

# Distributed Graph Processing

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# Outlook: Beyond Relational Data

- **Graph data**
- Data streams
- Spatial data

# Reading List

- "*Pregel: a system for large-scale graph processing*", SIGMOD 2010, G. Malewicz et al. [Google]
- "*One trillion edges: graph processing at Facebook-scale*", VLDB 2015, A. Ching et al. [Facebook]

# Motivation: Large Graphs

- Graphs may **exceed** resource limits of single machines
  - Graphs representing the entire **Web** (Google)
  - Graphs representing large **social networks** (FB)
  - ...
- This motivates graph processing in **clusters**

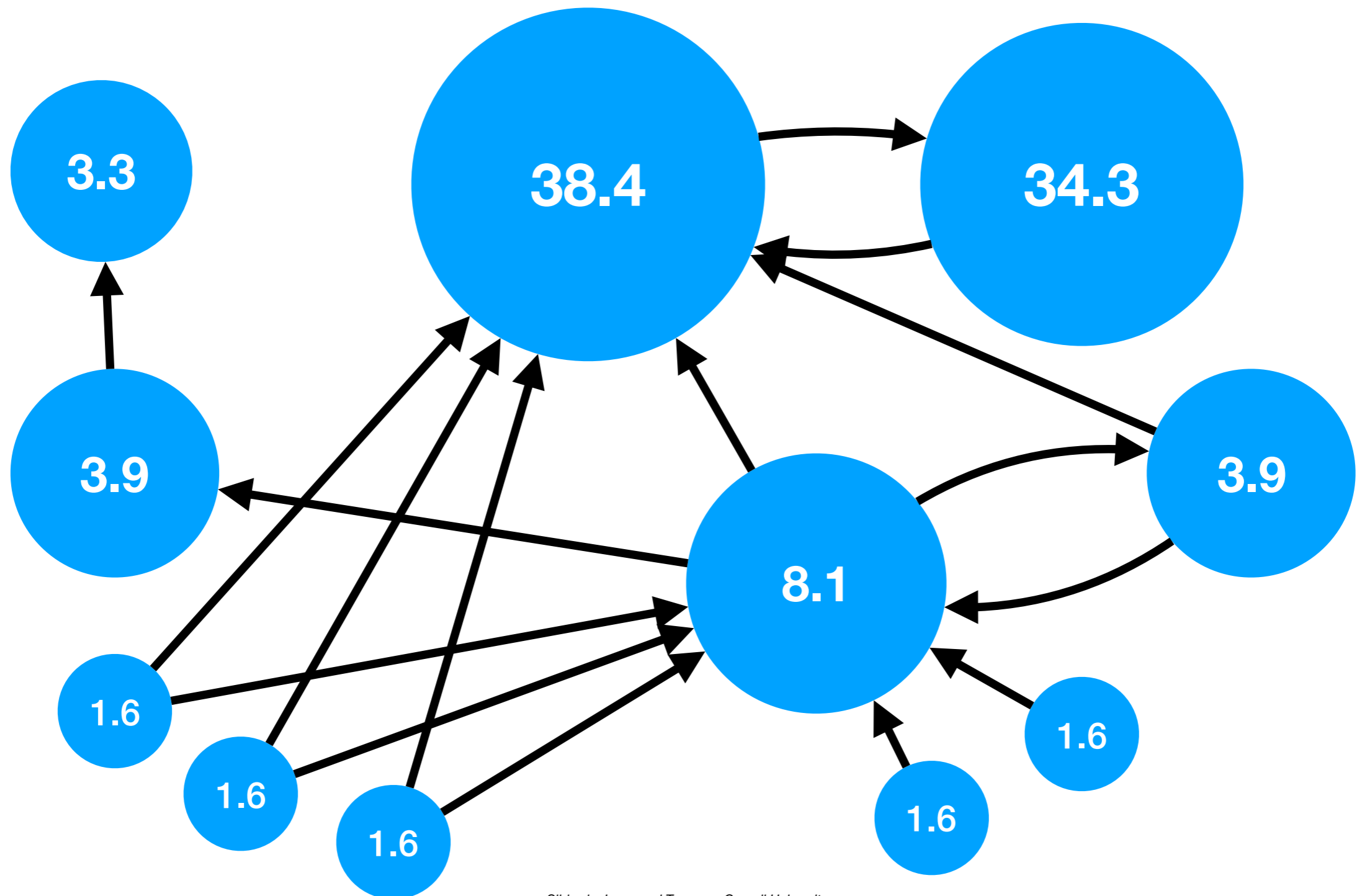
# Example: PageRank

- Google ranks search results via the **PageRank** algorithm
- Operates on a graph representation of the **Web**
  - **Nodes** represent Web sites
  - **Edges** represent links
- Pages with higher PageRank are **preferable**

