AN APPLICATION OF GIBBONS-ROSS-SHANKEN'S TEST OF THE EFFICIENCY OF A GIVEN PORTFOLIO

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Abstract

This paper provides an adaptation of the statistical tests of Gibbons, Ross, and Shanken (1989) to test for portfolio efficiency in two cases where theirs can not directly be used: 1) When the portfolio whose efficiency is being tested is not included in the set of securities generating the mean-standard deviation frontier and, 2) When testing for the existence of an efficient portfolio (of a given set of L portfolios) when none of these L portfolios is included in the set of securities generating the mean-standard deviation frontier. Our tests can be used to determine the efficiency of a variety of mutual funds

Resumen

En este artículo se adaptan las pruebas estadísticas de Gibbons, Ross, y Shanken (1989) para probar la eficiencia de un portafolio en dos casos en los que sus pruebas no pueden usarse directamente: 1) Cuando el portafolio cuya eficiencia está siendo probada no está incluído en el conjunto de instrumentos financieros que generan la frontera de media-desviación estándard y, 2) Cuando se prueba la existencia de un portafolio eficiente (de un conjunto dado de L portafolios), cuando ninguno de estos L portafolios está incluído en el conjunto de instrumentos financieros que generan la frontera de media-desviación estándard. Nuestros estadísticos pueden usarse para probar la eficiencia de fondos de inversión.

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1. Introduction

In this paper we adapt the statistical tests of Gibbons, Ross, and Shanken (1989) using the multivariate regression approach to test: i) Whether a given portfolio of N+1 securities is efficient with respect to this set of N+1 securities (in contrast to G-R-S's case, this set does not include the portfolio whose efficiency is being tested), and, ii) Whether there exists a portfolio of a given set of L portfolios of N+1 securities, (L<N), that is efficient with respect to the set of N+1 securities (in contrast to G-R-S's case, this set of N+1 securities does not include the L portfolios of whom the existence of an efficient portfolio is being tested).

Gibbons, Ross, and Shanken (1989) (see also MacKinlay (1987)) propose a test to determine "whether any particular portfolio is ex-ante mean-variance efficient", *i.e.* whether any particular portfolio is ex-ante on the mean-standard deviation frontier. They work with N+2 securities or portfolios: a riskless security with return R_0 , N linearly independent risky securities with returns R_1, R_2, \cdots, R_N and a portfolio p, whose efficiency is being tested, with return R_p . Portfolio p is not a portfolio of the first N+1 securities, *i.e.* R_p is not a weighted average of the returns R_0, R_1, \cdots, R_N . Their test is aimed at testing whether portfolio p is on the frontier generated by the N+1 securities with returns R_0, R_1, \cdots, R_N , and portfolio p, itself. This statistical test has frequently been used in testing the CAPM.

In Section 2 we work with N+2 securities or portfolios: the N+1 securities with returns R_0, R_1, \cdots, R_N and a portfolio q, whose efficiency is being tested, with return $R_q = \sum_{i=0}^N \alpha_i R_i$, where $\sum_{i=0}^N \alpha_i = 1$, (i.e. In our case, portfolio q is a portfolio of the first N+1 securities). The first of our objectives is to use G-R-S's statistical test to derive a corresponding statistical test to determine whether portfolio q is on the mean-standard deviation frontier generated by the N+1 securities with returns R_0, R_1, \cdots, R_N . This adapted version of the G-R-S's test can be used directly when testing whether a given portfolio of a set of securities is efficient with respect to this set of securities. G-R-S's test can not be applied directly in this case.

In Section 7 of Gibbons, Ross, and Shanken (1989), they propose a test to determine whether there exists a portfolio of a given set of L portfolios that is ex-ante mean-variance efficient, i.e. whether there exists a portfolio of a given set of L portfolios that is ex-ante on the mean-standard deviation frontier. They work with N+L+1 securities or portfolios: a riskless security with return R_0 , N linearly independent securities with returns R_1, R_2, \cdots, R_N and L linearly independent portfolios p_1, p_2, \dots, p_L , with returns $R_{p1}, R_{p2}, \dots, R_{pL}$. None of these L portfolios is a portfolio of the first N+1 securities, i.e. for all $k \in \{1, 2, \dots, L\}$, R_{pk} is not a weighted average of the returns R_0, R_1, \dots, R_N . Their test is aimed at testing whether there exists a portfolio of the L portfolios p_1, p_2, \dots, p_L on the frontier generated by the N+1 securities with returns R_0, R_1, \dots, R_N and portfolios $p_1, p_2 \cdots, p_L$, themselves. Their statistical test has been frequently used in testing some versions of the Arbitrage Pricing Theory, APT, in which the pricing restriction can be restated as "a linear combination of factor portfolios is mean-variance efficient" (Chamberlain (1983), Grinblatt and Titman (1987), Connor and Korajczyk (1988, 1995), Lehmann and Modest (1988), etc.).

