

# **The Dynamic Correlation and Hedging Effectiveness of Bitcoin against Brent Crude Oil under the Influence of Global Economic Policy Uncertainty**

**Macroeconomics**

**Master Thesis**

**Winter semester 2024/2025**



# Executive Summary



## Objective

Examine the dynamic correlation & hedging effectiveness of Bitcoin against Brent oil, considering the impact of GEPU



## Methodology

- DCC-MIDAS-X with structural breaks | Analyze the correlation dynamics
- HE ratio to assess the risk reduction | Analyze the hedging effectiveness



## Findings

- GEPU strengthens BTC – BRT correlation across different economic regimes
- BTC's hedging role is only suitable when the GEPU is low, with predicted negative BTC – BRT correlations
- BTC's hedging effectiveness is weak with % risk reduction  $< 3\%$



# Agenda

**Introduction**

**01**

**Literature review &  
Theoretical Framework**

**02**

**Methodology**

**03**

**Empirical Findings**

**04**

**Conclusion**

**05**

## Section 01:

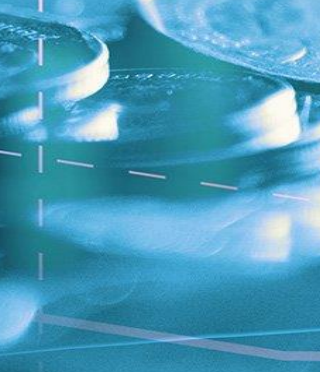
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## 1.1 Research Background

Bitcoin's role as a hedge against traditional financial assets, particularly Brent oil, has evolved significantly.

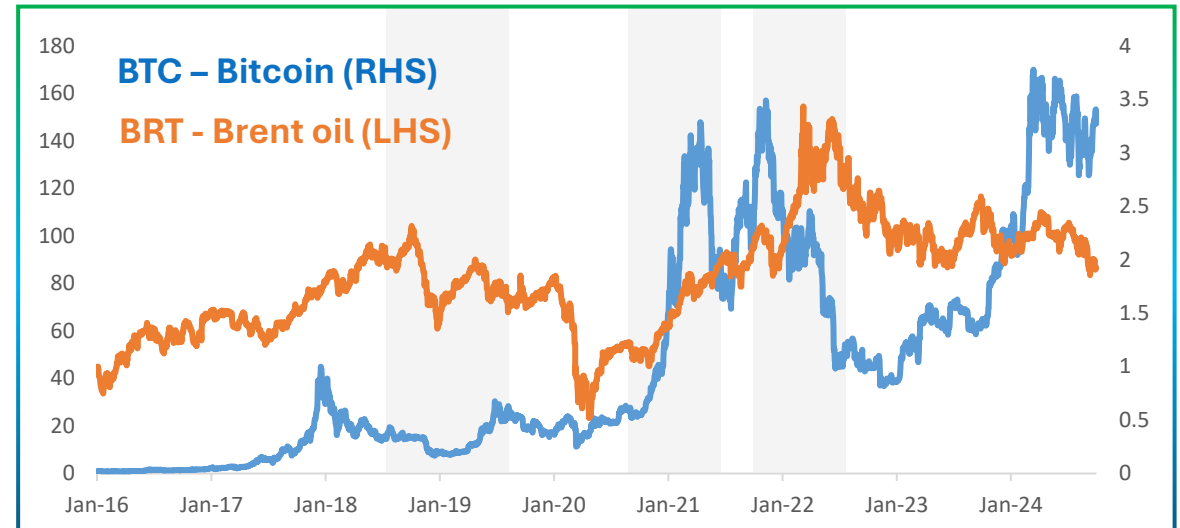
### Bitcoin as a hedge

Corr. matrix	Bitcoin	Gold	Commodity	Bond	Equity	Emerging Market
Bitcoin						
Gold	0.2					
Commodity	0.1	-0.03				
Bond	0.26	0.46	-0.12			
Equity	0.41	0.26	0.43	0.48		
Emerging Market	0.23	0.34	0.5	0.46	0.73	
AVERAGE	0.27	0.25	0.21	0.35	0.53	0.49