# Random Surfer

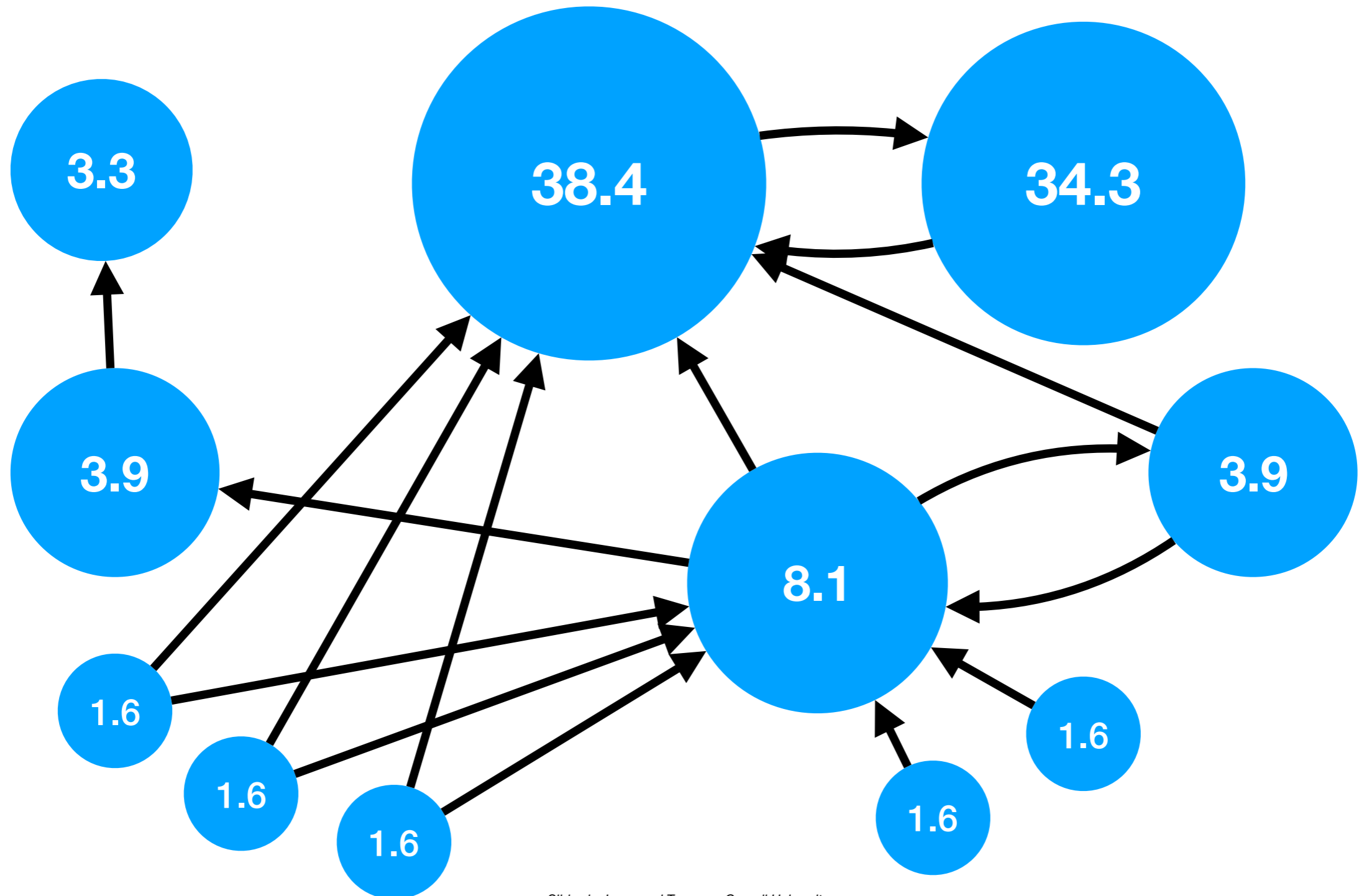
- PageRank is based on the **random surfer** model
- Random surfer **starts** from random Web site
- Randomly selects outgoing **links** to follow
  - May select **random** page with probability  $\alpha$
  - Selects random page if **no outgoing** links
- PageRank: **probability** to visit site at specific instant

