In Search of Real Rigidities

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In Search of Real Rigidities*

Gita Gopinath
Department of Economics,
Harvard University and NBER

Oleg Itskhoki
Department of Economics,
Princeton University

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Abstract

The closed and open economy literatures work on estimating real rigidities, but in parallel. We bring the two literatures together to shed light on this question. We use international price data and exchange rate shocks to evaluate the importance of real rigidities in price setting. We show that consistent with the presence of real rigidities the response of reset-price inflation to exchange rate shocks depicts significant persistence. Individual import prices, conditional on changing, respond to exchange rate shocks prior to the last price change. At the same time aggregate reset-price inflation for imports, like that for consumer prices, depicts little persistence. Competitors prices effect firm pricing and exchange rate pass-through into import prices are greater in response to trade-weighted as opposed to bilateral exchange rate changes. We quantitatively evaluate sticky price models (Calvo and menu cost) with variable markups at the wholesale level and constant markups at the retail level, consistent with empirical evidence. Variable markups alone generate sluggishness in price adjustment and increase the size of the contract multiplier, but their effects are modest.

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1 Introduction

A large literature has recently emerged that documents patterns of nominal price stickiness at the very micro level — the good level. The documented nominal durations are significantly shorter than the estimated real effects of money on output (e.g., see Christiano, Eichenbaum, and Evans, 1999). The long-lasting real effects of monetary shocks can be reconciled with moderate price stickiness if real rigidities are an important phenomena.\footnote{The literature following Taylor (1980) calls it the \textit{contract multiplier}, as this mechanism is based on the interaction between staggered fixed-price contracts and real rigidities. The term \textit{real rigidities} misleadingly points towards some imperfections. In fact, real rigidities may well be present in conventional neoclassical models (e.g., Atkeson and Burstein, 2008). The term originated from the fact that in a sticky price environment real rigidities work to amplify the effects of nominal rigidities which in turn constitute a departure from a neoclassical model.} These real rigidities dampen price responses even when firms adjust prices because of factors such as strategic complementarities in price setting, real wage rigidity, the dependence of costs on input prices that have yet to adjust, among others. It is not surprising then that an important empirical literature has emerged recently that evaluates the question: Are quantitatively important real rigidities present in the data? The answer appears to depend on what data one looks at.

In international economics there is a large and growing literature that estimates exchange rate pass-through from exchange rate shocks into prices. The estimated exchange rate pass-through is found to be incomplete, that is, if the dollar depreciates by 10% relative to the Euro, dollar prices of goods imported from the Euro area increase by less than ten percent even in the long-run. This incompleteness in pass-through is argued to be consistent with the presence of important real rigidities. Exchange rate changes generate relative price movements for the same good across markets despite costs being the same. This destination-specific markup is argued to be consistent with the presence of significant strategic complementarities in price setting.

The closed economy literature, on the other hand, uses indirect tests of real rigidities in the absence of well-identified and sizeable shocks like exchange rate shocks. The recent work based on micro evidence for retail prices argues that real rigidities are not an empirically important phenomena.

There are many developments in the measurement of real rigidities in the closed and open economy literatures, but these developments have taken place in parallel. In this paper we bring together the closed economy macro literature that focuses mainly on indirect tests of
real rigidities with the international pricing literature that uses an observable and sizeable shock, namely the exchange rate shock, to evaluate the behavior of prices, and in particular of strategic complementarities in pricing. We first review the recent evidence on real rigidities to evaluate if there is a consensus emerging on the importance of these rigidities in the data. Second, since the two literatures use different metrics to evaluate the importance of real rigidities we use unpublished international price data collected by the Bureau of Labor Statistics (BLS) to estimate both metrics using the same data. Third, we present new evidence on the dynamic response of international prices to exchange rate shocks and the response to competitor prices. Fourth, we calibrate sticky price macro models (Calvo and menu cost) with a retail and wholesale sector to the evidence on the variable markup channel of real rigidites. We evaluate its ability to match the behavior of prices in the data and to measure the extent of monetary non-neutrality this channel generates.

In reviewing the literature we group evidence based on whether the prices refer to retail (consumer) prices or wholesale prices. The wholesale prices can be alternatively viewed as the intermediate-good prices in the business-to-business transactions. Similarly the literature on exchange rate pass-through into at-the-dock prices of goods refers to wholesale prices. A review of the existing literature reveals one surprisingly consistent result across several studies, surprising since these studies use different methodologies and data sets. This result is that strategic complementarities, for example operating through variable markups, play little role for retail prices and appear to be quite important for wholesale prices.

We next use the BLS import price data to perform tests of real rigidity, where we use measures employed in the closed economy literature, namely the persistence of reset-price inflation (Bils, Klenow, and Malin, 2009, henceforth BKM), and measures employed in the open economy literature, namely the dynamic response of prices to exchange rate shocks. The actual import-price inflation series has a monthly persistence of 0.56, while the corresponding reset-price inflation series has a persistence of $-0.04$. In comparison, BKM estimate for retail prices that the inflation series has persistence of $-0.05$, while the reset-price inflation series has a persistence of $-0.41$. In comparison to retail prices import prices have greater persistence, but the magnitude of this persistence suggests very little sluggishness in price adjustment.

However, when we project the aggregate import reset-price inflation on lags of the trade-weighted nominal exchange rate changes, we find that the autocorrelation of the fitted series
is substantially higher as compared to that of unconditional reset-price inflation (0.33 versus −0.04). We find similar evidence using micro level price adjustments. We show that individual import prices, conditional on changing, respond to exchange rate shocks prior to the last time the price was adjusted and these lagged effects are large and statistically significant. The pass-through conditional on a price change to the cumulative exchange rate change since the last price adjustment is 0.11 and the response to the cumulative exchange rate over the previous price duration is 0.08. Both these pieces of evidence that evaluate the response to a specific shock suggests a more important role for real rigidities as compared to the point estimate of the autocorrelation of reset prices.

Next, we evaluate the importance of strategic complementarities in price setting for incomplete pass-through using some measures that capture the pricing behavior of competitors and measures that capture the extent of competition in sectors. These measures are not perfect but provide useful information about pricing behavior. We use the prices set by other firms in the same 10-digit or 4-digit harmonized code in the import price sample to control for the behavior of competitor prices and we find that they have an important positive effect on firms pricing, reducing the direct pass-through of exchange rate into prices. The point estimates suggest a markup elasticity of 1.5 which implies a 40% pass-through for purely idiosyncratic shocks. We also evaluate the sensitivity of firm pricing to shocks to competitors by measuring the response of prices to movements in the U.S. trade-weighted exchange rate that is orthogonal to the bilateral exchange rate for the country. We find the response to be sizeable and significant. In a similar vein we find that exchange rate pass-through is higher in response to a more aggregate shock as compared to more idiosyncratic shocks, when comparing the response to bilateral exchange rate shocks versus trade-weighted exchange rate shocks.

We also relate the incompleteness in pass-through to certain sectoral features that proxy for the level of competition among importers. An important distinction between retail prices and wholesale prices is that the latter captures business-to-business transactions. Consequently, the strength of bargaining power of the buyer can impact the extent of pass-through. We use unpublished measures of concentration in the import sector provided to us by the BLS—specifically, the Herfindahl index and the number of importers that make up the top 50% of trade—to evaluate this hypothesis. While the point estimates in many cases suggests that sectors dominated by a few large importers have lower pass-through from foreign firms, the estimated standard errors are large.
Lastly, we use estimates from the data to calibrate a closed economy model with different degrees of variable markup elasticity at the wholesale and retail level. In the existing monetary literature there is typically no interesting distinction made between the retail and wholesale sectors. We calibrate the parameters for the wholesale sector using the evidence from international prices. In the benchmark model we use Calvo price setting and later evaluate the case of menu cost pricing. First, we show that sluggishness in response to monetary shocks in wholesale prices feeds into slow adjustment of retail prices. However, the aggregate inflation and reset-price inflation series exhibit little persistence since their movement are dominated by more transitory shocks. However, conditional on monetary shocks or exchange-like shocks, inflation series exhibit considerable persistence. Similarly, output series can exhibit significant monetary non-neutralities. Second, while calibrated real rigidities in the form of variable markups increase the size of the contract multiplier, their effects are limited unless they are coupled with exogenous sources of persistence. The model however fails to match the slow dynamics in price adjustment that was documented in the empirical data, which suggest that additional sources of persistence are missing from the model.

An important question is why one observes differences in markup variability at the wholesale and retail level. We do not provide answers here but conjecture that this can be consistent with differences in the competitive environment at the two levels. That is the retail sector can be described as monopolistically competitive while the wholesale sector is better described as a bilateral bargaining environment. We present a static bargaining model of wholesale price setting that results in variable markups and incomplete pass-through of shocks into wholesale prices. Specifically, each final good producer bargains with its intermediate good suppliers regarding the price of the intermediate goods. Given these bargained prices, the final good producer is free to choose quantities of the intermediate inputs, as well as to set the price of its final good in the monopolistically competitive consumer market.

In this model the final goods producer and the intermediate good supplier bargain only on prices while quantities are determined by the demand curve faced by the retailer. Prices are therefore allocative. An important question is whether in the data it is the case that whole-sale prices are allocative and also whether contracts specify fixed prices at fixed quantities. While there is no simple way to test this Gopinath and Rigobon (2008) show that in the case of contracts for international prices they typically involve a fixed price with a quantity range specified, as opposed to a fixed quantity. Moreover, firms export the same good at the same price to multiple destinations and consequently the prices behave in many cases like a list price. Further the behavior of prices is consistent with models of monopolistic price setting, where prices are allocative, as discussed in the papers of Gopinath, Itskhoki, and Rigobon (2010) and Gopinath and Itskhoki (2010). Also, changes in intermediate-good prices effect final goods prices as they are fully passed-through into retail consumer prices. These separate pieces of evidence are consistent with whole-sale prices being allocative.
This model results in constant markups at the retail stage, but in variable markups at the wholesale level that depend among other things on the relative bargaining power of the final good producer and on the market share of the intermediate good supplier.

The paper is structured as follows. Section 2 provides a descriptive framework that spells out the sources of real rigidities that can result in sluggish price adjustment. Section 3 reviews the closed and open-economy literature on real rigidities that uses micro price data. Section 4 presents new empirical results on price adjustment using international data. Section 5 presents the closed economy model with differential markup variability in the retail and wholesale sector and sluggish price adjustment. Finally, Section 6 lays out a model of bargaining and variable markups in intermediate-good pricing. All derivations and technical details are relegated to the Appendix.

2 Real Rigidities: A Descriptive Framework

In this section we set-up some notation and spell out the sources of real rigidities that can result in sluggish price adjustment. This simple descriptive framework is used to organize the discussion of empirical evidence in the following two sections.

Define the desired price of a firm as the price it would set if it could adjust its prices flexibly in a given economic environment. With sticky prices, a forward-looking firm sets its price as a weighted average of future desired prices. The presence of real rigidities slows down the response of the desired price to a shock. Real rigidities, powered by nominal price stickiness and staggered price adjustment (Taylor, 1980; Calvo, 1983), can be either at the aggregate/industry level or at the micro/firm level. Real rigidities at the aggregate level include round-about production structure as in Basu (1995), real wage rigidity as in Blanchard and Galí (2007) and segmented input markets as in Woodford (2003). Real rigidities at the firm level, or strategic complementarities in pricing, arise either from non-constant marginal cost (i.e., decreasing returns to scale) as in Burstein and Hellwig (2007) or from variable markups. In turn, variable markups can be due to non-CES demand as in Kimball (1995) and Klenow and Willis (2006) or due to strategic complementarities in price setting between large firms as in Atkeson and Burstein (2008).

Empirical studies attempt to determine which of these channels of real rigidities if any are present in the data, as well as the plausible magnitudes of their effects. Since the empirical
literature has examined evidence at both retail and wholesale levels, we will maintain this
distinction in our descriptive framework.

**Retail (final-good) pricing:** The log desired price for a final good $i$ at time $t$ can be
written as a log desired markup over the marginal cost of the firm:

$$\tilde{p}_{it} = \mu^{R}_{it} + \alpha s_t + (1 - \alpha) w_t - z_{it},$$

where $\mu^{R}_{it}$ is the log desired markup, $s_t$ is the log price of the intermediate input, $w_t$ is the
log price of other inputs (e.g., labor), and $z_{it}$ represents the firm-specific marginal cost or
markup shock that may include an aggregate component common across a subset of firms.
Let $\alpha \in [0, 1]$ be the share of intermediate inputs in the production cost of the final-good
firms.

Throughout the paper we abstract from the non-constant marginal cost channel of strate-
gic complementarities in favor of the variable markup channel. For most purposes the two
mechanisms are largely substitutable, however variable markups can additionally explain
pricing to market, a phenomenon with strong empirical support as discussed in the next
section. We capture variable markups in the following reduced-form way:

$$\mu^{R}_{it} \approx \bar{\mu}^{R} - \Gamma^{R}_R (p_{it} - p_t),$$

where $p_t$ is the industry price index and $\Gamma^{R}_R \geq 0$ is the elasticity of the firm’s markup with
respect to its relative price. That is a higher relative price of the firm reduces its desired
markup. As we discuss in more detail in Section 5, this simple description of the desired
markup is consistent with various models of variable markups cited above.

Combining the above two equations, we arrive at

$$\tilde{p}_{it} = \frac{1}{1 + \Gamma^{R}_R} \left[ \bar{\mu}^{R} + \alpha s_t + (1 - \alpha) w_t - z_{it} \right] + \frac{\Gamma^{R}_R}{1 + \Gamma^{R}_R} p_t.$$

When $\Gamma^{R}_R > 0$, the desired price of the firm increases in the prices set by the competitors of
the firm, which we refer to as strategic complementarities in price setting. Under the same
circumstances a response to an idiosyncratic marginal cost/markup shock (such as $z_{it}$) is
incomplete: the pass-through of an idiosyncratic shock into the desired price equals $1/(1 +$
\( \Gamma \in [0,1] \).

To summarize, sluggishness in the response of desired final-good prices may arise either due to staggered price adjustment when \( \Gamma > 0 \), or due to sluggish adjustment in the marginal cost, i.e. due to a slow response of \( s_t \) and/or \( w_t \) to shocks.

