Endogenous Technology Adoption and R&D as Sources of Business Cycle Persistence

Anzoategui, Comin, Gertler, and Martinez

Discussion by Christopher Tonetti Stanford GSB

Conference on Macroeconomics and Monetary Policy Federal Reserve Bank of San Francisco 4 March 2016 Why did output growth slow down persistently during Great Recession?

- Reduced physical capital accumulation?
- Reduced labor supply?
- Reduced productivity growth?
- 2 TFP growth dropped strongly before and during recession
- 3 What contributed to the drop in TFP growth
 - Reduced expenditure on innovation (R&D)?
 - Reduced efficiency of innovation activity?
 - Reduced adoption of already invented ideas?
 - Exogenous movements beyond measured/filtered innovation and adoption behavior?
- 4 Key result: "Liquidity" shock reduced adoption activity

How to Judge the Paper?

- Plausible and intriguing idea about very important question
- Relatively easy to interpret the results
- Harder to understand what generates the results
- Free parameters or disciplined by data and cross-equation restrictions?
- Potential problem:
 - Fact: We measure productivity drops
 - Model: Add an unobserved factor with arbitrary time path that directly affects productivity
 - Result: Productivity drop in model is explained by drop in the unobserved factor
- The strength of the paper is determined by whether the unobserved factor of **adopted ideas** moves arbitrarily or not and if its movement is consistent with the movement of other observables

Main Suggestions to Authors

- More focus on showing the adoption process is tightly disciplined
 - More discussion on the calibration
 - More discussion on what is generating patterns in the filtered latent states
 - More analysis of behavior of non-targeted variables
 - More discussion of why so many bells and whistles are necessary and how they interact with novel features
- I need more help understanding what is generating quantitative results and convincing that this is the mechanism
- Great idea, very plausible, and consistent with interesting firm-level data (Andrews et. al. 2015)
- This will ultimately be a fantastic paper with big impact

What are the Results?

- The key outputs of the analysis are filtered latent states
 - Output growth is measured
 - TFP growth is calculated (basically) directly from measured variables
 - Shocks are filtered
 - Endogenous component of TFP, stock of potential ideas, and stock of adopted ideas are filtered
- Main result: Filtered time series for adoption rate falls sharply in recession driven by "liquidity" demand shock, driving the drop in GDP growth
- Filtered time series are the result of data + model + parameter values
- Next slides: What in the data/model/parameters is generating this result?

- Estimation to match model output to data on Y, C, I, L, W, π, r, R&D
- R&D is pro-cyclical, but not huge movement during recent recession
- Likely not decreased R&D that's causing the high frequency, large, prolonged drop in TFP growth
- Endogenous TFP is function of innovation (measured R&D) and adoption (no macro data)
- If not, R&D, maybe adoption decreased. Can't measure it, can try to infer if that is the case with **filtering**
- · First, some case-study evidence that adoption is pro-cyclical
 - This adoption data not used in structural analysis, just motivation

Model

- Ideas disappear at rate φ; skilled labor wage w_{st}
- Profit from producing intermediate good (monopolistic competition): Π_{mt}
- Value of adopted idea: $V_t = \prod_{mt} + \phi E_t[\Lambda_{t,t+1} V_{t+1}]$
- Adopters buy idea from innovator at cost J_t and adopt at rate λ_t (a function of L_{sat})
- Value of unadopted idea:

$$J_t = \max_{L_{sat}} E_t[\phi \Lambda_{t,t+1}[\lambda_t V_{t+1} + (1-\lambda_t)J_{t+1}] - w_{st}L_{sat}]$$

- Innovators create an idea with probability φ_t (a function of L_{sat})
- R&D FOC: $E_t[\Lambda_{t,t+1}\varphi_t J_{t+1}] = w_{st}$
- Adoption FOC: $Z_t \lambda' E_t [\phi \Lambda_{t,t+1} [V_{t+1} J_{t+1}]] = w_{st}$

The Mechanism: Procyclical Adoption

$$V_{t} = \prod_{mt} + \phi E_{t}[\Lambda_{t,t+1} V_{t+1}]$$

$$J_{t} = \max_{L_{sat}} E_{t}[\phi \Lambda_{t,t+1}[\lambda_{t} V_{t+1} + (1 - \lambda_{t})J_{t+1}] - w_{st}L_{sat}]$$

$$Z_{t}\lambda' E_{t}[\phi \Lambda_{t,t+1}[V_{t+1} - J_{t+1}]] = w_{st}$$

Adoption is procyclical

- *w*_{st} sticky
- In a boom, V_{t+1} increases more than $J_{t+1} \rightarrow \uparrow \lambda_t$
- V_{t+1} increases because profits Π_{mt} increases
- J_{t+1} increases because it is discounted V_{t+1} and persistence in boom
- J_{t+1} increases less because discounting, probability of adoption $\neq 1$, and mean reversion

