

# Introduction to data science and network science

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Artificial Intelligence, urbanistics and data science  
ITMO, 2024/2025

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## Contacts

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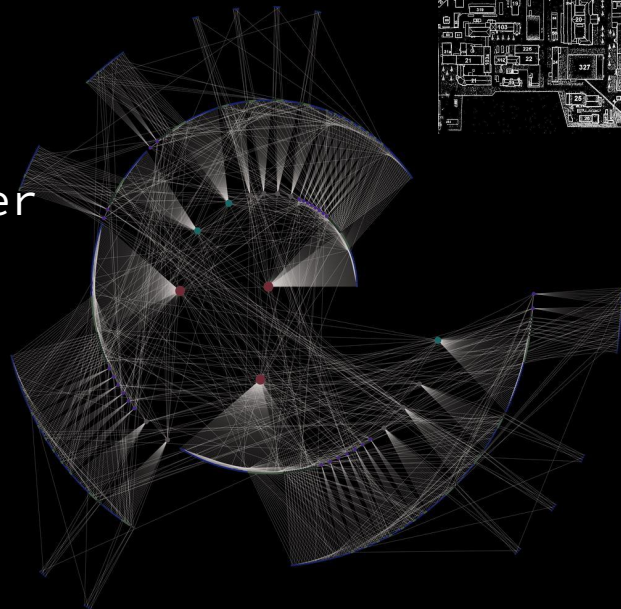
## Main topics:

Explainable AI (geometry of  
embeddings, analysis of higher-order  
data)

Geometry of graphs and hypergraphs  
(processes on graphs, transport  
applications)

Students contact:

Zahara Farook, Hritika Kathuria



LPI and embodied network science

# Outline and connection to other courses

1. Reversed classrooms
2. Project-based learning
3. Teaching and learning by doing (fablab access, art itmo projects, other external collaborations, e.g. lpi cri-paris.org collaboration)
4. Open science (github, X-posts on the results)
5. Open data policy if possible and needed
6. Course co-design (office hours by requirements), collective learning and self-learning (other platforms)
7. Sustainability oriented courses
8. Mathematics and other deep courses (Skoltech, Spbgu)
9. Network seminar subscription ([youtube](#) links)

# Codesing of the course

# Orientation every class



# Outline of Introduction to data science

# Outline of Introduction to data science

What are your main projects?

# Outline of Introduction to network science

1. Introduction to networks
2. Practical part: notebooks

Assignment for the next course



# Resources and libraries for the course

Standard libraries (Python): numpy, matplotlib, scikit learn, seaborn

Network libraries: networkx, osmnx (open street data)

## Network Science

by Albert-László Barabási

Personal Introduction

1. Introduction
2. Graph Theory
3. Random Networks
4. The Scale-Free Property
5. The Barabási-Albert Model

6. Evolving Networks

7. Degree Correlations
  8. Network Robustness
  9. Communities
  10. Spreading Phenomena
- Preface

Start Reading

# Resources and libraries for the course

Standard libraries (Python): numpy, matplotlib, scikit learn, seaborn

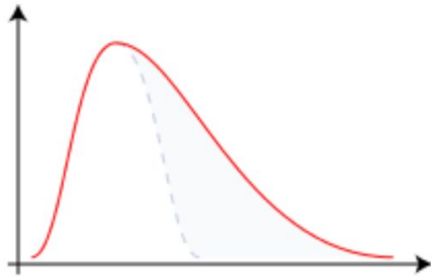
Network libraries: networkx, osmnx (open street data)

Support materials

- Big data course Marc and Liubov <https://github.com/Big-data-course-CRI/>
- Correlaid, Complex system conference CSS 2023 and TidyTuesday  
<https://github.com/rfordatascience/tidytuesday>
- Network science book <http://networksciencebook.com/>
- Network repository <https://networks.skewed.de/>
- Visualisation tools <https://gephi.org/users/download/>
- Network datasets <https://www.complex-networks.net/datasets.html#chap8>

# Statistics course

Two types of distributions we will work with very often



Positive Skew

Right skew

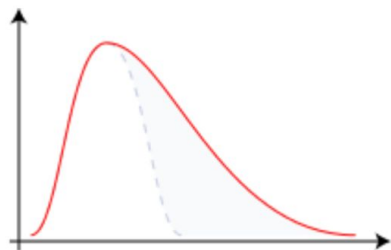
Long tail to the right

# Statistics course

Types of distributions we will work with very often  
(following the recommendations from Michael Szell on his introduction to single variable analysis)

[https://github.com/mszell/introdatasci/tree/main/unit12\\_singlevariableanalysis](https://github.com/mszell/introdatasci/tree/main/unit12_singlevariableanalysis)

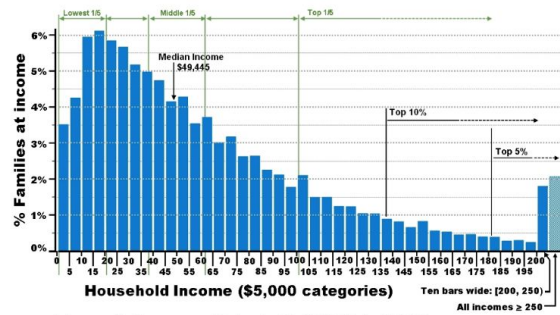
Follow the notebook



Positive Skew

Right skew

Long tail to the right



Data source: [http://www.census.gov/hhes/www/cpstables/032011/hhincnew08\\_000.htm](http://www.census.gov/hhes/www/cpstables/032011/hhincnew08_000.htm)

# Central limit theorem and its intuition

