

# **Economic Impact of the Most Drastic Lockdown During COVID-19 Pandemic Policy** -----the Experience of Hubei, China

Xiao Ke<sup>1</sup>, Cheng Hsiao<sup>\*2</sup> <sup>1</sup>Hubei University of Economics, <sup>2</sup>University of Southern California

## Introduction

- Pandemics have had strong adverse effects on economic prosperity.
- In Dec. 2019, a disease of unknown causes suddenly afflicted Wuhan, a megacity in Hubei province in central China with 60 million inhabitants, & quickly spread to the whole nation.

• Facing the raging COVID-19 epidemic, the Chinese government rolled out perhaps the most agile and aggressive disease containment effort (WHO, 2020).

## **Methods**

The contemporaneous covariance between  $y_{it}$  and  $y_{jt}$  is given by

 $cov(y_{it}, y_{jt}) = b'_i E(f_t f'_t) b_j$ 

As long as

 $E(u_{it}|d_{1t}) = 0, i = 2, \dots, N$ 

One can write

 $y_{1t}^{0*} = E(y_{1t}^{0*}|\tilde{y}_t) + \eta_{it} = a + c'\tilde{y}_t + \eta_{it}, \ t = 1, \dots, T$ 

where  $\tilde{y}_t = (y_{2t}, ..., y_{Nt})'$ ,  $E(\eta_{1t} | \tilde{y}_t) = 0$ . Hsiao et al. (2012) show that minimizing

$$\frac{1}{T_1}(y_1^0 - ea - Yc)'A(y_1^0 - ea - Yc)$$

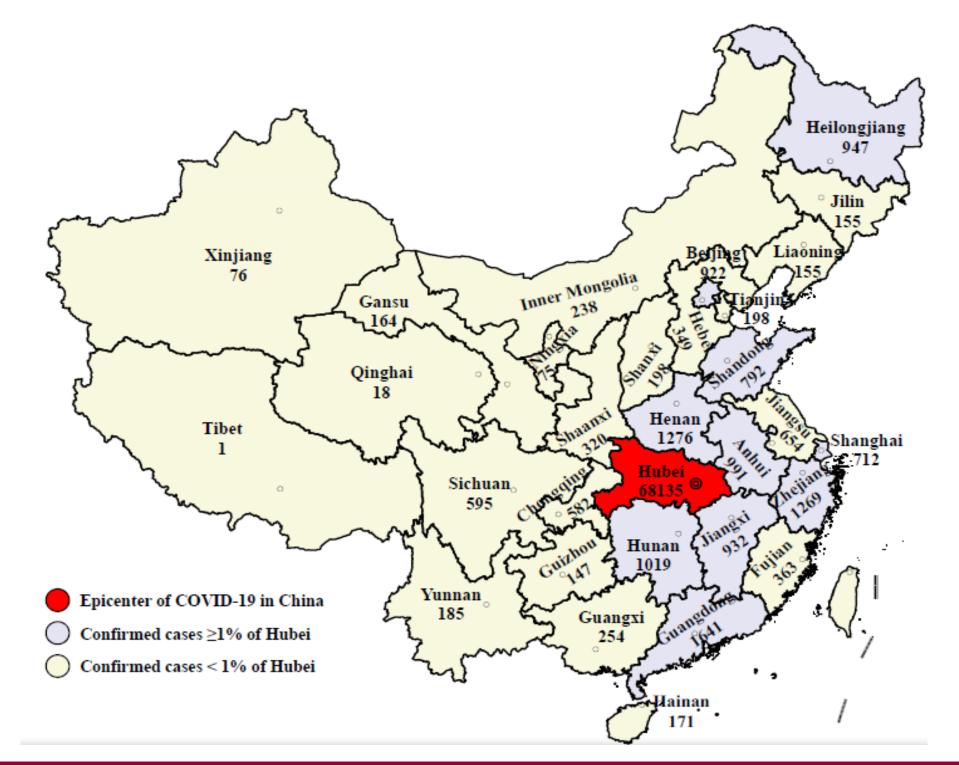
Summary of the loss/completion relative to counterfactual Hubei—Using AICC

