

Import Exchange Rate Pass-Through and Credit Constraints: Evidence from China

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Abstract

Exchange rate fluctuations are a key factor affecting international trade prices. This paper shows the incomplete exchange rate pass-through patterns in China and its linkage with importers' credit constraints. Using Chinese firm-level information and customs transaction records from 2000 to 2007, we find that (1) the average import price pass-through in China is about 35% to 40%, far below the near complete 95% export price pass-through; (2) for firms in financially more vulnerable industries, both import and export exchange rate pass-through tend to be more complete; (3) higher import source diversity of a firm can effectively reduce the import price pass-through and offset the effects of credit constraints.

(*JEL* F14, F31, F41)

1 Introduction

Why do exchange rate fluctuations not result in price changes of the same magnitude? This is one of the core questions among a set of “exchange rate disconnect” puzzles ([Obstfeld and Rogoff \(2000\)](#)). As price signals in the international trade market, exchange rates appear less informative for firms than expected. Exchange rate pass-through (ERPT), which describes the elasticity of local price changes to exchange rate fluctuations, varies widely across countries, industries, and time. Existing studies have generated widely

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varying estimates of the exchange rate pass-through. It is of particular interest to researchers both in the fields of international trade and open macroeconomics to demystify the “exchange rate disconnect” puzzles. Understanding the pattern of exchange rate pass-through has important implications for formulating macro policy, including monetary policy, inflation targeting, and the balance of payments.

In this paper, we focus on the role of importers in the determination of exchange rate pass-through and connect it with the importers’ financial constraints. This paper tends to fill a gap in the literature by linking both sides of the trade relationship and provides a novel perspective to study the nature of exchange rate disconnect for emerging markets, where firms are more vulnerable to credit constraints due to immature financial markets. In contrast to the conventional framework of exchange rate pass-through where importers are mostly price takers, we contribute to the literature by identifying the importers’ implicit sourcing power and comparing their heterogeneous capacity to absorb exchange rate shocks. Throughout the paper, we will compare firm-level import exchange rate pass-through with export pass-through to reflect the similarities and differences between the two. Going a step further, if a country’s export and import exchange rate pass-through patterns differ significantly, this may even affect terms of trade as well as current account imbalances. We believe that micro-evidence from import exchange rate pass-through provides a new perspective to study the “exchange rate disconnect” puzzle.

We estimate the exchange rate pass-through as the price elasticity of import prices concerning real exchange rates using Chinese micro-level data. Specifically, we merge the Chinese Industrial Enterprises datasets with the transaction data from China’s customs and adopted fixed effects panel regressions with first-order differences to capture the changes in product prices and real exchange rates. The average import exchange rate pass-through is between 35%-40%, which is obviously less complete compared to the over 95% export pass-through. Second, we identify the effects of credit constraints on importers’ exchange rate pass-through. We use US measures of sectors’ financial vulnerability (Manova, Wei and Zhang (2015)) in our main empirical analysis while use Chinese measures of credit needs (Fan, Li and Yeaple (2015)) for robustness checks. In our baseline results, import prices for firms in sectors with higher financial constraints are more sensitive to exchange rate shocks. Third, we calculate the proxy for an importer’s sourcing base and find that an importer with more alternative importing options can resist the effects of credit constraints and absorb exchange rate shocks better.

In the discussion part, we control several potential factors which may affect the import pass-through other than credit constraints. We first estimate the firm-level markup estimated following De Loecker and Warzynski (2012) and test how markups affect import ERPT. In addition, firms with different market shares also demonstrate heterogeneous price responses to exchange rate changes. We confirm that although firm heterogeneity in those aspects does affect exchange rate pass-through, credit constraints still play a role

that cannot be ignored even after we control those firm characteristics. In robustness checks, we further use alternative measures of credit constraints and alternative subsamples. For alternative measures, we compare sector-level credit constraints variables calculated from China data with those from US data. For alternative samples, we divide our subjects into two subsets: two-way traders (simultaneous import and export) and one-way traders (either only import or only export). We check our results using only the two-way traders who account for the vast majority of trading firms and the vast majority of China’s trade volume. These results are all significant and robust.

We show that importers’ credit constraints do influence price-setting patterns in international trade. We provide evidence for three key findings: (1) the average import exchange rate pass-through level in China is significantly less complete than the export one; (2) financial constraints will increase both of them to be more complete, and (3) importers who import a certain product from more sources have a less complete pass-through. In other words, financially constrained importers will absorb more price fluctuations caused by exchange rate changes, while financially constrained exporters pass through more exchange rate changes to prices, both compared to those unconstrained firms. This reflects that binding financial constraints will lead to not only narrow margins to adopt pricing-to-market strategies for the sellers but also limited sourcing power for the buyers. Importers with a wider sourcing base could get more access to alternative options and avoid bargaining disadvantages in exchange rate fluctuations to some extent.

This paper contributes to three strands of literature. First, this paper contributes to a wide literature on exchange rate disconnect ([Obstfeld and Rogoff \(2000\)](#)), particularly on trade pricing response to exchange rate changes ([Campa and Goldberg \(2005\)](#)). Trade prices denominated in the destination currency are usually much less volatile than exchange rates. Recently, more studies link the exchange rate pass-through to disaggregated firm-level characteristics. [Berman, Martin and Mayer \(2012\)](#) find that exporters with better performance react to an exchange rate shock by adjusting more their markup and less their export volume. [Amiti, Itskhoki and Konings \(2014\)](#) find that firms with higher import intensity and larger market share have less complete price pass-through. Many follow-up works find that exporters’ productivity ([Li, Ma and Xu \(2015\)](#)) and product quality ([Chen and Juvenal \(2016\)](#); [Auer, Chaney and Sauré \(2018\)](#)), as well as their imported inputs ([Wang and Yu \(2021\)](#)), market shares ([Auer and Schoenle \(2016\)](#); [Devereux, Dong and Tomlin \(2017\)](#)) and information access ([Garetto \(2016\)](#)), will all affect the export exchange rate pass-through. For Chinese exporters, [Li, Ma and Xu \(2015\)](#) find nearly complete exchange rate pass-through into RMB price for Chinese exporters, which means that their domestic currency price response is very weak. Yet the role of importers in determining exchange rate pass-through remains a novel field to study. The contribution of this paper is to provide micro-level evidence of importers’ heterogeneity in response to exchange rate shocks when firms are operating in industries with varying

degrees of credit constraints. This supports the idea that international trade pricing is influenced by the characteristics of both importers and exporters. We believe that the bargaining power of importers, an important yet underexplored force that determines pricing patterns in the international market, may help explain the exchange rate disconnect puzzle.

Second, we contribute to the literature by analyzing how credit constraints affect the exchange rate elasticity to international prices. We study the influence of sector-level credit constraints on the response of import prices to exchange rate fluctuations. The most relevant study to this paper is [Strasser \(2013\)](#), which discusses the relationship between exchange rate pass-through and exporters' credit conditions. [Strasser \(2013\)](#) uses a firm-level survey to show that financially constrained exporters pass exchange rate shocks to prices more completely. Subject to variable markups, firms under borrowing constraints must keep the pricing-to-market (PTM) strategy to a limited extent because they have less space to adjust their profit margin. One recent article [Dai et al. \(2021\)](#) finds that exchange rate pass-through to export prices denominated in home currency is lower for firms under tighter financial constraints. However, they focus on firm-level aggregate export behaviors while we assess the bilateral trade elasticity with each country instead. Our paper differs from earlier studies in that, first, we focus on importers, whose pricing behaviors are also affected by credit constraints, and second, with the detailed Chinese panel data, we could estimate the relationship between credit constraints and price response to exchange rate shocks at firm-source-product-year level controlling for various associated variables and their interactions. In addition, we provide a novel perspective to study the exchange rate disconnect puzzle in emerging markets, where firms are usually financially more vulnerable due to immature financial markets.

Third, this paper relates to the broader literature on credit constraints and international trade. The cross-border transportation of exporting or importing is riskier than domestic sales or purchasing, and contractual reliability in international transactions is weaker. Therefore, it is widely believed that trade participants rely on additional external capital to pay fixed costs to source from or export to other countries, which can not be fully covered by internal cash flows from operations ([Manova \(2013\)](#), [Chaney \(2016\)](#)). [Feenstra, Li and Yu \(2014\)](#), [Manova, Wei and Zhang \(2015\)](#), [Fan, Lai and Li \(2015\)](#) provide comprehensive theoretical explanations and empirical evidence from China about how credit constraints affect exports, through incomplete information, multinational links, and quality, respectively. Recently, [Xu and Guo \(2021\)](#) demonstrate that the estimated exchange rate elasticity of export volume is negatively correlated with the sector's degree of financial constraints. However, controlling markup adjustment could not fully eliminate the effect of financial constraints, which leaves room for further research. In the context of exchange rate shocks, our study deliberately examines how credit constraints affect the robustness of importers' global sourcing behavior. This provides

a novel perspective to study the exchange rate disconnect puzzle in emerging markets, where firms are usually financially more vulnerable due to immature financial markets.

The remainder of the paper is organized as follows. ?? describes the data we used. 3 introduces our empirical strategy and measures of key variables. Section ?? shows the main empirical results about import exchange rate pass-through and credit constraints. Section 6 discusses other factors affecting import exchange rate pass-through and directions for future research. Section 7 examines the robustness of the results. Section 7 concludes.

2 Data and Measurement

We conduct our empirical research using various data sources: (1) detailed transaction data provided by China’s General Administration of Customs, (2) the annual surveys from Chinese Industrial Enterprises provided by the National Bureau of Statistics of China, and (3) country-level macro data from Penn World Table 10.0. In this section, we will introduce the basic information of these datasets.

2.1 Customs Transaction Records

The first dataset we use is the transaction level records from the General Administration of Customs of China (GACC) as in [Manova and Zhang \(2012\)](#). The whole sample period ranges from 2000 to 2011. This dataset includes the most comprehensive information on all Chinese trade transactions including import and export values (denominated in US dollars), quantities, units, product names and codes, source and destination countries, and type of enterprises (e.g., state-owned, private, foreign-invested, and joint ventures), etc. Using these high-frequency trade records, we are able to compute firm–product unit values to study export or import price responses to exchange rate shocks.

We separate the full records into export and import parts. In our analysis, each unique transaction refers to a firm-product-country-year consolidation. The categories of products in China’s customs trade records are coded according to the Harmonized Coding and Description System (Harmonized System or HS) from World Customs Organization (WCO). The original data is subject to HS 8-digit classification. Since there are two major revisions of the HS system in 2002 and 2007, we aggregate HS8 product-level information to the HS6 level and then use conversion tables from the United Nations Trade Statistics to convert HS 2007 and HS 2002 codes into the older version of HS 1996 as in [Fan, Li and Yeaple \(2015\)](#).

