Modelling the supply of informal work

a microeconometric approach

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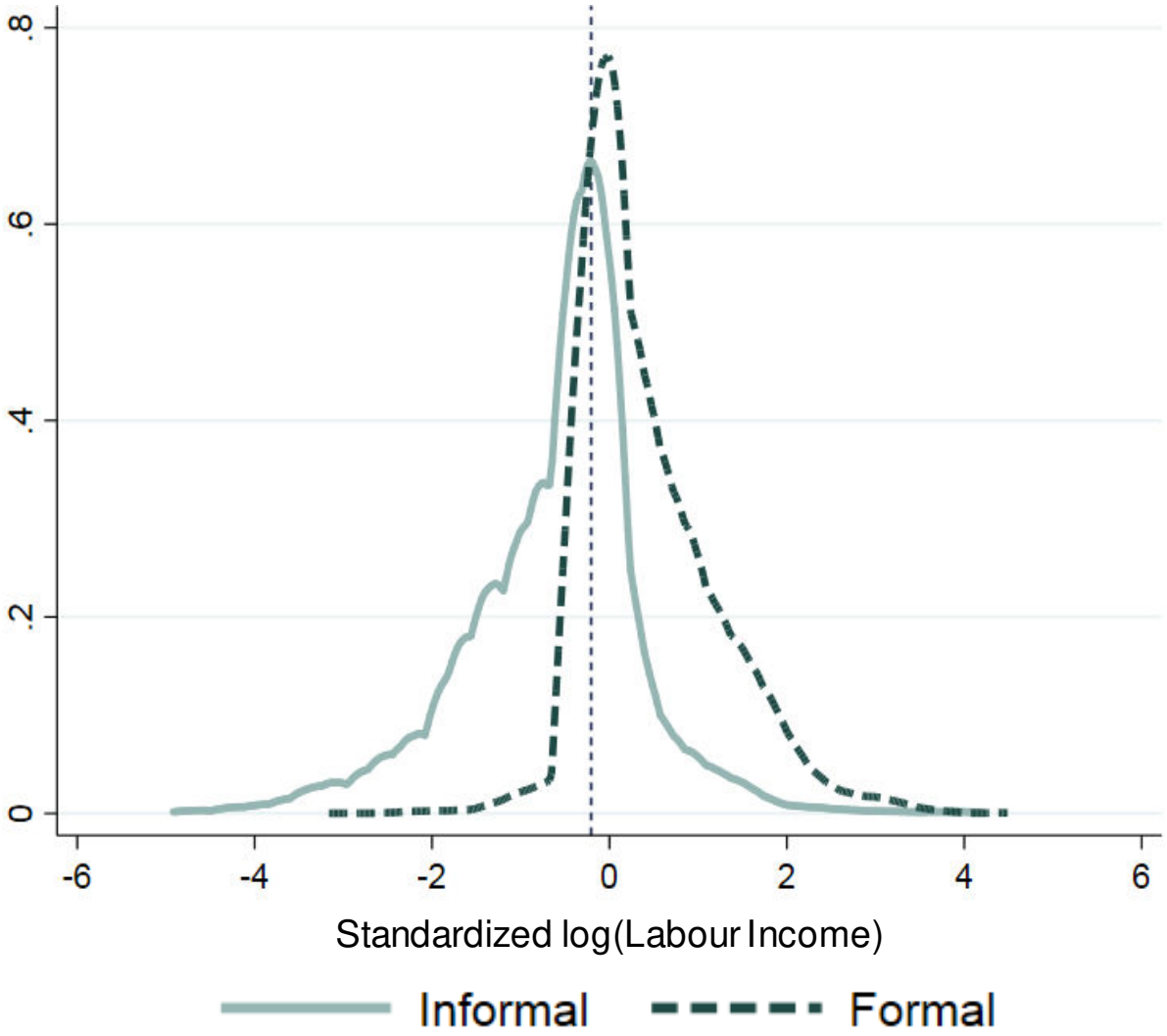
Motivation

- With more than half the workforce in the informal sector in Latin-American economies, understanding the nature of informality is critical
- If markets are segmented (i.e there are significant barriers to become formal) distortions in the formal sector are indeed large and the need for reform is compelling
- If informal workers' dynamics are similar to the formal sector, the focus shifts towards the cost-benefit analysis agents undertake choosing among sectors

Earnings distribution of formal and informal employees (2014)



*minimum wage vertical dotted line Source: Author's calculations

Research Questions

- To what extent barriers to enter the formal sector affect labour supply and sector choice?
- What would be the sectoral choice responses of informal workers to exogenous changes in tax-benefit system rules or education level?

