Heard the News? Environmental Policy and Clean Investments

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Objective and motivation

How has media attention to environmental policy evolved over the years?

How does information about environmental policy affect clean investments?

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- We aim to construct meaningful newspaper-based measures of US environmental policy over the last 40 years:
 - 1. general index of environmental policy
 - 2. 25 topic-specific indexes

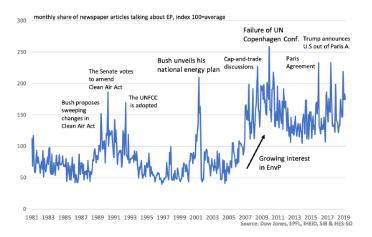


Contribution

- 1. News are a great tool to study information about environmental policy
 - high frequency data over long time periods (news arrive daily)
 - covers various topics and multi-dimensional facets of environmental policy (Brunel and Levinson, 2016; Botta and Kozluck, 2014)
 - captures aspects of policy design (e.g. uncertainty, complexity)
- 2. Novel methods relying on **text-mining techniques** (Gentzkow and Shapiro, 2010; Bybee et al, 2020)
- 3. Using our novel indices, we study whether more media attention to EnvP makes clean firms more attractive to investors.

Key findings

Historical evolution of US environmental policy + 25 sub-topics



Meaningful empirical association between our news index and proxies for clean investments (in VAR and firm-level estimations).

Outline

Measuring environmental policy

Data

Environmental Policy Index

Topic-specific indexes

EnvP news and clean investments

Aggregate clean investments - VAR models Firm-level estimations

Conclusions

Data

- News articles extracted from 10 US newspapers over 1981-2019.
- Monthly counts of articles relating to environmental and climate policy (EnvP) + total volume of articles.
- Source: automated access to Factiva, Dow Jones.

Share
22.5%
15.3%
13.8%
11.5%
10.8%
9.8%
6.2%
5.0%
3.4%
1.7%

Measuring environmental policy

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Environmental Policy Index

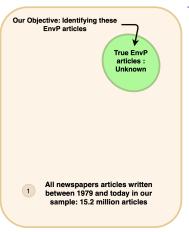
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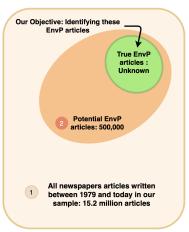
Conclusions

Identifying EnvP articles through text-mining (1)



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- 2. Broad environmental policy query to narrow down the universe of articles (i.e. 500,000)
- 3. Training set: random draw of 2,500 articles that we label manually. An article is coded as irrelevant in our codebook if:
 - ▶ No environment : "Brexit has caused changes in the business climate."
 - ► No policy : "New technological breakthrough for solar cells."

Identifying EnvP articles through text-mining (2)

Training a supervised ML algorithm for text classification

- ► Algorithm produces a rule predicting whether an article is about EnvP, based on words in a given article.
- ▶ Best performing algorithm: precision: 77% (close to average precision of annotators of 83%), and recall 65%.

► What are precision and recall?

Identifying EnvP articles through text-mining (2)

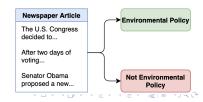
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Classifying our newspaper articles

► Using SVM prediction rule on our set of 500,000 articles, we identify 84,000 news articles as "true" EnvP



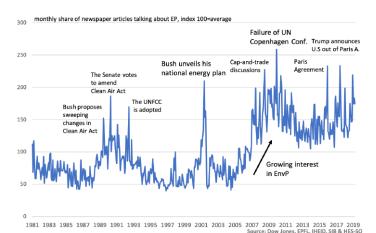
A glimpse into SVM top features

Table: 50 most discriminating words for predicting our EnvP index according to the trained SVM classifier.

