

PERFORMANCE ANALYSIS OF A COLLATERALIZED FUND OBLIGATION (CFO) EQUITY TRANCHE*

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Abstract

This article examines the performance of the junior tranche of a Collateralized Fund Obligation (CFO), i.e. the residual claim (equity) on a securitized portfolio of hedge funds. We use a polynomial goal programming model to create optimal portfolios of hedge funds, conditional to investor preferences and diversification constraints (maximum allocation per strategy). For each portfolio we build CFO structures that have different levels of leverage, and analyze both the stand alone performance as well as potential diversification benefits (low systematic risk exposures) of investing in the Equity Tranche of these structures. We find that the unconstrained mean-variance portfolio yields a high performance, but greater exposure to systematic risk. We observe the exact opposite picture in the case of unconstrained optimization where a skewness bias is added, thus proving the existence of a trade-off between stand alone performance and low exposure to systematic risk factors. We provide evidence that leveraged exposure to these hedge fund portfolios through the structuring of CFOs creates value for the Equity Tranche investor.

Keywords: Collateralized Fund Obligation (CFO), hedge funds, structured finance, portfolio optimization, performance analysis, multivariate linear regression, systematic risk.

JEL Classification: G11, G23.

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1. INTRODUCTION

In 2002, structured finance and hedge funds merged together for the first time. This union gave birth to *Collateralized Fund Obligations* (CFO) which consists in the securitization of hedge fund portfolios. This new category of asset seems to offer the best of both worlds. On the one hand it provides investors with a new vehicle through which to obtain exposure to hedge fund returns, and on the other hand offers to the financial engineers a new and quite unique pool of assets that they could securitize. The variety of strategies and low correlations with traditional assets make hedge funds an ideal candidate as collateral for securitization transactions. Although the theoretical justification of CFOs has been firmly established in financial literature (See Cheng (2002), Mahadevan and Schwartz (2002), Stone and Zissu (2004) and Missinhoun and Chacowry (2005)), there are still many misgivings as to their practical pertinence.

Thus, despite the rapid growth of the CDO market over the past two decades, there were no more than 20 CFO transactions prior to 2008. A lack of interest by investors in the Equity Tranche of these structures seems to be the source of the slow proliferation. The perception of low added-value and high inherent leverage, made it difficult to solicitate interest in the junior tranche of CFOs. That said, given the novelty and the complexity of these products, which belong to a vast family of derivatives, it is logical to suppose that the distinctive fundamental characteristics and benefits of the exposure to a CFO Equity Tranche have yet to be carefully scrutinized. This article provides a thorough analysis of the factors that might influence the performance of various hypothetical CFO Equity Tranches. Using data on historical hedge fund returns, the goal is to structure various CFOs based on a variety of underlying portfolios and investigate the returns of the Equity Tranche both in terms of stand alone performance and in terms of potential diversification benefits.

More specifically, the first objective, from the viewpoint of a CFO equity owner, is to define the optimal capital structure(s) as well as the general attributes for the diversification of the optimal portfolio of hedge funds for the securitization transaction.

The analysis is therefore far more thorough than that observed in existing literature on the subject of CFOs. In order to investigate the impact of strategy selection and diversification in the underlying hedge fund portfolio, we implement an optimization model that allows us to specify preferences for higher moments. This polynomial goal programming approach generates optimal allocations conditional to specific investor preferences. For each “optimal” portfolio, several debt structures are then considered so as to account for a far broader range of scenarios. This is done in order to identify, on the basis of a number of performance indicators, the optimal composition of the collateral and the appropriate leverage to which the exposure should be subjected.

The second objective is to study the exposure of the CFO Equity Tranche to systematic risk factors, such as market, credit and liquidity. In this sense, the study will determine the degree to which returns are defined by the returns of readily available risk premia, and therefore provide a better idea of their risk exposures. This is achieved using a multivariate linear regression model.

The results indicate that CFOs create value from the equity holder’s perspective. Nonetheless, there is a trade-off that must be made between the stand alone performance of a CFO Equity Tranche and its’ systematic risk exposure. We find that the unconstrained mean-variance portfolio yields a high performance but exhibits greater exposure to systematic risk factors. We observe the exact opposite in the case of unconstrained optimization where a preference for skewness is incorporated, thus proving the existence of a trade-off between performance and low-correlation with the financial markets. According to our results, an interesting compromise could be obtained by securitizing a well-diversified (constrained) underlying portfolio of funds.

The article is structured as follows. Section 2 presents the three stage methodology. Section 3 describes and analyzes the data and Section 4 focuses on the empirical results. Conclusions are presented in Section 5.

2. METHODOLOGY

The methodology is comprised of three steps. The first step concerns the allocation of hedge funds across the different strategies, the second relates to the structuring and evaluation of the CFO, and the final stage consists in the analysis of the systematic risk exposures of the resulting Equity Tranche of the CFO.

2.1 Allocation across hedge fund strategies

To decide on the asset allocation between the different investment strategies we use a polynomial goal programming (PGP) optimization model. This approach was introduced by Tayi and Leonard (1988), and has been employed by Chunnachinda, Dandapani, Hamid, and Prakash (1997) and Sun and Yan (2003) to incorporate the effect of skewness on portfolio allocation decisions. Davies, Kat, and Lu (2009) use this approach to incorporate investor preferences for higher moments into the construction of funds of funds. They extend the original model in order to account not only for skewness but also for the kurtosis that is prevalent in hedge fund return distributions. This approach incorporates multiple, and often conflicting, objectives and considers the impact of a change in investor preferences on asset allocation.

2.1.1 The PGP model

Consider an environment with m risky assets, each with random return \tilde{R}_i , and x_i being the percentage of wealth invested in the i^{th} asset. The risk free rate r is constant and no short selling of the risky assets is permitted. The percentage invested in the risk-free asset is determined by $x_{m+1} = 1 - \mathbf{1}^T \mathbf{X}$, where $\mathbf{1}$ is an identity vector of dimension $m \times 1$ and \mathbf{X} is the vector of dimension $m \times 1$ of percentages of wealth invested in the risky assets. V is the variance-covariance matrix for $\tilde{\mathbf{R}} = (\tilde{R}_1, \tilde{R}_2, \dots, \tilde{R}_m)$. This matrix is positive and of dimension $m \times m$. Thus, the problem of portfolio selection can be defined using the PGP model:

$$\text{MIN} \quad Z = (1 + d_1)^\alpha + (1 + d_3)^\beta + (1 + d_4)^\gamma \quad (1)$$

$$\text{Subject to} \quad E[\mathbf{X}^T \tilde{\mathbf{R}}] + x_{m+1}r + d_1 = Z_1^*, \quad (2)$$

$$E[\mathbf{X}^T (\tilde{\mathbf{R}} - E[\tilde{\mathbf{R}}])]^3 + d_3 = Z_3^*, \quad (3)$$

$$-E[\mathbf{X}^T (\tilde{\mathbf{R}} - E[\tilde{\mathbf{R}}])]^4 + d_4 = -Z_4^*, \quad (4)$$

$$d_1, d_3, d_4 \geq 0, \quad (5)$$

$$\mathbf{X}^T \mathbf{V} \mathbf{X} = 1, \quad x_i \geq 0, \quad x_{m+1} = 1 - \mathbf{1}^T \mathbf{X}, \quad (6)$$

where $Z_1^* = \text{Max} \{Z_1 \mid \mathbf{X}^T \mathbf{V} \mathbf{X} = 1\}$ is the average return of the optimal mean-variance portfolio with a unit variance, $Z_3^* = \text{Max} \{Z_3 \mid \mathbf{X}^T \mathbf{V} \mathbf{X} = 1\}$ is the skewness of the optimal skewness-variance portfolio with a unit variance, and $Z_4^* = \text{Max} \{Z_4 \mid \mathbf{X}^T \mathbf{V} \mathbf{X} = 1\}$ is the kurtosis of the optimal kurtosis-variance portfolio with a unit variance; and where α , β and γ are non-negative parameters specific to the investor's subjective degree of preference with respective regard to the average, the skewness and the kurtosis of the portfolio's returns. The form of the objective function ensures its monotone growth in d_1 , d_3 et d_4 for all possible values.

Even though the technique does not require an investor-specific utility function, it can still be inferred that the investors' utility functions are of higher order than quadratic. Importantly, the model's parameters α , β and γ have an explicit economic interpretation; they are directly linked to the concept of the marginal rate of substitution, which measures the desirability of foregoing an objective for the purpose of achieving another (trade-off of objectives). In short, the problem of a multi-objective PGP is solved in two stages. Firstly, the optimal values for Z_1^* , Z_3^* and Z_4^* are each obtained within a two-dimensional unit variance framework. Subsequently, these values are substituted in conditions (2) through (4) and the minimum value of (1) is found for a given set of investor preferences $\{\alpha, \beta, \gamma\}$ within a four-moment framework.

All resulting portfolios are composed of risky assets (hedge fund strategies) and a risk-free asset so as to ensure the unicity of each optimal portfolio. In order to ensure that the portfolio is solely invested in hedge funds, one must simply redistribute the allocation in the risk-free rate to the hedge fund strategies such that the total investment in the

risky assets equals 1. Thus $y_i = x_i / (x_1 + x_2 + \dots + x_m)$ will be the percentage invested in the i^{th} asset (hedge fund strategy i) for the optimal portfolio \mathbf{Y} .

Five sets of preference parameters $\{\alpha, \beta, \gamma\}$ are used for the optimization procedure. These are: $E_1 = \{\alpha = 1, \beta = 0, \gamma = 0\}$, $E_2 = \{\alpha = 1, \beta = 1, \gamma = 0\}$, $E_3 = \{\alpha = 1, \beta = 1, \gamma = 0.75\}$, $E_4 = \{\alpha = 2, \beta = 1, \gamma = 0.75\}$ and $E_5 = \{\alpha = 3, \beta = 1, \gamma = 0.25\}$. The goal is not to build portfolios with fully-representative parameters, but to determine the relative importance that must be attributed to each of the higher-order moments of the distribution. Simultaneously, one of the main objectives is to characterize an optimal underlying portfolio from the viewpoint of a CFO equity owner and it is therefore reasonable to suppose that the five sets of preferences should easily suffice. Note that the preference set E_1 represents a traditional mean-variance optimization. We also perform constrained optimizations for each set of preferences, with maximum allocation to each strategy limited to 10%, 15%, 20%, 25% and 50%. Finally, given the recent empirical evidence supporting equally-weighted portfolio (for example DeMiguel, Garlappi and Uppal (2005)) we also consider the $(1/m)$ portfolio allocation. In all 31 portfolios will be analyzed. The notation that will be employed to refer to a given preference set and diversification constraint is E_{hk} , where h refers to the preference set and k to the maximum allocation per strategy.

2.2 CFO Structures

In constructing the CFO structures we follow the approach of Mahadevan and Schwartz (2002). Overall, 20 structures are analysed, with each CFO $_j$ ($j = 1, 2, \dots, n$) being distinguished by its debt-to-equity ratio. CFO $_1$ represents the zero-debt case, which is simply a direct, unleveraged exposure to the hedge fund portfolio¹. The amount of debt within the structure is then increased in increments of 5% for each of the ensuing CFO structures, up to CFO $_{20}$ in which the Equity Tranche represents only 5% of the structure. Table 1 presents the 20 structures.

¹ Costs related to the structure's management and debt obviously do not apply in the case of this instrument.

<Table 1>

Each CFO has a par value of \$400M and a maturity of 7 years². As the proportion of debt increases, we notice that the number of tranches of rated debt also increases. To this end, a number of assumptions are made so as to comply with a same economic logic across all structures. We therefore assume that the first 200 million in debt (50% of the structure) are senior rated (AA), which explains why no junior debt tranche appears prior to CFO₁₂. This new tranche (BBB) is capped at 20% of the structure (\$80M), so that when more than \$280M in debt accrues, the BB tranche of debt is added. This new tranche is capped at 15%. The final tranche (BB-) only appears in the two riskiest structures, i.e. which use the most leverage.

It is important to note that the model does not include a coupon rate for each of the tranches. Instead, the procedure uses an “aggregated” cost of financing expressed in basis points in relation to a benchmark index that reflects the total, periodic remuneration of creditors, i.e. the owners of the CFO's rated debt tranches. Therefore, the greater the number of tranches, the greater the structure's financing costs. Indeed, with regard to structures whose debt represents 50% or less, the cost of financing is 150 basis points over the 1-month LIBOR. For the two structures that use the most leverage, the cost is 450 basis points more than the benchmark rate. Thus, this increased cost of capital on a per-level basis illustrates the need for increased returns on the underlying portfolio as the structure's leverage increases. Simultaneously, this aspect of a convex increase in costs (150 to 200 bps, 200 to 300 bps, 300 to 450 bps) allows us to observe the risk associated with the junior debt tranches.

