Structure and Evolution of GitHub Collaboration Network

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Data-Driven Modeling of Complex Systems

Introduction

What?

• Our objective is to analyze the **Collaboration Network** between users on Github to understand its **structure** and **evolution**

Why?

- To **understand the dynamics** of the interactions between users on github and how it has grown over time
- Seemed an interesting research study

How?

• Github Membership Network¹ (177.000 nodes, 440.000 edges)



The Starting Dataset

Node meaning

Edge meaning

Network format

Edge type

User, project

Membership

B Bipartite, undirected

Unweighted, no multiple edges

Size	n = 177,386
<u>Left size</u>	n ₁ = 56,519
Right size	n ₂ = 120,867
<u>Volume</u>	m = 440,237

The initial Dataset was a **bipartite**, **undirected graph** containing projects on one side and users on the other.

The Dataset also had **timestamps** for the **creation of projects**, ranging from mid-2008 to mid-2009

However, to analyze collaborations between users, we projected this bipartite network onto the user set, creating a **Unipartite**, **undirected weighted graph**

Building the Collaboration Graph



~ 11 milion edges

The Final Collaboration Graph was constructed as follows:

Nodes: represent **Users**. The attribute 'num_projects' indicates the number of unique projects for each user.

Edges: represent **Collaborations between users**. The weight is proportional to the number of projects shared by the users.

Isolated nodes were deleted for a more insightful analyses.

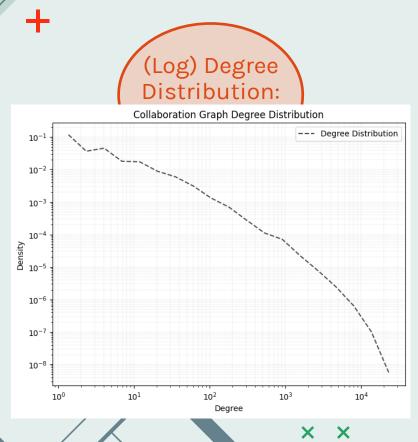
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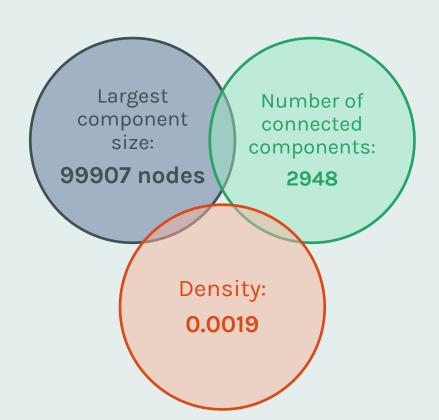
Research Questions

1. How is the Collaboration Network among GitHub users structured?

2. How does the Collaboration Network evolve over time? Are collaborations typically recurrent or users tend to seek new collaborators?

Overview of Graph Metrics





Scale-Free Property Analysis

The graph showed characteristics consistent with a scale-free network

The estimated exponent of the power law for this network is γ =2.3, which falls in the typical range for scale-free networks

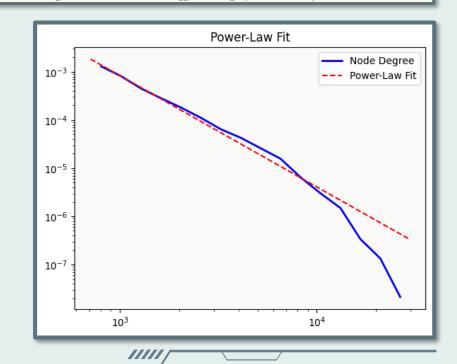
A comparison with exponential distribution showed a **p-value** of **0.00** and a **high R value**, indicating a significative and strong tendency towards a Scale-Free distribution

From the chart, we observe that the curve is consistent with a power law.



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=== Scale-Free Analysis ===

Gamma: 2.3089
p-value: 0.0000
R: 11.9656
There is significant evidence to suggest the graph follows a power-law distribution
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Community Detection

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N° Communities detected:

3382

Louvian algorithm Modularity:

0.2811

For this analysis we chose the **Louvian Community Detection algorithm**, as it's one of the fastest and most scalable:

3382 communities detected, with more central users often acting as **hubs**.

Among the communities found, there was a balance between smaller and larger communities.

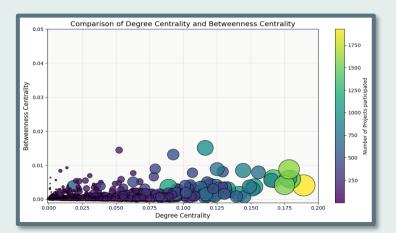
However, the modularity value indicates a **relatively**weak community structure.



Centrality Metrics

X	×	
	×	

id	degree_centrality	betweenness_centrality
user_17	0.272642	0.028591
user_299	0.189293	0.004054
user_607	0.179747	0.005841
user_645	0.178488	0.008774
user_8	0.177577	0.006048



Degree centrality highlighted users with the most connections in the network, confirming their role as **hubs**.

Betweenness centrality identified users who act as **bridges** connecting different parts of the network.

The chart confirms that most users cluster around low centrality measures and that users who participated in more projects tend to be more central.

