

# ICT and Future Productivity Evidence and Theory of a GPT

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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).
- Fernald et al. (2017): the Great Recession 2008 casts doubt on whether this is all of the story.

Argument

# This paper

- 1 Stick aggregate 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: Cholesky

The ICT shock is assumed to have

- no impact effect on TFP;
- maximal impact effect on ICT-I.

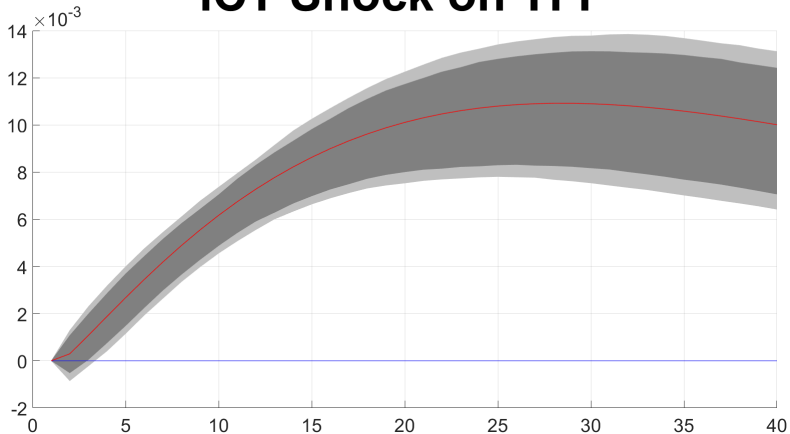
# Why this identification?

- 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.



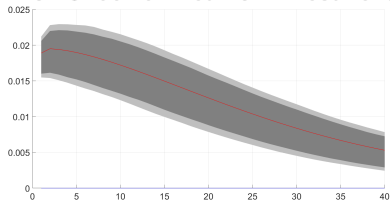
## Results

## ICT Shock on TFP

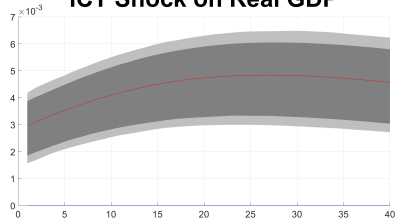


## Results II

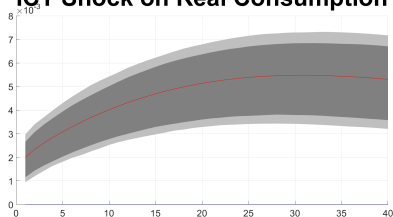
ICT Shock on Real ICT Investment



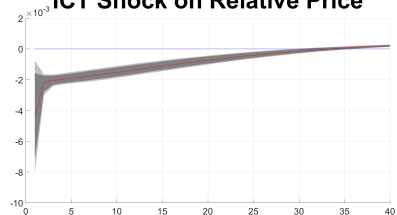
ICT Shock on Real GDP



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

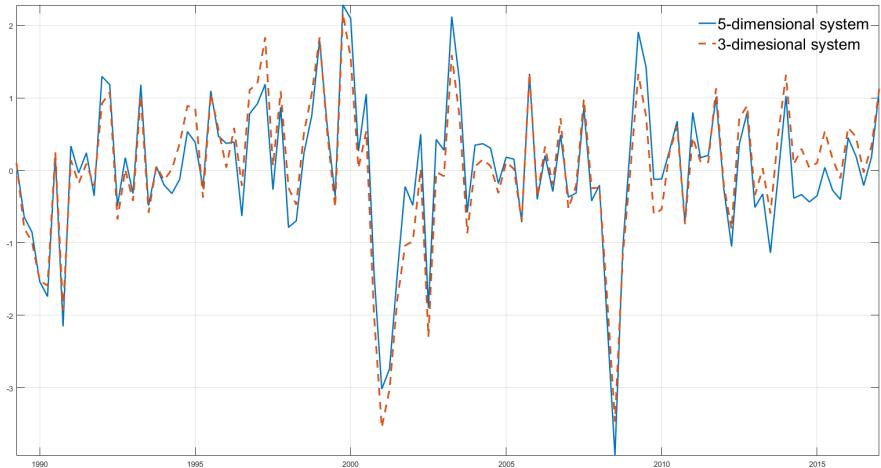
# Robustness checks

- 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

$$y_t^c(j) = A_t^c (k_t^c(j))^a (k_t^i(j))^b (l_t(j))^{1-a-b}, \quad 0 < a, b < 1 \quad (2)$$

ICT-good sector

$$y_t^i(q) = A_t^i (k_t^c(q))^a (k_t^i(q))^b (l_t(q))^{1-a-b}, \quad 0 < a, b < 1 \quad (3)$$

