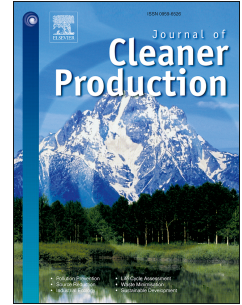


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CRediT author statement

Li: Conceptualization, Methodology, Software, Data curation, Writing- Original draft preparation. **Qi:** Supervision, Resources, Validation, Funding acquisition.
Shi: Conceptualization, Validation, Writing- Reviewing and Editing, Funding acquisition.

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The COVID-19 pandemic and energy transitions: Evidence from low-carbon power generation in China

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Abstract: The Corona Virus Disease 2019 (COVID-19) has led to a decline in carbon emissions or an improvement in air quality. Yet little is known about how the pandemic has affected the “low-carbon” energy transition. Here, using difference-in-differences (DID) models with historical controls, this study analyzed the overall impact of COVID-19 on China’s low-carbon power generation and examined the COVID-19 effect on the direction of the energy transition with a monthly province-specific, source-specific dataset. It was found that the COVID-19 pandemic increased the low-carbon power generation by 4.59% (0.0648 billion kWh), mainly driven by solar and wind power generation, especially solar power generation. Heterogeneous effects indicate that the pandemic has accelerated the transition of the power generation mix and the primary energy mix from carbon-intensive energy to modern renewables (such as solar and wind power). Finally, this study put forward several policy implications, including the need to promote the long-term development of renewables, green recovery, and so on.

Keywords: the COVID-19 pandemic; low-carbon power generation; energy transition; China;

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32

33 Highlights

34 The overall impact of COVID-19 on energy transitions from the perspective of
35 low-carbon power generations was analyzed

36 By using the stacked data, the COVID-19 pandemic increased the low-carbon
37 power generation by 4.59%

38 The overall results were mainly driven by solar and wind power generation,
39 especially solar power generation

40 The pandemic has accelerated the transition of the power generation mix and the
41 primary energy mix toward renewable energy sources

42

43

44 **1. Introduction**

45 The Corona Virus Disease 2019 (COVID-19) and the resulting strict containment
46 measures have resulted in huge economic contraction and social welfare losses for
47 many countries or regions (Baker et al., 2020; Ding et al., 2020; Nicola et al., 2020).
48 Most governments have called on people to self-isolate for the required period, forced
49 businesses to reduce their activity, and implemented city-wide lockdowns during the
50 pandemic (Fang et al., 2020; Liu et al., 2020). The year 2020 witnessed the sharpest
51 economic contraction since the great depression of the 1930s (IMF, 2020).

52 To prevent the spread of the virus, China has taken strong prevention and control
53 measures. These measures include but are not limited to the extension of the Spring
54 Festival holiday (from January 24 to February 10), maintaining social distance,
55 delaying the factory commencement dates, traffic control, and even blocking cities
56 (Kraemer et al., 2020; Tian et al., 2020; Li et al., 2021). No doubt that the outbreak
57 and spread of the virus are a tragedy and have exerted a tremendous impact on
58 China's economy and society.

59 This study filled the gap by investigating the COVID-19 effect on energy
60 transitions using the decarbonization of China's power generation sector as an
61 example. China is a distinguished case study due to its status as the world's largest
62 emitter of carbon emissions and thus faces unprecedented pressure to advance energy
63 transitions (Zhang and Chen, 2021). China committed to achieving the carbon peak by
64 2030 and carbon neutrality by 2060 ("Dual Carbon") (Zhang et al., 2021). Like others,
65 the power sector will be key to helping China meet its aggressive low-carbon
66 generation targets as well as the broader dual carbon target (Zhao et al., 2020; 2021).
67 The information on how the COVID-19 shock has affected the energy transition is a
68 piece of critical information for China to make its dual carbon policy. However, the
69 question is how to quantitatively analyze the COVID-19 effect on energy transitions
70 from the perspective of low-carbon power generations. Moreover, any attempt to
71 combat global warming depends critically on China's energy transition trajectory, and
72 the direction of China's energy transition has a leading impact globally (Jiang et al.,
73 2019). Therefore, from the perspectives of both academic research and industrial
74 practice, it is necessary to discuss in a timely manner how COVID-19 has affected the
75 direction of the energy transition under the current setting and how the energy
76 industry can find a path to rapid recovery during and after this crisis.

77 The present research is different from the relevant literature in at least two
78 aspects. To the best of our current knowledge, this is among the first empirical studies
79 that estimate the changes in low-carbon power generation levels before and during the
80 pandemic period relative to the previous period, which contributes to previous
81 empirical literature concentrating on economic variables and emission reductions
82 (Bekkers and Koopman, 2020; Dang and Trinh, 2021; Oskoui, 2020). Then, based on
83 the stacked data of solar power, wind power, nuclear power, and hydropower, this
84 study used a difference-in-differences (DID) model with historical controls to
85 quantitatively identify the overall effect of the COVID-19 pandemic on the energy
86 transition from low-carbon power generations. The method has recently been applied
87 in a few estimations of the COVID-19 impact (He et al., 2020; Chen et al., 2020;
88 Wang et al., 2021).

89 Second, this study assessed the heterogeneous impacts of the COVID-19 shock
90 on energy production and the energy mix of different types of energy sources. In the
91 literature, little emphasis has been placed on comparing impacts across different types
92 of power generation or primary energy sources even though such work is essential for
93 investigating the implications of the COVID-19 crisis on the direction of energy
94 transitions (Liu et al., 2021). In contrast, this study analyzed how the crisis has
95 affected the progress in expanding low-carbon or carbon-neutral energy sources.

96 The remainder of the study is organized as follows: In Section 2, we focused on
97 the literature review, and we introduced the data and statistical methodology in
98 Section 3. The overall results were presented in Section 4, which was followed by a
99 further discussion of the heterogeneous results in Section 5. Section 6 concluded and
100 provided some relevant policy implications.

101 **2. Literature review**

102 The shock of COVID-19 has stimulated intensive research activities. The
103 majority of these studies focused on investigating the economic effects of COVID-19
104 from multiple perspectives, such as economic output (Morgan et al., 2021;
105 Gharehgozli et al., 2020), household consumption (Martin et al., 2020), labor
106 employment (Hershbein and Holzer, 2021), supply chain (Shi et al., 2021) and
107 financial market (Ali et al., 2020; Baker et al., 2020; Ding et al., 2020). The COVID-
108 19 effect on carbon emissions or air quality (i.e., PM_{2.5}, PM₁₀, and SO₂) has also

109 been a hot topic. Recent studies have empirically discussed the reductions in global
110 CO₂ emissions (e.g., Liu et al., 2020; Forster et al., 2020; Le Quéré et al., 2020) and
111 the changes in China's urban air quality (e.g., Shi and Brasseur, 2020; Huang et al.,
112 2021; Chang et al., 2020) due to COVID-19. Most studies have found that the
113 COVID-19 crisis has lowered carbon emissions or improved air quality.

