

## Handout 6: Portfolio Variance with Many Risky Assets

Corporate Finance, Sections 001 and 002

### Case 1: Unsystematic risk only.

Recall that when the correlation  $\rho$  between two securities equals zero, the portfolio variance is given by:

$$\sigma_p^2 = X_1^2 \sigma_1^2 + X_2^2 \sigma_2^2$$

A simple generalization of this formula holds for many securities *provided that*  $\rho = 0$  *between all pairs of securities*:

$$\sigma_p^2 = X_1^2 \sigma_1^2 + X_2^2 \sigma_2^2 + \cdots + X_N^2 \sigma_N^2. \quad (1)$$

We will prove the following result. As  $N \rightarrow \infty$ , the portfolio standard deviation  $\sigma_p \rightarrow 0$ .

To make the notation simpler, assume that  $\sigma_1 = \sigma_2 = \cdots = \sigma_N = \sigma$ . This means that each asset is equally risky. Under those circumstances, we try the simple diversification strategy of dividing our wealth equally among each asset such that  $w_i = \frac{1}{N}$ .

These assumptions allow us to rewrite expression (1) as

$$\sigma_p^2 = \left(\frac{1}{N}\right)^2 \sigma^2 + \left(\frac{1}{N}\right)^2 \sigma^2 + \cdots + \left(\frac{1}{N}\right)^2 \sigma^2 \quad (2)$$

There are  $N$  identical terms in expression (2), which means:

$$\begin{aligned} \sigma_p^2 &= N \left(\frac{1}{N}\right)^2 \sigma^2 \\ \sigma_p^2 &= \frac{1}{N} \sigma^2 \end{aligned}$$

The expression in (3) shows that as  $N$  grows larger and larger, the variance the portfolio declines. As  $N \rightarrow \infty$ , the variance goes to zero.

### Case 2: Systematic and unsystematic risk

In fact, U.S. stocks do not have zero correlation with one another. We can capture the positive correlation of U.S. stocks with a factor model. Let  $R_i$  denote the return on an

individual stock, and  $R_M$  the return on a broad market index like the S&P 500. One way to capture the common source of variation is to run a regression of the values of  $R_i$  on  $R_M$ :

$$R_i = \alpha_i + \beta_i R_M + \epsilon_i \quad (3)$$

This is equivalent to fitting a line through a scatter plot of pairs of returns  $(R_M, R_i)$ . The slope of the line equals  $\beta_i$ . You may have heard in your statistics class that

$$\beta_i = \frac{\text{Cov}(R_i, R_M)}{\sigma_M^2}. \quad (4)$$

The coefficient  $\beta_i$  is a measure of how much the stock moves together with the market index  $R_M$ . The error term,  $\epsilon_i$ , measures the variability in  $R_i$  that is independent of all other securities in  $R_M$ .

Using (3), we can decompose the variance of a stock into its systematic and unsystematic components:

$$\begin{aligned} \sigma_i^2 &= \beta_i^2 \sigma_M^2 + \sigma_\epsilon^2 \\ \text{Total risk} &= \text{Systematic risk} + \text{Idiosyncratic risk} \end{aligned} \quad (5)$$

Here,  $\sigma_M^2$  is the variance of  $R_M$ . Equation (5) follows from the fact that  $\epsilon$  and  $R_M$  are independent random variables.

Equation (5) shows that U.S. stocks have both systematic risk ( $\beta_i^2 \sigma_M^2$ ) and unsystematic risk ( $\sigma_\epsilon^2$ ). The argument from Case 1 demonstrates that the unsystematic component of stock risk goes away in a well-diversified portfolio (i.e. a portfolio with a large number of securities  $N$ ). Only the systematic component remains.