# Intermediary Profits in a Time of Scarcity 

February 2024


#### Abstract

We study the impact of scarcity on intermediary profits through an analysis of the effects of the 2011 Japanese tsunami on the U.S. used car market. The tsunami devastated Japanese new car production, which we show led to a demand surge for used cars. Wholesale and retail prices on used Japanese cars rose, but dealer profits declined by $\$ 170$ per car, a drop of roughly $4 \%$. Consistent with theory, we find the decline in dealer profits is diminished for transactions where the dealer is relatively more indispensable.


JEL Classification: G23, G51, G33, D83
Keywords: Intermediation, Search, Supply-Chain Shocks

## 1 Introduction

In periods of scarcity, marked by either unexpected surges in demand or reductions in supply, those possessing scarce resources are positioned to profit. This observation is consistent with the predictions that follow from value-based strategy theories (Brandenburger and Stuart Jr, 1996; Gans and Ryall, 2017; Oberholzer-Gee, 2021), and comes as no surprise. Instead of examining how scarcity affects a producer's or asset holder's capacity to capture value, our study shifts its focus to how scarcity affects another key player in the value chain: the intermediary.

Intermediaries play a crucial role in decentralized markets, significantly contributing to both value creation and capture (Brealey et al., 1977; Fernandez-Mateo, 2007). Their economic significance is substantial; estimates suggest that intermediaries account for a quarter (Spulber, 1996) to a third of US Gross Domestic Product (Krakovsky, 2015). The presence of intermediaries in the economy, such as car dealers (Bennett, 2013; Israeli et al., 2022), real estate agents (Giustiziero, 2021; Gubler and Cooper, 2019), pharmaceutical benefit managers (Feng and Maini, 2023), and concert ticket resellers (Bennett et al., 2015) has been a subject of study in the strategy literature.

Intermediaries play a pivotal role in bridging gaps created by market frictions, which otherwise hinder value creating transactions from occurring (Gans and Ryall, 2017; Mahoney and Qian, 2013). Intermediaries create value by ensuring quality and reliability (certification) (Lizzeri, 1999), facilitating connections between buyers and sellers (matchmaking) (Khurana, 2002; Fernandez-Mateo, 2007; Rider and Samila, 2019), and influencing consumer preferences (shaping demand) (Cialdini, 2007; Jansen et al., 2023b). Take, for example, used car dealerships. These dealerships contribute value by acquiring, inspecting, refurbishing, sometimes certifying pre-owned vehicles, and actively engaging potential customers. Car dealers cap-
ture value that they added to the transaction through price discrimination, charging higher prices to customers with relatively high willingness to pay (Jansen et al., 2023a).

A core challenge that intermediaries face is adapting to a rapidly changing market environment (Teece et al., 1997; Argyres et al., 2019). Natural disasters (Oh and Oetzel, 2011), financial crises (Albuquerque and Bronnenberg, 2012), unanticipated policy shifts (Wang et al., 2020), global pandemics (Wenzel et al., 2020), or other macro-economic shocks are examples of instances where scarcity in markets increases and can affect intermediary performance. It is not clear what impact scarcity will have on the performance of such intermediaries, as they serve as both buyers and sellers. One natural intuition is that intermediaries thrive during periods of scarcity. As a recent example, the semiconductor chip shortage from 2020 to 2022 severely restricted the ability of auto manufacturers to produce new vehicles, inducing a scarcity of new cars.

Figure 1 shows that the chip shortage was correlated with substantially higher new car prices, as expected. Figure 1 also shows that the car dealers that act as intermediaries in this market profited magnificently during the period of scarcity. In terms of magnitudes, car dealer margins on new cars increased four times during the immediate post-pandemic period. Prior to the pandemic, shifts in margins were an order of magnitude smaller, suggesting that scarcity is a first order determinant of variation in intermediary performance.

Does the compelling visual evidence in Figure 1 establish that intermediaries generally benefit from scarcity? We think not, for three reasons. First, the recent change in dealer margins may be driven by different buying modes and experiences, shifting technology, changes in market structure, or other aspects of the macroeconomic environment. In other words, too much has changed over the last several years to attribute the increase in dealer profits solely to new car scarcity. Second, the medium-term patterns in Figure 1 from 2016 to

2020 suggest that dealer margins may not be monotonic in new car prices and that other, unobserved factors may be influencing both outcomes. Third, as shown in Figure 2, used car prices rose even more dramatically than new car prices during 2020 and 2021, but used car dealer margins increased much less than new car dealer margins. Clearly, there are market-specific features that determine the impact of scarcity on intermediary profits. Our empirical study of the relationship between intermediary performance and scarcity explores the role of search costs.

To provide theoretical guidance for our inquiry, we consider the predictions arising from economic models of search (Rubinstein and Wolinsky, 1987, 1990). If intermediaries are indispensable to trade (i.e., if buyers and sellers can trade only through them), then an increase in the number of buyers per seller leads to higher profits per unit for intermediaries. In such cases, the higher number of buyers strengthens the negotiating position of the intermediary and allows the intermediary to extract a higher price. Duffie et al. (2005), however, argue that scarcity can lead to a reduction in intermediary profits if intermediaries are not indispensable for trade. In these cases, the higher number of buyers (per available unit of the good) makes it easier for sellers to bypass intermediaries and sell directly to buyers. Recognizing their improved bargaining position, sellers increase the prices they charge intermediaries, which reduces the intermediaries' margins. We refer to this as the bypass mechanism.

We assess these competing theoretical arguments by evaluating the impact of scarcity on dealer profits in the U.S. used car market. The used car market is large and important. Dealers are active in this market, but sellers may also sell directly to buyers. The used car market therefore serves as a useful testing ground for the effects of scarcity. We study these effects by examining the impact of the 2011 Japanese tsunami. The tsunami had a
strong negative short-term impact on the production of new Japanese cars but little effect on non-Japanese manufacturers. This allows us to undertake a difference-in-difference analysis in which we contrast outcomes for Japanese and non-Japanese vehicles in the post-tsunami period.

We perform these tests on a data set that describes approximately 60,000 used car transactions in over 1,100 dealerships across 38 states over the period 2009-2013. The data include retail and wholesale prices and financing details. The purchase transactions were funded by a large automotive indirect-finance company.

We show that the tsunami generated a positive demand shock for used Japanese cars, as the shortage of new vehicles induced prospective buyers of Japanese cars to consider used autos instead. This led to higher wholesale prices for used Japanese cars for a period of roughly six months. We also find that retail prices for these cars increased.

Our central interest is in the impact of scarcity on intermediary profits. We find that the dealer profit per used Japanese car decreased by approximately $\$ 170$ (roughly 4\%) in the post-tsunami period; so, while wholesale and retail prices were rising, dealer profits were falling. Dealer margins relative to wholesale prices on used Japanese cars dropped by 2.1 percentage points (about 6\%) after the tsunami. This evidence clearly favors the Duffie et al. (2005) argument that a surge in the population of potential buyers reduces profits for search intermediaries (i.e., dealers) due to the bypass mechanism.

The bypass mechanism is most important in markets where sellers can easily circumvent dealers. A seller's ability to circumvent dealers depends partly on the amount of goods available. We define "thick" markets as those where a large volume of the good is potentially for sale; in "thin" markets, by contrast, the aggregate market size is small (Gavazza, 2011). For example, the popular Nissan Sentra sedan trades in a thick market, but the Nissan 350Z
sports car, which is produced at a relatively low volume, trades in a thin one. Although a similar set of dealers might be interested in either of these vehicles, a private seller is more likely directly to find a buyer for a Sentra than for a 350Z. Under the bypass mechanism, we would expect the intermediary margins to decrease more, post tsunami, for vehicles that trade in thick markets. We show that this is true.

These findings highlight the importance of the bypass mechanism in the used car market and provide an explanation for the sharp divergence, described in Figures 1 and 2, between recent dealer margins in new and used cars. Central to this explanation is the fact that car dealers are generally indispensable in new car sales but not in used car sales. Notably, approximately $45 \%$ of used car transactions in the first quarter of 2021 took place between private parties. ${ }^{1}$ Consistent with Rubinstein and Wolinsky (1987, 1990), we find that scarcity increased dealer margins in the new car market, where the dealers' participation is required for sales. In the used car market, where direct private sales are feasible and common, scarcity did not lead to higher dealer margins, which is consistent with the bypass mechanism outlined in Duffie et al. (2005).

