

Shock Therapy for Greener Growth

The Dynamics of Firms' R&D Investments

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Introduction

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- The fossil energy price or profitability is affected by:
 - ▶ Industrial policies directed towards these sectors
 - ▶ Carbon prices:
 - ★ Push consumer prices up: Increase clean innovation from the demand side
 - ★ Push producer prices down: Increase clean innovation from the supply side

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 - ▶ Industrial policies directed towards these sectors
 - ▶ Carbon prices:
 - ★ Push consumer prices up: Increase clean innovation from the demand side
 - ★ Push producer prices down: Increase clean innovation from the supply side
- A shock may lead to reallocation of resources:
 - ▶ Between firms and industries
 - ▶ Within firms

Stylized fact 1: Inertia in R&D

Table: Annual transition rates

Status year t	Status year t+1	
	No R&D	R&D
No R&D	0.862	0.138
R&D	0.098	0.902

Stylized fact 2: Increase in clean R&D

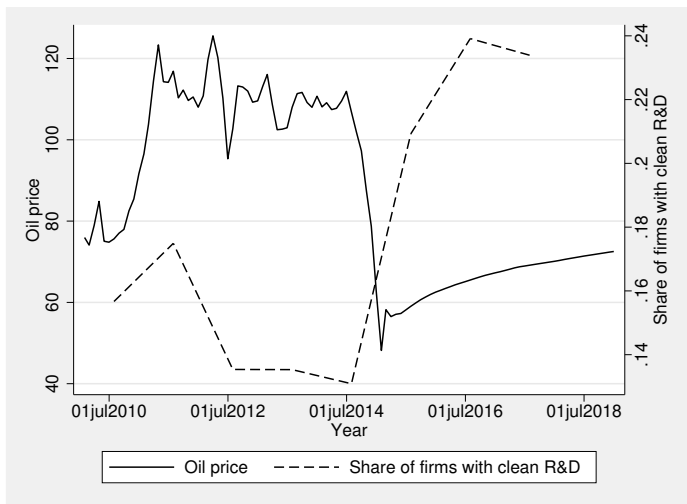
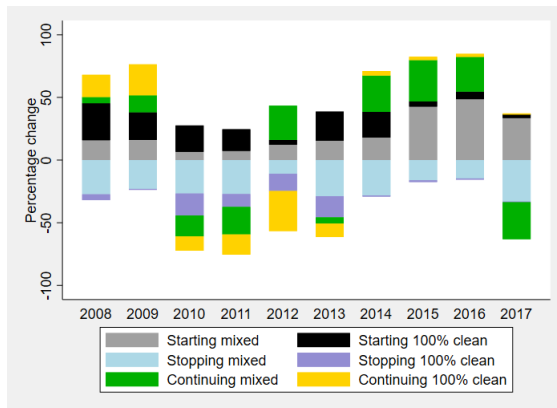


Figure: Oil prices and Clean Innovation

► Price evolution

Stylized fact 3: Increase in clean R&D driven by firms that do both



This paper

- Stylized theoretical model:
 - ▶ Directed technical change (clean and dirty) with heterogeneous firms
 - ▶ Explores how a persistent fall in price of oil may encourage clean innovation in the supply chain
 - ▶ Within-firm dynamics lead shock-exposed firms to react differently
- Empirical analysis
 - ▶ Uses rich firm-level data for Norway
 - ▶ Exploits that firms are differentially exposed to the 2014 oil price shock due to their supply linkages to the extractors of fossil energy
 - ▶ Findings indicate that shock-exposed firms react differently

Theoretical Framework

A stylized model of directed technical change

Production of two final (energy) goods: clean and dirty

- Exogenous final good prices (small, open economy)

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Production of a range of inputs for each final good

- Each variety produced by a monopolist
- One-period monopoly rights obtained by innovation for that variety

A stylized model of directed technical change

Production of two final (energy) goods: clean and dirty

- Exogenous final good prices (small, open economy)

