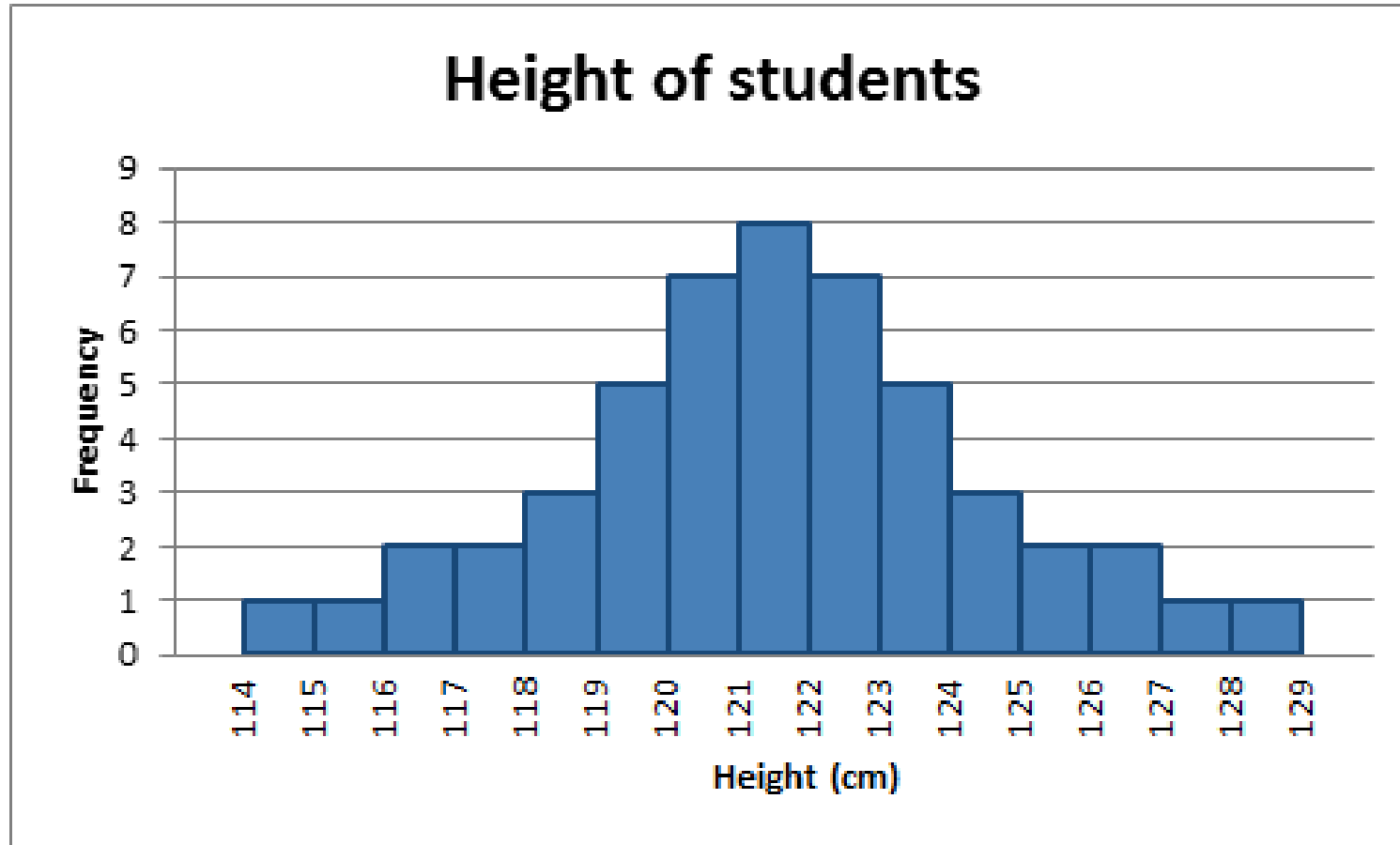
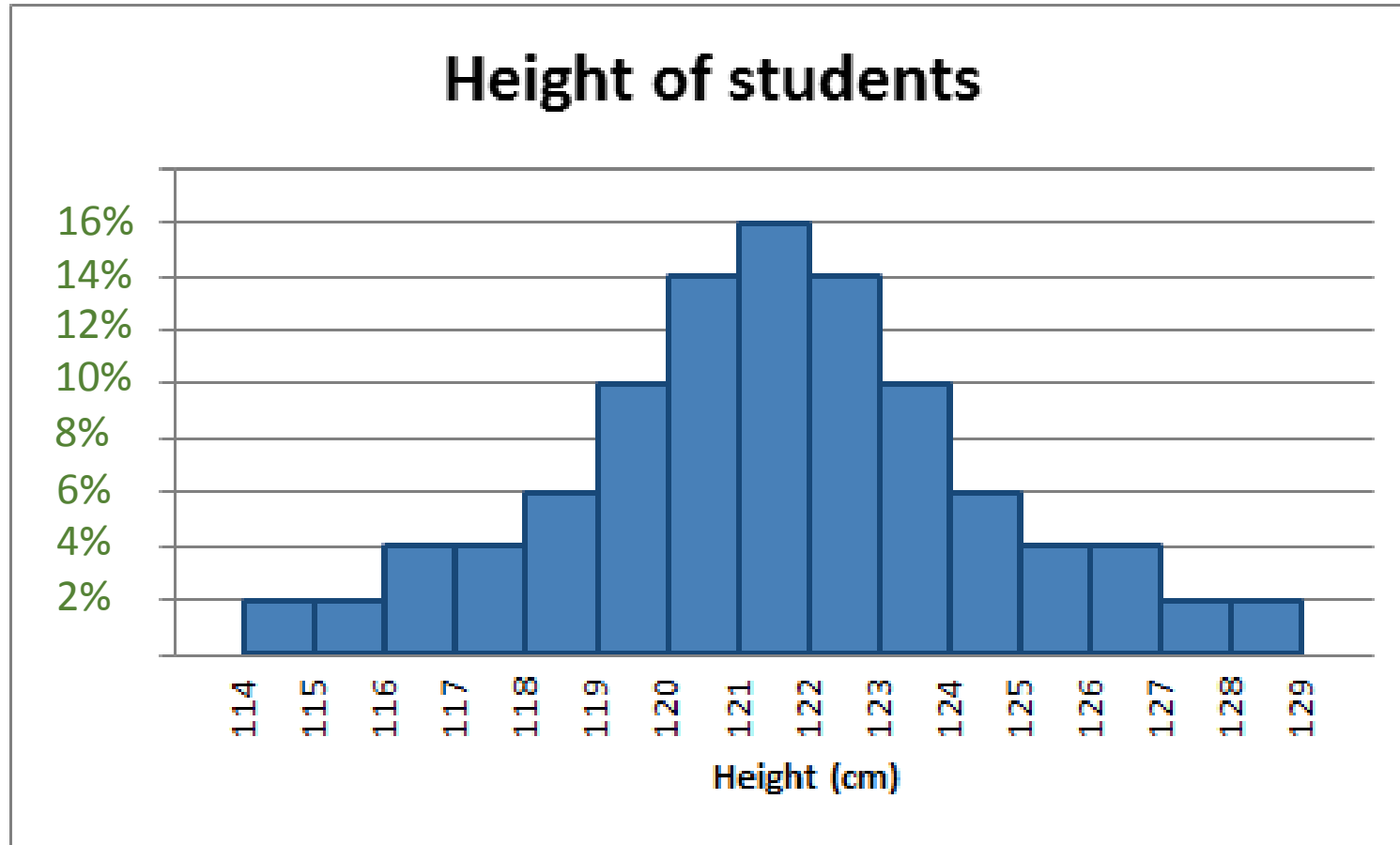