Low-mileage used cars are closer substitutes for new cars than high-mileage used cars are. We therefore expect that, under the bypass mechanism, the tsunami-induced demand surge will have a greater effect on lightly used Japanese cars. We find that this was the case, as dealer margins dropped more for low-mileage vehicles. This finding buttresses the argument that the key driver of our results was the post-tsunami scarcity of substitutes for new Japanese vehicles after new-vehicle production was disrupted.

Our context of the Japanese tsunami shows how scarcity in one market (new Japanese cars) can lead to a positive demand shock in related markets (used Japanese cars). A

[^0]common intuition is that when there is a positive (negative) demand shock, the holders of those goods will likely profit (lose). Numerous studies in the field of strategy support this intuition (Lim et al., 2023; Wang et al., 2020; Albuquerque and Bronnenberg, 2012); the primary contribution of these studies is to understand how firm-level heterogeneity influences the performance responses to demand shocks.

In contrast, our paper does not study the asset holders (usually producers of the goods) ${ }^{2}$ but rather the intermediaries who facilitate the exchange. These entities specialize in matching buyers and sellers, not in production. For intermediaries, we show that the impact of a positive demand shock can be ambiguous. Our finding that a positive demand shock in the used car market reduced intermediary profits is likely to hold in other settings where sellers and buyers can transact privately. Critical to this intuition is the role of the bypass mechanism; during times of peak demand, it becomes easier to bypass the intermediary entirely.

## 2 Data and Summary Statistics

To explore the role of scarcity on intermediary profits, we examine the retail and wholesale prices and financing of used vehicles that were purchased in 38 states in the period 20092013. Our data set consists of transaction-level information on pricing and financing. The loans to fund these purchases were made by a large automotive indirect-finance company. In total, we observe the prices and vehicle characteristics of 61,694 loans that were originated at 1,144 dealerships located in 693 U.S. ZIP codes, as described in Table 2. We also observe

[^1]the borrower characteristics and loan terms. ${ }^{3}$
Table 2 presents summary statistics for prices, vehicle characteristics, buyer characteristics, and loan terms. The average vehicle has approximately 41,380 miles on it at time of sale, a wholesale value (i.e., the price the dealer paid for that particular vehicle) of $\$ 13,798$, and a retail price of $\$ 17,541$. The 2010 new vehicle sales in the U.S. for each make and model are provided by Automotive News (a Crain Communications company).

### 2.12011 Japanese Tsunami

A magnitude 9.0 earthquake, the largest tremor to hit Japan since 1850, struck on March 11, 2011 off the coast of the Tōhoku region of Honshu, Japan, generating a tsunami with waves exceeding 100 feet in height. The resulting wave damage destroyed nearly everything along a 350-mile span of coastline, including over a dozen ports, and triggered a meltdown of the Fukushima nuclear reactor. Estimates of the economic damage wrought by the earthquake are on the order of US $\$ 360$ billion (Ferris and Solís, March 11, 2013).

All Japanese automotive manufacturers (e.g., Honda, Toyota, and Nissan) had plants in or near the affected area, and they closed those facilities. Japanese plants that were less directly affected by the earthquake and tsunami were also forced to stop production, as earthquake-related damage to approximately 500 tier-1 and tier-2 suppliers resulted in logistics difficulties and parts shortages. Both Toyota's and Honda's Japanese-built vehicle output declined by over $60 \%$ in March; other manufacturers also experienced steep decreases. Output levels in April were even lower (Wheatley and Ramsay, July 1, 2011). Production by Japanese automakers outside Japan was also severely affected. ${ }^{4}$ As shown in Table 1,

[^2]the production decrease for Japanese manufacturers was very substantial both in Japan and North America. In contrast, non-Japanese manufacturers essentially continued manufacturing without interruption. South Korean and American automakers met some of the demand that Japanese producers could not service (Edgerton, March 9, 2012). ${ }^{5}$

Table 1
Automotive Production by Month Post-tsunami automotive production decline relative to the Jan.-Feb. 2011 rate. Sources: International Organization of Motor Vehicle Manufacturers; Japan Automobile Manufacturers Association; VDA: German Association of the Automotive Industry; Automotive News, Crain Communications.

|  | Mar | Apr | May | Jun | Jul | Aug |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: |
| Japan Output | $-46 \%$ | $-61 \%$ | $-37 \%$ | $-4 \%$ | $3 \%$ | $-7 \%$ |
| North America Output by Japan Mfrs | $10 \%$ | $-42 \%$ | $-38 \%$ | $-19 \%$ | $-42 \%$ | $3 \%$ |
| North America Output by non-Japan Mfrs | $34 \%$ | $7 \%$ | $22 \%$ | $22 \%$ | $-12 \%$ | $25 \%$ |
| Germany Output | $27 \%$ | $3 \%$ | $23 \%$ | $1 \%$ | $1 \%$ | $-14 \%$ |

Lost production that was attributable to the tsunami amounted to almost $5 \%$ of global car production-approximately 2.8 million vehicles, of which $90 \%$ were from Japanese automakers (source: IHS Global Insight). The vehicle production losses of the Japanese manufacturers peaked in the second quarter, and all Japanese automakers returned to profitability from September 2011 to March 2012.
months. Toyota's North American production was impacted by a parts shortage, resulting a production loss of 500,000 vehicles. Honda's U.K. plant reduced its production capacity by $50 \%$ for almost two months, and Nissan's three U.S. plants were completely shuttered for a few months.
${ }^{5}$ Jesse Toprak, a vice president at TrueCar.com, observed that "part of the U.S. companies' ability to pick up sales resulted from being ready with small, high-MPG cars as gas prices were rising then as now. Sales shot up for models like the Chevrolet Cruze and Ford Fiesta, along with models from Korean maker Hyundai."

## 3 Theoretical Framework

Consider a model in which there are sellers, intermediaries, and buyers. We define a time of scarcity as a period in which there is an increase in the number of potential buyers per unit of the good, and we analyze the implications of scarcity for the price of the good and for intermediary profits. The motive for trade is that buyers place a higher valuation on the good than sellers do. There are also search frictions that impede the meeting of sellers and buyers when markets are not thick. Intermediaries can help relieve those frictions by buying the good from sellers at the wholesale price and selling it to buyers at the retail price. The intermediary profit is the difference between the retail and wholesale prices, and the intermediary margin is the profit scaled by the wholesale price.

The theoretical implications of scarcity depend on whether the intermediaries are indispensable to trade. We begin by assuming that trade can only occur through the medium of an intermediary (i.e., intermediaries are completely indispensable). In the simplified version of the model of Rubinstein and Wolinsky (1987) with one seller, one buyer, and one indispensable intermediary, trade takes place sequentially between the seller-intermediary and between the intermediary-buyer, and prices are determined via Nash bargaining. We introduce scarcity by increasing the number of buyers and allowing for random matching between the intermediary and any given buyer in the manner of Rubinstein and Wolinsky (1990). As shown in Rubinstein and Wolinsky (1990), as the number of buyers increases, the retail price rises. Under the bargaining model in Rubinstein and Wolinsky (1987), the single seller retains a fixed fraction of the retail price, so the wholesale price rises proportionately with the retail price. This gives rise to the implications below.

Hypothesis 1: If the intermediary is indispensable to trade, then as the number of potential

## buyers increases

a. wholesale and retail prices increase
b. the intermediary profit per unit increases
c. the intermediary margin is constant.

When buyers and sellers can meet on their own (i.e., when the intermediary is not indispensable), an increase in the number of potential buyers enhances the bargaining power of sellers and thereby alters the terms on which trade occurs. In the theory of Duffie et al. (2005), an increase in the number of interested buyers can be modeled as a higher hazard rate for the direct meeting of a seller with a potential consumer of the good. When the hazard rate is high enough, prices increase along with hazard rate intensity (i.e., as more potential buyers enter the market), eventually reaching a Walrasian price as search frictions dissipate. The intermediary profit, however, decreases in the number of potential buyers, as sellers can more easily circumvent the intermediary and trade directly with buyers. We refer to this strengthening of a seller's hand in negotiating with intermediaries as the bypass mechanism. It generates the following theoretical predictions.

Hypothesis 2: If buyers and sellers can meet directly (i.e., if the intermediary is not indispensable to trade), then as the number of potential buyers increases
a. wholesale and retail prices increase
b. the intermediary profit per unit decreases
c. the intermediary margin decreases.

