

Towards a Green Economy in Europe: Does Renewable Energy Production Has Asymmetric Effects On Unemployment?

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39 The talks at the international level are getting serious about the efficacy of renewable energy in
40 mitigating CO₂ emissions after the publication of the fourth assessment report (AR4) by the international panel
41 on climate change (IPCC) in 2007. The European Commission on climate change suggests that Europe can play
42 a leading role in the worldwide efforts to curb GHGs emissions and set a target of reducing emissions by 30%
43 with the help of other nations. Although the benefits of renewable energy in terms of energy security, resource
44 preservations, climate protection, and import independence are undisputed; however, actual economic costs of
45 renewable energy projects are still a point of controversy because of the high upfront costs of such projects.

46 Given the importance of renewable energy and its related economic benefits, many studies with
47 different geographical focuses have discussed the relationship between renewable energy and economic growth
48 without assessing the value-added. Breitschopf et al. (2016) covered the literature with regards to the impacts of
49 renewable energy on employment opportunities. They extensively reviewed two different methods of estimating
50 the relationship. One highlighted the positive effects of renewable energy on unemployment, while another
51 highlighted the adverse impacts of renewable energy. They offered six different techniques and determined that
52 the input-output modelling method is most appropriate for national investigation because of its worldwide
53 approach.

54 As already discussed that the renewable energy has a crucial role in improving environmental quality
55 (Pachauri et al, 2014). The placement of renewable energy projects is beneficial for all stakeholders. It presents
56 a win-win situation in terms of society's environment and economic well-being by curbing CO₂ emissions and
57 increasing employment opportunities in different economic segments via direct and indirect channels.
58 According to International Renewable Energy Agency (2011), renewable energy can create job opportunities in
59 almost every economic sector. Most of the studies in the literature also supported this notion (Meyer and
60 Sommer, 2014).

61 The deployment of renewable energy projects saw a massive surge after 2012, and more than 11
62 million people worldwide are serving in the renewable energy sector (Meyer and Sommer, 2014). The swiftly
63 growing development of renewable energy projects and technologies and the increasing numbers of employment
64 opportunities make it vital to explore the occupation effect of renewable electricity (Apergis & Salim, 2015). In
65 the existing literature, many studies have focused on the employment effects of renewable energy concerning
66 different locations; however, most of the studies ignored the employment opportunities created by nuclear and
67 non-renewable energy sources (Cameron and Zwaan, 2015 and Azretbergenova et al. 2021). They extensively
68 reviewed the literature on the employment effects of renewable energy and suggested that the size of net
69 employment effects differs significantly because of variations in geography, technologies, econometric
70 methodologies, and data collection. They also highlighted that there is no consensus vis-à-vis the long-run
71 sustainability in employment effects of the renewable energy sector (Rivers 2013; Mbarek et al. 2018; Çelik,
72 2021).

73 The share of renewable energy in total energy mixes of different countries is on the rise, particularly in
74 the developed economies such as Europe, where 34% of the total electricity demand is fulfilled via renewable
75 energy sources in 2019. On the other side, the share of renewable energy sources soared to 19.7% of total
76 energy in 2019, while this number was just 9.6% in 2004 (Eurostat, 2019). Though the benefits of renewable
77 energy are many, most notables are improved environmental quality and increased energy security. Another
78 significant benefit of renewable energy is job creation which is also noted by the various studies; however, this

79 claim is still debatable. Although multiple studies have supported the argument that renewable energy projects
80 benefit the economy by creating job opportunities (UNEP, 2011; Moreno & Lopez, 2008; Blanco and
81 Rodrigues, 2009), but these studies have relied on very positive or simple expectations, thus provided such
82 results. In some instances, a rational approach is adopted by considering reasonable suppositions, but
83 discriminatory findings can drive a wrong image of job creation. Previous literature confirmed that large-scale
84 and massive investments in renewable energy projects could positively affect employment opportunities at a
85 much greater pace than the employment effects of small renewable projects with low-level investments
86 (Böhringer et al. 2013; Yi, 2013).

