

The Government's Price Index  
Evidence and Implications for Governments' Productivity  
(Very preliminary)

Rodrigo Carril\*

Manuel García-Santana<sup>⊕</sup>

Sampreet S. Goraya<sup>◇</sup>

\*UPF, BSE, CEPR

<sup>⊕</sup>UPF, CREi, BSE, CEPR

<sup>◇</sup>Stockholm School of Economics

STEG, Nairobi

January 2026

# Introduction

## General motivation

- Governments are large producers of final goods and services (  $\frac{G}{Y} \approx 15 - 20\%$  )

## General motivation

- Governments are large producers of final goods and services ( $\frac{G}{Y} \approx 15 - 20\%$ )
- Assessing their efficiency as producers is difficult
  - Market value of their production not directly observable

## General motivation

- Governments are large producers of final goods and services ( $\frac{G}{Y} \approx 15 - 20\%$ )
- Assessing their efficiency as producers is difficult
  - Market value of their production not directly observable
- **This paper:** Could governments do more keeping their budget unchanged?

## General motivation

- Governments are large producers of final goods and services ( $\frac{G}{Y} \approx 15 - 20\%$ )
- Assessing their efficiency as producers is difficult
  - Market value of their production not directly observable
- **This paper:** Could governments do more keeping their budget unchanged?
- **Focus:** Government's performance as a buyer of intermediate inputs

## General motivation

- Governments are large producers of final goods and services ( $\frac{G}{Y} \approx 15 - 20\%$ )
- Assessing their efficiency as producers is difficult
  - Market value of their production not directly observable
- **This paper:** Could governments do more keeping their budget unchanged?
- **Focus:** Government's performance as a buyer of intermediate inputs
  - Are there government agencies paying “too much”?

## General motivation

- Governments are large producers of final goods and services ( $\frac{G}{Y} \approx 15 - 20\%$ )
- Assessing their efficiency as producers is difficult
  - Market value of their production not directly observable
- **This paper:** Could governments do more keeping their budget unchanged?
- **Focus:** Government's performance as a buyer of intermediate inputs
  - Are there government agencies paying “too much”?
  - Could governments provide more public services if they were to pay lower prices?

# Introduction

Fixing some ideas: governments' production in National Accounts

- The government component of GDP,  $G$ , is measured using a cost approach:

⇒  $\boxed{\text{Governments' value of production} = \text{Governments' incurred cost}}$

# Introduction

Fixing some ideas: governments' production in National Accounts

- The government component of GDP,  $G$ , is measured using a cost approach:

⇒ Governments' value of production = Governments' incurred cost

$$G \approx W_g L_g + \underbrace{\text{intermediate goods and services} + \text{structures} + \text{equipment}}_{\text{public procurement}}$$

## Introduction

Fixing some ideas: governments' production in National Accounts

- The government component of GDP,  $G$ , is measured using a cost approach:

⇒ Governments' value of production = Governments' incurred cost

$$G \approx W_g L_g + \underbrace{\text{intermediate goods and services} + \text{structures} + \text{equipment}}_{\text{public procurement}}$$

- Imagine a simple framework (without labor) where:

$$Y_g = M_g$$

## Introduction

Fixing some ideas: governments' production in National Accounts

- The government component of GDP,  $G$ , is measured using a cost approach:

⇒ Governments' value of production = Governments' incurred cost

$$G \approx W_g L_g + \underbrace{\text{intermediate goods and services} + \text{structures} + \text{equipment}}_{\text{public procurement}}$$

- Imagine a simple framework (without labor) where:

$$Y_g = M_g$$

- In this context:

$G = \text{public procurement} = P_g M_g$

# Introduction

Fixing some ideas: governments' production in National Accounts

- The government component of GDP,  $G$ , is measured using a cost approach:

⇒ Governments' value of production = Governments' incurred cost

$$G \approx W_g L_g + \underbrace{\text{intermediate goods and services} + \text{structures} + \text{equipment}}_{\text{public procurement}}$$

- Imagine a simple framework (without labor) where:

$$Y_g = M_g$$

- In this context:

$G = \text{public procurement} = P_g M_g$

- **This paper**: understanding the **factors** driving  $P_g$  and its **implications** for governments' productivity