In Section 3 we work with N+L+1 securities or portfolios: the N+1 securi-

ties with returns $R_0, R_1, R_2, \cdots, R_N$ plus L (L \leq N), linearly independent portfolios q_1, q_2, \cdots, q_L , with returns $R_{q1}, R_{q2}, \cdots, R_{qL}$, where $R_{qk} = \sum_{i=0}^N \alpha_i^k R_i$ and $\sum_{i=0}^N \alpha_i^k = 1$, for $k=1,2,\cdots,L$ (in our case, these L portfolios are portfolios of the first N+1 securities). The second of our objectives is to use G-R-S's statistical test to derive a corresponding statistical test to determine whether there exists a portfolio of portfolios q_1, q_2, \cdots, q_L that is on the mean-standard deviation frontier generated by the N+1 securities with returns R_0, R_1, \cdots, R_N . Our adapted version of G-R-S's statistical test can be used directly when testing if there exists a portfolio of a given set of portfolios that is efficient with respect to the set of securities included in the portfolios. G-R-S's test can not be applied directly in this case.

In Section 4 we mention cases in which the statistical tests presented here can be useful when testing the efficiency of a portfolio. In Section 5 we summarize the results.

2. Testing the Efficiency of a Portafolio

Given a set \mathcal{S} of K securities $(K \geq 1)$ a portfolio of these securities with expected return equal to $E \in \mathbb{R}$, is the frontier portfolio with expected return E if its return has the minimum variance among returns of portfolios (of securities in S) that have expected return E (see Constantinides and Malliaris (1995)). A frontier portfolio is also known as a mean-variance efficient portfolio. In case the frontier portfolio exists for $E \in \mathbb{R}$, let $\sigma(E)$ be its standard-deviation. The mean-standard deviation frontier generated by the K securities in \mathcal{S} is the set:

$$MSF \equiv \{ (\sigma(E), E) \in \mathbb{R}^2 \mid \sigma(E) \text{ exists for } E \in \mathbb{R} \}.$$
 (1)

Gibbons, Ross, and Shanken (1989) consider a given portfolio p which is not a portfolio of the riskless security with return R_0 and the N linearly independent securities with returns R_1, \dots, R_N . They propose a test to determine whether portfolio p is in the frontier generated by the securities with returns R_0, R_1, \dots, R_N , and portfolio p, itself. Here we consider a given portfolio q which is a portfolio of the N+1 securities with returns R_0, R_1, \dots, R_N . We use their test to generate a corresponding test to determine whether portfolio q is in the frontier generated by the securities with returns R_0, R_1, \dots, R_N .

To be able to apply G - R - S's test we need a way to go from their case to ours. This can be done using the following well known observation that states that repackaging the securities does not alter the mean-standard deviation frontier:

Lemma 1. The mean-standard deviation frontier generated by a set of M linearly independent securities is equal to the mean-standard deviation frontier generated by any M linearly independent portfolios of the M linearly independent securities.

Dm. Let R_x denote the M-dimensional vector of returns of M linearly independent securities, and $E(R_x)$ be the corresponding vector of expected returns. Let R_y denote the M-dimensional vector of returns of M linearly independent portfolios of the M securities, and $E(R_y)$ be the corresponding vector of expected returns. Then, $R_y = \Gamma' R_x$, where Γ is an invertible M×M matrix such that $\Gamma' 1_M = 1_M$, where 1_M is the M-dimensional vector of ones.

