

# Reference Pricing and Consumption Inequality

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# Motivation

- ▶ In the status quo, firms **sustain** large cross-country price gaps
  - ▶ Household heterogeneity shapes markups across products of varying qualities
  - ▶ I argue that across space, household heterogeneity can generate heterogeneous markups for **identical** products
  - ▶ Price gaps reveal differences in **demand composition** across markets

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  - ▶ I argue that across space, household heterogeneity can generate heterogeneous markups for **identical products**
  - ▶ Price gaps reveal differences in **demand composition** across markets
- ▶ In 2015, the EU proposed the Digital Single Market Strategy
  - ▶ Removes virtual borders across Europe
  - ▶ Requires retailers to **charge identical prices** to all EU customers
- ▶ Raises a natural question of welfare effects of such policies across space
  - ▶ I identify the **winners and losers** of the policy if it were implemented globally

# This Paper

- ▶ I document a middle ground between law of one price and full pricing-to-market in digital goods
  - ▶ Show **imperfect geo-blocking** enables price discrimination in wealthy markets
- ▶ Develop a model that embeds the **endogenous location choices** of heterogeneous customers with imperfect enforcement by the firm
  - ▶ Some customers engage in “risky arbitrage”
  - ▶ Prices increase in poorer countries
  - ▶ The policy would **reallocates surplus** from low income countries to high income countries
- ▶ Estimate the model on the video game market
  - ▶ Allows me to abstract away from differences in costs or quality
  - ▶ Under reasonable parameters, the imperfect enforcement is optimal for the firm
  - ▶ Eventually: compare prices, profits, and consumer surplus under different pricing regimes

Additional Examples

# Literature Review

- ▶ Violations of Law of One Price
  - ▶ Simonovska (2015), Crucini and Yilmazkuday (2014), Fajgelbaum et al (2011)
  - ▶ **Contribution:** in digital settings, cheaper markets enable price discrimination
- ▶ Uniform pricing and household sorting
  - ▶ DellaVigna Gentzkow (2019), Bils Klenow (2001), Jaimovich, Rebelo, Wong, and Zhang (2019)
  - ▶ **Contribution:** firms use low prices in smaller markets to segment households within the same country
- ▶ Reference Pricing
  - ▶ Dubois, Gandhi, and Vasserman (2022), Danzon and Chao (2000), Jensen (2007)
  - ▶ **Contribution:** cheaper online markets can act as reference prices
- ▶ Allocative Effects of Exchange Rate Shocks
  - ▶ Engel (2006), Drenik and Perez (2021), Cravino (2018), Gopinath et al (2011)
  - ▶ **Contribution:** exchange rate shocks change *where* goods are purchased

# Empirical Application

# Why Video Games?

- ▶ Focusing on video games allows me to rule out several traditional explanations of price variation across space:
  - ▶ No transport costs
    - ▶ Shuts down transport costs as a source of marginal cost heterogeneity
  - ▶ Products are identical across markets
    - ▶ Rules out quality differences across space
  - ▶ Goods cannot be resold across markets
    - ▶ Rules out arbitrage across customers in different locations

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- ▶ The video game market is an *ideal setting* Market Size

# The Video Game Market

- ▶ Video games are the **largest global digital media market**
- ▶ Steam is the largest PC video game retailer in the world, holding a **75% market share**
  - ▶ “Amazon” of video games
- ▶ Steam operates in many countries, including pricing in over 40 currencies

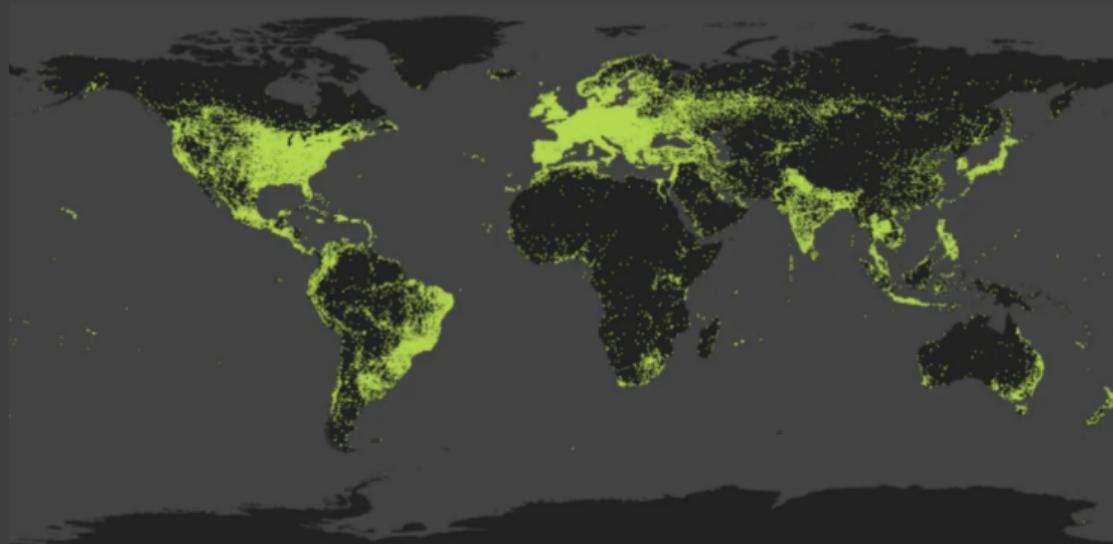


Figure: Map of Steam Users (2016)

