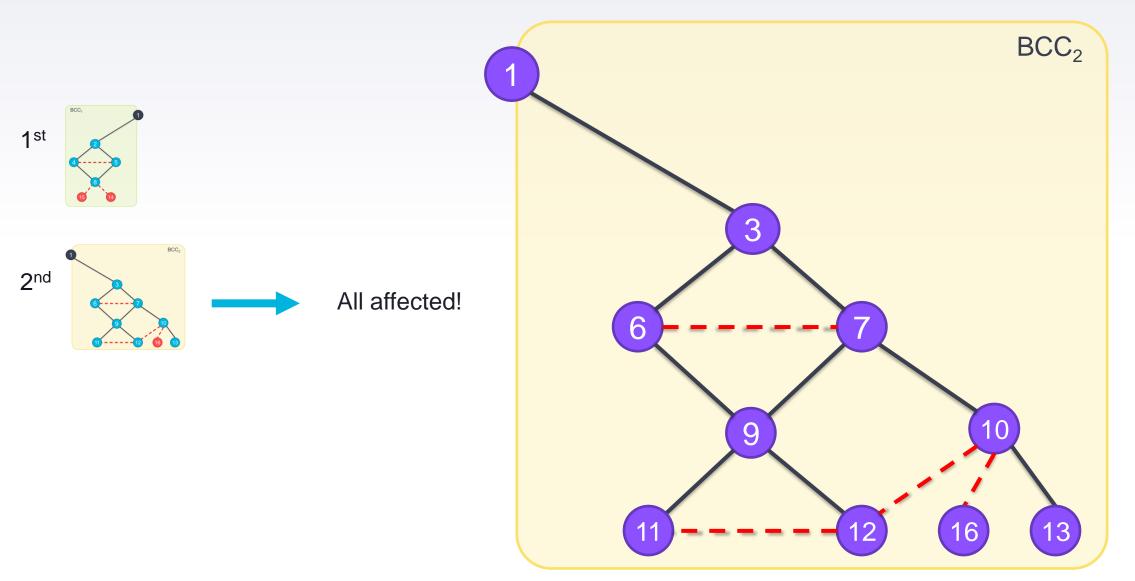


• Identify affected nodes for each BCC

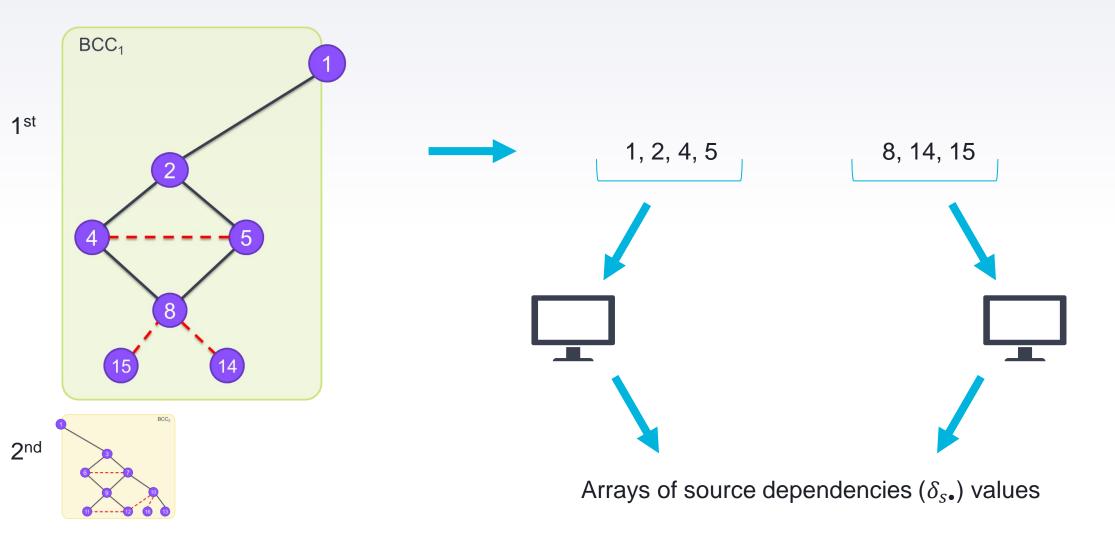




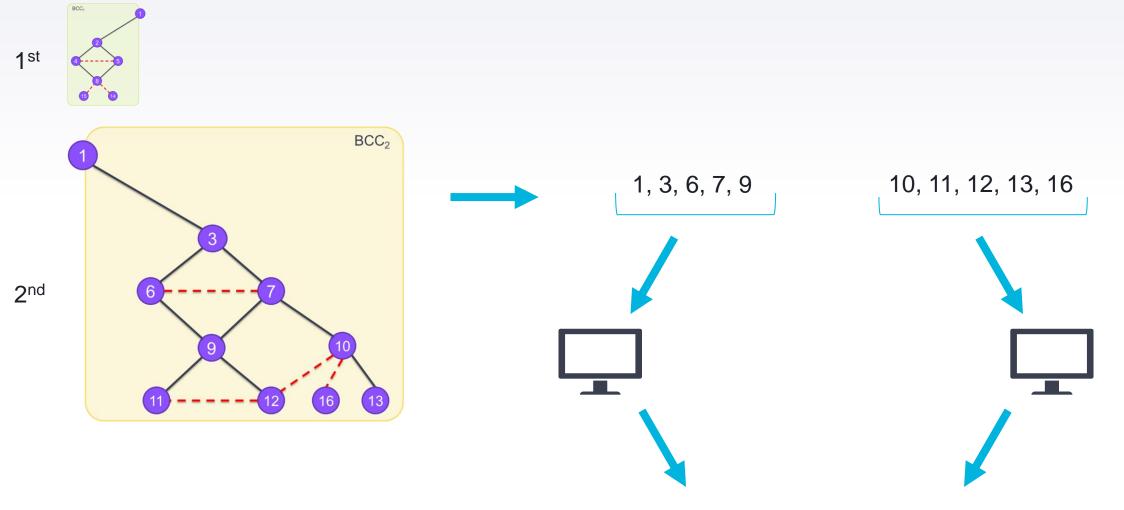
Identify affected nodes for each BCC