Production of a range of inputs for each final good

- Each variety produced by a monopolist
- One-period monopoly rights obtained by innovation for that variety
- Input producers can hire scientists for both types of R&D
 - ▶ Higher price of a final good gives higher profitability in R&D for inputs of that type
- Firms differ in their innovation probability:
 - ▶ Some firms have no R&D
 - ▶ Some firms do only one type of R&D
 - ▶ Some firms do both types of R&D

Within-firm dynamics

- Spillovers from mature (dirty) to clean R&D activity
 - ▶ Imply positive relation between the two types of R&D within the firm

Within-firm dynamics

- Spillovers from mature (dirty) to clean R&D activity
 - ▶ Imply positive relation between the two types of R&D within the firm
- Adjustment costs when rescaling total R&D activity
 - ▶ Imply negative relation between the two types of R&D within the firm

A persistent oil price drop

A fall in p_{dt} leads to:

- Lower dirty production
- Lower profits in dirty input production and thus in dirty R&D
- Lower dirty R&D in all exposed firms.

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Exposed firms after the shock:

- Spillovers suggest lower clean innovation (relative to other firms)
- Adjustment costs suggest higher clean innovation (relative to other firms)

Empirical Analysis

Empirical Analysis

Question:

What is the impact of the 2014 oil price shock on clean innovation in exposed firms relative to other firms?

Data and Sample

- Accounting data for all joint-stock firms in the manufacturing sector in Norway
 - ▶ operating income, operating profits, employment
- Product-level (HS8) trade data for the universe of firms
 - ▶ exports, imports
- R&D survey
 - ▶ R&D expenditure, R&D personell, Share of green R&D in total R&D
 - ▶ Clean R&D: renewable energy, other environment-related energy
- Sample
 - ▶ All joint stock firms in the manufacturing sector (nace #10 to 35) that are covered by the R&D survey
 - ▶ Unbalanced panel of approximately 1,300 firms per year
 - ▶ Covers 2007-2017

The 2014 Oil Price Shock

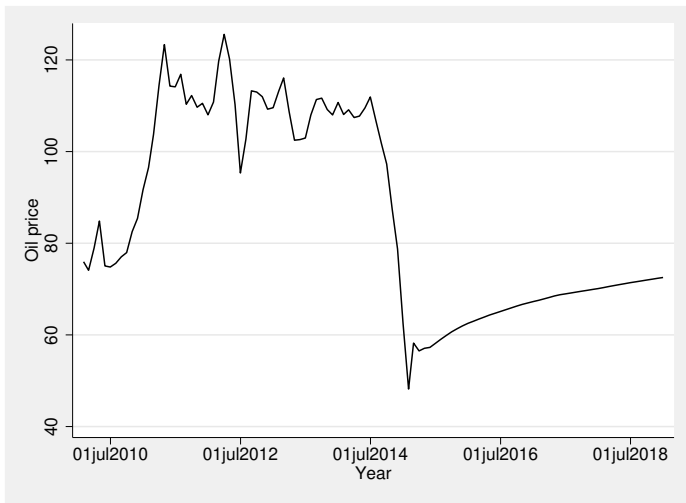


Figure: Oil price and Oil future prices (Source: Norges Bank)

The 2014 Oil Price Shock

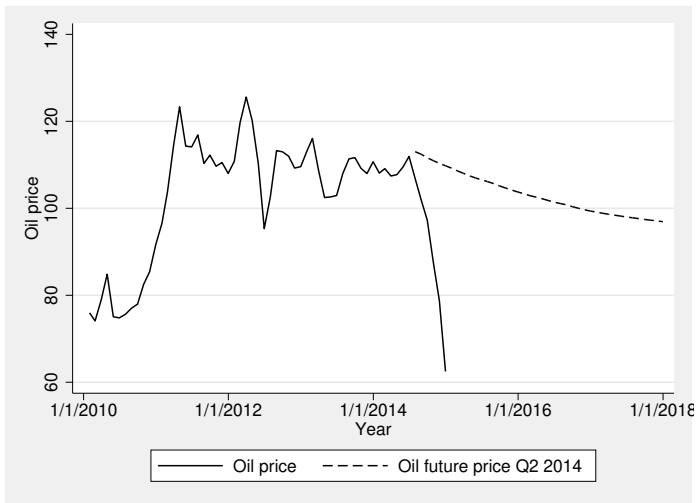


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Aggregate investments in clean R&D