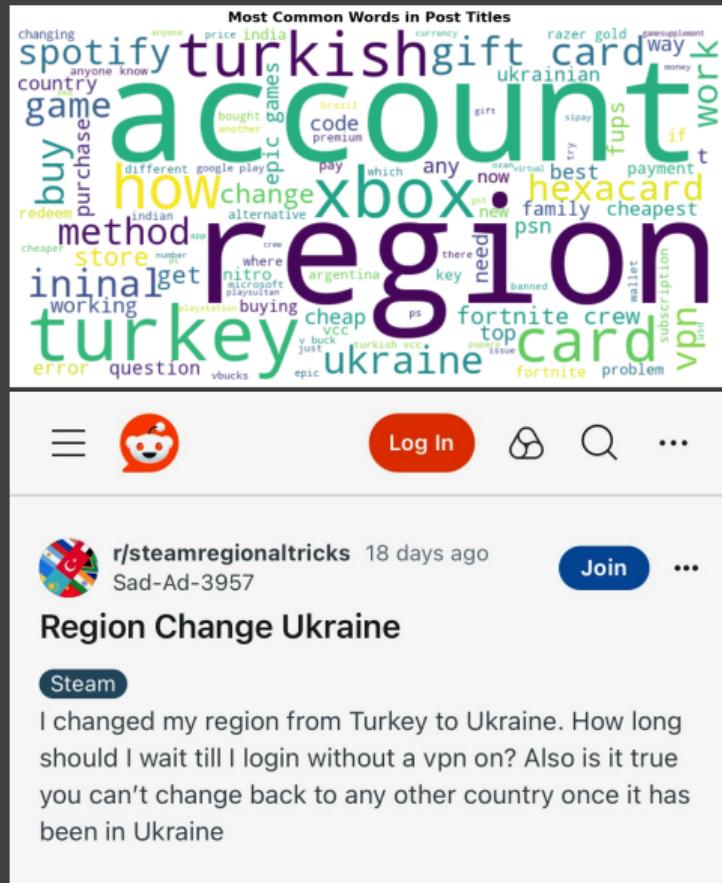
# Cross-Country Price Variation



CURRENCY	CURRENT PRICE	CONVERTED PRICE	
U.S. Dollar	\$59.99	\$59.99	
British Pound	£49.99	\$62.94	+4.92%
Euro	59,99€	\$64.22	+7.06%
Russian Ruble	1999 ₽	\$32.37	-46.04%
Brazilian Real	R\$ 199,99	\$41.05	-31.56%
Japanese Yen	¥ 6990	\$52.70	-12.15%
Indonesian Rupiah	Rp 699999	\$48.39	-19.33%
Malaysian Ringgit	RM219.00	\$49.82	-16.95%
Philippine Peso	₱2599.00	\$49.20	-17.99%
Singapore Dollar	S\$69.00	\$50.22	-16.28%
Thai Baht	฿1799.00	\$52.31	-12.80%
Vietnamese Dong	990000đ	\$42.68	-28.84%
South Korean Won	₩ 68000	\$52.59	-12.33%
Turkish Lira	₺249,00	\$14.78	-75.35%
Ukrainian Hryvnia	8998	\$30.42	-49.29%
Mexican Peso	Mex\$ 1299.00	\$66.33	+10.58%
Canadian Dollar	CDN\$ 79.99	\$63.82	+6.40%
Australian Dollar	A\$ 89.95	\$65.04	+8.43%
New Zealand Dollar	NZ\$ 99.99	\$64.89	+8.18%
Norwegian Krone	549.00 kr	\$58.00	-3.31%
Polish Złoty	199,00 zł	\$46.54	-22.42%
Swiss Franc	CHF 69.99	\$71.97	+19.98%
Chinese Yuan	¥ 298	\$44.67	-25.53%
Indian Rupee	₹ 2999	\$38.61	-35.64%
Chilean Peso	CLP\$ 39999	\$48.42	-19.28%
Peruvian Sol	S/199.00	\$53.20	-11.32%
Colombian Peso	COL\$ 199000	\$52.62	-12.28%
South African Rand	R 799.00	\$51.91	-13.46%
Hong Kong Dollar	HK\$ 399.00	\$50.85	-15.23%
Taiwan Dollar	NT\$ 1599	\$54.31	-9.46%
Saudi Riyal	229.00 SR	\$61.04	+1.75%
U.A.E. Dirham	229.00 AED	\$62.34	+3.93%
Argentine Peso	ARS\$ 2199,00	\$18.13	-69.77%
Israeli New Shekel	₪269.00	\$80.43	+34.08%
Kazakhstani Tenge	11499T	\$26.64	-55.58%

## Text Data

- ▶ Steam regional tricks subreddit
- ▶ Users discuss the best ways to exploit cross-country price variation on Steam
- ▶ Share payment methods, VPNs, etc
- ▶ Discuss consequences if detected by the firm



# Data Sources

- Daily Game price histories scraped from SteamDB in various currencies
  - Currencies: USD, Euro, Turkish Lira, Argentinian Peso, Brazilian Real, Colombian Peso, Japanese Yen, Uruguayan Peso, Chilean Peso, British Pound, Israeli New Shekel, and the Chinese Yuan
  - Characteristics: International release dates, developer, genre, AAA status

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  - ▶ I observe daily, country-level purchases of each game for large markets
  - ▶ I observe aggregate purchases over multiple smaller markets (e.g. Argentina)

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- ▶ Household Consumption Survey from Argentina
  - ▶ includes demographic information and expenditures
  - ▶ explicitly asks about video game purchases
- ▶ Currency data from FRED

# Model

# Home Customer Location Choice

- ▶ Customers choose a purchase location
- ▶ Focus on two countries: home and foreign
  - ▶ Foreign market customers *always* choose to purchase in foreign
  - ▶ Exposition of the household block focuses on home customers
- ▶ Each period, customers observe a **global menu of prices**, firm's strategy  $\pi$ , and exchange rates ( $E_t$ )

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- ▶ Each period, customers observe a **global menu of prices**, firm's strategy  $\pi$ , and exchange rates ( $E_t$ )
- ▶ Customers of type  $(\theta, L)$  vary along 4 exogenous attributes,  $\theta$ , and 1 state,  $L$ :
  - ▶  $c_i$ : physical location (home or foreign)
  - ▶  $y_i$ : income
  - ▶  $\alpha_i$ : preferences over video games
  - ▶  $\tau_i$ : lump-sum hassle cost to access foreign market prices
  - ▶  $L_{i,t}$ : library size that tracks the number of previously purchased goods

## “Risky” Arbitrage

- Customers earn per-period utility based on their purchase choice:

$$r(L, a; \theta) = u(c_a) + \alpha_i L$$

$$c_0 = y_i$$

$$c_H = y_i - p_H$$

$$c_F = y_i - (E_t p_F + \tau_i)$$

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- Purchases at home ( $H$ ) or in foreign ( $F$ ) add to the customer's library size
- Firms punish customers for purchasing in foreign with probability  $\pi$  by revoking access to the users' library based on Steam's terms and conditions
- The library law of motion is given by:

$$L'(L, a) = \begin{cases} L, & \text{if } a = 0 \\ L + 1, & \text{if } a = H \\ L + 1, & \text{if } a = F \text{ with probability } 1 - \pi \\ 0, & \text{if } a = F \text{ with probability } \pi \end{cases}$$

## Customer Location Choice

- A type  $\theta$  customer with existing library of size  $L$  chooses a location from which to purchase a good to solve:

$$V_\theta(L) = \max_a \{u(c_a) + \alpha L + \beta \mathbb{E} [V_\theta(L'; L, a)]\}$$

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- $\text{pol}_\theta(L)$  is the purchase location policy by a customer of type  $\theta$
- As the library size  $L$  grows, the shadow cost of purchasing in foreign rises
  - Life cycle pattern of initially buying in foreign and eventually switching to home
  - Customers exit the market with hazard rate  $s$  and new entrants are drawn from distribution  $\nu$  Hazard Rate Estimation
  - Changes to exchange rates break this monotonicity in library size by changing relative prices

## Demand

- Demand in the home location is given by integrating over all of the consumer types  $\theta$  and library sizes  $L$  for those purchasing at home

$$D_H(p_H, p_F, \pi, G) = \int_L \int_{\theta} \mathbb{1}\{\text{pol}_{\theta}(L) = H\} dG(\theta, L)$$

- Demand in the foreign location is given by:

$$D_F(p_H, p_F, \pi, G) = \int_L \int_{\theta} \mathbb{1}\{\text{pol}_{\theta}(L) = F\} dG(\theta, L)$$

