

Nuclear operations with a high penetration of renewables: the case of France

Nicolas Astier (PSE & ENPC) and Frank A. Wolak (Stanford University)

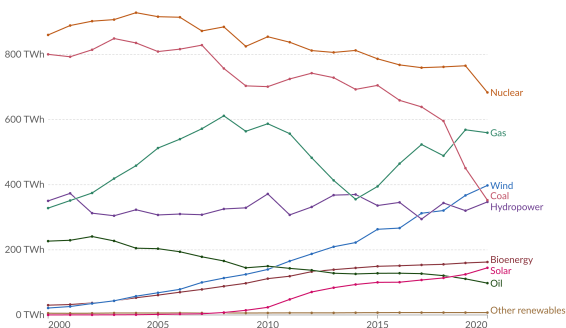
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-
6-7 June 2024

Nuclear and renewables as substitutes in the long-run

Electricity production by source, European Union (27)

Measured in terawatt-hours¹.

Our World
in Data



Data source: Ember (2024); Energy Institute - Statistical Review of World Energy (2023)

OurWorldInData.org/energy | CC BY

Note: Other renewables include waste, geothermal and wave and tidal energy.

1. **Watt-hour:** A watt-hour is the energy delivered by one watt of power for one hour. Since one watt is equivalent to one joule per second, a watt-hour is equivalent to 3600 joules of energy. Metric prefixes are used for multiples of the unit, usually: - kilowatt-hours (kWh), or a thousand watt-hours. - Megawatt-hours (MWh), or a million watt-hours. - Gigawatt-hours (GWh), or a billion watt-hours. - Terawatt-hours (TWh), or a trillion watt-hours.

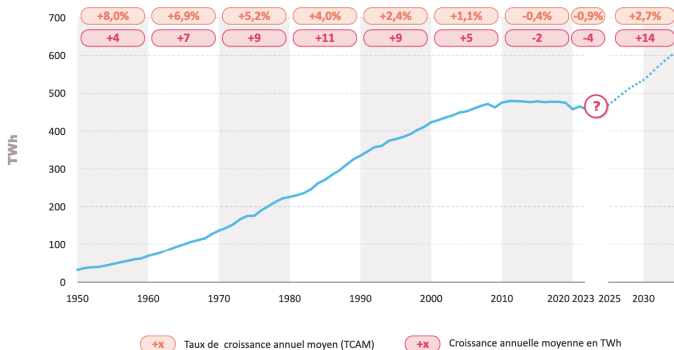
Nuclear and renewables: a (short-run) truce?



(source: <https://medium.com/@Wbgeist/nuclear-energy-a-path-to-sustainable-progress-1ca05b8790c9>)

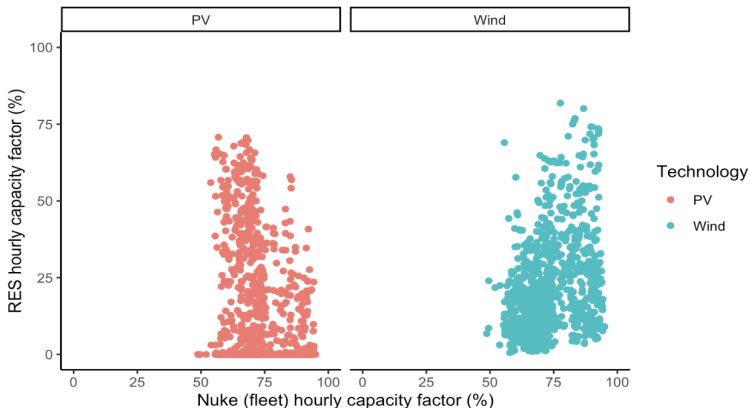
Nuclear and renewables: a (short-run) truce?

Figure 2.10 Consommation intérieure d'électricité en France continentale (données corrigées des aléas météorologiques, hors enrichissement d'uranium)



(source: Bilan prévisionnel 2023, RTE)

But wait a minute...?!



Sample of 1,000 randomly selected 15-min intervals in 2018 in France
(data source: RTE)

This paper

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- **Preliminary results:**
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 - ① an increased exposure to solar generation (mostly distributed in France) is associated with more frequently binding minimum output constraints.
 - ② despite flexible operations, the “supply response” to very low DA prices is far from perfect.

