

Macroeconomic Effects of COVID-19 Across the World Income Distribution

Titan Alon ¹ Minki Kim ¹ David Lagakos ² Mitchell VanVuren ¹

¹University of California San Diego

²Boston University and NBER

IMF ARC

Nov 4, 2021

Emerging Market Most Severely Affected by the Pandemic

- ▶ GDP fell more in emerging markets than in either richer or poorer countries
- ▶ Excess deaths per capita relatively higher in emerging markets
- ▶ Why? Different fundamentals? Different policies?

Emerging Market Most Severely Affected by the Pandemic: Why?

Our Approach: Quantitative analysis using macro model with disease spread

Key differences of emerging market economies

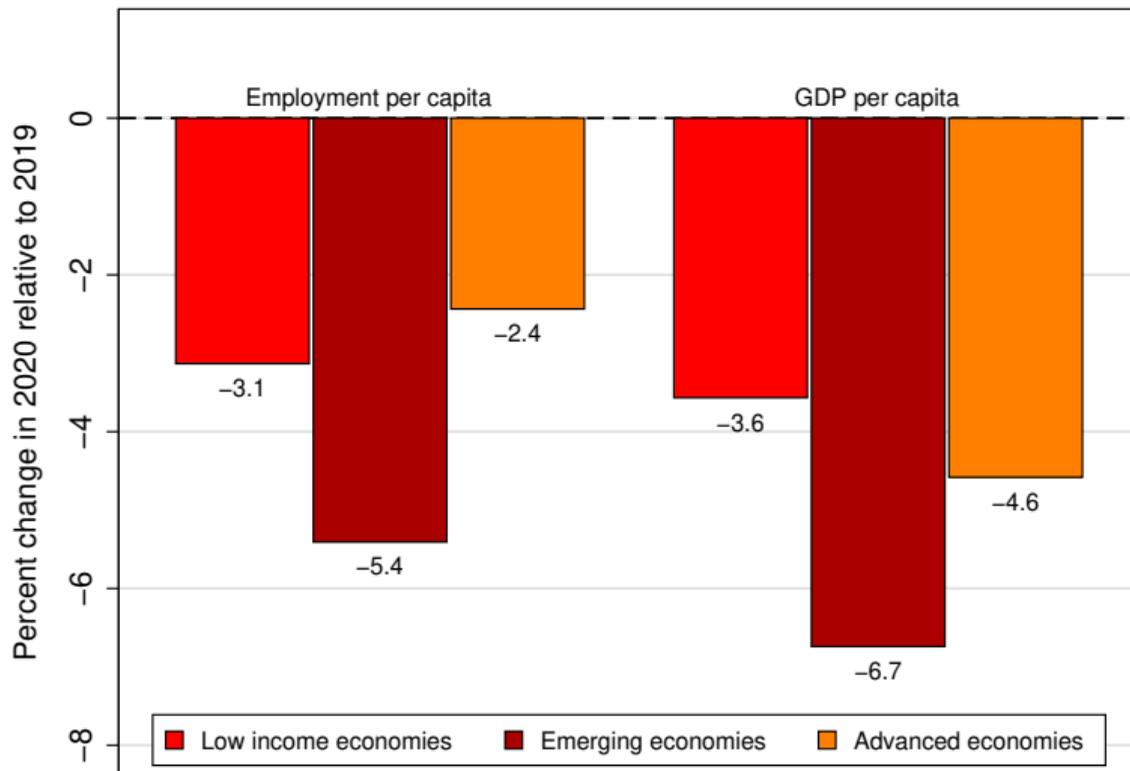
1. Large “social” sector requiring face-to-face interactions
2. Young populations (though not as young as low-income countries)
3. Strict lockdowns
4. Low healthcare capacity (though better than low-income countries)
5. Smaller transfers (work in progress)

