

# Reference Pricing and Consumption Inequality

Janet Stefanov

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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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  - ▶ 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 **reallocate 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

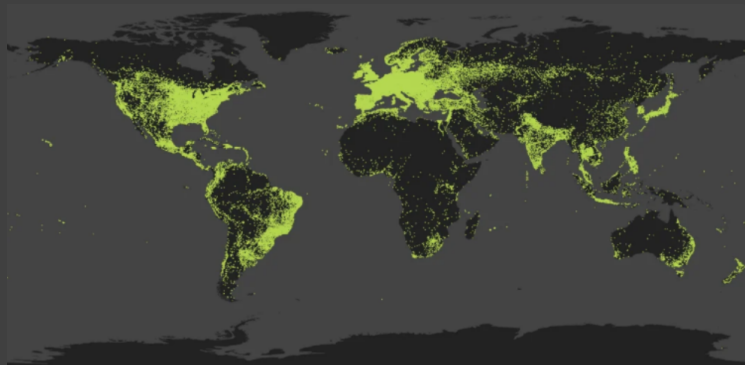
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- ▶ The video game market is an *ideal setting* Market Size



# 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)

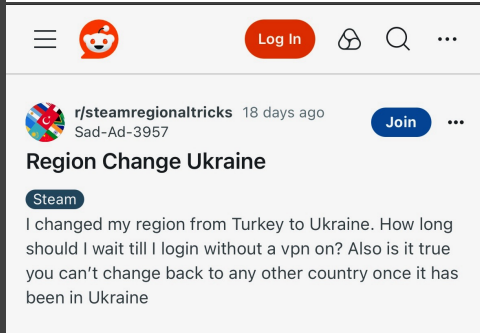
# Cross-Country Price Variation



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	₩ 66000	\$52.59 -12.33%
Turkish Lira	₺249.00	\$14.78 -75.35%
Ukrainian Hryvnia	899₴	\$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 Zloty	199,00zł	\$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%
Kazakhstan Tenge	11499₸	\$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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- ▶ 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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- ▶ 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



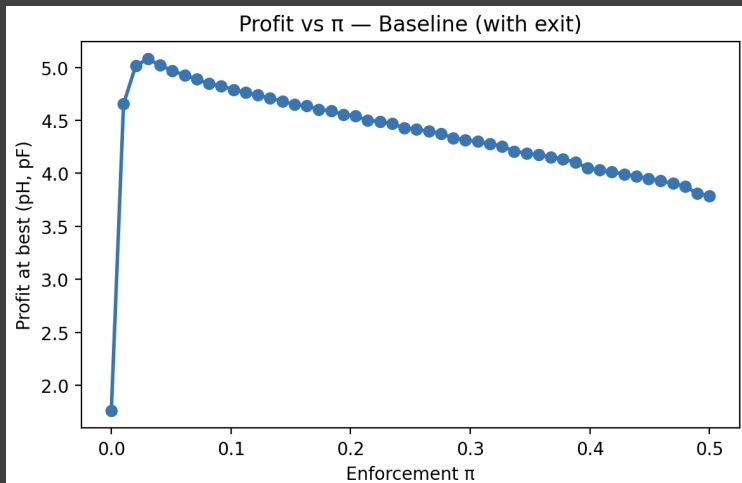
# Global Prices and Consumer Surplus

- ▶ 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

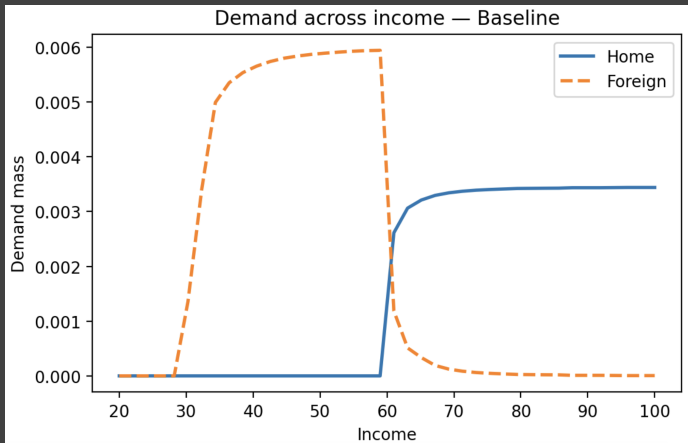
# Model Simulations

- ▶ 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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- ▶ **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

