

# How Does Economic Policy Uncertainty Affect CO2 Emissions? A Regional Analysis in China

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## Research Article

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# How does economic policy uncertainty affect CO<sub>2</sub> emissions? A regional analysis in China

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## Abstract

More recently, the COVID-19 pandemic outbreak has created massive economic policy uncertainty (EPU). EPU and its economic fallout have been a hot topic of study, however, the impact of EPU on CO<sub>2</sub> emissions has been seldom addressed to date. This paper investigates the direct impact of the EPU on CO<sub>2</sub> emissions and indirect effect via the environmental regulation at the national and regional levels using the panel data model and provincial panel data from 2003 to 2017 in China. The empirical results show that the central region is the most special one, which all explanatory variables except energy consumption are all non-significant even at the 10% level. For other samples, there is a significant positive correlation between EPU and CO<sub>2</sub> emissions, whether in the national or regional level. Additionally, environmental regulation alone can achieve the purpose of curtailing carbon emissions. However, when the EPU is taken into consideration, environmental regulation exerts a significantly positive effect on CO<sub>2</sub> emissions, leading to unintended increase in emissions. Moreover, the Environmental Kuznets Curve (EKC) hypothesis was confirmed in the national and eastern samples, while CO<sub>2</sub> emissions increase monotonically as economic level grows for western datasets. Based on the overall findings, some policy implications were put forward.

**Keywords** CO<sub>2</sub> emissions · Economic policy uncertainty · Environmental regulation · Panel data models

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## Abstract

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## Introduction

Climate change due to global warming is one of the defining issues of our time. Global temperature has been in constant rise since the pre-industrial era. Its well-known reason is the steady climb of greenhouse gas (GHG) emissions, particularly carbon dioxide (CO<sub>2</sub>) which is now 50% higher than pre-industrial levels. The member states of the United Nations have agreed to limit the increase in the global average temperature to 1.5°C, which requires substantial worldwide commitments to achieve these goals. As the world's largest carbon emitters, China has been paid more and more attention from all over the world on the issue of how the country to reduce greenhouse gas emissions and mitigate the climate change. Clarifying the impact of factors behind carbon emissions is significant for efforts to meet this challenge.

There is an extensive body of literature that analyzes the mechanism behind carbon emissions. Its well-established root cause is the consumption of a large amount of fossil fuel energy and economic activity human-made. Ehrlich and Ehrlich (1970) first propose the Impact = Population · Affluence · Technology (IPAT) framework to analyze environmental changes affected by anthropogenic driving forces. Dietz and Rosa (1994) reformulate the IPAT framework by considering nonproportional effects of driving forces, raising the stochastic version of this model, namely, STIRPAT (stochastic impacts by regression on population, affluence and technology). The stochastic model has been actively applied to examine all other factors affecting human impact on the environment for both developed and developing countries, such as industrial structure (Dong et al. 2020, Wang et al. 2016, Yu et al. 2018), foreign direct investment (FDI) (Demena and Afesorgbor 2020, Mahadevan and Sun 2020, Shahbaz et al. 2019, Tang and Tan 2015) and urbanization (Elliott and Clement 2014, Huo et al. 2020, Khan et al. 2019, Makido et al. 2012, Yao et al. 2018).

A considerable number of studies have investigated the nexus between CO<sub>2</sub> emissions and economic growth. Grossman and Krueger (1991) develop Environmental Kuznets Curves (EKC) to emphasize this nexus. Resembling Kuznets Curve of inequality (Kuznets 1955), the EKC also postulates an inverted-U hypothesis. According to EKC, environmental pollution or CO<sub>2</sub> emissions increase as economic level grows at the beginning and then decrease after a certain turning point. Since then, with various datasets and complex econometric methods employed, the EKC has been tested extensively, but the validity of itself is debatable. In the context of the EKC framework, the nonlinear relationship can be explained by three effects, that is, income, structural and abatement effect (Jaunky 2011). The results verified in many previous studies often appear to be contradictory and inconclusive both theoretically and empirically. The empirical studies like Chiu (2012), Farhani et al. (2014) confirm the existence of the EKC. The empirical findings of Alshehry and Belloumi (2017), Baek (2015), Fernández-Amador et al. (2017), Haq et al. (2016), however, fail

68 to provide such positive support. Moreover, Rodríguez et al. (2016) show a monotonic and positive relationship between  
69 environmental quality indicators and income. As mentioned, no clear conclusion can be drawn.

