On Competition in US banking when technology is noncovex

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The first slide: what this is all about

[measuring] banks' market power when technology is nonconvex.

Contributions.

- 1. A method to measure marginal costs under nonconvex technologies.
- 2. Application to the US banking data over 27 years: 1997-2023
- 3. The results are contrasted to a typical application
- 4. The results have important implications for reconsideration and effective implementation of policies to improve competition in banking

The first slide: what this is all about (cont.)

The issues

- Market power is measured by the Lerner index where the main challenge is to obtain marginal costs
- obtain marginal costs for a function that is not smooth
- comparison to approaches used in the literature

■ The Lerner Index (which measures the mark-up of a bank) is defined as

$$L = \frac{P - MC}{P} \tag{1}$$

- P is obtained from the data
- The marginal costs are not directly observable
- To calculate marginal costs in practice, it is necessary to assume a cost function

Marginal Cost

■ The concept implies the unit change of cost due to a change in output:

$$MC = \frac{\Delta C}{\Delta Y} \tag{2}$$

☐ If the cost function is differentiable,

$$MC = \frac{\partial C}{\partial Y} \tag{3}$$

An example

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How does competition affect bank systemic risk?



Deniz Anginer a,b,*, Asli Demirguc-Kunt b, Min Zhu b

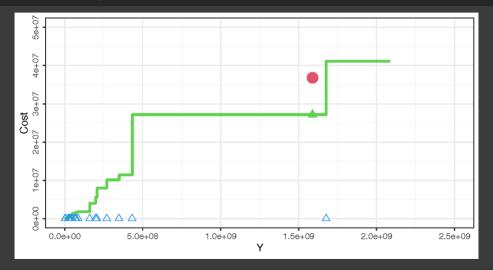
a Virginia Polytechnic Institute and State University, United States World Bank United States

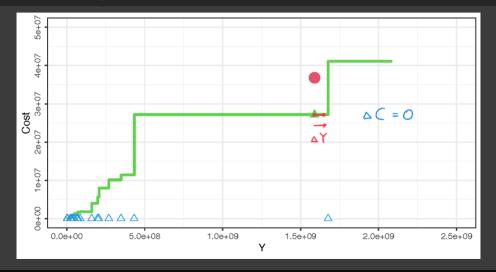
We follow the methodology used in Demirguc-Kunt and Martinez-Peria (2010) and first estimate the following log cost function for each country:

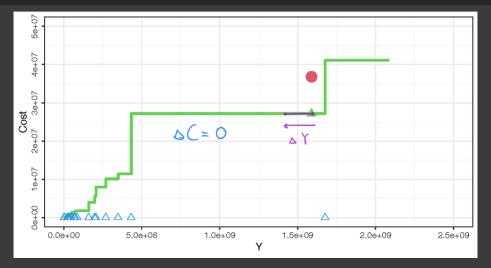
$$\begin{split} \log(\mathsf{C}_R) &= \alpha + \beta_1 \times \log(Q_R) + \beta_2 \times (\log(Q_R)^2 + \beta_3 \times \log(W_{1,R}) + \beta_4 \times \log(W_{2,R}) \\ &+ \beta_5 \times \log(W_{3,R}) + \beta_6 \times \log(Q_R) \times \log(W_{1,R}) + \beta_7 \times \log(Q_R) \times \log(W_{2,R}) \\ &+ \beta_8 \times \log(Q_R) \times \log(W_{3,R}) + \beta_9 \times (\log(W_{3,R}))^2 + \beta_{10} \times (\log(W_{2,R}))^2 \\ &+ \beta_{11} \times (\log(W_{3,R})^2 + \beta_{12} \times \log(W_{1,R}) \times \log(W_{2,R}) + \beta_{10} \times \log(W_{2,R}) \times \log(W_{3,R}) \\ &+ \beta_{14} \times \log(W_{2,R}) \times \log(W_{3,R}) + 0 \times \text{Var Dummies} \\ &+ \Omega_X \text{Bank Specialization Dummies} + \delta_R \end{split}$$

