

U.S. versus Europe: How Differential COVID-19 Policies Affect Inequality

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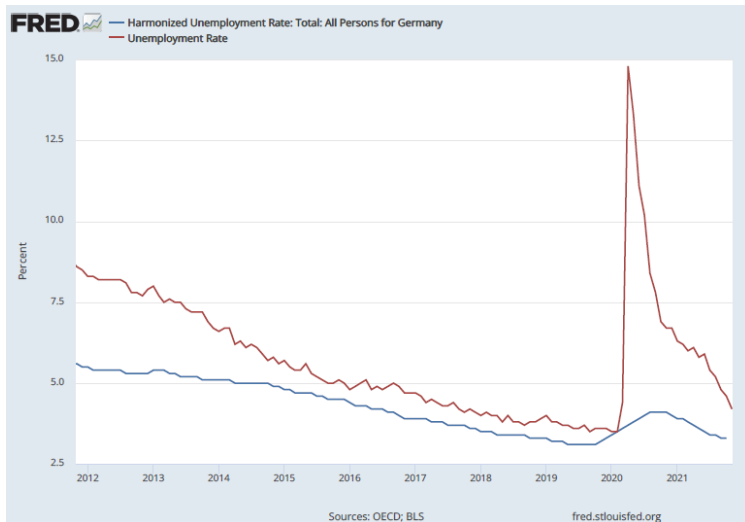
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Motivation

- COVID-19 = a large negative (supply and/or demand) shock
 - restrictions on labor supply (Kocherlakota, 2020)
 - temporary decline in productivity (Gregory et al. 2020)
 - infection shock (Kapicka and Rupert, 2020)
 - skill loss shock hitting unemployed (Jackson and Ortego-Marti, 2020)
 - job loss (Bernstein et al. 2020)
 - preference shock (Ravenna and Walsh, 2021)
- Differential government fiscal policies across countries
 - *U.S.*:
 - mostly transfers via the unemployment insurance system
 - *Europe*:
 - mostly transfers to firms to support employers-employees

Stylized Fact



Previous COVID-19 Heterogeneous-Agent Literature

- **Previous COVID-19 literature with heterogeneous agents (HA)**
 - Bayer, Born, Luetticke and Müller (2020): *HANK*
 - Gregory, Menzio and Wiczer (2020): *HA + search*
 - Ravenna and Walsh (2021): *HANK + search*
 - Guerrieri et al. (2020): *HANK + multi-sector*
 - steady state; all shocks are MIT; no calibrated exercise

This paper: quantifies the effects of COVID-19 policies in a calibrated, *multi-sector*, large-scale *HANK* model with *search*

HANK Model with Directed Search

- **Households**
 - Two types of consumption goods (imperfect substitutes)
 - Two types of assets: liquid (bonds) and illiquid assets
 - Two borrowing constraints, one for each asset
 - Idiosyncratic shocks to productivity level
 - Directed job search
 - Submarkets indexed by workers' characteristics
- **Two production sectors**
 - Fixed shares of population in each sector
 - Downward-sloping demand
 - Sticky prices a la Rotemberg
 - TFP levels subject to sector-specific shocks
 - CRS with capital and labor
 - One sector is COVID vulnerable
- **Labor firm**
 - Posts a job in the chosen submarket
 - Exogenous separation
 - Free entry for firms

HANK Model with Directed Search (Cont.)

- **Mutual fund**
 - One asset: capital
 - Subdivision of capital between two sectors
 - Owns final good producers and job posting firms
- **Fiscal policy**
 - Balanced budget
- **Automatic stabilizers**
 - Unemployment benefits
 - Progressive income tax
 - Food stamps
- **Discretionary policy**
 - Government transfers to households (U.S.)
 - Government transfers to labor firms (Europe)
- **Monetary policy**
 - Taylor rule with ZLB

Previous COVID-19 HA Literature versus Present Paper

- **Previous literature:**
 - On-the-job search without capital
 - No aggregate risk
 - Low-order perturbation
 - State space includes a small number of distributional moments
- **Present paper:** addresses the above limitations by using deep learning
 - On-the-job search with two assets – capital and bonds
 - Aggregate risk shocks in the solution procedure
 - COVID-19 – a sector-specific TFP shock
 - (discretionary fiscal policy as MIT shocks)
 - Global nonlinear solutions
 - State space includes variables of **all** agents

Deep Learning Analysis of Maliar, Maliar, Winant (2019)

1. **HANK model:**
$$\begin{cases} E_{\epsilon} [f_1 (X (s), \epsilon)] = 0 \\ \dots \\ E_{\epsilon} [f_n (X (s), \epsilon)] = 0 \end{cases}$$

s = state, $X (s)$ = decision function, ϵ = innovations.

