Family Matters: Globalization, Management and Firm Productivity[*]

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Abstract

How does globalization affect firm management and productivity? I investigate this question using a product-specific import competition shock in India, focusing on family-managed firms: the predominant form of corporate governance in the developing world. Utilizing a novel manager-firm matched dataset, I analyze tenure records and family ties of over six million company directors. Employing an event-study approach, I find that firms exposed to import competition replace family managers with unrelated professional executives, increasing firm productivity by 20 percent. To evaluate the aggregate implications, I construct a model of industrial equilibrium where family firms balance non-pecuniary private benefits of family management against higher profits achievable through professional management. Consistent with the empirical findings, the model predicts that import competition drives the least productive family firms to adopt professional management to avoid exit. These results underscore managerial restructuring as a critical mechanism linking competition to productivity gains.

Keywords: trade, competition, management, firm productivity, corporate governance, family firms, growth
JEL Codes: D24, F10, G30, L20, O32, O33 O40

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1   Introduction

It is commonly believed that the competitive pressures of globalization can serve as catalysts to boost firm productivity and reduce operational slack. A key channel through which firms can achieve higher productivity is managerial innovation and organizational change. However, in the absence of such pressures, the incentive to pursue these improvements may wane. As early as 1776, Adam Smith remarked that *"monopoly... is a great enemy to good management,"* capturing the intuition that competition plays a disciplining role in managerial behavior (Smith, 1776). Alfred Marshall later echoed a similar view, noting that many firms postpone productivity-enhancing changes until external forces compel them *"to exert themselves to the utmost to invent improved methods"* (Marshall, 1926). Despite their enduring influence, these ideas remain empirically underexplored. In today's global context– marked by a retreat from free trade and a renewed emphasis on protectionism and industrial policy– the question of how competition shapes internal firm organization has gained renewed urgency.

In this paper, I examine how globalization and import competition influence *within-firm* productivity. In particular, I demonstrate that exposure to global competition compels firms to overhaul their board of directors and top management, triggering fundamental changes in their organizational structure that drive within-firm productivity improvements. I focus on the management structures of family-managed firms in India, the predominant form of corporate governance in many developing countries (La Porta et al., 1999; Burkart et al., 2003; Morck et al., 2005).

Family firms typically emerge for two broad reasons (Bertrand and Schoar, 2006). First, in cultures with strong family values, founders may derive intrinsic satisfaction from involving kin in the management of their business. Second, in countries with weak legal institutions, family firms may represent an optimal response, as trust among family members can substitute for robust contract enforcement. In either case, unless family members are also the most talented managers, the dominance of family management often reflects a failure of meritocracy that leads to misallocation of talent *within* the firm. A substantial body of research has shown that family firms are less productive and follow poorer management practices than firms whose top managers are unrelated to the founder (Pérez-González, 2006; Bennedsen et al., 2007; Bloom et al., 2012, 2013a). In this paper, I show that trade exposure reduces talent misallocation within firms by compelling them to reassess the tradeoff between the non-monetary benefits of family management and the higher productivity potential of professional management, ultimately driving improvements in firm performance.

Prior research has not conclusively examined the role of competition and talent allocation within firms due to a combination of data limitations, empirical challenges, and theoretical constraints. First, a primary obstacle is the need for detailed data on a firm's *internal organization*; most firm-level datasets concentrate on balance sheet and financial variables, leaving information on organizational structure and management composition scarce. Second, identifying exogenous variation in competitive pressures across product markets is challenging. Even policies like trade liberalizations often generate export opportunities for domestic firms or occur alongside domestic reforms, thereby complicating efforts to isolate the impact of competition. Finally, on the theoretical side, classical price theory leaves no room for competition-induced improvements in productivity: there is no possibility for internal reallocation of resources within a firm

that is already operating optimally (Stigler, 1976).

I address these challenges in four steps. First, I identify and digitize old policy notifications from the archives of India's Commerce Ministry. This enables the analysis of a previously unexamined product-level import liberalization policy that exposed over a third of Indian manufacturing firms to heightened import competition. In the second step, I construct a novel manager-firm matched microdata with rich information on the tenure and family ties among the board of directors of firms. Third, I estimate event studies to provide a direct assessment of how trade shocks affect firms' production, organization, and productivity. In the final step, I develop a model of industry equilibrium where firms balance the non-pecuniary private benefits of employing family managers against the higher profitability achievable through professional management. For example, a firm owner may derive pleasure if their children or siblings run the firm. However, family managers may not be the most talented at running the firm. The model shows how increased import competition sharpens this trade-off between private benefits and profits, prompting low-productivity firms to professionalize their management to enhance productivity and avoid market exit. The theoretical framework also enables me to quantify the impact of within-firm productivity changes on aggregate TFP.

The foundation of my empirical analysis is a natural experiment stemming from the dismantling of India's Import License Raj in the late 1990s and early 2000s. This regime was among the most extensive protectionist policies ever implemented, effectively banning imports of consumer goods since the mid-1950s. Although the widespread economic reforms of 1991 significantly reduced tariff and non-tariff barriers for intermediate goods (Hasan et al., 2007; Topalova, 2010; Goldberg et al., 2010a; Topalova and Khandelwal, 2011; DeLoecker et al., 2016), non-tariff measures– in the form of quantitative restrictions (QRs)– remained on virtually all *final consumer goods*, accounting for roughly 30 percent of all tariff lines, until 2001.

These protections were enabled by specific provisions in the General Agreement on Tariffs and Trade (GATT) Articles that permitted developing countries to impose QRs to safeguard their external financial stability. For nearly half a century, these QRs insulated Indian manufacturing firms from foreign competition. However, in the late 1990s, the United States challenged India's continued use of these provisions by approaching the WTO, arguing that India's growing foreign reserves no longer justified such measures. Joined by several other countries, this challenge culminated in a WTO ruling– based on an independent IMF assessment affirming India's robust external position– that required the removal of QRs on balance-of-payments grounds across *all goods*. Consequently, nearly 3,000 goods at the HS 8-digit level, primarily final consumer goods, were liberalized.

Several features make this policy ideal for empirical analysis. First, the QR removal was externally imposed, with its timing and scope determined by the WTO and supported by an independent IMF assessment, minimizing domestic political or strategic influences. Second, unlike the comprehensive domestic reforms of 1991, the late-1990s QR removal occurred in relative isolation, reducing confounding effects and strengthening causal attribution to the policy change. Third, by targeting specific, narrowly defined HS-8 digit products (mostly final consumer goods), the reform generates quasi-experimental variation in import competition, enabling the use of an event study de-

sign. The focus on consumer goods ensures that the primary mechanism at work is a demand shock stemming from heightened import competition in consumer goods markets rather than cost reductions from cheaper intermediate inputs. Fourth, as a unilateral trade liberalization measure, it enhanced foreign access to the Indian market without concurrently improving export conditions for domestic firms, further isolating the impact of increased import competition. Finally, because this trade liberalization took place in the late 1990s and early 2000s, high-quality firm-level panel data are readily available– unlike the 1991 reforms in India or comparable policies in the 1980s and 1990s in other regions.

To examine how import competition influences firm-level operations and management, I assembled three novel datasets that, when integrated with comprehensive panel data on large and medium-sized firms, form the core of this analysis. To build the first of these datasets, I digitized archival documents from the Ministry of Commerce to create a detailed dataset of products subject to QRs, capturing the exact dates of their removal. This process enabled me to construct a novel HS 8-digit product-level database that chronicles QRs in India during the 1990s and their subsequent removal in the late 1990s and early 2000s.

A critical challenge arose from the fact that government notifications on QRs use HS product codes, which differ from the product nomenclature in the firm panel data. To address this, in the second step, I developed a novel concordance linking almost 3000 HS 8-digit codes to over 6,000 product categories in the firm panel data, thereby providing a precise measure of a firm's exposure to import competition– determined by whether a firm produces an affected product. Unlike previous studies that typically define exposure at broad industry levels– largely due to data constraints and the lack of detailed concordances– this approach identifies which firms produce the affected products directly, offering a much more granular and accurate measure of the import competition they face.

Finally, I obtained access to new administrative data from the Ministry of Corporate Affairs (MCA), which includes records on over 6 million company directors with complete tenure histories dating back to the 1970s. Researchers studying family firms often rely on last-name matching to infer family ties. However, this method can be imprecise in India, where last names may indicate caste rather than actual familial relationships. In contrast, the dataset used here includes managers' fathers' names. This additional detail allows for a more accurate identification of family connections within firm boards. By merging these administrative records with the firm panel data, I build a comprehensive dataset that captures firm production, product scope, and internal organizational structures– paving the way for a rigorous analysis of how the trade shock impacted firm operations, size, and most significantly, their internal organization and management structures.

The empirical analysis proceeds in three steps. First, I study aggregate HS-6 digit import and export data from the Indian customs to estimate the direct impact of QR removal on import competition in India. The event-study analysis reveals that, following the reform, treated products experienced a dramatic and persistent surge in imports– both in value and quantity. By the third year, the import value for products affected by the policy exceeded that of the control group by over 50 percent, while import quantities increased by roughly 1 log point, equivalent to over 150 percent. Notably, these

changes were observed only for imports, as export outcomes remained largely unaffected, underscoring the unilateral nature of the reform.

Second, at the firm level, the intensified import competition precipitated significant contractions in domestic firms' financial performance and scale. Both family and non-family firms producing QR-affected goods experienced sharp declines in total revenues relative to firms in the control group– about 20 percent by the third year and nearly 50 percent by the eighth year post-reform– accompanied by substantial reductions in operating profit margins. The shock also triggered a persistent decline in cost components, including labor and raw materials expenses, reflecting a systematic cutback in operations. These reductions translated into notable contractions in firm size, with treated firms' total assets initially falling by approximately 10 percent and nearly 30 percent over time. Importantly, these financial and scale adjustments were not solely driven by firm exit, as similar patterns emerge even when analyses are restricted to surviving firms.

Beyond the measurable contractions in firm scale and financial performance, the QR removal catalyzed a marked transformation in managerial turnover and corporate governance of family-managed firms that were exposed to import competition. Faced with deteriorating financial performance, these firms began to reconfigure their top management structures. The evidence shows a clear reduction in the share of family members on executive boards– dropping by almost 10 percentage points within a few years– with a corresponding increase in non-family professional managers. This managerial turnover was especially pronounced among firms that were less productive before the reform, indicating a selective move towards professionalization where it was most needed. Furthermore, firms that replaced family managers with external professionals experienced significant productivity improvements, with revenue and quantity productivity measures diverging positively from those that retained family-dominated management. This correlation between professionalizing management and productivity enhancements is suggestive of the role that external managerial talent may play in enabling firms to effectively respond to heightened import competition, unlock latent efficiencies, and drive long-term performance improvements. Concurrently, these professionalizing firms also reduced their average output prices, suggesting that the adoption of external expertise was correlated with cost-cutting. These results shed light on the broader question of why firms in more competitive markets exhibit higher productivity– a phenomenon commonly associated with what Leibenstein (1966) termed *"X-efficiency."* Yet, the mechanisms driving the correlation between productivity and competition remain poorly understood (Backus, 2020). Understanding these dynamics has implications across various economic applications.

If hiring external managers is productivity-enhancing, why do family firms wait for import competition to upgrade their management to outside professionals? To answer this question and quantify the impact of these firm-level adjustments, I develop a model of industrial equilibrium in which the allocation of top managerial talent within the firm takes center stage. I consider a closed economy operating under monopolistic competition, where firms produce differentiated products that together form a composite final good. Each firm begins as a family firm, with entrepreneurs enjoying *non-monetary private benefits* from retaining family control over management. For example, an entrepreneur may derive pleasure if their children or siblings run the firm. Such amenity value of family control of firm management has a long tradition in the

corporate finance literature ([Demsetz and Lehn, 1985](); [Burkart et al., 2003](); [Bertrand and Schoar, 2006]()). Naturally, the firm may only enjoy such private benefits as long as the firm is alive. If the firm exits, the firm's owner loses any private benefits associated with running a family firm. A firm can choose to upgrade its management by hiring external professionals– a decision that boosts its productivity by some constant multiplier. However, professionalization comes at a cost: the firm's owner loses all private benefits associated with running the business as a family firm.

The model predicts a dual selection mechanism into managerial upgradation that hinges on the firm's baseline productivity. Specifically, ex-ante low-productivity firms– hereafter referred to as *laggards*– professionalize management primarily as a *survival mechanism* because their weak performance makes them vulnerable to competition-induced exit. This form of negative selection arises when the urgency to remain viable outweighs the benefits of retaining family control. In contrast, ex-ante high-productivity firms–termed *frontier firms*– professionalize management due to its *efficiency-boosting mechanism*, whereby the gains from professional management further enhance their already high level of productivity, as in [Bustos (2011)](). Firms with intermediate ex-ante productivity levels, however, tend to retain family management since the incremental benefits of upgrading their management do not fully compensate for the loss of valuable non-monetary private benefits.

Import competition alters firms' incentives to professionalize management in a heterogeneous manner. For laggard firms, survival becomes even more challenging as increased competition erodes profits, prompting them to cede managerial control to professional managers to avoid exit. Although this shift entails the loss of cherished family benefits, it is essential for preserving the firm. Conversely, for frontier firms, professional management becomes less appealing under the trade shock, as lower profits make it difficult to justify the loss of the private benefits associated with family management. Thus, the model predicts that the least productive family firms are most likely to upgrade their management in response to the trade shock– a pattern that is mirrored in the empirical results.

Contribution to the Literature: Existing research on trade-induced productivity improvements has predominantly focused on production-side adjustments. These include changes in imported inputs ([Amiti and Konings, 2007](); [Topalova and Khandelwal, 2011]()), shifts in product scope ([Bernard et al., 2011]()), or innovation and technology upgrading ([Bustos, 2011](); [Bloom et al., 2016](); [Hombert and Matray, 2018](); [Perla et al., 2021]()). However, many of these studies focus on trade liberalization episodes that provide export opportunities to domestic firms rather than assess the impact of import competition. Moreover, with the notable exception of [Chen and Steinwender (2021)](), relatively little attention has been devoted to how managerial incentives and organizational structures respond to trade shocks. In contrast, this paper demonstrates that import competition also induces important changes in firms' internal management structures. Specifically, heightened competitive pressures lead to increased managerial turnover and a greater reliance on external professional managers, mechanisms that can unlock significant aggregate productivity gains. Moreover, while family-managed firms constitute a dominant form of corporate governance in many developing countries, they have received relatively little attention in the trade literature ([Atkin and Khandelwal, 2020](), [2022]()). This study fills that gap by directly examining how family

firms respond to trade shocks. Related literature in trade focuses on how organizational hierarchies respond to trade (Caliendo and Rossi-Hansberg, 2012; Marin and Verdier, 2014). This study complements these studies by uncovering additional dimensions through which trade reshapes organization structures.

A substantial body of work has documented the critical role of management practices in enhancing firm productivity (Bloom and Van Reenen, 2007; Bloom et al., 2012, 2021). For instance, Bloom et al. (2013b) finds in a randomized controlled trial that adopting improved management practices can raise productivity by as much as 17 percent among large and medium-sized Indian manufacturing firms. This naturally raises the question of why firms do not adopt better practices on their own. Prior research suggests that weak product market competition and the prevalence of family members in managerial roles may be to blame (Bloom and Van Reenen, 2007). In this paper, I combine both forces to tackle this question head-on. Moreover, past research indicates that trust issues between owners and external professionals can constrain the effective span of control. In support of this view, Akcigit et al. (2021) calibrate a model showing that inefficiencies in the delegation environment account for roughly 11 percent of the income per capita gap between the United States and India. While my work is complementary to theirs, it differs in mechanism: in my model, firms professionalize to avoid exit, whereas, in theirs, professionalization is driven by the need to overcome the constraints imposed by the limited span of control of family management.

The broader debate on the link between competition and productivity dates back to Hicks (1935)'s remark that *"the best of all monopoly profits is a quiet life."* Subsequent studies have proposed various mechanisms, including competition as an incentive to mitigate agency problems through the threat of liquidation (Schmidt, 1997), improved information flow that alleviates moral hazard (Hart, 1983; Nalebuff and Stiglitz, 1983), and the changing incentives for cost reduction in settings with endogenous market structures (Raith, 2003). This paper is complementary to these perspectives– particularly resonating with Schmidt (1997)– by illustrating how increased import competition compels firms to forgo non-monetary benefits associated with family management in favor of strategies that ensure survival.

Finally, there is rich literature on family firms. For example, Bertrand and Schoar (2006) provide a comprehensive survey, while Burkart et al. (2003) develop a model highlighting the role of the legal environment in family firm succession. Empirical studies have generally found that family-controlled firms underperform compared to their professionally managed counterparts (Pérez-González, 2006; Lemos and Scur, 2019; Bandiera et al., 2020), and Bennedsen et al. (2007) use instrumental variables to show that dynastic succession is linked to lower productivity in Denmark. By analyzing family firms during a period of dramatic import competition following the QR-removal policy, this paper offers new insights into managerial succession and turnover under crisis conditions– demonstrating that, when the going gets tough, family firms are more likely to seek external managerial talent to preserve firm viability.

The remainder of the paper is organized as follows. Sections 2 and 3 provide the details of the policy setting and data construction, setting the stage for the event study analysis. Section 4 lays out the empirical strategy and Section 5 presents event study results. Section 6 introduces the theoretical framework. Section 7 concludes.

2   Policy Background

In 1947, following independence from British rule, India's economic planning was characterized by a strong desire for self-reliance and minimal dependence on the West for its development objectives. A key outcome of this strategy was the implementation of a comprehensive import substitution and licensing regime, which involved direct control over foreign exchange utilization by Indian firms and households. A balance of payments (BOP) crisis in 1957 further intensified these import controls. Instead of relying on price controls such as tariffs, the Indian government employed quantitative restrictions (QRs) as its main policy instrument. A small group of bureaucrats in Delhi was responsible for allocating scarce foreign exchange across different sectors of the economy and among firms within each industry.[1] Imports of consumer goods were even more heavily regulated and virtually eliminated (Bhagwati and Srinivasan, 1975; Krueger, 2010). A complex web of overlapping agencies responsible for certifications and license issuance managed this process. During this period, there was a lucrative premium on import licenses, and foreign consumer goods were essentially absent from the market.

The framework of India's restrictive trade practices, particularly the use of QRs, was facilitated by specific exceptions within international trade agreements. Although the General Agreement on Tariffs and Trade (GATT) fundamentally prohibited QRs under Article XI, it provided crucial exceptions that India utilized. Article XVIII:B of the GATT allowed countries in the "early" stages of development to impose QRs to *"safeguard [their] external financial position and ensure a level of reserves adequate for the implementation of their program of economic development"*.[2] India utilized this provision of Article XVIII:B to support its QR regime since 1957 (Pursell and Sattar, 2004).

2.1   India's First Generation Trade Reform (1991)

By the 1980s, it was evident that India's regime of import-substituting industrialization had failed, yielding a per capita economic growth rate of only 1.7 percent. Although growth accelerated in the 1980s, India's public debt steadily increased throughout the decade, rendering its macro-fiscal situation vulnerable. The rising debt was exacerbated by a spike in oil prices during the Gulf War and a decline in remittances from workers in the Middle East, leading to a downgrade in India's credit rating. By 1991, India was on the brink of default. Consequently, India approached the International Monetary Fund (IMF) for emergency financing and agreed to implement macroeconomic stabilization and structural reforms.

The structural reforms of the early 1990s extended well beyond the scope of the IMF program (Krueger, 2010; Ahluwalia, 2019) and impacted several spheres of the economy. Revisions to the industrial licensing regime facilitated firm entry and capacity expansion, private firms were permitted to enter sectors previously reserved for state-owned enterprises, and foreign direct investment was eased in several industries.

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[1] For instance, firms could only obtain an import license if they demonstrated that their imports were essential for production and that the imported product was not manufactured domestically (Bhagwati and Srinivasan, 1975; Pursell and Sattar, 2004).

[2] For details, see https://www.wto.org/english/tratop_e/bop_e/bop_e.htm, accessed July 30, 2024.

In terms of trade policy, the exchange rate was devalued by over 20 percent, and both quantitative restrictions and tariffs were eased on *intermediate and capital goods*. With the removal of quantitative restrictions, tariffs became the primary restrictions on imports of these goods. Average tariffs were reduced from over 80 percent in 1990 to 36 percent by 1996 (Topalova, 2010). Several papers have studied the various impacts of these tariff reductions (Hasan et al., 2007; Topalova, 2010; Goldberg et al., 2010b; Topalova and Khandelwal, 2011).

Despite these reforms, India continued to impose stringent QRs on almost all consumer goods and a small number of intermediate products– almost 3000 products at the 8-digit HS level or 30 percent of all tariff lines (Panagariya, 2004). India justified these QRs under Article XVIII:B of the GATT, asserting that they were necessary to safeguard its external financial position due to inadequate foreign exchange reserves. QRs on consumer products were lifted a decade later and are the focus of this paper. One of the challenges in evaluating the impact of the 1991 trade reforms is precisely that they were implemented as part of a broad-based structural reform package. This makes it difficult to attribute post-policy changes in data to trade policy. While tariff changes were product-specific, as Topalova and Khandelwal (2011) note, there may be considerable complementarity between sectors that saw the highest tariff reductions and industries that benefited from other industrial reforms such as those mentioned above. This concern is less relevant to the removal of QRs during India's second generation of reforms in the late 1990s and early 2000s, as I will argue in the next section

2.2 Second Generation Trade Reform (1998-2001): Removal of Quantitative Restrictions (QRs)

While there are many studies on how India's 1991 liberalization affected various aspects of the Indian economy, there is virtually no work on the impact of continuing QRs on over 30 percent of tariff lies and the impact of their eventual removal a decade later. This is surprising given that even after the 1991 reforms, over two-thirds of India's tradable GDP remained protected by some kind of non-tariff import restrictions, most commonly QRs (Pursell and Sattar, 2004).

