

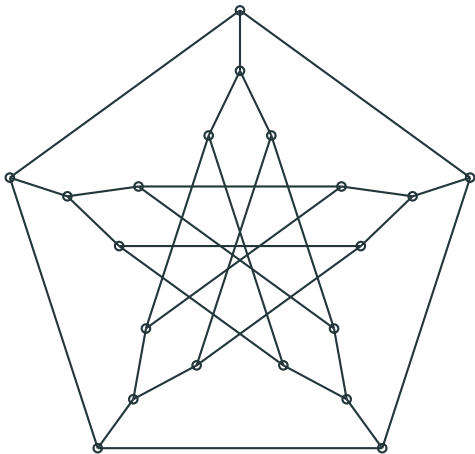
COORDINATION AND CONTROL OF MULTI-AGENT SYSTEMS

086730

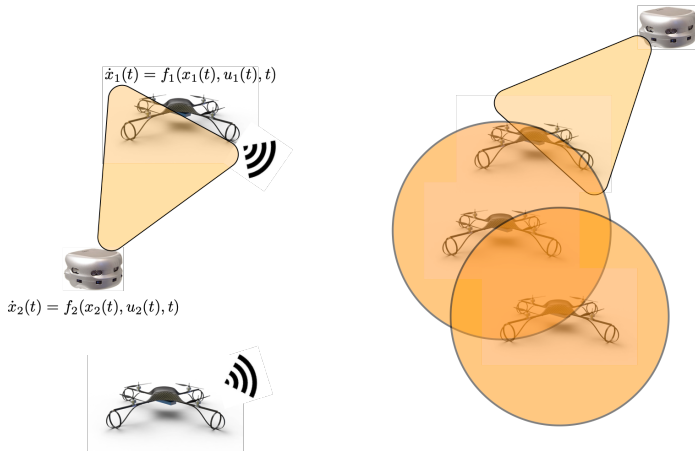
Daniel Zelazo

October 25, 2025

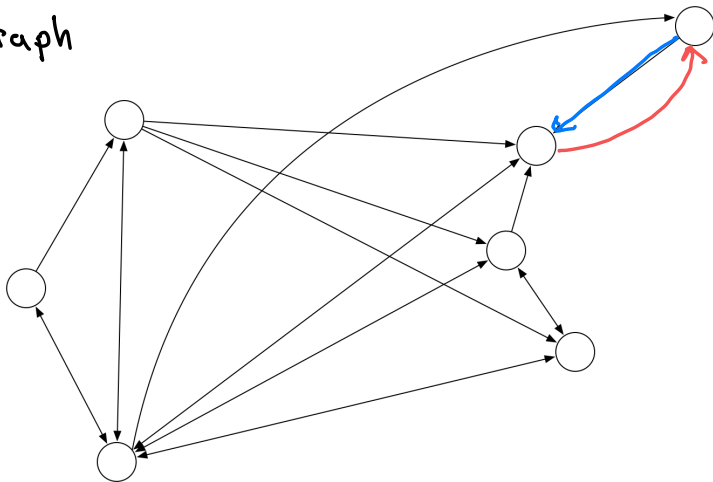
Introduction to Graph Theory



ABSTRACTION USING GRAPHS

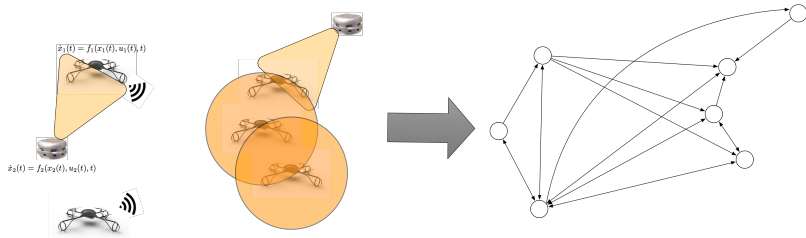


Graph



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

ABSTRACTION USING GRAPHS



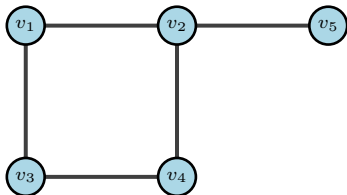
Definition

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

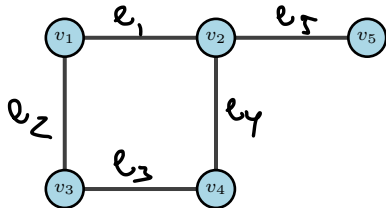
No arrows
on edges



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

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

UNDIRECTED GRAPHS



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

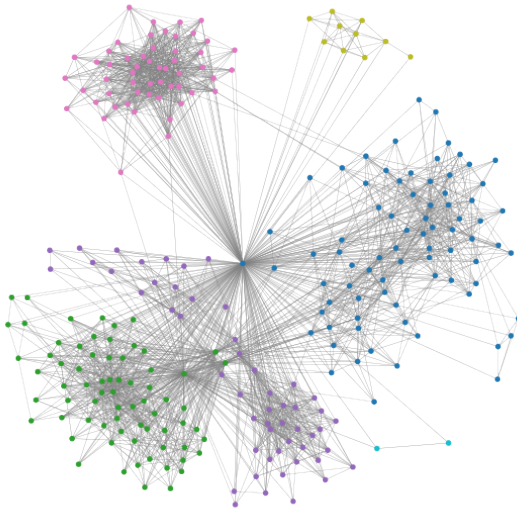
more terminology...

- ▶ **adjacent nodes** : $v_1 \sim v_2 \Leftrightarrow \{v_1, v_2\} \in \mathcal{E} \Leftrightarrow v_1 v_2 \in \mathcal{E}$
- ▶ **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: $\mathcal{N}(v_i) = \mathcal{N}_{v_i} = \{v_j \in \mathcal{V} : \{v_i, v_j\} \in \mathcal{E}\}$

