Measuring Productivity: Lessons from Researcher-Designed Surveys and Productivity Benchmarking

David Atkin, MIT Amit Khandelwal, Columbia GSB Adam Osman, UIUC

Motivation

Productivity differences can explain large differences in income per capita across and within countries

- Obvious Questions:
 - How large are productivity differences across firms?
 - What drives these differences?
 - What policies are most effective at raising productivity and reducing dispersion?
 - Etc.
- Active area of research (e.g. see Syverson 2011 review)

Central challenge for researchers analyzing productivity: it is *never observed*

Measuring TFPQ and TFPR

Researchers typically want physical output conditional on physical inputs (TFPQ)

- But input and output quantities typically not available
- When physical quantities are available, substantial measurement error concerns
- Even then, large differences in specifications and quality make comparisons across firms difficult
 - Administrative product codes often coarse (e.g., 1000+ barcodes within single HS10 code)

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Most firm-level datasets only provide expenditure/revenue data so lit. often uses TFPR

- Also captures differences in quality and markups
- If firm capabilities come from ability to produce both quantity and quality, TFPR may be reasonable

Existing literature has pursued various approaches to understand/mitigate these measurement issues

- Typically draw on additional variation in same datasets alongside (often strong) identifying assumptions
 - E.g. Ackerberg, Caves and Frazer (2015), Kim, Petrin and Song (2016), de Loecker et al. (2016), Garcia-Marin and Voigtlander (2019), de Loecker and Goldberg (2014) survey
 - Hallak and Sividasan (2013) and Sutton and Trefler (2015) explore multidimensional firm productivity; Foster, Haltiwanger and Syverson (2018) separate demand shocks from TFPQ

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- Detailed product specifications and external quality assessments allow us to calculate:
 - Quantity productivity (ability to produce quantity with a given set of inputs, TFPQ)
 - Quality productivity (ability to produce quality with a given set of inputs, TFPZ)
 - Capability (combine TFPQ and TFPZ using consumer's quality-quantity tradeoff, i.e. "TFPQ with quality-adjusted quantities")

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- To assess performance of various survey-based productivity measures, compare with productivity benchmarking exercises which we argue are closest to true productivity

Preview of Results

1. Standard TFPQ performs poorly at measuring quantity productivity

- Shows excessive dispersion across firms
- Is inversely correlated with quality productivity!

2. Controlling for product specifications–effectively making apples-to-apples comparisons– goes some way towards remedying deficiencies in (1)

3. Although standard TFPR does better than TFPQ at capturing a firm's broad capabilities, it performs worse than methods combining survey information with explicit quality measures

4. Quality-productivity dispersion similar to quantity-productivity dispersion, capability more dispersed

Outline

1. Industry Background

2. Survey Design and Productivity Benchmarks

3. Measuring Productivity

4. Comparing Productivity Measures: Six Results

5. Conclusion

Industry: Flat-Weave Rug Producers in Fowa, Egypt

Focus on firms producing a type of kilim rug called "duble"

Producers receive orders with particular set of specifications

- Codifiable attributes of the rug that are typically chosen by the buyer
- E.g. design, thread type, thread count, colors, duble subcategory
- Duble are already subset of 10 digit HS, we observe 435 spec combinations in surveys

Producers prepare inputs, install threads on double-treddle foot powered-looms, weave rug

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Rugs also differ in quality

- Depends on weaving technique and is difficult to codify or contract on
- E.g. how flat a rug lies determined by how skillfully thread is installed on loom, whether threads held correctly while weaving

Survey Design and Data

Survey all 219 firms between 1-5 workers

- Same sample as Atkin, Khandelwal and Osman (2017)
- Exclude baseline/pre-treatment rounds to ensure export status not changing

Quarterly firm surveys (5 post-baseline rounds):

- Surveys at product-line level (capturing rugs produced in last month)
- Output and input prices/quantities
- Labor hours, capital equipment, materials used
- Rug specifications: thread type, thread count, design, colors, market segment, duble subcategory
- External quality assessor graded each rug at time of survey on 11 dimensions
 - Corners, waviness, weight, touch, packedness, warp thread tightness, firmness, design accuracy, warp thread packedness, inputs, and loom quality

