

Hedge Fund Portfolio Management with Illiquid Assets

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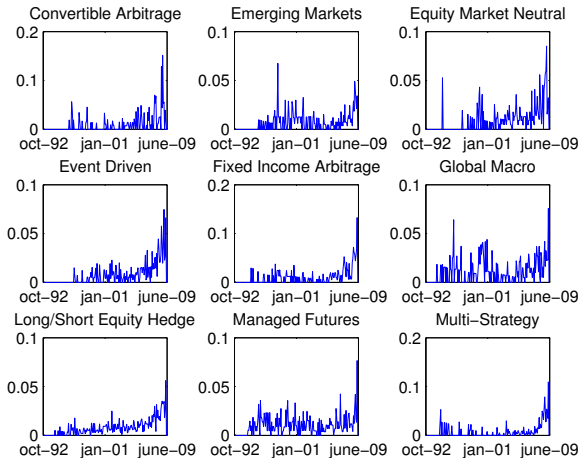
Contents

- 1 Economic Motivation
- 2 The model
 - Assumptions and notations
 - Solving the model
 - Model implications
- 3 Empirical application
 - The data
 - The estimation technique
- 4 Conclusion and further developments

Hedge funds liquidation rates

- The majority of Hedge funds have rather short lifetimes (compared to mutual funds)
- The liquidation rates, as well as the dependence among them, vary considerably according to:
 - The period of time
 - The management style

Frequency Counts of Liquidated Hedge funds



What are the drivers of Hedge fund liquidations ?

- Individual drivers
 - Illegal operations or frauds
 - Insufficient incentives for the Hedge fund manager
 - Poor performance: the Hedge fund is not attractive for investors
- Common drivers
 - Liquidity Crises
- How can we use liquidation date to infer fund exposures to liquidity risk ?

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 - **Liquidity Crises**
- How can we use liquidation date to infer fund exposures to liquidity risk ?

With Hedge funds data ...

- Large sample of different Hedge fund strategies ...
- ... with different exposures to market liquidity (market portfolios)
- ... with different exposures to funding liquidity (leverage, redemption policies)
- And then different liquidity mismatches !

Hedge funds and exposures to liquidity

i) Examples of funding liquidity shocks

- Large cash withdrawals of direct investors during **funding liquidity crisis**
- Deleveraging imposed by prime brokers
- Outflow of funds of funds investors

Effect by means of the **liability component of the balance sheet**

Hedge funds and exposures to liquidity

ii) Example of market liquidity shocks

Fund 1 is liquidated and sells illiquid asset A

→ "market price" of asset A decreases

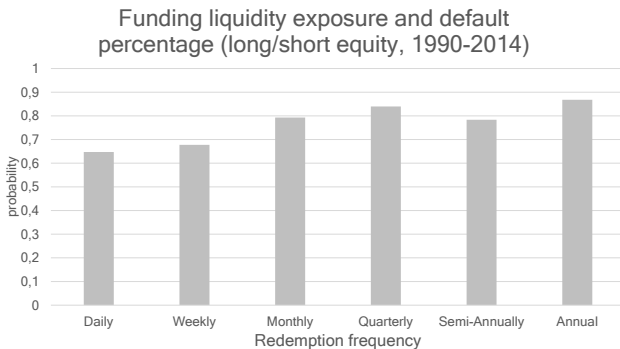
→ NAV of **Fund 2** invested in asset A decreases

→ likelihood of selling asset A by of **Fund 2** increases

Effect by means of the **asset component of the balance sheet**

Hedge funds and exposures to liquidity

- High level of *liquidity transformation*
- However, great resilience to depressed market liquidity and fire-sales during the *Financial Crisis* ...



Previous studies (with return data)

- Previous empirical studies have shown that both market **and** funding liquidity are priced in the cross-section of fund's returns.

Market: SADKA 2006-2010, TEO 2011.