Outline

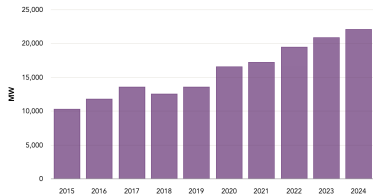
- 1 Motivation
- 2 Background
- 3 Data
- 4 Empirical strategy
- 5 Preliminary results
- 6 Next steps and discussion

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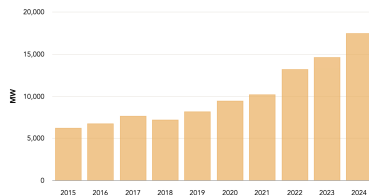
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Wind and solar PV growth

Cumulative on-shore wind capacity:



Cumulative solar PV capacity:

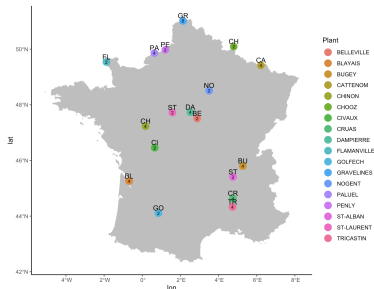


Total capacity of units of 1 MW or more, most of which connects (in France) to the distribution grid (source: RTE)

Remark: as of 2023, total solar PV capacity including installations < 1 MW was about 19 GW.

French nuclear fleet

- Commissioned **between 1979 and 2002**.
- Unit sizes:** 900 MW (32 units), 1300 MW (20 units) or 1500 MW (4 units).



As of today, 18 facilities composed of a total of **56 units** for an aggregate installed capacity of about **61 GW**. [Example](#)

Remark: the two units of Fessenheim were shut down in Feb 2020 and are dropped from the analysis.

Operating constraints on nuclear units

Nuclear units typically face:

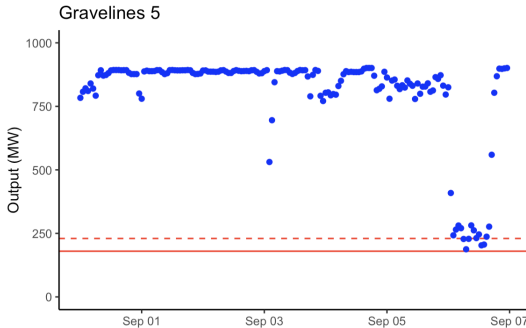
- a **minimum output constraint** (either nominal or driven by other factors), in a context where shutting them down is very costly.
- **ramping constraints**, limiting their ability to follow load variations.

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In France:



France and “flexible” nuclear operations



“Une question m’est régulièrement posée concernant la capacité de modulation des tranches du parc nucléaire français [...] En une trentaine de minutes, jusqu’à 90% d’avancement dans le cycle de production, il est possible d’ajuster à la hausse ou à la baisse 80% de la puissance d’un réacteur nucléaire.”

(Head of the nuclear division at EDF, March 2024)

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⇒ In what follows, we focus on the **minimum output constraint**.

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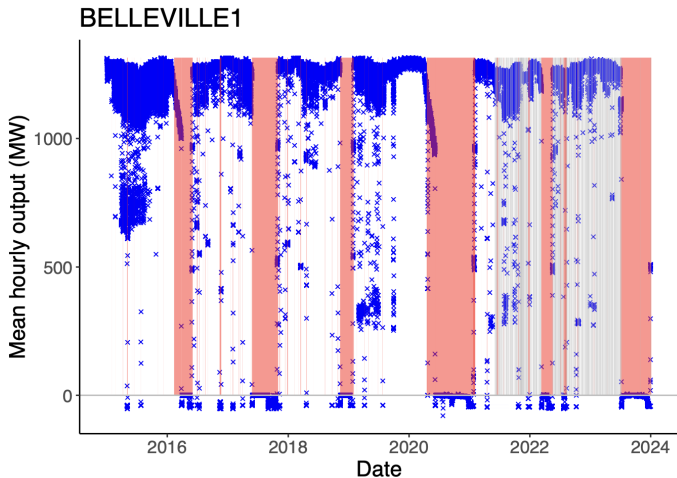
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Data Sources

We use several **public sources** of data:

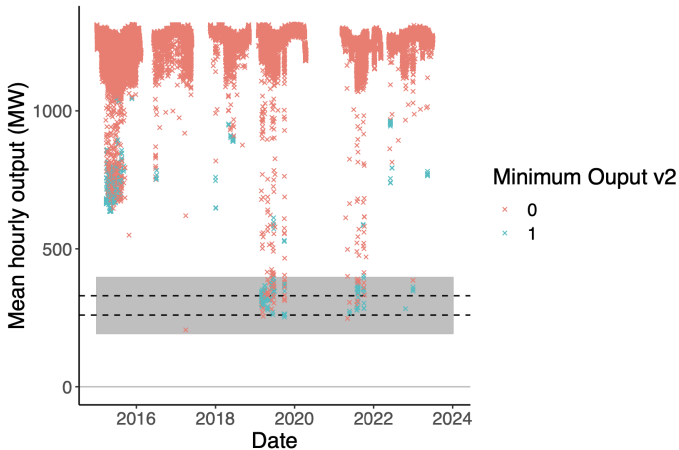
- 1 The **inventory of French power plants** provides the location (municipality), commissioning date, technology and installed capacity of the universe of power plants in France.
- 2 **Unit-level hourly output** for nuclear units between 1 January 2015 and 31 December 2023 (source: RTE). CF France CF USA
- 3 **Unit-level outages** for 2015-2023: start and end dates, type, description (source: RTE and EDF & scrapped data from ASN). Frequency
- 4 **Nuke unit characteristics**: nominal capacity, nominal minimum output level, maximum capacity available for secondary reserve, fuel type (source: EDF).
- 5 Hourly **day-ahead prices** (source: ENTSO-E).

Unit-level output

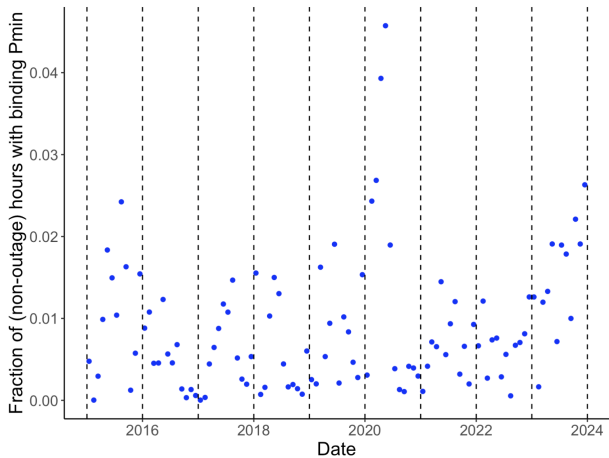


Minimum output constraint - Unit-level

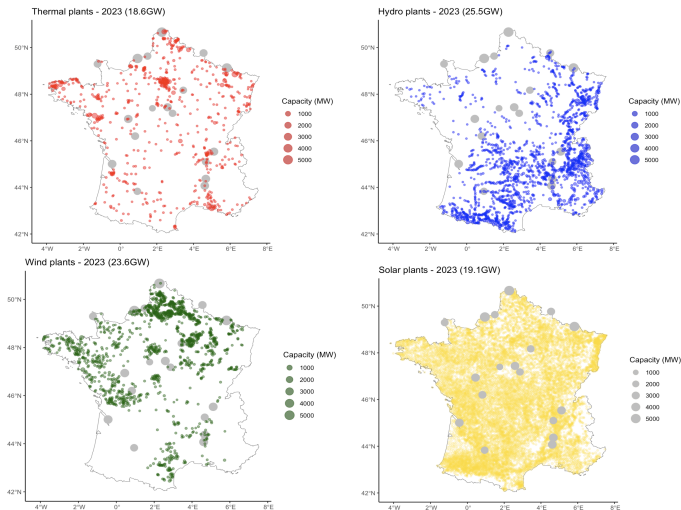
Data kept and inferred constrained region



Minimum output constraint - Frequency



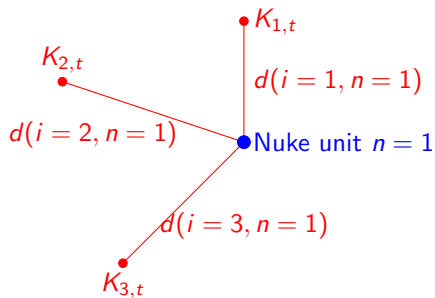
Location of nuke vs other facilities



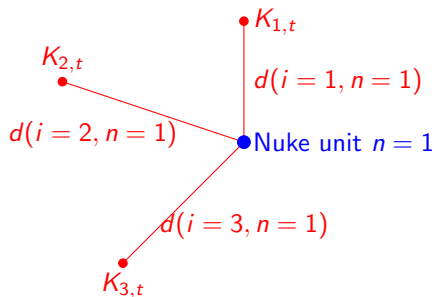
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Exposure of a nuke unit to another technology



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We define the “exposure” $X_{n,t,\tau}$ of nuclear unit n to technology τ at date t as:

$$X_{n,t,\tau} \equiv \sum_{i \in \text{Techno-}\tau \text{ at } t} \left(\frac{\max(d(i, n), d_0)}{d_0} \right)^{-\gamma} K_i$$

where $\gamma = 0.5$ and $d_0 = 1\text{km}$ are tuning parameters.