Quantitative Conclusions So Far

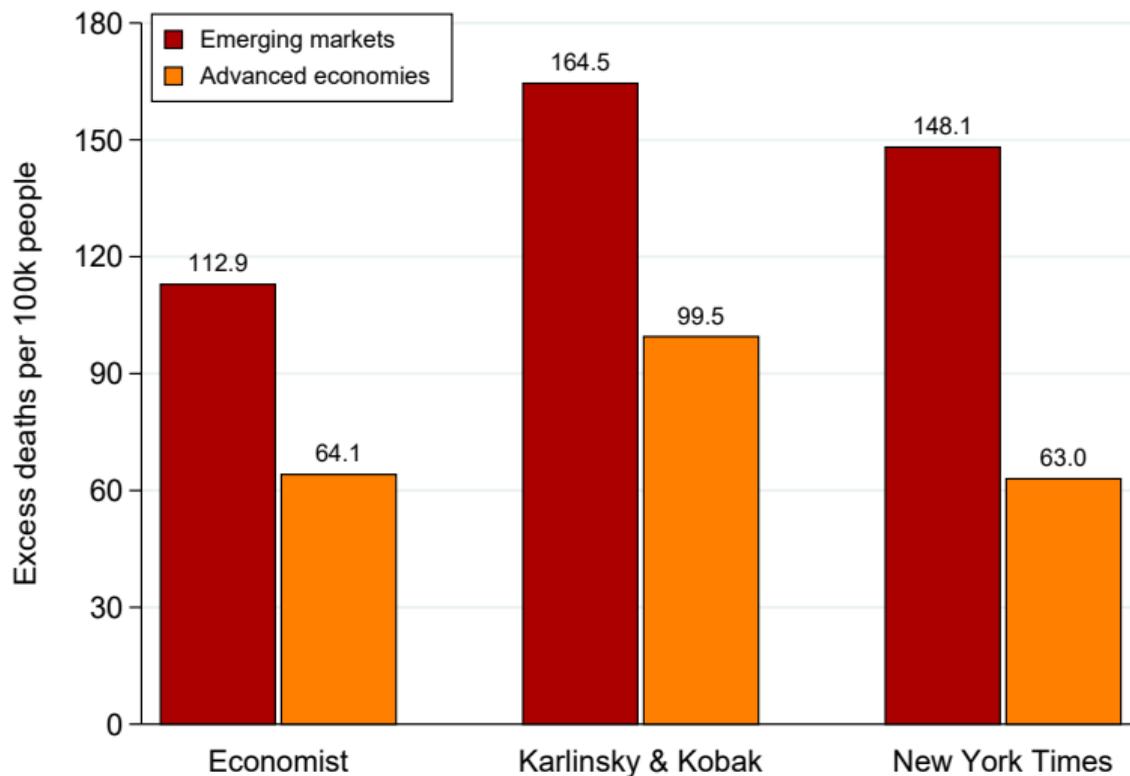
- ▶ Bigger GDP declines in emerging markets largely due to higher social employment shares & stricter lockdowns
- ▶ Model predicts higher mortalities in emerging markets but under-predicts data
- ▶ Low-income countries fared better largely due to much younger population and high "non-social" employment

Facts

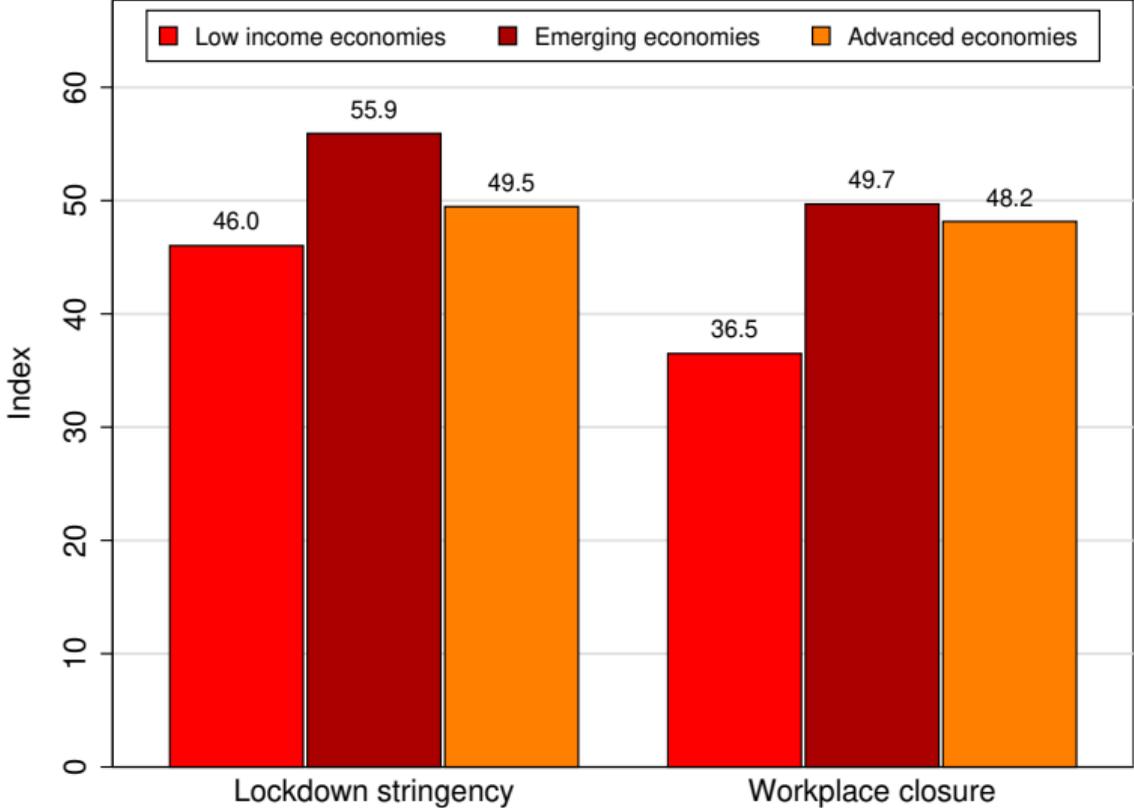
Employment and GDP Changes, 2019 to 2020



