

# Appendix: the Great Moderation in the Japanese Economy

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# Omitted Variables

# Omitted Variables

- 1 Capital?
  - Gali (1999)
  - assumption: Stationary Capital output ratio
- 2 Other Variables (Shocks)?
  - IST shocks, Monetary shock, and so on
  - Future work

# Equilibrium

# Self-Confirming Equilibrium 1)

- Fudenberg and Levine (1993), Cho and Sargent (2006), Sargent (2008)
- interactions among a collection of **adaptive agents** , each of whom averages past data to **approximate moments of the conditional probability distns**
- If outcome converge, a Law of Large Numbers implies agents' beliefs about conditional moments become correct on events that are infrequently observed.
- Beliefs are not necessarily correct about events that are infrequently observed.
- Where beliefs are correct, a self-confirming equilibrium is like a rational expectations equilibrium.

## Self-Confirming Equilibrium 2)

- Agent  $i$  with strategy space  $A_i$  and state space  $X_i$ .
- Probability distn  $P_i$  over  $A_i \times X_i$ :  
how actions and states are related.
- utility fcn is  $u_i : A_i \times X_i \rightarrow R$
- $\mu_i(\cdot : a_i)$ : a probability distn over  $X_i$   
represent  $i$ 's belief about the state conditional on action  $a_i$ .
- Agent  $i$ 's decision problem is to solve

$$\max_{a_i \in A_i} \int_{x_i} u_i(a_i, x_i) d\mu_i(x_i : a_i). \quad (1)$$

# Priors

# Priors

- 18 Models
- Robust Results



# Priors 1)

- the conditional prior density of  $\theta^T$  is given by

$$p(\theta^T | \alpha^T, h^T, Q, \Psi, \Xi) \propto I(\theta^T) f(\theta^T | \alpha^T, h^T, Q, \Psi, \Xi),$$

- $z^T$ : a sequence of  $z$ 's up to time  $T$ .
- $I(\theta^T) = \prod_{t=0}^T I(\theta_t)$ ,
- $f(\theta^T | \alpha^T, h^T, Q, \Psi, \Xi) = f(\theta_0) \prod_{t=1}^T f(\theta_t | \theta_{t-1}, \alpha^T, h^T, Q, \Psi, \Xi)$
- $I(\theta^T)$  takes a unit value  
if all the roots of the VAR polynomial associated with  $\theta_t$   
are larger than one in modulus  
and 0 otherwise,  
ruling out a non-stationary process.

## Priors 2) distributions

- Following Cogley and Sargent (2005) and Gali and Gambetti (2009), **prior distributions** and its **hyperparameters** :

$$p(\theta_0) \propto I(\theta_0)N(\hat{\theta}_{OLS}, \hat{V}(\hat{\theta}_{OLS}))$$

$$p(\log h_0) = N(\log \hat{h}_{OLS}, 10 \times I)$$

$$p(\alpha_0) = N(\hat{\alpha}_{OLS}, |\hat{\alpha}_{OLS}|)$$

$$p(Q) = IW(\bar{Q}^{-1}, T_0)$$

$$p(\Xi_{i,i}) = IG\left(\frac{\bar{\Xi}}{2}, \frac{1}{2}\right)$$

$$p(\Psi) = IW(\bar{\Psi}^{-1}, 2).$$

- not flat but uninformative

## Priors 3) values

- $\hat{\theta}_{OLS}$ : OLS estimates of VAR coefficients.
- $\hat{V}(\hat{\theta}_{OLS})$ : the estimate of their covariance matrix.
- $\hat{h}_{OLS}$ : vector containing elements of the diagonal matrix  $\hat{H}$ .
- $\hat{\alpha}_{OLS}$ : the element (2,1) of the lower triangular matrix  $\hat{A}$ .
- $\bar{Q} = k_Q \times \hat{V}(\hat{\theta}_{OLS})$ .
- $T_0$ : # of observations in the initial sample.
- $\bar{\Xi} = k_\xi$ .
- $\bar{\Psi} = k_\Psi \times |\hat{\alpha}_{OLS}|$ .
- Benchmark:  $k_Q = 0.005$ ,  $k_\xi = 0.0001$ ,  $k_\Psi = 0.001$ .

