

# Graph Theory and Complex Networks: An Introduction

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## Chapter 06: Network analysis

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02: Foundations	Basic terminology and properties of <a href="#">graphs</a>
03: Extensions	Directed & weighted graphs, colorings
04: Network traversal	Walking through graphs (cf. <a href="#">traveling</a> )
05: Trees	Graphs without <a href="#">cycles</a> ; routing algorithms
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# Introduction

## Observation

In real-world situations, graphs (or networks) may become very large, making it difficult to (visually) discover properties  $\Rightarrow$  we need **network analysis** tools.

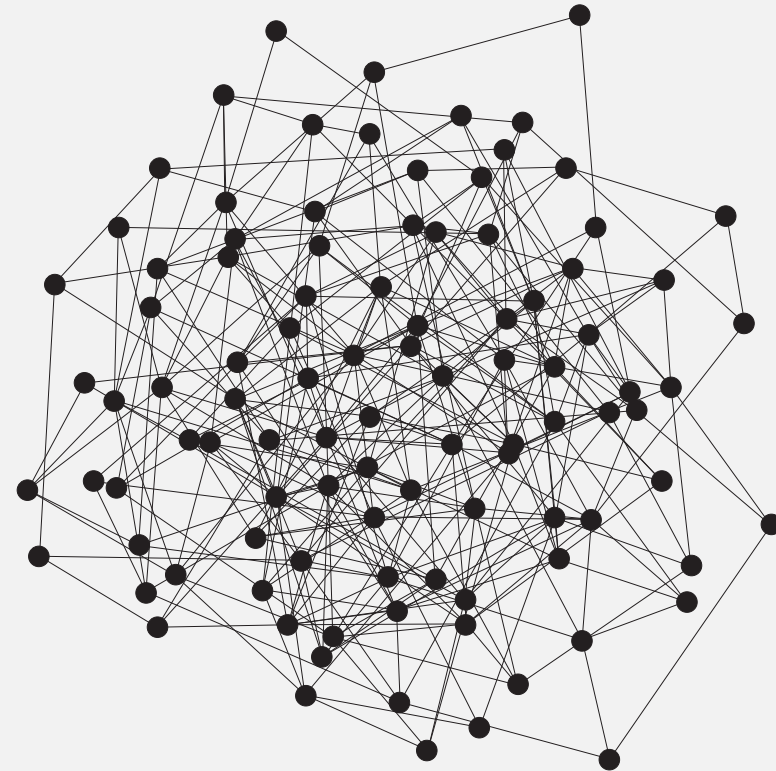
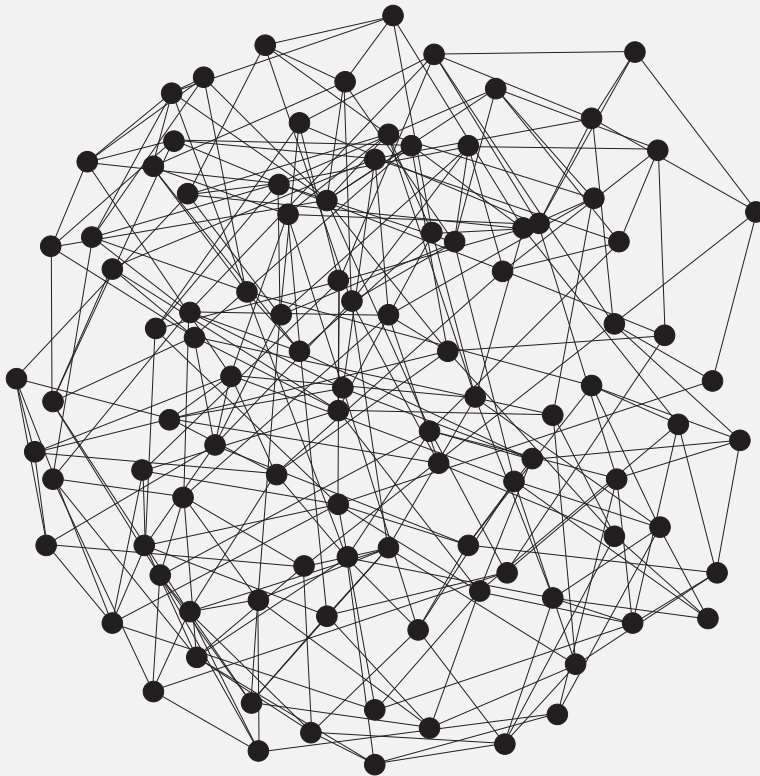
**Vertex degrees:** Consider the **distribution** of degrees: how many vertices have high degrees versus the number of vertices with low degrees.