Let $\sigma_x(E)$ be the standard deviation of the return of the frontier portfolio of the M securities with expected return $E \in \mathbb{R}$, and $\sigma_y(E)$ be the standard deviation of the return of the frontier portfolio of the M portfolios with expected return $E \in \mathbb{R}$. This means:

$$\sigma_x(\mathbf{E}) \equiv \min_{\omega \in \mathbb{R}^{\mathbf{M}}} \operatorname{var}(\omega' R_x)$$
 (2)

s.t.
$$\omega' 1_{\mathbf{M}} = 1$$
 and $\omega' \mathbf{E}(R_x) = \mathbf{E}$, (3)

and

$$\sigma_y(\mathbf{E}) = \min_{\alpha \in \mathbb{R}^M} \operatorname{var}(\alpha' \Gamma' R_x) \tag{4}$$

s.t.
$$\alpha' 1_{\mathbf{M}} = 1$$
 and $\alpha' \mathbf{E}(\Gamma' R_x) = \mathbf{E}$. (5)

It has to be shown that for $E \in \mathbb{R}$, $\sigma_x(E) = \sigma_y(E)$. If ω^* solves minimization problem (2), then $\alpha^* = \Gamma^{-1}\omega^*$ solves minimization problem (4). Otherwise, there exists $\alpha \in \mathbb{R}^M$ satisfying (5) such that $\operatorname{var}(\alpha'\Gamma'R_x) < \operatorname{var}(\alpha^{*'}\Gamma'R_x) = \operatorname{var}(\omega^{*'}R_x)$ Then, $\omega \equiv \Gamma\alpha$ satisfies (3) and $\operatorname{var}(\omega'R_x) < \operatorname{var}(\omega^{*'}R_x)$, which contradicts the fact that ω^* solves problem (2). Hence, $\alpha^* = \Gamma^{-1}\omega^*$ solves minimization problem (4). Hence,

$$\sigma_y(\mathbf{E}) = \operatorname{var}(\alpha^{*'}\Gamma'R_x) = \operatorname{var}(\omega^{*'}\Gamma'^{-1}\Gamma'R_x) = \operatorname{var}(\omega^{*'}R_x) = \sigma_x(\mathbf{E}) \blacksquare$$

Next, we apply Lemma 1 to our case. Portfolio q has return $R_q = \sum_{i=0}^{N} \omega_i R_i$, where $\sum_{i=0}^{N} = 1$. Without loss of generality, $\omega_1 \neq 0$. Let $R_x \equiv (R_0, R_1, \cdots R_N)'$, $\Gamma \equiv (e_1 \ \omega_q \ e_3 \ e_4 \cdots e_{N+1})$, where e_i is the (N+1) dimensional vector with 1 in the i-th row and zero elsewhere, $i=1,3,4,\cdots,N+1$, $\omega_q \equiv (\omega_0 \ \omega_1 \cdots \omega_N)'$, and $R_y \equiv \Gamma' R_x = (R_0 \ R_q \ R_2 \cdots R_N)'$. Det $(\Gamma) = \omega_1 \neq 0$. By Lemma 1, our hypothesis:

 H_0 : "Portfolio q is on the mean-standard deviation frontier generated by the N+1 securities with returns $R_0, R_1, R_2, \dots, R_N$ "

is equivalent to the hypothesis:

 H_0' : "Portfolio q is on the mean-standard deviation frontier generated by the N securities with returns $R_0, R_2, R_3, \dots, R_N$, and portfolio q, itself."

With this, the hypothesis we want to test is in G - R - S's frame and we can apply their test adapted as follows:

Following Gibbons, Ross and Shanken (1989), it is assumed we have data from T periods. Let r_{qt} be the excess return of portfolio q over the return of the riskless security on period t ($r_{qt} \equiv R_{qt} - R_{0t}$); let r_t be the (N – 1)-dimensional vector of excess returns of securities 2 through N over the return of the riskless security on period t ($r_t \equiv (R_{2t} - R_{0t} \quad R_{3t} - R_{0t} \cdots R_{Nt} - R_{0t})'$).

Consider the multivariate linear regression:

$$r_t = \alpha + br_{qt} + \varepsilon_t$$
 $t = 1, 2, \dots, T$ (6)

where

$$E(\varepsilon_t) = 0$$
, $E(\varepsilon_t r_{qt}) = 0$, $\alpha \in \mathbb{R}^{N-1}$, $b \in \mathbb{R}^{N-1}$.

It is assumed that $\{(r'_t \ r_{qt})\}_{t=1}^T$ are independent and identically distributed multinormal N-dimensional random vectors. This implies $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_T$ are independent and identically distributed $N_{N-1}(0, \Sigma_{\varepsilon})$ random vectors and ε_t is statistically independent of r_{qt} , for $t = 1, 2, \dots, T$. (Muirhead (1982)).

