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**RISK AND RETURN IN CONVERTIBLE
ARBITRAGE: EVIDENCE FROM THE
CONVERTIBLE BOND MARKET**

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Risk and Return in Convertible Arbitrage: Evidence from the Convertible Bond Market

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Abstract

In this paper, we identify and document the empirical characteristics of the key drivers of convertible arbitrage as a strategy and how they impact the performance of convertible arbitrage hedge funds. We show that the returns of a buy-and-hedge strategy involving taking a long position in convertible bonds (“CBs”) while hedging the equity risk alone explains a substantial amount of these funds’ return dynamics. In addition, we highlight the importance of non-price variables such as extreme market-wide events and the supply of CBs on performance. Out-of-sample tests provide corroborative evidence on our model’s predictions. At a more micro level, larger funds appear to be less dependent on directional exposure to CBs and more active in shorting stocks to hedge their exposure than smaller funds. They are also more vulnerable to supply shocks in the CB market. These findings are consistent with economies of scale that large funds enjoy in accessing the stock loan market. However, the friction involved in adjusting the stock of risk capital managed by a large fund can negatively impact performance when the supply of CBs declines. Taken together, our findings are consistent with convertible arbitrageurs collectively being rewarded for playing an intermediation role of funding CB issuers whilst distributing part of the equity risk of CBs to the equity market.

Keywords: Hedge funds, Convertible Bonds, Convertible arbitrage, Supply, Risk Factors

JEL classification: G10, G19, G23

1. Introduction

At the turn of the century, capitalization of the global convertible bond (“CB”) market stood at just under \$300 billion while the US equity market was more than 50 times higher at over \$1.5 trillion. Yet during the difficult market conditions between 2000 and 2002 (with events such as end to the dotcom bubble, September 11, and the accounting scandals at Worldcom and Enron), the new issues in both these markets were of similar orders of magnitude—close to \$300 billion. Even during the financial crisis of 2007-2008, firms managed to raise about \$118 billion in the US CB market.¹ This underscores the importance of the CB market as a source of capital for corporations during adverse economic conditions.²

To smooth the placement of such a large-scale issuance of CBs, economic agents willing to assume the inventory risk are clearly needed. Coincident to these macro events, the last decade has witnessed a rapid growth of convertible arbitrage (“CA”) hedge funds. In spite of the rapid growth, assets employed by the convertible arbitrage strategy averaged around 4.51% of assets across all hedge fund strategies between December 1993 and June 2007 (TASS Asset Flow Report). On the other hand, Mitchell et al. (2007) point out that convertible arbitrage and other hedge funds make up about 75% of the convertible market. Accounts in the financial press support this view; for example, Pulliam (2004) notes that in 2003, CA hedge funds purchased about 80% of newly issued convertible bonds. Brown et al. (2010) document that a large fraction

¹ Equity data are from Federal Reserve Bulletin (various issues). We thank Jeff Wurgler for making it available on his website <http://pages.stern.nyu.edu/~jwurgler/>. CB estimate is from the public and private proceeds of convertible debt from Thomson Reuters Financial’s SDC Platinum database, which we also use later on for our out-of-sample analysis.

² Apart from being a useful source of liquidity to issuers during adverse market conditions, the issuance of CBs depends also on the costs and benefits compared to other forms of securities issuance. Firms selling CBs incur costs including issuance costs (e.g., underwriter fees and discounts) and dilution of existing stockholders’ interest upon conversion. Firms can also enjoy certain benefits from issuing CBs. These include reduction in the agency cost of debt (Green, 1984; Jensen and Meckling, 1976), mitigation of underinvestment problem due to adverse selection (Brennan and Kraus, 1987; Brennan and Schwartz, 1988; Constantinides and Grundy, 1989), avoidance of high costs of direct equity sales (Brown et al., 2010; Stein, 1992), and reduction of the costs of sequential financing while controlling overinvestment incentives (Mayers, 1998). Firms will only issue CBs if these benefits exceed the costs of issuing CBs.

of convertible issues are sold to hedge funds by issuers with greater stock volatility and higher probability of financial distress, thereby avoiding high costs of issuing equity, which will be even greater during poor market condition. Among different hedge fund strategies, while CA may not be the strategy with the largest assets under management (being naturally limited by the supply of CBs in the market), CA hedge funds do play a significant role in funding the convertible bond market.

In this paper, we posit that a typical CA hedge fund manager assumes the role of an intermediary—financing the CB issuers while distributing part of the equity risk of CB ownership to the equity market through delta hedging. To test our hypothesis, we explicitly model a commonly used trading strategy that gives us direct insight into the performance of CA hedge funds. Specifically, we assume that CA hedge funds take a long position in the CBs and mitigate the inherent equity risk by shorting the equity of the CB issuers. We demonstrate empirically that such a model explains a substantial amount of CA hedge funds’ return dynamics.

More specifically, our model has two main components—passive and active. The passive component is similar to the “buy-and-hold” strategy commonly used by mutual funds, while the active component resembles the “buy-and-hedge” strategy used by hedge funds. The passive component differs from the active component in two dimensions--leverage and risk management. Since mutual funds rarely use leverage (e.g., Almazán et al. (2004)), the amount of liquidity they provide to CB issuers is limited to the amount of assets they manage. In contrast, through the use of leverage, CA hedge funds can purchase a quantity of CBs well in excess of their capital and therefore can provide greater liquidity to the CB issuers.³ From a risk

³ Anecdotal evidence suggests that some CA hedge funds employ a leverage ratio of up to \$5 borrowing to \$1 equity (Zuckerman (2008)). Gupta and Liang (2005) apply the Value-at-Risk Approach to evaluate the capital adequacy of hedge funds and find that convertible arbitrage funds are better capitalized than funds in emerging markets, long/short equity, and managed futures categories.

management perspective, unlike CB mutual funds that typically do not short securities, CA hedge funds can hedge the equity risk embedded in the CBs by shorting stocks. Thus, unlike mutual funds, CA hedge funds can use leverage and provide much more liquidity to CB issuers at only moderate levels of overall portfolio risk. We capture these two dimensions by specifying a “buy-and-hedge” strategy, which involves buying CBs at issuance and holding them until maturity (or till the end of our sample period, whichever is earlier) and shorting the shares of the CB issuers to hedge the equity risk.⁴

As a funding intermediary for CB issuers, a convertible arbitrageur’s performance depends on the supply of CBs as well as discrete liquidity events such as the Long Term Capital Management (LTCM) crisis. The supply of CBs will affect the investment opportunities and therefore profitability of CA funds. Liquidity events can negatively impact CA funds’ ability to borrow short-term capital from brokers and raise long-term capital from investors thereby adversely affecting their ability to effectively implement the CA strategy. Therefore, liquidity events can also affect the risk appetite of arbitrageurs. We test these hypotheses by incorporating the impact of changes in supply conditions and major market events while modeling the return of CA hedge funds.

Using the daily prices of 1,646 US CBs from January 1993 to April 2003, we have five major findings. First, we show that a combination of buy-and-hold and buy-and-hedge strategies explains a significant proportion of the variation in CA hedge fund returns. In addition, we show empirically that the returns of CA hedge funds are positively related to the supply of CBs. Second, responding to adverse liquidity events, we show that after the LTCM crisis, CA hedge funds do reduce their reliance on the buy-and-hold strategy thereby paring their directional

⁴ Since we cannot directly observe the extent to which CA hedge funds actively hedge the equity risk as opposed to buying and holding CBs, we allow for both the passive and active components in our model and empirically estimate their relative importance in determining the CA funds’ performance.

exposure to the CB market. Third, combining both supply conditions and market events, we find alphas to be either insignificant or negative and significant. At first glance, persistent negative alphas appear to be at odds with the growth in CA funds. We show empirically that these observed negative alphas depend on the assumption underlying the monetization of specific measures of CB supply. Fourth, we find that larger funds rely more on the buy-and-hedge factor relative to the buy-and-hold factor, and are affected more by supply shocks. This is consistent with economies of scale that large funds enjoy in accessing the stock loan market. However, the friction involved in adjusting the stock of risk capital managed by a large fund can negatively impact performance when the supply of CBs declines. Finally, going beyond the data used to calibrate our model which ended in April 2003, we are able to corroborate the in-sample results by using a different out-of-sample data source spanning the period May 2003 to June 2007.

Our paper contributes to the growing literature on identifying risk factors that drive different hedge fund strategies' returns such as trend-following strategy (Fung and Hsieh (2001)), merger arbitrage strategy (Mitchell and Pulvino (2001)), equity-oriented hedge fund strategies (Agarwal and Naik (2004)), equity pairs trading strategy (Gatev et al. (2006)), and fixed income arbitrage strategy (Duarte et al. (2007); Fung and Hsieh (2004)). Our empirical findings complement the recent work of Choi et al. (2009) and Choi et al. (2010). Choi et al. (2009) find that short selling of equity securities by CA hedge funds improves stock liquidity but does not affect price efficiency after CB issuance. Choi et al. (2010) analyze how the supply of capital from CA funds affects CB issuance activity. In this paper, we show in an integrated framework that CA hedge funds short stocks as part of their risk management process, as distinct from speculative motives, and that the supply of CB issuance in turn impacts these funds' performance.

The rest of the paper is organized as follows. Section 2 describes the data. Section 3 outlines the models of CA strategies used by CA hedge funds. Section 4 provides a description of our empirical methodology and our findings during the sample period. Section 5 discusses results from fund level regressions. Section 6 describes findings from our out-of-sample analysis and Section 7 concludes the paper.

2. Data

The initial sample of daily CB data comprises 2,243 US dollar-denominated CBs kindly provided to us by Albourne Partners, London.⁵ From this data set we are able to recover 1,646 CBs over the sample period from January 1993 to April 2003 that have complete contractual information and daily closing prices needed for our empirical analysis. Table 1 provides descriptive statistics of our sample. Panel A shows that there are 1,224 firms issuing a total of 1,646 CBs with a majority of the firms having a single CB issue. Panel B reports statistics on issue size. The mean and median issue sizes are \$323 million and \$175 million, respectively. Panel C tracks the growth in the number of CB issues in our sample. The issuance activity in the US increased steadily from 91 issues in 1993 to 179 issues in 2002, reaching the peak of 211 issues and 265 issues in 1997 and 2001 respectively.⁶ This variation in the supply of CBs has important implications for the investment opportunity of CA hedge funds--an issue we explore in Section 4.2 of the paper.

⁵ In an earlier version of our paper, we had conducted the entire analysis using Japanese CB data as well. Our results were similar for Japan but for the sake of brevity, we focus only on the analysis with the US bonds here. It would be interesting to examine the structural differences between different CB markets by including Europe too. These issues are part of our ongoing research agenda.

⁶ Since our sample period ends in April 2003, the issuance figures for 2003 are not comparable to the other years.

3. A model of convertible arbitrage strategies

In order to capture the dynamic nature of hedge fund strategies, Fung and Hsieh (1997) put forward a model where a portfolio of hedge funds can be represented as a linear combination of a set of basic, synthetic hedge fund strategies. Ideally, these synthetic hedge fund strategies are rule-based constructs involving only observable asset prices, and are similar in concept to the familiar Carhart (1997) and Fama and French (1993) long-short factors used in conventional asset pricing models. Our model is a variation of this approach adapted for CA hedge funds. However, our research application is in a market in which transaction costs need to be incorporated. This departure from previous research methodology is described in the next section where we specify the structure of our model.

3.1. A buy-and-hedge strategy—the *X* factor

In the spirit of Fung and Hsieh (2001, 2004), we construct a single asset-based-style (ABS) factor using all new CB issues to create a hedged CB portfolio over time. This involves forming an issue-size-weighted portfolio of CBs by buying each CB at the first available price and holding it till its natural maturity or till the end of the sample, whichever comes first.⁷ This simplified model has the benefit of having well-defined entry and exit points for the arbitrageurs and avoids complex and often costly secondary market trading strategies. Equity risk is estimated by computing the daily return on an issue-size-weighted portfolio of the underlying stocks associated with each CB in the portfolio – denoted as EQ_t .