Word	Weight	Word	Weight	Word	Weight
energy	3.16	crist	1.34	volkswagen	1.09
emission	3.06	air	1.33	refrigerator	1.08
environmental	2.95	ethanol	1.32	utility	1.07
epa	2.24	global warming	1.32	cleanup	1.06
solar	2.17	coal	1.30	federal	1.05
obama	2.05	climate	1.26	car	1.00
clean	1.89	regulation	1.24	penalty	0.99
pollution	1.83	program	1.18	house	0.98
waste	1.67	renewable	1.17	bannon	0.98
warming	1.62	reef	1.15	bill	0.98
recycle	1.47	protection	1.14	mercury	0.97
power	1.45	climate change	1.12	electric	0.96
global	1.38	env. protection	1.10	gasoline	0.94
standard	1.36	clean air	1.10	environment	0.94

▶ Top SVM articles

General EnvP Index

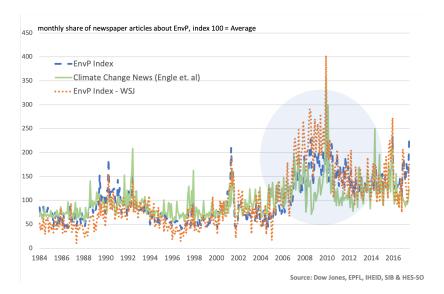


Historial evolution of EnvP media coverage (scaled by total volume of articles).





EnvP versus Climate Change Index



▶ Climate policy sentiment

Measuring environmental policy

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Aggregate clean investments - VAR models Firm-level estimations

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Identifying EnvP topics

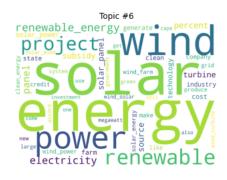
Unsupervised ML algorithm for topic identification

- 84,000 EnvP articles over 1981-2009.
- Topic modeling using Latent Dirichlet Allocation (LDA).

Identifying EnvP topics

Unsupervised ML algorithm for topic identification

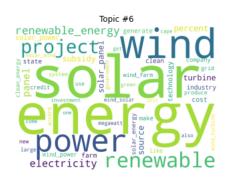
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Identifying EnvP topics

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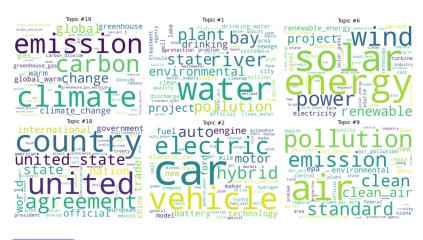
- 84,000 EnvP articles over 1981-2009.
- Topic modeling using Latent Dirichlet Allocation (LDA).
- ► LDA probablistically uncovers hidden content structure (here: 25 topics) based on co-occurrence of terms.
- Each news article is associated with multiple topics.



Example document

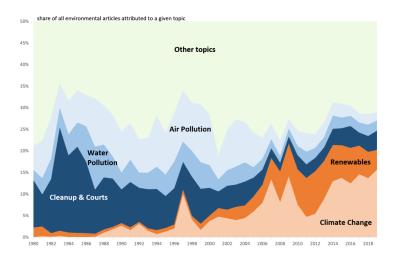


Wordclouds EnvP topics

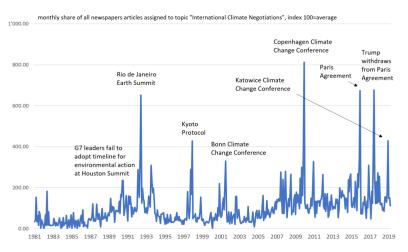


▶ Full topics list

Evolution of topic sub-indexes over time

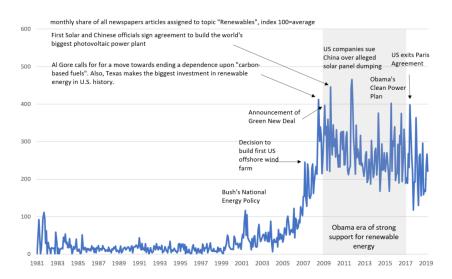


Topic International Agreements



Source: Dow Jones, EPFL, IHEID, SIB & HES-SO

Topic Renewables - EnvP-RE



Source: Dow Jones, EPFL, IHEID, SIB & HES-SO

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EnvP news and clean investments

How do our news-based environmental policy indexes relate to clean markets?

