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Asymmetric Information and the Death of ABS CDOs

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Abstract. A key feature of the 2007 financial crisis is that for some classes of securities trade has practically ceased. And where trade has occurred, it appears that market prices are well below their intrinsic values. This seems especially true for those securities where the payoff streams are particularly complex, for example, structured finance ABS CDOs. One explanation for this is that information about these securities' intrinsic values since the crisis has been asymmetric, with current holders having better information than potential buyers. We first characterize the information asymmetries that were present in the structured finance ABS CDO market. Because many of the CDO dealers had partially or fully integrated the pipeline from mortgage originations through CDO issuance, they had informational advantages over potential buyers that could well have disrupted trading in CDOs as the crisis took hold in August of 2007. Using a "workhorse" model for pricing securities under asymmetric information and a novel dataset for the intrinsic values of ABS CDOs, we show how the resulting adverse selection problem could explain why the bulk of these securities either trade at significant discounts to their intrinsic values or do not trade at all.

Keywords: CDO, securitization, asymmetric, lemons

JEL Classification: C63, D82, D43

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1 Introduction

Collateralized debt obligations (CDOs) were at the heart of the 2007-2008 financial crisis. By January 2009, global banks, insurers, and asset managers had written down \$218 billion in losses from their holdings of CDOs of asset backed securities (ABS), or 42 percent of their crisis-related losses (see Table 1). They had also taken \$84 billion of losses on residential mortgage backed securities (RMBS), mostly from private label U.S. subprime mortgages. These firms were being penalized by markets for having these “toxic” assets on their balance sheets, and would have probably sold them at prices close to the fair liquidation value of the underlying collateral if they could have. But secondary market prices were well below intrinsic values, and selling in this distressed market would have crystallized losses and eroded their capital position even further. When CDO markets froze, many large financial institutions had no choice but to keep these securities on their books. The uncertainty surrounding the size and location of these CDO-related losses contributed to heightened counterparty risk, which in turn led to the collapse of interbank funding markets, setting off the Panic of 2007 (Gorton (2009)).

Interlinkages between ABS prices and liquidity provision through the the shadow banking system likely amplified the collapse of ABS markets in the second half of 2007. Krishnamurthy (2010), Allen and Carletti (2008), and Bolton, Santos, and Scheinkman (2011) analyze how the inability of investors to raise funds to purchase these securities could have impaired normal market functioning and dislodged prices from fundamentals. Similarly, Gorton and Metrick (2012) and Caballero and Simsek (2009) discuss how concerns about the liquidity of asset-backed securities used as collateral in repo transactions, combined with increased counterparty risk led to a run in the repo market, which in turn impeded funding for ABS CDOs.³ Further contributing to the downward spiral in asset prices was the deleveraging cycle that took place as a result of procyclical active balance sheet management (Adrian and Shin (2010)). These factors certainly played a destabilizing role in the market for ABS CDOs. The contribution of this paper is to show that, even absent these factors, asymmetric information between buyers and sellers regarding the quality of ABS CDOs would have sufficed to shutter this market.

Gorton (2009) argues that this asymmetric information stemmed from a loss of information about the quality of the underlying loans that occurred as a result of the long chain of structures inherent in the subprime securitization process.⁴ We expand this argument by documenting

³Counterparty risk could have been further intensified by the asymmetry of information about the true financial positions of borrowers (Calomiris (2008)).

⁴In theory, securitization could ameliorate the effects of asymmetric information when the seller signals his private information regarding the quality of the assets underlying a CDO to the buyer by retaining the riskier junior tranche (DeMarzo (2005) and DeMarzo and Duffie (1999)). But in practice, CDO issuers typically retained the most senior tranche. And even if the issuer retained the junior tranche, there is no guarantee that he would not sell it later on, or keep it but short the tranche above it so that he actually profits when the CDO

how vertical integration in the securitization of subprime mortgages provided the CDO dealers informational advantages over potential buyers. We then assess the extent to which this asymmetric information could have contributed to the sudden evaporation of liquidity for structured finance ABS CDOs during the crisis, and the dislodging of secondary market prices of these securities from their intrinsic values.

Because the analysis proceeds in parts which are only joined together at the end, we provide the following roadmap: In the next section we summarize how subprime mortgages were securitized into ABS CDOs, and how this market collapsed in the second half of 2007, resulting in the sudden evaporation of liquidity and the dislodging of prices from intrinsic values. Having established the facts that we are trying to explain with asymmetric information, in Section 3 we describe the nature of the asymmetric information between sellers and potential buyers of ABS CDOs. In Sections 4 and 5, we lay out a “workhorse” model for pricing securities under asymmetric information, and describe a novel dataset for the intrinsic values of ABS CDOs. More details on the computation of these intrinsic values, which capture the fair liquidation value of the underlying collateral, can be found in Appendix A. In Section 6, we take these data directly to the model and show how asymmetric information between sellers and potential buyers could explain both the evaporation of liquidity and the dislodging of prices from intrinsic values. The last section concludes.

2 Securitization from Mortgage Loans to CDOs and the Onset of the Financial Crisis

The residential mortgage market experienced tremendous growth in securitization over the last decade. We focus on the securitization of private label securities, in particular structured finance ABS CDOs. These CDOs are backed, in part, by mortgages that do not conform to the criteria for purchase by Fannie Mae and Freddie Mac, which establish limits on debt-to-income ratio, documentation requirements, and maximum loan amount.

The securitization chain begins with the pooling of prime, Alt-A, subprime, and junior-lien mortgages into residential mortgage-backed securities (RMBS). These RMBS securities are tranching based on seniority, with senior note holders having priority over the principal and interest payments collected from the underlying loans. This senior-subordinate structure is often referred to as the “waterfall.” The most senior notes are rated AAA because they are the last to incur losses stemming from loan defaults. Generally, the AAA-, AA-, and

loses value, as happened with the correlation trades. Also, signaling would not fully ameliorate the lemon’s problem when computational complexity makes it difficult to calculate a CDO’s intrinsic value (Arora, Barak, Brunnermeier, and Ge (2009))

A-rated RMBS are purchased by so-called “High Grade” ABS CDOs, whereas the BBB-, and BB-rated RMBS are purchased by “Mezzanine” ABS CDOs. These CDOs are also issued with senior/subordinate structure. Finally, both High Grade and Mezzanine ABS CDOs are purchased by other CDOs, which are often referred to as CDO-squareds because they are primarily backed by CDOs.⁵

Even before the crisis, there were only moderate levels of secondary market activity for ABS-CDOs. This was so for several reasons. First, the lower-rated tranches of these CDOs were largely recycled into other CDOs or hedged through “correlation trades”.⁶ Second, several of the largest CDO dealers retained many of the AAA CDO bonds they issued (Financial Crisis Inquiry Commission (2011), Chapter 8). Finally, the remaining AAA bonds sold in the market were designed as buy-and-hold investments and, for the most part, were privately traded over-the-counter among institutional investors such as banks, hedge funds, mutual funds, pension funds, insurance companies, endowments, and local governments (IOSCO (2009)). Even so, secondary market activity for CDOs increased notably between 2003 and 2005, as spreads for seasoned CDOs tightened relative to those of new-issue CDOs (Lucas, Goodman, and Fabozzi (2006)).

On July 10, 2007, both Moody’s and S&P downgraded hundreds of subprime RMBS that had been issued in the fourth quarter of 2005 and in 2006, and placed others on watch citing higher than anticipated rates of early payment defaults, aggressive underwriting of mortgages, fraud, and slowing home price appreciation (Moody’s Investor Service (2007) and Standard & Poor’s (2007a)). Furthermore, S&P announced that they were changing their assumptions on house prices going forward, projecting that house prices would decline 8 percent nationwide between 2006 and 2008, bottoming out in the first quarter of 2008.

On July 10, the effect of these negative rating actions on the ratings and expected performance of CDOs was much more uncertain— it was unclear how many of these RMBS were placed into the CDOs, and how these downgrades would affect the underlying assumptions of the copula models used by the ratings agencies (and many investors). For example, in a conference call with investors (Standard & Poor’s (2007b)), S&P surveillance director Enestine Warner stated that “For CDO squareds we are very much analyzing those. The difficulty with trying to include them on an exposure list is that they’re more opaque and not as linear to analyze.” Regarding the model-based methodologies for rating CDOs, S&P managing director

⁵628 of the 727 CDOs in the Cordell, Huang, and Williams (2012) data had at least some CDOs as collateral, so they define a CDO-squared as having at least 50 percent of the total deal collateral as CDOs.

⁶As shown in Table 5 of Cordell, Huang, and Williams (2012), two-thirds of BBB- and AA-rated CDO tranches were recycled into other CDOs and CDO-squareds. The correlation trade was executed by buying the equity tranche of a CDO and hedging it by shorting other tranches, mainly the lower-rated tranches of the CDOs. See Financial Crisis Inquiry Commission (2011), p. 192 for a discussion.