From a formal standpoint, Hypothesis 2a requires that the number of potential buyers be high enough (i.e., that the hazard of meeting a potential buyer be sufficiently large), but
the general intuition that prices rise when counterparties find it easier to meet is straightforward and compelling. Rubinstein and Wolinsky (1987) do not consider changes in the relative numbers of buyers and sellers, but they do show that when buyers and sellers can meet on their own, an increase in the probability of their meeting reduces the markup of intermediaries, which is a similar intuition to Hypothesis 2b.

The bypass mechanism is likely to be more important in thick markets in which the quantity of goods available is large. Only for more familiar and standardized goods will direct connections of buyers and sellers be feasible.

Overall, it is clear from Hypotheses 1 and 2 that an influx of potential buyers raises prices; this is a general implication of scarcity. The impact of scarcity on intermediary per-unit profits, however, depends crucially on the ability of buyers and sellers to bypass middlemen and trade directly. We emphasize that Hypotheses 1 and 2 describe intermediaries' per-unit profits, not their overall profits. Scarcity that is driven by a negative supply shock could have quite different implications for overall profits than scarcity that is generated by a positive demand shock. In our empirical analysis, we analyze the role of intermediaries in the U.S. used car market and assess the effect of scarcity on their per-unit profits.

## 4 Empirical Methodology

We analyze prices and profits during times of scarcity using transaction-level data. We use the following regression specification for our analysis of a sale $l$ that takes place during period $t$ :

$$
\begin{equation*}
\text { Outcome }_{l t}=\beta_{0}+\beta_{1} \text { Japan }_{l}+\beta_{2} \text { Tsunami }_{t}+\beta_{3} \text { Japan }_{l} * \text { Tsunami }_{t}+\gamma \text { Controls }_{l t}+\epsilon_{l t}, \tag{1}
\end{equation*}
$$

$$
\text { Outcome }_{\text {imdt }}=
$$

where the variable Outcome represents Wholesale Price, Retail Price, Dealer Profit, Dealer Margin, and Loan-to-Value Ratio across various tables. In Equation (1), Japan is a dummy that equals 1 if the car is made by a Japanese manufacturer. ${ }^{6}$ Tsunami is a dummy variable that equals 1 if the sale took place between March 15, 2011, and Sept 15, 2011. The coefficient of interest is $\beta_{3}$, which describes the impact of the tsunami on used Japanese vehicles.

We control for make-model fixed effects (e.g., a dummy variable if the vehicle is a Honda Civic), dealership, and month of the year effects when predicting pricing and financing outcomes. The main effect on Japan is absorbed by the make-model fixed effects and is not reported in our tables. To control for time trends in the data, we use two different approaches. The first is to create a trend variable for the year, starting in 2009. This variable Year Trend takes the value of 0 in 2009, 1 in 2010, etc. Second, in many specifications we use year or year-month fixed effects. We also construct interactions between Year Trend and Japan, to control for Japanese car specific trends, and between the month and Japan, to control for seasonal differences between Japanese and non-Japanese vehicle sales.

## 5 Results

### 5.1 Prices and Dealer Margins

As described in Section 2.1, the March 2011 tsunami caused a dramatic short-term reduction in the supply of new Japanese cars. We propose that the tsunami therefore created a positive demand shock for used Japanese cars. Given that new Japanese cars were very hard to

[^3]find, we hypothesize that some of the consumers who had been considering buying a new Japanese car entered the market for a used Japanese car instead. These consumers most likely remained loyal to Japanese makes, as household-level brand preferences exhibit pronounced persistence (Bronnenberg et al., 2012). The used cars in our sample are quite young (with a median age of three years) and therefore probably served as reasonable substitutes for new cars.

To assess whether the tsunami caused a positive demand shock and created a period of scarcity for used Japanese cars, we estimate equation (1) for vehicle wholesale prices: we regress the price paid by a dealer for a used car on an indicator for the post-tsunami period (March 15-Sept 15, 2011), an interaction between the post-tsunami indicator and a dummy for Japanese cars, a linear control for the year of the sale, and fixed effects for the make-model of the car, the dealership, and the month of the year of the sale. Standard errors are double clustered by dealership and vehicle make. We find, in the first column of Table 3, a positive and significant effect of the post-tsunami-used Japanese car interaction (coefficient $=427.5$ and $t$-statistic $=3.02$ ): during the post-tsunami period, dealers purchased used Japanese cars at a premium of $\$ 427.5$. This increase represents a meaningful effect compared to the $\$ 13,400$ average vehicle wholesale price for used Japanese cars over the entire sample. This is clear evidence that the tsunami boosted the prices of the treated (i.e., used Japanese) cars and created a period of scarcity. As the overall supply of used Japanese cars in the U.S. was likely unaffected by the tsunami, this price increase can be attributed to a positive demand shock generated by the reduced availability of new Japanese cars. The resulting increase in wholesale prices is consistent with the first predictions of Hypotheses 1a and 2a in Section 3.

Including year-month fixed effects has little impact on the estimated effect on the in-
teraction, as detailed in the second column of Table 3. In the third column of Table 3, we present the results from including a control for the borrower credit score, an indicator for previous borrower bankruptcy, an indicator for borrower homeowner status, a control for the car mileage, an interaction between a Japanese car dummy and the years since 2008 (i.e., a Japan-specific time trend), and an interaction between a Japanese car dummy and the month (i.e., a Japanese seasonal control). We find a coefficient of $\$ 361.7$ and a $t$-statistic of 3.49. Decomposing the interaction into its month-by-month effects, we show, in the fourth column of Table 3, that vehicle wholesale prices were significantly higher for used Japanese cars in every month from April to August 2011.

In the fifth column of Table 3, we show that used Japanese cars did not sell at elevated wholesale prices in the period before the tsunami. A regression of vehicle wholesale price on an indicator for a sale in the six months before the tsunami and the interaction between this indicator and a Japanese vehicle dummy yields an insignificant result for the interaction (coefficient $=-68.1$ and $t$-statistic $=-0.50$ ). Thus, there is no apparent pre-trend in advance of the disaster.

The results in Table 3 establish that the tsunami caused a demand-driven scarcity in used Japanese cars. Hypotheses 1a and 2a both predict that retail prices (along with wholesale prices) should rise as a result of this scarcity. We test this prediction by regressing the retail price realized by the dealer on the post-tsunami-used Japanese car interaction and the standard controls. We find, in the first column of Table 4, that retail prices for used Japanese cars were indeed higher (coefficient $=274.1$ and $t$-statistic $=1.82$ ) in the post-tsunami period, confirming the retail price predictions of Hypotheses 1a and 2a. This result continues to hold when we include month-year fixed effects and borrower-level, vehicle-level, and time-trend controls, as detailed in the second and third columns of Table 4. We show, in the fourth
column of Table 4, that retail prices for used Japanese cars were generally higher from April to August 2011, but the monthly effects are only significant during two months. We do not find a significant retail price shift for these vehicles in the six months preceding the tsunami, as we detail in the fifth column of Table 4.

The tsunami raised wholesale and retail prices, as shown in Tables 3 and 4, which is consistent with Hypotheses 1a and 2a. We next assess the distinguishing predictions of Hypotheses 1 and 2, which focus on dealer profits and margins. If dealers are largely indispensable and cannot easily be circumvented, Hypothesis 1b predicts that dealer profits will increase in a period of scarcity, and Hypothesis 1c predicts that dealer margins will be constant. If, however, buyers and sellers can meet directly and with ease, Hypothesis 2b predicts that dealer profits will decrease in a period of scarcity, and Hypothesis 2c predicts that dealer margins will decline. We test these competing predictions by regressing dealer profits on the post-tsunami-used Japanese car interaction and the standard controls. We find, in the first column of Table 5, that dealer profits declined by $\$ 180.9$ ( $t$-statistic=-2.18) for used Japanese cars after the tsunami. This is clear evidence in favor of Hypothesis 2b. The decline in intermediary profits during the tsunami-induced scarcity and price surge is consistent with the bypass mechanism: sellers could negotiate directly with buyers, and an influx in potential buyers made it easier for sellers to avoid using intermediaries. The dealers' weakened bargaining position led to lower profits per vehicle for them.

Regression specifications including month-year fixed effects, borrower- and vehicle-level characteristics, and time trend controls confirm this basic finding, as shown in the second and third columns of Table 5. Including the full set of controls, we find that the tsunami reduced dealer profits on used Japanese vehicles by $\$ 170.1$ ( $t$-statistic $=-2.41$ ).

