Branding of beef retail products has gained momentum in recent years. In 2004, 42 percent of beef retail products were branded, a figure that grew to 63 percent in 2010 according to the 2010 National Meat Case Study conducted jointly by The Beef Check-off, the National Pork Board, and Cryovac® (National Cattlemen’s Beef Association 2010). As the potential value of differentiating and branding retail beef has become apparent, a proliferation of branding strategies has emerged. A review of retail data provided by Freshlook Marketing Group (2011) reveals that there are more than one hundred beef brands now present in U.S. retail markets. 

Product differentiation and branding are especially prevalent in beef steaks. The steak market is intriguing because numerous physical attributes and marketing characteristics are being used to differentiate the product. However, there is only limited information available on the implicit value of various steak product attributes and brand labels. In this study, we use revealed preference to determine implicit prices for retail steak products associated with descriptive package-label characteristics and product brands.

This study employs a two-step analysis. A hedonic model is used to reveal implicit prices for retail steak characteristics that include both physical (e.g., retail steak cut and bone presence) and credence (e.g., brand name, breed claim, organic production claim, religious processing claim) attributes. Understanding factors that affect implicit price is also of importance. For example, a brand can be thought of as a mix of hedonic, instrumental, and price preferences that represent value to the consumer and are reasonably consistent over time (Zeithaml 1988). To that end, we use estimated brand coefficients from the hedonic price model as a dependent variable to identify the factors associated with brand premiums. Knowing how a branding initiative affects value helps to identify branding strategies that successfully target consumers. Consumers benefit from brand strategies because the brand identifies a known set of attributes at a known level of quality and generally known price range relative to similar products (Owen, Wright, and Griffith 2000). As such, product brands reduce consumers’ search costs and uncertainty about product performance.

Previous Research

Previous studies have elicited various attribute values for retail beef products: eating quality (Hahn and Mathews 2007), fat content (Brester

Feldkamp, Schroeder, and Lusk (2005) evaluated consumers’ preferences for steaks labeled as generic, guaranteed tender, natural, USDA (U.S. Department of Agriculture) Choice, or Certified Angus Beef®. Consumers were willing to pay an economically important premium exceeding $2.00 per steak for Certified Angus Beef® relative to generic steak, suggesting name recognition of the Certified Angus Beef® brand.

Parcell and Schroeder (2007) analyzed panel-diary retail purchase prices for beef products to determine how pricing varied by product, geographic location, store type, sale items, composition (fresh, frozen, or cooked), and package size. Supermarket/grocery store branding was compared to Angus beef branding. Angus-branded medium-quality and high-quality steak cuts commanded premiums of $1.26 and $1.22 per pound, respectively, relative to supermarket/grocery-store-branded products. In contrast, low-quality Angus steak cuts appeared to be targeted at price-sensitive consumers as those cuts were sold at a lower price and the Angus brand premium dropped to $0.76 per pound.1

Ward, Lusk, and Dutton (2008) identified the value consumers place on observable characteristics of fresh beef sold at retail. Premiums for branded steaks and roasts ranged from $0.00 to $6.20 per pound relative to generic or unbranded products. In general, special and other brands were priced higher than generic and unbranded beef.

Martinez (2008) found a wide degree of variation in brand premiums for branded beef steaks—from −$0.44 to $4.15 per pound relative to unbranded steaks. Products receiving the largest premiums included branded beef allied with specific production requirements, including natural, organic, source verified, grass-fed, and breed.

Froehlich, Carlberg, and Ward (2009) analyzed consumers’ willingness to pay for fresh branded beef in an experimental auction framework. Survey participants preferred branded products to generic with hypothetical premiums that ranged from $1.12 to $1.32 per 12-ounce steak.

Past research has assessed valuation of branding in the national beef retail sector as well as other attributes that affect price. Our study differs from and builds upon information from previous research in several important ways. First, we rely on retail scanner data rather than hypothetical surveys or experiments to determine implicit market values of individual product characteristics. The scanner data provide a complete sample of sales of all steak products in the participating retail outlets over a five-year period. Second, rather than aggregating brands into arbitrary groupings, we estimate each individual brand premium after adjusting for other product characteristics. This allows us to determine the implicit price of the brand itself separate from the value of other product attributes bundled with the particular brand. Third, we estimate factors associated with implicit brand value. As such, we gain insight into the characteristics that contribute to brand equity and, thus, why consumers respond more favorably to certain brands.