# Introduction

Motivating examples: purchases of specific products in 2019

- *Electrocardiography EKG monitoring electrode patches (42181708-4)*
  - \$3.72 per unit (paid by Hospital Angol)
  - \$3.98 per unit (paid by Santa Cruz Hospital)
  - \$4.29 per unit (paid by Til Til Hospital)
  - \$7.30 per unit (paid by Hospital Pedro Morales)

# Introduction

Motivating examples: purchases of specific products in 2019

- *Electrocardiography EKG monitoring electrode patches (42181708-4)*
  - \$3.72 per unit (paid by Hospital Angol)
  - \$3.98 per unit (paid by Santa Cruz Hospital)
  - \$4.29 per unit (paid by Til Til Hospital)
  - \$7.30 per unit (paid by Hospital Pedro Morales)
  
- *Insulin (51183603-3)*
  - \$5.04 per unit (paid by Hospital Angol)
  - \$5.18 per unit (paid by Hospital Pitrufoquen)
  - \$5.44 per unit (paid by Hospital Victoria)

# Introduction

Motivating examples: purchases of specific products in 2019

- *Electrocardiography EKG monitoring electrode patches (42181708-4)*
  - \$3.72 per unit (paid by Hospital Angol, ≈ **234 beds**)
  - \$3.98 per unit (paid by Santa Cruz Hospital, ≈ **87 beds**)
  - \$4.29 per unit (paid by Til Til Hospital, ≈ **58 beds**)
  - \$7.30 per unit (paid by Hospital Pedro Morales, < **50 beds**)

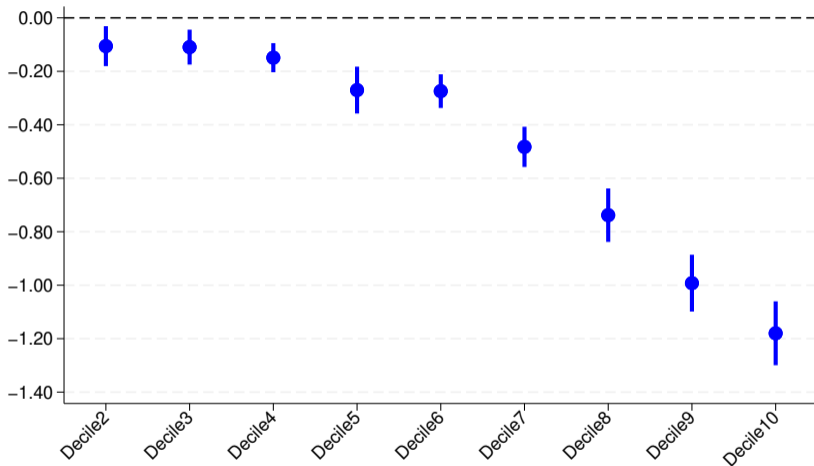
# Introduction

Motivating examples: purchases of specific products in 2019

- *Electrocardiography EKG monitoring electrode patches (42181708-4)*
  - \$3.72 per unit (paid by Hospital Angol, ≈ **234 beds**)
  - \$3.98 per unit (paid by Santa Cruz Hospital, ≈ **87 beds**)
  - \$4.29 per unit (paid by Til Til Hospital, ≈ **58 beds**)
  - \$7.30 per unit (paid by Hospital Pedro Morales, < **50 beds**)
  
- *Insulin (51183603-3)*
  - \$5.04 per unit (paid by Hospital Angol, ≈ 234 beds)
  - \$5.18 per unit (paid by Hospital Pitrufquen, ≈ 75 beds)
  - \$5.44 per unit (paid by Hospital Victoria, ≈ 165 beds)

# Introduction

Motivating examples: purchases of specific products in 2019



**Price discounts** (log points) in *Electrocardiography EKG monitoring electrode patches* by quantity decile

This paper

# This paper

1. **Data:** Universe of public procurement in Chile at the transaction-level
  - Almost 20 years of data: 2007-present
  - Approx. 70 million transactions
  - Products very narrowly defined ( $\approx$  188,000 products), e.g., 10 types of insulin
  - Value, quantity, prices
  - Info on agency buying and identity of the firm selling the product

# This paper

1. **Data:** Universe of public procurement in Chile at the transaction-level
  - Almost 20 years of data: 2007-present
  - Approx. 70 million transactions
  - Products very narrowly defined ( $\approx$  188,000 products), e.g., 10 types of insulin
  - Value, quantity, prices
  - Info on agency buying and identity of the firm selling the product
2. **Evidence:** Large dispersion in prices paid by agencies within narrowly defined categories
  - Currently exploring the role of buyers' size, quantity discounts, etc.

