

# Air Pollution and Innovation

Does air pollution hamper innovation-led growth ?

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# Air pollution

- ▶ Fine particulate matter –  $PM_{2.5}$
- ▶ For 80% caused by human activity
- ▶ Severe health effects – 6.5 mln. deaths annually (WHO)

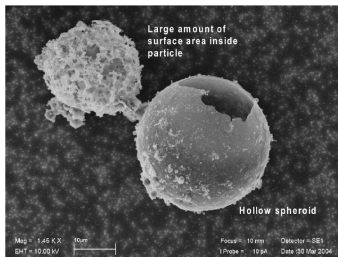


FIGURE 3.3 Scanning electron micrograph of coarse particles emitted from an oil-fired power plant. Diameters of the particles are greater than  $20\ \mu\text{m}$  optical diameter. Both particles are hollow, so their aerodynamic diameter is significantly smaller than if they were solid. Source characterization study by Stevens R, Lynam M, Proffitt D, 2004. Photo courtesy of Willis R, U.S. Environmental Protection Agency.

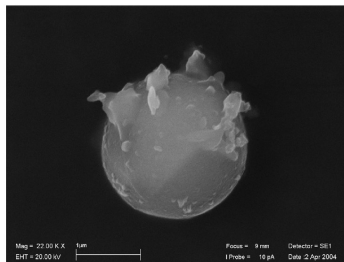


FIGURE 3.4 Scanning electron micrograph of spherical aluminosilicate fly ash particle emitted from an oil-fired power plant. Diameter of the particle is approximately  $2.5\ \mu\text{m}$ . Photo courtesy of Willis R, U.S. Environmental Protection Agency.

Source: Vallero (2014)

# Economic costs of air pollution

- ▶ Cost estimates important for regulation
- ▶ Common assessments:
  - ▶ Premature mortality: **0.092%** of GDP (Landrigan et al. 2018)
  - ▶ Accounting for absenteeism, health expenditures and agricultural yields: **0.3%** of GDP (OECD 2016)
- ▶ 'Hidden' productivity effects:
  - ▶ Labor supply (Hanna & Oliva 2015)
  - ▶ Productivity conditional on hours worked (Zivin & Neidell 2012, Chang et al. 2016, 2019)
- ▶ Aggregate economic output (Dechezleprêtre et al. 2019):
  - ▶ 10% less air pollution  $\Rightarrow$  **0.8%** higher GDP

# Overview

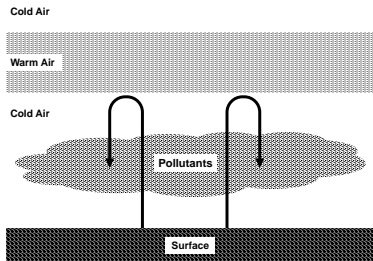
## What is the effect of air pollution on inventive output?

- ▶ Study design
  - ▶ Panel of 1288 NUTS-3 regions in Europe 2001-2013
  - ▶ Patents as measure for inventive output
  - ▶ Exposure to  $PM_{2.5}$  as measure of air pollution
  - ▶ Weather as source exogenous variation
- ▶ Findings
  - ▶  $0.17\mu\text{g}/\text{m}^3$  more  $PM_{2.5} \Rightarrow 1.7\%$  less patents
  - ▶ Back-of-the-envelope: 10% of effect on GDP
  - ▶ Heterogeneity regions
- ▶ Interpretation
  - ▶ **Not** inventor mobility (zero-sum)
  - ▶ **Not** mortality or R&D expenditures (in prior assessments)

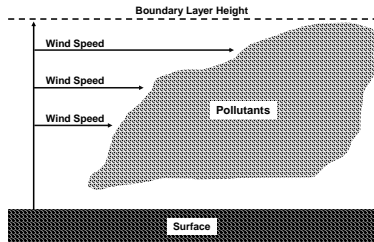
# Empirical strategy

- ▶ Measuring inventive output
  - ▶ Next year's number of patent filings in region
  - ▶ Lag between invention and patent application
- ▶ Exposure to air pollution
  - ▶  $PM_{2.5}$  concentration at high resolution
  - ▶ Weight for population density
- ▶ Exogenous variation in air pollution
  - ▶ Thermal inversions
  - ▶ Ventilation capability

# Instrumental variables



(a) Thermal Inversions



(b) Ventilation Coefficient

- ▶ **Relevance:** Positive correlation documented in atmospheric studies (Xu et al. 2017, Hou et al. 2018)
- ▶ **Exogeneity:** Determined by large-scale atmospheric dynamics
- ▶ **Exclusion Restriction:** Include a large set of weather controls

# IV regression

## First Stage

$$P_{it} = \alpha_1 TI_{it} + \alpha_2 VC_{it} + \alpha_3 Pop_{it} + \alpha_4 X_{it} + \omega_{ct} + \eta_i + \nu_{it}$$

