

# ARE 202: Welfare: Tools and Applications

## Spring 2018

Thibault FALLY

### Lecture notes 04 – Quantifying Consumer Welfare

## Plan

### 1. Tools

- EV, CV
- Consumer surplus
- Price indexes

### 2. Illustrations

- Atkin, Faber, Gonzales-Navarro (2016):  
impact of foreign store openings in Mexico
- PS3: Consumer surplus: Uber (Cohen et al 2016)

# Motivation

- Welfare is what we care about (eventually)
- But lots of difficulties:
  - How to quantify welfare changes?
  - How to compare effects across individuals?
- There are several ways to answer these questions:  
Definitions and properties of EV, CV, CS and ideal price index
- Important to know how to apply these tools, and know how they differ

## Quantifying welfare changes

Quantifying the effect of change in income:

- Easy: that's the change in income

Harder: quantifying the effect of change in prices.

Two approaches:

- 1) Change in income to compensate the change in prices?  
= **Compensating Variation (CV)**
- 2) Change in income equivalent to the change in prices?  
= **Equivalent Variation (EV)**

Both approaches make use of the expenditure function  $e(p, u)$ .

# Quantifying welfare changes

- Consider a change in prices from  $p$  to  $p'$  (fixed income  $w$ ). Utility goes from  $u = v(p, w)$  to  $u' = v(p', w)$ .
- The change in income that would *compensate* the change in prices would correspond to:

$$\text{Compensating Variation} = e(p, u) - e(p', u) = w - e(p', u)$$

[using: previous utility  $u$ , new prices  $p'$ ]

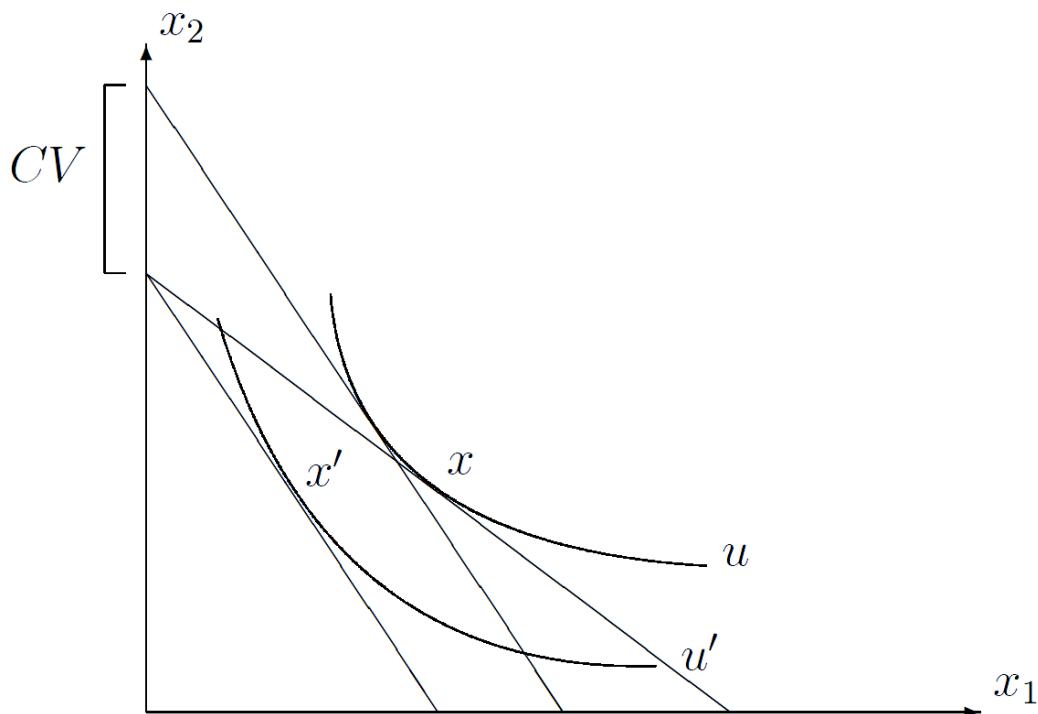
Note that we also have:  $v(p', w + CV) = v(p, w)$

- The change in income that would be *equivalent* to the change in prices would correspond to:

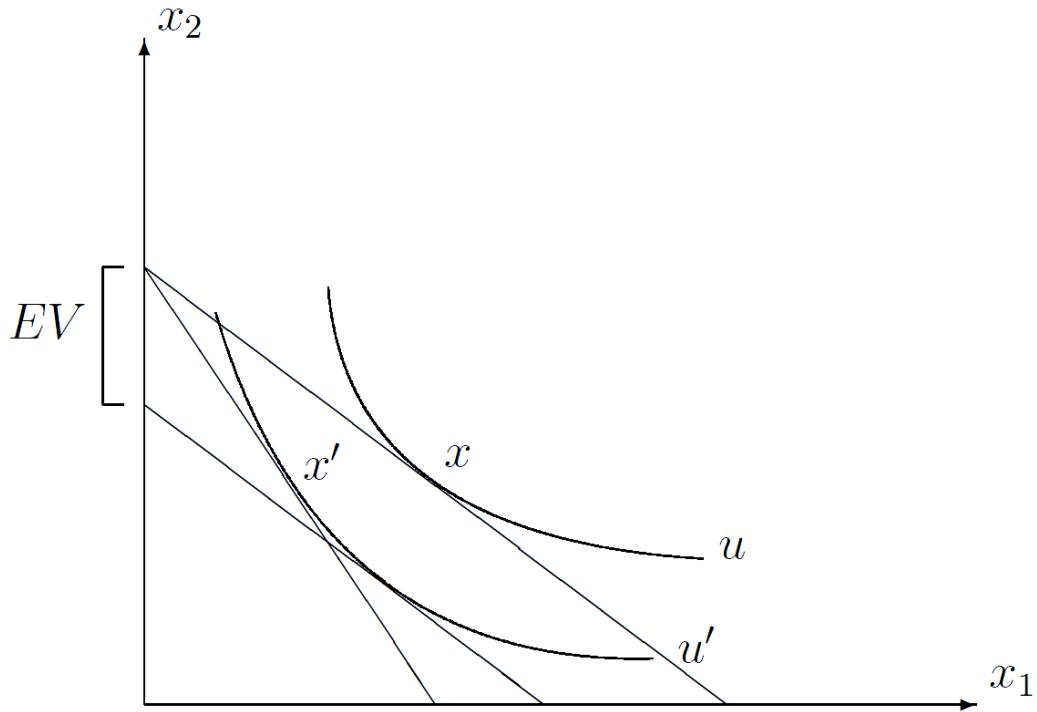
$$\text{Equivalent Variation} = e(p, u') - e(p', u') = e(p, u') - w$$

[using: new utility  $u'$ , previous prices  $p$ ]

## Compensating variation



## Equivalent variation



## Link to the shape of the demand curve

- Suppose that the prices change only for good  $i$
- Using Shephard's Lemma, we get:

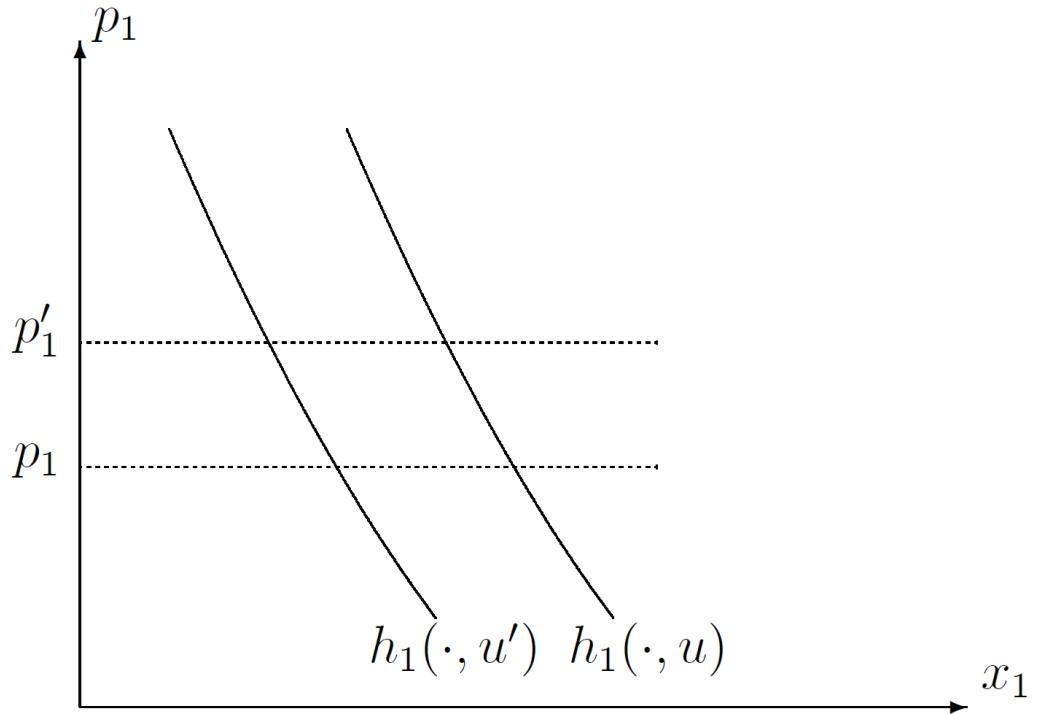
$$CV = e(p, u) - e(p', u) = \int_{p'_i}^{p_i} \frac{\partial e(p, u)}{\partial p_i} dp_i = \int_{p'_i}^{p_i} h_i(p, u) dp_i$$

- Similarly:

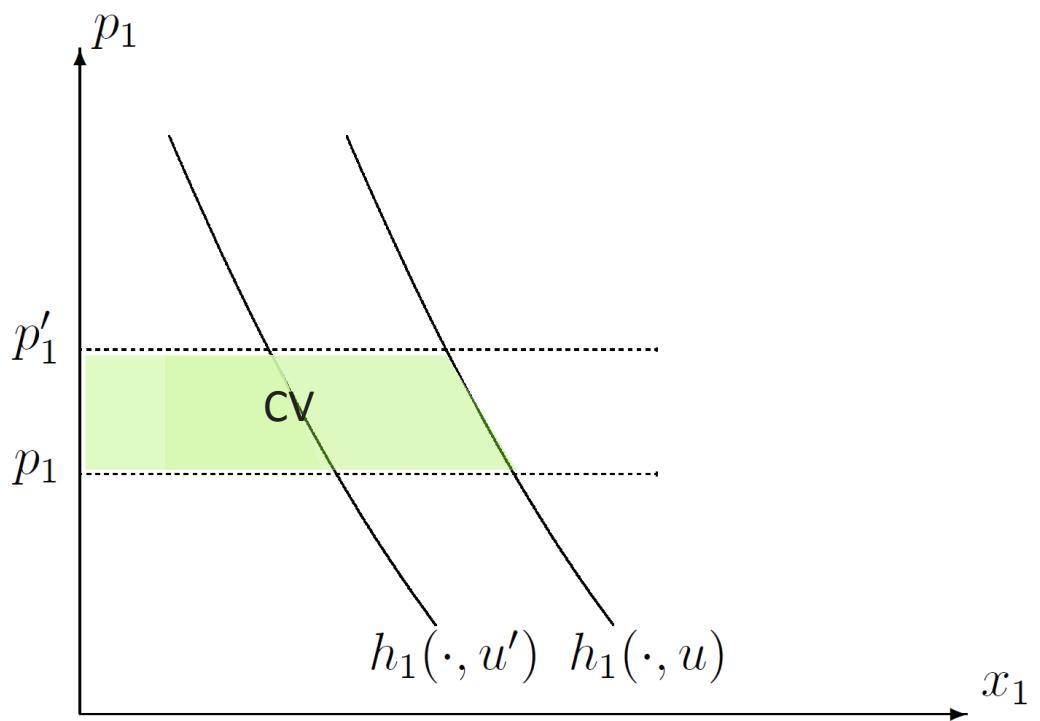
$$EV = e(p, u') - e(p', u') = \int_{p'_i}^{p_i} \frac{\partial e(p, u')}{\partial p_i} dp_i = \int_{p'_i}^{p_i} h_i(p, u') dp_i$$

- Graphically: areas “below” the Hicksian Demand  
(i.e. to the left since prices are on the Y-axis)

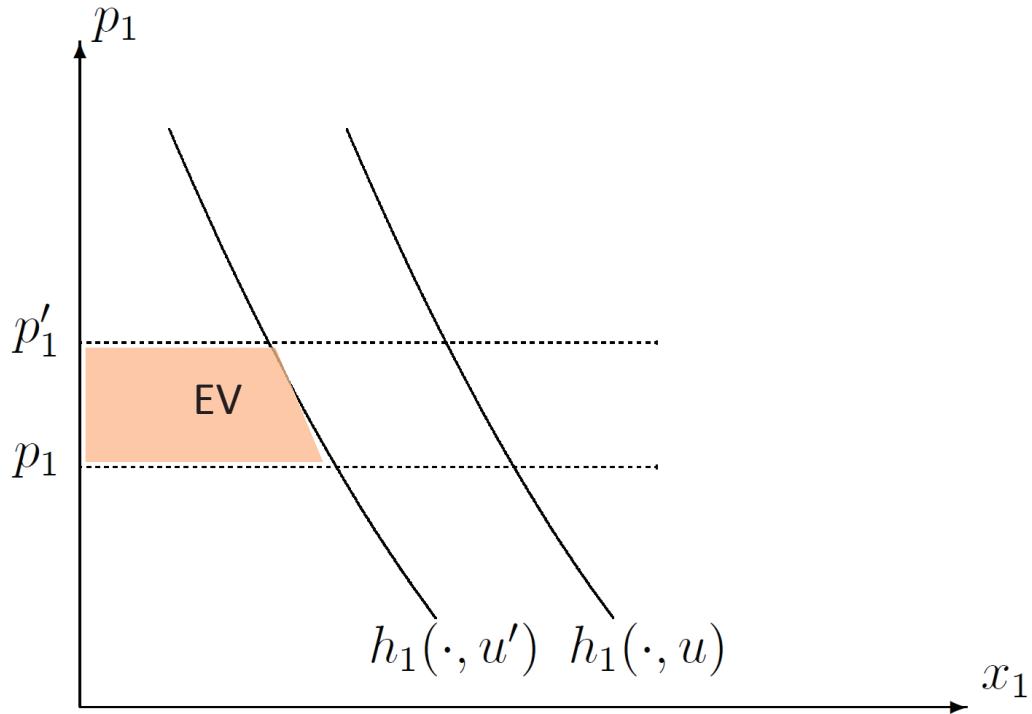
Hicksian demand for utility  $u$  and  $u'$ , assuming  $u' < u$  and normal good



Compensating variation



## Equivalent variation



## Consumer Surplus

- What if we use Marshallian instead of Hicksian Demand?
- Following the same idea, we define consumer surplus:

$$CS = \int_{p'_i}^{p_i} x_i(p, w) dp_i$$

- At the end points, notice that:

$$x_i(p', w) = h_i(p', u')$$

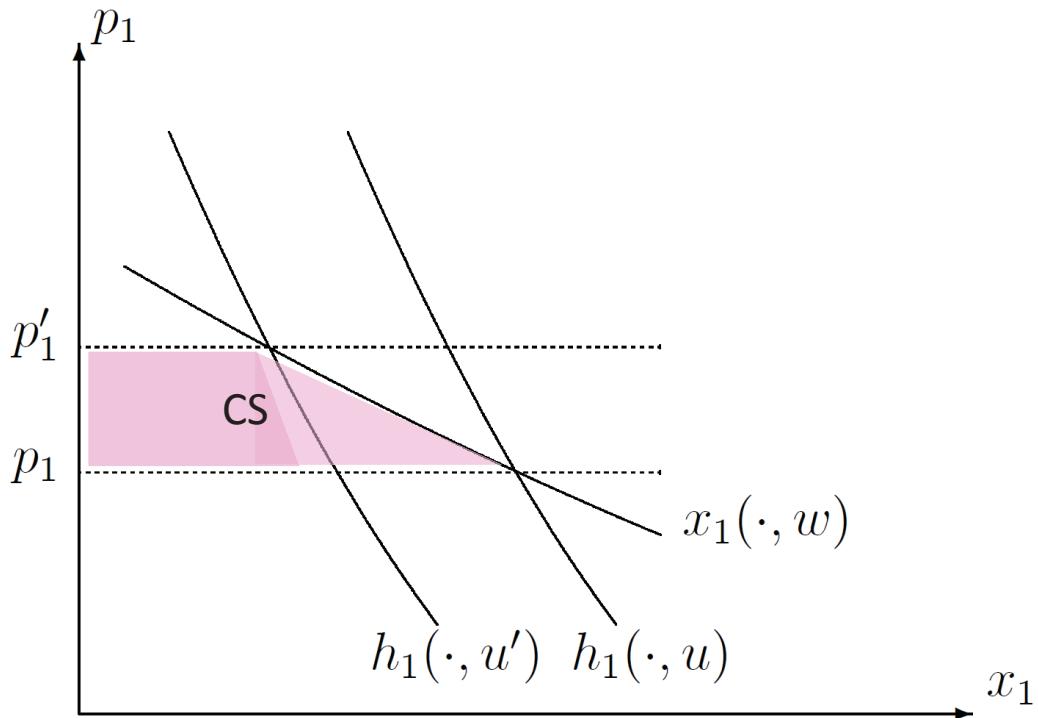
$$x_i(p, w) = h_i(p, u)$$

- With a normal good, we obtain:

$$EV < CS < CV$$

(reversed for an inferior good)

## Consumer surplus



## A simple case

- Assume quasi-linear preferences

$$U(x) = x_0 + \sum u_i(x_i)$$

- Recall some of the properties of quasi-linear prefs:
    - Lagrange multiplier  $\lambda = p_0 = 1$  (normalization of  $p_0$ )
    - Demand such that:  $u'_i(x_i) = p_i$   
Marshallian demand  $x_i$  only depends on price  $p_i$
    - **No wealth effect** (except for numeraire good  $x_0$ ),  
Hence same price effect for Hicksian and Marshallian Demand

$$\frac{\partial x_i}{\partial p_j} = \frac{\partial h_i}{\partial p_j}$$

- In this case, we get:

$$CV \equiv FV \equiv CS$$

# Willig (1976)

- Dilemma: CS easier to compute but has no theoretical foundation and differs from CV and EV as soon as income elasticity is non-zero
- However in practice:  
difference between CS, EV and CV are usually smaller than error due to estimation, and small when the effect on welfare is small.
- Willig (1976): for  $X \in \{EV, CV\}$

$$\frac{\eta^{\min}}{2} \cdot \frac{CS}{w} < \left| \frac{X - CS}{CS} \right| < \frac{\eta^{\max}}{2} \cdot \frac{CS}{w}$$

where  $\eta^{\min}$  and  $\eta^{\max}$  are the min and max income elasticity of demand

⇒ Relative error  $\left| \frac{X - CS}{CS} \right|$  is small with small shares in consumption  $\frac{CS}{w}$

## Comments on Willig (1976)

However, there are a number of reasons why the Willig result cannot always be used to justify the MCS as a good approximation to the CV and EV:

- (1) The Willig result doesn't carry over to the multiple prices changes, assumptions not always satisfied
- (2) Often we are trying to estimate the CS associated with a change in the prices and characteristics of some good or goods and/or a change in the level of non-market commodities, but the Willig result does not carry over to characteristics/non-market space (see Hanemann 1991, Shogrun et al 1994).
- (3) There is no need to approximate. We can get the exact CS measures. This is most easily seen by appealing to duality theory.

# Hausman (1981)

- Computes exact EV and CV (and DWL) rather than approximation
- Use Shephard's lemma and Roy's identity to retrieve Hicksian demand and expenditure function.

Steps:

1. Using Roy's identity, we can retrieve the indirect utility function (solve differential equation in  $v(w, p)$ )
  2. Invert the indirect utility to get the expenditure function:  
 $v(e(u, p), p) = u$
  3. Obtain the Hicksian demand using Shephard's Lemma:  
 $h_i(u, p) = \frac{\partial e(u, p)}{\partial p_i}$
  4. Use either the expenditure function or Hicksian demand to get CV or EV
- Note: Simple way = specify demand to estimate (e.g. CES) where the expenditure function can easily be computed from these estimates.

