## Social Networks & Recommendation Systems

II. Historical overview of the complex network science. Examples of the real-life networks.

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MSc program in Data Science has been developed as a part of task 10 of the project "NERW PW. Science - Education - Development - Cooperation" co-funded by European Union from European Social Fund.

## Before classes

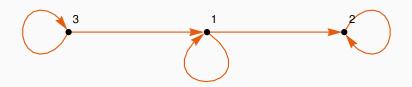
## Remember: graph representation methods

#### Adjacency matrix

$$A = \left[ \begin{array}{rrr} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{array} \right].$$

#### Adjacency list

$$L = \{\{1, 2\}, \{2\}, \{1, 3\}\}.$$



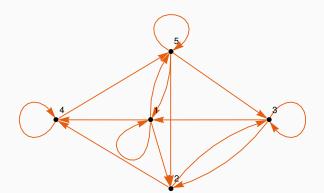
## Remember: vertex degree

#### Vertex degree

Number of incoming or outcoming edges.

$$k_{out} = \{4, 2, 3, 2, 4\},$$

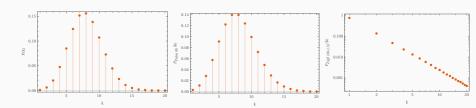
$$k_{in} = \{3, 3, 3, 3, 3\}.$$



## Remember: discrete probability distributions

## Discrete probability distributions – examples

- binomial distribution,
- · Poissona distribution,
- · Zipf distribution.



$$P(k) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$P(k) = \frac{\lambda^k e^{-\lambda}}{k!}$$

$$P(k) = \frac{1/k^{s}}{H_{N,s}}$$

Reminder

#### Exercise to think about - continuous distributions

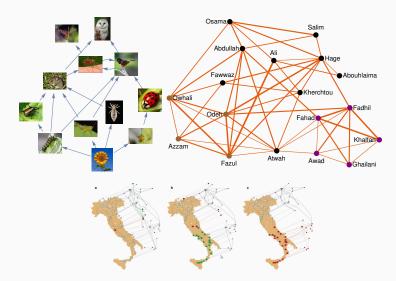
In complex networks, you often replace discrete distributions with continuous ones (this is the way we'll think during class). Find continuous analogs of distributions from the previous slide.

## Lecture

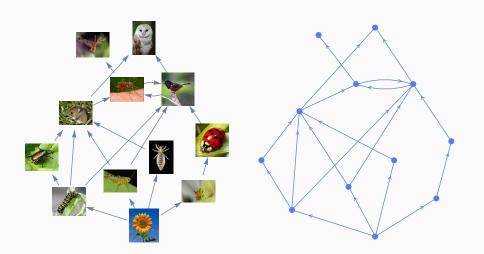
## Why are networks/graphs so useful?

#### Graphs are an illustration of a relationship

What relations are illustrated in the following graphs?



## What is the difference between graphs and networks?



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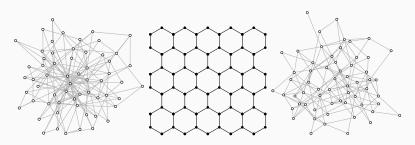
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Which of these graphs represents the real network?

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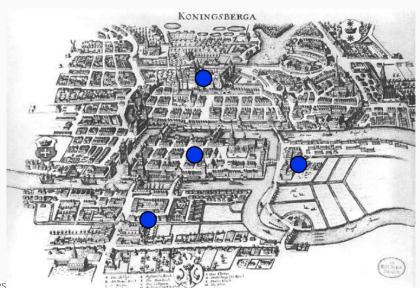
#### with expert knowledge from

- sociology,
- · economics,
- · biology,
- · medicine,
- · engineering,
- · and many other...

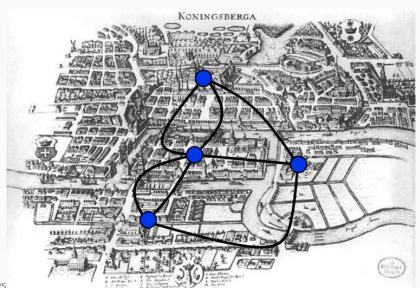
## The story of network science – Seven Bridges of Königsberg (1736)



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**SNARS** 

Source: 12

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- · we do not focus on this line of research (with one exception!).

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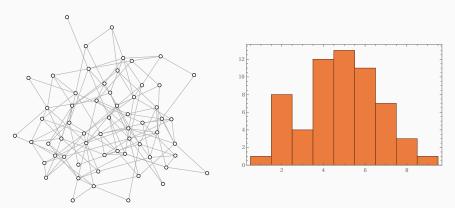
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More details on Lecture 5.

#### Degree distribution

P(k) = fraction of vertices with degree k



Is this a real network?

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More details during Lecture 4.

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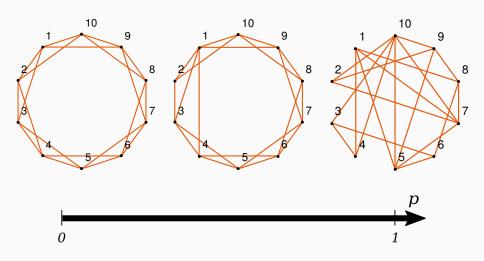
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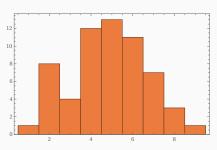
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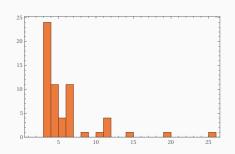
More details during Lecture 5.



#### Problem:

Real-world networks usually have power law like distributions.





#### What does it means?

- · No typical scale.
- · Fat tails.
- Fast spreading epidemics...

#### BA model

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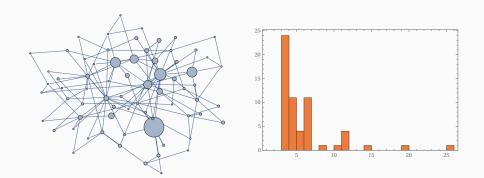
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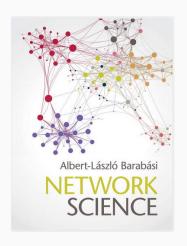
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## If you are interested in the history of complex networks read





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· Visualization of networks (both theoretical and real) (Lecture 3).

Thank you for your attention!

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