From 1998-2001, these remaining QRs were also removed. The impulse of this policy was external to India. In July 1997, the United States requested consultations with India under the World Trade Organization's (WTO) Understanding on Rules and Procedures Governing the Settlement of Disputes (DSU) to challenge India's QRs as being inconsistent with WTO obligations. In the following months, the US was joined by the European Communities, Switzerland, Australia, Canada, and New Zealand, leading the WTO Dispute Settlement Body to set up a panel to examine the validity of India's QRs in November 1997.The panel submitted its report a year later in December 1998. The panel ruled that India's foreign exchange reserves are adequate and "not facing a serious decline or threat" and concluded that India's QRs therefore do not constitute permissible "necessary" measures to address a weak BOP situation under Article XVIII:B.It further recommended that India remove *all* QRs that it maintains under Article VIII:B. Subsequently, India and the US mutually agreed for a phase-out plan where India agreed to remove all its outstanding BOP-related QRs by April 1, 2001.

Two features of the institutional procedure leading to the removal of India's QRs have

Figure 1: Composition of Liberalised Products By Year and Sector

(a) Liberalised Products by Year

(b) Liberalised Products by Sector
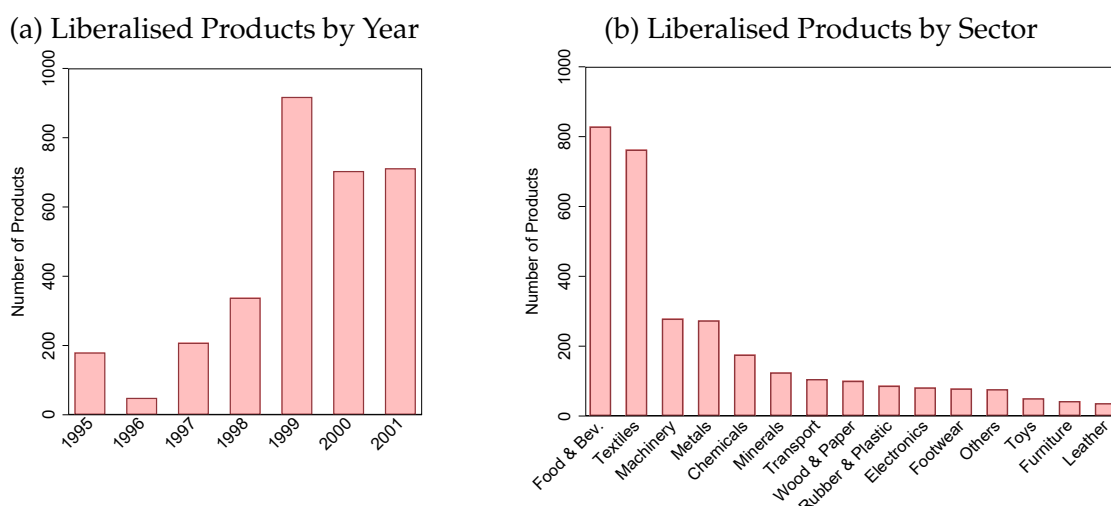
Notes: The figure displays the number of products that were liberalized over time (panel a) and across different sectors (panel b), based on a dataset created by digitizing archival policy documents from the Ministry of Commerce, Government of India, as outlined in Section 3. *Source:* Ministry of Commerce, Government of India.

important implications for the empirical strategy outlined in Section 4. First, India's case for maintaining QRs under Article XVIII:B largely hinged on showing that its BOP situation was vulnerable. For such an assessment, instead of relying on materials submitted by the disputing countries, the WTO relies almost exclusively on the independent determination of the International Monetary Fund (IMF), which is a permanent invitee in all Article XVIII consultations.[3] The IMF held that India is well placed to manage its external financial situation and that its reserves are adequate to remove all existing QRs over a relatively short period. Because the IMF was called upon to conduct this technical assessment, it is unlikely that this externally imposed policy reform was influenced by politicians or policymakers in either India or any of the disputing countries.

The second noteworthy feature of this policy is that the single technical assessment by the IMF applied uniformly to almost 3000 products on which India maintained QRs based on India's aggregate BOP position. It was an all-or-nothing approach. Neither the United States could selectively target certain products for the removal of QRs, nor could India selectively retain QRs on specific products. The policy did not allow for any selective application or exemptions. Consequently, India's loss in this dispute resulted in the removal of QRs on almost 3000 products across the board.

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[3]According to Article XV:2 of the GATT, "... *the contracting parties shall accept all findings of statistical and other facts presented by the IMF relating to foreign exchange, monetary reserves and balances of payments, and shall accept the determination of the Fund as to whether action by a contracting party in exchange matters is in accordance with the Articles of Agreement of the International Monetary Fund, or with the terms of a special exchange agreement between that contracting party and the contracting parties. The contracting parties in reaching their final decision in cases involving the criteria set forth in paragraph 2 (a) of Article XII or in paragraph 9 of Article XVIII, shall accept the determination of the Fund as to what constitutes a serious decline in the contracting party's monetary reserves, a very low level of its monetary reserves or a reasonable rate of increase in its monetary reserves, and as to the financial aspects of other matters covered in consultation in such cases.*"

3 Data

This study leverages three novel sources of data that are merged with comprehensive firm-level panel data. The construction of these data enables an in-depth analysis of how trade shocks affect both the production outcomes and the internal organization of firms.

CMIE Prowess. The primary data source is the CMIE Prowess database, which covers a substantial portion of India's formal economic activity. Firms in the Prowess database collectively account for 60 to 70 percent of the economic activity in the organized industrial sector, 75 percent of corporate taxes, and 95 percent of the excise duty collected by the Government of India (Goldberg et al., 2010a). CMIE has been compiling these data since 1988, primarily drawing on firms' annual reports and audited financial statements. The database contains information on manufacturing firms as well as on financial and non-financial services companies, spanning 775 5-digit industry codes based on the 2008 National Industrial Classification (NIC). Of these, 462 industries belong to manufacturing, which are further grouped into 20 2-digit sectors.

Prowess offers two distinctive features that have received relatively little attention from researchers. First, it provides detailed data on the product scope of each firm. Until 2011, the Companies Act (1956) required firms to report quantitative information– such as the value and quantity of production, revenues, stock, and capacity– for every product they produced. This obligation, coupled with the annual publication of financial statements, enabled Prowess to capture rich data on firms' product scopes. Prowess' product classification contains a total of 6,130 products. I use these data to determine whether a firm was exposed to the QR-removal policy based on the extent of its product offerings and operations in product markets affected by QR removals. Other studies that have employed these product scope data include Goldberg et al. (2010a); DeLoecker et al. (2016).

Second, and less widely recognized, is that Prowess publishes the names, designations, and tenures of all board members of each company for every year. Importantly, it also provides the Director Identification Number (DIN) for these directors, an 8-digit unique identifier issued by the Ministry of Corporate Affairs, Government of India. This feature facilitates matching with a larger administrative dataset from the Ministry of Corporate Affairs, as discussed below. The Prowess dataset includes data for a total of 383,779 directors across 51,125 manufacturing, financial, and non-financial services companies. Unlike the financial and product data, the board data are somewhat incomplete– names are often abbreviated to initials and last names, and designations may be missing. A key innovation of this project is the systematic cleaning and imputation of missing information for directors, performed on a firm-by-firm and director-by-director basis by cross-referencing the Prowess data with the administrative records from the MCA.

For cleaning the Prowess data, I follow a series of sample selection steps. First, I restrict the sample to manufacturing firms, thereby excluding service and financial companies. Next, I retain only those firms that began reporting data before the last pre-policy year (i.e., 2000) and that provide at least two observations during the pre-policy period. I also exclude a small number of firms that never report any information on the products

they produce. These steps yield a panel dataset comprising 4,994 firms observed over the period 1988-2010.

Prowess serves as the baseline dataset for this study. However, to investigate how the import competition shock impacts manufacturing firms in India– particularly regarding their internal organization and top management– I supplement Prowess with three novel datasets. This three-step data construction procedure is described below.

Novel Dataset on Product-level Quantitative Restrictions in India. The first step in identifying which firms are affected by the QR-removal policy is to determine which products are subject to the policy and when the restrictions were removed. There is no central database for this information, and the liberalization policy was implemented over several years (with a particularly heavy emphasis in the final three years– 1999-2001) through more than 26 government notifications issued in the Official Gazette of India.

I obtained these notifications from the archives of the Ministry of Commerce, Government of India, to create a detailed dataset of products under quantitative restrictions (QRs), including the exact removal dates. Figure A1 in Appendix A.1 shows an example of one such policy notification. These notifications typically provide a list of products along with their corresponding product codes, based on the 1996 Indian Trade Classification Harmonized System (ITCHS) nomenclature, the import licensing policy applicable to each product, and the date of notification. In most cases, the column for the new import policy indicates "*Free*", implying that there are no quantitative restrictions on the product after the notification date. In some instances, however, the policy may be less liberal– for example, quantitative restrictions might be partially lifted such that the product can be imported into India only by a State Trading Corporation (STC) or only imported through a Special Import License (SIL). I treat a product as liberalized if the imports of that product are completely free of any kind of quantitative or licensing restrictions.

Thus, after processing these data, I obtained a dataset at the 8-digit level that details the year in which QRs were removed for each product. Based on the above classification, out of a total of 10,839 ITCHS 8-digit products, 3,109 products appear in the QR-removal government notifications, and out of these 2,982 are made completely free of any kind of quantitative restriction between 1995 and 2001.

Novel Product Concordances. In the next step, I construct novel product concordances between the 1996 ITCHS product nomenclature used by the Indian customs authority with the product nomenclature in the firm-level CMIE Prowess data. These concordances were prepared by hand by going through the following resources: (1) detailed product descriptions of each of the 3,109 ITCHS 8-digit products contained in the QR notifications and matching them with the descriptions of 6,130 products reported by Prowess firms, (2) The HS-6 digit to Prowess product code concordance in Barrows and Ollivier (2021) for manufacturing products, and (3) an incomplete concordance between ITCHS products and Prowess products provided by CMIE. Significant manual work was required for this step despite existing concordances, because existing concordances use new vintages of HS products, while the QR notifications used older 1996 vintage of ITC-HS codes. Moreover, existing concordances are at the coarser 6-digit

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HS level. Table A1 in Appendix A.1 provides a few examples of how this matching is performed. As a result, I identified 1,168 unique Prowess products that correspond to at least one ITCHS product for which QRs were removed during the policy window. The next section discusses how these data are used to assign treatment at the firm level.

Table 1: Family Ties Among the Board Members and Top Managers of an Indian Firm

Name	Executive Member	Sex	Father's Name
A Jain	Yes	M	D K Jain
M Jain	Yes	M	D K Jain
P Jain	Yes	M	D K Jain
V K Sood	No	M	H R Sood
D K Jain	Yes	M	K L Jain
N K Jain	Yes	M	K L Jain
N Jain	Yes	M	N K Jain
M P Gupta	No	M	P D Gupta
S Sharma	No	M	R K Sharma
R Jain	No	F	S Raj
M L Mangla	No	M	T Chand

Notes: The table shows the directors of an example Indian firm, Cosco (India) Ltd, a sports goods producer. It lists all board members of Cosco from 2000 to 2010, illustrating the board's domination by the Jain family. The second column presents the name of each director (with first and middle names abbreviated for brevity), and the third column indicates whether the individual is on the executive board, i.e., top management positions like CEO, CFO, MD, etc. The final column includes the father's name of each person, allowing the identification of family ties among board members within the same firm. *Source:* CMIE Prowess and Ministry of Corporate Affairs, Government of India.

Administrative Data on Manager Family Ties. Finally, I obtained access to novel administrative data from the Ministry of Corporate Affairs (MCA), covering over 6 million board directors from approximately 1.3 million registered Indian companies. Each director is assigned a unique 8-digit Director Identification Number (DIN), and each firm has a unique 23-digit Company Identification Number (CIN). The dataset records the full legal names of individuals– with official spellings that are rigorously verified against government-issued IDs– along with key demographic details such as date of birth, gender, and place of birth.

Moreover, these data provide a rich historical record of board director tenures dating back to the 1970s. They capture the designations held by directors during their tenure at each firm. For example, while some directors serve solely as independent board members– attending meetings without involvement in daily operations– others hold executive roles (e.g., Managing Director, CEO, or CFO) that entail active participation in the firm's day-to-day management. The dataset also includes the start and end dates of each directorship, which is crucial for analyzing turnover among top managers.

Most crucially for this study, the data feature each director's father's name, thereby facilitating the identification of family ties within firm boards. Since the MCA registration process requires directors to provide their father's name, this information is

Table 2: Summary Statistics

	Obs	Mean	p10	p50	p90
Treated Firms (%)	83,726	47	0	0	100
Company Age	83,726	27	7	21	56
Wages	82,745	207	0	21	305
Gross Fixed Assets	82,067	2307	20	213	2618
Revenues	82,745	3399	1	363	4385
Expenses on Raw Materials	82,745	1386	0	136	1785
Atleast 1 Family Member on Board (%)	82,493	45	0	0	100
Family Share on Board (%)	39,644	38	10	33	75
Family Share on Executive Board (%)	39,644	64	0	100	100

Notes: This table presents summary statistics for firms included in our analysis, using data from CMIE Prowess and the Ministry of Corporate Affairs, Government of India. "Treated Firms (%)" represents the percentage of firms in the treatment group. "Company Age" is measured in years since incorporation. "Wages" denotes the total wage bill of the firm and is deflated using the GDP deflator. "Gross Fixed Assets" refers to the book value of fixed assets and is deflated using the gross fixed capital formation deflator. "Revenues" indicate total firm revenue and are deflated using NIC output deflators, while "Expenses on Raw Materials" reflect expenditures on material inputs and are deflated using NIC input deflators. "At least 1 Family Member on Board (%)" represents the percentage of firms identified as family-owned during pre-policy period. "Family Share on Board (%)" denotes the percentage of board members who belong to the controlling family. "Family Share on Executive Board (%)" represents the percentage of family members on the executive board of the firm. All monetary values are deflated using 2004 as the base year. Columns represent the total observations, mean, 10th percentile, median, and 90th percentile of each variable at the firm level. *Source:* CMIE Prowess and Ministry of Corporate Affairs, Government of India.

reliably recorded and enables the tracking of familial relationships across board members.

For instance, Table 1[4] illustrates the board composition of Cosco India Ltd., a sports goods producer, between 2000 and 2010. The table reveals that directors A Jain, M Jain, and P Jain share the same father's name (D K Jain), confirming that they are brothers. Additionally, N K Jain, identified as D K Jain's brother, and N Jain, noted as his nephew, are also listed. This example underscores how the inclusion of familial information in the administrative data enables a detailed analysis of family presence on firm boards and sheds light on the extent of familial networks among top managers in Indian companies.

3.1 Descriptive Statistics

This section provides a detailed description of the graphical results presented in Figures 2 and 3, focusing on the descriptive characteristics of family-managed firms in India. I define a firm as a family firm if it has at least two board members from the same family.

Figure 2 illustrates the distribution of family-managed firms across different industries and geographic regions. Panel (a) reveals some variation across industries in the share of family firms. While some sectors such as Food & Beverages, Textiles, and Rubber &

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[4]The initial and middle names have been abbreviated for brevity.

Plastic exhibit a higher prevalence of family management, all industries have at least 40 percent of firms with at least two board of directors who are from the same family. Panel (b) highlights the geographical distribution of family firms across Indian states. It shows that family firms are widespread in India: despite different local business culture and economic structures, nearly half of the firms are family firms in most states across India.

Figure 2: Share of Family Firms Across Industries and States

(a) Family Firms Across Industries

(b) Family Firms Across States
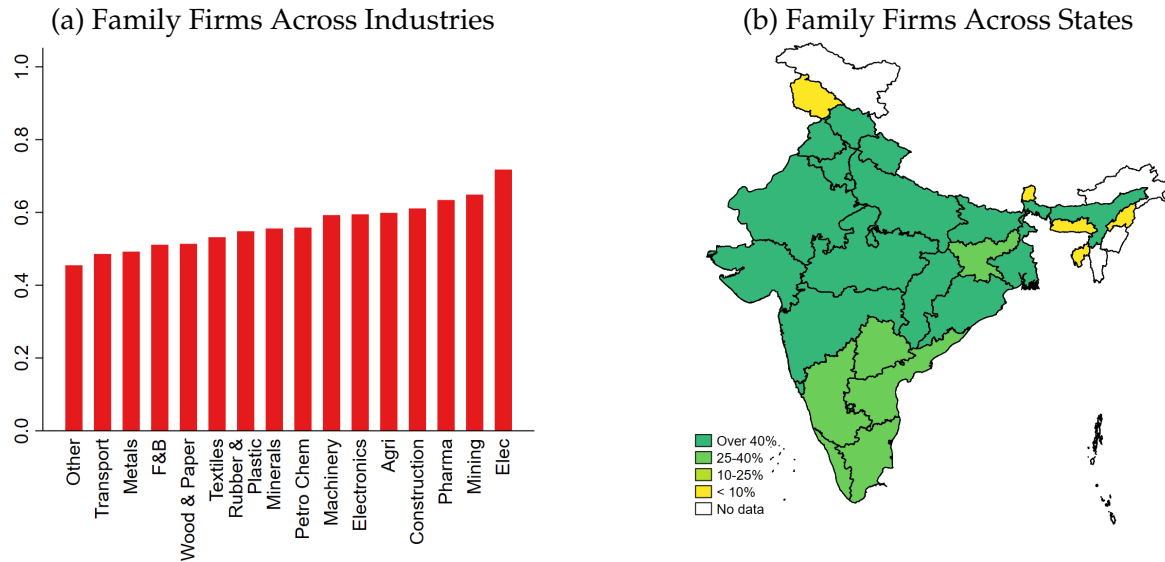
Notes: This figure displays the share of family firms across industries (panel a) and across states (panel b). *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

Figure 3 provides kernel density plots that compare firm-level outcomes based on the degree of family involvement in top management. Panels (a) and (b) depict the distribution of firm size measured by log revenues and log fixed assets, respectively. Firms with a higher share of family managers (top tercile of share of family among company directors, indicated in red) generally tend to be smaller, exhibiting lower revenues and fixed assets. Panel (c) indicates that these family-involved firms also have a lower wage bill, suggesting either lower employment levels or lower average wages. Panel (d) demonstrates that firms with extensive family management also tend to export less, as indicated by the higher density at the upper tail. Panels (e) and (f) address profitability, revealing that firms with significant family involvement tend to report lower profits, both in absolute terms and relative to revenues, indicating weaker financial performance.

Collectively, these descriptive statistics highlight distinctive differences in firm characteristics based on family involvement, underscoring the significant role of family management in influencing firm size, profitability, and market orientation in the Indian context.

4 Empirical Strategy

The removal of QRs from consumer goods in 1998-2001 offers a unique opportunity to analyze the impact of import competition on firm behavior. First, the reform was

Figure 3: Firms with High Share of Family Top Managers are Smaller and Less Profitable

(a) Log Revenue

(b) Log Fixed Assets

(c) Log Wage Bill

(d) Log Exports

(e) Arcsinh Profits
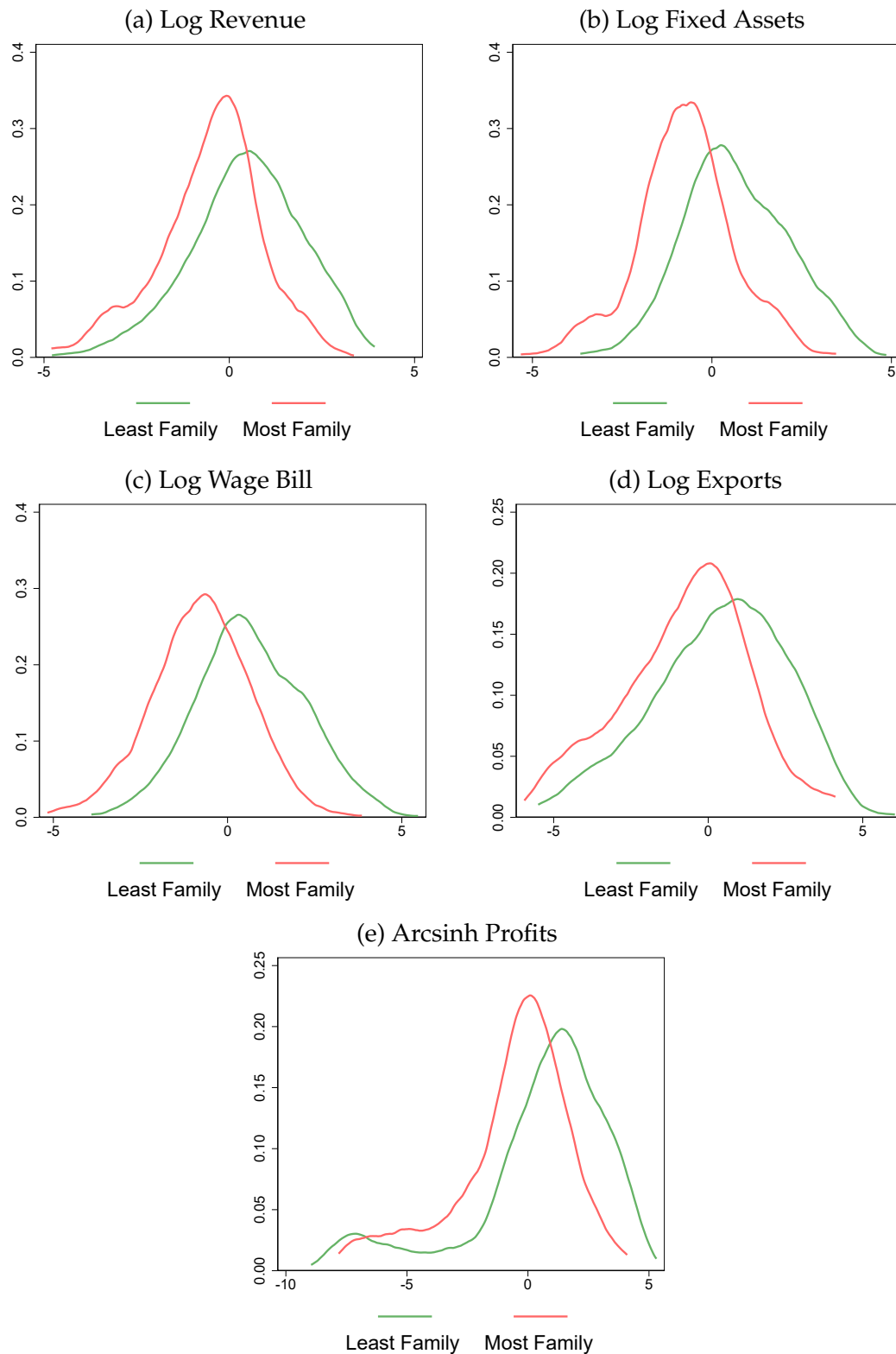
Notes: The figure plots kernel density plots of firm revenue (panel (a)), fixed assets ((panel (b)), wage bill (panel(c)), exports (panel (d)), and profits (panel (e)). The green (red) densities correspond to firms that have the bottom (top) tercile of share of family on the board in the pre-policy period. All variables are residualised by 5-digit industry dummies. *Source:* CMIE Prowess and Ministry of Corporate Affairs, Government of India. Prowess$_{dx}$.

product-specific, with India removing QRs on about 3000 products at the 8-digit HS level. The 8-digit HS classification provides a detailed breakdown of traded goods (about 10,000 goods in total). This quasi-experimental variation in exposure to import competition across the product space offers a natural setting for a difference-in-difference identification strategy.

