

NON-TARIFF BARRIERS AND BARGAINING IN GENERIC AND OFF-PATENT PHARMACEUTICALS*

Sharat Ganapati
Georgetown University
sg1390@georgetown.edu

Rebecca McKibbin
The University of Sydney
rebecca.mckibbin@sydney.edu.au

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Abstract

Pharmaceutical prices are widely dispersed across countries with comparable quality standards. We study two elements of this dispersion; non-tariff trade barriers and buyer bargaining power. Under monopoly, off-patent drug prices are 3-4 times higher in the United States. With 6 or more competitors, off-patent drug prices are similar across countries. Motivated by this, we use a bargaining model to examine two policy solutions to reduce drug prices. First, we remove barriers to trade in the form of a reciprocal approval arrangement, which could increase the number of competitors through increased market entry. Second, we explore the US government's unexploited purchasing power to negotiate drug prices. Regarding Medicaid, the first measure can reduce total expenditures by 8% and the second by 18%. There are little additional savings from doing both procedures in tandem.

Keywords: Law of One Price, Competition, Bargaining, Pharmaceuticals, Non-Tariff Barriers, Healthcare Economics, International Trade

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1 Introduction

Does the US government need to bargain over the prices of off-patent pharmaceutical prices? The US relies largely on competitive forces to set the price of pharmaceuticals that are no longer protected from competition by patents (primarily but not exclusively generic drugs). It is assumed that once patents expire, generic versions of the drugs will enter and drive prices to competitive levels. Recently, off-patent pharmaceutical prices have exhibited a series of extraordinary one off price increases (GAO, 2009). These broad-based and large price increases suggest that competition may not be a successful mechanism for achieving marginal cost pricing in all markets. In this paper we first document empirical evidence that the prices of US generic and off-patent-drugs in markets with few competitors are higher than in comparable countries. We then study two mechanisms behind this price variation. First, the role of competition between pharmaceutical suppliers that face barriers to market entry. Second, the role of downstream buyer market power and leverage.

In the United States, generic drugs account for 90% of prescriptions and 23.2% of expenditures in the \$324 billion prescription drug market (IQVIA, 2018). However, there is empirical evidence that off-patent drugs are not sold in perfectly competitive marketplaces at marginal cost. Since 2010, 20% of US generic or off-patent molecules have experienced a price increase of more than 100% (GAO, 2009). For example, the price of pyrimethamine, an anti-parasitic developed in the 1950’s, was infamously changed overnight from \$13 to \$750 per pill (Pollack, 2015). Although these drugs are no longer protected from competition by patents, many generic drug markets may not be large enough to attract the number of competitors needed to achieve marginal cost pricing (Berndt et al., 2017).¹ They may also be amenable to collusive behavior (Rowland, 2018).

Current policy is focused on encouraging more generic entry into the US market. For example, the Food & Drug Administration (FDA) 2019 Drug Competition Action Plan proposes that the FDA work with its international counterparts to harmonize the generic approval process (Gottlieb, 2018). This would allow suppliers to gain approval to sell in multiple countries essentially using one application process, thereby effectively removing a non-tariff trade barrier — a restriction on international trade that is not a tariff. Such policies are also advocated by the European Union in preliminary negotiations for the Transatlantic Trade and Investment Partnership (TTIP), seeking the harmonization of procedures to “entail significant cost savings” (European Commission, 2014). While this is a sensible policy direction, the US is unique in that it effectively relies solely on competition to achieve low generic drug prices. In contrast, many countries with comparable safety standards use a combination of competition and government purchasing power to attain low generic drug prices. Hence it makes sense to explore the effectiveness of both competition and bargaining based policies at reducing the prices of off-patent and generic drugs in these small markets.

Motivated by this observation, we compare final prices of pharmacy dispensed off-patent drugs across the US, Australia, Canada, New Zealand, and the United Kingdom.² We view this price as

¹In the US, Berndt et al. (2017) and Dave et al. (2017) show that 50% of such markets are monopolies or duopolies.

²These five developed nations were chosen primarily because there are no significant language barriers. For Canada

the welfare relevant price because it is the price paid by the consumer (either directly or indirectly through taxation).³ This comparison reveals that the final price of off-patent drugs in the US relative to each of these other countries declines as more suppliers enter the market.⁴ When there is only one American supplier, the price is on average 3-4 times higher.⁵ However, with six or more suppliers, retail prices are either similar or lower.⁶ We focus our analysis on prices paid by the US Medicaid program, however, results for Medicare Part D and private insurance are broadly similar. Our results suggest that while competition is an effective mechanism for lowering prices in markets with many suppliers, which tend to be large markets with many patients, bargaining may be necessary in markets with few suppliers, which tend to be smaller markets.

We study the mechanism behind this price variation using a structural model with Nash bargaining and a market entry game.⁷ The generic and off-patent drug market is modeled as a set of bilateral transactions between many suppliers and a representative buyer. The buyer and sellers split the surplus of the transaction based on bargaining weights and upstream seller competition. This setup nests bilateral Nash bargaining within a model of oligopoly, with sellers paying a fixed cost to enter a market, in the spirit of [Bresnahan and Reiss \(1991\)](#).⁸ Multiple buyers are parsimoniously captured through reduced bargaining leverage, a sufficient statistic that combines both bargaining power and the buyer's outside option. We focus on the buyer and consider the total cost paid, holding fixed the prescribing behavior of pharmacists and doctors. This model is identified using cross-country variation in prices, generic and off-patent drug suppliers, and market sizes.

Estimation using public data shows that the US currently applies substantially weaker bargaining leverage than any of the comparison countries. Furthermore, it is more expensive to enter the US market than a comparable country such as Australia or the UK, with the implied entry cost being \$6-14 million higher per year. The estimated costs cover not only regulatory costs, but the implied costs of collusion and bi-lateral downstream payments ([Loftus et al., 2016](#); [Hancock and](#)

we focus on the two largest English speaking provinces: Ontario and British Columbia. For the United Kingdom, we focus solely on England.

³The welfare implications of lowering the price are mostly distributional in this market. Since consumer demand is inelastic to the price (as a result of insurance), changes in the price represent a transfer between the consumer and the producer. Governments are the primary payer in drug markets and hence it is a policy objective to capture a greater share of the surplus for the consumer.

⁴While the US has higher prices for drugs with limited competition, it often has the lowest prices for the most widely disseminated drugs. For example, Gabapentin was first marketed in 1993 for partial epileptic seizures and went generic in 2004. By 2016, this drug had over 20 FDA-approved suppliers, with over 4 billion doses prescribed through Medicaid and Medicare Part D. Medicaid and Medicare Part D pharmacy reimbursements averaged \$0.17 and \$0.18 per dose respectively. By comparison, the Australian government reimbursed the equivalent of \$0.21 cents per dose.

⁵These results are based on generic and off-patent drugs sold in either tablet or capsule forms and consider per-dose prices, net of pharmacy dispensing fees.

⁶Research using data from the 1990s has shown that price reductions from a marginal entrant occur until there are eight or more suppliers of the drug ([Reiffen and Ward, 2005](#)).

⁷We use a Nash-in-oligopoly setup that allows us to jointly identify the outside option and the bargaining parameters separately from the market competition. This is adapted from the Nash-in-Nash games described in [Horn and Wolinsky \(1988\)](#). See [Chipty and Snyder \(1999\)](#) for an implementation in cable TV markets and [Ho and Lee \(2017\)](#) regarding hospital-insurance networks. [Collard-Wexler et al. \(2014\)](#) provide micro-foundations for such bargaining solutions.

⁸We flexibly model competition, allowing for possibilities ranging from pure collusion to homogenous product Bertrand competition.

Lupkin, 2019). We conduct four counterfactuals as they appertain to Medicaid: a reciprocity policy where once a generic drug is approved in one country it is approved in all; the US government directly bargaining prices; both; and the harmonization of entry costs. A reciprocity policy would reduce total drug expenditures by between 3%-8%. If the US bargained to the same degree as any comparison country, prices would be reduced by between 10%-18%. Implementing both policies has virtually the same effect as just the bargaining policy. If entry costs were brought in alignment with Australia or the UK, prices would be reduced by 6%-16%. These results indicate that that downstream buyer leverage and entry costs are equally important in explaining the price variation we observe.

This paper is the first to examine the relative roles of competition and bargaining in generic drug markets using a structural framework. Danzon and Chao (2000) show the existence of price dispersion across countries and estimate the effect of competition using a reduced form approach. We build on the spirit of Danzon and Chao (2000) but directly consider the role played by competition and bargaining in generic pharmaceutical pricing and entry.⁹ Other prior research on generic drugs considered the effect of exogenous entry on prices (Reiffen and Ward, 2005; Berndt et al., 2017; Grabowski and Vernon, 1992),¹⁰ or how competition-based policies could increase competition and hence lower prices (Berndt et al., 2017; Gupta et al., 2018; Bollyky and Kesselheim, 2017; Berndt et al., 2018). This latter literature has been reduced form and descriptive. Another reduced form literature studies the substitution patterns between generic entrants and the original branded drug under various types of price regulation (Kanavos et al. (2008); Puig-Junoy (2010); Kanavos et al. (2013)). Structural policy work has focused on studying the effect of parallel imports (goods that are imported by non-authorized resellers) of patented/protected pharmaceuticals (Malueg and Schwartz, 1994; Ganslandt and Maskus, 2004; Grossman and Lai, 2008; Dubois and Saethre, 2018). This literature focuses on price discrimination and cross-border effects.¹¹

This paper is also the first to quantify non-tariff trade barriers (in this case created by regulations that restrict entry in foreign markets) in pharmaceutical markets and structurally model the trade off between non-tariff barriers and competitive concerns with both a pricing and market entry margin. An extensive literature studies both deviations from the law of one price and non-tariff barriers. The lack of a single global price for goods is explained through many channels, such as incomplete exchange rate pass-through, local costs, pricing to market, market power, information frictions and

⁹Our results are not directly comparable with Danzon and Chao (2000) and have a different interpretation. Danzon and Chao (2000) compare average drug prices across countries and hence create a price index that takes into account the basket of drugs consumed in each country. With a different objective, in this paper we compare the price of the same drug across countries and then average this price difference across drugs that have the same number of suppliers.

¹⁰Reiffen and Ward (2005) structurally consider decisions of newly generic drugs versus the incumbents following the expiration of patents, taking demand as given. We exploit variation across both time and countries, considering a much wider set of drugs (including those that have been generic for long periods of time), consider bargaining jointly with competition, and directly consider counterfactuals relevant to current policy.

¹¹A related literature focuses on price regulation in patent protected markets. These papers consider both the effects on innovation and on cross-border price spillovers (Vernon, 2005; Kyle, 2007; Brekke et al., 2015). These negative effects on innovation and price-spillovers are less of an issue for generic drugs because generic drugs are copies of existing technology and do not require extensive and expensive drug trials, allowing for straightforward identification.

sticky prices.¹² Within health care, [Cooper et al. \(2018\)](#) and [Craig et al. \(2018\)](#) show that monopoly power and bargaining have significant relationships with procedure and procurement prices, but do not model endogenous market entry. Additionally, a largely separate literature considers the role of non-tariff barriers in hampering trade, often showing a quantitatively large aggregate effect.¹³ There are notable exceptions to these broad surveys, [Goldberg and Verboven \(2001\)](#) and [Goldberg and Hellerstein \(2008\)](#) consider the pricing decision of firms through a demand-side perspective.

2 Institutional Background

Competition in off-patent drug markets is shaped by the nature of manufacturing regulation, the insurance system, and pharmacy regulation. This section describes how these elements affect prices at each stage of the supply chain in each country as it applies to the drugs considered in this paper: patient administered drugs purchased in pharmacies. The supply chain in all five countries include suppliers or importers obtaining approval from a regulatory body to market their products. Manufacturers sell to wholesalers, who sell to pharmacies who retail the products to consumers.

The retail price paid depends on the consumer’s health insurance policy. In all five countries, the major insurers offer a set price schedule for generic drugs and pharmacies must take it or leave it. Reimbursements take two parts, the cost of the pharmaceutical and the dispensing fee. We focus on the cost of the pharmaceutical and net out the dispensing fee paid to the pharmacy. In the US, there are many insurers (or their subcontracted pharmacy benefit managers) arranging their own price agreements with pharmacies and suppliers. In Australia, the UK, and New Zealand, a single insurer — the government — uses its purchasing power to set pricing rules for drugs. Canada sits in the middle with a public-private system similar to the US, except that the government actively seeks to minimize drug prices for the public health plan. The covered generic formulary is rarely bargained over in any of the countries. Generic drugs are almost never excluded from public formularies for prescription coverage (unlike with on-patent drugs). Pharmacy regulation plays an important role in the competitiveness and size of the generic market in all five countries, and hence the number of suppliers, because it affects whether or not pharmacists can substitute branded drugs for generics. In 2013, generics were 84% of the market in the US, 83% in the UK, 77% in New Zealand, 70% in Canada in terms of volume ([OECD, 2015](#)).

2.1 Manufacturing and Entry Regulations

Before a drug can be sold to consumers, the supplier must apply for authorization from the local regulatory authority. These regulators monitor quality both in terms of the safety and efficacy of

¹²See [Isard \(1977\)](#), [Goldberg et al. \(1997\)](#), and [Burstein and Gopinath \(2014\)](#) for early work and two literature reviews. Selected recent examples include [Amiti et al. \(2014\)](#); [Auer and Schoenle \(2016\)](#); [Steinwender \(2018\)](#).

¹³For a description, see [Grieco \(1990\)](#). [Olarreaga et al. \(2006\)](#); [Egger et al. \(2015\)](#) perform quantification exercises on non-tariff barriers.

the drugs and in terms of manufacturing quality. The approval requirements depend on whether or not the drug is classified as an innovative drug or a generic. Innovative drugs require extensive (and costly) testing with clinical trials to demonstrate their safety and efficacy profile. Generic approval requires demonstration that the generic is bioequivalent to an innovative drug.¹⁴ Bioequivalence typically requires a clinical trial that shows the generic delivers the same amount of active ingredients into a patient’s bloodstream in the same amount of time as the original drug. The generic must also be identical in terms of dosage form, strength, route of administration, and intended use. All five countries follow the International Conference on Harmonization (ICH) Good Clinical Practice Guidelines in assessing bioequivalence.

In addition to demonstrating therapeutic equivalence, suppliers must also demonstrate compliance with production quality standards. All five countries have Good Manufacturing Practice (GMP) guidelines that comply with the ICH. These regulations ensure that products are properly produced, packaged, and safe (FDA, 2018). To ensure compliance, regulatory authorities conduct plant inspections. Such inspections are costly and complicated by the globalization of manufacturing.¹⁵ A large proportion of the US drug supply is manufactured overseas, with 40% of finished drugs and 80% of active ingredients produced abroad (GAO, 2016). To facilitate inspections abroad, the FDA has several foreign offices. Due to resource constraints, plants are selected for inspection using a risk-based approach. As of 2016, as many as a 1,000 foreign plants (33%) may have never been inspected at all (GAO, 2016). A reciprocity agreement could reduce the cost and/or improve the effectiveness of manufacturing quality regulation for the US. The EU (including the UK) Australia, Canada, and New Zealand already have reciprocal inspection agreements.¹⁶

2.2 Insurance & Pricing

The US has three major insurance regimes covering prescription drugs sold in pharmacies: private insurance, Medicare Part D, and Medicaid.¹⁷ However, all three frequently outsource the management of their prescription drug plans to pharmacy benefit managers (PBMs). PBMs have substantial market power with three companies controlling 66% of the market (Sood et al., 2017). PBMs act

¹⁴The innovative drug must already be approved in the applicable country.

¹⁵Many approved suppliers do not manufacture drugs themselves, they outsource production to contract manufacturers. For example Gilead outsources the majority of production to third parties (Miller, 2017). Our empirical analysis will focus on the approved supplier of the drug, not the physical manufacturer themselves, who simply need to follow safety regulations for all drugs they manufacture. We leave the relationship between the approved supplier and the contract manufacturer for future analysis. Even though the government-approved suppliers may not be the original manufacturer themselves, the literature simply terms these suppliers the “manufacturer.” We prefer the term “supplier”, but will refer to “manufacturers” in reference to the prior literature.

¹⁶Canada does not recognize New Zealand. In 2017 the FDA began a reciprocal inspection agreement with the European Union, which will be fully implemented in 2019 (EMA, 2018).

¹⁷Medicare is the US public insurance program for the elderly and long-term disabled. Medicaid is the US government insurance program for the poor and short-term disabled. In 2016, private insurance accounted for 43% of retail prescription drug spending, while Medicare and Medicaid accounted for 29% and 10%, respectively (the remainder is out of pocket payments (14%) and other US government insurance programs (4%)). Data on Medicaid allows us to net out the dispensing fee. We are unable to do this with Medicare and private insurance data, so we use these datasets as a robustness check.

as intermediaries in contracts between health plans, pharmacies, and suppliers, as well as set the formularies and network of covered pharmacies. For drugs with a single supplier, PBMs negotiate directly with suppliers for favorable placement on formularies (when there are drugs in the same therapeutic class that are close substitutes) in return for rebates, some or all of which they keep as profit. When there are multiple suppliers, PBMs provide a Maximum Allowable Cost (MAC) list to pharmacies, which states how much they will pay for the drugs. While MAC lists for private insurers and Part D are between the insurer and the PBM, Medicaid reimbursement rates are subject to statutory rules. These vary by state, but broadly, they compute potential prices using four different methods and set the reimbursement at the level of the lowest one.¹⁸ In addition, a federally mandated manufacturer rebate is set by statute.¹⁹

Australia, the UK, and New Zealand have government-operated universal health plans. Canada has a mixed public private system. Similar to the US, working age people have private insurance and each province has its own public drug plan, which cover disadvantaged segments of the population. We focus on the two largest English-speaking provinces: Ontario and British Columbia. Ontario covers the elderly, poor, and disabled, while British Columbia covers everyone but uses an income test to determine the subsidy rate of the beneficiary. Unlike the US, in all of these countries, the government uses its purchasing power to obtain lower drug prices. For on-patent drugs, governments directly bargaining with the supplier. In the case of generics, each country stipulates how it will pay for generics and the suppliers can take it or leave it.

The method used for setting generic drug prices differs across countries. Australia and the UK both use a reference price system, based on reported supplier prices. Suppliers report their sales prices and the health plan reimburses at the average. In Australia, prices are not revised upwards without approval, whereas in the UK, the price fluctuates with the market. Australia also has a mandatory price reduction of 16% when the first generic enters the market. New Zealand uses a competitive tendering system to obtain low generic drug prices. The supplier of the winning tender has their product exclusively eligible for reimbursement by the national health plan. All Canadian provinces except Quebec have agreed to set the price of generics entering the market from 2014 onwards using a tiered pricing system. The reimbursement rate is set at 85% of the price of branded drug when there is one generic, 50% with two generics and 25% when there are three or more. For generics introduced before 2014 the provincial reimbursement rates are used. Ontario sets the reimbursement rate at 25% while British Columbia sets it at 20%.

¹⁸The first is the acquisition cost plus a dispensing fee. States generally estimate the acquisition costs as either a discount on the average wholesale price (AWP) or a markup on the wholesale acquisition cost (WAC). Both the AWP and WAC are list prices and are not based on actual sales data. The second are usual and customary pharmacies charges to the public. Third is the federal upper limit (FUL), which is calculated by the CMS as 175% of the average manufacturer price (AMP). The AMP is based on data reported to the CMS quarterly for the purpose of computing rebates. Fourth is the State MAC price.

¹⁹During our study period this rebate was the greater of 15.1% of the AMP or the difference between the AMP and best price per unit for innovative drugs and 11% of AMP for generic drugs.

2.3 Pharmacies & Wholesalers

Pharmacies and wholesalers are not explicitly included in our model. However, these players affect the final price of drugs as they form part of the supply chain and receive a mark-up over the manufacturing price. Although the the wholesale market is very concentrated in all five countries — Australia, the US, the UK (Europe), and Canada have three dominant wholesalers, while NZ has just two wholesalers — wholesaler ability to increase prices is likely restricted by the market power of the insurers, particularly in countries with national health insurance plans, and do not appear to vary across drugs. A similar concern affects pharmacies.²⁰

3 Conceptual Framework

In this section we present a framework that links the institutional environment with key elements of the structural model as well as the empirical analysis and the available data. Our measure of welfare in this paper is changes in the price of off-patent drugs. Off-patent drugs are molecules that are no longer protected by patents. These drugs/molecules can be the original brand name or a generic copy. The welfare implications of price changes to off-patent drugs are primarily distributional. We consider this market to have a surplus that can be captured by the different actors in the market: consumers, insurers and their agents, governments, pharmacies, wholesalers and manufacturers. Our objective is to determine how consumers can capture a greater share of the surplus from suppliers and other upstream players. We view this is a policy relevant objective because in this market the government is ultimately the largest payer and its objective should be to minimize its expenditure.

Markets are defined in the analysis as individual molecules. Suppliers choose whether or not to produce a version of each of the molecules, which we refer to as products. Since the market is at the molecule level, all products are identical. Given that the product is homogenous in the market we assume that firms take prices as given when making the choice of whether or not to enter the market. We utilize a static entry model in the spirit of [Bresnahan and Reiss \(1991\)](#) and [Scott Morton \(1999\)](#). This model is chosen because we consider a stable market for molecules that are 30-40 years old. Thus a static model represents the relationships in the data well.

