

# *Time-varying dependence between Bitcoin and green financial assets: a comparison between pre- and post-COVID-19 periods*

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Time-varying Dependence between Bitcoin and Green Financial
 Assets: A Comparison between Pre- and Post-COVID-19 Periods

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#### 7 Abstract

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This paper studies the time-varying market linkages between Bitcoin and green assets be-8 fore and during the COVID-19 pandemic through a TVP-VAR model with stochastic volatil-9 ity. Both the roles of uncertainty and environmental attention related to cryptocurrency are 10 considered when modeling market linkages, which underlying asymmetry is detected from 11 three perspectives, i.e., bidirectionality of the impact direction, time points where the unit 12 shock of the IRF analysis is imposed, and before and after the pandemic. We find that 13 the investment sheltering role of Bitcoin for green assets is enhanced and expanded after 14 the onset of the pandemic, while green assets in turn consistently act as an effective hedge 15 for Bitcoin irrespective of the pandemic. Additional analyses confirm the robustness of our 16 findings, which possess implications for not only hedging against green portfolios but also 17 seeking green shelters. 18

Keywords: Bitcoin; Green financial assets; Time-varying dependence; Asymmetry;
 COVID-19

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#### 1. INTRODUCTION

Over the past number of decades, climate change has become one of the most serious issues 22 facing the world, calling for a global agenda for green and sustainable development in the 23 future. However, over the last decade, cryptocurrencies have been developed and become 24 more and more popular which are known to have high energy consumption with adverse en-25 vironmental impacts<sup>1</sup> while attention on the future green development of the cryptocurrency 26 market remains surprisingly scant (Corbet et al., 2021). As one of the leading cryptocur-27 rencies, Bitcoin has soared in value in recent times but still faces concerns from academics 28 and investors alike on whether to include this 'dirty currency' into the investment portfolio 29 or not (Naeem and Karim, 2021). The recent headline pulled by the strategic withdrawal of 30 Tesla's acceptance of cryptocurrencies purchase due to environmental concerns has further 31 drawn widespread attention of the power consumption and carbon emission issues of Bitcoin 32 transactions. Accordingly, there exists an ongoing debate in academia and financial markets 33 on whether to incorporate green assets into Bitcoin-related portfolios for dual goals of risk 34 diversification and green investment commitment. 35

In parallel, the unprecedented COVID-19 pandemic has not only incurred vast changes to 36 the everyday life to individuals but have had profound effects on operations of economic and 37 financial systems (Huang et al., 2021). Since literature suggests that financial assets have 38 become more correlated in the bust periods (Hartmann et al., 2004; Lee et al., 2011), acute 39 economic losses incurred by the ongoing Covid019 pandemic therefore provides a critical 40 test-bed for the dynamics of financial market co-movement and the potential investment 41 shelters in the downturn (Goodell, 2020). Recently, Bitcoin, being recognized as 'liquid 42 gold' due to its independence from political and economic tones of sovereign nations, has 43 raised widespread discussion on its potential sheltering role during a market crash (Yarovaya 44 et al., 2020). At the same time, it is pointed out that market dependence between Bitcoin 45

<sup>&</sup>lt;sup>1</sup>For instance, see https://www.forbes.com/advisor/investing/bitcoins-energy-usage-explained/.

and financial assets features a time-varying pattern, indicating that the degree of hedging and
safe-haven properties of Bitcoin, if any, tends to evolve over time instead of being constant
(Shahzad et al., 2019; Ren et al., 2022). Thus, the above entails careful reexamination of the
role of the pandemic in time-varying spillovers of financial markets, particularly the dynamic
dependence between Bitcoin and green assets in our case.

Against this background, this paper examines the time-varying market dependence be-51 tween Bitcoin and green financial assets before and during the COVID-19 pandemic through 52 a time-varying parameter VAR (TVP-VAR) model with stochastic volatility. To study the 53 role of the pandemic on the market dependence, we have followed extant literature (see, e.g., 54 Goodell and Goutte, 2021b; Huang et al., 2021) by considering two sub-samples before and 55 after the pandemic onset with the whole daily dataset spanning from September 2018 to 56 September 2021. The dynamics and asymmetry of the market dependence are studied from 57 three perspectives, i.e., various time horizons, before and during the pandemic, and bidi-58 rectional market relations, respectively. The time-varying impulse response function (IRF) 59 analysis identified by our TVP-VAR model with stochastic volatility is conducted to study 60 the potentially dynamic market dependence over time. In addition, green financial assets 61 are represented by individual indices including the Dow Jones Sustainability World Index 62 (SWI), S&P ESG Leader Index (ESGLI), S&P Green Bond Index (GBI), and S&P Global 63 Clean Energy Index (GCEI). 64

Moreover, the uncertainty in social and economic conditions could alter investors' an-65 ticipation, and the weakening of which further triggers the financial turmoil, leading to 66 fluctuations in the cryptocurrency market (Wu et al., 2022). At the same time, it is also 67 known that uncertainty from different perspectives could exert varying impacts and predic-68 tive power on the cryptocurrency market (Lucey et al., 2021b). While rising uncertainty 69 plays a key role in the cryptocurrency market dynamics, the commonly-used index such as 70 economic policy uncertainty involves various sectors in the economy, which may not offer a 71 sufficiently-accurate measure for the uncertainty in the field of cryptocurrency. There has 72

long been a devoid of an uncertainty indicator that is specifically designed for the cryp-73 tocurrency until the recently-developed cryptocurrency uncertainty indices by Lucey et al. 74 (2021b). We therefore employ the two indices by Lucey et al. (2021b) that capture the 75 sizes of unpredictable disturbances in price and policy of cryptocurrencies, respectively. The 76 two indices are Uncertainty of Cryptocurrency Policy (UCRY Policy) and Uncertainty of 77 Cryptocurrency Price (UCRY Price). Moreover, growing energy consumption caused by 78 cryptocurrency mining and its related emission issues have raised emerging environmental 79 concerns about the cryptocurrency. We therefore further incorporate the Cryptocurrency 80 Environmental Attention Index (ICEA) proposed by Lucey et al. (2021a) that measures 81 the extent of environmental sustainability concerns on the cryptocurrency trading by the 82 public. This enables us to determine how the linkages vary under different attention levels 83 for cryptocurrency environmental issues. Therefore, in our paper, the cross-market linkage 84 between Bitcoin and green assets will be investigated in a setting where the uncertainty and 85 environmental attention for the cryptocurrency are well considered. 86

Our paper is closely linked to the extant literature in the following three strands, market 87 dependence of Bitcoin with financial assets and commodities, investment sheltering role of 88 Bitcoin, and the role of COVID-19 pandemic in the market interaction, with a particular 89 focus on the time-varying market linkage between Bitcoin and green financial assets. Specif-90 ically, existing studies has examined the market interaction between Bitcoin and general 91 financial markets of assets and commodities with some further discussion on the sheltering 92 role of Bitcoin for financial assets and the other way round (e.g., Bouri et al., 2018; Urguhart 93 and Zhang, 2019; Conlon et al., 2020; Corbet et al., 2021; Wang et al., 2021; Duan et al., 94 2021a,b). Amongst others, Conlon et al. (2020) and Goodell and Goutte (2021b) examine the 95 potential role of cryptocurrencies as a safe-haven or diversifier on equity indices during the 96 COVID-19 period. Maghvereh and Abdoh (2020) employ a quantile cross-spectral approach 97 to examine the tail dependence between prices of Bitcoin and financial indicators including 98 S&P 500 and exchange rate. Regarding the linkage between cryptocurrencies and commod-99

ity markets, including gold and various commodities such as energy, metals, and agricultural
products, Le et al. (2021a) measure the spillover pattern between prices of Bitcoin and other
assets including stock and gold, and find the consistent safe-haven property of gold in the
post-COVID-19 period. Rehman and Kang (2021) point out a lead-lag connection between
Bitcoin and crude oil market.