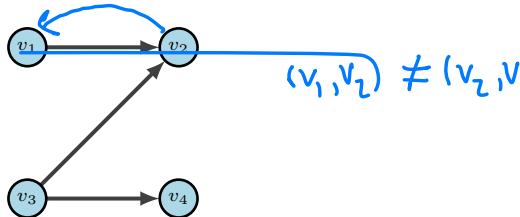
$$\mathcal{N}(v_3) = \{v_1, v_4\}$$

UNDIRECTED GRAPHS

Example: graphs can model *social interactions*



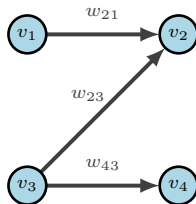
DIRECTED GRAPHS



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

- ▶ $\mathcal{V} = \{v_1, v_2, v_3, v_4, v_5\}$
- ▶ $\mathcal{E} = \{(v_1, v_2), (v_3, v_2), (v_3, v_4)\}$
 - 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



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

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

- **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

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

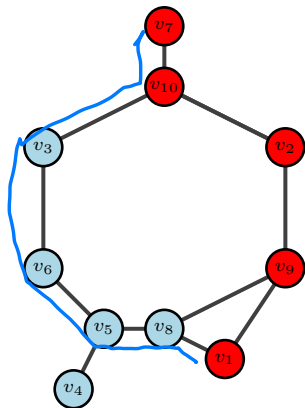
Definition

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

Example: path from v_1 to v_7

$$P(v_1, v_7) = v_1 v_9 v_2 v_{10} v_7$$

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



Definition

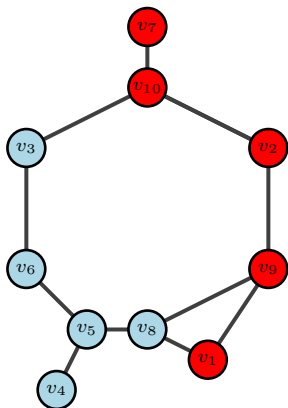
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Example: path from v_1 to v_7

$$P(v_1, v_7) = v_1 v_9 v_2 v_{10} v_7$$

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)

