## The long-term effect of renewable electricity on UK employment

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Image: A matrix

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#### Overview

► Literature assesses **employment impact of renewable** technologies using complex/data-intensive or simplistic methods → contradicting results

 $\blacktriangleright$  Contribution  $\rightarrow$  introduce a **transparent** and **easily reproducible econometric** methodology using aggregated and widely available data

- Results indicate that renewables:
  - generate the most sustainable jobs in the long-term period
  - stimulate 6 times more jobs in relation to nuclear

▶ What is the employment effect of the UKERC decarbonisation scenarios?

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#### Introduction

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## Introduction

- Increasing maturity of renewables and rising number of jobs created
- ► Are jobs created sustainable in the long run?
- ▶ Impact of **renewables** on net employment:
  - Literature focuses on specific technologies, location and plants
  - Complicated methods (CGE and IO) or very simple methods (employment factors)
  - Scarcity of data
  - $\bullet\,$  Importance of country specific characteristics  $\to\,$  cannot export employment factors from one country to another
- Proposing a simple and robust econometric method form aggregated data

 $\blacktriangleright$  UK electricity generation sector annual data from 1990 to 2016  $\rightarrow$  can be easily applied to other OECD countries

#### Literature review

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### Literature review

► Cameron and Zwaan (2015) identify 70 publications since early 2000:

- studies producing forecast or simulations based on theoretical models
- literature reviews
- input-output and employment factors
- But all of them focusing solely on renewable electricity
- $\blacktriangleright$  Employment effect of wind energy  $\rightarrow$  no consensus over the long-run

... Lack of robust empirical analyses comparing the macroeconomic net employment effect of conventional thermal power generation and renewable technologies

## **UK electricity supply**

## UK electricity supply

Employment

#### Employment in power generation sector (thousands jobs)



 $\blacktriangleright$  UK electricity market **privatised in 1990**  $\rightarrow$  a competitive bidding system lowered energy prices (DUKES 2017)

## UK electricity supply

Electricity supply per technology (GWhs)



 $\bullet\,$  "Dash for gas"  $\rightarrow$  massive transition from coal to gas in the 90s

#### UK electricity supply Electricity supply - Renewables (GWhs)



Renewable electricity has been steadily increasing in the late 2000s

UK has the largest global capacity in off-shore wind energy (BEIS, 2018)

#### UK electricity supply Variables and data sources

► Number of jobs is measured by "workforce jobs", with data sourced from employer surveys like the ONS Labour Force Survey

- ► GVA for the industrial sector incorporating all Major Power Producers (MPPs)
- ► Electricity supply (DUKES, 2017) generated by:
  - conventional thermal
  - CCGT
  - nuclear
  - renewables
- ► Electricity supply is equal to the level of electricity supplied to end users in the UK generated by MPPs
- ► Timespan: 1990-2016

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Theoretical modelling I

 $\blacktriangleright$  UK electricity system is driven by demand (DUKES 2017)  $\rightarrow$  supply is completely elastic

► Adaptive expectations (Nerlove, 1958)  $\rightarrow$  representative firm chooses labour inputs  $L_t$  based on the previous period's expectations about electricity supply in t

$$L_t = f(E_{t-1}[e_t|I_{t-1}])$$
(1)

Firm's expectations about  $e_t$  can be further distinguished in 4 subcategories: conventional thermal, CCGT, nuclear and renewable electricity

$$E_{t-1}[e_t|I_{t-1}] = E_{t-1}[con_t|I_{t-1}] + E_{t-1}[ccgt_t|I_{t-1}] + E_{t-1}[nuc_t|I_{t-1}] + E_{t-1}[ren_t|I_{t-1}]$$
(2)

Theoretical modelling II

▶ Representative firm forms expectations at t by considering demand for electricity observed at t - 1 and all past years j

$$E_{t-1}[e_t|I_{t-1}] = \beta(\sum_{j=0}^{\infty} (1-\beta_1)^j e_j),$$
(3)

where error-adjustment term  $\beta$  takes values between 0 an 1 and reflects deviations between expectations and reality

: expectation of higher  $e_t \rightarrow$  increase in production inputs and thus workforce.