114 Despite the proliferation of studies, how COVID-19 has affected energy
115 transitions is still not clear. On the one hand, COVID-19 could have slowed down
116 energy transitions. The COVID-19 crisis and the related containment measures have
117 significantly reduced energy consumption in many countries, which in turn has
118 influenced the deployment of renewables (IEA, 2020a; Chiaramonti and Maniatis,
119 2020; Zhong et al., 2020). Disruptions caused by the crisis have taken a big toll on the
120 investment and construction of renewable energy projects. In several countries, the
121 pandemic has made an already challenging investment environment worse,
122 specifically with regard to renewables (Selmi et al., 2021; Ivanov and Dolgui, 2021).
123 From an economic perspective, the crisis has exacerbated the financing challenges
124 that also slowed the support and dampened the enthusiasm of investors for energy
125 transitions (Karmaker et al., 2021; Mastropietro et al., 2020). Especially in countries
126 with a strong dependence on fossil fuel industries, the governments were likely to
127 transfer the funds originally used for the energy transition into the fields of health care
128 and social welfare, further slowing down the switching to low-carbon or carbon-
129 neutral energy sources (Birol, 2020; Emma, 2020).

130 On the other hand, COVID-19 may have accelerated energy transitions. In
131 today's world, a dramatic fall in the costs of renewable energy has speeded up the
132 large-scale utilization of renewable energy sources in power generation (Kåberger,
133 2018). During this pandemic, the power demand in various countries has generally
134 decreased (IEA, 2020a; Ghenai and Bettayeb, 2021). As a result, the power generation
135 capacity has exceeded the demand. Grid operators may have prioritized cheap, clean,
136 and environmentally friendly non-fossil energy. In addition, the deglobalization
137 caused by COVID-19 isolation measures has prompted some countries to enhance the
138 localization of supply chains or seek flexible solutions for resource development
139 (Quitow et al., 2021; Ba and Bai, 2020). Especially, many European countries were
140 continuing to deploy renewable energy sources, while continuous divestment trends in
141 the fossil fuel industries were accelerating in the wake of the crisis (European
142 Commission, 2020; Council of the European Union, 2020).

143 It can be seen from the above literature that the COVID-19 effect on energy
144 transitions is still controversial. However, the future of the energy system is going to
145 be in a more complex, diversified, and uncertain situation. Considering that the
146 transition from high-carbon energy to low-carbon energy sources is a fundamental
147 way of accelerating the power sector transformation (Wei et al., 2021), we used the
148 low-carbon power generations as the key indicator for this study. These low-carbon
149 generation sources include renewable energy, mainly solar and wind power, and
150 nuclear and hydropower, which are also actively promoted by the Chinese
151 government. Through the use of modified DID models, this study analyzed the overall
152 impact of COVID-19 on low-carbon power generations with a monthly province-
153 specific, source-specific dataset. Then, the study compared the productions of
154 different power generation and primary energy sources before and during the
155 pandemic and assessed how the recent COVID-19 pandemic has affected the direction
156 of the energy transition by fuel type.

157 **3. Data and methodology**

158 **3.1. Data**

159 This study used monthly power generations, energy production, and weather
160 conditions in China's 30 provinces from July 2018 to June 2020. The province-level
161 data for the generation of low-carbon power and the supply of other energy sources
162 were obtained from the National Bureau of Statistics of China (NBS). In this study,
163 low-carbon power mainly includes solar power, wind power, nuclear power, and
164 hydropower.^a Monthly meteorological data (average temperature, precipitation,
165 average relative humidity, and sunshine hours) for the 30 provinces were collected
166 from China statistical yearbooks and the National Meteorological Information Center.
167 In addition, this study measured the energy mix by calculating the ratio of specific
168 energy sources to the total energy supply and then examined the effects of the
169 COVID-19 pandemic on the direction of the energy transition. In measuring the
170 primary energy mix, the physical quantity of all primary energy sources has been

^a This is because the monthly power generation data from biomass, geothermal, or other renewables are not available. In addition, compared to wind and solar power generation, the power generated from the combined category for biomass, geothermal, and other renewables is at a negligible level. For example, in the first half of 2020 in China, the power generated from the combined category accounted for 0.0012% of the total power generation.

171 converted into standard coal equivalent.^a Table 1 presents the summary statistics of
 172 our key variables.

173 **Table 1.**

174 **Summary statistics of the main variables**

Variable	Description (unit)	Mean	Std. Dev.	Min	Max	Obs
<i>lcp</i>	stacked low-carbon power generations (10 ⁸ kWh)	14.11	37.28	0	366.8	2400
<i>prod_hp</i>	hydropower generation (10 ⁸ kWh)	33.48	66.49	0	366.8	600
<i>prod_wp</i>	wind power generation (10 ⁸ kWh)	9.766	21.7	0	114.2	600
<i>prod_np</i>	nuclear power generation (10 ⁸ kWh)	9.94	11.65	0	72.6	600
<i>prod_sp</i>	solar power generation (10 ⁸ kWh)	3.257	3.094	0	12.97	600
<i>mix_tpg</i>	the share of thermal power in the total power generation	0.715	0.25	0.04	0.995	600
<i>mix_hpg</i>	the share of hydropower in the total power generation	0.171	0.248	0	0.929	600
<i>mix_npg</i>	the share of nuclear power in the total power generation	0.045	0.09	0	0.391	600
<i>mix_wpg</i>	the share of wind power in the total power generation	0.05	0.045	0	0.225	600
<i>mix_spg</i>	the share of solar power in the total power generation	0.019	0.027	0	0.183	600
<i>mix_coal</i>	the share of raw coal in the total primary energy supply	0.479	0.341	0	0.993	600
<i>mix_oil</i>	the share of crude oil in the total primary energy supply	0.131	0.197	0	0.928	600
<i>mix_gas</i>	the share of natural gas in the total primary energy supply	0.127	0.214	0	0.976	600
<i>mix_sps</i>	the share of solar power in the total primary energy supply	0.013	0.019	0	0.12	600
<i>mix_wps</i>	the share of wind power in the total primary energy supply	0.034	0.041	0	0.426	600
<i>mix_hps</i>	the share of hydropower in the total primary energy supply	0.131	0.199	0	0.914	600
<i>mix_nps</i>	the share of nuclear power in the total primary energy supply	0.084	0.188	0	0.842	600

