

Advanced Topics in Control: Distributed Systems and Control

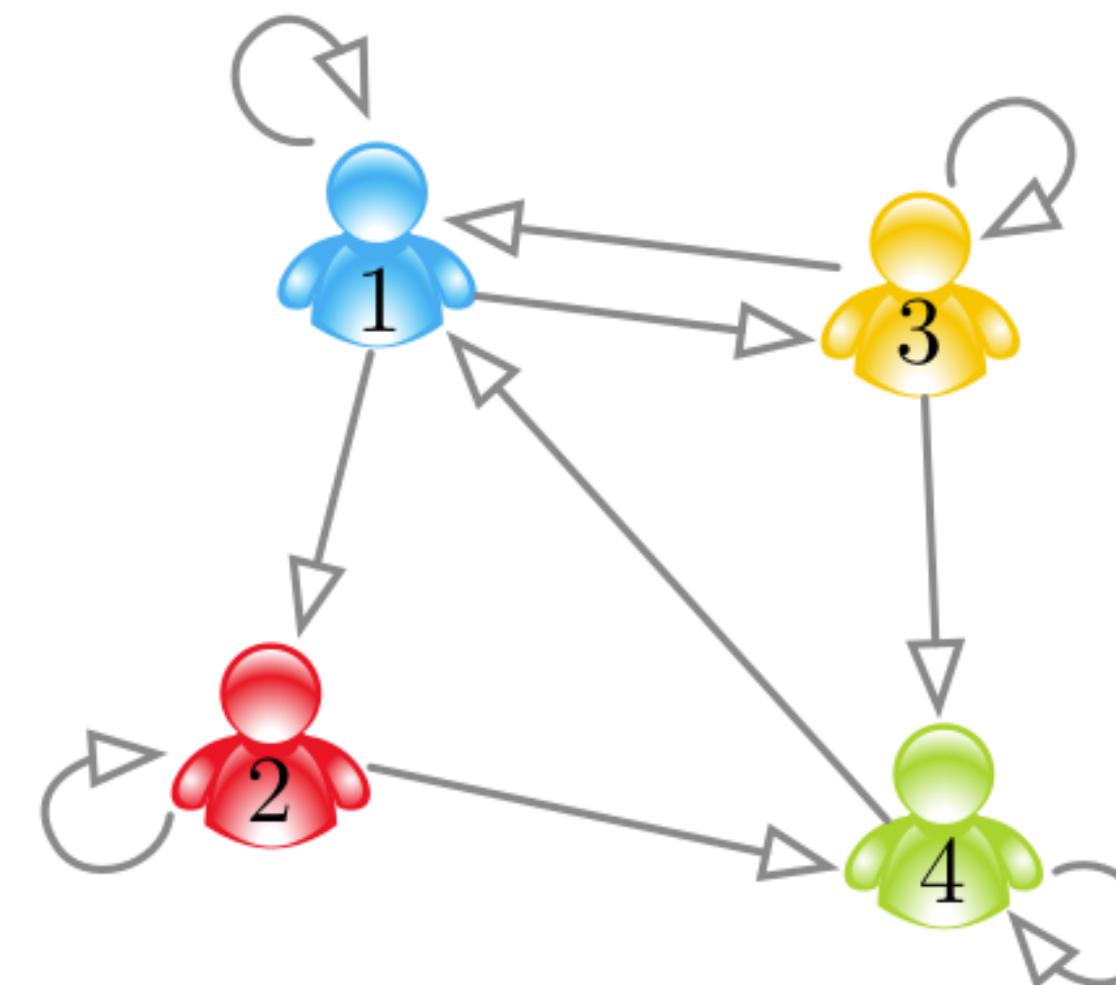
Lecture 3: Algebraic Graph Theory

Dr. Mohammad H. (Vahid) Mamduhi, March 7th, 2022



rawpixel

Quick review of Lecture 2



Distributed averaging algorithm:

$$x^+ = Ax$$

Properties:

- sparsity pattern induced by graph
- nonnegative coefficients: $a_{ij} \geq 0$
- row-stochasticity: $\sum_j a_{ij} = 1$

Observations regarding the solution $x(k) = A^k x_0$:

$$A = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} & 0 \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{3} & 0 & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & 0 & 0 & \frac{1}{2} \end{bmatrix} \quad A^2 = \begin{bmatrix} \frac{2}{9} & \frac{5}{18} & \frac{2}{9} & \frac{5}{18} \\ \frac{1}{4} & \frac{1}{4} & 0 & \frac{1}{2} \\ \frac{7}{18} & \frac{1}{9} & \frac{2}{9} & \frac{5}{18} \\ \frac{5}{12} & \frac{1}{6} & \frac{1}{6} & \frac{1}{4} \end{bmatrix} \quad \dots \quad A^\infty = \frac{6}{19} \begin{bmatrix} 1 & \frac{2}{3} & \frac{1}{2} & 1 \\ 1 & \frac{2}{3} & \frac{1}{2} & 1 \\ 1 & \frac{2}{3} & \frac{1}{2} & 1 \\ 1 & \frac{2}{3} & \frac{1}{2} & 1 \end{bmatrix}$$

⇒ agreement/sync/rendezvous: $\lim_{k \rightarrow \infty} x(k) = \left(\frac{6}{19} [1 \quad \frac{2}{3} \quad \frac{1}{2} \quad 1] \cdot x_0 \right) \cdot \mathbf{1}_4$

Quick review of Lecture 2

$$x(t+1) = Ax(t) \Rightarrow \lim_{t \rightarrow \infty} x(t) = \lim_{t \rightarrow \infty} A^t x(0) \text{ exists if } A \text{ is convergent} \Leftarrow$$

non-negative

dominant eigenvalue λ is simple

$\lambda > |\mu|$ for any other eigenvalue μ

Gershgorin disks theorem

Theorem 2.8 (Geršgorin Disks Theorem). For any square matrix $A \in \mathbb{R}^{n \times n}$,

$$\text{spec}(A) \subset \bigcup_{i \in \{1, \dots, n\}} \underbrace{\left\{ z \in \mathbb{C} \mid |z - a_{ii}| \leq \sum_{j=1, j \neq i}^n |a_{ij}| \right\}}_{\text{disk in the complex plane centered at } a_{ii} \text{ with radius } \sum_{j=1, j \neq i}^n |a_{ij}|}$$

or semi-convergent



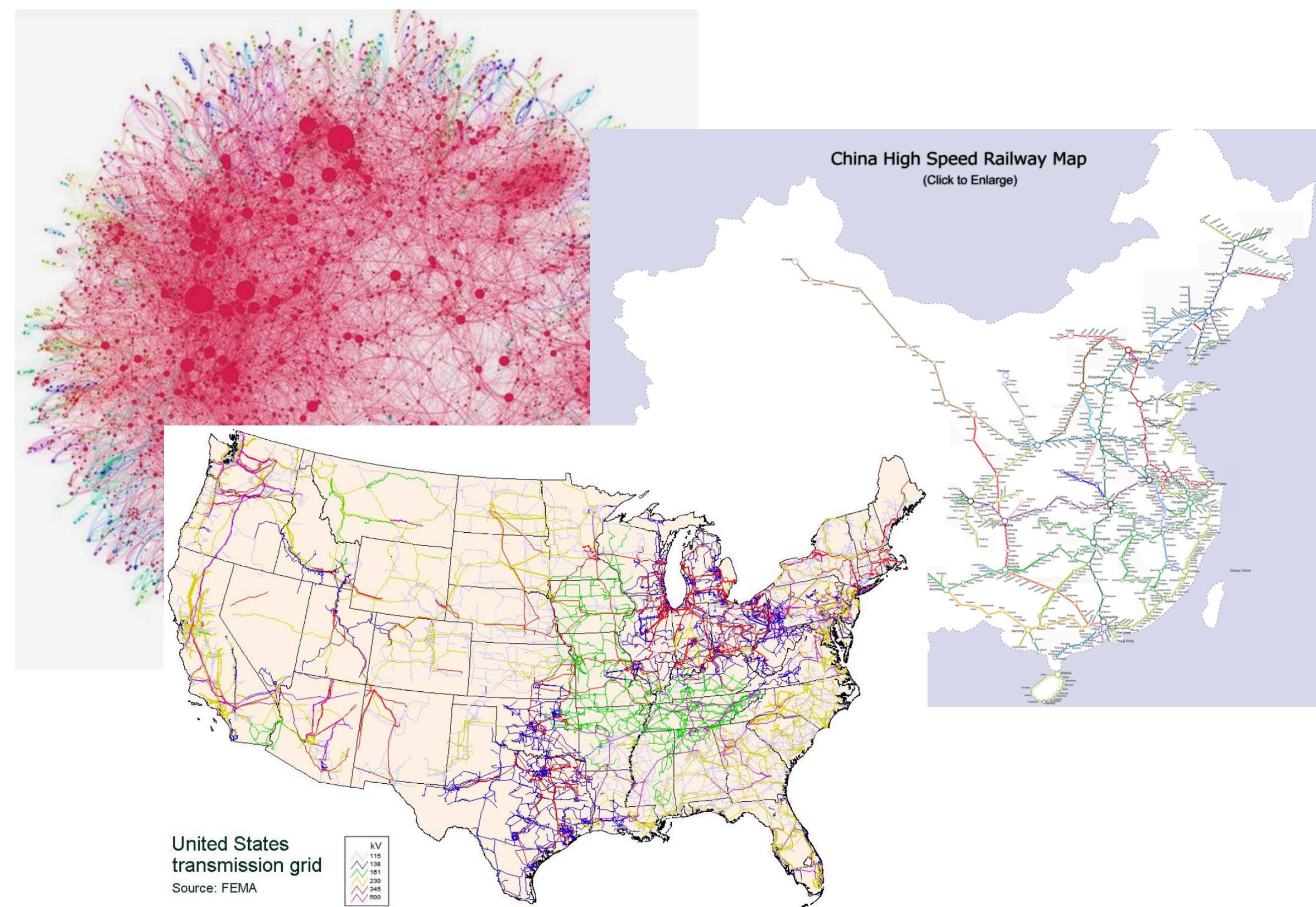
Theorem 2.13 (Powers of non-negative matrices with a simple and strictly dominant eigenvalue). Let A be a non-negative matrix. Assume the dominant eigenvalue λ is simple and strictly larger, in magnitude, than all other eigenvalues. Then A/λ is semi-convergent and

$$\lim_{k \rightarrow \infty} \frac{A^k}{\lambda^k} = vw^T,$$

where v and w are the right and left dominant eigenvectors of A normalized so that $v \geq 0$, $w \geq 0$ and $v^T w = 1$.

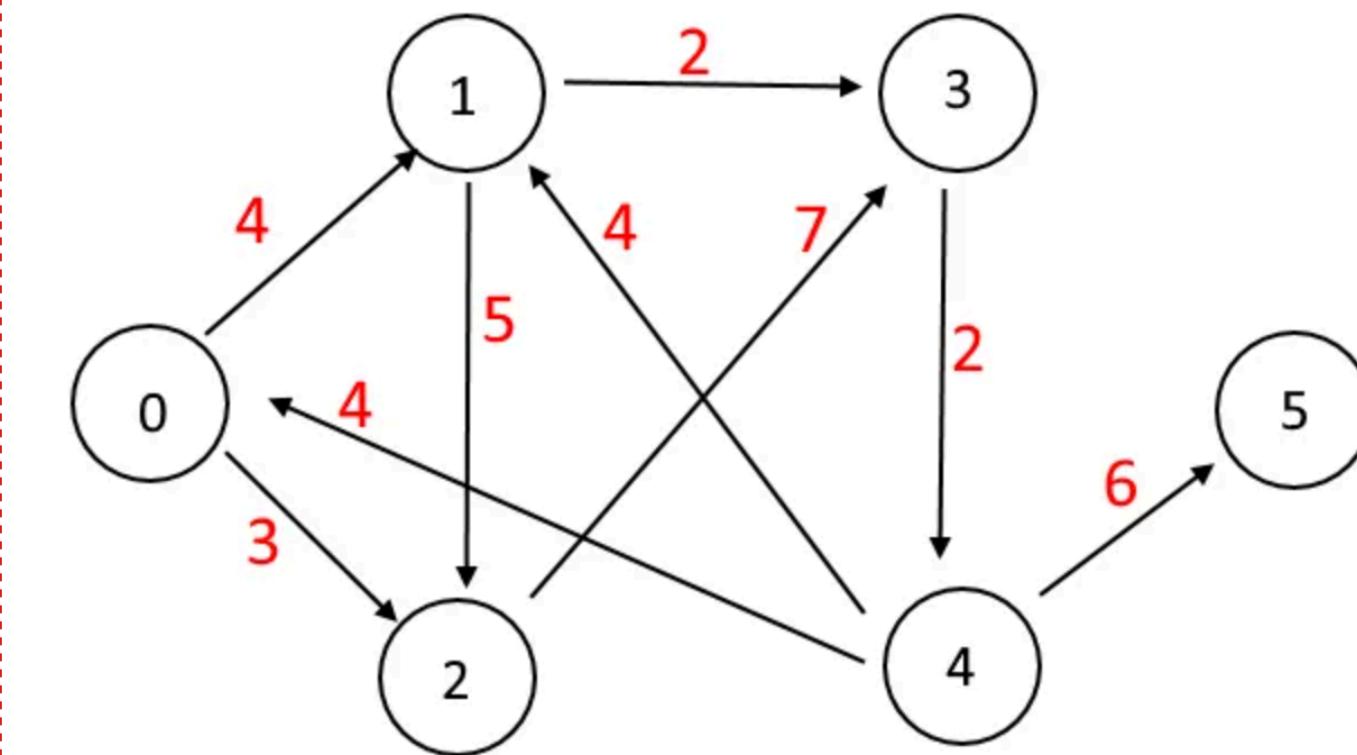
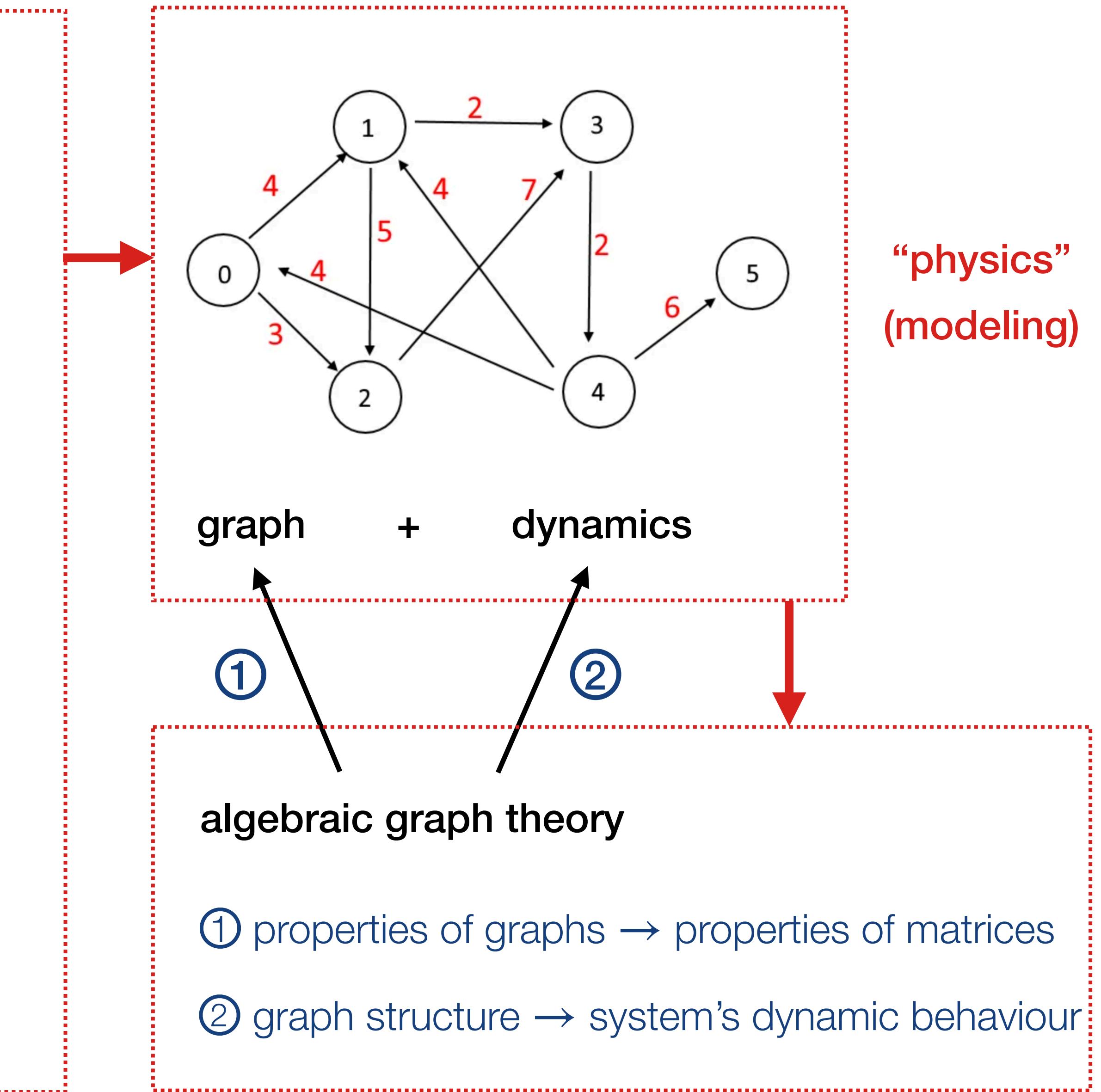