In this framework, suppliers of molecules make a decision to enter the market in each destination f and provide a product d if the profit of doing so ($\pi_{f,d}$) is greater than the fixed cost of entry F_f . We assume that there is a fixed cost of entry into each destination (F_f) that is (a) destination specific and (b) independent across destinations.²¹ Hence the a supplier s enters a destination d if and only

²⁰After directly accounting for pharmacy costs, we take these setups as given in estimation and do not change their structure in computing counterfactuals.

²¹The second part of this assumption is less restrictive than it initially seems. First, we consider marginal generic entrants - as opposed to the original innovator. Second, many of the implied fixed costs are country-specific, such as the regulatory lobbying, bilateral payments, and/or campaign contributions. Third, physical manufacturers are often “contract” manufacturer, which are separate from the official government-approved supplier. These contract manufacturers receive a specification and produce final packaged products for the government-approved supplier.

if:

$$\pi_{f,d}(s; S) > F_f.$$

Where $\pi_{f,d}(s; S)$ is the profit of the marginal s^{th} supplier over the set of S suppliers.

The profitability ($\pi_{f,d}$) of entering each market is the net present value of revenue less variable costs. This is the product of the price the supplier receives, which is a destination specific mark-up $\mu_{f,d}$ over marginal cost $c_{f,d}$, multiplied by the quantity sold $q_{f,d}$. Both the size of the mark-up and the quantity sold are a function of the number and identity of competitors in the market:

$$\pi_{f,d}(s; S) = \mu_{f,d}(s; S) \times c_{f,d} \times q_{f,d}(s; S).$$

We assume that the marginal cost within a market in a destination is constant across suppliers ($c_{f,d,s} = c_{f,d}$). This means that the marginal cost of producing a product is the same across products in a market.²²

Next we consider the welfare relevant price. As described in Section 2, there are many actors that are involved in setting the price of a product along the supply chain. These actors include but not limited to wholesalers, benefit managers, pharmacies, consumers, insurers and governments. Given our welfare objective, the most appropriate choice of outcome price is the retail price of each molecule (market) in each destination averaged across products $p_{f,d}$. The retail price p of a product comprises an amount paid by consumers (a copay δ_{copay}) and the amount the government contributes (δ_{gov}): $p = \delta_{gov} + \delta_{copay}$. The retail price $p_{s,f,d}$ of a product d in destination f can be represented as as a series of mark-ups ($\mu_{s,f,d}$) over the marginal cost of producing the product ($c_{s,f,d} = c_{f,d}$).

$$p_{s,f,d} = \mu_{s,f,d}^{pharmacy} \times \mu_{s,f,d}^{PBM} \times \mu_{s,f,d}^{wholesaler} \times \mu_{s,f,d}^{factory} \times c_{f,d},$$

We choose to use the entire retail price rather than just the government contribution as the government may subsidize markets for redistribution purposes. We take the average of the retail price across products because we view changes in the average price as best capturing what the government ultimately pays. Retail prices differ across pharmacies as individual consumers select the pharmacy to fill their prescription. There is generally no restriction on pharmacy choice in the non-US countries. In the US there is the possibility to exclude a pharmacy from a health plan. We take the view that no individual product would be large enough to affect an insurance plan's decision to include a pharmacy in its network.²³

These myriad contract manufacturers simply have to pass safety related inspection measures and are not responsible for the vast majority of entry costs, which are borne by the government approved supplier (Miller, 2017). Lastly, a survey of foreign factories for a sample of drugs shows that there are a large mass of potential suppliers that do not appear in our sample. For example there are only 10 FDA approved US suppliers for acyclovir, but at least 148 brands approved for sale in some country (MedIndia, 2019).

²²At this stage we flexible allow differences in the identity of the competitor s , thus branded suppliers and generic suppliers may have differences in both market shares and markups.

²³Moreover, most generic drugs are sold by the big four retail chain pharmacies.

The price we measure is therefore:

$$p_{f,d} = \sum_{s \in S} \text{share}(s) \times p_{s,f,d}.$$

This market weighted average share allows for differential pricing (and markup) between different suppliers. In some markets, there is substantially different pricing and market shares between the original branded entrant and other generic entrants (Kanavos et al., 2008). In our structural model, we directly model this behavior, recovering a welfare-relevant set of sufficient statistics, without directly modeling how market shares are determined.

In our framework the final buyer does not care about the exact division of surplus amongst firms along the supply chain, they simply care about the sales-weighted average markup μ :

$$p_{f,d} = \mu_{f,d}(S) \times c_{f,d}, \tag{1}$$

This mark-up is a function of the number of suppliers of the product in the market. Hence in our structural model we collapse the market into a problem between two agents: a buyer and a seller. The seller represents both the suppliers and all intermediaries in the supply chain. In our preliminary empirical analysis we estimate variants of Equation 1.

4 Data

The empirical analysis compares prices, competition structure, and market size across countries for off-patent patient administered drugs.²⁴ This requires data on the prices, quantities of prescriptions, and the number of suppliers of each drug for all five countries. We define a drug market as all products with the same molecule-dose-form. This follows the definition of a therapeutic substitute used by the US Food and Drug Administration (FDA) in determining off-patent classification.

Data on molecule-dose-forms in the US are obtained from the drugs@FDA database. We include molecules administered as “capsule” or “tablet”. We only include markets where the original product is off-patent and hence generics can enter (regardless of whether they have). A molecule-dose-form is classified as off-patent if an Abbreviated New Drug Application (ANDA) has been approved or if there are no patents or exclusivities listed in the FDA orange book. We compute similar statistics at the molecule level. As it takes time for the first generic to be reviewed, and the first approved generic receives 180 days of exclusivity against additional generic entry, we exclude markets with the first FDA approval within 20 years.²⁵ The number of suppliers supplying each included US market reflects the number of approved ANDAs. Data on drugs for corresponding foreign markets

²⁴We mix the usage of the term generic and off-patent, even though not all off-patent drugs are the generic copy.

²⁵This is similar, but not identical to patent protection, a period that typically starts at discovery, but before clinical trials and FDA approval, a process that can take upwards of 10 years. The US Hatch-Waxman Act effectively allows for five years of exclusivity after the patent expiry, allowing for 25 total years of near exclusivity. Our window effectively omits drugs in this period.

are obtained from the relevant regulatory authorities. We merge each of these data sets together and keep the subset of drugs that are available in both the US and the other country. Further details are provided in Appendix A.

The retail price and quantity prescribed in each off-patent market in the US is obtained from the Medicaid State Utilization data, the Medicare Provider Utilization and Payment Data: Part D Prescriber Public Use File, and the National Average Drug Acquisition Cost data (NADAC).²⁶ The Medicaid data contains the retail price paid for drugs covered through Medicaid, which we average across states. The Part D data contains the total expenditure and quantities, which we use to compute the average price. Medicare Part D data is limited as it does not distinguish between molecule dosages. For Medicaid, we compute per-unit prices net of dispensing fees and manufacturer rebates. Data on dispensing fees and manufacturer rebates are unavailable for Part D or NADAC. The NADAC data is collected by the Centers for Medicare and Medicaid Services and reflects the price paid by retail pharmacies to acquire prescription pharmaceuticals. Data from Medicaid is the gold standard, it can be corrected for manufacturer rebates and pharmacy reimbursements and is available at the molecule-dose-form level. Part D and NADAC data is much more limited, and cannot be corrected for manufacturer rebates and pharmacy reimbursements. Additionally, Part D data is only publicly reported at the molecule level, and NADAC data often has missing observations at the molecule-dose-form level. As Medicaid only accounts for 10% of US prescription drug spending, we include data on Medicare and private providers (through NADAC), for better coverage.

Data on retail drug prices and ex-manufacturer prices in Australia, Ontario, British Columbia, the UK, and NZ are obtained from the national health plan administrative statistics. We convert all prices into US dollars using the average annual exchange rate for each calendar year. In general we recover the per-unit price of a drug, net of a fixed per-prescription pharmacy dispensing charge. For Australia we use the price paid by Pharmaceutical Benefits Scheme (PBS) to the pharmacy, available from Pharmaceutical Benefits Scheme Schedule of Benefits for 2007-2017. This price excludes the dispensing fee and the allowed pharmacy mark-up.²⁷ For Ontario, we use the Ministry of Health and Long-Term Care’s drug benefit prices (DBP) 2017. This price omits the fixed dispensing fee and variable patient copayment, which can only offset the dispensing fee. Data on British Columbia from British Columbia PharmaCare for 2014-2017. We use the maximum allowable price, the maximum amount the drug benefit will reimburse, net of dispensing fees.²⁸ Any difference between the retail price and this price is paid by the patient. For New Zealand, we obtain data on retail prices from PHARMAC. This price excludes the patient copay, a flat rate of \$NZ5, and fees paid to the

²⁶Drug prices in the US are difficult to measure. There are many different types of drug prices available, many of which are not true prices but rather list prices. We choose to use the retail price in this paper because this is the only observable price available in the US. It is also a price that is policy relevant, as it is the final price paid by the government programs Medicare and Medicaid. We also obtain data on ex-manufacturer prices, which is the price received by the supplier. Wholesale prices are not available.

²⁷Patient copays in Australia are variable and are capped at \$AU30.70-\$AU38.80 during this period. Generic drugs often have patient copays below the cap, as maximum co-pays are capped at the combined cost of the pharmaceutical, dispensing, and preparation fees. Reported prices are inclusive of this variable patient copay.

²⁸Which are currently capped at \$C10.

pharmacist for dispensing. To be consistent, we add the copay to the reported price. Prices for the UK are obtained from the NHS England Drug Tariff.²⁹ These prices are the amount of the NHS subsidy. It does not include patient out of pocket costs or professional fees paid to pharmacists for dispensing the products.

Table 1 shows the number of observations for each country and the years included in the analysis. Prices for all non-US markets are net of fixed per-prescription pharmacy dispensing fees. The key variable of interest is the relative price of each drug in the US compared with the same drug in each of the other countries. The mean and variance of the US price ratio with respect to each base country are shown in the fifth and sixth columns of Table 1. On average, prices are extremely similar, with US prices only slightly higher than foreign prices. For the Medicaid sample, we present both adjusted and unadjusted prices. Adjusted prices subtract out the average pharmacy dispensing fee as well as the statutory manufacturer rebate of 13% for non-innovator drugs. Such data for Medicare and NADAC are unavailable.

Table 2 highlights the variation in suppliers in the United States (for 2017 Medicaid data) for our sample. While drugs with just one supplier account for 1% of doses, they make up 10% of off-patent Medicaid spending. Drugs with five or less suppliers account 25% of doses, but for 50% of total spending. While the majority of doses sold are in competitive markets, likely priced near marginal cost, many drugs have a limited number of suppliers.

5 Empirical Evidence

In this section, we present an exploratory analysis of prices and competition across countries. We start off with a key fact to motivate the entire analysis: *in markets with few domestic suppliers, Americans pay substantially higher prices than in foreign markets.* We then conduct several empirical exercises with a view to understand the potential for competition policy to be an effective tool at lowering prices in these markets through additional entry. First we show two facts that suggest a link between fixed costs, market entry, and competition, (1) Markets with few patients have few generic entrants and (2) these markets with few US patients have the highest prices in the US relative to foreign countries. This indicates that there may be natural bounds on the number of competitors that the market can support, which would limit the effectiveness of a competition policy. We then continue with two facts that underpin our counterfactual and the estimated model. (3) There is significant variation in the number of pharmaceutical providers across countries. (4) Demand in Medicare and Medicaid Part D are highly inelastic, highlighting the possible need for bargaining power to control buyer costs. These findings all have broad support using three different US data sources: Medicaid, Medicare Part D, and NADAC.

²⁹We focus only on England and leave Northern Ireland, Wales, and Scotland for further analysis.

Table 1: Summary Statistics

(a) Medicaid - Molecule-Dose-Form Comparison

Comparison	Obs	Start Year	End Year	$\log(P_{US}/P_{Dest})$ Adjusted		$\log(P_{US}/P_{Dest})$ Unadjusted		Mean First FDA Approval	Mean # US Sellers
				Mean	Std. dev.	Mean	Std. dev.		
AU	1706	2008	2017	0.505	1.685	1.139	1.195	1980	4.25
BC	858	2015	2017	0.282	1.783	0.735	1.314	1983	4.3
NZ	1470	2009	2017	0.332	1.571	1.090	1.033	1982	4.23
ON	344	2017	2017	0.346	1.578	0.886	1.110	1984	4.88
UK	1625	2010	2017	0.221	1.839	0.899	1.321	1981	4.17

(b) Medicare Part D - Molecule Comparison

Comparison	Obs	Start Year	End Year	$\log(P_{US}/P_{Dest})$		Mean First FDA Approval	Mean # US Sellers
				Mean	Std. dev.		
AU	764	2012	2016	1.076	1.410	1981	9.98
BC	288	2015	2016	0.838	1.529	1981	9.51
NZ	803	2012	2016	1.046	1.241	1981	9.78
ON	237	2016	2016	1.011	1.337	1983	10.17
UK	852	2012	2016	0.671	1.350	1981	9.32

(c) NADAC - Molecule-Dose-Form Comparison

Comparison	Obs	Start Year	End Year	$\log(P_{US}/P_{Dest})$		Mean First FDA Approval	Mean # US Sellers
				Mean	Std. dev.		
AU	1250	2013	2017	0.831	1.494	1982	5.63
BC	1099	2015	2017	0.230	1.510	1983	5.24
NZ	1180	2013	2017	0.629	1.288	1983	5.58
ON	463	2017	2017	0.349	1.341	1984	5.80
UK	1467	2013	2017	0.389	1.571	1983	5.40

Notes: Based on authors' calculations of public expenditure, price, and seller data. Raw prices for Medicaid refers to prices before accounting for rebates. Adjusted prices reflect rebates and accounting for dispensing fees. Data for Medicaid Part D and from NADAC are not adjusted. See text for further details.

Table 2: Medicaid Off-Patent Supplier Data - 2017

Suppliers	Doses Sold	Prescriptions	Value (\$)
1 Supplier	257,998,493	7,114,496	919,318,884
2 Suppliers	306,887,650	7,715,258	331,637,050
3 Suppliers	579,407,652	12,895,900	1,052,758,754
4 Suppliers	1,313,802,332	30,574,310	1,120,181,810
5 Suppliers	2,011,558,200	46,263,697	976,936,333
6-10 Suppliers	8,960,672,054	216,520,993	2,749,776,720
11+ Suppliers	4,064,083,412	81,402,692	1,529,467,082
Total	17,494,409,794	402,487,346	8,680,076,633

Notes: Based on authors' calculations. See text for further details.

5.1 Key Fact: US prices are only higher in markets with low competition

We compare the price differential between the US and foreign market by number of approved US suppliers. We compute the relative price differential between the US price and a foreign price as:

$$\text{premium}_{f dy} = \frac{\text{price}_{US dy}}{\text{price}_{f dy}}. \quad (2)$$

Subscript d references a molecule-dose-form combination for the Medicaid and NADAC data and a molecule for Medicare Part D data. Subscript y refers to the year of the observation and f refers to the country of the observation. The price ratio allow for the differencing out of multiplicative market-level fixed effects between country pairs. We non-parametrically recover the relationship between the premium and the number of suppliers participating in the US marketplace with US FDA regulatory approvals to distribute drugs:

$$\ln(\text{premium}_{f dy}) = \sum \beta_s \mathbb{I}_S(S_{US dy}) + \delta_{fy} + \epsilon_{f dy} \quad (3)$$

We use \mathbb{I}_S as an indicator function for the number of US approved suppliers: S_{dyUS} . β_s are a vector of relative prices differences, δ_{fy} are year-comparison market fixed effects, and $\epsilon_{f dy}$ represent measurement error. This is a version of Equation 1, using the foreign price to directly control of marginal costs.

Figure 1 shows the coefficients β_s from Equation 3 of the premium relative to each foreign country on the number of suppliers in the US market.³⁰ The premium is computed using prices for Medicaid (Sub-Figure 2a), Medicare Part D (Sub-Figure 2b), and average prices across all types of insurance (Sub-Figure 2c). Standard errors are robustly clustered at the molecule-level.

As the number of suppliers increases, the premium declines towards zero. The US has a higher relative price when there are fewer than 6 suppliers, after which there is only a small difference in average price. The relationship holds across all 5 comparison destinations and for all types of insurance in the US. This suggests that competition is effective in larger markets with more suppliers, but strong buyer bargaining achieves lower prices in smaller markets with fewer entrants.

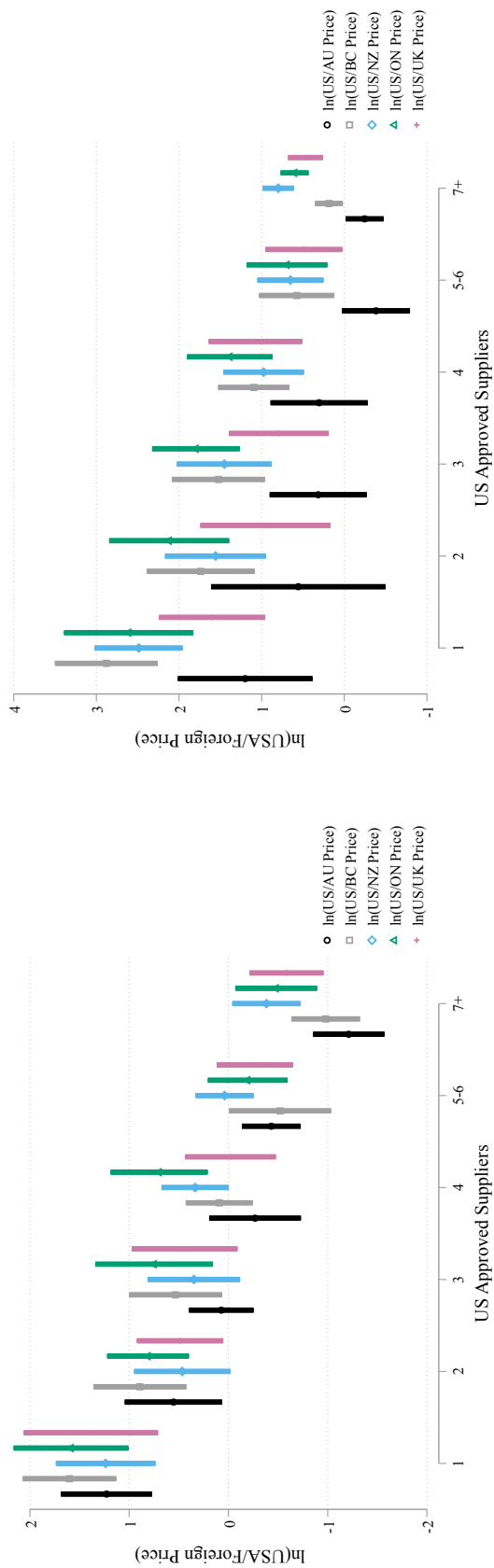
These results are formalized by estimating regressions of the log price differential between the US and another country ($\ln(\text{premium}_{f dy})$) on the log of the number of US suppliers ($\ln(S_{US dy})$). Fixed effects for the year are included to capture time trends. The regression is estimated separately for each country and each US data source.

$$\ln(\text{premium}_{f dy}) = \beta \ln(S_{US dy}) + \delta_{fy} + \epsilon_{f dy} \quad (4)$$

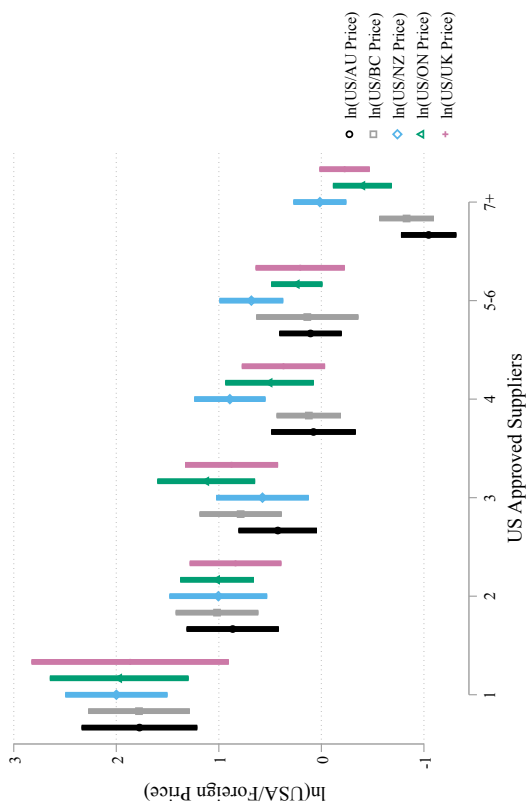
The coefficient of interest, the elasticity of the price premium to the number of suppliers in the US (β), is identified by variation in the number of suppliers across drug markets. As is standard

³⁰With the baseline being the first observed year.

Figure 1: Role of competition - Non-parametric Regressions



(a) Medicaid Data - Molecule-Dose-Form



(b) Medicare Data - Molecule

(c) NADAC Data - Molecule-Dose-Form

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B) with the number of FDA approved competitors of the US. Regression specification from Equation 3 with standard errors clustered at the molecule level. 95% confidence intervals displayed. See text for data sources and details. Year-destination fixed effects are removed by de-meaning. Results with all base years, omitting the 7+ supplier category produce similar results.

in demand estimation, we assume prices are more flexible than market entry and have no dynamic effects. Under these assumptions and if ϵ represents a mean zero shock after supplier entry, we can interpret β as a causal relationship between the number of suppliers and the pricing differences (but not absolute price levels).

Table 3 shows the results. The interpretation of the coefficients is as follows. A one percent increase in the number of suppliers in the US reduces the price differential between US Medicaid and Australia by 1.1% (British Columbia 1.1%; NZ 0.73%; Ontario 0.88%, UK 0.91%). Similar results hold for Medicare and NADAC data.