Will construct multiple productivity measures using these survey data

Productivity Benchmarking in a Laboratory

Set up a controlled laboratory:

- Firms paid a flat fee to produce an identical rug (domestic design, 140cm by 70cm, should weigh 1750g)
- We provide all firms with identical material inputs and loom in rented facility
- Recorded dimensions, weight, time to produce the rug
 - Provides direct measure of quantity productivity: m² per labor hour (Lab TFPQ)
 - "Benchmark" as lab ensures inputs, product specifications identical across firms
- Rug quality scored anonymously by quality assessor and prof of handicraft science
 - Will use these scores to benchmark quality productivity

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Measuring Productivity – Quantity Productivity

Cobb-Douglas production function

$$x=\phi_u l^{lpha_l} k^{lpha_k} e^\epsilon$$

- Output (*x*), Labor hours (*l*), Looms (*k*).
- For transparency, estimate via OLS (similar results using Olley Pakes control function)
- Recover **Unadjusted TFPQ** (ϕ_u) does not adjust for fact different firms produce different varieties

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Features of setting reduce measurement concerns:

- 1. Observe quantities of *x*, *l* and *k* rather than revenues and expenditures.
- 2. Given simple technology, plausibly no other inputs (e.g. no accounting, logistics, human resources).
- 3. Record inputs used for each specific rug produced so no error in allocating inputs to outputs
- 4. Leontief in materials.

Measuring Productivity – Quantity Productivity

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Adjusting for product specification: Consider $\phi_u = \phi(\lambda; \phi_a, \gamma) = \phi_a e^{\lambda \gamma}$

- Unit input requirements vary with vector of specifications (λ)
 - E.g. a high thread count rug requires more inputs
- Residual after conditioning on specification controls uncovers **Specification-Adjusted TFPQ** (ϕ_a)

Measuring Productivity – Quality Productivity

Similar specification rugs have very different quality, and like quantity, quality raise revenue But how to measure **Quality Productivity (TFPZ)**?

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Consumer valuation of quantity and quality (where q_i are our 11 quality dimensions):

$$U=((\Pi_j q_j^{ heta_j}x)^{rac{\sigma-1}{\sigma}}+(y)^{rac{\sigma-1}{\sigma}})^{rac{\sigma}{\sigma-1}}$$

Implied demand:

$$\ln x = (\sigma-1)\sum_j heta_j \ln q_j - \sigma \ln p + c$$

- Impose σ =2.74 (Broda & Weinstein 2006)
- θ 's capture consumer's valuation of 11 quality metrics vis a vis quantity

With quality aggregate, can formulate quality production function: $\Pi_j q_i^{ heta_j} = \zeta_u l^{eta_l} k^{eta_k} e^arepsilon$

- Recover **Unadjusted TFPZ** (ζ_u) via OLS
- Similarly, recover **Specification-Adjusted TFPZ** (ζ_a) after imposing: $\zeta_u = \zeta(\lambda; \zeta_a, \delta) = \zeta_a e^{\lambda \delta}$

Measuring Productivity – Capability

Aggregate quality and quantity production functions to form a firm's Capability (TFPC)

$$\Pi_j q_j^{ heta_j} x = \zeta_a \phi_a e^{\lambda(\gamma+\delta)} l^{lpha_l+eta_l} k^{lpha_k+eta_k} e^{\epsilon+arepsilon}$$

Unadjusted TFPC $(\phi_u \zeta_u)$ Specification-Adjusted TFPC $(\phi_a \zeta_a)$

Measuring Productivity - Lab Benchmarks

Lab TFPQ = 0.98m²/l

• Specifications fixed by lab design

Lab TFPZ = $\Pi_j q_{lab,j}^{\hat{\theta}_j}$

• For quality productivity, aggregate quality metrics using same θ 's as survey measures