Perron-Frobenius theorem: when A is primitive



Inter-connected systems: power grids
transportation networks
social networks, etc

real world (empirical observation)



graph + dynamics

① properties of graphs → properties of matrices

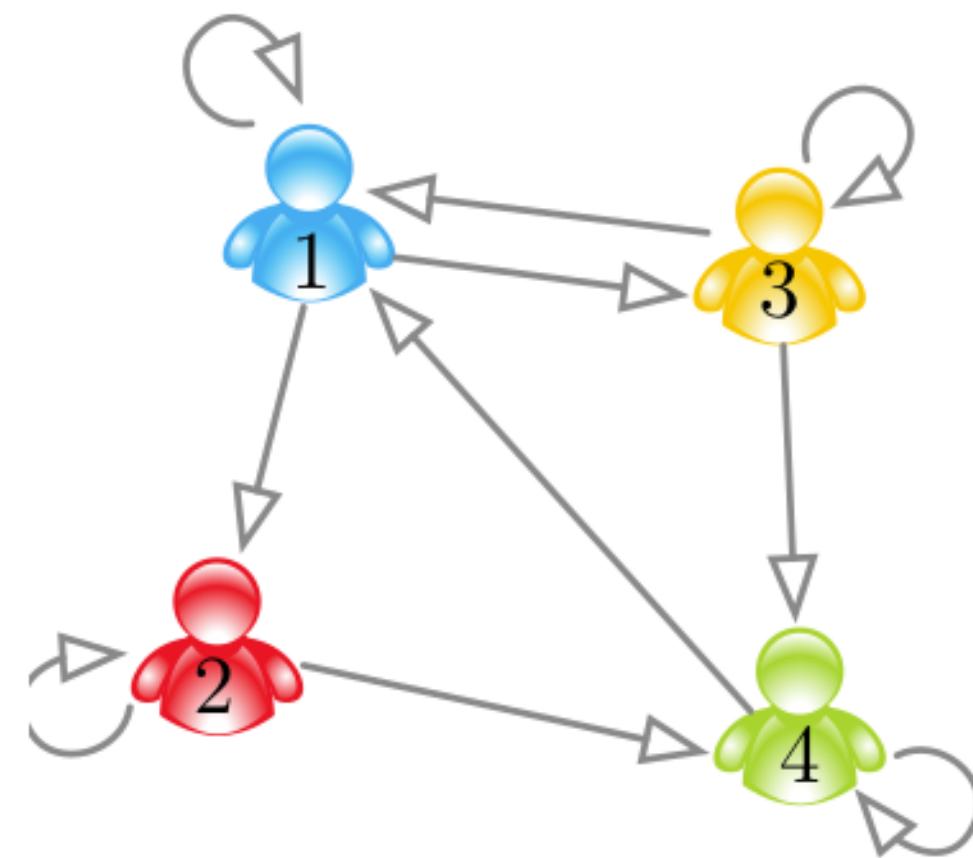
② graph structure → system's dynamic behaviour

algebraic graph theory

mathematics (analysis)

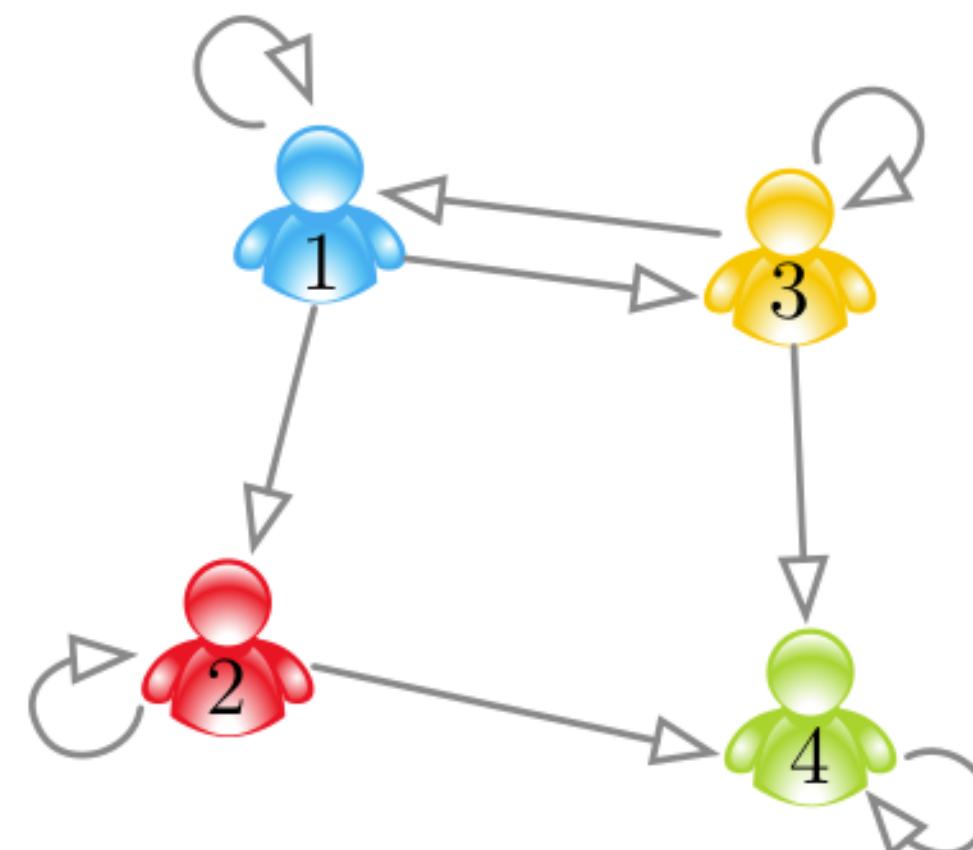
“physics”
(modeling)

This week and next week: deduce convergence from graph

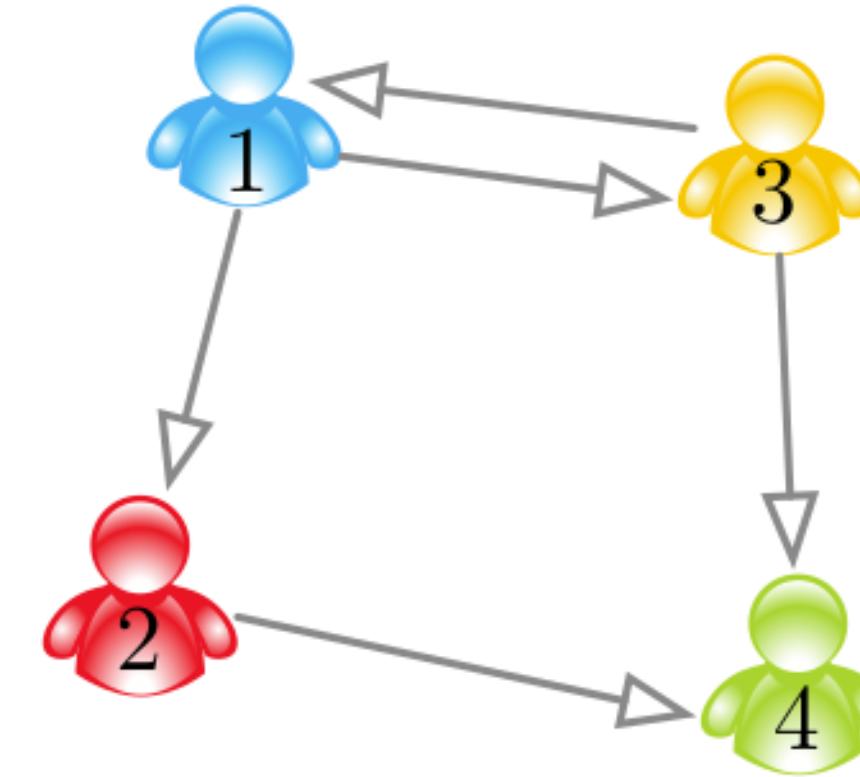


$$A = \begin{bmatrix} * & * & * & \\ & * & & \\ * & & * & * \\ * & & & * \end{bmatrix}$$

nonnegative and row-stochastic
⇒ converges to $\mathbb{1} \cdot [* * * *]$



converges to $\mathbb{1} \cdot [0 0 0 1]$



diverges & oscillates ...

Deduce convergence from graph: An example

① iterative algorithm: $x^+ = Ax$

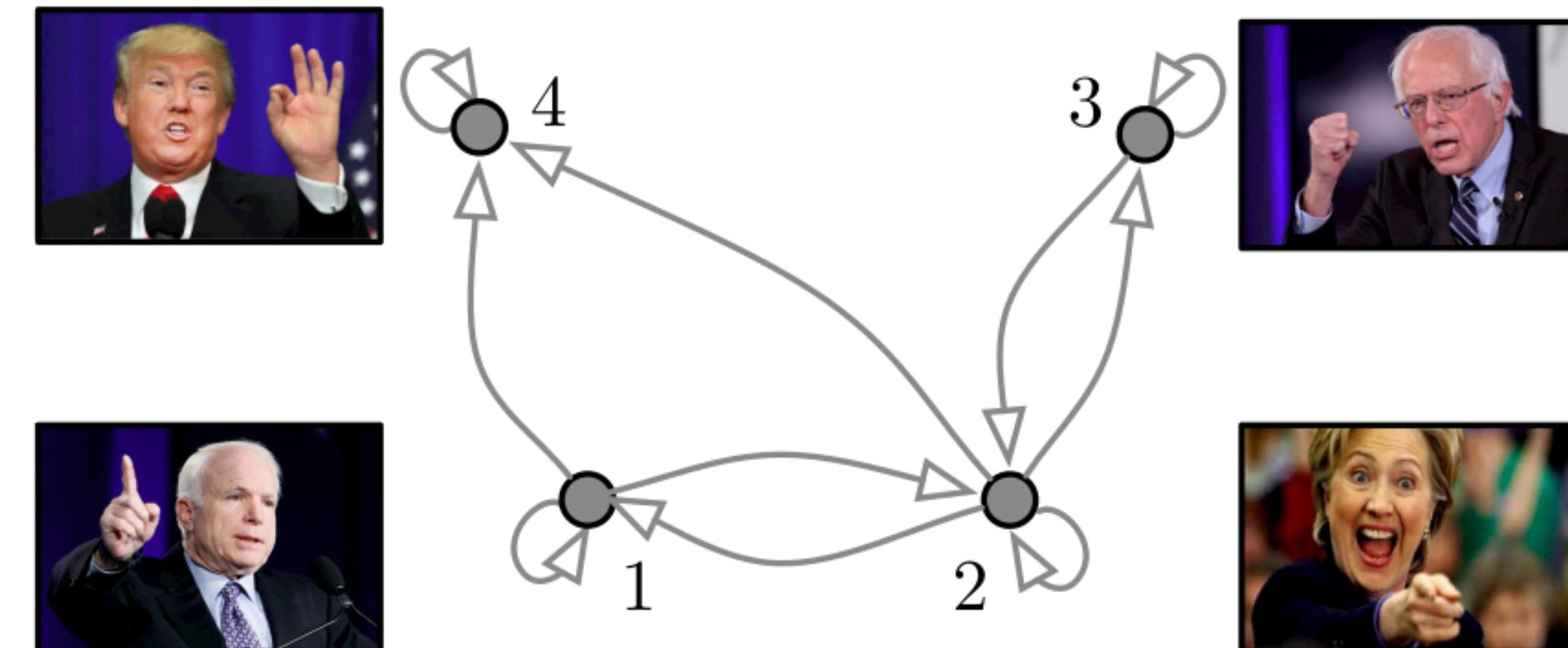
② non-negativity: $A \geq 0$

③ row-stochasticity: $A\mathbf{1} = \mathbf{1}$

④ diagonal self-weights: $a_{ii} > 0$

$$A = \begin{pmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & \\ & & & * \end{pmatrix}$$

$(* > 0$ unspecified)



Q1: Does the iteration converge?

⇒ yes ✓

Q2: What does it converge to?

