MIND-SHARE OR HEART-SHARE: COMPETING IN THE PURCHASE FUNNEL

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WHY STUDY ATTITUDE METRICS?

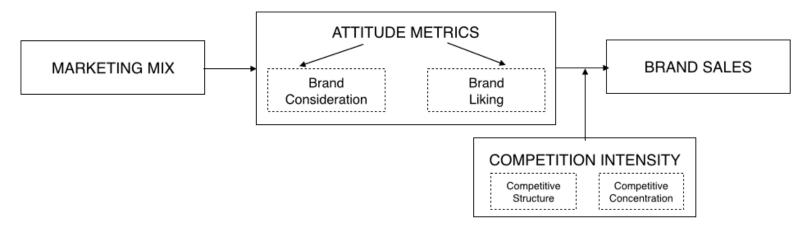
"The firm should monitor 3 variables when analyzing potential threats posed by competitors: share of market, share of mind and share of heart"

—Kotler & Keller 2011

- In efforts to enhance marketing's accountability in the firm, recent research focuses on the linkages of the unobservable consumer attitude metrics to market & financial outcomes
- Research results show consumer attitude metrics to have explanatory power in sales response models as they unlock the black box of marketing-sales effectiveness (e.g., Srinivasan et al. 2010; Hanssens et al. 2014)
- Particularly useful for (chief) marketing officers to understand the attitude metrics to:
 - identify specific remedial marketing actions
 - position their brands against competitive actions
 - monitor their brand performance over time
 - gain early signals of market performance
- Competition has been shown to have a significant role in consumer decision-making
 (Bronnenberg et al. 2000; Keller 1991) and the returns from firms' marketing mix efforts
 (Keller 1987; Kumar 2000; Unnava and Sirdeshmukh 1994): the common wisdom suggests that
 competition is not desirable for firm performance

WHAT DO WE STUDY?

- Referring to information processing and attitude research, we examine the process wherein marketing mix efforts generate sales through influencing consumer attitude metrics (Abelson and Levi 1985; Bettman 1979; Negungadi 1990)
- In response to *Srinivasan et al.'s (2010)* call for future research, we study an important question left unanswered: the role of *competition intensity* on the conversion of marketing actions to sales through consumer attitude metrics
- We analyze how the mediated effects of consumer attitude metrics on the marketing-sales link are moderated by the competition intensity in each category



MAJOR CONTRIBUTIONS

- This is the first study to provide quantitative evidence for the implications of competition intensity on the sales conversion of firms' marketing mix efforts through consumer attitudes
 - We add to the literature on marketing metrics, which is also of great interest to practitioners by employing on a rich dataset of 148 brands in 6 categories for over 10 years
- We contribute to research on the effects of competition in which the underlying assumption that it is not desirable by showing that competition may both desirable and adverse effects on the marketing mix-sales relationship
 - We show that competition intensity has contrasting effects on the upper-level (brand consideration) versus the lower-level (brand liking) of the purchase funnel
 - Higher <u>number of competing brands</u> in a category:
 - firms' marketing efforts lead to an increase in sales through the cognitive component of consumer attitude (brand consideration), but
 - ...a decrease in sales through the affective component of consumer attitude (brand liking)
 - Higher level of <u>competitive concentration</u>:
 - firms' marketing efforts lead to a decrease in sales through brand consideration, but
 - ...an increase in sales through brand liking

MAJOR FINDINGS

1. Main Effects

Sales Effect	Mindset Effect			
	Consideratio	Liking		
-0.005	-	_		
0.020	-	-		
0.014	0.019	-0.002		
0.030	0.026	0.013		
-0.176	0.036	0.040		
-0.007	0.019	-0.004		
-0.003	0.002	-0.006		
0.046	0.015	-0.035		
	-0.005 0.020 0.014 0.030 -0.176 -0.007 -0.003	Consideratio -0.005 - 0.020 - 0.014 0.019 0.030 0.026 -0.176 0.036 -0.007 0.019 -0.003 0.002		

2. Mediated Effects Conditional on Competition

In markets with **high (low) number of brands**, the change in sales due to **1 unit** change in:

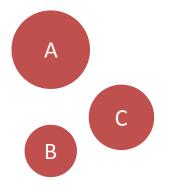
- advertising through consideration is 0.028 (-0.031)
 units; through liking is -0.007 (0.010) units
- promotion through consideration is 0.037 (-0.041)
 units; through liking is -0.006 (0.008) units