Funding: DUDLEY & NIMALENDRAN 2010, ARAGON & STRAHAN 2012.

Both: AGARWAL, ARAGON & SHI 2015.

- They mostly use regression evidence using market and funding liquidity indicators on the right-hand side.
- HOMBERT & THESMAR 2014 endogenize the funding liquidity exposure via constraints on withdrawals.

Endogeneity issue

The fund manager adjusts his portfolio [cash holdings] assessing future market and funding liquidity conditions.

In this paper

- We propose a two-period structural model of a single hedge fund portfolio management when it is subject to both funding and market liquidity shocks. [closest to ours is LIU & MELLO 2011]
- The funding liquidity shock is a random quantity of AUM being withdrawn, whereas the market liquidity shock is a random haircut applied on the fund's risky assets.
- The model produces closed-form solution for the fund's optimal cash holdings as a function of market and funding liquidity conditions, allowing to derive testable implications on hedge fund's data.

Results

- Our model produces endogenous default probabilities, default arising as the conjunction of big funding and market liquidity shocks together.
- The fund's optimal cash amount allows it to hedge partly against bad funding and market liquidity conditions ([Liquidity timing](#)).
- The fund's optimal cash amount allows it to mitigate the effect of market liquidity shocks when the market liquidity is very high and very low ([Liquidity resilience](#)).
- We confirm previous results showing that the fund's value is increasing in the funding liquidity conditions ([Mismatch reduction](#)).
- Taking the model to the data is easy and produces meaningful results with a new mismatch factor(*ongoing*).

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Fundamental assumptions

Consider a single hedge fund entity living for 2 periods:

- risk-neutral,
- price-taker,
- has a unit amount of assets under management,

which has access to two assets:

- **Cash:** zero interest rate, available every period;
- **Illiquid asset:** positive deterministic interest rate $\rho_1 + \rho_2$, available at period $t = 2$ only.

The HF balance-sheet is:

| Assets | Liabilities |
|----------------|-------------|
| Cash | AUM |
| Illiquid asset | |

Timing

$t = 0$ → The fund chooses a portfolio composition $(\delta, 1 - \delta)$ in cash/illiquid asset.

$t = 1$ → With probability π , a fraction θ of AUM is withdrawn by the investors.

- $\theta \sim \mathcal{U}[0, \bar{\theta}]$, where $\bar{\theta} \leq 1$.
- If $\theta \leq \delta$, the fund takes its cash to pay the investors.
- If $\theta > \delta$, the fund has to sell the illiquid asset on a secondary market.

$t = 2$ → The final value of the fund's portfolio is realized.

Funding liquidity shock

The value of θ is the funding liquidity shock. It is associated with parameters π and $\bar{\theta}$.

Liquidation cost on the secondary market

Remember that the illiquid asset has a known return $\rho_1 + \rho_2$ at period $t = 2$.

- ρ_1 alone would be the friction-less return at period $t = 1$.

Market liquidity shock

The realized return of the illiquid asset on the secondary market is equal to $\rho_1 - T$, where T is the market liquidity shock.

- $T \sim \mathcal{E}(\lambda)$, where $\lambda > 0$.
- λ is an indicator of market liquidity, $1/\lambda$ is the average rebate on the secondary market.

When does default arise?

Two situations can arise when a "big" θ is observed:

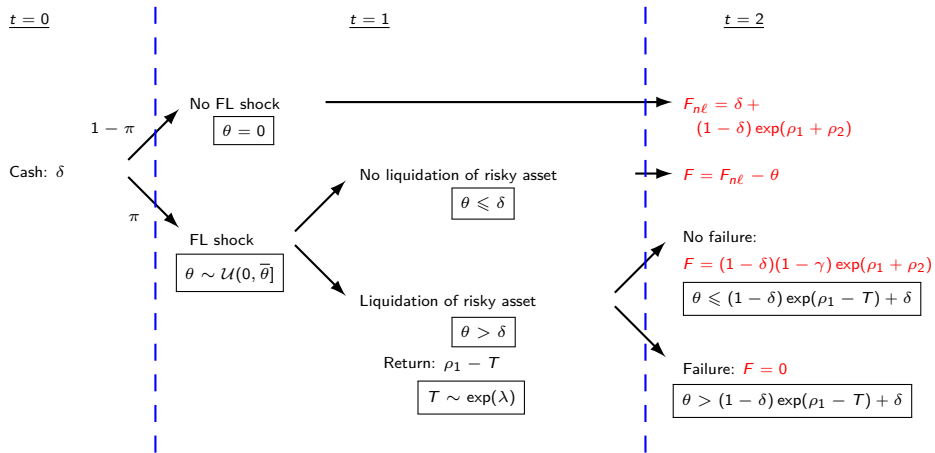
- The realized return $\rho_1 - T$ is sufficiently high to cover the discrepancy $\theta - \delta$. I sell a proportion γ of the illiquid asset equal to:

$$\gamma(1 - \delta) \exp(\rho_1 - T) = \theta - \delta \iff \gamma = \frac{\theta - \delta}{1 - \delta} \exp(T - \rho_1)$$

- If T is too high, selling all the illiquid asset is not sufficient to cover $\theta - \delta$. The default is triggered whenever:

$$(1 - \delta) \exp(\rho_1 - T) < \theta - \delta \iff \gamma > 1$$

Summary of the setup



The fund's maximization problem

- The fund is looking for the optimal quantity δ to maximize its expected portfolio value $\mathbb{E}(F)$ at period $t = 2$ from $t = 0$ perspective.

$$\delta^* = \operatorname{argmax}_{\delta} \mathbb{E}[F(\delta, \bar{\theta}, \lambda, \pi, \rho_1, \rho_2)] \quad \text{s.t.} \quad \delta \in [0, 1].$$

- The expected portfolio value is decomposed with the law of iterated expectations and needs the conditional distribution of γ given a big funding liquidity shock.

Default probability with exogenous cash

- Using the properties of the Uniform and the Exponential distributions, it is possible to obtain the conditional distribution of γ given a big liquidity shock.
- The unconditional default probability is a by-product:

Default probability

$$\mathbb{P}(\gamma > 1) = \frac{\pi(\bar{\theta} - \delta)^{\lambda+1}}{\bar{\theta}(\lambda + 1)(1 - \delta)^\lambda} e^{-\lambda\rho_1}$$

- This probability is decreasing in the market liquidity λ , increasing in the size of the maximal possible funding shock $\bar{\theta}$ and decreasing in the cash amount δ .

The expected fund's portfolio value

- The expected value of the fund's portfolio from $t = 0$ perspective is available in closed-form:

$$\begin{aligned}
 \mathbb{E}(F) &= \overbrace{(1 - \pi) [\delta + (1 - \delta) \exp(\rho_1 + \rho_2)]}^{\text{no liquidity shock}} + \overbrace{\pi \frac{\delta}{\bar{\theta}} \left((1 - \delta) \exp(\rho_1 + \rho_2) + \frac{\delta}{2} \right)}^{\text{small liquidity shock}} \\
 &+ (1 - \delta) \exp(\rho_1 + \rho_2) \pi \frac{\bar{\theta} - \delta}{\bar{\theta}} \underbrace{\left[1 - \frac{1}{\lambda + 1} \left(\frac{\bar{\theta} - \delta}{(1 - \delta) \exp(\rho_1)} \right)^\lambda \right]}_{\text{conditional survival probability}} \\
 &\times \underbrace{\left\{ 1 - \frac{\lambda}{1 - \lambda^2} \frac{\bar{\theta} - \delta}{(1 - \delta) \exp(\rho_1)} \left[\left(\frac{\bar{\theta} - \delta}{(1 - \delta) \exp(\rho_1)} \right)^{\lambda - 1} - \frac{\lambda + 1}{2} \right] \right\}}_{\text{conditional expectation of } 1 - \gamma}.
 \end{aligned}$$

A specific case: $\bar{\theta} = 1$.