**Distance statistics:** Focus on where vertices are **positioned** in the network: far away from each other, central in the network, etc.

**Clustering:** To what extent are my neighbors also adjacent to each other?

**Centrality:** Are there vertices that are **more important** than others?

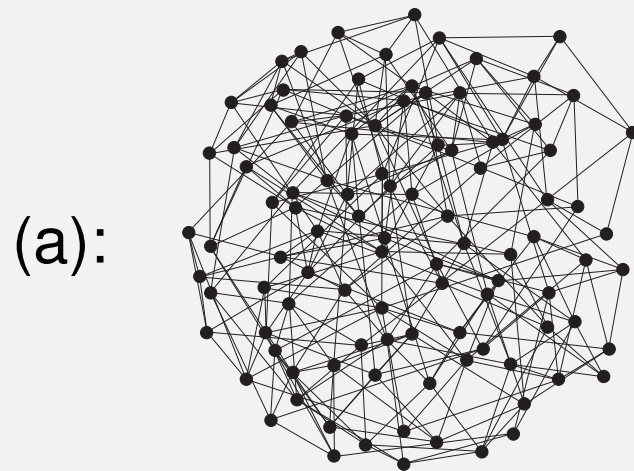
# Vertex degree



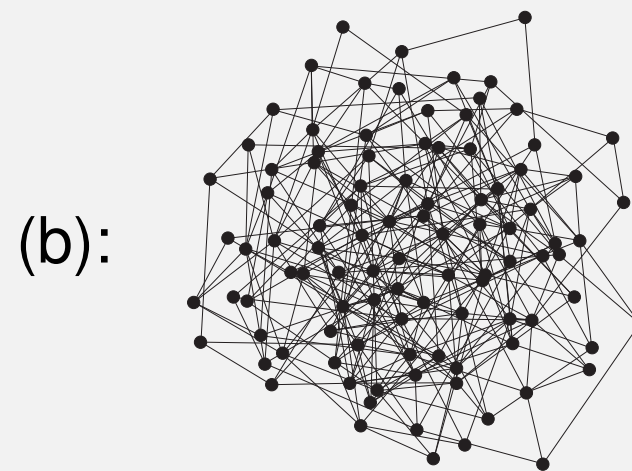
## Question

Can you visually observe real (nonisomorphic) differences?