Moreover, although being limited and ongoing, the market dependence between cryp-105 tocurrencies notably including Bitcoin and green financial indices have been investigated 106 with studies determining whether the linkage is uni- or bi-directional, and whether it varies 107 over time remains unclear (e.g., Okorie, 2021; Symitsi and Chalvatzis, 2018). For instance, 108 Symitsi and Chalvatzis (2018) show that there exists unidirectional return and volatility 109 interactions between Bitcoin and green financial assets represented by stock indices of en-110 ergy and technology firms, while shock interactions are found to be bi-directional. As for 111 the potential investment shelter role for green financial assets against adverse market fluc-112 tuations of cryptocurrencies, consensus has yet to be reached in existing empirical analyses 113 (e.g., Naeem and Karim, 2021; Lucey et al., 2021a). For example, Naeem and Karim (2021) 114 point out that green financial assets such as clean energy indices act as hedges for Bitcoin. 115 Ren and Lucey (2021) show that stock prices of clean energy would not be a direct hedge 116 but safe-haven for both 'dirty' and 'clean' cryptocurrencies. Thus, the above demonstrates 117 that consensus of existing findings on the market interdependence between cryptocurrencies 118 and green energy indices has not yet been reached. 119

In addition, growing attention focuses on the role of COVID-19 pandemic in the financial market linkage. Given that the later could become more connected during the economic downturn (Hartmann et al., 2004), the market depression associated with onset of the ongoing pandemic offers a critical testbed in this regard (Huang et al., 2021). Recent literature studies the impact of the pandemic on the co-movement between markets of financial assets including Bitcoin and commodities under uncertainty (Aloui et al., 2020; Sharif et al., 2020), as well as the sheltering role of Bitcoin against financial assets and commodities (Conlon et al., <sup>127</sup> 2020; Conlon and McGee, 2020). While it has been pointed out that the market dynamics
<sup>128</sup> of financial assets including cryptocurrencies could be altered by the pandemic (Goodell and
<sup>129</sup> Goutte, 2021a), research on the pandemic impact on the market nexus between Bitcoin and
<sup>130</sup> green financial remains surprisingly scant by far.

Our paper contributes to the extant related literature in the following aspects. First, 131 rather than a conventional VAR model that only reports fixed coefficient estimates, we 132 employ a TVP-VAR model with stochastic volatility. Through this, the potential time-133 varying property of variable relationships in both matrices of coefficient and covariance is 134 respectively identified. Moreover, our employed method offers a time-varying IRF analysis 135 that outperforms the conventional IRF by capturing the variable response in the face of a 136 unit structural shock being imposed at different time points, such as the first quarter, the half 137 year, and the third quarter, respectively, in our case. Overall, to the best of our knowledge, 138 our paper is among the first that investigates the possibly asymmetric and time-varying 139 market information spillovers between Bitcoin and green financial assets, while considering 140 related financial indicators notably including the cryptocurrency uncertainty index and the 141 index of cryptocurrency environmental attention. 142

We find that the market dependence between Bitcoin and the four considered green fi-143 nancial assets is asymmetric from three perspectives, namely impact directions, time points 144 where the unit shock is being imposed in the IRF analysis, as well as before and during the 145 COVID-19 pandemic. As for the contemporaneous market relationship, Bitcoin is found to 146 act as an investment shelter of effective hedge for the specific green asset GCEI before the 147 pandemic, while its effective hedging role is further enhanced and extended to three green 148 financial assets, i.e., SWI, ESGLI, and GCEI, after the pandemic onset at different time hori-149 zons. Conversely, the four green financial assets play an effective hedge for Bitcoin, and such 150 the role remains constant irrespective of the pandemic. The mutually reinforcing and en-151 hanced investment sheltering ability between Bitcoin and green assets over time corroborates 152 with the extant literature, and is supported by no or even a negative relationship between 153

the two agents. The underlying reason has been discussed from two perspectives, i.e., green economic transition and dynamics of production costs in terms of rising energy prices. In addition, several additional analyses are conducted to demonstrate that our findings are robust to various changes in the research design. Our research is of important implications as to whether there exists a green investment shelter against adverse fluctuations of Bitcoin, and the sheltering role of Bitcoin for green portfolios.

The rest of this paper is structured as follows. Section 2 discusses our empirical data with preliminary analysis. Section 3 introduces our employed methodology. Section 4 analyzes empirical results. Section 5 discusses the analysis of robustness checks. The last section concludes with a discussion on policy implications.

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#### 2. Data and Preliminary Analysis

Our research dataset includes Bitcoin price (BTC), four green financial assets (i.e., SWI, ES-165 GLI, GBI, and GCEI), three indices describing the cryptocurrency market condition through 166 both perspectives of uncertainty (i.e., URCY policy and URCY price) and environmental 167 sustainability (ICEA), and the CBOE Volatility Index (VIX). In spirit of the extant litera-168 ture (e.g., Goodell and Goutte, 2021b; Huang et al., 2021), the research period spans from 169 01 September 2018 to 30 September 2021, and is divided into two sub-samples on 11 March 170 2020, which is the first day when the COVID-19 was announced as a pandemic by the WHO<sup>2</sup>. 171 Accordingly, sub-samples of 01 September 2018 - 10 March 2020 and 11 March 2020 - 30 172 September 2021 are used to represent pre- and post-COVID-19 periods, respectively. 173

Regarding the Bitcoin price series, it is represented by the Coindesk Price Index from www.coindesk.com, which represents an average of Bitcoin prices across leading Bitcoin exchanges worldwide.<sup>3</sup> The four green financial assets involve Dow Jones Sustainability

 $<sup>^2 \</sup>rm See$  details about key dates of COVID-19 announced by the WHO at https://www.who.int/emergencies/diseases/novel-coronavirus-2019/interactive-timeline.