There are a few clear threats to identification, from both the supply side and the demand side. On the supply side, ϵ may allow for differences in marginal cost between markets. For example marginal costs may be systematically related to the number of suppliers. However we directly control for the marginal cost of production by considering the relative prices between two nations and absorb differences in distribution costs and exchange rate using the country-year fixed effect δ_{fy} .

On the demand side, there may be differences in substitutability between markets. For example, there may be many cardiovascular over the counter alternatives, but very few such anti-epileptic alternatives. However, our relative price differences are robust, controlling for the age of a drug, number of similar drugs (with country-pair fixed effects on Anatomical Therapeutic Chemical (ATC) code-year combinations), the number of other available dosage forms (controlling for molecule data), and using the lagged number of suppliers to control for sticky prices.³¹ Additionally, these relationships have been stable since 2010, with little observable changes in the relationship between suppliers and price premiums.³² Having controlled for the US supplier side, we now take a look at competition in foreign markets.

It is possible that the price differential can be explained by differences in the market size across destinations, due to differences in preferences for new versus generic drugs or differences in the prevalence of diseases. We examine this issue by estimating Equation 1, across two markets. Taking the logarithm of the price ratio, and assuming marginal cost are the same across countries, gives Equation 5.

$$\ln(\text{premium}_{f dy}) = \beta_{US} \ln(S_{f dy}) + \beta_F \ln(S_{f dy}) + \delta_{fy} + \epsilon_{f dy} \quad (5)$$

If the price premium can be explained by the foreign markets having more suppliers than the same US market, controlling for the foreign supplies should explain the price premium. If not, this suggests that bargaining plays a role in the price differential. Table 4 shows the results for the regression with

³¹See the Appendix for robustness checks. An ideal instrument for for S would be some sort of a shock that allows for an additional entrant. Using the home market effect as in Costinot et al. (2016) would be difficult as the home market effect also reveals underlying heterogeneity that violates the exogeneity restriction. Controlling for ATC codes (by year) should absorb country differences in preferences in pharmaceutical demand. We find that this has no statistical effect on the relationship of relative prices to the number of suppliers. Finally if we assume that demand is exogenous, we can use the number of doses sold as an instrument for the number of suppliers. Results are largely unchanged.

³²There are level changes, which can often be accounted for by large shifts in exchange rates. Australia and Canada are natural resource-driven economies and have volatile currencies. The United Kingdom initiated proceedings in 2016 to disengage from the global economy, which further influenced exchange rates.

Table 3: Relationship between Price Premiums and Number of US Suppliers

Medicaid - Molecule-Dose-Form

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.105*** (0.110)	-1.142*** (0.0898)	-0.736*** (0.104)	-0.875*** (0.106)	-0.909*** (0.127)
Adj. R-Square	0.281	0.257	0.134	0.162	0.129
Observations	1706	858	1470	344	1625
Fixed Effects	year	year	year	year	year

Medicare Part D - Molecule

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(Part D Suppliers)	-0.481*** (0.124)	-0.884*** (0.0930)	-0.494*** (0.0891)	-0.695*** (0.102)	-0.325** (0.106)
Adj. R-Square	0.133	0.339	0.157	0.229	0.0477
Observations	764	288	803	237	852
Fixed Effects	year	year	year	year	year

NADAC - Molecule-Dose-Form

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.252*** (0.0927)	-1.133*** (0.0709)	-0.819*** (0.0831)	-0.992*** (0.0849)	-0.855*** (0.113)
Adj. R-Square	0.407	0.310	0.234	0.248	0.149
Observations	1250	1099	1180	463	1467
Fixed Effects	year	year	year	year	year

Standard errors in parentheses, * p<.05, ** p<.01, *** p<.001.

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

both the number of US and comparison country suppliers. The foreign suppliers are insignificant for three countries (Canada, New Zealand, and the United Kingdom). These results imply that a change in the number of competitors in a foreign market are not correlated with either higher or lower prices (relative to the US). One possible reason is that buyers in foreign countries set strict price controls. These price controls limit producer surplus, irrespective of the number of sellers present. We will directly model this difference in outcomes between the US and foreign markets in our structural model.

The coefficients are negative for Australia when considering US Medicaid price data. An increase in the number of Australian suppliers means that Australian prices increase relative to the United States. Competition has the opposite expected effect. There are three straightforward explanations: one is a type-1 positive error (spurious correlation). Second, Australian drug approvals are correlated with US drug approvals (Australia quickly approves US drugs), thus the coefficients for US and Australian suppliers do not have sufficient independent variation. Lastly, Australian authorities may be quickly approving additional suppliers only when prices are high. We can reject the last explanation, as this relationship holds even considering new approvals with a lag of one or more years.³³

5.2 Motivating Facts

With the relationship between market entry and prices established, we provide a series of motivating facts to better understand the marketplace, support our model, and motivate counterfactual simulation.

5.2.1 Market size is highly correlated with the number of competitors

The main finding shows the relationship between the number of suppliers and the price differential between the United States and foreign markets. However, a critical piece of information revolves around the incentives of a generic pharmaceutical supplier to enter a marketplace. If there are few treatable patients requiring a small number of doses, there may be a smaller incentive for a supplier to enter the marketplace. We directly consider the relationship of suppliers to market size in Figure 2. This comparison is only done between Australia, the United States (under Medicare Part D and Medicaid), and the United Kingdom, as these systems make publicly available the total number of capsules/tablets sold.

As shown in Figure 2, an increase in the number of prescribed doses is correlated with an increase in the number of suppliers.³⁴ Across all three countries a doubling in the number of patients is correlated with an approximately 20-30% increase in the number of pharmaceutical suppliers. This

³³Furthermore, controlling for year-ATC fixed effects, reduces the importance of Australian suppliers on the US-Australia price difference. See Appendix for further discussion and tables.

³⁴These results use data for Medicaid. The results using Medicare Part D and NADAC data are broadly similar and are shown in the Appendix.

Table 4: Relationship between Price Differentials and Suppliers

Medicaid - Molecule-Dose-Form

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-0.928*** (0.126)	-1.022*** (0.110)	-0.765*** (0.122)	-0.834*** (0.122)	-0.966*** (0.173)
ln(Foreign Suppliers)	-0.314*** (0.0924)	-0.239* (0.0950)	0.0927 (0.114)	-0.0822 (0.108)	-0.0586 (0.171)
Adj. R-Square	0.298	0.275	0.127	0.168	0.136
Observations	1706	756	1408	335	273
Fixed Effects	year	year	year	year	year

Medicare Part D - Molecule

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-0.500*** (0.138)	-0.729*** (0.181)	-0.442*** (0.118)	-0.564** (0.179)	-0.408* (0.160)
ln(Foreign Suppliers)	-0.153 (0.101)	-0.146 (0.126)	0.0305 (0.114)	-0.0925 (0.122)	0.106 (0.164)
Adj. R-Square	0.153	0.230	0.0906	0.138	0.0245
Observations	755	244	760	226	191
Fixed Effects	year	year	year	year	year

NADAC - Molecule-Dose-Form

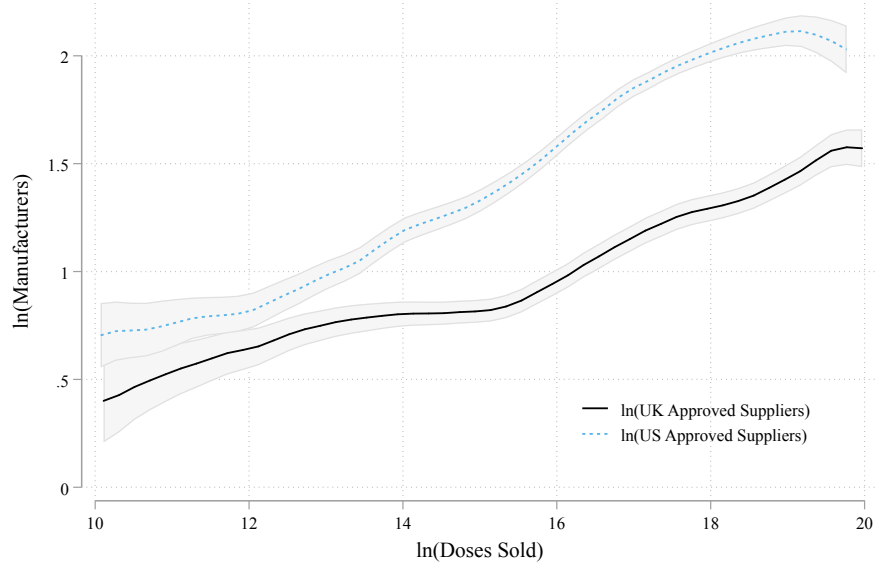
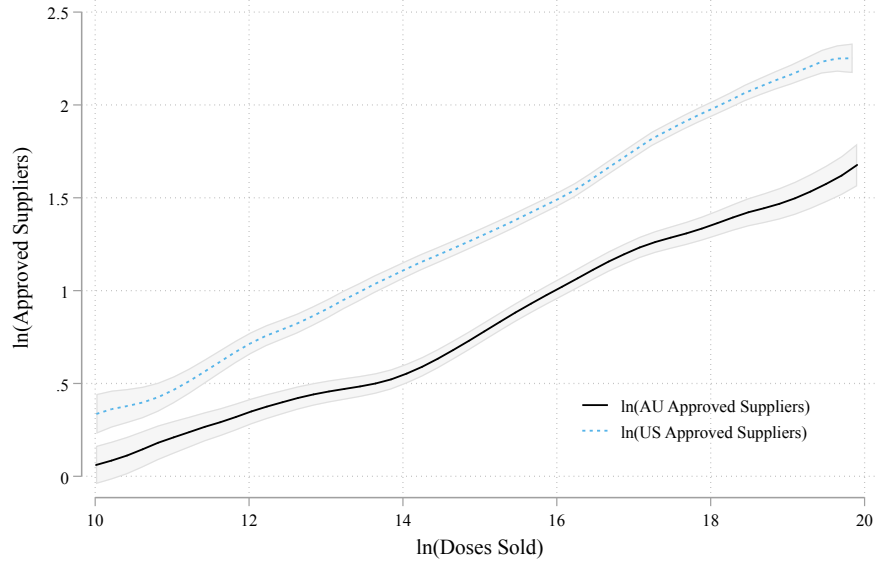
	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.138*** (0.113)	-0.994*** (0.0839)	-0.829*** (0.101)	-0.940*** (0.103)	-0.927*** (0.164)
ln(Foreign Suppliers)	-0.161* (0.0724)	-0.251*** (0.0681)	0.0135 (0.0989)	-0.0883 (0.0763)	0.0926 (0.131)
Adj. R-Square	0.413	0.337	0.231	0.258	0.124
Observations	1250	1023	1153	455	358
Fixed Effects	year	year	year	year	year

Standard errors in parentheses, * $p < .05$, ** $p < .01$, *** $p < .001$.

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 5 with standard errors clustered at the molecule level. See text for data sources and details.

Figure 2: Relationship between Suppliers and Market Size

(a) Australia-US Medicaid Matched Molecule-Dose-Form



(b) UK-US Medicaid Matched Molecule-Dose-Form

means that the marginal benefit of entry in terms of capturing profit from quantity (market size) is approximately the same across countries. This is in contrast to the results of Section 5.1, which showed that marginal benefit to entry to capture profit from higher prices is larger in the United States, than in either Australia or the United Kingdom. One possible rationalization is that the US has higher entry costs (or barriers to entry), preventing supplier entry and lower prices.

5.2.2 US Prices are relatively higher in markets with few patients

A natural policy response to insufficient competition would be to allow more entrants. In Figure 3 as market size increases, the US price premium decreases. This combined with the earlier finding that that drugs with few suppliers have large US price premiums indicates that the number of entrants is limited by the underlying demand for a particular molecule from patients and doctors. Reducing barriers to entry may allow more entrants, but there may be natural bounds on the total market size, limiting the effectiveness of such liberalization policies.

5.2.3 There is substantial variation in the number of approved drug suppliers by country

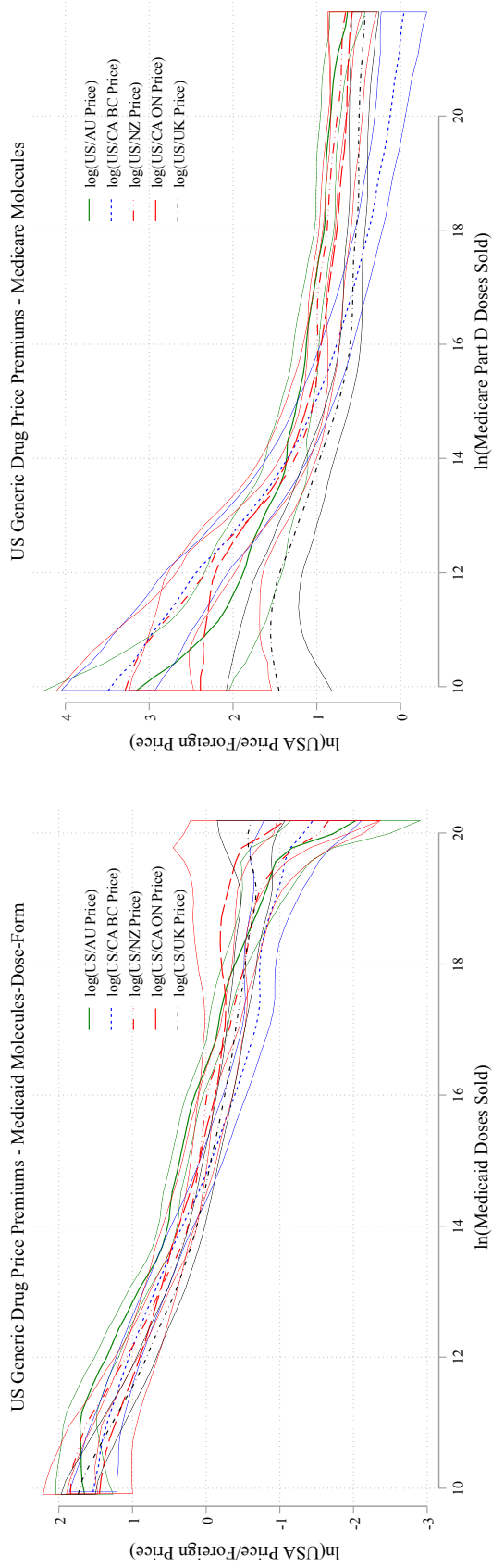
Harmonizing market entry across the English speaking world can only be effective at lowering prices if there exist additional entrants. Data from the five countries in our data sample show substantial variation in the number of suppliers at the molecule level. For example, the anti-epileptic generic gabapentin has 28 suppliers in Canada, but only 25 suppliers in the United States. Figure 4 systematically classifies this distribution. In some cases, the US, being a large market, has many more approved suppliers than our comparison countries. However, there is a significant left tail of molecules with suppliers that have entered foreign markets, but have not entered the US market. This suggests that allowing such suppliers better access to the US market could have the effect of reducing pharmaceutical prices through competition.

5.2.4 Validating Assumption: Inelastic Demand

To close out our exploration of off-patent and generic pharmaceuticals, we estimate the elasticity of quantity to the price.³⁵ This is a molecule-market level demand elasticity, not a firm-level elasticity. We want to understand how the market size changes with respect to market-level prices. For a model, we need this demand elasticity to understand the incentives of suppliers in their pricing decisions. In many non-American markets, generic demand can be considered completely inelastic as patients and prescribers do not observe costs. This is also largely true for Medicaid patients in the United

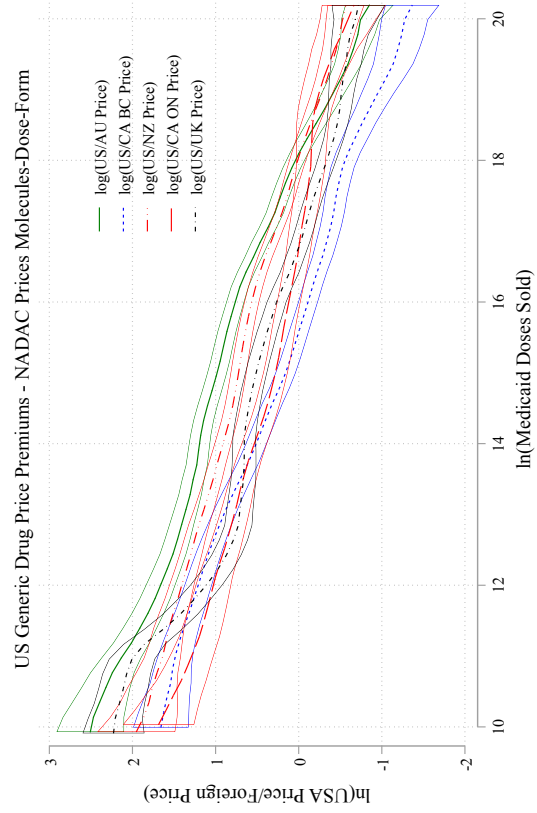
³⁵There are many studies that compute the elasticity of demand with respect to consumer prices in the United States under various insurance regimes (see [Abaluck et al. \(2018\)](#); [Einav et al. \(2016, 2015\)](#)). However, these elasticities are relative to consumer co-pays, not the supplier's price. Even still, they find extremely small elasticities not far from completely inelastic demand.

Figure 3: Role of Market Size



(a) Medicaid Data - Molecule-Dose-Form

(b) Medicare Data - Molecule

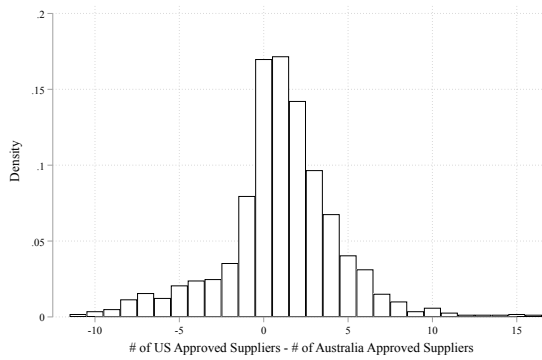


(c) NADAC Data - Molecule-Dose-Form

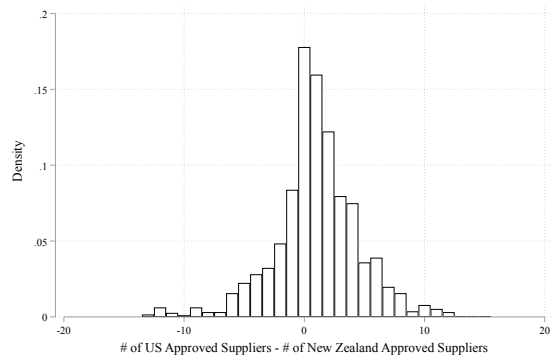
Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B) within the market size of the US.

Figure 4: Differentials in Supplier Numbers

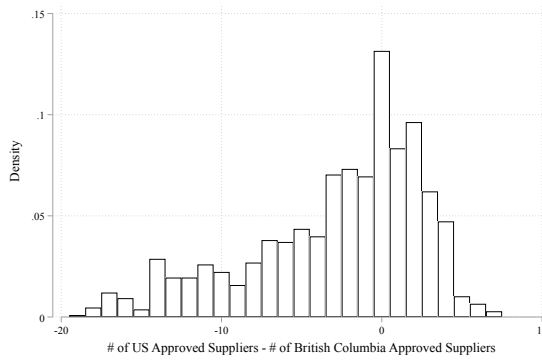
(a) Australia Comparison



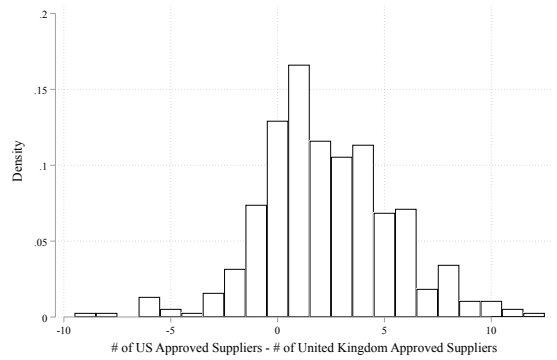
(b) New Zealand Comparison



(c) Canada Comparison



(d) UK Comparison



Notes: Supplier data sourced from public data of approvals. See text for further details.

States. A consumer may pay a nominal fixed co-pay that varies at the state level and is typically either constant or zero for all generic drugs.³⁶ However, patients in both Medicare Part D and private insurance plans face co-payments and/or co-insurance. For these patients, a straightforward regression of quantity on price at the molecule level is likely to suffer from simultaneity issues.³⁷

We propose and implement two independent strategies. We first consider at a subset of drugs with sudden price changes. We then consider global exchange rate fluctuations, that influence marginal costs, but not demand. Both produce similar results, substantially inelastic demand, less than one and near zero. While we are only estimating point elasticities, not the entire demand curve, we do not conduct counterfactual analyses with large upward price changes, and use this local approximation to guide our exercise.

Large Price Increases: We partially circumvent this issue by considering off-patent pharmaceuticals with an overnight price increase of more than 100%. This is plausibly exogenous variation because these price increases primarily occur when the molecules change hands (Berndt et al., 2017). Additionally, we do not believe that such increases in price are driven by demand side factors, but rather supply-side changes regarding drug ownership. We run the following first difference regression to remove the molecule/dose level fixed effects on this subset of pharmaceuticals:

$$\Delta_y \ln Q_{USdy} = \beta \Delta_y \ln P_{USdy} + \delta_y + \epsilon_{USdy} \quad (6)$$

The fixed effects δ_y control for time trends in overall prescription trends. We assume ϵ reflects a set of orthogonal shocks and that any demand shifters are differenced out.

