# Some Properties for A Portfolio Optimization Model

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#### Abstract

Portfolio optimization Problem is to find the securities portfolio minimizing the risk for a required return or maximizing the return for a given risk level. In this paper, we discuss a portfolio investment model with expected rate of return under non-negative constrains. we proved some properties of the model. Using these properties, the model solving will be simplified.

#### Mathematics Subject Classification: 91B28

Keywords: portfolio optimization, risk, expected rate of return, properties

## 1 Introduction

Portfolio theory has been an important part of modern finance theory. The traditional portfolio optimization problem is to find an investment plan for securities with a reasonable trade-off between the rate of return and risk. The mean-variance model of Markowitz is designed to obtain the portfolio which can achieve a specified average rate of return with the minimum risk<sup>[1]</sup>. Because of its computational complexity, many researchers study the method to solve the model[2,3]. The main aim of the present paper is to give some properties for the Markowitz model so as to simplified the solving process.

## 2 Mathematical Model

The following model is based upon mean-variance model by Markowitz<sup>[1]</sup>.

$$\begin{cases}
Min & f(w) = w^T V w \\
s. t. & Aw = b \\
& w \ge 0
\end{cases}$$
(1)

Where

$$w = (w_1, x_2, \dots, w_n)^T, \quad V = (\sigma_{ij})_{n \times n}$$
$$A = \begin{pmatrix} r_1 & r_2 & \cdots & r_n \\ 1 & 1 & \cdots & 1 \end{pmatrix}, b = \begin{pmatrix} r^* \\ 1 \end{pmatrix},$$

n is the number of securities selected by investors;  $w_i$  is the proportion of securities i in the portfolio;  $w = (w_1, w_2, \ldots, w_n)$  is the investment portfolio;  $\sigma_{ij}$  is the covariance between securities i and j; f(w) is the risk measured by the variance of the portfolio;  $r_i$  is the expected return rate of securities i;  $r^*$  is the expected return rate of the portfolio.

If we remove the constrains  $w \ge 0$ , then we get the model (2):

$$\begin{cases} Min \quad f(w) = w^T V w\\ s. \ t. \quad Aw = b \end{cases}$$
(2)

## **3** Some Properties for Mean-Variance Model

Model (1) and (2) are both quadratic programming model, for the covariance matrix V be positive semi-definite, the optimal solutions of model (1) and (2) are existed and unique. For simplicity we further assume the matrix V be positive definite, and the optimal solution is of uniqueness.<sup>[2]</sup>

We denotes the optimal solution of model (1) by  $w^{(1)} = (x_1, x_2, \ldots, x_n)^T$ , and denotes the optimal solution of model (2) by  $w^{(2)} = (y_1, y_2, \ldots, y_n)^T$ . Now we describe some useful properties for the two models.

**Lemma**  $\mathbf{1}^{[3]} w^{(2)}$  be the optimal solution of model (2), if and only if there exists a vector  $\alpha$  satisfy

$$\begin{cases} 2Vw^{(2)} = A^T \alpha \\ Aw^{(2)} = b \end{cases}$$

and the optimal solution of model (2) is  $w^{(2)} = V^{-1}A^T(AV^{-1}A^T)^{-1}b$ . **Lemma 2**<sup>[3]</sup>  $w^{(1)}$  be the optimal solution of model (1), if and only if there exists a vector  $\beta$  and a vector  $\gamma = (\gamma_1, \gamma_2, \ldots, \gamma_n)^T$  satisfy

$$\begin{cases} 2Vw^{(1)} = A^{T}\beta + \gamma \\ Aw^{(1)} = b \\ \gamma_{i}x_{i} = 0, \quad i = 1, 2, \dots, n \\ w^{(1)} \ge 0, \quad \gamma \ge 0 \end{cases}$$

**Theorem 1** If  $w^{(1)} \neq w^{(2)}$ , then there exists index *i* satisfy  $x_i = 0$  and  $y_i < 0$ . *Proof.* Because  $w^{(1)} \neq w^{(2)}$ , we have  $f(w^{(1)}) > f(w^{(2)})$ , and there is at least one index *i* satisfy  $y_i < 0$ . Define  $I = \{i | y_i < 0\} \neq \phi$ .