Let $\hat{\alpha}$ and \hat{b} the OLS estimators of α and \hat{b} .

$$\hat{\Sigma}_{\varepsilon}^{u} \equiv \frac{1}{T - 2} \sum_{t=1}^{T} \left[r_{t} - \hat{\alpha} - \hat{b}r_{qt} \right] \left[r_{t} - \hat{\alpha} - \hat{b}r_{qt} \right]', \tag{7}$$

$$\hat{\Sigma}_{\varepsilon}^{\text{MLE}} \equiv \frac{T-2}{T} \hat{\Sigma}_{\varepsilon}^{u}.$$

(i.e. $\hat{\Sigma}^u_{\varepsilon}$ is the unbaised estimator of Σ_{ε} and $\hat{\Sigma}^{\mathrm{MLE}}_{\varepsilon}$ is the maximum likelihood estimator of Σ_{ε} . N $\leq T-1$ so that $\hat{\Sigma}^u_{\varepsilon}$ is nonsingular).

$$\bar{r}_q \equiv \frac{1}{T} \sum_{t=1}^{T} r_{qt},\tag{8}$$

$$\hat{\sigma}_q^2 \equiv \frac{1}{T} \sum_{t=1}^T (r_{qt} - \bar{r}_q)^2, \tag{9}$$

$$\hat{\mathcal{S}}_q^2 \equiv \bar{r}_q^2 / \hat{\sigma}_q^2. \tag{10}$$

Applying Gibbons, Ross & Shanken (1989), we get the following: Proposition 1. The small-sample conditional distribution of

$$\frac{T(T-N)}{(N-1)(T-2)} \cdot \frac{\hat{\alpha}'\hat{\Sigma}_{\varepsilon}^{u-1}\hat{\alpha}}{1+\hat{S}_{g}^{2}} = \frac{T-N}{N-1} \cdot \frac{\hat{\alpha}'\hat{\Sigma}_{\varepsilon}^{MLE-1}\hat{\alpha}}{1+\hat{S}_{g}^{2}},$$
(11)

given r_q , is non-central $F_{N-1,T-N}(\lambda)$, where

$$\lambda \equiv T \cdot \frac{\alpha' \Sigma_{\varepsilon}^{-1} \alpha}{1 + \hat{S}_{q}^{2}} \quad \blacksquare \tag{12}$$

Under the hypothesis H_0 , the parameter λ is equal to zero (because $\alpha = 0$) and the statistic (11) has central $F_{N-1,T-N}$ distribution (uncondicionally). Given the observations $\{(r'_t \ r_{qt})\}_{t=1}^T$, the hypotesis H_0 is rejected if the statistic (11) is significantly different from zero.

G-R-S's geometrical interpretation of their test is preserved: $\alpha' \sum_{\varepsilon}^{-1} \alpha = S^2-S_q^2$, where this time, S is the slope of the frontier generated by the securities with returns $R_0, R_2, R_3, \dots, R_N$, and portfolio q, and $S_q = \mathrm{E}(r_q)/(\mathrm{var}(r_q))^{1/2}$. By Lemma 1 above and the fact that $\alpha' \sum_{\varepsilon}^{-1} \alpha = S^2 - S_q^2$, the statistical test presented in Proposition 1 is not affected by wich of the risky securities

is labeled as the security with return " R_1 ", as long as this one is included in portfolio q (i.e. $\omega_1 \neq 0$).

3. Testing the Efficiency of a Portafolio of L Portfolios

Gibbons, Ross and Shanken (1989) also consider a given set of $L \geq 1$ linearly independent portofolios p_1, p_2, \cdots, p_L , none of wich is a portfolio of the securities with returns R_0, R_1, \cdots, R_N . They propose a test to determine wether there exists a portfolio of these L portfolios in the frontier generated by the securities with returns R_0, R_1, \cdots, R_N and portfolios p_1, p_2, \cdots, p_L , themselves. Here we consider a given set of $1 \leq L \leq N$ linearly independent portfolios q_1, q_2, \cdots, q_L , where each of them is a portfolio of the N+1 securities with returns R_0, R_1, \cdots, R_N . We use their test to generate a corresponding test to determine whether there exist a portfolio of the L portfolios q_1, q_2, \cdots, q_L in the frontier generated by the securities with returns R_0, R_1, \cdots, R_N .

