# Reassessing China's Rural Reforms: The View from Outer Space

Joel Ferguson<sup>1</sup> Oliver Kim<sup>2</sup>

<sup>1</sup>Stanford University

<sup>2</sup>Open Philanthropy

Sep 5, 2024

➤ Since 1980, China has lifted 800 million people out of extreme poverty, accounting for 75% of global extreme poverty reduction and 23% of global GDP growth

- ➤ Since 1980, China has lifted 800 million people out of extreme poverty, accounting for 75% of global extreme poverty reduction and 23% of global GDP growth
- Economic reforms under Deng Xiaoping typically credited with the "Chinese miracle"

- ➤ Since 1980, China has lifted 800 million people out of extreme poverty, accounting for 75% of global extreme poverty reduction and 23% of global GDP growth
- Economic reforms under Deng Xiaoping typically credited with the "Chinese miracle"
- One of the central reforms was the post-1978 Household Responsibility System (HRS), which moved agriculture away from collective farming

- ➤ Since 1980, China has lifted 800 million people out of extreme poverty, accounting for 75% of global extreme poverty reduction and 23% of global GDP growth
- Economic reforms under Deng Xiaoping typically credited with the "Chinese miracle"
- One of the central reforms was the post-1978 Household Responsibility System (HRS), which moved agriculture away from collective farming
- Conventional view is that HRS was key to China's economic takeoff—Lin (1992) argues it caused half of agricultural output growth from 1978-84, when yields rose 40%

- ➤ Since 1980, China has lifted 800 million people out of extreme poverty, accounting for 75% of global extreme poverty reduction and 23% of global GDP growth
- Economic reforms under Deng Xiaoping typically credited with the "Chinese miracle"
- One of the central reforms was the post-1978 Household Responsibility System (HRS), which moved agriculture away from collective farming
- Conventional view is that HRS was key to China's economic takeoff—Lin (1992) argues it caused half of agricultural output growth from 1978-84, when yields rose 40%
- Perceived success of HRS led to other liberalizing reforms, kickstarting China's miracle

▶ But how much can we believe the conventional wisdom about the HRS?

- ▶ But how much can we believe the conventional wisdom about the HRS?
- Modern causal inference work on China's reform period is surprisingly rare—major statistical constraint is the lack of disaggregated economic data

- ▶ But how much can we believe the conventional wisdom about the HRS?
- ► Modern causal inference work on China's reform period is surprisingly rare—major statistical constraint is the lack of disaggregated economic data
- ► Even worse, Chinese economic statistics are and have been notoriously unreliable—can reflect political priorities as much as the ground truth

- ▶ But how much can we believe the conventional wisdom about the HRS?
- ► Modern causal inference work on China's reform period is surprisingly rare—major statistical constraint is the lack of disaggregated economic data
- Even worse, Chinese economic statistics are and have been notoriously unreliable—can reflect political priorities as much as the ground truth
  - Statistical misreporting by cadres partly caused Great Leap Forward famine—killing 15-55 million people (Yang 2013)

- ▶ But how much can we believe the conventional wisdom about the HRS?
- ► Modern causal inference work on China's reform period is surprisingly rare—major statistical constraint is the lack of disaggregated economic data
- Even worse, Chinese economic statistics are and have been notoriously unreliable—can reflect political priorities as much as the ground truth
  - Statistical misreporting by cadres partly caused Great Leap Forward famine—killing 15-55 million people (Yang 2013)
  - Amid China's recent slowdown, clampdown on publication of inconvenient statistics

- ▶ But how much can we believe the conventional wisdom about the HRS?
- Modern causal inference work on China's reform period is surprisingly rare—major statistical constraint is the lack of disaggregated economic data
- Even worse, Chinese economic statistics are and have been notoriously unreliable—can reflect political priorities as much as the ground truth
  - Statistical misreporting by cadres partly caused Great Leap Forward famine—killing 15-55 million people (Yang 2013)
  - Amid China's recent slowdown, clampdown on publication of inconvenient statistics
- Without good data, how can we understand the causes of the Chinese miracle?

- But how much can we believe the conventional wisdom about the HRS?
- ► Modern causal inference work on China's reform period is surprisingly rare—major statistical constraint is the lack of disaggregated economic data
- Even worse, Chinese economic statistics are and have been notoriously unreliable—can reflect political priorities as much as the ground truth
  - Statistical misreporting by cadres partly caused Great Leap Forward famine—killing 15-55 million people (Yang 2013)
  - ▶ Amid China's recent slowdown, clampdown on publication of inconvenient statistics
- Without good data, how can we understand the causes of the Chinese miracle?
- ► This project: Use historical satellites to measure what was actually going on on the ground, creating our own data to assess the effects of the HRS

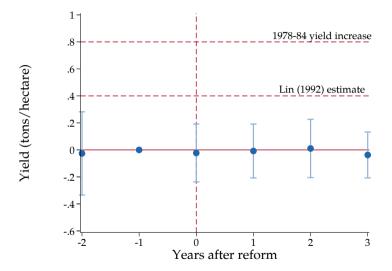
#### This Paper's Contributions

- 1. Create new measures of agricultural productivity, independent of the official Chinese data
  - > Apply machine learning methods to estimate yields (output over land area) from satellite imagery

