Labor Taxation and the Distribution of Income Shocks over the Cycle

Nicolò Dalvit, Julien Pascal Sciences Po, Sciences Po

Key Contributions

- Introduce non-linear income taxes in Robin (2011) with endogenous vacancy creation.
- Develop a solution algorithm based on Reiter (2009).
- Evaluate the contribution of income taxes to the distribution of cyclical income shocks.

Cyclical Income Risk

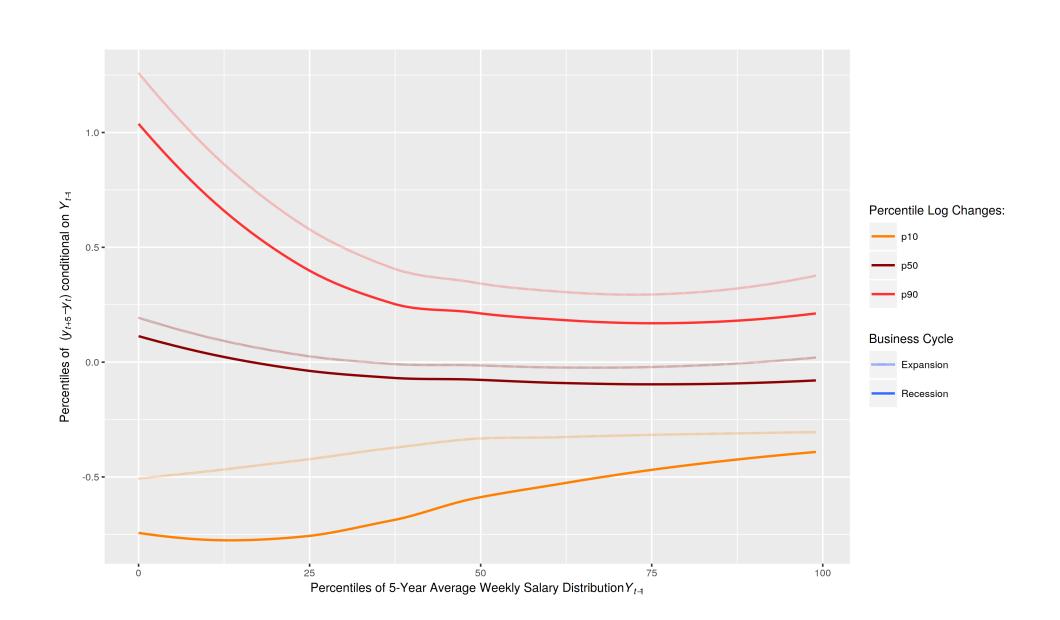


Figure 1: 5-year changes in log annual salary. Italy: 1977-2012. source: LoSai dataset, Estratti Conto.

Key facts on cyclical income risk:

- log-income changes are bigger and more cyclical for low-income workers.
- cyclical income risk driven mostly by extreme negative shocks (Guvenen et al. (2014)).
- Unemployment exits and entries play a key role.

Main Questions

- Can we reproduce the observed cyclical and distributional properties of labor income shocks?
- 2 How are these properties affected by alternative income tax schedules?

Model

- Continuum of workers with heterogeneous ability \boldsymbol{x} and homogeneous firms.

- Aggregate productivity z_t evolves stochastically.
- A firm-worker match produces output of value $p(x,z_t)$.
- The government taxes labor income w according to a tax schedule $\tau_w(w)$ and redistributes uniformly.
- Firms post vacancies V_t at cost $c(V_t)$.
- Per period number of meetings M_t is given by a matching function $M(L_t,V_t)$ with search effort

$$L_{t} = \int_{0}^{1} u_{t+}(x)dx + s \int_{0}^{1} h_{t+}(x)dx$$

• Unemployed (employed) workers meet a firm with probability λ_t ($s\lambda_t$), with

$$\lambda_t = \frac{M(L_t, V_t)}{L_t}$$

Let us define total and worker's private surplus from a match as $S_t(x, w)$ and $\Delta_t(x, w)$, respectively.

Wages are set following Robin (2011). Only two possible new wages per period and type:

$$\phi_t^0(x) : \Delta_t(x, \phi_t^0(x)) = 0$$

$$\phi_t^1(x) : \Delta_t(x, \phi_t^1(x)) = S_t(x, \phi_t^1(x))$$

Contrary to Robin (2011) and Lise Robin (2017) the surplus:

- depends on its allocation between workers and firms (i.e. on w) = partially transferable utility.
- ullet depends on the offer arrival rate λ_t

 λ_t , on the other hand, depends on L_t and

$$V_t = (c')^{-1} \left(\frac{M(L_t, V_t)}{V_t} J_t \right)$$

and therefore indirectly on the history-dependent distribution of matches $h_t(x) = \ell(x) - u_t(x)$.

Resolution Method

The model can be written as:

$$\begin{cases} \left(\underbrace{\triangle(x,w;\Gamma)}, \underbrace{S(x,w;\Gamma)} \right) = \Phi_1(\triangle(x,w,\Gamma), S(x,w;\Gamma)) \\ \text{Worker surplus} \end{cases}$$

$$= \Phi_2(h(.)|\triangle(x,w;\Gamma), S(x,w;\Gamma))$$
Distribution of Employment

where the aggregate state variable Γ contains z, h(.) and the tax schedule $\tau_w(.)$

1. Provide a finite representation of the model

Replace infinite dimensional (S, Δ, h) objects by discrete values on grids: $F(\mathbf{X_t}, \mathbf{X_{t-1}}, \eta_t, \varepsilon_t)$

 $\mathbf{X_t}$ contains values on grids $(S_{ij}, \Delta_{ij}, h_k)_t$, η_t are expectational errors and ε_t are shocks.

