Perceived Unemployment Risks over Business Cycles

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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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 - 2. ex-post channel: realized unemployment \rightarrow reduced income \rightarrow consumption \downarrow
- 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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 - full-information-rational-expectations (FIRE) assumes (a) perceived risk = (b) true risk
 - 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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 - (c) ex-post realized outcome: observed transition rates in Current Population Survey for job-finding rate and separation rate (the flow approach to unemployment)
- 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?"

Perceived and realized transition rates

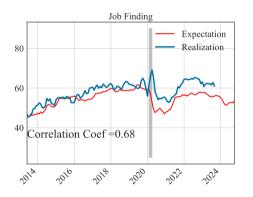
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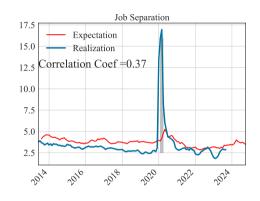
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- Time (dis)aggregation from monthly (12-month) to 3-month-horizon rates

Perceived and realized transition rates are highly correlated

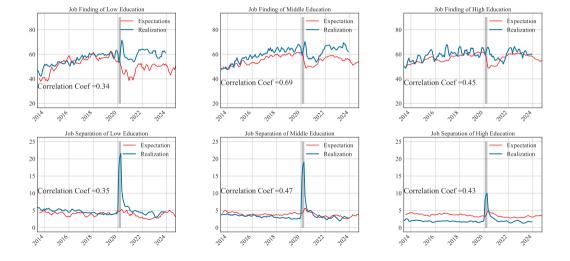




- In a 3-month horizon
- Even higher correlation without the pandemic
- Suggesting that perceptions do contain predictable future labor market movements

[employment status] 4 / 25

Perceived and realized transition rates remain correlated within education



Forecast errors of perceived unemployment risks

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

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

- $\widetilde{\mathsf{JF}}_{t+3|t}$ represents the perceived job-finding rate for 3 months ahead at time t
- $JF_{t,t+3}$ is the realization over the same horizon
- To test informational efficiency of perceived risks Coibion and Gorodnichenko (2015); Fuhrer (2018); Coibion et al. (2018)

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

- 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**	, ,	, ,	, ,	, ,
<u> </u>	(0.087)	(0.076)	(0.084)	(880.0)				
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 $JF_{t+3|t}=\Gamma^tX_t+\epsilon_t\to \widehat{\Gamma}^{t*}$ in the 10-year historical sample up to t
 - 2. Machine-efficient forecast $\widehat{\mathit{JF}}_{t+3|t}^* = \widehat{\Gamma}^{t*} X_t$ as a 3-month-ahead out-of-sample prediction for t

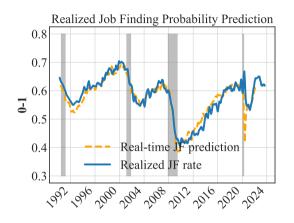
[real time] 8/25

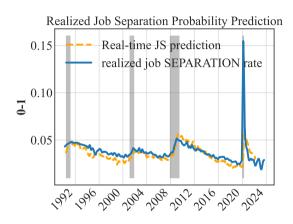
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- Data: 600+ time series
 - Real-time macroeconomic realizations, such as inflation, unemployment rate, GDP growth, etc.
 - Professional forecasts of the macroeconomy from Survey of Professional Forecasters (SPF)
 - Realized worker flow rates
 - Household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

[real time] 8 / 25

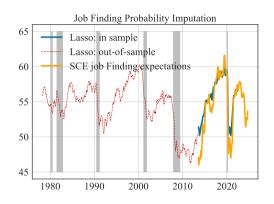
Machine-learning forecast of unemployment risks

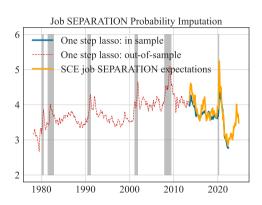




- Expectations in MSC and real-time UE rate are the most important predictors
 - e.g., income expectations, inflation expectations, news heard, durable/vehicle-buying intentions, household finance expectations, etc.