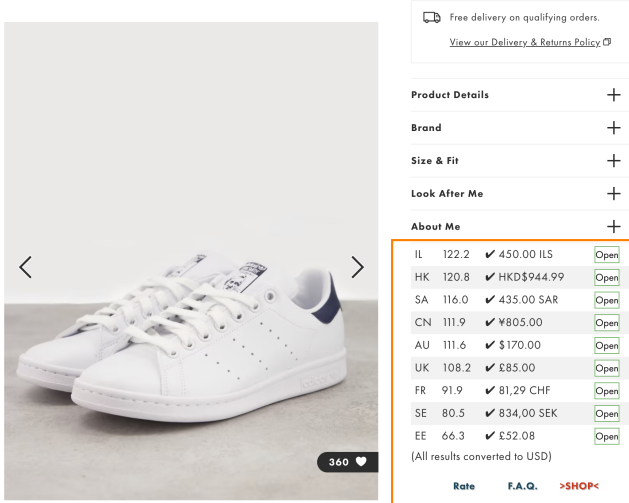
# EU's Digital Single Market

- ▶ 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 [Back](#)



The screenshot shows a product page for a pair of white sneakers. On the right side, there is a table titled 'Product Details' with expandable sections for 'Brand', 'Size & Fit', 'Look After Me', and 'About Me'. The 'About Me' section is expanded, showing a table of prices for different countries. The table has columns for Country Code, Price in Local Currency, Price in USD, and a link to 'Open' the product page for that country. The prices are listed for IL, HK, SA, CN, AU, UK, FR, SE, and EE. The table is highlighted with an orange border.

Country	Price (Local)	Price (USD)	Action
IL	122.2	✓ 450.00 ILS	<a href="#">Open</a>
HK	120.8	✓ HKD\$944.99	<a href="#">Open</a>
SA	116.0	✓ 435.00 SAR	<a href="#">Open</a>
CN	111.9	✓ ¥805.00	<a href="#">Open</a>
AU	111.6	✓ \$170.00	<a href="#">Open</a>
UK	108.2	✓ £85.00	<a href="#">Open</a>
FR	91.9	✓ 81,29 CHF	<a href="#">Open</a>
SE	80.5	✓ 834,00 SEK	<a href="#">Open</a>
EE	66.3	✓ £52.08	<a href="#">Open</a>

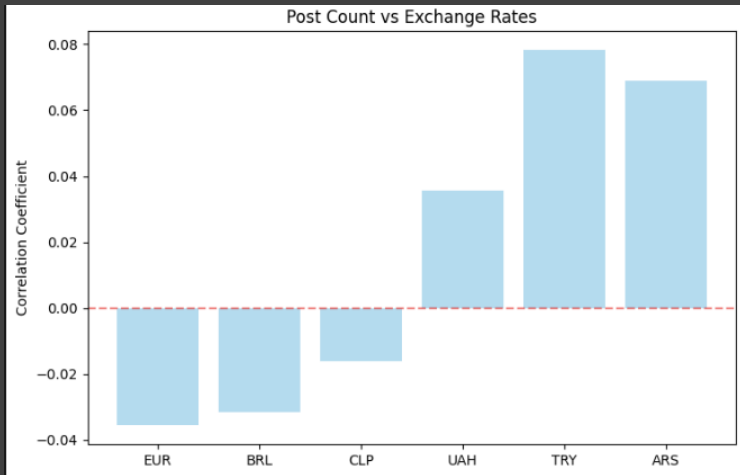
(All results converted to USD)