**Wholesale (intermediate-good) pricing:** Let the log desired price of an intermediate variety \( j \) be

\[
\tilde{s}_{jt} = \mu_{jt} + w_t + \phi_j e_t - a_{jt},
\]

where \( \mu_{jt} \) is the log desired markup, \( w_t \) is the price of inputs (e.g., labor), and \( a_{jt} \) is a firm-specific marginal cost or markup shock that again may contain an aggregate component. Furthermore, we allow the cost of intermediate firms to respond directly to exchange rate fluctuations \( e_t \) with various elasticities \( \phi_j \). For example, some of the intermediate varieties may be produced abroad so that their marginal cost in local currency fluctuates together with the exchange rate.\(^3\) Finally, the log price of the intermediate good, \( s_t \), is an aggregate of the prices of intermediate varieties, \( s_{jt} \), which can be approximated by a geometric average:

\[
s_t \approx \int_j s_{jt} \text{d}j.
\]

We similarly introduce the possibility of variable markups for the intermediate varieties:

\[
\mu_{jt} \approx \bar{\mu} - \Gamma (s_{jt} - s_t),
\]

where \( \Gamma \) is again the price elasticity of markup that may vary across varieties \( j \). We can rewrite the desired price of the intermediate variety as

\[
\tilde{s}_{jt} = \frac{1}{1 + \Gamma} \left[ \bar{\mu} + w_t + \phi_j e_t - a_{jt} \right] + \frac{\Gamma}{1 + \Gamma} s_t. \tag{2}
\]

Purely idiosyncratic shocks wash out in the aggregate and do not affect \( s_t \). Therefore, they get passed-through into desired prices immediately with a pass-through coefficient of \( 1/(1 + \Gamma) \). More aggregate shocks have two channels through which they affect the desired price — directly (\( w_t \) or \( e_t \)) and via the prices of competitors (\( s_t \)). Coupled with nominal price stickiness and staggered price adjustment, variable markups (\( \Gamma > 0 \)) may generate sluggish adjustment that lasts past the periods of nominal stickiness.

\(^3\)Of course, the exchange rate can have additional indirect (general equilibrium) effects via productivity or cost of inputs. We discuss this issue in more detail in Sections 3–4.
**Aggregate real rigidities:** All the above-mentioned sources of aggregate real rigidities enter through a sluggish adjustment to shocks of the marginal cost component denoted by \( w_t \). Specifically, real wage rigidities, Basu (1995) round-about production and segmented input markets can all be captured by a slow adjustment of \( w_t \) to shocks. For concreteness, one can use the following structure to think about aggregate real rigidities:\(^4\)

\[
w_t = \gamma m_t + (1 - \gamma) p_t,
\]

where \( m_t \) is aggregate nominal spending, \( p_t \) is the consumer price level and \( \gamma > 0 \) is the elasticity of the cost \( w_t \) with respect to monetary (nominal spending) shocks. The smaller is \( \gamma \), the more sluggish is the response of aggregate costs to monetary shocks, that is the stronger are aggregate real rigidities.

Equations (1)-(3) describe our simple framework to account for real rigidities.\(^5\) Sluggish adjustment in the desired final-good prices \( \tilde{p}_{it} \) can arise due to one of the three channels: (a) sluggish response of aggregate costs \( w_t \) (small \( \gamma \)); (b) sluggish adjustment of intermediate prices \( s_t \) (\( \Gamma > 0 \) provided \( \alpha > 0 \)); and (c) gradual adjustment in final-good desired prices (\( \Gamma_R > 0 \)). We now evaluate the evidence on these different sources of real rigidities.

### 3 Evidence on Real Rigidities: A Review

Although appealing at the intuitive level, real rigidities are hard to identify and measure in the data. Aggregate real rigidities imply a sluggish response of the marginal cost of firms to aggregate shocks. Firm-level real rigidities imply a muted response of the firms’ prices conditional on price adjustment to the marginal cost shocks. Since data on marginal costs is usually unavailable, it is hard to test these mechanisms directly. Therefore, the literature relies either on calibrations or indirect empirical tests.

In parallel, the international literature has a long tradition of estimating exchange rate pass-through, namely the response of international prices to changes in exchange rates.\(^6\) A

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\(^4\)See Section 5 for details.

\(^5\)In Section 5 we combine this framework with specific models of nominal stickiness to study the quantitative predictions of dynamic price-setting models.

\(^6\)Goldberg and Knetter (1997) provide a survey of the earlier empirical pass-through literature. The specification of pass-through regressions can either be in real or in nominal terms and accordingly include
standard assumption in the empirical pass-through literature is that changes in the exchange rate represent a shock to firm’s costs that are orthogonal to other shocks that effect the firm’s pricing decision and in reverse are not affected by firm pricing. This assumption is motivated by the empirical finding that exchange rate movements are disconnected from most macro-variables at the frequencies studied in the literature (1–2 year horizons). While this assumption might be more problematic for commodities such as oil or metals and for some commodity-exporting countries such as Canada, it is far less restrictive for most differentiated goods and most developed countries. Alternatively, the exchange rate change can be viewed not as a pure exogenous cost shock, but rather as an observable signal about the underlying fundamental macro shocks that differentially impact the costs of domestic and foreign firms. In this case, the coefficients on the exchange rate have less of a structural interpretation, but can still provide useful information about the nature of firm pricing.

The estimated exchange rate pass-through has typically been found to be incomplete. This incompleteness in pass-through can be consistent with various sources of firm-level and aggregate real rigidities, as well as with rational inattention and sticky information. The fact that firms export the same good to multiple destinations provides a mechanism to distinguish the source of the real rigidity. The pervasive evidence on pricing to market, the practice of charging different prices for the same good in different locations, provides support for the variable markup channel of incomplete pass-through.

In Sections 3.1 and 3.2 we review the empirical evidence on real rigidities in the closed and open economy literature respectively. We restrict attention to the most recent literature that uses micro-level price data.

3.1 Closed economy literature

In the closed economy literature the evidence on real rigidities typically arises from indirect identification strategies. Klenow and Willis (2006) evaluate the variable markup channel of real rigidities by calibrating a menu cost model with Kimball (1995) demand to match the U.S. Bureau of Labor Statistics (BLS) micro retail price evidence on the frequency and size of price adjustment. They conclude that the levels of real rigidity sufficient to generate the real or the nominal exchange rate. Since the two move closely at most horizons, either specification gives similar results. Typical pass-through regressions also include some controls for the foreign cost level, such as producer price index or manufacturing wage rate.

\footnote{See the literature following the seminal observation of Meese and Rogoff (1983).}
significant monetary non-neutrality has implausible implications for the required size of menu costs and idiosyncratic productivity shocks. Conceptually, significant real rigidities compress price dispersion so that much larger idiosyncratic shocks are required to match the size of price adjustment. At the same time, large menu costs are required to match the average durations of price rigidity given large idiosyncratic shocks.

Burstein and Hellwig (2007) replace Kimball demand with increasing marginal costs at the firm level to generate strategic complementarity in price setting. They calibrate the extent of decreasing returns to scale (elasticity of the marginal cost with respect to output) in order to match scanner data from a large chain of supermarkets in the Chicago area (Dominick’s) on the co-movement between prices and market shares. This calibration implies a moderate role for strategic complementarities generated via curvature in marginal costs.

BKM develop a test to assess a broad class of models with real rigidities. They examine the persistence of the actual and reset-price inflation series. Reset-price inflation is defined as an average reset price change within the group of goods that adjust prices in a given period, while reset prices for all other goods are indexed by reset-price inflation (for a formal definition see Section 4.1). In other words, this construct approximates desired-price inflation. If real rigidities are important, there should be significant persistence both in the actual and reset-price inflation series. The paper finds low persistence for actual price inflation and negative persistence for reset-price inflation (in both cases persistence is measured as the autoregression coefficient in an AR(1) specification for the inflation series), suggesting that the selection mechanism present in menu costs models (see Caplin and Spulber, 1987) offsets the effects of real rigidities and the real effects of money last less than nominal price durations.

In a different paper, Klenow and Willis (2007) find that prices respond to old information known prior to the most recent period of non-adjustment. They interpret this as evidence of sticky information in the spirit of Mankiw and Reis (2002), nevertheless, this evidence is also consistent with the presence of real rigidities and pricing complementarities that lead to incomplete price adjustment at the micro-level.

As previously mentioned, these tests for real rigidities are indirect relying mainly on calibrations or statistical properties of observable prices given that marginal costs and markups are not directly observable. The exception to this is a paper by Eichenbaum, Jaimovich, and Rebelo (2007) who use scanner data for a large grocery store chain in the U.S. They observe
both the wholesale price at which the store obtains the good (specific UPC) and the retail price at which the store sells it. At very short horizons the wholesale price can be viewed as the relevant marginal cost for the firm. They find that conditional on changing reference prices there is no evidence of variable markups, i.e. all of the reference wholesale cost change is passed through into reference retail prices.\(^8\)

### 3.2 Open Economy literature

The open economy literature evaluates the response of retail prices and at-the-dock prices to exchange rate shocks. While pass-through is less than one in both cases, pass-through into retail prices is always much lower than into at-the-dock prices. This is not surprising given that distribution costs, which are mainly non-traded costs unaffected by the exchange rate, are an important component of retail prices (e.g., see Campa and Goldberg, 2006). Goldberg and Knetter (1997) summarize a large body of the earlier empirical literature on pass-through.

A virtue of international price data is that one observes the prices at which the same firm sells its product in different destinations. There is considerable evidence of pricing to market, that is firms sell the same product at different prices in different destinations. This concept, first proposed in Dornbusch (1987) and Krugman (1987), attributes an important role to strategic complementarities that generate variable markups. In a recent paper Fitzgerald and Haller (2008) provide the most direct evidence of this phenomenon. They examine the pricing of Irish manufactures in domestic and export (U.K.) markets. They find a pronounced price differential response to exchange rate movements. Since the goods are manufactured in the same plant, the difference is attributable to variation in markups.\(^9\) Similarly, Burstein and Jaimovich (2008) use supermarket scanner data for stores in the U.S. and Canada and find that products with the same UPC code produced in the same country sell at different wholesale prices in the U.S. and Canada. More specifically, relative wholesale prices in the two markets move closely with the exchange rate.

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8Reference price (cost) is defined as the most often quoted price (cost) within a given time period.

9Specifically, Fitzgerald and Haller (2008) find that conditional on price adjustment in both markets the common-currency price differential across the two markets moves nearly one-to-one with the nominal exchange rate. According to the descriptive model (2) of Section 2, this is consistent with a \(\Gamma \gg 0\) provided that the competitor prices \(s_t\) in the two markets move little with the exchange rate as in the model of Atkeson and Burstein (2008).
In a recent set of papers, Gopinath and Rigobon (2008), Gopinath, Itskhoki, and Rigobon (2010) and Gopinath and Itskhoki (2010) use micro import price data collected by the U.S. Bureau of Labor Statistics for the period 1994-2005 to provide evidence of incomplete exchange rate pass-through at the dock even conditional on prices being changed. While incomplete exchange rate pass-through per se is consistent with evidence on variable markups it can arise from other sources of real rigidities as previously discussed. Gopinath and Itskhoki (2010) document a positive correlation between the frequency of price adjustment for a good and exchange rate pass-through conditional on price adjustment. They argue that this positive correlation is consistent with variation in markup elasticity across firms and cannot be consistent with other sources of firm heterogeneity. Neiman (2009) uses the same BLS import price data to document that consistent with the greater importance of the strategic complementarity channel for arms-length transactions as compared to intra-firm transactions, for differentiated products, intra-firm prices are characterized by less stickiness, less synchronization, and greater exchange rate pass-through.

Lastly, there are papers that evaluate the response of retail prices and wholesale prices for the same good to an exchange rate shock. Gopinath, Gourinchas, Hsieh, and Li (2009) use scanner data on retail and wholesale prices for stores of the same supermarket chain in the U.S. and Canada and find that in response to an exchange rate shock movements in relative retail prices in a common currency are explained mainly by movements in relative costs and very little by relative markups. Goldberg and Hellerstein (2006) and Nakamura and Zerom (2008) find that for beer and coffee sales respectively in a supermarket store the pass-through from exchange rate shocks to wholesale costs is incomplete, but the pass-through from wholesale costs to retail prices is close to complete. In a recent study, Berger, Faust, Rogers, and Steverson (2009) match goods in the BLS import price index to those in the BLS consumer price index for the period 1994-2007 and find that the overall distribution wedge, which is the percent difference between retail and at-the-dock prices does not vary systematically with the exchange rate which implies a nearly complete pass-through from at-the-dock to retail prices.\footnote{Additionally, a recent study by Berman, Martin, and Mayer (2009) compares the extent of markup variability across French exporters of different size. It finds that larger exporters have lower exchange rate pass-through, which is consistent with a model in which both the level and the variability of the markup increases with the productivity and hence size of the firm. Since exporting firms are typically larger, the implication of this finding is that the extent of markup variability can be greater in the international data than in the full sample of firms producing for the domestic market.}
Summary: Overall, a consistent finding across studies in the closed and open economy literature is that the variable markup channel of real rigidities is an important feature of the wholesale cost data but not of the retail price data. This is surprising consensus given the different approaches used in the closed and the open economy literature and the different datasets involved. In the terminology of Section 2, the empirical evidence is consistent with $\Gamma_R \approx 0$ and $\Gamma > 0$. Given the feedback from wholesale prices $s_t$ to retail prices $p_t$ in equation (1), one could expect sluggish adjustment in wholesale reset prices to generate slow adjustment in retail reset prices. The fact that BKM find a negative persistence for overall reset-price inflation could then be consistent with either a small $\alpha$ or a large transitory variance for $z_{it}$, as long as these shocks are not purely idiosyncratic. In the next section we explore explicitly using international price data the dynamics of price adjustment both unconditionally and conditioning on exchange rate shocks.

4 New Evidence on Real Rigidities

As discussed in Section 3.1, BKM find that regular and reset-price inflation for retail price data is consistent with the absence of important real rigidities. In this section we present evidence on the properties of regular and reset-price inflation for import price data using the BLS micro data on international prices. We then compare it to the properties of the inflation series conditional on exchange rate shocks, which allows us to evaluate the conditional response of prices to a given aggregate cost shock. We then study the dynamic properties of price adjustment at the firm-level.

First, we show that reset-price inflation for the import price data is less negatively autocorrelated as compared to that documented for retail prices. This is consistent with the conclusion that international prices depict higher real rigidities as compared to retail prices. Second, we present evidence regarding the sluggish response of price changes to exchange rate shocks. We do this in two ways. One, we project aggregate regular and reset-price inflation on lags of the exchange rate changes. We find that the autocorrelation of the fitted series is substantially higher as compared to that of the unconditional series. Two, using the micro data, we show that individual prices, conditional on changing, respond to exchange rate shocks that were realized prior to the previous price adjustment. Both pieces of evidence suggest a more important role for real rigidities as compared to the conclusions one
can draw from the analysis of unconditional aggregate inflation series.\footnote{This feature is consistent with the evidence in Boivin, Giannoni, and Mihov (2009) who find that while overall disaggregated prices are volatile they are sluggish in response to specific shocks, namely macroeconomic and monetary shocks.}

While this evidence is supportive of the presence of sizeable real rigidities, it does not discriminate between various sources of real rigidities. Therefore, we next try to assess specifically the importance of strategic complementarities in price setting. We do so by studying the response of firm’s prices to the shocks to its competitors. Our central result here is that firm’s prices respond strongly to the prices of its competitors and this channel explains a significant fraction of exchange rate pass-through into prices.

