

# Robust Assessment of Hedge Fund Performance through Nonparametric Discounting by Almeida and Garcia

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# Overview I

- Do hedge funds really provide risk – adjusted returns much higher than the risk - free rate?
- Or is the risk – adjustment inaccurate
- Will a better risk adjustment lead to lower risk – adjusted returns?
- How can we carry out the risk adjustment more accurately?

# Overview II

## Theoretical Asset Pricing

### Definition 1

$m$  is a SDF if

$$\forall \text{ assets } i, E[mR_i] = 1. \quad (1)$$

## Empirical Asset Pricing

- Using some empirical methodology [e.g. Hansen & Jagannathan, (1991)], find that

$$m = a + \sum_{j=1}^J b_j \underbrace{f_j}_{\text{factor, usually a return of a basis asset}} \quad (2)$$

- Using empirically determined SDF, price hedge fund return

$$E[mR_{HF}] = \alpha \quad (3)$$

- $\alpha$  large and positive
- hedge fund offers high risk adjusted return or the SDF has been determined inaccurately

# This paper

- New methodology for determining a **family of SDF's** from time series data on returns
- SDF can be a **non linear** function of factors (returns on basis assets)
  - presence of non linearities in hedge fund returns: Fung & Hsieh (2001), Mitchell & Pulvino (2001), Agarwal & Naik (2004)
  - usual approach: include options in set of basis assets, keep  $m$  linear
- relative to linear SDF, non linear SDF's give **lower  $\alpha$** 
  - some of the non linear SDF's capture better **higher moments** – reduction in  $\alpha$
  - accounting for **extreme events** leads to a reduction in  $\alpha$
- determine SDF via **Fenchel dual** problem – easier than primal approach
  - dual problem is a portfolio choice problem

# Estimating a SDF I

- for all assets,  $i$

$$E[mR_i] = 1 \quad (4)$$

- this is an **average over states** of the economy  $\sum_{k=1}^K p(\omega_k) m(\omega_k) R_i(\omega_k) = 1$
- if Ergodic Hypothesis is true (replace average over states with time series average – can use time series data), then

$$\underbrace{\lim_{T \rightarrow \infty} \sum_{t=1}^T \frac{m_t R_{i,t}}{T}}_{\text{time series average}} = \underbrace{E[mR_i]}_{\text{average over states}} \quad (5)$$

- For large  $T$

$$\sum_{t=1}^T \frac{m_t R_{i,t}}{T} = 1 \quad (6)$$

$$\frac{1}{T} \sum_{t=1}^T m_t (R_{i,t} - a^{-1}) = 0 \quad (7)$$

where  $a = \sum_{t=1}^T \frac{m_t}{T}$

- Need an extra condition to determine SDF

# Estimating a SDF II

- Hansen – Jagannathan (1991)

$$\hat{m}_{HJ} = \frac{1}{2} \underset{\{m_t\}_{t=1}^T}{\operatorname{argmin}} \overbrace{\frac{1}{T} \sum_{t=1}^T [(m_t)^2 - a^2]}^{\text{variance of SDF}} \quad (8)$$

s.t.

$$\forall i \in \{1, \dots, I\}, \frac{1}{T} \sum_{t=1}^T m_t (R_{i,t} - a^{-1}) = 0 \quad (9)$$

$$\sum_{t=1}^T \frac{m_t}{T} = a \quad (10)$$

$$\forall t \in \{1, \dots, T\}, m_t > 0. \quad (11)$$

- Alternatives to minimizing variance: Snow (1991), Stutzer (1995), Bansal & Lehmann (1997), Bernardo & Ledoit (2000), Cerny (2003)
- This paper: replace (8) by  $\hat{m}_{MD} = \frac{1}{2} \underset{\{m_t\}_{t=1}^T}{\operatorname{argmin}} \frac{1}{T} \sum_{t=1}^T \phi(m_t)$ , where  $\phi(\cdot)$  is a convex function of the form

$$\phi^\gamma(m) = \frac{m^{1+\gamma} - a^{1+\gamma}}{\gamma(1+\gamma)} \quad (12)$$

