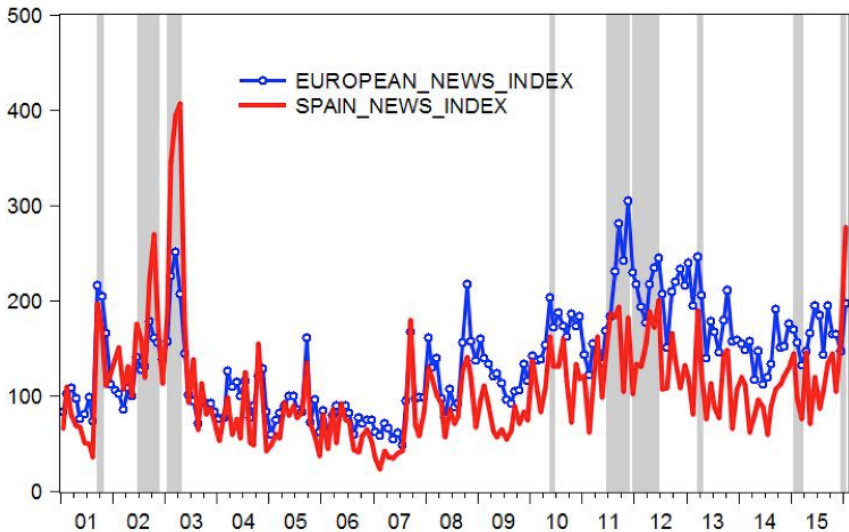


# Empirical Evidence

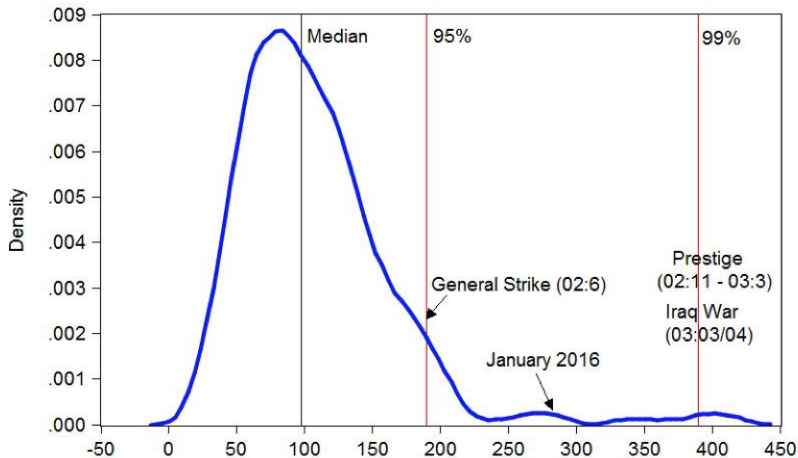
Jesús Fernández-Villaverde

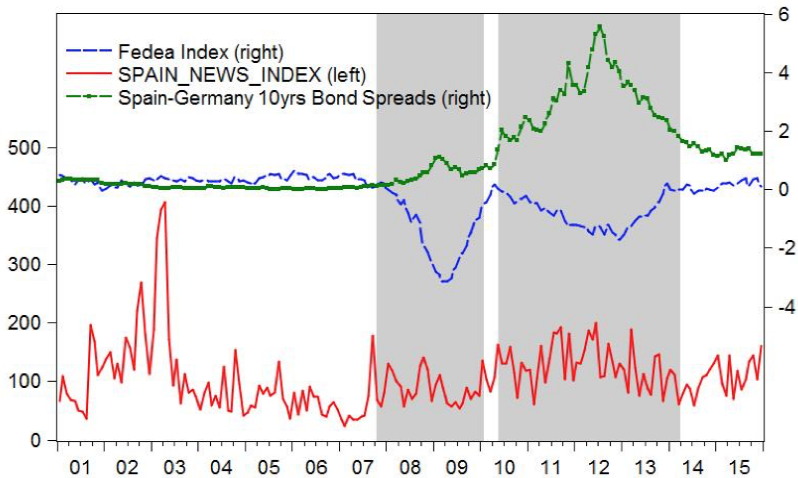
University of Pennsylvania

March 7, 2016

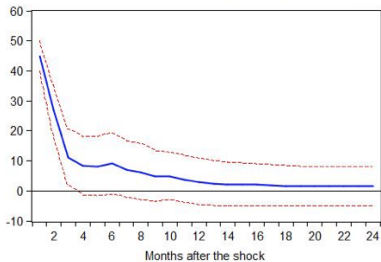


# Kernel Density of SPAIN\_NEWS\_INDEX (Full Sample: 2001:1-2016:1)

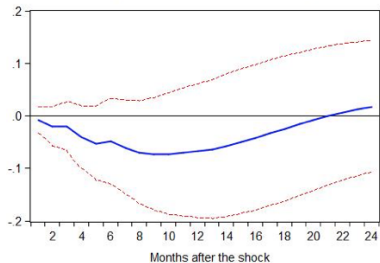




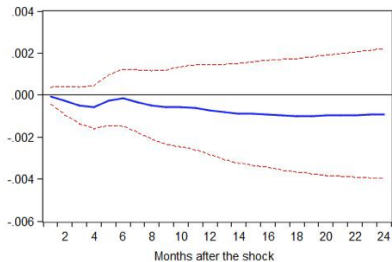
The size of the shock to SPAIN\_NEWS\_INDEX (1-sigma)



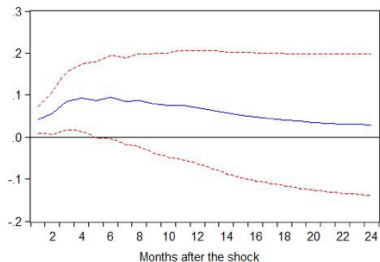
Response of FEDEA Index of Activity



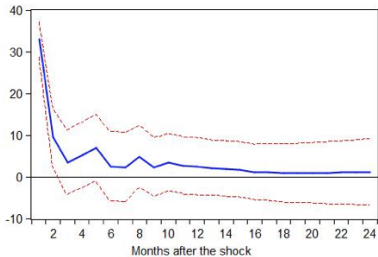
Response of Prices (HCPI)



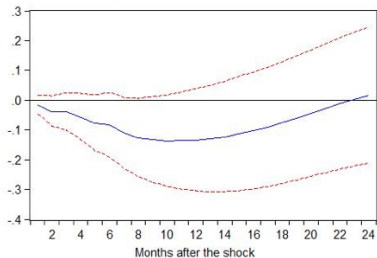
Response of Spreads



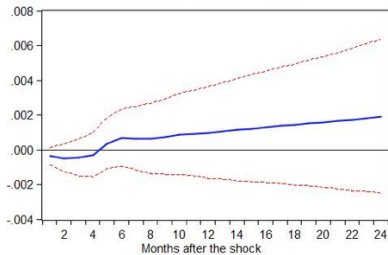
The size of the shock SPAIN\_NEWS\_INDEX (1-sigma)