- ▶  $P_{it}$ :  $PM_{2.5}$  concentration for NUTS-3  $i$  in year  $t$
- ▶  $TI_{it}$ : Share of days per year with thermal inversions
- ▶  $VC_{it}$ : Avg. number of consecutive days with low ventilation coefficient

## Second Stage

$$Y_{it} = \beta_1 \widehat{P}_{it} + \alpha_2 Pop_{it} + \beta_3 X_{it} + \lambda_{ct} + \alpha_i + \varepsilon_{it}$$

- ▶  $Y_{it}$ : Number of patent filings for NUTS-3  $i$  in year  $t + 1$

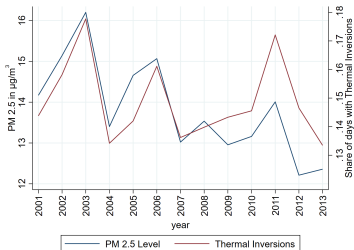
## Data

- ▶ Geocoded inventor addresses (de Rassenfosse et al. 2019) and PATSTAT
- ▶  $PM_{2.5}$  concentration (Van Donkelaar et al. 2019) and population data in Europe (Columbia University 2018)
- ▶ Atmospheric data from MERRA-2 (Gelaro et al. 2017)
- ▶ Atmospheric and climate data from ERA5 and ERA5-Land (Hersbach et al. 2020, Muñoz-Sabater 2019)
- ▶ Geographic indicators for NUTS-3 regions from Eurostat

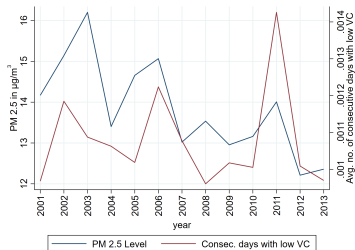


# Descriptive patterns

- $PM_{2.5}$  concentration falls on average by  $0.17\mu\text{g}/\text{m}^3$  every year




(c)  $PM_{2.5}$  and Thermal Inversions



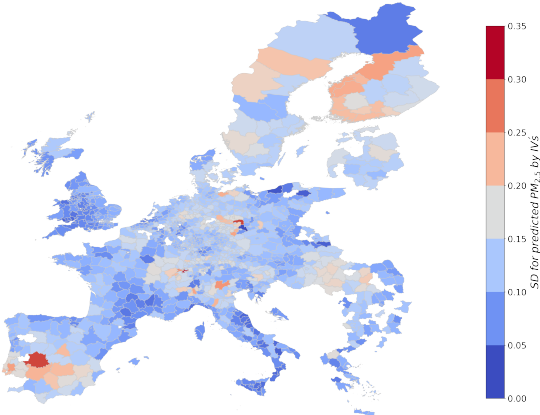
(d)  $PM_{2.5}$  and Ventilation Capability

## Summary statistics

- ▶ 1288 NUTS-regions in 30 countries for the period 2001 - 2013
- ▶ Trim  $PM_{2.5}$  at 1% at each side of the distribution
- ▶ Drop NUTS-3 regions equal or below 5th percentile of the distribution for total number of patent filings 

	N	Mean	Std. Dev.	10th	50th	90th
Nb. patent filings	16744	108.56	222.63	3	41	257
Population ( <i>in thd</i> )	14845	385.59	437.93	94.75	270.11	736.28
Pop-weighted PM 2.5	16407	13.62	4.79	8.31	12.88	20.38
Nb. thermal inversions	16484	54.38	32.58	13	51	100
Share thermal inversions	16484	0.15	0.09	0.04	0.14	0.27
Ventilation Coefficient ( $m^2/s$ )	16328	1642.95	784.28	599	1613	2667
Avg consecutive days with low VC	16328	0.00	0.0003	0.0008	0.0010	0.0015
Surface pressure ( $Pa$ )	16328	97969.59	3468.87	93819	98969	101094
Relative humidity (%)	16484	0.74	0.06	0.64	0.75	0.81
Temperature ( $K$ )	16328	283.03	2.67	280	283	286
Wind speed ( $m/s$ )	16328	2.49	0.93	1.24	2.47	3.62
Precipitation ( $m$ )	16328	0.13	0.04	0.09	0.12	0.18
<i>N</i>	16744					

# PM variation induced by the IVs



Predicted  $PM_{2.5}$  concentration by NUTS-3

## Main result

- ▶  $1\mu\text{g}/\text{m}^3$  increase in  $PM_{2.5} \Rightarrow 10\%$  less patent filings
- ▶ Europe's average yearly reduction in air pollution has led to 1.7% more innovations per year