**BTC's low correlation with traditional assets** reinforcing its role as a safe-haven during market crises

**47% of hedge funds invest in digital assets**, up from 29% in 2023, showing institutional increasing interest and BTC integration to financial system

### Bitcoin's unstable hedging effectiveness



**BTC-BRT correlation** is weak ( $\sim 0.1$ ), and during crises, **BTC performance** mirrored speculative assets, questioning its hedging reliability

**BTC relationship with oil** varies depending on market conditions, suggesting **its role as a hedge** is not linear and may only be effective in specific economic scenarios (Wang et al. 2022)





## 1.2 Problem Statements

There are 3 critical Research Gaps in the literature about the hedging effectiveness of Bitcoin against Brent oil

01

**Lack of empirical research on Bitcoin's role as a hedge against oil price volatility**



**Most studies have focused on BTC's correlation with equity markets and financial classes**, leaving the relationship with vital commodities like oil largely unexplored

*(Bouri et al., 2017; Baur et al., 2018)*

02

**Lack of consideration of macroeconomic conditions in Bitcoin's hedging behavior**



**BTC's behavior in response to macroeconomic uncertainty**, such as changes in financial distress or political instability, is crucial but underexplored

*(Bouri et al., 2017; Shahzad et al., 2019)*

03

**Inadequate methodological approaches in analyzing Bitcoin's hedging properties**



**Popular models estimate BTC's correlation with traditional assets**, such as DCC-GARCH, Wave Transformation, Copular,...  
...fail to account for both short-term & long-term macroeconomic influences, limiting their effectiveness in capturing correlation & hedging dynamic.



## 1.3 Research Objectives

This research targets **3 Objectives** to address the corresponding Research Gaps



**01**

**Evaluates the dynamic correlation and hedging effectiveness** of Bitcoin against Brent oil in 2016-2024 period



**02**

**Assess the influence of GEPU** on dynamic correlation & hedging effectiveness of Bitcoin against Brent oil



**03**

**Use DCC-MIDAS-X with structure break model**, decomposing the correlation into daily & monthly component, under different subsample set by breakpoints



## 1.4 Research Questions & Hypotheses

### Research Question

01

**How does the dynamic correlation of Bitcoin & Brent oil evolve**, under different levels of GEPUs, across different macroeconomic regimes set by GEPUs structural break?

02

**How effective is Bitcoin as a hedge against Brent oil price** considering GEPUs and different macroeconomic regimes set by GEPUs structural break?

### Hypotheses

H1

**The dynamic correlation between Bitcoin and crude oil** is time-varying and influenced by GEPUs.

*(Wang et al. 2024, Bouri et al., 2017)*

H2

**Bitcoin serves as an effective hedge** against Brent oil price fluctuations.

*(Shahzad et al., 2019, Bouri et al. 2017)*



## Section 02:

# Literature Review & Theoretical Framework



## 2.1 Theoretical Background of Bitcoins' Hedging Characteristics

While Bitcoin may offer benefits as a hedge against Brent oil price volatility, **its effectiveness is contingent on market conditions and susceptible to speculation**

The Hedging effectiveness	
<b>Hedging definition</b>	<b>Taking a position in an asset that inversely fluctuates</b> with another asset, minimizing the risk of holding the assets. <i>(CFA institutes, 2024)</i>
<b>Hedging driver</b>	<b>The more negative correlation</b> between assets, the more effective the hedge. <i>(Modern Portfolio Theory, Markovitz, 1952)</i>
<b>Optimal hedging ratio</b>	<b>Optimal Hedging theory</b> explains how the optimal allocation of a hedging asset can minimize total risk. <i>(René M. Stulz, 1984)</i>

Bitcoin's hedging ability	
<b>Theory</b>	BTC's decentralized structure & finite supply make it a potential store of value during economic crises, thus <b>offering diversification &amp; save-heaven ability</b> . <i>(Dyhrberg 2016; Bouri et al., 2017)</i>
<b>Potential</b>	BTC was poorly correlated with Brent oil during market stress events, such as the COVID-19 pandemic & the Russia-Ukraine war, <b>making it a potential hedge</b> . <i>(Dutta et al. 2020, Ren et al. 2022)</i>
<b>Reality</b>	<b>The effectiveness of BTC hedging against oil prices is complex and unstable</b> , driven by its speculative characteristic, and susceptible to economic shock. <i>(Conlon et al., 2024)</i>

## 2.2 The dynamic correlation of Bitcoin-Oil and the role of GPEU

The dynamic correlation between Bitcoin and Brent oil is shaped by a range of macroeconomic, geopolitical, and market factors which are captured by the GEPU.