In markets with **high (low) concentration**, the change in sales due to **1 unit** change in:

- advertising through consideration is -0.008 (0.005)
 units; through liking is not sig. (0.003) units
- promotion through consideration is -0.011 (0.007)
 units; through liking is not sig. (0.003) units

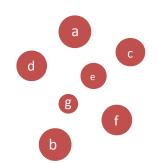
3. Total Effects

In markets with high concentration + low number of brands, 1-unit investment in marketing mix:



- marketing mix through consideration will result in -0.093 units sales decrease
- marketing mix through liking will result in 0.018 units sales increase

In markets with *low concentration + high* number of brands, 1-unit investment in marketing mix:



- marketing mix through consideration will result in 0.077 units sales increase
- marketing mix through liking will result in -0.007 units sales decrease

MANAGERIAL BEHAVIOR

COMPETITION INTENSITY	High # of Competitors	Low # of Competitors
Market with high concentration (Rival brands with heavy efforts)	Firms should carefully invest and manage both consideration and liking <i>(e.g., shampoo)</i>	Sales conversion of brand liking is the highest; invest on liking for higher sales <i>(e.g., Greek yogurt)</i>
Market with low concentration (Rival brands with low efforts)	Sales conversion from consideration is the highest; invest on consideration for higher sales (e.g., bottled water)	Firms should invest and carefully manage both consideration and liking (e.g., organic detergents)

DATA & VARIABLE OPERATIONALIZATION

Data provided by Prométhée (Kantar WorldPanel's brand performance tracker) includes complete weekly observations between Jan 2001 – Dec 2010 over 6 categories on 148 brands: bath gels (18 brands), bottled water (21 brands), dairy (29 brands), laundry detergents (32 brands), milk (24 brands) and soft drinks (24 brands)

Data variables include:

Consumer Attitudes (Srinivasan et al. 2010)

- 1. **Brand consideration** (consumers' mind/ cognitive states) as % of respondents who consider the brand for purchase among all other brands
- 2. **Brand liking** (consumers' hearts/ affective states) using the average of a 10-point scale rating ranging between 'like enormously' and 'not at all'

Marketing mix efforts

- 1. **Advertising** (the gross rating points of the brand)
- 2. **Promotion** (% of the brand sales in value with special display, special offers, special packs and/or price discounts)
- 3. **Price** (average price paid per unit)

Brand sales volume

• The evolution of the sales volume aggregated across all product forms of each brand: % of sales increase in liters or grams compared to that of the same period in the last year

OPERATIONALIZATION OF COMPETITION

Competitive Marketing Efforts

• Average of the marketing efforts of all other brands in the category weighted by marketshare

Competition Intensity (Bell et al. 1999; Gatignon 1984; Montgomery & Wernelfelt 1991; Morgan & Rego 2009; Nijs et al. 2001):

- 1. **Competitive structure** as the number of brands in the category
- 2. **Competitive concentration** as the *Herfindahl index* of promotional actions efforts
 - Sum of squares of the ratio of promotional actions efforts of the all other brands excluding the focal brand to the total efforts of the category
 - Higher (vs. lower) clutter indicates rival brands engage in heavy (vs. low) promotional actions

Calculating competitive concentration in a three-brand market

- Promotion efforts: Brand A: **50** Brand B: **30** Brand C: **20**
- Then, the *promotion-shares* will be: Brand A: **0.5** Brand B: **0.3** Brand C: **0.2**
- Market concentration for:
 - Brand A: $(0.3)^2 + (0.2)^2 = 0.13$
 - Brand B: $(0.5)^2 + (0.2)^2 = 0.29$
 - Brand C: $(0.5)^2 + (0.3)^2 = 0.36$

Therefore, irrespective of the number of competitors, Brand C has more active competitors compared to Brand A and Brand B, in terms of promotional activities.