### This Paper's Contributions

- 1. Create new measures of agricultural productivity, independent of the official Chinese data
  - Apply machine learning methods to estimate yields (output over land area) from satellite imagery
- 2. Identify the causal effects of the HRS using two different identification strategies:
  - A novel staggered differences-in-discontinuities design that identifies the effect at province boundaries
  - ► A staggered rollout design using county gazetteers from Almond et al (2019)

#### Preview: we find no evidence that the HRS increased yields in China



### This Paper's Contributions

- 1. Create new measures of agricultural productivity, independent of the official Chinese data
  - Apply machine learning methods to estimate yields (output over land area) from satellite imagery
- 2. Identify the causal effects of the HRS using two different identification strategies:
  - A novel staggered differences-in-discontinuities design that identifies the effect at province boundaries
  - ► A staggered rollout design using county gazetteers from Almond et al (2019)
- Main Finding: We do not find statistically significant evidence that decollectivization increase agricultural productivity
  - However, our satellite data indicates that aggregate yields still increased
  - Points to other policy reforms—most likely a major procurement price reform—as the main factor in China's agricultural takeoff

### Where This Paper Fits In

- Understanding the drivers of China's miracle
  - ► HRS: McMillan et al (1988), Lin (1992), Almond et al. (2019), Chen and Lan (2020)
  - Importance of agriculture in growth accounting: Young (2003), Brandt et al. (2008)
  - This paper: revises our understanding of one of the most important reforms in China

### Where This Paper Fits In

- Understanding the drivers of China's miracle
  - ► HRS: McMillan et al (1988), Lin (1992), Almond et al. (2019), Chen and Lan (2020)
  - Importance of agriculture in growth accounting: Young (2003), Brandt et al. (2008)
  - ▶ This paper: revises our understanding of one of the most important reforms in China
- ► The effects of land reform and farm size on agricultural productivity
  - Household farming is more efficient: Vollrath (2007), Kagin et al (2016)
  - Scale economies favor large farms: Foster and Rosenzweig (2017), Adamopoulos and Restuccia (2020), Kim and Wang (2024)
  - This paper: new evidence on the largest land reform/land privatization in history

### Where This Paper Fits In

- Understanding the drivers of China's miracle
  - ► HRS: McMillan et al (1988), Lin (1992), Almond et al. (2019), Chen and Lan (2020)
  - Importance of agriculture in growth accounting: Young (2003), Brandt et al. (2008)
  - ▶ This paper: revises our understanding of one of the most important reforms in China
- ► The effects of land reform and farm size on agricultural productivity
  - ▶ Household farming is more efficient: Vollrath (2007), Kagin et al (2016)
  - Scale economies favor large farms: Foster and Rosenzweig (2017), Adamopoulos and Restuccia (2020), Kim and Wang (2024)
  - This paper: new evidence on the largest land reform/land privatization in history
- Growing intersection between remote sensing and economics
  - Nighlights to measure growth in autocracies: Henderson et al (2012), Hodler and Raschky (2014), Martinez (2022)
  - Daytime + ML to measure poverty: Jean et al (2016), Yeh et al (2020), Huang et al (2021)
  - ▶ This paper: novel application of older satellites to economic history, with potential applications to other settings without reliable data

# Historical Background

#### What was Household Responsibility? Institutional Details



#### Pre-Reform

- Households organized into work teams of 20 - 30
- ► Households receive basic grain ration based on the number of family members. Work points determine an additional cash reward if there's a surplus (but these are rare)
- ► Team sells all grain to the state, for a fixed procurement price

#### What was Household Responsibility? Institutional Details



#### Pre-Reform

- Households organized into work teams of 20 - 30
- Households receive basic grain ration based on the number of family members. Work points determine an additional cash reward if there's a surplus (but these are rare)
- ► Team sells all grain to the state, for a fixed procurement price

#### Post-Reform

- Households keep surplus after fulfilling quota and paying agricultural taxes
- Two main varieties: baochan daohu. collective retains decision-making over cropping: baogan daohu, households get full responsibility over production
- In effect, a fixed-rent system where the landlord is the state—household becomes residual claimant on output

► Mao's death in 1976 created an opening for reform

- Mao's death in 1976 created an opening for reform
- Starting in Anhui Province in 1978, communes began trying responsibility systems—households free to sell their output after meeting the state quota

- Mao's death in 1976 created an opening for reform
- Starting in Anhui Province in 1978, communes began trying responsibility systems—households free to sell their output after meeting the state quota
- ▶ Pattern of "regional experiments" (Qian, Roland, and Xu 2006): provinces varied in their rates of reform, with provincial leaders in setting the pace (Teiwes and Sun 2016)

- ► Mao's death in 1976 created an opening for reform
- Starting in Anhui Province in 1978, communes began trying responsibility systems—households free to sell their output after meeting the state quota
- Pattern of "regional experiments" (Qian, Roland, and Xu 2006): provinces varied in their rates of reform, with provincial leaders in setting the pace (Teiwes and Sun 2016)
- Paradigm is Anhui's Wan Li: drought in 1978 spurred peasant experiments with household responsibility; unlike elsewhere, Wan permitted, then encouraged reform

