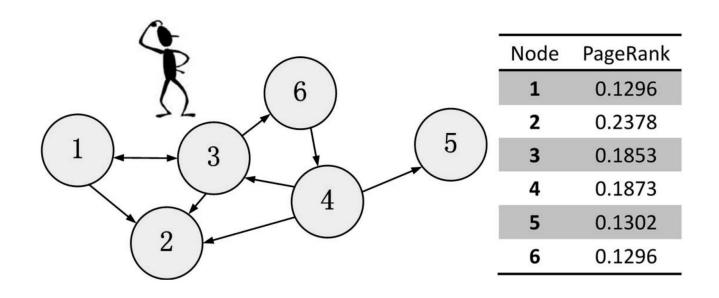
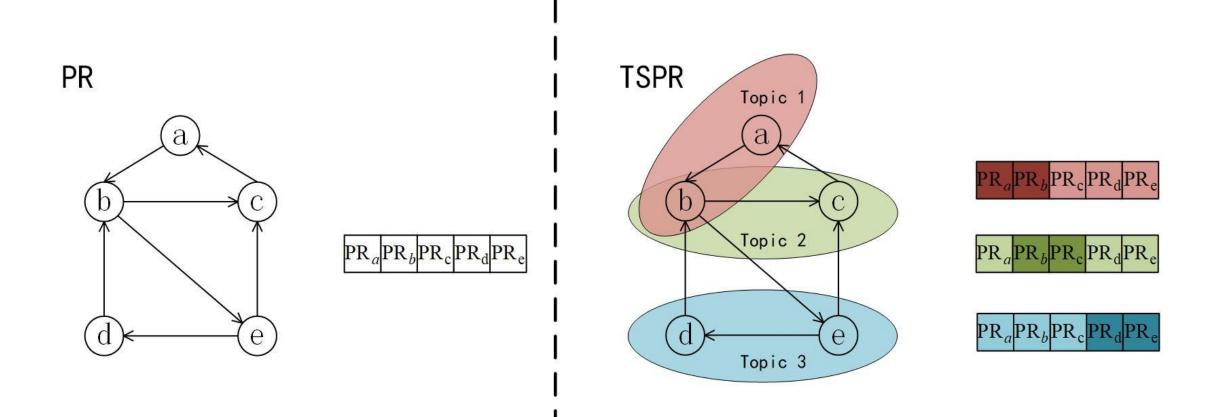
# Accelerating Topic-Sensitive PageRank By Exploiting the Query History

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## PageRank



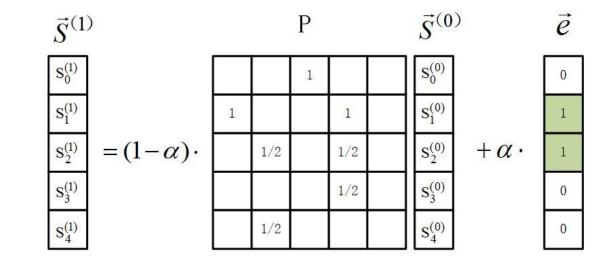
# **Topic Sensitive PageRank**



#### **Power Iteration**

- $\vec{S}$  : PageRank score vector
- P : transition matrix of graph
- $\vec{e}$  : indicator topic vector
- $\alpha$  : teleportation constant

first iteration:

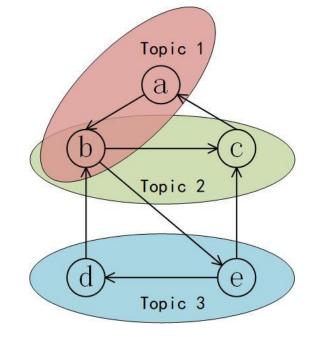


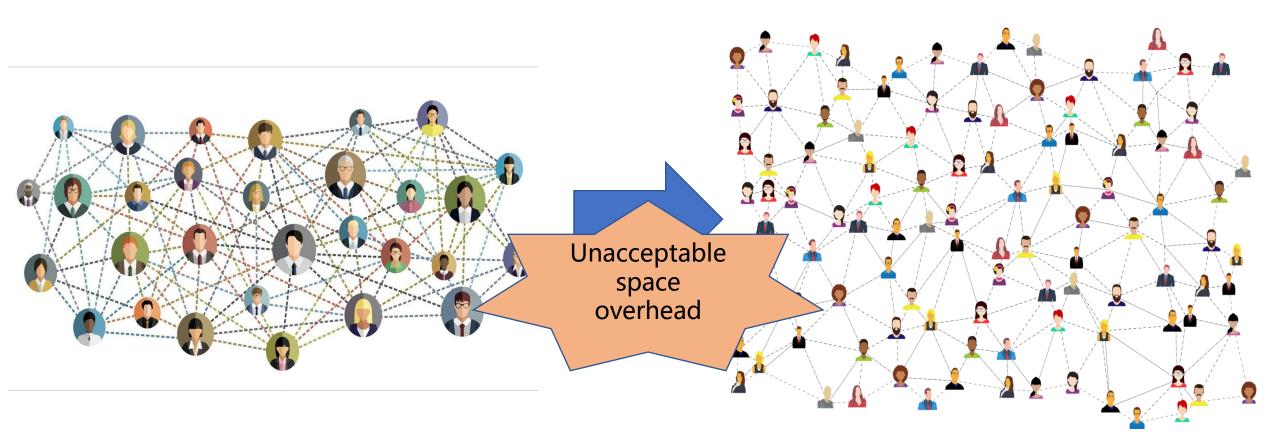
second iteration:

third iteration:

 $\vec{\mathbf{S}}^{(2)} = (1 - \alpha) \cdot P \cdot \vec{S}^{(1)} + \alpha \vec{e}$  $\vec{\mathbf{S}}^{(3)} = (1 - \alpha) \cdot P \cdot \vec{S}^{(2)} + \alpha \vec{e}$  $\cdots$ 

Stop when 
$$\left\|\mathbf{S}^{(t+1)} - \mathbf{S}^{(t)}\right\|_{1} < tolerance$$





- More topics  $\rightarrow$  more topic-specefic PR vector
- Lager graph  $\rightarrow$  Lager PR vector

# **Online Computation**

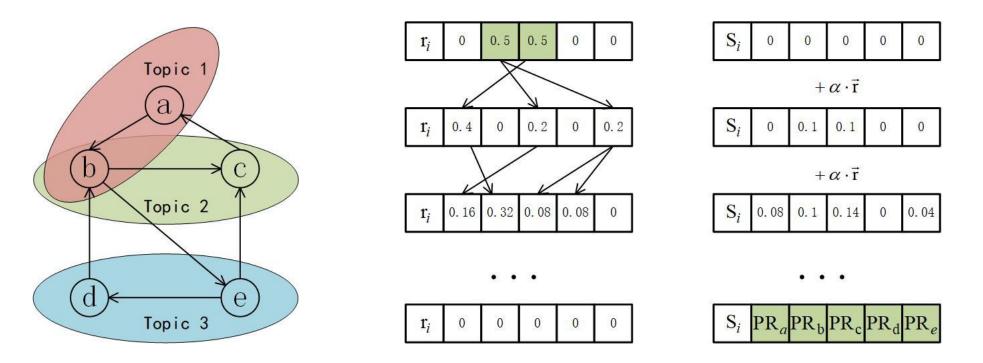
- How to accelerate iterative computation?
  - Approximation: Can't be used in high-precision situations
  - Index: Can't deal with dynamic graph

