

# TRADING WITH EXPERT DEALERS \*

Maria Chaderina

Vincent Glode

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## Abstract

We model investors' allocation of order flow across over-the-counter dealers jointly with dealers' costly acquisition of expertise that can be used to take advantage of investors across transactions. *Ceteris paribus*, investors prefer to allocate their order flow to dealers expected to intermediate large volumes of transactions and to acquire low levels of expertise, whereas dealers' benefits from acquiring expertise grow with the number of transactions they intermediate. Our model's equilibrium rationalizes why the most sought-after dealers often are those with the best data, technology, and skills, despite the significant adverse selection concerns triggered by their superior expertise.

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# 1 Introduction

A seminal insight from the literature on asymmetric information, dating back to at least Akerlof (1970), is that one should be wary of trading with a counterparty who possesses superior expertise (e.g., data, technology, and skills). Such counterparty can condition its trading decisions on superior information and take advantage of a less informed trader. Yet, in real financial markets, the dealers that intermediate most trades happen to be extremely sophisticated institutions. For example, Goldman Sachs remains the second most active over-the-counter (OTC) dealer of derivatives among U.S. banks while spending \$400M per year to acquire financial data from third-party sources and \$400K on average per employee to attract, compensate, and retain the best and brightest.<sup>1</sup> Why do so many investors and traders flock to such dealers despite the obvious adverse selection concerns associated with their superior expertise? Why don't dealers try to develop a reputation for lacking the private information and skills used to take advantage of their counterparties, thereby minimizing adverse selection concerns?

In this paper, we jointly model dealers' expertise acquisition and investors' order-flow allocation in OTC markets. Dealers acquire costly expertise to improve their ability to value assets they trade with investors — expertise allows dealers to take advantage of the less informed investors that send them order flow. Investors, however, take into account the expertise each dealer is expected to use against them when they choose how to split order flow across dealers.

An important and novel feature of our analysis consists of how we model dealers' expertise to capture the limited information spillovers across assets and transactions. In most models of information acquisition, a trader's expertise level dictates the informational *quality* of the signals this trader can observe for all transactions that take place. While this assumption might be realistic for highly specialized dealers or commodity markets, in many other settings a dealer's acquisition

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<sup>1</sup>See OCC (2021), Campbell (2018), and Goldman Sachs' 2021-Q4 earnings report for the specific numbers in the sentence. For evidence of concentration in OTC intermediation for various types of assets, see Cetorelli et al. (2007), Atkeson, Eisfeldt, and Weill (2014), Begenau, Piazzesi, and Schneider (2015), Di Maggio, Kermani, and Song (2017), Li and Schürhoff (2019), Siriwardane (2019), and Hendershott et al. (2020), among many others.

of superior information about a specific asset or transaction is likely to have limited pricing implications for other assets and transactions. For example, the majority of corporate bonds trade less than twice a year (see Chaderina, Muermann, and Scheuch 2022) while differences in maturity, duration, covenants, and collateral pledged render even bonds from the same issuer imperfect substitutes. Thus, if a dealer acquires an informational advantage about a specific corporate bond transaction, it does not imply that this dealer will also benefit from a similar advantage in its future transactions.

In our model, we instead assume that a dealer's expertise level dictates the *quantity* of transactions for which this dealer will have the resources needed to enjoy an informational advantage over its counterparties. By hiring smart traders, purchasing fast computers, and gaining access to various proprietary databases, a dealer ends up increasing its capacity or bandwidth to take advantage of counterparties across multiple transactions involving various types of assets within a short period of time. When investors are interested in trading different assets (e.g., corporate bonds, municipal bonds, and derivative products linked to different entities), a dealer's resources must be split to assess the terms of trade of all the proposed transactions. Thus, as more order flow gets directed to this dealer, its resources are spread out more thinly across potential trades, thereby weakening this dealer's ability to gain an informational advantage and assess the terms of trade associated with each proposed transaction. These liquidity externalities are the first of two essential forces in our model.

As hinted earlier, our model features a tension between investors' and dealers' preferences for dealers' expertise levels. *Ceteris paribus*, investors prefer to trade with dealers possessing low levels of expertise, as it reduces adverse selection concerns. Dealers, on the other hand, prefer to have access to high levels of expertise when trading with investors, as it allows them to extract a larger share of the surplus from trade. We show that, when the cost of expertise is sufficiently low, trading is concentrated around one dealer that acquires a high level of expertise in equilibrium. Investors are then indifferent between trading with this sophisticated dealer whose expertise is

thinly spread among a large number of transactions (resulting in high liquidity) and trading with a less sophisticated dealer whose scarcer resources can be used to gain informational advantages in a small number of transactions (resulting in low liquidity). Thus, paradoxically, in equilibrium order flow is concentrated around a dealer that employs smarter traders, owns faster computers, and has better data. This outcome arises because dealers' expertise levels are endogenous responses to investors' expected allocation of order flow. A dealer expecting to intermediate a high transaction volume cannot credibly commit to not acquiring the expertise needed to take advantage of all its unsophisticated counterparties. An outcome where most order flow is concentrated around a less sophisticated intermediary cannot be sustained in equilibrium as the large number of investors expected to do business with this "central" intermediary renders expertise acquisition too profitable for this intermediary. The endogeneity of dealers' expertise levels is the second essential force in our model.

Our paper contributes to the literature that studies the allocation of order flow in OTC markets. In contrast with predictions from the large literature that assumes random matching in OTC markets, Hendershott et al. (2020) empirically show that order-flow matching patterns are highly persistent, thereby highlighting the need to understand how traders select their counterparties. Green (2007) theoretically analyzes dealers' incentives to take advantage of investors who cannot compare terms of trade across dealers without incurring search costs. Unlike us, Green (2007) does not study the endogenous acquisition of expertise by dealers in response to the expected volume of transactions or the liquidity externalities of order-flow concentration in light of adverse selection concerns. Chacko, Jurek, and Stafford (2008), Lester, Rocheteau, and Weill (2015) and Sambalaibat (2018) highlight the execution-speed implications of order-flow concentration without modeling dealers' optimal response in terms of expertise acquisition. Glode and Opp (2016) show that OTC trading networks that sequentially involve differentially informed intermediaries as part of a transaction can reduce inefficient behaviors caused by asymmetric information. Our paper shows that order-flow concentration can also reduce trading inefficiencies due to a dealer's

superior information, thereby rationalizing why Li and Schürhoff (2019) find that the majority of OTC order flow is concentrated around central dealers, and why their involvement tends to reduce the number of intermediaries needed to complete a transaction. Pagano (1989) and Chaderina and Green (2014) endogenize market participation in light of liquidity externalities, but do not consider the role played by dealers' endogenous expertise. Babus and Hu (2017) show the optimality of a star network when intermediaries keep track of their counterparties' past actions and discipline them in case of misbehavior.

Like us, Li and Song (2022) jointly model dealers' expertise acquisition and investors' order-flow allocation in OTC markets. However, in their model, a dealer's superior information is directly shared with its customers, who can then make better portfolio decisions thanks to the expertise acquired by their chosen dealers. Whereas informed dealers effectively act as brokers or advisors in Li and Song (2022), they act as trading counterparties in our model, meaning that they use their expertise to take advantage of the investors who choose to transact with them.