Table 5 shows the results for different intervals of time for both Medicaid and Medicare Part D patients using various prices. Panel A considers Medicaid prices and Medicaid doses at the molecule-dose-form level. Panel B considers Medicare Part D prices and Medicare Part D doses at the molecule level. Panel C considers NADAC prices and Medicaid doses at the molecule-dose-form level.

All the results are either economically small or statistically insignificant, suggesting that quantity is very inelastic to price changes. It appears that in both Medicaid and Medicare Part D, off-patent and generic drug consumption is largely insensitive (to a first degree approximation) to prices. Due to institutional details, as Medicaid patients face minimal cost-sharing, we both expect and find almost no correlation with price to consumption. As for Medicare Part D, our results are possibly driven by our consideration of only off-patent pharmaceuticals. Even though there is cost-sharing in nearly all Part D plans, the vast majority of generic drugs appear in the lowest cost tiers and face the lowest levels of cost-sharing with patients.³⁸ These findings are in line with the extremely

³⁶While this is true for generic pharmaceuticals, this is not entirely true for on-patent and/or innovator drugs.

³⁷We also do not consider the entry and exit of related molecules, both generic and under patent. In a robustness exercise, we find that including ATC3-year fixed effects, to control for changes in such markets, has little effect.

³⁸Without individual data, it is impossible to determine the degree of patient-level cost-sharing. However, nearly all Part D plans from private insurance companies list most generic drugs (particularly those in tablet or capsule

Table 5: Estimated Price Elasticities - Using Price Jumps

Medicaid - Molecule-Dose-Form				
	(1)	(2)	(3)	(4)
	D.log(q)	D2.log(q)	D3.log(q)	D4.log(q)
Log Price Change	-0.00398 (0.00638)	-0.0371 (0.0271)	-0.0246 (0.0334)	-0.0189 (0.0369)
Observations	1081	1886	1430	1081
R^2	0.015	0.077	0.044	0.022
FE	year	year	year	year
Medicare Part D - Molecule				
	(1)	(2)	(3)	(4)
	D.log(q)	D2.log(q)	D3.log(q)	D4.log(q)
Log Price Change	0.0206 (0.116)	0.208 (0.143)	0.0911 (0.0724)	0.271 (0.219)
Observations	193	143	44	44
R^2	0.071	0.038	0.048	0.047
FE	year	year	year	year
NADAC - Molecule-Dose-Form				
	(1)	(2)	(3)	(4)
	D.log(q)	D2.log(q)	D3.log(q)	D4.log(q)
Log Price Change	0.138 (0.0913)	-0.0371 (0.0271)	-0.0246 (0.0334)	-0.0189 (0.0369)
Observations	1730	1886	1430	1081
R^2	0.019	0.077	0.044	0.022
FE	year	year	year	year

Standard errors in parentheses, * $p < .05$, ** $p < .01$, *** $p < .001$.

Notes: Based on authors' calculations of public expenditures and price data. Adjusted Medicaid prices reflect rebates and accounting for dispensing fees. Data for Medicaid Part D and from NADAC are not adjusted. See text for further details.

inelastic demand from [Einav et al. \(2016\)](#).³⁹

International Exchange Rate Fluctuations In a second exercise, we assume that exchange rates are determined largely exogenously from generic pharmaceutical prices. As a large proportion of US generics are manufactured abroad, they are subject to exchange rate fluctuations. As is standard in the literature, we assume that these changes affect marginal costs and are independent

form) as part of "Tier 1" and "Tier 2", which refer to "Preferred Generic Drugs" and "Generic Drugs" with low co-pays typically ranging from \$0-\$20 for 30 days of treatment. While this is not universal, and insurers may change the tier a generic drug appears on, in our sample this does not seem to affect the quantity produced and purchased.

³⁹We are only estimating demand elasticity for drugs that changed prices, over the range of the price change. These tend to be drugs that have limited substitutes. It may be possible that demand for other generic drugs is more elastic. However, a simple OLS regression using the entire pooled data finds similar results.

of any US-specific demand shocks. Thus we create an instrument:

$$IV_{dy} = \log \sum_{i \in I} \frac{\Delta EP_{d,y,i} \times q_{d,y-1,i}}{\sum_{i' \in I} q_{d,y-1,i'}}$$

This takes the expected price change due to exchange rate variation between y and $y - 1$ for drugs sourced from country i from the set of possible country sources I : $\Delta EP_{d,y,i}$ and weights by the lagged share of sales originating from country i . The first stage results in Table 6 show the explanatory power of this instrument. Primarily driven by exchange rate fluctuations to the Indian Rupee, Chinese Yuan, and the Euro, a weakening of the dollar leads to higher prices. The third column shows the estimates, with elasticities near zero.

Taken together, this inelastic nature of demand shows the possible welfare-enhancing role played by strong cost-controlling buyers, and further motivate both our model and our counterfactual with a strong bargaining monopolist buyer. In particular, with inelastic demand, a monopolist selling to a buyer without market power would not be profit maximizing. This motivates a model of bargaining between two sides, each with buying and selling power.

6 Structural Model

For the purpose of comparing policies and conducting counterfactuals, we construct a structural model of the off-patent and generic drug market. We first consider a canonical model with a single monopolist buyer and a single monopolist drug seller, and then extend this to cover different levels of oligopoly. As described in Section 3, we are interested in the actions and outcomes of two key players. Drug sellers maximize profits and drug buyers minimize costs. Prices are determined through Nash bargaining. Markets with more than one seller are defined implicitly, weighted between the results for a monopolist seller and perfect competition.⁴⁰ We model total demand as exogenously determined, due to the inelasticity of demand.⁴¹

The model has two stages, the first stage involves recovering the determinants of pricing differentials, conditional on the number of market participants. The second stage recovers marginal relative entry costs (on an discounted yearly-basis) to rationalize the number of market participants. Critically, we do not directly model the mode of competition between pharmaceutical companies, instead we recover a policy and counterfactual-relevant competition parameter (a sufficient statistic). Additionally, we do not model the set of drugs covered by various public plans (the formulary), as this set is rarely bargaining over in generic prescription drug markets.

⁴⁰Alternatively in the Appendix, markets with more than one seller can be defined recursively, with the outside option for sellers being zero (being shut out of the market), and the outside option for buyers being the price charged in a market with one less seller.

⁴¹For evidence from US markets, we show that price spikes of over 100% have very little effect on demand over a 1-2 year time horizon in Medicaid plans in Table 5. In particular, patients covered by US Medicaid programs face minimal cost sharing and are not constrained by sticker prices. In all non-US markets we consider, with rare exceptions, prescribers for non-patented medications do not share variable costs.

Table 6: Estimated Price Elasticities - Exchange Rate Shocks

Medicaid - Molecule-Dose-Form

	(1)	(2)	(3)
	Least Squares	First Stage	Instrumental Variables
D.ln(P)	-0.0308* (0.0134)		0.337 (0.182)
ln(Expected Price Change)		0.0515*** (0.00799)	
Observations	5556	5556	5556
FE	Year	Year	Year
F-stat	5.281	41.60	3.402

Medicare Part D - Molecule

	(1)	(2)	(3)
	Least Squares	First Stage	Instrumental Variables
D.ln(P)	0.0313 (0.0837)		-0.103 (0.435)
ln(Expected Price Change)		0.0249** (0.00762)	
Observations	1094	1094	1094
FE	Year	Year	Year
F-stat	0.140	10.69	0.0557

NADAC - Molecule-Dose-Form

	(1)	(2)	(3)
	Least Squares	First Stage	Instrumental Variables
D.ln(P)	-0.0485* (0.0237)		-0.280 (0.346)
ln(Expected Price Change)		0.0241* (0.0120)	
Observations	2490	2490	2490
FE	Year	Year	Year
F-stat	4.202	4.059	0.654

Standard errors in parentheses, * $p < .05$, ** $p < .01$, *** $p < .001$.

Notes: Based on authors' calculations of public expenditures and price data. Adjusted Medicaid prices reflect rebates and accounting for dispensing fees. Data for Medicaid Part D and from NADAC are not adjusted. The first column is a naive OLS estimate. The second column is the first-stage of the IV estimate. The last column presents the results of the IV estimate. See text for further details.

Our model essentially has four degrees of freedom: First, the average price difference across all drugs between two nations. Second, the rate of convergence in prices between two nations, as the number of suppliers increases. Third, we have the price differential as the number of competitors becomes large. Fourth, for a subset of countries, we have the number of doses sold. We can capture all these moments in a parsimonious way, allowing for the computation of our main counterfactuals: creating a single market in the English speaking world and allowing US buyers to bargain with drug makers in a similar fashion to the rest of the English speaking world.

6.1 Price Setting

6.1.1 Baseline Scenario

We model the simplest case with one buyer and one seller as a Nash bargaining game. Drug sellers maximize profits and drug buyers minimize costs. Profits for a monopolist seller are $\pi_s = (p - c)q$, where prices are p and constant marginal costs are c . The drug buyer is attempting to minimize costs $p \times q$. If the buyer and seller cannot come to an agreement the seller receives nothing. The buyer still needs to supply local patients and must pay per-unit price \bar{p} , where $\bar{p} \in [c, \infty)$. This ‘choke’ price \bar{p} represents the cost of a local insurer or government either compounding the drug locally, paying for a new supplier of the drug, or the imputed loss of social welfare from not providing the drug.⁴² We do not model the entire drug market, so \bar{p} also includes the substitutability of over-the-counter medications, alternative treatments, the usage of compounding pharmacies, other similar prescription pharmaceuticals, and non-treatment. As the relationship of US and foreign prices are robust to country pair-drug class-year fixed effects (see Appendix A.7), we omit direct inclusion or modeling of this outside market.⁴³

The price is determined by the Nash surplus:

$$NS = (p - c)q^{w_s} (\bar{p}q - pq)^{w_b},$$

where the weights w_s and w_b are the positive seller and buyer bargaining weights that sum to 1 (without loss of generality).⁴⁴

Taking a logarithmic transformation of the Nash Surplus and then first order conditions over the price, the monopoly bargaining price is:

$$p_m = w_s \bar{p} + (1 - w_s) c. \tag{7}$$

⁴²This treatment is isomorphic to considering a social welfare function: \bar{p}_b is a stand-in for the welfare-relevant term. Legally, the Medicaid choke price is bounded by a function of the private insurance price (the Federal Upper Limit or FUL). However the FUL is easily manipulated through two-part payments and rebates - which are not included in the calculation of the FUL.

⁴³An alternative approach would directly model $\bar{p} = \bar{p}(\mathbf{O}, G)$, where the set \mathbf{O} includes all alternatives and G represents political risk of broken down negotiations.

⁴⁴In some markets, a buyer may have the ability to drop a medication from the list of subsidized or covered molecules (formulary changes). We indirectly model this through the bargaining weights and leave detailed study for future work.

Thus when the buyer has no bargaining weight, $w_s = 1$, the seller appropriates all surplus and charges $p = \bar{p}$. When the seller has no bargaining weight, $w_s = 0$, the buyer gives a take-it-or-leave-it offer where $p = c$ and appropriates all the surplus.

6.1.2 Generalization

In the bulk of our considered markets, all non-US markets and the US Medicaid market, there are multiple drug sellers, but only one buyer. In classical economic markets, with infinite entrants and zero seller market power, the price in the marketplace is simply the average cost:

$$p_c = c \tag{8}$$

With an intermediate number of entrants, we directly model the price premium over marginal cost as a markup weighted between the monopolist solution from Equation 7 and the perfect competition solution in Equation 8. This competition function $\theta(S) : \mathbb{I}^+ \rightarrow \mathbb{R} \in [0, 1]$ weights between these two solutions in a parsimonious manner:

$$p = \theta(S) p_m + (1 - \theta(S)) p_c \tag{9}$$

This Nash-in-oligopoly setup makes three major assumptions. First, the competitive parameter is multiplicatively separable from monopolist price and the marginal cost. Second, we generalize away from any market size issues other than what is included in the function $\theta(\cdot)$ and assume that other than the outside option, firms compete in an identical manner across markets. Lastly, we assume that all pharmaceutical suppliers have identical cost structures. The price p charged does not need to be identical across sellers. Rather, p is simply the average price charged. The end payer (mostly governments), simply seek to minimize expenditures, holding the behavior of prescribers fixed.⁴⁵ The appendix highlights four cases, subsumed by this model, Bertrand competition with a homogenous good, Bertrand competition with differentiated goods, a repeated game, and an extension to multiple buyers.

Rewriting our generalized price from Equation 9, we obtain the price paid by a particular buyer as a function of the competition parameter:

$$p = \theta(S) [w_s \bar{p} + (1 - w_s) c] + (1 - \theta(S)) c$$

This requires the identification of the function θ , bargaining weight w_b , outside option price \bar{p} and marginal cost c . However our data is not refined enough to identify all these elements using only data on p and S . In particular, the outside option \bar{p} and w_s are hard to separately identify. We

⁴⁵In the market entry game, we take seriously the notion that there may be differences between the prices, markups, and market shares of the original legacy branded supplier and other generic suppliers. However this distinction is not relevant for this stage.

assume that \bar{p} is a linear function of marginal cost, with an unknown buyer-specific parameter γ_b :

$$\bar{p} = \gamma_b c.$$

This multiplicative relationship is motivated by the existence of small compounding pharmacies, who take stable widely available raw materials at cost c and package them for consumer use in small quantities, charging substantial markups to cover their overhead (Higby, 2003).

Simultaneously, we define the buyer-specific leverage parameter:

$$\kappa = [w_s \gamma_b + 1 - w_s] \in [1, \infty)$$

As noted before, as we do not know each buyer’s γ_b or ω_b , these two terms are only jointly identified through the sufficient statistic κ . We call this term the ‘buyer’s leverage’. As a buyer’s bargaining weight goes to zero, this term goes to γ_b , with the buyer facing the outside option. The term κ varies across different buyer-seller pairs, who may face differences in their outside options or bargaining weights, including political risk and variation in the prices or availability of substitute pharmaceuticals.⁴⁶

With this assumption, Equation 9 becomes:

$$p = c [\theta(S) \cdot \kappa + (1 - \theta(S))] = \mu_b(S) c \tag{10}$$

We parsimoniously define $\theta(S) = \exp(\alpha \log S)$, which implicitly defines $\theta(1) = 1$ and $\theta(\infty) = 0$ and matches our reduced form logarithmic relationships. Furthermore $\mu(S)$ is the average markup charged to the end seller in a market with S suppliers.

6.2 Market Entry: Maximum Effective Entry Costs

Our goal is to recover the entry cost of a marginal generic entrant. One of our counterfactuals will be to align these entry costs across countries to stimulate additional entry. We consider the imputed per-period fixed costs for entry into a single country. (As opposed to global entry costs of creating a new factory). This can be interpreted as the present discounted value of entry.⁴⁷ The US, using a strict interpretation of the law, does not appear to have higher regulatory entry costs than similar countries, such as Australia, which have similar safety and documentation requirements. In practice, the US takes a lengthy period to review the entry of generic pharmaceuticals, averaging 44 months in 2016 (Lupkin, 2015). However, the US is also larger country, so there may naturally be higher entry costs related to true fixed costs for delivery and distribution. Generic markets are also rife with collusion (Wall Street Journal 2016) and bi-lateral payments from incumbents to entrants (Kaiser

⁴⁶ Additionally, κ_b could include variation within buyers across different drug classes, with different off and on-patent substitutes.

⁴⁷ Our model here is static. We do not consider the inter-temporal dynamics of these complex markets.

Health News 2019). We conflate all these costs to infer imputed entry costs, recovering the effective market value of entry.

Market entry in the Australia, Canada, New Zealand, and the UK is straightforward as institutional rules effectively mandate uniform reimbursements across suppliers of a particular molecule. However, this is not necessarily true in the United States. In Medicare and Medicaid about 10% of doses are sold by the original branded supplier, charging 20% over the generic price.⁴⁸ While there are institutional rules in Medicaid and Medicare Part D to encourage the use of cheaper generics over nearly identical branded drugs, prescribers and patients may have other incentives. This is a complicated problem. Instead of directly modeling consumer behavior, we consider the direct role played by the number of suppliers on the share of the original pharmaceutical and the price premium charged by the original pharmaceutical supplier.

We furthermore assume that marginal costs are identical across suppliers. With the competition function $\theta(S)$ and buyer leverage κ_d recovered from demand estimation, we recover an estimate for average costs c for an observation at the drug (d), year (y), country/market (f) level:

$$c_{fdy} = \frac{R_{fdy} \times S_{fdy}}{\mu_f(S_{fdy}) \times q_{fdy}} \quad (11)$$

We use data on average revenues (R), pharmaceutical suppliers (S) and sold doses (q). The markup function $\mu_f(\cdot)$ is derived from Equation 10, and is a composite of the recovered competition function $\theta(\cdot)$ and buyer leverage κ_f .

Combining marginal cost and price data, we can recover average markups μ as a function of the markups of the original branded drug and the markups of the generic competitors (omitting drug and year subscripts for clarity):

$$\mu = \mu_n(S) \times s_n(S) + \mu_g(S) \times s_g(S) \quad (12)$$

This is function of the markup (μ_g) charged by the generic suppliers (which we assume are identical) and the the original branded supplier (μ_n), weighted by their respective total market shares $s_i(\cdot)$. We recover μ_n and μ_g by assuming common marginal costs, thus $\mu_n = p_n/c$ and $\mu_g = p_g/c$. As our counterfactual will not consider the role played by the original branded entrant (which happened at least 20 years prior to the generic entrants), we simply take the combined markup and resulting market share behavior jointly: $\mu s_n(S) = \mu_n(S) \times s_n(S)$. With the markup function $\mu_g(S)$ and total generic market share $s_g(S)$, the profits of a marginal entrant in the United States are bounded by:⁴⁹

$$\Pi_{US,generic} = \frac{q \times s_g(S) \times c \times [\mu_g(S) - 1]}{S - 1}$$

If there are no markups (a multiplicative markup of 1), then there are no excess profits over average

⁴⁸Similar patterns hold in Continental Europe(Kanavos et al., 2008).

⁴⁹Empirically there appears to be very little variation in price (and therefore markups) among generic competitors. Thus we treat these firms as identical competitors.

cost. However if markups are greater than one, we consider the notion of “excess” profits for generic suppliers in the USA market relative to another as:

$$\Pi_{excess} = \frac{q_{US} \times s_{g,US}(S_{US}) \times c_{US} \times [\mu_{g,US}(S_{US}) - 1]}{S_{US} - 1} - \frac{q_F \times c_F \times [\mu_F(S_F) - 1]}{S_F} \quad (13)$$

As an upper bound for the number of additional entrants, we consider how many more entrants the US market could support if $\Pi_{excess} \rightarrow 0$. For this we compute the largest S^* such that the following inequality is satisfied:⁵⁰

$$\Pi_{excess}(S_{US}^*) \geq 0$$

We consider this as an upper-bound as it is possible that $S_{US}^* > S_{World}$. This is due to two limitations, both data-based. First, we do not know the total number of suppliers that operate outside the English speaking world. Second, we do not know the total number of global suppliers. A marginal new global supplier must not only cover the average cost of manufacturing and distribution in a particular country (which we account for), but must also set up a new manufacturing line (for example in India, Puerto Rica, or Bulgaria).⁵¹

7 Identification & Estimation

Estimation works through backward induction. We first identify the policy relevant parameters in θ and κ , using observed variation in the number of suppliers present in various international markets. The central assumption here is that, conditional on the number of entrants, prices are fully determined in a market. This assumption is valid if there aren’t dynamic concerns, such as if prices today determine entry tomorrow. With data on prices and suppliers for a variety of countries and regulatory schemes, we seek to recover data on α and κ_b where $b \in \{US, UK, NZ, AU, CA_{ON}, CA_{BC}\}$. Identification stems for cross-country price comparisons stemming from variation in market entry across countries.

7.1 Pricing Identification and Estimation

For every pair of buyers, there are four types of parameters to be identified, competition α , buyer leverages κ_1, κ_2 , and relative average cost differences $c_{1/2}$. These parameters are identical across all countries. All four variables have natural simple analogs in the data, using variation in the number of firms participating and in relative prices.

All intuition can be derived from taking the ratios of the prices between two countries for an identical

⁵⁰As $\Pi_{excess}(S_{US}) > 0 \forall S$ and $\Pi'_{excess}(S_{US}^*) < 0$ in the empirical setting, there is always a solution where $S_{US}^* \in [S_{US}, \infty)$.

⁵¹As shown in the Appendix, all developed English-speaking counties have zero tariff barriers to entry for nearly all pharmaceutical products in tablet or capsule form.

molecule/dose:

$$\frac{p_1}{p_2} = \frac{\exp(\alpha \log S_1) [\kappa_1 - 1] + 1}{\exp(\alpha \log S_2) [\kappa_2 - 1] + 1} c_{1/2}$$

Here, S_1 and S_2 denote the number of suppliers in country 1 and 2, respectively. To identify cost differences $c_{1/2}$, take the $S_1, S_2 \rightarrow \infty$, then the price ratio asymptotes to the cost ratio $c_{12} = c_1/c_2$. When the number of entrants in both countries is extremely large, the market should resemble perfect competition. Thus the difference in prices is simply the difference in marginal costs. Costs such as distribution, labelling, and packaging may vary between countries. The parameter $c_{1/2}$ parsimoniously reveals relative differences in these costs between two countries.

For the buyer leverage κ_1 , fix the number of sellers in one country at 1 and let the number of sellers in another country grow larger. Then the price ratio asymptotes to $\kappa_1/c_{1/2}$.

