Financial Shocks, Productivity and Prices¹

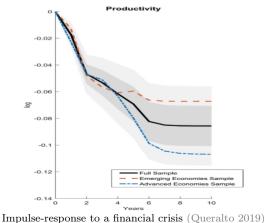
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 $^{^{1}}$ The views expressed in this paper are those of the authors and do not necessarily reflect the views of the National Bank of Belgium, the Eurosystem, or any other institutions to which the authors are affiliated.

Financial Crisis and Aggregate Productivity Growth

- Financial crises are followed by a persistent slowdown in economic activity (Cerra and Saxena 2008; Reinhart and Reinhart 2010; Ball 2014;)
- Large fraction of output decline is accounted for by persistent drop in aggregate TFP (Queralto 2019; Aguiar and Gopinath 2007)



Financial Crisis and Aggregate Productivity Growth

Several factors can account for slow aggregate productivity growth following financial crisis

- Contraction of demand (Mian and Sufi 2009;2014)
- Slowdown of innovation activity (Schmookler 1996; Aghion et al 2010; Bianchi et al 2017; Azoategui et al 2019; Reifschneider et al. 2015)
- Resource misallocation (Garcia-Macia 2015; Lenzu and Manaresi 2019; Sette et al 2019)
- Decline in business dynamism (Decker et al 2017)
- Loss of task-specific human capital (Labor hoarding, Giroud and Muller 2017)

Thanks to growing availability of micro-data, one explanation gained traction:

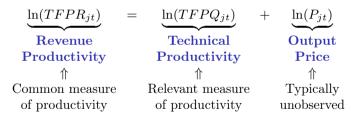
 $\mathbf{Credit\ supply\ shock} \Rightarrow \mathbf{Firm\text{-level\ productivity}} \downarrow \Rightarrow \mathbf{Aggregate\ productivity} \downarrow$

- Quantitative models (Aghion et al 2010; Midrigan and Xu 2014; Buera and Moll 2015; Queralto 2019)
- Micro-evidence (Duval et al. 2017; Dörr et al. 2018; Manaresi and Pierri 2018; Levine and Warusawitharana 2019)

Available empirical evidence is inherently inconclusive:

It fails to account/ignores that firm might **adjust prices** in response to financial shocks

Data limitation: typically firm-level output quantities and output prices not separately observed



Does the distinction between TFPR and TFPQ matter?

Disentangles the causal effect of credit shocks on $\ensuremath{\mathbf{TFPQ}}$ and prices

- 1. Measurement: Observe product-level quantities and prices; measure TFPR and TFPQ
- 2. Causality: Address identification issue constructing firm-level credit supply shifters

Disentangles the causal effect of credit shocks on TFPQ and prices

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Core results

- 1. Immediate and persistent contraction of TFPR growth in response to credit tightening
- 2. <u>Short-run</u>: TFPR response is *totally uninformative*:
 - Firms respond adjusting prices (heterogeneously), entirely driving TFPR response
 TFPQ is not affected
- 3. Long-run: Financial shocks have a persistent impact on TFPQ growth
 - TFPR and TFPQ co-move in the long-rum
 - Long-run TFPR response likely underestimates long-run TFPQ response (prices increase)

Understanding economic forces driving prices and productivity response

- 1. Analyze empirical findings through the lens of a theoretical model
- 2. Model offers testable implications linking financial factors to both firm price-setting behavior and productivity growth
 - Short-run price adjustment is heterogeneous (prices can increase/decrease)
 Direction of pricing response depends on interplay of firms' assets (inventories), liabilities, and competitive environment
 - Credit shock cubs firms innovation activity Slowdown of investments in technology adoption can explain long-term TFPQ growth

1. Data

- 2. Prices and productivity measurement
- 3. Identification of firm-level credit supply shocks
- 4. Effect of credit supply shocks on productivity and prices
 - Short-run effects VS Long-run effects
 - \bullet Economic channels (Theory + Empirics)
- 5. Taking stock

