

# Walk-classes, Centrality Collisions, and Spider Donuts

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# Preliminaries

Central theme:

**How do **walk structure** and **centrality behavior** relate?**

# Preliminaries

walk structure

centrality behavior

$$[e^{\beta \mathbf{A}}]_{jj}$$

Subgraph  
centrality

$$[(\mathbf{I} - \beta \mathbf{P})^{-1} \mathbf{e}_S]_j$$

Personalized  
PageRank

$$[e^{\beta \mathbf{A}} \mathbf{1}]_j$$

Total  
communicability

# Preliminaries

## walk structure

Closed  $L$ -walks

$$\left[ \mathbf{A}^L \mathbf{e}_j \right]_j$$

Transition probabilities

$$\left[ \mathbf{P}^L \mathbf{e}_S \right]_j$$

All  $L$ -walks

$$\left[ \mathbf{A}^L \mathbf{1} \right]_j$$

## centrality behavior

$$\left[ e^{\beta \mathbf{A}} \right]_{jj}$$

Subgraph centrality

$$\left[ (\mathbf{I} - \beta \mathbf{P})^{-1} \mathbf{e}_S \right]_j$$

Personalized PageRank

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Subgraph  
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All  $L$ -walks

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**General**

$$\left[ M^l \mathbf{s}(j) \right]_j$$

$$\left[ f(\beta M) \mathbf{s}(j) \right]_j$$

# Preliminaries

How do **walk structure** and **centrality behavior** relate?

Fix a graph matrix

$M$

diagonalizable

a vector function

$s(\cdot)$

e.g.  $\mathbf{1}, \mathbf{e}_j, \mathbf{e}_S$

and suitable function

$f(\cdot)$

pos power series

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Fix a graph matrix	$M$	diagonalizable
a vector function	$s(\cdot)$	e.g. $\mathbf{1}, \mathbf{e}_j, \mathbf{e}_S$
and suitable function	$f(\cdot)$	pos power series

## Positive Power Series:

An analytic function with a power series with

all positive coefficients:  $(1 - \beta x)^{-1}, e^{\beta x}$

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**Walk centrality**

$$C_f(j, \beta) = [f(\beta M)\mathbf{s}(j)]_j$$

$C(\cdot, \beta)$  when context is clear



# Questions

How do **walk structure** and **centrality behavior** relate?

$$C_f(j, \beta) = [f(\beta \mathbf{M})\mathbf{s}(j)]_j$$

- when can  $i, j$  have same score?
  - what does it mean if they do?
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# Rank-trajectory plot

Each curve represents a node in Zachary's Karate Club

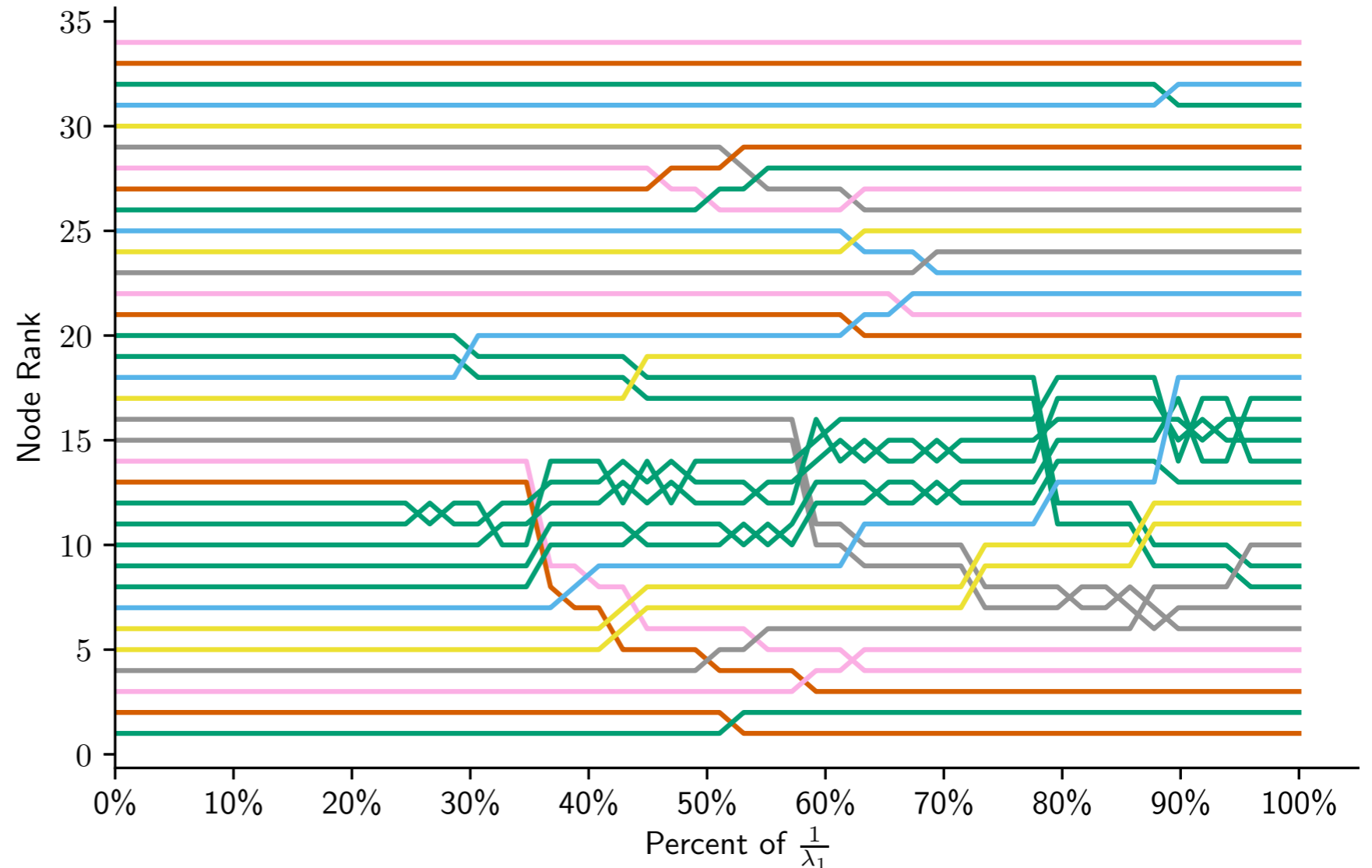
$x$  : Katz parameter

$$\beta \in (0, \frac{1}{\lambda_1})$$

$y$  :  $n$  - node's rank

according to

$$C_{Katz}(\cdot, \beta)$$



# Rank-trajectory plot

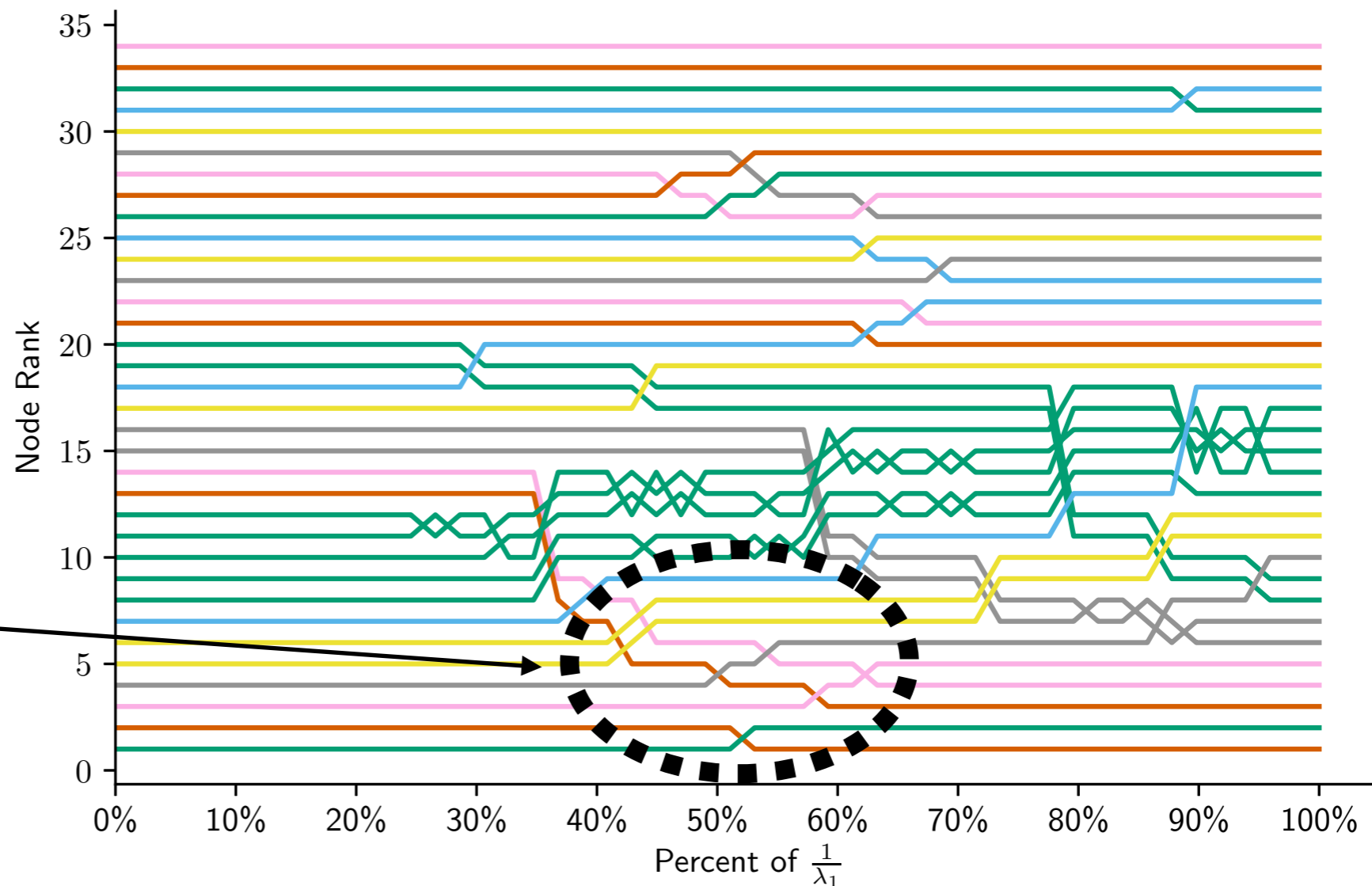
As  $\beta$  varies, node rankings vary. When nodes  $i, j$  swap rank,

by continuity,

$$C_f(j, \beta) = C_f(i, \beta)$$

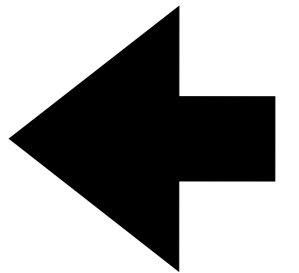
at some value  $\beta$ .

