A Framework for Identifying the Sources of Local-Currency Price Stability with an Empirical Application*

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Abstract

The inertia of the local-currency prices of traded goods in the face of exchange-rate changes is a well-documented phenomenon in International Economics. This paper develops a framework for identifying the sources of local-currency price stability. The empirical approach exploits manufacturers’ and retailers’ first-order conditions in conjunction with detailed information on the frequency of price adjustments in response to exchange-rate changes, in order to quantify the relative importance of markup adjustment by manufacturers and retailers, local-cost non-traded components, and nominal price rigidities, in the incomplete transmission of exchange-rate changes to prices. The approach is applied to micro data from the beer market. We find that on average, approximately 53% of the incomplete exchange rate pass-through is due to local non-traded costs; 7% to markup adjustment; 18% to the existence of own-brand price adjustment costs, and 22% to the indirect/strategic effect of such costs.

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1 Introduction

The incomplete transmission of exchange-rate shocks to the prices of imported goods has been the focus of a substantial amount of theoretical and empirical research. In his 2002 article in the *NBER Macroeconomics Annual*, Engel extensively discusses this research and identifies three potential sources for the incomplete exchange-rate pass-through: the existence of local costs (e.g., costs for non-traded services) even among goods that are typically considered to be “traded”; markup adjustment on the part of retailers and/or manufacturers; and pure nominal price rigidities (at times also referred to as “menu costs”) that lead to what Engel has labelled “local-currency-pricing”. Despite the significant amount of work and interest in this topic, evidence on the relative importance of each of the contributing factors remains mixed, in part because some of the key variables needed to identify these factors, such as markups or local costs, are not directly observable, especially not in aggregate data. Yet, in an era characterized by a continuing devaluation of the dollar against other major currencies, concerns about the impact of China’s exchange-rate policy on domestic prices, and general uncertainty about the effect of exchange rates on the unwinding of current imbalances, it is more important than ever to understand why import prices do not fully respond to exchange-rate changes, especially since different explanations have very different implications for exchange rate policy.¹

Aided by the increased availability of micro data sets, a set of recent studies has focused on the microeconomics of the cross-border transmission process, trying to identify the relative contribution of each of the sources of this price inertia within structural models of particular industries. The advantage of these studies is that the institutional knowledge of the industry can be used to inform modeling assumptions, which, applied to detailed consumer or product-level data, can deliver credible estimates of markups and local costs. The disadvantage is that the results are not generalizable without further work on other markets. Still, as we show below, the few studies available to date have been able to identify interesting empirical patterns that are surprisingly robust across markets, time, and specific modeling assumptions.

The general structure of the approach proposed in this strand of the literature is as follows. The starting point is an empirical model of the industry under consideration. The model has three elements: demand, costs, and equilibrium conditions. The demand side is estimated first, independently of the supply side, using either consumer level data on individual transactions, or product level data on market shares and prices. On the supply side, the cost function of a producer selling in a foreign country is specified in a way that allows for both a traded, and a non-traded, local (i.e., destination-market specific) component in this producer’s costs. The distinction between traded and non-traded costs is based on the currency in which these costs are paid. Traded costs are by definition incurred

¹A recent speech that summarizes policymakers’ concerns about these issues can be found at: http://www.federalreserve.gov/newsevents/speech/mishkin20080307a.htm.
by the seller in her home country. As such, they are subject to shocks caused by variation in the
nominal exchange rate when they are expressed in the destination market currency. In contrast, non-
traded costs are defined as those costs that are not be affected by exchange rate changes. Costs are
treated as unobservable. Assuming that firms act as profit maximizers, the market structure of the
industry in conjunction with particular assumptions regarding firms’ strategic behavior imply a set
of first-order conditions. Once the demand side parameters are estimated, these first-order conditions
can be exploited to back out the marginal costs and markups in the industry. Based on the specified
cost function, marginal costs are further decomposed into a traded and non-traded component.

With this decomposition in place, one then examines how the particular components of prices
(traded cost component, non-traded cost component, and markup) respond to exchange-rate changes.
The lack of price response is accordingly attributed to either markup adjustment, or to the existence
of a local, non-traded cost component. While the results of this decomposition naturally vary by
industry, it seems that existing studies are in agreement that markup adjustment is a big part of
the story. The observed exchange-rate pass-through is however too low to be explained by markup
adjustment alone; accordingly, the role attributed to non-traded costs in explaining the incomplete
price response is non-trivial.

While the above framework allows one to evaluate the relative contributions of markup adjustment
and non-traded costs in explaining incomplete exchange-rate pass-through, it is inherently unsuitable
to assessing the role of the third potential source of the incomplete price response: the existence of
fixed costs of repricing. There are two reasons for this inadequacy. The first reason is a conceptual one.
A key element of the framework described above is the premise that the firms’ first-order conditions
hold every period. Given that by assumption firms are always at the equilibrium implied by their
profit maximizing conditions, there is no role in this framework for price adjustment costs that would
cause firms to (temporarily) deviate from their optimal behavior. The second reason is a practical
one. Because the data used in most previous studies are either annual or monthly, and because they
are often the outcome of aggregation across more disaggregate product categories, we observe product
level prices changing in every period. But with prices adjusting every period, it is inherently impossible
to identify potential costs of repricing, which by nature imply that prices should remain fixed. Hence,
to the extent that such price rigidities are present, these may be masked by the aggregation across
different product lines, and across shorter time periods (e.g., weeks), over which nominal prices may
exhibit inertia. This may lead one to overstate the role of non-traded services: whatever portion
of incomplete pass-through cannot be accounted for by markup adjustment, will by construction be
attributed to non-traded costs, when in reality (and in a more general framework) it could be due to
the existence of price adjustment costs.

The current paper attempts to overcome this shortcoming by explicitly introducing price rigidities
into the model and suggesting an approach for quantifying their importance in explaining the docu-
mented incomplete cross-border transmission of exchange-rate shocks. To this end, we introduce two new elements.

The first one is to modify the standard framework of profit maximization to allow firms to deviate from their first-order conditions due to the existence of fixed costs of repricing. In this context we define costs of repricing in the broadest possible sense as all factors that may cause firms to keep their prices constant, and hence potentially deviate from the optimum implied by static profit maximization. Such factors may include the small costs of re-pricing (the so-called “menu-costs”) as well as the more substantive costs associated with the management’s time and effort in figuring out the new optimal price, the additional costs of advertising and more generally communicating the price change to the consumers, and – to the extent that one wants to incorporate dynamic considerations in the analysis – the option value of keeping the price unchanged in the face of ongoing uncertainty.

The second innovation of the paper is on the data side. In order to identify the potential role of nominal price rigidities we propose using higher frequency (weekly or bi-weekly) data on the prices of highly disaggregate, well-defined product lines. The advantage of using high-frequency data is that we observe many periods during which the price of a product remains utterly unchanged, followed by a discrete jump of the price to a new level. It is this discreteness in the price adjustment that we exploit in order to identify the role of nominal price rigidities.

The basic idea behind our approach is as follows. First, even with nominal price rigidities, we can estimate the demand and cost parameters of the model along the lines described in earlier papers by constraining the estimation to the periods for which we observe price adjustment; the underlying premise is that once a firm decides to incur the adjustment cost associated with a price change, it will set the product’s price according to the first-order conditions of its profit maximization problem. This of course does not imply that this firm’s behavior will be unaffected by the existence of price rigidities. Such rigidities may still have an indirect effect on the pricing behavior of firms that adjust their prices, as in any model of oligopolistic interaction firms take the prices (or quantities) of their competitors into account; if the competitor prices do not change in particular period (possibly because of price rigidities), this will affect the pricing behavior of the firms that do adjust prices. The estimation procedure takes this indirect effect into account.

Once the model parameters are estimated, we exploit information from both the periods in which prices adjust and periods in which prices remain unchanged to derive bounds on the adjustment costs associated with a price change. Our approach is based on the insight that in periods in which prices change, it has to be the case that the costs of price adjustment are lower than the additional profit the firm makes by changing its price; we can use this insight to derive an upper bound of this price adjustment cost. Similarly, in periods in which prices do not change, it has to be the case that the costs of adjustment exceed the extra profit associated with a price change; based on this insight, we can derive a lower bound for the price adjustment cost.
The costs of price adjustment are a concept that has a precise meaning within the context of our model; they are defined in the broadest possible sense as everything that prevents a firm from adjusting its price in a particular period, including concerns about losing customers in the future, rather than the literal labor or material costs a firm has to pay to change prices. As such, they are not directly comparable to estimates obtained in earlier studies using different methods (e.g., direct measurement or firm surveys). More importantly, the adjustments costs alone do not allow a full assessment of the impact of nominal price rigidities on exchange-rate pass-through; because such rigidities have both a direct and an indirect (operating through the competitor prices) effect on firms’ pricing behavior, it is possible that very small rigidities induce significant price inertia. To provide an overall assessment of the impact of price adjustment costs we therefore perform simulations that compare the pricing behavior with price rigidities to the one that would prevail with fully flexible prices. The differential response of prices in the two scenarios is attributed to the effect of nominal price rigidities. In the same step we also identify the role of markup adjustment and non-traded costs in generating incomplete pass-through.

We apply the framework described above to weekly, store-level data for the beer market. The beer market is well suited for investigating questions related to exchange-rate pass-through and price rigidities for several reasons: (1) a significant fraction of brands are imported and hence affected by exchange-rate fluctuations; (2) exchange-rate pass-through onto consumer prices is low, on the order of 7-10%; (3) there exist highly disaggregate, weekly data on both wholesale and retail prices; this allows us to examine how prices respond at each stage of the distribution chain; (4) both non-traded local costs and price rigidities are a-priori plausible; in particular, weekly data reveal that both wholesale and retail prices remain constant over the course of several weeks, suggesting the existence of price rigidities. The framework we propose is however not tailored to the beer market, and can be more generally applied to any market for which high frequency data are available so that the points of price adjustment can be identified.

Our analysis yields several interesting findings. First, at the descriptive level, we document infrequent price adjustment both at the wholesale and retail level. However, this price inertia seems to be primarily driven by the infrequent adjustment of wholesale rather than retail prices. In our data, there is not a single instance where a product’s retail price remains unchanged in response to a wholesale price change. In contrast, there are several instances in which the retail price changes, while

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2 There have been several attempts in the literature to measure price adjustment costs. Levy et al (1997) find menu costs to equal 0.70 percent of supermarkets’ revenue from time-use data. Dutta et al (1999) find menu costs to equal 0.59 percent of drugstores’ revenue. Levy et al have four measures of menu costs: 1. the labor cost to change prices; 2. the costs to print and deliver new price tags; 3. the costs of mistakes; 4. the costs of in-store supervision of the price changes. Some detailed microeconomic studies have cast doubt on the importance of menu cost in price rigidity. Blinder et al (1998) find in a direct survey that managers do not regard menu costs as an important cause of price rigidity. Both Carlton (1986) and Midrigan (2006) find that firms change prices frequently and in small increments, which is not consistent with a menu-cost explanation of price rigidity.
the wholesale price does not. Hence it seems that the primary reason that retail prices do not change from period to period is that there is little reason for them to change, as the underlying wholesale prices remain fixed.

As we discussed above, nominal price rigidities may affect the pricing decisions of a particular producer in two ways. First, they may prevent this producer from adjusting her price, because her own costs of repricing exceed the benefits, even when all other competing producers adjust their prices (direct effect). Second, such costs may induce other competing producers to keep their prices fixed, which may make price adjustment less profitable for the producer under consideration (indirect stratégic effect). Our simulations indicate that the direct effect is significant at the wholesale level, accounting for 17.9% on average for the incomplete pass-through. Interestingly, there is substantial variation in this estimate across brands; the own costs of price adjustment appear to be more important for brands with large market shares such as Corona and Heineken. In contrast, we find that at the retail level the own costs of repricing have no effect. There is also an indirect stratégic effect at the wholesale stage of the distribution chain that accounts for approximately 21.9% of the incomplete pass-through. Our final decomposition attributes 53% of the incomplete pass-through to local non-traded costs, 7.5% to markup adjustment, and 17.9% to the existence of own-brand price adjustment costs, and 21.9% which represent the indirect stratégic effect of such costs. As suspected at the beginning, costs of price adjustment appear to be substantially more important at the wholesale than retail level.

The approach we just described is static. While estimating a full dynamic model would of course be desirable, the complexity of our setup and model does not allow us to do so (at least not without simplifying parts of the model that we consider essential to our analysis, such as the strategic interactions between domestic and foreign firms as well as the retailer and manufacturers, the flexibility of the demand substitution patterns, and the consistent treatment of the cost side). However, we believe that there are two particular features of this market that allow us to abstract from dynamics without sacrificing the realism and relevance of our model. First, neither consumers nor sellers hold inventories for beer. 3 Second, the primary source of cost variation facing beer sellers and focus of our analysis is the exchange rate, which is highly persistent. 4 This implies that the premise of static expectations is not unreasonable here. If the exchange rate were the only source of cost variation, and exchange rate shocks were perceived as permanent, the static model would be equivalent to a dynamic one up to a discount factor: firms would be maximizing the sum over the (discounted) profits of each period, which in the absence of any cross-period price effects or inventories would be equivalent to maximizing period-by-period profits (Dixit 1991). If the exchange rate is the primary (but not only) source of

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3 The role of inventories has been emphasized in several recent papers by Hendel and Nevo (2006a and 2006b), Pesendorfer (2002), Aguirregabiria (1999), and Alessandria, Kaboski and Midrigan (2008). We refer the reader to section 3.3 for a detailed discussion of how the arguments put forth in these papers affect our analysis.

4 The consensus among International economists is that the exchange rate follows a random walk.
cost variation and it is highly persistent, our model provides a reasonable approximation to a dynamic setup. We discuss these issues formally in Section 3.3.

The remainder of the paper is organized as follows. To set the stage, we start by providing a brief description of the market and the data in the next section; in the same section, we also provide some descriptive statistics and discuss the price adjustment patterns evident in the retail and wholesale price data. Section 3 discusses the model and shows how it allows us to derive bounds for the price adjustment costs. Section 4 discusses the steps of the empirical implementation of the model in detail. Section 5 presents the estimation and simulation results, and Section 6 concludes.

2 The Market and the Data

In this section we describe the market our data cover. We present summary statistics and some preliminary descriptive results to build intuition for the results from the structural model. We then discuss some of the price-adjustment patterns in the data.

2.1 Market

The imported beer market first developed in the U.S. in the nineteenth century. As late as 1970, imported beers made up less than one percent of the total U.S. consumption of beer. Consumption of imported brands grew slowly in the 1980s and by double digits for each year in the 1990s – on average by 11 percent per year from 1993 to 2001. Beer is an example of one type of imported goods: packaged goods imported for consumption. Such imports do not require any further manufacture before reaching consumers and make up roughly half of the non-oil goods imports to the U.S. over the sample period.