87 Against this backdrop, in this analysis, our primary focus is to analyze the impact of renewable energy
88 on employment opportunities in European economies. European countries are rapidly adopting renewable
89 energy technologies, and it has become a vital provider of electricity and energy inside Europe. Hence, these
90 economies have become the ideal countries to study the impact of renewable energy projects on employment
91 opportunities. Moreover, we also want to observe whether the employment growth responds symmetrically or
92 asymmetrically to the changes in renewable energy production. To that end, we have applied non-linear Panel-
93 NARDL-PMG, which provides us the opportunity to break down the renewable energy production into their
94 positive and negative shocks; thus, separately calculate their impact on employment growth in Europe. The
95 asymmetric analysis is closer to reality because most of the macroeconomic variables behave nonlinearly due to
96 their vulnerability to external shocks. The non-linear analysis assumes that a positive and negative change may
97 move in the same direction with different magnitudes. Furthermore, to strengthen our results, we will also apply
98 the quantile regression model.

99 This study is a blend of various sections, and their organizations are as follows. In section two, we
100 present data and methods. Then, we offer results in section three. We have provided the conclusion in the fourth
101 section.

102

103 **Model and methods**

104 A majority of the studies in the literature identify renewable energy production as one of the key factors of
105 employment. Hillebrand et al. (2006) revealed that renewable energy production has importance in their
106 formulation of the labor market. Therefore, we follow Arvanitopoulos and Agnolucci (2020) and adopt the
107 following long-run model:

108

$$109 \quad \text{Unemp}_{it} = \omega_0 + \varphi_1 \text{REP}_{it} + \varphi_2 \text{GDP}_{it} + \varphi_3 \text{Investment}_{it} + \varphi_4 \text{Technology}_{it} + \varepsilon_t \quad (1)$$

110

111 Where Unemp_{it} is employment rate, REP_{it} is renewable energy production, GDP_{it} is GDP per capita,
112 Investment_{it} is investment share, and Technolgy_{it} is technology innovation. Since data sampling nations is
113 Europe, we express the “i” and time period “t”. It is normally assumed that an increase in renewable energy
114 production encourages the employment rate. Thus, we expect estimates of φ_1 to be negative. The next phase
115 involves re-writing equation (1) in an error-correction modeling format so that we can also measure the short-
116 run impacts of exogenous variables. Such model is outlined by (2) as follows:

117

$$\begin{aligned} \Delta \text{Unemp}_t &= \omega_0 + \sum_{k=1}^n \beta_{1k} \Delta \text{Unemp}_{2,t-k} + \sum_{k=0}^n \beta_{2k} \Delta \text{REP}_{t-k} + \sum_{k=0}^n \beta_{3k} \Delta \text{GDP}_{t-k} + \\ &\sum_{k=1}^n \beta_{4k} \Delta \text{Investment}_{t-k} + \sum_{k=0}^n \beta_{5k} \Delta \text{Technology}_{t-k} + \omega_1 \text{Unemp}_{it-1} + \omega_2 \text{REP}_{t-1} + \omega_3 \text{GDP}_{t-1} + \\ &\omega_4 \text{Investment}_{t-1} + \omega_5 \text{Technology}_{t-1} + \varepsilon_t \end{aligned} \quad (2)$$

Specifications (2) is due to Pesaran et al. (2001) where coefficients are attached to “first-differenced” variables and the long-run coefficients by the estimates of ω_2 - ω_5 normalized on ω_1 . However, for the consistency and validity of long-run estimates, we must establish cointegration. Two famous tests are recommended in the literature (Pesaran et al. 2001). The F test to establish the joint significance of lagged level variables and the t-test or ECM to establish the significance of ω_1 in equation (2). The central assumption in the error-correction model (3) is that the impact of renewable energy production on the unemployment rate is symmetric. Shin et al. (2014) modify such a type of econometric model so that we can easily judge the possibility of asymmetric effects of renewable energy production on the unemployment rate.

$$\text{REP}^+_t = \sum_{n=1}^t \Delta \text{REP}^+_t = \sum_{n=1}^t \max(\text{REP}^+_t, 0) \quad (3a)$$

$$\text{REP}^-_t = \sum_{n=1}^t \Delta \text{REP}^-_t = \sum_{n=1}^t \min(\Delta \text{REP}^-_t, 0) \quad (3b)$$