Literature Review

- Informality Choice
  - Macroeconomic and Growth models with two sectors
    - Raush (1991), Loayza (1996), Amaral and Quintin (2006), Loayza and Rigolini (2011), De Paula and Scheinkman (2011), Galiani and Weinschelbaum (2012)
  - Search and Matching Models
    - Zenou (2008), Albrecht, Navarro and Vroman (2009), Bosch and Maloney (2010)
- Labour supply with informality
  - Too few microeconometric studies
  - Pradhan and Van Soest (1997) Labour supply and sector choice of two spouses households in Bolivia (ordered vs multinomial models)
  - Gong, Van Soest and Villagomez (2004) study formality, informality or non-participation choice in Mexico with a dynamic multinomial logit panel data model

Current labour supply models

Following Aaberge and Colombino (2014) modern labour supply models can be classified in two groups:

- Discrete Choice Models (DC)
- Random Utility-Random Opportunity (RURO)

Discrete Choice Model (DC) aka Van Soest model (VS)

- Choice set of hours of worked consists of only k discrete alternatives $L_i \in [0, T] \; \forall \; i = 1,2, \ldots k$
- Stochastic utility function over only one Consumption good (disposable income) and Leisure

$$U_i = V(C_i , T - L_i | X) + \varepsilon_i$$

- X Individual measured characteristics such as age, gender, education or number of kids etc., that could directly affect preferences or indirectly affect disposable income (through tax and benefit rules)

- Function $f(.)$ translates each element in the discrete set of hours worked into disposable income for that choice, by adding benefits and subtracting taxes and social insurance (TB) from labour (wL_i) and non-labour $(\overline{Y}_i)$ original income

$$Y_i = C_i = f(L_i|X) = wL_i + \overline{Y}_i + TB(wL_i + \overline{Y}_i|X)$$

- Calculated with tax-benefit microsimulation models

- Assuming ε_i is an error term with an Extreme value distribution the individual likelihood takes the multinomial logit form (L_i observed choice):

$$p_i(L_i|X) = \frac{e^{v(L_i|X)}}{\sum_{j=1}^{k} e^{v(L_j|X)}}$$

- For non-workers wages are imputed
- For couples a unitary decision household is usually assumed, utility depends on individual labour supply but household disposable income
- The model may require some fixed utility costs of working to improve fit (Van Soest,1995)

Random Utility-Random Opportunity Models (RURO)

- Intuition 1: Wages are not fixed but part of the job offer alongside the time regime.
- Intuition 2: Utilities are weighted with the intensity with which job offers are made available to each agent

- Each job offer is a bundle consisting of working time to be supplied (L_i) and a wage to be paid by the employer (w_i).

- Each non-market activity is assumed to offer a wage $w_i = 0$ and to require no hours of work $L_i = 0$

- The arrival of job offers depends on personal characteristics, labour demand conditions, the wage and time regime the job offer stipulates.

- Arrival modelled by a Poisson process with intensity parameter given by

$$\lambda_1(\varepsilon_i, q) g_1(w_i) g_2(L_i)$$

  - The opportunities function λ_1 captures labour demand conditions given by $q = \exp(\beta_q X_q)$
  - g_1 is the density (lognormal) of jobs paying wages w_i which is assumed to depend on some covariates.
  - g_2 is the density (assumed piecemeal uniform with peaks in the most frequently observed time regimes) of jobs requiring L_i hours of work

- The resulting probability that the chosen time regime L_i and wage w_i are the observed for individual i observed working is given by

$$p_i(w_i, L_i) = \frac{q\, g_1(w_i)\, g_2(L_i)\, e^{V(w_i, L_i)}}{e^{V(0,0)} + \int_{w_j \in \mathbb{W}} \int_{L_k \in \mathbb{H}} q\, g_1(w_j)\, g_2(L_k)\, e^{V(w_j, L_k)}\, dL_k\, dw_j}$$

For non participation

$$p_i(0,0) = \frac{e^{V(0,0)}}{e^{V(0,0)} + \int_{w_j \in \mathbb{W}} \int_{L_k \in \mathbb{H}} q\, g_1(w_j)\, g_2(L_k)\, e^{V(w_j, L_k)}\, dL_k\, dw_j}$$

Where $\mathbb{W}$ and $\mathbb{H}$ are the sets of wage offers and hours supplied respectively which are not observed (Decoster et al., 2016)

Labour supply of informal work

- We extend RURO to take into account sectoral choice assuming that job offer intensities, and wage and hour densities are different for informal and formal workers.
- The resulting individual likelihood is given by

$$p_i(w_i, L_i, z_i) = \frac{q_z g_{1k}(w_i)\, g_{2k}(L_i)\, e^{V(w_i, L_i, z_i)}}{e^{V(0,0,0)} + \sum_{k=0}^{1} \int_{w_j \in \mathbb{W}} \int_{L_k \in \mathbb{H}} q_z\, g_{1z}(w_j)\, g_{2z}(L_k)\, e^{V(w_j, L_k, z_i)}\, dL_k\, dw_j}$$

For non participation

$$p_i(0,0,0) = \frac{e^{V(0,0,0)}}{e^{V(0,0,0)} + \sum_{k=0}^{1} \int_{w_j \in \mathbb{W}} \int_{L_k \in \mathbb{H}} q_z\, g_{1z}(w_j)\, g_{2z}(L_k)\, e^{V(w_j, L_k, z_i)}\, dL_k\, dw_j}$$