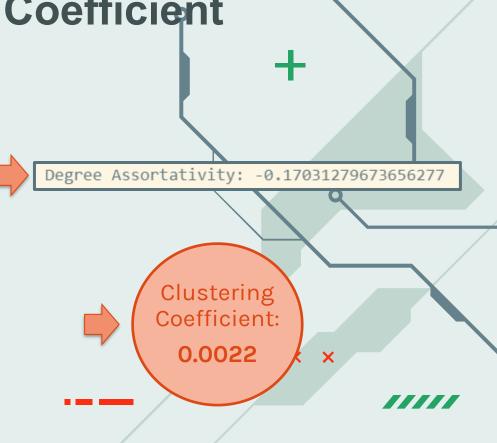


Assortativity and Clustering Coefficient

Next, to understand how highly connected users interact with less connected ones, we measured **Degree assortativity**:

The negative assortativity value indicates a slight **Dissortative Behaviour** in the Network

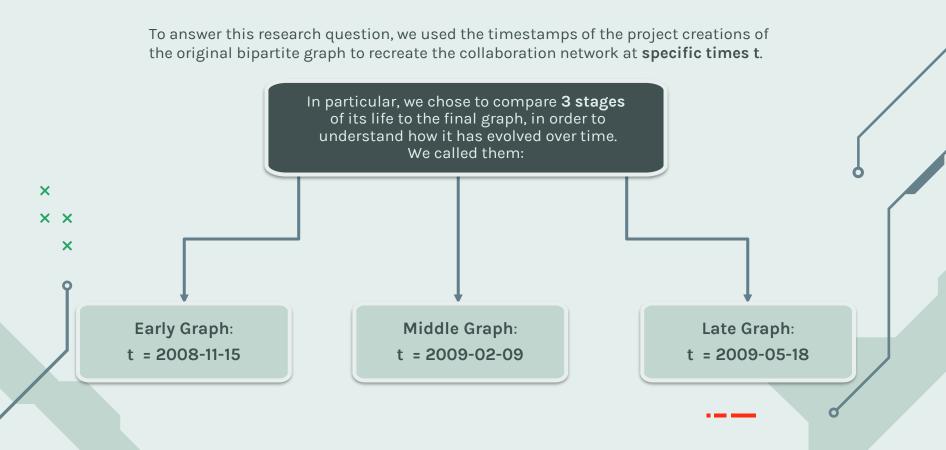
The Clustering Coefficient resulted low, indicating that nodes don't tend to form dense clusters with their neighbors



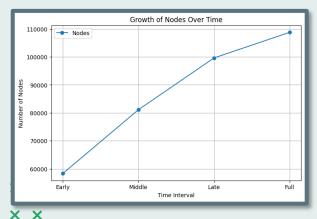
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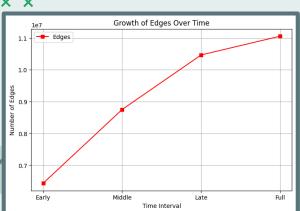
How does the Collaboration Network evolve over time? Are collaborations typically recurrent or users tend to seek new collaborators?

The Temporal Collaboration Network



Early, Middle, Late graphs





Firstly, we observed that the number of nodes and edges **grew almost linearly** over time, slowing down only during the last interval.

The dates span 3 months from each other.

We notice the graph nearly doubled its size in just the span of 6 months.

Comparison of Clustering Coefficient and Degree Distribution

Estimated Clustering Coefficient:

Early: 0.0039 Middle: 0.0030

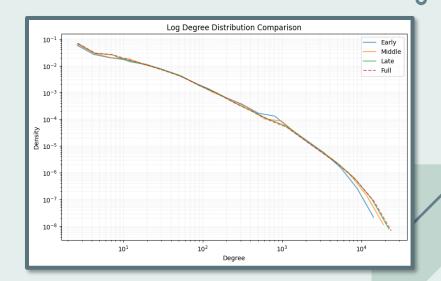
Late: 0.0025

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Comparing the degree distributions, we observe that the graph always maintained a similar trend, with slight variations to the high degree nodes.

This suggests that the graph has always been prone to the **formation of hubs.**

The clustering coefficient decreases over time, indicating that the network becomes less locally clustered as it grows. This is consistent with the decreasing density and modularity over time.



Community Detection Comparison

Community Detection with Louvian Algorithm was was used for each graph, and then the divisions for each interval were **compared to the Final graph**:

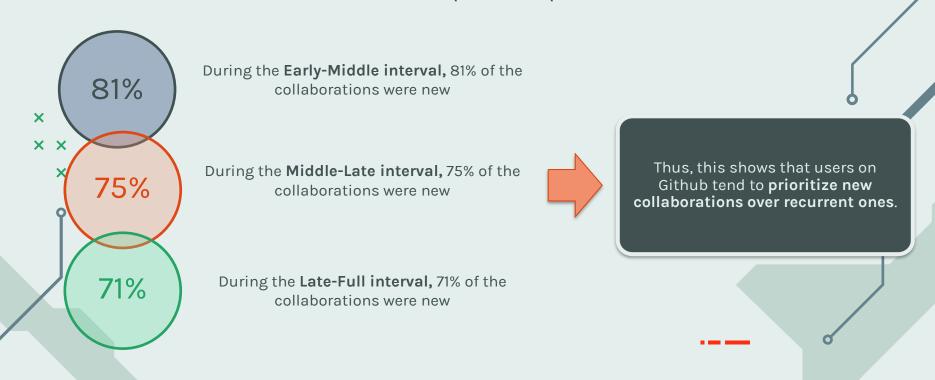
Modularity decreased over time. This was expected as the network grows in terms of nodes and edges, while the density of the graph decreased, reducing overall cohesion.

The number and size of communities divisions also varies a lot over time, indicating that communities are not stable but evolved and changed as the network grew.

Collaboration Recurrency

Lastly, we analyzed the new edges created during each interval to find out whether users tend to prefer recurrent collaborations over new collaborations.

The results showed that, **compared to the previous interval**:



Thank you for the attention!