Hedonic Model

An underlying assumption of the hedonic model is that goods can be distinguished by various product characteristics. As a result, marginal or implicit values can be estimated for each characteristic at the observed purchase price, which is linked to the presence of the particular characteristic.

Rosen (1974) hypothesized that the marginal implicit pricing schedule for a characteristic is a series of equilibriums between supply of and

---

1 The steak quality categories employed by Parcell and Schroeder (2007) represented aggregated primal cuts: high-quality (rib, ribeye, tenderloin, and filet mignon), medium-quality (T-bone, sirloin, New York strip, top loin, top sirloin, tip, porterhouse, and round), and low-quality (chuck, blade, arm, shoulder, flank, London broil, cube, and other).
demand for the characteristic over time or between markets. According to Rosen (1974), some have inappropriately interpreted the equilibrium points that represent marginal implicit values estimated from the standard hedonic pricing equation as the demand function for that characteristic. Rosen argued that the points are simply a sequence of supply and demand equilibriums that shift due to changes in exogenous supply and demand factors. He showed that standard hedonic price modeling overlooked changing marginal implicit values for different levels of characteristics because only consumer behavior had been considered and producer behavior had been ignored. “Estimated hedonic price-characteristics functions typically identify neither supply nor demand. In fact, those observations are described by a joint-envelope function and cannot by themselves identify the structure of consumer preferences and producer technologies that generate them” (Rosen 1974, p. 54). Accordingly, implicit values for specific product attributes are driven by a combination of consumer demand and product supply.

Following Rosen (1974), we suppose that a market good is composed of \( n \) characteristics:

\[
(1) \quad z = (z_1, z_2, \ldots, z_n).
\]

Prices can be related to the characteristics as

\[
(2) \quad p(z) = p(z_1, z_2, \ldots, z_n)
\]

where we assume that each product has a market price, \( p \), and that the summation of product attributes can be expressed by \( z \).

A vector of implicit marginal values is obtained by differentiating \( p(z) \) with respect to its \( i \)th argument, \( z_i \):

\[
(3) \quad p_{i}(z) = \frac{\partial p(z)}{\partial z_i}
\]

where \( p_{i} \) represents the characteristics’ marginal values.

We apply a two-step model. We first estimate implicit values for retail beef steak characteristics in the hedonic model and then estimate factors that affect a brand’s estimated value. Estimated brand coefficients from the standard hedonic price model are used as a dependent variable.

**Hedonic Model Using Retail Price and Product Characteristics**

We apply a hedonic pricing model to a panel of retail steak sales to estimate the impact of various physical attributes, product claims, and brand names on retail steak pricing. A fixed-effects estimator is hypothesized to control for the time-invariant, unobserved brand factors that may impact retail steak prices.\(^2\) Consider the model

\[
(4) \quad P_{ijt} = (\alpha + c_j) + x_{ijt} \beta + u_{ijt},
\]

\[
i = 1, \ldots, I, \quad j = 1, \ldots, J, \quad \text{and} \quad t = 1, \ldots, T
\]

where \( P_{ijt} \) denotes the price of the \( i \)th steak package with the \( j \)th brand for the \( t \)th time period, \( \alpha \) is the overall model intercept, \( c_j \) is the time-invariant, individual brand effect considered to be part of the intercept, \( x_{ijt} \) is a \( 1 \times K \) row vector of observable variables associated with product characteristics, \( \beta \) is a \( K \times 1 \) parameter vector of marginal effects of product attributes, and \( u_{ijt} \) represents the idiosyncratic errors that change across \( i, j, \) and \( t \) (Wooldridge 2002, Baltagi 2008). The \( c_j \) component consists of a dummy variable for each individual brand. The regression is a least squares dummy variable (LSDV) model—a fixed-effects model for which the slopes are constant but the intercepts differ according to the cross-sectional unit, which in this case is the brand.

The data set analyzed in this study includes repeated observations per cross-section and over time. As a result, errors are potentially serially correlated (i.e., correlation over \( t \) for a given \( i \) and \( j \)) and/or heteroskedastic. Inclusion of fixed individual-specific effects can reduce serial correlation in the errors (Cameron and Trivedi

---

\(^2\) The Durbin-Wu-Hausman test (Wu 1973) was used to determine if the time-invariant, unobservable factors should be treated as a fixed effect or as a random effect. We obtained group means for the time-invariant variables and added them to the estimated random-effects model. We then tested the joint hypothesis that the coefficients on the group means are all zero. The hypothesis that the individual effects are uncorrelated with the other regressors was rejected. This suggests that the effects are correlated with other variables in the model so the fixed-effects model is appropriate.
A Breusch-Pagan/Cook-Weisberg test rejects the null hypothesis that the error variances are constant. White’s heteroskedasticity-consistent covariance matrix is used to estimate standard errors.

