A Deep Systems Analysis of Bitcoin Working Paper

Shingai Thornton

Binghamton University

Version 0.5 – March 19, 2025

Abstract

This working paper applies Deep Systems Analysis (DSA) to Bitcoin, framing it as a complex adaptive system whose technical, economic, and social dynamics must be understood through an integrated lens. Using a three-phase methodology—system identification, environmental analysis, and recursive decomposition—we dissect Bitcoin's core subsystems (validation, mining, protocol, and development) and their interactions. The analysis illustrates how emergent properties like decentralization and censorship resistance arise from these interconnections, offering a rigorous holistic perspective beyond that provided by traditional siloed academic disciplines. With Bitcoin's growing adoption by individuals, corporations, and nation-states, this framework aims to provide a foundation for rigorous modeling, policy development, and future cryptoeconomic research. As a draft, I'm seeking feedback to refine this initial application of DSA to the Bitcoin network.

1 Introduction

The Bitcoin network consumes more energy than many nations, boasts a market capitalization of \$1.6 trillion (at publishing time), and has spawned an entire industry of blockchain technology along with a new academic field of cryptoeconomics. Yet, after fifteen years of operation, there are still fundamental questions being asked about the mysterious technology. What is it? What does it do? Can it provide any real value to society? Economists debate whether it is a purely speculative bubble or digital gold as computer scientists focus on its novel technical properties while trying to improve its scalability, security, flexibility, and privacy. Policymakers scrambling to construct new laws and regulatory frameworks argue about whether it is a bastion of financial innovation and freedom to be embraced, or a threat to financial stability that must be controlled or eradicated.

While these domain-specific perspectives are valuable and necessary, they only capture fragments of the whole story. Trying to understand Bitcoin within these disciplinary boundaries leads to tensions that are impossible to resolve when confined to the lens of a single scientific domain. Bitcoin has evolved far beyond Satoshi Nakamoto's original vision of peer-to-peer digital cash and now operates as a complex adaptive system whose behavior emerges from countless interactions between software, economic incentives, and human social dynamics. Proper understanding of Bitcoin demands a holistic analytical approach capable of bridging traditional disciplinary divides while preserving both technical rigor and systemic insight.

2 Background

2.1 Challenges in Cryptoeconomic Analysis

In 2015, Bitcoin was described in an early systematic survey of challenges facing cryptocurrencies as "a rare case where practice appears to be ahead of theory" [1]. The technology was adopted by a sizable group of early enthusiasts as real money long before mainstream economists had spent any time attempting to develop theoretical frameworks for reasoning about the complex dynamics of decentralized digital currency uncontrolled by any nation state. The authors of the survey noted that some early Bitcoin adopters dismissed the need for theoretical foundations, arguing that it demanded attention simply because 'it worked'. Meanwhile many academics and security researchers hesitated to engage with the technology because they saw the system as being dependent on "an unknown combination of socioeconomic factors hopelessly intractable to model."

Today, Bitcoin receives an abundance of attention from researchers who are curious about, excited by, and terrified of its potential impact on the global political economy. However, the academic community has yet to develop theoretical frameworks that are broad, flexible, and rigorous enough to simultaneously account for its social, economic, and technical dimensions in a formal manner. Given the increasing levels of retail, institutional, and governmental adoption of Bitcoin, this isn't just a theoretical problem—lack of understanding has practical consequences. Space Force officer and astronautical engineer Jason Lowery warns that over-reliance on purely economic frameworks creates systemic analytical bias that could impact policy decisions and even national security [2]. Lowery's novel Power Projection Theory highlights the way Bitcoin uses physical power in the form of energy to secure information in cyberspace, producing a compelling systemic perspective which frames Bitcoin as a form of "softwar", a new means of projecting real power into the virtual world. This unique framing of Bitcoin as a new form of electro-cyber power projection technology invokes theories of energy-backed money developed by Henry Ford and Nikola Tesla in the early 20th century.

Cryptoeconomic researchers who embrace the methods of complex adaptive systems and systems engineering argue that proper analysis of cryptoeconomic systems must span multiple disciplines: from systems engineering and computer science to economics, political science, and ethics [3]. A unified approach to understanding cryptoeconomic systems requires treating them as complex socioeconomic networks defined by individual autonomous actors acting at the micro level, economic policies embedded in software offering coordination at the meso level, and macro-level emergent properties arising from the interactions between these systems within the network. The structure and dynamics of blockchain networks cannot be properly understood without an explicitly interdisciplinary perspective.

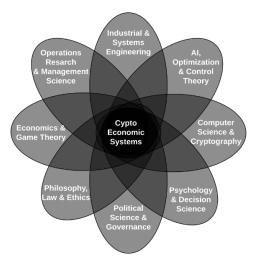


Fig. 2 Venn diagram of disciplines related to cryptoeconomic systems engineering.

Figure 1: Interdisciplinary nature of cryptoeconomics showing the overlap between computer science, economics, and cryptography. This intersection creates a complex domain requiring integration of multiple perspectives. Source: Voshmgir and Zargham [3]

The need for deep understanding is a very practical concern in light of mounting evidence which supports their assertion that blockchain networks are "mission-critical and safety-critical regulatory infrastructure for autonomous agents in untrusted economic networks." In addition to humans, there has been a recent surge in AI agents interacting with blockchains (Ante, 2025). As AI agents start to interact with humanity's economic systems in meaningful ways at scale using blockchains, it will be even more important to develop tools for reasoning systemically about blockchains because those interactions are likely to have substantial unpredictable and emergent impacts.

A key challenge facing interdisciplinary teams who seek to analyze and design these systems in a scientific manner is hampered by the lack of a common language to facilitate communication among specialists who are constrained in their capacity to coordinate due to varying disciplinary jargon. Effectively bridging disciplinary divides requires more than just collecting different viewpoints—it demands developing such a common analytical language or a "Rosetta Stone" to bridge disciplinary divides.

2.2 Systems Approaches in Cryptoeconomics

The language of systems is well-suited to support the work of interdisciplinary research teams. Systems science, though still in its formative stages and somewhat fragmented [4], provides theoretical frameworks and methodologies for describing general properties and principles found in systems across all scientific domains [5]. Born from the General System Theory movement that emerged after World War II, as scientists who experienced the power of interdisciplinary collaboration during wartime sought to formalize systems approaches during peacetime, systems science offers valuable tools for cryptoeconomic analysis. This section introduces three discipline-agnostic systems methods that have been successfully applied in cryptoeconomic research: agent-based modeling (ABM), system dynamics (SD), and network analysis [6].