The beer market is well suited for an exploration of the sources of local-currency price stability for the reasons discussed in the introduction: a significant fraction of brands are imported; exchange-rate pass-through to prices is generally low (between eight and ten percent); both non-traded local costs and price stickiness due to adjustments costs are \textit{a-priori} plausible; last but not least, we have a rich panel data set with weekly retail and wholesale prices. It is unusual to observe both retail and wholesale prices for a single product over time. These data enable us to separate us to isolate the role of local non-traded costs and of fixed adjustment costs in firms’ incomplete transmission of exchange-rate shocks to prices.

2.2 Data

Our data come from \textit{Dominick’s Finer Foods}, the second-largest supermarket chain in the Chicago metropolitan area in the mid 1990s with a market share of roughly 20 percent. The data record the retail and wholesale prices for each product sold by \textit{Dominick’s} over a period of four years. They were
gathered by the *Kilts Center for Marketing* at the University of Chicago’s Graduate School of Business and include aggregate retail volume market shares and retail prices for every major brand of beer sold in the U.S. Beer shipments in this market are handled by independent wholesale distributors. The model we develop in the next section of the paper abstracts from this additional step in the vertical chain, and assumes distributors are vertically integrated with brewers, in the sense that brewers bear their distributors’ costs and control their pricing decisions. It is common knowledge in the industry that brewers set their distributors’ prices through a practice known as *resale price maintenance* and cover a significant portion of their distributors’ marginal costs. This practice makes the analysis of pricing behavior along the distribution chain relatively straight-forward, as one can assume that distributors are, *de facto*, vertically integrated with brewers.

During the 1990s supermarkets increased the selection of beers they offered as well as the total shelf space devoted to beer. A study from this period found that beer was the tenth most frequently purchased item and the seventh most profitable item for the average U.S. supermarket. Supermarkets sell approximately 20 percent of all beer consumed in the U.S.

We aggregate data from each *Dominick’s* store into one of two price zones. For more details about this procedure, see Hellerstein (2008). We define a product as one six-pack serving of a brand of beer and quantity as the total number of servings sold per week. We define a market as one of *Dominick’s* price zones in one week. Products’ market shares are calculated with respect to the potential market which is defined as the total beer purchased each week in supermarkets by the residents of the zip codes in which each *Dominick’s* store is located. We define the outside good to be all beer sold by other supermarkets to residents of the same zip codes as well as all beer sales in the sample’s *Dominick’s* stores not already included in our sample. We have a total of 16 brands in our sample (5 domestic and 11 imported), each with 404 observations (202 weeks spanning the period from June 6, 1991 to June 1, 1995 in each of two price zones). We supplement the *Dominick’s* data with information on manufacturer costs, product characteristics, advertising, and the distribution of

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5 The data can be found at http://gsbwww.uchicago.edu/kilts/research/db/dominicks/.

6 Features of the *Dominicks*’ wholesale-price data confirm that brewers control distributors prices to the supermarket. Across individual *Dominicks*’ stores, which may each be served by a different distributor, each with an exclusive territory, the variation in UPC-level wholesale prices is less than one cent. Asker (2004) notes that one cannot distinguish distributors by observing the wholesale prices they charge to individual *Dominicks* stores. This supports the industry lore that distributors pricing is coordinated by brewers and is not set separately by each distributor to each retail outlet.

7 Canadian Trade Commissioner (2000).

8 As our data focus on one metropolitan statistical area, we do not need to control for variation in retail alcohol sales regulations. Such regulations can differ considerably across states.

9 The zones are defined by *Dominick’s* mainly on the basis of customer demographics. Although they do not report these zones, we identify them through zip-code level demographics (with a few exceptions, each *Dominick’s* store in our sample is the only store located in its zip code) and by comparing the average prices charged for the same product across stores. We classify each store according to its pricing behavior as a low- or high-price store and then aggregate sales across the stores in each pricing zone. This aggregation procedure retains some cross-sectional variation in the data which is helpful for the demand estimation. Residents’ income covaries positively with retail prices across the two zones.
consumer demographics. Product characteristics come from the ratings of a Consumer Reports study conducted in 1996. Summary statistics for the price data are provided in Table 1. Table 2 reports summary statistics for the characteristics data used in the demand estimation.

2.3 Preliminary Descriptive Results

We begin the analysis by documenting in several simple regressions whether Dominick’s imported-beer prices are systematically related to movements in bilateral nominal exchange-rates. These results can provide a benchmark against which we can measure the performance of the structural model. We estimate three price equations:

\[
\ln p_r^{jzt} = c_j + \zeta_z + \theta_t + \alpha \ln e_{jt} + \beta \ln co_{jzt} + \epsilon_{jzt}
\]

(1)

\[
\ln p_w^{jzt} = c_j + \zeta_z + \theta_t + \alpha \ln e_{jt} + \beta \ln co_{jzt} + \epsilon_{jzt}
\]

(2)

\[
\ln p_r^{jzt} = c_j + \zeta_z + \theta_t + \alpha \ln p_w^{jzt} + \epsilon_{jzt}
\]

(3)

where the subscripts \(j, z,\) and \(t\) refer to product, zone, and week respectively; \(p_r\) is the product’s retail price; \(p_w\) is the product’s wholesale price; \(c_j, \theta_t,\) and \(\zeta_z\) are product, week and zone dummies respectively that proxy among other things for demand shocks that may affect a brand’s price independent of exchange rates; \(e\) is the bilateral nominal exchange rate (domestic-currency units per unit of foreign currency); \(co_{jzt}\) denotes a set of variables that proxy for cost shocks that again may affect prices; such variables include measures of domestic (U.S.) wages, the price of barley in each country producing beer in our sample, the price of electricity in the Chicago area, and - for foreign brands - wages in each beer exporting country in our sample; \(\epsilon\) is a random error term. All variables are specified in levels, and not first differences, as our focus is on the long-run pass-through of exchange rate changes, and not the short-term dynamics.

Table 3 reports results from OLS estimation of the pricing equations. Columns 1 and 3 report results from specifications that include the full set of controls specified above, while in columns 2 and 4 the cost controls are omitted (since the latter do not vary at the weekly level). The results across the two specifications are remarkably similar. The average pass-through elasticity \(\alpha\) for the retail price is based on column 2 - 6.7 percent and is significant at the one-percent level. The regression establishes a roughly 7-percent benchmark for the retailer’s pass-through elasticity, that we will try to explain within the framework of the structural model. The fourth column of Table 3 reports similar results from estimation of the wholesale-price pricing equation, equation (2): Its pass-through elasticity is 4.7 percent, and the coefficient is again highly significant. Finally, the fifth column of Table 3 reports the results from an OLS regression of each brand’s retail price on its own wholesale price. The coefficient on the wholesale price is not significantly different from 100, which is consistent with the results from
the other columns: Exchange-rate shocks that are passed on by manufacturers to the retailer appear to be immediately and almost fully passed on to consumer prices.

This preliminary analysis reveals that local-currency price stability is an important feature of this market: only around 7 percent of an exchange-rate change is transmitted to a beer’s retail price. Where does the other 93 percent go? Existing literature on exchange rate pass-through has identified three potential sources of this incomplete transmission: a non-traded cost component in the manufacturing of traded goods, variable markups, and nominal price rigidities. The goal of our paper is to quantify the relative contribution of each of these sources in explaining incomplete pass-through.

2.4 Patterns of Price Adjustment in the Data

A rough idea of the timing and frequency of price changes in the beer market can be obtained from Figure 1, which plots the retail and wholesale prices for a six-pack of the British brand Bass Ale. The figure covers the full sample period, from the middle of 1991 to the middle of 1995. The plot serves to illustrate several interesting points.

First, the figure demonstrates the advantage of observing price data at a weekly frequency. Such data are ideal for analyzing the role of price stickiness, since we clearly see prices remaining constant for several weeks, and then jumping up (in a discrete step) to a new level. This pattern in the price adjustment process is exactly the one we would expect with price stickiness. That said, the infrequent adjustment of prices is by itself no definitive proof that price rigidities exist, as it is in principle possible that prices do not change simply because nothing else changes.

Second, a substantial fraction of the price variation reflects temporary price reductions (sales). To investigate if there was a systematic pattern in sales, we estimated sales determinant specifications similar to those in Pesendorfer (2002). Interestingly, in our sample of beer brands, sales appear to be random, in the sense that we did not find anything that could predict the timing of a sale. In particular, we did not find that the time that had elapsed since a previous sale, holiday dummies, or sales for other brands, could predict a current sale for a particular brand. This apparent randomness of sales in our sample is consistent with a Varian-type explanation of temporary price reductions. A potential problem with such an explanation is that in Varian’s model firms randomize prices each period, so that the notion of a “regular price” does not exist. However, in the Varian (1980) model there are no costs of price adjustment. Introducing costs of price adjustment (as we do in this paper) can explain the existence of “regular prices” that is evident in Figures 1 and 2. We should point out that we by no means claim that the patterns we observe in our data for beer generalize to other product categories. In fact, there is substantial evidence, based on Pesendorfer (2002) and recent papers by Hendel and Nevo (2006a and 2006b), that intertemporal considerations, such as those emphasized in models by Sobel (1984) and Pesendorfer (2002), are important in determining sales in other products.
that are more storable. Our findings regarding the timing of sales for beer are consistent with our premise throughout the paper that consumers do not store beer, so that intertemporal considerations, while potentially important in other markets, are not a first order concern here. We discuss this issue in more detail in Section 3.3.

Third, a striking feature of Figure 1 is that retail prices always adjust when wholesale prices adjust. So it seems that the main reason retail prices do not change in this market is that there is little reason for them to change (the cost facing retailers as measured by the wholesale price does not change). This is to be contrasted with the pattern we observe at the wholesale level: despite enormous variation in exogenous (to the industry) factors affecting manufacturer costs (i.e., exchange-rate fluctuations), wholesale prices remain unchanged for long periods of time.

A final point that Figure 1 together with similar plots for other brands illustrates is that price adjustment is not synchronized across brands. Given the strategic interactions between firms, this asynchronous price adjustment can generate significant price inertia, even if the nominal price rigidities facing each individual manufacturer or retailer are estimated to be small.

3 Model

This section describes the supply and demand sides of the model we use to identify the sources of incomplete exchange rate pass-through, and in particular the role of price rigidities.

3.1 Supply

We model the supply side of the market using a linear-pricing model in which manufacturers, acting as Bertrand oligopolists with differentiated products, set their prices followed by retailers who set their prices taking the wholesale prices they observe as given. Thus, a double margin is added to the marginal cost of the product before it reaches the consumer. Our framework builds on Hellerstein’s (2008) work on the beer market, but makes one key modification to her model: We introduce price rigidities both at the wholesale and retail level; the effect of these price rigidities is to cause firms to potentially deviate from their first-order conditions.

The strategic interaction between manufacturer and retailer is as follows. First, the manufacturer decides whether or not to change the product’s price taking into account the current period’s observables (costs, demand conditions, and competitor prices), and the anticipated reaction of the retailer. If she decides to change the price, then the new price is determined based on the manufacturer’s first-order conditions. Otherwise the wholesale price is the same as in the previous period. Next, the retailer observes the wholesale price set by the manufacturer and decides whether or not to change the product’s retail price. If the retail price changes, then the new retail price is determined according to the retailer’s first-order conditions. Otherwise the retail price is the same as in the previous period.
To characterize the equilibrium we use backward induction and solve the retailer’s problem first.

### 3.1.1 Retailer

Consider a retail firm that sells all of the market’s $J$ differentiated products. Let all firms use linear pricing and face constant marginal costs. The profits of the retail firm in market $t$ are given by:

$$
\Pi_r^t = \sum_j (p_r^j - p_w^j - ntc_r^j) s_{jt}(p_r^t) - A_{jt}^r
$$

The first part of the profit expression is standard. The variable $p_r^j$ is the price the retailer sets for product $j$, $p_w^j$ is the wholesale price paid by the retailer for product $j$, $ntc_r^j$ are local non-traded costs paid by the retailer to sell product $j$, and $s_{jt}(p_r^t)$ is the quantity demanded of product $j$ which is a function of the prices of all $J$ products. The new element in our approach is the introduction of the second term, $A_{jt}^r$, which captures the fixed cost of changing the price of product $j$ at time $t$. This cost is zero if the price remains unchanged from the previous period, but takes on a positive value, known to the retailer, but unknown to the econometrician, if the price adjusts in the current period:

$$
A_{jt}^r = 0 \text{ if } p_r^j = p_r^{j-1} \\
A_{jt}^r > 0 \text{ if } p_r^j \neq p_r^{j-1}
$$

We interpret the adjustment cost $A_{jt}^r$ as capturing all possible sources of price rigidity. These can include the management’s cost of calculating the new price; the marketing and advertising expenditures associated with communicating the new price to customers; the costs of printing and posting new price tags, etc... The particular interpretation of $A_{jt}^r$ is not important for our purposes. What is important is that this cost is independent of the sales volume; it is a discrete cost that the retailer pays every time the price adjusts from the previous period. The indexing of $A$ by product $j$ and time $t$ in our notation corresponds to the most flexible specification, in which the price adjustment cost is allowed to vary by product and time. One could potentially impose more structure by assuming that adjustment costs are constant over time, and/or constant across products.

The implication of the adjustment cost in the profit function is that it can cause firms to deviate from their first-order conditions, even if the retailer acts as a profit maximizer. Specifically, in the data we will observe one of two cases:

**Case 1: The price changes from the previous period, that is** $p_r^j \neq p_r^{j-1}$.

In this case the retailer solves the standard profit maximization problem to determine the new

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10We use the term “non-traded” to indicate that these costs are paid in dollars no matter what the origin of the product is. Hence, non-traded costs will not be affected by exchange rate shocks.
optimal price, and the observed retail price $p_{jt}^r$ will have to satisfy the first-order profit-maximizing conditions:

$$s_{jt} + \sum_{k=1,2,...,J} \left( p_{kt}^r - p_{kt}^w - ntc_{kt}^r \right) \frac{\partial s_{kt}}{\partial p_{jt}} = 0.$$  (6)

This gives us a set of $J$ equations, one for each product. One can solve for the retailer’s markups by defining a $J \times J$ matrix $\Omega_{rt}$, called the retailer reaction matrix (the matrix of demand substitution patterns), with element $S_{jk}^r = \frac{\partial s_{kt}}{\partial p_{jt}}$, $j,k = 1,...,J$, that is the marginal change in the $k$th product’s market share given a change in the $j$th product’s retail price (and, in theory, a $J \times J$ matrix $T^r$ with the $(j, k)$th element equal to 1 if both products $j$ and $k$ are sold by the retailer, and equal to zero otherwise. In our case $T^r$ is all ones so we omit it for simplicity going forward). The stacked first-order conditions can be rewritten in vector notation:

$$s_t + \Omega_{rt}(p_t^r - p_t^w - ntc_t^r) = 0$$  (7)

and inverted together in each market to get the retailer’s pricing equation, in vector notation:

$$p_t^r = p_t^w + ntc_t^r - \Omega_{rt}^{-1} s_t$$  (8)

where the retail price for product $j$ in market $t$ will be the sum of its wholesale price, non-traded costs, and markup.