Following Greene (2003), we use an F-test that resembles the structure of the F-test for R-square changes to test the hypothesis that the brand-specific constants are all equal, thereby testing the significance of the individual brand’s fixed effects. The F-ratio used for this test is

\[
F = \frac{(n - 1, \sum_{i=1}^{n} T_i - n - K)}{(1 - R^2_{LSDV}) \left( \sum_{i=1}^{n} T_i - n - K \right)}
\]

where \(LSDV\) and \(Pooled\) indicate LSDV and pooled models with only a single intercept for \(n\) brands and \(T\) time periods. Under the null hypothesis that the brand-specific constants are the same, this statistic is an F-random variable in which there are \(n-1\) degrees of freedom for the numerator and \(nT-n-K\) degrees of freedom for the denominator. The value of the F-random variable is \(F(61, 198,349) = 1,108.35\) (\(p\)-value = 0.000). This shows that the brand-specific constants differ and that a pooled model with one intercept is not appropriate. Overall, we conclude that the alternative pooled model omits important time-invariant brand effects so we use a fixed-effects model.

The retail steak price is modeled as

\[
Price_{ijt} = \alpha + \sum_{j=1}^{62} \beta_1 Brand_j + \beta_2 Breed_{it} + \beta_3 Organic_{it} + \beta_4 Religious_{it} + \beta_5 Bone_{it} + \sum_{c=1}^{31} \beta_6 Cut_{ict} + \sum_{j=1}^{273} \beta_7 Week_{it} + u_{ijt}
\]

where \(i\) refers to steak package (package here refers to weekly sales of the specific product), \(j\) refers to brand, and \(t\) refers to time period. Definitions for the rest of the variables are shown in Table 1.

### Brand Value Model

Brand value is the value beyond the physical and credence characteristics associated with the
product’s production or processing. There are numerous steak brands in the market aimed at appealing to different consumer preferences. It is therefore difficult to identify the characteristics that drive an individual brand’s premium using hedonic model parameter estimates on binary brand variables. For example, brands may differ across multiple dimensions, including the brand name’s longevity and the breadth of its distribution. The implicit values needed to determine brand value differences are obtained from estimating equation (6). Those implicit values ($\beta_{1j}$ estimates from equation (6) are used to determine the factors that contribute to a brand’s value. The brand value determination model takes the form

$$\text{(7)} \quad \text{BrandValue}_j = \alpha + \sum_{a=1}^{4} \beta_{a\cdot\text{BrandAge}_a} + \sum_{l=1}^{4} \beta_{l\cdot\text{Location}_l}$$

$$+ \sum_{p=1}^{4} \beta_{p\cdot\text{Positioning}_p} + \beta_{4\cdot\text{ChoicePlus}_j}$$

$$+ \beta_{5\cdot\text{MultiMeat}_j} + \epsilon_j$$

where $j$ refers to brand. Definitions for the rest of the variables are shown in Table 2.

