

## Week 7: Digital Assets & Market Analysis

# Learning Objectives

- ▶ Explain cryptocurrency fundamentals and blockchain technology basics
- ▶ Analyze crypto market structure, price formation, and liquidity
- ▶ Assess volatility characteristics and risk properties of digital assets
- ▶ Evaluate financial inclusion claims for cryptocurrencies critically
- ▶ Implement crypto data analysis using APIs and Python
- ▶ Discuss regulatory approaches and policy challenges

**Note on terminology:** We call these “digital assets” not “digital currencies” because, as we’ll demonstrate, they’ve failed as currencies but function as speculative assets.

# The Central Puzzle

**Bitcoin has existed for 15 years, yet remains marginal as actual currency:**

- ▶ **Virtually nobody** prices goods in Bitcoin
- ▶ **Few merchants** accept it for payment
- ▶ **Transaction volumes** tiny compared to traditional payments (Visa  $\sim 2,000\times$  more than Bitcoin)
- ▶ **Primary use:** Speculative trading, not currency

**This contradicts the original vision:**

Satoshi Nakamoto's 2008 whitepaper promised "*peer-to-peer electronic cash*" without financial institutions. The technology works: but adoption as currency has failed.

**Instead:** Bitcoin became a volatile speculative asset (20-30% monthly swings), unsuitable for currency functions

# Agenda

**Part I** : Cryptocurrency fundamentals and blockchain basics

**Part II** : Market structure, exchanges, and price formation

**Part III** : Volatility, risk characteristics, and correlations

**Part IV** : Financial inclusion claims vs evidence

**Part V** : Hands-on: Crypto data analysis and market behavior

# What is a Cryptocurrency?

## Core definition:

Digital currency using **cryptography** for security, operating on **decentralised networks** based on **blockchain** technology.

## How it differs from traditional electronic money:

- ▶ **Traditional** (bank transfer): Trusted central authority (bank) maintains ledger
- ▶ **Crypto**: Distributed network of computers maintains ledger via consensus
- ▶ **Goal**: Transfer value  $A \rightarrow B$  without trusted intermediaries

## Why this matters:

No bank can block your account, no government can shut it down (in theory)

# Cryptocurrency: The Ideal Properties

## Five key properties (as designed):

1. **Decentralisation** : No central authority controls the network
2. **Pseudonymity** : Transactions public, but users pseudonymous
3. **Immutability** : Historical transactions cannot be altered
4. **Censorship resistance** : No authority can block transactions
5. **Limited supply** : Fixed maximum supply (e.g., Bitcoin: 21M)

## The promise:

Digital cash that no government controls, no bank can freeze, and no inflation can devalue

# Reality Check: Gaps Between Ideal and Practice

## The spectrum of decentralisation:

- ▶ **Bitcoin:** Substantial decentralisation (15K+ nodes, distributed mining)
- ▶ **Altcoins:** Many effectively centralised (Ripple/XRP, Binance/BNB)
- ▶ **User practice:** Most use centralised exchanges (Coinbase, Binance)

**Result:** Intermediaries reintroduced despite technology's promise

## Other reality gaps:

- ▶ **Pseudonymity:** Blockchain analysis can trace transactions to identities
- ▶ **Immutability:** 51% attacks possible; hard forks can reverse history
- ▶ **Censorship resistance:** Miners can filter transactions; ISPs can block
- ▶ **Limited supply:** Rules can change via hard forks (social convention, not law)

## Pseudonymity vs Anonymity

### The privacy illusion:

- ▶ **What users see:** Anonymous addresses (“1A1zP1...”)
- ▶ **What blockchain reveals:** Complete transaction history, forever public
- ▶ **Blockchain forensics:** Chainalysis, Elliptic, CipherTrace track transactions

### Cybercrime realities (Cong et al. 2022):

- ▶ **33.5% of FTC fraud reports** involve cryptocurrency (\$728.8M in 2022)
- ▶ **BUT:** Organized ransomware operations successfully tracked and dismantled
- ▶ **Permanent public record** makes crypto potentially less private than cash

### Privacy coins (Monero, Zcash) enhance privacy but face:

- ▶ Regulatory pressure over money laundering concerns
- ▶ Exchange delistings
- ▶ Limited adoption

**Key insight:** Cryptocurrency is pseudonymous, not anonymous: and blockchain forensics increasingly effective

# The Platform Coordination Problem

## The chicken-and-egg dilemma:

New platforms face a critical challenge: **users won't join unless others are already there**

### Example: Launching a new messaging app

- ▶ **You alone:** App is worthless (nobody to message)
- ▶ **100 users:** Still limited value (most contacts not on platform)
- ▶ **Everyone joins:** Maximum value (can message anyone)

### The coordination failure:

- ▶ Everyone would benefit if everyone joined (social optimum)
- ▶ But individually rational to wait and see if others join first
- ▶ Result: Platform never reaches critical mass, even though it's valuable

**Traditional solution:** Subsidize early adopters (Uber's discounts, referral bonuses)

**Problem:** Expensive, temporary, attracts free-riders who leave when subsidies end

# Why Do Tokens Have Value? Tokenomics

**The puzzle:** Most tokens generate no cash flows, pay no dividends, grant no ownership rights. So why do they trade for positive prices?