The presence of the adjustment costs $A_{jt}^r$ in the profit function implies that for the retailer to change her price in the current period, it will have to be the case that the extra profits associated with the new price are at least as large as the adjustment cost:

$$\left( p_{jt}^r - p_{jt}^w - ntc_{jt}^r \right) s_{jt}(p_t^r) + \sum_k \left( p_{kt}^r - p_{kt}^w - ntc_{kt}^r \right) s_{kt}(p_t^r) - A_{jt}^r \geq$$

$$\left( p_{jt-1}^r - p_{jt}^w - ntc_{jt}^r \right) s_{jt}(p_{jt-1}^r, p_{kt}^r) + \sum_k \left( p_{kt}^r - p_{kt}^w - ntc_{kt}^r \right) s_{kt}(p_{jt-1}^r, p_{kt}^r), j \neq k$$  (9)

where $s_{jt}(p_{jt-1}^r, p_{kt}^r)$ denotes the counterfactual market share that product $j$ would have, if the retailer had kept the price unchanged from $p_{jt-1}^r$, and $p_{kt}^r$ denotes the prices of the other products $k$ that may or may not have changed from the previous period. We assume that the retailer considers whether to change only one of its product’s prices at a time, even if it is a multiproduct firm. The above inequality simply states that the profits the retailer makes by adjusting the price of product $j$ in the current period have to be greater than the profits the retailer would have achieved, if she had not changed the price (in which case the first-order condition of profit maximization would have been violated, but the
retailer would have saved on the adjustment costs $A_{jt}^r$. By rearranging terms we can use the above inequality to derive an upper bound $\overline{A}_{jt}^r$ for the price adjustment costs of product $j$:

$$A_{jt}^r \leq \overline{A}_{jt}^r = (p_{jt}^r - p_{jt}^w - ntc_{jt}) s_{jt}(p_{jt}^r) - (p_{jt-1}^r - p_{jt}^w - ntc_{jt}) s_{jt}(p_{jt-1}^r, p_{kt}^r) + \sum_k [(p_{kt}^r - p_{kt}^w - ntc_{kt}^r) (s_{kt}(p_{kt}^r) - s_{kt}(p_{jt-1}^r, p_{kt}^r))], j \neq k$$  \hspace{1cm} (10)

**Case 2: The price remains unchanged from the previous period, that is $p_{jt}^r = p_{jt-1}^r$.**

In this case the first-order conditions of profit maximization do not necessarily hold. If the retailer does not adjust the price of product $j$ in period $t$, it must be the case that the profits she makes from keeping the price constant are at least as large as the profits the retailer would have made if she had adjusted the price according to the first-order condition minus the adjustment costs associated with the price change:

$$\left( p_{jt-1}^r - p_{jt}^w - ntc_{jt}^r \right) s_{jt}(p_{jt-1}^r, p_{kt}^r) + \sum_k \left( p_{kt}^r - p_{kt}^w - ntc_{kt}^r \right) s_{kt}(p_{kt}^r) \geq$$

$$\left( p_{jt}^{rc} - p_{jt}^w - ntc_{jt}^{rc} \right) s_{jt}^{rc}(p_{jt}^{rc}, p_{kt}^r) + \sum_k \left( p_{kt}^{rc} - p_{kt}^w - ntc_{kt}^{rc} \right) s_{kt}^{rc}(p_{kt}^{rc}, p_{kt}^r) - A_{jt}^r, \ j \neq k$$  \hspace{1cm} (11)

where $p_{jt}^{rc}$ denotes the counterfactual price the retailer would have charged if he behaved according to the optimality conditions, and $s_{jt}^{rc}(p_{jt}^{rc}, p_{kt}^r)$ is the counterfactual market share that would correspond to this optimal price holding the prices of the competitor products at their observed levels. Just like in Case 1, we can rewrite the above inequality to derive a lower bound $\underline{A}_{jt}^r$ for the adjustment costs:

$$\underline{A}_{jt}^r \geq \overline{A}_{jt}^r = (p_{jt}^{rc} - p_{jt}^w - ntc_{jt}^{rc}) s_{jt}^{rc}(p_{jt}^{rc}, p_{kt}^r) - (p_{jt-1}^r - p_{jt}^w - ntc_{jt}) s_{jt}(p_{jt-1}^r, p_{kt}^r) + \sum_k [(p_{kt}^r - p_{kt}^w - ntc_{kt}^r) (s_{kt}^{rc}(p_{kt}^{rc}, p_{kt}^r) - s_{kt}(p_{kt}^r))], j \neq k$$  \hspace{1cm} (12)

The essence of our empirical approach to quantify the adjustment costs can be described as follows. First, we estimate the demand function. Once the demand parameters have been estimated, the market share function $s_{jt}(p_{jt}^r)$ as well as the own and cross price derivatives $\frac{\partial s_{jt}}{\partial p_{jt}^r}$ and $\frac{\partial s_{kt}}{\partial p_{jt}^r}$ can be treated as known. Next we exploit the first-order conditions for each product $j$ (6) to estimate the non-traded costs and markups of product $j$, but contrary to the approach typically employed in the Industrial Organization literature, we use only the periods in which the price of product $j$ adjusts, to back out costs and markups. In periods when the price does not adjust, the non-traded costs are not identified based on the first-order conditions; however, we can derive estimates of the non-traded costs for
these periods by imposing some additional structure on the problem, e.g., by modeling non-traded costs parametrically as a function of observables along the lines described in the next section. Once estimates of non-traded costs for these periods have been derived, we can calculate the counterfactual price \( p_{jt}^{rc} \) that the retailer would have charged if there were no price rigidities and she behaved according to the profit maximization conditions, as well as the associated counterfactual market share \( s_{jt}^{rc}(p_{jt}^{rc}, p_{kt}) \).

In the final step, we can exploit inequalities (10) and (12) to derive upper and lower bounds of the adjustment costs \( A_{jt}^{r} \).

Note that in the above framework price rigidities as captured by the adjustment cost \( A_{jt}^{r} \) affect pricing behavior in two ways. First, there is a direct effect: price rigidities may prevent the retailer from adjusting the price of any particular product if the adjustment cost associated with this product’s price change exceeds the additional profit. Second, there is an indirect effect that operates through the effect that price rigidities may have on the prices of competing products. When our retailer sets the price of product \( j \), she conditions on the prices of the other products with which product \( j \) competes. If these prices remain constant (potentially because of the existence of price rigidities), then the price change of product \( j \) may be smaller than the one we would have observed if price rigidities were altogether non-existent. The existence of this indirect effect implies that relatively small adjustment costs can potentially lead to significant price inertia. Accordingly, the magnitude of the adjustment costs cannot by itself provide a measure of the significance of price stickiness in explaining incomplete pass-through. To assess the overall impact of price adjustment costs it is necessary to perform simulations to compare the pricing behavior we observe to the one that would prevail with fully flexible prices.

### 3.1.2 Manufacturers

Let there be \( M \) manufacturers that each produce some subset of the market’s \( J \) differentiated products. Each manufacturer chooses its wholesale price \( p_{jt}^{w} \) taking the retailer’s anticipated behavior into account. Manufacturer \( w \)’s profit function is:

\[
\Pi_{jt}^{w} = \sum_{j} \left( (p_{jt}^{w} - c_{jt}^{w}(t_{cjt}, n_{tcjt})) \right) s_{jt}(p_{jt}^{r}(p_{jt}^{w})) - A_{jt}^{w} \tag{13}
\]

where \( c_{jt}^{w} \) is the marginal cost incurred by the manufacturer to produce and sell product \( j \); this cost is in turn a function of traded costs \( t_{cjt} \), and destination-market specific non-traded costs \( n_{tcjt} \). As noted above, the distinction between traded and non-traded costs is based on the currency in which these costs are paid; traded costs are by definition incurred in the manufacturer’s home country currency, and are subject to exchange rate shocks, while (dollar-denominated) non-traded costs are not. The term \( A_{jt}^{w} \) denotes the price adjustment cost incurred by the manufacturer. The interpretation of this cost is similar to the one for the retail adjustment cost; it is a discrete cost that is paid only when the
manufacturer adjusts the price of product \( j \):

\[
A^w_{jt} = \begin{cases} 
0 & \text{if } p^w_{jt} = p^w_{jt-1} \\
> 0 & \text{if } p^w_{jt} \neq p^w_{jt-1}
\end{cases}
\]

(14)

Given this structure, we can use the same procedure as the one we applied to the retailer’s problem in order to derive upper and lower bounds for the manufacturer adjustment cost. The derivation of the manufacturer bounds is however more complicated as the manufacturer needs to take into account the possibility that the retailer does not adjust her price due to the existence of the retailer adjustment cost.

As with the retailer, in the data we will observe one of two cases:

**Case 1: The wholesale price changes from the previous period, that is** \( p^w_{jt} \neq p^w_{jt-1} \).

Due to the existence of the retail adjustment cost, it is – in principle – possible in this case that the retail price does not adjust, while the wholesale price does adjust. However, in our data we do not observed a single instance of this happening. We therefore concentrate our discussion on the case where the retail price adjusts when the wholesale price adjusts.

Assuming that manufacturers act as profit maximizers, each wholesale price \( p^w_{jt} \) must satisfy the first-order profit-maximizing conditions given that it has been adjusted from the previous period:

\[
s_{jt} + \sum_{k=1,2,...,J} (p^w_{kt} - c^w_{kt}) \frac{\partial s_{kt}}{\partial p^w_{kt}} = 0.
\]

(15)

This gives us another set of \( J \) equations, one for each product. Let \( \Omega_{wt} \) be the manufacturer’s reaction matrix with elements \( \frac{\partial s_{jt}(p^w_{jt}(p^w_{jt})))}{\partial p^w_{jt}} \), the change in each product’s share with respect to a change in each product’s wholesale price. The manufacturer’s reaction matrix is a transformation of the retailer’s reaction matrix: \( \Omega_{wt} = \Omega_{pt}^t \Omega_{rt} \) where \( \Omega_{pt} \) is a \( J \)-by-\( J \) matrix of the partial derivative of each retail price with respect to each product’s wholesale price. Each column of \( \Omega_{pt} \) contains the entries of a response matrix computed without observing the retailer’s marginal costs. The properties of this manufacturer response matrix are described in greater detail in Villas-Boas (2007).\(^{11}\)

The manufacturers’ marginal costs (which are a function of the traded and non-traded costs, \( tc^w_t \) and \( ntc^w_t \) respectively) are then recovered by inverting the manufacturer reaction matrix \( \Omega_{wt} \) according to:

\[
p^w_t = c^w_t - \Omega_{wt}^{-1} s_t
\]

(16)

For product \( j \), the wholesale price is the sum of the manufacturer traded costs, non-traded costs, and

---

\(^{11}\)To obtain expressions for this matrix, one uses the implicit-function theorem to totally differentiate the retailer’s first-order condition for product \( j \) with respect to all retail prices and with respect to the manufacturer’s price \( p^f_j \).
markup function. The manufacturer of product $j$ can use her estimate of the retailer’s non-traded costs and reaction function to compute how a change in the manufacturer price will affect the retail price for the product.

For the manufacturer to have changed her price from the previous period, it has to be the case that the profits she makes from having changing the price (net of the price adjustment cost $A_{wjt}$) exceed the profits that the manufacturer would have made if she had left the wholesale price unchanged at $p_{wjt-1}$:

$$
(p_{wjt} - c_{jt}) s_{jt}(p_{rjt}(p_{wjt})) + \sum_k (p_{wkt} - c_{kt}) s_{kt}(p_{rkt}(p_{wkt})) - A_{wjt} \geq (p_{wjt-1} - c_{jt}) s_{jt}(p_{rjt}(p_{wjt-1}, p_{rkt})) + \sum_k (p_{wkt} - c_{kt}) s_{kt}(p_{rjt}(p_{wkt-1}, p_{rkt}), p_{rkt}), j \neq k \quad (17)
$$

This condition is similar to inequality (10) for the retailer, with a slight difference: the counterfactual market share $s_{jt}^c$, that the manufacturer would face if she left the price of product $j$ unchanged is a function of the counterfactual retail price $p_{rjt}^c$ that the retailer would charge when faced with an unchanged wholesale price $p_{wjt-1}$. But given the existence of the retail adjustment cost, this counterfactual price can follow one of two scenarios: the first one is that the retailer does not change the price from the previous period, so that $p_{rjt}^c = p_{rjt-1}$; the second possibility is that the retailer adjusts her price according to the retailer’s first-order conditions (6). Hence, before one can use the above inequality to infer the upper bound of the manufacturer’s adjustment cost, it is necessary to solve the retailer’s problem to determine how the retailer’s price response. Specifically, using notation for a single-product retailer for simplicity, if:

$$
(p_{rjt-1} - p_{wjt-1} - ntc_{jt}^r) s_{jt}^c(p_{rjt-1}, p_{rkt}) \geq (p_{rjt}^c - p_{wjt-1} - ntc_{jt}^r) s_{jt}^c(p_{rjt}(p_{wjt-1}), p_{rkt}) - A_{rjt}, j \neq k
$$

the retailer will leave her price unchanged. Otherwise, she will adjust her price to $p_{rjt}^c$, where $p_{rjt}^c$ is itself determined according to the first-order condition:

$$
s_{jt}^c + (p_{rjt}^c - p_{rjt-1} - ntc_{jt}^r) \frac{\partial s_{jt}^c}{\partial p_{rjt}^c} = 0
$$

Once the optimal pricing behavior of the retailer, conditional on the wholesale price being equal to $p_{wjt-1}$ has been determined, the upper bound of the manufacturer’s adjustment cost $A_{wjt}$ can be derived.
based on the inequality:

\[
A_{jt}^w \leq A_{jt}^w = \left[ (p_{jt}^w - c_{jt}) s_{jt}(p_{jt}^r, p_{jt}^w) + \sum_k (p_{kt}^w - c_{kt}) s_{kt}(p_{kt}^r, p_{kt}^w) \right] \\
- \left[ (p_{jt-1}^w - c_{jt}) s_{jt}(p_{jt-1}^r, p_{kt}^w) + \sum_k (p_{kt}^w - c_{kt}) s_{kt}(p_{kt-1}^r, p_{kt}^w) \right], \quad j \neq k \tag{18}
\]

where \( p_{jt}^{rc} \) is either equal to \( p_{jt-1}^r \) or determined according to the retailer’s first-order condition, and \( s_{jt}^c \) is evaluated accordingly.