When $\bar{\theta} = 1$, the expectation is simplified:

$$\begin{aligned} \mathbb{E}(F) &= (1 - \pi) \left[e^{\rho_1 + \rho_2} + \delta(1 - e^{\rho_1 + \rho_2}) \right] + \pi \left(\delta e^{\rho_1 + \rho_2} + \delta^2 \left(\frac{1}{2} - e^{\rho_1 + \rho_2} \right) \right) \\ &+ (1 - \delta)^2 e^{\rho_1 + \rho_2} \pi \left[1 - \frac{e^{-\lambda \rho_1}}{\lambda + 1} \right] \left\{ 1 - \frac{\lambda e^{-\lambda \rho_1}}{1 - \lambda^2} + \frac{\lambda e^{-\rho_1}}{2(1 - \lambda)} \right\}. \end{aligned}$$

which is **quadratic in δ** .

Model solutions

- When $\bar{\theta} = 1$, the problem can be solved in closed-form.
- When $\bar{\theta} < 1$, the problem has to be solved numerically.

Model solution when $\bar{\theta} = 1$

Optimal cash amount

$$\delta^*(1, \lambda, \pi, \rho_1, \rho_2) = \frac{G(\lambda, \rho_1) + H_1(\pi, \rho_1, \rho_2)}{G(\lambda, \rho_1) + H_2(\rho_1, \rho_2)} \quad \text{if this ratio is in } (0, 1),$$

with

$$G(\lambda, \rho_1) = \left[1 - \frac{e^{-\lambda\rho_1}}{\lambda + 1} \right] \left[1 - \frac{\lambda e^{-\lambda\rho_1}}{1 - \lambda^2} + \frac{\lambda e^{-\rho_1}}{2(1 - \lambda)} \right]$$

$$H_1(\pi, \rho_1, \rho_2) = \left(\frac{1}{\pi} - 1 \right) \frac{1 - e^{-\rho_1 - \rho_2}}{2} - \frac{1}{2}, \quad H_2(\rho_1, \rho_2) = \frac{e^{-\rho_1 - \rho_2}}{2} - 1.$$

- δ^* is a function of λ through $G(\cdot)$ only.
- δ^* is a function of π through $H_1(\cdot)$ only.

\implies this makes the influence of each parameter easier to calculate.

Properties of the optimal cash holding

The following properties hold when $\bar{\theta} = 1$:

Important results (1/3): liquidity timing

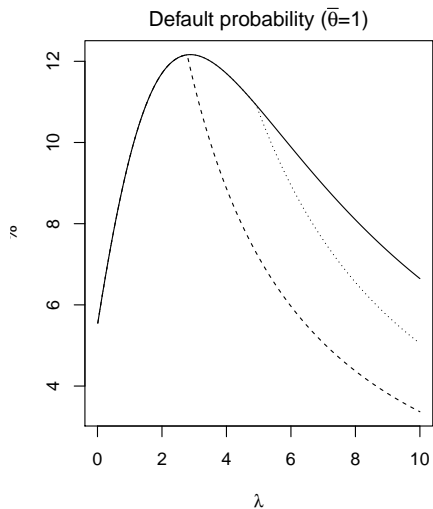
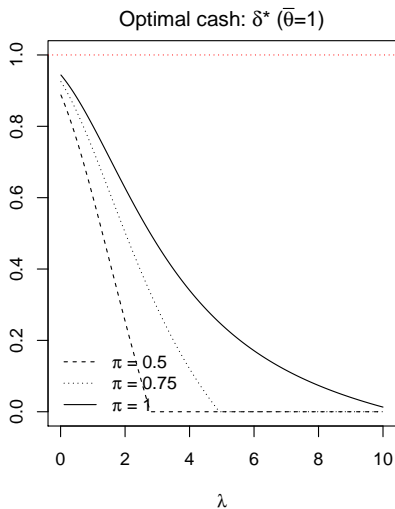
- δ^* is increasing with the probability of a funding liquidity shock π .
- δ^* is decreasing with the degree of market liquidity λ .

