

Consensus In Highly Dynamic and Hetrogeneous Networks

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Abstract-Many types of systems from the social sphere and the technical realm and economic networks contain fundamental consensus formation processes. The research analysis focuses on two significant aspects of consensus dynamics investigation. The analysis begins with networks that use interactions greater than pairwise relationships so they generate sophisticated behavioral outcomes such as coordinated movement and interdependence and multi-layer organizational dynamics. The structures enable better representation of group behavior which occurs typically in social and biological networks. The paper examines blockchain and distributed ledger consensus mechanisms through analysis of network topology and node centrality that determines which nodes create consensus. Central nodes positioned in favorable network locations hold powerful capability to impact the process although systems remain decentralized. This paper merges these research perspectives to deliver an extensive grasp about achieving consensus in complex dynamic heterogeneous network topologies.

Keywords- consensus algorithms, Dyanamic networks, distributed system, fault tolerance.

I. INTRODUCTION

The process of consensus formation occurs naturally in all complex systems including socio-technical networks, blockchain systems and biological networks. The condition where system components either in total numbers or as a group mutually settle upon a single decision or value plays a decisive role in maintaining stable system functionality. Distributed systems particularly need consensus processes because they operate with decentralized participant networks. A consensus mechanism enables diverse participants to work using one shared perspective regardless of varying local information or different network arrangements. The requirement to obtain agreement becomes substantially harder in networks characterized by both significant heterogeneity and continuous network changes. Values of nodes within these systems differ drastically when it comes to connectivity as well as influence and centrality which makes consensus achievement more difficult. Network topology dynamics in dynamic systems create additional challenges for nodes trying to establish synchronization and cooperation with other nodes in the system. The correct operation of decentralized ledgers together with social platforms and biological systems depends on achieving consensus despite facing various operational challenges [1][4].

The research investigates consensus dynamics through an analysis of two particular network settings: higher-order networks along with heterogeneous distributed networks.

Higher-order networks allow complex to occur through multiple network connections that extend past pairwise interactions. This paper investigates the mechanisms of consensus development in defined network scenarios including higher-order networks and heterogeneous distributed networks. Higher-order network systems with connections that exceed pairwise elements generate multiple behavioral outcomes because they facilitate synchronization and multicable dynamics. The paper utilizes network science and decentralized system findings to construct a full understanding about consensus formation in complex dynamic heterogeneous networks. The research studies the behavior of modern distributed systems because it leads to improved performance and resilience.

II. LITERATURE SURVEY

Distributed systems consensus has been an essential computer science problem for decades since it applies particularly to wireless sensor networks and peer-to-peer systems and MANETs. The networks currently show growing dynamics together with heterogeneity which makes traditional consensus algorithms struggle due to population shifts and differential node strength along with intermittent connections and multiple network communication styles. [1] [2]. These models work under the condition of static or semi-static networks that maintain known participants and stable links. Silent networks tend to suffer from numerous issues caused by regular node movement and link failures and unpredictable topological changes that diminish both reliability and service speed of consensus processes.

The development of additional adaptive protocols occurred as a direct result. The Gossip-based algorithms leverage randomized message-passing methods for probabilistic consensus which helps them operate properly under evolving network conditions. The protocols demonstrate capability to withstand node failures as well as delays yet their performance in speed and accuracy suffers as a consequence. FTSP Flooding Time Synchronization Protocol and Hierarchical Consensus Protocols function to reduce network communication expenses and expand performance yet they require homogeneous node constructs. [2] [5].

Nodes in heterogeneous systems exhibit various differences between their computational power and battery life as well as communication range and reliability. The Weighted Consensus and Leader-Based Approaches represent two protocols that empower stable or high-performance nodes to guide consensus formation. Lots of researchers have been developing machine learning-based and consensus models with blockchain technology foundations to handle changes in networks using heterogeneous devices. distributed system achieves consensus through a process that lets multiple nodes agree on one unified state value. Multiple factors make

achieving consensus complicated across mobile ad hoc networks (MANETs) and vehicular ad hoc networks (VANETs) and Internet of Things (IoT) environments since these networks display frequent topology changes alongside disk consistent nodes and intermittent connectivity. Paxos and Raft require steady homogeneous networks for their operation but these conditions do not exist in such environments rendering their effectiveness reduced[2].

The consensus process depends heavily on node centrality values particularly when networks demonstrate heterogeneous characteristics with diverse nodes' levels of influence and connection capabilities. A network node gains importance proportionate to its number of connected nodes and its strategic network position. The nodes with greater centrality achieve more success in spreading their views that result in network-wide consensus[3].