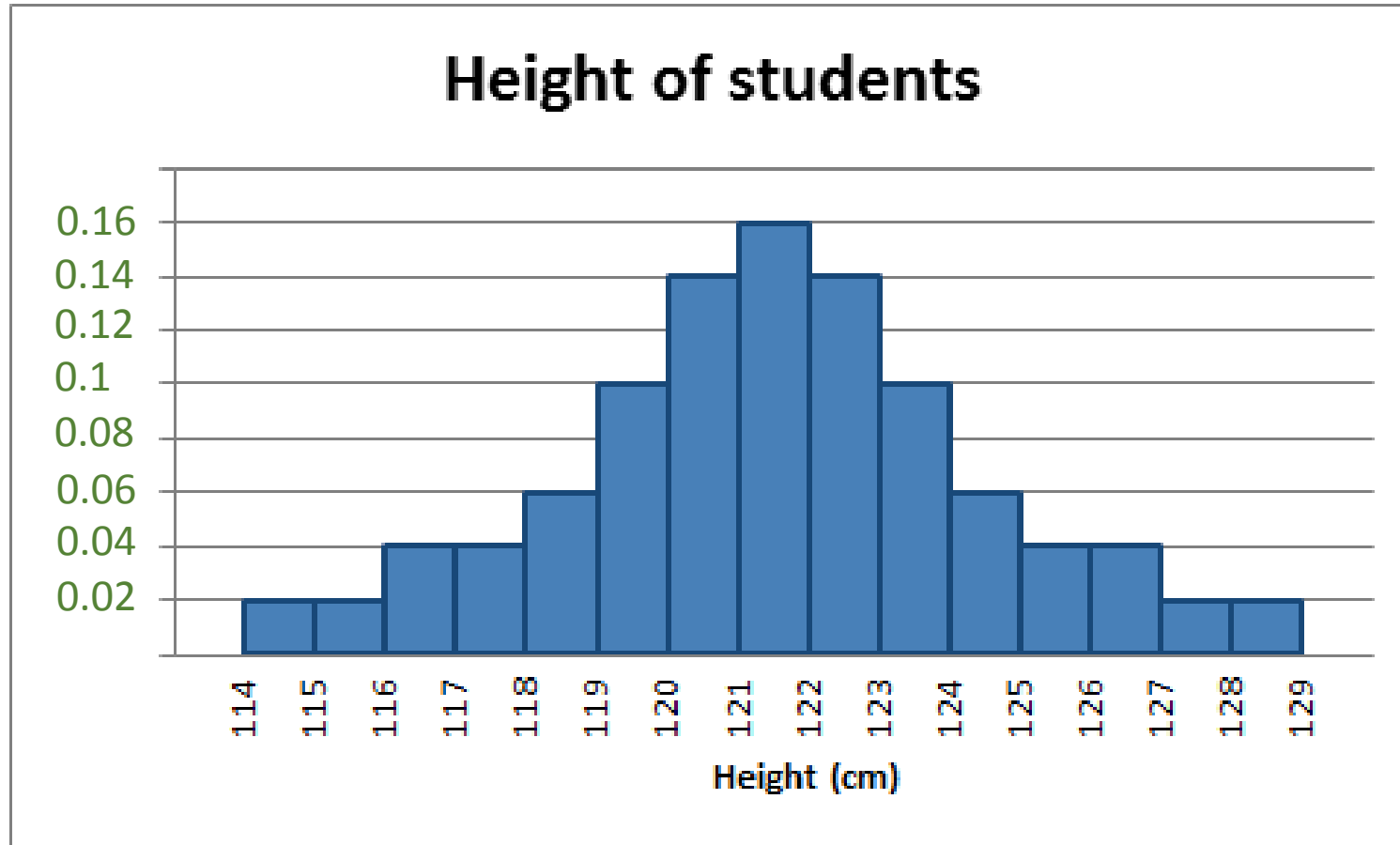
expected value of a random variable



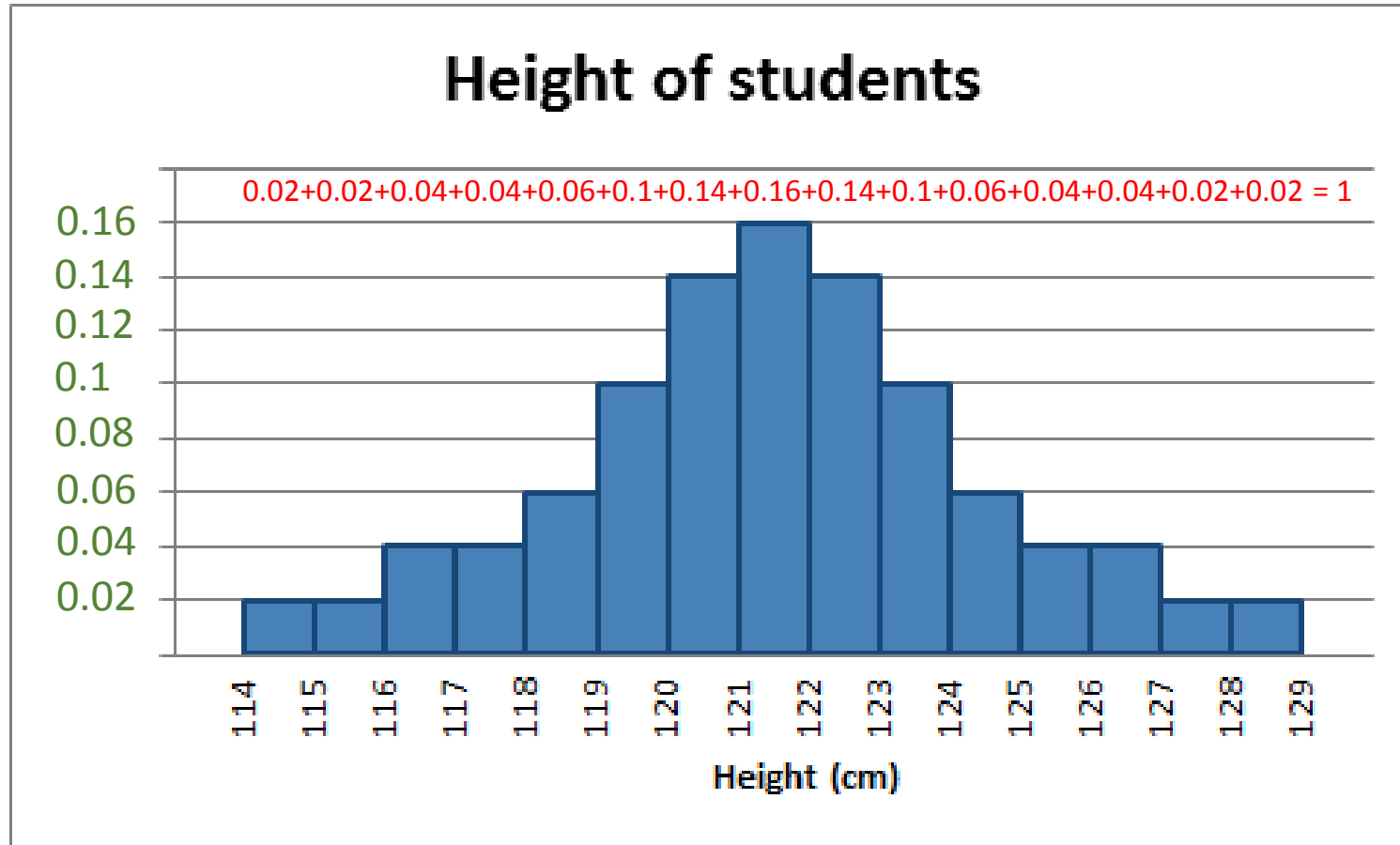
expected value of a random variable



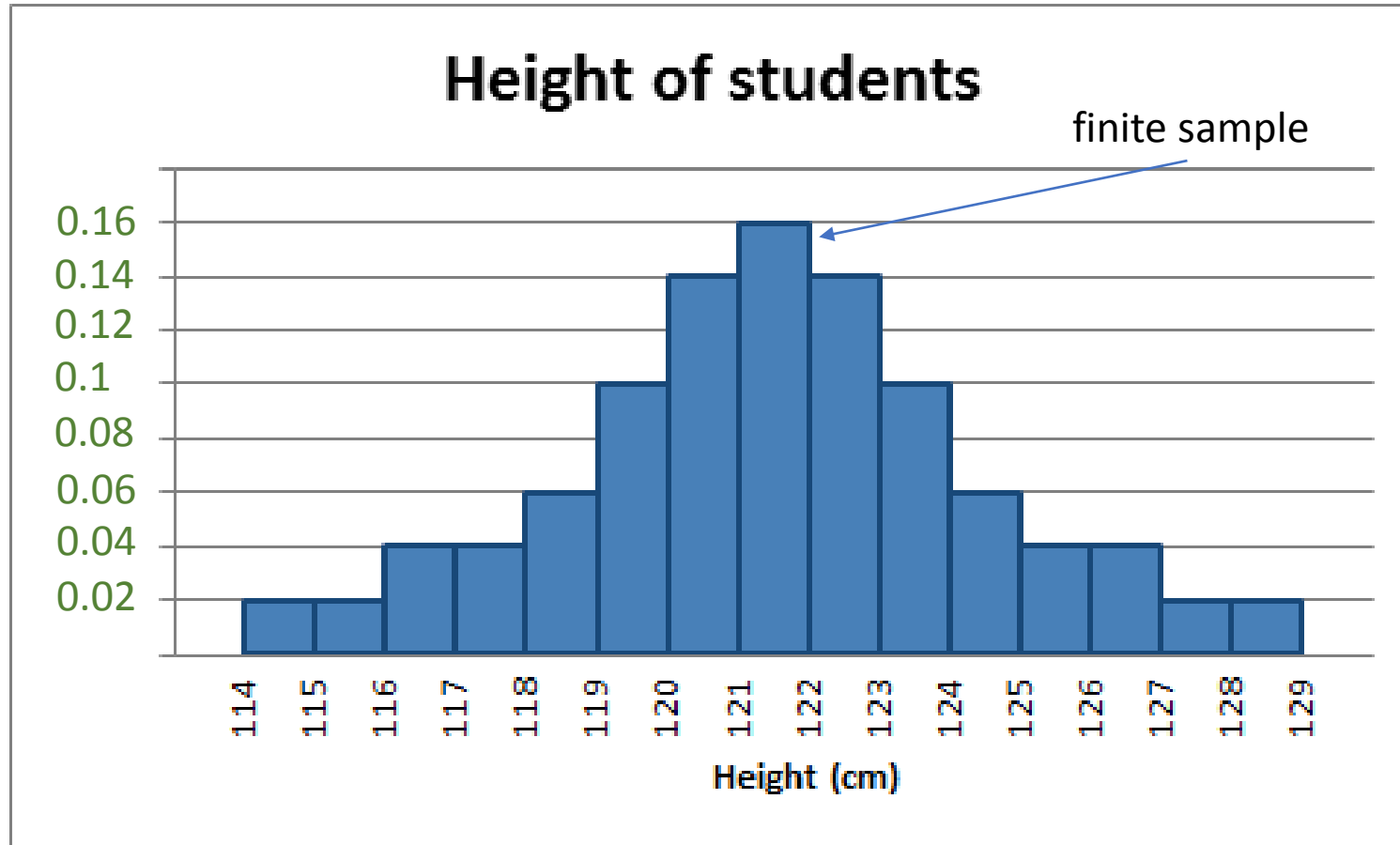
expected value of a random variable



