




# Learning with Graphs: Networks Embedding Algorithms


Luis Valenzuela  
Dennis Mejicanos  
Julio Guzmán  
Supervisor: Vincent Gauthier



INSTITUT  
POLYTECHNIQUE  
DE PARIS



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## Motivation

In order to apply machine learning models for graphs, we need to transform the vertices of a graph into a vector form. **By doing that we can take advantage of all ML algorithms (such as deep learning approaches) already developed to perform learning on graphs.**

We are going to compare three graph embedding techniques: DeepWalk, Node2Vec and Graph Convolutional Networks.



## Problem Statement

Given three graph datasets: Facebook Large Page-Page Network, Github Social Network, and Cora, we will test and compare different graph embedding techniques: DeepWalk, Node2Vec on the Facebook y GitHub datasets, and Node2Vec and and Graph Convolutional Networks (GCN) on the CORA dataset.

We perform semi-supervised learning in order to guess features of nodes in a graph.



# 1. Deep Walk



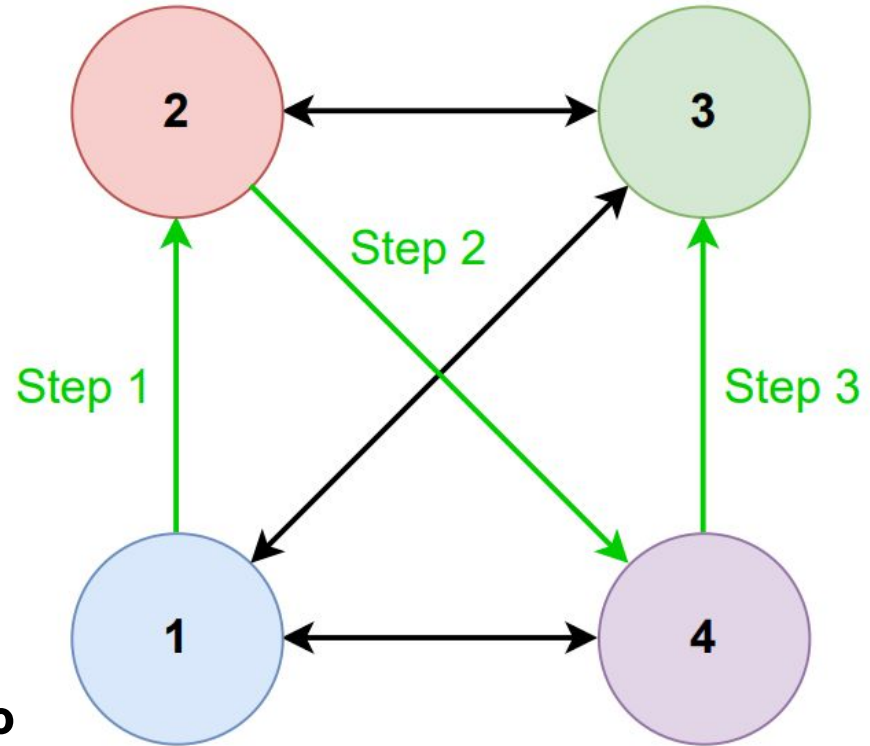
## DeepWalk

- ◎ Run through the graph randomly
- ◎ Walks per node
- ◎ Walk Length

Walk Set: ["1", "2", "3", "4"]



**Deep Learning classification model to predict the context: Skip-gram Model**



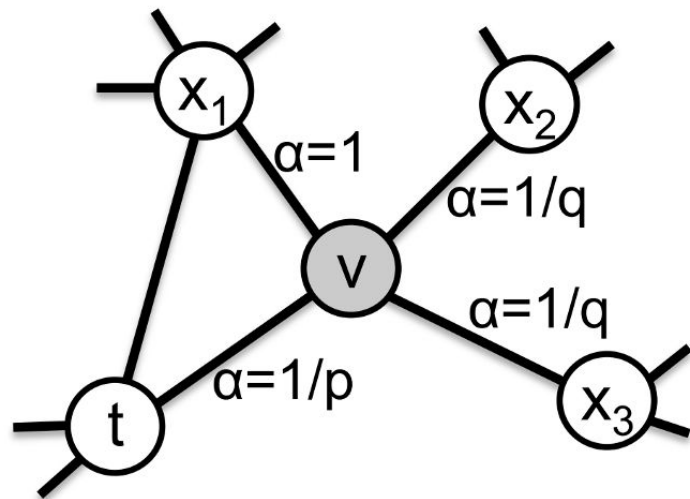


## 2. **Node2Vec**



## Node2Vec

- ◎ Biased Random Walker
  - P: Return Hyperparameter
  - Q: Inout Hyperparameter



Grover, A. (2016). Node2Vec: Scalable Feature Learning for Networks. [Figure]. Recovered from: <https://arxiv.org/pdf/1607.00653.pdf>



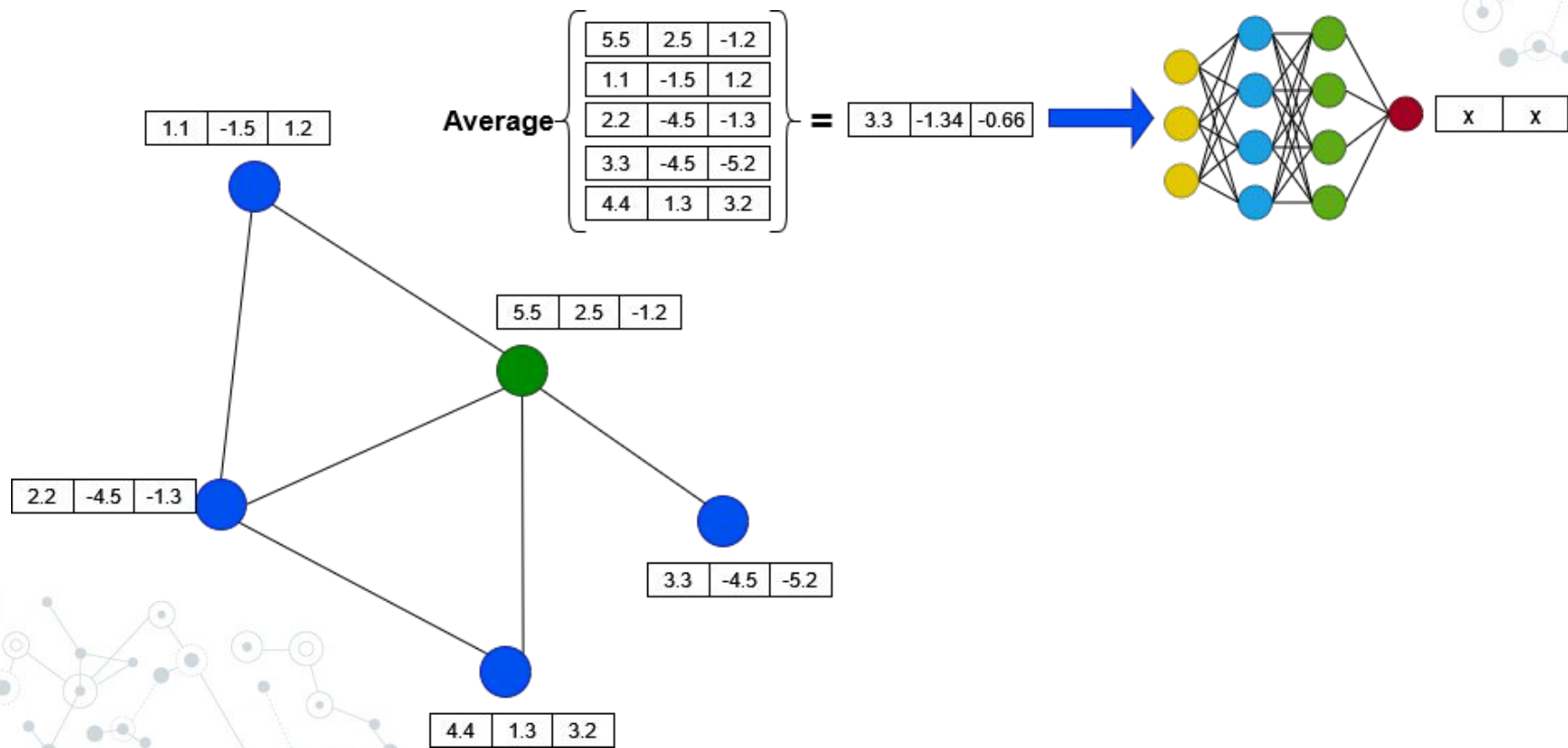


3.

