

# The Effect of Discount Option Quantity on Consumer Purchase Decisions

Group 2

Pilot Study Findings



# MOTIVATION

- **Choice overload:** too many options can depress purchase likelihood (Iyengar & Lepper 2000).
- **Discount framing:** identical monetary values drive different choices depending on “% off”, “\$ off”, etc. Wording and format can influence consumer behavior (Gendall et al., 2006)
- Online retailers often show several discounts at once – yet little is known about the joint effect of quantity × frame.



# CHOICE OVERLOAD IN CONSUMER ENVIRONMENTS

- Consumers face an overwhelming array of choices spanning product assortments, brands, features, and promotional offers (Broniarczyk & Griffin, 2014; Kahn et al., 2014).
- **Paradox of Choice:** excessive options can diminish consumer satisfaction and purchase likelihood (Schwartz, 2004; Scheibehenne et al., 2010)
  - Iyengar and Lepper's (2000) field experiment: Shoppers encountered either 6 or 24 varieties of jam at a grocery sampling display. While the extensive assortment initially attracted more visitors, the limited assortment generated substantially higher purchase rates (30% vs. 3%).

# MECHANISM OF CHOICE OVERLOAD

Several cognitive processes appear to contribute to this phenomenon:

- ➊ Larger choice sets demand greater cognitive resources to evaluate options, contributing to decision complexity and mental fatigue (Chernev et al., 2015).
- ➋ Expanded choice sets often heighten decision difficulty by increasing the number of attribute trade-offs consumers must navigate (Dhar & Nowlis, 1999; Tversky & Shafir, 1992).
- ➌ Extensive choice sets may elevate anticipated regret by increasing consumers' concerns about overlooking superior alternatives (Carmon et al., 2003; Sagi & Friedland, 2007)
- ➍ Larger assortments frequently amplify choice deferral, as consumers postpone decisions when confronted with choice complexity (Dhar, 1997; Jessup et al., 2009).

# DISCOUNT FRAMING

- Discount framing is how price reductions are presented to consumers, as is a particularly important dimension of promotional presentation (DelVecchio et al., 2007; Krishna et al., 2002).
- Economically equivalent discounts can evoke different consumer responses depending on their presentation format (Gendall et al., 2006; Chen et al., 1998; Hardesty & Bearden, 2003)
- Several common discount framing approaches dominate retail environments:
  - ① Percentage discounts (e.g., “20% off”)
  - ② Absolute or dollar-off discounts (e.g., “\$20 off”)
  - ③ Direct price comparisons (e.g., “Was \$100, now \$80”)
  - ④ Quantity-based promotions such as “buy one get one”

# EFFECT OF DISCOUNT FRAMING

The relative effectiveness of these framing approaches varies with contextual factors.

- Low-priced products: percentage discounts often appear more attractive than absolute discounts of equivalent value, as they create an impression of greater savings relative to the base price (Chen et al., 1998; DelVecchio et al., 2007).
- High-priced products, absolute discounts frequently outperform percentage discounts, as consumers focus on the absolute magnitude of savings rather than the proportion (Chen & Rao, 2007; González et al., 2016).



Cognitive processing depth also moderates framing effects.

- Under high cognitive load or time pressure, consumers often rely more heavily on heuristic processing, rendering them more susceptible to certain framing approaches (Chaiken, 1980; Lee & Chen, 2011).
- When cognitive resources are constrained, simplicity of calculation and immediate accessibility may determine which discount frame appears most attractive (Coulter & Coulter, 2007; Thomas & Morwitz, 2009).

# THE INTERSECTION OF CHOICE QUANTITY AND DISCOUNT FRAMING

- Existing gap: Despite parallel advances in understanding choice overload and discount framing, little research has systematically investigated their interaction.
- Importance of studying the intersection:
  - ① **Cognitive load theory:** Decision complexity  $\uparrow \Rightarrow$  Consumers rely more heavily on simplistic processing strategies (Payne et al., 1992; Bettman et al., 1998)
  - ② Limited attention resources may lead consumers to focus on certain framing characteristics (Kahneman, 1973; Milosavljevic et al., 2012)
  - ③ **Choice architecture research** : option arrangement significantly influences preference construction, particularly under conditions of decision complexity (Johnson et al., 2012; Dinner et al., 2011).