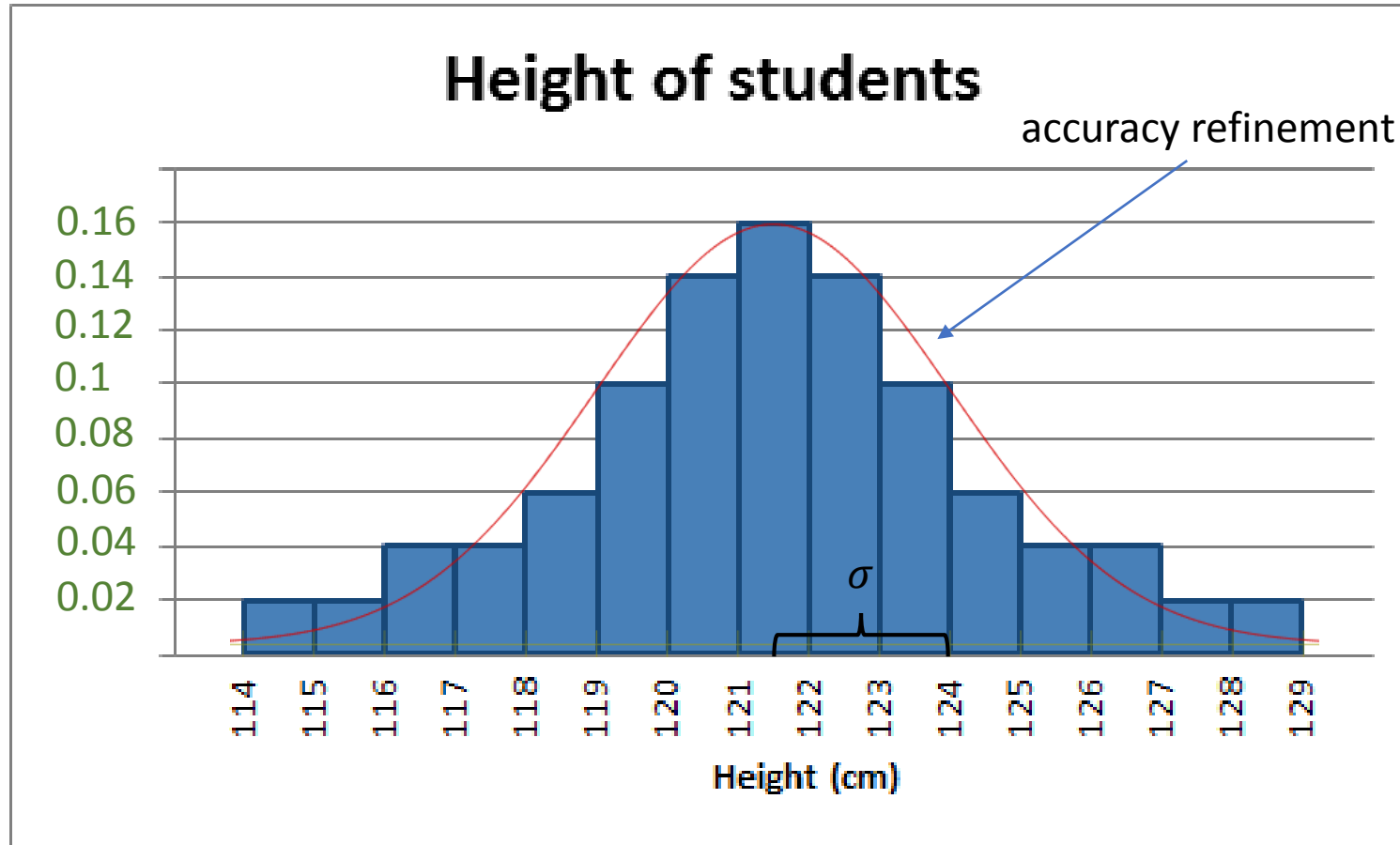
expected value of a random variable



expected value of a random variable



expected value of a random variable

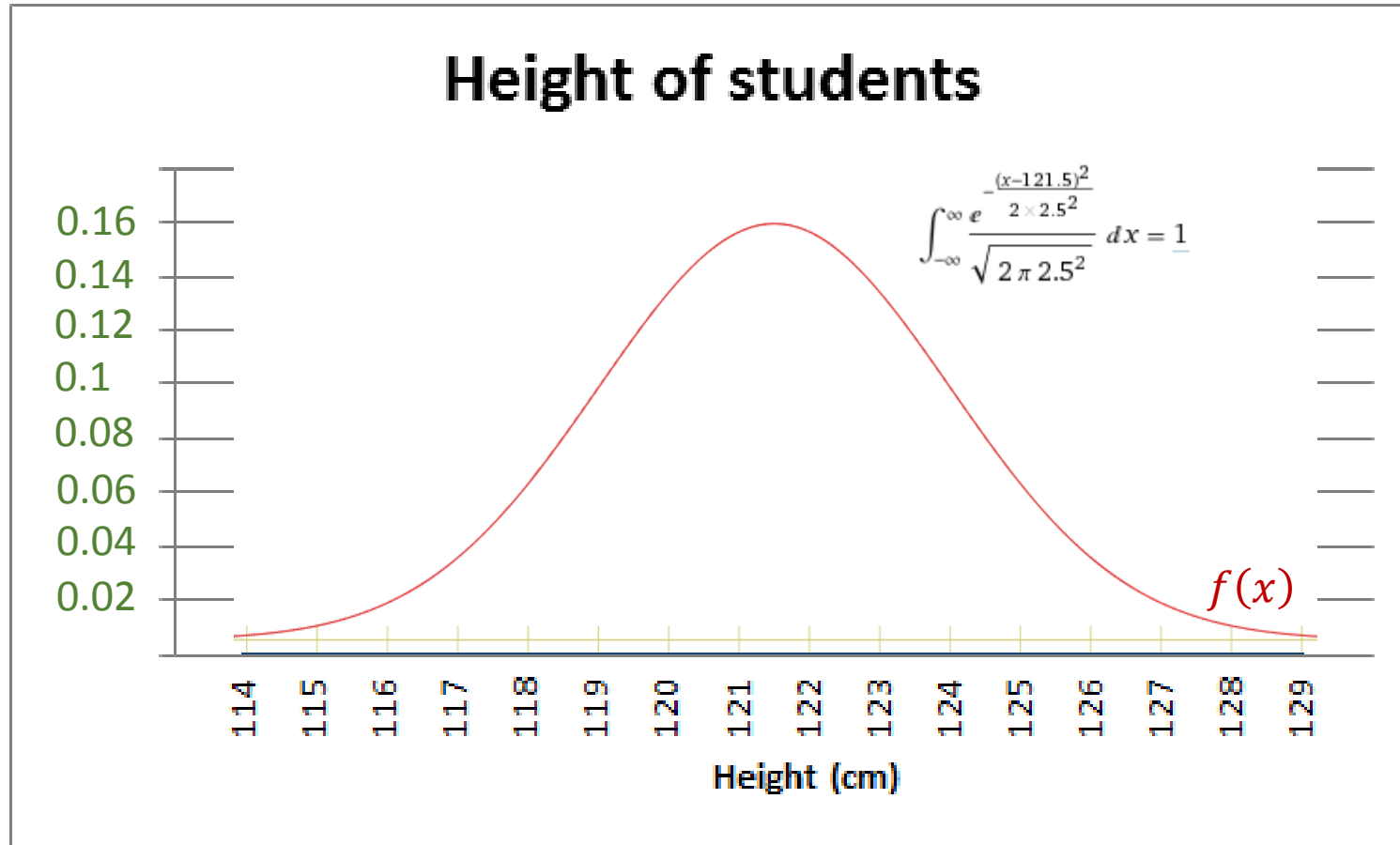


probability density function of the normal distribution:

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\sigma = 2.5; \mu = 121.5$$

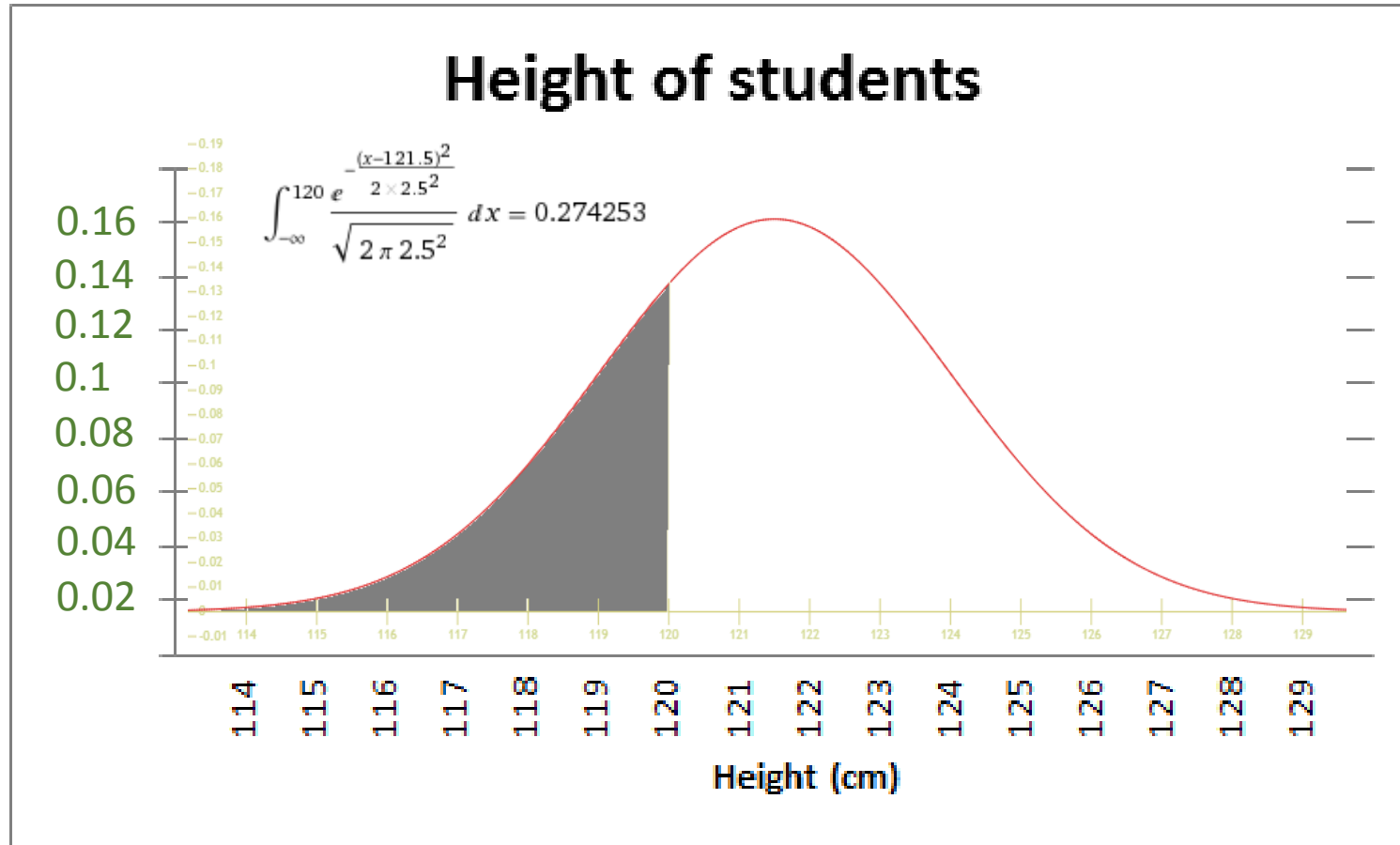
expected value of a random variable



probability density function

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

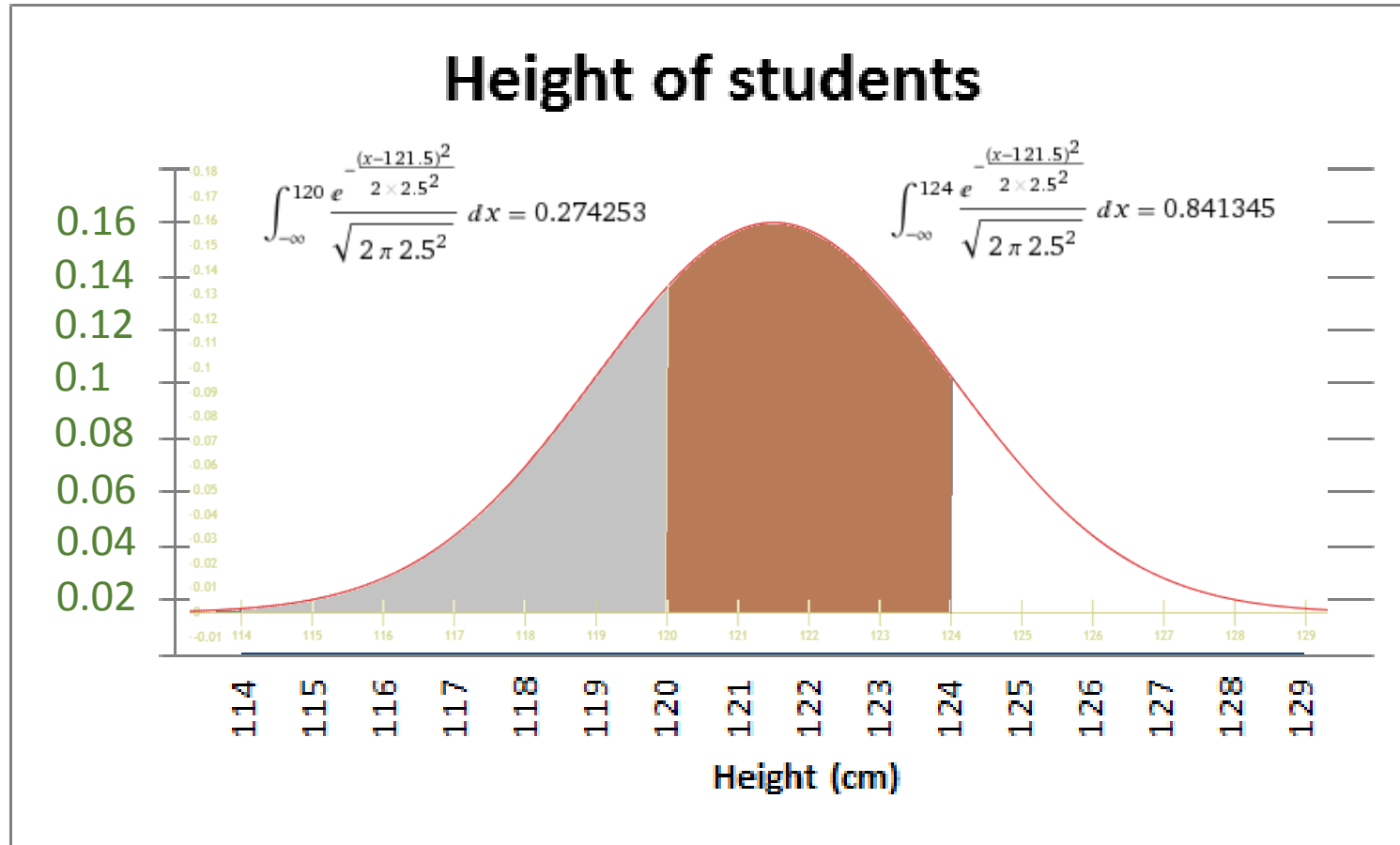
expected value of a random variable



probability density function

$$P[X \leq 120] = \int_{-\infty}^{120} f(x) dx$$

expected value of a random variable