**Three sources of value:**

1. **Transactional demand** : Must hold token to use platform (Cong, Li, and Wang 2021)
2. **Network effects** : More users  $\rightarrow$  higher value to each user
3. **Speculative demand** : Early buyers anticipate future adoption

**Key insight:** Tokens solve the coordination problem we just saw: and do it better than subsidies

# Token Value Mechanism: Transactional Demand

## How platforms create token value:

- ▶ **Requirement:** Users must hold tokens to access platform services
- ▶ **Feedback loop:** Better platform → more users → higher token demand → price increases
- ▶ **Differs from equity:** Value from transaction necessity, not profits

## Example: Cloud storage platform

- ▶ More users join → must buy tokens to use storage
- ▶ Token price rises → signals future value → attracts early adopters
- ▶ Early adopters profit from appreciation → validates platform

**Critical point:** Platform can lose money but have valuable tokens (if usage grows)

# How Tokens Solve the Coordination Problem

**Remember the problem:** Early adopters bear risk but don't capture the value they create for later users

**The token solution:**

- ▶ Early users buy tokens to access platform (requirement)
- ▶ As more users join → token price rises (network effects)
- ▶ Early users profit from token appreciation → they captured the value!

**The magic:** Turns early adopters into investors with skin in the game

**Research finding (Cong, Li, and Wang 2021):**

- ▶ **Tokenized platforms:** Can achieve 100% adoption (social optimum)
- ▶ **Tokenless platforms:** Reach only 40-60% potential (coordination failure)

**Why?** Token creates self-fulfilling expectations:

- ▶ “If others join, token price rises”
- ▶ “So I’ll join early to profit”
- ▶ Everyone thinks this, so everyone joins, so token price rises, so expectations

# Tokens vs Traditional Subsidies

## **Traditional platform subsidies (Uber model):**

- ▶ Require external venture capital (dilutive)
- ▶ Terminate when funding depletes
- ▶ Create free-riding (join for subsidy, leave after)

## **Token mechanism advantages:**

- ▶ No external funding needed (sell tokens to early users)
- ▶ Appreciation rewards early adopters automatically
- ▶ Users committed (token holders want platform to succeed)
- ▶ Liquid exit (sell to later adopters)

**Research finding:** Traditional subsidies require  $>100\%$  of transaction value to match token adoption: literally impossible

**The catch:** Only works with credible path to adoption. Many ICO failures = tokens for platforms without viable value propositions

# Blockchain Technology Basics

Block N-1

Block N

Block N+1

Prev Hash

Prev Hash

Prev Hash

Timestamp

Timestamp

Timestamp

Nonce

Nonce

Nonce

Txns

Txns

Txns

- Tx1

- Tx5

- Tx9

- Tx2

- Tx6

- Tx10

- Tx3

- Tx7

- Tx11

- Tx4

- Tx8

- Tx12

Hash

Hash

Hash

## Key concepts:

- ▶ **Block:** Container of transactions with metadata
- ▶ **Hash:** Cryptographic fingerprint linking blocks
- ▶ **Proof-of-Work:** Miners compete to find valid nonce

## Major Cryptocurrencies

| Cryptocurrency                 | Market Cap | Consensus                 | Key Features                                                 | Launched  |
|--------------------------------|------------|---------------------------|--------------------------------------------------------------|-----------|
| <b>Bitcoin (BTC)</b>           | ~\$500B    | Proof-of-Work             | Digital gold??, fixed supply (21M), store of value narrative | 2009      |
| <b>Ethereum (ETH)</b>          | ~\$200B    | Proof-of-Stake            | Smart contracts, DeFi ecosystem, programmable                | 2015      |
| <b>Stablecoins (USDT/USDC)</b> | ~\$150B    | Varies                    | Pegged to USD, used for trading/settlement                   | 2014/2018 |
| <b>Binance Coin (BNB)</b>      | ~\$35B     | Proof-of-Staked Authority | Exchange token, BSC network                                  | 2017      |
| <b>Others</b>                  | ~\$300B    | Varies                    | Thousands of altcoins, many with questionable                | -         |

# Bitcoin: Digital Gold or Digital Speculation?

## The narrative evolution:

- ▶ **2009-2013:** “Peer-to-peer electronic cash” (Satoshi’s whitepaper)
- ▶ **2013-present:** “Digital gold” / “store of value” (scaling limitations emerged)

## The digital gold argument:

- ▶ Fixed supply (21M cap) creates scarcity
- ▶ Easily divisible, portable, verifiable
- ▶ Censorship-resistant, no government control

## The empirical challenges:

- ▶ Gold: millennia of value; Bitcoin: 15 years
- ▶ Gold: stable purchasing power; Bitcoin: 60-80% annual volatility
- ▶ Gold: uncorrelated with equities; Bitcoin: increasingly correlated (0.3-0.4)
- ▶ Deflation spiral risk if widely adopted (fixed supply → hoarding)

**Yet Bitcoin persists:** Survived hacks, crashes, regulatory crackdowns, 80%+ drawdowns. BlackRock ETF (2024) legitimizes asset class.