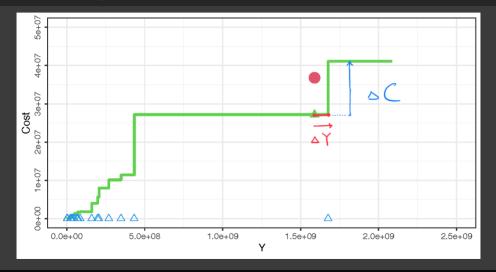
We then use the coefficient estimates from the previous regression to estimate marginal cost for bank i in calendar vear t:

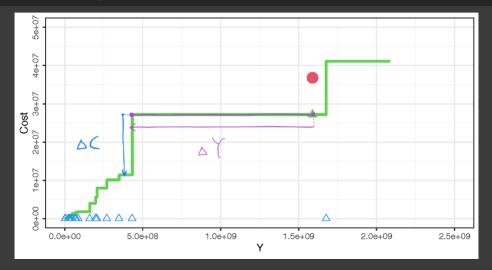
$$\begin{aligned} &MC_{it} = \partial C_{it}/\partial Q_{it} \\ &= C_{it}/Q_{it} \times [\beta_1 + 2 \times \beta_2 \times log(Q_{it}) + \beta_6 \times log(W_{1,it}) + \beta_7 \times log(W_{2,it}) + \beta_8 \times log(W_{3,it})] \end{aligned}$$

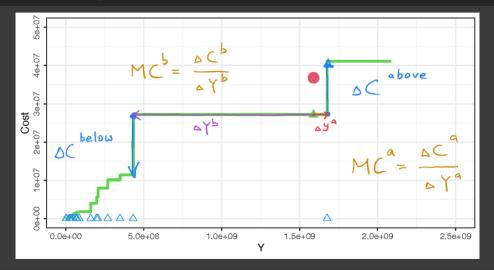












Find Change Points

- 1. Sort Y to get Y^s
- 2. For bank k, calculate C_{k}^{min}
- 3 Below:
 - \square Set i=1 calculate C_{k-1}^{min} at fixed input prices at P_k and Y_{k-1}^s
 - \Box If $C_{k-i}^{min} = C_k^{min}$, increase i=2 and calculate C_{k-i}^{min} at P_k and Y_{k-i}^s

repeat until $C_{k}^{min} \neq C_{k}^{min}$

- 4 Above:
 - \square Set i=1 calculate C_{k+1}^{min} at fixed input prices at P_k and Y_{k+1}^s
 - \Box If $C_{k+1}^{min} = C_k^{min}$, increase i = 2 and calculate C_{k+1}^{min} at P_k and Y_{k+1}^s

repeat until $C_{\nu \perp i}^{min} \neq C_{\nu}^{min}$

Algorithm (cont.)

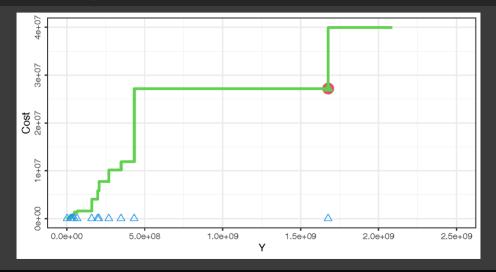
Calculate Marginal Costs

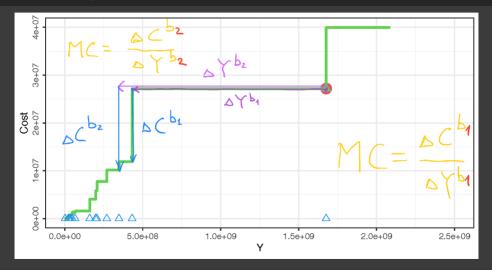
- 1. Below:
 - \square At the found change point k-i calculate

$$MC^{b} = \frac{C_{k}^{min} - C_{k-i}^{min}}{Y_{\nu}^{s} - Y_{\nu-i}^{s}} \tag{4}$$

- 2 Above.
 - \square At the found change point k+i calculate

$$MC^{a} = \frac{C_{k+i}^{min} - C_{k}^{min}}{Y_{s}^{s} - Y_{s}^{s}} \tag{5}$$





Algorithm (cont.)

