

MS04 Support Material

Estimators

For use with AQA A-level Mathematics Specification (6360)

1 Terminology

1.1 Sample Statistic

A numerical measure, calculated from a sample, is called a **sample statistic**.

Thus the following are examples of sample statistics:

$$\bar{x} \quad s^2 \quad r \quad a \quad b \quad \hat{p}$$

1.2 Population parameter

A numerical value that defines a population is called a **population parameter**.

Thus the following are examples of population parameters:

$$\mu \quad \sigma^2 \quad \rho \quad \alpha \quad \beta \quad p$$

It is important to remember that these are **not** random variables.

1.3 Estimator and Estimate

The **random variable** of a sample statistic is called an **estimator** for a population parameter.

A **value** of a sample statistic is called an **estimate** of a population parameter.

Thus \bar{X} and S^2 are estimators for μ and σ^2 , respectively, whereas \bar{x} and s^2 are estimates.

1.4 Sampling Distribution

The distribution of an estimator is called its **sampling distribution**.

Thus, if $X \sim N(\mu, \sigma^2)$, then the distribution $N\left(\mu, \frac{\sigma^2}{n}\right)$ would be the sampling distribution of the estimator, \bar{X} .

2 Properties of Estimators

2.1 Unbiased

The sample statistic T is an **unbiased** estimator for the population parameter θ when:

$$E(T) = \theta$$

2.2 Consistent

The sample statistic T is a **consistent** estimator for the population parameter θ when:

$$V(T) \rightarrow 0 \quad \text{as} \quad n \rightarrow \infty$$

2.3 Relative Efficiency

If R and T are two estimators for the parameter θ , then the **relative efficiency** of R relative to (with respect to) T is given by:

$$\text{RE } (R \text{ wrt } T) = \frac{[V(R)]^{-1}}{[V(T)]^{-1}}$$

3 Exemplar Examination Questions – Questions

3.1 Question 1

A random variable X is distributed with mean μ and standard deviation σ . Three independent observations, X_1 , X_2 and X_3 , of the random variable X are to be taken. The combined statistic:

$$aX_1 + bX_2 + cX_3,$$

where a , b and c are constants, is to be used as an estimator for μ .

- Obtain expressions for the mean and variance of this combined statistic.
- Explain why it is desirable that a , b and c are chosen so that $a + b + c = 1$.
- When $a = b = c = \frac{1}{3}$, the combined statistic is denoted by \bar{X} .
When $a = b = \frac{2}{3}$ and $c = -\frac{1}{3}$, the combined statistic is denoted by T .

Calculate the relative efficiency of \bar{X} with respect to T .

3.2 Question 2

Steel rods in a large batch have lengths that are distributed with mean μ and variance σ^2 . Two trainees are set the task of estimating μ . Trainee *A* intends to take a random sample of 20 rods and calculate the mean length, \bar{X}_A , of the rods in the sample.

- (a) Write down expressions for the mean and variance of \bar{X}_A .
- (b) Trainee *B* intends to take a random sample of 5 rods and calculate the mean length, \bar{X}_B , of the rods in the sample. Their instructor suggests that they could obtain a better estimate of μ by combining their individual estimates. The trainees consider two possible combinations, \bar{X}_M and \bar{X}_W , of their estimators, where:

$$\bar{X}_M = \frac{1}{2}\bar{X}_A + \frac{1}{2}\bar{X}_B \quad \text{and} \quad \bar{X}_W = \frac{4}{5}\bar{X}_A + \frac{1}{5}\bar{X}_B$$

- (i) Show that the estimator \bar{X}_M has mean μ and variance $\frac{\sigma^2}{16}$.
- (ii) State why \bar{X}_M is an unbiased estimator for μ .
- (iii) Show that \bar{X}_W is an unbiased estimator for μ .
- (iv) Calculate the relative efficiency of \bar{X}_M with respect to \bar{X}_W , and give a reason why \bar{X}_W should be preferred to \bar{X}_M as an estimator for μ .

3.3 Question 3

A manufacturing process produces a large number of items of which a proportion p are defective. Alan takes a random sample of n items from the output and records X defective items, whilst Betty independently takes a random sample of $2n$ items from the output and records Y defective items.