2.2.1 Agent-Based Modeling

Agent-based modeling is a "bottom-up" approach to building computer simulations of complex systems. Individual agents are programmed to follow simple behavioral rules, and complex dynamics emerge from their interactions. ABMs have proven useful for reasoning about system-wide emergent properties that arise from micro-level interactions, especially in situations where controlled experimentation isn't possible, as is the case in most societal systems.

Researchers at the U.S. Argonne National Laboratory developed a generalized ABM framework for modeling blockchain systems [7]. Their model describes and simulates the interactions between market agents making transactions, miner agents selecting and verifying transaction blocks, and the decentralized ledger of transactions. Their results illustrate how ABMs can formally describe essential elements of blockchain functioning while providing insights about the relationship between mining activities and energy efficiency.

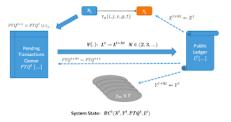


Figure 3: Elements of a generalized Blockchain system depicting the process flow for a single transaction

Figure 2: Generalized agent-based modeling framework for blockchain systems developed by Kaligotla et al. (2018). The model illustrates interactions between market agents, miners, and the distributed ledger. [7]

A team at Protocol Labs demonstrated how ABMs can support blockchain economy design through simulations of the Filecoin decentralized storage system that incorporate real-world statistical data via back-testing [8][9]. Similarly, Ethereum Foundation researchers used an ABM to study Ethereum's proof-of-stake consensus protocol after its historical switch from proof-of-work, helping them reason about how initial conditions affect consensus quality and identify specific attack vector research directions [10].

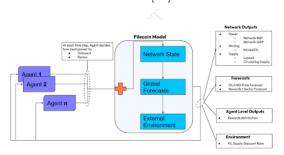


Fig. 2. Summary of the Agent Based Model of the Filecoin network, with arrows indicating direction of data flow.

Figure 3: Protocol Labs' agent-based model of the Filecoin network. This simulation incorporates real-world statistical data via back-testing to support blockchain economy design decisions.[9]

2.2.2 System Dynamics

System dynamics modeling is a "top-down" approach to building computer simulations. Rather than starting with individual agents, the modeler begins with a macrolevel conceptual model of relationships between system parts. The SD approach is useful for modeling systems described through quantities that change over time. SD models excel at exploring expected outcomes in various scenarios, evaluating proposed interventions, and anticipating potential negative side

effects.

One SD model aims to holistically describe Bitcoin's supply and demand dynamics [11]. Motivated by Bitcoin's explosive growth, the model captures various feedback loops and connects them to decisions made by different actor types (chartists, fundamentalists, and transactors) based on cost-benefit analyses. Calibrated to historical data, the model helps to explain Bitcoin's historical trajectory and informs future projections.

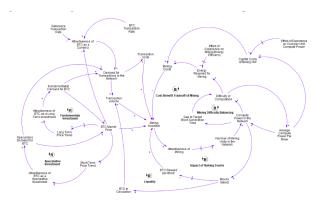


Figure 4: CLD of proposed BTC Ecosystem model

Figure 4: System dynamics model of Bitcoin's supply and demand mechanisms from Gopalakrishnan (2022). The model captures feedback loops between market actors, mining economics, and Bitcoin's monetary policy [11]

Another effort explored drivers of growth in bitcoin mining power and tested the hypothesis that miners mine blocks for profit [12]. This model anticipated that if Bitcoin prices and transaction fees remained at then-current levels, the Bitcoin network might see a decreased hashrate following the May 2020 halving event, implying the network was approaching "peak hash."

2.2.3 Network Analysis

Network analysis is an "in-between" method for examining relationship patterns or structures among system parts. Unlike ABM and SD approaches, it is not simulation-based but rather a means for formally describing complex system organization. With roots in graph theory (established by Leonhard Euler in 1735), network analysis is one of the most well-established systems approaches for studying blockchains.

A pioneering data-driven exploration of Bitcoin in 2016 used network analysis to group Bitcoin addresses into "super clusters" associated with distinct business categories [13]. After clustering, researchers created a network map showing

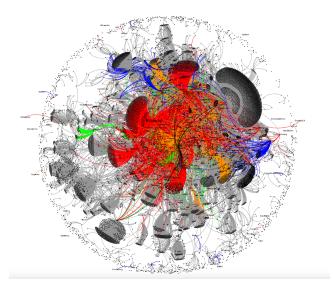


Figure 5: Network visualization of Bitcoin transaction clusters identified by Tasca et al. (2016). Different node colors represent distinct business categories, revealing patterns of transaction behaviors and ecosystem evolution. [13]

payment relationships among categories and identifying unique transaction behaviors in each cluster. The analysis revealed Bitcoin's economy had evolved through three distinct phases: an early prototype phase with early adopters, a growth stage populated by "sin" enterprises (gambling, black markets), and a maturation stage marked by progression toward "legitimate" enterprises.

More recent network analysis through 2023 also identified three distinct evolutionary periods with coherent network patterns: exploration, adaptation, and maturity [14]. This research indicated high centralization and wealth inequality within Bitcoin's network, suggesting potentially negative implications for long-term sustainability.

Similar network analyses of Ethereum have revealed user preferences for money transfers over smart contracts [15], classified different account types based on transaction statistics and network structure [16], analyzed decentralized application user behaviors [17], and identified network structures controlling evolutionary trajectories [18]. A comprehensive survey of graph approaches provides an overview of methodologies and insights while suggesting promising research directions like multi-resolution visualization and machine learning models to address data and model drift [19].

These cases demonstrate how systems-oriented methods help cryptoeconomic researchers explore emergent properties arising from micro-level behavior (ABM), causal relationships between macro-level variables (SD), and real-world organizational structures and dynamics (network analysis) in ways that integrate social, economic, and technical perspectives.

To accelerate rigorous interdisciplinary systems research in cryptoeconomics, two key issues must be addressed: accessibility and integration. These systems methods remain largely inaccessible to non-technical specialists, with results that aren't widely interpretable or usable. Additionally, there's a lack of tools and methodologies allowing seamless integration of insights across approaches. Ideally, researchers could shift between network, agent, and dynamical perspectives within a unified analytical framework.