## Priors 4) robustness

- 18 models
- $k_Q = \{0.005, 0.001, 0.1\}$ .
- $k_\xi = \{0.0001, 0.001\}$ .
- $k_\Psi = \{0.001, 0.01, 1\}$ .
- Robust in all cases!!!

# Estimation

# Estimation

- We use a Markov Chain Monte Carlo (MCMC) method, **the Gibbs sampling** .
  - The Gibbs sampler partitions the vector of unknowns into blocks.
  - The transition density is defined by the product of conditional densities.

## Step 1: $p(\theta^T | x^T, \alpha^T, h^T, Q, \Psi, \Xi)$

- Conditional on  $x^T, \alpha^T, h^T, Q, \Psi, \Xi$ , the unrestricted posterior of the states is normal.
- To draw from the conditional posterior, we employ the **algorithm of Carter and Kohn (1994)**.
- The conditional mean and variance of the terminal state  $\theta_T$  is computed using standard **Kalman filter recursions**.
- For all the other states, the following backward recursions are employed:

$$\theta_{t|t+1} = \theta_{t|t} + P_{t|t} P_{t|t+1}^{-1} (\theta_{t+1} - \theta_{t|t}), \quad (2)$$

$$P_{t|t+1} = P_{t|t} - P_{t|t} P_{t+1|t}^{-1} P_{t|t}, \quad (3)$$

where  $p(\theta^T | x^T, \alpha^T, h^T, Q, \Psi, \Xi) \sim N(\theta_{t|t+1}, P_{t|t+1})$ .

## Step 2: $p(\alpha^T | x^T, \theta^T, h^T, Q, \Psi, \Xi)$

- Conditional on  $\theta^T$ ,  
 $\hat{y}_t = x_t - \beta_{0,t} - \beta_{1,t}x_{t-1} - \dots - \beta_{p,t}x_{t-p}$  is observable.
- We can rewrite our system of equations as  $A_t \hat{y}_t = H_t v_t$ ,  
where  $v_t \sim N(0, I)$ .
- Conditional on  $h^T$ ,  
we use **the algorithm of Carter and Kohn (1994)**  
to obtain a draw for  $\alpha_t$   
taking the above system as observational equations and  
unobserved states equations.
- Given that the  $\alpha_t$  and the  $v_t$  are independent across equations,  
the algorithm can be applied equation by equation.



## Step 3: $p(h^T | x^T, \theta^T, \alpha^T, Q, \Psi, \Xi)$

- This is done by using the univariate algorithm by [Jacquier et al. \(1994\)](#) .

$$\begin{aligned} \text{Step 4: } & p(\Psi|x^T, \theta^T, \alpha^T, h^T, Q, \Xi), \\ & p(\Xi_{i,i}|x^T, \theta^T, \alpha^T, h^T, Q, \Psi), \\ & p(Q|x^T, \theta^T, \alpha^T, h^T, \Psi, \Xi) \end{aligned}$$

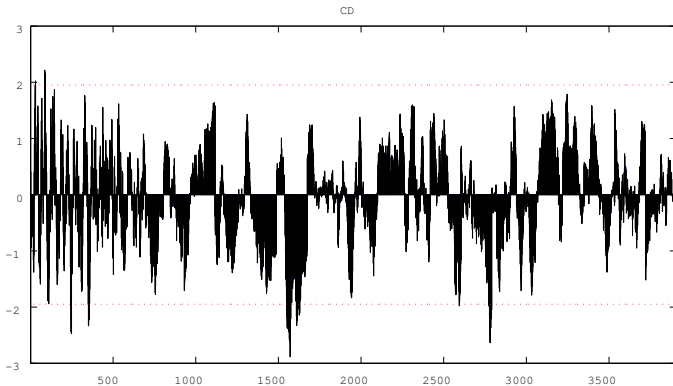
- Conditional on  $x^T, \theta^T, \alpha^T, h^T$ ,  
all the remaining hyperparameters,  
under conjugate priors,  
can be sampled in a standard way from  
Inverted Wishart and Inverted Gamma densities.