For later empirical studies, we drop unwanted observations referring to the standard of [Li, Ma and Xu \(2015\)](#): (1) products with inconsistent missing information of unit or

quantity; (2) special product categories such as arms (HS2=93), antiques (HS2=97), and special categories (HS2=98 and 99); (3) transactions existing for only one year without any change over time. Those outliers only make up a very small part of observations. We will refer to this sample as the "long sample" in subsequent analyses.

2.2 Chinese Firm-level Data

Our source of Chinese firm-level production and financial information is the Chinese Industrial Enterprises (CIE hereafter) database from the National Bureau of Statistics of China (NBSC). This database covers all state-owned enterprises and above-scale firms with annual sales of more than 5 million RMB. The dataset covers the period from 1999 to 2007. The number of firms each year ranges from about 130,000 in 1999 to 300,000 in 2007. The data provide details about firms' identification code, ownership, industry type, and about 80 other variables in the balance sheet. The company information variables we use in this project include the number of employees, total wage payments, the value of fixed assets, sales income, total operation inputs, etc.

To merge this firm-level survey data with customs records, we follow the standard procedure to match the identification codes based on the contact information of firms as in [Fan, Li and Yeaple \(2015\)](#). Manufacturing firms participating in international trade in the matched sample are uniquely identified by the FRDM codes and year. We drop unsatisfactory observations following the criteria of [Brooks, Kaboski and Li \(2021\)](#). This merged sample contains the overlapping time of the two datasets, i.e. from 2000 to 2007, and all indicators are in annual terms. We will refer to this combined dataset as the "matched sample" in the rest of this paper.

The summary statistics of the whole customs records, the firm information dataset, and the final matched sample are shown in panels A, B, and C in [Table 1](#), respectively.

A notable point is that for all variables involved in the amount of money in the table, the mean is much larger than the median, and the mean of some variables is even larger than its 90% quantile. This means that the distribution of trade value is very uneven, with a few large transactions accounting for the majority of trade volume.

2.3 Country-level Macro Data

We obtain bilateral nominal exchange rates and price level of household consumption from the newest Penn World Table (PWT 10.0) (referring to [Feenstra, Inklaar and Timmer \(2015\)](#)). From the whole table, we keep 183 countries (or districts) using 136 different fiat currencies, which have full records of exchange rates during the period from 1999 to 2011.

Table 1: Summary Statistics for Main Samples

	#observations	Mean	Median	Std. dev	P10	P90
Panel A: Customs records						
Export Value (USD)	18,581,221	424868	21692	1.04E+07	888	423436
Export Price (RMB)	18,581,221	22007.45	30.10417	2229173	4.564519	556.4724
Annual Export Price Change	11,400,795	0.025908	0.005982	0.665267	-0.50011	0.5709025
Import Value (USD)	14,172,315	439283	7721	1.98E+07	214	292720
Import Price (RMB)	14,172,315	49519.78	111.0406	1411944	5.159389	10247.12
Annual Import Price Change	8,580,234	0.023625	-0.00207	1.017117	-0.8523061	0.9388119
Panel B: Firm information						
Sales Income (thousand RMB)	1,745,511	78826.33	17630	714350.5	5318	111319
Employment	1,745,511	262.9454	108	964.6382	30	500
Fixed Asset (thousand RMB)	1,745,511	27437.2	4043	312024.8	573	36968
Operation Input (thousand RMB)	1,745,511	61682.99	13971	562923.1	4035	168810
Current wage payable (thousand RMB)	1,745,511	3730.157	1121	28699.16	266	6300
Panel C: Matched sample						
Export Value (USD)	3,168,876	880187.2	33693	4.66E+07	1376	712735
Export Price (RMB)	3,168,876	18326.68	28.31701	1893237	4.995613	398.6719
Annual Export Price Change	1,829,966	0.023539	0.006083	0.682097	-0.48284	0.550056
Import Value (USD)	3,280,928	1120261	11139	2.36E+07	266	529584
Import Price (RMB)	3,280,928	29955.95	76.56041	525990.1	4.966081	5614.432
Annual Import Price Change	1,827,983	-0.08694	-0.00105	1.34694	-1.20575	1.013989

* This table shows the summary statistics of some important variables in our three major datasets. Panel A and panel C describe the total annual values, annual average prices, and price changes for the whole customs records and the matched sample, respectively. The observations in panel A and panel C are at the firm-product-country-year level. The trade values in panel A and panel C are in US dollars while prices are in RMB. Panel B describes sales and costs information of Chinese manufacturing firms during 2000-2007. The money values in panel B are in thousands of RMB. The observations in panel B are at the firm-year level.

The bilateral nominal exchange rate is defined as the number of home currency units that can purchase a unit of foreign currency. An increase in NER_{ct} means a nominal depreciation of the Chinese RMB against this currency from country c . Following LMX Li, Ma and Xu (2015), the CPI-based real exchange rate (RER_{ct}) is defined as the nominal exchange rate multiplied by the foreign consumer price index (CPI) and divided by the Chinese consumer price index (CPI) at the same year, which is

$$RER_{ct} = NER_{ct} \cdot \frac{CPI_{ct}}{CPI_{CHN,t}}.$$

Again, an increase in RER_{ct} means a real depreciation of the Chinese RMB against the foreign country's c currency. In later specifications, we mainly use the first difference of the logarithm of the real exchange rate to represent exchange rate changes.

We saw substantial variations in RMB exchange rate fluctuations against different countries including its major trading partners during the period. We could observe that the real exchange rate against the US dollar did not change a lot in 2000-2004 due to the nominal pegging scheme of RMB to US dollars. In July 2005, the peg was lifted to a slight appreciation of RMB against US dollars as a result of the evolution of exchange policy. The currency exchange rates of China's other major trading partners also fluctuated to varying degrees during this period. Changes in nominal and real exchange rates for these countries (level in 1999 as base value 100) are shown in figures 1 and 2.

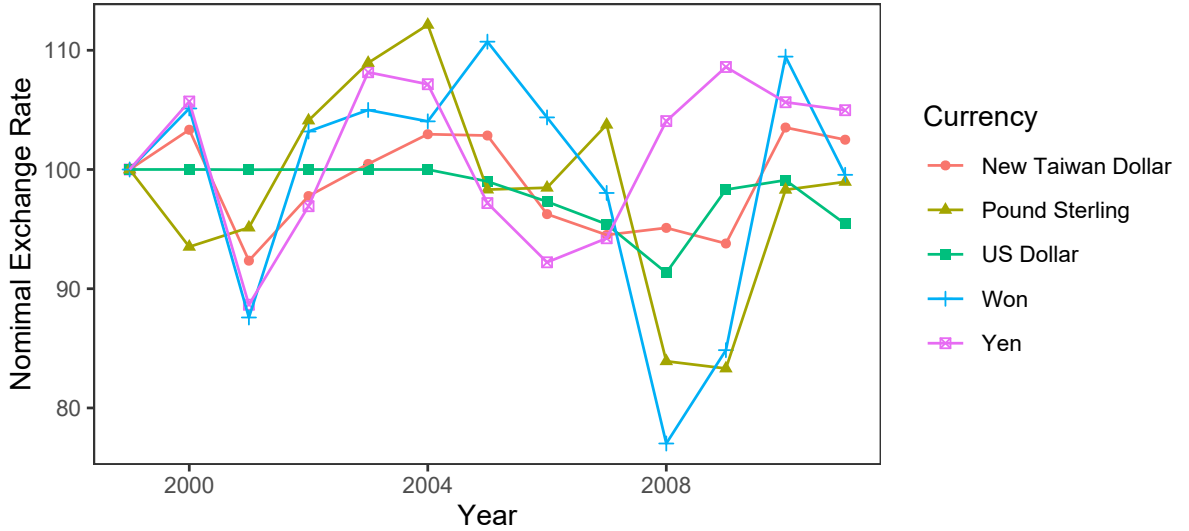


Figure 1: Nominal exchange rates of China's major trading partners (1999-2011)

In addition to nominal and real exchange rates, we also use the real GDP of the destination countries from PWT 10.0. The real GDP is computed with national-accounts growth rates. The controls of real GDP changes of the destination country, $\Delta RGDP_{ct}$, help us exclude the effect of economic growth on price movements. All macro variables including exchange rates and real GDPs are in annual terms to match the firm information

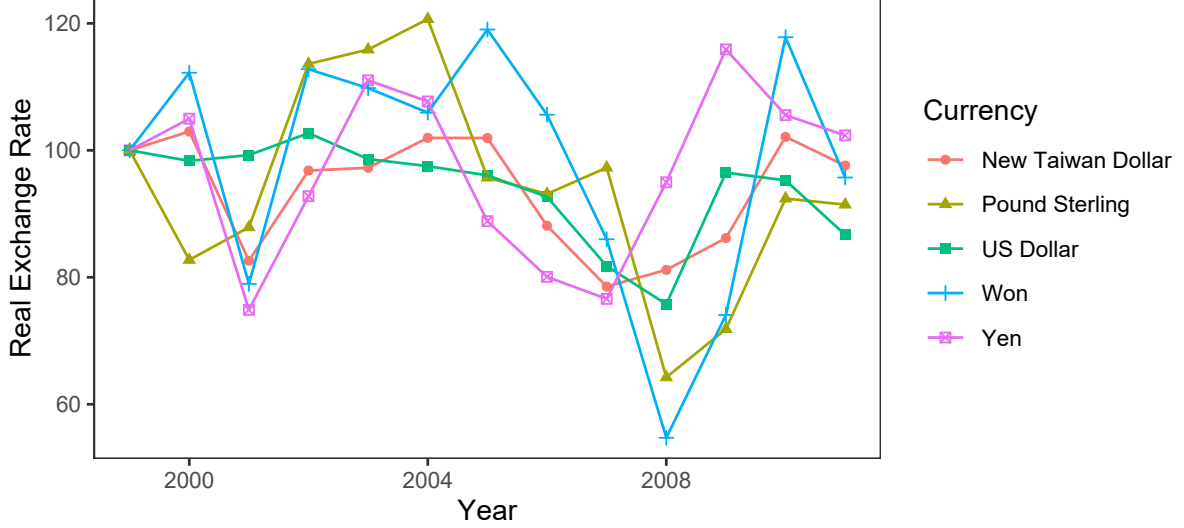


Figure 2: Real exchange rates of China’s major trading partners (1999-2011)

survey.

2.4 Credit constraints

Another critical issue in our empirical study is to measure the extent of financial constraints. To deal with potential concurrent endogeneity, our measures of credit constraints are applied to each firm across the whole period. Following a widely recognized literature on the role of credit constraints in international trade (Kroszner, Laeven and Klingebiel (2007); Manova, Wei and Zhang (2015); Fan, Lai and Li (2015)), we use multiple financial vulnerability measures at the sector level to proxy for credit needs (demand for outside capital) and ability to resist financial risks. These measures are designed to reflect the nature of each industry which should be regarded as exogenous for each firm. If a firm is in a more financially vulnerable industry, it tends to face a tighter credit constraint, regardless of its operating conditions.