# This paper

## 3. **Measurement:** government agencies' price indexes

- Nested CES structure

★ Whole government → Agency types → Agencies → Products → Suppliers

# This paper

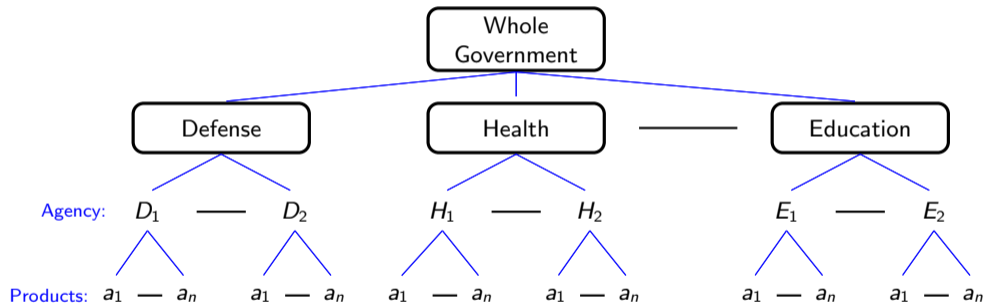
## 3. **Measurement:** government agencies' price indexes

- Nested CES structure

★ Whole government → Agency types → Agencies → Products → Suppliers

# This paper

A simple sketch of government structure



# This paper

## 3. **Measurement:** government agencies' price indexes

- Nested CES structure

★ Whole government → Agency types → Agencies → Products → Suppliers

# This paper

## 3. **Measurement:** government agencies' price indexes

- Nested CES structure
  - ★ Whole government → Agency types → Agencies → Products → Suppliers
- Structurally **estimate elasticities of substitution** across products, agencies, etc., (not today)

# This paper

## 3. **Measurement:** government agencies' price indexes

- Nested CES structure
  - ★ Whole government → Agency types → Agencies → Products → Suppliers
- Structurally **estimate elasticities of substitution** across products, agencies, etc., (not today)
- **Infer underlying markups** charged to different agencies
  - ★ Today (first step): infer markups as “**wedges**” within narrow categories, e.g., *product-supplier-month*
  - ★ Plan: estimate markups using **data on actual procurement auctions**

# This paper

## 3. **Measurement:** government agencies' price indexes

- Nested CES structure
  - ★ Whole government → Agency types → Agencies → Products → Suppliers
- Structurally estimate elasticities of substitution across products, agencies, etc., (not today)
- Infer underlying markups charged to different agencies
  - ★ Today (first step): infer markups as “wedges” within narrow categories, e.g., *product-supplier-month*
  - ★ Plan: estimate markups using **data on actual procurement auctions**

## 4. **Counterfactuals**

- Remove “markups” and compute resulting price indexes
- How much more  $Y_g$  could the government produce keeping expenditure unchanged?

Related literature

## Related literature

### **Waste in procurement**

- [Bandiera et al. \(AER, 2009\)](#)
  - Some public agencies pay systematically more than others in Italy
  - Focus on 21 standardized products (e.g., “Heating diesel”)
  - Mostly driven by “passive waste”

## Related literature

### Waste in procurement

- [Bandiera et al. \(AER, 2009\)](#)
  - Some public agencies pay systematically more than others in Italy
  - Focus on 21 standardized products (e.g., “Heating diesel”)
  - Mostly driven by “passive waste”
- [Best et al. \(AER, 2023\)](#)
  - Large dispersion in procurement prices in Russia
  - $\approx 40\%$  explained by bureaucrats FEs

## Related literature

### Waste in procurement

- [Bandiera et al.](#) (AER, 2009)
  - Some public agencies pay systematically more than others in Italy
  - Focus on 21 standardized products (e.g., “Heating diesel”)
  - Mostly driven by “passive waste”
- [Best et al.](#) (AER, 2023)
  - Large dispersion in procurement prices in Russia
  - $\approx 40\%$  explained by bureaucrats FEs