# Convergence

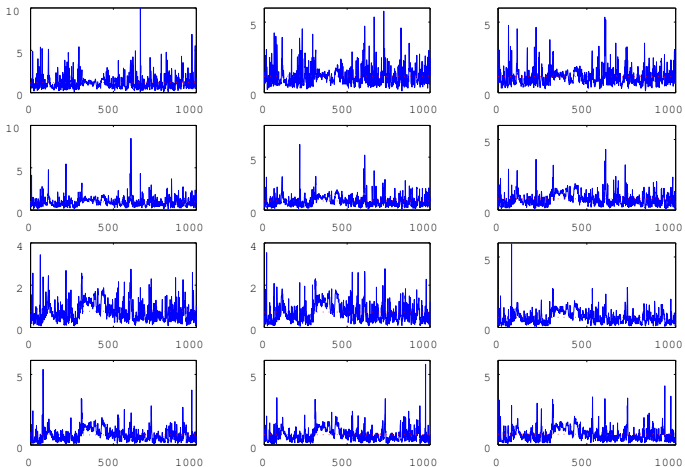
# Covergence

- Geweke (1992)
- Convergence Diagnostics (CD)

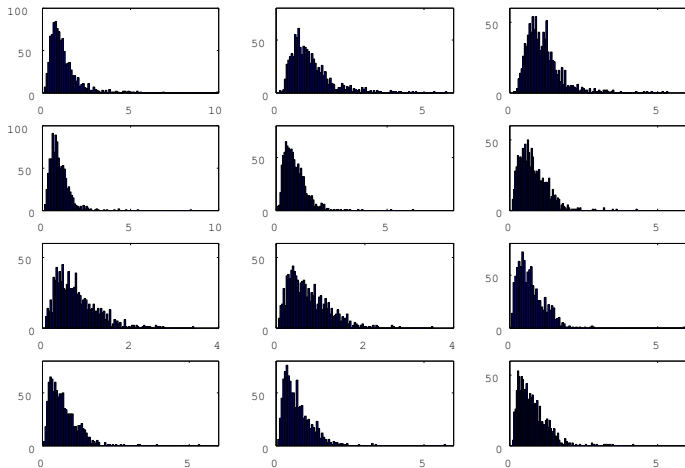
Fig: CD statistics



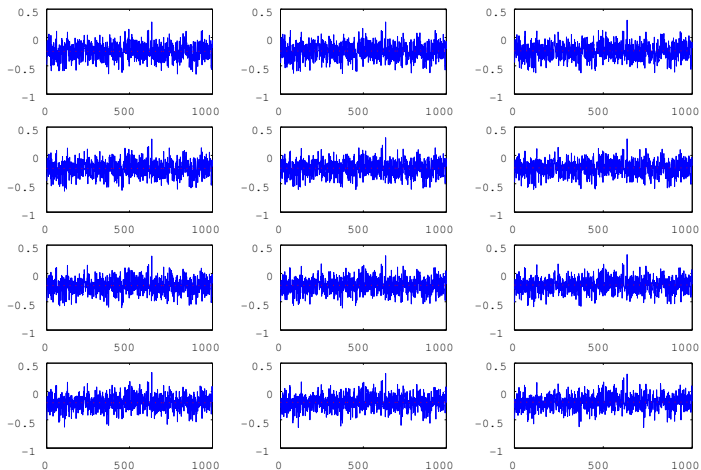
- Cannot reject the null hypothesis in most cases