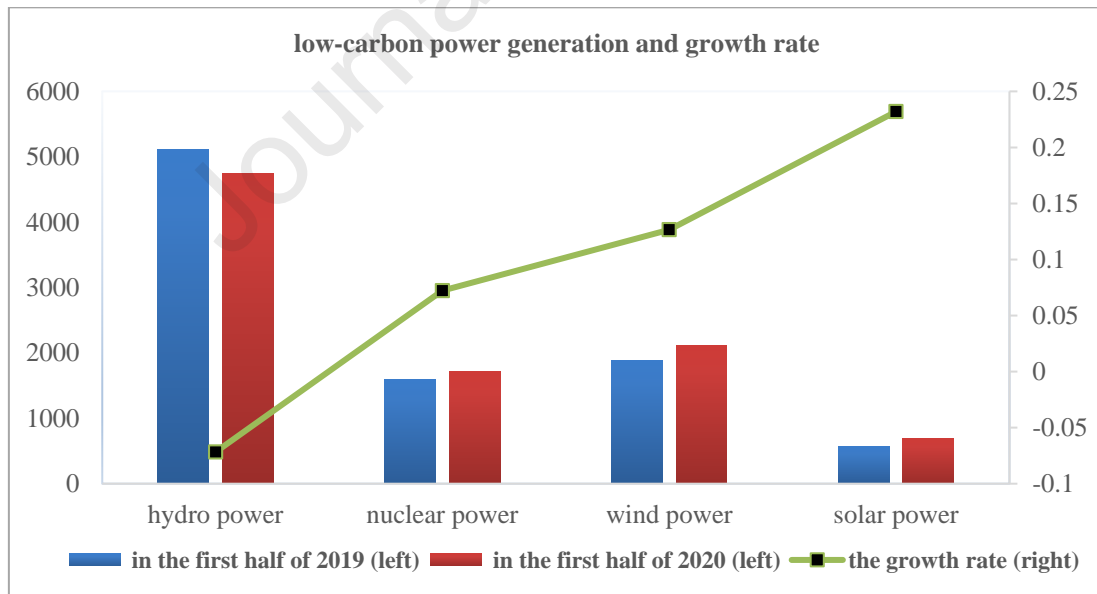
^a The primary energy supply was calculated by multiplying the activity data (i.e., energy production) and the conversion factors by energy types. Here, we used the standard coal conversion factor by different energy sources from the China energy statistical yearbooks to assess the total primary energy quantity. For example, the conversion factors of various low-carbon power generations are the same, namely, 10000 kWh of low-carbon power is equal to the power produced by burning 1.229 tons of standard coal.

<i>temp</i>	average temperature (°C)	17.29	9.251	-16	32.2	600
<i>humid</i>	average relative humidity (%)	66.1	15.47	1.4	93	600
<i>sun</i>	sunshine hours (h)	180.3	66.32	15.4	348.2	600
<i>preci</i>	precipitation (mm)	87.92	95	0	574	600

175 Notes: This study used data that include monthly power generations, energy production, and weather conditions in
 176 China's 30 provinces (excluding Hong Kong, Macao, Taiwan, and Tibet autonomous region), from July 2018 to
 177 June 2020 (excluding January and February).

178 The data show that renewable energy development initially had a certain ability
 179 to resist external shocks. In the first half of 2020, the global wind and solar power
 180 generation accounted for 9.8% of the total power generation, an increase of 14% over
 181 the same period in 2019 (IEA, 2020a). Also, the total installed capacity of global coal
 182 power decreased for the first time in history. In China, the most impressive progress
 183 has occurred in the power generation sector, where modern renewables (such as solar
 184 and wind power) have advanced significantly. When the total power, thermal power,
 185 and hydropower generation decreased by 0.08%, 0.59%, and 7.17%, respectively,
 186 year-on-year in the first half of 2020, the generation of domestic wind power and
 187 solar power increased by 12.65% and 23.20%, respectively (see Figure 1).

188



189

190 **Figure 1. The changes in low-carbon power generation during the first half-year of 2019 and**
 191 **2020**

192

Source: Author's own conception. Due to data availability, we defined four major low-carbon power sources:
 193 hydro, nuclear, wind, and solar in this study.

194 3.2. The modified DID models with historical controls

195 The study aimed to quantitatively identify the COVID-19 effect on energy
 196 transitions from the perspective of low-carbon power generations. As the COVID-19
 197 shock was a major public health emergency and the resulting containment measures
 198 were highly exogenous, the impacts on the energy supply and energy transition also
 199 met the main assumptions of a quasi-natural experimental design (Kanda and Kivimaa,
 200 2020). In this study, the DID model, using Stata software, version 15.1, was then
 201 applied to quantify power generation changes due to the pandemic.

202 However, the standard DID model needs to be modified for studying the
 203 COVID-19 pandemic. All Chinese provinces were in some degree of lockdown
 204 during the pandemic period, meaning that observational data at the province level
 205 provided no contemporary untreated controls, and it was difficult to estimate an
 206 average treatment effect according to the standard DID model. The literature proposed
 207 to identify a comparable group that could not receive treatment, e.g., historical
 208 controls prior to its availability (Newsome et al., 2021; He et al., 2020). With
 209 reference to Wang et al. (2021), how the COVID-19 or national-level pandemic-
 210 related measures have affected low-carbon generation relative to the trends in
 211 previous periods was examined and the first modified DID model with historical
 212 controls was as follows.

$$213 \quad lcp_{sit} = \alpha_0 + \alpha_1 treat \times post + controls + \gamma_s + \mu_i + \delta_t + \varepsilon_{sit} \quad (1)$$

214 where s , i , and t denote low-carbon power sources (solar power, wind power, nuclear
 215 power, or hydropower), provinces, and months, respectively. This study set the low-
 216 carbon power generations from July 2019 to June 2020 as a treatment group. This
 217 group was compared to a historical control group from July 2018 to June 2019. “*Treat*”
 218 is a grouping dummy variable, the value of which is set as 1 if it is in the period July
 219 2019 to June 2020, and set to 0 for July 2018 to June 2019. The value of “*post*” is set
 220 as 1 if it is a month during the pandemic period (March 2019 to June 2019, or March
 221 2020 to June 2020) within our study period.^a “*Controls*” describes the monthly
 222 weather condition variables (average temperature, precipitation, average relative
 223 humidity, and sunshine hours).

^a Because the power generation data for January and February were missing, this paper defined the pandemic period (the treatment period) as March to June (2019, 2020), and the period before the pandemic as July to December (2018, 2019). Also, based on existing evidence, excluding the Chinese Spring Festival holidays (from January to February) could avoid any power generation changes unrelated to the pandemic (Chen et al., 2020).