### Macroeconomic factors critically influence correlation

- **Inflationary periods** often lead to stronger BTC-BRT correlation as BTC is often viewed as a hedge against inflation, while BRT is positively related to economic growth (Smales, 2023; Fernandis, 2024).
- **Under policy uncertainty** period (e.g., the US-China Trade war 2019), oil prices can decline due to reduced production, while BTC's increasing volatility due to specific investor sentiment, complicates the assets' correlation.

### Disrupting non-economic events disrupt the BTC-BRT correlation

- **The Russian - Ukraine war** led to a sharp rise in oil prices due to supply disruptions, while BTC's price downtrend were driven more tightening interest rate, creating negative correlation (Cheah & Fry, 2022).
- **At the Covid 19 vaccination roll-out** (2021), Bitcoin – Oil's correlation surge amid brighter production outlook under quantitative easing (Wang et al. 2024).

### Financial Network Theory & Structural Break Theory, explain assets' correlation dynamics

- **Financial Network Theory** highlights the contagion effect, where shocks in one asset class can spill over into others (Allen & Gale, 2000). Cryptocurrencies increasing popularity & integration may amplify the effect.
- **Structural Break Theory** posits that extreme economic events can alter the relationship between asset classes significantly in response to the shift of market regimes (Pindyck & Rotemberg, 1990; Xiong et al., 2024).



## Section 03:

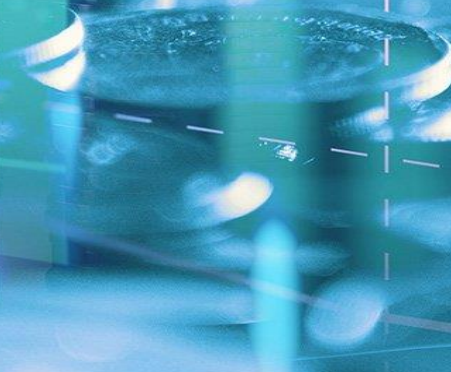
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```





## 3.1 Data

**Sample period** starts from Jan 1<sup>st</sup>, 2016 to Sep 30<sup>th</sup>, 2024

### Data Preparation

**Data  
input**

- **Daily BTC price** (CoinMarketCap)
- **Daily BRT price** (Investing.com)
- **Monthly GEPU** (PolicyUncertainty.com)

**Data  
transform**

- **BTC & BRT return** in natural logarithmic form (Stationary)
- **Raw GEPU** (Already stationary)

### Descriptive Statistic

	Ln_Return BTC	Ln_Return BRT	GEPU
Observation	3,196	3,196	105
Mean	0.002%	0.0002%	231
Median	0%	0%	228
Minimum	-50%	-28%	125
Maximum	23%	19%	432
Stdev.	4%	2%	61
Kurtosis	1344%	2636%	0.21
Skewness	-82%	-132%	0.54
<b>ADF Statistic</b>	<b>-41.5528***</b>	<b>-12.5398***</b>	<b>-3.1262***</b>



## 3.2 Methodology

### Model DCC-MIDAS-X with structural break

Assessing the dynamic correlation

1. GARCH-MIDAS-X
2. DCC-MIDAS-X
3. Structural break inclusion

*(Xiong et al. 2024)*

01

### Hedging Effectiveness (HE) ratio

Assessing the risk reduction of hedging strategy

1. Hedged vs unhedged portfolios
2. Optimal hedging ratio
3. Variance calculation

*(Kroner & Sultan 1993)*

02





## 3.2 Methodology (con't)

### Model DCC-MIDAS-X with structural break

Assessing the dynamic correlation

(Kroner & Sultan 1993)