Fig: Draws and Means of posterior  $\log h_t$ 

- stationary distr

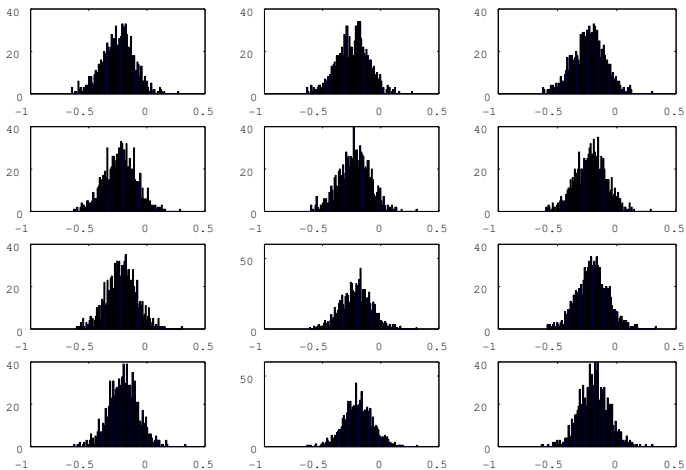
Fig: Density of Posterior  $\log h_t$ 

- stationary distn

Fig: Draws and Means of posterior  $\beta_{1,t}$ 

- stationary distn



Fig: Density of Posterior  $\beta_{1,t}$ 

- stationary distn

# Government and labor

# Labor Market Dynamics

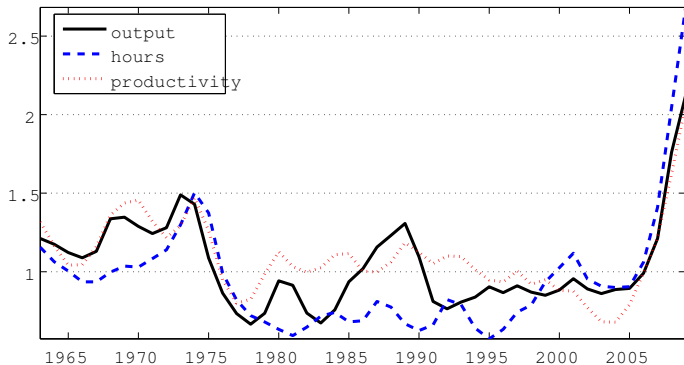
- Employment Protection by Case Laws (**Kaiko Kisei** )
  - 1 employers must meet four conditions before they fire an employee.
  - 2 Kawaguchi and Murao (2009)
- **Life-time employment system**
  - 1 Endo and Hirakata (2010)
  - 2 still survives
- Labor Standard Law (**Jitan** )
  - 1 Kawaguchi and Tsuru (2010): endogenous working hours ↓.
  - 2 Kuroda (2010): No effect on large firms
- Worker Dispatching Act (**Roudou Haken Hou** )
  - 1 1986, 1999, 2004

# Taxes

- Gali (2003), Uhlig (2004)
  - Capital tax: included in productivity shocks
  
- Suppose
  - 1 prod. fcn:  $Y = F(K, AN)$
  - 2 Homo. of degree 1:  $\frac{Y}{N} = AF(k, 1)$   
 where  $k \equiv (K/AN)$ : the ratio of capital to labor
  - 3  $(1 - \tau)F_k(k, 1) = \text{const.}$  hold along a B.G.P.
  - 4 stationary  $\tau$  may be unwarranted.
  
- But, at least
  - 1 theory:  $\tau \uparrow \rightarrow lp \downarrow N \uparrow$
  - 2 our result (IR):  $lp \uparrow N \uparrow$

# Phases

# Unconditional SDs

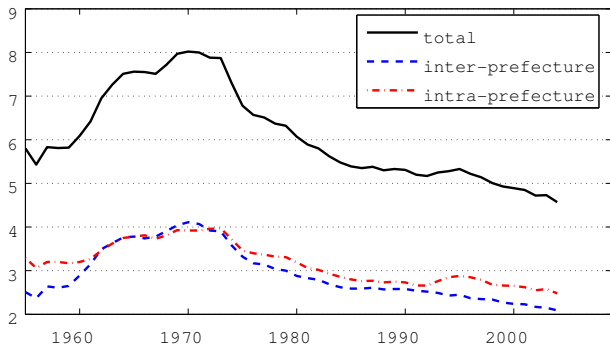


- 5 Phases

# Five Phases

- 5 distinct Phases in our economy
- *1st Phase: Until Mid-1970s*
  - ① participation of Baby-boomers
  - ② geographical and sectoral movement
- *2nd Phase: Mid-1970s to Late-1980s*
  - ① Very stable economy (GM period)
- *3rd Phase: Late-1980s to Early-1990s*
  - ① Bubble periods
  - ② volatile output while stable labor input and productivity
  - ③ disappearance of negative correlation

