Responses to rival exit: Product variety, market expansion, and preexisting market structure

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Research Summary: This study investigates incumbent responses to a main rival's exit. We argue that long-time rivals have developed an equilibrium by offering a mix of overlapping and unique products and by choosing geographic proximity to each other. A rival's exit, however, disrupts this equilibrium and motivates surviving firms to expand in both product and geographic spaces to seek a new equilibrium. Using data from all U.S. Best Buy stores before and after the exit of Circuit City, we find that Best Buy uses product variety expansion as its major response in markets where Circuit City was colocated, but it more often responds by opening new stores in non-colocated markets. Regardless of preexisting market structures, the magnitude of product variety expansion decreases with the opening of new stores.

Managerial Summary: How do surviving firms respond to a major rival's exit? By studying Best Buy's responses to Circuit City's withdrawal, we find the survivor expands in both product space (increasing product variety) and geographic space (opening new stores), due to two motives. First, the survivor strives to fill in "holes" left in the market. Second, the survivor experiences uncertainty in the postexit world wherein its reference point is gone, threat of potential entry looms, and it lacks information about new entrants. Thus, it must deter potential entry ex ante by preempting many prime product and geographic locations. Best Buy also responds according to preexisting market structures, primarily through product variety expansion in markets wherein Circuit City was colocated and through opening new stores in non-colocated markets.

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1 | INTRODUCTION

Firm exit is a common phenomenon across all industries (Porter, 1976). For example, in manufacturing industries, about 8.5–10% of firms exit in an average year (Dunne, Roberts, & Samuelson, 1988). The rate is found to be higher in banking (Wu & Knott, 2006) and substantially higher in retail trade (Jarmin, Klimek, & Miranda, 2004). In recent years, the retail industry has witnessed an escalation of exits, including those of well-known names such as Borders, Blockbuster Video, Circuit City, and Linens ‘n Things. Like market entry, the departure of major players from a market—especially a concentrated one in which only a few firms dominate—often causes dramatic changes in the competitive landscape and has strategic implications for surviving incumbents. Despite the prevalence and importance of market exit, however, extant literature has remained relatively silent on the subject, and no study has ever addressed how surviving firms respond to the exit of a main rival. Existing studies that examine market exit usually focus on exit decisions and their determinants (Chang, 1996; Chang & Wu, 2014; Lieberman, 1990; Lieberman, Lee, & Folta, 2017).

This lack of investigation into firm responses to rival exit is surprising, especially in light of the enormous body of adjacent literature on incumbent responses to market entry (Du, Li, & Wu, 2018; Geroski, 1995; Seamans & Zhu, 2013; Wang & Shaver, 2014). Notwithstanding the useful insights extant research offers on incumbent strategies in response to entry, we feel the special nature of exit requires the development of additional theory and insights because exit is not just a mirror image of entry. Although most literature views entry as a threat to incumbents' market share and profitability (for an exception, see McCann & Vroom, 2010), we cannot assume exit provides only opportunities for surviving incumbents. There is no question that a rival's exit offers a chance for surviving firms to capture the departing rival's customers, employees, suppliers, facilities, geographic locations, and even complex knowledge or innovations (Hoetker & Agarwal, 2007). However, rival exit can also pose a significant threat to surviving incumbents: the uncertainty associated with the disruption of an interdependent equilibrium. When rival firms have long-term and frequent interactions in a market, they serve as each other's reference points in their competitive space and resemble each other closely in their actions (Fiegenbaum & Thomas, 1995), thereby developing an interdependent equilibrium relationship. In particular, firms strive to achieve equilibrium by offering a certain mix of overlapping and unique products (Ren, Hu, Hu, & Hausman, 2011; Seamans & Zhu, 2013) and by locating close to or far from each other (Chung & Kalnins, 2001; McCann & Vroom, 2010). The equilibrium that rival firms develop with each other reinforces their competitive positions and creates barriers to entry (Lieberman & Asaba, 2006).

The exit of a main rival breaks this equilibrium. In the absence of a familiar “roommate” (Barnett, 2006) who shares the market and a history of repeated interactions, the surviving incumbent must face the uncertainty of having to deal with a possible new “roommate” who might be a total stranger. How, then, do surviving incumbents respond to rival exit? In this paper, we focus on the retail industry and examine both theoretically and empirically how surviving firms respond to this
disruption with two nonpricing strategies: product variety and market expansion. Product variety has become an increasingly important strategy, which is especially true in the retail industry wherein the prevalence of price match guarantees constrains retailers from using price reductions to attract consumers (Hess & Gerstner, 1991). Despite the cost of expanding product variety (Randall & Ulrich, 2001; Zhou & Wan, 2017a), a greater variety allows firms to serve heterogeneous customer preferences better (Sorenson, 2000), prevent customers from switching to competitors (Klemperer, 1995), and preempt competitive entry (Lancaster, 1998). Market expansion through the opening of new stores\(^1\) incurs higher fixed costs and may cannibalize the sales of existing stores (Kalnins, 2004), yet retailers still commonly use this strategy to gain market share and increase profits (Helfat & Lieberman, 2002). Moreover, by preemptively occupying prime geographic locations, expansion limits the amount of space for subsequent entrants and helps deter competitive entry (Lieberman & Montgomer, 1988).

We posit that in response to its rival's exit, a surviving firm expands in product space by increasing its product variety and in geographic space by opening new stores. The motives for the expansion are twofold. One motive is opportunity. Consistent with extant literature that emphasizes incumbent responses involving the pursuit of new opportunities (de Figueiredo & Silverman, 2007; Greve, 1995), the survivor is tempted to exploit the opportunity created by the withdrawal of the rival by filling in the “holes” left in the market. The other motive is uncertainty, which is a unique feature of the rival exit setting. The survivor is temporarily situated in a postexit world wherein its reference point is gone, threat from potential entry is looming, and it has little information about how the new entrants will behave. Out of this uncertainty and fear of the unknown, the survivor needs to preempt many locations in product and geographic spaces to deter potential competitors' entry ex ante.

We further posit that a surviving firm's postexit expansion is contingent upon the preexisting market structures it developed with the rival, because the benefits and costs associated with implementing each strategy vary when preexisting market structures are in place. We categorize a preexisting local market as either a colocated market or a non-colocated market, according to whether the nearest rival is colocated (i.e., within walking distance) or not. In the retail industry, direct rivals such as CVS and Walgreens, Best Buy and Circuit City, and Home Depot and Lowe's often locate their stores in the same shopping plaza or in very close proximity (e.g., across the street). Such colocation indicates the availability of customer-rich locations in a market (Wang & Shaver, 2014) and may affect the preexisting pattern of product offerings “coordinated” by rivals (Ren et al., 2011). We argue that while the costs of offering more product variety are comparable across market types, the benefits of doing so are higher in colocated markets. In non-colocated markets, opening new stores brings more revenue to the surviving firm by covering the other side of the market, while it incurs significantly lower cannibalization costs. We thus predict that product variety expansion is a more appealing strategy within colocated markets, while opening new stores is a more optimal strategy within non-colocated markets.

We have formalized these arguments by developing a simple analytical model that allows us to consider the benefits and costs of both strategies in different market structures. Based on this model, we have developed a set of hypotheses and tested them using data from the two specialty consumer electronics retail chains in the United States: Best Buy and Circuit City. This research setting is excellent for three reasons. First, both chains provided price match guarantees, so their competition

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\(^1\)A retail chain's market-entry strategy may include adding stores in a geographic market in which it is an incumbent, or in new geographic markets. Chevalier (1995) labels the former as expansion and the latter as de novo entry into a local market. Compared with de novo entry, expansion requires less time and planning (Chevalier, 1995), and it represents most of the store-opening cases in the context under investigation. We thus focus on expansion in this research.
focused less on pricing than on nonpricing dimensions. Second, we have appropriate and reliable data for use in our analysis. To facilitate consumer search, both Best Buy and Circuit City released their local store inventory information online and updated it in a timely manner (Levinson, 2005). As a result, the in-store product variety data of both chains was readily available. Advanced computer search and data collection techniques allowed us to retrieve and compile the information with enhanced accuracy as compared to manually collected data. Third, and perhaps most importantly, Circuit City’s rapid and nonselective exit from all geographic markets gave rise to exogenous changes in the structure of all of these markets, which significantly alleviates the empirical concern that market structure changes might be endogenous (Berry & Waldfogel, 2001). This enables us to not only study Best Buy's actions over a period of time spanning before and after Circuit City's demise, but also to explore cross-sectional variations across geographic markets.

We found that after the exit of Circuit City, Best Buy increased its store-level product variety, and the magnitude of increase was greater in colocated than in non-colocated markets. We also found that Best Buy was more likely to expand its market coverage by opening new stores, and that it engaged in such market expansion more aggressively in non-colocated markets than in colocated ones. Furthermore, regardless of preexisting market structures, the magnitude of product variety expansion decreases when new stores open. This suggests that while both product variety and convenient new store locations are effective strategies to attract consumers and fend off new entrants, there is a substitutive relationship between them.

