

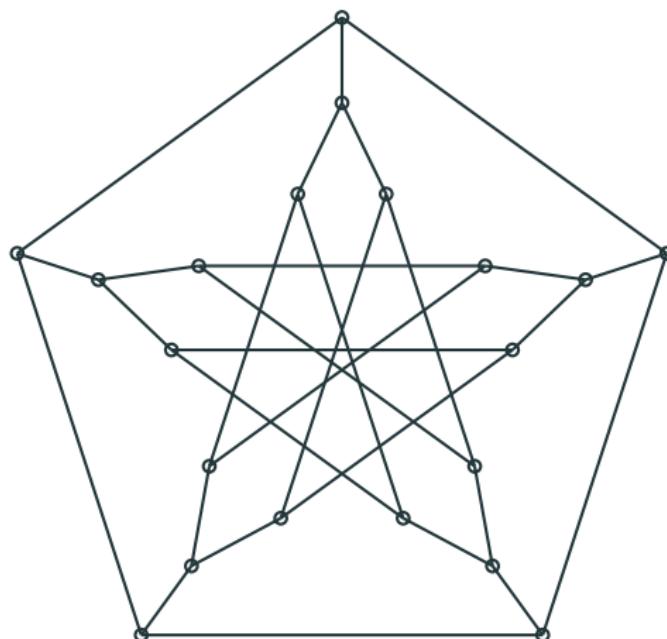
COORDINATION AND CONTROL OF MULTI-AGENT SYSTEMS

086730

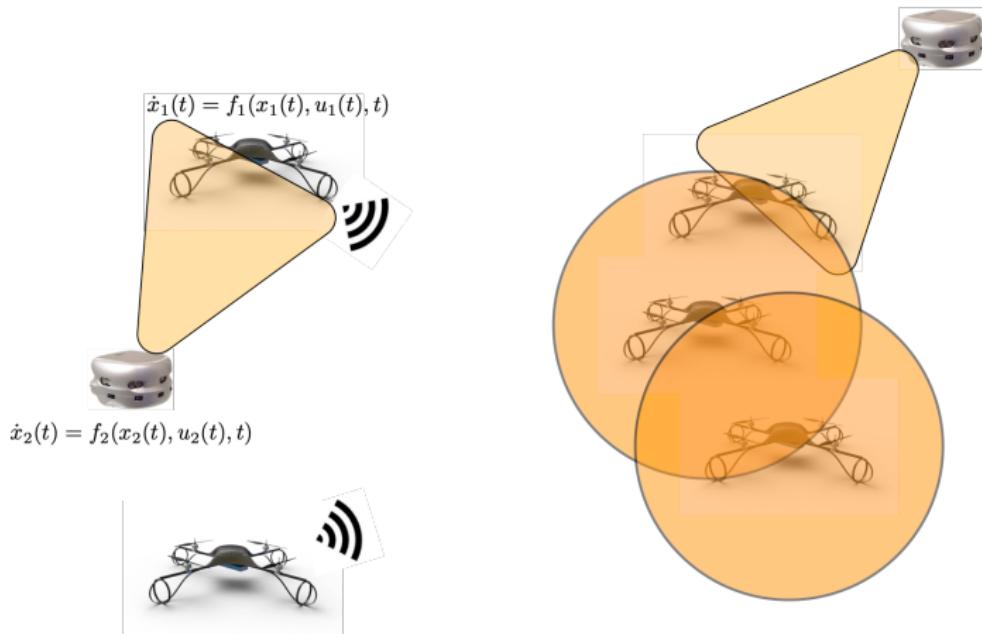
Daniel Zelazo

October 25, 2025

Introduction to Graph Theory

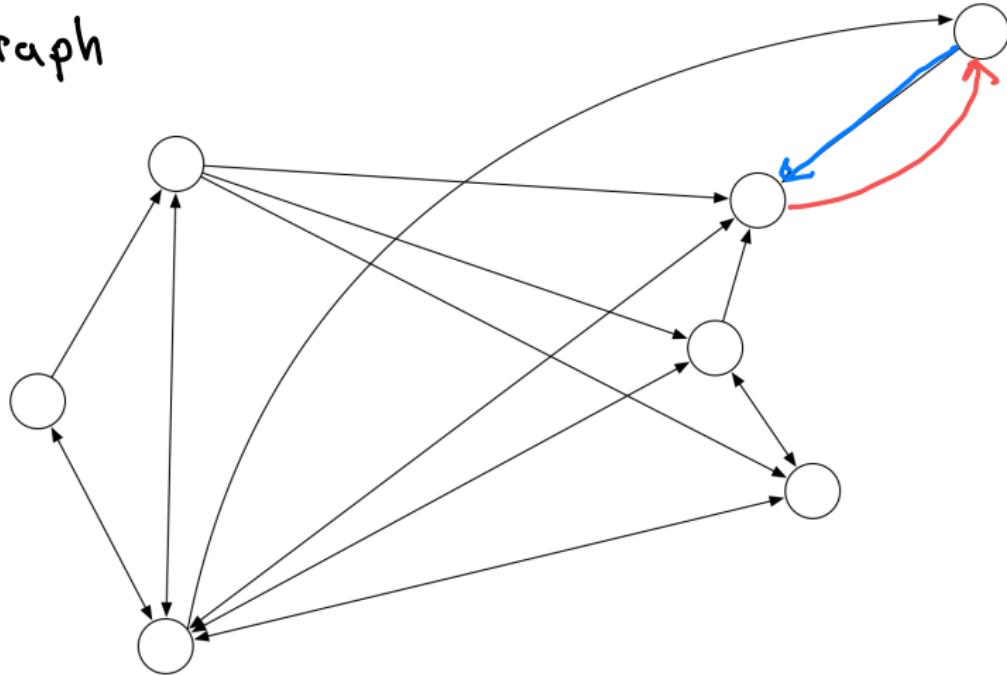


ABSTRACTION USING GRAPHS



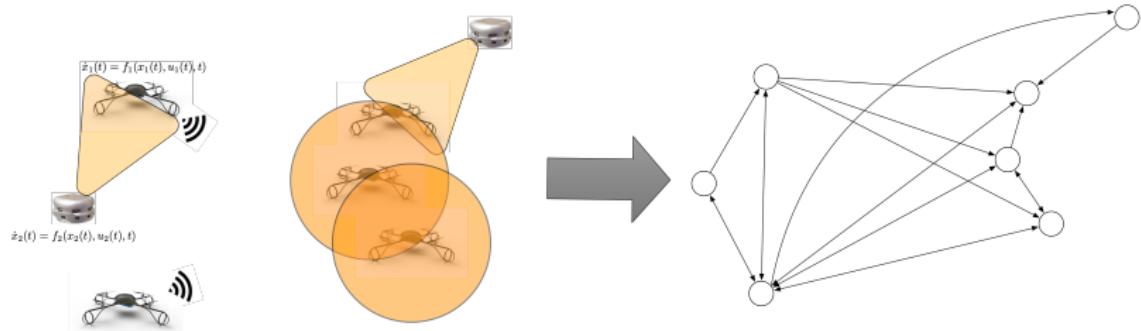
ABSTRACTION USING GRAPHS

Graph



- ○ - **nodes / vertices**
- → - **edges** (directed or undirected)
links

ABSTRACTION USING GRAPHS



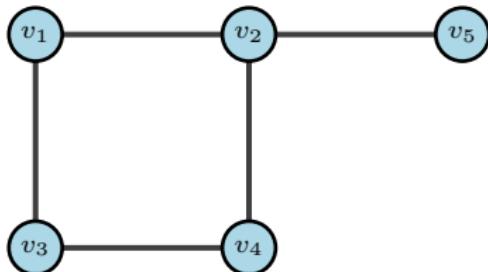
Definition

A **graph** is an ordered pair comprised of a set of **vertices** (or nodes), and a set of **edges** (or links).

