How Does Made in China 2025 Strategy Influence Green Innovation? A Quasi-Natural Experiment of Chinese Firms

Asif Razzaq

Senior Research Specialist Research Division, CAREC Institute Sep 24, 2025

Reference Paper



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Impact of "Made in China 2025" Industrial Strategy on Firms' Green Innovation: A Quasi-Natural Experiment

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The industrial policy measure "Made in China 2025" significantly promotes green innovation in enterprises.

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Presentation Outline

1. Background

2. Objectives

3. Analytical Framework

4. Empirical Results and Research Direction

Background

2015

Industrial policy is usually defined as addressing structural change of the economy sustained rapid industrialization.

China industrial development is not only due to market-oriented reform and a more open domestic market, but also due to the correct industrial development strategy and the changing industrial policies.

Four stages: system transition period (1978 - 1991), the period of initial establishment of market economy system (1992-2001), the period from China's accession to WTO to the 18th National Congress of the CPC (2001-2012) and the period since the 18th National Congress (2012-2019)

Pilot cities: Local governments. Leading Position in 10 Core Industries Manufactur.. Made in China 2025 (MIC) Core Material Target (70% by 2025) Implemented in **High tech- Industry World Factory** to Technology **Smart Manufacturing** powerhouse Independence from foreign

suppliers

...1 Continued

MIC 2025 in 2015 is highly institutionalized techno industrial policy, which targets key technologies and sectors in the next ten years. The new policy is consistent with the goal of "indigenous innovation"

MIC 2015

Reduce Taxation

Reduce Taxation

R&D for Large Manu.

Hightech Industrial Growth

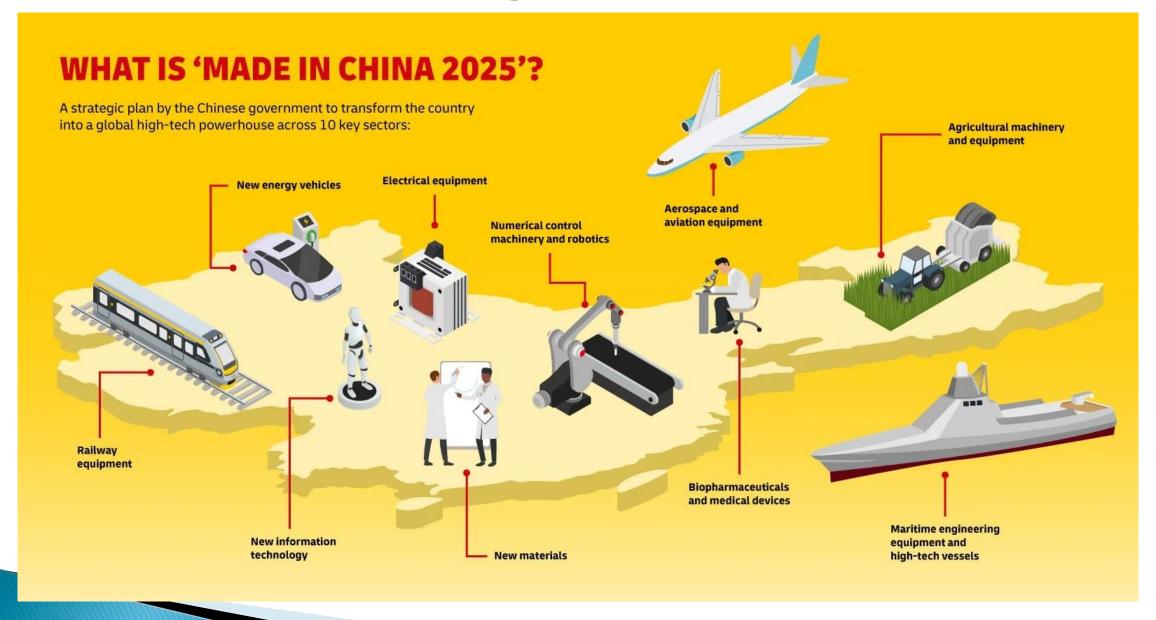
13th (2016-2020) and 14th (2021-2025) Five-year plans

- Innovation driven growth
- low-carbon development
- Reeducation in carbon intensity