<sup>&</sup>lt;sup>3</sup>The Coindesk Price Index is the data source commonly used in the literature (e.g. Kwon, 2020; Baur

World Index (SWI), S&P ESG Leader Index (ESGLI), S&P Green Bond Index (GBI), and
S&P Global Clean Energy Index (GCEI), and are from S&P Dow Jones Indices database
(www.spglobal.com/spdji). The three cryptocurrency market related indices contain Uncertainty of Cryptocurrency Policy Index (UCRY Policy), Uncertainty of Cryptocurrency Price
Index (UCRY Price), and Cryptocurrency Environmental Attention Index (ICEA). They are
sourced from Lucey et al. (2021a,b) that extracts news from the LexisNexis News & Business
database<sup>4</sup>. The VIX series is archived from COBE database (www.cboe.com).

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#### [Figure 1 about here.]

Temporal dynamics of the price series of Bitcoin, green financial assets, indices of cryp-185 tocurrency market condition, and VIX are drawn in Figure 1, respectively. Particularly, it 186 can be seen that all considered series except the VIX have witnessed a dramatic drop on 187 11 March 2020, i.e., the announcement date of COVID-19 as a pandemic. At the same 188 time, the VIX index peaked on the announcement date and then plunged, showing marked 189 volatility of financial markets induced by the pandemic. This also provides a visual demon-190 stration regarding the separation of our whole data to investigate the pandemic impact on 191 the cross-market linkages. For each of the considered series, it is transformed in the return 192 format as the first-order log difference between the current day and the last day times 100 of 193 the original series, i.e.,  $R_t = (\log (P_t) - \log (P_{t-1}))$ . Table 1 shows the descriptive statistics 194 for each transformed series in subsamples of pre- and post-COVID-19 periods, respectively. 195 Briefly, the return-transformed Bitcoin price index, green financial assets, and indices of 196 cryptocurrency market condition are found to possess larger mean values with higher stan-197 dard deviations since the COVID-19 is announced as a pandemic. At the same time the mean 198 value of return-transform VIX index turned to be negative with relatively lower volatility in 199 the post-COVID-19 period compared to that in the pre-COVID-19 period, being consistent 200

et al., 2018; Bouri et al., 2019; Kapar and Olmo, 2019)

<sup>&</sup>lt;sup>4</sup>See details about data of UCRY Policy, UCRY Price, ICEA at https://lnkd.in/egtcZvzS.

<sup>201</sup> with the existing findings (e.g., Lucey et al., 2021a,b; Naeem and Karim, 2021).

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[Table 1 about here.]

#### 3. Methodology

#### <sup>204</sup> 3.1. Time-Varying Parameter VAR with Stochastic Volatility

The time-varying parameter VAR (TVP-VAR) model with stochastic volatility is known 205 to outperform the conventional VAR method and has been widely applied to modelling 206 the dynamic relationship between macroeconomic and financial indicators (e.g., Chan and 207 Eisenstat, 2018; Clark and Ravazzolo, 2015; D'Agostino et al., 2013; Nakajima et al., 2011). 208 It not only considers the potential time-varying feature of the underlying structure, but also 209 accommodates the fluctuating disturbance, particularly in the financial turmoil. Therefore, 210 we employ the TVP-VAR model with stochastic volatility for clearer comprehension on 211 the relationship between price returns of Bitcoin and green financial assets by considering 212 the condition of the cryptocurrency market through aspects of uncertainty, environmental 213 sustainability, and financial volatility. 214

The model departs from a k-dimensional return vector  $R_t$  that involves Bitcoin prices BTN<sub>t</sub>, four green financial asset prices  $GFA'_t = (SWI_t, ESGLI_t, GBI_t, GCEI_t)'$ , three cryptocurrency market related indices  $CRI'_t = (UCRYPolicy_t, UCRYPrice_t, ICEA_t)'$ , and the financial market volatility  $VIX_t$  at period t (t = 1, ..., T). The TVP-VAR model with stochastic volatility is then formulated as: framework is formulated as

$$R_t = B_{1t}R_{t-1} + \ldots + B_{pt}R_{t-p} + \varepsilon_t, \quad \varepsilon_t \sim N\left(0, A_t^{-1}\Sigma_t \epsilon_t\right), \tag{1}$$

for t = p + 1, ..., n, where  $B_{1t}, ..., B_{pt}$  are  $k \times k$  autoregressive coefficient matrices with the number of variables k and lags of p.  $\epsilon_t$  refers to a  $k \times 1$  structural shocks with  $A_t$  being identified as a lower triangular covariance matrix,  $\Sigma_t$  being as a diagonal matrix of stochastic volatility, and  $\epsilon_t \sim N(0, I)$ . The structures of  $A_t$  and  $Sigma_t$  are presented as follows:

$$A_{t} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ a_{21,t} & 1 & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ a_{k1,t} & \cdots & a_{k,(k-1),t} & 1 \end{bmatrix} \quad \text{and} \quad \Sigma_{t} = \begin{bmatrix} \sigma_{1,t}^{2} & 0 & \cdots & 0 \\ 0 & \sigma_{2,t}^{2} & \ddots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & \sigma_{k,t}^{2} \end{bmatrix}.$$

Let  $X_t = (R'_{t-1}, \ldots, R'_{t-p})'$  and  $B'_t = (B_{1t}, \ldots, B_{pt})$ , the model can be rewritten in a more compact form:

$$R_t = B'_t X_t + \varepsilon_t. \tag{2}$$

To better model the potentially time-varying relationship between macro-financial variables in the TVP-VAR model, the coefficients of  $B_t$ , covariance matrix of  $A_t$ , and stochastic volatility of  $\Sigma_t$  are allowed to vary over time. To model the time-varying coefficients, we define  $\beta_t = \text{vec}(B_t)$ , by assuming that  $\beta_t$  follows a random walk process with  $t = p+1, \ldots, n$ ,

$$\beta_t = \beta_{t-1} + u_t, \quad u_t \sim N\left(0, \Sigma_\beta\right). \tag{3}$$

Similarly, we construct  $a_t$  as a row-wise stacked vector of the lower-triangular covariance elements of  $A_t$ , i.e.,  $a_t = (a_{21,t}, \ldots, a_{k,k-1,t})'$  and  $\sigma_t^2$  as a vector of the diagonal elements of  $\Sigma_t$ , i.e.,  $\sigma_t^2 = (\sigma_{1,t}^2, \ldots, \sigma_{k,t}^2)'$ , following Primiceri (2005). Both  $a_t$  and  $\sigma_t^2$  are assumed to follow a random walk process, and we have

$$a_t = a_{t-1} + \vartheta_t, \qquad \qquad \vartheta_t \sim N\left(0, \Sigma_a\right), \tag{4}$$

$$\log(\sigma_t) = \log(\sigma_{t-1}) + v_t, \qquad v_t \sim N(0, \Sigma_{\sigma}), \qquad (5)$$

Therefore, the vector involving the innovations to the structural shocks  $\varepsilon_t$  ( $\epsilon_t$ ), timevarying coefficients  $\beta_t$  ( $u_t$ ), and stochastic volatilities of  $a_t$  ( $\vartheta_t$ ) and log( $\sigma_t^2$ ) ( $v_t$ ) is known to <sup>228</sup> follow the below distribution

$$\begin{bmatrix} \epsilon_t \\ u_t \\ \vartheta_t \\ v_t \end{bmatrix} = \begin{bmatrix} I & 0 & 0 & 0 \\ 0 & \Sigma_\beta & 0 & 0 \\ 0 & 0 & \Sigma_a & 0 \\ 0 & 0 & 0 & \Sigma_\sigma \end{bmatrix}.$$
 (6)

Given that our TVP-VAR model with stochastic volatility is featured by large timevarying parameters as specified in Equations (3) to (5), we will discuss in Section 3.3 regarding how the Bayesian approach using a Markov chain Monte Carlo (MCMC) algorithm is employed to estimate the model.