70 In the process of achieving green development and high-quality economic growth, environmental regulation plays an extremely  
71 important role. During the “Outline of the Thirteenth Five-Year Plan for National Economic and Social Development of the People’s  
72 Republic of China” (13th Five-Year Plan) periods, the Chinese government promulgated a series of policies to strengthen  
73 environmental protection and build the ecological civilization. The relationship between environmental regulation and CO<sub>2</sub>  
74 emissions has been fiercely debated in recent years. As evidenced, although the field has accumulated a well-developed literature,  
75 there is no a uniform conclusion. According to the literature, there are two perspectives on this issue, that is, the “green paradox”  
76 and “curbing effect”. The concept of green paradox is first put forward by Sinn (2008), which is defined as a myriad of measures  
77 launched by politicians aimed at alleviating CO<sub>2</sub> emissions, eventually exacerbate the problem. Sinn attributes counter-productive  
78 policies to psychological expectation that the environmental regulation will gradually be tightened, resulting in accelerating the  
79 exploitation of fossil energy. Other researchers ascribe the emergence of the green paradox to unreasonable set of carbon taxes  
80 (Edenhofer and Kalkuhl 2011), support policies for alternative energy (van der Ploeg and Withagen 2012), policy's implementation  
81 lags (Di Maria et al. 2012) and unilateral climate policies (Sen 2016). On the contrary, other scholars argue that there is a significant  
82 positive correlation between environmental regulation and environmental pollution, which is also known as the “curbing effect”  
83 hypothesis. Zhang et al. (2017), Zhao et al. (2015) indicate that environmental regulations exert a positive effect on reducing carbon  
84 emissions. Nevertheless, some studies hold a more neutral viewpoint. Wang (2018) suggest that there is an inverted U-shaped curve  
85 relationship between environmental regulation and carbon emissions. Besides, some researchers confirmed that environmental  
86 regulations have indirect effects on carbon emission through transmission paths, such as industrial structure (Yang et al. 2020),  
87 energy structure (Wu et al. 2020), technological innovation (Cheng et al. 2017, Pei et al. 2019) and FDI (Wang et al. 2018). The  
88 influence mechanism, however, is paradoxical.

89 After perusing the aforementioned studies, we have found that researches on the driving forces of CO<sub>2</sub> emissions are relatively  
90 mature, while the macroeconomic institutional factor reflected by EPU has been neglected. EPU and its economic fallout have  
91 spurred an ongoing discussions and everlasting interest in academic field. Resembling the concept of risk, EPU refers to the inability  
92 of economic entities to forecast the timing, content, and potential economic consequences of policy decisions (Gulen and Ion 2015).  
93 Driven mainly by unpredictable political cost, EPU is an inevitable by-product of policy-making (Pastor and Veronesi 2012). While,  
94 it’s a daunting task to measure the overall level of EPU in the economy. A news-based index proxy for EPU developed by Baker et  
95 al. (2016) (henceforth BBD) makes it possible to capture aggregate economic policy uncertainty. Due to publicly available and  
96 scientific measure of the uncertainty present in the economy, the BBD index is widely accepted. Moreover, a coherent body of  
97 empirical research on the topic of this index has been conducted.

98 Understanding exactly how sensitive EPU is to CO<sub>2</sub> emissions is critically important for efforts to curtail emissions. However,  
99 there is a dearth of empirical literature on how EPU directly or indirectly affects CO<sub>2</sub> emissions. Based on US sector data, Jiang et  
100 al. (2019) implement a novel parametric test of Granger causality to assess the impact of EPU on CO<sub>2</sub> emissions and found that  
101 there is a Granger-causality running from the US EPU to carbon emissions. Pirgaip and Dincergok (2020) employ a bootstrap panel  
102 Granger causality test to investigate the causal relationship between EPU and energy consumption and CO<sub>2</sub> emissions in G7  
103 countries. They argued that EPU has adverse effects on energy saving and emission reduction. Adedoyin and Zakari (2020) examine  
104 the role of EPU played in the nexus between energy consumption, economic and CO<sub>2</sub> emissions in the UK. The results indicate that  
105 EPU plays a critical role in the effort of mitigating CO<sub>2</sub> emissions. Additionally, Adams et al. (2020) using data from economies  
106 characterized by resource-rich but crisis-prone demonstrate that there is a significant association between geopolitical risk, EPU,  
107 energy consumption, economic growth and CO<sub>2</sub> emissions in the long-run. This implies that higher levels of EPU adversely affect  
108 carbon abatement. This finding is consistent with the result of Anser et al. (2021). By constructing China provincial EPU index, Yu  
109 et al. (2021) indicate that China's provincial EPU level exerts a significantly positive impact on manufacturing firms' carbon  
110 emission intensity. The empirical analysis results show that manufacturing firms incline to choose traditional cheaper and dirty  
111 energy when the level of uncertainty rising. Besides, Abbasi and Adedoyin (2021) indicate that economic policy uncertainty has an  
112 insignificant impact on China’s CO<sub>2</sub> emissions.

113 The present paper contributes to the previous literature as follows. First, we account for EPU in the environmental regulation–  
114 CO<sub>2</sub> emissions nexus to examine the interaction effects of EPU. Understanding the impact of institutional factors behind carbon

emissions is a matter of utmost significance to realize a low-carbon economy. Second, previous studies have neglected the regional heterogeneity in China's CO<sub>2</sub> emission. The paper analyzes the interaction effects between core variable components based on relevant data of 30 provinces in China from 2003 to 2017. Accordingly, the objective of the study is to clarify how EPU directly affects CO<sub>2</sub> emissions and elucidate how the EPU affects the relationship between environmental regulation and CO<sub>2</sub> emissions. Based on panel data methods and STIRPAT model, this research takes EPU into account and examines the direct or indirect effect of EPU on CO<sub>2</sub> emissions.

The paper is structured as follows: Section 2 introduces the theoretical background and hypothesis of the study, Section 3 presents the data and methods. Section 4 demonstrates the empirical results and discussions. Section 5 concludes alongside with policy implications.

