

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

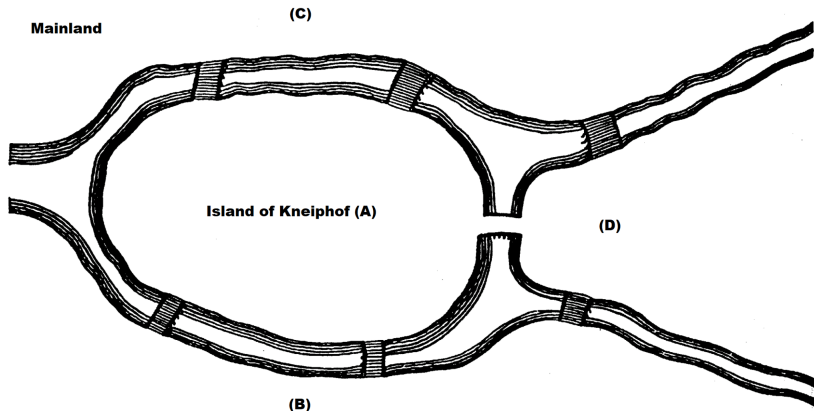


Figure 1: Representation of the diagram of the Seven Bridges of Königsberg presented by Euler in 1735 (Adapted from [1]).

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*, $E(G)$.

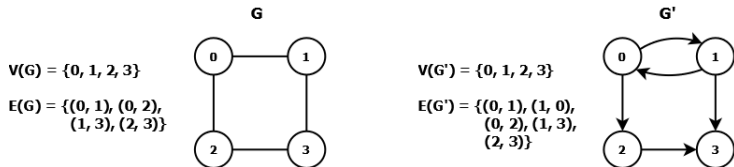


Figure 2: An undirected graph G (left) and a directed graph G' (right). Note how the direction of the edges affects the set of edges of G' .

Genesis of Graph Theory

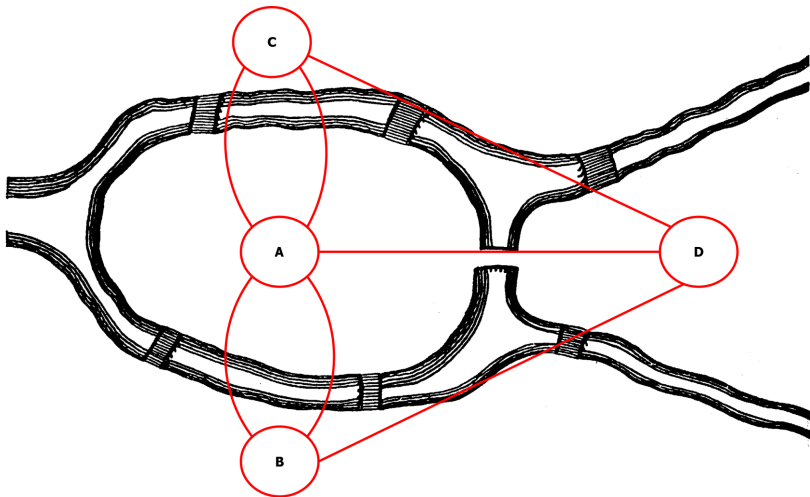
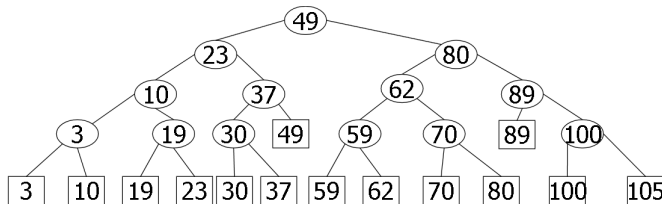


Figure 3: The city of Königsberg represented as a graph.