#### <sup>233</sup> 3.2. Shock Identification

To identify the structural shock in our TVP-VAR model with stochastic volatility, we conduct a Cholesky decomposition for the covariance matrix of  $\Sigma_t$ , as shown in Section 3.2. We order the target series following the existing literature (e.g., Lucey et al., 2021a,b; Naeem and Karim, 2021) as:

$$R_{t} = (UCRYPolicy_{t}, VIX_{t}, BTN_{t}, SWI_{t}, ESGLI_{t}, GBI_{t}, GCEI_{t}, UCRYPrice_{t}, ICEA_{t}).$$
(7)

Regarding the reason for the above ordering, in line with Lucey et al. (2021a,b), the series of the cryptocurrency policy uncertainty (UCRY policy) index is ordered first since that series such as VIX and Bitcoin can react instantaneously to uncertainty policy shocks. Given that other considered series are known to react faster than UCRY price index, the latter is accordingly ordered behind with the index of environmental sustainability of the cryptocurrency market (ICEA) being ordered last. Moreover, it is further worth noting that ordering Bitcoin before green financial assets is consistent with Naeem and Karim (2021).

#### <sup>245</sup> 3.3. Estimation Algorithm

The Bayesian method is applied to estimate our TVP-VAR model with stochastic volatility, wherein the MCMC algorithm is conducted using a simulation-based estimation that generates a sequence of draws from the full conditional distributions of all parameters. The time-varying parameters can be well estimated through a nested structure that computes the likelihood function for each iteration process. To enforce the MCMC algorithm, the following conditional distributions are formulated:

$$\beta_{t} | a_{t}, \sigma_{t}^{2}, \Sigma_{\beta},$$

$$a_{t} | \beta_{t}, \sigma_{t}^{2}, \Sigma_{a},$$

$$\sigma_{t}^{2} | \beta_{t}, a_{t}, \Sigma_{\sigma},$$

$$\Sigma_{\beta} | \beta_{t},$$

$$\Sigma_{a} | a_{t},$$

$$\Sigma_{\sigma} | \sigma_{t}^{2}.$$
(8)

Detailed discussions of drawing samples from the above conditional posterior distributions with priors follow the MCMC procedures of Chen et al. (2020) and Nakajima et al. (2011), where  $\sigma_t^2$  is illustrated as  $\exp(h_t)$ .

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#### 4. Empirical Results

#### <sup>256</sup> 4.1. Parameter Estimates

To examine the spillover pattern between price returns of Bitcoin and green financial assets while considering related financial indicators (i.e., cryptocurrency market related indices and VIX), and investigate its potential dynamics over time before and during the COVID-19 pandemic, we conduct the estimation by using the time-varying VAR (TVP-VAR) model with stochastic volatility in the two subsamples, respectively. The lag order of the TVP-VAR

system is set to be  $1,^5$  and jointly given that there are nine variables in total included in 262 the system, the estimated parameters therefore include 81 autoregressive coefficients ( $\beta_t$ ), 36 263 covariance parameters  $(a_t)$ , and 9 volatility parameters  $(\sigma_t^2)$  for the pre- and post-COVID-19 264 periods, respectively. Overall, with a particular focus on the time-varying relationships of 265 Bitcoin with green assets and related financial indicators, Figures 2 and 3 show dynamics 266 of the posterior draws of  $\beta_t$  over time before and during the COVID-19 pandemic; panels 267 (a) and (b) of 4 and 5 present dynamic plots of the posterior draws of  $a_t$  and  $\sigma_t^2$  over time 268 before and after the announcement of the pandemic, respectively. The above time-varying 269 dynamics are depicted in the x-axis of the corresponding figures where the period of 01 270 September 2018 - 11 March 2020 denotes the period before the pandemic and 12 March 2020 271 - 30 September 2021 denotes the one during the pandemic. 272

273

[Figure 2 about here.]

274

[Figure 3 about here.]

With regard to the time-varying posterior estimates of  $\beta_t$ , the role of Bitcoin in driving 275 other series and the other way around exhibit distinct results over time, as shown in Figures 276 2 and 3. We find the potentially bi-directional relationships of Bitcoin prices with green 277 financial asset prices (i.e., SWI, ESGLI, GBI, and GCEI) and related financial indicators in-278 cluding indices for the cryptocurrency market condition through perspectives of uncertainty 279 (URCY Policy and URCY Price) and environmental sustainability (ICEA), and VIX gener-280 ally remain stable and unchanged overtime before and after the outbreak of the COVID-19. 281 Interestingly, it is worth noting that time-varying coefficients that show the relationship be-282 tween Bitcoin prices and related financial indicators are found to be less significant at most 283 of the time. 284

<sup>&</sup>lt;sup>5</sup>While there is no theoretical guidance for the lag order selection in the time-varying parameter (TVP) VAR system, in the empirical analysis, we have tried different order numbers, e.g., 2, 3, 4, or 5. The results are consistent with that of our benchmark estimation, and they are available from the authors upon request.

#### [Figure 4 about here.]

Moreover, the time-varying posterior estimates of  $a_t$ , and  $\sigma_t^2$  experience a relatively 286 marked fluctuation before and after announcement of the COVID-19 pandemic. As shown in 287 Figures 3, it is clear that the systematic relationship (i.e., covariance shown by  $a_t$ ) of Bitcoin 288 prices with price indices of green financial assets and related financial indicators could be 289 time-varying and sensitive to the COVID-19 pandemic, such as the relationship of Bitcoin 290 with SWI and UCRY price; at the same time, the estimates of  $a_t$  for the correlations of 291 Bitcoin with URCY Policy and VIX tend to be time-invariant. In addition, the dynamic 292 time-varying pattern of the covariance of Bitcoin prices with prices of green financial assets 293 is found to be different and even reserved in pre- and post-COVID-19 periods, while the 294 counterpart of that with financial related indicators is shown to be relatively constant in the 295 two sub-samples. 296

297

#### [Figure 5 about here.]

As for the estimates of  $\sigma_t^2$  displayed in Figure 5, estimated variances of price indices of green financial assets and related financial indicators are shown to be time-varying with different patterns before and after the announcement of the pandemic. It is worth mentioning that  $\sigma_t^2$  of most of the green financial assets increases (except for that on ESGLI), while that on indices related to the cryptocurrency market condition tend to decrease as time evolving to the post-COVID-19 period.