Second, this policy is unusual in its primary focus on final consumer goods. Trade reforms typically affect product markets throughout the production network, influencing both intermediate inputs and final goods. Lowering import costs for intermediate goods directly reduces firm costs and may enhance firm productivity (Amiti and Konings, 2007; Goldberg et al., 2010a; Topalova and Khandelwal, 2011). However, this mechanism plays a limited role in the present case, as the QR removal primarily targeted final consumer goods. This narrower scope of the policy provides a unique opportunity to isolate and identify the impact of a specific *demand* shock– heightened import competition in consumer goods markets– on firm outcomes.

Third, the QR removal was a unilateral trade liberalization policy, granting foreign firms access to the Indian market without reducing export costs for Indian firms. Thus, the policy mainly reflects the effects of import competition, not export incentives.

Fourth, similar to the 1991 trade reform, and as discussed earlier, India's removal of QRs was externally imposed. The timing and scope of this liberalization were determined by the WTO, and critically, the IMF played a decisive role in the process. The IMF's decision to no longer allow India to rely on Article XVIII:B of the GATT was based on a technical assessment of India's external financial position. This assessment concluded that India's foreign exchange reserves were adequate, and the decision was made independently of political or other policy considerations. This externally driven process underscores that the QR removal was not influenced by domestic policy preferences or strategic interests, providing a uniquely exogenous shock for empirical analysis.

Finally, unlike the 1991 trade reforms, which coincided with widespread domestic liberalization, the removal of QRs in the late 1990s and early 2000s occurred in relative isolation, unaccompanied by other major domestic or trade policy changes. This limited scope reduces the likelihood of confounding effects, making it easier to attribute observed changes in firm behavior to the QR removal policy.

To analyze the effects of QR removal, I employ an event study framework at two levels:

1. **Aggregate product level:** To examine how the policy influenced aggregate imports.

2. **Firm level:** To assess how firms adjusted in response to increased import competition.

The aggregate product-level analysis allows me to quantify the policy's direct effect on trade flows. The firm-level analysis investigates its implications for the financial and managerial outcomes of exposed firms. Below, I outline the empirical specifications for each level of analysis.

Product-Level Event Studies To estimate the effect of QR removal on aggregate imports, I use the following event study specification at the product level:

$$y_{pt} = \sum_{k=\underline{T}}^{\overline{T}} \beta_k D_{pt}^k + \delta_p + \lambda_{qt} + \varepsilon_{pt} \tag{1}$$

where y_{pt} is the log import or export value or quantity of an 6-digit HS product p in year t, δ_p are HS-6 digit product fixed effects, and λ_{qt} are 4- digit HS product $\times$ year fixed effects.[5] The inclusion of λ_{qt} means that the β_k coefficients are identified using liberalized and unaffected HS-6 products within HS 4-digit product $\times$ time. Event time dummies D_{pt}^k are defined as $D_{pt}^k := 1\left[t = \tau_p + k\right] \forall k \in (\underline{T}, \overline{T})$ where τ_p is the year in which QRs were removed for product p. The coefficient for the event year $(k = 0)$ is normalized to zero. I set $\underline{T} = -5$ and $\overline{T} = +8$. Standard errors are clustered at the HS-6 digit product level.

The key identification assumption is that, in the absence of the QR removal, products affected by the policy would have followed similar import trends as unaffected products, after accounting for time-invariant differences between 6-digit products and common 4-digit product $\times$ year shocks.

Firm-Level Event Studies. To estimate the impact of the QR-removal policy on firm outcomes, I use an event study approach. The policy was implemented in a staggered manner from 1997 to 2001. All results presented in the next section rely on the estimator of Sun and Abraham (2021). For robustness, I also estimate the event study using a two-way fixed effects specification, which yields similar results (see Appendix A.2.2). The event study specification is as follows:

$$y_{it} = \sum_{k=\underline{T}}^{\overline{T}} \theta_k D_{it}^k + \delta_i + \lambda_{jt} + \varepsilon_{it}, \tag{2}$$

where y_{it} is an outcome of firm i in accounting year t, δ_i is a firm fixed effect, and λ_{jt} are three-digit industry $\times$ year fixed effects. Therefore, θ_k coefficients are estimated comparing treated and untreated firms within sector $\times$ time.[6] In robustness analysis, I show that the results are similar after controlling for location (state or district) $\times$ year fixed effects and firm size $\times$ year fixed effects. Event time dummies D_{it}^k are defined as follows. $D_{it}^k := 1[t = \tau_i + k] \forall k \in (\underline{T}, \overline{T}), D_{it}^{\overline{T}} = 1\left[t \geqslant \tau_i + \overline{T}\right]$, and $D_{it}^{\underline{T}} = 1\left[t \leqslant \tau_i + \underline{T}\right]$, where $1[.]$ is the indicator function and τ_i is the year in which QRs are removed on the highest-revenue product of firm i. ε_{it} is an error term. I normalize $\theta_0 = 0$ and set $\underline{T} = -5$ and $\overline{T} = +8$. Standard errors are clustered at the three-digit industry $\times$ year level.

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[5]I currently have access to annual trade flow data for India at the 6-digit HS level. a 6-digit HS product is classified as treated if any of its constituent 8-digit products are affected by QR-removal. I am in the process of procuring monthly trade flow data at the 8-digit HS level and plan to update these event studies once the new data becomes available.

[6]I am able to include three-digit sector $\times$ year fixed effects because the QR-removal policy was implemented at the more granular 8-digit HS code level. This ensures that within each three-digit sector, exposure to the policy varies across products, allowing for the inclusion of sector-level time controls without absorbing the treatment effect.

The key identification assumption for estimating θ_k is that, in the absence of the QR removal, firms operating in product markets exposed to import competition would have followed similar trends in outcome variables as firms in unaffected sectors. This implies that the latter serve as a reasonable counterfactual for the treated firms after accounting for time-invariant differences between firms and common 3-digit industry × year shocks.

5   Results

In this section, I first examine the impact of the QR-removal policy on product-level trade flows into India. After establishing that the policy significantly increased imports of liberalized products, I then analyze its effects on the financial and managerial behavior of manufacturing firms in India.

## 5.1   Impact of QR Removal on Aggregate Imports

Figure 4 presents event-study estimates from Equation (1), capturing how products exposed to the removal of QRs differed from unaffected products in terms of import and export outcomes. Panels (a) and (b) illustrate a substantial and persistent increase in both the value and the quantity of imports for the treated products. Notably, by the third year following QR removal, the import value of these products exceeds that of the control group by over 50 percent (panel (a)). The import value of treated products remains high through the eighth year after the reform, highlighting the persistence of the policy's effect. The impact on import quantities (panel (b)) follows a similar trajectory but is even more pronounced in magnitude, with the treated products reaching an increase of roughly 1 log point (over 150 percent) compared to the control group.

The stability of pre-trend coefficients suggests that treated and untreated products followed comparable trajectories before the reform. Moreover, the inclusion of HS-4 digit product × year fixed effects (where HS-4 is a broader product classification than HS-6) ensures that any time-varying shocks at the HS-4 digit level do not drive the results. Thus, the post-reform divergence in imports can be credibly attributed to the removal of quantitative restrictions.

The bottom panels confirm that the policy had little to no discernible effect on exports of the same set of products. This is precisely what one would expect from a reform that exclusively liberalized India's domestic market for foreign producers without granting any reciprocal benefits to Indian exporters. The absence of a parallel export response underscores the unilateral nature of the policy: it primarily heightened competition from foreign producers for Indian firms, without materially altering Indian firms' access to foreign markets. The divergence between imports (which rise sharply and persist) and exports (which remain unchanged) helps rule out alternative explanations related to changes in export opportunities.

These findings confirm that the removal of QRs substantially intensified import competition in India. Having established the surge in imports at the product level, I now turn to examining how firms responded to this heightened competition. Specifically, I employ an event-study framework comparing firms that produce goods affected by QR removal against firms whose product portfolio remained unaffected, enabling me

Figure 4: Value and Quantity of Imports Increase after QR Removal with No Impact on Exports

(a) Log Import Value

(b) Log Import Quantity

(c) Log Export Value

(d) Log Export Quantity
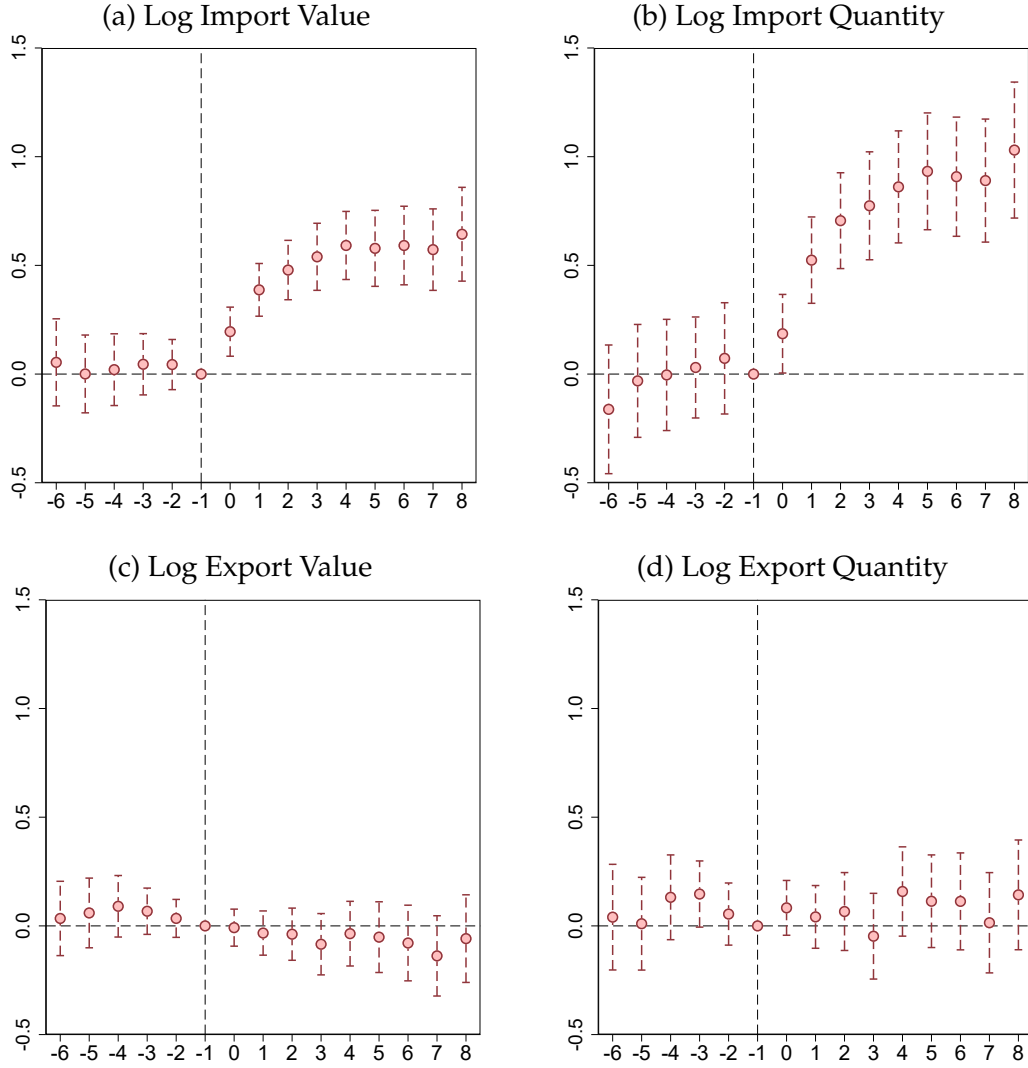
Notes: The figure presents β_k event study coefficients from Equation (1) using the Sun and Abraham (2021) estimator on annual HS-6 digit product-level panel data on imports and exports. The coefficients plotted correspond to Table A2, columns (1)-(4). The dependent variables are log import value (panel (a)), log import quantity (panel (b)), log export value (panel (c)), and log export quantity (panel (d)). An HS-6 digit product is identified as treated if QRs were removed from any of its constituent HS-8 digit products. β_{-1}, the coefficient prior to the year in which QRs were removed, is normalized to zero. The policy is staggered from 1995 to 2001, with the *x*-axis denoting years relative to the event. All regressions include HS-6 digit product fixed effects and HS-4 digit product × year fixed effects. Standard errors are clustered by HS-6 digit product. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Trade$_{dx}$.

to isolate the causal impact of import competition on firms' financial and managerial outcomes.

5.2   Impact on Firm Size and Financial Performance

Figure 5 presents event-study estimates from Equation (2), offering a 360-degree view of how intensified import competition influences key dimensions of firm size and financial performance. The panels cover firm revenues, costs, and capital structure, allowing us to trace the broad impact of foreign competition on domestic firms. Table A3 in a

This first panel shows a substantial decline in total revenue for firms exposed to QR removal, relative to unaffected firms. By the third year following the policy change, treated firms' revenues fell by approximately 20 percent compared with the control-group firms. The gap widens further in subsequent years: by the eighth year, revenues of treated firms are almost 50 percent lower than those of the control group. This pronounced and persistent decrease underscores the depth of the import-competition shock. In panel (b), I examine the operating profit-to-revenue ratio, which captures how effectively firms convert sales into operating profits. This ratio declines by about 0.04 for treated firms– equivalent to nearly halving the pre-policy average among control group firms of 0.09. Such a drop highlights that competitive pressures not only reduce overall revenue but also compress margins.

Turning to labor-related expenditures, panel (c) shows that the total wage bill experiences a decline comparable in magnitude to the drop in total sales. The effects manifest soon after the policy takes effect and persist through the eighth year. The protracted nature of this decline suggests that firms engage in sustained cost-cutting on labor, likely in response to shrinking market share and profitability. Expenditure on raw materials (panel (d)) also follows a downward trajectory, decreasing by as much as 30 percent by the eighth year. This reduction is consistent with firms scaling back production and operations in the face of heightened import competition, using fewer inputs in line with reduced output and sales.

The bottom two panels show changes in firms' capital structure. Panel (e) illustrates a substantial and growing decline in the total assets of treated firms. By the third year, assets are around 10 percent lower relative to the control group, and this disparity accelerates over time. By year eight, the total assets of treated firms have fallen by almost 30 percent compared with their unaffected counterparts. This pattern points to a long-term contraction in capacity, possibly due to underutilized assets, disinvestment, or an inability to generate sufficient cash flow to sustain capital stocks. Finally, panel (f) examines total firm borrowings. Unlike other indicators, borrowing remains relatively stable in the early years post-reform but eventually declines to around 25 percent below the level of control-group firms.

Figure 5 also helps confirm the lack of pre-trends in the outcome variables. Pre-policy event-study coefficients are insignificant and close to zero.

Overall, these results demonstrate that the QR-removal policy delivered a substantial negative shock to Indian firms' operations and balance sheets. As foreign products entered the domestic market at scale, many Indian firms struggled to protect their revenue base and profit margins, leading to cutbacks in labor, materials, and capital.

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These results are not driven by firm exits, although some firm exits do happen around the policy. Figure A5 in Appendix A.2.2 reproduces the event study estimates in Figure 5 conditioning on surviving firms. Out of 4994 total firms in the sample, 4147 firms survive till the end of the sample. The results are qualitatively similar to unconditional results presented in Figure 5.

In the next section, I explore how firms adapt organizationally in the face of this heightened competition, focusing on the turnover of top managerial positions– particularly among family-run firms that opt to bring in professional outside managers.[7]

5.3 Impact of QR Removal on Firm Management

As established in the previous section, the removal of QRs imposed a significant negative shock on the financial health of exposed firms. A natural question follows: what organizational actions did these firms undertake to mitigate the shock? In this paper, I focus specifically on top-management changes, motivated by both theoretical considerations and the empirical regularities documented in the corporate governance and organizational economics literature.

A large body of research demonstrates that negative shocks to profitability and poor firm performance often precipitate forced or voluntary departures of top executives (Jenter and Kanaan, 2015; Kaplan and Minton, 2011; Parrino, 1997). However, the mechanism of managerial restructuring in the context of family-owned or family-controlled firms differs in a critical way from standard CEO or executive turnover models. In many developing economies, including India, family members frequently occupy the most senior rolespositions such as CEO, CFO, or Managing Director– irrespective of whether they are the best-qualified individuals to navigate competitive challenges. Such arrangements may be beneficial when family managers possess significant firm-specific knowledge or when they help maintain continuity and trust. Yet, when adverse market shocks arisein this case, heightened import competitionthese same family-oriented hiring practices can become a liability.

Faced with a substantial erosion of profitability and revenue, domestically owned family firms may find themselves compelled to reassess the merit of keeping family members in top-level posts. Replacing family managers with external professionals can bring fresh expertise, more experience, access to wider networks, strategic thinking, and managerial skills that are often critical for adapting to intensified competitive pressures. In this study, I capture this phenomenon by looking beyond the traditional turnover measures (e.g., whether the CEO or CFO changes) and instead examining the extent of family involvement in senior management positions and the executive board of directors.

Figure 6 illustrates these organizational responses and shows how family firms respond to increased import competition by altering the composition of their executive boards. In panel (a), the dependent variable is the *share* of family members on the board, which has a pre-shock control-group mean of 0.60. By the third year after QR

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[7]Figure A4 in Appendix A.2.2 reproduces the event study estimates in Figure 5 for family firms (i.e. firms that have at least board member from the founder's family). Out of about 5,000 total firms in the sample, almost half of the firms meet this criteria. The results are qualitatively similar to results presented for the whole sample in Figure 5.