The idea behind using issue-size-weighting is two-fold. First, in terms of transaction costs, issue-size-weighting avoids the frequent and prohibitively expensive rebalancing induced

⁷ By natural maturity, we mean the point at which either the bond was converted to equity or redeemed by the issuer. In either case, there will be an exit price recorded in our data sample as in the case of a maturing bond where the initial principal is returned to the bond holder.

by daily bond price changes in an equally- (dollar-) weighted portfolio. Second, issue-size-weighting gives more weight to larger issues reflecting the impact of CB availability on arbitrageurs' returns. Note that our approach differs from a market-value weighted portfolio with a constant amount of capital invested. By increasing the invested capital of the hedged CB portfolio as new issues arrive, we avoid the need to sell existing holdings to make room for new issues that a fixed capital method entails. There is, however, an implicit assumption that CA hedge fund managers, in aggregate, are able to absorb new CB issues through capital infusion from their investors, through increased leverage, or both.

The buy-and-hedge strategy involves dynamically hedging the equity risk of the CB portfolio.⁸ For this purpose, we estimate the hedge ratios for equity risk by estimating the following regression over a 30-day rolling window:

$$R_{CB,t} = \gamma_0 + \gamma_1 EQ_t + v_t \quad (1)$$

where $R_{CB,t}$ is the day t return on the issue-size-weighted portfolio of CBs. The return on the buy-and-hedge strategy equals that on a *long* position in newly issued CBs and that on a *short* position in the corresponding stocks. The hedge ratios (or the short positions) are given by the slope estimates in equation (1).

⁸ Since CBs are also exposed to interest rate and credit risks, CA funds may also hedge these two risks in addition to the equity risk. Chan and Chen (2007), Davis and Lischka (1999), and Das and Sundaram (2004), among others, examine the impact of credit and interest rate risks on convertible bond pricing. We examine this variation of the buy-and-hedge strategy later in the paper. Further, CBs also may be subject to the risk of issuer calling the bonds. See for example Brennan and Schwartz (1977, 1980), Buchan (1997), Davis and Lischka (1999), Das and Sundaram (2004), Ingersoll (1977), McConnell and Schwartz (1986), and Tsiveriotis and Fernandes (1998). Brennan and Schwartz (1977, 1980) and Ingersoll (1977) also consider the optimal call strategies for convertible bonds. It is not obvious that call risk is a significant risk to CA funds. First, CB prices already reflect the likelihood of redemption (e.g., by trading at parity less the accrued interest). Second, there are also behavioral aspects of market agents that are hard to quantify. For example, Woodson (2002, p.28) notes that convertible issuers in Japan usually do not call their bonds to avoid upsetting their investors. Although this practice may be suboptimal with respect to short-term shareholder wealth maximization, it does suggest that CA funds are only subject to a limited amount of call risk in practice.

To capture the transaction costs in a buy-and-hedge strategy, we assume that the long position is financed by borrowing at the discount rate, $DISC_t$, which is proxied by the Fed Funds rate. The interest due on the cash balance from short positions in the underlying stocks (referred to as short rebate) is assumed to be at a lower rate than the borrowing rate by a spread, s , (or haircut) – for simplicity, we assume this to be 50 basis points for the US market.⁹ Another way of interpreting this 50 basis point spread is to think of it as the prime broker’s debit/credit spread for lending against the CB while paying interest on the cash balance from the short sale. This way, the reference level of interest rate cancels and the net cost to the CA hedge fund is this debit/credit spread of 50 basis points.¹⁰

After adjusting for these financing and transaction costs in this manner, the return corresponding to hedging the equity risk is $EQ_t - (DISC_t - s)$. We denote this cost-adjusted return by XEQ_t . The returns for our buy-and-hedge factor, X , is therefore given by

$$X_t = (R_{CB,t} - DISC_t) - \hat{\gamma}_1 XEQ_t \quad (2)$$

where X_t is day t return on following the convertible arbitrage strategy, $(R_{CB,t} - DISC_t)$ is the day t return on the bond portfolio adjusted for borrowing cost ($DISC_t$) associated with funding the long position in the hedged CB portfolio.

⁹ In general arbitrageurs earn a lower interest on the short stock position while paying higher interest on the amount borrowed for taking the long CB position. This borrowing-lending spread represents the cost of leveraging the position—call this the net spread for short. Note that the level of interest rate assumed is somewhat artificial as in a typical transaction the arbitrageur is both borrowing and lending money at the same time. Therefore the financing cost that matters in such transactions is the net spread which we assume to be 50bp. For example, an alternative way of reaching the same net spread of 50 bp could have been to borrow at the Fed funds rate plus 25 bp, and earn a short rebate at Fed funds rate minus 25 bp. Finally, we note that for the short rebate, negative interest rate is avoided as the Fed funds rate is higher than 25 bp throughout our sample period, ranging from 1.12% to 7.8% per annum.

¹⁰ Although it is quite possible that when liquidity conditions in the market is poor, this funding spread can rise. However, it is not clear that when credit markets are at extreme whether leverage will continue to be available at higher prices or simply dries up. Therefore, under normal market conditions, there may well be small variations to this simple assumption of 50 basis points funding spread the effect of which is unlikely to be material to our conclusions. Extreme market conditions are perhaps better modeled as discrete events separately and we do address this point separately in this paper.

3.2. *Descriptive statistics of buy-and-hedge strategies*

Table 2 reports the descriptive statistics of the X factor. Panel A shows that there are 411 CBs, on average, in the issue-size-weighted CB portfolio. These bonds have an average current yield of 13%, an average parity of 69%, and an average age (time since issuance) of 2.4 years.

Although the X factor has daily returns spanning the sample period from January 1993 to April 2003, we can only observe CA hedge fund returns on a monthly basis. Thus, we compound the daily returns on the X factor into monthly returns for our statistical analyses. Table 2 Panel B provides the descriptive statistics of these monthly returns over our sample period. The X factor has average monthly return of 0.17% during the sample period.¹¹ Although the average monthly return is positive, the factor return is quite volatile. The sample standard deviation of the X factor is about six times its monthly average return (1.01% vs. 0.17%).

3.3. *Buy-and-hold component of CA hedge fund returns*

We use the returns of the *Vanguard Convertible Securities* mutual fund (*VG*), one of the largest mutual funds investing in U.S. dollar denominated CBs, to proxy the performance of the passive *buy-and-hold* component for US CBs. Our choice of using Vanguard fund returns instead of a CB index to proxy for buy-and-hold returns is driven by the fact that Vanguard returns account for the costs associated with acquiring and carrying a long position in the CB market. Further, in contrast to a CB index, the Vanguard fund is investable.

¹¹ It should be noted that in theory, these returns only require an amount of equity margin pledged with the CA hedge funds' prime brokers in order to collateralize the long/short transactions. Therefore, depending on the implicit leverage used by a CA hedge fund, the return on equity margin can be significantly higher than the reported figures.

Panel A of Table 3 reports the descriptive statistics of the returns on the passive buy-and-hold strategy. Comparing the results in Tables 2 and 3, we observe that the mean monthly return of $VG(X)$ is 0.32% (0.17%) with a standard deviation of 3.72% (1.01%). These comparisons suggest that the buy-and-hedge strategy offers a better risk-return tradeoff than the buy-and-hold strategy in the CB market. For example, the ratio of mean return to sample standard deviation is 0.168 (0.086) for $X(VG)$.

3.4. *Representative portfolios of the CA hedge fund universe*

Since many hedge funds report solely to one database and there is little overlap between different databases (see Agarwal et al. (2009)), in order to construct a broad and more representative sample of CA hedge funds, we use data on individual hedge funds from three different databases — the *Centre for International Securities and Derivative Markets (CISDM)*, *CS Tremont (CT)* (now Lipper TASS), and *Hedge Fund Research (HFR)*. Specifically, we create two CA portfolios (equally-weighted (EW) and value-weighted (VW)) from the 155 unique funds out of a total of 207 funds that are classified as CA hedge funds spanning the sample period January 1993 to April 2003.¹² For the sake of completeness, we also repeat our analysis using easily available and widely used CA indexes based on these three databases. It is important to note that hedge fund indexes are generally constructed from widely diverging samples employing different index construction methods. For example, CISDM uses returns of the median fund, HFR uses equally-weighted returns, and CT employs value-weighted returns.

¹² From the 207 CA hedge funds from the three databases, we identify duplicates by matching funds by name and by comparing their return histories. When a fund appears in more than one database, we select the fund from the database that has the longest return history for that fund. Eliminating duplicate funds yields a final sample of 155 distinct CA hedge funds. The universe of CA hedge funds has grown substantially over our sample period from 29 funds managing just under \$1 billion at the beginning of 1993 to 119 funds managing about \$18 billion at the end of 2002.

Given the differences in index construction and different index inclusion requirements for funds, throughout the paper, we focus primarily on the results for our two CA portfolios (EW and VW), which are constructed using the union of these three databases. We also briefly discuss the results with the CISDM, CT, and HFR indexes.

Panel A of Table 3 reports the descriptive statistics of monthly returns of the two CA portfolios (EW and VW) and the three CA indexes. The mean (median) monthly returns for these indexes and portfolios range from 0.47% to 0.61% (0.68% to 0.79%) while the standard deviations (SDs) range from 0.67% to 1.38%. As expected, the broad-based variables targeting the central tendency of CA hedge funds' performance (EW, VW, CISDM, CT, and HFR) fall within a small range.

Panel B reports the correlations among the CA portfolios, the three CA indexes, VG, and the X factor. Not surprisingly, the CA portfolios and CA indexes are highly correlated with each other (correlations ranging from 0.80 to 0.93). Their correlations with VG are positive but lower in magnitude; the correlations range from a low of 0.34 (CT) to a high of 0.60 (EW). The CA portfolios and CA indexes are all positively correlated to the X factor, consistent with the use of the buy-and-hedge strategy by CA hedge funds.

Next we conduct a multivariate analysis of CA fund returns using a two-factor model consisting of both buy-and-hedge and buy-and-hold factors.

4. Empirical methodology and results

4.1. A model of CA hedge fund returns

Our analysis begins with the following regression model:¹³

¹³ Unless otherwise stated, we use Newey and West (1987) heteroskedasticity-and-autocorrelation consistent standard errors to compute t-statistics and *p*-values for regression coefficients.

$$CA_t = \theta_0 + \theta_1 X_t + \theta_2 VG_t + \psi_t \quad (3)$$

Here the time series of monthly returns for each CA portfolio and index, denoted by CA_t , are regressed against the monthly returns of the buy-and-hedge and buy-and-hold strategies. Since we cannot directly observe the extent to which CA hedge funds actively hedge the equity risk as opposed to buying and holding CBs, we allow for both the passive and active components in our model and empirically estimate their relative importance in determining the CA funds' performance.

Results in Table 4 show that the two CA portfolios (EW and VW) have significant exposures to X and VG , which is consistent with both buy-and-hedge and buy-and-hold strategies contributing to the CA returns. We also observe a significant intercept (or alpha) of 40 basis points per month for both EW and VW.¹⁴ The adjusted- R^2 for the EW and VW portfolios are 52.82% and 41.50%, respectively, suggesting that our simple model of using two asset-based strategies (buy-and-hedge and buy-and-hold) does a good job of explaining the dynamics of CA fund returns.¹⁵ The results for the three CA indexes are qualitatively similar to those for the two CA portfolios. We consistently observe significant loadings on X and VG , with explanatory power ranging from 24.20% to 48.18%.

As a robustness check, we repeat our analysis using only CBs that mature within the sample period to construct an alternative version of the buy-and-hedge factor.¹⁶ The results (not reported in the table) are qualitatively similar. Both the buy-and-hedge factor and the buy-and-

¹⁴ It is important to note that this specification assumes constant factor loadings through the sample period. Arguably, hedge funds may change their weights on the buy-and-hold and buy-and-hedge strategies in response to changes in investment opportunities and market conditions, an issue we explore in detail, later in the paper.

