

# From Graph Theory to Modern Network Science

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Critical TW / INESC TEC & CRACS  
Data Science Portugal Presentation

## Why Graphs Are Ubiquitous

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# Syllabus

## 1 Introduction

- Who am I?
- Genesis of graph theory and definitions
- Is the definition of Network necessary?

## 2 Finding Patterns

- The Subgraph Census Problem
- Network Motifs
- Graphlet Degree Distributions
- Use Cases

## 3 How to find these Patterns

- Enumerating Subgraphs
- Symmetry Breaking

## 4 Concluding Remarks

- Useful resources and tools
- Wrap up and Q&A

# Who am I?

- A Data Scientist specially enthusiastic about Deep Learning.
- Personal mission of teaching the computer to see and speak.
- Researcher in Complex Networks and sees graphs everywhere.
- Lover of music, philosophy, drinks and uh... **data** of course.

# Genesis of Graph Theory

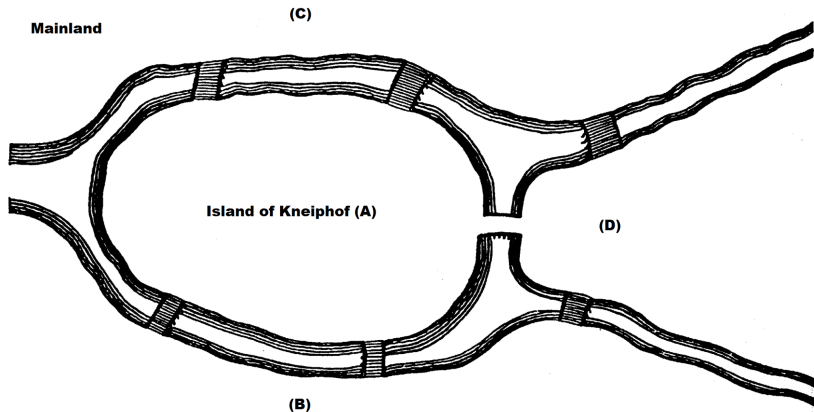


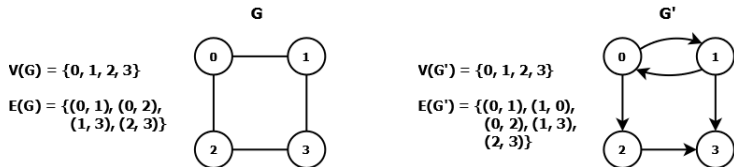
Figure 1: Representation of the diagram of the Seven Bridges of Königsberg presented by Euler in 1735 (Adapted from [1]).

# So what is a graph, strictly speaking?

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## Definition - Graph

A graph  $G = (V, E)$ , where  $V(G)$  is its set of *vertices* that are connected by a set of *edges*,  $E(G)$ .



**Figure 2:** An undirected graph  $G$  (left) and a directed graph  $G'$  (right). Note how the direction of the edges affects the set of edges of  $G'$ .

# Genesis of Graph Theory

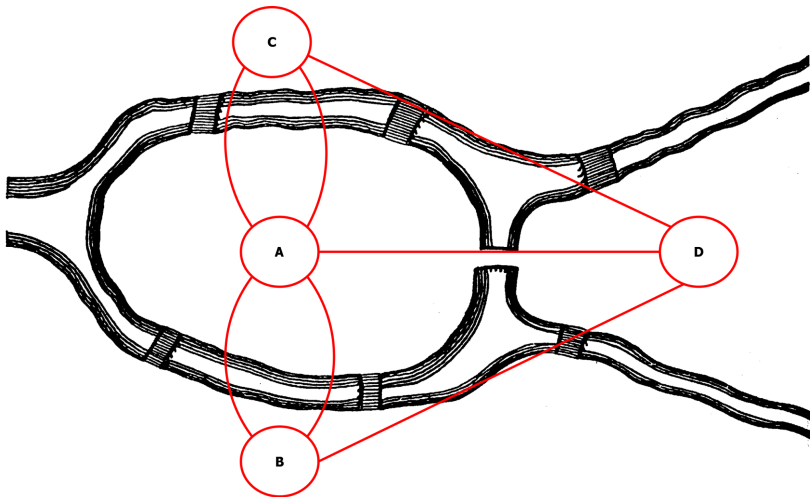
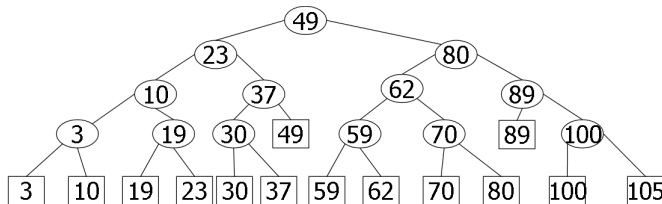


Figure 3: The city of Königsberg represented as a graph.