Figure 5: Domestic Firms Contract after QR Removal
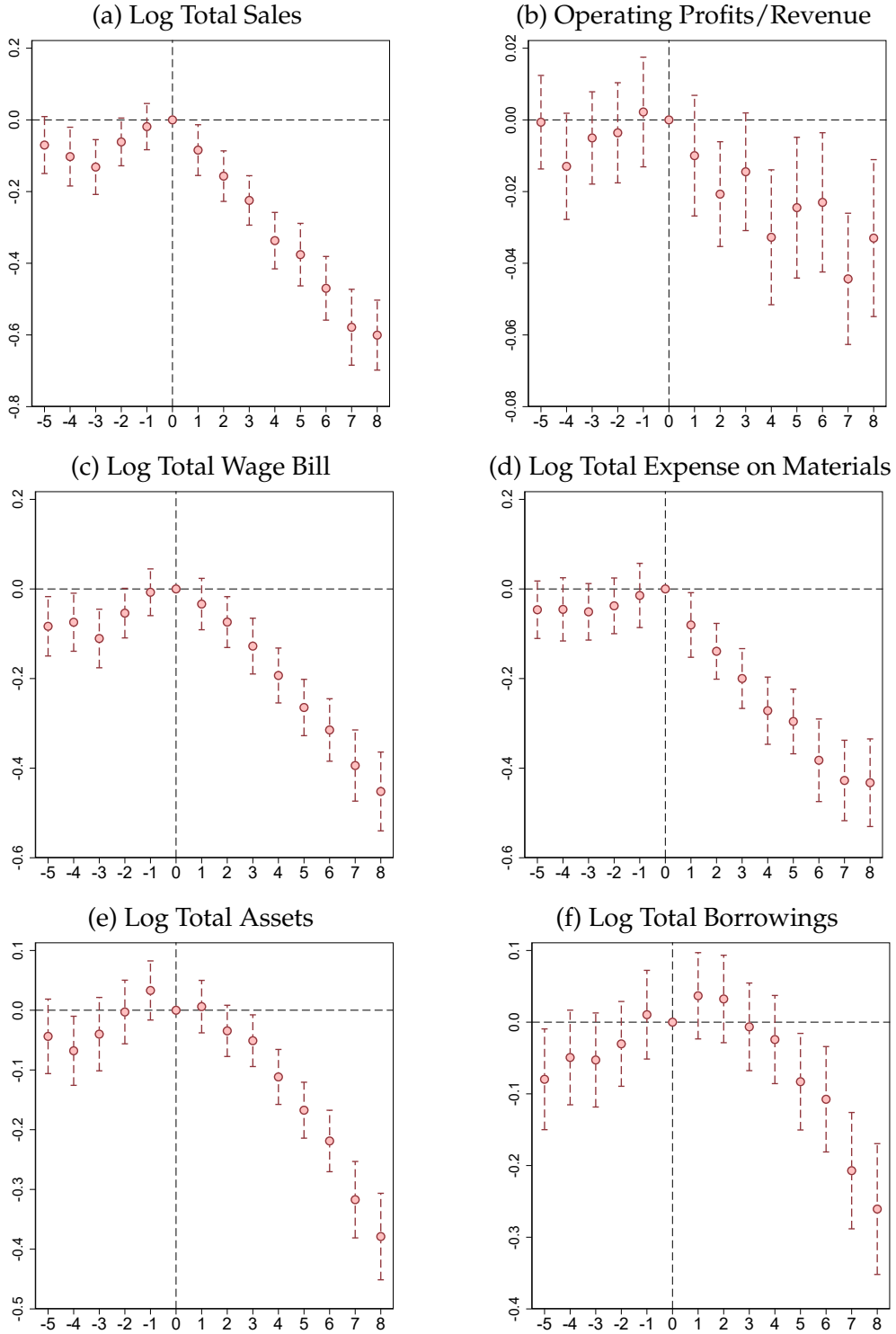
(a) Log Total Sales

(b) Operating Profits/Revenue

(c) Log Total Wage Bill

(d) Log Total Expense on Materials

(e) Log Total Assets

(f) Log Total Borrowings

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2), corresponding to coefficients in Table A3, columns (1)-(6). The dependent variables are: log total sales (panel (a)), operating profits/revenue (panel (b)), log total compensation (panel (c)), log total expense on raw materials (panel (d)), log total assets (panel (e)), and log total borrowings (panel (f)). A firm is identified as treated in a year if QRs were removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the *x*-axis denoting years relative to the event. All regressions include firm and three-digit industry × year fixed effects. Standard errors are clustered at three-digit industry × year level. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$.

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removal, this share declines by about 5 to 6 percentage points, widening to roughly 8 percentage points by the eighth year. In relative terms, these coefficients represent a substantial reduction of about 15 percent in the fraction of family executives at the top.

Panel (b) zooms in on which firms are *most likely* to shed family managers by comparing the bottom tercile of pre-policy productivity (red circles) to the rest of the sample (gray triangles).[8] The figure reveals that *almost all* of the reduction in the share of family members in top managerial roles is driven by firms that were relatively *less productive* before the QR removal. Indeed, these bottom-tercile firms show a pronounced and persistent decline in family share, while higher-productivity firms display little to no change. By the fourth year after QR removal, the share of top family managers in ex-ante declined by about 15 percentage points– almost three times as high as the overall impact for all firms in panel (a). Higher productivity firms showed no such change in their management structure. This pattern implies that the decision to replace family managers with external professionals is more prevalent– and more extensive– among firms that were initially weaker performers. Such selection into professionalization highlights a mechanism through which less-competitive firms may attempt to bolster their managerial capabilities when faced with heightened import competition. As elaborated in Section 6, this finding forms a core basis of my theoretical framework: those firms most in need of improving their productivity are the ones most likely to seek outside talent.
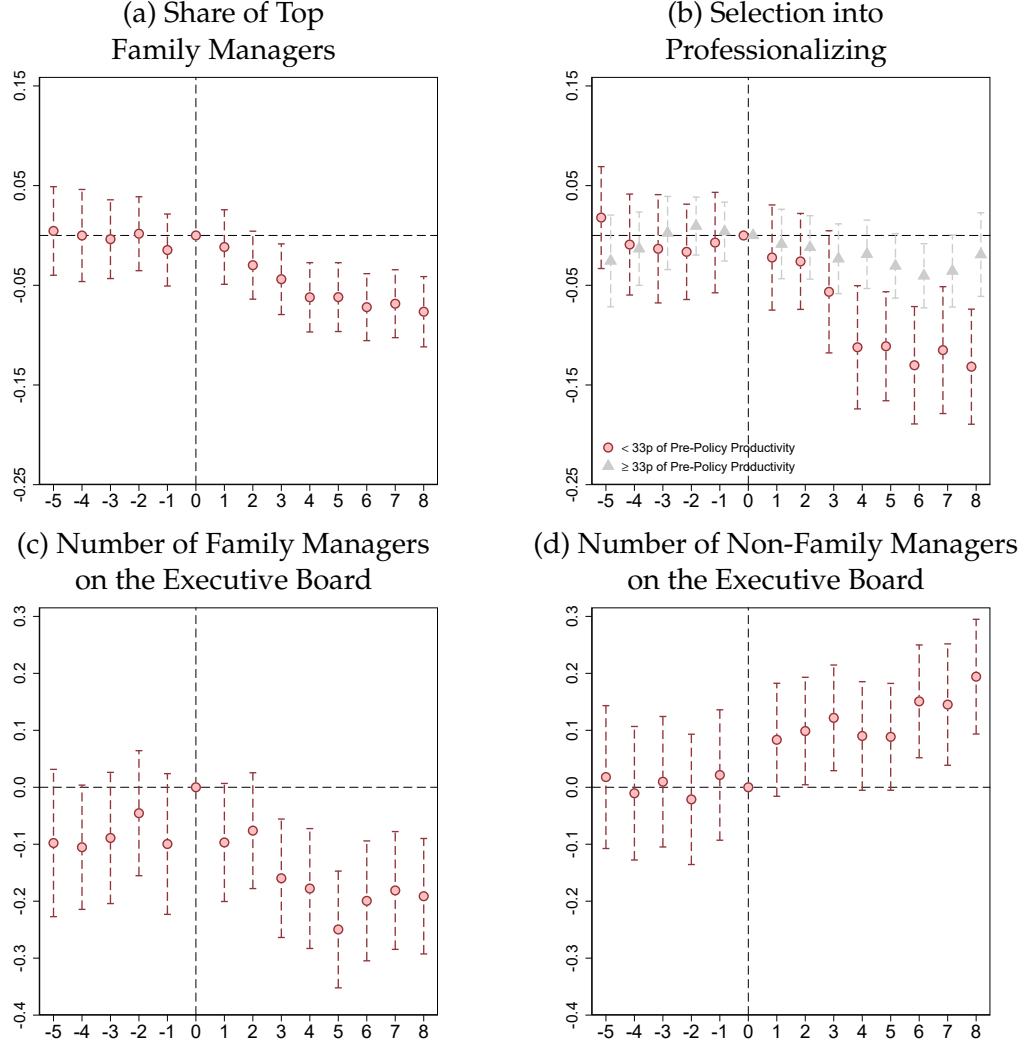
Panels (c) and (d) split the churn in top management into family and non-family managers. Before the shock, the average family firm in the control group had 1.3 family members on its executive board, compared to 0.5 non-family managers. After QR removal, the number of family managers steadily drops, culminating in a decrease of roughly 0.2 by the end of the sample window. Notably, panel (d) shows a nearly mirror-image increase in non-family professionals over the same horizon, pointing to a one-to-one replacement effect. In other words, for every family manager who exits, almost exactly one external professional is joining the board.

Taken together, these event-study results suggest that, under heightened competitive pressure, family-controlled firms do not simply shed family executives. Rather, they actively seek outside managerial talent to fill vacated positions, reconfiguring the firm's top hierarchy. To the best of my knowledge, this is the first study to show that globalization can trigger deeper changes in organizational structure within firms. An important point of comparison is Chen and Steinwender (2021), which studies how managers, particularly in family firms exert more effort in response to import competition. My focus is different in that I link a negative trade shock to the *composition* of senior management within the firm.

Such trade-induced change in corporate culture can be important, particularly in the context of developing countries where family-run firms and business groups are pervasive. As highlighted by recent work (Bloom and Van Reenen, 2007; Caliendo and Rossi-Hansberg, 2012; Akcigit et al., 2021), tight family control can constrain a firm's ability to adjust organizational layers or recruit external talent, potentially limiting the firm's capacity to respond effectively to competitive pressures. Bloom et al. (2013a) identify weak competitive pressure (for instance, due to protection from imports) and the predominance of family members in top management as major impediments to

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[8]Firm productivity is estimated using the method proposed in Petrin and Levinsohn (2012).

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Figure 6: Firms Reduce Family Members on the Executive Board of Directors after the QR Shock

(a) Share of Top
Family Managers

(b) Selection into
Professionalizing

○ < 33p of Pre-Policy Productivity
▲ ≥ 33p of Pre-Policy Productivity

(c) Number of Family Managers
on the Executive Board

(d) Number of Non-Family Managers
on the Executive Board

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2), corresponding to coefficients in Table A4 where (column (1)) corresponds to panel (a), (column (2)) corresponds to panel (c) and (column (3)) corresponds to panel (d). The dependent variables are: the share of family members on the executive board of directors of a firm (i.e., top management positions like CEO, CFO, MD, etc.) in panel (a), the share of top family managers dropped from the board while transitioning toward professional management in panel (b), the number of family members on the executive board of directors in panel (c) and the number of non-family professionals on the executive board of directors in panel (d). In panel (b), we examine selection into professionalizing by comparing firms based on their pre-policy productivity. Firms are divided into two groups: those in the bottom tertile of pre-policy productivity and the remaining firms. A firm is identified as treated in a year if QRs are removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the *x*-axis labels denoting years relative to the event. All regressions include firm and three-digit industry × year fixed effects. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

adopting effective management practicespractices that, in turn, can substantially boost firm performance. By showing that intensified import competition motivates family-owned firms to replace family managers with outside professionals, this paper offers fresh insights into how greater trade openness can reshape a firm's internal governance structure.

These results also speak to broader debates on whether business groups and family ownership in emerging markets facilitate or hinder growth. While such organizational forms may help mitigate imperfect capital markets or reputational frictions (Khanna and Yafeh, 2007), they can also exhibit weaker corporate governance, such as tunneling or underinvestment (Bertrand et al., 2002; Bertrand and Schoar, 2006). The evidence presented here suggests that, when faced with an exogenous shock such as the removal of QRs, family-controlled firms do not necessarily remain locked into potentially suboptimal leadership arrangements. Instead, they appear capable of adopting professional management structures to enhance competitiveness. Thus, the dismantling of protective barriers in India reveals how trade liberalization can catalyze deeper organizational changes, prompting even family-based firms to reconfigure their top management in pursuit of higher productivity and improved performance.

A natural question arising from the preceding analysis is: what are the implications of this push toward professionalization for firm productivity? In the following section, I explore this issue in greater detail.

5.4 The Impact of Professionalizing Management: Suggestive Evidence

Figure 7 offers indicative evidence that family-controlled firms that *professionalized* their top management– by reducing the share of family members in senior executive roles– enjoyed a greater post-reform boost in productivity relative to those that did not. I classify a firm as having professionalized if its share of family top managers declined between the pre-policy period and the end of the sample window (i.e., by $t = 8$ for the last-treated cohort). Figure 7 presents the event-study estimates from equation 2 separately for each group, focusing on productivity and prices.

Panels (a) and (b) in Figure 7 depict the evolution of two productivity measures– TFPR and TFPQ (both estimated following Petrin and Levinsohn 2012). For *professionalizing* firms (in red), both TFPR and TFPQ begin to diverge positively from zero in the first or second year following the policy and continue rising thereafter. By contrast, firms that retain family managers (in gray) show little change in either TFPR or TFPQ. Within about five to six years post-reform, TFPR for professionalizing firms lies roughly 20–30 percentage points above that of firms retaining family-dominated management, suggesting that the decision to bring in non-family managers may have facilitated substantial efficiency gains. The gap for TFPQ is over twice as large. These results are not driven by the fact that the least productive firms had a higher family share family managers before the policy. The mean share of family top managers for ex-ante low productivity firms is 0.62 (SD = 0.42) and 0.65 (SD = 0.41) for ex-ante high productivity firms.

Panel (c) shows the trajectory of average log prices, defined as the total value of a firm's output divided by the total quantity produced. Among professionalizing firms, prices

Figure 7: Increase in Productivity as Firms Shed Family Members after QR Removal
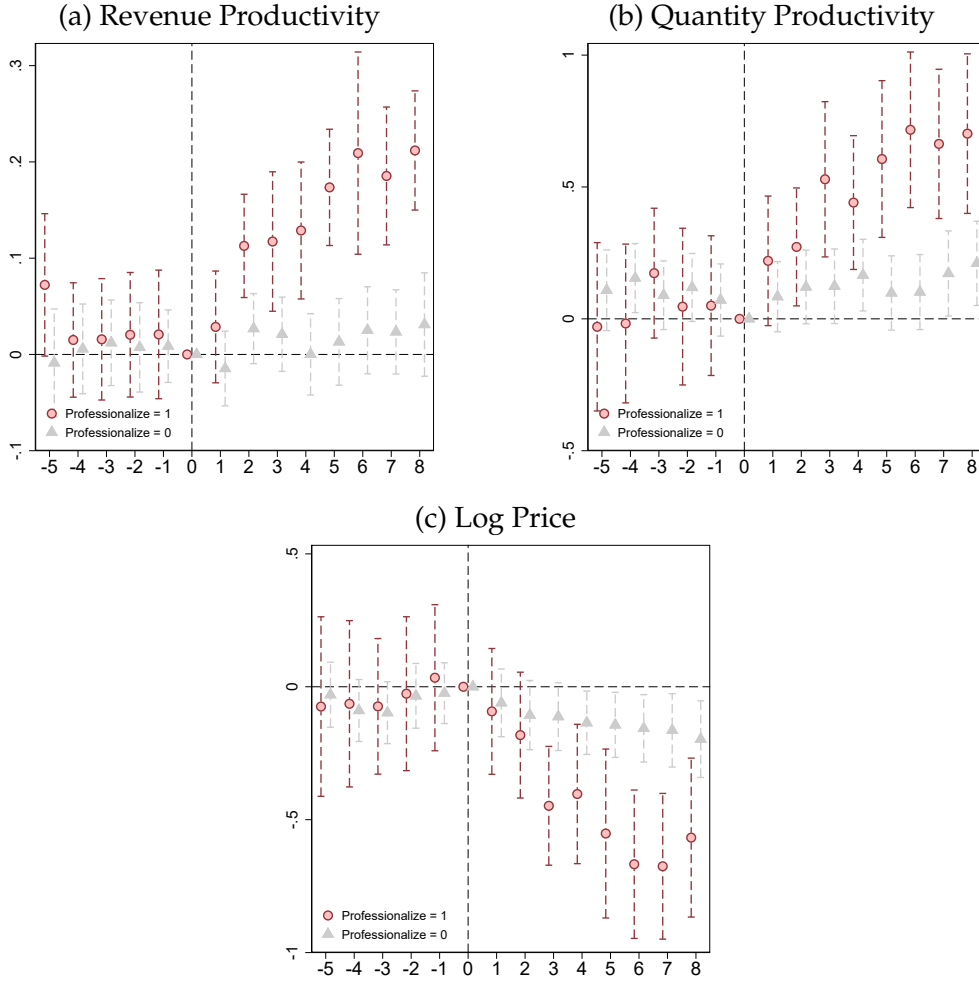
(a) Revenue Productivity

(b) Quantity Productivity

(c) Log Price

Notes: This figure presents the estimated θ_k event study coefficients from a regression specified in equation (2). Event studies are conducted separately for firms that professionalized their management after QR removal (in red) and those that did not (in grey). A firm is classified as treated in a year if QRs are removed on its highest-revenue product. A firm is considered to have professionalized if the share of family members on the executive board of directors declined in the post-policy period. The dependent variables are revenue productivity (TFPR) (panel (a)), quantity productivity (TFPQ) (panel (b)), and log price (panel (c)). TFPR and TFPQ are estimated using the method proposed in Petrin and Levinsohn (2012). Log price is defined as the ratio of a firm's total value of products produced and the total quantity of products produced. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the x-axis indicating years relative to the event. All regressions include firm and three-digit industry $\times$ year fixed effects. Standard errors are clustered at three-digit industry $\times$ year level. The vertical lines represent 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

drop notably after the reform, stabilizing at about 0.3 to 0.4 log points below pre-policy levels. By contrast, the non-professionalizing group exhibits minimal price adjustment. One interpretation is that firms with new (outside) management either implemented efficiency and cost-cutting measures that enabled price reductions or shifted toward lower-price varieties. This differential change in prices is a key reason for the much larger impact on TFPQ as compared to TFPR in the preceding two panels.

Overall, these patterns are consistent with the notion that heightened import competition catalyzes a deeper reorganization in firms that actively replace family managers with professional outsiders. Although the evidence in Figure 7 is inherently suggestive– firms self-select into professionalization, and not all organizational changes are captured. Some firms may have other unobserved advantages (e.g., more liquid credit lines, and stronger networks) that facilitate the hiring of external managers. These hidden characteristics could shape both the likelihood of professionalizing and subsequent performance improvements. Firms may also adapt in ways other than changing their top management, e.g., changes in mid-level managerial layers, shifts in organizational culture, etc. The evidence here is therefore *suggestive* rather than definitive.

Nevertheless, the event study results highlight two important themes. First, top-management turnover can be a critical margin of adjustment in response to negative trade shocks. Second, in family-run firms, bringing in external managerial talent appears to correlate with enhanced productivity performance.

6 Theoretical Framework

I consider a closed economy with one sector with monopolistic competition. Firms are heterogeneous in productivity and produce a unique variety indexed by i. The final good output, Y, is a CES aggregate of all intermediate varieties:

$$Y = \left(\sum_{i=1}^{N} y_i^{\frac{\sigma-1}{\sigma}} \right)^{\frac{\sigma}{\sigma-1}}.$$

where $\sigma > 1$ is the elasticity of substitution and N is the total number of varieties.

The aggregate price index is defined as $P \equiv \left(\sum_{i=1}^{N} p_i^{\frac{\sigma-1}{\sigma}} \right)^{\frac{\sigma}{\sigma-1}}$, where p_i is the price charged by the firm producing variety i. Cost minimization implies that the demand for variety i is given by

$$y_i = Y P^{\sigma} p_i^{-\sigma}. \tag{3}$$

Firm Entry, Management, and Firm Productivity. In this economy all firms start as family firms.[9] Each firm has the choice to professionalize firm management by hiring

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[9]While incorporating a firm, entrepreneurs, particularly in developing countries, often rely on social networks such as family, religion, caste, and geography, to manage their firm. There are many reasons for this, for e.g., trust among family members may substitute for weak legal institutions and contract

unrelated executives. The trade-offs associated with this choice are discussed below.

Firms pay a fixed cost f_e, paid in labor units, to enter the market and produce. After paying the fixed entry cost, the firm draws a productivity parameter $z \geq 1$ from a Pareto distribution

$$G(z) = 1 - z^{-k}$$

with $k > 1$ and $k > \sigma - 1$. Thus, z is the firm's productivity, if it decides to operate as a family firm. Owners of family-managed firms enjoy a non-monetary private benefit, $\mathcal{B}$, which is common for all firms. Private benefits measure the non-pecuniary utility that a firm's owner enjoys from running a firm as a family firm and holding the firm's *management* within the family. For example, a firm owner may derive pleasure if their children or siblings run the firm. Such amenity potential of family control of firm management has a long tradition in the corporate finance literature (Demsetz and Lehn, 1985; Burkart et al., 2003; Bertrand and Schoar, 2006). Firm owners can only enjoy private benefits if the firm is active. If the firm exits, its owner loses all private benefits associated with running the family enterprise.

Alternatively, firms may choose to professionalize their management by recruiting highly skilled external executives, thereby accessing a broader talent pool that surpasses the limitations of relying solely on family members. If the firm decides to do so, it loses its private benefit, $\mathcal{B}$, but at the same time, professionalization raises the productivity of the firm to γz, $\gamma > 1$, therefore earning the firm higher *monetary profits*. I assume that, other than loosing private benefits, there is no other cost of professionalization .[10]

Professionalization is costly to reverse. Once management is professionalized to unrelated managers, the firm pays cost κ to revert back to family management. κ captures several real-world frictions that make switching from professional management back to family difficult. Professional directors are typically bound by contractual commitments and fixed terms, which, along with the enhanced credibility and robust governance structures they provide, significantly boost the firm's market reputation and stakeholder confidence. Moreover, dismantling these established systems would not only disrupt the firm's operations but also risk reputational damage and a loss of investor trust. Together, these factors ensure that once professionalization occurs, the path back to family management is fraught with substantial costs, which our model captures with a high κ. Thus, the payoff from reverting becomes

$$\pi(z) + \mathcal{B} - \kappa,$$

instead of the usual family management payoff $\pi(z) + \mathcal{B}$.

Therefore, the decision to professionalize depends on the costs and benefits of delegating. The advantage of delegating is higher expected profits from the new productivity draw. The cost is foregoing private benefits.

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enforcement in developing countries (Bertrand and Schoar, 2006; Burkart et al., 2003).

[10]It's plausible that, in reality, firms incur fixed hiring or search costs when delegating management. However, introducing such costs into the model would not alter any qualitative predictions, as these costs are analytically similar to an increase in private benefits. Thus, for simplicity, I assume that the only cost of professionalization is the loss of private benefits.