Similarly, we can identify κ_2 . Identification uses differences in the number of suppliers between countries. One country (the country with $S \rightarrow \infty$ is assumed to be in perfect competition and the other country $S = 1$ is assumed to be a monopoly. While this is an extreme case, we can still use variation between duopolies, tri-opolies, and other combinations to identify κ_1 and κ_2 . Implicitly, we assume exogenous variation in total units demanded in each country. For example, there may be more patients that require the anti-malarial Chloroquine in Australia than in Canada. This will drive a difference in the number of entrants and thus help identify κ_1 and κ_2 . This difference can also be driven by differences in the number of suppliers due to variation in country and market-level fixed entry costs.

Intermediate variation in the number of entrants in each market S_1 and S_2 identify the competition parameter α .⁵² For intuition, fix $\kappa_1 = 2$, $\kappa_2 = 1$ and $c_{1/2} = 1$. Then, relative price differences identify α , as $p_1/p_2 = \exp(\alpha \log S_1) + 1$. This variation in the number of firms S_1 , when larger than 1 and less than ∞ , pins down α . We assume the competition parameter α is identical across countries.⁵³ Robustness checks, allowing for variations in α along pair-wise comparisons finds similar results using a common α . Effectively, when there is one strong bargainer ($\kappa_1 = 1$), α is primarily estimated by another country with weak bargaining. Empirically, this weak bargainer is the United States.

Critically, we do not observe marginal costs, and assume a marginal costs ratio $E\left(\log \frac{c_1}{c_2}\right) = c_{1,2}$. The estimated parameter $c_{1,2}$ defines the average difference in marginal cost between market 1 and market 2. Differences in marginal costs $c_{1,2}$ between markets reflect idiosyncratic terms that are uncorrelated with market entry.

Identification assumes that the the estimated price ratios deviated from predicted price ratios with

⁵²While we impose a logarithmic structure on $\theta(S)$, in principle we can estimate this function non-parametrically. We perform a simple extension in the Appendix and find consistent results.

⁵³The parameter α is only identified for countries where $\kappa > 1$. That means that the seller has some bargaining power. However, as shown below - in nearly all countries outside of the US, κ is very close 1; preventing α from being identified on a market-by-market basis.

a zero mean:

$$E \left(\log \frac{p_{b1}}{p_{b2}} - \log \hat{p}(\alpha, \kappa, c_r) \right) = 0$$

We stack moments for the price differentials of the United States with various comparison countries/provinces as $m(\cdot)$ and place them into a two-step generalized method of moments (GMM) estimation routine,

$$C = m(\alpha, \kappa, c_r) W m(\alpha, \kappa, c_r)'$$

Where W is the optimal weighting matrix, computed from a first stage using $W = I$. Standard errors are conducted with a bootstrap procedure, resampling over pharmaceutical observations. We bound $\kappa \geq 1$ to be consistent with the Nash bargaining and competition in Equation 10.⁵⁴

7.2 Entry Cost Estimation

Entry cost estimation is straightforward by comparison. We require one additional piece of data, the number of purchased doses. For Australia and the UK we have the entire universe of purchases made through their respective public health systems.⁵⁵ However we do not know the total market size by drug in the US, only the market size of Medicare and Medicaid. We compensate by adjusting purchases with aggregate market data.⁵⁶ While not perfect, as Medicare, Medicaid, and private insurers serve different patients suffering from different maladies; it is a first order approximation of the total US market size.

From Equation 12, we require the function $\mu_{s_n}(N)$, $\mu_g(N)$, and $s_g(N)$ for the US market. The first is the market share-weighted markup for the existing brand-name drug, the second is the markup for generic suppliers, and the last the aggregate market share of generic suppliers. For all three objects, we non-parametrically recover these function, using available data. Thus we compute:

$$\begin{aligned} \mu_{s_n} &= \sum_s \beta_s^{\mu_{s_n}} \mathbb{I}(S = s) + \epsilon \\ \mu_g &= \sum_s \beta_s^{\mu_g} \mathbb{I}(S = s) + \epsilon \\ s_g &= \sum_s \beta_s^{s_g} \mathbb{I}(S = s) + \epsilon \end{aligned}$$

We then construct counterfactual profits for an marginal generic entrant following Equation 13. By comparing figure between the United States and both Australia and the UK, we compute

⁵⁴Results without this constraint are highlighted in the Appendix.

⁵⁵We not have sales for all EU countries, our estimate will measure the marginal cost of entering just the English market.

⁵⁶We use the Medical Expenditure Panel Survey (MEPS) Prescribed Medicines File for this. In 2014, Medicaid paid for 11% of doses, Medicare paid for 33% of doses and other sources paid for 56% of doses. Other includes all-cash, private insurance, and specialized programs such as the US Defense Department's Tricare and through Veterans Affairs.

$\Pi_{excess}(S_{US}^*)$. Standard errors are conducted with a bootstrap procedure, resampling over pharmaceutical observations.

8 Results & Counterfactuals

8.1 Competition and Buyer Leverage Results

Table 7 shows the estimates for the two parameters of the model, α and κ as well as the average residual \hat{c}_{12} . We estimate the parameters using four different versions of the data for the US: Medicaid with the market as molecule-dose-route; Medicaid with the market at the molecule-level; Medicare Part D, which is only available at the molecule level; and NADAC data at the molecule-dose-route level. There is a leverage parameter (κ) for each country and a cost difference for the US compared with each country ($\hat{c}_{US,F}$).⁵⁷ The competition parameter α is the rate at which the weight between the monopoly and competitive price ($\theta_b(S)$) declines as the number of competitors increases. A larger value indicates a speedier approach to competition. Medicare Part D prices approach the competitive price more quickly than Medicaid prices. This is not surprising as Part D plans are privately operated. The buyer leverage (κ) parameters are increasing in the outside option \bar{p} and the buyer bargaining parameter w_b , and decreasing in the seller bargaining parameter w_s . A lower κ is interpreted as the country having greater bargaining power or a better outside option than the base country.⁵⁸ The average cost difference captures the unexplained variation in the data. The results are similar across the versions of the data. The UK and New Zealand have the largest average differences with the US.⁵⁹

8.2 Excess Entry

Estimation proceeds in two stages. First, for the US, we estimate the functions μ_{s_n} , μ_g and s_g in Figure 5. Second we then estimate Equation 13 in Table 8.

Figure 5 shows trends concerning the interplay between branded and unbranded generics in US Medicare data. First the branded market-share weighted markup function μ_{s_n} quickly decreases with the first non-branded generic entrant. Similarly, the average markup of generic suppliers μ_g falls to nearly marginal cost pricing, while the market share of the non-branded suppliers nears 100%.

A reason for high US generic drug prices could be higher implied costs of entry. Although the entry requirements across countries are broadly similar, differences in how long or cumbersome the

⁵⁷Since we assume that competition functions the same in each country for identification there is just one α parameter.

⁵⁸ κ often is at the constraint where $\kappa = 1$. If κ was unbounded, this would mean governments have the ability to force firms to sell with $\gamma < 1$, under the estimated marginal cost. While this may occur, we leave this discussion to the appendix.

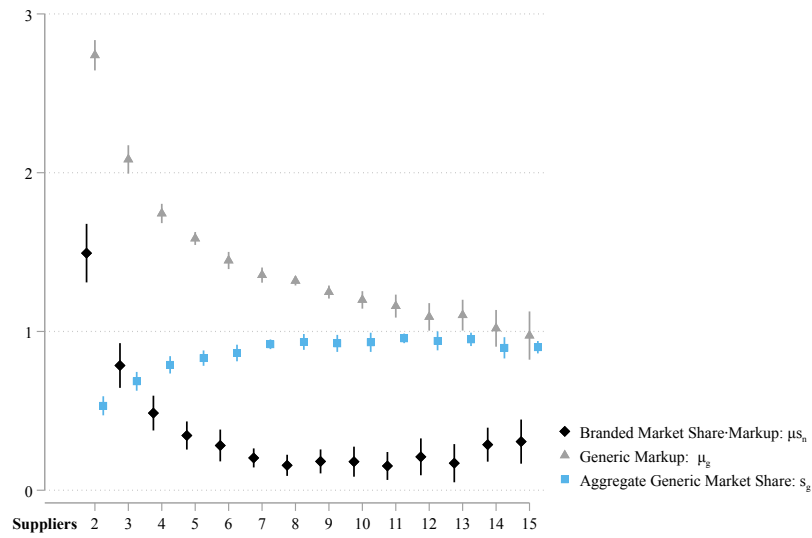
⁵⁹Comparing prices with the UK/NZ and Australia confirm this trend, as the number of entrants increases, UK/NZ prices fall faster than Australian prices.

Table 7: Joint Estimation Table

	Medicaid Molecule-Dose	Medicaid(d) Molecule	Medicare(d) Molecule	NADAC Molecule
Competition α	-1.18 (0.08)	-1.68 (0.23)	-1.43 (0.12)	-1.25 (0.19)
Leverage κ US	5.50 (0.39)	4.50 (0.32)	6.38 (0.75)	5.87 (0.68)
Leverage κ AU	1.00 (0.00)	1.00 (0.00)	1.02 (0.11)	1.00 (0.00)
Leverage κ BC	1.00 (0.00)	1.00 (0.00)	1.00 (0.00)	1.00 (0.00)
Leverage κ NZ	1.73 (0.12)	1.20 (0.10)	1.18 (0.13)	1.02 (0.08)
Leverage κ ON	1.09 (0.18)	1.00 (0.01)	1.00 (0.14)	1.11 (0.21)
Leverage κ UK	1.66 (0.18)	1.67 (0.17)	1.75 (0.24)	1.81 (0.24)
Cost Diff US/AU	1.09 (0.09)	1.78 (0.12)	1.63 (0.11)	0.99 (0.13)
Cost Diff US/BC	0.63 (0.05)	1.02 (0.05)	0.96 (0.10)	0.50 (0.06)
Cost Diff US/NZ	1.16 (0.09)	1.58 (0.10)	1.48 (0.11)	0.79 (0.09)
Cost Diff US/ON	0.85 (0.08)	1.36 (0.13)	1.29 (0.16)	0.77 (0.10)
Cost Diff US/UK	2.02 (0.13)	2.67 (0.16)	2.39 (0.18)	1.38 (0.15)
N	7654	4743	2820	2810

Notes: This table uses a GMM estimator to estimate three types of parameters: (1) the estimated intensity of competition, (2) the joint outside option/bargaining parameter, and (3) mean logarithmic cost differences. Competition intensity α is estimated with a single parameter. The leverage parameter κ is increasing in the outside option \bar{p} and the buyer bargaining parameter w_b , and decreasing in the seller bargaining parameter w_s . Results for column 1 are for Medicaid data aggregated to the molecule-dose-route level. Results for column 2 are for Medicare Part D data aggregated to the molecule-level. Results for column 2 are for Medicare Part D price data combine with US supplier data from Medicaid suppliers. United Kingdom supplier data based on 2017 supplier data, not considering possible cross-importation policies. Bootstrapped standard errors reported in brackets.

Figure 5: Branded versus Generic Markup and Share Functions using US Medicare



Notes: The x-axis depicts the number of suppliers in markets with both branded and generic competitors. There are three trends here. First, the triangles show that generic markups fall as the number of suppliers increase, effectively reaching marginal cost pricing at 15 suppliers. Second, the squares show that generic market share, compared to the branded off-patent market share, increases as the number of suppliers increase. Lastly, the diamonds show that the effect of the branded off-patent drug on average markups falls as the number of total suppliers increases. Robust standard errors with 95% CI.

Table 8: Imputed Excess Entry Cost Estimates

scenario/est (\$M)	Medicaid Molecule-Dose	Medicaid(d) Molecule	Medicare(d) Molecule	NADAC Molecule
AU	13.68 (0.42)	10.78 (0.45)	9.54 (1.20)	5.65 (1.26)
UK	7.94 (0.38)	11.13 (0.41)	8.00 (1.12)	7.11 (1.14)

Notes: This table takes the estimates and data from the GMM estimation and computes “excess entry costs”, which are necessary to justify the fixed costs of entry between the United States and a comparison nation. Bootstrapped standard errors reported in brackets.

application process is can affect the entry cost. For example, in the years leading up to 2012 resource constraints at the FDA resulted in a large backlog of applications for generic approval (Berndt et al., 2018). This has since largely been addressed through the introduction of application fees (GAO, 2017), but the issues of collusion and pay-to-stay-out remain (Wall Street Journal 2016) (Kaiser Health News 2019). To examine differences in entry costs we compute the implied cost of a supplier entering the US relative to Australia and the United Kingdom.⁶⁰ Under various specifications, there must be excess costs of \$6-\$14 million per year to justify the limited number of US entrants in the marketplace.⁶¹ Using estimates for Medicare and Medicaid costs must be between \$8-\$14 million per year (shown in Columns (1)-(3) in Table 8),. Under estimates using NADAC prices and Medicaid quantities, which do not include retailer markups, we find estimates between \$6-\$7 million per year (shown in Column (4) in Table 8).

8.3 Counterfactuals

Our counterfactuals revolve around two major concepts. The first involves regulatory changes on entry and the second involves structural changes in the nature of bargaining and buyer power. Both sets of policies have attracted substantial political support. We explore various implementations and combinations of these policies. In Table 9 we show the aggregate affect of these policies on total generic and off-patent drug expenditures. In Table 10 we show the effect on the average drug. In general, the aggregate effects of these policies are smaller than the average effect, as popular drugs with millions of doses face small or no changes in counterfactual prices, operating close to perfect competition.

Counterfactual 1: Single Market The first counterfactual simulates a reciprocity policy, effectively eliminating non-tariff barriers within the rich English-speaking world. It supposes that once a

⁶⁰These costs represent entering the England, not the entire European Union, and should be interpreted as an upper bound for the entire European market.

⁶¹Entry is a largely one-time cost to pay for regulatory compliance. As we use annual data, the results reflect the discounted cost of entry. If we assume an annual discount rate of 10-20%, all these figures should be multiplied by between 5 and 10.

Table 9: Percentage Change in **Total** Drug Expenditures

scenario/est (%)	Medicaid Molecule-Dose	Medicaid(d) Molecule	Medicare(d) Molecule	NADAC/Medicaid Molecule
CF 1: Single Market	-7.8 (0.8)	-3.2 (0.5)	-2.9 (0.4)	-3.8 (0.8)
CF 2: Foreign Negotiation	-18.3 (4.2)	-10.4 (3.1)	-12.3 (5.5)	-20.8 (7.6)
CF 3: Both	-18.6 (4.2)	-10.6 (3.2)	-12.4 (5.5)	-21.0 (7.6)
CF 4: Free Entry	-15.1 (1.7)	-6.3 (0.8)	-8.1 (1.3)	-5.0 (0.5)
N	7654	4743	2820	2810

Notes: Each statistic represent the total cost savings across all off-patent drugs in a particular counterfactual using various data sources. Bootstrapped standard errors reported in brackets. This table takes the estimates and data from the GMM estimation and computes the policy relevant counterfactuals. The columns represent different estimates using different data sources. The rows represent the different counterfactuals. See text for further details.

Table 10: Percentage Change in **Average** Drug Costs

scenario/est (%)	Medicaid Molecule-Dose	Medicaid(d) Molecule	Medicare(d) Molecule	NADAC/Medicaid Molecule
CF 1: Single Market	-10.6 (0.4)	-5.2 (0.3)	-5.0 (0.4)	-6.1 (0.4)
CF 2: Foreign Negotiation	-33.3 (3.7)	-27.2 (3.5)	-29.2 (5.7)	-39.1 (6.7)
CF 3: Both	-33.8 (3.6)	-27.5 (3.5)	-29.4 (5.7)	-39.4 (6.7)
CF 4: Free Entry	-31.8 (1.9)	-19.7 (1.3)	-24.7 (2.6)	-15.8 (0.7)
N	7654	4743	2820	2810

Notes: Each statistic represent the average cost savings across all off-patent drugs in a particular counterfactual using various data sources. Bootstrapped standard errors reported in brackets. This table takes the estimates and data from the GMM estimation and computes the policy relevant counterfactuals. The columns represent different estimates using different data sources. The rows represent the different counterfactuals. See text for further details.

supplier was approved in Australia, Canada, New Zealand or the UK then it can sell its product in the US without any further entry costs. We simulate this policy by increasing the number of entrants in each market to the maximum unique entrants across all 4 foreign markets. Mathematically, we consider the change in the US price to be:

$$\frac{p_{CF1}}{p_{US}} = \frac{\exp(\alpha \log S_{World}) [\kappa_{US} - 1] + 1}{\exp(\alpha \log S_{US}) [\kappa_{US} - 1] + 1}.$$

To be conservative, S_{World} is the maximum number of suppliers approved to sell that drug in any other country.

The first row of Table 9 shows the percentage change in total drug expenditure for Medicaid using Medicaid price data, Medicare Part D using Medicaid price data, and Medicare using NADAC price data. The first and fourth columns uses the molecule-dose-route market definition, while columns (2) and (3) use the molecule-only definition. A single market would reduce total Medicaid expenditure by 3-8%, Medicare Part D by 3%, and Medicaid-weighted NADAC prices by 4%.

Intuitively, results comes from increasing the number of competitors S_{US} and considering the effect of competition through the parameter α on pharmaceutical prices. While these figures seem small, they are pulled towards zero by a set of extremely popular drugs with many competitors. Table 10 replicates Table 9, but considers the change in the average drug price. In general, we find larger cost savings.

Counterfactual 2: Strong US Buyer Leverage (Bargaining/Outside Option) This counterfactual estimates the savings to Medicaid, Medicare Part D, and across all buyers with NADAC data by using improving buyer leverage to bargain drug prices. Results are simply derived from the observed price change when κ_{US} is changed to the average of κ in all other countries. It is implemented by considering the average cost savings if the US has the same buyer leverage as competitor nations:

$$\frac{p_{CF2}}{p_{US}} = \frac{\exp(\alpha \log S_{US}) [\bar{\kappa}_f - 1] + 1}{\exp(\alpha \log S_{US}) [\kappa_{US} - 1] + 1},$$

Where κ_f is the average buyer leverage over all foreign markets. Since the US is a much larger market, this is likely a lower bound. The results in the second row of Table 9 show the savings estimates in Medicaid are between 10-18%, Medicare Part D by 12%, and Medicaid-weighted NADAC prices by 21%. When considering cost saving for the average drug, savings are much larger, ranging from 27-40%.

Counterfactual 3: Both single market and US Buyer Leverage (Bargaining) Next we estimate the effect of both increasing competition through reciprocity and the US improving their buyer leverage. This is simulated assuming that all potential foreign entrants enter and that the US

buyer leverage is changed to the average of all other countries:

$$\frac{p_{CF3}}{p_{US}} = \frac{\exp(\alpha \log S_{World}) [\bar{\kappa}_f - 1] + 1}{\exp(\alpha \log S_{US}) [\kappa_{US} - 1] + 1}.$$

The results, in row three of Table 9, show savings similar to those from just implementing the bargaining policy, reducing Medicaid expenditure by 10-19%, Medicare Part D by 12%, and Medicaid-weighted NADAC prices by 21%. When considering cost saving for the average drug, savings are much larger, ranging from 28-40%.

Counterfactual 4: Elimination of Excess Entry Costs We show in Section 8.2 that the US has higher implied entry costs than Australia and the UK. A reciprocity policy would inherently lower the cost of entry. Therefore our estimates of the number of new entrants to the US in Counterfactual 1, which keeps the fixed cost of entry constant, may underestimate the effect. In this counterfactual we show the effect of lowering the fixed entry costs (the excessive profits) to foreign levels:

$$\frac{p_{CF4}}{p_{US}} = \frac{\exp(\alpha \log S_{Entry}) [\kappa_{US} - 1] + 1}{\exp(\alpha \log S_{US}) [\kappa_{US} - 1] + 1},$$

Where S_{Entry} is the maximum number of profitable entrants if US entry costs for generic suppliers were the maximum of either Australia or the United Kingdom.

This provides an estimate of the maximum number of entrants that would be likely to enter following a liberalization of drug policy. To be as conservative as possible, we consider buyer-leverage to be computed as the maximum from comparison destinations.⁶² We find gains larger than those from under a single market, and comparable to better bargaining by the US government. The estimates show a reduction in Medicaid expenditure of 8-15%, Medicare Part D by 6%, and Medicaid-weighted NADAC prices by 5%. This occurs as the massive market US, dwarfing the size of Australia and the United Kingdom, would allow for significant freer entry. We consider these estimates a hard upper bound on the gains from regulatory barriers, as it is possible that $S_{US}^* > S_{World}$. We do not know the total number of suppliers operating outside the English speaking world. A marginal new global supplier must not only cover the average cost of manufacturing and distribution in a particular country (which we account for), but also set up a new production facilities.

9 Discussion & Conclusion

We study generic and off-patent drug marketplaces across five nations, focusing on a broad set of pharmaceuticals sold in either tablet or capsule form. This paper empirically shows that the average price in the US is 200-400% higher than the average in four countries where the government

⁶²If we use the minimum, we effectively get the same result as perfect competition. We find that Australian and British Columbia effectively have no markups. Thus, we would have an infinite number of entrants in the US counterfactual. In addition, we convert the dollar figures from Table 8 to percentages when calculating counterfactuals.

intervenes in generic drug prices — Australia, the UK, Canada and New Zealand — when there is only one supplier in the US market. This price differential declines as there are more suppliers in the market. Once there are six or more suppliers, there is little difference in the price.

We use a structural model of the generic drug market to compare the relative effectiveness of two policy solutions: international reciprocity and government bargaining. A well-implemented bargaining policy is far more effective than a reciprocity policy, with savings to Medicaid of as much as 18% and to Medicare Part D of 12% compared with savings of 8% and 3% respectively. Better bargaining can also reduce the harms from collusion between suppliers, as buyers can band together and make “take-it-or-leave-it” offers near marginal cost.