Recent advances in applied systems science offer promising directions for starting to address some of these challenges.

3 Deep Systems Analysis Methodology

Deep Systems Analysis (DSA) is a novel systems science method that provides formal procedures for analyzing complex, adaptive, and evolvable systems while bridging historical divides in systems science [20]. It synthesizes George Klir's mathematically rigorous methods with Peter Checkland's Soft Systems Methodology, creating a unified framework that accommodates both precision and the inherent uncertainty involved when dealing with human and social complexity.

What distinguishes DSA is its intention to enforce structured and comprehensive analysis before proceeding to modeling. By prioritizing deep understanding of system structure, behaviors, and relationships, DSA creates a solid foundation that makes subsequent modeling efforts—whether using agent-based modeling, system dynamics, or network analysis—more coherent and reliable. DSA is grounded in a formal systems ontology and mathematical definition of system that provides a precise language for describing systems across domains. This nascent "system language" aims to establish a common vocabulary and syntax for representing entities, relationships, and dynamics, enabling truly interdisciplinary analysis and communication. The ontological foundation ensures that analysts can represent both technical components and social processes within the same framework.

The DSA process unfolds iteratively through three phases:

- 1. **System Identification**: Defining the system of interest, its purpose, primary outputs, and boundaries
- 2. Environmental Analysis: Mapping key environmental entities, their relationships with the system, and critical input/output flows
- 3. Recursive Decomposition: Breaking down the system into subsystems and analyzing their interactions and interfaces

Following this process produces a comprehensive knowledge base spanning raw data, graphical system maps, organizational hierarchies, and mathematical models of system dynamics. The resulting analysis can be captured and visualized using tools like BERT, an open-source software tool designed specifically for DSA [21]. This approach is particularly well-suited to Bitcoin, where essential properties like censorship resistance and decentralization emerge not from individual components in isolation, but from their interactions. The following sections will demonstrate how applying DSA to Bitcoin generates a holistic perspective while maintaining analytical rigor.

4 Deep Systems Analysis of Bitcoin

4.1 Introduction

This section applies DSA to Bitcoin through three iterative phases: system identification, environmental analysis, and recursive decomposition. Each phase builds on the insights from the previous, creating a comprehensive understanding of Bitcoin's structure and dynamics. The analysis reveals how Bitcoin functions as a complex adaptive system with emergent properties arising from subsystem interactions.

4.2 System Identification

4.2.1 Purpose

The analysis process starts with identifying a system of interest (SOI) and determining its purpose. Bitcoin, the SOI, was created to enable trustless digital cash transactions secured through decentralized consensus [22]. However, its emergent role in the Global Financial System (GFS) has evolved far beyond simply serving as p2p electronic cash—it now has multiple distinct and complementary functions. Individuals and corporations buy Bitcoin to act as a store of value, or "digital gold". Nations leverage Bitcoin as a global settlement network and neutral monetary infrastructure. Technologists view it as a general-purpose platform for building decentralized, open, and programmable digital financial infrastructure. A systems perspective allows us to account for all of these perspectives simultaneously.

4.2.2 Primary System Output

Systems are largely defined by their outputs; the things they work to produce and maintain. Bitcoin's primary product is the set of confirmed transactions contained in blocks which make up the blockchain.

The confirmed transactions can be measured directly by hosting and inspecting a local Bitcoin node, or by using a third-party service designed specifically

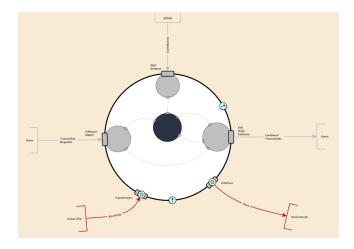


Figure 6: Visualization of Bitcoin's system structure using the BERT analysis tool. This representation shows Bitcoin's key subsystems, interfaces, and flow relationships identified through Deep Systems Analysis.

to present block chain data in a digestible manner. The flow of blocks can be measured through various parameters. $\ \ \, \ \ \,$

Table 1: Confirmed Transactions Flow Parameters

| Parameters | Units of Measurement |
|-------------------------|----------------------|
| Blocksize | Megabyte (MB) |
| Transactions per block | Transaction Count |
| Transaction Throughput | Seconds |
| Total Value Transferred | BTC |

4.3 Environmental Analysis

The next step is to identify key entities in the environment that use the system's primary product output. Understanding the system's purpose requires understanding what sort of agents receive its output and for what reasons. In the case of Bitcoin, there are four well-defined types of users who require confirmed transactions for different reasons. We can identify them based on their unique patterns of activity using the methods of network science [13].

1. P2P Transactors: Focused on using Bitcoin as originally intended; to transfer digital value without relying on centralized intermediaries. These users primarily value Bitcoin's censorship resistance and permissionless nature.

- Investors: Primarily engaged in acquiring and storing Bitcoin to hold for the long term, treating it as a store of value and potential hedge against inflation. Their blockspace usage primarily consists of large value transfers to secure storage.
- 3. Traders: Mainly interested in exploiting fluctuations in the relative price of BTC to other currencies to create a profit. They generate consistent demand for blockspace through frequent transfers between exchanges and trading venues.
- 4. Developers: Use blockspace to store code or serve as inputs into separate blockchain systems, building unique applications and protocols. This includes time chain stamping and cross-chain bridge operations.

One key characteristic of the users operating in Bitcoin's immediate environment is that they play a dual role. They simultaneously act as sinks that receive confirmed transactions and sources responsible for initiating transaction requests. This creates a significant overlap between the set of users who demand confirmed transactions and those who provide the key input of transaction requests. Transaction requests can be measured through a few parameters.

Table 2: Transaction Requests Flow Parameters

| Parameter | Units of Measurement |
|--------------------------|----------------------|
| Request acceptance rate | Percentage |
| Total pending value | BTC |
| Average transaction size | Bytes |

Another key input comes from contributors—a globally distributed group of developers, researchers, and community members interested in improving how Bitcoin functions. These contributors aggregate primarily on Github where they coordinate improvements through several measurable channels:

- 1. Code contributions: Direct improvements to Bitcoin's implementation through pull requests and code reviews in the reference client repository.
- 2. Testing and verification: Systematic testing of proposed changes, bug reporting, and security auditing.
- 3. Documentation: Technical specifications, implementation guides, and user updates.
- 4. Bitcoin Improvement Proposals (BIPs): Formal protocol enhancement proposals that follow standardized processes for review and potential adoption.