¹⁵ To further highlight the economic importance of the X factors, we also compare the adjusted- R^2 from our model with adjusted- R^2 's from other models used in the hedge fund literature including the Carhart (1997) model augmented with out-of-the-money call and put options on S&P 500 index as in Agarwal and Naik (2004) and the Fung and Hsieh (2004) seven-factor model. The results in Appendix A show that the explanatory power from these models ranges from 7% to 30%, which is substantially lower than the range of 24% to 53% obtained with our more parsimonious model.

¹⁶ We thank the referee for suggesting this possibility.

hold factor continue to be positively related to CA returns; the coefficient of the buy-and-hedge factor is statistically significant in all cases but one and the coefficient of VG is always statistically significant. The two-factor model produces adjusted R-squares ranging from 11% to 39% but lower than those reported in Table 4. This suggests that the buy-and-hedge factor that uses all CBs does a better job of explaining CA returns during our sample period.

Since CBs are also exposed to interest rate risk and credit risk, in theory, CA funds may also hedge these two risks in addition to the equity risk. However, the extent to which they do so in practice must depend on the costs and benefits of hedging these risks which is an empirical question. To answer this question, we repeat our analysis using an alternative buy-and-hedge factor, X^A , which hedges all three sources of risk (interest, credit and equity). X^A is constructed as follows. Equity risk is captured in EQ_t , as in the case of factor X . The impact of credit risk inherent in CBs is proxied by the daily change in the spread between the yields of Baa corporate bonds and 10-year U.S. Treasury bonds. To incorporate the impact of interest rate risk, we use the daily yield of 5-year U.S. Treasury bonds.¹⁷ The risks of the portfolio of CBs are hedged using the following regression model:

$$R_{CB,t} = \gamma_0 + \gamma_1 EQ_t + \gamma_2 IR_t + \gamma_3 CR_t + \eta_t \quad (4)$$

Here $R_{CB,t}$ is the day t return on the issue-size-weighted portfolio of CBs, EQ_t is the day t return on the issue-size-weighted portfolio of underlying stocks, IR_t is the day t interest rate proxy, and CR_t is the day t credit risk proxy. The return to the strategy that hedges all three risks follows that of a portfolio comprising a *long* position in the issue-size-weighted portfolio of CBs, a short position in the issue-size-weighted portfolio of the corresponding stocks, a short position in

¹⁷ We obtain these from the Federal Reserve Board website and Datastream.

government bonds, and a short position in the spread between corporate and government bonds.¹⁸ The short quantities are obtained from the slope estimates in equation (4).

After adjusting for financing and transaction costs, the returns corresponding to hedging the equity, credit, and interest rate risks are $EQ_t - (DISC_t - s)$, $CR_t - (DISC_t - s)$, and $IR_t - (DISC_t - s)$ respectively. We denote these cost-adjusted returns by XEQ_t , XCR_t , and XIR_t . The cost-adjusted returns from the alternative buy-and-hedge factor, X^A , is given by

$$X_t^A = (R_{CB,t} - DISC_t) - \hat{\gamma}_1 XEQ_t - \hat{\gamma}_2 XIR_t - \hat{\gamma}_3 XCR_t \quad (5)$$

where X_t^A is day t return on following the convertible arbitrage strategy, $(R_{CB,t} - DISC_t)$ is the day t return on the bond portfolio adjusted for borrowing cost ($DISC_t$) associated with funding the long position in the CB portfolio, and XEQ_t , XCR_t , and XIR_t are as defined above.

Table 5 reports the results from the regression model of equation (3), re-estimated using VG and X^A . We observe that the loadings on the two factors are similar to those in Table 4. However, using the alternative buy-and-hedge factor lowers explanatory power across all CA portfolios and indexes. For example, for the EW CA portfolio, the two-factor model with the original buy-and-hedge factor (X) achieves an adjusted-R² of 52.82%, but the two-factor model with the alternative buy-and-hedge factor (X^A) produces a lower adjusted-R² of 43.68%. This suggests that our specification of hedging equity risk alone is a better characterization of the buy-and-hedge strategy used by CA hedge funds in practice. Therefore, in the following analyses, our buy-and-hedge strategy involves taking a long position in CBs and delta hedging the equity risk.

¹⁸ The stocks in the stock portfolio receive the same weight as the corresponding CBs. The interest rate and credit variables are indexes. Hence, there is no weighting for these variables.

These empirical results suggest that CA hedge funds behave in a manner consistent with that of an intermediary who provides funding to the CB issuers while acting as an equity-risk transfer agent distributing risk from the CB to the equity market.

4.2. *How does the supply of CBs affect convertible arbitrageurs?*

Viewed as a funding and risk intermediary for CB issuers, a convertible arbitrageur's performance in month t should depend on the supply of new CB issues available for trading month t , NI_t . However, simply relying on CB issuance dates recorded in databases produces an inaccurate measure of NI_t as these bonds often trade in the "when-issued" (WI) market--which exists between the announcement date of a new issue and ends on the issuance date that can be several months hence.¹⁹ The existence of the WI market means that the profit/loss effect of a new CB issue can enter into a CA fund's performance well before the recorded issuance date of newly issued CBs as WI CB positions are marked to market for accounting purposes. The combination of when issued trading and a potentially long settlement period between announcement and issuance dates give rise to a measurement problem if one uses issuance dates to construct a supply proxy. In other words, NI_t contains bonds with issuance dates in months $t+1$, $t+2$, or beyond. Therefore, the first step to constructing an accurate proxy of CB supply is to gauge how far in time CB deals are announced in advance of eventual issuance. To that end, we randomly sample (without replacement) 100 CBs that are issued during our sample period and manually collect the earliest announcement date from Factiva news search supplemented with

¹⁹ Chacko et al. (2005) study the determinants of liquidity in the US corporate bond market using a dataset of when-issued and secondary market trades. Nyborg and Sundaresan (1996), among others, study when-issued trading in the US Treasury market. In private correspondence, a Financial Industry Regulatory Authority (FINRA) representative confirms the existence of when-issued trading in convertible bonds.

searches on the SEC EDGAR database and Google.²⁰ Ninety-three of the 100 randomly selected bonds have complete information.

Table 6 displays the frequency distribution of the announcement month relative to the issuance month. About two-thirds of the announcements are made in the same month as the issuance month while the remaining announcements are made before the issuance month. In particular, about 24% of the CB deals are announced one month before the issuance month and another 6% are announced two months before. Of the two remaining CB issues, one is announced four months before issuance while another is announced five months before issuance. This distribution suggests that a reasonable proxy of NI_t includes CBs with issuance dates in months $t+1$ and $t+2$. Accordingly, we construct our month t supply variable, $LDeal_t$, as

$$LDeal_t = \log(1 + CB_{t+1} + CB_{t+2}) \quad (6)$$

where $\log(\cdot)$ is the natural logarithm function, and CB_{t+1} (CB_{t+2}) is the number of CB issues with issuance dates in month $t+1$ ($t+2$).²¹ To test whether supply conditions affect CA performance, we estimate the following regression model:

$$CA_t = \alpha + \beta_1 X_t + \beta_2 VG_t + \gamma LDeal_t + \chi_t \quad (7)$$

where $LDeal_t$ is the supply variable in month t and the other variables are as in equation (3).

Table 7 reports the results from applying equation (7) to the two CA portfolios and the three CA indexes. The coefficients for X and VG with respect to the two CA portfolios, EW and VW, remain positive and statistically significant. In addition, the new supply variable $LDeal$ has a positive and highly significant coefficient for both the portfolios. After accounting for supply effects via the $LDeal$ variable, both CA portfolios exhibit significant negative alphas. This is in

²⁰ Our CB data set from Albourne Partners contains the issuance dates, but not the announcement dates. In unreported work, we find that the distribution of sampled bonds across years is similar to the distribution for the entire sample of 1,646 bonds.

²¹ For robustness, we repeat our tests using the logarithm of the number of CB issues in month t , and find our results to be qualitatively similar.

sharp contrast to the positive alphas previously reported in Table 4 before the supply effects of CBs are incorporated into our model. Furthermore, despite the reduction in the degrees of freedom, the introduction of the supply variable leads to an improvement in the explanatory power of our model—adjusted-R² increases from 52.82% to 59.70% for EW, and from 41.50% to 55.26% for VW. The results for the three CA indexes are qualitatively similar. Taken together, the evidence highlights the importance of changing investment opportunities on the performance of CA hedge funds.

At first glance, the occurrence of significant negative alphas appears to be at odd with the growth in CA hedge funds. To explore this phenomenon, we examine alternative ways of incorporating the effect of the WI market on the *LDeal* variable. Unlike other factors in our model, *LDeal* is a non-price construct and is not directly investable. The monetization of *LDeal*'s effect on performance is imprecise. The possibility exists that we may have over compensated for the when-issued trading effect of the *LDeal* variable. Short of precisely quantifying the exact announcement dates of thousands of bonds manually, there is no obvious way to accurately assess this possibility. Here we report our results based on a measure of *LDeal* that errs on the side of conservatism. Overall, without the *LDeal* variable, one observes positive alpha. Using a conservative estimate of *LDeal* leads to the observed negative alpha. Both sets of results point to the presence of a supply effect.²²

Thus far the empirical analyses indicate that convertible arbitrageurs benefit from providing funding to CB issuers while simultaneously offsetting the bond's equity risk utilizing the liquidity of the equity markets. However, for CA hedge funds to perform this function effectively they have to secure funding, and sometimes have to leverage their portfolio to

²² We repeat our analyses using a version of *LDeal* that is based only on supply in month $t+1$ (i.e., $LDeal = \log(CB_{t+1})$) and find that the estimated alphas fall between the alphas we currently report and the alphas based only on supply in month t .

purchase the CBs. The ability of CA hedge funds to borrow short-term capital from prime brokers as well as to attract longer term capital from their investors will in turn depend on prevailing liquidity conditions in the market place. In this context, the question arises as to how convertible arbitrageurs respond to market upheavals that interrupt their ability to fund their portfolios? To gain insight into this question, we examine how CA hedge funds manage extreme events such as the LTCM crisis.

4.3. *The impact of extreme market events on CA hedge funds*

We investigate the effect of the LTCM event on CA fund returns by testing for structural breaks to our model. Following Brown et al. (1975), we apply the CUSUM procedure and confirm that the LTCM crisis indeed lead to consistent boundary violations across the two CA portfolios and the three CA indexes. Similar to Fung and Hsieh (2004) and Fung et al. (2008), we use the following structural break model to account for the LTCM crisis:

$$CA_t = D1(\omega_0 + \omega_1 X_t + \omega_2 VG_t + \gamma LDeal_t) + D2(\omega_0 + \omega_3 X_t + \omega_4 VG_t + \pi LDeal_t) + \kappa_t \quad (8)$$

Here the variables are as defined before in equations (3) and (7).

The results from this joint estimation are reported in Table 8. There is a general decline in the exposure to the buy-and-hold strategy uniformly across the CA portfolios and indexes compared to the results reported in Table 7. The coefficient on $D1 \times VG$ is smaller than that on $D2 \times VG$ for EW, CISDM, and HFR at varying levels of statistical significance. For the two AUM-weighted constructs, the VW portfolio and the CT index, post-LTCM exposure to the directional factor, VG , appear to have weakened both in slope coefficient as well as statistical significance. These results are broadly consistent with an increase in risk aversion post LTCM. Collaborative evidence can be seen in post-LTCM increases in the exposures to the buy-and-

hedge factor across all cases. However, we can only confirm the statistical significance for all the CA indexes but not for the CA portfolios (see the F-test results for $DI \times X$ in the lower panel of Table 8). In terms of the effect of CB supply on returns, the results are mixed. Both the CA portfolios and indexes exhibit significant positive loading on the supply variable (*LDeal*) before the LTCM crisis similar to the results reported in Table 7. However, post LTCM, the coefficient on *LDeal* is positive but only weakly significant (at the 10% level) for EW. But the F-tests failed to reject the null that *LDeal* has identical coefficients before and after the LTCM crisis, with the exception of VW. Taken together, there seems to be a preponderance of evidence pointing towards a decrease in risk appetite (or an increase in risk aversion) after the LTCM crisis. However, the variability of the results across the two portfolios (EW and VW) and three indexes (CISDM, CT, and HFR) indicate potential behavioral differences in the cross section of funds—an issue we investigate in more detail in the next section.