The data used in this section is the micro import price data underlying the construction of the U.S. import price index. This covers the period 1994-2005. For details regarding the data see Gopinath and Rigobon (2008).

4.1 Reset-price inflation

We follow BKM in estimating the reset-price inflation series for U.S. imports. The log price of a good $i$ at time $t$ is denoted by $p_{it}$. The log of the reset price at time $t$ for good $i$ is denoted by $p_{it}^*$ and defined as:

$$
p_{it}^* = \begin{cases} 
p_{it} & \text{if } p_{it} \neq p_{i,t-1}, \\
p_{i,t-1}^* + \pi_{i,t}^* & \text{if } p_{it} = p_{i,t-1}, \end{cases}
$$

where $\pi_{i,t}^*$ is the average reset-price inflation of those goods whose prices change at time $t$. The inflation of the actual price series is referred to as \textit{regular-price inflation}. In a Calvo pricing environment, where firms only differ in the exogenous frequency of price adjustment, the behavior of reset-price inflation will capture the extent of real rigidities. As discussed in detail in BKM, in the presence of real rigidities reset prices will adjust sluggishly as firms have multiple price adjustments before they fully respond to a shock. The regular-price inflation series display even greater persistence as each firm waits for the random arrival of the opportunity to change its price. In the case of state-dependent pricing or variations in desired price responses across sectors, the behavior of measured reset prices is not as direct because there is selection of which firms change prices. Gopinath and Itskhoki (2010) show that sectors that have a higher frequency of price adjustment also have higher long-run pass-
through (less real rigidities). Consequently, the reset-price series is affected by those goods that change prices more frequently and pass-through eventually a lot more.

It is important to point out that the sample sizes in the import price data are not as large as in the CPI data, and given the low frequency of price adjustment, the number of actual price changes used to impute the reset price series is low.\textsuperscript{12} This would suggest that the import reset-price inflation series is more subject to noise as compared to the construct of BKM for the U.S. CPI.

The persistence and volatility of the regular and reset-price inflation series within various sub-samples are reported in Table 1.\textsuperscript{13} We measure persistence as the AR(1) coefficient for the series and we report the standard deviation of the series as a measure of volatility.\textsuperscript{14} The first row of Table 1 reports these moments for consumer-price inflation from BKM. The rest of the rows provide these statistics for the BLS import-price data.\textsuperscript{15}

Columns 1 and 2 report statistics for the unconditional inflation series. For all subsamples considered, import price inflation is more persistent relative to consumer-price inflation as calculated by BKM.\textsuperscript{16} For example, for the dollar-priced imports, the persistence of inflation is 0.56, while for consumer prices it is close to zero (−0.05). The import-price inflation is also more volatile. A similar comparison holds for reset-price inflation. While BKM find that reset-price inflation for consumer prices is negatively autocorrelated (−0.41), we find essentially zero autocorrelation for import prices (e.g., −0.04 for the dollar-priced imports). This difference in persistence of consumer-price and import-price inflation is consistent with the different nature of pricing at the consumer and intermediate-good levels (since most

\textsuperscript{12}In the sample that excludes petrol classifications the median number of price observations per month is 6,335 and the median number of price observations whose price is different from the previous month is 770.

\textsuperscript{13}The small number of price changes limits the analysis to large groups of goods.

\textsuperscript{14}We estimate an AR(1)-coefficient so as to compare our results directly to BKM. If, as argued by Stock and Watson (2007), CPI inflation is better modeled as an ARMA(1,1) or an IMA(1,1), then the first order autocorrelation understates the long-run persistence of the series. We have estimated other measures of persistence such as the variance ratio to the long-run variance of the series. Although these measures suggest greater persistence for the inflation series, our comparative results for the conditional and unconditional inflation still hold.

\textsuperscript{15}In all specifications we exclude petrol classifications. For each series we use 2002 weights at the 4-digit level to aggregate across prices. More precisely, for actual inflation, we estimate mean price change by 4-digit harmonized code for each month, then we average across the different harmonized codes using weights at the 4-digit level. For the reset-price inflation, we assume that at the start of our sample in January 1994 all prices were reset prices, which is an initial condition assumption. Then we follow the formula in equation (4) to construct the reset-price inflation, $\pi^*_t$.

\textsuperscript{16}BKM sample is longer, 1989-2008. They also exclude energy, fresh fruit and vegetables and eggs. We report their results for the sample that excludes sales price.
Table 1: Volatility and persistence of regular and reset-price inflation

<table>
<thead>
<tr>
<th></th>
<th>Unconditional</th>
<th>Conditional on ER</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>AR(1) St.D.</td>
<td>AR(1) St.D.</td>
</tr>
<tr>
<td>Regular-price inflation</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Consumer prices (from BKM)</td>
<td>−0.05 0.14%</td>
<td>— —</td>
</tr>
<tr>
<td>Import prices</td>
<td>0.51 0.33%</td>
<td>0.55 0.25%</td>
</tr>
<tr>
<td>− Dollar-priced goods</td>
<td>0.56 0.27%</td>
<td>0.79 0.18%</td>
</tr>
<tr>
<td>− Market transactions</td>
<td>0.43 0.30%</td>
<td>0.70 0.20%</td>
</tr>
<tr>
<td>Reset-price inflation</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Consumer prices (from BKM)</td>
<td>−0.41 0.95%</td>
<td>— —</td>
</tr>
<tr>
<td>Import prices</td>
<td>0.02 1.60%</td>
<td>0.31 0.82%</td>
</tr>
<tr>
<td>− Dollar-priced goods</td>
<td>−0.04 1.20%</td>
<td>0.33 0.75%</td>
</tr>
<tr>
<td>− Market transactions</td>
<td>−0.03 1.70%</td>
<td>0.17 0.91%</td>
</tr>
</tbody>
</table>

Note: Import prices exclude petrol classifications; the rows for market transactions include only dollar-priced goods. The last two columns project the inflation series on the current and 24 lags of the log changes of the U.S. trade-weighted exchange rate and compute the moments for the projected series.

An import feature of the international data is that we can examine the response of inflation series to a specific shock, namely the exchange rate shock. This has advantages over just looking at reset-price inflation that aggregates (imperfectly in small samples) across idiosyncratic, sectoral and aggregate shocks. We accordingly project the regular and reset-price inflation series on current and 24 lags of the log changes of the U.S. trade-weighted nominal exchange rate. We use the fitted values from this regression and estimate the AR(1) coefficient and standard deviation for the fitted series. The results are reported in the last two columns of Table 1. In all cases regular and reset-price inflation conditional (projected) on the exchange rate shocks exhibits more persistence. For instance, in the case of dollar-priced imports the conditional regular-price inflation series has an overall persistence of 0.79, while its unconditional persistence is 0.56. Similarly, for the reset-price inflation series the conditional persistence is 0.33 as opposed to the unconditional persistence of −0.04.

This evidence is consistent with the presence of multiple shocks of different degrees of

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17 As documented by Stock and Watson (2007) among others, the short-run persistence of consumer-price inflation decreased in the 1990s.
persistence driving the inflation process. Under these circumstances the unconditional persistence of the inflation series might not accurately reflect the underlying sluggishness in the micro-level price adjustment. In the next subsection we present further evidence of the sluggish response of prices to exchange rate shocks by examining the behavior of individual prices and their response to lagged exchange rates changes.

4.2 Micro-dynamics of price adjustment

At the good level, we estimate pass-through into prices of exchange rate shocks realized during the most recent period of price non-adjustment and of those that were realized prior to the previous price adjustment. In the absence of real rigidities, all adjustment should take place at the first instance of price change and hence the coefficient on the exchange rate change prior to the previous price adjustment should be zero.

Formally, we estimate the following regression:

$$\Delta \bar{p}_{i,t} = \beta_1 \Delta \tau_1 e_{i,t} + \beta_2 \Delta \tau_2 e_{i,t-\tau_1} + Z_{i,t}' \gamma + \epsilon_{it}, \quad (5)$$

where $i$ indexes the good, $\Delta \bar{p}_{i,t}$ is the change in the log dollar price of the good, conditional on price adjustment in the currency of pricing. $\Delta \tau_1 e_{i,t} \equiv e_{i,t} - e_{i,t-\tau_1}$ is the cumulative change in the log of the bilateral nominal exchange rate over the duration when the previous price was in effect (which we denote $\tau_1$). Similarly, $\tau_2$ denotes the duration of the previous price of the firm so that $\Delta \tau_2 e_{i,t-\tau_1} \equiv e_{i,t-\tau_1} - e_{i,t-\tau_1-\tau_2}$ is the cumulative exchange rate change over the previous (the one prior to the previous price change) period of non-adjustment. Figure 1 illustrates a hypothetical price series: if $\Delta \bar{p}_{i,t}$ is the price change between $t_3$ and $t_{LL}$, $\Delta \tau_1 e_{i,t}$ is then the exchange rate change between $t_3$ and $t_{LL}$ and $\Delta \tau_2 e_{i,t-\tau_1}$ is the exchange rate change between $t_2$ and $t_3$. Finally, $Z_{i,t}$ includes controls for the cumulative change in the foreign consumer price level, the US consumer price level and fixed effects for every BLS-defined primary strata (mostly 2–4-digit harmonized codes) and country pair. We allow $Z_t$ to include lagged foreign and domestic inflation. The standard errors are clustered at the level of the fixed effects. We restrict the sample to non-petrol, dollar-priced goods and

---

17 In the BLS database, the original reported price (in the currency of pricing) and the dollar converted price are both provided. We use the latter, conditional on the original reported price having changed. Since the first price adjustment is censored from the data, we also perform the analysis excluding the first price change and find that the results are not sensitive to this assumption.
market transactions. Note that this specification requires the goods to have at least two price adjustments during their life. Since there are several goods that have only one price change during their life, we lose about 30% of the goods.

![Figure 1: Hypothetical good-level price series and nominal exchange rate](image)

By conditioning on a price change we get past the period of nominal rigidity, which is essential to understanding the role of real rigidities. In a Calvo pricing environment, since the decision to change prices is exogenous, there are no selection issues to be concerned with. However, in an environment with endogenous frequency of price adjustment, conditioning on a price change will induce a bias in the exchange rate pass-through estimates as it generates a conditional correlation between the exchange rate and the residual even if the unconditional correlation is zero. This is problematic if one tries to provide a structural interpretation to the coefficient. This is not our purpose here. We use this specification to provide a relation in the data between the response of prices conditional on adjusting to lagged exchange rate changes. Later we will estimate these regressions in the model simulated data, where frequency is chosen endogenously and infer how well models with real rigidities perform in matching the facts in the data. This exercise is accordingly similar to that of trying to match the behavior of the reset-price inflation series, a series that also is effected by selection issues. Furthermore, when the selection bias is strong the pass-through coefficient on first adjustment ($\beta_1$) is biased upwards while the pass-through coefficient on second adjustment ($\beta_2$) is biased downwards, which makes it harder to identify the presence of real rigidities.
Table 2: Dynamic response to exchange rate shocks

<table>
<thead>
<tr>
<th></th>
<th>$\beta_1$</th>
<th>s.e.($\beta_1$)</th>
<th>$\beta_2$</th>
<th>s.e.($\beta_2$)</th>
<th>$N_{obs}$</th>
<th>$R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>All countries</td>
<td>0.11</td>
<td>(0.02)</td>
<td>0.08</td>
<td>(0.01)</td>
<td>69,917</td>
<td>0.01</td>
</tr>
<tr>
<td>Non-OECD</td>
<td>0.06</td>
<td>(0.02)</td>
<td>0.04</td>
<td>(0.01)</td>
<td>37,108</td>
<td>0.01</td>
</tr>
<tr>
<td>High-income OECD</td>
<td>0.23</td>
<td>(0.02)</td>
<td>0.18</td>
<td>(0.02)</td>
<td>32,809</td>
<td>0.02</td>
</tr>
<tr>
<td>Euro area</td>
<td>0.22</td>
<td>(0.04)</td>
<td>0.14</td>
<td>(0.03)</td>
<td>5,933</td>
<td>0.02</td>
</tr>
<tr>
<td>Japan</td>
<td>0.26</td>
<td>(0.04)</td>
<td>0.24</td>
<td>(0.03)</td>
<td>4,249</td>
<td>0.06</td>
</tr>
<tr>
<td>Canada</td>
<td>0.28</td>
<td>(0.16)</td>
<td>0.34</td>
<td>(0.05)</td>
<td>14,620</td>
<td>0.01</td>
</tr>
<tr>
<td>Differentiated goods</td>
<td>0.14</td>
<td>(0.02)</td>
<td>0.10</td>
<td>(0.02)</td>
<td>21,360</td>
<td>0.02</td>
</tr>
<tr>
<td>No missing prices</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>All countries</td>
<td>0.10</td>
<td>(0.02)</td>
<td>0.08</td>
<td>(0.01)</td>
<td>45,765</td>
<td>0.01</td>
</tr>
<tr>
<td>High-income OECD</td>
<td>0.19</td>
<td>(0.03)</td>
<td>0.13</td>
<td>(0.02)</td>
<td>22,436</td>
<td>0.01</td>
</tr>
</tbody>
</table>

Note: $\beta_1$ and $\beta_2$ are the pass-through coefficients at the first and second rounds of price adjustment respectively, estimated from regression (5). Standard errors in brackets are clustered at the country $\times$ 4-digit-sector level. $N_{obs}$ is the number of price changes in the sample. Results under “No missing prices” in the lower panel exclude from the sample all price changes which were followed or preceded by a missing price.

Table 2 reports the results from estimation of regression (5). We provide evidence for various sub-samples of the data. Across all specifications we find that exchange rate shocks that took place prior to the current period of non-adjustment have a significant effect on current price adjustments. This is consistent with the existence of real rigidities in pricing. The strength of these lagged effects is much stronger than what would be suggested purely by the reset-price inflation series. The first row of Table 2 points out that the elasticity of current price changes to lagged exchange rate shocks for dollar-priced goods is 0.08, which is only slightly smaller than the response to the contemporaneous exchange rate movement (equal to 0.11). The importance of these lagged effects is consistently present in all sub-samples. For the high-income OECD sample the contemporaneous and lagged responses are 0.23 and 0.18 respectively. For the non-OECD sample the pass-through rates are overall lower but there are still important lagged effects. This is similarly documented for the Euro-area countries, Japan and Canada.

