Perceived Unemployment Risks over the Business Cycle

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Introduction

- Unemployment risk amplifies business cycles fluctuations in state-of-the-art INCOMPLETE-MARKET HA-MACRO MODELS (Bayer et al., 2019; Den Haan et al., 2018; Broer et al., 2021; Graves, 2020)
  1. ex-ante channel: fears of unemployment $\rightarrow$ precautionary saving $\rightarrow$ consumption $\downarrow$
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- Standard approach does not distinguish (a) perceived risk, (b) true risk, (c) realized outcome
  - full-information-rational-expectations (FIRE) assumes (a) perceived risk = (b) true risk
  - empirical implementation assumes (b) true risk = (c) realized outcome

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  - empirical implementation assumes (b) true risk = (c) realized outcome

- This paper aims to:
  - measure these three conceptually different objects
  - quantify the consumption response to unemployment risks and decompose it into (a), (b), (c)

This paper

1. Separately measure
   (a) ex-ante perceived risk: survey expectations in Survey of Consumer Expectations
   (b) ex-ante true risk: real-time machine-efficient forecasts à la Bianchi et al. (2022)
   (c) ex-post realized outcome: observed transition rates in Current Population Survey

   for job-finding rate and separation rate (the flow approach to unemployment)

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2. Plug into the workhorse heterogeneous-agent model with unemployment risk and quantify consumption response to unemployment risk due to
   - precautionary behavior from (a)
   - ex-post response from (c)
   - **under**/over insurance due to misperception (a) − (b)

Data

Perceived and realized transition rates

- Realized job-finding and separation rates from Current Population Survey:

$$JF_t = \frac{UE_t}{U_{t-1}}, \quad JS_t = \frac{EU_t}{E_{t-1}}$$

where gross flows from U to E and E to U are measured using CPS panel dimension

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- Perceived job-finding and separation rates from Survey of Consumer Expectations:
  - $\widetilde{JF}_t$: *"Suppose you were to lose your main job this month, what do you think is the percent chance that you will find a job within the following 3 months?"*
  - $\widetilde{JS}_t$: *"What do you think is the percent chance that you will lose your main (for those with multiple jobs) or current (for those with single job) job during the next 12 months?"*

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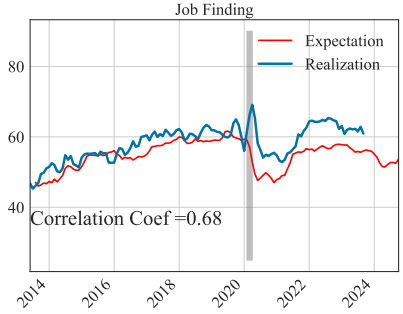
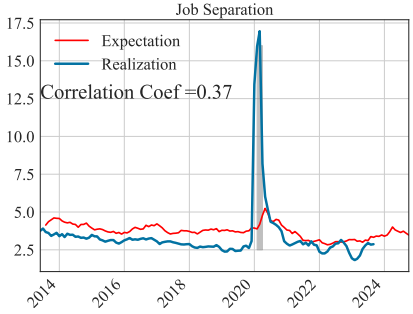
- Time (dis)aggregation from monthly (12-month) to 3-month-horizon rates

Perceived and realized transition rates are highly correlated





- Over 3-month horizon
- A higher correlation if the Covid shock period excluded
- Suggesting that perceptions do contain predictable future labor market movements
- Such pattern remains within each group: ⏩ by employment status ⏩ by education

Forecast errors of perceived unemployment risks

- To systematically assess the relationship between perceived and realized risks, define

$$\mathrm{FE}^{JF}_{t,t+3} = \widetilde{\mathrm{JF}}_{t+3|t} - \mathrm{JF}_{t,t+3}$$

  - $\widetilde{\mathrm{JF}}_{t+3|t}$ represents the perceived job-finding rate for 3 months ahead at time t
  - $\mathrm{JF}_{t,t+3}$ is the realization over the same horizon

- To test informational efficiency of perceived risks Coibion and Gorodnichenko (2015); Fuhrer (2018); Coibion et al. (2018)

$$\mathrm{FE}^{JF}_{t,t+3} = \alpha + \beta \mathrm{FE}^{JF}_{t-3,t} + \gamma X_{t-3} + \epsilon_t$$

  - Null hypothesis under FIRE: $\beta = 0$
  - $\beta > 0$: past errors persist into future forecasts, reflecting information rigidity

Auto-regressions of forecast errors (FE) imply perceptions are inefficient

	JF	JF LowEdu	JF MidEdu	JF HighEdu	JS	JS LowEdu	JS MidEdu	JS HighEdu
Constant	-0.027***	-0.027***	-0.038***	-0.024***	0.003*	0.076***	0.079***	0.051***
	(0.004)	(0.007)	(0.005)	(0.004)	(0.002)	(0.009)	(0.010)	(0.009)
lag_FE_jf	0.256***	0.545***	0.272***	0.183**				
	(0.087)	(0.076)	(0.084)	(0.088)				
lag_FE_js					0.131	0.202**	0.267***	0.554***
					(0.091)	(0.089)	(0.088)	(0.075)
Observations	121	124	124	124	121	124	124	124
R^2	0.068	0.295	0.079	0.034	0.017	0.040	0.070	0.308
Adjusted R^2	0.060	0.289	0.071	0.026	0.009	0.032	0.062	0.302
F Statistic	8.628***	51.049***	10.452***	4.297**	2.062	5.103**	9.197***	54.322***