### Table 2. Description of Variables and Summary Statistics of Brands

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
<th>Model 1 (no adjustment)</th>
<th>Model 2 (DFBETAS)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>Standard Deviation</td>
<td>Mean</td>
</tr>
<tr>
<td>Dependent Variable</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>BrandValue$_j$</td>
<td>Implicit value of brand (dollars per pound)</td>
<td>2.40</td>
<td>1.74</td>
</tr>
<tr>
<td>Independent Variables</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>BrandAge$_a$</td>
<td>Binary variables for age of brand $j$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$a =$ 1–3 years</td>
<td>0.07</td>
<td>0.25</td>
<td>0.02</td>
</tr>
<tr>
<td>4–6 years</td>
<td>0.18</td>
<td>0.39</td>
<td>0.19</td>
</tr>
<tr>
<td>7–10 years</td>
<td>0.05</td>
<td>0.22</td>
<td>0.07</td>
</tr>
<tr>
<td>greater than 10 years</td>
<td>0.70</td>
<td>0.46</td>
<td>0.72</td>
</tr>
<tr>
<td>Location$_l$</td>
<td>Binary variables for geographic scope of brand $j$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$l =$ Local</td>
<td>0.13</td>
<td>0.34</td>
<td>0.12</td>
</tr>
<tr>
<td>Regional</td>
<td>0.45</td>
<td>0.50</td>
<td>0.42</td>
</tr>
<tr>
<td>National</td>
<td>0.42</td>
<td>0.50</td>
<td>0.47</td>
</tr>
<tr>
<td>Positioning$_p$</td>
<td>Binary variables for positioning of brand $j$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$p =$ Special</td>
<td>0.33</td>
<td>0.48</td>
<td>0.33</td>
</tr>
<tr>
<td>Program</td>
<td>0.07</td>
<td>0.25</td>
<td>0.07</td>
</tr>
<tr>
<td>Store</td>
<td>0.23</td>
<td>0.43</td>
<td>0.21</td>
</tr>
<tr>
<td>Other</td>
<td>0.37</td>
<td>0.49</td>
<td>0.40</td>
</tr>
<tr>
<td>ChoicePlus$_j$</td>
<td>Proportion of brand $j$’s total sale of pounds labeled as Choice or Prime over entire data set</td>
<td>0.18</td>
<td>0.38</td>
</tr>
<tr>
<td>MultiMeat$_j$</td>
<td>Binary variable = 1 if multiple meat species per brand, 0 otherwise</td>
<td>0.63</td>
<td>0.49</td>
</tr>
</tbody>
</table>

Note: Observations that were deemed to be influential and removed from model 2 data were |DFBETAS| > 2√−n where n is the number of observations used (Belsley, Kuh, and Welsch 1980).

Using coefficient estimates as dependent variables in subsequent regression analyses is common (e.g., Bowman and Ethridge 1992, Chiu, Chen, and Capps 1993, Jusko and Shively 2005). Achen (2005) reviewed several alternative estimation techniques for two-step regression modeling. He concluded that the method we employ of estimating each equation separately by ordinary least squares is acceptable and even preferable under certain conditions to other potential methods. Achen (2005) noted that the second-stage model estimated using ordinary least squares for each equation can be delicate if the first-stage is not well specified. Specification errors in the first-stage will cause biases and inconsistencies in the second-stage even if the second-stage has no problem of its own. He further suggested that the first-stage sample size should be orders of magnitude larger than the second-stage sample and that there should be no large influential points in the second-stage parameters. Our models and associated estimates pass all of these conditions as discussed in the results.

Data

Scanner data representing steak purchases in U.S. retail outlets from January 1, 2004, through March 31, 2009, were obtained from Freshlook Marketing Group, which collects InfoScan data on meat department random-weight sales from retail food stores nationwide. The data set contains 198,719 weekly aggregated sales observations as value of pounds sold by brand name, retail cut, quality grade, breed claim, organic production claim, religious processing claim, and presence of bone.

Due to confidentiality restrictions, we cannot identify the specific brand names. We therefore label the brands numerically as \textit{Brand 1} through \textit{Brand 62}. Likewise, we simply note whether a specific breed claim was present or not. Brands carrying an organic claim must be certified by an accredited certifying agent, which verifies that the meat was produced and handled according to guidelines set forth in the Organic Foods Production Act (OFPA) of 1990 (Title 21, P.L. 101-624). Synthetic chemicals are not allowed and the producer and handlers must follow a plan agreed to by the certifying agent (USDA 2010). Religious processing claims consist of Kosher, Kosher-Glatt, Halal, and no religious claim. We combined all religious claims into a single binary variable equal to one if the product had a religious processing claim and zero otherwise.

Previous studies (e.g., Parcell and Schroeder 2007, Ward, Lusk, and Dutton 2008) included USDA quality grades of Prime, Choice, and Select and ungraded meat to categorize meat quality. There is considerable collinearity present in our data set between individual brands and quality grades. Consequently, we estimate a model that excludes quality grade variables since they are embedded in the brand effects.

Thirty-three steak cuts (Figure 1) are represented in the data. Cuts that are considered Premium are expected to have positive coefficients for the cut-specific binary variables included in the hedonic model while Everyday steak cuts are expected to have negative coefficient estimates (the Premium and Everyday categories come from The Beef Check-off Program (2008)). In addition, cuts that retain a bone (\textit{Bone} = 1) are expected to have a lower retail price per pound than boneless cuts. To adjust for changing aggregate steak prices over time, we include weekly binary variables.