	Proportion of loss to counterfactual	Proportion of completion to counterfactual			
	Lockdown quarter	Re-opening quarters	First half of 2020	Whole year 2020	
GDP	37%	100%	81%	92%	
Primary industry value added	17%	94%	94%	92%	
Second industry value added	46%	95%	77%	84%	
Tertiary industry value added	31%	90%	82%	84%	
Industry value added cumulative growth rate	54%	96%	82%	91%	
Fixed capital investment	82%	64%	46%	57%	
Retail sales	30%	87%	77%	82%	
Export and import	28%	96%	82%	92%	
Export	36%	92%	77%	88%	
Import	25%	96%	81%	92%	
Real estate expenditure	71%	106%	63%	94%	
Road passenger ridership	92%	44%	38%	45%	
Road freight volume	99%	95%	63%	88%	

• We uses the Hsiao, Ching, and Wan (2012) panel approach and Hsiao and Zhou (2019) to assess the evolution of economic consequences of the drastic lockdown policy in the epicenter of COVID-19---the Hubei Province of China.

### The lockdown treatment

Severality of the COVID-19 in the epicenter & nonepicenter regions in China



yields consistent estimates of a and c, where  $y_1^0 = (y_{11}, \ldots, y_{1T_1})'$ , A is a  $T_1 \times T_1$  positive definite matrix. Define the predictor for counterfactual  $y_{1t}^0$ without the COVID-19 outbreak and the decisive lockdown in Hubei

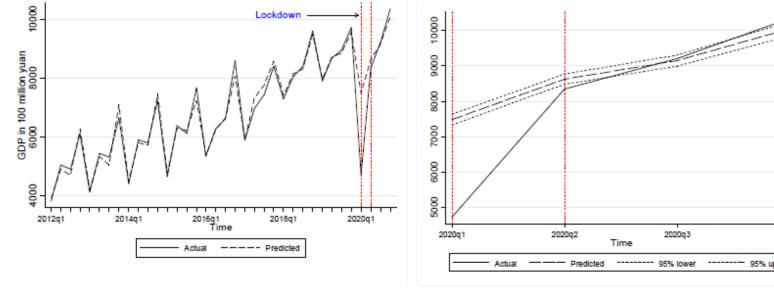
 $\hat{y}_{1t}^{0*} = \hat{a} + \hat{c}' \tilde{y}_t, \ T_1 + 1, \dots, T$ 

Prediction for the drastic COVID-19 lockdown policy treatment effect:

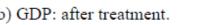
 $\hat{\Delta}_{1t} = y_{1t} - \hat{y}_{1t}^{0*}$  for  $t = T_1 + 1, \dots, T$ 