We call these  
**collisions.**



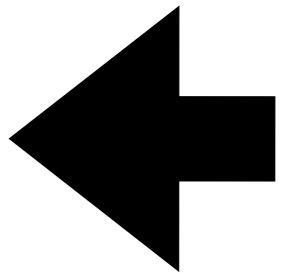
# Observations about questions...

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If  $i, j$  never collide, then  $C_f(j, \beta) > C_f(i, \beta)$  for all  $\beta$ ,  
i.e.,  $j$  is always ranked above  $i$ .

**Sufficient Condition:** If for  $\ell \in \mathbb{N}$   $[M^\ell \mathbf{s}(j)]_j > [M^\ell \mathbf{s}(i)]_i$

Then for all  $f$  and  $\beta$ ,  $j$  ranks above  $i$ . We say node  $j$   
**majorizes** node  $i$ .

# Nodes that majorize always rank higher

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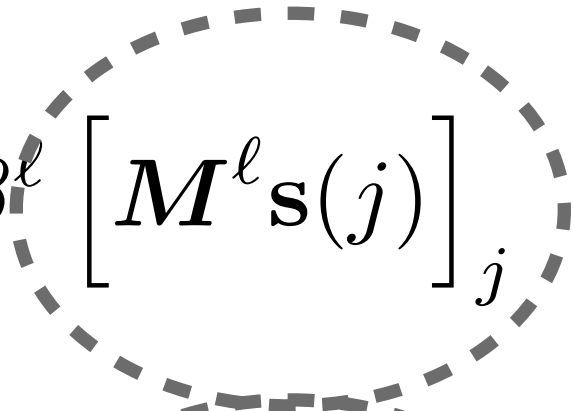
Proof sketch:

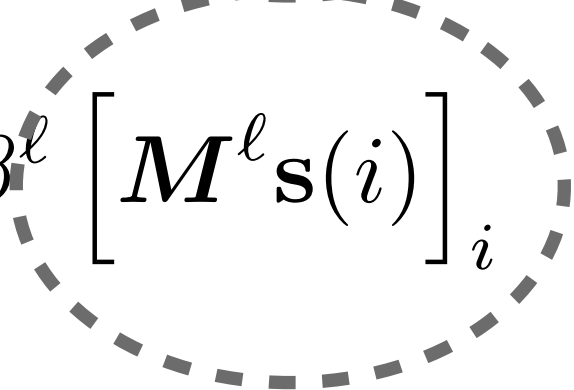
# Nodes that majorize always rank higher

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Proof sketch:

$$C_f(j, \beta) = \sum_{\ell=0}^{\infty} f_\ell \beta^\ell \left[ \mathbf{M}^\ell \mathbf{s}(j) \right]_j$$


$$C_f(i, \beta) = \sum_{\ell=0}^{\infty} f_\ell \beta^\ell \left[ \mathbf{M}^\ell \mathbf{s}(i) \right]_i$$


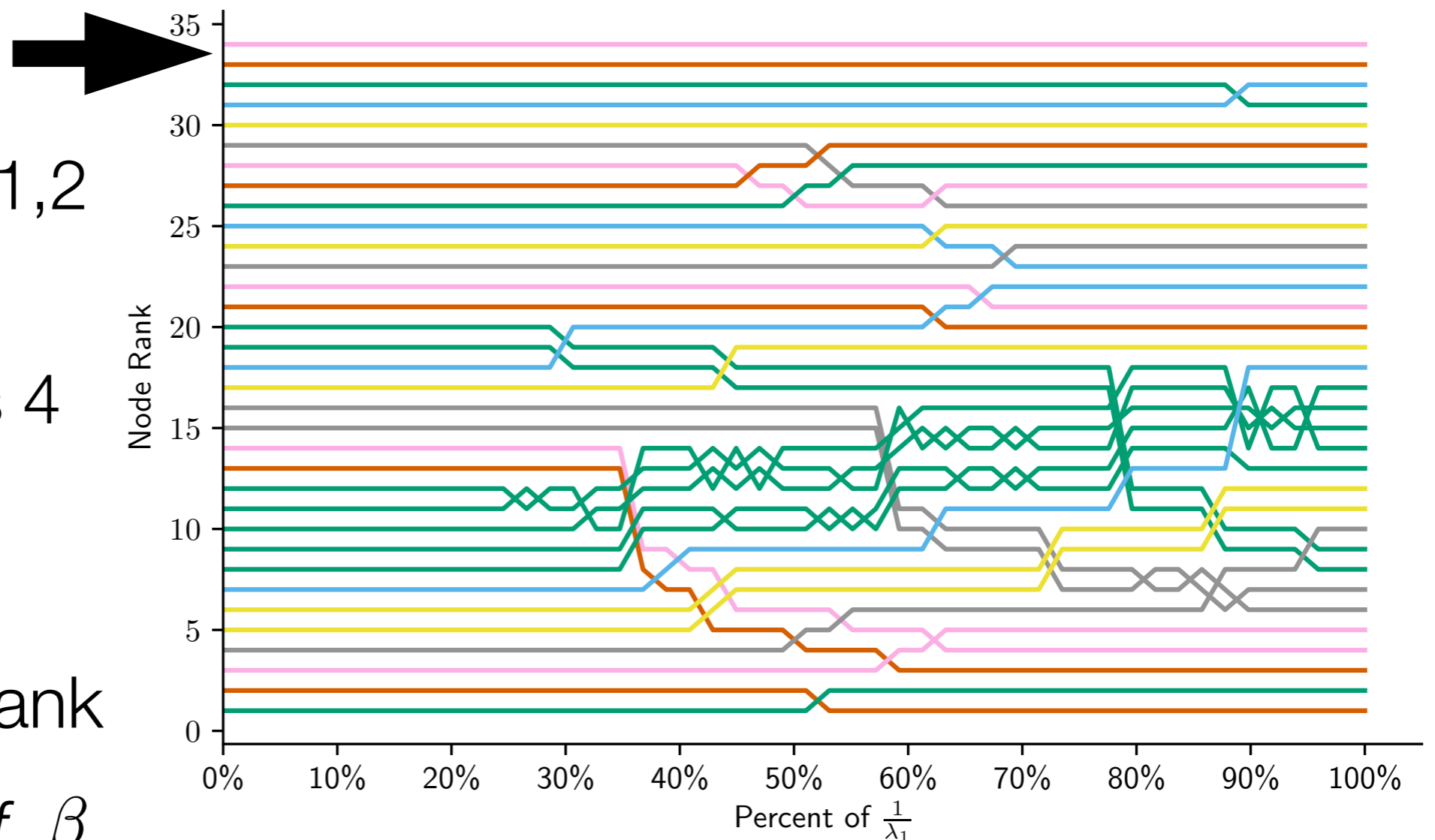


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Nodes ranked 1,2  
in Karate each  
majorize nodes 4  
through  $n$

These nodes rank  
in top 3 for all  $f, \beta$



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What if  $j$  doesn't have more  $k$ -walks for **all**  $k$ , just some?

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Recall: Katz domain is  $\left(0, \frac{1}{\lambda_1}\right)$

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**Corollary:** if  $d(j) > d(i)$  then  $j$  ranks above  $i$  for  $\beta$  near 0

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**Theorem:** [Klymko, Benzi 2015] “On the Limiting behavior...”

For  $f$ ,  $\mathbf{M}$ , nonnegative  $\mathbf{s}$ ,

$$C_f(\cdot, \beta) \rightarrow \begin{cases} \text{degree centrality, as } \beta \rightarrow 0^+ \\ \text{eigenvector centrality, as } \beta \rightarrow \beta_{max}^- \end{cases}$$

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Our bound makes explicit the rate of convergence here:

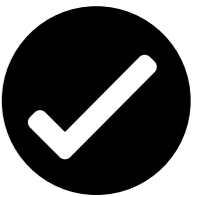
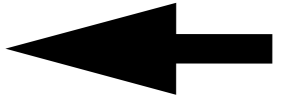
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# Number of collisions is bounded

**Theorem:** [Benzi, 2014]

If  $e_{jj}^{\beta A} = e_{ii}^{\beta A}$  for all  $\beta$  in a set with an accumulation point, then it holds for all  $\beta$ .