### **Forward Push**

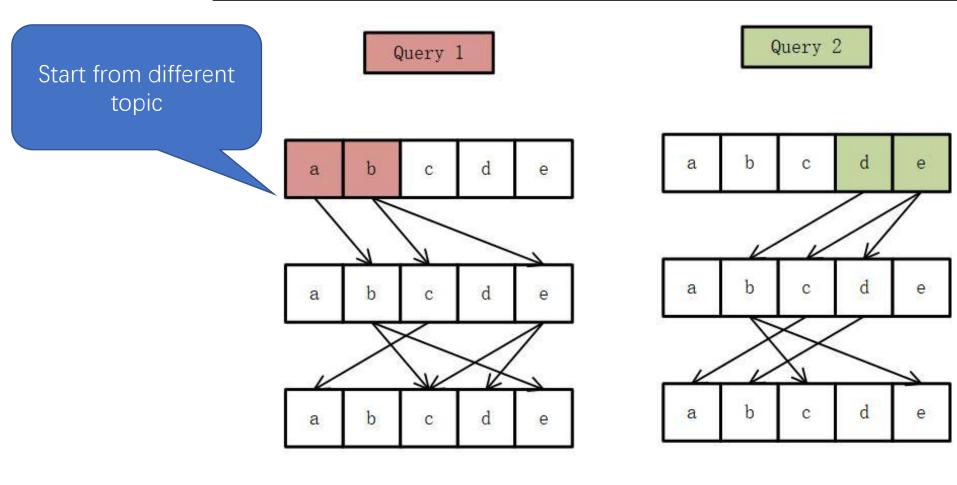
for each vertex :

- 1. reserve  $1 \alpha$  proportion of residual it has received;
- 2. evenly push the remaining  $\alpha$  proportion to its target vertices.

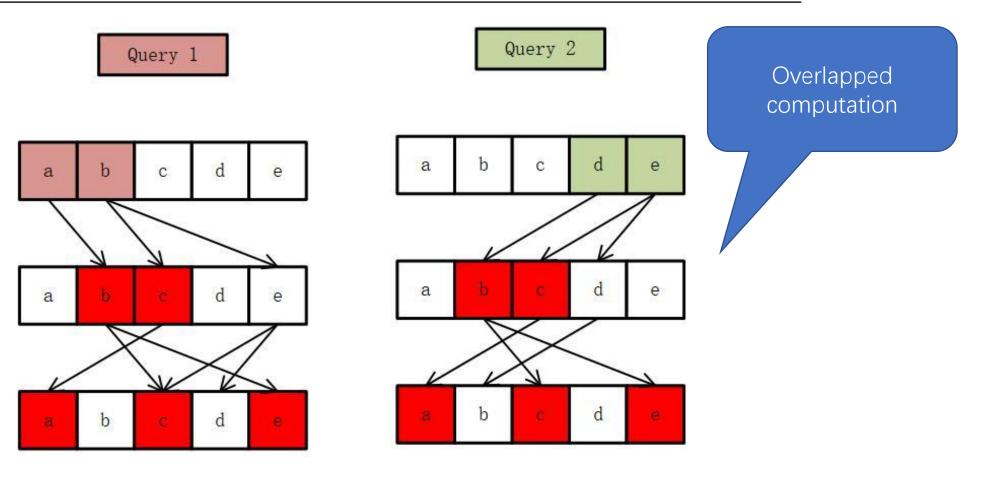
until: no vertex holds valid pushing mass