- ► Mao's death in 1976 created an opening for reform
- Starting in Anhui Province in 1978, communes began trying responsibility systems—households free to sell their output after meeting the state quota
- Pattern of "regional experiments" (Qian, Roland, and Xu 2006): provinces varied in their rates of reform, with provincial leaders in setting the pace (Teiwes and Sun 2016)
- Paradigm is Anhui's Wan Li: drought in 1978 spurred peasant experiments with household responsibility; unlike elsewhere, Wan permitted, then encouraged reform
- Anhui, Sichuan, Guizhou were early leaders; after 1982, when HRS became national policy, Jiangsu, Heilongjiang were notable laggards

# When did provinces have 50% of work teams adopt HRS?





# Data and Measurement

#### Remotely Sensed Data

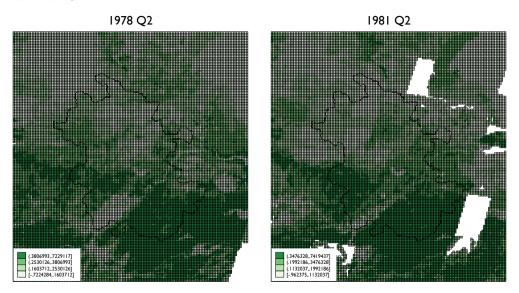
- Satellite sources: Advanced Very High Resolution Radiometer (AVHRR), 1978-2013
  - Roughly 1km resolution, twice (!) every day
  - Use red and near-infrared bands to construct the Normalized Difference Vegetation Index:

$$NDVI = \frac{NIR - Red}{NIR + Red}$$

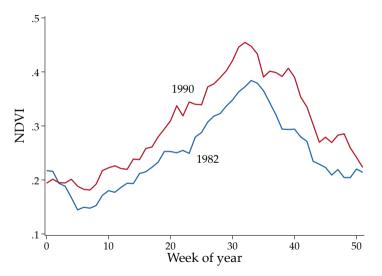
- Large literature in remote sensing + environmental science showing NDVI strongly predicts yields (agricultural output over area)<sup>1</sup>
- ▶ Our data: aggregate up to 0.5-degree grid-cells (roughly 5km x 5km at Anhui's latitude)

<sup>&</sup>lt;sup>1</sup>Sharma et al (1993), Hamar et al (1996), Qiu et al. (2003), Moriondo, Maselli, and Bindi (2007), Liu et al (2013), Vittek et al (2014), Wang et al (2014), Lobell, et al. (2015), Burke and Lobell (2017), Kasampalis et al (2018), Cao et al (2020), Bognar et al (2022)

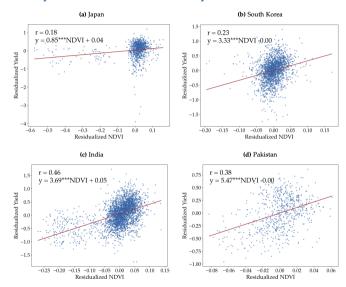
#### NDVI in Anhui



# NDVI in China: Seasonal Cycles



#### Peak NDVI predicts yield across a variety of contexts



#### Predicting Yields with Random Forests

► Can we do better than peak NDVI? Yields unlikely just to be just a linear function of peak NDVI; we're throwing out a lot of information

#### Predicting Yields with Random Forests

- ► Can we do better than peak NDVI? Yields unlikely just to be just a linear function of peak NDVI; we're throwing out a lot of information
- But if we use a more sophisticated model, we don't want to use Chinese data to train, which may re-introduce bias

### Predicting Yields with Random Forests

- Can we do better than peak NDVI? Yields unlikely just to be just a linear function of peak NDVI; we're throwing out a lot of information
- But if we use a more sophisticated model, we don't want to use Chinese data to train, which may re-introduce bias
- Solution: train a random forest to predict yields from satellite data of the previously shown neighboring countries, use that model to predict

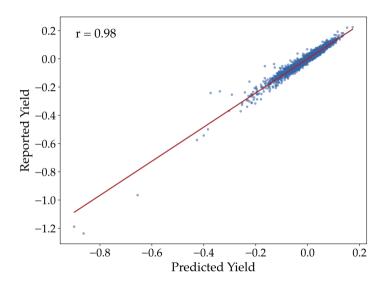
### Predicting Yields with Random Forests

- ► Can we do better than peak NDVI? Yields unlikely just to be just a linear function of peak NDVI; we're throwing out a lot of information
- But if we use a more sophisticated model, we don't want to use Chinese data to train, which may re-introduce bias
- Solution: train a random forest to predict yields from satellite data of the previously shown neighboring countries, use that model to predict
- Random forests have been successfully used to predict agricultural yields in a wide variety of contexts (Jeong et al. 2016; Cao et al. 2020; Marques Ramos et al. 2020)

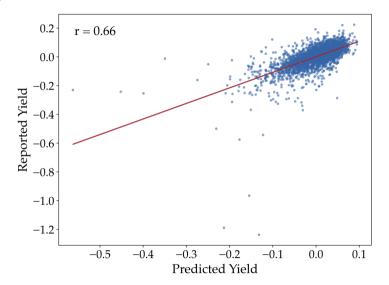
### Training the Random Forest

- ► Training Data: Pooled sample of sub-national data from neighboring countries with similar agricultural mix: Japan, South Korea, India, Pakistan
- Use harmonic regression to smooth NDVI, take deciles within each geographic unit—creates 160 features for each yearly observation
- **Prediction:** creates yearly observations of predicted grain yield  $\hat{y}$
- ➤ Validate both in-sample and using 5-fold cross validation (hold out 1/5 of data in training, validate against that 1/5, repeat over all 5 holdout "folds")

# In-Sample Fit



# Out-of-Sample Fit (5-fold cross validation)



### One Last Test

Are our satellite-based measures sensitive enough to predict yield changes within China from known exogenous shocks?