**Case 2: The wholesale price does not change from the previous period, that is \( p_{jt}^w = p_{jt-1}^w \).**

The lack of price adjustment in this case implies that the wholesale price is not necessarily determined based on the manufacturer first-order condition. Regarding the retail price, it is again possible that the retailer adjusts the retail price in periods when the wholesale price remains unchanged. However, in practice we rarely observe this case in the data. Hence, we concentrate on the case where both wholesale and retail prices remain unchanged, that is \( p_{jt}^w = p_{jt-1}^w \) and \( p_{jt}^r = p_{jt-1}^r \).

Given that the manufacturer does not adjust the wholesale price, it has to be the case that the profits she makes at \( p_{jt-1}^w \) are at least as large as the profit she *would have* made if she had changed the price to a counterfactual wholesale price \( p_{jt}^{wc} \) according to the profit maximization condition and paid the associated adjustment cost

\[
(p_{jt-1}^w - c_{jt}) s_{jt}(p_{jt-1}^r, p_{kt}^w) + \sum_k (p_{kt}^w - c_{kt}) s_{kt}(p_{kt-1}^r, p_{kt}^w) \geq \]

\[
(p_{jt}^{wc} - c_{jt}) s_{jt}^c(p_{jt}^r, p_{jk}^w, p_{kt}^w) + \sum_k (p_{kt}^w - c_{kt}) s_{kt}^c(p_{kt}^r(p_{jt}^{wc}, p_{kt}^w), p_{kt}^w) - A_{jt}^w, \quad k \neq j \tag{19}
\]

As with the case of the retailer, we can exploit this insight to derive a lower bound \( A_{jt}^w \) for the price adjustment cost:

\[
A_{jt}^w \geq A_{jt}^w = \left[ (p_{jt}^{wc} - c_{jt}) s_{jt}^c(p_{jt}^r, p_{jk}^w, p_{kt}^w) + \sum_k (p_{kt}^w - c_{kt}) s_{kt}^c(p_{kt}^r(p_{jt}^{wc}, p_{kt}^w), p_{kt}^w) \right] \\
- \left[ (p_{jt-1}^w - c_{jt}) s_{jt}(p_{jt-1}^r, p_{kt}^w) + \sum_k (p_{kt}^w - c_{kt}) s_{kt}(p_{kt-1}^r, p_{kt}^w) \right], \quad k \neq j \tag{20}
\]

The determination of the counterfactual optimal wholesale price \( p_{jt}^{wc} \) and the associated counterfactual market share \( s_{jt}^c \) is however more involved in this case, as the manufacturer has to take into account the reaction of the retailer, who may or may not adjust her price in response to a wholesale price change.

To find the price \( p_{jt}^{wc} \) the manufacturer would set if she were willing to incur the adjustment cost,
we proceed as follows. First, we consider the case in which the retail price would have changed in response to the wholesale price change. In this case $p_{jkt}^{wc}$ would be determined according to equation (16) which reflects the manufacturer’s first-order condition; the inverted manufacturer reaction matrix $\Omega_{wt}^{-1}$ in this equation incorporates the optimal pass-through of the wholesale price change onto the retail price.

Next we consider the case in which the retailer does not adjust her price in response to the wholesale price change. Even though as noted above we never observe this case in the data, the possibility that the wholesale price change does not get passed through by the retailer is factored in when manufacturers set prices. If the manufacturer anticipates an equilibrium in which the retailer does not adjust her price, the optimal manufacturer behavior will be to change the wholesale price up to the point where the retailer is just indifferent between changing the retail price and leaving it the same as in the previous period, that is, for the retailer (written for a single-product firm for simplicity):

$$
(p_{jkt}^{r} - p_{jkt}^{wc} - ntc_{jkt}^{r}) s_{jkt}^{r}(p_{jkt}^{r}, p_{kkt}) = (p_{jkt}^{rc} - p_{jkt}^{wc} - ntc_{jkt}^{r}) s_{jkt}^{r}(p_{jkt}^{rc}, p_{kkt}) - A_{jkt}^{r}, \ k \neq j
$$

(21)

The left hand side of the above equation denotes the profits the retailer would make if she did not pass-through the change in the wholesale price. The right hand side represents the profits the retailer would make if she changed the retail price to $p_{jkt}^{rc}$, where the latter is determined based on the retailer’s first-order condition $s_{jkt}^{r} + \left(p_{jkt}^{rc} - p_{jkt}^{wc} - ntc_{jkt}^{r}\right) \frac{\partial s_{jkt}^{r}}{\partial p_{jkt}^{rc}} = 0$. To find the wholesale price $p_{jkt}^{wc}$ the manufacturer would charge in this case, equation (21) can be solved simultaneously with the retailer’s first-order condition for $p_{jkt}^{wc}$ and $p_{jkt}^{rc}$.

The final step in determining the counterfactual optimal wholesale price $p_{jkt}^{wc}$ that the manufacturer would choose if she changed the wholesale price from the previous period is to compare the manufacturer profits for the case where the retailer adjusts the price, to the manufacturer profits for the case where the retailer does not pass-through the wholesale price change, in which case the wholesale price will be set according to (21). The manufacturer will pick the $p_{jkt}^{wc}$ that corresponds to the higher profits. Once the wholesale price is found, the optimal retail price response and associated market share can be determined as well, and inserted in (20) in order to infer the manufacturer adjustment cost lower bound.

### 3.2 Demand

The estimation of costs, markups, and adjustment costs requires consistent estimates of the demand function as a first step. Market demand is derived from a standard discrete-choice model of consumer behavior. Given that the credibility of all our results will ultimately depend on the credibility of the demand system, it is imperative to adopt as general and flexible a framework as possible to model
consumer behavior. We use the BLP random-coefficients model described in Hellerstein (2008), as this model was shown to fit the data well, while imposing very few restrictions on the substitution patterns. In the following we provide a brief overview of the model, and refer the reader to Nevo (2001) and Hellerstein (2008) for a more detailed discussion of the implementation.

Let the indirect utility $u_{ijt}$ that consumer $i$ derives from consuming product $j$ at time $t$ take the quasi-linear form:

$$u_{ijt} = x_{jt} \beta_i - \alpha_i p_{jt} + \xi_{ijt} + \varepsilon_{ijt}, \quad i = 1, ..., I, \quad j = 1, ..., J, \quad t = 1, ..., T. \quad (22)$$

where $\varepsilon_{ijt}$ is a mean-zero stochastic term. The utility from consuming a given product is a function of a vector of product characteristics $(x, \xi, p)$ where $p$ are product prices, $x$ are product characteristics observed by the econometrician, the consumer, and the producer, and $\xi$ are product characteristics observed by the producer and consumer but not by the econometrician. Let the taste for certain product characteristics vary with individual consumer characteristics:

$$\left( \begin{array}{c} \alpha_i \\ \beta_i \end{array} \right) = \left( \begin{array}{c} \alpha \\ \beta \end{array} \right) + \Pi D_i + \Sigma v_i \quad (23)$$

where $D_i$ is a vector of demographics for consumer $i$, $\Pi$ is a matrix of coefficients that characterize how consumer tastes vary with demographics, $v_i$ is a vector of unobserved characteristics for consumer $i$, and $\Sigma$ is a matrix of coefficients that characterizes how consumer tastes vary with their unobserved characteristics. Conditional on demographics, the distribution of consumer unobserved characteristics is assumed to be multivariate normal. The demographic draws give an empirical distribution for the observed consumer characteristics $D_i$. Indirect utility can be expressed in terms of mean utility $\delta_{jt} = \beta x_{jt} - \alpha p_{jt} + \xi_{jt}$ and deviations (in vector notation) from that mean $\mu_{ijt} = [\Pi D_i \Sigma v_i] * [p_{jt} \ x_{jt}]$:

$$u_{ijt} = \delta_{jt} + \mu_{ijt} + \varepsilon_{ijt} \quad (24)$$

Finally, consumers have the option of purchasing an “outside” good; that is, consumer $i$ can choose not to purchase any of the products in the sample. The price of the outside good is assumed to be set independently of the prices observed in the sample. The mean utility of the outside good is normalized to be zero and constant over markets. The indirect utility from choosing to consume the outside good is:

$$u_{i0t} = \xi_{0t} + \pi_0 D_i + \sigma_0 v_{i0} + \varepsilon_{i0t} \quad (25)$$

Let $A_j$ be the set of consumer traits that induce purchase of good $j$. The market share of good $j$ in

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12 The existence of an “outside” good means that the focus on a single retailer (Dominick’s) does not imply that this retailer has monopoly power in the retail market; consumers faced with a price increase at Dominick’s have the option of switching to beer sold in other supermarkets, which represents the “outside” good in our framework.
market $t$ is given by the probability that product $j$ is chosen:

$$s_{jt} = \int_{\zeta \in A_j} P^*(d\zeta)$$

(26)

where $P^*(d\zeta)$ is the density of consumer characteristics $\zeta = [D \ \nu]$ in the population. To compute this integral, one must make assumptions about the distribution of the error term $\varepsilon_{ijt}$. Assuming that $\varepsilon_{ijt}$ is i.i.d. with a Type I extreme-value distribution, the market share function becomes:

$$s_{jt} = \int_{\mu_{it}} e^{\delta_{jt} + \mu_{ijt}} 1 + \sum_k e^{\delta_{kt} + \mu_{ikt}} f(\mu_{it}) d\mu_{it}$$

(27)

The integral is approximated by the smooth simulator which, given a set of $N$ draws from the density of consumer characteristics $P^*(d\zeta)$, can be written:

$$s_{jt} = \frac{1}{N} \sum_{i=1}^{N} \frac{e^{\delta_{jt} + \mu_{ijt}}}{1 + \sum_k e^{\delta_{kt} + \mu_{ikt}}}$$

(28)

Given these predicted market shares, we search for the demand parameters that implicitly minimize the distance between these predicted market shares and the observed market shares by using a generalized method-of-moments (GMM) procedure.

### 3.3 Discussion

The framework we just described is static. In this section we explore the implications of dynamic considerations for our analysis.

Dynamic considerations may affect the analysis in several ways. First, to the extent that consumers and/or retailers hold inventories of beer, the demand and supply side parameter estimates obtained by the static approach may be biased. Specifically, on the demand side, Hendel and Nevo (2006a, 2006b) have shown that when consumers stockpile in response to temporary price reductions, static demand estimates may overstate the long-run price elasticities of demand by a factor of 2 to 6.\textsuperscript{13} On the supply side, Aguirregabiria (1999) has analyzed the pricing behavior of a monopolistically-competitive retailer that holds inventories in a central store and delivers goods from this store to individual outlets. He shows that in the presence of fixed ordering costs and nominal price rigidities inventory dynamics have a significant effect on the retailer’s decision to change a brand’s price; ignoring such dynamics may hence lead to biased estimates of the importance of nominal price rigidities.\textsuperscript{14} Fortunately, these

\textsuperscript{13} Their numbers refer to laundry detergents, however, which consumers stockpile much more intensively than they do beer.

\textsuperscript{14} The role of inventories is also emphasized in a recent paper by Alessandria, Kaboski and Midrigan (2008), which investigates exchange rate pass-through directly.
concerns that both build on the importance of inventories appear to be less relevant in our case. The industry wisdom is that consumers typically consume beer within a few hours after its purchase, so that consumer stockpiling is not a first-order concern. On the supply side, state and local regulations concerning the distribution of all alcohol, including beer, in Illinois stipulate that it is illegal for the central store of a retail chain to maintain inventories of beer and to deliver them to individual outlets. This must be done by firms exclusively licensed to be distributors. It is also illegal for beer to be transported from one outlet to another by the central store. So from the point of view of the central store or the individual outlet, there is no inventory problem associated with beer, unlike most other products which are distributed by the central store. As the central store does not keep inventories of beer (indeed cannot by law), there is no relationship between inventory decisions and prices. And there is no incentive for individual outlets to maintain inventories over a period of longer than a week, as they can get a new shipment each week from the distributor, rather than bearing the costs of holding inventories themselves. An additional piece of evidence that stores do not hold beer inventories comes from the wholesale price series in Figures 1 and 2. In the Dominick’s data, reported wholesale prices are based on average acquisition costs. If inventories were large and moved slowly, we would not expect to see the sharp "round-trip" drops we see in our wholesale price graphs.

A second way in which dynamics can enter the analysis is through intertemporal effects on the demand side, stemming for example from habit formation. The implications of such dynamics are explored in Slade (1998) and in a recent theoretical paper by Ravn, Schmitt-Grohe and Uribe (2007). While we believe that such dynamics are potentially important, it is not feasible to incorporate them into our current framework in a way that would allow us to estimate the model. To our knowledge, the only paper that has succeeded in estimating a model similar to ours but with dynamics on the demand side is the one by Slade (1998), which employs a linear demand specification. In comparison to Slade’s paper, our approach puts more emphasis on the modelling of the demand side by allowing for a high degree of product differentiation and consumer heterogeneity and flexible substitution patterns. We believe that this is important in the context of the question we are trying to address, since one of the explanations for incomplete pass-through is markup adjustment by retailers and manufacturers, and there is a close link between markups and demand structure. However, the price we pay for the complexity of the demand system is that we have to abstract from demand side dynamics.

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16 For more on Illinois liquor regulations, see the homepage of the Illinois Liquor Control Commission at http://www.state.il.us/LCC/.
17 As Aguirregabiria argues, “There are some brands for which the central store does not keep inventories. Some of them are very perishable goods which are delivered daily from wholesalers to outlets (e.g. fresh vegetables, fish, some types of bread, etc.) In other cases, they are brands from manufacturers with efficient distribution networks that allow them to deliver their brands to individual outlets. From the point of view of the company’s central store, there is not any inventory problem associated with those brands. Since we are interested in the relationship between price and inventory decisions we only consider those brands for which the central store keeps inventories” (Aguirregabiria 1999, p. 286).
18 We should note however that our static approach is standard in the Industrial Organization literature.
Finally, dynamic considerations may impact the analysis through the modeling of the price adjustment costs. As pointed out by Dixit (1991), under the assumption of rational expectations there is an option value in not adjusting prices, which magnifies the effects of even small costs of adjustment. The static approach we adopt, which is most similar to the models of Akerlof and Yellen (1985) and Mankiw (1985), fails to model this option value and hence may result in estimates of adjustment costs that are biased upwards. Nakamura (2007) in a recent paper adopts a dynamic framework to estimate menu costs, which are modeled as random i.i.d. variables. While the menu costs are derived within a dynamic framework, computational constraints force her to estimate the cost parameters based on static profit maximization conditions that do not take menu costs into account, so that there is a tension between the model used to obtain menu costs and conduct simulations, and the static model that is used to obtain the cost estimates. Our approach has the advantage that it enables us to maintain a theoretically consistent approach throughout the analysis. Still, we are concerned about the implications of Dixit’s arguments for our analysis, so we turn to those next.