*p<0.1; **p<0.05; ***p<0.01

Ex-ante Comparison

(Proxy for) true ex-ante transition risk

- Machine-learning efficient forecasts à la Bianchi et al. (2022):
  1. Lasso forecasting model $JF_{t+3|t} = \Gamma^t X_t + \epsilon_t \rightarrow \widehat{\Gamma}^{t*}$ in the 10-year historical sample up to t
  2. Machine-efficient forecast $\widehat{JF}^*_{t+3|t} = \widehat{\Gamma}^{t*} X_t$ as a 3-month-ahead out-of-sample prediction for t

[real time]

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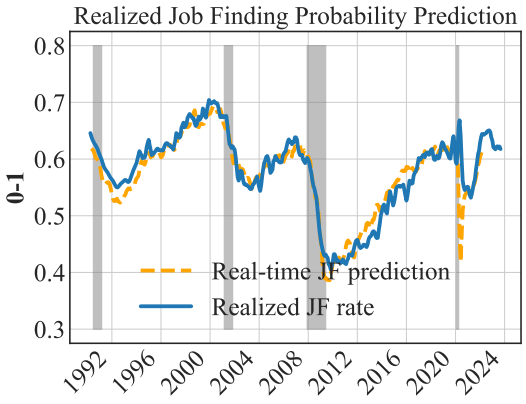
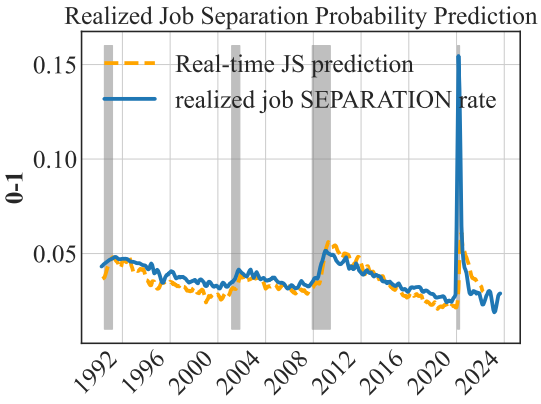
- Data: $600+$ time series
  - Real-time macroeconomic realizations, such as inflation, unemployment rate, GDP growth, etc.
  - Professional forecasts of the macroeconomy from Survey of Professional Forecasters (SPF)
  - Realized worker flow rates
  - Household expectations from Michigan Survey of Consumers (MSC)

[real time]

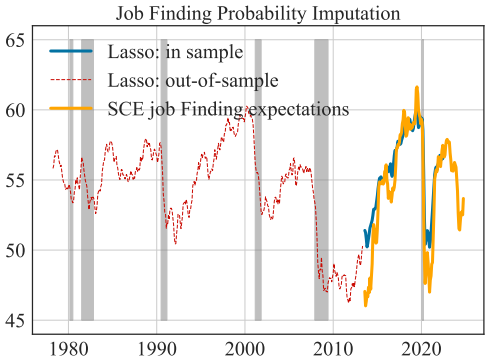
Machine-learning forecast of unemployment risks

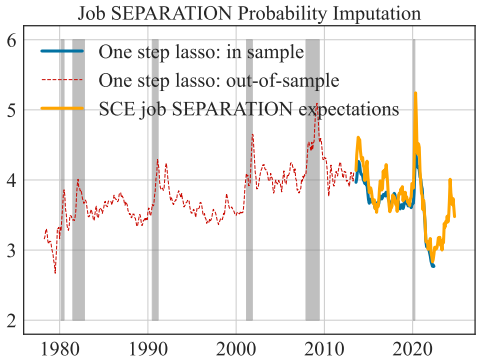




- Expectations in the MSC and real-time UE rate are the most important predictors
  - e.g., income expectations, inflation expectations, news heard, durable/vehicle-buying intentions, household finance expectations, etc.

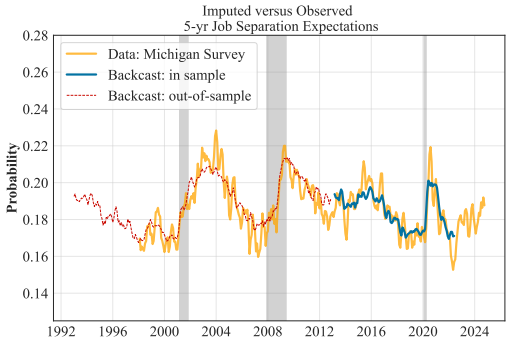
Backcasting beliefs: what were people thinking before the SCE?