Our paper also relates to the literature that studies information acquisition in OTC markets. Glode, Green, and Lowery (2012) highlight OTC traders' incentives to overinvest in expertise prior to their trading interactions in order to take advantage of their counterparties, but assume an exogenous order-flow allocation. Glode and Opp (2020) show how predictable trading interactions can incentivize costly information acquisition by OTC traders. Galeotti and Goyal (2010) and Herskovic and Ramos (2020) study network formation games where information is shared within connections, leading to complementarities and concentration. All these papers are, however, silent about the liquidity externalities of order-flow concentration in light of adverse selection concerns, which are the focus of our analysis.

## 2 Model

Our model has two stages. In the first stage, dealers acquire costly expertise while investors choose the dealers with whom they will trade. In the second stage, trade takes place between investors and dealers. We now introduce the trading game and analyze agents' optimal trading behavior. Using these results, we then solve for the optimal levels of expertise that dealers acquire and the optimal allocations of order flow that investors choose.

### 2.1 Trading Stage

Since the focus of our paper is on how dealers and investors behave in the first stage, we keep our model of the second stage as simple as possible. While the model makes formal assumptions in terms of how trading occurs among agents, we will later emphasize the generic properties of trading games that our central insights rely on.

Consider a trading game between a prospective buyer and the current owner of a financial asset. The current owner values the asset for its common value  $v$ , whose realization can either be  $v_l$  or  $v_h$  ( $> v_l$ ) with equal probabilities based on public information. The expected value of the asset is then:  $E(v) \equiv \frac{v_h + v_l}{2}$ . The prospective buyer reaches out to the current owner of the asset (a.k.a., the seller) because, in addition to its common value  $v$ , the buyer would collect a private benefit  $b > 0$  from acquiring and holding the asset (e.g., due to unmodeled liquidity or diversification reasons). The existence of gains to trade  $b$  is public knowledge and implies that trade taking place is welfare improving.

Before the transaction occurs, the seller receives a signal  $s \in \{v_l, v_h\}$  about the value of the asset which is correct with probability  $\alpha = \frac{1}{2} + e$ . We use  $e_i \in [0, \frac{1}{2}]$  to denote the seller's level of expertise for this specific transaction. If  $e = 0$ , the seller's signal is uninformative about the value of the asset, whereas if  $e = \frac{1}{2}$ , the seller knows perfectly the value of the asset. Both the seller and the buyer know the level of  $e$ . While the seller's transaction-specific expertise  $e$  is taken as given in

this stage, we will later formalize dealers' incentives to invest in expertise (e.g., technology, human capital, and data) and boost their  $e$  as a function of the expected allocation of investors' order flow.

To avoid signaling and equilibrium multiplicity concerns, we assume that the buyer makes a take-it-or-leave-it offer to purchase the asset at a price  $P$  from the potentially privately-informed seller. When deciding which price  $P$  to offer in exchange for an asset, the prospective buyer faces an intuitive tradeoff. Offering a higher price means that the buyer is more likely to get the asset and realize the gains to trade  $b$ . However, offering a higher price also means that, conditional on getting the asset, the buyer shares more of its surplus with the seller.

Specifically, the buyer considers offering a price:

$$P_h = \alpha v_h + (1 - \alpha)v_l = E(v) + e(v_h - v_l), \quad (1)$$

which is equal to how much the seller would value the asset after observing a signal  $s = v_h$ . If this price is offered, the seller accepts to trade with the buyer regardless of whether the signal is  $s = v_h$  or  $s = v_l$ , which is the socially optimal trading outcome. The buyer also considers offering a price:

$$P_l = \alpha v_l + (1 - \alpha)v_h = E(v) - e(v_h - v_l), \quad (2)$$

which is equal to how much the seller would value the asset after observing a signal  $s = v_l$ . The seller accepts this price after observing a signal  $s = v_l$  but not after observing a signal  $s = v_h$ , which means that the gains to trade are destroyed half of the time.

Overall, the buyer's expected surplus from offering a price  $P \in \{P_l, P_h\}$  is:

$$\begin{aligned} \Pi(\alpha, b, P) &\equiv \begin{cases} \frac{1}{2}(1 - \alpha)(v_h + b - P_l) + \frac{1}{2}\alpha(v_l + b - P_l) & \text{if } P = P_l, \\ \left[\frac{1}{2}\alpha + \frac{1}{2}(1 - \alpha)\right](v_h + b - P_h) + \left[\frac{1}{2}(1 - \alpha) + \frac{1}{2}\alpha\right](v_l + b - P_h) & \text{if } P = P_h, \end{cases} \\ &= \begin{cases} \frac{b}{2} & \text{if } P = P_l, \\ b - e(v_h - v_l) & \text{if } P = P_h. \end{cases} \end{aligned} \quad (3)$$

A buyer never finds it optimal to offer either  $P > P_h$  (which is dominated by offering  $P = P_h$ ) or  $P < P_l$  (which is dominated by offering  $P = P_l$ ), neither the buyer ever offers  $P_l < P < P_h$  as it is strictly dominated by offering  $P_l$  which yields the same probability of trade but at a lower cost to the buyer. We now characterize the buyer's optimal bidding strategy when facing a seller with transaction-specific expertise  $e$ .

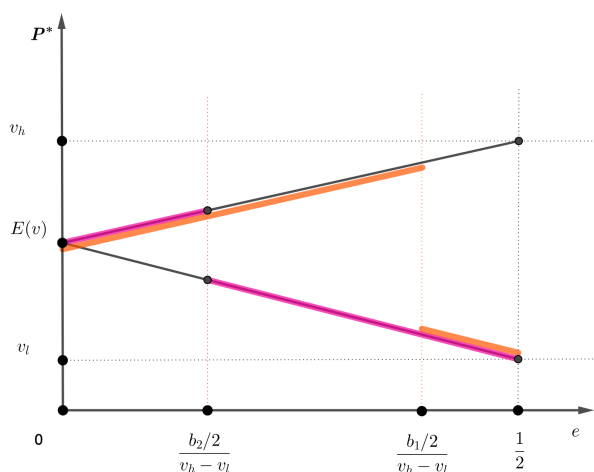
**Proposition 1.** *In equilibrium, the buyer offers the price:*

$$P^* = \begin{cases} P_l = E(v) - e(v_h - v_l) & \text{if } e > \frac{b/2}{v_h - v_l}, \\ P_h = E(v) + e(v_h - v_l) & \text{if } e \leq \frac{b/2}{v_h - v_l}, \end{cases} \quad (4)$$

Throughout the paper, proofs of our formal results are relegated to the Appendix. As Proposition 1 shows, the solution for  $P^*$  allows for two cases that differ in how adverse selection impacts the liquidity of trade. For low levels of  $e$ , adverse selection concerns are mild and, as a result, the buyer finds it optimal to make a high offer  $P_h$  that the seller accepts with probability one. In this case, liquidity is high and the full surplus from trade is split between the buyer and the seller through the optimal bid  $P^*$ , which is increasing in the seller's expertise level  $e$ . However, for high levels of expertise, the seller's informational advantage becomes too costly for the buyer who prefers to offer the low price  $P_l$ , making the optimal bid  $P^*$  decreasing in the seller's expertise



level  $e$ . In this case, liquidity is low as half of the surplus from trade is destroyed due to adverse selection concerns. Altogether, the seller benefits from boosting its transaction-specific expertise when the buyer is willing to offer  $P_h$ , but having too much of that expertise can impede trade and result in the seller being worse off. Moreover, the buyer's incentives to offer a high price that sustains liquid trading are increasing in  $b$ , the private benefit of holding the asset. These comparative statics are illustrated in Figure 1, which plots the buyer's optimal bid as a function of the seller's transaction-specific expertise level  $e$  for two levels of private benefit.