The first measure we use is external finance dependence ($ExtFin_j$), the share of capital expenditures not financed by operational cash flows. If external finance dependence is high, the industry is more financially vulnerable and firms in this industry are more credit constrained. The second measure is asset tangibility ($Tang_j$), which describes the share of the net value of tangible assets that firms can pledge as collateral to raise external finance, in its total book value. The third measure is the inventory-to-sales ratio ($Invent_j$), which measures the production cycle duration and the necessary working capital to maintain inventories and meet demand.

To utilize the U.S. industry-level credit measures in the literature, we match the CIC industry code system used in China to the International Standard Industrial Classification (ISIC) system. We first convert the older ISIC Revision 2 3-digit and 4-digit industries

from [Manova, Wei and Zhang \(2015\)](#) to match the newest ISIC Revision 3 codes; then we link the ISIC Revision 3 codes to the adjusted CIC codes in CIE datasets. Finally, we could match firms in the merged sample to those sector-level financial vulnerability measures. One-to-many situations may occur in the process of encoding matching. In this case, we construct the target variable for the new industry by averaging its source industries.

Although we construct three measures of credit constraints as in the literature, we will focus on the external finance dependence and tangibility in our later analysis. One important reason is that their interpretation can be linked to firms' exposure and resistance to financial frictions directly. In contrast, the inventory ratio may be connected to inventory management efficiency rather than liquidity and financial reasons. Following [Manova, Wei and Zhang \(2015\)](#), we also construct the first principal component of external finance dependence and asset tangibility FPC_j , which increases with the former and falls with the latter. An industry with a higher FPC_j is more financially sensitive if firms in it require more outside funds but own less collateralizable assets. Therefore, we could use FPC_j as an aggregate measure to combine information about financial vulnerability from $ExtFin_j$ and $Tang_j$.

We have two major reasons why we use credit constraint measures based on US data in our main regressions. First, we want to remove the distortion by the limited credit supply in China and focus on the credit demand associated with sectoral characteristics. Second, the U.S. patterns of sectoral credit demand are proved persistent in a cross-country setting in the literature ([Kroszner, Laeven and Klingebiel \(2007\)](#); [Manova, Wei and Zhang \(2015\)](#); [Fan, Lai and Li \(2015\)](#)), especially when the industry classification is broadly defined. Intuitively, the financial needs of an industry may differ in level across countries, but the relative ranking between industries is supposed to be the same across countries, due to technical reasons specific to the industry itself.

Alternatively, we also compute credit needs based on Chinese firm-level information from CIE data. In addition to the already mentioned three measures, external finance dependence ($ExtFin_j$), asset tangibility ($Tang_j$), and inventory ratio ($Invent_j$), we include the fourth measure is R&D intensity (RD_j), defined as the ratio of research and development expenditure to the total sales. Usually, R&D activities are capital-intensive so it requires firms to pay a large fixed cost before production and sales. Therefore, firms in an R&D-intensive industry should be more financially vulnerable. However, since we only have the information on firms' R&D expenditure in and after 2005, which narrows the range of available samples, we will only use R&D intensity as an auxiliary proxy variable.

We adopt the measure of external finance dependence used by [Fan, Lai and Li \(2015\)](#). Then we calculate the inventory ratio as the value of inventory over sales income, the asset tangibility as the value of fixed assets over total assets, and the R&D intensity as

R&D spending over total sales income. To avoid credit constraints being endogenously affected by other corporate factors, we take the median of the firm-level credit constraint measure in the same CIC 2-digit industry as the industry-level credit constraint measure. Since the R&D investment of a considerable number of companies is equal to 0, we choose to take the average rather than the median when calculating the R&D intensity of the industry. The regression results using the Chinese industry measures are provided as robustness checks.

The summary statistics of the credit constraints measures in the firm-level data are shown in panels A and B in Table 2, respectively.

Table 2: Summary Statistics of Credit Constrains Measures

	#observations	Mean	Median	Std. dev	P10	P90
Panel A: US Measures						
FPC_j	1,745,511	-7.12e-09	-0.2706642	1	-1.071394	1.072687
$ExtFin_j$	1,745,511	-.0036698	-0.05	0.3112002	-0.25	0.28
$Tang_j$	1,745,511	0.3106788	0.32	0.0944181	0.1866667	0.43
$Invent_j$	1,745,511	0.1594069	0.1633333	0.0292352	0.115	0.1933333
	#observations	Mean	Median	Std. dev	P10	P90
Panel B: Chinese Measures						
$ExtFin_j$	1,745,511	-0.6479498	-0.47	0.6746751	-1.32	-0.1
$Tang_j$	1,745,511	0.3332769	0.3268749	0.0648019	0.2390799	0.4317028
$Invent_j$	1,745,511	0.1102537	0.1030875	0.0274747	0.0778921	0.1348336
$R\&D_j$	1,745,511	0.0168278	0.012111	0.0142106	0.0053125	0.0281532

* This table shows the summary statistics of credit constraint measures. Panel A describes the measures calculated using US data while panel B shows the alternative Chinese version. All variables are unitless, the numerical size only means relative rank.

3 Empirical Framework

This section describes our empirical specifications. We start from the baseline estimations of exchange rate pass-through and then include

3.1 Estimating Equations

3.1.1 Baseline estimations of exchange rate pass-through

Since exchange rate pass-through is not an indicator that can be directly measured, we need to use panel data regression to estimate it. The first step goal is to estimate exchange rate pass-through as the elasticity of unit values changes to exchange rate changes using the firm-product-country details. Our strategy is based on the fixed effects regression. Specifically, we run a regression of import price changes on the bilateral real exchange rate changes between China and the source country, controlling the change of real GDP

in the source country. The baseline equation refers to [Amiti, Itskhoki and Konings \(2014\)](#) and [Li, Ma and Xu \(2015\)](#) as below:

$$\Delta \ln P_{ijct}^D = \alpha + \beta^D \Delta \ln RER_{ct} + \gamma \Delta \ln RGDP_{ct} + \xi_{ijc} + \tau_t + \varepsilon_{ijct} \quad (1)$$

where P_{ijct} represents the trade price of the product i bought (sold) by firm j from (to) country c during year t . $D \in \{\text{Import, Export}\}$ denotes the trade direction. Therefore, this equation could be used to estimate both import and export exchange rate pass-through with different values of D . RER_{ct} is the bilateral real exchange rate between Chinese RMB and currency in the country c . $RGDP_{ct}$ represents the real GDP of the source country deflated to the constant price level, which proxy for market demand. ξ_{ijc} denotes the firm-product-country level fixed effects to capture any time-invariant unobserved factors for a combination of firm, product, and destination. These multi-dimensional fixed effects restrict unit value changes to price adjustments, rather than other changes in corporate trade decisions. τ_t , the year dummies, control for macro-shocks that are common to all firms. We will alter the fixed effects setting in robustness checks.

To deal with possible non-stationarity of the panel, we use the first difference of the logarithms for prices $\Delta \ln P_{ijct}^D$, real exchange rates $\Delta \ln RER_{ct}$ and real GDP $\Delta \ln RGDP_{ct}$ to represent their annual rates of change. In this way, we transform the dynamic panel into a fixed effects regression. Therefore, using import price changes, the estimated coefficient of interest β is the elasticity of price changes to exchange rate changes, i.e. import exchange rate pass-through. We also provide estimations for export exchange rate pass-through which is more common in past literature, where import prices are now replaced by export prices and the information of sources is replaced by which of destination markets.

The real prices for export and import P_{ijct}^{Import} and P_{ijct}^{Export} are both denominated by the Chinese RMB in this paper. Using RMB as the denomination currency will make the coefficients on the import side and export side have different meanings. The level of coefficient $\beta^{D=Import}$ measures the completeness of import exchange rate pass-through, i.e. a higher β means Chinese importers face more volatile import RMB prices during exchange rate shocks. However, for the export side, $\beta^{D=Export}$ means the "incompleteness" of the export pass-through because a higher β means Chinese exporters pass less exchange rate change to the destination market while having more volatile domestic currency prices. To be noticed, a majority of firms in our sample are both importers and exporters so they will appear in estimations of both exchange rate pass-through.

3.1.2 Estimations with credit constraints

Since our main focus is how firms' credit constraints affect exchange rate pass-through, we then include an interaction term of sectors' financial vulnerability into the estimation function. Intuitively, firms operating in those financially vulnerable industries tend to have less access to enough funds to support their international trade activities, that is, they are subject to tighter credit constraints.

Therefore, we study the credit constraint effects on exchange rate pass-through across sectors with the following panel regression:

$$\Delta \ln P_{ijct}^D = \alpha + \beta_1^D \Delta \ln RER_{ct} + \beta_2^D \Delta \ln RER_{ct} \cdot FV_j + \gamma \Delta \ln RGDP_{ct} + \xi_{ijc} + \tau_t + \varepsilon_{ijct} \quad (2)$$

where the variable FV_j represents the financial vulnerability of the sector to which the firm j belongs and the rest are the same as those in the baseline equation. The interaction coefficient β_2 represents the effect of credit constraints on exchange rate pass-through. A positive β_2^{Import} for importers implies more credit-constrained importers have a more complete import exchange rate pass-through, while a positive β_2^{Export} implies more credit-constrained exporters have a less complete exchange rate pass-through. The overall import ERPT for an importer j is given by $\beta_1^D + \beta_2^{Import} FV_j$ and the export ERPT for an exporter j is $\beta_1^D + \beta_2^{Export} FV_j$.

Through this estimation strategy, we hope to scrutinize how the pricing behavior of Chinese importers in response to the exchange rate is affected by credit constraints and compare it with that of exporters. Although the functional forms for export and import pass-through are similar, the underlying mechanism could be different, which is one of our key innovation points. While credit-constrained exporters' pricing decisions to deal with exchange rate shocks are mainly related to production and profit margin, the penetration effect of credit constraints on import prices is through a more direct channel, as the shortage of funds directly affects purchasing choices and bargaining.

3.1.3 Estimations with additional factors

After estimating exchange rate pass-through at the firm level and the impact of credit constraints on it, we need to go a step further to explore why. Through what channels will credit constraints affect the ability of importers to cope with exchange rate shocks? What other factors would exacerbate or diminish this effect? Are the effects of credit constraints fully explained by these controls?

