

Benhabib-Gali

$$Y = f(K, HL), L = 1, H(0) = \phi,$$

$$H(t) = \phi e^{\gamma t}, y_i = \frac{Y_i(t)}{H(t)} = f\left(\frac{K}{H}, L\right) = f(k, L)$$

$$\frac{y_i(t)}{y_i(0)} = \frac{\frac{Y_i(t)}{H(t)}}{\frac{Y_i(0)}{\phi}} = \frac{\frac{Y_i(t)}{\phi e^{\gamma t}}}{\frac{Y_i(0)}{\phi}} = \frac{Y_i(t)}{Y_i(0)} e^{-\gamma t}$$

$$\ln\left(\frac{y_i(T)}{y_i(0)}\right) = \ln\left(\frac{Y_i(T)}{Y_i(0)}\right) - \gamma T$$

$$\left(\frac{1}{T}\right) \ln\left(\frac{Y_i(T)}{Y_i(0)}\right) = \ln\left(\frac{y_i(T)}{y_i(0)}\right) \left(\frac{1}{T}\right) + \gamma$$

But since from convergence we have

$$\begin{aligned}\ln(y(T)) - \ln(y(0)) &= \ln\left(\frac{y_i(T)}{y_i(0)}\right) \\ &= (1 - e^{-\lambda T})(\ln(y^*) - \ln(y(0)))\end{aligned}$$

we get

$$\begin{aligned}\left(\frac{1}{T}\right)\ln\left(\frac{Y_i(T)}{Y_i(0)}\right) \\ = \left(\frac{1}{T}\right)(1 - e^{-\lambda T})(\ln(y^*) - \ln(y(0))) + \gamma\end{aligned}$$

or,if

$$dY_i = \left(\frac{1}{T}\right)\ln\left(\frac{Y_i(T)}{Y_i(0)}\right), \quad \pi = \left(\frac{1}{T}\right)(1 - e^{-\lambda T})$$

,

$$dY_i = (\gamma + \pi\ln(y^*)) - \pi\ln(y_i(0)) + \varepsilon_i$$

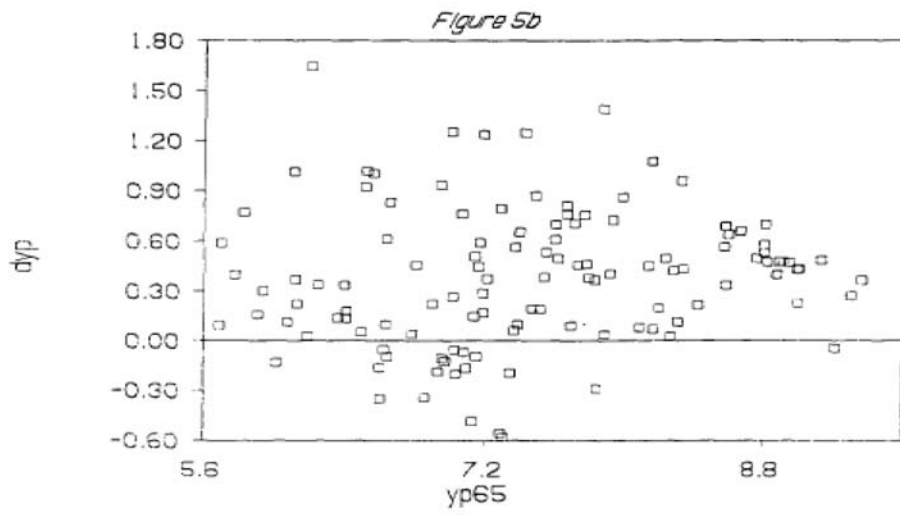
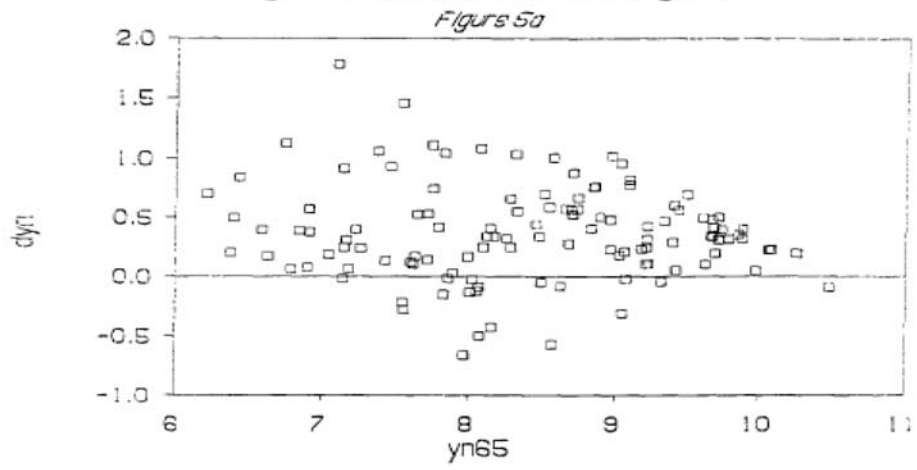
$$dY_i = (\gamma + \pi\ln(y^*)) - \pi\ln(Y_i(0)) + \pi\ln\phi + \varepsilon_i$$