Data

Comprehensive micro-level database merging administrative records of Belgian manufacturing

- 1. Prices and quantity of output at the firm-product level
 - PRODCOM database from Belgian statistical agency
- 2. Firm-level data on firm assets, liabilities and income statement
 - Firms' annual accounts
 - Value added fiscal declarations
- 3. Firm-bank credit relationships
 - Corporate credit register of the National Bank of Belgium
- 4. Bank balance sheets
 - Supervisory records from National Bank of Belgium
- 5. Firm-to-firm transactions (firm-level input-output matrix)
 - Universe of business-to-business transactions (customer-seller) from VAT declarations

Panel a: Firm characteristic	s		Panel b: Credit, Prices, and TFP growth rates				
	Mean	$^{\mathrm{SD}}$, , ,	Mean	$^{\mathrm{SD}}$		
Total assets _{j}	92.722	308.623	$g(\operatorname{Credit}_j)$	153	.580		
$Employees_j$	215.407	1123.363	$g_5(\operatorname{Credit}_j)$	822;	1.05		
Age_j	32.401	19.044	$\Delta \ln(P_j)$.007	.156		
Bank leverage _{j}	.210	.205	$\Delta_5 \ln(P_j)$.110	.322		
Share liabilities due_j	.347	.236	$\Delta \ln(TFPR_j)$.022	.112		
Cash_i	.064	.083	$\Delta_5 \ln(TFPR_i)$.028	.137		
ROA_{j}	.022	.113	$\Delta \ln(TFPQ_j)$.016	.188		
Number of products _{j}	2.485	2.944	$\Delta_5 \ln(TFPQ_j)$	082	.338		
Inventory <i>i</i>	.145	.113					
Multiple relationships $_{i}$.635	.482					
Number of relationships _{j}	2.090	1.097	Panel c: Lender characteris	\mathbf{tics}			
Length of relationships	9.748	11.239	Sov_j	.142	.045		
R&D rate _{j}	.062	.212	Bank Size _{j}	17.692	.457		
Technology adoption rate i	.290	.933	Tier1 ratio _{i}	.217	.010		
Innovation rate _{i}	.351	.948	Net interbank liabilities ratio _{j}	091	.073		
Length customer relationships _{j}	4.383	1.337	Liquidity ratio _j	.847	.245		
Customer share i	.069	.071	Deposits ratio i	1.594	.122		
Number of direct competitors $_{i}$	27.256	27.256	Bad loans ratio _{i}	.004	.001		
Customized $goods_j$.094	.293	U U				
Firms	1,	017	▷ Industry partition				
Firm-year observations	5,	486	Firm-bank relationships	2,057			

Prices and TFP estimation

$$\Delta \ln(TFPR_j) = \Delta \ln(TFPQ_j) + \Delta \ln(P_j)$$

- <u>Prices</u> Measurement of firms' "output price" is challenging for multi-product firms We construct different price measures building on product-level prices
 - Product-level prices
 - Firm-level price index
- **TFPR** We estimate revenue productivity as residual from non-parametric gross output production functions (Gandhi et al 2019)

$$\ln(\widehat{TFPR}_{jt}) = \ln(PQ_{jt}) - f(k_{it}, l_{it}, m_{it}; \hat{\Theta})$$

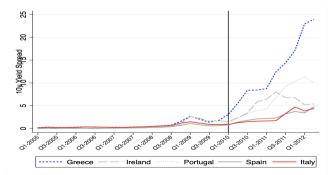
• **TFPQ** -
$$\Delta \ln(\widehat{TFPQ}_j) = \Delta \ln(\widehat{TFPR}_j) - \Delta \ln(P_j)$$

Estimation of credit supply shocks

The burst of the sovereign crisis

We follow Bottero, Lenzu and Mezzanotti (2019), and exploit heterogeneity of banks' exposure to sovereign securities issued by GIPSI countries in the wake of Greek bailout request