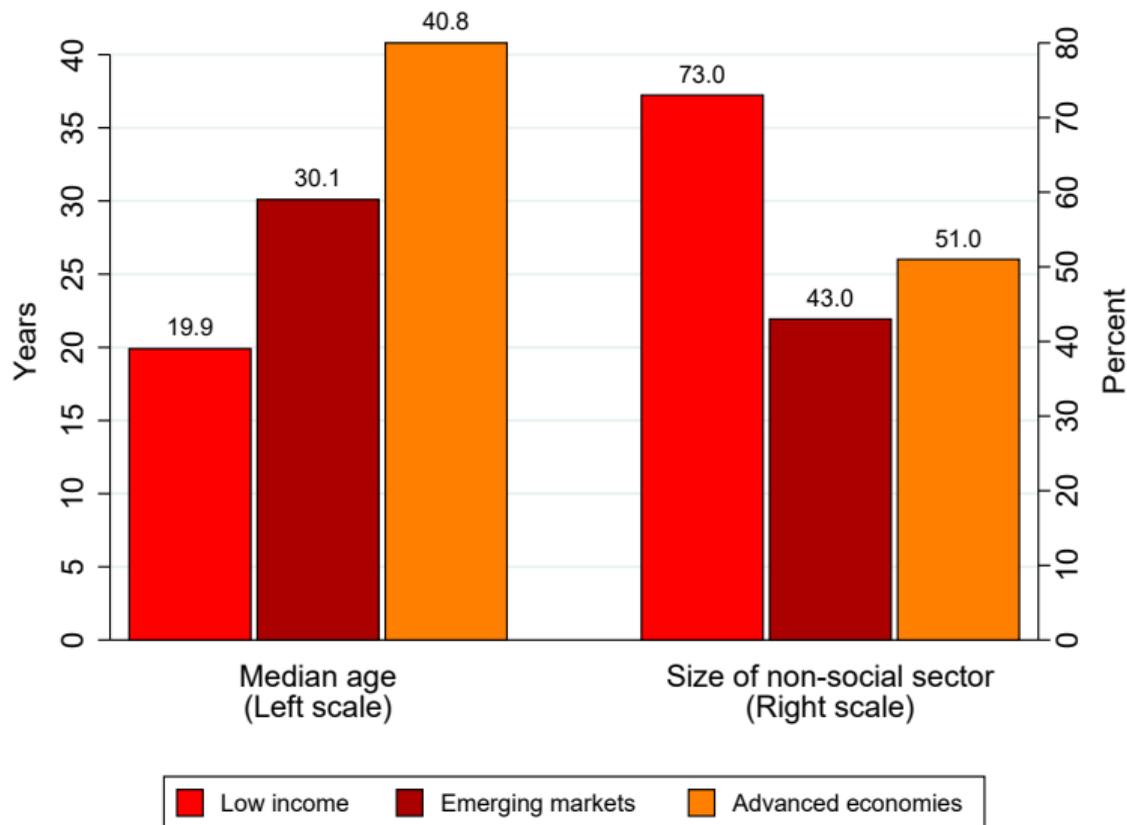
Excess Deaths per 100,000 People



Lockdown Intensity Index



Median Age and Non-Social Share in Employment



Model

Two-Sector Heterogeneous Agent Model + **Epidemiology**

Epidemiology

- ▶ SICR with age heterogeneity as in Glover et al. (2020): two age groups

Households

- ▶ Face uninsured idiosyncratic labor income risk and health risk
- ▶ Accumulate assets endogenously, face credit constraint

Two sectors

- ▶ Social: remote work entails large productivity loss
- ▶ Non-social: can work remotely with little productivity loss

Government

- ▶ Imposes containment policies and administer vaccines
- ▶ Collects taxes and makes transfers but with limited fiscal capacity

Households and Preferences

- ▶ Two age groups: young adults (ω_y) and old (ω_o)
- ▶ Preferences (of the living):

$$\mathbb{E} \left[\sum_{t=0}^{\infty} \beta_j^t \log(c_t) + \bar{u} \right]$$

- ▶ β_j is discount factor of age group j , where $j \in \{y, o\}$
- ▶ \bar{u} is the flow value of being alive

Sector, Permanent Productivities and Idiosyncratic Shocks

- ▶ Individuals are assigned to one of the two sectors; $s \in \{S, N\}$:
 1. Social sector (ω_S) : occupations with little room for remote work
Examples: Waitress, hair dressers
 2. Non-social (ω_N): occupations which can be done remotely
Examples: Professors, subsistence farmers
- ▶ Individuals are also endowed with a permanent productivities $z \sim G$
- ▶ Individuals face idiosyncratic productivity shock as in Aiyagari (1994)

$$\log v_{t+1} = \rho_v \log v_t + \varepsilon_{t+1} \quad \text{with} \quad \varepsilon_{t+1} \stackrel{iid}{\sim} F(0, \sigma_v)$$

Regular and Remote Work

- ▶ Workers can choose to work remotely
- ▶ Remote work involves less social contact, hence safer
- ▶ However, remote work also entails income loss

$$\text{Labor income} = \begin{cases} w_t \times v \times z & \text{if go to workplace} \\ \phi_s \times w_t \times v \times z & \text{if work remotely} \end{cases}$$

- ▶ ϕ_s : productivity penalty of remote work, $0 \leq \phi_s \leq \phi_N \leq 1$