In this part, we explore additional factors that affect firm-level import exchange rate pass-through. To do so, we introduce a vector \mathbb{Z}_{jt} (or its lagged form \mathbb{Z}_{jt-1}) to include

those additional factors and apply it to both control terms and interaction terms with real exchange rate changes:

$$\Delta \ln P_{ijct}^D = \alpha + [\beta_1^D + \beta_2^D \cdot FV_j + \beta_3^D \cdot \mathbb{Z}_{jt-1}^D]' \Delta \ln RER_{ct} + \gamma \Delta \ln RGDP_{ct} + \mathbb{Z}_{jt}^{D'} \eta + \xi_{ijc} + \tau_t + \varepsilon_{ijct}. \quad (3)$$

$$\begin{aligned} \Delta \ln P_{ijct}^D = & \alpha + [\beta_1^D + \beta_2^D \cdot FV_j + \beta_3^D \cdot \mathbb{Z}_{jt-1}^D + \beta_4^D \cdot FV_j \cdot \mathbb{Z}_{jt-1}^D]' \Delta \ln RER_{ct} \\ & + \gamma \Delta \ln RGDP_{ct} + \mathbb{Z}_{jt}^{D'} \eta + \xi_{ijc} + \tau_t + \varepsilon_{ijct}. \end{aligned} \quad (4)$$

We will then use the estimation strategy of the form 3 and 4 to take into account various factors that may directly or indirectly affect exchange rate pass-through. We want to verify whether the effects of credit constraints are fully explained by some of these channels. Our firm-level data has the merit of containing information on its production, so we could connect the exchange rate pass-through with estimated firm-specific markup. We will also include trading partners later to measure the flexibility to choose alternative import sources. The coefficient of the interaction term between additional factors and real exchange rate movement β_3 represents the direct effects of those factors on the exchange rate pass-through other than through financial constraints. In equation 4, the triple interaction coefficient represents the indirect effects of those factors on the exchange rate pass-through through financial constraints. The same sign of β_2 and β_4 means that the factor enhances the effect of credit constraints, while the opposite sign means that it alleviates credit constraints.

Market share is another popular proxies for firm size or market power. For example, [Amiti, Itskhoki and Konings \(2014\)](#) uses destination-specific market shares proxying for markup elasticity. When using market share as the additional factor, equation 5 becomes the derivative form as below:

$$\begin{aligned} \Delta \ln P_{ijct}^D = & \alpha + [\beta_1^D + \beta_2^D \cdot FV_j + \beta_3^D \cdot S_{ijct}^D + \beta_4^D \cdot S_{ijct}^{D^2}] \Delta \ln RER_{ct} \\ & + \gamma \Delta \ln RGDP_{ct} + \eta S_{ijct}^D + \xi_{ijc} + \tau_t + \varepsilon_{ijct}. \end{aligned} \quad (5)$$

The quadratic term for market share is used to test whether there is a non-monotonic relationship between exchange rate pass-through and market shares, such as the U-shape relationship documented by [Garetto \(2016\)](#) and [Devereux, Dong and Tomlin \(2017\)](#). In addition to this, controlling for market share improves our estimation accuracy of the effect of credit constraints on exchange rate pass-through.

3.2 Measurements

3.2.1 Unit value as trade price

The customs records contain disaggregated trade values (denominated by US dollars) and quantities for each HS6 product i , each firm j , from (or to) each country c , in each year t , V_{ijct} , and Q_{ijct} . We first convert the value of the goods into RMB using the average exchange rate for the year. Then, the import and export prices we use are computed as unit values, defined as

$$P_{ijct}^D = \frac{V_{ijct}^D \cdot NER_{US,t}}{Q_{ijct}^D}$$

where $D \in \{\text{Import, Export}\}$ and $NER_{US,t}$ is the annualized nominal exchange rate of US dollars in terms of RMB in year t . Because product categories are highly subdivided, we believe that the unit value is an ideal proxy for the transaction price.

Similar to real exchange rates, we take the first difference of the logarithm to represent price changes of a certain product across years. We will exclude observations with the annual growth rate of unit value in the top or bottom 1 percentile in the distribution, by HS2 product category and year, to avoid results being affected by extreme idiosyncratic factors other than some exchange rate adjustments.

3.2.2 Import sources and export markets

Following the literature about import sourcing, an importer's sourcing diversity could increase its bargaining power in import prices in addition to its production characteristics. We want to test how importers' sourcing diversity affects exchange rate pass-through. We provide a simple measure for the firm-product level sourcing diversity $Source_{ijt}$ as the number of source countries from which an importer j imports a certain HS6 product type i in year t .

Similarly, for exporters, we count the number of destination countries to which an exporter exports a certain HS6 product type i as the firm-product-level selling measure, $Market_{ijt}$. Controlling other variables, the number of export markets for the same product can measure its export network diversity. In a robustness test, we use the firm-year fixed effect to control for differences in import and export diversity caused by firm size.

4 Empirical Results

In this chapter, we will present our major empirical results using the samples and strategies described above. First, we show the estimated import exchange rate pass-through is rather incomplete while the export exchange rate pass-through is very close to complete. Second, we show that both importers and exporters with tighter credit constraints have more complete pass-throughs. Finally, we present the results with firm heterogeneity to explore factors affecting importers' capacity to absorb exchange rates and the role of credit constraints among them.

4.1 Import Pass-through vs Export Pass-through

Past literature on firm-level evidence of exchange rate pass-through mainly focuses on exporters' price-setting behaviors. Importers are likely to be more than simple price takers, which gives us a new perspective on exchange rate pass-through. Now we take a closer look at how real exchange rate fluctuations affect import prices and export prices differently in China. The results for import exchange rate pass-through versus export exchange rate pass-through are shown in Table 3 using different samples.

We report the baseline estimates of import exchange rate pass-through in panel A. Column (1) shows the import exchange rate pass-through using equation (1) for the long sample from 2000 to 2011, including all companies that appear at least once in customs records, whether or not they are registered in the CIE database. Column (2) shows the results for import exchange rate pass-through for the matched sample (importers registered in the CIE database from 2000 to 2007). The pass-through coefficient in column (2) is larger than the one in column (1), yet both are incomplete. The average import exchange rate pass-through for China in the long sample is around 18%, while in the matched sample about 39%. The latter ERPT means that the import prices denominated in RMB will increase by about 3.9% during a 10% real depreciation and decrease by the same amount during a 10% appreciation. Column (3) and column (4) report the import pass-through among a subset of the matched sample including only the top 50 and top 20 trading partners by import value. The results in those subsamples of top partners are slightly less complete than those in column (2), yet the levels are of similar magnitude. Import price fluctuations will reflect nearly one-third of exchange rate fluctuations, with the remainder absorbed by foreign currency price fluctuations. These results imply that Chinese importers have to bear less than half but still considerable cost fluctuations due to exchange rate shocks.

Accordingly, estimates of export exchange rate pass-through are recorded in panel B. Similar to which in panel A for imports, column (1) and column (2) in panel B show the export exchange pass-through using the long sample and the matched sample. Column (3)

Table 3: Baseline Estimations of Exchange Rate Pass-Through

	(1)	(2)	(3)	(4)
Panel A	Import			
	Whole	Matched	Top 50	Top 20
$\Delta \ln RER_{ct}$	0.179*** (0.003)	0.357*** (0.015)	0.354*** (0.015)	0.344*** (0.016)
$\Delta \ln RGDR_{ct}$	-0.133*** (0.026)	0.263*** (0.090)	0.282*** (0.091)	0.333*** (0.097)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	8409682	1792020	1781948	1684798
Panel B	Export			
	Whole	Matched	Top 50	Top 20
$\Delta \ln RER_{ct}$	0.050*** (0.002)	0.031*** (0.005)	0.039*** (0.006)	0.065*** (0.009)
$\Delta \ln RGDR_{ct}$	-0.102*** (0.010)	-0.083** (0.037)	-0.118*** (0.042)	-0.082 (0.056)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	11173463	1793974	1611410	1251147

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Column (1) uses the long sample from 2000 to 2011. Column (2) uses the matched sample from 2000 to 2007. Columns (3) and (4) use finer sub-samples with only top 50 and top 20 partners ranked by total trade value. Panel A shows the results of import ERPT while panel B shows the results of export ERPT. All regressions include firm-product-country fixed effects and year fixed effects.

and column (4), in turn, show the export price pass-through to major export destinations. The estimated export pass-through in each column equals one minus the coefficient at the row of $\Delta \ln RER_{ct}$. The export pass-through ranges from 93.3% for the top 20 countries to 96.6% for the matched sample, which is near complete. Specifically, 95% export ERPT means that a 10% real appreciation of RMB concerning a destination market is associated with a 0.5% decrease in the RMB price while a 9.5% increase in foreign currency price. That is to say, export price in RMB adjusts very little with exchange rate fluctuations.

The strikingly almost complete ERPT into RMB export price echoes the finding of LMXLi, Ma and Xu (2015). One explanation could be that most Chinese exporters are at the lower end of the value chain, where high ERPT exposes their low profit margins, leaving no room for pricing-to-market strategy. Most Chinese exporters have no choice but to pass all exchange rate swings to their destination prices, regardless of potential better monopolistic competition strategies. In addition, since RMB is pegged to the dollar until 2005, the complete pass-through to price in destination currency is consistent with a rigid price model with the US dollar as the invoice currency, at least in those countries with freely floating exchange rates relative to the U.S. dollar. We will later provide results using samples excluding the US and other countries whose currency is pegged to the US dollar.

From the comparison, it can be seen that the exchange rate-price pass-through on China's import side is much less complete than that on the export side. That is to say, for Chinese firms, when the RMB depreciates against the currencies of major trading partners, export prices denominated in RMB will not rise significantly, but their import costs will rise sharply; on the contrary; when the real exchange rate of RMB appreciates, export prices in RMB will decrease only to a limited extent, and their import costs will increase dramatically drop. If we consider a typical two-way trader in China who simultaneously imports and exports from two groups of countries with strong correlations in exchange rate fluctuations, a devaluation of the local currency will reduce his unit profits, while an appreciation of the local currency will widen his profit margins. We call this phenomenon the asymmetry of the import and export exchange rate pass-through, which exposes Chinese trading companies to two-way exchange rate risks.

4.2 Effects of Credit Constraints

Another goal of our paper is to assess how importers with varying degrees of financial vulnerability absorb exchange rate fluctuations when the home currency depreciates or appreciates. We evaluate the consequences of credit constraints on the firms' price responses to exchange rate shocks using equation 2. Table 4 presents differences in exchange rate pass-through into import prices and export resulting from the industry-level credit demand heterogeneity. Panel A reports the results for credit constraints and import

pass-through and panel B reports the comparing results for the export side.

We are particularly interested in the coefficients of the interaction terms. Note that a larger elasticity coefficient on the import side implies a more complete exchange rate-price pass-through, and similarly a positive cross-term coefficient implies that the magnitude of this variable is positively related to exchange rate pass-through. Using the first principal component of external finance dependence and asset tangibility FPC_j to measure financial vulnerability FV_j , we see that import exchange rate pass-through is more complete in financially more vulnerable sectors, relative to financially less vulnerable sectors (column 1, row 3). Columns (2) and (3) separately show the effects of external finance dependence and asset tangibility on importers' exchange rate pass-through. Consistent with the definition that higher external finance dependence implies tighter credit constraints faced by firms while higher asset tangibility can alleviate them, we observe a positive coefficient for the former (row 5) and a negative coefficient for the latter (row 6). When we use the auxiliary measure $Invent_j$, we further observe that the effect on exchange rate pass-through is positive (column 4). Overall, the coefficient β_2^{Import} of the interaction term $\Delta \ln RER_{ct} \cdot FV_j$ is positive and significant at the 1% level. Our evidence supports the intuition that exchange rate fluctuations are more likely to be reflected in unstable import costs for importers in more financially vulnerable industries because they have weak bargaining power in the international market. External financing dependence, internal collateral capacity, and inventory turnover all act jointly and separately on the exchange rate pass-through of importers.