Finally, in terms of explanatory power, there is a uniform increase in adjusted- R^2 compared to the results in Table 7. For example, with CISDM as the dependent variable, equation (8) produces an adjusted- R^2 of 75.49%, which is substantially higher than the adjusted- R^2 in Table 7 (55.44%). More importantly, both the pre-LTCM and post-LTCM alphas are either insignificantly different from zero or negative. Overall, these findings confirm a lack of abnormal performance from CA funds on average, after adjusting for the changes in investment opportunities and the structural changes arising from extreme market events such as the LTCM crisis. The natural question that arises is whether these conclusions apply uniformly across the funds in our sample.²³ This is the subject of the next section.

²³ We thank the referee for directing us towards analyzing individual CA funds.

5. Individual fund level analyses

The empirical results in the previous section point to potential differences in return characteristics between the small funds (weighted higher in the EW portfolio as well as indices such as the HFR and CISDM) and the larger funds, which have greater weighting in the VW portfolio and the CT index. To investigate these differences, we begin by identifying all CA hedge funds with at least 24 months of data on returns and assets under management during our sample period. We then compute the time-series average of the assets under management for each fund to sort funds into size terciles (Small, Medium, and Large). This procedure yields 36 CA hedge funds in each tercile, for a total of 108 funds. Within each size tercile, we estimate equation (7), the two-factor plus supply effect model, for each fund. The summary statistics of the respective size terciles and regression results are reported in Table 9 Panel A.

The sample moments of returns reveal differences across size terciles. Larger funds (Medium and Large terciles) appear to generate higher returns than small funds. The mean (median) monthly excess return on the Large, Medium, and Small portfolio is 0.51%, 0.65%, and 0.39% (0.66%, 0.78% and 0.63%), respectively. The largest and smallest funds experience greater performance variability than the medium funds. The Large and Small portfolios have monthly standard deviations of 1.25% and 1.29%, respectively, while the Medium portfolio has a monthly standard deviation of 1.05%. Higher performance variability implies more extreme returns that can create greater divergence between the mean and median. We do observe such a pattern in the data. The Small and Large portfolios, with higher standard deviations, exhibit larger differences between mean and median returns than the Medium portfolio. In terms of Sharpe ratio, the Medium portfolio has the highest Sharpe ratio (0.62) while the Small portfolio has the lowest Sharpe ratio (0.30).

Turning to the regression results, we see that the average coefficient of X is positive and statistically significant at the 1% level, indicating that the buy-and-hedge factor continues to be a significant explanatory variable among CA hedge funds of varying sizes. The average coefficient of VG is positive and statistically significant for the small and medium terciles but not for larger CA hedge funds. The average coefficient of $LDeal$ is positive for all terciles, but is statistically significant only for the medium and large terciles, suggesting that the supply variable is more important in explaining the returns of medium and large funds. These findings are consistent with larger funds being less dependent on security selection to support performance for their capital base. This observation is, in turn, consistent with a diseconomy of scale to skill-based investment strategies resulting in larger funds being more dependent on systematic, factor-related returns for performance. Consequently, declines in the supply of convertible bonds are likely to negatively impact them more than their smaller brethren. Consistent with the results in Table 7, the non-factor related returns from this three-factor model are not statistically different from zero for smaller funds and negative for medium and larger funds. However the model's explanatory power, adjusted R^2 , for each of the three terciles is notably lower than those reported for the broader averages in Table 7.

Next, we examine how well the structural break model explains fund returns across the size terciles. For this analysis, we apply a different selection procedure to ensure that the funds selected have a sufficiently long return history that spans the LTCM crisis. Specifically, we only include CA hedge funds with at least 48 months of data on returns and assets under management during our sample period to ensure sufficient degrees of freedom for efficient estimation. In addition, the funds must have existed before, during, and after the LTCM crisis. In other words, the return series must span the LTCM crisis. This second condition ensures that $D1$ and $D2$, as

well as all associated interaction terms can be estimated for each fund. The selection procedure yields 15 CA hedge funds for each size tercile.²⁴ Panel B of Table 9 reports the average regression coefficients, adjusted- R^2 , and fund characteristics. Because our selection procedure yields a subset of the original size terciles (employed in Table 9 Panel A), we also report summary statistics of monthly excess returns on these new size terciles. To summarize tests of differences of coefficients among the funds, we report the number of funds for which there is a statistically significant increase or decrease in coefficient after the LTCM crisis.²⁵

In terms of sample moments of returns, we continue to observe similar patterns to the results in Table 9 Panel A. Specifically, larger funds (Medium and Large terciles) appear to generate higher returns than small funds. For example, the median monthly excess return on the Large, Medium, and Small portfolio is 0.77%, 0.67%, and 0.54%, respectively. Performance variability is higher among the largest and smallest funds than among the midsize funds; the Large (Small) portfolio has a monthly standard deviation of 2.12% (1.25%) while the Medium portfolio has a monthly standard deviation of 0.91%. As expected, greater return variability in the Large and Small portfolios result in more pronounced differences between mean and median returns. The Medium portfolio has the highest Sharpe ratio (0.69) and the Large portfolio has the lowest (0.23).

The fund-level regression results add clarity to those of CA indexes and portfolios in several respects. First, they clarify that the exposure to the directional variable, VG , reported in Table 8 for the broad-based portfolios and indexes are primarily driven by the smaller funds.

²⁴ In unreported work, we compute the grand average age for each size tercile and find no systematic relation between size and age. Grand average age (in years) in the Small, Medium, and Large tercile is 5.7, 4.1, and 4.4, respectively. There is no apparent age bias in the terciles. Interestingly, the higher age of small funds is consistent with their greater survival rate and less negative abnormal performance based on the results in Table 9 Panel A.

²⁵ We identify a change in coefficient as statistically significant if the F-statistic implies a rejection of the null of no change at the 10% level or lower.

Panel B of Table 9 now confirms that on average, neither the Medium nor the Large tercile of funds exhibit statistically significant exposure to the VG factor. This result is not affected by the LTCM event. These findings also clarify that the exposure to the buy-and-hedge factor, X , reported in Table 8 for the broad-based portfolios and indexes are primarily driven by the larger funds both pre and post the LTCM crisis. Panel B of Table 9 reports collaborative evidence for this observation. Second, there is an increase in explanatory power across all size terciles compared to the basic two-factor model. The average adjusted- R^2 rises from 23.60% to 31.84% for the small tercile (comparing Panel A to Panel B of Table 9), 25.41% to 43.99% for the Medium tercile and more than doubling for the Large tercile going from 18.01% to 39.29%. Third, as a gauge of whether there is a change in risk-taking behavior, we examine the number of funds for which there is a statistically significant change in coefficient on VG after the LTCM crisis. In both Small and Medium terciles, the number of statistically significant decreases in VG 's coefficient is higher than the number of statistically significant increases. For example, among the smallest funds, there are 4 cases of a decrease in coefficient and only 1 case of an increase. Thus, for the Small tercile, the decline in the average coefficient is representative of changes at the individual fund level. This is consistent with our earlier findings at the average (index) level showing that greater risk aversion post the crisis reduces the use of the buy-and-hold strategy among the arbitrageurs. Fourth, pre LTCM, the supply variable figures significantly as a return driver for all three terciles of funds. Post LTCM, the supply variable no longer figures significantly for the broad-based portfolios and CA indexes (see Table 8), but here the variable continues to figure significantly for the Medium size tercile. In order to further explore the divergent impact of the supply of CBs on returns post LTCM across CA hedge funds, we appeal to a cumulative residual analysis of our model around the LTCM event.

For each tercile, we first estimate equation (8) *without* the $LDeal$ variable for each fund and use the factor loadings to compute each fund's monthly residuals. Specifically, in the period prior to and including the LTCM month (from March 1998 to September 1998), fund i 's residual in month t , $Residual_{i,t}$, is the difference between its excess return and the product of the pre-LTCM factor loadings and factor realizations, i.e.,

$$Residual_{i,t} = CA_{i,t} - (d1X_i \times X_t + d1VG_i \times VG_t) \quad (9)$$

where $CA_{i,t}$ is fund i 's excess return in month t , $d1X_i$ ($d1VG_i$) is fund i 's pre-LTCM coefficient on X (VG), and X_t (VG_t) is the buy-and-hedge (buy-and-hold) factor realization in month t . Similarly, for the 10 months after LTCM (from October 1998 to July 1999)²⁶, $Residual_{i,t}$ is computed as the difference between its excess return and the product of the post-LTCM factor loadings and factor realizations.

$$Residual_{i,t} = CA_{i,t} - (d2X_i \times X_t + d2VG_i \times VG_t) \quad (10)$$

where $CA_{i,t}$ is fund i 's excess return in month t , $d2X_i$ ($d2VG_i$) is fund i 's post-LTCM coefficient on X (VG), and X_t (VG_t) is the buy-and-hedge (buy-and-hold) factor realization in month t .

Next we average the fund level residuals across funds each month to obtain the equally-weighted residual for each tercile. We then cumulate the equally-weighted residuals to obtain cumulative residuals for each tercile. In Figure 1, we plot the cumulative residuals for the size terciles from March 1998 to July 1999. The cumulative residual plots show that all three groups appear to have recovered from the LTCM event, albeit at varying speeds, by the middle of 1999.

Accompanying the residual plots is the plot of cumulative CB issuance ('cumulative supply'), which shows that CB issuance dropped dramatically during the crisis (the flat portion

²⁶ In July 1999, 12 CBs were issued, a number that is comparable to the pre-LTCM issuance of 13 CBs in July 1998 and 12 CBs in August 1998. Thus, we choose to end the post-LTCM period in July 1999 to examine whether the recovery of CB supply to approximately pre-LTCM levels has differential impact on fund performance after the LTCM crisis.

of the plot) and gradually recovered to pre-crisis issuance level by July 1999. Since our regression results in Panel B of Table 9 suggest that smaller funds are less reliant on CB issuance than larger funds, it is interesting to explore the extent to which the different size terciles rely on CB supply for their post-crisis recovery. To that end, we construct cumulative residual plots that account for supply effects. Specifically, we compute monthly residuals at the individual fund level using fund excess return, factor loadings from estimating equation (8), factor realizations, and realizations of $LDeal$. We then compute equally-weighted residuals for each tercile and cumulate those residuals for the plots. In the period prior to and including the LTCM month (from March 1998 to September 1998), fund i 's residual in month t is computed as,

$$Residual_{i,t} = CA_{i,t} - (d1X_i \times X_t + d1VG_i \times VG_t + d1LDeal_i \times LDeal_t) \quad (11)$$

where $d1LDeal_i$ is fund i 's pre-LTCM coefficient on $LDeal$ and $LDeal_t$ is the realization of $LDeal$ in month t and all other variables have been previously defined. From October 1998 to July 1999, $Residual_{i,t}$ is computed as,

$$Residual_{i,t} = CA_{i,t} - (d2X_i \times X_t + d2VG_i \times VG_t + d2LDeal_i \times LDeal_t) \quad (12)$$

where $d2LDeal_i$ is fund i 's post-LTCM coefficient on $LDeal$ and all other variables have been previously defined.

The cumulative residual plots net of supply effects (Figure 2) show that, conditional on survival, the surviving small funds appear to recover from the LTCM event by April 1999 while the surviving Medium and Large funds continue to show losses up till July 1999, at which point CB issuance had recovered to pre-crisis levels. The post-LTCM upward trend in Small's cumulative residuals plot and the downward trend in the plots of Medium and Large confirm the regression results in Panel B of Table 9 that small funds are less affected by supply changes but this is not the case for the larger funds in the Medium and Large terciles. Both larger groups

depend on new CB issues without which, the cumulative residual continues to decline, suggesting little or no recovery of losses on non-factor related return from the LTCM event itself.