# Spot the Graph





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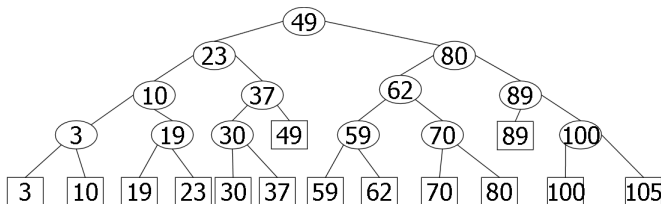
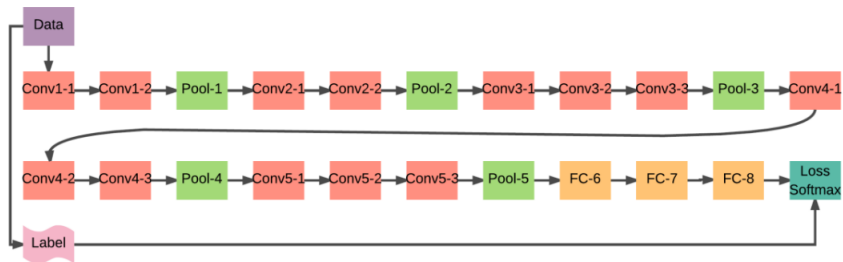


Figure 4: A balanced tree.

# Spot the Graph



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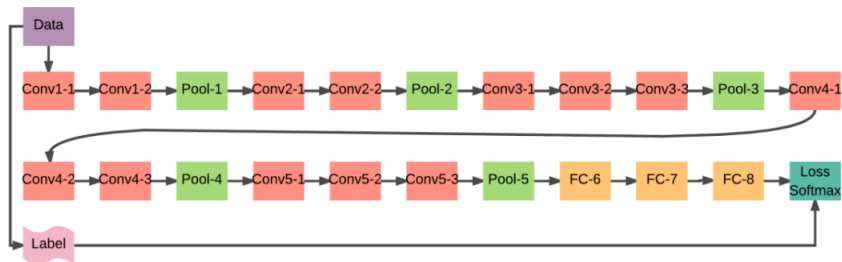
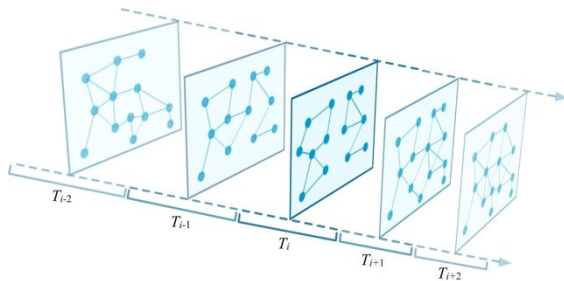


Figure 5: VGG16 Convolutional NN (Source [2])

# Spot the Graph



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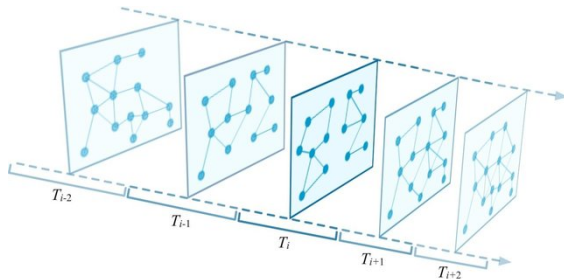
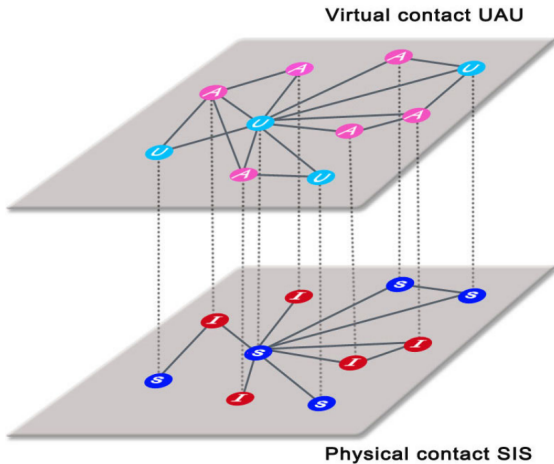


Figure 6: Abstraction of a time-evolving Network (Source [3])

