SOLVING FOR THE EFFICIENT FRONTIER

Recall T is point on the efficient frontier in most counterclockwise directions. Equation of line Tangent at T is:

$$\overline{R}_{i} = R_{F} + \frac{\overline{R}_{T} - R_{F}}{\sigma_{T}} \sigma_{i}$$

Moving in most counter-clockwise directions is equivalent to maximizing slope. Thus, the objective is to find portfolio where:

$$\theta = \frac{\overline{R}_{T} - R_{F}}{\sigma_{T}}$$

is maximum.

$$\Theta = \frac{\sum X_{i} \overline{R}_{i} - 1R_{F}}{\left[\sum X_{i}^{2} \sigma_{i}^{2} + \sum \sum X_{i} X_{j} \sigma_{ij}\right]^{1/2}}$$

Subject to
$$\Sigma X_{1} = 1$$

$$\Theta = \frac{\sum_{i} \overline{R}_{i} - \left[\sum_{i} R_{F}\right]}{\left[\sum_{i} x_{i}^{2} \sigma_{i}^{2} + \sum_{i} x_{j} \sigma_{ij}\right]^{1/2}}$$

$$= \frac{\sum X_{i} \left[\overline{R}_{i} - R_{F} \right]}{\left[\sum X_{i}^{2} \sigma_{i}^{2} + \sum \sum X_{i} X_{j} \sigma_{ij} \right]^{1/2}}$$

 Θ can be written as:

$$\Theta = \left(\overline{R}_T - R_F\right) \left(\sigma^2\right)^{-\frac{1}{2}}$$

recall how to take a derivative using the chain rule namely:

$$d\Theta = f(X,Y) = XdY + YdX$$

and that

$$dX^{N} = NX^{N-1}$$

thus

$$\frac{d\Theta}{dX_{k}} = \left(\overline{R}_{T} - R_{F}\right)\left(-\frac{1}{2}\right)\left(\sigma^{2}\right)^{-\frac{3}{2}}\frac{d\sigma^{2}}{dX_{k}} + \frac{d\left[\overline{R}_{T} - R_{F}\right]\left(\sigma^{2}\right)^{-\frac{1}{2}}}{dX_{k}}$$

or

$$\frac{d\Theta}{dX_k} = \left[\overline{R}_T - \overline{R}_F \right] - \frac{1}{2} \left[\frac{3}{2} \left[2X_k \sigma_k^2 + 2\sum_{k \neq i} X_k \sigma_{ik} \right] + \left[\overline{R}_k - \overline{R}_F \left(\sigma^2 \right)^{-1/2} \right] \right]$$

$$\frac{d\Theta}{dX_{k}} = -\frac{\overline{R}_{T} - R_{F}}{\sigma^{2}} \left[X_{k} \sigma_{k}^{2} + \sum_{k \neq i} X_{k} \sigma_{ik} \right] + \left(\overline{R}_{k} - R_{F} \right)$$

$$\left(\overline{R}_{k} - R_{F}\right) = \lambda \left[X_{k}\sigma_{k}^{2} + \sum_{k \neq i} X_{k}\sigma_{ik}\right]$$

where:

$$\lambda = \frac{\overline{R}_{T} - R_{F}}{\sigma^{2}}$$

First order conditions

$$\overline{R}_k - R_F = \lambda \left[X_1 \sigma_{1k} + X_2 \sigma_{2k} + ... X_k \sigma_k^2 + ... + X_N \sigma_{Nk} \right]$$
 for all i

where
$$\lambda = \frac{\overline{R}}{\sigma_T^2} \frac{-R}{F}$$

define
$$Z_1 = \lambda X_1$$
 then

$$\overline{R}_k - R_F = Z_1 \sigma_{ik} + Z_2 \sigma_{2k} + \dots Z_k \sigma_k^2 + \dots + Z_N \sigma_{Nk} \quad \text{ all } k$$

Example 3 securities:

$$\overline{R}_{1} - R_{F} = Z_{1}\sigma_{1}^{2} + Z_{2}\sigma_{21} + Z_{3}\sigma_{31}$$

$$\overline{R}_{2} - R_{F} = Z_{1}\sigma_{21} + Z_{2}\sigma_{2}^{2} + Z_{3}\sigma_{32}$$

$$\overline{R}_{3} - R_{F} = Z_{1}\sigma_{31} + Z_{2}\sigma_{23} + Z_{3}\sigma_{3}^{2}$$

X's are
$$X_{\underline{i}} = \frac{Z_{\underline{i}}}{\sum_{\underline{j}} Z_{\underline{j}}}$$

$$\overline{R} = \begin{bmatrix} \overline{R} \\ \overline{R} \\ \overline{R} \\ 2 \\ \overline{R} \\ 3 \end{bmatrix}$$

$$\Sigma = \begin{bmatrix} \sigma_1^2 \sigma_1 \sigma_1 \\ 1 & 12 & 13 \\ \sigma_1 \sigma_2^2 \sigma_2 \\ \sigma_1 \sigma_2 \sigma_3 \\ \sigma_1 \sigma_2 \sigma_3 \end{bmatrix}$$

$$1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

$$Z = \begin{bmatrix} Z \\ 1 \\ Z \\ Z \\ 3 \end{bmatrix}$$

$$\overline{R} - R_F 1 = \Sigma Z$$

and

$$Z=\Sigma^{-1}\overline{R}+R_{F}\Sigma^{-1}1$$

Typical Element:

$$Z_i = C_0 + C_1 R_F$$

Thus

Short Sales Not Allowed

Minimize σ^2

Subject to:

1.
$$\Sigma X_{1} = 1$$

2.
$$\Sigma X_i \overline{R}_i = \overline{R}_p$$

3.
$$\Sigma X_1 \ge 0$$

Quadratic Programming Problem

Since have squared and cross product terms.