It is worth noting that the exchange rate pass-through here is estimated by the at-the-dock import price in our specification. This excludes any impact of credit constraints on post-landing costs, such as local distribution and logistics costs. In other words, credit constraints imposed on Chinese importers can affect home import costs by directly affecting the supplier's pricing behavior. Intuitively, firms in more credit-constrained industries may have weaker import bargaining power, and products are more likely to be pegged to the dollar or the currency of the exporting country (less pricing-to-market), so at-the-dock prices are more affected by exchange rate fluctuations, regardless of any domestic market factor.

By substituting the superscript D in equation 2 to export, we obtain a comparative result of the effect of credit constraints on export price pass-through. Estimates in columns (1), (2), and (4) all show significantly negative coefficients on interaction terms while column (3) shows a negative significant coefficient. The estimates suggest that financial constraints lead export exchange rate pass-through to a more complete degree, although the original result is already close to complete. These results verified the conclusion of [Strasser \(2013\)](#) who argues that financially constrained firms have higher export price pass-through compared to unconstrained firms. That is to say, credit constraints restrict exporters from absorbing exchange rate shocks, potentially because firms need external

Table 4: Effects of Credit Constraints on Exchange Rate Pass-Through

	(1)	(2)	(3)	(4)
Panel A	Import			
	FPC	External Finance	Tangibility	Inventory
$\Delta \ln RER_{ct}$	0.123*** (0.016)	0.218*** (0.016)	1.175*** (0.033)	-0.739*** (0.069)
$\Delta \ln RGDR_{ct}$	0.314*** (0.090)	0.323*** (0.090)	0.283*** (0.090)	0.273*** (0.090)
$\Delta \ln RER_{ct} * FPC_j$	0.379*** (0.010)			
$\Delta \ln RER_{ct} * ExtFin_j$		1.159*** (0.029)		
$\Delta \ln RER_{ct} * Tang_j$			-3.305*** (0.117)	
$\Delta \ln RER_{ct} * Invent_j$				6.305*** (0.389)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1792020	1792020	1792020	1792020
Panel B	Export			
	FPC	External Finance	Tangibility	Inventory
$\Delta \ln RER_{ct}$	0.039*** (0.006)	0.034*** (0.005)	-0.030** (0.015)	0.102*** (0.030)
$\Delta \ln RGDR_{ct}$	-0.084** (0.037)	-0.083** (0.037)	-0.084** (0.037)	-0.083** (0.037)
$\Delta \ln RER_{ct} * FPC_j$	-0.019*** (0.004)			
$\Delta \ln RER_{ct} * ExtFin_j$		-0.045*** (0.013)		
$\Delta \ln RER_{ct} * Tang_j$			0.230*** (0.053)	
$\Delta \ln RER_{ct} * Invent_j$				-0.412** (0.171)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1793974	1793974	1793974	1793974

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (1)-(4) use different measures of credit constraints calculated using U.S. data. Panel A shows the results of import ERPT while panel B shows the results of export ERPT. All regressions include firm-product-country fixed effects and year fixed effects.

finance to apply pricing-to-market strategies in foreign markets.

Comparing panel A and panel B, although the import ERPT is still less complete than the export ERPT, we can reach a consistent conclusion that credit constraints steer both of them toward a more complete direction. Following the our analysis, credit constraints expose Chinese manufacturing firms to greater exchange rate risk in international trade. Exporters with more vulnerable credit are forced to lower destination prices when RMB depreciates compared to those with unrestricted credit, while RMB income remained relatively unchanged, and importers' costs rose more significantly; in contrast, when RMB appreciates, restricted exporters will increase destination prices more, even if it means losing their competitive advantages, and importers' costs will be reduced at this time. For credit-constrained two-way traders, given import sources and export markets cannot be adjusted quickly, the unit profit margin is more sensitive to exchange rate fluctuations.

Nonetheless, the direction in which credit constraints affect the exchange rate pass-through on the export side and the import side is the same, the underlying channels may work differently. Following [Strasser \(2013\)](#), a higher external finance premium causes higher marginal costs. Thus, firms with binding financial constraints have no choice but set higher prices and face a higher price elasticity of demand. When there is an exchange rate shock, the optimal choice is to adjust their markups but credit-constrained firms can do so only to a limited extent because they have narrower profit margins. However, for import ERPT, credit constraints affect how buyers pay in the transactions. Adequate credit or cash reserves give importers a better bargaining chip, for example, by allowing them to negotiate longer-term purchase agreements, where exchange rate fluctuations will be more borne by international sellers. In contrast, a credit-strapped importer may not have the buffers to transfer risk, so it must accept current exchange rate settlements, although that means taking on more volatile prices.

4.3 Sourcing Diversity and Credit Constraints

In this section, we further study the factors that directly affect the purchasing side. As emerged in an intuitive guess, a potential mechanism through which financial constraints affect an importer's bargaining power with foreign suppliers is its outside sourcing options. Companies with more trading partners can flexibly adjust the weight of imports from different countries. Firms with heterogeneous sourcing capacity may thus be affected by credit constraints to a different extent.

Therefore, we employ equation 4 to include the number of import sources. The estimation results are reported in the below Table 5 and confirm the empirical relevance of differences in sourcing diversity across firms. Similarly, results for sales destination diversity on the export side are provided in panel B 13 as a comparison.

Table 5: Import Sources and Effects of Credit Constraints on Import Exchange Rate Pass-Through

	(1)	(2)	(3)	(4)
Panel A		Import		
	#Sources	#Sources+ FPC	#Sources+ External Finance	#Sources+ Tangibility
$\Delta \ln RER_{ct}$	0.433*** (0.017)	0.177*** (0.019)	0.274*** (0.018)	1.386*** (0.040)
$\Delta \ln RGDR_{ct}$	0.250*** (0.090)	0.292*** (0.090)	0.297*** (0.090)	0.267*** (0.090)
$\#Source_{ijt}$	-0.021*** (0.002)	-0.016*** (0.003)	-0.016*** (0.003)	-0.050*** (0.006)
$\Delta \ln RER_{ct} * FPC_j * \#Source_{ijt}$		-0.014*** (0.002)		
$\Delta \ln RER_{ct} * FPC_j$		0.443*** (0.012)		
$\Delta \ln RER_{ct} * ExtFin_j * \#Source_{ijt}$			-0.054*** (0.006)	
$\Delta \ln RER_{ct} * ExtFin_j$			1.410*** (0.037)	
$\Delta \ln RER_{ct} * Tang_j * \#Source_{ijt}$				0.104*** (0.026)
$\Delta \ln RER_{ct} * Tang_j$				-3.790*** (0.148)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1792020	1792020	1792020	1792020

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A shows the results of import ERPT while panel B shows the results of export ERPT. Panel B is shown in Appendix 13. All regressions include firm-product-country fixed effects and year fixed effects.

The estimates for intersection terms between import sources and real exchange rate changes are displayed in column (1). We find that importers who import a certain product from more sources will have a less complete pass-through. This is consistent with our hypothesis that importers with more alternative sourcing options will have less complete pass-through. Interestingly, exporters who export to more destinations (both for a certain product or in total) will have a slightly more complete pass-through. In other words, the diversity of import sources for the same product can significantly enhance the stability of import prices, but the diversity of export markets does not.

In columns (2)-(4), after adding interactions, we find the effects of credit constraints still exist while the triple interaction terms with the number of sources have the opposite and significant coefficients. That means a wider sourcing base will mitigate the effects of credit constraints, in addition to its effect on pass-through. We continue to observe that this triple interaction effect only works for the import side in panel A but not for the export side in panel B (attached in appendix 13). The opposing effects of credit constraints and purchasing diversity on exchange rate pass-through confirm our conjecture about the bargaining power of importers. If a firm can import the same product from more sources, it has more flexibility in the face of bilateral exchange rate shocks in individual markets. In other words, constrained firms with more import sources have more ways to escape the unfavorable exchange rate risk. A more diverse importer can either switch from one source to another to reduce costs (trade diversion effect) or make a more credible threat to negotiate a more stable price.

5 Robustness

In this chapter, we further check the robustness of our main results with alternative measures and sub-samples.

5.1 Alternative Measures of Credit Constraints

As the first robustness test to verify our baseline results, we use alternative credit constraint measures from CIE database. The purpose is to avoid potential bias from differences in the attributes of industry credit demand in different countries. The details of constructing these Chinese variables are discussed in section 3.2. Although Chinese financial market is less mature than that of the US, the rankings of industries in credit constraints are comparable. Thus, the results based on the credit constraints measures from Chinese data are expected to be consistent with our main findings. Our results are reported in Table 6 which can be easily compared with the results using US measures.

Columns 1-4 of panel A of table 6 present the import-side results for external finance

Table 6: Alternative Estimations with Chinese Measures of Credit Constraints

	(1)	(2)	(3)	(4)
Panel A	Import			
	External Finance	Tangibility	Inventory	R&D Intensity
$\Delta \ln RER_{ct}$	0.502*** (0.018)	2.032*** (0.053)	-0.752*** (0.043)	0.043** (0.019)
$\Delta \ln RGDP_{ct}$	0.249*** (0.090)	0.240*** (0.090)	0.333*** (0.090)	0.304*** (0.090)
$\Delta \ln RER_{ct} * ExtFin_j$	0.223*** (0.014)			
$\Delta \ln RER_{ct} * Tang_j$		-5.776*** (0.176)		
$\Delta \ln RER_{ct} * Invent_j$			9.797*** (0.352)	
$\Delta \ln RER_{ct} * R\&D_j$				16.398*** (0.573)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1792020	1792020	1792020	1792020
Panel B	Export			
	External Finance	Tangibility	Inventory	R&D Intensity
$\Delta \ln RER_{ct}$	0.016** (0.007)	-0.135*** (0.023)	0.086*** (0.019)	0.043*** (0.007)
$\Delta \ln RGDP_{ct}$	-0.081** (0.037)	-0.082** (0.037)	-0.081** (0.037)	-0.082** (0.037)
$\Delta \ln RER_{ct} * ExtFin_j$	-0.021*** (0.007)			
$\Delta \ln RER_{ct} * Tang_j$		0.557*** (0.074)		
$\Delta \ln RER_{ct} * Invent_j$			-0.504*** (0.165)	
$\Delta \ln RER_{ct} * R\&D_j$				-0.627*** (0.243)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1793974	1793974	1793974	1793974

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (1)-(4) use different measures of credit constraints calculated using Chinese data. Panel A shows the results of import ERPT while panel B shows the results of export ERPT. All regressions include firm-product-country fixed effects and year fixed effects.

dependence, tangibility, inventory ratio, and R&D intensity, respectively. Panel B reports the export-side results with the same variables. All regressions include firm-product-country fixed effects and year fixed effects as before. Nevertheless, most of the interaction term coefficients exhibit the same signs as above, confirming the validity of our baseline findings of the effects of credit constraints on exchange rate pass-through. We can still conclude that financially more constrained firms (both importers and exporters) have more complete exchange rate pass-through than those less constrained, even with Chinese measures.