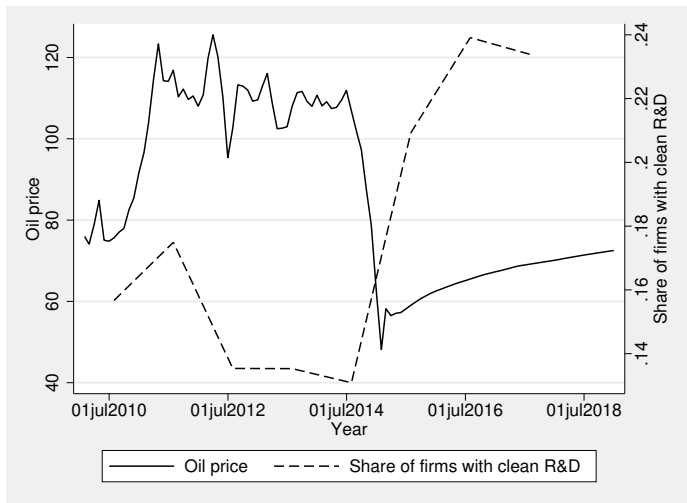


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Identification

- Challenge: identify which firms are most exposed to the shock
 - ▶ Standard approach in the literature: Input-output tables
 - ▶ Our approach: Firm-specific exposure measure based on trade data

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- Challenge: identify which firms are most exposed to the shock
 - ▶ Standard approach in the literature: Input-output tables
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- Firm-specific oil industry exposure measure:
 - ▶ use firm-level imports by oil producers to identify which products they use in their production:
 - ★ identify the HS8 products imported by the oil extraction industry (nace #6) in the pre-shock period (2007-2013)
 - ▶ use firm-level exports by suppliers to identify which firms sell these products:
 - ★ identify the firms in manufacturing that export products imported by the oil extraction industry ($j \in o$)
 - ▶ calculate firm-level exposure, $x_{oi} \in [0, 1]$, as the share of “oil products” in firms’ total export basket in 2013
 - ★ $x_{oi} = \sum_{j \in o} x_{ijt} / \sum_j x_{ijt}$ for $t = 2013$

Exposure measure: Pros and cons

Benefits:

- 1 Captures the fact that the oil price shock was global
- 2 Gives firm-level variation
- 3 Allows controlling for industry-level trends

Drawbacks:

- 1 Reduces sample to only manufacturing + exporting firms
- 2 Misses indirect exposure
- 3 Relies on assumption that share of exports is informative of share of production

Shock exposure

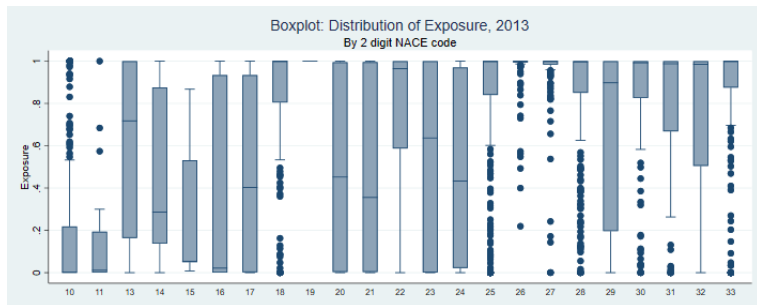


Figure: Within-industry Variation in Exposure

► Details

Empirical model

Diff-in-diff: Compare firms affected by the oil price shock through the supply chain to other firms pre/post 2014:

$$y_{it} = \alpha_i + \beta x_{oi} \times Post_t + \gamma Z_{it} + \delta_{st} + \varepsilon_{it},$$