**Figure 1**

Buyer's Optimal Bidding Function. This graph illustrates the two cases of the buyer's optimal bidding function, as formalized by Proposition 1. The x-axis represents the seller's expertise level and the y-axis represents the price the buyer finds optimal to offer to the seller. The two vertical lines highlight the levels of expertise for which trading breaks down with probability 1/2 for two different levels of private benefit from trade  $b_2 < b_1$ .

The next corollary establishes that the buyer is made better off by being paired with a seller with lower transaction-specific expertise.

**Corollary 1.** *If  $e \leq \frac{b/2}{v_h - v_l}$ , the buyer's expected surplus  $\Pi$  is decreasing in the seller's expertise level  $e$ . If  $e > \frac{b/2}{v_h - v_l}$ , the buyer's expected surplus  $\Pi$  is unaffected by the seller's expertise level  $e$ .*

For the remainder of the paper, we impose a parametric restriction on  $b$  that results in adverse selection impeding trade for high enough levels of transaction-specific expertise:

**Assumption 1.**  $0 \leq b < v_h - v_l$ .

Imposing  $b < v_h - v_l$  guarantees that  $\frac{b/2}{v_h - v_l} < \frac{1}{2}$  and that  $P^* = P_l$  is the optimal bidding strategy in the region where  $\frac{b/2}{v_h - v_l} < e \leq \frac{1}{2}$ . This restriction will help capture the notion that dealers must limit how much expertise they acquire given how investors respond to the risk of being adversely selected.

## 2.2 Expertise Acquisition and Order-Flow Allocation Stage

Using the trading game analyzed above, we can study how dealers choose how much expertise they acquire and how investors choose the dealers with whom they trade. While the analysis of the trading stage in subsection 2.1 focused on the seller being privately informed, the payoff functions we derived would not change if we instead assumed that an uninformed seller quotes a price to a buyer with expertise level  $e$ . Thus, we can interpret the buyer's expected profit  $\Pi$  more broadly as the payoff of an unsophisticated investor trading with an expert dealer, regardless of who is long and who is short the asset. To capture the idea that dealers' investments in expertise are made in anticipation of investors' order-flow allocation and investors' order flow is allocated in anticipation of dealers' expertise levels, we solve for a Nash equilibrium when investors' and dealers' decisions are made simultaneously.

As alluded to in our introduction, our model is designed to capture the notion that a dealer's investment in expertise improves its capacity to take advantage, within a short period of time, of various counterparties across multiple transactions involving different types of assets. A financial firm cannot value all possible assets and assess the terms of trade of all deals offered simply by hiring one trader, purchasing one computer, or gaining access to one proprietary database. The more resources it invests to boost its expertise, the higher is the dealer's bandwidth for assessing the profitability of a variety of proposed deals. To depict in a tractable manner this idea that more scalability requires additional investments, we assume that the market is fragmented in the sense

that there are multiple investors each interested in trading a different asset (e.g., a specific corporate bond, municipal bond, or derivative product). Each dealer starts with  $\underline{k}$  units of total expertise (e.g., data, human capital, computing power) to value assets and gain an informational advantage over investors. Each dealer  $j$  can then boost its total expertise to  $k_j > \underline{k}$  at a cost  $c \cdot (k_j - \underline{k})$ .

Each dealer is assumed to start with a loyal client base of measure  $\underline{n}$ . These are unsophisticated investors who always trade with the same (e.g., local) dealer, in the spirit of Green (2007). We refer to the  $N$  investors who are not loyal to one specific dealer and who optimize based on the expected terms of trade as independent investors. A dealer  $j$ 's total order flow  $n_j \equiv \underline{n} + n_j^*$  combines the transaction volume from its loyal investors and from the independent investors who chose this dealer. In the simultaneous-move game of this stage, each independent investor chooses which dealer to do business with and each dealer chooses how much to invest in boosting its total expertise. When considering doing business with a dealer  $j$ , investor  $i$  conjectures the transaction-specific expertise  $e_j$  that this dealer will be able to use to gain an informational advantage. Then trade occurs as described in subsection 2.1, taking these decisions as given.

Dealer  $j$ 's transaction-specific expertise is assumed to be  $e_j = \frac{1}{2} \min\left(\frac{k_j}{n_j}, 1\right)$ , reflecting the dealer  $j$ 's total expertise  $k_j$  normalized by its total order flow  $n_j$  while ensuring that the signal's probability of being correct,  $\frac{1}{2} + e_j$ , cannot be greater than one. When dealer  $j$  receives weakly less order flow  $n_j$  than its expertise capacity  $k_j$ , dealer  $j$  is able to produce a perfect signal about the value of the asset traded with each investor. However, if the order flow  $n_j$  is larger than the expertise capacity  $k_j$ , dealer  $j$ 's resources are spread out across all transactions and the resulting signals produced for each transaction are imperfect. Dealers are allowed to be short or long any asset in their inventory (recall: the model behaves symmetrically whether the dealer is a buyer or a seller).

Substituting the optimal bid  $P^*$  derived in Proposition 1 into the dealer's profit function, we

can write the dealer's total expected profit as:

$$\Delta(k_j, n_j) = \begin{cases} -c \cdot (k_j - \underline{k}) & \text{if } e_j > \frac{b/2}{v_h - v_l}, \\ n_j e_j (v_h - v_l) - c \cdot (k_j - \underline{k}) & \text{if } e_j \leq \frac{b/2}{v_h - v_l}. \end{cases} \quad (5)$$

The next lemma shows that when expertise is sufficiently cheap to acquire dealer  $j$  is better off increasing  $k_j$  until  $e_j = \frac{b/2}{v_h - v_l}$ .

**Lemma 1.** For a given  $n_j$ ,  $\frac{\partial \Delta(k_j, n_j)}{\partial k_j} > 0$  as long as  $e_j < \frac{b/2}{v_h - v_l}$  and  $c < \frac{1}{2}(v_h - v_l)$ .

A dealer benefits in two ways from increasing its expertise in our setting. A first benefit comes from making better decisions when responding to offers. This benefit is most evident in the range of  $e_j > \frac{b/2}{v_h - v_l}$  where investors are so concerned about adverse selection that they make offers that result in illiquid trade. For a given price offer, the more accurate the signal is, the more profitable is the dealer's decision whether to refuse or execute the transaction. However, the improved-decision benefit is offset by worse prices being offered by investors who grow more concerned about being adversely selected. A second benefit of expertise is triggered when  $e_j \leq \frac{b/2}{v_h - v_l}$ . In that region, if the dealer's signal becomes marginally more precise, investors respond by making more generous offers to ensure that the dealer agrees to trade with probability one.

While our model makes formal assumptions about how trading occurs among agents, the central insights we highlight throughout the paper mainly rely on the generic property that, ceteris paribus, a dealer's expertise increases the surplus from trade it can appropriate and decreases the surplus from trade its counterparties can retain.

