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11      **Main Manuscript for**

12      Alexandria Ocasio-Cortez Bluesky follower's network.

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14      Wirginia Szoltysek.

15      Email: [s251186@dtu.dk](mailto:s251186@dtu.dk)

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29    **Abstract**

30    This analyzes the social network of samples of Alexandria Ocasio-Cortez's followers on the  
31    Bluesky platform. Using a dataset of 50, 000 followers and their posts, I examined how users  
32    interact through replies and posts. The network is very sparse, with most users having few  
33    connections, and only small groups showing strong mutual interactions. These findings show  
34    patterns of engagement and community structure among followers on BlueSky. The document  
35    starts with introduction and following results. It ends with discussion about findings.

36    **Significance Statement**

37    The study contains large database of BlueSky users posts with parent/roots hierarchy.  
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40    **Introduction**

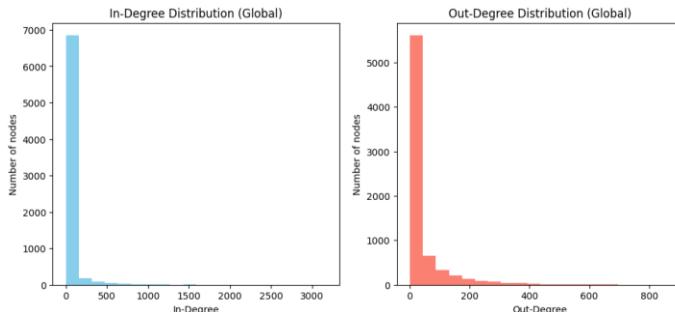
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42    BlueSky is a microblogging social media service with over 40 million users who publish short  
43    posts and interact through replies, reposts or follows. It is often used by public figures to promote  
44    their views. One of them is Alexandria Ocasio-Cortez, an American politician, activist, and  
45    member of the Democratic Party. This study examines the interactions between small fractions of  
46    her follower base of over 2 million accounts and explores the community structures and roles  
47    different accounts play within the network.

48    To explore the data more closely, data of 50 000 followers were extracted including all their  
49    BlueSky posts/replies. The dataset was cleaned to keep only the posts which connected  
50    minimum two users what resulted in around 800 000 posts of 7300 unique members.

51    **Results**

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54    The **Global Blusky** network of the tested sample consists of 7266 nodes and 300257 edges  
55    with a low density of 0.0057 which indicates very sparse network. It is mostly weakly connected,  
56    with large component of 7264 nodes, but has small strongly connected groups (2560 SCCs),  
57    reflecting limited mutual interactions.

