# Predicting Policy Change with Text Data

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The views expressed here are the authors' and do not represent the views of Bates White, the Mercatus Center, or their other employees.

#### Our use case: China

The Policy Change Index (PCI): a machine learning model that predicts **policy changes** by "reading" this:



## How can it possibly work?!

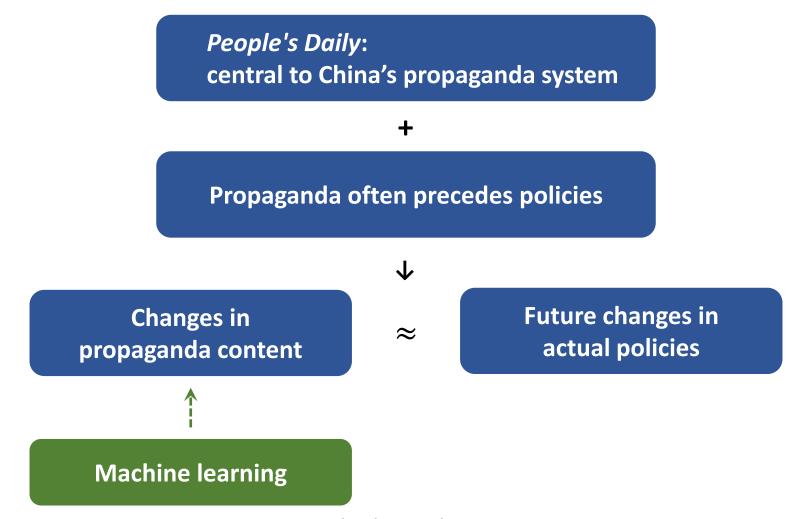
Because propaganda often precedes policies.

## How can it possibly work?!

Because propaganda often precedes policies.



## The "value" of propaganda



## Challenge: lack of labels

- Propaganda is abstract and subjective.
- We narrow it down to editorial priorities.
- Still, no labels—editorial priorities are in the editor's mind!

#### What's in the editor's mind??

Transform estimation problems to anomaly detection problems.

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#### Estimation: infeasible

- Newspaper text comes with page number.
- The infeasible estimation problem:

```
g: \{(text, FrontPage)\} \rightarrow \{"editorial priorities"\}
```

- g("pvt sector is great", on front page) = (reform is high-priority);
- g("central planning is great", on front page) = (reform is low-priority).
- Only if *g* can be learned...

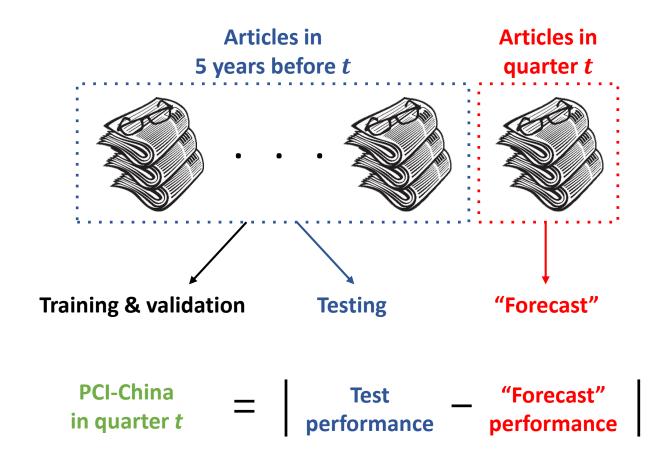
#### Anomaly detection: feasible

• Treat "editorial priorities" as a latent variable:

```
f_{\text{"editorial priorities"}}: \{text\} \rightarrow \{FrontPage\}
```

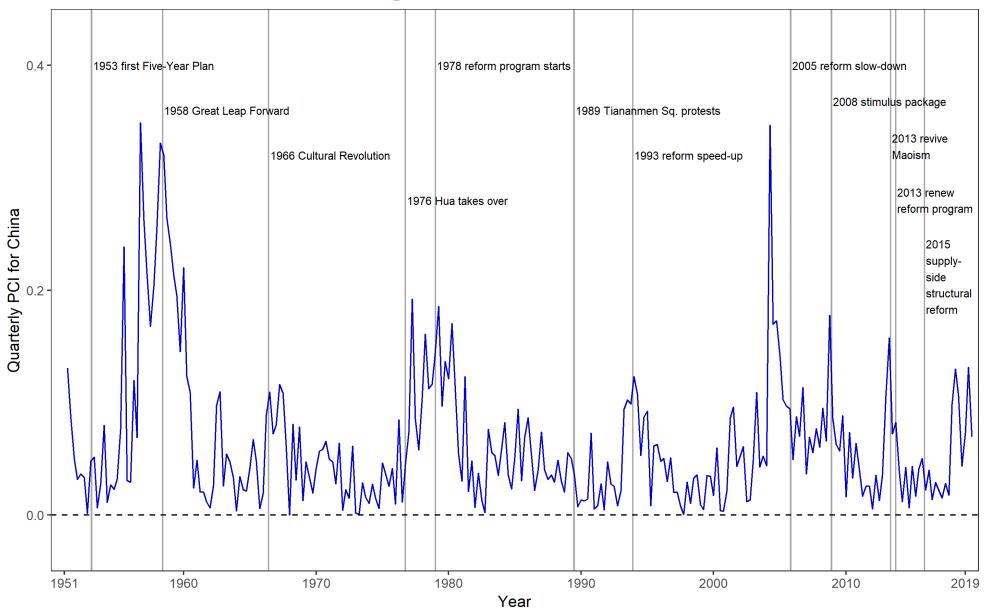
- Lots of labels for f.
- Detecting changes in f: a feasible anomaly detection problem.
- Change in  $f \Rightarrow$  change in propaganda priorities  $\Rightarrow$  change in policy!