#### One Last Test

- Are our satellite-based measures sensitive enough to predict yield changes within China from known exogenous shocks?
- Estimate the effect of temperature shocks on our predicted yields in China:

	(1)	(2)
>1SD temp. shock in growing months	0.0847** (0.0295)	
<-1SD temp. shock in growing months		-0.0728***
		(0.0152)
Observations	3768602	3768602
$R^2$	0.474	0.472

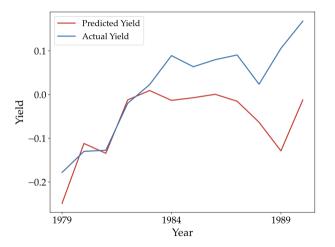
#### One Last Test

- Are our satellite-based measures sensitive enough to predict yield changes within China from known exogenous shocks?
- Estimate the effect of temperature shocks on our predicted yields in China:

	(1)	(2)
>1SD temp. shock in growing months	0.0847** (0.0295)	
<-1SD temp. shock in growing months		-0.0728***
		(0.0152)
Observations	3768602	3768602
$R^2$	0.474	0.472

 $\blacktriangleright$  Elasticities consistent with climate literature—indicate we can detect  $\approx$  5% changes in yields

# Aggregate yields did appear to increase throughout the late 1970s



# **Empirical Results**

▶ We use differential timing of HRS across provinces to identify its effects

- ▶ We use differential timing of HRS across provinces to identify its effects
- ► Treatment Variable: indicator  $D_i = 1$  when more than 50% of province i's work teams adopt HRS Map Event Study

- We use differential timing of HRS across provinces to identify its effects
- ► Treatment Variable: indicator  $D_i = 1$  when more than 50% of province i's work teams adopt HRS Map Event Study
- ► Selection concern: HRS was intended as emergency relief—places that adopted earlier were poorer and at risk of famine. Perhaps growth just captures mean reversion?

- We use differential timing of HRS across provinces to identify its effects
- ► Treatment Variable: indicator  $D_i = 1$  when more than 50% of province i's work teams adopt HRS Map Event Study
- ► Selection concern: HRS was intended as emergency relief—places that adopted earlier were poorer and at risk of famine. Perhaps growth just captures mean reversion?
- ➤ Solution: Compare cells just on either side of provincial borders (geographic regression discontinuity), include cell-level fixed effects, to control for fixed differences of places (difference-in-discontinuities)

### **Estimating Equation**

Formally, for cell *i* in border group *b* in year *t*, at horizon *h*, we estimate

$$y_{i,b,t+h} - y_{i,b,t-1} = \underbrace{\beta_h \Delta D_{i,b,t}}_{\text{Pooled treatment effect}} + \underbrace{\gamma_{b,t+h} (R_i \times B_{b,t})}_{\text{Control province fit}} + \underbrace{\delta_{b,t+h} (R_i \times B_{b,t} \times \Delta D_{i,b,t})}_{\text{Treated province fit}} + \underbrace{B_{b,t}}_{\text{border-year FE}} + e_{i,b,t}^h$$

### Estimating Equation

Formally, for cell *i* in border group *b* in year *t*, at horizon *h*, we estimate

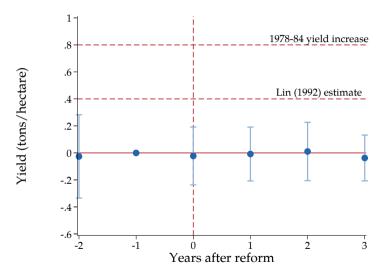
$$y_{i,b,t+h} - y_{i,b,t-1} = \underbrace{\frac{\beta_h \Delta D_{i,b,t}}{Pooled \ treatment \ effect}}_{Pooled \ treatment \ effect} + \underbrace{\frac{\gamma_{b,t+h} (R_i \times B_{b,t})}{Control \ province \ fit}}_{Control \ province \ fit} + \underbrace{\frac{\delta_{b,t+h} (R_i \times B_{b,t} \times \Delta D_{i,b,t})}{Treated \ province \ fit}}_{Double \ treatment} + \underbrace{\frac{\beta_{b,t}}{\beta_{b,t}}}_{Double \ treatment} + e_{i,b,t}^h.$$

To prevent negative weighting, for each time period *t* we restrict the sample to just-treated units and "clean control" (or never-treated) units:

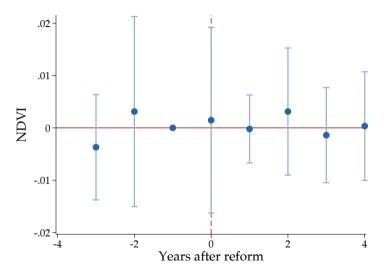
$$\begin{cases} \text{newly treated:} & \Delta D_{i,t} = 1 \\ \text{or clean control:} & D_{i,t+h} = 0 \end{cases}$$