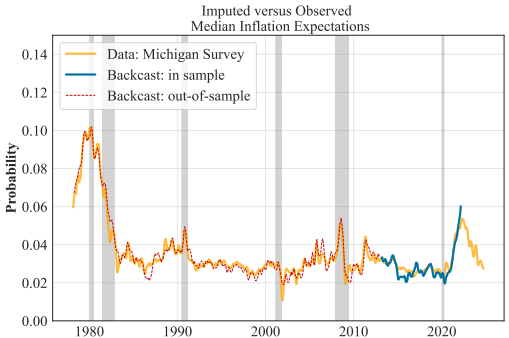




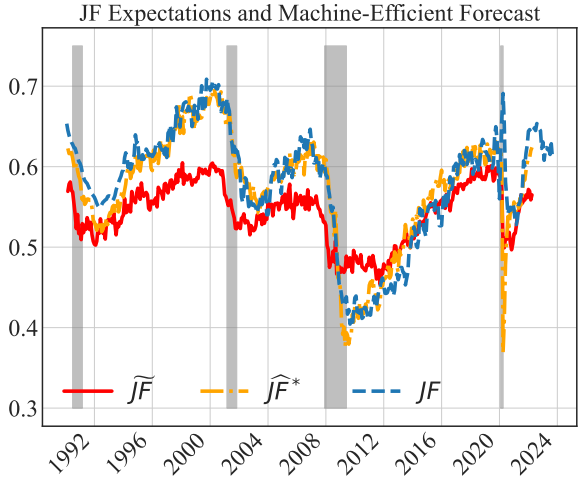
Validating the backcasting method

Imputed Beliefs versus Observed Expectations in the MSC



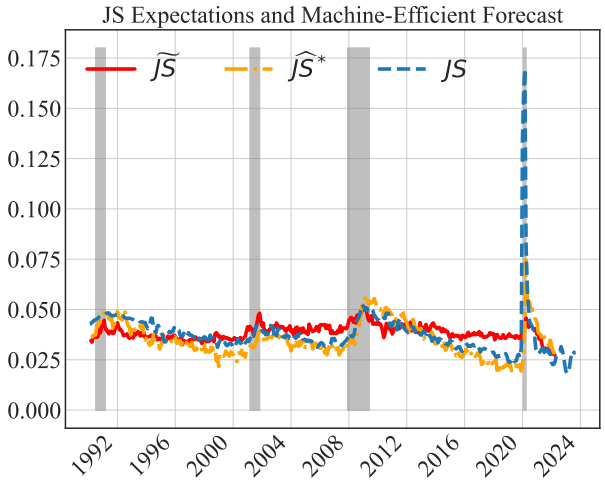


Surveys versus machine: job-finding



$$\log(\widetilde{JF}_{t+3|t}) = 1.92 + \mathbf{0.51}\log(\widehat{JF}^*_{t+3|t}) + \epsilon_t,$$ In-sample

Surveys versus machine: job-separation



$$\log(\widetilde{JS}_{t+3|t}) = 1.13 + \mathbf{0.19}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t,$$

In-sample

Heterogeneity in risks and perceptions

- Repeat the exercise with q-th percentile perceived risks $\widetilde{JF}^q$ and $\widetilde{JS}^q$, $\forall q \in \{0.25, 0.5, 0.75\}$
- Whose expectations react to their real-time unemployment risks the most?

$$\log(\widetilde{JF}^{0.25}_{t+3|t}) = -1.55 + \mathbf{1.22}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t \qquad \log(\widetilde{JS}^{0.25}_{t+3|t}) = -0.42 + \mathbf{0.46}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JF}^{0.5}_{t+3|t}) = 1.54 + \mathbf{0.63}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t \qquad \log(\widetilde{JS}^{0.5}_{t+3|t}) = 1.06 + \mathbf{0.68}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JF}^{0.75}_{t+3|t}) = 3.62 + \mathbf{0.20}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t \qquad \log(\widetilde{JS}^{0.75}_{t+3|t}) = 2.57 + \mathbf{0.27}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

In-sample

Observable heterogeneity: education

$$\log(\widetilde{JF}^{LEdu}_{t+3|t}) = 1.28 + \mathbf{0.66}\log(\widehat{JF}^{*LEdu}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JF}^{MEdu}_{t+3|t}) = 2.53 + \mathbf{0.36}\log(\widehat{JF}^{*MEdu}_{t+3|t}) + \epsilon_t$$

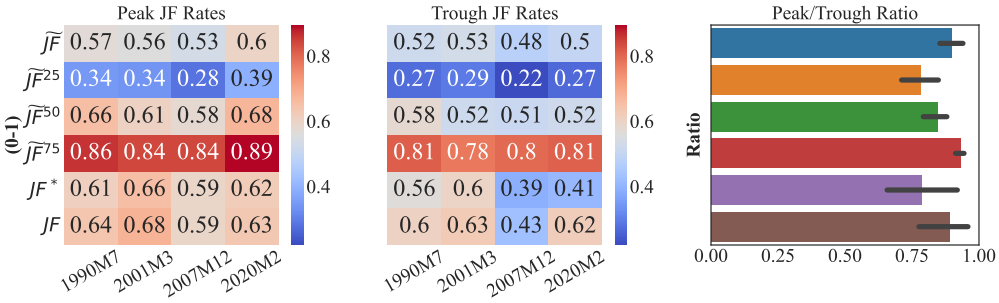
$$\log(\widetilde{JF}^{HEdu}_{t+3|t}) = 1.87 + \mathbf{0.53}\log(\widehat{JF}^{*HEdu}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}^{LEdu}_{t+3|t}) = 1.1 + \mathbf{0.17}\log(\widehat{JS}^{*LEdu}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}^{MEdu}_{t+3|t}) = 0.95 + \mathbf{0.35}\log(\widehat{JS}^{*MEdu}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}^{HEdu}_{t+3|t})) = 1.08 + \mathbf{0.33}\log(\widehat{JS}^{*HEdu}_{t+3|t}) + \epsilon_t$$