# PageRank Example



# PageRank Example

*Do We Sometimes Select Random Pages?*



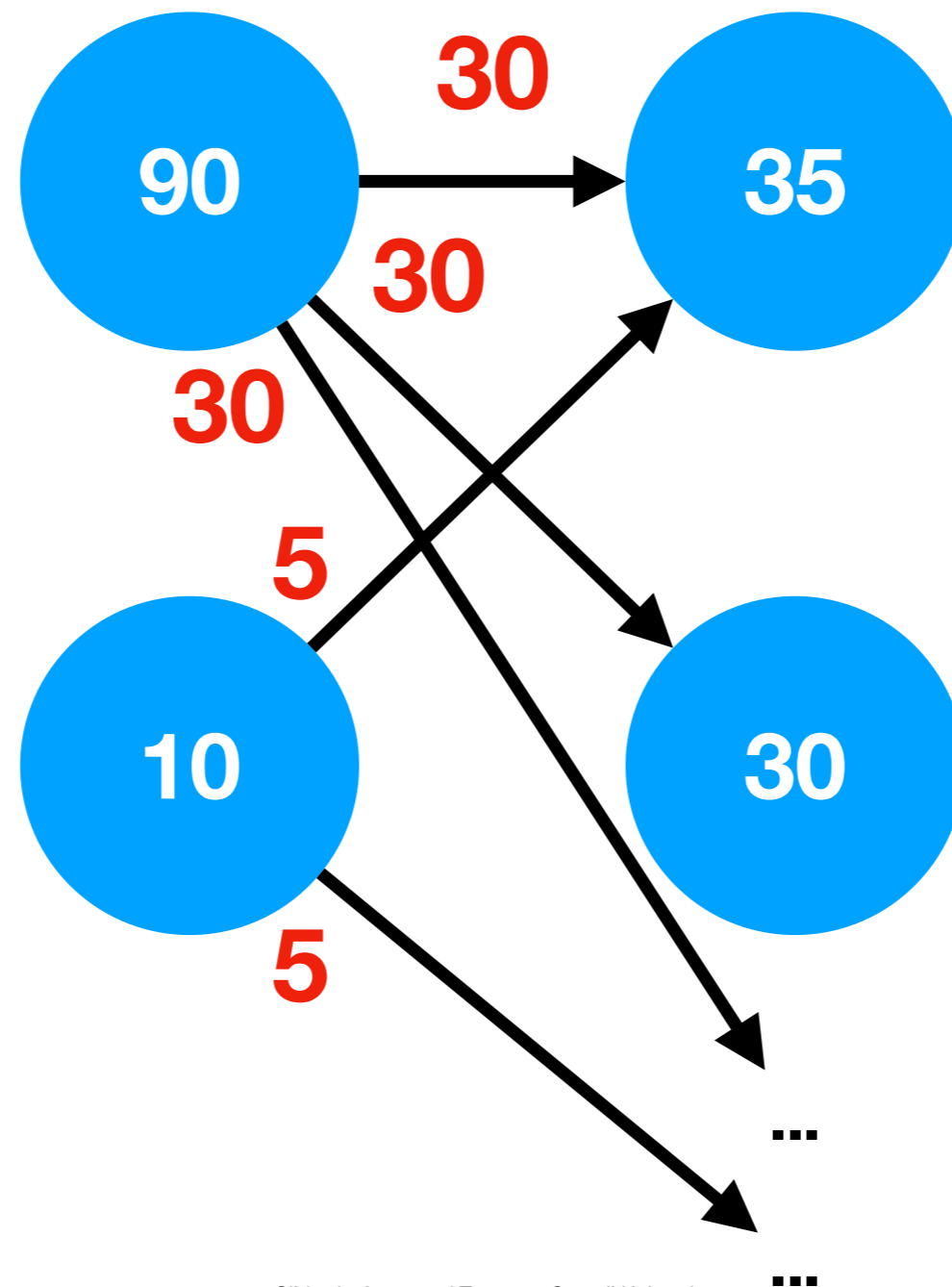


# Calculating PageRank

- We can calculate PageRank via an **iterative** algorithm
- We initialize each node's PageRank to  **$1/NrNodes$**
- In each iteration, we **redistribute** PageRank over links
  - Each node partitions PageRank among outgoing links
  - PageRank in next iteration is sum over incoming links

# PageRank Iterative Updates

*Iteration I    Iteration I+1*



# Pregel Overview

- **Pregel** is a system for distributed graph processing
- Proposed in 2010 (Google), **PageRank** is use case
- Pregel **distributes** graph partitions over cluster nodes
- Worker nodes process their partition in **parallel**