## **Overlapped Computation**

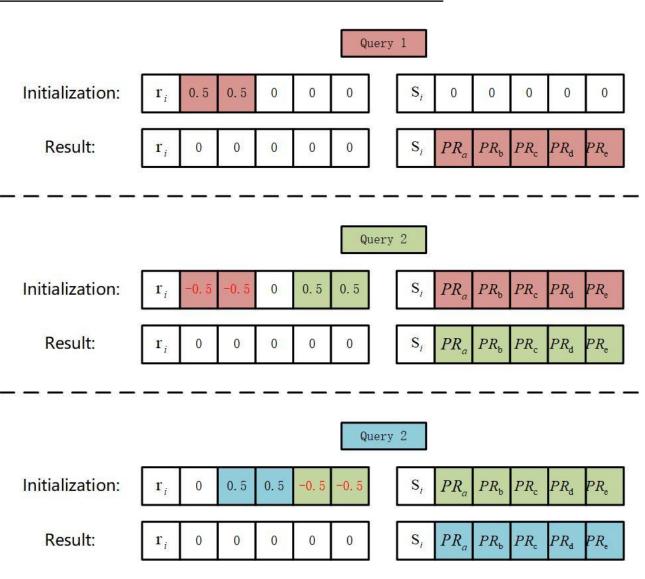


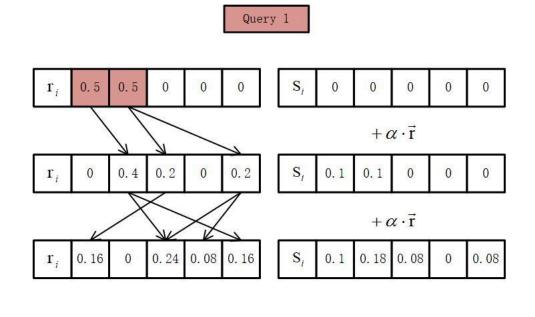
# **Overlapped Computation**



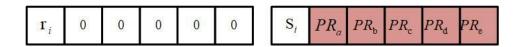
**Core idea**: accelerate the current computation by utilzing the results of the previous query

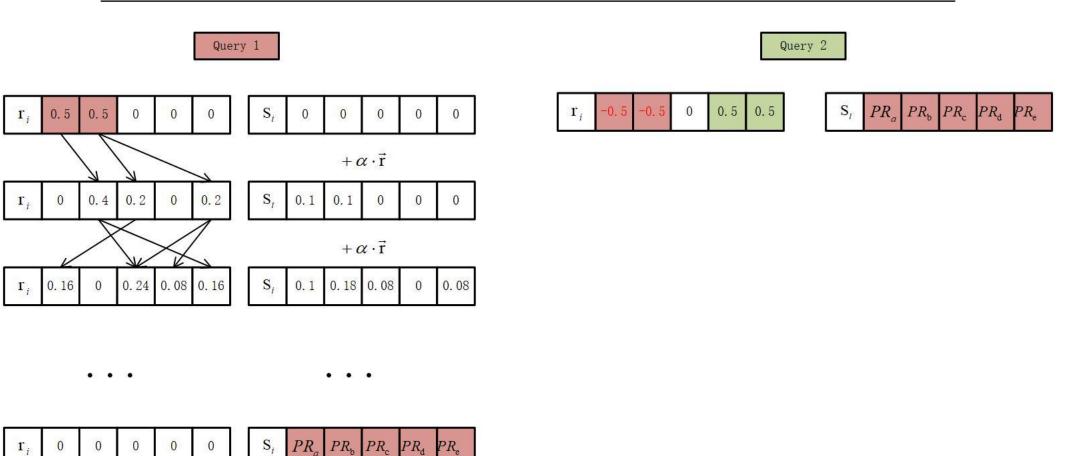
- First query: execute the standard forward push method Subsequent query:
- 1. initialize residual vector: set the current query's topic to a positive value and the previous query's topic to a negative value.
- 2. initialize socre vector: set score vector to the result from the previous query
- 3. perform forward push opearation