# Fenchel dual problem

- Instead of finding  $\{m_t\}_{t=1}^T$  find  $\{\lambda_i\}_{i=1}^I$ :  $I < T$  (easier)

$$\hat{\lambda} = \underset{\epsilon \in \mathbb{R}, \lambda \in \mathbb{R}^I}{\operatorname{argsup}} a\epsilon - \frac{1}{T} \sum_{t=1}^T \psi^\gamma(\epsilon + \lambda^T (R_t - \frac{1}{a} \mathbf{1}_I)) \quad (13)$$

$$\psi^\gamma(z) = \underbrace{\sup_{w > 0} zw - \phi^\gamma(w)}_{\text{Fenchel dual}} \quad (14)$$

$$\hat{m}_{t,MD} = a \frac{\left(a^\gamma + \gamma \hat{\lambda}^T (R_t - \frac{1}{a} \mathbf{1}_I)\right)^{1/\gamma}}{\frac{1}{T} \sum_{t=1}^T \left(a^\gamma + \gamma \hat{\lambda}^T (R_t - \frac{1}{a} \mathbf{1}_I)\right)^{1/\gamma}} \quad (15)$$

- SDF is **non linear**
  - Prior literature: SDF linear but with returns from assets with non linear payoffs
  - $\gamma = 1$ : reduces to linear SDF. Hansen – Jagannathan (1991) is a special case of this paper.
  - Dual problem can be interpreted as a portfolio choice problem
    - investor with HARA utility:  $-\gamma$  cautiousness parameter
    - $\{\lambda_i\}_{i=1}^I$  are units of wealth invested in each risky asset

# What is going on?

- Researcher chooses a set of basis assets:  $\{R_i\}_{i=1}^I$
- HARA investor with cautiousness,  $-\gamma$  chooses portfolio of the basis assets, i.e.  $\lambda$
- $\lambda$  plus the basis assets gives us the nonlinear SDF,  $m$
- for each  $\gamma$  get a different SDF: a family of SDF's



# Family of SDF's

## Portfolios ( $\lambda$ )

- More cautious investors ( $-\gamma > 0$  [see Eq (14) in paper]) and larger: put more weight on skewness and kurtosis
- $-\gamma > 0$  and larger: more weight on skewness and kurtosis in SDF
- returns: S&P 500 (equity), RU2000 (equity, size – spread), MSCI (emerging market index), 10yrTr, BAA,  $r_f$ 
  - all investors: short BAA, long all others
  - more cautious investors: more weight on  $r_f$  rises, less weight on risky assets

## SDF volatility $\equiv$ market price of risk

- options added to set of basis assets: higher market price of risk and more sensitivity to extreme events
- More cautious HARA investor: higher market price of risk
- higher market price of risk: will lower performance of any risky strategy correlated with SDF
- extreme events v. important for correct analysis of hedge fund performance if HF's sell puts!

## Pricing errors

- Lowest for  $-\gamma = 3, 2, 0$
- Large for  $-\gamma = 1$  (HJ)

# Hedge fund performance

- increasing cautiousness of HARA investor: SDF more volatile, lower pricing errors
- adding options to basis assets: SDF higher during very bad states of the world and SDF more volatile
- increasing cautiousness of HARA investor and adding options to basis assets lowers hedge fund performance

# Suggestions: Fenchel duality

- Fenchel duality appears in many places: physics (with v clear interpretations), finance (with less clear interpretations)
  - Consumption – portfolio choice problems with constraints: more easily solved using Fenchel dual [Cvitanic & Karatzas (1992)]
  - Social planner's optimization problem where individual agents have recursive preferences: more easily solved using Fenchel dual [Dumas, Uppal & Wang (2000)]
    - Fenchel dual can be interpreted as problem of a robust investor
- This paper: why does the Fenchel dual problem in this paper appear in the form of a portfolio choice problem
- Can we find a unified economic interpretation for Fenchel dual problems in general?

## Suggestions: Testing the methodology

- Compare your methodology for estimating SDF with others in a framework where we actually know what the SDF is
  - consumption – based asset pricing model, complete markets
    - e.g.,  $N$  dividend trees subject to rare disasters, single agent with power utility or several agents with catching up with Jones preferences (different curvature parameters), complete markets
    - compute unique SDF,  $m$ , directly
    - use simulated time series data on returns to estimate SDF,  $\hat{m}$ : measure  $\|\hat{m} - m\|$
  - consumption – based asset pricing model, incomplete markets
    - each agent's SDF can be computed directly,  $m_h$
    - use simulated time series data on returns to estimate SDF,  $\hat{m}$ : which SDF is it closest to?