### <sup>304</sup> 4.2. Time-varying impacts of Bitcoin on green financial assets and related <sup>305</sup> financial indicators

Whether and how the impact of Bitcoin prices on green financial asset prices and related financial indicators (i.e., indices of the cryptocurrency market condition from perspectives of uncertainty and environmental sustainability and the VIX index) varies over time before

and after the announcement of the COVID-19 as a pandemic? To answer this question, we 309 conduct the time-varying impulse response function (IRF) analysis based on the parameter 310 estimates of the employed TVP-VAR model with stochastic volatility. Rather than the 311 conventional IRF that only observes variable response after receiving a unit shock to a 312 target variable at the initial time point (i.e., time 0), the IRF produced by the TVP-VAR 313 model with stochastic volatility captures potential variations of the variable response after 314 receiving a unit shock at different time points. That is, the variable response is allowed to 315 be time-varying once receiving a shock to a target variable at different time points, viz., 316 the first quarter, the half year, and the third quarter, respectively, in our case. Since its 317 initial development by Cogley and Sargent (2001, 2005) and Primiceri (2005), it has been 318 raised an increasing attention by related literature (e.g., Chan and Eisenstat, 2018; Clark 319 and Ravazzolo, 2015; D'Agostino et al., 2013). Time variations of responses of green financial 320 asset prices and related financial indicators are plotted as shown in Figures 6 and 7 once 321 receiving a unit shock to Bitcoin prices in the first quarter, the half year, and the third 322 quarter, respectively. The potential different pattern of the time-variations in the pre- and 323 post-periods of the COVID-19 pandemic is compared and shown in Panels (a) and (b) of 324 the two figures, respectively. Overall, the variable responses are shown to have relatively 325 consistent tendency while having different magnitudes once receiving the Bitcoin price shock 326 on the three different time points. The variable responses during the COVID-19 period 327 demonstrate a different dynamic pattern compared to that before the COVID-19 period. 328

329

#### [Figure 6 about here.]

Specifically, once receiving a unit shock to Bitcoin prices before the COVID-19 pandemic, as shown in Panel (a) of Figure 6, the contemporaneous responses of three of the four selected green financial asset prices, i.e., prices of SWI, ESGLI, and GBI, are shown to be positive, quickly turn to become negative after a short period, and eventually revert to zero after around four horizons. At the same time, the response pattern of the specific green financial

asset (GCEI) price is shown to be reversed that it receives a negative and instant move, and 335 quickly reverts to zero within around four horizons. The above response pattern is found 336 to be consistent regardless of when the unit shock to Bitcoin prices is received, while the 337 response magnitude varies over horizons with different time points for the shock impulse. 338 Once receiving a unit shock to Bitcoin prices after the COVID-19 pandemic, as shown in 339 Panel (b) of Figure 6, the time-varying responses of both GBI and GCEI are shown to have 340 a broadly similar pattern as their counterparts shown before the COVID-19 pandemic; in 341 contrast, the responses of SWI and ESGLI reverse after the pandemic announcement. In 342 addition, the sizes of responses of each of the four green financial asset prices are found to 343 be relatively similar once the shock to Bitcoin prices is at different time points. 344

345

#### [Figure 7 about here.]

Regarding the time-varying responses of related financial indicators to a unit shock of 346 Bitcoin prices, both response path and magnitude are shown to be distinct before and after 347 the announcement of the COVID-19 pandemic. In Panel (a) of Figure 7, it can be seen that 348 cryptocurrency market related indices for policy and price uncertainty, i.e., URCY policy 349 and URCY price, depict a negative instant response although with different degrees and 350 then revert to zero after four horizons in the pre-COVID-19 period. At the same time, the 351 cryptocurrency market related index for environmental sustainability (ICEA) and the VIX 352 index broadly exhibit a positive response and then go back to zero with different response 353 sizes; this pattern for ICEA particularly applies only to short time points. In contrast, 354 after announcement of the COVID-19 pandemic, as shown in Panel (b) of Figure 7, time-355 varying patterns of the responses of the four related financial indicators differ compared 356 with that before the pandemic. For example, the response of the VIX index totally turns 357 become negative against a positive response before the pandemic. Importantly, it is worth 358 mentioning that responses of each of the green financial asset prices and related financial 359 indicators are shown to be time-varying instead of being constant, depending on the time 360

points when a unit shock is imposed on the Bitcoin price, i.e., the first quarter, the half year,
and the third quarter, respectively, in our case.

Importantly, regarding the contemporaneous relationship, before the COVID-19 pan-363 demic, Bitcoin prices exert positive impacts on prices of three out of the four target green 364 financial assets, i.e., SWI, ESGLI, and GBI, except for the price of GCEI, which is influenced 365 negatively, at various time horizons. Such positive impacts on prices of SWI and ESGLI turn 366 to become negative during the pandemic, while the respective positive and negative impacts 367 on GBI and GCEI are shown to be consistent in the face of the pandemic outbreak. It can 368 be therefore found that Bitcoin acts as an investment shelter on the specific green financial 369 asset GCEI before the pandemic, while its sheltering role is enhanced and further expanded 370 to three green financial assets, i.e., SWI, ESGLI, and GCEI, during the pandemic. As for 371 indices related to the cryptocurrency market condition, Bitcoin is found to perform a con-372 sistent zero contemporaneous impact on URCY policy and VIX in both subsamples before 373 and after the pandemic onset, while its sheltering role for UCRY price is only effective before 374 the pandemic. Our results of the investment sheltering property of Bitcoin against various 375 financial indicators corroborate with existing literature (Huang et al., 2021; Le et al., 2021a). 376 The hedging role of Bitcoin against green assets is found to be further enhanced after the 377 pandemic onset over time as supported by the weakened relation between Bitcoin and green 378 assets. The corresponding reason can be attributed to both aspects of green transition and 379 dynamics of production costs, and will be further discussed in the next section. 380

### <sup>381</sup> 4.3. Time varying impacts of green financial assets and related financial <sup>382</sup> indicators on Bitcoin

Having studied the spillover of Bitcoin price fluctuations on prices of green financial assets
and related financial indicators, we investigate the spillover impact from the other direction
by using the time-varying IRF produced by the TVP-VAR model with stochastic volatility.
Through this, the potentially bi-directional relationship of Bitcoin prices with green financial

asset prices and related financial indices is identified. The corresponding results regarding 387 the time-varying Bitcoin price response at different time points in pre- and post-COVID-388 19 periods are respectively reported in Figures 8 and 9. Specifically, once receiving a unit 389 price shock to green financial assets before the pandemic, as shown in Panel (a) of Figure 390 8, Bitcoin prices respond positively from zero at the initial stage to shocks from SWI and 391 GCEI, while the impacts of ESGLI and GBI on Bitcoin prices turn negative from zero. The 392 response of Bitcoin prices to the shocks from green financial assets at different time points 393 (i.e., the first-quarter, the half year, and the third-quarter periods) depicts a consistent 394 and overlapping pattern. Moreover, as depicted in Panel (b) of Figure 8, Bitcoin prices 395 demonstrate a consistent response to shocks from all green financial assets before and after 396 the pandemic onset except for GBI. For the latter, it experienced a reversed pattern when 397 facing Bitcoin price shocks before and after the pandemic. 398