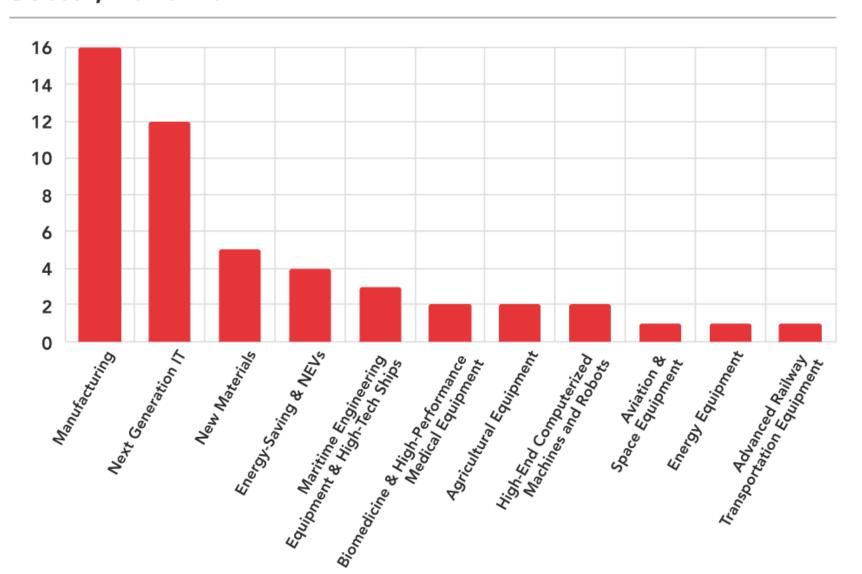
Green industrial transformation and energy transition are two fundamental objectives of 13th and 14th five years plans.

Background



Background

Number of National Level Action and Development Plans by Sector, 2015-2019



China's total expenditure on research and development (R&D) amounted to nearly 3.09 trillion yuan (about 456 billion U.S. dollars, 2.55% of GDP) in 2022, up 10.4 percent year on year, according to the National Bureau of Statistics

^{*77%} to Business,

^{*7%} to Universities

^{*16%} to Government

Objectives

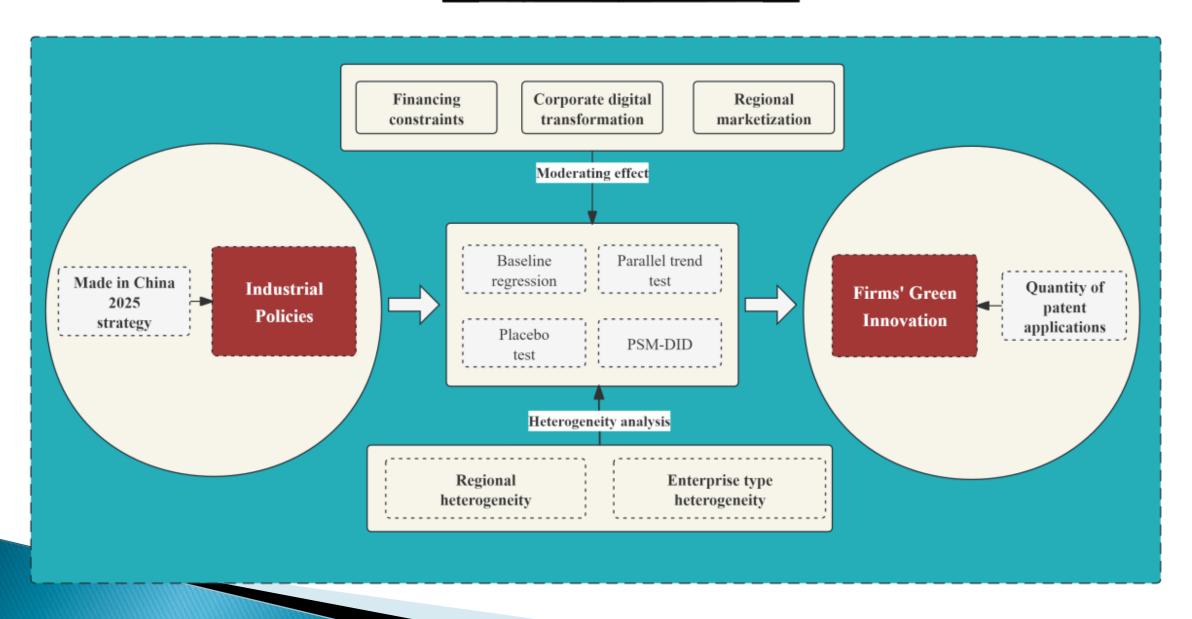
To assess the impact of the "MIC" strategy on firms' green innovation using a natural quasi-experiment.