Rate F.A.Q. >SHOP<

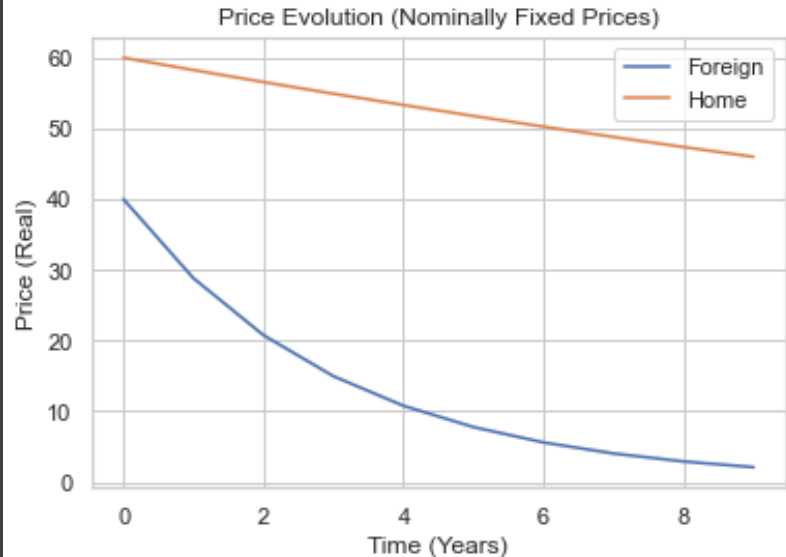
# Reddit Posts vs Exchange Rates

- 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



# Timing

- In each time period  $t$ :

Exchange rate  $E_t$  realized

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

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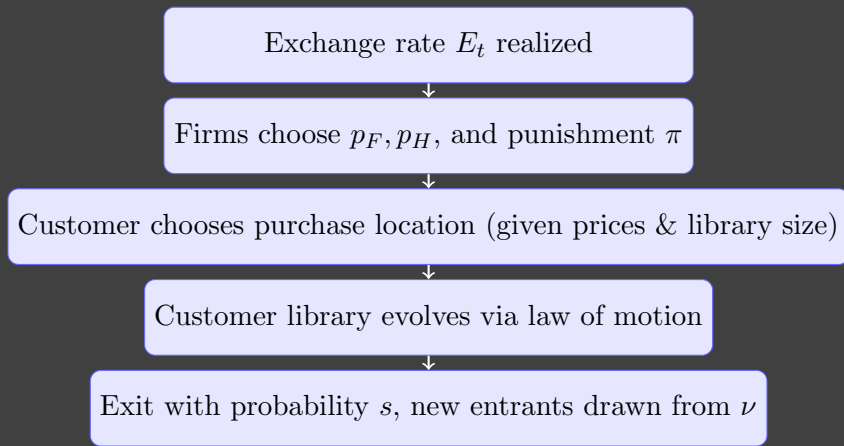
Firms choose  $p_F, p_H$ , and punishment  $\pi$

Customer chooses purchase location (given prices & library size)

Customer library evolves via law of motion

# Timing

- In each time period  $t$ :



# 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:

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

Home customers that purchase in the home market

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Home customers that purchase in the home market

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

Home customers that buy at foreign prices

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Home customers that buy at foreign prices

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

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

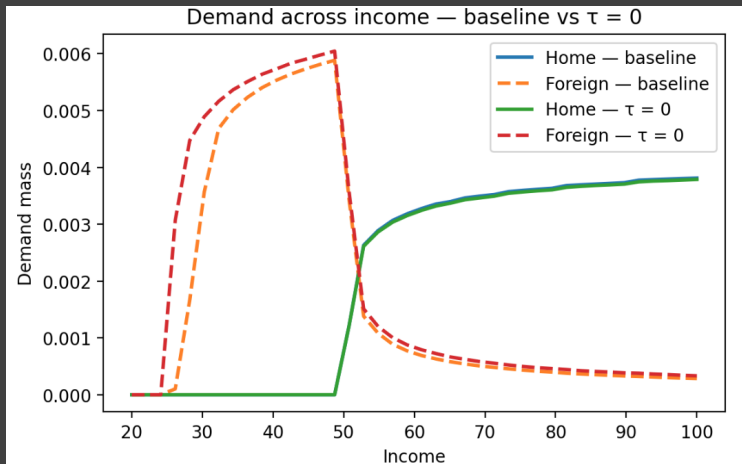
[Hazard Estimation](#)

