Policy Experimentation in China: the Political Economy of Policy Learning

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 - From point to surface: trying out policies in a number of localities before deciding to roll out as national policies.
 - Many speculated that policy experimentation plays a vital role in China's reform and growth. (e.g., Rawski 1995; Cao et al. 1999; Roland 2000; Qian 2003)
- However, little is understood about the characteristics of policy experimentation, and how they may affect policy learning and policy outcomes.

Policy experimentation: examples

B.1 Carbon emission trading

During 2011-2021 Experimentation in 1 wave 7 provinces / prefectures as experimentation sites



B.2 Separation of permits and licenses

During 2015-2018 Experimentation in 3 waves 24 prefectures as experimentation sites



Policy experimentation: examples

B.1 Carbon emission trading

During 2011-2021 Experimentation in 1 wave 7 provinces / prefectures as experimentation sites



B.3 Agriculture catastrophe insurance

During 2017-2021 Experimentation in 2 waves 14 provinces as experimentation sites



B.2 Separation of permits and licenses

During 2015-2018 Experimentation in 3 waves 24 prefectures as experimentation sites



B.4 County fiscal empowerment reform

During 2002-2015 Experimentation in 10+ waves 1,246 counties as experimentation sites



This project

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We ask, in the context of China's policy experimentation:

- 1. Is sample selection of experimentation representative?
- 2. Does experimentation induce excessive efforts?
- 3. Is central government naive when interpreting experimentation outcomes?
- 4. What are the implications on learning from experimentation, and national policy outcomes?

1. Are experimentation sites representative?

Is the selection of experimentation sites representative?

- For each experimentation, we compare pre-experimentation characteristics between "treatment" and "control" sites:
 - Baseline: pre-experiment local fiscal revenue;
 - Robust to using alternative characteristics including local GDP per capita, as well as those specific to policy domains.

Use t-statistics of the comparison as a summary stats:

$$t_i = rac{\hat{Y}_i(1) - \hat{Y}_i(0)}{\sqrt{rac{\hat{S}_i^{\,2}(1)}{n_{i,1}} + rac{\hat{S}_i^{\,2}(0)}{n_{i,0}}}}.$$

 Underlying t-distribution with degrees of freedom adjusted for each experimentation.

Experimentation sites substantially positively selected



▶ >90% of experiments implemented in richer localities;

Can reject null of representativeness at 90% level for at least 67% of the experiments. • Results

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- Robust to just examining subsample of early-round sites;
- Robust to other test procedures such as permutation tests.
- Specification curve visualizing all combinations.
- Share of experiments with positive exp. sites selection mildly decreases over time. Results

2. Endogenous efforts during experimentation?

Does domain-specific fiscal expenditure increase during experimentation?

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	Share of fiscal expenditure on experiment-related domains					
	(1)	(2)	(3)	(4)	(5)	(6)
Panel A: Fiscal input among experimentation sites						
# of experiments	0.003*** (0.001)	0.002*** (0.0004)	0.002*** (0.0005)	-0.013*** (0.003)	-0.002* (0.001)	-0.003 (0.002)
$\# \times {\rm career}$ incentive				0.043*** (0.007)	0.009** (0.004)	0.011** (0.005)

# of obs.	142,116	142,116	142,116	142,116	142,116	142,116
Mean of DV	0 174	0 174	0 174	0 174	0 174	0 174
County by domain FE	No	Yes	Yes	No	Yes	Yes
County by year FE	Yes	No	Yes	Yes	No	Yes
Domain by year FE	Yes	Yes	Yes	Yes	Yes	Yes

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Panel A: Fiscal input among experimentation sites						
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	(0.001)	(0.0004)	(0.0005)	(0.003)	(0.001)	(0.002)
$\# \times \text{career incentive}$				0.043***	0.009**	0.011**
				(0.007)	(0.004)	(0.005)
Panel B: Fiscal input among non-experimentation sites during national policy roll-out						
# of rolled out policies	0.001	0.001	0.001	0.001	0.001	0.001
	(0.001)	(0.0004)	(0.001)	(0.003)	(0.001)	(0.002)
$\# \times career$ incentive	. ,	. ,	. ,	-0.001	-0.0004	-0.0003
				(0.005)	(0.002)	(0.003)
# of obs.	142,116	142,116	142,116	142,116	142,116	142,116
Mean of DV	0.174	0.174	0.174	0.174	0.174	0.174
County by domain FE	No	Yes	Yes	No	Yes	Yes
County by year FE	Yes	No	Yes	Yes	No	Yes
Domain by year FE	Yes	Yes	Yes	Yes	Yes	Yes

- Event study specification: no increase in expenditure prior to exp.; Results
- Results hold among exp. w/o fiscal support from central govt.; Results
- Larger results if exp. may not require additional fiscal inputs; Pesults
- Results robust to alternative measures of career incentives.
- Results unlikely to be driven by data manipulation (exaggerated reporting).