- Includes foreign customers and arbitrageurs

## Firm Problem

- ▶ Firm selects  $(p_H, p_F, \pi)$  to maximize profits subject to a convex punishment strategy cost for implementing  $\pi$
- ▶ The firm solves:

$$\max \quad p_H D_H + p_F D_F - C(\pi; k_1, k_2)$$

- ▶ where  $C(\pi; k_1, k_2)$  is a convex punishment strategy cost:

$$C(\pi; k_1, k_2) = k_1 \pi + k_2 \pi^2$$

- ▶ taking as given induced customer demand Timing

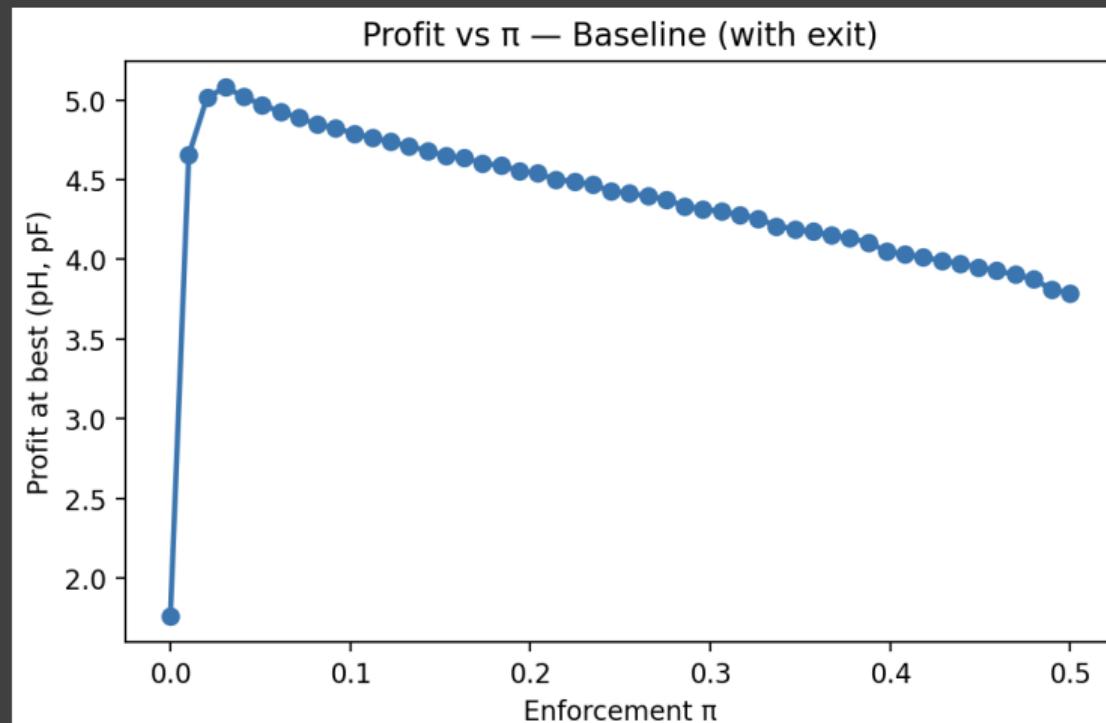
# Global Prices and Consumer Surplus

- The firm solves for a global price vector and punishment strategy via a fixed point that trades off: **Equilibrium Concept**
  - Market expansion effect **from the price sensitive group**: home customers attracted by the lower prices
  - Decreased revenue **from the savvy group**: home customers that instead purchase in foreign
  - Decreased revenue from customers located in the foreign market
- Both **savvy** and **price sensitive** customers gain from imperfect enforcement
- Consumer surplus of foreign customers *decreases* relative to fully segmented markets

# Estimation

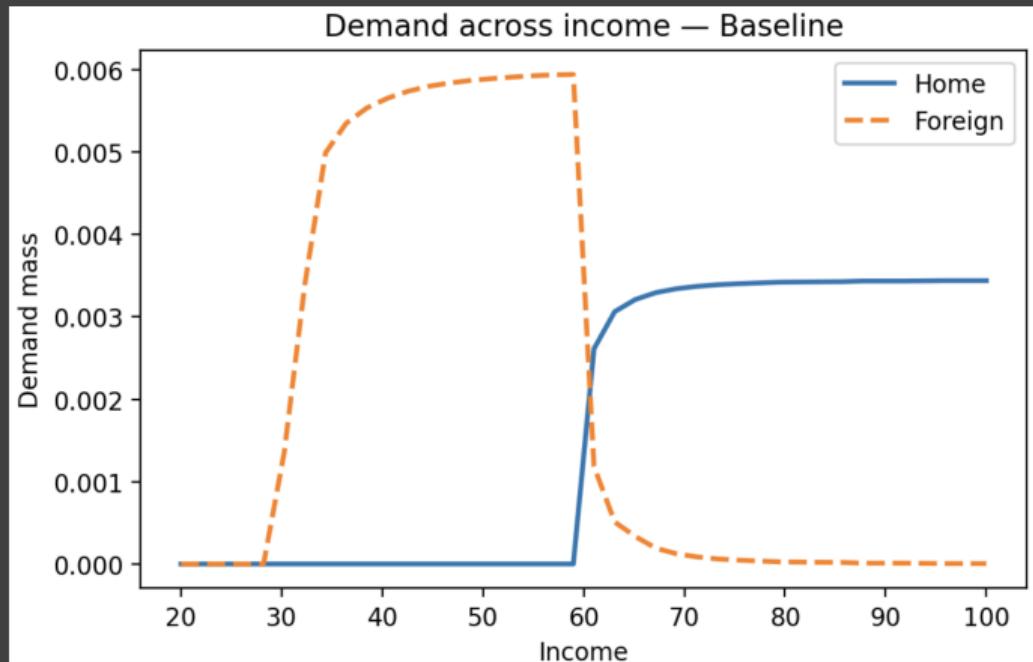
## Model Simulations

- At each level of punishment probability  $\pi$ , find the highest profit by varying  $p_F$  and  $p_H$  Calibrated Parameters
- At the optimal  $\pi$ , the recovered prices are 59.58 for home and 15.00 for foreign



## Demand Across Income Levels

- ▶ For prices of 59.58 and 15.00,  $\pi^*$  of .03 efficiently sorts home households
- ▶ Lower income households largely purchase in the foreign market:
  - ▶ Higher marginal utility of non-video game consumption
  - ▶ Smaller accumulated library sizes



## Estimation Strategy: Simulated Method of Moments

- ▶ I assume that observed prices ( $p_H, p_F$ ) are profit maximizing given consumer heterogeneity, exchange-rate risk, and library dynamics

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- ▶ I assume that observed prices  $(p_H, p_F)$  are profit maximizing given consumer heterogeneity, exchange-rate risk, and library dynamics
- ▶ Outer loop: choose candidate parameter vector  $\theta$ , firm cost vector  $k_1, k_2$
- ▶ Inner loop: solve the firm–consumer environment given  $\theta, k_1, k_2$ :
  - ▶ Consumer side: solve value functions and policies; aggregate to demand  $D_H(\theta), D_F(\theta)$
  - ▶ Firm side: given induced demand, solve for optimal  $(p_H(\theta), p_F(\theta), \pi(\theta))$

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  - ▶ Firm side: given induced demand, solve for optimal  $(p_H(\theta), p_F(\theta), \pi(\theta))$
- ▶ Simulate model moments  $m^{\text{model}}(\theta)$  to match data moments  $m^{\text{data}}$
- ▶ Minimize  $Q(\theta) = [m^{\text{data}} - m^{\text{model}}(\theta)]^\top W [m^{\text{data}} - m^{\text{model}}(\theta)]$  Moments