The common costs underlying our CFO structures differ slightly from those of Mahadevan and Schwartz (2002). Specifically, the periodic cost of debt is directly calculated by adding the variation in basis points for each level to the 1-month LIBOR rate. Consequently, it is assumed that the structures are perfectly hedged in terms of interest rate risk, which reduces the number of variables handled without however undermining the robustness of the analysis. The senior management fees of 0.5% per

² These are representative of the size and maturity of a typical CFO transaction.

annum and the up-front fees of 3% add to the financial burden of the structure's liabilities. To prevent the up-front fees from disproportionately impacting the first month's results, the latter are financed by the structure via a gradually-amortized loan at the 7-year Swap rate in effect at the time the vehicle is issued (5.85%). For the \$12M million originally due (3% x \$400M), the monthly installment amounts to \$174,441.

Thus, from the above information and a series of returns on a collateralized portfolio, it is possible to derive a periodic return on the Equity Tranche of any structure for any underlying portfolio. To calculate the monthly return of the Equity Tranche of CFO_{j, hk} at time t using the return on E_{hk}, the following formulas are used:

$$Equity_{j,hk,t} = \left\{ \left[NAV_{t-1}(1 + r_{hk,t}) - D_j \left(\frac{LIBOR_t + AFC_j}{12} \right) - \left(NAV_{t-1} * \frac{0,5\%}{12} \right) - PMT \right] - D_j \right\} \quad (7)$$

$$r_{j,hk,t} = \frac{Equity_{j,hk,t} - Equity_{j,hk,t-1}}{Equity_{j,hk,t-1}} \quad (8)$$

where:

$Equity_{j, hk, t}$ is the value of the equity of the structure j, hk at the end of month t ,

NAV_{t-1} is the structure's total value at the beginning of the month t ,

$r_{hk,t}$ is the return on portfolio E_{hk} for the month t ,

D_j is the value of the debt of CFO_j;

$LIBOR_t$ is the value of the 1-month LIBOR at the end of month t ,

AFC_j is the aggregated financing cost of CFO_j;

0.5%/12 corresponds to the senior management fees reported monthly;

PMT corresponds to the monthly payment of \$174,441 concerning the amortization of up-front fees.

A monthly frequency is favored herein; all costs are therefore reported on a monthly basis and the income generated by the model is represented by a series of returns, each with 84 observations (7 years x 12 months). It is also important to mention that contrary to Mahadevan and Schwartz (2002), no distribution of market value gains will be performed during the 7 years of the vehicles' existence. Implicitly, the assumption of

a 7-year lock-up period is in effect with regard to the equity owners. In short, this model is not intended as a perfect representation of the operations observed in practice; the objective herein is to record our findings with regard to the performance of the Equity Tranche in relation to the structure's leverage and the composition of the underlying portfolio. Therefore, the assumptions put forward do not prevent us from meeting this objective. The model presented, which is purposely simple and robust, enables us to meet the objectives of our process without encumbering the analysis with the CFO's structural details.

2.2.1 Evaluating the Equity Tranche

Given the 31 hedge fund portfolios and 20 possible CFO structures, we obtain 620 series of returns (31 × 20 CFO structures) containing 84 observations each.

An initial analysis of the results is performed using a number of performance measures. In addition to calculating the usual descriptive statistics and normality tests for return distributions³, several performance measures are calculated to evaluate the risk-adjusted performance of the different CFO structures. We employ two conventional measures, specifically, the Sharpe and Treynor ratios. We also calculate three more comprehensive measures, namely the Sortino ratio (see Sortino and Price (2004)), the modified Sharpe ratio (see Gregoriou and Gueyi (2003) and Lee (2007)) and the Omega measure (see Keating and Shadwick (2002) and De Souza and Gokcan (2004)).

In sum, after having determined the CFO structures and optimal portfolios from the equity owner's viewpoint using the performance measures stated above, the attributes of capital structure (D/E ratio) and the impact of the diversification constraints on the underlying strategies are analyzed. These results allow us to draw a first round of conclusions.

³ For a complete description of the tests of normality used, refer to Jarque and Bera (1980), Lilliefors (1967) and Genest and Rémillard (2004).

2.3 Analysis of the Systematic Risk Exposure

In order to verify whether the Equity Tranche of each CFO is exposed to systematic risk factors (market, credit and liquidity) we estimate a multivariate linear regression. If the resulting regression coefficients are not significant, this would imply that returns on CFO Equity Tranches are not determined by systematic risk factors and hence could offer significant diversification benefits to traditional portfolios. The same analysis is performed on the returns of the underlying hedge fund portfolios so as to compare the diversification benefits resulting from a direct exposure to the hedge fund portfolios versus that of a CFO. The variables for the regression are presented in Section 4.3.

3. DATA

The hedge fund data was provided by Desjardins Global Asset Management and includes HFR and TASS databases. As of January 31, 2008, 7,533 hedge funds were included in the database. The sample consists of monthly returns of hedge funds over a period of 17 years, from February 1991 to January 2008. The database includes specific information about each fund, including the self-reported investment style.

Of the original 7,533 hedge funds, many had to be eliminated from the study. Firstly, 2,523 funds of hedge funds were withdrawn because they are not strategy-specific funds. Similarly, 126 HFRI and HFRX indexes were deleted. Next, we eliminated 73 funds that did not report their after-fee returns. Also, to ensure the reliability of data used and avoid dealing with self-selection bias, 632 funds were withdrawn from the sample as they registered either less than 12 consecutive months of performance or the disclosure of their results had been interrupted at some point. Finally, 33 "Regulation D" funds were discarded since the latter category only appears as of the late '90s.

Given that this article focuses on the optimal allocation of capital across the various hedge fund strategies, it is necessary to further examine the distribution of funds across the strategies. We observe that across the 29 initial strategies, certain strategies have

more than 300 funds while others have less than 30. To address this relative imbalance, we opt to aggregate similar strategies. After these deletions and aggregations, the returns of 4,146 hedge funds make up the sample over a period of 204 months (17 years x 12 months). Table 2 presents the distribution of hedge funds across strategies before and after aggregation.

<Table2>

We observe that the process results in three new aggregated strategies: *Emerging markets (Asia, E. Europe/CIS, Global and Latin America)*, *Fixed Income (Arbitrage, Convertible Bonds, Diversified, High Yield and Mortgage Backed)* and *Sectorial (Energy, Financial, Health Care/Biotechnology, Miscellaneous, Real Estate and Technology)*. The number of strategies therefore decreases from 29 to 16 and the hedge funds are subsequently better dispersed across the strategies.

In order to circumvent the issue of relative size of the strategies, avoid working with 4,146 assets and further counter the issue of survivorship bias, an equally-weighted index is constructed for each strategy. In other words, an average return per strategy is calculated for each date of the sample. This allows us to perform the portfolio optimization with 16 assets and obtain easily-interpretable results in relation to the number, attributed weight and type of the different strategies so as to create the greatest possible value for the owners of a CFO Equity Tranche. Hence, from a technical point of view, this approach ensures the investment in active (or living) funds for each of the strategies without exposure to the specific risk of a given fund. This therefore avoids periodically rolling funds, allowing us to maintain the necessary weights, i.e. to put forward a passive management strategy.

Furthermore, in order to meet the objectives of the proposed empirical approach, the sample is split into two periods. The first ten years of the sample are used for the optimization of hedge fund portfolios *ex ante*, and the final seven years provide out-of-

sample data to which the CFO structures will be applied. Table 3 presents the statistics of each strategy index over the first ten years and over the final seven years.

<Table 3>

If we first make a general comparison between the data of the two periods, we note that hedge fund returns over period 1 are higher, more volatile and generally demonstrate a greater level of skewness and kurtosis. This result can mainly be explained by the fact that period 1 witnessed a very bullish market while period 2 was more unpredictable and generated less value on the financial markets⁴. In this regard, it should be noted that the data only cover the beginning of the 2008 financial crisis. Nonetheless, despite the obvious disparities between the two periods, the strategies demonstrate similar behavior. We observe that the tests of Jarque-Bera, Lilliefors, and Genest and Remillard reject normality in the respective proportions of 81% (13/16), 56% (9/16) and 63% (10/16) for period 1 and 38% (6/16), 38% (6/16) and 56% (9/16) for period 2. This must be considered when testing the normality of returns of the constructed portfolios and CFOs.

As for the comparison between the strategies themselves, we first observe that the *Emerging Markets* and *Sectorial* strategies demonstrate the best average monthly returns while the *Short Selling* strategy is at the bottom of the ladder. In terms of volatility, the above three strategies are, quite logically, at the top of the list, alongside with the *Equity Non-Hedge* and *Managed Futures* strategies while the majority of market-neutral strategies demonstrate the least volatile returns. Secondly, *Foreign Exchange* strategy stands out with its positive skewness while the *Merger Arbitrage* and *Event Driven* strategies are distinguished by higher kurtosis. When the time comes to dissect the composition of the optimal portfolios, it will be interesting to analyze the weight assigned in light of these preliminary descriptive statistics.

⁴ The *Dow Jones Industrial Average Index* recorded an increase of approximately 300% (14.9% on an annual basis) between February 1991 and January 2001, and less than 20% (2.5% on an annual basis) between February 2001 and January 2008.

Several other financial time series covering the period over which the CFO structures are distributed (2001-2008) are required to perform all the necessary calculations. These series, all expressed on a monthly basis, are obtained from *Bloomberg*. Table 4 summarizes the descriptive statistics for these secondary variables. The 1-month LIBOR and *T-bill* series both have the same average, but the returns on the latter are far more volatile. The 10-year *on-* and *off-the-run* government securities demonstrate almost identical statistical behavior while 20-year *on-the-run* securities offer better average returns and are less volatile. Finally, we note that these three assets exhibit positive skewness (with the exception of the Russell 3000).

<Table 4>

4. EMPIRICAL RESULTS

4.1 Optimal hedge fund allocations

Table 5 presents, for three of the sets of preferences $E_h = \{\alpha, \beta, \gamma\}$, the composition of the optimized hedge fund portfolios.⁵ We observe, conditional on the diversification constraint imposed, the weight y_i assigned to each strategy.

<Table 5>

In analyzing the various distributions, we notice first of all that in the case of mean-variance optimization ($\alpha = 1, \beta = 0, \gamma = 0$), with no weight constraint, one-third of capital is allocated to *Relative Value Arbitrage* and *Foreign Exchange* strategies, respectively. This is logical since these two strategies have the highest reward-risk ratios for period 1, i.e. the time window allotted for the optimization process. As the diversification constraint on the optimized portfolio is increased, the *Equity Hedge* and *Distressed Securities* strategies represent the majority of the composition, with the two previous strategies situated at the highest possible level permissible under the said constraint. It should be noted here that the two least constrained portfolios are identical. This means that even if

⁵ For the purpose of brevity, only the results for preference sets E_1, E_2 and E_3 are presented in Tables 5 and 6.

we allow weights in excess of 50% of the portfolio, no portfolio will be positioned above this threshold for this set of preferences.

With regard to the set of preferences $E_2 = \{\alpha=1; \beta=1; \gamma=0\}$, there is equal preference between returns and skewness; with no importance placed on the level of kurtosis. When the weights y_i are unconstrained, almost 45% of the capital is allocated to *Foreign Exchange* strategy, i.e. the one with greatest level of positive skewness. However, it may be surprising to see that the *Equity Market Neutral* and *Convertible Arbitrage* strategies, which are both negatively skewed, are ranked second and third respectively. This clearly demonstrates the trade-off between skewness and returns for this preference set. It is also interesting to note that as the maximum weight constraint becomes more significant, thereby forceably reducing the allocation in the *Foreign Exchange* strategy, the allocation to *Equity Market Neutral* and *Convertible Arbitrage* strategies practically disappear. It is at this point that the *Equity Hedge*, *Distressed Securities*, *Market Timing*, *Short Selling* and *Managed Futures* strategies start to emerge. These strategies all exhibit positive skewness and lower skewness-return ratios.

As for $E_3 = \{\alpha = 1, \beta = 1, \gamma = 0.75\}$, the relative significance of the kurtosis is evident for the first time. In this case, we note a more harmonious balance of funds across strategies from the very outset. Despite this, the *Foreign Exchange* strategy is still predominant (20% in the least-constrained case) along with *Distressed Securities* (13%) and *Market Timing* (16%). No other strategy obtains more than 10% of the capital. This allocation certainly takes root in the weak kurtosis observed in relation to the significance of the skewness levels of the three main strategies. For this set of preferences, the evolution of diversification constraints has very little impact on the allocation of capital. In this regard, it will be interesting to compare the performance of these portfolios, where there is optimal allocation through all strategies, to that of equally-weighted funds that allocate the same proportion of capital to each strategy.