# Graph Convolutional Networks (GCN)



## GCN



A decorative network diagram in the top-left corner, featuring a complex web of interconnected nodes and lines. Some nodes are highlighted with blue circles, and others with blue dots.

# 4. **Experimental Results**

A decorative network diagram in the bottom-right corner, featuring a complex web of interconnected nodes and lines. Some nodes are highlighted with blue circles, and others with blue dots.

A decorative network graph pattern in the top-left corner, featuring a complex web of interconnected nodes and edges. Some nodes are highlighted with blue circles, and others with solid blue dots. The pattern is composed of light gray lines and dots.

4.1

# Github Dataset

A decorative network graph pattern in the bottom-right corner, featuring a complex web of interconnected nodes and edges. Some nodes are highlighted with blue circles, and others with solid blue dots. The pattern is composed of light gray lines and dots.

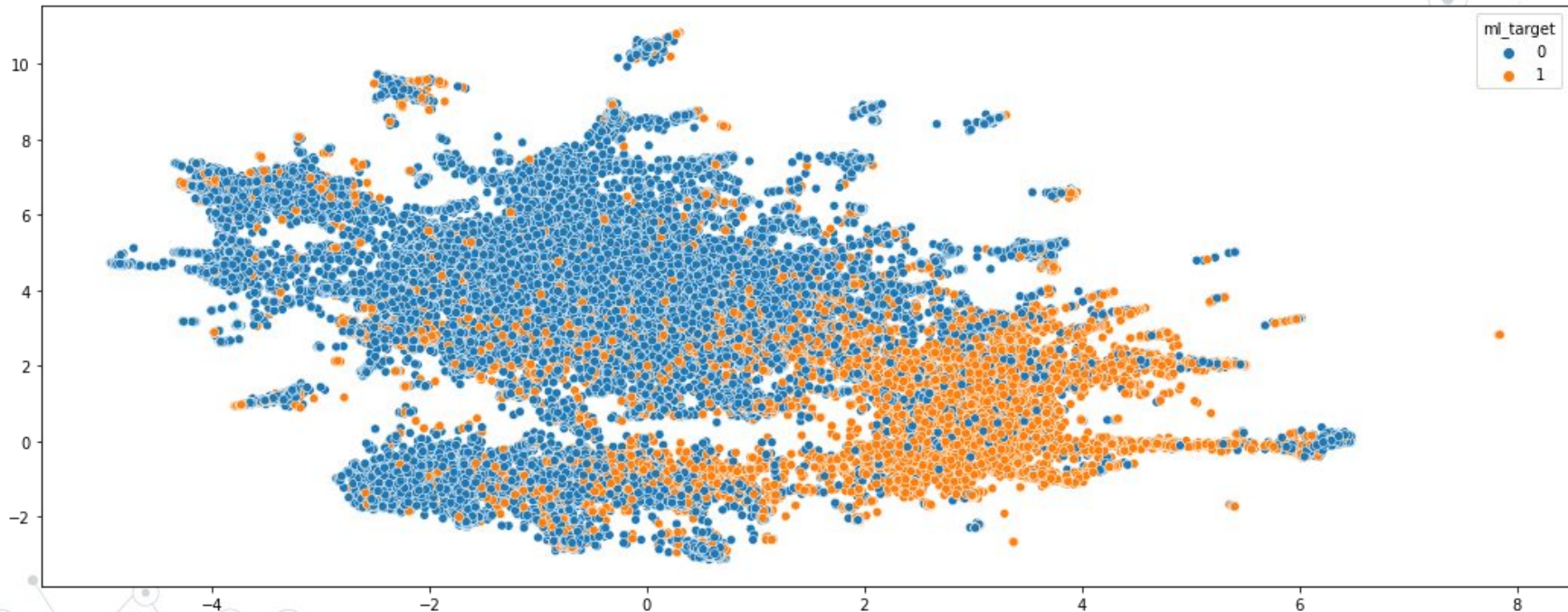
## Github Dataset

- Nodes are developers who have at least 10 repositories on github.
- Edges are mutual follower relationships between the developers.
- Features are extracted based on the location, employer and e-mail address.
- Nodes: 37,700: 27,961 Web Dev, 9,739 ML Dev.
- Edges: 289,003.

Predict if a GitHub user is web or machine learning developer.

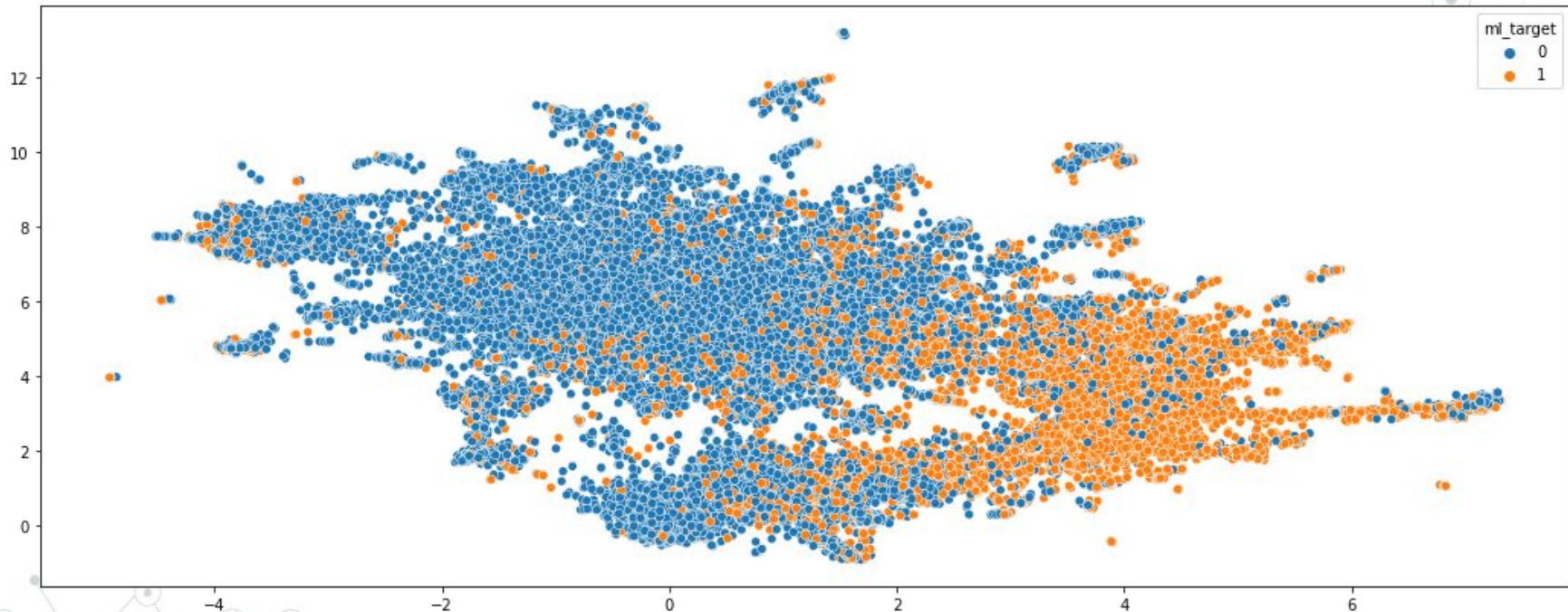
Source: <https://github.com/benedekrozemberczki/datasets>