To examine regional/industrial disparities in the influence of industrial policies on green innovation.

To analyze the role of *financing constraints*, digital transformation, and marketization in moderating the effects of MIC on green innovation

To analyze the other factors of corporate green innovation and lesson for peer countries

Analytical Framework



Analytical Framework: Sample

Cathay Pacific (CSMAR) data set explored Unbalanced panel from 2000 to 2021

Companies in the financial category are excluded Samples of companies:

- -ST, *ST, or PT during the observation period are excluded
- -Companies that are missing the main variables are excluded

1,594 firms are selected for the study 815 treated and 779 control group Observations ranged from 8850 to 10538

Analytical Framework: Sample Sample Industry Catagories

In dustant Classification	# of total	# treat	# control
Industry Classification	firms	group	group
A. Agriculture and Livestock	19	0	19
B. Mining	62	0	62
C.Manufacturing	966	815	151
D. Energy and water production and supply	70	0	70
E. Construction	45	0	45
F. Wholesale and retail trade	90	0	90
G. Transportation, storage and postal services	64	0	64
H. Accommodation and catering	6	0	6
I. Information transmission, software and information		0	
technology services	82		82
K.Real Estate	72	0	72
L. Rental and business services	27	0	27
N. Water conservancy, environment and public facilities		0	
management industry	16		16
O. Residential Services, Repair and Other Services	19	0	19
P. Education	2	0	2
Q.Health and social work	1	0	1
R.Culture, sports and recreation	8	0	8
S. Public administration, Social security and social			
organizations	45	0	45
Total	1594	815	779

Analytical Framework: Sample Treat Group Sectors

Industry Code	Industry Name	Number of enterprises
C26	Chemical raw materials and chemical products manufacturing	106
C27	Pharmaceutical Manufacturing	103
C28	Chemical fiber manufacturing	15
C29	Rubber and plastic products industry	20
C30	Non-metallic mineral products industry	40
C31	Ferrous metal smelting and rolling processing industry	25
C32	Non-ferrous metal smelting and rolling processing industry	43
C33	Metal Products Industry	20
C34	General Equipment Manufacturing	43
C35	Specialty Equipment Manufacturing	75
C36	Automobile Manufacturing	45
C37	Railroad, ship, aerospace and other transportation equipment manufacturing industry	30
C38	Electrical machinery and equipment manufacturing	95
C39	Computer, communications and other electronic equipment manufacturing	140
C40	Instrumentation Manufacturing	15
	Total	815

Analytical Framework: Sample 10 key sectors

Name of key sectors involved	Industry codes involved	Number of enterprises
Railroad Equipment	C37+C40	45
New Energy Vehicles	C36	45
New Information Technology	C39	140
Electrical Equipment	C38	95
New Materials	C26+C28+C29+C30+C31+C32+C33	269
CNC Machinery and Robotics	C35	75
Biopharmaceuticals and Medical Devices	C27	103
 Aerospace Equipment 	`-	-
Agricultural machinery and equipment	C34	43
Total		815

Analytical Framework: Model

Foundational model is as follows:

$$CGI_{it} = \alpha + \beta treat_i \times post_t + \gamma X_{it} + \delta_t + \mu_i + ind_k + \varepsilon_{it}$$
 (1)

The explanatory variable CGI_{it} is the green innovation level of enterprise i in year t. $treat_i$ is a dummy variable for the experimental group, which takes 1 if enterprise i is a pilot enterprise and 0 otherwise. $post_t$ is a dummy variable for the experimental period, which takes 1 in 2015 and after, and 0 before 2015.