Suppose that instead of modeling the manufacturer’s problem according to equation (13), we adopted a dynamic framework, in which (for the sake of simplicity) a single-product firm seeks to maximize the expected discounted value of future profits:

$$
\sum_{t=0}^{\infty} E_t \delta^t \Pi_{jt}^w = \sum_{t=0}^{\infty} E_t \delta^t \{p_{jt}^w - c_{jt}(t_{c_{jt}}^w, n_{tc_{jt}}^w) s_{jt}(p_t^r(p_t^w)) - A_{jt}^w\}
$$

where $\delta$ denotes the discount rate and as before:

$$
A_{jt}^w = \begin{cases} 
0 & \text{if } p_{jt}^w = p_{jt-1}^w \\
> 0 & \text{if } p_{jt}^w \neq p_{jt-1}^w
\end{cases}
$$

Given that there are no dynamics on either the demand or cost side (the market shares depend only on current and not past prices), it follows that if the firm were faced with cost shocks that were considered permanent, the dynamic problem would be equivalent to the static one up to a discount factor: firms would be maximizing the sum over the (discounted) profits of each period, which in the absence of any cross-period price effects or inventories would be equivalent to maximizing period-by-period profits. In contrast, if shocks are perceived as temporary, firms will have an incentive to wait to adjust prices, so that the dynamic solution will be different from the static one.

The exchange rate shocks that move the traded component of costs ($t_{c_{jt}}^w$) around, and are the focus of our analysis, are highly persistent; empirical work in International Economics has consistently...
failed to reject the unit root hypothesis for exchange rates. We therefore believe that the premise that such shocks are perceived as permanent is not unrealistic. Of course, exchange rates are not the only source of cost variation. Beer producers also face cost variation caused by changes in the prices of inputs such as barley, glass and energy. However, there is nothing that comes close to the variation caused by exchange rate volatility! In Figure 3, we plot the exchange rate facing beer producers over our sample period along with the prices for glass and energy (all series are normalized to start at 1 at the beginning of our sample). While there is some variation in the prices of energy and glass, the exchange rate variation trumps everything else.

In sum, the static approach we adopt provides a reasonable approximation to a dynamic setup if the main source of cost variation is exchange rates, and if exchange rates are highly persistent. We believe that both premises are supported by the data.

Finally, we should point out that even though the static model is equivalent to a dynamic model with static expectations in terms of their implications for pass-through, the exact magnitudes of the price adjustment costs will differ depending on which of the two models we use to derive them. In particular, if we use the dynamic model (with static expectations) to derive adjustment cost bounds, a single-product version of inequality (18) becomes:

$$A_{jt}^w \leq \frac{\delta}{1-\delta} \sum_{t=0}^{\infty} \delta^t \{(p_{jt}^w - c_{jt}^w) s_{jt}(p_{rt}^w) - (p_{jt-1}^w - c_{jt-1}^w) s_{jt}(p_{jt-1}^w, p_{kt}^w)\}, \ k \neq j$$

which, assuming a constant discount rate, would reduce to:

$$A_{jt}^w \leq \frac{\delta}{1-\delta} \{(p_{jt}^w - c_{jt}^w) s_{jt}(p_{rt}^w) - (p_{jt-1}^w - c_{jt-1}^w) s_{jt}(p_{jt-1}^w, p_{kt}^w)\}, \ k \neq j$$

That is the firm would compare the fixed cost of adjusting its price today, to the present discounted value of all future additional profits that it would make if it changed its price. A similar argument applies to the lower bound $A_{jt}^w$. This implies that if we adopted a dynamic framework, in order to obtain the bounds for the price adjustment costs consistent with the dynamic model we would have to multiply the static estimates by the factor $\frac{\delta}{1-\delta}$, which is of the same order of magnitude as the inverse of the interest rate. This would “blow up” the static estimates for the price adjustment costs. Note however, that this would leave our simulations and decomposition results at the end of the paper completely unchanged: in the dynamic setup, both sides of the inequalities (10) and (18) that determine whether the firm will adjust its price in response to an exchange rate shock or not, would be multiplied by the same factor $\frac{\delta}{1-\delta}$, so that this factor cancels out. This is why we emphasize that the term “price adjustment costs” is meaningful only within the context of the particular model we use as the basis of the empirical analysis, and that our results regarding pass-through are robust as long as this term is used consistently throughout the analysis.
4 Empirical Approach

Our empirical approach has two components: estimation and simulation. At the estimation stage, we estimate the demand parameters, the traded and non-traded costs and markups of the retailer and manufacturers, and the upper and lower bounds for the price adjustment costs. As noted above, these bounds are not by themselves informative regarding the role of price rigidities in explaining the incomplete cross-border cost shock transmission. To see why, suppose we estimate the adjustment cost of changing the price of a particular product \( j \) to be very small at the retail level. Still, as long as the adjustment cost is nonzero, it will cause the price of product \( j \) to remain unchanged in some periods. This in turn will affect the pricing of competing products: if the price of \( j \) does not change, then the prices of the products that do change may change by less than they would if all prices adjusted. Similarly at the wholesale level, the presence of a small adjustment cost at the retail level may cause the manufacturer to keep the wholesale level price constant if she anticipates that the retailer will not pass-through the change. Hence, a small adjustment cost may cause significant price inertia at both the retail and wholesale levels.

To assess the overall impact of adjustment costs on pricing behavior we employ simulation. In particular, we compute the industry equilibrium that would emerge if the dollar appreciated (depreciated) and prices were fully flexible, that is all adjustment costs were set to zero. Next we compare this equilibrium to the one that prevails in the presence of price rigidities. We interpret the differential response of prices across the two cases as a measure of the overall impact of nominal price rigidities. In the following we describe each step of our empirical approach in more detail.

4.1 Estimation

The estimation stage consists of the following steps:

1. Demand Estimation

The estimation of the demand system follows Hellerstein (2008). We model the mean utility associated with product \( j \) at time \( t \) as follows:\footnote{The demand model is also indexed by price zone \( z \). In each period we have observations for two separate price zones. To keep the exposition simple, we omit the subscript \( z \) from our notation.}

\[
\delta_{jt} = \beta d_j - \alpha p_{jt} + \Delta \xi_{jt}
\]

where the product fixed effects \( d_j \) proxy for both the observed characteristics \( x_{jt} \) in the term in equation (22) and the mean unobserved characteristics. The residual \( \Delta \xi_{jt} \) captures deviations of the unobserved product characteristics from the mean (e.g., time-specific local promotional

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20 The demand model is also indexed by price zone \( z \). In each period we have observations for two separate price zones. To keep the exposition simple, we omit the subscript \( z \) from our notation.
activity) and is likely to be correlated with the price $p_{jt}$; for example, an increase in the product's promotional activity may simultaneously increase the mean evaluation of this product by consumers and a rise in its retail price. Addressing this simultaneity bias requires finding appropriate instruments, that is a set of variables $z_{jt}$ that are correlated with the product price $p_{jt}$ but are orthogonal to the error term $\Delta \xi_{jt}$. Factor prices and exchange rates satisfy this condition as they are unlikely to have any relationship to promotional activities while they are by virtue of the supply relation correlated with product prices. To construct our instruments we interact hourly wages in each country's beverage industry with weekly bilateral exchange rates and indicator variables for each brand; this allows each product's price to respond differently to a given supply shock.

2. **Back out the non-traded retail costs** $ntc_{jt}^r$ **and retail markups using data only for the periods in which retail prices adjust.**

Once the parameters of the demand system have been estimated, we compute the market share function $s_{jt}(p_t^r)$ as well as the own and cross price derivatives $\frac{\partial s_{jt}}{\partial p_{jt}}$. Then we use the retailer's first-order conditions for each product $j$ (6) to estimate the non-traded retail costs of product $j$.

3. **Model these non-traded costs parametrically as a function of observables (e.g., zone dummies, month dummies, local wages), and estimate the parameters of this function using data from the periods for which we observe retail price adjustment.**

The procedure described under Step 2 allows us to back out the retailer's non-traded costs for the periods for which we observe the price of a product adjusting, so that we can reasonably assume that the retailer sets the new price according to the first-order conditions. However, this approach does not work in periods in which the price does not change. To get estimates of the non-traded costs for these periods we employ the following procedure:

First, we collect the data on the non-traded costs $ntc_{jt}^r$ in Step 2 for the periods in which the price of product $j$ adjusted. Then we model these costs parametrically as a function of observables:

$$ntc_{jt}^r = c_j + \gamma_z d_z + \gamma_w w_t^d + \eta_{jt}$$

where $c_j$ are brand fixed effects, $d_z$ are price zone dummies, and $w_t^d$ denote local wages. We run the above regression using data from the periods we observe price adjustment, and then use the parameter estimates to construct the predicted non-traded costs for the periods for which we do not observe price adjustment.

4. **Derivation of upper and lower bounds for the retailer price adjustment costs** $A_{jt}^r$. 

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With the demand parameter and non-traded cost estimates in hand, we employ (10) and (12) to derived the upper and lower bounds of the retailer adjustment costs $A_{jt}$. The computation of the upper bound is straightforward: in (10) all variables are observed, except for the counterfactual market share $s_{jt}^c(p_{jt-1}^r, p_{kt}^r)$ that product $j$ would have if the retailer did not change her price from the previous period. This counterfactual share can however be easily evaluated once the demand parameters are estimated, given that the market share function is known.

The computation of the lower bound based on (12) requires the derivation of the counterfactual optimal price $p_{jt}^c$ that the retailer would charge if she changed the retail price from the previous period, and the associated market share $s_{jt}^c(p_{jt}^c, p_{kt}^r)$. These are computed using (8) which reflects the first-order condition of the profit-maximizing retailer.

5. **Back out the manufacturer marginal cost $c_{jt}^m$ using data only for the periods in which wholesale prices adjust.**

The procedure here is similar as the one we employ to derive the non-traded costs for the retailer. In periods when the wholesale price changes, manufacturers behave according to their first-order conditions. Hence, we can use equation (16) to back out the manufacturer marginal cost $c_{jt}^m$.

6. **Model the manufacturer marginal cost parametrically as a function of observables (e.g., time dummies, local and foreign wages), and estimate the parameters of this function using data from the periods for which we observe wholesale price adjustment.**

The manufacturer first-order conditions we utilize under Step 5 allow us to back out the total marginal cost of the manufacturer; however they do not tell us how to decompose this cost into a traded and non-traded component. Furthermore, it is not possible to back out the marginal manufacturer costs for the periods when wholesale prices do not adjust based on this procedure, given that the first-order conditions do not necessarily hold then. To accomplish the above tasks, we model the total manufacturer costs parametrically as a function of observables, and estimate this function using data from the periods of wholesale price adjustment only. Specifically, we assume that the manufacturer marginal cost $c_{jt}^m$ takes the form:

$$c_{jt}^m = \exp(\theta_j + \omega_{jt})(w_t^d)^{\theta_{dw}}(e_{jt}w_t^f)\hat{F}_j\theta_{fw}(p_{bjt})\hat{D}_j\theta_{dp}(e_{jt}p_{bjt})\hat{F}_j\theta_{fp}$$

or, in log-terms:

$$\ln c_{jt}^m = \theta_j + \theta_{dw} \ln w_t^d + F_j \theta_{fw} \ln(e_{jt}w_t^f) + D_j \theta_{dp} \ln(p_{bdt}) + F_j \theta_{fp} \ln(e_{jt}p_{bjt}) + \omega_{jt}$$
where \( w_d^l \) and \( w_f^l \) denote local domestic and foreign wages respectively, \( e_{jt} \) is the bilateral exchange rate between the producer country and the U.S., \( p_{bjt} \) is the price of barley in the country of production of brand \( j \), \( F_j \) is a dummy that is equal to 1 if the product is produced by a foreign supplier, and zero otherwise, and \( D_j \) is a dummy that is equal to 1 if the product is produced by a domestic supplier, and zero otherwise. For the function to be homogeneous of degree 1 in factor prices, we require \( \theta dw + F_j \cdot \theta fw + D_j \cdot \theta dp + F_j \cdot \theta fp = 1 \). Equation (30) can be easily estimated by Least Squares.

The estimation of the above equation for the manufacturer marginal cost serves two purposes. First, it allows us to decompose the total marginal cost into a traded and a non-traded component. Recall that by definition the traded component refers to the part of the marginal cost that is paid in foreign currency and hence is subject to exchange-rate fluctuations. For domestic producers the traded component will be (by definition) zero. Foreign producers selling in the U.S. will generally have both traded and local non-traded costs. The latter are captured in the above specification by the term \( (w_f^l)^d w \) that indicates the dependence of foreign producers’ marginal costs on the local wages in the U.S.. The specification in (29) can be used to demonstrate two important facts regarding foreign suppliers’ costs. First, foreign producers selling to the U.S. will typically experience substantially more volatility than domestic producers due to their exposure to exchange-rate shocks. Second, as long as the local non-traded cost component is nonzero (so that \( \theta fw + \theta fp < 1 \)), the dollar denominated marginal cost of foreign producers will change by a smaller proportion than the exchange rate. This incomplete marginal cost response may partially explain the incomplete response of exchange-rate changes on to prices.

Estimation of the marginal cost equation (30) furthermore allows us to use the parameter estimates to construct predicted values for the manufacturer traded and non-traded costs for the periods in which wholesale price adjustment is not observed.

7. Derivation of upper and lower bounds for the wholesale price adjustment costs \( A_{jt}^w \).

The final step is to use all parameter estimates obtained in the previous steps to compute the upper and lower bounds of the manufacturer price adjustment costs based on (18) and (20). Consider inequality (18) first that determines the adjustment cost upper bound. Once steps 1-6 are completed, all variables in this inequality are known, except for the counterfactual retail price \( p_{rjt}^c \) that the retailer would charge if the manufacturer did not change her price in that period. The counterfactual price \( p_{rjt}^c \) can take on one of two values: it is either equal to \( p_{rjt-1}^c \).

\[ 21 \text{Given the assumption of } \theta_{d+w} + F_j \cdot \theta_{f+w} = 1 \text{ which guarantees homogeneity of degree 1 of the marginal-cost function in factor prices, if local nontraded costs are zero, then } \theta_{d+w} = 0 \text{ and } \theta_{f+w} = 1. \text{ In contrast, with positive nontraded costs we will have } \theta_{f+w} < 1. \]
or it is determined according to the retailer's first-order condition, conditional on the retailer observing the wholesale price $p_{wt-1}^w$. To determine which of the two prices the retailer will choose, we first solve for the optimal price that the retailer would pick if she behaved according to her profit maximization condition. Then we compare the retail profits evaluated at this retail price, to the profits that the retailer would make if she kept the retail price unchanged at $p_{jt-1}^r$. The retailer will choose the price associated with the higher retail profits. Once the counterfactual retail price $p_{jt}^c$ has been determined this way, the associated counterfactual market share $s_{jt}^c(p_{jt}^w(p_{jt-1}^w,p_{kt}^r))$, $k \neq j$, can easily be evaluated.