To estimate the model we drawn from a priory density functions. The simulated likelihood for the observed time regime L_i and wage w_i is

$$p_i(w_i, L_i, z_i) = \frac{\frac{\mathbb{P}(0,0,0)}{\mathbb{P}(w_i, L_i, z_i)} q_z \, g_{1z}(w_i) \, g_{2z}(L_i) \, e^{V(w_i, L_i, z_i)}}{e^{V(0,0,0)} + \sum_{w_j, L_k, z_n \in \mathbb{D}} \frac{\mathbb{P}(0,0,0)}{\mathbb{P}(w_j, L_k, z_n)} q_n \, g_{1n}(w_j) \, g_{2n}(L_k) \, e^{V(w_j, L_k, z_n)}}$$

For non participation

$$p_i(w_i, L_i, z_i) = \frac{e^{V(0,0,0)}}{e^{V(0,0,0)} + \sum_{w_j, L_k, z_n \in \mathbb{D}} \frac{\mathbb{P}(0,0,0)}{\mathbb{P}(w_j, L_k, z_n)} q_n \, g_{1n}(w_j) \, g_{2n}(L_k) \, e^{V(w_j, L_k, z_n)}}$$

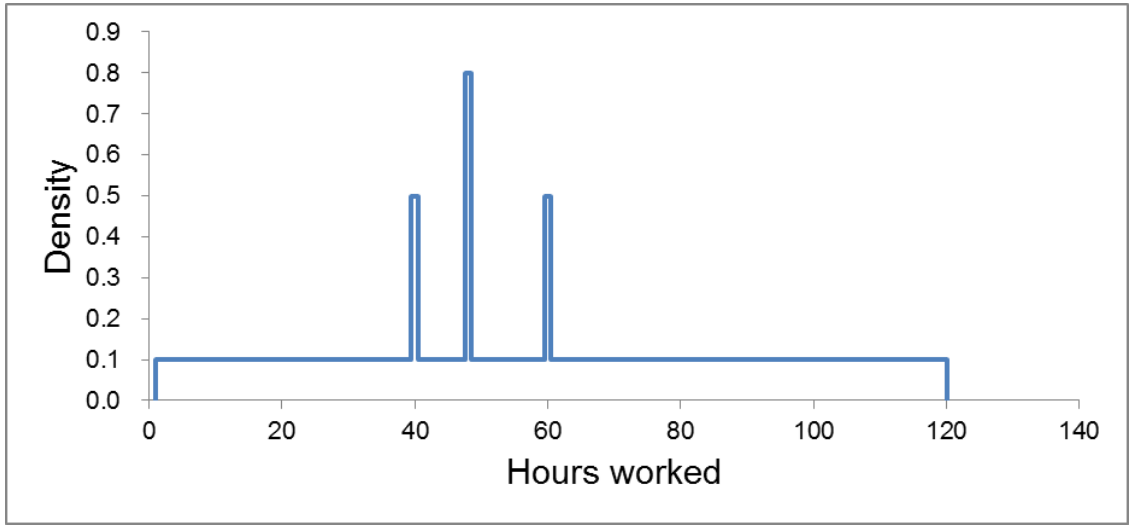
The probability of a job offer being drawn is included in the model $\left(\frac{\mathbb{P}(0,0,0)}{\mathbb{P}(w_j, L_k, z_i)}\right)$ and the observed choice must be included in the $\mathbb{D}$ subset (Train, 2009).

- For the three models the systematic part of the utility function is of the Box-Cox type

$$V(C_i, T - L_i|X) = V(w_i, L_i) = (\beta'_H X_H)\left(\frac{\left(\frac{T-L_i}{T}\right)^{\alpha_H} - 1}{\alpha_H}\right) + (\beta'_Y X_Y)\left(\frac{Y_i{}^{\alpha_Y} - 1}{\alpha_Y}\right)$$

- X_H and X_Y are a vector of parameters that shift the intensity of preference for leisure and income
- α_H and $\alpha_Y < 1$ determine the curvature of the indifference curves
  - lower values imply less substitutability between leisure and income.

- Hours density piecemeal uniform (Only for RURO type)

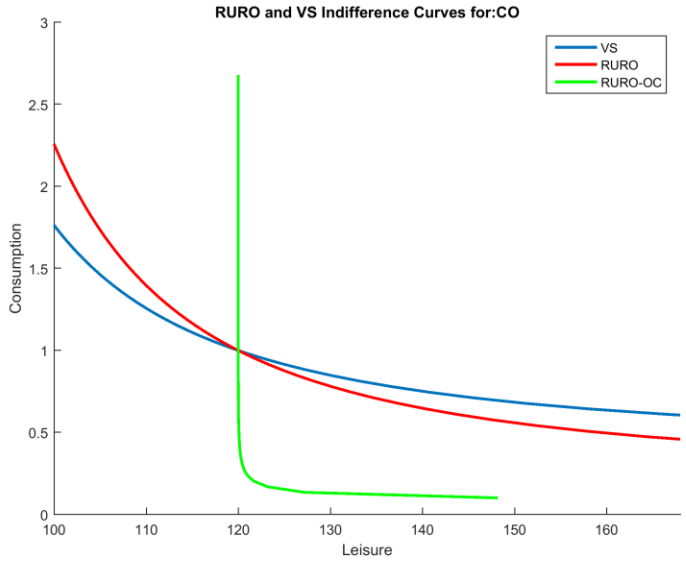


Data

- Quality of Life National Survey (ENCV)
  - Income and Expenditure Household Survey for Colombia 2014
  - Cross Section
  - 67.332 observations