# CONTRIBUTIONS

The present research aims to address this theoretical gap by systematically manipulating both the number of discount options presented to consumers and the framing approaches employed across these options. By examining which discount frames consumers select when faced with varying choice set sizes, we contribute to a more integrative understanding of how choice architecture and promotional framing jointly shape consumer decision-making in digital commerce environments.

# EXPERIMENT PROCEDURE

This study will be conducted on the Gorilla online experiment platform. The experimental procedure follows this sequence:

- ① Consent and demographic survey
- ② Initial Block (9 trials):
  - 2-4 discount options with varying base prices
  - Randomized trial order and discount positions
  - No follow-up questions
- ③ Main Block (6 trials):
  - 2 trials for each condition (2-4 discount options)
  - Identical base prices and final prices within each trial
  - Follow-up questions
  - 2 Attention checks with explicit instructions and memory-based attention checks
- ④ Debriefing and Completion

# STIMULI

For all stages of the experiment, participants will be presented with between 2 and 4 discounted products in each trial. All options include

- A photo of the product
- Base price
- One of four discount framings:
  - ① Now-price,
  - ② \$-off,
  - ③ %-off,
  - ④ Buy-one-get-one.









# FOLLOW-UP QUESTIONS

Please answer the following questions about the task you just completed.

How many discount options did you see on the previous screen?

How likely would you be to buy the product you chose?

Unlikely

1

2

3

4

5

6

7

Likely

How confident are you in the choice you made?

Not  
Confident

1

2

3

4

5

6

7

Confident

If given more time to consider the options, how likely are you to make the same choice?

Unlikely

1

2

3

4

5

6

7

Likely

Why did you make this decision? Please answer in at least 10 words.

# REVISIONS

Please answer the following questions about the task you just completed.

How many options did you see on the previous screen?

How likely would you be to buy the product shown?

Unlikely

1	2	3	4	5	6	7
---	---	---	---	---	---	---

Likely

How confident are you in option you chose?

Not  
Confident

1	2	3	4	5	6	7
---	---	---	---	---	---	---

Confident

Please select 7 for this response.

Not  
Confident

1	2	3	4	5	6	7
---	---	---	---	---	---	---

Confident

If given more time to consider the options, how likely are you to make the same choice?

Unlikely

1	2	3	4	5	6	7
---	---	---	---	---	---	---

Likely

Why did you make this decision? Please try to answer in at least 5 words.

Next

# KEY MEASURES

- **Primary:** chosen frame, screen position.
- **Process:** reaction time (RT), total mouse path (px).
- **Post-choice:** buy likelihood, confidence, choice stability (7-pt scales).
- Demographics for context (age, gender, income, shopping frequency).

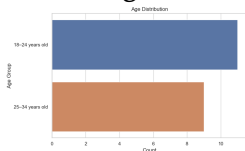
# PILOT SAMPLE (N = 20)

- U.S. convenience sample.
- Age:  $M = 28.6$  ( $SD = 6.4$ ); 60 % female.
- Ethnicity: 75 % Asian; Education: 75 % Bachelor's+.
- Goal = protocol debugging and effect-size estimation.

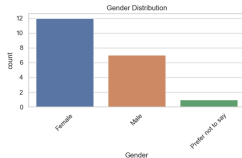
As expected, the sample skewed toward young adults, highly educated individuals, majority Asian racial identity, lower income brackets (likely students). While this limits generalizability, it served the pilot's purpose of identifying technical and procedural issues prior to scaling the study to a more diverse population.

# DESCRIPTIVE STATISTICS OVERVIEW

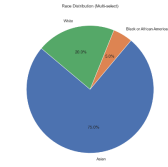
## Age



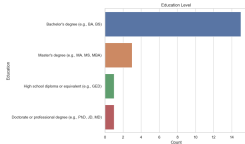
## Gender



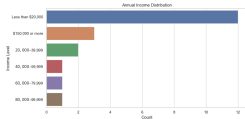
## Race



## Education



## Income

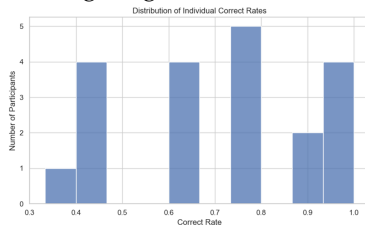
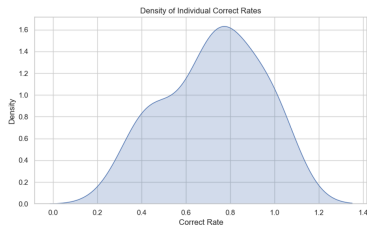