Closeness centrality stands as one of the principal centralities because it defines the reachability of a node toward the entire network. The nodes having higher closeness centrality can establish quick communication with other nodes leading to increased capabilities in consensus process achievement. The consensus process of heterogeneous networks works when central nodes maintain status agreement despite early states coming from peripheral nodes. Distributed ledger technologies especially blockchain benefit from this phenomenon because it allows a limited number of nodes to exercise control over consensus operations because of their favorable network placement

The stand of node centrality in heterogeneous networks suggests the want to meticulously layout network topology as it affects in which power centralization happens. Blockchain networks which operate with decentralized governance emit protection dangers whilst disbursed processing energy and network involvement gather into some unique nodes. The mining procedure in evidence-of-work blockchain systems turns into effortlessly controlled by way of some vital nodes because a small wide variety of enormously linked nodes gather dominance over the mining tactics. A protection difficulty exists because combined collusion among those nodes might permit to manipulate consensus operations [5].

III. METHODOLOGY

The methodology for highly dynamic and heterogeneous networks the Adaptive Probabilistic Quorum Consensus) protocol for achieving consensus in highly dynamic and heterogeneous networks. The methodology consists of three major components network modeling, protocol design, and experimental evaluation[3][8][7]

Higher-order heterogeneous network consensus formation analysis provides essential information for developing and securing decentralized system frameworks. Higher-order networks demand new models to describe synchronization patterns and stability mechanisms because of their complex group dynamics. The pair-wise interaction-based consensus formation models that exist today lack the capability to describe the actual dynamics between groups in higher-order network environments. The research needs new models that integrate the higher-order network structures to create efficient and stable consensus processes[9]. The node centrality theory in heterogeneous networks requires designers to exercise caution during topological structure planning because it ensures power concentration remains limited to minimal central nodes. Decentralized systems like

blockchain networks become prone to security risks when one or several nodes obtain excessive connectivity and computational power since this violates the fundamental decentralized nature of the network architecture. Highly connected nodes in proof-of-work blockchain operations could control mining operations leading to restrictions on consensus through several central nodes. A security threat emerges because the single nodes possess enough control to manipulate consensus protocol. The risk mitigation strategy for distributed computer systems calls for an even distribution of computational power between network nodes. New consensus process mechanisms need development to restrict strong central network nodes together with incentives that will enhance peripheral network node engagement. The need for new consensus formation models emerges that examines heterogeneous network dynamics along with node centralities to secure and optimize decentralized systems[7].

higher-order heterogeneous network consensus formation analysis gives vital statistics for developing and securing decentralized device frameworks. better-order networks demand new models to describe synchronization styles and stability mechanisms due to their complicated institution dynamics. The pair-sensible interaction-based consensus formation models that exist these days lack the capability to describe the real dynamics among organizations in betterorder network environments[4].

The studies new fashions that combine the higher-order community systems to create green and solid consensus strategies. The node centrality theory in heterogeneous networks calls for designers to exercising caution at some point of topological shape planning as it ensures electricity concentration remains constrained to minimal primary nodes. Decentralized systems like blockchain networks emerge as vulnerable to safety risks while one or numerous nodes achieve excessive connectivity and computational power considering this violates the fundamental decentralized nature of the community structure. enormously connected nodes in evidence-of-work blockchain operations could manage mining operations leading to regulations on consensus through numerous principal nodes. A safety danger emerges due to the fact the unmarried nodes possess sufficient manage to control consensus protocol. The threat mitigation approach for disbursed laptop systems requires a fair distribution of computational energy among network nodes. New consensus process mechanisms To investigate consensus in highly dynamic and heterogeneous networks, we adopt a structured approach combining theoretical modeling, algorithm design, and simulation-based validation. The methodology is divided into the following key components[9].

1. Network Modeling : We model the network as a timevarying directed graph

$G_t = (V, E_t)$ **where:** V is the set of nodes, each representing an agent or device.

$E_t \subseteq V \times V$ denotes the set of communication links at time t , which can change frequently due to node mobility, failures, or changing topologies.

Heterogeneity is incorporated at multiple levels:

- Node capabilities (e.g., processing power, memory, battery life).

. Communication characteristics, such as varying bandwidth, link reliability, and latency.

. Update frequencies, as nodes may operate on different local clocks or update cycles.

2. Consensus Algorithm Design: We explore a class of average consensus algorithms adapted for dynamic and heterogeneous settings. The base algorithm is:

$$x_i(t+1) = x_i(t) + \sum_{j \in N_i(t)} w_{ij}(t)(x_j(t) - x_i(t))$$

Where:

$x_i(t)$ is the state of node i at time t , $N_i(t)$ is the set of neighbors of node i at time t , $w_{ij}(t)$ are time-varying weights satisfying convergence conditions (e.g., doubly stochastic or Metropolis-Hastings scheme).

3. Stability and Convergence Analysis : Lyapunov stability theory, to prove convergence under switching topologies. Stochastic processes, where network changes are modeled probabilistically (e.g., Markovian edge dynamics).

