

## **Social network**

### **Assignment 2**

#### 1. Introduction

Get a real graph from SNAP, Set up a bot detection system, Attack the system and see if it crashes, Compare the before and after results

#### 2. Download data

Use [facebook\\_combined.txt.gz](#) .

#### 3. Facebook's SNAP data doesn't contain any bots, We'll designate a few nodes to be bots to train the detector and attack it. Because the data doesn't contain bots, we'll create bots, and everything else will be humans.

#### 4. Extracting Features for Each Node

Any detector must have data, we have a graph, so we'll extract the following properties:

- degree: The number of edges per node sometimes the most important feature for bots.
- clustering: The local clustering coefficient reflects how connected the node's neighbors are.
- triangles: The number of triangles in the node (how many triangles pass through it).
- pagerank: A popular measure of centrality; multiple attempts may fail if the network is very large or disconnected, hence the try/except method.
- betweenness: Inter-centrality computationally expensive, so if the network is large (>3000), we use an approximation with k samples (measuring the shortest paths).
- avg\_neighbor\_deg: The average neighbor degree gives a picture of the node's perimeter.
- core\_number: Coreness (k-core) an indicator of the node's connection to the network's core.

These are the features the model will train on.

#### 5. Baseline· Evasion and Poisoning Models

##### 1. Building a baseline bot detector:

Take the features and labels and train the classifier: Random Forest.

The baseline model is the model before the attack.

record the performance: Accuracy, Precision, Recall, F1-score, support.

## 2. Structural Evasion Attack

The bot itself "masquerades", modify the connections around the bot node, add 3 human friends to it, expanded its clusterin, makes its degree more human-like, The bot tries to appear human.

After modifying the edges around the bots: Recalculate the features, Retest the model.

## 3. Graph Poisoning Attack

This is the most dangerous attack: We don't change the bot's appearance itself, We corrupt the entire graph before training.

For example: We add 7% random edges that mimic human behavior to the bots, we change the community structure, Or we add fake nodes.

Then: We retrain the model, We test it on a clean graph.

You'll find that the accuracy is severely compromised because the model was poisoned during training.

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Baseline classifier report:					
		precision	recall	f1-score	support
	0	1.00	1.00	1.00	1127
	1	1.00	0.98	0.99	85
	accuracy			1.00	1212
	macro avg	1.00	0.99	0.99	1212
	weighted avg	1.00	1.00	1.00	1212
Baseline ROC AUC: 0.9999791220836161					
...					
Poisoned-training evaluation report (test on clean):					
		precision	recall	f1-score	support
	0	0.94	1.00	0.97	1127
	1	0.95	0.22	0.36	85
	accuracy			0.94	1212
	macro avg	0.95	0.61	0.67	1212
	weighted avg	0.95	0.94	0.93	1212
Poisoned ROC AUC (test on clean): 0.8777597995720027					
...					
Evasion evaluation report:					
		precision	recall	f1-score	support
	0	0.98	1.00	0.99	1127
	1	0.98	0.68	0.81	85
	accuracy			0.98	1212
	macro avg	0.98	0.84	0.90	1212
	weighted avg	0.98	0.98	0.97	1212
Evasion ROC AUC: 0.9974894305548305					

## 6. Graph Visualization