Next consider inequality (20) that determines the adjustment cost lower bound. Again, all variables in this inequality can be treated as known once steps 1-6 are completed, except for the counterfactual retail and wholesale prices, $p_{jt}^c$ and $p_{jt}^w$ respectively, which we would observe if the manufacturer changed her price from the previous period. To determine those, we consider two cases. In the first case the retail price changes from the previous period; the optimal prices $p_{jt}^w$ and $p_{jt}^c$ are then determined according to the manufacturer and retailer first-order conditions, equations (16) and (6) respectively, with the inverted manufacturer reaction matrix $\Omega_{wt}^{-1}$ reflecting the optimal pass-through of the wholesale price change onto the retail price. Let $\pi_{1}^{wc}$ denote the manufacturer profits associated with the so-computed prices $p_{jt}^w$ and $p_{jt}^c$.

Next, consider the case in which the retail price does not change, even though the wholesale price does. As noted earlier, the optimal manufacturer pricing behavior in this case will involve changing the wholesale price up to the point where the retailer is just indifferent between changing her price and keeping it constant at $p_{jt-1}^r$. The optimal wholesale price will then be determined based on equation (21) along the lines discussed in the previous section. Let $\pi_{2}^{wc}$ denote the manufacturer profits associated with the prices $p_{jt}^w$ and $p_{jt-1}^r$ in this case.

If $\pi_{1}^{wc} > \pi_{2}^{wc}$, the manufacturer will set the wholesale price anticipating that the retailer will adjust her price too. Hence, the counterfactual prices $p_{jt}^{w\text{c}}$ and $p_{jt}^{c\text{c}}$ will satisfy the conditions described under the first case above. If $\pi_{1}^{wc} < \pi_{2}^{wc}$, the manufacturer will price the product anticipating that the retailer will not adjust her price. The resulting counterfactual wholesale price will then satisfy the indifference condition discussed under the second case, while the retail price will remain unchanged at $p_{jt-1}^r$. Once the counterfactual wholesale and retail prices have been determined, evaluation of the adjustment cost lower bound based on (20) is straightforward.

### 4.2 Simulations and Decomposition of Incomplete Exchange-Rate Pass-Through

To assess the overall impact of adjustment costs on pricing behavior we employ simulations. First, we compute the industry equilibrium that would emerge if a particular firm faced an exchange rate shock and prices were fully flexible, that is, all adjustment costs were equal to zero. In a second set of
simulations, we derive the industry equilibrium under the presence of nominal rigidities. We interpret the differential response of prices across the different cases as a measure of the impact of nominal price rigidities.

To be more specific, to identify the channels through which nominal rigidities affect prices, we conduct three separate simulations that we briefly describe below.

Simulation 1: Simulate the effect of a 5% exchange-rate change when there are no price rigidities. To recover manufacturer and retailer pass-through coefficients we derive the effect of a shock to foreign firms’ marginal costs (e.g., an exchange rate shock) on all firms’ wholesale and retail prices by computing a new Bertrand-Nash equilibrium. Suppose that an exchange rate shock hits the traded component of the $j$th product’s marginal cost (that is the component that is denominated in foreign currency). To compute the transmission of this shock to wholesale prices, we substitute the new vector of traded marginal costs, $tc^w_t$, into the system of $J$ nonlinear equations that characterize manufacturer pricing behavior, and then search for the wholesale price vector $p^*_w$ that will solve the system:

$$p^*_w = c^w_{jt}(tc^w_t, nt^w_{jt}) - \sum_{k \in \Gamma_m} (S_{wt} * T_{w})^{-1} s_{kt} \text{ for } j = 1, 2, ..., J_t. \tag{31}$$

To compute pass-through coefficients at the retail level, we substitute the derived values of the vector $p^*_w$ into the system of $J$ nonlinear equations for the retail firms, and then search for the retail price vector $p^*_r$ that will solve it:

$$p^*_r = p^*_w + nt_{j,t} - \sum_{k \in \Gamma_r} (S_{wt} * T_{w})^{-1} s_{kt} \text{ for } j, k = 1, 2, ..., J_t. \tag{32}$$

The transmission of the original marginal-cost shock to the retail price is given by $\left(\frac{dp^r}{dp^w}\right)' \frac{dp^w}{dtc^w_t}$.

This first simulation gives us the benchmark pass-through elasticities that would emerge in the absence of nominal rigidities. Using our estimation results on the traded and non-traded components of manufacturer and retailer costs respectively, we can further decompose these pass-through elasticities into the part that is due to the presence of traded costs and the part that reflects markup adjustment. Given the Cobb-Douglas specification for the manufacturer marginal cost described earlier, the contribution of the traded costs will be given by the sum of the coefficients $\theta_{fw} + \theta_{fp}$ (or, in other words, the contribution of local costs to generating incomplete pass-through will be captured by the coefficient on domestic wages $\theta_{dw}$). The difference between the derived wholesale pass-through and the one attributed to traded costs reflects markup adjustment on the part of the manufacturer. Similarly at the retail level, we can use our estimates of the retailer non-traded costs to compute the effect that such costs have in generating incomplete pass-through of wholesale to retail prices; this effect will be
given by \( (d \ln (p_{wj} + ntc_{j})/d \ln p_{wj}) \). The difference between the derived retailer pass-through and the one attributed to retailer non-traded costs captures the markup adjustment on the part of the retailer.

Next we consider the case where nominal price rigidities are present. Because firms in our framework are not symmetric, and price changes will not be synchronized, characterizing the equilibrium in this case becomes extremely involved. To keep the problem tractable and get a sense of how price rigidities affect prices, we confine our discussion to two extreme cases; one in which the firm facing the exchange rate shock assumes that all other competitors will adjust their prices, and one in which the firm under consideration assumes that competitor prices will remain fixed as a result of nominal rigidities. These cases correspond to the following two simulations:

**Simulation 2:** Simulate the effect of a 5% exchange-rate change assuming that the foreign-brand producer facing the exchange rate shock faces no price adjustment costs and also assumes that competitor prices will remain fixed. In this case, the new industry equilibrium is computed as in Simulation 1, but by additionally imposing that all other product prices remain unchanged. This simulation captures the indirect or strategic aspect of repricing costs; even if nominal rigidities do not prevent a particular firm from adjusting its price, this adjustment may be smaller if the firm assumes that nominal rigidities will keep competitor prices fixed compared to the case without any rigidities. A comparison between the results from Simulation 2 and Simulation 1 allows us to capture precisely this effect.

**Simulation 3:** Simulate the effect of a 5% exchange-rate change assuming that the foreign-brand producer facing the exchange rate shock also faces price adjustment costs. In this case, the new industry equilibrium is computed taking into account the fixed adjustment costs for the foreign brand affected by the exchange-rate shock included in equations (31) and (32). As discussed earlier, our approach does not allow us to pin down these adjustment costs, but only derive upper and lower bounds for these costs. In all simulations involving price adjustment costs we employ the estimated upper bounds.

5 Results

This section first discusses results from the estimation of the demand system. It then describes estimates of brand-level markups, non-traded costs, and upper and lower bounds for the retailer and the manufacturer adjustment costs. Finally, it reports the results from the simulations that allow us to decompose the incomplete transmission of exchange rate shocks into its sources in order to quantify the relative contribution of local costs, markup adjustment, and repricing costs in generating the documented local currency price stability.
5.1 Demand Estimation: Logit Model

Table 4 reports results from estimation of demand using the multinomial logit model. Due to its restrictive functional form, this model will not produce credible estimates of pass-through. However, it is helpful to see how well the instruments for price perform in the logit demand estimation before turning to the full random-coefficients model.22

Table 4 suggests that the instruments have power. The first-stage F-test of the instruments, at 34.45, is significant at the one-percent level. The consumer’s sensitivity to price should increase after we instrument for unobserved changes in characteristics. That is, consumers should appear more sensitive to price once we instrument for the impact of unobserved (by the econometrician, not by firms or consumers) changes in product characteristics on their consumption choices. It is promising that the price coefficient falls from -0.93 in the OLS estimation to -2.43 in the IV estimation. Note that the 95-percent confidence interval of the latter coefficient does not include the value of the former. Results reported in columns 2 and 4 of the table show that including holiday dummies does not affect the demand coefficients in either the OLS or the IV estimation.

5.2 Demand: Random-Coefficients Model

Table 5 reports results from estimation of the demand system. We allow consumers’ income to interact with their taste coefficients for price and percent alcohol. As we estimate the demand system using product fixed effects, we recover the mean consumer-taste coefficients in a generalized-least-squares regression of the estimated product fixed effects on product characteristics (maltiness, bitterness, hoppiness and percent alcohol).

The coefficients on the characteristics generally appear reasonable. As consumers’ income rises, they become less price sensitive. The random coefficient on income, at 40.37, is significant at the five-percent level. The mean preference in the population is not amenable to a bitter taste in beer, which has a negative and significant coefficient. As the percent alcohol rises across brands, the mean utility in the population also rises, an intuitive result. There is heterogeneity in the population with respect to this characteristic: Those with higher incomes get less utility from a high percent of alcohol in their beer, that is, prefer light beers, as one can glean from the negative and significant random coefficient of -51.84. This is consistent with industry lore: Higher income individuals tend to prefer light and imported beers.23

22 An appendix (available online at http://www.ny.frb.org/research/economists/hellerstein/papers.html) reports the first-stage results for demand. Most of the coefficients have the expected sign: as hourly compensation increases, the retail price of each product should increase. T-statistics calculated using Huber-White robust standard errors indicate that most of the coefficients are significant at the 5-percent level.

23 Our demand side estimates are similar to those obtained in Hellerstein (2008), who estimated a random-coefficients demand model with the same data, but after aggregating the data up to the monthly frequency. This makes us confident that our price elasticities of demand are not affected by the type of arguments made in Hendel and Nevo (2006b).
5.3 Retail Markups and Non-Traded Costs

Table 6 reports retail and wholesale prices and markups for selected imported brands. The markups are derived using firms’ first-order conditions. Under the assumption of no adjustment costs, the markups would be derived using the first-order conditions of every product in every period. Under the alternative assumption of some adjustment costs, the markups are derived in each period by using the first-order conditions of only those products whose prices adjust from the previous period. As discussed earlier, many of the price changes in our data reflect promotions, during which the price of a particular brand is reduced for a few weeks (see also Figures 1 and 2). A striking characteristic of these promotions is that product prices return to their exact pre-promotion level once the promotion is over. In theory, the transition from the discount price to the pre-promotion level is a price change that could be handled in the same manner as a level change in price (after all, firms do incur some cost every time they change the posted price); yet, given that firms seem to charge exactly the same price that they were charging before the promotion, we were skeptical about the plausibility of the assumption that the post-promotion prices are determined based on firms’ first-order conditions. To be safe, we conducted the empirical analysis both ways, first applying the FOC’s to all periods in which the price changed (including changes associated with promotions), and then excluding those time periods during which firms charged the same price as before the promotion. The results did not differ in any significant manner across the two approaches, but the second approach significantly reduces the number of observations associated with a price change than we can exploit in the empirical analysis. Still, in the remainder of the paper we report results based on this second, more conservative approach, as we are more comfortable with the assumption that FOC’s hold only in those periods during which a firm charges a price that is genuinely different from the price charged in earlier periods. In general, the markups that are derived based on this approach appear reasonable and consistent with industry wisdom as evident from Table 6. The same applies to the retailer’s non-traded costs that were derived based on the regression reported in Table 8.

Using a similar procedure we derive the marginal costs facing each beer manufacturer, and then employed the regression described in equation (30) in Section 4.1 to obtain an estimate of the “local content” of foreign manufacturers’ marginal cost. This local content is reflected in the magnitude of the “domestic U.S. wages” coefficient that captures the cost share accounted for by domestic labor. As discussed earlier (see p. 8, in particular footnote 7), because distributors pricing is coordinated by brewers, we treat the manufacturer and distributor as one entity, so that the “local” manufacturer costs include the marketing and distribution costs incurred by the distributor. With a highly significant coefficient of 0.53 the share of local costs appears to be substantial. It implies that a big part of foreign manufacturers’ costs of selling in the U.S. market are not affected by exchange-rate fluctuations. Hence it comes as no surprise that foreign producers do not fully adjust their U.S. dollar prices in response
to exchange rate changes. This finding implies that even without menu costs, the existence of local non-traded costs can generate a significant degree of inertia in local currency prices. However, whether local costs can by themselves fully explain the observed patterns in the price data remains an open question. We therefore turn to an examination of the price adjustment costs next.

5.4 Adjustment Costs

Table 11 reports the mean estimates by brand of the upper and lower bounds on the retailer’s adjustment costs. The upper and lower bounds generally are consistent for each brand as well across the brands. The mean lower bounds on adjustment costs to retail prices range from $13.20 for Harp to $112.54 for Heineken, with a mean lower bound across foreign brands of $46.14, and a mean upper bound of almost $255. The entries in the third and fourth columns report the sum of the upper or lower bounds for each brand’s price-adjustment costs divided by the retailer’s total revenue from that brand over the full sample period. These numbers are more comparable to those of the Levy et al (1997) and Dutta et al (1999) studies, which divide the costs of repricing calculated for only those periods when prices change divided by the revenue earned by the firm across all periods, whether prices change or not. Manufacturer adjustment costs, reported in Table 12, are roughly the same order of magnitude as those for retailers, though smaller as a share of revenue because of firms’ less frequent price adjustment. Their upper bounds range from 0.2 percent of revenue for Heineken to 4.4 percent of revenue for Molson Golden, and their lower bounds from 0.01 percent of revenue for Harp to 0.67 percent of revenue for Beck’s. The sum of the upper bounds of repricing costs across all foreign brands is 3.5 and 1.85 percent of total revenue for the retailer and manufacturer, respectively, and for the lower bounds is 0.63 and 0.11 percent of total revenue for the retailer and manufacturer, respectively.

As pointed out earlier, the particular estimates of price adjustment costs are not of interest here, as these numbers are meaningful only in the particular context of our model, and are accordingly meant to be used only in the context of our model in order to perform simulations. Nevertheless, we find it interesting that the order of magnitude of our estimates is similar to the one obtained in other studies that used a completely different methodology.