Pat Jordan stated that “As far as correlations in CDOs within our stress models, you know, we do have an assumption for correlations. We are studying the actual experience now and over time it will determine whether that should result in any kind of change to either the correlation or any other component of our stress models for CDOs.” It was only over the ensuing weeks that S&P would discover that most of the publicly traded ABS CDOs contained the downgraded BBB bonds. Investors learned that there were lemons in the market for CDOs, but figuring out which CDOs were the lemons would prove to be tricky.

Rating agencies and many other market participants relied on copula models to estimate expected collateral losses on ABS CDOs, similar to the ones used to value corporate CDOs. However, unlike the models used for RMBS securities which capture the sensitivity of the performance of the underlying mortgage loans to changes in house prices and other economic fundamentals, the copula models relied instead on correlations among bonds in the CDOs to estimate losses. The probability of default and expected losses produced by these models are extremely sensitive to model parameters based on historical performance of the underlying bonds, which, prior to 2007, was exceptionally good (Heitfield (2010)). Crucially, these models were not designed to directly capture the sensitivity of the performance of the underlying assets to systemic changes in house prices or changes in attributes that affect mortgage loan performance.⁷

Issuance and secondary-market trading of ABS CDOs declined considerably by August 2007. On August 9, BNP Paribas froze subscriptions and redemptions to several of its funds, and suspended computation of their net asset values because of “the complete evaporation of liquidity in certain market segments of the U.S. securitisation market,” which “has made it impossible to value certain assets fairly regardless of their quality or credit rating.” (New York Times (2007)) A survey of industry participants and market authorities by the International Organization of Securities Commissions (IOSCO) found that “secondary market trading in SFPs [structured finance products] appears to have declined significantly since the onset of the credit crisis.” (IOSCO (2009), p.10) As Ross Heller, an executive director at J.P. Morgan Securities Inc., put it: “We’re definitely in a period of very low liquidity at the moment.” (Bloomberg (2008))

And where trades did occur, it appeared that market prices were well below what one might have believed to be the intrinsic value for that class of security. The above-mentioned IOSCO survey of industry participants noted that:

“Respondents observe that, although there are some potential buyers in the market currently, most holders of SFPs [structured finance products] are reluctant to sell.

⁷For a discussion of the copula framework used by the rating agencies and others, see Duffie and Garleanu (2001), Moody’s Investor Service (2000), and Elizalde (2006).

Respondents noted that this could be ascribed to a general belief that the hold-to-maturity value of many SFPs exceeds the value that could be attained in the current secondary market.” (IOSCO (2009), p.10)

Similarly, in a White Paper accompanying the March 23, 2009 press release announcing the details of the Public-Private Investment Program to purchase troubled assets, the Treasury stated that “while fundamentals have surely deteriorated over the past 18-24 months, there is evidence that current prices for some legacy assets embed substantial liquidity discounts.” (U.S. Department of the Treasury (2009)) In sum, the secondary market for ABS CDOs virtually disappeared, and where trades did occur, secondary market prices dislodged from fundamentals.

By the end of 2007, Citigroup and Merrill Lynch reported spectacular losses of \$24 billion and \$25 billion, respectively, largely from their extensive CDO businesses (Financial Crisis Inquiry Commission (2011)). Billions more in CDO-related losses were reported by other large financial institutions such as Bank of America, Morgan Stanley, J.P. Morgan, Bear Stearns, as well as by insurance companies and hedge funds. Had these institutions been able to sell these CDOs at prices close to their intrinsic values, they likely would have done so because market participants began discriminating between firms with “toxic” assets on their balance sheets and those without. But, as confirmed by a survey of market participants, “selling into the current market would *crystallize* losses and could potentially erode the sellers’ capital position.” (IOSCO (2009), p.10)

3 Information asymmetries in the market for CDOs

Starting in summer of 2007, sellers (i.e. all the dealers and some of the largest financial firms around the world) wanted to sell their CDOs to avoid mark-to-market losses and the uncertainty that these assets were generating on their balance sheets. But when ABS CDOs began to suffer downgrades in July 2007, potential buyers could no longer trust the ratings or the copula models that broke down. Also, it soon became apparent that RMBS and CDO ratings were grossly inflated. Because rating agencies were only paid if they were selected to rate the deal, they were incentivized to inflate ratings in order to participate in more deals and increase their market share.⁸ As demonstrated by Cordell, Huang, and Williams (2012) (hereafter CHW), the only way to correctly value a CDO was by digging deep into its collateral

⁸On Feb. 4, 2013, the U.S. Department of Justice filed a law suit against S&P alleging that S&P “devised, participated in, and executed a scheme to defraud investors in RMBS and CDO tranches... to obtain money from these investors by means of material false and fraudulent pretenses, representations, and promises, and the concealment of material facts.” U.S. Department of Justice (2013)

and valuing it using loan-level data. But most investors either relied on rating agency ratings or used the standard copula models to value the CDOs. Others did not have the expertise, the resources, or the time to compute the intrinsic values of these securities. The sellers, on the other hand, having originated many of these CDOs, had large warehouse lines of subprime loans as well as many newly issued subprime securities, and were better equipped to value these securities because they had better knowledge of their underlying risks.

This superior knowledge of the underlying risks derives mainly from the fact that the largest CDO dealers were also in the business of originating the underlying RMBS (Table 2). But the vertical integration in the securitization process went even further than that. As shown in Table 3, most CDO dealers also entered the business of originating and servicing mortgages by acquiring large subprime originators and servicers. Most were either fully integrated (doing originations, issuance of RMBS, issuance of CDOs, and, finally, servicing the mortgages) or were partially integrated by engaging in some part of the mortgage chain. By owning an originator, CDO dealers would have information on the quality of the originations. By owning a servicer, they could see firsthand the early payment defaults (EPDs) and rise in delinquencies that started after house prices peaked in late 2005.

Further complicating the investor's ability to penetrate the CDO's portfolio is the fact that a CDO portfolio is not static; the CDO manager may purchase new assets, with constraints, to replace securities that pay down or mature. For example, when the financial crisis deepened and it became difficult to issue new CDOs, Citigroup and other large underwriters were holding large "warehouses" of RMBS intended to serve as collateral for future CDO issuances. To avoid losses on these RMBS, CDO underwriters began selling them into existing CDOs (Financial Crisis Inquiry Commission (2011), p. 197), essentially transferring the credit risk of these warehouse bonds to previous CDO noteholders.

This informational asymmetry could have been further amplified by the fact that CDOs are extremely complex, making it difficult for the average investor to forecast how a specific level of subprime losses would impact a CDO (Goodman, Li, Lucas, Zimmerman, and Fabozzi (2008) and Brunnermeier (2009)). For example, the existence of triggers and other complicated cash flow mechanics make it difficult to determine which bonds will be written down even if one knows the amount of collateral losses a deal will suffer.⁹ Goodman, Li, Lucas, Zimmerman, and Fabozzi (2008) state that "Calculating expected bond losses on mezzanine CDOs is a

⁹Most subprime RMBS deals use an excess spread/overcollateralization (ES/OC) structure to protect investors against losses on the underlying loans. Excess spread (ES) is the difference between the average rate on the loans securitized in the deal and the rate on the RMBS bonds issued from the securitization. Overcollateralization (OC) provides an additional buffer against losses. The lower-rated tranches will begin incurring losses only after the ES and OC are exhausted. At three years, a trigger test is performed to evaluate if the deal is performing well. If the deal passes this trigger test, some of the OC can be released to pay down principal.

complex, highly data-intensive task that requires access to a large database and a large amount of computing power, as well as knowledgeable analysts. Hence it is difficult for the average investor to forecast how a specific level of subprime losses will impact a mezzanine CDO.” (p.317)

In addition, the relative returns of the various tranches could be manipulated by the CDO manager to favor those tranches he held. For example, Lucas, Goodman, and Fabozzi (2006) describe how a CDO manager is able to manipulate the overcollateralization ratio over time to disguise par-building trades in an attempt to continue to redirect cash flows to the lower tranches. Also, the CDO manager may have financial ties with some note-holders that could result in a conflict of interest or outright fraud (Tavakoli (2003)). There are now many lawsuits underway in which CDO investors allege that the CDO issuers committed fraud by: (1) not disclosing conflicts of interest such as the short positions of other investors who helped pick the assets that went into the CDO, or (2) selectively placing risky assets into the CDOs in order to offload them from their own balance sheets, or (3) misinforming the investors about the true quality of the assets underlying the CDO, or (4) improperly obtaining millions of dollars in advisory fees and undisclosed profits at the expense of clients and investors.

Having established the sudden evaporation in liquidity in the market for ABS CDOs, and the informational advantage that sellers had over potential buyers, we now turn to our model of asymmetric information.