SEs computed using a 2-stage bootstrap: (1) bootstrapping estimated relationship for yields (Sexton and Laake 2009), then plugging in predicted yield estimates into RD; (2) bootstrapping RD by border clusters (B=73)

# Effect of HRS adoption on estimated yield



# Effect of HRS adoption on Max NDVI TWEE VERSION



# Taking Stock

- Initial results very surprising—we observe a null effect of the HRS at provincial boundaries
- Does this reflect a quirk of the identification strategy? Are border areas unrepresentative in some way?
- ▶ (If time): alternative identification strategy, using county-level rollout data from gazetteers, collected by Almond et al. (2019), finds a similar answer

Can run alternate design looking at means of treated areas

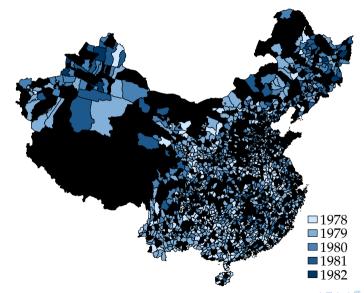
- Can run alternate design looking at means of treated areas
- ▶ Use alternate source of treatment variation from Almond et al (2019), who collect gazetteers from 914 counties across China, compiled by local officials for historical purposes

- Can run alternate design looking at means of treated areas
- ▶ Use alternate source of treatment variation from Almond et al (2019), who collect gazetteers from 914 counties across China, compiled by local officials for historical purposes
- They code a reform event  $(D_{i,t} = 1)$  as "the year when collectively owned land was first contracted to individual households in a few villages for each county; it usually took 2–3 years for it to spread to the entire county."

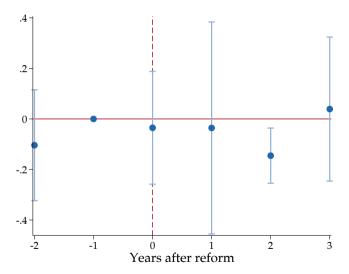
- Can run alternate design looking at means of treated areas
- ▶ Use alternate source of treatment variation from Almond et al (2019), who collect gazetteers from 914 counties across China, compiled by local officials for historical purposes
- ▶ They code a reform event  $(D_{i,t} = 1)$  as "the year when collectively owned land was first contracted to individual households in a few villages for each county; it usually took 2–3 years for it to spread to the entire county."
- Estimate similar Dube et al (2023) staggered rollout: for county *i* at year *t* at horizon *h* after reform,

$$y_{i,t+h} - y_{i,t-1} = \beta \Delta D_{i,t} + \delta_t^h + \varepsilon_{i,t}^h$$

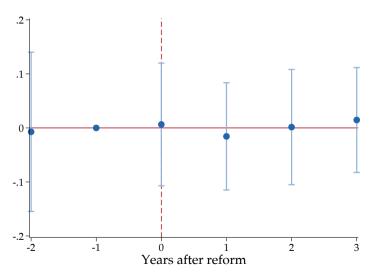
# County Reform Dates in Almond et al (2019)



# NDVI after Almond (2019) county decollectivization



# Predicted yield after Almond (2019) county decollectivization



► Two different sources of variation—the same surprising answer. What else could have raised yields?

- Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops

- Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops
  - Even in the official macroeconomic data, yields started rising from 1978-1982—but land reform wasn't implemented nationally until 1983

- Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops
  - Even in the official macroeconomic data, yields started rising from 1978-1982—but land reform wasn't implemented nationally until 1983
  - Lin estimates that around 16% of output growth can be attributed to procurement price rises

- ► Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops
  - Even in the official macroeconomic data, yields started rising from 1978-1982—but land reform wasn't implemented nationally until 1983
  - Lin estimates that around 16% of output growth can be attributed to procurement price rises
- Qualitative Evidence: skepticism around governments' claims around HRS:

- ► Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops
  - Even in the official macroeconomic data, yields started rising from 1978-1982—but land reform wasn't implemented nationally until 1983
  - Lin estimates that around 16% of output growth can be attributed to procurement price rises
- Qualitative Evidence: skepticism around governments' claims around HRS:
  - ▶ A 1983 CIA report attributes yield growth instead to "good weather, increased use of fertilizer and equipment, and other related reforms, such as a more rational state pricing system"

- ► Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops
  - Even in the official macroeconomic data, yields started rising from 1978-1982—but land reform wasn't implemented nationally until 1983
  - Lin estimates that around 16% of output growth can be attributed to procurement price rises
- Qualitative Evidence: skepticism around governments' claims around HRS:
  - ▶ A 1983 CIA report attributes yield growth instead to "good weather, increased use of fertilizer and equipment, and other related reforms, such as a more rational state pricing system"
  - Bramall (2003) notes that national agricultural growth accelerated in 1976-80; points out that early decollectivizers did not grow faster than late decollectivizers; points to procurement price reform instead

