

Discussion of: Marginal Tax Rates and Income: New Time Series Evidence, by K. Mertens

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- τ_t^S = average marginal tax rate

OLS Regression Results

Table II OLS Regression Results

	All Tax Units		Top 1%	Top 5%	Top 10%	Top 5-1%	Top 10-5%	Btm. 99%	Btm. 90%
	Series 1	Series 2							
<i>A. 1951-2010</i>									
Wage Inc.	-0.57*	-0.63**	0.45	0.13	-0.01	-0.15	-0.32	-0.75**	-0.97***
	(-1.18, 0.03)	(-1.19, -0.07)	(-0.10, 0.99)	(-0.32, 0.59)	(-0.43, 0.42)	(-0.40, 0.11)	(-0.86, 0.22)	(-1.33, -0.16)	(-1.66, -0.28)
Total Inc.	-0.33	-0.42	0.58*	0.28	0.14	-0.14	-0.27*	-0.60**	-0.80***
	(-0.89, 0.23)	(-0.95, 0.11)	(-0.08, 1.25)	(-0.37, 0.94)	(-0.44, 0.72)	(-0.46, 0.17)	(-0.58, 0.04)	(-1.08, -0.12)	(-1.37, -0.23)

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- For most income groups, estimates are negative, that is tax increases \Rightarrow higher income growth
- The problem is that there might be endogeneity, and thus the estimate may be biased.

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- 1. Review of analysis
- 2. Other potentially interesting/useful things to do? Things I am curious about, etc...

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- The paper proposes SVAR identification to overcome the bias

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- e_t^S = aggregate wage income, τ_t^S = average marginal tax rate, u_t are reduced-form shocks.

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- We care about:
 - (i) v_t^τ = the exogenous variation in average marginal tax rates
 - (ii) H_{12} – in fact, e.g. VAR no lags:

$$\begin{pmatrix} \ln(e_t) \\ \ln(1 - \tau_t) \end{pmatrix} = \begin{pmatrix} H_{11} & \mathbf{H}_{12} \\ H_{21} & H_{22} \end{pmatrix} \begin{pmatrix} v_t^o \\ v_t^\tau \end{pmatrix}$$

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- Let $m_t =$ exogenous narrative policy shocks s.t.:

$$E(v_t^\tau m_t) = \alpha \neq 0 \text{ (relevance)}$$

$$E(v_t^o m_t) = 0 \text{ (exogeneity)}$$

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$$\underbrace{\alpha \begin{pmatrix} H_{12} \\ 1 \end{pmatrix}'}_{=[0,1]} \underbrace{H'^{-1} H^{-1} u_t}_{v_t} = \alpha [0, 1] v_t = \alpha v_t^\tau$$

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 - (iii) $E(u_t^e m_t) = H_{12} \alpha$ gives an estimate of H_{12}
 - (iv) estimate impulse responses by tracing the effect of v_t^τ on Y_t .

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- One of the complications following the three stage procedure is that impulse response bands are quite convoluted to calculate, and the author has to rely on Goncalves and Kilian's bootstrap (2004), whereas Montiel-Olea, Stock and Watson's procedure provides a simple formula...
- **Bottom line: why not comparing the results of alternative procedures? Are there advantages relative to the simpler procedure?**

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Three-stage Regression Estimated in This Paper

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- Regress the variables you are interested in onto $v_t^\tau, v_{t-1}^\tau, \dots$ to get IRFs

Question II. About the Instruments

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 - The paper reports first-stage F-tests on elasticities estimated from $\Delta \ln(e_t^s) = \varepsilon \Delta \ln(1 - \tau_t^s) + r_t^s$ using v_t^T as instrument – more later...

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- However, the responses may be nonlinear (threshold models)
- Test whether there is empirical evidence of non-linearities?

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- 2 the paper also looks at dynamic responses of different income percentiles to the shock:

$$\Delta \ln (e_t^s) = \gamma_0 v_t^\tau + \gamma_1 v_{t-1}^\tau + \dots$$

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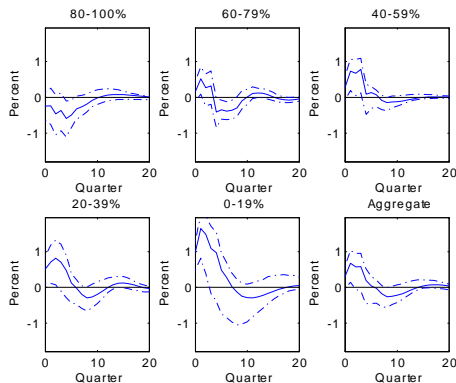
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- 2 Are they different across income levels? Confidence bands?
- 3 The paper reports first-stage F-tests on elasticities estimated from $\Delta \ln(e_t^s) = \varepsilon \Delta \ln(1 - \tau_t^s) + r_t^s$ using v_t^τ as instrument – but v_t^τ is estimated in the aggregate VAR – has the first stage F-statistics been corrected for that?

- 4. Heterogeneity. Anderson, Inoue and Rossi (2012) show that the response of consumption is very different across income groups.

IRF of Consumption to Gov. Spending Shock by Income Group



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- I wonder if it would be useful to support the claims in this paper with micro evidence based on survey data (if available)

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- Hope the comments are useful