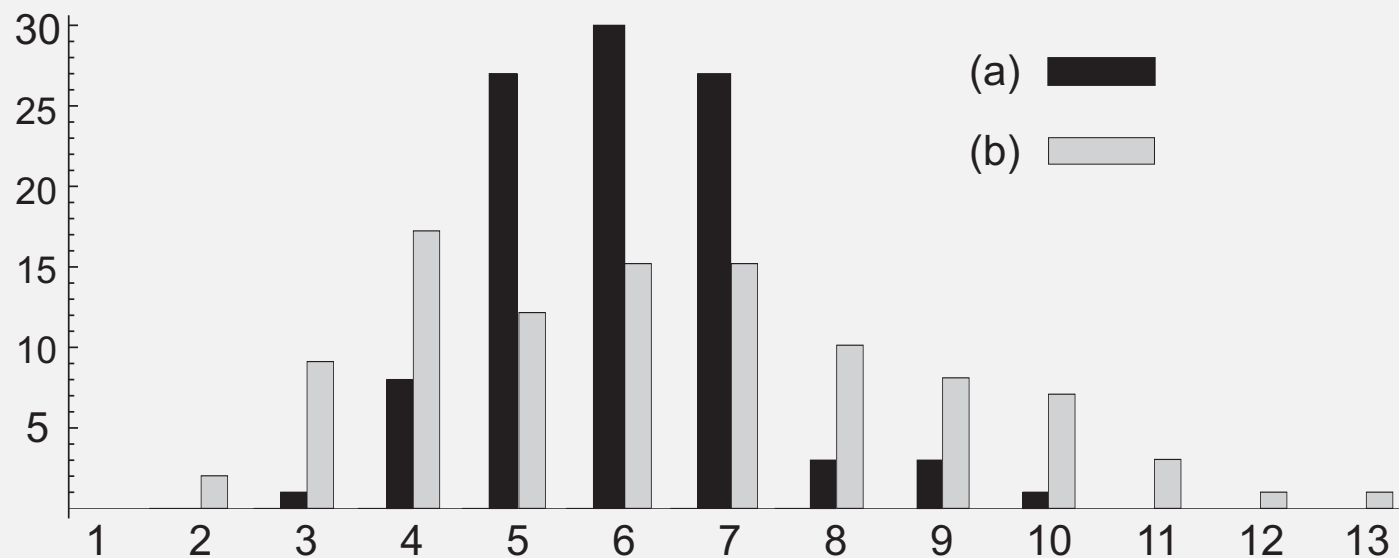
# Vertex degree: Histogram



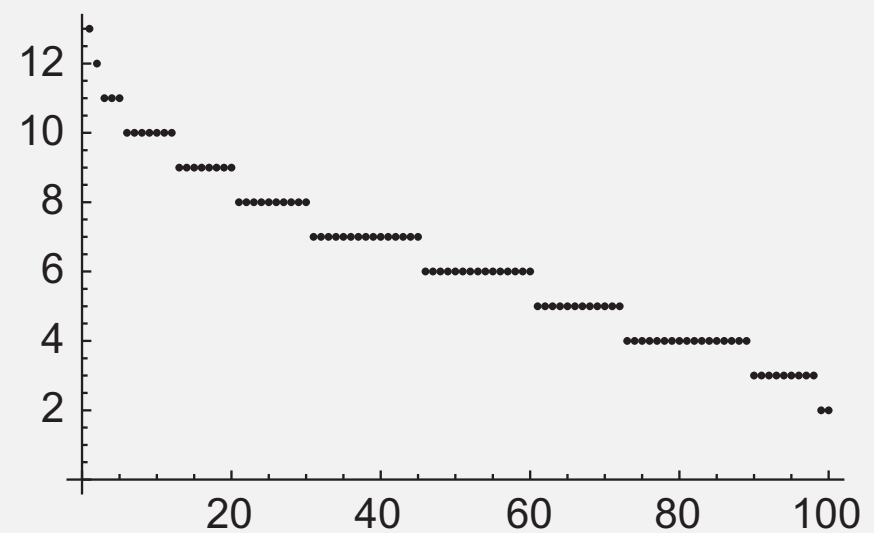
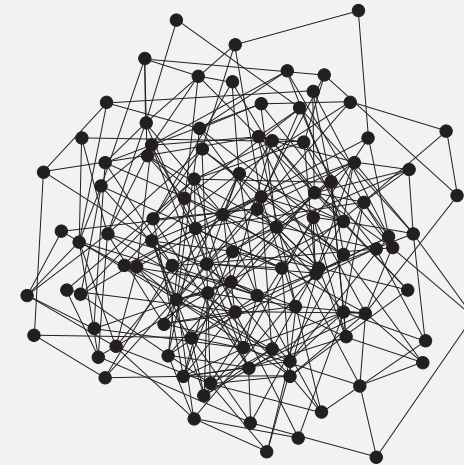
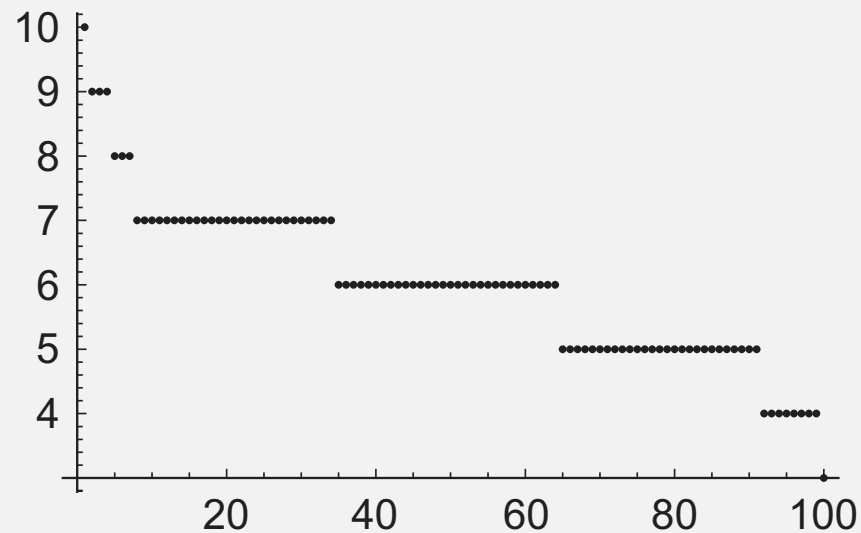
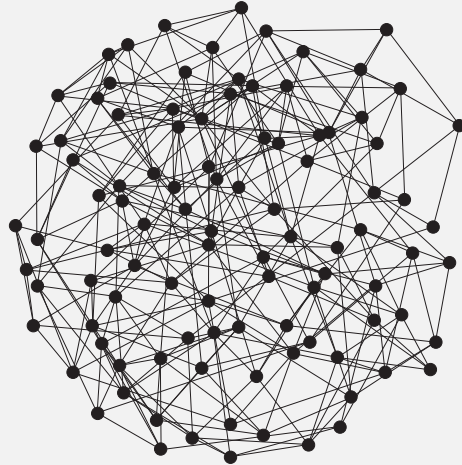
$n = 100, m = 300$



$n = 100, m = 317$



# Vertex degree: Ranked histogram



# Distance statistics

## Definition

$G$  is connected,  $d(u, v)$  is distance between vertices  $u$  and  $v$ : the **length** of a shortest path between  $u$  and  $v$ .

**Eccentricity**  $\varepsilon(u)$ :  $\max\{d(u, v) \mid v \in V(G)\}$

**Radius**  $rad(G)$ :  $\min\{\varepsilon(u) \mid u \in V(G)\}$

**Diameter**  $diam(G)$ :  $\max\{d(u, v) \mid u, v \in V(G)\}$

## Note

Note that these definitions apply to directed as well as undirected graphs.

# Path lengths

## Definition

$G$  is connected with vertex  $V$ ;  $\bar{d}(u)$  is average **length** of shortest paths from  $u$  to any other vertex  $v$ :

$$\bar{d}(u) \stackrel{\text{def}}{=} \frac{1}{|V| - 1} \sum_{v \in V, v \neq u} d(u, v)$$

The **average path length**  $\bar{d}(G)$ :

$$\bar{d}(G) \stackrel{\text{def}}{=} \frac{1}{|V|} \sum_{u \in V} \bar{d}(u) = \frac{1}{|V|^2 - |V|} \sum_{u, v \in V, u \neq v} d(u, v)$$



# Path lengths

## Definition

The **characteristic path length** is the **median** over all  $\bar{d}(u)$ .

