

# Rank-Dependent Probability Weighting and the Macroeconomy: Insights from a Model with Incomplete Markets and Aggregate Shocks

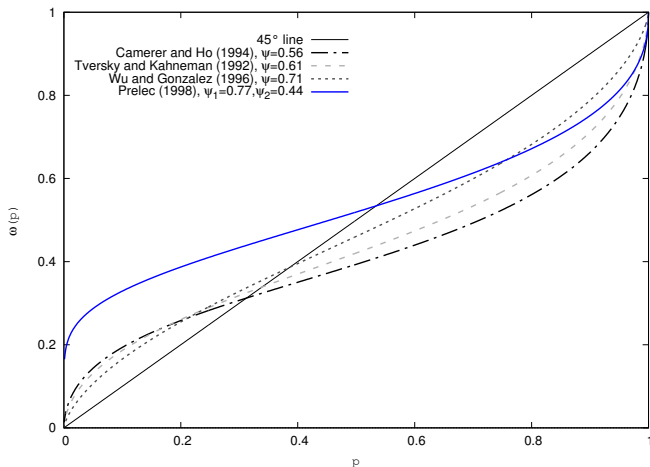
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# Motivation

- A vast experimental literature documents **rank-dependent probability weighting** in economic decisions characterized by risk.
- Individuals **overestimate** the likelihood of low-probability (bad) events, such as the probability of losing a job.
- It is not known whether this behavioral bias matters for **macroeconomic outcomes**, i.e. for business cycle analysis, and stabilization (fiscal) policies.
- I develop a model of **Rank-Dependent Expected Utility (RDEU)**, with Incomplete Markets and Aggregate Shocks.
- Standard models with Incomplete Markets and Aggregate Shocks imply grossly unrealistic unemployment rates, I specify a better formulation of **labor market dynamics**.

# Estimated Probability Weighting Functions



**Figure:** Probability weighting functions  $\omega(p)$ . One-parameter function (CH '94, TK '92, and WG '96):  $\omega(p) = (p^\psi / (p^\psi + (1-p)^\psi))^{1/\psi}$ . Two-parameter function (Prelec '98):  $\omega(p) = \exp(-\psi_1 * (-\ln(p))^{\psi_2})$ .

# Objective and Distorted Job Separations and Job Creations

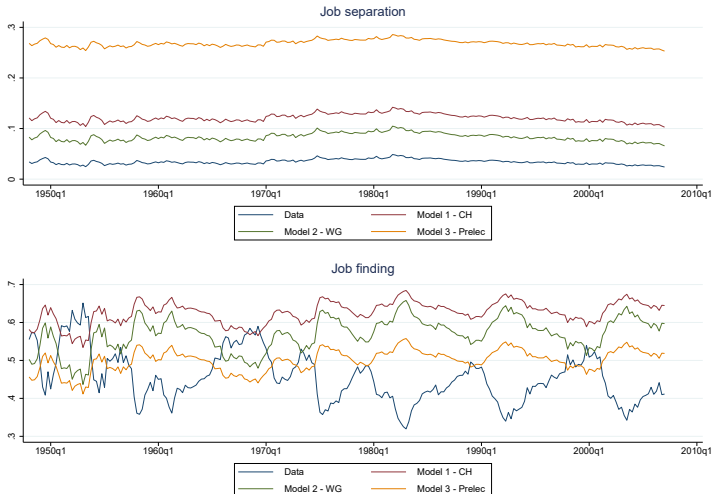


Figure: Job separations (JSP, top panel) and Job creations (JFP, bottom panel): Times series of the data Vs. distorted (RDEUs) probabilities.

# Contributions

- Implement a tractable –yet realistic– formulation of **labor market dynamics** in the processes driving the job finding and separation probabilities. → Individuals are facing an empirically relevant **level of risk**.
- **Computationally** quite challenging, but **feasible and easier** to work with than: a) prospect theory, and b) stochastic-volatility.
- Embed **RDEU** into a rich model with incomplete markets and aggregate shocks. → Explicit measurement of effects on **business cycles**.
- Apply the **Krusell-Smith algorithm** regarding household forecasting of future unemployment. → Implied bounded rationality allows not to rely on the **objective transition probabilities** for the aggregate dynamics.
- Quantify the importance of **RDEU** for household saving. → Explicit measurement of **wealth inequality** determinants.

