Towards a format for describing networks
Format elements

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Abstract. The key components that a common format for describing networks should include are discussed.

Keywords: First keyword · Second keyword · Third keyword

1 Introduction

2 Nets and NetsJSON

For dealing with networks with properties with structured values (for example, temporal quantities) we are developing a Python package Nets [**?**].

For describing temporal networks we initially, extending Pajek format, defined and used the Ianus format.

In 2015 we started to develop a new format based on JSON – we named it netJSON. On February 26, 2019 the format was renamed to NetsJSON because of the collision with http://netjson.org/rfc.html.

NetsJSON has two versions: a *basic* and a *general* version. Current implementation of the Nets / TQ library supports only the basic version. Nets.

Besides for a *description* of networks with structured values, NetsJSON should *envelope* (most of) existing network description formats [**?**] (archiving, conversion) and provide input data for D3.js *visualization*s.

2.1 Informal description of the basic netsJSON format

```
{
"netsJSON": "basic",
```

```
"info": {
   "org":1, "nNodes":n, "nArcs":mA, "nEdges":mE,
   "simple":TF, "directed":TF, "multirel":TF, "mode":m,
   "network":fName, "title":title,
   "time": { "Tmin":tm, "Tmax":tM, "Tlabs": {labs} },
   "meta": [events], ...
   },
"nodes": [
   { "id":nodeId, "lab":label, "x":x, "y":y, ... },
   ***
   ]
"links": [
   { "type":arc/edge, "n1":nodeID1, "n2":nodeID2, "rel":r, ... }
   ***
   ]
}
```

where ... are user-defined properties and *** is a sequence of such elements.

Basic netsJSON format An event description can contain fields:

```
{  "date": date,
   "title": short description,
   "author": name,
   "desc": long description,
   "url": URL,
   "cite": reference,
   "copy": copyright
}
```

for describing temporal networks a node element and a link element has an additional required property `tq`

Example 1, Franzosi's violence network / UTF-8 no sig

2.2 Description of networks using a spreadsheet

How to describe a network $\mathcal{N}$? In principle the answer is simple – we list its components $\mathcal{V}$, $\mathcal{L}$, $\mathcal{P}$, and $\mathcal{W}$.

The simplest way is to describe a network $\mathcal{N}$ by providing $(\mathcal{V}, \mathcal{P})$ and $(\mathcal{L}, \mathcal{W})$ in a form of two tables.

As an example, let us describe a part of network determined by the following works:

Generalized blockmodeling, Clustering with relational constraint, Partitioning signed social networks, The Strength of Weak Ties

There are nodes of different types (modes): persons, papers, books, series, journals, publishers; and different relations among them: author_of, editor_of, contained_in, cites, published_by.

Both tables are often maintained in Excel. They can be exported as text in CSV (Comma Separated Values) format.

bibNodes.csv

```
name;mode;country;sex;year;vol;num;fPage;lPage;x;y
"Batagelj, Vladimir";person;SI;m;;;;;;809.1;653.7
"Doreian, Patrick";person;US;m;;;;;;358.5;679.1
"Ferligoj, Anuška";person;SI;f;;;;;;619.5;680.7
"Granovetter, Mark";person;US;m;;;;;;145.6;660.5
"Moustaki, Irini";person;UK;f;;;;;;783.0;228.0
"Mrvar, Andrej";person;SI;m;;;;;;478.0;630.1
"Clustering with relational constraint";paper;;;1982;47;;413;426;684.1;380.1
"The Strength of Weak Ties";paper;;;1973;78;6;1360;1380;111.3;329.4
"Partitioning signed social networks";paper;;;2009;31;1;1;11;408.0;337.8
"Generalized Blockmodeling";book;;;2005;24;;1;385;533.0;445.9
"Psychometrika";journal;;;;;;;;741.8;086.1
"Social Networks";journal;;;;;;;;321.4;236.5
"The American Journal of Sociology";journal;;;;;;;;111.3;168.9
"Structural Analysis in the Social Sciences";series;;;;;;;;310.4;082.8
"Cambridge University Press";publisher;UK;;;;;;534.3;238.2
"Springer";publisher;US;;;;;;;884.6;174.0
```

bibNodes.csv

In large networks, to avoid the empty cells, we split a network to some subnetworks – a collection.