### Efficiency losses from price dispersion in the private sector

- [Burstein, Cravino, Rojas](#) (WP, 2025)
  - Use firm-to-firm transaction data from Chile
  - Some firms pay systematically more than others for the same input
  - Significant productivity gains from equating markups paid by buyers

# 1. Data

# 1. Data

## Universe of public procurement in Chile at the transaction-level

- Almost 20 years of data: 2007-present
- Approx. 70 million transactions
- Products at the *United Nations Standard Products and Services Code* ( $\approx$  188,000 products)
- Value, quantity, prices
- Info on agency buying and identity of the firm selling the product

# 1. Data

Aggregated transaction data accounts for a significant fraction of GDP

Table: Summary Statistics

Year	All	Goods	Services	All/GDP (%)	Goods/GDP (%)	Services/GDP (%)
2014	4,154	2,139	2,015	2.81	1.45	1.36
2015	5,203	2,302	2,900	3.28	1.45	1.83
2016	5,782	2,773	3,009	3.43	1.64	1.78
2017	6,685	2,685	4,001	3.73	1.50	2.23
2018	6,607	2,547	4,060	3.49	1.34	2.14
2019	11,316	2,923	8,393	5.79	1.50	4.29

- Today: focus on **Goods**

# 1. Data

## Classifying agencies into "Types"

Table: Distribution of Sectors

Entity Sector	Num entities	Total Value	Share (%)
Defense	377	1,970	4.96
Health - Governance	8	3,699	9.31
Health - Public Services	222	9,253	23.28
Education - Governance	67	504	1.27
Education - Public Services	22	2,505	6.30
Central Government	383	4,445	11.18
Local Government	455	14,437	36.32
Other	266	1,717	4.32
NA	227	1,217	3.06

- Today: focus on **Health - Public Services**

## 2. Evidence Measurement issues

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:
  - Different agencies buy different brands/quality

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:
  - Different agencies buy different brands/quality
  - Measurement error/mistakes in the **units of measurement**
- Our current approach: apply a standard ***k*-means clustering algorithm**

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:
  - Different agencies buy different brands/quality
  - Measurement error/mistakes in the **units of measurement**
- Our current approach: apply a standard **k-means clustering algorithm**
  - (i) Cluster prices by assigning them to the closest mean
  - (ii) Update means to minimize within-cluster distance

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:
  - Different agencies buy different brands/quality
  - Measurement error/mistakes in the **units of measurement**
- Our current approach: apply a standard ***k*-means clustering algorithm**
  - (i) Cluster prices by assigning them to the closest mean
  - (ii) Update means to minimize within-cluster distance
- Use the outcome to create new “product” definitions. For example:

## 2. Evidence

### Measurement issues

- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:
  - Different agencies buy different brands/quality
  - Measurement error/mistakes in the **units of measurement**
- Our current approach: apply a standard **k-means clustering algorithm**
  - (i) Cluster prices by assigning them to the closest mean
  - (ii) Update means to minimize within-cluster distance
- Use the outcome to create new “product” definitions. For example:
  - “*Plain savory biscuits (50181903-1)*”

## 2. Evidence

### Measurement issues

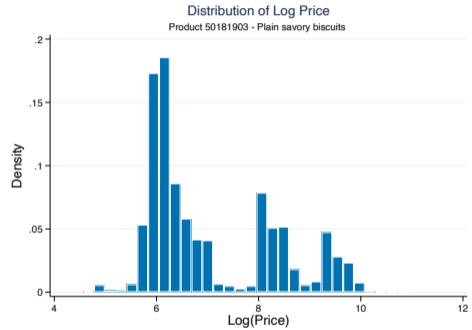
- Comparing prices paid by different agencies problematic even within narrowly defined products
- For example, take the case of “*Plain savory biscuits (50181903)*”. Comparing prices problematic if:
  - Different agencies buy different brands/quality
  - Measurement error/mistakes in the **units of measurement**
- Our current approach: apply a standard **k-means clustering algorithm**
  - (i) Cluster prices by assigning them to the closest mean
  - (ii) Update means to minimize within-cluster distance
- Use the outcome to create new “product” definitions. For example:
  - “*Plain savory biscuits (50181903-1)*”
  - “*Plain savory biscuits (50181903-2)*”
  - “*Plain savory biscuits (50181903-3)*”

