

ICT and Future Productivity: Evidence and Theory of a GPT

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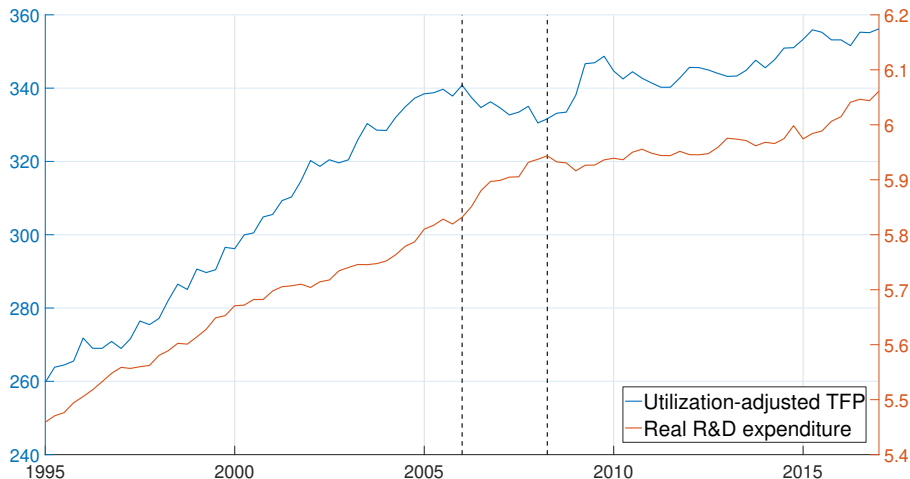
Boston College

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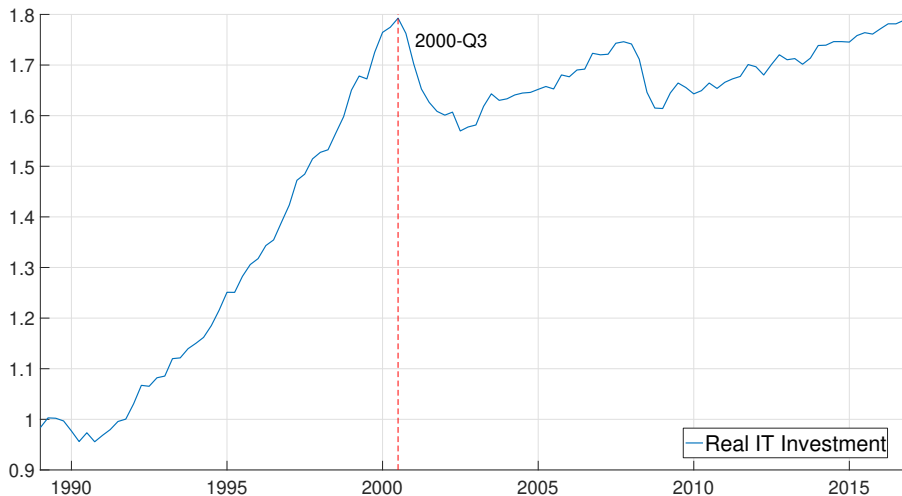
Motivation

- Point of departure: medium-run business cycle models à la Comin & Gertler (2006).
- Key prediction: BC fluctuations of a particular kind of investment (research & development (R&D) → adoption) lead total factor productivity (TFP).
- The Great Recession 2008 casts doubt on whether this is all of the story (Fernald et al. (2017)).

Wrong timing



Fernald et al. (2017): “R&D and adoption can’t be the whole story”.



→ A suspect is investment in information and communication technologies (ICT).

This paper

- 1 Stick ICT investment (ICT-I) in a VAR and explore how an identified shock to this affects TFP.

ICT shock = a shock to the productivity of the ICT sector today

→ A shock to ICT leads to substantial TFP increases over the medium-run.

- 2 Draw on the conclusions of the ICT literature to build a structural model to interpret the results.

→ ICT literature: ICT is a general-purpose technology (GPT).

⇒ Estimate a two-sector endogenous growth model to ask whether aggregate data supports this interpretation.