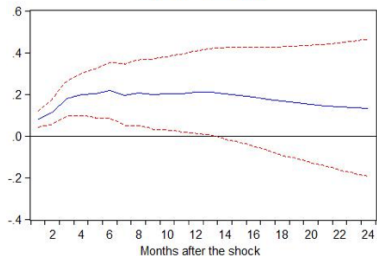
Response of FEDEA Index of Activity



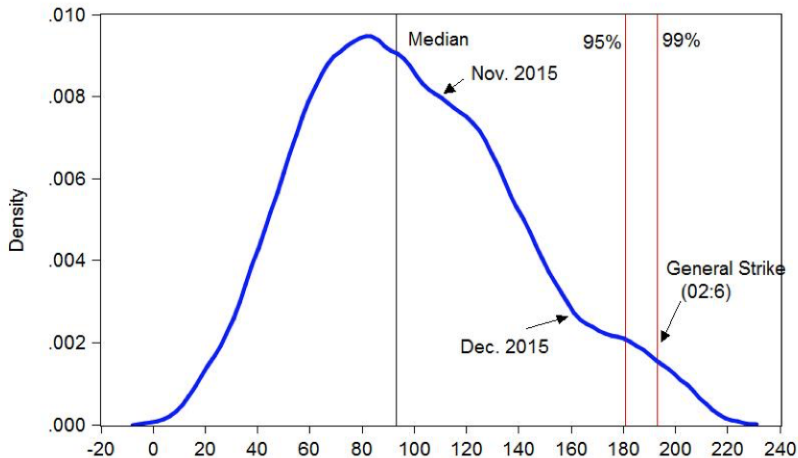
Response of Prices (HICP)



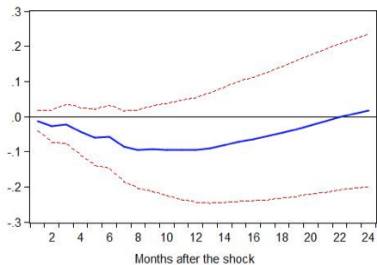
Response of Spreads



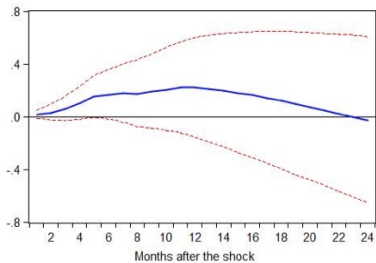
### Kernel Density of SPAIN\_NEWS\_INDEX (Sample Period: 2005:1- 2015:12)



Response of FEDEA Index of Activity



Response of Unemployment Changes





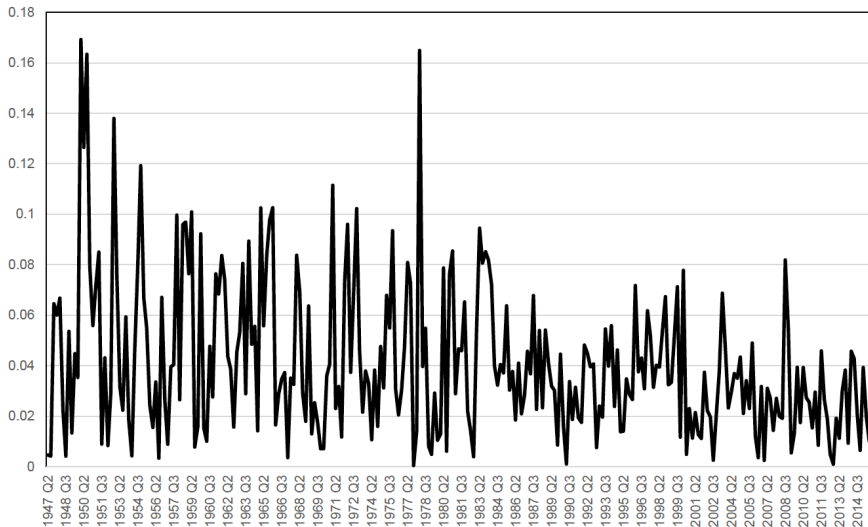
## A long tradition...

- Applied macroeconomic is concerned with the effects of shocks on certain key variables.
- Shocks have been characterized by temporary changes in the conditional mean of stochastic processes feeding our models.
  - ① The RBC program analyzes the consequences of temporary changes in the conditional mean of productivity (Kydland and Prescott, 1982).
  - ② Monetary models are focused on the effects of temporary changes in the conditional mean of innovations to the nominal interest rates (Woodford, 2003, or Christiano, Eichenbaum, and Evans, 2005).
  - ③ International devotes time to understand temporary changes in the conditional mean of the real interest rate (Mendoza, 1991 or Neumeyer and Perri, 2005) or the terms of trade (Mendoza, 1995).

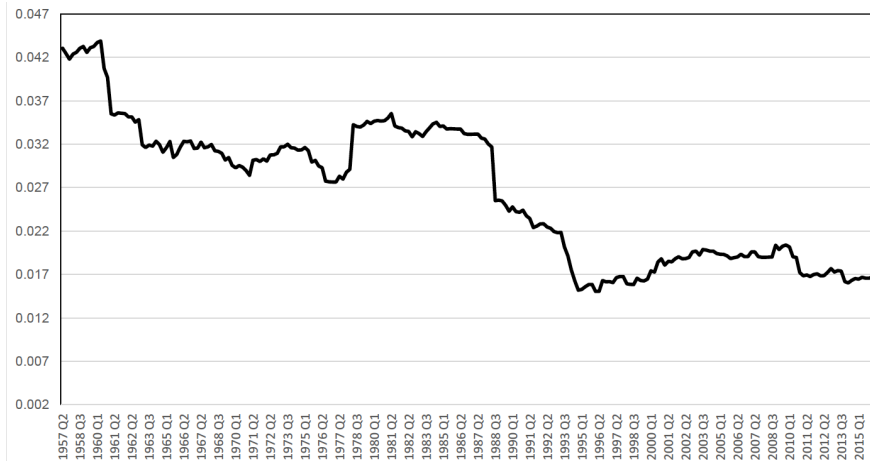
## ...and a continuation

- More recently, applied macroeconomists have started moving their attention towards situations where shocks are characterized by temporary changes in the conditional second moments of the stochastic processes.
- In particular, time-varying standard deviations.
- A first motivation for this move comes from the realization that time series have a strong time-varying variance component.
- Perhaps the most famous of those episodes was “the great moderation” of aggregate fluctuations that the U.S. economy.