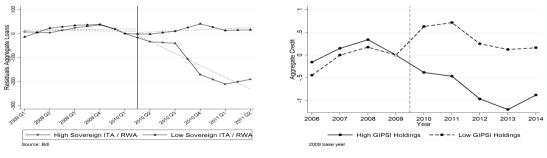
- April 2010: Bailout request advanced by Greek government
- Events in Greece triggered reassessment of country-specific risk in sovereign bond market
- Shortly after, investors began to be concerned with solvency of other peripheral EU countries



Yield-spread of sovereigns GIPSI countries vis-a-vis German sovereigns

Transmission of the sovereign shock to the banking sector ¹⁴

- The sudden change in the risk profile of sovereigns issued by GIPSI countries had a direct negative effect on the balance sheets of banks holding these assets
- Banks passed the shock through to their borrowers in the form of a credit tightening Bottero, Lenzu, Mezzanotti (2019), Acharya et al. (2018), and Balduzzi et al. (2018)



Italy (Bottero et al. 2019)

Belgium (This paper)

- Construction of firm-level credit supply shifters and identification strategy follows Bottero, Lenzu, Mezzanotti (2019)
- Exploit variation in the presence and importance of firms' relationships with banks more/less exposed to GIPSI sovereigns (measured before Greek bailout (2010:Q1))

$$\operatorname{Sov}_{j} = \sum_{b \in \mathcal{B}_{j}} \omega_{jb} \cdot \operatorname{Sov}_{b,2010:Q1} \qquad w_{jb} = \frac{\operatorname{Credit}_{jb,2010:Q1}}{\operatorname{Credit}_{j\mathcal{B},2010:Q1}}$$

$$\mathbf{Y_j} = \boldsymbol{\beta} \cdot \mathrm{Sov}_{\mathbf{j}} + \boldsymbol{\Gamma_1'} \mathbf{X_j} + \boldsymbol{\Gamma_2'} \mathbf{K_j} + \boldsymbol{\Gamma_3'} \mathbf{Z_j} + \mathbf{i_{ind}} + \mathbf{i_{reg}} + \mathbf{u_j}$$

 $\mathbf{X}_{j} = \text{Firm-level controls (Age, size, leverage, liquidity, number of products, productivity)}$ $\mathbf{K}_{j} = \text{Bank-level controls (Size, capitalization, Interbank liabilities, ROA, non-performing loans)}$ $\mathbf{Z}_{j} = \text{Relationship-level controls (Number and average length of lending relationships)}$

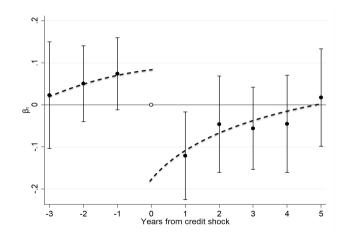
> First Stage: $Y_j = g(Credit_j)$ Second Stage: $Y_j = \{\Delta ln(TFPR_j), \Delta ln(P_j), \Delta ln(TFPQ_j)\}$

Credit response to the sovereign shock (first-stage)

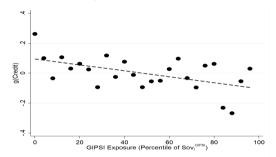
		g(Cre		$g(\operatorname{Credit}_{jb})$		
		Firm	-level		Relation	ship-level
	(1)	(2)	(3)	(4)	(5)	(6)
Sov_j	121	0.56	022	.010	158	178
	(.050)	(.044)	(.049)	(.047)	(.062)	(.044)
$Sov_j \ge Bank \ Leverage_j$		343		268		
		(.102)		(.085)		
$Sov_i \ge Share liabilities due_i$			300	241		
0 0			(.102)	(.090)		
R^2	.132	.146	.148	.156	.123	.487
Observations	1,017	1,017	1,017	1,017	2,021	2,532
Firm controls	Y	Y	Y	Y	Ν	Ν
Bank controls	Y	Y	Y	Y	Υ	Y
Relationship controls	Y	Y	Y	Y	Υ	Y
Industry FE	Y	Y	Y	Y	Ν	Ν
Geography FE	Y	Y	Y	Y	Ν	Ν
Firm FE	Ν	Ν	Ν	Ν	Ν	Y