224 To capture the overall effect of the pandemic on the energy transition from the
 225 low-carbon power supply, this study followed the approach of Duflo et al. (2013) and
 226 Li et al. (2020) and used the stacked low-carbon power generations as the explained
 227 variable (*lcp*).^a The parameter of interest is α_t , which reflects the COVID-19 effect
 228 on low-carbon generation. Specifically, we calculated the changes in low-carbon
 229 generation during the pandemic versus before the pandemic period, from 2019 to
 230 2020, and compared these findings with corresponding changes in the same periods
 231 from 2018 to 2019. γ_s is the set of power source fixed effects, controlling for any
 232 time-invariant source heterogeneity. μ_i is the set of province fixed effects, controlling
 233 for time-invariant, unobserved province characteristics across provinces, such as
 234 geographic features. δ_t is the set of month fixed effects, controlling for the monthly
 235 shocks common to all provinces, such as business cycles. ε_{sit} is an error item. We
 236 estimated Eq. (1) allowing for province-level clustering of the errors.

237 The baseline DID model identifies the average differences in low-carbon
 238 generations between the treatment and control groups. On this basis, the monthly
 239 differences in low-carbon generation measures between the two groups were further
 240 compared. Based on Eq. (2), this study performed whether the DID model met the
 241 parallel trend requirements during the pre-pandemic period, and dynamic analysis of
 242 the COVID-19 effect. The test model is set as:

$$243 \quad lcp_{sit} = \beta_0 + \sum_t \beta_t treat \times d_t + controls + \gamma_s + \mu_i + \delta_t + \varepsilon_{sit} \quad (2)$$

244 where d_t is a series of month dummy variables. In Eq. (2), the dummy variable
 245 indicating one month before the treatment (December) was omitted from the
 246 regression, the focus was on the month-to-month changes in the coefficients β_t within
 247 the event window. More importantly, the conditions under which the outcome
 248 variable follows a common trend are as follows: the coefficients β_t (from July to
 249 November) were nonsignificant. During the treatment period, by comparing the
 250 changes in β_t (from March to June), it is possible to analyze the dynamic effect of the
 251 COVID-19 shock on low-carbon generation.

252 Next, to explore whether the COVID-19 effect varies across different types of
 253 power sources or energy sources, this study tested for the existence and direction of
 254 causality between the COVID-19 pandemic and energy supply in China at

^a In unstacked data, each power sample is in a separate column. Alternatively, all the data can be stacked in one column, that is, the four power sources are pooled together. Of course, we also added a column of grouping indicators (numbers or text) that define each power sample.

255 disaggregated levels, like solar power, wind power, nuclear power, hydropower, and
 256 so on. Note that the heterogeneity analyses help us to understand what drives the
 257 overall effects (Nicolli and Vona, 2016) and to compare the influence on the
 258 production of various energy sources. In this study, the heterogeneity analysis is based
 259 on Eq. (3) below:

$$260 \quad prod_{it} = \theta_0 + \theta_1 treat \times post + controls + \mu_i + \delta_t + \varepsilon_{it} \quad (3)$$

261 where the explained variable *prod* is one of the energy production indexes in province
 262 *i* at month *t*, including low-carbon power sources and other primary energy sources
 263 (such as raw coal, crude oil, and natural gas). Province and month fixed effects are
 264 included in all specifications in order to control for time-unvarying province attributes
 265 and nationwide common time shocks, respectively.

266 Each energy source type is associated with a bundle of environmental effects.
 267 Moving further upstream in the energy supply chain, the transition toward low-carbon
 268 or carbon-neutral energy sources involves the gradual reduction of the exploitation of
 269 fossil fuel resources (Davidson, 2019; York and Bell, 2019). To better understand the
 270 impacts on the direction of the energy transition, this study measured the energy mix
 271 by calculating the ratio of specific energy sources to the total energy supply. Then, the
 272 heterogeneous effects of COVID-19 on the energy mix were examined. The
 273 specification for the energy mix of each type of energy is:

$$274 \quad mix_{it} = \pi_0 + \pi_1 treat \times post + controls + \mu_i + \delta_t + \varepsilon_{it} \quad (4)$$

275 where the dependent variable *mix* is either the share of a certain type of power source
 276 in the total electricity generation or the share of a certain type of primary energy in the
 277 total primary energy supply in province *i* at month *t*. Each regression implements
 278 model (4) and controls for the weather condition variables, province and month fixed
 279 effect.

280 **4. Overall effects**

281 **4.1. Baseline estimation**

282 The DID model (Eq. (1)) was used to estimate the changes in low-carbon power
 283 generation levels before and during the pandemic period, relative to the previous
 284 period, and to quantitatively assess the overall effect of COVID-19 on energy
 285 transition from the perspective of low-carbon power generations. Column (1) of Table
 286 2 shows the effect of the COVID-19 pandemic on low-carbon power generations

287 through the stacked data of solar and wind power. Using the stacked data of two
 288 different combinations of the three low-carbon power sources, the estimation results
 289 were reported in columns (2) and (3) of Table 2. When all four low-carbon power
 290 sources are pooled together, column (4) presents the benchmark results for the overall
 291 effect of COVID-19 on low-carbon power generations. All regressions include
 292 controls for province fixed effects, month fixed effects, source-specific fixed effects,
 293 and weather conditions. However, only the coefficients of the interaction term
 294 ($treat \times post$) were discussed here, due to limited space.

295 **Table 2.**

296 **Overall effects of COVID-19 on low-carbon generations**

Column	(1)	(2)	(3)	(4)
Variable	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>
Type	Solar and wind power	Solar, wind, and hydro power	Solar, wind, and nuclear power	Solar, wind, nuclear, and hydro power
$treat \times post$	1.122*** (0.213)	0.547* (0.272)	1.063*** (0.269)	0.648** (0.247)
controls	Yes	Yes	Yes	Yes
province FE	Yes	Yes	Yes	Yes
month FE	Yes	Yes	Yes	Yes
source FE	Yes	Yes	Yes	Yes
Obs	1,200	1,800	1,800	2,400
R-squared	0.666	0.377	0.331	0.285

297 Notes: This table presents estimates of DID regressions of the energy transition on the COVID-19 pandemic and
 298 weather condition variables. The dependent variable is the stacked low-carbon power generations (*lcp*) for all
 299 columns (1)-(4) with different power source types. The weather condition controls are the monthly average
 300 temperature (*temp*), monthly precipitation (*preci*), monthly average relative humidity (*humid*), and monthly
 301 sunshine hours (*sun*) for each province. All the specifications control for province fixed effects, month fixed
 302 effects, and source-specific fixed effects. The estimates of weather variables, fixed effects dummies, and constant
 303 terms are suppressed for brevity. Reported in parentheses are robust standard errors clustered by province. ***p <
 304 0.01, **p < 0.05, *p < 0.1.