5.2 Sub-sample Test: Two-way traders

Since most of China’s imports go through two-way traders, who conduct both export and import simultaneously, import exchange rate pass-through is likely to be related to export behaviors. Specifically, importers who also export may pass part of the price fluctuations of imported intermediate goods caused by exchange rate shocks to their export destination to buffer the impact of exchange rate risks. To test whether the pattern of China’s import pass-through still holds for two-way traders, we use the sub-sample consisting only of two-way traders for a robustness check. Table 7 presents the results for this restricted sample.

Table 7 shows that the results for the subset of two-way traders are highly similar to the results for the entire matched sample, indicating that the exchange rate pass-through pattern of two-way traders dominates China’s imports and exports. The results from the sample of two-way traders are still typical, whether estimating the exchange rate pass-through of imports and exports or examining the role of credit constraints. Combined with our discussion later, the nature of two-way traders should be the focus of future research on import-export connections.

5.3 Sub-sample Test: Excluding US Dollar Peg

Due to China’s exchange rate peg to the US dollar until July 2005, we observed very weak movements in the exchange rate of RMB against the US dollar and other currencies pegged to the US dollar as shown in 1 and 2. Therefore, in this test, we exclude the US and other countries that use the US dollar as their official currency or whose currency is pegged to the US dollar from our baseline matched sample.¹ We examine whether

¹The excluded countries and regions are Belize, Liberia, Qatar, Ecuador, Djibouti, Saint Lucia, Saint Kitts and Nevis, St. Vincent and the Grenadines, Dominica, Antigua and Barbuda, Palestine, Bahamas, Barbados, Panama, Bahrain, Curaçao, Grenada, Saudi Arabia, Zimbabwe, Macao SAR of China, Turks and Caicos Islands, Bermuda, Jordan, United States, British Virgin Islands, Dutch Sint Maarten, El Salvador, Montserrat, United Arab Emirates, Oman, Aruba, Hong Kong SAR of China, Maldives, Lebanon.

Table 7: Alternative Estimations with Two-way Traders

	(1)	(2)	(3)	(4)
Panel A	Import (Two-way traders)			
	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.394*** (0.015)	0.136*** (0.016)	0.231*** (0.015)	1.158*** (0.031)
$\Delta \ln RGDP_{ct}$	0.406*** (0.086)	0.459*** (0.086)	0.469*** (0.086)	0.427*** (0.086)
$\Delta \ln RER_{ct} * FPC_j$		0.388*** (0.009)		
$\Delta \ln RER_{ct} * ExtFin_j$			1.246*** (0.028)	
$\Delta \ln RER_{ct} * Tang_j$				-3.138*** (0.112)
Year FE		Yes	Yes	Yes
Firm-product-country FE		Yes	Yes	Yes
Observations	1712289	1712289	1712289	1712289
Panel B	Export (Two-way traders)			
	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.040*** (0.006)	0.051*** (0.006)	0.044*** (0.006)	-0.034** (0.016)
$\Delta \ln RGDP_{ct}$	-0.144*** (0.041)	-0.145*** (0.041)	-0.144*** (0.041)	-0.147*** (0.041)
$\Delta \ln RER_{ct} * FPC_j$		-0.022*** (0.005)		
$\Delta \ln RER_{ct} * ExtFin_j$			-0.048*** (0.015)	
$\Delta \ln RER_{ct} * Tang_j$				0.284*** (0.059)
Year FE		Yes	Yes	Yes
Firm-product-country FE		Yes	Yes	Yes
Observations	1415415	1415415	1415415	1415415

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. We only include those firms that both import and export during the same year. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A shows the results of import ERPT while panel B shows the results of export ERPT. All regressions include firm-product-country fixed effects and year fixed effects.

the previous main results are robust after excluding those countries whose currencies are denominated in or pegged to the US dollar.

Table 8 demonstrates that the results without the U.S. and those pegged countries are not overturned. The estimated import ERPT for this subsample is 38.2%, which is slightly more complete than 35.7% in the matched sample shown in Table 3. In turn, the estimated export ERPT for this subsample is 96.0%, which is less complete than 96.9% in the matched sample. All coefficients on interaction terms remain robust and significant, suggesting that higher external finance dependence makes exchange rate pass-through more complete while higher asset tangibility plays an opposite role. Therefore, the peg of RMB to the dollar before 2005 did not interfere with the main conclusions of this paper.

6 Discussions

6.1 Firm Heterogeneity in Markup

Given that credit constraints play an important role in China’s import exchange rate pass-through, we proceed to analyze how different factors participate in determining import exchange rate pass-through, and whether they can replace credit constraints or not. One major argument in the literature is that firms with different sales markup may have heterogeneous responses to exchange rate shocks as suggested by [Berman, Martin and Mayer \(2012\)](#) and [Li, Ma and Xu \(2015\)](#). The same logic could also apply to import exchange rate pass-through. Besides, [Li, Liao and Zhao \(2018\)](#) provide microeconomic evidence that both internal finance and external credit supply significantly promote firms’ sales growth rates. Following the empirical framework in section 3.1, we add estimated firm-level markup into interactions, and the results are shown in Table 9.

Panel A shows how markup affects import ERPT in addition to credit constraints and panel B (attached in 14) shows the comparison results for export ERPT. Column (1) shows the direct effects of firm-level markup on exchange rate pass-through. Columns (2) to (4) add the first principal component, external finance dependence, and asset tangibility to the regression respectively. All columns control for the lag term of markup.

From the coefficients of interaction terms in panel A, firms with higher markup have less complete import pass-through while firms. The effects of financial constraints are still significant and robust as in section 4.2 after controlling for markup. The coefficients in panel B 14 imply a similar conclusion. Exporters with higher markup have less complete export pass-through. In other words, higher markup on the exchange rate pass-through work in the opposite direction with tighter credit constraints. However, the coefficients in rows (4)-(6) are significant respectively, indicating that markup cannot fully explain the effect of credit constraints, which is consistent with [Xu and Guo \(2021\)](#).

Table 8: Alternative Estimations Excluding US Dollar-pegged countries

	(1)	(2)	(3)	(4)
Panel A	Import (Excluding US Dollar Peg)			
	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.382*** (0.017)	0.164*** (0.018)	0.256*** (0.018)	1.173*** (0.035)
$\Delta \ln RGDP_{ct}$	0.709*** (0.106)	0.778*** (0.106)	0.797*** (0.106)	0.730*** (0.106)
$\Delta \ln RER_{ct} * FPC_j$		0.362*** (0.010)		
$\Delta \ln RER_{ct} * ExtFin_j$			1.103*** (0.030)	
$\Delta \ln RER_{ct} * Tang_j$				-3.186*** (0.121)
Year FE		Yes	Yes	Yes
Firm-product-country FE		Yes	Yes	Yes
Observations	1489984	1489984	1489984	1489984
Panel B	Export (Excluding US Dollar Peg)			
	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.031*** (0.006)	0.038*** (0.006)	0.034*** (0.006)	-0.021 (0.016)
$\Delta \ln RGDP_{ct}$	-0.023 (0.049)	-0.024 (0.049)	-0.023 (0.049)	-0.025 (0.049)
$\Delta \ln RER_{ct} * FPC_j$		-0.016*** (0.005)		
$\Delta \ln RER_{ct} * ExtFin_j$			-0.039*** (0.014)	
$\Delta \ln RER_{ct} * Tang_j$				0.196*** (0.055)
Year FE		Yes	Yes	Yes
Firm-product-country FE		Yes	Yes	Yes
Observations	1371760	1371760	1371760	1371760

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. We do not include countries that use the U.S. dollar or currencies pegged to the U.S. dollar. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A shows the results of import ERPT while panel B shows the results of export ERPT. All regressions include firm-product-country fixed effects and year fixed effects.

Table 9: Heterogeneous Markup and Effects of Credit Constraints on Import Exchange Rate Pass-through

	(1)	(2)	(3)	(4)
Panel A		Import		
	Markup	Markup+ FPC	Markup+ External Finance	Markup+ Tangibility
$\Delta \ln RER_{ct}$	0.459*** (0.045)	0.310*** (0.045)	0.431*** (0.045)	1.419*** (0.058)
$\Delta \ln RGDP_{ct}$	0.329*** (0.106)	0.376*** (0.106)	0.389*** (0.106)	0.343*** (0.106)
$\Delta \ln RER_{ct} * Markup_{jt-1}$	-0.073** (0.030)	-0.157*** (0.030)	-0.163*** (0.031)	-0.112*** (0.030)
$\Delta \ln RER_{ct} * FPC_j$		0.414*** (0.011)		
$\Delta \ln RER_{ct} * ExtFin_j$			1.232*** (0.034)	
$\Delta \ln RER_{ct} * Tang_j$				-3.692*** (0.139)
$Markup_{jt-1}$	0.013** (0.006)	0.009 (0.006)	0.007 (0.006)	0.012** (0.006)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1411106	1411106	1411106	1411106

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1%. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A shows the effects of markup and credit constraints on import ERPT while panel B shows their effects on export ERPT. Panel B is shown in Appendix 14. All regressions include firm-product-country fixed effects and year fixed effects.

The explanation for import price absorptive capacity seems more complicated than which for export ERPT. BMM [Berman, Martin and Mayer \(2012\)](#) documents that more productive firms react to depreciation (or appreciation) by adjusting more markup and less export volume, keeping local market prices relatively stable, which means a less complete pass-through. This explanation hinges on endogenous markup over marginal costs where less elastic demand allows them to adjust markups more extensively during currency fluctuations. However, on the import side, other factors influence the sourcing capacity upon exchange rate movements. In any case, the effects of credit constraints are not offset or replaced by the sales factors, so the conclusions in section 4.2 about credit constraints remain valid.

6.2 Firm Heterogeneity in Market Share

In this section, we first provide the regression results of equation 5 with the market share and its square term constructed in section 3.2. The results are presented in Table 10. The coefficient estimates for β_3 and β_4 can be used to map out the ERPT–import market share relationship.

In columns (1) and (2), we add the primary and quadratic terms of market share to the baseline estimation of ERPT in turn. In columns (3)-(4), we further include the effects of external finance dependence and asset tangibility on top of column (2). Panel B shows the results for the export side. We find that there is evidence of a negative relationship between import pass-through and market share; however, the coefficient for the linear interaction term is positive, while the strong negative effect lies on the squared interaction term, suggesting some curvature in this relationship. The effect of market share on export ERPT is not significant.