The contribution flow is essential to Bitcoin's ability to evolve while maintaining stability. The process is deliberately conservative, with multiple stages

of peer review and testing before changes are accepted. This creates a measurable pipeline of improvements where contribution quality and consensus can be tracked through metrics like pull request acceptance rates, review participation, and testing coverage.

Table 3: Contributions Flow Parameters

| Parameter | Units of Measurement |
|---------------------|-------------------------------|
| Pull Request Rate | PRs/Month |
| Code Merge Rate | # of Accepted pull requests |
| BIP Submission Rate | Proposals/month |
| Active Developers | Unique contributors per month |

Contributors interact with the system through well-defined interfaces that enforce programming standards, documentation requirements, and testing protocols. This structured approach allows the decentralized development process to be monitored and measured while preserving Bitcoin's core properties.

In order to produce confirmed transactions, Bitcoin needs a constant input of electricity. Its security model depends on a globally distributed network of computers solving increasingly computational intensive puzzles, requiring a significant flow of electricity to operate. While precise overall electricity usage and internal energy flows between subsystems are complex and difficult to track precisely, aggregate measurements enable analysis of Bitcoin's energy-security tradeoffs and broader environmental impact. They also provide essential metrics for stakeholders interested in projecting future power requirements and evaluating network sustainability. Bitcoin's energy consumption can be measured and analyzed across several key dimensions using tools such as The University of Cambridge's Bitcoin Electricity Consumption Index [23].

Table 4: Electricity Flow Parameters

| Parameter | Units of Measurement |
|-------------------------|----------------------|
| Daily power consumption | TWh |
| Energy source mix | Percentage |
| Geographic distribution | Percentage |
| Energy efficiency | J/hash |

4.3.1 Boundary Definition

Boundary definition can be one of the most difficult aspects of systems analysis, especially when dealing with complex systems that blend human and technological aspects. How do we delineate what is inside Bitcoin and what is outside? What, exactly, is Bitcoin? The DSA methodology guides analysts to define boundaries by specifying what subsystems are responsible for managing interactions with the environment. Since Bitcoin has been defined as a system that

produces blocks of confirmed transactions, the interfaces are subsystems that manage flows between internal Bitcoin processes and external entities. Five key interfaces manage Bitcoin's boundary interactions, handling three input flows and two output flows:

Input Interfaces:

- Software Wallet: Manages incoming transaction requests from users, validating format and propagating valid requests to internal processing systems.
- 2. Web Browser: Controls how external contributions of code, documentation, and protocol improvements enter the development process.
- 3. Transformer: Converts external electrical power into usable energy for internal computational processes.

Output Interfaces:

- 1. Full Node Software: Broadcasts confirmed transactions in new blocks to the network, enabling external entities to verify and consume these outputs.
- 2. Heat Dissipation: Manages the release of waste heat generated by internal computational processes into the environment.

Each interface implements specific protocols that govern how internal and external systems can interact, creating measurable boundaries while enabling essential flows of transactions, contributions, energy, and waste heat. This boundary definition allows us to clearly distinguish between Bitcoin's internal processes and its environmental interactions.

4.4 Recursive Decomposition

After identifying Bitcoin's primary output and conducting a preliminary environmental analysis, we can now examine the internal architecture and dynamics that enable the block production process. Analysis reveals four primary interacting subsystems that maintain Bitcoin's existence as a decentralized network: Validation, Mining, Protocol, and Development. These subsystems work in continuous coordination, with each playing a distinct but interconnected role in Bitcoin's operation:

- 1. Validation: Processes and propagates transactions.
- Mining: Confirms transactions into blocks through energy-intensive proofof-work.
- 3. Protocol: Provides formal software-encoded rules governing the entire system.

4. Development: Enables controlled adaptation and evolution of the network.

These subsystems maintain strict coordination through well-defined interfaces and flows. We can trace how information and work moves through the system, starting with transaction processing and following through to protocol evolution.

4.4.1 Validation

Bitcoin's architecture separates key responsibilities. The Validation subsystem is responsible for taking the transaction request inputs and feeding them into the network for processing and confirmation. It maintains network decentralization through a distributed network of full nodes that independently verify and propagate transactions and blocks. It serves as the foundation of Bitcoin's trustless consensus model by ensuring each node maintains an identical representation of the blockchain state while operating independently.

Key Interfaces

There are two key interfaces which manage the subsystem's boundary interactions and facilitate its role in the broader network. The validation subsystem implements strict protocol rules for transaction and block validation, maintaining a local mempool of unconfirmed transactions and participating in the network-wide propagation of validated blockchain data. After receiving transactions through the software wallet interface, the validation subsystem verifies their format and signatures. Valid transactions are added to the node's mempool - a waiting area for unconfirmed transactions. Through the transaction propagation interface (see Table 7 in Appendix A), these transactions are broadcast to other nodes and, crucially, to the mining subsystem where they become candidates for inclusion in new blocks.

Internal Mechanics

The validation subsystem implements strict protocol rules for transaction and block validation, maintaining a local mempool of unconfirmed transactions and participating in the network-wide propagation of validated blockchain data. After receiving transactions through the software wallet interface, the validation subsystem verifies their format and signatures. Valid transactions are added to the node's mempool - a waiting area for unconfirmed transactions. Through the transaction propagation interface, these transactions are broadcast to other nodes and, crucially, to the mining subsystem where they become candidates for inclusion in new blocks.

Interaction with Mining

Once validated, carefully defined transactions flow into the mining subsystem in a process managed by interfaces following strict protocols on both ends. This flow (detailed in Table 8 in Appendix A) demonstrates a key principle of Bitcoin's architecture—the separation of validation from consensus determination. While the validation subsystem ensures transactions meet all protocol rules, it's the mining subsystem that determines their ultimate ordering and inclusion in the blockchain.

4.4.2 Mining

Purpose and Components

The mining subsystem is responsible for taking validated transactions as input and feeding them into the complex confirmation process which produces confirmed transactions in blocks as output. It secures the Bitcoin network through proof-of-work computation, while also driving the selection of transactions and enforcing monetary policy. Mining connects to the system's broader environment through the energy interface, consuming significant electrical power to perform proof-of-work calculations.