This study makes four major contributions. First, it provides the only comprehensive analysis of incumbent responses to rival exit that has yet been conducted. We highlight the importance of market exit as a strategic phenomenon that offers unique challenges and opportunities for surviving firms. Extant research on entry, which represents the most adjacent stream of work, has advanced our understanding of when and how incumbents will respond to market entry. The major insights from recent studies in this area include that incumbent responses to entry depend on characteristics of the market (Seamans & Zhu, 2013; Wang & Shaver, 2014), characteristics of incumbents (Semadeni, 2006; Simon, 2005), or characteristics of entrants (McCann & Vroom, 2010; Seamans, 2013). Above all, response to entry is a defense story in that incumbents are incentivized to defend their current market share, positioning, and profitability. In contrast, incumbent response to rival exit is a story of expansion and preemption. Surviving incumbents expand to pursue new opportunities, as suggested by extant literature (de Figueiredo & Silverman, 2007; Greve, 1995; Wang & Shaver, 2014). They may expand excessively by preempting many prime locations in product and geographic spaces, driven by the uncertainty associated with the disappearance of a reference point, threat from potential entrants, and lack of knowledge of future competitors' strategies and actions.

Second, among the various strategic options that firms may pursue in response to a rival's exit, we deviated from examining pricing response, which has been the traditional focus of competitive strategy scholars, to join a recent stream of work (Bennett, Pierce, Snyder, & Toffel, 2013; Simon, 2005) that explores nonpricing responses. Moreover, similar to Casadesus-Masanell and Zhu (2010), we selected multiple nonpricing responses that involve different levels of commitment and allowed them to interact. The substitutive relationship we found between the two responses-increasing product variety and opening new stores—provides further evidence of the limits of firms' coordination capacity (Zhou, 2011; Zhou & Wan, 2017b).

2Best Buy and Circuit City had long been perceived as duopoly competitors since the 1990s. However, due in part to waning consumer spending, tight credit markets, and the economic crisis, Circuit City filed for bankruptcy in November 2008 (Rosenbloom, 2009). Many bankrupt retailers close stores gradually or in select markets. Circuit City instead shut down its retail stores quickly and relatively indiscriminately in all geographic markets; its last brick-and-mortar store closed on March 8, 2009.
Third, we emphasized the role of preexisting market structures in determining a firm's postexit responses. Our theory and empirical evidence demonstrates that market structure distinguished by geographic proximity has a strong influence on how firms respond to a competitor's exit: The surviving firm tends to resort to product variety expansion as its primary response in colocated markets, but it will more likely respond by opening new stores in non-colocated markets. Prior studies have argued that incumbent responses to entry are affected by the distribution of customer preferences (Seamans & Zhu, 2013; Wang & Shaver, 2014). Our findings suggest the possibility that preexisting market structures can be used to infer the distribution of customer preferences in geographic spaces. We also extended the work by Ren et al. (2011) by uncovering an important factor driving postexit product variety expansion: the equilibrium a firm has developed with its rivals in preexisting market structures.

Fourth, by focusing on drastic changes in market structure, our study adds important insights to the multidisciplinary literature pertaining to the relationship between market structure and product variety. Despite some relatively abundant theoretical contributions relating to structure and variety (Lancaster, 1990, 1998), empirical research remains scant, perhaps due to the endogeneity of market structure. Recent research (Olivares & Cachon, 2009; Watson, 2009) within this domain focuses on cross-sectional variations. Berry and Waldfogel’s (2001) study of radio broadcasting stations is the only empirical investigation that addresses how a radical change in market structure affects product variety. We have extended their pioneering research by exploring rival exit as another trigger of market structure change (vs. exogenous policy change, which their study deals with) and by focusing on store-level product variety (vs. market-level variety, as their study explores). We will detail all four contributions in Section 5.

2 | THEORY AND HYPOTHESES

2.1 | Theoretical background and model overview

Our research emerged from the understanding that rival firms have developed an equilibrium by coordinating their level of overlapping products and by choosing to colocate or locate at a distance from each other (Ren et al., 2011; Seamans & Zhu, 2013), and that this equilibrium helps preserve the status quo of the rivals and fend off new entrants (Clarkson & Toh, 2010; Lieberman & Asaba, 2006). We argue that the exit of a main rival disrupts the earlier equilibrium achieved by rival firms. Without the rival, a surviving firm seems to enjoy a temporary monopoly. However, this temporary monopoly is fraught with uncertainty and threat, as it attracts potential new entrants with whom the surviving incumbent has not yet interacted, thereby putting the incumbent at an informational disadvantage (Barnett, 2006; Seamans, 2013). Against this background, we focus on a set of duopoly retailers and develop a simple analytical model to examine how a surviving firm repositions itself in both product and geographic spaces to respond to the departure of its rival. Because they differ in their level of commitment, these two responses—those made in product and in geographical space—may not follow a simple sequential order. We argue that prior to engaging in new store openings, a surviving retailer would consider adjusting its product variety, as this alternative entails significantly lower costs than opening a new store, can be implemented more quickly, and is more reversible. After opening a new sister store, the retailer would re-optimize the existing store's product variety.

We provide an overview of our formal model by highlighting the key assumptions and insights derived from it. We built our model by letting the surviving store choose its optimal level of product variety to maximize its store profit. Relatedly, our model is based on a consumer demand function in
which a representative consumer derives her utility from product varieties. We note two assumptions in regard to our demand function. One assumption is that prices are the same across rivals.\(^3\) In light of the prevalence of price match guarantees in the retail industry, we restricted stores from using price differences to attract customers. In the second assumption, we decomposed a store’s product variety into overlapping and unique products. Overlapping products demonstrate a certain level of similarity between rival firms, which sends signals to stakeholders that these firms conform to industry norms and are legitimate players in a market, while unique products differentiate a firm from its rival and thereby serve to reduce competition (Deephouse, 1999). Consistent with existing studies (Dobrev, 2007; Lynch Jr. & Ariely, 2000), we assume unique products contribute more to store profits than do overlapping products.

In our model, we considered the benefits and costs of adding product variety following a rival’s exit. Adding retail variety incurs costs such as inventory, transportation, setups, change-overs, and opportunity costs of shelf space (Ryzin & Mahajan, 1999). Compared to the cost side, the benefits of postexit product variety expansion are more nuanced and have strategic implications. First, in the case of a duopoly, any additional products a store considers carrying are likely to be overlapping products. After the exit of the rival, however, any additional products the store carries become unique ones. No longer facing the discounting constraint of product overlap, the store has incentives to increase products, as it now can enjoy a higher level of marginal profit for carrying additional products. Second, when a firm lacks knowledge about the entry plans of potential rivals, having a broader product variety can establish barriers to entry, in that the incumbent firm has preemptively filled all potential market gaps in which a new firm could have entered (Lancaster, 1998). If a new firm decides to enter the market, the products it offers are more likely to be overlapping ones when the surviving store has already expanded its variety, which reduces the entrant’s postentry profit.

While the costs of postexit product variety expansion do not vary much between non-colocated and colocated markets, the benefits are greater for stores in colocated markets. First, the inferiority of overlapping products in terms of contribution to store profit is more severe in colocated markets. As a result, the profit enhancement gained by switching from overlapping to unique products is greater for stores in colocated markets. Second, because colocation results in fiercer competition, colocated rivals are more likely than distant rivals to differentiate themselves by carrying fewer overlapping products, yet each carries a smaller product variety to save on stocking costs (Ren et al., 2011). As such, a colocated market provides more room to offer unique products and appears more attractive to potential rivals, thereby posing a more serious threat to the surviving store. Together all these arguments suggest that a surviving store in a colocated market has more incentive to increase its product variety than a surviving store in a non-colocated market. To share a better understanding of the nuances of the benefits, Appendix S1 illustrates different patterns of preexisting product variety and postexit expansion in non-colocated and colocated markets.

We also model the cost–benefit tradeoffs in store opening to illuminate the circumstances under which the surviving retailer opens a new store. Although opening a new store is rarely financially trivial (Zhu, Singh, & Manuszak, 2009) and can cannibalize sales of existing stores (Kalnins, 2004), it allows the surviving retailer to occupy the vacant side of the market and deter potential competitors from entry (Liu, Gupta, & Zhang, 2006). Moreover, opening new stores enables the retailer to redeploy its resources and capabilities, including by replicating operating procedures or store configurations, sharing supply channels, distribution and advertising, and applying market-specific knowledge.

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\(^3\)To validate the claim, we collected price data on digital cameras available in March 2006 for all Best Buy and Circuit City stores. A paired \(t\)-test on the prices of digital cameras showed no significant difference in prices \((p = 0.53)\). The average paired price difference between the two chains is \(0.23\%\) of the price, and \(47\%\) of the matched products have the same prices at both chains.
(Helfat & Lieberman, 2002). We take these benefits and costs into account in our model and make a convenient and plausible assumption that the new store will be located at the same site as the exiting store. Our model suggests that the net gain of a new store opening is greater in non-colocated markets. Why? When preexisting rival stores were located farther from each other, each store held a geographical advantage for consumers in the proximate part of the market. A new store opened in such a market is thus more likely to generate incremental sales by covering that side of the market without incurring high cannibalization costs. As cannibalization risk significantly escalates with geographic proximity (Pancras, Sriram, & Kumar, 2012), the negative impacts of cannibalization surpass all possible benefits in colocated markets.