Definition

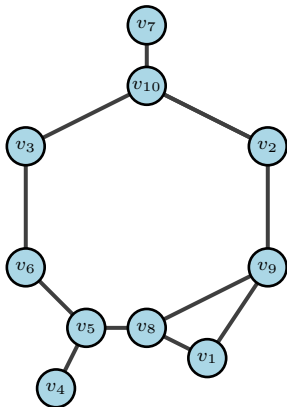
A **walk** (of length k) is a non-empty alternating sequence $v_0 e_0 v_1 e_1 \cdots e_k v_k$ of vertices and edges in $\mathcal{G}$ such that $e_i = \{v_i, v_{i+1}\}$ for all $i < k$. If $v_0 = v_k$, the walk is **closed**.

Example: possible walk from v_4 to v_6

$$v_4 e_{45} v_5 e_{56} v_6 \text{ (length:2)}$$

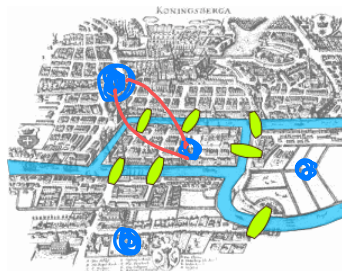
or

$$v_4 e_{45} v_5 e_{58} v_8 e_{81} v_1 e_{81} v_8 e_{58} v_5 e_{56} v_6 \text{ (length:6)}$$



SEVEN BRIDGES OF KÖNIGSBERG

9



- 7 bridges problem led to the development of graph theory

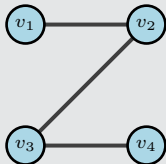
Is there a *walk* through the city of Königsberg that crosses each bridge once and only once?

CONNECTIVITY OF GRAPHS

Undirected Graphs

Connected Graph

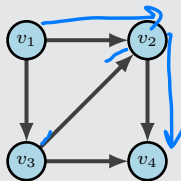
an undirected graph is **connected** if for every pair of vertices, there exists a path connecting them



Directed Graphs

Strongly Connected Graph

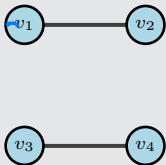
a directed graph is **strongly connected** if for every pair of vertices, there exists a *directed* path connecting them



Undirected Graphs

Disconnected Graph

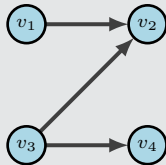
a graph is **disconnected** if it is not (weakly) connected



Directed Graphs

Weakly Connected Graph

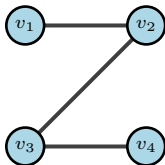
a directed graph is **weakly connected** if the graph obtained by replacing each directed edge with an undirected edge is connected



Undirected Graphs

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

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



$$\begin{aligned}d_{v_1} &= 1 \\d_{v_3} &= 2\end{aligned}$$

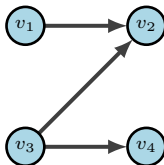
Directed Graphs

- **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

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



$$d_{v_1}^{in} = 0$$

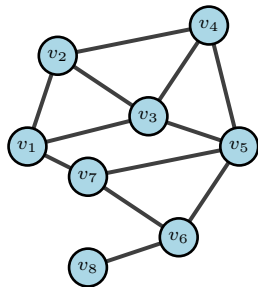
$$d_{v_2}^{in} = 2$$

$$d_{v_3}^{out} = 2$$

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

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

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



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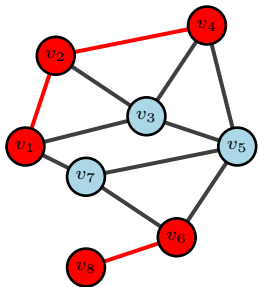
Subgraph

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

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

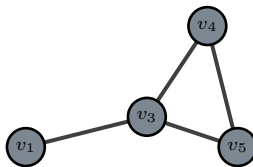
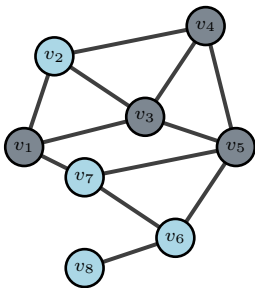
$$\mathcal{V}' = \{v_1, v_2, v_4, v_6, v_8\}$$