399

#### [Figure 8 about here.]

In terms of the Bitcoin price response to a unit shock from related financial indicators, 400 as shown in Figure 9, the responses to shocks from indices of the cryptocurrency market 401 condition from perspectives of price uncertainty (UCRY Price) and environmental sustain-402 ability (ICEA) tend to have a similar pattern that evolves from zero at the initial time point, 403 peaks at the short horizons, and then revert to zero within around four horizons. The above 404 responses are highly consistent not only in the face of shocks from different time points (i.e., 405 the first-quarter, the half year, and the third-quarter periods), but also in pre- and post-406 COVID-19 periods. Moreover, the responses of Bitcoin prices to shocks from both indices of 407 the cryptocurrency market condition from perspectives of policy uncertainty (URCY Policy) 408 and VIX are shown to have different dynamic patterns in the face of shocks from different 409 points. At the same time, the announcement of the pandemic tends to reverse the Bitcoin 410 price response to shocks from URCY Policy and VIX at various time points. 411

412

[Figure 9 about here.]

Importantly, as for the contemporaneous price impacts of the four-target green financial 413 assets on Bitcoin, they are shown to be zero and remain consistent before and during the 414 COVID-19 pandemic at different time points (i.e., the first quarter, the half year, and the 415 third quarter) as depicted in Figure 8. The above demonstrates the effective hedging role 416 of green financial assets for Bitcoin. Regarding that of the financial indicators related to 417 the cryptocurrency market condition, indices of VIX, UCRY price, and ICEA consistently 418 depict negative or no significant relationship with Bitcoin before and during the pandemic 419 at different time points, showing as an investment hedge. At the same time, compared to 420 the pre-pandemic period, the impacts of UCRY policy on Bitcoin reverses to be negative 421 after the pandemic onset with relatively large magnitudes. Furthermore, it is worth noting 422 that although being within the same and broad category, different types of green assets 423 could impact Bitcoin differently, and vice versa. This is consistent with the viewpoint that 424 various green assets could result in distinct environmental benefits from green transition, 425 further leading to their different relations with Bitcoin (Naeem and Karim, 2021; Symitsi and 426 Chalvatzis, 2018). In addition, since financialization of green-related projects/investments 427 features different degrees, this might also lead to different relation patterns of various green 428 assets with Bitcoin (Naeem et al., 2021). 429

Overall, our results are in line with existing findings that green markets provide effective 430 diversification and hedging potential for other financial assets notably including Bitcoin 431 (Naeem et al., 2021; Nguyen et al., 2021; Lucey et al., 2021b). In the light of the extant 432 literature (Chan et al., 2019; Cheah et al., 2022; Dyhrberg, 2016; Grobys, 2021; Liu et al., 433 2022; Pal and Mitra, 2019), such the sheltering role of green assets can be explained by two 434 underlying mechanisms, i.e., the process of green economic transition, and the dynamics of 435 production costs. In the context of a globally rising energy consumption with large CO2 436 emissions, the environmental threats have encouraged the finance of cleaner energy, while 437 also calling for green transition of the energy-intensive development mode including Bitcoin 438 trading and mining. Accordingly, although the presence of active Bitcoin trading pursuing 439

for economic profits, green investors would instead adhere to cleaner investments, leading to 440 negligible or even reverse relationship between dynamics of the two types of assets (Naeem 441 and Karim, 2021). At the same time, from the perspective of production costs, it has to be 442 acknowledged that fossil energy is still the major source that drives the economic operation, 443 and the its price along with the electricity price have recently experienced a marked rising 444 due to various reasons such as pandemic recovery, political conflicts, and energy structure 445 adjustment, etc. This would further encourage the usage of renewable and clean energy 446 financed by green projects, while Bitcoin activities would instead encounter higher costs 447 due to the high electricity usage (Le et al., 2021b). Thus, our findings demonstrate the 448 effectiveness of the investment sheltering role of green assets for Bitcoin-related portfolios 449 that inclusion of the former would hedge against the financial risk associated with the latter. 450

451

#### 5. Robustness Checks

How robust are our findings to changes in the research design? In this section, we conduct additional analyses to reassure the robustness of our findings in the face of inclusion of the price series of another important cryptocurrency, i.e., Ethereum, replacement of the green asset indices with alternative ones, and changes in the sample period, respectively.

#### <sup>456</sup> 5.1. Inclusion of the Ethereum price series

How sensitive are our obtained results regarding the market interaction between Bitcoin and 457 green assets when considering market dynamics of other leading cryptocurrencies? Our ro-458 bustness check starts by including the Ethereum price in the empirical estimation. Ethereum 459 is known as the cryptocurrency Ether issued in a decentralized computing platform to reward 460 mining nodes, which is another widely-traded cryptocurrency in addition to Bitcoin (Conlon 461 et al., 2020). Specifically, we re-estimate the market interaction between Bitcoin and green 462 asset by using the TVP-VAR model with the stochastic volatility, while considering the role 463 of Ethereum price. 464

Overall, as an intuitive illustration through the time-varying impulse response function 465 (IRF) analysis, it is clear that the results are broadly consistent in the estimations with and 466 without considering the Ethereum price.<sup>6</sup> In particular, the IRF plot regarding the Bitcoin 467 - green asset market interaction when considering the Ethereum price before and during the 468 COVID-19 pandemic is exhibited in Figure 10, which results are similar to the corresponding 469 counterparts in the main analysis shown in Figure 5. In terms of the contemporaneous 470 relationship, the response of green financial assets after receiving a unit shock to Bitcoin 471 prices before the pandemic is presented in Panel (a) of Figure 10 where Bitcoin acts as 472 an investment shelter on the specific green financial asset GCEI. During the pandemic, its 473 sheltering role is further enhanced and extended to three green assets, i.e., SWI, ESGLI, 474 and GCEI, as shown in Panel (b) of Figure 10. The dynamic and potentially sheltering role 475 of the Bitcoin price tends to be consistent when the unit shock is impulse in different time 476 points, i.e., the first quarter, the half year, and the third quarter, further demonstrating that 477 our results are not sensitive to inclusion of additional cryptocurrency prices. 478

479

#### [Figure 10 about here.]

#### 480 5.2. Replacement of green asset indices with alternative ones

To further examine the robustness of our main results regarding the time-varying market dependence of Bitcoin with green financial assets, we follow Ren and Lucey (2021) by considering the WilderHill Clean Energy Index (CEI) to measure the overall performance of clean energy sector. CEI is then used to replace the S&P Global Clean Energy Index (GCEI) in the re-estimation of our TVP-VAR model with stochastic volatility. Our obtained results after the replacement of GCEI by CEI have been visually presented by the time-varying IRF

<sup>&</sup>lt;sup>6</sup>For each of the following three robustness checks, the IRF results about all the time-varying interactions of Bitcoin prices with green financial asset prices and related financial indices before and during the COVID-19 pandemic are shown to be broadly consistent with our main findings. While the full version of the IRF plots is compressed due to limited space and is available from the authors upon request, we provide some typical illustrations that show the particular market interactions under focus.