# From Model to Data Moments

- ▶ Market shares: pin down income and switching cost distributions by capturing substitution patterns
  - ▶ Baseline home vs. foreign shares
  - ▶ Response to exchange rate shocks
  - ▶ Lags in switching
- ▶ Text data moments (Reddit)
  - ▶ Reflect the product of enforcement intensity  $\pi$  and switching volume
  - ▶ Frequency of punishment discussions
  - ▶ Correlation with large shocks
  - ▶ Baseline frequency / false positives
- ▶ Market shares identify the volume of switches, while Reddit data identifies how often switches trigger punishment

# Counterfactuals

- ▶ Compare the prices and profits under imperfectly enforced digital boundaries to two benchmarks
  - ▶ Digital Single Market regime where law of one price must hold
  - ▶ Segmented Market regime where firms can price to each market individually
- ▶ Preliminary simulation results yield regime-dependent prices:
  - ▶ Imperfect enforcement:  $p_H = 59.58$ ,  $p_F = 15.00$
  - ▶ Uniform prices:  $p = 47.917$
  - ▶ Segmented markets:  $p_H = 53.75$ ,  $p_F = 11.25$
  - ▶ Unified markets raise prices by > 300% in poor countries and lower prices by 20% in rich countries
- ▶ Zero hassle cost for customers Zero hassle cost
- ▶ Zero enforcement cost for the firm Zero Enforcement Cost

# Conclusion

- ▶ I document a new empirical pattern that firms allow price-sensitive consumers to access lower foreign market prices
- ▶ I develop and estimate a model to rationalize these new empirical findings
- ▶ To estimate the welfare effects of uniform price mandates, I consider prices under counterfactual pricing regimes (in progress)
  - ▶ Fully segmented market benchmark
  - ▶ Digital single market benchmark

# Appendix

# EU's Digital Single Market

- The EU's Single Digital Market prohibits geoblocking to ensure equal access to digital goods.
- Cross-country price differences for video games remain substantial, despite regulatory efforts.
- Increased competition and access to consumer goods are key goals of the Single Digital Market.

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# Key Mechanism: Endogenous Purchase Location Choice

- ▶ Price sensitive customers can **change their digital market** and access cheaper prices
- ▶ Cross-country price gaps reveal information about the underlying demand curves of customers that take each action
- ▶ Similar features show up in Netflix, Spotify, other digital goods

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FR	91.9	✓ 81,29 CHF	<a href="#">Open</a>
SE	80.5	✓ 834,00 SEK	<a href="#">Open</a>
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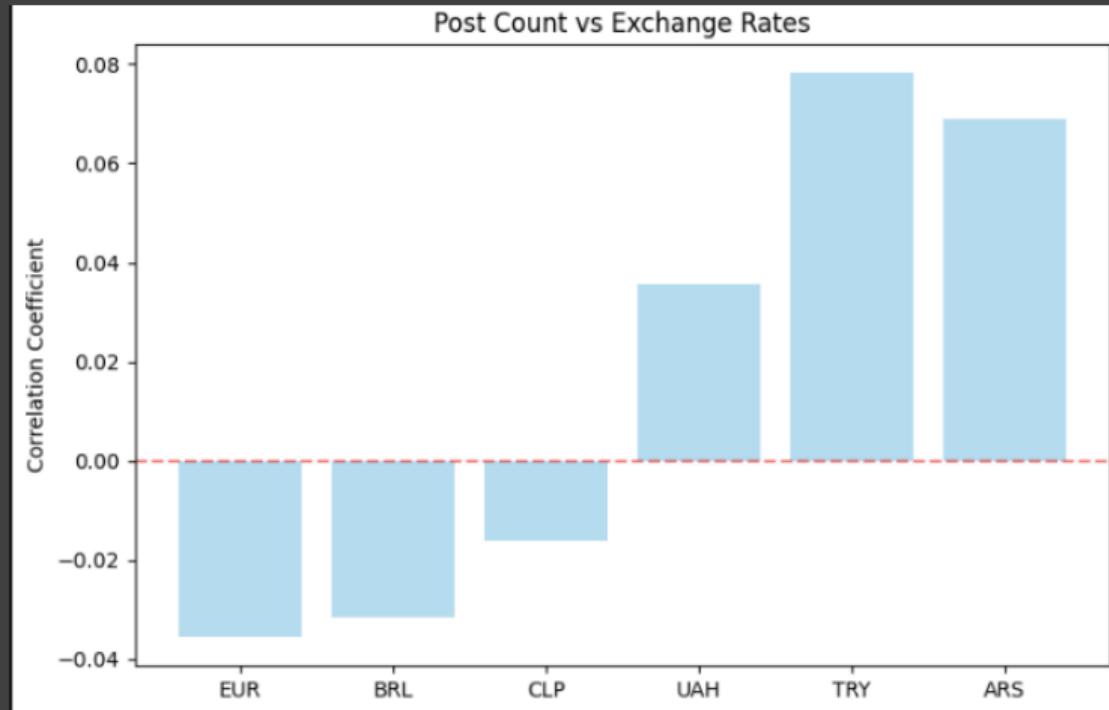
(All results converted to USD)

[Rate](#) [F.A.Q.](#) [>SHOP<](#)