Our model further shows how the surviving retailer re-optimizes its product variety when a new sister store opens in the market. The objective now becomes maximizing the joint profits of the surviving store and the new sister store by determining both stores' optimal levels of product variety and level of product overlap. Both decisions involve high coordination costs. Offering more product variants alone increases the difficulty of coordinating inventory management and increases the odds of stock-outs and overstocking (Ryzin & Mahajan, 1999), which reduces customer satisfaction and hurts store sales and future demand (Zhou & Wan, 2017a). To find the optimal level of overlap, the surviving firm needs to balance a number of factors including the joint revenues of both sister stores, the fixed cost of opening a sister store, the economy of scale benefit enjoyed by sister stores, and the cannibalization risk associated with overlapping products. The complexity of coordination might make exercising both strategies concurrently too costly for the firm (Zhou & Wan, 2017a, 2017b), which may lead the retailer to use them as substitutes.

2.2 Baseline model: Product variety in a preexisting duopoly market

We first use the classic differentiated duopoly model of Singh and Vives (1984) to model consumers' purchasing decisions across two stores: Store 1 of the surviving firm and Store 2 of the rival. Store i's (i = 1, 2) product variety, ti, is presented as ti = vi + vo, where vi denotes the number of unique products carried by Store i and vo denotes the number of overlapping products carried by both stores. Parameter V denotes the product variety universe comprised of the total available products offered by the manufacturers from which the stores choose. The value that Store i's product variety provides for a consumer is modeled as \( \Phi_i = \phi(v_i + mvo) \), with \( \phi > 0 \) reflecting the fact that more product variety brings higher value to consumers (Lancaster, 1990), and m captures a utility discount (\( \frac{1}{2} < m < 1 \)).
because consumers can derive only a fraction of the value from overlapping products that they can gain from unique products. The utility discount of overlapping products is more severe for colocated stores \( (m_i) \) than for non-colocated stores \( (m_j) \). In other words, \( m_i < m_j \). As consumers can visit both stores without incurring transport costs and may purchase from either, the two stores are akin to one big store to consumers. The joint value of overlapping products at both stores is almost the same as the value when these products are available at only one store (i.e., \( m \to 1/2 \)).

In a duopoly market with two stores, a representative consumer's utility function is given by

\[
u = \alpha_1 q_1 (\Phi_1 + 1) + \alpha_2 q_2 (\Phi_2 + 1) - \frac{1}{2} \left( \beta_1 q_1^2 - 2 \gamma q_1 q_2 + \beta_2 q_2^2 \right) - pq_1 - pq_2, \tag{1}\]

where \( p \) is price, \( q_i \) is demand of store \( i \), \( \alpha_i > 0, \beta_i > 0, \gamma \geq 0, \beta_1 \beta_2 - \gamma^2 > 0 \), and \( \alpha \beta_j - \alpha \gamma > 0 \) for \( i = 1, 2 \) and \( j = 3 - i \). Parameter \( \gamma \) measures the competition effect between stores. \( \alpha q_i (\Phi_i + 1) \) captures two aspects of a consumer's utility from purchasing \( q_i \) units of products from store \( i \). First, regarding the quantity of purchased products \( \alpha q_i \), a consumer's utility increases as she obtains more units of products from the store, given the store's fixed product variety. Second, regarding the quality of selected products \( \alpha q_i \Phi_i \), a consumer's utility increases with the store's product variety because higher product variety makes it more possible for the consumer to find suitable products to satisfy her needs. The consumer's decreasing marginal utility from purchasing quantity \( q_i \) is reflected by the quadratic term \( \frac{1}{2} (\beta_i q_i^2) \), wherein the coefficient \( \frac{1}{2} \) is added for mathematical convenience. The consumer chooses quantities \( (q_1 \text{ and } q_2) \) to purchase from Stores 1 and 2 to maximize her utility, \( u \). Following the way of rewriting by Singh and Vives (1984), we simplify the demand functions as follows:

\[
\begin{align*}
q_1 &= a_o + a(v_1 + mv_o) - e(v_2 + mv_o) - (b-d)p \\
q_2 &= a_o + a(v_2 + mv_o) - e(v_1 + mv_o) - (b-d)p
\end{align*}
\tag{2}
\]

where \( a > e \) because it is reasonable to assume that a store's product variety has a higher impact on its own demand than on the other store's demand.

We solve for the optimal levels of product variety in a duopoly market. The cost for Store \( i \) to carry a product variety of \( t_i = v_i + v_o \) is given by \( c r_i^2 = c(v_i + v_o)^2 \), which suggests that Store \( i \)'s cost of carrying products increases with product variety and becomes 0 if product variety is given by 0. We use a quadratic cost function to reflect the increasing marginal cost of carrying more product variety (Mas-Colell, Whinston, & Green, 1995). We assume that the cost of carrying products is independent of the composition of the variety. Because the setting we are examining is retail, we assume zero marginal production cost for stores. Store \( i \)'s profit is thus given by

\[
\Pi_i = pq_i - c(v_i + v_o)^2 = p \left[ a_o + a(v_i + mv_o) - e(v_j + mv_o) - (b-d)p \right] - c(v_i + v_o)^2. \tag{3}\]

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6When a product is uniquely available at one store, a consumer can visit only that store to purchase the product. In contrast, when a product is available at both stores, the consumer can visit either store. Therefore, a unique product at store \( i \) increases that store's attractiveness to a consumer to a greater extent than does an overlapping product.

7For \( \Phi = 0 \), our model will revert back to the pricing model of Singh and Vives (1984) because it does not capture the effect of a consumer's utility increase from higher product variety.

8Like Singh and Vives (1984), we assume \( a_o = 1, a = a, b_i = b \), and \( e_i = e \). They assume that consumers have the same elasticity to price across firms. Since firms' decision variable in our model is product variety, our assumption implies that consumers have the same elasticity to product variety. See Appendix S2 for the details of the rewriting and simplification.

9This assumption fits settings such as consumer electronics retail stores wherein the sizes of products do not significantly differ from each other (Ryzin & Mahajan, 1999). We thank an anonymous reviewer for suggesting the addition of this assumption.
Consistent with Moorthy (1988), we assume Store 1, the surviving store, to be the market leader and move first in deciding on its product variety $t_1$ before Store 2 decides on $t_2$. Lemma 1 below summarizes the results in which the conditions for $cV$ ensure the existence of preexisting product overlap. Appendix S2 provides proof of all lemmas and propositions.

**Lemma 1:** In a duopoly market wherein the product variety universe is fully covered by both stores, for $\max\left\{\frac{c}{1}, \frac{pam + pe(1-m)}{2}\right\} \leq cV \leq pam + pe(1-m)$, the optimal levels of product variety are given by:

$$t_D^1 = t_D^2 = \frac{p}{2c}(am + e[1-m]).$$

(4)

### 2.3 Product variety change in the postexit market

We now solve for the optimal level of product variety of the surviving store after the rival exits the market, assuming the surviving firm does not open any new stores in the market. If the surviving firm does not open a new store in the postexit market, only Store 1’s products will be available for the consumer to purchase. The consumer’s utility function in Equation (1) consists of $q_1$ only. Accordingly, the demand function for the surviving store is given by:

$$q_1 = a_o + at_1 - bp.$$  

(5)

Note that when there is only one store in the market, $m = 1$ since all of the products carried by Store 1 are unique in the market. Solving Store 1’s profit maximization problem leads to the optimal level of product variety for Store 1, as follows:

$$t_S^1 = v_S^1 = \frac{pa}{2c}.$$  

(6)

To investigate how Store 1 adjusts its product variety after the rival’s exit, we compared the optimal product variety of the postexit single-store case ($t_S^1$) with that of the preexisting duopoly case ($t_D^1$). As shown in Proposition A1 in Appendix S2, $t_S^1 > t_D^1$, as long as product overlap existed prior to rival exit. Whereas a surviving store tends to increase its product variety following the exit of its rival, the extent of product variety expansion differs by market structure. The difference in product variety increase between a colocated market and a non-colocated one is given by

$$(t_S^1 - t_D^1) - (t_S^n - t_D^n) = \frac{p}{2c}(a-e)(m_n - m_c),$$

which is positive because $a > e$ and $m_n > m_c$. We therefore submit the following hypotheses:

**H1A.** After the exit of the rival store, a surviving store increases its product variety.

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10Two points are noteworthy regarding our theoretical model. First, we focus on unique pure strategy equilibrium in this study. Second, in a duopoly market, there are two possible scenarios: some level of product variety overlap exists between the two stores (Scenario 1), and no overlap exists in product variety (Scenario 2). Since it is common in many industries for rival stores to hold at least some overlapping products, our model and empirical analysis focus on Scenario 1, in which there exists some level of overlapping product variety (i.e., the product variety universe is fully covered by both stores). The analysis of the zero-overlap scenario is also provided in Appendix S2.