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



SUBGRAPHS

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

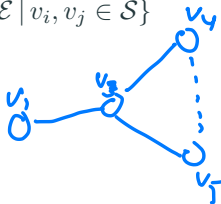


$$\mathcal{G}_S = (\mathcal{S}, \mathcal{E}_S) \subseteq \mathcal{G}$$

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

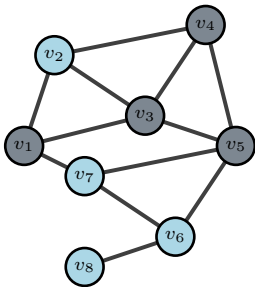
Generate a subgraph that
is **induced** by a set of
nodes

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



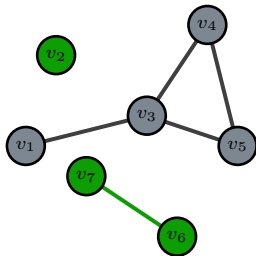
SUBGRAPHS

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

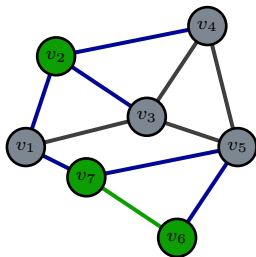
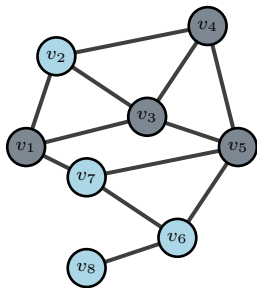
$$\partial \mathcal{G}_{\mathcal{S}} = (\partial \mathcal{S}, \mathcal{E}_{\partial \mathcal{S}})$$

$$\partial \mathcal{S} = \{v_i \in \mathcal{V} \mid v_i \notin \mathcal{S}, \exists v_j \in \mathcal{S} \text{ s.t. } \{v_i, v_j\} \in \mathcal{E}\} = \{v_2, v_7, v_8\}$$

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

SUBGRAPHS

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



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

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

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

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

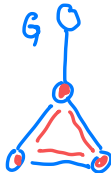
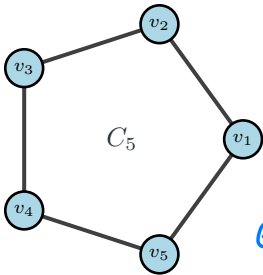
Closure of a subgraph

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

SPECIAL GRAPH CLASSES

Trees and Cycles

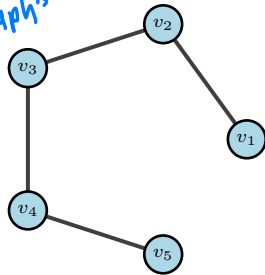
A **cycle** is a connected graph where each node has degree 2



"subgraph"

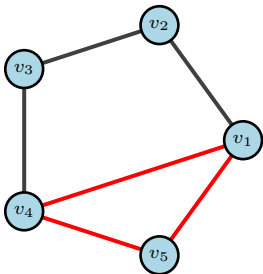
$C_3 \subset G$

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

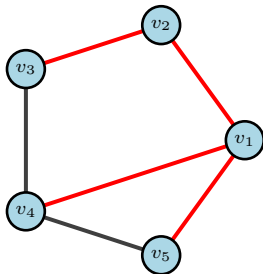


Trees and Cycles

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



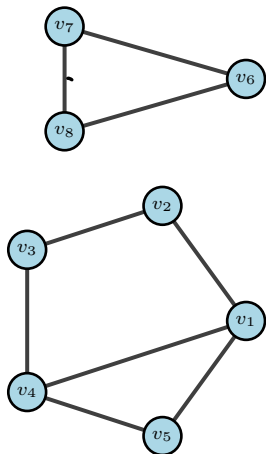
A **spanning tree** of a connected graph is a subgraph that is a tree