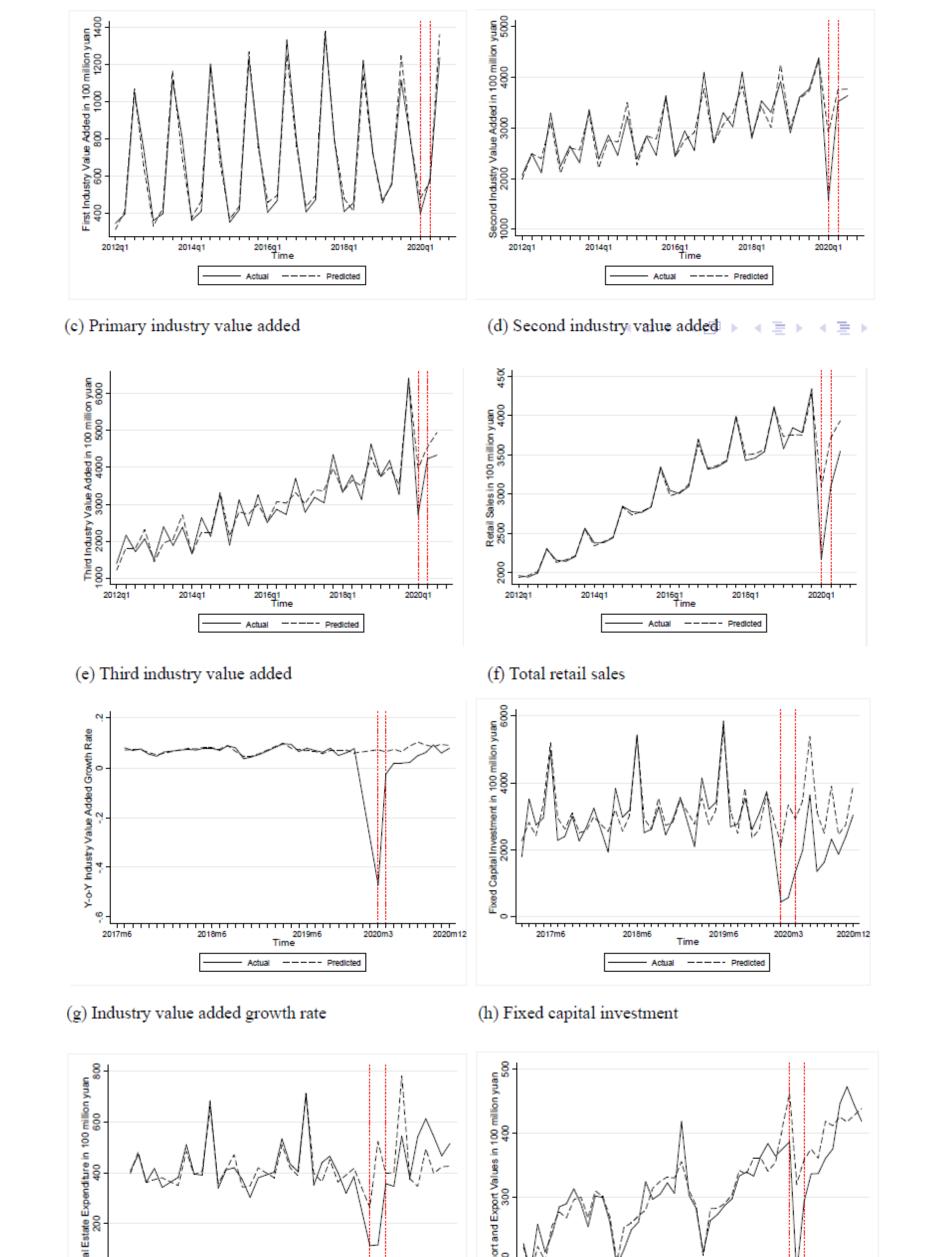
### **Results**

Treatment effects on GDP, total retail, industry value added growth rate, fixed capital investment, real estate investment, export, import, road passenger ridership, and road freight transport volume



#### (a) GDP: full sample





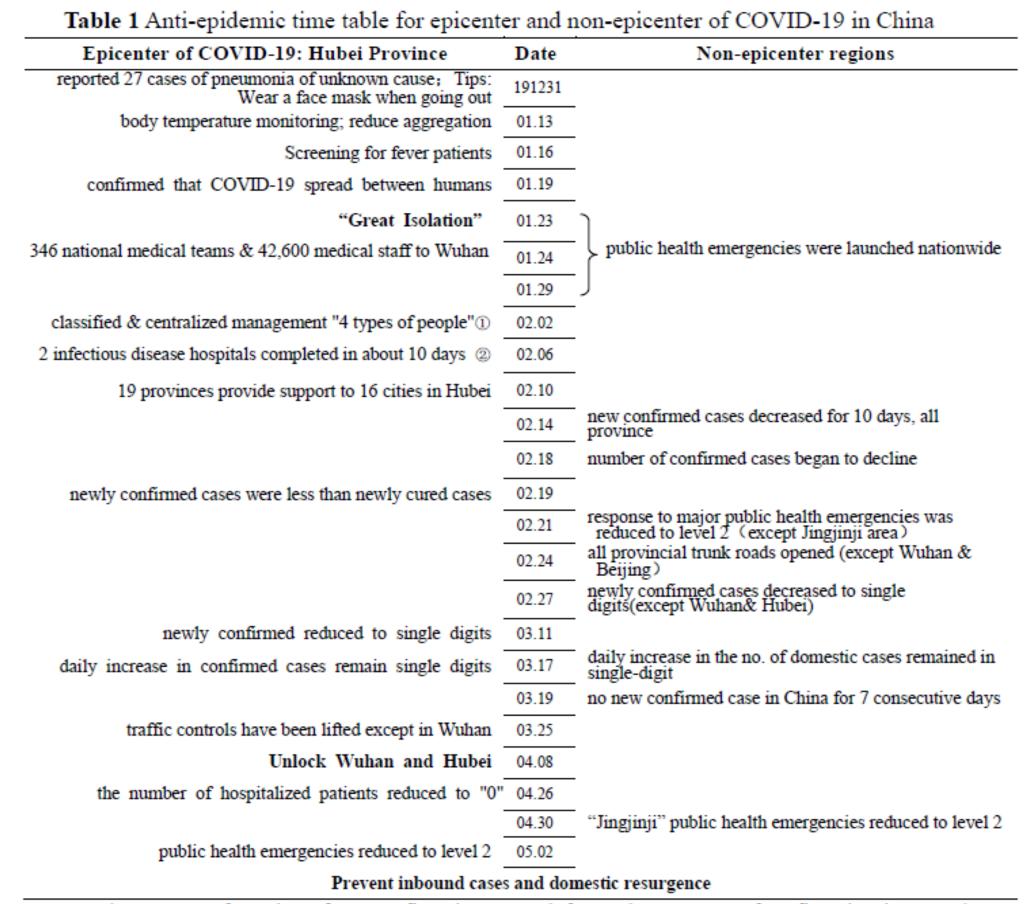
Notes: 1. In the table, we have two types of data: quarterly and monthly. For quarterly data, the lockdown quarter is 2020:q1, re-opening quarters are 2020:q2-2020:q4 for GDP and 2020:q2-2020:q3 for value added of the primary, secondary, tertiary industries, and retail sales respectively. For monthly data (fixed capital investment, export, import, real estate expenditure, road passenger ridership, and road freight volume), the lockdown quarter is 2020:m2-2020:m3; and for industry value added cumulative growth rate, the lockdown quarter is 2020:m3. The re-opening quarters are 2020:m4-2020:m12 for all the variables. The first half of the year refers to policy evaluation period from the lockdown quarter to 2020:m6 or 2020:q2. And the whole year 2000 refers to the whole policy evaluation period from the lockdown quarter to the end of the re-opening quarters for each macroeconomic indicator respectively.