01

<p><b>01</b></p> <p><b>GARCH-MIDAS-X</b></p>	<p><b>Decompose volatility</b> into daily and monthly components considering the impact of GEPU on monthly volatility.</p> <p>(Engel et al. (2013) and Conrad &amp; Loch (2015))</p>	<p>(1) <math>r_{i,t} = \mu_i + \sqrt{\tau_t g_{i,t}} \varepsilon_{i,t}</math></p> <p>(2) <math>g_{i,t} = (1 - \alpha_i - \beta_i) + \alpha_i \frac{(r_{i-1,t} - \mu_i)^2}{\tau_t} + \beta_i g_{i-1,t}</math></p> <p>(3) <math>\log(\tau_t) = m_i + \theta_{i,RV} \sum_{k=1}^K \varphi_k(\omega_v^{RV}) RV_{t-k} + \theta_{i,X} \sum_{k=1}^K \varphi_k(\omega_v^X) X_{t-k}</math></p>
<p><b>02</b></p> <p><b>DCC-MIDAS-X</b></p>	<p><b>Decompose correlation</b> into daily and monthly components considering the impact of GEPU on monthly correlations.</p> <p>(Colacito et al., 2011, Conrad et al. 2014)</p>	<p>(4) <math>q_{ij,t} = (1 - a - b) \overline{\rho_{ij,\tau}} + a \xi_{i,t-1} \xi_{j,t-1} + b q_{ij,t-1}</math></p> <p>(5) <math>\overline{\rho_{ij,\tau}} = \frac{\exp(2z_{ij,\tau}) - 1}{\exp(2z_{ij,\tau}) + 1}</math></p> <p>(6) <math>z_{ij,\tau} = m_c + \theta_{RC} \sum_{k=1}^{K_c} \varphi_k(\omega_c^{RC}) RC_{t-k} + \theta_X \sum_{k=1}^{K_c} \varphi_k(\omega_c^X) X_{t-k}</math></p>
<p><b>03</b></p> <p><b>Structural break inclusion</b></p>	<ul style="list-style-type: none"> <li><b>Bai Perron test</b> for structural break of GEPU</li> <li>Divide the sample into subsamples</li> <li><b>Include dummy variables</b> representing the GEPU incremental impact on the correlation.</li> </ul> <p>(Xiong et al. 2024)</p>	<p>GARCH-MIDAS-X with structure break in long term component</p> <p>(3') <math>\log(\tau_{i,t}) = m_v + (\theta_{RV} + \sum \theta_{RV,j} D_j) \sum_{k=1}^K \varphi_k(\omega_v^{RV}) RV_{t-k} + (\theta_X + \sum \theta_{X,j} D_j) \sum_{k=1}^K \varphi_k(\omega_v^X) X_{t-k}</math></p> <p>DCC-MIDAS-X with structure break in long term component</p> <p>(6') <math>z_{ij,\tau} = m_c + (\theta_{RC} + \sum \theta_{RC,j} D_j) \sum_{k=1}^{K_c} \varphi_k(\omega_c^{RC}) RC_{t-k} + (\theta_X + \sum \theta_{X,j} D_j) \sum_{k=1}^{K_c} \varphi_k(\omega_c^X) X_{t-k}</math></p>



## 3.2 Methodology (con't)

### Hedging Effectiveness (HE) ratio

Assessing the risk reduction of hedging strategy

(Kroner & Sultan 1993)

02

<b>01</b> The Hedging Effectiveness	<ul style="list-style-type: none"><li>HE ratio is utilized to evaluate the performance of hedging strategies.</li><li>It is risk reduction achieved by the hedging portfolio (using BTC) vs the unhedged portfolio (BRT only).</li></ul>	$(1) HE = \frac{var_{unhedged} - var_{hedged}}{var_{unhedged}}$
<b>02</b> Return & Variance	<ul style="list-style-type: none"><li>The return of the hedged portfolio is (2):</li><li>The variance of the hedged portfolio is (3):</li></ul>	$(2) R_{H,t} = R_{O,t} - \gamma_t R_{B,t}$ $(3) var(R_{H,t}   O_{t-1}) = var(R_{B,t}   O_{t-1}) - 2\gamma_t cov(R_{B,t}, R_{O,t}   O_{t-1}) + \gamma_t^2 var(R_{O,t}   O_{t-1})$
<b>03</b> Optimal Hedge Ratio	<ul style="list-style-type: none"><li>Minimizes the conditional variance of the hedged portfolio:</li></ul>	$(4) \gamma_t^*   O_{t-1} = \frac{cov(R_{B,t}, R_{O,t}   O_{t-1})}{var(R_{B,t}   O_{t-1})}$