- (a) Write down the mean and variance of **each** of X and Y .
- (b) Their supervisor considers two possible estimators for p :

$$\hat{p}_1 = \frac{1}{2} \left(\frac{X}{n} + \frac{Y}{2n} \right) \quad \text{and} \quad \hat{p}_2 = \frac{X+Y}{3n}$$

- (i) Show that **each** of these estimators for p is unbiased.
- (ii) Show that the variance of \hat{p}_1 is $\frac{3p(1-p)}{8n}$.
- (iii) Calculate the relative efficiency of \hat{p}_1 relative to \hat{p}_2 .

3.4 Question 4

The random variable X has a rectangular distribution on the interval $0 \leq x \leq 2\theta$.

(a) State, in terms of θ , the variance of X .

Independent observations, $X_1, X_2, X_3, \dots, X_n$, are taken of the random variable X . The mean of such a sample of observations is denoted by \bar{X} .

(b) Write down, in terms of θ and n , expressions for the mean and variance of \bar{X} .

(c) Explain why \bar{X} is:

(i) an unbiased estimator for θ ,

(ii) a consistent estimator for θ .

(d) For a random sample of size 5, the median, M , is an unbiased estimator for θ with variance $\frac{\theta^2}{7}$.

For such a sample, calculate the relative efficiency of \bar{X} with respect to M , and give a reason why \bar{X} should be preferred to M as an estimator for θ .

(e) A particular random sample of 5 observations of X gave the following values.

1.4 2.5 0.2 0.4 1.3

(i) Use this sample to write down an inequality for 2θ , the upper boundary of the interval for X .

(ii) Obtain the values of \bar{x} and m for this sample.

(iii) Explain which, of \bar{x} and m , is the more appropriate estimate of θ in this case.

4 Exemplar Examination Questions – Answers

4.1 Question 1

(a)

$$E(aX_1 + bX_2 + cX_3) = aE(X_1) + bE(X_2) + cE(X_3)$$

$$= a\mu + b\mu + c\mu$$

$$= \underline{(a + b + c)\mu}$$

$$V(aX_1 + bX_2 + cX_3) = a^2V(X_1) + b^2V(X_2) + c^2V(X_3)$$

$$= a^2\sigma^2 + b^2\sigma^2 + c^2\sigma^2$$

$$= \underline{(a^2 + b^2 + c^2)\sigma^2}$$

(b)

$$a + b + c = 1 \quad \Rightarrow \quad E(aX_1 + bX_2 + cX_3) = \mu$$

Thus combined statistic is an unbiased estimator for μ

(c)

$$V(\bar{X}) = \left(\frac{1}{9} + \frac{1}{9} + \frac{1}{9} \right) \sigma^2 = \underline{\frac{\sigma^2}{3}}$$

$$V(T) = \left(\frac{4}{9} + \frac{4}{9} + \frac{(-1)^2}{9} \right) \sigma^2 = \underline{\sigma^2}$$

$$\text{RE}(\bar{X} \text{ wrt } T) = \frac{[V(\bar{X})]^{-1}}{[V(T)]^{-1}} = \frac{\left[\frac{\sigma^2}{3} \right]^{-1}}{[\sigma^2]^{-1}} = \underline{3}$$

4.2 Question 2

(a)

$$E(\bar{X}_A) = \mu \qquad V(\bar{X}_A) = \underline{\underline{\frac{\sigma^2}{20}}}$$

(b)

(i)

$$E(\bar{X}_M) = \frac{1}{2}E(\bar{X}_A) + \frac{1}{2}E(\bar{X}_B) = \frac{1}{2}\mu + \frac{1}{2}\mu = \mu$$

$$V(\bar{X}_M) = \frac{1}{4}V(\bar{X}_A) + \frac{1}{4}V(\bar{X}_B) = \frac{1}{4} \times \frac{\sigma^2}{20} + \frac{1}{4} \times \frac{\sigma^2}{5} = \underline{\underline{\frac{\sigma^2}{16}}}$$

(ii)

$$E(\bar{X}_M) = \mu \Rightarrow \bar{X}_M \text{ is unbiased}$$

(iii)

$$E(\bar{X}_W) = \frac{4}{5}E(\bar{X}_A) + \frac{1}{5}E(\bar{X}_B) = \frac{4}{5}\mu + \frac{1}{5}\mu = \mu \Rightarrow \bar{X}_W \text{ is unbiased}$$

$$V(\bar{X}_W) = \frac{16}{25}V(\bar{X}_A) + \frac{1}{25}V(\bar{X}_B) = \frac{16}{25} \times \frac{\sigma^2}{20} + \frac{1}{25} \times \frac{\sigma^2}{5} = \underline{\underline{\frac{\sigma^2}{25}}}$$

