

The Longshot bias in market data: Evidence from Counter-Strike: Global Offensive

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Loot Boxes

- ▶ Many video games have chosen to sell cosmetic alterations to their games using randomization mechanisms called “loot boxes”
- ▶ Economic Literature tells us that there is no benefit to randomization for risk-neutral consumers, so the benefit must come from risk-loving consumers.
- ▶ What aspect of these lotteries is generating the revenue for the companies selling them?
- ▶ How much more revenue-generating is this compared to traditional selling mechanisms?

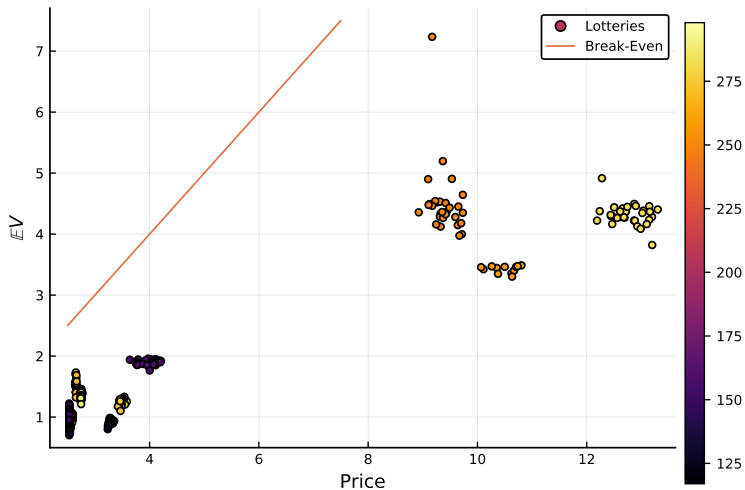
Why do we care?

- ▶ We are interested in discovering what drives this market to feature randomization mechanisms.
- ▶ Are consumers inherently more risk-loving when they play video games?
- ▶ Is this driven by consumers over-weighting tiny probabilities as cumulative prospect theory suggests?
- ▶ Are consumers weighing benefits and losses differently?
- ▶ What is the magnitude of these gains from randomization?

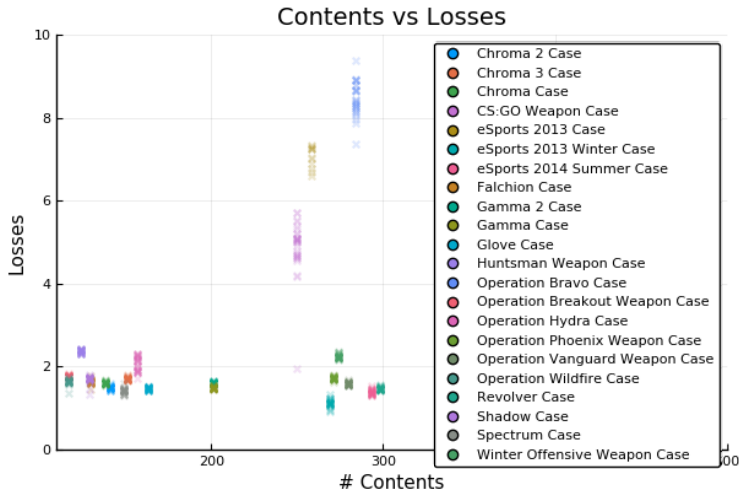
The Data

- ▶ Contains complete market history for all items sold in the Steam Community Market for *Counter-Strike: Global Offensive*
- ▶ Market history is specific to the hour for the last 30 days, specific to the day for the remaining time the item has existed.
- ▶ Contains all active buy and sell orders for each of these items as of March 31st 2018.
- ▶ Number of active players per day and unique twitch viewers per day

Pictures



Does Size Matter?