## Related Literature

- **RDEU/Prospect Theory:** Quiggin (JEBO 81), Segal, Spivak, and Zeira (EL 88), Prelec (ECTA 98), Tversky and Kahneman (JRU 92), Kahneman and Tversky (ECTA 79).
- **“Exotic” Macroeconomics:** Backus, Routledge, and Zin (NBER 05), Backus, Ferriere, and Zin (JME 15).
- **Incomplete Markets and Aggregate Shocks:** Krusell and Smith (JPE 98, RED 02), Krusell, Mukoyama, Sahin and Smith (RED 09), Castaneda, Diaz-Gimenez and Rios-Rull (JME 98), den Haan, Judd and Juillard (JEDC 10), and many more...

# Ingredients of the Economic Model

- Incomplete markets, exogenous borrowing constraint and labor market risk (idiosyncratic and aggregate).
- Self-insurance through a (risky) asset to smooth income fluctuations.
- Ex-ante homogeneous infinitely-lived agents.
- Ex-post heterogeneity: employment status ( $\epsilon$ ) and assets ( $a$ ) (high degree of wealth concentration can be obtained).
- Time varying labor market conditions and associated risk.
- Aggregate states: capital ( $K$ ), unemployment rate ( $U$ ), productivity ( $Z$ ), job separation probability ( $s$ ) and job finding probability ( $\phi$ ).
- Households maximize Expected Utility (EU), but they distort objective risks via Rank-Dependent Probability Weighting (RDEU).
- This formulation avoids: the need to estimate reference points, time-inconsistency (dynamic programming can be used).

# The Model - Production

- The supply side is kept as simple as possible.
- Aggregate Cobb-Douglas production function in aggregate capital  $K_t$  and labor  $L_t$ .
- Total labor is  $L_t = I * N_t = I * (1 - U_t)$ .
- For tractability, productivity is deterministically linked to the unemployment rate:  $Z_t = 1 - \zeta * (U_t / \bar{U} - 1)$ .
- Output is  $Y_t = Z_t K_t^\alpha L_t^{1-\alpha}$ .
- Usual expressions for the net real return to capital

$$r_t = \alpha Z_t \left( \frac{L_t}{K_t} \right)^{1-\alpha} - \delta \text{ and the wage rate } w_t = (1 - \alpha) Z_t \left( \frac{K_t}{L_t} \right).$$



# The Model - Households

$$V(a, \epsilon, s, \phi, K, U) = \max_{c \geq 0, a' \geq a} \left\{ \frac{c^{1-\gamma} - 1}{1-\gamma} + \beta \mathbb{E}_{\epsilon', s', \phi', K', U' | \epsilon, s, \phi, K, U} V(a', \epsilon', s', \phi', K', U') \right\}$$

s.t.

$$c + a' = (1+r)a + (1-\tau)wl, \text{ if } \epsilon = e$$

$$c + a' = (1+r)a + \chi wl, \text{ if } \epsilon = u$$

$$\ln s' = (1 - \rho_s)\mu_s + \rho_s \ln s + \varepsilon'_s, \varepsilon_s \sim N(0, \sigma_s^2)$$

$$\ln \phi' = (1 - \rho_\phi)\mu_\phi + \rho_\phi \ln \phi + \varepsilon'_{\phi,t}, \varepsilon_\phi \sim N(0, \sigma_\phi^2)$$

$$\ln K' = \theta_{K,0} + \theta_{K,1} \ln K + \theta_{K,2} U + \theta_{K,3} Z$$

$$U' = \theta_{U,0} + \theta_{U,1} U + \theta_{U,2} \pi(u, e') * U + \theta_{U,3} \pi(e, u') * N$$

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*Exogenous Stochastic Processes*

$$\ln s' = (1-\rho_s)\mu_s + \rho_s \ln s + \varepsilon'_s, \varepsilon_s \sim N(0, \sigma_s^2)$$

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*Aggregate Laws of Motion*

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# Labor Market Dynamics

- **Markov Chains:** the typical formulation à-la Krusell and Smith (JPE 98) has Markov Chains on  $s$  and  $\phi$  such that the economy jumps randomly between  $U_{low} = 4\%$  in booms and  $U_{high} = 10\%$  in recessions.
- **AR(1)'s:** Using the Shimer (AER 05 and RED 12) data, I postulate and estimate with MLE two AR(1) processes for the job separation probability (JSP,  $s$ ) and the job finding probability (JFP,  $\phi$ ):

$$\ln s' = (1 - \rho_s)\mu_s + \rho_s \ln s + \varepsilon'_s, \varepsilon_s \sim N(0, \sigma_s^2)$$

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- **VAR(1):** I tried it, but it is not needed (small parameter estimates, no Granger causality, and low correlation of residuals).