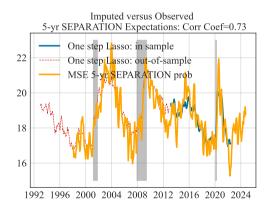
Backcasting beliefs: what were people thinking before SCE?

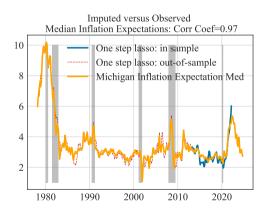




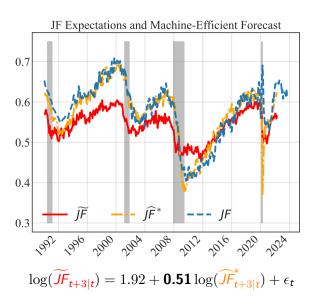
Validating the backcasting method

Imputed Beliefs versus Observed Expectations in MSC

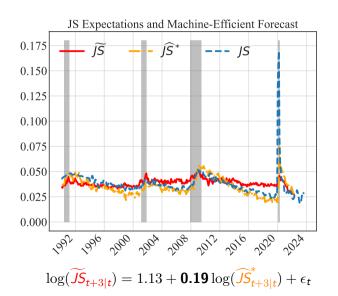




Surveys versus machine: job-finding



Surveys versus machine: job-separation



Heterogeneity

$$\begin{aligned} & \mathsf{JF}_{i,t} = \eta_{i,t} \mathsf{JF}_t \\ & \widetilde{\mathsf{JF}}_{i,t} = \mathbb{E}_i (\mathsf{JF}_{i,t}) = \mathbb{E}_i (\eta_{i,t} \mathsf{JF}_t) \\ & \widetilde{\mathsf{JF}}_t = \frac{\sum \mathbb{E}_i (\mathsf{JF}_{i,t})}{N} \stackrel{?}{=} \mathsf{JF}_t \end{aligned}$$

- Depends on distribution of $\eta_{i,t}$ and the shape of $\mathbb{E}_i()$
- Mueller et al. (2021): ex-ante heterogeneity + underreaction to variations in JF across workers and over unemployment spells \rightarrow underinsurance \rightarrow amplified dispersion in JF rates
- This paper: underreaction to variations in unemployment risks over business cycles

Distribution of unemployment risks

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

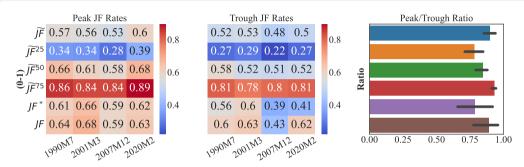
$$\begin{split} \log(\widetilde{\mathit{JF}}_{t+3|t}^{0.25}) &= -1.55 + \mathbf{1.22} \log(\widehat{\mathit{JF}}_{t+3|t}^*) + \epsilon_t \\ &\log(\widetilde{\mathit{JS}}_{t+3|t}^{0.5}) = -0.42 + \mathbf{0.46} \log(\widehat{\mathit{JS}}_{t+3|t}^*) + \epsilon_t \\ \log(\widetilde{\mathit{JF}}_{t+3|t}^{0.5}) &= 1.54 + \mathbf{0.63} \log(\widehat{\mathit{JF}}_{t+3|t}^*) + \epsilon_t \\ \log(\widetilde{\mathit{JF}}_{t+3|t}^{0.75}) &= 3.62 + \mathbf{0.20} \log(\widehat{\mathit{JF}}_{t+3|t}^*) + \epsilon_t \\ \end{split} \\ \log(\widetilde{\mathit{JF}}_{t+3|t}^{0.25}) &= 2.57 + \mathbf{0.27} \log(\widehat{\mathit{JS}}_{t+3|t}^*) + \epsilon_t \end{split}$$