 β is the parameter of interest in this paper, whose sign and value represent the direction and magnitude of the impact of implementing the "Made in China 2025" industrial policy on corporate green innovation. β is the parameter of interest in this paper, whose sign and value represent the direction and size of the impact of implementing the "Made in China 2025" industrial policy on enterprises' green innovation. X_{it} is the control variable included in this paper. Meanwhile, this paper controls for the time-fixed effect (δ_t) , individual fixed effect (u_i) , and industry-fixed effect (ind_k) , and ε_{it} is the random error term.

Analytical Framework: Variables

Dependent variable: Corporate green innovation (CGI) is the # of enterprise green development patent applications (approved)

Independent variable: Employs group dummy variables and staged dummy variables within the framework of a quasi-natural experiment, based on the initiation of the "Made in China 2025" industrial policy in 2015.

To establish the experimental and control groups, grouping dummy variables are employed, denoted as "Treat." Companies falling within MIC focal areas are classified as the experimental group, represented by a value of 1 for Treat. Conversely, companies outside these areas are assigned to the control group, with a Treat value of 0.

The staging dummy variable is denoted as "Post." A year subsequent to 2015 indicates the implementation phase of the strategy, assigning a value of 1 to Post. Conversely, a year preceding the implementation denotes a value of 0 for Post.

Analytical Framework: Variables

Control variables

To avoid estimation bias caused by omitted variables, the following control variables are introduced concerning the studies of scholars such as (Pan et al., 2022; Xu et al., 2023): firm size (Ta), expressed in terms of total assets, number of employees in the firm (Ne), total profits of the firm (Tp), turnover of total assets (Ato), net profit margin of total assets (Roa), return on net assets (Roe), Tobin's Q (Tbq), and the percentage of shares held by the first largest shareholder (Sha).

Analytical Framework: Variables

Mechanism Variables:

Financing constraints (SA): The present study selects the exogenous SA index (Hadlock & Pierce, 2010) as the preferred measure for capturing enterprises' financing constraints. Below equation demonstrates the calculation method employed, where SA denotes the firm's financing constraint, Size represents the logarithm of the firm's assets, and Age indicates the number of years the firm has been listed.

$$SA=-0.737\times Size+0.043\times Size^2-0.04\times Age$$

Corporate digital transformation : Following Wu et al. (2022), it analyse word frequency statistics encompassing the term "enterprise digital transformation" found in the annual reports of listed companies using web crawling.

Regional marketization: Adopts the relevant index compiled by Fan et al. (2011) as a metric for assessing the level of marketization. The selected index serves as a representative measure of the quality of the external governance environment, with larger values indicating a more favorable regional market environment.

Analytical Framework: Method

MIC industrial policy focuses on industries with well-defined boundaries, and the interval between its introduction and implementation is relatively short, thereby serving as an exogenous policy shock to enterprises and creating favorable conditions for employing the Difference-in-Differences (DID) model.

D-DID is an advanced econometric technique used to estimate the causal effect of a treatment or intervention when multiple groups are exposed to the treatment at different times. It builds upon the traditional Difference-in-Difference (DID) approach by introducing an additional time dimension or treatment group, allowing for more complex comparisons.

The study will employ a **double-difference-in-difference (D-DID)** model to estimate the impact of industrial policies on firms' green innovation.

Parallel trend assumption: The difference in the means of the explanatory variables between the experimental and control groups remains consistent across time. It requires that the time trends of the explanatory variables in both groups exhibit a consistent pattern.