- We focus on singles aged 18-60 years, not in education or disabled and in urban areas
  - They must be living without other working age family members
  - They are either working as employees (formal or informal) or not working i.e self-employed are excluded
  - If working, they report only one job
  - 697 observations

Preferences



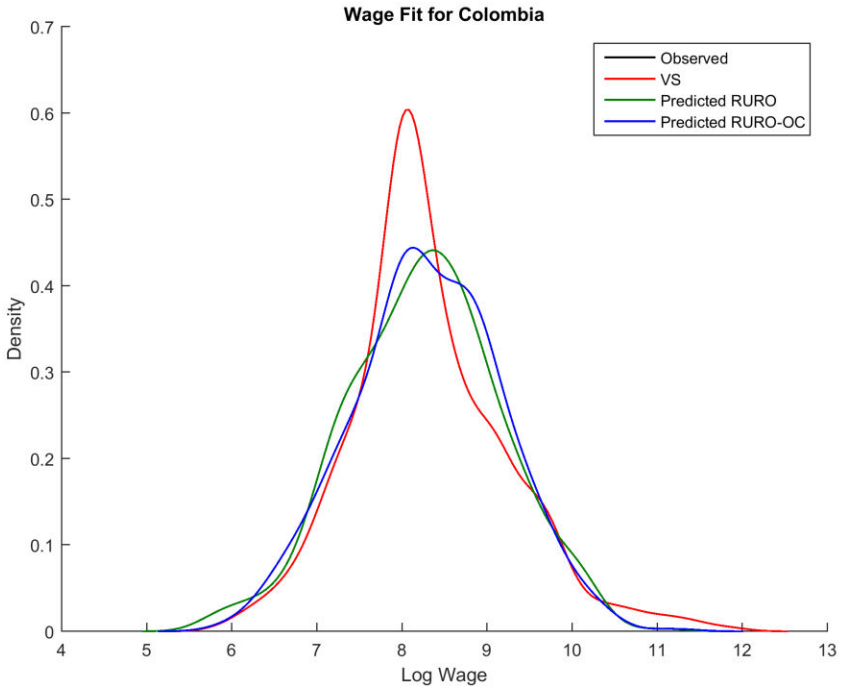
		VS		RURO		RURO_OC	
	Variable	coef.	t-value	coef.	t-value	coef.	t-value
All	leisureXconst	5.49	0.38	4.48	0.27	0.21	0.18
	leisure_male	-0.50	-1.94	-0.68	-1.76	-0.01	-0.47
	leisure_age	-2.56	-0.31	-2.28	-0.24	-0.16	-0.23
	leisure_age2	0.48	0.42	0.46	0.34	0.03	0.33
	leisure_child02	0.23	0.43	0.64	1.03	0.13	0.92
	leisure_child34	-0.31	-0.70	-0.45	-0.91	0.01	0.18
	leisure_child512	0.55	1.67	0.25	0.66	0.09	1.29
	alfa_leisure	-2.17	-6.48	-1.67	-2.92	-5.41	-5.04
	incomeXconst	2.30	10.96	1.07	5.48	0.00	0.21
	alfa_income	-0.36	-6.04	-0.29	-3.36	-3.19	-1.85
	fcXconst	42.94	1.97				
	fcXchild02	0.69	1.43				
	fcXage	-23.33	-1.93				
	fcXage2	3.29	1.96				