Observable heterogeneity: education

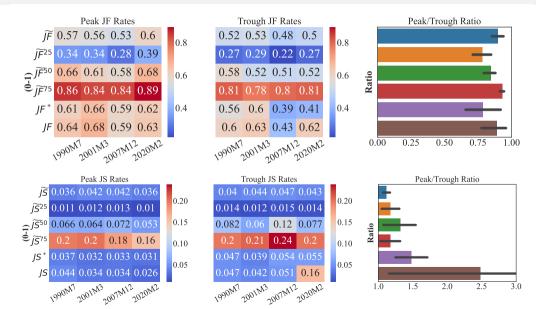
$$\begin{split} \log(\widetilde{\mathit{JF}}_{t+3|t}^{\mathit{LEdu}}) &= 1.28 + \textbf{0.66} \log(\widehat{\mathit{JF}}_{t+3|t}^{\mathit{*LEdu}}) + \epsilon_t \\ \log(\widetilde{\mathit{JF}}_{t+3|t}^{\mathit{MEdu}}) &= 2.53 + \textbf{0.36} \log(\widehat{\mathit{JF}}_{t+3|t}^{\mathit{*MEdu}}) + \epsilon_t \\ \log(\widetilde{\mathit{JF}}_{t+3|t}^{\mathit{HEdu}}) &= 1.87 + \textbf{0.53} \log(\widehat{\mathit{JF}}_{t+3|t}^{\mathit{*HEdu}}) + \epsilon_t \end{split}$$

$$\begin{split} \log(\widetilde{JS}_{t+3|t}^{LEdu}) &= 1.1 + \mathbf{0.17} \log(\widehat{JS}_{t+3|t}^{*LEdu}) + \epsilon_t \\ \log(\widetilde{JS}_{t+3|t}^{MEdu}) &= 0.95 + \mathbf{0.35} \log(\widehat{JS}_{t+3|t}^{*MEdu}) + \epsilon_t \\ \log(\widetilde{JS}_{t+3|t}^{HEdu})) &= 1.08 + \mathbf{0.33} \log(\widehat{JS}_{t+3|t}^{*HEdu}) + \epsilon_t \end{split}$$

Business cycle patterns of risks and perceptions



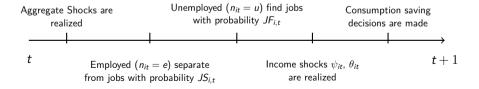
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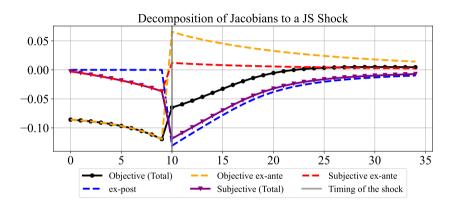
Model

Model elements

- Buffer-stock consumers
- Uninsured idiosyncratic income risks: persistent job spells, persistent+ transitory wage risks
- No-borrowing constraints
- Homogeneous/heterogeneous unemployment risks
- monthly frequency



Decomposition of Subjective Consumption Jacobians



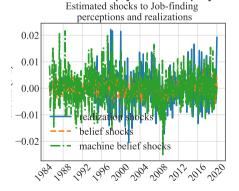
• Subjective under-perceiving true risks \rightarrow underinsurance ex-ante and sharper post-shock drop in consumption

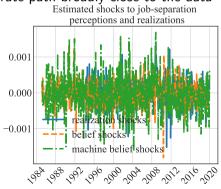
Quantifying consumption impacts of unemployment risks

Overal Impact = Sensitivty \times (cumulative impacts of) shocks

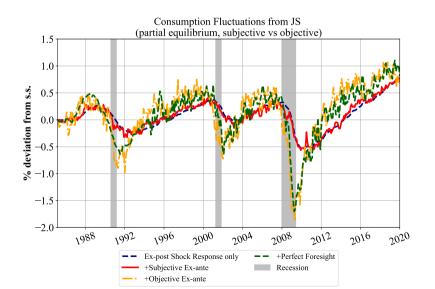
• Simulating the partial equilibrium aggregate consumption deviations from the steady state based on estimated AR(1) and shocks of realized rates, perceived risks, and machine forecast