Dep. Variable:	OLS	IV with TI		IV with VC		IV with TI & VC	
	Log(Patents)	Log(Patents)	$PM_{2.5_{t-1}}$	Log(Patents)	$PM_{2.5_{t-1}}$	Log(Patents)	$PM_{2.5_{t-1}}$
$PM_{2.5_{t-1}}$	-0.0057 (0.0037)	-0.10** (0.048)		-0.10** (0.050)		-0.10*** (0.036)	
<i>Share Inversions</i> $_{t-1}$			3.70*** (0.52)				3.65*** (0.52)
<i>Share low VC</i> $_{t-1}$					261.3*** (39.4)		256.8*** (39.6)
NUTS-3 FE	YES	YES		YES		YES	
Country-Year FE	YES	YES		YES		YES	
Controls	YES	YES		YES		YES	
KP F-stat.		50.4		43.9		44.0	
Hansen J stat. P-Val						0.96	
N	12864	12864	12864	12864	12864	12864	12864

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

# Additional results

## Regional variation

- ▶ Effect larger in regions with above-median pollution [▶ Results](#)
- ▶ Similar effect for regions with medium vs. high levels of urbanization [▶ Results](#)

## Inventor Migration

- ▶ Geocoded data on disambiguated inventors (Morrison et al. 2017)
- ▶ Track mobility by changing addresses across patents (Zacchia 2018)
- ▶ Economically and statistically insignificant effect on inventor mobility [▶ Results](#)

# Mechanisms I

## Are these effects captured in prior assessments?

- ▶ Mortality
- ▶ R&D spending

### Mortality ▶ Results

- ▶ Total mortality statistics per NUTS-3 region (Eurostat)
- ▶ Conservative estimate because inventors probably *less* likely to pass away (socioeconomic factors)
- ▶ Economically and statistically insignificant effect on total mortality

# Mechanisms II

## R&D spending $\Rightarrow$ Science ▶▶ Results

- ▶ Science funding should react less to air pollution
- ▶ Geocoded data on PubMed author affiliations (Torvik 2015)
- ▶  $1\mu\text{g}/\text{m}^3$  increase in  $PM_{2.5} \Rightarrow 9\%$  less journal articles

## R&D spending $\Rightarrow$ Invention quality ▶▶ Results

- ▶ Less R&D expenditures  $\Rightarrow$  Less low quality inventions  $\Rightarrow$  Higher observed quality
- ▶ Weight patent counts by forward citation counts
- ▶ Quality does not increase significantly

## Robustness checks

- ▶ Poisson fixed effects regression [Results](#)
- ▶ Accounting for spatial and serial correlation [Results](#)
- ▶ Altering threshold for the definition of a low Ventilation Coefficient [Results](#)
- ▶ Including all NUTS-3 regions for main results [Results](#)
- ▶ Altering the sample period up to the interval 2001 - 2008 [Results](#)
- ▶ Excluding weather covariates [Results](#)
- ▶ None of the weather controls are significantly correlated with # of patent filings [Results](#)



## Conclusion

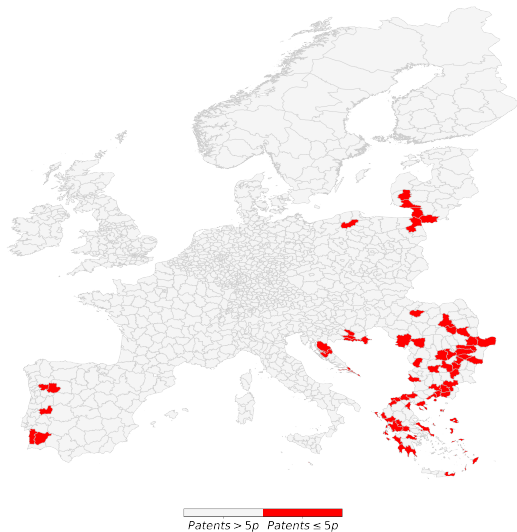
- ▶ Negative effect of air pollution on innovation
  - ▶ Driven by more polluted regions
  - ▶ In regions with medium to high levels of urbanization
  - ▶ No evidence that migration drives the effect
- ▶ Cost not accounted for in previous assessments
  - ▶ Mortality and R&D spending unlikely to drive effect
- ▶ Back-of-the-envelope:
  - ▶ Use estimates of private and social value of innovation (Gambardella et al. 2008, Bloom et al. 2013, Zacchia 2018)
  - ▶ 21.3 bln. euros annually
  - ▶ Previous assessment using GDP understated with at least 10%

**Evidence for another hidden cost of air pollution**

**Thank you!**

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## Dropped NUTS-3 regions



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## High- vs. low-pollution regions

Dep. Variable:	Low PM 2.5 below Median		High PM 2.5 above Median	
	Log(Patents)	PM2.5	Log(Patents)	PM2.5
PM2.5	-0.049 (0.049)		-0.15** (0.067)	
<i>Share Inversions</i>		3.17*** (0.52)		3.89*** (0.94)
<i>Share low VC</i>		225.7*** (44.9)		210.0*** (63.4)
Mean # Patents	89.19		130.81	
Mean PM 2.5	10.49		17.05	
NUTS-3 FE	YES		YES	
Country-Year FE	YES		YES	
KP F-stat.	32.1		14.0	
Hansen J stat. P-Val	0.59		0.94	
N	6605	6605	6307	6307