In the data there can be spells that have missing prices and where the new price follows or precedes a missing price. In this case the exact timing of the price change is not known, so lagged effects can arise from getting the timing wrong. The last two rows of Table 2 checks
for the robustness of the results by including in the sample only those price changes that were not followed or preceded by a missing price. This changes the sample composition, but lagged responses are still strongly evident.

The results in this section are consistent with the evidence in Gopinath, Itskhoki, and Rigobon (2010) that long-run pass-through is much higher than pass-through conditional on the first adjustment to the exchange rate shock. Here we present explicitly the dynamics and extend the sample to more countries.

We also divide goods into four equally-sized bins based on their frequency of price adjustment and estimate equation 5 within each bin separately. The purpose of this exercise is to evaluate if the importance of lags varies across goods with different frequencies of price adjustment. One conjecture may be that it is only the very high-frequency goods that have multiple price adjustments to respond to a shock. In fact this is not the case as we find that lags are important even for goods that adjust prices very infrequently: for example, the first quartile contains goods that adjust prices less than once a year and in the first round the pass-through is 0.12, while it is 0.08 in the second round. This finding further assuages concerns about measurement issues with the timing of price adjustment.19

<table>
<thead>
<tr>
<th>Table 3: Dynamic response to exchange rate shocks: by end-use sectors</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
</tr>
<tr>
<td><strong>Panel A: All countries</strong></td>
</tr>
<tr>
<td>Food, feed and beverages</td>
</tr>
<tr>
<td>Industrial supplies and materials</td>
</tr>
<tr>
<td>Capital goods, except automotive</td>
</tr>
<tr>
<td>Consumer goods (non-food)</td>
</tr>
<tr>
<td><strong>Panel B: High-income OECD</strong></td>
</tr>
<tr>
<td>Food, feed and beverages</td>
</tr>
<tr>
<td>Industrial supplies and materials</td>
</tr>
<tr>
<td>Capital goods, except automotive</td>
</tr>
<tr>
<td>Consumer goods (non-food)</td>
</tr>
</tbody>
</table>

*Note: see notes to Table 2.*

We also break the sample down by the ‘end use’ of the product. Again we find the second

19 Refer to Gopinath and Itskhoki (2010) for a detailed analysis of pass-through conditional on first adjustment and pass-through conditional on many rounds of adjustment across goods with different frequencies.
rounds of price adjustment to be significant for most end-use categories. In the case of ‘Food, feed and beverages’ exported from the non-OECD counties (which dominate the sample of all countries in Panel A), the pass-through is generally very low and insignificant from zero. In the case of ‘Consumer goods (non-food and excluding automotive)’ the dynamics is less evident. However, one should be careful about interpreting the results for this sub-sample because these goods more often have a fixed price during their life and then get discontinued and replaced. Since we do not observe price changes across discontinuations we might be excluding important adjustments that take place at the time of product replacement.

Overall, the micro evidence is consistent with the aggregate level evidence of sluggish adjustment to exchange rate changes. This sluggishness is consistent with many forms of real rigidities including variable markups, the Basu (1995) intermediate input channel wherein each firm’s output is used as an input in production, sluggish response of other factor costs (like wages) to the underlying source of exchange rate shocks. Equivalently, it could arise from rational inattention or sticky information. The next subsections evaluates how the extent of product market competition affects the patterns of pass-through in order to identify the effects of strategic complementarities in price setting.

4.3 Competition and pass-through

In this section we evaluate the importance of the strategic complementarity in price setting for incomplete pass-through using some measures that capture the pricing behavior of competitors and measures that capture the extent of competition within sectors. Ideally, one would need to perform this analysis with detailed industry data for each product and information on prices and market shares of different firms. This data however does not exist for the large number of products included in our study. Consequently, we use some proxies here and they are necessarily imperfect. Nevertheless, we find evidence that is consistent with the presence of significant strategic complementarities at the firm level.

**Trade-weighted versus bilateral exchange rate**: First, we evaluate the response of each firm’s pricing to its own bilateral exchange rate as compared to its response to the trade-weighted exchange rate. Movements in the trade-weighted exchange rate can be viewed as a more aggregate shock that effects a larger fraction of firm’s competitors as compared to a shock that only effects the bilateral exchange rate. An alternative interpretation could be
that prices of a firm that uses production inputs from the rest of the world are sensitive to the movements in the trade-weighted exchange rate because it impacts the firm’s costs.

Figure 2: Impulse responses to bilateral and U.S. trade-weighted exchange rate

More specifically we estimate the following standard pass-through regression,

\[ \Delta p_{k,t} = \alpha_k + \sum_{j=0}^{n} \beta_j \Delta e_{k,t-j} - j + \sum_{j=0}^{n} \gamma_j \pi_{k,t-j} + \epsilon_{k,t}, \]  

(6)

where \( k \) indexes the country, \( \Delta p \) is the average monthly log price change in dollars, \( \pi \) is the monthly foreign-country inflation using the consumer price index, \( n \) is the number of monthly lags that varies from 1 to 24. \( \Delta e_{k,t-j} - j \) is either a bilateral nominal exchange rate or the U.S. trade-weighted nominal exchange rate. Figure 2 plots \( \sum_{j=0}^{n} \beta_j \) as a function of \( n \) for each case, where we estimate a pooled regression restricting the coefficients \( \beta_j \) to be the same across countries. Panel (a) of the figure plots the results for the all-countries sample, while Panel (b) does it for the high-income OECD sub-sample. In both figures it is evident that the pass-through from the trade-weighted exchange rate exceeds the bilateral exchange rate pass-through, consistent with the hypothesis that firm’s prices are responsive to cost shocks of firm’s competitors. In further analysis we find that this pattern is evident for countries in the Euro area as well as the non-OECD countries, while it is less evident for Japan, Canada and the United Kingdom.

We also perform the analysis using the individual price data, conditional on a price change. We evaluate the response to the bilateral exchange rate change since the last time the price was adjusted and to movements in the U.S. trade-weighted exchange rate that is orthogonal to the bilateral exchange rate for the country. More specifically, we run a
Table 4: Pass-through of bilateral and U.S. trade-weighted exchange rate

<table>
<thead>
<tr>
<th></th>
<th>Bilateral ER</th>
<th>T-W ER</th>
<th>N_{obs}</th>
<th>R^2</th>
</tr>
</thead>
<tbody>
<tr>
<td>All countries</td>
<td>0.11 (0.01)</td>
<td>0.19 (0.02)</td>
<td>83,064</td>
<td>0.01</td>
</tr>
<tr>
<td>Non-OECD</td>
<td>0.07 (0.02)</td>
<td>0.18 (0.02)</td>
<td>46,420</td>
<td>0.01</td>
</tr>
<tr>
<td>High-income OECD</td>
<td>0.22 (0.02)</td>
<td>0.17 (0.05)</td>
<td>32,809</td>
<td>0.02</td>
</tr>
<tr>
<td>Euro area</td>
<td>0.27 (0.03)</td>
<td>0.31 (0.07)</td>
<td>7,856</td>
<td>0.03</td>
</tr>
<tr>
<td>Japan</td>
<td>0.21 (0.04)</td>
<td>0.17 (0.06)</td>
<td>5,733</td>
<td>0.02</td>
</tr>
<tr>
<td>Canada</td>
<td>0.23 (0.12)</td>
<td>0.12 (0.10)</td>
<td>16,221</td>
<td>0.01</td>
</tr>
<tr>
<td>Differentiated</td>
<td>0.12 (0.01)</td>
<td>0.17 (0.03)</td>
<td>21,360</td>
<td>0.02</td>
</tr>
</tbody>
</table>

Note: The first column reports pass-through conditional on price adjustment of the bilateral exchange rate shocks. The second column reports the pass-through of the component of the U.S. trade-weighted exchange rate orthogonal to the bilateral exchange rate (i.e., a residual from the projection of the trade-weighted exchange rate on the bilateral exchange rate). Clustered standard errors in brackets.

first-stage regression where we regress the trade-weighted exchange rate on the bilateral exchange rate. We calculate the residual and then estimate a second stage regression where we regress the price change, conditional on adjustment, on the cumulative change in the bilateral exchange rate and in the residual.\(^\text{20}\) We include a control for the cumulative change in foreign-country CPI inflation since the last price change. The results are reported in Table 4. Consistent with the evidence in Figure 2 using aggregate price changes, the effect of the residual is almost as large as the direct effect of the bilateral exchange rate, and in some cases it is even larger.

Competitor prices: A more direct test of the presence of strategic complementarities is to evaluate whether changes in the competitor prices affect the pricing decisions of the firm. We do so by estimating the following regression:

\[
\Delta \bar{p}_{i,k,t} = \beta_e \Delta \tau_{1e,i,k,t} + \beta_I \Delta \tau_I P_{k,t} + \gamma Z_{i,t} + \epsilon_{i,t},
\]

where \(\Delta \bar{p}_{i,k,t}\) is the change in the log dollar price of good \(i\) in sector \(k\), conditional on price adjustment, and \(\Delta \tau_{1e,i,k,t}\) is the cumulative change in the log of the bilateral nominal exchange rate over the duration for which the previous price was in effect. Now \(\Delta \tau_I P_{k,t}\) is

\(^{20}\)The coefficient on the residual will be equivalent to the coefficient on the trade-weighted exchange rate obtained from regressing the price change on the bilateral and the trade-weighted exchange rate, but the coefficient on the bilateral exchange rate will be different across the two specification.
a measure of the cumulative price change by firms other than firm $i$ in sector $k$.$^{21}$ We also estimate the same regression for the life-long change in the price of the good and refer to the coefficients in this case as *long-run pass-through*. In terms of Figure 1, this corresponds to having $p_{t_{LL}} - p_{t_0}$ on the left-hand side of (7) and corresponding cumulative changes in variables on the right-hand side.$^{22}$ Finally, $Z_{i,t}$ represents the cumulative change in the consumer price index in the foreign country. We again restrict the sample to non-petrol, dollar-priced goods and market transactions. We include fixed effects for every BLS-defined primary strata (mostly 2–4-digit harmonized codes) and country pair; the standard errors are clustered at the level of the fixed effects.

Table 5: Response to competitor prices

<table>
<thead>
<tr>
<th>Panel A: Pass-through conditional on first price change</th>
<th>$\beta_e$</th>
<th>s.e.($\beta_e$)</th>
<th>$\beta_I$</th>
<th>s.e.($\beta_I$)</th>
<th>$N_{obs}$</th>
<th>$R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>No $\Delta_1 P_{k,t}^I$</td>
<td>0.13</td>
<td>(0.01)</td>
<td>—</td>
<td>—</td>
<td>83,056</td>
<td>0.01</td>
</tr>
<tr>
<td>$\Delta_1 P_{k,t}^I$ (Primary strata)</td>
<td>0.07</td>
<td>(0.01)</td>
<td>0.61</td>
<td>(0.02)</td>
<td>78,942</td>
<td>0.13</td>
</tr>
<tr>
<td>$\Delta_1 P_{k,t}^I$ (10-digit HTS)</td>
<td>0.04</td>
<td>(0.01)</td>
<td>0.61</td>
<td>(0.02)</td>
<td>59,972</td>
<td>0.25</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Panel B: Long-run pass-through</th>
<th>$\beta_e$</th>
<th>s.e.($\beta_e$)</th>
<th>$\beta_I$</th>
<th>s.e.($\beta_I$)</th>
<th>$N_{obs}$</th>
<th>$R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>No $\Delta_1 P_{k,t}^I$</td>
<td>0.31</td>
<td>(0.03)</td>
<td>—</td>
<td>—</td>
<td>16,145</td>
<td>0.06</td>
</tr>
<tr>
<td>$\Delta_1 P_{k,t}^I$ (Primary strata)</td>
<td>0.13</td>
<td>(0.02)</td>
<td>0.66</td>
<td>(0.03)</td>
<td>15,273</td>
<td>0.24</td>
</tr>
<tr>
<td>$\Delta_1 P_{k,t}^I$ (10-digit HTS)</td>
<td>0.16</td>
<td>(0.01)</td>
<td>0.62</td>
<td>(0.03)</td>
<td>11,379</td>
<td>0.34</td>
</tr>
</tbody>
</table>

Note: Results from estimation of equation (7). The coefficient $\gamma$ on the consumer-price inflation in the foreign country also shrinks along with $\beta_e$ when we include the control for competitor prices.

The results are reported in Table 5. The first row of each panel (labeled ‘No $\Delta_1 P_{k,t}^I$ ’) presents the results where we exclude any industry competition effect. The next two rows include industry price effects aggregated at the BLS-defined primary strata level (mostly 2–4-digit harmonized codes) and at the 10-digit harmonized code level respectively. As is evident in all specifications, the effect of competitor prices is large and highly significant. Moreover,

$^{21}$For each good $i$ we calculate the average monthly import price change for all goods in the same 10-digit or BLS-defined primary strata classification, excluding good $i$. Then we add 1 to the cumulates change over time to arrive at an industry price index for each good. In our main specification we include non-adjacent price changes, that is if prices are available for January and March, but are missing for February, the price change in March refers to the percentage difference between the March and January price. We also perform the analysis where we include only price changes across adjacent months and obtain qualitatively the same results.

$^{22}$Results are unaffected if we exclude the first price change that can be censored. For more details on the comparison between life-long pass-through and pass-through conditional on first price adjustment see Gopinath and Itskhoki (2010).
it significantly reduces the direct response of prices to the exchange rate shock. If one
were to do a back-of-the-envelope calculation of the extent of strategic complementarities
using expression (2) of our accounting framework in Section 2, one obtains a measure of
markup elasticity of $\Gamma \approx 1.5$ (from $\Gamma / (1 + \Gamma) \approx 0.6$). This value is consistent with the
required markup elasticity to match the evidence of incomplete long-run pass-through, as
we discuss in Gopinath, Itskhoki, and Rigobon (2010) and Gopinath and Itskhoki (2010).
Furthermore, note that the direct impact of exchange rate changes ($\beta_e$) still increases from
the specification conditional on one price adjustment (Panel A) to the the life-long specification
(Panel B), even when we control for competitor prices. This suggests that, although strategic
complementarities are an important feature of price setting, it does not fully explain the
delayed pass-through of exchange rate shocks and there are other sources of real rigidity
present in the data.