Exposure of a nuke unit to another technology

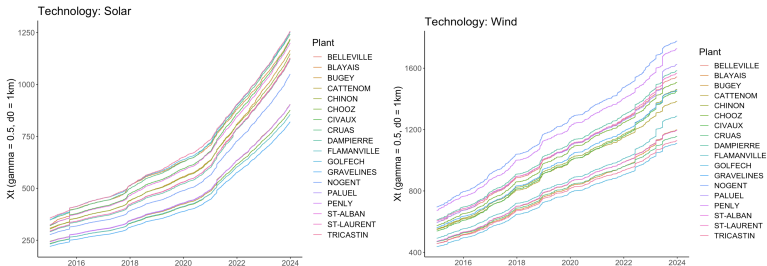


Figure: Obtained exposure metric (in MW) to solar (left panel) and wind (right panel) for the 18 nuclear facilities between 1 January 2015 and 31 December 2023 ($d_0 = 1$ km and $\gamma = 0.5$).

Approach

- **Outcome variable:** $Y_{h,n} \equiv$ dummy variable taking the value 1 if the minimum output constraint is assessed to be binding in hour h for unit n .
- **Main explanatory variables:** $X_{\tau,h,n} \equiv$ exposure metric of unit n in hour h to technology τ , with $\tau \in \{\text{wind, solar, hydro, thermal}\}$.

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- **Models:**

$$Y_{h,n} = \sum_{\tau} \beta_{\tau} X_{\tau,h,n} + \mu_n + \delta_y + \delta_m + \delta_d + \delta_h + \epsilon_{h,n}$$

where the error term is assumed either i.i.d. normal (linear probability model) or i.i.d. Gumbel (logit).

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Regression results

	Dependent variable: Is minimum constraint binding? (Definition 1)							
	Linear probability model				Logit			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
X_PV	0.003 (0.006)	0.01 (0.010)	0.01 (0.011)	0.04 (0.022)	0.3 (0.691) [0.003]	0.8 (1.148) [0.007]	1.5 (1.091) [0.01]	6.0 (2.729) [0.05]
X_Wind	0.004 (0.005)	0.01 (0.009)	0.00 (0.010)	0.01 (0.023)	0.5 (0.545) [0.004]	0.8 (0.957) [0.007]	-0.4 (1.075) [-0.004]	2.0 (2.327) [0.02]
X_Hydro				0.04 (0.075)				9 (9.547) [0.08]
X_Therm				-0.02 (0.016)				-4 (2.565) [-0.03]
Unit FE	N	N	Y	Y	N	N	Y	Y
Year FE	N	Y	N	Y	N	Y	N	Y
Month of year FE	N	Y	N	Y	N	Y	N	Y
Day of week FE	N	Y	N	Y	N	Y	N	Y
Hour of day FE	N	Y	N	Y	N	Y	N	Y
Observations	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091
R ²	0.0003	0.013	0.006	0.019				
Log Likelihood					-134,063	-119,213	-126,495	-111,141
Note:	Robust standard errors clustered at the unit level							

Note:

Robust standard errors clustered at the unit level

Hourly effects LPM

Alternative definition

$\gamma = 0.25$

$\gamma = 0.75$

Remark: X_T have been scaled (/1000) and range from 0.22 to 1.26 for PV and from 0.44 to 1.78 from wind.

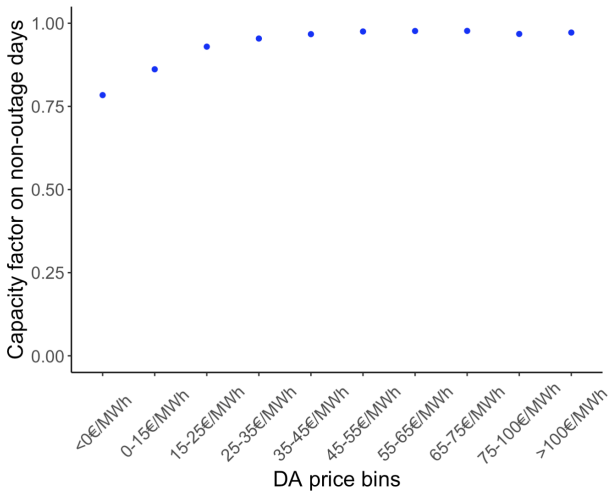
Opportunity cost of inflexibility?