In addition to details of aggregate weekly sales, the data set contains information about how the characteristics of each individual brand are likely to affect the brand’s value. The variables defined in Table 2 are used in step two of the analysis to regress the estimated brand premiums from the hedonic model against factors associated with each brand name. Brand longevity is the continued presence of a brand in the relevant market (Banbury and Mitchell 1995, Li 1995). The longevity of a brand is essential for a firm’s survival because it is linked to performance measures such as profitability and market share (Kanter and Brinkerhoff 1981, Suarez and Utterback 1995). We categorized brand longevity into five segments: (i) three years and less (7 percent), (ii) four to six years (18 percent), (iii) seven to ten years (5 percent), and (iv) eleven years and greater (70 percent). Brands having a longer presence in the market are expected to have greater consumer recognition and thus higher brand value.

The data set consists of 60 steak brands. We classify brands into the following geographic distribution categories. A \textit{local} brand is distributed only within a local geographic area and is privately

\footnote{Unbranded products and products that are included in a conglomerate “store” grouping are not included in the data set for stage two of the analysis. Thus, the number of brands is reduced from 62 to 60.}
owned and controlled by a small company. A regional brand is distributed regionally to retail outlets and is owned and controlled by a private company. Distribution is to one or more regions but is not nationwide. A national brand is distributed to retail locations nationwide and is controlled by the company or the supplier(s) that owns the brand. Of the 60 brands in this analysis, 8 were local, 27 were regional, and 25 were national.

A brand’s geographic prominence can have a positive or a negative relationship with its value. Previous research supports this assertion of mixed expectations regarding the geographic scope of a brand and the effect of that distribution on its value. Jekanowski et al. (2000) surveyed consumers in Indiana and concluded that they were willing to pay a premium for locally produced meat. This is consistent with similar results obtained from California (McGarry-Wolf and Thulin 2000), Colorado (Thilmany, Grammis, and Sparling 2003), and Chicago and Denver (Umberger et al. 2003). A national brand has a much larger overall volume and benefits from greater advertising expenditures, and, as such, it garners broader general consumer awareness that can secure greater brand value and a higher price (Parcell and Schroeder 2007). Previous studies (Darby et al. 2006, Hu 2007) have shown that taste is the single most important attribute in repeat purchases of a food, and consumers are more likely to have had experience with a nationally branded product than with a locally or regionally distributed one. In the early days of branded beef in the United States, some of the major beef brands were spearheaded by producer groups (e.g., the American Angus Association developed Certified Angus Beef®). The early emergence of nationally branded beef products in the United States encouraged retailers and packers to establish brands.

Because individual branded products are positioned to appeal to specific groups of consumers, we include brand positioning in our analysis to determine how types of brands differ in value. Brand types used in the study include special (33 percent), program (7 percent), store (23 percent), and other (37 percent). These

---

**Figure 1. Retail Steak Cut Proportion of Sales**

Notes: Following The Beef Check-off (2008), we classify retail steak cuts as Premium (porterhouse, T-bone, tender loin, top loin, lip-on ribeye, and ribeye) and Everyday (arm, blade, chuck eye, cross rib, neck, seven bone, tender, under blade, flank, ball tip, flap, sirloin, tri-tip, top sirloin, skirt, rib cap-on, bottom, full cut, round eye, tip, top, clod, flat iron, petite tender, top blade, cube, and other). Standard deviations are shown in parentheses.
categories are consistent with those of Ward, Lusk, and Dutton (2008). Special brands carry labels that relate information about practices used in production, such as “natural” or “organic,” that involve higher production costs (Yanik et al. 1999). Meats labeled as special are expected to have a high brand value. Program brands are breed-specific products (e.g., Angus beef). Generally, we expect that a breed name on the package helps to promote consumer confidence and loyalty due to the accountability and product assurance that come with the breed name. Store brands are specific to certain retail stores or chains of stores. Other brands are meat brands that could not be classified readily into one of the other three types; they were most often owned by a processor or meat market.

Quality grade variables are excluded from the retail price estimation because they are embedded with the brand effects. Because Prime and Choice quality grades signal high quality, we expect those labels to increase a brand’s value. We thus included the proportion of pounds sold by a particular brand that was labeled as Choice or Prime (ChoicePlus) in the brand-value-determination model.