⇒ $\lim_{k \rightarrow \infty} x(k) = \mathbf{1} \cdot [0 \ 0 \ 0 \ 1] \cdot x_0$

Deduce convergence from graph: An example

What if node 4 was also listening to somebody?

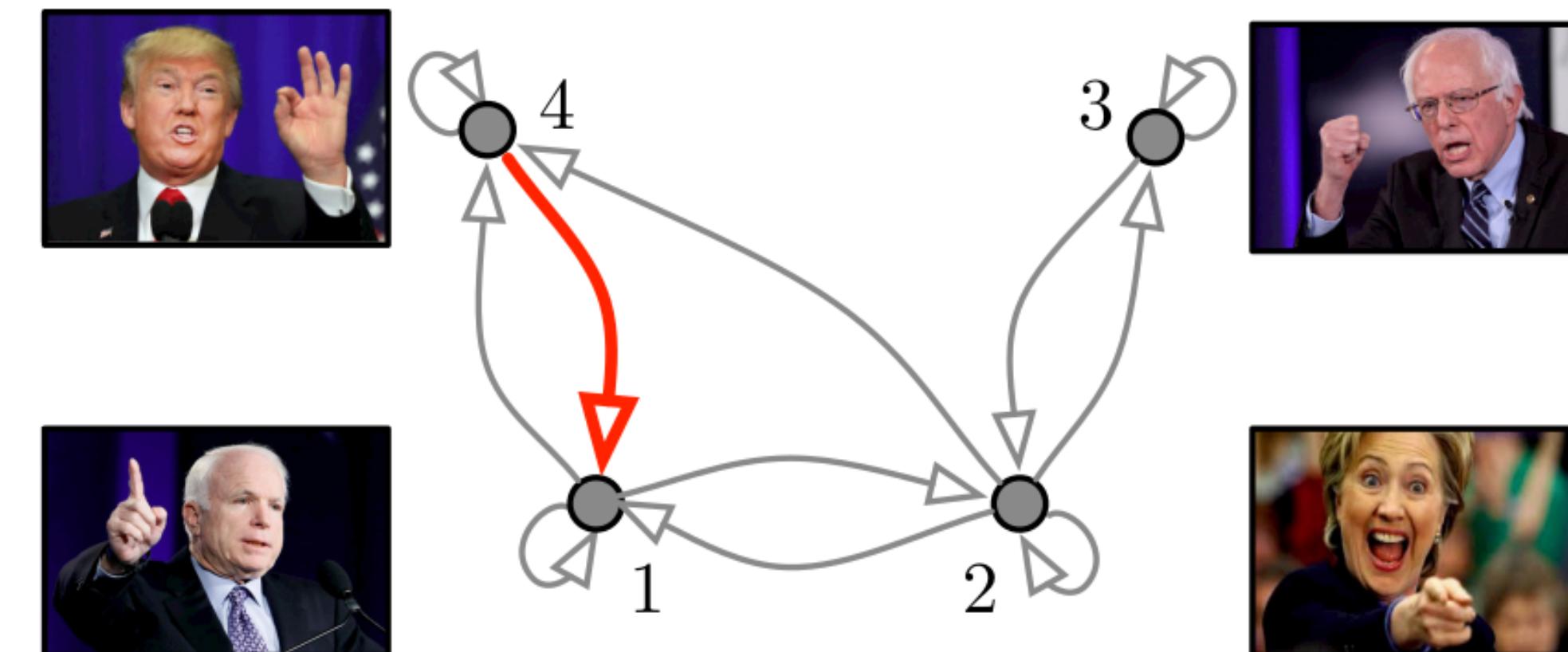
① iterative algorithm: $x^+ = Ax$

② non-negativity: $A \geq 0$

③ row-stochasticity: $A\mathbf{1} = \mathbf{1}$

④ diagonal self-weights: $a_{ii} > 0$

$$A = \begin{pmatrix} \star & \star & \star & \star \\ \star & \star & \star & \star \\ \star & \star & \star & \star \\ \star & & & \star \end{pmatrix} \quad (\star > 0 \text{ unspecified})$$



Q1: Does the iteration converge?

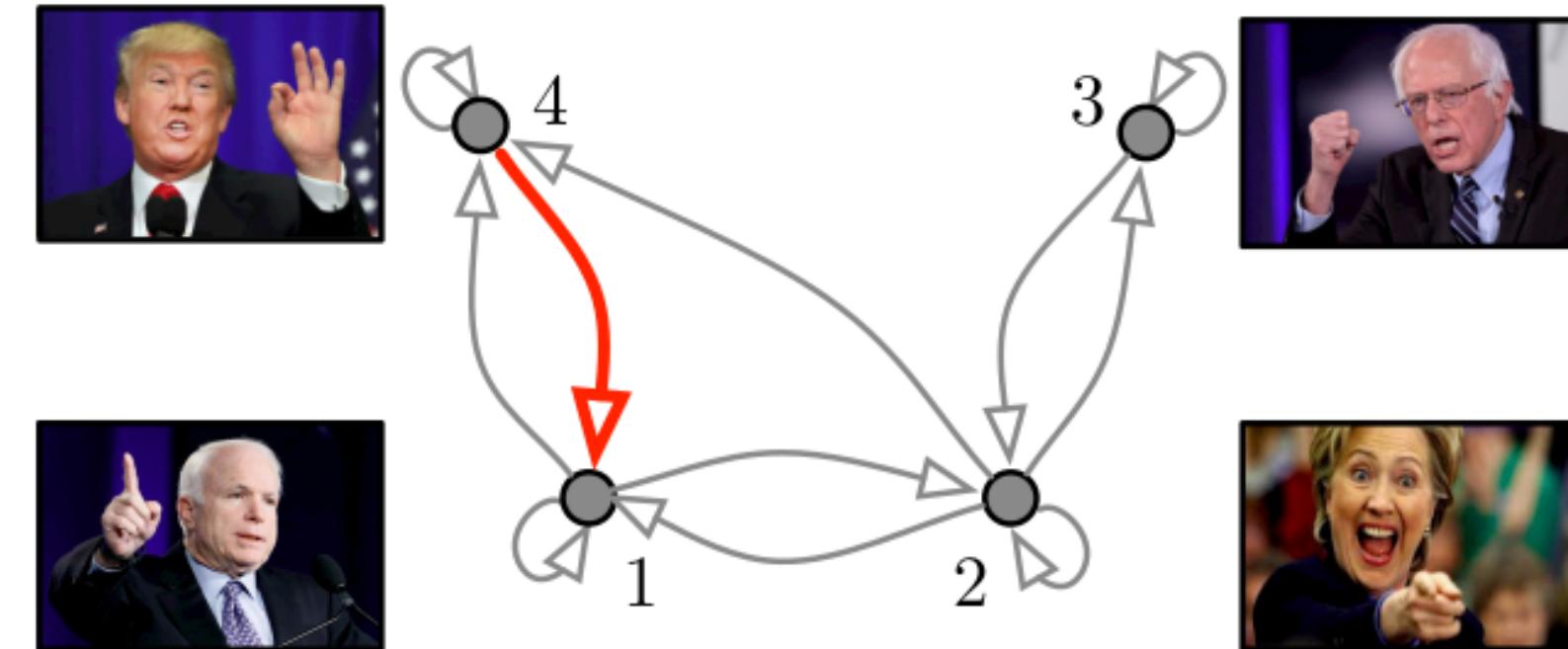
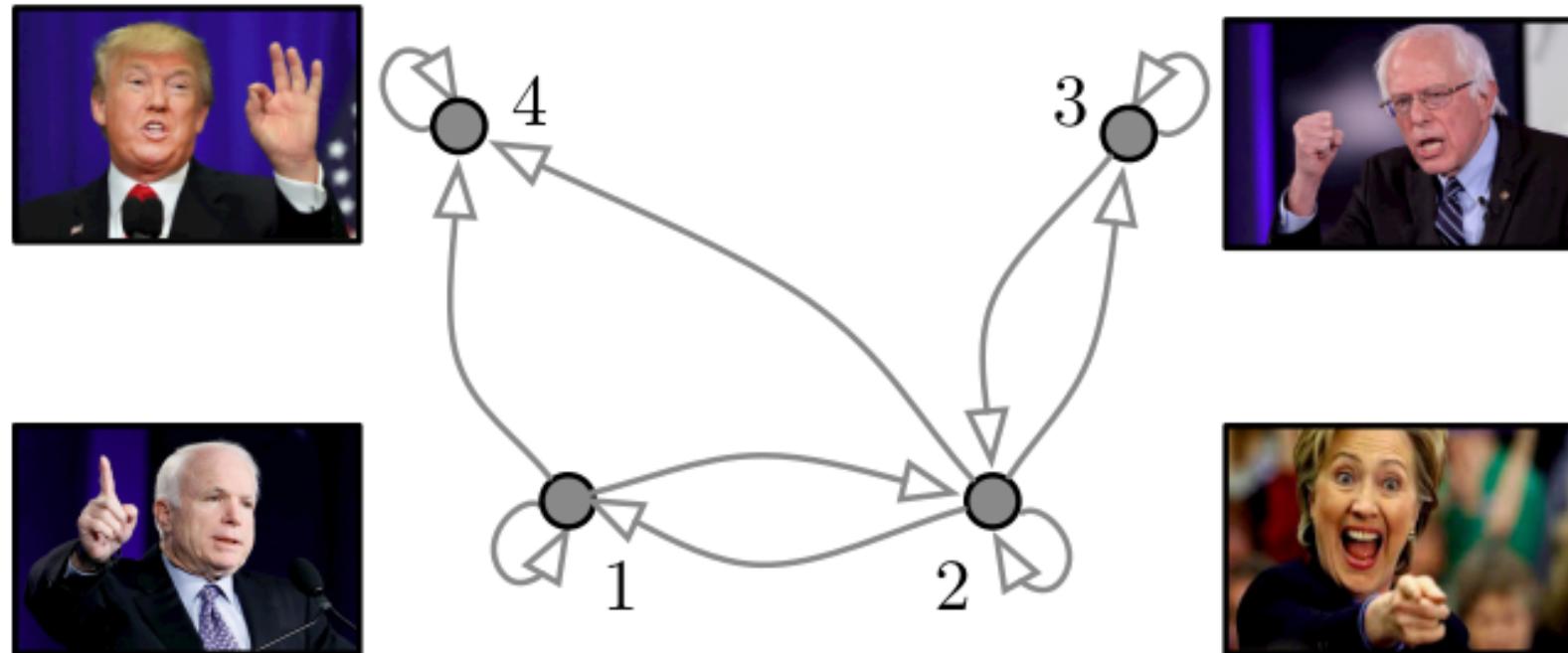
\Rightarrow yes ✓

Q2: What does it converge to?

$\Rightarrow \lim_{k \rightarrow \infty} x(k) = \mathbf{1} \cdot [\star \ \star \ \star \ \star] \cdot x_0$

Deduce convergence from graph: An example

What is the distinguishing feature? Remember this example!



$$A = \begin{pmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & \\ & & & * \end{pmatrix}$$

$$\lim_{k \rightarrow \infty} x(k) = \mathbf{1} \cdot [0 \ 0 \ 0 \ 1] \cdot x_0$$

$$A = \begin{pmatrix} * & * & * & * \\ * & * & * & * \\ * & * & * & \\ \color{red}{*} & & & * \end{pmatrix}$$

$$\lim_{k \rightarrow \infty} x(k) = \mathbf{1} \cdot [\color{red}{*} \ \color{red}{*} \ \color{red}{*} \ \color{red}{*}] \cdot x_0$$

Graph theory: basic concepts

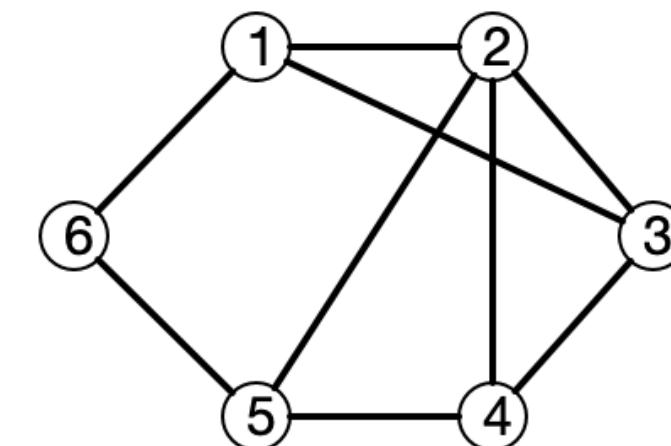
1. Basic elements: **nodes** (labels as $1, 2, \dots, n$)
links (undirected/directed, unweighted/weighted)

Unweighted $G = (V, E)$: node set $V = \{1, \dots, n\}$

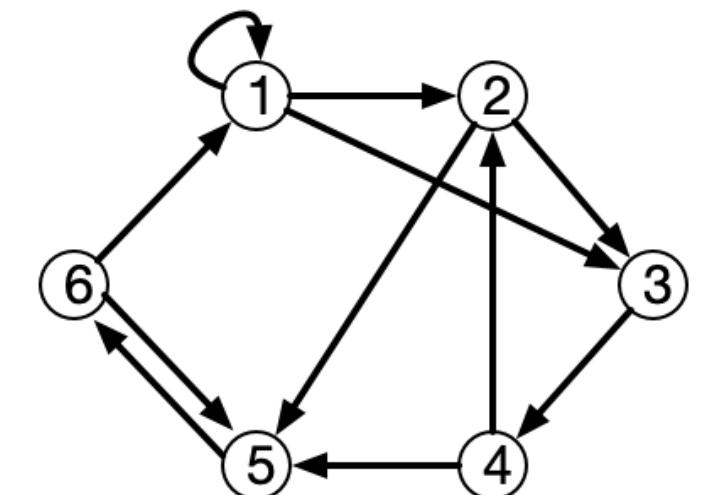
e.g. coauthor net. \leftarrow link set $E \subseteq \{\{i, j\} \mid i, j \in V\}$ (undirected graphs)

e.g. citation net. \leftarrow link set $E \subseteq V \times V = \{(i, j) \mid i, j \in V\}$ (directed)

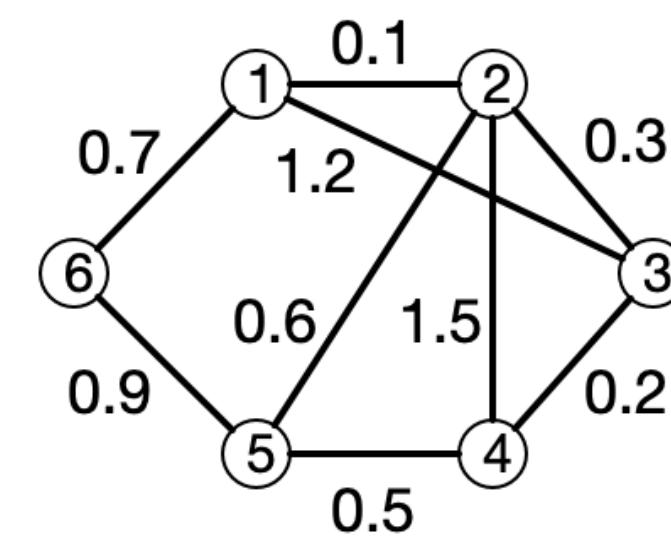
Weighted $G = (V, E, \{a_e\}_{e \in E})$: a_e is the weight of link $e \in E$
usually assumed to be **non-negative**



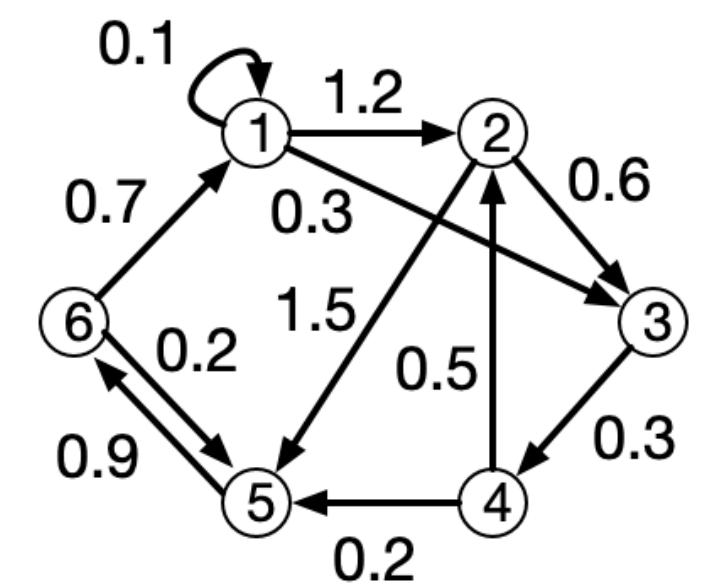
undirected unweighted



directed unweighted



undirected weighted



directed weighted

* Conventionally, self-loops are not allowed in undirected graphs.