11$c = \frac{2m(a-e)^2 - 3m^2(a-e)^2 - 2am(a-e) + 2ae - e^2}{4[1 + (1-m)(a-e)]^{\frac{1}{2}}}$.
2.4 New store opening in the postexit market

We have modeled the benefits and costs associated with a new store opening in the postexit market. First, the new store generates incremental sales for the retailer. Second, the surviving retailer pays a fixed cost of $F$ for opening a new store. Third, we account for both the cannibalization risk, $m$, between overlapping products across both stores (Kalnins, 2004) and the economy of scale benefit, $w$, of carrying products that overlap with those of sister stores (Helfat & Lieberman, 2002). Fourth, we consider the competition effect, $e$, between sister stores.

Suppose a store incurs a cost of $c(v_i + wv_o)^2$ for carrying a product variety of $t_i = v_i + v_o$. The parameter $w$ ($\frac{1}{2} < w \leq 1$) captures the benefits of economy of scale when a firm's sister store sells overlapping products. $w > \frac{1}{2}$ suggests that the marginal cost of carrying an overlapping product is higher than half of the marginal cost of carrying a unique product, and $w = 1$ refers to the duopoly case of rival stores. Since two geographically proximate sister stores are more likely to both benefit from economy of scale and suffer cannibalization risks than two distant sister stores, both $w$ and $m$ are smaller for closer stores.

In contrast to two rival stores, two sister stores can fully coordinate with each other to maximize their total profit in the market. In the analysis of sister stores, we assume that the new store will stock unique products before carrying any overlapping ones. Lemma A1 in Appendix S2 shows the conditions under which the surviving firm would choose to open a new store. It suggests that the firm may not always want to open a new store, even when the fixed cost of opening a new store, $F$, is minimal. Proposition A2 summarizes the results of Lemma A1 and concludes that the surviving firm is less likely to open new stores when the cannibalization effect between overlapping products is stronger. Since the cannibalization effect is stronger for colocated sister stores than for non-colocated sister stores ($m_c < m_n$), we submit the following hypothesis:

H2. Following the exit of the rival store, the surviving firm is less likely to open new stores in a colocated market than in a non-colocated market.

2.5 Product variety change moderated by a new store opening in the postexit market

To study how the surviving store's product variety increase is affected by opening a new store, we compared the store's product variety increase in the case wherein it opens a new store $(t_i^T - t_i^P)$ to the case wherein it does not open a new store $(t_i^S - t_i^P)$. Lemma A2 summarizes the optimal levels of product variety when the surviving firm operates sister stores $(t_i^T)$.

The detailed analysis of Lemma A2 and Proposition A3 shows that $(t_i^T - t_i^P) - (t_i^S - t_i^P) < 0$, suggesting that product variety increase at a surviving store accompanied by a sister store is smaller than that of a standalone surviving store. The rationale is as follows. After the rival's exit, if a sister store opens in the market, the surviving retailer needs to decide on both stores' product varieties, as well as the level of overlapping products, to maximize the joint profits of both stores. Coordinating the optimal level of overlap between sister stores is not easy: The firm needs to

---

12This is true as long as the cost parameter, $w$, is no smaller than the parameter, $m$, that denotes the utility discount from product overlap and measures cannibalization in the case of sister stores.
balance its resources (product variety universe) and constraints (carrying cost of product variety) while taking into account the complex benefit–cost tradeoffs including additional revenues from more product offerings, economy of scale benefits, and cannibalization costs associated with overlapping products. The coordination difficulty increases even more when more product variants are offered at each store. Although both product variety and convenient new store locations are effective means of attracting consumers, such coordination challenges might make it too costly for the firm to adopt both strategies simultaneously. The results lead to our third hypothesis:

**H3. When the surviving firm opens a new store after the rival's exit, its existing store increases product variety to a lesser extent than when the firm does not open a new store.**

### 3 | EMPIRICAL ANALYSIS

#### 3.1 | Data

We assembled our data set from four major sources. First, we obtained the store locations of every Best Buy and Circuit City retail store in the United States from their websites, www.bestbuy.com and www.circuitcity.com. Additionally, we used Python to write web crawlers to acquire product variety data from the Best Buy website on the digital cameras available at each store. We chose digital cameras as the focal product category for several reasons. The first is the importance of digital cameras to the consumer electronics retail industry as well as to Best Buy. Digital cameras were one of the hottest markets for consumer electronics products in 2006 (Chen & Xie, 2008) and continued to grow until 2010, according to the digital cameras shipment history from the Camera and Imaging Products Association. Best Buy has consistently highlighted digital cameras as a major product category in its annual reports. Furthermore, product variety is highly valued in this category because consumers choose digital cameras based on a list of key product attributes (such as brand, optical vs. digital zoom, camera type, and megapixels), with respect to which their tastes and preferences vary substantially. Moreover, digital cameras require a reasonably high inventory cost (e.g., shelf space, warehouse cost, inventory turnover speed) that prevents stores from expanding product variety without limits. In addition, we were able to acquire product variety data for digital cameras with a high level of precision and accuracy. In some consumer electronics categories, manufacturers routinely provide slight attribute differences and different model numbers across retailers in an attempt to thwart price match guarantees. In extreme cases, they even put different model numbers on identical products. We did not observe such a practice in the digital camera category within the U.S. market, however.

We monitored our web crawlers closely in case changes occurred to the website. For example, we used 20–30 computers to extract web pages simultaneously so that the data crawling processes finished on the same day to minimize any possible changes in the product information. Each round of data collection took roughly 8 hr. Before formally launching the crawling processes, we interviewed several store managers, who confirmed the consistency between their actual in-store product variety and the information listed online. After the data was crawled, we checked the accuracy of the data crawl by visiting several local stores and manually collecting product variety information for digital cameras. We found no difference between the hand-collected information and the data collected with the crawling program. We collected the product variety and store location data in March.
2006, when Circuit City was still profitable, and again in July 2010, when Best Buy had fully responded to the complete exit of Circuit City. Of the 710 Best Buy retail stores that existed in 2006, 709 remained open in 2010 (i.e., the surviving stores). In addition, Best Buy opened 370 new stores between March 2006 and July 2010.

Second, we collected long-term market-level demographics data (i.e., that which had been stable for 10 years) from the Missouri Census Data Center (MCDC). Following the concentric method used by Ren et al. (2011), we delineated a local market by drawing a circle with a 10-mile radius around each store's location, with the presumption that this distance represents a relevant competitive area. With the raw demographic information at the tract level provided by Census 2000, MCDC aggregates the tract-level data to the market level by taking weighted averages, using the tract-level population as the weight. The MCDC website also provides an interface from which we used Python to extract the relevant demographic information of the 10-mile circle surrounding each specified store location.

Third, because local market conditions may have changed rapidly during the 2006–2010 period, we acquired annual demographic information from the American Community Survey (ACS) of the Census. The annual demographic data of the ACS is collected at the county level, and we matched this data with each surviving store's location.

Fourth, to reflect the absolute level of and changes in online retailing prevalence in each market, we collected search data for Amazon.com through Google Trends, which provides data on the relative frequency of search terms entered by Google users across time and geographic units (for applications of Google Trends, see Ginsberg et al., 2009; Hu, Du, & Damangir, 2014). Google Trends provides volume indexes for queries consumers have entered into the Google search engine, free of charge and going as far back as January 2004. Given the ubiquity of consumer online searches and Google's dominance, the volume of Google searches can plausibly be viewed as a reflection of the collective interests of Internet users. We conducted a Google search for Amazon.com at the county level for both 2006 and 2010. As the rate of Internet penetration in a geographic region may influence consumers' adoption of online retailing, we also collected state-level Internet penetration rates for the years of 2006 and 2010 from the Federal Communications Commission and matched the data with each surviving Best Buy store's location.

3.2 | Measures

Before describing our variable measures, we will explain how we define a pair of rivals. Using the latitude and longitude of each store, we calculated the spherical distance between a surviving store and all rival stores. We then selected the rival store closest to the surviving store in distance and defined these two stores as a pair of preexisting rivals.

Our dependent variable is based on the product variety at each surviving store. Following Zhou and Wan (2017a, 2017b), we measured total product variety in each store as the number of stock-keeping units (SKUs) in the digital camera product category. SKU has several advantages as a measure of product variety. First, it serves as the unit for measuring inventory from the retailer's perspective. Second, compared to other measures such as product attributes, SKU is more objective.

13When we first collected the data in March 2006, Circuit City showed no sign of trouble. During the first fiscal quarter that ended on May 31, 2006, Circuit City reported a $6.4 million profit and earned 4 cents per share, exceeding analysts' expectations for four consecutive quarters. Moreover, it expected growth in domestic same-store sales to average between 5 and 7% (Associated Press, 2006).