4 The model

In a straightforward generalization of the Akerlof (1970) lemon’s model, we start with a set of existing securities $A = \{a(t) : t = 1, \dots, N\}$ where t denotes a particular security. The quantity of security t is given by $a(t)$. The intrinsic value of security t , defined as the fair liquidation value of the underlying collateral, is $T(t)$, $L \leq T(t) \leq U$. Importantly, this intrinsic value of the security does not depend on who holds the security. There may be several types of securities holders, which we index by i . An individual’s portfolio is denoted $\mathbf{A}(i) = \{a(i, t) : 1 \leq t \leq N\}$ and $a(t) = \sum_i a(i, t)$. The total intrinsic value of agent i ’s portfolio is

$$TT[\mathbf{A}(i)] = \sum_t a(i, t) \cdot T(t). \tag{1}$$

Agents are risk neutral and have preferences over the money they have on hand (M) and the intrinsic value of the securities they hold. Their utility functions have the form

$$U_i[M, \mathbf{A}(i)] = M + x(i) \cdot TT[\mathbf{A}(i)] \tag{2}$$

where $x(i)$, $0 < x(i) \leq 1$, $i = \{1, 2, \dots, I\}$ is an agent-specific discount factor. This discount factor reflects the fact that some agents are willing to sell their securities at a discount to their intrinsic value.

For now we assume the $\mathbf{A}(i)$ and $x(i)$ are known to everyone. It will be handy to also have $\mathbf{T} = \{t : L \leq T(t) \leq U\}$ represent the set of intrinsic values and \mathbf{P} represent the set of possible prices for these securities.

4.1 No disclosure and delayed revelation

To keep the model simple, we assume it is impossible for the current holder of a security to credibly reveal its intrinsic value to a potential buyer. In addition, we assume that after a security trades there is a significant delay before the intrinsic value is revealed to the buyer. This assures that in a single period a given security only changes hands once. In Section 6, we argue that our results would still hold under the more realistic assumption that buyers attach some weight to the possibility that sellers are better informed. Also, when we take the model to the data, we will give the potential buyers the benefit of the doubt by endowing them with information on the vintage, seniority, and CDO type, as well as information on the distribution of the CDO intrinsic values.

The sellers Utility maximization implies that the minimum price that a current holder would require to relinquish a security with true payoff T is $V_s(i)[T] = x(i)T$, which we refer to as the value to the seller. Note that $V_s(i)[\cdot]$ maps a true payoff into a price, $V_s(i) : \mathbf{T} \rightarrow \mathbf{P}$.

Because $V_s(i)[\cdot]$ is monotonic, we can define $T^{\max}(i)[p] = V_s(i)^{-1}[p] = p/x(i)$ as the maximal value of $T(t)$ that seller i will offer when given the opportunity to sell securities for price p .

Define $\mathbf{S}(i)[p] = \{a(i, t) : L < T(t) < T^{\max}[p]\} \subseteq \mathbf{A}(i)$. This is the set of securities that seller i will offer at price p . The total set of securities that will be offered by all sellers at price p is $\mathbf{S}[p] = \bigcup_i \mathbf{S}(i)[p]$. Note that $\mathbf{S}[p]$ maps $\mathbf{P} \rightarrow \mathbf{T}$. We call the total set of securities offered at price p the ‘fruit bowl’; it is our basic supply relationship.

We can now define the average intrinsic value of the securities offered at a given price as

$$AT[\mathbf{S}[p]] = E[T(t) : t \in \mathbf{S}[p]] = \frac{\sum_{t \in \mathbf{S}[p]} a(t) \cdot T(t)}{\sum_{t \in \mathbf{S}[p]} a(t)}. \quad (3)$$

The buyer(s) Like the sellers, the representative buyer only cares about the average, or total, true payoff of the portfolio he acquires. However, the buyer may require that the price

he pays for the securities represent a discount, x_b , to their true payoff. Thus, when offered a set of securities \mathbf{Z} , he will value a draw from it at some fraction, x_b , of the elements' expected true payoff.¹⁰

$$V_b[\mathbf{Z}] = x_b \cdot E[T(t) : t \in \mathbf{Z}] = \frac{x_b \cdot \sum_{t \in \mathbf{Z}} a(t) \cdot T(t)}{\sum_{t \in \mathbf{Z}} a(t)} = x_b \cdot AT[\mathbf{Z}]. \quad (4)$$

V_b is the maximal per-unit price the buyer is willing to pay for the securities in \mathbf{Z} . We assume there are enough buyers so that in principle the entire set of securities could trade hands.

4.1.1 The equilibrium

We assume that sellers cannot coordinate with each other and that prices are determined in an exchange following a Dutch auction.¹¹ The auction begins with an auctioneer calling out the highest price, which results in the sellers throwing all their securities into the fruit bowl. Assuming some dispersion in the initial intrinsic values, the average value of the fruit bowl is less than the maximum price called, so there will be no trade at this price. As the auctioneer lowers the price, some informed sellers withdraw their higher valued securities from the fruit bowl, causing the average intrinsic value ($AT[\mathbf{S}[p]]$) of the securities in the fruit bowl to fall. One can think of this as a race. As the price is lowered it comes closer to the average value of what is in the fruit bowl. At the same time the value of the bowl is declining and moving away from the price as the better securities are removed. Eventually, the price called out will be such that buyers value a draw from the fruit bowl at the same price that elicited the fruit bowl to be filled. That is, the equilibrium is the $(p^*, \mathbf{S}[p^*])$ pair such that

$$V_b[\mathbf{S}[p^*]] = p^* \quad (5)$$

Buyers know that they will be “overpaying” for some low-value securities and “underpaying” for some high-value securities in the fruit bowl they purchase, but in the end they pay exactly what the bowl is worth to them, and all utility gains from trade go to the sellers. There will always be at least one equilibrium, although it may be trivial in that only the securities with the lowest value trade at a price equal to L , which may be zero. For now we admit

¹⁰We will stick with tradition and frame this in terms of expectations. However, in the current setup there will be no uncertainty about the securities traded, so it is just as natural to think of this as the mean or average value of the securities traded rather than the expected value of a draw from the portfolio.

¹¹Although buyers do not know the intrinsic values of the individual securities, there is no winner’s curse in our setup because all buyers value the bundle of securities offered into the fruit bowl equally, based on their perfect knowledge of the distribution of the intrinsic values and their knowledge of the sellers’ discount factor.

these ‘niggling’ equilibria and trades, although they can be eliminated by introducing a small transaction cost. It is easy to show that: (1) If $L > 0$ and $x(i) < x_b$ for any $i = \{1, \dots, I\}$, then there will be non-trivial trades. And (2), if $V_b[\mathbf{A}] \geq x(i)U$ for all $i = \{1, \dots, I\}$, then all securities will be traded at a price $p^* = V_b[\mathbf{A}]$. Also, there may be more than one pair $(p^*, \mathbf{S}(p^*))$ and the Dutch auction implies that the equilibrium with the highest price will be chosen. At the equilibrium price p^* , we denote the maximal value of $T(t)$ offered by any seller by T^{\max^*} .

If all sellers have the same discount factor, x_s , and all buyers have the same discount factor, x_b , then the set of securities traded in equilibrium is determined by the ratio x_s/x_b . That is, by combining equations 4 and 5 and the fact that in equilibrium, $T^{\max^*} = p^*/x_s$, we get that

$$\frac{\sum_{t \in \mathbf{S}[p^*]} a(t) \cdot T(t)}{\sum_{t \in \mathbf{S}[p^*]} a(t) T^{\max^*}} = \frac{x_s}{x_b}. \quad (6)$$

The left-hand side of equation 6 measures the “junkiness” of the fruit bowl at the equilibrium price p^* —the numerator is the total value of the securities in the fruit bowl, and the denominator is maximum possible value of the securities in the fruit bowl if they were all valued at T^{\max^*} . Equation 6 therefore implies that the junkier the securities offered into the fruit bowl, the larger the discount that sellers would have to accept (lower x_s) for trade to occur. Thus, the ratio x_s/x_b uniquely determines which securities trade in equilibrium, but the equilibrium price p^* will depend on the level of either x_b or x_s .

Graphical illustration of the equilibrium We demonstrate the equilibrium conditions graphically using the examples illustrated in Figures 1 through 3. The top part of each panel shows how the equilibrium price (p^*) is determined. The bottom part shows the distribution of securities in the market, and those that are traded at the equilibrium price.

Turning to the top portion of each panel, on the horizontal axis we plot the intrinsic value of the securities, T . On the vertical axis we plot the average price of a set of securities or the value of some set of securities to the holders. It will be handy to have a 45° line ($p = T$) which maps any security with intrinsic value T into a price equal to T . When the buyer’s discount $x_b = 1$, the equilibria are the $(p^*, \mathbf{S}[p^*])$ pairs such that $AT[\mathbf{S}[p^*]] = p^*$. On the graph, the equilibria occur when the $AT[\mathbf{S}[p]]$ line touches the 45° line. Without regard to the distribution of securities we can also plot $V_s[T]$, which gives the minimum price that the seller would require to relinquish a security with payoff T . If all sellers have the same discount factor, call it x_s , then $V_s[T] = x_s T$ is just a ray from the origin with slope x_s , which is less than one. At the equilibrium price p^* , the corresponding T^{\max^*} is given by $V_s^{-1}[p^*]$, and all securities with payoffs $T < T^{\max^*}$ will be offered into the fruit bowl and traded at that price.