## **Theoretical background and hypothesis**

More recently, the COVID-19 pandemic outbreak has created massive economic uncertainty (Jordà et al. 2020). Without question, the EPU is significant to the economic system. EPU has dampening impact both at macro level, i.e. economic growth (Baker et al. 2016) and at micro level, i.e. corporate investments (Akron et al. 2020, Gulen and Ion 2015), corporate innovation (He et al. 2020, Xu 2020), household consumption (Aaberge et al. 2017, Levenko 2020) and stock market (Arouri et al. 2016, Das et al. 2019). In all, uncertainty reshapes the environment in which the economic entities operate. Besides, when the external environment changes, the decision of economic entities that link to carbon emissions is altered. To summarize, we speculate that EPU may have an impact on CO<sub>2</sub> emissions.

Actually, the EPU has harmful effect on the whole economic and impeded the recovery from recession (Baker et al. 2016). The depressed economic situation inevitably leads to consumers curbing household consumption and enterprises postponing corporate investment. With the economic condition changed, the demand for energy consumption will be affected. Traditional cheaper energy, for instance, coal and oil would be chosen by more industries due to sluggish economic situation (Jiang et al. 2019, Yu et al. 2021). Undoubtedly, the more cheap and dirty fossil fuels energy consumed, the more carbon emissions. In addition, the level of uncertainty has a certain degree of impact on energy prices, which could affect consumers' energy choices. Balcilar et al. (2017) imply that EPU is an important driver of oil price volatility and there is an asymmetric connection between uncertainty and oil returns. Yang (2019) demonstrates that oil price shocks are significantly influenced by EPU and the causality relationship between them intensifies as time scales increase. Olanipekun et al. (2019) show that there is an asymmetric causal relationship between gasoline prices and EPU in the sampled countries. On the one hand the reduction in consumer production and living caused by higher level of EPU has partly reduced carbon emissions, on the other hand the use of more cheap and dirty energy increases carbon emissions. The net effects are unclear.

A plethora of works explore that the economic policy-related uncertainty depresses corporate investment (Chen et al. 2019, Dibiasi et al. 2018, Liu and Zhang 2020). What is more, higher level of uncertainty encourages firms to increase their cash holdings (Cheng et al. 2018, Demir and Ersan 2017, Phan et al. 2019). On the basis of the real option theory, corporate investment projects are irreversible to varying degrees and usually costly, economic policy uncertainty increases the value of the waiting option, so firms become more cautious and prefer to delay investment until getting more information for decision making, in the face of greater uncertainty (Bernanke 1983, Pindyck 1991). The main contributory paths of EPU affecting the corporate investment are attributed to financial market frictions (Liu et al. 2020a). Higher uncertainty increases external financing costs and the debt default risk. The investment of renewable energy industry for example hydropower, wind, and biomass has the nature of higher degree of irreversibility, which usually have the characteristics of larger initial investment scale and longer investment return cycle. That is, renewable energy's investment lead to higher sunk costs and renewable energy enterprises inevitably delay investment when economic policy uncertainty higher. There is no doubt that the decrease in investment, especially in the renewable energy industry, has increased carbon emissions.

Furthermore, according to signal theory, the increase of carbon information disclosure can reduce the burden of relevant government and effectively promote energy saving and emission reduction (Connelly et al. 2011). However, firms have more incentive to reduce the transparency of carbon information disclosure when the EPU is at high level. By incorporating difference-in-difference (DID) estimation, Pan et al. (2020) find that policy uncertainty has a more significant inhibition effect on corporate

162 environmental information disclosure. Beyond question, a reduction in carbon disclosure will increase carbon emissions.

163 Based on the foregoing analytic mechanisms commented, this study proposes the following hypothesis.

164 Hypothesis 1a: The EPU index has a negative correlation with CO<sub>2</sub> emissions.

165 Hypothesis 1b: The EPU index has a positive correlation with CO<sub>2</sub> emissions.

166 As briefly analyzed in the introduction, the influencing mechanism between environmental regulation and CO<sub>2</sub> emissions is paradoxical. Undoubtedly, the unpredictability of policy changes, including environmental policies is a major source of uncertainty. 167 With high level of uncertainty, the attention on green development from government will be distracted, resulting in environmental 168 protection policies executed poorly. The governments' determination to reduce emissions may be doubted, which in turn leads to 169 the efforts for carbon abatement by enterprises relaxed, eventually resulting in the increase of emissions. We infer that the indirect 170 effect of EPU on CO<sub>2</sub> emissions may be through environmental regulation, that is to say, EPU exerts interaction effects on the 171 environmental regulation–CO<sub>2</sub> emissions nexus. Guo et al. (2019a) contend that uncertainty has a significantly effect on 172 environmental policy making. They argued that both underestimation and overestimation of uncertainties make against to formulate 173 a reasonable carbon mitigation design by policy makers. Lecuyer and Quirion (2019) argue that high level of uncertainty has a 174 positive incentive effect on renewable energy subsidies effectiveness, while these subsidies are not welfare-improving at the low 175 level of uncertainty. Considering the forward-looking policy uncertainty, Contreras and Platania (2019) put forward a zero mean 176 reverting model to estimate the effectiveness of the London Environment Strategy. Against this backdrop, the EPU is a key 177 determinant in understanding the impact of environmental regulation on CO<sub>2</sub> emission. Based on the discussion above, this study 178 puts forward the following hypothesis:

180 Hypothesis 2: EPU plays an important role in moderating the relationship between environmental regulation and CO<sub>2</sub> emissions.

## 182 **Data and methods**

### 184 **Data**

185  
186 In the present study, considering the availability and completeness of the data, 30-provincial panel data from 2003 to 2017 in China 187 are investigated. Given that the CO<sub>2</sub> emissions are heterogeneous in different regions, these 30 provincial-level administration 188 regions can be divided into three groups, namely, eastern, central, and western regions, according to the level of economic and social 189 development. The three economic zones are at different stages of economic development. Due to this, the subsequent analysis is 190 conducted at both national and regional level.