- a graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- vertex set $\mathcal{V} = \{v_1, \dots, v_n\}$
- edge set $\mathcal{E} \subseteq [\mathcal{V}]^2$ (all 2-element subsets of \mathcal{V})

UNDIRECTED GRAPHS

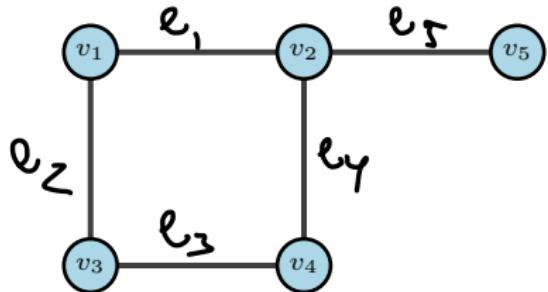
No arrows
on edges



$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

- ▶ $\mathcal{V} = \{v_1, v_2, v_3, v_4, v_5\}$
- ▶ $[\mathcal{V}]^2 = \{\{v_1, v_2\}, \{v_1, v_3\}, \{v_1, v_4\}, \{v_1, v_5\}, \{v_2, v_3\}, \{v_2, v_4\}, \{v_2, v_5\}, \{v_3, v_4\}, \{v_3, v_5\}, \{v_4, v_5\}\}$
- ▶ $\mathcal{E} = \{\{v_1, v_2\}, \{v_1, v_3\}, \{v_2, v_4\}, \{v_3, v_4\}, \{v_2, v_5\}\}$

UNDIRECTED GRAPHS



$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

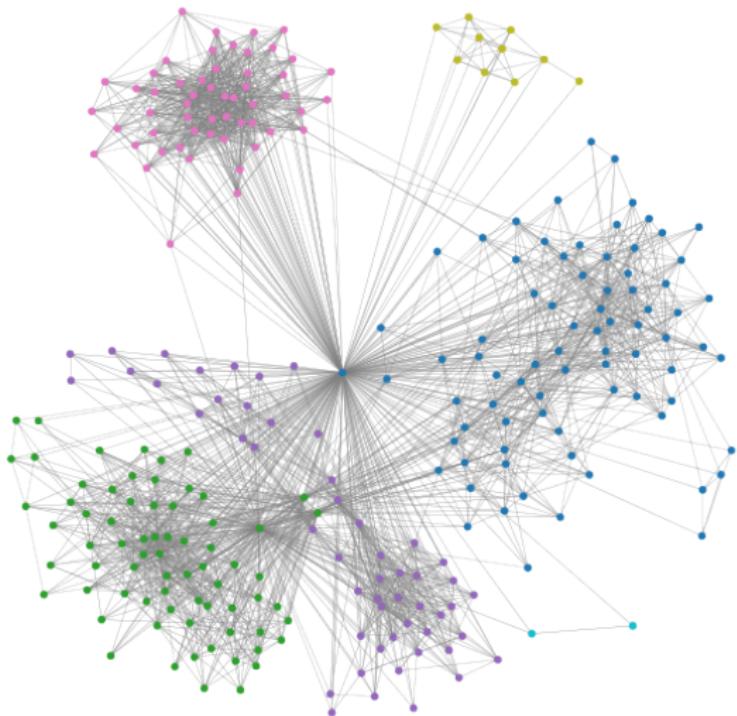
more terminology...

- ▶ **adjacent nodes**: $v_1 \sim v_2 \Leftrightarrow \{v_1, v_2\} \in \mathcal{E} \Leftrightarrow v_1v_2 \in \mathcal{E}$
- ▶ **incident nodes**: $v_i \in \mathcal{V}$ is *incident* to edge $e_j \in \mathcal{E}$, i.e., $e_j = \{v_i, v_k\} \in \mathcal{E}$ for some $v_k \in \mathcal{V}$
- ▶ **neighborhood** of a node: $\mathcal{N}(v_i) = \mathcal{N}_{v_i} = \{v_j \in \mathcal{V} : \{v_i, v_j\} \in \mathcal{E}\}$

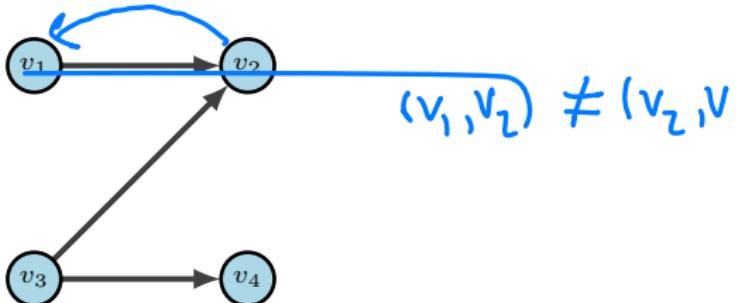
$$\mathcal{N}(v_3) = \{v_1, v_4\}$$

UNDIRECTED GRAPHS

Example: graphs can model *social interactions*



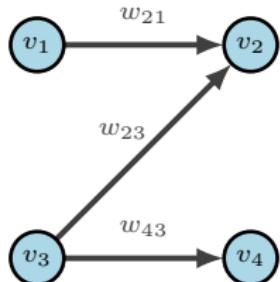
DIRECTED GRAPHS



$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

- $\mathcal{V} = \{v_1, v_2, v_3, v_4, v_5\}$
- $\mathcal{E} = \{(v_1, v_2), (v_3, v_2), (v_3, v_4)\}$
 - edges are **ordered pairs** with a **tail** (initial) and **head** (terminal) node
 - edges are said to have an **orientation**
- can define **(in)-** and **(out)-**Neighborhoods

WEIGHTED GRAPHS



$$G = (\mathcal{V}, \mathcal{E})$$

$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

- ▶ **weights** can be assigned to each edge (directed or undirected)
- ▶ $\mathcal{W} : \mathcal{E} \rightarrow \mathbb{R}$
 i.e., $\mathcal{W}((v_3, v_4)) = w_{43}$
- can collect weights into a diagonal matrix

$$W = \begin{bmatrix} & & \\ & \ddots & \\ & & w_{ji} \\ & & & \ddots \end{bmatrix} \in \mathbb{R}^{|\mathcal{E}| \times |\mathcal{E}|}$$

PATHS AND WALKS

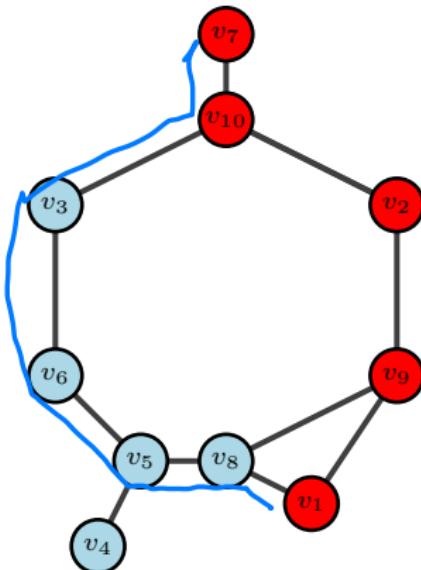
Definition

A (simple) path is a sequence of distinct vertices such that consecutive vertices are adjacent.

Example: path from v_1 to v_7

$$P(v_1, v_7) = v_1v_9v_2v_{10}v_7$$

- ▶ path length is the number of edges traversed
- ▶ paths are not unique!
 - shortest path



PATHS AND WALKS

Definition

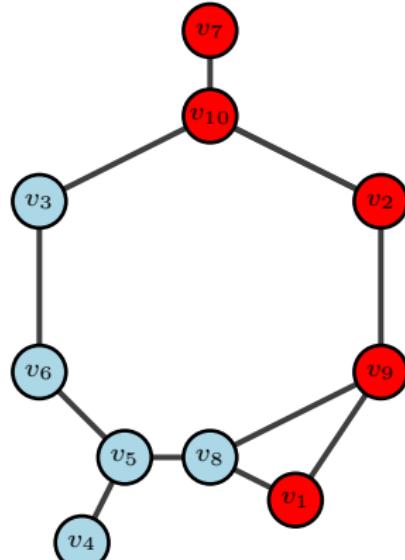
A (simple) **path** is a sequence of distinct vertices such that consecutive vertices are adjacent.