Key Interfaces

Three critical interfaces manage its boundary interactions (see Table 9 in Appendix A for details):

- 1. **Mempool Transaction Selection**: Selects and prioritizes unconfirmed transactions based on fee rates and block constraints
- 2. **Protocol Rule Reception**: Receives and implements consensus rules and blockchain state information
 - 3. Block Dissemination: Propagates newly mined blocks to the network

Internal Mechanics

The mining subsystem first interacts with the validated transaction pool through its mempool transaction Selection interface. This interface actively monitors the mempool, selecting and prioritizing transactions based on fee rates and block size constraints. To properly construct valid blocks, the mining subsystem must adhere to strict consensus rules received through its protocol rule reception interface. These rules govern critical parameters including:

- 1. Current block reward calculation
- 2. Difficulty target requirements
- 3. Block size and weight limits
- 4. Transaction ordering requirements

Once a valid proof-of-work solution is found, the block dissemination interface handles propagating the newly mined block to the network. This initiates a cascade of validation and synchronization across the network. This processing cycle represents Bitcoin's core operational loop.

Core Operational Loop

This processing cycle represents Bitcoin's core operational loop:

- 1. Users submit unconfirmed transactions through the validation subsystem.
- 2. The protocol subsystem confirms their validity against current rules.
- 3. The mining subsystem selects and orders transactions based on Protocol's parameters.
- 4. Once a valid block is found, the protocol subsystem validates it and updates the network.
- 5. The validation subsystem propagates these changes across the network.

This cycle demonstrates how Bitcoin achieves decentralized consensus through the coordinated but independent actions of its subsystems.

Interaction with Protocol

The interaction between mining and protocol subsystems (detailed in Table 10 in Appendix A) represents a critical feedback loop in Bitcoin's operation.

4.4.3 Protocol Subsystem

Purpose and Components

The protocol subsystem orchestrates inputs and outputs from all internal subsystems and ensures that the decentralized subsystems are all operating according to the same set of code-defined rules. It serves as Bitcoin's fundamental rule engine, enabling decentralized consensus without central coordination. Unlike the other subsystems, protocol operates purely internally, coordinating interactions between mining, validation, and development to maintain consistent network behavior through strict code-defined rules.

Key Interfaces

Analysis reveals four critical interfaces that manage aspects of protocol operation.

Table 5: Protocol Subsystem Key Interfaces

| Interface | Purpose | Key Functions |
|------------------------------|---------------------------|---|
| Protocol Version Management | Upgrade coordination | Version compatibility Security patch integration Update distribution |
| Block & Transaction Rules | Consensus rule definition | Block validation rules Transaction validity criteria |
| | | • Monetary policy parameters |
| Block Validation/Integration | Chain state management | Block verificationTransaction validationState transitions |
| Block & State Distribution | Network synchronization | Block propagationState distributionNetwork consistency |

Critical Feedback Loops

The protocol subsystem coordinates several critical feedback loops:

1. Difficulty Adjustment: Monitors block times between mining and validation to maintain consistent block production

- $\hbox{2. Block Reward Schedule: Controls monetary policy through predetermined subsidy adjustments } \\$
- 3. Consensus Rules: Updates network parameters based on observed network conditions
- 4. Upgrade Activation: Manages coordinated deployment of protocol improvements

These feedback mechanisms enable Bitcoin to self-regulate and maintain stability without central control.

Code-as-Law Implementation

The protocol subsystem is unique in that it exists as pure code—software rules written in the Bitcoin Core implementation that govern how all other subsystems must behave. These rules are replicated across every node in the network, creating an immutable framework that converts human economic behavior into deterministic computer operations.

Unlike traditional financial systems where rules might be interpreted or selectively enforced by human authorities, Bitcoin's protocol rules are explicit in code and automatically enforced by every network participant. This code-as-law principle ensures that critical parameters like monetary policy (coin issuance), transaction validity, and consensus mechanisms operate with mathematical precision rather than human discretion.

Table 6: Protocol-Validation Interaction Flow

| Flow | Description | Key Parameters |
|-------------------------|---|------------------------------------|
| Block & State Updates | Network state information and | |
| | new block announcements dis- tributed from Protocol to Vali- | Block propagation time |
| | dation subsystem. | Nodes receiving block |
| | | Block size |
| | | Network difficulty |
| Network Synchronization | State and block propagation | |
| | data from the validation network. | Node version info |
| | | • Chain state |
| | | • Block header verification status |

Interaction with Validation

This interaction maintains network-wide consistency while preserving decentralization through:

- 1. Independent verification of all state changes by validation nodes
- 2. Rapid propagation of new blocks across the network
- 3. Continuous monitoring of network health metrics
- 4. Automatic adjustment of propagation parameters

4.4.4 Development Subsystem

Purpose and Components

The development subsystem takes in proposed enhancements to Bitcoin as inputs and feeds them into a complex process of political deliberation where accepted enhancements are produced as outputs for the protocol. It enables Bitcoin's controlled evolution through a decentralized process of protocol enhancement. It interfaces with the external developer community through the contribution interface, receiving code submissions, bug reports, and improvement proposals. The key interface for this process, the Protocol Enhancement Release interface, is described in detail in Table 11 in Appendix A.

Development Process

The development process emphasizes rigorous peer review and consensusbuilding to ensure changes maintain Bitcoin's fundamental properties and network security. Core developers propose, review, and implement protocol improvements through the Bitcoin Improvement Proposal (BIP) process.

Interaction with Protocol

The relationship between protocol and development subsystems is bidirectional, creating a crucial feedback loop that governs Bitcoin's evolution (see Table 12 in Appendix A for details).

The protocol subsystem influences development through:

- 1. Backward compatibility requirements for all changes
- 2. Existing consensus rules that must be preserved
- 3. Network-wide deployment considerations
- 4. Security model constraints

5 Discussion

This paper has demonstrated how applying Deep Systems Analysis to Bitcoin reveals a holistic view of its functioning as a complex adaptive system. Rather than reducing Bitcoin to simplistic framings—whether as mere peer-to-peer currency, digital gold, or a speculative bubble—DSA enables a more neutral, comprehensive perspective that acknowledges Bitcoin's many dimensions. DSA clearly shows how Bitcoin operates at the intersection of physical, digital, and social realms. The analysis identifies Bitcoin's primary function—the production of confirmed transactions within blocks—and systematically traces the internal subsystems and environmental interactions that enable this process. By mapping key inputs (electricity, code contributions, transaction requests) through clearly defined interfaces to outputs (confirmed transactions, waste heat), we establish measurable flows that can be monitored, analyzed, and modeled. This structured decomposition reveals how Bitcoin's essential properties like decentralization and censorship resistance emerge not from any single component but from the dynamic interactions between subsystems.