Pharmaceutical manufacturing is a global business. Other than regulatory barriers, pharmaceuticals face low effective tariff and non-tariff barriers. Hence when a new supplier decides to set up manufacturing, the decision is not a purely local decision, it is a global decision, taking to account the fixed costs of production, number of competitors and each market’s regulatory conditions. If pharmaceutical entry is liberalized in the United States, it could stimulate additional entry into production and manufacturing. Accordingly we consider how this affects our results using two counterfactual exercises. Counterfactual #1 presents an lower bound. Here we assume that all supplier already approved can now costlessly enter the US market. Counterfactual #4 provides an upper bound by liberalizing entry. Lowering entry costs in one country might lead to completely new suppliers entering the market due to the global nature of the supplier’s entry decision.

Underlying our conclusions are some important assumptions. We are implicitly assuming that the US government has the state capacity to effectively bargain with pharmaceutical suppliers.⁶³ Mistakes in bargaining and political entanglements could lead to prices below free-market levels and market exit. In contrast, a free-market approach may be easier to implement with less political interference, and has little to no downside. In addition, a market-based policy like that currently being pursued by the FDA through the 2019 Drug Action Plan, could have additional benefits that we have not captured in our model. First and foremost increased competitors could reduce the risk of shortages caused by plant specific quality failures. Furthermore, an important weakness of our approach is that we are not able to decompose the components underlying fixed costs of entry; can they be directly addressed by the FDA or are they the result of complex downstream bargaining between PBMs, suppliers, and pharmacies. These are things that we do not observe and take as given in our approach.

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⁶³See Carril and Duggan (2018) for an example with respect to defense procurement and market concentration.

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Online Appendix

A Data Details

A.1 United States Data

Each market contains a set of similar pharmaceuticals with an associated NDC code. The manufacturer can be identified from the first 5 digits of the NDC code. A list of approved molecules are obtained from the drugs@FDA database. We include molecules for which the form of administration is listed as “capsule” or “tablet”. The application numbers are matched with the 2009 FDA Orange Book. The orange book lists all unexpired patents and exclusivities. A molecule is classified as off-patent if there are no exclusivities listed in the Orange Book or if there is ANDA approval.

There are several potential sources of price data. For the primary analysis we use data from Medicaid, however, our results are robust to data from Medicare Part D and private insurance data. The data on reimbursement and quantities comes from the State Utilization Data:

Since the start of the Medicaid Drug Rebate Program, states report drug utilization for covered outpatient drugs paid for by state Medicaid agencies. In order for pharmaceutical manufacturers to obtain Medicaid coverage for their prescription drugs, they must sign a rebate agreement with the Secretary of Health and Human Services (HHS) to provide rebates for those drugs when purchased by Medicaid. The reported data includes two fields: Medicaid Amount Reimbursed and Non-Medicaid Amount Reimbursed. The sum of these two fields should generally equal the Total Amount Reimbursed on an NDC by NDC basis; Rebates are paid on both brand name and generic drugs. Total amount reimbursed by both Medicaid and non-Medicaid entities to pharmacies or other providers for the 11-digit NDC drug in the period covered (two previous fields added together). This total is not reduced or affected by Medicaid rebates paid to the state. This amount represents both federal and state reimbursement and is inclusive of dispensing fees.⁶⁴

A.1.1 FDA Approvals and Drug Codes

Historical National Drug Code (NDC) and approval data is acquired from the National Bureau of Economic Research (NBER). Each drug for sale package includes data on the manufacturer, active ingredients, dosage, form, and packaging details. This data additionally includes details regarding the firm that applied for approval, the date of approval, and the current status of the drug. We stack this data across time to merge to the Medicare, Medicaid, and NADAC data. We retain data on all approved drugs that appear in either tablet or capsule form.

⁶⁴From <https://www.medicaid.gov/medicaid/prescription-drugs/state-drug-utilization-data/state-drug-utilization-data-faq/index.html>

We additionally make use of the NBER’s NDC crosswalk that concords two different version of drug codes. An NDC is a composite of three variables: the drug code, the manufacturer, and the package’s details. One version reports all three variables separately and another concatenates the variables in a semi-arbitrary fashion. We convert all data to the former form.

A.1.2 State Medicaid Data

State-level Medicare data is sourced from the Centers for Medicaid and Medicare Services (CMS) and is published at the monthly level.⁶⁵ We collapse this data by year, molecule, dose and form, harmonizing names across years. This data is originally collected by individual states and reported to the CMS.

With this aggregated data, we perform three procedures to align it better with global data. First, the CMS reports that “Amounts reimbursed by Medicaid are pre-rebate, not net of rebates.” The 1993 Medicare and Healthcare Act instituted a system of system of manufacturer rebates of the net price. For multiple source drugs, the statutory reimbursement rate is 13% of the average manufacturers price. For single-source, non-innovator drugs, the statutory reimbursement rate is calculated with a similar formula based on the average manufacturer. We directly account for these differences.

Second, the CMS reports that “This amount represents both federal and state reimbursement and is inclusive of dispensing fees.” As all other countries separately report dispensing fees, we further account for this. In 2007, the median dispensing fee was approximately \$5 and in 2017 the median dispensing fee was approximately \$10. We code this in.

Finally, as part of the Affordable Care Act (ACA), US states instituted Federal Upper Limits (FUL), the maximum allowable price. We consider any prices above these FUL in ACA years to be computed in error and thus exclude them from our analysis.

A.1.3 Federal Medicare Part D Data

Medicare Part D plans are regulated prescription drug plans that cover 75% of the United States Medicare population (Cubanski et al., 2018). We use national-level aggregates from the Centers for Medicaid and Medicare Services (CMS) that are constructed from the complete universe of Part D patients. CMS maintains a Chronic Condition Data Warehouse (CCW), “a database with 100% of Medicare enrollment and final-action Part D prescription drug event (PDE) data.”

Publicly available information does not report on individual dose sizes and forms. Rather, this data is reported at only the molecule level. We aggregate this data by year in order to perform analysis. In addition, there are two further caveats: the molecule level data is inclusive of manufacturer rebates and dispensing fees. This makes it harder to align with foreign data, leaving this as a robustness check with our mainline results.

⁶⁵We leave the separate state-by-state analysis to further work and consolidate the data to the national level.

A.1.4 National Average Drug Acquisition Cost (NADAC) Data

For a subset of drugs, we additionally append price data with National Average Drug Acquisition Cost (NADAC) data. This data is collected by the CMS to help aid states in setting Medicaid reimbursement costs. Prior to 2011, many states used the average wholesale price (AWP) as a benchmark. However, the Office of Inspector General found that that the AWP methodology was “fundamentally flawed”. It was replaced by the NADAC, which reflects the average drug acquisition costs for retail, consumer-facing pharmacies from a nationally representative sample. While sampling variation is not revealed by the CMS, they reveal that the NADAC sampling average margin of error is below 2.5%.

This data is collected by a federal subcontractor and released weekly and reflects the price paid by the retailer to the pharmaceutical distributor. This data does not include the dispensing fee, any upstream rebates, nor the retailer markup. It includes all transactions paid using cash, private insurance (including Medicare Part D plans), and public insurance.⁶⁶

Data is released at the National Drug Code level, and includes both selected over-the-counter medications and prescription pharmaceuticals. We aggregate this data to a year-molecule-dose-form using national Medicare market shares for all prescription pharmaceuticals that appear in either a tablet or capsule form.

A.2 Canadian Data

A.2.1 Canadian Manufactures Data

Data on approved Canadian pharmaceutical providers is obtained from Health Canada, which maintains their Drug Product Database (DPD). This database includes all “human, veterinary, disinfectant, and radiopharmaceutical products approved for use by Health Canada”. The database includes all approved, marketed, cancelled, and dormant products. We consider all pharmaceutical products that are produced in a tablet or capsule form. Then we consider the start year of Canadian availability, i.e. the year the drug was first approved or marketed. The last day of Canadian availability is the first year when a drug is flagged as “cancelled” or made “dormant”. Since active ingredients in the DPD have slight variations with corresponding US and provincial drug names, we perform manual cleaning to align drug names. We create a manufacturing database with the (1) list of active ingredients, (2) year, and (3) number of approved companies.

A.2.2 British Columbia Price Data

Data on British Columbia pharmaceutical prices is obtained from PharmaCare, which has eight plans that cover all patients under the Medical Services Plan of B.C. (MSP), the statewide sing-

⁶⁶Patients on 340b plans are excluded.

payer health insurance plan for all province residents that are either Canadian citizens or legal permanent residents. This plan is targeted to achieve universal health care.

PharmaCare covers both pharmacy dispensing fees as well the maximum price it will recognize for each drug in the British Columbia PharmaCare formulary. The PharmaCare plan operates on an income-based subsidy scheme, where lower income participants are reimbursed for a proportion of all pharmaceutical costs. The provincial government maintains an online database for the last three years of prices, including data on discontinued drugs. We consider only the unsubsidized price, which is the maximum that can be charged. Any subsidy simply offsets this cost.⁶⁷

For drugs in this formulary, we consider all doses markets as either a tablet or capsule. For each active ingredient we extract out the dosage of each active ingredient. As active ingredients in PharmaCare have slight variations with corresponding US and other provincial drug names, we perform manual cleaning to align drug names. We create a provincial database with the (1) list of active ingredients, (2) form, (3) dosages, (4) year, and (5) average price. We additionally create a second database that averages drug prices across forms and dosages for comparability to US Medicare Part D data. Data is converted from Canadian Dollars to United States Dollars using exchange rate data from the Board of Governors of the Federal Reserve System access through the FRED database.

A.2.3 Ontario Price Data

Data on Ontario pharmaceutical prices is obtained from the six main Ontario Public Drug Programs that cover 43% of all provincial spending on prescription drugs. This plan has coverage primarily targeted at the poor, disabled, and elderly. This is roughly comparable to the population covered by the US Medicaid and Medicare programs. These plans cover the cost of drugs in monthly published formularies.

We use the 42nd Edition edition formulary for 2017, considering all drugs classified as either a tablet or capsule. We then extract out both the active ingredients and respective dosages. As active ingredients in Ontario's Public Drug Formulary have slight variations with corresponding US and other provincial drug names, we perform manual cleaning to align drug names. We create a provincial database with the (1) list of active ingredients, (2) form, (3) dosages, (4) year, and (5) average price. We additionally create a second database that averages drug prices across forms and dosages for comparability to US Medicare Part D data. Data is converted from Canadian Dollars to United States Dollars using exchange rate data from the Board of Governors of the Federal Reserve System access through the FRED database.

⁶⁷As with all non-US sources, the final dispensing fees are not included.

A.3 Australian Data

A.3.1 Australian Price and Manufacturer Data

Australian data comes from the Australian Government’s Department of Health, which runs the national and universal Pharmaceutical Benefits Scheme (PBS). It maintains a monthly database of drug prices both inclusive and exclusive of the government subsidy. We take this data for January of each year and concord molecule names, routes, and dosages across time. These prices include both the price to the manufacturer and the price to the pharmacy.⁶⁸ In contrast to databases maintained by some other nations, this data also includes names on all of the approved manufacturers.

For drugs in this formulary, we consider all doses markets classified as either a tablet or capsule. We then extract out the dosage of each active ingredient. As active ingredients in the PBS have slight variations with corresponding US drug names, we perform manual cleaning to align drug names. We create a database with the (1) list of active ingredients, (2) form, (3) dosages, (4) year, (5) average price across packages. We additionally create a second database that averages drug prices across forms and dosages for comparability to US Medicare Part D data. Data is converted from Australian Dollars to United States Dollars using exchange rate data from the Board of Governors of the Federal Reserve System access through the FRED database.

A.3.2 Australian Prescribed Doses Data

Until 2015, the Drug Utilisation Sub-Committee (DUSC) of the Pharmaceutical Benefits Advisory Committee published an annual report on all prescription drugs by dose, form and molecule. This annual database, the Australian Statistics on Medicines (ASM), is sourced from all the underlying reimbursed dosages from the PBS. This includes both the number of prescriptions written and the total expenditure at a molecule, dose, and route level.

The ASM data is only is currently available through 2015. For data on 2016 and 2017, we turn to a slightly different database, the Section 85 PBS, RPBS Section 85 Date-of-Processing and Date-of-Supply data. This data is updated monthly and includes the number of prescriptions written and the total expenditure at a molecule, dose, and route level.

A.4 United Kingdom

A.4.1 Manufacturer Data

We obtain manufacturer data for the entirety of the United Kingdom from the electronic Medicines Compendium (eMC). This resource is run by the non-profit Datapharm, which coordinates data between the UK Medicines and Healthcare products Regulatory Agency (MHRA) or the European Medicines Agency (EMA), and drug companies.

⁶⁸As with all non-US sources, the final dispensing fees are not included.

Since this data is not available for bulk download, we web scraped the website in 2017 for data on nearly 2,000 active molecules available for sale. We create a database that includes all products that are available in either tablet or capsule form. As we do not have time variation, we use this database for all years of United Kingdom Data.

A.4.2 National Health Service

English National Health Service price data is not directly available online from NHS Business Services Authority (NHSBSA). We repurpose data acquired from a freedom of information request from the Centre for Evidence-Based Medicine (EBM Datalab) within the Nuffield Department of Primary Care Health Sciences at University of Oxford. EBM Datalab cleaned the messy and inconsistent data into a usable format. As active ingredients in the NHS data have slight variations with corresponding US and other provincial drug names, we perform manual cleaning to align drug names. We collapse the data down to year-molecule-dose-form with average prices.⁶⁹ We create a similar, second database with data collapsed to the year-molecule form. Data is converted from British Pounds to United States Dollars using exchange rate data from the Board of Governors of the Federal Reserve System access through the FRED database.

A.4.3 UK Prescribed Doses Data

While the formulary data from the UK does not include the number of prescribed doses, a separate database is maintained by the NHS Business Services Authority for England. This data is classified as the Prescription Cost Analysis (PCA) data and collected monthly. We collect all data and collapse by year to get the total number of prescribed doses by “Pharmacy & Appliance Contractor”, which refer to local pharmacies. Data is converted from British Pounds to United States Dollars using exchange rate data from the Board of Governors of the Federal Reserve System access through the FRED database.

A.5 New Zealand

A.5.1 Manufacturers

New Zealand manufacturer data for 2006-2018 is obtained from Medicines and Medical Devices Safety Authority, which operates under the auspices of the Ministry of Health. From their online product search, we obtain a list of all pharmaceutical products, with data on the active ingredient, manufacturer, status, approval date, and notification date (for drugs with lapsed or not available approvals). As this data does not exist in an easily readable format, we use a web-scraping program to download the relevant data. We first download a list of all drugs by their first letter and then iterate through all the listed molecules.

⁶⁹As with all non-US sources, the final dispensing fees are not included.

New Zealand additionally collects information, not only on the marketer of the drug, but also details on the original manufacturer, product tester, and packaging plant. We leave the analysis of this data for future work.

A.5.2 Prices

Prices for consumers in New Zealand are obtained from the Pharmaceutical Management Agency, a government agency referred to as Pharmac. It maintains a monthly database of drug prices both inclusive and exclusive of the government subsidy. We take this data for January of each year and concord molecule names, routes, and dosages across time. In New Zealand, consumers are also required to pay a \$5 per prescription co-pay. We add this price back into the cost of the drug.⁷⁰

For drugs in this formulary, we consider all doses sold as either a tablet or capsule. We then extract out the dosage of each active ingredient. As active ingredients in Pharmac have slight variations with corresponding US drug names, we perform manual cleaning to align drug names. We create a database with the (1) list of active ingredients, (2) form, (3) dosages, (4) year, and (5) average price across packages. We additionally create a second database that averages drug prices across forms and dosages for comparability to US Medicare Part D data.

Data is converted from New Zealand Dollars to United States Dollars using exchange rate data from the Board of Governors of the Federal Reserve System access through the FRED database.

A.6 Pharmaceutical Tariffs

One possible source of pricing difference would be differential tariff regimes. However, in our sample, tariffs are zero or negligible. As of 2017, Australia, Canada, the European Union, New Zealand, and the United States all apply zero or near zero tariff rates for all pharmaceutical imports. In particular the European Union does not apply any tariffs whatsoever on all pharmaceutical supplies. Canada, New Zealand, and the United States applies small tariffs on surgical and medical devices, which are not included in our sample. Australia additionally places a 2.5% tariff on all immunological products. These products are primarily vaccines that are largely injected, but there are a few generic vaccines available in pill or capsule dosages, such as Ty21a for typhoid. (UNCTAD, 2018; WTO, 2018)

B Competition Examples

We highlight three plausible micro-foundations for $\theta(S)$: Bertrand competition with a homogenous good, Bertrand competition with discrete choice, and Bertrand competition as a repeated game (which can be made equivalent to variants of Cournot competition).

⁷⁰This fee is subsidized by the government for special categorizes of patients, such as those under the age of 13, and those with over 20 prescriptions.

⁷¹As with all non-US sources, the final dispensing fees are not included.

Table A.1: Summary of Data Sources

Data Type	Data Type	Observation Level
US FDA Manufacturer Data (via National Bureau of Economic Research)	Manufacturers	Molecule-Dose-Form and Molecule
US Medicaid Data (Centers for Medicaid and Medicare Services)	Prices, Quantities	Molecule-Dose-Form and Molecule
US Medicare Part D (Centers for Medicaid and Medicare Services)	Prices, Quantities	Molecule
US National Average Drug Acquisition Cost (NADAC) Data	Prices	Molecule-Dose-Form and Molecule
British Columbia PharmaCare formulary	Prices	Molecule-Dose-Form and Molecule
Ontario Public Drug Programs	Prices	Molecule-Dose-Form and Molecule
Health Canada Drug Product Database (DPD)	Manufacturers	Molecule-Dose-Form and Molecule
UK National Health Service (England) - Centre for Evidence-Based Medicine Datalab (via Oxford)	Prices	Molecule-Dose-Form and Molecule
UK National Health Service (England) - Prescription Cost Analysis	Prices, Quantities	Molecule-Dose-Form and Molecule
UK Datapharm	Manufacturers	Molecule-Dose-Form and Molecule
Australia Pharmaceutical Benefits Scheme	Prices, Manufacturers	Molecule-Dose-Form and Molecule
Australia Australian Statistics on Medicines/Section 85	Quantity	Molecule-Dose-Form and Molecule
New Zealand Pharmac	Prices	Molecule-Dose-Form and Molecule
New Zealand Medicines and Medical Devices Safety Authority	Manufacturers	Molecule
US Federal Reserve	Exchange Rates	

Notes: See text for full details.

Table A.2: Example Data - British Columbia - 10 Largest Price Differences - 2016

Molecule	Dose	Form	FDA Approval	US Suppliers	Medicaid Price	BC Price	Medicaid Prescriptions
pyrimethamine	25	tablet	1953	1	605.51	1.43	981
mebendazole	100	tablet	1996	1	312.69	5.91	5970
penicillamine	250	capsule	1970	1	224.24	3.89	732
penicillamine	250	tablet	1970	1	90.04	0.68	205
procarbazine	50	capsule	1985	1	57.06	0.44	799
morphine sulfate	200	capsule	1987	1	54.05	1.19	287
methoxsalen	10	capsule	1954	2	49.72	0.65	226
oxymetholone	50	tablet	1972	1	35.24	1.77	58
hydromorphone	32	tablet	1926	1	37.20	11.49	1269
ethacrynic acid	25	tablet	1967	2	18.46	0.97	2408

Bertrand Competition: Homogenous Good This case nests Bertrand competition with homogenous goods, where:⁷²

$$\theta(S) = \begin{cases} 1 & S = 1 \\ 0 & S \geq 2. \end{cases}$$

While this case seems to be rejected in the data, as there is not a stark difference between markets with one and two suppliers, but rather a smooth continuum; this provides a canonical baseline for our other possibilities.

Bertrand Competition: Discrete Choice Analogously, if pharmacists make a discrete choice between S identical suppliers on the basis of price, and buyers have no market power, then:⁷³

$$\theta(S) = \frac{1}{\alpha \left(1 - \frac{1}{S}\right)} \times \frac{1}{p_m - p_c}$$

Where the logit discrete choice probability of choosing supplier s conditional on prices \vec{p} is (following [McFadden et al. \(1973\)](#)):

$$Pr(s; S) = \frac{\alpha p_s}{\sum_{s' \in S} \alpha p_{s'}}.$$

The parameter α governs the dispersion/sensitivity to price for the downstream pharmacist and is derived from the variance of a standard Gumbel random variable.

Bertrand Competition: Repeated Game Drug makers (both large developers and generic suppliers) tend to be long lived, exist in highly regulated environments, have high barriers to entry, and utilize extremely long and entrenched supply chains. Furthermore, there is extensive information sharing on both prices and quantities through third party organizations that report data with extremely high frequency ([IHS Markit, 2018](#)).⁷⁴ All of these characteristics ([Ivaldi et al., 2003](#)) make the drug supply amenable to tacit collusion, especially when the number of competitors is relatively low. Monitoring rivals is straightforward and threats are credible. We do not take an explicit stance on how this occurs, other than that we can model it through our function $\theta(S)$. The ability to maintain a collusive monopoly-like price is a function of the number of current market participants.

