

# Cross-Asset Contagion in Stress Regimes

*Evidence from Bitcoin Crash Events and Equity Correlation Breakdown (2020–2026)*

## ABSTRACT

Traditional portfolio theory assumes that defensive, low-beta equities provide diversification benefits during periods of market stress. This study examines whether those benefits persist during crypto-led liquidity shocks.

Using over 2,200 trading days between 2020 and 2026, I analyze regime-dependent correlations between Bitcoin and selected equity groups. Stress regimes are defined as weekly Bitcoin drawdowns exceeding 10%, consistent with institutional market conventions for correction-level declines.

The results reveal a **structural breakdown in diversification**. During stress regimes, correlations between Bitcoin and defensive equities rise from a baseline of **0.07** to **0.42**, approaching the correlation levels observed in high-beta, sentiment-driven equities. This convergence indicates that Bitcoin functions as a **high-frequency proxy for systemic liquidity stress**, signaling periods when cross-asset correlations tighten, and traditional diversification temporarily weakens.

## PROBLEM STATEMENT

- **Theoretical Background**
  - Modern Portfolio Theory (MPT) relies on **low or stable correlations** across assets to reduce portfolio risk. Defensive equities such as Consumer Staples and Healthcare—are widely held for their perceived resilience during market downturns.
- **Motivation**
  - With the financialization of cryptocurrencies, Bitcoin has become deeply intertwined with global liquidity conditions. Its 24/7 trading structure and sensitivity to leverage make it a potential early indicator of risk-off environments.

### *Research Question*

*Do Bitcoin crash events coincide with a regime-dependent breakdown in the correlation structure between high-beta and defensive equities?*

## ASSET SELECTION RATIONALE

### Why These Stocks?

The analysis deliberately uses a **small, representative asset set** rather than a broad index universe to maintain interpretability and clarity.

#### Group A: Sentiment-Driven / High-Beta Equities

- Coinbase (COIN)
- MicroStrategy (MSTR)
- Tesla (TSLA)
- NVIDIA (NVDA)

**Rationale:**

These stocks exhibit:

- High volatility and growth orientation
- Strong sensitivity to retail sentiment and liquidity conditions
- Elevated beta during speculative market phases

They serve as proxies for **risk-on equity behavior**.

**Group B: Defensive / Low-Beta Equities**

- **Johnson & Johnson (JNJ)**
- **Coca-Cola (KO)**
- **Procter & Gamble (PG)**

**Rationale:**

These companies are:

- Cash-flow stable
- Low-beta, dividend-paying
- Commonly used for capital preservation and defensive allocation

They represent assets where **diversification is explicitly expected to hold**.

**REGIME DEFINITION & THRESHOLD JUSTIFICATION****Why a 10% Bitcoin Drawdown?**

The stress regime is defined as:

**Weekly Bitcoin return  $\leq -10\%$**

This threshold aligns with **institutional market conventions**:

**Market Term Typical Definition**

Pullback      ~5% decline

**Correction      ~10% decline**

Bear Market    ~20% decline

A 10% weekly move in Bitcoin corresponds to **tail-risk behavior**, capturing extreme liquidity stress without being overly restrictive. This approach prioritizes **transparency and interpretability** over model complexity.

## METHODOLOGY

### Data

- Daily adjusted closing prices (Yahoo Finance)
- Period: **January 2020 – 2026**

### Steps

1. Compute daily log returns
2. Identify stress regimes based on Bitcoin weekly returns
3. Segment returns into **Normal** and **Crisis** regimes
4. Compute Pearson correlations between Bitcoin and each equity
5. Compare correlation structures across regimes

## RESULTS

### Key Finding: Correlation Convergence During Stress

Regime	High-Beta Equities	Defensive Equities
Normal	0.36	<b>0.07</b>
Crisis	0.45	<b>0.42</b>
Absolute Change	+0.09	<b>+0.35</b>

### Interpretation

- Defensive equities are effectively uncorrelated with Bitcoin in normal conditions.
- During Bitcoin crash regimes, defensive stocks become **moderately correlated**, nearly matching high-beta equities.
- The correlation gap between “risky” and “defensive” assets narrows from **0.29** to **0.03**, indicating a **temporary failure of diversification**.

This does **not** imply that defensive stocks become speculative.

It implies that **their diversification benefit weakens precisely when it is most needed**.

## CAUSALITY CLARIFICATION

This study **does not claim**:

- Bitcoin causes equity sell-offs
- Crypto drives institutional portfolio decisions

Instead:

- Bitcoin crashes are treated as a **symptom of broader risk-off and liquidity-tightening environments**

- In such regimes, market participants de-risk across asset classes simultaneously

## FAQ's (SELECTED)

### Why Bitcoin instead of VIX?

Bitcoin trades continuously and reacts rapidly to retail leverage and liquidity shocks, often before equity markets open. It complements—not replaces—traditional volatility indices.

### Isn't correlation increase from 0.07 misleading in percentage terms?

Yes, which is why the analysis emphasizes **absolute changes** rather than relative percentages. A +0.35 shift represents a material structural change in the correlation matrix.

### What is the practical implication?

Static correlation assumptions underestimate portfolio risk during stress regimes, potentially leading to higher drawdowns and VaR breaches.

## LIMITATIONS & EXTENSIONS

### Limitations

- Threshold-based regime definition
- Small, representative asset set
- Static Pearson correlations

### Future Extensions

- VIX-based regime validation
- Rolling correlation analysis
- Portfolio-level VaR / Expected Shortfall backtesting
- Time-varying correlation models (e.g., DCC-GARCH)

## AREAS FOR IMPROVEMENT / SUGGESTIONS:

### 1. Robustness Checks

- Consider testing alternative crisis thresholds (e.g., -8%, -12%) to ensure results are not sensitive to your chosen cutoff.
- You could also try using rolling correlations or time-varying models (e.g., DCC-GARCH) to validate the regime-based findings.

### 2. Statistical Significance

- While you show correlation changes, adding confidence intervals or p-values for correlation differences would strengthen inferential claims.

### 3. Model Extension

- The LSTM section is a nice touch, but it feels somewhat disconnected from the main correlation analysis. Consider integrating regime labels as features in the LSTM or using it to forecast correlation regimes.

#### 4. **Broader Context**

- Briefly compare your findings to existing literature on cross-asset contagion (e.g., during the 2008 crisis or COVID-19 crash). This would contextualize your contribution.

#### 5. **Limitations Section**

- In the PDF, you note limitations well. In the notebook, you could add a brief reflection on data quality, survivorship bias, or the assumption of linear correlation (Pearson).