4.1.1 Performance Analysis

Table 6 presents the descriptive statistics, the tests of normality and the performance measures attributable to the returns of the portfolios examined in period 2. These results are presented for the same sets of preferences as presented earlier, and for the equally-weighted portfolio.

<Table 6>

In analyzing Table 6 with regard to the descriptive statistics for each series of returns, we first see that they are in line with those of their main components, i.e. the strategy indexes. For example, the funds of the set E_2 demonstrate the greatest level of skewness and the non-weight-constrained mean-variance portfolio dominates in terms of the return-variance ratio. These characteristics become diluted as diversification broadens, while the presence of other strategies gradually normalizes the allocations. It is also interesting to note that the equally-weighted portfolio demonstrates the best average returns and the most variance. In this sense, it is not surprising to see average returns decrease with the diversification of the set of preferences E_1 while variance also shows a sharp decrease. However, this same phenomenon does not apply to the other optimized portfolios for which allocation considers both skewness and kurtosis. In short, the mean of all average returns is 0.79 and there is mostly negative skewness and positive excess kurtosis.

With regard to the normality of returns, we first notice that normality is not rejected in all cases according to the Jarque-Bera test. With respect to the other two tests, the returns of five portfolios out of 31 (16.1%) reject the assumption of normality under the Lilliefors test, while six out of 31 (19.4 %) do the same when the Genest and Remillard procedure is used. This is an interesting result. Although most hedge fund strategy indexes reject normality on an individual level, it would appear that when these strategies are combined together, the distributions of the portfolios are considerably more Gaussian. Let us also recall that the strategy indexes exhibit greater normality in period 2, which may be another plausible explanation. In sum, without completely casting aside rejection of the null hypothesis, the decrease in the proportion of rejections is undisputable.

These results may seem small for a hedge fund environment, but they are still significant and confirm the relevance of using comprehensive performance measures, or at least those measures that account for all information contained in the tails of the distributions. It is also interesting to emphasize that normality is only rejected in the case of highly-diversified portfolios; the normality of returns of the equally-weighted portfolio is rejected by the Lilliefors, and Genest and Remillard tests.

With regard to traditional performance measures, it is the unconstrained portfolio of set E_1 that has the highest Sharpe ratio at 0.80 while the equally-weighted portfolio comes in last at 0.61. For the other portfolios, the value of the Sharpe ratio varies between 0.62 and 0.67. Consequently, it appears that no distinction can be made between the portfolios using this performance measure. On the other hand, the unconstrained portfolios of sets E_2 and E_5 stand out in terms of the best Treynor ratio with a value of around 0.80 and 0.20, respectively. Their good performance in terms of Treynor ratio is largely due to the low beta coefficient, i.e. the covariance of returns with those of the market. It should once again be noted that the equally-weighted portfolio has the lowest ratio at 0.04. In sum, if we base the analysis on the Sharpe and Treynor ratios, we would conclude that incorporating preferences for higher moments does not generate value. Equal weighting is systematically an underperforming strategy and diversification generally undermines the performance of a hedge fund portfolio.

If we focus on the more comprehensive performance measures, it is possible to better evaluate and discern the advantages of accounting for higher moments. It should be noted here that for the Sortino ratio and the Omega measure, the monthly minimal accepted returns (MAR) used for the calculations are 0.8% and 1.2%. Table 7 summarizes the performance for all portfolios with regard to the comprehensive performance measure. In addition, a classification of each measure is included.

<Table 7>

For the Sortino ratio and the Omega measure, the equally-weighted portfolio dominates, which is the exact opposite of the observation made using the two simple performance measures. Also, we deduce that equal weighting may be better than the highly-diversified optimizations found using the PGP model. This is similar to the conclusions of DeMiguel, Garlappi and Uppal (2005). Nevertheless, certain funds originating from the optimization procedure are always among the best performers. As for the lowest values, the unconstrained portfolios for the set of preferences E_2 and E_5 are ranked near the bottom, which directly contradicts the results of the Treynor ratio presented earlier. The mean-variance optimal portfolio (preference set E_1) no longer outperforms.

The modified Sharpe ratio offers a similar classification as the conventional Sharpe ratio. Unlike the Omega measure, the modified Sharpe ratio simply incorporates estimates of the skewness and kurtosis coefficients. It does not incorporate all information regarding the actual empirical distribution. Nonetheless, given the non-normality of the returns distributions of certain portfolios, this version of the ratio allows us to establish a more precise classification of the different portfolios of funds.

In short, there are seven portfolios that stand out from our performance analysis. These are: E_1 ; 100%, E_2 ; 100%, E_2 ; 20%, E_4 ; 10%, E_5 ; 50%, E_5 ; 15% and E_6 ; 6.25%. It is clear that the underlying interests appear at two extremes of diversification. Indeed, they are either very little or highly diversified, which means that no unilateral deduction can be made unless it accounts for investor preference. In the following section, we shall determine whether these findings still apply from the viewpoint of a CFO equity owner.

4.2 Performance of the CFOs' Equity Tranche

Table 8 presents the results of the equity of CFO structures 11, 15 and 18. These structures are those deemed optimal from the viewpoint of junior tranche owners. That said, the following remarks arise from the study of all of the structures. As in the previous section, the descriptive statistics, the results of tests of normality and the performance measures are included.

<Table 8>

In analyzing the descriptive statistics, we observe that regardless of the securitized underlying portfolio, the linear increase of leverage within a CFO structure results in a relatively linear increase, at least at the outset, of the first four centered moments of the distribution of equity returns. This finding seems logical *a priori*. However, for a given aggregate cost of financing, where the maximum debt level has been reached, average returns seem to experience a local peak.

With regard to tests of normality, we see that the Jarque-Bera, Lilliefors, and Genest and Remillard tests reject normality in the respective proportions of 5% (28/589), 17% (102/589) and 21% (125/589). In all three cases, this represents an increase compared to the case of unlevered portfolio of hedge funds, which reinforces the interest of using more comprehensive performance measures. Generally speaking, it is possible to conclude that the securitization of hedge funds reduces the normality of the return distribution of the Equity Tranche. When we look more closely at the distribution of "non-normality" across structures, we see a more frequent rejection of the null hypothesis as the structure's leverage increases. Indeed, we note for example that using the method of Genest and Remillard, four out of 31 funds rejected the normality of equity returns in the case of CFO 2; the result is ten out of 31 for CFO 14.

As for the performance indicators, we will focus on the more sophisticated measures. First of all, the Sortino ratio increases as leverage increases, reaching a maximum for CFO 15, for both levels of minimal accepted returns considered. In the case of the unconstrained mean-variance portfolio, the measure is close to 1 with a monthly *MAR* at 0.8%, and 0.82 with a monthly *MAR* at 1.2%. Beyond CFO 15, the ratio decreases toward the weaker levels of CFO 20. According to this measure however, there seems to be an optimal debt level for the equity owner, beyond which it is no longer beneficial to add leverage. With regard to the measure's overall behavior as leverage increases, we once again observe the phenomenon of local peaks for CFOs 11 and 18.

The Omega measure behaves in a similar manner to the Sortino ratio except for the fact that it decreases beyond a certain threshold when we consider a monthly $MAR = 0.8\%$. More specifically, when the minimal accepted return is at this level, the maximum occurs at CFO 18 for a value of almost 1.96, and CFOs 11 and 15 are also distinguished by their local optimality. With monthly $MAR = 1.2\%$, CFO 20 shows the highest values at levels above 1.40. However, if we compare this structure to that of CFO 18, we see that the highest values are very close and that the lowest values of CFO 20 are substantially lower than those of CFO 18. Thus, the benefits of leverage decrease for CFOs 19 and 20. In summary, the three structures using a maximum level of debt still dominate, thus CFOs 11, 15 and 18. Also, it is interesting to note that at this level of minimal accepted returns, it is the equally-weighted portfolio that results in maximum “intra-CFO” values rather than the portfolios resulting from the unconstrained optimization.

Finally, the modified Sharpe ratio behaves much the same way as the conventional Sharpe ratio. It systematically decreases between CFO 2 and CFO 20, and the rate at which the decrease occurs for CFOs 12, 16 and 19 is far more pronounced. The estimates of the higher moments used in the calculation of the modified Sharpe ratio are however subject to significant estimation error as the size of the sample is relatively small. Ultimately, we will rely exclusively on the Sortino and Omega measures given that they account more accurately for the actual empirical distribution of the returns.

In short, the performance analysis indicates that CFOs 11, 15 and 18, which maximize the debt levels for a given aggregate cost CFO structures, are clearly distinguished from the others. The optimal debt-to-equity ratios are therefore 1, $7/3$ and $17/3$. For every dollar of equity, there must be at least one dollar of debt. In this sense, this supports the existence of an optimal debt level from the equity owner’s viewpoint, the merit of CFOs and thus, the added-value of the latter for the investor. These results also show the outperformance of the CFO Equity Tranche, regardless of the structure, when compared to a direct exposure to the portfolio of hedge funds. Indeed, if we compare the results of the Sortino ratio and the Omega measure from one portfolio of hedge funds to another versus the CFO, the latter always outperform the former. It is important to emphasize

that if we rely solely on simple performance measures, we would not come to these conclusions.

With regard to the relationship between the underlying portfolio and the performance of the CFO Equity Tranche, the results show more or less the same pattern as in the previous section. More specifically, the funds all behave in a similar manner from the equity owner's viewpoint, regardless of the amount of leverage. This means that the same seven portfolios are of interest to the CFO Equity Tranche owner and that once again, on the sole basis of risk-adjusted performance, no general conclusion can be drawn as to the diversification requirements for an equity owner. It all depends on the objectives pursued. As a result, there appears to be independence between decisions concerning the underlying portfolio and the debt structure. This is indeed very interesting as CFOs are flexible instruments, meaning that it is possible to choose the types of underlying portfolio according to one's need for diversification and the extent of leverage based on one's appetite for risk.

In short, from an initial universe of 589 potential CFOs, the number is now a mere 21. Indeed only seven portfolios and three structures are of interest. By including the unlevered hedge fund portfolios as CFO 1, we test the systematic risk exposure of 28 of the 620 initial instruments using a multivariate linear regression model.

4.3 Systematic Risk Exposure of the Equity Tranche

The following multivariate linear regression model for the verification of the benefits of diversification was estimated using the returns of the 28 CFO structures identified in the previous section.

$$CFO_{j,hk,t} = \delta_{j,hk,t} + \beta_{1j,hk} STOCK_t + \beta_{2j,hk} DEF_t + \beta_{3j,hk} LIQ_t + \beta_{4j,hk} TERM_t + \beta_{5j,hk} TREND_t + \varepsilon_{j,hk} \quad (9)$$

where:

$\delta_{j,hk,t}$ is the constant term for the CFO_j, subjected to the set of preference E_n and the diversification constraint k , at time t ;

$CFO_{j,hk,t}$ is the monthly return of CFO_j , constructed using the hedge fund portfolio subjected to the set of preference E_h and the diversification constraint k , at time t .

$STOCK_t$ is the monthly return of the Russel 3000 stock index at time t . It is used as an indicator of financial market performance and is included so as to determine whether systematic exposure to equity returns has an explicative power with regard to the performance of a securitized hedge fund portfolio.

DEF_t is the spread in basis points of the LUCI Total OAS⁶ index at time t and is used as an indicator of the systematic default risk of credit markets. It is included in the model so as to ascertain whether systematic default risk in the credit markets influences CFO returns. To this end, several variables are used in the literature to capture this notion of default. However, as stipulated in Longstaff, Mithal and Neis (2005), among others, since the advent of the Credit Default Swaps (CDS) market, it has become common practice to use the spreads on these instruments to identify the default risk embedded in credit risk⁷.

LIQ_t is the spread in basis points between 10-year on-the-run and off-the-run government securities at time t . The variable is used as an indicator of the systematic liquidity risk of bond markets. This indicator is further explained in Longstaff, Mithal and Neis (2005).

$TERM_t$ is the spread in basis points between long-term (20-year) and short-term (1-month t-bills) government securities at time t . It is used as an indicator of systematic interest rate risk. This indicator is the same as that used in Fama and French (1993).

⁶ LUCI is the acronym for *Liquid US Corporate Indices*, a series of bond indexes developed by Cr dit Suisse. The Total OAS (*Option Adjusted Spread*) version captures the entire market, eliminating the specific nature of securities such as embedded options.

⁷ The original intention was to use the CDX.NA.IG index spread. However, a thorough study of the series shows that there is a lack of liquidity prior to 2004, which could bias the results.

TREND_t is the typical time trend that is used when the variables are distributed over time. The idea is to eliminate any time trend effect that may result in autocorrelation in the data series.