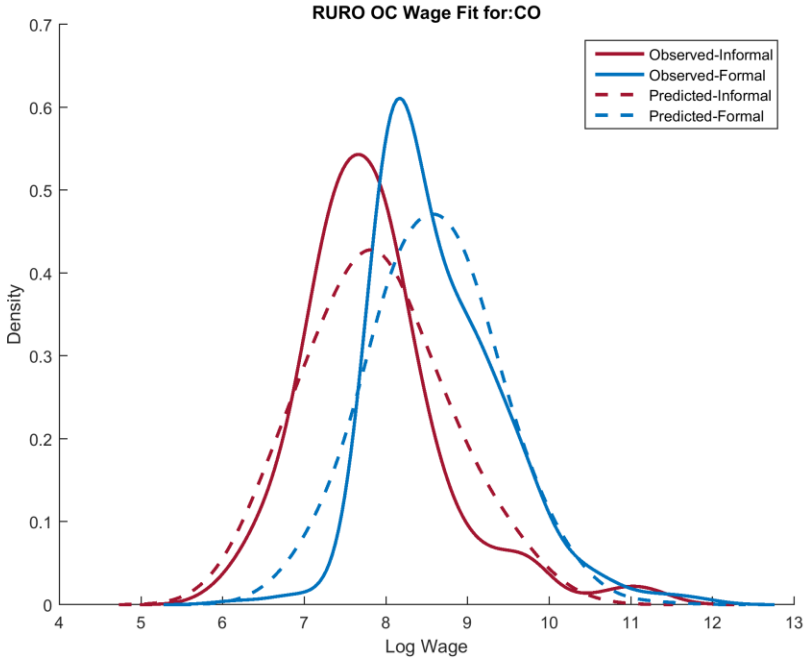
Source: Author's calculations

Wages

		VS		RURO		RURO_OC	
	Variable	coef.	t-value	coef.	t-value	coef.	t-value
All	w_const			6.71	40.41		
	w_exp			3.17	2.54		
	w_exp2			-5.50	-2.20		
	w_eduhigh			1.01	11.44		
	w_male			-0.02	-0.21		
	w_capital			0.37	3.44		
	w_socsec			0.52	6.14		
	rmse			0.74	26.56		
Informal	w0_const					7.31	36.77
	w0_exp					2.06	1.17
	w0_exp2					-3.44	-1.03
	w0_eduhigh					1.18	6.65
	w0_male					0.25	2.23
	w0_capital					0.57	2.89
	rmse0					0.77	19.52
Formal	w1_const					7.82	45.85
	w1_exp					4.14	2.69
	w1_exp2					-7.51	-2.30
	w1_eduhigh					0.77	9.30
	w1_male					-0.17	-2.08
	w1_capital					0.16	1.52
	rmse1					0.65	23.56

Source: Author's calculations





Source: Author's calculations

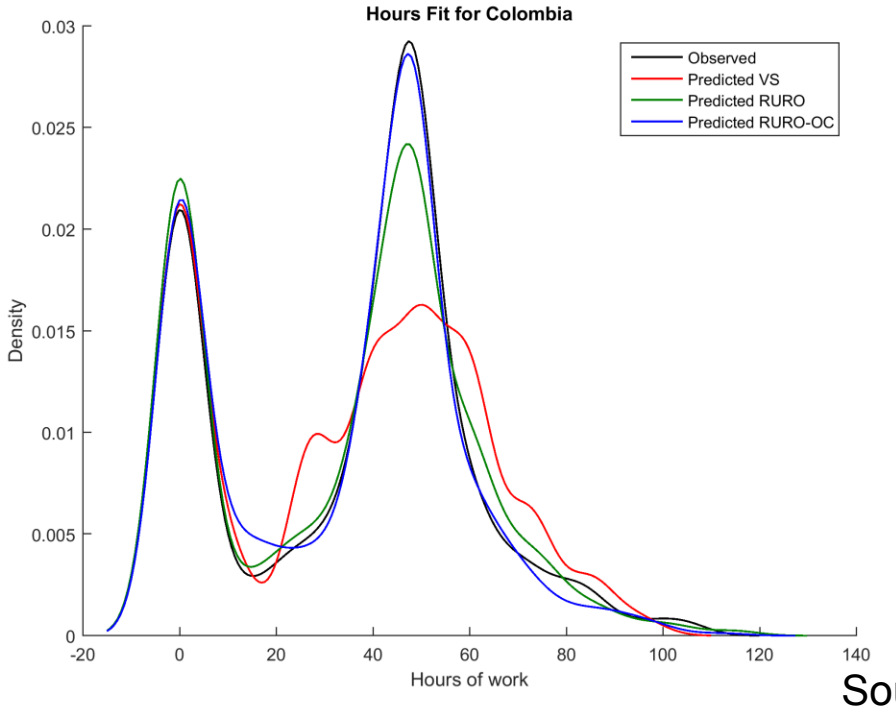
Opportunities

		VS		RURO		RURO_OC	
	Variable	coef.	t-value	coef.	t-value	coef.	t-value
All	opp_const			-24.24	-1.13		
	opp_age			10.98	0.91		
	opp_age2			-1.64	-0.98		
	opp_eduhigh			0.53	0.92		
	opp_formrate			0.89	0.86		
	opp_male			-0.61	-1.56		
Informal	opp0_const					-41.04	-2.48
	opp0_age					21.57	2.32
	opp0_age2					-3.21	-2.47
	opp0_eduhigh					-0.22	-0.71
	opp0_male					0.69	2.56
Formal	opp1_const					-84.70	-4.78
	opp1_age					43.43	4.40
	opp1_age2					-6.06	-4.43
	opp1_eduhigh					0.17	0.47
	opp1_form					4.68	6.88
	opp1_male					1.02	3.95