Business cycle patterns of risks and perceptions



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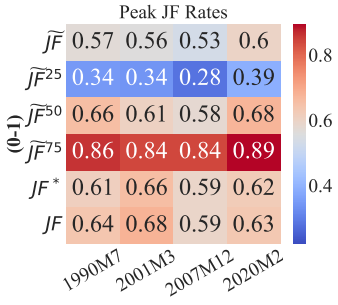
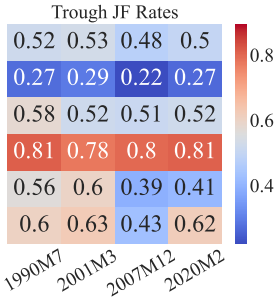
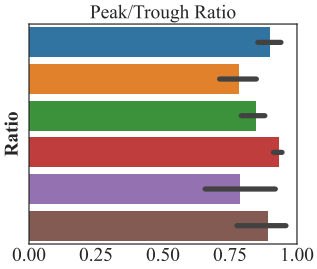
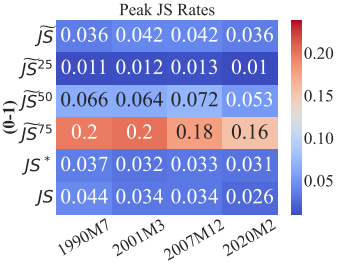
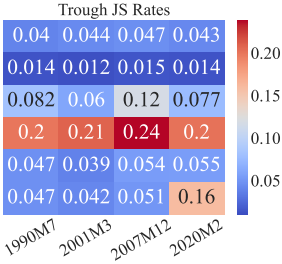
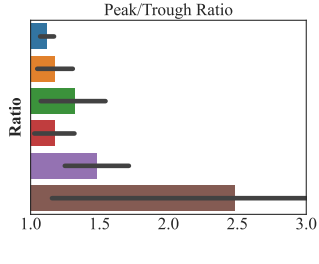








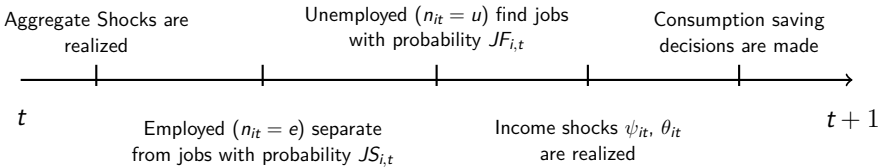




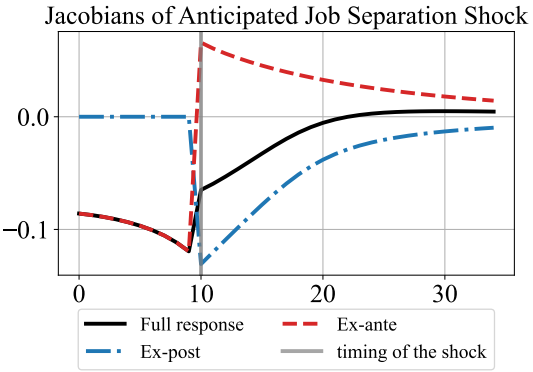
Model

Model elements

- Buffer-stock consumers
- Uninsured idiosyncratic income risks: persistent job spells, persistent+ transitory wage risks
- Zero-Borrowing constraint
- Homogeneous/heterogeneous unemployment risks
- monthly frequency



Decomposition of aggregate consumption response



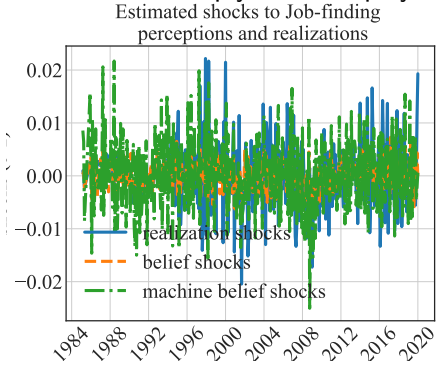
- Sequence-space Jacobian method Auclert et al. (2021)
- Jacobian decomposed into (a) ex-ante precautionary response (b) ex-post shock response

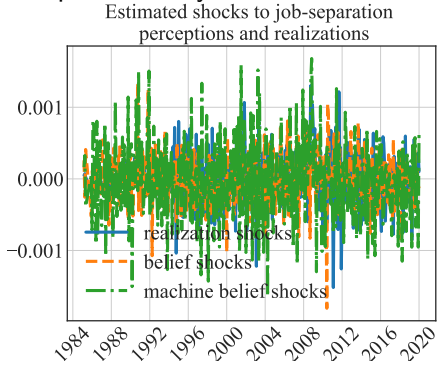
Quantifying consumption impacts of unemployment risks

$$\text{Overal Impact} = \text{Sensitivty} \times (\text{cumulative impacts of}) \text{ shocks}$$