Example: path from v_1 to v_7

$$P(v_1, v_7) = v_1v_9v_2v_{10}v_7$$

Example: *Shortest Path Problem*

Given a graph with two nodes identified as the *start* node and the *terminal* node, find the shortest length path between them.



- ▶ Waze and other navigation software
- ▶ optimization over graphs
(Network Optimization)

PATHS AND WALKS

Definition

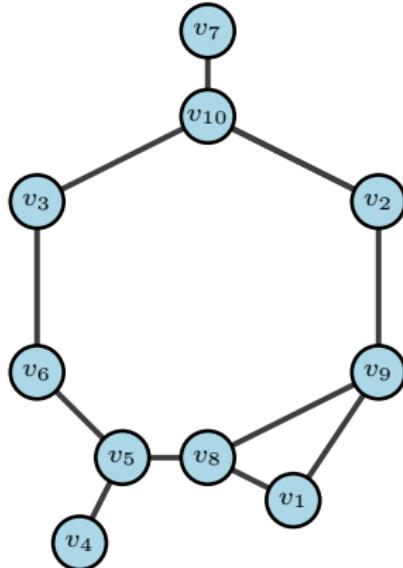
A **walk** (of length k) is a non-empty alternating sequence $v_0e_0v_1e_1 \cdots e_{k-1}v_k$ of vertices and edges in \mathcal{G} such that $e_i = \{v_i, v_{i+1}\}$ for all $i < k$. If $v_0 = v_k$, the walk is **closed**.

Example: possible walk from v_4 to v_6

$v_4e_{45}v_5e_{56}v_6$ (length:2)

or

$v_4e_{45}v_5e_{58}v_8e_{81}v_1e_{81}v_8e_{58}v_5e_{56}v_6$ (length:6)

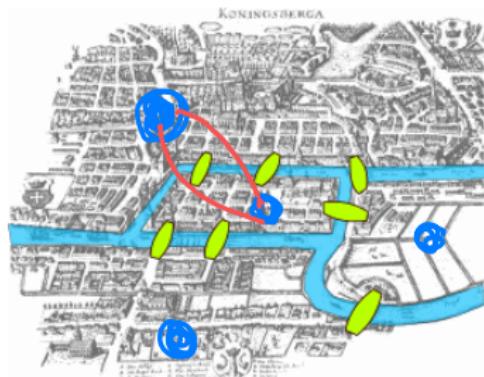


SEVEN BRIDGES OF KÖNIGSBERG

9



- ▶ 7 bridges problem led to the development of graph theory



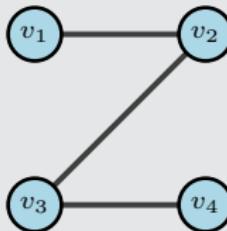
Is there a *walk* through the city of Königsberg that crosses each bridge once and only once?

CONNECTIVITY OF GRAPHS

Undirected Graphs

Connected Graph

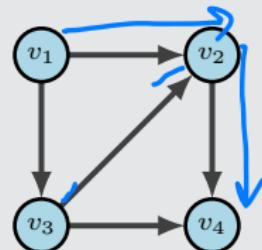
an undirected graph is **connected** if for every pair of vertices, there exists a path connecting them



Directed Graphs

Strongly Connected Graph

a directed graph is **strongly connected** if for every pair of vertices, there exists a *directed* path connecting them

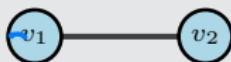


CONNECTIVITY OF GRAPHS

Undirected Graphs

Disconnected Graph

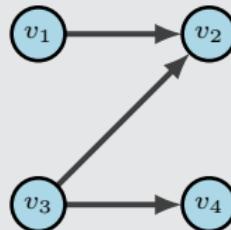
a graph is **disconnected** if it is not (weakly) connected



Directed Graphs

Weakly Connected Graph

a directed graph is **weakly connected** if the graph obtained by replacing each directed edge with an undirected edge is connected

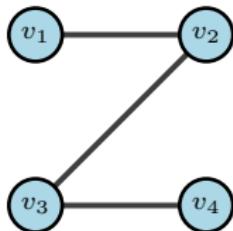


NODE DEGREE

Undirected Graphs

- **degree** of a vertex $v_i \in \mathcal{V}$ is the cardinality of its neighbor set

$$d_{v_i} = |\mathcal{V}(v_i)|$$



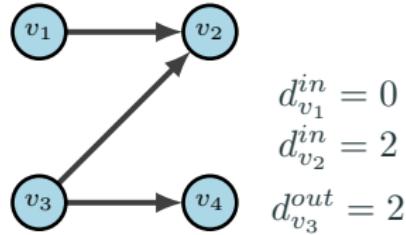
Directed Graphs

- **in-degree** of a vertex $v_i \in \mathcal{V}$ is the cardinality of its in-neighbor set

$$d_{v_i}^{in} = |\mathcal{V}^{in}(v_i)|$$

- **out-degree** of a vertex $v_i \in \mathcal{V}$ is the cardinality of its out-neighbor set

$$d_{v_i}^{out} = |\mathcal{V}^{out}(v_i)|$$

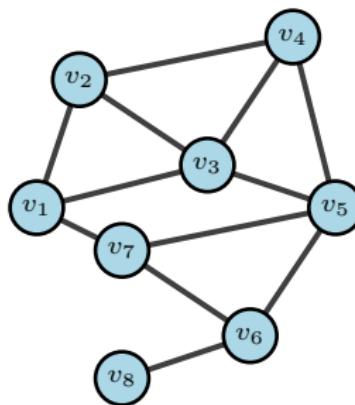


SUBGRAPHS

Graphs are a **set-theoretic** object!

$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

$$\mathcal{V} = \{v_1, \dots, v_8\}$$



SUBGRAPHS

Graphs are a **set-theoretic** object!

$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

$$\mathcal{V} = \{v_1, \dots, v_8\}$$

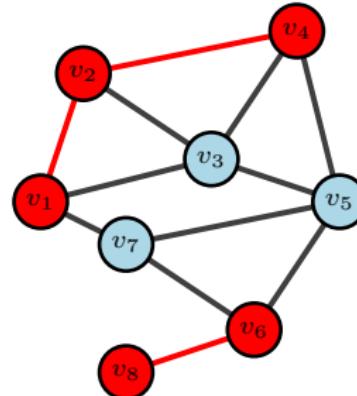
Subgraph

$$\mathcal{G}' = (\mathcal{V}', \mathcal{E}') \subset \mathcal{G}$$

$$\Rightarrow \mathcal{V}' \subseteq \mathcal{V} \text{ and } \mathcal{E}' \subseteq \mathcal{E}$$

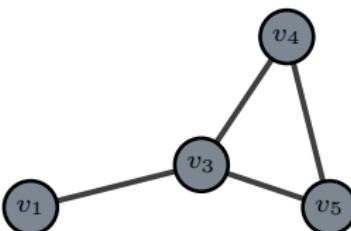
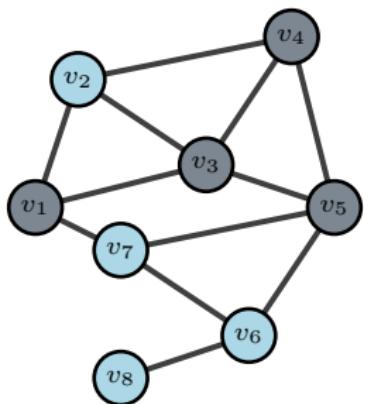
$$\mathcal{V}' = \{v_1, v_2, v_4, v_6, v_8\}$$

$$\mathcal{E}' = \{\{v_1, v_2\}, \{v_2, v_4\}, \{v_6, v_8\}\}$$



SUBGRAPHS

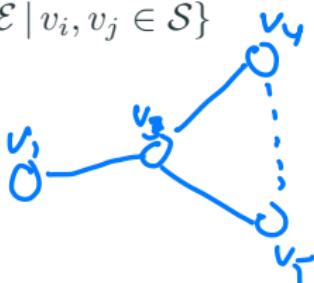
Graphs are a **set-theoretic** object!