Source: Author's calculations

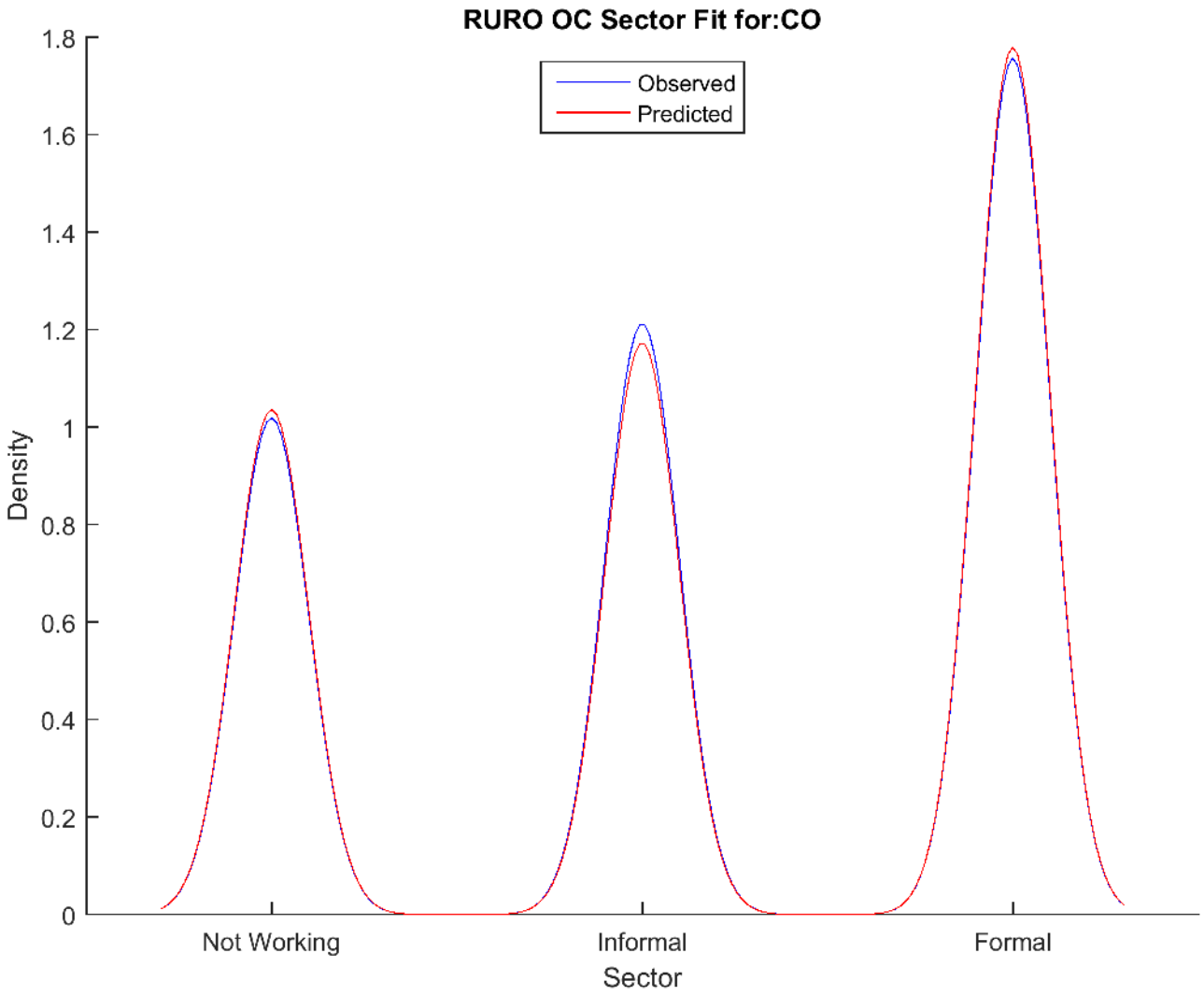
Hours

	Variable	VS coef.	VS t-value	RURO coef.	RURO t-value	RURO_OC coef.	RURO_OC t-value
All	hours_peak1 (40 h/w)			2.64	16.65		
	hours_peak2 (48 h/w)			3.65	31.72		
	hours_peak3 (60 h/w)			2.01	9.48		
Informal	hours0_peak1 (40 h/w)					2.16	7.96
	hours0_peak2 (48 h/w)					3.15	16.23
	hours0_peak3 (60 h/w)					2.02	6.66
Formal	hours1_peak1 (40 h/w)					3.33	17.01
	hours1_peak2 (48 h/w)					4.60	31.93
	hours1_peak3 (60 h/w)					2.48	8.76