Assets and Individual's Budget Constraint

- ▶ Borrowing limit: $a \geq \bar{a}$
- ▶ Individuals in sector s have the following budget constraint:

$$c + a' \leq (1 - \tau)w_s z v n + (1 + r)a + T$$

$$n = \begin{cases} 1 & \text{if go to workplace} \\ \phi_s & \text{if work remotely or under lockdown} \end{cases}$$

- ▶ Final good technology:

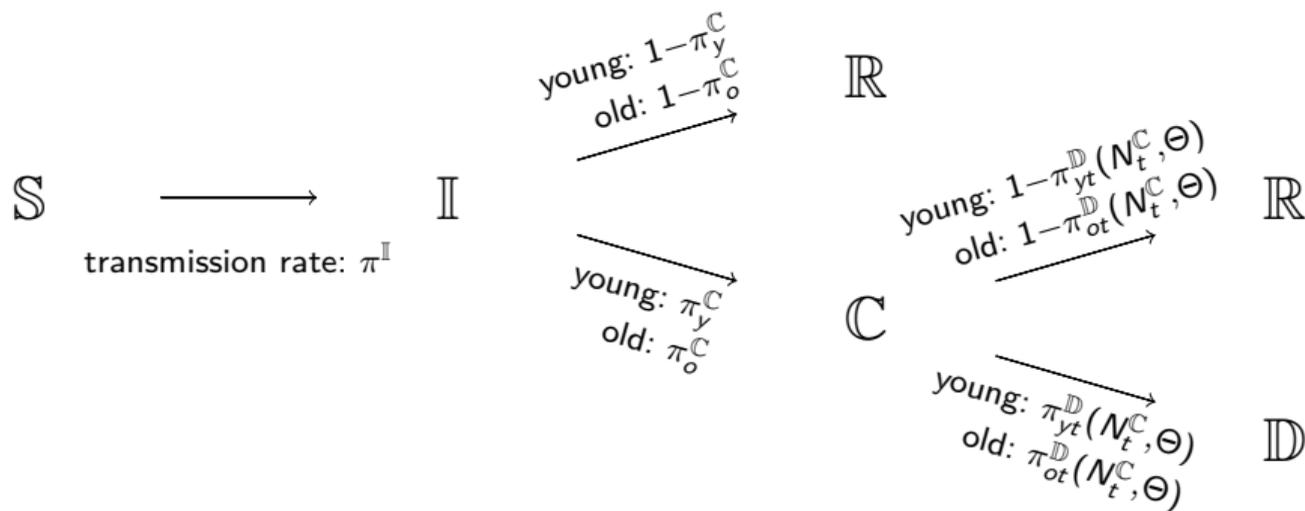
$$Y = AL^\alpha K^{1-\alpha}, \quad 0 < \alpha \leq 1$$

$$L = L_S + L_N$$

- ▶ K rented at exogenously given international rental rate r^F

Health States and Transitions

- ▶ Being infected drops all productivities by fraction $0 < \eta \leq 1$ until recovery
- ▶ Being critical drops all productivities to 0 until recovery



- ▶ Baseline probability a susceptible person becomes infected is:

$$\pi_t^{\text{I}} = \beta_t^{\text{I}} \times N_t^{\text{I}} / N_t$$

- ▶ β_t^{I} is exogenous time-varying infection rates
 - ▶ E.g. more social distancing or masking, better treatment

Voluntary Substitution Away From Workplace and Lockdowns

Voluntary Substitution

- ▶ Working remotely lowers the probability of infection:

$$\pi^{\text{I}} = \begin{cases} \beta^{\text{I}} \times N^{\text{I}}/N & \text{if go to workplace} \\ \beta^{\text{I}} \times N^{\text{I}}/N \times \xi & \text{if work remotely} \end{cases}$$

$0 \leq \xi < 1$: represents how much safer remote work is

Lockdown

- ▶ Randomly select λ fraction of each (young, old) \times (social, non-social) groups
- ▶ Selected individuals are forced to work from home
- ▶ $0 \leq \lambda \leq 1$: lockdown intensity, which varies across countries

Voluntary Substitution Away From Workplace and Lockdowns (continued)

- ▶ In each period, susceptible individuals can choose between working remotely and going to workplace by choosing the option with higher value

$$V = \max\{V^w + \varepsilon_w, V^r + \varepsilon_r\}$$

where ε_x , $x \in \{w, r\}$: independently distributed Gumbel taste shock

- ▶ Recovered / vaccinated individuals are not subject to lockdown

Hospital Capacity

- ▶ Θ is maximum ICU capacity per capita ($0 < \Theta < 1$)
- ▶ Probability of receiving an ICU bed is $\min\{\frac{\Theta}{N_t^C}, 1\}$
- ▶ Fatality rate $\pi_{jt}^{\mathbb{D}}$:

$$\pi_{jt}^{\mathbb{D}}(N_t^C, \Theta) = \begin{cases} \pi_j^{\mathbb{D}} & \text{if assigned ICU bed} \\ \kappa \times \pi_j^{\mathbb{D}} & \text{if not assigned} \end{cases}$$