## Shopping Frequency



# INDIVIDUAL ACCURACY PERFORMANCE

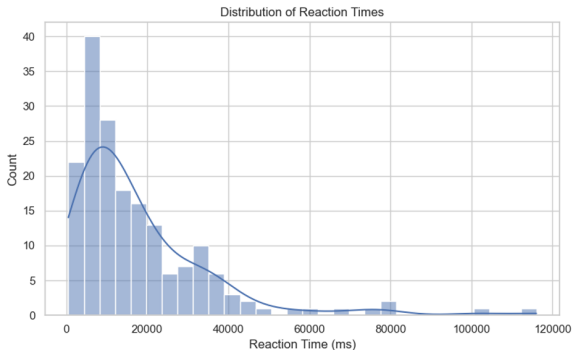
Distribution of individual correct rates during fatigue trials:



Most participants had correct rates between 0.6 and 0.9, with fewer falling below 0.5.

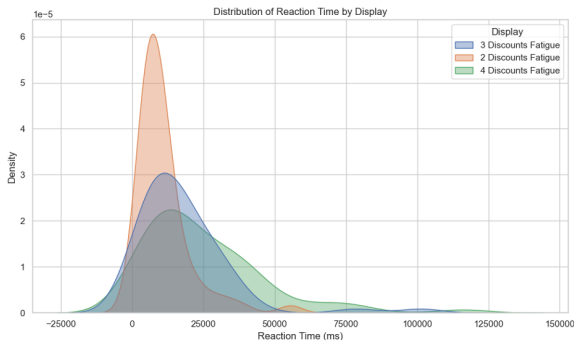
# REACTION TIME DISTRIBUTIONS

Reaction times displayed a strong right-skew, as shown in the overall histogram:



- Most reactions were under 30,000 ms, with a sharp drop after 10,000 ms.
- A small number of very long responses (outliers above 80,000 ms) pulled the average up.

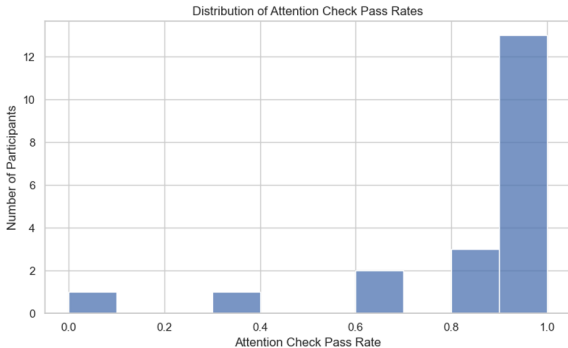
# REACTION TIME BY DISPLAY CONDITION



- 2 Discount Fatigue trials: tightest and fastest response times, with most under 15,000 ms
- 3 Discount Fatigue trials: slightly more spread and a modest increase in slower reactions.
- 4 Discount Fatigue trials: much wider range and higher reaction times, with some peaking well beyond 50,000 ms



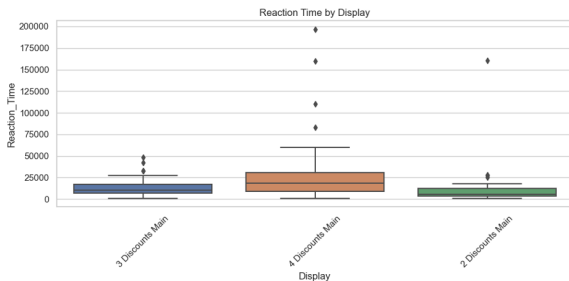
# ATTENTION CHECK PASS RATES



- Most participants performed very well. The majority achieving rates above 80%.
- Only a few participants failed more than one attention check.