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

- ▶ Non-linear relationship between air pollution and innovation

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## Urban vs. rural regions

Dep. Variable:	Predominantly Urban		Intermediate		Predominantly Rural	
	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5
<i>PM2.5</i>	-0.092** (0.043)		-0.17*** (0.062)		-0.0033 (0.094)	
<i>Share Inversions</i>		5.43*** (1.08)		2.66*** (0.80)		3.91*** (0.76)
<i>Share low VC</i>		303.7*** (83.3)		238.2*** (57.4)		148.7** (67.9)
Mean # Patents	208.26		96.59		40.84	
Mean PM 2.5	14.14		14.15		13.06	
NUTS-3 FE	YES		YES		YES	
Year FE	YES		YES		YES	
KP F-stat.	18.9		13.7		16.2	
Hansen J stat. P-Val	0.60		0.56		0.25	
N	3370	3370	5746	5746	3704	3704

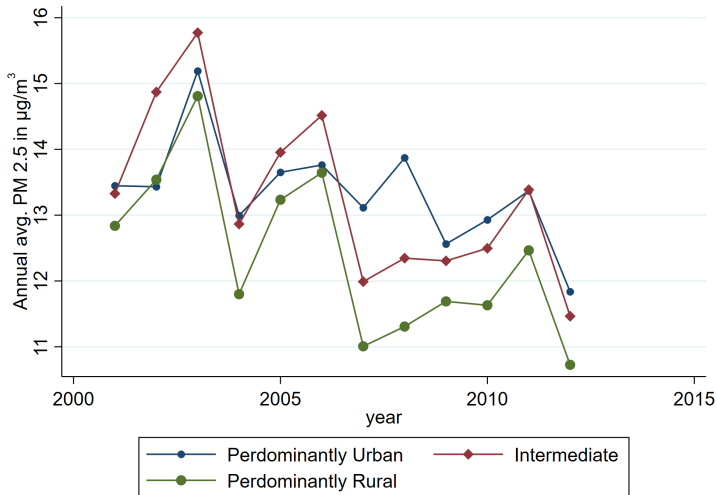
SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

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▶▶ PM levels

## PM level in urban vs. rural regions



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# Inventors' relocation and air pollution

Dep. Variable:	OLS		IV		OLS		IV	
	Log(Emigr. Inv.)	Log(Emigr. Inv.)	$PM2.5_{t-1}$	Log(Emigr. Inv. Patents)	Log(Emigr. Inv. Patents)	$PM2.5_{t-1}$		
$PM2.5_{t-1}$	-0.011*** (0.0039)	0.031 (0.046)		-0.017** (0.0069)	0.027 (0.084)			
<i>Share Inventions</i> <sub>t-1</sub>			3.78*** (0.50)				3.78*** (0.50)	
<i>Share low VC</i> <sub>t-1</sub>			227.2*** (40.3)				227.2*** (40.3)	
NUTS-3 FE	YES	YES		YES	YES			
Country-Year FE	YES	YES		YES	YES			
Controls	YES	YES		YES	YES			
KP F-stat.		42.5			42.5			
Hansen J stat. P-Val		0.25			0.27			
N	13671	13671	13671	13671	13671	13671	13671	13671

Dep. Variable:	OLS		IV		OLS		IV	
	Log(Immigr. Inv.)	Log(Immigr. Inv.)	$PM2.5_{t-1}$	Log(Immigr. Inv. Patents)	Log(Immigr. Inv. Patents)	$PM2.5_{t-1}$		
$PM2.5_{t-1}$	-0.0059 (0.0039)	-0.010 (0.043)		-0.0055 (0.0068)	-0.056 (0.079)			
<i>Share Inventions</i> <sub>t-1</sub>			3.78*** (0.50)				3.78*** (0.50)	
<i>Share low VC</i> <sub>t-1</sub>			227.2*** (40.3)				227.2*** (40.3)	
NUTS-3 FE	YES	YES		YES	YES			
Country-Year FE	YES	YES		YES	YES			
Controls	YES	YES		YES	YES			
KP F-stat.		42.5			42.5			
Hansen J stat. P-Val		0.56			0.85			
N	13671	13671	13671	13671	13671	13671	13671	13671

SE clustered at NUTS-3

\* p<0.1, \*\* p<0.05, and \*\*\* p<0.01

## Science

Dep. Variable:	OLS	IV	
	Log(PubMed)	Log(PubMed)	PM2.5
<i>PM2.5</i>	-0.0023 (0.0041)	-0.086** (0.042)	
<i>Share Inversions</i>			3.78*** (0.50)
<i>Share low VC</i>			227.2*** (40.3)
NUTS-3 FE	YES	YES	
Country-Year FE	YES	YES	
Controls	YES	YES	
KP F-stat.		42.5	
Hansen J stat. P-Val		0.26	
N	13671	13671	13671