Related literature

① Medium-run business cycles

- Comin & Gertler (2006), Bianchi et al. (2014), Moran & Queralto (2017), Guerron & Jinnai (2015)

② ICT and productivity

- Jorgenson & Stiroh (1999), Oliner & Stichel (2000), Brynjolfsson & Lorin (2000), Stiroh (2002)

③ Identification of news shocks in SVARs

- Beaudry & Portier (2006), Barsky & Sims (2011)

④ Multi-sector growth models

- Greenwood, Hercowitz & Krusell (1997), Oulton (2007), Fisher (2006), Whelan (2003)

Roadmap

1 SVAR Analysis

2 Structural Model

1. SVAR analysis

We run a SVAR using aggregate, quarterly US data. The data vector is:

$$\mathbf{X}_t = \begin{bmatrix} TFP_t \\ ICTI_t \\ GDP_t \\ C_t \\ RP_t \end{bmatrix} \quad (1)$$

- $RP = \pi^{IT} / \pi^{CPI}$.
- All variables are real (except price indexes) and in log levels (except for RP, which is in growth rates).
- The dataset ranges from 1989:q1 - 2017:q2.

Baseline identification

$$\max_{\gamma_j} \Pi_{2,j}(0) = e_2' \tilde{A}_0 \gamma_j \quad (2)$$

subject to

$$\Pi_{1,j}(0) = e_1' \tilde{A}_0 \gamma_j = 0, \quad \text{and} \quad (3)$$

$$\gamma_j' \gamma_j = 1 \quad (4)$$

where j represents the arbitrary position of the ICT shock.

- (2) maximal impact effect on ICTI.
- (3) no impact effect on TFP.
- (4) γ_j comes from an orthogonal matrix.

VAR

Why this identification?

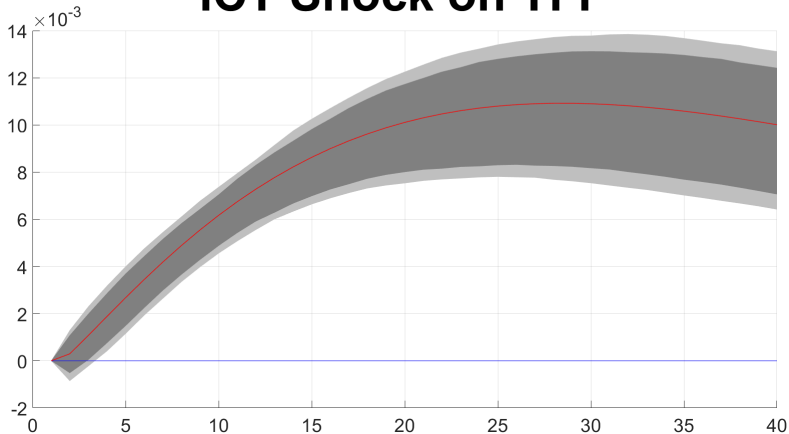
- ICT value added is less than 5% of GDP (BEA, April 2018).
→ ICT-I increases shouldn't affect TFP on impact.
- Prediction of multisector models (GHK): sectoral productivity increase leads to sectoral output becoming cheaper.
→ ICT-I should rise after ICT productivity shock.

Our favorite specification

- Recall: dataset is quarterly and covers 1989:q1-2017-q2.
- One lag (as suggested by BIC and HQ).
- Horizon of FEV-maximization: 60 quarters.

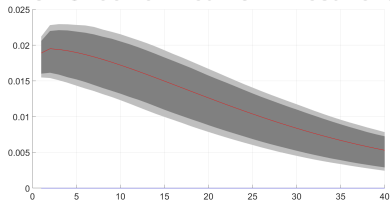
Results

ICT Shock on TFP

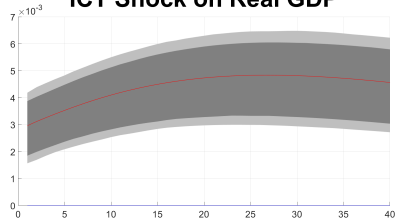


Results II

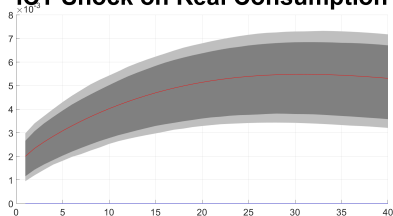
ICT Shock on Real ICT Investment



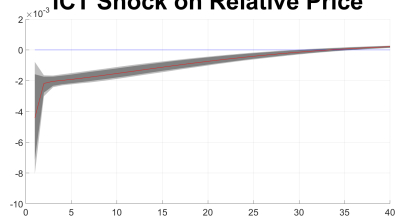
ICT Shock on Real GDP



ICT Shock on Real Consumption



ICT Shock on Relative Price



Results for a larger-scale VAR

Results III

	$h = 1$	$h = 4$	$h = 8$	$h = 16$	$h = 24$	$h = 40$
TFP	0	0.0023	0.0194	0.1088	0.2273	0.3382
ICT-I	0.9997	0.9038	0.7964	0.6320	0.5310	0.4371
Real GDP	0.2620	0.3061	0.3486	0.3936	0.4046	0.3881
Real C	0.1952	0.2638	0.3219	0.3931	0.4188	0.4064
Relative Prices	0.0618	0.0967	0.1276	0.1511	0.1516	0.1467