“Une question m’est régulièrement posée concernant la capacité de modulation des tranches du parc nucléaire français [...] En une trentaine de minutes, jusqu’à 90% d’avancement dans le cycle de production, il est possible d’ajuster à la hausse ou à la baisse 80% de la puissance d’un réacteur nucléaire.”

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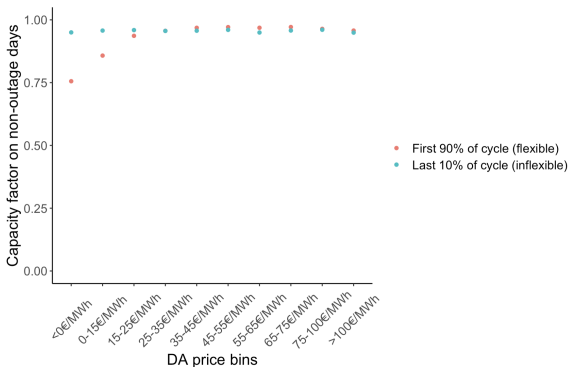


Opportunity cost of inflexibility?

- 1 We try to retrieve unit-level fuel cycles from outage data and keep the most credible ones ($\pm 20\%$ of the unit-level median aggregate output during an imputed cycle).
- 2 We compute capacity factors conditional on being during the 90% first hours and the 10% last hours of the fuel cycle.

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Next steps (non-exhaustive)

1 Back of the envelope monetary costs:

- Minimum output constraint:

$$\underbrace{0.04}_{\uparrow \text{ freq}} \times \underbrace{5\text{€/MWh}}_{\mathbb{E}_h[(MC-p_h)^+]} \times \underbrace{200\text{MW}}_{P_{min}} \times \underbrace{8760}_{\#h/yr} = 350 \text{ k€/unit-yr}$$

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- ## 2 Dynamic minimum output constraint: from mid-June 2021 onwards, minimum output constraints that significantly deviate from the nominal value are publicly reported by EDF. [Example](#)

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① Back of the envelope monetary costs:

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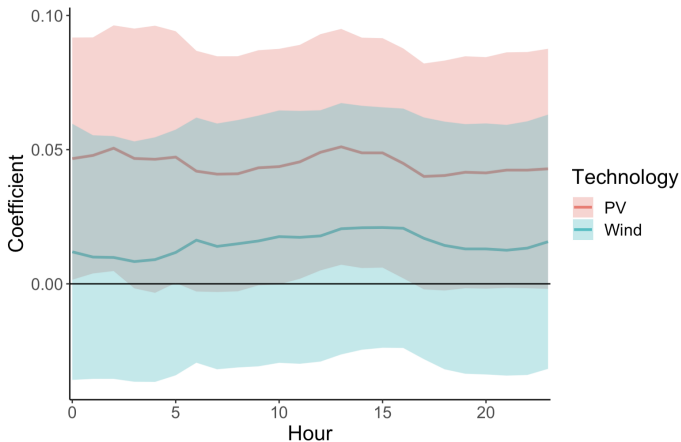
③ Spatial vs electrical proximity: benchmark our “exposure approach” against an approach grounded in electrical engineering (power flow model).

④ Expand the period of analysis: add year 2024 and (possibly) years prior to 2015.

Thank you!

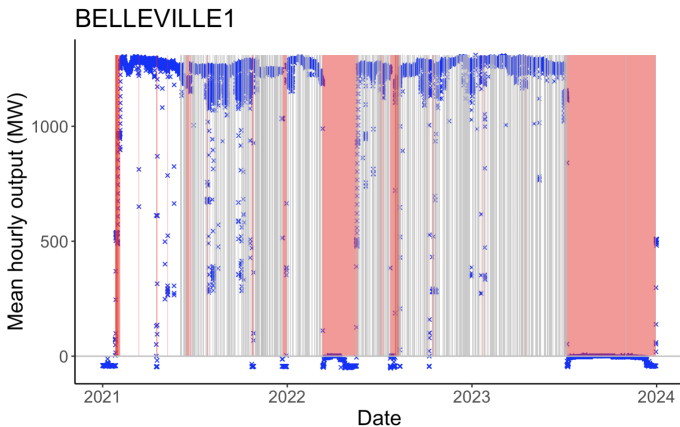
nicolas.astier@psemail.eu

Hourly interactions (LPM)

