Managed Funds

Mark Hendricks

January 2025

MLP Training: Portfolio Management

Outline

Performance

Measuring Risk

Hedge Funds

Fees and Flow

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Active under-performance?

For mutual funds.

- ► Studies beginning with Jensen (1968) find negative excess returns.
- ▶ Net of fees even worse.
- ► Average under-performance of about 1%.

Yet, huge industry for active management. Why?

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Momentum in mutual funds

Carhart (1997) finds

- ► Fund with good return faces close to 50/50 chance of excess returns in next period.
- ► Momentum strategy works here.
- ▶ About 1% per month excess return, but lots of variation.
- ▶ Only a 55/45 bet that last period's "winners" will have excess return next period.

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Finding skill or avoiding disaster?

The momentum results of Carhart (1997) find the worst "losers" do very poorly.

- ▶ 8% per year difference between top decile and bottom decile.
- ▶ 2.65% is the difference just between decile 9 and 10.
- ► Much of the momentum return comes from shorting these poor funds.

Fees and expenses are strongly related to poor performance.

▶ About 1% of the momentum return come from lower fees and expenses of the "winners".

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Uncertainty of a fund's mean return

Suppose we observe a fund with returns satisfying the following:

- ► are independent, identically distributed (iid)
- ▶ have volatility, $\sigma_r = 15\%$.
- ▶ have been observed for T = 5 years.
- ▶ have an observed sample average of $\overline{r} = 10\%$

As investors, we cannot observe the population mean of returns, μ . We must try to infer it.

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CLT and inferring the mean of returns

According to the Central Limit Theorem (CLT), the sample mean has the following distribution:

$$\sigma\left(\overline{r}\right) = \frac{\sigma_r}{\sqrt{T}}$$

Thus, we have the following confidence interval on the fund's mean return:

$$CI(\bar{r}, 95\%) = \bar{r} \pm 2\frac{15\%}{\sqrt{5}} \approx 10\% \pm 13\%$$

We are not sure whether the fund's mean return is -3% or 23%!

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Biased mean estimate

Not only is mean estimate uncertain, but biased for several reasons:

- ► Survivor bias: Data often only available for surviving funds.
- ▶ Backfill bias: At time *t*, we look back at full history of any fund still surviving.
- Incubator bias: Funds will be incubating for a time before they are launched. Successful incubations launch and report their incubated history.

For hedge funds, death rate of about 20% annually, so significant.

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Backfill and survivor bias

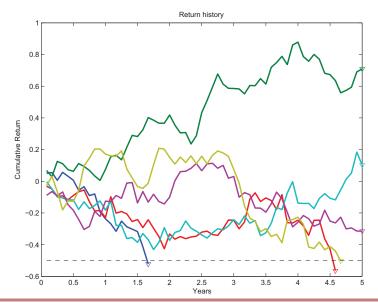
Table: Mean return data, 1994-2003.

Backfilled Not Backfilled	Hedge Funds 14.65% 7.34%	Mutual Funds — —
Surviving	13.74%	9.73%
All	9.32%	8.49%

Source: Malkiel and Saha (2005)

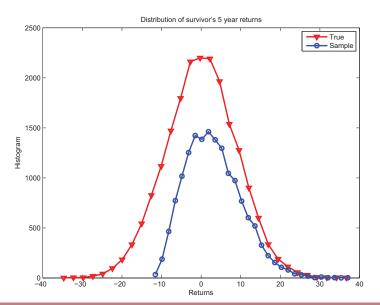
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Simulation: cumulative returns and survival



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Simulation: return distribution



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Simulation: survival bias

From the simulation above,

Statistics across funds	Raw	Selected
Mean	0.05%	3.12%
Std. Dev	8.87%	7.00%

Figure: Source: Cochrane

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Reporting returns

Returns reported by hedge funds may be "stale", rather than reflective of current market info.