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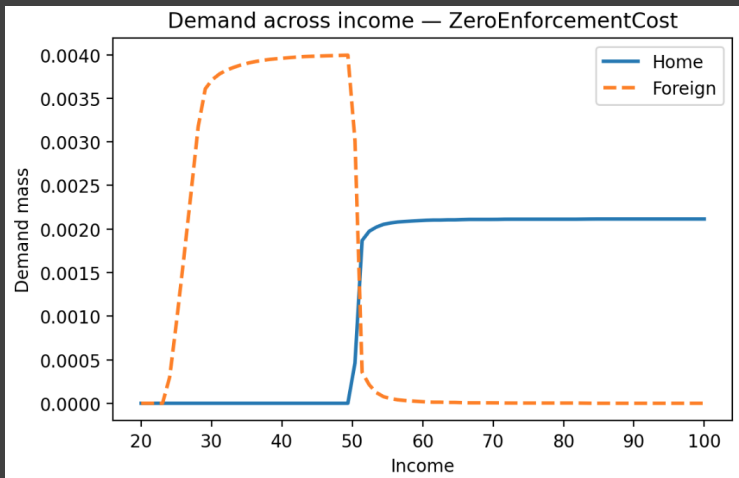


# Counterfactual: Information Shock

- ▶ 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

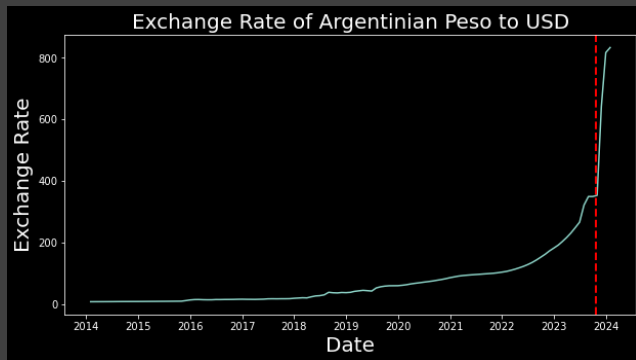
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# Parameter Estimation via SMM (in progress)

- ▶ 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

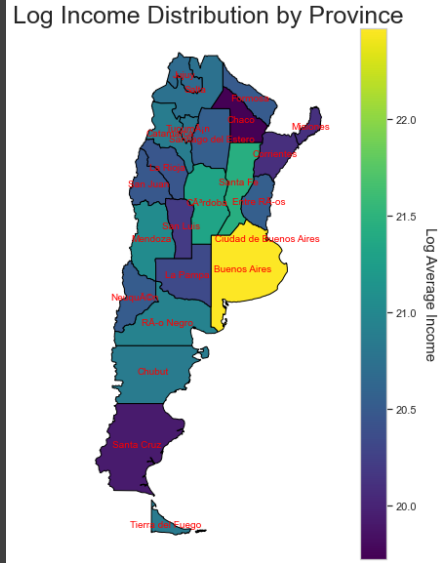
- ▶ Currency volatility exacerbates the tradeoff between arbitrage and price discrimination
- ▶ Consider Argentina [Back](#)



# Demand-Side Assumption: Heterogeneity

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

[Back](#)