Examples with a uniform distribution To plot $AT[\mathbf{S}[p]]$ we need to know the distribution of the intrinsic values of the securities. To keep this first example simple, we will assume a uniform distribution on T with lower and upper bounds L and U . That is $\forall a(t) \in \mathbf{A}, a(t) = a$. In this case

$$AT[\mathbf{S}[p]] = E[T : L < T(t) < T^{\max}] = \frac{\sum_{t \in \mathbf{S}[p]} a(t) \cdot T(t)}{\sum_{t \in \mathbf{S}[p]} a(t)} = \frac{(L + T^{\max}[p])}{2} = \frac{(L + p/x_s)}{2},$$

and the resulting equilibrium price is $p^* = \frac{Lx_s}{2x_s - 1}$.

The left panel of Figure 1 illustrates the uniform case when $L > 0$. The equilibrium is such that some of the securities—those valued at $T^{\max*} = p^*/x_s$ or less—trade at a price (p^*) equal to the average intrinsic value of all securities in the fruit bowl.

So far, we assumed that it takes a long time for the new buyer of a security to ascertain its intrinsic value. We clearly need something like this to keep the new buyers from aggravating the adverse selection problem. To see this, consider the same example shown on the left panel of Figure 1. What happens to the market after the securities with payoffs $T < T^{\max*}$ change hands?

Suppose all of these first-round trades have taken place and we now allow those who purchased securities to learn the intrinsic value of the securities they acquired. In the second round of trading, the potential sellers comprise all security holders: those who bought securities in the first-round and those who had securities in the first round but never sold them. These two groups of potential sellers have different discount factors. Those who bought securities in the first round when they had a discount factor of $x_b = 1$, have the same discount factor of 1 in the second round because they are not liquidity constrained. That is, they have no incentive to sell their securities at a price that is less than their intrinsic values. As such, their $V_s[]$ schedule will be the 45° line between L and $T^{\max*}$, shown by the green line in the right panel of Figure 1. For the other sellers who never sold their securities in the first round, their $V_s[]$ schedule (shown in red) remains $x_s T$ (where $x_s = 0.8$). The blue $AT[\mathbf{S}[p]]$ line representing the average intrinsic value of the basket of securities offered at a given price p is now above the 45° line for all $T > L$. That is, only securities with $T = L$ would possibly trade in the second round.¹²

Figure 2 illustrates the two uniform cases when $L = 0$. In the left panel of Figure 2, $x_s > 1/2$ and no securities are traded (except maybe those worth zero). In the right panel, $x_s < 1/2$ and all securities are traded at an indeterminate price somewhere between $x_s U$ and

¹²Since all the securities with a value of L were traded in the first round, and they are now held by agents with identical discount factor x_b , there are no gains to be had for these holders to trade the securities among themselves.

$(U + L)/2$.

Examples with a more general distribution If the payoffs are uniformly distributed with $L = 0$, trade is limited to cases where the sellers' discount is large ($x_s < 1/2$). For more general distributions, this is not the case. The left panel of Figure 3 illustrates an example where the distribution of securities has a small mass between $T = 0$ and $T = 7$, the seller's discount is not too small ($x_s = 0.85$) and yet there are no (non-trivial) trades in equilibrium. This example shows how a few lemons can ruin the market.

The right panel of Figure 3 illustrates multiple equilibria. The Dutch auction results in trades at the equilibrium with the highest price (p^{**}). However, one could easily impose alternative rules on the bidding process.

5 Data

We use data on the intrinsic values of 4,865 individual CDO securities (from 672 CDO deals), computed as of March 2011 by CHW. Appendix A provides more details on how these intrinsic values are computed. The original CHW sample comprises 727 CDO deals, but we exclude 213 deals from the 2000-2005 vintages that were liquidated as of March 2011 because they were likely liquidated before the financial crisis erupted in 2007. To compute these intrinsic values, CHW use the Intex engine which covers the universe of publicly traded private-label RMBS and ABS CDO deals in the United States. The Intex engine not only warehouses data on the underlying collateral backing each CDO, it also programs into its cash-flow engine the structural features of each deal, such as loss allocations among the tranches, triggers, prepayment penalty allocations, and interest rate hedges. This allows buyers and sellers a common set of tools to generate their own valuations. For each CDO deal, the securities are tranced based on their seniority. At the tranche level, there are 4,865 individual CDO securities in our sample: 734 Senior AAA, 1,009 Junior AAA, 1,387 AA and A, 828 BBB, 252 BB/B, and 655 NR (not rated, equity, or "first-loss" tranche). Mezzanine CDOs make up about two-thirds of the CDOs in our sample; the rest are High Grade CDOs.

Roughly three-quarters of the 4,865 CDO bonds in the CHW data were originated between the second half of 2005 and the end of 2007. These CDOs experienced higher losses than those issued in earlier years because they are primarily backed by subprime securities and other ABS CDOs. As described above, underwriting standards for subprime mortgages deteriorated significantly in the second half of 2005, and further in 2006 and 2007, as can be seen by the dramatic upward shift in the cumulative default rate curves for subprime loans starting in 2005 (Figure 4). CHW and Goodman, Li, Lucas, Zimmerman, and Fabozzi (2008) find that CDO

losses increase by year of origination. Also, during the first half of 2005 ISDA published a template for executing credit default swaps of ABS on a Pay-As-You-Go basis, which made it easier to invest CDO collateral synthetically and brought about an explosion in synthetic CDO issuance starting in the second half of the year (see Figure 5). We therefore take CDO vintage into account when we examine how these securities trade under asymmetric information. To this end, we label CDOs that were issued on or before June 30, 2005 as the “early” vintage, and those issued after this date as the “late” vintage.

Table 4 summarizes the distributions of intrinsic values for the various CDO categories in our sample. Only the most senior tranches of the early vintage CDOs have average intrinsic values greater than 50. All CDO categories with seniority AA/A and below, and all of the late vintage CDOs experienced heavy losses, as can be seen by the low means and medians for their intrinsic values. Furthermore, the high standard deviations of the intrinsic values indicate that they are highly dispersed between 0 and 100.

6 Trading CDOs under asymmetric information

We now examine how these 4,865 CDOs would trade when potential buyers cannot penetrate a CDO’s portfolio to determine its intrinsic value, only the sellers can. Even so, to give the buyers the benefit of the doubt, we assume that they are equipped with some valuable information. That is, when considering the purchase of a CDO tranche they know its seniority, its vintage, and whether it is a High Grade or Mezzanine CDO. But more importantly, we assume that each buyer knows the true distribution of intrinsic values for the type of CDO he is considering to buy. One can think of each CDO as having a label (much like the label on a bottle of wine) with three stamps on it: a High Grade or Mezzanine stamp, a Seniority stamp (e.g. Sr-AAA), and a vintage stamp (early or late). For example, a CDO’s label may read “High Grade, Jr-AAA, early vintage.” Another CDO’s label may read “Mezzanine, BBB, late vintage.” We will consider trading in markets for CDOs that have the same label. Sellers are assumed to know the intrinsic value of a CDO, but cannot credibly reveal this to the buyer. Referring back to the model in Section 4, buyers value a draw from the fruit bowl at the same price that elicited the fruit bowl to be filled. Assuming that prices are called following a Dutch auction, we compute the resulting trading equilibria for the various CDO markets. For a given CDO market, it will be useful to define ‘meaningful’ trades as those involving securities other than the ones with the lowest intrinsic value.

For simplicity, we set the buyers discount factor (x_b) to 0.9. That is, a buyer is willing to pay at most 90 for a CDO with $T = 100$. This normalization is not important for our results because, as we showed earlier, it is the ratio between the sellers’ and buyer’s discounts (x_s/x_b)

that uniquely determines which securities trade in equilibrium. The junkier the securities offered into the fruit bowl, the more desperate sellers would have to be (relative to the buyers) in order for trade to occur. To determine the price at which they trade, we would still need knowledge of, or an assumption about, one of the discount factors. Instead of arbitrarily fixing x_s and examining the resulting equilibria, we ask the following question: given the distribution of intrinsic values in our data, how desperate do the sellers have to be in order for there to be trade under asymmetric information? More specifically, we will search for x_s^{max} , which is the maximum value of x_s at which meaningful trades occur. The minimum haircut that sellers would be willing to take on the intrinsic value of their securities for meaningful trades to occur is therefore $1 - x_s^{max}$.

We first consider trading under asymmetric information in the high grade, senior-AAA, early vintage CDOs, which were regarded as relatively safe before July 2007. For these securities, our model implies that they will only trade if the sellers' discount factor is 0.64 or less, which is a hair-cut of at least 36 percent on the intrinsic values of their securities. The left panel of Figure 6 illustrates the graphical interpretation of the equilibrium when $x_s = x_s^{max} = 0.64$ (the buyers discount is fixed at 0.9). At the equilibrium price $p^* = 45$ the maximum value of T that sellers will offer into the fruit bowl is $T^{max*} = 70$. As indicated by the grey-shaded bars in the bottom part of the panel, those securities having intrinsic values greater than 70 will not trade. Holders who did not sell would rightfully feel aggrieved if the market assigns a value of 45 to their securities which are all worth more than 70.