2. Parameterize $X (s) \simeq \varphi (s; \theta)$ with a **deep neural network**.
3. Construct **objective function** for DL training

$$\min_{\theta} (E_{\epsilon} [f_1 (\varphi (s; \theta), \epsilon)])^2 + \dots + (E_{\epsilon} [f_n (\varphi (s; \theta), \epsilon)])^2 \rightarrow 0$$

4. **All-in-one expectation** operator is a critical novelty:

$$(E_{\epsilon} [f_j (\varphi (s; \theta), \epsilon)])^2 = E_{(\epsilon_1, \epsilon_2)} [f_j (\varphi (s; \theta), \epsilon_1) \cdot f_j (\varphi (s; \theta), \epsilon_2)]$$

with ϵ_1, ϵ_2 = two independent draws.

5. **Stochastic gradient descent** for training (random grids)
6. Google **TensorFlow** platform – software that leads to ground breaking applications (image, speech recognition, etc).

Household Consumption-Saving Problem

Stage 1. Given \bar{C}_t , choose a vector (c_t^X, c_t^Z) maximizing total consumption

$$\begin{aligned} \max_{\{c_t^X, c_t^Z\}} & \left[(1 - \varepsilon)^{\frac{1}{\eta}} (c_t^X)^{1 - \frac{1}{\eta}} + \varepsilon^{\frac{1}{\eta}} (c_t^Z)^{1 - \frac{1}{\eta}} \right]^{\frac{\eta}{1 - \eta}} \\ \text{s.t.} & P_t^X c_t^X + P_t^Z c_t^Z \leq \bar{C}_t, \end{aligned}$$

Stage 2. Given optimal (c_t^X, c_t^Z) , choose a utility-maximizing combination of consumption and assets subject to

$$(1) \quad a' = (1 + r^m) a + i$$

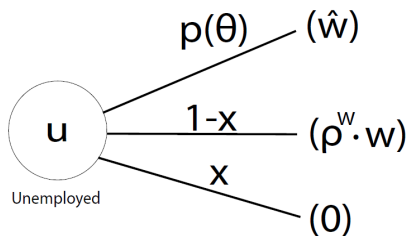
$$(2) \quad \frac{b'}{R} = b + y_{min} + (1 + r^m) a + \varpi - c + \Psi(i, a)$$

$$(3) \quad a' \geq 0, \quad b' \geq -\bar{b}$$

- $\varpi = \mathcal{W}^\nu \times \exp(\eta_\ell - \bar{\eta}_\ell) \times w$ if employed where ν = sector, \mathcal{W} = efficiency wage, w = wage share and $\varpi_t = b_t^{ump}$ if unemployed

Unemployed Search Decision

Agent j searches in a submarket $(a'(j), b'(j), \hat{w}, \eta_e(j), \nu(j), \Sigma)$



θ : market tightness;

$p(\theta)$: job finding rate;

x : probability of unemployment insurance expiring

Unemployed Search Decision

- Bellman equation:

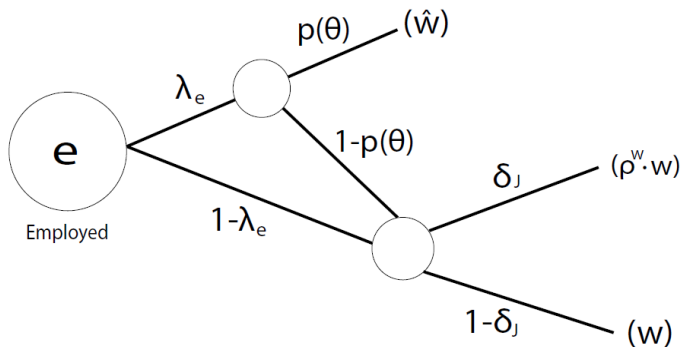
$$Q(w(j), \epsilon(j) = u, \dots) = \max_{\hat{w}} \{ p(\theta) EV(\hat{w}, \epsilon'(j) = e, \dots) + \\ \{ (1 - \chi) EV(\rho^w w(j), \epsilon'(j) = u, \dots) \\ + \chi EV(0, \epsilon'(j) = u, \dots) \} \times (1 - p(\theta)) \}$$

subject to

$$q(\theta) \beta E[J(\hat{w}, \dots)] = k$$

- $\theta(a'(j), b'(j), \hat{w}, \eta_\ell(j), \nu(j), \Sigma)$
- $q(\theta)$: job filling rate
- k : flow cost of posting a vacancy