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Proof: We show  $C$  is analytic and use the Identity Theorem.

**Corollary:**

This proves that the number of distinct rankings  $C_f(\cdot, \cdot)$

Produces is finite!

# Number of collisions is bounded

**Theorem:** [Horton, K., Sullivan]

If  $f$  is the resolvent,  $(1 - \beta x)^{-1}$ , then  $i, j$  collide  $\leq m-1$  times, unless they collide for all  $\beta$ .

Degree of min poly



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**Corollary:** [Horton, K., Sullivan]

PageRank and Katz induce at most  $O(n^3)$  distinct rankings.

- Trivial upper bound is  $(n!)$
- Do other  $f$  induce more rankings? Is that better?
- Find a bound for  $f(x) = e^x$

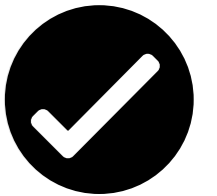


# Questions

How do **walk structure** and **centrality behavior** relate?

$$C_f(j, \beta) = [f(\beta \mathbf{M})\mathbf{s}(j)]_j$$

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**Nodes  $i, j$  have “too many” collisions *iff* same walk structure**

# Walk structure determines centrality behavior

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## Proposition:

degree of  
min poly of  $\mathbf{M}$

If for  $k = 0, \dots, m-1$

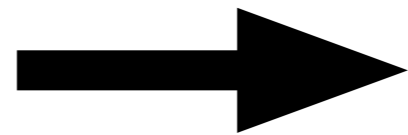
then for all  $f$  and  $\beta$

$$\left[ \mathbf{M}^k \mathbf{s}(j) \right]_j = \left[ \mathbf{M}^k \mathbf{s}(i) \right]_i$$

$$C_f(j, \beta) \equiv C_f(i, \beta)$$

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Identical **walk structure**...



Identical **walk centrality**

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# Walk-classes and iso-centrality

## Definition:

A **walk-class** = all nodes in  $G$  with identical “walk tuples”

$$\left( [M^0 \mathbf{s}(j)]_j, [M^1 \mathbf{s}(j)]_j, \dots, [M^{m-1} \mathbf{s}(j)]_j \right)$$

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**Theorem:** [Horton, K., Sullivan] The following are equivalent:

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3. Nodes  $i, j$  have identical **walk centrality:**

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We say nodes satisfying these properties are **isocentral**.

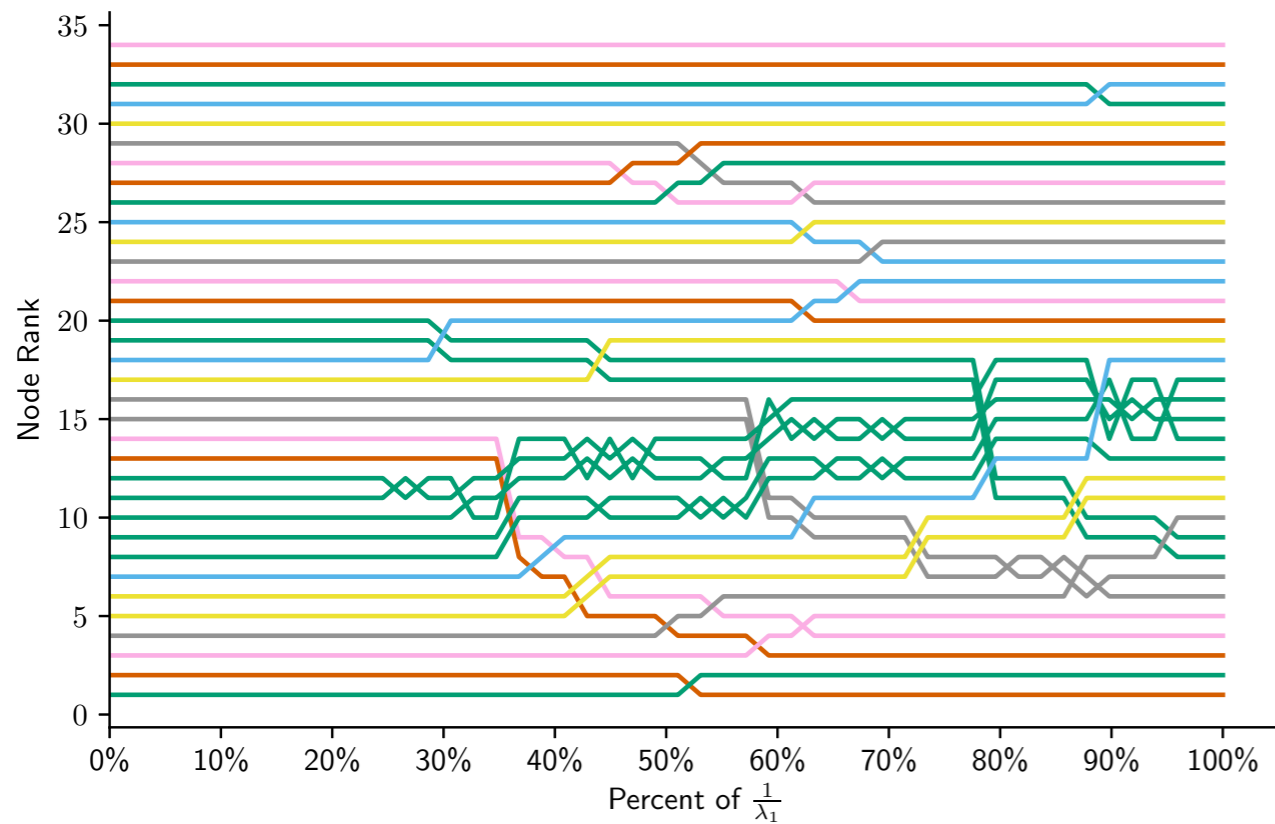
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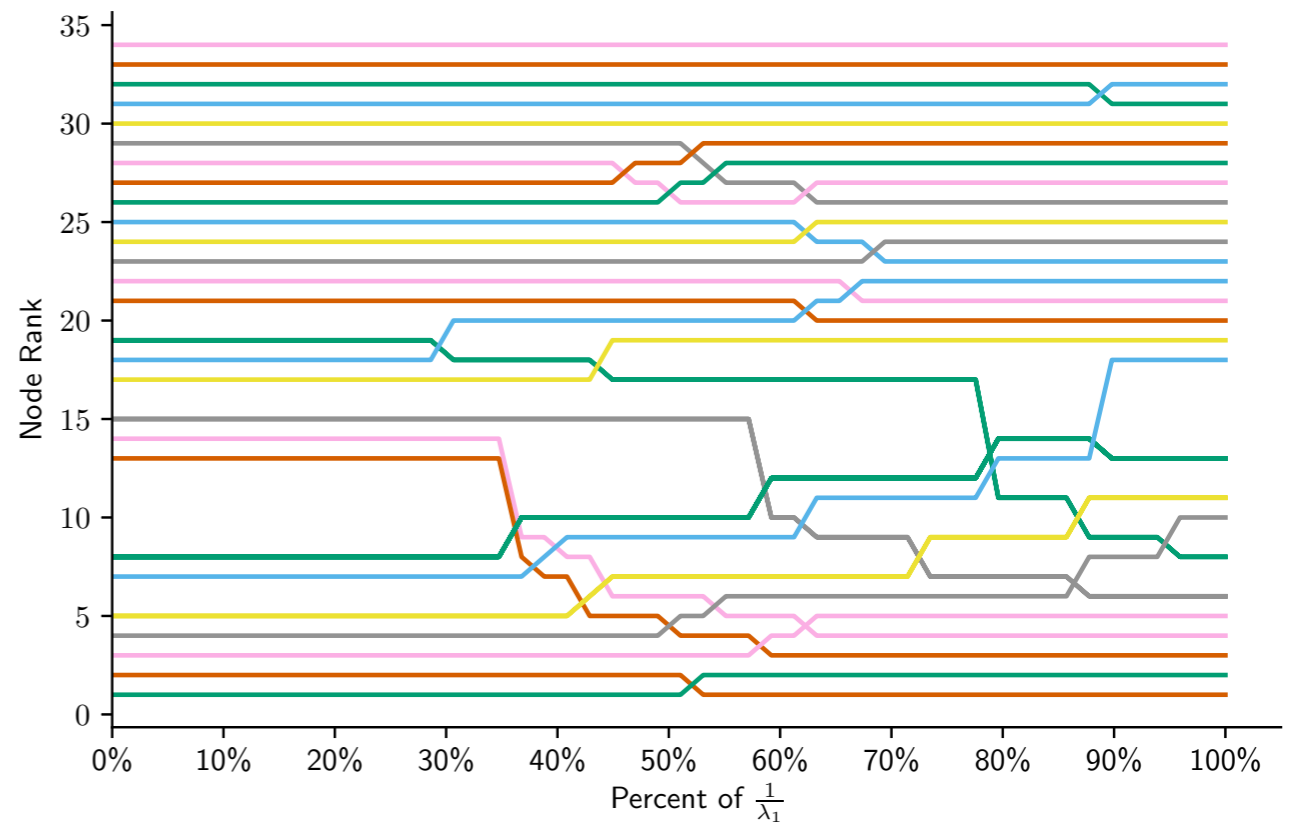
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# Walk-classes in Zachary's Karate Club