The mezzanine, senior-AAA, early vintage CDOs in our sample experienced the lowest losses of all categories. The average intrinsic value for this group is 84.5. For these CDOs, trade can occur under asymmetric information as long as the sellers discount factor is no higher than 0.87, which implies a minimum haircut of 13 percent. The resulting equilibrium when $x_s = x_s^{max} = 0.87$ is shown in the right panel of Figure 6. Although this equilibrium results in some meaningful trade, only 4 of the 126 securities trade. If $x_s = x_s^{max} = 0.76$, that is, if the sellers were willing to incur a haircut of 24 percent, then all securities would trade at an equilibrium price of $p^* = 76$.

There are many more absolute lemons in the market for late vintage CDOs— which puts more mass in the far left tail of the distribution of values and imparts more “damaging” dispersion to the distribution. This implies that sellers would have to be more desperate to sell these securities (lower x_s) in order for there to be meaningful trades. For example, meaningful trades only occur for the high-grade, senior-AAA, late vintage CDOs when the sellers' discount factor is 0.42 or less, which would imply a haircut of at least 58 percent. The resulting equilibrium with $x_s = x_s^{max} = 0.42$ is shown in the left panel of Figure 7. Only securities with $T \leq 58$ are offered into the fruit bowl and traded, at a price $p^* = 25$. For the

mezzanine, senior-AAA, late vintage CDOs, there are so many lemons that meaningful trades only occur when the sellers' discount is 0.29 or smaller (a haircut of at least 71 percent). As shown in the right panel of Figure 7, the resulting equilibrium with $x_s = x_s^{max} = 0.29$ is such that everything trades. This is a knife-edge equilibrium in the sense that raising the sellers discount by a smidgen would result in no trades.

Table 5 shows the maximum sellers' discount at which meaningful trades occur (x_s^{max}) for each of the CDO categories. x_s^{max} is only above 0.50 for the high grade and mezzanine, Sr-AAA, early vintage categories. For all other categories, x_s^{max} is below 0.5, implying that trading in these securities would only occur if sellers would be willing to accept a haircut of at least 50 percent.

All told, given the distribution of intrinsic values of CDOs, our model implies that unless the sellers were extremely desperate to rid themselves of these securities, only the highest rated, early vintage CDOs would trade under asymmetric information. If trading did take place for the late vintage and less senior securities, our model would predict that these securities would have traded at prices well below their intrinsic values. Both of these findings are consistent with the sharp decline in secondary market activity for ABS CDOs that took place in the second half of 2007, and the earlier-cited IOSCO survey response that "the hold-to-maturity value of many structured finance products exceeds the value that could be attained in the current secondary market." In our view, the downgrades of hundreds of subprime mortgage backed securities on July 10, 2007 by Moody's and S&P was a wake-up call to investors who learned that there were potentially many lemons among the CDOs. The sudden realization about the presence of lemons in the CDO market came at a time when neither the ratings nor the copula models being used to value CDOs could be trusted. Buyers and sellers learned that the only way to accurately value a CDO was by building machinery to value its underlying assets using loan-level data, in much the same way it was done for the underlying RMBS bonds themselves. Because sellers had underwritten many of these CDOs, they had better knowledge of the underlying loans and recently issued RMBS bonds, and thus better understood their inherent risks. This introduced asymmetric information between buyers and sellers, which in turn, would explain why the market for ABS CDOs shut down. The severe losses suffered in the market combined with the way it was concentrated in the largest financial firms explain how this information, once revealed, could set off the Panic of 2007.

Before concluding, we should emphasize that it is not necessarily the lack of information about the CDOs' intrinsic values that shutter these markets; rather, it is the asymmetric distribution of this information. For example, suppose the sellers are as poorly informed as the buyers: They know the distribution of intrinsic values within any label, but they do not know the intrinsic values of the individual securities, including the ones they hold. For now

we also assume the buyers know the extent of the sellers' ignorance, and that $x_b > x_s$. In this case, the risk neutral seller would value his securities at the average intrinsic value of all securities within the label. As such, any price at or above $x_s \cdot TT[\mathbf{A}]$ will elicit all securities for that label to be supplied to the fruit bowl. Since buyers would be willing to pay anything less than $x_b \cdot TT[\mathbf{A}]$, which is above $x_s \cdot TT[\mathbf{A}]$, the entire stock of securities would trade. Thus it is not the lack of information that shuts the market, but the asymmetric distribution of that information.

Finally, we note that the possibility that the sellers are better informed than buyers is sufficient to shutter the market even if sellers are not actually better informed. Developing this fully is beyond the scope of this paper, but the argument is straightforward: If buyers fear (or place some probability on the possibility) that sellers know the intrinsic value of their individual securities, then the buyers will value a draw from the fruit bowl at some weighted average of (1) the average intrinsic value of all the securities for a given label assuming sellers are ignorant of individual securities' values, and (2) the lower average value of what informed sellers would offer to the fruit bowl for a given price. Thus, if the weight attached to the possibility (2) that sellers are informed is high enough, the market will collapse, even if few or no sellers are in fact well informed.

7 Conclusion

The 2007-2008 financial crisis led to a freezing of CDO markets and a disconnect between market prices (for the few trades that did occur) and fundamentals. Home mortgages were securitized through a long chain of financial intermediaries, which culminated in the creation of CDOs. Many of the largest CDO dealers were fully integrated (doing originations, issuance of RMBS, issuance of CDOs, and, finally, servicing the mortgages), which gave them an informational advantage over potential buyers. Also, the complexity and opaqueness of these securities made it easy for asymmetric information to arise between current holders and potential buyers. We show how vertical integration, opacity, and complexity in the securitization process and the resulting asymmetric information between buyers and sellers could have amplified the stresses already present in CDO markets, by making these markets illiquid and dislodging prices from intrinsic values.

Using a novel dataset on the distribution of intrinsic values of structured finance ABS CDOs, we examine which securities would trade under asymmetric information. The likelihood of trade depends not only on the distribution of intrinsic values of the CDOs but also on the ratio between the sellers and buyers' discount factor. For the 4,865 CDOs in our sample, we find that the informational problems were so severe that most of the CDOs would not trade

unless the sellers were willing to sell their securities at a very steep discount to their intrinsic values.

We take this as evidence that asymmetric information about the intrinsic values of the CDOs was sufficient to cause CDOs to stop trading and their prices to be well below what most holders valued them at. This is not to imply that asymmetric information was the only factor leading to the market's breakdown. Indeed there were likely many factors, the question we ask is if asymmetric information alone would be sufficient to cause this. The experiment is to give the potential buyers less information than current holders and determine the conditions under which there would be significant trade. Because we give the buyers perfect information about the distribution of the intrinsic values within relative narrow groups of securities, the experiment is rigged to favor a negative answer. That is, it is rigged to find that asymmetric information alone is not sufficient to shutter these markets. However, given the distribution of the CDO intrinsic values, we find that asymmetric information would be sufficient to close the markets. For a couple of labels we find trade even when sellers are not massively desperate for cash; their discount factors are greater than 0.5. But in these two cases only the junkiest of the securities trade. We therefore conclude that the information asymmetries were sufficient to shut down the CDO market. These securities are still not trading in any meaningful way and, if one buys our results, they may never.

Appendix A

This appendix summarizes the methodology used by CHW to value ABS CDOs. These CDO values are the intrinsic values used in our analysis. They are calculated as the liquidation value of the underlying ABS (mostly RMBS) collateral— the value that a CDO bondholder would receive if the CDO manager sold all the underlying ABS bonds and allocated the proceeds according to the CDO bondholders' positions in the waterfall structure. Valuing the 213 CDOs that were already liquidated as of March 2011 is fairly straightforward, since this involves simply summing up the actual principal payments made to the CDO bondholders as reported in the trustee's final liquidation report provided to Intex (or from other sources if it was not). Table 7 in CHW shows that the 213 liquidated SF CDO deals experienced losses of \$162 billion, a loss rate of 78 percent.

For the 514 non-liquidated, or active, CDOs (as of March 2011), the idea is to compute the expected fair liquidation values of the still-active underlying ABS (mostly RMBS) collateral, and then feed these as recovery amounts into the CDO waterfall. The fair liquidation values of the underlying RMBS are, in turn, computed using realized and expected write-downs on the underlying mortgage loans and securities, computed from models described below. First, CHW determine that, of the original \$432.6 billion original balance of CDO collateral, \$75.9

billion in principal had already been paid back to the CDO bondholders, while \$134.4 billion had already been written down as losses as of March 2011. This leaves \$222.3 billion in ABS collateral backing the 514 still active CDOs.