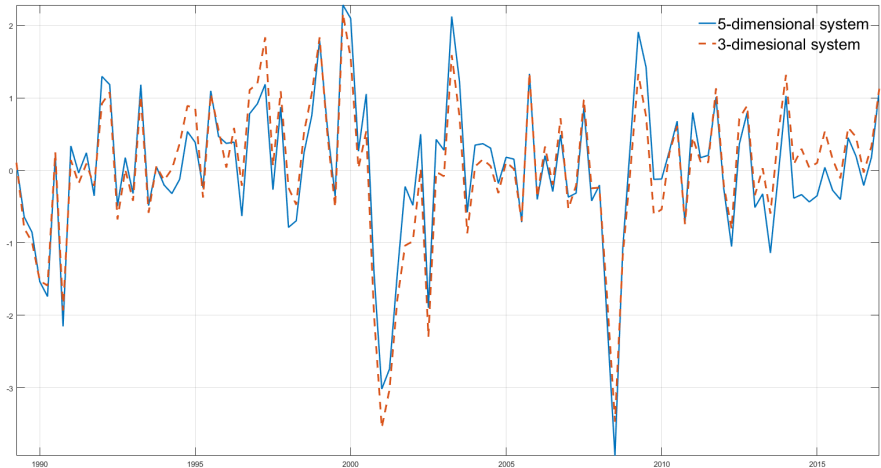
Robustness checks

- Main critique: reverse causality from news about future TFP.
- Alternative specification filters out news to recover an alternative ICT shock series.

Alternative specification

The two recovered ICT shock series

Series of ICT shocks for different specifications



Interpretation of results

- ICT-I leads to significant and persistent TFP increases in the medium run.
 - ICT literature: it's a general purpose technology (GPT)!
- embed in a structural model and estimate whether data favors the GPT-interpretation.

Roadmap

1 SVAR Analysis

2 Structural Model

2. Model - in a nutshell

- Belongs to the class of Greenwood, Hercowitz & Krusell (GHK) models.
(We use isomorphic formulation of Oulton (2007).)
- Key: two sectors with identical production functions
→ with an externality capturing the GPT-nature of ICT-capital.
- Rest of model perfectly standard.

Two sectors

Consumption-good sector

$$y_t^c(j) = A_t^c (k_t^c(j))^a (k_t^i(j))^b (l_t(j))^{1-a-b}, \quad 0 < a, b < 1 \quad (5)$$

ICT-good sector

$$y_t^i(q) = A_t^i (k_t^c(q))^a (k_t^i(q))^b (l_t(q))^{1-a-b}, \quad 0 < a, b < 1 \quad (6)$$

with

$$A_t^c = \eta_t \theta_t^c (k_t^i)^\gamma$$

$$A_t^i = \eta_t \theta_t^i (k_t^i)^\gamma$$

Uses of outputs and GDP

Consumption-good sector

$$y_t^c = c_t + i_t^c$$

ICT-good sector

$$y_t^i = i_t^i$$

GDP is

$$GDP_t = (1 - w)y_t^c + wy_t^i \quad \text{where} \quad w = \frac{p_t y_t^i}{y_t^c + p_t y_t^i}$$

with

$$p_t = \frac{p_t^i}{p_t^c} \quad \text{where we normalize} \quad p_t^c = 1$$

TFP in the model

Can be computed 2 ways:

$$1) \ TFP_t = (1 - w)TFP_t^c + wTFP_t^i$$

$$2) \ g_{TFP} = g_{GDP} - ag_{k^c} - bg_l - (1 - a - b)g_{k^i}$$

where for a variable X , $g_x := \ln \left(\frac{X_t}{X_{t-1}} \right)$.

In the model, the latter is equivalent to

$$g_{TFP} = g_\eta + wg_{\theta^c} + (1 - w)g_{\theta^i} + \gamma g_{k^i} \quad (7)$$

Impulse-response matching - does data support $\gamma > 0$?