305 The results show that the interaction term was significantly positive when
 306 considering weather condition variables and the fixed effects of the three dimensions.
 307 This finding means that the COVID-19 crisis had a significant promotion effect on the
 308 low-carbon energy supply, compared with the same period in 2018-2019. The
 309 benchmark estimate in column (4) of Table 2 demonstrates that, across the four
 310 measures of low-carbon energy supply, the COVID-19 pandemic on average

311 increased the low-carbon power generation by 0.0648 billion kWh (by 4.59%).^a
 312 These positive impacts of COVID-19 on low carbon generation could be due to the
 313 following factors. First, the output of low-carbon power is largely unaffected by the
 314 weak demand, because low-carbon power generation has low operating costs and
 315 priority dispatch (Quitow et al., 2021; Liu et al., 2021). Moreover, the installed
 316 capacity of wind and solar power generation continues to expand in China, further
 317 increasing the advantages of variable renewable energy sources. Therefore, low-
 318 carbon energy has ushered in an unconventional development opportunity (Hoang et
 319 al., 2021).

320 4.2. Robustness checks

321 4.2.1. Parallel trend hypothesis test and dynamic effect analysis

322 When applying the DID model, one validity test commonly used involves
 323 examining whether the treatment and control groups exhibit parallel pre-treatment
 324 trends. This study adopted the event study approach by estimating a series of
 325 coefficients for each month to investigate how the trends in the low-carbon
 326 generation between the two groups evolved before and during the pandemic period.

327 The estimated coefficients for each month within the event window, along with
 328 the 95% confidence intervals, were presented in Figure 2. The dummy variable for
 329 December (one month before the treatment) was omitted from the regression. After
 330 introducing the interactions of month dummy variables and the term *treat*, all the
 331 estimates for the five months before the treatment were statistically insignificant at the
 332 5% level. The results suggest that the trends in the low-carbon generation before the
 333 pandemic period were similar to those in 2018. This finding inspires confidence that
 334 the historical control group (2018.7-2019.6) provided a good counterfactual for the
 335 treatment group (2019.7-2020.6). Meanwhile, the interactive term after the treatment
 336 ($treat \times d_{Mar}$) was significantly positive, with the low-carbon generation increasing by
 337 0.1260 billion kWh (Column (1) of Table 3). Despite an abnormal two or three
 338 months down after the spring festival, the value quickly becomes positive. These
 339 results confirm the conclusion that the COVID-19 pandemic significantly increased
 340 low-carbon generation (Supplementary Note).

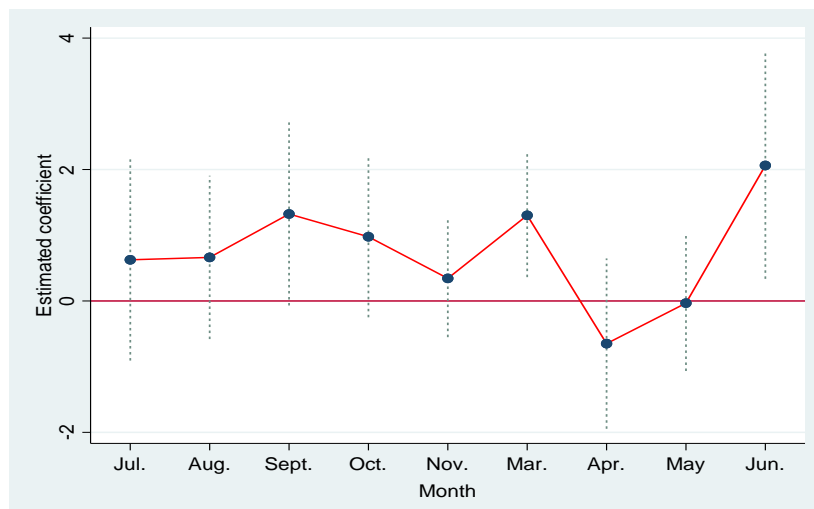
341

^a The most important thing of causal identification is to ensure the consistent estimation of causal effects (Cinelli et al., 2021). In this study, the values of R^2 in Table 2 are acceptable after considering a series of robust tests that followed.

342 **Table 3.**343 **Robustness tests based on model specifications**

Column	(1)	(2)	(3)	(4)	(5)
Variable	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>
Type	Dynamic effects	Province-time trend	Province-energy effects	Adding the square of temperature	Adding the square terms of temperature and rainfall
<i>treat</i> × <i>post</i>		0.667** (0.256)	0.669*** (0.239)	0.623** (0.250)	0.508** (0.244)
<i>treat</i> × <i>d</i> _{Mar}	1.260** (0.458)				
<i>treat</i> × <i>d</i> _{Apr}	-0.615 (0.633)				
<i>treat</i> × <i>d</i> _{May}	-0.0568 (0.513)				
<i>treat</i> × <i>d</i> _{Jun}	2.006** (0.827)				
controls	Yes	Yes	Yes	Yes	Yes
province FE	Yes	Yes	Yes	Yes	Yes
month FE	Yes	Yes	Yes	Yes	Yes
source FE	Yes	Yes	Yes	Yes	Yes
Obs	2,400	2,400	2,400	2,400	2,400
R-squared	0.285	0.286	0.933	0.285	0.285

344 Notes: This table reports the estimation results for robustness tests based on model specifications. The dependent
 345 variable is the stacked low-carbon power generations for all columns (1)-(6) with four energy types. Other notes as
 346 Table 2.



347 **Figure 2. Parallel trend hypothesis test and dynamic effect analysis**

348

349 Source: Author's own conception based on Stata software. Low-carbon generation levels are compared between
350 2018.7-2019.6 and 2019.7-2020.6. The dummy variable for December (one month before the treatment) is omitted
351 from the regression. Also, excluding the Chinese Spring Festival holidays (from January to February) could avoid
352 any changes in power generation that were unrelated to the pandemic. Each estimate shows the difference in low-
353 carbon generation relative to the difference one month before the treatment. The red and dashed lines represent the
354 estimated coefficients and 95% confidence intervals, respectively.

355 *4.2.2. Province-month trend and province-energy effects*

356 The province-month trend terms were added to the regression model to control
357 some of the provincial factors that may have been omitted or changed over time (Liu
358 and Qiu, 2016). After introducing the crossovers of the province dummy variables and
359 the monthly trend term, the COVID-19 effect in column (2) of Table 3 was still
360 significant, thereby confirming the robustness of the baseline results. In column (3), in
361 addition to the fixed effects considered in the baseline scenario, this study controlled
362 for province-source fixed effects and thus rules out any bias from unobserved changes
363 affecting specific power generations in each province. The key findings regarding the
364 COVID-19 effect on low-carbon generations were broadly consistent.