Table 9 presents the value of the estimated coefficients and their respective *p-value*. We recall that the ideal scenario is that of completely uncorrelated returns, i.e. statistically null coefficients. In the case of the constant $\delta_{j,hk,t}$ its statistical significance implies the presence of a fixed effect not captured by the model's variables.

<Table 9>

We first note that the constant $\delta_{j,hk,t}$ and the coefficient β_5 , which is attributed to the TREND variable, are always significant. This means, on the one hand, that returns are specific to each of the instruments. In addition, we see that the values taken by $\delta_{j,hk,t}$ are always positive and increase as leverage increases, thus reflecting greater added-value. On the other hand, the significance of β_5 confirms that there is indeed a trend in terms of CFO returns. It is also interesting to note that this coefficient is always negative, implying that returns decrease over the period studied (2001-2008).

When we consider the estimated coefficients of the variables of interest, β_1 to β_4 , it is important to emphasize at the very outset that the statistical significance of the coefficients follows a rather consistent pattern across the different CFO structures tested. Indeed, the same pattern in terms of the statistical significance of the regression coefficients is exhibited in the case of direct exposure to the portfolio of hedge fund and CFO 11, whereas CFOs 15 and 18 present fewer significant coefficients. It can therefore be inferred that leverage has very little effect on systematic diversification. The benefits of diversification are therefore not an argument in favor of CFOs.

Regardless of the choice of underlying portfolio and of the selected capital structure, the coefficient estimates indicate that the liquidity (LIQ) and slope of the term structure (TERM) variables do not significantly impact CFO returns. With respect to the STOCK variable, which relates to systematic equity market risk, we note that the estimated

coefficient is almost always significant. The estimated coefficient for the default risk variable is less stable across the different CFO structures. In the case of exposure to CFO 1 and CFO 11, we see that the DEF variable has a significant, negative impact in the case of underlying portfolios $E_{1;100\%}$, $E_{2;20\%}$ and $E_{5;15\%}$. For CFOs 15 and 18, there is a gradual loss of significance of the estimated coefficient for the last two underlying portfolios ($E_{2;20\%}$ et $E_{5;15\%}$).

Thus, it is the unconstrained mean-variance portfolio that fairs the most poorly in terms of systematic risk exposures, while the unconstrained version of preference set E_2 demonstrates the least dependence to the selected systematic risk factors. We recall however that this underlying portfolio underperforms in terms of most performance measures. It therefore appears necessary to reach a compromise between performance and decorrelation with the capital market as the best of both worlds can not be achieved simultaneously. Two interesting alternatives appear to be that of $E_{4;10\%}$ and the equally-weighted portfolio that both exhibit a slight positive correlation with equity markets. Although not an optimal solution, these results provide strong evidence that it is possible to construct CFOs that generate value in terms of both performance and low systematic risk exposure.

On the basis of the results above, when the objective pursued by the investor in terms of CFO equity is either performance, or decorrelation, an unconstrained portfolio in which *Foreign Exchange* strategy is predominant is essential. Where one wishes to combine the two objectives, there must be substantial diversification of the underlying portfolio. With regard to the debt-to-equity ratio of the capital structure selected, there is almost no impact, notwithstanding exceptions. This therefore implies that CFOs represent added-value compared to a direct exposure to the portfolio of hedge funds, only in terms of performance however, not with regard to diversification in capital markets.

5. CONCLUSION

In response to suggestions that CFO Equity Tranches do not offer value to investors, we decided to undertake a thorough examination of the performance of this type of investment. The objective was to assess the performance of the Equity Tranche of CFOs both in terms of risk-adjusted return as well as systematic risk exposures. For this purpose, 30 optimal portfolios (each conditional to a set of preferences and weight constraints) and an equally-weighted portfolio were constructed using 16 hedge fund strategy indexes. All of these underlying portfolios were then securitized using 19 capital structures, which allowed us to analyze the series of returns of 620 CFO Equity Tranches.

Interestingly, we observe that if we consider the overall distribution of returns of a CFO Equity Tranche in analyzing the performance, securitizing and tranching the underlying portfolio of hedge funds adds value for the end investor. Our analysis also finds that there was no direct relationship between the optimal portfolio to securitize and the capital structure decisions. With regard to the debt-to-equity ratio, we conclude that leverage is beneficial for the CFO equity holder when the level of debt is maximized for a given funding cost. In addition, our analysis shows that a trade-off takes place between performance and systematic diversification; if one considers a combination of the two objectives, we conclude that the underlying portfolio must be broadly diversified across the various hedge fund strategies. Thus, these conclusions suggest that market participants might have been too hasty in dismissing CFOs, and not taking greater advantage of the benefits offered by these investment vehicles.

If the reputation of securitization had not already been undermined by its involvement in the current financial crisis, the impact of this two-dimensional analysis could have been rather different. However, given the present stigma attached to structured products, we should not expect to witness new CFO transactions anytime soon. This is unfortunate considering that the nature of the problem is one of inaccurate valuation rather than one of overexuberant financial engineering, as explained in Longstaff and Myers (2009).

Global financial markets are currently seeking to rebuild a sound financial system and it will be interesting to see whether CFOs will be favored among the range of investments available to institutional investors. For this to transpire, securitization and financing practices must be rehabilitated in conjunction with proper governance rules that promote greater transparency.

Since it is only a matter of time before financial markets have had the chance to digest the consequences of the current crisis, it is relevant to pursue this research further. Under this new analysis, it would be consistent to consider alternative ways in which to further expand on the conclusions. For example, an actively-managed underlying portfolio could be of interest. Indeed, rather than turning to strategy indexes that are not readily investible, it would be suitable to use investments in specific funds and make readjustments as necessary. From a risk management viewpoint, it would be appropriate to examine the potential tools that would allow exposure to a CFO Equity Tranche to provide strong risk-adjusted performance with negligible exposure to systematic risk factors. Finally, it would be important to assess the impact of high market stress (like the one that prevailed in 2007 and 2008) in order to analyze the behavior of CFOs in such circumstances.

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Performance Analysis of a CFO Equity Tranche

Table 1
Structural Assumptions for the 20 CFOs Considered

Table 1 presents the 20 capital structures considered. The characteristics of each structure appear on the vertical axis, while the 20 CFO structures are presented horizontally. The upper portion of the table shows the features of the five possible tranches and the lower one relates to cost assumptions. Shaded areas designate the absence of a certain tranche for a specific structure.

	CFO 1	CFO 2	CFO 3	CFO 4	CFO 5	CFO 6	CFO 7	CFO 8	CFO 9	CFO 10	CFO 11	CFO 12	CFO 13	CFO 14	CFO 15	CFO 16	CFO 17	CFO 18	CFO 19	CFO 20
Senior Notes																				
Rating		AA																		
Par Value		\$20 M	\$40 M	\$60 M	\$80 M	\$100 M	\$120 M	\$140 M	\$160 M	\$180 M	\$200 M									
% of Structure		5%	10%	15%	20%	25%	30%	35%	40%	45%	50%	50%	50%	50%	50%	50%	50%	50%	50%	50%
Advance Rate		60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%	60%
Class II Notes																				
Rating												BBB								
Par Value												\$20 M	\$40 M	\$60 M	\$80 M					
% of Structure												5%	10%	15%	20%	20%	20%	20%	20%	20%
Advance Rate												75%	75%	75%	75%	75%	75%	75%	75%	75%
Class III Notes																				
Rating																BB	BB	BB	BB	BB
Par Value																\$20 M	\$40 M	\$60 M	\$60 M	\$60 M
% of Structure																5%	10%	15%	15%	15%
Advance Rate																82%	82%	82%	82%	82%
Class IV Notes																				
Rating																				BB-
Par Value																				\$20 M
% of Structure																				5%
Advance Rate																				90%
Total Debt	- \$	\$20 M	\$40 M	\$60 M	\$80 M	\$100 M	\$120 M	\$140 M	\$160 M	\$180 M	\$200 M	\$220 M	\$240 M	\$260 M	\$280 M	\$300 M	\$320 M	\$340 M	\$360 M	\$380 M
Equity Tranche																				
Par Value	\$400 M	\$380 M	\$360 M	\$340 M	\$320 M	\$300 M	\$280 M	\$260 M	\$240 M	\$220 M	\$200 M	\$180 M	\$160 M	\$140 M	\$120 M	\$100 M	\$80 M	\$60 M	\$40 M	\$20 M
% of Structure	100%	95%	90%	85%	80%	75%	70%	65%	60%	55%	50%	45%	40%	35%	30%	25%	20%	15%	10%	5%
% of MV Gains Distributed on a Current Basis	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%	0%
Total Par Value of CFO	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M	\$400 M
D/E Ratio	-	0.053	0.111	0.176	0.25	0.333	0.429	0.538	0.667	0.818	1.00	1.222	1.50	1.857	2.333	3.00	4.00	5.667	9.00	19.00
Costs and Fees																				
All in Funding Cost (Spread in bps)	-	150	150	150	150	150	150	150	150	150	150	200	200	200	200	300	300	300	450	450
Senior Management Fee	N/A	50 bps																		
Up-Front Fees	N/A	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%	3%
Benchmark	N/A	1M LIBOR																		
7-year Swap rate @ t=0	N/A	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%	5.85%
Stated Maturity (All Securities)	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years	7 Years

Table 2
Distribution of Hedge Fund Strategies within Sample

The table shows the absolute and relative frequencies of each of the hedge fund strategies in our sample. Panel A describes the sample before the aggregation of certain strategies while Panel B presents the sample after aggregation.

PANEL A
Sample BEFORE Strategy Aggregation

	Strategy	Frequency	Relative Frequency
1	Convertible Arbitrage	100	2.39%
2	Distressed Securities	112	2.68%
3	Emerging Markets: Asia	88	2.11%
4	Emerging Markets: E. Europe/CIS	71	1.70%
5	Emerging Markets: Global	109	2.61%
6	Emerging Markets: Latin America	26	0.62%
7	Equity Hedge	1359	32.52%
8	Equity Market Neutral	263	6.29%
9	Equity Non-Hedge	135	3.23%
10	Event-Driven	248	5.93%
11	Fixed Income: Arbitrage	75	1.79%
12	Fixed Income: Convertible Bonds	25	0.60%
13	Fixed Income: Diversified	86	2.06%
14	Fixed Income: High Yield	67	1.60%
15	Fixed Income: Mortgage-Backed	56	1.34%
16	Foreign Exchange	75	1.79%
17	Macro	293	7.01%
18	Managed Futures	299	7.15%
19	Market Timing	20	0.48%
20	Merger Arbitrage	37	0.89%
21	Regulation D	33	0.79%
22	Relative Value Arbitrage	288	6.89%
23	Sector: Energy	81	1.94%
24	Sector: Financial	33	0.79%
25	Sector: Health Care/Biotechnology	47	1.12%
26	Sector: Miscellaneous	35	0.84%
27	Sector: Real Estate	41	0.98%
28	Sector: Technology	58	1.39%
29	Short Selling	19	0.45%
	TOTAL	4179	100%

PANEL B
Sample AFTER Strategy Aggregation

	Strategy	Frequency	Relative frequency
1	Convertible Arbitrage	100	2.41%
2	Distressed Securities	112	2.70%
3	Equity Hedge	1359	32.78%
4	Equity Market Neutral	263	6.34%
5	Equity Non-Hedge	135	3.26%
6	Event-Driven	248	5.98%
7	Foreign Exchange	75	1.81%
8	Macro	293	7.07%
9	Managed Futures	299	7.21%
10	Market Timing	20	0.48%
11	Merger Arbitrage	37	0.89%
12	Relative Value Arbitrage	288	6.95%
13	Short Selling	19	0.46%
14	Emerging Markets	294	7.09%
15	Fixed Income	309	7.45%
16	Sectorial	295	7.12%
	TOTAL	4146	100%

Table 3
Statistics of Hedge Fund Strategy Indexes

Table 3 presents the descriptive statistics and the results of the Jarque-Bera (JB), Lilliefors, and Genest-Remillard (G & R) normality tests for every hedge fund strategy in the sample. Period 1 runs from February 1991 to January 2001, while period 2 covers the seven following years to January 2008. For both panels, the nature of the statistics appears on the left-hand-side and the strategies feature horizontally. The descriptive statistics are monthly and include the first four moments of the distributions, as well as the minimum and maximum returns. With regard to the tests for normality, the value of the test statistic and a dummy variable indicating whether or not the value is significant at a confidence level of 95% are presented*. The dummy variable equals 1 when the null hypothesis of normality is rejected.