Extension to multiple buyers While all our foreign markets and US Medicaid have a single buyer representing a particular consumer, US Medicare Part D and other US private insurance markets have multiple buyers. These sales are run through many firms, each competing for customers on price. This changes the Nash problem; downstream pharmaceutical insurance firm profits directly

⁷²We cannot allow Cournot competition, since in our setup, the output quantity is exogenously set at Q_b as we do not directly model the decision of a prescriber. In Appendix X, we allow for an arbitrary demand elasticity and show how this can nest Cournot competition in that case.

⁷³This is obtained since $p = \frac{1}{\alpha \left(1 - \frac{1}{S}\right)} + c$ ([Berry, 1994](#)).

⁷⁴An extensive economics and medical literature uses various slices of this data, however it is prohibitively expensive to acquire the entire dataset for analysis.

enter into both the payoff and outside option (as opposed to cost minimization). However, we are unable to identify the profits of individual Medicare Part D Insurers. Rather, we continue modeling choices as with a monopolist buyer and consider different bargaining weights as well as different outside options.

C Additional Empirical Results

C.1 Medicare Data Results

A summary of the Medicaid data using medians, binned by the number of suppliers is available in Table 6 and Table A.3.

C.2 Using Molecule-Level Data

Medicare Part D data is not publicly released to the public by molecule-dose-form. Rather, data is released by molecule, averaging across all doses. This in principle should not hinder our analysis, however the US FDA regulates drugs at the molecule-dose-form level, which is how we define markets. We replicate our analysis from before using this average Medicare Part D data. As shown in Figure 9, international price comparisons at the molecule level are extremely similar between Medicare Part D and Medicaid. While the number of observations is reduced, as different dosages are aggregated, trends are statistically similar.

C.3 Medicaid Result Robustness

Table A.4 conducts robustness for Medicaid results. The first panel reports the baseline results. The second panel considers 'un-adjusted' data, where Medicaid data is not adjusted for manufacturer rebates and pharmacy dispensing fees. The last panel considers the same regressions, but with molecule-level fixed effects.

C.4 Weighted Regression Results

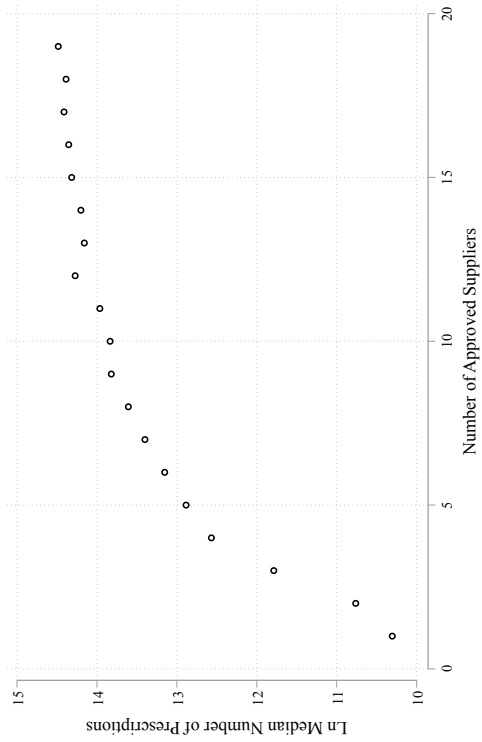
In Table A.5, we recreate Table 3 using weights for the number of doses sold. For NADAC data, we simply use the number of Medicaid doses sold.

C.5 With ATC Code Fixed Effects

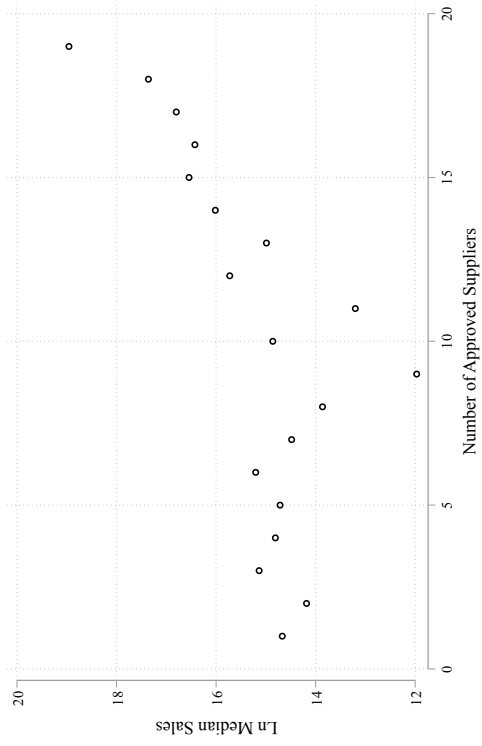
In Table A.6, we recreate 3 with controls for the 3-digit ATC code, to control for differences in outside options between different drug types.

Figure 6: Medicaid Summary - Molecule - Dose - Form level data

(a) Suppliers and the median number of prescriptions



(b) Suppliers and the median sales revenue



(c) Median sales revenue and median prescriptions, binned by the number of suppliers

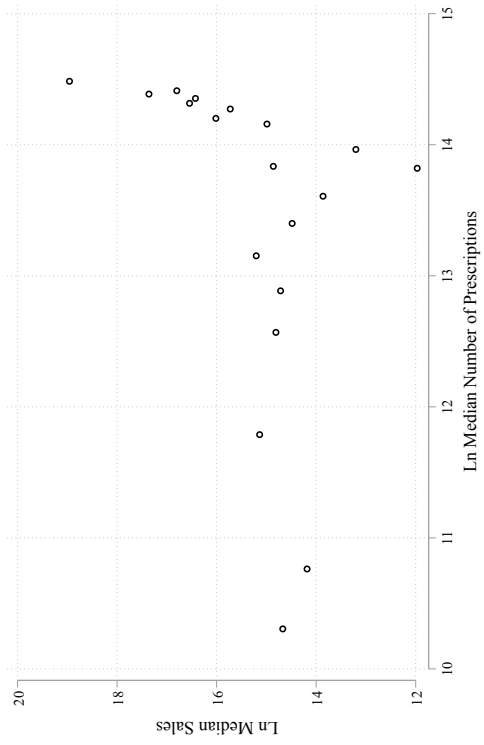


Table A.3: Medicaid Data Summary

Revenue Data:

	Revenues	Observations
1 Suppliers	6.908	2,938
2 Suppliers	2.729	1,891
3-5 Suppliers	8.419	2,749
6-10 Suppliers	3.418	1,515
11+ Suppliers	1.968	262
Total	23.441	9,355

Yearly Data:

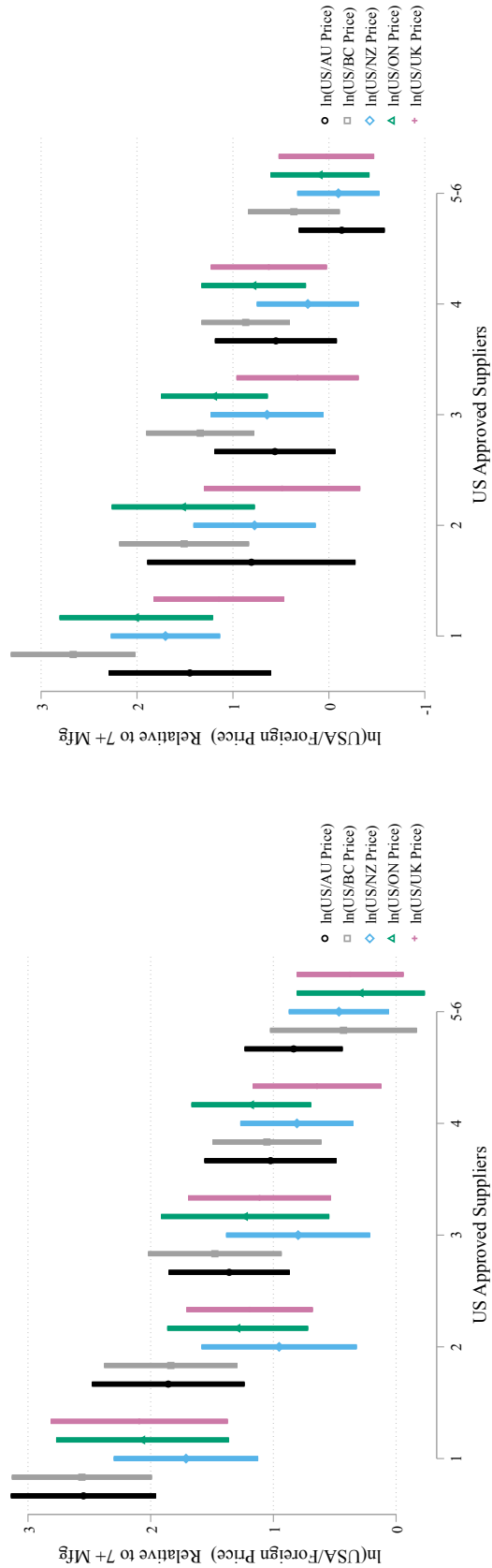
	Sales	Observations
2008	1.355	708
2009	1.496	746
2010	2.090	816
2011	2.194	789
2012	2.101	858
2013	1.880	916
2014	2.375	973
2015	2.880	1,044
2016	3.542	1,190
2017	3.527	1,315
Total	23.441	9,355

Price Jump Data:

	Price Jumps	Observations
2008	0	708
2009	20	746
2010	48	816
2011	52	789
2012	96	858
2013	76	916
2014	160	973
2015	125	1,044
2016	76	1,190
2017	57	1,315
Total	710	9,355

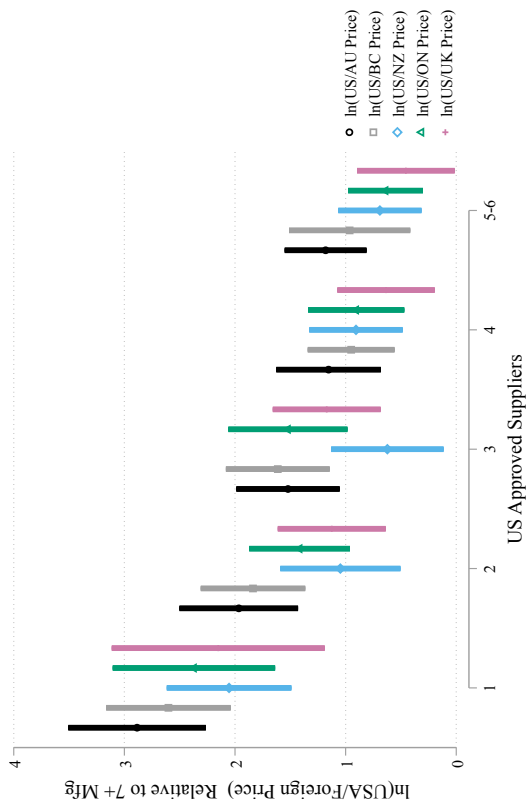
Notes: Based on authors' calculations of public Medicaid data.

Figure 7: Role of competition - Non-parametric Regressions - Relative to Baseline of 7+ Suppliers



(b) Medicare Data - Molecule

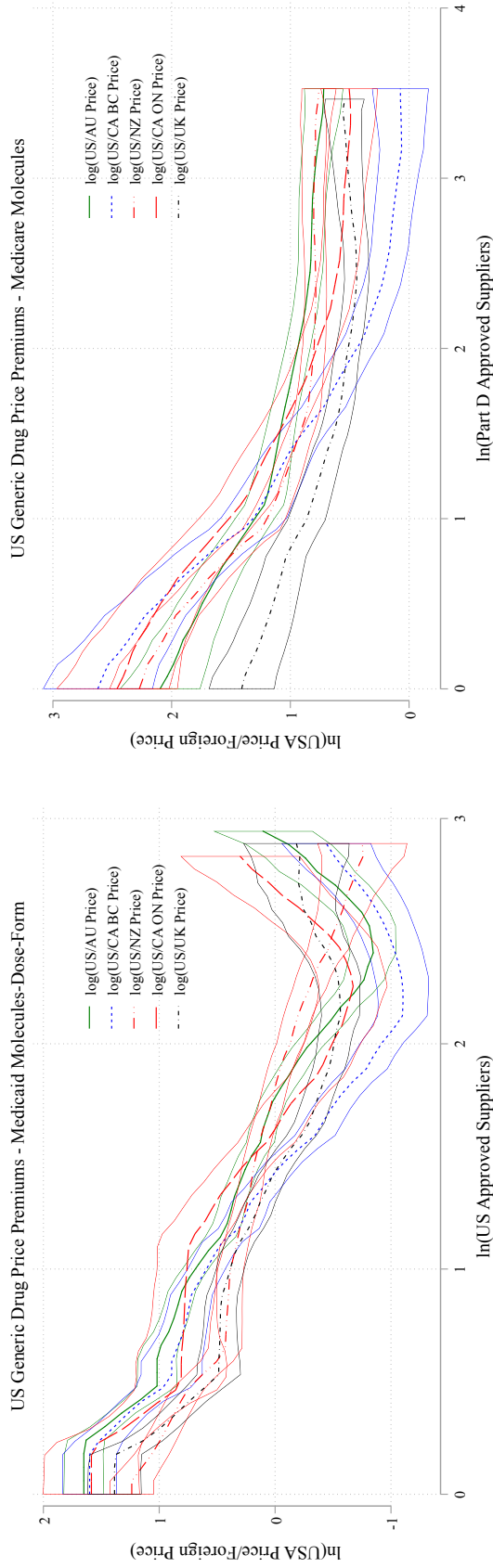
(a) Medicaid Data - Molecule-Dose-Form



(c) NADAC Data - Molecule-Dose-Form

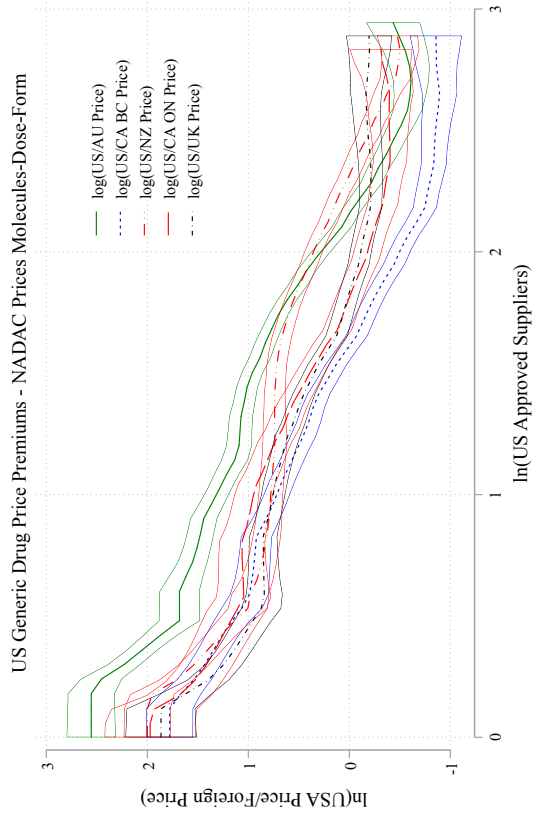
Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B) with the number of FDA approved competitors of the US. Regression specification from Equation 3 with standard errors clustered at the molecule level. 95% confidence intervals displayed. See text for data sources and details. Year-destination fixed effects are removed by de-meaning.

Figure 8: Role Of Competition



(a) Medicaid Data - Molecule-Dose-Form

(b) Medicare Data - Molecule

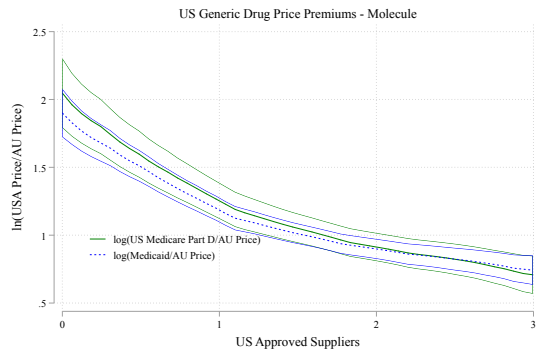


(c) NADAC Data - Molecule-Dose-Form

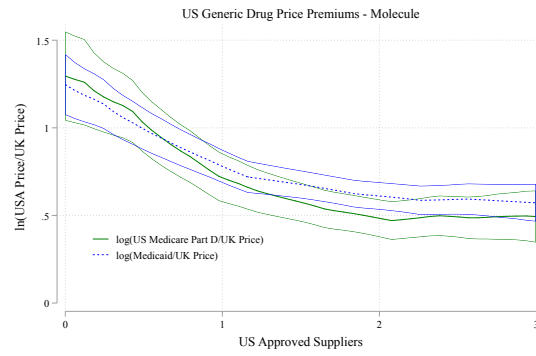
Notes: Average price difference between the two countries taken across drugs (molecule-dose-route) within the market size of the US.

Figure 9: Medicare and Medicaid International Price Differences

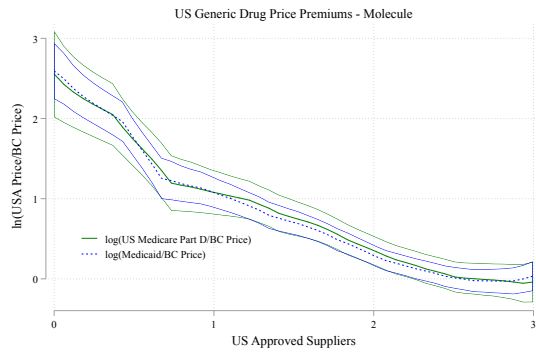
(a) Australia



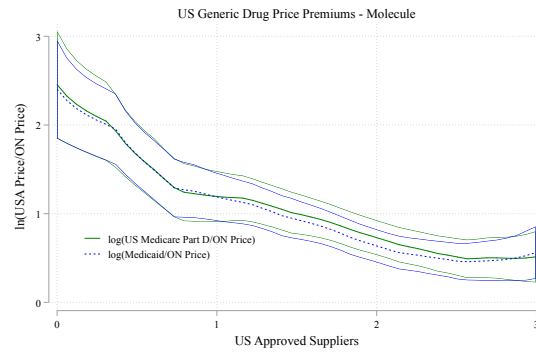
(b) United Kingdom



(c) Canada - British Columbia



(d) Canada - Ontario



(e) New Zealand

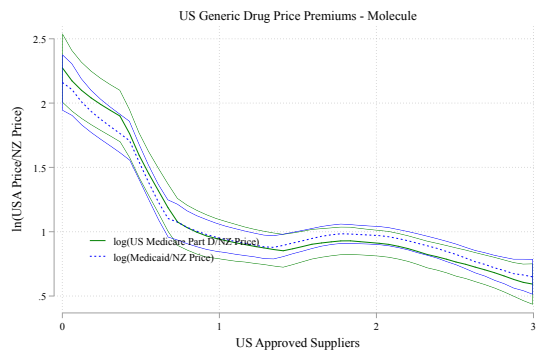


Table A.4: Relationship between Price Premiums and Number of US Suppliers - Medicaid Robustness

Medicaid - Molecule-Dose-Form - Baseline

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.105*** (0.110)	-1.142*** (0.0898)	-0.736*** (0.104)	-0.875*** (0.106)	-0.909*** (0.127)
Adj. R-Square	0.281	0.257	0.134	0.162	0.129
Observations	1706	858	1470	344	1625
Fixed Effects	year	year	year	year	year

Medicaid - Molecule-Dose-Form - Medicaid Prices Unadjusted for Manufacturer Rebates and Dispensing fees

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-0.754*** (0.0799)	-0.869*** (0.0618)	-0.380*** (0.0681)	-0.629*** (0.0692)	-0.450*** (0.0902)
Adj. R-Square	0.275	0.267	0.102	0.164	0.0680
Observations	2172	1173	1979	488	2171
Fixed Effects	year	year	year	year	year

Medicaid - Molecule-Dose-Form - Baseline - Molecule Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-0.973*** (0.127)	-1.051*** (0.126)	-0.631*** (0.167)	-0.828*** (0.242)	-1.281*** (0.122)
Adj. R-Square	0.689	0.824	0.635	0.776	0.677
Observations	1706	858	1470	344	1625
Fixed Effects	year,drug	year,drug	year,drug	year,drug	year,drug

Notes: Average price difference between the two countries taken across drugs. Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details. The sample size changes in the second panel, as the data adjustment procedure leads to a subset of observations with negative prices. It is possible that some states net out dispensing fees, even though the CMS data claims otherwise.