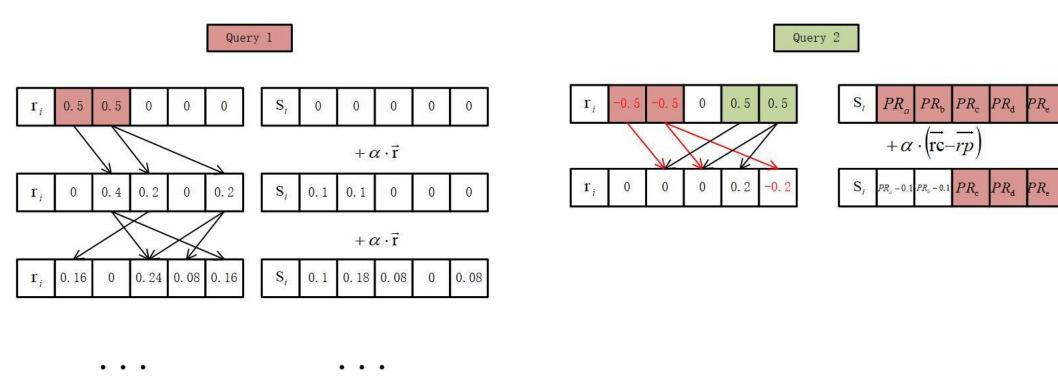




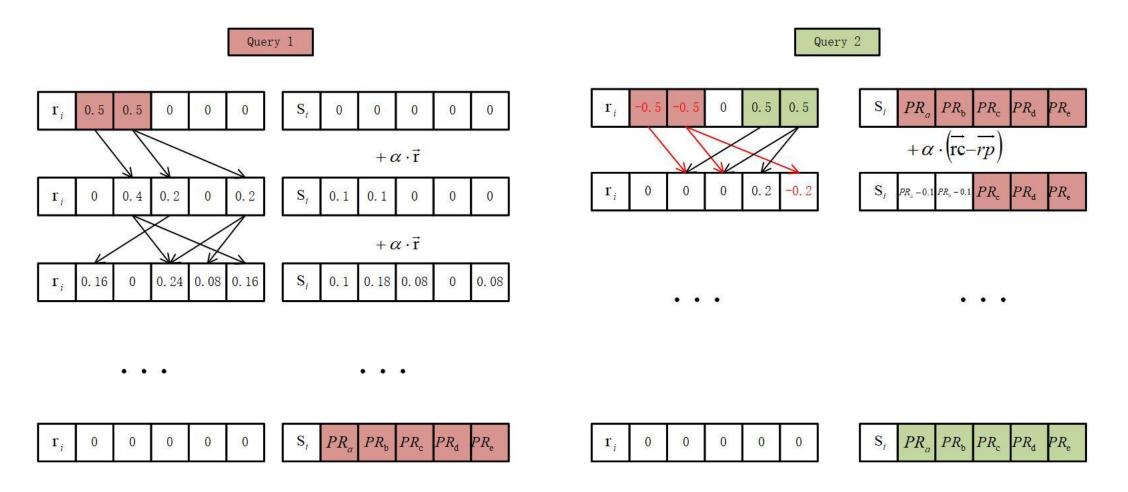
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 $\mathbf{r}_i = 0 = 0 = 0 = 0$   $\mathbf{S}_i = PR_{\alpha} = PR_{b} = PR_{c} = PR_{d} = PR_{e}$ 



#### **Experiments**

#### • Competitors

Traditional Power Iteration Forward Push, hybrid method PowerPush

#### • Environment

Linux server, 32 GB RAM, Ubuntu 22.04 (64-bit), GCC 11.3

• Datasets

Graph	# of vertex	# of edge
Catster (CT)	149684	10896394
DBLP (DL)	317080	2099732
Google (GL)	916428	12156500
Wiki-Talk (WT)	2394385	8505513
cit-Patents (CP)	3774768	16518948