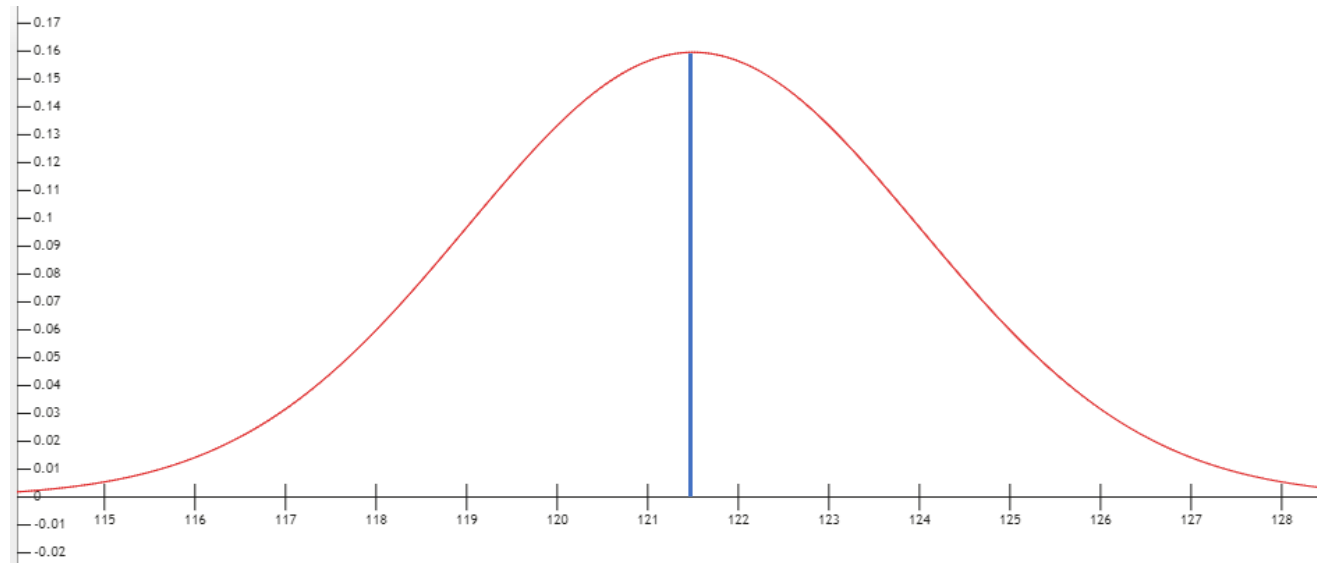


probability density function

$$\begin{aligned} P[120 < X \leq 124] &= \\ &= \int_{120}^{124} f(x) dx = \\ &= 0.841345 - 0.274253 = \\ &= 0.567092 \end{aligned}$$

expected value of a random variable

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx$$



probability density function
of the normal distribution:

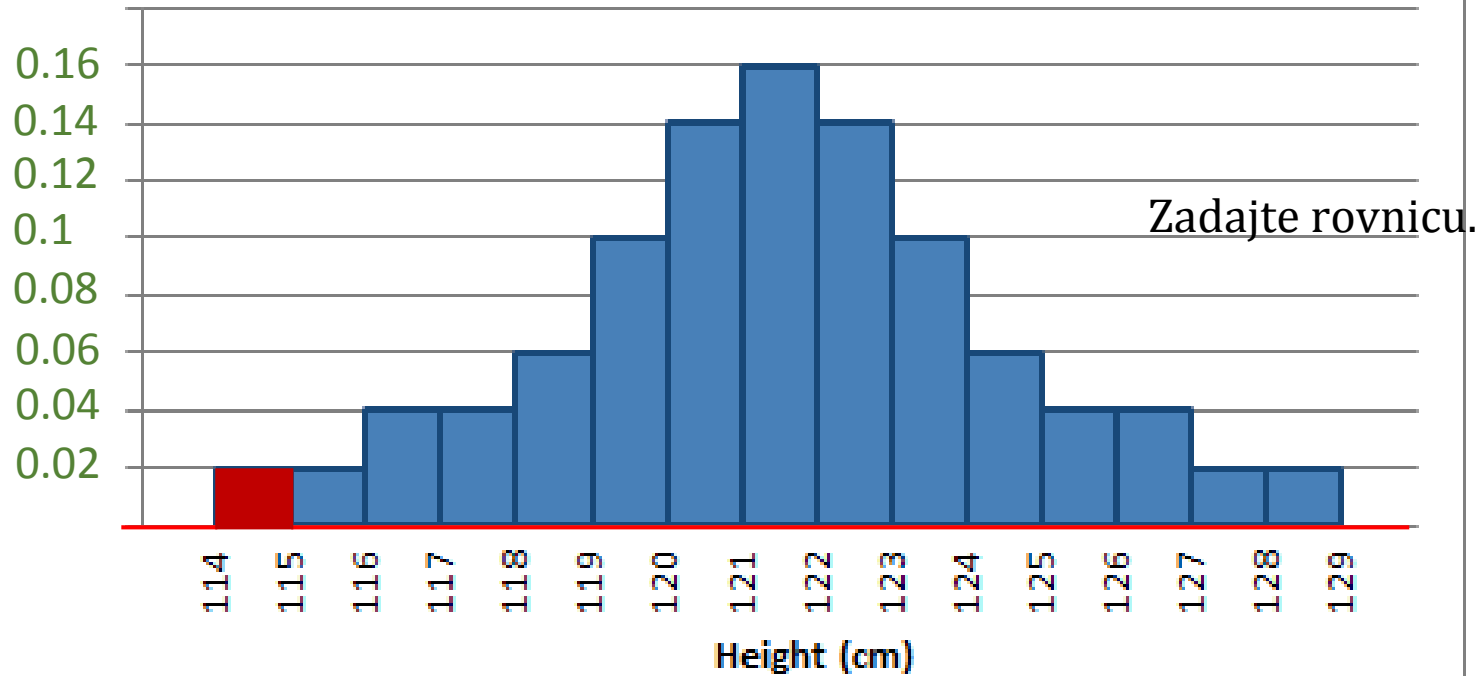
$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\sigma = 2.5; \mu = 121.5$$