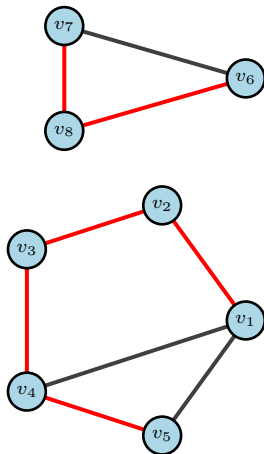
SPECIAL GRAPH CLASSES

Forests

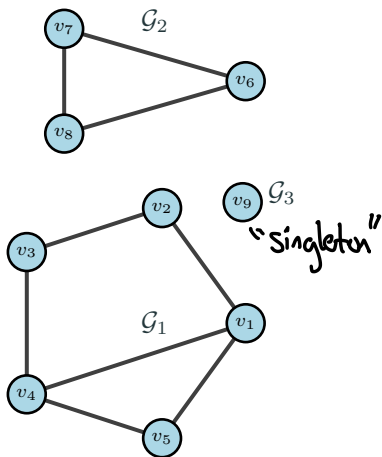
2 components



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



Connected Components



A **connected component** is a connected subgraph of G

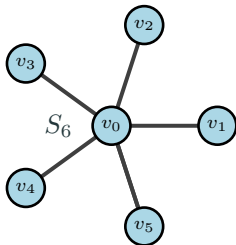
$$G = G_1 \cup G_2 \cup G_3$$

G has 3 connected components

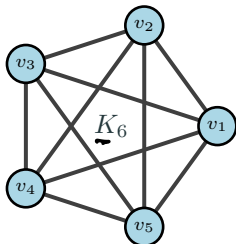
$$G = (V, \cup V_1 \cup V_2 \cup V_3, E_1 \cup E_2 \cup E_3)$$

SPECIAL GRAPH CLASSES

Star Graph

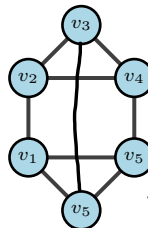


Complete Graph



k -Regular Graph

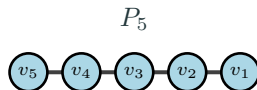
↑
degree of
each node



3-regular

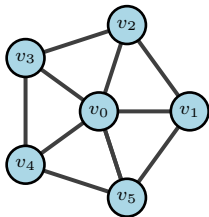
3-regular

Path Graph

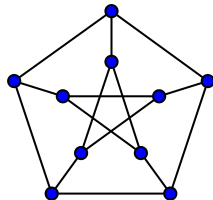


SPECIAL GRAPH CLASSES

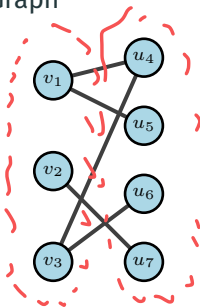
Wheel Graph



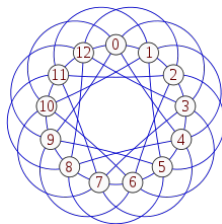
Peterson Graph



Bipartite Graph

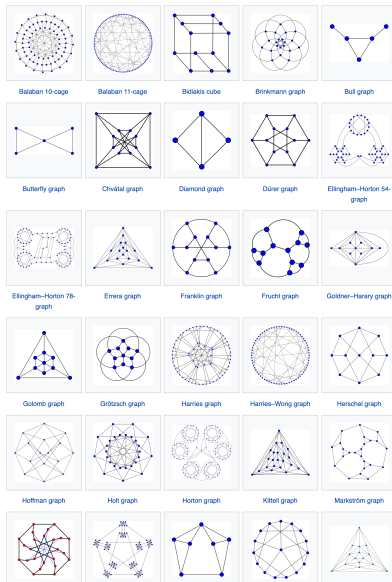


Payley Graph



SPECIAL GRAPH CLASSES

so many named graphs!