An equilibrium of the expertise acquisition and order-flow allocation stage is defined by dealers' expertise choices  $k_j^* \geq \underline{k}$  and investors' dealer choices  $Dealer_i$  such that:

- for any investor  $i$  that chooses to rout its trade to dealer  $j$ , i.e.,  $Dealer_i = j$ , we have:

$$\Pi \left( \frac{1}{2} + \frac{1}{2} \min \left( \frac{k_j^*}{n_j}, 1 \right), b, P_j^* \right) \geq \Pi \left( \frac{1}{2} + \frac{1}{2} \min \left( \frac{k_{j'}^*}{n_{j'} + 1}, 1 \right), b, P_{j'}^* \right) \quad \forall j' \neq j, \quad (6)$$

where  $P_j^*$  and  $P_{j'}^*$  denote the optimal price offers from investors when trading with dealer  $j$  and dealer  $j'$  respectively (see Proposition 1),

- for any dealer  $j$  that chooses to acquire an expertise capacity  $k_j^*$  and expects to receive order flow  $n_j$  as a result of investors' dealer choices  $Dealer_i$ , we have:

$$\Delta(k_j^*, n_j) \geq \Delta(k_{j'}^*, n_j) \quad \forall k_{j'}^* \neq k_j^*. \quad (7)$$

As already stated, we model dealers' and investors' decisions as a simultaneous-move game. This timeline captures the idea that expertise acquisition and order-flow allocation are endogenous to each other and none of these decisions unilaterally drives the other (as they would if they were sequential). While the initial level of expertise as well as the loyal client base of each dealer are well-known to all market participants, we assume that investors cannot know for sure dealers' future levels of expertise, whereas dealers cannot be certain of investors' future allocation of order flow. Our equilibrium definition imposes that there is no systematic deviation between conjectured and realized outcomes when agents make their decisions.

For the remainder of this section, we make the following assumptions about dealers and loyal investors:

**Assumption 2.** *All dealers are ex-ante identical with respect to their initial levels of expertise  $\underline{k}$  and order flow from loyal investors  $\underline{n}$ . Moreover,  $\underline{k} \geq \underline{n} + 1$ .*

We now introduce the basic intuition for why concentrating order flow around one dealer can be optimal for independent investors.

**Lemma 2.** *If in equilibrium at least one dealer whose transaction-specific expertise is low enough to allow for liquid trading (i.e., there is at least one dealer whose  $e_j \leq \frac{b/2}{v_h - v_l}$ ), then all independent investors' order flow is allocated to the same dealer and, as a result, other dealers only receive order flow from their loyal investors.*

Intuitively, this lemma shows, by contradiction, that if independent investors were expected to split their transaction volume among two or more dealers, then each independent investor would benefit from re-routing its order flow to a different dealer. By doing so, a deviating investor would decrease the expertise available to be used by the targeted dealer for each transaction (i.e., this dealer's  $e_j$ ). Therefore, it would be in the best interest of every independent investor to allocate its order flow to a single dealer in order to stretch this dealer's expertise capacity and minimize its informational advantage in each transaction. The existence of one dealer whose  $e_j \leq \frac{b/2}{v_h - v_l}$  ensures that while an additional independent investor switching to dealer  $j$  would further decrease its  $e_j$ , it would also benefit this investor as well as all other investors that trade with this dealer. Recall from Corollary 1 that an investor's trading profit is decreasing in  $e_j$  whenever  $e_j \leq \frac{b/2}{v_h - v_l}$  and trading remains liquid.

Lemma 2 thus highlights how liquidity externalities can incentivize independent investors to concentrate their order flow around one dealer. Each investor prefers to trade with a dealer that is attracting a lot of other independent investors, because this dealer is less likely to use its expertise capacity to take advantage of this specific investor. However, as we show in the next proposition, dealers take the allocation of order flow into account when choosing how much expertise to acquire. But before deriving this result, we need to rule out extreme outcomes by imposing a few restrictions on the aggregate distribution of dealers and investors.

**Assumption 3.** *The market is populated with  $L$  dealers and  $N$  independent investors, such that*

$$\frac{k}{n + \frac{N}{L}} \leq \frac{b}{v_h - v_l}.$$

This restriction limits how much expertise dealers start with for the case where aggregate order flow is split evenly across all dealers. Dealers' initial level of expertise  $\underline{k}$  is low enough to allow for liquid trading as investors still find it optimal to make generous offers, despite their adverse selection concerns. This restriction also implies that, if all independent investors were to choose to do business with a single dealer, spreading the initial level of expertise  $\underline{k}$  across this larger transaction volume would also result in a low enough transaction-specific expertise  $e_j$  to yield liquid trade.

In the next proposition we combine the insights of Lemmas 1 and 2 with Assumption 3 and pin down the equilibrium allocation of order flow and the equilibrium levels of dealer expertise.

**Proposition 2.** *When  $c < \frac{1}{2}(v_h - v_l)$ , in equilibrium one dealer, say  $j^*$ , receives all the order flow from independent investors and acquires the highest level of expertise that allows trading to remain liquid, whereas all other dealers trade only with their loyal investors and do not acquire additional expertise above the initial level  $\underline{k}$ .*

Proposition 2 establishes that, when  $c < \frac{1}{2}(v_h - v_l)$ , in equilibrium order flow is concentrated around one central dealer that makes the largest investment in expertise among all dealers. While its investors would prefer this dealer to have a weaker informational advantage, alternative dealers who acquire less expertise than the central dealer also happen to participate in fewer transactions. In fact, Assumption 2 implies that these alternative dealers have excess expertise bandwidth they could use against any deviating independent investor — deviating by reallocating an investor's order flow towards them would result in illiquid trading. From Lemma 1, we know that for a small enough cost parameter  $c$  a dealer benefits from increasing its expertise as long as it does not prevent trading to remain liquid. Thus, the central dealer optimally reaches their expertise cutoff given their equilibrium level of order flow, whereas alternative dealers have some unused expertise and are in position to use their limited resources to take advantage of any investor that trades with them. As a result, independent investors are indifferent between doing business with

the expert dealer  $j^*$  that attracts all independent investors' order flow and doing business with any of the other dealers whose levels of expertise and order flow are limited — in both cases, investors collect an expected profit of  $\frac{b}{2}$  as shown in subsection 2.1.

Proposition 2 thus highlights that in equilibrium the large investments in expertise made by the central dealer do not deter independent investors from sending their order flow its way. Investors recognize that trade is more likely to happen with the central dealer than with peripheral dealers who possess too much expertise for the limited order flow that they obtain from loyal investors. In equilibrium, investors trading with the central dealer have their offers accepted with probability one, implying maximal levels of liquidity and surplus from trade, whereas peripheral dealers accept to trade with their investors only half of the time, implying inefficiently low levels of liquidity and surplus from trade.

We can interpret this equilibrium as featuring a sophisticated (i.e., expert) dealer attracting the transaction volume of attentive (i.e., independent) investors and multiple less sophisticated dealers benefiting from the inattentiveness of their loyal investor clienteles. Yet, for sufficiently low  $c$ , investors are indifferent about which dealer they trade with in equilibrium because all dealers have  $e_j = \frac{b/2}{v_h - v_l}$  or above. The central dealer provides liquid trading and earns a positive expected profit, whereas peripheral dealers provide illiquid trading and earn no profit.

