# Twitter Community detection

Unveiling the Twittersphere

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### Outline

- Introduction
- Communities in social media
- Dataset description
- Feature extraction
- Clustering phase
- Results & Experimentation
- Conclusion

### Introduction



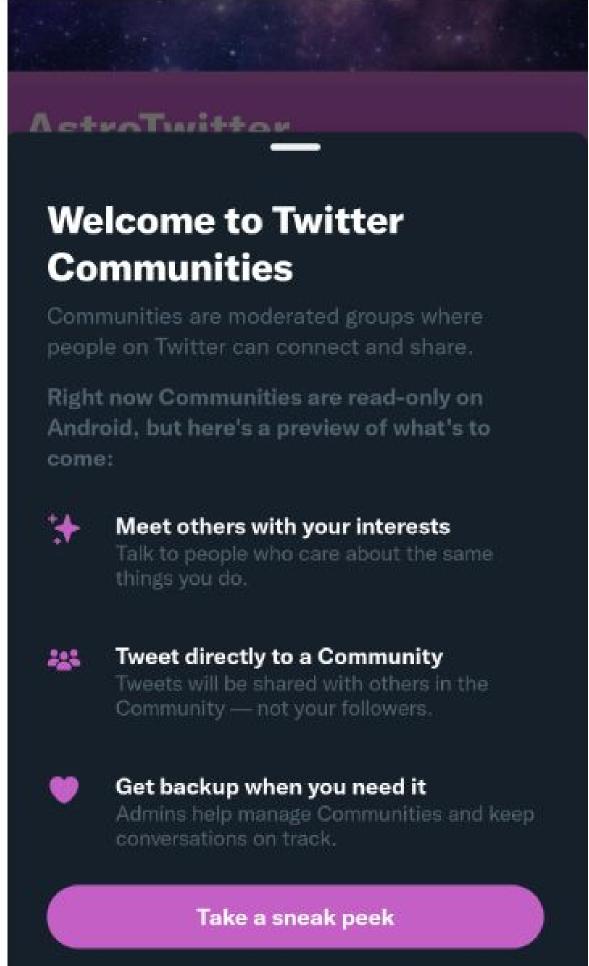


## Communities in social media

• The social hierarchy of social media users

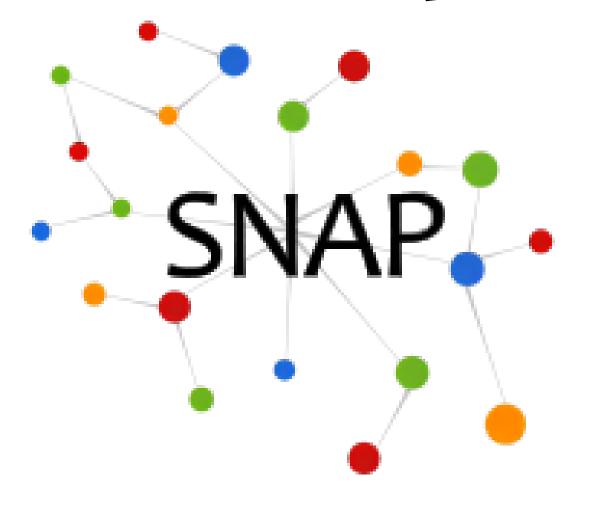








## Dataset Description



Dataset statistics				
Nodes	81306			
Edges	1768149			
Nodes in largest WCC	81306 (1.000)			
Edges in largest WCC	1768149 (1.000)			
Nodes in largest SCC	68413 (0.841)			
Edges in largest SCC	1685163 (0.953)			

Average clustering coefficient	0.5653
Number of triangles	13082506
Fraction of closed triangles	0.06415
Diameter (longest shortest path)	7
90-percentile effective diameter	4.5