# Pregel Computation Model

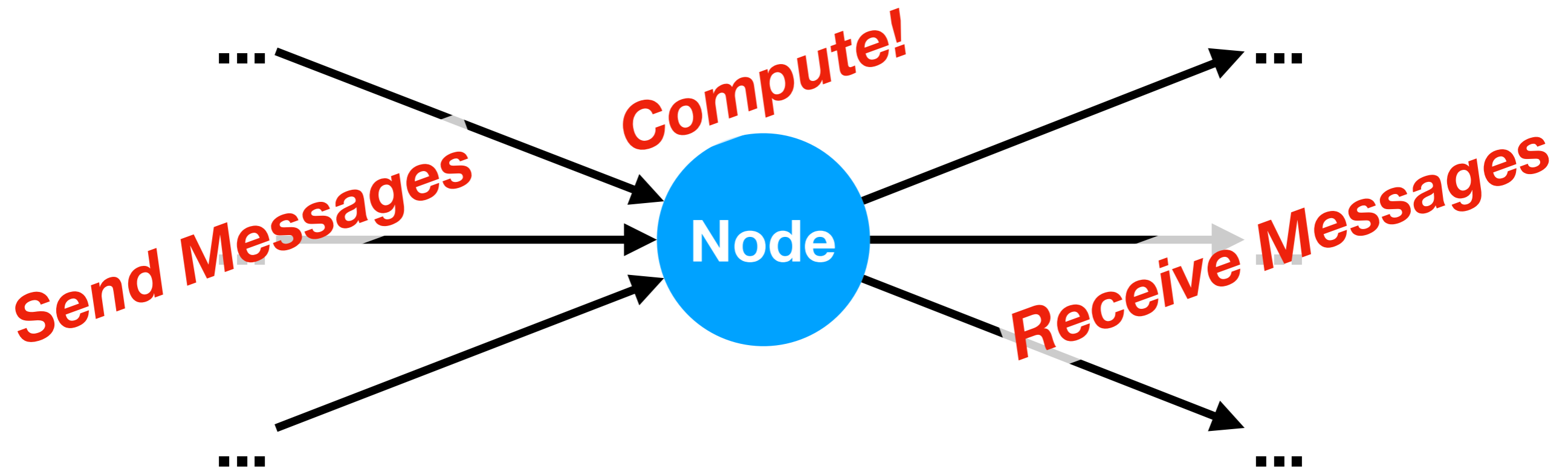
- Computation is divided into **iterations** ("supersteps")
- In each iteration, we invoke **Compute** for each node
  - Compute function can be **customized** by user
  - **Input**: messages sent to this vertex in prior iteration
  - Can **message** other nodes, delivered in next iteration
- Computation ends once all nodes vote to **halt**