## Consumer welfare with discrete-choice models

- The same tools can be used (McFadden 1978, 1981, Small Rosen 1981)
- Aggregating many consumers  $z$  with indirect utility across choices  $i$ :

$$U_z = \min_i \{ \alpha(y - p_i) + \phi(Z_i) + \epsilon_{zi} \} = \min_i \{ V_{zi} + \epsilon_{zi} \}$$

with  $\epsilon_{zi} \sim e^{-e^{-\epsilon}}$ , we get:

$$EV = \int \frac{U_{zt'} - U_{zt}}{\alpha} dF(\epsilon) = \frac{1}{\alpha} \log \left( \frac{\sum_i \exp V_{zit'}}{\sum_i \exp V_{zit}} \right)$$

- But becomes quickly messy if we aggregate across consumers with heterogeneous  $\alpha$ 's interacting with many product characteristics  $Z_i$

# Plan

## 1. Tools

- EV, CV
- Consumer surplus
- Price indexes

## 2. Illustrations

## Ideal price index

- We've already seen Laspeyres and Paasche price indexes (using initial and new consumption as respective weights)

$$P^{\text{Laspeyres}} = \frac{x \cdot p'}{x \cdot p} \quad P^{\text{Paasche}} = \frac{x' \cdot p'}{x' \cdot p}$$

- More generally, an **ideal** price index is defined as:

$$\text{Ideal Index} = \frac{e(p', u)}{e(p, u)} = \text{Ideal}(u)$$

- With homothetic preferences,  $\text{Ideal}(u)$  does not depend on  $u$

# Comparison to Paasche and Laspeyres

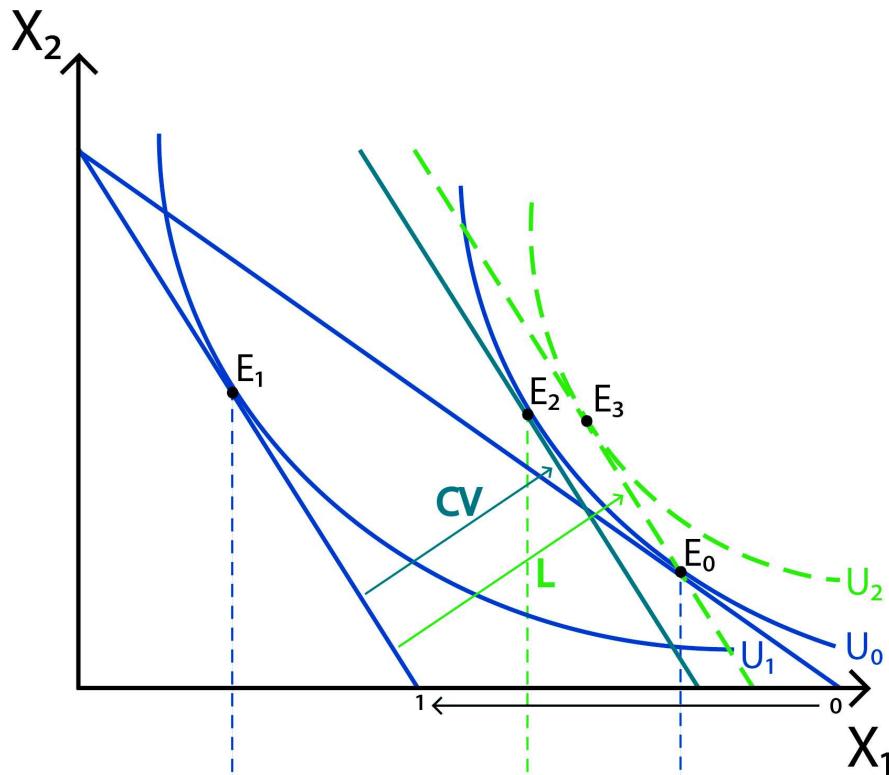
- Notice the “substitution bias”:

$$P_{Laspeyres} = \frac{x \cdot p'}{w} = \frac{x \cdot p'}{e(p, u)} \geq \frac{e(p', u)}{e(p, u)} = Ideal(u)$$

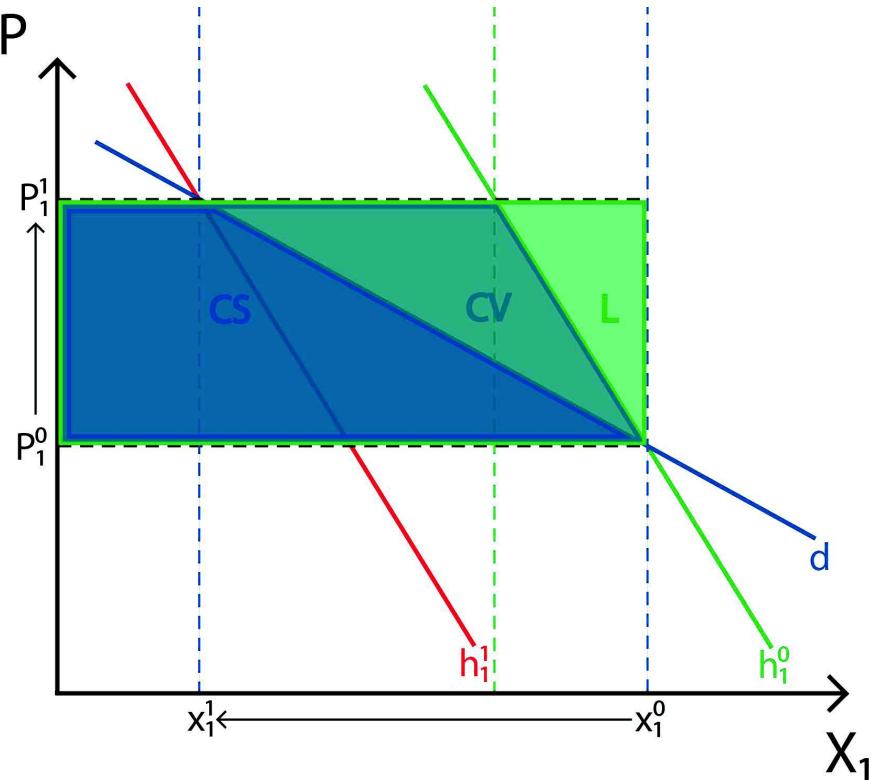
$$P_{Paasche} = \frac{w}{x' \cdot p} = \frac{e(p', u')}{x' \cdot p} \leq \frac{e(p', u')}{e(p, u')} = Ideal(u')$$

- Laspeyres and Paasche are ideal (or “exact”) only for Leontief preferences
- We can show that:  $P < EV < CS < CV < L$  for normal goods (graphical proof in the next slides)

Compensating variation vs. Laspeyres price index, when price of good 1 increases:

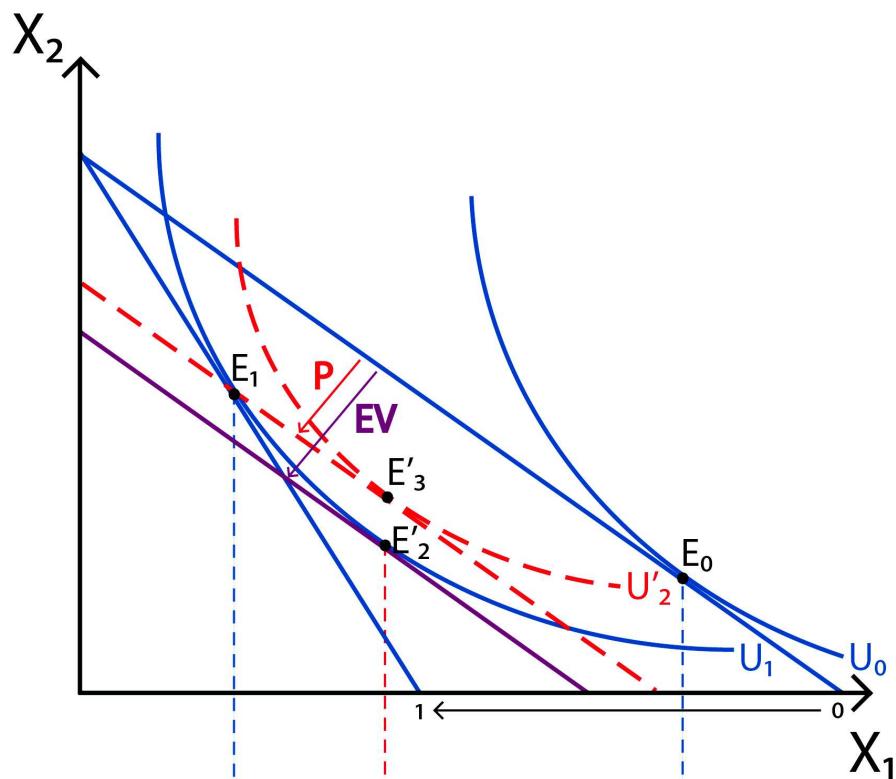


## Compensating variation vs. Laspeyres price index



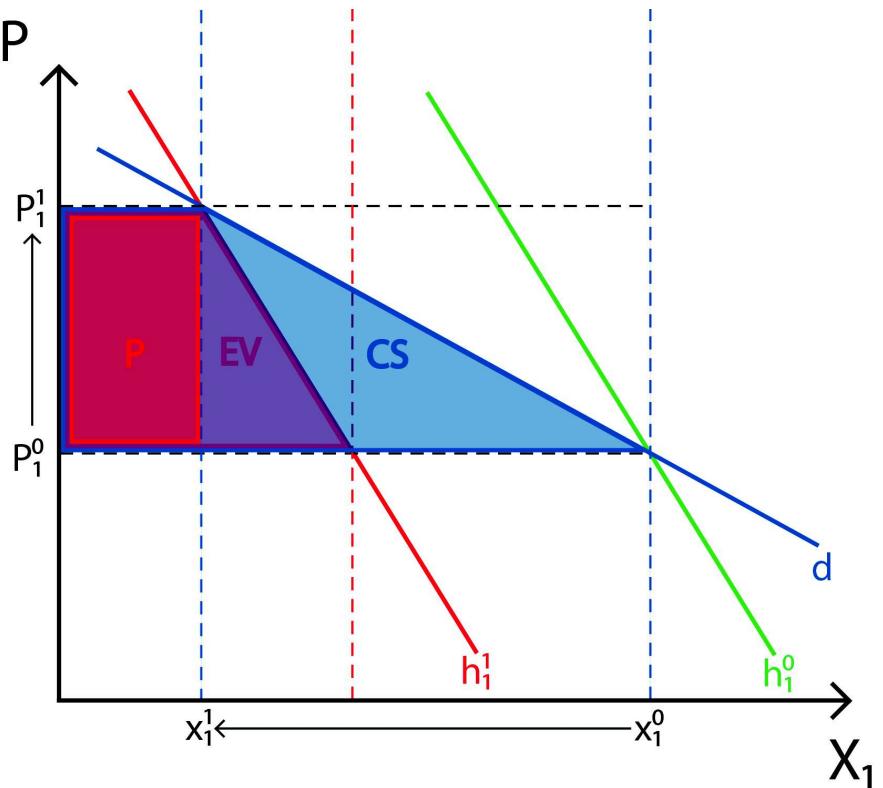
◀ □ ▶ ⏪ ⏩ ⏴ ⏵ ⏷ ⏸ ⏹ ⏺ ⏻ ⏻ ⏻

## Equivalent variation vs. Paasche price index

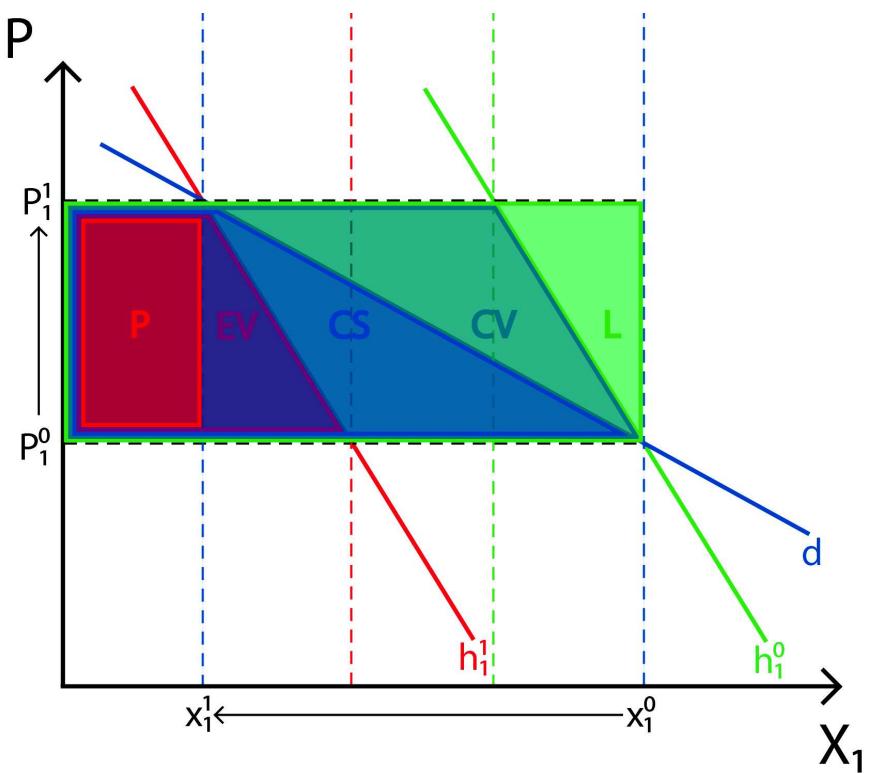


◀ □ ▶ ⏪ ⏩ ⏴ ⏵ ⏷ ⏸ ⏹ ⏻ ⏻ ⏻

## Equivalent variation vs. Paasche price index



$P < EV < CS < CV < L$  for normal goods:



## Simple example

With CES preferences  $U = \left[ \sum_i (b_i x_i)^{\frac{\sigma-1}{\sigma}} \right]^{\frac{\sigma}{\sigma-1}}$

- Expenditure function:  $e(U, p) = UP$ , defining  $U$  as above and  $P$  as:
- CES ideal price index:  $P = \left[ \sum_i b_i^\sigma p_i^{1-\sigma} \right]$
- Equivalent variation:  $EV = P.U' - w = (P - P').U'$
- Compensating variation:  $CV = w - P'.U = (P - P').U$
- Generally, with homothetic preferences, it is easier and more direct to describe changes in price indexes  $P'/P$  than EV, CV and CS

## More price indexes

- Fisher price index: geometric average of Paasche and Laspeyres

$$\log P^{Fisher} = \frac{1}{2} \left( \log P^{Laspeyres} + \log P^{Paasche} \right)$$

- Stone price index (using consumption shares  $s_{ti}$ , exact for CD prefs):

$$\log P^{Stone} = \sum_i s_{i1} \log \left( \frac{p_{i1}}{p_{i0}} \right)$$

- Tornqvist price index (frequently used, exact for translog preferences):

$$\log P^{Tornqvist} = \sum_i \left( \frac{s_{i1} + s_{i0}}{2} \right) \log \left( \frac{p_{i1}}{p_{i0}} \right)$$

- + Various “tests” that price indexes should satisfy (Diewert 93)

# Price indexes with CES

- CES ideal price index:  $P = [\sum_i b_i^\sigma p_i^{1-\sigma}]$   
accounting for tastes parameters  $b_i$  (e.g. differences in quality)  
but  $\sigma$  is not directly observed (and hard to estimate)
- Sato-Vartia price index (exact for CES!)

$$\log P^{SV} = \sum_i w_i \log \left( \frac{p_{i1}}{p_{i0}} \right) \quad \text{with: } w_i = \frac{\left( \frac{s_{i1}-s_{i0}}{\ln s_{i1}-\ln s_{i0}} \right)}{\sum_j \left( \frac{s_{j1}-s_{j0}}{\ln s_{j1}-\ln s_{j0}} \right)}$$

*Elements of proof:* with CES:  $\log s_i = \sigma \log b_i + (1 - \sigma)(\log p_i - \log P)$ .  
Summing over  $i$  with weights  $w_i$  to be determined, and taking the difference  
bw periods, we get:  $\log \left( \frac{P_1}{P_0} \right) = \sum_i w_i \log \left( \frac{p_{i1}}{p_{i0}} \right) + \frac{1}{\sigma-1} \sum_i w_i \log \left( \frac{s_{i1}}{s_{i0}} \right)$ . For  
 $\sum_i w_i \log \left( \frac{p_{i1}}{p_{i0}} \right)$  to be a price index, we need  $\sum_i w_i \log \left( \frac{s_{i1}}{s_{i0}} \right) = 0$ . In the limit  
case  $s_{i1} = s_{i0}$ , we also need  $w_i = s_i$ .

## Two other issues:

- “**Outlet bias**”:
  - We also need to account for variations in prices for the *same* good, and taking an average is not a good solution. Prices vary across outlets, consumers tend to buy in large quantities from cheap stores (e.g. Costco).
- “**New goods bias**”:
  - Price indexes above are based on comparison of prices before/after.  
With new goods: weights? prices?
  - More generally, there is a large literature aiming at quantifying the welfare gains from new goods, with various structures on the supply and demand side (see e.g. Hausman 2003, Nevo 2003)

# New goods with CES

Q: How to account for new product varieties not available before?

- Feenstra (1994) extends SV to account for extensive margin:

$$P^{SV+} = \left( \frac{\sum_{i \in \Omega_c} s_{i1}}{\sum_{i \in \Omega_c} s_{i0}} \right)^{\frac{1}{\sigma-1}} \times P^{SV}$$

Across *continuing* varieties  $\Omega_c$ , hence with  $\sum_{i \in \Omega_c} s_{i1} < 1$

- See Problem Set 5 for simple case with homogeneous products
- Application: Broda and Weinstein (2006) estimate gains from increased import varieties (1972-2001) as 2.6% of GDP

## Separability of expenditure function

- Suppose that we have two sets of goods: grocery vs. non-grocery

Q: Under which condition can we summarize the vector of prices  $p$  of grocery goods into a price index  $P_G(p)$  such that consumption in non-grocery goods only depend on non-grocery prices and  $P_G$ ?

A: If the expenditure function is separable, i.e. if we can write:

$$e(u, p, p') = \hat{e}(u, P_G(p), p')$$

where  $P_G(p)$  is a grocery price index and  $p'$  vector of non-grocery prices

- Notes:

- In this case:  $\frac{h_i}{h_j} = \frac{\frac{\partial e}{\partial p_i}}{\frac{\partial e}{\partial p_j}} = \frac{\frac{\partial P_G}{\partial p_i}}{\frac{\partial P_G}{\partial p_j}}$  for any two grocery goods  $i$  and  $j$
- Separability in expenditure is neither sufficient or necessary for separability in utility

# Plan

1. Tools

2. **Illustrations**

## Welfare analysis in practice

Problem set 3 related to Cohen et al (2016) measuring CS for Uber

- PS3 highlights issues computing total CS rather than changes in CS
- Integrability issues given Cohen et al (2016)'s price elasticity estimates

# Welfare analysis in practice

Atkin, Faber and Gonzalez (2016) as a good practical example.

- Foreign entry in the retail sector in Mexico, 2001-2014

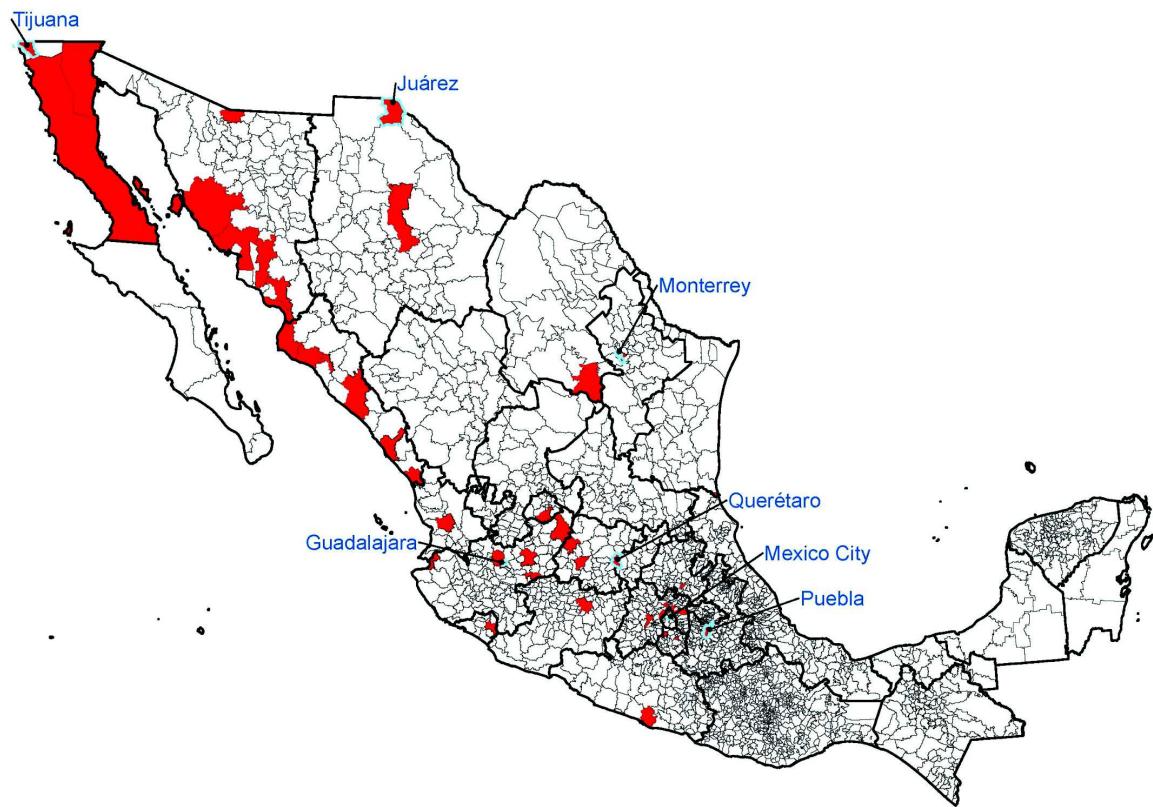
They mainly ask three questions:

- ① What is the effect of foreign retail entry on household welfare?
- ② What are the channels underlying this effect? (availability of new products, competition, entry/exit of local retailers, etc.)
- ③ Does the effect differ across the income distribution?