Lotteries

Case	Values		Number of Contents				
	$\mathbb{E}[V]$	Price	#Blue	#Purple	#Pink	#Red	#Gold
Operation Wildfire	0.89891	2.5307	26	18	14	9	50
Operation Breakout	0.77011	2.5305	24	15	12	10	56
Falchion Case	0.95072	2.5323	27	24	11	9	59
Shadow Case	0.85299	2.5349	29	17	14	10	59
Huntsman Weapon Case	0.95531	3.3181	25	17	12	8	62
Spectrum Case	0.98146	2.53	34	23	15	9	68
Chroma 2 Case	1.0058	2.53	25	13	13	9	81
Chroma 3 Case	0.66099	2.53	30	19	11	10	81
Chroma Case	0.83215	2.55	23	20	10	4	81
Glove Case	0.84301	2.53	27	26	9	12	89
Operation Hydra	1.5465	4.0827	25	20	14	9	89
Gamma 2 Case	0.68335	2.53	31	22	13	7	128
Gamma Case	0.80717	2.53	31	21	11	10	128

High Content Lotteries

Case	Values		Number of Contents				
	$E[V]$	Price	#Blue	#Purple	#Pink	#Red	#Gold
CS:GO Weapon	4.4611	9.3248	7	6	7	2	228
eSports 2013 Case	3.2708	10.354	8	13	7	2	228
eSports 2013 Winter	1.5687	2.6441	18	9	11	3	228
eSports 2014 Summer	1.4136	2.7414	21	19	16	9	228
Operation Bravo	4.3567	12.628	26	15	9	6	228
Operation Phoenix	0.85507	2.5416	15	12	9	7	228
Operation Vanguard	1.038	2.5928	17	13	12	10	228
Revolver Case	1.1045	2.53	24	25	12	9	228
Winter Offensive	1.299	3.5079	14	14	12	6	228

Discrete Choice - Berry (1994)

Utility for these lotteries is quasi-linear

$$u_{ijt} = V(x_{jt}, p_{jt}; \theta) + \xi_{jt} + \epsilon_{ij} \quad \epsilon_{ij} \sim \text{Gumbel}$$

Consumers choose the lottery that has the highest utility for them:

$$\Pr(i \rightarrow j) = \frac{\exp(V(x_{jt}, p_{jt}; \theta) + \xi_{jt})}{\sum_{k \in \mathcal{F}} \exp(V(x_{kt}, p_{kt}; \theta) + \xi_{kt})}$$

Using an outside option that is normalized so that it has zero utility:

$$\log s_{jt} - \log s_{0t} = V(x_{jt}, p_{jt}; \theta) + \xi_{jt}$$

Implications

- ▶ Differentiated Goods
- ▶ Prediction based on market shares
- ▶ Homogeneous Consumers - Is this reasonable?
- ▶ No structure placed on ξ

Cumulative Prospect Theory

- ▶ Four main components: Reference dependence, loss aversion, diminishing sensitivity, and probability weighting
- ▶ Diminishing sensitivity and loss aversion are summarized by the valuation function for each content of the lottery.
- ▶ x is not the content of the lottery, but the value of the gain or loss of that content relative to some reference point.

$$v(x) = \begin{cases} x^\alpha & x \geq 0 \\ -\lambda(-x)^\alpha & x < 0 \end{cases}$$

Reference dependence and Loss Aversion

- ▶ What is the proper reference point?
- ▶ Can it be estimated?
- ▶ How is loss aversion tied to the reference point?

Picture

Probability Weighting Function

$$w(p) = \frac{\gamma p^\delta}{\gamma p^\delta + (1 - p)^\delta}$$

- This weight is applied to cumulative probabilities - Interpretations of γ and δ ?

Picture

Valuation of a Lottery

- ▶ The "viewed" probability of each person obtaining an item from the lottery is

$$\Pi_{s_i} = \sum_{j=1}^{s_i} \pi_{s_j}$$

$$p_i = w(\Pi_{s_i}) - w(\Pi_{s_{i-1}})$$

$$F(x_i) = [w(\Pi_{s_i}) - w(\Pi_{s_{i-1}})] v(x_i - R)$$

$$V = \sum_{i=1} F(x_i)$$

Estimation Procedure

Estimation

- ▶ Price is determined by intersection of supply and demand and is therefore endogenous
- ▶ Instrument with the changes in daily player base from the average number of players

$$\xi_{jt} = \log s_{jt} - \log s_{0t} - V(x_{jt}, p_{jt}; \theta)$$

Using the orthogonality of ξ_{jt} to the instruments and exogenous parameters:

$$\min_{\xi_{j,t}, \xi_{j,t}} \sum_{j,t} \xi'_{j,t} \Omega \xi_{j,t}$$

$$\text{subject to: } \xi_{j,t} = \log s_{jt} - \log s_{0t} - V(x_{jt}, p_{jt}; \theta)$$

$$\xi_{j,t} = \xi_{j,t} \mathbf{Z}_{j,t}$$