

Linear Combinations of Random Variables¹

UW

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¹HMS, 2017, v1.0

Chapter References

- ▶ Navidi, Chapter 2.5

Linear Combinations: Adding a Constant

- ▶ Adding a Constant value, for example 2.0:

Consider: 2, 6, 4, 1 $\Rightarrow \mu = 3.25$

Consider: 4, 8, 6, 2 $\Rightarrow \mu = 5.25$

- ▶ Mean has also increased by 2.0. In general:

$$\mu_{X+a} = \mu_X + a \quad \text{or} \quad E(X + a) = E(X) + a$$

Linear Combinations: Adding a Constant

- ▶ Adding a Constant value, for example 2.0:

Consider: 2, 6, 4, 1 $\Rightarrow \sigma^2 = 3.69$

Consider: 4, 8, 6, 2 $\Rightarrow \sigma^2 = 3.69$

- ▶ Variance is unchanged. In general:

$$\sigma_{X+a}^2 = \sigma_X^2 \quad \text{or} \quad \text{Var}(X + a) = \text{Var}(X)$$

Linear Combinations: Multiply by a Constant

- ▶ Multiply by a Constant value, for example 2.0:

Consider: 2, 6, 4, 1 $\Rightarrow \mu = 3.25$

Consider: 4, 8, 6, 2 $\Rightarrow \mu = 6.5$

- ▶ Mean has change by the same factor. In general:

$$\mu_{aX} = a\mu_X \quad \text{or} \quad E(aX) = aE(X)$$

Linear Combinations: Multiply by a Constant

- ▶ Multiply by a Constant value, for example 2.0:

Consider: 2, 6, 4, 1 $\Rightarrow \sigma^2 = 3.69$

Consider: 4, 8, 6, 2 $\Rightarrow \sigma^2 = 14.76$

- ▶ Variance has changed by a factor of the square of the constant. In general:

$$\sigma_{aX}^2 = a^2 \sigma_X^2 \quad \text{or} \quad \text{Var}(aX) = a^2 \text{Var}(X)$$

Linear Combinations: Add and Multiply by a Constant

$$\mu_{aX+b} = a\mu_X + b \quad \text{or} \quad E(aX + b) = aE(X) + b$$

$$\sigma_{aX+b}^2 = a^2\sigma_X^2 \quad \text{or} \quad \text{Var}(aX + b) = a^2\text{Var}(X)$$

Linear Combinations of Means

- ▶ The mean of $X_1 + X_2 + \dots$ is

$$\mu_{X_1} + \mu_{X_2} + \dots = \mu_{X_1+X_2+\dots}$$

- ▶ In general:

$$c_1X_1 + c_2X_2 + \dots = c_1\mu_{X_1} + c_2\mu_{X_2} + \dots$$

- ▶ In expectation form:

$$E(c_1X_1 + c_2X_2 + \dots) = c_1E(X_1) + c_2E(X_2) + \dots$$

Linear Combinations of Variances

- ▶ The variance of $X_1 + X_2 + \dots$ is

$$\sigma_{X_1}^2 + \sigma_{X_2}^2 + \dots = \sigma_{X_1+X_2+\dots}^2$$

Note you **cannot** add standard deviations

- ▶ In general:

$$\sigma_{c_1X_1+c_2X_2+\dots}^2 = c_1^2\sigma_{X_1}^2 + c_2^2\sigma_{X_2}^2 + \dots$$

- ▶ In Var form:

$$\text{Var}(c_1X_1 + c_2X_2 + \dots) = c_1^2\text{Var}(X_1) + c_2^2\text{Var}(X_2) + \dots$$

Application

- ▶ Find the variance of a sample of means

$$\text{Var}(\bar{X}) = \text{Var}\left(\frac{1}{n} \sum X_i\right)$$

- ▶ Use $\text{Var}(aX) = a^2\text{Var}(X)$:

$$\text{Var}(\bar{X}) = \frac{1}{n^2} \text{Var}\left(\sum X_i\right)$$

- ▶ Assume X_i are identically distributed with a common variance σ^2 , then by $\text{Var}(X_1 + X_2 + \dots) = \text{Var}(X_1) + \text{Var}(X_2) + \dots$

$$\text{Var}(\bar{X}) = \frac{1}{n^2} \sum (\text{Var}(X_i)) = \frac{1}{n^2} n\sigma^2 = \frac{\sigma^2}{n}$$

Application

- ▶ The variance of the sample of mean is

$$\text{Var}(\bar{X}) = \frac{\sigma^2}{n}$$

- ▶ The standard deviation of this is called the standard error:

$$\text{SE} = \frac{\sigma}{\sqrt{n}}$$

- ▶ Standard deviation measures: Precision (how bunched up your measurements are)
- ▶ Standard error measures: Accuracy (assuming there is no fixed bias)

Example

- ▶ A researcher dispenses a medium into 96 wells using a manual single tip pipette.
- ▶ Ten well plates are selected at random and the mean volume and standard deviation measured:

$$\bar{x} = 1.05 \text{ ml} \quad \sigma = 0.01 \text{ ml}$$

- ▶ The standard error is given by:

$$SE = \frac{\sigma}{\sqrt{n}} = \frac{0.01}{\sqrt{10}} = 0.0032 \text{ ml}$$

- ▶ Mean = 1.05 ± 0.0032