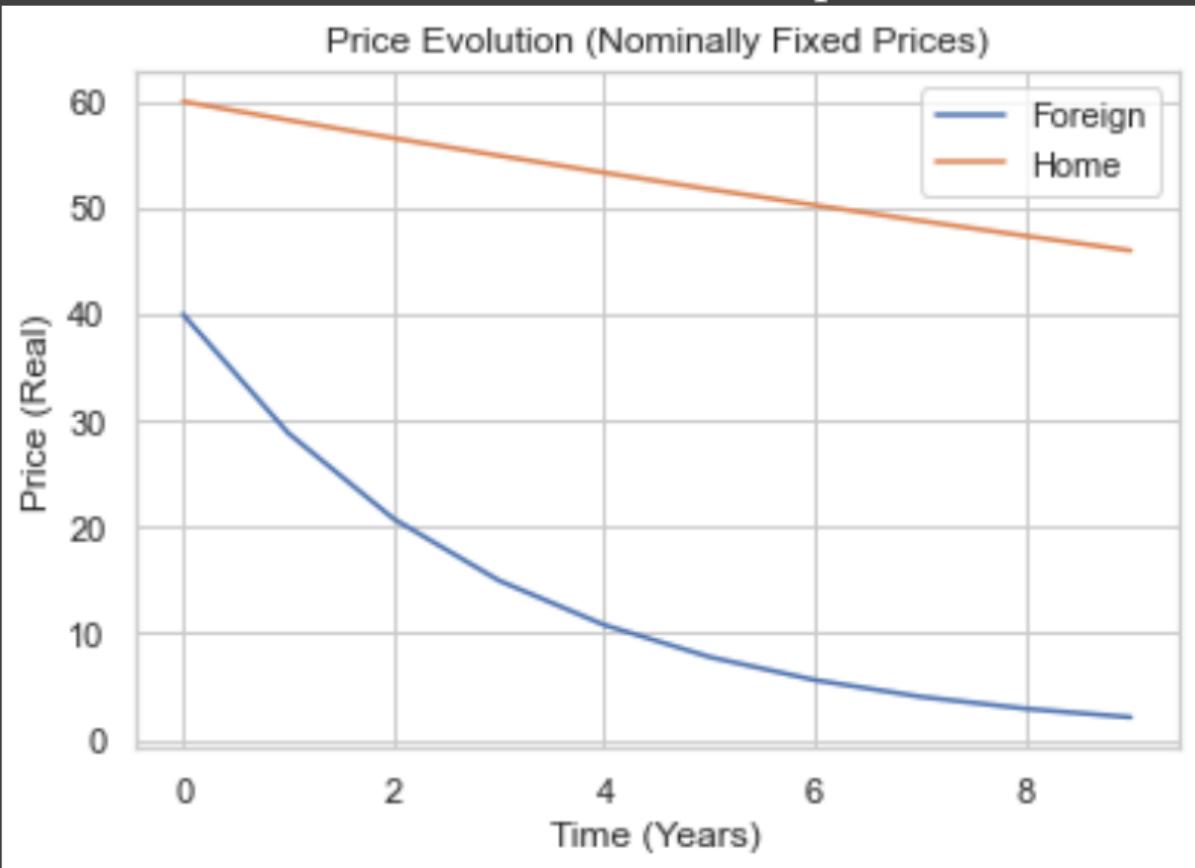
# Reddit Posts vs Exchange Rates

- Counts of Steam regional tricks Reddit posts increase when the USD appreciates relative to the Ukrainian hryvnia, Turkish lira, and Argentinian peso

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## Movements of Relative Prices Example



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# Timing

- In each time period  $t$ :

Exchange rate  $E_t$  realized

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Customer chooses purchase location (given prices & library size)

Customer library evolves via law of motion

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Exchange rate  $E_t$  realized



Firms choose  $p_F, p_H$ , and punishment  $\pi$



Customer chooses purchase location (given prices & library size)



Customer library evolves via law of motion



Exit with probability  $s$ , new entrants drawn from  $\nu$

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# Recursive Stationary Equilibrium

A recursive stationary equilibrium is a collection of prices, punishment strategy, purchase policies such that

1.  $\forall \theta$ ,  $V_\theta(L)$  and  $\text{pol}_\theta(L)$  solve the consumer Bellman equation
2. Firm maximizes profits given induced demand curves by selecting  $p_H, p_F, \pi$
3.  $\forall \theta$   $\mu_\theta^*$  satisfies  $\mu_\theta^* = (1 - s)\mu_\theta^* P_\theta + s\nu$
4.  $G(\theta, L) = w(\theta)\mu_\theta^*(L)$

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# Total Revenue

- The firm chooses prices  $p_H$  and  $p_F$  to maximize total revenue:

$$\sum_t \beta^t p_H \int_L \int_{\theta} \underbrace{\mathbb{1}\{\text{pol}_{\theta}(L) = H\} dG(\theta, L)}_{\text{Purchase at home}}$$

Home customers that purchase in the home market

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$$+ \sum_t \beta^t p_F \int_L \int_{\theta: \theta_c = H} \underbrace{\mathbb{1}\{\text{pol}_{\theta}(L) = F\} dG(\theta, L)}_{\text{Home customer, purchase in foreign}}$$

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$$+ \sum_t \beta^t E_t p_F \int_L \int_{\theta: \theta_c = F} \underbrace{\mathbb{1}\{\text{pol}_{\theta}(L) = F\} dG(\theta, L)}_{\text{Foreign customer purchasing at F price}}$$

Foreign customers purchasing in foreign

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# Externally Calibrated Parameters

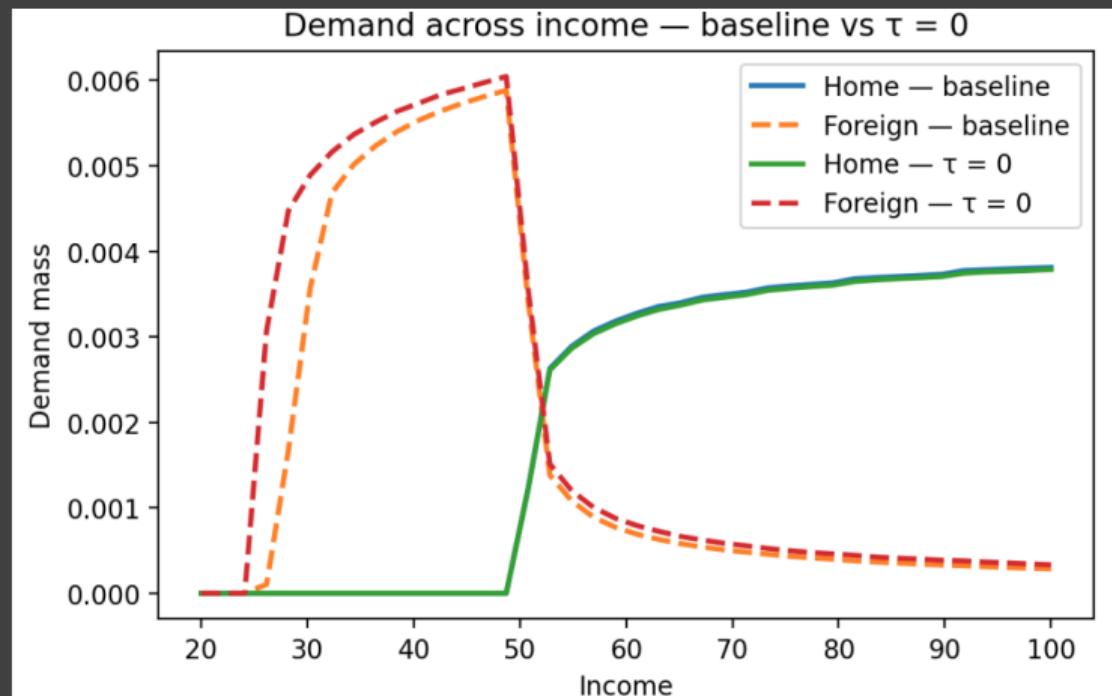
- ▶ Discount factor  $\beta = .98$
- ▶ Exchange rate process
  - ▶ Matched to monthly exchange rate data for Argentina
- ▶ Hazard rate  $s = .0037$ 
  - ▶ Fit to match the hazard rate of playing video games over the lifecycle

[Hazard Estimation](#)