Graph theory: basic concepts

$$G = (V, E, \{a_e\}_{e \in E})$$

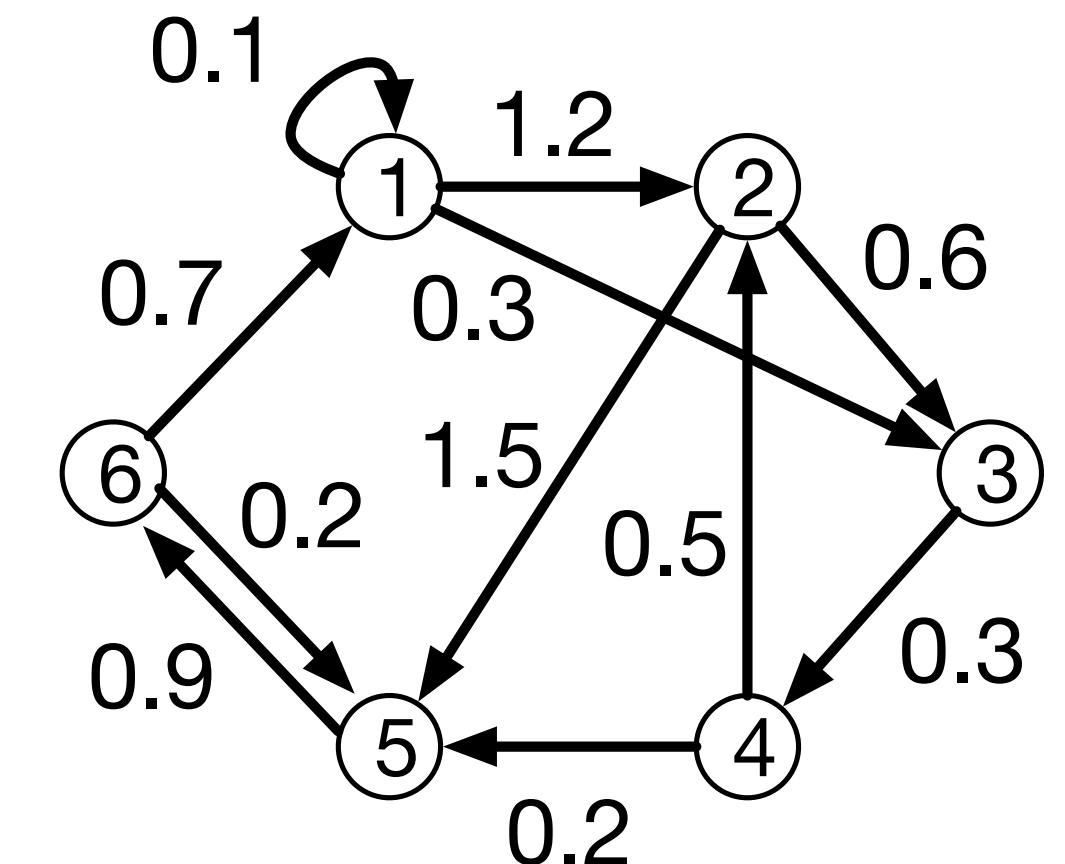
2. Subgraph $G' = (V', E')$: $V' \subseteq V, E' \subseteq E \cap \{\{i, j\} \mid i, j \in V'\}$ (undirected)

$$E' \subseteq E \cap V' \times V' \quad (\text{directed})$$

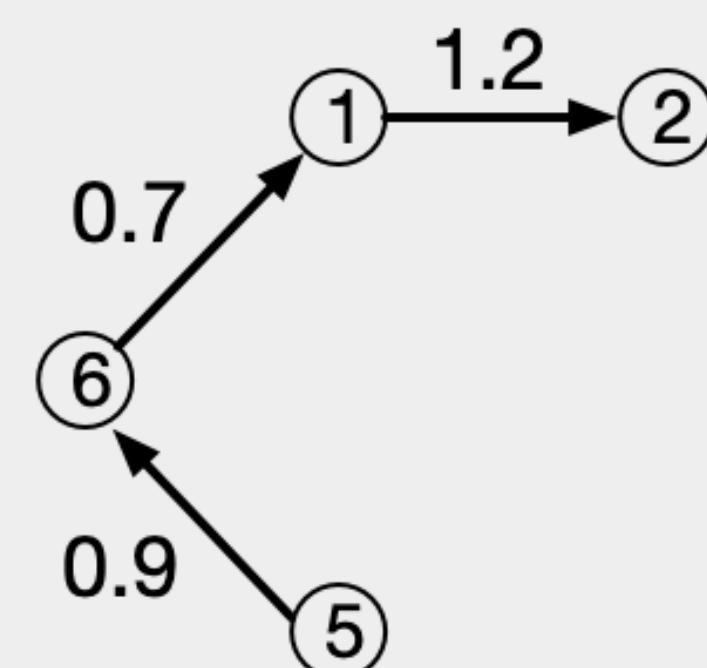
Induced subgraph $G' = (V', E')$: $V' \subseteq V, E' = E \cap \{\{i, j\} \mid i, j \in V'\}$ (undirected)

$$E' = E \cap (V' \times V') \quad (\text{directed})$$

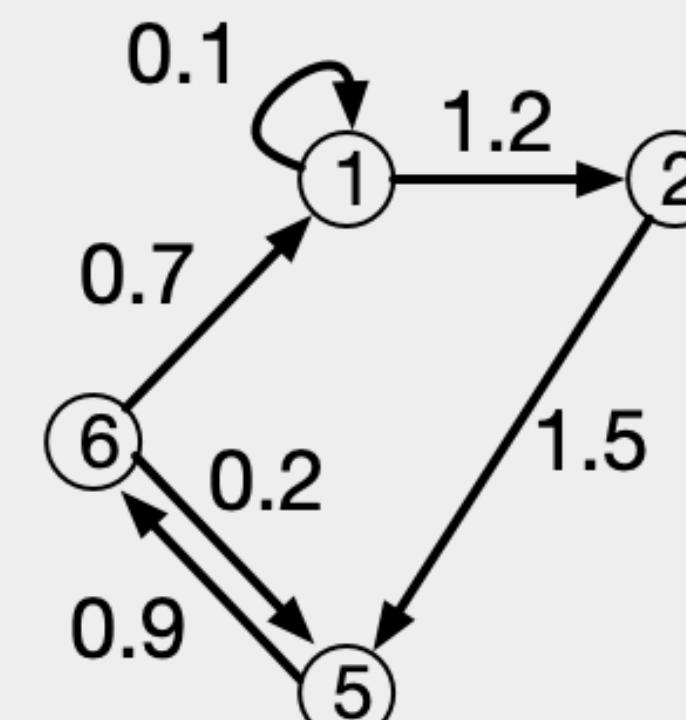
Spanning subgraph $G' = (V, E')$: $E' \subseteq E$



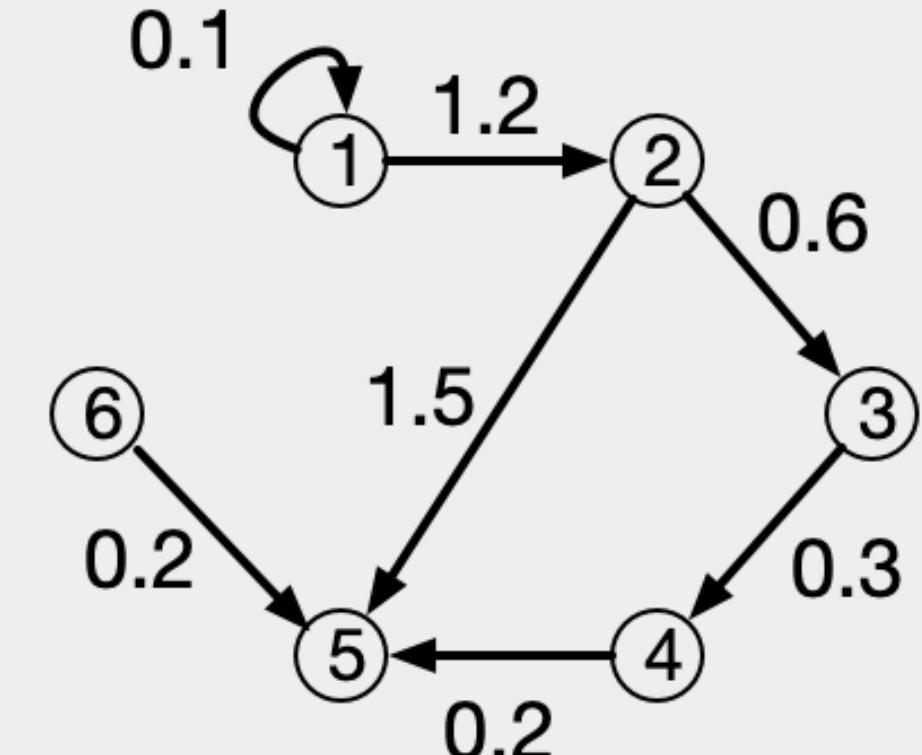
A subgraph



The induced subgraph



A spanning subgraph



Graph theory: basic concepts

1. Basic elements: **nodes** (labels as $1, 2, \dots, n$)

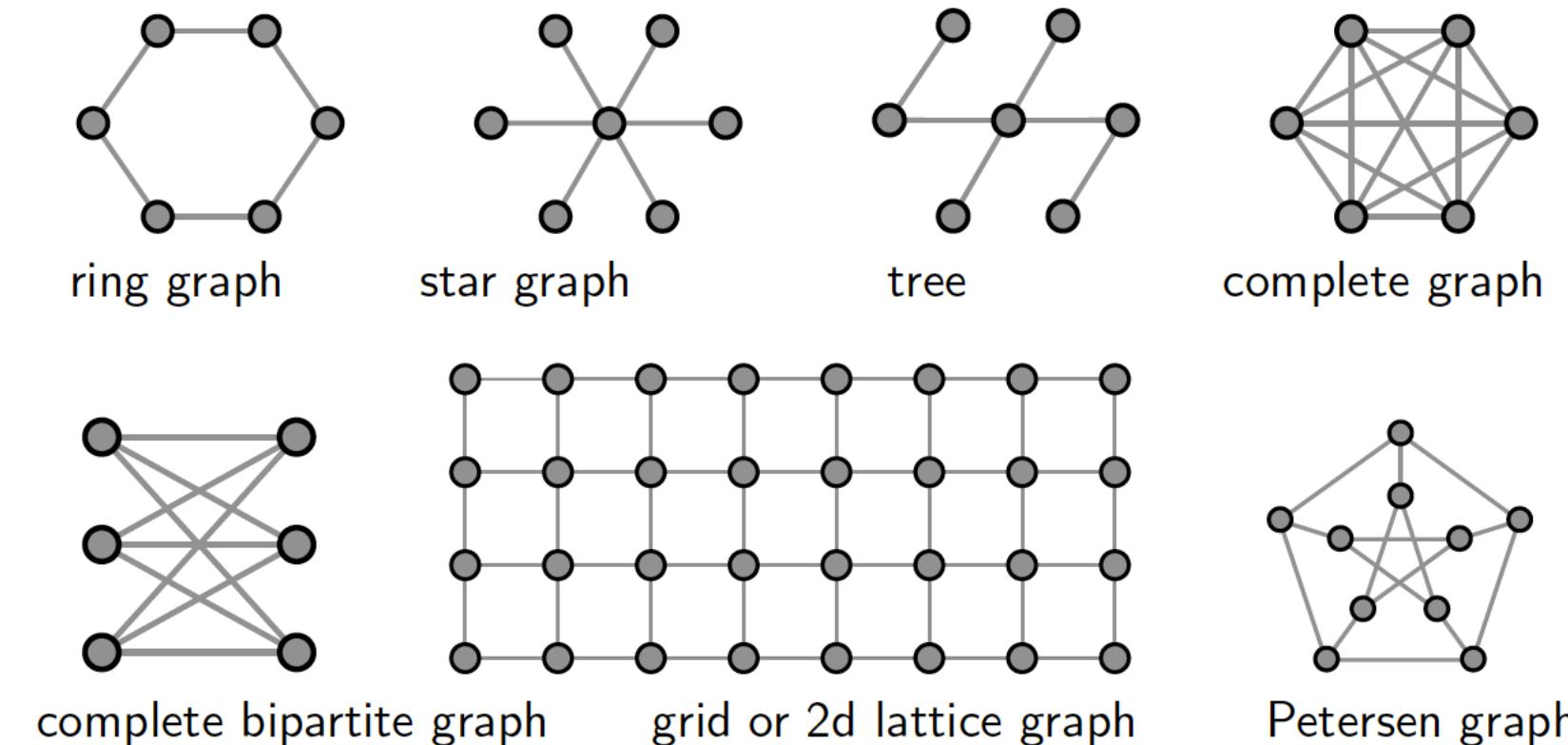
links (undirected/directed, unweighted/weighted)