Brand recognition can be strengthened by branding multiple food products. For example, a number of brands offer beef, poultry, and/or pork products. Multiproduct brands may enjoy greater brand equity because of broader consumer recognition of and loyalty to the brand name across food products. Sixty-three percent of the brands in our sample represent products from companies offering multiple meat brands.

**Implicit Prices of Retail Beef Steak Characteristics**

In empirical estimation, the theoretical foundation for hedonic models provides little guidance on appropriate functional form. Here, steak is assumed to be separable and additive in the various characteristics (e.g., breed claim, organic production claim, religious processing claim, cut), suggesting a linear relationship. This implies that individual steak characteristics can be bundled, packaged, and purchased in any combination. Empirical results for our hedonic pricing model are presented in Table 3.7 The coefficient estimates refer to changes in retail steak prices in dollars per pound from one-unit changes in the independent variable, *ceteris paribus*. A positive coefficient represents a premium for the particular steak characteristic while a negative coefficient indicates a discount.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient Estimate</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intercept</td>
<td>6.39*** (0.08)</td>
</tr>
<tr>
<td>Brandj (default: unbranded) (Figure 2)</td>
<td></td>
</tr>
<tr>
<td>Breed</td>
<td>–1.27*** (0.03)</td>
</tr>
<tr>
<td>Organic</td>
<td>2.98*** (0.05)</td>
</tr>
<tr>
<td>Religious</td>
<td>1.18*** (0.04)</td>
</tr>
<tr>
<td>Bone</td>
<td>–0.77*** (0.02)</td>
</tr>
<tr>
<td>Cut (default: sirloin) (Figure 3)</td>
<td></td>
</tr>
<tr>
<td>Observations</td>
<td>198,179</td>
</tr>
<tr>
<td>R-square</td>
<td>0.73</td>
</tr>
</tbody>
</table>

Notes: Coefficient estimates refer to a change in the retail steak price in dollars per pound from a one-unit change in the independent variable, *ceteris paribus*. Three (*** ) asterisks denote coefficients significantly different from zero at the 0.01 level. Standard errors are shown in parentheses under the coefficient estimates.

---

6 We also considered a log-linear model. Box-Cox regressions suggest that a log-linear functional form is more appropriate. However, the difference in “fit” is slight. In this case, the linear functional form is preferred because the price-per-pound interpretation is more straightforward for step-one and step-two modeling. General conclusions from each model specification are qualitatively the same.

7 Influence diagnostics (Belsley, Kuh, and Welsch 1980) determined that the coefficient estimates are not significantly influenced by a specific subset of outlier data.
determined that some brands with breed claims garner a premium while others are discounted. Thus, to predict the price of a steak that has a breed claim, the brand parameter estimate and breed claim estimate must be considered together.

Organic production of steaks garners a premium of $2.98 per pound relative to nonorganic steak products. These results are consistent with expectations; organic products tend to exhibit higher prices because they represent a particular niche market that is costly to supply relative to conventionally produced products. The organic price premium also may indicate growing consumer demand for organically produced products. The intuition for religious processing claims is similar, and those products exhibit a premium of $1.18 per pound.

As expected, bone-in retail cuts are discounted relative to boneless products, which have a $0.77 premium per pound. Shongwe et al. (2007) found that bone-in T-bone and rump steaks were discounted 23 percent relative to boneless products.

Figure 2. Estimated Steak Brand Premium Compared to Unbranded Product, January 1, 2004, through March 31, 2009

Notes: Coefficient estimates refer to a change in retail steak price in dollars per pound for a particular brand, *ceteris paribus*. One (*), two (**), and three (***) asterisks denote coefficients significantly different from zero at the 0.10, 0.05, and 0.01 level, respectively. Standard errors are shown in parentheses.
Unnevehr and Bard (1993) found that the per-pound price for bone-in cuts was sharply lower at $1.11 for rib steaks, $0.91 for loin steaks, and $0.30 for sirloin steaks.

The steak-cut coefficients in our model coincide with classifications by The Beef Check-off (2008) for Premium and Everyday steaks. Steaks in the Premium category are tenderloin, porterhouse, T-bone, ribeye, top loin, and lip-on ribeye, and they garner premiums (Figure 3) relative to sirloin steaks with an average premium of $3.43 per pound. The tenderloin cut receives the highest premium, $7.60 per pound. The premium for steaks in the Everyday category is negative, ranging from $0.05 to $4.67 per pound. Everyday steaks are likely discounted because consumers perceive these cuts as being less flavorful and less tender. Additional processing and preparation often are necessary when cooking such steaks.