- On average, $+1\sigma$ exposure to sovereign shock leads to a reduction of **12 percent** in bank credit in the year following the burst of the sovereign crisis
- Effect larger for firms with more leverage and with a large share of debt due within one year (roll-over effect)

Credit shock was transitory in nature

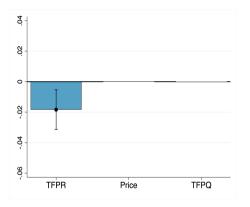


• Relationship between credit holds along the entire distribution



- Effect not driven by sorting of high-sovereign banks with worst borrowers
- Effect similar using alternative measures of sovereign exposure
- Placebo: exposure to sovereigns of "Core" EU countries have positive or no effect on lending

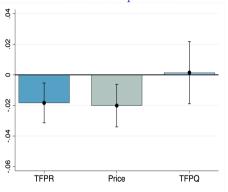
Effect of credit supply shock on productivity and pricing



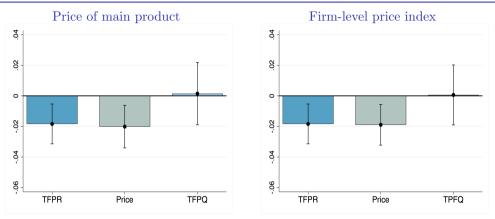
Result #1:

Sharp drop in firm-level TFPR growth (-10% credit \Rightarrow -1.5% TFPR growth) (consistent with findings of previous literature)

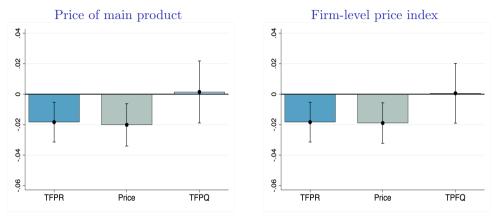




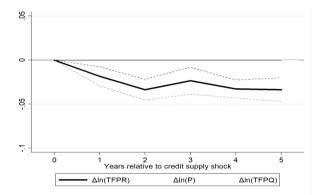
Result #2:Price adjustment drives short-run effect on TFPR (-10% credit \Rightarrow -1.6% price)Technical productivity growth is unaffected in the short-run (!!)



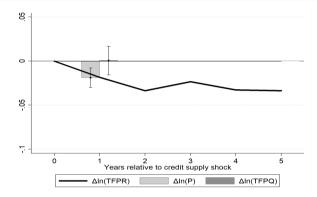
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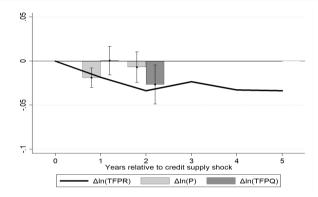


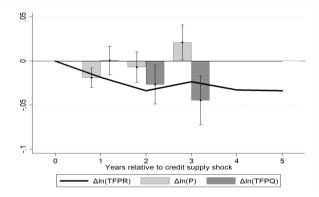
- Results are not affected by how prices are measured/aggregated \triangleright Alternative prices
- \bullet Results are also robust to method of productivity measurement $\quad \triangleright \ {\rm Alternative \ TFP}$
- Hold for single-product firms and multi-product firms that don't change product composition
- Are robust to control for firm products' quality adjustment (forthcoming)