Unweighted $G = (V, E)$: node set $V = \{1, \dots, n\}$

link set $E \subseteq \{\{i, j\} \mid i, j \in V\}$ (undirected graphs)

link set $E \subseteq V \times V = \{(i, j) \mid i, j \in V\}$ (directed)

Weighted $G = (V, E, \{a_e\}_{e \in E})$: a_e is the weight of link $e \in E$

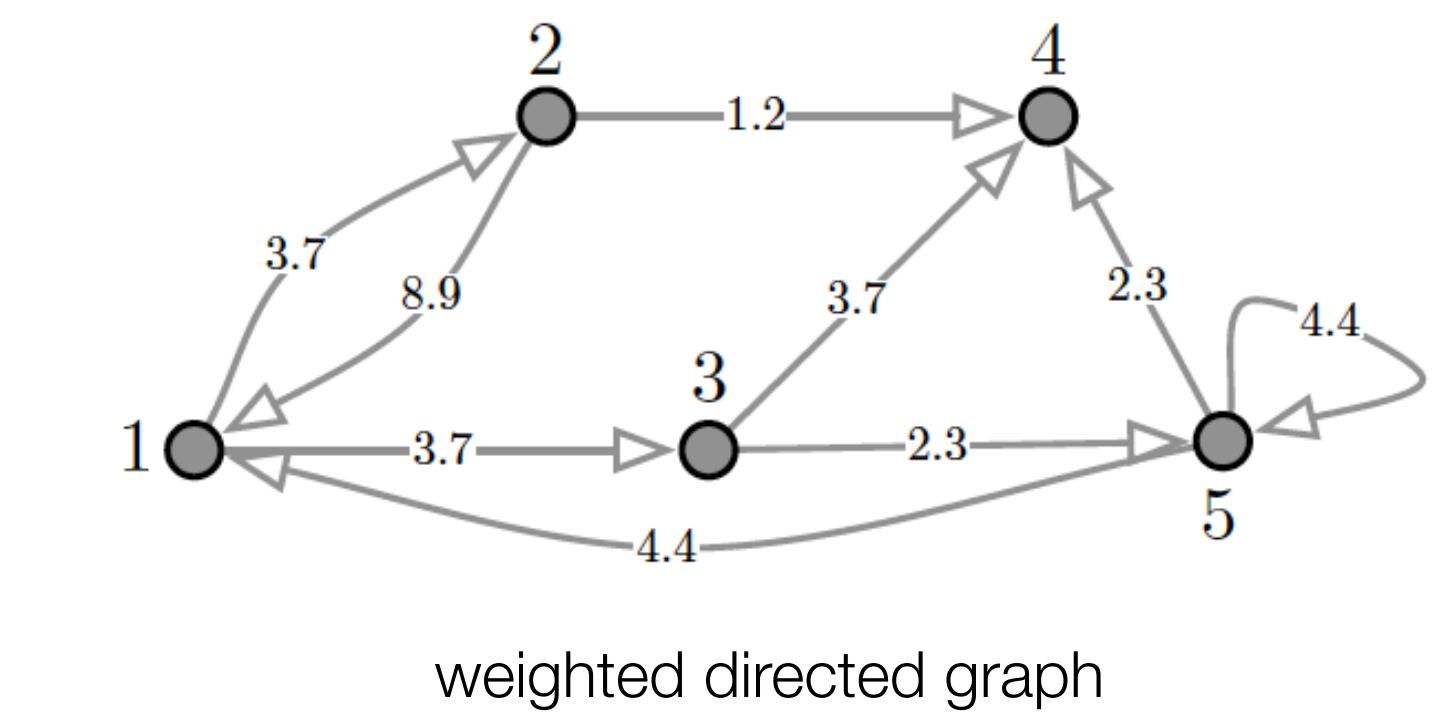
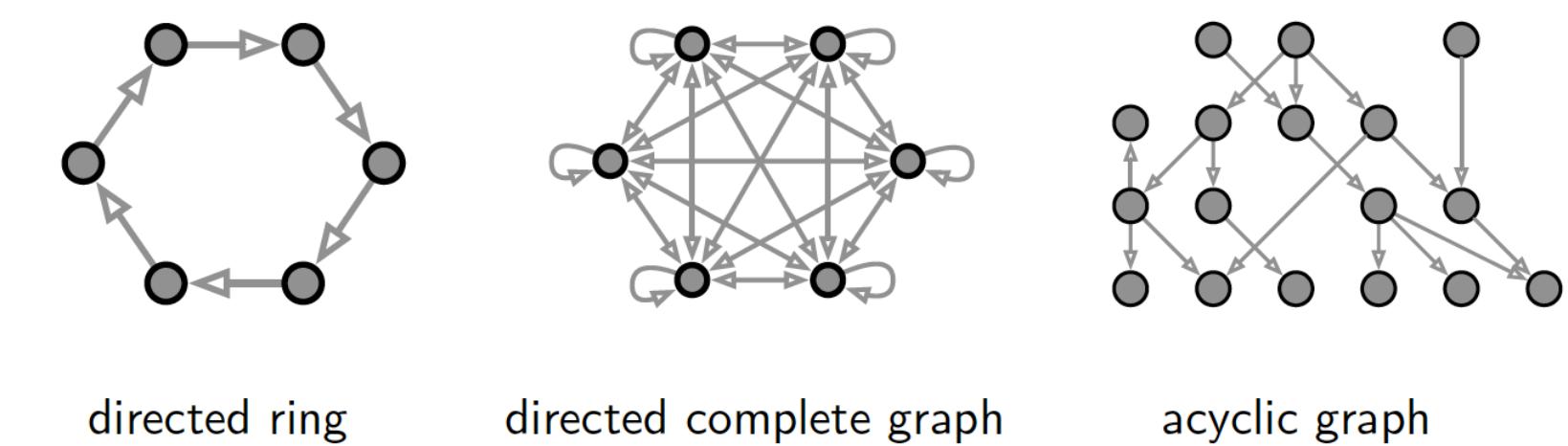


2. **Subgraph** $G' = (V', E')$: $V' \subseteq V, E' \subseteq E \cap \{\{i, j\} \mid i, j \in V'\}$ (undirected)

$E' \subseteq E \cap V' \times V'$ (directed)

Induced subgraph $G' = (V', E')$: $V' \subseteq V, E' = E \cap \{\{i, j\} \mid i, j \in V'\}$ (undirected)

$E' = E \cap (V' \times V')$ (directed)



Graph theory: basic concepts

3. Directed graphs (weighted or unweighted)

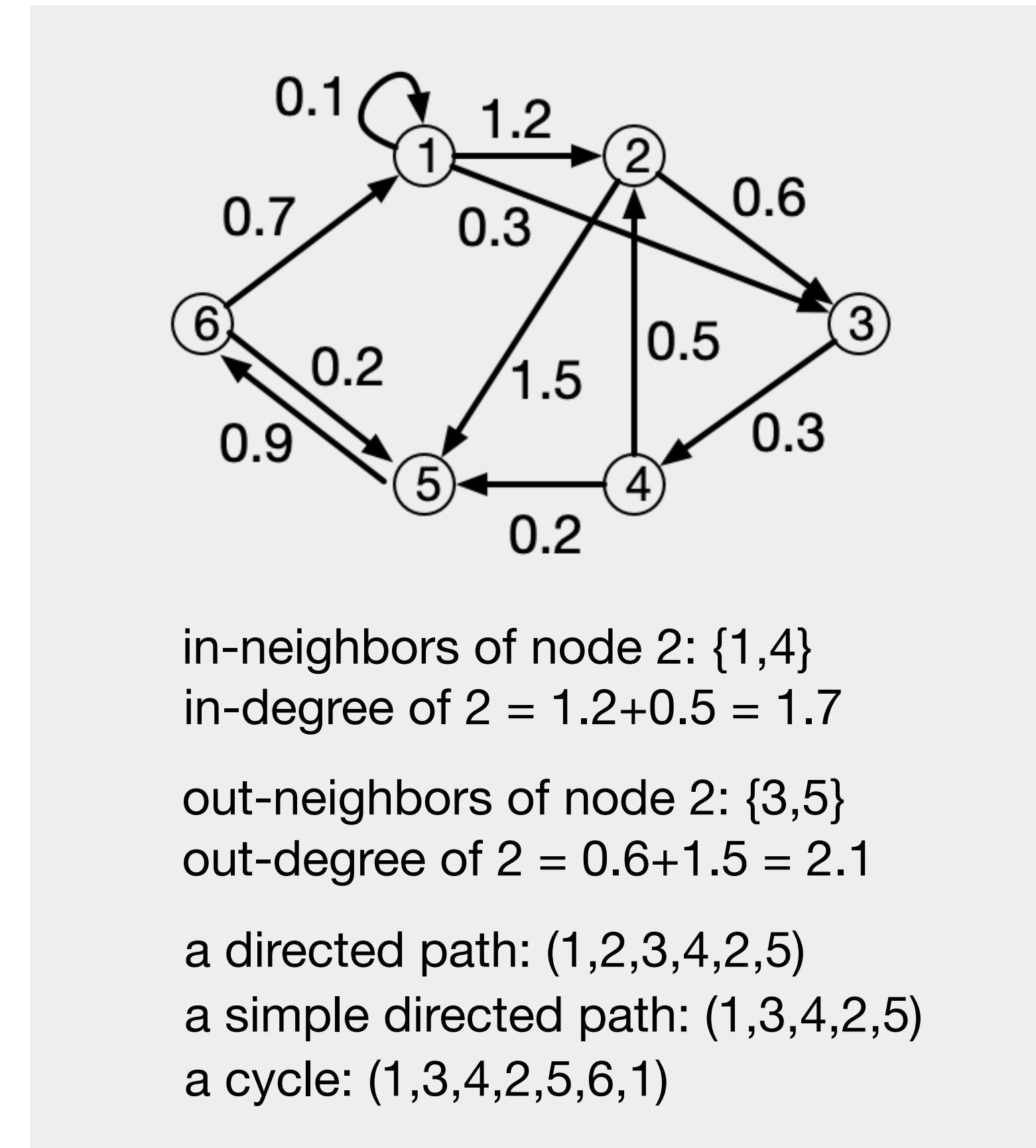
- **in-neighbor**: node j is an in-neighbor of node i if $(j, i) \in E$
- **out-neighbor**: node j is an out-neighbor of node i if $(i, j) \in E$
- **in-degree**: number of in-neighbors (unweighted)
sum of the weights of links from in-neighbors (weighted)
- **out-degree**: number of out-neighbors (unweighted)
sum of the weights of links to out-neighbors (weighted)
- **directed path**: an ordered sequence of nodes i_1, i_2, \dots, i_m such that
$$(i_k, i_{k+1}) \in E, \text{ for any } k = 1, 2, \dots, m - 1$$

simple directed path: no repeating node, except possibly for $i_1 = i_m$

- **cycle**: simple directed path with $i_1 = i_m$, with **length** equal to m
- **acyclic graph**: a directed graph with no cycle

* A self loop at node i makes i both its own in-neighbor and own out-neighbour.

Please carefully read Ch. 3 of Prof. Bullo's lecture notes regarding undirected graphs & concepts not covered in this lecture.



Graph theory: basic concepts

3. Directed graphs (weighted or unweighted)

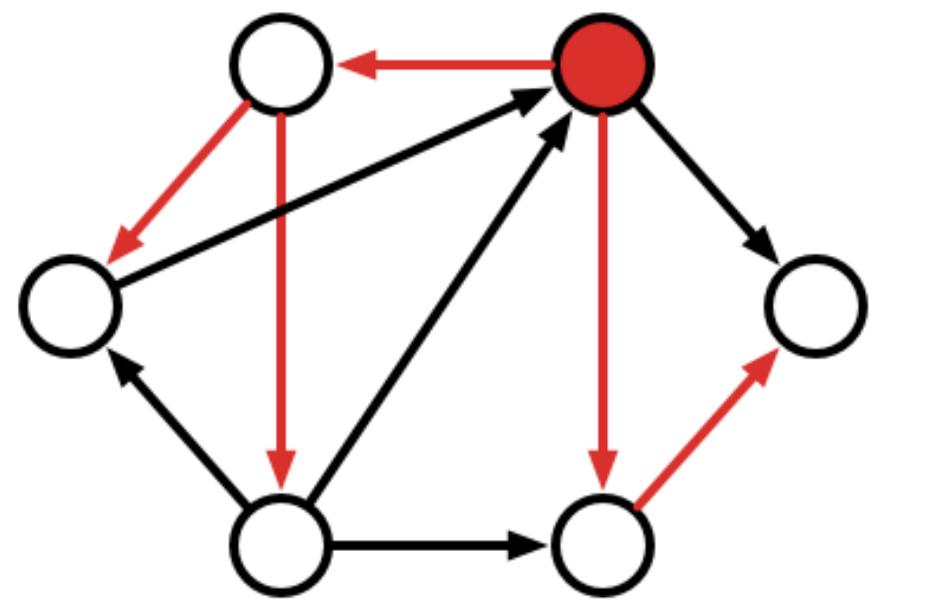
— **source**: node with no in-neighbor

sink: node with no out-neighbor

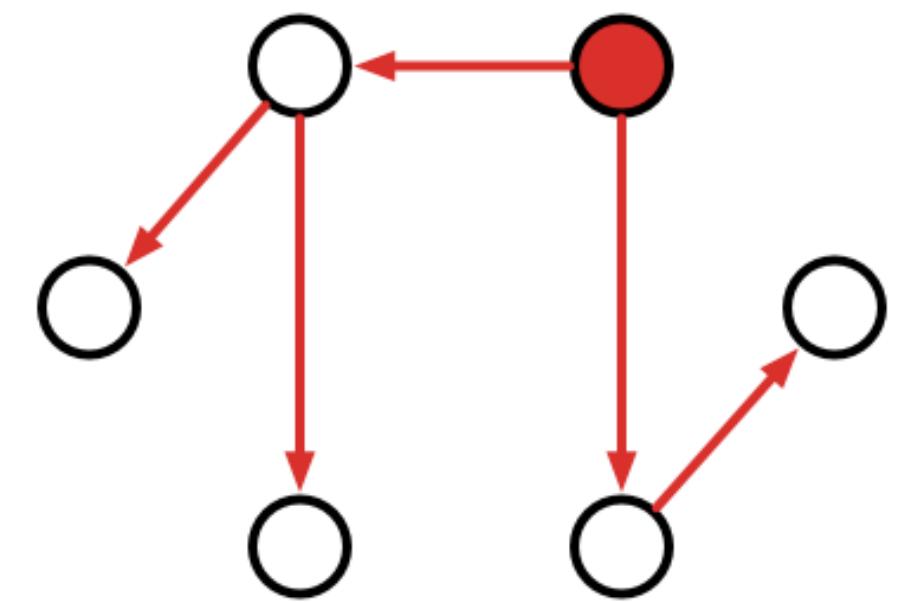
— **directed tree**: an **acyclic** graph satisfying: \exists a node (the **root**), such that it can reach any other node by one and only one directed path.

e.g. academic genealogy: $i \rightarrow j \Leftrightarrow$ individual i is the PhD advisor of j

— **directed spanning tree** of a directed graph: spanning subgraph & directed tree

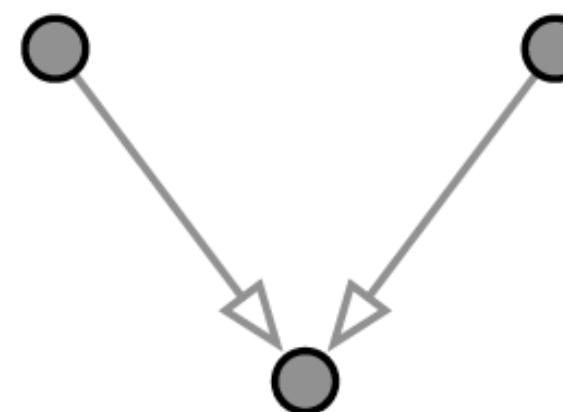


$$G = (V, E)$$



$$G' = (V, \{\text{red links}\})$$

G' is a directed spanning tree of G
(but not the only one).