2. Solve for a steady-state of the discrete model

- Solve for S and \triangle holding fixed h
- Solve for h holding fixed S and \triangle

3. Linearize F around its non-stochastic steady-state and use a rational expectation solver

$$F_1(\mathbf{X_t} - \mathbf{X_{ss}}) + F_2(\mathbf{X_{t-1}} - \mathbf{X_{ss}}) + F_3\eta_t + F_4\varepsilon_t = \mathbf{0}$$

$$F_1 = \frac{\partial F}{\partial \mathbf{X_t}} | \mathbf{X_{ss}}, \ F_2 = \frac{\partial F}{\partial \mathbf{X_{t-1}}} | \mathbf{X_{ss}}, \ F_3 = \frac{\partial F}{\partial \eta_t} | \mathbf{X_{ss}}, \ F_4 = \frac{\partial F}{\partial \varepsilon_t} | \mathbf{X_{ss}}$$

The outcome is a linear model:

$$\mathbf{X_{t+1}} = A_{\tau}\mathbf{X_t} + B_{\tau}\varepsilon_{t+1}$$

(Preliminary) Counter-Factual

We calibrate the model using Italian administrative data for the period 1977-2012. We use our model to asses two alternative income tax regimes:

• Italian income tax regime in 2010.

• Revenue equivalent flat tax (24% flat rate).

Table 2: Counter-Factual - Aggregate

	1-Year Log Income Change			
	Level			
	P10	P50	P90	Std
Step	-0.349	0.002	0.345	0.327
Flat	-0.449	0	0.434	0.42
	(Time Series) St. Deviation			
\mathbf{Step}	0.489	0.244	0.469	0.264
Flat	0.649	0.337	0.609	0.307

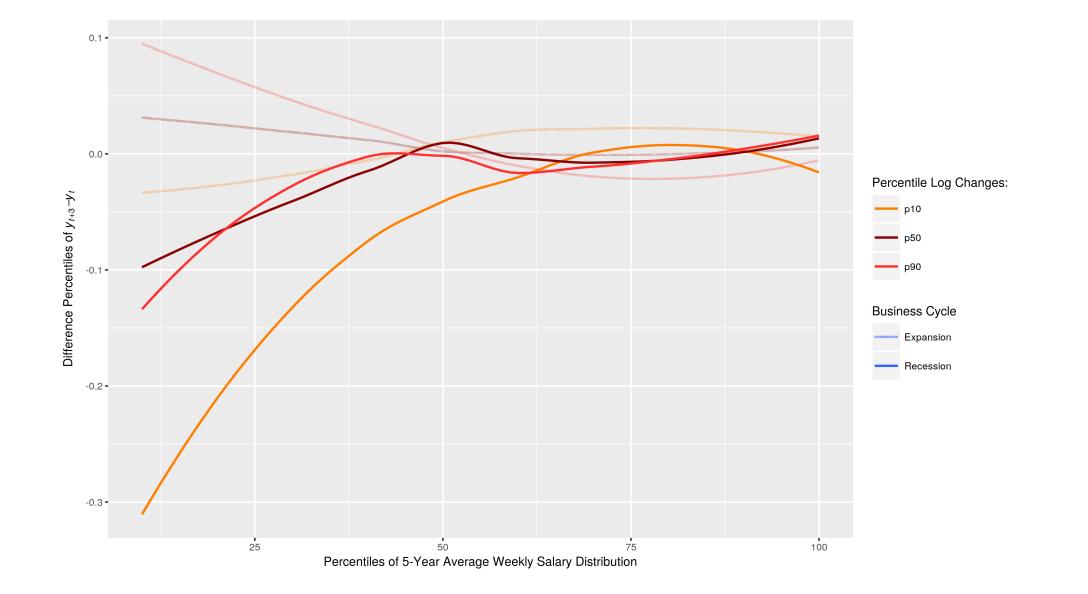


Figure 2: 3-year changes in log annual salary (simulated data)

Surplus Function

$$\begin{split} S_t(x,w) &= p(x,z_t) - \tau_w(w)w - b(x) + \frac{1-\delta}{1+r} \mathbb{E}_t \left[\mathbbm{1}\{S_{t+1}(x,w) < 0\} R_{t+1}^w(x) \right. \\ &+ \mathbbm{1}\{S_{t+1}(x,w) \geq 0\} [s\lambda_{t+1} S_{t+1}(x,\phi_{t+1}^1(x)) + (1-s\lambda_{t+1}) A_{t+1}(x,w)] \right] \\ R_t^w(x) &= \begin{cases} S_t(x,\phi_t^1(x)) & \text{if } S_t(x,\phi_t^0(x)) \geq 0 \\ 0 & \text{if } S_t(x,\phi_t^0(x)) < 0 \end{cases} \\ A_t^w(x) &= \begin{cases} S_t(x,w) & \text{if } 0 \geq \Delta_t(x,w) \geq S_t(x,w) \\ S_t(x,\phi_t^1(x)) & \text{if } \Delta_t(x,w) > S_t(x,w) \\ S_t(x,\phi_t^0(x)) & \text{if } \Delta_t(x,w) < 0 \end{cases} \end{split}$$

Conclusion

- An income tax introduces an additional level of complexity in a model à la Lise, Robin (2017).
- Reiter (2009) allows to efficiently solve and estimate the model (estimation is ongoing).
- Preliminary results show that, on aggregate, a revenue equivalent **flat tax**:
- increases the aggregate volatility and dispersion of income changes.
- the effect is driven by low income workers.