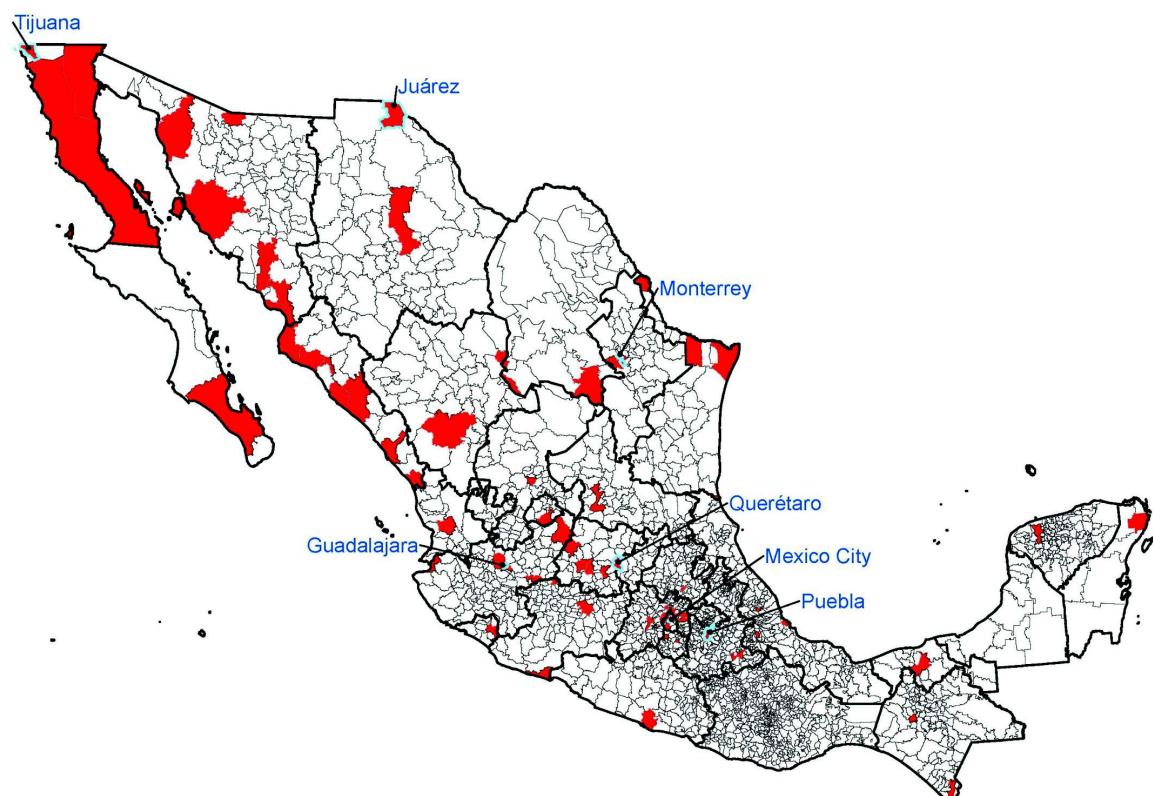
## Motivation and context

- Intense policy debates in various countries:  
e.g. India hesitates to ban foreign entry in retail
- Retail in an important sector in developing economies:  
10-15% of GDP, > 15% of employment, > 50% expenditures
- Foreign retail FDI:  
Developing country share grew from 10% to 25% in two decades
- Large expansion of foreign retail in Mexico:  
From 365 stores in 2001 to 1335 stores in 2014.

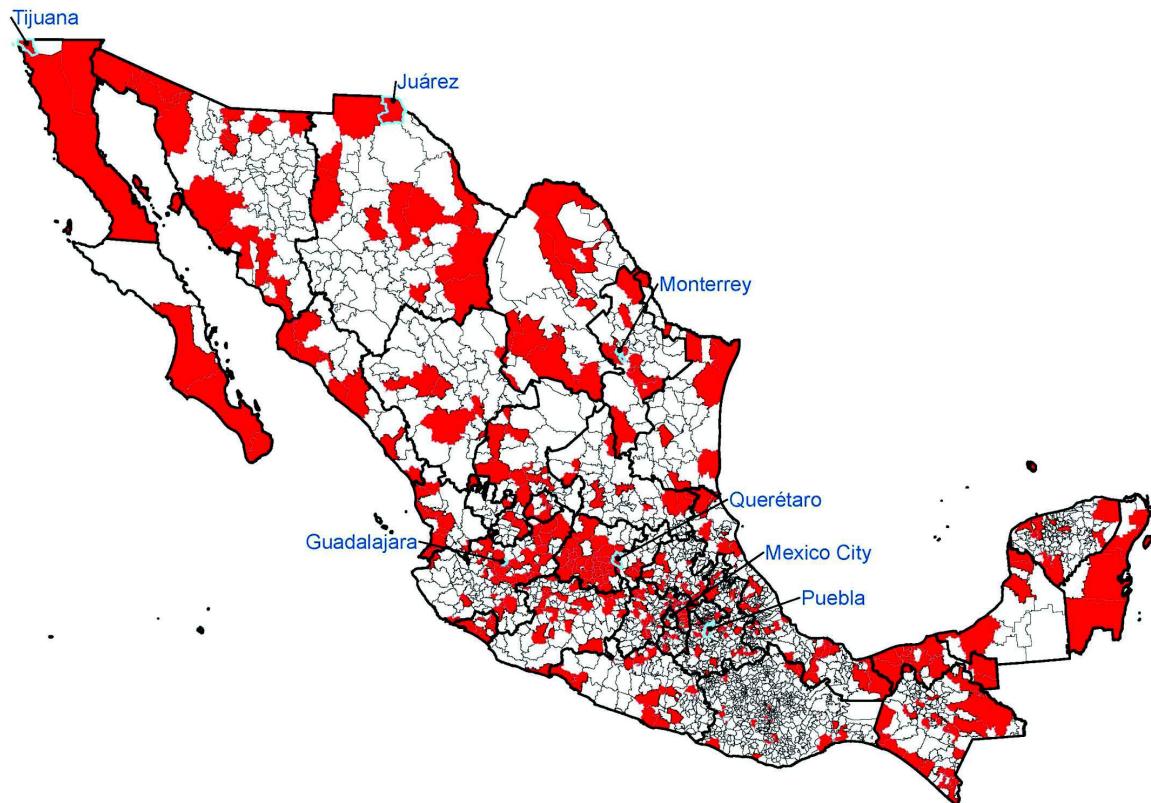
## Localization of foreign stores – 204 stores in 1995



Localization of foreign stores – 365 stores in 2001



## Localization of foreign stores – 1335 stores in 2014



Data

- Universe of supermarket locations, opening dates (2002-14)
  - Barcode/store Mexican CPI microdata (2002-14) (INEGI)
  - Household/barcode/store level Consumer Panel data (2011-14)
  - ENIGH Household survey data on budget shares at product-group/store-type level (2006-12)
  - Worker level data on income sources (2002-12)
  - Store revenues, costs: Mexican Retail Census (2003 and 08)

## How do foreign retailers differ *ex post*?

Dependent Variable:	(1) Log Price	(2) Log Price	(3) Log Number of Barcodes	(4) Log Floor Space
Foreign Store Dummy	-0.118*** (0.00913)	0.249*** (0.0160)	1.612*** (0.0671)	1.911*** (0.0416)
Municipality-By-Year FX	✓	✓	✓	✓
Municipality-By-Product-By-Month FX	✓	✓	✗	✗
Municipality-By-Barcode-By-Month FX	✓	✗	✗	✗
Observations	18,659,777	18,659,777	10,393	11,113
R-squared	0.923	0.368	0.139	0.302
Number of Municipalities	151	151	151	499

## Challenges

- Availability of consumption data (only available for later years at barcode level) calls for Paasche indexes?
- Income effect: incomes may have changed due to foreign entry
  - Approx: neglect how changes in income affects substitution
- Price effects:
  - Direct negative effect on prices?
  - Differences in quality?
  - Entry / exit of stores and product variety?

# General expression for welfare effects

$$\begin{aligned} \bullet \quad CV &= e(\mathbf{P}^1, u_h^0) - y_h^1 \\ &= \underbrace{[e(\mathbf{P}^1, u_h^0) - e(\mathbf{P}^0, u_h^0)]}_{\text{Cost of living effect (CLE)}} - \underbrace{[y_h^1 - y_h^0]}_{\text{Income effect (IE)}} \end{aligned}$$

- While effects on incomes can in principle be estimated without imposing additional structure, this is not the case for cost of living.
  - Can observe price changes of products in continuing domestic stores ( $\mathbf{P}_{dc}^1 - \mathbf{P}_{dc}^0$ ).
  - Cannot observe price changes for consumption at entering foreign retailers ( $\mathbf{P}_f^1 - \mathbf{P}_f^{0*}$ ) or exiting domestic retailers ( $\mathbf{P}_{dx}^{1*} - \mathbf{P}_{dx}^0$ ).