Bitcoin's architecture separates concerns across four primary subsystems (Validation, Mining, Protocol, and Development), with each maintaining distinct responsibilities while coordinating through well-defined interfaces. This separation serves as a defense against centralization—no single subsystem can control the network, yet together they maintain coherent functioning through carefully designed flows and feedback mechanisms. This helps Bitcoin maintain a balance between stability and adaptability that has proven remarkably resilient over Bitcoin's fifteen-year history.

The Bitcoin protocol serves as the system's "constitution"—encoding basic rules as immutable code while providing mechanisms for coordinated evolution through the Development subsystem. The Mining subsystem converts electricity into digital security, the Validation subsystem transforms digital rules into social trust, and the Development subsystem channels social consensus into protocol evolution. These cross-domain interactions create unique feedback loops that traditional disciplinary approaches struggle to capture in isolation.

The framework established in this analysis provides a foundation for more rigorous, empirically-grounded research on Bitcoin's behavior, evolution, and potential futures. By identifying measurable flows at system boundaries and interfaces, we enable data collection that can inform both theoretical models and practical applications, from engineering updates to regulatory approaches.

6 Future Research Directions

6.1 Refined Analysis

The subjective nature of systems analysis means that my initial attempt at decomposition represents just one of many possible perspectives. Future work should engage a diverse group of domain experts—from core developers and miners to economists and regulators—to refine the system boundaries, interfaces, and flow definitions. Comparing multiple candidate models would strengthen confidence in the framework while highlighting areas where competing perspectives exist.

This paper has pursued decomposition only to the first major level of subsystems. Each identified subsystem (Validation, Mining, Protocol, and Development) could be further decomposed to reveal more detailed internal structures and dynamics. For example, exploring the internal workings of the Mining subsystem would illuminate how mining pools, individual miners, and ASIC manufacturers interact, potentially revealing additional feedback loops and emergent behaviors. Another important element of the DSA process not fully addressed in this paper involves estimating system transformation functions that describe relationships between inputs and outputs. With appropriate data collection guided by the framework established here, researchers could develop and test mathematical models of how Bitcoin transforms inputs (electricity, transaction requests, contributions) into outputs (confirmed transactions, network security).

Bitcoin offers a unique opportunity for evolutionary analysis due to its completely transparent transaction history and extensive archives of community discussions. The DSA framework could inform historical analysis of key evolutionary milestones—from the early hardening of the protocol (2010-2012) to the scaling debates (2015-2017) to recent developments in layer-2 technologies. This historical perspective would shed light on how Bitcoin's subsystems co-evolve and how governance mechanisms have developed over time.

An important next step towards facilitating these research directions involves developing the technical infrastructure to collect and organize historical and real-time network data aligned with the flows identified in this analysis. Building shared structured knowledgebases would support ongoing research while enabling more sophisticated modeling and prediction efforts.

6.2 Modeling and Simulation

The core premise of DSA is that thorough analysis prior to modeling leads to more robust and useful models. This hypothesis deserves empirical testing. The Bitcoin system decomposition presented here provides a foundation for developing integrated models that combine agent-based, system dynamics, and network approaches.

Agent-based models could simulate how individual miners, developers, and users interact within the constraints of Bitcoin's protocol rules, potentially revealing emergent behaviors not obvious from the system description alone. System dynamics models could explore feedback loops between subsystems, helping predict how changes in one area (e.g., development governance) might affect others (e.g., mining incentives). Network models could map and analyze the evolving relationships between system components, from transaction patterns to mining pool distributions.

These modeling approaches could be particularly valuable for exploring Bitcoin's emergent properties—those fundamental characteristics that don't belong to any single component but arise from system-wide interactions. Key emergent properties include decentralization, censorship resistance, trust, and value creation [24]. These properties form interconnected feedback loops; for instance, increased value drives mining participation, enhancing security and building trust, which enables further adoption. A deeper understanding of these emergent dynamics would not only improve theoretical frameworks but also inform practical decisions by protocol developers, miners, and users.

Two computational simulations were developed to illustrate and validate the idea that structured files derived from applying DSA can be used to directly inform the design of simulations [25]. An ABM was developed to capture emergent behaviors from individual actors and an SD model was used to help understand system-wide feedback loops and macro-level behavior. Future research could focus on developing a more robust pipeline from structured systems analysis to useful dynamic models and simulations.

6.3 Generalized Cryptoeconomic Systems Research

While Bitcoin served as a logical starting point for this research due to its long history and relatively simple design, the DSA approach demonstrated here could be extended to analyze other public blockchain systems. Comparative analysis of systems including Bitcoin, Ethereum, Cosmos Hub, and Solana would reveal how different architectural choices influence system behavior and emergent properties. Such comparative analysis could build on existing blockchain governance research while adding systematic depth through the DSA framework [26]. This approach could be particularly valuable for addressing open problems in Decentralized Autonomous Organizations (DAOs), where the interaction between technical, economic, and social systems poses unique challenges [27].

The theoretical frameworks underlying DSA could also be enriched through integration with other promising approaches in cryptoeconomic theory. These include the contribution systems framework [28], mathematical models based on dynamical systems and control theory [29], and set-theoretic approaches to consensus mechanisms [30]. Such integration would strengthen the mathematical

foundations of DSA while broadening its applicability across the cryptoeconomic landscape.

7 Conclusion

This paper has demonstrated how Deep Systems Analysis provides a valuable framework for understanding Bitcoin as a complex adaptive system. By decomposing Bitcoin into its constituent subsystems and mapping their interactions, we've revealed the intricate mechanisms that enable its core functions while maintaining decentralization and security. The DSA approach bridges traditional disciplinary divides, allowing technical, economic, and social perspectives to be integrated within a single analytical framework. This integration is particularly valuable in the cryptoeconomic domain, where emergent system properties arise from interdependent technical designs, economic incentives, and human behaviors.