### Files hierarchy

### Project directory

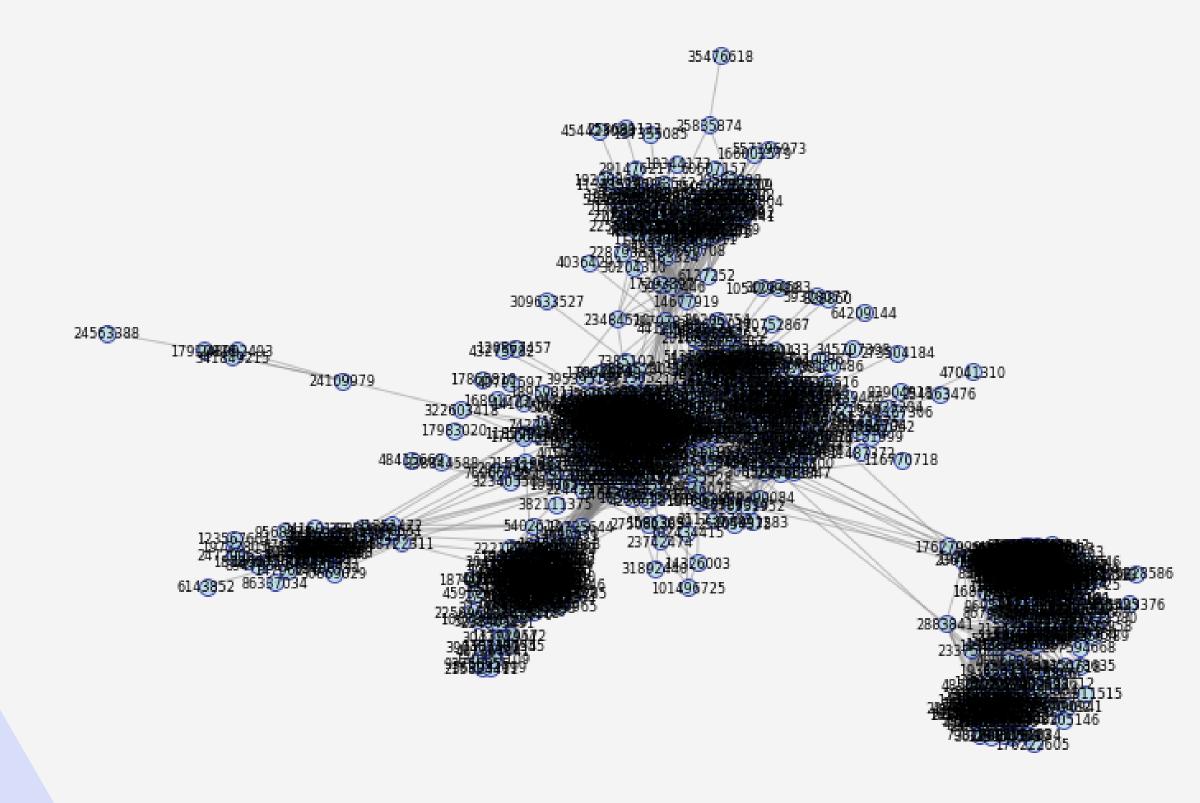
- twitter\_combined.txt
  - Twitter directory