This allows us to derive the default probability when cash holdings are at the optimal value:

Important results (2/3): liquidity resilience

- The PD is insensitive to the probability of a funding liquidity shock π .
- The PD is a bell-shaped function of the degree of market liquidity λ .

Graphical illustration

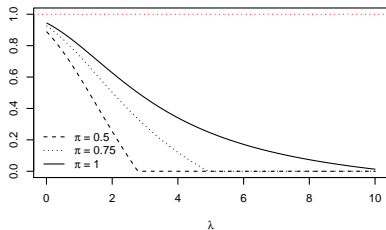
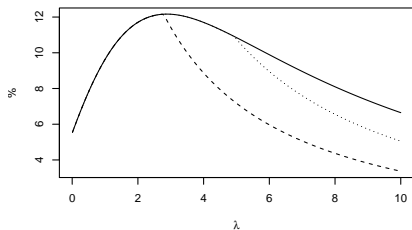
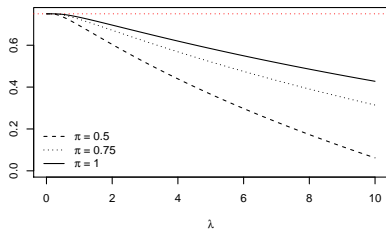
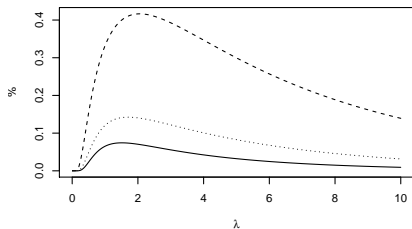


Model implications when $\bar{\theta} < 1$

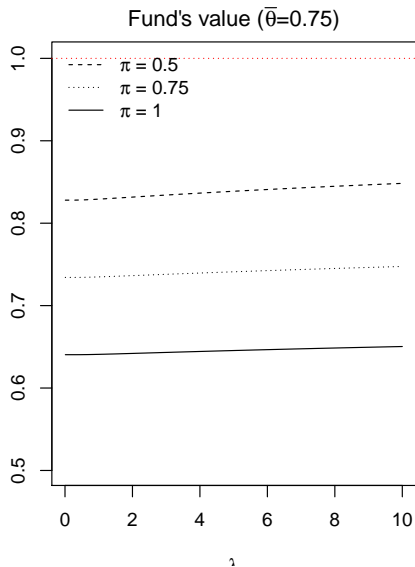
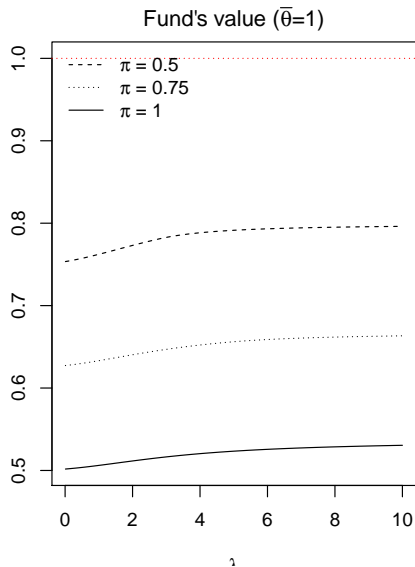
Important results (3/3): mismatch reduction

- All previous results still apply when $\bar{\theta} < 1$ but the PD is now decreasing with π .
- The expected portfolio value is decreasing in $\bar{\theta}$. [HOMBERT & THESMAR 2014]

Graphical illustration (1/2)

Optimal cash: δ^* ($\bar{\theta}=1$)Default probability ($\bar{\theta}=1$)Optimal cash: δ^* ($\bar{\theta}=0.75$)Default probability ($\bar{\theta}=0.75$)

Graphical illustration (2/2)



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Description of the data

We use the Lipper TASS database from 2000 to 2014 where we select the intersection of:

- Funds whose NAV is in USD, with monthly frequency.
- Funds reporting their AUM monthly.
- Funds that are not FoF or multiple management classes.