- x_{oi} measures firm i 's exposure
- α_i firm FE, δ_{st} industry-year FE (NACE 2-digit)
- $Post_t = 1$ if $t > 2013$

- Z_{it} : firm level controls
 - ▶ log employment, log tangible assets, export share, energy share and a dummy for public funding
 - ▶ measured at baseline and interacted with year dummies

Outcome variables

- Measures of clean (renewable energy) R&D activity
 - ▶ Clean energy R&D dummy
 - ▶ Clean energy share of R&D expenditures
 - ▶ Clean R&D expenditures
 - ▶ (Switchers)
- (Measures of sales and profit:)
 - ▶ sales per employee
 - ▶ operating profits categorical variable (-1, 0, 1)
- (Measures of R&D activity:)
 - ▶ R&D dummy
 - ▶ log R&D employment

Results: Clean R&D

Table: Clean R&D

Variable:	Dummy (1)	Share (3)	Log Value (5)
$Post_t * X_{oi}$	0.055*** (0.020)	1.575* (0.899)	0.346** (0.140)
Controls	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes
Ind.*year FE	Yes	Yes	Yes
Obs.	11,695	11,695	11,695

▶ W/o controls

▶ PPML

Standard errors in parenthesis are clustered on firm. Log Value is measured as $\log(1 + \text{Green R\&D expenditures})$. Controls include baseline levels of log employment, log tangible assets, export share, energy share and a dummy for public funding, all interacted with year dummies. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Mechanisms & robustness

- R&D in general [▶ Link](#)
- Sales and profits [▶ Link](#)
- Pre-trends [▶ Link](#)
- Placebo using other technology types [▶ Link](#)
- Renewable prices [▶ Link](#)

Conclusion

- Carbon pricing can affect clean innovation not only through higher demand for clean alternatives, but also through higher supply
- We show theoretically that within-firm dynamics may lead directly exposed firms in the energy sector to react differently to a drop in the oil price, compared to less exposed firms
- We show empirically that exposed firms in the Norwegian oil supply sector increase clean innovation in response to the 2014 drop in the oil price

Conclusion

- Carbon pricing can affect clean innovation not only through higher demand for clean alternatives, but also through higher supply
- We show theoretically that within-firm dynamics may lead directly exposed firms in the energy sector to react differently to a drop in the oil price, compared to less exposed firms
- We show empirically that exposed firms in the Norwegian oil supply sector increase clean innovation in response to the 2014 drop in the oil price
- Carbon pricing (and other policy measures) may induce reallocation not only across, but also within, firms

Thank you!

Innovation and growth

- A_{jt} : Aggregate state of the technology of type j at time t :

$$A_{jt} \equiv \int_0^1 A_{jit} di.$$

- A successful innovation increases the quality of the input by: $(1 + \gamma) > 1$

▶ Back

How may an oil price drop affect clean innovation?

A persistent drop in p^{oil} will:

- reduce profitability in fossil energy production and its supply chain
- lead to lower fossil energy related R&D
- lead to free resources for R&D activity in the market
⇒ higher clean innovation in “all” firms.

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- lead to free resources for R&D activity in the market
⇒ higher clean innovation in “all” firms.
- In addition: Directly exposed firms (in supply chain):
 - ▶ may invest more in clean R&D if rescaling of R&D activity is costly.
 - ▶ may invest less in clean R&F if there are within-firm technological spillovers from fossil to clean innovation.