Spot the Graph



Spot the Graph

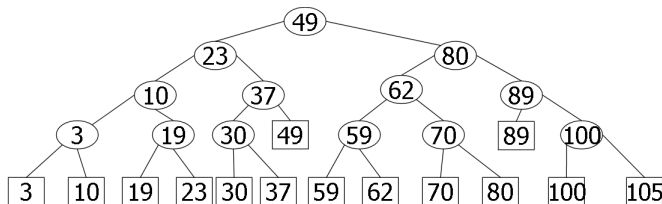
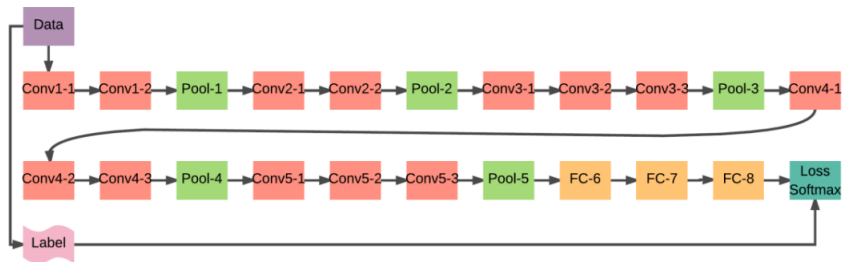


Figure 4: A balanced tree.

Spot the Graph



Spot the Graph

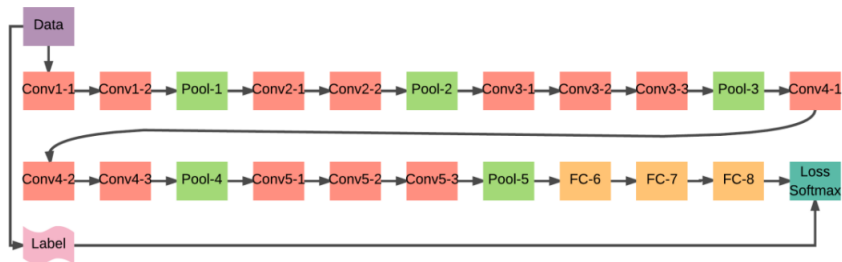
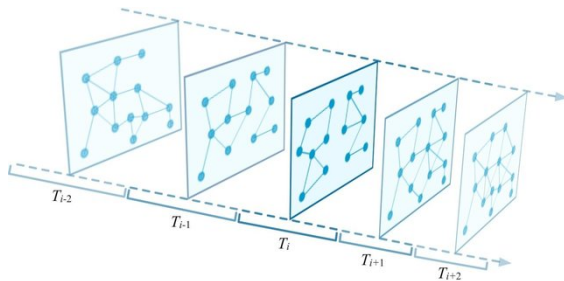


Figure 5: VGG16 Convolutional NN (Source [2])

Spot the Graph



Spot the Graph

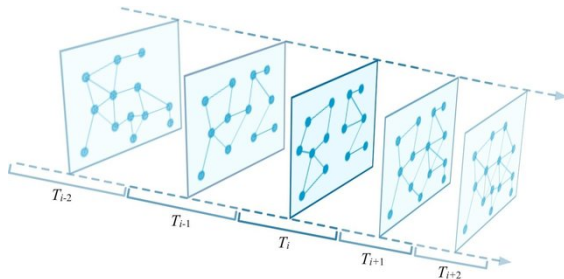
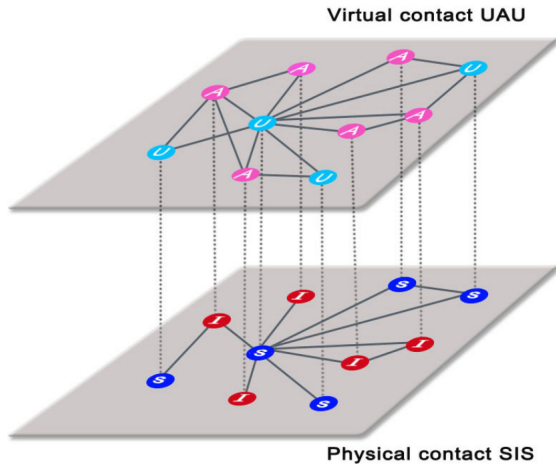


Figure 6: Abstraction of a time-evolving Network (Source [3])