Overall, our analysis at the individual fund level reveals substantial cross-sectional variation in the use of buy-and-hedge strategy relative to the buy-and-hold strategy, and the impact of supply and market conditions on the abnormal performance of CA funds. As the description of CA hedge funds suggests, there is a growing (as measured in the assets under management) dependence on hedged convertibles bonds as a strategy for larger funds. In contrast, smaller CA hedge funds are more dependent on the directional performance of CBs and less reliant on hedging. Consequently when a major market upheaval such as the LTCM crisis comes along, the surviving small funds are able to reap the benefits sifting through the wreckage left behind from funds that had to deleverage or were liquidated entirely. For larger funds, security selection is not the main performance driver given their bigger capital base and their dependence on hedged convertibles, factor X , as a strategy. Pre and post a major market crisis such as LTCM, any loss due to deleveraging appears to be almost permanent or at least to take a much longer time to recoup (judging by the cumulative residual analysis) compared to their smaller, more nimble, brethrens that survived. As such, marginal increases/decreases in the supply of bonds would appear less relevant to their long-term performance. In contrast, the Medium size funds continue to exhibit, pre and post LTCM, consistent exposures to the buy-and-hedge factor as well as being performance reliant on the supply of CBs. Thus far, these empirical conclusions are based on data from hedge funds and CBs over two five-year periods surrounding the LTCM event. The natural question to ask is whether the return drivers we have identified continue to be the key determinants of CA investors' portfolios.

6. Out-of-sample analysis of funded CA portfolios

6.1. Construction of the variables for the out-of-sample period

In this section we focus on an out-of-sample period between May 2003 and June 2007. During this period, investable versions of the CA indexes from CT and HFR (“CTX” and “HFRX” for short) were launched and had successfully attracted investors’ capital. These products are designed to offer investors a broad exposure to CA hedge funds with explicitly defined, rule-based portfolio constructions. As such, they inherit all the trading frictions of a real-life CA hedge fund portfolio bringing in some of the practical considerations that are typically assumed away in statistical indexes.²⁷

For consistency, we adjusted the two return factors to better reflect the trading friction of the dependent variables. In terms of the buy-and-hold factor, VG , is available and remained open as a mutual fund throughout the out-of-sample period. To proxy an investable version of the buy-and-hedge factor, X , we employ an alternative approach. We use VG to proxy the long-CB component of the buy-and-hedge factor. To estimate the hedge ratio for the short-equity component, we regress VG on the Russell 2000 index:

$$VG_t = \gamma_0 + \gamma_1 RUS_t + v_t \quad (13)$$

where RUS_t is the month t return on the Russell 2000 index.²⁸ To obtain the equity hedge ratio for month t , we perform an OLS estimation of equation (13) using data from the previous 24 months.

The slope coefficient estimate for RUS_t , $\hat{\gamma}_1$, serves as the equity hedge ratio for month t . We then compute the buy-and-hedge factor in month t as:

²⁷ Practical considerations include issues such as whether a fund is open to new investors, the friction involved in rebalancing a portfolio with funds that stipulate different redemption terms and free of survivorship biases that may exist in indexes of hedge funds—see Fung and Hsieh (2004) for discussions of these issues.

²⁸ Both ETFs and Futures Contract on the Russell 2000 exist and closely track the index. However, to ensure we match the data period in our analysis with the investable hedge fund indices and to avoid basis considerations that may arise from the futures contract, we opted to use the index return series instead.

$$X_{O,t} = VG_t - \hat{\gamma}_1 RUS_t \quad (14)$$

A rolling 24-month window is used to repeat the two-step procedure. In particular, we use the monthly returns on VG and RUS from May 2001 to June 2007 with the first (last) estimation window corresponding to May 2001 to April 2003 (June 2005 to May 2007).²⁹ Table 10 Panel A shows that the CA indexes have average monthly returns ranging from -0.12% to 0.21% in the out-of-sample period. The negative average return corresponds to HFRX while all other indexes have positive average returns. Median returns portray a similar picture: CISDM, CT, CTX, and HFR indexes all have higher medians than HFRX. In terms of dispersion, CISDM, CTX, and HFR exhibit lower volatility compared to CT and HFRX. Overall, there are discernable differences between indexes, which are primarily statistical constructs, and the investable portfolios. Consistent with the pattern observed in-sample, Table 10 Panel B shows that the CA indexes are highly correlated with each other while VG and X_o continue to exhibit strong positive correlations with the CA indexes.

6.2. *Out-of-sample analysis with the buy-and-hedge and buy-and-hold factors*

To verify that the investable factors X_o and VG continue to be significant determinants of CA hedge funds in the out-of-sample period, we regress CA index returns on X_o and VG :

$$CA_t = \theta_0 + \theta_1 X_{O,t} + \theta_2 VG_t + \psi_t \quad (15)$$

The results are reported in Table 11. The coefficients of X_o and VG are positive and highly significant for all CA indexes with adjusted- R^2 ranging from 41.08% to 44.93%.

²⁹ To illustrate that the out-of-sample version of the X factor is similar to the in-sample version, we repeat the regression in equation (3) with the out-of-sample version and report the results in Appendix B. We find that the results are similar to those in Table 4 with the X factor continuing to be significant in all cases and the explanatory power ranging from 15% to 37%, which is lower than the range of 24% to 53% obtained earlier with the in-sample version. This is to be expected as the out-of-sample version of the X factor is unlikely to be as precise as the in-sample version, which is created by delta-hedging the portfolio of individual stocks corresponding to the different CBs as opposed to delta-hedging the Russell 2000 index. Considering the lack of data on individual CBs during the out-of-sample period, we believe this is the best proxy we can construct.

During our sample period from January 1993 to April 2003, we use the non-investable convertible arbitrage indexes as their investable counterparts are not available. However, during the out-of-sample period from May 2003 to June 2007, we notice and recognize the introduction of investable indexes. Hence, as a sensitivity check, we repeat our analysis with the investable indexes. Both HFR indexes (HFR and HFRX) load positively and significantly on the buy-and-hold and buy-and-hedge factors, which is consistent with HFR's objective of constructing an investable index that has maximum correlation with its investable counterpart. In contrast, regression results for the CS Tremont indices do diverge in one respect. While the non-investable index, CT, has positive and significant loadings on the buy-and-hold and buy-and-hedge factors, its investable counterpart, CTX, only shows a positively significant exposure to the buy-and-hold factor. The coefficient on the buy-and-hedge factor is positive, but has a t-statistic of 1.67 (p-value of 0.102). One possible explanation for the slight loss of statistical power is the fact CTX has a somewhat shorter return history than CT during the out-of-sample period; CT has 50 monthly observations whereas CTX has 47 monthly observations.³⁰ It is interesting to note that when we augment the two-factor model with the supply variable (see Table 12), both CT and CTX have positive and statistically significant loadings on the buy-and-hold and buy-and-hedge factors.

Next we examine the effect of CB supply on performance.

6.3. *Out-of-sample analysis with the buy-and-hedge and buy-and-hold factors along with the supply variable*

The supply variable, *LDeal*, used in the sample period is constructed from individual CB data, which end in April 2003. To construct the same variables for the out-of-sample period, we

³⁰ Another possible explanation is that CT and CTX have different construction methodologies, resulting in different constituent funds in each index and divergent return properties. CT is a value-weighted index of CA hedge funds while CTX is a value-weighted index of the six largest eligible funds making up CT. However, Table 10 Panel B shows that both indices have a high correlation of 0.95, suggesting that index construction is a less likely explanation.

collect CB issuance data from Thomson Reuters Financial’s SDC Platinum database. To validate our choice and verify the consistency between the Albourne and SDC databases, we compare monthly CB issuance from the individual CB data from Albourne Partners with that from SDC during the sample period (January 1993 to April 2003). We find that both variables are closely matched in terms of sample moments and are highly correlated.³¹ This provides us comfort in using the supply variable constructed from the SDC data to estimate the following regression in the out-of-sample period:

$$CA_t = \alpha + \beta_1 X_t + \beta_2 VG_t + \gamma LDeal_t + \chi_t \quad (16)$$

The regression results in Table 12 indicate significant loadings on the buy-and-hedge factor across all non-investable CA indexes, corroborating our in-sample results. There is a notable increase in the explanatory power of our model. The adjusted-R² now ranges from 48.72% to 56.44% (compared to 41.08% to 44.93% in Table 11), which underscores the importance of accounting for supply variable in explaining the returns of CA strategy. Interestingly, the relevance of the buy-and-hold factor has weakened in the out-of-sample period. Supply effects continue to play an important role in explaining CA index returns during the out-of-sample period; the coefficient on *LDeal* is positive and significant for all CA indexes. Consistent with the earlier evidence of supply effects explaining the abnormal returns, we observe that all alphas are now significantly negative. In terms of the investable indexes, the supply variable figures significantly in the regression model and exhibits positive loadings. Including the supply variable enhances the explanatory power of the model. Exposures to the investable buy-and-hedge and buy-and-hold factors remain similar. There is, however, a notable increase in both the magnitude as well as statistical significance of the negative alpha term.

³¹ The summary statistics (mean, median, minimum, maximum, and the standard deviation) of the number of deals per month in the US are 13.3, 12.0, 1.0, 48.0, and 8.3 using Albourne data. The corresponding figures are 11.6, 11.0, 1.0, 41.0, and 6.6 for SDC. The correlation between the two series is 0.76.

Overall, these out-of-sample analyses using investable CA portfolios and factors confirm the earlier findings. Taken together, they show that our model can be used to capture the risk-return characteristics of investing in CA funds in a real-life setting.

7. Conclusion

In this paper, we provide a rare glimpse into the strategies used by convertible arbitrageurs. We posit that these arbitrageurs assume the role of an intermediary that provides capital to the CB issuers while transmitting the equity risk of CB ownership to the equity market through delta hedging. Using daily data from the underlying convertible bond and stock markets in the US, we compute the returns to a buy-and-hedge strategy, which involves taking a long position in the convertible bonds and hedging the equity risk by shorting the shares of CB issuers. We show that such a strategy together with a simple buy-and-hold strategy can explain a large proportion of the return variation in convertible arbitrage hedge funds. In addition, our results reveal the important role played by supply conditions in determining the returns of convertible arbitrage hedge funds. Furthermore, we show that convertible arbitrageurs are sensitive to extreme market events such as the LTCM crisis. In response to such events, we demonstrate that there can be an increase in the risk aversion of economic agents, which results in less reliance on the buy-and-hold strategy. Examining the returns of individual convertible arbitrage funds uncovers interesting cross-sectional variation in the use of buy-and-hedge strategy and the impact of CB supply. Larger funds tend to rely more on the buy-and-hedge strategy and exhibit less directional exposure to CBs, which is consistent with the economies of scale they enjoy in accessing the stock loan market. Large funds also are more vulnerable to supply shocks, which is suggestive of frictions involved in adjusting the stock of risk capital in response to the decline in

CB supply. Finally, out-of-sample analysis using investable CA portfolios and factors confirm that our key findings persist in more recent times as well as with real-life portfolios.

Taken together, these results suggest that convertible arbitrageurs collectively play the role of an intermediary who provides funding to convertible bond issuers whilst transferring the equity risk of CB ownership to the equity market through hedging. This conclusion appears to be robust to sample periods as well as the frictions of investing in CA funds.