- ► Two different sources of variation—the same surprising answer. What else could have raised yields?
  - In 1979, the state raised procurement prices for grain, oil crops, cotton, and other crops
  - Even in the official macroeconomic data, yields started rising from 1978-1982—but land reform wasn't implemented nationally until 1983
  - Lin estimates that around 16% of output growth can be attributed to procurement price rises
- Qualitative Evidence: skepticism around governments' claims around HRS:
  - ▶ A 1983 CIA report attributes yield growth instead to "good weather, increased use of fertilizer and equipment, and other related reforms, such as a more rational state pricing system"
  - Bramall (2003) notes that national agricultural growth accelerated in 1976-80; points out that early decollectivizers did not grow faster than late decollectivizers; points to procurement price reform instead
- ► Highlights the need for careful causal identification—contribution of this paper

### **Broader Takeaways**

- ► We should treat broad claims founded on data from autocratic regimes with skepticism—particularly when there are political incentives to misreport
- Satellite data presents an invaluable window in the past for big questions in growth and development
- ▶ We have presented evidence that suggests we need to revise our view of one of the central causes of China's economic takeoff
- Future work will continue to explore the policies behind the Chinese Miracle

# Thank you!

oliverwkim@gmail.com

# Appendix Slides

# Estimating Continuous Treatment

$$y_{i,b,t+h} - y_{i,b,t-1} = \frac{\alpha_h \Delta D_{i,b,t} + \beta_h \Delta D_{i,b,t} \times HRS_{i,b,t}}{+ \gamma_{b,t+h} (R_i \times B_{b,t}) + \delta_{b,t+h} (R_i \times B_{b,t} \times \Delta D_{i,b,t})}$$

$$+ B_{b,t} + e_{i,b,t}^h.$$

$$(1)$$

#### Continuous Effects

	Horizon (h years after reform)			
	0	1	2	3
Linear				
$lpha_{h}$	1.16	1.05	1.15	-0.00
	(0.78)	(0.67)	(0.94)	(0.94)
$eta_{h}$	-1.33	-1.29	-1.45	-0.01
	(0.93)	(0.78)	( 1.11)	( 1.07)
Binned				
$50\% \le HRS < 60\%$	-0.01	-0.01	-0.02	-0.02
	(0.02)	(0.01)	(0.01)	(0.0)
$60\% \leq HRS < 70\%$	-0.01	-0.06	-0.05	-0.03
	(0.01)	(0.04)	(0.05)	(0.04)
$70\% \le HRS < 80\%$	0.00	0.01	-0.02	-0.01
	(0.01)	(0.01)	(0.02)	(0.01)
$80\% \le HRS < 90\%$	0.00	0.02	0.02	0.02
	(0.01)	(0.02)	(0.03)	(0.02)
$90\% \le HRS < 100\%$	0.01	0.03	-0.01	0.02
	(0.04)	(0.04)	(0.04)	(0.02)

#### Threats to Identification

- ► Confounding treatment
- Spillovers
- Core vs. border differences

▶ Were there other policy changes at the same time that affected yields at the border?

- ▶ Were there other policy changes at the same time that affected yields at the border?
  - Biggest policy change was 1979 rise in procurement prices, but this was largely uniform across provinces
  - Can look at province-level event studies of inputs
  - Given null finding, requires somewhat convoluted story of cancelling out positive effect of HRS

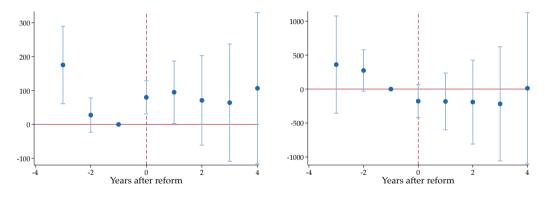


Figure 1: Draft animals

Figure 2: Farm machinery horsepower

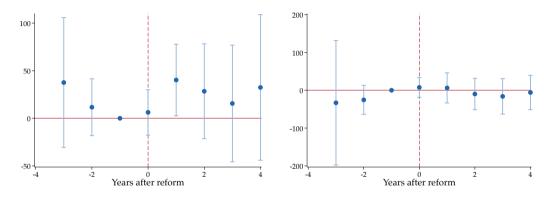


Figure 3: Fertilizer application

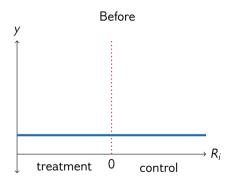
Figure 4: Total agricultural investment

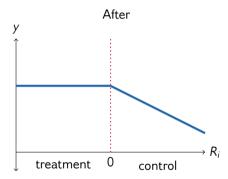
- Did the reform spread across boundaries?
  - ► Several institutional features restrain spillovers:

- Did the reform spread across boundaries?
  - Several institutional features restrain spillovers:
    - Totalitarian control over labor mobility under the hukou system Detail

- Did the reform spread across boundaries?
  - Several institutional features restrain spillovers:
    - ► Totalitarian control over labor mobility under the *hukou* system Detail
    - Strong ideological control over province boundaries (Teiwes and Sun 2016)

- Did the reform spread across boundaries?
  - Several institutional features restrain spillovers:
    - ► Totalitarian control over labor mobility under the hukou system Detail
    - Strong ideological control over province boundaries (Teiwes and Sun 2016)
  - Can test for spillovers under the weak assumption that effects on yields would decay with distance to the border