Spot the graph



Spot the graph

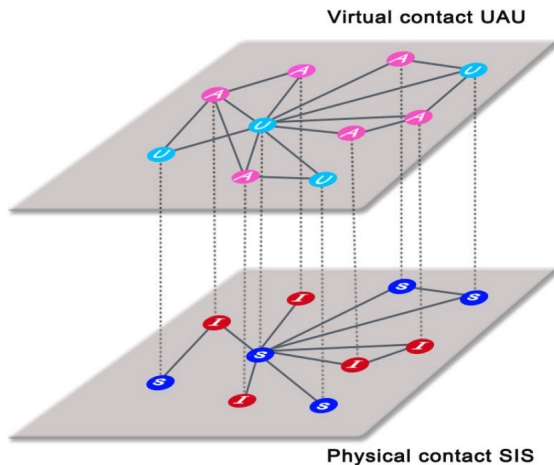


Figure 7: Epidemic multiplex Network (source [4])

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.

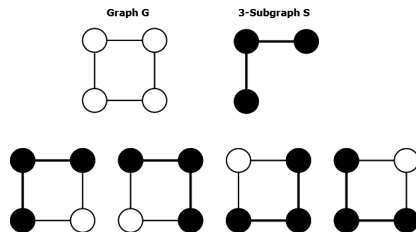


Figure 8: All occurrences of a subgraph of size 3 (right) on graph G (left).

Network Motifs

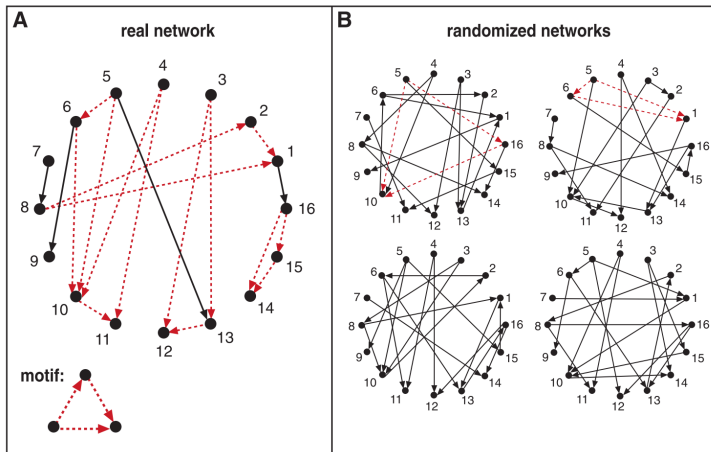


Figure 9: Given a network G a motif M is an induced subgraph that appears more often than expected (source: [5]).

Network Motifs

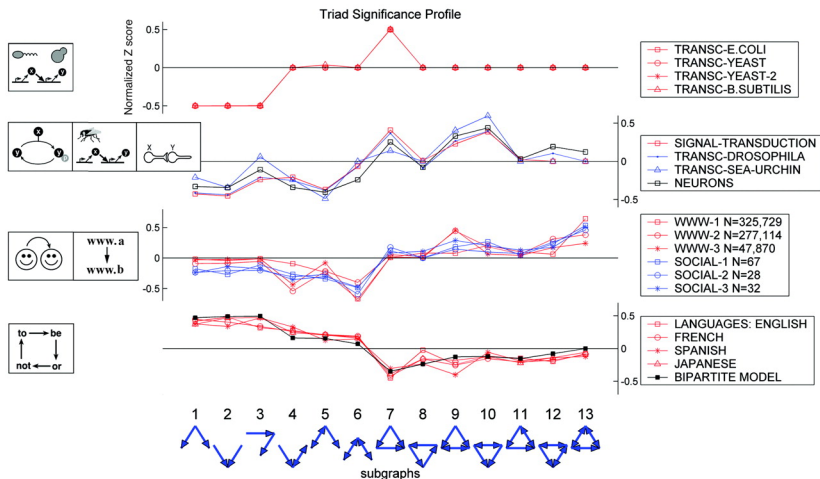


Figure 10: The triad significance profile from from various disciplines (source: [6]).