NodelD.egofeat

NodelD.featnames

## Graph network

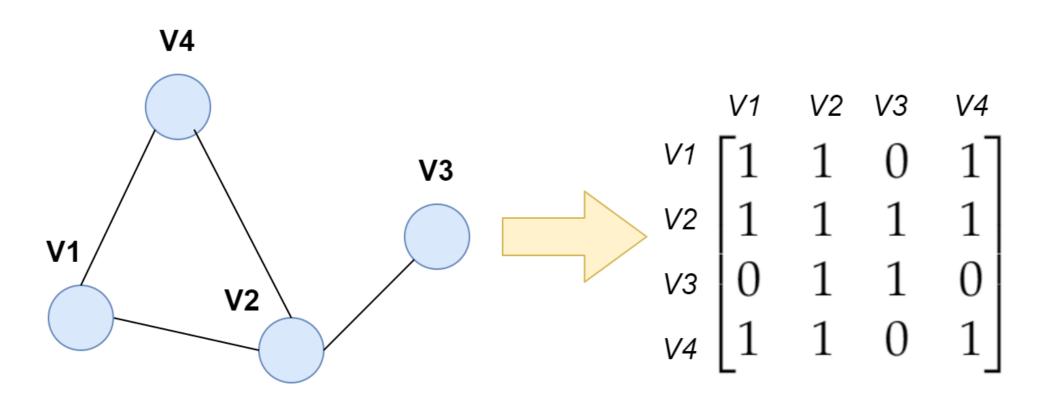




#### Directed graph A--->B

Node A follows Node B

## Feature Extraction





## Edge-based Approach

#### **Extract Adjacency Matrix**

## Load the dataset

• Extract the edges from the twitter\_combined.txt

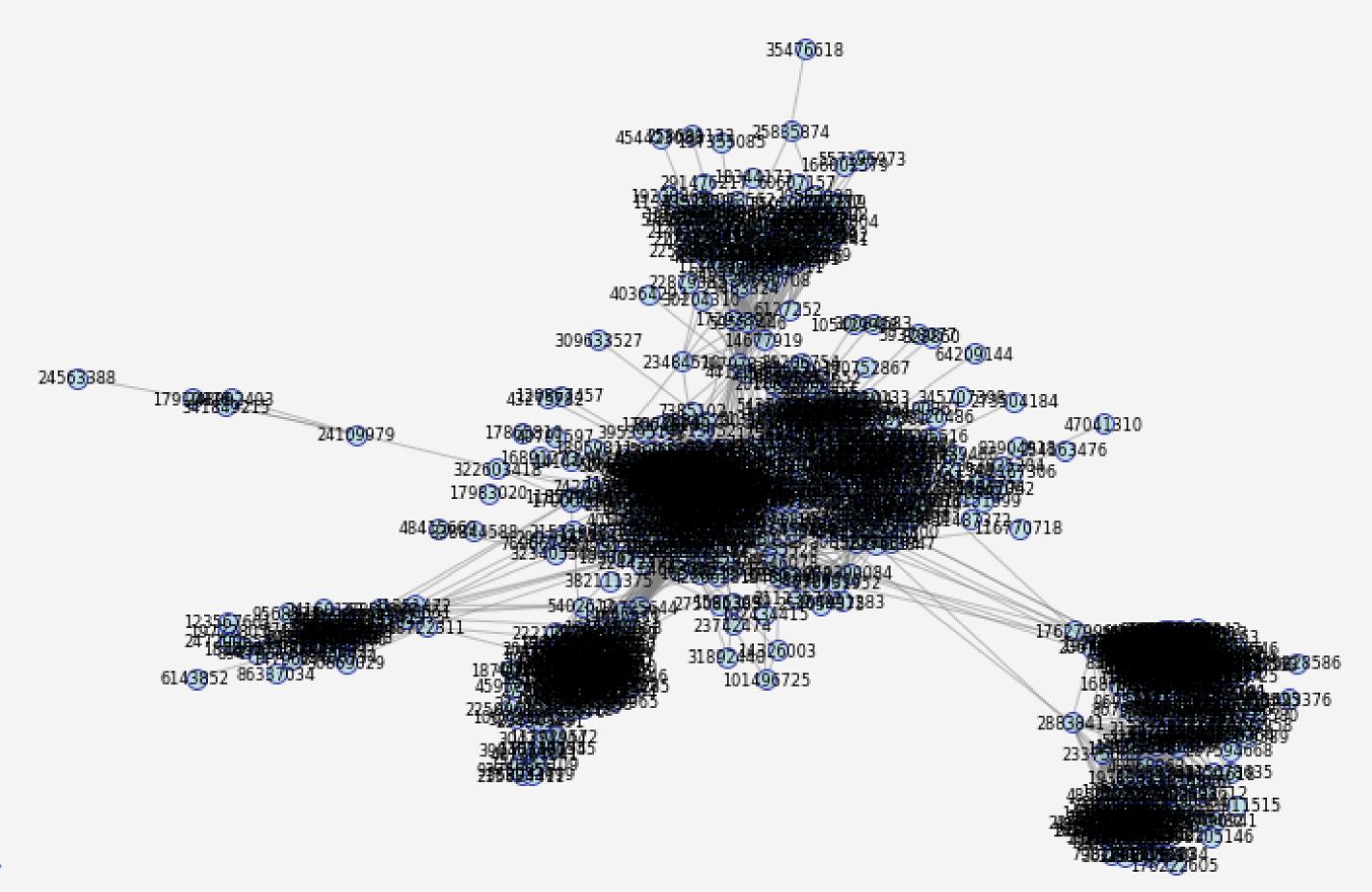
## Form the list of edges (Node i ,Node j)