14We customized Google Trends to extract queries filtered by geographic areas (e.g., metropolitan areas), time ranges (e.g., 2006 and 2010), and categories (e.g., shopping). As an indicator of Google's standing as the dominant search engine in the United States, according to www.Compete.com, Google accounted for 63.8% of online searches in May 2011, handling more than 9.5 billion queries from over 138 million unique users during that month. The reliability of Google search-term data as proxies for underlying economic and social data has been established in a variety of contexts, such as predicting car sales, home purchases, international travel, and religious affiliation (Ginsberg et al., 2009; Hu et al., 2014).
(Fader & Hardie, 1996). When using this measure, researchers are not engaging in a subjective attribute selection process. Moreover, unlike with conjoint analysis or panel data, we cannot directly observe or interpret how much weight consumers place on each researcher-defined attribute. An attribute-based variety measure would have been even less desirable.

Two independent variables measure preexisting market structures. Following Rosenthal and Strange (2003) and Ren et al. (2011), we coded COLOCATE as equal to 1 if the preexisting nearest rival store is within a half-mile radius of the surviving store, and 0 otherwise. We defined NONCOL as equal to 1 if the distance between a surviving store and its nearest former rival store was greater than half a mile but no greater than 10 miles, and 0 otherwise. These two variables describe the two mutually exclusive competitive market structures prior to the exit of Circuit City. Of the 709 Best Buy surviving stores, 397 previously were in non-colocated competitive markets (NONCOL = 1), 186 had colocated rivals (COLOCATE = 1), and the remaining 126 stores used to be in monopoly markets (NONCOL = 0 and COLOCATE = 0). Our third independent variable, ENTRY, specifies whether Best Buy made new entries into a given market after Circuit City exited. We coded ENTRY as equal to 1 if at least one new Best Buy store opened within the 10-mile radius of a surviving Best Buy store, and 0 otherwise.

Further, we controlled for a number of variables that affect long-term market-level demand and consumer taste heterogeneity. Specifically, INCOME (log of median household income in a market) and POPDEN (population density in a market) relate to a market's product variety; the former indicates potential income constraints on purchasing behavior (Hoch, Kim, Montgomery, & Rossi, 1995), and the latter affects the market's demand level and relative profitability (Watson, 2009). Other demographic variables include the fraction of the population with a college-level or higher education (EDUCATION), the fraction of the population over 18 years of age (ADULT), the fraction of the population older than 65 years (SENIOR), the fraction of male residents (MALE), the fraction of married population (MARRIED), and the fraction of consumers who are nonwhite (NONWHITE). All of these demographic variables are based on Census 2000 data computed by MCDC. Some strongly urban areas, such as New York City, Chicago, and Boston, may reflect a different market structure than the rest of the United States. On the one hand, these areas are extremely populated, so rival stores may be more likely to colocate within one mile. On the other hand, because the expensive real estate in these areas causes inventory costs to rise quickly, new store openings and product variety levels might be limited in super-urban areas. To control for the effects of these super-urban regions, we followed Chung and Kalnins (2001) by generating an indicator variable, SUPERURBAN, that equals 1 if the tract-level population density of the focal store (also from Census 2000) is greater than 4,400 people per square mile. We also accounted for each market's outside competition from brick-and-mortar stores by including the logged number of Walmart stores in 2010 (WM) based on the Walmart store location data we retrieved from Walmart’s website. Finally, we controlled for the prevalence of online retailing by adding the 2010 Google search index for Amazon (AMAZON) and the 2010 Internet penetration rate of each market (INTERNET).

Additionally, we included a set of controls that reflect the changes in local market conditions from 2006 to 2010. These change variables are consistent with the static demographic variables that we included. For example, local demographics might change, and Internet retailing may become more common. Using the annual demographic data collected from ACS, we calculated the percentage growth from 2006 to 2010 of population (ΔPOPULATION), income (ΔINCOME), and fraction of adult, senior, male, and nonwhite population (ΔADULT, ΔSENIOR, ΔMALE, and ΔNONWHITE). To control for the changes in both offline and online competition, we also calculated the percentage growth from 2006 to 2010 of the logged number of Walmart stores (ΔWM), search intensity for Amazon (ΔAMAZON), and Internet penetration rate (ΔINTERNET). We calculated ΔEDUCATION
as the percentage growth from 2006 to 2011 because the 2010 data on the fraction of the population with a college-level or higher education was not available. We did not include ΔMARRIED because annual data for this variable is not available in ACS, nor did we include ΔSUPERURBAN because no change occurred for this variable from 2006 to 2010. We provide detailed definitions and descriptive statistics for all of these variables in Table 1.

3.3 | Statistical model and analysis

Normally it is difficult to measure change in market structure because it represents the endogenous outcome of a competitive process (Berry & Waldfogel, 2001). However, Circuit City’s complete and quick

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
<th>Mean</th>
<th>SD</th>
<th>Min</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>ΔPV</td>
<td>Change in the logged number of a surviving store's total product variety (SKUs) after the rival chain exits</td>
<td>0.282</td>
<td>0.215</td>
<td>−0.633</td>
<td>0.842</td>
</tr>
<tr>
<td>NONCOL</td>
<td>Indicator = 1 if a surviving store's distance from its preexisting nearest rival store &gt;0.5 mile and ≤10 miles</td>
<td>0.560</td>
<td>0.497</td>
<td>0.000</td>
<td>1.000</td>
</tr>
<tr>
<td>COLOCATE</td>
<td>Indicator = 1 if a surviving store's distance from its preexisting nearest rival store ≤0.5 mile</td>
<td>0.262</td>
<td>0.440</td>
<td>0.000</td>
<td>1.000</td>
</tr>
<tr>
<td>ENTRY</td>
<td>Indicator = 1 if the surviving chain opens at least one new store in the market of a surviving store</td>
<td>0.394</td>
<td>0.489</td>
<td>0.000</td>
<td>1.000</td>
</tr>
<tr>
<td>INCOME</td>
<td>Median household income (US$, logged units), Census 2000</td>
<td>10.788</td>
<td>0.223</td>
<td>10.194</td>
<td>11.445</td>
</tr>
<tr>
<td>EDUCATION</td>
<td>Decimal fraction of the population with a college school or higher education, Census 2000</td>
<td>0.112</td>
<td>0.030</td>
<td>0.043</td>
<td>0.200</td>
</tr>
<tr>
<td>ADULT</td>
<td>Decimal fraction of the population older than 18 years, Census 2000</td>
<td>0.257</td>
<td>0.025</td>
<td>0.181</td>
<td>0.362</td>
</tr>
<tr>
<td>SENIOR</td>
<td>Decimal fraction of the population older than 65 years, Census 2000</td>
<td>0.117</td>
<td>0.031</td>
<td>0.046</td>
<td>0.280</td>
</tr>
<tr>
<td>MALE</td>
<td>Decimal fraction of male population, Census 2000</td>
<td>0.489</td>
<td>0.009</td>
<td>0.468</td>
<td>0.549</td>
</tr>
<tr>
<td>MARRIED</td>
<td>Decimal fraction of married population, Census 2000</td>
<td>0.514</td>
<td>0.074</td>
<td>0.318</td>
<td>0.708</td>
</tr>
<tr>
<td>NONWHITE</td>
<td>Decimal fraction of consumers who are nonwhite, Census 2000</td>
<td>0.272</td>
<td>0.156</td>
<td>0.024</td>
<td>0.741</td>
</tr>
<tr>
<td>SUPERURBAN</td>
<td>Indicator variable for super-urban market, Census 2000</td>
<td>0.226</td>
<td>0.418</td>
<td>0.000</td>
<td>1.000</td>
</tr>
<tr>
<td>WM</td>
<td>Number of Walmart stores in 2010 (logged units)</td>
<td>1.430</td>
<td>0.508</td>
<td>0.000</td>
<td>2.708</td>
</tr>
<tr>
<td>AMAZON</td>
<td>Google search index for Amazon in 2010</td>
<td>38.563</td>
<td>8.244</td>
<td>12.962</td>
<td>51.923</td>
</tr>
<tr>
<td>INTERNET</td>
<td>Internet penetration in 2010</td>
<td>80.550</td>
<td>3.537</td>
<td>70.870</td>
<td>90.100</td>
</tr>
<tr>
<td>ΔPOPULATION</td>
<td>Percentage change of population between 2006 and 2010</td>
<td>0.039</td>
<td>0.047</td>
<td>−0.079</td>
<td>0.231</td>
</tr>
<tr>
<td>ΔINCOME</td>
<td>Percentage change of median household income between 2006 and 2010</td>
<td>0.026</td>
<td>0.057</td>
<td>−0.150</td>
<td>0.242</td>
</tr>
<tr>
<td>ΔEDUCATION</td>
<td>Percentage change of population with a college or higher education between 2006 and 2010</td>
<td>−0.052</td>
<td>0.076</td>
<td>−0.278</td>
<td>0.338</td>
</tr>
<tr>
<td>ΔADULT</td>
<td>Percentage change of population for those 18 and older between 2006 and 2010</td>
<td>0.009</td>
<td>0.014</td>
<td>−0.050</td>
<td>0.062</td>
</tr>
<tr>
<td>ΔSENIOR</td>
<td>Percentage change of population for those 65 and older between 2006 and 2010</td>
<td>0.060</td>
<td>0.062</td>
<td>−0.324</td>
<td>0.438</td>
</tr>
<tr>
<td>ΔMALE</td>
<td>Percentage change of male population between 2006 and 2010</td>
<td>−0.004</td>
<td>0.010</td>
<td>−0.051</td>
<td>0.055</td>
</tr>
<tr>
<td>ΔNONWHITE</td>
<td>Percentage change of nonwhite population between 2006 and 2010</td>
<td>0.005</td>
<td>0.154</td>
<td>−0.719</td>
<td>0.978</td>
</tr>
<tr>
<td>ΔWM</td>
<td>Percentage of the number of Walmart stores between 2006 and 2010</td>
<td>0.131</td>
<td>0.321</td>
<td>0.000</td>
<td>4.000</td>
</tr>
<tr>
<td>ΔAMAZON</td>
<td>Percentage change of Google search index for Amazon between 2006 and 2010</td>
<td>0.679</td>
<td>0.285</td>
<td>−0.329</td>
<td>1.353</td>
</tr>
<tr>
<td>ΔINTERNET</td>
<td>Percentage change of Internet penetration between 2006 and 2010</td>
<td>0.129</td>
<td>0.038</td>
<td>0.048</td>
<td>0.252</td>
</tr>
</tbody>
</table>
market exit represented an exogenous change to the market structures of all of Best Buy’s geographic markets. To compare the product variety at each Best Buy store before and after Circuit City’s exit, we included in our analysis a number of demographic changes that occurred during this period of time. However, to ensure stricter control of other possible changes in local market conditions and trends (for example, which products are popular may change from 2006 to 2010), we used the product variety change of Best Buy stores in monopoly markets as a baseline for our empirical analysis. The external changes that influence product variety, other than Circuit City’s exit, apply to Best Buy stores in both monopoly and competitive markets. Appendix S3 provides a comparison of market-level characteristics between monopoly and competitive markets to better illustrate the basic conditions of the two types of markets.