## Github Dataset - DeepWalk Embedding



Walks per node = 10, walk length = 80

## Github Dataset - Node2Vec Embedding



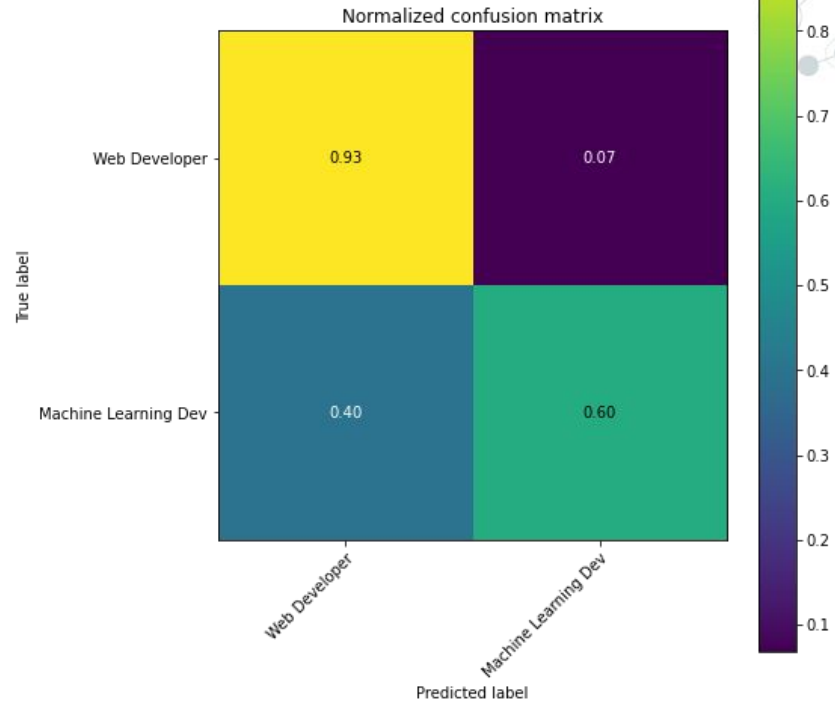
Walks per node = 10, walk length = 80,  $p = 0.25$ ,  $q = 0.25$

# Github Dataset Results

## DeepWalk

F1 Score: 0.8407

Walks per node = 10, walk Length = 80





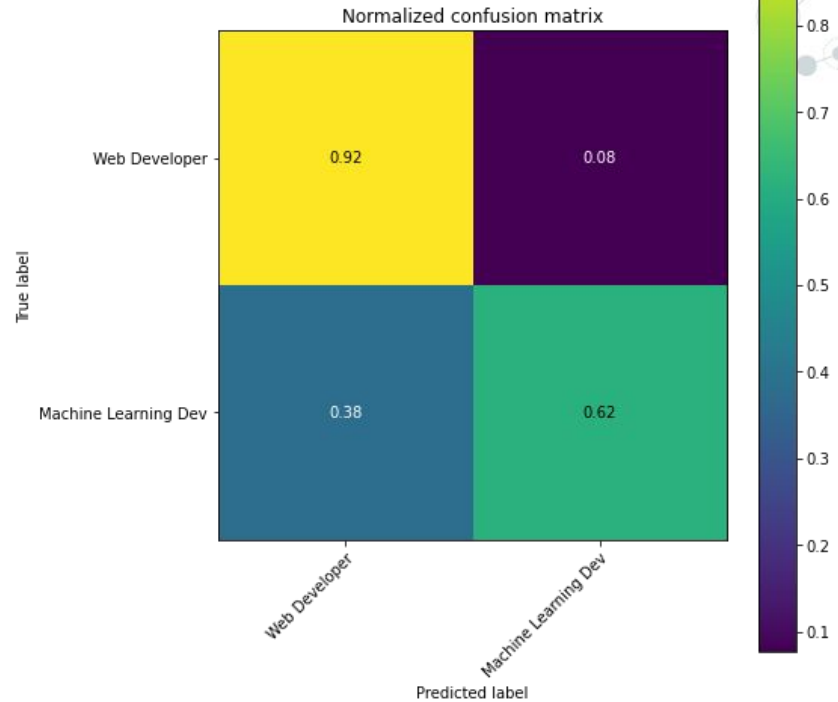
# Github Dataset Results

## Node2Vec

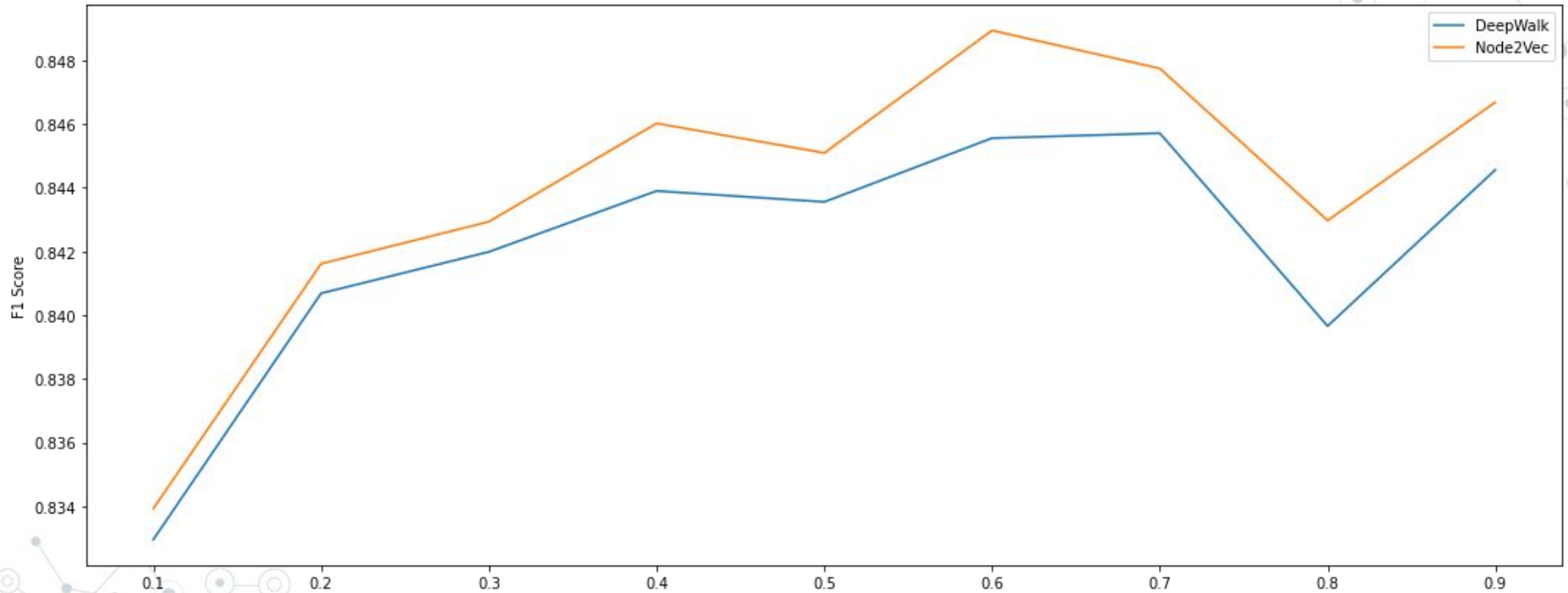
F1 Score: 0.8448

Walks per node = 10, walk Length = 80

$p = 0.25$ ,  $q = 0.25$



## Github Dataset Results



Source Code: [https://github.com/julioegb/DeepWalk\\_vs\\_Node2Vec](https://github.com/julioegb/DeepWalk_vs_Node2Vec)

A decorative network graph in the top-left corner, featuring a complex web of interconnected nodes and edges. Some nodes are highlighted with blue circles, and others with blue dots. The graph is rendered in a light gray color.