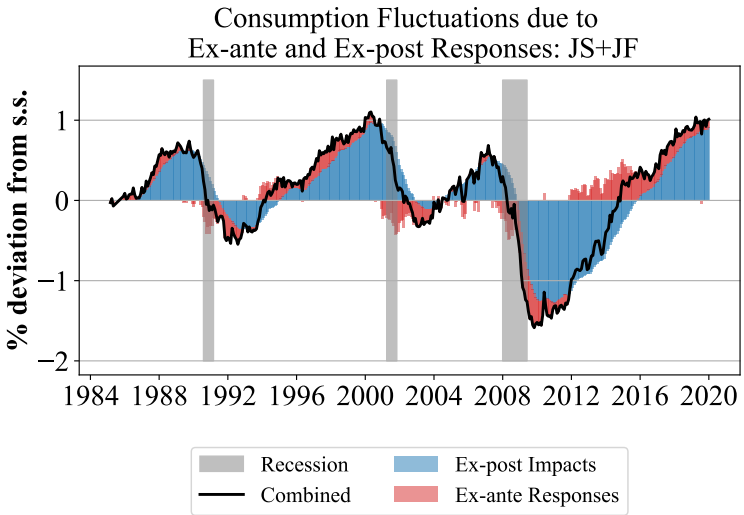
- Simulating the partial equilibrium aggregate consumption deviations from the steady state based on estimated AR(1) and shocks of realized rates, perceived risks, and machine forecast

- Validation: shocks imply an unemployment rate path broadly close to the data

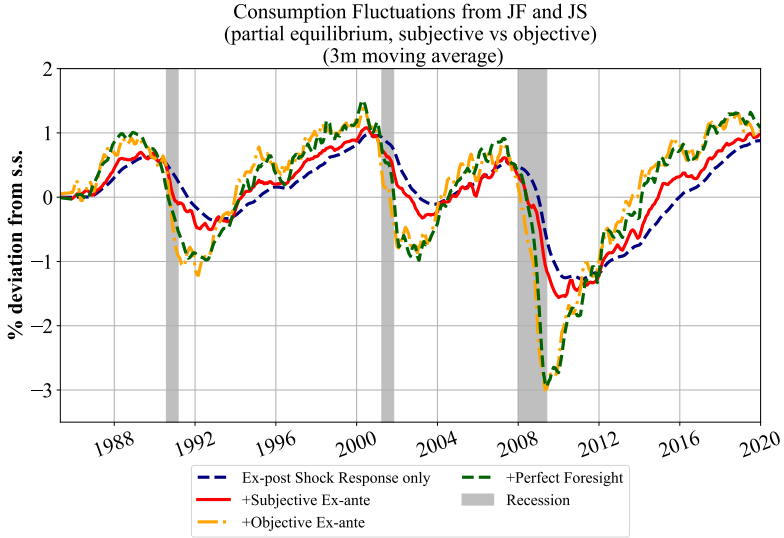




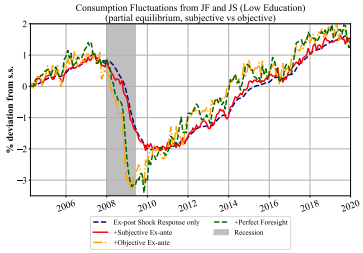
Consumption impacts of unemployment (risks): ex-ante versus ex-post

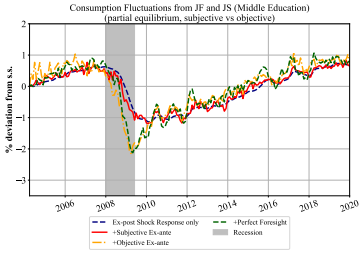


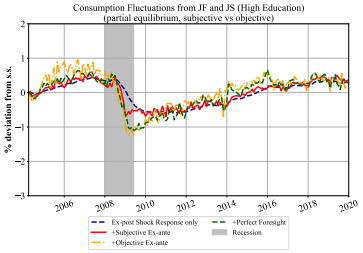
Compared to two counterfactuals



Quantification by education: JS+JF



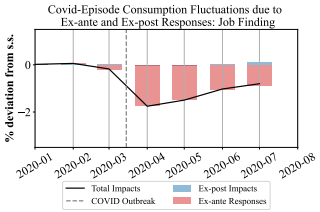




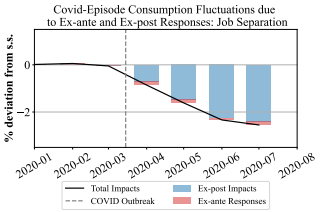
- Calibrated to match education-specific MPCs
- Group with the larger risk exposure has stickier belief, hence more underinsured
- Amplification due to heterogeneous risk exposures + heterogeneous belief distortions