Calculate Marginal Costs

- 1. Below 1
- 2. Below 2

Calculate Marginal Costs

- 1. Above 1
- 2. Above 2

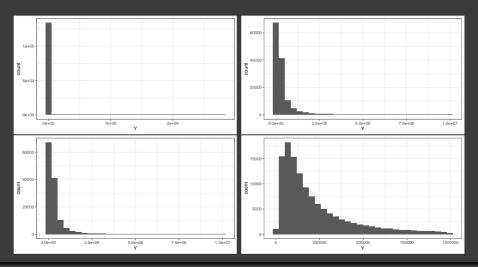
Calculate Marginal Costs

Get a minimum of two

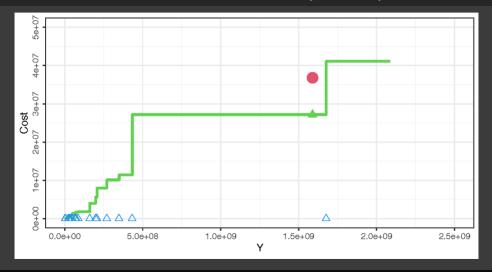
Data, sources

- Call reports on US commercial banks
- Sample:
 - Years 1997-2023
 - \Box Total assets > \$1m in at least one of the 27 years
 - $\sum_{T} = 87K$ (about 3.3K per year)
- Output Y: total assets
- Utput price P: operating income divided by total assets
- 4 Inputs [prices]: purchased funds, deposits, labor, capital

Total assets (zoomed in)



Nonconvex Cost Function (k = 188)



The Lerner Index

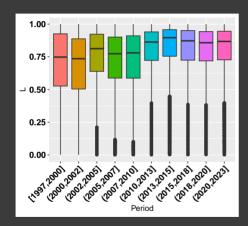


Figure: Left based on the NC technology

The Lerner Index

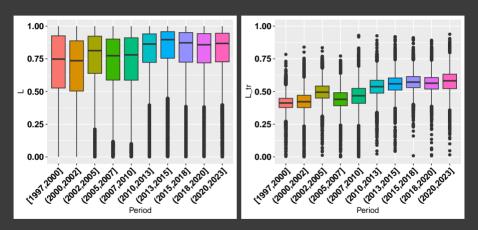


Figure: Left based on the NC technology; right: traditional.

Empirical Results 0000000000

The Lerner Index

Figure:

correlation: 0.22

Stability

$$Z = \frac{(ROA + E/A)}{\operatorname{sd}_{ROA}},\tag{6}$$

ROA is the return on assets, E/A is the equity to assets ratio, and sd_{ROA} is the standard deviation of ROA over the last 3 periods. Higher values of the Z-score come from either greater income or higher capital and indicate lower probabilities of bank insolvency, thus providing a direct measure of bank stability.

$$\ln Z_{it} = \sum_{t} \beta_{i} D_{t} L_{i,t-1} + controls_{it} + macro_{i,t-1}$$
(7)

Traditional Approach

Cost function

$$\ln C_{i} = \beta_{0} + \beta_{y} \ln Y_{i} + 0.5 \beta_{yy} (\ln Y_{i})^{2} + \sum_{j} \beta_{wj} \ln w_{ji}$$

$$+ \sum_{i} \sum_{k} \beta_{wjk} \ln w_{ji} \ln w_{ki} + \sum_{i} \beta_{wjk} \ln w_{ji} \ln Y_{i}$$
(8)

The marginal costs are calculated as

$$MC_{i} = \frac{\partial C_{i}}{\partial Y_{i}} = \frac{\partial \ln C_{i}}{\partial \ln Y_{i}} \frac{C_{i}}{Y_{i}} = \frac{\partial \ln C_{i}}{\partial \ln Y_{i}} AC_{i}$$
(9)

Literature 1 / 3



Literature 2 / 3



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4.2. Competition and systemic risk