4.2

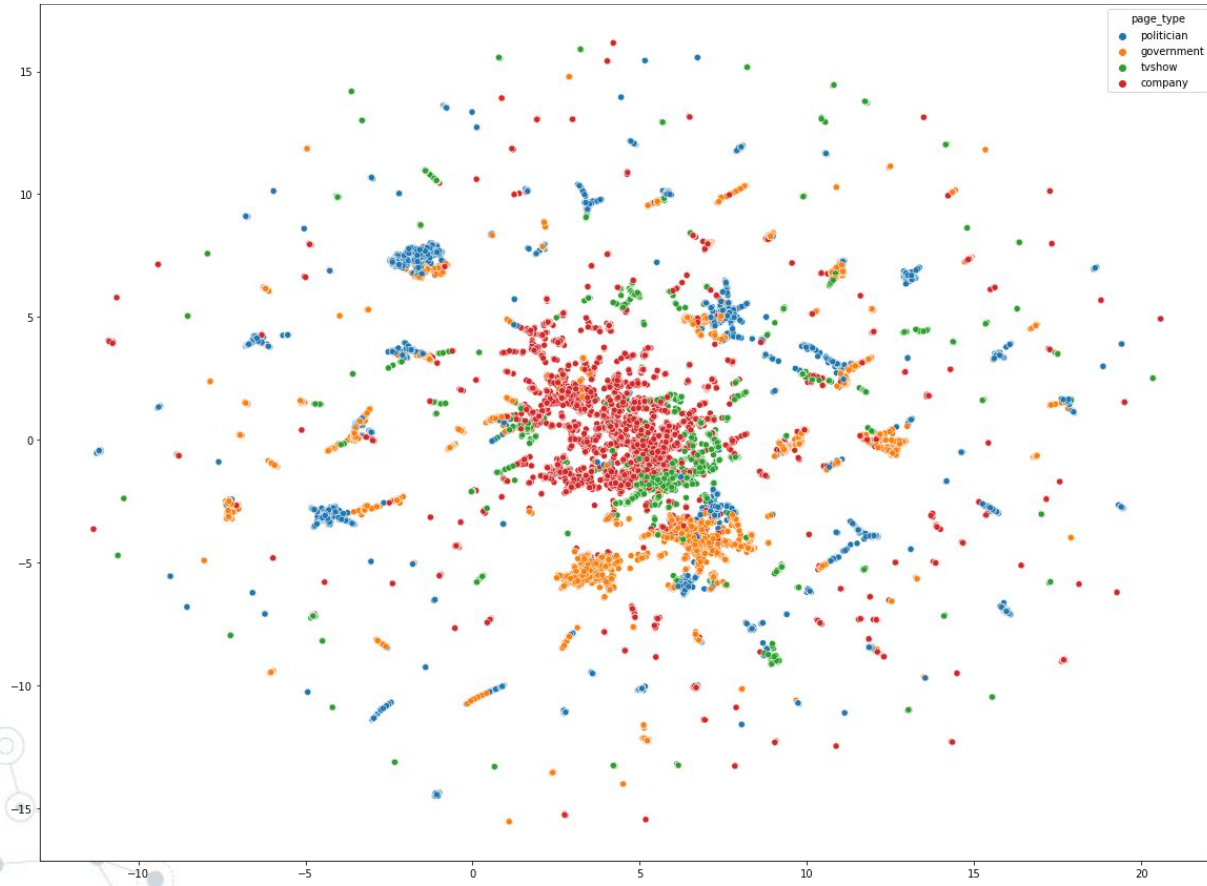
# Facebook Large Page-Page Network

A decorative network graph in the bottom-right corner, featuring a complex web of interconnected nodes and edges. Some nodes are highlighted with blue circles, and others with blue dots. The graph is rendered in a light gray color.

## Facebook Large Page-Page Network

- Nodes: 22,470
- Edges: 171,002
- Categories:
  - Politicians
  - Governmental Organizations
  - Television Shows
  - Companies
- Features are the site descriptions that summarize the purpose of the site.