We now analyze how the cost of expertise more generally affects the central dealer's investments in expertise, while taking into account the endogeneity of order flow. As discussed above, a dealer benefits from acquiring expertise through the better terms of trade investors offer and through the ability to make better decisions in response to these offers. In order to decide how much expertise to acquire, a dealer must then compare these benefits with the cost of expertise. The proposition that follows generalizes Proposition 2 to allow for  $c \geq \frac{1}{2}(v_h - v_l)$ .

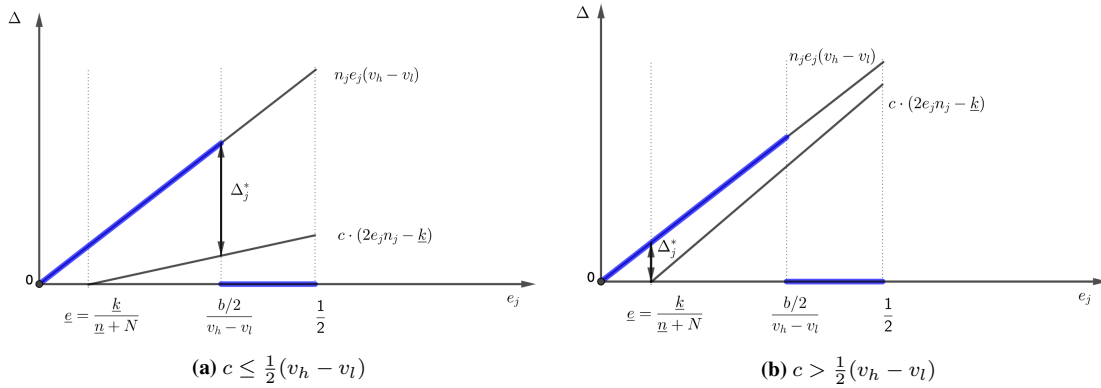
**Proposition 3.** *In equilibrium, the central dealer's level of transaction-specific expertise is:*



$$e^* = \begin{cases} \underline{e} & \text{if } c > \frac{1}{2}(v_h - v_l), \\ \frac{b/2}{v_h - v_l} & \text{if } c \leq \frac{1}{2}(v_h - v_l), \end{cases} \quad (8)$$

where  $\underline{e} \equiv \frac{1}{2} \left( \frac{k}{n+N} \right)$ .

This proposition shows that the central dealer's equilibrium level of expertise is weakly decreasing in the cost of expertise. This prediction may seem natural at first, but the analysis must account for how the dealer's choice of expertise is complicated by investors' optimal bidding response. If investors expect their dealer to have a low level of expertise, they offer good terms of trade to ensure that the dealer agrees with probability one, thereby maximizing the liquidity of trade. Acquiring a marginally higher level of expertise then solely benefits the dealer by improving the terms of trade and thereby re-allocating a higher share of the surplus towards the dealer. Acquiring substantially more expertise might, however, trigger the opposite response. If the dealer's transaction-specific expertise is high, adverse selection concerns are too severe and investors offer worse terms of trade, resulting in illiquid trading. At that point, the benefits of higher expertise are guaranteed to be lower than the costs.



**Figure 2**

Benefits and Costs of Acquiring Expertise. Panel (a) compares the dealer's gross profit from trade to the cost of expertise when the cost of expertise is low. Panel (b) compares the dealer's gross profit from trade to the cost of expertise when the cost of expertise is high.

Figure 2 illustrates the benefits and costs of expertise acquisition for high and low levels of

c. Panel (a) shows that when it is inexpensive to boost expertise, the central dealer acquires the maximum level of expertise that still allows for liquid trading. The graph illustrates that in this case the dealer earns a sizable profit. When the expertise costs are high, however, the central dealer finds it optimal not to invest in expertise. Panel (b) illustrates that in this case the central dealer still earns a positive profit, albeit smaller than in panel (a).

Overall, a central dealer finds it optimal to boost its expertise because it allows to make better decisions and receive better terms of trade from investors. The dealer compares the resulting higher trade payoff with the cost of acquiring expertise. In equilibrium, the dealer acquires a level of expertise that does not discourage order flow from independent investors, who recognize that other (less busy) dealers would also take advantage of them as their lower expertise bandwidth is not spread as thinly. In all cases, the equilibrium can still be thought of featuring a sophisticated (i.e., expert) dealer providing high liquidity to a large pool of attentive investors and multiple less sophisticated dealers only attracting their loyal clienteles and trading less frequently with each of them, thereby providing less liquidity. Moreover, when the cost of expertise is high, independent investors are collecting strictly higher trading profit from their central dealer, who provide high liquidity, than loyal investors are getting from their peripheral dealers, who provide low liquidity.

### **3 Extensions and Model Implications**

We now extend our baseline analysis to enrich the cross-sectional and time-series implications of our model.

#### **3.1 Payment for Order Flow**

In the equilibrium of our baseline model, peripheral dealers collect no profit from trading with their loyal investors, in contrast with models like Green (2007) where dealers take advantage of their loyal/naive clienteles. This difference is due to the fact that while by construction loyal

investors cannot switch dealers in either setting, in our setting loyal investors can still worsen the terms of trade they offer if their local dealer happens to possess a high level of expertise. As an implication of this result, if we were to allow dealers to pay for order flow, peripheral dealers would be unwilling to offer any compensation aimed at attracting independent investors' order flow, unless the additional order flow was so large that it would reduce dealer expertise below the point where  $e_j = \frac{b/2}{v_h - v_l}$  and trading becomes illiquid. Thus, subject to an intuitive parametric restriction, the concentrated allocation of order flow around a central dealer would survive in an environment where payment for order flow is allowed.

### 3.2 Ex Ante Dealer Heterogeneity

In the baseline analysis, we showed that the equilibrium allocation of order flow disproportionately favors one dealer even though dealers are homogeneous ex ante. When the cost of expertise is small enough, the central dealer finds it optimal to acquire more expertise than its peers. This prediction rationalizes why investors tend to (paradoxically) allocate their order flow to the most sophisticated dealers, despite standard adverse selection concerns. A natural limitation of our equilibrium predictions so far is that any of the ex-ante identical dealers can be selected by investors to become the market leader – our analysis has been silent about how independent investors may coordinate on picking a single dealer and how this chosen dealer knows to expect more order flow when deciding how much expertise to acquire. To shed light on how the ex-ante characteristics of the dealer that turns out ex post to be the central one impact investor welfare, we now allow for ex-ante heterogeneity in the per-unit cost of expertise acquisition and rank possible equilibria in terms of investors' welfare. We denote by  $c_j$  the cost of acquiring expertise for dealer  $j$ .

**Proposition 4.** *Independent investors weakly prefer an equilibrium in which the dealer with the highest cost of expertise  $c_j$  is the central one.*

Proposition 3 shows that the central dealer's level of expertise is weakly decreasing in  $c$ , while

Corollary 1 shows that investors weakly prefer to trade with dealers with a lower  $e_j$ . It is therefore intuitively clear that independent investors are weakly better off when the central dealer is the one facing the highest cost of expertise.