We now turn to the results obtained by considering the SRISK as the dependent variable. As emphasized in introduction, to the best of our knowledge, only the recent paper of Anginer et al. (2014) has previously investigated the link between competition and systemic risk at the bank level. However, unlike our study, Anginer et al. (2014) do not consider the SRISK as a measure of systemic risk, but use the $\Delta CoVar$ and a measure based on the correlation between the distance-to-default of each bank and the distance-to-default of the market. Table 4 presents our regression results. As above, specifications (1) to

Table 2

ompetition and bank individual risk: results obtained with the Z-score

Dependent variable	Z-score FE	Z-score FE	Z-score FE	Z-score FE	Z-score IV	Z-score IV
Lerner	2.911***	3.148*** (0.588)	2.851*** (0.568)	2.338*** (0.662)	6.300***	7.352***
Size	0.419	(410-07)	0.145	0.095	0.539	(1.001)

Literature 3 / 3



Bank competition and stability: Cross-country heterogeneity *

Thorsten Beck^a, Olivier De Jonghe^{b,*}, Glenn Schepens^c

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Table 5

The market power-bank soundness relationship:	[iii] esimple pervections									
Variables	In(Z-score)	-In(sd(ROA))	Eq/TA	In(Z-score5)	In(Z-score)	In(Z-score)	In(Z-score)	In(Z-score)	ln(Z-score)	
Lerner	1.946*** (0.105)	0.841*** (0.117)	7.443*** (0.487)	1.845*** (0.114)	1.814*** (0.0862)		1.987*** (0.109)	1.988*** (0.109)	2.324*** (0.139)	

$$\ln Z_{it} = \sum_{i} \beta_{t} D_{t} L_{i,t-1} + controls_{it} + macro_{i,t-1} \Big| T_{i} > 5$$
 (10)

where

controls: (i) In TA, (ii) diversification, (iii) Ilp, (iv) In assets growth, (v) Eq/Assets:

macro: lagged (i) rgdpg, (ii) interest rate, (iii) unempl, (iv) all banks

credit. (v) all banks TA

Market power and stability

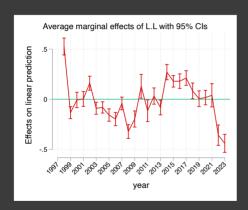
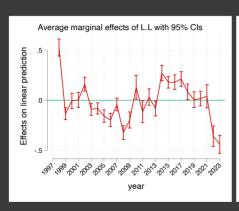


Figure: Left based on the NC technology

Market power and stability



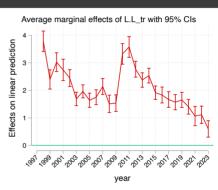


Figure: Left based on the NC technology; right: traditional.

Profitability

$$ROA_{it} = \sum_{i} \beta_{2d} D_t L_{i,t-1} + controls_{it} + macro_{i,t-1} \left| T_i > 5 \right|$$
 (11)

where

controls: (i) In TA, (ii) diversification, (iii) Ilp, (iv) In assets growth, (v)

Eq/Assets:

macro: lagged (i) rgdpg, (ii) interest rate, (iii) unempl, (iv) all banks

credit, (v) all banks TA

Market power and profitability

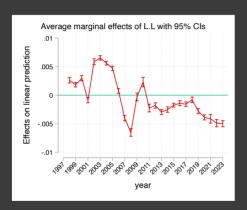
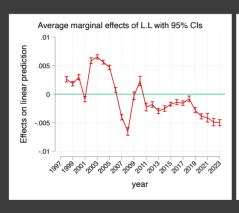


Figure: Left based on the NC technology

Market power and profitability



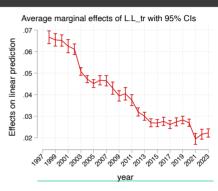


Figure: Left based on the NC technology; right: traditional.

Concluding Remarks

- Seemingly competitive environment in banking
- Deceitful if market power is calculated using an improper cost function
- We offer a method of calculating the Lerner Index when the cost function is nonconvex
- □ Huge differences with the traditional approach

The Last Slide

Thank you for your attention