## 2. Evidence

### Measurement issues. An example



(a) Time series



(b) Histogram

## 2. Evidence

### Measurement issues. Clustering results

#### Coverage:

Products	Product-Clusters
11,765	15,854

**Takeaway:** ~67% of observations belong to a single price cluster; ~33% exhibit price heterogeneity

#### Distribution of Cluster Counts:

No. Clusters	Share
1	66.86%
2	22.75%
3	6.06%
4	3.52%
5	0.82%

## 2. Evidence

Price dispersion within narrowly defined categories

## 2. Evidence. Price dispersion

### Variables definition

- Let's  $p_{ib\tau}$  denote the price paid by agency  $b$  for product-cluster  $i$  in transaction  $\tau$
- $p_i$  be monthly average of prices paid for product  $i$  across all buyers

## 2. Evidence. Price dispersion

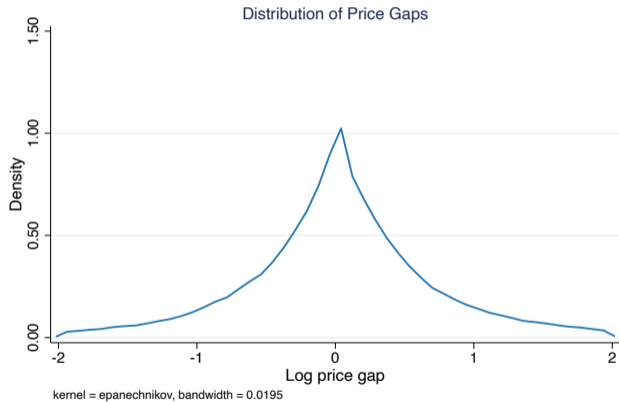
### Variables definition

- Let's  $p_{ib\tau}$  denote the price paid by agency  $b$  for product-cluster  $i$  in transaction  $\tau$
- $p_i$  be monthly average of prices paid for product  $i$  across all buyers
- Define the “price gap” as:

$$\psi_{ib\tau} \equiv \log \left( \frac{p_{ib\tau}}{p_i} \right)$$

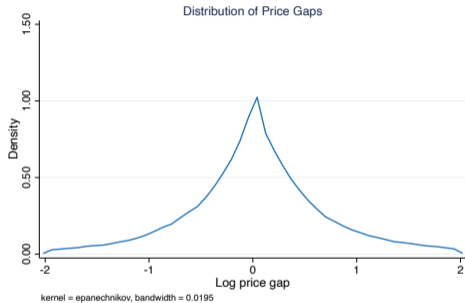
## 2. Evidence. Price dispersion

### A. Large dispersion in "price gaps" across agencies

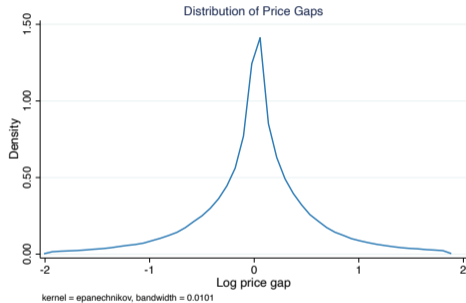


## 2. Evidence. Price dispersion

### A. Large dispersion in "price gaps" across agencies



(a) Price gaps within products

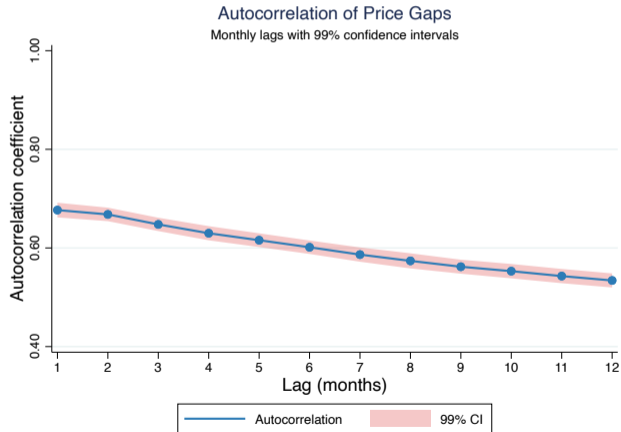


(b) Price gaps within products-seller

## 2. Evidence. Price dispersion

B. These “price gaps” are persistent over time

### Persistence in Price Gaps



## 2. Evidence. Price dispersion

C. These “price gaps” are lower when the buyer is large

- Compute buyer size as the **total quantity of the product purchased** across all time periods
- Compare average price gaps across buyers of different sizes.