Empirical Results

Descriptive Statistics

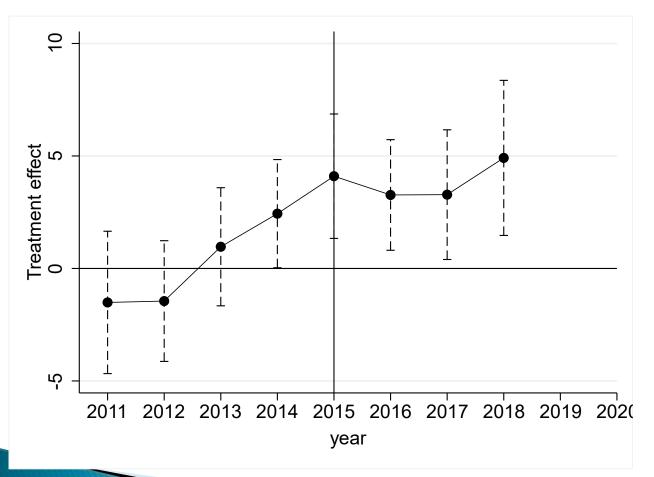
Classification	Variable	Mean	Std. Dev.	Min	Max
Dependent variable	CGI	3.80	25.46	0.00	996
Independent	treat	0.52	0.50	0.00	1.00
variables	post	0.69	0.46	0.00	1.00
	Та	22.30	1.67	16.52	31.31
	Ne	7.76	1.44	2.08	13.13
	Тр	15.33	105.37	-244.97	4248.99
Control variables	Ato	0.62	0.52	0.00	8.25
Control variables	Roe	0.06	3.00	-53.96	281.99
	Roa	0.03	0.15	-3.99	2.64
	Tbq	2.05	3.29	0.68	122.19
	Sha	36.06	16.21	0.39	98.86
Mechanism	SA	-3.80	0.30	-5.32	-2.11
	DT	10.45	28.46	0.00	544
variables	MAR	9.088	1.74	-0.16	11.49

Empirical Results- Baseline Regression

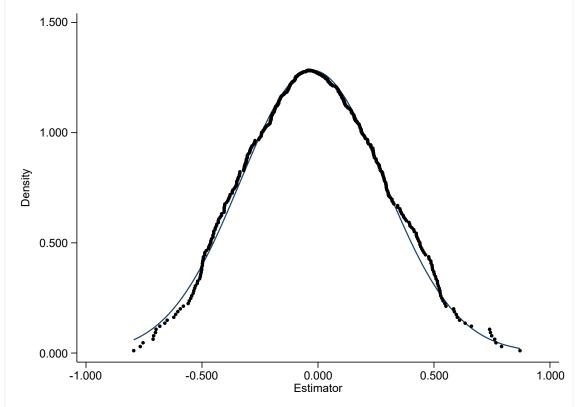
	(1)	(2)	(3)	(4)
	CGI	CGI	CGI	CGI
post Xtreat	4.613***	4.634***	0.915**	1.168*
	(0.527)	(0.529)	(0.430)	(0.693)
Ta		1.008***	-0.159	-0.159
		(0.245)	(0.226)	(0.240)
Ne		0.919***	0.142	0.153
		(0.271)	(0.205)	(0.221)
Tp		0.003	0.004	0.004
		(0.003)	(0.003)	(0.003)
Ato		-0.430	-0.437	-0.445
		(0.525)	(0.355)	(0.365)
Roe		0.077	-0.043	-0.042
		(0.283)	(0.042)	(0.043)
Roa		-0.987	1.404	1.344
		(3.341)	(1.693)	(1.718)
Tbq		0.054	0.027	0.028
		(0.095)	(0.025)	(0.026)
Sha		-0.016	0.027	0.029
		(0.017)	(0.021)	(0.022)
Constant	1.704***	-27.275***	5.070	4.838
	(0.365)	(4.497)	(4.310)	(4.599)
Firm/Year	No	No	YES	YES
irm/Ind/Year	No	No	No	YES
R-squared	0.008	0.019	0.669	0.667

Empirical Results-Prerequisite Tests

Parallel Trend Test



Placebo Test



Empirical Results-Prerequisite Tests

Counterfactual testing

	(1)	(2)
	CGI	CGI
post×treat_2012	0.824	0.913
	(0.700)	(0.701)
Control	No	YES
Firm/Ind/Year effect	YES	YES
Constant	3.564***	4.992
	(0.363)	(4.600)
Observations	9514	9514
R -squared	0.667	0.667

Empirical Results-DID Model

Average treatment effect estimation results

Variable	ATT	Standard error	t-value
CGI	5.518***	0.413	9.25

Endophytic treatments result

	(1)	(2)
	Psm-DID	Psm-DID
post ×treat	3.343***	3.417***
	(0.867)	(0.871)
Control	No	YES
Firm/Ind/Year effect	YES	YES
Constant	2.731***	3.707
	(0.355)	(4.601)
Observations	9508	9508
R -squared	0.668	0.668