#### Policy Change Index for China (PCI-China)

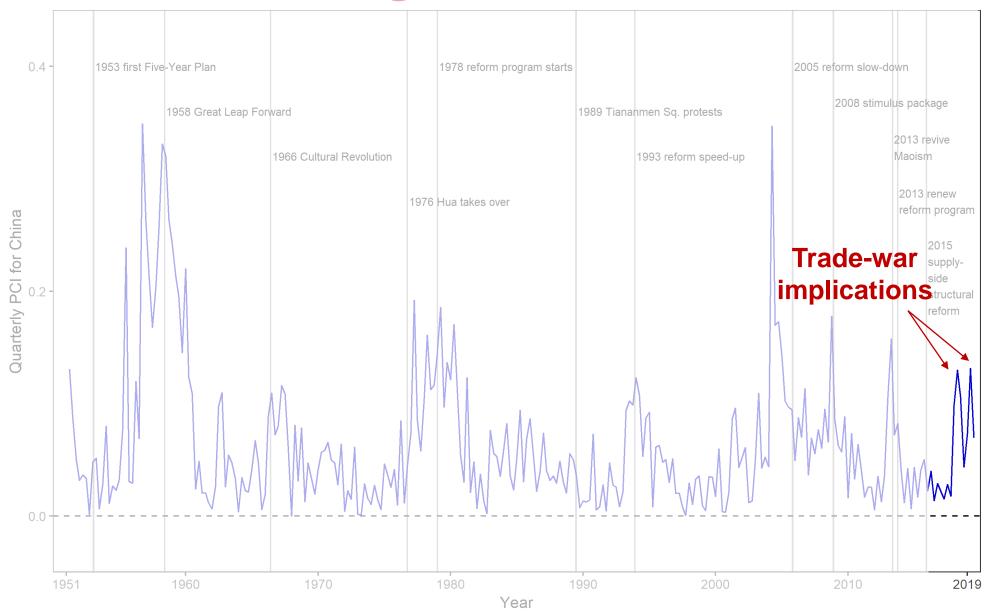


Large PCI-China ⇒ Structural change in editorial priorities

## PCI-China: result v ground truth



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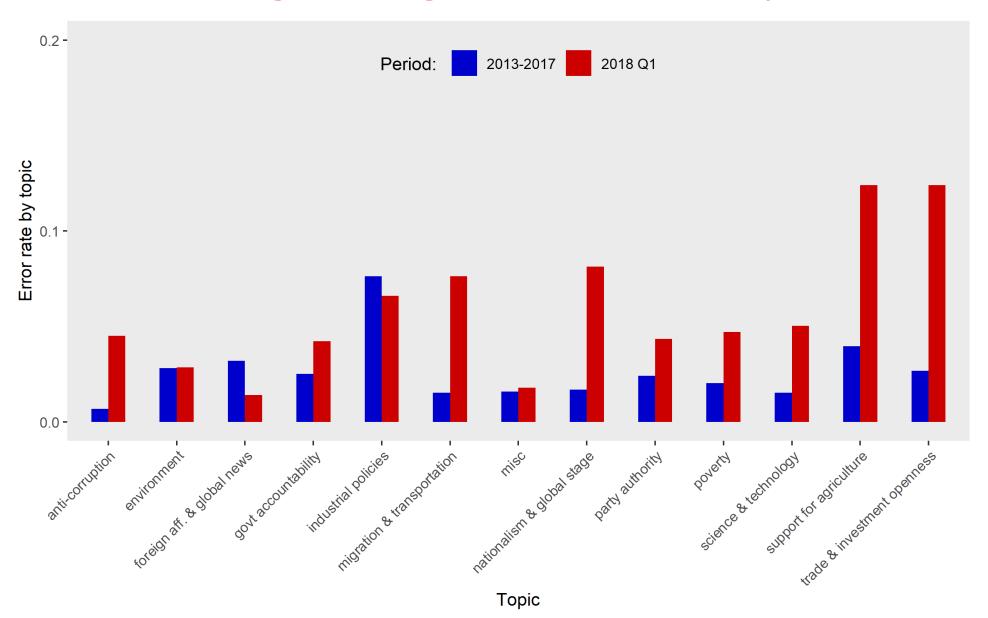
## Understanding changes