365 *4.2.3. Adding the square terms of weather variables*

366 To verify whether a non-linear relationship exists between weather variables and
367 power generations, referring to Zheng et al. (2019), column (4) added the square term
368 of temperature to the model. The results show that the square term was not significant,
369 and the interaction term was significantly positive. Column (5) further added the
370 square terms of temperature and precipitation to the model. The direction and
371 magnitude of the interaction term coefficient were consistent with those in Table 3.

372 *4.2.4. Adding additional control variables*

373 The commissioning of new renewable energy facilities and energy market
374 fluctuations during the sample period could lead to estimation errors. We therefore
375 included the renewable power commissioning indicator (measured by the "newly
376 added renewable power capacity") and the energy price indicator (measured by the
377 "fuel and power price index" at 2018 constant prices) in the regression to control for
378 the potential impact of these variables. The estimation results provided in columns (1-
379 2) of Table S1 reveal that, adding additional control variables did not alter our
380 conclusions of the baseline regression.

381 *4.2.5. Sample adjustment*

382 In light of the extent and pace of the expansion of the COVID-19 outbreak in
383 various provinces, an infection index was applied that allows taking into account the

384 magnitude of the pandemic (Zhu et al., 2020). This index was constructed as the
 385 natural logarithm of one plus the number of accumulated confirmed cases each
 386 month.^a The corresponding results reported in Column (1) of Table 4 indicate that the
 387 estimated coefficient for the interaction term between the treatment group and the
 388 infection index was significantly positive. This finding confirms that the severity of
 389 the pandemic has tended to impact the low-carbon energy supply positively.

390 **Table 4.**

391 **Robustness tests based on sample adjustment**

Column	(1)	(2)	(3)	(4)	(5)
Variable	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>	<i>lcp</i>	$\ln(lcp + 1)$
Type	Using pandemic reporting data	Deleting the samples from Hubei	Deleting the samples with “0” values	Deleting data for July and August	Taking the logarithm value
<i>treat</i> × <i>post</i>	0.0912* (0.0527)	0.696** (0.267)	0.831** (0.320)	0.737** (0.311)	0.0653*** (0.0133)
controls	Yes	Yes	Yes	Yes	Yes
province FE	Yes	Yes	Yes	Yes	Yes
month FE	Yes	Yes	Yes	Yes	Yes
source FE	Yes	Yes	Yes	Yes	Yes
Obs	2,400	2,320	1,898	1,920	2,400
R-squared	0.285	0.275	0.389	0.282	0.399

392 Notes: This table presents the estimation results for robustness tests based on sample adjustment. The dependent
 393 variable is the stacked low-carbon power generations for all columns (1)-(5) with four power sources. Other notes
 394 as Table 2

395 Hubei province, where the new virus was first detected and strict epidemic
 396 prevention measures were imposed in China, has also been excluded from this study.
 397 It can be seen from column (2) of Table 4 that the results were not dominated by the
 398 province that was most affected by the virus. In addition, there are some “0” values in
 399 the data. Especially, this applies to marginal power generation technologies, such as
 400 nuclear power. After deleting the samples with “0” values, the regression results
 401 shown in column (3) of Table 4 suggest that the basic conclusions were not affected
 402 obviously.

^a The number of COVID-19 confirmed cases for 30 provinces is obtained from China Stock Market & Accounting Research Database (CSMAR), which tracks the real-time confirmed cases all over the country.

403 We used a different starting sample month to check the sensitivity, i.e., we
404 dropped two months at the head and changed the start of the sample period to
405 September. After deleting data for July and August, the results shown in column (4)
406 of Table 4 were consistent with the benchmark results, i.e., the level of low-carbon
407 generations increased substantially due to the pandemic.

408 To mitigate potential outliers, the baseline tests were repeated with the natural
409 logarithm of one plus the total low-carbon generation as the dependent variable. The
410 logarithm transformation allows one to capture the percentage change in total low-
411 carbon generation. Similar estimation results were found after the inclusion of this
412 relative measure (column (5)), i.e., the estimated parameter for the interaction term
413 was significantly positive.

414 **5. Further discussion**

415 **5.1. Heterogeneous effects on the energy production by primary energy sources**

416 Despite the significance of the COVID-19 pandemic related to overall low-
417 carbon generation, it hides significant heterogeneity across low-carbon power sources.
418 To better understand the evolution of low-carbon power and other primary energy
419 sources, this study took a step forward and compared the influence of the COVID-19
420 pandemic on energy production by different primary energy sources.

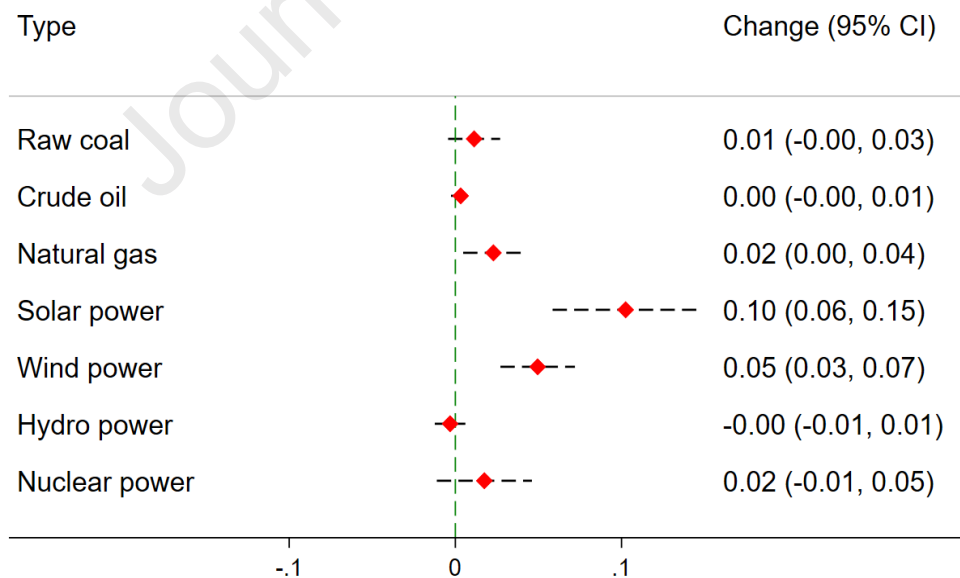
421 Figure 3 displays the regression results of Eq. (3) for seven different primary
422 energies (raw coal, crude oil, natural gas, solar power, wind power, hydropower, and
423 nuclear power). The standardized regression coefficient was reported for each primary
424 energy source by employing a pooled panel with weather variables and fixed effects
425 dummies. The change in energy production level was estimated before and during the
426 pandemic period, relative to the previous period.

