

A First Course in Network Theory

Reminders and Basics about Graphs and Networks

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Generalities and Interest

Complex Networks are Graphs used to represent **complex systems**.

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Complex Systems *Systems composed of a **large number of simple elements in interaction** and exhibiting emerging phenomena¹.*

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Examples of complex networks

- **Social networks,**
- Biological networks,
- Transports,
- Language,
- Technological,
- Etc.

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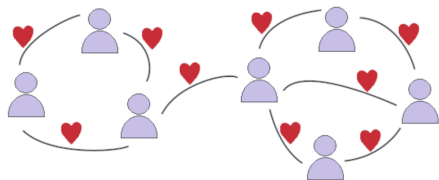
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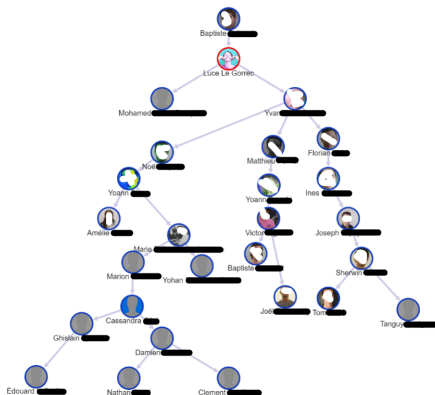
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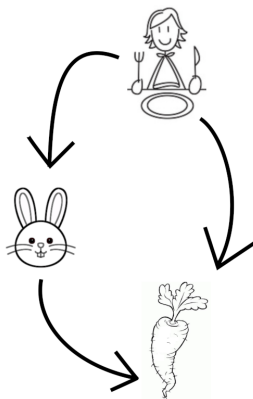
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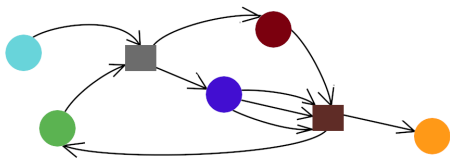
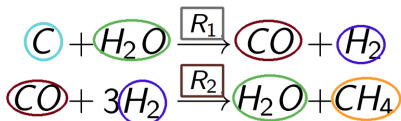
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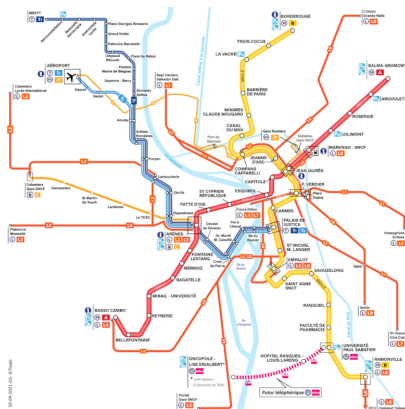
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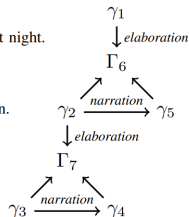
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γ_1 : John had a lovely evening last night.
 γ_2 : He had a great meal.
 γ_3 : He ate salmon.
 γ_4 : He devoured lots of cheese.
 γ_5 : He won a dancing competition.
 Γ_6 : $\gamma_2 \cup \gamma_5$
 Γ_7 : $\gamma_3 \cup \gamma_4$



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Basics – Definitions

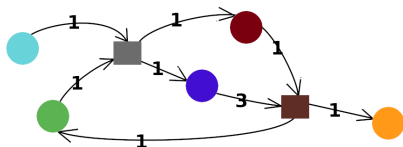
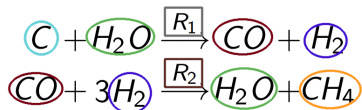
Definition A **Graph or Network** $G = (V, E, \omega)$ is a tuple of:

- a set V , called the vertex set (or node set). Elements of V are called **vertices or nodes**.
- a set $E \subset V \times V$, called the edge set (or link set). Elements in E are called **edges or links**.
- an application $\omega : E \rightarrow \Omega$, called the **edge weight function**.

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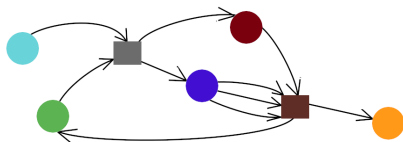
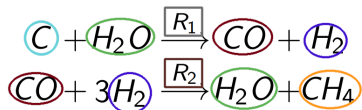
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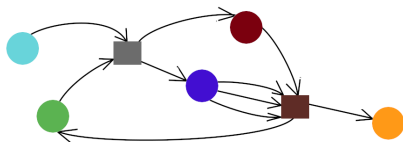
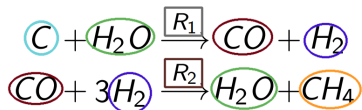


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Remark With this definition, information about nodes is lost.

Basics – Definitions

Definitions Given a graph $G = (V, E, \omega)$, and an edge $e \in E$:

- If $e(1) = e(2)$, then e is called a **(self-)loop**.

NB If $\forall e \in E$, e is **not** a self-loop, then G is called **anti-reflexive**.