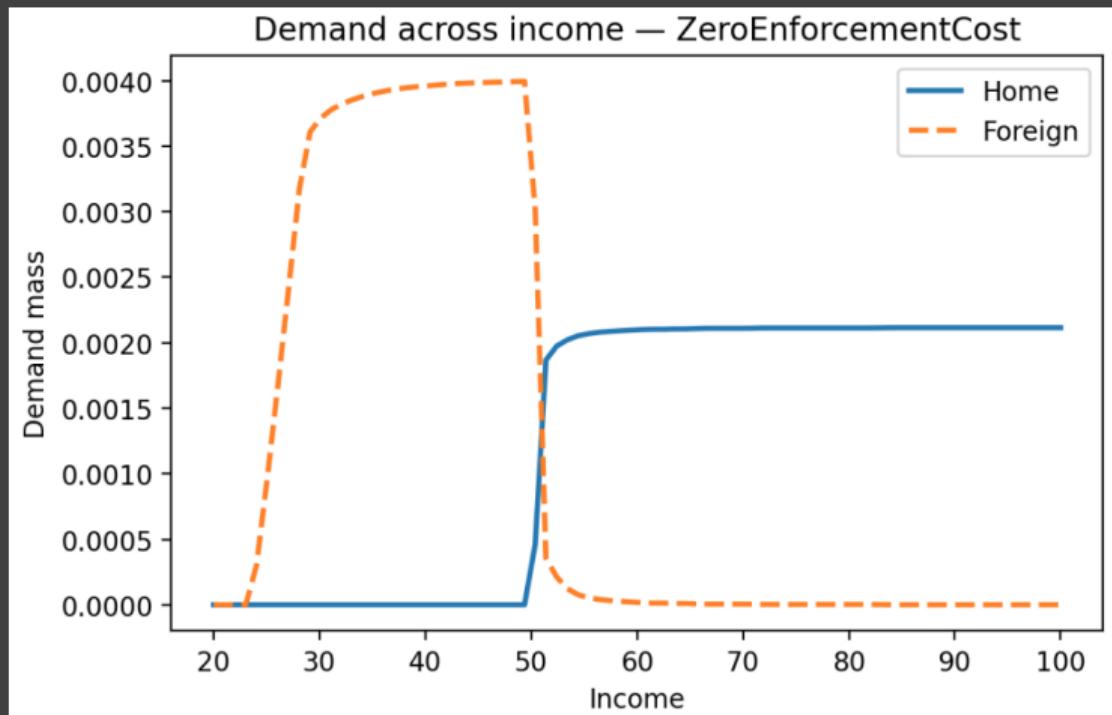
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## Counterfactual: Information Shock

- ▶ Suppose information about accessing foreign markets becomes easier to access
- ▶ The benefits of this accrue to lower income home customers
- ▶ Increased democratization of video games [Back](#)



# Zero Enforcement Cost



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# From Model to Data Moments

Model Object / Parameter	Informative Data Moments
Switching costs ( $\underline{\tau}, \bar{\tau}$ )	Lag between exchange rate shocks and changes in foreign share
Income distribution ( $\mu_y, \sigma_y$ )	Baseline foreign vs. domestic shares by income group
Taste for library size ( $\mu_\alpha, \sigma_\alpha$ )	Correlation between library size and foreign purchasing; baseline foreign share levels
Firm enforcement strategy $\pi$	Correlation between large exchange rate shocks and frequency of punishment discussions; baseline punishment rate
Punishment cost curvature ( $k_1, k_2$ )	Magnitude and frequency of observed punishment actions; false positive rate

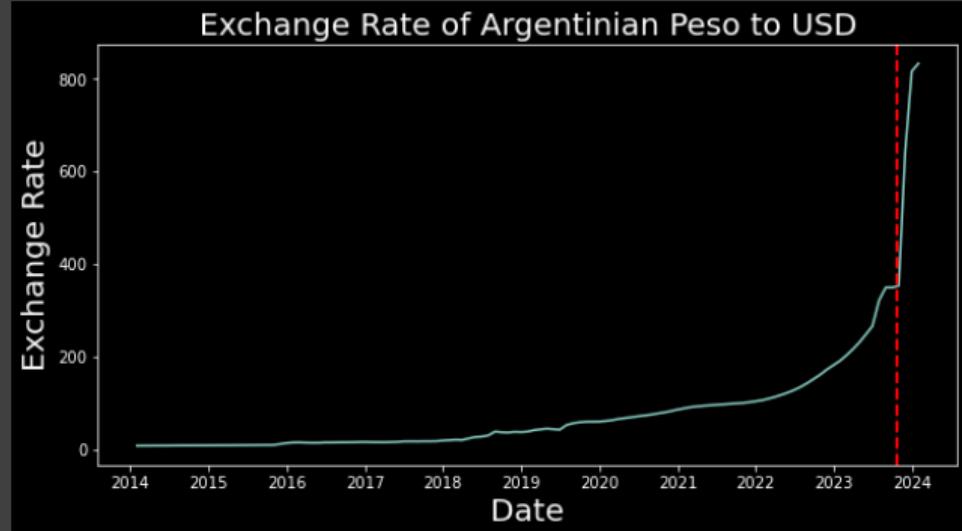
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# Parameter Estimation via SMM (in progress)

- I am currently estimating customer heterogeneity and firm punishment strategy via SMM:
  - Normal distribution of risk aversion:  $\mu_\gamma, \sigma_\gamma$
  - Uniform distribution of fixed switching costs:  $\underline{\tau}, \bar{\tau}$
  - Normal distribution of income (can be Pareto as long as shape > 1):  $\mu_y, \sigma_y$
  - Normal distribution of taste for library size:  $\mu_\alpha, \sigma_\alpha$
  - Firm strategy:  $\pi$
  - Convex punishment cost:  $k_1, k_2$
- Data moments that inform identification
  - Response of foreign purchase share (level) to exchange rate shocks
  - Correlation of size of exchange rate shocks and discussions of punishment actions
  - Baseline foreign and domestic shares
  - Baseline frequency of punishment discussion
  - Lag between exchange rate shocks and changes in foreign share ( $\tau_s$ )
  - False positive rate on punishment [Back](#)

## Demand-Side Assumption: Volatility

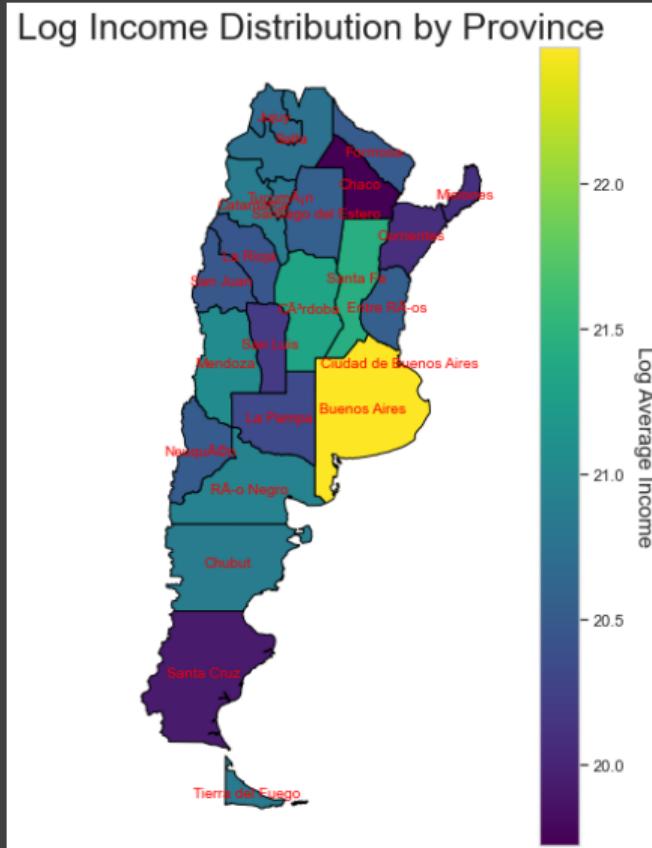
- ▶ Currency volatility exacerbates the tradeoff between arbitrage and price discrimination
- ▶ Consider Argentina [Back](#)