Estimate three parameters: $\Omega = (\gamma, \sigma_\iota^2, \rho_\iota)$ with

- σ_ι^2 = the variance of an ICT technology shock
- ρ_ι = the persistence of the same shock
- γ = the size of the spillover effect of ICT capital on TFP

We estimate Ω as

$$\min_{\Omega} [\hat{\Psi} - \Psi(\Omega)]' \Lambda [\hat{\Psi} - \Psi(\Omega)] \quad (8)$$

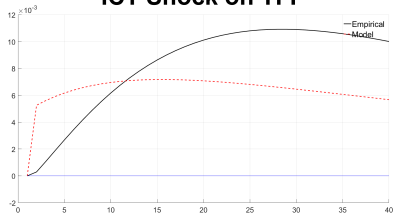
- $\Psi(\Omega)$ = mapping from Ω to the theoretical impulse responses
- $\hat{\Psi}$ = the empirical impulse responses of an ICT shock to TFP, ICTI, C and RP

IR-matching results I

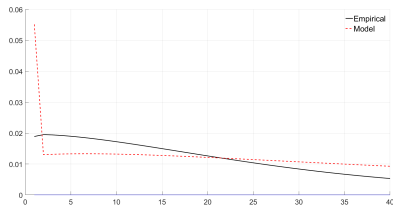
Symbol	Economic Interpretation	Estimated Value
σ_{ι}^2	Variance of ICT technological shock	0.01
ρ_{ι}	Persistence of ICT technological shock	0.9
γ	Size of spillover of ICT capital on TFP	0.5881

IR-matching results II

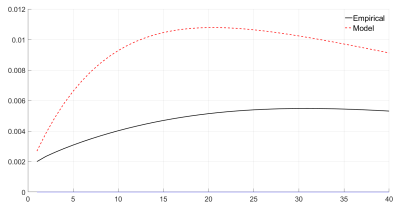
ICT Shock on TFP



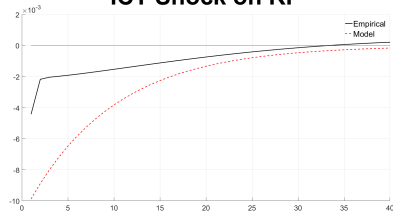
ICT Shock on ICTI



ICT Shock on C

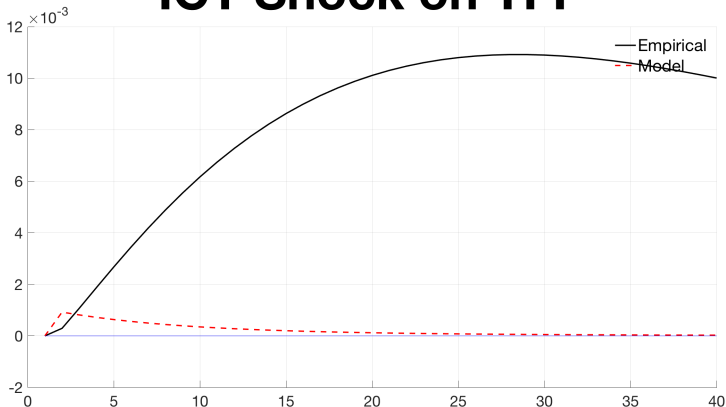


ICT Shock on RP



The spillover drives the TFP-response

ICT Shock on TFP

Figure: Setting $\gamma = 0.1$

Conclusion

- SVAR analysis uncovered interesting pattern between ICT productivity and TFP.
 - An ICT shock leads to delayed but persistent TFP increases.
- Two-sector structural model with a spillover from ICT-capital can rationalize results.
 - Estimation of the model suggests that data supports the GPT-interpretation of ICT.

THANK YOU!

Notation in detail

The reduced-form VAR is

$$y_t = B(L)y_{t-1} + i_t$$

Mapping between innovations i_t to structural shocks s_t

$$A_0 s_t = i_t$$

→ structural-form VAR is

$$A_0^{-1} y_t = C(L)y_{t-1} + s_t$$

where $C(L) = A_0^{-1}B(L)$ and $s_t = A_0^{-1}i_t$, and the impact matrix A_0 satisfies $\Sigma = A_0 A_0'$.

For any arbitrary orthogonalization $\tilde{A}_0 : \Sigma = \tilde{A}_0 \tilde{A}_0'$, a rotation using an orthogonal matrix D ($DD' = I$) allows us to back out impact matrix as $A_0 = \tilde{A}_0 D$.