- When $\omega : E \rightarrow \Omega = \{\omega_0\}$, then G is called an **unweighted graph** and ω may be omitted: $G = (V, E)$.

- If $\exists f \in E : \begin{cases} f(1) = e(2) \\ f(2) = e(1) \\ \omega(f) = \omega(e) \end{cases}$, then e is called a **bi-directed edge**.

NB If all edges are bi-directed, then G is said to be **undirected or non-directed**. One can choose to count each edge only once.

Definition A graph is called **simple** if it is i) non-directed; ii) unweighted; iii) anti-reflexive.

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Representing Graphs – Sets, Drawings and Matrices

Sets and Drawings

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Matrix representations We enforce $V = \{1, \dots, n\}$ with $n = |V|$.

- **Incidence:** Given $G = (V, E)$ anti-reflexive and unweighted, with $m = |E = \{e_1, \dots, e_m\}|$, the so-called **incidence matrix of G** is a

$$\text{matrix } \mathbf{B} \in \{-1, 0, 1\}^{n \times m} \text{ s.t. } b(i, k) = \begin{cases} -1 & \text{if } e_k(1) = i \\ 1 & \text{if } e_k(2) = i \\ 0 & \text{otherwise} \end{cases}$$

NB For undirected graphs: two definitions.

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Question: How many adjacency matrices/incidence matrices possible for one graph ?

Representing Graphs – Isomorphisms

Definition Two graphs $G_1 = (V_1, E_1, \omega_1)$ and $G_2 = (V_2, E_2, \omega_2)$ are called **isomorphic** if

- $|V_1| = |V_2|$.
- $|E_1| = |E_2|$.
- $\exists s : V_1 \rightarrow V_2$ a bijection s.t. $(i, j) \in E_1 \iff (s(i), s(j)) \in E_2$.
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Characterisation Two graphs of adjacency matrices respectively \mathbf{A}_1 and \mathbf{A}_2 are isomorphic iff \mathbf{A}_1 can be obtained from **simultaneous permutations of rows and columns** of \mathbf{A}_2 .

Representing Graphs – Permutation matrices

Definition A matrix $\mathbf{M} \in \mathbb{R}_+^{n \times n}$ is called **bi-stochastic** if $\mathbf{M}\mathbf{1} = \mathbf{M}^T\mathbf{1} = \mathbf{1}$.

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Exercise Prove that $\mathbf{P}^T = \mathbf{P}^{-1}$.

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$$\forall \mathbf{M} \in \mathbb{R}^{n \times n} \begin{cases} \mathbf{P}\mathbf{M} = \mathbf{M}([\sigma(1), \dots, \sigma(n)], :) \\ \mathbf{M}\mathbf{P}^T = \mathbf{M}(:, [\sigma(1), \dots, \sigma(n)]) \end{cases}$$

Proof Exercise

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Connectivity – Neighbourhood and degrees

Definition Given a graph $G = (V, E, \omega)$ and a node $v \in V$, we call:

- The **out-neighbourhood of v** the set $\mathcal{N}_{out}(v) = \{u : (v, u) \in E\}$,
Its cardinal is called **the out-degree of v** : $d_{out}(v) = |\mathcal{N}_{out}(v)|$.
- The **in-neighbourhood of v** the set $\mathcal{N}_{in}(v) = \{u : (u, v) \in E\}$.
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Exercise Given $G = (V, E)$ an unweighted graph, express the in- and out-degrees of a node $v \in V$ using G 's adjacency matrix, then G 's incidence matrix.

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Exercise Given $G = (V, E)$ an unweighted graph, express the in- and out-degrees of a node $v \in V$ using G 's adjacency matrix, then G 's incidence matrix.

Connectivity – Interlude on weighted degrees

Definition Given a graph $G = (V, E, \omega)$ and a node $v \in V$, thus:

- The **weighted out-degree of v** is

$$d_{out}^{\omega}(v) = \sum_{u:(v,u) \in E} \omega((v, u)).$$

- The **weighted in-degree of v** is

$$d_{in}^{\omega}(v) = \sum_{u:(u,v) \in E} \omega((u, v)).$$

- If G is **non-directed**, the **weighted degree of v** is

$$d^{\omega}(v) = \sum_{u:\{u,v\} \in E} \omega((u, v)).$$

Exercise Explain the limitation of weighted degree to non-directed graph. Any idea for extending this notion to directed graphs?

Connectivity – Paths

Definition Given a graph $G = (V, E, \omega)$, and $u, v \in V$, a **path from u to v** is a sequence of edges $e_1, \dots, e_k \in E$ s.t.

$$i) : e_1(1) = u, \quad ii) : e_k(2) = v, \quad iii) : \forall i \in \{1, \dots, k-1\}, e_i(2) = e_{i+1}(1).$$

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Property Given $\mathbf{A} \in \{0, 1\}^{n \times n}$ the adjacency matrix of an unweighted graph, **the value of $a^k(i, j)$ is the number of k -paths from node i to node j .**

Proof Exercise.

Connectivity – Irreducibility

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Characterisation **Irreducible matrices** are adjacency matrices of **strongly connected graphs**.

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