Period 1: 120 months (February 1991 to January 2001)																
	Convert. Arbit.	Dist. Securities	Equity Hedge	E. Mkt Neutral	Equity N-Hedge	Event-Driven	FX	Macro	Mnged Fut.	Market Timing	Merger Arbit.	Relat. Val Arbit.	Short Selling	Emerg. Mkts	Fixed Inc.	Sectorial
Mean	0.0119	0.0151	0.0180	0.0101	0.0165	0.0165	0.0161	0.0138	0.0146	0.0152	0.0118	0.0145	-0.0039	0.0244	0.0109	0.0200
Standard deviation	0.0116	0.0216	0.0273	0.0111	0.0355	0.0225	0.0206	0.0213	0.0456	0.0257	0.0118	0.0114	0.0828	0.0659	0.0114	0.0410
Skewness	-0.1963	0.9182	0.0413	-0.2320	-0.6419	-0.5284	1.5295	0.2341	0.9591	0.7721	-2.5619	-0.7081	0.1929	-0.1586	-0.6013	-0.5642
Kurtosis	5.3599	7.4489	4.1026	2.5946	5.8056	8.8399	7.6032	3.9269	5.9506	4.9391	19.3466	4.501	4.8889	3.2656	6.2689	6.7705
Min	-0.0321	-0.0647	-0.0809	-0.0195	-0.1473	-0.0948	-0.0204	-0.0561	-0.0924	-0.0464	-0.0684	-0.0230	-0.3036	-0.1946	-0.0377	-0.1733
Max	0.0510	0.1086	0.1025	0.0338	0.1116	0.1094	0.1178	0.0826	0.2281	0.1154	0.0393	0.0402	0.2783	0.1874	0.0450	0.1399
JB Test (df=2)	28.62	115.82	6.11	1.90	47.60	176.11	152.74	5.39	61.93	30.72	1467.32	21.30	18.58	0.86	60.66	77.45
α=5% ; x ≤ 5.99	1	1	1	0	1	1	1	0	1	1	1	1	1	0	1	1
Lilliefors Test	0.007	0.001	0.50	0.02	0.50	0.16	0.001	0.17	0.04	0.16	0.008	0.003	0.008	0.50	0.10	0.01
α=5% ; p-val ≥ 5%	1	1	0	1	0	0	1	0	1	0	1	1	1	0	0	1
G & R Test	0.00	0.00	0.89	0.09	0.59	0.01	0.00	0.19	0.01	0.06	0.00	0.00	0.00	0.95	0.04	0.02
α=5% ; p-val ≥ 5%	1	1	0	0	0	1	1	0	1	0	1	1	1	0	1	1

Period 2: 84 months (February 2001 to January 2008)																
	Convert. Arbit.	Dist. Securities	Equity Hedge	E. Mkt Neutral	Equity N-Hedge	Event-Driven	FX	Macro	Mnged Fut.	Market Timing	Merger Arbit.	Relat. Val Arbit.	Short Selling	Emerg. Mkts	Fixed Inc.	Sectorial
Mean	0.0064	0.0114	0.0086	0.0056	0.0103	0.0089	0.0069	0.0094	0.0105	0.0064	0.0052	0.0085	0.0049	0.0167	0.0067	0.0113
Standard deviation	0.0111	0.0134	0.0180	0.0064	0.0345	0.016	0.0099	0.0136	0.0279	0.0243	0.0086	0.0065	0.0470	0.0295	0.0083	0.0214
Skewness	-0.4950	-0.1774	-0.6148	-1.2205	-0.4870	-0.7719	1.4167	-0.0101	-0.0549	-0.0420	-0.1752	-0.6253	0.1632	-0.9236	-0.6435	-0.3815
Kurtosis	3.5464	3.2546	2.9455	5.7064	2.7287	3.6133	5.9688	2.7894	2.7363	2.5760	3.9317	3.6192	2.8247	3.3725	3.4356	2.5501
Min	-0.0284	-0.0240	-0.0414	-0.0196	-0.0774	-0.0342	-0.0067	-0.0224	-0.0570	-0.0514	-0.0210	-0.0127	-0.1013	-0.0679	-0.0176	-0.0395
Max	0.0339	0.0459	0.0444	0.0164	0.0774	0.0387	0.0444	0.0433	0.0719	0.0596	0.0260	0.0206	0.1281	0.0649	0.0236	0.0570
JB Test (df=2)	4.48	0.67	5.30	46.49	3.58	9.66	58.95	0.16	0.29	0.65	3.47	6.82	0.48	12.43	6.46	2.75
α=5% ; x ≤ 5.99	0	0	0	1	0	1	1	0	0	0	0	1	0	1	1	0
Lilliefors Test	0.12	0.18	0.05	0.02	0.07	0.003	0.006	0.50	0.50	0.38	0.02	0.13	0.39	0.001	0.09	0.15
α=5% ; p-val ≥ 5%	0	0	1	1	0	1	1	0	0	0	1	0	0	1	0	0
G & R Test	0.03	0.44	0.01	0.01	0.09	0.00	0.00	0.92	0.99	0.78	0.04	0.09	0.35	0.00	0.02	0.03
α=5% ; p-val ≥ 5%	1	0	1	1	0	1	1	0	0	0	1	0	0	1	1	1

* P-values are available upon request.

Table 4
Descriptive Statistics of Secondary Data

The table presents the first four moments of the distribution as well as the maximum and minimum. The statistics appear on the vertical axis, while the time series feature horizontally. They are all presented on a monthly return basis. 1M LIBOR and 1M T-Bill are relative to the money market while the Russell 3000 is relative to the stock market. CDX.NA.IG_5Y and LUCI_OAS relate to the corporate credit market while the three whose title starts with USA are related to the U.S. government securities, 10- and 20-year maturities. For the 10-year maturity, on-the-run (on-t-r) and off-the-run (off-t-r) rates are presented.

Period 2: 84 months (February 2001 to January 2008)								
	LIBOR 1M	T-Bill 1M	Russel 3000	CDX.NA.IG_5Y ^a	LUCI_OAS*	USA_10Y (on-t-r)	USA_10Y (off-t-r)	USA_20Y
Mean	0.0026	0.0026	0.0015	4.4948	12.0387	0.0037	0.0037	0.0042
Variance	1.87E-06	5.10E-07	0.0016	1.5875	14.6199	1.50E-07	1.60E-07	1.10E-07
Skewness	0.1906	0.3597	-0.4743	0.9061	0.7873	-0.0265	-0.0301	0.3986
Excess kurtosis	-1.6071	-1.2216	0.3473	1.485	-0.5299	-0.4791	-0.5402	-0.5446
Max	0.0048	0.0042	0.0803	9.0417	21.7833	0.0045	0.0045	0.0049
Min	0.0009	0.0016	-0.1064	2.5419	7.5167	0.0028	0.0028	0.0036

* Data are shown in terms of spread in basis points.

^a The statistics relative to this series are calculated from 58 months of available data.

Table 5
Composition of Optimal Portfolios for Sets of Preferences E₁, E₂ and E₃

Table 5 shows the composition of optimal portfolios for the first three sets of preferences. All hedge fund strategies appear on the vertical axis and the different portfolios (set of preference and diversification constraint) are presented horizontally. The sum of each column equals to 1, as the weights are listed as a proportion of the total value and are all comprised between 0 and 1 (no-short constraint). Blanks refer to null allocations.

HF Strategy	E ₁ : { $\alpha = 1$; $\beta = 0$; $\gamma = 0$ }						E ₂ : { $\alpha = 1$; $\beta = 1$; $\gamma = 0$ }						E ₃ : { $\alpha = 1$; $\beta = 1$; $\gamma = 0.75$ }					
	y <= 100%	y <= 50%	y <= 25%	y <= 20%	y <= 15%	y <= 10%	y <= 100%	y <= 50%	y <= 25%	y <= 20%	y <= 15%	y <= 10%	y <= 100%	y <= 50%	y <= 25%	y <= 20%	y <= 15%	y <= 10%
Convertible Arbitrage				0.0267	0.0124	0.0828	0.1364	0.1372		0.0150	0.0799	0.0775	0.0501	0.0501	0.0531	0.0510	0.0569	0.0659
Distressed Securities	0.0941	0.0941	0.0988	0.1154	0.1500	0.1000			0.2500	0.2000	0.1500	0.1000	0.1326	0.1326	0.1347	0.1315	0.1340	0.1000
Equity Hedge	0.1195	0.1195	0.2023	0.1579	0.1500	0.1000			0.1288	0.2000	0.1500	0.1000	0.0688	0.0688	0.0690	0.0678	0.0700	0.0792
Equity Market Neutral						0.0522	0.1780	0.1774			0.0579	0.0659	0.0488	0.0488	0.0566	0.0503	0.0574	0.0714
Equity Non-Hedge												0.0270	0.0154	0.0154	0.0133	0.0148	0.0197	0.0316
Event-Driven	0.0175	0.0175		0.0537	0.0484	0.0889				0.0003		0.0436	0.0360	0.0360	0.0247	0.0353	0.0367	0.0430
Foreign Exchange	0.2986	0.2986	0.2500	0.2000	0.1500	0.1000	0.4418	0.4418	0.2500	0.2000	0.1500	0.1000	0.2055	0.2055	0.2018	0.2000	0.1500	0.1000
Macro						0.0077						0.0454			0.0335			0.0569
Managed Futures				0.0013	0.0106	0.0241					0.1115	0.1000	0.0538	0.0538	0.0606	0.0602	0.0926	0.0870
Market Timing	0.0482	0.0482	0.0582	0.0829	0.1046	0.0926	0.0419	0.0421	0.2500	0.2000	0.1500	0.1000	0.1616	0.1616	0.1642	0.1633	0.1500	0.1000
Merger Arbitrage	0.0524	0.0524	0.0645	0.0777	0.1074	0.0949	0.0931	0.0935	0.0050		0.0226	0.0238	0.0226	0.0226	0.0253	0.0237	0.0248	0.0312
Relative Value Arbitrage	0.3154	0.3154	0.2500	0.2000	0.1500	0.1000	0.0147	0.0157	0.0248	0.0271	0.0412	0.0531	0.0480	0.0480	0.0460	0.0488	0.0513	0.0572
Short Selling	0.0344	0.0344	0.0475	0.0486	0.0629	0.0587	0.0307	0.0307	0.0914	0.1103	0.0850	0.0923	0.0739	0.0739	0.0681	0.0735	0.0782	0.0828
Emerging Markets	0.0198	0.0198	0.0211	0.0128	0.0273	0.0271				0.0209	0.0009	0.0153	0.0268	0.0268	0.0159	0.0257	0.0199	0.0209
Fixed Income						0.0204	0.0635	0.0616			0.0011	0.0295	0.0279	0.0279	0.0198	0.0278	0.0337	0.0417
Sectorial			0.0077	0.023	0.0265	0.0505					0.0265	0.0266	0.0283	0.0283	0.0133	0.0263	0.0248	0.0311

Table 6
Performance of Optimal Portfolios for Sets of Preferences E₁, E₂, E₃ and E₆

The table presents the descriptive statistics, the results of the Jarque-Bera (JB), Lilliefors, and Genest-Remillard (G & R) tests for normality and the performance measures of the optimal portfolios for sets 1, 2 and 3 of preference for the moments of distributions; the equally-weighted (EW) portfolio is also presented. Each of the indicators appears on the vertical axis and the different portfolios (set of preference and diversification constraint) are presented horizontally. The descriptive statistics are monthly and include the first four moments of the distributions, as well as the maximum and minimum returns. With regard to the tests for normality, the value of the test statistic and a dummy variable indicating whether or not the value is significant at a confidence level of 95% are presented*. The dummy variable equals 1 when the null hypothesis of normality is rejected. The performance measures (Sharpe, Treynor, Sortino, Modified Sharpe and Omega (Ω)) are also presented on a monthly basis. Minimal Accepted Return (MAR) levels of 0.8% and 1.2% monthly are used for the calculations of the Sortino ratio and the Omega measure. The Modified Sharpe is computed based on a Value-at-Risk calculation with a 95% confidence level.