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

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# Invention quality

Dep. Variable:	5-year fw. cites		Fw. cites normalized	
	Log(Cites)	$PM2.5_{t-1}$	Log(Cites)	$PM2.5_{t-1}$
$PM2.5_{t-1}$	-0.12*		-0.13***	
	(0.062)		(0.042)	
<i>Share Inversions</i> $_{t-1}$		3.65***		3.65***
		(0.52)		(0.52)
<i>Share low VC</i> $_{t-1}$		256.8***		256.8***
		(39.6)		(39.6)
NUTS-3 FE	YES		YES	
Country-Year FE	YES		YES	
Controls	YES		YES	
KP F-stat.	44.0		44.0	
Hansen J stat. P-Val	0.73		0.89	
N	12864	12864	12864	12864

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

## Total deaths and air pollution

Dep. Variable:	OLS	IV		OLS	IV	
	Log(Deaths)	Log(Deaths)	PM2.5	Death Rate	Death Rate	PM2.5
<i>PM2.5</i>	-0.0011*** (0.00036)	0.0027 (0.0038)		-0.0012*** (0.00038)	0.0028 (0.0040)	
<i>Share Inversions</i>			3.43*** (0.50)			3.43*** (0.50)
<i>Share low VC</i>			241.0*** (40.7)			241.0*** (40.7)
NUTS-3 FE	YES	YES		YES	YES	
Country-Year FE	YES	YES		YES	YES	
Controls	YES	YES		YES	YES	
KP F-stat.		38.6			38.6	
Hansen J stat. P-Val		0.24			0.30	
N	13230	13230	13230	13230	13230	13230

Death rate does not include population as control. Standard errors are reported in parentheses and clustered at the NUTS-3 level.

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

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## Count model

- ▶ Poisson fixed effects regression with control function approach to account for endogenous  $PM_{2.5}$  (Lin & Wooldridge 2019)

Dep. Variable:	Poisson	Poisson CF
	Nb. Patents	Nb. Patents
$PM_{2.5}$	-0.0077** (0.0036)	-0.071*** (0.027)
$\hat{\nu}$		0.064** (0.027)
NUTS-3 FE	YES	YES
Country-Year FE	YES	YES
Log likelihood	-56582.7	
Pseudo R2	0.96	
Replications		500
Clusters	1196	1365
N	13599	16380

SE are bootstrapped

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

## Conley SE

- ▶ Adjust SE in line with Conley (1999, 2010) to account for spatial and serial correlation
- ▶ Assumption 1 (2) assumes a spatial correlation range of 500 (1000) kilometers and a serial correlation range of 5 (10) years

	<i>Clustered SE</i>	<i>SE – Assumption 1</i>	<i>SE – Assumption 2</i>
Dep. Variable:	Log(Patents)	Log(Patents)	Log(Patents)
<i>PM2.5</i>	-0.087** (0.036)	-0.087** (0.044)	-0.087** (0.042)
NUTS-3 FE	YES	YES	YES
Country-Year FE	YES	YES	YES
Controls	YES	YES	YES
N	13671	13671	13671

SE are bootstrapped

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

## Different thresholds for a low VC

Dep. Variable:	<i>Low VC</i> ≤ <i>p</i> 10		<i>Low VC</i> ≤ <i>p</i> 15		<i>Low VC</i> ≤ <i>p</i> 20		<i>Low VC</i> ≤ <i>p</i> 25	
	Log(Patents)	<i>PM</i> 2.5	Log(Patents)	<i>PM</i> 2.5	Log(Patents)	<i>PM</i> 2.5	Log(Patents)	<i>PM</i> 2.5
<i>PM</i> 2.5	-0.10** (0.047)		-0.088** (0.043)		-0.10*** (0.036)		-0.054* (0.028)	
<i>Share Inversions</i>		3.70*** (0.52)		3.67*** (0.51)		3.66*** (0.51)		3.70*** (0.51)
<i>Share p</i> 10 ≥ <i>VC</i>		90.6 (82.0)						
<i>Share p</i> 15 ≥ <i>VC</i>				181.1*** (53.4)				
<i>Share p</i> 20 ≥ <i>VC</i>						255.6*** (39.5)		
<i>Share p</i> 25 ≥ <i>VC</i>								279.7*** (31.5)
NUTS-3 FE	YES		YES		YES		YES	
Country-Year FE	YES		YES		YES		YES	
Controls	YES		YES		YES		YES	
KP F-stat.	26.5		31.5		44.9		63.3	
Hansen J stat. P-Val	0.15		0.81		0.78		0.27	
N	12924	12924	12924	12924	12924	12924	12924	12924