## Rank-plot for Katz

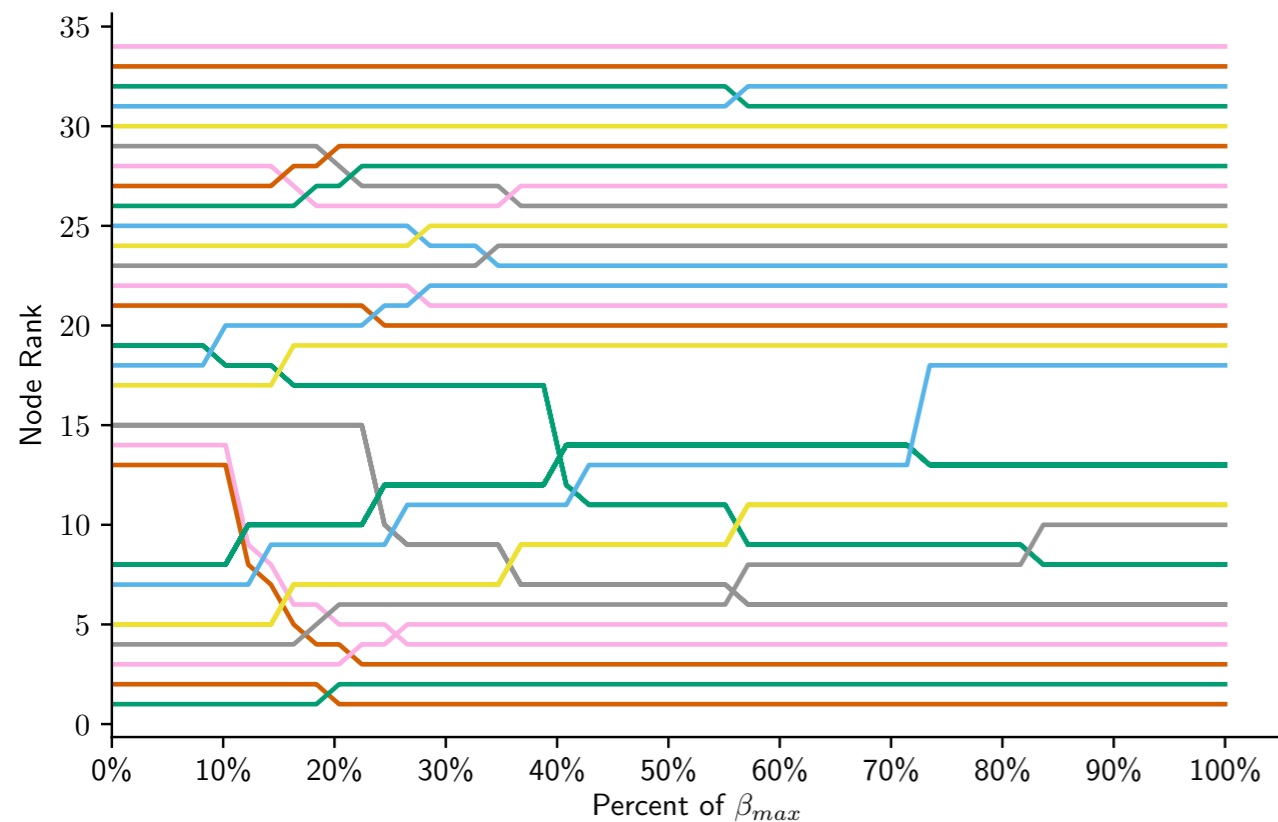


## Katz, walk-classes merged

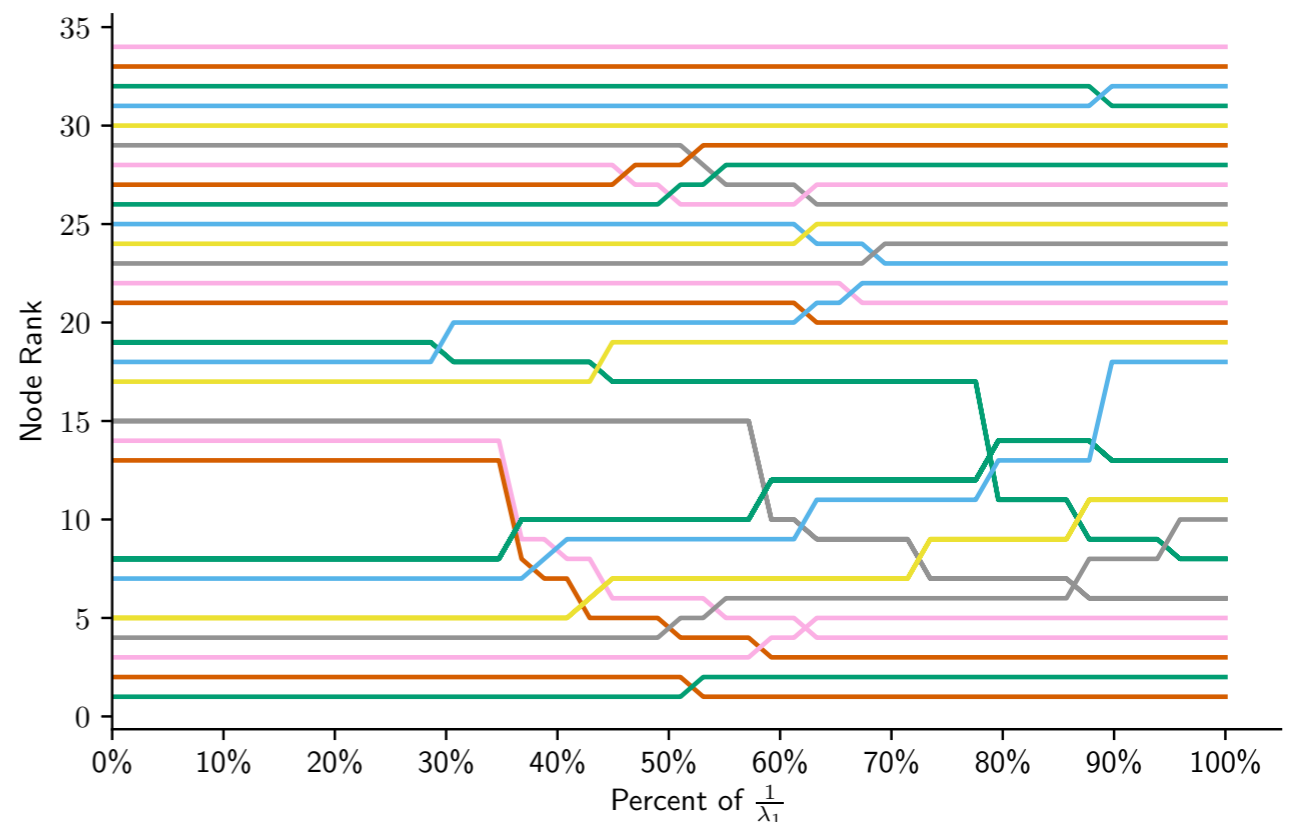


# Walk-classes in Zachary's Karate Club

## Total communicability, merged



## Katz, walk-classes merged



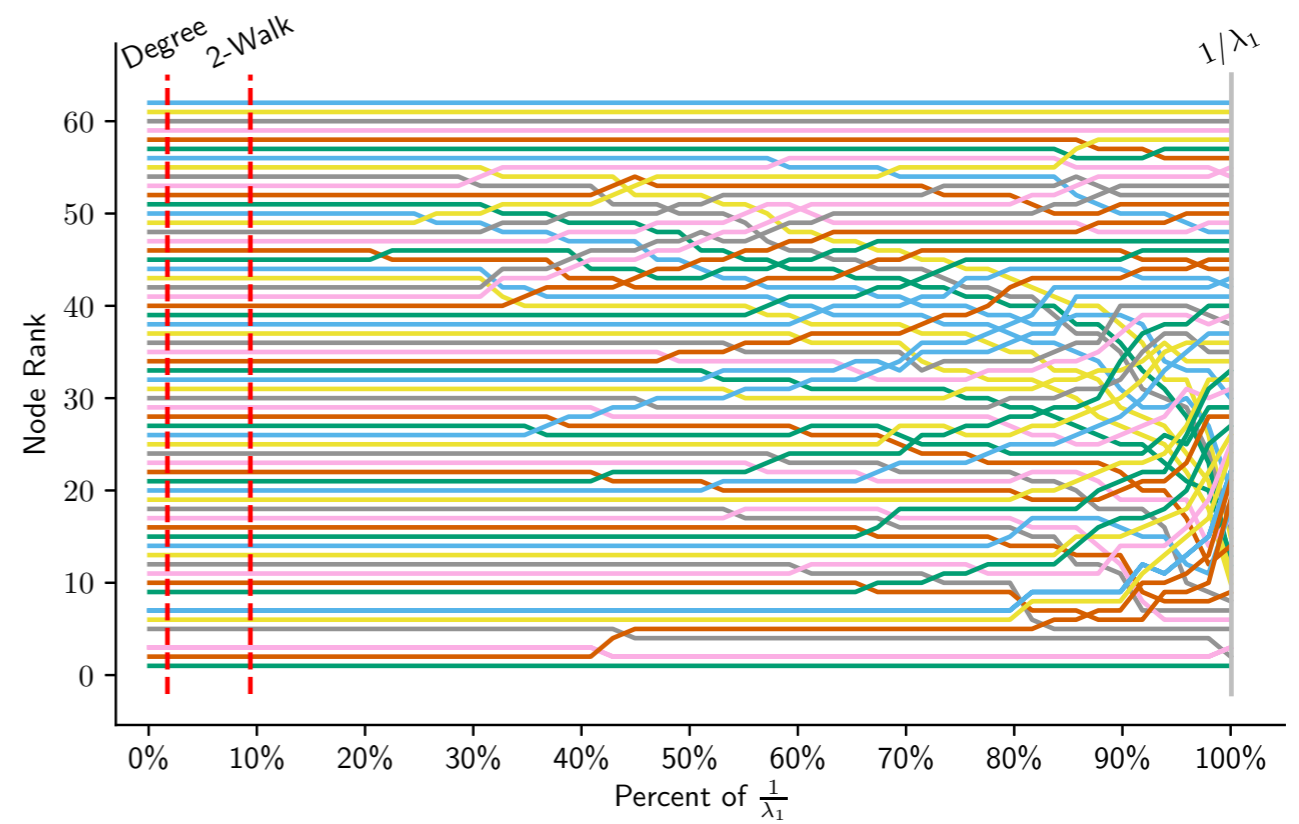