$$\begin{aligned}\mathcal{G}_{\mathcal{S}} &= (\mathcal{S}, \mathcal{E}_{\mathcal{S}}) \subseteq \mathcal{G} \\ \mathcal{E}_{\mathcal{S}} &= \{\{v_i, v_j\} \in \mathcal{E} \mid v_i, v_j \in \mathcal{S}\}\end{aligned}$$

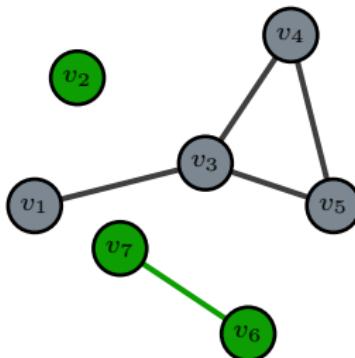
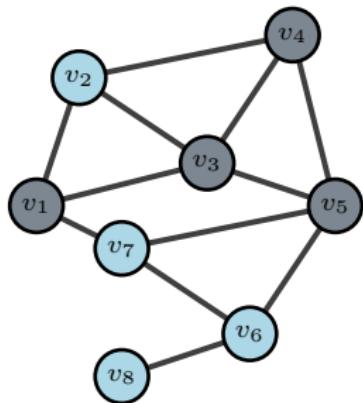
Generate a subgraph that is **induced** by a set of nodes

$$\mathcal{S} = \{v_1, v_3, v_4, v_5\}$$



SUBGRAPHS

Graphs are a **set-theoretic** object!



Generate a subgraph that is **induced** by a set of nodes

$$\mathcal{S} = \{v_1, v_3, v_4, v_5\}$$

$$\mathcal{G}_{\mathcal{S}} = (\mathcal{S}, \mathcal{E}_{\mathcal{S}}) \subseteq \mathcal{G}$$

$$\mathcal{E}_{\mathcal{S}} = \{\{v_i, v_j\} \in \mathcal{E} \mid v_i, v_j \in \mathcal{S}\}$$

Boundary of a subgraph

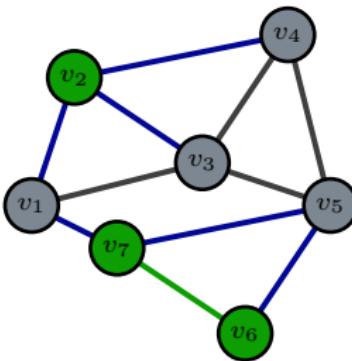
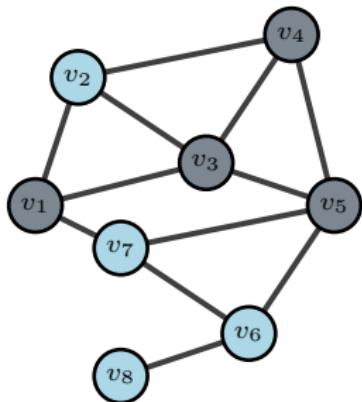
$$\partial\mathcal{G}_{\mathcal{S}} = (\partial\mathcal{S}, \mathcal{E}_{\partial\mathcal{S}})$$

$$\partial\mathcal{S} = \{v_i \in \mathcal{V} \mid v_i \notin \mathcal{S}, \exists v_j \in \mathcal{S} \text{ s.t. } \{v_i, v_j\} \in \mathcal{E}\} = \{v_2, v_6, v_7\}$$

$$\mathcal{E}_{\partial\mathcal{S}} = \{\{v_i, v_j\} \in \mathcal{E} \mid v_i, v_j \in \partial\mathcal{S}\}$$

SUBGRAPHS

Graphs are a **set-theoretic** object!



Generate a subgraph that
is **induced** by a set of
nodes

$$\mathcal{S} = \{v_1, v_3, v_4, v_5\}$$

$$\mathcal{G}_{\mathcal{S}} = (\mathcal{S}, \mathcal{E}_{\mathcal{S}}) \subseteq \mathcal{G}$$

$$\mathcal{E}_{\mathcal{S}} = \{\{v_i, v_j\} \in \mathcal{E} \mid v_i, v_j \in \mathcal{S}\}$$

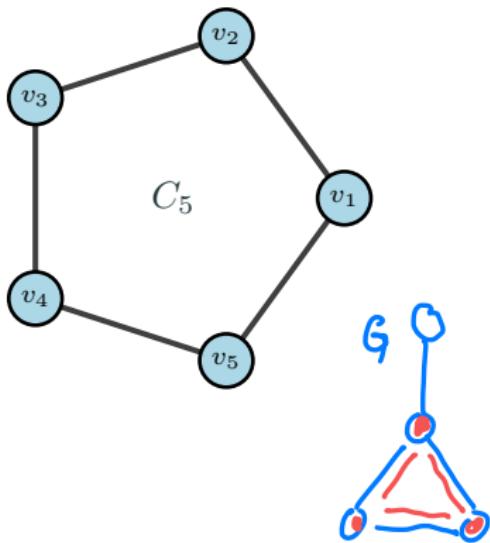
Closure of a subgraph

$$\text{cl}\mathcal{G}_{\mathcal{S}} = \mathcal{G}_{\mathcal{S} \cup \partial\mathcal{S}}$$

SPECIAL GRAPH CLASSES

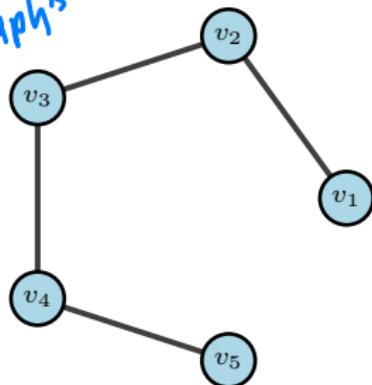
Trees and Cycles

A **cycle** is a connected graph where each node has degree 2



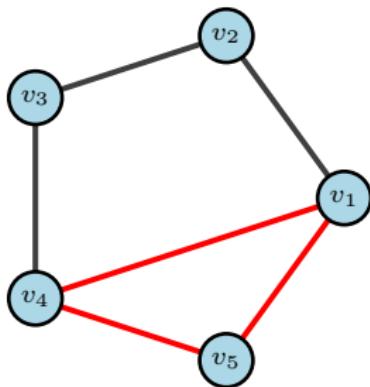
A **tree** is a connected graph containing no cycles (acyclic)

"subgraph"

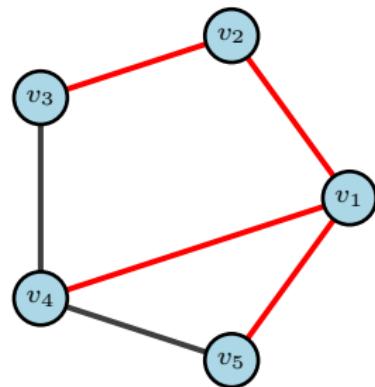


Trees and Cycles

A graph **contains cycles** if there is a subgraph that is a cycle



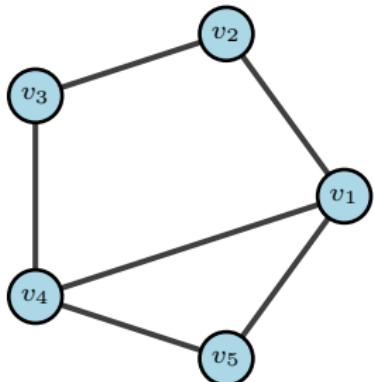
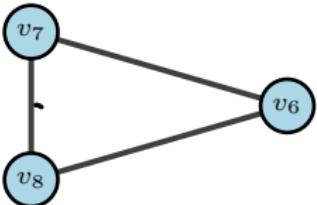
A **spanning tree** of a connected graph is a subgraph that is a tree