## Note

The median over  $n$  nondecreasing values  $x_1, x_2, \dots, x_n$ :

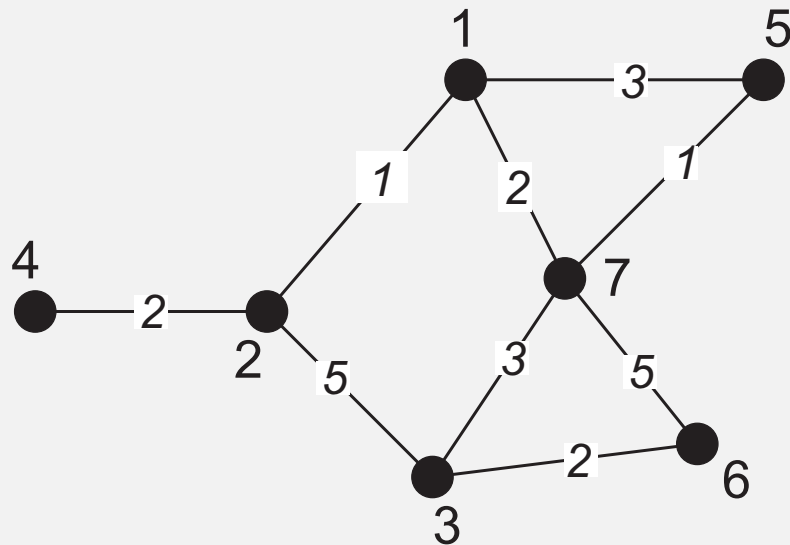
- $n$  odd  $\Rightarrow x_{(n+1)/2}$
- $n$  even  $\Rightarrow (x_{n/2} + x_{n/2+1})/2$

The median separates the higher values from the lower values into two equally-sized subsets.

## Example

$\{3, 4, 4, 6, 0, 6, 1\} \Rightarrow [0, 1, 3, 4, 4, 6, 6] \Rightarrow M = x_{(7+1)/2} = x_4 = 4$

# Example distance statistics



Vertex	1	2	3	4	5	6	7	$\varepsilon(u)$	$\sum_{v \neq u} d(u, v)$	$\bar{d}(u)$
1	0	1	5	3	3	7	2	7	21	3.50
2	1	0	5	2	4	7	3	7	22	3.67
3	5	5	0	7	4	2	3	7	26	4.33
4	3	2	7	0	6	9	5	9	32	5.33
5	3	4	4	6	0	6	1	6	24	4.00
6	7	7	2	9	6	0	5	9	36	6.00
7	2	3	3	5	1	5	0	5	19	3.17

# Clustering coefficient

## Observation

Many networks show a high degree of **clustering**: my neighbors are each other's neighbors.

## Note

An extreme case is formed by having all my neighbors be adjacent to each other  $\Rightarrow$  neighbors form a **complete graph**.

## Question

What is the other extreme case?

# Clustering coefficient

## Definition

$G$  is simple, connected, undirected. Vertex  $v \in V(G)$  with neighborset  $N(v)$ .

- Let  $n_v = |N(v)|$ .

**Note:** max. number of edges between neighbors is  $\binom{n_v}{2}$ .

- Let  $m_v$  is number of edges in subgraph induced by  $N(v)$ :  
 $m_v = |E(G[N(v)])|$ .

**Clustering coefficient**  $cc(v)$ :

$$cc(v) \stackrel{\text{def}}{=} \begin{cases} m_v / \binom{n_v}{2} = \frac{2 \cdot m_v}{n_v(n_v-1)} & \text{if } \delta(v) > 1 \\ \text{undefined} & \text{otherwise} \end{cases}$$

# Clustering coefficient

## Definition

$G$  is simple, connected and undirected.

Let  $V^* \stackrel{\text{def}}{=} \{v \in V(G) \mid \delta(v) > 1\}$ .