58    The Global Degree Distribution graph (Figure 1.) is highly skewed. Most nodes have very low  
59    degree.

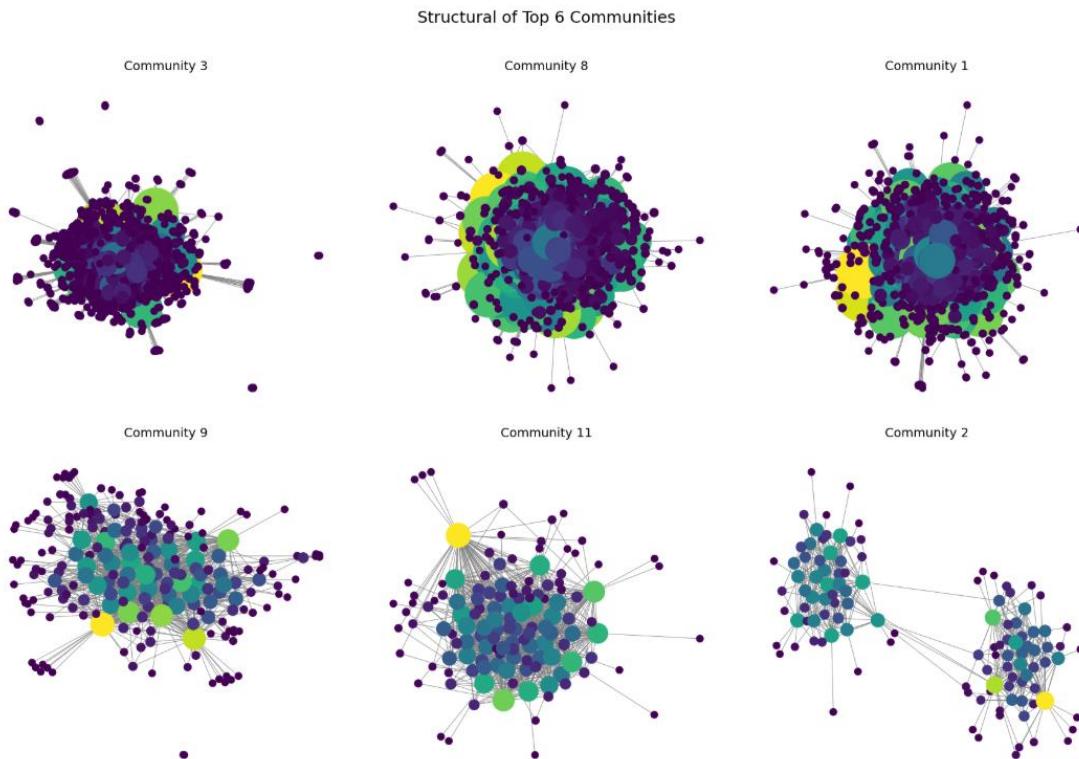


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61    *Figure 1 Global Degree Distributions*

62    This structure motivates us to do community-level analysis to uncover more meaningful connections and  
63    influences.

Using Louvain algorithm, the global network was divided into **multiple communities** and only six largest were selected for further analysis. For clearer visualization purposes, disparity filter was applied to extract the backbone of each community. The size and color of the node is dependent on how high the node degree is.

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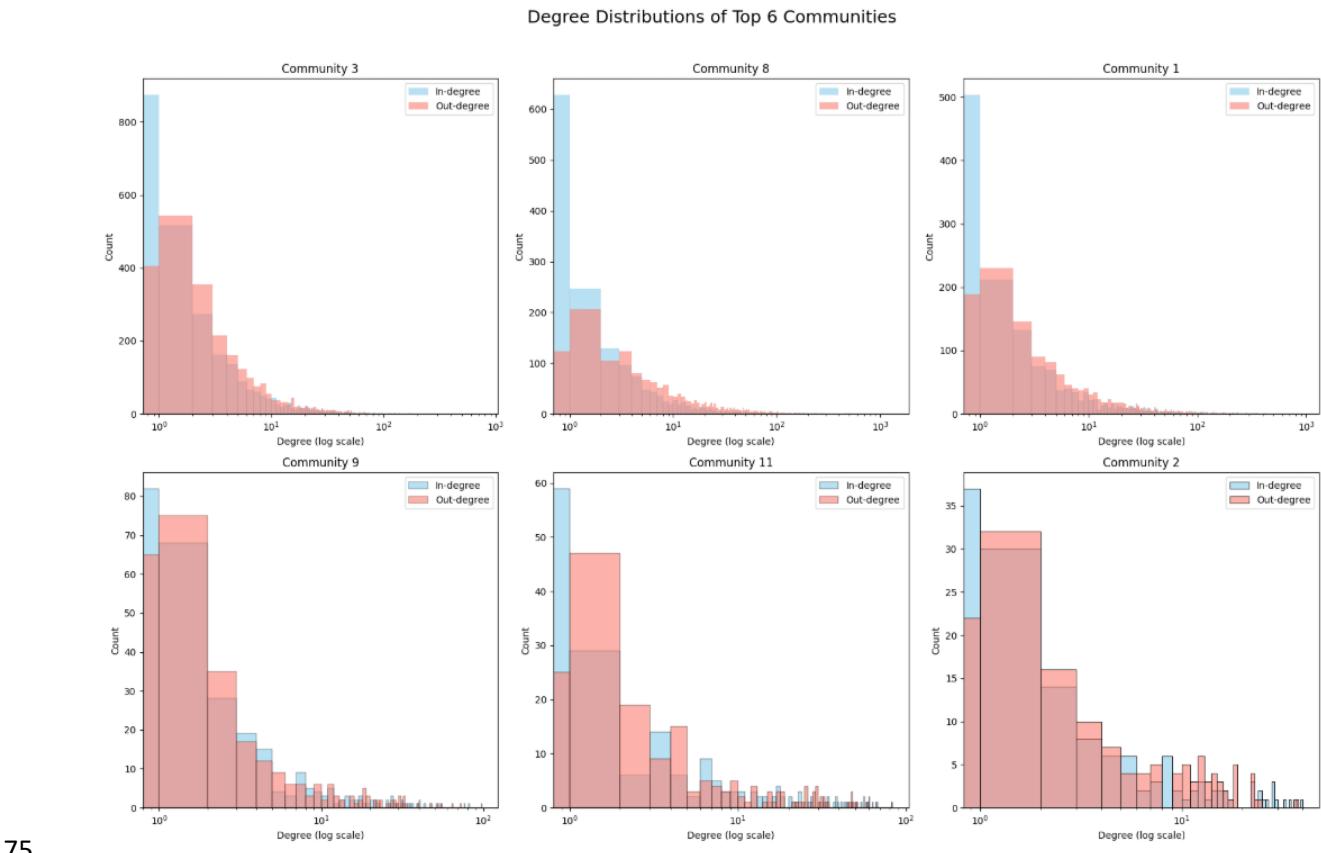
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Figure 2 Social Network graphs based on Community. We can observe Community 3, 8 and 1 are densely connected forming tight clusters. There are a few big nodes with high degree nodes. Community 9, 11 and 2 are less dense and more heterogeneous.