# Ethereum: Programmable Money and Smart Contracts

## Beyond simple value transfer:

Ethereum (2015) = “World computer” executing arbitrary code via **smart contracts**

**Smart contracts** = Programs on blockchain that execute automatically when conditions met

## What this enables:

- ▶ **DeFi**: Lending, trading, derivatives without banks
- ▶ **NFTs**: Digital ownership certificates
- ▶ **DAOs**: Decentralised governance organisations
- ▶ **Dapps**: Applications running on blockchain

## The Merge (Sept 2022): Proof-of-Work → Proof-of-Stake

- ▶ 99% energy reduction
- ▶ Changed from inflation to potential deflation
- ▶ Increased centralisation concerns (Lido, Coinbase, Binance control majority stake)

**Ongoing challenges:** Scalability (< 15 TPS), censorship, regulation, uncertainty

# Stablecoins: The Cryptocurrency Paradox

## The ironic admission:

Cryptocurrencies supposedly free users from government fiat, **yet the most-used cryptocurrencies are pegged to the US dollar**

## What are stablecoins?

Tokens pegged 1:1 to fiat currency (usually USD), used for crypto trading/settlement

- ▶ **Tether (USDT)**: ~\$90B market cap
- ▶ **USD Coin (USDC)**: ~\$25B market cap
- ▶ Combined: More than most cryptocurrencies except Bitcoin/Ethereum

## Three types:

1. **Fiat-collateralized** (USDT, USDC): Backed by cash/treasuries, centralised issuers
2. **Crypto-collateralized** (DAI): Backed by ETH with over-collateralization
3. **Algorithmic**: Maintain peg through incentives: **most failed catastrophically**

**Terra/UST collapse (May 2022)**: \$40B algorithmic stablecoin death spiral in 48 hours

# The Altcoin Universe: Innovation or Exploitation?

**Beyond Bitcoin and Ethereum:** Thousands of alternative cryptocurrencies (“altcoins”)

**Some legitimate experiments:**

- ▶ **Cardano:** Peer-reviewed academic development
- ▶ **Solana:** High throughput (65K TPS target) at cost of centralisation
- ▶ **Monero:** Privacy-focused (fungible, untraceable)

**But massive fraud at scale:**

- ▶ **ICO boom (2017-18):** Raised billions, most projects failed or were scams
- ▶ **Pattern repeats:** IEOs, IDOs, NFT drops, play-to-earn, DeFi “food tokens”
- ▶ **Common scams:** Pump-and-dump, Ponzi tokenomics, rug pulls

**The information asymmetry problem:**

- ▶ **Founders know:** Code quality, tokenomics, true intentions
- ▶ **Retail investors see:** Marketing materials, price charts, FOMO

**Lack of investor protections:**

# Cryptocurrency Exchanges

## **Centralized exchanges (CEX):**

- ▶ Coinbase, Binance, Kraken, etc.
- ▶ Hold customer funds (custody risk)
- ▶ Order book matching
- ▶ Fiat on/off ramps
- ▶ Regulated (varying degrees)

## **Decentralized exchanges (DEX):**

- ▶ Uniswap, SushiSwap, PancakeSwap
- ▶ Users retain custody
- ▶ Automated market makers (AMMs)
- ▶ Limited fiat access
- ▶ Regulatory uncertainty

## **Key issues:**

Fragmentation, price discrepancies, arbitrage opportunities, custody risk, regulatory gaps

# Research Evidence: Arbitrage and Manipulation

## Arbitrage opportunities (Makarov and Schoar 2020):

- ▶ **Cross-border spreads 3-15%:** US-Europe (3%), Japan-US (10%), US-Korea (15-40%)
- ▶ **Within-country <1%:** Price consistency within same jurisdiction
- ▶ **“Kimchi premium”:** Korean prices peaked 40% above US (2017-18)
- ▶ Reflects genuine frictions (transfer times, capital controls, regulatory barriers)
- ▶ Spreads have narrowed over time (<0.5% now vs. 5-10% earlier)

## Manipulation evidence (Griffin and Shams 2020):

- ▶ **Tether used to systematically purchase Bitcoin** during price declines
- ▶ Prices on suspicious exchanges diverge from regulated venues
- ▶ Not all price discrepancies represent genuine arbitrage opportunities

