Reading China:

Predicting Policy Change with Machine Learning

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Policy Change Index (PCI) for China

The first *leading* indicator of China's policy changes.

- Covers 1951 Q1 2018 Q3.
- Can be updated in the future.

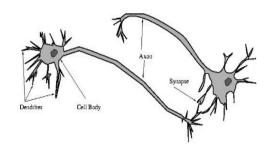
How to predict policy changes?

Build a machine learning algorithm to

- "read" the *People's Daily*;
- detect changes in how it prioritizes policy issues.



Official newspaper, 1946-2018



(Artificial) neural networks

Source of predictive power

The Leninist tradition:

- "[T]he whole task of the Communists is to be able to convince the backward elements."
- It is a fundamental necessity "to transform the press from an organ which primarily reports the political news of the day into a serious organ for the economic education of the mass of the population."

Source of predictive power

People's Daily: nerve center of China's propaganda system

+

Propaganda often precedes policies.

 $\mathbf{1}$



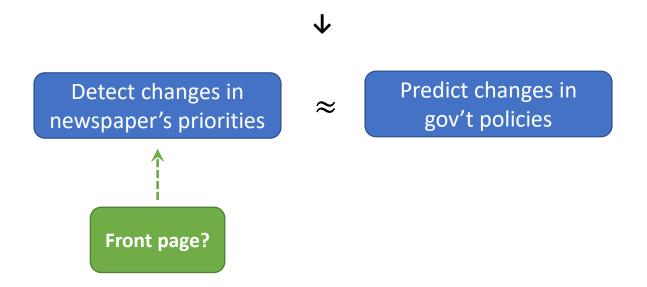
Predict changes in gov't policies

Source of predictive power

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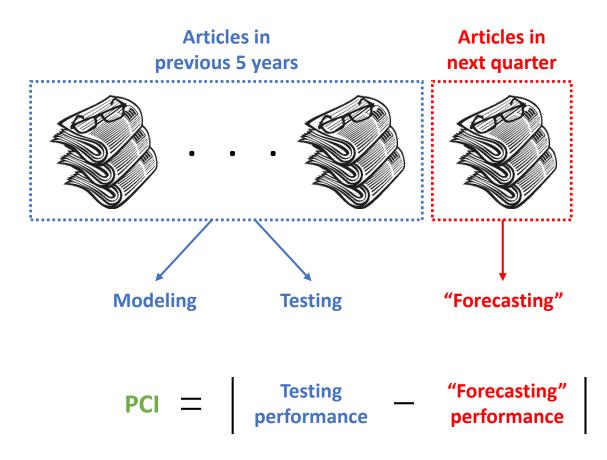


Methodology

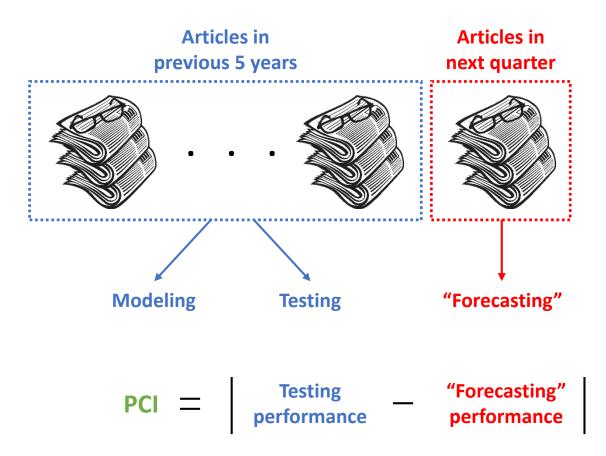
Imagine an avid reader of the People's Daily who

- 1. reads recent articles;
- 2. forms a paradigm about front-page content;
- 3. tests the paradigm on new articles.

Methodology



Methodology



"Language-free" — it does **not** require the reading of the Chinese text.