We constructed an empirical analysis that in spirit is a difference-in-difference comparison. For each store, \( i \), that existed in both 2006 and 2010, we know its product variety both before and after Circuit City’s exit (\( PV_{i,2006} \) and \( PV_{i,2010} \)). We log-transformed the product variety counts and defined the dependent variable as

\[
\Delta PV_i = \log(PV_{i,2010}) - \log(PV_{i,2006})
\]  

Equation (7) produces a logged ratio directly, which removes any variety change related to magnitude. Regardless of how large or small a given store’s product variety is, we are more interested in the relative than the absolute change.\(^{15}\) This definition captures the variety change that occurred after the exit event and accounts for the temporal difference in a difference-in-difference analysis. The second “difference” here refers to product variety variations across geographic regions—namely, the difference between competitive markets and the baseline monopoly markets.

We modeled product variety change as a function of dummy independent variables and a set of market-level controls. Specifically, the functional form is as follows:

\[
\Delta PV_i = \beta_0 + \beta_1 \text{NONCOL}_i + \beta_2 \text{COLOCATE}_i + \beta_3 \text{ENTRY}_i + W_i \delta + \epsilon_i
\]  

where \( W_i \) is a vector of control variables reflecting the long-term, stable market conditions as well as changes in market conditions from 2006 to 2010 (as specified in Table 1) that might influence product variety change, \( \delta \) is a coefficient vector, and \( \epsilon_i \) is the error term following an i.i.d. normal distribution with a mean of 0.

Store entry decisions (ENTRY) are typically long-term-oriented, difficult to reverse, and very costly; in most cases, they are not made simultaneously with short-term decisions on product variety (\( \Delta PV \)). There is a slight possibility that the ENTRY variable may still be endogenous in that both product variety and entry decisions may be determined by some unobservable market-level characteristics, leaving the variable ENTRY possibly correlated with the error term \( \epsilon_i \). We adopted the instrumental variable (IV) approach to address such potential endogeneity. Specifically, we adopted a two-step procedure as suggested by Angrist and Pischke (2009). In the first step, we applied a probit model to analyze Best Buy’s new store opening. The probability that Best Buy would open a new store in market \( i \) (where surviving store \( i \) exists) is a function of market structures and a set of exogenous market-level demographic variables:

\[
\Pr(\text{ENTRY}_i = 1|X_i) = \Phi(X_i \alpha)
\]  

where \( \Phi(\cdot) \) is the cumulative distribution function of the standard normal distribution and \( X_i \) includes market structure variables (NONCOL and COLOCATE) as well as demographic variables that may affect market demand and thus are related to the store-opening decision in a given market. In the second step, we then used the predicted value of the nonlinear regression (Equation (9)) as an instrument for ENTRY and applied the conventional 2SLS procedure to estimate the model of product variety change (Equation (8)).

\(^{15}\)We also checked the result without the log transformation; our results did not change substantively.
Such estimation requires at least one “extra” explanatory variable that drives entry decisions but does not affect product variety change directly (Angrist & Pischke, 2009). Following Olivares and Cachon (2009), we used the market-level population size as our “extra” variable. Population is correlated with market entry since more firms would enter as a market's population increases, which suggested that we should include population in the probit model. Nevertheless, population may not be correlated with unobserved consumer characteristics that influence product variety following our use of a set of controls \((W_i)\) to capture observed consumer characteristics. This variable not only makes intuitive sense but also satisfies two statistical conditions, making it a reasonable “extra” variable. First, it has a positive and significant impact (coefficient = 0.445, \(p = 0.000\)) on the first-step dependent variable ENTRY. Second, we regressed \(\Delta PV\) on POPULATION, two market structure dummies, and the set of exogenous market-level controls \((W_i)\) and found the coefficient of POPULATION to be insignificant (coefficient = −0.004, \(p = 0.757\)), suggesting that this variable does not directly affect the change in product variety.

4 | RESULTS

4.1 | Main results

Table 2 presents the general pattern of product variety change in the three preexisting market structures. Overall, the average store-level product variety for surviving Best Buy stores increased from 33.25 in 2006 to 44.93 in 2010. Consistent with our theory, following the exit of Circuit City, Best Buy’s store-level product variety increased in both non-colocated and colocated markets (from 33.43 to 46.01 and from 33.57 to 47.41, respectively), as compared to the baseline monopoly markets (from 32.22 to 37.81). The magnitude of increase was larger in colocated markets (\(\Delta PV = 13.84\)) than in non-colocated markets (\(\Delta PV = 12.60\)). We also considered where Best Buy opened its new stores. As shown in Table 3, with only ten exceptions in monopoly markets, all store openings occurred in

<table>
<thead>
<tr>
<th>TABLE 2</th>
<th>Product variety levels at surviving Best Buy stores in 2006 and 2010</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(N)</td>
</tr>
<tr>
<td></td>
<td></td>
</tr>
<tr>
<td>Monopoly market (baseline) (NONCOL = 0; COLOCATE = 0)</td>
<td>126</td>
</tr>
<tr>
<td>Non-colocated competitive market (NONCOL = 1; COLOCATE = 0)</td>
<td>397</td>
</tr>
<tr>
<td>Colocated competitive market (NONCOL = 0; COLOCATE = 1)</td>
<td>186</td>
</tr>
<tr>
<td>Notes. SDs appear in parentheses. SKU: stock-keeping units.</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>TABLE 3</th>
<th>Where did Best Buy open new stores after Circuit City exited?</th>
</tr>
</thead>
<tbody>
<tr>
<td>AFTER: Did Best Buy open a new store in the market?</td>
<td>No</td>
</tr>
<tr>
<td>A: Across all markets ((N = 709))</td>
<td></td>
</tr>
<tr>
<td>BEFORE Monopoly market</td>
<td>116</td>
</tr>
<tr>
<td>Competitive market</td>
<td>314</td>
</tr>
<tr>
<td>B: Competitive markets ((N = 583))</td>
<td></td>
</tr>
<tr>
<td>BEFORE Non-colocated market</td>
<td>195</td>
</tr>
<tr>
<td>Colocated market</td>
<td>119</td>
</tr>
</tbody>
</table>
competitive markets. The number of new stores opened in non-colocated markets (202) was greater than that in colocated markets (67).