**Implication:** Markets are not just inefficient: they're partially manipulated

## 24/7 Trading & Thin Liquidity

### Continuous trading creates unique dynamics:

- ▶ **24/7/365 markets:** No close, no circuit breakers, global access
- ▶ **Time-zone effects:** Volume peaks during US-Europe overlap (1-9pm London)
- ▶ **Weekend volatility:** Crashes can occur when traditional markets closed
- ▶ **News anytime:** Elon tweets, hacks, regulations: no pause for digestion

### Thin order books create price impact:

- ▶ **\$10M Bitcoin order:** Moves price 0.5-1%
- ▶ **\$10M Apple order:** Moves price 0.01%
- ▶ **Small altcoins:** Entire order book might be \$10K-\$100K

### Liquidity hierarchy:

Bitcoin/Ethereum (deep) > Major altcoins (moderate) > Small-caps (virtually none)

**Manipulation opportunities:** Spoofing, layering, whale moves

# Market Manipulation: Wash Trading

## The volume fabrication problem:

70-90% of reported trading volume on unregulated exchanges may be **fake**

## Why fake volume?

- ▶ Exchanges compete for users and listings
- ▶ Higher volume attracts traders and projects
- ▶ Creates false liquidity impression

## Bitwise (2019) analysis for SEC:

- ▶ Analyzed 81 exchanges
- ▶ **Only 10 had genuine volume**
- ▶ 71 showed patterns consistent with wash trading

## Implications:

- ▶ Market cap rankings misleading
- ▶ Liquidity assumptions false
- ▶ Research using volume data compromised

# Market Manipulation: Tether & Price Intervention

## **Griffin and Shams (2020) forensic evidence:**

Tether Limited issued USDT and used it to purchase Bitcoin during downward price pressure

### **The mechanism:**

1. Bitcoin price falling → negative sentiment
2. Tether issues new USDT stablecoins
3. Uses USDT to purchase Bitcoin (absorbing supply)
4. Bitcoin price stabilizes/recovers despite bad news

### **Why this works:**

- ▶ Tether is dominant stablecoin for crypto trading
- ▶ Sellers accept USDT instead of fiat
- ▶ Large purchases absorb supply and support prices

### **The concerns:**

- ▶ Tether lacks transparency about USD backing (limited audits)

# Technical Trading & Market Efficiency

## Why technical analysis dominates:

No fundamental valuation anchor: what is Bitcoin “worth”?

- ▶ No cash flows to discount
- ▶ No assets backing tokens
- ▶ No financial statements
- ▶ Price becomes its own signal

## Self-fulfilling dynamics:

- ▶ Traders believe 200-day MA is support → **becomes support** (they buy there)
- ▶ Momentum works → traders chase trends → **reinforces momentum**
- ▶ Resistance levels matter → traders sell there → **creates resistance**

**Market structure creates patterns that technical analysis exploits**

## Research evidence (mixed):

- ▶ Momentum, mean reversion, predictable patterns exist
- ▶ Markets in some regions are efficient

# Bitcoin Volatility: The Numbers

## Historical volatility metrics:

- ▶ Daily volatility: 3-5% (vs. 1-1.5% for equities)
- ▶ Annual volatility: 60-80% (vs. 15-20% for S&P 500)
- ▶ Maximum drawdown: 80-85% in bear markets
- ▶ Skewness: Slightly negative (large downside moves)
- ▶ Kurtosis: High (fat tails, extreme events common)

## Implications:

- ▶ Unsuitable as currency (purchasing power instability)
- ▶ Challenges for portfolio allocation
- ▶ High margin requirements, liquidation risks
- ▶ Options pricing difficult (model assumptions violated)

# Bitcoin Price History: Extreme Volatility in Action

## Key observations:

- ▶ **Multiple 50%+ crashes** within 5 years (2020 Covid, 2022 bear market)
- ▶ **Daily moves of  $\pm 5\%$**  occur regularly (red lines on returns chart)
- ▶ **80-85% peak-to-trough drawdowns** make long-term holding psychologically difficult
- ▶ Compare to S&P 500: Largest crash since 2020 was  $\sim 35\%$  (Covid), recovered within months

# Return Distribution: Fat Tails Everywhere

## Statistical evidence:

- ▶ **Kurtosis:** `{python} f'{{kurt:.2f}}'` (normal = 3.0) → Fat tails
- ▶ **Skewness:** `{python} f'{{skew:.2f}}'` (normal = 0.0) → Slightly negative (crash risk)
- ▶  **$\pm 5\%$  daily moves:** `{python} extreme_events events in {python} total_days days ({python} f'{{100*extreme_events/total_days:.1f}}'%)`
- ▶ **Normal distribution predicts:**  $< 1\%$  probability of  $\pm 5\%$  moves
- ▶ **Reality:** `{python} f'{{100*extreme_events/total_days:.0f}}' ×` more frequent than normal distribution