Summary of the loss/completion relative to counterfactual Hubei—Using LASSO

Table 12 Summary of the loss/completion due to LASSO method of selecting the predictor or changes in control units

changes in control units								
-			Proportion of completion to counterfactual					
Lockdown quarter		First half of 2020		Whole year 2020				
LASSO	Control group	LASSO	Control group	LASSO	Control group			
35%	36%	81%	82%	92%	91%			
21%	_	94%	_	92%	_			
47%	_	77%	—	90%	—			
31%	32%	83%	83%	86%	89%			
54%	54%	82%	82%	92%	91%			
80%	80%	44%	47%	56%	56%			
29%	31%	78%	76%	77%	81%			
28%	_	84%	_	94%	_			
34%	_	79%	_	88%	_			
16%	_	91%	_	97%	_			
73%	73%	62%	59%	92%	86%			
92%	_	38%	—	45%	—			
99%	_	63%	_	88%	_			
-	Proportio to counte Lockdow LASSO 35% 21% 47% 31% 54% 80% 29% 28% 34% 16% 73% 92%	C     Proportion of loss to counterfactual     Lockdown quarter     LASSO   Control group     35%   36%     21%      47%      31%   32%     54%   54%     80%   80%     29%   31%     28%      34%      16%      73%   73%     92%	Proportion of loss   to counterfactual   Lockdown quarter First half of   LASSO Control group LASSO   35% 36% 81%   21% — 94%   47% — 77%   31% 32% 83%   54% 54% 82%   80% 44% 29%   28% — 84%   34% — 79%   16% — 91%   73% 73% 62%	Proportion of loss to counterfactual   Proportion to count for counterfactual     Lockdown quarter   First half of 2020     LASSO   Control group   LASSO   Control group     35%   36%   81%   82%     21%   —   94%   —     47%   —   77%   —     31%   32%   83%   83%     54%   54%   82%   82%     80%   80%   44%   47%     29%   31%   78%   76%     28%   —   84%   —     34%   —   91%   —     73%   73%   62%   59%     92%   —   38%   —	Proportion of loss Proportion of completing to counterfactual   Lockdown quarter First half of 2020 Whole year   LASSO Control group LASSO Control group LASSO   35% 36% 81% 82% 92%   21%  94%  92%   47%  77%  90%   31% 32% 83% 83% 86%   54% 54% 82% 92%   80% 84% 47% 56%   29% 31% 78% 76% 77%   28% - 84% - 94%   34% - 91% - 88%   16% - 91% - 97%   73% 73% 62% 59% 92%   92% - 38% - 45%			

#### Anti-epidemic measures for epicenter annon-epicenter of COVID-19 in Chin



Notes: 1. Notes are the same as those in Table 11. 2. As for the robustness checks with respective to changes in control group, we re-estimate the impact of COVID-19 lockdown on variables which select Hubei's neighboring provinces or Eastern coastal provinces with none-zero weights in our baseline analysis. See text for details

## Discussions

Used the HCW and Hsiao and Zhou (2019) panel data program evaluation approach to assess the economic impacts of Hubei Provinces lockdown policy in times of COVID-19 pandemic.