We present the main estimation results in Table 4. Model 1 represents the probit analysis of Best Buy’s new store opening, which is also used to generate the instrument for the potentially endogenous variable ENTRY. Model 2 and Model 3 contain the ordinary least squares (OLS) and IV results of our analysis of product variety change from before to after Circuit City’s market exit. We focus on

<table>
<thead>
<tr>
<th>Table 4</th>
<th>Results of Best Buy’s postexit new store opening and product variety change (N = 709)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Model 1</strong></td>
<td><strong>Model 2</strong></td>
</tr>
<tr>
<td>Dependent variable</td>
<td>ENTRY</td>
</tr>
<tr>
<td>Estimation method</td>
<td>Probit</td>
</tr>
<tr>
<td>Intercept</td>
<td>−33.234 (8.137)</td>
</tr>
<tr>
<td>NONCOL</td>
<td>0.478 (0.228)</td>
</tr>
<tr>
<td>COLOCATE</td>
<td>0.189 (0.242)</td>
</tr>
<tr>
<td>ENTRY</td>
<td>0.002 (0.018)</td>
</tr>
<tr>
<td>POPULATION</td>
<td>0.445 (0.115)</td>
</tr>
<tr>
<td>INCOME</td>
<td>1.033 (0.591)</td>
</tr>
<tr>
<td>EDUCATION</td>
<td>1.0275 (0.491)</td>
</tr>
<tr>
<td>ADULT</td>
<td>16.067 (0.638)</td>
</tr>
<tr>
<td>SENIOR</td>
<td>15.452 (0.484)</td>
</tr>
<tr>
<td>MALE</td>
<td>21.478 (0.491)</td>
</tr>
<tr>
<td>MARRIED</td>
<td>−4.221 (0.206)</td>
</tr>
<tr>
<td>NONWHITE</td>
<td>1.612 (0.082)</td>
</tr>
<tr>
<td>SUPERURBAN</td>
<td>0.276 (0.020)</td>
</tr>
<tr>
<td>WM</td>
<td>0.457 (0.082)</td>
</tr>
<tr>
<td>AMAZON</td>
<td>−0.004 (0.001)</td>
</tr>
<tr>
<td>INTERNET</td>
<td>−0.022 (0.003)</td>
</tr>
<tr>
<td>ΔPOPULATION</td>
<td>2.337 (0.223)</td>
</tr>
<tr>
<td>ΔINCOME</td>
<td>2.849 (0.160)</td>
</tr>
<tr>
<td>ΔEDUCATION</td>
<td>0.473 (0.105)</td>
</tr>
<tr>
<td>ΔADULT</td>
<td>12.280 (0.729)</td>
</tr>
<tr>
<td>ΔSENIOR</td>
<td>0.941 (0.157)</td>
</tr>
<tr>
<td>ΔMALE</td>
<td>−8.921 (0.970)</td>
</tr>
<tr>
<td>ΔNONWHITE</td>
<td>−0.764 (0.049)</td>
</tr>
<tr>
<td>ΔWM</td>
<td>−0.182 (0.026)</td>
</tr>
<tr>
<td>ΔAMAZON</td>
<td>0.524 (0.035)</td>
</tr>
<tr>
<td>ΔINTERNET</td>
<td>−0.948 (0.220)</td>
</tr>
<tr>
<td>Log pseudo-likelihood</td>
<td>−314.061</td>
</tr>
<tr>
<td>Wald χ²(24)</td>
<td>226.25</td>
</tr>
<tr>
<td>F statistics</td>
<td>F(24, 684) = 7.24</td>
</tr>
<tr>
<td>R²</td>
<td>0.176</td>
</tr>
<tr>
<td>Endogeneity test</td>
<td>NA</td>
</tr>
</tbody>
</table>

Notes. Robust SEs appear in parentheses; p-values in italics. All tests are two-tailed. IV: instrumental variable; OLS: ordinary least square.

* In the IV estimation of Model 3, we use the predicted value of ENTRY from the probit analysis in Model 1 as the instrument for ENTRY.

b We report R² for the OLS estimation. We do not report R² for the IV estimation since R² is not an appropriate measure of goodness of fit for IV regressions.
the IV analysis results (Model 3); the OLS estimation results (Model 2) are presented only for comparison, because we found that ENTRY is empirically endogenous (endogeneity test $\chi^2 = 4.977$, $p = 0.026$). Throughout all the specifications, we calculated robust SEs to deal with issues of heteroscedasticity.

H1 investigates how the surviving stores' postexit product variety change varies with preexisting market structures. In Model 3, both NONCOL and COLOCATE have positive and significant coefficients ($\beta_1 = 0.122$, $p = 0.000$; $\beta_2 = 0.158$, $p = 0.000$), suggesting that the surviving stores in both non-colocated and colocated markets tend to increase their product variety, which supports H1A. Moreover, the coefficient of COLOCATE is significantly greater than that of NONCOL ($\chi^2 = 3.72$, $p = 0.027$ for a one-tailed test of $\beta_2 > \beta_1$),16 suggesting that the surviving stores in colocated markets increase their variety to a greater extent than those in non-colocated markets. H1B thus is supported.

We now turn to the probit analysis of ENTRY (Model 1) to further explore a firm's store-opening decisions (H2). A number of control variables have significant coefficients. For example, Best Buy is more likely to open new stores in markets characterized by larger population size (POPULATION), higher median household income (INCOME), more Walmart stores (WM), and a higher level of urbanization (SUPERURBAN), as these variables may signal a higher demand level. Store opening is also more likely in markets with a more educated, adult, senior, male, nonmarried, and nonwhite population. Among the change variables from 2006 to 2010, if a market has a higher growth rate in income and adult population and if Amazon became a stronger competitor in 2010 in the market, we would observe a higher likelihood for Best Buy to open new stores. Regarding how Best Buy's entry pattern is affected by preexisting market structures, we find that compared to the baseline monopoly markets, the tendency for Best Buy to open new stores is not significantly higher in colocated markets ($\alpha_{COLOCATE} = 0.189$, $p = 0.434$), but Best Buy is more likely to do so in non-colocated markets ($\alpha_{NONCOL} = 0.478$, $p = 0.036$). Among the two types of competitive markets, Best Buy is more likely to open new stores in non-colocated than in colocated markets ($\chi^2 = 4.55$, $p = 0.016$ for a one-tailed test of $\alpha_{NONCOL} > \alpha_{COLOCATE}$). H2 thus is supported.

H3 examines how new store opening affects the firm's product variety adjustment. In Model 3, the coefficient of ENTRY is negative and significant ($\beta_3 = -0.145$, $p = 0.044$), suggesting that Best Buy's postexit product variety increase was negatively affected by new store entry, in support of H3. When Best Buy enters its own store's market by opening at least one new store, the surviving store does not increase product variety as much as it would in the absence of a new Best Buy store. This outcome implies that a market expansion strategy (i.e., opening new stores to attract consumers) and a product variety strategy (i.e., stocking more product variety to make the store more attractive to consumers) act as substitute strategies for a single firm. Among the control variables, we found that markets that have a higher median household income, are located in super-urban areas, have more Walmart stores, and have experienced a greater population rise from 2006 to 2010 tend to display greater increase in product variety. We also found that Best Buy's product variety change is affected by market-level consumer heterogeneity: Greater variety increase is found in markets with a higher fraction of adult, senior, male, unmarried, and nonwhite population.

16Our results also held when we conducted a two-tailed test to compare the coefficients of NONCOL and COLOCATE. Specifically, when we defined COLOCATE as a radius of 0.5 mile and NONCOL as a radius of 0.5–10 mile, $p = 0.054$ for a two-tailed test of $\beta_2 > \beta_1$. In another two-tailed test, $p = 0.002$ when COLOCATE was defined as a radius of 0.5 mile and NONCOL as a radius of 0.5–12 miles. Finally, in a third two-tailed test, $p = 0.020$ when we defined COLOCATE as a radius of 1 mile and NONCOL as a radius of 1–12 miles. We thank an anonymous reviewer for this valuable suggestion.
4.2 Robustness tests

We conducted several robustness checks and reported the results in Appendix S4. First, we varied the market definition by changing it to span a 12-mile radius. Correspondingly, we defined NONCOL as a dummy variable indicating the presence of the nearest rival in a 0.5–12-mile radius, while defining ENTRY as signifying whether or not a new store opened in a 12-mile radius (Model 4). Second, we kept the 10-mile market definition but expanded the radius of COLOCATE to 1 mile. Accordingly, we defined NONCOL as a 1–10-mile radius (Model 5). Third, we used the 1-mile cutoff point for COLOCATE and defined NONCOL as constituting a 1–12-mile radius. The entry cutoff point (ENTRY) was changed to 12 miles as well (Model 6). None of these alternative specifications substantively altered our findings. In Model 5, the magnitude of product variety increase is not significantly higher in colocated markets than in non-colocated markets ($\chi^2 = 1.39, p = 0.119$ for a one-tailed test of $\beta_2 > \beta_1$), yet we found strong support for H1B in all other specifications (0.5 mile vs. 10 miles, 1 mile vs. 12 miles, 0.5 vs. 12 miles). This finding suggests that H1B is supported in a slightly more polarized market definition. Fourth, our results remained robust when ENTRY was dropped from the regression (Model 7) and in the subsample wherein no store opening occurred (ENTRY = 0) (Model 8).

Fifth, we replaced the two dummy variables reflecting market structures (COLOCATE and NONCOL) with a continuous variable measuring a surviving store's distance (logged) from its nearest rival store (Model 9). The DISTANCE variable is negative and significant, suggesting the surviving store increases its product variety to a greater extent when the former nearest rival store is geographically closer, in support of both H1A and H1B. Our results remained supported with two alternative measures of DISTANCE: (a) DISTANCE was defined as the logged distance from the nearest Circuit City store for competitive markets and 0 for monopoly markets (Model 10), and (b) DISTANCE was defined as the exponential of negative distance from the nearest Circuit City store for competitive markets and as 0 for monopoly markets (Model 11). In both Model 10 and Model 11, a dummy variable COMPETITIVE (=1 for competitive markets, =0 for monopoly markets) was included to test H1A.