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

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# Different thresholds for dropping NUTS-3 regions

Dep. Variable:	No trimming		Trim NUTS-3 at 5%		Trim NUTS-3 at 10%		Trim NUTS-3 at 15%	
	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5
<i>PM2.5</i>	-0.087** (0.036)		-0.099*** (0.036)		-0.14*** (0.037)		-0.14*** (0.036)	
<i>Share Inversions</i>		3.78*** (0.50)		3.65*** (0.51)		3.27*** (0.50)		3.24*** (0.52)
<i>Share low VC</i>		227.2*** (40.3)		253.7*** (39.5)		258.6*** (40.6)		259.5*** (41.8)
NUTS-3 FE	YES		YES		YES		YES	
Country-Year FE	YES		YES		YES		YES	
Controls	YES		YES		YES		YES	
KP F-stat.	42.5		44.9		39.0		37.1	
Hansen J stat. P-Val	0.50		0.82		0.70		0.96	
N	13671	13671	12984	12984	12267	12267	11699	11699

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

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## Different sample periods

Dep. Variable:	Log(Patents)	Log(Patents)	Log(Patents)	Log(Patents)	Log(Patents)
<i>PM2.5</i>	-0.10*** (0.036)	-0.097*** (0.034)	-0.16*** (0.055)	-0.16*** (0.047)	-0.14*** (0.046)
Period	2001-2012	2001-2011	2001-2010	2001-2009	2001-2008
NUTS-3 FE	YES	YES	YES	YES	YES
Country-Year FE	YES	YES	YES	YES	YES
KP F-stat.	44.9	46.5	18.3	23.0	24.9
Hansen J stat. P-Val	0.78	0.65	0.21	0.79	0.51
N	12924	11778	10659	9565	8475

SE clustered at NUTS-3

\*  $p < 0.1$ , \*\*  $p < 0.05$ , and \*\*\*  $p < 0.01$

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# Main results with individual weather controls

Dep. Variable:	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5	Log(Patents)	PM2.5
PM2.5	-0.075** (0.030)		-0.084** (0.035)		-0.079** (0.031)		-0.084** (0.036)		-0.076** (0.031)		-0.086*** (0.030)		-0.10*** (0.036)	
Share Inventions		4.05*** (0.49)		4.03*** (0.49)		4.16*** (0.51)		3.56*** (0.50)		3.97*** (0.49)		4.06*** (0.49)		3.66*** (0.51)
Share low VC		300.6*** (39.2)		224.0*** (38.1)		295.4*** (39.1)		256.0*** (38.8)		301.4*** (39.2)		326.5*** (41.3)		255.6*** (39.5)
Log(Pop)	0.32 (0.26)	0.74 (0.79)	0.34 (0.26)	0.83 (0.79)	0.32 (0.26)	0.75 (0.79)	0.33 (0.26)	0.75 (0.78)	0.33 (0.26)	0.77 (0.79)	0.33 (0.26)	0.72 (0.79)	0.36 (0.26)	0.88 (0.78)
Surface Pressure			0.0014 (0.0011)	0.020*** (0.0033)									0.0022 (0.0014)	0.024*** (0.0041)
Surface Pressure <sup>2</sup>			-5.0e-09 (4.9e-09)	-0.00000007*** (0.000000017)									-8.9e-09 (6.8e-09)	-0.000000092*** (0.000000021)
Relative humidity					1.53 (3.51)	22.4** (9.19)							-0.57 (3.49)	2.49 (9.57)
Relative humidity <sup>2</sup>					-0.89 (2.48)	-17.8*** (6.58)							0.46 (2.46)	-5.30 (6.78)
Temperature							0.56 (1.06)	5.30 (3.60)					-0.39 (1.37)	-5.06 (4.50)
Temperature <sup>2</sup>							-0.0010 (0.0019)	-0.010 (0.0064)					0.00063 (0.0024)	0.0081 (0.0080)
Precipitation									1.27 (69.0)	-236.2 (201.4)			-0.86 (75.5)	-159.6 (217.9)
Precipitation <sup>2</sup>									13426.1 (19623.1)	165018.1*** (50695.6)			20722.2 (21520.4)	208616.1*** (51402.7)
Wind speed											-0.0095 (0.13)	0.15 (0.38)	0.035 (0.13)	0.71* (0.43)
Wind speed <sup>2</sup>											-0.014 (0.019)	0.035 (0.053)	-0.012 (0.020)	0.070 (0.065)
NUTS-3 FE	YES		YES		YES		YES		YES		YES		YES	
Country-Year FE	YES		YES		YES		YES		YES		YES		YES	
KP F-stat.	60.9		49.5		57.8		45.3		59.9		62.5		44.9	
Hansen J stat. P-Val	0.84		0.91		0.77		0.88		0.81		0.86		0.78	
N	12924	12924	12924	12924	12924	12924	12924	12924	12924	12924	12924	12924	12924	12924