- ▶ $\pi_j^{\mathbb{D}}$: baseline fatality rate of an age group j patient
- ▶ κ governs the impact of hospital overuse on fatality rate

Calibration Summary

Calibrate model to match US time series

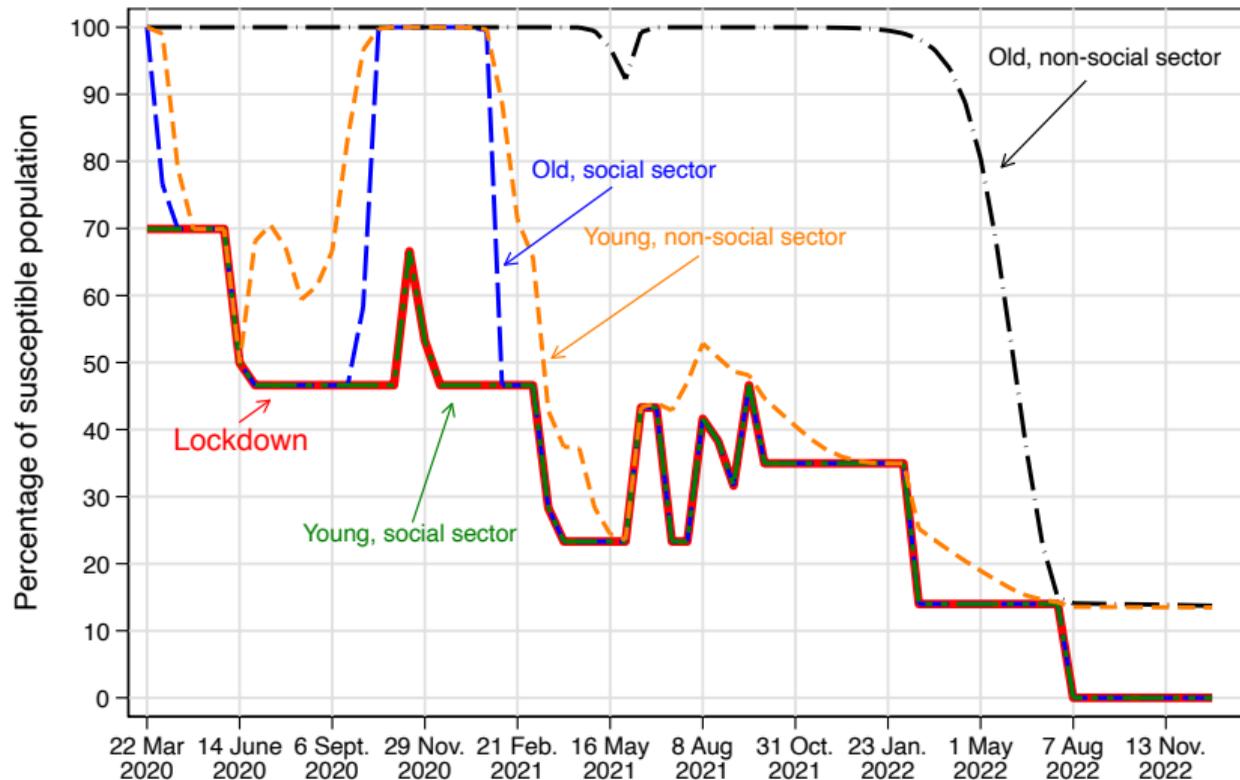
Death rates by age taken from epidemiology literature

Income process and macro parameters taken from literature

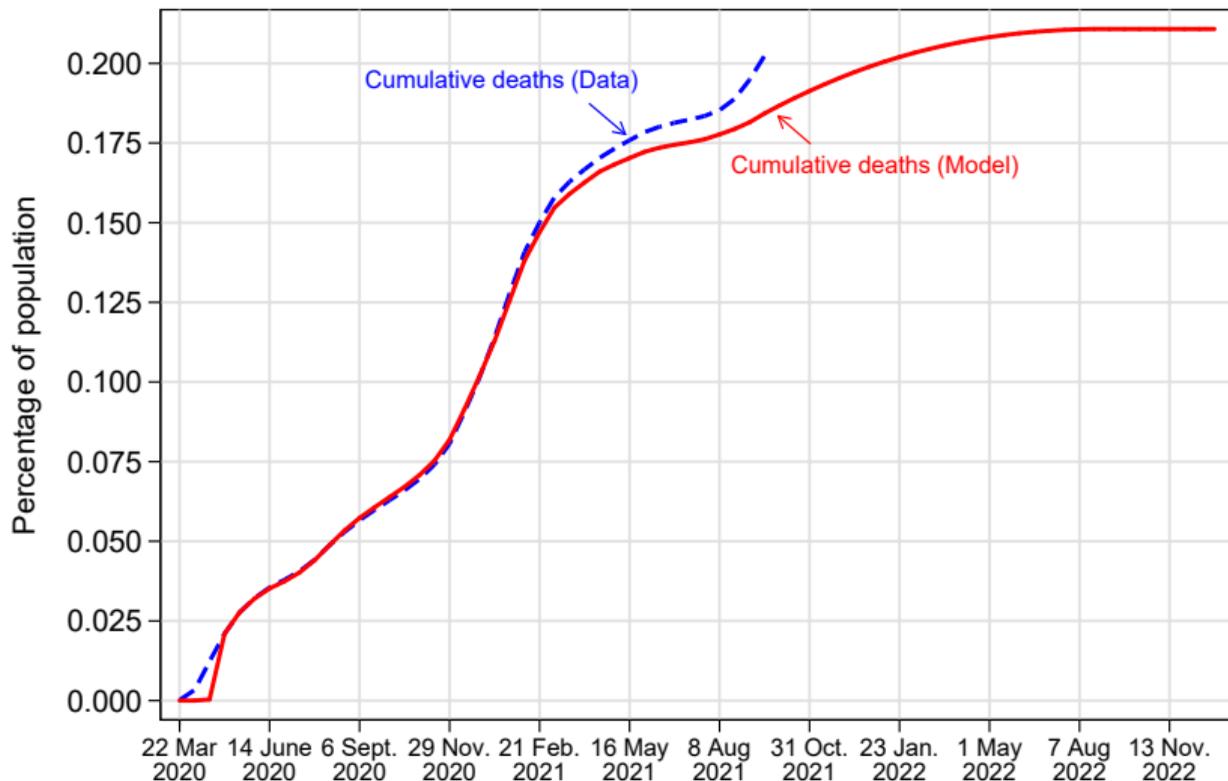
Penalties for work from home: 0% for Non-social; 28% for Social