Note that, although dealer profits on used Japanese cars were lower in the post-tsunami
period, dealers did not lose money, on average, on these transactions. The average dealer profit on all used cars over the entire sample period was $\$ 4,488$ per vehicle, and dealer profits on used Japanese cars declined by $3.8 \%$ after the tsunami. So, average profits clearly remained positive.

The results in the fourth column of Table 5 show that dealer profits were broadly lower from April to August 2011, although only one of the monthly indicators is significant. Taking the post-tsunami period as a whole, it is clear that scarcity decreased dealers' per-vehicle profits on used Japanese cars. The results in the fifth column of Table 5 show that there was no discernible pre-trend in profits on used Japanese autos in the six months before the tsunami (coefficient $=-44.1$ and $t$-statistic $=-0.64$ ).

Given the results in Tables 3 and 5 showing that vehicle wholesale prices were higher and dealer profits were lower on used Japanese cars in the post-tsunami period, one would expect that dealer margins were also lower. We show that this is indeed true in the first column of Table 6: the tsunami reduced margins on used Japanese cars by 2.2 percentage points $(t$-statistic $=-2.81)$. This finding is robust to the inclusion of month-year, borrowerlevel, vehicle-level, and time trend controls, as displayed in the second and third columns of Table 6. In the specification with all controls, we find a reduced margin of 2.1 percentage points ( $t$-statistic $=-3.23$ ). Relative to the average dealer margin of $35.1 \%$, this represents a drop of $6.0 \%$.

Dealer margins were significantly lower in four out of the five months in the period from April to August 2011, as we detail in the fourth column of Table 6. We find, in the fifth column of Table 6, no evidence of a pre-trend (coefficient $=-0.003$ and $t$-statistic $=-0.56$ ); this finding is confirmed by the graphical evidence in Figure 3.

Tables 5 and 6 show that the bypass mechanism plays an important role in the used car
market as a whole. We propose that the effect of the bypass mechanism will be stronger in the subset of vehicles for which dealers have a weaker (or no) role in facilitating sales.

In what types of markets are intermediaries dispensable? In general, search frictions for individual agents are less severe in thick markets, where a relatively large number of assets are available for trade (Gavazza, 2011). In these markets, it is feasible for buyers and sellers to find each other, and intermediaries are not central in facilitating trade. Intermediaries are more important in thin markets, where the number of assets is relatively small.

We describe the thickness of a market for a given auto make and model by measuring the total sales of that make-model in the year preceding the tsunami (i.e., 2010). ${ }^{7}$ Vehicles with greater previous year sales trade in thicker markets. One implication of the bypass mechanism is that the post-tsunami scarcity in used Japanese vehicles should reduce dealer margins more substantially in thick markets.

We assess the relative impact of the tsunami on thick versus thin markets by regressing the wholesale price on the triple interaction between the post-tsunami indicator, the used Japanese car dummy, and the 2010 sales of the make-model, along with all the relevant double interactions and the full set of controls. We find, in the first column of Table 7, that the effect of the triple interaction is positive (coefficient $=329.8$ and $t$-statistic $=2.76$ ). The tsunami led to higher wholesale prices especially for Japanese cars that traded in thick markets with a large number of vehicles. We also show, in the second column of Table 7, that the impact of the triple interaction on retail prices was positive.

The main test, however, regresses dealer margins on the triple interaction. We find that dealer margins were particularly low for used Japanese cars in thick markets (coefficient=0.011 and $t$-statistic $=-2.62$ ), as displayed in the third column of Table 7.

[^4]Table 7 uses a continuous measure of sales. In Figure 4a and Figure 4b we employ a 100,000 -unit threshold to denote thick markets; those with sales below that threshold represent thin markets. Consistent with the results in Table 7, Figure 4a shows that the tsunami had no measurable impact on dealer margins for used Japanese cars in thin markets, while Figure 4b illustrates the large negative impact of the tsunami on dealer margins on used Japanese cars in thick markets.

The results in Tables 3-7 are all broadly consistent with Hypothesis 2. They support the argument that direct sales from buyers to sellers are a viable alternative to the use of dealers. An increase in the set of potential buyers of used Japanese cars during the post-tsunami period therefore served to reduce dealer profits and margins, via the bypass mechanism, at a time when wholesale and retail prices increased. As expected, this effect was strongest in thick markets.

### 5.2 Substitution Effects

Our central claim is that the post-tsunami demand surge for used Japanese cars arose from the scarcity of new Japanese cars. If this is correct, then demand should have increased the most for low-mileage used Japanese cars, as they were the closest substitutes for new Japanese cars.

We test this claim by regressing wholesale prices on the triple interaction between the post-tsunami indicator, the used Japanese car dummy, and the mileage of the car (scaled by 10,000 ), along with all the related double interactions and the full set of controls. We find, in the first column of Table 8, that wholesale prices rose less, post tsunami, on highmileage used Japanese cars (coefficient $=-108.0$ and $t$-statistic $=-2.77$ ) than on low-mileage used Japanese cars, consistent with the argument that the low-mileage vehicles were closer
substitutes. We find that the retail prices of used Japanese cars did not vary significantly (coefficient=-61.1 and $t$-statistic=-1.57) with mileage, as detailed in column two of Table 8.

## 6 Conclusion

Scarcity is a common and important feature of markets, and we study its effects on the profits of intermediaries through an analysis of the U.S. used auto market. We show that after the 2011 tsunami, the disruption of new Japanese auto manufacturing led to a demand surge for used Japanese vehicles, resulting in elevated wholesale and retail prices. We find that during the post-tsunami period, the dealers that act as sales intermediaries for used Japanese cars experienced a $3.8 \%$ drop in profits and a $6.0 \%$ reduction in margins. The effects were particularly strong in thick markets in which there were a large quantity of vehicles available. These findings are consistent with the search theory argument that, during scarcity, intermediary profits decrease due to a bypass mechanism in which the introduction of new potential buyers makes it easier for sellers to circumvent dealers.

Dealers possess superior industry knowledge, a degree of market power, and relatively stable access to financing. In all these respects, they seem well-suited to thrive during an episode of scarcity. We show instead that when dealers are not indispensable, scarcity causes their market power to deteriorate, resulting in lower margins. This suggests that the 20202022 supply chain disruptions will generate disparate profit outcomes for intermediaries, depending on the feasibility of private party sales. In the used auto market in particular, the emergence of very large used car specialist firms may limit the negative impact of scarcity on dealer profits, relative to the 2011 post-tsunami experience.

A careful analysis of the complex effects of scarcity is important for a full understanding
of the propagation of price fluctuations through the economy and the relative impacts of those fluctuations on producers, intermediaries, and consumers.

## References

Albuquerque, Paulo, and Bart J. Bronnenberg, 2012, Measuring the impact of negative demand shocks on car dealer networks, Marketing Science 31, 4-23.

Argyres, Nicholas, Joseph T. Mahoney, and Jackson Nickerson, 2019, Strategic responses to shocks: Comparative adjustment costs, transaction costs, and opportunity costs, Strategic Management Journal 40, 357-376.

Bennett, Victor M., Robert Seamans, and Feng Zhu, 2015, Cannibalization and option value effects of secondary markets: Evidence from the us concert industry, Strategic Management Journal 36, 1599-1614.

Bennett, Victor Manuel, 2013, Organization and bargaining: Sales process choice at auto dealerships, Management Science 59, 2003-2018.

Brandenburger, Adam M, and Harborne W Stuart Jr, 1996, Value-based business strategy, Journal of economics \& management strategy 5, 5-24.

Brealey, Richard, Hayne E. Leland, and David H. Pyle, 1977, Informational asymmetries, financial structure, and financial intermediation, The Journal of Finance 32, 371-387.

Bronnenberg, Bart J, Jean-Pierre H Dubé, and Matthew Gentzkow, 2012, The evolution of brand preferences: Evidence from consumer migration, American Economic Review 102, 2472-2508.

Cialdini, Robert B, 2007, Influence: The psychology of persuasion, volume 55 (Collins New York).

Duffie, Darrell, Nicolae Gârleanu, and Lasse Heje Pedersen, 2005, Over-the-counter markets, Econometrica 73, 1815-1847.