Model-implied consumption dynamics during COVID

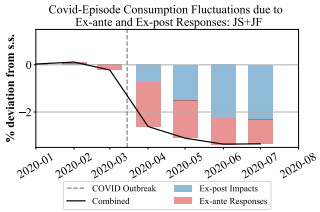
JF



JS



JS+JF



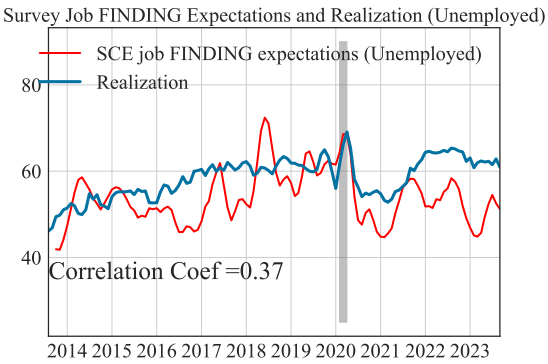
- Perception and realization shocks filtered from pre-COVID persistence

Conclusion

- We quantify the aggregate consumption fluctuations due to
  - Perceived risks → ex-ante responses
  - Realized shocks → ex-post impacts
  - True risks (a counterfactual benchmark as opposed to perceived risks)
- Ex-ante responses are important and sizable in past recessions
- But the **stickiness** of risk perceptions limited the role of precautionary saving motives
- Both risks and perceptions are widely **heterogeneous**
- The correlation pattern of risk exposure and belief distortion as an **amplification** mechanism

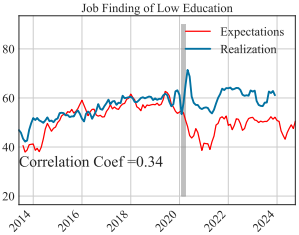
Appendix

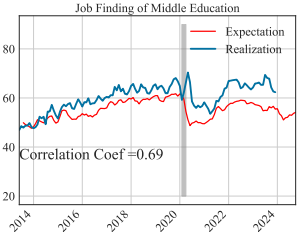
JF perceptions by the unemployed and employed

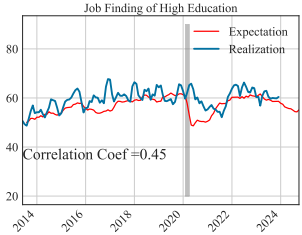


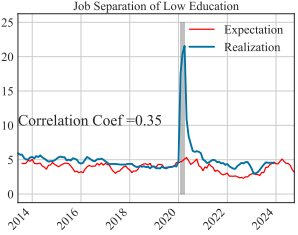
Back

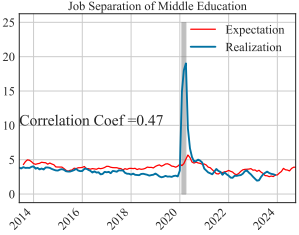
Perceived and realized transition rates remain correlated within education

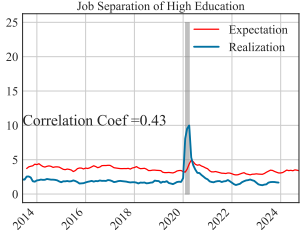












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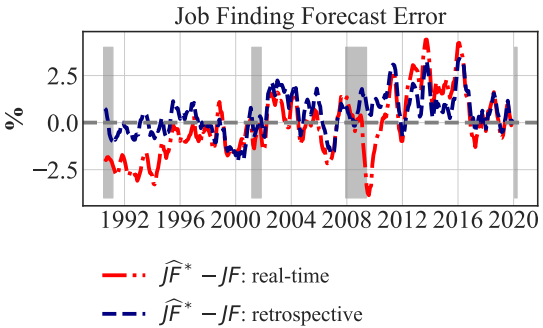
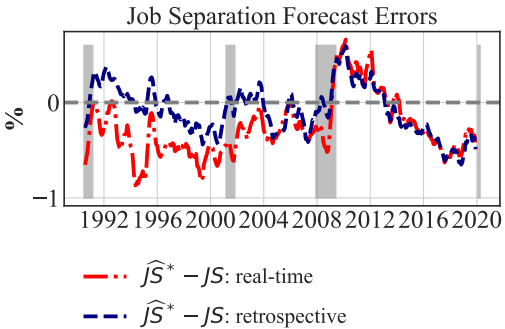
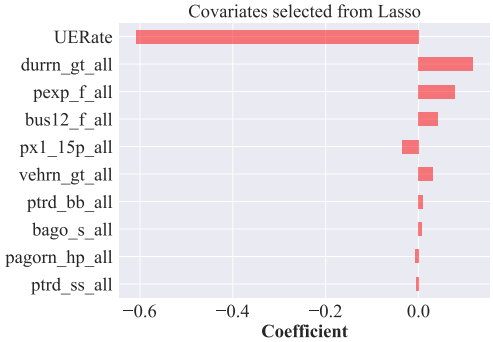
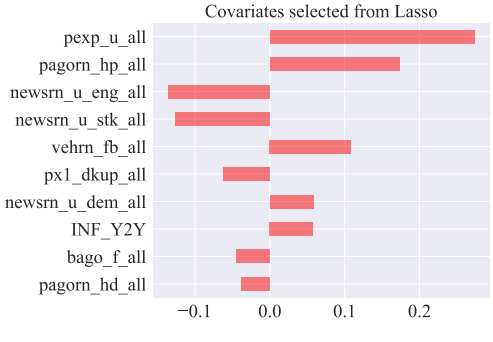
Why is real-time important?