		Classified on front page?	
		No	Yes
Front page?	No		false positives
	Yes	false negatives	

#### Content of *misclassified* articles has policy substance:

- False negatives suggest new policies;
- False positives suggest phasing-out policies.

#### Understanding changes: 2018 Q1 uptick



## Trade-war implications

Two upticks in 2018 Q1 and 2019 Q1:

- Internally: strengthening party authority;
- Externally: nationalism and global leadership.

A trade deal with structural reforms? Curb your enthusiasm...

## How can these be useful to you?

#### Remember Trick #2?

When there's no label for the target variable, transform the estimation problem to an anomaly detection problem.

Many other use cases share the same structure!

#### Future PCI projects

Stay tuned for more in the PCI pipeline:

- PCIs for other countries: North Korea, Russia, Cuba,...
- PCI sub-indices, by subject: trade policy, fiscal policy,...
- And more!

#### Back to Trick #1

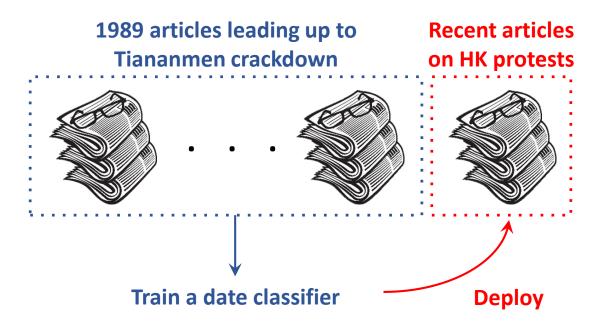
Propaganda often precedes policies.

More generally, words often foreshadow actions.

#### Consider the Hong Kong protests

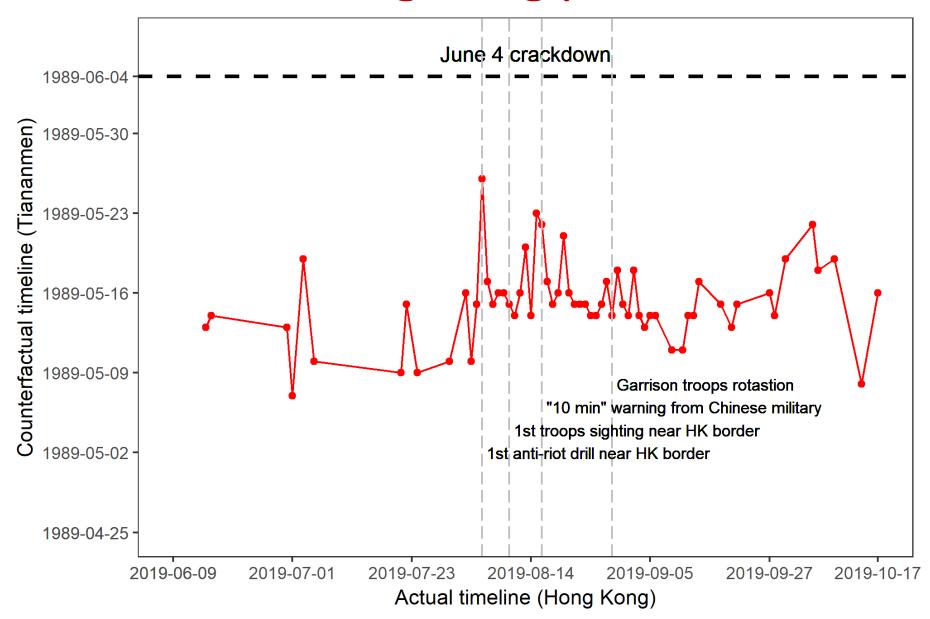
- Will the protests be met with a military crackdown by China?
  - → How hard is the People's Daily slamming Hong Kong?
- There's a unique, well-formed precedent: Tiananmen crackdown.

#### Policy Change Index for Crackdown (PCI-Crackdown)



PCI-Crackdown: mapping each current date to an "as-if" date

## PCI-Crackdown: Hong Kong protests



## **Takeaways**

1. Words often foreshadow actions.

2. No labels? Try transforming estimation problems to anomaly detection problems.

Website: policychangeindex.org

(papers, code, newsletter,...)