• These edges will be used in the process of creating the adjacency matrix

## Construct the Adjacency

 Check the connection between nodes and fill in the adjacency matrix

### Graph network



# Feature-based approach

#### **Extract Adjacency Matrix**

Load the dataset

Form the list of edges (Node i ,Node j)

Extract account features from nodeld.feature names

• Check the connection between

matrix

nodes and fill in the adjacency

• Extract the edges from the twitter\_combined.txt

• These edges will be used in the process of creating the adjacency matrix

Store extracted data in a Dataframe (972,24250)

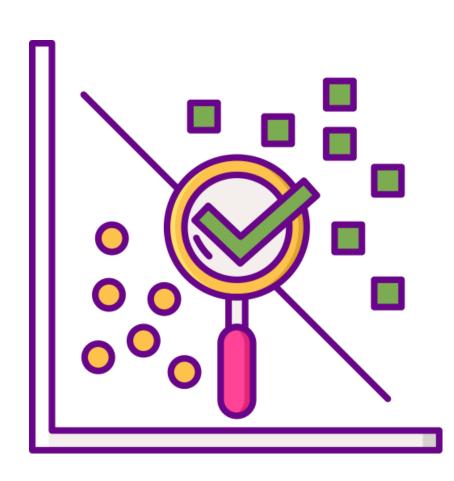
Compute degree, closeness and betweenness centrality

17/30

#### Final Dataset

	nodeld	#OCTAVIA	#THEHELP	#ff	@BAFTA	@FuckYes	sEmma	@JUD#	AOcombr	@astowe	ellcom @er	nmastonebr	@helpmovie	@	ochococat
968	14528221	0	0	1	0		0		0		0	0	0		0
969	14840869	0	0	0	0		0		0		0	0	0		0
970	82726142	0	0	0	0		0		0		0	0	0		0
971	255790981	0	0	0	0		0		0		0	0	0		0
972	36618690	0	0	0	0		0		0		0	0	0		1
@hel	pmovie	@chococa	t @danny	Bstyl	e, @jon_	blaze55	@kylepı	ulver:	@terryca	wanagh	@twitchtv.	Degree Centrality			veenness Centrality
	0	(	0		0	0		0		0	0	0.001030	0.238165		0.000000
	0	(	0		0	0		0		0	0	0.003090	0.239990		0.001701
	0	(	0		0	0		0		0	0	0.002060	0.206245		0.000958
	0	(	0		0	0		0		0	0	0.009269	0.273984		0.002988
	0	:	1		1	1		1		1	1	0.010299	0.294064		0.010228

## Clustering Phase





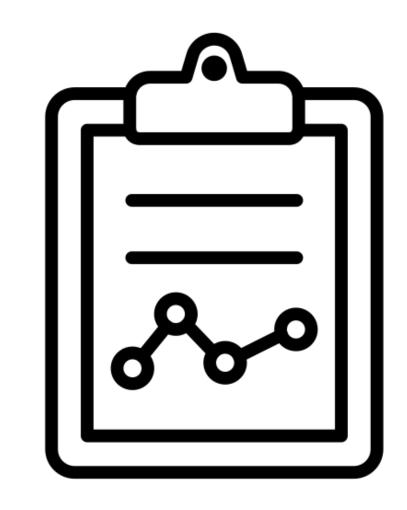
## Edge-based Approach

Parameter	Value
Number of clusters	3
Affinity	Precomputed
Algorithm	Spectral Clustering