The next step is to determine the fair liquidation value of this remaining \$222.3 billion of CDO collateral by estimating expected losses, and then feeding these recovery values into the CDO waterfall to determine how much each ABS CDO bond recovers, and ultimately loses. Table 8 in CHW describes the methodologies used for valuing the 23,147 unique securities underlying these active CDOs. Roughly two-thirds are RMBS, which are backed by subprime, Alt-A and prime mortgages. These RMBS were valued using the proprietary prepayment, default, and loss models of Andrew Davidson & Company, a leading market modeling vendor, which is seamlessly integrated with the Intex cash flow engine for valuation and loss estimation.¹³ Each ABS collateral asset (or loan) contributes its fair value liquidation proceeds to the ABS deal, with the aggregate recovery amounts allocated in the Intex software to the various tranches of the ABS CDO deals according to the CDO waterfall.¹⁴ All told, CHW estimated \$123 billion in expected write-downs stemming from these ABS bonds underlying the still active CDOs (on top of the \$134.4 billion that had already been written down).

As of March 2011 the actual write downs of \$162 billion for the 213 liquidated SF CDOs combined with \$134 billion actual and \$123 billion expected write-downs for the 514 still active SF CDOs sum up to a grand write-down total of \$420 billion, 65 percent of the \$641 billion of ABS of CDO issuance (Table 10 in CHW). While different valuation methods can be used to estimate ABS CDO losses, it is noteworthy that as of March 2011, 71 percent of the write-downs had already been realized. Of the remaining 29 percent of estimated write-downs (\$123 billion), most came from still-active BBB- and A-rated subprime and alt-A RMBS securities that were trading at values at or new zero. Based on this, loss estimates will not likely vary much by the valuation method chosen.

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¹³See <http://www.ad-co.com/>. The advantage of using a leading market vendor is that their fair values are good representations of what investors will pay in the market for these RMBS securities.

¹⁴For CDOs that have other CDOs as collateral, CHW first compute the fair liquidation values of the underlying CDOs, and then feed these values together with those of all other assets in the CDO’s portfolio into the CDO’s waterfall to determine the tranche payoffs.

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Tables

Firm Type	ABS CDOs	Corp. Credit	RMBS	Other or Undefined	Total Write- downs
Insurers/asset managers	61,074	6,320	10,386	3,8347	116,126
North American banks	84,319	23,702	42,272	59,011	209,305
European banks	63,464	18,579	26,423	62,634	171,100
Asia/emerging market banks	9,358	4,724	5,728	3,743	23,553
Totals	218,216	53,324	84,810	163,735	520,084
Share of total writedowns	42%	10%	16%	32%	100%

Table 1: Crisis-Related ABS CDO writedowns through January 26, 2009 (in millions of dollars). Source: Creditflux Ltd. (2009). Based on publicly disclosed company information compiled by Creditflux Ltd. According to Creditflux, ABS CDO writedowns were no longer reported separately after February 2008. By this time, most firms had substantially written down their holdings.

CDO Dealers	Subprime RMBS Issuance	ABS CDO Issuance
Merrill Lynch	76,747	91,767
Citigroup	54,032	70,552
UBS Warburg	20,024	65,409
Goldman, Sachs & Co.	52,727	59,103
Credit Suisse First Boston	99,081	38,209
Deutsche Bank Securities	59,635	37,362
Wachovia Securities	12,528	25,033
RBS Greenwich Capital	31,643	23,869
Calyon Securities	0	22,825
Banc of America Securities	28,524	22,617
Barclays Capital	33,812	21,424
Lehman Brothers	170,342	19,116
Bear Stearns	61,698	19,111
Morgan Stanley	121,289	14,237
WestLB Securities	0	12,125
SG Corporate & Inv. Banking	4,796	11,609
Dresdner Kleinwort Wasserstein	0	9,875
JP Morgan	53,628	7,584
Totals	880,507	571,826

Table 2: CDO Dealers' Issuance of Subprime RMBS and ABS CDOs (in millions of dollars).
Source: Intex Solutions, Inc.

CDO Dealers	Originators	Servicers	RMBS Shelves	Level of Integration
Merrill	First Franklin (2006), Ownit (2005)	Wilshire Credit, First Franklin (2006)	SURF, MLHEL, MLMI, MLFF, MLFF in 2007	Full
Citigroup	CitiFinancial (2000), Argent (2007)	CitiMortgage, AMC Mtge. Services (2007)	Citigroup Mtge. Trust, Salomon Brother Mtge., Citifinancial Mtge., Citicorp Res.	Full
UBS	--	--	MASTR	Partial
Goldman Sachs	Southern Pacific (1999), Senderra Funding (2007)	Litton Servicing (2007)	GSABS, GSAMP	Partial
Credit Suisse	DLJ Mtge. (2000)	SPS (2004)	ABSHE, HEAT, HEMT, CSFB Mtge	Full
Deutsche Bank	Chapel Funding (2006), MortgageIT (2006)	Chapel Funding (2006)	ACE	Full
Wachovia	--	--	Wachovia Asset Securitization	Partial
RBS	--	--	SVHE	Partial
Calyon	--	--	--	None
Bank of America	Countrywide (2008)	Countrywide (2008)	BAFC	Partial
Barclays	Equifirst (2007)	HomeEq(2006)	SABR, EquiFirst Loan Sec. Trust	Full
Lehman Brothers	BNC (2003), Aurora (2003), Finance America (2004), SIB Mtge. (2004), Preferred Mtge. (2004)	Aurora (2003)	SAIL, ARC, SASCO, SASC, BNC Mtge. Loan Trust, Finance America Mtge. Loan Trust	Full
Bear Stearns	Encore Credit (2007)	EMC	BSABS, EMC Mtge. Loan Trust, Encore Credit	Full
Morgan Stanley	Saxon (2006), CityMortgage(2006)	Saxon	MSAB, MSHE, MSDW, MS Mtge. Loan Trust	Full
WestLB	--	--	--	None
SG Corporate	--	--	SG Mtge. Securities Trust	Partial
Dresdner	--	--	--	None
JP Morgan	Chase Home Finance, WAMu (2008)	Chase Home Finance, WAMu (2008)	JPM Mtge. Acquisition Trust, CFLAT, CFAB	Full

Table 3: CDO Dealer Vertical Integration of Subprime Securitization. Notes: The table shows the acquisition of subprime originators and servicers (along with dates where available), and lists names of RMBS issuance shelves run by each dealer. Sources: Intex for issuance volume, numerous industry sources for other data.

CDO category	N	Mean	Median	Max	Min	Std.Dev.
HG, Sr-AAA, early vintage	56	66.3	67.8	100	0	25
HG, Jr-AAA, early vintage	67	28.9	0.8	100	0	38.4
HG, AA/A, early vintage	79	4.5	0	100	0	15.2
HG, BBB, early vintage	44	8.3	0.4	58.6	0	14
HG, BB/B, early vintage	9	0	0	0	0	0
HG, NR, early vintage	41	2.4	0	98.6	0	15.4
HG, Sr-AAA, late vintage	236	34.5	33.8	100	0	26.2
HG, Jr-AAA, late vintage	349	8.6	0	100	0	20.6
HG, AA/A, late vintage	412	1.7	0	100	0	9.2
HG, BBB, late vintage	178	1.8	0	100	0	10.7
HG, BB/B, late vintage	61	1.2	0	15.7	0	3.2
HG, NR, late vintage	185	1	0	100	0	8.7
Mez, Sr-AAA, early vintage	126	84.5	88.4	100	42.1	16.3
Mez, Jr-AAA, early vintage	184	55.2	68.2	100	0	42.8
Mez, AA/A, early vintage	254	18.4	0	100	0	34.6
Mez, BBB, early vintage	208	13.8	0	100	0	25.8
Mez, BB/B, early vintage	49	1.9	0	66.2	0	9.8
Mez, NR, early vintage	142	2.2	0	100	0	14
Mez, Sr-AAA, late vintage	316	32.4	19.1	100	0	34.1
Mez, Jr-AAA, late vintage	409	7.7	0	100	0	21.4
Mez, AA/A, late vintage	642	3.2	0	100	0	14.7
Mez, BBB, late vintage	398	4.1	0.2	100	0	12.8
Mez, BB/B, late vintage	133	4.4	0	100	0	12.2
Mez, NR, late vintage	287	1.5	0	100	0	11.8

Table 4: Summary statistics for distribution of intrinsic values by CDO type and vintage.