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Community	Hashtags	Sentiments
3	#ithashappenedhere, #nokings, #trump	5.431
8	#epstein, #nokings, #epsteinfiles	5.369
1	#epstein, #epsteinfiles, #releasetheepsteinfiles	5.400
9	#survivormusic, #breakfastcrew, #strongertogether	5.798

74 Figure 3 Most common tags per Community and average Sentiment.

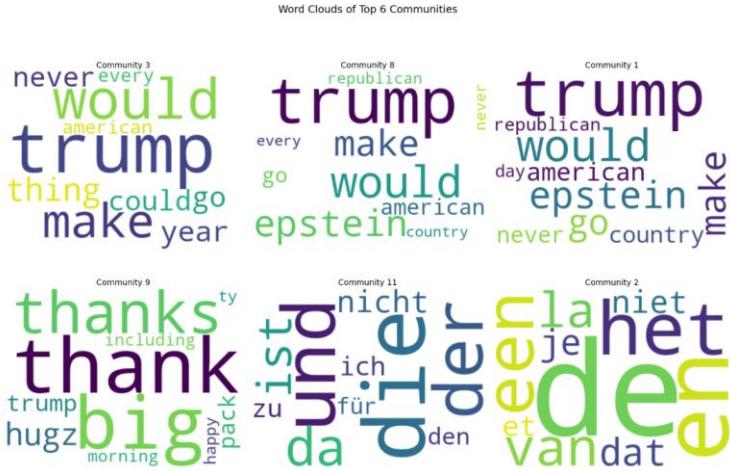


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Figure 4 Degree Distribution graphs. Community 3,8 and 1 contain highly skewed In-degree numbers illustrating users whose posts are getting high numbers of replies and may be popular or influential. Community 9,11 and 2 are also skewed but with much more balanced way. Both, in and out degrees look relatively similar.

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Figure 5 Word Clouds. Community 3, 8 and 1 have mainly words related to political topics. Community 9 contains optimistic words, and Community 11 and Community 2 are random German and Dutch words.



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85 **Discussion**

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87 The result shows that even after detecting communities, the graphs are sparse and the  
88 interactions are not very high. There are few users who interact much more and they have also  
89 many followers. Based on WordClouds we can see most of the biggest communities were  
90 involved into politics topics and only one seemed to have bit different areas of interest.

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92 The study shows only small percentage of followers and not all the interactions due to lack of time  
93 and large size of the source data. If reposts and likes were added to the set, we could probably  
94 see denser network.

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98 **References**

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100 1 Wikipedia(n.d.)*AlexandriaOcasio-Cortez*. Available at:  
101 [https://en.wikipedia.org/wiki/Alexandria\\_Ocasio-Cortez](https://en.wikipedia.org/wiki/Alexandria_Ocasio-Cortez) (Accessed: 10 December 2025).  
102 2 Wikipedia (n.d.) *Bluesky*. Available at: <https://en.wikipedia.org/wiki/Bluesky>. Accessed: 10  
103 December 2025.

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105 **Figures and Tables**

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107 **Figure 1.** Global in/out Distribution Graph

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109 **Figure 2.** Network Graphs for top 6 Communities

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111 **Figure 3.** Distribution graphs for top 6 Communities.

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113 **Figure 4.** World Clouds for top 6 Communities

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