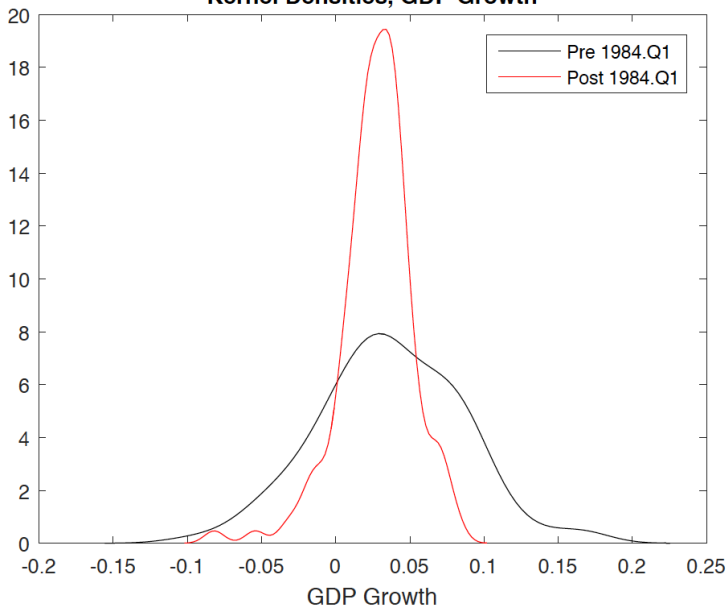
# GDP growth



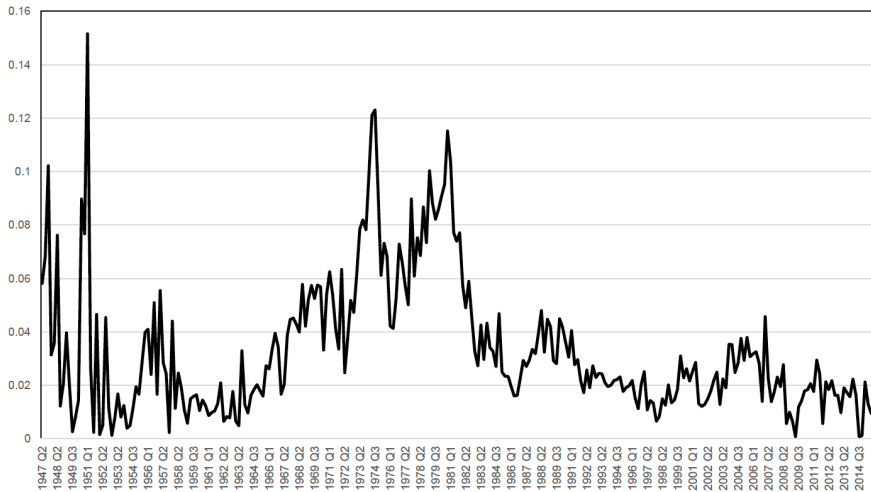
# GDP volatility



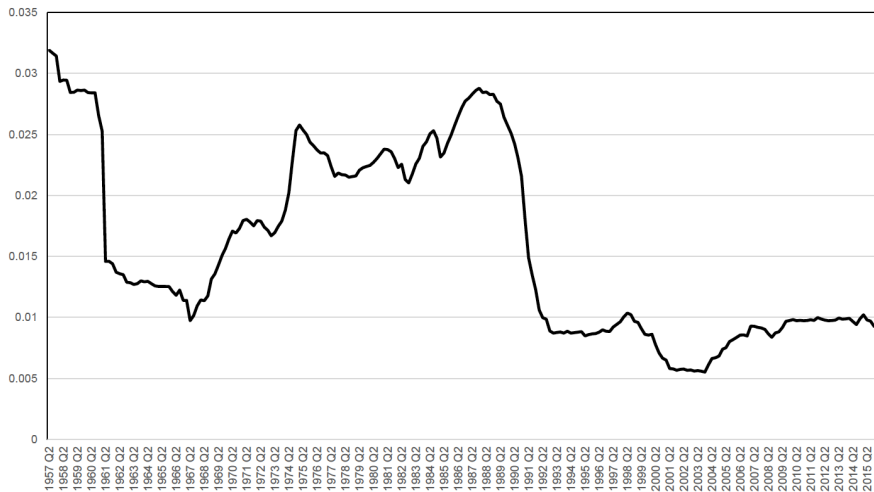
## Kernel Densities, GDP Growth



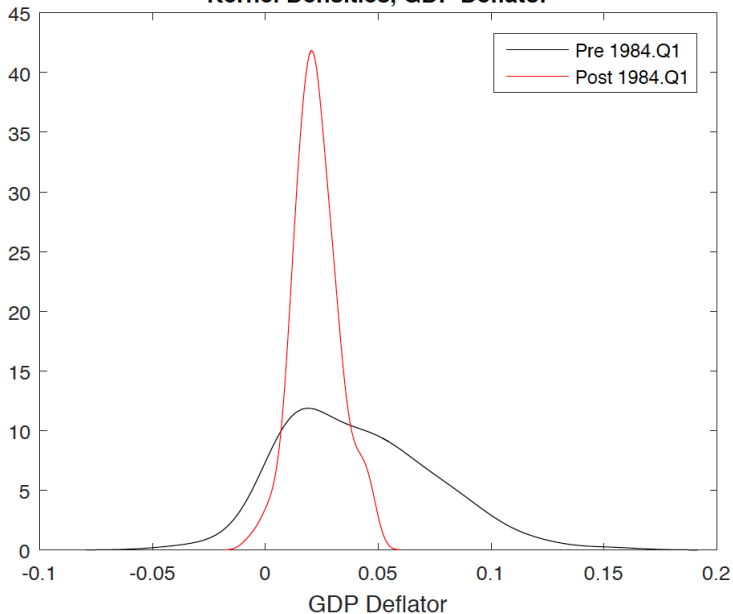
# GDP deflator



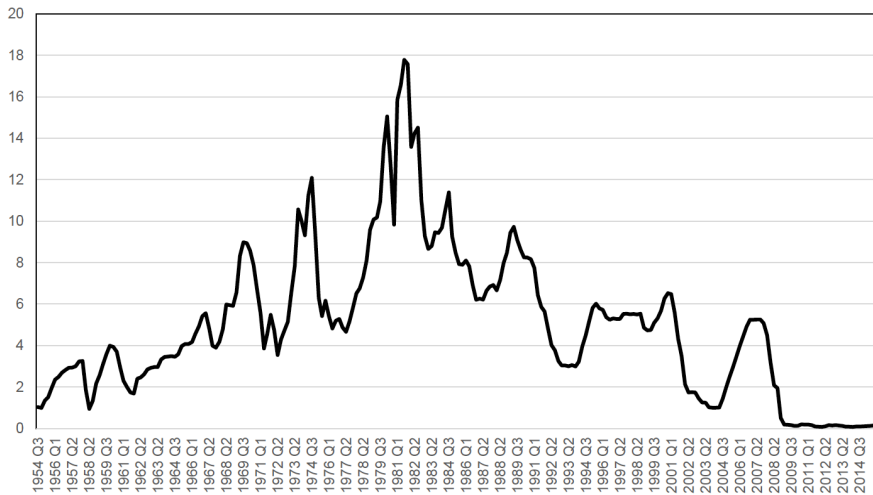
# GDP deflator volatility



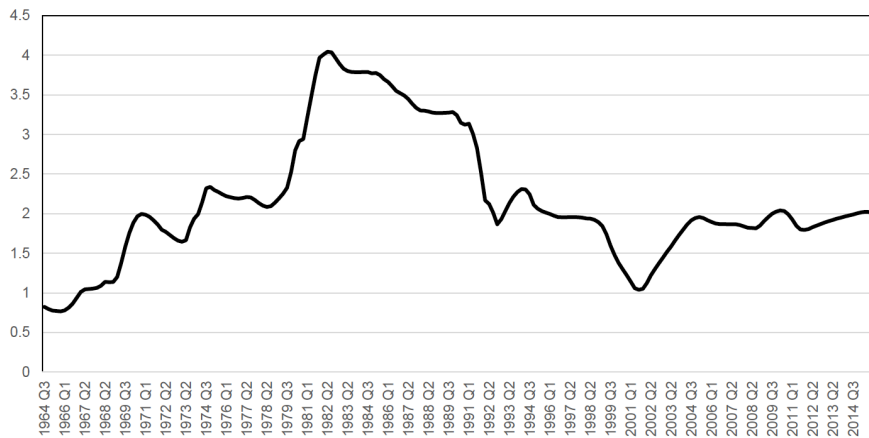
## Kernel Densities, GDP Deflator



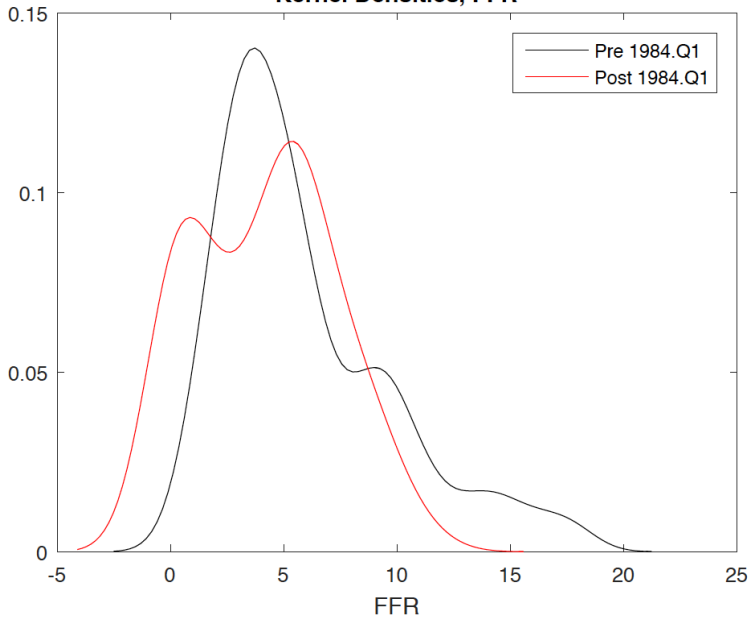




# FFR volatility



## Kernel Densities, FFR

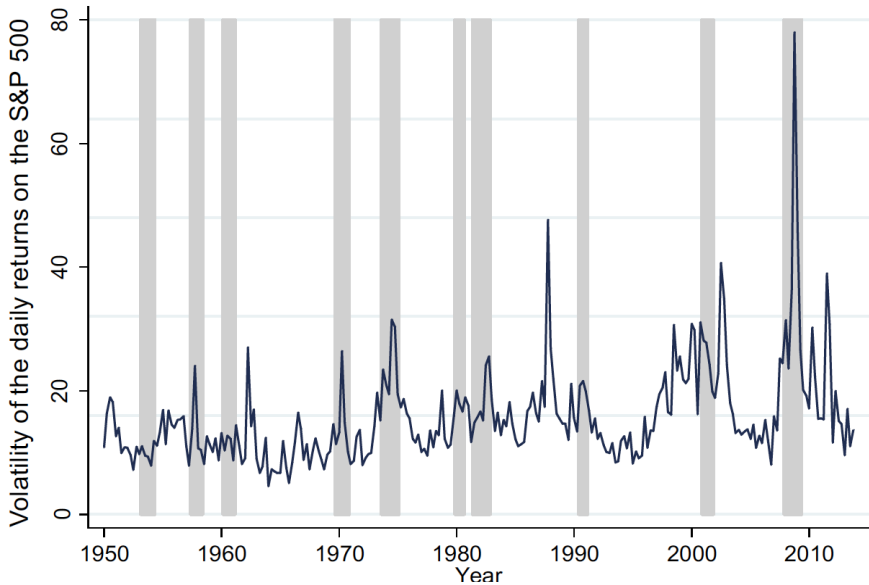


## Changes in Volatility of U.S. Aggregate Variables

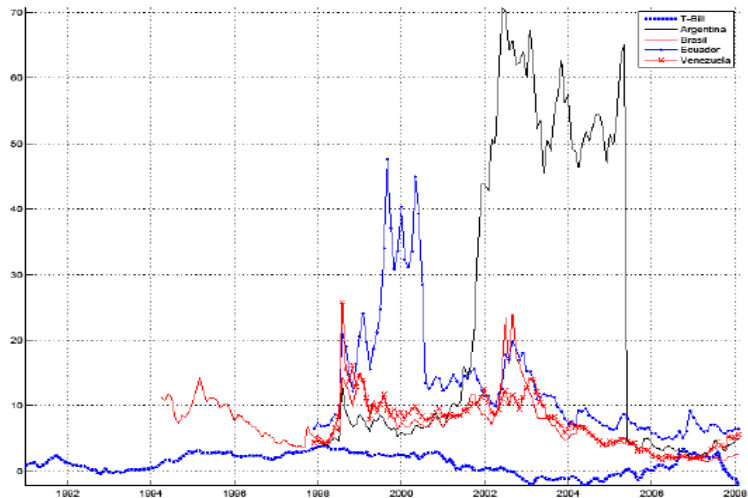
	Means		
	Inflation	Output Growth	FFR
All sample	3.2427	3.2375	5.0157
Pre 1984.Q1	4.1082	3.6742	5.9683
After 1984.Q1	2.2488	2.7359	4.1449
Post-1984.Q1/pre-1984.Q1	0.5474	0.7446	0.6945

## Changes in Volatility of U.S. Aggregate Variables

	Standard Deviations		
	Inflation	Output Growth	FFR
All sample	2.6360	3.9327	3.5662
Pre 1984.Q1	3.2440	4.8338	3.8809
After 1984.Q1	1.016	2.4561	3.0128
Post-1984.Q1/pre-1984.Q1	0.3130	0.5081	0.7763



# Interest rates



# Stochastic volatility I

- Stochastic volatility:

$$x_t = \rho x_{t-1} + \sigma_t \varepsilon_t, \quad \varepsilon_t \sim \mathcal{N}(0, 1).$$

and