As nation-states and major corporate institutions continue adopting Bitcoin and blockchain technologies for treasury reserves, payments infrastructure, and settlement systems, comprehensive analytical frameworks become essential for sound policy and strategy development. Similarly, as blockchain systems increasingly mediate human and AI agent interactions, the need for rigorous systems-based understanding becomes more critical. The recent surge in AI agents interacting with blockchains points to a future where these systems may serve as primary coordination mechanisms for complex economic networks involving both human and artificial intelligence.

By providing a framework that spans technical, economic, and social dimensions, DSA offers a promising approach for navigating this complex future. The Bitcoin analysis presented here represents an initial step toward a more sophisticated understanding of cryptoeconomic systems. By continuing to refine these analytical methods and applying them across diverse blockchain architectures, researchers can develop more robust theories, models, and practical applications that account for the full complexity of these revolutionary technologies.

AI Declaration

In developing this paper, I utilized generative AI tools to assist with various aspects of the research and writing process. Specifically, Claude AI and Grok were employed for literature review, data organization, text editing, structural refinement, and simulation development. The AI systems helped identify relevant sources, suggest methodological improvements, refine technical explanations, and format the document according to academic standards. All simulations and analyses referenced were designed with conceptual assistance from these AI systems, though the intellectual direction, core arguments, interpretations, and

final editorial decisions remained entirely my own.

References

- [1] Joseph Bonneau, Andrew Miller, Jeremy Clark, Arvind Narayanan, Joshua A. Kroll, and Edward W. Felten. SoK: Research Perspectives and Challenges for Bitcoin and Cryptocurrencies. In 2015 IEEE Symposium on Security and Privacy, pages 104–121, May 2015. doi: 10.1109/SP.2015.14. URL https://ieeexplore.ieee.org/document/7163021/.
- [2] Jason P. Lowery. Softwar: A Novel Theory on Power Projection and the National Strategic Significance of Bitcoin. Thesis, Massachusetts Institute of Technology, February 2023. URL https://dspace.mit.edu/handle/ 1721.1/153030.
- [3] Shermin Voshmgir and Michael Zargham. Foundations of cryptoeconomic systems. Working paper, Vienna University of Economics and Business, 2020. URL https://research.wu.ac.at/en/publications/foundations-of-cryptoeconomic-systems-6.
- [4] Andreas Hieronymi. Understanding Systems Science: A Visual and Integrative Approach. Systems Research and Behavioral Science, 30(5): 580-595, 2013. ISSN 1099-1743. doi: 10.1002/sres.2215. URL https://onlinelibrary.wiley.com/doi/full/10.1002/sres.2215.
- [5] George E. Mobus and Michael C. Kalton. Principles of Systems Science. Understanding Complex Systems. Springer, New York, NY, 2015. ISBN 978-1-4939-1919-2 978-1-4939-1920-8. doi: 10.1007/978-1-4939-1920-8.
- [6] Zachary Neal. Handbook of Applied System Science. https://www.routledge.com/Handbook-of-Applied-System-Science/Neal/p/book/9780415843348, 2016.
- [7] Chaitanya Kaligotla and Charles M Macal. A GENERALIZED AGENT BASED FRAMEWORK FOR MODELING A BLOCKCHAIN SYSTEM. In 2018 Winter Simulation Conference (WSC), pages 1001–1012, Gothenburg, Sweden, December 2018. IEEE. ISBN 978-1-5386-6572-5. doi: 10.1109/WSC.2018.8632374.
- [8] Kiran Karra, Tom Mellan, Maria Silva, Juan P. Madrigal-Cianci, Axel Cubero Cortes, and Zixuan Zhang. An Agent-Based Model Framework for Utility-Based Cryptoeconomies, July 2023. URL https://arxiv. org/abs/2307.15200.
- [9] Protocol Labs. Filecoin Economics Explorer · Streamlit. https://filcryptoecon.streamlit.app/Supply_Exploration, 2023.

- [10] Benjamin Kraner, Nicolò Vallarano, Claudio J. Tessone, and Caspar Schwarz-Schilling. Agent-Based Modelling of Ethereum Consensus, May 2023. URL https://arxiv.org/abs/2305.13906.
- [11] Vignesh Gopalakrishnan. Modeling the Trajectory of Bitcoin Using System Dynamics. Thesis, Massachusetts Institute of Technology, May 2022. URL https://dspace.mit.edu/handle/1721.1/144998.
- [12] Davide Lasi and Lukas Saul. A System Dynamics Model of Bitcoin: Mining as an Efficient Market and the Possibility of "Peak Hash", April 2020. URL https://arxiv.org/abs/2004.09212.
- [13] Paolo Tasca, Shaowen Liu, and Adam Hayes. The Evolution of the Bitcoin Economy: Extracting and Analyzing the Network of Payment Relationships, July 2016. URL https://papers.ssrn.com/sol3/papers.cfm?abstract_id=2808762.
- [14] Marco Venturini. Mapping network structures and dynamics of decentralised cryptocurrencies: The evolution of Bitcoin (2009-2023). https://arxiv.org/html/2501.11416v1, 2025.
- [15] Ting Chen, Yuxiao Zhu, Zihao Li, Jiachi Chen, Xiaoqi Li, Xiapu Luo, Xiaodong Lin, and Xiaosong Zhange. Understanding Ethereum via Graph Analysis. In *IEEE INFOCOM 2018 IEEE Conference on Computer Communications*, pages 1484–1492, April 2018. doi: 10.1109/INFOCOM.2018.8486401.
- [16] Yijun Xia, Jieli Liu, Jiatao Zheng, Jiajing Wu, and Xiaokang Su. Portraits of Typical Accounts in Ethereum Transaction Network. In Hong-Ning Dai, Xuanzhe Liu, Daniel Xiapu Luo, Jiang Xiao, and Xiangping Chen, editors, Blockchain and Trustworthy Systems, pages 44–56, Singapore, 2021. Springer. ISBN 9789811679933. doi: 10.1007/978-981-16-7993-3_4.
- [17] Tian Min and Wei Cai. Portrait of decentralized application users: An overview based on large-scale Ethereum data. *CCF Transactions on Pervasive Computing and Interaction*, 4(2):124–141, June 2022. ISSN 2524-5228. doi: 10.1007/s42486-022-00094-6.
- [18] Dan Lin, Jingjing Yang, Jialan Chen, Jiajing Wu, and Zibin Zheng. Evolution of Global Driving Factors in Ethereum Transaction Networks. In Jiajing Wu, Dan Lin, and Zibin Zheng, editors, Blockchain Transaction Data Analytics: Complex Network Approaches, pages 51–72. Springer Nature, Singapore, 2025. ISBN 978-981-9744-30-5. doi: 10.1007/978-981-97-4430-5_3.
- [19] Arijit Khan. Graph Analysis of the Ethereum Blockchain Data: A Survey of Datasets, Methods, and Future Work. In 2022 IEEE International Conference on Blockchain (Blockchain), pages 250-257, August 2022. doi: 10.1109/Blockchain55522.2022.00042. URL https://ieeexplore.ieee.org/document/9881605.