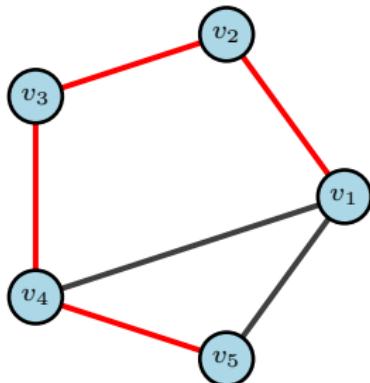
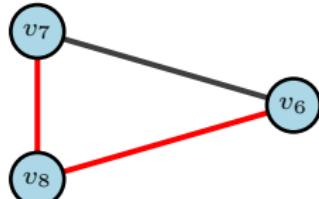
SPECIAL GRAPH CLASSES

Forests

2 components

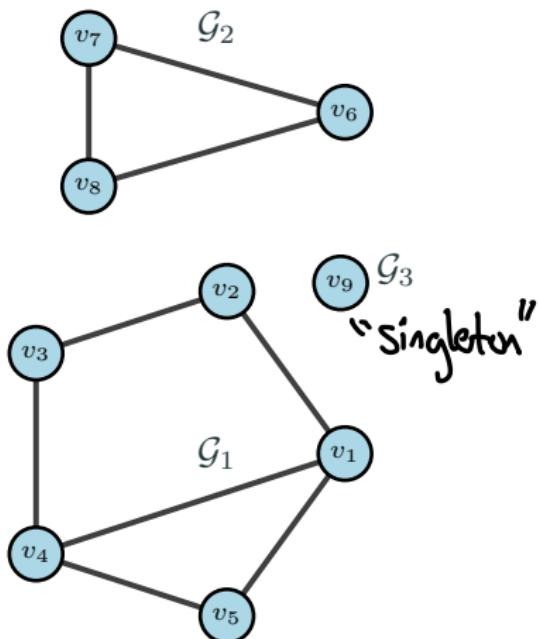


A **spanning forest** is a maximal acyclic subgraph



SPECIAL GRAPH CLASSES

Connected Components



A **connected component** is a connected subgraph of \mathcal{G}

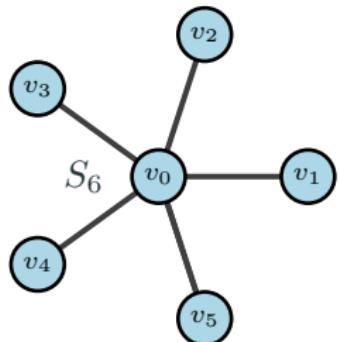
$$\mathcal{G} = \mathcal{G}_1 \cup \mathcal{G}_2 \cup \mathcal{G}_3$$

\mathcal{G} has 3 connected components

$$\mathcal{G} = (V_1 \cup V_2 \cup V_3, E_1 \cup E_2 \cup E_3)$$

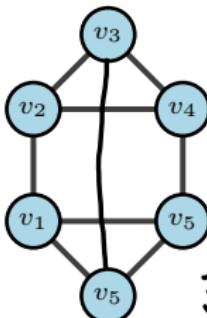
SPECIAL GRAPH CLASSES

Star Graph



k -Regular Graph

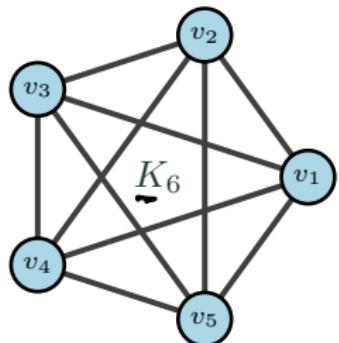
↑
degree of
each node



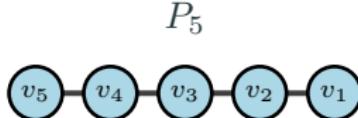
3-regular

3-regular

Complete Graph

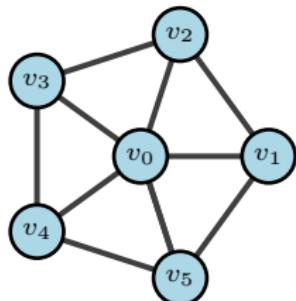


Path Graph

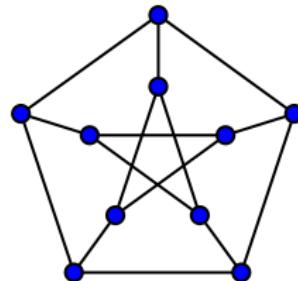


SPECIAL GRAPH CLASSES

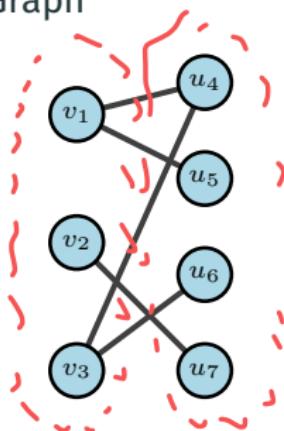
Wheel Graph



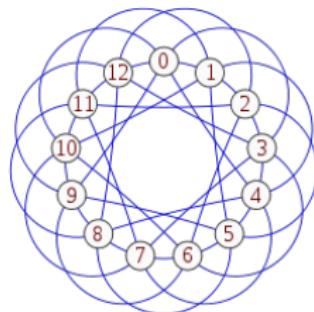
Peterson Graph



Bipartite Graph

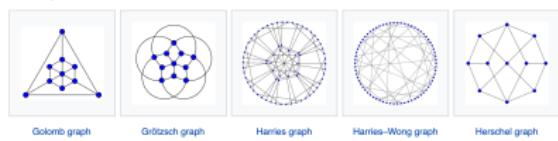
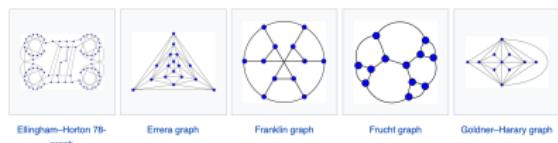
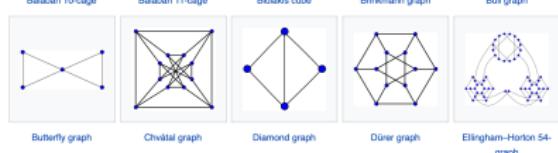
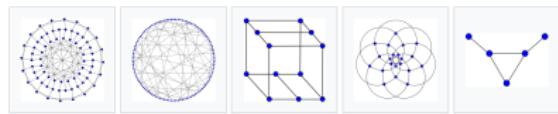


Payley Graph



SPECIAL GRAPH CLASSES

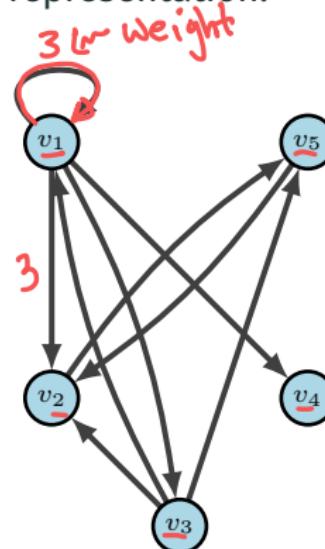
so many named graphs!