$$\log \sigma_t = (1 - \rho_\sigma) \log \sigma + \rho_\sigma \log \sigma_{t-1} + (1 - \rho_\sigma^2)^{\frac{1}{2}} \eta u_t, \quad u_t \sim \mathcal{N}(0, 1).$$

- This can be a process for many observable  $x_t$ : productivity, taxes, asset returns.
- Level innovations vs. volatility innovations.
- Interpretation.
- Non-linear structure.
- Discrete time process. Alternative with diffusion processes in continuous time.



# Stochastic volatility II

- Richer specifications:
  - ① More lags and moving average components.
  - ② Additional regressors.
  - ③ VAR(MA)-SV.
  - ④ Non-Gaussian innovations.
  - ⑤ Correlation among innovations.
  - ⑥ Threshold effects.
  - ⑦ Asymmetries.

## Other specifications I

- Markov-regime switching models:

$$\sigma_t \in [\sigma_1, \dots, \sigma_n]$$

with transition matrix

$$P_{ij} = \begin{bmatrix} p_{11} & \dots & p_{1n} \\ \vdots & & \vdots \\ p_{n1} & \dots & p_{nn} \end{bmatrix}$$

- Advantages and disadvantages (econometric and theoretical).
- Mixed-models.

## Other specifications II

- GARCH(p,q):

$$x_t = \rho x_{t-1} + a_t$$

where

$$a_t = \sigma_t \varepsilon_t, \quad \varepsilon_t \sim \mathcal{N}(0, 1)$$

and

$$\sigma_t = \sqrt{\omega + \sum_{i=1}^p \alpha_i a_{t-i}^2 + \sum_{i=1}^q \beta_i \sigma_{t-i}^2}$$

- Advantages and disadvantages (econometric and theoretical).
- Dozens of possible variations.