# Probability Densities of JSP and JFP (quarterly)

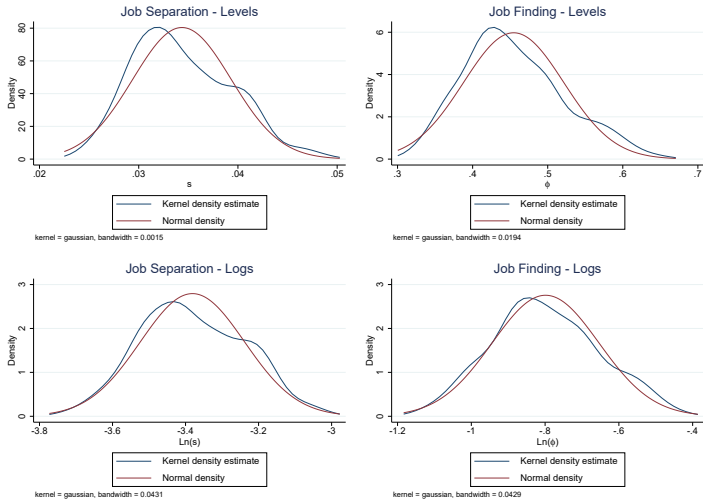


Figure: Limiting distributions of the two estimated AR(1) processes (in levels and logs) and kernel densities of the Shimer data.

# Probability Densities of Quarterly Unemployment Rates

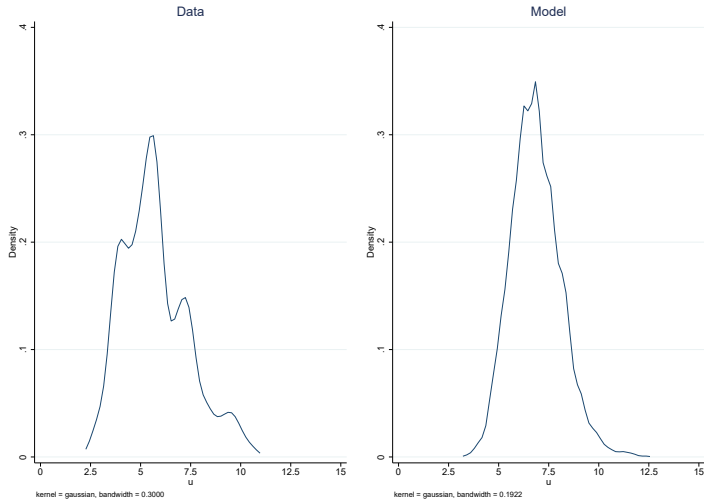


Figure: The data span the 1948Q1-2020Q1 period; the simulation draws long shock sequences from the estimated JFP and JSP AR(1) processes.



# Quarterly Unemployment Rates, Data Vs. Simulated path

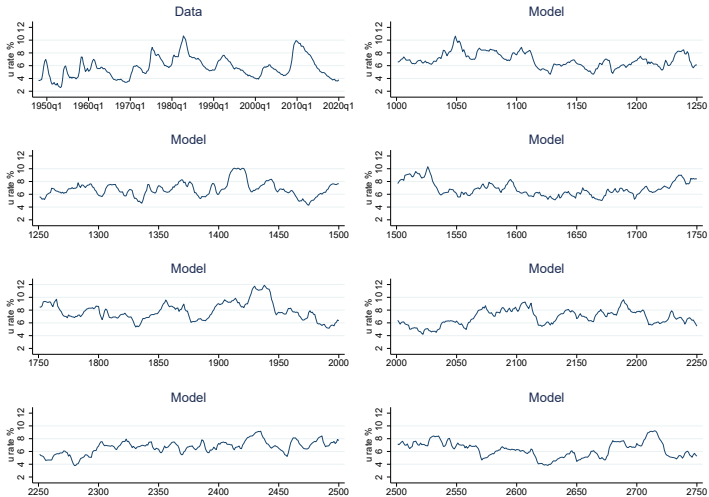


Figure: The simulation draws long shock sequences from the estimated JFP and JSP AR(1) processes, and the law of motion for the unemployment rate. ▶

# The Model - Quasi-Aggregation

- Also this model features Quasi-Aggregation: first moments deliver an almost perfect fit of the forecasting equations.
- Both the Capital and Unemployment Aggregate Laws Motion (ALM) have  $R^2 > 99.6\%$ .
- The Unemployment ALMs based on the objective transition probabilities have a marginally higher  $R^2 = 99.9\%$ .