- ► **Hypothesis**: More media attention to EnvP makes clean firms more attractive to investors.
- ► **Channels**: increased awareness of investment opportunities in clean markets **and** reflection of increased policy stringency.

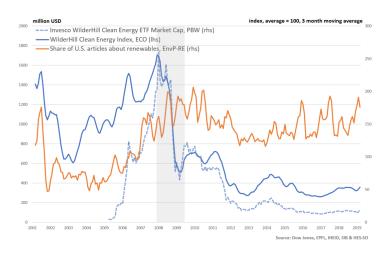
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- ► Channels: increased awareness of investment opportunities in clean markets and reflection of increased policy stringency.
- Dynamic relationship between our news index and demand for the main benchmark clean-energ ETF and aggregate VC deals in VAR models
 - Impulse response of to a shock in our renewable policy news index.
- 2. Firm-level regressions identification strategy differentiate firms by exposure to environmental policy.

► EnvP vs OECD Policy Stringency

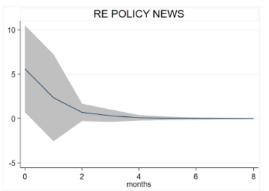
Clean-energy stocks (1)



ECO and PBW-ETF WilderHill Clean-energy Index, 2001-2019 monthly. Source: Datastream

Clean-energy stocks (2)

Impulse response function: effect of a shock in EnvP-RE news on clean-energy fund demand (AuM of PBW ETF).

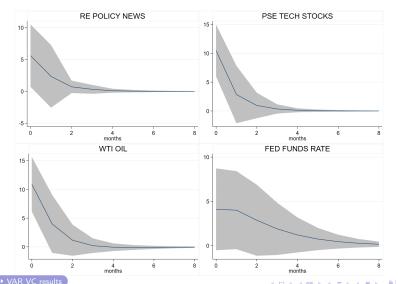


A one-SD shock to EnvP-RE news growth is associated with an increase in 5 million USD (0.1 SD) in AuM of the PBW ETF

► Robustness to different VAR specifications Y ► Variables and Cholesky ordering

Clean-energy stocks (3)

Impulse response function: effect of other shocks on clean-energy fund demand (AuM of PBW ETF).



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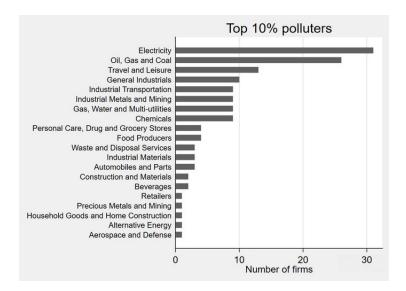
Firm-level estimations

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Firm-level stock returns (1)

- ► **Hypothesis**: We expect polluting (green) firms to be negatively (positively) affected by EnvP news.
- ▶ **Identification**: differentiate firms by exposure to environmental policy (annual scope-1 GHG emissions / green revenues)
- ► Sample: 1,400 US firms, Jan 2004-Mar 2019.
 - Monthly firm-level stock returns and annual firm-level balance sheet data from Datastream.
 - Annual firm-level GHG emissions (scope 1) from Trucost.
 - For a subsample, we obtain additional data on firms' yearly green revenue shares (FTSE Russell)

Firm-level GHG emissions



Firm-level stock returns (2)

How does media attention to environmental policy affect firm-level stock returns?

► LHS: continuously compounded log returns in excess of safe interest rate for each firm.

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- ▶ RHS: unanticipated (white-noise) component of EnvP: $EnvP_t = \alpha + \sum_{k=1}^7 \beta_k EnvP_{t-k} + \varepsilon_t^{EnvP}$ (Brogaard & Detzel, 2015).

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Additional controls

- Firm heterogeneity: size, profitability, leverage, dividends per share and fixed effects.
- ► **Technological progress**: industry-year time trend.
- ► Month FE or Fama-French risk factors (MKTRF, SMB, HML, RMW and CMA).