**Sector concentration:** Finally, we relate the incompleteness in exchange rate pass-through
to certain sectoral features that proxy for the level of competition among importers. An
important distinction between retail and wholesale prices is that the latter originate from
business-to-business transactions. Consequently, the strength of bargaining power of the
buyer can impact the extent of pass-through. To evaluate this hypothesis we use measures
of concentration in the import sectors provided to us by the BLS. The BLS constructs a
Herfindahl index for importers and a measure of the number of importers that make up the
top 50% of trade using census data on all imports entering the U.S.\(^\text{23}\) We were provided
estimates at the level of BLS-defined primary strata. We estimate the following long-run
pass-through regression where we interact the exchange rate change with a measure of concentra-
tion:

$$
\Delta \bar{p}_{i,k,t} = \beta \Delta \tau_{r_e} e_{i,k,t} + \psi \left( \Delta \tau_{r_e} e_{i,k,t} \cdot C_k \right) + Z'_{i,k,t} \gamma + \epsilon_{i,t},
$$

where the second regressor is the interaction of the exchange rate change with a given
concentration measure. In $Z_{i,k,t}$ we include separate controls for the concentration measure,
the change in CPI inflation (both stand alone and interacted with the concentration measure)
and country fixed effects. All standard errors are clustered at the primary-strata level.

The results are reported in Table 6 for the two measures of concentration. While the
point estimates in both regressions suggest that sectors that are dominated by a few large

\(^{23}\)The BLS constructs this to help create their sampling frame.
importers (high Herfindahl index and small number of firms in the top 50%) have lower pass-through from foreign firms, the standard errors on these estimates are large. Overall, the evidence is inconclusive. We also performed this exercise for pass-through conditional on first price change, as well as restricted the sample to differentiated goods only, and in all cases found no clear relationship in the data.

5 Model

In this section we quantitatively evaluate a reduced-form sticky price model with a retail and a wholesale sector. Consistent with the data, we allow for variable markups at the wholesale level and constant markups at the retail level. In the next section we discuss a bargaining-based micro-foundation for this reduced-form assumptions. In the existing monetary literature there is typically no interesting distinction made between the retail and wholesale sectors. The goal is to evaluate the behavior of regular and reset-price inflation, both unconditional and conditional on aggregate shocks, as well as the dynamic response of good-level prices conditional on changing so as to compare it to the evidence in Section 4. We also evaluate the extent of monetary non-neutrality generated by this source of real rigidities.

The model generates sluggishness in response to monetary shocks in wholesale prices and this feeds into the slow adjustment of retail prices. However, aggregate inflation and reset-price inflation exhibit little persistence since their movements are dominated by more transitory shocks. On the other hand, conditional on monetary shocks or exchange-rate-like shocks, inflation series exhibit considerable persistence. This is consistent with the data. The model however fails to match the slow dynamics in price adjustment documented in the empirical data suggesting a need for an additional source of persistence in prices.

Similarly, the output series can exhibit significant monetary non-neutralities if the money

<table>
<thead>
<tr>
<th></th>
<th>$\beta$</th>
<th>s.e.$(\beta)$</th>
<th>$\psi$</th>
<th>s.e.$(\psi)$</th>
<th>$N_{\text{obs}}$</th>
<th>$R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Herfindahl index</td>
<td>0.30</td>
<td>(0.03)</td>
<td>-0.02</td>
<td>(0.02)</td>
<td>12,432</td>
<td>0.06</td>
</tr>
<tr>
<td>No. importers in top 50%</td>
<td>0.23</td>
<td>(0.02)</td>
<td>0.02</td>
<td>(0.02)</td>
<td>12,435</td>
<td>0.06</td>
</tr>
</tbody>
</table>
growth process is sufficiently persistent. While calibrated real rigidities in the form of variable markups increase the size of the contract multiplier, their effects are modest unless they are coupled with exogenous sources of persistence.

We begin by laying out the familiar equilibrium conditions of the model while the details are relegated to the appendix. We first consider the model with Calvo price setting and later evaluate the robustness of the predictions in a menu cost model of price setting in the wholesale sector. We should clarify that this is only a numerical exercise where the empirical evidence disciplines some parameters of the model.

5.1 Setup of the model

Wholesale sector: Wholesale firms use labor and a constant returns to scale production function to produce intermediate goods. Therefore, a wholesale firm \( j \) faces a constant marginal cost \( mc_{jt} \) (all variables in logs):

\[
mc_{jt} = w_t + \phi_j e_t - a_{jt},
\]

where \( w_t \) is the nominal wage rate and \( e_t \) captures an exogenous exchange rate-like shock that affects the wholesale firm with elasticity \( \phi_j \) that varies across firms. Further, \( a_{jt} \equiv \bar{a}_t + \tilde{a}_{jt} \) is the sum of an aggregate (wholesale-sector-wide) and idiosyncratic (firm-specific) shock to the firm; \( a_{jt} \) represents some combination of shocks to the marginal cost and the markup that affects the firm’s desired price.

The desired log-price of a wholesale producer equals a log desired markup over the marginal cost:

\[
\tilde{s}_{jt} = \mu_{jt} + mc_{jt}.
\]

By desired prices we mean prices that a firm would set if it could adjust prices every period in a given general equilibrium environment; desired price is not the same as a reset price, which is set for a number of future periods.

We assume variable markups that depend on the firm’s relative price:

\[
\mu_{jt} = \bar{\mu} - \Gamma(s_{jt} - s_t),
\]
where $\bar{\mu}$ is the steady state level of markup, $\Gamma$ is the elasticity of markup with respect to price, and

$$s_t \equiv \int s_j \, dj$$

is (an approximation to) the price index in the wholesale sector. This specification is a first-order approximation to a more general model of variable markups. For example, it can be obtained from the Kimball demand with non-constant elasticity (e.g., see Klenow and Willis, 2006; Gopinath and Itskhoki, 2010), or a model of strategic interactions between large firms (e.g., see Atkeson and Burstein, 2008, and the bargaining model of Section 6).

**Retail sector:** In the retail sector, firms combine labor and intermediate goods supplied by the wholesale sector to produce a final good. Specifically, firm’s $i$ marginal cost is given by:

$$mc^R_{it} = \alpha s_t + (1 - \alpha)w_t - z_{it},$$

where $\alpha$ is the production-cost share of intermediate goods, and $z_t \equiv \bar{z}_t + \tilde{z}_t$ is the sum of aggregate (retail-sector-wide) and idiosyncratic (firm-specific) marginal cost and/or markup shocks that affect the firm’s desired price. Note that we assume that the exchange rate shock, $e_t$, does not affect the retail sector directly, and each retail firm uses a full bundle of intermediate goods as input in production.

We assume constantmarkup pricing in the retail sector (e.g., monopolistic competition and CES demand) so that the desired price of firm $i$ is given by

$$\tilde{p}_{it} = \bar{\mu} + mc^R_{it}.$$  

In the notation of Section 2 it is equivalent to assuming $\Gamma_R = 0$. This assumption along with $\Gamma > 0$ is consistent with the evidence discussed in Section 3. In Section 6 we provide one economic explanation that can rationalize this difference.

**Wage rate and real output:** We assume that the nominal wage rate depends on the consumer (final-good) price level, $p_t$, and aggregate nominal spending, $m_t$:

$$w_t = \gamma m_t + (1 - \gamma)p_t, \quad \gamma > 0.$$  

(9)
This reduced form model of wages is common in macroeconomics (e.g., see Chari, Kehoe, and McGrattan, 2000; Burstein and Hellwig, 2007) and can be derived, for example, from a cash-in-advance model of money demand and the intratemporal optimality condition for consumption-leisure choice. Small values of $\gamma$ imply a more gradual response of wages to aggregate nominal spending shocks and hence are a stand-in for various unmodeled aggregate real rigidities such as real wage rigidity (Blanchard and Galí, 2007), segmented labor markets (Woodford, 2003), and round-about production structure (Basu, 1995).

With our definition of $m_t$ as aggregate nominal spending, real output is given by

$$y_t = m_t - p_t.$$  

Therefore, the extent of monetary non-neutrality can be measured as the persistence of $y_t$ in response to nominal spending shocks since in a flexible-price world exogenous $m_t$-shocks have no effect on real output.

**Exogenous shock processes:** As commonly assumed in the literature (e.g., see Chari, Kehoe, and McGrattan, 2000, BKM), nominal spending $m_t$ is assumed to follow an exogenous AR(1)-process in first differences:

$$\Delta m_t = \rho_m \Delta m_{t-1} + \sigma_m \epsilon^m_t,$$

where $\rho_m \geq 0$ is the measure of exogenous persistence in the model.

All other exogenous shocks follow persistent but stationary AR(1)-processes:

$$x_t = \rho_x x_{t-1} + \sigma_x \epsilon^x_t,$$

where $x_t \in \{e_t, z_t, \tilde{a}_t, \tilde{z}_{it}, \tilde{a}_{jt}\}$. All innovations ($\epsilon^m_t$ and $\epsilon^x_t$’s) are mean-zero unit-variance i.i.d. random variables.

---

24 Specifically, a model with log-utility of consumption and linear disutility of labor results in $\gamma = 1$, provided there is no additional source of aggregate real rigidities. Golosov and Lucas (2007) derive the same specification in a money-in-the-utility model. Ball and Romer (1990) refer to this benchmark as the case of strategic neutrality. Aggregate real rigidities work to reduce the value of $\gamma$. 

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29
5.1.1 Calvo price-setting

In the case of Calvo price setting, a given wholesale firm $j$ adjusts prices with probability $(1 - \theta)$ each period, while for any retail firm $i$ the adjustment probability equals $(1 - \theta_R)$. Up to a first order approximation, at the instances of adjustment the firms set their prices to the discounted expectation of their future desired prices (see Appendix for a formal derivation):

$$
s_{jt} = (1 - \beta \theta) \sum_{\ell=0}^{\infty} (\beta \theta)^\ell \mathbb{E}_t \tilde{s}_{j,t+\ell},$$
$$p_{it} = (1 - \beta \theta_R) \sum_{\ell=0}^{\infty} (\beta \theta_R)^\ell \mathbb{E}_t \tilde{p}_{i,t+\ell},$$

where $\beta$ is the discount factor, $\tilde{s}_{jt}$ and $\tilde{p}_{it}$ are the (theoretical) reset prices and $\tilde{s}_{j,t+\ell}$ and $\tilde{p}_{i,t+\ell}$ are the future desired prices (derived above) for the wholesale and retail firms respectively.

Under Calvo-pricing assumptions, the dynamics of aggregate wholesale and retail prices is given respectively by $s_t = \theta s_{t-1} + (1 - \theta)\mathbb{E}_j \tilde{s}_{jt}$ and $p_t = \theta_R p_{t-1} + (1 - \theta_R)\mathbb{E}_i \tilde{p}_{it}$, where $\mathbb{E}_j$ and $\mathbb{E}_i$ denote the cross-sectional expectations. Combining these equations with the expressions for reset prices and substituting in the expressions for desired prices, we arrive at the familiar forward-looking Phillips curves—dynamic equations for aggregate wholesale and retail inflation (see Appendix):

$$\Delta s_t = \beta \mathbb{E}_t \Delta s_{t+1} + \frac{\lambda}{1 + \Gamma} \left[ \gamma (m_t - p_t) - (s_t - p_t) + \tilde{\phi} e_t - \tilde{a}_t \right],$$
$$\Delta p_t = \beta \mathbb{E}_t \Delta p_{t+1} + \lambda_R \left[ \alpha (s_t - p_t) + (1 - \alpha) \gamma (m_t - p_t) - \tilde{z}_t \right],$$

where the expressions in square brackets are the average marginal costs of retail and wholesale firms with $w_t$ substituted in from (9). Note that all idiosyncratic shocks wash out from the aggregate price dynamic equations. The slopes of the Phillips curves equal $\lambda \equiv (1 - \beta \theta)(1-\theta)/\theta$ and analogously for $\lambda_R$. Finally, $\tilde{\phi} \equiv \int \phi_j d\mu_j$ is the sensitivity of the average wholesale marginal cost to exchange rate shock, $e_t$.

Dynamic equations (10)-(11) together with the specifications for the exogenous shock processes fully describe equilibrium dynamics in the case of Calvo pricing. The solution to this dynamic system can be obtained numerically using a conventional method.25 Finally, in the Calvo case, reset-price inflation can be measured simply as $\Delta s_t^* = (\Delta s_t - \theta \Delta s_{t-1})/(1 - \theta_R)$.

25When final-good prices are flexible ($\theta_R = 0$), this dynamic system has a simple closed-form solution which we discuss in the Appendix.
θ) and \( \Delta p_t^* = (\Delta p_t - \theta_R \Delta p_{t-1})/(1 - \theta_R) \), since the adjusting firms are selected randomly. Given the equilibrium dynamics of the aggregate variables, we can simulate firm-level prices by using the expressions for optimal reset prices provided above. Aggregating firm-level prices we arrive at the sample measures of regular and reset-price inflation, the counterparts to the empirical measures studied in Section 4.

Nominal and real rigidities in the model: We now discuss the sources of nominal and real rigidity in the model. First, nominal stickiness enters through the Calvo parameters \( \theta \) and \( \theta_R \) that reduce the slopes of the Phillips curves (\( \lambda \) and \( \lambda_R \)) and increase the persistence of inflation. Real rigidities in the form of variable markups as measured by \( \Gamma \) further reduce the slope of the wholesale inflation Phillips curve and contribute to the sluggish adjustment of wholesale prices. Furthermore, aggregate real rigidities measured inversely by \( \gamma \) slow down the pass-through of monetary shocks into the marginal costs of both types of firms and reduce the slopes of the Phillips curves. Finally, the share of intermediate inputs in the final good production costs \( \alpha \) links retail marginal costs to wholesale prices and constitutes a channel through which sluggish adjustment in wholesale prices translates into persistence in retail prices.

5.1.2 Calibration

We calibrate the model to monthly data and summarize the benchmark parameters in Table 7. We set the discount rate to 4% annually that implies a monthly discount factor \( \beta = 0.961^{1/12} \). We calibrate the money growth process and exchange rate process to the data. Specifically, we use the monthly BEA data on M2 supply to calibrate \( \rho_m = 0.5 \) and \( \sigma_m = 0.25\% \). Other papers in the literature use different numbers for the persistence of money growth. For example, Chari, Kehoe, and McGrattan (2000) use \( \rho_m = 0.571^{1/3} \approx 0.83 \), while BKM use \( \rho_m = 0 \). Therefore, for robustness we also simulate the model for \( \rho_m = 0 \) and 0.8. Next we let the exchange rate follow a very persistent AR(1) process with the standard deviation of innovation equal to \( \sigma_e = 2\% \) and autocorrelation parameter \( \rho_e = 0.995 \), consistent with the data on bilateral nominal exchange rates for developed countries.