## A decomposition

$$\begin{aligned} CLE &= \underbrace{[e(\mathbf{P}_{dc}^1, \mathbf{P}_{dx}^{1*}, \mathbf{P}_f^1, u_h^0) - e(\mathbf{P}_{dc}^1, \mathbf{P}_{dx}^{1*}, \mathbf{P}_f^{1*}, u_h^0)]}_{\text{1: Direct effect (DE)}} + \underbrace{[e(\mathbf{P}_{dc}^1, \mathbf{P}_{dx}^{1*}, \mathbf{P}_f^{1*}, u_h^0) - e(\mathbf{P}_{dc}^0, \mathbf{P}_{dx}^{0*}, \mathbf{P}_f^{0*}, u_h^0)]}_{\text{2: Pro-competitive intensive margin (PEI)}} \\ &\quad + \underbrace{[e(\mathbf{P}_{dc}^0, \mathbf{P}_{dx}^{0*}, \mathbf{P}_f^{0*}, u_h^0) - e(\mathbf{P}_{dc}^0, \mathbf{P}_{dx}^0, \mathbf{P}_f^{0*}, u_h^0)]}_{\text{3: Pro-competitive exit margin (PEX)}} \end{aligned}$$

$$\begin{aligned} IE &= \underbrace{\sum_{i \in \{\tau, \mu\}} [l_{ih}^1 - l_{ih}^0]}_{(4) \text{ Retail labor income effect}} + \underbrace{\sum_{i \in \{\tau, \mu\}} [\pi_{ih}^1 - \pi_{ih}^{i0}]}_{(5) \text{ Retail profit effect}} \\ &\quad + \underbrace{\sum_{i \in \{\sigma\}} [(l_{ih}^1 - l_{ih}^0) + (\pi_{ih}^1 - \pi_{ih}^{i0})]}_{(6) \text{ Other income effect}} \end{aligned}$$

- Where \*'s denote unobserved prices for products in entering/exiting retailers.

## Two alternative approaches

### ① Assuming multi-tier CES preferences:

- Advantages: Exact price index, quantification of gains from new varieties
- Disadvantages: Imposing structure on consumer preferences

### ② First-order approximation:

- Advantages: Paasche index as approximation without imposing specific preferences
- Disadvantages: Holds post-entry market shares fixed, solely based on observed store price differences  
Assumes away gains from variety or shopping amenities

## Using exact approach

Use a multi-tier asymmetric CES utility function:

$$U = \prod_{g \in G} [Q_g]^{\alpha_{gh}} : \text{ Cobb-Douglas over product groups } g$$

$$Q_g = \left( \sum_{s \in S_g} \beta_{gsh} q_{gs}^{\frac{\eta_{gh}-1}{\eta_{gh}}} \right)^{\frac{\eta_{gh}}{\eta_{gh}-1}} : \text{ CES over stores } s$$

$q_{gs}$  : preferences within store-good unspecified for now

Under our multi-tier CES, the CLE becomes:

$$\bullet \frac{CLE}{e(\mathbf{P}_d^{0*}, \mathbf{P}_f^{0*}, u_h^0)} = \prod_{g \in G} \left\{ \left( \frac{\sum_{s \in S_g^{dc}} \phi_{gsh}^1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta-1}} \prod_{s \in S_g^{dc}} \left( \frac{p_{gs}^1}{p_{gs}^0} \right) \omega_{gsh} \right\}^{\alpha_{gh}} - 1$$

# Notation

g=product group, s=store, b=barcode, m=municipality, t=month

$r_{gsh}^t$ : Price index of product-specific prices  $p_{gsb}^t$

$$\phi_{gsh}^t = r_{gsh}^t q_{gsh}^t / \sum_{s \in S_g} r_{gsh}^t q_{gsh}^t$$

$$\tilde{\phi}_{gsh}^t = r_{gsh}^t q_{gsh}^t / \sum_{s \in S_g^{dc}} r_{gsh}^t q_{gsh}^t$$

$\prod_{s \in S_g^{dc}} \left( \frac{r_{gsh}^1}{r_{gsh}^0} \right)^{\omega_{gsh}}$ : Sato-Vartia price index

$$\omega_{gsh} = \left( \frac{\tilde{\phi}_{gsh}^1 - \tilde{\phi}_{gsh}^0}{\ln \tilde{\phi}_{gsh}^1 - \ln \tilde{\phi}_{gsh}^0} \right) / \sum_{s \in S_g^{dc}} \left( \frac{\tilde{\phi}_{gsh}^1 - \tilde{\phi}_{gsh}^0}{\ln \tilde{\phi}_{gsh}^1 - \ln \tilde{\phi}_{gsh}^0} \right)$$

## Using exact approach

Uses price changes and consumption basket changes to estimate  
(in particular: effect on (Stone) price index  $r_{gs}$  by store/product)

Uses preference parameters to estimate:  $\eta_{gh}$

$$\begin{aligned}
 & \frac{CV}{e(\mathbf{P}_d^0, \mathbf{P}_f^{0*}, u_h^0)} = \\
 & \underbrace{\left[ \prod_{g \in G} \left\{ \left( \frac{\sum_{s \in S_g^{dc}} \phi_{gsh}^1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta_{gh}-1}} \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} - \prod_{g \in G} \left\{ \left( \frac{1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta_{gh}-1}} \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} \right]}_{(1) \text{ Direct effect (DE)}} \\
 & + \underbrace{\left[ \prod_{g \in G} \left\{ \left( \frac{1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta_{gh}-1}} \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} - \prod_{g \in G} \left\{ \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} \right]}_{(3) \text{ Pro-competitive exit (PEX)}} + \underbrace{\left[ \prod_{g \in G} \left\{ \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} - 1 \right]}_{(2) \text{ Pro-competitive price (PEI)}} \\
 & - \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{ih}^0 \left( \frac{i_{ih}^1 - i_{ih}^0}{i_{ih}^0} \right) \right]}_{(4) \text{ Retail labor income effect}} - \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^0}{\pi_{ih}^0} \right) \right]}_{(5) \text{ Retail profit effect}} - \underbrace{\sum_{i \in \{\sigma\}} \left[ \theta_{ih}^0 \left( \frac{i_{ih}^1 - i_{ih}^0}{i_{ih}^0} \right) + \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^0}{\pi_{ih}^0} \right) \right]}_{(6) \text{ Other income effect}}
 \end{aligned}$$

# Using first-order general approach

Using Shephard's Lemma to approximate pro-competitive price effects (PP' below) and direct price effects (DE' below):

$$PP' \approx \sum_b \sum_{s \in S_b^{dc}} \left( q_{bs}^1 (p_{bs}^1 - p_{bs}^0) \right)$$

$$\frac{PP'}{e(\mathbf{P}_f^1, \mathbf{P}_{dc}^1, \mathbf{P}_{dx}^{1*}, u_h^0)} \approx \sum_b \sum_{s \in S_b^{dc}} \left( \phi_{bs}^1 \left( \frac{p_{bs}^1 - p_{bs}^0}{p_{bs}^1} \right) \right)$$

Similarly:

$$\frac{DE'}{e(\mathbf{P}_f^1, \mathbf{P}_{dc}^1, \mathbf{P}_{dx}^{1*}, u_h^0)} \approx \sum_b \sum_{s \in S_b^f} \left( \phi_{bs}^1 \left( \frac{p_{bf}^1 - p_{bds}^0}{p_{bf}^1} \right) \right)$$

# Using first-order general approach

Uses price changes to estimate

Holds ex post consumption shares constant ( $\approx$  Paasche)

$$\frac{CV}{e(\mathbf{P}_d^0, \mathbf{P}_f^{0*}, u_h^0)} \approx \underbrace{\sum_b \sum_{s \in S_b^f} \left[ \phi_{bs}^1 \left( \frac{p_{bf}^1 - p_{bds}^0}{p_{bf}^1} \right) \right]}_{(1) \text{ Direct effect } (DE)} + \underbrace{\sum_b \sum_{s \in S_b^{dc}} \left[ \phi_{bs}^1 \left( \frac{p_{bs}^1 - p_{bs}^0}{p_{bs}^1} \right) \right]}_{(2) \text{ Pro-competitive effect } (PE)}$$

$$- \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{ilh}^0 \left( \frac{l_{ih}^1 - l_{ih}^0}{l_{ih}^0} \right) \right]}_{(4) \text{ Retail labor income effect}} - \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^{i0}}{\pi_{ih}^0} \right) \right]}_{(5) \text{ Retail profit effect}} - \underbrace{\sum_{i \in \{\sigma\}} \left[ \theta_{ilh}^0 \left( \frac{l_{ih}^1 - l_{ih}^0}{l_{ih}^0} \right) + \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^{i0}}{\pi_{ih}^0} \right) \right]}_{(6) \text{ Other income effect}}$$

# What we need to estimate

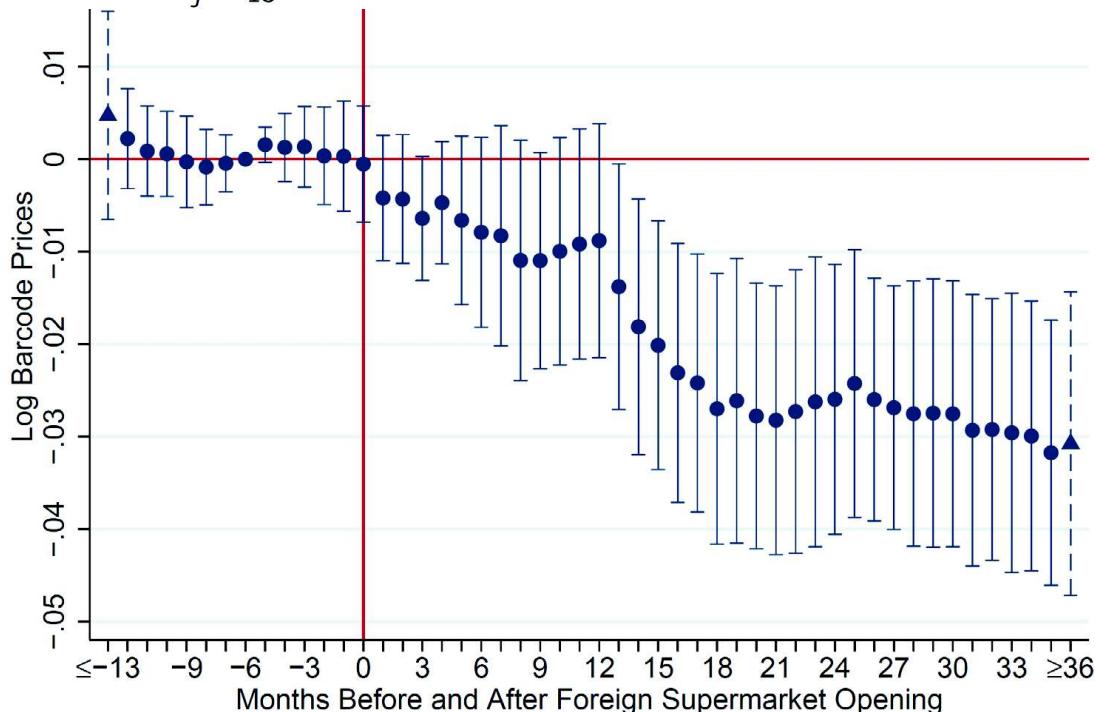
- Estimate direct effect on prices  $\frac{r_{gs}^1}{r_{gs}^0}$
- Differences in prices across stores  $p_{bf}^1 - p_{bds}^0$
- Effect on quantities
- Effect on the number of local stores
- Effect on income, by source (retail labor, retail profits, other)
- CES preferences: estimate elasticity of substitution  $\eta_{gh}$