191 The CO<sub>2</sub> emission data were collected from the China Emission Accounts and Datasets (CEADS) website. Sponsored by domestic 192 and international professional bodies research institutes, such as National Natural Science Foundation of China, Science and 193 Technology Research Council UK, CEADs devotes to China's emission accounting methods and applications (Liu et al. 2020b). 194 The carbon emission datasets published by CEADs are characterized by fully open access, multi-scale coverage and free to download. 195 In addition to the traditional energy consumption carbon emission database, CEADs database also covers the detailed data of carbon 196 emissions from industrial processes, including cement production, lime production, glass, and so on. The emission data is accurate 197 and most up-to-date, effectively avoiding measurement errors.

198 The regional difference of CO<sub>2</sub> emissions can be clearly seen from Fig.2. The fluctuation trend of emissions is generally on the 199 rise. National CO<sub>2</sub> emissions increase roughly one and a half times throughout the study period. Due to superior in resource 200 endowment, CO<sub>2</sub> emissions in the eastern region has been in a dominant position in 2003–2017. The western region's emission 201 levels have exceeded those of the central region since 2013. Since 2011, the growth of CO<sub>2</sub> emissions has become slower. Apparently, 202 CO<sub>2</sub> emissions in China has been effectively controlled.

203 Based on (Baker et al. 2013b), the BBD index is a weighted average of three components, namely, the news-based component, 204 the tax component and the forecaster disagreement component. The weights are one-half, one-sixth and one-third, respectively. BBD 205 go to great lengths to verify the BBD index is a reliable and accurate picture of EPU. According to the newspaper-based methods, 206 the EPU indices have been constructed for major economies including China. Based on the South China Morning Post (SCMP), 207 Hong Kong's leading English-language newspaper, Baker et al. (2013b) develop a monthly EPU index for China, running from 208 January 1995 to the present. The index may have some limitations. The Hong Kong-based newspaper may not overall picture the

level of EPU in China and only one newspaper in the sample may magnify measurement error that could induce a bias in empirical analysis. To conquer those shortcomings, following BBD approach, Davis et al. (2019) (henceforth DLS) construct monthly indices of EPU for China by using two mainland Chinese newspapers: the Renmin Daily and the Guangming Daily.

The two indices are compared in Fig. 1. Both indices look like a sensible proxy for true policy-related economic uncertainty. In addition, the data show that the EPU curve of China is not smooth, and the phenomena of large growth and the sharp decline occurring alternately. They tend to move together with spikes around events that are ex ante predicable to increase uncertainty, such as financial crises, stimulus package, elections. Currently, these two indices rose to historic highs after the global pandemic. However, there are significant divergences between two indices. As a result, this study takes the DLS index as the benchmark indicator.

In this paper, we use cost-based environmental regulation indices as the proxy variable for the level of environmental regulations (Guo et al. 2019b, Hu and Wang 2020, Lanoie et al. 2008, Wang et al. 2019, Wu et al. 2020, Yang et al. 2020). We adopt the ratio of industrial pollution governance expenditures collected by each province to its industrial added value (denoted as ER) as the environmental regulation variable. It is easy to find that the cost-based index focus on the expenditure (input) of pollution control in each province. There is a positively correlation between cost-based index and the enforcement of government environmental regulation. The larger the value of the index, the greater environmental regulation intensity. Therefore, pollution control expenditure or treatment investments can well reflect the environmental regulation intensity.

To eliminate the impact of price factors, all variables related to the price index, including environmental regulation, per capita GDP, are converted to the constant 2000 price. Energy consumption is expressed by the total energy consumption of each province at the end of the year. In order to be consistent with the time horizon, the monthly Chinese DLS EPU index is converted into the annual EPU index by the arithmetic average method. The 30-province shares the same average EPU index. In addition, this study takes Chinese BBD EPU index as the auxiliary index measuring the level of economic policy uncertainty to test results robustness.

The data sources are described in Table 1. Table 2 shows the descriptive statistics for all variables.

## Methods

The primary purpose of this paper is to shed light on the mechanisms how uncertainty affects the CO<sub>2</sub> emissions. By examining the direct or indirect effect of EPU on CO<sub>2</sub> emissions, we establish the following models based on panel data methods and STIRPAT model. CO<sub>2</sub> emissions were selected as the explanatory variable. Economic policy uncertainty and environmental regulation were the core explanatory variable.

$$CO2 = f(RGDP, RGDP2, ENC, EPU, ER)$$

$$\ln CO2_{it} = \alpha_0 + \beta_1 \ln RGDP_{it} + \beta_2 RGDP2_{it} + \beta_3 \ln ENC_{it} + \beta_4 \ln EPU_{it} + \beta_5 \ln ER_{it} + \lambda_t + \mu_i + \varepsilon_{it} \quad (1)$$

$$CO2 = f(RGDP, RGDP2, ENC, EPU, ER, EPU * ER)$$

$$\ln CO2_{it} = \alpha_0 + \beta_1 \ln RGDP_{it} + \beta_2 RGDP2_{it} + \beta_3 \ln ENC_{it} + \beta_4 \ln EPU_{it} + \beta_5 \ln ER_{it} + \beta_6 \ln EPUER_{it} + \lambda_t + \mu_i + \varepsilon_{it} \quad (2)$$

Equation (1) is constructed to elucidate the directly effect of EPU on CO<sub>2</sub> emissions. Based on equation (1), equation (2)

introduced the concept of  $EPU_{it} * ER_{it}$ , namely the interacted items of EPU and ER to examine the interaction effects of EPU on the environmental regulation–CO<sub>2</sub> emissions nexus. To ensure the linearity of the model and eliminate the heteroscedasticity phenomenon, logarithmic processing is carried out for all variables.