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Table 1
Descriptive statistics of convertible bonds

Panel A: Breakdown of sample by the number of issues per issuer			
	Number of issues per issuer	Number of issuers	Total number of bonds
	1	954	954
	2	178	356
	3	55	165
	4	20	80
	5	15	75
	7	1	7
	9	1	9
		1,224	1,646
Panel B: Summary statistics of issue size			
	Average issue size (mil)	\$ 323	
	25th percentile (mil)	\$ 100	
	50th percentile (mil)	\$ 175	
	75th percentile (mil)	\$ 350	
Panel C: Number of issues each year			
	1993	91	
	1994	70	
	1995	78	
	1996	185	
	1997	211	
	1998	177	
	1999	148	
	2000	176	
	2001	265	
	2002	179	
	Until April 2003	66	

This table provides the summary statistics of our sample of 1,646 U.S. convertible bonds during our sample period from January 1993 to April 2003. Panel A provides the breakdown of our sample by the number of issues per issuer. Panel B provides the summary statistics of issue size (in \$ millions). Panel C reports the number of issues during each year by summing the issues across 12 months for 1993-2002 and 4 months for the last year, 2003.

Table 2
Descriptive statistics of the buy-and-hedge factor

Panel A: Bonds used to construct X factor	
Average number of bonds	411
Current yield (%)	13
Parity (%)	69
Age of bonds (in years)	2.4
Panel B: Summary statistics of the monthly returns of X factor	
Mean (%)	0.17
Median (%)	0.22
SD (%)	1.01
Min. (%)	-1.78
Max. (%)	1.95
Skew	-0.14
Kurt	-0.48

We report descriptive statistics of the buy-and-hedge factor, X. Panel A provides the daily average of the number of convertible bonds, current yield, parity, and age of convertible bonds in the portfolio generating the X factor. Current yield is computed using a full year's coupon divided by the bond's price (which incorporates accrued interest). Parity is the bond's conversion value as a percentage of the par value. Age is the number of years since the bond was issued. Panel B provides summary statistics of the buy-and-hedge factor's monthly returns during our sample period from January 1993 to April 2003.

Table 3

Descriptive statistics of the long-biased convertible bond portfolio and the convertible arbitrage portfolios

Panel A: Summary Statistics

	Mean (%)	Median (%)	SD (%)	Min. (%)	Max. (%)	Skew	Kurt
VG	0.32	0.22	3.72	-13.25	10.11	-0.43	1.33
EW	0.56	0.68	1.03	-4.00	2.55	-1.08	2.51
VW	0.50	0.74	1.28	-4.66	3.04	-1.24	2.59
CISDM	0.61	0.74	0.67	-2.35	2.20	-1.36	3.98
CT	0.47	0.79	1.38	-5.10	3.07	-1.69	4.55
HFR	0.57	0.71	0.98	-3.65	2.93	-1.39	3.99

Panel B: Correlations between VG, the two CA portfolios, the three CA indexes and the X factor

	VG	EW	VW	CISDM	CT	HFR	X
VG	1.00						
EW	0.60 ^{***}	1.00					
VW	0.49 ^{***}	0.90 ^{***}	1.00				
CISDM	0.55 ^{***}	0.92 ^{***}	0.83 ^{***}	1.00			
CT	0.34 ^{***}	0.81 ^{***}	0.85 ^{***}	0.80 ^{***}	1.00		
HFR	0.51 ^{***}	0.90 ^{***}	0.84 ^{***}	0.93 ^{***}	0.80 ^{***}	1.00	
X	0.39 ^{***}	0.62 ^{***}	0.58 ^{***}	0.61 ^{***}	0.48 ^{***}	0.60 ^{***}	1.00

Panel A of this table provides the descriptive statistics of the monthly returns of the Vanguard Convertible Securities mutual fund (VG), equally-weighted and value-weighted portfolios of all CA hedge funds (EW and VW) and three convertible arbitrage (CA) indexes (CISDM, CT, and HFR) from January 1993 to April 2003 with the exception of the CT index, which starts in 1994. The descriptive statistics include mean, median, standard deviation (SD), minimum (Min.), maximum (Max.), skewness (Skew), and kurtosis (Kurt). Panel B provides the correlations among the above return series and the buy-and-hedge factor, X. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 4

Regression analysis with the buy-and-hedge and buy-and-hold factors

	CA Portfolios				CA Indexes					
	EW Coeff	t-stat	VW Coeff	t-stat	CISDM Coeff	t-stat	CT Coeff	t-stat	HFR Coeff	t-stat
Intercept	0.004***	(4.756)	0.004***	(2.654)	0.005***	(8.505)	0.004**	(2.339)	0.005***	(5.135)
X	0.465***	(4.969)	0.581***	(3.952)	0.311***	(5.079)	0.558***	(3.600)	0.453***	(4.562)
VG	0.118***	(5.499)	0.108***	(3.866)	0.066***	(4.279)	0.065*	(1.902)	0.087***	(3.588)
Adj R ²	52.82%		41.50%		48.18%		24.20%		43.96%	

This table provides the results of the following OLS regression during our sample period from January 1993 to April 2003:

$$CA_t = \theta_0 + \theta_1 X_t + \theta_2 VG_t + \psi_t$$

where, CA_t are the excess returns (in excess of risk-free rate, i.e., US Fed Funds rate) during month t on the equally-weighted and value-weighted indexes of all the CA funds in our sample (EW and VW) or on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR), X_t is the return during month t on the buy-and-hedge factor and VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 5
Results with the alternative buy-and-hedge factor, X^A

	<u>CA Portfolios</u>				<u>CA Indexes</u>					
	EW Coeff	t-stat	VW Coeff	t-stat	CISDM Coeff	t-stat	CT Coeff	t-stat	HFR Coeff	t-stat
Intercept	0.003***	(2.884)	0.002	(1.327)	0.005***	(5.783)	0.003	(1.214)	0.004***	(2.961)
X^A	0.345***	(3.886)	0.447***	(3.271)	0.226***	(3.915)	0.405***	(2.693)	0.361***	(4.036)
VG	0.115***	(4.281)	0.102***	(3.015)	0.065***	(3.385)	0.059	(1.423)	0.081***	(2.858)
Adj R ²	43.68%		32.75%		38.05%		16.51%		35.73%	

This table provides the results of the following OLS regression during our sample period from January 1993 to April 2003:

$$CA_t = \eta_0 + \eta_1 X_t^A + \eta_2 VG_t + \pi_t$$

where, CA_t are the excess returns (in excess of risk-free rate, i.e., US Fed Funds rate) during month t on the equally-weighted and value-weighted indexes of all the CA funds in our sample (EW and VW) or on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR), X_t^A is the return during month t on the alternative-X factor where all the three risks are hedged, i.e., long convertible bonds and short the equity, credit, and interest rate risks and VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund. We proxy the credit risk in CBs by the daily change in the spread between the yield on Baa corporate bonds and the yield on 10-year U.S. Treasury bonds. We capture the interest rate risk in CBs through the daily yield on 5-year U.S. Treasury bonds. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 6
Information content of supply

Announcement month is	No. of convertible bond issues	Percent (%)
Same as issuance month	63	67.74
One month before issuance month	22	23.66
Two months before issuance month	6	6.45
Four months before issuance month	1	1.08
Five months before issuance month	1	1.08
Total	93	100

We randomly sample 100 US convertible bonds and hand collect the earliest announcement date from Factiva news search supplemented with searches on SEC EDGAR and Google. Announcement dates are successfully collected for 93 of the 100 bonds. For these bonds, we compare the announcement month and the issuance month to determine the extent to which the announcement date precedes reported issuance date. We report the frequency distribution of the announcement month relative to the issuance month.

Table 7
Regression analysis allowing for supply

	<u>CA Portfolios</u>				<u>CA Indexes</u>					
	EW Coeff	t-stat	VW Coeff	t-stat	CISDM Coeff	t-stat	CT Coeff	t-stat	HFR Coeff	t-stat
Intercept	-0.010**	(2.169)	-0.022***	(4.078)	-0.005	(1.432)	-0.026***	(3.197)	-0.011**	(2.236)
X	0.442***	(5.279)	0.535***	(4.396)	0.294***	(5.622)	0.453***	(3.696)	0.427***	(4.800)
VG	0.099***	(5.994)	0.074***	(3.444)	0.053***	(4.179)	0.030	(1.015)	0.068***	(3.248)
LDeal _t	0.005***	(3.415)	0.008***	(5.033)	0.003***	(3.360)	0.010***	(3.910)	0.005***	(3.516)
Adj R ²	59.70%		55.26%		55.44%		39.19%		51.90%	

This table provides the results of the following OLS regression during our sample period from January 1993 to April 2003:

$$CA_t = \alpha + \beta_1 X_t + \beta_2 VG_t + \gamma LDeal_t + \chi_t$$

where CA_t are the excess returns (in excess of risk-free rate - US Fed Funds rate) during month t on the equally-weighted and value-weighted indexes of all the CA funds in our sample (EW and VW) or on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR), X_t is the return during month t on the buy-and-hedge factor and VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund. $LDeal_t$ is the natural log of one plus the number of convertible issues in months $t+1$ and $t+2$. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 8

Structural break model after allowing for supply of convertible bonds

		CA Portfolios				CA Indexes					
		EW		VW		CISDM		CT		HFR	
		Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat
PRE-LTCM	PERIOD										
	D1	-0.011**	(2.085)	-0.027***	(4.643)	-0.005	(1.359)	-0.032***	(4.321)	-0.012**	(2.344)
	D1 x X	0.349***	(3.557)	0.420***	(2.911)	0.235***	(3.841)	0.282**	(2.347)	0.293***	(3.083)
	D1 x VG	0.154***	(5.152)	0.129***	(3.563)	0.085***	(3.913)	0.067	(1.336)	0.128***	(4.473)
	D1 x LDeal _t	0.005***	(3.117)	0.010***	(5.418)	0.003***	(3.062)	0.011***	(5.058)	0.005***	(3.570)
POST-LTCM	PERIOD										
	D2	-0.003	(0.644)	-0.007	(0.811)	-0.001	(0.158)	-0.010	(0.568)	0.0001	(0.010)
	D2 x X	0.552***	(6.228)	0.683***	(5.231)	0.376***	(6.773)	0.694***	(4.269)	0.643***	(6.029)
	D2 x VG	0.075***	(4.160)	0.050*	(1.873)	0.039***	(2.936)	0.017	(0.429)	0.041*	(1.809)
	D2 x LDeal _t	0.002*	(1.814)	0.004	(1.496)	0.002	(1.564)	0.005	(0.954)	0.002	(0.782)
	Adj R ²	69.29%		62.40%		75.49%		45.52%		65.70%	
	<i>Test of differences of coefficients</i>										
		F-value	p-value	F-value	p-value	F-value	p-value	F-value	p-value	F-value	p-value
	D1 = D2 = 0	3.44**	0.035	11.07***	0.000	1.37	0.259	12.10***	0.000	3.49**	0.034
	D1 x X = D2 x X	2.68	0.104	2.24	0.138	2.84*	0.095	3.70*	0.057	6.34**	0.013
	D1 x VG = D2 x VG	3.72*	0.056	2.61	0.109	2.96*	0.088	0.37	0.544	5.09**	0.026
	D1 x LDeal _t = D2 x LDeal _t	1.39	0.240	3.97**	0.049	0.62	0.434	1.43	0.234	1.87	0.175

This table provides the results of the following OLS regression during our sample period from January 1993 to April 2003:

$$CA_t = D1(\omega_0 + \omega_1 X_t^{US} + \omega_2 VG_t + \gamma LDeal_t) + D2(\omega_0 + \omega_3 X_t^{US} + \omega_4 VG_t + \pi LDeal_t) + \kappa_t$$

where, CA_t are the excess returns during month t on the equally-weighted and value-weighted indexes of all the CA funds in our sample (EW and VW) or on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR), X_t are the return during month t on the buy-and-hedge factor and VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund. $LDeal_t$ is natural log of one plus the number of convertible issues in month $t+1$ and $t+2$. The pre-LTCM (post-LTCM) period dummy, D1 (D2), takes the value of 1 (0) before (after) the LTCM crisis in September 1998 and equals 0 (1) otherwise. There is no intercept term as the regression includes both the dummies. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. F-value and p-value are, respectively, the F-statistic and associated p-value from the test of difference of coefficients. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 9
Individual fund level analyses