GRAPHS AND MATRICES

All square matrices have a (directed) graph representation!

$$M = \begin{bmatrix} 3 & 3 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 10 \\ 2 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{bmatrix} \Leftrightarrow \begin{matrix} 5 \times 5 \\ \text{5x5} \end{matrix}$$



For a matrix $M \in \mathbb{R}^{n \times n}$

$$\mathcal{G}(M) = (\mathcal{V}(M), \mathcal{E}(M))$$

$$|\mathcal{V}(M)| = n$$

$$e = (v_i, v_j) \in \mathcal{E}(M) \Leftrightarrow [M]_{ij} \neq 0$$

0 or *

WHEN GRAPH THEORY AND LINEAR ALGEBRA MEET

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} b \\ a \\ c \end{bmatrix}$$

Definition

A matrix $M \in \mathbb{R}^{n \times n}$ is said to be **irreducible** if there does not exist a permutation matrix P and an integer r such that

$$P^T M P = \begin{bmatrix} B & C \\ 0 & D \end{bmatrix},$$

with $B \in \mathbb{R}^{r \times r}$, $C \in \mathbb{R}^{r \times n-r}$, and $D \in \mathbb{R}^{n-r \times n-r}$.

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

WHEN GRAPH THEORY AND LINEAR ALGEBRA MEET

Definition

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with $B \in \mathbb{R}^{r \times r}$, $C \in \mathbb{R}^{r \times n-r}$, and $D \in \mathbb{R}^{n-r \times n-r}$.

What is a permutation matrix?

Definition

A **permutation matrix** P is a square matrix of order n such that each row and column contains one element equal to 1, with remaining elements equal to 0. Furthermore, permutation matrices satisfy the property $P^T = P^{-1}$.

Example:

$$P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$
$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$$

WHEN GRAPH THEORY AND LINEAR ALGEBRA MEET

Definition

A matrix $M \in \mathbb{R}^{n \times n}$ is said to be **irreducible** if there does not exist a permutation matrix P and an integer r such that

$$P^T M P = \begin{bmatrix} B & C \\ 0 & D \end{bmatrix},$$

with $B \in \mathbb{R}^{r \times r}$, $C \in \mathbb{R}^{r \times n-r}$, and $D \in \mathbb{R}^{n-r \times n-r}$.

Which matrix is irreducible?

$$\overline{\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}} \begin{bmatrix} 0 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix}$$

$$M = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \text{or} \quad M = \begin{bmatrix} 0 & 0 \\ 1 & 2 \end{bmatrix}$$

what is the permutation matrix?

$$\begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 0 & 0 \end{bmatrix}$$

Theorem

Let $M \in \mathbb{R}^{n \times n}$. The following statements are equivalent:

- i) M is irreducible.
- ii) The digraph associated with M , $\mathcal{G}(M)$, is strongly connected.

M is irreducible $\Rightarrow G[M]$ is S.C.

• Assume $G[M]$ is not S.C.

$\Rightarrow \exists$ at least one pair of nodes
with no directed path between them

\Rightarrow say there is no directed path from
node v to u

Define 3 sets

i) $R(v) = \{ s \in V[M] \mid \exists \text{ a directed path from } v \text{ to } s \} \cup \{v\}$

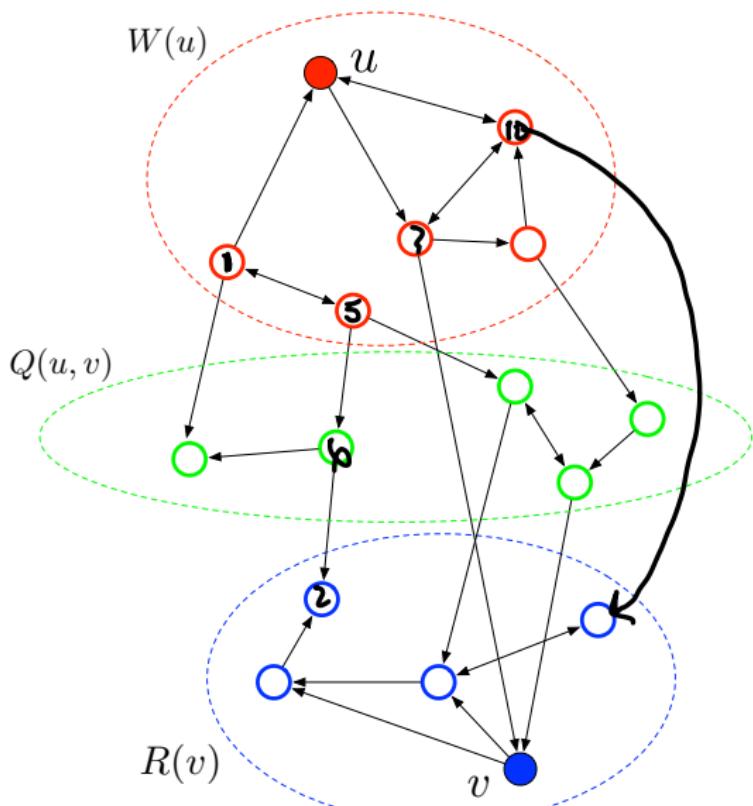
ii) $W(u) = \{ s \in V[M] \mid \exists \text{ a directed path from } s \text{ to } u \} \cup \{u\}$

$$\Rightarrow R(v) \cap W(u) = \emptyset$$

iii) $Q(u, v) = V[M] \setminus (R(v) \cup W(u))$

↑
Set Minus

IRREDUCIBILITY AND STRONG CONNECTEDNESS



$$P^T M P = \bar{M}$$

$$\bar{M} =$$

$$\begin{bmatrix} W(u) & Q(u,v) & R(v) \\ \vdots & \vdots & \vdots \\ \ast & \ast & \ast \\ \vdots & \vdots & \vdots \\ \circ & \ast & \ast \\ \vdots & \vdots & \vdots \\ \circ & \ast & \ast \\ \vdots & \vdots & \vdots \\ \circ & 0 & 0 \\ \vdots & \vdots & \vdots \\ \circ & 0 & \ast \end{bmatrix}$$

W(u) Q(u,v) R(v)

W(u)

Q(u,v)

R(v)

Structured Linear Systems

A *structured linear system* is a description of a dynamic system that considers only the interaction and influence between system states, control, and outputs independent on any realization of parameter values.

$$\begin{bmatrix} \dot{p} \\ \dot{\theta} \\ \ddot{p} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & \frac{-m_p g}{m_c} & \frac{-K_1^2}{R_m m_c} & 0 \\ 0 & \frac{(m_p+m_c)g}{m_c I_p} & \frac{K_1^2}{R_m m_c I_p} & 0 \end{bmatrix} \begin{bmatrix} p \\ \theta \\ \dot{p} \\ \dot{\theta} \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \frac{K_1}{R_m m_c} \\ \frac{-K_1}{R_m m_c I_p} \end{bmatrix} F$$

$$y = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} p \\ \theta \\ \dot{p} \\ \dot{\theta} \end{bmatrix}$$

STRUCTURED LINEAR SYSTEMS

Structured Linear Systems

A *structured linear system* is a description of a dynamic system that considers only the interaction and influence between system states, control, and outputs independent on any realization of parameter values.