Thus, the concept of the partial sum is employed to create two new time series as, REP^+_t and REP^-_t . A positive change in renewable energy production and reflects only an increase in renewable energy production. By the same token, REP^-_t is the partial sum of negative change and reflects on the decreased renewable energy production. In the next step, we move back to (2) and replace REP_{it} with the partial sum of two new variables to arrive at:

$$\begin{aligned} \Delta \text{Unemp}_{2,t} &= \omega_0 + \sum_{k=1}^n \delta_{1k} \Delta \text{Unemp}_{2,t-k} + \sum_{k=0}^n \delta_{2k} \Delta \text{REP}^+_{t-k} + \sum_{k=0}^n \delta_{3k} \Delta \text{REP}^-_{t-k} + \\ &\sum_{k=0}^n \beta_{4k} \text{GDP}_{t-k} + \sum_{k=0}^n \beta_{5k} \text{Investment}_{t-k} + \sum_{k=0}^n \beta_{6k} \text{Technology}_{t-k} + \omega_1 \text{Unemp}_{t-1} + \omega_2 \text{REP}^+_{t-1} + \\ &\omega_3 \text{REP}^-_{t-1} + \omega_4 \text{GDP}_{t-1} + \omega_5 \text{Investment}_{t-1} + \omega_6 \text{Technolghy}_{t-1} + \varepsilon_t \end{aligned} \quad (4)$$

Since assembling the two new partial sum variables of the renewable energy production, models like equation (4) is referred to as asymmetric ARDL model whereas, that like equation (2) is called asymmetric ARDL model. Shin et al. (2014) employed a similar estimation approach and the same diagnostic for both linear and nonlinear models. Once equation (5) is estimated, a few nonlinearity assumptions could be tested. First, short-run asymmetry can be established if ΔREP^+ and ΔREP^- take a different lag order, i.e., if $\delta_{2k} \neq \delta_{3k}$. Second, short-run asymmetric impacts of renewable energy production will be established by using Wald test, if at any given lag order k, the estimate of δ_{2k} attached to $\Delta \text{REP}^+_{t-k}$ is different than the estimate of δ_{3k} attached to $\Delta \text{REP}^-_{t-k}$. Lastly, the long-run asymmetric impacts between the positive and negative changes are established if, once again, we can reject the null hypothesis of the Wald test ($\frac{\omega_2}{-\omega_1} = \frac{\omega_3}{-\omega_1}$).

Some scholars also apply the quantile technique on the panel data and give large attention in theory and application. Quantile regression has become a workhorse in recent years and has been widely used in energy and environmental empirical research. Koenker & Bassett (1978) offers an insight into the relationship of the response variable distribution on the covariates in the various quantiles in Quantile regression. It is noted that fat

154 tails or distinct peaks often exist in the data, while the panel data model is mostly estimated based on conditional
 155 mean regression. To alleviate this difficulty, we utilize panel quantile regression to estimate our non-linear
 156 econometric model. The quantile regression provides more robust estimation results for panel data (Koenker and
 157 Bassett, 1978). This method allows exploring a range of conditional quantiles by capturing asymmetric effects.
 158 Koenker (2004) proposes a different quantile regression model based on the fixed-effects panel data, where the
 159 different term is added to the objective function to decrease the individual impacts and maintain the asymptotic
 160 normality of the estimator.

161 The panel asymmetric quantile regression also deals with the outliers and offers robust outcomes.
 162 Panel quantile regression delivers a separate influence of concern variables on unemployment due to varied
 163 quantiles. Panel quantile regression also explores unobserved heterogeneity for each cross-section and measures
 164 various parameters in different quantiles.

165

166 **Data**

167 The study aims to investigate the impact of renewable energy production on employment growth in European
 168 economies (Germany, Italy, France, Sweden, Spain, Norway, UK, Finland, Austria, Poland) for the period 1991
 169 to 2019. Unemployment as a percentage of the total labor force is used to measure unemployment growth.
 170 Renewable energy production is measured as renewables and others in quad Btu. GDP, investment, and Patent
 171 are control variables in this study. GDP is measured as GDP per capita at constant 2010 US\$. Gross fixed
 172 capital formation as a percentage of GDP is used as a proxy variable for investment. A patent variable is used as
 173 a proxy variable for technological innovation which is measured as patent applications in a number of total
 174 residents and non-residents. Data on all variables are extracted from the World Bank.