### 3.3 Increasing Accessibility of OTC Markets

In the baseline analysis, we assumed a market populated by many independent investors. In particular, we focused on the homogenous case where  $\underline{k} < \frac{b}{v_h - v_l} \left( \underline{n} + \frac{N}{L} \right)$ , meaning that there were many more investors and corresponding total order flow (i.e.,  $L\underline{n} + N$ ) than dealers' combined expertise before considering any investment (i.e.,  $L\underline{k}$ ). We now relax this restriction to shed light on the implications of decreasing the number of independent investors or increasing the initial level of dealer expertise.

In what follows, we analyze an equilibrium for the case with ex-ante homogenous dealers and  $\underline{k} > \left( \underline{n} + \frac{N}{L} + 1 \right) \left( \frac{b}{v_h - v_l} \right)$ . When this condition holds, all dealers start with a level of expertise that would result in illiquid trading if independent investors were to allocate their transactions evenly among dealers. The following proposition demonstrates that an even allocation of order flow can now be observed in equilibrium.

**Proposition 5.** *If  $\underline{k} > \frac{b}{v_h - v_l} \left( \underline{n} + \frac{N}{L} + 1 \right)$ , there exists an equilibrium in which all independent investors allocate their order flow evenly among dealers and no dealer acquires any additional expertise.*

Proposition 5 describes a market outcome in sharp contrast to the one discussed in the baseline analysis: we now have an even allocation of order flow across all dealers. Independent investors know that all dealers will have informational advantages that result in illiquid trading, yet since such business model is prevalent, investors have no incentives to switch dealers.

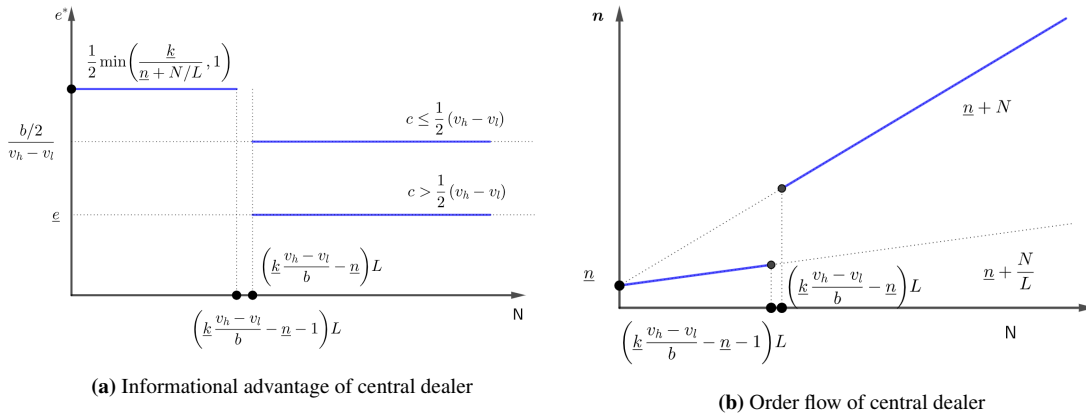
Unlike when Assumption 3 holds, here none of the dealers expects to receive enough order flow to warrant acquiring additional expertise. The dispersed allocation of order flow in this equilibrium

does not, however, benefit independent investors. That is, they are all trading with dealers who are so informed that their investors only earn profits of  $\frac{b}{2}$  in equilibrium, due to the illiquidity of trade.

Figure 3 illustrates how the informational advantage and order-flow allocation of a central dealer change when the number of independent investors crosses the threshold:

$$\left[ \frac{k}{b} \left( \frac{v_h - v_l}{b} \right) - \underline{n} - 1 \right] L. \quad (9)$$

When the number of independent investors is small, an even allocation of order flow across all dealers is sustainable in equilibrium. Panel (a) shows that  $e^* > \frac{b/2}{v_h - v_l}$  for small  $N$ , while Panel (b) shows that each dealer's order flow is  $\underline{n} + \frac{N}{L}$ . As the number of independent investors increases beyond the threshold above, this equilibrium is no longer sustainable. The order flow concentrates around a central dealer, which receives a total order flow  $\underline{n} + N$  and either acquires enough expertise to have the most precise signal without impeding liquid trading (i.e.,  $e^* = \frac{b/2}{v_h - v_l}$ ), or does not acquire any extra expertise (i.e.,  $e^* = \underline{e}$ ), depending on the cost of expertise.



**Figure 3**

Equilibrium Impact of Increasing the Number of Independent Investors in the Economy. Panel (a) plots the quality of the central dealer's information signal as a function of the number of independent investors, for two levels of expertise cost. Panel (b) plots the transaction volume allocated to a central dealer as a function of the number of independent investors.

This extension can thus shed light on the consequences of the recent liberalization of OTC markets. Our results suggest that a disproportionately faster growth in the number of investors

than in the initial level of expertise for dealers (e.g.,  $N$  growing faster than  $\underline{k}$ ) could have led to increased concentration of order flow. Moreover, this inflow of participants may have increased the central dealer's incentives to acquire expertise above and beyond what is or was standard in the industry. Yet, since the dispersed equilibrium described in Proposition 5 features  $e^* > \frac{b/2}{v_h - v_l}$  for all dealers whereas the concentrated equilibrium from Proposition 2 may feature  $e_{j^*} \leq \frac{b/2}{v_h - v_l}$  for some levels of expertise acquisition costs, independent investors may or may not have benefited from this liberalization. We formalize this result in the corollary that follows.

**Corollary 2.** *An increase in the number of independent investors from  $N < \left[ \underline{k} \left( \frac{v_h - v_l}{b} \right) - \underline{n} - 1 \right] L$  to  $N > \left[ \underline{k} \left( \frac{v_h - v_l}{b} \right) - \underline{n} \right] L$  benefits independent investors if and only if  $c > \frac{1}{2}(v_h - v_l)$ .*

While new investors may greatly benefit from gaining access to OTC markets, our analysis in Corollary 2 shows that incumbent investors are made better off only if the dealers' costs of expertise are high enough. In particular, with high enough costs, the central dealer is unwilling to invest in expertise and reach  $e = \frac{b/2}{v_h - v_l}$  following an increase in OTC market accessibility. Yet, our analysis shows that potential advancements in technology, and the associated decreases in the costs of expertise dealers face, may have allowed a central dealer to extract a large fraction of the social surplus created by providing investors with easier access to OTC markets.

## 4 Conclusion

We jointly model dealers' acquisition of expertise and investors' allocation of order flow across dealers in OTC markets. An important and novel feature of our analysis consists of how we model dealers' expertise: a dealer's investment in expertise determines the resources that this dealer can allocate towards gaining an informational advantage over its counterparties across various transactions. We show that, under intuitive conditions, the equilibrium allocation of order flow is concentrated towards one dealer that invests significant resources to gain informational advantages

against its investors. Despite the adverse selection concerns associated with the dealer's large investments in expertise, investors prefer to funnel their transactions to this expert dealer rather than trading with less popular dealers who would provide less liquidity. Liquidity externalities arise in our model since order-flow concentration spreads out the resources that the central dealer can use to assess each proposed transaction, thereby weakening each investor's adverse selection concerns and increasing the liquidity of trade. Yet, expecting this high concentration of order flow, the central dealer finds it profitable to acquire the highest level of expertise that does not push investors away. Our analysis sheds light on the drivers behind the concentration of OTC intermediation as well as on the welfare implications of easier access to OTC markets.