CDO Category	x_s^{max}	T^{max}	# of Secs with $T > 0$	# of Secs with $T > 0$ Traded	$p^*/AT[\mathbf{A}]$
HG, Sr-AAA, early vintage	0.64	71	55	31	0.68
HG, Jr-AAA, early vintage	0.25	100	33	33	0.90
HG, AA/A, early vintage	0.10	22	15	13	0.51
HG, BBB, early vintage	0.22	15	21	15	0.42
HG, BB/B, early vintage	0	0	0	0	n.a.
HG, NR, early vintage	0.02	99	1	1	0.90
HG, Sr-AAA, late vintage	0.42	59	206	175	0.71
HG, Jr-AAA, late vintage	0.08	68	107	99	0.70
HG, AA/A, late vintage	0.03	41	57	54	0.71
HG, BBB, late vintage	0.01	100	26	26	0.90
HG, BB/B, late vintage	0.10	11	9	8	0.92
HG, NR, late vintage	0.01	100	3	3	0.90
Mez, Sr-AAA, early vintage	0.87	48	126	4	0.49
Mez, Jr-AAA, early vintage	0.50	100	137	137	0.90
Mez, AA/A, early vintage	0.16	100	82	82	0.90
Mez, BBB, early vintage	0.15	31	98	72	0.35
Mez, BB/B, early vintage	0.02	67	5	5	0.90
Mez, NR, early vintage	0.02	100	6	6	0.90
Mez, Sr-AAA, late vintage	0.29	100	251	251	0.90
Mez, Jr-AAA, late vintage	0.07	100	115	115	0.90
Mez, AA/A, late vintage	0.03	100	106	106	0.90
Mez, BBB, late vintage	0.06	36	167	161	0.63
Mez, BB/B, late vintage	0.06	56	48	47	0.80
Mez, NR, late vintage	0.01	100	7	7	0.90

Table 5: Trading under asymmetric information with $x_b = 0.9$. x_s^{max} is the maximum value of x_s for which meaningful trades occur. T^{max} is the maximal value of T that sellers will offer when given the opportunity to sell securities at the equilibrium price p^* .

Figures

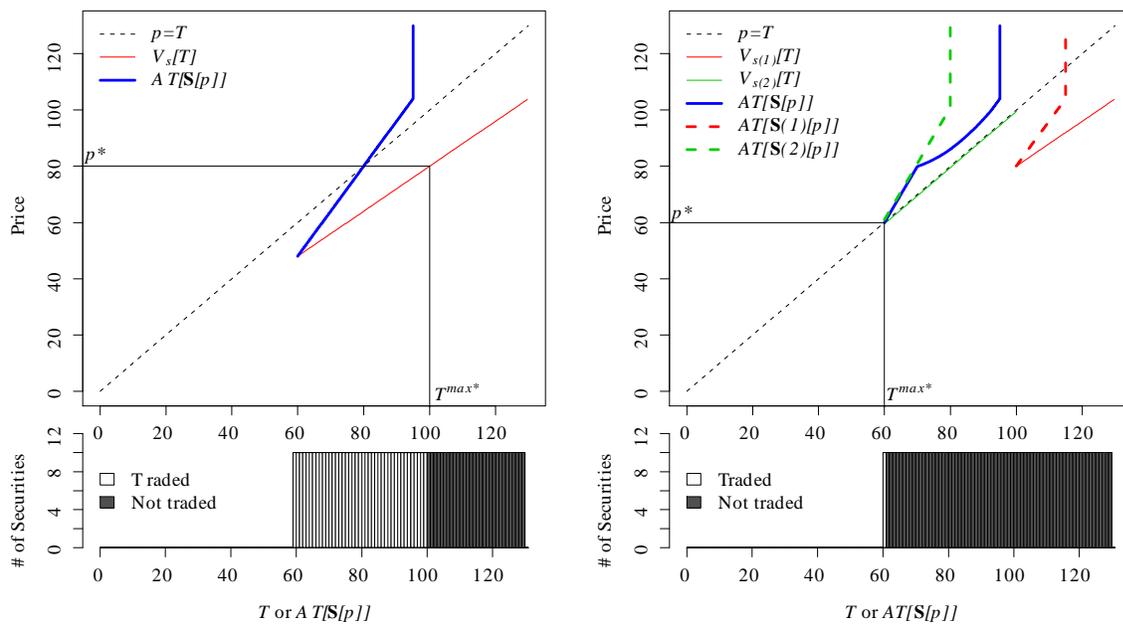


Figure 1: Trading when securities are uniformly distributed between $L = 60$ and $U = 130$. Left panel: Seller's discount (x) is set to 0.8, and only some securities will trade. Right panel: Second round of trading.

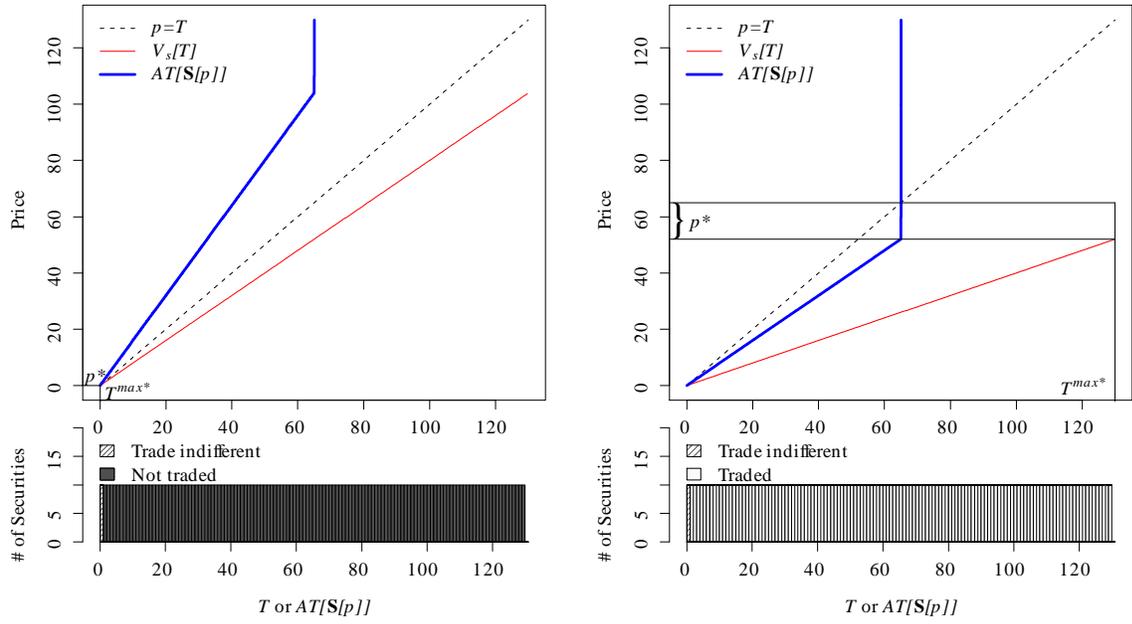


Figure 2: Trading when securities are uniformly distributed between $L = 0$ and $U = 130$. Left panel: Seller's discount (x) is set to 0.8, and only securities with $T = 0$ will trade. Right panel: $x = 0.4$ and all securities trade.

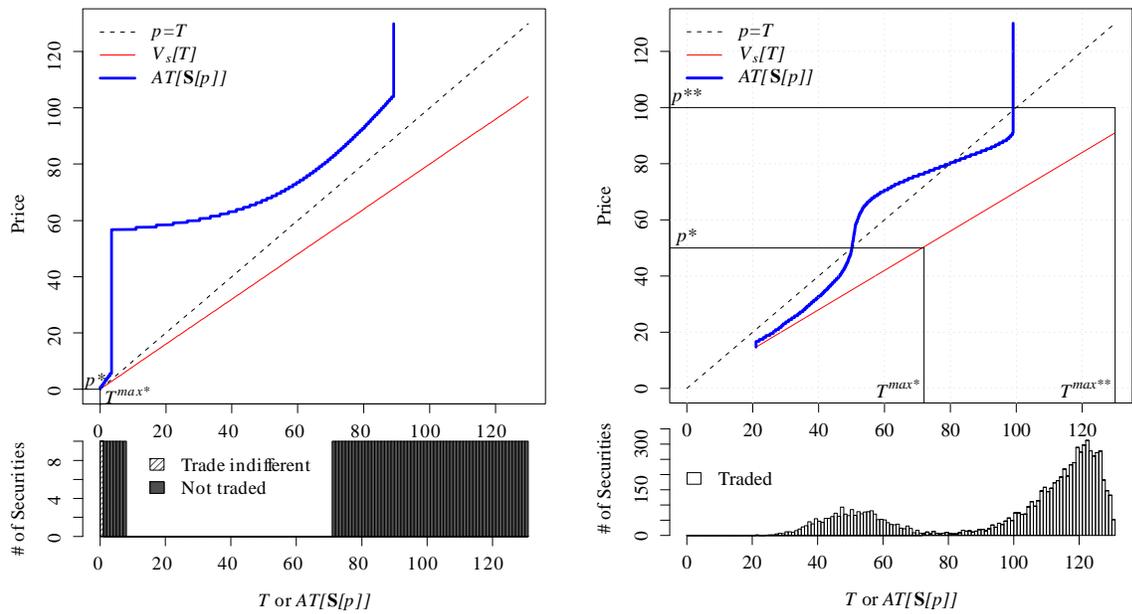


Figure 3: Left panel: A significant mass of payoffs close to zero can preclude all trade. Right panel: Multiple equilibria.