# Spot the graph



# Spot the graph

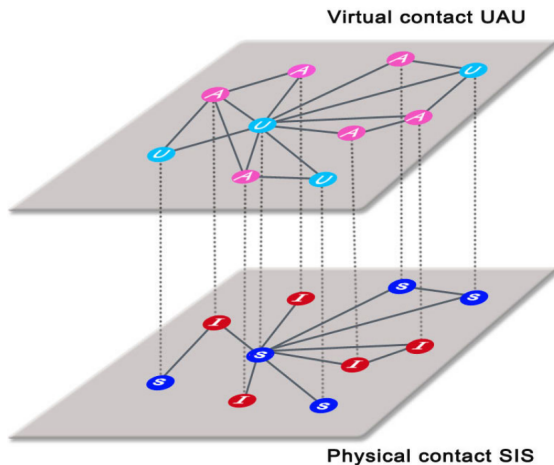


Figure 7: Epidemic multiplex Network (source [4])

## Definition - The Subgraph Census Problem

Given some positive integer  $k$  and a graph  $G$ , count the exact number of distinct occurrences of each of all possible connected induced  $k$ -subgraphs of  $G$ . Two occurrences are distinct if there is at least one vertex that they do not share.

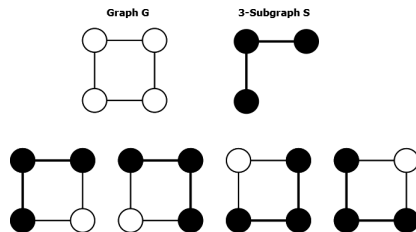


Figure 8: All occurrences of a subgraph of size 3 (right) on graph  $G$  (left).



# Network Motifs

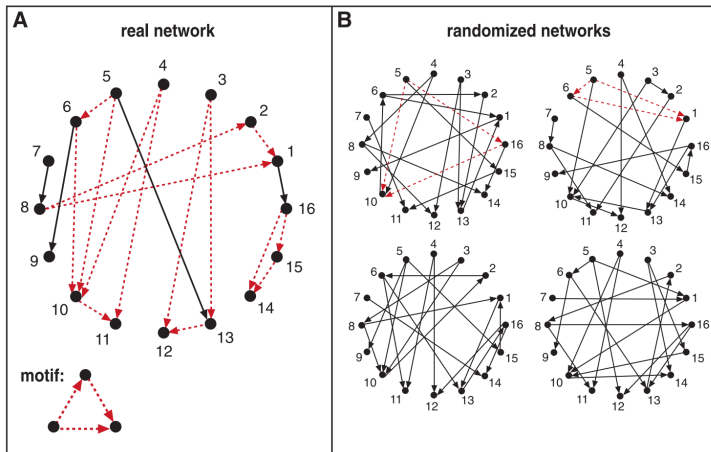


Figure 9: Given a network  $G$  a motif  $M$  is an induced subgraph that appears more often than expected (source: [5]).

# Network Motifs

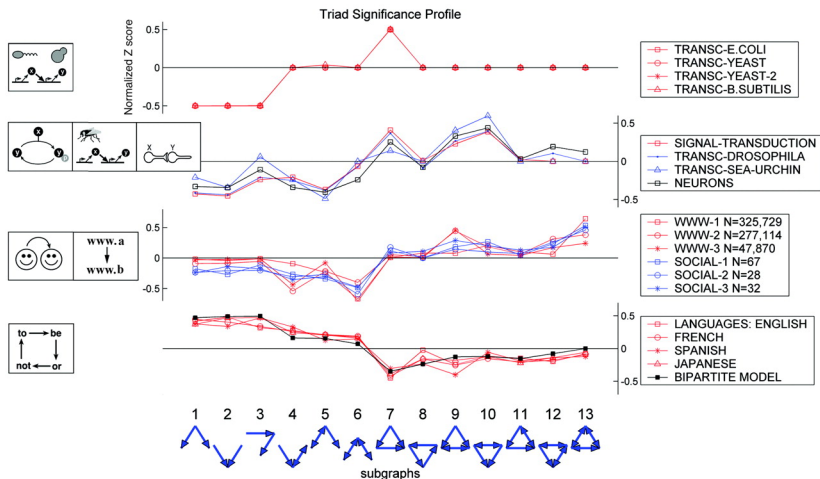


Figure 10: The triad significance profile from various disciplines (source: [6]).

# Graphlet Degree Distributions

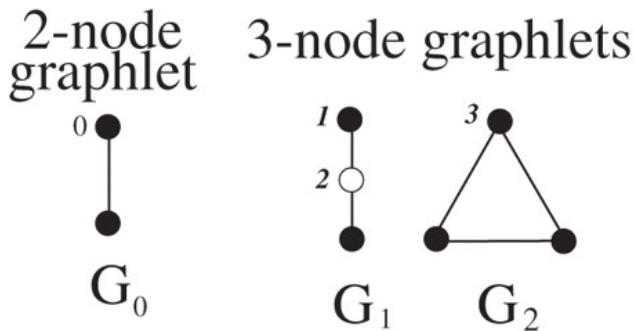


Figure 11: Orbits for all possible graphlets from sizes 2 to 3 (Adapted: [7]).