Clustering coefficient  $CC(G)$  for  $G$ :

$$CC(G) \stackrel{\text{def}}{=} \frac{1}{|V^*|} \sum_{v \in V^*} cc(v)$$

# Clustering coefficient: triangles

## Definition

A **triangle** is a **complete (sub)graph** with exactly 3 vertices. A **triple** is a (sub)graph with exactly 3 vertices and 2 edges.

## Definition

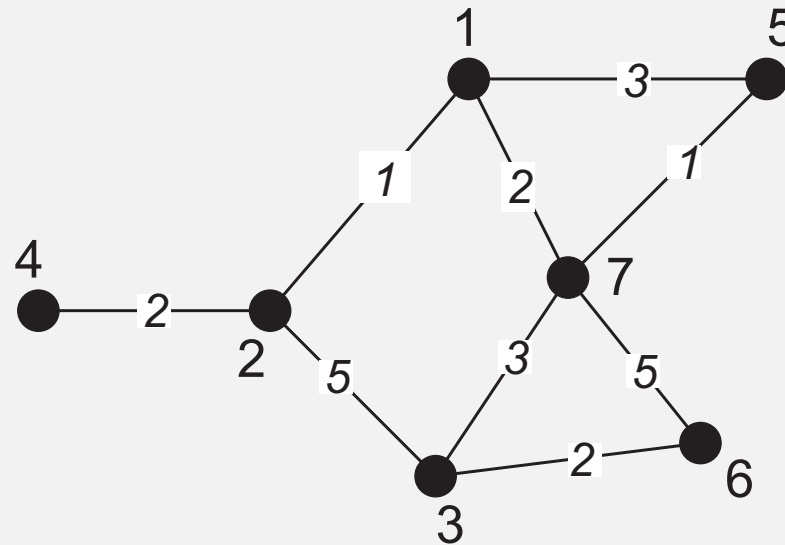
$G$  is simple and connected with  $n_{\Delta}(G)$  distinct triangles and  $n_{\wedge}(G)$  distinct triples.

The **network transitivity**  $\tau(G) \stackrel{\text{def}}{=} n_{\Delta}(G)/n_{\wedge}(G)$ .

## Notation

A **triple at  $v$** :  $v$  is incident to both edges (“in the middle”).  $n_{\wedge}(v)$ : number of triples at  $v$ .

# Clustering coefficient: example



Vertex:	1	2	3	4	5	6	7
$cc:$	$1/3$	0	$1/3$	<i>undefined</i>	1	1	$1/3$
$n_{\Lambda}:$	3	3	3	0	1	1	6

**Vertex 1**  $N(1) = \{2, 5, 7\}; E(G[N(1)]) = \langle 5, 7 \rangle \Rightarrow cc(1) = \frac{1}{3}$   
 Triples at 1:  $G[\{2, 1, 5\}], G[\{2, 1, 7\}], G[\{5, 1, 7\}]$