⇒ We concentrate in the current application on the management style *long/short equity*, though we have 8 others.

Data series

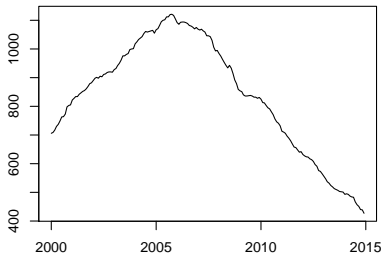
Default: $d_{j,t} = 1$ if j^{th} fund's NAV is not reported at t whereas it was reported before.

Flows: $f_{j,t}$ are the growth rate of AUM minus the performance.

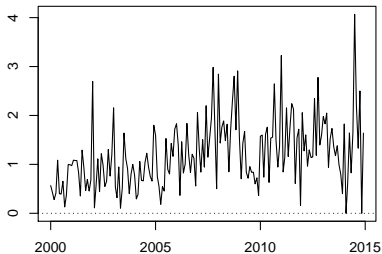
Performance: Growth rate of the NAV.

The data (2000-2014)

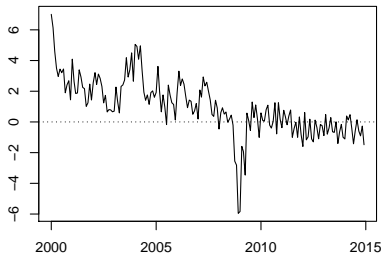
Panel (a): number of alive funds



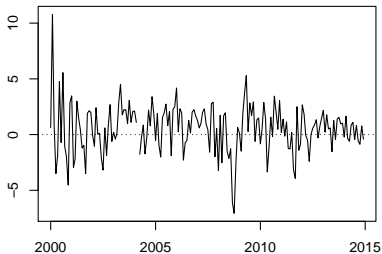
Panel (b): default probability



Panel (c): average inflows/outflows



Panel (d): average performance



Estimated quantities

We parameterize the following quantities:

- $\rho_t = \bar{\rho}$ is the average performance of the funds not experiencing outflows (cross-section avg first, time-series avg then).
- π_t is the cross-section average of number of funds experiencing outflows at period t .
- $\bar{\theta}_t$, λ_t and δ_t^* are estimated with a GMM-type estimator.

Cost functions

- $\bar{\theta}_t$ is bigger and as close as possible to the maximum observed outflow across funds.
- λ_t is positive and minimizes the distance between the model-implied and the measured default probability.

Formalization

- n_t is the number of alive funds at t , $f_{j,t}$ is the observed flows of fund j at t .
- The cost function of $\bar{\theta}_t$ is given by:

$$m_1 = \left(\bar{\theta}_t - \max_{j \in \{1, \dots, n_t\}} (-f_{j,t}) \right)^2 \quad \text{with} \quad \bar{\theta}_t \geq \max_{j \in \{1, \dots, n_t\}} (-f_{j,t}).$$

- The cost function of λ_t is given by:

$$m_2 = \left(\frac{\pi_t [\bar{\theta}_t - \delta^*(\bar{\theta}_t, \lambda_t, \pi_t, \bar{\rho})]^{\lambda_t+1}}{\bar{\theta}_t(\lambda_t + 1) [1 - \delta^*(\bar{\theta}_t, \lambda_t, \pi_t, \bar{\rho})]^{\lambda_t}} \exp(-\lambda_t \bar{\rho}) - \frac{1}{n_{t-1}} \sum_{j=1}^{n_{t-1}} d_{j,t} \right)^2.$$

Algorithm: for each t , pick $(\bar{\theta}_t, \lambda_t)$, maximize the expected fund value to get δ^* , compute $\mathcal{C} = \omega m_1 + (1 - \omega) m_2$ and minimize.