- [20] George E. Mobus. Systems Science: Theory, Analysis, Modeling, and Design. Springer International Publishing, Cham, 2022. ISBN 978-3-030-93481-1 978-3-030-93482-8. doi: 10.1007/978-3-030-93482-8.
- [21] Halcyonic Systems. halcyonic-systems/bert, 2025. URL https://github.com/halcyonic-systems/bert.
- [22] Satoshi Nakamoto. Bitcoin: A Peer-to-Peer Electronic Cash System. 2008. URL https://bitcoin.org/bitcoin.pdf.
- [23] Cambridge University. Cambridge Blockchain Network Sustainability Index. https://ccaf.io/cbnsi/cbeci, 2025.
- [24] Michael Norman. The Emergence of Trust and Value in Public Blockchain Networks. https://www.researchgate.net/publication/325552991, June 2018.
- [25] Shingai Thornton. rsthornton/btc_dsa, March 2025. URL https://github.com/rsthornton/btc_dsa. original-date: 2025-03-18T12:27:58Z.
- [26] Gabriella Laatikainen, Mengcheng Li, and Pekka Abrahamsson. A system-based view of blockchain governance. *Information and Software Technology*, 157:107149, May 2023. ISSN 09505849. doi: 10.1016/j.infsof.2023.107149.
- [27] Joshua Tan, Tara Merk, Sarah Hubbard, Eliza R. Oak, Helena Rong, Joni Pirovich, Ellie Rennie, Rolf Hoefer, Michael Zargham, Jason Potts, Chris Berg, Reuben Youngblom, Primavera De Filippi, Seth Frey, Jeff Strnad, Morshed Mannan, Kelsie Nabben, Silke Noa Elrifai, Jake Hartnell, Benjamin Mako Hill, Tobin South, Ryan L. Thomas, Jonathan Dotan, Ariana Spring, Alexia Maddox, Woojin Lim, Kevin Owocki, Ari Juels, and Dan Boneh. Open Problems in DAOs, June 2024. URL https://arxiv.org/html/2310.19201v2.
- [28] Ellie Rennie and Jason Potts. Contribution Systems, November 2024. URL https://papers.srn.com/sol3/papers.cfm?abstract_id=5018758.
- [29] Zixuan Zhang, Michael Zargham, and Victor M. Preciado. On modeling blockchain-enabled economic networks as stochastic dynamical systems. Applied Network Science, 5(1):1–24, December 2020. ISSN 2364-8228. doi: 10.1007/s41109-020-0254-9.
- [30] Mike Neuder. A set-theoretic view of Ethereum coteries. https://notes.ethereum.org/@mikeneuder/set-theoretic-ethereum, October 2023.

A Detailed Interface and Flow Specifications

This appendix provides detailed specifications for the interfaces and flows between Bitcoin's subsystems that were referenced in the main text. These specifications include comprehensive parameter listings and technical details that support the Deep Systems Analysis methodology.

Table 7: Validation Subsystem Key Interfaces

| Interface | Purpose | Key Functions |
|-------------------------|--|----------------------------|
| Transaction Propagation | Network distribution of unconfirmed transactions | |
| | uncommined transactions | • Transaction verification |
| | | • Network broadcast |
| | | • Fee evaluation |
| | | |

Table 8: Validation-Mining Interaction Flow

| Flow | Description | Key Parameters |
|----------------------|--|---|
| Mempool Transactions | Validated but unconfirmed transactions passed from the Validation subsystem to the Mining subsystem for potential block inclusion, representing the current pool of pending transactions managed through strict interface protocols. | Number of unconfirmed transactions Average transaction fee Propagation time (Seconds) |

Table 9: Mining Subsystem Key Interfaces

| Interface | Purpose | Key Functions |
|-------------------------------|------------------------------------|---|
| Mempool Transaction Selection | Block content determination | • Selects and prioritizes unconfirmed transactions based on fee rates and block constraints |
| Protocol Rule Reception | Block template creation | • Receives and implements consensus rules and blockchain state information |
| Block Dissemination | Network distribution of new blocks | Propagates newly mined blocks to the network |

Table 10: Mining-Protocol Interaction Flow

| Flow | Description | Key Parameters |
|-----------------------------|---|---|
| Protocol Rules & Parameters | Consensus rules and chain state parameters transmitted from Protocol to Mining subsystem. Guides valid block creation and reward calculation. | Current block subsidy Network difficulty Chain height Block template rules |
| Mined Blocks | Newly created blocks transmitted from Mining to Protocol subsystem for validation and network distribution. Represents the core proof-of-work output. | Block hashTransaction countBlock reward |

Table 11: Development Subsystem Key Interface

| Interface | Purpose | Key Functions |
|------------------------------|--------------------------|-----------------------------------|
| Protocol Enhancement Release | Protocol upgrade coordi- | Issues new protocol versions that |
| | nation | have achieved community con- |
| | | sensus through the Bitcoin Im- |
| | | provement Proposal (BIP) pro- |
| | | cess, manages: |
| | | BIP implementation |
| | | • Version release coordination |
| | | Upgrade deployment |
| | | Backward compatibility testing |

Table 12: Development-Protocol Interaction Flow

| Flow | Description | Key Parameters |
|----------------------|---|--|
| Protocol Constraints | Existing consensus rules and pro- tocol architecture that constrain and guide development decisions | Current protocol version Active soft forks Network compatibility requirements Security boundaries |
| Protocol Updates | Protocol upgrades and security patches flowing from Develop- ment to Protocol | Protocol version number BIP implementation status Activation height Backward compatibility requirements |