Time-specific infection probability taken to match U.S. cumulative deaths

Patterns of Remote / Regular Work Along the Path of the Pandemic



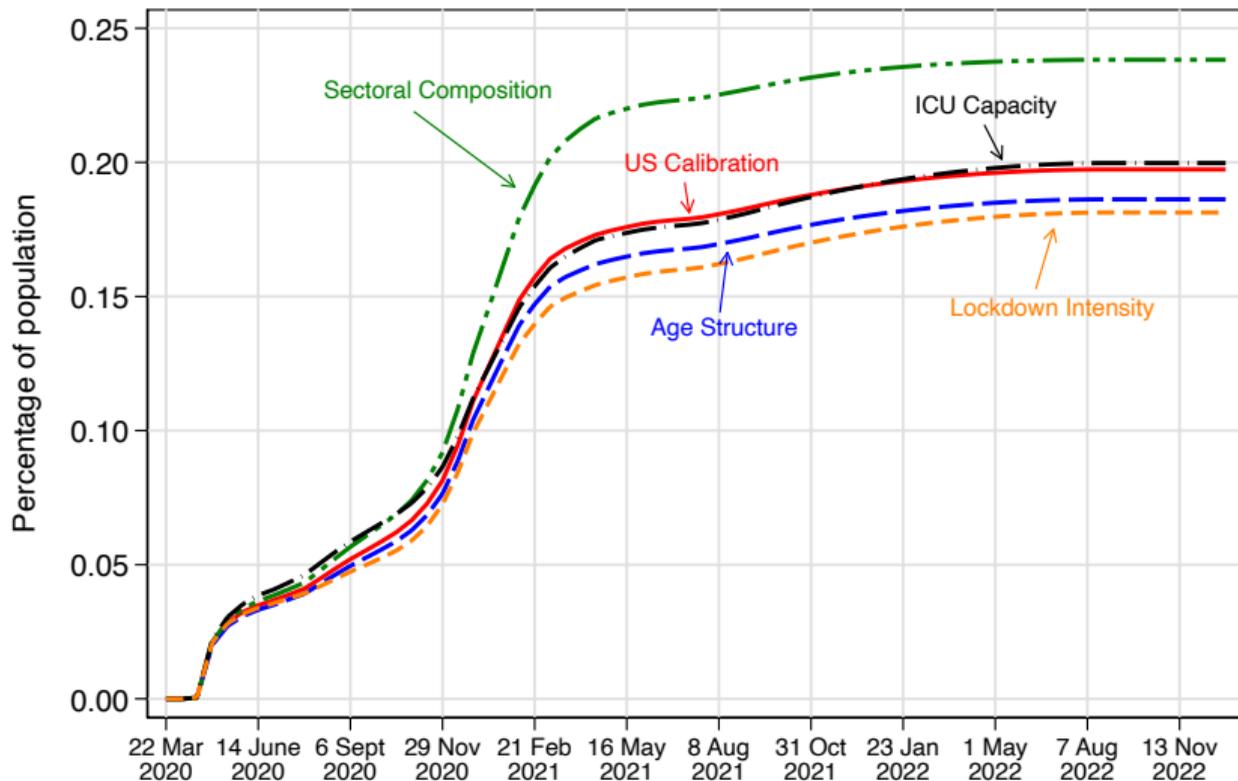
Predicted vs. Actual U.S. COVID-19 Deaths



