network *mining*

introduction to network analysis in Python (NetPy)

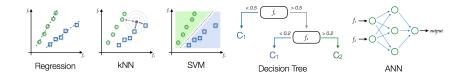
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mining overview

how to *mine* different *graphs/networks*?

how to mine network nodes/links?

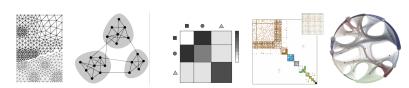
 $node/link\ clustering
ightarrow revealing\ similarity\ clusters$ $node/link\ classification
ightarrow predicting\ discrete\ labels$ $node/link\ regression
ightarrow predicting\ numerical\ values$



mining clustering

how to *cluster* network *nodes/links*?

- graph partitioning and community detection methods for connected assortative clusters based on homophily
- (stochastic) blockmodeling and role discovery methods also disconnected disassortative clusters based on equivalence



for survey/user guide see [For10, FH16]

mining classification/regression

how to *classify/regress* network *nodes/links*?

- relational learning and link mining methods machine learning methods using network structure
- node/link structure used as learning features centrality, bridging, fragments, egonets, clusters etc. node features from local random walk exploration \rightarrow

DeepWalk [PARS14] node2vec [GL16] struc2vec [FRS17] only homophily

also equivalence

only structure

for *survey* see [BCM11, ZPS⁺16]

mining classification

— classification by clustering in APS citation network [Šub15b]

		7 journals		91 sections	
class	method	NMI	CA	NMI	CA
clusters	spectrum	0.361	59.8%	0.380	38.6%
	modularity	0.339	68.1%	0.426	37.1%
	map equation	0.232	71.3%	0.416	48.1%
	block model	0.243	69.6%	0.392	45.3%
baselines	neighbors	-	63.9%	-	46.5%
	2-neighbors	-	71.5%	-	50.4%
	network	-	27.6%	-	17.9%

— ... by clustering in WikiLeaks reference network [Šub15a]

		3 privacies		246 embassies	
class	method	NMI	CA	NMI	CA
clusters	spectrum	0.003	49.1%	0.658	47.9%
	modularity	0.048	59.2%	0.699	52.3%
	map equation	0.088	33.1%	0.654	37.1%
	block model	0.035	56.5%	0.625	37.6%
baselines	neighbors	-	14.2%	-	15.0%
	2-neighbors	-	27.7%	-	31.6%
	network	-	49.1%	-	1.4%

mining regression

- regression by centrality in Slovenian highways network

		traffic loads	
class	centrality	Pearson	Spearman
spectral	degree	0.278	0.275
	eigenvector	0.010	0.241
	PageRank	0.121	-0.149
distance	closeness	0.623	0.634
uistance	betweenness	0.628	0.651
clustering	standard	0.000	0.000
	μ -corrected	0.000	0.230

— ... by centrality in Wikipedia web of trust network [MAC11]

		users trust		
class	centrality	Pearson	Spearman	
	degree	0.013	0.316	
spectral	eigenvector	0.026	0.118	
	PageRank	0.001	0.182	
clustering	standard	0.134	0.115	
Clustering	μ -corrected	0.127	0.199	

mining references



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