**Q-Q plot interpretation:** Systematic deviation from diagonal = non-normal distribution

## Empirical Evidence: Returns and Risk Factors

**Liu and Tsyvinski (2021) comprehensive analysis (2011-2018):**

| Frequency      | Mean Return | Std Dev | Sharpe Ratio |
|----------------|-------------|---------|--------------|
| <b>Daily</b>   | 0.46%       | 5.46%   | 0.08         |
| <b>Weekly</b>  | 3.44%       | 16.50%  | 0.21         |
| <b>Monthly</b> | 20.44%      | 70.80%  | 0.29         |

**Annualized volatility: ~167%** (vs. 15-20% for stocks)

**Extreme events occur frequently:**

- ▶ **-20% daily loss:** 0.48% probability (once every 200 days, ~2x/year)
- ▶ **+20% daily gain:** 0.9% probability (almost 2x as often as crashes)
- ▶ **High kurtosis, fat tails:** Standard risk models systematically underestimate tail risk

**Key insight:** Sharpe ratios competitive with equities at monthly horizon, but extraordinary volatility and tail risk make standard portfolio theory inadequate

# Correlation Patterns: Limited Diversification

## **Within-crypto correlations:**

- ▶ Bitcoin-altcoin correlation: 0.5-0.7 (high)
- ▶ Most altcoins highly correlated with Bitcoin
- ▶ “Bitcoin dominance” drives market
- ▶ Diversification within crypto limited

## **Cross-asset correlations:**

- ▶ Bitcoin-equity: 0.2-0.4 (time-varying, increasing)
- ▶ Bitcoin-bonds:  $\sim 0.0$  (no relation)
- ▶ Bitcoin-gold:  $\sim 0.0$  to 0.1 (weak “digital gold” support)
- ▶ During stress: correlations increase (“flight to liquidity”)

**Diversification:** Limited within crypto; modest cross-asset benefits

## Correlation Visualization: Crypto as “Risk-On” Asset

### Key findings:

- ▶ **Bitcoin-Ethereum: 0.85** → Diversifying across crypto provides minimal benefit
- ▶ **Bitcoin-S&P 500: Increasing** from 0.15 (early) to 0.45 (recent) → “Digital gold” narrative weakens
- ▶ **Bitcoin-Gold: 0.10** → Very weak, contradicts “digital gold” marketing
- ▶ **Bitcoin-Bonds: -0.03** → No relationship, neither diversifier nor risk asset behavior

**Implication:** Bitcoin increasingly behaves as risk-on asset (like tech stocks), not safe haven

# Volatility Comparison: Crypto vs Traditional Assets

## Stark differences:

- ▶ **Bitcoin volatility:** `{python} f'{{vol_btc:.0f}}'%` →  $\sim 4\times$  higher than S&P 500
- ▶ **Ethereum:** `{python} f'{{vol_eth:.0f}}'%` → Even more volatile than Bitcoin
- ▶ **S&P 500:** `{python} f'{{vol_sp500:.0f}}'%` → Baseline equity volatility
- ▶ **Gold:** `{python} f'{{vol_gold:.0f}}'%` → True “safe haven” behavior
- ▶ **Bonds:** `{python} f'{{vol_bonds:.0f}}'%` → Lowest volatility, capital preservation

**Portfolio implication:** 1% Bitcoin allocation contributes more volatility than 20% equity allocation

# Modeling Crypto Volatility: Statistical Tests

**Before modeling, test for key properties:**

**1. Autocorrelation in squared returns** (Ljung-Box test) - Tests if volatility clusters (today's volatility predicts tomorrow's) -  $H_0$ : No autocorrelation |  $H_1$ : Volatility clustering exists

**2. Normality of returns** (Jarque-Bera test) - Tests if returns are Gaussian (normal distribution) -  $H_0$ : Normal distribution |  $H_1$ : Fat tails, skewness

**Typical results for Bitcoin:** - **Ljung-Box:**  $p < 0.001 \rightarrow$  Strong evidence of volatility clustering - **Jarque-Bera:**  $p < 0.001 \rightarrow$  Reject normality (fat tails)

# GARCH Models: Capturing Volatility Clustering

## Connection to Week 3 foundations:

Volatility **clusters**: high volatility today  $\rightarrow$  high volatility tomorrow. Standard deviation assumes constant volatility (wrong for crypto!).

## GARCH(1,1) model: (Week 3, §3.4)

$$\sigma_t^2 = \omega + \alpha \epsilon_{t-1}^2 + \beta \sigma_{t-1}^2$$

- ▶  $\omega$ : Long-run variance baseline
- ▶  $\alpha$ : News impact (yesterday's shock)
- ▶  $\beta$ : Persistence (yesterday's variance)
- ▶ **Persistence**:  $\alpha + \beta$  (close to 1 = highly persistent)

💡 Connection to Ch 03, §3.4: GARCH Models

GARCH models **time-varying volatility**: volatility today depends on recent shocks and past volatility. This captures volatility clustering observed in Bitcoin returns.