	(1) ln(r_excess)	(2) ln(r_excess)	(3) ln(r_excess)	(4) ln(r_excess)	(5) ln(r_excess)	(6) ln(r_excess)	(7) ln(r_excess)
EnvP	0.0231*** (0.0006)	0.0231*** (0.0006)	0.0205*** (0.0008)	0.0205*** (0.0008)	0.0351*** (0.0022)	0.0224*** (0.0016)	0.0219*** (0.0013)
EnvP × AVG Emissions	-0.0026*** (0.0004)		-0.0023*** (0.0004)				-0.0013*** (0.0004)
$EnvP \times AVG$ Emission Intensity		-0.0021*** (0.0004)		-0.0016*** (0.0005)			
$EnvP \times Quartile of emissions=2$					-0.0107*** (0.0025)		
EnvP \times Quartile of emissions=3					-0.0169*** (0.0025)		
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EnvP \times Quartile of emission intensity=2						-0.0012 (0.0022)	
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EnvP \times Quartile of emission intensity=4						-0.0043** (0.0020)	
Green Revenue Share							0.0025 (0.0027)
$\operatorname{EnvP} \times \operatorname{Green} \ \operatorname{Revenue} \ \operatorname{Share}$							0.0060*** (0.0015)
Firm FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Industry-Year Trend	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm controls	No	No	Yes	Yes	Yes	Yes	Yes
Risk factors	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	69,668	69,668	34,689	34,689	34,689	34,689	9,579
Firms	1,400	1,400	613	613	613	613	230
\mathbb{R}^2	0.95	0.95	0.96	0.96	0.96	0.96	0.95

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EnvP net sentiment \times AVG Emissions	0.0056*** (0.0011)
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Conclusions and next steps

- News provide a very rich dataset covering many environmental topics at high frequency
- Novel methods based on text-mining ML algorithms to extract information from news
- Meaningful relationship between our news-based index and clean markets → firms exposed to environmental policy respond to EnvP news
- Ongoing work on news-index to measure uncertainty about environmental policy (rollbacks, awaiting decisions from courts, etc)

Supplementary Slides

Our codebook

0	Article is about foreign (non US) environmental policy (and is not discussed/compared to or in relation to US env policy)	
	RELEVANT below	
1	Article is about environment/climate, with minor but significant (=explicit and specific, not a general statement) reference to environmental policy. (even if it's a opinion piece) - Article is about local environmental impacts in a very specific geographical area, with some reference to state or federal env/climate policy - Note: env policy implies legislation, laws, but also the financing of large demonstration projects, renewable power plants, etc by a (local) public authority.	

Figure: Excerpt from our codebook



Support Vector Machines (SVM)

SVM maximizes the distance between the two closest articles on both sides of the decision boundary:

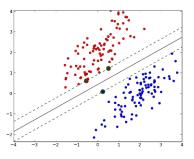
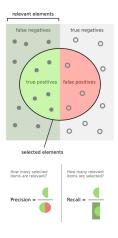