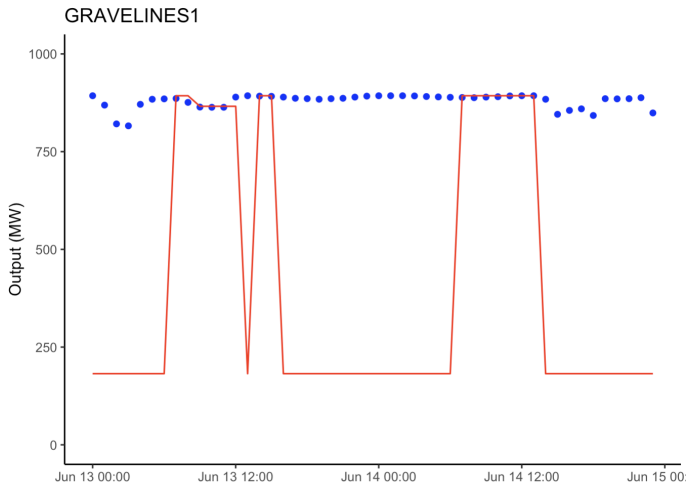


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Dynamic minimum output constraint



Dynamic minimum output constraint



$$\gamma = 0.25$$

<i>Dependent variable: Is minimum constraint binding? (Definition 1)</i>								
	<i>Linear probability model</i>				<i>Logit</i>			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
X_PV	0.001 (0.003)	0.004 (0.005)	0.003 (0.003)	0.020 (0.008)	0.07 (0.273)	0.4 (0.525)	0.3 (0.358)	3 (0.892)
X_Wind	0.001 (0.002)	0.003 (0.004)	-0.001 (0.003)	0.01 (0.009)	0.1 (0.236)	0.3 (0.425)	-0.1 (0.352)	1 (0.908)
X_Hydro				0.05 (0.022)				7.5 (2.827)
X_Therm				-0.008 (0.004)				-1.7 (0.632)
Unit FE	N	N	Y	Y	N	N	Y	Y
Year FE	N	Y	N	Y	N	Y	N	Y
Month of year FE	N	Y	N	Y	N	Y	N	Y
Day of week FE	N	Y	N	Y	N	Y	N	Y
Hour of day FE	N	Y	N	Y	N	Y	N	Y
Observations	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091
R ²	0.0003	0.013	0.006	0.019				
Log Likelihood					-134,091	-119,216	-126,541	-110,796
<i>Note:</i>					Robust standard errors clustered at the unit level			

Back

$$\gamma = 0.75$$

<i>Dependent variable: Is minimum constraint binding? (Definition 1)</i>								
	<i>Linear probability model</i>				<i>Logit</i>			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
X_PV	0.01 (0.017)	0.03 (0.028)	0.04 (0.032)	0.1 (0.060)	1.3 (1.887)	2.8 (3.138)	5.2 (3.043)	13 (7.710)
X_Wind	0.0177 (0.014)	0.027 (0.028)	-0.010 (0.031)	0.005 (0.065)	2 (1.424)	3 (2.718)	-1 (3.068)	1 (6.286)
X_Hydro				-0.1 (0.231)				0 (27.932)
X_Therm				-0.04 (0.063)				0 (27.932)
Unit FE	N	N	Y	Y	N	N	Y	Y
Year FE	N	Y	N	Y	N	Y	N	Y
Month of year FE	N	Y	N	Y	N	Y	N	Y
Day of week FE	N	Y	N	Y	N	Y	N	Y
Hour of day FE	N	Y	N	Y	N	Y	N	Y
Observations	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091
R ²	0.0004	0.013	0.006	0.019				
Log Likelihood					-133,981	-119,123	-126,478	-111,356
Note:	Robust standard errors clustered at the unit level							