• We find a strict lockdown could result in huge economic loss in times of pandemic but it also showed that as long as the epidemic could be quickly contained, the

Notes: 1. The "4 types of people" refer to confirmed, suspected, fever, close contacts of confirmed patients.2. The infectious disease hospitals specially built for the treatment of critically ill patients are "Leishenshan" and "Huoshenshan" hospitals.

## **Methods**

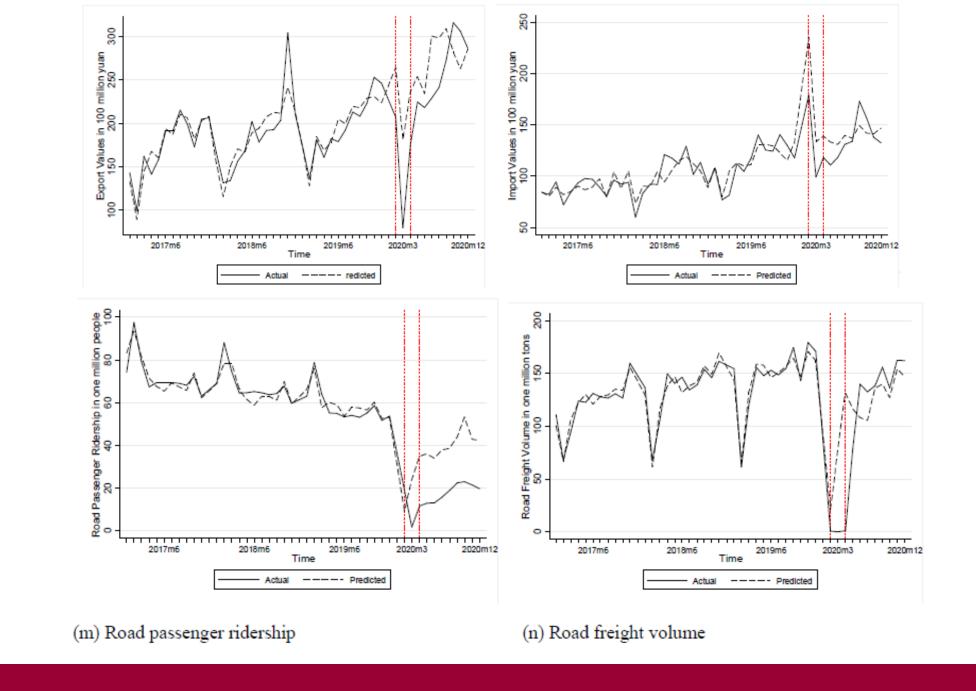
Model of untreated outcome  $y_{it}^0$  for i = 1, ..., N, t = 1, ..., T

 $y_{it}^{0} = \mathbf{b}'_{i}f_{t} + \alpha_{i} + u_{it}$ 

province-specific effects  $\alpha_i =$ 

- $K \times 1$  (unobserved) common factors that vary over time  $f_t =$
- $K \times 1$  vector of constant loadings that may vary across i  $b_i =$
- K = number of common factors

 $u_{it} = i$ th unit idiosyncratic error with  $E(u_{it}) = 0, E(u_{it}u_{is}) = 0$  for  $i \neq j$ 



2018m6

2017m6

(j) Import & export

2019m6

Actual — Predicte

2019m6 Time

- Actual ---- Predicted

(i) Real estate expenditure

2020m3

economic loss was temporary and controllable.

- How did Hubei's economy recover so fast after the COVID-19 epidemic breakout?
- 1. The "lockdown" is temporary. It has not affected the fundamental structure of the economy;
- 2. The Chinese government's "lockdown" policy brought the spread of COVID-19 to a halt in three months;
- 3. The resilience of people in Hubei also played a pivotal role.

## **Literature Cited**

- Hsiao, C., Ching, H. S., & Wan, S. K. (2012). A panel data approach for program evaluation--Measuring the benefits of political and economic integration of Hong Kong with mainland China. Journal of Applied Econometrics, 27, 705-740.
- Li, K., & Bell, D. (2017). Estimation of average treatment effects with panel data: Asymptotic theory and implementation. Journal of Econometrics, 197, 65-75.

## Xiao Ke, assistant professor, HUE, China (xiaokeproperty@163.com)