- ► They hold many illiquid securities, including OTC products which are hard to price.
- ► Thus, hedge funds often have discretion in how to mark prices on their book for month-end reporting.
- ► Stale prices could cause the effects of market movements to show up in later months.

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Illustration of return smoothing

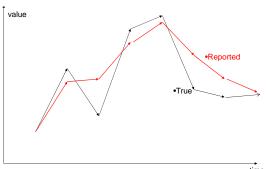


Figure: Illustration of smoothed hedge-fund returns. Source: Cochrane

- Due to illiquidity, (or managed accounting,) reported fund returns may be too smooth.
- ▶ Lowers volatility, market beta. Induces serial correlation.

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Mis-measured risk

Asness (2001) argues funds report stale prices, bias market beta down.

- ▶ Runs tests on combined return over several periods.
- ► If these lagged returns have market exposure, indicative that smoothing is hiding important factor exposure.
- Alternatively, factor exposure on these lagged returns could indicate market timing.

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Hedge-funds and lagged market risk

	(1)	(2)	(3)	(4)	(4) - (1)
	Simple	Betas from Lagged S&P 500 Regressions (Exhibit 4A)			
Portfolio	Monthly Regression Beta (Exhibit 2)	Contemporaneous Beta (β_0)	Sum of Lagged Betas $(\beta_1 + \beta_2 + \beta_3)$	Total Summed Beta $(\beta_0 + \beta_1 + \beta_2 + \beta_3)$	Difference in Beta
Aggregate Hedge Fund Index	0.37	0.40	0.44	0.84	0.47
Convertible Arbitrage	0.04	0.08	0.35	0.43	0.38
Event-Driven	0.28	0.31	0.30	0.61	0.33
Equity Market-Neutral	0.12	0.13	0.08	0.20	0.09
Fixed-Income Arbitrage	0.02	0.05	0.31	0.36	0.33
Long/Short Equity	0.55	0.57	0.42	0.99	0.45
Emerging Markets	0.74	0.79	0.46	1.25	0.51
Global Macro	0.37	0.41	0.57	0.98	0.61
Managed Futures	0.01	-0.01	-0.17	-0.19	-0.20
Dedicated Short Bias	-0.99	-1.01	-0.25	-1.27	-0.28

Figure: Source: Asness (2001). Data covers 1994-2000.

Estimates of

$$R_t^i = \alpha_i + \beta_{i,0} R_t^m + \beta_{i,1} R_{t-1}^m + \beta_{i,2} R_{t-2}^m + \beta_{i,3} R_{t-3}^m + \beta_{i,4} R_{t-4}^m + \epsilon_t^i$$

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Hedge funds taking tail risk

Many hedge-funds take on tail risk.

- ➤ Sometimes done explicitly through trades such as merger-arbitrage or the other "event" strategies mentioned in the table of slide ??.
- ▶ But also take tail risk without explicitly trading options.
- As we know from the proof of Black-Scholes, actively trading a stock and bond can lead to option-like returns.

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Why trade in the distribution tails?

- First, selling puts earns a significant premium in normal times. (Hence the volatility "smile".)
- ▶ Nothing wrong with selling insurance to the market. In some ways, it is sensible that rich investors are selling "disaster" insurance to the market.
- But it is important that we understand the risks associated with the trades!

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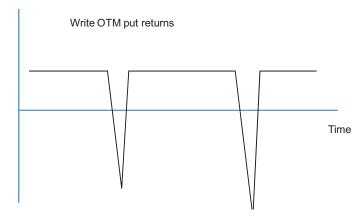
Measuring distribution tails

We mentioned above, that for any security, measuring mean returns can be difficult given the high volatility.

- ▶ Measuring the tail of a distribution is considerably harder!
- ► In short samples, we may no very little about rare events. Even in longer samples, how "rare" of events have we seen?
- Such nonlinear payoffs are challenging for statistical inference. The betas, correlations, etc. will jump in large market downturns.
- ► Thus, estimation will overestimate excess returns, underestimate risk.