We select the parameters for the idiosyncratic shock processes (\( \tilde{\sigma}_a, \tilde{\sigma}_z, \tilde{\rho}_a \) and \( \tilde{\rho}_z \)) to match the micro-data on price adjustment. Specifically, we set the persistence of idiosyncratic shocks to match the high autocorrelation of new prices in the BLS IPP data (0.77 for import
<table>
<thead>
<tr>
<th>Parameter</th>
<th>Symbol</th>
<th>Value</th>
<th>Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>Discount factor</td>
<td>$\beta$</td>
<td>$0.96^{1/12}$</td>
<td>Monthly data</td>
</tr>
<tr>
<td>Money growth process, $\Delta m_t$</td>
<td>volatility</td>
<td>$\sigma_m$</td>
<td>0.25%</td>
</tr>
<tr>
<td></td>
<td>persistence</td>
<td>$\rho_m$</td>
<td>0 or 0.5</td>
</tr>
<tr>
<td>Exchange rate process, $\Delta e_t$</td>
<td>volatility</td>
<td>$\sigma_e$</td>
<td>2%</td>
</tr>
<tr>
<td></td>
<td>persistence</td>
<td>$\rho_e$</td>
<td>0.995</td>
</tr>
<tr>
<td>Retail idiosyncratic shocks, $\tilde{z}_{it}$</td>
<td>volatility</td>
<td>$\tilde{\sigma}_z$</td>
<td>8%</td>
</tr>
<tr>
<td></td>
<td>persistence</td>
<td>$\tilde{\rho}_z$</td>
<td>0.90</td>
</tr>
<tr>
<td>Wholesale idiosyncratic shocks, $\tilde{a}_{jt}$</td>
<td>volatility</td>
<td>$\tilde{\sigma}_a$</td>
<td>10%</td>
</tr>
<tr>
<td></td>
<td>persistence</td>
<td>$\tilde{\rho}_a$</td>
<td>0.95</td>
</tr>
<tr>
<td>Retail aggregate shocks, $\tilde{z}_t$</td>
<td>volatility</td>
<td>$\tilde{\sigma}_z$</td>
<td>5%</td>
</tr>
<tr>
<td></td>
<td>persistence</td>
<td>$\tilde{\rho}_z$</td>
<td>0.50</td>
</tr>
<tr>
<td>Wholesale aggregate shocks, $\tilde{a}_t$</td>
<td>volatility</td>
<td>$\tilde{\sigma}_a$</td>
<td>4%</td>
</tr>
<tr>
<td></td>
<td>persistence</td>
<td>$\tilde{\rho}_a$</td>
<td>0.75</td>
</tr>
<tr>
<td>Calvo parameters</td>
<td>Retail</td>
<td>$\theta_R$</td>
<td>0.75</td>
</tr>
<tr>
<td></td>
<td>Wholesale</td>
<td>$\theta$</td>
<td>0.90</td>
</tr>
<tr>
<td>Share of intermediate inputs</td>
<td>$\alpha$</td>
<td>0.5</td>
<td></td>
</tr>
<tr>
<td>Wholesale markup elasticity</td>
<td>$\Gamma$</td>
<td>1.5</td>
<td></td>
</tr>
<tr>
<td>Aggregate real rigidities</td>
<td>$\gamma$</td>
<td>0.75</td>
<td></td>
</tr>
<tr>
<td>Sensitivity to the ER shock</td>
<td>$\phi$</td>
<td>0.225</td>
<td></td>
</tr>
</tbody>
</table>
prices) and we set the standard deviation of idiosyncratic shocks to match the absolute size of price adjustment (7.5% for import prices and 8.5% for consumer prices). This results in the standard deviation of idiosyncratic shocks equal to 10% and 8% for wholesale and retail prices respectively, while the persistence is set to 0.95 and 0.9 respectively.

Next we set the parameters for the aggregate shock processes ($\bar{\sigma}_a, \bar{\sigma}_z, \bar{\rho}_a$ and $\bar{\rho}_z$) to match the standard deviation and autocorrelation of regular and reset-price inflation series, reported previously in Table 1. This requires fairly large and transitional aggregate shocks at both the wholesale and retail levels (standard deviation of 4% and 5% and persistence of 0.75 and 0.5 respectively). These processes are a stand-in for all unmodeled shocks that hit the economy, including various economy-wide and industry-level marginal cost and markup shocks.

We set the Calvo probabilities of non-adjusting prices ($\theta$ and $\theta_R$) to match the micro-data on nominal price durations. Specifically, we choose parameters to produce 10 month durations in the wholesale sector (consistent with the evidence in Nakamura and Steinsson, 2008; Gopinath and Rigobon, 2008) and 4 month durations in the retail sector (consistent with Bils and Klenow, 2004). Next we calibrate $\gamma$, the slope of the wage equation (9) and the aggregate real rigidity parameter of the model. The literature uses a wide variety of values for $\gamma$ ranging between 0.1 in models with segmented labor markets and round-about production and 4 in models with no real rigidities and strong concavity in the utility function. We set the benchmark value for $\gamma$ to be 0.75 and for robustness we also use a greater value of 1.5, so that these parameters lie on both sides of the strategic neutrality case of $\gamma = 1$ and depart only moderately from it. We view this as a conservative choice for an aggregate parameter for which we have little direct information.

We choose the benchmark value for markup elasticity to be $\Gamma = 1.5$. This number implies a 40% pass-through of idiosyncratic shocks and is consistent with the evidence in Gopinath, Itskhoki, and Rigobon (2010) and Gopinath and Itskhoki (2010) on long-run exchange rate pass-through of about 50%. Moreover, it is consistent with the coefficients on the competitor prices reported in Table 5 of Section 4. We additionally evaluate the robustness of our results using the values of $\Gamma$ of 0 and 4, the former being the case of constant markups and the latter being the case of strong strategic complementarities at the firm level.\footnote{Our benchmark number of $\Gamma = 1.5$ is considerably smaller than the markup elasticity of 2.5 implied by the Klenow and Willis (2006) calibration.}

Finally, we set the share of intermediate inputs in the final good production, $\alpha$, to equal

\footnote{For details see Gopinath and Itskhoki (2010).}
50%, at the conservative end of the spectrum of calibrations considered in Nakamura and Steinsson (2010). The sensitivity of the aggregate marginal cost to exchange rate shocks is set to $\bar{\phi} = 0.225$, which is consistent with most domestic firms being unaffected by this shock directly, while a small fraction of importers in the industry (for example, 30%) being affected strongly by this shock (for example, $\phi_j = 0.75$).

### 5.2 Simulation results

First, we study the persistence of regular and reset-price inflation generated by our model. We compute all inflation series as sample averages of the simulated firm prices using a procedure close to the one used on the BLS data in Section 4. The results are reported in Table 8 for different values of parameters $\rho_m$, $\Gamma$ and $\gamma$. The first two columns report the results for the final-good (retail) inflation series, the next two columns provide the results for the unconditional wholesale inflation series, while the last two columns provide the results for the projection of wholesale inflation series onto lags of exchange rate changes.

<table>
<thead>
<tr>
<th>Unconditional</th>
<th>Conditional on ER</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\Delta p_t$</td>
<td>$\Delta p^*_t$</td>
</tr>
<tr>
<td>$\Gamma = 0$</td>
<td>0.41</td>
</tr>
<tr>
<td>$\Gamma = 1.5$</td>
<td>0.32</td>
</tr>
<tr>
<td>$\Gamma = 4$</td>
<td>0.37</td>
</tr>
<tr>
<td>$\rho_m = 0$</td>
<td>0.39</td>
</tr>
<tr>
<td>$\rho_m = 0.8$</td>
<td>0.48</td>
</tr>
<tr>
<td>$\gamma = 1.5$</td>
<td>0.37</td>
</tr>
</tbody>
</table>

Note: the entries are AR(1) coefficients for each series ($\Delta p$ refers to final-good inflation and $\Delta s$ refers to wholesale inflation; $^*$ indicated reset-price inflation); in the last two columns the series are the projections on the current and 24 lags of the exchange rate changes. All inflation series are sample averages of the simulated firm prices, approximating the procedure used in the data. The default parameters are the benchmark parameters from Table 7 (i.e, $\Gamma = 1.5$, $\gamma = 0.75$, $\rho_m = 0.5$).

The first pattern that emerges from Table 8 is that the aggregate consumer-price inflation

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28 For more details refer to the calibration in Gopinath and Itskhoki (2010).

29 The results in this section are preliminary: while the general patterns are robust across a variety of simulations that we ran, we are still working to put standard errors around the reported numbers.

30 Specifically, for the wholesale prices we compute the inflation series for the subsample of “foreign firms” (those which are affected directly by the exchange rate shock, i.e. which have $\phi_j > 0$) to make this exercise as close as possible to our empirical evidence in Section 4 which uses the data on import prices.
series may not be very persistent even when wholesale inflation is significantly more persistent. Secondly, wholesale inflation is significantly less persistent than the wholesale inflation projected on the exchange rate. Next we examine reset-price inflation. Both for retail and wholesale prices, reset-price inflation is negatively autocorrelated, while when projected on the exchange rate the autocorrelation becomes positive. These patterns are consistent with the empirical findings in Section 4.1. In our calibration, negative autocorrelation of the reset-price inflation series arises due to sampling error combined with transitory semi-aggregate shocks affecting wholesale and retail pricing.

The results in Table 8 are largely similar across different parameter values considered. Higher values of markup elasticity $\Gamma$ result in a more persistent inflation series, particularly when projected on the exchange rate shock, while variation in $\gamma$ and $\rho_m$ has relatively little effect on persistence of the inflation series. This is because in our calibration monetary shocks are not the key drivers of the inflation series in the short-run (at the monthly frequency), which appears to be a reasonable description of reality.\footnote{This feature appear to be consistent with the aggregate data where inflation is well approximated by an ARMA(1,1) process with both large AR and MA roots (see, for example, Stock and Watson, 2007). While monetary shocks are likely to be responsible for the AR component (long memory, low-frequency movements), there needs to be a source of relatively large transitory shocks to explain the large MA component.}

Table 9: Half-life of output in response to a monetary shock (in months)

<table>
<thead>
<tr>
<th>$\rho_m$</th>
<th>$\rho_m = 0$</th>
<th>$\rho_m = 0.5$</th>
<th>$\rho_m = 0.8$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Panel A: $\gamma = 0.75$</td>
<td>5.3</td>
<td>17.2</td>
<td>56.1</td>
</tr>
<tr>
<td>$\Gamma = 0$</td>
<td>7.0</td>
<td>23.7</td>
<td>83.0</td>
</tr>
<tr>
<td>$\Gamma = 4$</td>
<td>8.9</td>
<td>31.3</td>
<td>114.8</td>
</tr>
<tr>
<td>Panel B: $\gamma = 1.5$</td>
<td>3.6</td>
<td>11.8</td>
<td>40.0</td>
</tr>
<tr>
<td>$\Gamma = 0$</td>
<td>4.4</td>
<td>15.5</td>
<td>58.1</td>
</tr>
<tr>
<td>$\Gamma = 4$</td>
<td>5.4</td>
<td>19.8</td>
<td>80.0</td>
</tr>
</tbody>
</table>

Note: half-life is defined as $\log(0.5)/\log(\varrho_{y|m})$, where $\varrho_{y|m}$ equals the AR(1)-coefficient of output $y_t \equiv m_t - p_t$ conditional on monetary shocks $\epsilon_t^m$ (i.e., when other sources of shocks are shut down).

We next evaluate the extent of monetary non-neutrality produced by the model. Table 9 reports the half-lives of output in response to a monetary shock for different values of the parameters. Specifically, we calculate the AR(1)-coefficient of the real output series when all shocks other than monetary shocks are shut down; based on this AR(1)-coefficient we
back out a measure of half-life that we report. From Table 9 it is evident that the model can produce a wide range for the extent of non-neutrality, with half-lives of output ranging from about one quarter to over 20 quarters. However, this variation is largely driven by $\rho_m$, the exogenous persistence introduced through the autocorrelation of the money growth rate. On the other hand, variation in the amount of real rigidity, $\Gamma$ and $\gamma$, has a relatively modest effect on the extent of non-neutrality: an increase in $\Gamma$ from 0 to 4 nearly doubles the half-live, while a decrease in $\gamma$ from 1.5 to 0.75 increases the half-live by around 50%.

When we fix parameters at their benchmark values, the model produces a fairly large half-live of slightly below 8 quarters, while shutting down the variable markup channel drops the half-live to less than 6 quarters. Without exogenous persistence (i.e., $\rho_m = 0$), however, the model produces very little monetary non-neutrality (a half-life of around 1 quarter). We conclude that the empirically calibrated variable markup channel of real rigidities goes a fair way in amplifying the real effects on output, however without exogenous persistence it’s absolute effect is modest.

Note that the variation in the persistence of output deviation in the model is not very tightly linked to the persistence of inflation which does not vary much with the amount of real rigidities or the extent of exogenous persistence. Therefore, fairly long periods of monetary non-neutrality can be consistent with transitory inflation dynamics and negatively autocorrelated reset-price inflation. Again this is because monetary shocks are not the main drivers of inflation at very high frequencies.

Table 10: Exchange rate pass-through

<table>
<thead>
<tr>
<th></th>
<th>$\Gamma = 0$</th>
<th>$\Gamma = 1.5$</th>
<th>$\Gamma = 4$</th>
</tr>
</thead>
<tbody>
<tr>
<td>First adjustment</td>
<td>61%</td>
<td>44%</td>
<td>35%</td>
</tr>
<tr>
<td>Second adjustment</td>
<td>-5%</td>
<td>4%</td>
<td>7%</td>
</tr>
</tbody>
</table>

_Note: the entries are coefficients from the regression at the firm-level of price change conditional on adjustment on cumulative exchange rate change during the most recent and the previous price duration respectively (i.e., a counterpart to regression (5) of Section 4)._}

Our final results evaluate the success of the model at capturing the slow response of firm-level prices to exchange rate shocks at the micro-level, conditional on price adjustment. These results are reported in Table 10. When there are no strategic complementarities,

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32This finding is consistent with the results in Carvalho and Nechio (2008) on the persistence of real exchange rates.
the second adjustment is negative (due to some mean-reversion that we assumed in the exchange rate process). When strategic complementarities are present, pass-through at the second round of price adjustment is positive, however it is much smaller than it is in the data (see Table 2 in Section 4), where the second rounds of price adjustment are almost as large as the first. This failure of the model persists even when we assume strong strategic complementarities ($\Gamma = 4$) or shorter nominal price durations (not reported). Although the model with strategic complementarities captures incomplete pass-through in the long-run, it predicts very fast dynamics of pass-through relative to the data. This leads us to conclude that our model misses some important sources of persistence, such as rational inattention or sticky information, which may further contribute to the extent of monetary non-neutrality produced by the model. Matching the very slow adjustment of prices to aggregate shocks at the micro-level that we document in Section 4 is an important challenge that we leave for future work.

5.3 A menu cost model

In this section we briefly describe the setup and provide the simulation results of a menu cost model of price setting. The details are relegated to the Appendix and further discussion of the estimation procedure can be found in Gopinath and Itskhoki (2010). This exercise is important in order to evaluate the severity of the selection effects present in menu cost models and absent in the time-dependent pricing models.