## Estimating GARCH(1,1) for Bitcoin

**Fit GARCH to Bitcoin returns:**

**Key insight:** GARCH captures **volatility spikes** (2018 crash, 2020 COVID, 2021 bull run)

**Persistence + 0.98:** Volatility shocks decay **very slowly** (half-life ~35 days)

## Asymmetric Volatility: GJR-GARCH

**Problem:** GARCH(1,1) treats good/bad news equally. Reality: **Bad news increases volatility more** (leverage effect)

**GJR-GARCH model:** (Week 3, §3.4.2)

$$\sigma_t^2 = \omega + \alpha \epsilon_{t-1}^2 + \gamma \epsilon_{t-1}^2 \mathbb{I}_{\epsilon_{t-1} < 0} + \beta \sigma_{t-1}^2$$

- ▶  $\gamma$ : **Asymmetry parameter** (extra impact if  $\epsilon_{t-1} < 0$ , i.e., negative shock)
- ▶ If  $\gamma > 0$ : Bad news hits harder than good news

**Key insight:** -5% Bitcoin drop increases volatility **~1.5x more** than +5% rally

! Connection to Ch 03, §3.4.2: Asymmetric GARCH

**Leverage effect:** Negative returns increase volatility more than positive returns. Why? - **Risk aversion:** Bad news triggers panic selling → more volatility - **Leverage:** Price drops → debt/equity ratio rises → perceived risk increases  
GJR-GARCH's **news impact curve** shows this asymmetry visually.

## Volatility Forecasting and Out-of-Sample Validation

**Question:** Does GARCH predict future volatility accurately?

**Test:** Rolling-window forecast + **Mincer-Zarnowitz regression** (Week 1, §0.6)

$$\text{Realized Vol}_t = \alpha + \beta \times \text{Forecast Vol}_t + \epsilon_t$$

- ▶ If  $\alpha = 0$  and  $\beta = 1$ : **Unbiased forecast**
- ▶ If  $R^2$  high: **Accurate forecast**

**Key insight:** GARCH forecasts Bitcoin volatility reasonably well ( $R^2 \sim 0.5-0.7$ ), but **underestimates** during extreme events

💡 Connection to Week 1, §0.6: Out-of-Sample Validation

**Mincer-Zarnowitz** tests forecast unbiasedness: -  $\alpha = 0, \beta = 1$ : Forecast is unbiased - High  $R^2$ : Forecast explains realized volatility well

This is **honest evaluation**: forecast on past data, test on future data (no look-ahead bias).

## Structural Breaks vs GARCH Persistence

**Key question:** Is Bitcoin's high GARCH persistence ( $+ 0.98$ ) **real** or an **artifact of regime shifts**?

**Two interpretations:**

1. **True persistence:** Volatility shocks decay slowly (GARCH is correct model)
2. **Structural breaks:** Volatility shifts between **regimes** (calm vs turbulent), GARCH mistakes this for persistence

**Test:** Compare **full-sample GARCH** vs **sub-period GARCH** models

If persistence **drops** in sub-periods  $\rightarrow$  regime shifts, not true persistence

**!** Connection to Ch 03, §3.5: Structural Breaks

**Structural breaks:** Volatility regime changes (calm  $\rightarrow$  turbulent  $\rightarrow$  calm). GARCH fitted to full sample can **confuse regime shifts with persistence**.

**Key concepts:** - **Chow test:** Test if parameters differ across periods - **Rolling volatility:** Visual regime identification - **Model comparison** (AIC/BIC): Does

## Regime Identification: Rolling Volatility

**Visual inspection:** Plot rolling 30-day volatility to identify regimes

**Key insight:** Bitcoin alternates between **calm** ( $<30\%$  vol) and **turbulent** ( $>60\%$  vol) regimes

## Sub-Period GARCH: Testing for Regime Stability

**Test:** Fit GARCH separately to first half vs second half of sample

If persistence ( + ) **similar** → true persistence

If persistence **different** → regime shifts

**Typical result:** Sub-period persistence **varies** (0.90 vs 0.96) → evidence of regime shifts



Implication: GARCH Persistence is Partly Spurious

If persistence differs across sub-periods, **full-sample GARCH overestimates** true persistence by confusing regime shifts with gradual mean reversion.

**Better models:** Markov-switching GARCH, regime-dependent volatility, threshold models

**Practical impact:** Risk models using full-sample GARCH will **overestimate** volatility persistence → wrong hedging ratios, wrong VaR estimates

# Summary: Volatility Modeling for Cryptocurrency

## What we demonstrated:

1. **Statistical tests** (Ljung-Box, Jarque-Bera): Bitcoin has volatility clustering, fat tails
2. **GARCH(1,1)** (Week 3, §3.4): Models time-varying volatility, high persistence ( $+ 0.98$ )
3. **GJR-GARCH** (Week 3, §3.4.2): Asymmetric: bad news increases vol  $1.5\times$  more than good news
4. **Volatility forecasting** (Week 1, §0.6): Mincer-Zarnowitz test shows decent accuracy ( $R^2 \sim 0.6$ )
5. **Structural breaks** (Week 3, §3.5): High persistence is **partly spurious**: regime shifts matter!