Methodology: data

sample_data

	date	year	month	day	page	title	body	id
0	2018- 10-01	2018	10	1	1	习近平在会见四川航空"中国民航英雄机组"全体成员时强 调 学习英雄事迹 弘扬英雄精神 将非凡英	中共中央总书记、国家主席、中央军委主席习近平专门邀 请四川航空"中国民航英雄机组"全体成员	2018100000
1	2018- 10-01	2018	10	1	1	烈士纪念日向人民英雄敬献花篮仪式在京隆重举行 习近 平李克强栗战书汪洋王沪宁赵乐际韩正王岐山出席	9月30日上午,党和国家领导人习近平、李克强、栗战书、 汪洋、王沪宁、赵乐际、韩正、王岐山	2018100001
2	2018- 10-01	2018	10	1	1	庆祝中华人民共和国成立69周年 国务院举行国庆招待 会 习近平栗战书汪洋王沪宁赵乐际王岐山等	9月30日晚,国务院在北京人民大会堂举行国庆招待会,热 烈庆祝中华人民共和国成立六十九周年	2018100002
3	2018- 10-01	2018	10	1	2	习近平就印度尼西亚中苏拉威西省地震海啸向印尼总统佐 科致慰问电	新华社北京9月30日电 9月30日,国家主席习近平就印度 尼西亚中苏拉威西省发生强烈地震及	2018100003
4	2018- 10-01	2018	10	1	2	在庆祝中华人民共和国成立六十九周年招待会上的致辞 中华人民共和国国务院总理 李克强 (二〇	各位来宾、各位朋友、同志们: 今天,我们隆重庆祝 中华人民共和国成立六十九周年。新中国波澜壮	2018100004
5	2018- 10-01	2018	10	1	2	用奋斗成就复兴伟业(社论) ——热烈庆祝中华人民共和国 成立69周年	时间的年轮,刻印下奋斗者的足迹。当10月的阳光照耀大 地,我们迎来了人民共和国69岁华诞。…	2018100005
6	2018- 10-01	2018	10	1	2	国务院印发《决定》 进一步压减工业产品生产许可证管 理目录和简化审批程序	新华社北京9月30日电 经李克强总理签批,国务院日前印 发《关于进一步压减工业产品生产许可	2018100006
7	2018-	2018	10	1	2	谱写新时代乡村全面振兴新篇章 ——论学习习近平总书 记光工家饰长封框兴战歌乐声进活转神	本报评论员 乡村振兴既是一场攻坚战,更是一场持久 战 应汤服空信心 哈空中提 莱士家士	2018100007

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Methodology: data

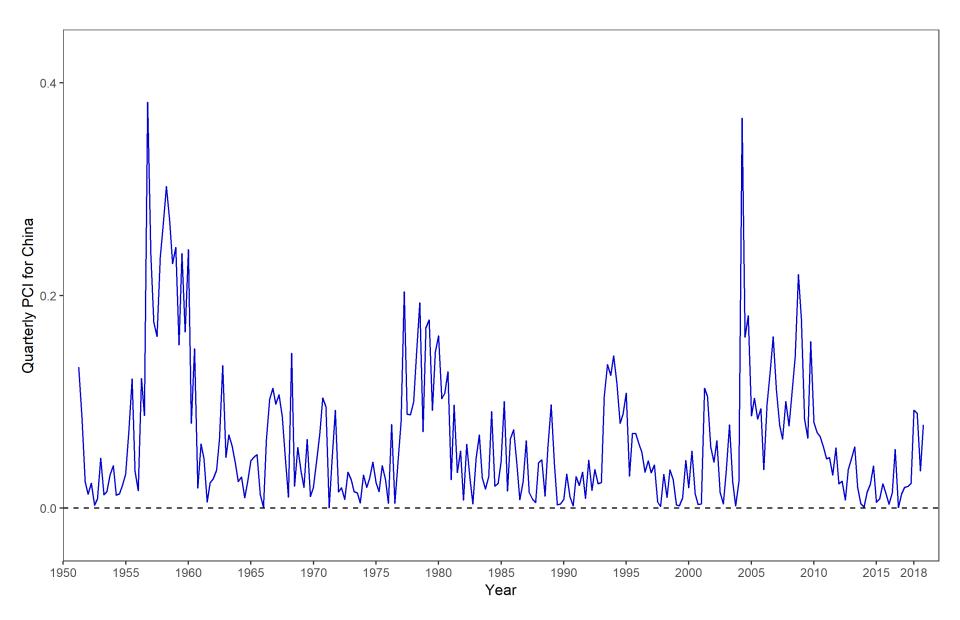
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Methodology: modelling Input Metadata **Texts** $\boldsymbol{\mathcal{X}}$: each article as an observation. Word embedding Recurrent Neural Multilayer *f* : a complicated function. neural networks perceptron networks **Multilayer** perceptron y = f(x)Output Front page?

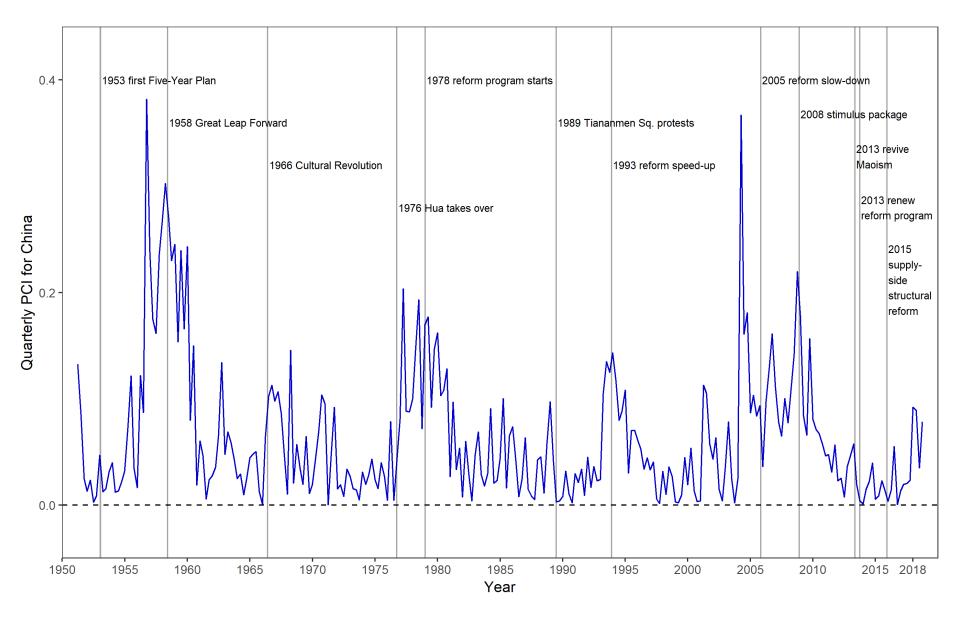
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Results