Herein, lnCO<sub>2it</sub> means the carbon emission per capita; lnRGDP and lnRGDP2 denote Real Gross Domestic Product per capita in logarithms and its squared term. The quadratic term of RGDP is incorporated in the models to test the EKC hypothesis in the context of China. lnENC is energy consumption; lnER expresses environmental regulation; lnEPU represents economic policy uncertainty, where subscripts *i* and *t* denote the *i*-th province and the *t*-th year, respectively.  $\alpha_0$  is the constant term;  $\beta_1, \beta_2, \dots, \beta_6$  are the evaluation coefficients, with  $\beta_5$  and  $\beta_6$  being the focus of interest coefficients in this study.  $\lambda_t$  and  $\mu_i$  are time fixed effects and individual fixed effects respectively.  $\varepsilon_{it}$  is the residual error term.

251  
252 **Results and discussion**  
253

254 Table 3 presents the Pearson correlation matrix to capture the intensity of the correlation between the variables and indicate no high  
255 correlations. As it is shown, both the EPU and the level of environmental regulation are positively and significantly correlated with  
256 CO<sub>2</sub> emission, while the relationship between them are negative. Besides, energy consumption and environmental regulation is  
257 insignificantly related.

258 In this paper, 30-province panel data from 2003 to 2017 in China are employed. Normally, in terms of short panel data, there are  
259 three types of estimation method: pooled regression, fixed effects model and random effects model. This paper performs F test and  
260 Hausman (1978) test to choose the most appropriate estimation method. The F test is run to inspect individual specific effects with  
261 the null hypothesis, that is all  $\mu_i = 0$ . The lower the p-value of the F-test statistic, the higher the significance level. With the null  
262 hypothesis is rejected, the estimator of fixed effects model is more efficient than the pooled regression. The Hausman test is applied  
263 to determine whether to choose the fixed effects model or the random effects model. Similarly, the larger Hausman test value means  
264 that the corresponding p-value is smaller than preset significance levels. The null hypothesis should be rejected and a fixed effects  
265 model established. Table 4 reports the F test results. The empirical results of Hausman test are shown in Table 5.

266 As shown in Table 4 and Table 5, the results of the F test and the Hausman test implied that the null hypothesis was strongly  
267 rejected, suggesting that the fixed effects model should be selected to examine the directly or indirectly effect of EPU on CO<sub>2</sub>  
268 emissions for the whole nation and the three regions.

269 This paper decomposes the effect of the EPU on CO<sub>2</sub> emissions from the national and regional level, respectively. We mainly  
270 focus on the empirical results of Equation 2. The empirical results are reported in Table 6. The columns present the results of the  
271 whole dataset and the three regional datasets, respectively.

272 As shown in Table 6, the empirical results of the fixed effect estimation at the national level show that all the explanatory variables  
273 are statistically significant at 1% level (Eq. (2)). While in model 1, all explanatory variables are statistically significant except for  
274 the EPU. Due to natural logarithmic form taken in this model, the coefficients of the explanatory variables could be directly  
275 explained as elasticities. According to the estimation results of Eq. (2), the EPU positively increases the carbon emissions. A 1%  
276 increase of uncertainty would increase emissions 0.3326% in China when other variables remain constant. The estimate results  
277 conform to both our expectations and the findings of previous studies (Adams et al. 2020, Pirgaip and Dincergok 2020). They argued  
278 that EPU has adverse effects on emission reduction. In accord with economic intuition, economic policy uncertainty disrupts  
279 economic situation, inhibiting household consumption and corporate investment, especially in the renewable energy industry. All of  
280 consequences do harm to economic restructuring, industrial upgrading and green innovation. Depressed economic environment  
281 drives enterprises to choose traditional cheaper and dirty energy for production, leading to increased carbon emission. Ultimately,  
282 the EPU exert negative effect on CO<sub>2</sub> emissions. Other possible explanation for the negative correlation was that higher uncertainty  
283 has a more significant inhibition effect on corporate environmental information disclosure. The less environmental disclosure  
284 information, the worse it is for carbon reduction. Thereby, hypothesis 1a was verified.

285 Our results show that environmental regulation was negatively correlated with CO<sub>2</sub> emissions. This implies that environmental  
286 regulation tools meet the basic requirements of emissions reduction at the national level. Through a forced mechanism, the  
287 improvement of environmental regulation is conducive to mitigate emissions. China's current legal and supervision systems on  
288 environmental protection have being more and more adequate with various regulations and measures promulgated. This result is in  
289 accord with the findings of Pei et al. (2019), Zhang et al. (2017). Furthermore, this paper uses the interactive term (EPU\*ER) to  
290 capture the effect of uncertainty on the functional mechanisms of environmental regulation. Meanwhile the signs of interactive term  
291 are positive and statistically significant, indicating that under the influence of uncertainty, environmental regulations have  
292 significantly promoted CO<sub>2</sub> emissions and that a green paradox occurs. The possible reasons why EPU exerts positive impact upon  
293 the environmental regulation-CO<sub>2</sub> emissions nexus is that the priority for governments is economic stability rather than  
294 environmental conservation or other environmental policies under high level of uncertainty, resulting in the standards of  
295 environmental policy implementation relaxed and carbon emissions increased. Therefore, hypothesis 2 is verified.