## Key lessons:

- ▶ **GARCH works** for crypto: captures clustering, fat tails, asymmetry
- ▶ **But:** Single-regime GARCH **overestimates persistence** due to regime shifts
- ▶ **Better models:** Regime-switching GARCH, threshold models, DCC (dynamic correlation)

# Systematic Return Predictors

**Beyond volatility: Factor structure (Liu, Tsyvinski, and Wu 2022)**

**Three-factor model** explains cross-sectional cryptocurrency returns:

1. **Market factor:** Overall crypto market movements
2. **Size factor:** Smaller coins outperform (liquidity premium, convenience yield trade-off)
3. **Momentum factor:** Past winners continue outperforming (attention-driven trading)

**Key finding:** 10 of 24 characteristics predict returns significantly

**Implication:** Not pure bubbles: systematic risk-return relationships exist

# The Inclusion Narrative

## **Crypto advocates claim:**

- ▶ Banking the unbanked (2B people without accounts)
- ▶ Reducing remittance costs (vs. Western Union 5-7%)
- ▶ Enabling censorship-resistant transactions
- ▶ Empowering individuals in oppressive regimes
- ▶ Financial services without permission or discrimination

## **Reality check questions:**

- ▶ Who actually uses cryptocurrency?
- ▶ For what purposes?
- ▶ In which countries and demographics?
- ▶ What evidence supports welfare benefits?
- ▶ What barriers prevent mainstream adoption?

## Crypto vs. Mobile Money: Evidence Comparison

| Dimension               | Mobile Money (M-Pesa)                      | Cryptocurrency                |
|-------------------------|--------------------------------------------|-------------------------------|
| <b>Target users</b>     | Poor, unbanked, rural                      | Unbanked (narrative)          |
| <b>Actual adoption</b>  | 90%+ Kenyan adults, previously excluded    | Wealthy, tech-savvy, male     |
| <b>Primary use</b>      | Payments, remittances, savings             | Speculation                   |
| <b>Welfare evidence</b> | 2pp poverty reduction (Suri and Jack 2016) | None (no causal studies)      |
| <b>Cost structure</b>   | ~1% transaction fees                       | \$0.01-\$50+ (variable)       |
| <b>Technology</b>       | USSD on basic phones                       | Smartphone, internet required |

### Key findings from Auer et al. (2025):

- ▶ **5M owners (2016) → 220M (2021):** Growth follows price momentum, not financial need
- ▶ **Retail pattern:** Enter when prices rise, exit when they fall (speculation)
- ▶ **Mobile money** is fundamentally different from cryptocurrencies

# Who Actually Uses Cryptocurrency?

**Survey evidence contradicts inclusion narrative:**

## **Demographics:**

- ▶ **Income:** Ownership increases with wealth (5-15% in developed countries, 1-5% in developing)
- ▶ **Education:** Correlates with tech literacy and higher education
- ▶ **Gender:** 70-80% male users
- ▶ **Age:** Concentrated 18-40 years old

## **Geographic patterns:**

- ▶ **Highest adoption:** US, EU, wealthy Asian countries (functioning banking systems)
- ▶ **Developing countries:** Lower absolute adoption despite need
- ▶ **Venezuela, Nigeria:** Some elevated interest, but stablecoins (USD access) not Bitcoin

**Profile conclusion:** “Risk-seeking tech enthusiasts” NOT “financially excluded seeking services”

## Usage Patterns: Speculation Not Utility

### Blockchain analytics reveal actual behavior:

#### Transaction volumes:

- ▶ **Bitcoin on-chain:** 200K-400K transactions/day
- ▶ **Exchange trading:** Billions in daily volume
- ▶ **Implication:** Most Bitcoin never leaves exchanges: traded, not used

#### Merchant acceptance:

- ▶ **Peak 2017-18:** BitPay processed for Microsoft, Overstock, Steam
- ▶ **Current:** Merchants dropping support (volatility, fees)
- ▶ **Result:** Minimal real-world payment utility

#### Ethereum gas fees:

- ▶ **Congestion periods:** \$50-100+ per transaction
- ▶ **Excludes poor users** by cost alone

**Stablecoins** show most utility (trading, some remittances) but users still mostly traders, not checked

# Barriers to Inclusion: Technical Complexity

## Infrastructure requirements:

- ▶ **Internet + smartphone:** Not universal in developing countries
- ▶ **M-Pesa comparison:** Works on basic phones via USSD
- ▶ **Immediate exclusion:** Hundreds of millions without required technology

## Private key management: a critical vulnerability:

- ▶ **Requirement:** Securely store 12-24 word seed phrases
- ▶ **No recovery mechanism:** Lose seed phrase = lose funds permanently
- ▶ **Contrast:** Banks and M-Pesa have account recovery
- ▶ **Single mistake = permanent loss**