$$\int_{-\infty}^{\infty} \frac{x e^{-\frac{(x-121.5)^2}{2 \times 2.5^2}}}{\sqrt{2\pi \cdot 2.5^2}} dx = 121.5$$

expected value of a random variable

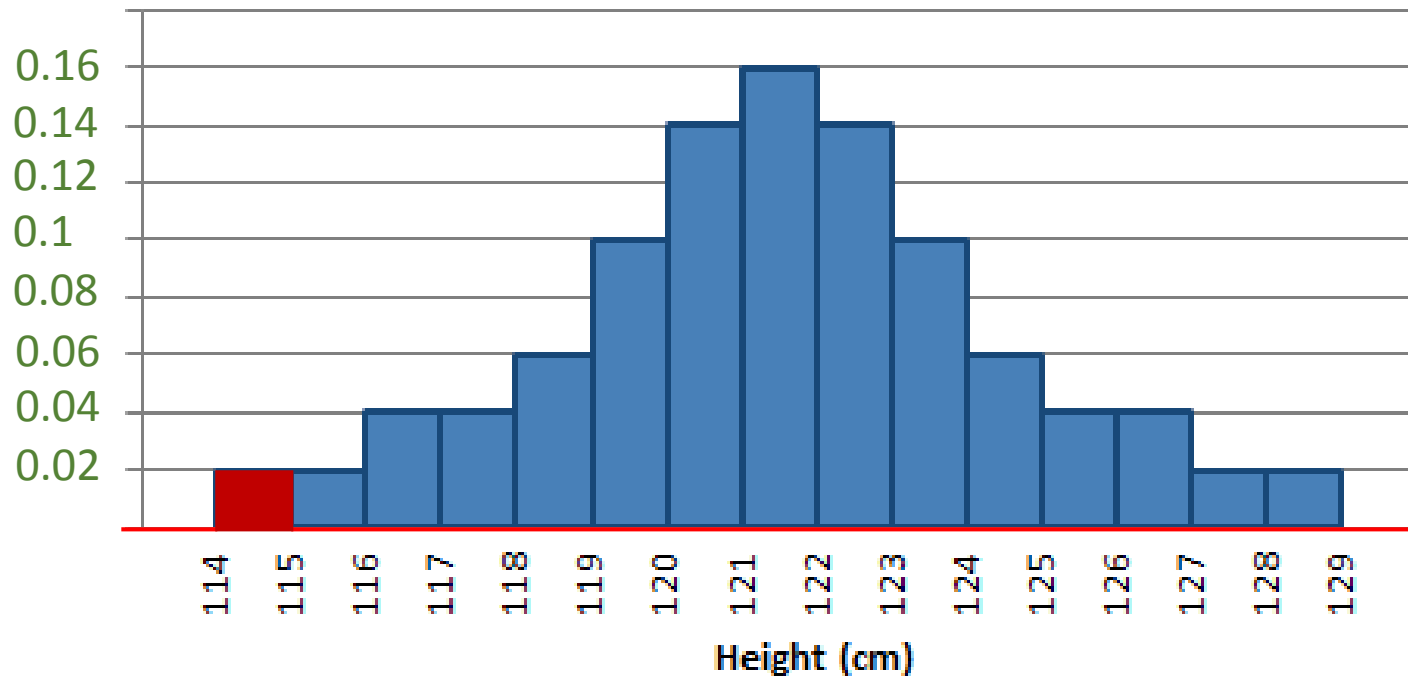
Height of students



$$\int_{-\infty}^{\infty} x f(x) dx$$
$$f(x) = \begin{cases} 0 & x \in (-\infty, 114] \\ 0.02 & x \in (114, 115] \\ 0 & x \in (115, \infty) \end{cases}$$

expected value of a random variable

Height of students



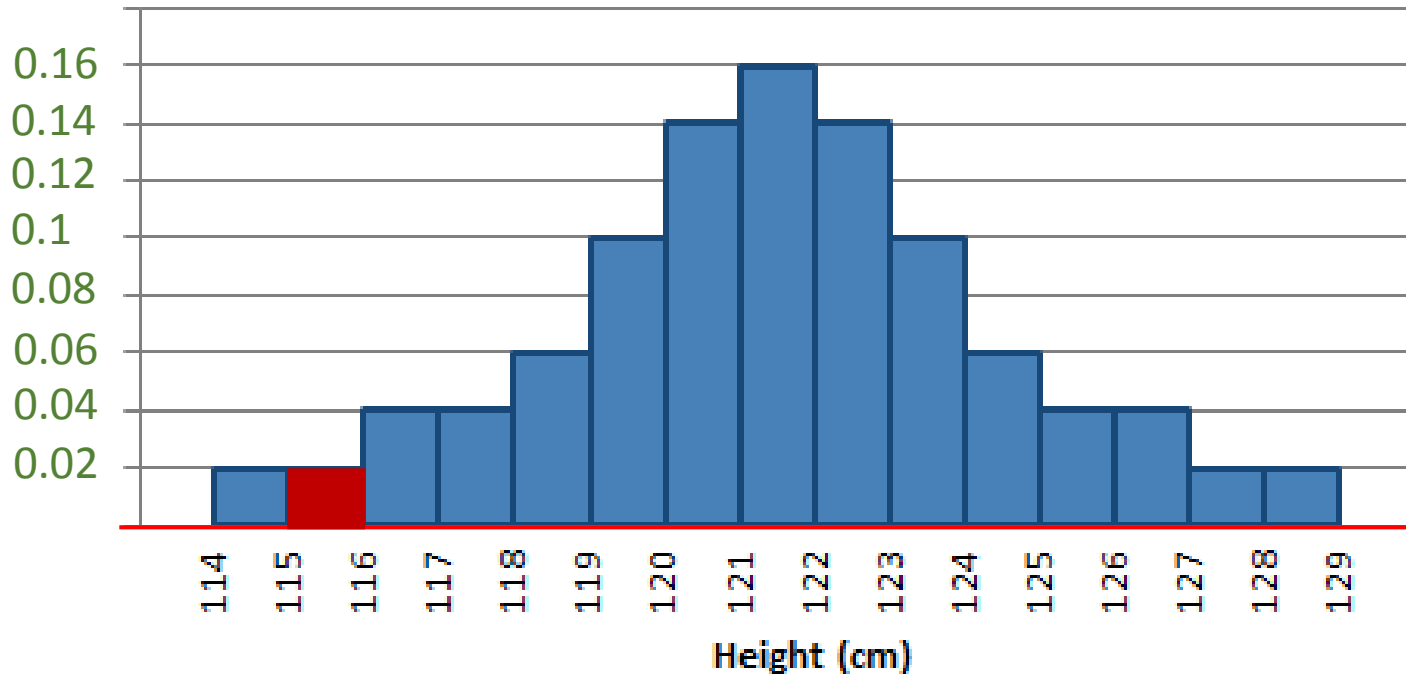
$$0.02 \times 114.5$$

$$\int_{-\infty}^{\infty} x f(x) dx$$
$$f(x) = \begin{cases} 0 & x \in (-\infty, 114] \\ 0.02 & x \in (114, 115] \\ 0 & x \in (115, \infty) \end{cases}$$