$$dY_i = \kappa - \pi\ln(Y_i(0)) + \varepsilon_i$$

where ε_i is the error term, and

$$\kappa = \gamma + \pi\ln(y^*) + \pi\ln\phi.$$

Figure 5: unconditional convergence



What if y_i^* differ accross countries, or if initial values $Y_i(0)$ are correlated with technology levels ϕ_i ?

Let γ , and y^* be common. Note that

$$\begin{aligned} \ln(y_i(0)) &= \alpha \ln\left(\frac{k_i(0)}{y_i(0)}\right) + \alpha \ln(y_i(0)) \\ &= \left(\frac{\alpha}{1-\alpha}\right) \ln\left(\frac{k_i(0)}{y_i(0)}\right) \end{aligned}$$

Accordingly, write

$$\begin{aligned} dY_i &= (\gamma + \pi \ln(y^*)) - \pi \ln(y_i(0)) + \varepsilon_i \\ &= \Gamma - \left(\frac{\pi\alpha}{1-\alpha}\right) \ln\left(\frac{K_i(0)}{Y_i(0)}\right) + \varepsilon_i \end{aligned}$$

This gets rid of non-common ϕ_i . We estimate

$$\begin{aligned} dY_i^n &= 1.199 + 0.130 \ln\left(\frac{K_i(0)}{Y_i^n(0)}\right) + \varepsilon_i \\ &\quad (0.687) \quad (0.107) \end{aligned}$$

$$dY_i^p = 1.475 + 0.172 \ln\left(\frac{K_i(0)}{Y_i^n(0)}\right) + \varepsilon_i$$

(0.705) (0.110)

Figure 6: omega-conditional convergence

Figure 6a

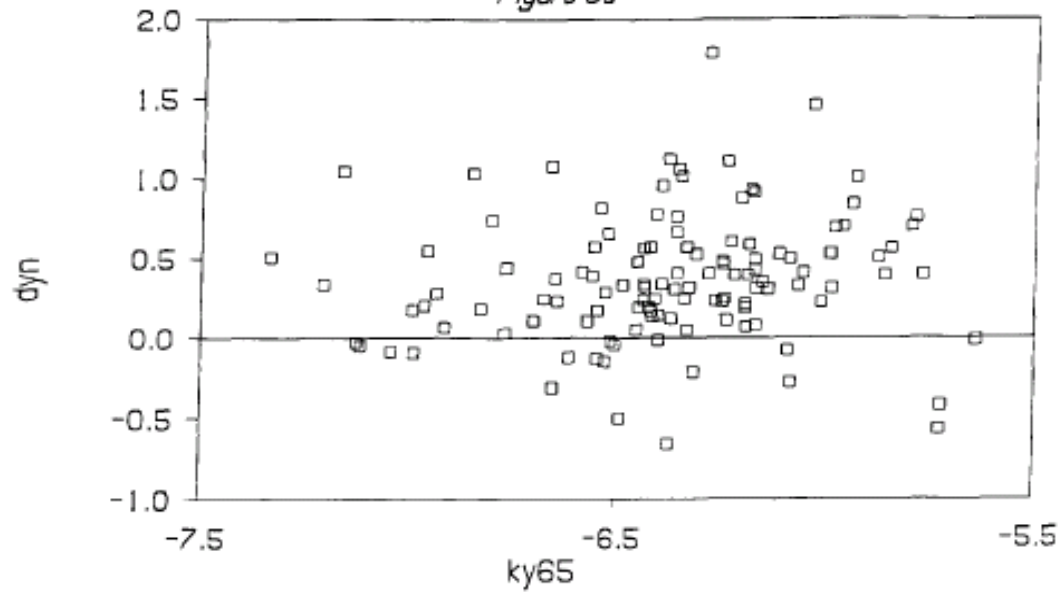
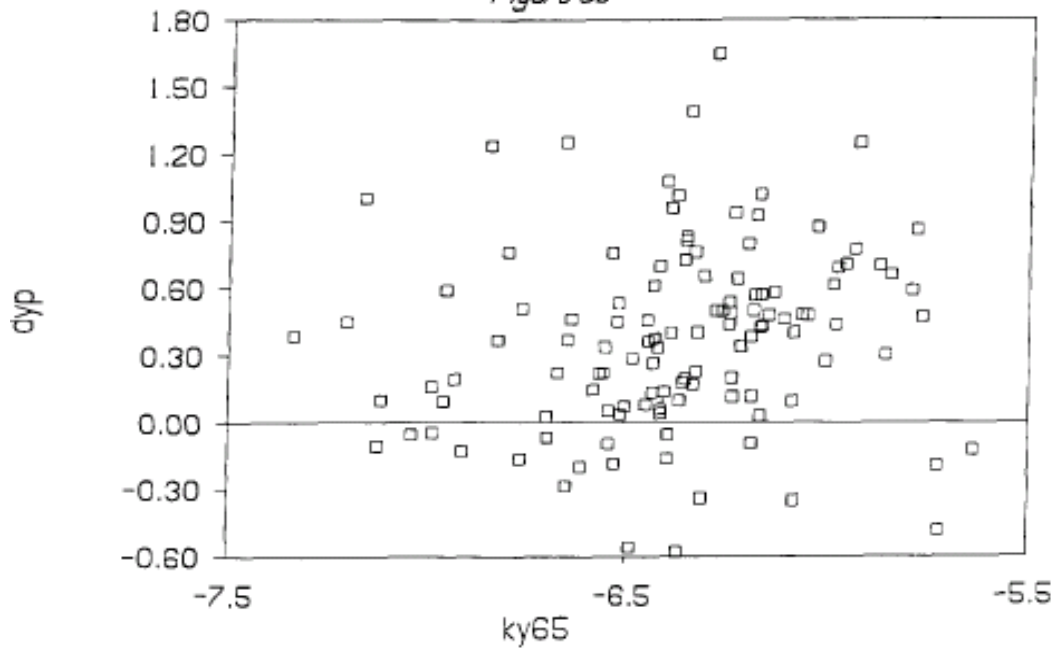