## A real life example

- Decomposition of interest rates:

$$r_t = \underbrace{r}_{\text{mean}} + \underbrace{\varepsilon_{tb,t}}_{\text{T-Bill shocks}} + \underbrace{\varepsilon_{r,t}}_{\text{Spread shocks}}$$

- $\varepsilon_{tb,t}$  and  $\varepsilon_{r,t}$  follow:

$$\varepsilon_{tb,t} = \rho_{tb}\varepsilon_{tb,t-1} + e^{\sigma_{tb,t}}u_{tb,t}, \quad u_{tb,t} \sim \mathcal{N}(0, 1)$$

$$\varepsilon_{r,t} = \rho_r\varepsilon_{r,t-1} + e^{\sigma_{r,t}}u_{r,t}, \quad u_{r,t} \sim \mathcal{N}(0, 1)$$

- $\sigma_{tb,t}$  and  $\sigma_{r,t}$  follow:

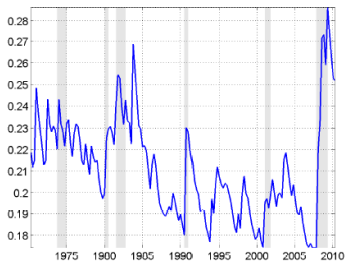
$$\sigma_{tb,t} = \left(1 - \rho_{\sigma_{tb}}\right)\sigma_{tb} + \rho_{\sigma_{tb}}\sigma_{tb,t-1} + \eta_{tb}u_{\sigma_{tb,t}}, \quad u_{\sigma_{tb,t}} \sim \mathcal{N}(0, 1)$$

$$\sigma_{r,t} = \left(1 - \rho_{\sigma_r}\right)\sigma_r + \rho_{\sigma_r}\sigma_{r,t-1} + \eta_r u_{\sigma_{r,t}}, \quad u_{\sigma_{r,t}} \sim \mathcal{N}(0, 1)$$

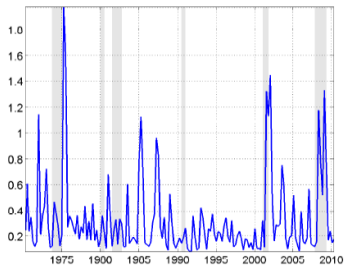
## An alternative motivation

- A second motivation for this move is that temporary changes in the conditional standard deviation of shocks can capture the spreading out of distributions of events in the future.
- For example, an increase in the variance of future paths of fiscal policy can be captured by a temporary increase in the standard deviation of the innovations to some fiscal policy rules.
- Similarly, the higher volatility of sovereign debt markets as the one currently observed can be included in our models as a temporary increase in the standard deviation in the innovations to a country-specific spread.

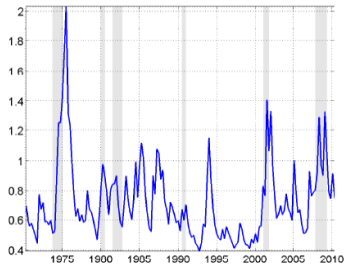
## Government spending



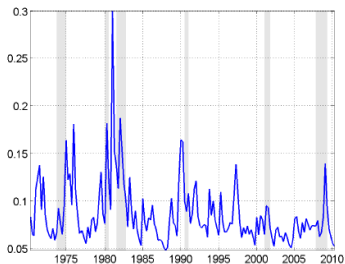
## Labor Tax



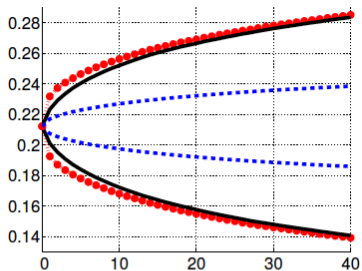
## Capital Tax



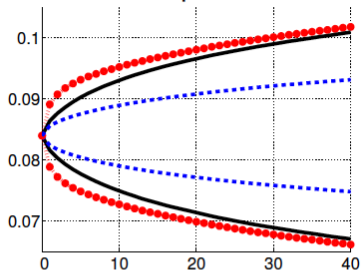
## Consumption Tax



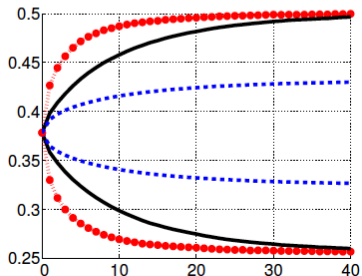
### Labor Tax



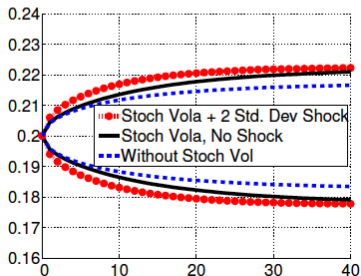
### Consumption Tax



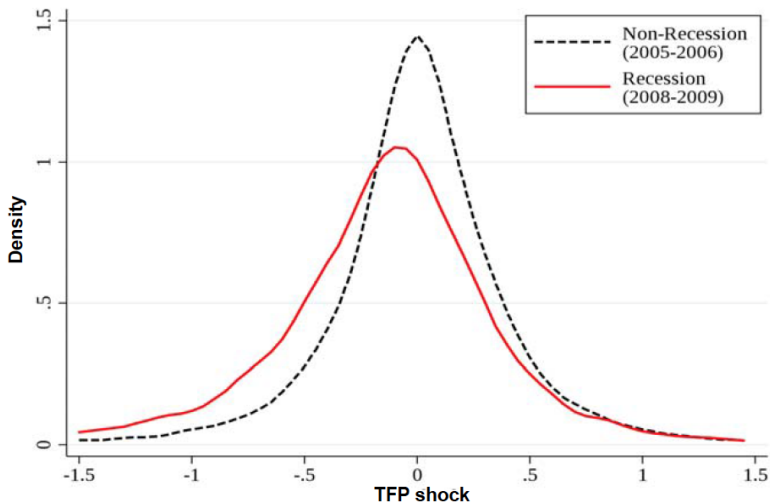
### Capital Tax



### Government spending



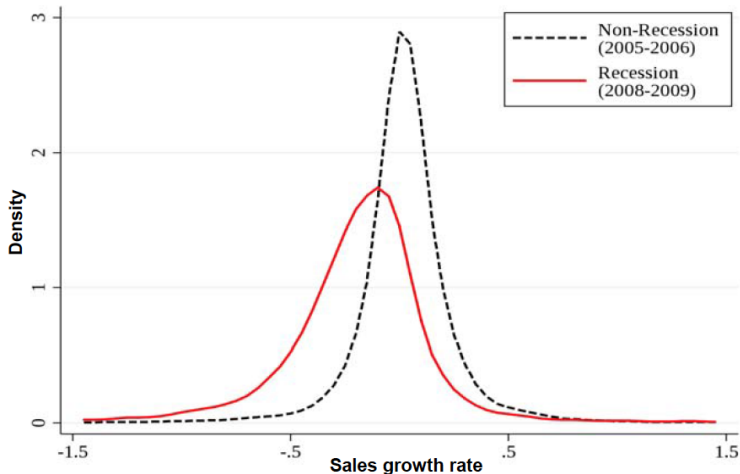
**Figure 1: The variance of establishment-level TFP shocks increased by 76% in the Great Recession**



**Notes:** Constructed from the Census of Manufactures and the Annual Survey of Manufactures using a balanced panel of 15,752 establishments active in 2005-06 and 2008-09. Moments of the distribution for non-recession (recession) years are: mean 0 (-0.166), variance 0.198 (0.349), coefficient of skewness -1.060 (-1.340) and kurtosis 15.01 (11.96). The year 2007 is omitted because according to the NBER the recession began in December 2007, so 2007 is not a clean “before” or “during” recession year.