Figure: Support Vector Machines



Precision & Recall



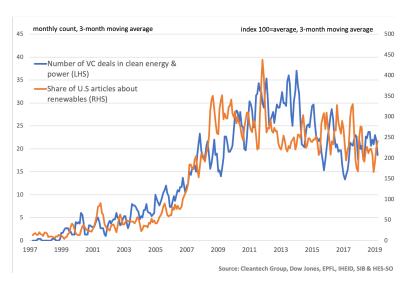


Articles with highest SVM score

Title	Date	SVM	Newspaper	Excerpt
Time to Confront Climate Change	Dec 28, 2012	4.78	New York Times	"That ruling, known as the endangerment finding, made possible the administration's historic new emissions standards for cars and light rucks. It also provided the basis for the first steps toward regulating emissions from new power plants, and, possibly, further steps requiring existing plants to reduce global warming pollution."
Environmentalists, Industry Air Differences on Pollution	Oct 17, 1999	4.66	Washington Post	"As a result, environmental groups are pressing states and Congress for specific environmental protections against increased pollution, financial incentives for energy efficiency and renewable energy, and federal pollution guidelines to be part of the overall deregulatory effort."
Trump can't do much to worsen climate change	Apr 2, 2017	4.64	Washington Post	"Tump does not want to regulate carbon or other fossil-fuel -related pollutants under the Clean Air Act, but the statute and the Supreme Court say that he must. As Tump repeals the Clean Power Plan and updated limits on mercury, coal ash and smog, he will face legal challenges that he may well lose."
On Environmental Rules, Bush Sees a Balance, Critics a Threat	Feb 23, 2003	4.55	New York Times	"Whether rejecting a treaty on global warming, questioning Clinton-era rules on forest protection or pressing for changes in landmark environmental laws, Mr. Bush has imposed a distinctive stamp on a vast landscape of issues affecting air, water, land, energy and the global climate.
Candidates Agree World Is Warming, but Talk Stops There	Oct 26, 2012	4.48	New York Times	"Mr. Obama has supported broad climate change legislation, financed extensive clean energy projects and pushed new regulations to reduce global warming emissions from cars and power plants."

→ Back

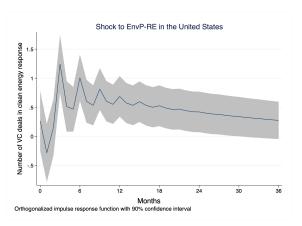
Clean-energy VC deals (1)



Source VC deals: Cleantech Group i3 database, 1997-2019, monthly number of deals in Clean-energy.

Clean-energy VC deals (2)

Impulse response function: effect of a shock in Env-RE news on number of VC deals.

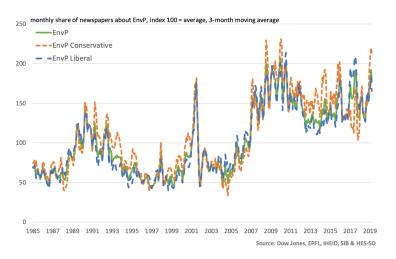


A one-SD increase in EnvP-RE news is associated with nearly one more VC deal in Clean-energy three months after the shock.

▶ Robustness to different VAR specifications \(\) \(\) Variables and Cholesky ordering

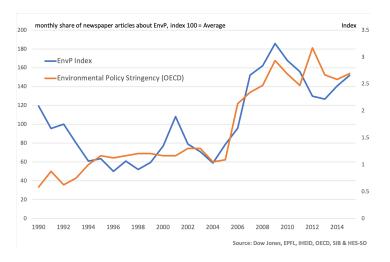


Political slant?



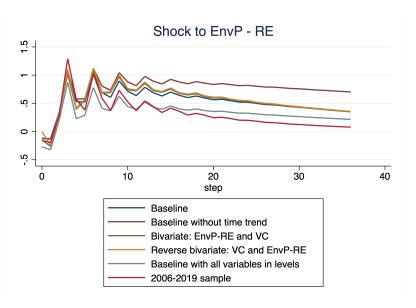
Conservative-leaning – WSJ, Houston Chronicle, Boston Herald, Dallas Morning News. **Liberal-leaning** – NYT, Washington Post, SFC, Tampa Bay Times, San Diego Union Tribune and San Jose Mercury News.

EnvP Index vs OECD Stringency Index





Robustness VAR VC



Variables and Cholesky ordering in VAR VC deals

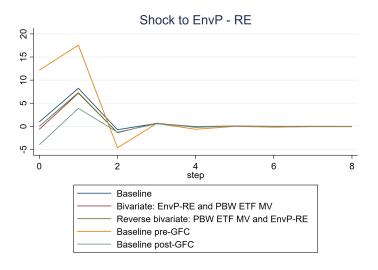
Table: Baseline VAR VC deals

Variables	Version used	Cholesky ordering
Our EnvP-RE policy index	Levels	1
US West Texas Intermediate crude oil	Log diff	2
spot price		
GDP	Log diff	3
Federal funds effective rate	First diff	4
Number of VC deals in Clean-energy	Levels	5

Time trend; 3 lags.