<sup>487</sup> plot with the impact of the market dynamics of green assets on that of Bitcoin particularly <sup>488</sup> shown in Figure 11. Overall, the response patterns of the Bitcoin price when receiving a <sup>489</sup> unit shock to each of the four green financial assets shown in Figure 11 generally mimic that <sup>490</sup> of the counterparts in our main findings reported in Figure 8 both before and during the <sup>491</sup> COVID-19 periods. This further supports our argument that the green financial assets are <sup>492</sup> typically regarded as the investment shelter as a (weak) hedge for Bitcoin. Therefore, our <sup>493</sup> results are not sensitive to the variable replacement of the employed green financial assets.

494

#### [Figure 11 about here.]

#### $_{495}$ 5.3. Changes in the sample period

As an additional robustness check, we conduct a re-estimation by changing the sample period. 496 Following the extant literature (e.g., Conlon et al., 2020; Conlon and McGee, 2020), updated 497 sub-samples before and after the announcement date of the COVID-19 pandemic with the 498 length for each sub-sample being as a whole year, i.e., from 01 March 2019 to 11 March 2020, 490 and from 12 March 2020 to 30 March 2021. It has been checked that the market relationship 500 of Bitcoin with green assets and related financial indicators using the updated subsamples 501 is consistent with that of our main findings. Particularly, the time-varying IRF results for 502 the response of Bitcoin prices in the face of a unit shock to related financial indicators 503 using the updated sub-samples of before and during the pandemic are respectively depicted 504 in Figure 12, being broadly in line with the corresponding counterparts obtained in our 505 main results shown in Figure 9. Importantly, the contemporaneous price impacts of related 506 financial indices on Bitcoin are found to be zero except the positive impact of URCY policy 507 on Bitcoin before the pandemic although its contemporaneous impact turns to be zero during 508 the pandemic. The above speaks in favor of the weak sheltering role of typical and related 509 financial indicators such as VIX, UCRY price, and ICEA for Bitcoin in both the sub-samples 510 of before and during the COVID-19, further reassuring the robustness of our main findings. 511

512

[Figure 12 about here.]

#### 6. CONCLUSIONS

In the face of rapid climate warming worldwide, the pursuit of future dynamics of the Bitcoin market within a green and sustainable environment has become increasingly important, driving ongoing but still limited attention on whether to add green assets to the Bitcoinrelated portfolio for risk diversification while contributing to the green commitment. At the same time, the onset of the COVID-19 pandemic and its associated financial turmoil call for a reinvestigation of the cross-sectional dynamics of the financial markets, among which the connectedness tends to become more correlated during the downturn.

Our paper, therefore, examines the dynamics of the market dependence between Bitcoin 521 and green financial assets over time by applying a recently-developed time-varying param-522 eter VAR (TVP-VAR) model with stochastic volatility. Four indices such as Dow Jones 523 Sustainability World Index (SWI), S&P ESG Leader Index (ESGLI), S&P Green Bond In-524 dex (GBI), and S&P Global Clean Energy Index (GCEI) are considered to represent the 525 dynamics of the green assets through four perspectives in a comprehensive manner. The 526 role of two uncertainty indices specifically designed for the cryptocurrency market by Lucey 527 et al. (2021b), along with an index of cryptocurrency environmental attention developed by 528 Lucey et al. (2021a), are particularly considered for accurate interpretation of the Bitcoin 529 market interaction with green asset markets. 530

Consistent with our expectations, the potential asymmetry of the market dependence is 531 found through three perspectives, i.e., impact directions, time points where the unit shock 532 is being imposed in the IRF (i.e., short-, medium-, and long-runs), as well as before and 533 during the pandemic. The investment sheltering role of Bitcoin/green assets against adverse 534 fluctuations in one another is shown by the contemporaneous market linkage. Specifically, 535 before the pandemic, Bitcoin is found to consistently play a sheltering role of effective hedge 536 on the specific green financial asset GBI at various time horizons as indicated by its negative 537 price impact on GBI. The sheltering role of Bitcoin is further enhanced after the pandemic 538

<sup>539</sup> outbreak that the price of Bitcoin negatively affects that of three out of four green assets <sup>540</sup> under research, i.e., SWI, ESGLI, and GBI, during the pandemic. On the other hand, the <sup>541</sup> four green assets are shown to consistently exhibit an investment shelter as a weak hedge for <sup>542</sup> Bitcoin at different time horizons before and during the pandemic. The robustness of our <sup>543</sup> findings is further reassured by a series of additional analysis.

While it is known that assets tend to become increasingly co-moved during economic 544 downturns (e.g., Bekaert et al., 2009; Goodell and Goutte, 2021b), in the financial turmoil 545 related to the COVID-19 bear market, Bitcoin has depicted its independence with other 546 assets, showing effective investment sheltering role especially against green assets. These 547 findings contribute to the currently-ongoing debate as to whether green assets could act 548 as an investment shelter for the Bitcoin-related portfolio, and offer Bitcoin investors with 549 clear insights for both risk mitigation and green commitment in their portfolios. In turn, 550 comprehension on the underlying investment sheltering role of Bitcoin for green portfolios 551 is also gained. Therefore, our findings are important for not only hedging against green 552 portfolios but also seeking green shelter. The findings also help policymakers promote the 553 green and sustainable development of the financial market, notably the Bitcoin market, 554 effectively combating global climate change 555

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Figure 1: Dynamics of target factors over time

*Note:* This figure plots the time-varying dynamics of nine target series under research including the Bitcoin price (BTC), four green financial assets (i.e., SWI, ESGLI, GBI, and GCEI), three indices describing the cryptocurrency market condition through both perspectives of uncertainty (i.e., URCY policy and URCY price) and environmental sustainability (ICEA), and the CBOE Volatility Index (VIX).



Figure 2: Time-varying posterior estimates for autoregressive coefficients  $\beta_t$  before the COVID-19 pandemic

Note: This figure reports the time-varying dynamics of the posterior estimates of  $\beta_t$  defined in Equation (3) before the pandemic. Sub-figures on left two columns show the dynamic impacts of Bitcoin prices on prices of green assets (i.e., SWI, ESGLI, GBI, and GCEI) and related financial indicators (i.e., UCRY policy, UCRY price, ICEA, and VIX) over time. Conversely, the ones on right two columns show the response of Bitcoin when facing impacts from green assets and related financial indicators over time. In each sub-figure, the blue solid line stands for the time-varying dynamics of the posterior mean of  $\beta_t$ , and the red dotted lines denote dynamics of the corresponding 95% intervals. The x-axis denotes time periods, and the y-axis denotes the impact magnitude.