SE clustered at NUTS-3

\* p<0.1, \*\* p<0.05, and \*\*\* p<0.01

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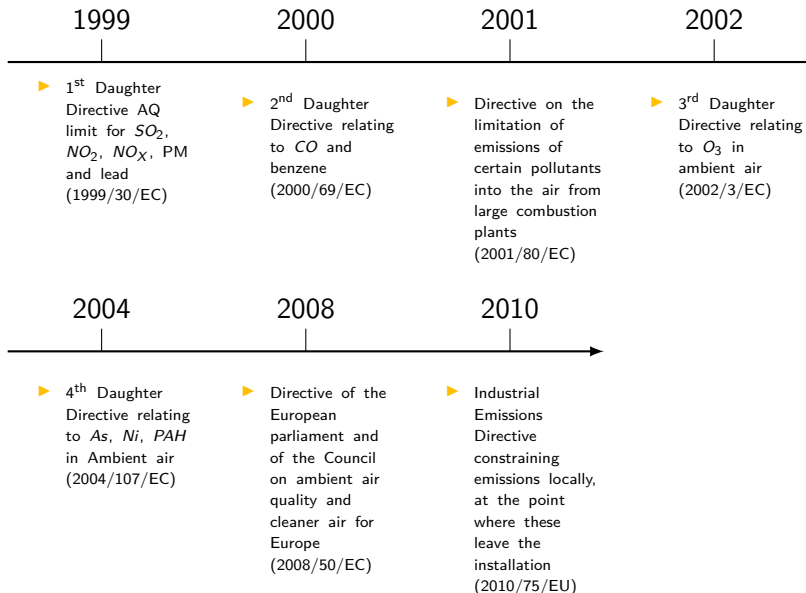
# Patent filings and weather controls

Dep. Variable:	Log(Patents)	Log(Patents)	Log(Patents)	Log(Patents)	Log(Patents)	Log(Patents)
<i>Log(Pop)</i>	0.15 (0.27)	0.29 (0.25)	0.16 (0.27)	0.16 (0.27)	0.16 (0.27)	0.25 (0.26)
<i>Surface Pressure</i>	-0.00020 (0.00080)					-0.000044 (0.00100)
<i>Surface Pressure</i> <sup>2</sup>	4.0e-10 (4.0e-09)					-5.9e-10 (5.2e-09)
<i>Relative humidity</i>		0.27 (3.20)				0.031 (3.31)
<i>Relative humidity</i> <sup>2</sup>		0.027 (2.26)				0.23 (2.34)
<i>Temperature</i>			0.018 (0.91)			-0.19 (1.13)
<i>Temperature</i> <sup>2</sup>			0.0000082 (0.0016)			0.00040 (0.0020)
<i>Precipitation</i>				35.3 (63.8)		50.1 (70.2)
<i>Precipitation</i> <sup>2</sup>				-2336.3 (17437.3)		-6023.3 (18345.3)
<i>Wind speed</i>					-0.0022 (0.10)	-0.035 (0.11)
<i>Wind speed</i> <sup>2</sup>					-0.014 (0.014)	-0.015 (0.014)
NUTS-3 FE	YES	YES	YES	YES	YES	YES
Country-Year FE	YES	YES	YES	YES	YES	YES
F-stat.	0.42	0.68	0.64	0.49	1.45	1.15
R-sq	0.96	0.96	0.96	0.96	0.96	0.96
N	13270	13399	13270	13270	13270	13152

SE clustered at NUTS-3

\* p<0.1, \*\* p<0.05, and \*\*\* p<0.01

# Air pollution policy in Europe



## References i

- Bloom, N., Schankerman, M. & Reenen, J. V. (2013), 'Identifying Technology Spillovers and Product Market Rivalry', *Econometrica* **81**(4), 1347–1393.
- Chang, T. Y., Zivin, J. G., Gross, T. & Neidell, M. (2019), 'The effect of pollution on worker productivity: Evidence from call center workers in China', *American Economic Journal: Applied Economics* **11**(1), 151–172.
- Chang, T., Zivin, J. G., Gross, T. & Neidell, M. (2016), 'Particulate pollution and the productivity of pear packers', *American Economic Journal: Economic Policy* **8**(3), 141–169.
- Columbia University, C. f. I. E. S. I. N. C. (2018), 'Gridded population of the world, version 4 (gpwv4): Population count, revision 11'.  
**URL:** <https://doi.org/10.7927/H4JW8BX5>
- Conley, T. G. (1999), 'Gmm estimation with cross sectional dependence', *Journal of econometrics* **92**(1), 1–45.