▶ Back

Shock exposure

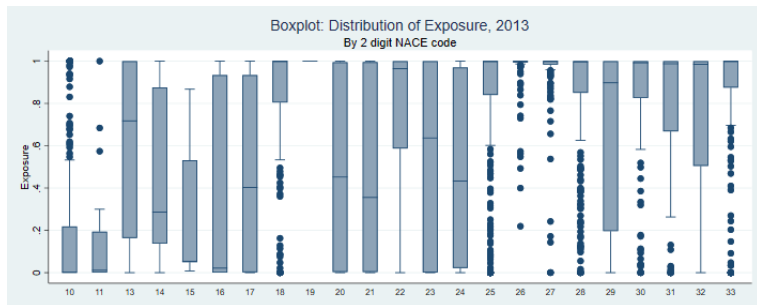


Figure: Within-industry Variation in Exposure

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► Back

Box plot

- Median value: the line splitting the box in two represents the median value
 - ▶ shows that 50 % of the data lies below the median value and 50 % lies above
- Lower Quartile: the bottom edge of the box represents the lower quartile
 - ▶ shows the value at which the first 25 % of the data falls up to
- Upper Quartile: the upper edge of the box shows the upper quartile
 - ▶ shows that 25 % of the data lies to the right of the upper quartile value
- Upper and lower values of the data: the horizontal lines stop at are the values of the upper and lower values of the data
- Outliers: the single points on the diagram

▶ Back

Switchers

- “New Green”
 - ▶ Dummy=1 if no Green energy R&D=0 in $t - 1$ & Green energy R&D>0 in t
 - ▶ otherwise Dummy=0
 - ▶
- “From R&D to Green”
 - ▶ Dummy=1 if R&D=1 in $t - 1$ & New Green=1 in t
 - ▶ Dummy=0 if R&D=1 $t - 1$ & & New Green=0 in t
 - ▶ Column 2: Dummy=0 if R&D=0 in $t - 1$
 - ▶ Column 3: Dummy=missing if R&D=0 in $t - 1$

▶ Back

Results: Clean R&D

Table: Clean R&D

Variable:	Dummy (1)	Dummy (2)	Dummy (3)	Dummy (4)
$Post_t * X_{oi}$	0.044** (0.019)	0.039** (0.020)	0.042** (0.020)	0.055*** (0.020)
Controls excl. energy	No	No	Yes	Yes
Controls incl. energy	No	No	No	Yes
Firm FE	No	Yes	Yes	Yes
Ind.*year FE	Yes	Yes	Yes	Yes
Obs.	11,695	11,695	11,695	11,695

▶ Back

Standard errors in parenthesis are clustered on firm. Controls include baseline levels of log employment, log tangible assets, export share, energy share and a dummy for public funding, all interacted with year dummies. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Results: R&D

Table: R&D

Variable:	R&D dummy (1)	log R&D emp (2)	R&D emp share (3)
$Post_t \times X_{oi}$	-0.015 (0.028)	-0.008 (0.007)	-0.001 (0.001)
Controls	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes
Industry*year FE	Yes	Yes	Yes
Observations	11,695	11,695	11,695

▶ Back

Standard errors in parenthesis are clustered on firm. Log R&D employment is $\log(1+\text{R\&D employees})$. Controls include baseline levels of log employment, log tangible assets, export share, energy share and a dummy for public funding, all interacted with year dummies. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Results: Sales and profits

Table: Sales per employee and Profits indicator

Variable:	Sales per emp. (1)	Sales per emp. (2)	Profits (3)	Profits (4)
$Post_t \times X_{oi}$	-0.082*** (0.026)	-0.043* (0.026)	-0.179*** (0.049)	-0.081 (0.060)
Controls	Yes	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes	Yes
Year FE	Yes	No	Yes	No
Industry FE	Yes	No	Yes	No
Industry*year FE	No	Yes	No	Yes
Observations	11,695	11,695	11,695	11,695

▶ Back

Standard errors in parenthesis are clustered on firm. Controls include baseline levels of log employment, log tangible assets, export share, energy share and a dummy for public funding, all interacted with year dummies. The indicator for operating profits takes on 0, -1 and 1 depending on whether the firms makes zero, negative or positive profits. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Results: Placebo using other tech fields