# Feature-based approach

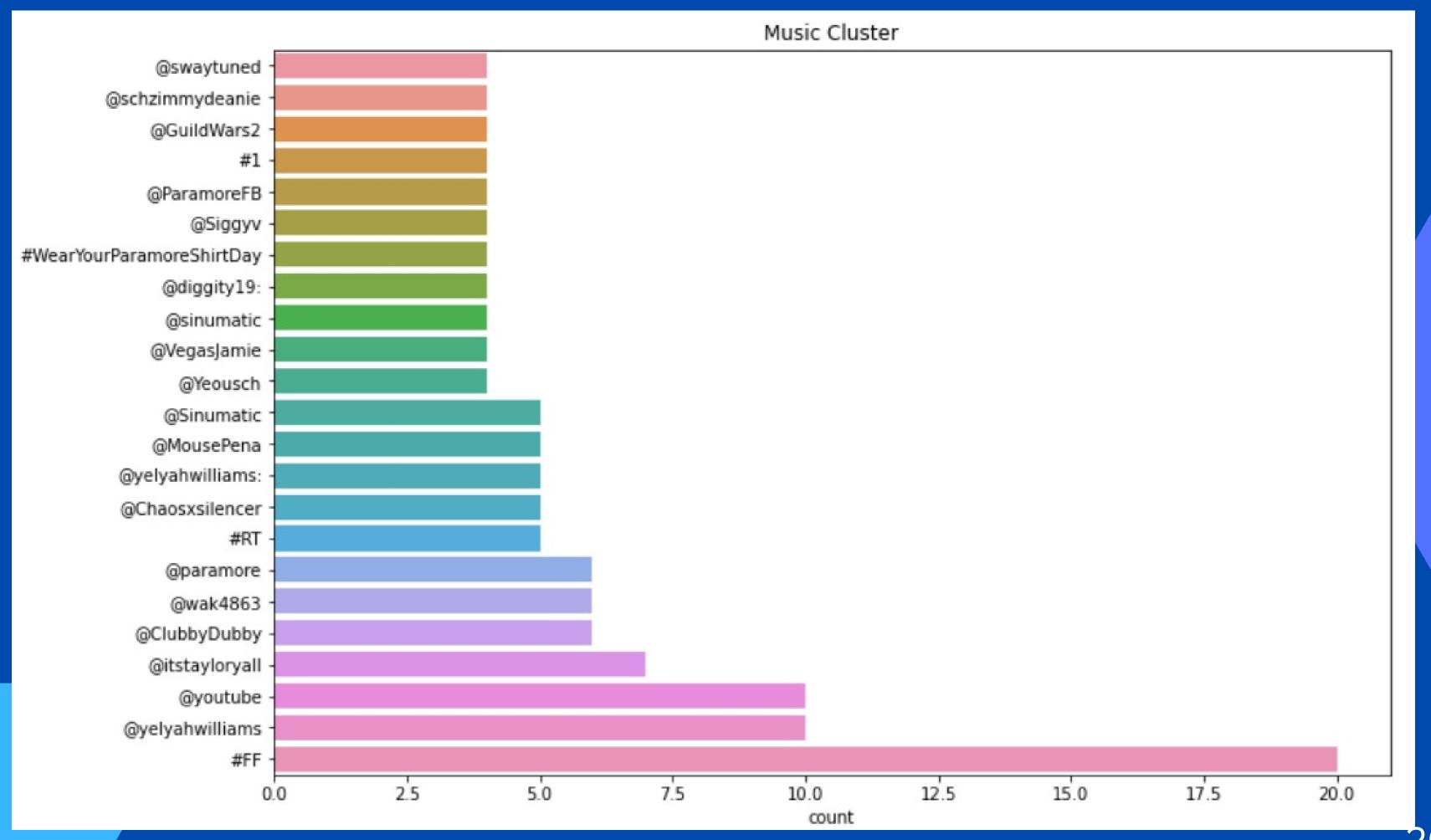
Algorithm	Number of Clusters	Data	Random Seed
k-means	3	df	42
hierarchical	3	df	
spectral	3	df	42