Table: Price Gaps and Buyer Size

	(1)	(2)
	(1)	(2)
Log(Total Quantity)	-0.052*** (0.001)	-0.069*** (0.001)
Product-cluster FE	Yes	Yes
Buyer FE	No	Yes
Observations	375,452	375,451
$R^2$	0.264	0.279

## 2. Evidence. Price dispersion

D. These “price gaps” are lower when relationships are longer

Table: Price Gaps and Length of Relationships

	(1)	(2)
Recent Relationship (<1 Year)	-0.005 (0.004)	-0.006** (0.003)
Established Relationship ( $\geq 1$ Year)	-0.009** (0.004)	-0.013*** (0.003)
Product-cluster FE	Yes	Yes
Buyer-Seller FE	No	Yes
Observations	7,153,240	7,122,308
$R^2$	0.179	0.329

## 2. Evidence. Price dispersion

E. These “price gaps” are lower when transactions are larger

Table: Price Gaps and Buyer-Seller Relationships

	(1)	(2)	(3)	(4)
Log(Quantity)	-0.182*** (0.003)	-0.190*** (0.004)	-0.098*** (0.002)	-0.135*** (0.004)
Log(Quantity) × Log(Quantity)		0.001*** (0.000)		0.005*** (0.000)
Product-cluster FE	Yes	Yes	No	No
Buyer-Seller FE	Yes	Yes	No	No
Product-Buyer-Seller FE	No	No	Yes	Yes
Observations	7,122,308	7,122,308	6,777,667	6,777,667
$R^2$	0.391	0.391	0.530	0.530

## 2. Evidence. Price dispersion across buyers

### F. Differences in “price ranges” across products

- $\max(p_{ibt})$  = maximum price paid for product-cluster  $i$  across all buyers in month  $t$
- $\min(p_{ibt})$  = minimum price paid for product-cluster  $i$  across all buyers in month  $t$

## 2. Evidence. Price dispersion across buyers

### F. Differences in “price ranges” across products

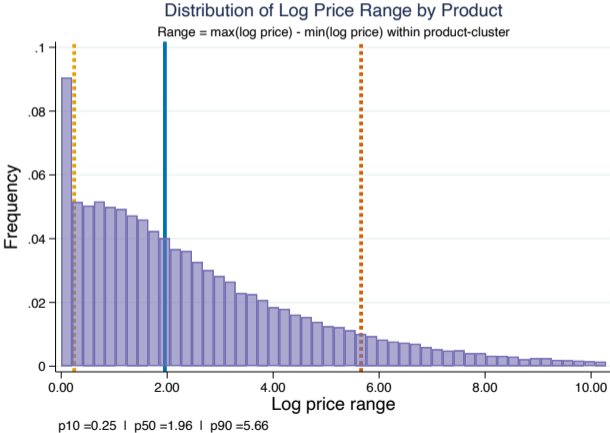
- $\max(p_{ibt})$  = maximum price paid for product-cluster  $i$  across all buyers in month  $t$
- $\min(p_{ibt})$  = minimum price paid for product-cluster  $i$  across all buyers in month  $t$
- Compute the “price range” of a product-cluster  $i$  as:

$$\text{range}_i \equiv \log \left( \frac{\max(p_{ibt})}{\min(p_{ibt})} \right)$$

## 2. Evidence. Price dispersion across buyers

### F. Differences in "price ranges" across products

#### Range of Price Gaps



### 3. Measurement: governments' price indexes

### 3. Measurement: governments' price indexes

Construct "baseline" price indexes in "Health - Public Services"

- In "Health - Public Services", the government buys through many agencies, i.e., hospitals  $g$

$$Y = \left[ \left( \frac{1}{G} \right)^{1/\sigma_g} \sum_g Y_g^{(\sigma_g-1)/\sigma_g} \right]^{\sigma_g/(\sigma_g-1)} ; \quad P = \left[ \frac{1}{G} \sum_i P_g^{1-\sigma_g} \right]^{1/(1-\sigma_g)}$$