→ The matrix of impact responses to all shocks is:

$$\Pi(0) = \tilde{A}_0 D$$

Specifically, denoting the responses of variable i to shock j , it is

$$\Pi_{i,j}(0) = e_i' \tilde{A}_0 D e_j$$

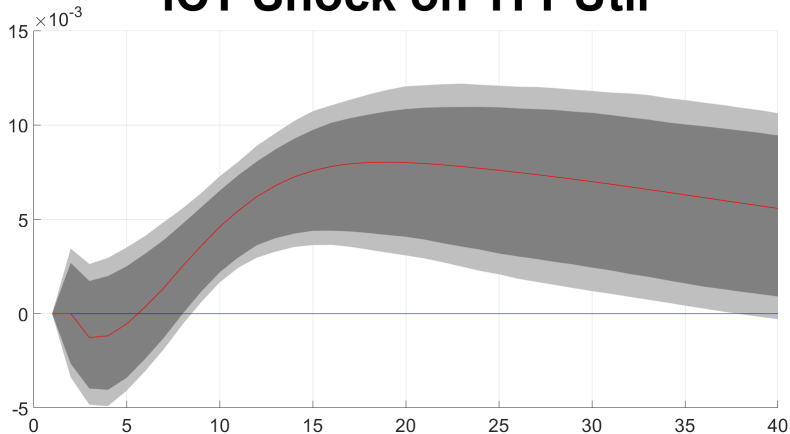
where e_k is a selector column vector.

Denote $\gamma_j := D e_j$, a specific column of D .

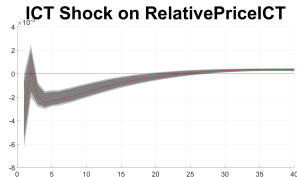
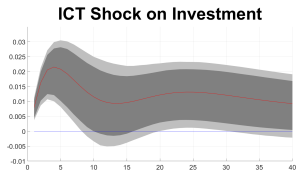
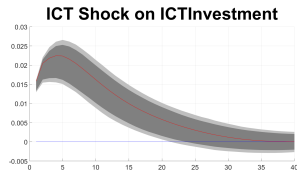
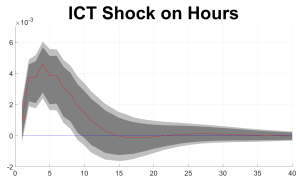
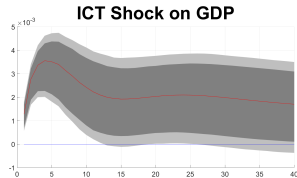
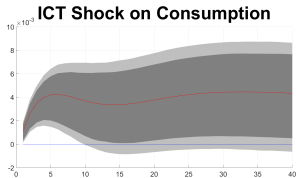
→ $\tilde{A}_0 \gamma_j$ = the vector of impact responses of all variables to shock j .

IRFs for a larger-scale VAR I (2 lags)

ICT Shock on TFPUtil



IRFs for a larger-scale VAR II (2 lags)



Controlling for news

Step 1 - Identification of γ_{news}

$$\max_{\gamma_{news}} \Omega_{1,news}(h) = \frac{\sum_{t=0}^h e_1' B^t \tilde{A}_0 \gamma_{news} \gamma_{news}' \tilde{A}_0' B'^t e_1}{e_1' (\sum_{\tau=0}^H B^t \Sigma B'^t) e_1}$$

subject to

$$\Pi_{1,news}(0) = 0,$$

$$\Pi_{6,news}(0) = \Pi_{6,news}(1) = \Pi_{6,news}(2) = 0, \quad \text{and}$$

$$\gamma_{news} \gamma_{news}' = 1.$$

Controlling for news

Step 2 - Identification of γ_{ICT}

$$\max_{\gamma_{ICT}} \Pi_{2,ICT}(0) = e_2' \tilde{A}_0 \gamma_{ICT}$$

subject to

$$\Pi_{1,ICT}(0) = 0, \\ \gamma_{news} \gamma_{ICT}' = 0, \text{ and } \gamma_{ICT} \gamma_{ICT}' = 1.$$

◀ Return

Robustness checks for the news specification

- Different variables
 - Add the Michigan index of consumer confidence (expected business conditions 5 years ahead)
 - Replace IT prices with capital prices (following Comin & Gertler)
 - Replace CPI inflation with PCE inflation
- Different horizons at which we impose the restriction on relative prices for the news shock
→ ran 6, 8, 10, 12 and 16 quarters.
- Increase the number of lags (2)
- Check whether VAR is information-sufficient to identify the news shock (Forni-Gambetti test) (p-val of 12%)