Using Lemma 1 and Lemma 2 , by calculation it is easy to get

$$\sum_{i=1}^{i} y_i \gamma_i = \sum_{i \in I} y_i \gamma_i + \sum_{i \notin I} y_i \gamma_i < 0.$$

Since  $\sum_{i \notin I} y_i \gamma_i \ge 0$ , therefore  $\sum_{i \in I} y_i \gamma_i < 0$ .

Thus there is at least one index  $i \in I$  makes  $\gamma_i > 0$ , and further from Lemma 2, we know that the corresponding  $x_i = 0$ .

For  $x_i = 0$ , we call the securities i is superfluous for model (1).

Without loss of generality, we assume the vector  $\gamma$  in Lemma 2 satisfy  $\gamma_i = 0$  (i = 1, 2, ..., m) and  $\gamma_i > 0$  (i = m + 1, m + 2, ..., n). Accordingly, denote the optimal solution of model (1) by  $w^{(1)} \equiv ((x^{(1)})^T, (x^{(2)})^T)^T$ , where  $x^{(1)} = (x_1, x_2, \dots, x_m)^T$ ,  $x^{(2)} = (x_{m+1}, x_{m+2}, \dots, x_n)^T$ . Then we have the following conclusions.

**Theorem 2** Let  $A_1 = \begin{pmatrix} r_1 & r_2 & \cdots & r_m \\ 1 & 1 & \cdots & 1 \end{pmatrix}, \quad V_{11} = (\sigma_{ij})_{m \times m}$ , then

 $x_1 \ge 0, x_2 \ge 0, \dots, x_m \ge 0; x_{m+1} = 0, x_{m+2} = 0, \dots, x_n = 0;$  and  $x^{(1)} = 0, \dots, x_n = 0$  $(x_1, x_2, \ldots, x_m)^T$  is the optimal solution of the following model (3)

$$\begin{cases} Min \ f_1(w_1, w_2, \dots, w_m) = (w_1, w_2, \dots, w_m) V_{11}(w_1, w_2, \dots, w_m)^T \\ s. t. \ A_1(w_1, w_2, \dots, w_m)^T = b \end{cases}$$
(3)

*Proof.* According to Lemma 2, the  $x_1 \ge 0, x_2 \ge 0, \ldots, x_m \ge 0; x_{m+1} =$  $0, x_{m+2} = 0, \dots, x_n = 0$  is evident, and  $2Vw^{(1)} = A^T\beta + \gamma$  and  $Aw^{(1)} = b$ , through a simple calculation we know that

 $2V_{11}x^{(1)} = A_1^T\beta$  and  $A_1x^{(1)} = b$ It is obtained from lemma 1 that  $x^{(1)} = (x_1, x_2, \dots, x_m)^T$  is the optimal solution of model (3).

**Theorem 3** When the number of securities in the portfolio is reduced, the risk of the optimal portfolio will not fall.

*Proof.* Without loss of generality, we consider model (2) and model (3)(0 <m < n).Let  $w^{(2)} = (y_1, y_2, \dots, y_n)^T$  be the optimal solution of model (2),  $w^{(3)} = (\bar{y}_1, \bar{y}_2, \dots, \bar{y}_m)^T$  be the optimal solution of model (3). Because  $((w^{(3)})^T, 0)^T$  is the feasible solution of model (2), so we have

 $f(w^{(2)}) = (w^{(2)})^T V w^{(2)} \le ((w^{(3)})^T, 0) V ((w^{(3)})^T, 0)^T = (w^{(3)})^T V_{11} w^{(3)} = f_1(w^{(3)}).$ Remark: Theorem 2 and Theorem 3 show that, remove the superfluous securities of model (1) from model (2), the simplified model (2) [That is model (3), and without regard the non-negative of  $w_i$  can contribute the optimal solution of model (1).

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Received: July, 2012