

Javier Sanz-Cruzado Glasgow IR Away Day December 12, 2022



Development



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Outline

- 1. Introduction to networks
- 2. What is RELISON?
- 3. Installation
- 4. How to use it
- 5. Future plans



Introduction to networks

The basics – Networks



- Two types of objects:
 - Nodes
 - Relations (connections between nodes)
- Mathematically modelled as graphs
- Networks are everywhere!

- Social networks (connecting people)
 - Twitter
 - Facebook
 - LinkedIn
 - Goodreads
 - TikTok
 - ...



Transport / energy networks







• Search engine



- In summary, many things can be modelled as a network / graph
- So you might need some tool to:
 - Create and manipulate networks
 - Analyze the structure of those networks
 - Predict future links
 - Etc.
- RELISON fits that gap!



What is RELISON?

What is **RELISON**?

- **RELISON:** REcommending LInks in SOcial Networks
- Java library



- Focus:
 - Link recommendation
 - Effects over network structure

- Is RELISON only useful for link recommendation? No! It offers much more! We'll see in the next slide
- Is RELISON only useful for social networks? No! You can use it over general networks
- Are you sure it is only available in Java? Yes, as of now (version 1.0.0) it is only available in Java

Functionalities provided by RELISON



RELISON useful links



OFFICIAL WEBSITE





DOCUMENTATION





Installation

Installation



Command line

- RELISON provides command line programs to execute multiple functionalities
- Limitations:
 - Graphs must be read from .csv files
 - Node types can only be integers
- Advantage: no need for code

How to use it?

• Step 1: Download the compiled JAR file

curl -L https://github.com/ir-uam/RELISON/releases/download/v1.0.0maven/relison.jar --output relison.jar

• Step 2: Execute program

java [VM Options] -jar relison.jar [PROGRAM_CODE] [arguments]
We'll explore some of the
 programs during this
 presentation!

Developer mode

- Used for implementing new methods, metrics, etc.
- Recommendation: Maven
 - RELISON is available in Maven Central since last month
 - Add the dependency as:

```
<dependency>
   <groupId>io.github.ir-uam</groupId>
    <artifactId>RELISON-[module-name]<artifactId>
    <version>1.0.0</version>
</dependency>
```

• Where [module-name] is the name of the module to use

Note about this tutorial

- For time constraints, we will focus here on command line programs
- If you want to use developer mode, refer to the documentation of the library.
- Would you be interested on a presentation like this for "developer mode"?



How to use it

Hands-on

• Do you want to try RELISON as we advance?







Network creation

Types of graphs supported

Edge number between nodes



Edge weights 3.5 -2 4 Weighted



Unweighted

Undirected

Manually crafted networks

- Default format: tab separated .csv
- Origin \t Destination \t Weight (opt) \t Type (opt)
- Example: type less weighted directed network





Α	В	1.0
Α	С	2.0
В	D	1.5
С	В	3.0

Automatically created random graphs

- Graph models
 - Define the properties of a network
 - Nodes and edges are created accordingly to a stochastic algorithm.
- In RELISON:
 - Random Erdös networks
 - Preferential attachment
 - Watts-Strogatz
 - Non-stochastic: complete graph, empty graph

Automatically created random graphs (II)

• The *graphgen* program

graphgen output_file directed? num_nodes algorithm_name [parameters]

- Input:
 - **output_file:** file on which to store the graph
 - directed: true if the graph is directed, false otherwise
 - num_nodes: the number of nodes
 - **algorithm_name:** the name of the graph generation algorithm
 - [parameters]: the parameters for the algorithm
- Output:
 - A tab separated .csv file with the graph





Link recommendation & link prediction

Recommendation task



Link / People / Contact recommendation



Link / People / Contact recommendation



- Items = users
- Availability of social relationships
- Rating matrix = adjacency matrix

Importance of link recommendation

- Particular characteristics w.r.t. classic recommendation
 - Development of new methods
 - Use of social network analysis
- Creation of new links
 - Main asset of networks
 - Communication channels
 - Source of information
 - Increase engagement of users

Link / People / Contact recommendation



- Items = users
- Availability of social relationships
- Rating matrix = adjacency matrix

Link prediction



- Which edges will appear in the network in the future?
- Classification problem
- Unique ranking for all possible links

Link recommendation vs. prediction

Link recommendation Link prediction

- Social networks (mostly)
 Any network



Functionalities for link recommendation



Algorithms

Unsupervised		Supervised		
Trivial	RandomPopularity	Classifiers	Random forestLinear regression	
Friends of friends	MCNIR models	Learning to rank	LambdaMART	
Path based	Graph distanceKatz similarity			
Random walk	 Personalized PageR Money (Twitter) 	Rank		
Collaborative filtering	 User / Item-based k Implicit MF 	NN	More than 50 algorithms!	
Content based	 Twittomender Centroid-based CB 			

Evaluation metrics

- Accuracy: IR-based metrics
 - Precision
 - Recall
 - nDCG
 - MAP

• Diversity:

- Intra-list diversity
- ERR-IA
- Predicted Gini complement

• Novelty:

- Long tail novelty
- Unexpectedness
- Mean prediction distance

• Effects on networks: Later, on SNA section

How to use

• From command line: the *recommendation* program

recommendation train test multigraph directed weighted selfloops readtypes config output rec-length [optional parameters]