bibLinks.csv

```
from;relation;to
"Batagelj, Vladimir";authorOf;"Generalized Blockmodeling"
"Doreian, Patrick";authorOf;"Generalized Blockmodeling"
"Ferligoj, Anuška";authorOf;"Generalized Blockmodeling"
"Batagelj, Vladimir";authorOf;"Clustering with relational constraint"
"Ferligoj, Anuška";authorOf;"Clustering with relational constraint"
"Granovetter, Mark";authorOf;"The Strength of Weak Ties"
"Granovetter, Mark";editorOf;"Structural Analysis in the Social Sciences"
"Doreian, Patrick";authorOf;"Partitioning signed social networks"
"Mrvar, Andrej";authorOf;"Partitioning signed social networks"
"Moustaki, Irini";editorOf;"Psychometrika"
"Doreian, Patrick";editorOf;"Social Networks"
"Generalized Blockmodeling";containedIn;"Structural Analysis in the Social Sciences"
"Clustering with relational constraint";containedIn;"Psychometrika"
"The Strength of Weak Ties";containedIn;"The American Journal of Sociology"
"Partitioning signed social networks";containedIn;"Social Networks"
"Partitioning signed social networks";cites;"Generalized Blockmodeling"
"Generalized Blockmodeling";cites;"Clustering with relational constraint"
"Structural Analysis in the Social Sciences";publishedBy;"Cambridge University Press"
"Psychometrika";publishedBy;"Springer"
```

bibLinks.csv

2.3 Factorization and description of large networks

To save space and improve computing efficiency we often replace values of categorical variables with integers. In R this encoding is called a *factorization*.

We enumerate all possible values of a given categorical variable (coding table) and afterward replace each value with the corresponding index in the coding table.

This approach is used in most programs dealing with large networks. Unfortunately, the coding table is often a kind of meta-data.

CSV2Pajek.R

```
# transforming CSV file to Pajek files
# by Vladimir Batagelj, June 2016
# setwd("C:/Users/batagelj/work/Python/graph/SVG/EUSN")
# colC <- c(rep("character",4),rep("numeric",7)); nas=c("","NA","NaN")
colC <- c(rep("character",4),rep("numeric",5)); nas=c("","NA","NaN")
nodes <- read.csv2("bibNodes.csv",encoding='UTF-8',colClasses=colC,na.strings=nas)
n <- nrow(nodes); M <- factor(nodes$mode); S <- factor(nodes$sex)
mod <- levels(M); sx <- levels(S); S <- as.numeric(S); S[is.na(S)] <- 0
links <- read.csv2("bibLinks.csv",encoding='UTF-8',colClasses="character")
```

```
F <- factor(links$from,levels=nodes$name,ordered=TRUE)
T <- factor(links$to,levels=nodes$name,ordered=TRUE)
R <- factor(links$relation); rel <- levels(R)
net <- file("bib.net","w"); cat('*vertices ',n,'\n',file=net)
clu <- file("bibMode.clu","w"); sex <- file("bibSex.clu","w")
cat('%',file=clu); cat('%',file=sex)
for(i in 1:length(mod)) cat(' ',i,mod[i],file=clu)
cat('\n*vertices ',n,'\n',file=clu)
for(i in 1:length(sx)) cat(' ',i,sx[i],file=sex)
cat('\n*vertices ',n,'\n',file=sex)
for(v in 1:n) {
  cat(v,' "',nodes$name[v],'"\n',sep='',file=net);
  cat(M[v],'\n',file=clu); cat(S[v],'\n',file=sex)
}
for(r in 1:length(rel)) cat('*arcs :',r,' "',rel[r],'"\n',sep='',file=net)
cat('*arcs\n',file=net)
for(a in 1:nrow(links))
  cat(R[a],': ',F[a],' ',T[a],' 1 1 "',rel[R[a]],'"\n',sep='',file=net)
close(net); close(clu); close(sex)
```