**Retail Value of Beef Steak Brands**

The preceding discussion highlights the value that consumers place on descriptive characteristics of steak and identifies individual brand values. But what factors influence brand value? The second step of the analysis provides insight into this question. According to Achen (2005), two-step estimation is viable when (i) first-stage models are well specified, (ii) first-stage samples (of size $n_j$) are much larger than second-stage samples (of size $m$), and (iii) influential points in the second stage are
thoroughly assessed. We assert that our first-stage specification is appropriate. Next, there should be more retail steak prices by an order of magnitude than there are brands. Achen (2005) found that simple examples indicate that consistency requires \( m/n \to 0 \) asymptotically. We clearly adhere to the sample-size principle as \( 60 / 198,179 = 0.0003 \). Finally, to determine if the results of the second-step regression analysis are significantly influenced by outlier observations, we compute influential diagnostics, namely, DFBETAS. According to Belsley, Kuh, and Welsch (1980, p. 263), “DFBETAS point to characteristics of the data to which the coefficient estimates or their estimated standard errors are particularly sensitive and are especially useful for examining the suitability of the data for structural estimation.”

Our discussion of results follows our base model, which did not eliminate any potential outlier observations (model 1). However, we highlight specific results that may be impacted by outlier data by comparing the base model to model 2. In model 2, we delete any influential outlier observations as determined by the DFBETAS.

The results of estimating equation (7) are presented in Table 4. New brands—brands that have existed in the industry for three years or less—have a $1.69 per pound premium relative to brands that have been in the industry for greater than ten years. When consumers can choose among many brands, producers tend to emphasize development of new and different product attributes rather than the value found in traditional products (Outlaw, Anderson, and Padberg 1997). Our estimates indicate that new brands are typically introduced at premium prices. Perhaps newly launched brands are targeting specific, emerging consumer trends.

How many of these brands will be successful in sustaining a premium value over time is unknown but some likely will fail. The estimates generated for medium-aged brands are not statistically significant.

Regional brands garnered a $0.76 premium per pound relative to national brands. The alternative model estimated after deleting influential outliers indicates that the regional premium could be as large as $1.02 per pound relative to national brands. This is a surprising result since regional brands have smaller market shares and presumably less general consumer recognition. In contrast,

### Table 4. Determinants of Brand Value per Pound

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient Estimates</th>
<th>Model 1</th>
<th>Model 2</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>(no adjustment)</td>
<td>(DFBETAS)</td>
</tr>
<tr>
<td>Intercept</td>
<td>0.41</td>
<td>0.09</td>
<td>0.63</td>
</tr>
<tr>
<td>BrandAge₀ (default: greater than 10 years)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>1–3 years</td>
<td>1.69**</td>
<td>1.94*</td>
<td>(0.82)</td>
</tr>
<tr>
<td>4–6 years</td>
<td>–0.71</td>
<td>–0.60</td>
<td>(0.59)</td>
</tr>
<tr>
<td>7–10 years</td>
<td>–0.28</td>
<td>–0.76</td>
<td>(1.02)</td>
</tr>
<tr>
<td>Location, (default: national)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>local</td>
<td>0.09</td>
<td>0.85</td>
<td>(0.62)</td>
</tr>
<tr>
<td>regional</td>
<td>0.76*</td>
<td>1.02***</td>
<td>(0.43)</td>
</tr>
<tr>
<td>Positioning, (default: other)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>special</td>
<td>2.32***</td>
<td>2.54***</td>
<td>(0.48)</td>
</tr>
<tr>
<td>program</td>
<td>1.54*</td>
<td>2.91***</td>
<td>(0.90)</td>
</tr>
<tr>
<td>store</td>
<td>1.31***</td>
<td>2.34***</td>
<td>(0.64)</td>
</tr>
<tr>
<td>ChoicePlus</td>
<td>0.73</td>
<td>0.32</td>
<td>(0.61)</td>
</tr>
<tr>
<td>MultiMeat</td>
<td>0.56</td>
<td>0.51</td>
<td>(0.48)</td>
</tr>
<tr>
<td>Observations</td>
<td>60</td>
<td>43</td>
<td></td>
</tr>
<tr>
<td>R-square</td>
<td>0.43</td>
<td>0.70</td>
<td></td>
</tr>
</tbody>
</table>

Notes: Coefficient estimates refer to a change in brand value in dollars per pound from a one-unit change in the independent variable, ceteris paribus. One (*), two (**), and three (***) asterisks denote coefficients significantly different from zero at the 0.10, 0.05, and 0.01 level, respectively. Standard errors are in parentheses under the coefficient estimates. Observations that were deemed to be influential and removed from model 2 estimation are \(|DFBETAS| > 2\sqrt{n}/n\) where \( n \) is the number of observations used (Belsley, Kuh, and Welsch 1980).