► Nevertheless, a positive employment effect in time *t* might not only be the outcome of higher electricity consumption but also of higher economic activity

... We also control for gross value added (GVA) in the electricity generation sector

Econometric modelling I

▶ Unit root (DFGLS and Zivot and Andrews 1992) and cointegration (Johansen 1998, 1991) testing  $\rightarrow$  variables integrated of order I(1) and cointegrated

► Cointegration analysis using a VAR approach (Johansen 1988, 1991), and estimate a Vector Error Correction (VECM) model of order p, where all variables are treated as endogenous

$$\Delta x_t = \Gamma_0 + \Pi x_{t-1} + \sum_{i=1}^p \Gamma_i X_{t-1},$$
(4)

where  $x_t$  is a 6 x 1 vector containing the logarithms of employment, GVA, and electricity generated by 1) conventional thermal, 2) CCGT, 3) nuclear and 4) renewable technologies,  $\Pi$  and  $\Gamma_i$  are 6 x 6 coefficient matrices and  $\Gamma_0$  contains the deterministic terms

► Impulse response functions (IRFs)

► We examine the **long-term response of employment** to 1 GWh shock applied to each type of electricity supply independently and then we compare the employment effects

► Generalised impulse response function (Koop et al, 1996) which is invariant to the ordering of the variables in the VAR and "fully takes into account the historical patterns of the correlation observed amongst the different shocks" (Pesaran and Smith, 1998)

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▶ Table 1. Cointegrating vectors β from the VECM specifications VECM 1 and VECM 2

VEC	CM 1							
	Jobs	GVA	Conventional	CCGT	Nuclear	Renewables	Trend	Constant
$\beta_1$	1	-0.96					-0.35	2.8
$\beta_2$			1	0.31			0.89	-28.27
$\beta_3$			1		1.36		-0.12	-25.34
$\beta_4$					1	0.22	-1.02	1.58
VEC	CM 2							
	Jobs	GVA	Conventional	CCGT	Nuclear	Renewables	Trend	Constant
$\beta_1$	1	-1.0					0.06	-2.32
$\beta_2$			1	0.46	0.76	0.3	-0.001	-28.25

- All variables I(1) and Johansen test indicates at least 4 coint relationships
- Positive scale effect between output and employment
- Substitution effect between conventional thermal and rest of technologies
- Concerns about 4 vectors  $\rightarrow$  try 2 vectors

► **Table 2**. P-values of the Likelihood Ratio tests for the coefficients in the cointegrating vectors presented in Table 1

	GVA	CCGT	NUC	REN	Trend $\beta_1$	Trend $\beta_2$	Trend $\beta_3$	Trend $\beta_4$	All trends
VECM 1	0	0	0	0	0.04	0.31	0.15	0.07	0
VECM 2	0	0	0	0	0	0.86			0

► Table 3. Diagnostic tests for the residuals of the VECMs cointegrating vectors presented in Table 1

	Lags	Serial correlation	Heteroskedasticity
VECM 1	1	0.04	0.31
VECM 2	1	0.17	0.4

- long-term coefficients are all strongly statistically
- VECM 1 residuals serially correlated
- VECM 2 no heteroskedasticity and serial correlation in the residuals

- ▶ 4 cointegrating vectors:
  - unexpectedly high values of the trend in  $\beta_3$  and  $\beta_4$
  - residuals serially correlated
  - restricting one coefficient implies considerable **instability** in the other  $\beta$  coefficients
  - plenty of indication about spurious results
- ► 2 cointegrating vectors:
  - restricting one coefficient does **not** affect the other  $\beta$  coefficients
  - failure to detect heteroskedasticity and serial correlation in the residuals
  - restricting one coefficient does not affect the stability of the model
  - VECM 1 and VECM 2 have coefficients of the same sign and similar value  $\rightarrow$  proves the overall stability of the model

▶ Employment response to 1 GWh electricity supply permanent increase



 $\blacktriangleright$  1GWh permanent increase in renewable electricity supply creates 3.5 jobs in the long run  $\rightarrow$  6 times more than nuclear

## **Policy implications**

## **Policy Implications**

► Employment effect for the UKERC energy security scenarios (Watson et al, 2018)

Scenario	CCGT	Nuclear	Renewables	Net employment	Gross employment
1) Energy island	-21,492	19,303	-10,219	-12,408	118,432
2) Slow decarbonisation	-9,278	-19,089	60,034	31,668	162,418
3) Low carbon	4,499	-19,089	53,549	38,960	169,024
4) Low carbon (no BECCS)	14,956	-19,089	16,124	11,992	142,742
5) Technology optimism	-17,886	-19,089	75,136	38,162	168,912

- We use the long-run employment effect estimates on the 2030 UK decarbonisation scenarios for the UK energy generation sector
- Energy island which is "inward-looking" and potentially related to no-deal BREXIT  $\rightarrow$  reduction in employment (-12,000 jobs)
- Low carbon UK meets climate change goals  $\rightarrow$  highest level of employment (+39,000 jobs)

... Scenarios based overwhelmingly on renewable electricity result in a significant net employment effect

#### **Conclusions**

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## Conclusions

▶ 1 GWh permanent increase in renewable electricity supply creates 3.5 jobs in the long-term, i.e. about six times the number of jobs created by an equally sized increase in nuclear generation.