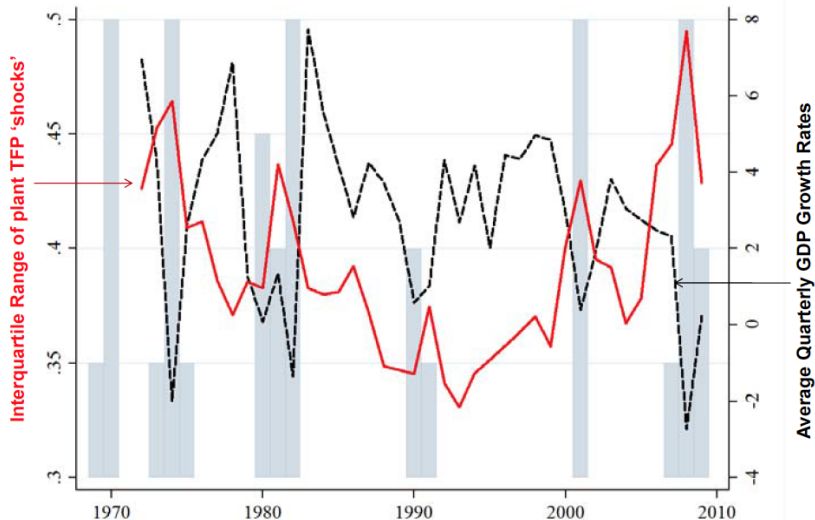


**Figure 2: The variance of establishment-level sales growth rates increased by 152% in the Great Recession**



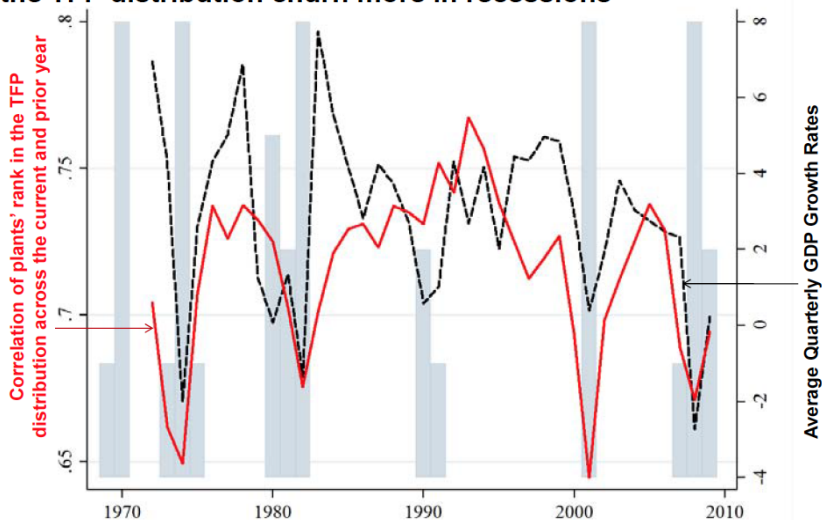
**Notes:** Constructed from the Census of Manufactures and the Annual Survey of Manufactures using a balanced panel of 15,752 establishments active in 2005-06 and 2008-09. Moments of the distribution for non-recession (recession) years are: mean 0.026 (-0.191), variance 0.052 (0.131), coefficient of skewness 0.164 (-0.330) and kurtosis 13.07 (7.66). The year 2007 is omitted because according to the NBER the recession began in December 2007, so 2007 is not a clean "before" or "during" recession year.

**Figure 3: TFP 'shocks' are more dispersed in recessions**



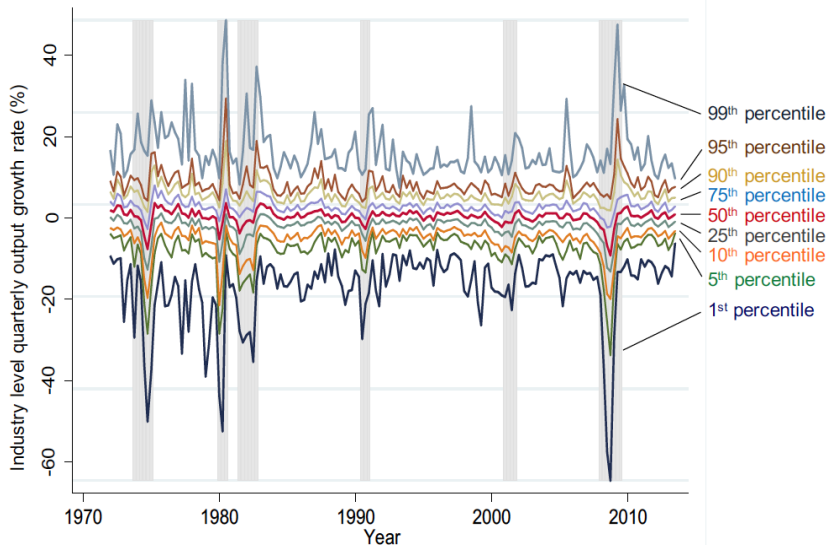
**Notes:** Constructed from the Census of Manufactures and the Annual Survey of Manufactures establishments, using establishments with 25+ years to address sample selection. Grey shaded columns are share of quarters in recession within a year.

**Figure 4: Recessions increase turbulence: plant rankings in the TFP distribution churn more in recessions**



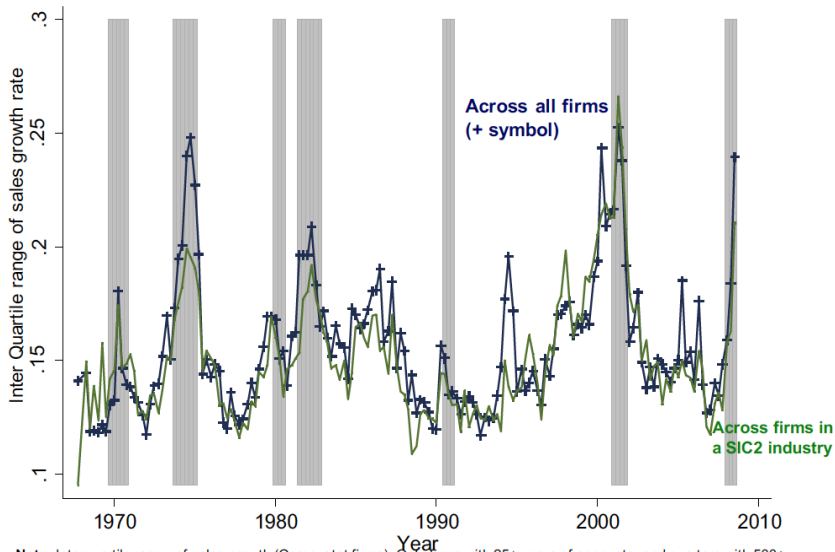
**Notes:** Constructed from the Census of Manufacturers and the Annual Survey of Manufacturing establishments, using establishments with 25+ years to address sample selection. Grey shaded columns are share of quarters in recession within a year. Plants' rank in the TFP distribution is their decile within the industry and year TFP ranking.