# Facebook Large Page-Page Network - Deepwalk

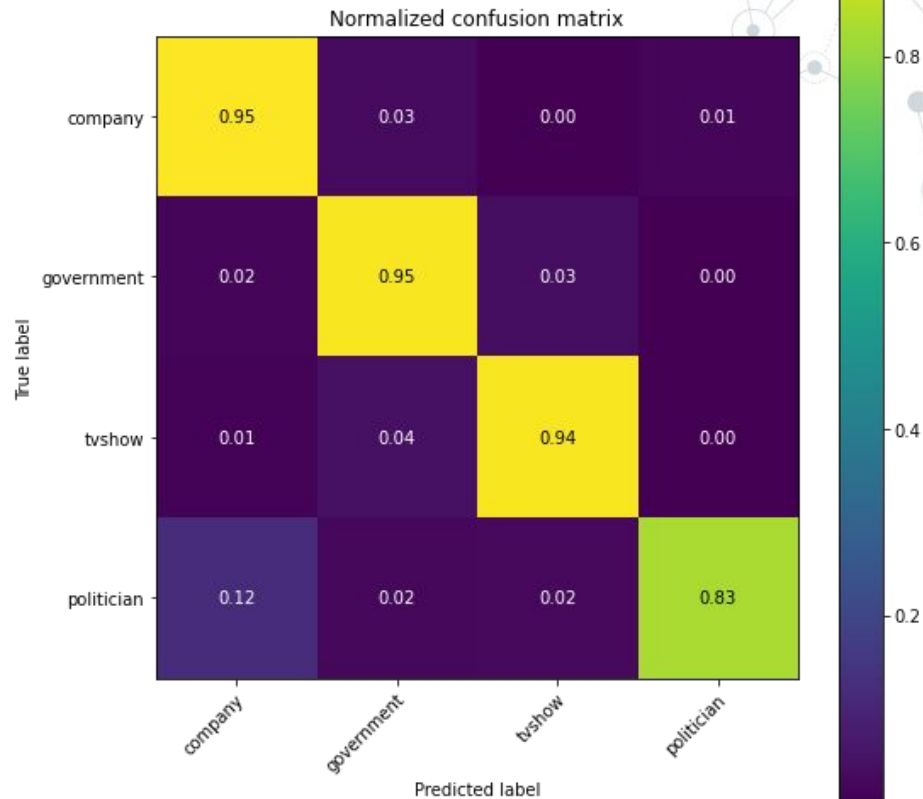


# Facebook Large Page-Page Network - Deepwalk (1)



## Deep Walk

- Random Forest
- Not Balanced
- Accuracy = 0.931

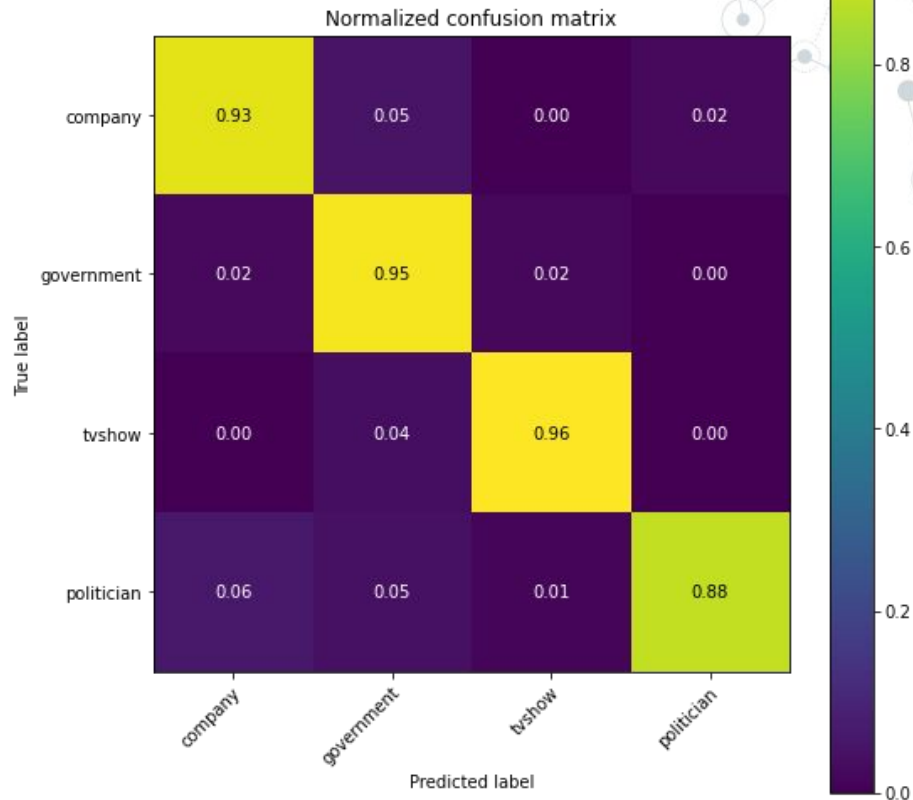


## Facebook Large Page-Page Network - Deepwalk (2)

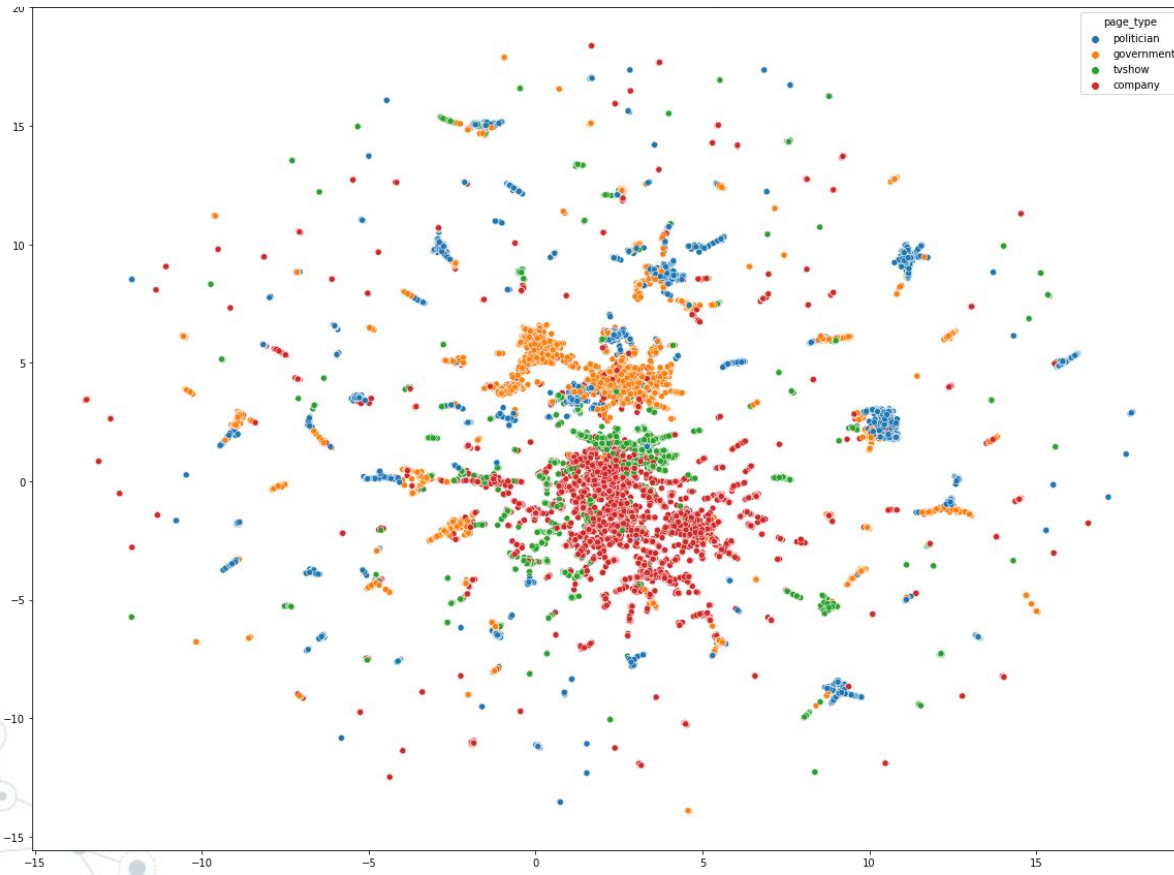


### Deep Walk

- Random Forest
- Balanced
- Accuracy = 0.937