	E ₁ : {α = 1; β = 0; γ = 0}						E ₂ : {α = 1; β = 1; γ = 0}						E ₃ : {α = 1; β = 1; γ = 0.75}						E ₆ : EW
	y <= 100%	y <= 50%	y <= 25%	y <= 20%	y <= 15%	y <= 10%	y <= 100%	y <= 50%	y <= 25%	y <= 20%	y <= 15%	y <= 10%	y <= 100%	y <= 50%	y <= 25%	y <= 20%	y <= 15%	y <= 10%	y = 6.25%
Mean	0.008077	0.008077	0.008095	0.008058	0.008215	0.008003	0.006376	0.006378	0.007971	0.008180	0.007902	0.008110	0.008031	0.008031	0.007940	0.008021	0.008091	0.008156	0.008597
Variance	4.78E-05	4.78E-05	5.60E-05	6.35E-05	6.94E-05	7.10E-05	3.81E-05	3.82E-05	7.04E-05	7.03E-05	6.86E-05	6.88E-05	6.90E-05	6.90E-05	6.92E-05	6.95E-05	7.18E-05	7.08E-05	9.88E-05
Skewness	-0.3793	-0.3793	-0.4507	-0.4436	-0.4165	-0.4582	0.2493	0.2492	-0.0005	-0.1584	-0.1268	-0.2366	-0.1314	-0.1314	-0.0869	-0.1205	-0.1384	-0.2457	-0.4270
Excess kurtosis	0.2581	0.2581	0.3661	0.3442	0.2639	0.2604	-0.1148	-0.1160	0.1839	0.1600	-0.1919	-0.2506	0.0175	0.0175	-0.0213	0.0046	-0.1070	-0.1832	-0.0592
Max	0.0224	0.0224	0.0239	0.0249	0.0268	0.0265	0.0228	0.0228	0.0300	0.0298	0.0277	0.0270	0.0278	0.0278	0.0279	0.0279	0.0280	0.0271	0.0291
Min	-0.0113	-0.0113	-0.0133	-0.0137	-0.0139	-0.0138	-0.0068	-0.0068	-0.0149	-0.0136	-0.0110	-0.0115	-0.0126	-0.0126	-0.0124	-0.0125	-0.0118	-0.0116	-0.0150
JB Test (df = 2)	2.25	2.25	3.31	3.17	2.67	3.18	0.92	0.92	0.12	0.44	0.35	1.00	0.24	0.24	0.11	0.20	0.31	0.96	2.57
α=5% ; x ≤ 5.99	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Lilliefors Test	0.50	0.50	0.09	0.03	0.08	0.15	0.50	0.50	0.50	0.24	0.05	0.10	0.38	0.38	0.50	0.38	0.49	0.12	0.04
α= 5% ; p-val ≥ 5%	0	0	0	1	0	0	0	0	0	0	1	0	0	0	0	0	0	0	1
Genest & Remillard Test	0.40	0.45	0.11	0.06	0.03	0.02	0.60	0.61	0.69	0.20	0.21	0.04	0.33	0.47	0.58	0.33	0.24	0.03	0.00
α= 5% ; p-val ≥ 5%	0	0	0	0	1	1	0	0	0	0	0	1	0	0	0	0	0	1	1
Sharpe	0.7972	0.7972	0.7393	0.6898	0.6785	0.6457	0.6174	0.6174	0.6446	0.6702	0.6447	0.6689	0.6583	0.6583	0.6466	0.6549	0.6525	0.6649	0.6070
Treynor	0.0702	0.0702	0.0568	0.0481	0.0474	0.0461	0.7915	0.7981	0.0522	0.0553	0.1018	0.0916	0.0586	0.0586	0.0621	0.0597	0.0684	0.0731	0.0418
Sortino MAR=0	4.1199	4.1199	3.4138	3.0041	2.9266	2.6633	4.8478	4.8465	3.2773	3.2394	3.2702	3.3239	3.2284	3.2284	3.2339	3.2186	3.2002	3.2059	2.3109
Sortino MAR=Rf	2.0198	2.0198	1.7276	1.5519	1.5263	1.3883	1.6841	1.6842	1.6107	1.6178	1.5771	1.6238	1.5932	1.5932	1.5808	1.5872	1.5851	1.5893	1.2827
Sortino MAR=0.004	1.2536	1.2536	1.0929	0.9896	0.9847	0.8910	0.8132	0.8136	1.0037	1.0289	0.9707	1.0153	0.9993	0.9993	0.9812	0.9944	0.9968	1.0040	0.8658
Sortino MAR=0.008	0.0151	0.0151	0.0171	0.0098	0.0350	0.0005	-0.3131	-0.3128	-0.0049	0.0301	-0.0161	0.0182	0.0052	0.0052	-0.0100	0.0035	0.0150	0.0254	0.0829
Sortino MAR=0.012	-0.5255	-0.5255	-0.4957	-0.4775	-0.4495	-0.4623	-0.6963	-0.6961	-0.4793	-0.4573	-0.4834	-0.4620	-0.4733	-0.4733	-0.4814	-0.4733	-0.4610	-0.4539	-0.3614
Sortino MAR=0.016	-0.7610	-0.7610	-0.7360	-0.7178	-0.6970	-0.7016	-0.8462	-0.8461	-0.7101	-0.6996	-0.7142	-0.7033	-0.7090	-0.7090	-0.7129	-0.7084	-0.6995	-0.6971	-0.6174
Modified Sharpe 95% (α=5%)	1.3294	1.3294	1.0387	0.8847	0.8526	0.7606	1.1171	1.1169	0.9324	0.9382	0.8796	0.8952	0.9174	0.9174	0.9025	0.9110	0.8887	0.8778	0.6542
Ω (MAR=0)	16.3885	16.3885	12.6570	10.8673	10.2516	9.0137	16.8982	16.8859	11.0174	10.5652	10.6381	10.7006	10.5186	10.5186	10.7024	10.4917	10.3898	10.1550	7.4065
Ω (MAR=Rf)	6.9654	6.9654	5.8272	5.2087	4.9906	4.6330	4.9259	4.9265	5.0842	5.1061	4.7611	4.8767	4.9892	4.9892	4.8924	4.9530	4.8828	4.8584	4.1459
Ω (MAR=0.004)	4.2247	4.2247	3.7307	3.4195	3.3414	3.1304	2.6800	2.6812	3.2992	3.3972	3.1035	3.2112	3.2872	3.2872	3.2009	3.2638	3.2174	3.2272	2.9930
Ω (MAR=0.008)	1.0288	1.0288	1.0337	1.0190	1.0682	1.0010	0.5219	0.5224	0.9912	1.0568	0.9711	1.0335	1.0097	1.0097	0.9817	1.0065	1.0276	1.0471	1.1620
Ω (MAR=0.012)	0.2149	0.2149	0.2368	0.2548	0.2856	0.2639	0.1044	0.1046	0.2932	0.3002	0.2666	0.2812	0.2822	0.2822	0.2760	0.2829	0.2954	0.2918	0.3955
Ω (MAR=0.016)	0.0443	0.0443	0.0525	0.0600	0.0683	0.0638	0.0208	0.0208	0.0808	0.0772	0.0705	0.0702	0.0766	0.0766	0.0764	0.0775	0.0802	0.0738	0.1117

* P-values are available upon request.

Table 7
Performance and Rank of Portfolios According to the Comprehensive Measures

The table shows the results, on a monthly basis, of the Sortino ratio, the Modified Sharpe ratio and the Omega measure (Ω) for all 30 optimal portfolios as well as the equally-weighted fund. The 31 portfolios and their characteristics (sets of preference and diversification constraint) appear on the vertical axis and the performance measures are presented horizontally. Minimal Accepted Return (MAR) levels of 0.8% and 1.2% monthly are used for the calculations of the Sortino ratio and the Omega measure. The Modified Sharpe was computed based on a Value-at-Risk calculation with a 95% confidence level. To the right of each measure is the rank of the portfolio for a given indicator; 1 indicates the best performing portfolio and 31, the worst performing one.

E_h	k	Sortino MAR=0,008	Rank	Sortino MAR=0,012	Rank	Modified Sharpe 95%	Rank	Ω (MAR=0,008)	Rank	Ω (MAR=0,012)	Rank
$E_1: \{\alpha=1; \beta=0; \gamma=0\}$	$y \leq 100\%$	0.0151	12	-0.5255	26	1.3294	1	1.0288	12	0.2149	26
	$y \leq 50\%$	0.0151	12	-0.5255	26	1.3294	1	1.0288	12	0.2149	26
	$y \leq 25\%$	0.0171	11	-0.4957	25	1.0387	7	1.0337	10	0.2368	25
	$y \leq 20\%$	0.0098	15	-0.4775	21	0.8847	22	1.0190	15	0.2548	24
	$y \leq 15\%$	0.0350	2	-0.4495	4	0.8526	29	1.0682	2	0.2856	12
	$y \leq 10\%$	0.0005	24	-0.4623	12	0.7606	30	1.0010	24	0.2639	23
$E_2: \{\alpha=1; \beta=1; \gamma=0\}$	$y \leq 100\%$	-0.3131	31	-0.6963	31	1.1171	5	0.5219	31	0.1044	31
	$y \leq 50\%$	-0.3128	30	-0.6961	30	1.1169	6	0.5224	30	0.1046	30
	$y \leq 25\%$	-0.0049	25	-0.4793	22	0.9324	10	0.9912	25	0.2932	9
	$y \leq 20\%$	0.0301	5	-0.4573	9	0.9382	9	1.0568	5	0.3002	6
	$y \leq 15\%$	-0.0161	27	-0.4834	24	0.8796	24	0.9711	27	0.2666	22
	$y \leq 10\%$	0.0182	10	-0.4620	11	0.8952	17	1.0335	11	0.2812	17
$E_3: \{\alpha=1; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0052	21	-0.4733	19	0.9174	11	1.0097	21	0.2822	15
	$y \leq 50\%$	0.0052	21	-0.4733	19	0.9174	11	1.0097	21	0.2822	15
	$y \leq 25\%$	-0.0100	26	-0.4814	23	0.9025	15	0.9817	26	0.2760	21
	$y \leq 20\%$	0.0035	23	-0.4733	18	0.9110	13	1.0065	23	0.2829	14
	$y \leq 15\%$	0.0150	14	-0.4610	10	0.8887	21	1.0276	14	0.2954	8
	$y \leq 10\%$	0.0254	8	-0.4539	7	0.8778	26	1.0471	8	0.2918	10
$E_4: \{\alpha=2; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0067	19	-0.4711	15	0.8944	18	1.0126	19	0.2811	18
	$y \leq 50\%$	0.0067	19	-0.4711	15	0.8944	18	1.0126	19	0.2811	18
	$y \leq 25\%$	0.0070	18	-0.4722	17	0.9029	14	1.0133	18	0.2794	20
	$y \leq 20\%$	0.0082	17	-0.4691	13	0.8907	20	1.0153	17	0.2832	13
	$y \leq 15\%$	0.0278	7	-0.4514	5	0.8797	23	1.0520	7	0.3019	5
	$y \leq 10\%$	0.0337	3	-0.4449	2	0.8531	28	1.0632	3	0.2977	7
$E_5: \{\alpha=3; \beta=1; \gamma=0.25\}$	$y \leq 100\%$	-0.2761	29	-0.6781	29	1.1416	4	0.5693	29	0.1170	29
	$y \leq 50\%$	-0.2754	28	-0.6779	28	1.1439	3	0.5702	28	0.1172	28
	$y \leq 25\%$	0.0087	16	-0.4700	14	0.9402	8	1.0156	16	0.3046	3
	$y \leq 20\%$	0.0293	6	-0.4515	6	0.8965	16	1.0553	6	0.3097	2
	$y \leq 15\%$	0.0318	4	-0.4476	3	0.8680	27	1.0603	4	0.3038	4
	$y \leq 10\%$	0.0236	9	-0.4572	8	0.8780	25	1.0439	9	0.2864	11
$E_6: EW (1/m)$	$y = 6.25\%$	0.0829	1	-0.3614	1	0.6542	31	1.1620	1	0.3955	1

Table 8
Performance of CFOs 11, 15 and 18

Table 8 presents the descriptive statistics, the results of the Jarque-Bera (JB), Lilliefors, and Genest-Remillard (G & R) tests for normality and the performance measures of CFO structures 11 (panel A), 15 (Panel B) and 18 (Panel C) constructed from the 31 hedge fund portfolios. For each of the three CFO structures, the nature of the underlying portfolio (set of preference and diversification constraint) appears on the vertical axis and their statistics within the particular CFO structure are presented horizontally. As a reference, the value of the debt portion of the CFO's \$400 M and the all-in funding cost, in basis points over the 1M LIBOR, appear at the top of each panel. The descriptive statistics are monthly and include the first four moments of the distributions, as well as the minimum and maximum returns. With regard to the tests for normality, the value of the test statistic and a dummy variable indicating whether or not the value is significant at a confidence level of 95% are presented*. The dummy variable equals 1 when the null hypothesis of normality is rejected. The performance measures (Sharpe, Treynor, Sortino, Modified Sharpe and Omega (Ω)) are also presented on a monthly basis. For the Sortino ratio and the Omega measure, the calculations are made for Minimal Accepted Return (MAR) levels of 0.8% and 1.2% monthly. The Modified Sharpe is computed based on a VaR with a 95% confidence level.