# Demand-Side Assumption: Heterogeneity

- Customer heterogeneity creates price discrimination motives
- US customers are heterogeneous
- Argentinian customers also exhibit high heterogeneity

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# Dollarization and Steam's Policy Change

- The model predicts that the incentive constraints become more difficult to satisfy when there are exchange rate shocks
- When exchange rate shocks are more frequent than price changes, the firm anticipates possible exchange rate trends
- To reduce arbitrage, the firm can either:
  1. Reduce price gap  $\eta$  Model
  2. Price in the home currency (e.g. dollars)
- Under a stable exchange rate, (2) is rarely optimal Proof
  - Firm cedes currency exchange frictions that customers face

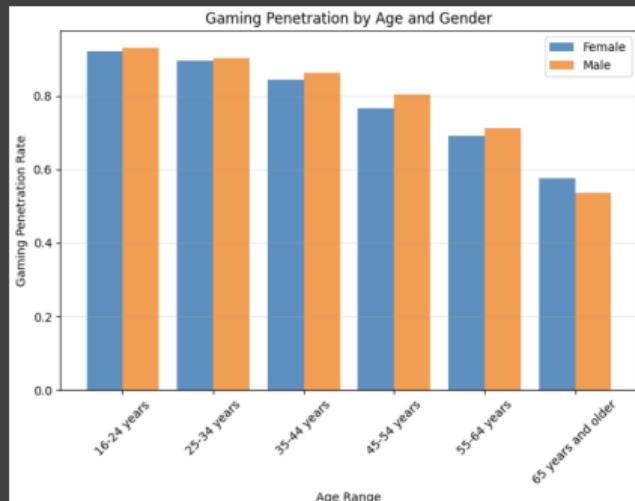
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- October 25, 2023: Steam announces that all sales in Argentina and Turkey will be in **US Dollars** starting November 20, 2023
- The policy change **reset cross-country price gaps and currency simultaneously**

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# Estimating the Hazard Parameter

- ▶ Customers in the model have a constant hazard  $s$  of exiting the market
- ▶ Since the model has a lifecycle component in terms of the size of the video game library, natural to think of  $s$  as the rate of exiting the video game market writ large
- ▶ Fit a constant hazard rate to the cross-sectional fractions of American adults that play video games in different age buckets
- ▶ End up with a hazard rate of .0044 for women and .0037 for men [Back](#)



# Demand Estimation Results: Argentina

Table: Estimation Results

	Coefficient	Confidence Interval
<b>constant</b>	-3.597*** (0.258)	[-4.102, -3.092]
<b>AAA</b>	1.316*** (0.261)	[0.804, 1.828]
<b>price</b>	-0.600*** (0.092)	[-0.781, -0.420]

Table: \*

Notes: \*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ . Standard errors in parentheses.

Developer-level fixed effects are included.

## Demand Estimation Results: US

Table: Estimation Results

	Coefficient	Confidence Interval
<b>constant</b>	-3.191*** (0.188)	[-3.561, -2.821]
<b>AAA</b>	.636*** (0.114)	[0.412, 0.860]
<b>price</b>	-0.051*** (0.009)	[-0.070, -0.033]

Table: \*

Notes: \*\*\*  $p < 0.01$ , \*\*  $p < 0.05$ , \*  $p < 0.1$ . Standard errors in parentheses.

Developer-level fixed effects are included.