Edgerton, Jerry, March 9, 2012, How the japanese tsunami changed the auto industry, Technical report, CBS News, www.cbsnews.com/news/how-the-japanese-tsunami-changed-the-auto-industry/.

Feng, Josh, and Luca Maini, 2023, Demand inertia and the hidden impact of pharmacy benefit managers, Management Science (forthcoming) .

Fernandez-Mateo, Isabel, 2007, Who pays the price of brokerage? transferring constraint through price setting in the staffing sector, American Sociological Review 72, 291-317.

Ferris, Elizabeth, and Mireya Solís, March 11, 2013, Earthquake, tsunami, meltdown - the triple disaster's impact on japan, impact on the world, Technical report, Congressional Research Service, https://www.brookings.edu/blog/up-front/2013/03/11/earthquake-tsunami-meltdown-the-triple-disasters-impact-on-japan-impact-on-the-world/.

Gans, Joshua, and Michael D Ryall, 2017, Value capture theory: A strategic management review, Strategic Management Journal 38, 17-41.

Gavazza, Alessandro, 2011, The role of trading frictions in real asset markets, American Economic Review 101, 1106-43.

Giustiziero, Gianluigi, 2021, Is the division of labor limited by the extent of the market? opportunity cost theory with evidence from the real estate brokerage industry, Strategic Management Journal 42, 1344-1378.

Gubler, Timothy, and Ryan Cooper, 2019, Socially advantaged? how social affiliations influence access to valuable service professional transactions, Strategic Management Journal 40, 2287-2314.

Israeli, Ayelet, Fiona Scott-Morton, Jorge Silva-Risso, and Florian Zettelmeyer, 2022, How market power affects dynamic pricing: Evidence from inventory fluctuations at car dealerships, Management Science 68, 895-916.

Jansen, Mark, Samuel Kruger, and Gonzalo Maturana, 2023a, Dealer financing in the subprime auto market: Markups and implicit subsidies, in American Finance Association Annual Meeting Paper.

Jansen, Mark, Lamar Pierce, Jason Snyder, and Hieu Nguyen, 2023b, Product sales incentive spillovers to the lending market, Management Science (forthcoming).

Khurana, Rakesh, 2002, Market triads: A theoretical and empirical analysis of market intermediation, Journal for the Theory of Social Behaviour 32, 239-262.

Krakovsky, Marina, 2015, The middleman economy: How brokers, agents, dealers, and everyday matchmakers create value and profit (Palgrave Macmillan).

Lim, Najoung, Seojin Kim, and Rajshree Agarwal, 2023, Weathering a demand shock: The impact of prior vertical scope on post-shock firm response, Strategic Management Journal 44, 1965-2004.

Lizzeri, Alessandro, 1999, Information revelation and certification intermediaries, The RAND Journal of Economics 214-231.

Mahoney, Joseph T., and Lihong Qian, 2013, Market frictions as building blocks of an organizational economics approach to strategic management, Strategic Management Journal 34, 1019-1041.

Oberholzer-Gee, Felix, 2021, Better, Simpler Strategy: A Value-Based Guide to Exceptional Performance (Harvard Business Review Press).

Oh, Chang Hoon, and Jennifer Oetzel, 2011, Multinationals' response to major disasters: how does subsidiary investment vary in response to the type of disaster and the quality of country governance?, Strategic Management Journal 32, 658-681.

Rider, Christopher, and Sampsa Samila, 2019, Envisioning value: Certification, matchmaking, and returns to brokerage, Working Paper.

Rubinstein, Ariel, and Asher Wolinsky, 1987, Middlemen, Quarterly Journal of Economics 102, 581-593.

Rubinstein, Ariel, and Asher Wolinsky, 1990, Decentralized trading, strategic behaviour and the walrasian outcome, Review of Economic Studies 57, 63-78.

Spulber, Daniel F., 1996, Market microstructure and intermediation, Journal of Economic Perspectives 10, 135-152.

Teece, David J., Gary Pisano, and Amy Shuen, 1997, Dynamic capabilities and strategic management, Strategic Management Journal 18, 509-533.

Wang, Tang, Vikas A. Aggarwal, and Brian Wu, 2020, Capability interactions and adaptation to demand-side change, Strategic Management Journal 41, 1595-1627.

Wenzel, Matthias, Sarah Stanske, and Marvin B Lieberman, 2020, Strategic responses to crisis, Strategic Management Journal 41, 3161.

Wheatley, Malcolm, and Malcolm Ramsay, July 1, 2011, After the disaster in japan, Technical report, Automotive Logistics, https://www.automotivelogistics.media/after-the-disaster-in-japan/7408.article.


Figure 1. This figure reports new vehicle prices and dealership margins over 2016-2021 scaled so that Q1, $2016(=100)$. The blue dashed line represents average retail transaction prices of new vehicle purchases excluding any manufacturers' incentives. The blue solid line represents the average front-end gross margins at U.S. dealerships on new car sales. Sources: Kelley Blue Book, Cox Automotive; Power Information Network, J.D. Power; Data accessed November 12, 2021

(a) New and Used Car Prices (2016 to 2021)

(b) New and Used Car Margins (2016 to 2021))

Figure 2. This figure reports (a) new and used car prices and (b) dealership margins by calendar quarter. The figures are scaled so that Q1, $2016(=100)$. In Figure (a), the blue dashed line represents average retail transaction prices of new vehicle purchases excluding any manufacturers' incentives. The red dashed line represents the Manheim Used Car Index, which is an index of used car prices based on more than 5 million used vehicle transactions per year. Manheim reports that the used vehicle price index is independent of underlying shifts in the characteristics of vehicles being sold. In Figure (b), the blue solid line represents the average front-end gross margins at U.S. dealerships on new car sales. The red solid line represents average gross margins at U.S. dealerships on used car sales. Sources: Manheim, Cox Automotive; Power Information Network, J.D. Power; data accessed November 12, 2021.


Figure 3. This figure reports differences in the dealership profit margins for the sale of used cars between Japanese makes and non-Japanese makes. The figure shows a discontinuity in the first quarter of 2011, the quarter in which the tsunami occurred (March 15, 2011). Controls include monthly income, credit score, and prior bankruptcy. Standard errors are double clustered by dealership and vehicle make.
(a) Dealer Margin for Low Sales Volume Vehicles

(b) Dealer Margin for Large Sales Volume Vehicles


Figure 4. This figure reports differences in the dealership profit margins for the sale of used cars between Japanese makes and non-Japanese makes. The figure shows a discontinuity in the first quarter of 2011, the quarter in which the tsunami occurred (March 15, 2011). Panel a shows the results for sales of thinly traded vehicles (i.e., vehicles with annual sales volume less than 100,000 vehicles in 2010). Panel b shows the results for sales of thickly traded vehicles (i.e., vehicles with annual sales volume in excess of 100,000). Controls include monthly income, credit score, and prior bankruptcy. Standard errors are double clustered by dealership and vehicle make.
(a) Dealer Margin for Low Mileage Vehicles Vehicles

(b) Dealer Margin for High Mileage Vehicles Vehicles


Figure 5. This figure reports differences in the dealership profit margins for the sale of used cars between Japanese makes and non-Japanese makes. The figure shows a discontinuity in the first quarter of 2011, the quarter in which the tsunami occurred (March 15, 2011). Panel a (b) shows the results for sales of vehicles with fewer (more) than 40,000 miles. Controls include monthly income, credit score, and prior bankruptcy. Standard errors are double clustered by dealership and vehicle make.