Table: Placebo

Variable:	Bio tech R&D (1)	ICT R&D (2)
$Post_t \times X_{oi}$	-0.015 (0.015)	-0.032 (0.024)
Controls	Yes	Yes
Firm FE	Yes	Yes
Industry*year FE	Yes	Yes
Observations	11,695	11,695

▶ Back

Standard errors in parenthesis are clustered on firm. Controls include baseline levels of log employment, log tangible assets, export share, energy share and a dummy for public funding, all interacted with year dummies. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Falling Renewable Prices

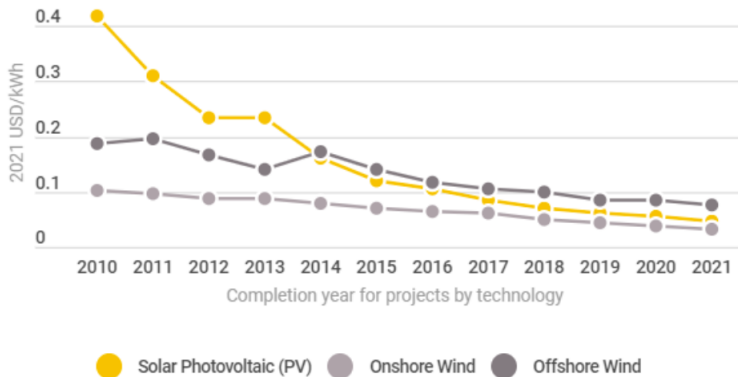


Figure: Global renewable prices (IRENA)

▶ Back

The Oil Price Shock 2014

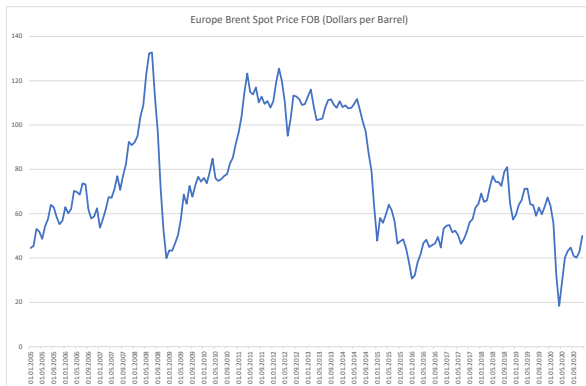


Figure: Oil price: Brent Blend (Source: U.S. Energy Information administration)

▶ Back

Results: Clean R&D PPML

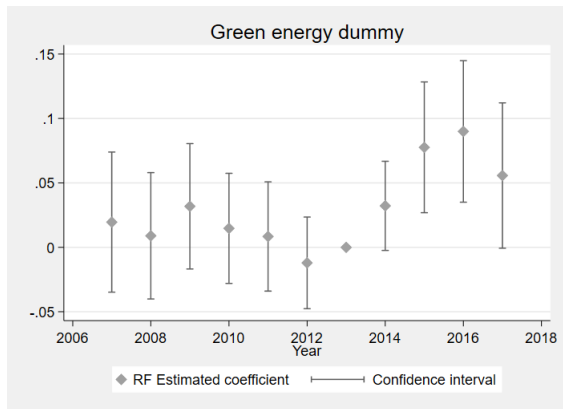
Table: Clean R&D PPML

Variable:	Dummy (1)	Share (3)	Value (5)
$Post_t * X_{oi}$	0.499** (0.199)	0.358* (0.218)	0.716 (0.470)
Controls	Yes	Yes	Yes
Firm FE	Yes	Yes	Yes
Ind.*year FE	Yes	Yes	Yes
Obs.	3,024	3,024	3,024