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

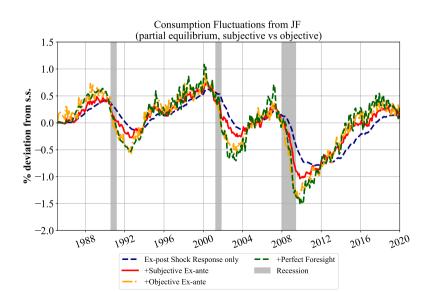




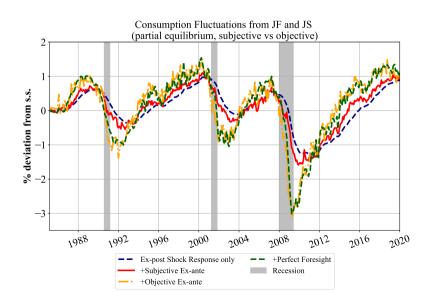
Consumption impacts of unemployment risks: job separation



Consumption impacts of unemployment risks: job finding



Consumption impacts of unemployment risks: separation+finding



Quantification by education: JS+JF



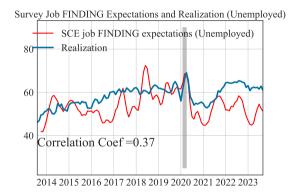
- Group with the larger risk exposure has stickier belief, hence more underinsured
- Amplification due to heterogeneous risk exposures + heterogeneous belief distortions

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

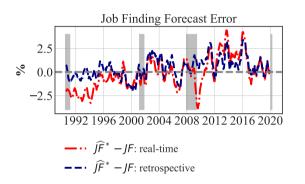
Appendix

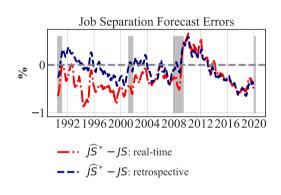
JF perceptions by the unemployed and employed



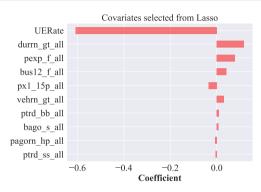


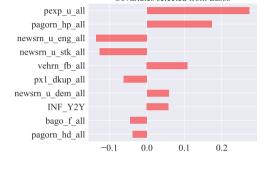
Why is real-time important?





The most important covariates of perceived unemployment risks





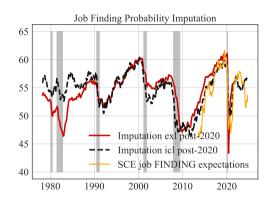
- UERate: real-time unemployment rate.
- Durrn gt all: good time to buy durables.
- Pexp f all: expecting better finance.
- Bus12 f all: better business conditions.
- Px1 15p all: expected inflation above 15 percent.
- Vehrn gt all: good time to buy vehicles.
- ptrd bb all: better off financially now 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.

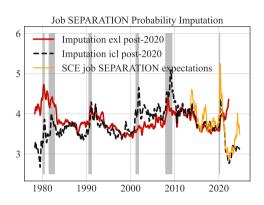
- Pexp_u_all: expecting worse personal finance.
- Newsrn u eng all: heard unfavorable news about energy crisis.
- Newsrn u stk all: heard about unfavorable news regarding stock market.

Covariates selected from Lasso

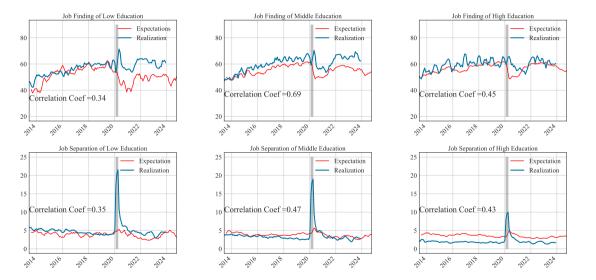
- Vehrn fb all: bad time to buy vehicles due to uncertain future.
- Px1_dkup_all: do not know about future inflation.
- Newsrn u dem all: heard unfavorable news about lower consumer demand
- INF Y2Y: real-time inflation rate.
- Bago f all: better business conditions.
- Pagorn hd all: worse personal finance due to higher debt.