Panel A: Two-factor model with supply

Size Tercile:	Small		Medium		Large	
	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat
Intercept	-0.002	(-0.552)	-0.015***	(-4.281)	-0.009***	(-2.863)
X	0.394***	(5.909)	0.419***	(6.624)	0.451***	(7.806)
VG	0.161***	(3.686)	0.127***	(3.185)	0.056	(1.688)
LDeal	0.002	(1.343)	0.006***	(5.914)	0.004***	(4.735)
Adj R ²	23.60%		25.41%		18.01%	
AUM	8.390		51.088		243.696	
# Funds	36		36		36	
Mean (%)	0.39		0.65		0.51	
Median (%)	0.63		0.78		0.66	
SD (%)	1.29		1.05		1.25	

Panel B: Two-factor model with supply and structural break for the LTCM crisis							
Size Tercile:		Small		Medium		Large	
		Coeff	t-stat	Coeff	t-stat	Coeff	t-stat
PRE-LTCM	D1	-0.013 ^{***}	(-3.360)	-0.018 ^{***}	(-3.531)	-0.028 ^{**}	(-2.589)
	D1 x X	0.154 [*]	(1.955)	0.407 ^{***}	(4.547)	0.413 ^{**}	(2.500)
	D1 x VG	0.141 ^{**}	(2.678)	0.025	(0.715)	0.079	(1.141)
	D1 x LDeal _t	0.005 ^{***}	(5.107)	0.008 ^{***}	(5.387)	0.010 ^{***}	(3.080)
POST-LTCM	D2	0.017 [*]	(1.985)	-0.012 ^{**}	(-2.596)	-0.009	(-1.638)
	D2 x X	0.697	(1.753)	0.677 ^{***}	(4.347)	0.411 ^{***}	(3.869)
	D2 x VG	0.098 ^{***}	(3.419)	0.035	(0.953)	0.084	(1.039)
	D2 x LDeal _t	-0.004	(-1.509)	0.005 ^{***}	(4.018)	0.004	(1.618)
	Adj R ²	31.84%		43.99%		39.29%	
	AUM	13.214		63.616		225.036	
	# Funds	15		15		15	
	Mean (%)	0.47		0.63		0.48	
	Median (%)	0.54		0.67		0.77	
	SD (%)	1.25		0.91		2.12	
<i>Test of differences of coefficients</i>		Increase	Decrease	Increase	Decrease	Increase	Decrease
D1 = D2 = 0		3	0	3	0	4	1
D1 x X = D2 x X		3	0	7	1	2	1
D1 x VG = D2 x VG		1	4	2	4	3	1
D1 x LDeal _t = D2 x LDeal _t		0	2	0	5	1	4

Panel A of this table summarizes the fund level results of the following OLS regression during our sample period from January 1993 to April 2003: $CA_t = \theta_0 + \theta_1 X_t + \theta_2 VG_t + \theta_3 LDeal_t + \psi_t$

where, CA_t is the excess return (in excess of risk-free rate, i.e., US Fed Funds rate) during month t on a convertible arbitrage hedge fund with at least 24 monthly returns during the sample period, X_t is the return during month t on the buy-and-hedge factor, VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund (VG) and $LDeal_t$ is natural log of one plus the number of convertible issues in months $t+1$ and $t+2$. We sort funds into size terciles based on their average assets under management during the sample period and summarize the regression outputs of funds in each tercile. Coeff and t-stat are, respectively, the mean and t-statistic of individual fund regression coefficients. Adj R² is the average of funds' adjusted R². AUM reports the grand average assets under management (in \$ millions) in each tercile and # Funds reports the number of funds in each tercile. Mean, Median, and SD are, respectively, the mean, median and standard deviation of the monthly excess return on the equally-weighted size tercile. Panel B of this table provides the results of the following OLS regression during our sample period from January 1993 to April 2003:

$$CA_t = D1(\omega_0 + \omega_1 X_t + \omega_2 VG_t + \gamma LDeal_t) + D2(\omega_0 + \omega_3 X_t + \omega_4 VG_t + \pi LDeal_t) + \kappa_t$$

where, CA_t is the excess return (in excess of risk-free rate - US Fed Funds rate) during month t on a convertible arbitrage hedge fund with at least 48 monthly returns during the sample period and which existed before, during and after the LTCM crisis. The pre-LTCM (post-LTCM) period dummy, D1 (D2), takes the value of 1 (0) before (after) the LTCM crisis in September 1998 and equals 0 (1)

otherwise. There is no intercept term as the regression includes both the dummies. As in Panel A, we summarize the regression outputs of funds in each tercile. Increase (Decrease) is the number of funds for which there is a statistically significant increase (decrease) in coefficient after the LTCM crisis. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 10

Descriptive Statistics of Convertible Arbitrage Indexes and Factors during the out-of-sample period

Panel A: Summary Statistics							
	Mean (%)	Median (%)	SD (%)	Min. (%)	Max. (%)	Skew	Kurt
VG	0.84	0.95	2.03	-4.25	5.03	-0.17	0.16
CISDM	0.17	0.28	0.83	-2.68	1.98	-0.83	1.77
CT	0.21	0.35	1.03	-3.36	2.39	-0.91	2.08
CTX	0.20	0.24	0.93	-2.59	2.11	-0.37	0.87
HFR	0.13	0.29	0.89	-2.87	2.04	-0.93	1.57
HFRX	-0.12	0.14	1.14	-3.68	2.67	-0.78	1.49

Panel B: Correlations between VG, CA indexes and the X factor							
	VG	CISDM	CT	CTX	HFR	HFRX	
VG	1.00						
CISDM	0.57 ^{***}	1.00					
CT	0.60 ^{***}	0.95 ^{***}	1.00				
CTX	0.64 ^{***}	0.91 ^{***}	0.95 ^{***}	1.00			
HFR	0.55 ^{***}	0.97 ^{***}	0.97 ^{***}	0.91 ^{***}	1.00		
HFRX	0.49 ^{***}	0.87 ^{***}	0.89 ^{***}	0.87 ^{***}	0.89 ^{***}	1.00	
X_o	0.51 ^{***}	0.57 ^{***}	0.60 ^{***}	0.57 ^{***}	0.62 ^{***}	0.66 ^{***}	

Panel A of this table provides the descriptive statistics of the monthly returns of the Vanguard Convertible Securities mutual fund (VG), three convertible arbitrage (CA) indexes (CISDM, CT, HFR) and two convertible arbitrage investable indexes (CTX and HFRX) during the out-of-sample period (May 2003 to June 2007). VG, CISDM, CT, HFR and HFRX are available for the entire out-of-sample period while CTX starts in August 2003. The descriptive statistics include mean, median, standard deviation (SD), minimum (Min.), maximum (Max.), skewness (Skew), and kurtosis (Kurt). Panel B provides correlations between VG, CISDM, CT, CTX, HFR, HFRX and the buy-and-hedge factor, X_o . *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 11
Out-of-sample analysis with the buy-and-hedge and the buy-and-hold factors

	CISDM		CT		CTX		HFR		HFRX	
	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat
Intercept	-0.001	(0.536)	-0.001	(0.610)	-0.001	(0.571)	-0.001	(0.902)	-0.005**	(2.250)
X_o	0.307**	(2.681)	0.401***	(2.889)	0.280	(1.669)	0.400***	(3.325)	0.627***	(3.892)
VG	0.156**	(2.527)	0.201***	(3.164)	0.214***	(3.693)	0.141**	(2.232)	0.116*	(1.849)
Adj R ²	41.08%		44.93%		44.82%		43.57%		44.92%	

This table provides results of the following OLS regression during the out-of-sample period from May 2003 to June 2007:

$$CA_t = \theta_0 + \theta_1 X_{o,t} + \theta_2 VG_t + \psi_t$$

where CA_t are the excess returns (in excess of risk-free rate, i.e., US Fed Funds rate) during month t on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR) or convertible arbitrage investable indexes (CTX, or HFRX), $X_{o,t}$ is the return during month t on the buy-and-hedge (i.e., X) factor, and VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Table 12

Out-of-sample analysis with the buy-and-hedge and the buy-and-hold factors after allowing for supply

	CISDM		CT		CTX		HFR		HFRX	
	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat
Intercept	-0.024**	(2.293)	-0.034**	(2.409)	-0.036***	(3.422)	-0.027**	(2.247)	-0.030**	(2.047)
$X_{o,t}$	0.322***	(3.353)	0.410***	(3.594)	0.257*	(1.827)	0.411***	(4.175)	0.641***	(4.265)
VG	0.110*	(1.755)	0.139*	(1.922)	0.172***	(2.709)	0.088	(1.366)	0.067	(0.869)
$LDeal_t$	0.007**	(2.277)	0.009**	(2.412)	0.010***	(3.406)	0.007**	(2.197)	0.007*	(1.783)
Adj R ²	48.72%		54.03%		56.44%		50.93%		49.17%	

This table provides results of the following OLS regression during the out-of-sample period from May 2003 to June 2007:

$$CA_t = \alpha + \beta_1 X_{o,t} + \beta_2 VG_t + \gamma LDeal_t + \chi_t$$

where CA_t are the excess returns (in excess of risk-free rate, i.e., US Fed Funds rate) during month t on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR) or convertible arbitrage investable indexes (CTX, or HFRX), $X_{o,t}$ is the return during month t on the X factor, VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund (VG), $LDeal_t$ is natural log of one plus the number of convertible issues in month $t+1$ and $t+2$. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Appendix A

Regression analysis using other multifactor models

Panel A

	CA Portfolios				CA Indexes					
	EW Coeff	t-stat	VW Coeff	t-stat	CISDM Coeff	t-stat	CT Coeff	t-stat	HFR Coeff	t-stat
Intercept	0.005***	(3.43)	0.004**	(2.32)	0.006***	(6.00)	0.004*	(1.76)	0.005***	(3.84)
MKTRF	-0.010	(-0.22)	-0.021	(-0.31)	0.002	(0.08)	-0.026	(-0.34)	-0.017	(-0.42)
SMB	0.108***	(4.07)	0.103***	(3.01)	0.068***	(3.76)	0.118**	(2.56)	0.089***	(3.64)
HML	0.053*	(1.72)	0.048	(1.29)	0.043**	(2.23)	0.104**	(2.05)	0.050*	(1.70)
MOM	0.008	(0.56)	0.001	(0.03)	-0.003	(-0.31)	-0.003	(-0.17)	-0.0003	(-0.02)
OTM CALL	0.248	(1.64)	0.004*	(1.89)	0.133	(1.22)	0.161	(0.75)	0.209	(1.43)
OTM PUT	-0.471**	(-1.99)	-0.004	(-1.14)	-0.251	(-1.54)	-0.432	(-0.95)	-0.405*	(-1.77)
Adj R ²	29.54%		15.97%		23.22%		6.91%		18.63%	

Panel B

	CA Portfolios				CA Indexes					
	EW Coeff	t-stat	VW Coeff	t-stat	CISDM Coeff	t-stat	CT Coeff	t-stat	HFR Coeff	t-stat
Intercept	0.005***	(4.24)	0.004***	(2.83)	0.006***	(7.44)	0.005**	(2.28)	0.005***	(4.84)
SPMRF	0.081***	(3.89)	0.077**	(2.55)	0.044***	(3.58)	0.019	(0.54)	0.061***	(3.32)
SML	0.089***	(4.00)	0.076**	(2.48)	0.050***	(3.14)	0.071**	(2.03)	0.071***	(2.96)
TSYMRF	0.086*	(1.97)	0.091	(1.38)	0.069*	(1.96)	0.079	(1.27)	0.090*	(1.95)
HYMTSY	0.137	(1.05)	0.167	(0.92)	0.120	(1.59)	0.218	(1.03)	0.115	(1.00)
PTFSBD	-0.016**	(-2.23)	-0.017*	(-1.77)	-0.010**	(-2.11)	-0.017	(-1.53)	-0.016**	(-2.36)
PTFSFX	0.001	(0.34)	-0.001	(-0.12)	0.001	(0.55)	-0.001	(-0.08)	0.002	(0.40)
PTFSCOM	0.006	(0.94)	0.005	(0.71)	0.003	(0.68)	0.007	(0.72)	0.005	(0.98)
Adj R ²	28.29%		16.89%		25.78%		7.42%		21.33%	