# Facebook Large Page-Page Network - Node2Vec



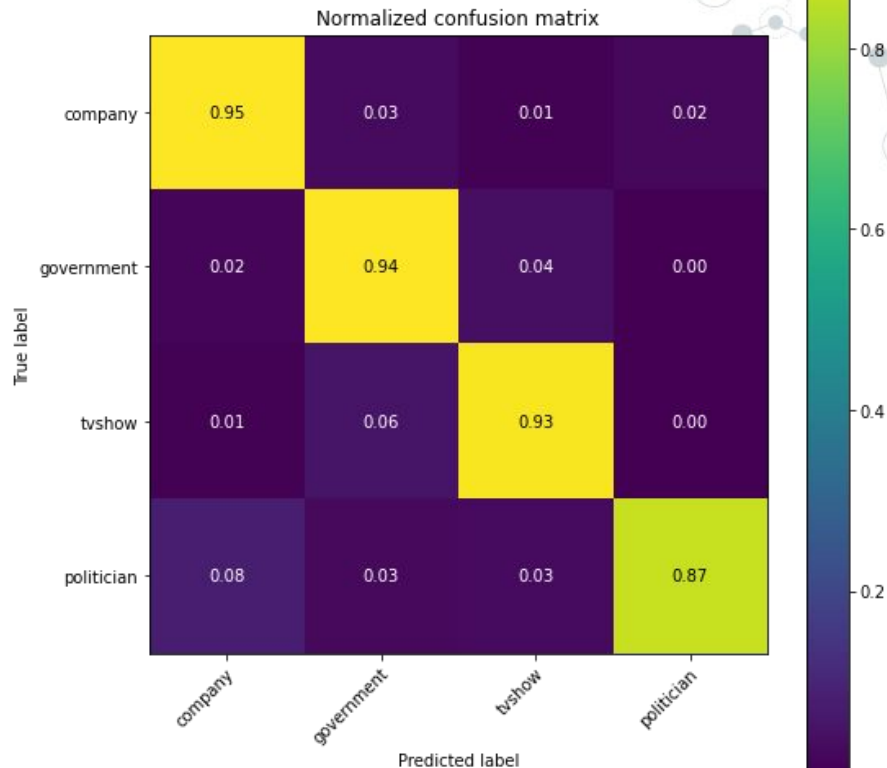


# Facebook Large Page-Page Network - Node2Vec (1)



Node2Vec

- $P = 10$
- $Q = 0.1$
- Random Forest
- Not Balanced
- Accuracy = 0.931

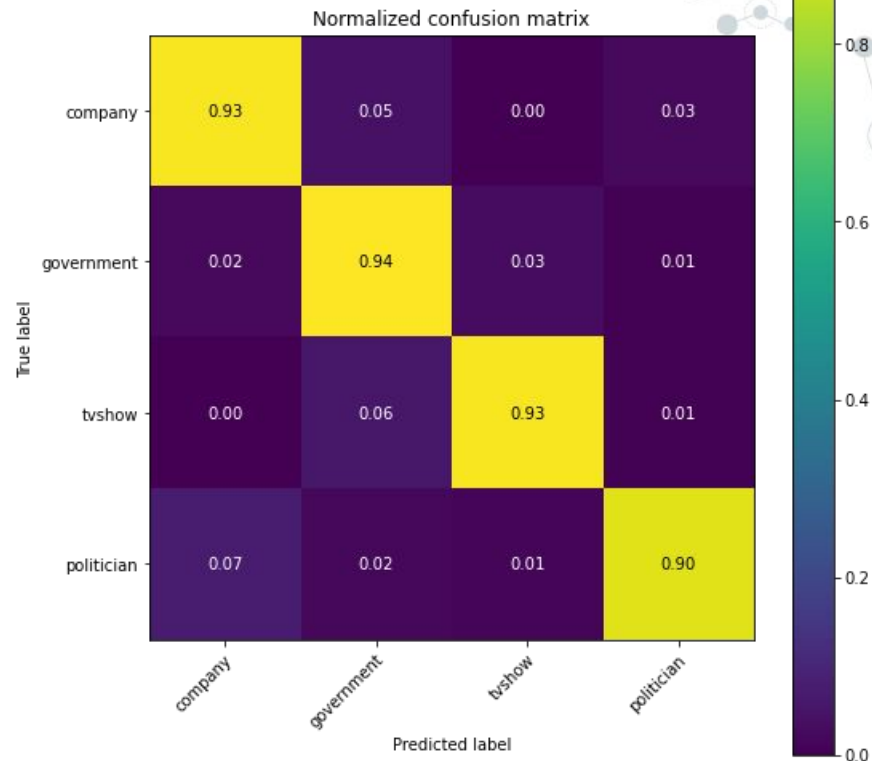


## Facebook Large Page-Page Network - Node2Vec (2)

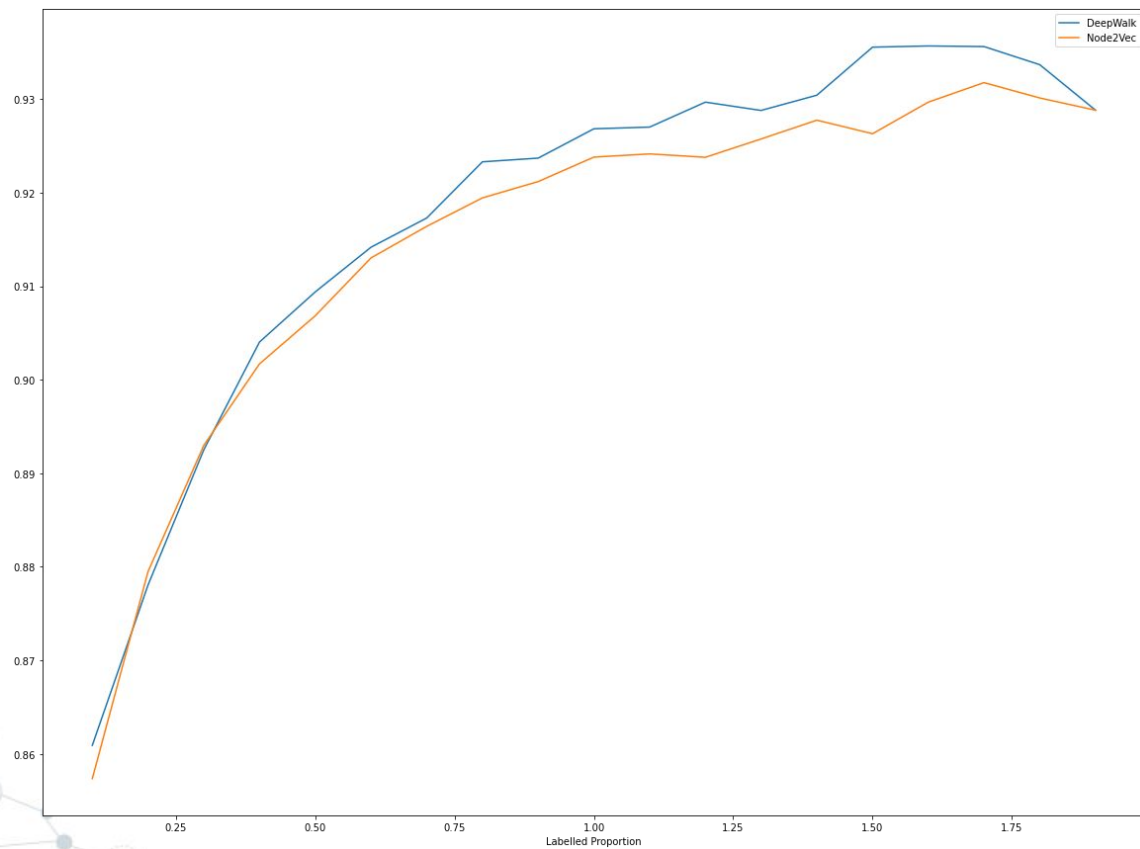


Node2Vec

- $P = 10$