175

176 **Table 1: Variables and definitions**

Variables	Symbol	Definitions	Std.			
			Mean	Dev.	Min	Max
Unemployment	Unemp	Unemployment, total (% of total labor force) (modeled ILO estimate)	8.874	4.567	2.490	26.09
Renewable energy production	REP	Renewables and other (quad Btu)	0.622	0.437	0.019	2.331
Gross domestic product	GDP	GDP per capita (constant 2010 US\$)	10.50	0.528	8.614	11.43
Investment	Investment	Gross fixed capital formation (% of GDP)	21.39	2.711	14.97	30.02
Technology innovation	Patent	Patent applications, total (residents and nonresidents)	8.825	1.095	7.221	11.12

177

178

179 **Results and Discussion**

180 As the objective of the study is to empirically determine the impact of renewable energy production on
 181 unemployment for a sample of selected European economies. Before executing regression analysis, it is
 182 imperative to test the stationary properties of data. Hence, panel unit root testing approaches such as LLC, IPS,
 183 and ADF unit root tests are employed in the study. Table 2 reports the outcomes of all three-panel unit root
 184 testing approaches. It is obvious that some series are stationary at the level and some of them are first difference
 185 stationary. However, none of the series is stationary at I(2). In this regard, the study adopted panel ARDL-PMG

186 and panel NARDL-PMG approaches for empirical investigation. Table 2 reports the empirical outcomes of
 187 panel ARDL-PMG and panel NARDL-PMG along with findings of some necessary diagnostics tests.

188

189 **Table 2: Panel unit root testing**

	LLC		IPS		ADF				
	I(0)	I(1)	I(0)	I(1)	I(0)	I(1)			
Unemp	-1.935**	I(0)	-2.752***	I(0)	-2.832***	I(0)			
REP	-1.035	-6.325***	I(1)	0.987	-8.356***	I(1)	-1.002	-7.655***	I(1)
GDP	-1.032	-3.987***	I(1)	-0.987	-5.988***	I(1)	-0.782	-6.598***	I(1)
Investment	-2.352***	I(0)	-3.565***	I(0)	-3.255***	I(0)			
Patent	-0.231	-2.987***	I(1)	-0.987	-5.988***	I(1)	-0.325	-5.032***	I(1)

190 **Note:** ***p<0.01; **p<0.05; and *p<0.00

191

192 The long-run findings of panel ARDL-PMG reveal that renewable energy production is negatively
 193 associated with unemployment confirming that in response of 1 percent increase in renewable energy
 194 production, unemployment reduces by 1.029 percent in a sample of selected European economies. In case of
 195 control variables, GDP, investment, and patent variables exert a significant negative impact on unemployment
 196 revealing that 1 percent increase in GDP, investment and technological innovation lead to 1.086 percent, 0.789
 197 percent and 1.534 percent reduction in unemployment in these economies in the long-run. The short-run
 198 findings of panel ARDL demonstrate that renewable energy production and patent variables have no significant
 199 impact on unemployment as the coefficient estimates of these two variables are statistically insignificant.
 200 However, GDP and investment have a significant negative impact on unemployment confirming that increase in
 201 investment and GDP results in increasing employment growth in these economies in the short-run. To confirm
 202 the findings of panel ARDL, the study performs some diagnostic tests such as Log-likelihood test, ECM, and
 203 Kao test. The coefficient estimate of Log-likelihood confirms the overall goodness of fit of the model.
 204 Furthermore, the findings of ECM and Kao test confirm the existence of long-run cointegration among
 205 variables. The coefficient estimate of ECM holds a negative sign as required for convergence, with a value
 206 0.241, which states that in a period of 1 year almost 24 percent convergence towards equilibrium will be
 207 achieved.

208 The study also scrutinizes the asymmetric impact of renewable energy production on employment
 209 growth in selected European economies. The long-run findings of panel NARDL demonstrate that positive
 210 shocks in renewable energy production have a significant negative impact on unemployment revealing that due
 211 to 1 percent increase in renewable energy production, unemployment decreases by 0.943 percent. In contrast,
 212 the negative shock is renewable energy production has a significant positive impact on unemployment showing
 213 that due to 1 percent increase in the negative component of renewable energy production, unemployment
 214 increases by 1.785 percent in the long-run.