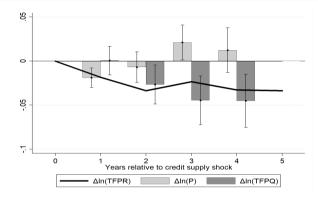


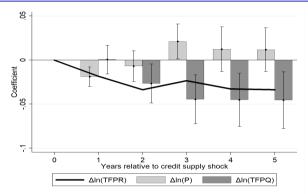
• Credit tightening impairs long-run TFPR growth











Result #3: Credit tightening impairs long-run productivity growth

- TFPQ effect kicks-in after 2-3 years
- \bullet +1 σ credit hock \Rightarrow -6% TFPQ after 5 years
- Pricing effects reverses over time (although noisy)
- TFPR response *underestimates* true TFPQ response

Economic channels

What drives short-run pricing response and long-run TFPQ response?

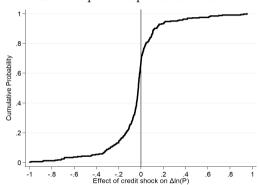
Model highlights

- Leveraged firm sequentially decides production, pricing, and investments as uncertainty resolves
- Realization of unanticipated credit supply shock updates firm's prior about likelihood of default (default is costly, Modigliani & Miller does not apply)
- Firms can take actions to contrast effect of shock on probability of default:
 - \bullet Increase sales revenues ($P\downarrow$)
 - \bullet Reduce costs (investments in innovation $\downarrow)$
- But lower P and innovation have costs (trade-off):
 - \bullet Lower P reduces short-term profits
 - Lower innovation reduces productivity growth and therefore long-term profits

- 1. A firm's ability to sell more output (capacity constraint)
- 2. Effectiveness of price adjustment in reducing default probability

Prediction 2 - Investments in innovation contract

Prediction 1 - Pricing response to credit tightening is heterogeneous



Distribution of price response to credit shock

- On average, firms reduce prices in response to a tightening of financing conditions
- But, about 35 percent of firms appear to respond increasing preces

	(1)	(2)	(3)
		$\Delta \ln(P_j)$	
Sov_j	012	024	024
	(.007)	(.008)	(.008)
$\operatorname{Sov}_j \times \operatorname{Inventory}_j$	051		003
	(.018)		(.039)
$Sov_j \times High Default Risk_j$.027	.056
		(.009)	(.018)
$Sov_j \times Inventory_j$			191
\times High Default Risk _j			(.084)
Marginal effects:			
High Default Risk and Low inventory			0.032
			(.014)
High Default Risk and High inventory			194
			(.056)
R^2	.124	.124	.131
Observations	1,017	1,017	1,017

Inventory fire-sale channel + strength of liabilities

• Inventory depletion allows firms to increase sales revenues (at no extra production cost) \Rightarrow **Price decrease**

Inventory fire-sale channel + strength of liabilities

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$Sov_i \times Inventory_i$	051	· /	003
5 65	(.018)		(.039)
$Sov_i \times High Default Risk_i$	· /	.027	.056 [´]
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Inventory depletion allows firms to increase sales revenues (at no extra production cost) ⇒ Price decrease
Firms w/ weak balance sheets (high net liabilities) are hit hard by credit shock: extra revenues from price reduction do little to mitigate default ⇒ Price increase

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Elasticity of customers' demand

• Take advantage of firm-to-firm transaction data to construct measurable proxies of residual elasticity of demand of individual firms

	(1)	(2)	(3)
		$\Delta \ln(P_j)$	
Sov_j	-0.043	-0.019	-0.025
	(0.001)	(0.005)	(0.008)
$Sov_j \times Length customer relationships_j$	0.006		
	(0.001)		
$Sov_j \times Customized goods_j$		0.052	
		(0.023)	
$Sov_j \times Customer share_j$			0.115
			(0.049)
R2	0.126	0.131	0.125
Observations	1,017	1,017	1,017

If consumers' demand is inelastic

- Lower incentives to lower prices in response to credit shock: sales revenues do not increase much if $P \downarrow$
- ${ullet}$ Greater incentives to increase prices to raise liquidity: sales revenues do not drop much if $P\uparrow$