$$\begin{aligned} &0.02 \int_{114}^{115} x dx = \\ &= 0.02 \left[\frac{x^2}{2} \right]_{114}^{115} = \\ &= 0.02 \frac{115^2 - 114^2}{2} = \\ &= 0.02 \times 114.5 \end{aligned}$$

expected value of a random variable

Height of students



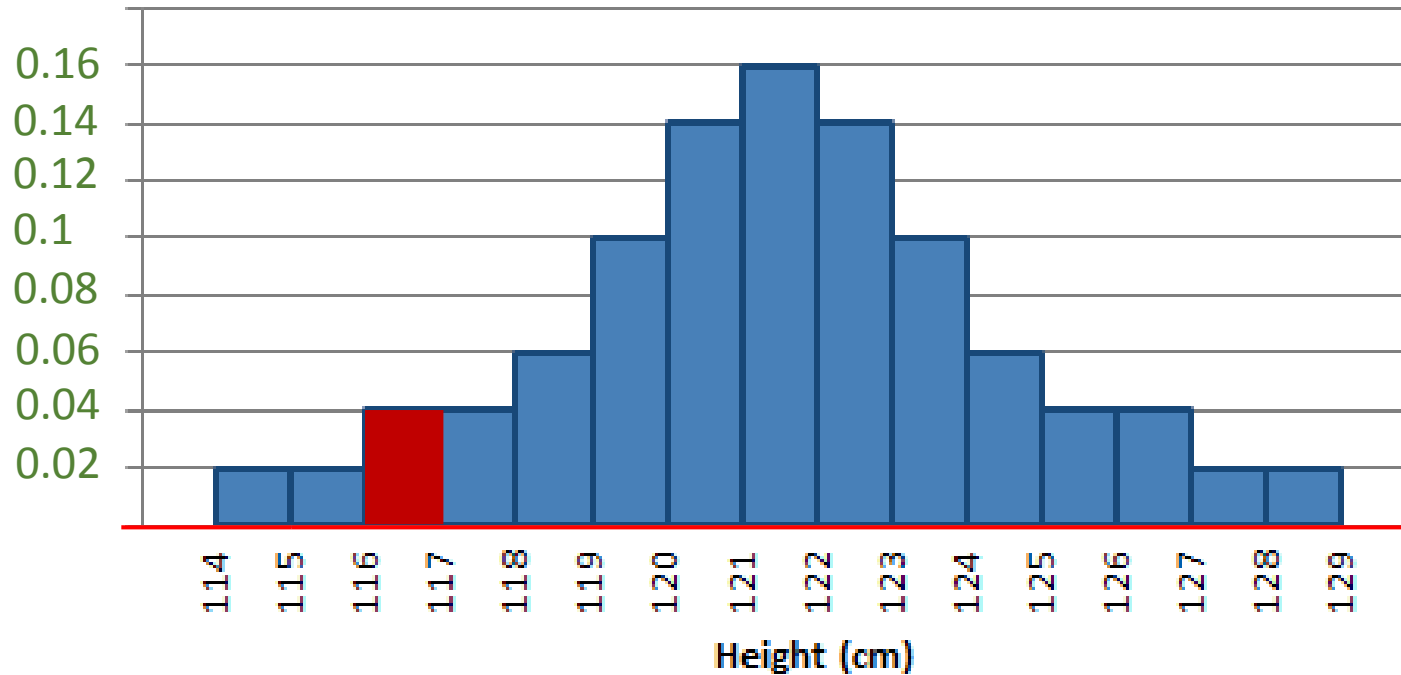
$$0.02 \times 114.5 + 0.02 \times 115.5$$

$$\int_{-\infty}^{\infty} x f(x) dx$$
$$f(x) = \begin{cases} 0 & x \in (-\infty, 115] \\ 0.02 & x \in (115, 116] \\ 0 & x \in (116, \infty) \end{cases}$$

$$\begin{aligned} 0.02 \int_{115}^{116} x dx &= \\ &= 0.02 \left[\frac{x^2}{2} \right]_{115}^{116} = \\ &= 0.02 \frac{116^2 - 115^2}{2} = \\ &= 0.02 \times 115.5 \end{aligned}$$

expected value of a random variable

Height of students



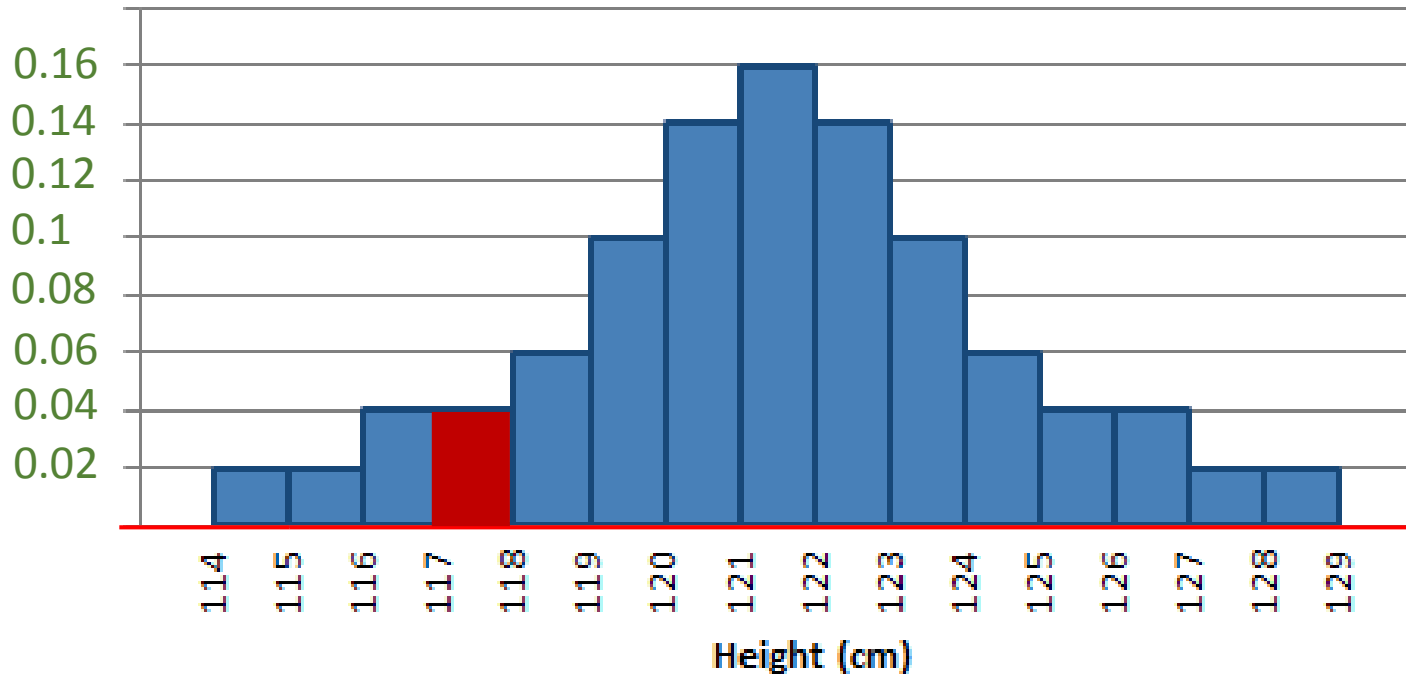
$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5$$