(iv)

$$RE(\bar{X}_M \text{ wrt } \bar{X}_W) = \frac{[V(\bar{X}_M)]^{-1}}{[V(\bar{X}_W)]^{-1}} = \frac{\left[\frac{\sigma^2}{16}\right]^{-1}}{\left[\frac{\sigma^2}{25}\right]^{-1}} = \frac{16}{25} \text{ or } \underline{0.64} \text{ or } \underline{64\%}$$

Prefer \bar{X}_W since $V(\bar{X}_W) < V(\bar{X}_M)$

or $RE(\bar{X}_M \text{ wrt } \bar{X}_W) < 1$ or 100%

4.3 Question 3

(a)

$$E(X) = np \qquad V(X) = np(1-p)$$

$$E(Y) = 2np \qquad V(Y) = 2np(1-p)$$

(b)

(i)

$$E(\hat{p}_1) = \frac{1}{2} \left[\frac{1}{n} E(X) + \frac{1}{2n} E(Y) \right] = \frac{1}{2} [p + p] = p \quad \Rightarrow \quad \text{unbiased}$$

$$E(\hat{p}_2) = \frac{1}{3n} [E(X) + E(Y)] = \frac{1}{3n} [np + 2np] = p \quad \Rightarrow \quad \text{unbiased}$$

(ii)

$$\begin{aligned} V(\hat{p}_1) &= \frac{1}{4} \left[\frac{1}{n^2} V(X) + \frac{1}{4n^2} V(Y) \right] \\ &= \frac{1}{4} \left[\frac{np(1-p)}{n^2} + \frac{2np(1-p)}{4n^2} \right] = \frac{1}{4} \left[\frac{6p(1-p)}{4n} \right] = \frac{3p(1-p)}{8n} \end{aligned}$$

(iii)

$$\begin{aligned} V(\hat{p}_2) &= \frac{1}{9n^2} [V(X) + V(Y)] \\ &= \frac{1}{9n^2} [np(1-p) + 2np(1-p)] = \frac{p(1-p)}{3n} \end{aligned}$$

$$\text{RE}(\hat{p}_1 \text{ wrt } \hat{p}_2) = \frac{[V(\hat{p}_1)]^{-1}}{[V(\hat{p}_2)]^{-1}} = \frac{\left[\frac{3p(1-p)}{8n} \right]^{-1}}{\left[\frac{p(1-p)}{3n} \right]^{-1}} = \frac{8}{9} \text{ or } \underline{0.889} \text{ or } \underline{88.9\%}$$

4.4 Question 4

(a)

$$V(X) = \frac{(2\theta - 0)^2}{12} = \frac{4\theta^2}{12} = \frac{\theta^2}{3}$$

(b)

$$E(\bar{X}) = \theta \qquad V(\bar{X}) = \frac{\theta^2}{3n}$$

(c)

(i)

$$E(\bar{X}) = \theta \Rightarrow \text{unbiased}$$

(ii)

$$V(\bar{X}) = \frac{\theta^2}{3n} \rightarrow 0 \text{ as } n \rightarrow \infty \Rightarrow \text{consistent}$$

(d)

$$\text{RE}(\bar{X} \text{ wrt } M) = \frac{[V(\bar{X})]^{-1}}{[V(M)]^{-1}} = \frac{\left[\frac{\theta^2}{3 \times 5}\right]^{-1}}{\left[\frac{\theta^2}{7}\right]^{-1}} = \frac{15}{7} \text{ or } \underline{2.14} \text{ or } \underline{214\%}$$

Prefer \bar{X} since $V(\bar{X}) < V(M)$

or $\text{RE}(\bar{X} \text{ wrt } M) > 1$ or 100%

(e)

(i)

$$\text{Largest sample value} \leq 2\theta \Rightarrow 2.5 \leq 2\theta \quad \text{or} \quad \underline{2\theta \geq 2.5}$$

(ii)

$$\bar{x} = \underline{1.16} \qquad m = \underline{1.3}$$

(iii)

$$\text{From (i)} \quad \theta \geq 1.25$$

$$\text{Now } m = 1.3 > 1.25 \quad \text{but } \bar{x} = 1.16 < 1.25$$

Thus choose m since:

\bar{x} is an impossible value for θ

or

θ must be closer to m than to \bar{x}

or

\bar{x} is less than 1.25