# REACTION TIME

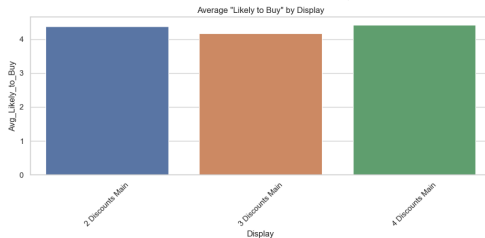
Reaction times increased with the number of discounts



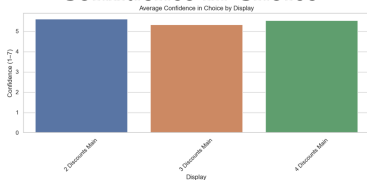
- 2 Discount Fatigue trials: lowest median reaction time and the least variability
- 3 Discount Fatigue trials: slightly higher and more variable response times.
- 4 Discount Fatigue trials: highest median and largest variance, with some extreme outliers reaching over 150,000 ms.

# DECISION-RELATED MEASURES

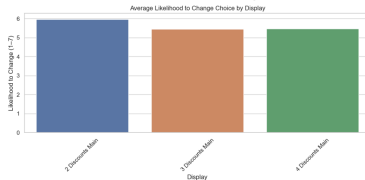
## Likelihood to Buy



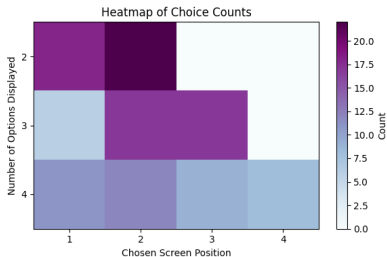
## Confidence in Choice



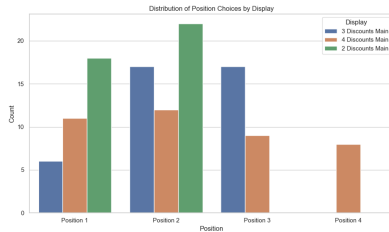
## Likelihood to Change Choice



# DESCRIPTIVE FREQUENCY TABLE OF CHOSEN POSITION



(a) heatmap



(b) histogram

# $\chi^2$ TEST: IS THE OVERALL DISTRIBUTION ACROSS POSITIONS UNIFORM?

- 2 options  $\rightarrow \chi^2(1) = 0.40, p = 0.527$
- 3 options  $\rightarrow \chi^2(2) = 6.05, p = 0.049$
- 4 options  $\rightarrow \chi^2(3) = 1.00, p = 0.801$

# DIRECT TEST OF “EARLIER / CENTRAL” BIAS

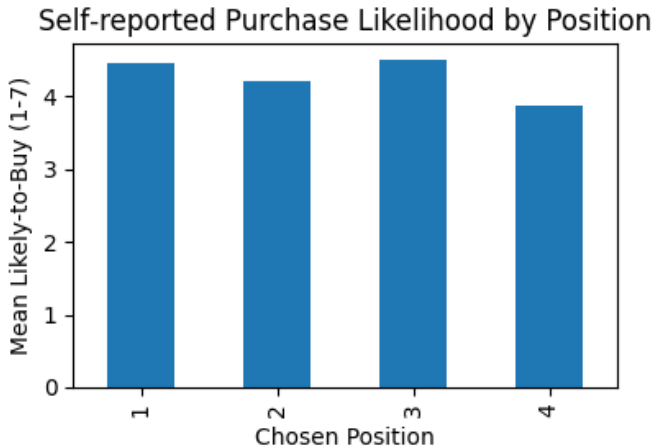
One-sided z-tests vs. chance for bias positions:

- 2 options → 18/40 bias choices ( $\hat{p} = 0.45, p_0 = 0.50$ )  $z = -0.64, p = 0.737$
- 3 options → 23/40 bias choices ( $\hat{p} = 0.57, p_0 = 0.67$ )  $z = -1.17, p = 0.880$
- 4 options → 23/40 bias choices ( $\hat{p} = 0.57, p_0 = 0.50$ )  $z = 0.96, p = 0.169$



# ANOVA TEST

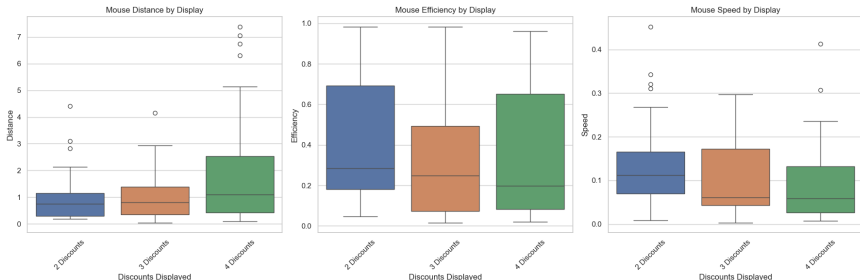
- Reaction Time ~ Position ANOVA  $F = 2.49$   $p = 0.0638$
- Likely to buy ~ Position ANOVA  $F = 0.45$   $p = 0.7159$