PANEL A

CFO 11 (Debt = \$200 M ; All-in funding cost = 150 bps over LIBOR 1 month)															
E_h	k	Mean	Variance	Skewness	Excess kurtosis	JB $\alpha=5\%$	Lilliefors $\alpha=5\%$	G&R $\alpha=5\%$	Sharpe	Treynor	Sortino MAR=0.008	Sortino MAR=0.012	Modified Sharpe $\alpha=5\%$	Ω MAR=0.008	Ω MAR=0.012
$E_1: \{\alpha=1; \beta=0; \gamma=0\}$	$y \leq 100\%$	0.0098	1.255E-04	-0.2858	-0.0136	0	0	0	0.6472	0.0602	0.9963	0.769	0.7509	1.5024	0.6018
	$y \leq 50\%$	0.0098	1.255E-04	-0.2858	-0.0136	0	0	0	0.6472	0.0602	0.9963	0.769	0.7509	1.5024	0.6018
	$y \leq 25\%$	0.0098	1.460E-04	-0.4064	0.0752	0	0	0	0.6017	0.0473	0.8977	0.713	0.6216	1.4668	0.6245
	$y \leq 20\%$	0.0098	1.667E-04	-0.4104	0.0762	0	0	0	0.5586	0.0391	0.8231	0.6628	0.5444	1.4090	0.6377
	$y \leq 15\%$	0.0100	1.817E-04	-0.4135	-0.0191	0	0	1	0.5524	0.0388	0.8173	0.6635	0.5292	1.4435	0.6826
	$y \leq 10\%$	0.0097	1.861E-04	-0.4443	-0.0123	0	1	1	0.5225	0.0377	0.7553	0.6162	0.4802	1.3567	0.6439
$E_2: \{\alpha=1; \beta=1; \gamma=0\}$	$y \leq 100\%$	0.0071	1.157E-04	0.4734	0.2580	0	0	0	0.4201	-0.5235	0.5943	0.4390	0.4786	0.8059	0.3329
	$y \leq 50\%$	0.0071	1.158E-04	0.4729	0.2559	0	0	0	0.4202	-0.5181	0.5946	0.4393	0.4788	0.8064	0.3333
	$y \leq 25\%$	0.0096	1.858E-04	0.0043	-0.2065	0	0	0	0.5179	0.0428	0.8107	0.6420	0.5497	1.3531	0.6461
	$y \leq 20\%$	0.0100	1.817E-04	-0.1852	-0.2746	0	0	0	0.5482	0.0467	0.8407	0.6739	0.5654	1.4338	0.6780
	$y \leq 15\%$	0.0095	1.808E-04	-0.1361	-0.4612	0	1	0	0.5194	0.1051	0.7842	0.6247	0.5270	1.3207	0.6303
	$y \leq 10\%$	0.0099	1.794E-04	-0.2355	-0.5394	0	0	1	0.5460	0.0909	0.8255	0.6596	0.5498	1.3960	0.6706
$E_3: \{\alpha=1; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0097	1.790E-04	-0.1204	-0.2906	0	0	0	0.5359	0.0497	0.8216	0.6545	0.5585	1.3785	0.6504
	$y \leq 50\%$	0.0097	1.790E-04	-0.1204	-0.2906	0	0	0	0.5359	0.0497	0.8216	0.6545	0.5585	1.3785	0.6504
	$y \leq 25\%$	0.0096	1.803E-04	-0.0757	-0.3115	0	0	0	0.5236	0.0531	0.8019	0.6374	0.5459	1.3396	0.6350
	$y \leq 20\%$	0.0097	1.803E-04	-0.1110	-0.3004	0	0	0	0.5329	0.0508	0.8170	0.6509	0.5547	1.3718	0.6497
	$y \leq 15\%$	0.0098	1.864E-04	-0.1286	-0.3860	0	0	0	0.5326	0.0609	0.8170	0.6528	0.5483	1.3868	0.6698
	$y \leq 10\%$	0.0099	1.835E-04	-0.2351	-0.4667	0	1	1	0.5448	0.0668	0.8258	0.6615	0.5476	1.4111	0.6826
$E_4: \{\alpha=2; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0097	1.791E-04	-0.1920	-0.1883	0	0	0	0.5366	0.0422	0.8139	0.6516	0.5473	1.3829	0.6489
	$y \leq 50\%$	0.0097	1.791E-04	-0.1920	-0.1883	0	0	0	0.5366	0.0422	0.8139	0.6516	0.5473	1.3829	0.6489
	$y \leq 25\%$	0.0097	1.775E-04	-0.1939	-0.1886	0	0	0	0.5393	0.0429	0.8185	0.6546	0.5517	1.3858	0.6481
	$y \leq 20\%$	0.0098	1.803E-04	-0.1847	-0.2076	0	0	0	0.5360	0.0431	0.8145	0.6519	0.5473	1.3840	0.6521
	$y \leq 15\%$	0.0100	1.870E-04	-0.1854	-0.3037	0	0	0	0.5406	0.0491	0.8283	0.6635	0.5513	1.4187	0.6849
	$y \leq 10\%$	0.0100	1.864E-04	-0.2818	-0.3614	0	1	1	0.5463	0.0531	0.8258	0.6639	0.5412	1.4294	0.6959
$E_5: \{\alpha=3; \beta=1; \gamma=0.25\}$	$y \leq 100\%$	0.0074	1.159E-04	0.3280	-0.0388	0	0	0	0.4488	0.2628	0.6398	0.4747	0.5098	0.8689	0.3538
	$y \leq 50\%$	0.0074	1.159E-04	0.3277	-0.0381	0	0	0	0.4496	0.2637	0.6412	0.4757	0.5111	0.8704	0.3543
	$y \leq 25\%$	0.0097	1.890E-04	0.0167	-0.1873	0	0	0	0.5225	0.0403	0.8251	0.6536	0.5587	1.3806	0.6639
	$y \leq 20\%$	0.0100	1.899E-04	-0.1505	-0.2202	0	0	0	0.5364	0.0404	0.8271	0.6637	0.5512	1.4256	0.6860
	$y \leq 15\%$	0.0100	1.872E-04	-0.2307	-0.2506	0	0	0	0.5430	0.0456	0.8271	0.6650	0.5470	1.4285	0.6887
	$y \leq 10\%$	0.0099	1.795E-04	-0.2604	-0.3925	0	1	1	0.5497	0.0705	0.8313	0.6654	0.5522	1.4137	0.6768
$E_6: EW (1/m)$	$y = 6.25\%$	0.0106	2.565E-04	-0.3987	-0.1921	0	1	1	0.5033	0.0345	0.7522	0.6283	0.4491	1.4854	0.8058

Table 8
Performance of CFOs 11, 15 and 18 (continued)

PANEL B

CFO 15 (Debt = \$280 M ; All-in funding cost = 200 bps over LIBOR 1 month)															
E_h	k	Mean	Variance	Skewness	Excess kurtosis	JB $\alpha=5\%$	Lilliefors $\alpha=5\%$	G&R $\alpha=5\%$	Sharpe	Treynor	Sortino MAR=0.008	Sortino MAR=0.012	Modified Sharpe $\alpha=5\%$	Ω MAR=0.008	Ω MAR=0.012
$E_1: \{\alpha=1; \beta=0; \gamma=0\}$	$y \leq 100\%$	0.0120	2.540E-04	-0.1768	-0.0826	0	0	0	0.5936	0.0568	0.9984	0.8212	0.6269	1.8739	1.0037
	$y \leq 50\%$	0.0120	2.540E-04	-0.1768	-0.0826	0	0	0	0.5936	0.0568	0.9984	0.8212	0.6269	1.8739	1.0037
	$y \leq 25\%$	0.0120	2.953E-04	-0.3433	0.0178	0	0	0	0.5520	0.0435	0.8855	0.7464	0.5221	1.7975	1.0072
	$y \leq 20\%$	0.0120	3.400E-04	-0.3646	0.0435	0	0	0	0.5104	0.0354	0.8034	0.6847	0.4571	1.6978	0.9963
	$y \leq 15\%$	0.0123	3.710E-04	-0.3883	-0.0460	0	1	1	0.5047	0.0351	0.7930	0.6801	0.4430	1.7162	1.0374
	$y \leq 10\%$	0.0119	3.789E-04	-0.4098	-0.0714	0	1	1	0.4779	0.0342	0.7347	0.6319	0.4063	1.6207	0.9830
$E_2: \{\alpha=1; \beta=1; \gamma=0\}$	$y \leq 100\%$	0.0082	2.646E-04	0.6583	0.6494	1	0	0	0.3473	-0.2364	0.5518	0.4412	0.3476	1.0343	0.5617
	$y \leq 50\%$	0.0082	2.647E-04	0.6577	0.6468	1	0	1	0.3475	-0.2350	0.5522	0.4416	0.3478	1.0349	0.5621
	$y \leq 25\%$	0.0118	3.824E-04	0.0606	-0.3049	0	0	0	0.4703	0.0390	0.7901	0.6657	0.4552	1.6278	0.9696
	$y \leq 20\%$	0.0122	3.666E-04	-0.1560	-0.4277	0	0	0	0.5035	0.0431	0.8248	0.6991	0.4732	1.7107	1.0270
	$y \leq 15\%$	0.0117	3.664E-04	-0.0863	-0.5361	0	1	1	0.4762	0.1154	0.7742	0.6517	0.4446	1.5821	0.9598
	$y \leq 10\%$	0.0121	3.595E-04	-0.1839	-0.6645	0	1	1	0.5042	0.0957	0.8214	0.6914	0.4686	1.6662	1.0157
$E_3: \{\alpha=1; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0119	3.594E-04	-0.0623	-0.4081	0	0	0	0.4935	0.0466	0.8147	0.6855	0.4733	1.6559	0.9894
	$y \leq 50\%$	0.0119	3.594E-04	-0.0623	-0.4081	0	0	0	0.4935	0.0466	0.8147	0.6855	0.4733	1.6559	0.9894
	$y \leq 25\%$	0.0117	3.626E-04	-0.0169	-0.4277	0	0	0	0.4814	0.0503	0.7952	0.6676	0.4617	1.6103	0.9655
	$y \leq 20\%$	0.0119	3.620E-04	-0.0530	-0.4179	0	0	0	0.4907	0.0479	0.8103	0.6816	0.4702	1.6481	0.9868
	$y \leq 15\%$	0.0121	3.736E-04	-0.0663	-0.4851	0	0	0	0.4914	0.0593	0.8123	0.6841	0.4679	1.6557	1.0078
	$y \leq 10\%$	0.0122	3.663E-04	-0.1794	-0.5968	0	1	1	0.5041	0.0660	0.8225	0.6942	0.4691	1.6829	1.0269
$E_4: \{\alpha=2; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0119	3.607E-04	-0.1434	-0.3167	0	0	0	0.4932	0.0387	0.8023	0.6786	0.4623	1.6623	0.9907
	$y \leq 50\%$	0.0119	3.607E-04	-0.1434	-0.3167	0	0	0	0.4932	0.0387	0.8023	0.6786	0.4623	1.6623	0.9907
	$y \leq 25\%$	0.0119	3.570E-04	-0.1445	-0.3204	0	0	0	0.4959	0.0394	0.8078	0.6825	0.4662	1.6675	0.9913
	$y \leq 20\%$	0.0120	3.624E-04	-0.1323	-0.3354	0	0	0	0.4932	0.0396	0.8045	0.6799	0.4636	1.6622	0.9938
	$y \leq 15\%$	0.0122	3.729E-04	-0.1244	-0.4274	0	0	0	0.5001	0.0462	0.8244	0.6962	0.4725	1.6962	1.0282
	$y \leq 10\%$	0.0123	3.709E-04	-0.2267	-0.4916	0	1	1	0.5063	0.0506	0.8229	0.6968	0.4658	1.7089	1.0403
$E_5: \{\alpha=3; \beta=1; \gamma=0.25\}$	$y \leq 100\%$	0.0087	2.595E-04	0.4900	0.2461	0	0	0	0.3784	0.3634	0.6023	0.4826	0.3799	1.1096	0.5994
	$y \leq 50\%$	0.0087	2.594E-04	0.4902	0.2475	0	0	0	0.3792	0.3658	0.6039	0.4838	0.3810	1.1115	0.6002
	$y \leq 25\%$	0.0119	3.888E-04	0.0769	-0.2680	0	0	0	0.4750	0.0365	0.8048	0.6781	0.4637	1.6552	0.9910
	$y \leq 20\%$	0.0122	3.844E-04	-0.1036	-0.3333	0	0	0	0.4922	0.0369	0.8116	0.6889	0.4644	1.7013	1.0274
	$y \leq 15\%$	0.0123	3.732E-04	-0.1756	-0.3896	0	0	0	0.5025	0.0425	0.8217	0.6959	0.4688	1.7050	1.0348
	$y \leq 10\%$	0.0122	3.569E-04	-0.1913	-0.5283	0	1	1	0.5097	0.0705	0.8319	0.7010	0.4765	1.6917	1.0250
$E_6: EW (1/m)$	$y = 6.25\%$	0.0131	5.175E-04	-0.3569	-0.1135	0	1	1	0.4648	0.0315	0.7350	0.6429	0.3909	1.7174	1.1298

