

1 The Summation Operator

Because many elementary propositions in econometrics involve the use of sums of numbers, it should be useful to review the summation operator, i.e. Σ . Assume a random variable (henceforth r.v) denoted X from which a sample of n quantities are observed, i.e., $X_i, i = 1, \dots, n$.

Then the total sum of the observations ($X_1 + X_2 + \dots + X_n$) can be represented as

$$\sum_{i=1}^n X_i = X_1 + X_2 + \dots + X_n \quad (1)$$

The following summation operator rules are useful.

Rule 1. The summation of a constant k times a r.v X_i is equal to the constant times the summation of that r.v.

$$\sum_{i=1}^n kX_i = k \sum_{i=1}^n X_i \quad (2)$$

PROOF)

$$\sum_{i=1}^n kX_i = kX_1 + kX_2 + \dots + kX_n = k(X_1 + X_2 + \dots + X_n) = k \sum_{i=1}^n X_i$$

Rule 2. The summation of the sum of observations on two r.v's is equal to the sum of their summations.

$$\sum_{i=1}^n (X_i + Y_i) = \sum_{i=1}^n X_i + \sum_{i=1}^n Y_i \quad (3)$$

PROOF)

$$\begin{aligned} \sum_{i=1}^n (X_i + Y_i) &= (X_1 + Y_1) + (X_2 + Y_2) + \dots + (X_n + Y_n) \\ &= (X_1 + X_2 + \dots + X_n) + (Y_1 + Y_2 + \dots + Y_n) = \sum_{i=1}^n X_i + \sum_{i=1}^n Y_i \end{aligned}$$

Rule 3. The summation of a constant over n observations equals the product of the constant and n .

$$\sum_{i=1}^n k = nk \quad (4)$$

PROOF)

$$\sum_{i=1}^n k = \underbrace{k + k + \dots + k}_{n \text{ times}} = nk$$

Using the above 3 rules, useful results concerning the mean, variance, and covariance of r.v's can be obtained. First, let's define the average of n observations of the r.v X as

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \quad (5)$$

Using this definition, we can prove Rule 4.

Rule 4. The summation of the deviations of observations on X about its mean is zero.

$$\sum_{i=1}^n (X_i - \bar{X}) = 0 \quad (6)$$

Lowercase letters is used to represent deviations form, that is, $x_i = X_i - \bar{X}$. Hence,

$$\sum_{i=1}^n x_i = 0 \quad (7)$$

PROOF)

$$\bar{x} = \frac{\sum x_i}{n} = \frac{\sum (X_i - \bar{X})}{n} = \frac{\sum X_i}{n} - \bar{X} = \bar{X} - \bar{X} = 0$$

Now, let's define the variance of X to be

$$Var(X) = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 \quad (8)$$

and the covariance of X and Y to be

$$Cov(X, Y) = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y}) \quad (9)$$

Using these definitions and earlier results, we can prove two more summation rules.

Rule 5. The covariance between X and Y is equal to the mean of the products of observations on X and Y minus the product of their means.

$$\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y}) = \frac{1}{n} \sum_{i=1}^n X_i Y_i - \bar{X}\bar{Y} \quad (10)$$

PROOF)

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y}) &= \frac{1}{n} \sum_{i=1}^n X_i Y_i - \frac{1}{n} \sum_{i=1}^n \bar{X} Y_i \\ &\quad - \frac{1}{n} \sum_{i=1}^n X_i \bar{Y} + \frac{1}{n} \sum_{i=1}^n \bar{X} \bar{Y} \end{aligned}$$

and using Rule 1, we get

$$Cov(X, Y) = \frac{1}{n} \sum_{i=1}^n X_i Y_i - \frac{1}{n} \bar{X} \sum_{i=1}^n Y_i - \frac{1}{n} \bar{Y} \sum_{i=1}^n X_i + \frac{1}{n} \sum_{i=1}^n \bar{X} \bar{Y}$$

Now, recalling the definition of the mean of X and the mean of Y ,

$$\begin{aligned} Cov(X, Y) &= \frac{1}{n} \sum_{i=1}^n X_i Y_i - \bar{X} \bar{Y} - \bar{Y} \bar{X} + \frac{1}{n} \sum_{i=1}^n \bar{X} \bar{Y} \\ &= \frac{1}{n} \sum_{i=1}^n X_i Y_i - 2\bar{X} \bar{Y} + \bar{X} \bar{Y} \\ &\text{since } \sum_{i=1}^n \bar{X} \bar{Y} = n\bar{X} \bar{Y} \text{ by Rule 3} \\ &= \frac{1}{n} \sum_{i=1}^n X_i Y_i - \bar{X} \bar{Y} \end{aligned}$$

Rule 6 follows easily from Rule 5, since it applies to the case in which X and X again are two variables.

Rule 6. The variance of X is equal to the mean of the squares of observations on X minus its mean squared.

$$\frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 = \frac{1}{n} \sum_{i=1}^n X_i^2 - \bar{X}^2 \quad (11)$$

Note, incidentally, that when X and Y happen to have zero means (as occurs when they are measured in deviations about their means), the definitions of covariance and variance become

$$Cov(x, y) = \frac{1}{n} \sum_{i=1}^n x_i y_i \quad \text{and} \quad Var(x) = \frac{1}{n} \sum_{i=1}^n x_i^2$$

In certain situations it will be necessary to use summations which apply to two random variables, called *double summations*. Specifically, let X_{ij} be a r.v. which takes on n values for each outcome of i and j . There will, of course, be n^2 total outcomes. Now we define the double summation of these n^2 outcomes as

$$\begin{aligned} \sum_{i=1}^n \sum_{j=1}^m X_{ij} &= \sum_{i=1}^n (X_{i1} + X_{i2} + \cdots + X_{im}) \\ &= (X_{11} + X_{12} + \cdots + X_{1m}) + (X_{21} + X_{22} + \cdots + X_{2m}) \\ &\quad + \cdots + (X_{n1} + X_{n2} + \cdots + X_{nm}) \end{aligned}$$

The following two double-summation rules will be useful.

Rule 7.

$$\sum_{i=1}^n \sum_{j=1}^m X_i Y_j = \left(\sum_{i=1}^n X_i \right) \left(\sum_{j=1}^m Y_j \right) \quad (12)$$

Note that the double summation in Rule 7 is very different from the single summation $\sum_{i=1}^n X_i Y_i$, which contains n (rather than n^2) terms.

Rule 8.

$$\sum_{i=1}^n \sum_{j=1}^m (X_{ij} + Y_{ij}) = \sum_{i=1}^n \sum_{j=1}^m X_{ij} + \sum_{i=1}^n \sum_{j=1}^m Y_{ij} \quad (13)$$