# Does Round Pricing Lead to More Favorable Product Perception?

Mike Fan, Prakhar Maini, Sean Norris, Claudia Ventresca

### **Research Question**

#### Does Round Pricing Lead to More Favorable Product Perception?

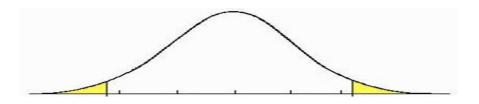
Price may not only only affect perceived quality but also affect the actual satisfaction of owning and using the product

## **Hypothesis and Direction of Outcomes**

 $H_0 = 0$   $H_a \neq 0$ 

Price rounding, in a hedonic context, has **no effect** on perception of product quality

Price rounding, in a hedonic context, <u>has an effect</u> on perception of product quality



### **Treatment**

#### **Pre-Treatment**



5 Questions, 10 pt. Scale

**Treatment** 



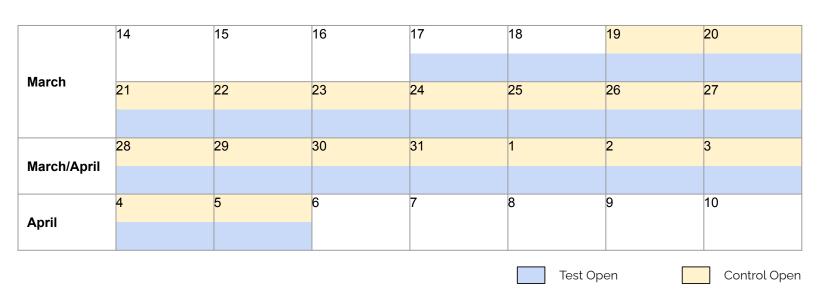
Checkout Prompt + Pricing

#### **Post-Treatment**



Same Survey as Pre-Treatment

## **Randomization**



We used this strategy to try to minimize excludability issues on mTurk

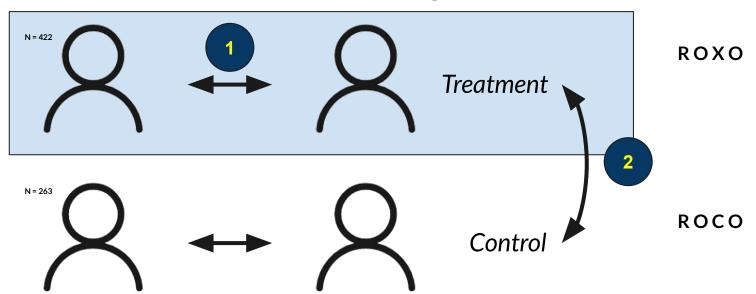
#### **Measurement Units**

People, specifically 1,437 mTurk respondents:

- **Control:** High Non-Round = 348 **(119)**
- **Control:** Low Non-Round = 413 (144)
  - De-duplicated Control = 263
- **Treatment:** Round = 677 **(422)**



# Before & After Placebo Design



#### **Outcome Measures**

Difference in responses to the question-topics below for each person and between test and control

Will Not Require Maintenance

Will Live Up to Stated Expectations

Possess a Rich Set of Features

Exhibits Beautiful Design and Craftsmanship

**Anticipated Satisfaction** 

Each question was answered on a scale from 0 to 10



# **Analysis and Results**

Bin_group	С		Т		_
Variable	N	Percent	N	Percent	Test
gender	263	17.50/	422	50.50/	X2=2.362
Female	125	47.5%	213	50.5%	
Male	130	49.4%	201	47.6%	
Non-binary	5	1.9%	3	0.7%	
Prefer not to say	3	1.1%	5	1.2%	
education	261		422		X2=5.475
Bachelors	125	47.9%	220	52.1%	
Doctorate	9	3.4%	6	1.4%	
High school/GED	65	24.9%	112	26.5%	
Masters	60	23%	79	18.7%	
Prefer not to say	2	0.8%	5	1.2%	
age_group	261		420		X2=2.958
18-25	31	11.9%	40	9.5%	
26-35	89	34.1%	161	38.3%	
36-45	62	23.8%	94	22.4%	
46-55	43	16.5%	61	14.5%	
56+	36	13.8%	63	15%	
Prefer not to say	0	0%	1	0.2%	
income_group	263		421		X2=6.371
\$1,000,000+	3	1.1%	8	1.9%	
\$100,001 - \$200,000	45	17.1%	72	17.1%	
\$200,001 - \$300,000	18	6.8%	26	6.2%	
\$300,001 - \$500,000	20	7.6%	15	3.6%	
\$500,001 - \$1,000,000	12	4.6%	23	5.5%	
0 - \$100,000	158	60.1%	266	63.2%	
Prefer not to say	7	2.7%	11	2.6%	

	Dependent variable:					
	Camera - Post treatment	Headphone - Post treatment	Laptop - Post treatment			
Treatment	0.093 (0.347)	0.599 (0.389)	0.433 (0.414)			
Previous Score - Camera	0.841*** (0.040)					
Previous Score - Headphone		0.907*** (0.028)				
Previous Score - Laptop			0.885*** (0.033)			
Constant	5.462*** (1.848)	3.767*** (1.429)	1.802 (2.473)			
Age buckets	Yes	Yes	Yes			
Gender buckets	Yes	Yes	Yes			
Education buckets	Yes	Yes	Yes			
Income buckets	Yes	Yes	Yes			
Observations	679	679	679			
R2	0.624	0.684	0.610			
Adjusted R2	0.613	0.675	0.598			
Residual Std. Error (df = 658)	4.474	4.864	5.303			
F Statistic (df = 20; 658)	54.615***	71.265***	51.471***			

# **Analysis and Results - HTE (Cont..)**

	Camera	Headphone	Laptop	Bonferroni corrected threshold
Age	0.423	0.546	0.978	0.01
Income	NA	0.263	0.024	0.0083
Gender	0.578	0.132	0.128	0.0167
Education	0.036	0.0118	0.035	0.0125

#### **Questions and Concerns**

- How can we improve our randomization given the time, platform, and monetary constraints?
- Can you think of a way to perform the utilitarian arm of the experiment given the above constraints?
- How can we improve the generalizability of our findings to other product and purchase contexts?
- How can the findings be used to price products in an economic and ethically sound manner?