Figure 3: Time-varying posterior estimates for autoregressive coefficients  $\beta_t$  in during the COVID-19 pandemic

Note: This figure reports the time-varying dynamics of the posterior estimates of  $\beta_t$  defined in Equation (3) during the pandemic. Sub-figures on left two columns show the dynamic impacts of Bitcoin prices on prices of green assets (i.e., SWI, ESGLI, GBI, and GCEI) and related financial indicators (i.e., UCRY policy, UCRY price, ICEA, and VIX) over time. Conversely, the ones on right two columns show the response of Bitcoin when facing impacts from green assets and related financial indicators over time. In each sub-figure, the blue solid line stands for the time-varying dynamics of the posterior mean of  $\beta_t$ , and the red dotted lines denote dynamics of the corresponding 95% intervals. The x-axis denotes time periods, and the y-axis denotes the impact magnitude.



## Figure 4: Time-varying posterior estimates for covariance parameters $a_t$ before and during the COVID-19 pandemic

Note: This figure reports the time-varying dynamics of the posterior estimates of t defined in Equation (4). The estimated covariance parameters (i.e., t) of Bitcoin prices with green asset prices and related financial indicators before and during the pandemic are shown in panels (a) and (b), respectively. In each sub-figure, the blue solid line stands for the time-varying dynamics of the posterior mean of  $\beta_t$ , and the red dotted lines denote dynamics of the corresponding 95% intervals. The x-axis denotes time periods, and the y-axis denotes the impact magnitude.



#### Figure 5: Time-varying posterior estimates for volatility parameters $\sigma_t^2$ in the preand post-COVID-19 periods

Note: This figure reports the time-varying dynamics of the posterior estimates of  $\sigma_t^2$  defined in Equation (5). The estimated volatility parameters (i.e.,  $\sigma_t^2$ ) of each of the considered series before and during the pandemic are shown in panels (a) and (b), respectively. In each sub-figure, the blue solid line stands for the time-varying dynamics of the posterior mean of  $\beta_t$ , and the red dotted lines denote dynamics of the corresponding 95% intervals. The x-axis denotes time periods, and the y-axis denotes the impact magnitude.



Figure 6: The time-varying impulse response of green financial asset prices to a unit shock of Bitcoin prices in the pre- and post-COVID-19 periods

Note: This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the specific green asset when facing a unit shock to Bitcoin prices at different time points, i.e., the first quarter, the half year, and the third quarter, respectively. 35



Figure 7: The time-varying impulse response of related financial indicators to a unit shock of Bitcoin prices in the pre- and post-COVID-19 periods

*Note:* This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the specific related financial indicator when facing a unit shock to Bitcoin prices at different time points, i.e., the first quarter, the half year, and the third quarter, respectively.



Figure 8: The time-varying impulse response of Bitcoin prices to unit shocks of green financial asset prices in the pre- and post-COVID-19 periods

*Note:* This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the Bitcoin price when facing a unit shock to the price of the specific green asset at different time points, i.e., the first quarter, the half year, and the third quarter, respectively.



Figure 9: The time-varying impulse response of Bitcoin prices to unit shocks of related financial indicators in the pre- and post-COVID-19 periods

*Note:* This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the Bitcoin price when facing a unit shock to the related financial indicator at different time points, i.e., the first quarter, the half year, and the third quarter, respectively. 38



Figure 10: Robustness 1: IRF plots of green asset prices to a unit shock of Bitcoin prices in the pre- and post-COVID-19 when including the Ethereum prices

*Note*: This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the specific variable when facing a unit shock to Bitcoin prices at different time points, i.e., the first quarter, the half year, and the third quarter, respectively. 39



Figure 11: Robustness 2: IRF plots of Bitcoin prices to a unit shock of green asset prices in the pre- and post-COVID-19 with alternative green asset prices

Note: This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the specific variable when facing a unit shock to a target green asset at different time points, i.e., the first quarter, the half year, and the third quarter, respectively. 40



Figure 12: Robustness 3: IRF plots of Bitcoin prices to a unit shock of related financial indicators in the pre- and post-COVID-19 with changes in the sample period

*Note:* This figure shows the IRF plots before and during the pandemic as shown in panels (a) and (b), respectively. The gray dotted line, blue dashed line, and red solid line refer to the response of the specific variable when facing a unit shock to a target related financial indicator at different time points, i.e., the first quarter, the half year, and the third quarter, respectively.

	Mean	Std. Dev.	Minimum	Maximum	Skewness	Kurtosis
Panel A: Pre-Covid-19						
BTC	0.0222	3.5610	-14.9442	15.2987	-0.1216	3.6718
SWI	-0.0295	0.8499	-10.6051	2.8620	-4.5717	49.5546
ESGLSI	-0.0368	1.0774	-10.8401	4.4731	-2.8730	26.4139
GBI	0.0098	0.2101	-2.3671	0.8167	-2.3432	30.2399
GCEI	0.0177	1.1340	-12.4971	5.5528	-4.1089	43.9267
UCRY Policy	0.0022	0.1295	-0.9091	0.9041	0.2036	18.3164
UCRY Price	0.0021	0.1519	-1.1347	1.0598	0.3207	21.1604
ICEA	-0.0006	0.0611	-0.4383	0.7623	2.7226	57.5845
VIX	0.3171	7.2506	-19.8144	38.2167	1.3691	5.5387
Panel B: Post-Covid-19						
BTC	0.3473	3.7728	-13.4276	16.1041	0.1385	1.8977
SWI	0.0754	0.9812	-8.7034	7.6939	-0.5779	21.2194
ESGLSI	0.0883	1.4428	-12.9915	11.2908	-0.5905	21.8781
GBI	0.0124	0.2861	-2.4099	2.0127	-0.8740	15.6212
GCEI	0.1366	1.8595	-10.5579	11.0330	-0.2783	6.9841
UCRY Policy	0.0091	0.4410	-3.5751	4.4182	2.0752	37.0057
UCRY Price	0.0083	0.4646	-3.2736	3.9308	2.7952	35.0517
ICEA	0.0120	0.3533	-2.3672	5.6834	6.4870	127.4807
VIX	-0.1618	6.7453	-20.8405	48.0214	1.6034	9.2081

 Table 1: Descriptive statistics

*Note*: This table provides descriptive statistics of the variables used in this study, including the price series of Bitcoin (BTC), the four green financial assets, i.e., SWI, ESGLSI, GBI, GCEI, the two uncertainty index of the cryptocurrency market through aspects of policy and price, i.e., UCRY policy and UCRY price, the index of Cryptocurrency Environmental Attention (ICEA), and the CBOE Volatility Index (VIX).