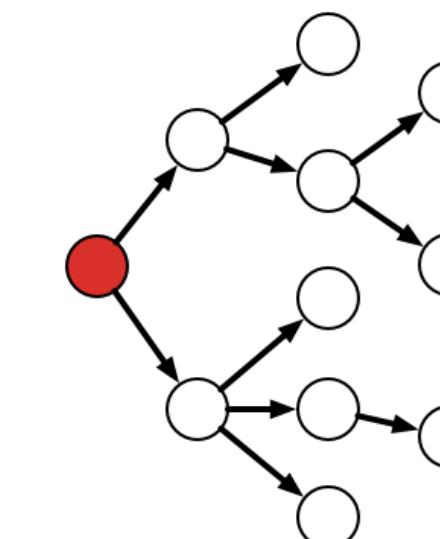


(a) An acyclic digraph with one sink and two sources

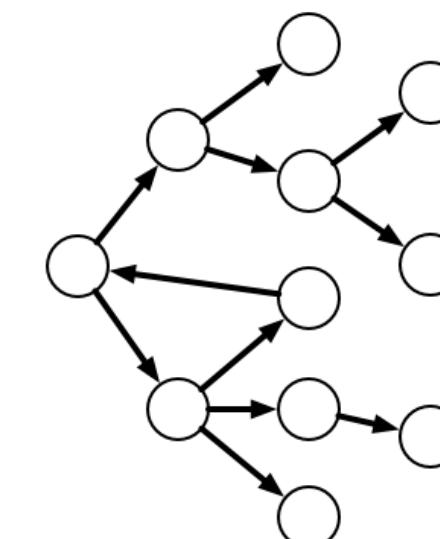
Q1: Every acyclic digraph has at least one source and at least one sink. Why?

Q2: The root is the only source in a directed tree. Why?

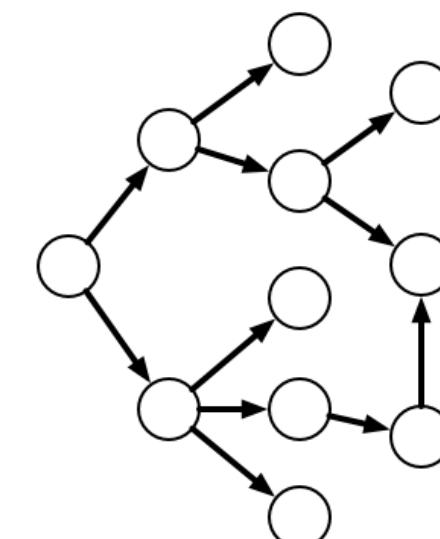
Q3: Does every digraph have a directed spanning tree?



A directed tree



Not a directed tree



Not a directed tree

Graph theory: basic concepts

3. Directed graphs (weighted or unweighted)

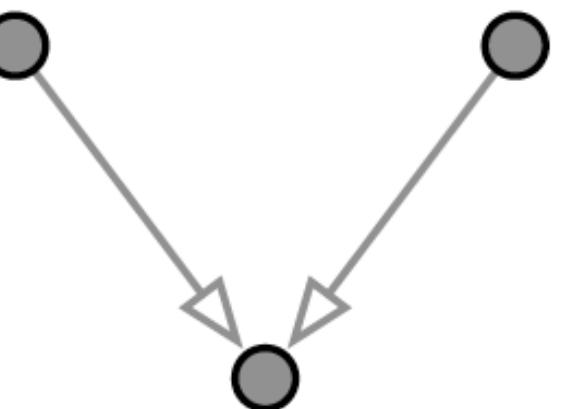
— **source**: node with no in-neighbor

sink: node with no out-neighbor

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e.g. academic genealogy: $i \rightarrow j \Leftrightarrow$ individual i is the PhD advisor of j

— **directed spanning tree** of a directed graph: spanning subgraph & directed tree



(a) An acyclic digraph with one sink and two sources

Q1: Every acyclic digraph has at least one source and at least one sink. Why?

Q2: The root is the only source in a directed tree. Why?

Q3: Does every digraph have a directed spanning tree?

Graph theory: basic concepts

3. Directed graphs (weighted or unweighted)

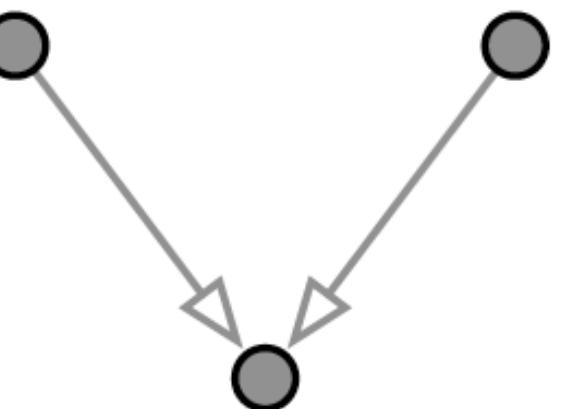
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(a) An acyclic digraph with one sink and two sources

Q1: Every acyclic digraph has at least one source and at least one sink. Why?

Q2: The root is the only source in a directed tree. Why?

Q3: Does every digraph have a directed spanning tree?

Graph theory: basic concepts

3. Directed graphs (weighted or unweighted)

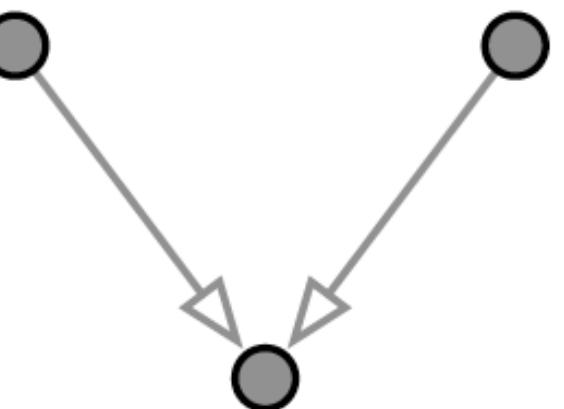
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(a) An acyclic digraph with one sink and two sources

Q1: Every acyclic digraph has at least one source and at least one sink. Why?

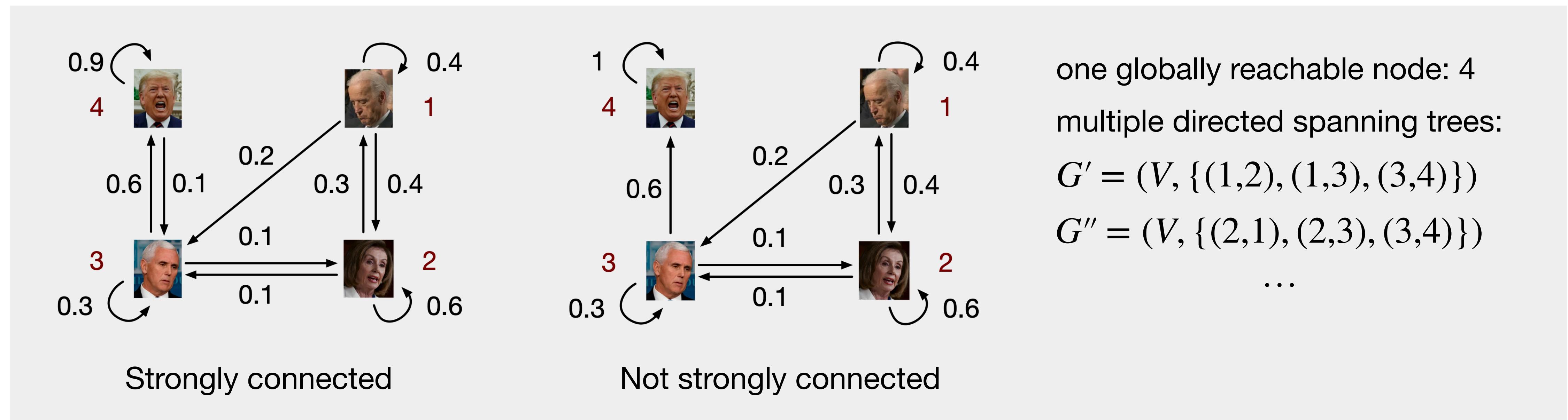
Q2: The root is the only source in a directed tree. Why?

Q3: Does every digraph have a directed spanning tree?

Graph theory: basic concepts

4. Connectivity properties of directed graphs

- G is strongly connected: \exists at least one directed path from any node to any other node
- G has a globally reachable node: \exists a node that can be reached from any other node via at least one directed path
- G has a directed spanning tree $\Leftrightarrow \exists$ a node that can reach any other node via at least one directed path



Graph theory: basic concepts

4. Connectivity properties of directed graphs

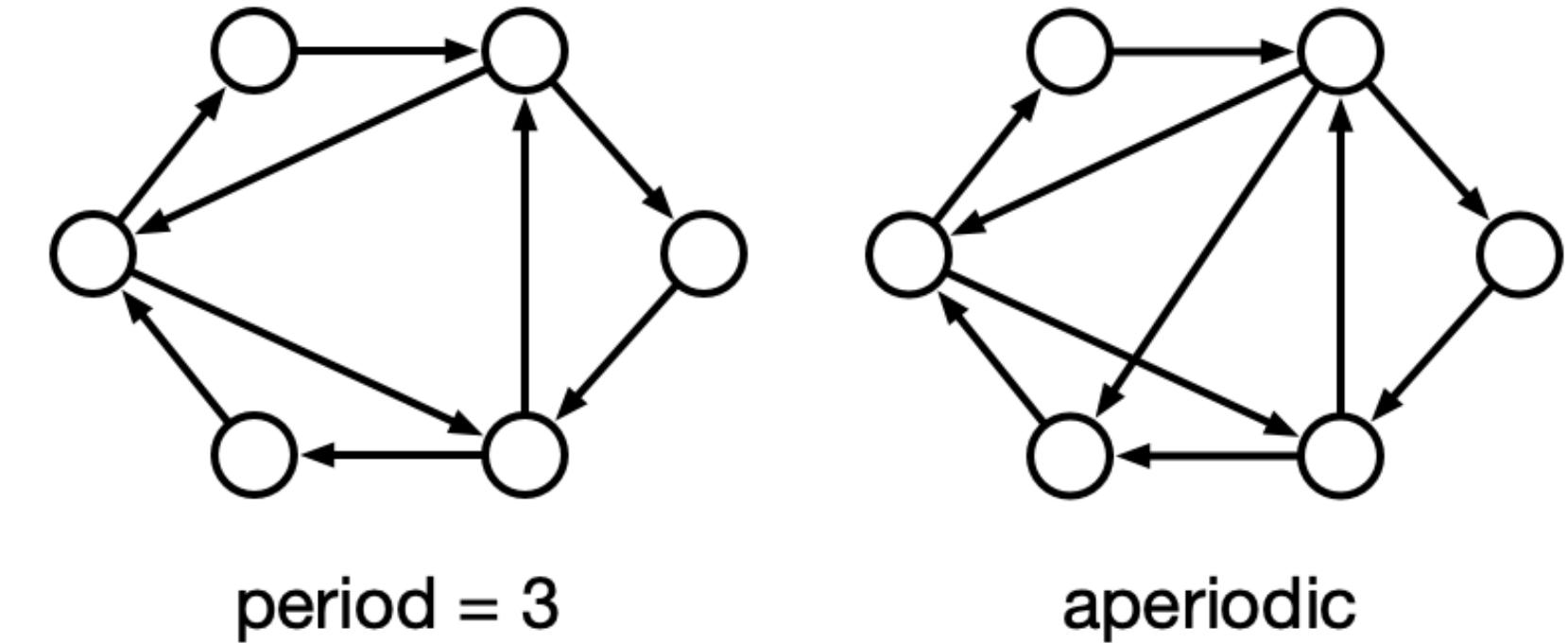
- G is strongly connected: \exists at least one directed path from any node to any other node
- G has a globally reachable node: \exists a node that can be reached from any other node via at least one directed path
- G has a directed spanning tree $\Leftrightarrow \exists$ a node that can reach any other node via at least one directed path

A strongly connected graph always has at least one directed spanning tree.