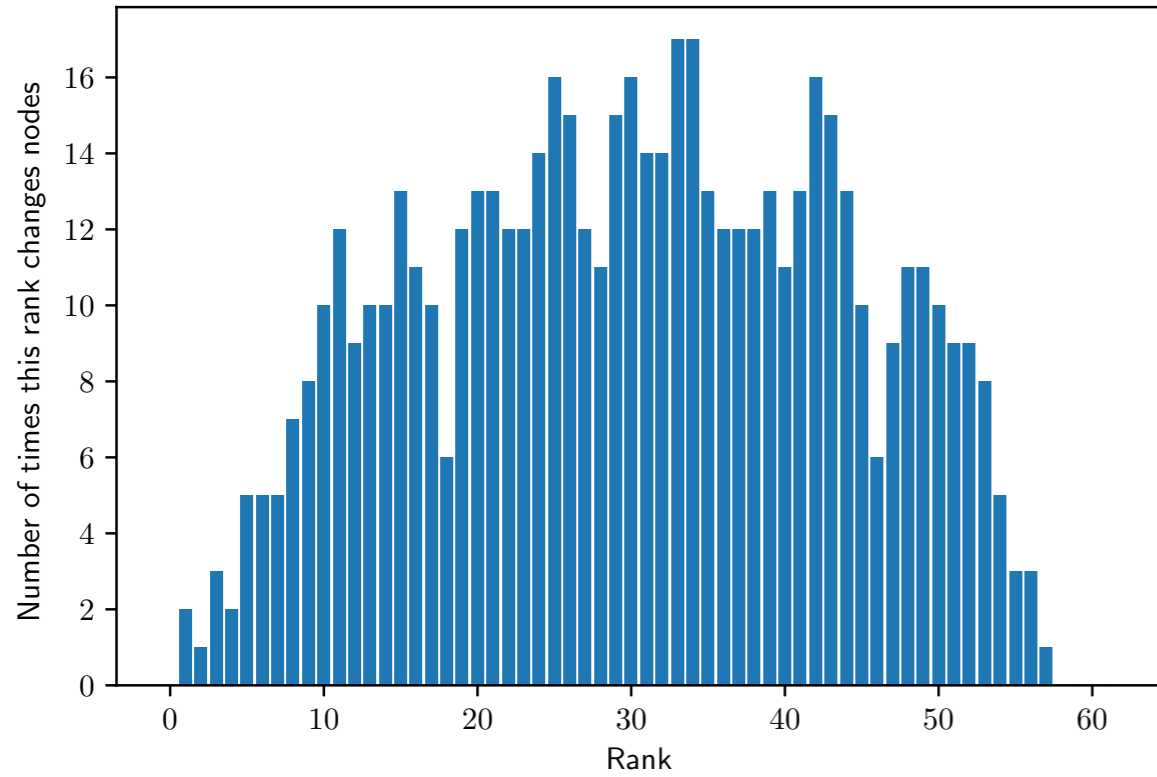
**Same node curves are merged in both plots  
because walk-classes are regardless of  $f$**

# Collisions are abundant; Iso-centrality is not

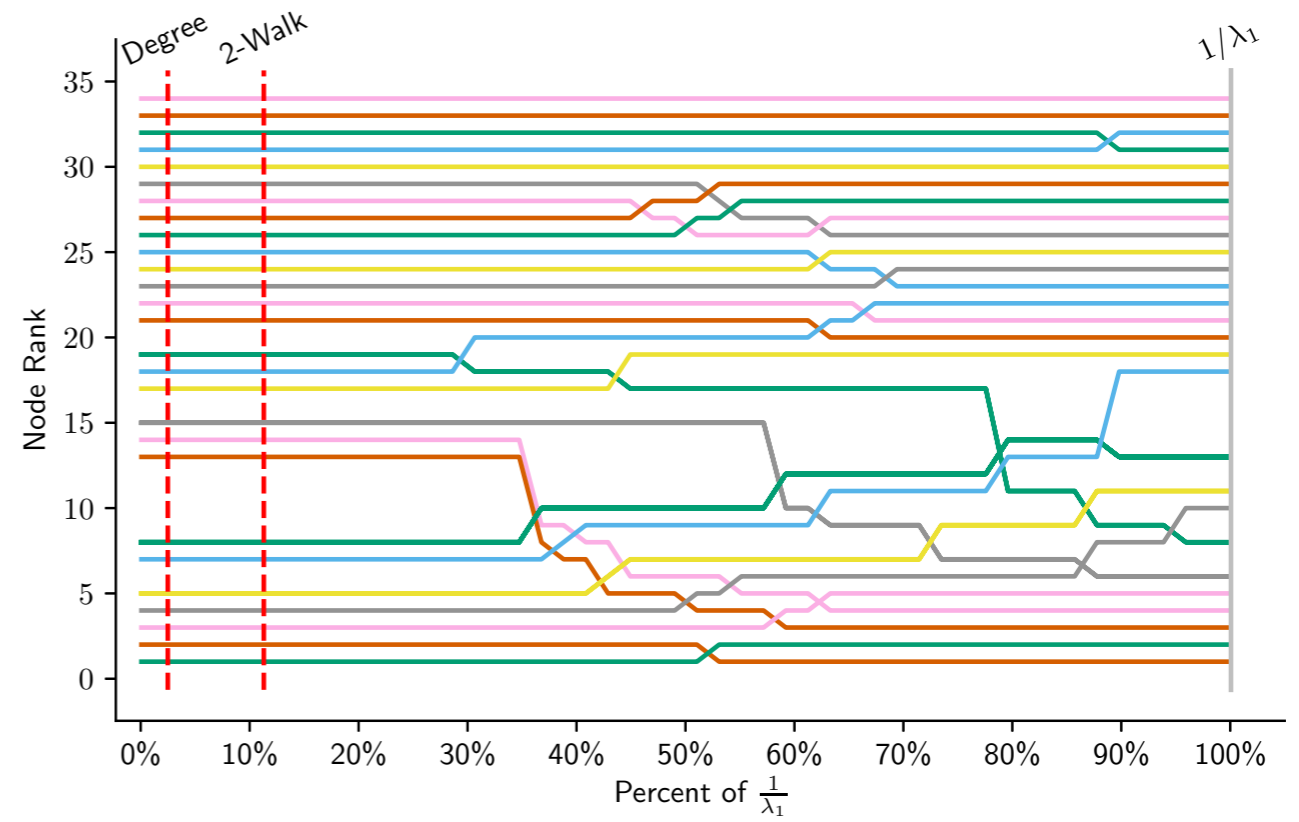
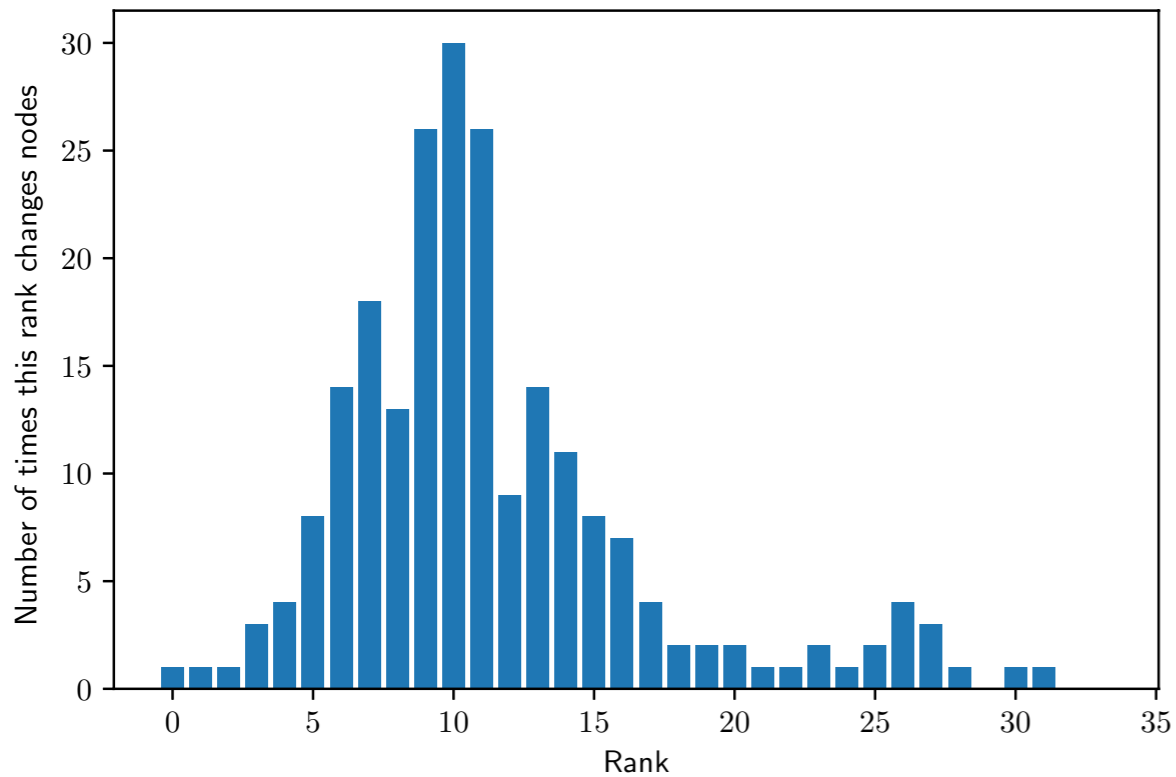
Collisions