Technology. The production function for each intermediate variety i is

$$y = z\ell.$$

where ℓ is labor employed. To produce, the firm pays a fixed cost of wf, which is paid in labor units. w is the wage rate and is used as the numeraire ($w = 1$). The firm maximizes profits subject to demand for its product (3), leading to the usual expression of equilibrium prices being a constant markup over marginal cost:

$$p_i = \frac{\sigma}{\sigma - 1}\left(\frac{w}{z}\right) = \frac{w}{\rho z}. \tag{4}$$

where $\rho = \frac{\sigma - 1}{\sigma}$. This implies that firm profits are given by:

$$\begin{aligned}
\pi(z) &= p(z)y(z) - w\ell(z) - wf \\
&= Az^{\sigma-1} - wf, \tag{5}
\end{aligned}$$

where $A = \frac{1}{\sigma}\rho^{\sigma-1}EP^{\sigma-1}w^{1-\sigma}$ and $E = YP$ is the aggregate expenditure in the economy.

Labor demand can be expressed as a function of firm profits

$$\begin{aligned}
\ell(z) &= \frac{y(z)}{z} \\
&= \frac{(\sigma - 1)}{w}(\pi(z) + f) \tag{6}
\end{aligned}$$

In this setup, the total payoff to the firm owner from an active firm is the sum of monetary profits, π, and non-monetary private benefits, $\mathcal{B}$, that the firm owner enjoys only if management is held within the family.

Define an indicator variable $\mathcal{P}$, where $\mathcal{P} = 1$ if the firm professionalizes management and $\mathcal{P} = 0$ if it retains family management. Then firms' optimal payoff which is given by:

$$\text{Firm's payoff} = \begin{cases} \pi(z) + \mathcal{B} & \text{if } \mathcal{P} = 0 \\ \pi(\gamma z) & \text{if } \mathcal{P} = 1 \\ 0 & \text{if firm exits} \end{cases}$$

Liquidity Constraints and Hand-to-Mouth Owners. To endogenize the exit decision, I assume that firm owners are *hand-to-mouth*, that is, they have no liquid wealth (formally, I set owner wealth, $\omega = 0$ for all firms). In each period, a firm must cover its operating costs (including the fixed cost wf) solely from its monetary profits. Because the non-pecuniary benefit $\mathcal{B}$ provides no liquidity, a negative monetary profit (i.e., $\pi(\cdot) < 0$) leaves the firm unable to cover its operating expenses, forcing it to exit. This mechanism endogenizes the no-negative-profits condition: a firm cannot continue operating with $\pi(\cdot) < 0$ because it cannot finance its fixed costs.

$$z_f = \left(\frac{wf}{A}\right)^{\frac{1}{\sigma-1}} \tag{7}$$

Thus, for a firm to be able to operate as a family firm, its productivity must be at least z_f. For firms that have professionalized management, the exit threshold, denoted z_e, is defined by by $\pi(\gamma z_e) = 0$:

$$z_e = \frac{1}{\gamma}\left(\frac{wf}{A}\right)^{\frac{1}{\sigma-1}} = \frac{z_f}{\gamma} \tag{8}$$

Professionalization to Survive among Laggard Firms. Note that the survival productivity cutoff for family firms is higher than that of firms that have professionalized management. This is because firms that professionalize management enjoy a productivity boost of $\gamma > 1$. This observation implies that any time a firm draws a productivity parameter less than z_e, it will immediately exit. For such a firm, even professionalization of management does not sufficiently boost its productivity to make enough profits to pay its fixed cost of operation. If a firm draws a productivity parameter $z \in (z_e, z_f)$, it will always choose to professionalize. This is because such a firm cannot make sufficient profits to pay its fixed operating costs as a family firm. However, it can survive if it decides to professionalize management, in which case its productivity will rise to γz, resulting in higher profits, potentially avoiding exit. Thus, the firm chooses to forgo its private benefits as professionalization is essential for preserving the firm. To summarize, for laggard firms with low productivity, there is negative selection into professionalizing. Conditional on survival ($z > z_e$), firms professionalize if they draw an initial productivity parameter $z < z_f$.

Professionalization to Boost Profits among Frontier Firms. If both $\pi(z)$ and $\pi(\gamma z)$ are positive, a firm chooses to professionalize its management if this choice yields a higher payoff than from staying a family firm. Since the firm gives up private benefits if it professionalizes, $\pi(\gamma z)$ must be sufficiently larger than $\pi(z)$ to compensate the firm for the loss of private benefits. Thus, the professionalization decision rests on the trade-off between retaining private benefits associated with running the firm as a family firm and higher profits associated with professionalization . This decision yields another productivity threshold, z_d, at which the firm is indifferent between remaining a family firm and upgrading its management by professionalizing:

$$\pi(z_d) + \mathcal{B} = \pi(\gamma z_d)$$
$$\implies z_d = \left(\frac{\mathcal{B}}{wf(\gamma^{\sigma-1}-1)}\right)^{\frac{1}{\sigma-1}} \cdot z_f \tag{9}$$

Thus, frontier firms professionalize management if their initial productivity $z > z_d$, i.e. we have more productive firms delegating– there is positive selection into professionalization . Firms with intermediate levels of productivity, i.e., productivity firms with

$z \in [z_f, z_d)$, choose to retain family management as the gains from delegating are not sufficient to compensate the firm owner for the loss of private benefits.

The following proposition summarizes the dual selection mechanism of the model.

Proposition 1 (Dual Selection Mechanism)**.** *Consider a firm drawing an initial productivity parameter z from the Pareto distribution*

$$G(z) = 1 - z^{-k}, \quad k > 1 \text{ and } k > \sigma - 1.$$

Define

$$z_f = \left(\frac{wf}{A} \right)^{\frac{1}{\sigma - 1}},$$

as the minimum productivity level required for profitable operation under family management. The corresponding survival threshold under professional management is given by

$$z_e = \frac{z_f}{\gamma}.$$

Moreover, let z_d denote the productivity cutoff at which a firm is indifferent between retaining family management (with the associated non-pecuniary benefit $\mathcal{B}$) and delegating to professional management, so that

$$\pi(z_d) + \mathcal{B} = \pi(z_d)$$

Then, the firm's optimal production and managerial decisions are characterized as follows:

1. ***Exit:*** *If $z < z_e$, the firm cannot generate positive monetary profits even after the managerial productivity boost from professionalization . Consequently, the firm exits the market.*

2. ***Professionalization to Avert Exit (Negative Selection):*** *If $z \in [z_e, z_f)$, the firm would incur losses under family management but can achieve positive profits by delegating management. In this range, professionalization is a necessary survival strategy.*

3. ***Retention of Family Management:*** *If $z \in [z_f, z_d)$, the firm is sufficiently productive to cover its fixed costs as a family firm. However, the incremental profit gain from professionalization does not offset the loss of the non-pecuniary benefit $\mathcal{B}$; hence, the firm opts to retain family management.*

4. ***Professionalization for Profit Enhancement (Positive Selection):*** *If $z \geq z_d$, the additional profits from delegating management more than compensate for the forfeited private benefits. As a result, highly productive firms choose to professionalize management.*

Figure 8 shows the professionalization thresholds in a diagram. In this diagram, the horizontal axis measures firm productivity, $z^{\sigma - 1}$, and the vertical axis reports both the firm's *monetary profits* and its *total payoff* (monetary profits plus non-pecuniary private benefits). The three vertical dashed lines divide the productivity space into four distinct regions that capture how firms decide whether to exit, remain family-managed, or professionalize management.

Starting from the left, the first vertical dashed line at z_e represents the *survival cutoff under professionalization* . Any firm drawing productivity $z < z_e$ will exit immediately because even with the productivity boost from professionalization (from z to γz), it

Figure 8: Professionalization Choice and Firm Productivity
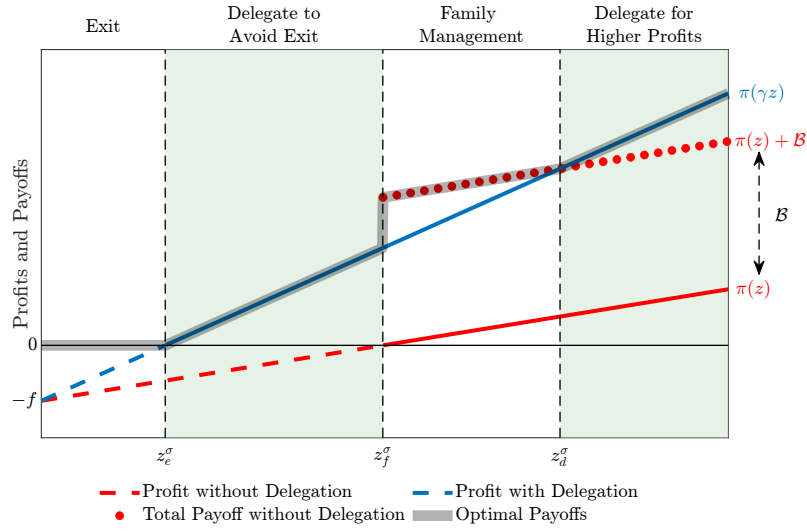
would still not earn enough revenue to cover the fixed cost of operating. Thus, in that *exit* region, monetary profits are below zero, and no strategy can prevent the firm's failure.

Between z_e and z_f (the second dashed line), firms survive *only* if they professionalize. As the red dashed line (profit without professionalization) remains below zero in this region, these "laggard" firms cannot cover their fixed costs under family management; however, by delegating management to outside professionals, they can increase their productivity parameter to γz, thereby generating positive profits. I refer to this as *negative selection into professionalizing management* since relatively unproductive firms upgrade their management out of necessity rather than choice.

Once firm productivity surpasses z_f, indicated by the second dashed line, the firm would be profitable even as a family firm. The red dashed line ($\pi(z)$) is now above zero, reflecting positive profits. However, in the intermediate region $[z_f, z_d)$, it is optimal for the owner to *retain family management*: while monetary profits from delegating (solid blue line, $\pi(\gamma z)$) are higher than monetary profits from running the firm as a family firm (solid red line), they are not large enough to offset the owner's loss of the private benefit $\mathcal{B}$ (depicted as the vertical gap in the figure).

Finally, at the third vertical dashed line, z_d, we reach a threshold beyond which high-productivity "frontier" firms *professionalize to increase profits* rather than to avoid exit. Past this cutoff, the profit increase from γz exceeds the loss of the private benefit. Hence, the blue dashed curve representing $\pi(\gamma z)$ lies above the red dashed curve plus the shaded private-benefit segment. This is termed *positive selection into professionalizing management*, capturing how the most productive firms choose professional management to further boost their earnings. The thick black line shows optimal firm payoffs for different levels of productivity.

6.1 Comparative Static: Unilateral Import Competition

In this section, I analyze the partial equilibrium impact of a *unilateral trade liberalization*, like the removal of QRs, that takes the form of an exogenous increase in foreign varieties, lowering the aggregate price index, P. From equation (5), recall that firm profits depend on the composite market demand parameter

$$ A \;=\; \frac{1}{\sigma}\,\rho^{\sigma-1}\,E\,P^{\sigma-1}\,w^{1-\sigma}. $$

A fall in P directly reduces A, thereby lowering per-period profits for *all* firms in the domestic market. As profits shrink, the key productivity thresholds derived in the previous section respond as follows:

(i) **Rise in the Exit Thresholds.** From equations (7) and (8), both z_f (the minimum productivity required for profitable operation under family management) and z_e (the minimum productivity under professionalization) increase when A decreases. Intuitively, each firm's revenues drop, so it becomes more difficult to cover the fixed production cost wf. This implies:

   - **More exit at the bottom:** As z_e rises, some firms in the lower tail now fall below this new, higher productivity threshold and must exit the market, even if they would have been able to survive under professionalization prior to the shock.

   - **More negative-selection professionalization :** Because z_f also rises, there is a broader range of "laggard" firms whose productivity *as a family firm* would no longer cover fixed costs. Such firms now *must* professionalize to boost productivity to γz in order to avoid exit.

(ii) **Rise in the Frontier Professionalization Threshold.** Turning to equation (9), the cutoff z_d at which a firm is indifferent between retaining family management and delegating (purely to enhance profits) also increases as A falls. In essence, lower market-wide profitability makes the private benefit $\mathcal{B}$ more valuable in relative terms, while the additional profit gain from professionalization (beyond covering costs) is smaller. Consequently, fewer *frontier* firms (i.e. those above z_d) wish to professionalize purely for revenue expansion.

To explicitly connect professionalization decisions across periods $t = 0, 1$, define indicator variables $\mathcal{P}_t$, where $\mathcal{P}_t = 1$ if the firm professionalizes management in period t, and $\mathcal{P}_t = 0$ otherwise. Professionalization boosts productivity from z to γz, with $\gamma > 1$, but entails the loss of private benefits B. Reverting to family management after previously delegating incurs a switching cost κ. Thus, period-1 profits are given by:

$$ \pi(z, \mathcal{P}_0, \mathcal{P}_1) = \mathcal{P}_1 \cdot \pi(\gamma z) + (1 - \mathcal{P}_1)\big[\pi(z) + \mathcal{B} - \kappa\mathcal{P}_0\big] $$

Case 1: $\mathcal{P}_1 = 1$. If the firm professionalizes in period $t = 1$, profits are $\pi(\gamma z)$, irrespective of previous choices.

Case 2: $\mathcal{P}_1 = 0$. If the firm remains family-managed in period $t = 1$, profits depend on past choices. For firms family-managed in period $t = 0$ ($\mathcal{P}_0 = 0$), profits are $\pi(z) + \mathcal{B}$. However, firms previously delegating in period $t = 0$ ($\mathcal{P}_0 = 1$) incur the switching cost κ, earning profits $\pi(z) + \mathcal{B} - \kappa$. As discussed in the section, the empirical evidence presented in Section 5 is indicative of such a high value of κ.

Figure 9: Professionalization Choice after Import Competition Shock

(a) Comparative Static for Laggard Firms: Trade-Induced Professionalization to Avoid Exit
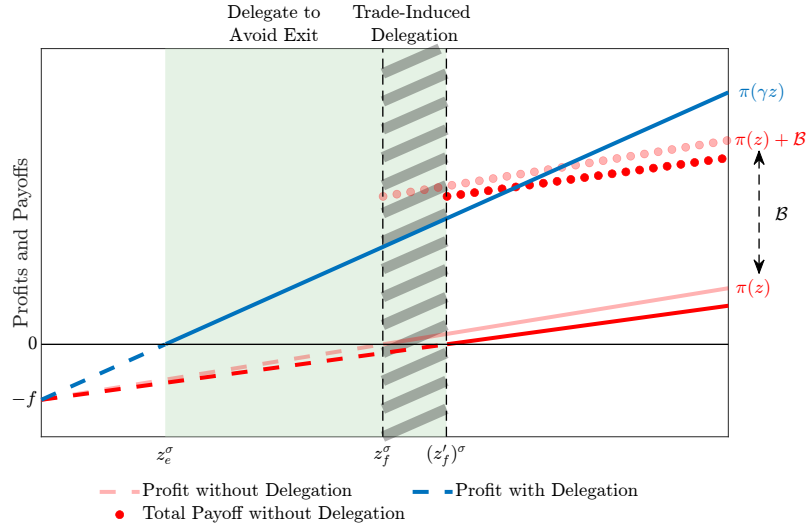
(b) Comparative Static for Frontier Firms: Regret Past Professionalization
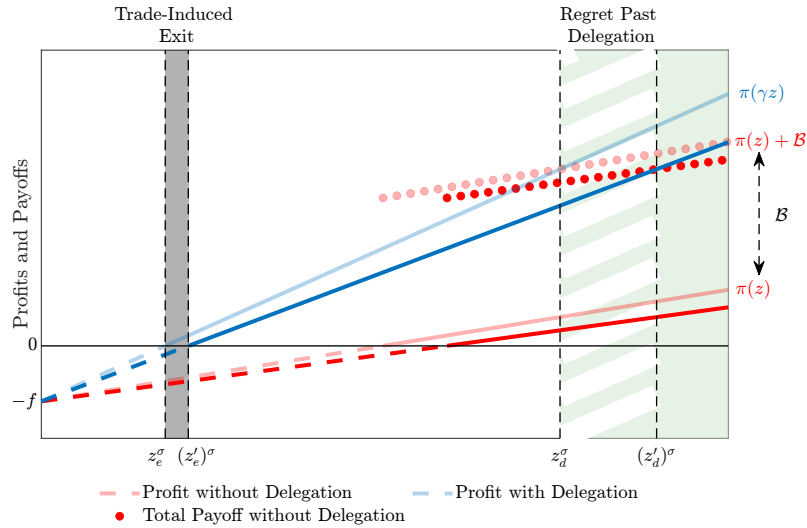
Figure 9 illustrates the heterogeneous responses of firm profits to the import competition shock induced by the removal of quantitative restrictions (QRs). The dashed lines indicate pre-shock profit levels, while solid lines represent post-shock profits. Panel (a) shows that for low-productivity 'laggard' firms, the import competition shock reduces profits significantly below their pre-shock levels, compelling these firms to delegate management to external professionals as a survival strategy. Panel (b) depicts that for high-productivity 'frontier' firms, lower profits following the shock, reduce the appeal of professionalization relative to private benefits.

6.2   Comparing Model Predictions with Empirical Evidence

The comparative-statics results of the previous section suggest that a unilateral trade shock– modeled here as an exogenous increase in foreign varieties that lowers the aggregate price index P– leads to two central predictions about professionalization among domestic firms:

  (i) *Exit and "Negative-Selection" Professionalization Rise for Laggard Firms.* Because the shock reduces the composite profitability parameter A, both the exit cutoffs z_e and z_f increase. More low-productivity (or "laggard") firms find themselves at risk of making negative profits if they retain family management; these firms can survive only by delegating and thus enjoying the productivity boost γz. This group therefore exhibits a surge in professionalization driven by the need to avert exit.

  (ii) *Professionalization Becomes Less Attractive for Frontier Firms.* The higher "frontier" threshold z_d also moves upward, reducing the fraction of large, highly-productive firms that choose to professionalize solely for profit enhancement. In other words, lower overall profitability tightens the trade-off between forgoing private benefits and realizing higher productivity, leading fewer of the most productive firms to initiate professionalization *ex-post*. Whether this change in incentives induces professional firms to switch back to family to not depends on how high the re-switching costs, κ are.

These predictions align closely with the empirical results shown in Section 5, where the removal of quantitative restrictions (QRs) in India increased import competition and generated new incentives for organizational change in domestic firms. Below, I highlight how the main findings map to the theoretical comparative statics:

A Rise in Professionalization among Laggard Firms.   Figure 6 documents that, after the removal of QRs, there is a notable decline in the share of family members occupying top executive positions. Moreover, panel (b) in Figure 6 clarifies that this *aggregate* shift is driven almost entirely by firms in the *bottom tercile of pre-policy productivity*. This selection pattern is precisely the "negative selection into professionalization " mechanism described in the model: less-productive firms– those closest to the exit threshold– are the ones that restructure their top management in order to boost productivity and mitigate the profitability shock. In short, the firms whose survival is most imperiled by rising foreign competition are precisely the ones that replace family managers with external professionals. The fact that the total number of professionalized firms increases, while highly productive incumbents *do not* systematically shift from family to professional management, confirms that it is predominantly the negative-selection professionalization margin that shapes the new equilibrium under unilateral trade liberalization.

No Change in Professionalization among Frontier Firms.   As discussed in the previous section, import competition makes professionalization less appealing for frontier firms. As heightened competition reduces overall profitability, the private benefit $\mathcal{B}$ looms larger in the firm's payoff calculation, making the incremental profit gains from professionalization insufficient to justify forgoing family control. In the data, there is

little evidence that the most productive firms respond to the shock by reversing professionalization . Empirically, higher-productivity enterprises do not show any change in the share of family members on the executive board (Figure 6(b)). This is indicative of the high cost of switching back from professional to family firms, κ.

6.3 Future Extension: Calibration and Policy Counterfactuals

I plan three key extensions to the theoretical framework that make the model suitable for estimation.

1. **Heterogeneity in Private Benefits:** Instead of assuming a common $\mathcal{B}$ for all family-managed enterprises, the model should allow for firm-specific private benefit draws. In the data, firms with similar productivity levels sometimes report different delegation decisions. By introducing heterogeneity in private benefits, the model can more accurately capture how individual firms' unique valuations of retaining family control influence their delegation choices. I also assume that if a firm professionalizes, it retains some portion of its private benefits rather than losing all its private benefits when it professionalizes.

2. **Uncertainty in the Import Competition Shock:** Incorporate uncertainty regarding the magnitude of the import competition shock. In reality, firms are often uncertain about the severity of the competitive pressure they face. This added uncertainty makes the delegation decision more sensitive to the private benefit parameter, as firms weigh the risk of the shock more directly in their managerial restructuring.

3. **Continuous Delegation Decision:** Extend the model by replacing the current binary delegation decision with a continuous one. This modification will better align the model's predictions with the data, which capture not only whether firms delegate but also the extent to which they do so.

6.3.1 Model Extension

I outline extensions 1 and 2 below.

Let private benefits $\mathcal{B}_i$ vary by firm. Furthermore, assume that the firm does not perfectly observe shocks to market demand A. If a firm stays a family firm it enjoys all its private benefits, $\mathcal{B}_i$. If it professionalize, it only retains $\delta \mathcal{B}_i$, where $\delta < 1$. On professionalizing the firm receives a boost to its productivity, by $\gamma > 1$, as before.