Innovation and Adoption FOCs

$$E_t[\Lambda_{t,t+1}J_{t+1}] = \frac{W_{st}}{\chi_t Z_t L_{srt}^{\rho_z - 1}}$$
$$E_t[\phi \Lambda_{t,t+1}[V_{t+1} - J_{t+1}]] = \frac{W_{st}}{Z_t \lambda'} = \frac{W_{st}}{\psi_t Z_t L_{sqt}^{\rho_\lambda - 1}}$$

- Recall, R&D didn't change much (data), adoption tanks (filtered)
- Recall main finding is negative liquidity shock drives fall in adoption
- How will negative shock to Λ_{t,t+1} not affect R&D much but have big effect on adoption?
 - ϕ changes effective discount rate for adopters vs. innovators
 - Different elasticities of success w.r.t. labor expenditure (ρ_λ vs. ρ_z) means same change in SDF differentially affects adoption vs. innovation activity

The SDF Links Financial Prices and Adoption Behavior

$$E_{t}[\Lambda_{t,t+1}J_{t+1}] = \frac{W_{st}}{\chi_{t}Z_{t}L_{stt}^{\rho_{z}-1}}$$

$$E_{t}[\phi\Lambda_{t,t+1}[V_{t+1} - J_{t+1}]] = \frac{W_{st}}{\psi_{t}Z_{t}L_{sat}^{\rho_{\lambda}-1}}$$

$$1 = E_{t}[\Lambda_{t,t+1}R_{k,t+1}]$$

$$1 = E_{t}[\Lambda_{t,t+1}R_{t+1}] + \zeta_{t}$$

- SDF prices all assets: ideas, bonds, physical capital, etc.
- Does the SDF shock that matches R&D and TFP data hit short-term interest rate **levels** well?
- Is this long vs. short risk exposure? Does SDF shock hit long-short spreads well?
- Evidence on profit flows (V) vs. risk premia (Λ) over Great Recession?
- Can you provide more specifics on link between R_k and λ_t
- Great opportunity to use more finance analysis to prove not just free parameters to hit desired targets

Free parameters to hit desired targets or external discipline?

- Elasticity of adoption rate w.r.t. adoption expenditures to hit R&D to GDP ratio of 2.2%
 - ▶ ρ_λ = 0.95
 - Comin and Gertler (2006): "admittedly, this estimate is crude...[but] provides a plausible benchmark"
 - Elasticity is key parameter in business cycle analysis
 - Analysis of FOCs suggests robustness to this parameter is essential!
 - Especially if don't have strong evidence for baseline value
- 2. Average time from invention to adoption
 - $\bar{\lambda}$ s.t. 7 years on average
 - Great use of adoption data from Comin and Hobijn (2010)

The Parameters 2/2

- 3. Rate at which ideas disappear

 - Comin and Gertler (2006) use three percent annual
 - "Falls in the middle of a broad range of estimates... in the literature"
- 4. Elasticity of innovation rate. w.r.t. innovation expenditures part of joint estimation
 - ▶ *ρ_z* = 0.34
 - Low relative to literature: Girliches 0.6-1.0
 - Comin and Gertler (2006) use 0.8
 - Maybe not comparable due to quarter vs. year, but then also for adoption elasticity
 - Suggest calibrating this variable just like above three to 0.6 or 0.8 and check robustness
 - Ultimately: R&D won't move but adoption will move strongly because elasticities are low for R&D and high for adoption based on above calibration/estimation

The Results: TFP Drop is Endogenous



• Filtered endogenous TFP shock drives measured TFP

The Results: SDF Shock Works through Endogenous TFP



Filtered "liquidity" demand shock drives endogenous TFP

The Results: Adoption Drops Precipitously



· Filtered adoption rates drives drop in filtered endogenous TFP

Where do the Results Come From?

- Need to get drop in calculated TFP since it is in the data
- R&D doesn't drop much
- So either adoption drops hard or decrease in TFP is exogenous
- With very high elasticity of adoption, adoption drops hard
- Results crucially rest on elasticity of adoption, which is number picked by authors who previously said its measure is "admittedly crude."
 - Probably can't get better direct measures and need to make progress and not just give up
 - Authors are best game in town and on the right approach, but need to do more checks on implications of ρ_λ on observables to add convincing evidence this number is roughly right
 - Also do robustness to this number to check sensitivity of results
- Chance to make predictions:
 - $Z_t A_t$ is really big now
 - Predict fast recovery now that liquidity shock is gone (financial spreads are more normal)?