427 In Figure 3, the dependent variables are the energy production indices. Among
428 the four electricity generation sources, the coefficients of the interaction term between
429 the treatment group and pandemic period were significantly positive for solar power
430 and wind power. This finding indicates that the COVID-19 pandemic improved solar
431 and wind power generation compared with the same period in 2018-2019. Moreover,
432 it should be pointed out that the overall results were mainly driven by solar and wind
433 power. Especially, the pandemic had the most significant effect on solar power, with a
434 standardized estimated coefficient of 0.103. The pandemic or the pandemic-related

435 measures appear to have had a major driving effect on renewable project development
436 in China.

437 In fact, the operation of renewable power generation was less affected by
438 fluctuations in raw materials and manpower and has had apparent advantages during
439 the COVID-19 pandemic (Kelvin and Brindley, 2020). The technological
440 advancement and electricity market reform have substantially reduced the costs and
441 affordability of renewable energy. Thus, the competitiveness of modern renewable
442 energy sources (such as solar and wind power) has increased significantly (IRENA,
443 2021; Amir and Khan, 2021). However, no significant effect was observed for
444 hydropower and nuclear power. For technologies with a long lead time for
445 development, such as hydropower and nuclear power, electricity generation may not
446 be significantly affected by the outbreak.

447 For other primary energy sources (fossil fuels), the pandemic significantly
448 increased the supply of natural gas, at a significance of 5% and a standardized
449 estimated coefficient of 0.02. Yet, the production of raw coal and crude oil that
450 remain China's base energy sources have not changed significantly during the
451 COVID-19 period. This finding at least shows that the pandemic has been more
452 inclined to push the development of clean and low-carbon energy.



453
454 **Figure 3. Heterogeneous effects of COVID-19 on the production of various primary energy**
455 **sources**

456 Source: Author's own conception based on Stata software. Red diamonds mark the standardized estimated
457 coefficients of the interaction term and the dashed black lines represent the 95% confidence intervals of the
458 estimate.

459 5.2. Heterogeneous effects on the energy mix

460 The COVID-19 crisis has already had significant effects on low-carbon power
461 generations, but how has it influenced the direction of the energy transition? As the
462 electricity sector is an important contributor to carbon dioxide emissions (Li et al.,
463 2017), this study additionally considered a relative power generation indicator, instead
464 of the absolute amount of energy production, i.e., the ratio of specific power sources
465 to total power generation was used. Through variables transformation, the COVID-19
466 effect on the direction of the energy transition was examined.

467 5.2.1. On power generation mix

468 Given that the same set of weather control variables and fixed effects dummies
469 are included in each regression, Table 5 presents the heterogeneous results of the
470 COVID-19 effect on the electricity generation mix by fuel type. Specifically, the
471 pandemic has led to a rise in the proportions of solar and wind power, while there has
472 been a decline in the proportion of hydropower (significant at the 5% level). This
473 finding implies that the direction of the electricity generation mix transition has
474 shifted from hydropower to solar and wind power. From the power supply side, the
475 decline in demand is intensifying the competition among various power generation
476 technologies and fuels. The non-dispatching ability of modern renewable energy
477 (including wind and solar) and renewable energy's priority in China's power system
478 have enabled it to buck the trend and become a beneficiary in the increasingly fierce
479 competition among various power sources. The impact of the pandemic has revealed
480 an important message, namely that renewable energy power generation is becoming
481 the baseload supply of electricity, due to the low marginal cost and priority grid
482 access.

483 Although hydropower accounts for a large proportion of non-fossil energy
484 generation in China, the creation of new hydropower generation has shown a
485 downward trend in the past few years. The estimated coefficient on the interaction
486 term of -0.011 in the hydro regression was likely due to low precipitation in
487 hydropower regions in the first half of 2020. In addition, the estimated COVID-19
488 effect on the thermal and nuclear power shares of the power generation mix has been
489 statistically insignificant. Compared with modern renewable energy power generation
490 with a low marginal cost, fossil fuel energy power generation has experienced more
491 frequent start-up/shutdown and has not had economic advantages during the pandemic.
492 However, thermal power has strong flexibility, continuous production, and strong

493 overall anti-risk ability. Nuclear energy cannot compete with renewable energy in
 494 terms of cost and construction speed and has been unaffected by the pandemic,

495 The regression results provide strong evidence that COVID-19 has advanced the
 496 transition of the power generation mix. Specifically, due to the pandemic, the power
 497 generation mix is likely to move, in relative terms, from hydropower (generated using
 498 domestic resources) toward modern, capital-intensive renewables. From the current
 499 situation, the COVID-19 crisis did not necessarily crowd out decarbonization efforts
 500 in the power industry, instead, it accelerated the electricity transition (Pianta et al.,
 501 2020).

502 **Table 5.**

503 **Heterogeneous effects of COVID-19 on the power generation mix**

Column	(1)	(2)	(3)	(4)	(5)
Variable	mix_tpg	mix_spg	mix_wpg	mix_hpg	mix_npg
Type	Thermal power	Solar power	Wind power	Hydropower	Nuclear power
<i>treat</i> × <i>post</i>	0.00136	0.00316**	0.00510**	-0.0110**	0.00136
	(0.00434)	(0.00121)	(0.00193)	(0.00405)	(0.00171)
controls	Yes	Yes	Yes	Yes	Yes
province FE	Yes	Yes	Yes	Yes	Yes
month FE	Yes	Yes	Yes	Yes	Yes
Obs	600	600	600	600	600
R-squared	0.913	0.919	0.907	0.919	0.900

504 Notes: This table presents the estimation results for the heterogeneous effects of COVID-19 on the power
 505 generation mix by fuel type. The dependent variable is the electric mix for all columns (1)-(5) with different power
 506 types. Other notes as Table 2.

507 *5.2.2. On primary energy mix*

508 To further understand the impacts of the COVID-19 pandemic on the primary
 509 energy mix by fuel type, this study measured the primary energy mix by calculating
 510 the ratio of specific energy sources to the total primary energy supply (10000 tons of
 511 standard coal). From the empirical results shown in Table 6, the COVID-19 effect on
 512 the transition of the primary energy mix away from carbon-intensive energy was
 513 significant. Specifically, the estimated COVID-19 effect was negative for the shares
 514 of raw coal and crude oil in the primary energy mix during the study period and was
 515 positive for solar and wind power. The expansion of solar and wind power was
 516 closely linked to a concurrent decline in the shares of raw coal and crude oil, the most

517 carbon-intensive forms of primary energy supply. This finding demonstrates that the
 518 primary energy mix tended to switch from raw coal and crude oil to solar and wind
 519 power. The estimates indicate that the pandemic's impacts on the shares of natural gas,
 520 hydropower, and nuclear power have been insignificant. In a word, the heterogeneous
 521 results reveal that the pandemic has accelerated the transition of the primary energy
 522 mix from high-carbon energy (i.e., raw coal and crude oil) to modern renewables,
 523 such as solar and wind power.