## Argentina's Video Game Market

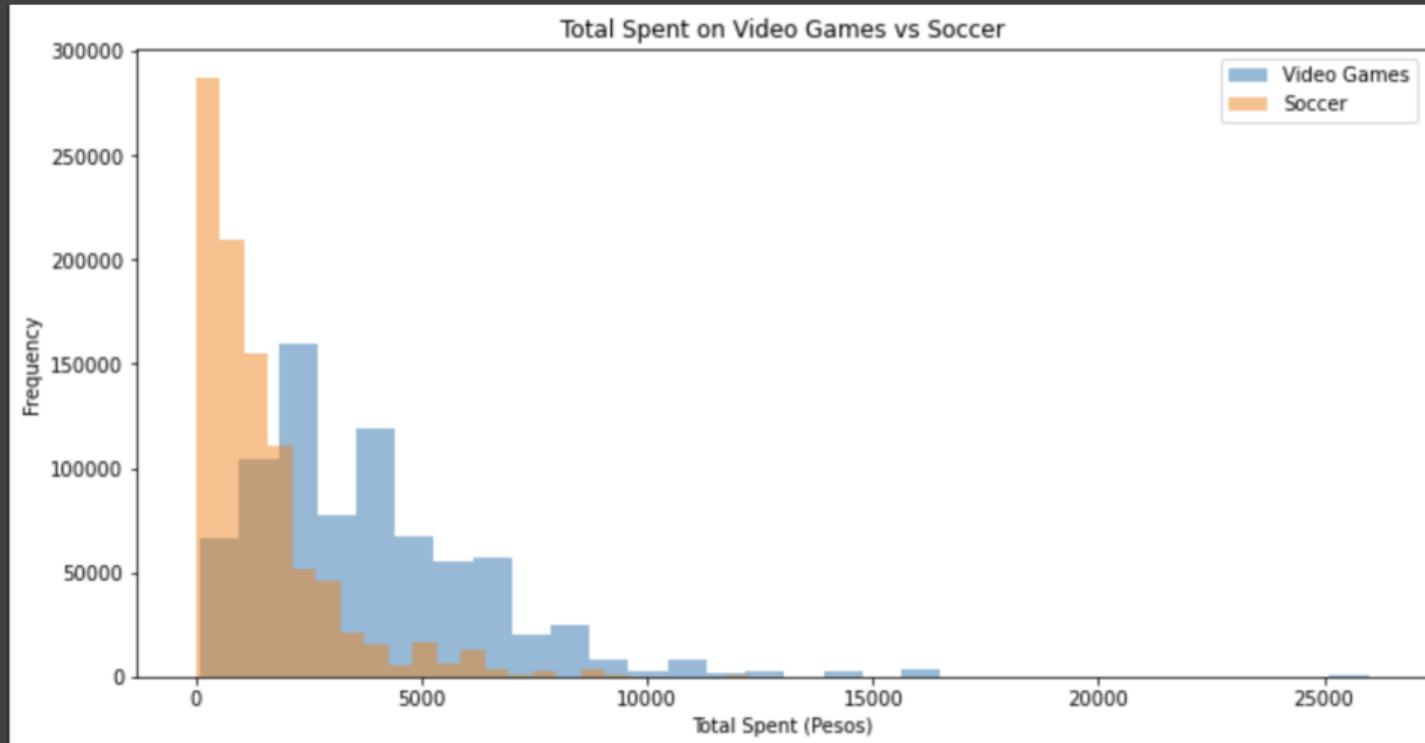


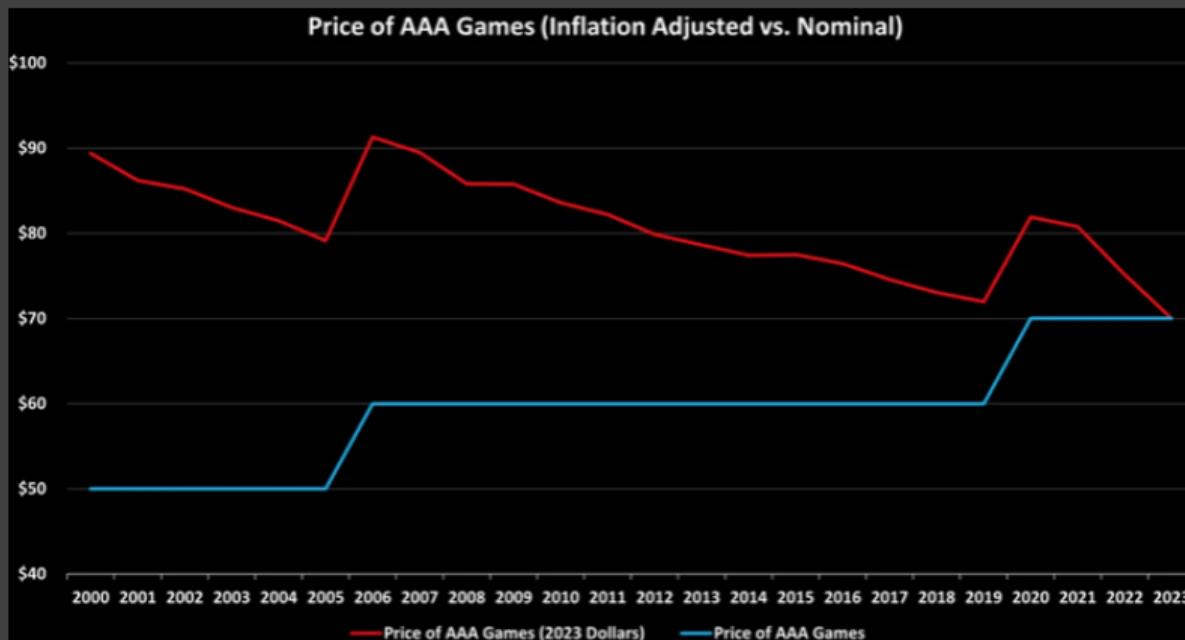
Figure: Argentinians spend more on video games than on soccer 

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# Nominal Price Rigidity

- Nominal prices for AAA games do not move much over time

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## Demand Estimation: Argentina

- ▶ Discrete choice in each period between buying a game or selecting the outside option
- ▶ Think of each choice as a game-market pair
- ▶ Estimate a logit demand model with indirect utility:

$$u_{idjt} = \alpha \ln p_{jt} + \beta \text{AAA}_j + \phi_d + \mu_t + \xi_{jt}$$

- ▶ I estimate separately for Argentina and for the US
- ▶ Coefficient on price is -.600 (se: 0.092)
- ▶ Coefficient on AAA is 1.316 (se: 0.261)
- ▶ Full demand estimation results [AR Demand Estimation](#) [Back](#)

# Static Model in GE

## Model Overview

- ▶ An alternative formulation of the model embeds the Mongey-Waugh pricing inequality framework in space with multiple locations
- ▶ Infinite-horizon economy with households, heterogeneous firms, and a government.
- ▶ Two goods:
  - ▶ Composite good: produced competitively.
  - ▶ Differentiated product: produced by heterogeneous firms (quality  $\psi_{jt}$ , productivity  $z_{jt}$ ).
- ▶ Finite set of markets  $m = 1, \dots, M$  with market-specific prices and exchange rates  $e_{mt}$ .
- ▶ Households can buy from any market by paying hassle cost  $\tau_{imt}$ .

# Households

Preferences:

$$E \left[ \sum_{t=0}^{\infty} \beta^t \sum_{m \in M} \sum_{j \in J} \tilde{u}_{ijmt} \right]$$

where

$$\tilde{u}_{ijmt} = \begin{cases} u(c_{it}) + \psi_j + \xi_{jmt}, & \text{if } j \text{ purchased from } m, \\ 0, & \text{otherwise.} \end{cases}$$

- ▶ Taste shocks  $\xi_{ijmt}$ : i.i.d. Type I Extreme Value with parameter  $\theta$ .
- ▶ Effective price:  $\tilde{p}_{ijmt} = \frac{p_{jm}}{e_{mt}} + \tau_{imt}$ .
- ▶ Labor: supplied inelastically, evolves via Markov process  $P(l, l')$ .
- ▶ Budget constraint:

$$c_{ijmt} + \tilde{p}_{ijmt} + a_{i,t+1} \leq R_{t+1} a_{it} + w_t l_{it} + \Pi_t.$$

## Firms

- ▶ Produce differentiated product with:

$$y_{jt} = z_{jt} n_{jt}^\alpha$$

- ▶ Profits:

$$\Pi_{jt} = \sum_m p_{jm} y_{jmt} - W_t n_{jt}$$

- ▶ Bertrand competition: choose  $p_j$  across all markets to maximize profits.
- ▶ First-order condition in matrix form:

$$x_j = -J^\top (p_j - mc_j)$$

with elasticity matrix  $E_j$  and revenue vector  $R_j$ .

# Markup Equation

From FOCs:

$$\mu_j = - \left( E_j^\top \text{diag}(R_j) \right)^{-1} R_j$$

Element  $k$ :

$$\mu_k = \frac{1 + \sum_{m \neq k} \epsilon_{mk} \frac{R_m}{R_k} \mu_m}{-\epsilon_{kk}}$$

- ▶ Positive cross-elasticities  $\epsilon_{mk} > 0 \Rightarrow$  diversion raises markups.
- ▶ Firms internalize that higher prices in one market may shift demand to other markets they control.

# Government

- ▶ Provides elastic supply of assets.
- ▶ Budget constraint:

$$R_t B_t = B_{t+1}.$$

## Household Problem in Bellman Form

Let  $M_t = a_t + w_t l_t + \pi_t - \frac{a_{t+1}}{R}$  be effective expenditure.

$$v_{jm}(a, l, \tau) = \max \{u(M_t - \tilde{p}_{jm}) + \psi_j + \beta E[v(\cdot)]\}$$

Choice probability:

$$\rho_{jm}(M_t) = \frac{\exp [\theta(u(M_t - \tilde{p}_{jm}) + \psi_j)]}{\sum_{k \in J} \sum_{n \in M} \exp [\theta(u(M_t - \tilde{p}_{kn}) + \psi_k)]}$$

# Aggregation and Equilibrium

- Household law of motion:

$$\Lambda(a', l', \tau') = \int \rho_{jm}(a, l, \tau) \Lambda(a, l, \tau) P(l, l') da dl d\tau$$

- Aggregate demand for firm  $j$  in market  $m$ :

$$x_{jm} = \int \rho_{jm}(a, l, \tau) \Lambda(a, l, \tau) da dl d\tau$$

- Stationary recursive equilibrium: household optimization, firm optimization, market clearing, stationary  $\Lambda$ , government budget constraint.