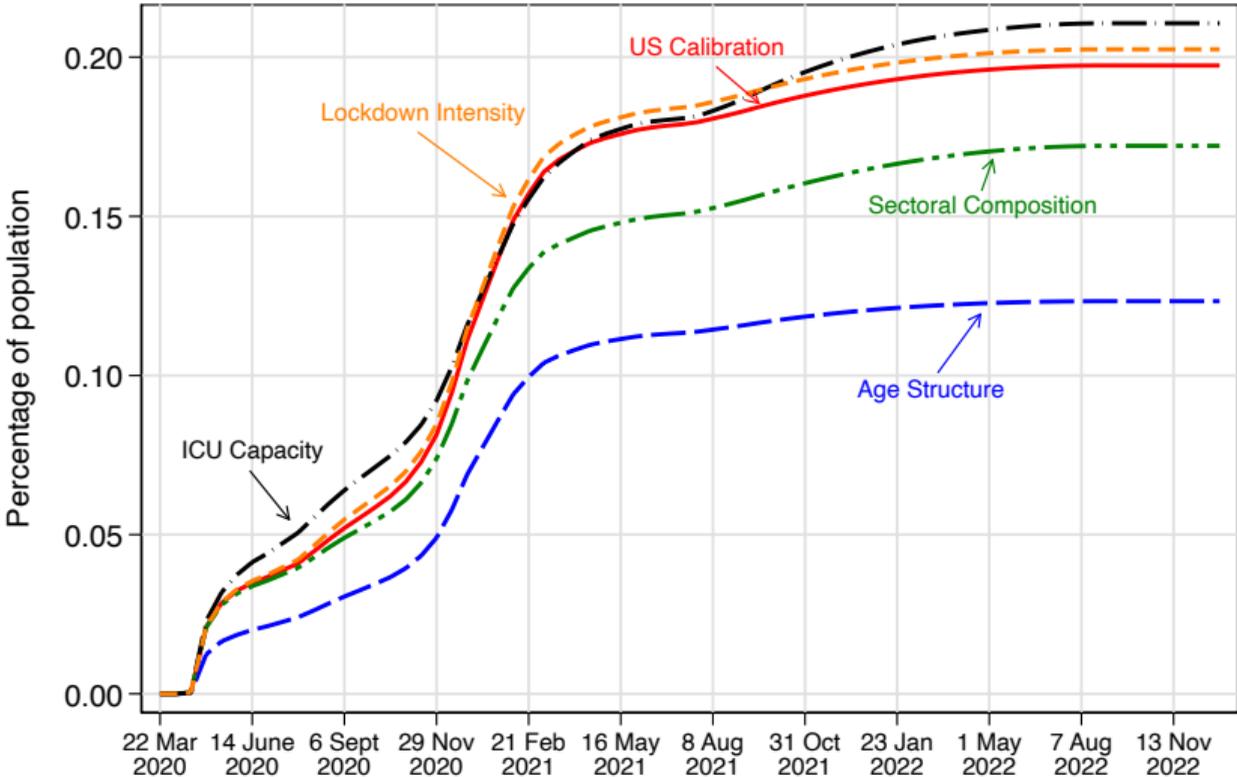
Counterfactual Simulations

1. What would U.S. deaths have been if the U.S. had the characteristics of
 - ... emerging market economies?
 - ... low-income economies?
2. What would U.S. GDP have been?

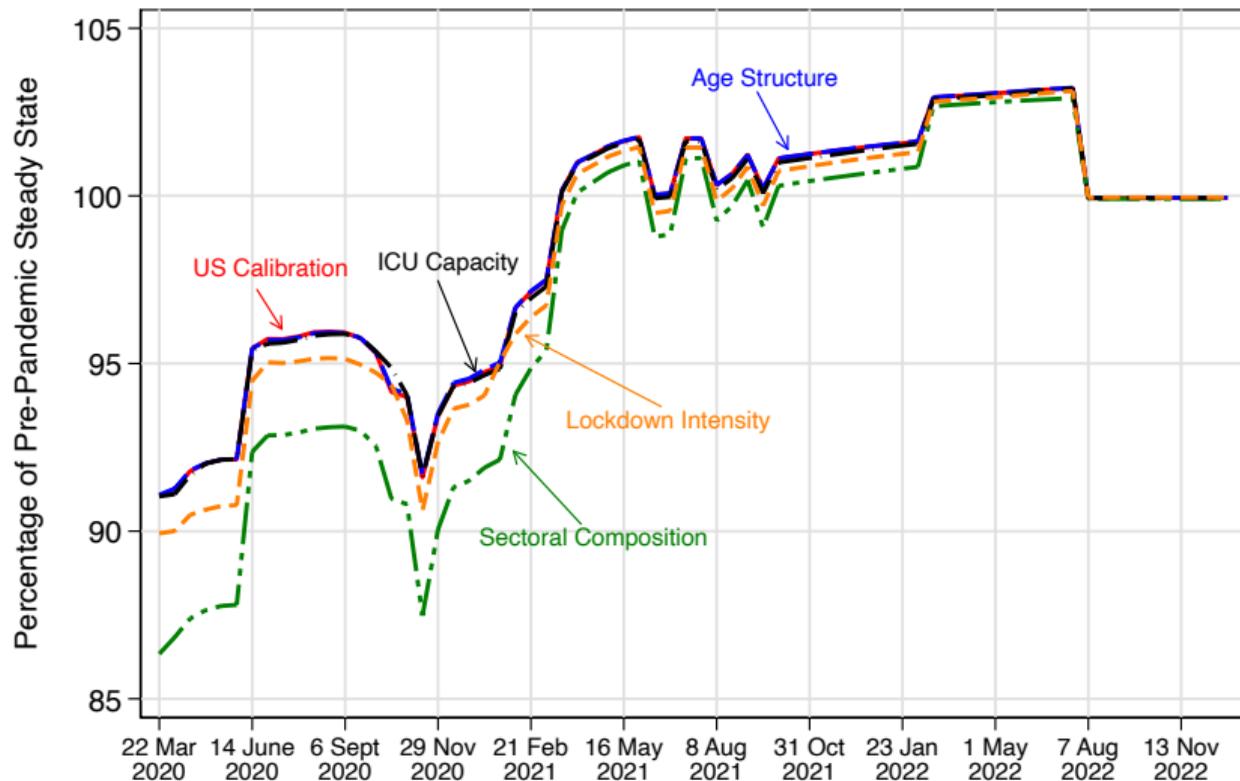
Counterfactual Deaths: U.S. with Emerging Markets' Features