# 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

# Dollarization and Steam's Policy Change

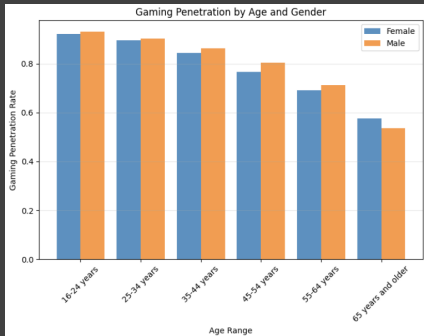
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- ▶ Under a stable exchange rate, (2) is rarely optimal [Proof](#)
  - ▶ Firm cedes currency exchange frictions that customers face
- ▶ October 25, 2023: Steam announces that all sales in Argentina and Turkey will be in **US Dollars** starting November 20, 2023
- ▶ The policy change **reset cross-country price gaps and currency** simultaneously

[Back](#)



# 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
constant	-3.597*** (0.258)	[-4.102, -3.092]
AAA	1.316*** (0.261)	[0.804, 1.828]
price	-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
constant	-3.191*** (0.188)	[-3.561, -2.821]
AAA	.636*** (0.114)	[0.412, 0.860]
price	-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

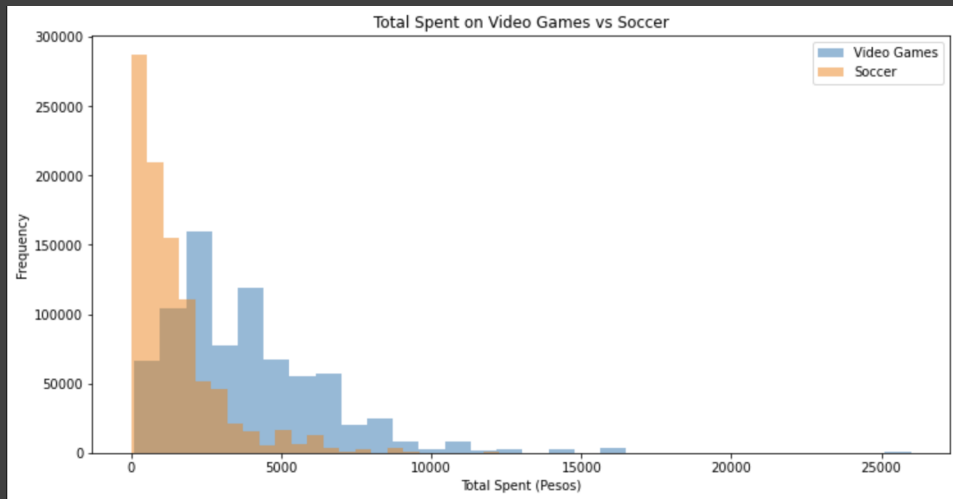
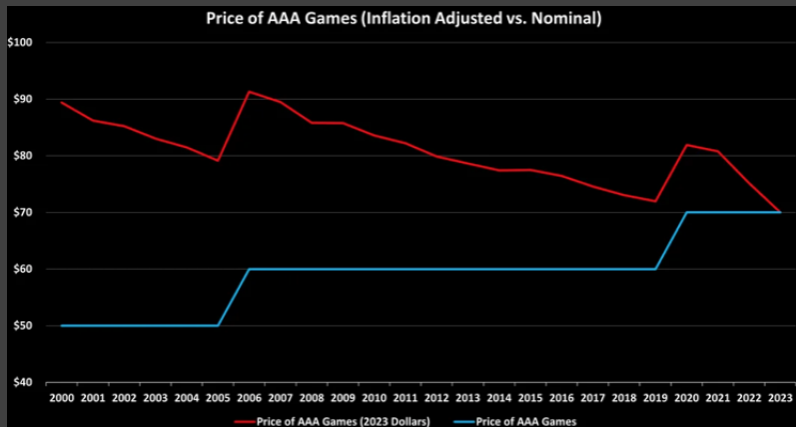


Figure: Argentinians spend more on video games than on soccer [Back](#)

# Nominal Price Rigidity

- ▶ **Nominal prices** for AAA games do not move much over time [Back](#)



# 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](#)

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# 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.