

Variance

Let X be a **discrete random variable**, then the variance of X is;

$$\text{Var}(X) = \sum (x - \mu)^2 p(x)$$

where $p(x) = P(X = x) \geq 0$

$$\mu = E(X)$$

Note: $x - \mu$ is the **deviation** of x from μ

$\text{Var}(X)$ is a measure of spread

e.g. Using the data of the milk marketing survey, find the variance of the number of litres of milk consumed in a week by a family.

x	0	1	2	3	4	5	Σ
$p(x)$	$\frac{2}{25}$	$\frac{5}{25}$	$\frac{9}{25}$	$\frac{5}{25}$	$\frac{3}{25}$	$\frac{1}{25}$	1
$xp(x)$	0	$\frac{5}{25}$	$\frac{18}{25}$	$\frac{15}{25}$	$\frac{12}{25}$	$\frac{5}{25}$	2.2
$(x - \mu)^2$	$\frac{121}{25}$	$\frac{36}{25}$	$\frac{1}{25}$	$\frac{16}{25}$	$\frac{81}{25}$	$\frac{196}{25}$	—
$(x - \mu)^2 p(x)$	$\frac{242}{625}$	$\frac{180}{625}$	$\frac{9}{625}$	$\frac{80}{625}$	$\frac{243}{625}$	$\frac{196}{625}$	1.52

μ

$$\begin{aligned} \text{Var}(X) &= \sum (x - \mu)^2 p(x) \\ &= \underline{1.52} \end{aligned}$$

alternatively $\text{Var}(X) = \sum (x - \mu)^2 p(x)$

$$= \text{E}[(X - \mu)^2]$$

$$= \text{E}(X^2 - 2\mu X + \mu^2)$$

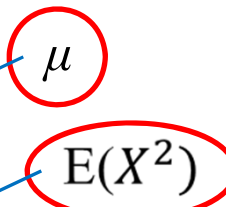
$$= \text{E}(X^2) - 2\mu \text{E}(X) + \mu^2$$

$$= \text{E}(X^2) - 2\mu \times \mu + \mu^2$$

$$\text{Var}(X) = \text{E}(X^2) - \mu^2$$

where $\text{E}(X^2) = \sum x^2 p(x)$

x	0	1	2	3	4	5	Σ
$p(x)$	$\frac{2}{25}$	$\frac{5}{25}$	$\frac{9}{25}$	$\frac{5}{25}$	$\frac{3}{25}$	$\frac{1}{25}$	1
$xp(x)$	0	$\frac{5}{25}$	$\frac{18}{25}$	$\frac{15}{25}$	$\frac{12}{25}$	$\frac{5}{25}$	2.2
$x^2p(x)$	0	$\frac{5}{25}$	$\frac{36}{25}$	$\frac{45}{25}$	$\frac{48}{25}$	$\frac{25}{25}$	6.36



$$\text{Var}(X) = \text{E}(X^2) - \mu^2$$

$$= 6.36 - (2.2)^2 = \underline{1.52}$$

Standard Deviation (σ)

If the random variable, X , is measured in units
then variance ($E(X^2) - \mu^2$) is measured in units².

Standard deviation is used to measure spread using the same units
as the random variable.

$$\sigma = \sqrt{\text{Var}(X)}$$

or

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

e.g. $\sigma = \sqrt{1.52}$

$$= 1.23288\dots$$

$$= \underline{1.23} \text{ (to 2 dp)}$$

Standardising Data

Data standardisation is the process of making sure your data set can be compared to other data sets

Original data

$$E(X) = \mu \quad \text{and} \quad \text{Var}(X) = \sigma^2$$

adding a constant amount to each piece of data

$$Y = X + c$$

$$E(Y) = E(X + c)$$

$$= E(X) + E(c)$$

$$= \mu + c$$

$$\text{Var}(Y) = \text{Var}(X + c)$$

$$= E[(X + c)^2] - (\mu + c)^2$$

$$= E(X^2 + 2cX + c^2) - (\mu^2 - 2c\mu + c^2)$$

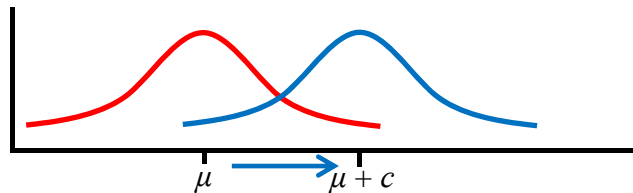
$$= E(X^2) + 2cE(X) + c^2 - \mu^2 + 2c\mu - c^2$$

$$= E(X^2) + 2c\mu + c^2 - \mu^2 + 2c\mu - c^2$$

$$= E(X^2) - \mu^2$$

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

simply adding a constant amount does not change the shape of the data, it only translates it c units



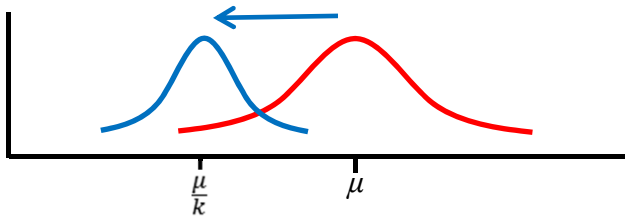
dividing each piece of data by a constant amount

$$Z = \frac{X}{k}$$

$$\begin{aligned} E(Z) &= E\left(\frac{X}{k}\right) \\ &= \frac{E(X)}{k} \\ &= \frac{\mu}{k} \end{aligned}$$

$$\begin{aligned} \text{Var}(Z) &= \text{Var}\left(\frac{X}{k}\right) \\ &= E\left[\left(\frac{X}{k}\right)^2\right] - \left(\frac{\mu}{k}\right)^2 \\ &= \frac{E(X^2)}{k^2} - \frac{\mu^2}{k^2} \\ &= \frac{E(X^2) - \mu^2}{k^2} \\ &= \frac{\text{Var}(X)}{k^2} \\ &= \frac{\sigma^2}{k^2} = \left(\frac{\sigma}{k}\right)^2 \end{aligned}$$

dividing by a constant amount changes the shape of the data, as well as altering the mean



z-scores

A common standardisation is to “normalise” the data by creating a z-score

$$z = \frac{x - \mu}{\sigma}$$

$$E(Z) = E\left(\frac{X - \mu}{\sigma}\right)$$

$$= \frac{E(X - \mu)}{\sigma}$$

$$= \frac{\mu - \mu}{\sigma}$$

$$= 0$$

$$\therefore \mu_Z = 0 \quad \text{and} \quad \sigma_Z = 1$$

$$\text{Var}(Z) = \text{Var}\left(\frac{X - \mu}{\sigma}\right)$$

$$= \frac{\text{Var}(X - \mu)}{\sigma^2}$$

$$= \frac{\sigma^2}{\sigma^2} = 1$$

Exercise 13C; 1, 2, 3bd, 4, 6, 8, 9, 10, 11, 12abc, 13