This table provides the results of OLS regressions for the five CA portfolios – the equally-weighted and value-weighted portfolios of individual funds (EW and VW) and three CA indexes (CISDM, CT, and HFR). Panel A reports the results from the Carhart (1997) 4-factor model augmented with the out-of-the-money call (OTM CALL) and put (OTM PUT) option factors as in Agarwal and Naik (2004). The 4-factor model of Carhart (1997) includes MKTRF, SMB, HML, and MOM that refer to excess returns on the market (value-weighted return on all NYSE, AMEX, and NASDAQ stocks minus the one-month Treasury bill rate), Fama-French (1993) size and book-to-market factors, and momentum factor. Panel B reports the results from the Fung and Hsieh (2004) 7-factor model including excess return on S&P 500 (SPMRF), spread between Wilshire Small Cap 1750 index and Wilshire Large Cap 750 index (SML), excess returns on 10-year Treasury (TSYMRF), credit spread, i.e., difference between CSFB High-Yield index returns and 10-year Treasury returns (HYMTSY), lookback straddles on bond futures (PTFSBD), lookback straddles on currency futures (PTFSFX), and lookback straddles on commodity futures (PTFSCOM). Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

Appendix B

In-sample results with the out-of-sample buy-and-hedge factor, X_o

	<u>CA Portfolios</u>				<u>CA Indexes</u>					
	EW		VW		CISDM		CT		HFR	
	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat	Coeff	t-stat
Intercept	0.005 ^{***}	(4.471)	0.004 ^{***}	(2.907)	0.006 ^{***}	(7.467)	0.004 ^{**}	(2.106)	0.005 ^{***}	(4.634)
X_o	0.090 [*]	(1.811)	0.151 ^{**}	(2.432)	0.062 [*]	(1.780)	0.183 ^{**}	(2.459)	0.104 ^{**}	(1.994)
VG	0.150 ^{***}	(6.040)	0.142 ^{***}	(4.463)	0.088 ^{***}	(4.601)	0.093 ^{**}	(2.301)	0.116 ^{***}	(4.032)
Adj R ²	36.79%		26.64%		31.20%		15.05%		27.82%	

This table provides the results of the following OLS regression during our sample period from January 1993 to April 2003:

$$CA_t = \eta_0 + \eta_1 X_{ot} + \eta_2 VG_t + \pi_t$$

where, CA_t are the excess returns (in excess of risk-free rate, i.e., US Fed Funds rate) during month t on the equally-weighted index of all the CA funds in our sample (EW) or on the convertible arbitrage (CA) hedge fund indexes (CISDM, CT, or HFR), X_{ot} is the return during month t on the out-of-sample X factor constructed using the Russell 2000 index instead of the individual stock data as described in Section 6.1. VG_t is the excess return during month t on the Vanguard Convertible Securities mutual fund. Coeff are regression estimates and t-stat are t-statistics computed with Newey and West (1987) standard errors. *, **, and *** indicate that the coefficient is significantly different from zero at the 10, 5, and 1% levels respectively.

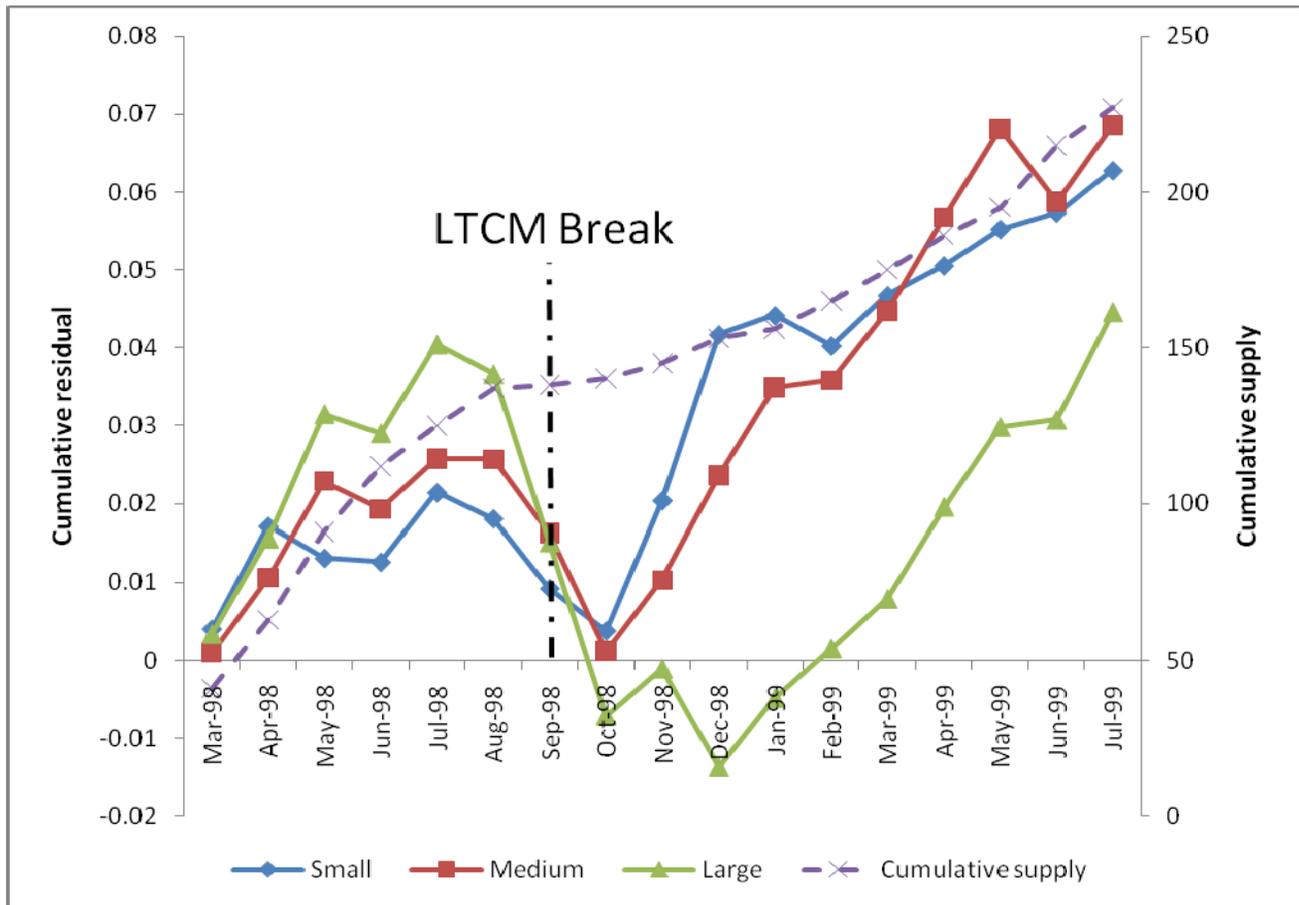


Fig. 1. Cumulative residuals. This figure presents cumulative residual plots of the equally-weighted size terciles (Small, Medium, and Large) and the plot of cumulative convertible bond (CB) supply. The plots cover the period surrounding the LTCM break (March 1998 – July 1999). For each tercile, we first compute monthly residuals for each fund using the fund’s excess returns, its factor loadings from estimating the structural break model without supply effect, and factor realizations. Once we have the fund level residuals, we average the residuals across funds each month to obtain the equally-weighted residual for each tercile. We then cumulate the equally-weighted residuals to obtain cumulative residuals for each tercile.

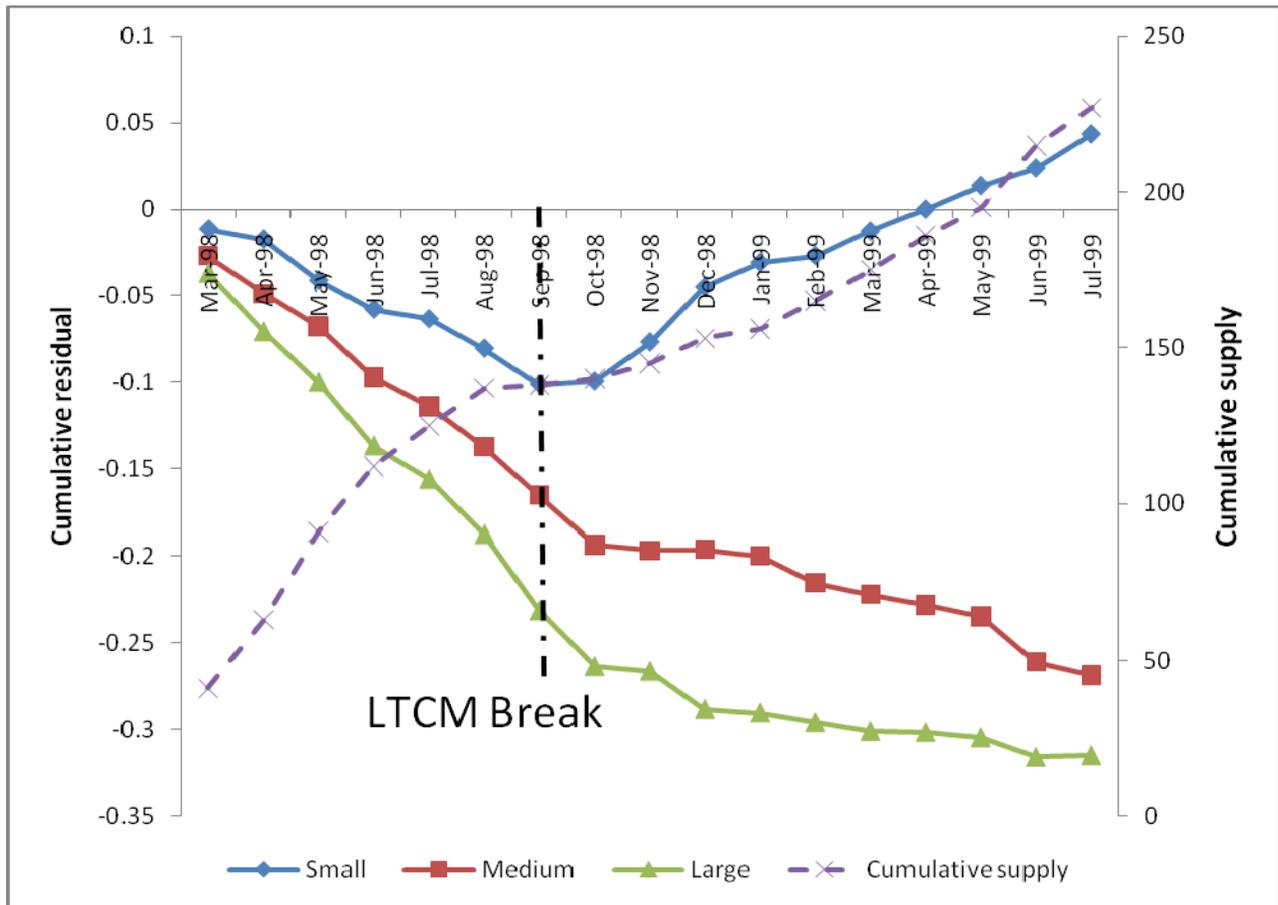


Fig. 2. Cumulative residuals net of supply effect. This figure presents net of supply effect cumulative residual plots of the equally-weighted size terciles (Small, Medium, and Large) and the plot of cumulative convertible bond (CB) supply. The plots cover the period surrounding the LTCM break (March 1998 – July 1999). For each tercile, we first compute monthly residuals for each fund using the fund’s excess returns, its factor loadings from estimating the structural break model with supply effect, factor realizations, and realizations of supply variable (*LDeal*). Once we have the fund level residuals, we average the residuals across funds each month to obtain the equally-weighted residual for each tercile. We then cumulate the equally-weighted residuals to obtain cumulative residuals for each tercile.

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