# Calibration - RDEU Model

<i>Parameter</i>	<i>Value</i>	<i>Target</i>
$\alpha$ - Capital share	0.36	Labor share of output = 64%
$\delta$ - Capital depreciation rate	0.025	Avg. inv. share of output = 25%
$\theta$ - Risk Aversion	2.0	EIS = 0.5
$\beta$ - Discount factor	0.988	Avg. quarterly interest rate = 1%
$l$ - Hours worked	0.327	Time share of work = 33%
$\zeta$ - Productivity sens. to $u$ .	0.02	S.d. of aggregate income = 1.9%
$\chi$ - UI Replacement Rate	0.4	Benefits are 40% of labor earnings
$\underline{a}$ - Borrowing limit	-0.8	Households can borrow up to 20%
$\psi$ - Curv. of prob. weighting	-	Estimates in the literature
$\rho_\phi$ - Pers. of the AR(1) JFP	0.936	MLE estimates on CPS data
$\sigma_\phi$ - S.d. of the JFP shocks	0.051	MLE estimates on CPS data
$\mu_\phi$ - Uncond. avg. of the JFP	-0.8	MLE estimates on CPS data
$\rho_s$ - Pers. of the AR(1) JSP	0.923	MLE estimates on CPS data
$\sigma_s$ - S.d. of the JSP shocks	0.056	MLE estimates on CPS data
$\mu_s$ - Uncond. avg. of the JSP	-3.4	MLE estimates on CPS data

## Results - Business Cycle Statistics

<i>Variable j</i>	$\sigma_j$	$\sigma_j/\sigma_Y$	$\rho_j$	$\rho_{j,Y}$
<i>HP-filtered Data</i>				
<i>Y</i>	0.0189	1.000	0.892	1.000
<i>C</i>	0.0111	0.586	0.850	0.901
<i>I</i>	0.0658	3.487	0.847	0.937
<i>L</i>	0.0173	0.916	0.915	0.821
<i>EU Model</i>				
<i>Y</i>	0.0162	1.000	0.975	1.000
<i>C</i>	0.0138	0.850	0.929	0.911
<i>I</i>	0.0315	1.948	0.841	0.852
<i>L</i>	0.0138	0.855	0.910	0.944
<i>RDEU M1 (CH)</i>				
<i>Y</i>	0.0195	1.000	0.982	1.000
<i>C</i>	0.0137	0.702	0.953	0.837
<i>I</i>	0.0504	2.577	0.901	0.886
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<i>RDEU M2 (WG)</i>				
<i>Y</i>	0.0188	1.000	0.981	1.000
<i>C</i>	0.0138	0.732	0.949	0.859
<i>I</i>	0.0453	2.408	0.895	0.883
<i>L</i>	0.0138	0.735	0.910	0.907
<i>RDEU M3 (P)</i>				
<i>Y</i>	0.0192	1.000	0.982	1.000
<i>C</i>	0.0131	0.681	0.949	0.837
<i>I</i>	0.0477	2.488	0.901	0.902
<i>L</i>	0.0138	0.722	0.910	0.902



# Conclusions

- In terms of business-cycle statistics, the RDEU models attain lower root mean squared errors than the EU one.
- RDEU helps improving the fit of the **volatilities** of the macro aggregates.
- RDEU does not help improving the fit of the **autocorrelations** of some macro aggregates ( $C$  and  $Y$ ).
- In terms of long-run outcomes (e.g., capital level and wealth inequality), the RDEU models display mixed results: quantitatively important, but can go in either direction.
- Future work: consider the **welfare costs** of business cycles, exploiting a counterfactual that shuts down only the aggregate unemployment risk. → Develop a recursive methodology to address the so-called **integration principle**.