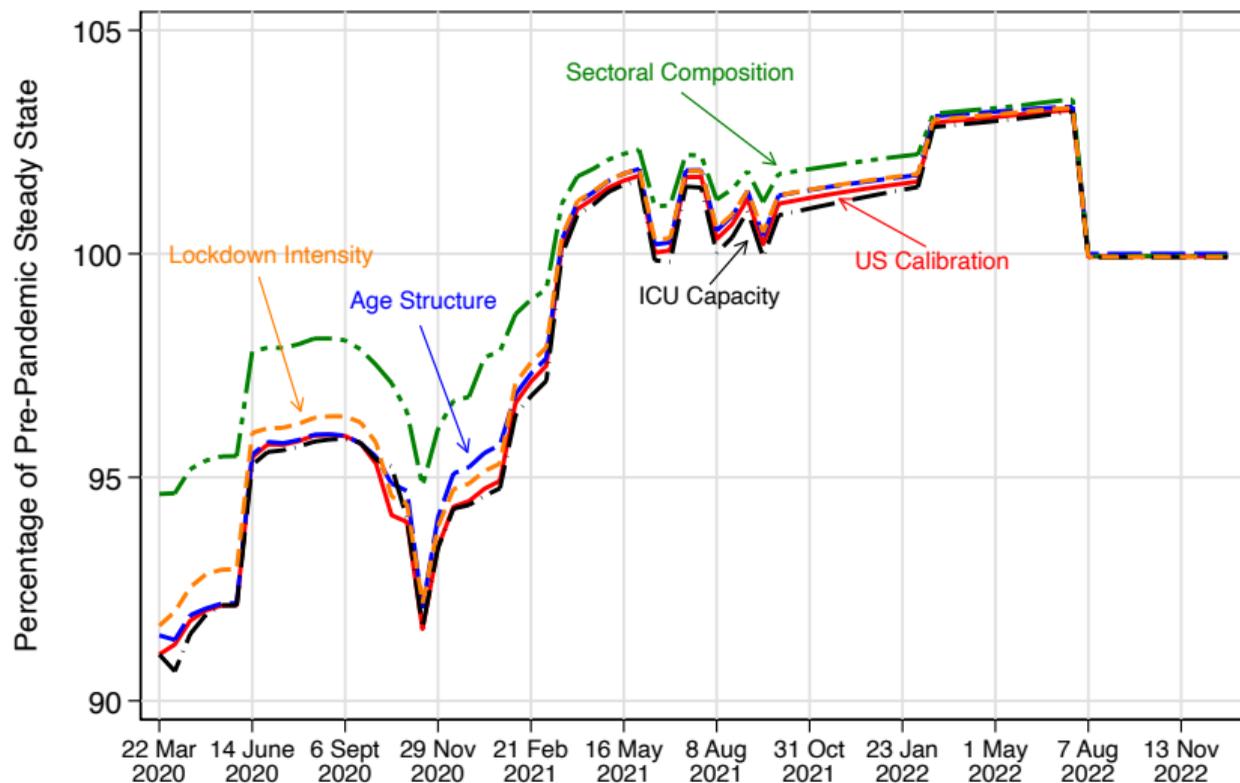
Counterfactual Deaths: U.S. with Low-Income Economies' Features



Counterfactual GDP: U.S. with Emerging Markets' Features



Counterfactual GDP: U.S. with Low-Income Economies' Features



Conclusions So Far

- ▶ Macroeconomic outcomes worse during pandemic in emerging markets than in either advanced economies or low-income economies
- ▶ Quantitative results so far from macro-epidemiology model:
 - Emerging markets worse in large part due to higher shares of “social” employment & stricter lockdowns
 - Low-income economies escaped largely due to lower social employment shares & younger age structure

Future Work (Hopefully For Others too...)

- ▶ Model still greatly under-predicting deaths in emerging markets
- ▶ Pandemic transfers differed across countries; still need to add this
- ▶ More broadly, other factors absent here likely relevant for why emerging markets did particularly worse
- ▶ These include: mask prevalence, other comorbidities, school closing policies, vaccine rollouts ...