Result: PCI



Result: PCI — with ground truth



Understanding substance of change

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

• Content of *mis*-classified articles has policy substance.

Supervised learning: a digression

Supervised learning

 $mapping: X \to Y$

- Trained on $\{x_i, y_i\}_{i \in training}$.
- Goal: from $\{x_j\}_{j \in new}$, to predict $\{y_j\}_{j \in new}$.
- Challenge: need lots of training data.

an *in*feasible approach

 $g: \{(Article, FrontPage)\} \rightarrow \{(Policy, Priority)\}$

- With the learned function *g*:
 - g("pvt sector is important", front page) = (reform, high priority);
 - *g*("central planning is great", front page) = (reform, low priority); ...

an *in*feasible approach

 $g: \{(Article, FrontPage)\} \rightarrow \{(Policy, Priority)\}$

- With the learned function *g*:
 - *g*("pvt sector is important", front page) = (reform, high priority);
 - *g*("central planning is great", front page) = (reform, low priority); ...
- But where are the training data?

a feasible approach

• Think of priorities as a latent variable:

 $f_{\{(Policy, Priority)\}} : \{Article\} \rightarrow \{FrontPage\}$

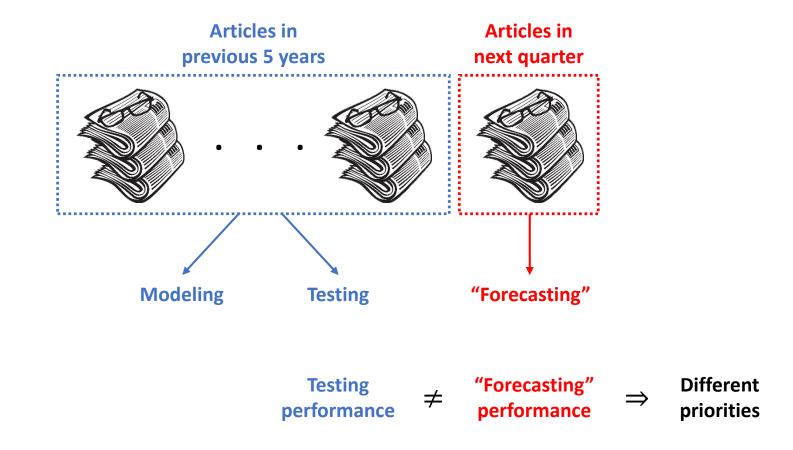
a feasible approach

• Think of priorities as a latent variable:

 $f_{\{(Policy, Priority)\}} : \{Article\} \rightarrow \{FrontPage\}$

- Lots of training data to learn each function *f*.
- Difference in function \Rightarrow difference in priorities.
- "Language-free!"

a feasible approach



Other applications

App 1: PCIs for other countries

Predicting other (ex-)Communist regimes' policies:

- Soviet Union's *Pravda*
- East Germany's Neues Deutschland
- North Korea's *Rodong Sinmun*
- Cuba's Granma
- Vietnam's Nhân Dân

App 1: PCIs for other countries

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Work in progress.

App 2: measuring media bias

Measuring media bias in the US:

- Replicate the same analysis on US newspapers.
- Compare "*PCI*"_{*WaPo}, "PCI*"_{*NYT*}, "*PCI*"_{*WSJ*}, ...</sub>
- Divergence among "*PCI*"s ---> Polarization in media

App 3: predicting vote change

Predicting vote change in legislation:

- Newspaper texts ----> Legislators' public statements
- Page numbers ---> Legislators' names
- What if Sen. A's statement is mistaken as Sen. B's?

Source of predictive power:

• Political necessity to justify vote changes by making different statements *in advance*.

Other apps

Omitted here. See our research paper.

Interested in DIY?

- Website: policychangeindex.com (newsletter sign-up)
- Paper: policychangeindex.com/pdf/Reading_China.pdf
- Source code: github.com/PSLmodels/PCI
- A simulated example to show how the PCI works.

Questions?