### Spillovers into untreated provinces, in distance to the border

$$y_{i,b,t+h} - y_{i,b,t-1} = \beta_h \Delta D_{i,b,t} + \gamma_{t+h} R_i + \delta_{t+h} (R_i \times \Delta D_{i,b,t}) + B_{b,t} + e_{i,b,t}^h$$
 (2)

	Horizon (h years after reform)			
	0	1	2	3
<u>Linear</u>				
$\gamma_{t+h}$ (in 1000 km)	0.03	-0.08	0.02	-0.12
	(0.07)	(0.14)	( 0.19)	( 0.16)
<u>Binned</u>				
$0 \text{km} \leq R_i < 10 \text{km}$	-0.00	0.00	0.00	0.01
	(0.00)	(0.04)	(0.06)	(0.02)
$10 \text{km} \leq R_i < 20 \text{km}$	-0.00	0.00	0.00	0.01
	(0.00)	(0.04)	(0.06)	(0.01)
$20 \text{km} \le R_i < 30 \text{km}$	-0.00	0.00	0.00	0.00
	(0.00)	(0.04)	(0.06)	(0.01)
$30 \text{km} \le R_i < 40 \text{km}$	-0.00	0.00	0.00	0.00
	(0.00)	( 0.04)	( 0.06)	( 0.01)

#### Institutional Details about the HRS

- ▶ Land Rights: the collective/village kept ownership of the land (Benjamin and Brandt 2002). In most villages, the village executive committee controlled tenancy/use rights, with occasional reallocations despite the law mandating 15 years of tenure (Brandt, Rozelle and Turner 2004).
- ► Capital: the communes' capital—tractors, agricultural machines—was distributed to households (Eisenman 2018, p.258)
- Land Markets: "Land... can also move among households in rental transactions. Although 70 percent of surveyed villages report that households enjoy unencumbered rights to rent land in 1995, the land rental market is thin." (Brandt, Rozelle and Turner 2004).
- ▶ Labor Markets: "The market for agricultural labor is equally thin. Only half of all villages report the use of hired farm labor in 1995, up from one quarter in 1988" (Brandt, Rozelle and Turner 2004). However, nonfarm labor was widespread: "employment in local village and township-run enterprises, family businesses, and long-term employment outside the village doubled between 1988 and 1995" (Brandt, Rozelle and Turner 2004)

# Politically Correlated Measurement Error in Official Yields

▶ We can model the official yield  $\tilde{y}_{i,t}$  as a function of the true yield  $y_{i,t}$ :

$$\tilde{y}_{i,t} = \lambda y_{i,t} + \varepsilon_{i,t}$$

Suppose the true relationship between yield and the HRS is

$$y_{i,t} = \beta_0 + \beta_1 HRS_{i,t} + \eta_{i,t}$$

but since we only have official yield data what we actually estimate is

$$\tilde{y}_{i,t} = \tilde{\beta}_0 + \tilde{\beta}_1 HRS_{i,t} + \tilde{\eta}_{i,t}.$$
 (3)

▶ Then the coefficient will be the true  $\beta$  plus a classic OVB term:

$$ilde{eta}_1 = \lambda eta_1 + rac{\operatorname{Cov}(\mathit{HRS}_{i,t}, arepsilon_{i,t})}{\mathbb{V}(\mathit{HRS}_{i,t})}.$$



### Lots of Measurement Error Math (continued)

▶ If  $Cov(HRS_{i,t}, \varepsilon_{i,t}) \neq 0$  or  $\lambda \neq 1$ , then  $\beta_1 \neq \tilde{\beta}_1$  and our estimate of  $\beta_1$  will be biased. We can check if the first of these conditions holds by bringing in the satellite data, which can be modeled as

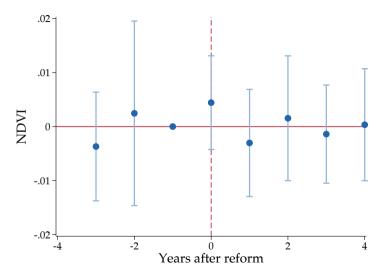
$$\bar{y}_{i,t} = \kappa y_{i,t} + \theta_{i,t}$$

- We know that  $Cov(\theta, HRS) = 0$ , which implies that  $\bar{\beta}_1 = \kappa \beta$ .
- lacktriangle The satellite-based regression results in column 4 suggest that  $ar{eta}_1 pprox 0$
- If we rule out the degenerate case where  $\kappa = 0$ , then the true relationship between NDVI and HRS is  $\beta_1 = 0$ . Since we found a positive  $\tilde{\beta}_1$ ,  $Cov(HRS_{i,t}, \varepsilon_{i,t}) > 0$
- Consistent with a story where provinces sought to present good economic news about the rollout of reform and exaggerated their grain yields as the HRS spread Lin 1992 table

# More Detailed Sample Description

- ▶ 22 provinces (drop municipalities, Tibet, Xinjiang, Hong Kong, Macau)
- ▶ 73 border groups

# Peak NDVI effect (no clean controls) Back



# Qualitative Evidence about the Spread of the HRS

- Unger (1985) interviewed 28 migrants to Hong Kong from 28 villages in Anhui, Fujian, Hubei, Jiangsu, Jiangxi, Shandong, Tianjin, and Zhejiang:
  - "Fully twenty-six out of the twenty-eight villages in my sample had indeed decollectivized into family smallholdings by the end of 1982. But official proclamations to the contrary, twenty-four of the twenty-eight interviewees report that in their own villages the decision as to precisely what type of system would be adopted was made exclusively by officials at levels far above the village. In only two villages had the team cadres and peasants themselves taken the initiative"
- Thaxton (2008) looks at Da Fo village:
  - ▶ "Da Fo's farmers launched themselves into private trade and began to reclaim market space before the Cultural Revolution had ended, before the CCP's Eleventh Congress, and before the baochan daohu system gained acceptance by the party's Central Committee. They also rejected all central and county government attempts to enlist their participation in the household responsibility system between 1979 and 1982, engineering their own silent land- to-the-tiller movement in this period."