Suppose that aggregate demand falls to $\varepsilon \cdot A$, $\varepsilon \in [0,1]$ due to import competition. But the firm is uncertain about the extent of the shock. If the shock is very severe, the firm exits. For family firms, this happens when $\varepsilon < \varepsilon_f^*(A) \equiv \frac{f}{Az^{\sigma-1}}$ because at such low ε, a family firm is unable to make positive profits. Similarly, a professional firm exits if $\varepsilon < \varepsilon_p^*(A) \equiv \frac{f}{A(\gamma z)^{\sigma-1}}$. Thus the probability of exit for a family firm is given by $\mathbb{P}(\varepsilon < \varepsilon_f^*)$ and that of a professional firm is $\mathbb{P}(\varepsilon < \varepsilon_p^*)$.

The expected utility for a family firm is given by

$$U_F(z) = \int_{\varepsilon_F^*}^1 \left[\varepsilon \cdot Az^{\sigma-1} - f + \mathcal{B}_i \right] f(\varepsilon) d\varepsilon$$

$$= \underbrace{\left(\int_{\varepsilon_F^*}^1 f(\varepsilon) d\varepsilon \right)}_{\substack{\text{Probability of} \\ \text{Survival}}} \times \left[\underbrace{\frac{\int_{\varepsilon_F^*}^1 \varepsilon f(\varepsilon) d\varepsilon}{\int_{\varepsilon_F^*}^1 f(\varepsilon) d\varepsilon}}_{\mathbb{E}(\varepsilon | \varepsilon > \varepsilon_F^*)} Az^{\sigma-1} - f + \mathcal{B}_i \right]$$

$$= \mathbb{P}(\varepsilon > \varepsilon_F^*) \cdot \left(\mathbb{E}(\varepsilon \mid \varepsilon > \varepsilon_F^*) Az^{\sigma-1} - f + \mathcal{B}_i \right)$$

Similarly, the expected payoff of professional firm is

$$U_P(z) = \mathbb{P}(\varepsilon > \varepsilon_P^*) \cdot \left(\mathbb{E}(\varepsilon \mid \varepsilon > \varepsilon_P^*) A(\gamma z)^{\sigma-1} - f + \delta \mathcal{B}_i \right)$$

A firm decides to professionalize if $U_P(z) - U_F(z) > 0$. If $f(\varepsilon)$ is Uniform distribution, that is, $f(\varepsilon) = \begin{cases} 1, & \varepsilon \in [0,1] \\ 0, & \text{otherwise} \end{cases}$, then it can be shown that

$$U_P(z) - U_F(z) = \frac{1}{2} \left(\gamma^{\sigma-1} - 1 \right) \cdot \left(Az^{\sigma-1} - \left(\frac{z_f}{\gamma z} \right)^{\sigma-1} f \right)$$

$$+ \mathcal{B}_i \left(\delta - 1 + \left(\frac{z_f}{z} \right)^{\sigma-1} \cdot \left(1 - \frac{\delta}{\gamma^{\sigma-1}} \right) \right)$$

$$= \Delta(z, A) + \mathcal{B}_i \Phi(z, A) \tag{10}$$

Thus $U_P(z) - U_F(z)$ depends on two terms.

1. $\Delta(z, A) = \frac{1}{2} \left(\gamma^{\sigma-1} - 1 \right) \cdot \left(Az^{\sigma-1} - \left(\frac{z_f}{\gamma z} \right)^{\sigma-1} f \right)$: this is the direct increase in monetary profits as a result of legation, which boosts firm productivity. This effect is similar to the value of cost reduction in the framework of Schmidt (1997).

2. $\mathcal{B}_i \Phi(z, A) = \mathcal{B}_i \left(\delta - 1 + \left(\frac{z_f}{z} \right)^{\sigma-1} \cdot \left(1 - \frac{\delta}{\gamma^{\sigma-1}} \right) \right)$. $\mathcal{B}_i$ private benefits have two opposite effects on likelihood of delegation:

   (a) $\delta - 1 < 0$ term: when B_i is large, a family firm has a lot of private benefits to lose from hiring an outside manager. Indeed, professionalizing means you only keep $\delta \mathcal{B}_i$ instead of the full $\mathcal{B}_i$. Everything else equal, this term by itself reduces the incentive to professionalize. As the firm's private benefits go up, they are more reluctant to delegate authority to an outside professional because they forfeit $(1 - \delta)\mathcal{B}_i$.

(b) $\left(\frac{z_f}{z}\right)^{\sigma-1} \cdot \left(1 - \frac{\delta}{\gamma^{\sigma-1}}\right)$: Professionalizing improves survival odds, so the firm might preserve some fraction of the private benefits at all– whereas if the firm stayed family-managed and ended up exiting, it would get zero. The higher productivity (via γ) lowers the exit threshold, and that makes it more likely the firm keeps whatever fraction δ of the private benefits. If the firm did not professionalize, it might lose everything by dropping out of the market in a deep shock. This effect is larger, the larger is $\frac{z_f}{z}$, i.e. the close firm productivity z is to the exit threshold. Firms close to the exit threshold particularly care about this term. In the limit when $\frac{z_f}{z} = 1$, i.e. the firm is at the verge of exit, the whole term reduces to $\delta \mathcal{B}_i \left(1 - \frac{1}{\gamma^{\sigma-1}}\right)$. Since $1 - \frac{1}{\gamma^{\sigma-1}} > 0$, this implies that firms with higher $\mathcal{B}_i$ are more likely to delegate.

Thus, large $\mathcal{B}_i$ does not necessarily mean the net effect is negative. Although the direct part from $\delta - 1$ tilts against professionalization, the firm also wants to factor in the effect that professionalizing raises the chance those benefits do not get destroyed by exit.

6.3.2 Comparative Static

As with the baseline model, the impact of import competition on delegation is ambiguous and depends on firm productivity and private benefits.

- $\frac{\partial}{\partial A}\left(\Delta(z; A)\right) < 0$: value of cost-reduction falls when there's lower market demand. This is because lower market demand A reduces profits for all productivity levels.

- $\frac{\partial}{\partial A}\left(\Phi(z; A)\right) > 0$: value of avoiding exit goes up because the firm is more likely to exit as market demand falls.

Thus, the overall effect of increased import competition is ambiguous.

6.3.3 Structural Estimation

There are three aims of structural estimation. First, estimate the distribution of private benefits. Second, to measure the aggregate productivity impact of trade-induced professionalization. Anf Finally, estimated model opens the door to policy experiments that vary private-benefit parameters or delegation costs in ways that go beyond historical events,

For estimation purposes, I make additional assumptions regarding the distributions of firm productivity and private benefits. I assume that firms draw productivity from a Pareto distribution $G(z) = 1 - z^{-k}$ with $k > 1$ and $k > \sigma - 1$. I plan to estimate k by based on the empirical distribution of firm productivity in the data. I assume that private benefits are drawn from a Log Normal distribution with mean μ and variance ν.

The set of parameters to estimate is $\Theta = \{f, \sigma, \gamma, k, \mu, \nu\}$. The fixed cost f will be estimated by matching firm exit rates observed in the data. I set $\sigma = 5$ based on the literature (Fan et al., 2023). I set $\gamma = 1.2$ based on the empirical evidence provided in

Figure 7 (a). This can be thought of an upper bound for the impact of professionalization on firm productivity. I plan to conduct sensitivity analysis based on a range of values for γ. The Pareto tail parameter for firm productivity, k, will be estimated by matching the empirical distribution of firm productivity.

Identifying the Distribution of Private Benefits. A central empirical objective is to use firms' observed delegation decisions during periods of intensified import competition to back out the distribution of $\mathcal{B}_i$. In particular, once I specify that private benefits follow a Log Normal distribution with mean μ and variance ν, I can exploit the discrete decision rule in (10).

This rule states that a firm will delegate if and only if $\Delta(z, A) + \mathcal{B}_i \Phi(z, A) > 0$. Conditional on observing a firm's productivity z and its management choice (professionalize vs. remain family-managed), I learn whether its realization of $\mathcal{B}_i$ is above or below the threshold $-\Delta(z, A)/\Phi(z, A)$. While a single binary outcome for an individual firm pins down only an inequality rather than an exact $\mathcal{B}_i$, assembling choices from many firms at different productivity levels yields constraints on the entire distribution of private benefits.

A practical way to implement this in the data is via a *simulation-based* method of moments or likelihood approach. First, I draw a large sample of firms' productivities z from a Pareto distribution and private benefits $\mathcal{B}$ from a Log Normal $\ln \mathcal{B} \sim N(\mu, \nu)$. For each simulated firm, I compute $\Delta(z)$ and $\Phi(z)$, check whether $\Delta(z) + \Phi(z)\mathcal{B} > 0$, and thus determine a "model-implied" delegation decision. By grouping these simulated firms into productivity bins and computing the fraction that delegate in each bin, I obtain the predicted delegation shares as a function of $\{\mu, \nu\}$. Finally, I choose the parameters μ, ν to minimize the gap between these simulated shares and the actual delegation rates observed across productivity bins in the data, thereby identifying the distribution of private benefits.

Quantifying Aggregate Effects. A second goal of the estimation is to measure the impact of trade-induced delegation on aggregate productivity. The structural estimation allows me to compute the counterfactual distribution of active firms across different management modes and productivity levels– both with and without trade liberalization. I will track how total output, average productivity, and the mass of active firms change when a trade shock induces additional delegation among lower-z businesses.

Policy Counterfactuals. Finally, the estimated model opens the door to policy experiments that vary private-benefit parameters or delegation costs in ways that go beyond historical events. For example, I can set all $\mathcal{B}_i = 0$ and re-solve the model to see how many more firms would delegate and how aggregate productivity and welfare would evolve under the same import shock. Alternatively, I can impose a hypothetical subsidy or tax on the delegation to see how reducing (or increasing) the cost of finding and recruiting professional managers changes firms' incentives to professionalize in the face of import competition. Because the model captures both *negative-selection* and *positive-selection* delegation, one can evaluate whether subsidizing delegation disproportionately aids lower-productivity firms– or whether it also spurs additional restructuring among higher-productivity businesses.

Thus, moving to a setting with heterogeneous private benefits and uncertainty provides a richer and more realistic account of firms' delegation decisions. The quantitative exercise described above enables me to evaluate alternative policy scenarios, such as subsidies for professional managers and their potential to reshape corporate governance and boost aggregate productivity in the wake of trade shocks.

7 Conclusion

This study demonstrates that trade liberalization, through a productspecific import competition shock, reshapes firm management structures in a profound way. Focusing on family firms– the predominant mode of corporate governance in many developing countries– I find that heightened import competition compels these firms to undertake significant managerial turnover. Empirically, firms facing increased foreign competition are more likely to replace family managers with professional executives, a shift that is closely associated with improved withinfirm productivity. The evidence, drawn from a novel managerfirm matched dataset and detailed board director tenure records for over 6 million Indian directors, reveals that the restructuring of top management is not merely a byproduct of declining sales or contracting firm size. Instead, it reflects a deliberate organizational response to external competitive pressures.

The event study analysis highlights that following the removal of quantitative restrictions, family-controlled firms– particularly those with lower pre-policy productivity– experience a marked decline in the share of family members on their executive boards. This replacement is accompanied by a corresponding rise in non-family professional managers, suggesting a one-to-one substitution effect. Furthermore, these changes in management composition are linked to subsequent productivity improvements, as evidenced by rising revenue and quantity productivity measures and declining average output prices. Such findings support the broader hypothesis that organizational reform is a key channel through which competition enhances X-efficiency.

To further interpret these empirical results, I develop a simple model of industrial equilibrium in which family firms face a trade-off between the non-monetary private benefits of retaining family management and the monetary gains from delegating management to professionals. The model predicts a dual-selection mechanism: less productive family firms delegate management out of necessity to avert exit, while more productive firms delegate to further boost efficiency. The data primarily reflect the negative–selection channel, with laggard firms undergoing managerial changes to survive under harsher competitive conditions.

These insights have important policy implications. In contexts where family firms dominate, trade liberalization can trigger internal restructuring that not only improves firm-level productivity but also contributes to aggregate efficiency gains. Future work should extend this analysis by structurally estimating the model, incorporating heterogeneous private benefits across firms, and evaluating policy counterfactuals– such as subsidies for managerial professionalization– to better understand the aggregate impact of these organizational adjustments.

In sum, this paper contributes to the literature on trade-induced productivity improvements by highlighting an often-overlooked internal adjustment mechanism. It under-

scores the importance of managerial innovation in response to external shocks and suggests that policies promoting transparency and competitive pressures can indirectly stimulate organizational reforms that bolster firm performance.

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A Appendix

A.1 Appendix: Data

Figure A1: Example of a 1998 Government of India Policy Notification Mentioning Product Codes for which QRs were Removed

S.O. 321(E).—In exercise of the powers conferred by section 5 of the Foreign Trade (Development and Regulation) Act, 1992 (No. 22 of 1992) read with paragraph 4.1 of Export and Import Policy, 1997-2002, the Central Government hereby makes the following amendments in the ITC (HS) Classifications of Export and Import Items, 1997-2002, published on 31st March, 1997 and as amended from time to time. In respect of following Exim Code Nos., policy indicated in columns 3, 4 and 5 shall be amended to read as under.

Exim Code	Item Description	Policy	Conditions relating to the Policy	Import under SIL/Public Notice
03061301	Shrimp (scampi) macrobactium frozen	Free		
03061302	AFD shrimp frozen	Free		
03061303	Prawns frozen	Free		
03061400	Crabs	Free		
03061900	Other, including flours, meals and pellets of crustaceans. fit for human	Free		

Notes: This figure shows an example of a government notification issued by the Ministry of Commerce in 1998, detailing the products for which quantitative restrictions were adjusted. The first column lists the 8-digit ITC HS codes, with corresponding product descriptions in the second column. The third column, labeled "Policy", indicates the status of restrictions; "Free" signifies that QRs on that specific product have been lifted. More than 30 such notifications, spanning over 1,000 pages, were digitized to create a novel dataset on product-level quantitative restrictions in India. *Source:* Ministry of Commerce, Government of India.

Table A1: Novel Product Concordances

Prowess Products		ITC HS Products		Industry
3008040800	Sunflower seed oil	15121910	Sunflower oil edible grade	15142	Manufacture of vegetable oils and fats, excluding corn oil.
3008040804	Sunflower seed oil, refined				
4012080400	Suitcases	420212.04	Plastic moulded suit-cases	19121	Manufacture of travel goods like suitcases, bags and holdalls etc.
5024200404	Distempers	321000.01	Distempers	24222	Manufacture of paints, varnishes, enamels or lacquers.
6308361216	Washing Machines/Laundry Mach	84501100	Fully - automatic washing machines (upto 10kg)	29308	Manufacture of other electric domestic appliances n.e.c.: dishwashers, household type laundry equipment, electric razors including parts and accessories for electrical domestic appliances
		84501200	Other washing machines with built-in centrifugal drier (upto 10kg)		
		84501300	Other washing machines (upto 10kg)		

Notes: This table shows a mapping of ITC HS codes to both NIC industry codes and Prowess product codes. Two separate mapping exercises were conducted to achieve this concordance. First, 8-digit ITC HS codes were mapped to NIC industry codes at the 4-digit level using the HI to I3 concordance provided by the World Bank, which served as a foundation for further manual extension to the 5-digit NIC 1998 codes. Second, an ITC HS to Prowess mapping was created at the most granular level, linking 8-digit ITC HS codes to 10-digit Prowess codes based on product descriptions. This concordance links over 2,700 HS products to more than 6,000 Prowess products across 400 industries, providing a comprehensive framework for analyzing product-level and industry-level relationships. *Source:* CMIE Prowess and Ministry of Corporate Affairs, Government of India.

A.2   Appendix: Results

A.2.1   Aggregate Product-Level Analysis

Table A2: Value and Quantity of Imports Increase after QR Removal with No Impact on Exports

	(1) Import Value	(2) Import Quantity	(3) Export Value	(4) Export Quantity
6 years before event	0.054	-0.16	0.034	0.040
	(0.10)	(0.15)	(0.087)	(0.12)
5 years before event	0.00060	-0.031	0.060	0.010
	(0.091)	(0.13)	(0.082)	(0.11)
4 years before event	0.020	-0.0040	0.090	0.13
	(0.084)	(0.13)	(0.072)	(0.099)
3 years before event	0.045	0.030	0.067	0.15*
	(0.072)	(0.12)	(0.054)	(0.078)
2 years before event	0.044	0.072	0.035	0.054
	(0.059)	(0.13)	(0.045)	(0.073)
Year of event	0.20***	0.19**	-0.0081	0.083
	(0.058)	(0.092)	(0.043)	(0.064)
1 year after event	0.39***	0.52***	-0.033	0.041
	(0.062)	(0.10)	(0.052)	(0.074)
2 years after event	0.48***	0.71***	-0.038	0.066
	(0.070)	(0.11)	(0.061)	(0.091)
3 years after event	0.54***	0.77***	-0.084	-0.047
	(0.079)	(0.13)	(0.072)	(0.10)
4 years after event	0.59***	0.86***	-0.036	0.16
	(0.080)	(0.13)	(0.076)	(0.10)
5 years after event	0.58***	0.93***	-0.052	0.11
	(0.089)	(0.14)	(0.083)	(0.11)
6 years after event	0.59***	0.91***	-0.079	0.11
	(0.092)	(0.14)	(0.089)	(0.11)
7 years after event	0.57***	0.89***	-0.14	0.014
	(0.096)	(0.14)	(0.094)	(0.12)
8 years after event	0.64***	1.03***	-0.059	0.14
	(0.11)	(0.16)	(0.10)	(0.13)
HS-6 digit FE	✓	✓	✓	✓
HS-4 digit × Year FE	✓	✓	✓	✓
Observations	90364	89911	93222	92503
R^2	0.89	0.91	0.86	0.89

Notes: The table shows the results of running the event study specification (1). The dependent variables are log import value (column (1)), log import quantity (column (2)), log export value (column (3)), and log export quantity (column (4)). An HS-6 digit product is identified as treated if QRs were removed from any of its constituent HS-8 digit products. β_{-1}, the coefficient prior to the year in which QRs were removed, is normalized to zero. The policy is staggered from 1995 to 2001. All regressions include HS-6 digit product fixed effects and HS-4 digit product × year fixed effects. Clustering of standard errors is done at HS-6 digit product. Standard errors are in parenthesis. All columns report event study estimates plotted in Figure 4. ***, ** and * denote statistical significance at the 1%, 5%, and 10% levels, respectively. *Source:* CMIE Trade$_{dx}$.