Graph theory: basic concepts

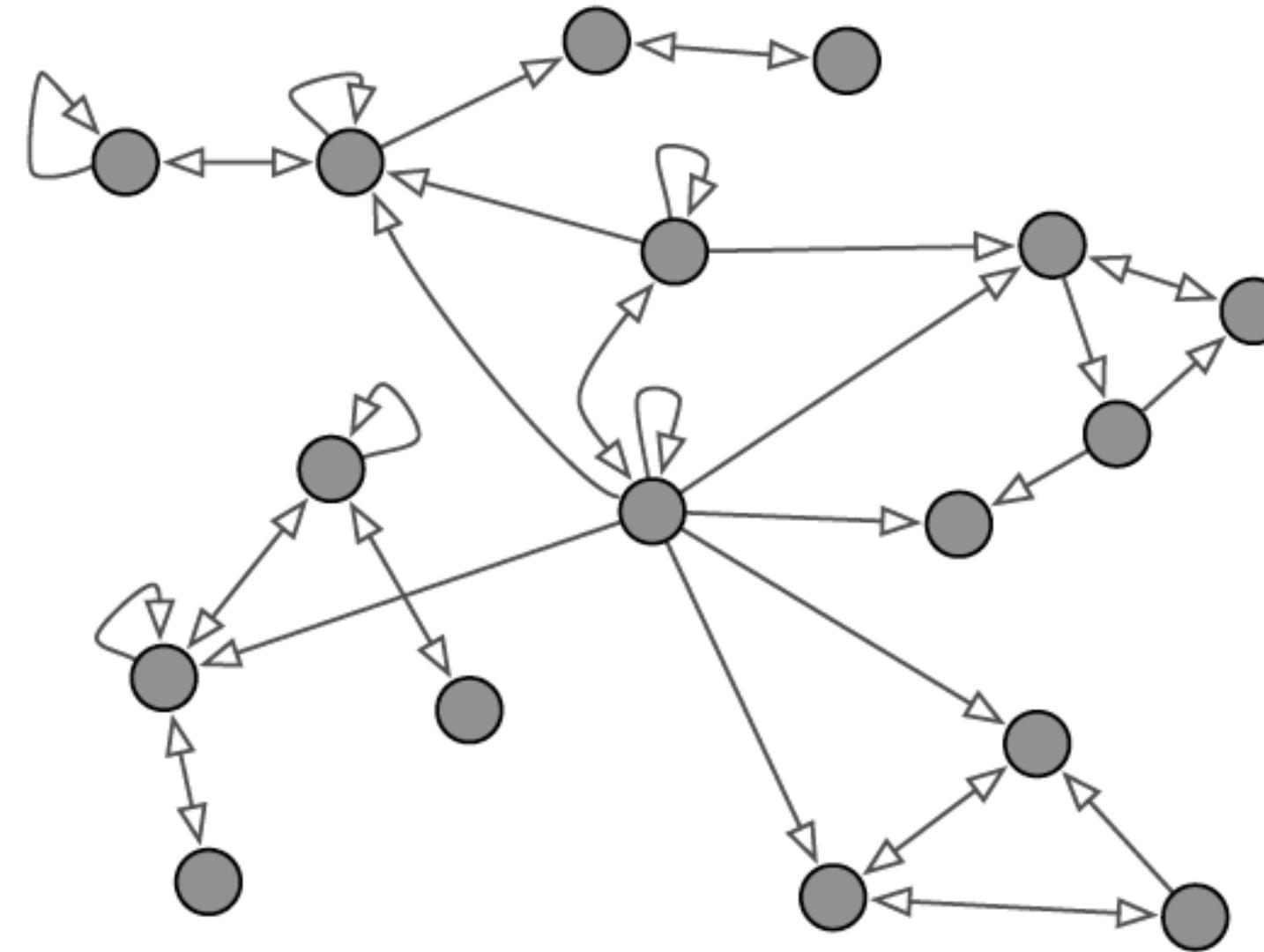
5. Periodicity of strongly connected directed graphs

- **period**: the largest common divisor of the lengths of all the cycles
- G is **periodic** if its period > 1
- G is **aperiodic** if its period $= 1$

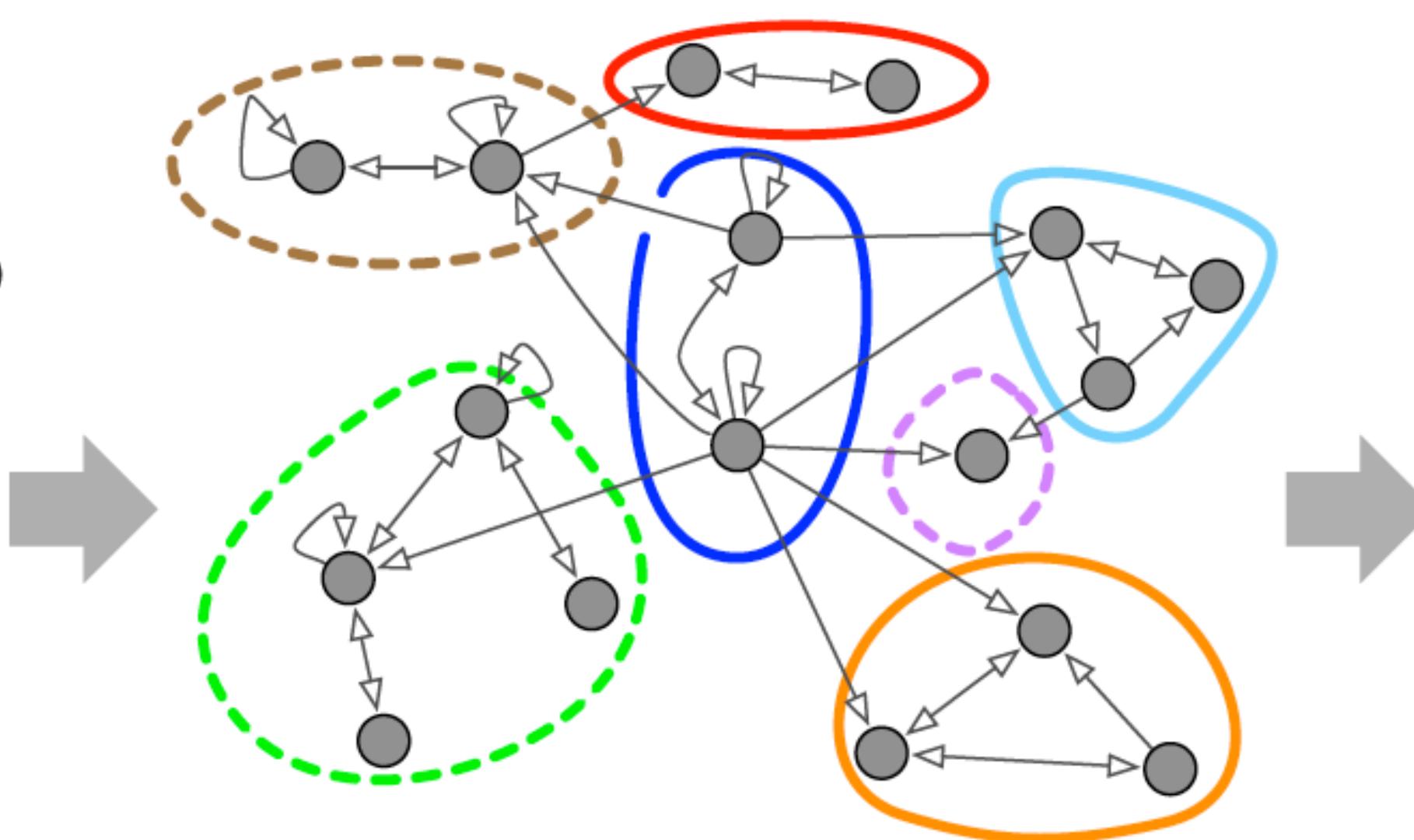


6. Condensation of directed graphs

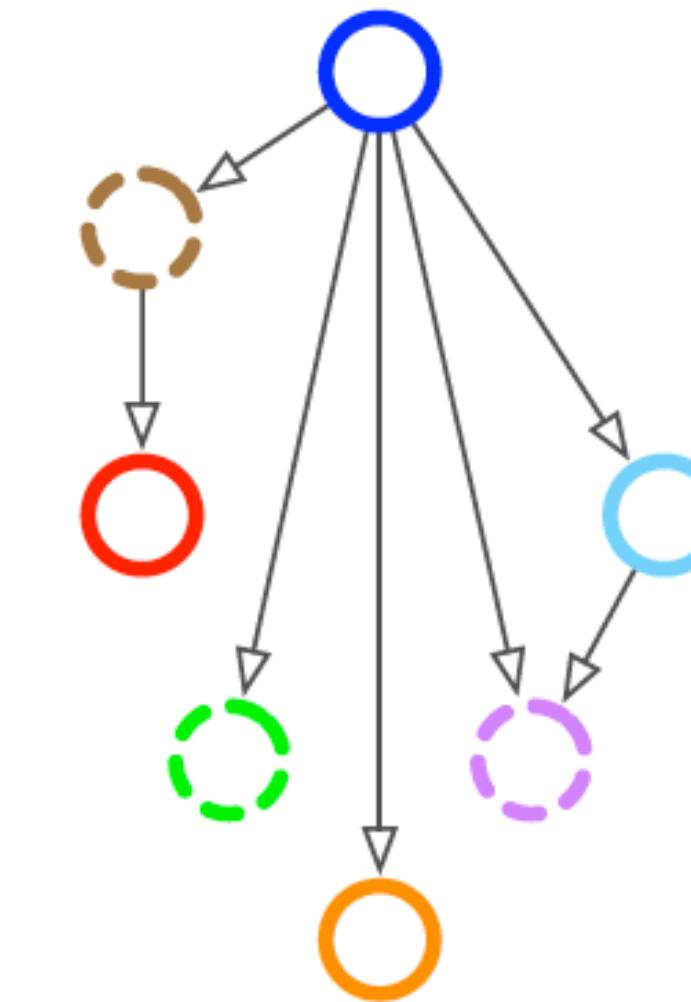
- **strongly connected component**: induced subgraph H & strongly connected
& any other subgraph strictly containing H is not strongly connected
- **condensation digraph**: $C(G)$ is a directed graph
 - Each node H is a strongly connected component in G
 - \exists a link from H_i to H_j if in G there exists a link from one node in SCC H_i to another node in SCC H_j



(a) An example digraph G



(b) The strongly connected components of the digraph G



(c) The condensation digraph $C(G)$

Figure 3.7: An example digraph, its strongly connected components and its condensation digraph.

Properties of the condensation digraph:

- 1) $C(G)$ is acyclic;
- 2) G has a globally reachable node $\Leftrightarrow C(G)$ has a globally reachable node $\Leftrightarrow C(G)$ has a unique sink;
- 3) G has a directed spanning tree $\Leftrightarrow C(G)$ has a directed spanning tree $\Leftrightarrow C(G)$ has a unique source;

Algebraic graph theory

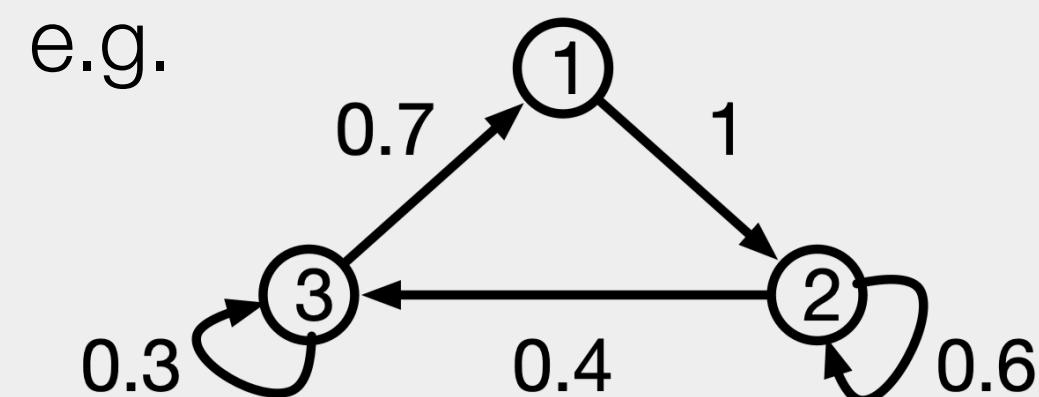
- graph properties \leftrightarrow matrix properties
- graph properties \leftrightarrow (matrix properties) \leftrightarrow dynamic behaviour of $x(t+1) = Ax(t)$ (Next lecture)

1. adjacency matrix: matrix representation of directed weighted graphs

Graph $G = (V, E, \{a_e\}_{e \in E})$

\leftrightarrow

Adjacency matrix $A = (a_{ij})_{n \times n} \in \mathbb{R}_{\geq 0}^{n \times n}$, where



$$a_{ij} = \begin{cases} \text{the weight of the link from } i \text{ to } j, & \text{if } (i, j) \in E \\ 0, & \text{if } (i, j) \notin E \end{cases}$$

\rightarrow

Binary adjacency matrix $[A] = ([a]_{ij})_{n \times n} \in \{0, 1\}^{n \times n}$, where

$$[a]_{ij} = \begin{cases} 1, & \text{if } (i, j) \in E \\ 0, & \text{if } (i, j) \notin E \end{cases}$$

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0.6 & 0.4 \\ 0.7 & 0 & 0.3 \end{bmatrix} \quad [A] = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix}$$

* $[A]$ is just for simplicity of notations in this lecture, not a conventional symbol for binary adjacency matrices.

Algebraic graph theory

2. properties of graphs \leftrightarrow properties of adjacency matrices

(1) in-neighbor set of node i , out-neighbour set of node i

(2) in-degree of node i , out-degree of node i

(3) strongly connected

(4) The out-degree of every node is 1

(1) $\{j \in V \mid a_{ji} > 0\}$, $\{j \in V \mid a_{ij} > 0\}$

$$(2) \sum_{j=1}^n a_{ji} = (1_n^\top A)_i, \quad \sum_{j=1}^n a_{ij} = (A 1_n)_i$$

$$(3) I + A + A^2 + \dots + A^{n-1} > 0$$

(4) A is row-stochastic

Key Results:

Corollary 4.5 (Connectivity properties of the digraph and positive powers of the adjacency matrix: cont'd). Let G be a weighted digraph with n nodes, weighted adjacency matrix A and a self-loop at each node. The following statements are equivalent:

(i) G is strongly connected; and

(ii) A^{n-1} is positive, so that A is primitive.

Algebraic graph theory

2. properties of graphs \leftrightarrow properties of adjacency matrices

Lemma 4.2 (Directed paths and powers of the adjacency matrix). *Let G be a weighted digraph with n nodes, with adjacency matrix A and binary adjacency matrix $A_{0,1} \in \{0, 1\}^{n \times n}$. For all $i, j \in \{1, \dots, n\}$ and $k \in \mathbb{N}$*

- (i) *the (i, j) entry of $A_{0,1}^k$ equals the number of paths of length k from node i to node j ; and*
- (ii) *the (i, j) entry of A^k is positive if and only if there exists a path of length k from node i to node j .*

(Paths here are directed paths that possibly include self-loops.)