MODEL

Sales conversion model

- How attitudinal changes are translated into behavioral responses and generate sales
- Competition intensity and promotion clutter moderates the conversion

$$SALES_{t} = \alpha_{0} + \alpha_{1} * SALES_{t-1} + \left[\beta_{11}\beta_{12}\beta_{13}\right] \begin{bmatrix} ADV_{jst} \\ PROMO_{jst} \\ PRICE_{jst} \end{bmatrix} + \left[\beta_{21}\beta_{22}\beta_{23}\right] \begin{bmatrix} ADV_{js(t-1)} \\ PROMO_{js(t-1)} \\ PRICE_{js(t-1)} \end{bmatrix} + \left[\gamma_{11}\gamma_{12}\gamma_{13}\right] \begin{bmatrix} COMP_ADV_{jst} \\ COMP_PROMO_{jst} \\ COMP_PRICE_{jst} \end{bmatrix} + \left[\gamma_{21}\gamma_{22}\gamma_{23}\right] \begin{bmatrix} COMP_ADV_{js(t-1)} \\ COMP_PROMO_{js(t-1)} \\ COMP_PRICE_{jst} \end{bmatrix} + \left[\beta_{21}\beta_{22}\gamma_{23}\right] \begin{bmatrix} COMP_PROMO_{js(t-1)} \\ COMP_PRICE_{js(t-1)} \end{bmatrix} + \left[\beta_{21}\beta_{22}\gamma_{23}\right] \begin{bmatrix} COMP_PROMO_{js(t-1)} \\ COMP_PROMO_{js(t-1)} \end{bmatrix} + \left[\beta_{$$

Mindset/market response model

- How marketing mix efforts influence consumer attitude metrics
- Competition intensity and promotion clutter moderates the conversion

$$\begin{bmatrix} CONS_{jst} \\ LIKING_{jst} \end{bmatrix} = \alpha_0 + \begin{bmatrix} \alpha_1\alpha_2 \end{bmatrix} \begin{bmatrix} CONS_{js(t-1)} \\ LIKING_{js(t-1)} \end{bmatrix} + \begin{bmatrix} \beta_{11}\beta_{12}\beta_{13} \end{bmatrix} \begin{bmatrix} ADV_{jst} \\ PROMO_{jst} \\ PRICE_{jst} \end{bmatrix} + \begin{bmatrix} \beta_{21}\beta_{22}\beta_{23} \end{bmatrix} \begin{bmatrix} ADV_{js(t-1)} \\ PROMO_{js(t-1)} \\ PRICE_{js(t-1)} \end{bmatrix} + \begin{bmatrix} \gamma_{11}\gamma_{12}\gamma_{13} \end{bmatrix} \begin{bmatrix} COMP_ADV_{jst} \\ COMP_PROMO_{jst} \\ COMP_PRICE_{jst} \end{bmatrix} + \begin{bmatrix} \gamma_{21}\gamma_{22}\gamma_{23} \end{bmatrix} \begin{bmatrix} COMP_ADV_{js(t-1)} \\ COMP_PRICE_{jst} \end{bmatrix}$$

ESTIMATION

- Employed three-stage least square (3SLS) (Zellner and Theil 1962):
 - Allowed the errors in different equations to correlate
 - Estimated all equations simultaneously
 (c.f., Bruno et al. 2012; Luo et al. 2012; Van Heerde et al. 2013)
- Used the lagged-level values of endogenous variables as instruments, to avoid endogeneity bias (Arellano and Bond 1991; Tuli and Bharadwaj 2009)
- Applied pooling using the Constant Coefficient Model (CCM), to reduce collinearity and increase model parsimony and degrees of freedom (c.f., Hanssens et al. 2014; Horváth and Wieringa 2008)
- Included variables in their differences (rather than levels) to control for unobserved heterogeneity due to time & category effects (Gruca and Rego 2005; Luo et al. 2010)
- We checked model assumptions with Durbin-Watson, White's, RESET, Jarque-Bera, VIF, & Breusch-Pagan tests
 (Morgan and Rego 2006, 2009; Luo et al. 2012): no violation

STRATEGIC IMPLICATIONS

- If a firm is operating in a market with many players, it may be more beneficial for firms to be amongst the brands that are considered by consumers
 - Because the higher the alternative brands, the higher the consumers' motivation to simplify information processing and limit purchase alternatives to brands in their restricted consideration set
 - Hence, firms may need to develop products with visible higher benefits than substitutes, and communicate these benefits of their products through informative advertising
- If a firm is operating in a concentrated market in which a few strong players invest heavily in promotions, then it is better to have the brands that are most liked by consumers
 - Because the intensified information content may lead consumers to disproportionately put more weight on unfavorable information in competing ads and attach more to the brand they like, making brand liking a shield against all other brands
 - Hence, firms may need to create emotional ads, build self-brand connections and develop brand personalities to match with the consumer expectations