Table 2
Summary Statistics This table reports summary statistics for the sample of 61,694 auto sales transactions extended by 1,144 dealerships. Means and standard deviations are reported. The data summarizes vehicle, and dealership characteristics, and also borrower characteristics.

|  | mean | sd | count |
| :--- | :---: | :---: | :---: |
| Vehicle and Dealership Characteristics |  |  |  |
| Japan | 0.452 | 0.498 | 62580 |
| Vehicle Wholesale Price | 13816 | 3659 | 62580 |
| Vehicle Retail Price | 17539 | 3879 | 62577 |
| Vehicle Mileage ('0,000) | 4.138 | 2.092 | 62579 |
| Vehicle Model 2010 Sales ('0,000) | 1.323 | 1.100 | 62194 |
| Dealer Profit Margin (\%) | 0.351 | 0.197 | 62577 |
| Dealer Profit Margin (\$) | 4488 | 2307 | 62577 |
| Borrower Characteristics |  |  |  |
| Credit Risk | 19.712 | 1.590 | 61096 |
| Credit Score | 533.999 | 44.744 | 61108 |
| Loan-to-Value ratio | 1.307 | 0.173 | 62574 |
| Ch. 7 Bankruptcy | 0.291 | 0.454 | 62580 |
| Homeowner | 0.094 | 0.292 | 62580 |
| Gross Monthly Income (ln) | 8.332 | 0.372 | 62580 |

Table 3
Difference in vehicle wholesale prices of Japanese- and non-Japanese-manufactured vehicles This table reports estimates from panel regressions of vehicle wholesale prices of used cars on whether the vehicle is manufactured by a Japanese manufacturer during the tsunami period. The dependent variable is the wholesale price of the vehicle at the time of sale. Japan is an indicator equal to one if the vehicle was manufactured by Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Scion, Mazda, Mitsubishi, Suzuki, Acura, or Isuzu. The regressions control for borrower characteristics (credit score, homeownership, prior bankruptcy, and income) and vehicle characteristics (mileage). Tsunami is defined as one if the date of sale falls within March 15-Sept 15, 2011, and zero otherwise. Fixed effects are included for make-model, dealership, month of the year, year, month-year and month-Japan, as indicated. The regressions also control for a Japan dummy interacted with the year trend. Standard errors are double clustered by dealership and vehicle make. T-statistics are shown in parentheses below the coefficient estimates. ${ }^{* * *},{ }^{* *}$, and ${ }^{*}$ indicate statistical significance at the $1 \%, 5 \%$, and $10 \%$ levels, respectively.

| Dep Var: Wholesale price | (1) | (2) | (3) | (4) | (5) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Japan * Tsunami | 427.5*** | 427.2** | $361.7^{*}$ |  |  |
|  | (3.02) | (2.96) | (3.49) |  |  |
| Tsunami | $556.3{ }^{* * *}$ | -146.1 | -82.3 |  |  |
|  | (4.68) | (-0.96) | (-0.69) |  |  |
| Year Trend | $215.7^{* * *}$ |  |  |  |  |
|  | (5.43) |  |  |  |  |
| Credit score |  |  | 0.49 | 0.49 | 0.49 |
|  |  |  | (1.64) | (1.66) | (1.65) |
| Ch. 7 BK prior to origination |  |  | 93.0*** | $93.4{ }^{* * *}$ | $93.2^{* * *}$ |
|  |  |  | (2.91) | (2.93) | (2.93) |
| Homeowner (=1) as of Date_Booking |  |  | -32.3 | -32.7 | -32.6 |
|  |  |  | (-1.28) | (-1.30) | (-1.29) |
| Gross Monthly Income (ln) |  |  | 1164.9*** | 1165.0*** | 1165.6*** |
|  |  |  | (15.83) | (15.86) | (15.90) |
| 10,000s Miles |  |  | -937.5*** | -937.4*** | -937.4*** |
|  |  |  | (-25.42) | (-25.41) | (-25.43) |
| Japan * Apr |  |  |  | $305.7{ }^{* *}$ |  |
|  |  |  |  | (2.62) |  |
| Japan * May |  |  |  | 409.9** |  |
|  |  |  |  | (2.24) |  |
| Japan * Jun |  |  |  | $531.7^{* *}$ |  |
|  |  |  |  | (2.65) |  |
| Japan * Jul |  |  |  | $574.6^{* * *}$ |  |
|  |  |  |  | (3.08) |  |
| Japan * Aug |  |  |  | $339.8{ }^{* *}$ |  |
|  |  |  |  | (2.42) |  |
| Japan * 6 Month Prior to Tsunami |  |  |  |  | -68.1 |
|  |  |  |  |  | (-0.50) |
| 6 Month Prior to Tsunami |  |  |  |  | 0.73 |
|  |  |  |  |  | (0.00) |
| Adjusted $R^{2}$ | 0.407 | 0.414 | 0.671 | 0.671 | 0.671 |
| Vehicle_Model | YES | YES | YES | YES | YES |
| Dealership | YES | YES | YES | YES | YES |
| Month | YES | YES | YES | YES | YES |
| Year | NO | YES | YES | YES | YES |
| MonthXYear | NO | YES | YES | YES | YES |
| JapanTrend | NO | NO | YES | YES | YES |
| MonthXJapan | NO | NO | YES | YES | YES |
| Observations | 58773 | 58773 | 57397 | 57397 | 57397 |

## Table 4

Difference in vehicle retail prices of Japanese- and non-Japanese-manufactured vehicles This table reports estimates from panel regressions of vehicle retail prices of used cars on whether the vehicle is manufactured by a Japanese manufacturer during the tsunami period. The dependent variable is the retail price of the vehicle at the time of sale. Japan is an indicator equal to one if the vehicle was manufactured by Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Scion, Mazda, Mitsubishi, Suzuki, Acura, or Isuzu. The regressions control for borrower characteristics (credit score, homeownership, prior bankruptcy, and income) and vehicle characteristics (mileage). Tsunami is defined as one if the date falls within March 15-Sept 15, 2011, and zero otherwise. Fixed effects are included for makemodel, dealership, month of the year, year, month-year, and month-Japan, as indicated. The regressions also control for a Japan dummy interacted with the year trend. Standard errors are double clustered by dealership and vehicle make. T-statistics are shown in parentheses below the coefficient estimates. ${ }^{* * *}$, ${ }^{* *}$, and ${ }^{*}$ indicate statistical significance at the $1 \%, 5 \%$, and $10 \%$ levels, respectively.

| Dep Var: Retail Price | (1) | (2) | (3) | (4) | (5) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Japan * Tsunami | 274.1* | $272.6^{*}$ | $221.6^{* *}$ |  |  |
|  | (1.82) | (1.79) | (2.06) |  |  |
| Tsunami | $593.2{ }^{* * *}$ | -22.9 | 31.4 |  |  |
|  | (4.94) | (-0.18) | (0.28) |  |  |
| Year Trend | $\begin{gathered} 377.6^{* * *} \\ (8.22) \end{gathered}$ |  |  |  |  |
| Credit score |  |  | 1.19*** | 1.20 *** | 1.19*** |
|  |  |  | (2.99) | (3.00) | (2.99) |
| Ch. 7 BK prior to origination |  |  | 102.5*** | 102.7*** | 102.7*** |
|  |  |  | (3.68) | (3.68) | (3.68) |
| Homeowner (=1) as of Date_Booking |  |  | -50.8* | -51.4** | -51.2** |
|  |  |  | (-2.02) | (-2.05) | (-2.03) |
| Gross Monthly Income (ln) |  |  | $1643.9^{* * *}$ | 1644.4*** | 1644.5*** |
|  |  |  | (22.82) | (22.86) | (22.88) |
| 10,000s Miles |  |  | -971.1*** | -970.9*** | -971.1*** |
|  |  |  | (-27.15) | (-27.14) | (-27.17) |
| Japan * Apr |  |  |  | 279.8* |  |
|  |  |  |  | (1.99) |  |
| Japan * May |  |  |  | 276.5 |  |
|  |  |  |  | (1.41) |  |
| Japan * Jun |  |  |  | 149.0 |  |
|  |  |  |  | (0.58) |  |
| Japan * Jul |  |  |  | 538.2 *** |  |
|  |  |  |  | (3.09) |  |
| Japan * Aug |  |  |  | 64.0 |  |
|  |  |  |  | (0.53) |  |
| Japan * 6 Month Prior to Tsunami |  |  |  |  | -132.8 |
|  |  |  |  |  | (-0.96) |
| 6 Month Prior to Tsunami |  |  |  |  | 23.4 |
|  |  |  |  |  | (0.14) |
| Adjusted $R^{2}$ | 0.404 | 0.411 | 0.645 | 0.645 | 0.645 |
| Vehicle_Model | YES | YES | YES | YES | YES |
| Dealership | YES | YES | YES | YES | YES |
| Month | YES | YES | YES | YES | YES |
| Year | NO | YES | YES | YES | YES |
| MonthXYear | NO | YES | YES | YES | YES |
| JapanTrend | NO | NO | YES | YES | YES |
| MonthXJapan | NO | NO | YES | YES | YES |
| Observations | 58770 | 58770 | 57396 | 57396 | 57396 |