• Send groups of affected nodes to separate threads/machines



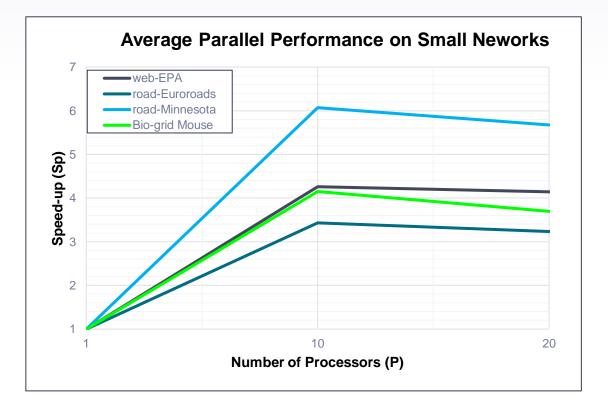
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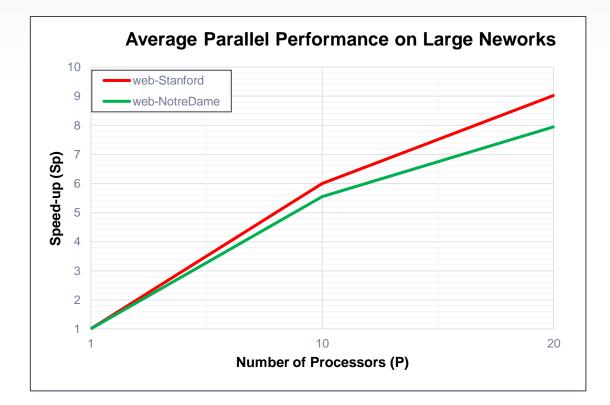