Walk-classes merged

dolphins



karate-cc



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$$C_f(j, \beta) = C_f(i, \beta) \text{ for all } f \text{ and } \beta$$

4. Nodes  $i, j$  have the “same” eigenvector centrality

# Walk-class determines centrality behavior

**Theorem.** *Let  $M \in \mathbb{R}^{n \times n}$  be diagonalizable with  $|\lambda_1| > \dots > |\lambda_m|$ , and eigen-decomposition  $M = \sum_{k=1}^m \lambda_k \mathbf{V}_k \mathbf{U}_k^T$ . Then nodes  $i, j$  are in the same walk-class if and only if  $[\mathbf{V}_k \mathbf{U}_k^T \mathbf{s}(i)]_i = [\mathbf{V}_k \mathbf{U}_k^T \mathbf{s}(j)]_j$  holds for  $k = 1, \dots, m$ .*

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## **Theorem:**

Nodes in a walk-class have “same” **eigenvector centrality:**

$i, j$  in the same walk-class if and only if

$$[\mathbf{V}_k \mathbf{U}_k^T \mathbf{s}(i)]_i = [\mathbf{V}_k \mathbf{U}_k^T \mathbf{s}(j)]_j \quad \text{for each } k = 1, \dots, m$$

# Walk-class determines **centrality behavior**

Depending on  $\mathbf{s}$  and  $\mathbf{M}$ , this means  $i, j$  have same...

- Perron-Frobenius — eigenvector score
- Fiedler vector — score magnitude
- spectral clustering — positions symmetric in embedding

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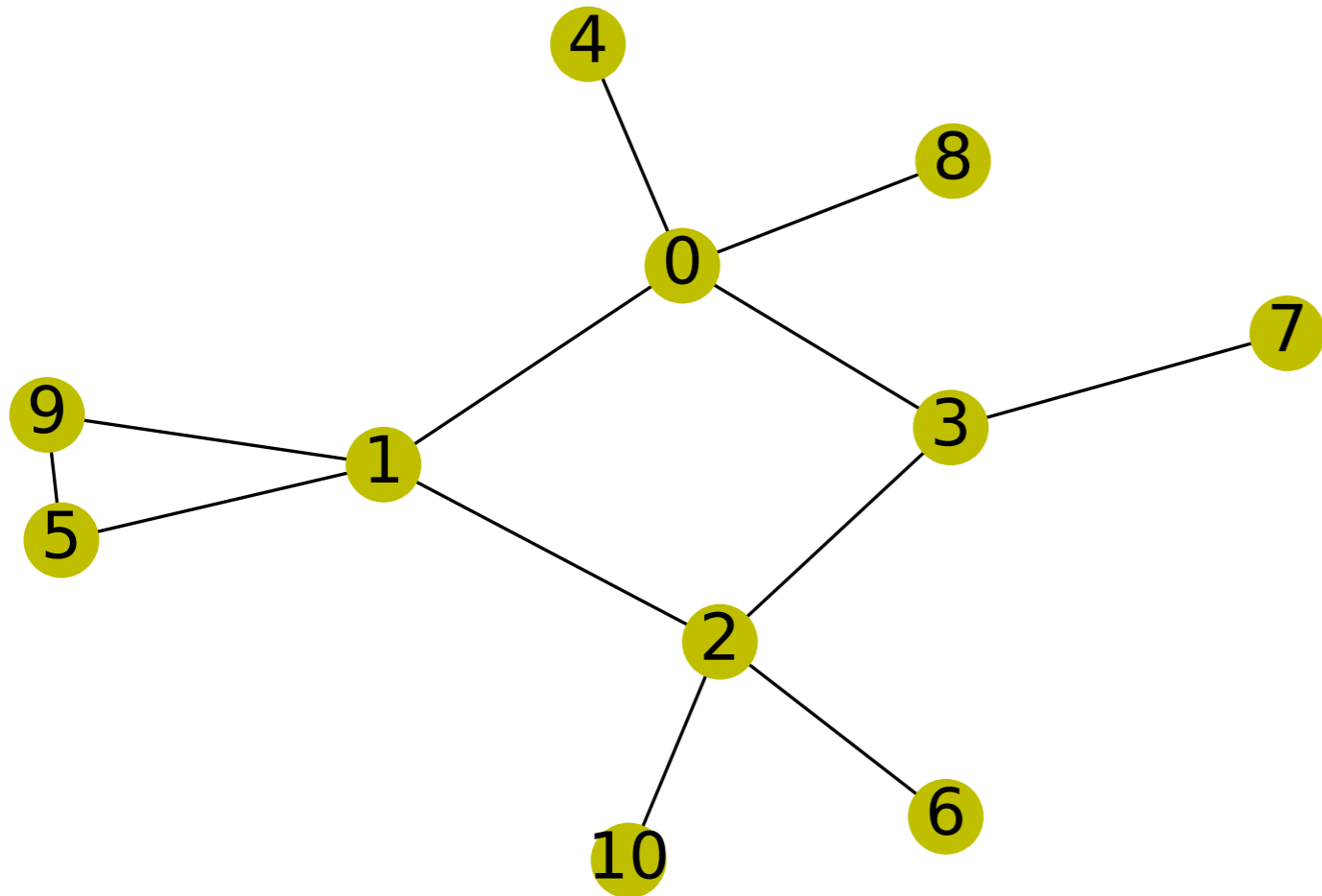
$i, j$  in the same walk-class if and only if

$$[\mathbf{V}_k \mathbf{U}_k^T \mathbf{s}(i)]_i = [\mathbf{V}_k \mathbf{U}_k^T \mathbf{s}(j)]_j \quad \text{for each } k = 1, \dots, m$$



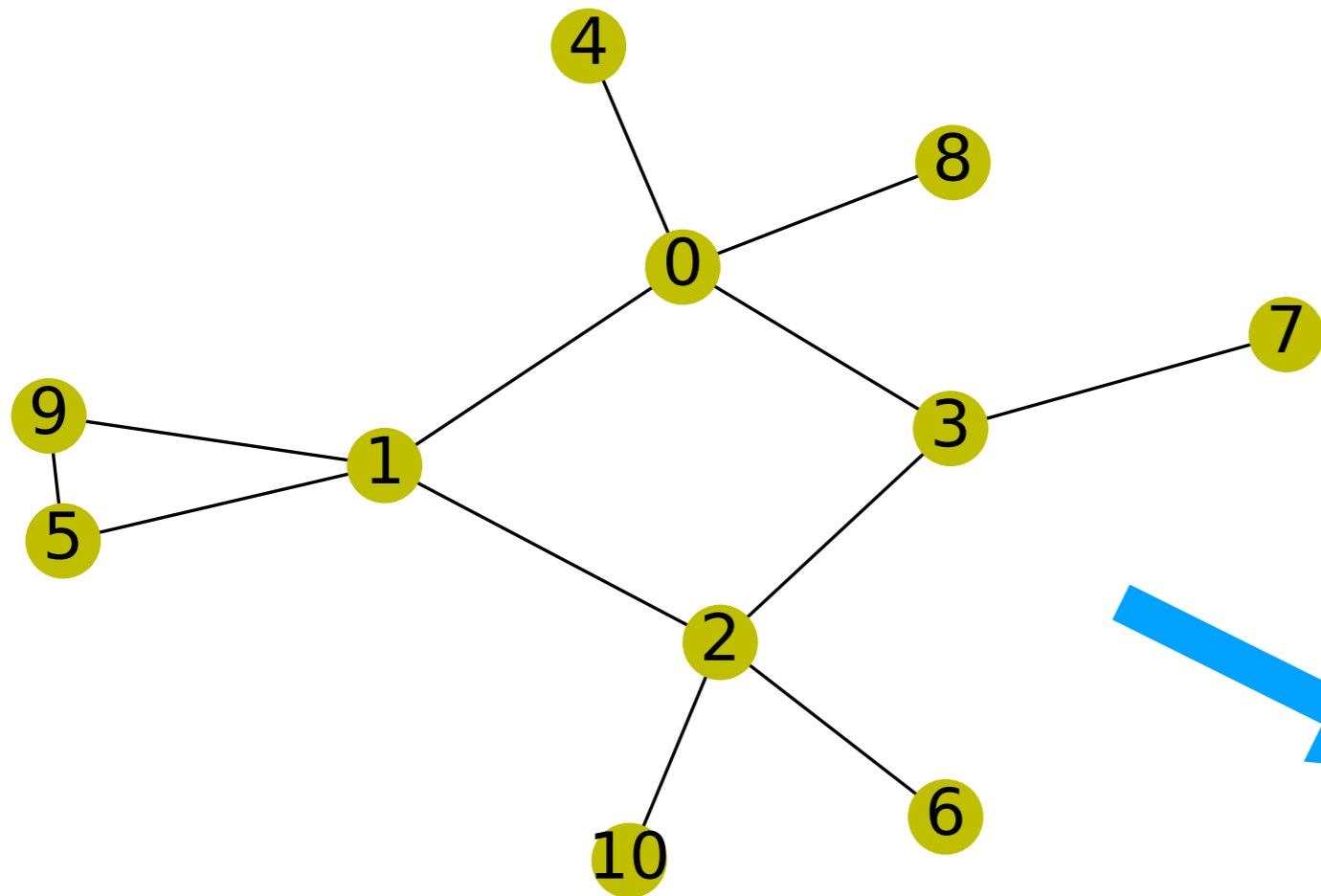
# Walk-class determines centrality behavior