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

Consumption-good sector

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

ICT-good sector

$$y_t^i = i_t^i$$

TFP in the model

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

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

- $\sigma_\iota^2$  = the variance of an ICT technology shock
- $\rho_\iota$  = the persistence of the same shock
- $\gamma$  = the size of the spillover effect of ICT capital on TFP

and calibrate the rest. Calibration

We estimate  $\Omega$  as

$$\min_{\Omega} [\hat{\Psi} - \Psi(\Omega)]' \Lambda [\hat{\Psi} - \Psi(\Omega)] \quad (4)$$

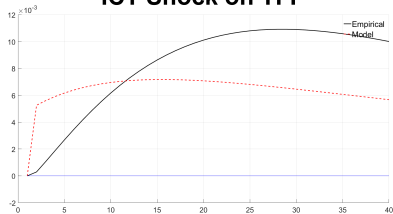
- $\Psi(\Omega)$  = mapping from  $\Omega$  to the theoretical impulse responses
- $\hat{\Psi}$  = the empirical impulse responses of an ICT shock to TFP, ICTI, C and RP

## IR-matching results I

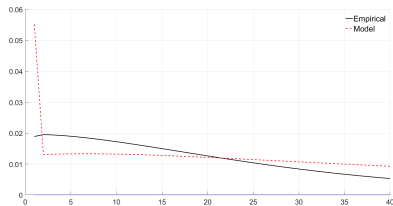
Symbol	Economic Interpretation	Estimated Value
$\sigma_{\iota}^2$	Variance of ICT technological shock	0.01
$\rho_{\iota}$	Persistence of ICT technological shock	0.9
$\gamma$	Size of spillover of ICT capital on TFP	0.5881

## IR-matching results II

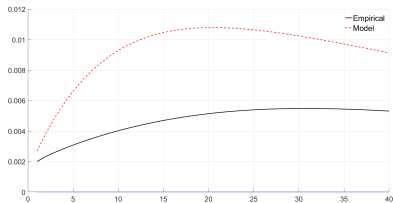
ICT Shock on TFP



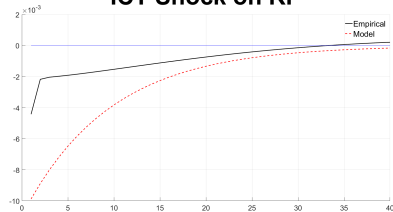
ICT Shock on ICTI



ICT Shock on C



ICT Shock on RP

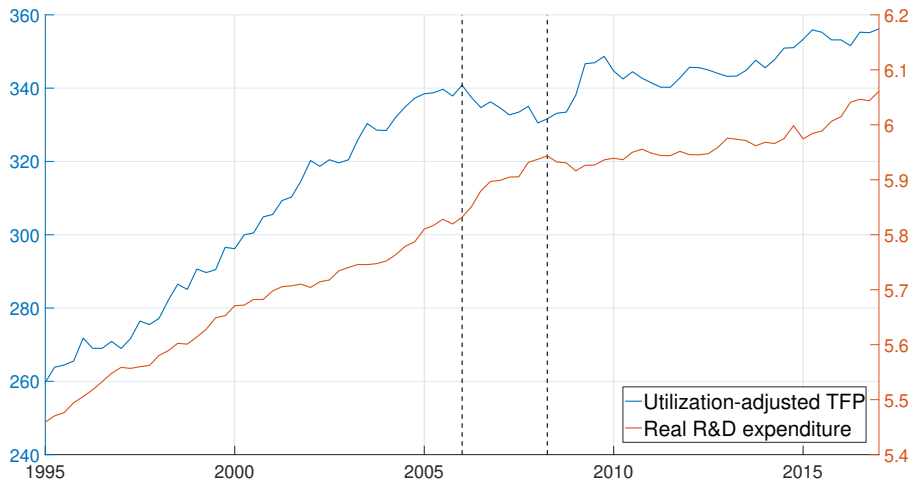


# Conclusion

- 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.

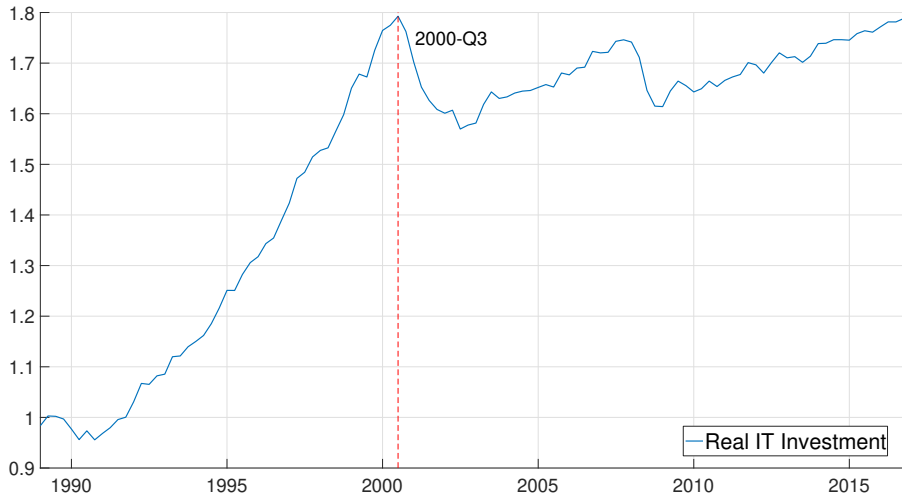
THANK YOU!

