## Topology-Aware Node Ranking -Motivational Example



## Topology-Aware Node Ranking -Basic Idea



### PageRank:

The importance of a web page is determined not only by its own contents but also its neighbors'

Observation:

The importance of a substrate node is determined not only by its own resource but also its neighbors'

## Topology-Aware Node Ranking -Details

- A node has a higher rank if it has more highly-ranked neighbors
- Combined Node resources
- Initial node rank  $NR^{(0)}(u) = \frac{H(u)}{\sum_{v \in V} H(v)}$

$$H(u) = CPU(u) \sum_{l \in L(u)} BW(l),$$
$$= \frac{H(u)}{D}$$

- Jumping probability  $p_{uv}^J = \frac{H(v)}{\sum\limits_{w \in V} H(w)}$
- Forwarding probability  $p_{uv}^F = \frac{H(v)}{\sum H(w)}$

$$NR^{(t+1)}(v) = \sum_{u \in V} p_{uv}^J \cdot p_u^J \cdot NR^{(t)}(u) + \sum_{u \in nbr_1(v)} p_{uv}^F \cdot p_u^F \cdot NR^{(t)}(u)$$

## Topology-Aware Node Ranking -Details

• network of n nodes  $V = \{v_1, v_2, \cdots, v_n\}$   $NR_i^{(t)} = NR^{(t)}(v_i)$ 

 $NR^{(t+1)} = \mathbf{T} \cdot NR^{(t)}$ 

• where T is a one step transition matrix of the

Markov chain

$$\mathbf{T} = \begin{pmatrix} p_{11}^{J} & p_{12}^{J} & \cdots & p_{1n}^{J} \\ p_{21}^{J} & p_{22}^{J} & \cdots & p_{2n}^{J} \\ \vdots & \vdots & \ddots & \vdots \\ p_{n1}^{J} & p_{n2}^{J} & \cdots & p_{nn}^{J} \end{pmatrix} \cdot \begin{pmatrix} p_{1}^{J} & 0 & \cdots & 0 \\ 0 & p_{2}^{J} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & p_{n}^{J} \end{pmatrix} + \\ \begin{pmatrix} 0 & p_{12}^{F} & \cdots & p_{1n}^{F} \\ p_{21}^{F} & 0 & \cdots & p_{2n}^{F} \\ \vdots & \vdots & \ddots & \vdots \\ p_{n1}^{F} & p_{n2}^{F} & \cdots & 0 \end{pmatrix} \cdot \begin{pmatrix} p_{1}^{F} & 0 & \cdots & 0 \\ 0 & p_{2}^{F} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & p_{n}^{F} \end{pmatrix}$$

## Topology-Aware Node Ranking -Details

• network of n nodes  $V = \{v_1, v_2, \cdots, v_n\}$   $NR_i^{(t)} = NR^{(t)}(v_i)$ 

 $NR^{(t+1)} = \mathbf{T} \cdot NR^{(t)}$ 

where T is a one step transition matrix of the Markov chain

Algorithm 1 The NodeRank Computing Method

$$\begin{array}{c} 1: \text{ Given a positive value } \epsilon, i \leftarrow 0 \\ 2: \text{ repeat} \\ 3: & NR^{(i+1)} \leftarrow \mathbf{T} \cdot NR^{(i)} \\ 4: & \delta \leftarrow \|NR^{(i+1)} - NR^{(i)}\| \\ 5: & i++ \\ 6: \text{ until } \delta < \epsilon \end{array} \qquad \mathbf{T} = \begin{pmatrix} p_{11}^J & p_{12}^J & \cdots & p_{1n}^J \\ p_{21}^J & p_{22}^J & \cdots & p_{2n}^J \\ \vdots & \vdots & \ddots & \vdots \\ p_{n1}^J & p_{n2}^J & \cdots & p_{nn}^J \end{pmatrix} \cdot \begin{pmatrix} p_1^J & 0 & \cdots & 0 \\ 0 & p_2^J & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & p_n^J \end{pmatrix} + \\ \begin{pmatrix} 0 & p_{12}^F & \cdots & p_{1n}^F \\ p_{21}^F & 0 & \cdots & p_{2n}^F \\ \vdots & \vdots & \ddots & \vdots \\ p_{n1}^F & p_{n2}^F & \cdots & 0 \end{pmatrix} \cdot \begin{pmatrix} p_1^F & 0 & \cdots & 0 \\ 0 & p_2^F & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & p_n^F \end{pmatrix} \end{array}$$