# Appendix

## A Proofs Omitted from the Text

**Proof of Proposition 1:** From (3), it is optimal to bid  $P_l$  if and only if  $\Pi(\alpha, b, P_l) > \Pi(\alpha, b, P_h)$ .

Hence, the buyer's optimal bidding strategy can be written as:

$$P^* = \begin{cases} P_l & \text{if } e > \frac{b/2}{v_h - v_l}, \\ P_h & \text{if } e \leq \frac{b/2}{v_h - v_l}. \end{cases} \quad (\text{A1})$$

□

**Proof of Corollary 1:** Substitute the optimal bidding policy of the buyer into the buyer's profit:

$$\Pi(\alpha, b, P) = \begin{cases} \frac{b}{2} & \text{if } e > \frac{b/2}{v_h - v_l}, \\ b - e(v_h - v_l) & \text{if } e \leq \frac{b/2}{v_h - v_l}. \end{cases} \quad (\text{A2})$$

This is a weakly-decreasing function of  $e$  without jumps:

$$\frac{\partial \Pi(\alpha, b, P)}{\partial e} = \begin{cases} 0 & \text{if } e > \frac{b/2}{v_h - v_l}, \\ -(v_h - v_l) & \text{if } e \leq \frac{b/2}{v_h - v_l}. \end{cases} \quad (\text{A3})$$

□

**Proof of Lemma 1:** First of all, for a given level of order flow  $n_j$  and expertise bandwidth  $k_j$ , a dealer's expected profit, when the investor is the buyer and the dealer is the seller, is given by:

$$\Delta(k_j, n_j) \equiv \begin{cases} \frac{n_j}{2} [\alpha_j(P_l - v_l) + (1 - \alpha_j)(P_l - v_h)] - c \cdot (k_j - \underline{k}) & \text{if } P = P_l, \\ \frac{n_j}{2} [\alpha_j(P_h - v_h) + (1 - \alpha_j)(P_h - v_l) + \alpha_j(P_h - v_l) + (1 - \alpha_j)(P_h - v_h)] - c \cdot (k_j - \underline{k}) & \text{if } P = P_h. \end{cases}$$



where  $\alpha_j = \frac{1}{2} + e_j = \frac{1}{2} + \frac{1}{2} \min\left(\frac{k_j}{n_j}, 1\right)$  and  $b$  is from the range of parameters consistent with Assumption 1. We can substitute optimal bidding function from Proposition 1 to arrive at Equation (5). Then we can show that when  $e_j < \frac{b/2}{v_h - v_l}$ :

$$\frac{\partial \Delta(k_j, n_j)}{\partial k_j} = \frac{1}{2}(v_h - v_l) - c. \quad (\text{A4})$$

Thus, as long as  $c < \frac{1}{2}(v_h - v_l)$ ,  $\Delta(k_j, n_j)$  is strictly increasing in  $k_j$  as long as  $e_j < \frac{b/2}{v_h - v_l}$ .  $\square$

**Proof of Lemma 2:** Suppose, by contradiction, that given all dealers' choices of expertise, independent investors allocate their order flow among at least two dealers, i.e., there exist dealers  $j'$  and  $j''$  such that  $n_{j'}^* > 0$  and  $n_{j''}^* > 0$ . Then, if  $e_{j'} < e_{j''} < \frac{b/2}{v_h - v_l}$  or  $e_{j'} < \frac{b/2}{v_h - v_l} < e_{j''}$ , it is optimal for an independent investor who was planning to do business with dealer  $j''$  to switch to dealer  $j'$  because of the weaker informational advantage that dealer  $j'$  has in this specific transaction with this investor, following Corollary 1. Thus, this allocation of order flow cannot be part of an equilibrium. If instead  $\frac{b/2}{v_h - v_l} > e_{j'} > e_{j''}$  or  $e_{j'} > \frac{b/2}{v_h - v_l} > e_{j''}$ , the reverse is true, and similarly this order-flow allocation cannot be part of an equilibrium.

Now if  $e_{j'} = e_{j''} < \frac{b/2}{v_h - v_l}$ , an independent investor who was planning to do business with either dealer is strictly better off switching because  $\frac{k_{j'}}{n_{j'}+1} < \frac{k_{j''}}{n_{j''}-1}$  and  $\frac{k_{j'}}{n_{j'}-1} > \frac{k_{j''}}{n_{j''}+1}$ . Finally, if  $e_{j'} \geq \frac{b/2}{v_h - v_l}$  and  $e_{j''} \geq \frac{b/2}{v_h - v_l}$ , then an independent investor is better off switching to the dealer(s) whose  $e_j < \frac{b/2}{v_h - v_l}$  (which exist(s) given the statement of the lemma). Hence, no matter what the dealers' expertise levels are, it is always optimal for a subset of independent investors to reallocate their order flow as long as the order flow of independent investors is allocated among two or more dealers and at least one dealer in the economy has  $e_j < \frac{b/2}{v_h - v_l}$ .  $\square$

**Proof of Proposition 2:** To show that the stated equilibrium is indeed an equilibrium, we first rule out that a dealer  $j \neq j^*$  would acquire  $k_j > \underline{k}$  of expertise. When  $\underline{k} \geq \underline{n} + 1$ , this dealer's  $e_j$  is already greater than  $\frac{b/2}{v_h - v_l}$ , which means that this dealer's informational advantage creates so much adverse selection that investors find it optimal to offer prices that do not yield liquid trading (remember: dealers  $j \neq j^*$  do not attract any order flow from independent investors). Thus, dealer

$j$ 's profit is zero and there is no benefit to acquiring more expertise.

We next rule out that an independent investor would leave dealer  $j^*$  and allocate order flow to a different dealer. If that was the case, this investor would go from trading with a dealer with  $e_{j^*} = \frac{b/2}{v_h - v_l}$  to trading with a dealer with  $e_j = \frac{1}{2} \min\left(\frac{k}{\underline{n}+1}, 1\right) > \frac{b/2}{v_h - v_l}$ . Since an investor's profit is equal to  $\frac{b}{2}$  whenever its dealer's expertise satisfies  $e > \frac{b/2}{v_h - v_l}$ , thus, there would be no benefit to switching dealers.

Moreover, when  $c < \frac{1}{2}(v_h - v_l)$ , dealer  $j^*$  finds it optimal to increase its expertise level until  $e_{j^*} = \frac{b/2}{v_h - v_l}$  (see Lemma 1). Once dealer  $j^*$  reaches an expertise bandwidth of  $k_{j^*} = (\underline{n} + N) \frac{v_h - v_l}{b}$ , the transaction-specific expertise reaches the maximum level that does not impede full trade. At that point, there would be no benefit to acquiring more expertise. Thus, the outcome described in the proposition is indeed an equilibrium.

Next, we show that an outcome where two dealers acquire expertise above  $\underline{k}$  such that  $\frac{k_j}{n_j} = \frac{k_j^*}{n_j^*}$  cannot be an equilibrium. That is, we cannot have multiple dealers receiving order flow from independent investors and acquiring enough expertise to take full advantage of their order flow. In such a situation, similar to the logic in Lemma 2, independent investors would have an incentive to go from dealer  $j$  to  $j^*$  or the other way around. By switching from dealer  $j$  to  $j^*$ , investors would increase the order flow going to dealer  $j^*$  above this dealer's expertise capacity, thereby spreading its resources further and decreasing the average quality of its signals, i.e.,  $\frac{k_j}{n_j - 1} > \frac{k_j^*}{n_j^* + 1}$ .