Table 5
Difference in dealer profits of Japanese- and non-Japanese-manufactured vehicles This table reports estimates from a panel regressions of dealer profit margins in dollars on the sale of used cars on whether the vehicle is manufactured by a Japanese manufacturer during the tsunami period. The dependent variable is dealership margin measured in dollars. Japan is an indicator equal to one if the vehicle was manufactured by Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Scion, Mazda, Mitsubishi, Suzuki, Acura, or Isuzu. The regressions control for borrower characteristics (credit score, homeownership, prior bankruptcy, and income) and vehicle characteristics (mileage). Tsunami defined as one if the date falls within March 15-Sept. 15, 2011 and zero otherwise. Fixed effects are included for makemodel, dealership, month of the year, year, month-year, and month-Japan, as indicated. The regressions also control for a Japan dummy interacted with the year trend. Standard errors are double clustered by dealership and vehicle make. T-statistics are shown in parentheses below the coefficient estimates. ${ }^{* * *}$, ${ }^{* *}$, and ${ }^{*}$ indicate statistical significance at the $1 \%, 5 \%$, and $10 \%$ levels, respectively.

| Dep Var: Dealer Profits | (1) | (2) | (3) | (4) | (5) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Japan * Tsunami | -180.9** | -183.8** | -170.1** |  |  |
|  | (-2.18) | (-2.19) | (-2.41) |  |  |
| Tsunami | 1.87 | 93.7 | 56.0 |  |  |
|  | (0.04) | (0.96) | (0.65) |  |  |
| Year Trend | $250.0{ }^{* * *}$ |  |  |  |  |
|  | (12.97) |  |  |  |  |
| Credit score |  |  | 4.80*** | 4.80*** | 4.80*** |
|  |  |  | (21.99) | (21.97) | (21.89) |
| Ch. 7 BK prior to origination |  |  | 52.0* | 51.9* | 51.8* |
|  |  |  | (1.82) | (1.82) | (1.83) |
| Homeowner ( $=1$ ) as of Date_Booking |  |  | 11.7 | 11.8 | 11.9 |
|  |  |  | (0.56) | (0.56) | (0.57) |
| Gross Monthly Income (ln) |  |  | 1139.8*** | 1139.9*** | 1139.4*** |
|  |  |  | (24.18) | (24.22) | (24.25) |
| 10,000s Miles |  |  | -0.63 | -0.58 | -0.71 |
| Japan * Apr |  |  | (-0.04) | (-0.04) | (-0.05) |
|  |  |  |  | -93.7 |  |
|  |  |  |  | (-0.81) |  |
| Japan * May |  |  |  | -221.3 |  |
|  |  |  |  | (-1.62) |  |
| Japan * Jun |  |  |  | -348.6** |  |
|  |  |  |  | (-2.67) |  |
| Japan * Jul |  |  |  | -43.2 |  |
|  |  |  |  | (-0.38) |  |
| Japan * Aug |  |  |  | -209.6 |  |
|  |  |  |  | (-1.64) |  |
| Japan * 6 Month Prior to Tsunami |  |  |  |  | -44.1 |
|  |  |  |  |  | (-0.64) |
| 6 Month Prior to Tsunami |  |  |  |  | -24.2 |
|  |  |  |  |  | (-0.18) |
| Adjusted $R^{2}$ | 0.178 | 0.183 | 0.223 | 0.223 | 0.223 |
| Vehicle_Model | YES | YES | YES | YES | YES |
| Dealership | YES | YES | YES | YES | YES |
| Month | YES | YES | YES | YES | YES |
| Year | NO | YES | YES | YES | YES |
| MonthXYear | NO | YES | YES | YES | YES |
| JapanTrend | NO | NO | YES | YES | YES |
| MonthXJapan | NO | NO | YES | YES | YES |
| Observations | 58770 | 58770 | 57396 | 57396 | 57396 |

## Table 6

Difference in dealer percent margin of Japanese- and non-Japanese-manufactured vehicles This table reports estimates from panel regressions of dealer profit margins in percent on the sale of used cars on whether the vehicle is manufactured by a Japanese manufacturer during the tsunami period. The dependent variable is dealership margin measured in percent relative to the wholesale price of the vehicle. Japan is an indicator equal to one if the vehicle was manufactured by Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Scion, Mazda, Mitsubishi, Suzuki, Acura, or Isuzu. The regressions control for borrower characteristics (credit score, homeownership, prior bankruptcy, and income) and vehicle characteristics (mileage). Tsunami is defined as one if the date falls within March 15-Sept. 15, 2011, and zero otherwise. Fixed effects are included for make-model, dealership, month of the year, year, month-year, and month-Japan, as indicated. The regressions also control for a Japan dummy interacted with the year trend. Standard errors are double clustered by dealership and vehicle make. T-statistics are shown in parentheses below the coefficient estimates. ${ }^{* * *},{ }^{* *}$, and * indicate statistical significance at the $1 \%, 5 \%$, and $10 \%$ levels, respectively.

| Dep Var: Dealer Margin | (1) | (2) | (3) | (4) | (5) |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Japan * Tsunami | $-0.022^{* * *}$ | -0.022*** | -0.021*** |  |  |
|  | (-2.81) | (-2.79) | (-3.23) |  |  |
| Tsunami | -0.016** | 0.008 | 0.004 |  |  |
|  | (-2.70) | (0.82) | (0.49) |  |  |
| Year Trend | $0.012^{* * *}$ |  |  |  |  |
|  | (6.68) |  |  |  |  |
| Credit score |  |  | 0.00039*** | 0.00039*** | 0.00039*** |
|  |  |  | (18.12) | (18.11) | (18.02) |
| Ch. 7 BK prior to origination |  |  | 0.0012 | 0.0012 | 0.0012 |
|  |  |  | (0.44) | (0.43) | (0.43) |
| Homeowner (=1) as of Date_Booking |  |  | 0.0027 | 0.0027 | 0.0027 |
|  |  |  | (1.23) | (1.24) | (1.24) |
| Gross Monthly Income (ln) |  |  | 0.060 *** | 0.060 *** | 0.060 *** |
|  |  |  | (12.80) | (12.82) | (12.83) |
| 10,000s Miles |  |  | $0.025^{* * *}$ | $0.025^{* * *}$ | $0.025^{* * *}$ |
|  |  |  | (14.71) | (14.70) | (14.70) |
| Japan * Apr |  |  |  | -0.016* |  |
|  |  |  |  | (-2.01) |  |
| Japan * May |  |  |  | -0.025* |  |
|  |  |  |  | (-1.74) |  |
| Japan * Jun |  |  |  | -0.035*** |  |
|  |  |  |  | (-3.31) |  |
| Japan * Jul |  |  |  | -0.018 |  |
|  |  |  |  | (-1.52) |  |
| Japan * Aug |  |  |  | -0.026** |  |
|  |  |  |  | (-2.23) |  |
| Japan * 6 Month Prior to Tsunami |  |  |  |  | -0.0030 |
|  |  |  |  |  | (-0.56) |
| 6 Month Prior to Tsunami |  |  |  |  | 0.0020 |
|  |  |  |  |  | (0.19) |
| Adjusted $R^{2}$ | 0.212 | 0.216 | 0.272 | 0.272 | 0.272 |
| Vehicle_Model | YES | YES | YES | YES | YES |
| Dealership | YES | YES | YES | YES | YES |
| Month | YES | YES | YES | YES | YES |
| Year | NO | YES | YES | YES | YES |
| MonthXYear | NO | YES | YES | YES | YES |
| JapanTrend | NO | NO | YES | YES | YES |
| MonthXJapan | NO | NO | YES | YES | YES |
| Observations | 58770 | 58770 | 57396 | 57396 | 57396 |

Table 7
Triple Interaction: Sales Volume This table reports estimates from panel regressions of wholesale price, retail price, and dealer profit margins on the triple interaction between Japan, Tsunami, and Sales 2010 ('0,000). Japan is an indicator equal to one if the vehicle was manufactured by Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Scion, Mazda, Mitsubishi, Suzuki, Acura, or Isuzu. The regressions control for borrower characteristics (credit score, homeownership, prior bankruptcy, and income) and vehicle characteristics (mileage). Tsunami defined as one if the date falls within March 15-Sept. 15, 2011, and zero otherwise. Fixed effects are included for make-model, dealership, month of the year, year, month-year, and month-Japan, as indicated. The regressions also control for a Japan dummy interacted with the year trend. Standard errors are double clustered by dealership and vehicle make. T-statistics are shown in parentheses below the coefficient estimates. ${ }^{* * *},{ }^{* *}$, and * indicate statistical significance at the $1 \%, 5 \%$, and $10 \%$ levels, respectively.