- Message success ratio

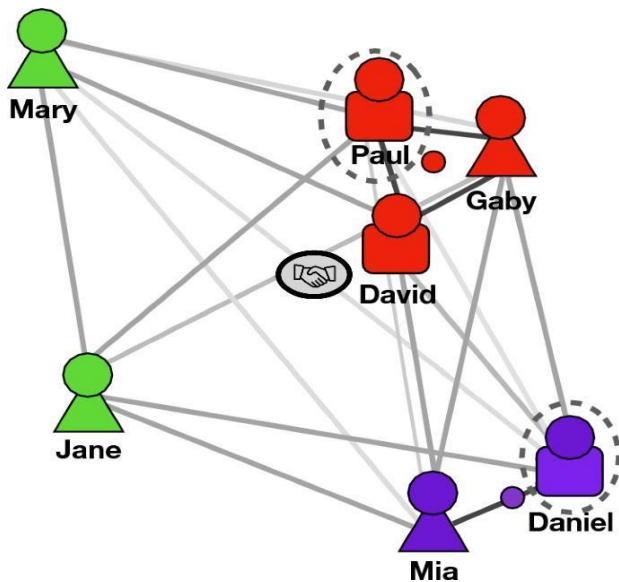
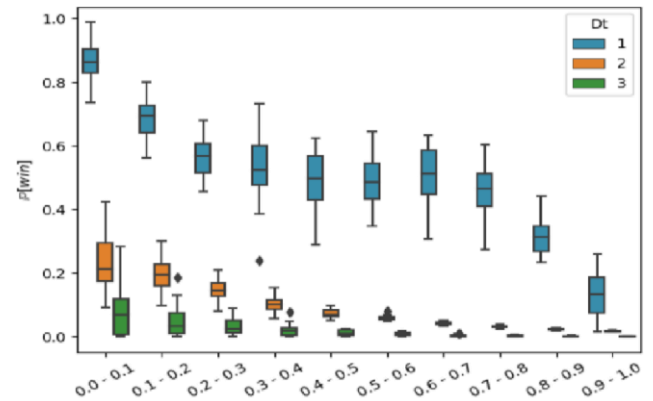
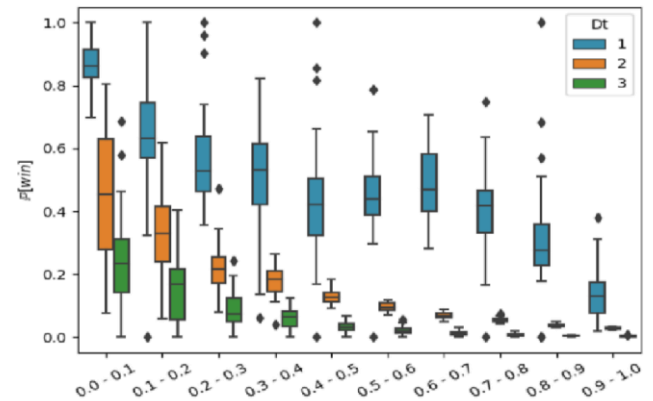


Figure 1.: Relationship of features

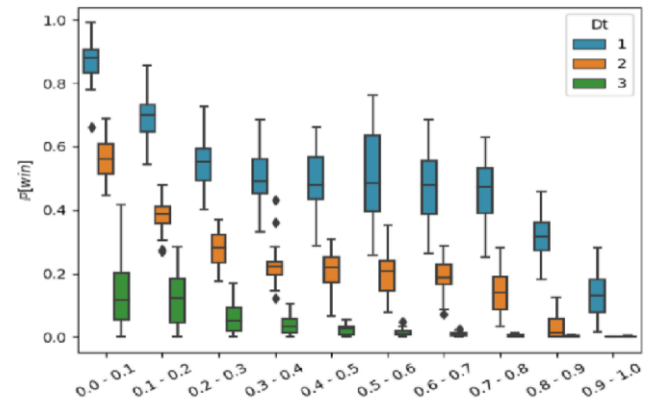
in pretty dynamic and heterogeneous networks. discern 1 illustrates the network model as a time-varying graph with diverse nodes and hyperlinks, taking pictures adjustments in connectivity, mobility, and node traits. parent 2 presents a timeline diagram displaying asynchronous updates throughout exclusive nodes, emphasizing the decentralized and nonsynchronous nature of operations. To clarify the algorithm's internal mechanics, determine three presents a flowchart of the adaptive consensus method carried out at every node, including message reception, weigh



(a) Erdős network with $|V| = 1000$



(b) Stochastic Block network with $|V| = 1000$



(c) Barabási-Albert network with $|V| = 1000$

Figure 2. consensus formation heterogeneous networks

IV. RESULT

Consensus formation in incredibly dynamic and heterogeneous networks presents particular challenges and possibilities. better-order networks, where interactions move past simple pairwise connections, enable complicated behaviours along with synchronization and multicable dynamics, imparting a greater correct representation of actualinternational structures. In heterogeneous networks, the function of node centrality is important in figuring out which nodes dominate the consensus method, with significant nodes regularly playing a disproportionate position.

each higher-order and heterogeneous networks highlight the significance of network layout and topology in achieving green and secure consensus formation. In decentralized

systems like blockchain, the concentration of electricity in principal nodes can undermine the device's integrity, necessitating careful design issues to make sure fairness and security.