We proceed by applying Lemma 1 to our case. For K= 1, 2, · · · ,L, portfolio q_k has return $R_{q_k} = \sum_{i=0}^{N} \omega_i^k R_i$, where $\sum_{i=0}^{N} \omega_i^k = 1$. Let us assume that $\omega_k^k \neq 0$. (i. e. that security "k" is included in portfolio q_k , K = 1, 2, · · · ,L). Let $R_x \equiv (R_0, R_1, \cdots, R_N)'$, $\Gamma \equiv (e_1 \ \omega_{q_1} \ \omega_{q_2} \cdots \omega_{q_L} \ e_{L+2} \ e_{L+3} \cdots e_{N+1})$, where e_i is the (N + 1) dimensional vector with 1 in the i-th row and zero elsewhere, $\omega_{q_k} \equiv (\omega_0^k \ \omega_1^k \cdots \omega_N^k)'$ and $R_y \equiv \Gamma' R_x = (R_{g_0} \ R_{q_1} \ R_{q_2} \cdots R_{q_L} \ R_{L+1} \ R_{L+2} \cdots R_N)'$. Det $(\Gamma) = \omega_1^1 \times \omega_2^2 \times \cdots \times \omega_L^L \neq 0$. By Lemma 1, the hypothesis: H_L : "There exists a portfolio of the L portfolios q_1, q_2, \cdots, q_L in the mean-

 H_L : "There exists a portfolio of the L portfolios q_1, q_2, \dots, q_L in the mean-standard deviation frontier geneated by the N+1 securities with returns R_0, R_1, \dots, R_N "

is equivalent to the hypothesis:

 H'_L : "There exists a portfolio of the L portfolios q_1, q_2, \dots, q_L in the mean-standard deviation frontier generated by the N-L+1 securities with returns $R_0, R_{L+1}, R_{L+2}, \dots, R_N$, and the L portfolios q_1, q_2, \dots, q_L , themselves".

With this, the hypothesis we want to test is in the G - R - S' frame and we can apply their test.

Following Gibbons, Ross, and Shanken (1989), and Shanken (1986), let r_{qt} be the L-dimensional vector of excess returns of portfolios q_1, q_2, \cdots, q_L over the return of the riskless security on period t ($r_{qt} \equiv (R_{q1t} - R_{0t} \ R_{q2t} - R_{0t} \cdots R_{qLt} - R_{0t})'$); and ρ_t be the (N-L) dimensional vector of excess returns of securities L + 1 through N over the return of the riskless security on period t ($\rho_t \equiv (R_{L+1,t} - R_{0t} \ R_{L+2,t} - R_{0t} \cdots R_{N,t} - R_{0,t})'$).

Consider the multivariate linear regression:

$$\rho_t = \delta + Br_{qt} + e_t \qquad t = 0, 1, 2, \cdots, T \tag{13}$$

where $\delta \in \mathbb{R}^{\mathrm{N-L}}$, B is a (N-L)× L matrix, $\mathrm{E}(e_t) = 0$ and $\mathrm{E}(e_t r'_{qt}) = 0$. It is assumed that $\{(\rho'_t \ r_{qt})\}_{t=1}^T$ are independent and identically distributed multinormal N-dimensional random vectors. This implies e_1, e_2, \cdots, e_T are independent and identically distributed $N_{\mathrm{N-L}}(0, \Sigma_e)$ random vectors and e_t is statistically independent of r_{qt} for $t=1,2,\cdots,T$ (Muirhead (1982)).

Let $\hat{\delta}$ and \hat{B} the OLS estimators of δ and \hat{B}

$$\hat{\Sigma}_{e}^{u} \equiv \frac{1}{T - (L+1)} \sum_{t=1}^{T} \left[r_{t} - \hat{\delta} - \hat{B}r_{qt} \right] \left[r_{t} - \hat{\delta} - \hat{B}r_{qt} \right]', \tag{14}$$

$$\hat{\Sigma}_e^{\text{MLE}} \equiv \frac{T - (L+1)}{T} \hat{\Sigma}_e^u.$$

(i. e. $\hat{\Sigma}_e^u$ is the unbiased estimator of Σ_e and $\hat{\Sigma}_e^{\text{MLE}}$ is the maximum likelihood estimator of Σ_e).