|  | Wholesale price | Retail price | Dealer Margin |
| :--- | :---: | :---: | :---: |
| Tsunami | 53.9 | 154.6 | 0.0022 |
|  | $(0.32)$ | $(0.99)$ | $(0.23)$ |
| Japan * Tsunami | -95.3 | -49.6 | -0.0033 |
| Sales 2010 ('0,000) * Tsunami | $(-0.73)$ | $(-0.34)$ | $(-0.50)$ |
| Sales 2010 ('0,000) * Japan * Tsunami | -150.6 | -132.0 | 0.0025 |
|  | $(-1.43)$ | $(-1.45)$ | $(0.55)$ |
| Credit score | $329.8^{* * *}$ | $216.3^{* *}$ | $-0.011^{* *}$ |
|  | $(2.76)$ | $(2.04)$ | $(-2.62)$ |
| Ch. 7 BK prior to origination | 0.49 | $1.21^{* * *}$ | $0.00039^{* * *}$ |
|  | $(1.64)$ | $(2.98)$ | $(18.17)$ |
| Homeowner (=1) as of Date_Booking | $91.5^{* * *}$ | $102.0^{* * *}$ | 0.0012 |
|  | $(2.90)$ | $(3.66)$ | $(0.43)$ |
| Gross Monthly Income (ln) | -32.3 | $-51.8^{* *}$ | 0.0027 |
|  | $(-1.30)$ | $(-2.07)$ | $(1.22)$ |
| 10,000s Miles | $1166.0^{* * *}$ | $1645.3^{* * *}$ | $0.060^{* * *}$ |
|  | $(15.77)$ | $(22.66)$ | $(12.80)$ |
| Adjusted $R^{2}$ | $-937.6^{* * *}$ | $-971.1^{* * *}$ | $0.025^{* * *}$ |
| Vehicle_Model | $(-25.52)$ | $(-27.23)$ | $(14.66)$ |
| Dealership | 0.671 | 0.644 | 0.272 |
| Month | YES | YES | YES |
| Year | YES | YES | YES |
| MonthXYear | YES | YES | YES |
| JapanTrend | YES | YES | YES |
| ObservXJapan | YES | YES | YES |

## Table 8

Triple Interaction: Vehicle Miles This table reports estimates from panel regressions of wholesale price, retail price, dealer profit margins and loan-to-value ratios on the triple interaction between Japan, Tsunami, and Vehicle Mileage ('0,000). Japan is an indicator equal to one if the vehicle was manufactured by Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Scion, Mazda, Mitsubishi, Suzuki, Acura, or Isuzu. The regressions control for borrower characteristics (credit score, homeownership, prior bankruptcy, and income) and vehicle characteristics (mileage). Tsunami defined as one if the transaction date falls within March 15-Sept. 15, 2011, and zero otherwise. Fixed effects are included for make-model, dealership, month of the year, year, month-year, and month-Japan, as indicated. The regressions also control for a Japan dummy interacted with the year trend. Standard errors are double clustered by dealership and vehicle make. T-statistics are shown in parentheses below the coefficient estimates. ${ }^{* * *},{ }^{* *}$, and ${ }^{*}$ indicate statistical significance at the $1 \%, 5 \%$, and $10 \%$ levels, respectively.

|  | Wholesale price | Retail price | Dealer margin | LTV |
| :---: | :---: | :---: | :---: | :---: |
| Tsunami | -173.5 | -274.9 | -0.0100 | -0.0028 |
|  | (-0.88) | (-1.28) | (-0.97) | (-0.36) |
| Japan * Tsunami | $831.5^{* * *}$ | $463.5^{* * *}$ | -0.055*** | -0.055*** |
|  | (4.98) | (2.87) | (-4.87) | (-4.99) |
| 10,000s Miles * Tsunami | 21.5 | 72.8* | 0.0034** | 0.0022 |
|  | (0.61) | (1.99) | (2.72) | (1.56) |
| Japan * 10,000s Miles | $213.7^{* * *}$ | $205.9^{* * *}$ | -0.0061** | -0.0057** |
|  | (6.27) | (6.75) | (-2.44) | (-2.21) |
| Japan * 10,000s Miles * Tsunami | -108.0*** | -61.1 | 0.0073*** | $0.0070^{* * *}$ |
|  | (-2.77) | (-1.57) | (3.66) | (3.33) |
| Credit score | 0.50 | 1.20 *** | $0.00039^{* * *}$ | $0.00014^{* * *}$ |
|  | (1.67) | (2.98) | (18.01) | (10.26) |
| Ch. 7 BK prior to origination | 94.7 ${ }^{* * *}$ | 104.0*** | 0.0011 | 0.0043* |
|  | (2.95) | (3.67) | (0.42) | (1.95) |
| Homeowner (=1) as of Date_Booking | -35.6 | -53.3** | 0.0028 | -0.0069*** |
|  | (-1.45) | (-2.17) | (1.31) | (-2.95) |
| Gross Monthly Income (ln) | 1165.9*** | 1644.8*** | 0.060*** | $0.040^{* * *}$ |
|  | (15.94) | (22.78) | (12.90) | (8.96) |
| 10,000s Miles | -1053.9*** | -1088.4*** | $0.028^{* * *}$ | $0.026^{* * *}$ |
|  | (-50.34) | (-44.77) | (21.12) | (22.77) |
| Adjusted $R^{2}$ | 0.674 | 0.647 | 0.273 | 0.304 |
| Vehicle_Model | YES | YES | YES | YES |
| Dealership | YES | YES | YES | YES |
| Month | YES | YES | YES | YES |
| Year | YES | YES | YES | YES |
| MonthXYear | YES | YES | YES | YES |
| JapanTrend | YES | YES | YES | YES |
| MonthXJapan | YES | YES | YES | YES |
| Observations | 57397 | 57396 | 57396 | 57391 |

# Internet Appendix to: "Intermediary Profit in a Time of Scarcity"* 

For Online Publication

July 2022

[^5]

Figure IA. 1
This figure reports monthly new car sales volume from January 2010 to December 2011, scaled so that January $2010(=100)$. The red (blue) dashed line represents of new vehicle sales of Japanese (non-Japanese) manufacturers. The black dotted line represents the tsuanmi event in March 2011. Sources: Automotive News, Crain Communications

## Table IA. 1

Change in automotive production by country. Column 1 reports the change in production for 2011 relative to the 2010 rate. Column 2 reports the change in global market share of each country from 2010 to 2011. Source: Automotive News, Crain Communications

|  | $(1)$ | $(2)$ |
| :--- | ---: | :---: |
| China | $4.2 \%$ | $+1.3 \%$ |
| Japan | $-13.9 \%$ | $-16.2 \%$ |
| Germany | $5.8 \%$ | $+2.8 \%$ |
| South Korea | $9.2 \%$ | $+6.2 \%$ |
| India | $7.4 \%$ | $+4.4 \%$ |
| USA | $9.0 \%$ | $+6.0 \%$ |
| Rest of World | $5.2 \%$ | $+2.3 \%$ |


[^0]:    ${ }^{1}$ https://www.coxautoinc.com/market-insights/cox-automotive-13-month-rolling-used-vehicle-saar/, accessed January 10, 2022.

[^1]:    ${ }^{2}$ See for example, Bennett et al. (2015), which studies the thickness of the secondary market and its influence on producer behavior. Our study focuses on the performance of the intermediaries.

[^2]:    ${ }^{3}$ The raw data extend from the 1990 s to 2021 and include approximately 343,000 loans. We exclude loans outside of the five-year time window surrounding the tsunami. We also exclude a small number of observations for new cars and for loans with incomplete origination data.
    ${ }^{4}$ For example, Toyota's plants in China, Europe, and North America ceased production for up to three

[^3]:    ${ }^{6}$ The Japanese makes in our data are Toyota, Nissan, Lexus, Infiniti, Subaru, Honda, Acura, Mazda, Mitsubishi, Isuzu, Suzuki, and Scion.

[^4]:    ${ }^{7}$ Data comes from the website https://www.goodcarbadcar.net which aggregates make-model sales data.

[^5]:    * Mark Garmaise, Anderson School of Management at UCLA; Mark Jansen, University of Utah, Eccles School of Business; Jason Snyder, University of Utah, Eccles School of Business: University of Utah, Eccles School of Business