Lemma 4.4 (Global reachability and powers of the adjacency matrix). *Let G be a weighted digraph with $n \geq 2$ nodes and weighted adjacency matrix A . For any $j \in \{1, \dots, n\}$, the following statements are equivalent:*

- (i) *the j th node of G is globally reachable, and*
- (ii) *the j th column of $\sum_{k=0}^{n-1} A^k$ is positive.*

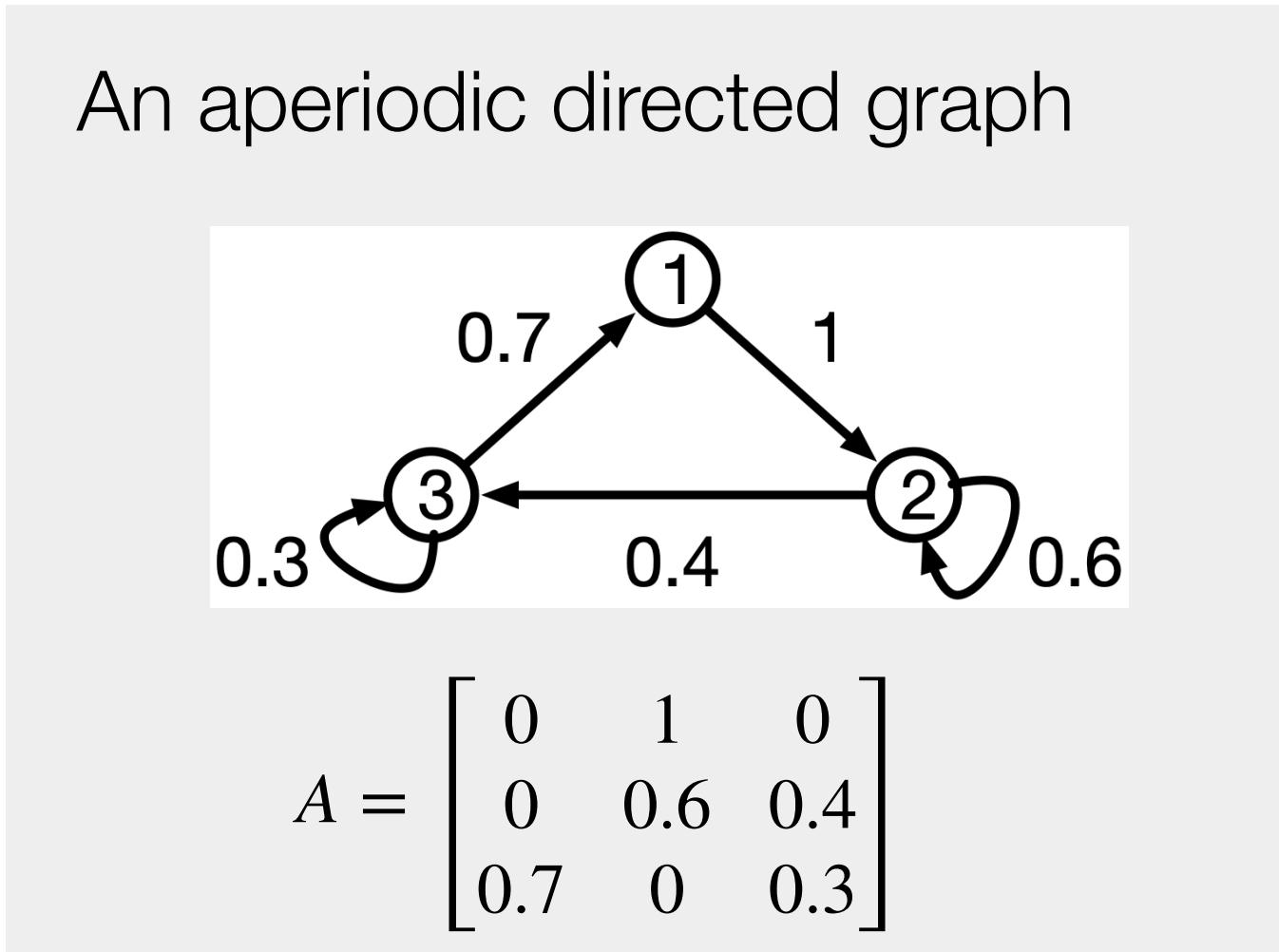
Algebraic graph theory

2. properties of graphs \leftrightarrow properties of adjacency matrices

Theorem 4.7 (Strongly connected and aperiodic digraphs and primitive adjacency matrices). *Let G be a weighted digraph with $n \geq 2$ nodes and with weighted adjacency matrix A . The following two statements are equivalent:*

- (i) G is strongly connected and aperiodic; and
- (ii) A is primitive, that is, there exists $k \in \mathbb{N}$ such that A^k is positive.

Example:

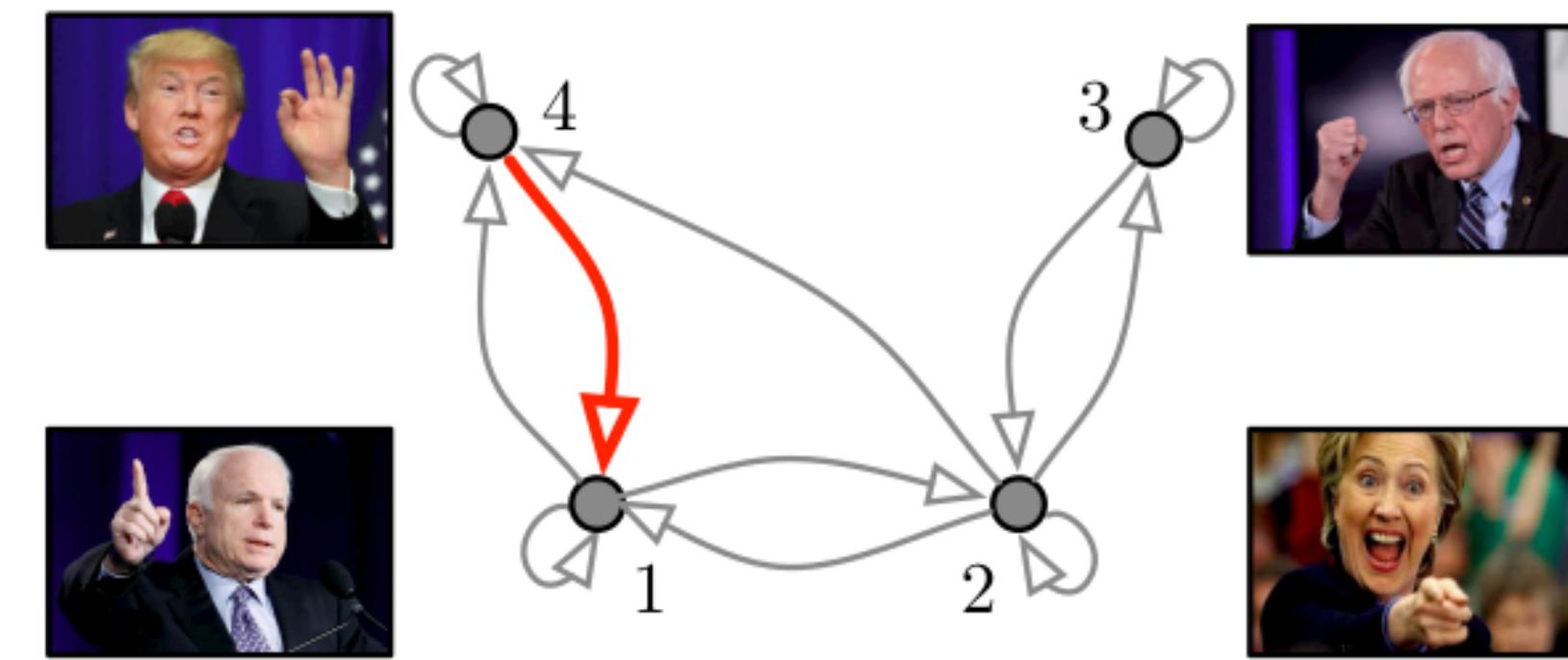
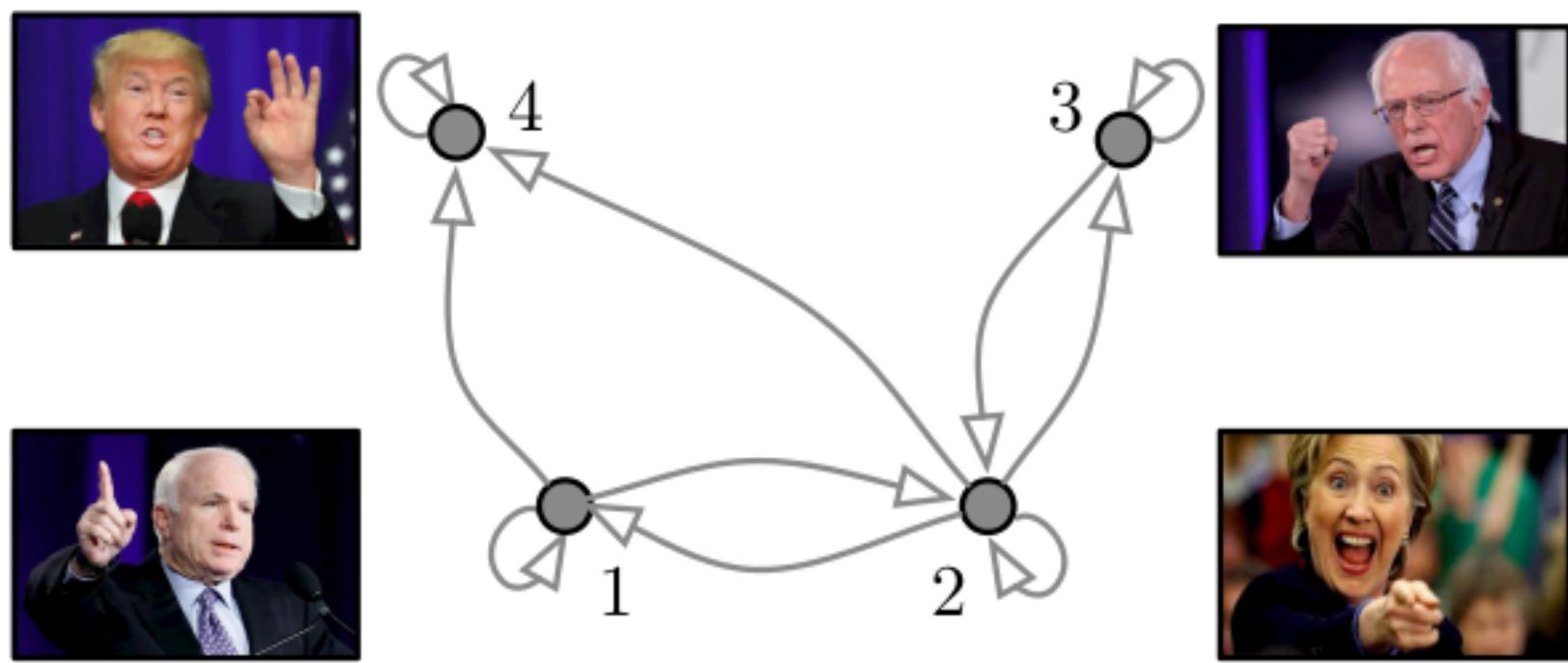


Graph is strongly connected and aperiodic \rightarrow there should be a $k \in \mathbb{N}$ such that A^k is positive!

$$\text{Compute } A^2 = \begin{bmatrix} 0 & 0.6 & 0.4 \\ 0.28 & 0.36 & 0.36 \\ 0.21 & 0.7 & 0.09 \end{bmatrix} \rightarrow \text{Not positive!}$$

$$\text{Compute } A^3 = \begin{bmatrix} 0.28 & 0.36 & 0.36 \\ 0.252 & 0.496 & 0.252 \\ 0.063 & 0.63 & 0.307 \end{bmatrix} \rightarrow \text{Positive!}$$

Remember this example? Do you see the distinguishing feature now?



$$A = \begin{pmatrix} \star & \star & \star & \star \\ \star & \star & \star & \star \\ \star & \star & \star & \\ & & & \star \end{pmatrix}$$

$$\lim_{k \rightarrow \infty} x(k) = \mathbf{1} \cdot [0 \ 0 \ 0 \ 1] \cdot x_0$$

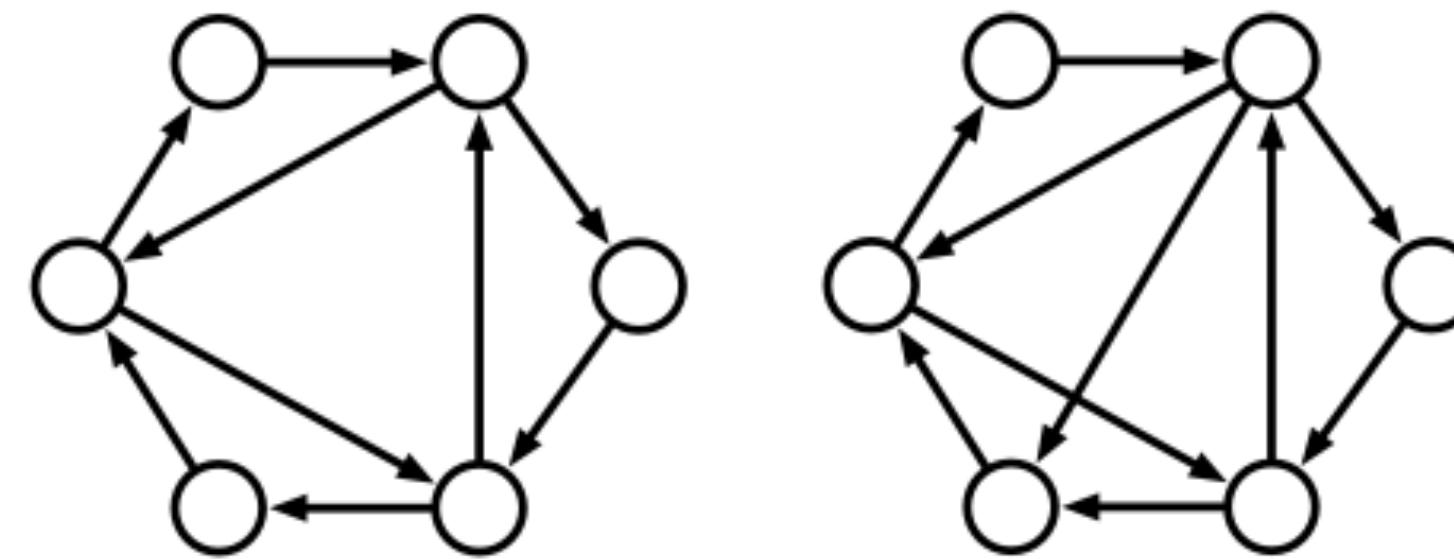
$$A = \begin{pmatrix} \star & \star & \star & \star \\ \star & \star & \star & \star \\ \star & \star & \star & \\ \star & & & \star \end{pmatrix}$$

$$\lim_{k \rightarrow \infty} x(k) = \mathbf{1} \cdot [\star \ \star \ \star \ \star] \cdot x_0$$

Summary

1. basic concepts in graph theory
 - different types of graphs
 - neighbour, degree, subgraphs
 - path, cycle, period, tree, etc
2. connectivity conditions
 - strongly connected
 - spanning tree, globally reachable node
3. algebraic graph theory I: describe graph properties
 - adjacency matrix
 - graph properties \leftrightarrow matrix properties, e.g., existence of paths, number of path, strongly connected, aperiodic

Next lecture: $x(t + 1) = Ax(t)$, convergence? consensus?



Algebraic graph theory II

Infer the behaviour of $x(t + 1) = Ax(t)$
directly from the graph structure.