296 For the control variables, the elasticity of energy consumption is the greatest (0.8495), indicating that a 1% increase in energy  
297 consumption level would lead to 0.8495% increase in CO<sub>2</sub> emissions. GDP growth rate are positively associated with CO<sub>2</sub> emissions.

298 By judging from the quadratic term of per capita GDP, the regression coefficients are negative at the significance level of 1%. In  
299 other words, the environmental Kuznets curve (EKC) theory holds.

300 Considering the typical regional differences in China, the relationship between considered variables may also have regional  
301 heterogeneity characteristics. Thus, we should analyze the effects of the EPU upon emissions at the region level. Compared with  
302 the national aggregate case, the eastern region has similar findings. The influences of variables appear to be greater due to high  
303 economic intensity and population density in this region. In the eastern region, all the independent variables are statistically  
304 significant at the 1% level (Model 4). The elasticity of energy consumption is 0.5923, which is the lowest in the sample selected.  
305 Environmental regulation passes t-test with elasticity of -0.418. That is to say, for every 1% increase in environmental regulations,  
306 carbon emissions fall by 0.418%. The EPU index has a significant role in promoting eastern CO<sub>2</sub> emissions, taking the value of  
307 0.5802. From the perspective of the moderating effect of uncertainty, the coefficient of interaction term between EPU and  
308 environmental regulation is significantly positive, indicating that uncertainty can increase CO<sub>2</sub> emissions from environmental  
309 regulation.

310 As shown in Column 6, the central region is quite different. It is worth noting that all explanatory variables except energy  
311 consumption are all non-significant even at the 10% level. One possible explanation is that the heterogeneity of sample distribution  
312 causes insignificance of most variables. Owing to fast urbanization and industrialization process, the elasticity of energy  
313 consumption in the central region is the highest (0.9600). In the western region, energy consumption is the biggest element with  
314 elasticity of 0.8456, followed by uncertainty. The interactive term (EPU\*ER) has similar story to that of the nation level and the  
315 eastern region. The findings indicate that environmental regulation can effectively mitigate CO<sub>2</sub> emissions without consideration of  
316 uncertainty. However, when the EPU taken into consideration, environmental regulation exerts a significantly positive effect on CO<sub>2</sub>  
317 emissions, leading to unintended increase in emissions. Environmental regulations and the EPU are statistically significant with  
318 elasticity of - 0.3006 and 0.5060, respectively. It indicates that the increase of the intensity of environmental regulations is conducive  
319 to improve carbon emission reduction, while the positive sign of uncertainty means that high level of EPU would definitely lead to  
320 more carbon emissions. Both environmental regulation and EPU index are lower than that of the eastern case. In a word, it can be  
321 found from Table 6 that uncertainty and the interactive term (EPU\*ER) have significantly contributed to the increase in CO<sub>2</sub>  
322 emissions, while environmental regulations intensity shows a significantly negative association with CO<sub>2</sub> emissions, whether in the  
323 nation level or the regions.

324 The variable GDP per capita presents a positive and significant effect on emissions for eastern and western regions except for  
325 central region for which the relation isn't significant. In the eastern region, the coefficient of GDP per capita is the largest (0.7243),  
326 which is much higher than other samples. It suggests that there is a strong link between economic growth and CO<sub>2</sub> emissions. The  
327 coefficient of the quadratic term of GDP per capita is negative and statistically significant in the eastern region, while it is statistically  
328 nonsignificant in the central and western regions. This means that the results in the eastern region is in line with the theory of EKC,  
329 however the inverted U-shaped relationship between economic growth and emissions cannot hold any more in the central and  
330 western regions.

## 332 **Robustness test**

333  
334 In this study, the auxiliary economic policy uncertainty variable (EPU2) developed by Baker et al. (2016) is used for the robustness  
335 test. The results of robust test are shown in Table 7. The sign and significance of these variables' coefficients have seldom changed.  
336 Overall, the findings are consistent with the results of Table 6. Therefore, it can be concluded that the aforementioned results of  
337 empirical analysis are relatively robust.

338 Despite informative findings yielded, this study has some limitations that are worth noting. First of all, this study employed a  
339 same set of EPU indices as proxy for Chinese provinces uncertainty level, which is not in line with the reality. For better estimating  
340 the impact of economic policy uncertainty on CO<sub>2</sub> emissions, more rational uncertainty accounting for Chinese provinces level is  
341 expected in future studies. Moreover, economic policy uncertainty could affect CO<sub>2</sub> emissions through many channels, from which  
342 this paper only considers the elements of environmental regulation, and other factors such as energy consumption, foreign direct  
343 investment would be studied in future research. Second, the cost-based environmental regulation indices only reveal one aspect of  
344 environmental governance. Therefore, a more reasonable and comprehensive index to fully reflect the level of environmental

345 regulations would become increasingly important. Furthermore, this study only employed a conventional econometric method,  
346 further research would take spatial spillover effects into consideration. The exploration of spatial spillover effects is conducive to  
347 the in-depth study of the driving forces behind CO<sub>2</sub> emissions (Feng et al. 2020, Hu and Wang 2020, Meng et al. 2017).  
348

## 349 **Conclusions and policy implications**

350  
351 Using panel data of 30 provinces of China from 2003 to 2017, this study empirically analyzes the direct impact of EPU and the  
352 indirect impact of environmental regulation on CO<sub>2</sub> emissions. The main conclusions can be summarized as follows.