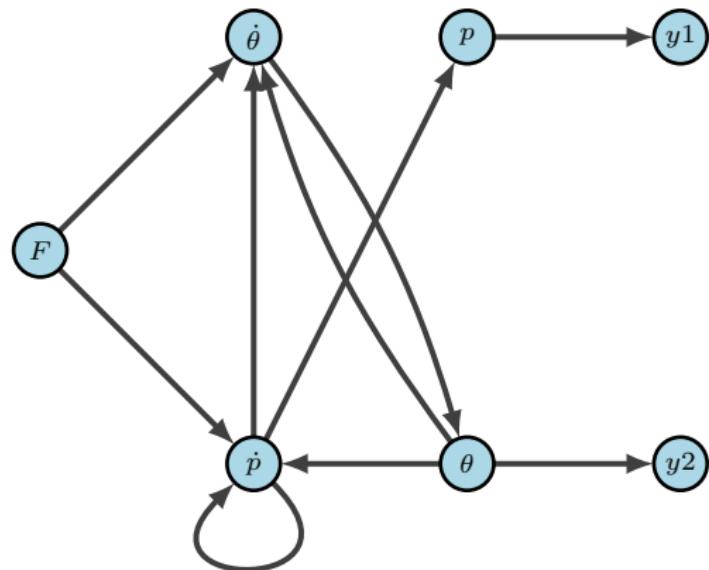
- ▶ Express dynamics as a matrix

$$\begin{array}{l} \text{deriv} \rightarrow \\ \text{outpt} \rightarrow \end{array} \begin{bmatrix} \dot{\mathbf{x}} \\ \mathbf{y} \end{bmatrix} = \underbrace{\begin{bmatrix} A & B \\ C & 0 \end{bmatrix}}_M \begin{bmatrix} \mathbf{x} \\ F \end{bmatrix}$$

state
input

- ▶ Define the digraph associated with M

$$\mathcal{V} = \{F, p, \dot{p}, \theta, \dot{\theta}, y_1, y_2\}$$



STRUCTURAL CONTROLLABILITY

the structure of a system

- ▶ system states and controls are either related (non-zero entry in state-space) or not (0-entry)
- ▶ values of parameters are neglected

$$\begin{bmatrix} \dot{p} \\ \dot{\theta} \\ \ddot{p} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} 0 & 0 & * & 0 \\ 0 & 0 & 0 & * \\ 0 & * & * & 0 \\ 0 & * & * & 0 \end{bmatrix} \begin{bmatrix} p \\ \theta \\ \dot{p} \\ \dot{\theta} \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ * \\ * \end{bmatrix} F$$

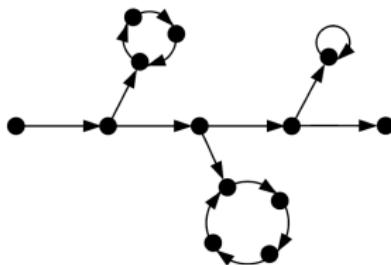
Definition

A system (A, B) is **structurally controllable** if there exists a system structurally equivalent to (A, B) which is controllable in the usual sense.

Theorem [Lin '74]

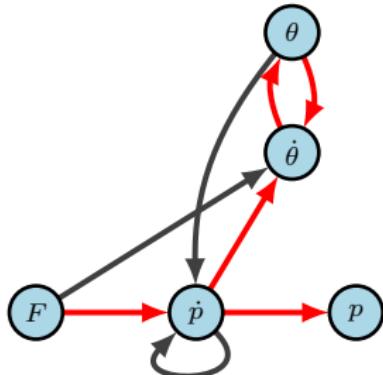
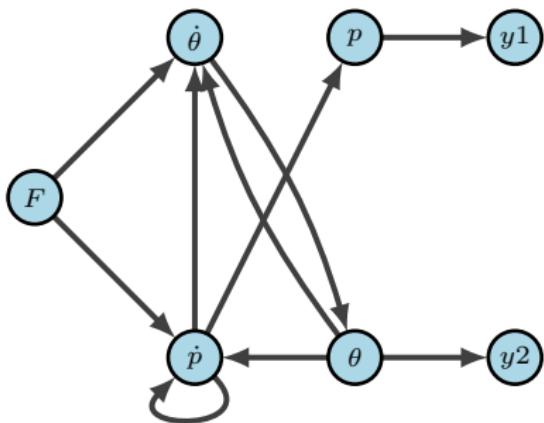
The following statements for a structured system (A, B) are equivalent:

- i) (A, B) is structurally controllable
- ii) In the graph $\mathcal{G}(A, B)$, there exists a disjoint union of cacti that covers all the state vertices.



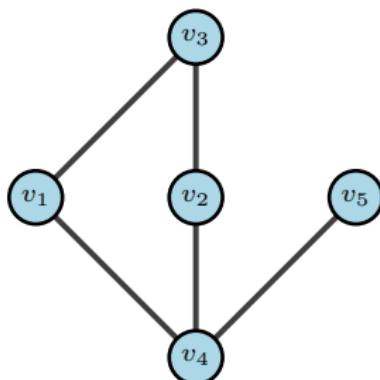
a cactus graph with 3 buds

STRUCTURAL CONTROLLABILITY



the graph of the system contains a cactus! the system is structurally controllable!

Graphs and their properties can be studied using matrices and constructs from linear algebra

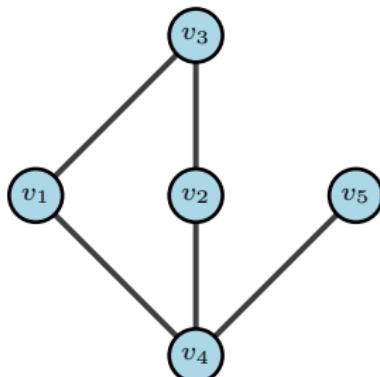


Degree Matrix: $\Delta(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$
 A diagonal matrix with the degree of each node on the diagonal

$$[\Delta(\mathcal{G})]_{ij} = \begin{cases} d(v_i), & i = j \\ 0, & \text{otherwise} \end{cases}$$

$$\Delta(\mathcal{G}) = \begin{bmatrix} 2 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Graphs and their properties can be studied using matrices and constructs from linear algebra



Adjacency Matrix: $A(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$
A symmetric matrix encoding the adjacency relationship of nodes in the graph

$$[A(\mathcal{G})]_{ij} = \begin{cases} 1, & i \sim j \\ 0, & \text{otherwise} \end{cases}$$

$$A(\mathcal{G}) = \begin{bmatrix} 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$

NUMBER OF WALKS LEMMA

Lemma

Let \mathcal{G} be a graph with adjacency matrix $A(\mathcal{G})$. The number of walks from node v_i to v_j of length r is $[A(\mathcal{G})^r]_{ij}$.

Proof:

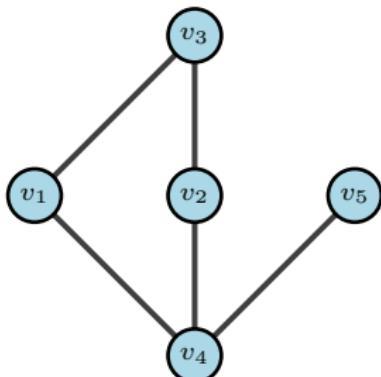
Homework

Corollary

Let \mathcal{G} be an undirected graph with e edges, t triangles, and adjacency matrix $A(\mathcal{G})$. Then

- i) $\text{tr } A(\mathcal{G}) = 0$
- ii) $\text{tr } A(\mathcal{G})^2 = 2e$
- iii) $\text{tr } A(\mathcal{G})^3 = 6t$

Graphs and their properties can be studied using matrices and constructs from linear algebra



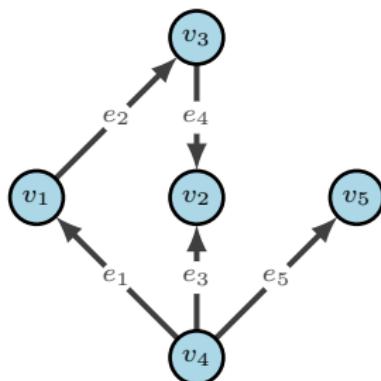
Incidence Matrix: $E(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{E}|}$

A matrix encoding the incidence relation between nodes and edges

$$[E(\mathcal{G})]_{ij} = \begin{cases} 1, & v_i \text{ is tail of edge } e_j \\ -1, & v_i \text{ is head of edge } e_j \\ 0, & \text{otherwise} \end{cases}$$

$$E(\mathcal{G}) = \begin{bmatrix} -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & -1 & -1 & 0 \\ 0 & -1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & -1 \end{bmatrix}$$

Graphs and their properties can be studied using matrices and constructs from linear algebra



assign an **arbitrary** orientation to each edge

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Theorem

Let \mathcal{G} be a graph with n vertices, c connected components, and an arbitrary orientation assigned to each edge. Then $\text{rank } E(\mathcal{G}) = n - c$.