Table 8
Performance of CFOs 11, 15 and 18 (continued)
 PANEL C

CFO 18 (Debt = \$340 M ; All-in funding cost = 300 bps over LIBOR 1 month)															
E_h	k	Mean	Variance	Skewness	Excess kurtosis	JB $\alpha=5\%$	Lilliefors $\alpha=5\%$	G&R $\alpha=5\%$	Sharpe	Treynor	Sortino MAR=0.008	Sortino MAR=0.012	Modified Sharpe $\alpha=5\%$	Ω MAR=0.008	Ω MAR=0.012
$E_1: \{\alpha=1; \beta=0; \gamma=0\}$	$y \leq 100\%$	0.0149	6.779E-04	-0.0187	0.0959	0	0	0	0.4749	0.0474	0.8314	0.7326	0.4418	1.9585	1.3345
	$y \leq 50\%$	0.0149	6.779E-04	-0.0187	0.0959	0	0	0	0.4749	0.0474	0.8314	0.7326	0.4418	1.9585	1.3345
	$y \leq 25\%$	0.015	7.978E-04	-0.2632	0.2770	0	0	0	0.4394	0.0348	0.7237	0.6497	0.3682	1.8608	1.3082
	$y \leq 20\%$	0.0149	9.408E-04	-0.3193	0.3499	0	0	0	0.4018	0.0276	0.6445	0.5839	0.3196	1.7455	1.2680
	$y \leq 15\%$	0.0154	1.038E-03	-0.3688	0.3002	0	0	0	0.3974	0.0275	0.6325	0.5759	0.3092	1.7514	1.2960
	$y \leq 10\%$	0.0147	1.047E-03	-0.3740	0.1657	0	1	0	0.3764	0.0267	0.5891	0.5366	0.2870	1.6642	1.2325
$E_2: \{\alpha=1; \beta=1; \gamma=0\}$	$y \leq 100\%$	0.0087	8.573E-04	0.8688	1.2782	1	1	1	0.2080	-0.0990	0.3433	0.3004	0.1762	1.0600	0.7474
	$y \leq 50\%$	0.0087	8.573E-04	0.8682	1.2748	1	1	1	0.2082	-0.0986	0.3438	0.3008	0.1764	1.0607	0.7479
	$y \leq 25\%$	0.0146	1.092E-03	0.1824	-0.0811	0	0	0	0.3632	0.0306	0.6259	0.5633	0.3135	1.6629	1.2187
	$y \leq 20\%$	0.0152	1.004E-03	-0.0837	-0.3054	0	0	0	0.3996	0.0347	0.6682	0.6024	0.3343	1.7538	1.2895
	$y \leq 15\%$	0.0145	9.800E-04	0.0330	-0.4012	0	0	0	0.3799	0.1286	0.6379	0.5707	0.3214	1.6425	1.2108
	$y \leq 10\%$	0.0151	9.415E-04	-0.0809	-0.6626	0	1	1	0.4096	0.0978	0.6893	0.6157	0.3444	1.7311	1.2769
$E_3: \{\alpha=1; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0148	9.627E-04	0.0564	-0.3723	0	0	0	0.3943	0.0385	0.6726	0.6013	0.3400	1.7103	1.2484
	$y \leq 50\%$	0.0148	9.627E-04	0.0564	-0.3723	0	0	0	0.3943	0.0385	0.6726	0.6013	0.3400	1.7103	1.2484
	$y \leq 25\%$	0.0145	9.726E-04	0.1058	-0.4254	0	0	0	0.3832	0.0421	0.6553	0.5846	0.3304	1.6637	1.2176
	$y \leq 20\%$	0.0148	9.691E-04	0.0682	-0.3887	0	0	0	0.3921	0.0397	0.6693	0.5981	0.3381	1.7027	1.2436
	$y \leq 15\%$	0.0150	9.911E-04	0.0678	-0.4240	0	0	1	0.3960	0.0523	0.6781	0.6059	0.3421	1.7125	1.2634
	$y \leq 10\%$	0.0153	9.583E-04	-0.0742	-0.6260	0	0	0	0.4101	0.0600	0.6915	0.6187	0.3456	1.7465	1.2889
$E_4: \{\alpha=2; \beta=1; \gamma=0.75\}$	$y \leq 100\%$	0.0148	9.786E-04	-0.0544	-0.2381	0	0	0	0.3916	0.0307	0.6538	0.5876	0.3285	1.7100	1.2484
	$y \leq 50\%$	0.0148	9.786E-04	-0.0544	-0.2381	0	0	0	0.3916	0.0307	0.6538	0.5876	0.3285	1.7100	1.2484
	$y \leq 25\%$	0.0148	9.663E-04	-0.0537	-0.2516	0	0	0	0.3943	0.0313	0.6595	0.5922	0.3316	1.7162	1.2506
	$y \leq 20\%$	0.0149	9.791E-04	-0.0351	-0.2727	0	0	0	0.3927	0.0316	0.6586	0.5912	0.3311	1.7112	1.2508
	$y \leq 15\%$	0.0153	9.873E-04	-0.0033	-0.3939	0	1	0	0.4042	0.0385	0.6891	0.6171	0.3461	1.7537	1.2874
	$y \leq 10\%$	0.0154	9.708E-04	-0.1297	-0.5004	0	0	0	0.4125	0.0432	0.6923	0.6210	0.3439	1.7710	1.3017
$E_5: \{\alpha=3; \beta=1; \gamma=0.25\}$	$y \leq 100\%$	0.0094	8.231E-04	0.7112	0.8357	1	0	0	0.2395	0.7232	0.3967	0.3468	0.2053	1.1371	0.7980
	$y \leq 50\%$	0.0095	8.222E-04	0.7120	0.8381	1	0	0	0.2403	0.7433	0.3984	0.3482	0.2061	1.1393	0.7993
	$y \leq 25\%$	0.0148	1.111E-03	0.1996	-0.0070	0	0	0	0.3680	0.0284	0.6386	0.5749	0.3203	1.6894	1.2410
	$y \leq 20\%$	0.0153	1.059E-03	-0.0150	-0.1675	0	0	0	0.3901	0.0292	0.6576	0.5935	0.3295	1.7425	1.2836
	$y \leq 15\%$	0.0153	9.901E-04	-0.0701	-0.3556	0	0	0	0.4061	0.0349	0.6845	0.6145	0.3426	1.7594	1.2945
	$y \leq 10\%$	0.0152	9.250E-04	-0.0578	-0.5521	0	0	1	0.4165	0.0654	0.7057	0.6298	0.3553	1.7602	1.2916
$E_6: EW (1/m)$	$y = 6.25\%$	0.0167	1.412E-03	-0.3207	0.3804	0	0	0	0.3771	0.0255	0.6070	0.5588	0.2906	1.7599	1.3623

* P-values are available upon request.

Table 9
Results of Multivariate Linear Regressions for CFOs 1, 11, 15 and 18

The table shows the results of the multivariate linear regression model applied to the returns of CFO structures 1, 11, 15 and 18. For every structure, the value of the coefficient respective to each independent variable and their corresponding p-value are presented. The coefficient δ is the model's constant. Within each panel (A to D), the independent variables appear on the vertical axis and the underlying funds (set of preference and diversification constraint) are presented horizontally. The shaded values indicate a *p-value* below 5%.

PANEL A							
CFO 1 (Direct exposure to Portfolio of HF)							
coefficients							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.025818	0.021177	0.024543	0.025677	0.020373	0.026133	0.026712
STOCK	0.061247	-0.008493	0.082277	0.077032	0.006984	0.087423	0.129787
DEF	-0.000675	-0.000480	-0.000671	-0.000603	-0.000451	-0.000659	-0.000639
LIQU	-21.324371	-20.243579	-11.967612	-21.047440	-20.423683	-17.897796	-16.064178
TERM	-2.305055	-2.320236	-2.013981	-2.712104	-2.214730	-2.580198	-2.646475
TREND	-0.000144	-0.000127	-0.000122	-0.000143	-0.000116	-0.000142	-0.000151
p-values							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.0000	0.0002	0.0005	0.0004	0.0005	0.0002	0.0005
STOCK	0.0016	0.6536	0.0006	0.0018	0.7177	0.0003	0.0000
DEF	0.0097	0.0649	0.0353	0.0671	0.0889	0.0398	0.0677
LIQU	0.3550	0.3830	0.6723	0.4734	0.3886	0.5297	0.6060
TERM	0.0511	0.0512	0.1633	0.0713	0.0678	0.0770	0.0969
TREND	0.0014	0.0049	0.0249	0.0119	0.0112	0.0098	0.0123

PANEL B							
CFO 11							
coefficients							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.037932	0.031745	0.035875	0.037632	0.030135	0.038319	0.038088
STOCK	0.098863	-0.022917	0.134674	0.122659	0.004186	0.142237	0.217518
DEF	-0.001056	-0.000803	-0.001073	-0.000942	-0.000749	-0.001033	-0.000944
LIQU	-49.959631	-46.930344	-35.537299	-49.574431	-48.093856	-45.137863	-41.680773
TERM	-3.254628	-3.444272	-2.654086	-3.858305	-3.212852	-3.641032	-3.606458
TREND	-0.000253	-0.000230	-0.000217	-0.000250	-0.000211	-0.000249	-0.000257
p-values							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.0000	0.0013	0.0013	0.0011	0.0024	0.0006	0.0017
STOCK	0.0015	0.4848	0.0005	0.0019	0.8991	0.0002	0.0000
DEF	0.0119	0.0744	0.0353	0.0738	0.0986	0.0429	0.0884
LIQU	0.1794	0.2437	0.4326	0.2925	0.2362	0.3205	0.3996
TERM	0.0861	0.0935	0.2495	0.1082	0.1203	0.1162	0.1529
TREND	0.0005	0.0032	0.0131	0.0061	0.0073	0.0047	0.0073

PANEL C							
CFO 15							
coefficients							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.050856	0.044229	0.048016	0.050308	0.041476	0.051277	0.049340
STOCK	0.138615	-0.044203	0.191964	0.169802	-0.003430	0.201255	0.316547
DEF	-0.001428	-0.001170	-0.001481	-0.001271	-0.001080	-0.001401	-0.001199
LIQU	-81.254411	-76.995044	-62.397691	-80.775945	-79.772810	-75.565411	-69.526505
TERM	-4.251777	-4.744886	-3.253458	-5.024286	-4.336980	-4.723197	-4.471656
TREND	-0.000370	-0.000350	-0.000320	-0.000366	-0.000318	-0.000364	-0.000368
p-values							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.0001	0.0028	0.0022	0.0019	0.0048	0.0011	0.0037
STOCK	0.0017	0.3712	0.0004	0.0023	0.9445	0.0002	0.0000
DEF	0.0163	0.0841	0.0401	0.0868	0.1102	0.0507	0.1235
LIQU	0.1245	0.2043	0.3311	0.2243	0.1885	0.2378	0.3185
TERM	0.1135	0.1243	0.3183	0.1375	0.1594	0.1470	0.2073
TREND	0.0004	0.0029	0.0099	0.0045	0.0066	0.0034	0.0064

PANEL D							
CFO 18							
coefficients							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.074028	0.070676	0.069972	0.072836	0.064612	0.074630	0.066990
STOCK	0.218948	-0.099010	0.313963	0.262845	-0.027730	0.324006	0.534591
DEF	-0.002167	-0.002082	-0.002329	-0.001924	-0.001884	-0.002143	-0.001600
LIQU	-146.250322	-145.031986	-121.184595	-145.829989	-152.364230	-140.468397	-126.841879
TERM	-6.021820	-7.488111	-4.120773	-7.066956	-6.627699	-6.635288	-5.641108
TREND	-0.000590	-0.000613	-0.000512	-0.000579	-0.000546	-0.000579	-0.000563
p-values							
	{ $\alpha=1; \beta=0; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 100\%$	{ $\alpha=1; \beta=1; \gamma=0$ $y \leq 20\%$	{ $\alpha=2; \beta=1; \gamma=0.75$ $y \leq 10\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 50\%$	{ $\alpha=3; \beta=1; \gamma=0.25$ $y \leq 15\%$	1/m $y = 6.25\%$
δ	0.0005	0.0075	0.0068	0.0056	0.0133	0.0034	0.0158
STOCK	0.0026	0.2660	0.0005	0.0037	0.7524	0.0003	0.0000
DEF	0.0264	0.0873	0.0514	0.1113	0.1178	0.0671	0.2110
LIQU	0.0934	0.1838	0.2559	0.1786	0.1592	0.1801	0.2700
TERM	0.1723	0.1766	0.4457	0.1990	0.2269	0.2119	0.3334
TREND	0.0005	0.0037	0.0127	0.0057	0.0088	0.0043	0.0112