Sixth and finally, to distinguish Best Buy's expansion that occurred before versus after Circuit City's exit, we sorted all the new stores that opened after 2006 by their store IDs, split them in half, coded the first half as early entries (i.e., before Circuit City's exit in late 2008), and coded the second half as late entries. The dummy variable LATE ENTRY (=1 for late entries, and 0 otherwise) in Model 12 is insignificant, suggesting that there were no systematic differences between pre-2008 and post-2008 entries. Our results still held in the subsample of only post-2008 entries (i.e., the subsample of LATE ENTRY = 1), as presented in Model 13.

5 DISCUSSION AND CONCLUSIONS

5.1 Contributions

Our paper provides several important contributions to the literature. First and foremost, our study is the first to examine how incumbents respond to a main rival's exit. Our formal model and empirical

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17The rationale behind Best Buy's store openings before versus after Circuit City's exit might be different. Unfortunately, data on the opening dates of Best Buy stores is not available. After an extensive study of Best Buy store IDs and store-opening sequences, we found that Best Buy has been using monotonically increasing integers as store IDs, which suggests that stores with smaller IDs were opened earlier and stores with larger IDs opened later. We thus used store IDs as a proxy for store-opening sequence. Knowing that Best Buy opened 80 stores in 2007, 101 in 2008, 100 in 2009, 46 in 2010, and 30 in 2011, we split the new stores in half to identify early and late entries.
study show that following the rival’s exit, the survivor is motivated to expand in both product and geographic spaces by increasing its store-level product variety and opening new stores. Such expansions involve a joint consideration of filling in the “holes” in the market and preempting attractive locations to deter potential and unknown entrants. At a broad level, our findings contribute to the literature on incumbent responses to entry (Geroski, 1995; Seamans, 2013; Seamans & Zhu, 2013) by highlighting the difference between entry and exit. Incumbents play the role of defenders when responding to entry. In comparison, surviving incumbents following a rival’s exit act more like aggressors due to their limited information and engage in preemption without specific potential rivals in mind. Our findings of postexit expansion thus add to the literature on ex ante entry deterrence (Goolsbee & Syverson, 2008; Seamans, 2013). Moreover, this study contributes to a nascent body of research on incumbent repositioning (Du et al., 2018; Seamans & Zhu, 2017; Wang & Shaver, 2014). Repositioning, or response in general, can be opportunity-driven (de Figueiredo & Silverman, 2007; Greve, 1995) as well as competition-driven (Wang & Shaver, 2014). The former is argued to enhance future payoffs, as the new position is more attractive than the current one, and the latter is argued to mitigate immediate losses by making the current position less attractive (Wang & Shaver, 2014). We add uncertainty resulting from the rival’s exit as the third motive to the repositioning literature. Because information about the postexit world is at best incomplete, it is problematic to categorize uncertainty-driven repositioning as either pursuit of gains or avoidance of losses. For the same reason, we speculate that surviving firms may over-expand shortly after the exit and then contract after new firms enter and more information about the new entrants becomes available.

Second, our study considers multiple nonpricing responses and the relationship between them. Whereas the traditional focus has been on pricing responses (Geroski, 1995; Goolsbee & Syverson, 2008; McCann & Vroom, 2010; Simon, 2005), a growing stream of research has recently turned to nonpricing responses, such as store location (Thomadsen, 2007), product overlap (Hwang, Bronnberg, & Thomadsen, 2010), the decision to go online or stay offline (Seamans & Zhu, 2013), service quality (Bennett et al., 2013), and product variety (Casadesus-Masanell & Halaburda, 2014). Notably, Casadesus-Masanell and Zhu (2010) have examined how an incumbent chooses between multiple responses, including changing its business model and two tactical measures (pricing and advertising intensity), to cope with competition from new entrants. Similar to their study, we also focus on multiple responses that differ in terms of cost, reversibility, time consumption, and long-term orientation. We show that with the opening of a new sister store, the surviving store increases its product variety to a lesser extent. Such a substitutive relationship is consistent with the tradeoffs between product variety (horizontal scope) and vertical integration (vertical scope) found in the soft-drink industry (Zhou & Wan, 2017b). Our study thus provides additional evidence on the constraint of firms’ coordination capacity (Zhou, 2011; Zhou & Wan, 2017a).

Third, we explore how firm responses vary across different market structures. We focus on the distinction between colocated and non-colocated markets, partly because colocation has been particularly prevalent and important in retail yet has received little study in this context (Chung & Kalnins, 2001). Preexisting market structures reflect the distribution of customer preferences in the market and affect the equilibrium that rival firms are able to achieve, both of which influence the surviving firm’s postexit responses. We find the pattern of new store openings varies across market types in that the surviving firm is less willing to open new stores in colocated markets. This likely occurs because customers are more concentrated in colocated markets, resulting in fewer attractive locations for the surviving firm to choose from (Greve, 1995). In contrast, in non-colocated markets wherein customers are more evenly distributed, the surviving firm has more latitude to choose locations to balance the benefits of market coverage and risks of cannibalization (Anderson, Goeree, & Ramer, 1997).
Moreover, the different patterns we find for postexit product-variety expansion are consistent with the preexisting equilibrium identified by Ren et al. (2011). They find that colocated rivals carry a lower level of product variety than non-colocated ones, but they differentiate themselves more from each other by carrying fewer overlapping products. Our analytical model (represented in Lemma 1, as shown in Appendix S2) corroborates their empirical finding. We further show that because of this preexisting equilibrium on product variety, the surviving store in a colocated market tends to increase product variety to a greater extent than occurs in a non-colocated market. By adding temporal elements and studying the linkages between preexisting and postexit competitive dynamics, our finding thus extends Ren et al.'s (2011) work that emphasizes only cross-market-type variations.

Fourth, our study contributes to the literature on the relationship between product variety and market structure by focusing on radical changes in market structure. Most extant studies either focus on cross-sectional data (Olivares & Cachon, 2009; Watson, 2009) or treat market structure as a relatively stable element in panel data and do not explore the dynamic nature of the industries (Bayus & Putsis, 1999). We note, however, that market structure may experience abrupt or drastic changes. The changes can be driven by a new business model (Seamans & Zhu, 2013), consumer demand shifts (de Figueiredo & Silverman, 2007), competitive entry (Zhu et al., 2009), or policy change (Berry & Waldfogel, 2001). Berry and Waldfogel's (2001) seminal study of radio broadcasting stations after the Telecommunications Act of 1996 is the only extant research on how a drastic change in market structure affects product variety. Likewise, we also find that greater concentration in a geographic market results in more product variety. In the setting explored by Berry and Waldfogel (2001), market structure becomes consolidated when an exogenous policy change triggers a major wave of mergers. We complement their work by exploring increased market concentration triggered by the exit of a major rival. Moreover, with their focus on the total product variety available to consumers at the market level, their findings have implications for consumer welfare and public policies. With a firm-level analysis, our study extends their findings and sheds light on firms' strategies in response to market structure changes.

5.2 Limitations and future research directions

We note several caveats to our analysis and thus directions for future research. First, the findings of our study are based on firms in a consolidated industry over two periods of time; the results thus might reflect some factors specific to this industry or the study periods. Replications in other settings, such as booksellers (Barnes & Noble and Borders) could help ensure the generalizability of our findings. Second, we restricted our analysis to duopoly cases. Although the duopoly setting is common to many retail markets, allowing for multiple competitors and incorporating the effect of firm characteristics (e.g., size, age, and organizational structure) would extend our research. Consider, for example, the competitive dynamics of Netflix, Redbox, and Blockbuster. Do multiple surviving firms compete more aggressively and expand product variety more than if there were only one major surviving firm? How does the inclusion of additional players alter the interaction of product variety with market entry? If surviving firms are heterogeneous in characteristics and strategies, which kinds of firms are more likely to benefit from rivals' exit?

Third, as the first study to examine how surviving firms respond to rival exit, we define exit in a general way and our findings are generalizable to all markets undergoing drastic changes in market structure due to rival exit. We do not equate exit with failure, nor do we imply the industry is declining. We note that exit is a complex phenomenon that can be triggered by many factors such as high competitive intensity in the market (Baum & Korn, 1996), instability of overall market demand (Harrigan, 1980), the viability of the exiting rival's business model (Chuang & Baum, 2003), or
managers' limited cognitive abilities to achieve optimization objectives (Meyer et al., 2010). Exploring the fundamental reasons behind exit and distinguishing different types of exits will advance our understanding of the motives and patterns of incumbent responses to rival exit. Fourth and finally, we suggest that surviving firms expand out of the motives of meeting demand and preempting entry from unfamiliar entrants. Future research can further investigate how these mechanisms play out. We also suggest that surviving firms might overly expand out of uncertainty. Future research can examine the performance implications of incumbent responses to exit and use empirical evidence to validate this speculation. In summary, we hope our study ignites more scholarly interest in questions related to how surviving firms are affected by and act on the exit of rivals, regardless of whether they operate within thriving or declining industries.

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REFERENCES


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