# Illustration of Computation

*Iteration I-1*

*Iteration I*

*Iteration I+1*

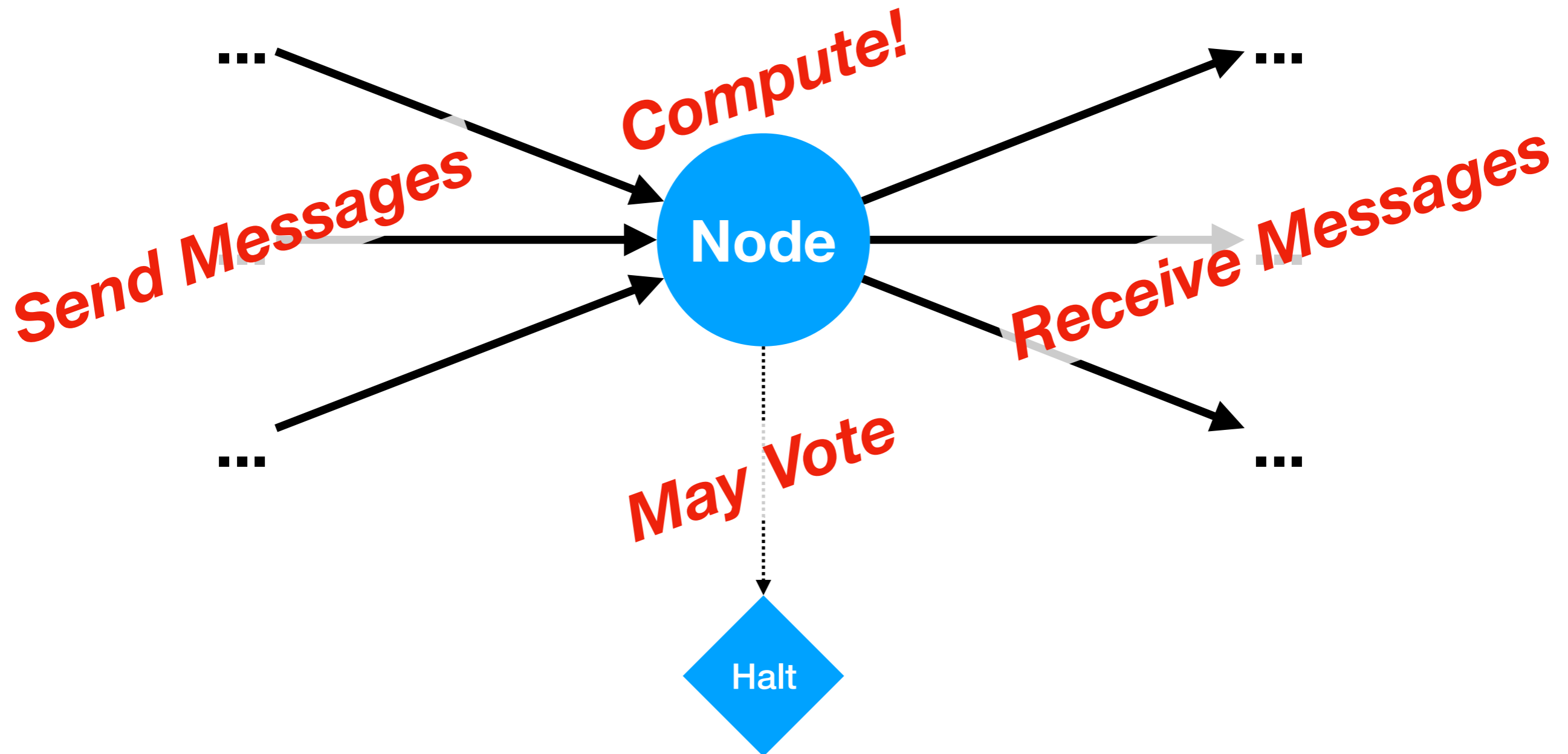


# Illustration of Computation

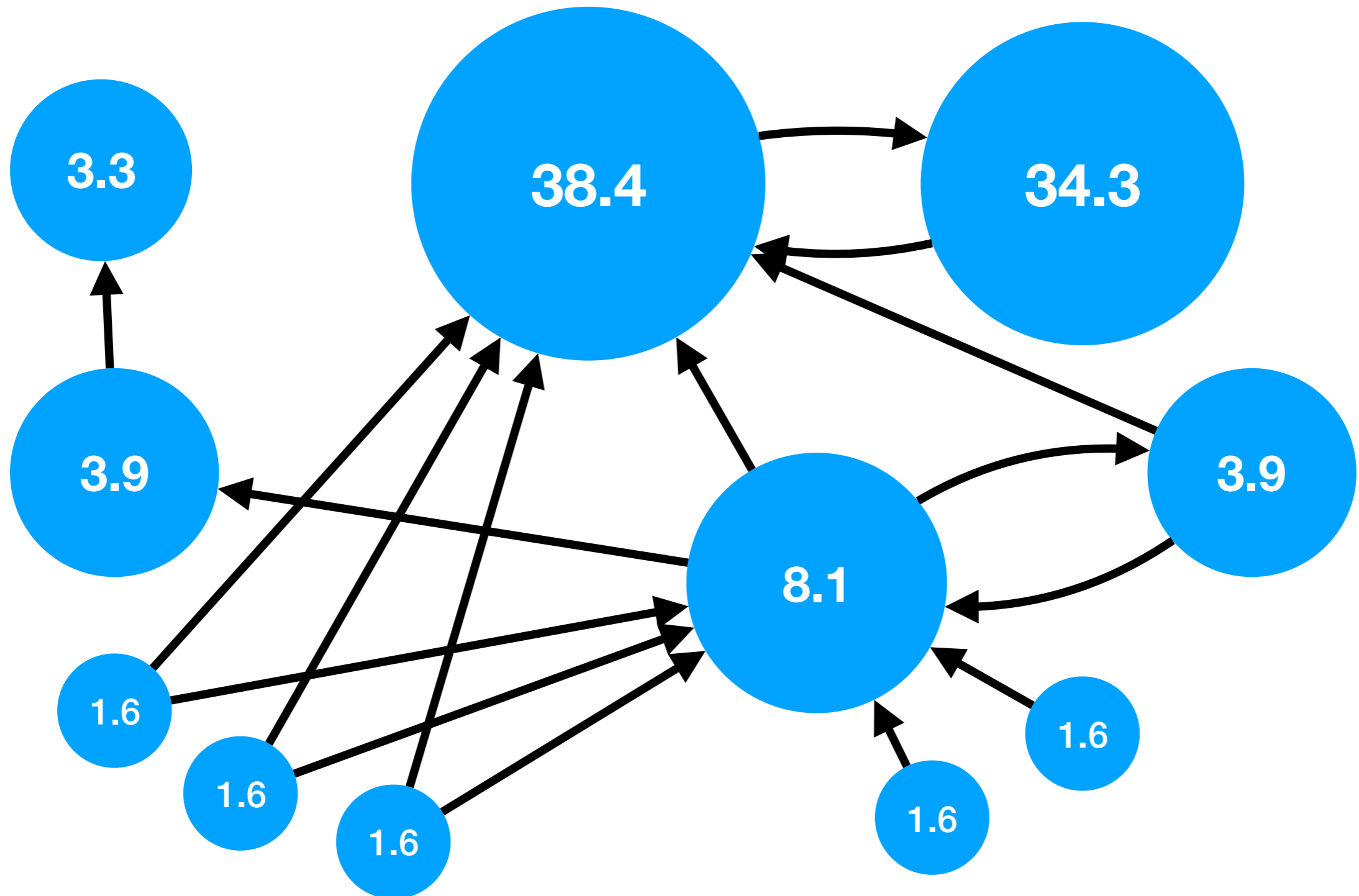
*Iteration I-1*

*Iteration I*

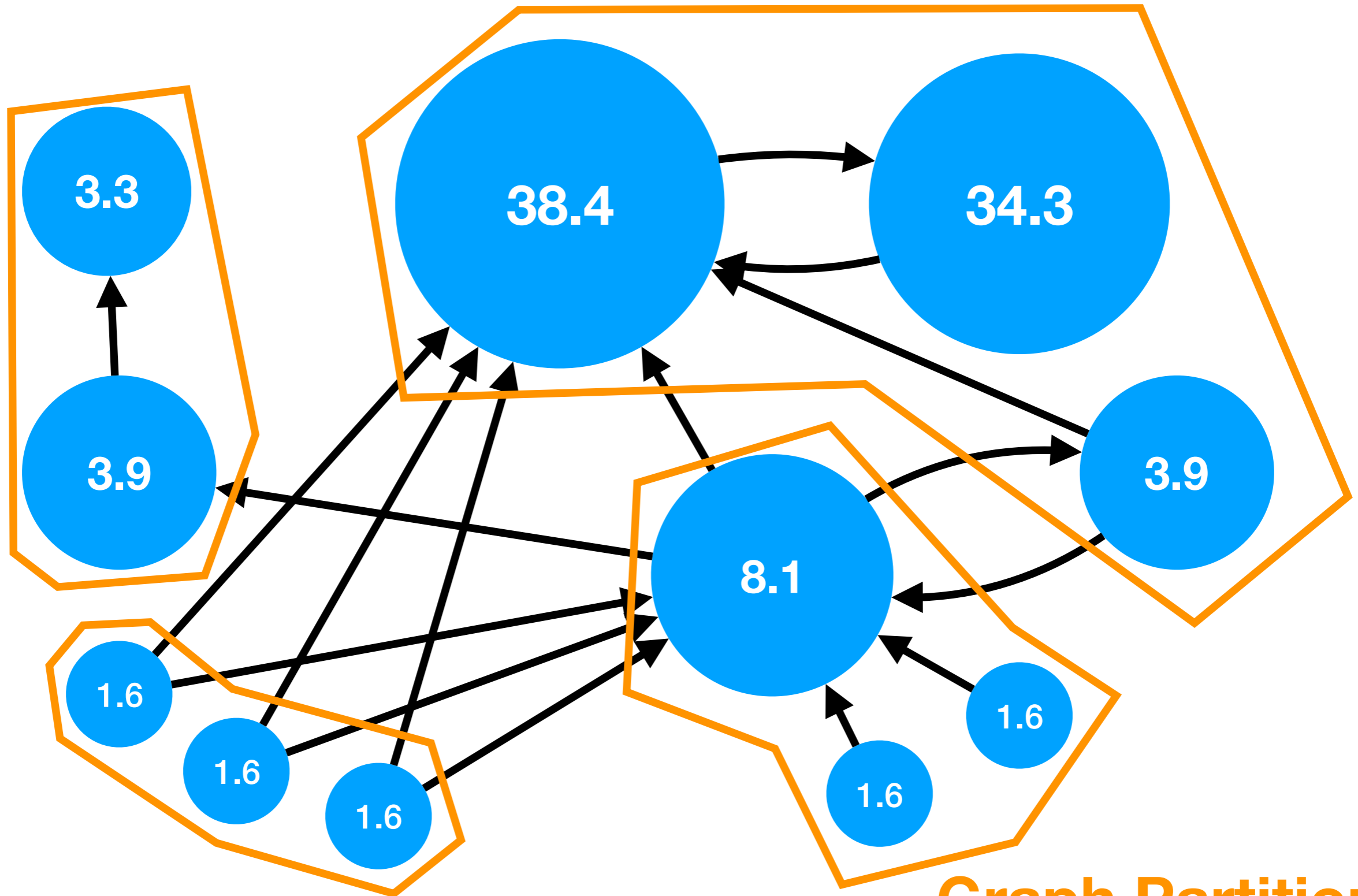
*Iteration I+1*