Figure A2: Value and Quantity of Imports Increase after QR Removal with No Impact on Exports: Robustness to Two-Way Fixed Effects Estimator

(a) Log Import Value

(b) Log Import Quantity

(c) Log Export Value

(d) Log Export Quantity
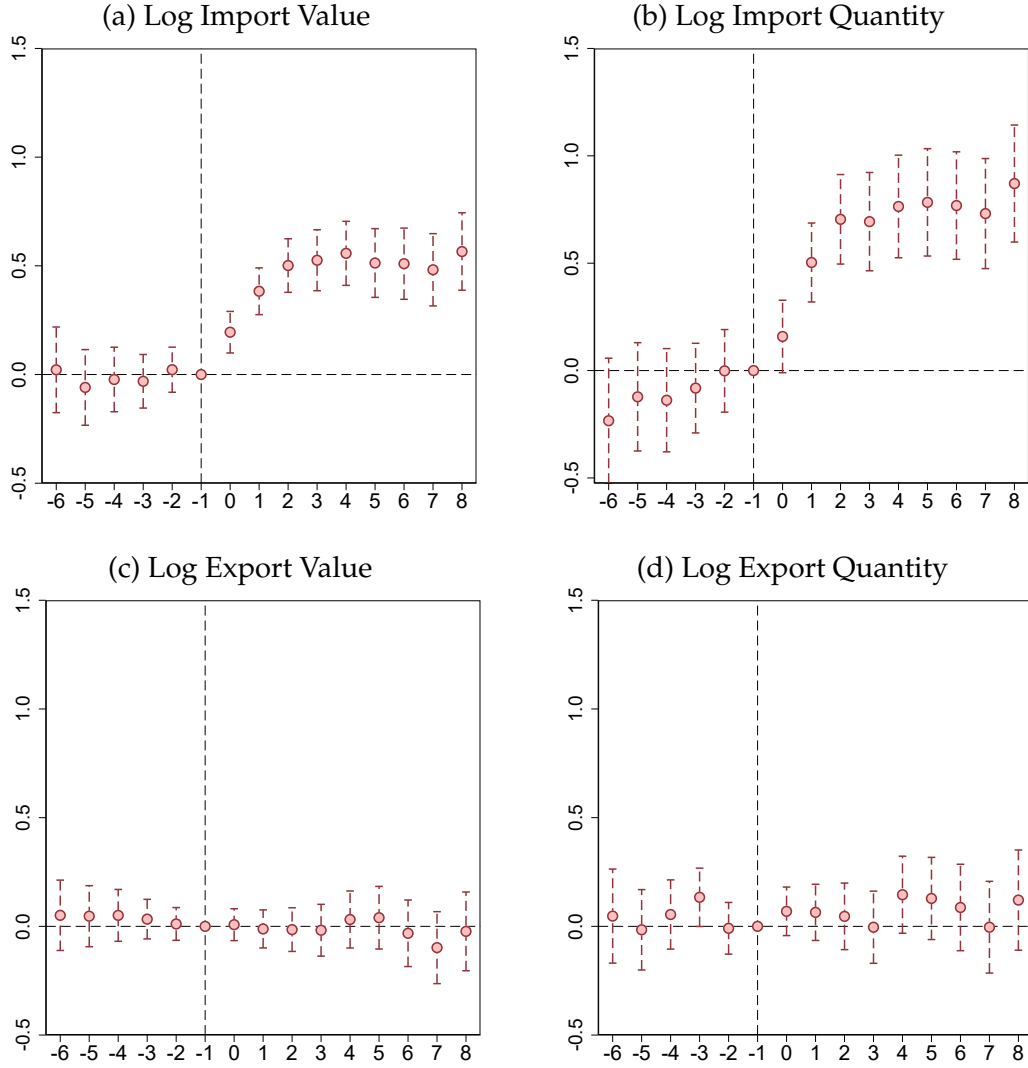
Notes: The figure presents β_k event study coefficients from Equation (1) using the two-way fixed estimator on annual HS-6 digit product-level panel data on imports and exports. The dependent variables are log import value (panel (a)), log import quantity (panel (b)), log export value (panel (c)), and log export quantity (panel (d)). An HS-6 digit product is identified as treated if QRs were removed from any of its constituent HS-8 digit products. β_{-1}, the coefficient prior to the year in which QRs were removed, is normalized to zero. The policy is staggered from 1995 to 2001, with the *x*-axis denoting years relative to the event. All regressions include HS-6 digit product fixed effects and HS-4 digit product $\times$ year fixed effects. Standard errors are clustered by HS-6 digit product. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Trade$_{dx}$.
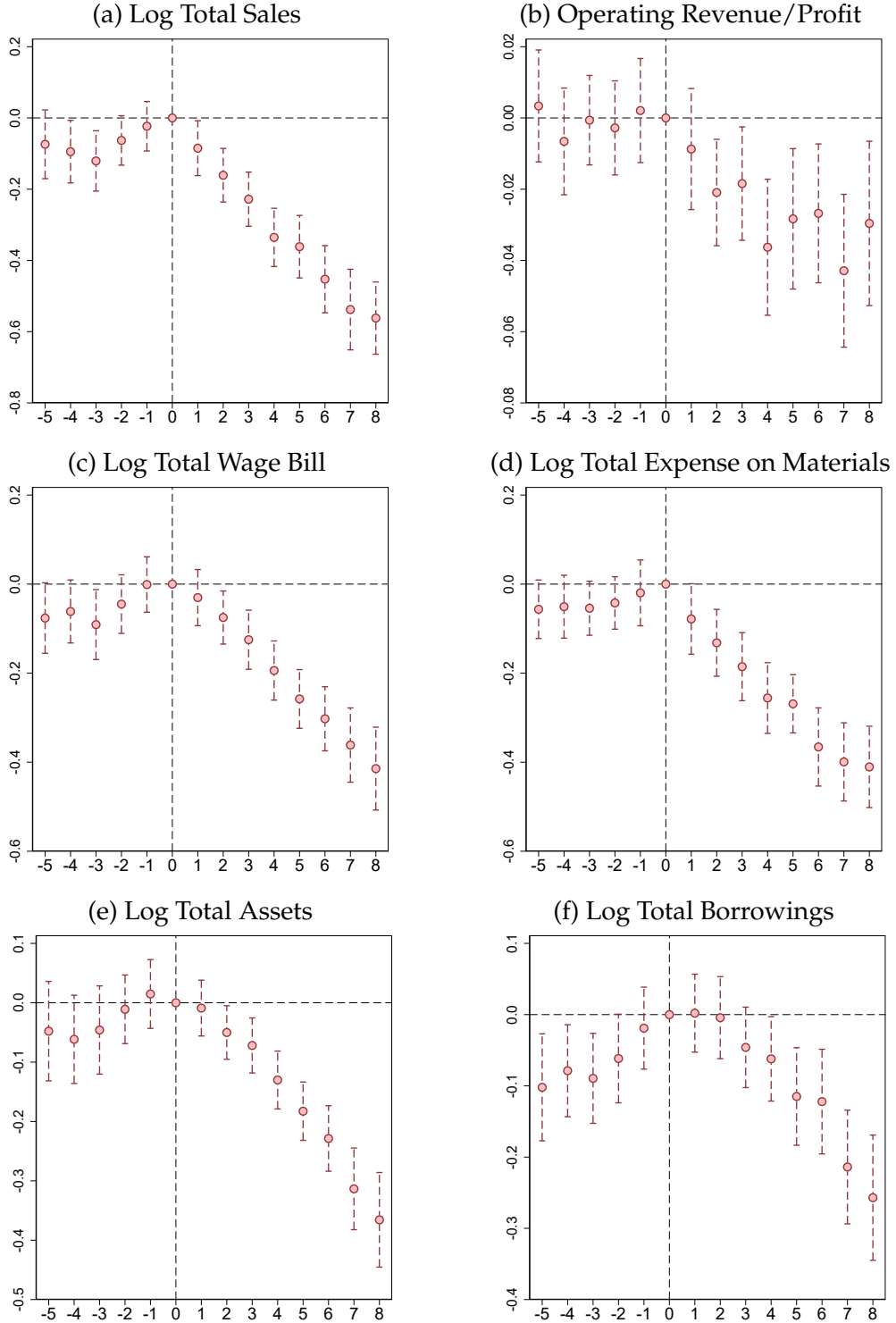
A.2.2 Firm-Level Analysis

A.2.3 Firm Level Event Studies: Financial Indicators

Table A3: Domestic Firms Contract after QR Removal

|	(1)	(2)	(3)	(4)	(5)	(6)
	Total Sales	Operating Profits/ Revenue	Total Wage Bill	Total Expense on Materials	Total Assets	Total Borrowings
5 years before event	-0.081*	-0.0031	-0.098***	-0.060*	-0.051	-0.078**
	(0.042)	(0.0065)	(0.034)	(0.035)	(0.033)	(0.038)
4 years before event	-0.11**	-0.016**	-0.074**	-0.057	-0.075**	-0.057
	(0.042)	(0.0071)	(0.036)	(0.037)	(0.031)	(0.037)
3 years before event	-0.14***	-0.0090	-0.11***	-0.061*	-0.050	-0.056
	(0.040)	(0.0064)	(0.035)	(0.033)	(0.034)	(0.037)
2 years before event	-0.069*	-0.0065	-0.056*	-0.036	-0.010	-0.044
	(0.035)	(0.0067)	(0.031)	(0.035)	(0.029)	(0.033)
1 year before event	-0.025	-0.0014	-0.0092	-0.019	0.031	0.011
	(0.035)	(0.0074)	(0.029)	(0.038)	(0.027)	(0.033)
1 year after event	-0.082**	-0.010	-0.048	-0.088**	-0.0041	0.027
	(0.036)	(0.0081)	(0.030)	(0.036)	(0.024)	(0.033)
2 years after event	-0.14***	-0.015**	-0.081***	-0.15***	-0.041*	0.026
	(0.037)	(0.0074)	(0.031)	(0.034)	(0.023)	(0.033)
3 years after event	-0.21***	-0.014*	-0.14***	-0.19***	-0.052**	-0.0073
	(0.036)	(0.0086)	(0.033)	(0.037)	(0.023)	(0.035)
4 years after event	-0.32***	-0.024***	-0.20***	-0.26***	-0.11***	-0.033
	(0.040)	(0.0090)	(0.032)	(0.038)	(0.024)	(0.034)
5 years after event	-0.36***	-0.021**	-0.26***	-0.28***	-0.16***	-0.085**
	(0.042)	(0.0091)	(0.032)	(0.036)	(0.025)	(0.035)
6 years after event	-0.43***	-0.021**	-0.30***	-0.36***	-0.21***	-0.11***
	(0.044)	(0.0091)	(0.034)	(0.047)	(0.028)	(0.039)
7 years after event	-0.53***	-0.032***	-0.38***	-0.40***	-0.30***	-0.20***
	(0.050)	(0.0089)	(0.041)	(0.044)	(0.033)	(0.042)
8 years after event	-0.56***	-0.024**	-0.44***	-0.41***	-0.35***	-0.25***
	(0.048)	(0.010)	(0.044)	(0.048)	(0.035)	(0.049)
Firm FE	✓	✓	✓	✓	✓	✓
Year FE	✓	✓	✓	✓	✓	✓
Observations	64686	64686	65638	57562	69461	65138
R^2	0.76	0.36	0.82	0.78	0.83	0.75

Notes: The table shows the results of running the event study specification (2). The dependent variables are: log total sales (column (1)), operating profits/revenue (column (2)), log total wage bill (column (3)), log total expense on materials (column (4)), log total assets (column (5)), and log total borrowings (column (6)). A firm is identified as treated in a year if QRs were removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001. Clustering of standard errors is at the four-digit industry × year. Standard errors are in parenthesis. ***, ** and * denote statistical significance at the 1%, 5%, and 10% levels, respectively. *Source:* CMIE Prowess$_{dx}$.

Figure A3: Domestic Firms Contract after QR Removal: Robustness to Two-way Fixed Effects Estimator
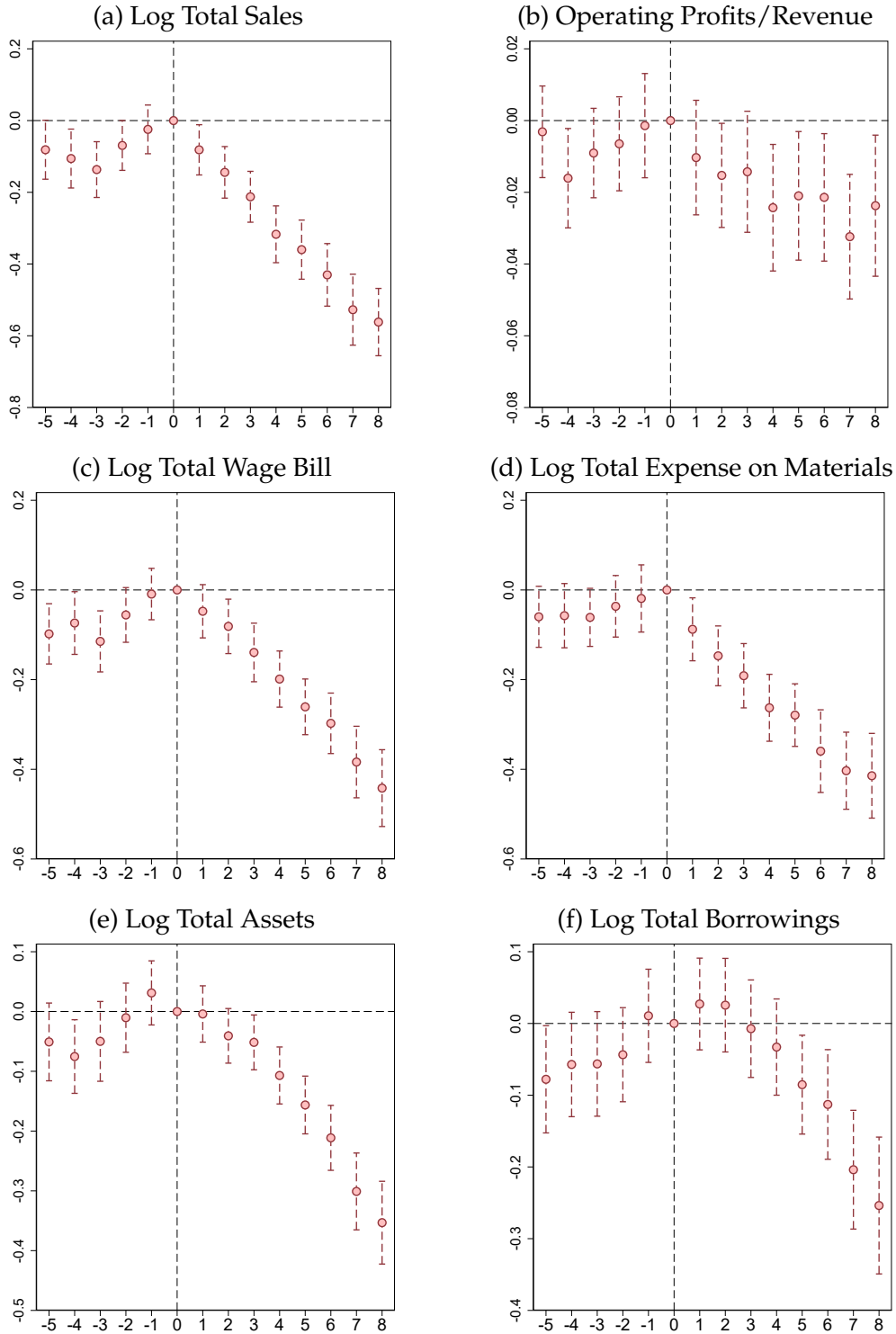

(a) Log Total Sales

(b) Operating Revenue/Profit

(c) Log Total Wage Bill

(d) Log Total Expense on Materials

(e) Log Total Assets

(f) Log Total Borrowings

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2), estimated using two-way fixed estimator. The dependent variables are: log total sales (panel (a)), operating profits/revenue (panel (b)), log total compensation (panel (c)), log total expense on raw materials (panel (d)), log total assets (panel (e)), and log total borrowings (panel (f)). A firm is identified as treated in a year if QRs were removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the *x*-axis denoting years relative to the event. All regressions include firm and three-digit industry × year fixed effects. Standard errors are clustered at the three-digit industry × year level. The vertical lines are the 95 percent confidence intervals. *Source:* Prowess.

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Figure A4: Domestic Firms Contract after QR Removal: Robustness to Conditioning on Family Firms
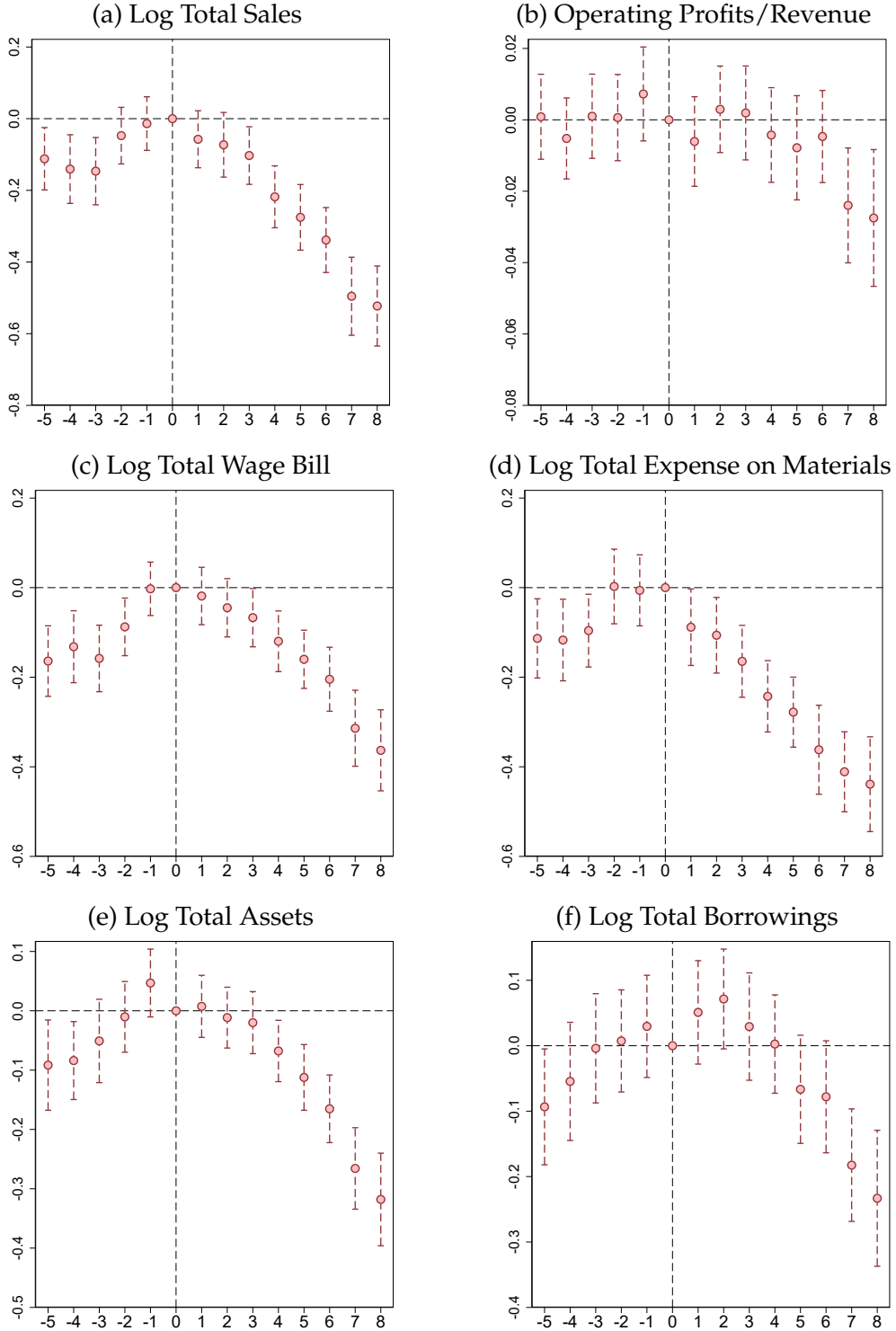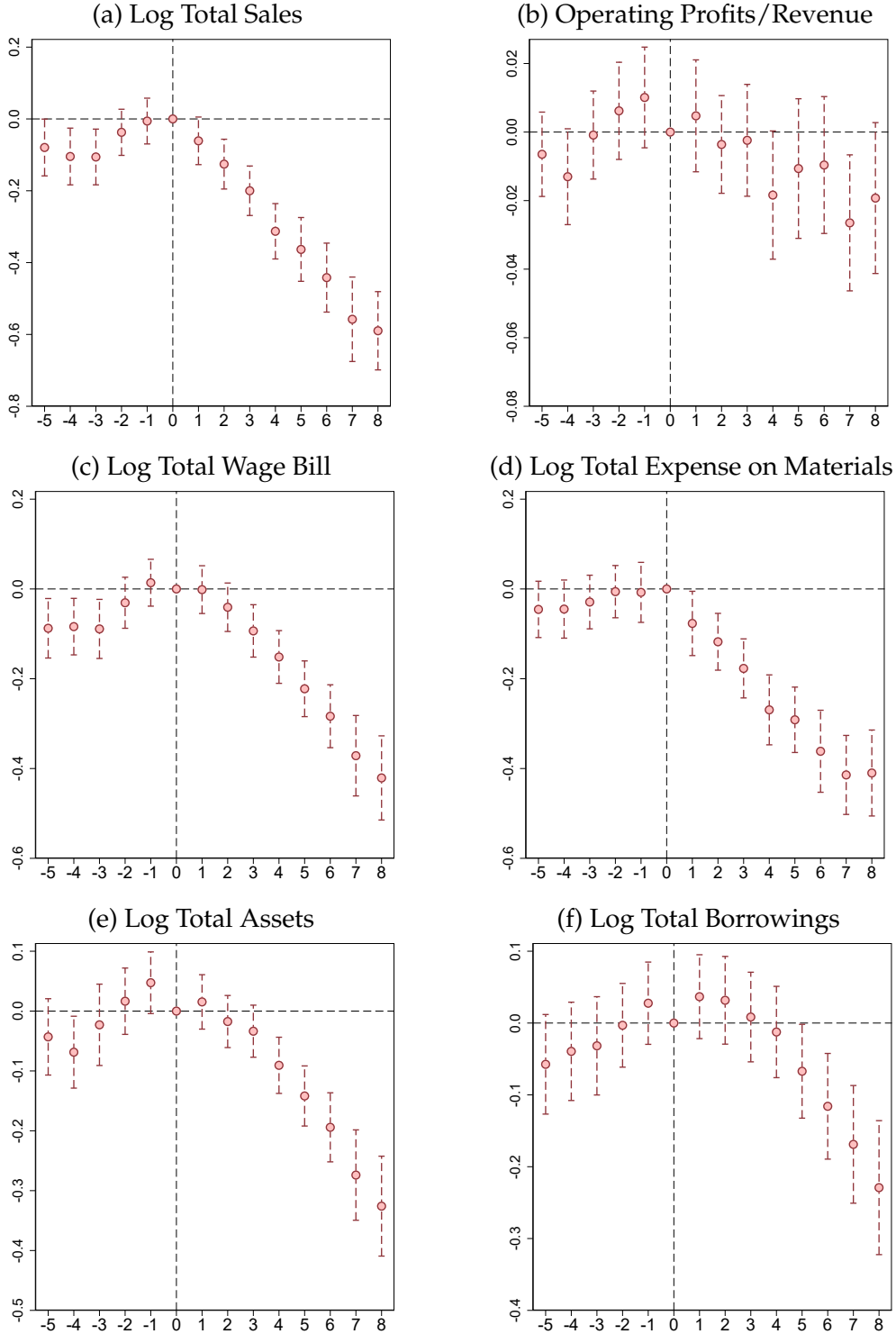

(a) Log Total Sales

(b) Operating Profits/Revenue

(c) Log Total Wage Bill

(d) Log Total Expense on Materials

(e) Log Total Assets

(f) Log Total Borrowings

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2) for firms which were identified as family firms beginning from pre-policy period, where the dependent variables are: log total sales (panel (a)), operating profits/revenue (panel (b)), log total compensation (panel (c)), log total expense on raw materials (panel (d)), log total assets (panel (e)), and log total borrowings (panel (f)). A firm is identified as treated in a year if QRs were removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the *x*-axis denoting years relative to the event. All regressions include firm and three-digit industry × year fixed effects. Standard errors are clustered at the three-digit industry × year level. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$.