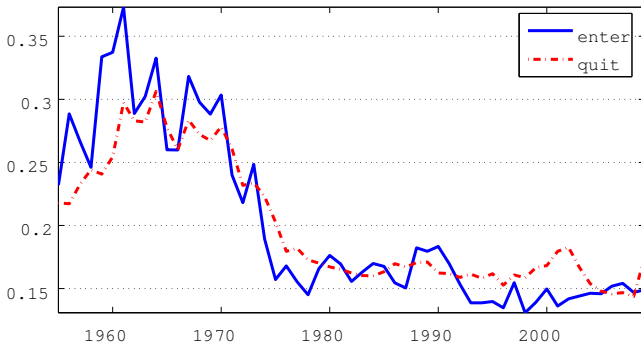
Fig: Migration Rates



- upper trend in the 1st Phase

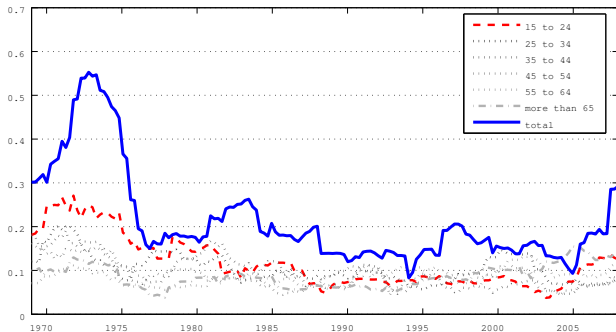


Fig: Turnover Rates



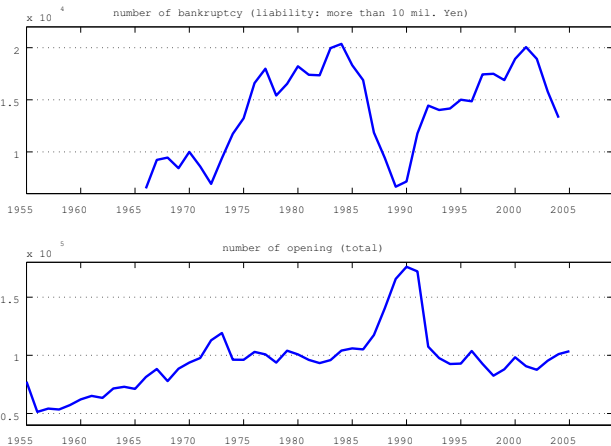
- volatile in the 1st Phase

Fig: SD of employment rate among ages



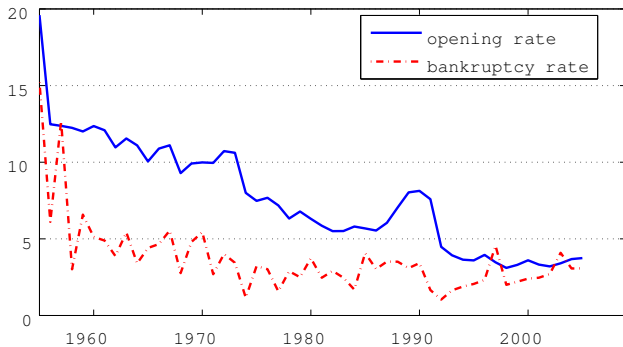
- main role of 15-24 in the 1st Phase
- consistent with Jaimovich and Siu (2009)