Source: Author's calculations

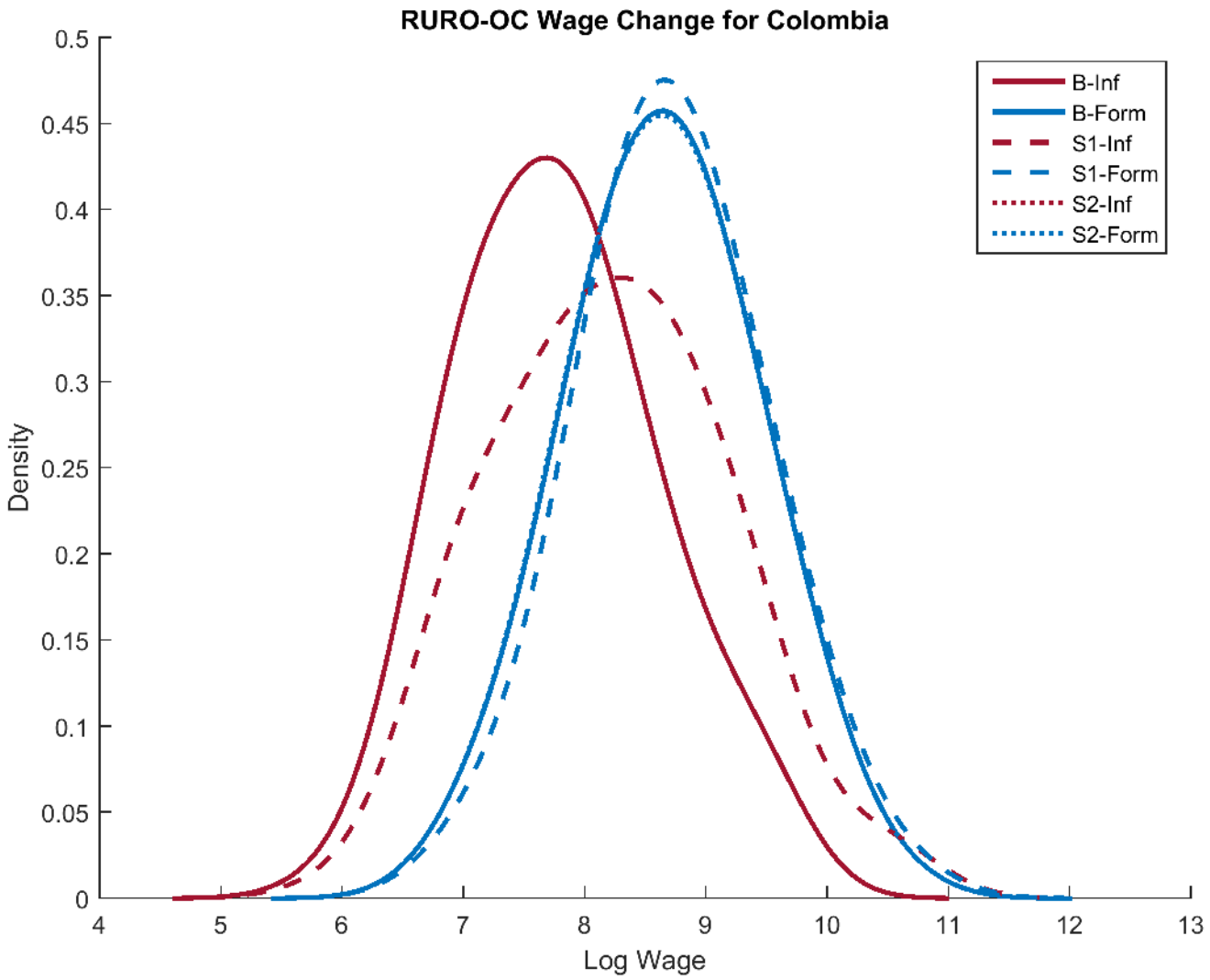
Sector Choice



Source: Author's calculations

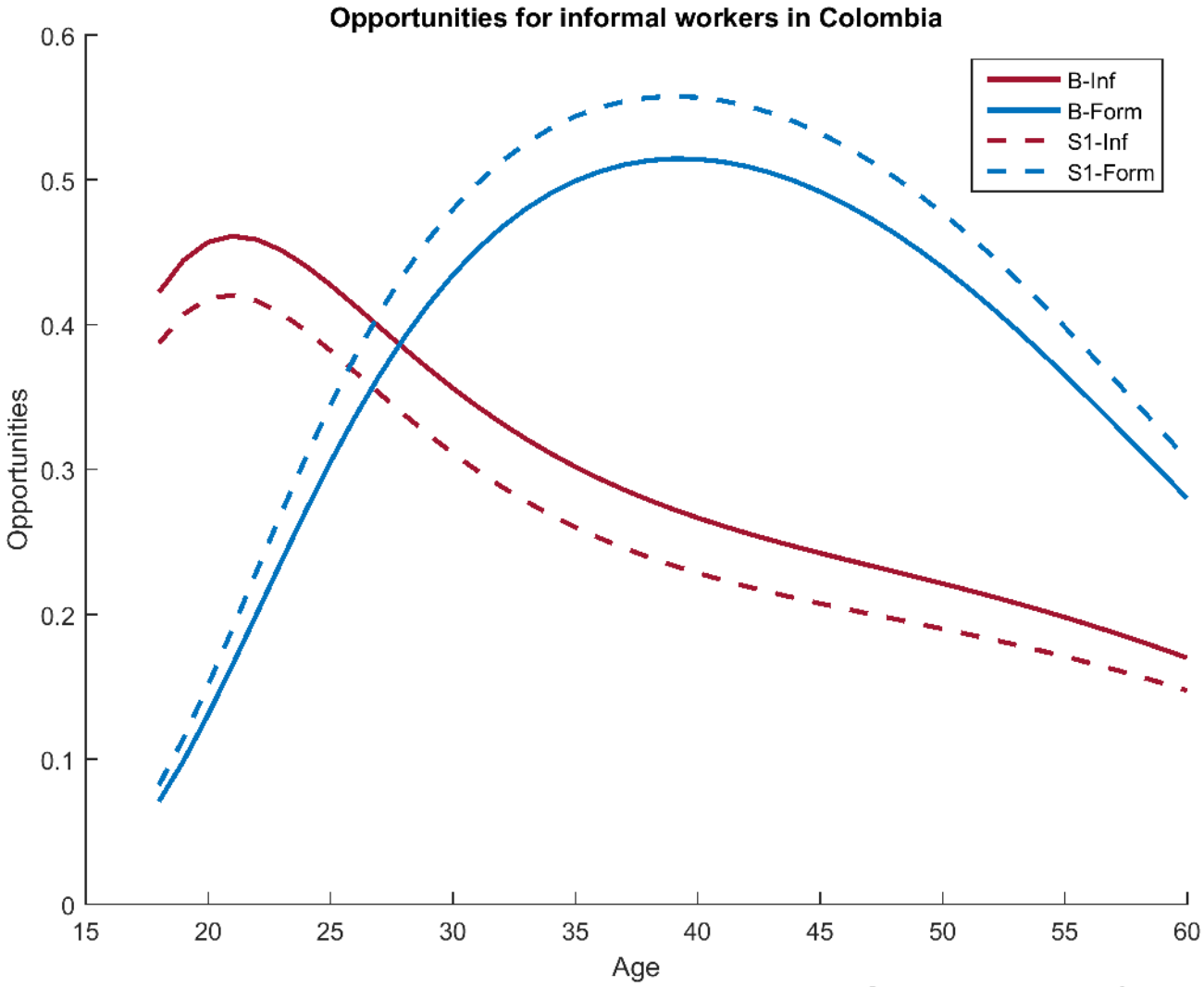
Simulation

- Simulation1: We increase the observed education of informal workers aged 18-40 years to the highest level resulting in informal workers with HE 10% →70%
  - Increases opportunities, wages and incomes
- Simulation2: We eliminate social insurance contributions for formal workers earning less than 3 monthly minimum wages
  - Increases disposable incomes in the formal sector



Source: Author's calculations

- Opportunities B and S1



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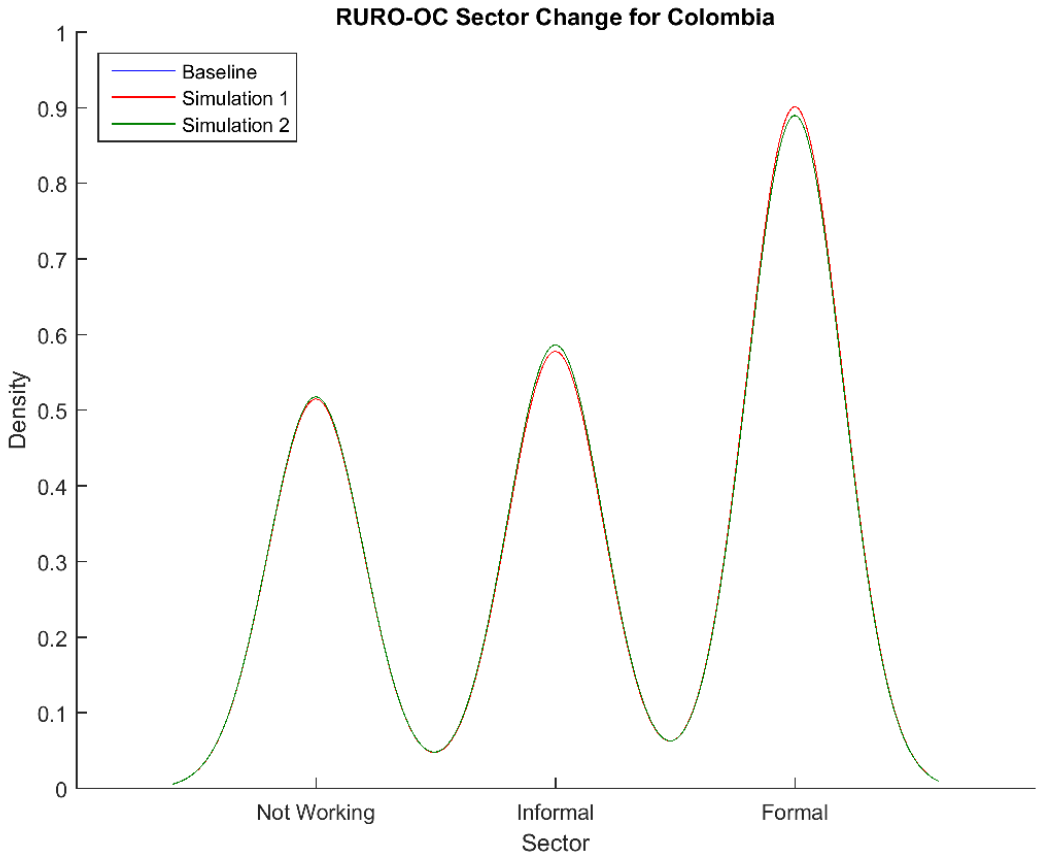

		Simulation 1		
		Not Working	Informal	Formal
Baseline	Not Working	0.994	0.000	0.006
	Informal	0.000	0.961	0.039
	Formal	0.000	0.016	0.984

Simulation 2			
	Not Working	Informal	Formal
OLS	1.000	0.000	0.000
Informal	0.000	1.000	0.000
Formal	0.000	0.000	1.000

Source: Author's calculations

Conclusions

- Very limited sector movement after important exogenous changes. This result favours the hypothesis of labour market segmentation
- Unobservables determine a great deal of opportunities for informal workers in the formal sector

Next steps

- Explore alternative determinants of opportunities
- Explore alternative dataset with more observations
- Calibrate a micro-founded macroeconomic model to better capture the interactions between supply and demand of formal and informal labour