Imputing beliefs including or excluding the Covid era



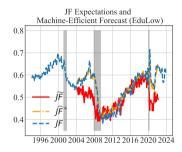


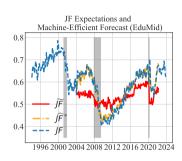
Observable heterogeneity: education

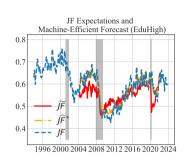


In recessions, low-educ workers lower perceived job finding more than high-educ ones but the

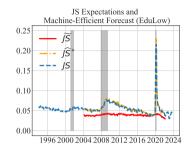
Belief distortions by education: job finding

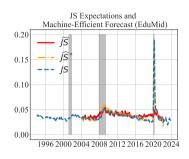


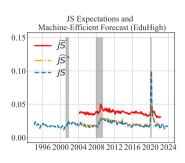




Belief distortions by education: job separation







Household block of the model

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

$$s.t. \quad \mathbf{a}_{it} = \mathbf{m}_{it} - \mathbf{c}_{it}$$

$$\mathbf{a}_{it} + \mathbf{c}_{it} = \mathbf{z}_{it} + (1 + r_{t}^{a})\mathbf{a}_{it-1}$$

$$\mathbf{a}_{it} > 0$$

Household block of the model: income process

Wage

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

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

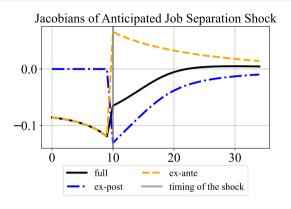
$$\zeta_{it} = \begin{cases} \theta_{it}, & \text{if employed} : n_{i,t} = e \\ \\ \theta_{it}\gamma, & \text{if unemployed} : n_{i,t} = u \end{cases}$$

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	$ ho_{e}$	0.997	Kekre (2023)
Std Dev of idiosyncratic income process	σ_{e}	0.057	Kekre (2023)
Std Dev of Log Transitory Shock	$\sigma_{ heta}$	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 $= 0.21$

References I

- Auclert, Adrien, Bence Bardóczy, Matthew Rognlie, and Ludwig Straub, "Using the sequence-space Jacobian to solve and estimate heterogeneous-agent models," *Econometrica*, 2021, 89 (5), 2375–2408.
- Bayer, Christian, Ralph Lütticke, Lien Pham-Dao, and Volker Tjaden, "Precautionary savings, illiquid assets, and the aggregate consequences of shocks to household income risk," *Econometrica*, 2019, 87 (1), 255–290.
- **Bianchi, Francesco, Sydney C Ludvigson, and Sai Ma**, "Belief distortions and macroeconomic fluctuations," *American Economic Review*, 2022, 112 (7), 2269–2315.
- Broer, Tobias, Jeppe Druedahl, Karl Harmenberg, and Erik Öberg, "The unemployment-risk channel in business-cycle fluctuations," 2021.
- **Coibion, Olivier and Yuriy Gorodnichenko**, "Information rigidity and the expectations formation process: A simple framework and new facts," *American Economic Review*, 2015, *105* (8), 2644–78.

References II

- _ , _ , and Saten Kumar, "How do firms form their expectations? new survey evidence," *American Economic Review*, 2018, *108* (9), 2671–2713.
- **Fuhrer, Jeffrey C**, "Intrinsic expectations persistence: evidence from professional and household survey expectations," *Available at SSRN 3296152*, 2018.
- **Graves, Sebastian**, "Does Unemployment Risk Affect Business Cycle Dynamics?," *International Finance Discussion Paper*, 2020, (1298).
- Haan, Wouter J Den, Pontus Rendahl, and Markus Riegler, "Unemployment (fears) and deflationary spirals," *Journal of the European Economic Association*, 2018, 16 (5), 1281–1349.
- **Kekre, Rohan**, "Unemployment insurance in macroeconomic stabilization," *Review of Economic Studies*, 2023, *90* (5), 2439–2480.
- Mueller, Andreas I, Johannes Spinnewijn, and Giorgio Topa, "Job seekers' perceptions and employment prospects: Heterogeneity, duration dependence, and bias," *American Economic Review*, 2021, 111 (1), 324–63.