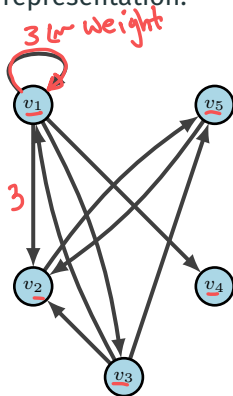


GRAPHS AND MATRICES

All square matrices have a (directed) graph representation!

5×5

$$M = \begin{bmatrix} \underline{3} & \underline{3} & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 10 \\ 2 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{bmatrix} \Leftrightarrow$$



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 *

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

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

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

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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:

$$P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}_{22}$$

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

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

Which matrix is irreducible?

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

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

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

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

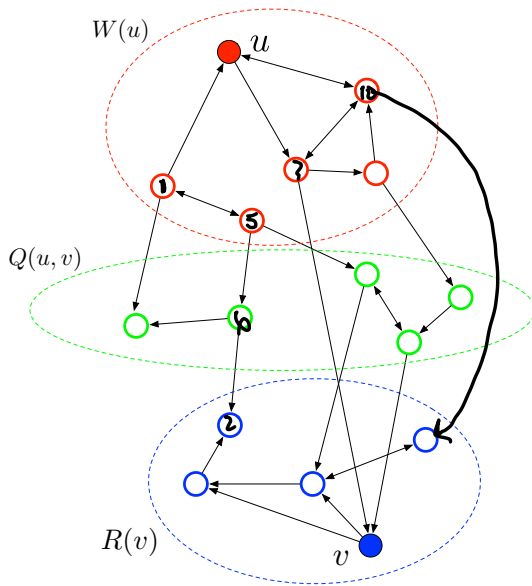
$$ii) W(u) = \left\{ s \in V[M] \mid \exists \text{ a directed path from } s \text{ to } u \right\} \cup \{u\}$$

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

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

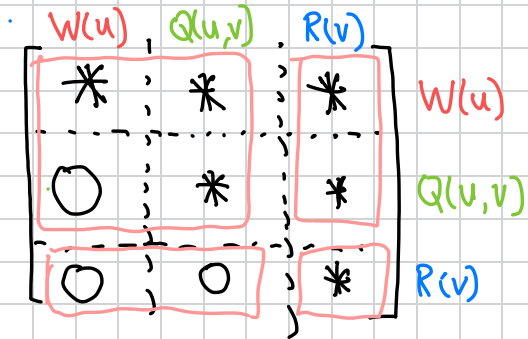
↑
set minus

IRREDUCIBILITY AND STRONG CONNECTEDNESS



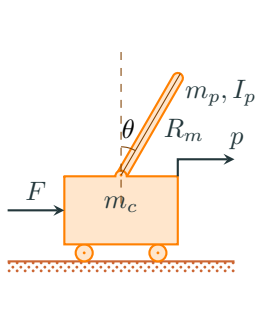
$$P^T M P = \bar{M}$$

$$\bar{M} =$$



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

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

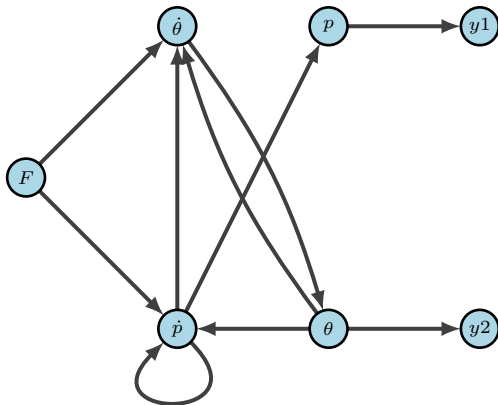
deriv. $\rightarrow$ $\dot{\mathbf{x}}$
output $\rightarrow$ y

$$\begin{bmatrix} \dot{\mathbf{x}} \\ y \end{bmatrix} = \underbrace{\begin{bmatrix} A & B \\ C & 0 \end{bmatrix}}_M \begin{bmatrix} \mathbf{x} \\ F \end{bmatrix}$$

state $\rightarrow$ $\mathbf{x}$
input $\rightarrow$ F

- Define the digraph associated with M

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



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

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

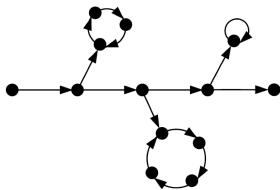
Definition

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]