We adopt a two-sector model (wholesale and retail as above) with three types of shocks: idiosyncratic marginal cost shocks, semi-aggregate marginal cost shocks (a stand-in for exchange rate shocks) in the wholesale sector and aggregate monetary (nominal spending) shocks. In order to maintain computational feasibility, we assume that retail prices are completely flexible. We introduce the variable markup channel using the Klenow and Willis (2006) specification of the Kimball (1995) preferences. The rest of the setup is similar to the one discussed above in Section 5.1.

With flexible prices, no strategic complementarities (e.g., CES demand) and no aggregate shocks in the retail sector, the final-good price level is given by $p_t = \bar{\mu}^R + \alpha s_t + (1 - \alpha)w_t$. The wage rate is still assumed to satisfy (9). These two equations allow one to solve for the final price level $p_t$ and the nominal wage rate $w_t$ as functions of aggregate nominal spending.
m_t and the wholesale price level s_t. 

A wholesale firm j faces a marginal cost \(mc_{jt} = w_t + \phi_j e_t - a_{jt}\), where \(a_{jt}\) is an idiosyncratic shock and \(e_t\) is the semi-aggregate shock that affects the firm with elasticity \(\phi_j\) distributed on \([0, 1]\). The firm also faces a demand schedule with elasticity \(\sigma\) and the elasticity of elasticity \(\varepsilon\), evaluated when firm’s relative price equals 1 (for details see Gopinath and Itskhoki, 2010, and the Appendix). This implies a desired markup of \(\sigma/(\sigma - 1)\) with markup elasticity equal to \(\Gamma = \varepsilon/(\sigma - 1)\). A firm maximizes its discounted present value by optimally choosing the instances of price adjustment at a menu cost \(\kappa\) and optimally resetting prices at these instances. This problem can be formalized with a standard Bellman equation (see Appendix) which we solve numerically. We then use the derived policy functions to simulate a panel of prices on which we conduct similar empirical tests to those in Section 4.

For calibration we use the same benchmark parameters as in Table 7. Additionally, we set \(\sigma = 5\) and \(\varepsilon = 6\) so that the level of wholesale markup is 25% and the elasticity of markup is equal as before to \(\Gamma = 1.5\). The menu cost is chosen to match the duration of 10 months of wholesale prices that implies a \(\kappa = 3.5\%\) of the steady state revenue conditional on adjustment (equivalently 0.35% of annual revenues), a number consistent with the literature. Finally, in order to match the absolute size of price adjustment of 7.5%, we set the standard deviation of the idiosyncratic shocks to \(\sigma_a = 6\%\).

To keep this section brief, we report the results only for the benchmark values of parameters, \(\Gamma = 1.5\), \(\gamma = 0.75\) and \(\rho_m = 0.5\). The unconditional autocorrelation of the wholesale-price inflation is 0.29, and it is 0.38 conditional on the exchange rate shock, both numbers being substantially smaller than in the case of the Calvo model (see Table 8). The corresponding reset-price inflation series is strongly negatively autocorrelated—with autocorrelation of \(-0.89\) unconditionally and \(-0.66\) conditional on the exchange rate shock—emphasizing the powerful selection effects of the menu cost models. This negative autocorrelation of reset-price inflation conditional on the exchange rate shock goes against our empirical findings in Section 4.

Similarly, in the type of micro-level pass-through regressions run in Section 4.2, the menu cost model generates a pass-through coefficient conditional on the first round of price adjustment equal to 55%, while the coefficient for the second round of price adjustment is \(-8\%\). This is also due to the strong selection effect of the menu cost models which dominated the persistence introduced through strategic complementarities in pricing. Recall that in the
Calvo model these two pass-through coefficients were 44% and 4% respectively (see Table 10), and this is in contrast to our empirical findings that pass-through at the second adjustment is nearly as large as the first one. Finally, the half-life of output in response to monetary shocks is 15.1 months in the menu cost model as opposed to 23.7 months in the Calvo model.\footnote{The approach to computing the conditional half-life of output in the menu cost model is different: Since we cannot simply shut down other sources of shocks in the menu cost model (as its dynamics is non-linear in shocks), we estimate econometrically the impulse response of output $y_t$ to current and lagged monetary shocks $\varepsilon_{t-j}$ and then use it to compute the projection of output series onto these shocks. The reported number is based on the AR(1)-coefficient for the projected series.} This illustrates a well known fact that the selection effect of the menu cost models reduces substantially the contract multiplier relative to the time-dependent models of price setting.

**Summary:** A number of insights come out of our simulation exercise. First, transitory aggregate inflation series are consistent with persistent response of prices to certain aggregate shocks, including exchange rate and monetary shocks. Second, properties of the aggregate inflation series may be largely disconnected from the size of the contract multiplier for monetary shocks. Third, quantitatively exogenous persistence appears to be substantially more important than real rigidities in generating long half-lives and large contract multipliers. Although strategic complementarities work to increase the size of the contract multiplier, their effects are modest unless coupled with exogenous sources of persistence. Fourth, the analyzed models (in particular the menu cost model, but also the Calvo model) cannot match very sluggish responses of prices to shocks at the micro-level conditional on adjusting prices. This suggests a need for additional sources of persistence lacking from the model.

### 6 Bargaining Model of Price Setting

We now consider a setup in which final good producers bargain with a number of intermediate good suppliers. This is a more realistically characterization of business-to-business transactions. We show that in a static bargaining model there are strategic complementarities in wholesale price setting so that markups are variable and wholesale prices exhibit incomplete pass-through. On the other hand, final good producers compete monopolistically and subject to a CES consumer demand charge constant markups. This provides a microfoundation for the reduced-form assumptions on wholesale and retail markups imposed in the
dynamic model of Section 5. We leave the extension of this bargaining model to a dynamic price-setting setup for further research.

Although it is quite natural to think that intermediate good prices are set via bargaining, we are unaware of any macroeconomic models of intermediate-good price-setting via bargaining. In a recent paper, Goldberg and Tille (2009) propose a bargaining model of currency choice in international transactions. Our model here is most-closely related to Atkeson and Burstein (2008), which is a special case of our model when the intermediate-good suppliers have full bargaining power and hence act as price setters.\footnote{We note that Atkeson and Burstein (2008) choose not to interpret their two-tier demand structure as a sequence of the wholesale and retail sectors, but this is an equally coherent interpretation.}

Consider a final good producer $i$. In what follows, we omit the final good producer’s identifier $i$ where it leads to no confusion. The final good producer uses intermediate inputs to assemble the final good according to a CES technology

$$y = \left[ \sum_{j=1}^{N} q_j^\eta \right]^{1/\eta},$$

where $\eta \in (0,1)$ controls the elasticity of substitution between intermediate varieties, $N$ is the number of intermediate varieties used for assembly and $q_j$ is the input quantity of intermediate variety $j$.\footnote{Note that in this section, as opposed to the rest of the paper, lower case letters denote the levels of variables rather than logs.} Note that labor is not used for production of the final good, which is produced using intermediates only. This corresponds to the special case of $\alpha = 1$ in the terminology of Sections 2 and 5, and we adopt it for simplicity.

The revenue of the final good producer is given by

$$R = py = Ay^\zeta,$$

where $p$ is the final good price, $A$ is a demand shifter and $\zeta \in (0,1)$ is a parameter that controls the elasticity of demand. This revenue function specification arises from CES preferences over the final good with the elasticity of substitution equal to $1/(1 - \zeta)$. Finally, the intermediate good producer $j$ has a constant marginal cost of $c_j$. Therefore, the total
surplus to be shared between the final and intermediate good producers is

\[ Ay^\zeta - \sum_{j=1}^{N} c_j q_j, \quad (13) \]

where \( y \) is given in (12).

The surplus in (13) is divided according to bilateral Nash bargaining between the final good producer and each of the intermediate good suppliers. Specifically, we assume that prices are determined through bargaining, while given prices the final good producer is free to choose any quantity of the intermediate good supply.\(^\text{36}\)

Formally, denote by \( s_j \) the price of intermediate goods determined via bargaining. The final good supplier will maximize his revenues minus cost given by \( \sum_{j=1}^{N} s_j q_j \) when choosing quantities. As a result, quantities lie on the demand curve given by

\[ \zeta Ay^{\zeta-n} q_j^{n-1} = s_j. \quad (14) \]

This implies (upon aggregation over \( j \)) that the final good’s price is a constant markup over the cost index of the intermediate goods:

\[ p = Ay^{\zeta-1} = \frac{s}{\zeta}, \quad \text{where} \quad s \equiv \left[ \sum_{j=1}^{N} s_j - \frac{1}{n} \right]^{-\frac{1}{n}}. \]

Therefore, we arrive at the first assumption of Section 5 that retail prices are set as a constant markup over marginal cost (i.e., \( \Gamma_R = 0 \)). Note that the final good quantities and prices respond to the intermediate good prices, \( s_j \), which play an allocative role in this bargaining model.

The Nash bargaining between the final good producer and supplier \( j \) determines the price \( s_j \). We assume that the bargaining power of the final good producer is \( \phi \in (0, 1) \) and we denote his relative bargaining power by \( \lambda \equiv \phi/(1 - \phi) \). We do not provide here a micro-foundations for the source of variation in the bargaining power. It may come from the

\(^{36}\)If the parties could also bargain over quantities, the bargaining game would result in efficient supply of the intermediate goods with prices playing a role of transfers without any allocative role. However, if for example the value of the demand shifter \( A \) were unknown at the bargaining stage, it could be optimal to set prices without restricting quantities and let the quantities adjust \( \text{ex post} \) in response to the movements in \( A \). Empirically, intermediate-goods prices appear to play an allocative role since changes in intermediate-good prices get fully passed-through into retail consumer prices (see Section 3).
differential patience of the final good producer and suppliers that in turn may be related to the extent of liquidity constraints that different firms face or from the tightness of the supplier market (i.e., how easy it is to replace a given supplier).

If bargaining breaks down, the supplier receives zero, while the final good producer has to assemble the final good without the input of this supplier. Therefore, the surplus of the supplier is $s_j - c_j q_j$. From (14) it follows that the profit of the final good producer equals $(1 - \zeta) A y^\zeta$, where $y = (\zeta A / s)^{1/(1 - \zeta)}$ is the optimal output of the final good given intermediate price index $s$. If bargaining with supplier $j$ breaks down, the cost index of the intermediates becomes

$$s_{-j} = \left[ \sum_{k \neq j} s_k^{-\eta/(1-\eta)} \right]^{-\frac{1-\eta}{\eta}} > s.$$

Consequently, the optimal output of the final good will be $y_{-j} = \left( \zeta A / s_{-j} \right)^{1/(1 - \zeta)} < y$ resulting in the profit of the final good producer of $(1 - \zeta) A y_{-j}^\zeta$. Under these circumstances, the incremental surplus of the final good producer from supplier $j$ is given by

$$(1 - \zeta) A [y^\zeta - y_{-j}^\zeta] = (1 - \zeta) \zeta^{\frac{1}{1-\zeta}} A^{\frac{1}{1-\zeta}} [s^\zeta - s_{-j}^\zeta].$$

Under these circumstances, we can write the Nash bargaining problem formally as:

$$\max_{s_j} \left\{ \left[ (s_j - c_j) q_j \right]^{1-\phi} \left[ (1 - \zeta) \zeta A^{\frac{1}{1-\zeta}} (s^\zeta - s_{-j}^\zeta) \right]^{\phi} \right\},$$

where $q_j$ is subject to (14) and $s$ and $s_{-j}$ are the cost indexes defined above. Taking the first order condition, one can demonstrate that the bargained price needs to satisfy the following condition:

$$\frac{s_j}{s_j - c_j} = \frac{1}{1 - \eta} + \frac{\zeta}{1 - \zeta} \theta_j \left[ \frac{\lambda}{1 - (1 - \theta_j) \chi} - \eta - \zeta \right],$$

where $\chi \equiv \frac{\zeta}{1 - \zeta} \frac{1 - \eta}{\eta}$ and $\theta_j$ is the cost (market) share of supplier $j$ equal to

$$\theta_j \equiv \frac{s_j q_j}{s y} = \left( \frac{s_j}{s} \right)^{-\frac{\eta}{1-\eta}}.$$

Note that (15) implies a markup pricing rule in which the markup depends on the market share $\theta_j$, the relative bargaining power $\lambda$ and the parameters of the model $\zeta$ and $\eta$. Moreover, since prices affect the market shares of firms, in general markups are not constant and there is strategic non-neutrality in pricing (i.e., $\Gamma \neq 0$ in the terminology of Section 2).
To keep this section brief, we state here only the main results on the properties of the bargained prices which follow directly from (15):

1. When $\theta_j = 0$ (infinitesimal supplier), the markup is constant and given by $\mu_j \equiv s_j/c_j = \phi + (1 - \phi)/\eta$, which is a convex combination of 1 and $1/\eta$ weighted by the bargaining power.

2. When all bargaining power is with the final good producer (i.e., $\phi = 1$ or $\lambda = \infty$), bargaining results in marginal-cost pricing $s_j = c_j$.

3. In the opposite case, when full bargaining power is with intermediate goods producers ($\phi = \lambda = 0$), markup depends on the market share:

$$\frac{\mu_j}{\mu_j - 1} \equiv \frac{s_j}{s_j - c_j} = \frac{1}{1 - \eta} - \theta_j \frac{1}{1 - \eta} \frac{\eta - \zeta}{1 - \zeta}.$$ 

Note that when $\eta > \zeta$ (the baseline case), the markup is increasing in the market share and hence decreasing in the relative price of the firm. This corresponds to the case $\Gamma > 0$.

4. In general, when $\theta_j \in (0, 1)$ and $\lambda < \infty$, the markup is variable and pass-through is incomplete. Moreover, pass-through is not necessarily monotonic in the market share.

To summarize, this bargaining model results in variable-markup pricing at the wholesale level and constant-markup pricing at the retail level, with wholesale markups depending on the market shares of the suppliers, their relative bargaining power, as well as parameters of the model. Therefore, the model appears to be consistent with the broad features of the data discussed in Sections 3–4. We leave to future research testing the qualitative success of this model in capturing the dynamics of wholesale and retail prices.

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37In fact, this case exactly corresponds to the Atkeson and Burstein (2008) model since $1/(1 - \eta)$ equals the elasticity of substitution between intermediate varieties and $1/(1 - \zeta)$ equals the elasticity of substitution between final good varieties.
APPENDIX

A Calvo Price Setting

In this section we derive a general log-linear approximation for the price setting equation and aggregate inflation dynamics (Phillips curve) in a Calvo model. Since these derivations are well-known, we keep the exposition brief.