▶ Back

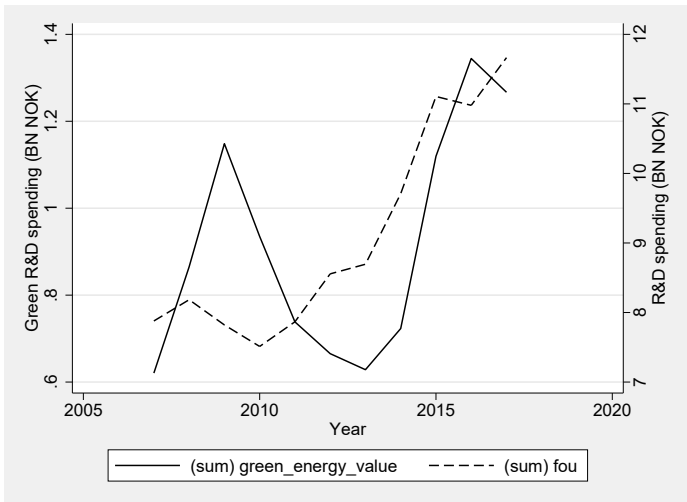
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Dynamic DID

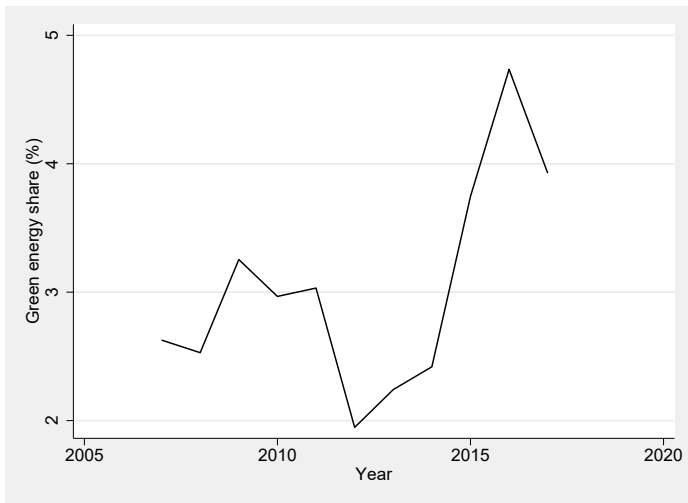


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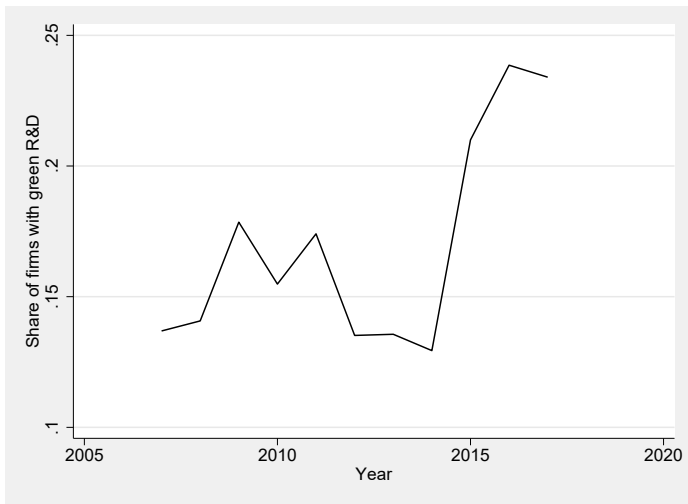
R&D expenditure



Green energy share in R&D



Share of firms with green R&D



Two propositions

Proposition

W/o spillovers or adjustment costs:

$$\frac{ds'_{kct}}{dp_{dt}} = \frac{ds'''_{kct}}{dp_{dt}}$$

→ All firms engaged in clean R&D, independently of whether the firms are also active in dirty R&D or not, will respond equally

Proposition

With spillovers and adjustment costs:

$$\frac{ds'''_{kct}}{dp_{dt}} < \frac{ds'_{kct}}{dp_{dt}}$$

→ Exposed firms will respond more or less depending on whether the spillover effect dominates the effect of the adjustment costs or vice versa