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24 This comparison is subject to the same caveats mentioned in the Introduction, p. 6, footnote 2.
25 For example, if we interpreted these numbers in the context of the dynamic model outlined at the end of Section 3.3, then we would have to divide the current estimates by the interest rate to obtain the price adjustment costs corresponding to the dynamic setup. For example, if we used an interest rate of 10%, the resulting lower bounds on price adjustment cost estimates would be around 6% and 11% of revenue for retailers and manufacturers respectively. Levy et al and Dutta et al’s estimates for retailer costs, which were based on the sum of the labor and material costs associated with a price change, were 0.70% and 0.59% of the revenue respectively.
26 This procedure involves summing up the estimated upper bounds for the price adjustment costs for those periods within a year in which prices did change, and then dividing this sum by the firm’s revenue over the entire year. We should emphasize that because our “price adjustment costs” are defined in the most general sense to include all factors that may prevent firms from changing their nominal prices (and not just the literal labor and material costs of changing prices), our numbers are not directly comparable to Levy et al’s. Still, it is interesting that despite these differences the
Finally, Table 13 reports the results of a fixed-effects panel regression of the derived retail menu costs on a dummy for a level change in a brand’s price as well as a dummy for sales, that is, temporary price reductions. Adjustment costs appear to be significantly higher for level changes in prices, averaging about $900 for a “permanent” level price change, and closer to $50 for a temporary price reduction. This finding provides empirical support for a recent paper by Kehoe and Midrigan (2007) that is based on the premise that the fixed cost of changing a regular price is larger than the fixed cost of a temporary reduction.

5.5 Simulations

Using the full random-coefficients model and the derived measures of traded, nontraded, and repricing costs, we conduct the counterfactual experiments described earlier to analyze how firms react to exchange-rate shocks. This subsection presents and discusses the results from these experiments. We consider the effect of a five-percent foreign currency appreciation on foreign brands’ prices in three scenarios, each with a different assumption about the nature of the repricing costs faced by foreign brands. We use the results from all three scenarios to assess the relative importance of foreign firms’ local non-traded costs, markup adjustment, and repricing costs in their incomplete pass-through of exchange-rate fluctuations.

The counterfactual experiments consider the effect of a five-percent appreciation of the relevant foreign currency on the prices of a British, German, Mexican, and Dutch brand (Bass, Beck’s, Corona, and Heineken, respectively) in twelve exercises reported in Table 14. There are three panels in the table, each one corresponding to one of the simulations we described above. The first column of the table reports for each simulation the manufacturer pass-through elasticity of the original shock that is due to local dollar-denominated costs incurred by the manufacturer. The second column reports the pass-through of the original shock to the wholesale price that is attributable to manufacturer markup adjustment. The third column reports the (incomplete) pass-through of the original shock to the retail price due to the presence of a local component in retail costs. The last column reports the incomplete pass-through of the original shock to the retail price due to the retailer’s markup adjustment.

**Simulation 1: Simulate the effect of a 5% appreciation of the relevant foreign currency when there are no repricing costs.** The first counterfactual experiment examines the manufacturers’ and the retailer’s pass-through following a 1-percent appreciation of the relevant foreign currency when they face no repricing costs. Its results are reported in the top panel of Table 14. The median pass-through of the exchange rate shock on manufacturer’s total marginal cost is 47 percent, which is determined by the coefficient on local wages (0.53) from the regressions reported in Table 10: The
average local non-traded cost incurred by a foreign manufacturer is approximately 50 percent of her total costs. Thus, a nontrivial amount of non-traded value is added at this stage of the distribution chain. Next, manufacturer markup adjustments are large and vary quite a bit across brands in this counterfactual: With markup adjustments taken into account, the median pass-through elasticity of the exchange rate shock to the wholesale price ranges from 34.4 percent for Bass to 44.5 percent for Heineken. It is 39.8 percent across all brands. With retailer local costs taken into account, the median pass-through becomes 30.7 percent and ranges from 26.3 percent for Corona to 40.0 percent for Becks. Finally, the retailer appears to adjust her markup significantly for some brands and only marginally for others: The median retailer pass-through elasticity across all brands is 20.0 percent and ranges from 0 percent for Bass to 26.7 percent for Heineken.

**Simulation 2:** Simulate the effect of a 5% appreciation of the relevant foreign currency assuming that the foreign firm affected by the exchange-rate shock faces no repricing costs and also assuming that competitor prices will remain fixed. The second counterfactual experiment considers whether manufacturers and the retailer adjust their prices following a 5-percent appreciation of the relevant foreign currency if they take their competitor prices to be fixed. In this case, other manufacturer repricing costs faced by Bass and Becks prevent the wholesale prices of these two brands from adjusting, which translates to zero pass-through elasticities. But what is interesting in this counterfactual is the fact that the pass-through elasticities in those cases where prices do adjust (Corona and Heineken) are now lower compared to Simulation 1. This additional reduction in the pass-through elasticities also captures the indirect or strategic effect of repricing costs: because each brand assumes that repricing costs will prevent its competitors from changing their prices, the brand’s own response to the exchange rate shock is less pronounced than it would be with flexible prices. Overall, this effect accounts for the reduction of the manufacturer pass-through elasticity from 39.8% in Simulation 1, to 17.8% in Simulation 2, and the reduction of the retail pass-through elasticity from 20.0% in Simulation 1, to 0% in Simulation 2.

**Simulation 3:** Simulate the effect of a 5% appreciation of the relevant foreign currency assuming that the foreign brand affected by the exchange-rate shock faces also fixed repricing costs. The third counterfactual experiment considers how manufacturers and the retailer adjust their prices following a 5-percent appreciation of the relevant foreign currency if they must incur fixed repricing costs to alter their prices. Its results are reported in the bottom panel of Table 14. The median pass-through of the exchange rate change on manufacturer total marginal cost is again 50 percent as the share of non-traded costs is unaffected by the nature of the counterfactual. But final manufacturer pass-through elasticities now are zero across brands. Thus taking into account a brand’s own price adjustment costs reduced the average pass-through elasticity from 17.9% in Simulation 2 to
0% in Simulation 3. Note however that this reduction is due to the zero transmission of the exchange rate shock on the wholesale prices of *Corona* and *Heineken* due to these two brands’ manufacturer repricing costs. In contrast, retail repricing costs do not contribute to any further reductions in the pass-through elasticities. This is consistent with the patterns we documented earlier that suggest that retail prices always adjust whenever wholesale prices adjust.

### 5.6 Decomposition of the Incomplete Transmission

Table 15 decomposes the sources of the incomplete transmission of the exchange-rate shock to retail prices that is documented in Table 14. The first column of the table reports the share of the incomplete transmission that can be attributed to a local dollar-denominated cost component in manufacturers’ marginal costs. The second column reports the share that can be attributed to markup adjustment by manufacturers following the shock (separate from any costs of repricing faced). Columns three and seven report the shares of the incomplete transmission attributable to the fixed costs of price adjustment incurred by the manufacturer and retailer, respectively, when they change their own prices (the direct effect of repricing costs). Columns four and eight report the shares of the incomplete transmission attributable to the effect that the fixed costs of repricing faced by competitors have on the manufacturer and retailer’s pricing behaviors (the indirect or strategic effect). The fifth column reports the share attributable to a local-cost component in the retailer’s marginal costs, and the sixth column the share attributable to the retailer’s markup adjustment, separate from any markup adjustment associated with repricing costs.

Manufacturers’ local non-traded costs play the most significant role in the incomplete transmission of the original shock to retail prices. Following a 5-percent appreciation of the relevant foreign currency, it is responsible for roughly half, or 53 percent, of the observed retail-price inertia. Manufacturers’ markup adjustment accounts for 7.5 percent of the remaining adjustment, their competitors price rigidities for 22 percent, and their own repricing costs for another 17.9 percent. After accounting for manufacturers own repricing costs, the retailer’s markup adjustment and own repricing costs have a negligible role in explaining the incomplete transmission. These results support the initial intuition conveyed by Figures 1 and 2 that the effects of fixed repricing costs are most evident in the infrequent adjustment of wholesale prices, while such costs play only a minor role in explaining the inertia of retail prices.

### 6 Conclusions

This paper set out to build a framework that can be used to identify the determinants of local currency price stability in the face of exchange rate fluctuations. The empirical model we developed incorporates the three main potential sources identified in the literature: local non-traded costs; markup adjustment;
and fixed costs of repricing. Our analysis yields several interesting findings.

First, at the descriptive level, we document that while both wholesale and retail prices do not change every period, retail prices always respond to changes in wholesale prices. Hence, it appears that infrequent price adjustment is primarily driven by the behavior of wholesale prices. Second, when we use our model to derive upper and lower bounds for the fixed costs of price adjustment facing retailers and manufacturers, we find that these costs are roughly of the same order of magnitude for manufacturers and retailers in absolute terms, though smaller for manufacturers as a share of their total revenue. Third, the counterfactual simulations we conduct in order to decompose the incomplete transmission of exchange rate shocks into its sources suggest that both local non-traded costs and markup adjustment are important in generating local currency price stability. However, these two factors alone cannot completely explain the incomplete pass-through of exchange rate changes to consumer prices. To generate the pass-through observed in the data we need to allow for price adjustment costs. To be more specific, local costs and markup adjustment put together generate an exchange rate pass-through of approximately 20%. The pass-through observed in our data is between 6 and 7%. The missing 11-12% represent the overall impact of costs of repricing, both firms’ own and the strategic effects of their competitors’ repricing costs. These rigidities affect primarily the adjustment of wholesale prices; their effect on retail prices is very minor, and operates in large part through the strategic interaction of competing brands. Why nominal price rigidities operate primarily at the wholesale but not retail level is - in our opinion - an intriguing question worth further exploration. One possible explanation is that wholesale prices are set through long-term contracts and are therefore less responsive to changes in economic conditions. We hope that future research can shed more light into this issue.
References


Figure 1: Weekly retail and wholesale prices for Bass Ale. Prices are for a single six-pack and are from Zone 1. 202 observations. Source: Dominick’s.
Figure 2: Weekly retail and wholesale prices for Corona. Prices are for a single six-pack and are from Zone 1. 202 observations. Source: Dominick’s.

Figure 3: The nominal exchange rate varies by more than do typical input prices for German brewers.
<table>
<thead>
<tr>
<th>Description</th>
<th>Mean</th>
<th>Median</th>
<th>Standard Deviation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Retail prices ($ per six-pack)</td>
<td>5.44</td>
<td>5.79</td>
<td>1.28</td>
</tr>
<tr>
<td>Wholesale prices ($ per six-pack)</td>
<td>4.50</td>
<td>4.92</td>
<td>1.09</td>
</tr>
<tr>
<td>Dummy for retail-price change (=1 if yes)</td>
<td>.18</td>
<td>0</td>
<td>.24</td>
</tr>
<tr>
<td>Dummy for wholesale-price change (=1 if yes)</td>
<td>.06</td>
<td>0</td>
<td>.16</td>
</tr>
</tbody>
</table>

Table 1: Summary statistics for prices for the 16 products in the sample. 6464 observations. Source: *Dominick’s*.

<table>
<thead>
<tr>
<th>Description</th>
<th>Mean</th>
<th>Median</th>
<th>Std</th>
<th>Minimum</th>
<th>Maximum</th>
</tr>
</thead>
<tbody>
<tr>
<td>Percent Alcohol</td>
<td>4.52</td>
<td>4.60</td>
<td>.68</td>
<td>2.41</td>
<td>6.04</td>
</tr>
<tr>
<td>Bitterness</td>
<td>2.50</td>
<td>2.10</td>
<td>1.08</td>
<td>1.70</td>
<td>5.80</td>
</tr>
</tbody>
</table>

### Table 3: Some preliminary descriptive results.

The dependent variable is the retail or the wholesale price for a six-pack of each brand of beer. The exchange-rate is the average of the previous week’s bilateral spot rate between the foreign manufacturer’s country and the United States (is the number of dollars per unit of foreign currency). All regressions include brand, price-zone, and week fixed effects. The second and fourth columns of the table report results from regressions with controls for domestic and foreign costs. The coefficients on the control variables, and their sources, are reported in the appendix. Robust standard errors in parentheses: Those starred are significant at the * 5-percent or ** 1-percent level. Source: Authors’ calculations.

<table>
<thead>
<tr>
<th>Variable</th>
<th>OLS</th>
<th>OLS</th>
<th>IV</th>
<th>IV</th>
</tr>
</thead>
<tbody>
<tr>
<td>Exchange rate</td>
<td>5.96</td>
<td>6.72</td>
<td>4.27</td>
<td>4.74</td>
</tr>
<tr>
<td></td>
<td>(1.50)**</td>
<td>(1.56)**</td>
<td>(1.50)**</td>
<td>(1.52)**</td>
</tr>
<tr>
<td>Wholesale price</td>
<td>1.83</td>
<td>1.79</td>
<td>1.66</td>
<td>1.67</td>
</tr>
<tr>
<td></td>
<td>(.02)**</td>
<td>(.16)**</td>
<td>(.08)**</td>
<td>(.10)**</td>
</tr>
<tr>
<td>Observations</td>
<td>3636</td>
<td>3636</td>
<td>3636</td>
<td>3636</td>
</tr>
<tr>
<td>(R^2)</td>
<td>.65</td>
<td>.65</td>
<td>.81</td>
<td>.81</td>
</tr>
</tbody>
</table>

### Table 4: Diagnostic results from the logit model of demand.

Dependent variable is \(\ln(S_{jt}) - \ln(S_{ot})\). Both regressions include brand fixed effects. Huber-White robust standard errors are reported in parentheses. Costs are domestic wages in the beverage industry interacted with weekly nominal exchange rates for foreign brands. Source: Authors’ calculations.

<table>
<thead>
<tr>
<th>Variable</th>
<th>OLS</th>
<th>OLS</th>
<th>IV</th>
<th>IV</th>
</tr>
</thead>
<tbody>
<tr>
<td>Price</td>
<td>-.93</td>
<td>-.92</td>
<td>-2.43</td>
<td>-2.43</td>
</tr>
<tr>
<td></td>
<td>(.01)**</td>
<td>(.01)**</td>
<td>(.35)**</td>
<td>(.35)**</td>
</tr>
<tr>
<td>Holiday</td>
<td>.06</td>
<td>.01</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(.02)**</td>
<td>(.01)</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>First-Stage Results</th>
<th>F-Statistic</th>
<th>34.45</th>
<th>34.24</th>
</tr>
</thead>
<tbody>
<tr>
<td>Observations</td>
<td>6464</td>
<td>6464</td>
<td>6464</td>
</tr>
<tr>
<td>Instruments</td>
<td>wages</td>
<td>wages</td>
<td></td>
</tr>
</tbody>
</table>
Variable | Mean in Population | Interaction with Income
---|---|---
Constant | -16.39* | (.20)
Price | -2.48* | 40.37* (18.17)
Bitterness | -5.57* | (.20)
Percent Alcohol | 7.35* | -51.84* (20.32)

Minimum-Distance $R^2$ | 0.18

Table 5: Results from the full random-coefficients model of demand. Based on 6464 observations. Asymptotically robust standard errors in parentheses. Starred coefficients are significant at the 5-percent level. Source: Authors’ calculations.