The most important covariates of perceived unemployment risks





- UERate: real-time unemployment rate.
- Durrn_gt_all: good time to buy durables.
- Pexp_f_all: expecting better finance.
- Bus12_f_all: better business conditions.
- Px1_15p_all: expected inflation above 15 percent.
- Vehrn_gt_all: good time to buy vehicles.
- ptrd_bb_all: better off financially now and future.
- bago_s_all: same business conditions.
- Pagorn_hp_all: worse finance due to higher prices.
- Ptrd_ss_all: same personal finance now and future.

- Pexp_u_all: expecting worse personal finance.
- Newsrn_u_eng_all: heard unfavorable news about energy crisis.
- Newsrn_u_stk_all: heard about unfavorable news regarding stock market.
- Vehrn_fb_all: bad time to buy vehicles due to uncertain future.
- Px1_dkup_all: do not know about future inflation.
- Newsrn_u_dem_all: heard unfavorable news about lower consumer demand.
- INF_Y2Y: real-time inflation rate.
- Bago_f_all: better business conditions.
- Pagorn_hd_all: worse personal finance due to higher debt.

Surveys versus machine (2013-2023)

$$\log(\widetilde{JF}_{t+3|t}) = 0.71 + \mathbf{0.81}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JF}^{0.25}_{t+3|t}) = -5.73 + \mathbf{2.26}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JF}^{0.5}_{t+3|t}) = -0.84 + \mathbf{1.22}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JF}^{0.75}_{t+3|t}) = 2.66 + \mathbf{0.44}\log(\widehat{JF}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}_{t+3|t}) = 1.11 + \mathbf{0.14}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}^{0.25}_{t+3|t}) = -0.91 + \mathbf{0.61}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}^{0.5}_{t+3|t}) = 0.12 + \mathbf{0.34}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

$$\log(\widetilde{JS}^{0.75}_{t+3|t}) = 1.40 + \mathbf{0.06}\log(\widehat{JS}^{*}_{t+3|t}) + \epsilon_t$$

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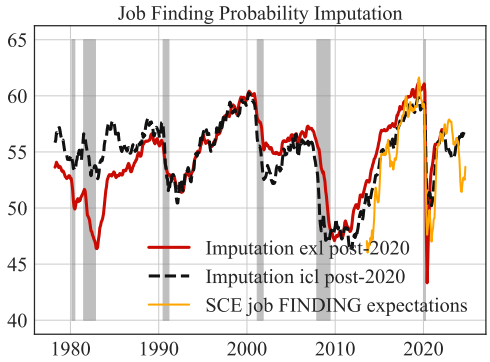
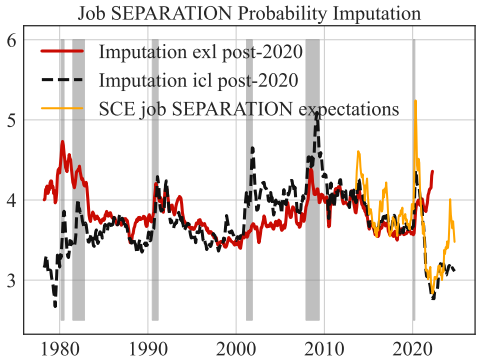
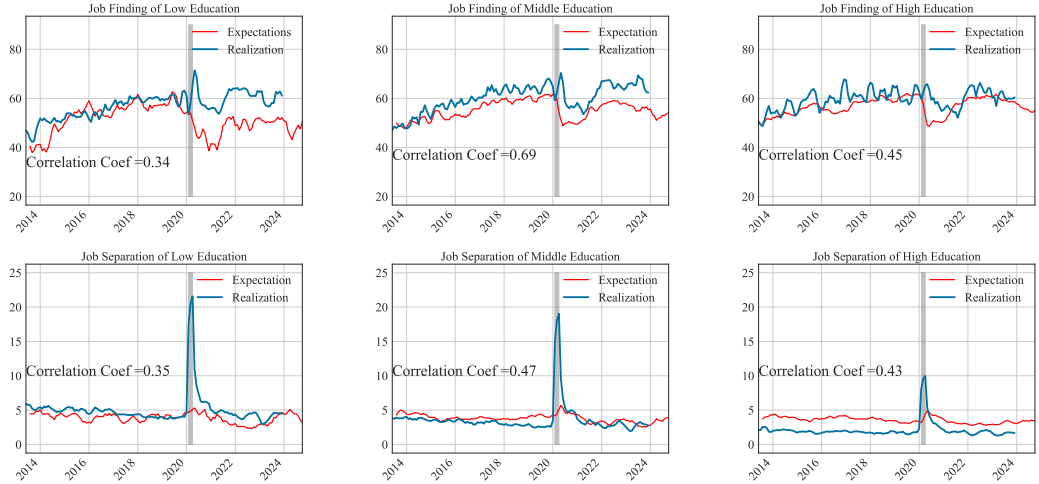
Imputing beliefs including or excluding the Covid era