3. Naive interpretation of experimentation outcomes?

Policy learning and location-specific shocks

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- Land sale revenue was a major fiscal windfall in the 2000s;
- Following Chen and Kung (2019), we instrument land revenue with the interaction between: (i) land suitable for commercial development due to terrain features; and (ii) demand shock due to interest rate changes.

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	National roll-out			
	(1)	(2)	(3)	
Land revenue (instrumented)	0.020***	0.039***	0.029***	
	(0.002)	(0.003)	(0.003)	
# of obs.	18,464	18,464	18,464	
Mean of DV	0.509	0.509	0.509	
Ministry FE	No	No	Yes	
Year FE	Yes	Yes	Yes	
County FE	No	Yes	Yes	

Future interest rates do not predict contemporaneous land revenues; Results

 Future revenue windfall is *not* associated with increase in policy national roll-out. Does policy's national roll-out depend on exp. outcomes?

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A. Pre vs. post

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A. Pre vs. post

B. Controlling for provincial trend

C. Synthetic control

▶ Details

4. Consequences on policy learning and policy outcomes?

Among exp. that eventually become national policies ... Systematic shrinkage in policy effects when rolled out to the entire nation?

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- Example: net zero overall effects of county fiscal empowerment exp. Details
- Results robust controlling for selection bias and endogenous efforts.
- National policy effects: what do exp. effects predict? Results

Among exp. that eventually become national policies ... Are regions similar to exp. sites benefit *more* from national policies?

Among exp. that eventually become national policies ...

Are regions similar to exp. sites benefit more from national policies?

- For each policy, calculate difference between each non-experimentation locality and experimentation sites using Mahalanobis distance: <a href="https://www.non-experimentations-active-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-complexity-c
 - Socioeconomic conditions prior to experimentation;
 - Politicians' incentives during experimentation.
- Policy and county FEs: identifying variations from composition of experimentation sites.

Among exp. that eventually become national policies ...

Are regions similar to exp. sites benefit more from national policies?

- For each policy, calculate difference between each non-experimentation locality and experimentation sites using Mahalanobis distance: • Robustness
 - Socioeconomic conditions prior to experimentation;
 - Politicians' incentives during experimentation.
- Policy and county FEs: identifying variations from composition of experimentation sites.

	GDP per capita (pre vs. post)					
	(1)	(2)	(3)			
Panel A: Selection of experimentation sites						
M-distance on local development	-0.007***	-0.007***	-0.006***			
	(0.001)	(0.001)	(0.001)			
# of obs.	77,588	77,588	77,588			
Mean of DV	0.0806	0.0806	0.0806			
Panel B: Endogenous efforts during experimentation						
M-distance on career incentives	-0.001***	-0.002***	-0.0001			
	(0.0002)	(0.0003)	(0.0002)			
# of obs.	86,221	86,221	86,221			
Mean of DV	0.0930	0.0930	0.0930			
Policy FE	No	No	Yes			
Year FE	No	Yes	Yes			
County FE	Yes	Yes	Yes			

Discussion

Policy experimentation in China

We examine one of the largest scale of systematic policy experimentation in recent history.

- 1. Experimentation sites are substantially positively selected; misaligned incentives across political hierarchies account for much of the observed positive selection.
- 2. Unrepresentative experimental situation during experimentation: local politicians exert strategic efforts and allocate more resources that may exaggerate policy impacts.
- **3**. Positive selection and experimenter effects not fully accounted for when evaluating experiments, biasing policy learning and national policies.

Policy experimentation in complex, political environment

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- Strong bureaucracy and centralization could facilitate implementation of systematic experimentation.
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Policy experimentation in complex, political environment

- Strong bureaucracy and centralization could facilitate implementation of systematic experimentation.
 - Contrast with under-experimentation under federalism. (Besley & Coate 2003; Mukand & Rodrik 2005; Callander & Harstad 2015)
- Misaligned incentives across political hierarchy could bias learning, inevitable in complex experimentation where bureaucracy needs to be involved.
 - Another form of capture and distortion in policy making (Stigler 1971; Peltzman 1976; Glaeser & Shleifer 2003; Bertrand et al. 2020): policy-making can be captured at the learning stage.
 - Trade-off: inducing efforts to improve policy outcomes, while minimizing experimenter effects that may bias learning. (Vivalt 2020; DellaVigna and Lino 2021)