Graphlet Degree Distributions

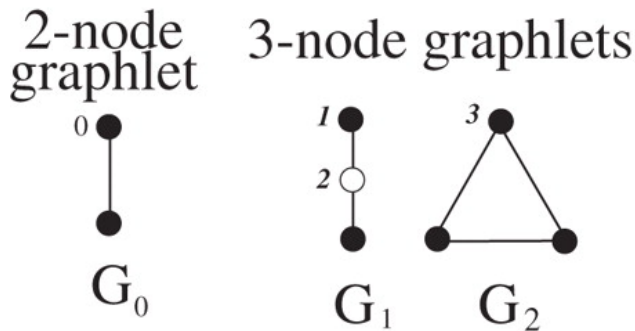


Figure 11: Orbits for all possible graphlets from sizes 2 to 3 (Adapted: [7]).

Graphlet Degree Distributions

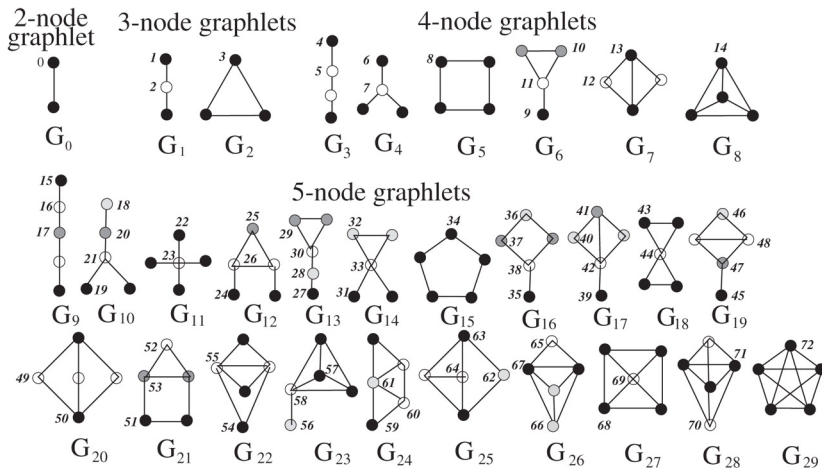


Figure 12: Orbits for all possible graphlets from sizes 2 to 5 (Source: [7]).

Graphlet Degree Distributions

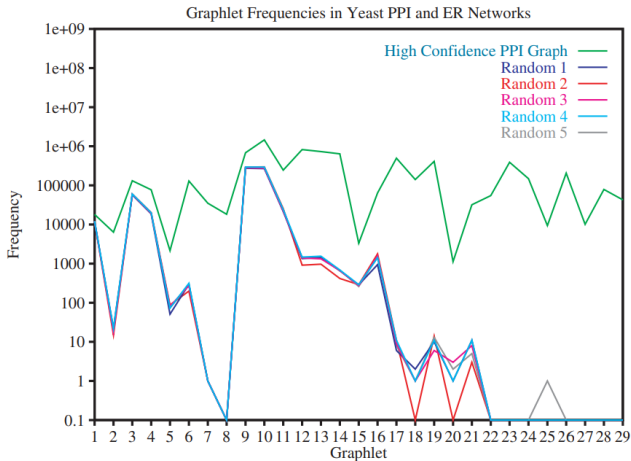


Figure 13: PPI Networks versus Random Networks (Source: [8]).

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

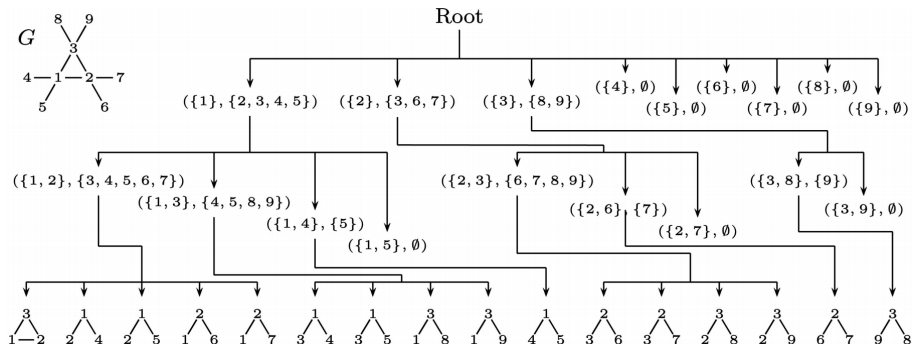


Figure 14: ESU-TREE of a graph G (Source: [9]).

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`

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

References I



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