524 **Table 6.**

525 **Heterogeneous effects of COVID-19 on the primary energy mix**

Column	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Variable	mix_coal	mix_oil	mix_gas	mix_sps	mix_wps	mix_hps	mix_nps
Type	Raw coal	Crude oil	Natural gas	Solar power	Wind power	Hydro power	Nuclear power
<i>treat</i> × <i>post</i>	-0.0128* (0.00649)	-0.00399* (0.00197)	0.00600 (0.00410)	0.00346** (0.00138)	0.00750** (0.00317)	-0.00340 (0.00321)	0.00327 (0.00315)
controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes
province FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
month FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Obs	600	600	600	600	600	600	600
R-squared	0.905	0.906	0.902	0.897	0.630	0.903	0.900

526 Notes: This table presents the estimation results for the heterogeneous effects of COVID-19 on the primary energy
 527 mix by fuel type. The dependent variable is the primary energy mix for all columns (1)-(7) with different energy
 528 types. Other notes as Table 2.

529 The results of this study are consistent findings from the literature. The previous
 530 studies did not quantitatively estimate the changes in low-carbon power generations
 531 induced by the COVID-19 pandemic, although they reached a near consensus that
 532 China's energy transition has been altered by the pandemic to a great extent (Quitow
 533 et al., 2021; Liu et al., 2021; Hoang et al., 2021). For example, Quitow et al. (2021)
 534 and Hoang et al. (2021) showed that the crisis caused unprecedented decarbonization
 535 of the power system. Similarly, we found that the COVID-19 shock significantly
 536 increased low-carbon power generation. Meanwhile, several studies argued that the
 537 crisis might have tremendous consequences on the direction of the energy transition
 538 (European Commission, 2020; Pianta et al., 2020; Kuzemko et al., 2020). In a similar
 539 vein, this study further revealed that COVID-19 has promoted the adoption of low-

540 carbon power sources on the upper rungs of the electricity ladder (modern renewables
541 such as solar and wind power). The results of this study provided direct empirical
542 evidence on the COVID-19 effect on China's low-carbon energy transition, as well as
543 important cross-cutting insights not only for China but also for other large and
544 emerging economies.

545 **6. Conclusions and policy implications**

546 COVID-19 has profoundly changed the economy, society, and people's lives
547 worldwide. As a crucial part of the economy, China's energy sector should have also
548 been altered by the pandemic. Understanding the effects of COVID-19 on low-carbon
549 energy transitions in China is necessary for China to make its plan toward "Dual
550 Carbon" targets. However, while there are quite a few studies on the COVID-19, no
551 one has investigated how it affected energy transitions.

552 On the one hand, investigating the epidemic's treatment effect on energy
553 transitions can enrich the main contents of the impact assessment of the epidemic,
554 without limiting the analysis to the economy and human well-being. On the other
555 hand, when assessing a major public safety and health event such as COVID-19, it is
556 necessary to consider the possible deductions caused by the virus in terms of welfare
557 losses. To achieve more accurate and comprehensive evaluation results, consideration
558 is also given in this study to the impact on the low-carbon power supply and the
559 direction of the energy transition.

560 It was found that, by using the stacked low-carbon power generations (we
561 defined four major low-carbon power sources: solar, wind, nuclear, and hydro), the
562 COVID-19 pandemic had a significant promotion effect on low-carbon power
563 generations, compared with the same period in 2018-2019. In terms of economic
564 magnitude, the COVID-19 pandemic on average, increased the low-carbon power
565 generation by 4.59% (0.0648 billion kWh). This result was robust when considering
566 the parallel trend hypothesis test, dynamic effects, province-month trend, province-
567 energy effects, other model specifications, and changes in sample adjustment.

568 The heterogeneous analysis of the effect on energy production indicates that the
569 COVID-19 pandemic improved solar and wind power generation. It is also worth
570 noting that the overall results were mainly driven by solar and wind power generation,
571 especially solar power generation. The heterogeneous analysis of the effect on the

572 energy mix indicates that the pandemic has fostered the transition of the power
573 generation mix and the primary energy mix from high-carbon energy to modern
574 renewables (such as solar and wind power).

575 Our results have the following policy implications. China needs to seize the
576 momentum to promote the low-carbon energy transition during the COVID-19 crisis.
577 While the pandemic disrupted the world from all aspects, our results suggest that it
578 accelerated decarbonization efforts in the power industry, and promoted the power
579 mix toward renewable energy sources. Since renewables will play a vital role in
580 advancing low-carbon energy transition and achieving dual carbon targets, they
581 require a continued medium-term and long-term policy vision. Accordingly, the
582 development strategy of the next round of the energy industry should be scientifically
583 planned.

584 In addition, promoting energy transitions should be a part of the recovery plan.
585 In order to realize the dual carbon goals, China's post-pandemic economic stimulus
586 measures should be closely combined with long-term low-carbon development and
587 climate policies, such as market-oriented reform and energy transitions, so as to
588 promote green recovery. Investment in energy transitions may not only achieve
589 economic recovery in the short term (after COVID-19) but could also contribute to
590 long-term social development (Khan et al., 2021).

591 This study concluded by proposing several directions for future research. The
592 short-term effects of COVID-19 on the energy transition were only considered in the
593 present work, and it is still unclear whether the impacts were just a one-time shock or
594 have permanently altered the development model of the power system. As the
595 COVID-19 pandemic is still spreading all over the world, the long-term effects of
596 COVID-19 on the low-carbon power generation and the transition to renewables
597 remains to be seen, which is an important field of energy transition research (Zhong et
598 al., 2020). Also, while monthly source-specific data do provide a knowledge base for
599 assessing the decarbonization efforts of the power sector, information on day-to-day
600 energy production and generation patterns induced by COVID-19 is unfortunately
601 omitted. Therefore, a dataset on source-specific power generations with high time
602 frequency is urgently needed to understand how the pandemic has affected the low-
603 carbon power supply and generation patterns. Finally, the present study only focused
604 on energy production and energy transition in the context of China, where the
605 government sticks to the dynamic zero-covid policy in stopping the large-scale spread

606 of the virus, which is quite different from most other countries. Future studies could
607 continue to explore emerging generation patterns and cross-country differences,
608 which can help provide additional insight to understanding the COVID-19 effects on
609 global efforts to address energy transition.

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613 **Conflict of Interest statement**

614 The authors declare no conflict of interest.

615
616

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Highlights

The overall impact of COVID-19 on energy transitions from the perspective of low-carbon power generations was analyzed

By using the stacked data, the COVID-19 pandemic increased the low-carbon power generation by 4.59%

The overall results were mainly driven by solar and wind power generation, especially solar power generation

The pandemic has accelerated the transition of the power generation mix and the primary energy mix toward renewable energy sources

Declaration of interests

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

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