Lab TFPC = Lab TFPQ × Lab TFPZ

• Capabilities is product of quality and quantity productivity

Measuring Productivity: Summary

Quantity Productivity: Unadjusted TFPQ (ϕ_u), Specification-Adjusted TFPQ (ϕ_a)

Quality Productivity: Unadjusted TFPZ (ζ_u), Specification-Adjusted TFPZ (ζ_a)

Capability: Unadjusted TFPC ($\phi_u \zeta_u$), Specification-Adjusted TFPC ($\phi_a \zeta_a$)

Also: estimate **TFPR** using revenues and input expenditures instead of quantities *x*, *l*, *k*

Compare to Lab Benchmarks: Lab TFPQ, Lab TFPZ, Lab TFPC

Outline

1. Industry Background

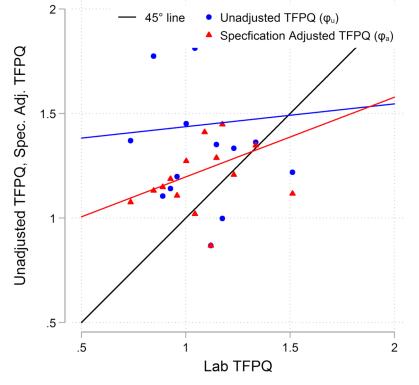
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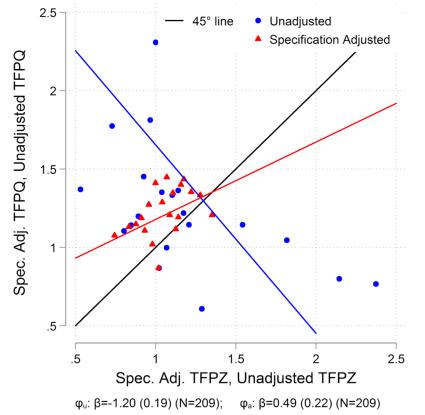
Result 1: Importance of Adjusting for Product Spec.



• Shows value of finer product-category controls for measuring quantity productivity

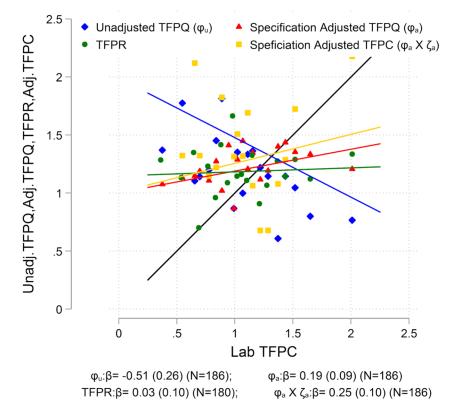
 $φ_u$: β=0.11 (se=0.62) (N=186); $φ_a$: β=0.38 (se=0.21) (N=186)

Result 2: Quantity versus Quality Productivity



TFPQ and TFPQ positively correlated
But only after spec adjusting, as more capable firms make varieties with more demanding specs

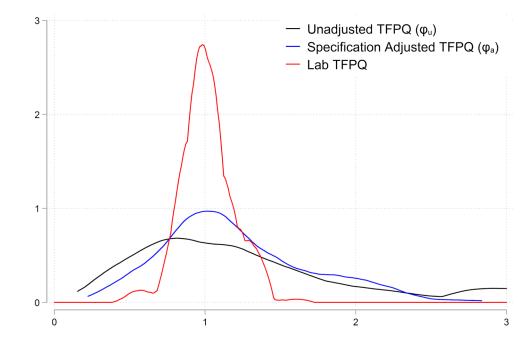
Result 3: TFPR as a Proxy for Capabilities



- Unadjusted TFPQ is misleading capability measure, TFPR more suitable proxy
- But specification adjusting, and/or combining quality and quantity measures superior