## Industry growth dispersion (by month)



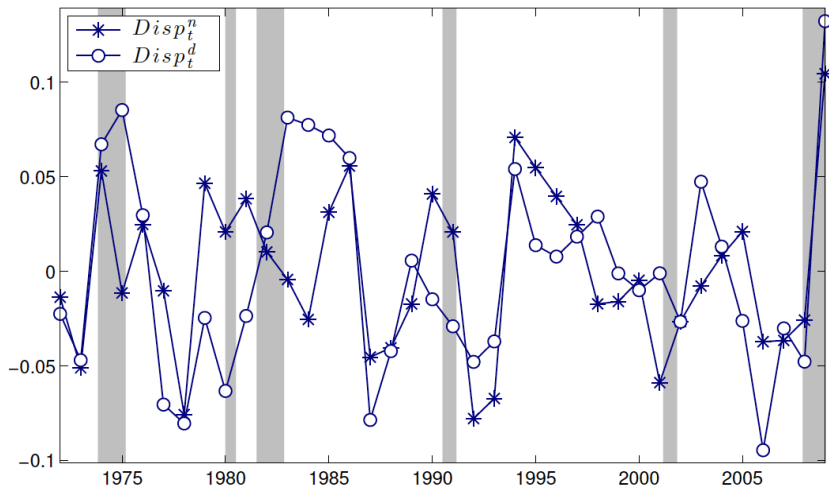
**Note:** 1<sup>st</sup>, 5<sup>th</sup>, 10<sup>th</sup>, 25<sup>th</sup>, 50<sup>th</sup>, 75<sup>th</sup>, 90<sup>th</sup>, 95<sup>th</sup> and 99<sup>th</sup> percentiles of 3-month growth rates of industrial production within each quarter. All 196 manufacturing NAICS sectors in the Federal Reserve Board database. Source: Bloom, Floetotto and Jaimovich (2009)

## Firm growth dispersion (by quarter)



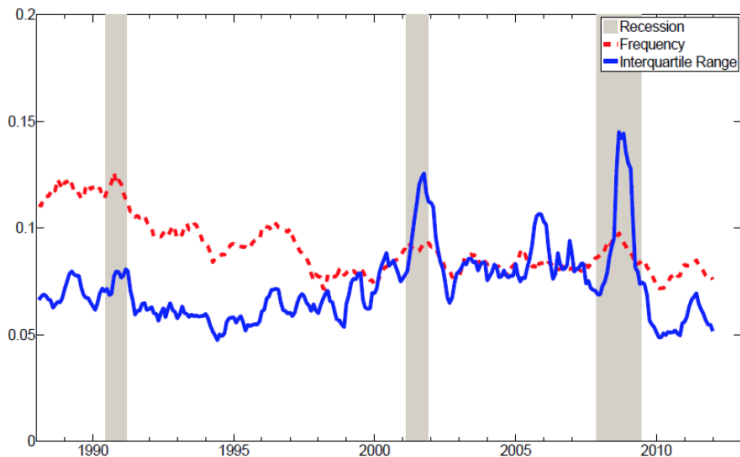
**Note:** Interquartile range of sales growth (Compustat firms). Only firms with 25+ years of accounts, and quarters with 500+ observations. SIC2 only cells with 25+ obs. SIC2 is used as the level of industry definition to maintain sample size. The grey shaded columns are recessions according to the NBER. Source: Bloom, Floetotto, Jaimovich, Saporta and Terry (2011)

Figure 1: Dispersion in productivity levels



# Product level price dispersion (by quarter)

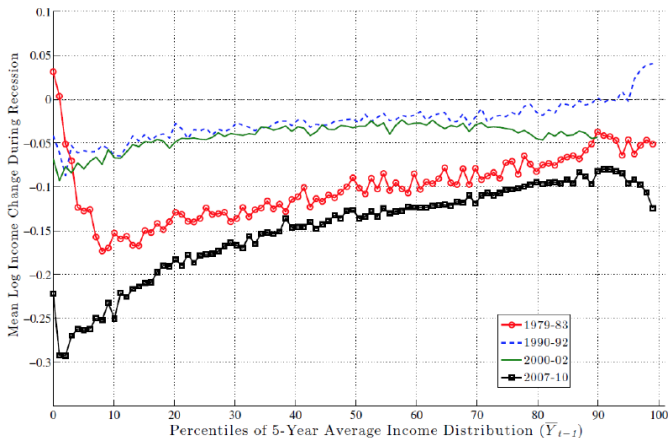
Figure 1: Price Changes Across Time



Data is seasonally adjusted using 12 monthly dummies and smoothed with a 6 month moving average.  
Frequency is the Median Frequency of Adjustment.

**Source:** Joe Vavra (2014, QJE) "Inflation dynamics and time varying volatility"

## But SSA data on several million individuals shows rising 3<sup>rd</sup> moment but flat 2<sup>nd</sup> moment in recessions

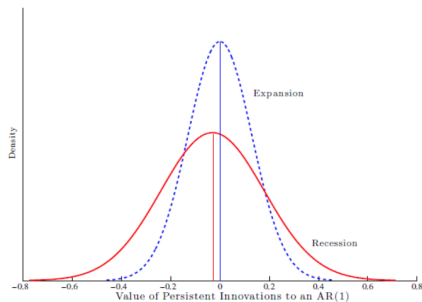


Guvenen, Ozkan & Song, "The nature of countercyclical income risk" (2014, JPE)

Notes: Uses about 5m obs per year from the US Social Security Administration earnings data

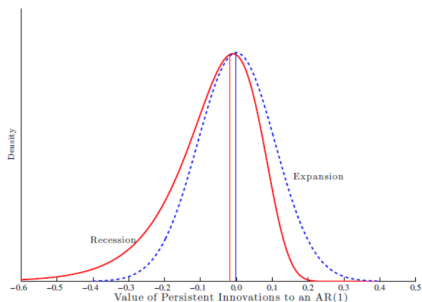


## Countercyclical Variance



Macro, industry, firms,  
plants and prices

## Countercyclical Left-Skewness

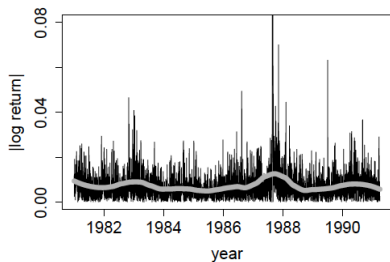


Wages

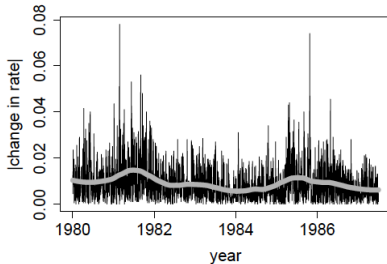
# Literature I

- In one form or another, economists have talked for a long time about time-varying volatility.
- A breakthrough came with Engle's (1982) paper on autoregressive conditional heteroscedasticity, or ARCH.
- Engle postulated that the evolution of variance over time of time series  $x_t$  was an autoregressive process that is hit by the square of the (scaled) innovation on the level of  $x_t$ .
- The application in Engle's original paper was the estimation of an ARCH process for British inflation.
- Early indication that this was a central issue in macroeconomics.
- But it was not in macro where ARCH models came to reign: the true boom was in finance.

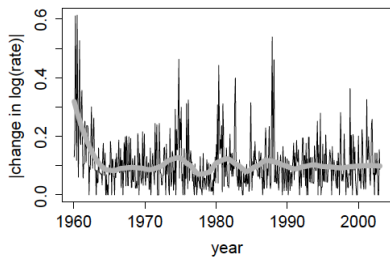
**S&P 500 daily return**



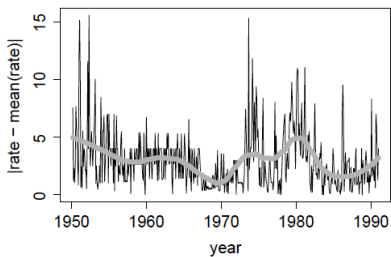
**BP/dollar exchange rate**



**Risk-free interest rate**



**Inflation rate**



## Literature II

- The situation changed after **Kim and Nelson (1998)**, **McConnell and Pérez-Quirós (2000)**, and **Blanchard and Simon (2001)**.
- Documented that the volatility of U.S. aggregate fluctuations had changed over time. **Stock and Watson (2002)** named this phenomenon “the great moderation.”
- **Sims and Zha (2006)** estimated a structural vector autoregression (SVAR) with Markov-regime switching both in the autoregressive coefficients and in the variances of the disturbances.
- They concluded that models with shocks that have time-varying volatilities are a key in applied macroeconomics.