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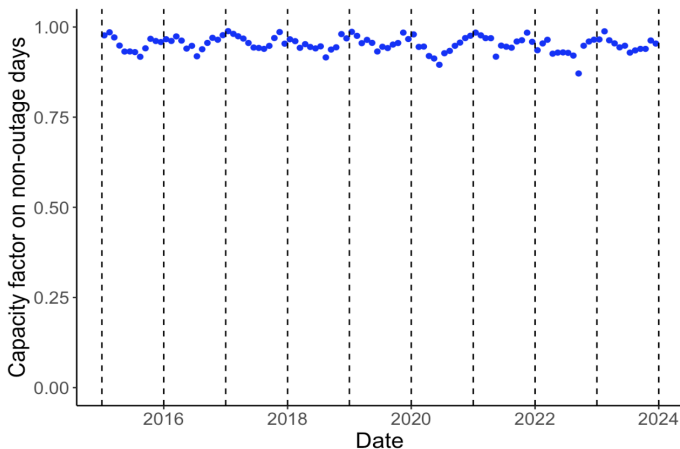
Alternative definition of the minimum output constraint

Minimum output constraint binds in a given hour of a non-outage day if:

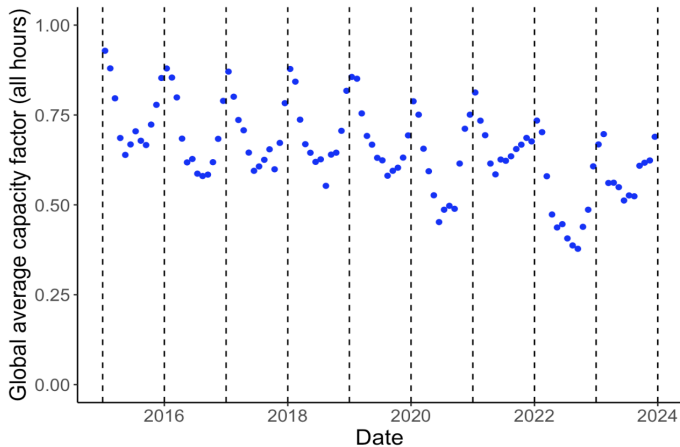
- ❶ output is between 10% and 80% of nominal capacity.
- ❷ the absolute change in output in hour $h - 1$ vs h is lower than 20 MW.

Dependent variable: Is minimum constraint binding? (Definition 2 - $\gamma = 0.5$)								
	Linear probability model				Logit			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
X_PV	0.001 (0.007)	0.00 (0.009)	0.01 (0.014)	0.04 (0.022)	0.1 (0.556)	0 (1.273)	0.5 (1.233)	3.7 (2.158)
X_Wind	0.007 (0.005)	0.00 (0.009)	0.01 (0.014)	0.01 (0.022)	0.6 (0.414)	0.3 (0.735)	0.4 (1.205)	1 (1.909)
X_Hydro				0.0 (0.080)				0 (8.167)
X_Therm				0.01 (0.019)				0 (2.385)
Unit FE	N	N	Y	Y	N	N	Y	Y
Year FE	N	Y	N	Y	N	Y	N	Y
Month of year FE	N	Y	N	Y	N	Y	N	Y
Day of week FE	N	Y	N	Y	N	Y	N	Y
Hour of day FE	N	Y	N	Y	N	Y	N	Y
Observations	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091	2,736,091
R ²	0.0004	0.016	0.006	0.022				
Log Likelihood					-171,658	-153,365	-164,562	-146,055
Note:	Robust standard errors clustered at the unit level							

Capacity factor conditional on being on a non-outage day



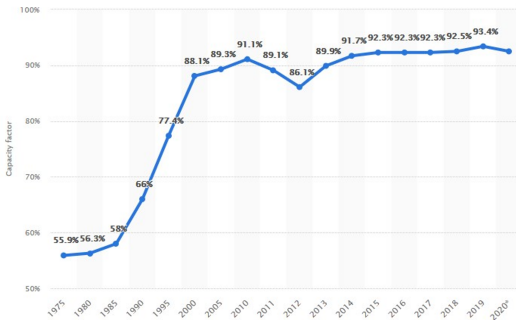
Unconditional capacity factor



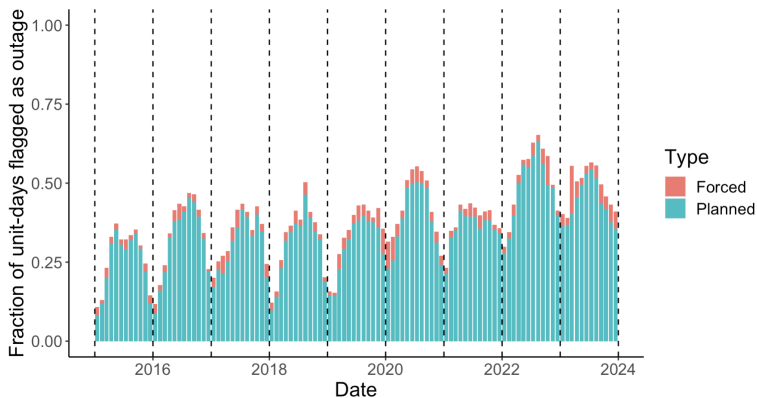
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Unconditional capacity factor