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Time-series of returns for writing OTM puts



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Detecting option-like strategies: quadratic

Treynor-Mazuy regression:

$$\tilde{r}_t^i = \alpha + \beta_0 \tilde{r}_t^m + \beta_1 (\tilde{r}_t^m)^2 + \epsilon_t$$

In a down market,

- $\beta_1 > 0$ is like buying put options
- ▶ β_1 < 0 is like selling put options.

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Detecting option-like strategies: piecewise options

Option-based benchmarks:

$$\begin{split} \tilde{r}_t^i &= \alpha + \beta_0 \tilde{r}_t^m + \beta_1 \max (\tilde{r}^m - k_1, 0) + \\ &+ \beta_2 \max (k_2 - \tilde{r}_t^m, 0) + \epsilon \end{split}$$

- \triangleright $\beta_1 \neq 0$, is like being long (short) a call option.
- $ightharpoonup \beta_2 \neq 0$, is like being long (short) a put option.

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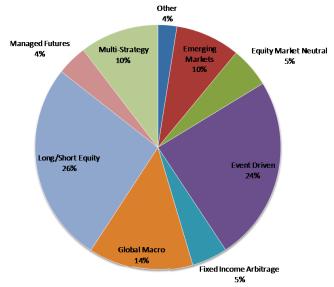
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Hedge fund composition - by strategy



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Why hedge funds?

Some oft-cited reasons...

- enhance returns
- add diversification
- get into alternative investments

What do you think?

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Alternative assets?

Commonly cited as diversifying into "alternative assets".

- ➤ You cannot be more diversified than holding the market portfolio!
- ▶ Only diversify if they offer some new asset in this market portfolio that you could not get before.
- ▶ If long assets *A* and *B*, and then through hedge-fund add a position long *A* and short *B*, no gain in diversification.

Diversifying over managers

Diversifying over investment managers just combines a lot of active positions back into a passive market position.

- ▶ But now you pay large fees to hold a passive position!
- Same idea applies to diversifying over mutual funds, but lower fees.

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Are hedge funds "hedged"?

Are hedge-fund strategies hedged against market movements?

- ▶ What about other commonly known factors?
- ► Could get large returns through levering on market beta.
- Or could load on other risk factors: carry trade, momentum, value, etc.

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Fund growth

How do investors respond to realized returns?

- ▶ Investors seek funds with not just large returns, but large α .
- ► Funds seek investors, as the fee includes 2% of assets under management.

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Fund flow as a function of returns

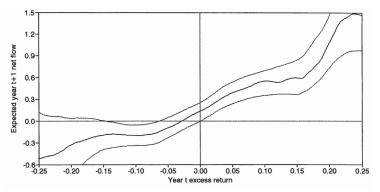


Figure: Flow of assets as function of the fund's prior return, for **young** funds. With 90% confidence interval. Flow measures proportional asset growth, net of internal growth from portfolio gains.

 $\mathsf{Flow}_{t+1}^i \equiv \left(\mathsf{Assets}_{t+1}^i - \mathsf{Assets}_t^i\right)/\mathsf{Assets}_t^i - R_{t+1}^i$. Source: Chevalier and Ellison (1997).

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Fund flow as a function of returns

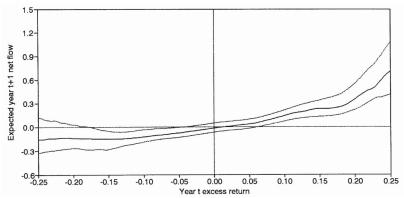


Figure: Flow of assets as function of the fund's prior return, for **old** funds. Flow measures proportional asset growth, net of internal growth from portfolio gains. Source: Chevalier and Ellison (1997).

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Chasing performance

From the previous two figures, it appears that investors chase performance.