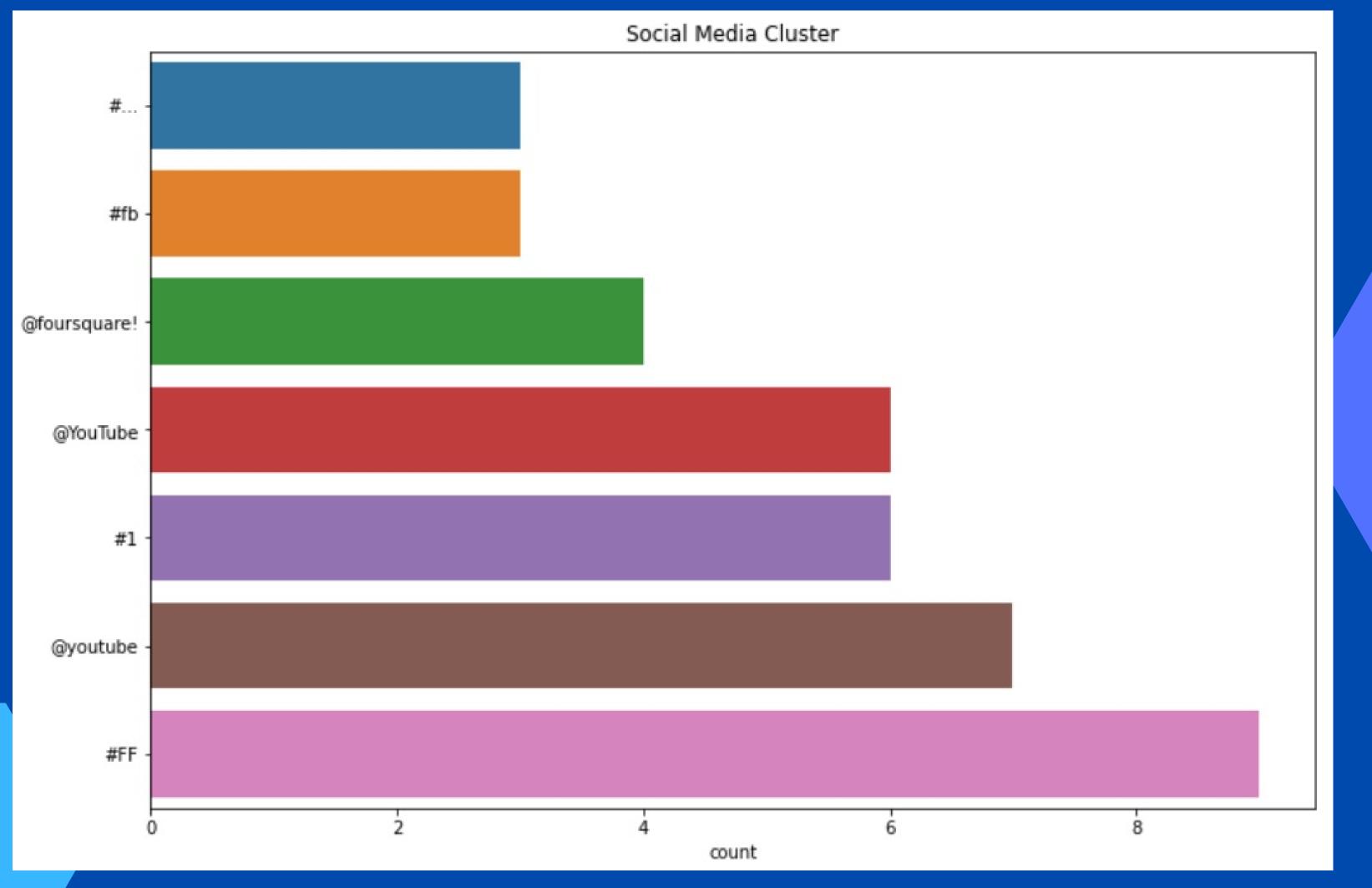
## Results & Experimentation

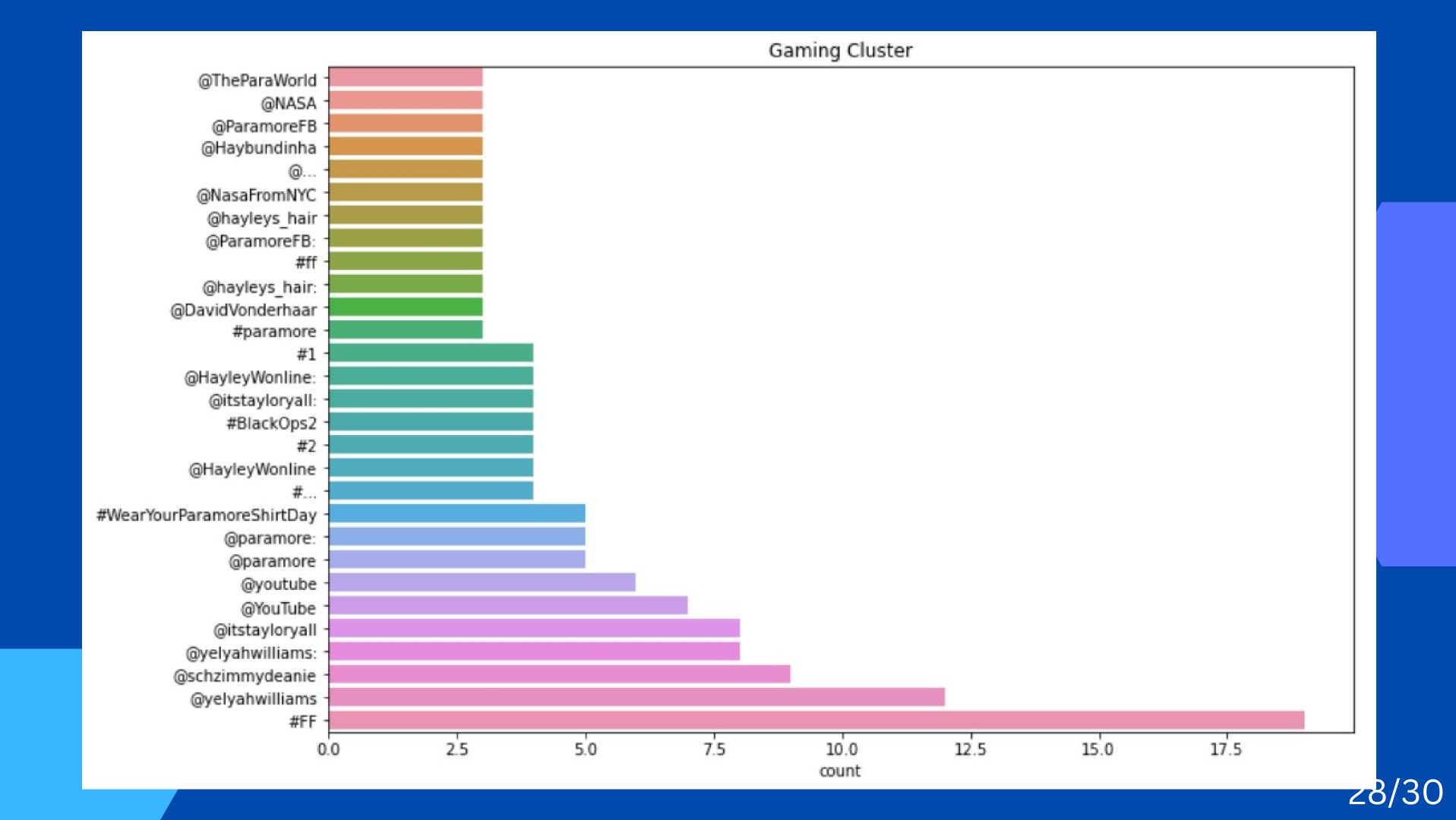




Method	Clustering Algorithm	Silhouette Score	
Method 1: Edge- based Approach	Spectral Clustering	-0.1887	
	Spectral Clustering	-0.0343	
Method 2: Feature- based Approach	Hierarchical Clustering	0.6756	
	K-MEANS	0.6917	







## Conclusion





## Thank you for your attention

Feel free to ask questions