Employed Search Decision



λ_e : probability of being chosen to look for a better job;

δ_J : probability of being hit by an exogenous separation shock and becomes unemployed

Employed Search Decision

- Bellman equation:

$$Q(w(j), \epsilon(j) = e, \dots) = \max_{\hat{w}} \{ \lambda_e p(\theta) EV(\hat{w}, \epsilon'(j) = e, \dots) + \{ (1 - \delta_J) EV(w(j), \epsilon'(j) = e, \dots) + \delta_J EV(\rho^w w(j), \epsilon'(j) = u, \dots) \} \times (1 - \lambda_e p(\theta)) \}$$

subject to

$$q(\theta) \beta E[J(\hat{w}, \dots)] = k$$

Two Production Sectors

- Technology of a firm in sector ν

$$Y_t^\nu = \exp(\eta_{\theta,t}^\nu - \bar{\eta}_\theta^\nu) (K_{t-1}^\nu)^\alpha (H_t^\nu)^{1-\alpha}$$

H_t^ν : efficiency labor; K_{t-1}^ν : capital; $\exp(\eta_{\theta,t}^\nu)$: TFP level

- AR(1) process for TFP

$$\eta_{\theta,t}^\nu = \rho^{\theta,\nu} \eta_{\theta,t-1}^\nu + \sigma_\theta^\nu \varepsilon_{\theta,t}^\nu, \quad \varepsilon_{\theta,t}^\nu \sim \mathcal{N}(0, 1)$$

- Downward sloping demand

$$Y_t^\nu = \varepsilon \left(\frac{P_t^\nu}{P_t} \right)^{-\eta} Y_t$$

- Profit

$$(1 - \tau_t) \frac{P_t^\nu}{P_t} Y_t^\nu - mc_t^\nu Y_t^\nu$$

Job Posting Firm

- Profit

$$D_t^{J,\nu} = \mathcal{W}_t^\nu \exp(\eta_{\ell,t} - \bar{\eta}_\ell) (1 - w_t) + \tau_t^J$$

- The value of a firm

$$J(a, b, w, \eta_\ell, \nu, \Sigma) = D^{J,\nu} + (1 - \delta_J) (1 - \lambda_e p(\theta)) \beta EJ(a', b', w, \eta'_\ell, \nu, \Sigma')$$

- In equilibrium,

$$q(\theta) \beta EJ(a, b, w, \eta_\ell, \nu, \Sigma) = k$$

Government

- A balanced government budget in each period

$$\int \mathcal{W}_t^{\nu(j)} \exp(\eta_{\ell,t}(j)) \rho^w w_t(j) \mathcal{I}_{\{\epsilon(j)=u\}} dj + y_{\min} \\ + \int \tau^J(j) \mathcal{I}_{\{\epsilon(j)=u\}} dj = \int_h \tau_t Y_t^X(h) dh + \int_h \tau_t Y_t^Z(h) dh$$

- y_{\min} : transfers to households; τ^J : transfers to firms
- We plan to extend to unbalanced budget

Impulse Response Functions

- Koop's et al. (1996) methodology
- In the figures,
 - "Relative":

$$\frac{X^{innovation} - X^{no_innovation}}{X^{no_innovation}} \times 100$$

- "Absolute":

$$\frac{X^{innovation} - X^{no_innovation}}{\text{number of simulations}} \times 100$$

Impulse Response Functions (IRF)

- **No-innovation** series:

"No TFP shock in COVID-vulnerable sector" plus "No Transfers"

- 3 **innovation** series, each of which has

"A TFP shock in the COVID vulnerable sector"

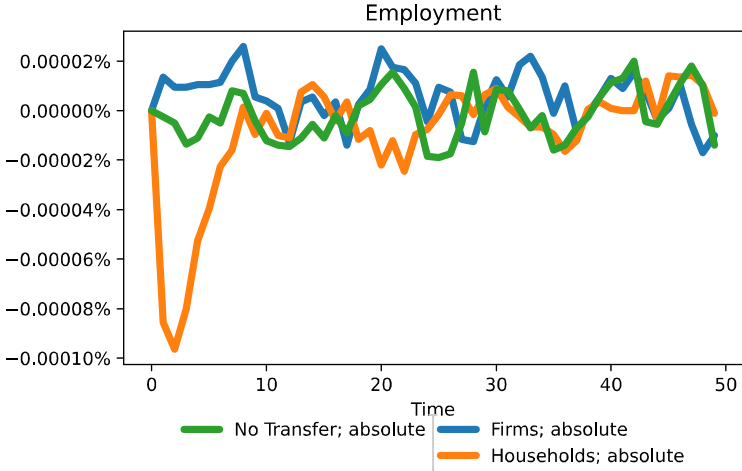
plus

1. *"No Transfer"*: no additional transfers to households or firms
2. *"Households"*: transfers to households
3. *"Firms"*: transfers to firms