$$\bar{r}_{q} \equiv \frac{1}{T} \sum_{t=1}^{T} r_{qt},\tag{15}$$

$$\hat{\Sigma}_{q} \equiv \frac{1}{T} \sum_{t=1}^{T} (r_{qt} - \bar{r}_{q})(r_{qt} - \bar{r}_{q})', \tag{16}$$

$$\hat{\theta}_q^2 \equiv \bar{r}_q' \hat{\Sigma}_q^{-1} \bar{r}_q. \tag{17}$$

Applying Gibbons, Ross, and Shanken (1989), we get the following: Proposition 2. The small-sample conditional distribution of

$$\frac{T(T-N)}{(N-L)(T-L-1)} \cdot \frac{\hat{\delta}' \hat{\Sigma}_{\varepsilon}^{u^{-1}} \hat{\delta}}{1+\hat{\theta}_{g}^{2}} = \frac{T-N}{N-L} \cdot \frac{\hat{\delta}' \hat{\Sigma}_{\varepsilon}^{MLE^{-1}} \hat{\delta}}{1+\hat{\theta}_{g}^{2}},$$
(18)

given r_q , is non-central $F_{N-L,T-N}(\lambda)$, where

$$\lambda \equiv T \cdot \frac{\delta' \Sigma_e^{-1} \delta}{1 + \hat{\theta}_a^2} \quad \blacksquare \tag{19}$$

Under the hypothesis H_L , the parameter λ is equal to zero (because $\delta=0$) and the statistic (18) has a central $F_{N-L,T-N}$ distribution (unconditionally). Given the observations $\{(r'_t \ r_{qt})\}_{t=1}^T$ the hypotesisi H_L is rejected if the statistic (18) is significantly different from zero.

The geometrical interpretation of the test is as follows: $\delta' \Sigma_{\varepsilon}^{-1} \delta = \mathcal{S}^2 - \mathcal{S}_{L}^2$, where S is the slope of the frontier generated by the securities with returns $R_0, R_{L+1}, R_{L+2}, \dots, R_N$, and the portfolios q_1, q_2, \dots, q_L , and \mathcal{S}_L is the slope of the frontier generated by R_0 and the portfolios q_1, q_2, \dots, q_L .

By Lemma 1 above and the fact that $\delta'\Sigma_{\varepsilon}^{-1}\delta=\mathcal{S}^2-\mathcal{S}_{\mathbf{L}}^2$, the statistical test presented in Proposition 2 is not affected by which of the risky securities are labeled as securities with returns " R_1 ", " R_2 ", ..., " $R_{\mathbf{L}}$ ", as long as the security with return " R_k " is included in portfolio q_k , for $k=1,2,\cdots,\mathbf{L}$ (i. e. $\omega_k^k\neq 0,\ k=1,2,\cdots,\mathbf{L}$).

4. Applications

Our statistical tests presented above have very practical applications. The test in Proposition 1 can be used to determine if a given portfolio of a set of N securities is efficient with respect to this set of N securities. This can be applied in the cases of a mutual fund, a money market fund, an index fund, etc. Given data on the returns of a fund and of its underlying securities, we can test for the efficiency of the fund with respect to its set of underlying securities.

The statistical test presented in Proposition 2, above, can be applied to test for the existence of a portfolio of a given set of L portfolios that is efficient with respect to the set of securities included in the L portfolios. Given data on the returns of a set of mutual funds, money market funds, index funds, etc. and the returns of their underlying securities, we can therefore test for the existence of an efficient portfolio of the funds with respect to the whole set of underlying securities.

5. Summary

The statistical tests of Gibbons, Ross, and Shanken (1989) are used to determine whether a given security or portfolio is efficient with respect to a set of linearly independent securities that includes the security or portfolio whose efficiency is being tested, and to determine whether there exists a portfolio of a given set of L portfolios that is efficient with respect to a set of linearly independent securities that includes the set of L portfolios. In this paper we use their tests to generate two corresponding statistical tests to determine whether a given portfolio of N securities is efficient with respect to the set of N securities and whether there exists a portfolio of a given set of L portfolios of N securities (L < N) that is efficient with respect to the set of N securities. We replace one of the securities (L of the securities) generating the mean-standard frontier from the multivariate regression with the portfolio (L portfolios) under consideration to obtain the adapted tests.

Our tests can then be applied to test the efficiency of a mutual fund with respect to its set of underlying securities or to test for the existence of an efficient portfolio of mutual funds with respect to their whole set of underlying securities.

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