Additional diagnostic tests were conducted for the brand-value model. The presence of heteroskedasticity was tested using the Breusch-Pagan/Cook-Weisberg test, which failed to reject the null hypothesis of constant variances. Statistics for variance inflation factors indicated no problem with multicollinearity. Box-Cox regressions that compared the goodness-of-fit of models in which the dependent variable was in levels or logs suggest that a linear functional form is more appropriate.
local brands do not differ statistically in terms of brand equity relative to national brands.

Our results support recent changes firms have made to differentiate products through brand positioning. Estimates of positioning in our analysis show that special brands sell for $2.32 per pound more than other brands; while program and store brands have premiums of $1.54 and $1.31 per pound, respectively. The alternative model adjusted for influential observations shows that these premiums could be as large as $2.54 for special, $2.91 for program, and $2.34 for store brands relative to other brands.

Both models fail to show a statistically significant effect for (i) the proportion of pounds labeled Choice or Prime (ChoicePlus) and (ii) whether the company that owns the brand also has meat from other species carrying the same brand name (MultiMeat).

Conclusions and Implications

The objective of this study was to determine how product attributes of retail beef steak affect prices. We were particularly interested in determining implicit price premiums for steak brands. Certain brands garner premiums while others receive discounts relative to unbranded products. We found that 55 of the 62 retail steak brands evaluated received positive premiums; the remaining brands were discounted relative to unbranded products. Characteristics other than brand that garner a premium include organic production claims, religious processing claims, and boneless products. Premium steaks, such as tenderloin, porterhouse, T-bone, ribeye, top loin, and lip-on ribeye, exhibit premiums when compared to sirloin steaks. Steak cuts perceived to be of lower quality were discounted. Consumers exhibit complex purchasing behavior that various beef industry sectors are taking into consideration in their efforts to provide desired products. Our results identify implicit prices of numerous product attributes, providing every sector of the beef industry with information regarding how to improve branding strategies.

Newly introduced brands tend to have higher prices than existing products. They are often developed in response to consumer trends regarding quality and safety and involve additional production costs. Products that successfully align with emerging consumer preferences may be able to capture premium prices. We lacked data necessary to determine failure rates for newly developed products, a factor that must be kept in mind when determining pricing strategies.

Branding often targets consumers who prefer specific product characteristics, such as certified breed and natural and organic production claims. Also, branding tied to a particular retailer or store can help to promote and ensure confidence and loyalty. Tonsor, Schroeder, and Pennings (2009) concluded that store reputation influences consumer perceptions about products’ quality and safety and that store branding is an additional way to leverage value from a store’s reputation. We expect store brands to play a substantial and growing role in branded beef products in the future.

Brand value is not strongly associated with the quality grade of the meat. Most branded beef products do not use quality grade as a substantial part of the product labeling scheme even though the grade is often part of the brand’s specifications. Consumers may not understand or may be poorly informed about formal beef grades (Cox, McMullen, and Garrod 1990).

For a branded steak product to be successful, there must be a strong link between consumers’ attitudes and the attributes offered by the brand. Marketers must realize that consumer perceptions change over time as a result of new information, increased competition in a product category, and changing expectations. The dynamic nature of the retail beef market suggests that marketers must track perceptions over time and continually align their product and branding strategies with changing views.

An important limitation of this analysis is that we could not determine whether implicit premiums for product attributes are driven more by consumer demand or by supply. That is, a product attribute such as an organic production claim has a statistically significant premium of $2.98 per pound. However, organic products represent only 4 percent of the steak market. Consequently, determining whether to target a steak product to this market requires understanding not only the added cost of doing so but also the elasticity of demand for the product. Similarly, a particular steak brand may enjoy a premium price in the market that is associated with both demand and supply for that brand. In designing strategies for expanding or contracting a brand’s presence, elasticity of demand for the specific branded
product must be known. Determining demand elasticities for specific steak product attributes is a logical step for future research.

References


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