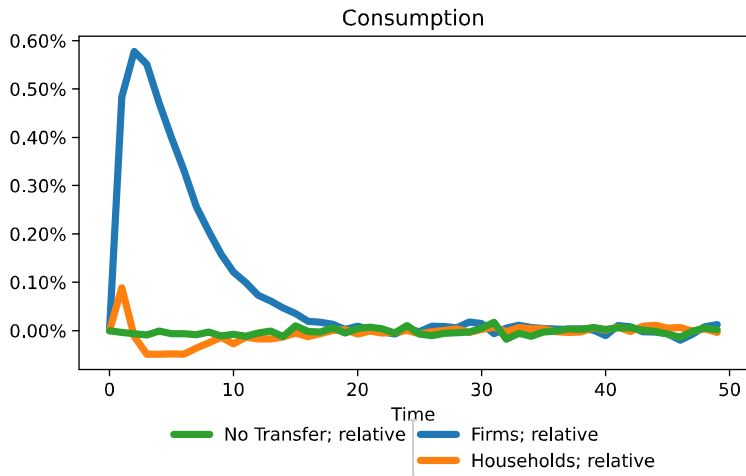
Model's Solution and IRF

- Model's period: 1 month
- 200 agents (100 in each sector)
- "*TFP shock in the COVID vulnerable sector*":
 - a 1 standard deviation negative shock
- "*Households*":
 - y_{min} increases by a factor of 5 in the first period and then decays by a factor of .5 each period
- "*Firms*":
 - τ^J increases from 0 to $4y_{min}$ in the first period and then decays by a factor of .5 each period