**Example** Nodes in a walk-class  $\leftrightarrow$  same spectral clustering

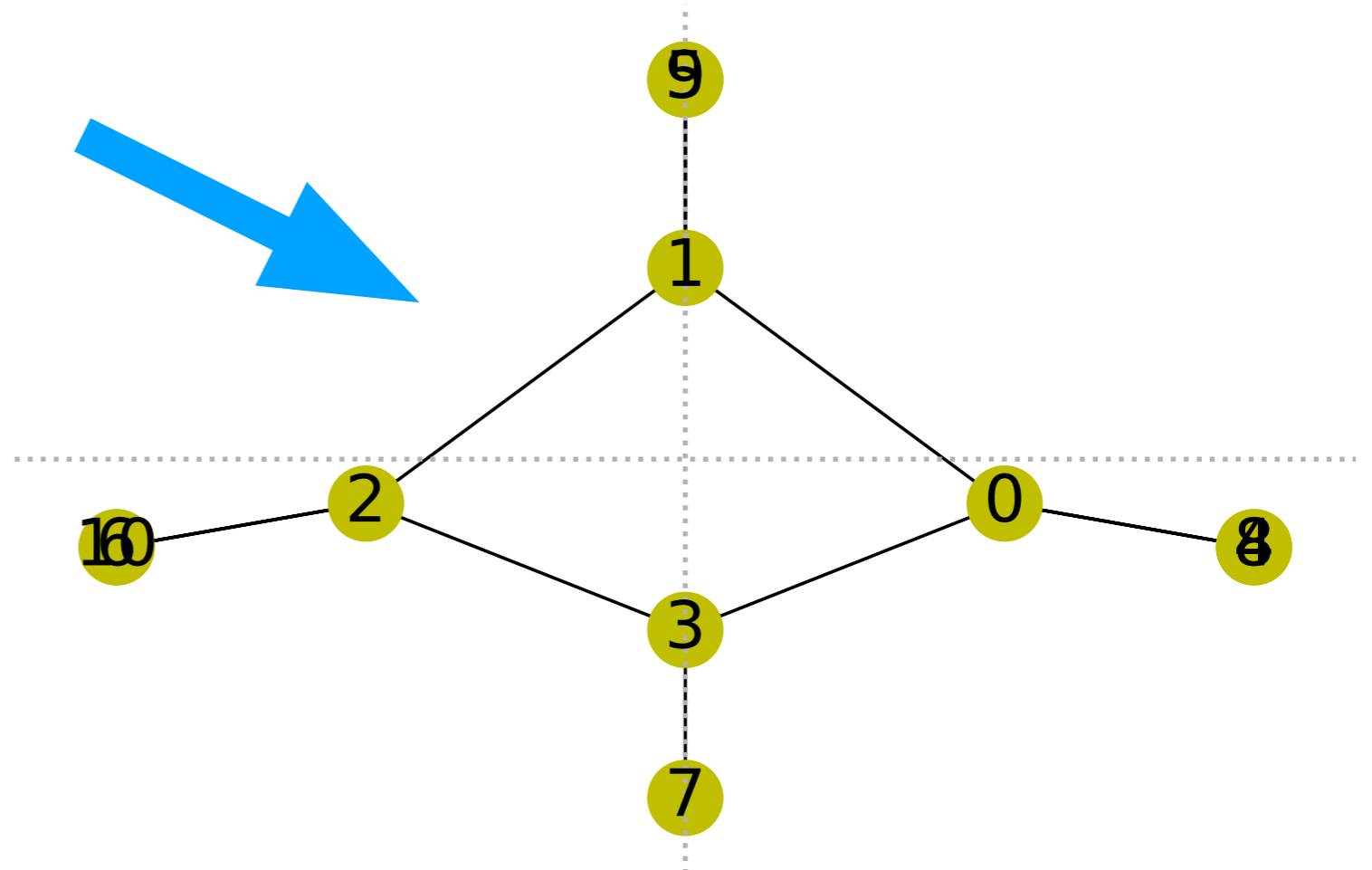


# Walk-class determines centrality behavior

**Example** Nodes in a walk-class  $\leftrightarrow$  same spectral clustering



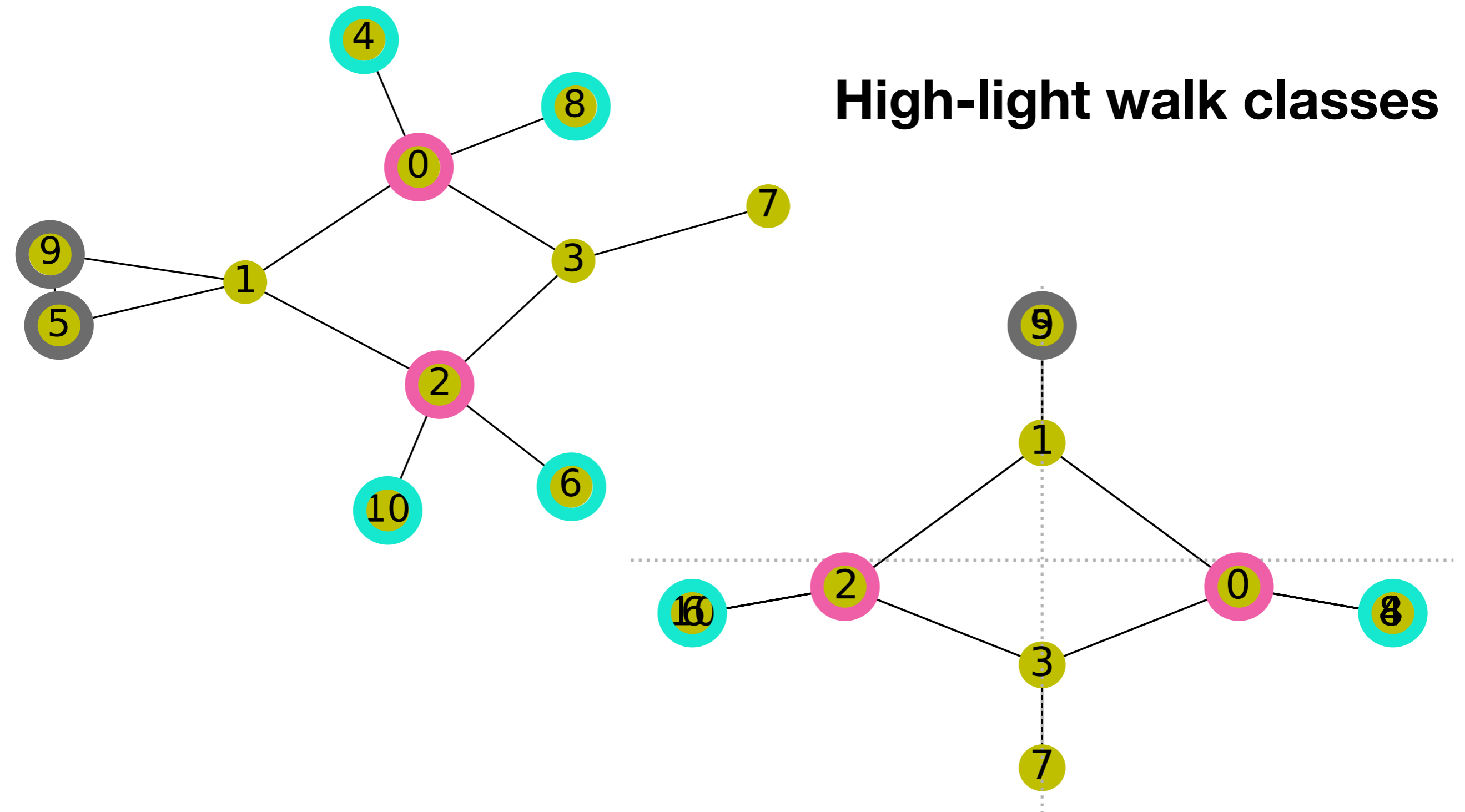
Embed nodes in  $\mathbb{R}^d$   
according to  $e_i^T U_k$