353 At the national level, the EPU positively increases the carbon emissions and environmental regulation alone is negatively  
354 correlated with CO<sub>2</sub> emissions. However, under the influence of the EPU, environmental regulations significantly promote carbon  
355 emissions, causing a green paradox. That is to say, economic policy uncertainty provided interaction effects between environmental  
356 regulations and CO<sub>2</sub> emissions. The mechanism of the interaction effects was that the high level of uncertainty will lead to the  
357 current environmental regulations executed poorly, ultimately promoting CO<sub>2</sub> national emissions.

358 At the region level, the central region is the most special one. All explanatory variables except energy consumption are all non-  
359 significant even at the 10% level. The eastern and western regions have the similar findings to the whole nation. Moreover, the  
360 results demonstrate that the re-examination of the EKC hypothesis is inconclusive. Kuznets relationship between economic growth  
361 and CO<sub>2</sub> emissions for the national and eastern samples was confirmed. While CO<sub>2</sub> emissions rises monotonically as GDP grows  
362 for western datasets.

363 Given the main results we obtained, the related policy implications can be drawn. First and foremost, this study demonstrated that  
364 high level of uncertainty exerts negative impact upon mitigating emissions, resulting in environment deteriorate. For this reason,  
365 the government should do their best to minimize the adverse effects of uncertainty. Unexpected spike in the level of economic policy  
366 uncertainty could not affect the credibility of government. The determination to environmental conservation and emission abatement  
367 of government would not be shaken even under high levels of uncertainty. Necessary supportive measures should be taken into  
368 consideration to decrease uncertainty and maintain transparency and stability of policies.

369 Besides, environmental regulations play an important role in the emission abatement. The intensity of environmental regulations  
370 must be appropriate. An overly restrictive emission-reduction policy would be counterproductive. Besides, the environmental  
371 regulations should not be one size fits all, due to the discrepancies of economic development and energy resource endowments.  
372 Therefore, different regions should implement different environmental policies according to regional conditions. Moreover, these  
373 regulations should be applied without hampering policy stability, since economic policy uncertainty goes against to the effectiveness  
374 of environmental regulations and the achievement of carbon mitigation targets. In addition, environmental protection is not just a  
375 matter for the government. The role of public participation and supervision should be emphasized. Local residents should improve  
376 the awareness of energy conservation and low-carbon lifestyle. Furthermore, there is a need to strengthen the interregional  
377 cooperation and communication on carbon reduction experiences and practices to maximize technology spillovers effects. All  
378 regions should work hand in hand to tackle the issue.

379 Last but not least, the empirical findings uncovered the adverse effect of energy consumption on CO<sub>2</sub> emissions. Accordingly, the  
380 government concerned should promote the use of renewable energy or clean energy, for instance, nuclear power, hydropower, wind  
381 power to optimize energy consumption structure and lessen the dependence on fossil fuel energy. Meanwhile, the government should  
382 increase the fiscal support and stimulate capital investment for low-carbon technologies innovation to improve energy efficiency,  
383 decoupling economic growth from environmental pollution.

## 384 **Declarations**

### 385 **Ethics approval and consent to participate**

386 Not applicable

### 387 **Consent for publication**

388 Not applicable

### 389 **Availability of data and materials**

390 The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request.  
391

## 392 **Competing interests**

393 The authors declare that they have no competing interests.

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## 397 **Authors' contributions**

398 All authors contributed to the study conception and design. Material preparation, data collection and analysis were performed by  
399 YL and ZPZ. The first draft of the manuscript was written by ZPZ and all authors commented on previous versions of the manuscript.  
400 All authors read and approved the final manuscript.