Products of estimation

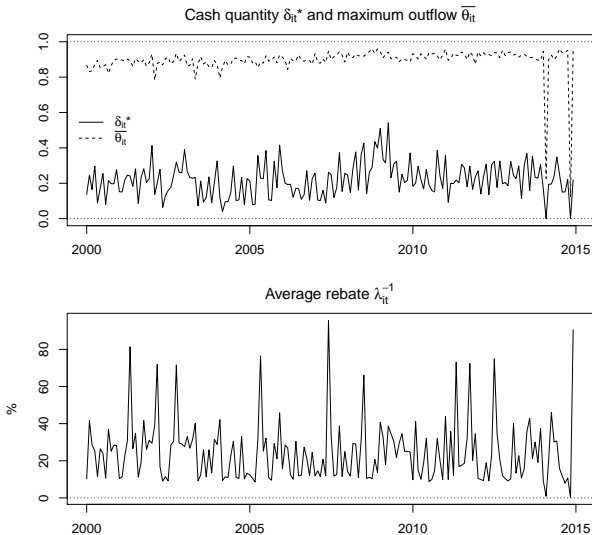
With the estimation, we get:

- A time-series for the maximum funding liquidity shock $\bar{\theta}_t$,
- A time-series for the optimal cash-holdings δ_t^* ,
- A time-series for the average market liquidity λ_t ,
- A risk factor taking both funding and market liquidity risks into account, which is computed as:

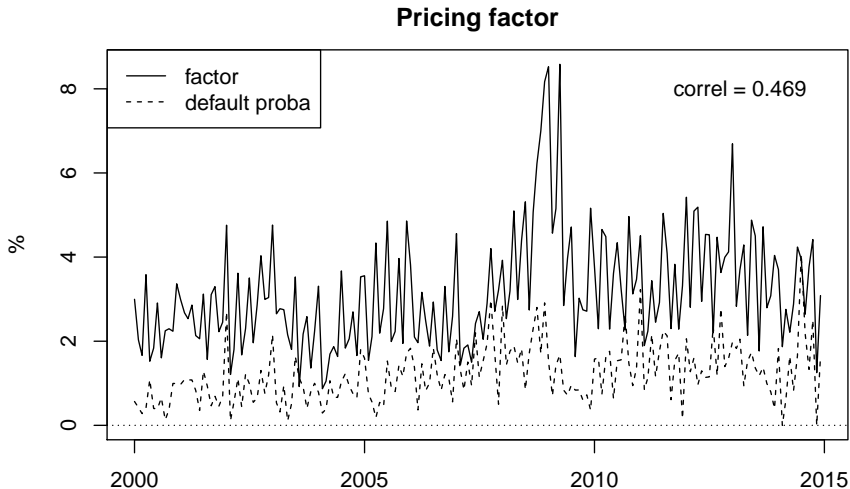
$$f_t = \bar{\rho} - \frac{1}{2} \log \mathbb{E} [F(\delta_t^*, \bar{\theta}_t, \lambda_t, \bar{\rho})]$$

f_t is homogeneous to a one-period net return.

Long/short equity results (1/2)



Long/short equity results (2/2)



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Further empirical developments

- Use the factor in asset pricing regression to show that both market and funding liquidity shocks are priced in the hedge fund returns.
- Develop the same exercise on the different management style we possess.

What we have done:

- We developed a simple structural model to relate fund's optimal portfolio composition to **both** funding and market liquidity conditions.
- We produce a set of testable conditions for the fund's optimal portfolio.
- We take the model to the data and construct liquidity indicators per management style.