# Mapping

- Phase 1: node-to-node
  - Sort VN nodes according to their CPU requirements

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- Sort SN nodes according to their MCRank
- Maximum first matching
- Phase 2: link-to-link
  - shortest path
    - y-z: G-H-D ?
      - G-F-E-D ?
  - k-shortest path
    - multiple edges



## Mapping RW-MaxMatch

Algorithm 2 RW-MaxMatch Node Mappin	ng Stage
1: Compute the NodeRank values of all no	des in both $G_s$
and $G_v$ using Algorithm 1 with a given v	value of $\epsilon$ .
2: Sort the nodes in $G_s$ according to their No	odeRank values
in non-increasing order.	
3: Sort the nodes in $G_v$ according to their Net	odeRank values
in non-increasing order.	
4: Map virtual nodes to substrate nodes	using the L2S2
mapping procedure.	
5: if node resource constraints satisfied the	en
6: return NODE_MAPPING_SUCCESS	
7: else	
8: return NODE_MAPPING_FAILED	Algorithm 3 RW-MaxMatch Link Mapping Stage
9: end if	1: if path unsplittable then
	- 2: Map virtual links using the <i>k</i> -shortest path algorithm.
	3: else
	4: Map virtual links using the multi-commodity flow al-
	gorithm.
	5: end if
	6: if link resource constraints satisfied then
	7: return LINK_MAPPING_SUCCESS
	8: else
	9: return LINK_MAPPING_FAILED
	10: end if

## Mapping RW-BFS

#### Algorithm 4 RW-BFS

- 1: Compute the NodeRank values of all nodes in both  $G_s$ and  $G_v$  using Algorithm 1 with a given value of  $\epsilon$ .
- 2: Construct the breadth-first searching tree of  $G_v$ .
- 3: Sort all the nodes in  $G_v$  in non-increasing order in each level of the breadth-first tree according to their NodeR-ank values.
- 4: backtrack\_count = 0
- 5: Select a positive integer  $\Delta$ .
- 6: for each node  $N_v^i$  in  $G_v$  do
- 7: Build the candidate substrate node list for
- 8: if  $Match(N_v^i) = 1$  then
- 9:  $\operatorname{Match}(N_v^{i+1})$
- 10: else
- 11: if backtrack\_count  $\leq \Delta$  then
- 12:  $\operatorname{Match}(N_v^{i-1})$
- 13: backtrack\_count++
- 14: else
- 15: return BFS\_FAILED
- 16: end if
- 17: end if
- 18: end for
- 19: return BFS\_SUCCESS

### Algorithm 5 The Details of Match Function

- 1: if  $N_v^i$  is the root then
- 2: map  $N_v^i$  onto the substrate node with the largest NodeRank
- 3: return MATCH\_SUCCESS
- 4: end if
- 5: k = 1
- 6: while  $k < Max\_Hop$  do
- 7: for each  $N_s^j$  which satisfies the k-hop constraint from its mapped parent node in the candidate substrate node list of  $N_v^i$  do
- 8: if  $pair(N_v^i, N_s^j)$  meets all the capacity constraints then
  - return MATCH\_SUCCESS
- 10: end if
- 11: end for
- 12: k++

9:

- 13: end while
- 14: return MATCH\_FAILED