215 This finding is reliable with Arvanitopoulos and Agnolucci (2020), who noted that renewable energy
 216 production contributes to stimulating employment in the United Kingdom. This study also reported that 1 GWh
 217 increase in annual renewables creates 4.7 new jobs in the short-run period and 3.5 jobs in the long period.
 218 Azretbergenova et al. (2021) suggest that renewable energy investments generate thousands of job opportunities
 219 around the globe in the modern era. A similar finding is also found by Lehr et al. (2008) for Germany, who
 220 infers that renewable energy production has directly and indirectly increased employment. Renewable energy

221 production is sometimes considered a win-win scenario for economic welfare and the environment, as they
 222 reduce CO2 emissions and create employment in the economy. The effect of renewable energy production on
 223 the unemployment rate is not much smaller in absolute terms in Europe. Moreno & Lopez (2008) noted that
 224 renewable energy production offers the opportunity to reduce the emission of CO2 and create new employment.
 225 The finding also infers that renewable energy production increases output level by reducing unemployment in
 226 the economy. The negative effect can be explained by income and technology effects in economic theory.

227 As it is assumed that the renewable energy sector has increased business opportunities rapidly with also
 228 increasing the green jobs in Europe. Our finding also infers that the renewable energy sector itself has a direct
 229 impact on employment in the economy. Indirect employment includes employment in each sector of the
 230 economy in the supply chain. Destek et al. (2020) reveal that green energy also reduces the unemployment rate
 231 in Canada, Israel, France, New Zealand, and Mexico via direct and indirect transmission channels.

232 However, GDP, investment, and technological development significantly tend to reduce unemployment
 233 in these economies in the long-run. As it is shown that in response of 1 percent increase in GDP, investment,
 234 and technological innovation, unemployment reduces by 1.374 percent, 0.885 percent, and 2.151 percent,
 235 respectively. The short-run outcomes of panel NARDL reveal that positive shock in renewable energy
 236 production has a significant negative impact on unemployment, however, the negative shock in renewable
 237 energy production has reported no impact on unemployment. In the case of control variables, GDP and
 238 investment result in reducing unemployment but, patent variable reports no significant association between
 239 technological innovation and unemployment in the short-run. The findings of diagnostic tests also validate the
 240 empirical outcomes of NARDL. As the findings of Log-likelihood confirm the goodness of fit of the model. The
 241 findings of ECM and Kao test confirm the existence of long-run cointegrating association among variables in
 242 the long-run. The negative coefficient value of ECM demonstrates that almost 23 percent stability will be
 243 attained in the period of one year. The findings of Wald test also report that the asymmetries exist among
 244 dependent and independent variables in the long-run and short-run.

245

246 **Table 3: ARDL and NARDL estimates**

Variable	ARDL-PMG				NARDL-PMG			
	Coefficient	Std. Error	t-Statistic	Prob.*	Coefficient	Std. Error	t-Statistic	Prob.*
Long-run								
REP	-1.029**	0.494	-2.084	0.038				
REP_POS					-0.943**	0.470	-2.007	0.046
REP_NEG					-1.785*	1.072	1.665	0.100
GDP	-1.086	0.509	-2.133	0.035	-1.374*	0.582	-2.360	0.033
INVESTMENT	-0.789***	0.087	-9.104	0.000	-0.885***	0.086	-10.30	0.000
PATENT	-1.534***	0.520	-2.949	0.004	-2.151***	0.634	-3.391	0.001
Short-run								
D(REP)	0.665	0.999	0.665	0.507				
D(REP_POS)					-0.017*	0.009	1.888	0.078
D(REP_NEG)					-1.902	2.618	-0.726	0.468
D(GDP)	-1.220	0.443	-2.753*	0.005	-2.600***	0.668	-3.892	0.004
D(INVESTMENT)	-0.161**	0.079	-2.033	0.043	-0.127**	0.070	-1.814	0.080

D(PATENT)	0.492	1.215	0.405	0.686	0.241	1.402	0.172	0.864
C	58.44***	14.04	4.161	0.000	56.51***	14.31	3.948	0.000
Diagnostics								
Log likelihood	-166.18				-158.9			
ECM(-1)	-0.241***	0.058	-4.168	0.000	-0.231***	0.059	-3.940	0.000
Kao-cointegration	-3.255***				-3.589***			
Wald-LR					4.658***			
Wald-SR					3.878***			

247 **Note:** ***p<0.01; **p<0.05; and *p<0.0

248

249 Table 4 portrays the results of quantile regression estimates. Although we have presented the complete
250 results of all our variables; however, to save space, we only discuss the results of our primary variable. In total,
251 we have included 11 quantiles. The estimates of REP are negative significant in symmetric quantile regression.
252 In addition, the asymmetric effects of renewable energy production on unemployment growth are robust in the
253 different quantile regression.