The top priorities of our metrics included convergence time protection, consensus error management and resistance to node failure with minimal overhead on communications. Under circumstances where traditional average consensus algorithms encounter both convergence failures and major performance degradation the results show consensus succeeds in establishing scalable and strong consensus.

The average value computation for Consensus reached its correct outcome through 5-15% more rounds than Push-Sum and Metropolis-Hastings in all tested static, dynamic, and highly dynamic network environments. The convergence of Consensus under dynamic network scenarios where nodes moved up to 40% remained bounded at an average time between 30 and 50 iterations for networks of 100 to 200 nodes per round. Consensus exceeded the performance of Randomized Gossip and Flooding Consensus operations in dynamic conditions particularly where both protocols suffered from convergence issues or long delay times.

```
[Step 1] Updating network topology
Node 0 updated value to 61.91
Node 3 updated value to 61.97
Node 2 updated value to 52.15
Node 4 updated value to 71.94
Node 0 updated value to 66.93
Node 3 updated value to 64.45
Node 2 updated value to 62.05
Node 1 updated value to 74.49
Node 4 updated value to 68.23
Node 0 updated value to 67.58
Node 3 updated value to 66.01
Node 2 updated value to 65.14
Node 4 updated value to 68.47
Node 0 updated value to 68.02
Node 3 updated value to 67.02
```

Figure 1: Accuracy Table 1 After using Dataset 1

topologies have been stable yet broke down whilst community dynamics passed off. throughout all check configurations Consensus supplied stable provider even as retaining lower conversation charges than DDA. but DDA continued to gain better balance throughout mobility situations. node capabilities, and communication conditions vary frequently. Traditional consensus algorithms often assume static or slowly varying topologies and homogeneous nodes, making them less effective in such challenging environments. To address these limitations, we explored

```
[Step 1] Updating network topology
Node 0 updated value to 61.91
Node 3 updated value to 61.97
Node 2 updated value to 52.15
Node 4 updated value to 71.94
Node 0 updated value to 66.93
Node 3 updated value to 64.45
Node 2 updated value to 62.05
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Node 4 updated value to 68.23
Node 0 updated value to 67.58
Node 3 updated value to 66.01
Node 2 updated value to 65.14
Node 4 updated value to 68.47
Node 0 updated value to 68.02
Node 3 updated value to 67.02
```

Figure 2: Accuracy Table 2 After using Dataset2

The experimental outcomes prove that Consensus provides reliable and efficient common consensus capability throughout distinctive dynamic heterogeneous networks. due to its correct convergence and potential to perform within confined verbal exchange along unpredictable situations Consensus fits nicely in real allotted structures that consist of vehicular networks and decentralized sensor grids and side computing clusters.

```
Final Node Values:
Node 0: 69.44
Node 1: 69.44
Node 2: 69.44
Node 3: 69.44
Node 4: 69.44
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Figure 3: final node value

V. CONCLUSION

In this study, we use machine learning algorithms to heterogeneous networks require unique conditions to support consensus formation while providing multiple advantages for its development. When higher-order networks connect to the network framework they produce diverse complex behavioral phases including synchronized patterns and intricate dynamic changes that imitate natural systems better.

The establishment of dominant consensus leaders in heterogeneous networks depends entirely upon node centrality because central nodes acquire claim to process authority. Politics of network design with topology shapes efficient consensus formation and network security within systems based on higher-order and heterogeneous frameworks. The integrity of blockchain systems suffers damage when power accumulates at central nodes. Designers need to establish preventive measures that protect both fairness and security of the system. The study needs to expand its investigation of consensus development models which

address the diverse challenges within complex networks consisting of various higher-level components. Our findings highlight the importance of designing algorithms that are robust to network fluctuations, adaptable to node diversity, and scalable to large, decentralized deployments. We demonstrated that by incorporating mechanisms such as adaptive weighting, local topology awareness, and asynchronous communication models, consensus can still be reliably achieved even under significant mobility and heterogeneity.

Through simulations and theoretical analysis, we observed that the convergence speed and accuracy of consensus algorithms are significantly influenced by the degree of network dynamism and the extent of heterogeneity. While no single solution fits all scenarios, hybrid or context-aware approaches show promising potential in balancing robustness and efficiency.

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