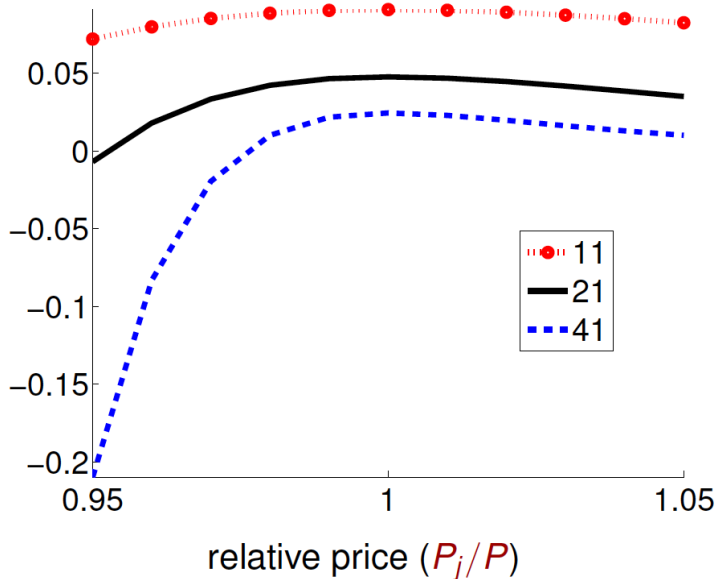
## Literature III

- Big boom, however, is after Bloom (2009).
- Many papers after it (including mine!).
- We will discuss some of them as we go along.
- There are:
  - ① Methodological issues (solution, estimation).
  - ② Data.
  - ③ Conceptual: endogenous vs. exogenous uncertainty, beliefs vs. DGP.
  - ④ Economic intuition.

# Mechanisms behind uncertainty shocks

- ① Utility function.
- ② Price decisions.
- ③ Oi-Hartman-Abel effect.
- ④ Option value effect.
- ⑤ Ss-rules.
- ⑥ Non-conventional preferences, [Gilboa and Schmeidler \(1989\)](#).

## Period profits



## Oi-Hartman-Abel effect

- Oi (1961), Hartman (1972) and Abel (1983).
- A higher variance of productivity increases investment, hiring, and output because the optimal capital and labor choices are convex in productivity.
- Example:

$$y = Ak^\alpha l^\beta$$

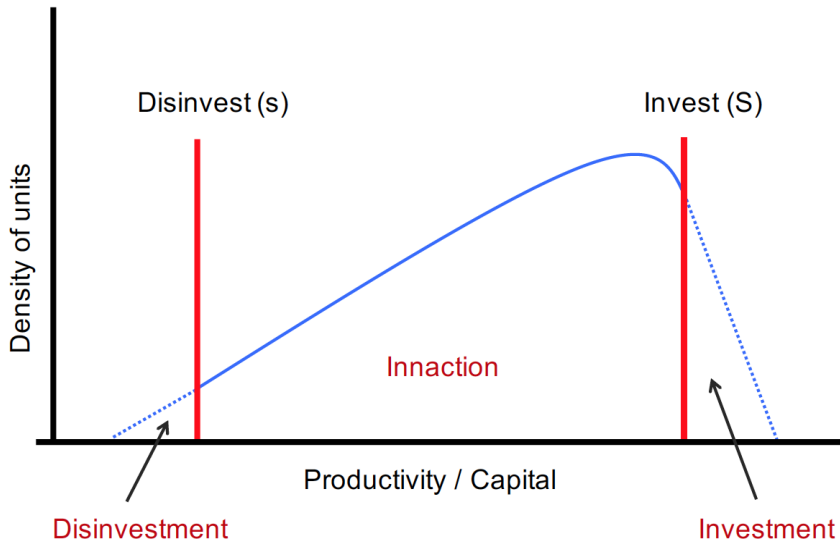
where  $\alpha + \beta < 1$ .

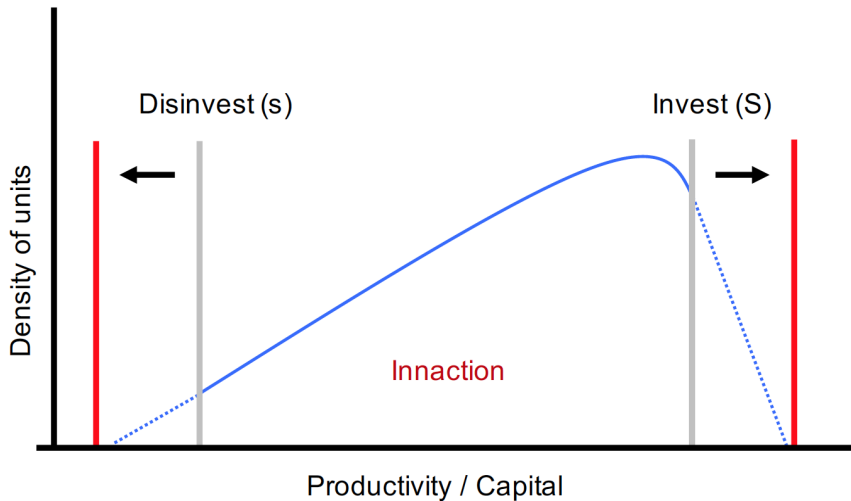
- Then:

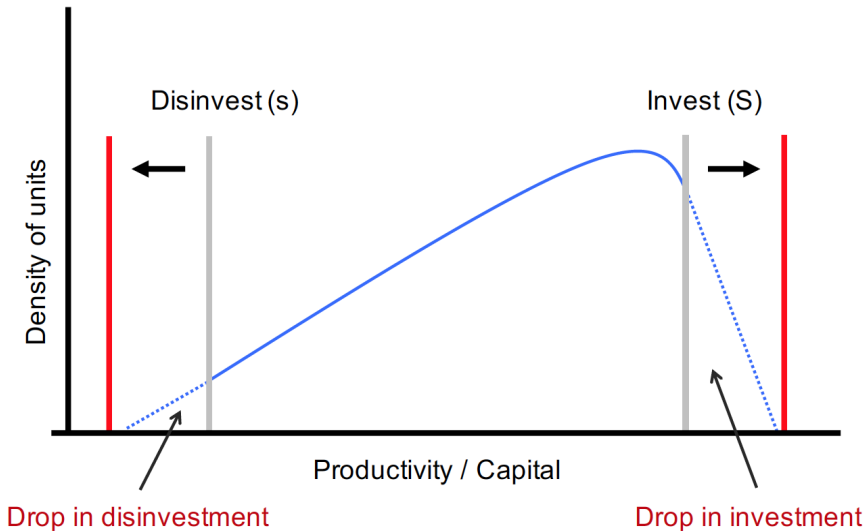
$$k^* = \mu_1 A^{\frac{1}{1-\alpha-\beta}}$$

$$l^* = \mu_2 A^{\frac{1}{1-\alpha-\beta}}$$









# Ambiguity aversion

- Agents do not know dispersion of shocks.
- Problem:

$$V(k, z) = \max_{c, l, k'} \left[ u(c_t, l_t) + \beta \min_{\lambda \in [\underline{\sigma}_t, \bar{\sigma}_t]} \mathbb{E}_{k, z} V(k', z') \right]$$
$$\text{s.t. } c + k' = e^z k^\alpha l^{1-\alpha} + (1 - \delta)k$$
$$z' = \lambda z + \sigma_t \varepsilon'$$

- Intuition.