Finally, Assumption 3 rules out as an equilibrium a situation where independent investors allocate their order flow equally among dealers that all have excess expertise capacity despite not making any investment, i.e.,  $\frac{k_j}{n_j} > \frac{b}{v_h - v_l}$  for all  $j$ . □

**Proof of Proposition 3:** Denote by  $n_{j^*} = \underline{n} + N$  the total order flow that the dealer selected by independent investors, say  $j^*$ , expects to receive. Recall from Equation (5) that the dealer's profit

is:

$$\Delta(k_{j^*}, n_{j^*}) = \begin{cases} -c \cdot (k_{j^*} - \underline{k}) & \text{if } e_{j^*} > \frac{b/2}{v_h - v_l}, \\ n_{j^*} k_{j^*} (v_h - v_l) - c \cdot (k_{j^*} - \underline{k}) & \text{if } e_{j^*} \leq \frac{b/2}{v_h - v_l}; \end{cases} \quad (\text{A5})$$

or alternatively,

$$\Delta(k_{j^*}, n_{j^*}) = \begin{cases} -c \cdot (k_{j^*} - \underline{k}) & \text{if } k_{j^*} > n_{j^*} \frac{b}{v_h - v_l}, \\ n_{j^*} k_{j^*} (v_h - v_l) - c \cdot (k_{j^*} - \underline{k}) & \text{if } k_{j^*} < n_{j^*} \frac{b}{v_h - v_l}. \end{cases} \quad (\text{A6})$$

Hence,  $\Delta(k_{j^*}, n_{j^*})$  is a piece-wise function of  $k_{j^*}$  for a given  $n_{j^*}$  with a jump at  $k_{j^*} = n_{j^*} \frac{b}{v_h - v_l}$ .

The cost parameter  $c$  determines the  $e_{j^*}$  that the central dealer aims to achieve in equilibrium, for a given level of order flow  $n_{j^*}$ . In particular, this dealer's marginal benefit from acquiring more expertise is given by  $\frac{\partial \Delta(k_{j^*}, n_{j^*})}{\partial k_{j^*}}$ . If  $k_{j^*} \leq n_{j^*}$  (thus:  $e_{j^*} \leq \frac{1}{2}$ ), this marginal benefit can be written as:

$$\Delta(k_{j^*}, n_{j^*}) = \begin{cases} -c & \text{if } k_{j^*} > n_{j^*} \cdot \frac{b}{v_h - v_l}, \\ \frac{1}{2}(v_h - v_l) - c & \text{if } k_{j^*} < n_{j^*} \cdot \frac{b}{v_h - v_l}. \end{cases} \quad (\text{A7})$$

If, however,  $k_{j^*} > n_{j^*}$ , the marginal benefit is:

$$\frac{\partial \Delta(k_{j^*}, n_{j^*})}{\partial k_{j^*}} = -c. \quad (\text{A8})$$

Hence, given the anticipated order flow of  $n_{j^*}$ , the optimal choice of  $k_{j^*} \in \left[ \underline{k}, n_{j^*} \cdot \frac{b}{v_h - v_l} \right]$ . In order to find the optimal  $k_{j^*}$  we need to see if the  $c$  is above or below  $\frac{1}{2}(v_h - v_l)$ . If  $c > \frac{1}{2}(v_h - v_l)$ , then  $\Delta(k_{j^*}, n_{j^*})$  as a function of  $k_{j^*}$  is globally decreasing, meaning that it is optimal for the dealer not to acquire any extra expertise. Then  $e^* = \underline{e} \equiv \frac{1}{2} \left( \frac{\underline{k}}{n+N} \right)$ .

If  $c \leq \frac{1}{2}(v_h - v_l)$ , then the first segment of the function is increasing, meaning that it is optimal for the dealer to invest in expertise until  $k_{j^*} = n_{j^*} \frac{b}{v_h - v_l}$ . Then  $e_j^* = \frac{b/2}{v_h - v_l}$ .

Altogether, the equilibrium features the following level of transaction-specific expertise for the

central dealer:

$$e^* = \begin{cases} \underline{e} & \text{if } c > \frac{1}{2}(v_h - v_l), \\ \frac{b/2}{v_h - v_l} & \text{if } c \leq \frac{1}{2}(v_h - v_l). \end{cases} \quad (\text{A9})$$

Since  $\underline{e} < \frac{b/2}{v_h - v_l}$  as implied by Assumption 3,  $e^*$  is weakly decreasing in  $c$ .  $\square$

**Proof of Proposition 4:** From Corollary 1 we know that independent investors weakly prefer to trade with a dealer with the smallest  $e_{j^*}$ , thereby minimizing exposure to the information advantage of a dealer. Since  $e_{j^*}$  is weakly decreasing in  $c_{j^*}$  according to Proposition 3, the higher the cost of acquiring expertise is for the dealer, the lower is the equilibrium level of informativeness of the central dealer. Hence, independent investors are better off if the central dealer is the dealer  $j$  with the highest  $c_j$ .  $\square$

**Proof of Proposition 5:** Consider a dealer  $j$  that receives  $\frac{N}{L}$  trades from independent investors and does not acquire additional expertise, i.e.,  $k_j = \underline{k}$ . The proposition statement implies that:

$$e_j = \frac{1}{2} \min \left( \frac{\underline{k}}{\underline{n} + \frac{N}{L}}, 1 \right) > \frac{b/2}{v_h - v_l}, \quad (\text{A10})$$

for all dealers. Hence, if an independent investor were to switch from dealer  $j$  to dealer  $j'$ , dealer  $j'$ 's transaction-specific expertise would become:

$$e_{j'} = \frac{1}{2} \min \left( \frac{\underline{k}}{\underline{n} + \frac{N}{L} + 1}, 1 \right) > \frac{b/2}{v_h - v_l}. \quad (\text{A11})$$

Hence, there is no incentive for any of the independent investors to deviate away from the current choice of dealers. Therefore,  $k_j = \underline{k}$  for all  $j$  and independent investors allocating evenly their order flow among dealers is an equilibrium.  $\square$

**Proof of Corollary 2:** Before the increase in  $N$ , the equilibrium was consistent with Proposition 5, which means that all investors were trading with a highly informed dealer, i.e.,  $e_j > \frac{b/2}{v_h - v_l}$

for all  $j$ . After the increase in  $N$ , Proposition 3 states that for a central dealer  $j^*$  that receives all independent investors' order flow, we have  $e_{j^*} < \frac{b/2}{v_h - v_l}$  as long as  $c > \frac{1}{2}(v_h - v_l)$ . Hence, independent investors benefit from a liberalization of OTC that reduces their dealer's transaction-specific expertise. Otherwise, if  $c \leq \frac{1}{2}(v_h - v_l)$  we still have that  $e_{j^*} = \frac{b/2}{v_h - v_l}$  for the central dealer and  $e_j = \frac{1}{2} \min\left(\frac{k}{n}, 1\right) > \frac{b/2}{v_h - v_l}$  for all other dealers, thus none of the incumbent investors is made better off. □

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