# MOUSE MOVEMENT

Mouse movement increased with discount number, but efficiency and speed were not statistically different



One-way ANOVA:

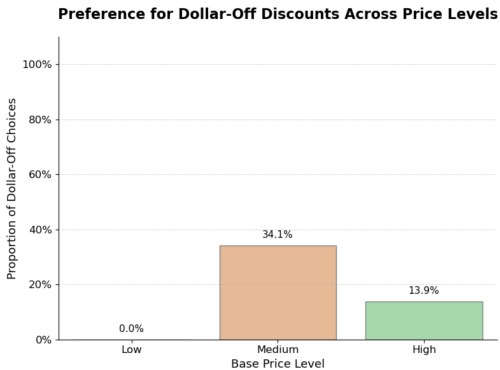
- Distance:  $F(2, 117) = 6.07, p = .003$
- Efficiency:  $F(2, 117) = 0.34, p = .71$
- Speed:  $F(2, 117) = 1.97, p = .14$

# DEAL TYPE DISTRIBUTION

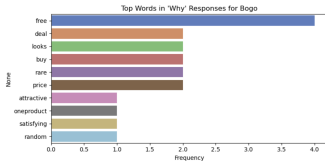


- The most frequently chosen deal type was Percentage off (nearly 60 occurrences), followed by Now price, Dollar off, and BOGO.

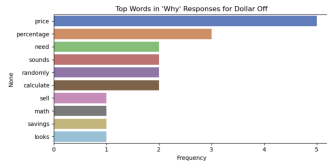
# DEAL TYPE



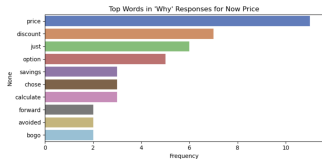
# DEAL TYPE



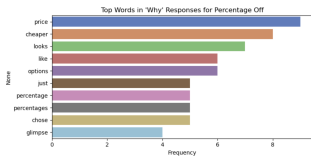
(a) Top Words of Bogo



(b) Top Words of \$ of



(c) Top Words of Now Price



(d) Top Words of Percentage off

Figure: Top Words in "Why" Responses by Deal Type

# DESCRIPTIVE STATISTICS BY OPTION COUNT

	2 opts	3 opts	4 opts
Mean RT (s)	1.82	2.35	2.94
SD RT	0.61	0.77	0.93
Mean mouse distance (px)	540	710	865
SD distance	154	201	242
Valid trials (%)	100	97	95

RT trimmed at  $\pm 2.5$  SD; two timeout trials (8 s cap) removed.

# MANIPULATION CHECK

- One-way ANOVA on RTs:  $F(2, 117) = 5.52, p = .005\%$
- Median RT rises steadily from 2 → 3 → 4 options.
- Confirms choice-quantity successfully induces cognitive load.

# HEADLINE PILOT FINDINGS

- % off frames chosen most ( $\sim 60$  selections)  $\rightarrow$  contrary to H3.
- **Position 2** (central) dominates across all conditions  $\rightarrow$  layout salience matters.
- RT and variance grow with option count, supporting H2.
- Attention-check pass rate high ( $> 80\%$ )  $\rightarrow$  data usable.

# IMPLICATIONS

- Need larger, representative sample to test *framing*  $\times$  *price* interactions.
- Counter-balance numeric magnitude of discounts more tightly.
- Redesign UI to separate ordinal order from centrality.



# FEEDBACK ACKNOWLEDGMENTS

- Henry: Clear labeling of pages, Clarify wording of instructions, and added more explicit attention check
- Yunfei: Clarify that you always make one selection
- Vera: Refer to choices as simply "options"
- Andy, Zhoujun, John: Many expressed frustration about the repetitiveness or confusion about the equal prices, which were all confirmations of our design

Introduction  
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Literature Review  
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Methodology  
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Descriptive Results  
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Pilot Results  
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Summary  
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# THANK YOU

## Q&A