# Graphlet Degree Distributions

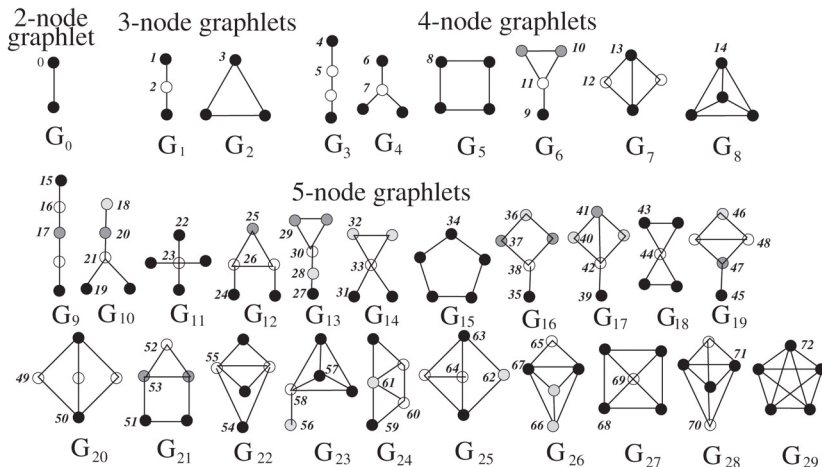


Figure 12: Orbits for all possible graphlets from sizes 2 to 5 (Source: [7]).

# Graphlet Degree Distributions

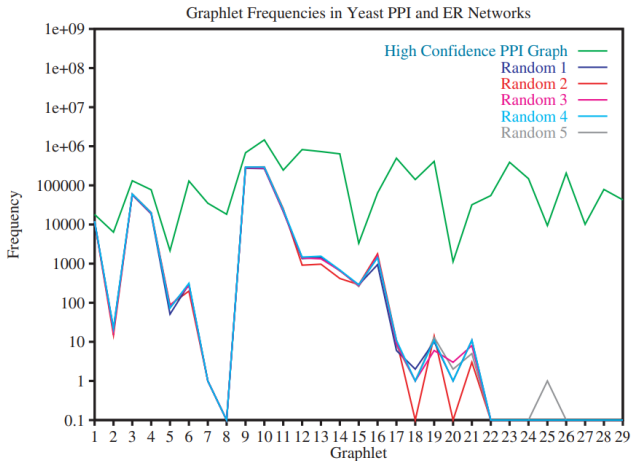


Figure 13: PPI Networks versus Random Networks (Source: [8]).

# How to find these patterns efficiently? **ESU**

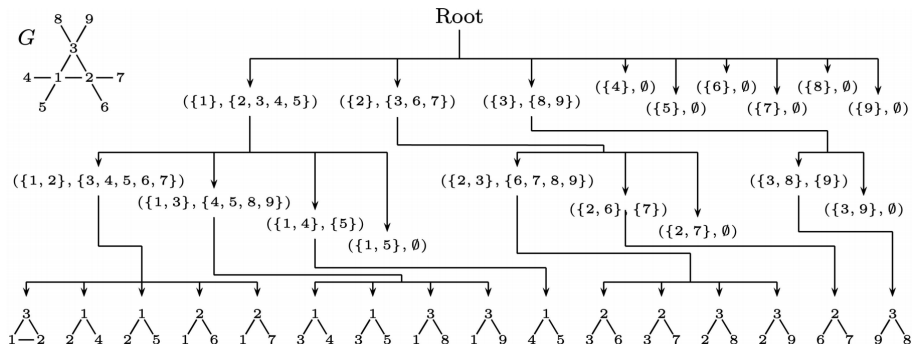


Figure 14: ESU-TREE of a graph  $G$  (Source: [9]).

# Are Graphs/Networks ubiquitous or not?

# Useful Resources and tools

- [Network Science Book by Barabási](#).
- [NetworkX](#): a Python Package for Network Analysis.
- [geffy](#): an open-source Network visualization tool.
- [gTrieScanner](#): original code for G-Tries.
- Check FasE [10], an algorithm which offers a general framework using an adaptive version of G-Tries.
- [Condensation Decondensation Framework](#): a POC extending these algorithms by yours truly.
- All the references in this presentation.
- Any questions, feel free to contact me at:  
[miguelopesmartins@gmail.com](mailto:miguelopesmartins@gmail.com)



**That's all Folks!**  
Thank you for your time :)

# References I



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