# Fig: Number of Bankruptcy and Opening



- different feature in the 3rd Phase

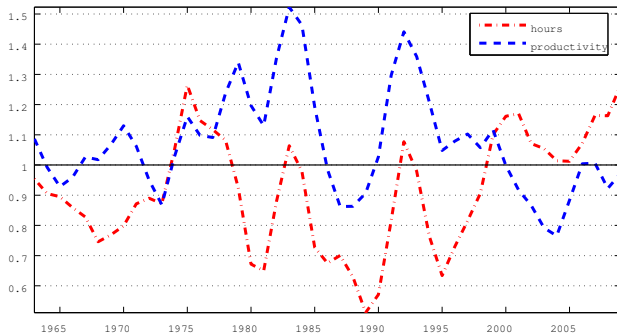
Fig: Rates of Bankruptcy and Opening



# Five Phases

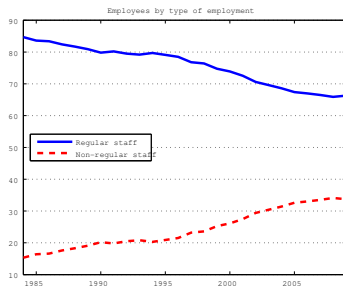
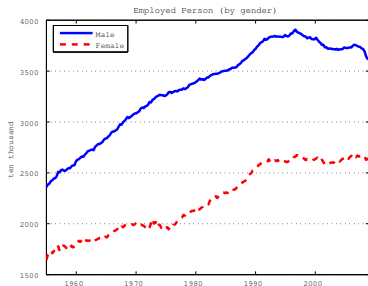
- *4th Phase: Early-1990s to Mid-2000s*
  - 1 stable output (GM period)
  - 2 labor input  $\uparrow$
  - 3 labor productivity  $\downarrow$
  
- *5th: Late-2009s*
  - 1 Global crisis
  - 2 all volatile

Fig: Relative SDs



- SD of labor input  $\uparrow$  in the 4th and 5th Phases

Fig: Employed Person (by gender) and Employees (by type of employment)

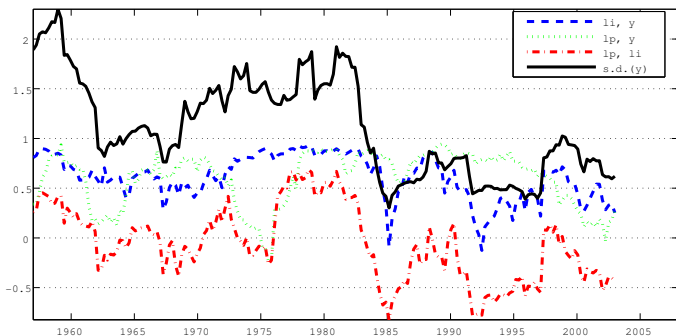


- Participation Rate: Male  $\uparrow$  until 1975, Female  $\uparrow$  from 1975
- Participation Rate of Non-regular staff  $\uparrow$
- Saito (2003): home production: husband  $\downarrow$   $\rightarrow$  wife  $\uparrow$
- Gaston and Kishi (2007): long time working by part-time workers

# Comparison with the U.S.



Fig: Rolling Correlations of the U.S.



- procyclical movement of productivity under NT shocks
- sign changes of correlation btw  $l_i$  and  $l_p$

## Relationship with theory

- Relationship with labor hoarding models
  - ① Disutility from work depends on employment ( $n$ ), hours ( $h$ ), and effort ( $e$ ):
 
$$v(n_t, h_t, e_t) \equiv n_t \left( \frac{\lambda_h}{1+\eta_h} h_t^{1+\eta_h} + \frac{\lambda_e}{1+\eta_e} e_t^{1+\eta_e} \right)$$
  - ②  $y_t = a_t n_t (h_t e_t)^\alpha$   
 $= a_t n_t h_t^\phi$  where  $\phi = \alpha(1 + \frac{\eta_h}{1+\eta_e})$ .
  
- U.S. with the labor-search model
  - ① Hiring cost  $\Downarrow$ .
  - ② substitution from  $e_t$  to  $n_t$  and  $h_t$ .

# Hours and Employment

- Negative Correlations of  $l_p$  and  $l_i$  in all periods  
(Contributed by NT shocks)
- the labor-search model: Not Our Story in Japan