## Section 04:

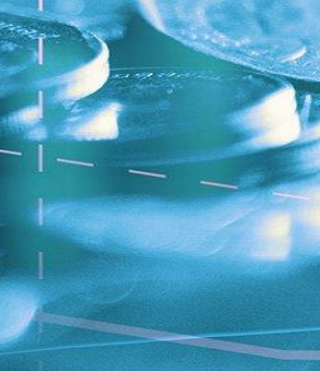
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## 4. Empirical Findings

**Structure Break Test**

**01**

**GEPU Impact on the  
Dynamic Correlation**

**02**

**Quantified BTC's Hedging  
effectiveness on BRT**

**03**



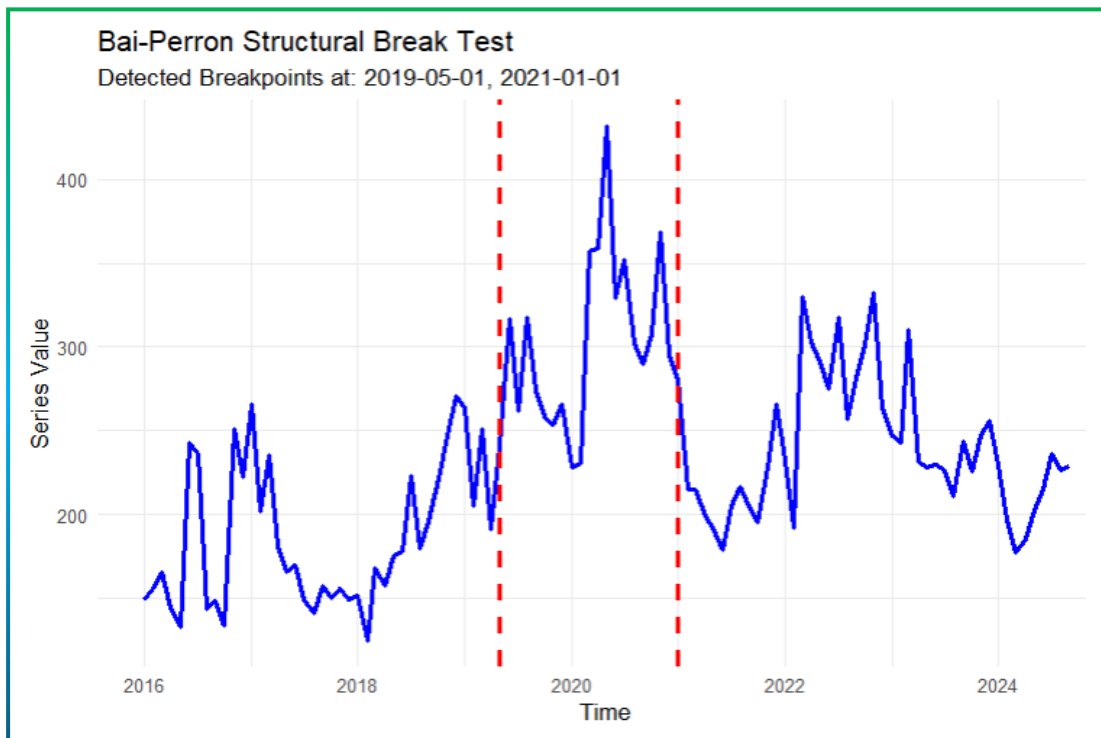


## 4.1 Empirical Findings | Structure Break Test

The structural break tests for the GEPU index identify **two key breakpoints** that correspond to major global economic events

### Test Statistic

Breakpoint 1	Breakpoint 2	SupF Test	ExpF Test	AveF Test
May-19	Jan-21	63.693***	28.393***	19.997***



### Breakpoint Analysis

#### First breakpoint | May 2019

- Coincident with the US-China trade war which led to the significant upward volatility of the GEPU afterward

#### Second breakpoint | Jan 2021

- Corresponds to the post-pandemic economic recovery, which temporarily cools down the GEPU afterward