### 3. Measurement: governments' price indexes

Construct "baseline" price indexes in "Health - Public Services"

- In "Health - Public Services", the government buys through many agencies, i.e., hospitals  $g$

$$Y = \left[ \left( \frac{1}{G} \right)^{1/\sigma_g} \sum_g Y_g^{(\sigma_g-1)/\sigma_g} \right]^{\sigma_g/(\sigma_g-1)} ; \quad P = \left[ \frac{1}{G} \sum_i P_g^{1-\sigma_g} \right]^{1/(1-\sigma_g)}$$

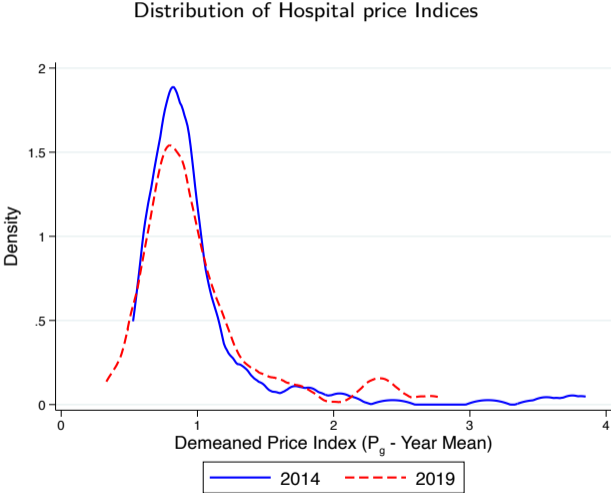
- Each hospital  $g$  buys different combination of products  $i$

$$Y_g = \left[ \left( \frac{1}{I_g} \right)^{1/\sigma_f} \sum_i y_{gi}^{(\sigma_f-1)/\sigma_f} \right]^{\sigma_f/(\sigma_f-1)} ; \quad P_g = \left[ \frac{1}{I_g} \sum_i P_{gi}^{1-\sigma_f} \right]^{1/(1-\sigma_f)}$$

- We use  $\sigma_f = 5$  and  $\sigma_g = 3$ .

### 3. Measurement: governments' price indexes

Large dispersion in price indexes across hospitals



## 4. Counterfactual price indexes

## 4. Counterfactual price indexes

A simple (preliminary) approach

- Define a markup charged to hospital  $g$  as:

$$\mu_{ig} = \frac{p_{ig}}{\min(p_{ig})}$$

## 4. Counterfactual price indexes

### A simple (preliminary) approach

- Define a markup charged to hospital  $g$  as:

$$\mu_{ig} = \frac{p_{ig}}{\min(p_{ig})}$$

- The counterfactual price index is

$$P_g^C = \left[ \frac{1}{I_g} \sum_i (p_{gi} \mu_{ig}^{-1})^{1-\sigma_f} \right]^{1/(1-\sigma_f)}$$

## 4. Counterfactual price indexes

### A simple (preliminary) approach

- Define a markup charged to hospital  $g$  as:

$$\mu_{ig} = \frac{p_{ig}}{\min(p_{ig})}$$

- The counterfactual price index is

$$P_g^C = \left[ \frac{1}{I_g} \sum_i (p_{gi} \mu_{ig}^{-1})^{1-\sigma_f} \right]^{1/(1-\sigma_f)}$$

- Counterfactual 1:  $\min(p_{ig})$  is the minimum price paid by hospitals within a month-product category.

## 4. Counterfactual price indexes

### A simple (preliminary) approach

- Define a markup charged to hospital  $g$  as:

$$\mu_{ig} = \frac{p_{ig}}{\min(p_{ig})}$$

- The counterfactual price index is

$$P_g^C = \left[ \frac{1}{I_g} \sum_i (p_{gi} \mu_{ig}^{-1})^{1-\sigma_f} \right]^{1/(1-\sigma_f)}$$

- Counterfactual 1:  $\min(p_{ig})$  is the minimum price paid by hospitals within a month-product category.