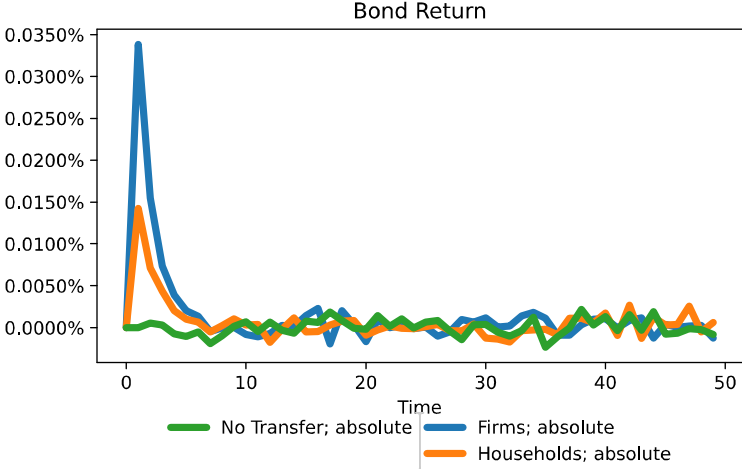
Employment



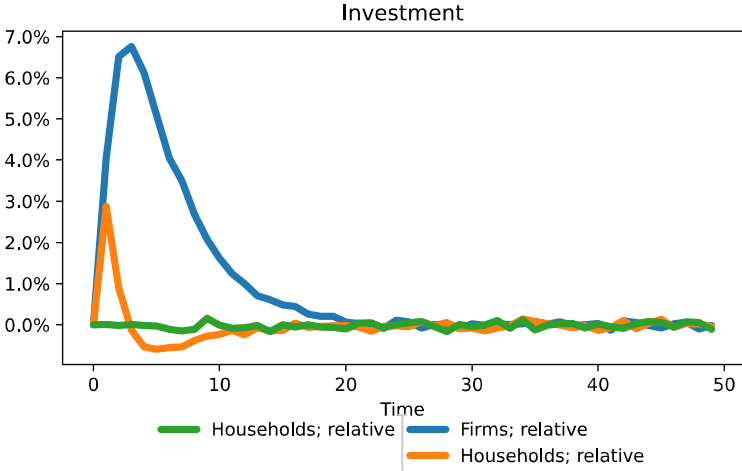
Consumption



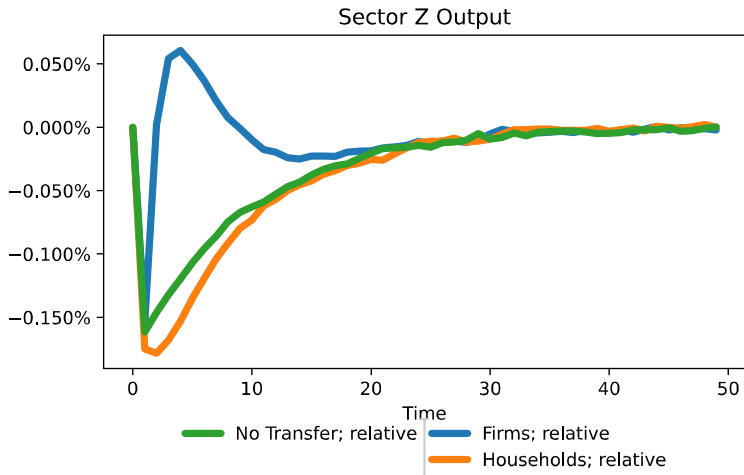
Bond Return



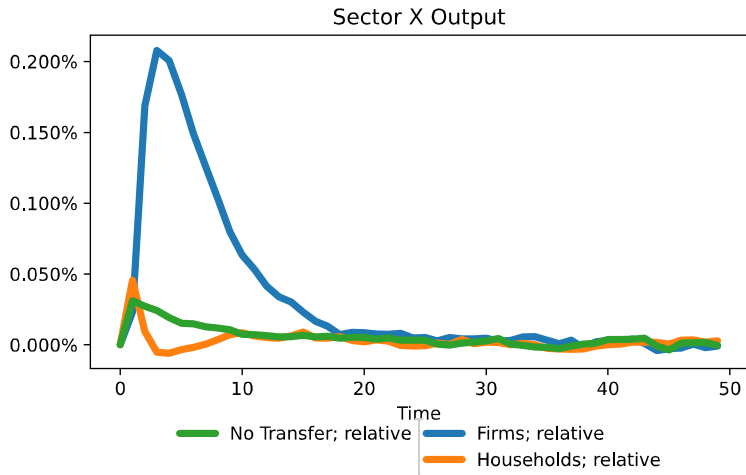
Investment



Output in the COVID Vulnerable Sector



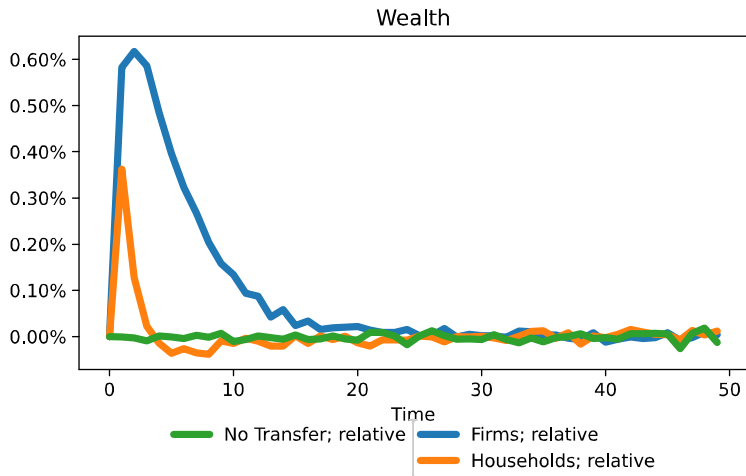
Output in the COVID Robust Sector



Wage Share



Wealth



Conclusion

- The type of transfers used is important for how output responds to TFP shock in one sector:
 - with the firm transfers, output in the COVID-vulnerable sector starts increasing almost immediately, after the initial drop
- The responses of employment and wage share differ between firm and household transfers
- All effects are significantly larger for firm transfers than for household transfers
- Preliminary results for the real economy are promising
- Need to solve the full new Keynesian model

Thank you!

Krusell and Smith (1998) versus the Present Paper

- **Krusell and Smith (1998)** use a reduced state space:
 X_i (*variables of agent i , aggregate moments*)
⇒ few state variables
- **The present paper** uses the full state space:
 X_i (*variables of all agents, distributions*)
⇒ **hundreds of state variables**

How do we deal with such a large state space?

1. Neural network automatically performs the model reduction
– it learns to summarize information from many inputs into a smaller set of hidden layers.
2. Neural network deals with ill conditioning
– it learns to ignore collinear and redundant variables.