#### Russian-Ukraine war | No breakpoint

- Its impact might be more regional, geopolitical than global economic



## 4.2 Empirical Findings | GEPU Impact on Dynamic Correlation

**GEPU has positive impact on the BTC-BRT correlation & has been strengthening over time.**

	Estimate	Std. Error	t value	Pr(> t )
$\alpha$	0.0072	0.0051	1.4354	0.1512
$\beta$	0.9757***	0.0137	71.4138	0.0000
$m$	-1.915*	1.1090	(1.7273)	0.0841
$\theta_{RC}$	-0.579	0.9458	(0.6127)	0.5401
$\theta_{RC1}$	0.1029	1.0795	0.0954	0.9240
$\theta_{RC2}$	0.2577	0.3109	0.8290	0.4071
$\omega_{RC}$	1.0014	4.0869	0.2450	0.8064
$\theta_X$	0.3704*	0.2073	1.7867	0.0740
$\theta_{X1}$	1.5183**	0.6679	2.2735	0.0230
$\theta_{X2}$	0.7486*	0.4435	1.6880	0.0914
$\omega_X$	7.7766	4.8534	1.6023	0.1091
AIC: 11,167   BIC: 11,234   Log(L): -5,572.84				

⇒ MA model:  $y_t = \theta + \alpha * \text{error}^2$

⇒ AR model:  $y_t = \theta + \beta * y_{t-1}$

⇒ ARIMA =  $y_t = \theta + \alpha * \text{error}^2 + \beta * y_{t-1} + \text{error}$

⇒ GARCH =  $\sigma_t = \theta + \alpha * \text{error}^2 + \beta * \sigma_{t-1}$

Alpha represent the impact of a random shock of volatility (residual volatility) in t-1 on t

Beta represent the impact of previous estimated volatility in t-1 on t

⇒ Persistent of volatility over time

M = unconditional volatility

Theta = sensitivity to RV and to GEPU

The positive impact of GEPU on BTC-BRT correlations suggests that **BTC may only be an effective hedge when economic uncertainty is low, and the predicted BTC-BRT correlation is negative.**





## 4.3 Empirical Findings | BTC's Hedging effectiveness on BRT

**Bitcoin's hedging ability is weak** despite slight improvements when accounting for the impact of GEPU and structural breaks

Without structural break		With structural break				
HE	1.410%	2.770%				
	Bitcoin	Gold	Commodity	Bond	Equity	Emerging Market
Bitcoin						
Gold	0.2					
Commodity	<b>0.1</b>	-0.03				
Bond	0.26	0.46	-0.12			
Equity	0.41	0.26	0.43	0.48		
Emerging Market	0.23	0.34	0.5	0.46	0.73	
AVERAGE	0.27	0.25	0.21	0.35	0.53	0.49

**The DCC-MIDAS-X model with structural breaks** marginally increases the HE ratio, but BTC ability to hedge BTC is limited, with HE values below 3% in both cases

**BTC's low realized correlation with oil (0.077)** throughout the testing period limits its effectiveness as a hedge.

**BTC's high volatility and speculative nature contrasting to oil's stable relationship with macroeconomic factors** might make it inconsistent as a hedge against BRT (Wang et al., 2024).

## Section 05:

# Final Thoughts

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## 5. Conclusion

This study investigates the dynamic correlation and hedging effectiveness of Bitcoin against Brent oil under the influence of GEPU index.

### Summary Findings

**GEPU significantly increases** the long-run correlation between BTC and BRT, with the magnitude strengthening across economic regimes.  
*(Financial Network Theory)*

**Bitcoin's role as a hedge** is contingent upon low GEPU, with negative BTC-BRT correlations.  
*(Modern Portfolio Theory - Markowitz, 1952; Wang et al. 2024)*

**Bitcoin's hedging effectiveness remains weak**, with the hedging effectiveness ratio staying below 3%.  
*(Wang et al. 2024; Allen & Gale, 2000)*

### Limitation

- Model complexity and vulnerable to overfitting
- Lack considerations about other exogenous factors impacting correlation like interest rate, inflation, market sentiment

### Recommendation

- Expand the analysis to other assets,
- Formally test alternative risk management strategies like diversification.





**Thank you for  
your attention!**