- Bramall (1995):
  - The process of decollectivization the best-known part of the reform package did not begin in Sichuan until the autumn of 1977. Moreover... the dominant characteristic of this institutional change was its gradualism. Experimentation was very much the norm, and many of the province's most prosperous counties shifted to family farming (baogan dao hu) only after the autumn of 1982 and then only at the behest of central government. It is therefore difficult to use the phrase "almost spontaneous" to describe either the process of institutional change, or the rural reform package as a whole."

### The Hukou System (pre-1984)

- To move from a rural to urban area, needed a certificate of employment or school admission, or police permission
- State monopoly on urban housing
- Hotels or inns required official travel documents
- After 1960, you were required to show official travel certificates before buying a train, bus, or boat ticket
- In urban areas, residents had to show a ration card to draw grain tickets, which could then be exchanged for food
- Travellers had to bring their own grain to state grain stations where they would be exchanged for grain tickets

Back to ID threats

# Replicating Lin (1992)

We can return to the original Lin (1992) regression, which estimates for province i in year t,

$$\begin{aligned} &\ln(Y_{it}) = \alpha_1 + \alpha_2 \ln(\mathsf{Land}_{it}) + \alpha_3 \ln(\mathsf{Labor}_{it}) + \alpha_4 \ln(\mathsf{Capital}_{it}) + \alpha_5 \ln(\mathsf{Fertilizer}_{it}) \\ &+ \alpha_6 \mathsf{HRS}_{it} + \alpha_7 \mathsf{MP}_{t-1} + \alpha_8 \mathsf{GP}_t + \alpha_9 \mathsf{NGCA}_{it} + \alpha_{10} \mathsf{MCI}_{it} + \alpha_{11} T_t + \sum_{j=12}^{39} \alpha_j D_j + \varepsilon_{it} \end{aligned} \tag{4}$$

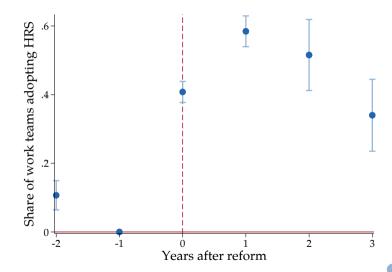
 $Y_{it}$  is agricultural output,  $HRS_{i,t}$  is the share of work teams in a province that have adopted household farming,  $MP_{i,t-1}$  is the lagged index of market prices divided by input prices,  $GP_t$  is the index of above-quota prices divided by manufactured input prices,  $NGCA_{it}$  is the share of total sown area dedicated to nongrain crops,  $MCI_t$  is a multiple cropping index,  $T_t$  is a linear time trend, and  $D_i$  is a province-level dummy variable.

# Running Lin (1992) with satellites Measurement Error

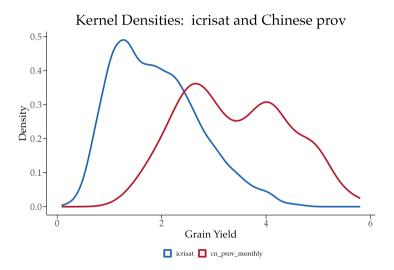
	(1)	(2)	(3)	(4)
	VFO	VFO (year effects)	Grain yield	Q2 NDVI mean
HRS share	0.220***	0.177***	0.221***	-0.0502
	(0.0300)	(0.0459)	(0.0535)	(0.0985)
In(Land)	0.594***	0.470***	-0.502*	0.0655
	(0.115)	(0.110)	(0.205)	(0.377)
In(Labor)	0.104***	0.125***	0.0493	-0.0548
	(0.0288)	(0.0275)	(0.0513)	(0.100)
In(Capital)	0.0675	0.173**	0.0354	-0.0359
·	(0.0607)	(0.0593)	(0.108)	(0.229)
In(Fertilizer)	0.172***	0.149***	0.165***	0.206*
	(0.0246)	(0.0240)	(0.0438)	(0.0872)
Observations	364	364	364	300
$R^2$	0.842	0.865	0.536	0.104

Standard errors in parentheses. \* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001

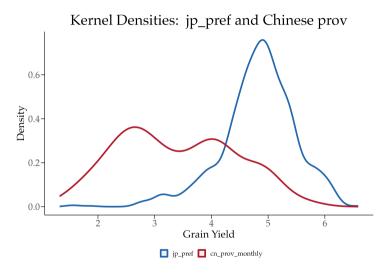
# **Event Study**



# Icrisat Density Comparison (Back)



# Japanese Prefecture Density Comparison (Back)



# Korea Density Comparison (Back)

