# A Machine Learning Perspective

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## Introduction

# The world's platform for change

212,279,858 people taking action. Victories every day.

Start a petition

### Project aim:

Analyze emerge and spread phenomena of dynamic sustainable pro-social behavior in social network.

#### Data sets:

Collected from Change.org, one of the most active petition platform.

#### Methods:

Node2vec, word2vec & combination of both.

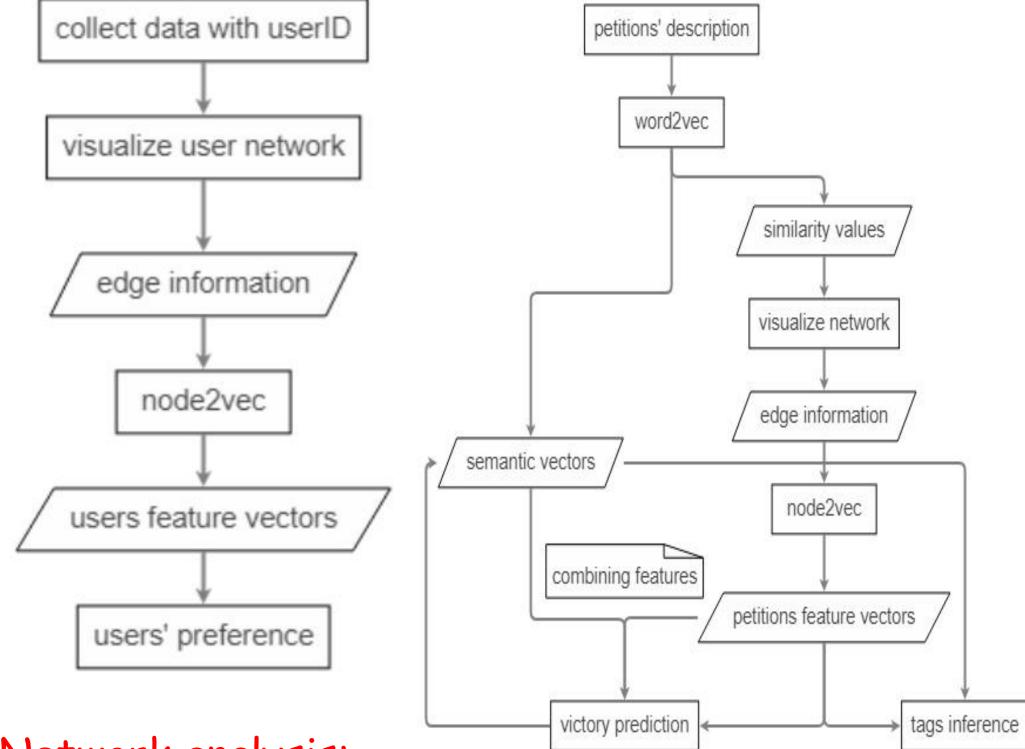
# Approach

### Users research:

node2vec & linear regression

### Petition research:

word2vec & multi-lable regression



#### Network analysis:

Community detection Degree distribution

#### Set up:

Most frequent 20 tags

#### Evaluation metric:

Precision and recall in binary task Macro and Micro in multilabel task

# Problem description

Problem 1: How participants influence each other.

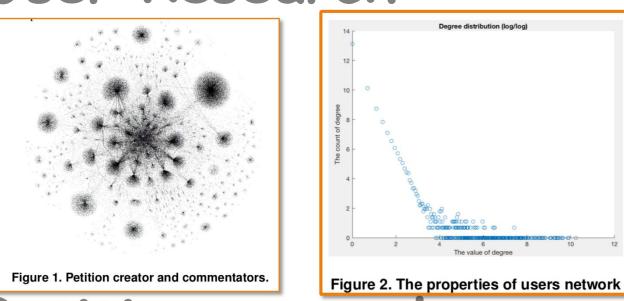
Problem 2: How large online social network brings single behavior others and influence them.

Problem 3: How much influential users will determine the number of signatures a petition could get since all followers of an influential user will also sign a petition that this influential user signs.

Problem 4: Is there any possibility that certain content may lead to the victory of the petition.

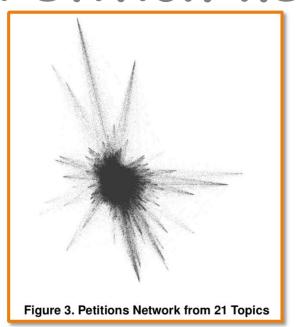
## Results

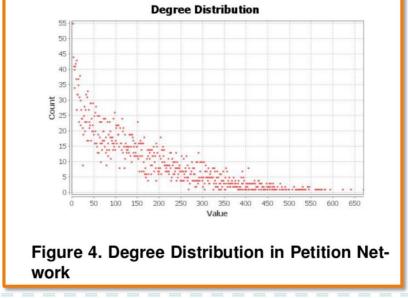
### User Research:

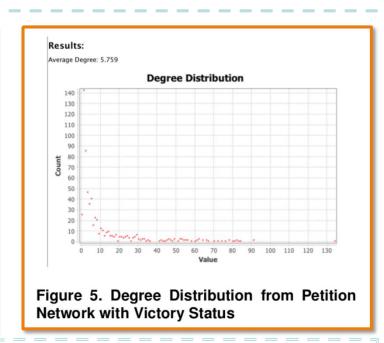


For user network, we have 650000 nodes and 3400000 edges, and degree distribution follows power law, so user network is a scale free network.

Petition network:







For petition network, we have 4545 nodes and 341352 edges, and degree distribution is power law, so petitions also forms into a scale free network.

## Petition multi-lable predict:

	word2vec	node2vec	word2vec & node2vec	feature all 0
macro	0.479	0.480	0.481	0.481
micro	0.923	0.924	0.922	0.481

Macro and micro in combination reduce.

## Petition victory predict:

word		Prediction		node ,	de	Prediction		com	nbi	Р	n	all		Prediction					
		Failed	Victory	Total	Houc		Failed	Victory	Total	ne	Failed	Victory	Total	zer	os	Failed	Victory	Total	
Actual	Failed	231	121	352	nal	Failed	265	87	352	Actual	Failed	258	94	352		Failed	352	0	352
	Victory	154	89	243	Acti	Victory	167	76	243		Victory	142	101	243		Victory	243	0	243

Precision: 0.42

Precision: 0.47

Precision: 0.52

Precision: 0

We combine features for binary and multilabel, and find that simply combining was not good for multilabel classification while could improve the performance for victory prediction.