# Parallel Processing



# Parallel Processing



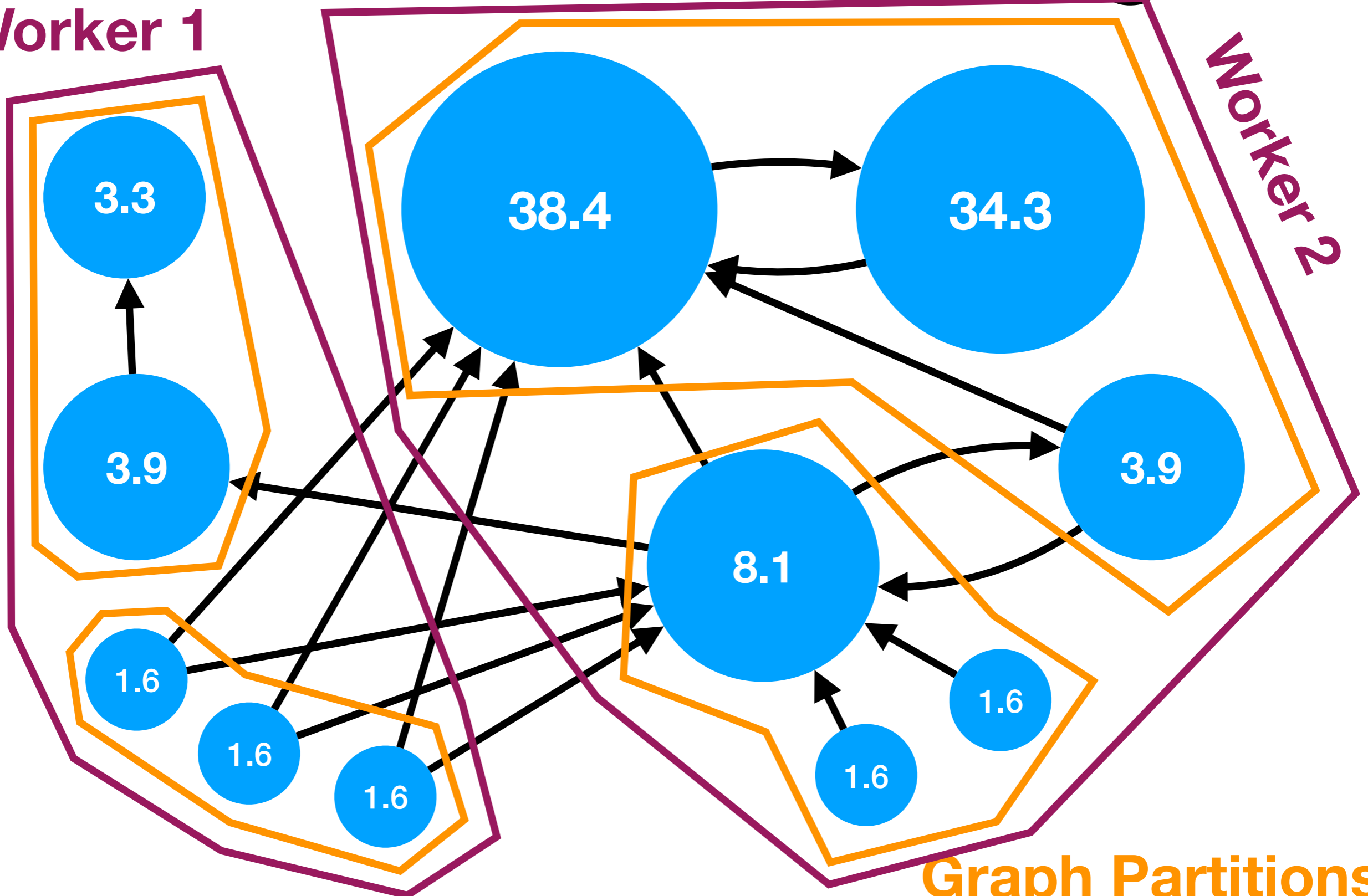
**Graph Partitions**



# Parallel Processing

Worker 1

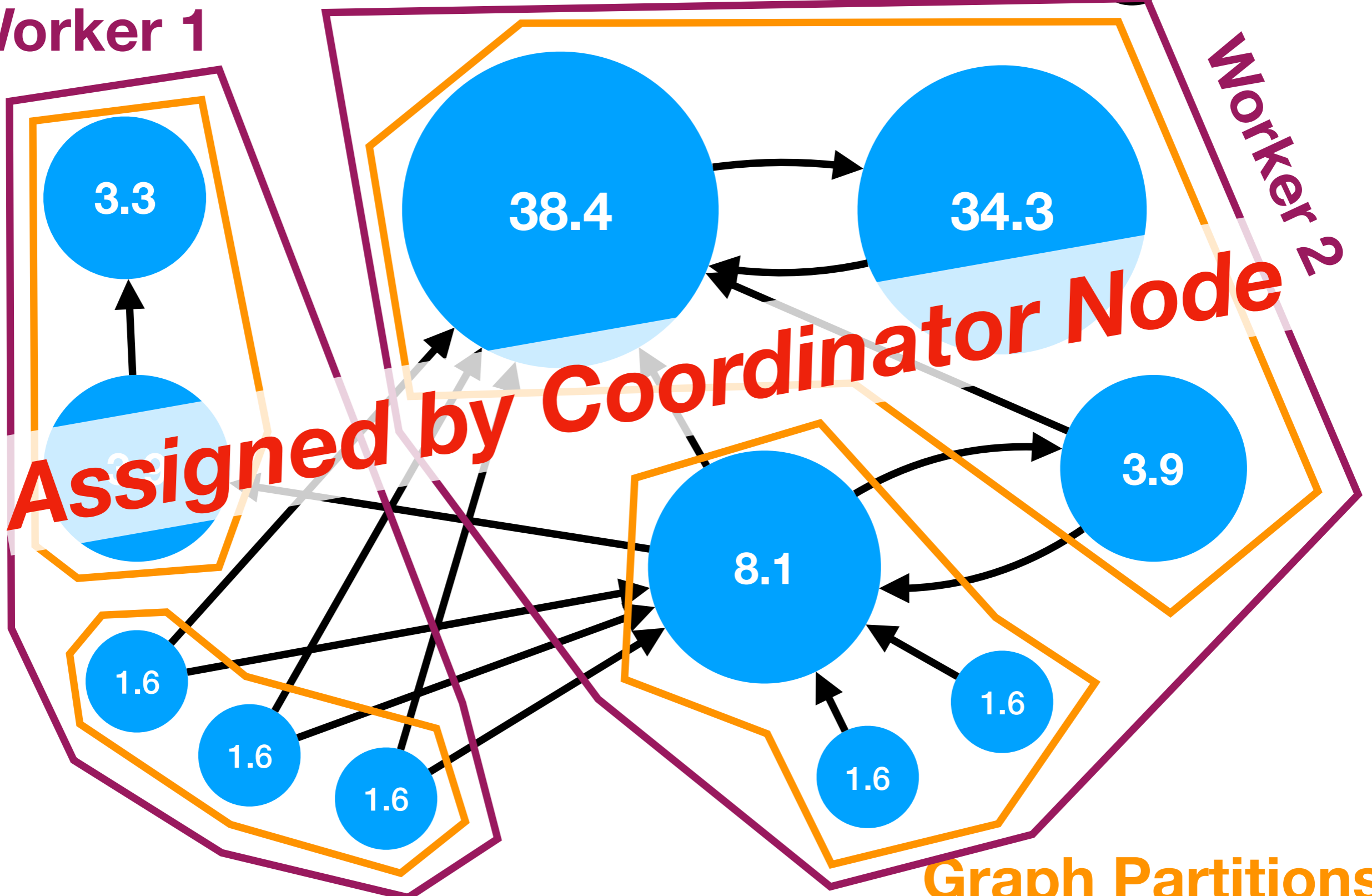
Worker 2



# Parallel Processing

Worker 1

Worker 2



# Fault Tolerance

- Workers **persist** input and state at iteration start
- Coordinator **detects** worker failures via pings
  - **Recovery** may start several supersteps earlier
  - **Re-partition** graph to replace failed workers
- "**Confined recovery**" restricted to failed partitions
  - Requires persisting outgoing messages as well

# PageRank in Pregel

Compute(**ReceivedPR** : int[]):

**NewPR** = sum(**ReceivedPR**)

For o in **OutgoingLinks**:

Send(o.target, **NewPR**/|**OutgoingLinks**|)

*(Extensions required for random jumps and handling "dead ends")*

# Better Performance with Combiners

- Basic version **sends** lots of page rank values
- Can **aggregate** messages via custom "Combiners"
- Here: can combine page rank for same target as **sum**