- $Q = 0.1$
- Random Forest
- Balanced
- Accuracy = 0.928



# Deep Walk vs Node2Vec





## 4.3

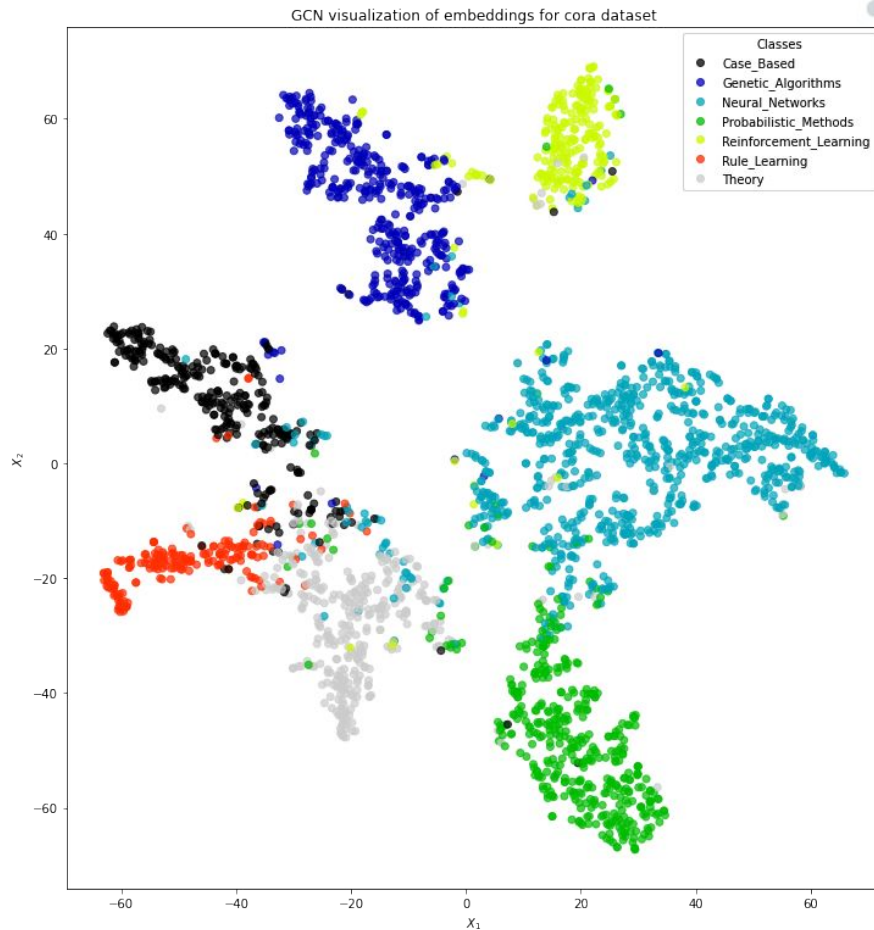
# Cora Dataset



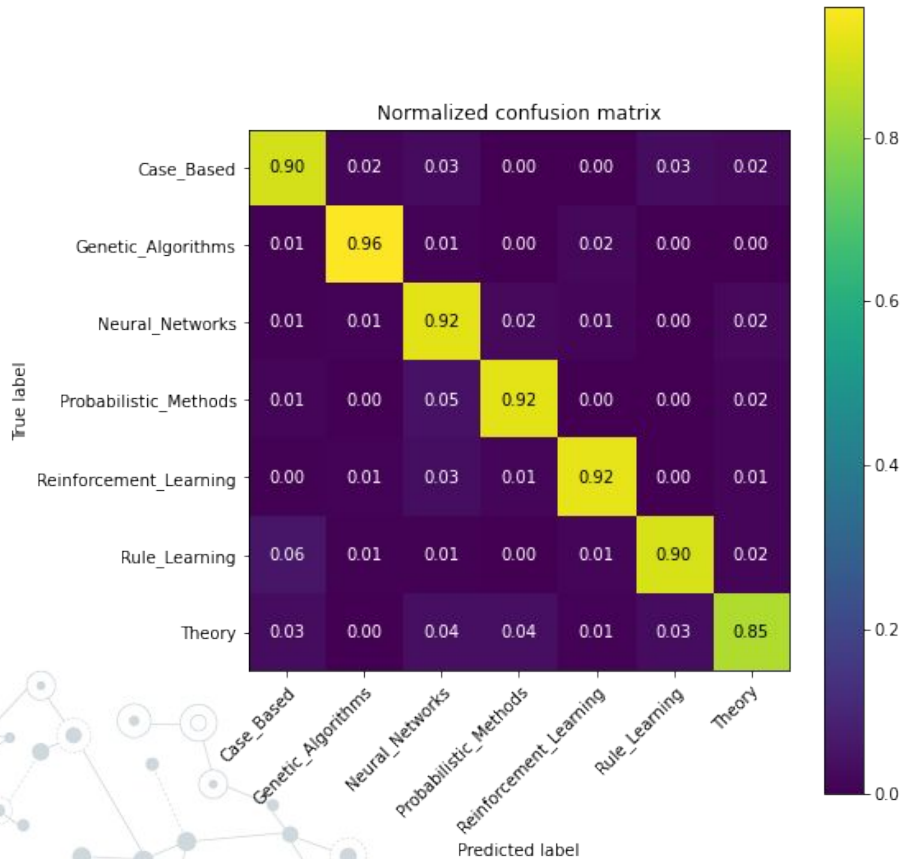
## Cora Dataset

- Nodes: 2,780
- Edges: 5,429
- Categories:
  - Neural\_Networks
  - Probabilistic\_Methods
  - Genetic\_Algorithms
  - Theory
  - Case\_Based
  - Reinforcement\_Learning
  - Rule\_Learning