- **train / test:** the training/test networks
- multigraph/directed/weighted: true/false if network is multigraph/directed/weighted
- **selfloops:** true if we want to read self-loops
- readtypes: true if edges have types and we want to read them
- **config:** YAML configuration file
- **output:** output directory in which to store the files
- **rec-length:** cutoff of the recommendation

Program configuration (YAML)

```
algorithms:
   algorithm name:
       param_name:
           type: int/double/boolean/string/long/orientation
           values: [value1,...,valueN] or value
           range:
           - start: start val
             end: end val
             step: step_val
metrics:
   metric_name:
       param_name:
           ...
```

Program output

- The recommendation program provides two classes of outputs:
 - **Recommendations:** the generated recommendations. Csv files with the format:

Target_user \t Candidate_user \t Score

- **Metrics:** the metrics for the recommendations. A .csv file with the format

Variant \t Fraction \t Metric1 \t Metric2 \t ... \t MetricN

Example configuration (YAML)

algorithms: iMF: **k**: type: int range: - start: 10 end: 300 step: 10 alpha: type: double values: [10.0, 40.0] lambda: type: double values: 150.0

metrics: nDCG: cutoff: type: int values: 10 Predicted Gini complement: cutoff: type: int values: [1,5,10]





Network structural analysis

Studying the structure of networks

• Different networks present different structures



- Understanding our networks might provide insights on our algorithms
- Many properties can be measured

Types of properties

Node



- Degree
- PageRank
- Closeness

Edge/pair



- Distance
- Neighbor overlap
- Betweenness





- Clust. coefficient
- Density
- Diameter

Types of properties (community based)

 For now, communities are just partitions of nodes (we'll provide a better definition later)

Individual community



- Degree
- Size
- Volume

Graph (community-based)



- Modularity
- Edge Gini complement

How to use

- From command line: the sna program
- Computes structural metrics of a network

sna network multigraph directed weighted selfloops config output
(-communities comm1,...,commN --distances)

- **network:** the network to analyze
- multigraph/directed/weighted: true/false if network is multigraph/directed/weighted
- **selfloops:** true if we want to read self-loops
- config: YAML configuration file
- output: output directory in which to store the files
- **communities:** community files (see later)
- --distances: a flag to pre-compute distances between nodes

Program configuration (YAML)

metrics:

```
metric name:
   type: vertex/edge/pair/graph/indiv. community/global community
   params:
       param_name:
          type: int/double/boolean/string/long/orientation
          values: [value1,...,valueN] or value
          range:
           - start: start val
            end: end val
```

step: step_val

YAML example

```
metrics:
   Clustering coefficient:
      type: graph
      params:
         uSel:
             type: orientation
             values: IN
         vSel:
             type: orientation
             values: OUT
   Eccentricity:
      type: vertex
```

```
Embeddedness:
   type: edge
   params:
      uSel:
         type: orientation
          values: IN
      vSel:
          type: orientation
          values: OUT
```

Program output

- The recommendation program provides many outputs:
- global.txt: a .csv file containing
 - Global metrics
 - Averaged vertex/node/indiv. community metrics
 - Format:

```
Metric \t Value
```

- Directories for every other metric, containing a file per metric
 - For vertex metrics: Node \t Value
 - For pair metrics: Origin \t Destination \t Value
 - For indiv. community metrics: Community \t Value

Effects of recommendations

- There is also a program for measuring the effect of recommendations
- The program is named *effects*
- Configuration file is the same
- It only computes global properties
- More information in the documentation of the library





Community detection

Communities

- Homophily: similar nodes tend to relate to each other
- It is a characteristic of real world networks
 - Groups of tightly connected nodes
 - Barely connected to each other
- We name those clusters of nodes **communities**



Community detection algorithms

Connectedness-based approaches

- How many nodes can we reach from a particular node?
- In undirected networks: connected components
- In directed networks: strongly / weakly connected components

Modularity-based approaches

- Modularity measures how good a clustering is
- Compares:
 - Links inside communities
 - Links between communities we would have in a random graph keeping the degree distribution
- Methods: Louvain, Infomap

How to use

• From command line: the *communities* program

communities graph multigraph directed weighted self_loops config
output_dir

- graph: the network on which we want to detect communities
- multigraph/directed/weighted: true/false if network is multigraph/directed/weighted
- **selfloops:** true if we want to read self-loops
- **config:** YAML configuration file
- **output:** output directory in which to store the files

Program configuration (YAML)

```
algorithms:
   algorithm name:
      param_name:
         type: int/double/boolean/string/long/orientation
         value: value
Example:
algorithms:
   Louvain:
      threshold:
         type: double
         value: 0.0001
```

Program output

- The recommendation program provides an output for each selected algorithm
- It includes the community partition
- CSV file (tab separated)
- Format

Node \t Community





Information diffusion

Information diffusion

- Edges in networks (and in special social networks) are channels for communicating information
- Example: We read tweets thanks to our followed users
- The information diffusion process is complex
- Mechanisms of it are not exactly known
- Some models have been proposed

Information diffusion in RELISON

- Based on simulation
- Simulate the simultaneous diffusion of different pieces of information
- Highly configurable simulation
- Multiple components



Information spread



Information reception





Future directions

Future directions

- Knowledge graph support
- Node / graph embeddings
- Python wrapper (pyRELISON)
- Network visualization
- Suggestions?



Do you want to help?

- Help to improve RELISON wanted!
- Contact me!
 - Office: F111
 - E-mail: javier.sanz-cruzadopuig@glasgow.ac.uk
 - Skype
 - Teams
 - Twitter....
- Together we can make RELISON better!



Do you want to know more?



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