## 4. Counterfactual price indexes

### A simple (preliminary) approach

- Define a markup charged to hospital  $g$  as:

$$\mu_{ig} = \frac{p_{ig}}{\min(p_{ig})}$$

- The counterfactual price index is

$$P_g^C = \left[ \frac{1}{I_g} \sum_i (p_{gi} \mu_{ig}^{-1})^{1-\sigma_f} \right]^{1/(1-\sigma_f)}$$

- Counterfactual 1:  $\min(p_{ig})$  is the minimum price paid by hospitals within a month-product category.
- Counterfactual 2:  $\min(p_{ig})$  is the minimum price paid by hospitals within a month-seller-product category.

## 4. Counterfactual price indexes

### A simple (preliminary) approach

- Define a markup charged to hospital  $g$  as:

$$\mu_{ig} = \frac{p_{ig}}{\min(p_{ig})}$$

- The counterfactual price index is

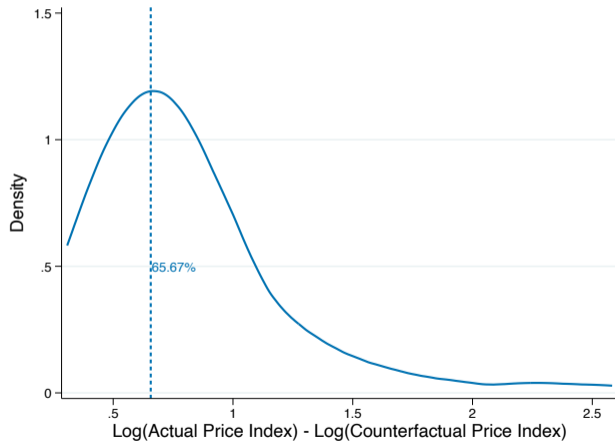
$$P_g^C = \left[ \frac{1}{I_g} \sum_i (p_{gi} \mu_{ig}^{-1})^{1-\sigma_f} \right]^{1/(1-\sigma_f)}$$

- Counterfactual 1:  $\min(p_{ig})$  is the minimum price paid by hospitals within a month-product category.
- Counterfactual 2:  $\min(p_{ig})$  is the minimum price paid by hospitals within a month-seller-product category.
- Note: If no variation within a given category, price is kept the same

## 4. Counterfactual price indexes

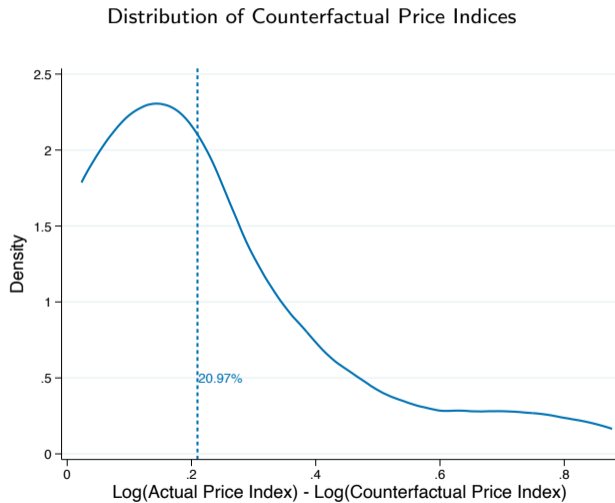
A simple (preliminary) approach

Distribution of Counterfactual Price Indices



## 4. Counterfactual price indexes

A simple (preliminary) approach



Final remarks

## Final remarks

- Our (super) preliminary numbers suggest that
  - Governments could produce 20-65% more if they were to pay less for their intermediate inputs
- Lots of work to be done. Examples:
  - Measurement issues, quality, etc.
  - Markups estimation
  - Estimation of elasticities of substitution
- Policy implications?

## Clustering Methodology

**Objective:** Identify homogeneous goods based on price distribution within each *productid*

**Approach:**

- Focus on log of price, pooling all buyers and sellers
- Outliers removed beforehand

**Algorithm:** PAM (Partitioning Around Medoids) / CLARA

- K-medoids clustering: each cluster represented by an actual data point (medoid)
- CLARA extension used for large samples

**Optimal  $k$  selection:**

- Unsupervised: test all possible  $k$  from 1 to  $n - 1$  buyers
- Selection criterion: **Silhouette score** (ranges  $-1$  to  $1$ ; higher = better fit)