Notation:

g=product group, s=store, b=barcode, m=municipality, t=month

## Direct effect on prices

$$\ln p_{gsbmt} = \sum_{j=-13}^{36} \beta_j I(\text{MonthsSinceEntry}_{mt} = j) + \delta_{gsbm} + \eta_t + \varepsilon_{gsbmt}$$



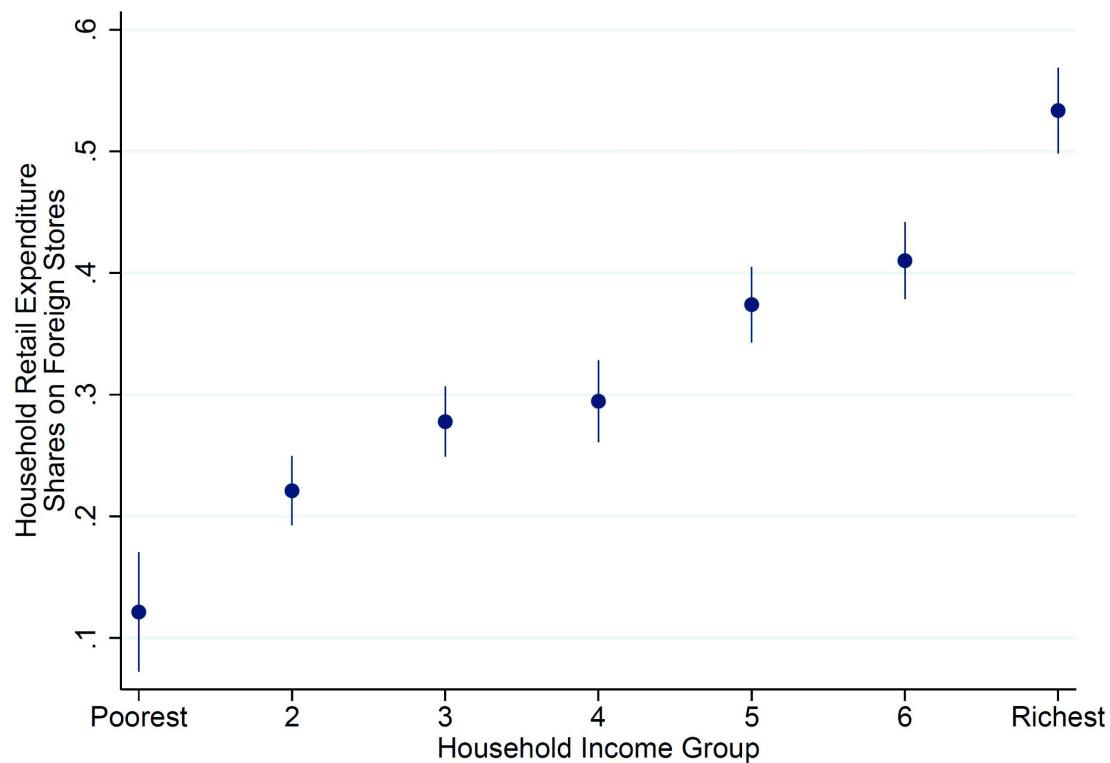
# Differences in prices across stores

(to be used for first-order approximation)

Dependent Variable:	(1) Log Price	(2) Log Price	(3) Log Price	(4) Log Price
Domestic Store	0.118*** (0.00913)			
Domestic Store X Food		0.124*** (0.00979)		
Domestic Store X Non-Food			0.0744*** (0.00765)	
Domestic Store X Traditional				0.173*** (0.00874)
Domestic Store X Modern				0.0397*** (0.0113)
Domestic Store X Food X Traditional				0.174*** (0.00942)
Domestic Store X Non-Food X Traditional				0.170*** (0.0108)
Domestic Store X Food X Modern				0.0431*** (0.0124)
Domestic Store X Non-Food X Modern				0.0189*** (0.00713)
Municipality-By-Barcode-By-Month FX	✓	✓	✓	✓
Observations	18,659,777	18,659,777	18,659,777	18,659,777
R-squared	0.923	0.923	0.923	0.923
Number of Municipalities	151	151	151	151



## Ex post foreign retail share by income group



# Effect on store exit

$$d\ln(N\_Establishments_m^{08-03}) = \beta_1 ForeignEntry_m^{08-03} + \beta_2 ForeignEntry_m^{Pre 04} + \gamma X_m + \varepsilon_m$$

Panel A: Unweighted regressions	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
	ΔLog(Number Stores) 2003-08 Traditional Store Formats				ΔLog(Number Stores) 2003-08 Modern Store Formats			
ΔForeign Entry 2003-2008	-0.019 (0.014)	-0.023 (0.014)	-0.025* (0.014)	-0.024* (0.014)	0.0088 (0.067)	-0.0065 (0.068)	-0.036 (0.069)	-0.035 (0.069)
Foreign Entry Pre 2003	-0.055*** (0.013)	-0.057*** (0.015)	-0.035** (0.015)	-0.032** (0.016)	0.20*** (0.053)	0.16*** (0.058)	0.17*** (0.060)	0.17*** (0.062)
Δlog(Public Expenditures)			0.12*** (0.028)	0.12*** (0.028)			0.37*** (0.12)	0.38*** (0.12)
Δlog(GDP per Capita)				-0.020 (0.014)				-0.012 (0.066)
Geographical Region FX	✗	✓	✓	✓	✗	✓	✓	✓
Municipality Size FX	✗	✓	✓	✓	✗	✓	✓	✓
Observations	608	608	564	564	608	608	564	564
R-squared	0.022	0.056	0.107	0.110	0.015	0.085	0.107	0.107
Median Stores/Municipality	2088	2088	2088	2088	33.5	33.5	33.5	33.5

# Effect on income

No effect on average income (see paper), but some heterogeneity:

$$\ln(Income)_{jimt} = \sum_i \beta_i (ForeignEntry_{mt} \times Occupation_i) + \gamma X_{jimt} + \delta_{mt} + \eta_{im} + \theta_{it} + \varepsilon_{jimt}$$

Dependent Variable:	(1) Log (Monthly Income)	(2) Log (Monthly Income)	(3) Log (Monthly Income)	(4) Log (Employment)	(5) Log (Employment)	(6) Log (Employment)
Foreign Entry X Modern Retail Workers	-0.000278 (0.0192)	-0.0348* (0.0204)	-0.0278 (0.0212)	-0.00396 (0.0653)	0.0369 (0.0714)	0.0392 (0.0561)
Foreign Entry X Traditional Retail Workers	-0.0356* (0.0199)	-0.0571*** (0.0216)	-0.0592** (0.0240)	-0.104* (0.0531)	-0.0942 (0.0571)	-0.113** (0.0552)
Foreign Entry X Agriculture	0.0265 (0.0264)	0.0218 (0.0311)	0.0202 (0.0307)	-0.0597 (0.0809)	-0.0285 (0.101)	-0.00811 (0.106)
Foreign Entry X Manufacturing	-0.00513 (0.0174)	-0.00612 (0.0186)	0.0117 (0.0187)	-0.166*** (0.0379)	0.00572 (0.0368)	-0.0166 (0.0380)
Person Controls	✓	✓	✓	✗	✗	✗
Municipality-by-Quarter FX	✓	✓	✓	✓	✓	✓
Municipality-by-Group Fixed Effects	✓	✓	✓	✓	✓	✓
Group-by-Quarter FX	✗	✓	✓	✗	✓	✓
State-by-Group Time Trends	✗	✗	✓	✗	✗	✓
Observations	3,878,561	3,878,561	3,878,561	47,666	47,666	47,666
R-squared	0.340	0.340	0.341	0.963	0.965	0.967
Number of Individuals	1,455,911	1,455,911	1,455,911	1,455,911	1,455,911	1,455,911
Number of Municipality-by-Quarter Cells	8,574	8,574	8,574	8,574	8,574	8,574
Number of State-by-Group Time Trends	160	160	160	160	160	160
Number of Municipality Clusters	273	273	273	273	273	273

# Using exact approach

Uses price changes and consumption basket changes to estimate

Uses preference parameters to estimate:  $\eta_{gh}$

$$\frac{CV}{e(\mathbf{P}_d^0, \mathbf{P}_f^{0*}, u_h^0)} =$$

$$\underbrace{\left[ \prod_{g \in G} \left\{ \left( \frac{\sum_{s \in S_g^{dc}} \phi_{gsh}^1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta_{gh}-1}} \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} - \prod_{g \in G} \left\{ \left( \frac{1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta_{gh}-1}} \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} \right]}_{(1) \text{ Direct effect (DE)}}$$

$$+ \underbrace{\left[ \prod_{g \in G} \left\{ \left( \frac{1}{\sum_{s \in S_g^{dc}} \phi_{gsh}^0} \right)^{\frac{1}{\eta_{gh}-1}} \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} - \prod_{g \in G} \left\{ \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} \right]}_{(3) \text{ Pro-competitive exit (PEX)}} + \underbrace{\left[ \prod_{g \in G} \left\{ \prod_{s \in S_g^{dc}} \left( \frac{r_{gs}^1}{r_{gs}^0} \right)^{\omega_{gsh}} \right\}^{\alpha_{gh}} - 1 \right]}_{(2) \text{ Pro-competitive price (PEI)}}$$

$$- \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{ih}^0 \left( \frac{i_{ih}^1 - i_{ih}^0}{i_{ih}^0} \right) \right]}_{(4) \text{ Retail labor income effect}} - \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^0}{\pi_{ih}^0} \right) \right]}_{(5) \text{ Retail profit effect}} - \underbrace{\sum_{i \in \{\sigma\}} \left[ \theta_{ih}^0 \left( \frac{i_{ih}^1 - i_{ih}^0}{i_{ih}^0} \right) + \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^0}{\pi_{ih}^0} \right) \right]}_{(6) \text{ Other income effect}}$$