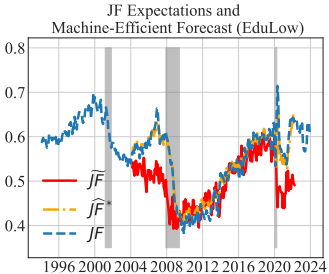


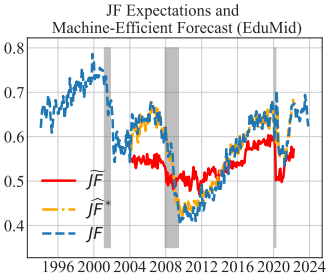
Observable heterogeneity: education

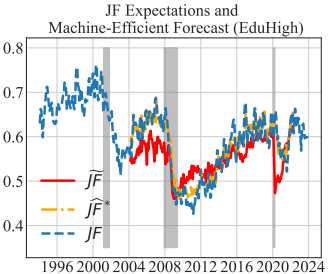


- low education group faces higher separation rate, but perceived separation risks did not go up as much

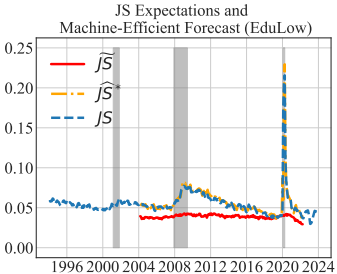
Belief distortions by education: job finding

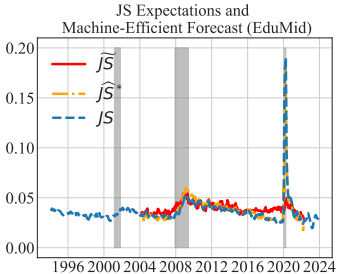


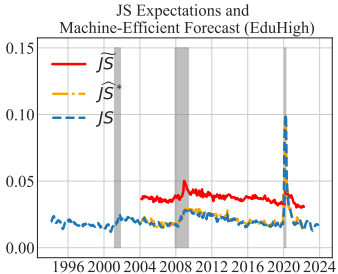




Belief distortions by education: job separation







Household block of the model

$$v_t\left(\mathbf{m}_{it}, e_{it}, n_{it}\right) = \max_{\{\mathbf{c}_{it}, \mathbf{a}_{it}\}} \left\{U\left(\mathbf{c}_{it}\right) + \beta_i(1-D)\mathrm{E}_{\mathrm{t}}\left[v_{t+1}\left(\mathbf{m}_{t+1}, e_{it+1}, n_{it+1}\right)\right]\right\}$$

$$\begin{aligned} s.t. \qquad \mathbf{a}_{it} &= \mathbf{m}_{it} - \mathbf{c}_{it} \\ \mathbf{a}_{it} + \mathbf{c}_{it} &= \mathbf{z}_{it} + (1 + r_t^a)\mathbf{a}_{it-1} \\ \mathbf{a}_{it} &\geq 0 \end{aligned}$$

Household block of the model: income process

Wage

$$\mathbf{z}_{i,t} = e_{i,t}\zeta_{it}$$

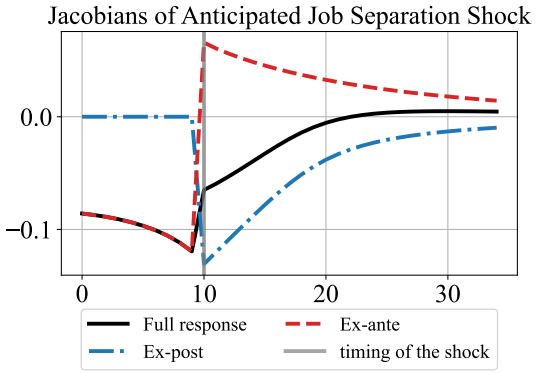
$$\log e_{i,t} = \rho_e \log e_{i,t-1} + \eta_{i,t}, \quad \eta_{i,t} \sim \mathcal{N}(0, \sigma_e^2)$$

$$\zeta_{it} = \begin{cases} \theta_{it}, \text{ if employed : } n_{i,t} = e \\ \\ \theta_{it}\gamma, \text{ if unemployed : } n_{i,t} = u \end{cases}$$

Labor market transitions

$$p(n_{i,t+1} = e | n_{i,t} = u) = JF_{i,t}$$
$$p(n_{i,t+1} = u | n_{i,t} = e) = JS_{i,t}$$

Decomposition of aggregate consumption response



- Sequence-space Jacobian method Auclert et al. (2021)
- Jacobian decomposed into (a) ex-ante precautionary response (b) ex-post shock response
- $\beta = 0.97$ matched to a target average quarterly MPC of 0.21
- UI replacement ratio $\gamma = 0.5$

Calibration

Description	Parameter	Value	Source/Target
CRRA	CRRA	2	Standard
Real Interest Rate	r	$1.05^{\frac{1}{12}} - 1$	5% annualized real rate
UI replacement rate	γ	0.5	50% replacement rate
Persistence of idiosyncratic income process	ρ_e	0.997	Kekre (2023)
Std Dev of idiosyncratic income process	σ_e	0.057	Kekre (2023)
Std Dev of Log Transitory Shock	σ_θ	0.244	Kekre (2023)
Steady state Job-Finding Rate	JF	0.25	CPS
Steady state Job-Separation Rate	JS	0.017	CPS
Discount Factor	β	0.988	Quarterly MPC $= 0.21$

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