Result 4: Unadjusted TFPQ Overstates Dispersion more than Spec-Adjusted TFPQ

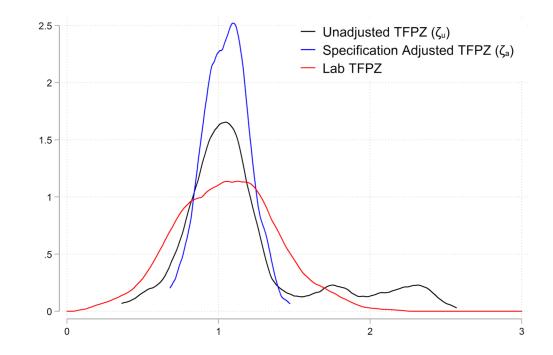
<u>90-10 Ratios:</u> Unadjusted TFPQ: 4.7 Spec-Adjusted TFPQ: 3.0 Lab TFPQ: 1.6



Dispersion in standard datasets may partially reflect product differentiation, rather than differences in underlying productivity

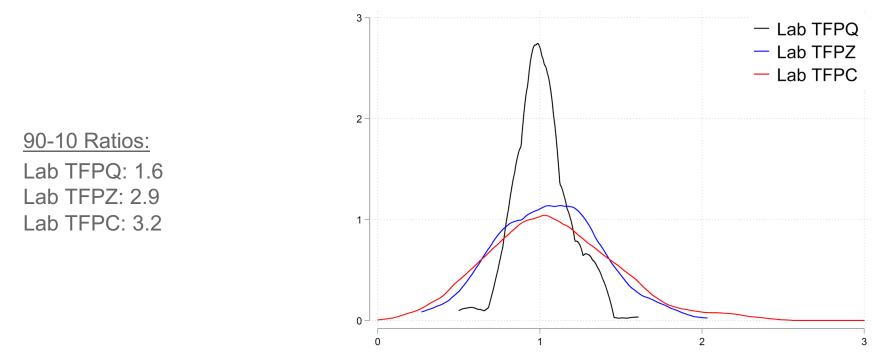
Result 5: TFPZ Dispersion is Large

90-10 Ratios: Unadjusted TFPZ: 2.6 Spec-Adjusted TFPZ: 1.5 Lab TFPZ: 2.9



Even within very narrowly defined product category, large variation in quality across firms

Result 6: TFPC More Dispersed than TFPQ & TFPZ



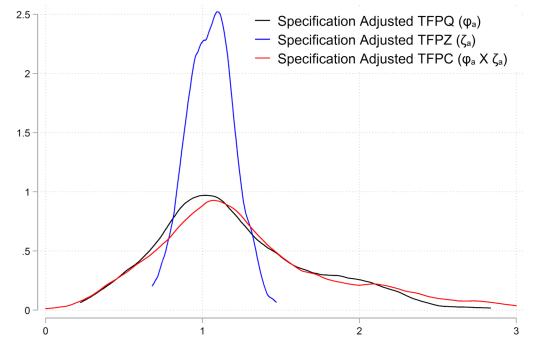
Broad capabilities of firms are more dispersed than a single dimension of productivity

Implication of fact quantity and quality productivity positively correlated

Result 6: TFPC More Dispersed than TFPQ & TFPZ

90-10 Ratios:

Spec-Adjusted TFPQ: 3.0 Spec-Adjusted TFPZ: 1.5 Spec-Adjusted TFPC: 3.6



Broad capabilities of firms are more dispersed than a single dimension of productivity

Implication of fact quantity and quality productivity positively correlated

Conclusions

Combination of tailored surveys and productivity benchmarks provides insight into productivity measurement

- 1. Standard TFPQ performs poorly at measuring quantity productivity
 - Weakly corr. with Lab TFPQ, inversely corr. with quality productivity
 - Using product specifications to make an apples-to-apples comparisons reduces measured productivity dispersion by a third
- 2. Substantial variation in quality productivity and even greater dispersion in capabilities
 - Implication of positive correlation between quality and quantity productivity
- 3. If researchers are interested in broader capabilities of firms:
 - TFPR—for all its imperfections—may be better proxy than (unadjusted) TFPQ
 - But tailored surveys and/or benchmarking may be best way to understand productivity differences across firms