## References ii

- Conley, T. G. (2010), Spatial econometrics, in 'Microeconometrics', Springer, pp. 303–313.
- de Rassenfosse, G., Kozak, J. & Seliger, F. (2019), 'Geocoding of worldwide patent data', *Scientific Data* **6**(1), 260.  
**URL:** <https://doi.org/10.1038/s41597-019-0264-6>
- Dechezleprêtre, A., Rivers, N. & Stadler, B. (2019), 'The economic cost of air pollution: Evidence from europe'.
- Gambardella, A., Harhoff, D. & Verspagen, B. (2008), 'The value of European patents', *European Management Review* **5**(2), 69–84.
- Gelaro, R., McCarty, W., Suárez, M. J., Todling, R., Molod, A., Takacs, L., Randles, C. A., Darmenov, A., Bosilovich, M. G., Reichle, R. et al. (2017), 'The modern-era retrospective analysis for research and applications, version 2 (merra-2)', *Journal of climate* **30**(14), 5419–5454.

## References iii

- Hanna, R. & Oliva, P. (2015), 'The effect of pollution on labor supply: Evidence from a natural experiment in Mexico City', *Journal of Public Economics* **122**, 68–79.
- Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, A., Muñoz-Sabater, J., Nicolas, J., Peubey, C., Radu, R., Schepers, D. et al. (2020), 'The era5 global reanalysis', *Quarterly Journal of the Royal Meteorological Society* **146**(730), 1999–2049.
- Hou, X., Fei, D., Kang, H., Zhang, Y. & Gao, J. (2018), 'Seasonal statistical analysis of the impact of meteorological factors on fine particle pollution in china in 2013–2017', *Natural Hazards* **93**(2), 677–698.

## References iv

- Landrigan, P. J., Fuller, R., Acosta, N. J., Adeyi, O., Arnold, R., Basu, N. N., Baldé, A. B., Bertollini, R., Bose-O'Reilly, S., Boufford, J. I., Breyse, P. N., Chiles, T., Mahidol, C., Coll-Seck, A. M., Cropper, M. L., Fobil, J., Fuster, V., Greenstone, M., Haines, A., Hanrahan, D., Hunter, D., Khare, M., Krupnick, A., Lanphear, B., Lohani, B., Martin, K., Mathiasen, K. V., McTeer, M. A., Murray, C. J., Ndahimananjara, J. D., Perera, F., Potočník, J., Preker, A. S., Ramesh, J., Rockström, J., Salinas, C., Samson, L. D., Sandilya, K., Sly, P. D., Smith, K. R., Steiner, A., Stewart, R. B., Suk, W. A., van Schayck, O. C., Yadama, G. N., Yumkella, K. & Zhong, M. (2018), 'The Lancet Commission on pollution and health'.
- Lin, W. & Wooldridge, J. M. (2019), Testing and correcting for endogeneity in nonlinear unobserved effects models, *in* 'Panel Data Econometrics', Elsevier, pp. 21–43.

## References v

- Morrison, G., Riccaboni, M. & Pammolli, F. (2017), 'Disambiguation of patent inventors and assignees using high-resolution geolocation data', *Scientific data* **4**(1), 1–21.
- Muñoz-Sabater, J. (2019), 'Era5-land hourly data from 2001 to present'.  
**URL:** <https://cds.climate.copernicus.eu/doi/10.24381/cds.e2161bac>
- OECD (2016), The economic consequences of outdoor air pollution., Technical report.
- Torvik, V. I. (2015), Mapaffil: A bibliographic tool for mapping author affiliation strings to cities and their geocodes worldwide, in 'D-Lib magazine: the magazine of the Digital Library Forum', Vol. 21, NIH Public Access.
- Vallero, D. (2014), Chapter 3 - the science of air pollution, in D. Vallero, ed., 'Fundamentals of Air Pollution (Fifth Edition)', fifth edition edn, Academic Press, Boston, pp. 43–81.  
**URL:**  
<https://www.sciencedirect.com/science/article/pii/B9780124017337000037>

## References vi

- Van Donkelaar, A., Martin, R. V., Li, C. & Burnett, R. T. (2019), 'Regional estimates of chemical composition of fine particulate matter using a combined geoscience-statistical method with information from satellites, models, and monitors', *Environmental science & technology* **53**(5), 2595–2611.
- Xu, G., Jiao, L., Zhang, B., Zhao, S., Yuan, M., Gu, Y., Liu, J., Tang, X. et al. (2017), 'Spatial and temporal variability of the pm<sub>2.5</sub>/pm<sub>10</sub> ratio in wuhan, central china', *Aerosol and Air Quality Research* **17**(3), 741–751.
- Zacchia, P. (2018), 'Benefiting colleagues but not the city: Localized effects from the relocation of superstar inventors', *Research Policy* **47**(5), 992–1005.
- Zivin, J. G. & Neidell, M. (2012), 'The impact of pollution on worker productivity', *American Economic Review* **102**(7), 3652–3673.