#### **Overall Performance**

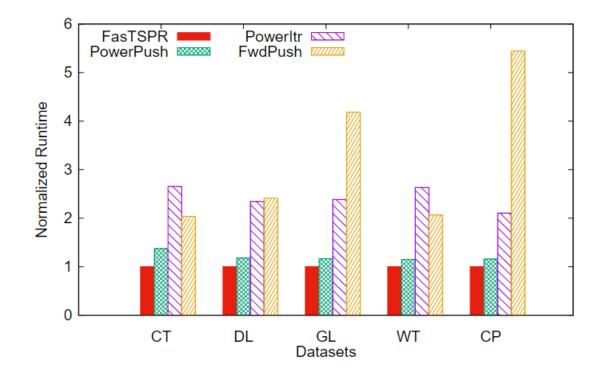


Fig. 1: The comparison of runtime

Achieved a 1.37× - 5.44× speedup

#### **Convergence Speed**

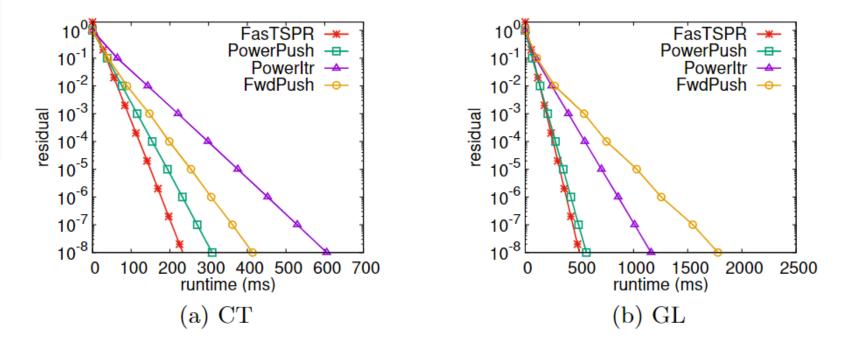


Fig. 2: The comparison of convergence speed

FasTSPR converges the fastest

#### **Impact of Different Queries**

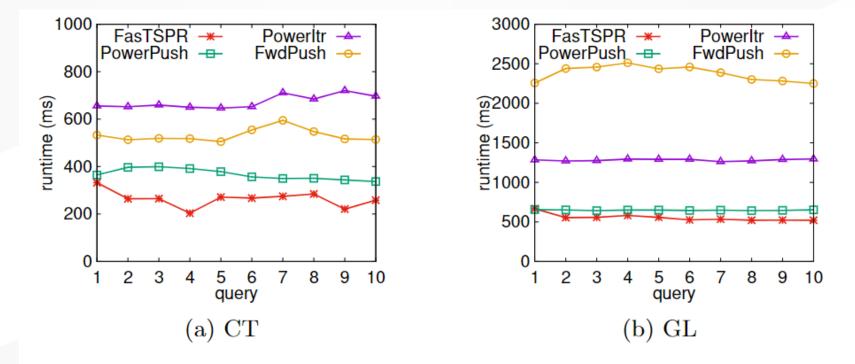


Fig. 3: The runtime when varying the topics

### Impact of topic size

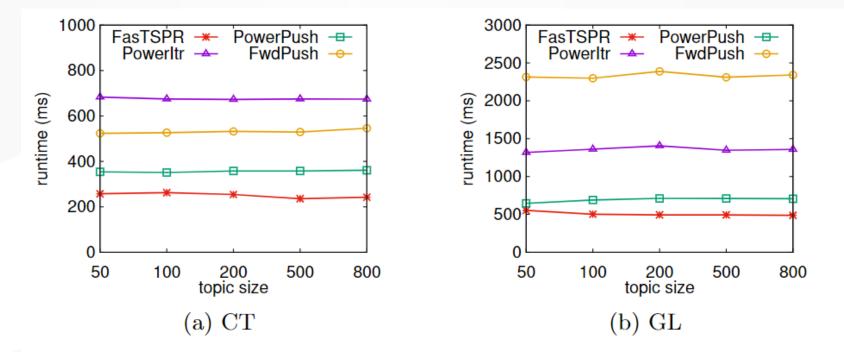


Fig. 4: The runtime when varying the size of topics

FasTSPR outperforms other algorithm in all query size

### Conclusion

- We propose an efficient TSPR algorithm, FasTSPR
- We provide a formal proof to prove the correctness of

FasTSPR

#### Thank you for listening!