# Clustering coefficient versus transitivity

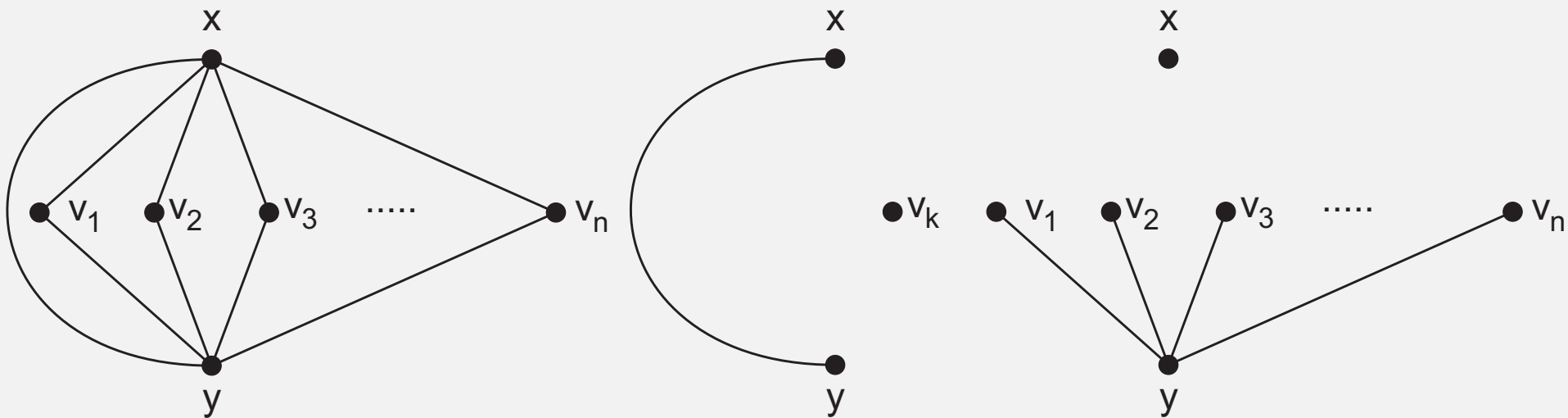
## Observation

Let  $n_{\Delta}(v)$  be the number of triangles of which  $v$  is member  $\Rightarrow$

- $cc(v) = \frac{n_{\Delta}(v)}{n_{\Lambda}(v)}$
- $n_{\Lambda}(v) = \binom{\delta(v)}{2}$
- $n_{\Delta}(G) = \frac{1}{3} \sum_{v \in V^*} n_{\Delta}(v)$  (Note:  $V^* = \{v \in V \mid \delta(v) > 1\}$ )



# Clustering coefficient versus transitivity

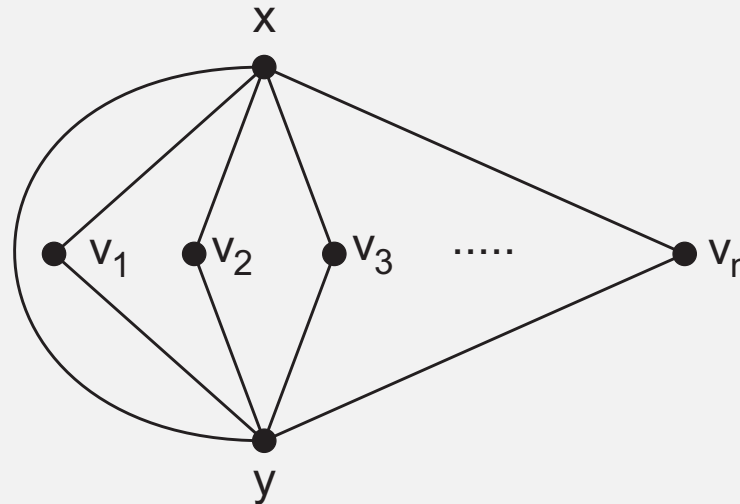


$$G_k = G[\{x, y, v_1, v_2, \dots, v_k\}] \Rightarrow:$$

$$cc(u) = \begin{cases} 1 & \text{if } u = v_1, \dots, v_k \\ \frac{k}{\binom{k+1}{2}} = \frac{k}{\frac{1}{2} \cdot k(k+1)} = \frac{2}{k+1} & \text{if } u = x \text{ or } u = y \end{cases}$$

$$CC(G_k) = \frac{1}{k+2} \left( 2 \cdot \frac{2}{k+1} + k \cdot 1 \right) = \frac{k^2 + k + 4}{k^2 + 3k + 2} \Rightarrow \lim_{k \rightarrow \infty} CC(G_k) = 1$$

# Clustering coefficient versus transitivity



$$G_k = G[\{x, y, v_1, v_2, \dots, v_k\}] \Rightarrow$$

$$n_{\Delta}(u) = \begin{cases} 1 & \text{if } u = v_1, \dots, v_k \\ \binom{\delta(u)}{2} = \binom{k+1}{2} & \text{if } u = x, y \end{cases}$$

$$\tau(G_k) = \frac{n_{\Delta}(G_k)}{\sum n_{\Delta}(u)} = \frac{k}{2 \cdot \frac{1}{2} \cdot k(k+1) + k} = \frac{1}{k+2} \Rightarrow \lim_{k \rightarrow \infty} \tau(G_k) = 0$$

# Centrality

## Issue

Are there any vertices that are more important than the others?