In addition, we also perform group regressions by market share quartile and report the results in Table 11. Columns (1)-(4) show the import exchange rate pass-through for importers within each quartile of the market share distribution (0-25%, 25%-50%, 50%-75%, 75%-100%), respectively. Panel B in turn shows the results for exporters within different quartiles in the market share distribution. Results for each quartile group with interaction terms of credit constraints are not provided here but are available upon request. Roughly speaking, the import exchange rate pass-through and market share show a hump-shaped (inverted U-shaped) relationship, and the export exchange rate pass-through coefficient (weakly) decreases with the market share (towards complete price pass-through in upper quartiles).

Compared with the literature, [Auer and Schoenle \(2016\)](#) suggest that the direct response of prices to an exchange rate shock is U-shaped in exporter market share while [Devereux, Dong and Tomlin \(2017\)](#) supplement it by arguing that the market share of the importing firm is negatively correlated with pass-through and positively with local cur-

Table 10: Market Share and Effects of Credit Constraints on Import Exchange Rate Pass-through

	(1)	(2)	(3)	(4)	(5)
Panel A	MS	MS^2	Import FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.392*** (0.016)	0.381*** (0.016)	0.119*** (0.018)	0.220*** (0.017)	1.175*** (0.033)
$\Delta \ln RGDP_{ct}$	0.247*** (0.090)	0.251*** (0.090)	0.314*** (0.090)	0.321*** (0.090)	0.280*** (0.090)
$\Delta \ln RGDP_{ct} * MS_{ijct}$	-0.305*** (0.042)	0.155 (0.156)	0.677*** (0.156)	0.556*** (0.156)	0.546*** (0.156)
$\Delta \ln RGDP_{ct} * MS_{ijct}^2$		-0.531*** (0.173)	-0.941*** (0.173)	-0.837*** (0.173)	-0.846*** (0.173)
$\Delta \ln RER_{ct} * FPC_j$			0.379*** (0.010)		
$\Delta \ln RER_{ct} * ExtFin_j$				1.156*** (0.029)	
$\Delta \ln RER_{ct} * Tang_j$					-3.290*** (0.118)
MS_{ijpt}	-0.012 (0.012)	-0.009 (0.012)	-0.002 (0.012)	-0.004 (0.012)	-0.003 (0.012)
Year FE	Yes	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes	Yes
Observations	1792020	1792020	1792020	1792020	1792020

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5%, and 1%. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (3)-(5) use different measures of credit constraints calculated using U.S. data. Panel A shows the effects of market share and credit constraints on import ERPT while panel B shows their effects on export ERPT. Panel B is shown in Appendix 15. All regressions include firm-product-country fixed effects and year fixed effects.

Table 11: Estimations of Exchange Rate Pass-Through by Market Share Quartile

	(1)	(2)	(3)	(4)
Panel A		Import		
	1st	2nd	3rd	4th
$\Delta \ln RER_{ct}$	0.222*** (0.056)	0.378*** (0.042)	0.404*** (0.032)	0.242*** (0.020)
$\Delta \ln RGDP_{ct}$	-0.371 (0.324)	0.444* (0.241)	0.314* (0.179)	-0.054 (0.129)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	372447	450728	492016	476829
Panel B		Export		
	1st	2nd	3rd	4th
$\Delta \ln RER_{ct}$	0.101*** (0.023)	0.096*** (0.014)	0.032*** (0.010)	0.007 (0.008)
$\Delta \ln RGDP_{ct}$	-0.099 (0.188)	0.082 (0.104)	-0.062 (0.070)	-0.054 (0.052)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	367524	464827	508742	452881

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Results for each quartile group with interaction terms of credit constraints are not provided here but are available upon request. All regressions include firm-product-country fixed effects and year fixed effects.

rency pricing (LCP). The insignificant relationship we find between export exchange rate pass-through and market share may be because China’s original export exchange rate pass-through is nearly complete. Yet, the hump-shaped relationship between Chinese importers is interesting and worthy of further discussion.

6.3 Firm Heterogeneity in Ownership

An important feature of the financing environment in China is that links to the government and ease of access to credit are highly correlated. Therefore, ownership may be another potential factor affecting credit constraints, as state-owned enterprises (SOEs), domestic private firms, foreign-owned firms and joint ventures in China have different credit features in addition to their heterogeneous credit demand at the industry level. In this section, we check different types of firms and see whether their ownership types would affect the current main results. First, we provide baseline estimations of exchange rate pass-through for those firms with different ownership. Next, we check whether credit needs have effects with the same direction on exchange rate pass-through among all those firms.

Table 12 shows the results for importers of different ownership types and Table 16 shows the results for exporters. From the baseline coefficients in columns (1) and (5) in both panels of Table 12, we find that state-owned enterprises (SOEs) have the most complete import exchange rate pass-through around 62%, while domestic private enterprises have the most incomplete import pass-through at about 14.6%. Firms with foreign capital (including multinational enterprises and joint ventures) perform in the middle, with 31.6% and 40.2% import ERPT respectively. As for the impact of industry-level credit demand constraints on exchange rate pass-through (columns (2)-(4) and (6)-(8) in both panels), the signs and significance of regression coefficients for all types of firms are the same as in the previous main results. This suggests that although ownership type itself can cause differences among firms, the effect of credit constraints on exchange rate pass-through cannot be ignored in all types of firms.

6.4 Discussion on Import-export Linkage

Following the analysis of two-way traders, we briefly discuss the potential relationship between import and export pass-through of exchange rate shocks here. On the one hand, for two-way traders, export exchange rate pass-through could act as a ”pressure-reducing valve” for import price pass-through. When a firm has the ability to pass more exchange rate fluctuations to destination prices, it has more room to absorb price fluctuations of imported inputs. In other words, the firm-level export pass-through will have a positive effect on all product-level import pass-throughs of the same firm.

Table 12: Estimations with Importers of Different Ownership Types

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A-1	State-owned Enterprises				Domestic Private Enterprises			
	Baseline	FPC	External Finance	Tangibility	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.620*** (0.177)	0.462** (0.186)	0.488*** (0.184)	1.339*** (0.360)	0.146** (0.074)	0.053 (0.078)	0.100 (0.075)	0.682*** (0.169)
$\Delta \ln RGDP_{ct}$	-1.675 (1.257)	-1.846 (1.258)	-1.893 (1.259)	-1.747 (1.257)	0.195 (0.525)	0.212 (0.525)	0.189 (0.525)	0.236 (0.525)
$\Delta \ln RER_{ct} * FPC_j$		0.252*** (0.095)				0.181*** (0.049)		
$\Delta \ln RER_{ct} * ExtFin_j$			0.786*** (0.299)				0.448*** (0.145)	
$\Delta \ln RER_{ct} * Tang_j$				-2.761** (1.205)				-2.069*** (0.588)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm-product -country FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	17537	17537	17537	17537	88729	88729	88729	88729
Panel A-2	Multinational Enterprises				Joint Ventures			
	Baseline	FPC	External Finance	Tangibility	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.316*** (0.021)	0.008 (0.022)	0.121*** (0.021)	1.342*** (0.044)	0.402*** (0.024)	0.256*** (0.026)	0.322*** (0.024)	0.919*** (0.051)
$\Delta \ln RGDP_{ct}$	0.516*** (0.117)	0.511*** (0.117)	0.504*** (0.117)	0.519*** (0.117)	0.086 (0.150)	0.167 (0.150)	0.189 (0.150)	0.107 (0.150)
$\Delta \ln RER_{ct} * FPC_j$		0.501*** (0.013)				0.232*** (0.015)		
$\Delta \ln RER_{ct} * ExtFin_j$			1.625*** (0.041)				0.671*** (0.044)	
$\Delta \ln RER_{ct} * Tang_j$				-4.138*** (0.157)				-2.102*** (0.185)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm-product -country FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	1067958	1067958	1067958	1067958	617796	617796	617796	617796

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A-1 and panel A-2 show results of import ERPT for state-owned enterprises, domestic private enterprises, multinational enterprises, and joint venture firms, respectively. Panel B-1 and panel B-2 show the results of export ERPT for those firms. Panel B is shown in Appendix 16. All regressions include firm-product-country fixed effects and year fixed effects.

On the other hand, big importers are also big exporters (Amiti, Itskhoki and Konings (2014)). Therefore, advantages in some firm characteristics, either explicit ones such as size, market share, or productivity, or implicit ones like foreign networks may lead them to have greater bargaining power on both the import side and the export side and thus cause less complete export and import price pass-through at the same time.

To study whether and how those two channels affect the import exchange rate pass-through, ideally we need to control export price pass-through when calculating import price pass-through for all two-way traders. However, it is not available to estimate the price pass-through of each individual firm. We could only check potential influential factors individually using current strategies. However, this provides a future direction for studies to determine the factors of import pass-through.

6.5 Discussion on the Trend of China's Exchange Rate Pass-through

This article focuses on the horizontal comparison of import and export exchange rate pass-through and its causes. In the future, we can extend our methodology to reveal the time-series trend by varying periods. Actually, in our preliminary research, we found that China's export exchange rate pass-through has shown an overall downward trend (to be more incomplete) during the 2000s while the trend of China's import pass-through is vague. However, a more detailed investigation requires a longer panel covering recent years.

Devereux, Dong and Tomlin (2017) attempt to study how changes in import market shares over time may be related to changes in aggregate exchange rate pass-through over time. They run weighted rolling regressions on 12-month windows moving up by month covering 70 months. Although they find no perfect coincidence between the increase in import market share of large importers and the decrease in pass-through, the general trend does show that the aggregate pass-through is incomplete when the import market share of large importers is large. Therefore, we have reason to suspect that the evolution in China's exchange rate pass-through over time may also be due to the concentration of market share, that is, large exporters and importers occupy more market shares and may thus have stronger bargaining power.

To further ask whether the trend of exchange rate pass-through could be at least partially affected by credit constraints, there are two possible channels to discuss. First, the credit constraints on Chinese exporters are gradually loosening. It may be because of the decreasing credit needs of Chinese exporters or the improvement of the immature financial market in China. Second, China's exports switch from more credit-constrained to less-constrained industries. Credit-constrained firms find it harder to survive in export markets (extensive margin) or export less in value (intensive margin).

7 Conclusion

In this paper, we provide evidence at a highly disaggregated level for the incomplete import exchange rate pass-through in China. Our research contributes to the literature by revealing how importers' characteristics, especially the degree of financial constraints that they face, affect exchange rate price pass-through patterns. Utilizing unit value information from Chinese Customs and comparing imports with exports, we find that (1) the average import price pass-through in China is around 35-40%, far below the 95% export price pass-through; (2) for firms in industries with more stringent credit constraints, both import and export exchange rate pass-through tend to be more complete; (3) import source diversity can effectively reduce import price pass-through and offset the effects of credit constraints. The main novelty of our empirical strategy is to focus on the role of importers. We believe that micro import price pass-through measures China's ability to withstand risks in the international trade market from a new perspective.