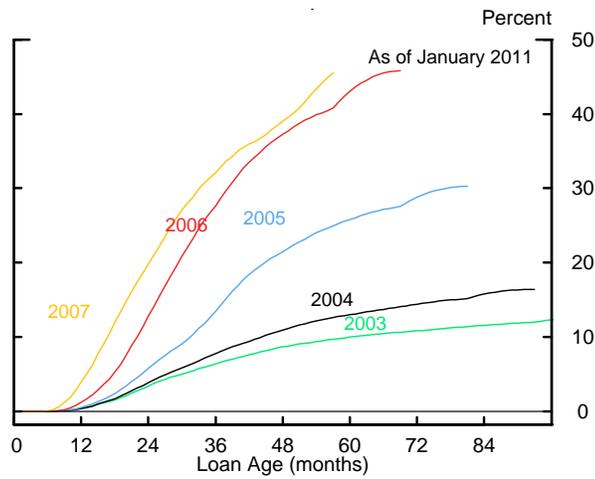


Figure 4: Cumulative defaults on subprime loans. Source: Federal Reserve Board calculations using CoreLogic data.

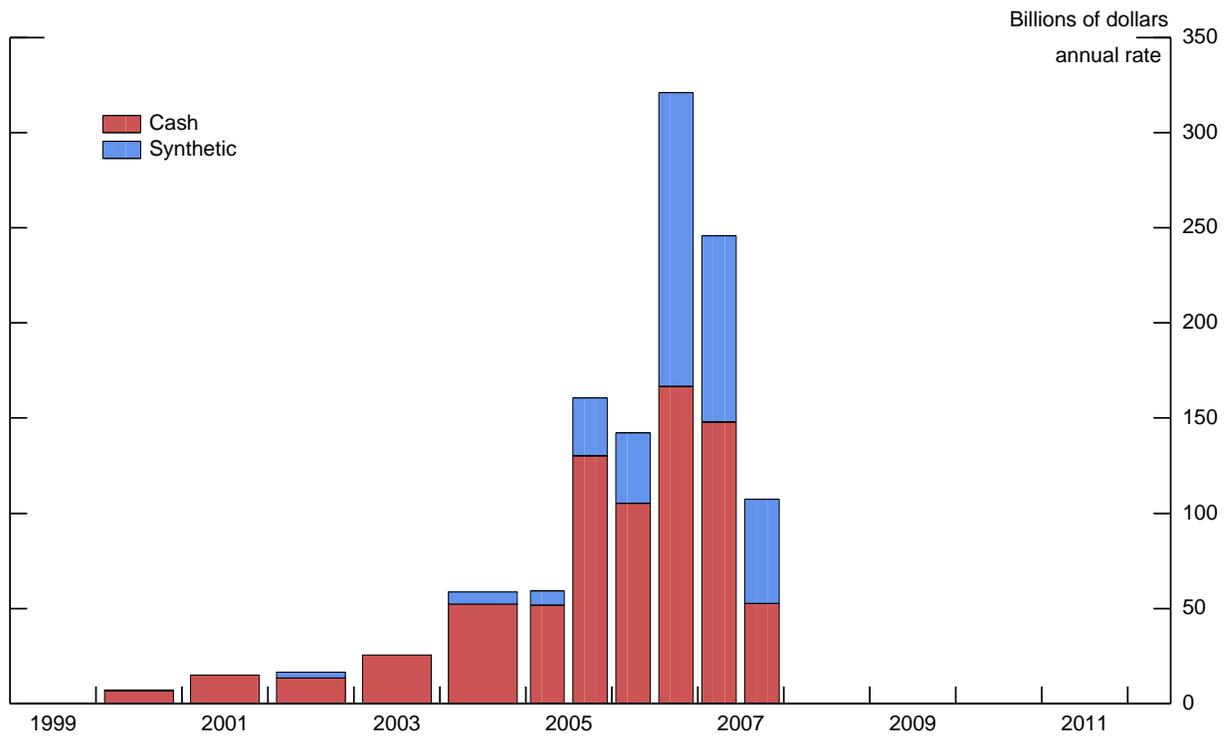


Figure 5: Issuance of publicly traded SF ABS CDOs. Source: Cordell, Huang, and Williams (2012).

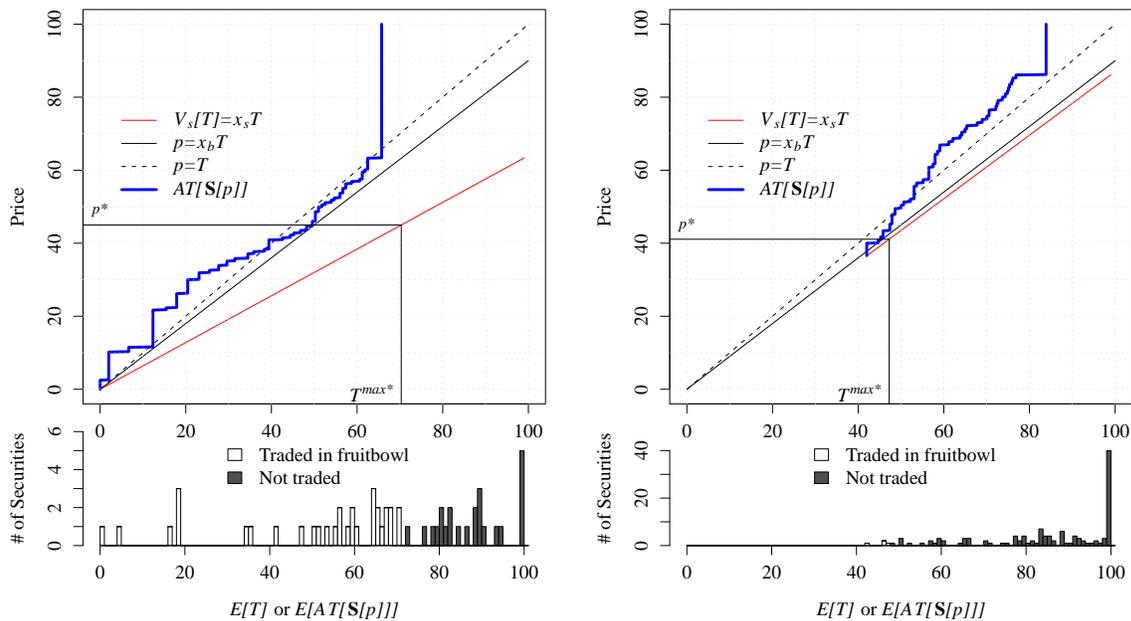


Figure 6: Trading of early vintage Sr-AAA rated CDOs. Left panel: High grade CDOs, with $x_s = 0.64$. Right panel: Mezzanine CDOs, with $x_s = 0.87$.

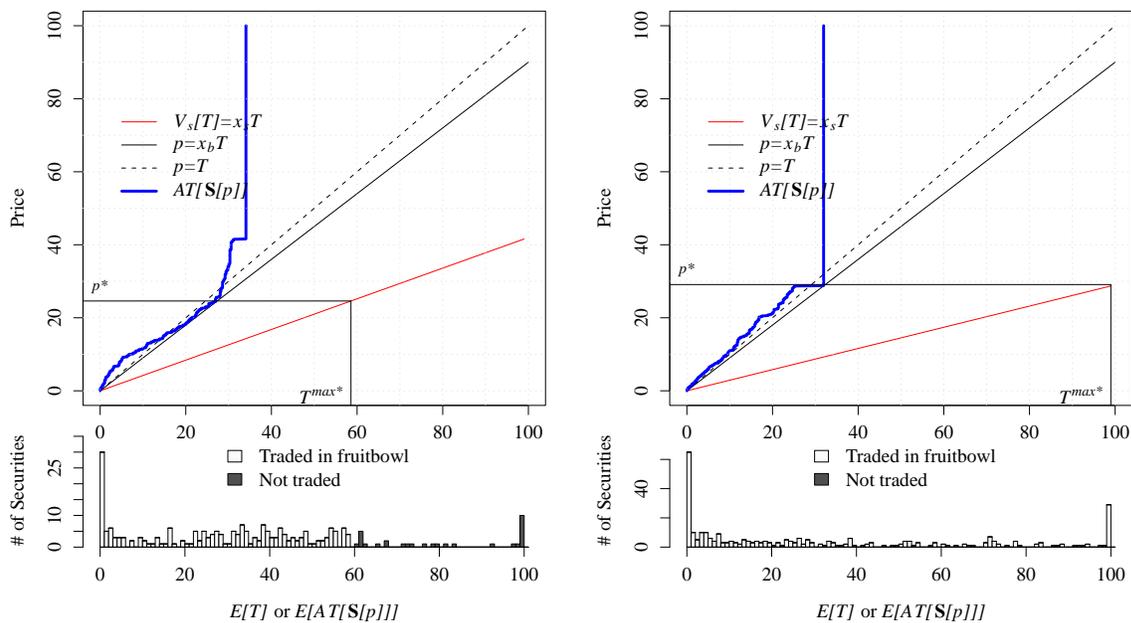


Figure 7: Trading of late vintage Sr-AAA rated CDOs. Left panel: High grade CDOs, with $x_s = 0.42$. Right panel: Mezzanine CDOs, with $x_s = 0.29$.