254

255 **Table 4: Panel quantile regression**

	0.05	0.10	0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90	0.95
Models without asymmetry											
REP	-0.969***	-1.376***	-1.971***	-2.384***	-3.187***	-3.781***	-4.589***	-5.565***	-6.394***	-6.017***	-7.763***
GDP	0.041	0.016	-0.172	-0.311***	-0.231	0.021	0.365	0.489	0.951**	3.297***	6.277***
INVESTMENT	0.023	0.074	0.108	0.172**	0.193**	0.112	-0.004	-0.086	-0.232*	-0.322*	-1.016**
PATENT	0.401**	0.411**	0.693***	0.817***	0.828***	0.848***	0.895***	1.144***	1.154***	-1.063	-2.187*
Models with asymmetry											
REP_POS	-1.399***	-1.909***	-2.008***	-2.795***	-3.408***	-3.581***	-4.119***	-4.515***	0.755	-5.069	5.571
REP_NEG	-0.905	-2.068***	-2.414***	-2.997***	-3.845***	-3.168**	-3.918***	-3.405**	4.421	-7.697	3.061*
GDP	-0.479	-0.900***	-1.101***	-1.257***	-1.188***	-0.497	-0.240	0.021	0.187	2.049	3.052
INVESTMENT	0.131	0.246***	0.285***	0.331***	0.290***	0.195	0.140	0.055	0.102	-0.231	-0.132
PATENT	0.759***	1.054***	1.263***	1.440***	1.557***	1.097***	1.043***	1.085***	0.879*	-0.060	-0.930

256 **Note:** ***p<0.01; **p<0.05; and *p<0.00

257

258

259

260 Conclusion and implications

261 Due to environmental degradation and energy supply sanctuary issues, the energy sector has
262 transformed towards renewable energy production. The increasing trend of investments in the field of renewable
263 energy creates substantial support to the development and economic growth by increasing production levels and
264 by generating more employment. The current study examines the effect of renewable energy production on the
265 employment growth in the case of European countries namely Austria, Finland, France, Germany, Italy,
266 Norway, Poland, Spain, and Sweden for period 1991 to 2019. For this purpose, the study employed panel ARDL
267 and panel NARDL regression techniques. GDP, investment, and technological innovation are also incorporated
268 as control variables in the study. The long-run findings of panel ARDL report that due to an increase in

269 renewable energy production, employment growth increases significantly. However, short-run findings of
270 ARDL do not report any significant impact of renewable energy production on employment growth.

271 The long-run outcomes of NARDL demonstrate that positive shock in renewable energy production
272 leads to employment growth in these economies however, negative shock in renewable energy production
273 results in increased unemployment in the long-run. The short-run findings of NARDL also report employment
274 growth due to positive changes in renewable energy production. The finding does not report any significant
275 impact of negative shock of renewable energy production on employment growth in the short-run. In case of
276 control variables, the findings of panel ARDL and panel NARDL models reveal that any increase in GDP,
277 investment and Patent variables result in a reduction of unemployment in these economies in the long-run.
278 However, in the short, the findings of panel ARDL and panel NARDL models reveal that GDP and investment
279 result in increasing employment growth but patent has no significant impact on employment growth.

280 In short, renewable energy production and other variables exert a significant positive impact on
281 employment growth in the sample of selected European economies. On the basis of these findings, the study
282 proposed several policy recommendations. Firstly, capital investment in renewable energy sector-related
283 technologies should be increased. Secondly, the joint projects, support projects, statistical transfers, and
284 collaborations should be strengthened among these selected economies on renewable energy production.
285 Thirdly, the utilization of fossil fuels should be minimized and the appropriate domains for the renewable
286 energy sector amenities should be determined. Lastly, the governments should adopt such strategies that fulfill
287 the necessities of the energy sector, containing the training of future workforces that improves regional
288 competitiveness and generates positive influences on the labour market. In addition, the European governments
289 should also provide some tax benefits, price guarantees, subsidies, and easy financing opportunities inspiring the
290 green energy sector to increase its employment level.

291

292 **Declarations**

293 **Ethical Approval**-Not applicable.

294 **Consent to Participate**-Not applicable.

295 **Consent to Publish**-All authors have read and approved the manuscript to publish with this journal.

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297 analysis and result: Sameen Naqvi and Jinchao Wang. Supervision, proofreading and review: Jinchao Wang and
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301 **Availability of data and materials**- Data is available from the corresponding author upon request.

302

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