

# Univariate and Multivariate Summary Statistics

## A Review of Univariate Summary Statistics

### Random Sample

Let  $x_1, x_2, \dots, x_N$  be a random sample  
from some distribution.

Random Sample  
expressed as a vector

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{bmatrix}$$

Sample Mean

$$\hat{\mu} = \bar{x} = \frac{1}{N} \sum_{r=1}^N x_r = \frac{\mathbf{x}'\mathbf{1}}{N} = \frac{\mathbf{x}'\mathbf{1}}{\mathbf{1}'\mathbf{1}} = \mathbf{x}'\mathbf{1}(\mathbf{1}'\mathbf{1})^{-1}$$

## Sample Variance

$$s^2 = \frac{1}{N-1} \left[ \sum_{r=1}^N (x_r - \hat{\mu})^2 \right]$$

## Sample Variance in matrix notation

The deviation of each observation  
from the sample mean is

$$\begin{aligned} \mathbf{x} - \hat{\mu}\mathbf{1} &= \mathbf{x} - \mathbf{1}\bar{x} = \mathbf{x} - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1} \mathbf{1}'\mathbf{x} \\ &= \mathbf{I}_N \mathbf{x} - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1} \mathbf{1}'\mathbf{x} \\ &= \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1} \mathbf{1}' \right) \mathbf{x} \end{aligned}$$

## Sample Variance in matrix notation

Alternatively, the deviation of each observation from the sample mean can be expressed as

$$\mathbf{x} - \hat{\mu}\mathbf{1} = \mathbf{x} - \mathbf{1}\bar{X} = \left( \mathbf{I}_N - \frac{1}{N}\mathbf{J}_N \right) \mathbf{x}$$

## Sample Variance in matrix notation

The sum of the squared deviations of each observation from the sample mean is

$$\begin{aligned} & \left( \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1}\mathbf{1}' \right) \mathbf{x} \right)' \left( \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1}\mathbf{1}' \right) \mathbf{x} \right) \\ &= \mathbf{x}' \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1}\mathbf{1}' \right)^2 \mathbf{x} \\ &= \mathbf{x}' \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1}\mathbf{1}' \right) \mathbf{x} \end{aligned}$$

## Sample Variance in matrix notation

$$\begin{aligned} s^2 &= \frac{1}{N-1} \mathbf{x}' \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1} \mathbf{1}' \right) \mathbf{x} \\ &= \frac{1}{N-1} \mathbf{x}' \left( \mathbf{I}_N - \frac{1}{N} \mathbf{J}_N \right) \mathbf{x} \end{aligned}$$

## Multivariate Observations

Let

- $p$  = number of response variables
- $N$  = number of experimental units
- $x_{rj}$  = value of the  $j^{\text{th}}$  response variable on the  $r^{\text{th}}$  experimental unit

## Data Matrix

$$\mathbf{X}_{N \times p} = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1p} \\ x_{21} & x_{22} & \cdots & x_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ x_{N1} & x_{N2} & \cdots & x_{Np} \end{bmatrix} = (x_{rj})$$
$$= \begin{bmatrix} \mathbf{x}'_1 \\ \mathbf{x}'_2 \\ \vdots \\ \mathbf{x}'_N \end{bmatrix}$$

## A Multivariate Observation

$$\mathbf{x}'_r = (x_{r1} \quad x_{r2} \quad \cdots \quad x_{rp})$$

## Multivariate Summary Statistics

### Random Sample

Let  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N$  be a random sample from a multivariate distribution.

### Sample Mean Vector

$$\begin{aligned}\hat{\boldsymbol{\mu}} = \bar{\mathbf{x}} &= \frac{1}{N} \sum_{r=1}^N \mathbf{x}_r = \frac{1}{N} \mathbf{X}'\mathbf{1} \\ &= \mathbf{X}'\mathbf{1}(\mathbf{1}'\mathbf{1})^{-1}\end{aligned}$$

## Sample Variance-Covariance Matrix

$$\begin{aligned}\hat{\Sigma} = \mathbf{S} &= \frac{1}{N-1} \left[ \sum_{r=1}^N (\mathbf{x}_r - \hat{\boldsymbol{\mu}})(\mathbf{x}_r - \hat{\boldsymbol{\mu}})' \right] \\ &= \frac{1}{N-1} \mathbf{X}' \left( \mathbf{I}_N - \mathbf{1}(\mathbf{1}'\mathbf{1})^{-1} \mathbf{1} \right) \mathbf{X} \\ &= \frac{1}{N-1} \mathbf{X}' \left( \mathbf{I}_N - \frac{1}{N} \mathbf{J}_N \right) \mathbf{X}\end{aligned}$$

## Sample Correlation Matrix

$$\mathbf{R} = \begin{bmatrix} 1 & r_{12} & \cdots & r_{1p} \\ r_{12} & 1 & \cdots & r_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ r_{1p} & r_{2p} & \cdots & 1 \end{bmatrix}$$

## The Sample Correlation Matrix

$$\text{Let } \mathbf{D}_s = \text{diag} \left( \sqrt{s_{11}}, \sqrt{s_{22}}, \dots, \sqrt{s_{pp}} \right)$$

Then the Sample Correlation Matrix  $\mathbf{R}$  is given by

$$\mathbf{R} = \left( r_{ij} \right) = \mathbf{D}_s^{-1} \mathbf{S} \mathbf{D}_s^{-1}$$