## Transaction complexity:

- ▶ **Bitcoin sending:** Access wallet → generate address → verify address → wait 10-60 min → variable fees
- ▶ **M-Pesa sending:** Dial \*334# → send → phone number → amount → PIN → instant
- ▶ **Usability gap is enormous**

## Barriers to Inclusion: Knowledge Requirements

### **Cryptocurrency demands sophisticated financial literacy:**

Must understand:

- ▶ Blockchain technology concepts
- ▶ Different cryptocurrency types and use cases
- ▶ Wallet types (hot, cold, custodial, non-custodial)
- ▶ Exchange mechanisms and order types
- ▶ Volatility, risk, and portfolio implications
- ▶ Tax treatment and reporting requirements
- ▶ Security threats (phishing, malware, rug pulls, smart contract risks)

### **The systematic exclusion problem:**

- ▶ Financial literacy correlates with education and income
- ▶ **Lowest among populations crypto claims to serve**
- ▶ “Do your own research” (DYOR) effectively blames victims
- ▶ **Information asymmetry enables systematic scams**

# Barriers to Inclusion: Institutional Gatekeeping

## The on/off ramp paradox:

To use cryptocurrency, you typically need:

1. **Bank account** (to transfer fiat to exchange)
2. **KYC/AML verification** (ID, proof of address)
3. **Exchange account approval** (days to weeks wait)
4. **Minimum deposits** (\$10-50, significant for poor)

**If you're unbanked, you can't easily access cryptocurrency**

## Alternative on-ramps exist but:

- ▶ **Bitcoin ATMs**: 5-10% fees, urban concentration
- ▶ **Peer-to-peer trading**: High premiums, limited scale
- ▶ **Cash-for-crypto**: Geographic constraints

## Merchant acceptance:

- ▶ Negligible even in developed countries
- ▶ Virtually nonexistent in developing countries (outside tourism)

## Barriers to Inclusion: Economic Unsuitability

### Volatility destroys savings function:

- ▶ Monthly salary \$100 → save \$20 in Bitcoin
- ▶ Next month: might be worth \$15 or \$25
- ▶ **Poor households cannot accept this savings risk**
- ▶ M-Pesa maintains stable value in local currency

### Credit access impossible for target users:

- ▶ **DeFi lending**: Requires over-collateralization (deposit \$150 to borrow \$100)
- ▶ **Unbanked poor have no collateral** to begin with
- ▶ **Predatory crypto lending**: Worse terms than traditional options

### No recourse for errors:

- ▶ Wealthy investor loses \$1K to scam: painful but not devastating
- ▶ Poor person loses month's savings: catastrophic
- ▶ **Irreversibility + anonymity = attacker-friendly, victim-hostile**
- ▶ Banks and M-Pesa provide fraud protection and recovery

## Lab 7 Preview

### **Exercise 1: Accessing Crypto Market Data (30 min)**

- ▶ CoinGecko/CoinMarketCap APIs
- ▶ Exchange APIs (Coinbase, Binance)
- ▶ Historical price data retrieval
- ▶ Order book and volume analysis

### **Exercise 2: Return and Volatility Analysis (40 min)**

- ▶ Calculate returns across cryptocurrencies
- ▶ Measure and compare volatilities
- ▶ Identify tail risk and fat tails
- ▶ Correlation analysis

### **Exercise 3: Market Efficiency Testing (30 min)**

- ▶ Autocorrelation and momentum
- ▶ Mean reversion tests
- ▶ Arbitrage opportunity detection
- ▶ Price prediction attempts

## Summary and Key Takeaways

1. Cryptocurrencies use blockchain for decentralized transactions but face scalability and governance challenges
2. Markets are fragmented, volatile, and exhibit unique microstructure features (24/7, thin liquidity, manipulation)
3. Bitcoin volatility (~60-80% annualized) makes it unsuitable as currency and challenging as investment
4. Financial inclusion claims are not supported by evidence: adoption concentrates among wealthy speculators
5. Multiple barriers (technical, knowledge, institutional, economic) prevent mainstream adoption
6. Cryptocurrency data analysis reveals market inefficiencies and risk characteristics requiring careful management

# References

## Core readings:

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- ▶ Böhme, R., Christin, N., Edelman, B., & Moore, T. (2015). "Bitcoin: Economics, Technology, and Governance," *Journal of Economic Perspectives*
- ▶ Vives (2019) : Digital disruption in banking (Chapter 4 on crypto)
- ▶ Makarov, I., & Schoar, A. (2020). "Trading and Arbitrage in Cryptocurrency Markets," *Journal of Financial Economics*
- ▶ Sockin, M., & Xiong, W. (2023). "A Model of Cryptocurrencies," *Management Science* (forthcoming)

## Additional resources:

- ▶ CoinGecko/CoinMarketCap for market data
- ▶ Chainalysis/Elliptic for blockchain analytics reports
- ▶ BIS, IMF, World Bank reports on CBDCs and crypto financial stability
- ▶ Academic working papers on crypto market efficiency, volatility, adoption

## References

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