Arrays of source dependencies ( $\delta_{s}$ ) values

### Parallel Performance

- Non-linear speed-up
- Better parallel performance on larger graphs to be expected

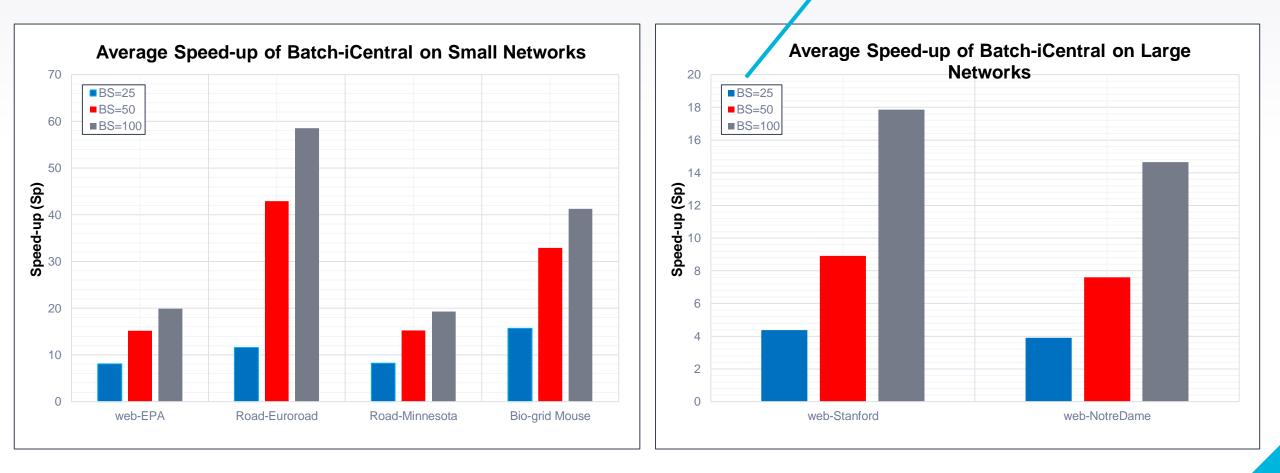




## **Comparison of Results**

BS – Batch Size

Batch-iCentral vastly out-performs regular iCentral



# Thanks For Listening! Any questions?

# Question 1:

• What type of graph traversal is used when calculating betweenness centrality?

# Question 2:

 When recalculating BC in a dynamic graph, is it more effective to process edges one-by-one or in a batch?

# Question 3:

• What is one section/element of the graph that betweenness centrality can be parallelized on?

## **References:**

- [1] L. C. Freeman, "A Set of Measures of Centrality Based on Betweenness," Sociometry, vol. 40, no.
   1, pp. 35–41, 1977, doi: 10.2307/3033543.
- [2] M. Grandjean, English: Graph representing the metadata of thousands of archive documents, documenting the social network of hundreds of League of Nations personals. 2013.
- [3] U. Brandes, "A faster algorithm for betweenness centrality," *The Journal of Mathematical Sociology*, vol. 25, no. 2, pp. 163–177, Jun. 2001, doi: 10.1080/0022250X.2001.9990249.
- [4] K. Shukla, S. C. Regunta, S. H. Tondomker, and K. Kothapalli, "Efficient parallel algorithms for betweenness- and closeness-centrality in dynamic graphs," in *Proceedings of the 34th ACM International Conference on Supercomputing*, New York, NY, USA, Jun. 2020, pp. 1–12, doi: 10.1145/3392717.3392743.
- [5] F. Jamour, S. Skiadopoulos, and P. Kalnis, "Parallel Algorithm for Incremental Betweenness Centrality on Large Graphs," *IEEE Transactions on Parallel and Distributed Systems*, vol. 29, no. 3, pp. 659–672, Mar. 2018, doi: 10.1109/TPDS.2017.2763951.