There are several directions for future improvement. First, we need to explore the underlying mechanism by which credit constraints affect exchange rate pass-through. We only verify this effect based on a reduced-form approach at this stage. Even after controlling for some potential channels claimed by literature, we are not yet clear about how the remaining effects of credit constraints work. Future work should build a structural model to identify the detailed channels. Second, we could study how a firm's import and export behaviors influence each other. The dominance of two-way traders in China's international trade volume is a key fact that we cannot ignore. Adjustments on the import side and export side are two sides of the same coin for companies to face exchange rate shocks. Third, we should pay attention to the trend of China's exchange rate pass-through over time. The trend may reflect changing market power of Chinese firms and their patterns of pricing to market behaviors. Ideally, we expect to distinguish the contribution of each factor to the trend in exchange rate pass-through.

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A Extra Measures

A.1 Firm-level markup

In the discussion, we argue that credit constraints will affect the “absorptive capacity” of exchange rate shocks other than the firm’s attributes in sales. In the following work, we will control markup to test the conjectures concretely. Referring to [Brooks, Kaboski and Li \(2021\)](#), even without direct measures of prices and marginal cost, we can still estimate the firm-level markup, using the structural assumptions of [De Loecker and Warzynski \(2012\)](#) (DLW hereafter) and GMM estimation method.

Simply put, DLW (2012) [De Loecker and Warzynski \(2012\)](#) derives the firm-specific markup as the ratio of an input factor’s output elasticity to its firm-specific factor payment share $\mu_t = \theta_t^X (\alpha_t^X)^{-1}$, where α_t^X is the share of expenditures on input X in total sales and θ_t^X denotes the output elasticity on an input X. The major difficulty is calculating the firm-specific output elasticity concerning materials, which requires estimating firm-specific production functions. We apply the methodology of [Akerberg, Caves and Frazer \(2015\)](#) to address the endogeneity of inputs, assuming a 3rd-order translog gross output production function in capital, labor, and material inputs:

$$y_t = \beta_k k_t + \beta_l l_t + \beta_m m_t + \beta_{k2} k_t^2 + \beta_{l2} l_t^2 + \beta_{m2} m_t^2 + \beta_{kl} k_t l_t + \beta_{km} k_t m_t + \beta_{lm} l_t m_t + \beta_{k3} k_t^3 + \dots + \omega_t + \epsilon_t.$$

In practice, we need to construct four production variables in log form: real output value y_t , persons engaged l_t , real fixed assets at current value k_t , and real material inputs m_t . Real output values are deflated by output deflators, while real fixed assets and real material inputs are deflated by investment deflators and input deflators, respectively. The deflators are constructed as in [Brandt, Van Biesebroeck and Zhang \(2012\)](#).

A.2 Market share

In addition to the extensive diversity measured by the number of import sources or export markets, we also use a firm’s share in a specific import or export market to describe its intensive competitiveness.

Following [Amiti, Itskhoki and Konings \(2014\)](#) and [Devereux, Dong and Tomlin \(2017\)](#), we define import market share as a firm’s value share in the import market, within a given HS6 product category. Therefore, a single firm can have multiple import market shares for multiple products. Our definition of import market share is also year specific, and so a firm’s import market share can vary over time.

$$S_{ijct}^D \equiv \frac{v_{ijct}^D}{\sum_{j' \in J_{ict}} v_{ij'ct}^D}$$

where $D \in \{\text{Import, Export}\}$. The capital letter J denotes the set of potential competitors in the same product-specific market.

The destination-specific export market share proxy is similarly defined as the value share of a firm relative to all Chinese exporters in our sample who export the same product to the same market. Since we only have data from China Customs, our export market share S_{ijct}^D is relative to other Chinese firms. The external competitive stance in a particular sector-destination pair is also common for all Chinese exporters in that country and hence our measure captures all relevant variation in market share between firms in our sample.

B Extra Tables

Table 13: Export Markets and Effects of Credit Constraints on Export Exchange Rate Pass-Through

	(1)	(2)	(3)	(4)
Panel B	Export			
	#Markets	#Markets+ FPC	#Markets+ External Finance	#Markets+ Tangibility
$\Delta \ln RER_{ct}$	0.055*** (0.007)	0.058*** (0.008)	0.056*** (0.007)	0.028 (0.021)
$\Delta \ln RGDR_{ct}$	-0.079** (0.037)	-0.080** (0.037)	-0.079** (0.037)	-0.081** (0.037)
$\#Market_{ijt}$	-0.001*** (0.000)	-0.001*** (0.000)	-0.001*** (0.000)	-0.003*** (0.001)
$\Delta \ln RER_{ct} * FPC_j * \#Market_{ijt}$		-0.000* (0.000)		
$\Delta \ln RER_{ct} * FPC_j$		-0.009 (0.007)		
$\Delta \ln RER_{ct} * ExtFin_j * \#Market_{ijt}$			-0.001 (0.001)	
$\Delta \ln RER_{ct} * ExtFin_j$			-0.019 (0.020)	
$\Delta \ln RER_{ct} * Tang_j * \#Market_{ijt}$				0.006** (0.003)
$\Delta \ln RER_{ct} * Tang_j$				0.102 (0.075)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1793974	1793974	1793974	1793974

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A is shown in Table 5. All regressions include firm-product-country fixed effects and year fixed effects

Table 14: Heterogeneous Markup and Effects of Credit Constraints on Export Exchange Rate Pass-through

	(1)	(2)	(3)	(4)
Panel B	Markup	Export Markup+ FPC	Export Markup+ External Finance	Export Markup+ Tangibility
$\Delta \ln RER_{ct}$	-0.046** (0.021)	-0.042** (0.021)	-0.048** (0.021)	-0.112*** (0.027)
$\Delta \ln RGDP_{ct}$	-0.074* (0.043)	-0.076* (0.043)	-0.075* (0.043)	-0.076* (0.043)
$\Delta \ln RER_{ct} * Markup_{jt-1}$	0.061*** (0.016)	0.066*** (0.016)	0.067*** (0.016)	0.063*** (0.016)
$\Delta \ln RER_{ct} * FPC_j$		-0.023*** (0.005)		
$\Delta \ln RER_{ct} * ExtFin_j$			-0.061*** (0.016)	
$\Delta \ln RER_{ct} * Tang_j$				0.243*** (0.062)
$Markup_{jt-1}$	-0.005* (0.003)	-0.005 (0.003)	-0.004 (0.003)	-0.005* (0.003)
Year FE	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes
Observations	1411116	1411116	1411116	1411116

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A is shown in Table 9. All regressions include firm-product-country fixed effects and year fixed effects.

Table 15: Market Share and Effects of Credit Constraints on Import Exchange Rate Pass-through

	(1)	(2)	(3)	(4)	(5)
Panel B	MS	MS^2	Export FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.035*** (0.006)	0.037*** (0.007)	0.045*** (0.007)	0.039*** (0.007)	-0.022 (0.015)
$\Delta \ln RGDP_{ct}$	-0.079** (0.037)	-0.079** (0.037)	-0.080** (0.037)	-0.079** (0.037)	-0.080** (0.037)
$\Delta \ln RGDP_{ct} * MS_{ijct}$	-0.008 (0.012)	-0.040 (0.049)	-0.043 (0.049)	-0.036 (0.049)	-0.051 (0.049)
$\Delta \ln RGDP_{ct} * MS_{ijct}^2$		0.035 (0.050)	0.037 (0.050)	0.033 (0.050)	0.041 (0.050)
$\Delta \ln RER_{ct} * FPC_j$			-0.019*** (0.004)		
$\Delta \ln RER_{ct} * ExtFin_j$				-0.045*** (0.013)	
$\Delta \ln RER_{ct} * Tang_j$					0.229*** (0.053)
MS_{ijpt}	0.070*** (0.004)	0.070*** (0.004)	0.070*** (0.004)	0.070*** (0.004)	0.070*** (0.004)
Year FE	Yes	Yes	Yes	Yes	Yes
Firm-product-country FE	Yes	Yes	Yes	Yes	Yes
Observations	1793974	1793974	1793974	1793974	1793974

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (3)-(5) use different measures of credit constraints calculated using U.S. data. Panel A is shown in Table 10. All regressions include firm-product-country fixed effects and year fixed effects.

Table 16: Estimations with Exporters of Different Ownership Types

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel B-1	State-owned Enterprises				Domestic Private Enterprises			
	Baseline	FPC	External Finance	Tangibility	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	-0.012 (0.034)	-0.032 (0.036)	-0.034 (0.035)	0.028 (0.089)	0.031*** (0.011)	0.032*** (0.011)	0.030*** (0.011)	-0.008 (0.032)
$\Delta \ln RGDP_{ct}$	0.310 (0.225)	0.301 (0.225)	0.298 (0.225)	0.308 (0.225)	0.079 (0.076)	0.079 (0.076)	0.079 (0.076)	0.078 (0.076)
$\Delta \ln RER_{ct} * FPC_j$		0.034 (0.021)				-0.004 (0.009)		
$\Delta \ln RER_{ct} * ExtFin_j$			0.116** (0.058)				0.008 (0.026)	
$\Delta \ln RER_{ct} * Tang_j$				-0.146 (0.304)				0.143 (0.109)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm-product -country FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	26894	26894	26894	26894	484522	484522	484522	484522
Panel B-2	Multinational Enterprises				Joint Ventures			
	Baseline	FPC	External Finance	Tangibility	Baseline	FPC	External Finance	Tangibility
$\Delta \ln RER_{ct}$	0.032*** (0.009)	0.045*** (0.009)	0.038*** (0.009)	-0.048** (0.024)	0.031*** (0.009)	0.042*** (0.010)	0.036*** (0.009)	-0.026 (0.025)
$\Delta \ln RGDP_{ct}$	-0.183*** (0.059)	-0.187*** (0.059)	-0.185*** (0.059)	-0.188*** (0.059)	-0.080 (0.062)	-0.079 (0.062)	-0.078 (0.062)	-0.080 (0.062)
$\Delta \ln RER_{ct} * FPC_j$		-0.028*** (0.007)				-0.026*** (0.007)		
$\Delta \ln RER_{ct} * ExtFin_j$			-0.072*** (0.023)				-0.078*** (0.022)	
$\Delta \ln RER_{ct} * Tang_j$				0.308*** (0.086)				0.218** (0.089)
Year FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Firm-product -country FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	711685	711685	711685	711685	570873	570873	570873	570873

* Standard errors in parentheses; *, **, and *** indicate significance at 10%, 5% and 1% levels. The dependent variable is the price change $\Delta \ln P_{ijct}$. Columns (2)-(4) use different measures of credit constraints calculated using U.S. data. Panel A-1 and A-2 are shown in Table 12. All regressions include firm-product-country fixed effects and year fixed effects.