Figure 6b



But what if γ_i, y_i^* etc. hidden in Γ are non-common, and what if $\frac{K_i(0)}{Y_i(0)}$ is correlated with Γ ?

Split sample into two equal parts and difference, so that

$$dY_i(t, T) - dY(0, t) = -\Psi_i(dK_i(0, t) - dY_i(0, t))$$

where

$$dX(a, b) = (1/(a - b))(\ln(X_i(b)) - \ln(X_i(a)))$$

This gets rid of Γ . But what about α and π , inside Ψ_i ? Are they common? Then use

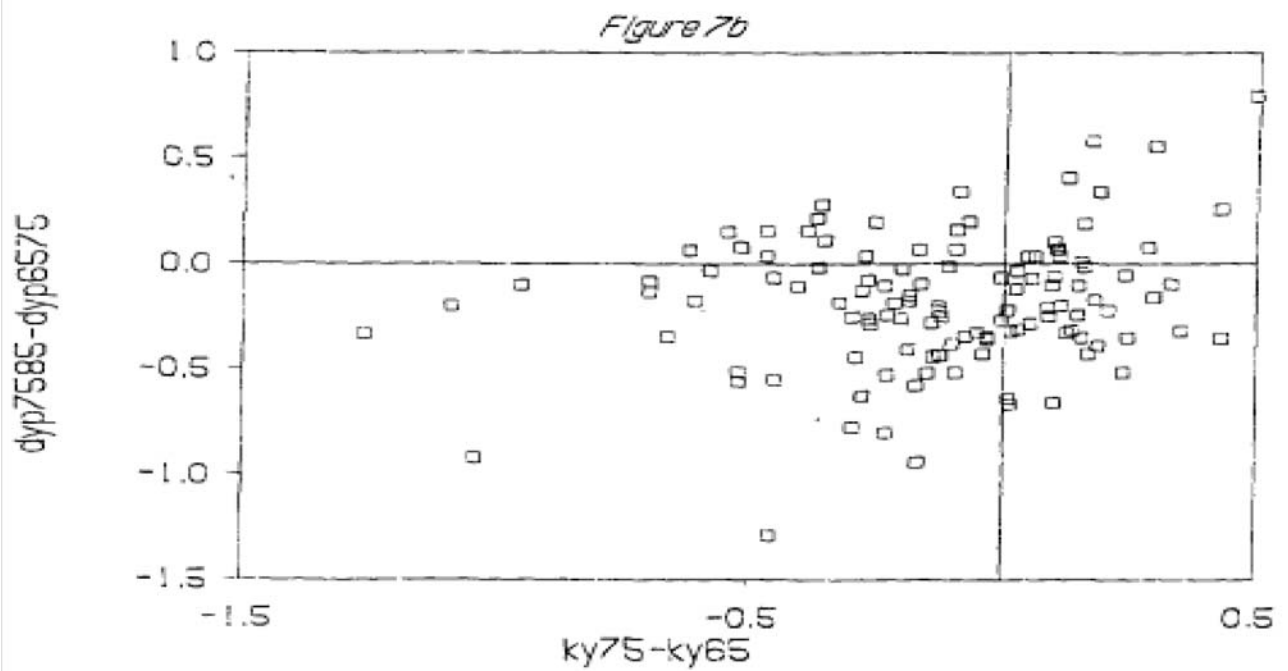
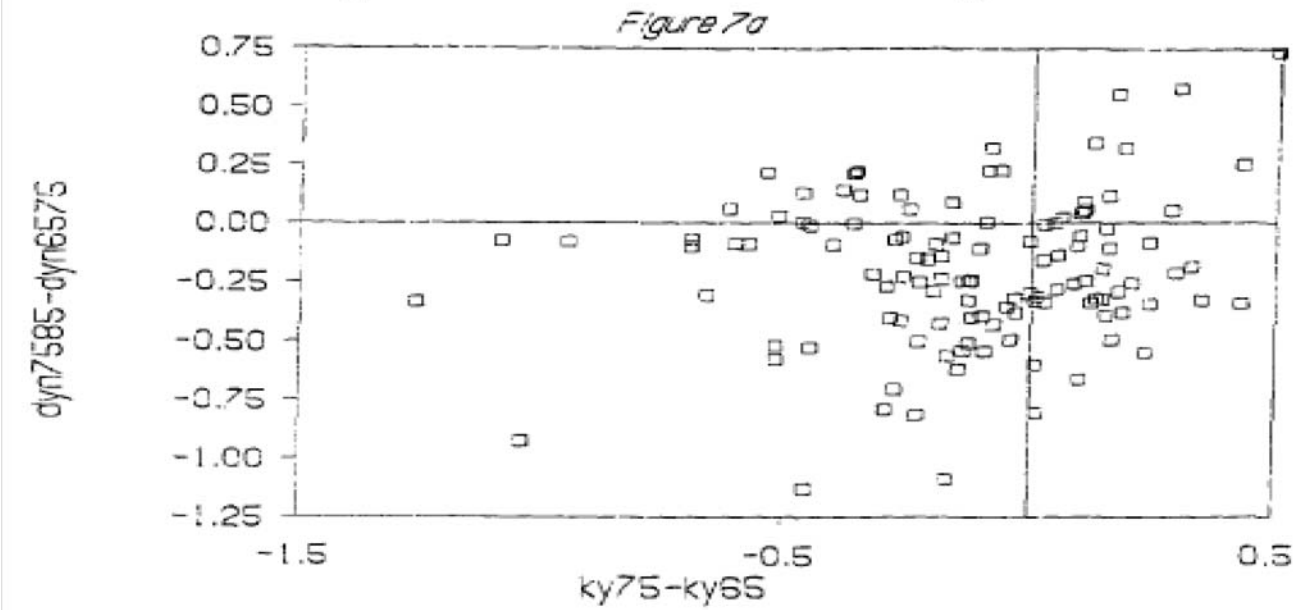
$$\sum_i (dY_i(t, T) - dY(0, t))(dK_i(0, t) - dY_i(0, t))$$

since we expect

$$(dY_i(t, T) - dY(0, t))(dK_i(0, t) - dY_i(0, t)) < 0$$

for all i

Figure 7: weak-conditional convergence



But more than 50% of observations are in

the southwest and northeast quadrants!

Quah's Analysis

If M is an transition matrix of incomes (25 year transition)

$$M = \begin{array}{c} \begin{array}{ccccc} 0 - .25 & .25 - .5 & .5 - 1 & 1 - 2 & 2 - \infty \\ .97 & .3 & 0 & 0 & 0 \\ .05 & .92 & .04 & 0 & 0 \\ 0 & .04 & .92 & .04 & 0 \\ 0 & 0 & .04 & .94 & .02 \\ 0 & 0 & 0 & .01 & .99 \end{array} \end{array}$$

The row eigenvector of the transpose of the transition matrix is the ergodic distribution:

$$.24 \quad .18 \quad .16 \quad .16 \quad .27$$

But the thinning middle may be due to differences in underlying parameters affecting savings, technology etc. Lets do it with $\frac{K}{Y}$:

	0 – 0.8	0.8 – 1.2	1.2 – ∞
$M =$	0.7	0.24	0.06
	0.32	0.34	0.34
	0.09	0.21	0.7

Ergodic Distribution

0.3809 0.2545 0.3645

Some thinning in the middle.