Figure A5: Domestic Firms Contract after QR Removal: Robustness to Conditioning on Surviving Firms

(a) Log Total Sales

(b) Operating Profits/Revenue

(c) Log Total Wage Bill

(d) Log Total Expense on Materials

(e) Log Total Assets

(f) Log Total Borrowings

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2) for firms that survive after the shock (i.e., firms that do not exit after the shock). The dependent variables are: log total sales (panel (a)), operating profits/revenue (panel (b)), log total compensation (panel (c)), log total expense on raw materials (panel (d)), log total assets (panel (e)), and log total borrowings (panel (f)). A firm is identified as treated in a year if QRs were removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the x-axis denoting years relative to the event. All regressions include firm and three-digit industry $\times$ year fixed effects. Standard errors are clustered at the three-digit industry $\times$ year level. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$.

Figure A6: Domestic Firms Contract After QR Removal: Robustness to Alternate Treatment Assignment Based on All Products Contributing $\geq$ 10% of Firm Revenue

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2), where the dependent variables are: log total sales (panel (a)), operating profits/revenue (panel (b)), log total compensation (panel (c)), log total expense on raw materials (panel (d)), log total assets (panel (e)), and log total borrowings (panel (f)). A firm is identified as treated in a year if QRs were removed on any product that accounts for at least 10% of the firm's total production value. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the *x*-axis denoting years relative to the event. All regressions include firm and three-digit industry $\times$ year fixed effects. Standard errors are clustered at the three-digit industry $\times$ year level. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$.
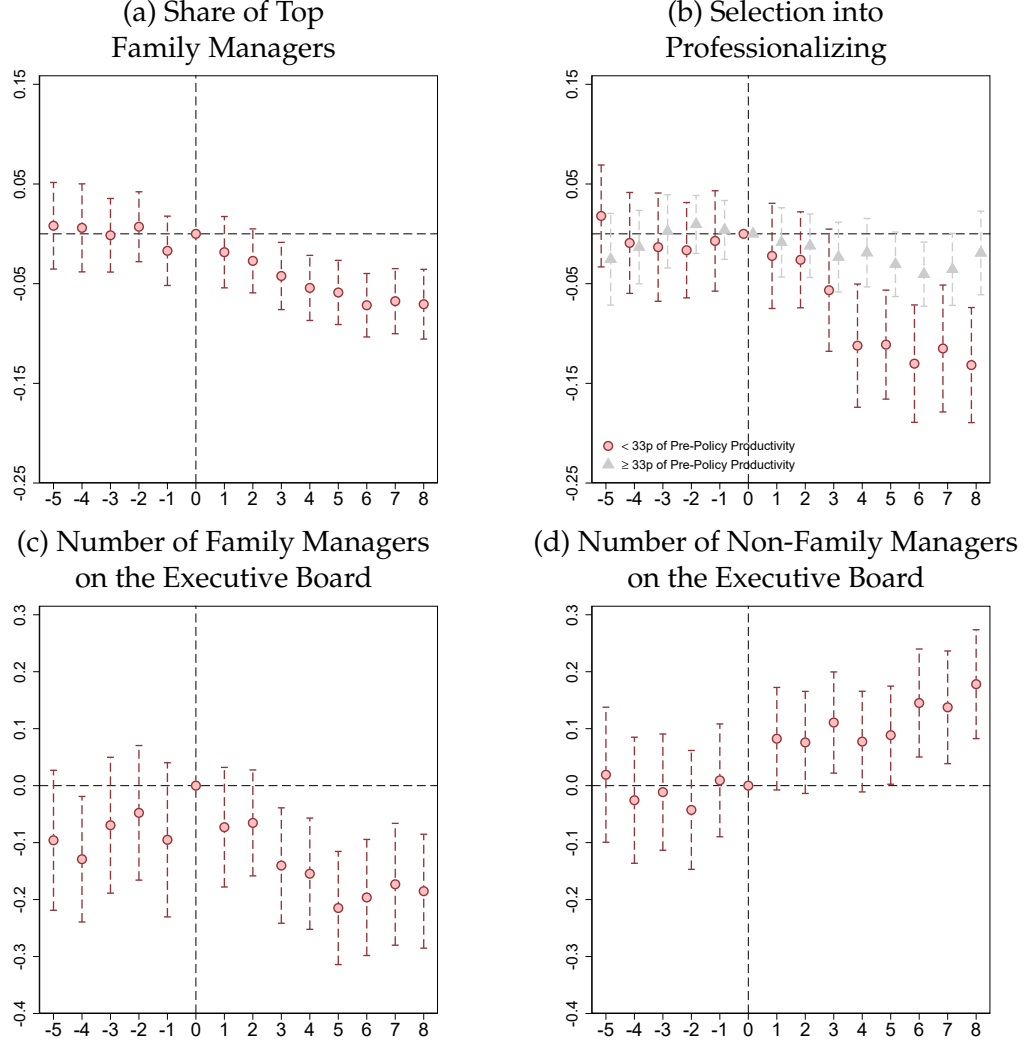
A.2.4 Firm Level Event Studies: Managerial Indicators

Table A4: Firms Reduce Family Members on the Executive Board of Directors after the QR Shock

|	(1) Share of Top Family Managers	(2) Number of Family Managers	(3) Number of Non-Family Managers
5 years before event	0.0045	-0.098	0.018
	(0.023)	(0.066)	(0.064)
4 years before event	-0.000048	-0.11*	-0.010
	(0.024)	(0.056)	(0.060)
3 years before event	-0.0037	-0.089	0.0097
	(0.020)	(0.059)	(0.058)
2 years before event	0.0017	-0.046	-0.021
	(0.019)	(0.056)	(0.058)
1 year before event	-0.015	-0.100	0.022
	(0.018)	(0.063)	(0.058)
1 year after event	-0.012	-0.097*	0.083*
	(0.019)	(0.053)	(0.051)
2 years after event	-0.030*	-0.076	0.099**
	(0.017)	(0.052)	(0.048)
3 years after event	-0.044**	-0.16***	0.12**
	(0.018)	(0.053)	(0.047)
4 years after event	-0.062***	-0.18***	0.090*
	(0.018)	(0.054)	(0.049)
5 years after event	-0.062***	-0.25***	0.089*
	(0.018)	(0.052)	(0.048)
6 years after event	-0.072***	-0.20***	0.15***
	(0.017)	(0.054)	(0.050)
7 years after event	-0.068***	-0.18***	0.15***
	(0.017)	(0.053)	(0.054)
8 years after event	-0.076***	-0.19***	0.19***
	(0.018)	(0.052)	(0.051)
Firm FE	✓	✓	✓
Industry × Year FE	✓	✓	✓
District × Year FE	✓	✓	✓
Observations	23445	23800	23800
R^2	0.83	0.77	0.75

Notes: The table shows the results of running the event study specification (2). The dependent variables are: the share of family members on the executive board of directors of a firm (i.e., top management positions like CEO, CFO, MD, etc.) in (column (1)), the number of family members on the executive board of directors in (column (2)) and the number of non-family professionals on the executive board of directors in (column (3)). A firm is identified as treated in a year if QRs were removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001. Clustering of standard errors is at the three-digit industry × year. Standard errors are in parenthesis. ***, ** and * denote statistical significance at the 1%, 5%, and 10% levels, respectively. *Source:* CMIE Prowess$_{dx}$.

Figure A7: Firms Reduce Family Members on the Executive Board of Directors after the QR Shock: Robustness to Two-way Fixed Effects Estimator

(a) Share of Top
Family Managers

(b) Selection into
Professionalizing

○ < 33p of Pre-Policy Productivity
▲ ≥ 33p of Pre-Policy Productivity

(c) Number of Family Managers
on the Executive Board

(d) Number of Non-Family Managers
on the Executive Board

Notes: The figure plots the estimated θ_k event study coefficients from a regression of the form given in (2), estimated using the two-way fixed estimator. The dependent variables are: the share of family members on the executive board of directors of a firm (i.e., top management positions like CEO, CFO, MD, etc.) in panel (a), share of top family managers dropped from the board while transitioning toward professional management in panel (b), the number of family members on the executive board of directors in panel (c) and the number of non-family professionals on the executive board of directors in panel (d). In panel (b), we examine selection into professionalizing by comparing firms based on their pre-policy productivity. Firms are divided into two groups: those in the bottom tertile of pre-policy productivity and the remaining firms. A firm is identified as treated in a year if QRs are removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the *x*-axis labels denoting years relative to the event. All regressions include firm and three-digit industry × year fixed effects. Standard errors are clustered at the three-digit industry × year level. The vertical lines are the 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

Figure A8: Increase in Productivity as Firms Shed Family Members after QR Removal: Triple Specification

(a) Revenue Productivity

(b) Quantity Productivity

(c) Log Price
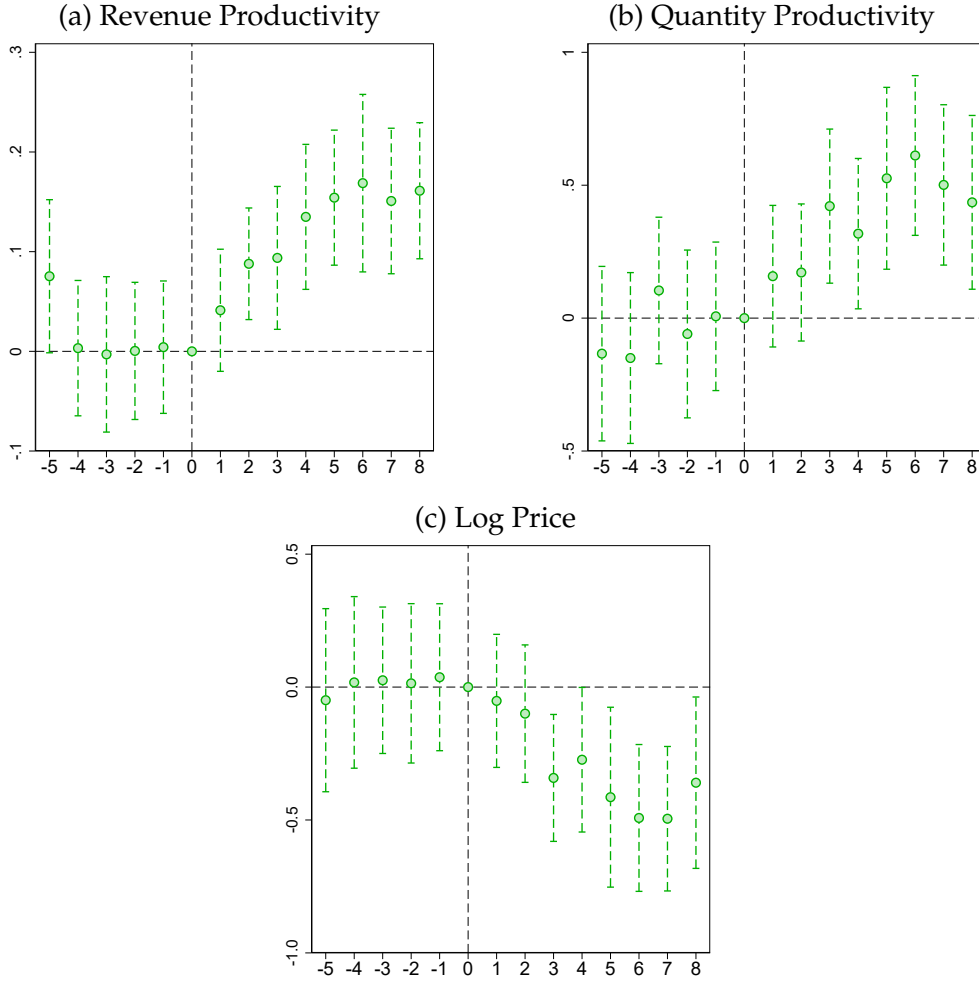
Notes: This figure presents the estimated θ_k event study coefficients from a regression specified in equation (2), with coefficients in Table A5. Event studies are conducted by interacting and taking differences between firms that professionalized their management after QR removal and those that did not. A firm is classified as treated in a year if QRs are removed on its highest-revenue product. A firm is considered to have professionalized if the share of family members on the executive board of directors declined in the post-policy period. The dependent variables are revenue productivity (TFPR) (panel (a)), quantity productivity (TFPQ) (panel (b)), and log price (panel (c)). TFPR and TFPQ are estimated using the method proposed in Petrin and Levinsohn (2012). Log price is defined as the ratio of a firm's total value of products produced and the total quantity of products produced. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the x-axis indicating years relative to the event. All regressions include firm and three-digit industry × year fixed effects. Standard errors are clustered at the three-digit industry × year level. The vertical lines represent 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

Table A5: Increase in Productivity as Firms Shed Family Members after QR Removal: Triple Specification

|	(1) Revenue Productivity	(2) Quantity Productivity	(3) Log Price
5 years before event	0.075*	-0.13	-0.049
	(0.039)	(0.17)	(0.18)
4 years before event	0.0032	-0.15	0.018
	(0.035)	(0.16)	(0.16)
3 years before event	-0.0030	0.10	0.026
	(0.040)	(0.14)	(0.14)
2 years before event	0.00044	-0.060	0.014
	(0.035)	(0.16)	(0.15)
1 year before event	0.0042	0.0068	0.037
	(0.034)	(0.14)	(0.14)
1 year after event	0.041	0.16	-0.052
	(0.031)	(0.14)	(0.13)
2 years after event	0.088***	0.17	-0.100
	(0.029)	(0.13)	(0.13)
3 years after event	0.094**	0.42***	-0.34***
	(0.037)	(0.15)	(0.12)
4 years after event	0.13***	0.32**	-0.27**
	(0.037)	(0.14)	(0.14)
5 years after event	0.15***	0.53***	-0.41**
	(0.035)	(0.17)	(0.17)
6 years after event	0.17***	0.61***	-0.49***
	(0.045)	(0.15)	(0.14)
7 years after event	0.15***	0.50***	-0.50***
	(0.037)	(0.15)	(0.14)
8 years after event	0.16***	0.44***	-0.36**
	(0.035)	(0.17)	(0.16)
Firm FE	✓	✓	✓
Year FE	✓	✓	✓
Observations	25473	25473	25473
R^2	0.87	0.86	0.86

Notes: The table presents the estimated θ_k event study coefficients from a regression specified in equation (2), with corresponding estimates plotted in A8. Event studies are conducted by interacting and taking differences between firms that professionalized their management after QR removal and those that did not. A firm is classified as treated in a year if QRs are removed on its highest-revenue product. A firm is considered to have professionalized if the share of family members on the executive board of directors declined in the post-policy period. The dependent variables are revenue productivity (TFPR) (column (1)), quantity productivity (TFPQ) (column (2)), and log price (column (3)). TFPR and TFPQ are estimated using the method proposed in Petrin and Levinsohn (2012). Log price is defined as the ratio of a firm's total value of products produced and the total quantity of products produced. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001. All regressions include firm and year fixed effects. Clustering of standard errors is at the three-digit industry × year. Standard errors are in parenthesis. ***, ** and * denote statistical significance at the 1%, 5%, and 10% levels, respectively. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

Figure A9: Increase in Productivity as Firms Shed Family Members after QR Removal: Robustness to Two-Way Fixed Effects Estimator

(a) Revenue Productivity

(b) Quantity Productivity

(c) Log Price
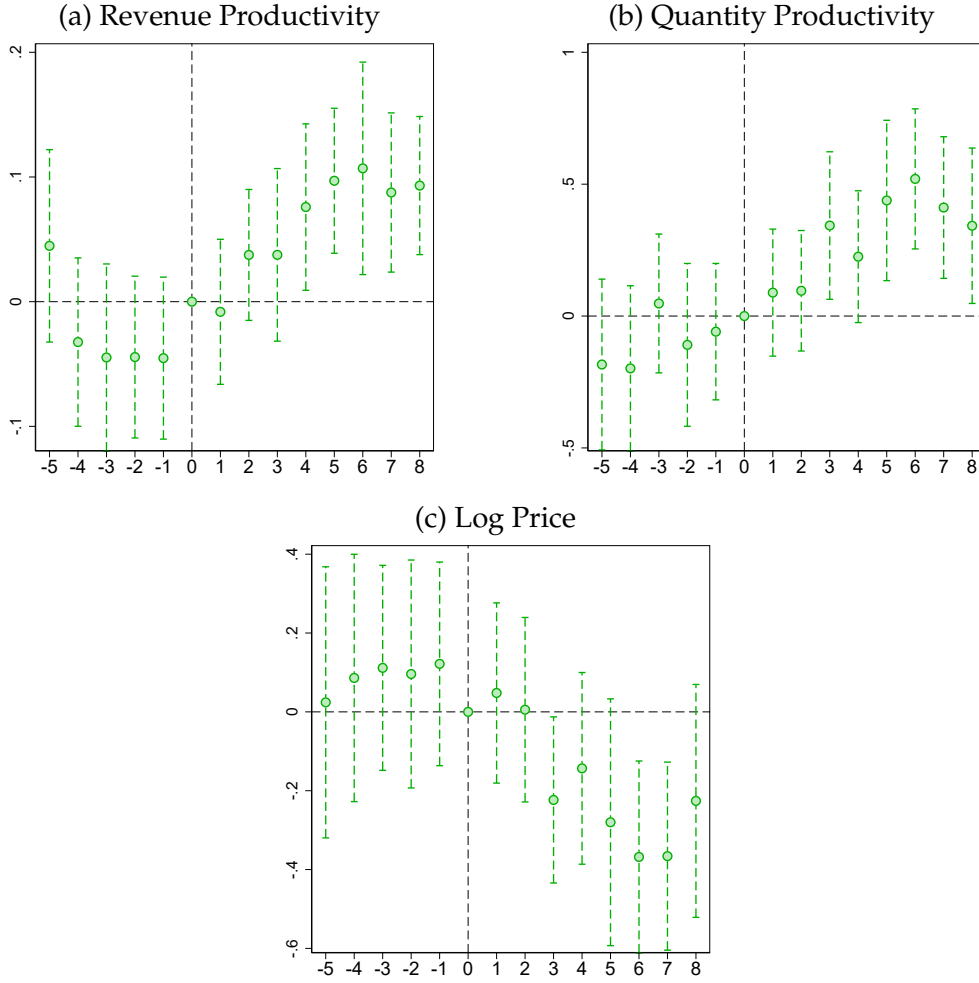
Notes: This figure presents the estimated θ_k event study coefficients from a regression specified in equation (2), estimated using two-way fixed estimator. Event studies are conducted separately for firms that professionalized their management after QR removal (in red) and those that did not (in grey). A firm is classified as treated in a year if QRs are removed on its highest-revenue product. A firm is considered to have professionalized if the share of family members on the executive board of directors declined in the post-policy period. The dependent variables are revenue productivity (TFPR) (panel (a)), quantity productivity (TFPQ) (panel (b)), and log price (panel (c)). TFPR and TFPQ are estimated using the method proposed in Petrin and Levinsohn (2012). Log price is defined as the ratio of a firm's total value of products produced and the total quantity of products produced. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the x-axis indicating years relative to the event. All regressions include firm and three-digit industry × year fixed effects. Standard errors are clustered at the three-digit industry × year level. The vertical lines represent 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.

Figure A10: Increase in Quantity Productivity as Firms Shed Family Members after QR Removal: Robustness
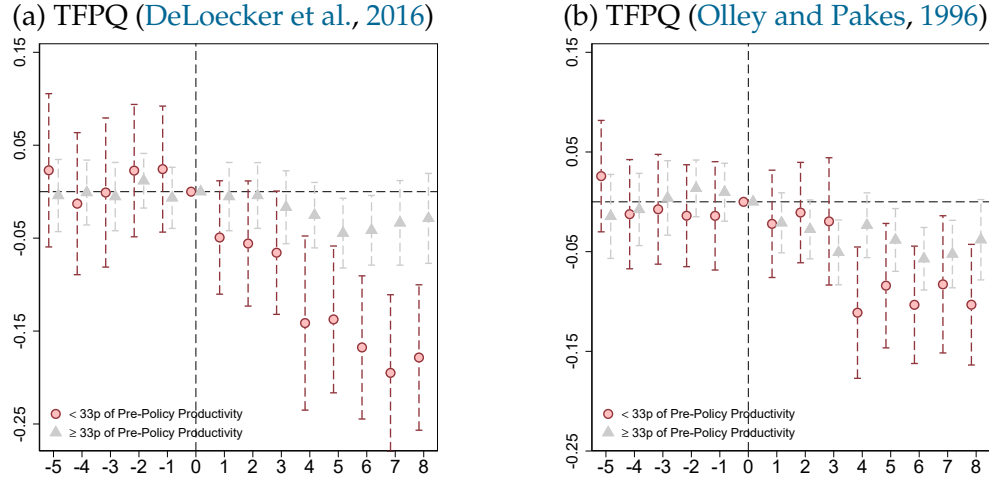
(a) TFPQ (DeLoecker et al., 2016)

(b) TFPQ (Olley and Pakes, 1996)

Notes: This figure presents the estimated θ_k event study coefficients from a regression specified in equation (2). The dependent variable in both panels is the share of top family managers dropped from the board while transitioning toward professional management. Here, we examine selection into professionalizing by comparing firms based on their pre-policy productivity levels. Firms are divided into two groups: those in the bottom tertile of pre-policy productivity and the remaining firms. Productivity (TFPQ) is estimated using the method proposed in DeLoecker et al. (2016) in panel (a), while Panel (b) employs the method from Olley and Pakes (1996). A firm is identified as treated in a year if QRs are removed on its highest-revenue product. θ_0, the coefficient for the year in which QRs were removed, is normalized to zero. The event is staggered from 1995 to 2001, with the x-axis indicating years relative to the event. All regressions include firm and three-digit industry $\times$ year fixed effects. Standard errors are clustered at three-digit industry $\times$ year level. The vertical lines represent 95 percent confidence intervals. *Source:* CMIE Prowess$_{dx}$ and the Ministry of Corporate Affairs, Government of India.