# Cora Dataset Classification GCN



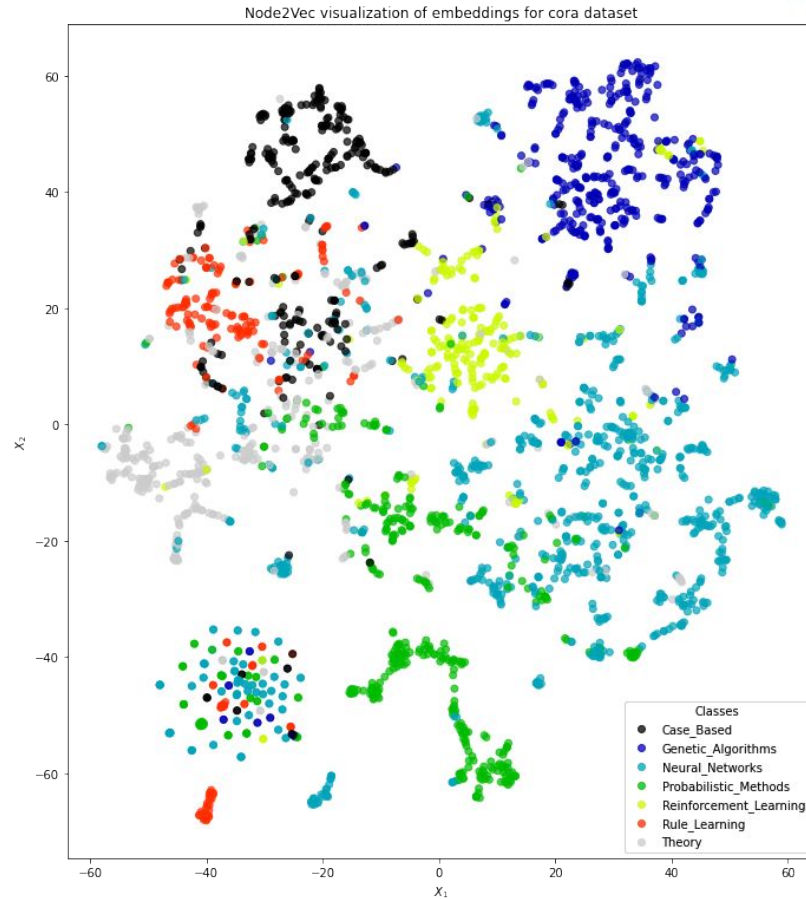
# Cora Dataset - Classification - GCN



GCN

○ Accuracy = 0.913

# Cora Dataset Classification Node2Vec



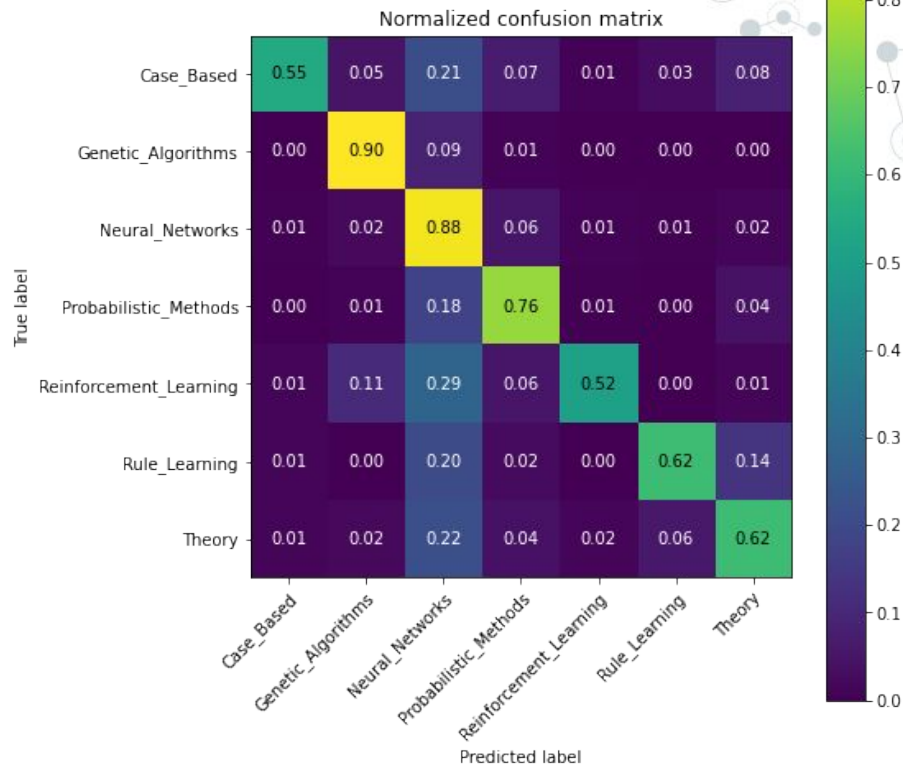


## Cora Dataset - Classification - Node2Vec

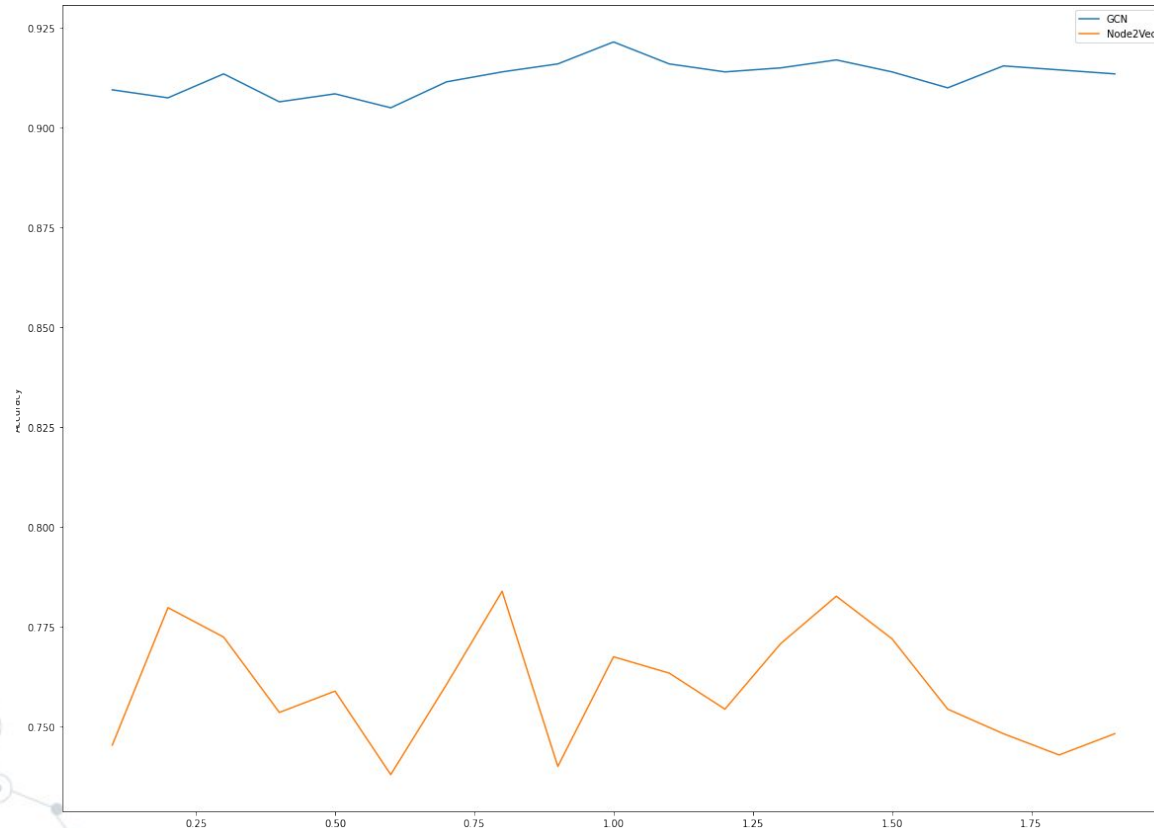


Node2Vec

- $P = 0.5$
- $Q = 0.25$
- Random Forest
- Accuracy = 0.748



# GCN vs Node2Vec



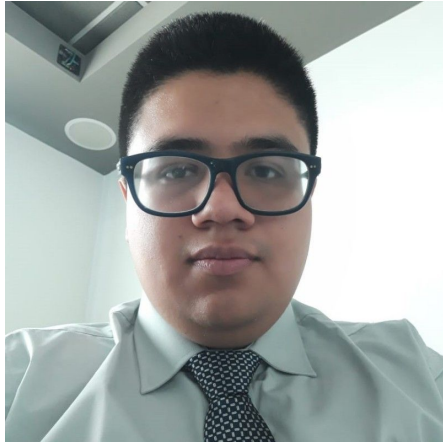
## Conclusion

- ◎ Node2Vec is an improvement to DeepWalk. Graph Convolutional Networks is an improvement of Node2Vec.
- ◎ For community detection it is better to use Node2Vec. Ranking improves compared to Deep Walk.
- ◎ For classification after comparing the different methods, GCN obtained a more precise accuracy

## References

- ❑ A. Grover and J. Leskovec, “Node2vec: Scalable Feature Learning for Networks,” Arxiv.org. [Online]. Available: <https://arxiv.org/pdf/1607.00653.pdf>. [Accessed: 09-Feb-2022].
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- ❑ Leskovec, J. (s. f.). SNAP: Network datasets: Wikipedia Article Networks. SNAP: Stanford Network Analysis Project. <https://snap.stanford.edu/data/facebook-large-page-page-network.html>
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- ❑ Mejicanos, D., 2022. *Research/Research\_GCN\_Dennis\_Mejicanos.ipynb at main · DennisM23/Research*. [online] GitHub. Available at: <[https://github.com/DennisM23/Research/blob/main/Research\\_GCN\\_Dennis\\_Mejicanos.ipynb](https://github.com/DennisM23/Research/blob/main/Research_GCN_Dennis_Mejicanos.ipynb)> [Accessed 13 February 2022].
- ❑ Jupyter Notebooks DeepWalk and Node2Vec Github Dataset. Available at: <[https://github.com/julioegb/DeepWalk\\_vs\\_Node2Vec](https://github.com/julioegb/DeepWalk_vs_Node2Vec)> [Accessed 13 February 2022].
- ❑ GitHub Social Network. *Dataset*. [online] Available at: <<https://github.com/benedekrozemberczki/datasets>> [Accessed 13 February 2022].

## Our Team



**Dennis Mejicanos**

Telecommunications Engineer  
dennis.mejicanos\_garcia@telecom-sudparis.eu



**Luis Valenzuela**

Telecommunications Engineer  
luis.valenzuela\_navarro@telecom-sudparis.eu



**Julio Guzmán**

Computer Science Engineer  
julio.guzman\_barraza@telecom-sudparis.eu