# Walk-class determines centrality behavior

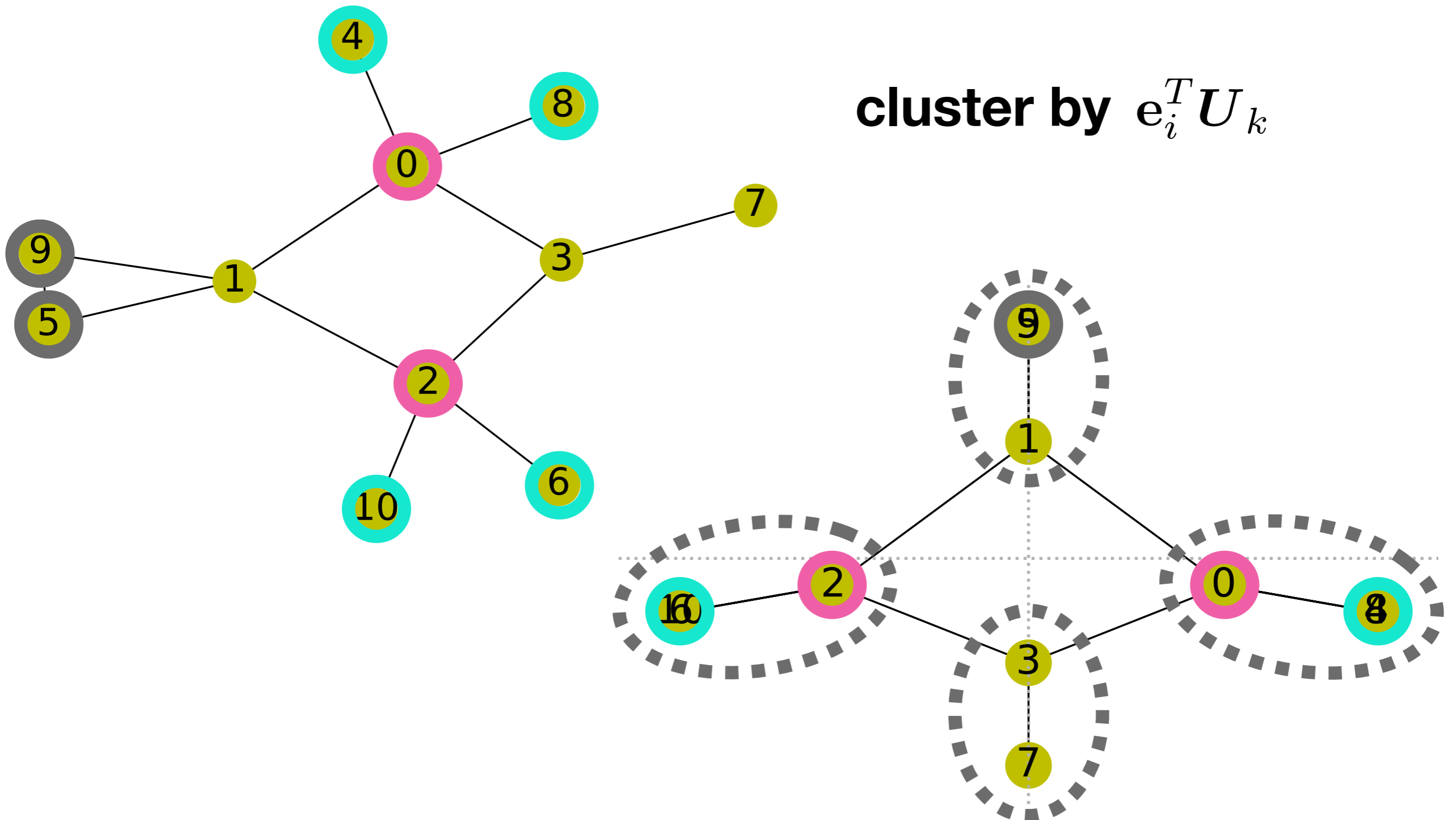
**Example** Nodes in a walk-class  $\leftrightarrow$  same spectral clustering

**High-light walk classes**



# Walk-class determines centrality behavior

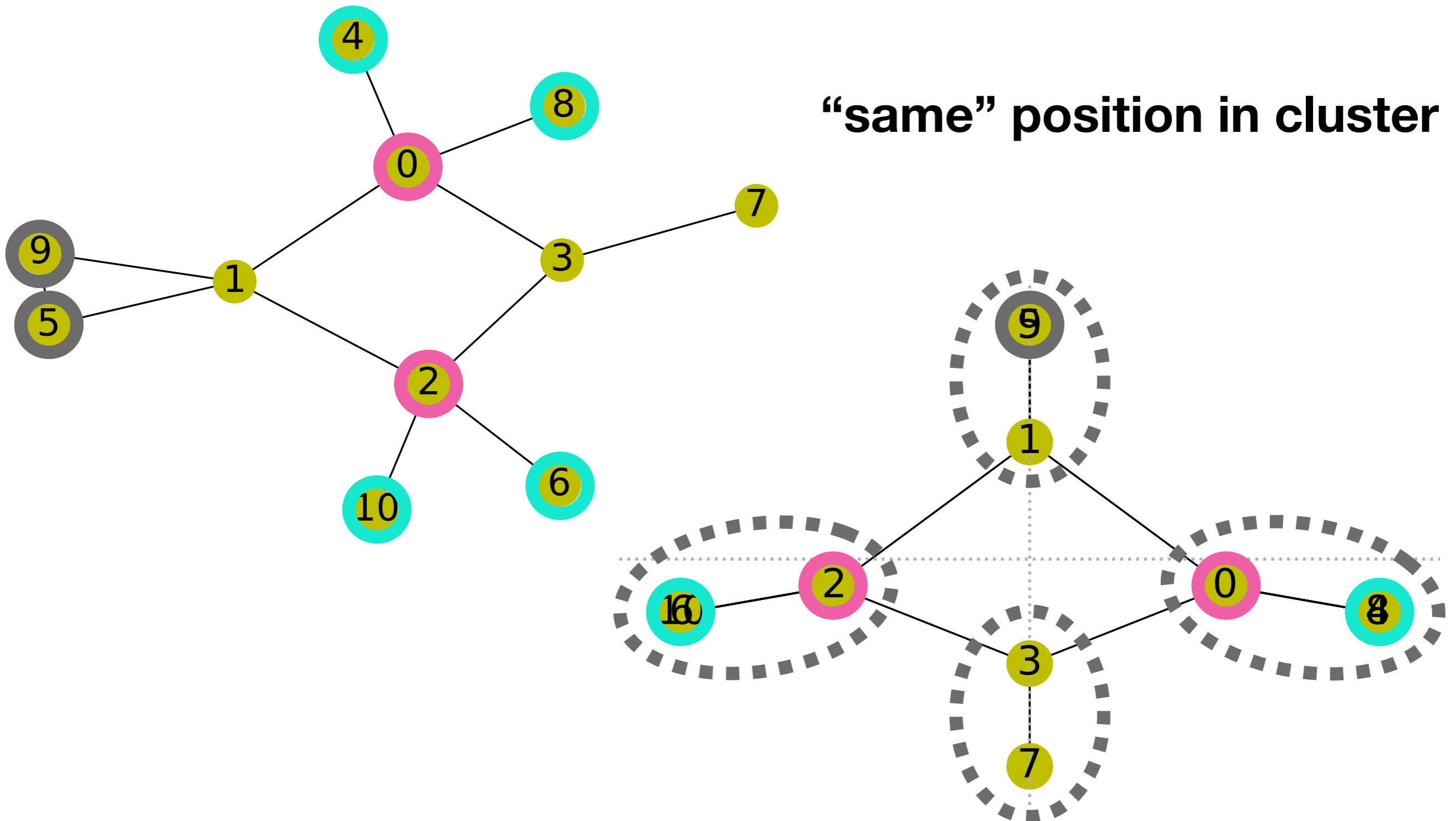
**Example** Nodes in a walk-class  $\leftrightarrow$  same spectral clustering



# Walk-class determines centrality behavior

**Example** Nodes in a walk-class  $\leftrightarrow$  same spectral clustering

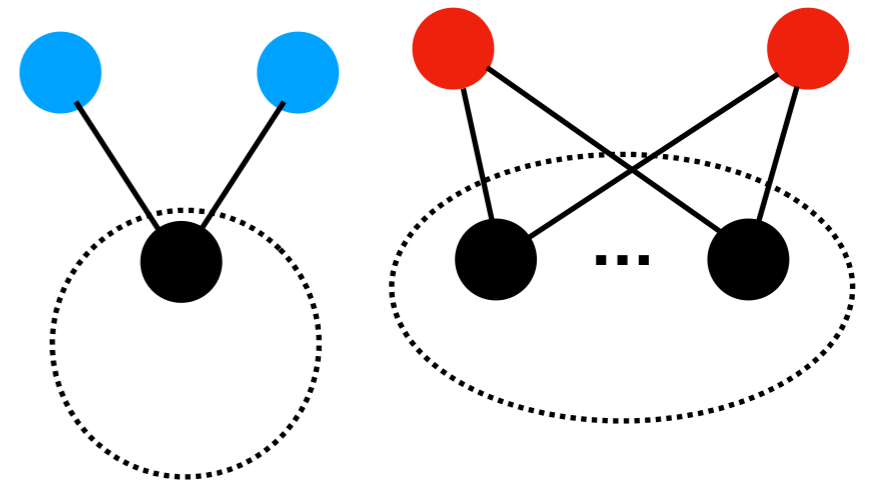
“same” position in cluster



# Walk-class -> centrality: take-aways

- Iso-centrality can occur in real networks
- But it's rare
- Usual cause: small symmetries, e.g.

two nodes that have same neighborhood



- Iso-central nodes have identical walk-centrality, walk structure, and “same” eigen-centrality

# Thank you!

## Concluding thoughts:

- The more  $j$  (partially) majorizes  $i$ ,  
the larger interval near 0 where  $j$  ranks above  $i$
- # collisions is finite, unless nodes are iso-central
- Iso-central nodes rare; same walk- and eigen-centrality

Future work: Better understand collision bounds, effect on ranking

**Bonus slides, if time**



# Walk-classes in Spider Donuts

The spider donut graph family:

- This spider donut has exactly 3 walk-classes
- Constructed so they collide simultaneously
- Also constructed a 4-walk-class spider donut, all 4 class collide simultaneously
- Motivation: wanted to understand how many walk-classes can be involved in one collision at a single parameter value.