Empirical Results-Robustness Tests

Robustness and endogeneity tests

VARIABLES	(1)	(2)	(3)
post×treat	3.415***	3.822***	3.795***
	(0.868)	(0.872)	(0.875)
control variable	Yes	Yes	Yes
Constant	3.710	3.352	3.300
	(4.591)	(4.641)	(4.672)
Firm/Ind/Year effect	Yes	Yes	Yes
Firm/Ind/Year/City effect	No	Yes	No
R-squared	0.668	0.669	0.669

- Beijing and Shanghai are excluded (Column 1)
- City-level fixed effects are incorporated as controls (Column 2)
- Additional control variables are included (Column 3)

Empirical Results-Heterogeneity Results

External environment heterogeneity results

	(1)	(2)
	Eastern Region	Central and Western Regions
post ×treat	1.748*	0.346
	(1.040)	(0.278)
Fisher's permutation test		-1.402***
Control	YES	YES
Firm/Ind/Year effect	YES	YES
Constant	1.224	18.514**
	(5.828)	(7.714)
Observations	6551	2957
R -squared	0.677	0.538

Empirical Results-Heterogeneity Results

External environment heterogeneity results

	(1)	(2)
	Lightly polluting enterprises	Heavily polluting enterprises
post×treat	2.381*	-0.004
	(1.304)	(0.298)
Fisher's permutation test	2.38	85***
Control	YES	YES
Firm/Ind/Year effect	YES	YES
Constant	7.771	4.795
	(5.084)	(10.352)
Observations	6430	3064
R -squared	0.649	0.789

Empirical Results-Mechanism

VARIABLES	(1)	(2)	(3)
post ×treat	1.891**	-0.055	1.059
	(0.957)	(0.711)	(0.679)
post Xtreat XSA	4.516**		
	(2.267)		
SA .	13.837***		
	(4.019)		
post ×treat ×DT		0.129**	
		(0.052)	
DT		0.070***	
		(0.025)	
oost ×treat ×MAR			0.692*
			(0.400)
MAR			-0.516
			(0.328)
control variable	YES	YES	YES
Constant	57.576***	6.025	9.378*
	(15.378)	(5.036)	(5.375)
irm/Ind/Year effect	YES	YES	YES
R-squared	0.668	0.670	0.667

Results and Explanations

Key Take Aways

- MIC Contributes to corporate green innovation, however, its marginal impact varies at industrial level, regional disparity and nature of firms.
- MIC influence firm's green innovation through removing financing constraints, digital transformation, and increasing regional marketization.

- Prior literature produced mixed outcomes, mainly they studied the MIC impact on firm's performance, productivity, overall R&D input intensity, and financing availability, however no conclusive outcome has been drawn due to various reasons.
- So far, no noticeable study explore its impact on green innovation through unique mechanism

Future consideration

Further Consideration

• This working paper later would be extended considering foreign spillovers, especially inward FDI spillovers considering specific firms and industries, rather than taking all sectors

• Single firm/industry level case study could be explored to navigate the direct benefits received form MIC 2025 strategy. A comparative analysis of different industrial initiatives could also be considered.

Recommendations

Recommendation s

• Enhance Industrial Policy Quality: Shift focus from scale and quantity to improving the quality and productivity of industries. Increase support for advanced manufacturing and enhance resource allocation efficiency to boost green innovation levels.

• Establish a Green Innovation System: Develop a comprehensive innovation framework that promotes green innovation. Support enterprises with financial assistance, innovation incentives, and tax benefits to enhance their green competitiveness and sustain industry growth.

Recommendations

Recommendation s

• Tailor Regional Policies: Create region-specific industrial policies considering regional disparities and pollution levels. Offer greater support and incentives to enterprises in the eastern regions and those with lower pollution to accelerate their green innovation efforts.

• Reduce Financing Constraints: Establish a green finance fund and implement green credit policies to ease financing for green innovation. Encourage digital transformation in enterprises by providing support and training, and strengthen market competition mechanisms to ensure fair practices.