Economic channels: long-term TFPQ response

Investments in innovation and technology adoption

Previous literature

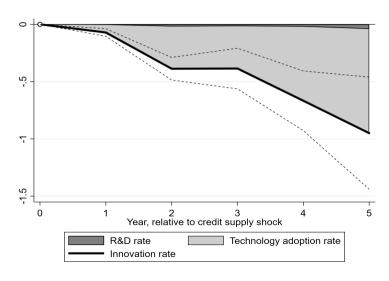
- Links productivity growth and productivity enhancing investments (Aghion et al. 2012, Garcia-Macia 2017, Huber 2018; Anzoategui et al. 2019)
- Provides evidence consistent with financial constraints discouraging innovation (Hall and Lerner 2010; Kerr and Nanda 2015; Bond et al. 2005; Howell 2017; Comin and Nanda 2018)

We connect the two strands of the literature

- 1. Credit shock \rightarrow investments in innovation (R&D and technology adoption)
- 2. Investments in innovation (driven by exposure to credit shock) \rightarrow TFPQ growth

Economic channels: long-term TFPQ response

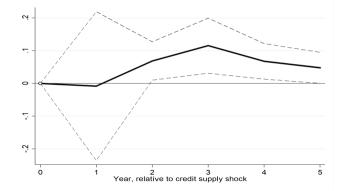
Prediction 2 - Credit shock leads to contraction of investments in innovation



Economic channels: long-term TFPQ response

Linking financial shocks to TFPQ via investment in innovation

IV regression, instrumenting cumulative investments in innovation with credit supply shock



• Contraction of innovation driven by tightening of financing conditions explains long-run TFPQ growth slowdown

Taking stock

Financing, pricing decisions, and productivity dynamics are tightly related.

- 1. Distinguishing between TFPQ and TFPR is of first-order importance to understand relationship between finance, productivity growth, and economic growth (Aghion et al. 2012, Garcia-Macia 2017; Huber 2018; Anzoategui et al. 2019; Queralto 2019)
 - The use of TFPR measures as proxies of TFPQ is ubiquitous
 - But confounding TFPR and TFPQ movements offers misleading insights into how firms respond to financial shocks
 - Observed short-run drop in TFPR is good (price adjustment is constrained optimum)
- Emphasis on distinction b/w TFPR and TFPQ shared w/ empirical studies in other settings (Foster et al. 2008; Hsieh and Klenow 2009; Haltiwanger et al. 2018; Katayama et al. 2009; Eslava et al. 2013; Garcia-Marin and Voigtlander 2018; Eslava and Haltiwanger 2018)
- 3. This paper bridges the productivity literature with separate literatures
 - Response of prices/markups to financial shocks: Prices increase (Chevalier and Scharfstein 1995; Gilchrist et al 2017) vs Price decrease (Zingales 1998; Phillips and Sertsios 2013; Kim 2018)
 - Financing innovation (Hall and Lerner 2010; Kerr and Nanda 2015; Howell 2017; Comin and Nanda 2018)

Financial Shocks, Productivity and Prices

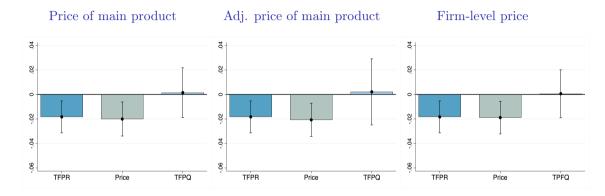
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Industry partition