## Price elasticity of demand

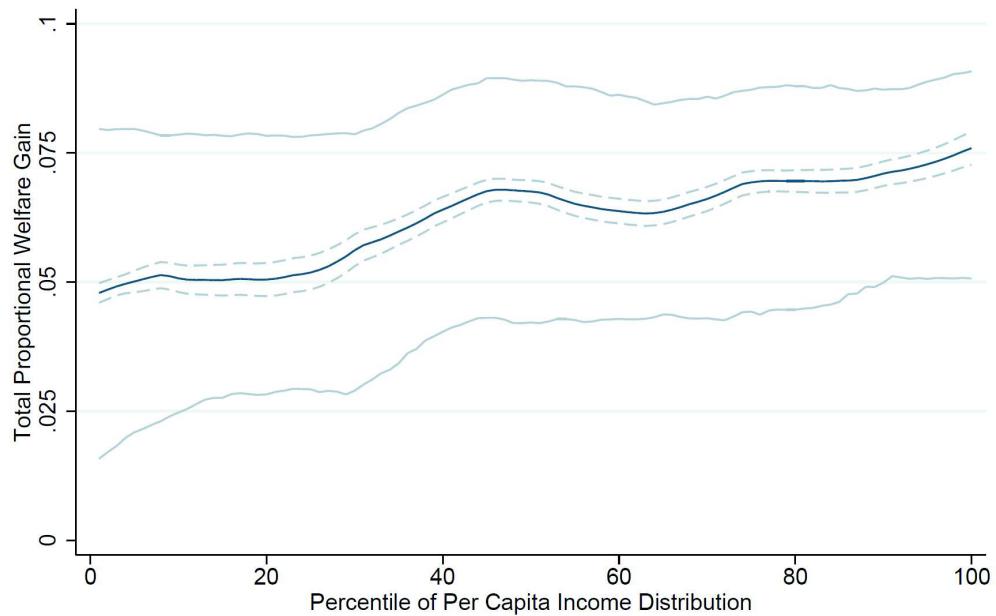
It's a challenge to get large enough elasticities  $\eta_{gh}$ :

$$\ln \phi_{gshmt} = (1 - \eta_{gh}) \ln r_{gshmt} - (1 - \eta_{gh}) \ln c_{ghmt} + \eta_{gh} \ln \beta_{gshmt}$$

Panel A: Average Coefficient Estimates	(1)	(3)	(5)	(7)	(9)
	Average Prices	Average Prices	Average Prices	Average Prices	Average Prices
	OLS	National IV	Regional IV	National IV	Regional IV
Dependent Variable: Log Budget Shares (Phi)					
Log(Store Price Index)	0.214*** (0.006)	-1.341*** (0.145)	-1.856*** (0.608)	-2.648*** (0.338)	-3.362*** (1.038)
Product Group-by-Income Group-by-Municipality-by-Quarter FX	✓	✓	✓	✓	✓
Retailer-by-Product Group-by-Quarter FX	✓	✓	✓	✓	✓
Retailer-by-Municipality FX	✓	✓	✓	✓	✓
Retailer-by-Municipality-by-Quarter FX	✗	✗	✗	✓	✓
Retailer-by-Municipality-by-Product Group FX	✗	✗	✗	✗	✗
Observations	304,885	304,885	297,624	304,885	297,624
First-Stage F-Statistic	184.884	14.833	87.951	15.52	

# Welfare gains with CES

## Distribution of the Gains from Retail FDI



- Large and significant average gains from foreign entry.
- Gains are regressive (richest gain approximately 1.5 times as much).

# Welfare gains with CES

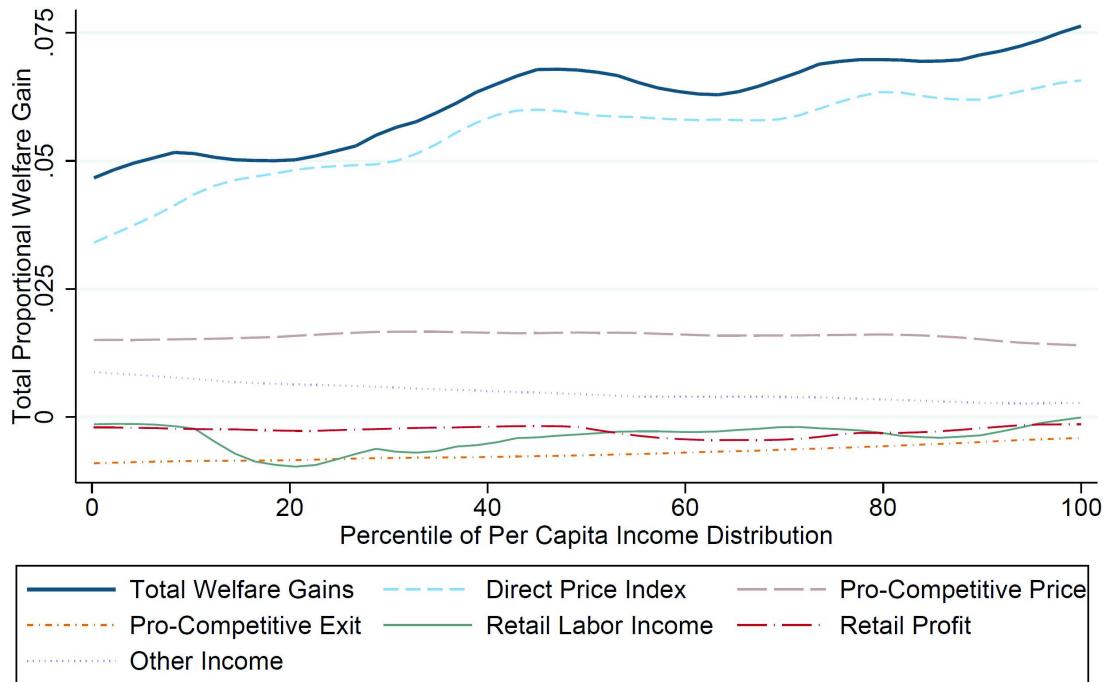
Decomposition of the 6.2% average welfare gains:

- most of the gains from cost of living effect (CLE)
- 3/4 direct effect (lower prices, higher quality at foreign stores)
- 1/4 driven by pro-competitive effects on domestic stores

Dependent Variable:	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Total Effect	Direct Price Index Effect	Pro-Comp Price Effect	Pro-Comp Exit	Labor Income Effect	Profit Effect	Other Income Effect
Average Effect	0.0621*** (0.0104)	0.0551*** (0.0006)	0.0158*** (0.0050)	-0.00705 (0.0053)	-0.00397** (0.0020)	-0.00269** (0.0013)	0.0049 (0.0078)
Max	0.730	0.177	0.055	0.000	0.692	0.000	0.020
Min	-0.986	0.000	0.000	-0.014	-1.000	-1.000	0.000
Proportion Negative	0.0203	0	0	0.999	0.0736	0.0581	0
Observations (Households)	12,293	12,293	12,293	12,293	12,293	12,293	12,293
Number of Municipality Clusters	240	240	240	240	240	240	240

# Welfare gains with CES

Percentile of Per Capita Income Distribution



## Using first-order general approach

Uses price changes to estimate

Holds ex post consumption shares constant ( $\approx$  Paasche)

$$\frac{CV}{e(\mathbf{P}_d^0, \mathbf{P}_f^{0*}, u_h^0)} \approx \underbrace{\sum_b \sum_{s \in S_b^f} \left[ \phi_{bsh}^1 \left( \frac{p_{bf}^1 - p_{bds}^0}{p_{bf}^1} \right) \right]}_{(1) \text{ Direct effect (DE)}} + \underbrace{\sum_b \sum_{s \in S_b^{dc}} \left[ \phi_{bsh}^1 \left( \frac{p_{bs}^1 - p_{bs}^0}{p_{bs}^1} \right) \right]}_{(2) \text{ Pro-competitive effect (PE)}}$$

$$- \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{ih}^0 \left( \frac{l_{ih}^1 - l_{ih}^0}{l_{ih}^0} \right) \right]}_{(4) \text{ Retail labor income effect}} - \underbrace{\sum_{i \in \{\tau, \mu\}} \left[ \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^{i0}}{\pi_{ih}^0} \right) \right]}_{(5) \text{ Retail profit effect}} - \underbrace{\sum_{i \in \{\sigma\}} \left[ \theta_{ih}^0 \left( \frac{l_{ih}^1 - l_{ih}^0}{l_{ih}^0} \right) + \theta_{i\pi h}^0 \left( \frac{\pi_{ih}^1 - \pi_{ih}^{i0}}{\pi_{ih}^0} \right) \right]}_{(6) \text{ Other income effect}}$$

## Lower estimated gains with first-order approximation

- No effect of exit (using ex post consumption shares)
- Smaller direct effect (neglects quality  $\neq$  bw domestic vs foreign stores)
- Smaller pro-competitive effects (neglects quality upgrading)

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Dependent Variable:	Total Effect	Direct Price Index Effect	Pro-Comp Price Effect	Pro-Comp Exit	Labor Income Effect	Profit Effect	Other Income Effect
Average Effect	0.0621*** (0.0104)	0.0551*** (0.0006)	0.0158*** (0.0050)	-0.00705 (0.0053)	-0.00397** (0.0020)	-0.00269** (0.0013)	0.0049 (0.0078)
	(8)	(9)	(10)	(11)	(12)	(13)	(14)
Dependent Variable:	Total Effect	Direct Price Index Effect	Pro-Comp Price Effect	Pro-Comp Exit	Labor Income Effect	Profit Effect	Other Income Effect
Average Effect	0.0295*** (0.0093)	0.0204*** (0.0014)	0.0109*** (0.0037)	0 (0.0000)	-0.00397** (0.0020)	-0.00269** (0.0013)	0.0049 (0.0078)
Max	0.715	0.060	0.031	0.000	0.692	0.000	0.020
Min	-0.995	0.000	0.000	0.000	-1.000	-1.000	0.000
Proportion Negative	0.0527	0	0	0	0.0736	0.0581	0
Observations (Households)	12,293	12,293	12,293	12,293	12,293	12,293	12,293
Number of Municipality Clusters	240	240	240	240	240	240	240

## Concluding remarks

- Large positive effects of foreign entry in retail sector (6.2% gains on average for Mexican households)
- Gains 50% larger for rich consumers (see paper for decompositions of these differences in gains)
- Mostly driven by effects on cost of living  
Small effects on income, affects only a minority
- Quality of stores and products matter quantitatively:  
important to account for it (e.g. with CES exact price indexes)