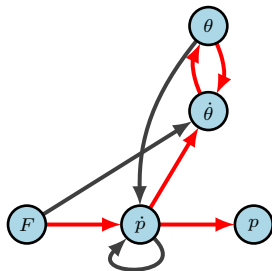
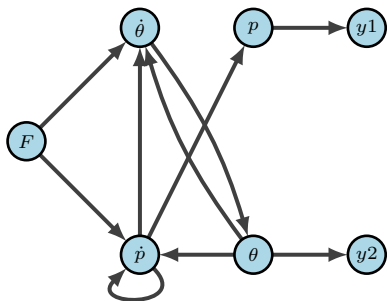
The following statements for a structured system (A, B) are equivalent:

- 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.



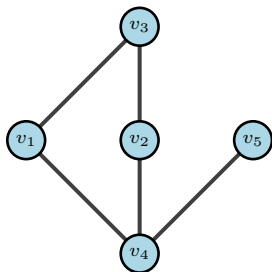
a cactus graph with 3 buds

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



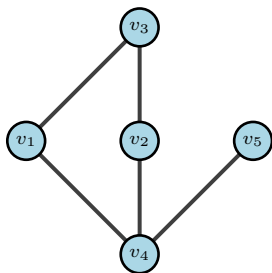
Degree Matrix: $\Delta(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$

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

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

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:

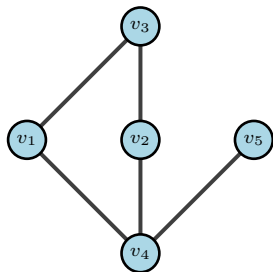
Homework

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



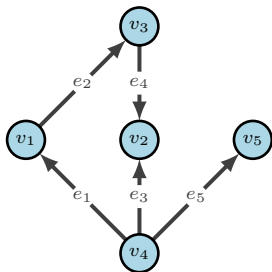
Incidence Matrix: $E(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{E}|}$

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: $\mathcal{G} = \cup_{i=1}^c \mathcal{G}_i$
- ▶ 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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- ▶ 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.

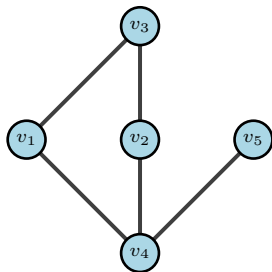


$$\dot{x} = f(x_i, u_i)$$
$$y = \begin{bmatrix} \vdots \\ x_i - x_j \\ \vdots \end{bmatrix}$$

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 $x_i - x_j$

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

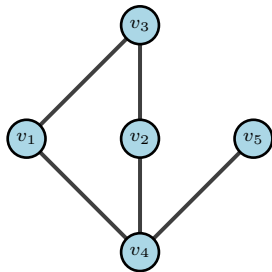
Graphs and their properties can be studied using matrices and constructs from linear algebra



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

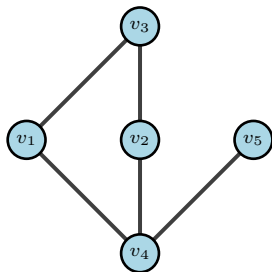


Combinatorial Graph Laplacian: $L(\mathcal{G}) \in \mathbb{R}^{|\mathcal{V}| \times |\mathcal{V}|}$
Constructions

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

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

- ▶ $\text{rank } L(\mathcal{G}) = |\mathcal{V}| - 1 \Leftrightarrow \mathcal{G} \text{ is connected}$
- ▶ $\mathcal{G}$ is connected, then 0 is a simple eigenvalue and $L(\mathcal{G})\mathbf{1} = 0$
- ▶ $L(\mathcal{G})$ 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) $\lambda_2(\mathcal{G}) > 0$.

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

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

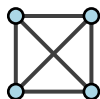
- $\det M_{(ij)}$ is called the ij -minor of M

MATRIX-TREE THEOREM

Theorem

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

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



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