Consider a firm $j$ with a real profit function $\Pi^j(x_j|S)$, where $x_j$ is the firm’s log-price and $S$ is the state of the economy. The desired price of the firm is $\tilde{x}_j(S) \equiv \arg \max_{x_j} \Pi^j(x_j|S)$ with the necessary condition $\Pi^j_j(\tilde{x}_j(S)|S) = 0$, where the subscript denotes a partial derivative. We assume that the marginal cost of the firm does not depend on the price of the firm, i.e. a firm faces a constant returns to scale in production where productivity depends on the state of the world. Then we can decompose the desired price as

$$\tilde{x}_j(S) = \mu_j(\tilde{x}_j - x, S) + mc_j(S),$$

where $mc_j(S)$ is the log nominal marginal cost of the firm and $\mu_j(x_j-x, S)$ is the log (desired) markup which we allow to depend on the relative price of the firm, with $x$ denoting the log of the relevant price index.

A general first order approximation to the markup can be written as

$$\mu_j(x_j, S) \approx \bar{\mu} - \Gamma(x_j - x) + \epsilon_j(S),$$

where $\bar{\mu}$ and $\Gamma$ are some constants (assumed to be common across all firms at the point of approximation) and $\epsilon_j(S)$ is some linear function of the state $S$. It is natural to assume that $\epsilon_j(S)$ is stationary, while $mc_j(S)$ is co-integrated with the nominal variables of the model. With this approximation, we can solve explicitly for the desired price of the firm:

$$\tilde{x}_j(S) = \frac{1}{1+\Gamma} [\bar{\mu} + mc_j(S) + \epsilon_j(S)] + \frac{\Gamma}{1+\Gamma} x.$$  

\[38\]Since we look at the real profit function, that is a profit function normalized by the price level in the economy, it is without loss of generality to assume that the nominal variables enter the sufficient state vector $S$ only normalized by the price level. Therefore, we can treat $S$ as having a stationary distribution even though monetary variables may be trending. For an example see Appendix C.
A given firm sets prices to maximize its discounted expected value. In a Calvo pricing environment the firm may adjust its price in every period with exogenous probability \((1 - \theta)\). Therefore, we can write the problem of the firm recursively as:

\[
\bar{x}_{jt}(S) = \arg \max_{x_j} \mathbb{E} \left\{ \Pi^j(x_j|S_t) + \sum_{\ell=1}^{\infty} Q_{t,t+\ell}(S) \theta^{\ell-1} \left[ \theta \Pi^j(x_j|S_{t+\ell}) + (1 - \theta) \Pi^j(\bar{x}_{j,t+\ell}(S)|S_{t+\ell}) \right] S_t \right\},
\]

where \(Q_{t,t+\ell}(S)\) is the stochastic discount factor for real variables, \(S_t \equiv (S_0, \ldots, S_t)\) is the history of the states and \(S \equiv (S_t, S_{t+1}, \ldots)\). The first order condition for the price setting can be written as:

\[
\sum_{\ell=0}^{\infty} \theta^\ell \mathbb{E} \left\{ Q_{t,t+\ell} \Pi^j(\bar{x}_{jt}|S_{t+\ell}) \right\} S_t = 0,
\]

where we omit the explicit dependence on \(S\). Taking a first order approximation of this optimality condition around a non-stochastic steady state with zero inflation, we obtain

\[
\sum_{\ell=0}^{\infty} (\beta \theta)^\ell \mathbb{E} \left\{ \bar{x}_{jt} - \bar{x}_j(S_{t+\ell}) \right\} S_t = 0,
\]

where \(\beta^\ell\) is the non-stochastic steady state value of \(Q_{t,t+\ell}\). The price setting formulas in Section 5.1 are direct implications of this linearized optimality condition. Now using the expression for the desired price, we have:

\[
\bar{x}_{jt} = (1 - \beta \theta) \sum_{\ell=0}^{\infty} (\beta \theta)^\ell \mathbb{E}_t \left\{ \frac{1}{1 + \Gamma} \left( mc_{j,t+\ell} + \epsilon_{j,t+\ell} \right) + \frac{\Gamma}{1 + \Gamma} x_{t+\ell} \right\},
\]

where we switched notation for conditional expectation, suppressed the explicit dependence on the state of the economy and omitted the constant by implicitly relabeling the variables to denote the deviations from the non-stochastic steady state.

Finally, since the nominal marginal cost is possibly integrated, we need to scale this expression by some monetary variably co-integrated with the marginal cost. A natural candidate is the competitor price index \(x_t\) or a sector price level (in a number of models, including the Kimball demand model, the two variables coincide). We therefore define \(x_t = \int_j x_{jt} dj\). With some manipulation, we rewrite the deflated price-setting equation as:

\[
\bar{x}_{jt} - x_{t-1} = \frac{1 - \beta \theta}{1 + \Gamma} \sum_{\ell=0}^{\infty} (\beta \theta)^\ell \mathbb{E}_t \left\{ mc_{j,t+\ell} - x_{t+\ell} + \epsilon_{j,t+\ell} \right\} + \sum_{\ell=0}^{\infty} (\beta \theta)^\ell \mathbb{E}_t \Delta x_{t+\ell},
\]

45
or in a recursive form:

\[
(\bar{x}_{jt} - x_{t-1}) - \beta \theta \mathbb{E}_t (\bar{x}_{jt+1} - x_t) = \frac{1 - \beta \theta}{1 + \Gamma} \left( mc_{jt} - x_t + \epsilon_{jt} \right) + \Delta x_t.
\]

Next, with Calvo pricing, the dynamics of the aggregate price level can be written as

\[
x_t = \theta x_{t-1} + (1 - \theta) \mathbb{E}_j \bar{x}_{jt} \quad \Rightarrow \quad \Delta x_t = (1 - \theta) \mathbb{E}_j \{ \bar{x}_{jt} - x_{t-1} \},
\]

where \( \mathbb{E}_j \) is the cross-sectional average across all firms. Combining the above two equations and rearranging, we arrive at the traditional New-Keynesian Phillips curve:

\[
\Delta x_t - \beta \mathbb{E}_t \Delta x_{t+1} = \frac{\lambda}{1 + \Gamma} \mathbb{E}_j \{ mc_{jt} - x_t + \epsilon_{jt} \}, \quad \lambda \equiv \frac{(1 - \beta \theta)(1 - \theta)}{\theta}.
\]

Equations (10)-(11) in the text are special cases of this Phillips curve with the expressions for marginal costs substituted in (note that the cross-sectional expectation averages out all purely idiosyncratic shocks).

**B Aggregate Dynamics under Calvo Pricing**

The aggregate dynamic system contains three equations — the two Phillips curves for the wholesale and retail prices and the aggregate wage equation — for three variables \((s_t, p_t, w_t)\):

\[
\Delta s_t = \beta \mathbb{E}_t \Delta s_{t+1} + \frac{\lambda}{1 + \Gamma} \left\{ w_t - s_t + \bar{e}_t - \bar{a}_t \right\},
\]

\[
\Delta p_t = \beta \mathbb{E}_t \Delta p_{t+1} + \lambda_R \left\{ \alpha s_t + (1 - \alpha) w_t - p_t - \bar{z}_t \right\},
\]

\[
w_t = \gamma m_t + (1 - \gamma) p_t,
\]

where \( \Delta m_t, e_t, z_t \) and \( a_t \) follow exogenous stationary processes. This system can be solved using conventional method, which results in the expressions for the endogenous variables \((s_t, p_t, w_t)\) as functions of the shocks to the exogenous variables. This solution allows to study the statistical properties of the endogenous variable time series, including their volatility and persistence.

When the retail prices are set flexibly (i.e., \( \theta_R = 0 \) or \( \lambda_R = \infty \)), there exists a tractable

\[\text{This is the step which requires stationarity of the right-hand side variables.}\]
analytical solution for the aggregate dynamics. We discuss it briefly here. In this case, the expression for the consumer price level becomes static:

\[ p_t = \alpha s_t + (1 - \alpha)w_t - \bar{z}_t. \]

Together with the wage equation it allows to solve for \( p_t \) and \( w_t \) as linear functions of \( m_t, s_t \) and \( \bar{z}_t \). Substituting these expressions into the wholesale-price Phillips curve, we obtain a second-order difference equation in \( s_t - m_t \):\(^{40}\)

\[ \Delta s_t - \beta E_t \Delta s_{t+1} = \kappa (s_t - m_t) + \xi_t, \]

where

\[ \xi_t \equiv \frac{\lambda}{1 + \Gamma} (\frac{\phi e_t - \bar{a}_t}{1 + \Gamma} - \frac{1 - \gamma}{\alpha(1 - \gamma) + \gamma} \bar{z}_t \]

is the summary measure of all shocks other than \( m_t \) and \( \kappa \equiv \frac{\lambda}{1 + \Gamma} \frac{1 - \gamma}{\alpha(1 - \gamma) + \gamma} \) is the summary measure of nominal and real rigidities in the model. This difference equation can be solved forward. Assuming for simplicity that \( \xi_t \) follows an AR(1), the process for \( s_t - m_t \) is an ARMA(3,1). Therefore, \( s_t \) is co-integrated with \( m_t \) and movements in \( m_t \) dominate the low-frequency movements in \( s_t \), however, the short-run dynamics of \( s_t \) (around slow moving \( m_t \)) may be dominated by transitory shocks to \( \xi_t \). In particular, the MA-component may reduce significantly the short-run persistence, while it does not affect the long-run persistence. This logic is consistent with the empirical findings of Stock and Watson (2007).\(^{41}\) Finally, one can show that one of the AR-roots is given by \( \rho_m \), while the other root is decreasing in \( \kappa \) and converging to 1 as \( \kappa \to 0 \). Furthermore, one can show that these roots also drive the persistence of the output response to monetary shocks.

C Menu Cost Model

**Kimball demand:** To simulate the menu cost model we first need to specify the explicit source of variable markups. We generate variable markups by introducing the Klenow and Willis (2006) specification of the Kimball (1995) demand. The demand function for firm \( j \)

\(^{40}\)This is under the assumption of a unit root in \( m_t \); otherwise, the difference equation is second order in \( s_t \).

\(^{41}\)If \( \xi_t \) is absent from the model, the process for \( s_t - m_t \) is an AR(2) and therefore \( s_t \) exhibits both high short-run and long-run persistence.
in this case is given by:

\[ \psi(s_{jt} - s_t) = \left[ 1 - \varepsilon(s_{jt} - s_t) \right]^{\sigma/\varepsilon}, \quad \sigma \geq 1, \quad \varepsilon > 0, \]

where \( s_{jt} \) is the log price of the firm and \( s_t = \int s_{jt} \, dj \) is the sectoral log price index. In Gopinath and Itskhoki (2010) we show that this price index is a valid second order approximation to the ideal price index with this demand system. The price elasticity of this demand is given by

\[ \tilde{\sigma} = \frac{\sigma}{1 - \varepsilon(s_{jt} - s_t)} \]

which equals \( \sigma \) when the relative price of the firm is 1. With this demand, the desired price is equal to a markup \( \tilde{\sigma}/(\tilde{\sigma} - 1) \) over the marginal cost. The elasticity of the markup is given by

\[ \tilde{\Gamma} = \frac{\tilde{\varepsilon}}{\tilde{\sigma} - 1}, \]

where

\[ \tilde{\varepsilon} \equiv \frac{\varepsilon}{1 - \varepsilon(s_{jt} - s_t)} \]

is the elasticity of the elasticity of demand. Therefore, the markup elasticity evaluated at the relative price of 1 is given by \( \Gamma = \varepsilon/(\sigma - 1) > 0 \).

**Problem of the firm:** The real profit of the firm is given by

\[ \Pi(s_{jt}|S_{jt}) = \psi(s_{jt} - s_t) \left[ \exp\{s_{jt} - p_t\} - \exp\{mc_{jt} - p_t\} \right], \]

where \( S_{jt} \) is the state vector for the firm and \( mc_{jt} \) is the log nominal marginal cost of the firm and \( p_t \) is the final-good price level. As discussed in the text, the marginal cost of the firm equals

\[ mc_{jt} = w_t + \phi_j e_t - a_{jt}. \]

Therefore, the state vector for the firm includes \((p_t, s_t, w_t, e_t, a_{jt})\).

We can write the firm’s problem recursively as:

\[
\begin{align*}
V^N(S_{jt}) &= \Pi(s_{jt-1}|S_{jt}) + \mathbb{E} \{ Q(S_{j,t+1})V(S_{j,t+1}|S_{jt}) \}, \\
V^A(S_{jt}) &= \max_{\bar{s}_{jt}} \left\{ \Pi(s_{jt}|S_{jt}) + \mathbb{E} \{ Q(S_{j,t+1})V(S_{j,t+1}|S_{jt}) \} \right\}, \\
V(S_{jt}) &= \max \left\{ V^N(S_{jt}), V^A(S_{jt}) - \kappa \right\},
\end{align*}
\]
where $V$ is the value of the firm, $V^N$ is the value of the firm if it does not adjust its price and $V^A$ is the value of the firm if it adjusts its price; $\kappa$ is the menu cost, $Q$ is the stochastic discount factor for real variables (which we set to equal $\beta$ in the simulation), and $S_{jt}$ includes in addition the previous price of the firm $s_{j,t-1}$.

**General equilibrium:** We assume flexible prices for the final good and no aggregate productivity shocks in the final-good sector. This implies that $p_t = \bar{\mu}_R + \alpha s_t + (1 - \alpha)w_t$, where $\bar{\mu}_R$ is the constant markup in the final-good sector. In turn, the wage is given by $w_t = \gamma m_t + (1 - \gamma)p_t$. This allows us to solve for $p_t$ and $w_t$ as a function of $m_t$ and $s_t$ and reduce the aggregate state space to $(s_t, m_t, e_t)$. The state vector for an individual wholesale firm additionally includes $(s_{j,t-1}, a_{jt})$.

In the general equilibrium of the model, firms optimally decide to adjust prices given their current state vector and rational expectations about the evolution of the state vector, while aggregated individual firm pricing decisions are consistent with the aggregate dynamics of the wholesale price level $s_t$.

**Simulation procedure:** We iterate the Bellman equation for the firm pricing problem on the grid given a forecasting rule for the evolution of the state vector.\textsuperscript{42} This produces a policy function for firm pricing decisions, which allows us to simulate a panel of firm prices. In each period of the simulation we make sure that the wholesale price index is consistent with the firm pricing decisions (which constitutes a static fixed point problem). As a result, we obtain a time series for the equilibrium wholesale price level. Given this time series, we update the forecasting equation. We iterate this procedure until the forecasting equation converges. With the equilibrium forecasting rule we simulate a panel of firm prices and we use it to estimate various statistical moments as in Section 4. Additional details of the simulation procedure can be found in Gopinath and Itskhoki (2010).

\textsuperscript{42}To ensure stationarity of the grids, we normalize all nominal variables in the model by $m_{t-1}$. 

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References