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- \mathcal{G} has c connected components: $\mathcal{G} = \cup_{i=1}^c \mathcal{G}_i$

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Proof:

- \mathcal{G} has c connected components: $\mathcal{G} = \cup_{i=1}^c \mathcal{G}_i$
- with appropriate relabelling of nodes/edges, can write

$$E(\mathcal{G}) = \begin{bmatrix} E(\mathcal{G}_1) & & \\ & \ddots & \\ & & E(\mathcal{G}_c) \end{bmatrix}$$

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Proof:

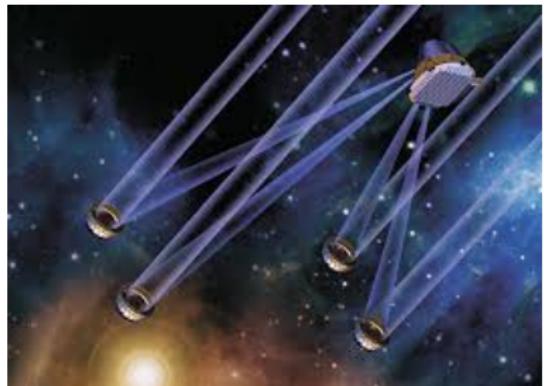
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- with appropriate relabelling of nodes/edges, can write

$$E(\mathcal{G}) = \begin{bmatrix} E(\mathcal{G}_1) & & \\ & \ddots & \\ & & E(\mathcal{G}_c) \end{bmatrix}$$

- let $\mathcal{H} = (\mathcal{V}, \mathcal{E})$ be a connected graph. Show that $\text{rank } E(\mathcal{H}) = |\mathcal{V}| - 1$

RELATIVE SENSING NETWORKS

Interferometry is a technique used for imaging in deep space. Rather than using 1 large (and expensive!) telescope, a team of smaller (and cheaper!) sensors can achieve the same goal. This requires high accuracy and precision of relative spacing between satellites.



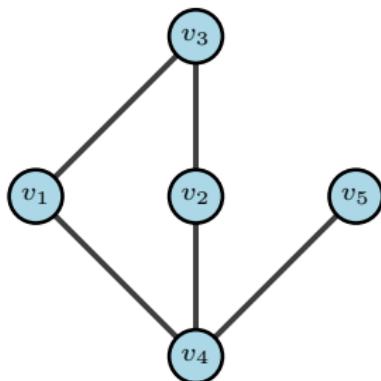
$$\dot{x} = f(x_i, u_i)$$

$$y = \begin{bmatrix} \vdots \\ x_i - x_j \\ \vdots \end{bmatrix}$$

For the sensing graph $\mathcal{G} = (\mathcal{V}, \mathcal{E})$, each edge $e_i = (v_i, v_j) \in \mathcal{E}$ encodes the relative measurement $x_i - x_j$

$$y = E(\mathcal{G})^T x$$

Graphs and their properties can be studied using matrices and constructs from linear algebra



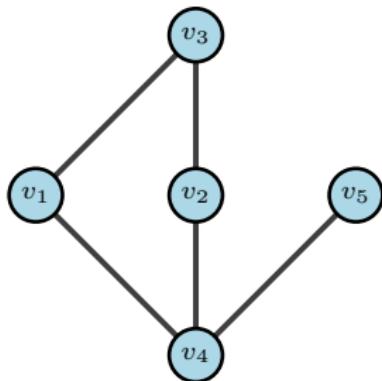
Combinatorial Graph Laplacian: $L(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$
A symmetric matrix

$$[L(\mathcal{G})]_{ij} = \begin{cases} d(v_i), & i = j \\ -1, & \{i, j\} \in \mathcal{E} \end{cases}$$

$$L(\mathcal{G}) = \begin{bmatrix} 2 & 0 & -1 & -1 & 0 \\ 0 & 2 & -1 & -1 & 0 \\ -1 & -1 & 2 & 0 & 0 \\ -1 & -1 & 0 & 3 & -1 \\ 0 & 0 & 0 & -1 & 1 \end{bmatrix}$$

Graphs and their properties can be studied using matrices and constructs from linear algebra

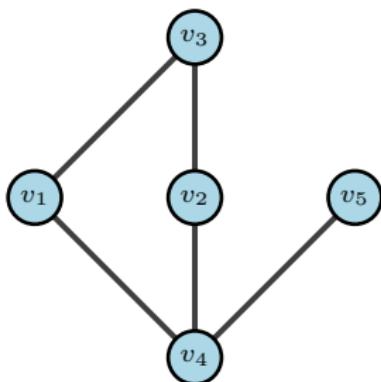
Combinatorial Graph Laplacian: $L(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$
Constructions



$$\begin{aligned} L(\mathcal{G}) &= \Delta(\mathcal{G}) - A(\mathcal{G}) \\ &= E(\mathcal{G})E(\mathcal{G})^T \end{aligned}$$

- ▶ using incidence matrix, construction is independent of the edge orientation!

Graphs and their properties can be studied using matrices and constructs from linear algebra



Combinatorial Graph Laplacian: $L(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$

- ▶ rank $L(\mathcal{G}) = |\mathcal{V}| - 1 \Leftrightarrow \mathcal{G}$ is connected
- ▶ \mathcal{G} is connected, then 0 is a simple eigenvalue and $L(\mathcal{G})\mathbb{1} = 0$
- ▶ $L(\mathcal{G})$ is a positive semi-definite matrix

$$x^T L(\mathcal{G}) x \geq 0 \quad \forall x \in \mathbb{R}^{|\mathcal{V}|}$$

- ▶ ordered eigenvalues

$$0 = \lambda_1(\mathcal{G}) \leq \lambda_2(\mathcal{G}) \leq \dots \leq \lambda_{|\mathcal{V}|}(\mathcal{G})$$

- ▶ Algebraic Connectivity (Fiedler Eigenvalue) :
 $\lambda_2(\mathcal{G})$

Theorem

For a graph \mathcal{G} , the following statements are equivalent:

- i) \mathcal{G} is connected
- ii) $\lambda_2(\mathcal{G}) > 0$.

MATRIX-TREE THEOREM

Theorem

Let $\tau(\mathcal{G})$ be the number of spanning trees in \mathcal{G} . Then

$$\tau(\mathcal{G}) = \det L(\mathcal{G})_{(ij)}.$$

- For a matrix $M \in \mathbb{R}^{n \times n}$, $M_{(ij)} \in \mathbb{R}^{n-1 \times n-1}$ is obtained by deleting the i th row and j th column of M

$$M = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \\ 13 & 14 & 15 & 16 \end{bmatrix} \Rightarrow M_{(23)} = \begin{bmatrix} 1 & 2 & 4 \\ 9 & 10 & 12 \\ 13 & 14 & 16 \end{bmatrix}$$

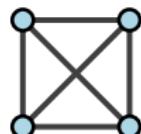
- $\det M_{(ij)}$ is called the ij -minor of M

MATRIX-TREE THEOREM

Theorem

Let $\tau(\mathcal{G})$ be the number of spanning trees in \mathcal{G} . Then

$$\tau(\mathcal{G}) = \det L(\mathcal{G})_{(ij)}.$$



$$L(\mathcal{G}) = \begin{bmatrix} 3 & -1 & -1 & -1 \\ -1 & 3 & -1 & -1 \\ -1 & -1 & 3 & -1 \\ -1 & -1 & -1 & 3 \end{bmatrix} \quad \tau(\mathcal{G}) = 16$$