Robustness VAR stock



Estimated effect of a shock in EnvP-RE news index on PBW-ETF market cap changes, monthly



Variables and Cholesky ordering in VAR stock

Table: Baseline VAR stock

Variables	Version used	Cholesky ordering
Our EnvP-RE policy index	Log diff	1
US West Texas Intermediate crude oil	Log diff	2
spot price		
Federal funds effective rate	First diff	3
NYSE Arca Technology Stock Index	Log diff	4
WilderHill Clean-energy Stock Index	Log diff	5

No time trend; 2 lags.



Firm-level VC investments (1)

- ➤ Crunchbase: 31,808 active startup firms, venture capital funding rounds over Jan 1998-Mar 2019, firm-quarter panel dataset
- Estimate impact on EnvP news on probability of receiving VC funding (and amount) in next quarter
- Identification strategy differentiates startups by exposure to environmental policy → cleantech vs. other startups
- ightharpoonup Cleantech = 4%, Clean-energy = 2.4% of VC deals
- Controls: GDP growth, Fed funds rate, oil price, firm's age

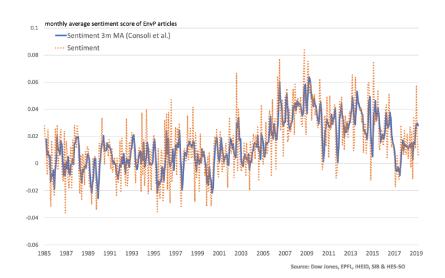
$$VC_{i,t+s} = \alpha + \beta_1 EnvP_t + \beta_2 EnvP_t \cdot Cleantech_i + \beta_3 Controls_{i,t} + \beta_4 TimeTrend_t + \gamma_{quarter/year/industry/state/series} + \epsilon_{i,t}$$

Firm-level VC investments (2)

	(1) Funded (Q+1)	(2) Funded (Q+1)	(3) Funded (Q+1)	(4) Amount (Q+1)	(5) Amount (Q+1)
Log EnvP index	0.00470*** (0.00121)	-0.00529*** (0.00195)	-0.00522** (0.00214)	-0.114** (0.0467)	-0.127*** (0.0467)
Cleantech	-0.126*** (0.0162)	-0.124*** (0.0197)	-0.113*** (0.0200)	-1.707*** (0.586)	-2.736*** (0.614)
Log EnvP x Cleantech	0.0272*** (0.00332)	0.0278*** (0.00461)	0.0209*** (0.00481)	0.391*** (0.119)	0.760*** (0.136)
Log Sentiment Index			-0.00148*** (0.000573)		
Log Sentiment x Cleantech			0.00504*** (0.00171)		
Industry controls	Yes	Yes	Yes	Yes	Yes
Other controls	Yes	Yes	Yes	Yes	Yes
Quarter FE	No	Yes	Yes	Yes	Yes
Year FE	No	Yes	Yes	Yes	Yes
State FE	No	Yes	Yes	Yes	Yes
Industry-time trend	No	Yes	Yes	No	Yes
Series FE	No	Yes	Yes	No	Yes
Observations R^2	1056221 0.003	1056221 0.006	980975 0.006	57319 0.133	57319 0.135

A doubling of EnvP media coverage is associated with an increase of receiving VC funding by 1.6%-pt (=26% of average probability that a clean-tech startup gets funded). Back

Climate Policy Sentiment





Topics list

Topic	#	Topic	#	Topic	#
Climate Change	19	Oil & Gas production	15	Vehicle Fuels	12
EPA & Federal Gov.	5	Intl. Climate Negotiations	18	Waste & Recycling	26
Cleanups & Courts	17	Texas	11	Green Buildings	25
Government Budgets	3	Renewables	6	North-East Region	8
Air Pollution	9	Env. Conservation	4	Offshore Oil Drilling	7
Congress & Policy	13	Water Pollution	1	Nuclear Power	21
Businesses & Investments	22	Climate Science	16	Coal Industry	10
Presidents & Campaigns	23	California	14		
Power & Utilities	24	Automobile Industry	2		

Table 4: Topic interpretation and classification (ranked by size). Topic # refers to labels in Figures 5 and 6.

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