Industry Code	Industry	Percentage of firms
7	Mining of metal ores	0.1
8	Other mining and quarrying	1.97
10	Manufacture of food products	25.07
11	Manufacture of beverages	2.06
12	Manufacture of tobacco products	0.88
13	Manufacture of textiles	7.87
14	Manufacture of wearing apparel	0.59
15	Manufacture of leather and related products	0.29
16	Manufacture of wood and of products of wood and cork, except furniture	3.93
17	Manufacture of paper and paper products	4.72
18	Printing and reproduction of recorded media	1.67
20	Manufacture of chemical and chemical products	12.88
21	Manufacture of basic pharmaceutical products and pharmaceutical preparations	0.2
22	Manufacture of rubber and plastic products	8.36
23	Manufacture of other non-metallic mineral products	7.18
24	Manufacture of basic metals	3.93
25	Manufacture of fabricated metal products, except machinery and equipment	6.59
26	Manufacture of computer, electronic, and optical products	2.95
27	Manufacture of electrical equipment	3.15
28	Manufacture of machinery and equipment nec	5.21
29	Manufacture of motor vehicles, trailers, and semi-trailers	0.2
31	Manufacture of furniture	0.1
32	Other manufacturing	0.1

Production Function Estimates

NACE Code	Prod.		stimatio	n, NP	Prod. Fun. Estimation, CD			Revenue Shares, CD					
(2-digits)	θ^L	θ^{K}	θ^M	\mathbf{RS}		θ^L	θ^{K}	θ^M	\mathbf{RS}	θ^L	θ^{K}	θ^M	\mathbf{RS}
14	0.24	0.075	0.647	0.962	(0.316	0.03	0.614	0.96	0.20	2 0.161	0.637	1
15	0.201	0.061	0.769	1.031	(0.298	0.05	0.719	1.067	0.22	4 -0.157	0.934	1
16										0.11	9 0.088	0.793	1
17	0.295	0.037	0.683	1.015	(0.277	0.128	0.66	1.065	0.25	6 - 0.058	0.686	1
18	0.276	0.057	0.726	1.059	(0.482	0.109	0.601	1.192	0.32	9 - 0.033	0.638	1
19										0.26	2 - 0.053	0.685	1
20	0.226	0.06	0.718	1.005	(0.152	0.201	0.676	1.029	0.23	9 - 0.069	0.692	1
21	0.207	0.076	0.702	0.985	(0.335	0.005	0.692	1.032	0.22	2 0.081	0.697	1
22	0.265	0.047	0.699	1.01		0.22	0.025	0.685	0.93	0.22	5 0.058	0.718	1
23										0.15	2 - 0.073	0.775	1
24	0.238	0.059	0.702	0.998	(0.346	0.027	0.707	1.08	0.21	1 -0.057	0.847	1
25	0.273	0.05	0.68	1.004	(0.314	0.033	0.687	1.034	0.22	1 0.071	0.708	1
26	0.289	0.09	0.643	1.022	(0.263	0.045	0.674	0.983	0.22	9 - 0.082	0.689	1
27	0.185	0.067	0.741	0.993	(0.263	0.032	0.725	1.019	0.20	1 0.029	0.771	1
28	0.281	0.06	0.663	1.004	(0.308	0.1	0.615	1.023	0.30	5 0.05	0.644	1
29	0.331	0.048	0.654	1.034	(0.349	0.051	0.625	1.025	0.29	5 0.031	0.674	1
30										0.2	B 0.076	0.693	1
31	0.321	0.049	0.648	1.017	(0.388	0.028	0.625	1.041	0.29	1 0.055	0.654	1
32	0.324	0.078	0.607	1.009	(0.414	0.031	0.622	1.066	0.28	0.054	0.666	1
33	0.325	0.034	0.651	1.011	(0.425	-0.036	0.609	0.999	0.30	6 0.057	0.637	1



• Results are not affected by how prices are measured/aggregated

Non-parametric **Cobb-Douglas Cobb-Douglas** Prod. function estimation Prod. function estimation Revenue shares 4 8 8 8 20 8 0 0 0 -.02 8 80 64 8 04 90 90 90 TEPB Price TPFO TEPB Price TPFO TEPR Price TPFO

• Results are also robust to method of productivity measurement

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