- ► But little evidence of persistence!
- ► Though above we did note a potential momentum strategy.
- ▶ Is moving between funds due to learning manager's skill?
- Or could be avoiding funds that are prone to managerial gambling due to being below high-water.

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Fragile funds

Managed funds are fragile due to investor's chasing performance.

- ▶ When asset values are already low, investors want to pull out.
- ▶ Due to marking-to-market, first investors to leave will redeem at a higher price.
- Later withdrawals will be forced to liquidate at "fire-sale" prices.
- ► This makes funds prone to bank-runs.

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Lock-ups

Most hedge funds deal with this fragility with lockups.

- Lockups may restrict that withdrawals can only come once-per-year, or perhaps after some vesting.
- ▶ While they are restrictive, these lock-ups could indeed be enhancing value for individual investors.
- ► They keep your fellow-investors locked in so that a run does not begin, with "fire-sale" prices as a result.
- ► Also may allow fund to enjoy rebound, as bad times may have high expected returns.

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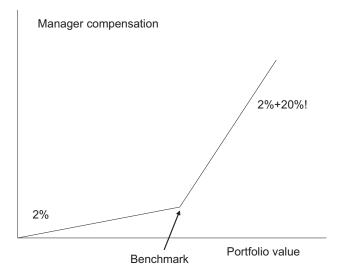
Hedge fund fees

We have discussed the risk and return characteristics of hedge fund returns—but the gross and net are very different.

- ► Typical fees are 2% of assets under management plus 20% of profits.
- High-water marks require that a fund make up previous losses before designating profits.
- ► Funds of funds charge another 1% plus 10% on top of all this!

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Hedge fund management payoff



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Managerial incentives

The payoff to management looks a lot like a call option!

- Management would prefer volatility of returns.
- High-water marks, and large positions of the managing partners are meant to re-align incentives.
- But maybe high-water marks just increase the "strike" of the compensation and encourage greater volatility!
- Besides, funds to close rather than work back up to high-water.

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Example for compensation

Have \$1000 assets under management and 2/20 fee.

- ▶ Do nothing, then \mathbb{E} [fee] = 20.
- ▶ Take a 50/50 bet on \$500, then \mathbb{E} [fee] = 70.
- ► Take a 99/1 bet on plus \$1000, minus \$100,000. Then

$$\mathbb{E}\left[\text{fee}\right] = \$1000 \times 2\% + 0.99 \times \$1000 \times 20\% = \$218$$

So in last case, getting a big reward for an investment with negative expected value!

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Rewarding risk or skill?

Fees should reward skilled allocation—not for loading up risk.

- ► A manager could always earn excess returns just by loading up on commonly known factors.
- ie. Don't reward a fund simply leveraging the market index, value stocks, small stocks, etc.
- Nor do we want to reward known strategies.

If regression of fund return on factors gives high R^2 , then we are paying fees for the equivalent of a mutual fund!

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Benchmarking

Benchmark to obvious factors, such as the market:

$$R_t^i = \alpha + \beta^m R_t^m + \epsilon_t$$

- ▶ Benchmarking compensates managers for only the unexplained portion of returns: $\alpha + \epsilon_t$.
- No reason to compensate for the factor-driven portion, $\beta^m R_t^m$, which we can easily replicate.

Unfortunately, little statistical power in this test. We have a small time series of fund returns, yet many factors which we would like to benchmark.

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Tracking mutual funds

With mutual funds, the opposite: supposed to track factors and charge low fees.

- For a portfolio of funds, Carhart (1997) finds R^2 of up to 95% using just the market index.
- Mutual fund variation driven by difference in style—not selection of stocks within a style.
- ► That is, most funds have high tracking R² using just a few common factors.

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Skill in factors?

Is it really proper to benchmark for every known factor?

- ▶ Maybe the skill of a fund is to manage all these known factors.
- ▶ Just because a return can be explained by factors, it is not clear that investors would find it convenient to build such returns themselves.

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