CSV2Pajek.R

bib.net

```
*vertices  16
1 "Batagelj, Vladimir"
2 "Doreian, Patrick"
3 "Ferligoj, Anuška"
4 "Granovetter, Mark"
5 "Moustaki, Irini"
6 "Mrvar, Andrej"
7 "Clustering with relational constraint"
8 "The Strength of Weak Ties"
9 "Partitioning signed social networks"
10 "Generalized Blockmodeling"
11 "Psychometrika"
12 "Social Networks"
13 "The American Journal of Sociology"
14 "Structural Analysis in the Social Sciences"
15 "Cambridge University Press"
16 "Springer"
*arcs :1 "authorOf"
*arcs :2 "cites"
*arcs :3 "containedIn"
*arcs :4 "editorOf"
*arcs :5 "publishedBy"
```

```
*arcs
1: 1 10 1 1 "authorOf"
1: 2 10 1 1 "authorOf"
1: 3 10 1 1 "authorOf"
1: 1 7 1 1 "authorOf"
1: 3 7 1 1 "authorOf"
1: 4 8 1 1 "authorOf"
4: 4 14 1 1 "editorOf"
1: 2 9 1 1 "authorOf"
1: 6 9 1 1 "authorOf"
4: 5 11 1 1 "editorOf"
4: 2 12 1 1 "editorOf"
3: 10 14 1 1 "containedIn"
3: 7 11 1 1 "containedIn"
3: 8 13 1 1 "containedIn"
3: 9 12 1 1 "containedIn"
2: 9 10 1 1 "cites"
2: 10 7 1 1 "cites"
5: 14 15 1 1 "publishedBy"
5: 11 16 1 1 "publishedBy"
```

bib.net, bibMode.clu, bibSex.clu; bib.paj, bib.ini.

Reading Pajek files in R

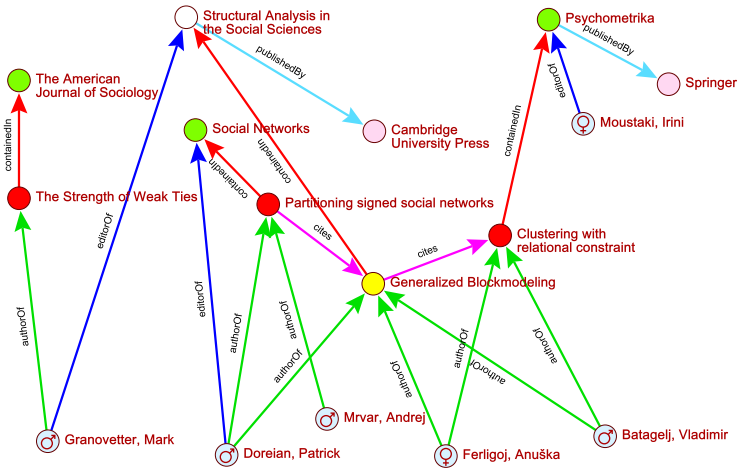

Fig. 1. Bibliographic network – picture / Pajek

3 Network formats

3.1 Format

Gephi – Supported Graph Formats
GEXF 1.3 primer
Cons matrix – large networks
Cons neighbors – node properties
Tomaž Pisanski et al. – Vega
Primož Potočnik et al. – catalogues
N = (V,L,P,W) tables (V,P) (L,W)
Mixed links directed/undirected
Multiple links
Additional data about values, algebraic structures
Metadata (Dublin Core, FAIR, Schema)
Python page on network DS Python Patterns - Implementing Graphs
Factorization
Values – semiring weights (Semirings)
Structured values: TQs, records, distributions
Values – functions: PERT, circuits, modeling
IDs - in network science often missing; individuals/classes -is a-
Alternative labels for display (short, other languages)
Collections of networks
Operations and transformations on networks: extensions (new yearly data), constrictive description (NetML), intersection, union, ... product; derived networks
Contexts !!! – matching IDs in different contexts
Generalizations: multiway, hypernets, bikes
Size: huge networks
Metadata, updates, comments
Default values
KG ¡-¿ networks (single relation, type of units)

4 Conclusions

1.
2.
3.

4.1 Acknowledgments

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