► Jobs created by the deployment of **renewable technologies** are the **most sustainable in the long-term period** 

▶ UKTM scenarios for electricity generation in 2030  $\rightarrow$  scenarios based on renewables can generate net effect of 39.000 jobs

... Policy-makers should incentivise and support the deployment of renewable electricity technologies

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### Thank you!



#### ► Unit root test results for Major Power Producers (MPPs)

	Logs			First differences			Logs		First differences	
	DF-GLS test	lags	Determ comp	DF-GLS test	lags	Determ comp	ZA test	lags	ZA test	lags
Jobs	-1.46	1	Trend	-3.91 (*)	0	Trend				
GVA	-1.18	2	Trend	-6.43 (**)	0	Trend				
Conv. thermal	-1.43	0	Trend	-1.49	2	Trend	-1.86	2	-7.31 (**)	0
CCGT	-1.97	1	Trend	-2.81	1	Trend	-3.67	0	-4.59 (*)	1
Nuclear	-2.27	0	Trend	-1.85	3	Trend	-3.45	3	-10.19 (**)	0
Renewables	-1.57	0	Trend	-6.15 (**)	0	Trend				
(+),(*),(**) in the superscripts indicate significance of the test statistics of the unit root tests at 90%, 95% and 99% significance level, respectively.										



#### ► Johansen test cointegration results

	Trace				Max Eigenvalue			
	H0	H1	trace	p-value	H0	H1	max	p-value
MPPs	r = 0	$r \geq 1$	0.973 (**)	0	r = 0	r = 1	0.973 (**)	0
	$r \leq 1$	$r \geq 2$	0.897 (**)	0	r = 1	r = 2	0.897 (**)	0
	$r \leq 2$	$r \geq 3$	0.827 (**)	0	r = 2	r = 3	0.827 (**)	0
	$r \leq 3$	$r \geq 4$	0.737 (**)	0	r = 3	r = 4	0.737 (**)	-0.01
	$r \leq 4$	$r \geq 5$	0.507	-0.14	r = 4	r = 5	0.507	-0.11

#### Appendix VECM A

 $\blacktriangleright$  Cointegrating vectors  $\beta$  from alternative VECM specifications to those presented in Table 1

	Jobs	GVA	Conv	CCGT	Nuclear	Renewables	Trend	Constant
$\beta_1$	1	-0.73					0.05	-5.32
$\beta_2$			1	0.56				-18.17
$\beta_3$			1		1.3		0.03	-26.9
$\beta_4$					1	0.64	-0.02	-16.8
VEC	CM 1B							
	Jobs	GVA	Conv	CCGT	Nuclear	Renewables	Trend	Constant
$\beta_1$	1	-1.03					0.21	-4.76
$\beta_2$			1	0.35			-0.3	-11.38
$\beta_3$			1		1.15			-24.71
$\beta_4$					1	0.27	0.37	-19.09
VEC	CM 2A							
	Jobs	GVA	Conv	CCGT	Nuclear	Renewables	Trend	Constant
$\beta_1$	1	-1.06					0.06	-2.27
$\beta_2$			1	0.44	0.77	0.28		-28.19

VECM 1A

#### Appendix VECM A

▶ P-values of the Likelihood Ration tests for the coefficients in the coint vectors  $\beta$ 

	GVA	CCGT	NUC	REN	Trend $\beta_1$	Trend $\beta_2$	Trend $\beta_3$	Trend $\beta_4$	All trends
VECM 1A	0	0	0	0	0.01		0	0.4	0
VECM 1B	0	0	0	0	0.02	0		0	0
VECM 2A	0	0	0	0	0				

Diagnostic tests for the residuals of the VECMs cointegrating vectors

	Lags	Serial correlation	Heteroskedasticity
VECM 1A	1	0.37	0.32
VECM 1B	1	0.03	0.32
VECM 2A	1	0.2	0.38

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#### Appendix Variables



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