## Definition

$G$  is (strongly) connected. The **center**  $C(G)$  is the set of vertices with minimal eccentricity:

$$C(G) \stackrel{\text{def}}{=} \{v \in V(G) \mid \varepsilon(v) = \text{rad}(G)\}$$

## Intuition

At the center means at minimal distance to the farthest node.

# Vertex centrality

## Definition

$G$  is (strongly) connected. The (eccentricity based) vertex centrality  $c_E(u)$  of  $u$ :

$$c_E(u) \stackrel{\text{def}}{=} \frac{1}{\varepsilon(u)}$$

## Intuition

The higher the centrality, the “closer” to the center of a graph.

# Closeness

## Definition

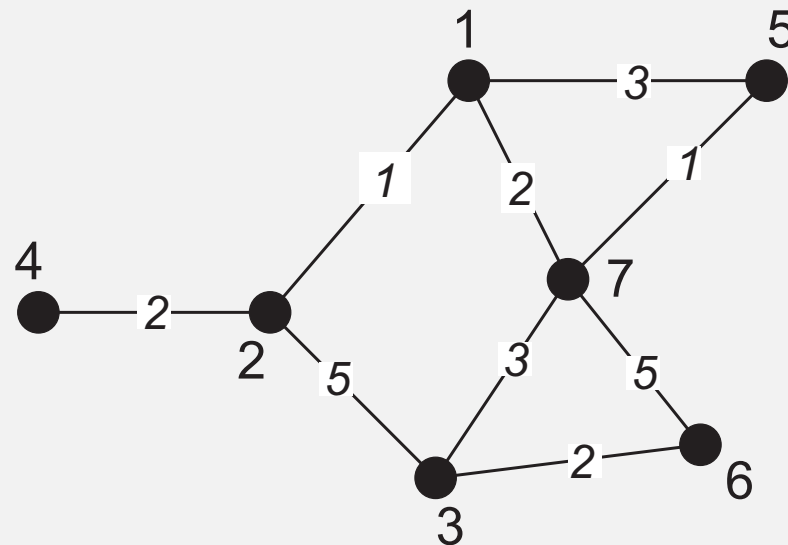
$G$  is (strongly) connected. The closeness  $c_C(u)$  of  $u$ :

$$c_C(u) \stackrel{\text{def}}{=} \frac{1}{\sum_{v \in V(G)} d(u, v)}$$

## Intuition

How close is a vertex to **all** other nodes?

# Centrality: example



Vertex:	1	2	3	4	5	6	7
$\varepsilon(u)$	7	7	7	9	6	9	5
$\sum d(u, \cdot)$	21	22	27	32	24	37	29
$c_C(u)$ :	0.048	0.045	0.037	0.031	0.042	0.027	0.034

# Betweenness

## Intuition

Important vertices are those whose removal significantly increases the distance between other vertices. **Example:** cut vertices.

## Definition

$G$  is simple and (strongly) connected.  $S(x, y)$  is set of shortest paths between  $x$  and  $y$ .  $S(x, u, y) \subseteq S(x, y)$  paths that pass through  $u$ .

**Betweenness centrality**  $c_B(u)$  of  $u$ :

$$c_B(u) \stackrel{\text{def}}{=} \sum_{x \neq y \neq u} \frac{|S(x, u, y)|}{|S(x, y)|}$$