# Wrong timing



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





## 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

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

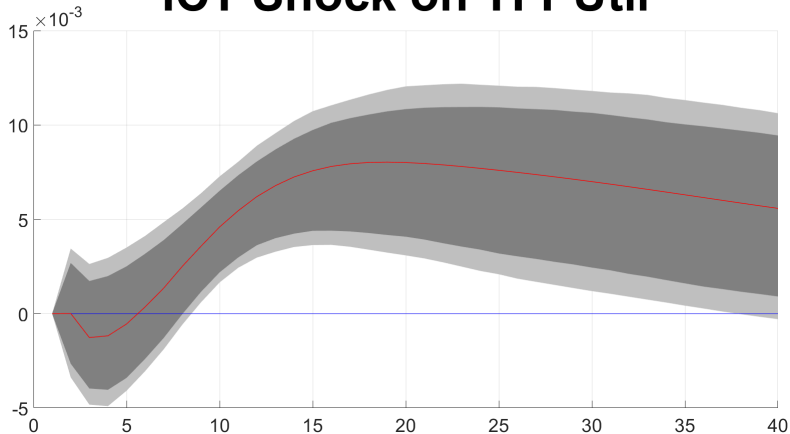
where  $e_k$  is a selector column vector.

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

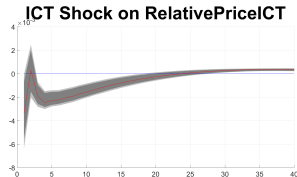
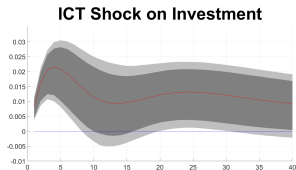
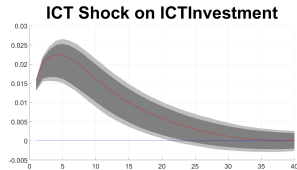
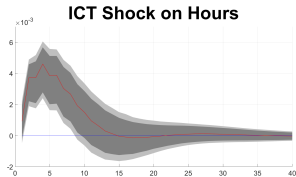
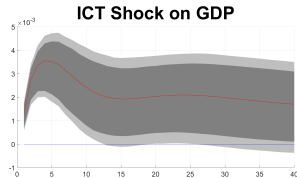
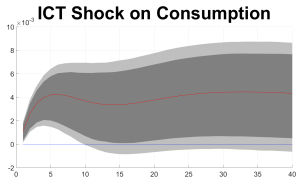
→  $\tilde{A}_0 \gamma_j$  = the vector of impact responses of all variables to shock  $j$ .

## 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 $\gamma_{news}$

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

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 $\gamma_{ICT}$

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

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%)



# 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

$$g_{TFP} = g_\eta + wg_{\theta^c} + (1 - w)g_{\theta^i} + \gamma g_{k^i} \quad (5)$$

# GDP and relative prices in the model

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

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

◀ Return

# Calibration

	Economic Interpretation	Value	Reference
$\beta$	Discount factor of households	$0.97^{\frac{1}{4}}$	Match 3M-T-bill return
$\delta^c$	Depreciation of hard capital	0.0206	BEA
$\delta^i$	Depreciation of ICT capital	0.0398	BEA
$a$	Hard capital share	0.3	Standard value
$b$	ICT capital share	0.031	Oulton (2012)
$\Gamma^c$	Steady state growth rate of hard capital	1.0034	See below
$\Gamma^i$	Steady state growth rate of ICT capital	1.0160	See below
$\chi$	Frisch elasticity	1	Standard value

Get  $(\Gamma^c, \Gamma^i)$  by solving the following model-implied steady state system

$$\begin{cases} \frac{p_t}{p_{t-1}} = \frac{\Gamma^c}{\Gamma^i} \\ \frac{c_t}{c_{t-1}} = (\Gamma^c)^{\frac{1-b-\gamma}{1-a-b-\gamma}} (\Gamma^i)^{\frac{b+\gamma}{1-a-b-\gamma}} \end{cases} \quad (6)$$