<table>
<thead>
<tr>
<th>Brand</th>
<th>Retail price</th>
<th>Manufacturer price</th>
<th>Retail markup</th>
<th>Manufacturer markup</th>
</tr>
</thead>
<tbody>
<tr>
<td>Bass</td>
<td>$6.29</td>
<td>$5.71</td>
<td>$0.41</td>
<td>$0.40</td>
</tr>
<tr>
<td>Beck’s</td>
<td>$5.22</td>
<td>$4.44</td>
<td>$0.40</td>
<td>$0.40</td>
</tr>
<tr>
<td>Corona</td>
<td>$5.04</td>
<td>$4.31</td>
<td>$0.40</td>
<td>$0.39</td>
</tr>
<tr>
<td>Guinness</td>
<td>$6.63</td>
<td>$5.74</td>
<td>$0.44</td>
<td>$0.42</td>
</tr>
<tr>
<td>Heineken</td>
<td>$5.55</td>
<td>$4.83</td>
<td>$0.40</td>
<td>$0.39</td>
</tr>
<tr>
<td>All Imports</td>
<td>$5.45</td>
<td>$4.77</td>
<td>$0.40</td>
<td>$0.40</td>
</tr>
</tbody>
</table>

Table 6: Mean prices and price-cost markups for selected brands. The markup is price less marginal cost with the marginal costs derived from the structural model and with units in dollars per six-pack. Source: Authors’ calculations.
<table>
<thead>
<tr>
<th>Brand</th>
<th>Backed-out</th>
<th>Fitted</th>
</tr>
</thead>
<tbody>
<tr>
<td>Amstel</td>
<td>.29</td>
<td>.27</td>
</tr>
<tr>
<td></td>
<td>(.078)**</td>
<td>(.078)**</td>
</tr>
<tr>
<td>Bass</td>
<td>.21</td>
<td>.18</td>
</tr>
<tr>
<td></td>
<td>(.078)**</td>
<td>(.076)**</td>
</tr>
<tr>
<td>Beck's</td>
<td>.37</td>
<td>.33</td>
</tr>
<tr>
<td></td>
<td>(.055)**</td>
<td>(.054)**</td>
</tr>
<tr>
<td>Corona</td>
<td>.37</td>
<td>.34</td>
</tr>
<tr>
<td></td>
<td>(.064)**</td>
<td>(.066)**</td>
</tr>
<tr>
<td>Guinness</td>
<td>.51</td>
<td>.50</td>
</tr>
<tr>
<td></td>
<td>(.162)**</td>
<td>(.162)**</td>
</tr>
<tr>
<td>Harp</td>
<td>.40</td>
<td>.41</td>
</tr>
<tr>
<td></td>
<td>(.089)**</td>
<td>(.091)**</td>
</tr>
<tr>
<td>Heineken</td>
<td>.43</td>
<td>.41</td>
</tr>
<tr>
<td></td>
<td>(.086)**</td>
<td>(.087)**</td>
</tr>
<tr>
<td>Molson Golden</td>
<td>.30</td>
<td>.33</td>
</tr>
<tr>
<td></td>
<td>(.056)**</td>
<td>(.060)**</td>
</tr>
<tr>
<td>St. Pauli Girl</td>
<td>.61</td>
<td>.56</td>
</tr>
<tr>
<td></td>
<td>(.092)**</td>
<td>(.082)**</td>
</tr>
<tr>
<td>All Imports</td>
<td>.38</td>
<td>.37</td>
</tr>
<tr>
<td></td>
<td>(.060)**</td>
<td>(.063)**</td>
</tr>
</tbody>
</table>

Table 7: Backed-out and fitted non-traded costs incurred by the retailer by brand. Each entry reports the mean across weeks of the backed-out or fitted measure of a brand's non-traded cost to the retailer in cents per six-pack. Standard errors from bootstrap simulations with 200 draws reported in parentheses under each coefficient. Those starred are significant at the *5- or **1-percent level.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chicago-Area Grocery Wages</td>
<td>.50</td>
</tr>
<tr>
<td></td>
<td>(.06)**</td>
</tr>
<tr>
<td>Constant</td>
<td>-4.34</td>
</tr>
<tr>
<td></td>
<td>(.51)**</td>
</tr>
</tbody>
</table>

Table 8: Results from regressions of backed-out retailer non-traded costs on determinants. Dependent variable is retailer’s non-traded cost which varies by week. Huber-White robust standard errors are reported in parentheses. Those starred are significant at the *5- or **1-percent level. Source: Authors’ calculations.
<table>
<thead>
<tr>
<th>Brand</th>
<th>Backed-out</th>
<th>Fitted</th>
</tr>
</thead>
<tbody>
<tr>
<td>Amstel</td>
<td>4.53</td>
<td>4.56</td>
</tr>
<tr>
<td></td>
<td>(.06)**</td>
<td>(.06)**</td>
</tr>
<tr>
<td>Bass</td>
<td>5.36</td>
<td>5.28</td>
</tr>
<tr>
<td></td>
<td>(.06)**</td>
<td>(.05)**</td>
</tr>
<tr>
<td>Beck's</td>
<td>4.02</td>
<td>4.02</td>
</tr>
<tr>
<td></td>
<td>(.04)**</td>
<td>(.03)**</td>
</tr>
<tr>
<td>Corona</td>
<td>3.46</td>
<td>3.43</td>
</tr>
<tr>
<td></td>
<td>(.06)**</td>
<td>(.06)**</td>
</tr>
<tr>
<td>Guinness</td>
<td>5.74</td>
<td>5.49</td>
</tr>
<tr>
<td></td>
<td>(.07)**</td>
<td>(.07)**</td>
</tr>
<tr>
<td>Harp</td>
<td>4.98</td>
<td>5.15</td>
</tr>
<tr>
<td></td>
<td>(.06)**</td>
<td>(.06)**</td>
</tr>
<tr>
<td>Heineken</td>
<td>4.60</td>
<td>4.51</td>
</tr>
<tr>
<td></td>
<td>(.06)**</td>
<td>(.06)**</td>
</tr>
<tr>
<td>Molson Golden</td>
<td>3.25</td>
<td>3.27</td>
</tr>
<tr>
<td></td>
<td>(.04)**</td>
<td>(.04)**</td>
</tr>
<tr>
<td>All Imports</td>
<td>4.18</td>
<td>4.41</td>
</tr>
<tr>
<td></td>
<td>(.00)**</td>
<td>(.10)**</td>
</tr>
</tbody>
</table>

Table 9: Backed-out and fitted total costs incurred by foreign manufacturer by brand. Each entry reports the mean across weeks of the backed-out or fitted measure of a brand’s total cost to the manufacturer in cents per six-pack. Standard errors from bootstrap simulations with 200 draws reported in parentheses under each coefficient. Those starred are significant at the *5- or **1-percent level.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
</tr>
</thead>
<tbody>
<tr>
<td>Domestic U.S. wages</td>
<td>.53</td>
</tr>
<tr>
<td></td>
<td>(.09)**</td>
</tr>
<tr>
<td>Price foreign barley</td>
<td>.17</td>
</tr>
<tr>
<td></td>
<td>(.04)**</td>
</tr>
<tr>
<td>Foreign wages</td>
<td>.30</td>
</tr>
<tr>
<td></td>
<td>(.05)**</td>
</tr>
<tr>
<td>Observations</td>
<td>316</td>
</tr>
</tbody>
</table>

Table 10: Results from constrained linear regression of foreign manufacturer total backed-out costs on determinants. Dependent variable is manufacturer’s total marginal cost for periods when the wholesale price changes which varies by week. Includes brand fixed effects. Starred coefficients are significant at the *5- or **1-percent level. Source: Authors’ calculations.
<table>
<thead>
<tr>
<th>Brand</th>
<th>Mean Cost Share of Brand’s Revenue</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Upper Bounds</td>
<td>Lower Bounds</td>
</tr>
<tr>
<td>Amstel</td>
<td>$107.18</td>
<td>$34.94</td>
</tr>
<tr>
<td>Bass</td>
<td>$101.92</td>
<td>$38.18</td>
</tr>
<tr>
<td>Beck’s</td>
<td>$531.26</td>
<td>$49.96</td>
</tr>
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<td>Corona</td>
<td>$139.24</td>
<td>$72.62</td>
</tr>
<tr>
<td>Guinness</td>
<td>$269.34</td>
<td>$35.06</td>
</tr>
<tr>
<td>Harp</td>
<td>$19.18</td>
<td>$13.20</td>
</tr>
<tr>
<td>Heineken</td>
<td>$652.58</td>
<td>$112.54</td>
</tr>
<tr>
<td>Molson Golden</td>
<td>$101.36</td>
<td>$28.14</td>
</tr>
<tr>
<td>St. Pauli Girl</td>
<td>$38.48</td>
<td>$35.20</td>
</tr>
<tr>
<td>All</td>
<td>$254.98</td>
<td>$46.14</td>
</tr>
</tbody>
</table>

Table 11: Bounds for the retailer’s adjustment costs by brand. The entries in the first two columns report the mean over time of the dollar value of adjustment costs, and in the third and fourth columns the mean over revenue for that brand over all markets whether the price changed or not.
<table>
<thead>
<tr>
<th>Brand</th>
<th>Mean Cost Share of Brand's Revenue</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Upper Bounds</td>
</tr>
<tr>
<td>Amstel</td>
<td>$76.14</td>
</tr>
<tr>
<td>Bass</td>
<td>$16.68</td>
</tr>
<tr>
<td>Beck's</td>
<td>$499.14</td>
</tr>
<tr>
<td>Corona</td>
<td>$376.86</td>
</tr>
<tr>
<td>Guinness</td>
<td>$310.00</td>
</tr>
<tr>
<td>Harp</td>
<td>$27.40</td>
</tr>
<tr>
<td>Heineken</td>
<td>$210.78</td>
</tr>
<tr>
<td>Molson Golden</td>
<td>$278.56</td>
</tr>
<tr>
<td>All</td>
<td>$281.72</td>
</tr>
</tbody>
</table>

Table 12: *Bounds for the manufacturers’ adjustment costs by brand.* The entries in the first two columns report the mean over time of the dollar value of adjustment costs, and in the third and fourth columns the mean over revenue for that brand over all markets whether the price changed or not.
<table>
<thead>
<tr>
<th>Variable</th>
<th>Coefficient</th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Dummy for level change in retail price</td>
<td>907.00</td>
<td>(34.03)***</td>
</tr>
<tr>
<td>Dummy for all retail price changes</td>
<td>55.44</td>
<td>(10.21)***</td>
</tr>
<tr>
<td>Overall $R^2$</td>
<td>.19</td>
<td></td>
</tr>
<tr>
<td>Observations</td>
<td>3636</td>
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</tbody>
</table>

Table 13: Regression of retailer’s fixed adjustment costs on a dummy for a level retail-price change. The regression includes brand and zone fixed effects. Source: Authors’ calculations.
<table>
<thead>
<tr>
<th></th>
<th>Manufacturer:</th>
<th>Retailer:</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Traded</td>
<td>Markup Adjustment</td>
</tr>
<tr>
<td>No repricing costs</td>
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<td></td>
</tr>
<tr>
<td>Bass</td>
<td>47.0</td>
<td>34.4</td>
</tr>
<tr>
<td>Beck’s</td>
<td>47.0</td>
<td>44.2</td>
</tr>
<tr>
<td>Corona</td>
<td>47.0</td>
<td>36.0</td>
</tr>
<tr>
<td>Heineken</td>
<td>47.0</td>
<td>44.5</td>
</tr>
<tr>
<td>All</td>
<td>47.0</td>
<td>39.8</td>
</tr>
<tr>
<td>Competitor-brand repricing costs</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bass</td>
<td>47.0</td>
<td>0</td>
</tr>
<tr>
<td>Becks</td>
<td>47.0</td>
<td>0</td>
</tr>
<tr>
<td>Corona</td>
<td>47.0</td>
<td>35.7</td>
</tr>
<tr>
<td>Heineken</td>
<td>47.0</td>
<td>35.7</td>
</tr>
<tr>
<td>All</td>
<td>47.0</td>
<td>17.9</td>
</tr>
<tr>
<td>Own-brand repricing costs</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Bass</td>
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<td>0</td>
</tr>
<tr>
<td>Becks</td>
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<td>0</td>
</tr>
<tr>
<td>Corona</td>
<td>47.0</td>
<td>0</td>
</tr>
<tr>
<td>Heineken</td>
<td>47.0</td>
<td>0</td>
</tr>
<tr>
<td>All</td>
<td>47.0</td>
<td>0</td>
</tr>
</tbody>
</table>

Table 14: Counterfactual experiments: median pass-through of a 5-percent depreciation of the relevant foreign currency. Median over 404 markets. Retailer’s incomplete pass-through: the retail price’s percent change for the given percent change in the exchange rate, attributed to the presence of local dollar-denominated costs or to the retailer’s markup adjustment. Manufacturer’s incomplete pass-through: the manufacturer price’s percent change for a given percent change in the exchange rate, attributed to the share of local dollar-denominated costs in the manufacturer’s total costs or to the manufacturer’s markup adjustment. Source: Authors’ calculations.
Table 15: *Counterfactual experiments: Decomposition of the incomplete transmission of a 1-percent appreciation of the relevant foreign currency to consumer prices.* Median over 404 markets. Local costs: the share of the incomplete transmission explained by the presence of a local dollar-denominated component in foreign manufacturers’ or the retailer’s marginal costs. Markup adjustment: the share of the incomplete transmission explained by the retailer or manufacturer’s markup adjustment excluding markup adjustment due to fixed costs of price adjustment. Repricing costs own: Fixed costs of price adjustment incurred by the manufacturer or retailer to change its own price. Repricing costs, other: The effect of competitors’ costs of price adjustment on the manufacturer or retailer’s own price adjustment behavior. Source: Authors’ calculations.

<table>
<thead>
<tr>
<th></th>
<th>Manufacturer</th>
<th>Retailer</th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
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</thead>
<tbody>
<tr>
<td></td>
<td>Local Costs</td>
<td>Markup</td>
<td>Costs of</td>
<td>Markup</td>
<td>Costs of</td>
<td>Other</td>
<td>Own</td>
<td>Local</td>
<td>Markup</td>
<td>Costs of</td>
<td>Other</td>
<td>Own</td>
</tr>
<tr>
<td></td>
<td>Adj.</td>
<td>Repricing:</td>
<td></td>
<td>Adj.</td>
<td>Repricing:</td>
<td></td>
<td></td>
<td>Costs</td>
<td>Adj.</td>
<td>Repricing:</td>
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</tr>
<tr>
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<td>2.8</td>
<td>44.2</td>
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<td>35.7</td>
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<tr>
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