U.S. Nuclear Industry Capacity Factors 1975 – 2020, Percent



Frequency of outages

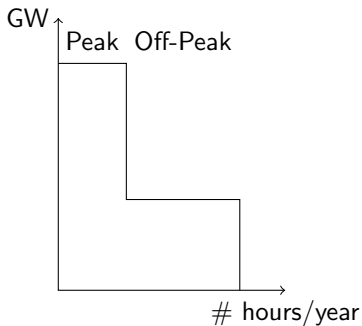


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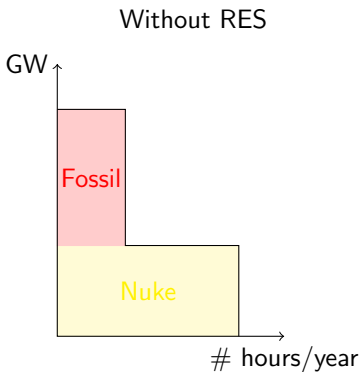
Example of 2-unit nuke facility (Golfech)



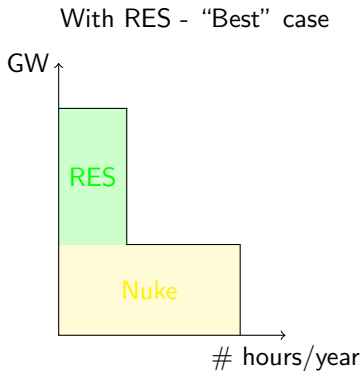
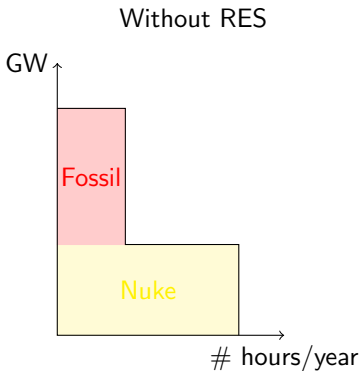
Nuclear and renewables as substitutes in the long-run



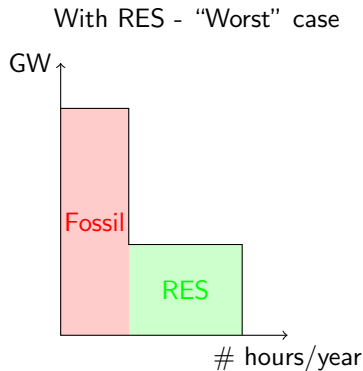
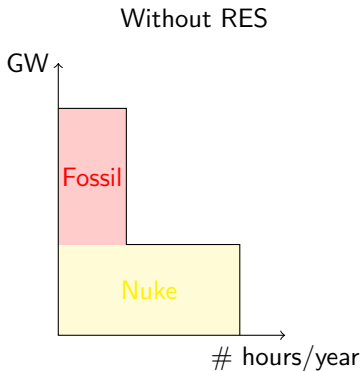
Nuclear and renewables as substitutes in the long-run



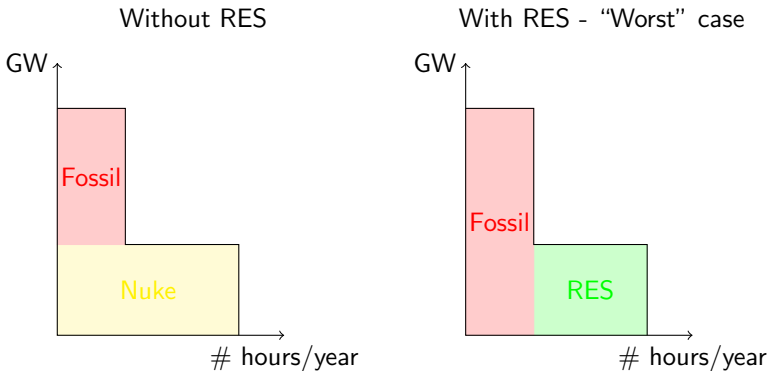
Nuclear and renewables as substitutes in the long-run



Nuclear and renewables as substitutes in the long-run



Nuclear and renewables as substitutes in the long-run



⇒ See for example Green and Léautier (2015).