Table A.5: Relationship between Price Premiums and Number of US Suppliers - Weighted Regressions

Medicaid - Molecule-Dose-Form - Weighed by Doses

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.105*** (0.110)	-1.142*** (0.0898)	-0.736*** (0.104)	-0.875*** (0.106)	-0.909*** (0.127)
Adj. R-Square	0.281	0.257	0.134	0.162	0.129
Observations	1706	858	1470	344	1625
Fixed Effects	year	year	year	year	year

Medicare Part D - Molecule - Weighed by Doses

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(Part D Suppliers)	-0.481*** (0.124)	-0.884*** (0.0930)	-0.494*** (0.0891)	-0.695*** (0.102)	-0.325** (0.106)
Adj. R-Square	0.133	0.339	0.157	0.229	0.0477
Observations	764	288	803	237	852
Fixed Effects	year	year	year	year	year

NADAC - Molecule-Dose-Form - Weighed by Medicaid Doses

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.252*** (0.0927)	-1.133*** (0.0709)	-0.819*** (0.0831)	-0.992*** (0.0849)	-0.855*** (0.113)
Adj. R-Square	0.407	0.310	0.234	0.248	0.149
Observations	1250	1099	1180	463	1467
Fixed Effects	year	year	year	year	year

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

Table A.6: Relationship between Price Premiums and Number of US Suppliers - 3 Digit ATC Code Fixed Effects

Medicaid - Molecule-Dose-Form - Year and ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.087*** (0.100)	-1.068*** (0.0941)	-0.719*** (0.105)	-0.858*** (0.116)	-0.887*** (0.133)
Adj. R-Square	0.438	0.392	0.282	0.321	0.263
Observations	1703	858	1468	344	1624
Fixed Effects	year,ATC3	year,ATC3	year,ATC3	year,ATC3	year,ATC3

Medicare Part D - Molecule - Year and ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(Part D Suppliers)	-0.540*** (0.115)	-0.816*** (0.103)	-0.475*** (0.0874)	-0.701*** (0.128)	-0.253** (0.0953)
Adj. R-Square	0.410	0.413	0.366	0.308	0.260
Observations	764	288	803	237	852
Fixed Effects	year ATC3	year ATC3	year ATC3	year ATC3	year ATC3

NADAC - Molecule-Dose-Form - Year and ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.249*** (0.0862)	-1.052*** (0.0710)	-0.806*** (0.0864)	-0.962*** (0.0840)	-0.909*** (0.123)
Adj. R-Square	0.565	0.471	0.374	0.406	0.299
Observations	1250	1099	1180	463	1467
Fixed Effects	year,ATC3	year,ATC3	year,ATC3	year,ATC3	year,ATC3

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

Table A.7: Relationship between Price Premiums and Number of US Suppliers - 3 Digit ATC Code-Year Fixed Effects

Medicaid - Molecule-Dose-Form - Year-ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.030*** (0.115)	-1.067*** (0.0976)	-0.661*** (0.118)	-0.858*** (0.116)	-0.892*** (0.154)
Adj. R-Square	0.433	0.366	0.238	0.321	0.210
Observations	1703	858	1468	344	1624
Fixed Effects	year-ATC3	year-ATC3	year-ATC3	year-ATC3	year-ATC3

Medicare Part D - Molecule - Year-ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(Part D Suppliers)	-0.536*** (0.130)	-0.828*** (0.105)	-0.469*** (0.0957)	-0.701*** (0.128)	-0.244* (0.107)
Adj. R-Square	0.276	0.396	0.249	0.308	0.110
Observations	764	288	803	237	852
Fixed Effects	year-ATC3	year-ATC3	year-ATC3	year-ATC3	year-ATC3

NADAC - Molecule-Dose-Form - Year-ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.265*** (0.0935)	-1.063*** (0.0724)	-0.824*** (0.0946)	-0.962*** (0.0840)	-0.931*** (0.131)
Adj. R-Square	0.529	0.451	0.311	0.406	0.246
Observations	1250	1099	1180	463	1467
Fixed Effects	year-ATC3	year-ATC3	year-ATC3	year-ATC3	year-ATC3

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

C.6 With ATC Code-Year Fixed Effects

In Table A.7, we recreate Table 3 with controls for the 3-digit ATC code-year combinations, to control for differences in time-varying outside options between different drug types. Table A.8, we recreate Table 4 including foreign entry.

C.7 Allowing for Lagged Number of Suppliers

In Table A.9, we recreate Table 3 with controls for the 3-digit ATC code-year combinations, to control for differences in time-varying outside options between different drug types.

Table A.8: Relationship between Price Premiums and Number of US and Foreign Suppliers - 3 Digit ATC Code-Year Fixed Effects

Medicaid - Molecule-Dose-Form - Year-ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-0.885*** (0.134)	-0.924*** (0.124)	-0.644*** (0.143)	-0.729*** (0.138)	-0.943*** (0.190)
ln(Foreign Suppliers)	-0.285* (0.120)	-0.351** (0.119)	-0.0451 (0.165)	-0.221 (0.127)	0.0299 (0.208)
Adj. R-Square	0.443	0.378	0.199	0.325	0.261
Observations	1703	756	1406	335	273
Fixed Effects	year-ATC3	year-ATC3	year-ATC3	year-ATC3	year-ATC3

Medicare Part D - Molecule - Year-ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(Part D Suppliers)	-0.508*** (0.149)	-0.855*** (0.185)	-0.501*** (0.119)	-0.727*** (0.174)	-0.225 (0.171)
ln(Foreign Suppliers)	-0.0565 (0.129)	0.00729 (0.144)	0.117 (0.153)	0.0300 (0.132)	0.127 (0.194)
Adj. R-Square	0.276	0.382	0.200	0.280	0.0474
Observations	764	245	770	230	192
Fixed Effects	year-ATC3	year-ATC3	year-ATC3	year-ATC3	year-ATC3

NADAC - Molecule-Dose-Form - Year-ATC 3-digit Fixed Effects

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.150*** (0.111)	-0.982*** (0.0866)	-0.774*** (0.112)	-0.924*** (0.101)	-1.081*** (0.176)
ln(Foreign Suppliers)	-0.184* (0.0878)	-0.180* (0.0765)	-0.0918 (0.130)	-0.0682 (0.0870)	0.208 (0.159)
Adj. R-Square	0.536	0.482	0.312	0.414	0.263
Observations	1250	1023	1153	455	358
Fixed Effects	year-ATC3	year-ATC3	year-ATC3	year-ATC3	year-ATC3

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

Table A.9: Relationship between Price Premiums and Number of US Suppliers - Lagged Number of Suppliers

Medicaid - Molecule-Dose-Form - Lagged Number of Suppliers-

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
L.ln(US Suppliers)	-1.170*** (0.128)	-1.201*** (0.105)	-0.787*** (0.116)	-1.030*** (0.119)	-1.042*** (0.143)
Adj. R-Square	0.293	0.257	0.144	0.207	0.151
Observations	1321	748	1263	295	1388
Fixed Effects	year	year	year	year	year

Medicare Part D - Molecule - Lagged Number of Suppliers

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
L.ln(Part D Suppliers)	-0.488*** (0.132)	-0.897*** (0.0963)	-0.508*** (0.0932)	-0.696*** (0.109)	-0.339** (0.113)
Adj. R-Square	0.117	0.348	0.157	0.231	0.0468
Observations	580	262	610	215	647
Fixed Effects	year	year	year	year	year

NADAC - Molecule-Dose-Form - Lagged Number of Suppliers

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
L.ln(US Suppliers)	-1.277*** (0.0983)	-1.203*** (0.0767)	-0.839*** (0.0830)	-1.112*** (0.0858)	-0.915*** (0.118)
Adj. R-Square	0.440	0.341	0.254	0.324	0.173
Observations	1078	972	1024	408	1266
Fixed Effects	year	year	year	year	year

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

Table A.10: Relationship between Price Premiums and Number of US Suppliers

Medicaid - Molecule-Dose-Form - Number of Approved Sellers Instrumented by number of doses sold

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(US Suppliers)	-1.402*** (0.168)	-1.466*** (0.138)	-1.330*** (0.155)	-1.152*** (0.168)	-1.593*** (0.170)
Adj. R-Square	0.263	0.237	0.0550	0.145	0.0595
Observations	1706	858	1470	344	1625
Fixed Effects	year	year	year	year	year
First Stage F	285.2	704.1	330.7	357.0	298.0

Medicaid- Molecule- Number of Approved Sellers Instrumented by number of doses sold

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
ln(Part D Suppliers)	-0.483** (0.150)	-1.004*** (0.128)	-0.514*** (0.109)	-0.587*** (0.121)	-0.366* (0.148)
Adj. R-Square	0.126	0.261	0.146	0.188	0.0464
Observations	755	285	793	233	842
Fixed Effects	year	year	year	year	year
First Stage F	427.9	619.1	465.4	496.5	508.5

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

C.8 Instrumenting number of entrants with number of patients.

We may worry that the number of suppliers is correlated with a shock that differentially affects marginal cost across different destinations. To control for that possibility, we assume that the number of patients is exogenous. See Table A.10.

C.9 Non-parametric Results

In Table A.11, we recreate Table 3 dummy variables for the number of competitors.

C.10 Market Size Facts

Next we estimate the elasticity of demand to the number of suppliers in the US. These results are shown in Table A.12. Column (1) is the elasticity without any fixed effects, its shows that a one percent increase in the number of suppliers results is correlated with a 0.75% decrease in the price. Columns (2)-(4) include different combinations of fixed effects (year, year and drug, year and drug-dose).

Table A.11: Relationship between Price Premiums and Number of US Suppliers - Broken Out

Medicaid - Molecule-Dose-Form

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
2	-0.690** (0.228)	-0.726** (0.238)	-0.758** (0.260)	-0.776** (0.278)	-0.900** (0.295)
3-5	-1.422*** (0.211)	-1.429*** (0.197)	-1.028*** (0.234)	-1.145*** (0.255)	-1.304*** (0.297)
6-10	-2.246*** (0.233)	-2.698*** (0.216)	-1.445*** (0.236)	-2.138*** (0.254)	-2.027*** (0.312)
11+	-2.509*** (0.399)	-2.215*** (0.260)	-2.226*** (0.322)	-1.528*** (0.333)	-1.864*** (0.369)
Adj. R-Square	0.268	0.271	0.130	0.167	0.129
Observations	1706	858	1470	344	1625
Fixed Effects	year	year	year	year	year

Medicare Part D - Molecule

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
2	-0.641 (0.676)	-1.154* (0.460)	-0.928* (0.409)	-0.489 (0.540)	-0.659 (0.512)
3-5	-1.009* (0.433)	-1.653*** (0.354)	-1.376*** (0.309)	-1.154** (0.431)	-0.636 (0.364)
6-10	-1.320** (0.436)	-2.493*** (0.348)	-1.718*** (0.300)	-1.863*** (0.414)	-1.303*** (0.357)
11+	-1.573*** (0.440)	-2.773*** (0.332)	-1.733*** (0.298)	-2.138*** (0.410)	-1.117** (0.356)
Adj. R-Square	0.138	0.366	0.193	0.251	0.0787
Observations	764	288	803	237	852
Fixed Effects	year	year	year	year	year

NADAC - Molecule-Dose-Form

	(1)	(2)	(3)	(4)	(5)
	$\ln(p_{US}/p_{AU})$	$\ln(p_{US}/p_{BC})$	$\ln(p_{US}/p_{NZ})$	$\ln(p_{US}/p_{ON})$	$\ln(p_{US}/p_{UK})$
2	-0.918*** (0.255)	-0.766*** (0.216)	-1.006*** (0.263)	-0.955*** (0.285)	-1.026* (0.406)
3-5	-1.534*** (0.242)	-1.325*** (0.185)	-1.289*** (0.230)	-1.315*** (0.262)	-1.281** (0.388)
6-10	-2.471*** (0.241)	-2.446*** (0.182)	-1.748*** (0.225)	-2.227*** (0.255)	-2.102*** (0.390)
11+	-3.352*** (0.277)	-2.644*** (0.229)	-2.619*** (0.245)	-2.340*** (0.307)	-2.165*** (0.396)
Adj. R-Square	0.394	0.314	0.251	0.231	0.158
Observations	1250	1099	1180	463	1467
Fixed Effects	year	year	year	year	year

Notes: Average price difference between the two countries taken across drugs (molecule-dose-route for panel A and C and molecule for Panel B). Regression specification from Equation 4 with standard errors clustered at the molecule level. See text for data sources and details.

Table A.12: Supplier Elasticity

Medicaid - Molecule-Dose-Form

	(1)	(2)	(3)	(4)
	$\ln(\text{price}_{US})$	$\ln(\text{price}_{US})$	$\ln(\text{price}_{US})$	$\ln(\text{price}_{US})$
$\ln(\text{manufacturers}_{US})$	-1.073*** (0.0584)	-1.122*** (0.0569)	-1.142*** (0.0613)	-0.320*** (0.0624)
Observations	7878	7878	7878	7878
R^2	0.166	0.223	0.665	0.889
FE		Y	Y,Drug	Y,Drug-Dose

Standard errors in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$ **NADAC - Molecule-Dose-Form**

	(1)	(2)	(3)	(4)
	$\ln(\text{price}_{US})$	$\ln(\text{price}_{US})$	$\ln(\text{price}_{US})$	$\ln(\text{price}_{US})$
$\ln(\text{manufacturers}_{US})$	-1.119*** (0.0540)	-1.148*** (0.0537)	-1.015*** (0.0712)	-0.170** (0.0535)
Observations	4807	4807	4807	4807
R^2	0.262	0.280	0.776	0.976
FE		Y	Y,Drug	Y,Drug-Dose

Standard errors in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$ **Australia - Molecule-Dose-Form**

	(1)	(2)	(3)	(4)
	$\ln(\text{price}_{AU})$	$\ln(\text{price}_{AU})$	$\ln(\text{price}_{AU})$	$\ln(\text{price}_{AU})$
$\ln(\text{manufacturers}_{AU})$	-0.250*** (0.0723)	-0.263*** (0.0726)	0.0191 (0.0800)	0.00639 (0.0812)
Observations	2172	2172	2172	2172
R^2	0.028	0.033	0.871	0.967
FE		Y	Y,Drug	Y,Drug-Dose

Standard errors in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

C.11 Unconstrained optimization

To be consistent with Nash bargaining, the competition parameter must be greater than 1. In the main text, estimation follows this restriction. In Table A.13 and A.14, this restriction is loosened. Results do slightly violate the bounding condition, however entry cost estimates and US leverage parameters are largely unchanged.

D Pair-wise Country Results

In Table A.15, we estimate pairwise results between the US and one other destination for Medicaid. In general, our results echo the baseline results.⁷⁵

E Monte Carlo Results

In Table A.16, we run a Monte Carlo simulation of our estimation procedure for two countries. One country has bargaining leverage $\kappa_1 = 5$ and the other has $\kappa_2 = 1$. The competition parameter α equals -1, illustrating a high level of competition. We additionally add variation in marginal costs between the countries, drawn from a normal distribution and simulate 500 drugs over 1000 simulations. We summarize the results in Table A.16. We find that the Monte Carlo estimates largely mirror the actual parameters.

F Multiple Heterogenous Sellers

F.0.1 Monopolist buyer and oligopolist sellers

In the bulk of our considered markets, all non-US markets and the US Medicaid market, there are multiple drug sellers, but only one buyer. For simplicity and due to data limitations, we assume that all S drug sellers are ex-ante identical. A drug seller maximizes profits $\pi_s = (p_s - c) m_s(\vec{p}) q$, where $m_s(\vec{p})$ is the market share of seller s conditional on the vector of prices \vec{p} . A buyer minimizes costs again, where $C_b = \sum p_s m_s(\vec{p}) q$. The outside option of the seller is to exit the market and make zero profits and the outside option of the buyer is to pay the equilibrium price in a market with $S - 1$ sellers. This outside price $\bar{p}_{s,S-1}$ is defined recursively, starting from Equation 7.

As before, the price is determined by the Nash Surplus:

$$NS = (\pi_s)^{w_s} (\bar{p}_b q - C_b)^{w_b}.$$

⁷⁵We do not estimate cost differences for simplicity.

Table A.13: Joint Estimation Table - Unconstrained Estimation

	Medicaid Molecule-Dose	Medicaid(d) Molecule	Medicare(d) Molecule	NADAC Molecule
Competition α	-1.12 (0.09)	-1.64 (0.25)	-1.41 (0.13)	-1.12 (0.20)
Leverage κ US	5.14 (0.37)	3.98 (0.24)	6.18 (0.78)	5.26 (0.55)
Leverage κ AU	0.85 (0.04)	0.77 (0.06)	1.01 (0.13)	0.75 (0.06)
Leverage κ BC	0.57 (0.07)	0.51 (0.07)	0.67 (0.13)	0.63 (0.11)
Leverage κ NZ	1.68 (0.11)	1.15 (0.08)	1.17 (0.12)	0.98 (0.15)
Leverage κ ON	1.04 (0.22)	0.74 (0.20)	0.94 (0.20)	1.05 (0.16)
Leverage κ UK	1.64 (0.18)	1.61 (0.17)	1.73 (0.24)	1.75 (0.24)
Cost Diff US/AU	1.01 (0.09)	1.60 (0.13)	1.64 (0.15)	0.85 (0.14)
Cost Diff US/BC	0.55 (0.05)	0.88 (0.06)	0.86 (0.10)	0.42 (0.07)
Cost Diff US/NZ	1.17 (0.09)	1.60 (0.10)	1.48 (0.10)	0.77 (0.11)
Cost Diff US/ON	0.84 (0.08)	1.29 (0.13)	1.28 (0.15)	0.74 (0.12)
Cost Diff US/UK	2.03 (0.14)	2.72 (0.16)	2.40 (0.18)	1.35 (0.17)
N	7654	4743	2820	2810

Notes: This table uses a GMM estimator to estimate three types of parameters: (1) the estimated intensity of competition, (2) the joint outside option/bargaining parameter, and (3) mean logarithmic cost differences. Competition intensity α is estimated with a single parameter. The leverage parameter κ is increasing in the outside option \bar{p} and the buyer bargaining parameter w_b , and decreasing in the seller bargaining parameter w_s . Results for column 1 are for Medicaid data aggregated to the molecule-dose-route level. Results for column 2 are for Medicare Part D data aggregated to the molecule-level. Results for column 2 are for Medicare Part D price data combine with US supplier data from Medicaid suppliers. United Kingdom supplier data based on 2017 supplier data, not considering possible cross-importation policies. Bootstrapped standard errors reported in brackets.

Table A.14: Imputed Excess Entry Cost Estimates - Unconstrained Estimation

scenario/est (\$M)	Medicaid Molecule-Dose	Medicaid(d) Molecule	Medicare(d) Molecule	NADAC Molecule
AU	13.64 (0.47)	10.42 (0.35)	9.35 (1.12)	6.15 (1.48)
UK	7.86 (0.40)	10.70 (0.27)	7.80 (0.91)	7.27 (1.26)

Notes: This table takes the estimates and data from the GMM estimation and computes “excess entry costs”, which are necessary to justify the fixed costs of entry between the United States and a comparison nation. Bootstrapped standard errors reported in brackets.

Table A.15: Separate Estimation Table - Unconstrained Estimation

	US/AU	US/BC	US/NZ	US/ON	US/UK
Leverage κ US	5.72 (0.58)	4.87 (0.53)	4.39 (0.43)	5.75 (0.90)	9.85 (0.89)
Leverage κ Foreign	0.91 (0.07)	0.78 (0.11)	1.45 (0.13)	1.25 (0.33)	1.76 (0.20)
Competition α	-1.14 (0.07)	-2.31 (0.17)	-0.74 (0.08)	-1.47 (0.16)	-0.66 (0.05)

Notes: This table uses a GMM estimator to estimate two types of parameters: (1) the estimated intensity of competition, (2) the joint outside option/bargaining parameter. Competition intensity α is estimated with a single parameter. The bargaining parameter κ is increasing in the outside option \bar{p} and the buyer bargaining parameter w_b , and decreasing in the seller bargaining parameter w_s . Results are for Medicaid data aggregated to the molecule-dose-route level. United Kingdom supplier data based on 2017 supplier data, not considering possible cross-importation policies. Bootstrapped standard errors reported in brackets.

Table A.16: Monte Carlo Results

	Actual Value	Estimated Mean	Standard Deviation
κ_1	5.00	5.05	0.30
κ_2	1.00	1.02	0.03
α	-1.00	-1.00	0.08
c_{12}	2.00	2.00	0.13

Notes: This table presents Monte Carlo simulation results with 500 drugs, with the number of suppliers ranging from 1 to 20, and the marginal cost difference between the two countries drawn from a normal distribution with a mean of 2 and a standard deviation of 1/4. We run 1000 simulations.

We assume the same bargaining weights w as before. The major concern here is the determination of the function $m_s(\vec{p})$. Many consider the competition in drug marketplaces to be Bertrand, thus if $p_s > p_{s'}$, then $m_s = 0$. This would result in $p_s = c$ for $S > 1$. However, this seems contrary to our descriptive results, where the US price premium is strictly monotonically decreasing in sellers S and not constant for $S > 1$. Instead we offer two other options, a discrete choice micro-foundation, where pharmacists make allocations based on price or Bertrand competition with capacity constraints.⁷⁶

Taking the logarithm of the Nash Surplus, we obtain:

$$\log NS = w_s \log((p_s - c) m_s(\vec{p}) q) + w_b \log\left(\bar{p}_{N-1} q - \sum_{s \in S} p m_s(\vec{p}) q\right).$$

For brevity, we focus on the the first micro-foundation. Pharmacists allocate prescriptions between drug sellers according to both price and a random-component. Following [McFadden et al. \(1973\)](#), we assume the random component is distributed as a standard Gumbel random variable and that market shares for firm s are:

$$m_s(\vec{p}) = \frac{\exp(\alpha_b \log p_s)}{\sum_{s' \in S} \exp(\alpha_b \log p_{s'})},$$

where the parameter α_b governs the important of the price in the pharmacists choice and/or the final payers ability to control costs.

Taking first order conditions, we generalize Equation 7, implicitly defining the price:

$$w_b m_s (p_s - c) = w_s (\bar{p}_{N-1} - p_s) \left[\alpha \left(1 - \frac{c}{p_s} \right) (1 - m_s) + 1 \right].$$

When there is only one seller, then $m_s = 1$, mechanically simplifying to Equation 7. When buyers have zero bargaining weights, we have results that are aligned with [Berry \(1994\)](#):

$$p_s = \min \left\{ \bar{p}_{S-1}, c \frac{\alpha (1 - s)}{\alpha (1 - s) + 1} \right\}$$

When sellers have zero bargaining power, $p_s = c$, as before.

While such estimation is appealing, it has a much higher data requirement for full estimation. In particular, the competition function is not only a function of the competitive structure, but also the relative bargaining weights. We currently do not have enough data, nor enough degrees of freedom, to estimate such a function with any confidence.

⁷⁶Alternative explanations include Bertrand competition supported by a repeated game or standard cournot competition where capacity is pre-set.