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# Figures

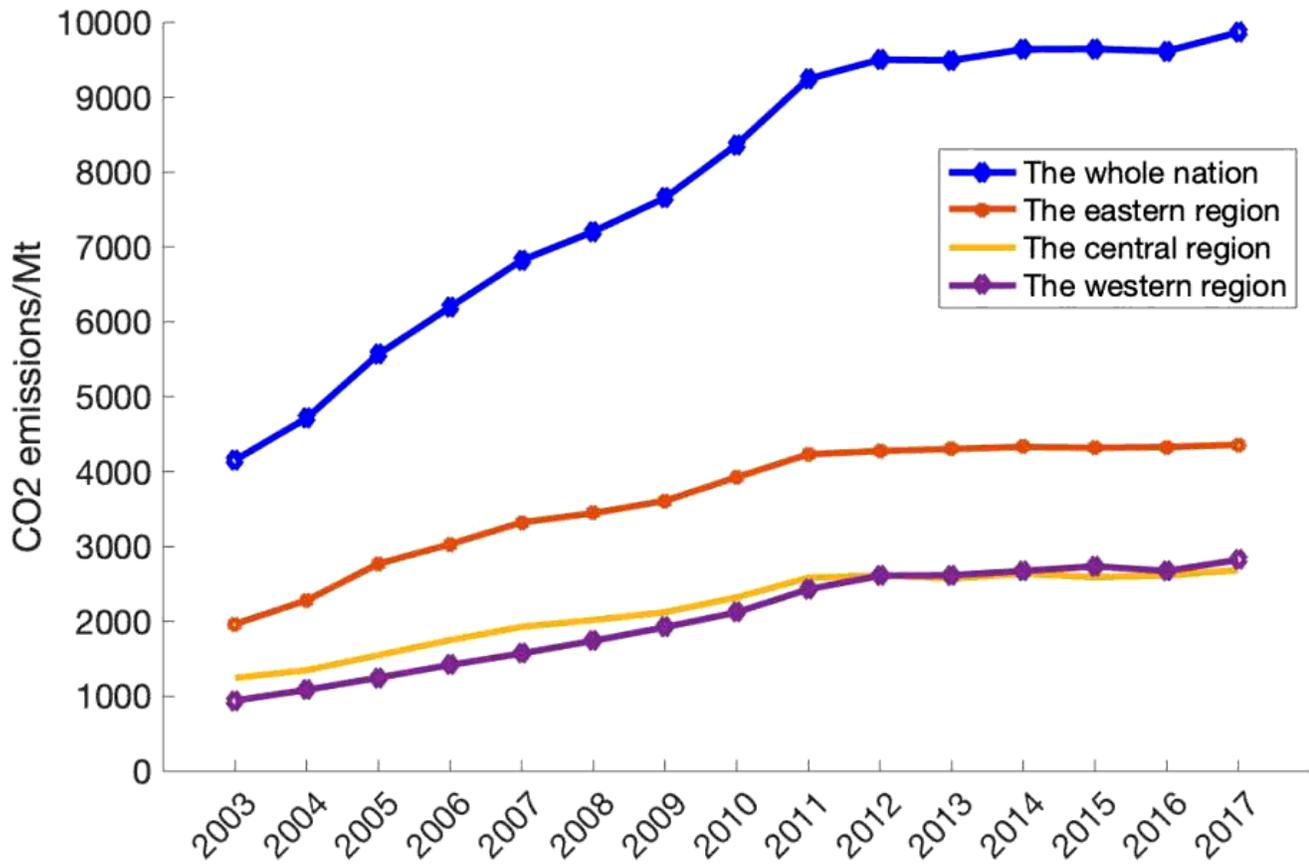
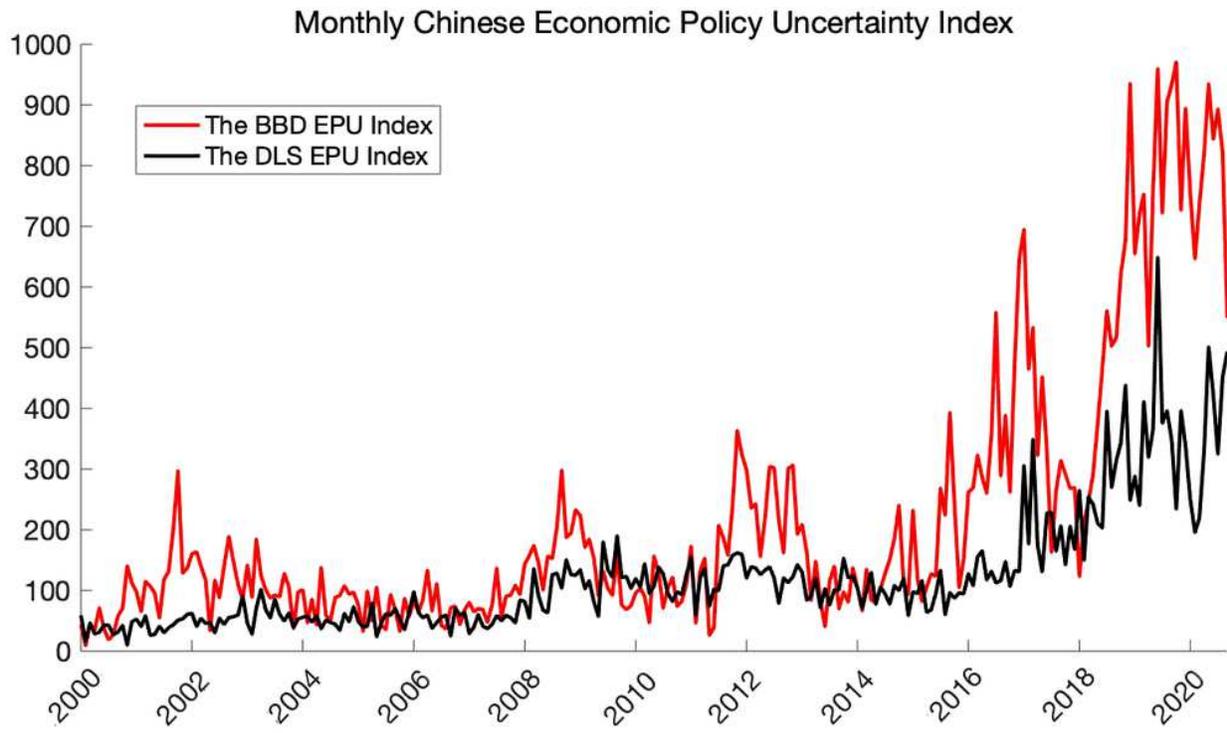


Figure 1

Changes in China's national and regional CO2 emissions from 2003 to 2017



**Figure 2**

The figure shows monthly Chinese policy-related economic uncertainty Index from January 2000 to October 2020