$$\int_{-\infty}^{\infty} x f(x) dx$$
$$f(x) = \begin{cases} 0 & x \in (-\infty, 116] \\ 0.04 & x \in (116, 117] \\ 0 & x \in (117, \infty) \end{cases}$$

$$\begin{aligned} 0.04 \int_{116}^{117} x dx &= \\ &= 0.04 \left[\frac{x^2}{2} \right]_{116}^{117} = \\ &= 0.04 \frac{117^2 - 116^2}{2} = \\ &= 0.04 \times 116.5 \end{aligned}$$

expected value of a random variable

Height of students



$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5$$

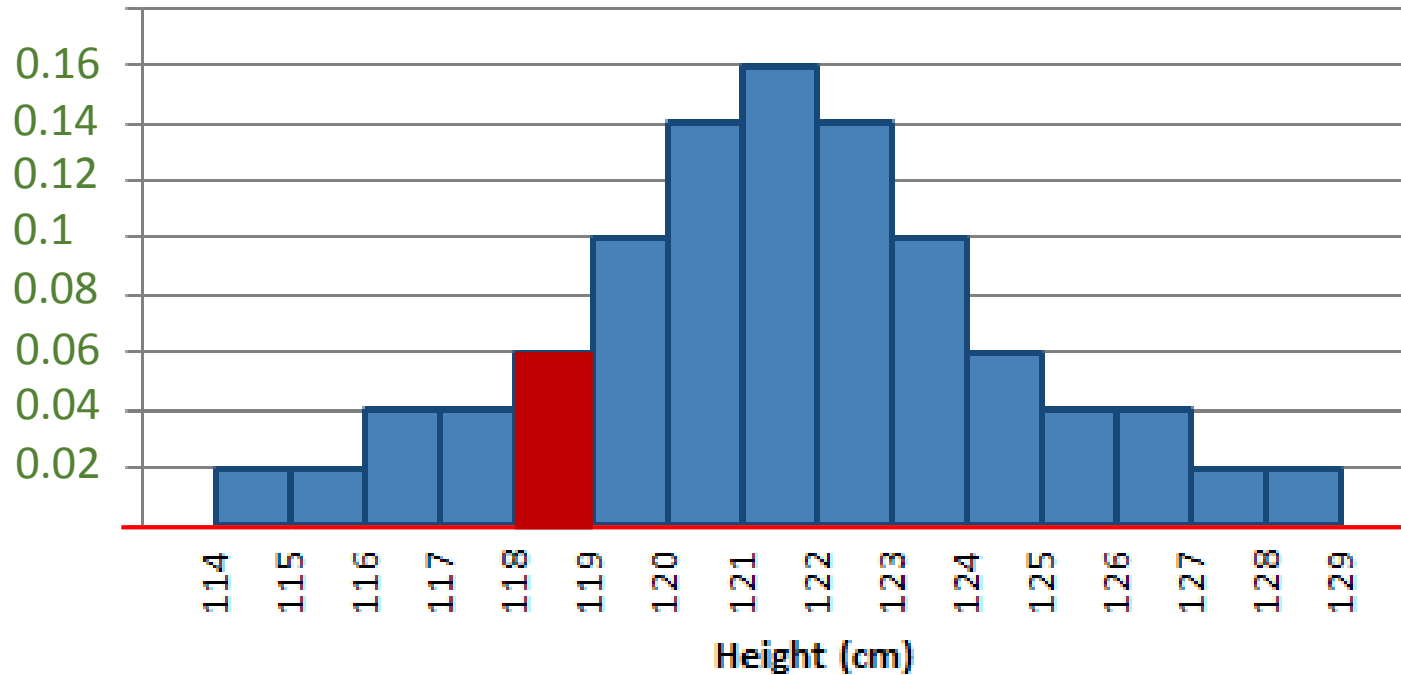
$$\int_{-\infty}^{\infty} x f(x) dx$$

$$f(x) = \begin{cases} 0 & x \in (-\infty, 117] \\ 0.04 & x \in (117, 118] \\ 0 & x \in (118, \infty) \end{cases}$$

$$\begin{aligned} &0.04 \int_{117}^{118} x dx = \\ &= 0.04 \left[\frac{x^2}{2} \right]_{117}^{118} = \\ &= 0.04 \frac{118^2 - 117^2}{2} = \\ &= 0.04 \times 117.5 \end{aligned}$$

expected value of a random variable

Height of students



$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5$$

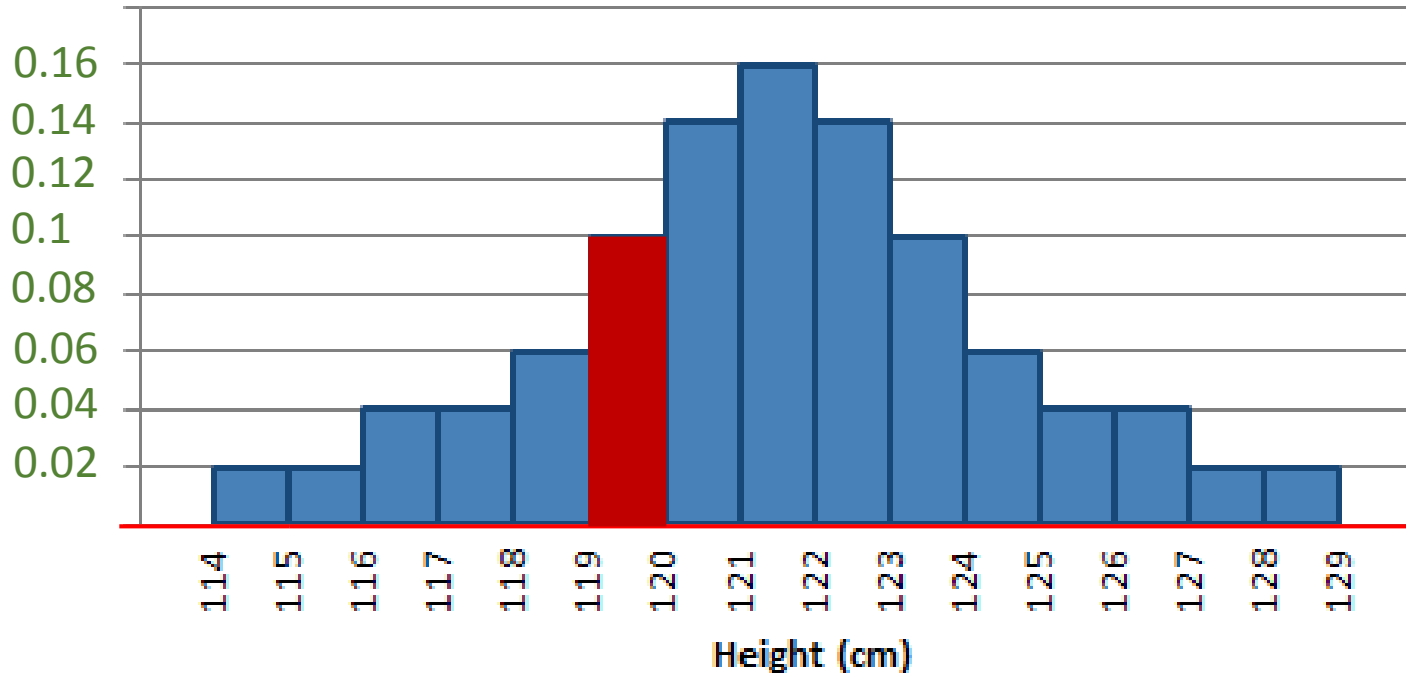
$$\int_{-\infty}^{\infty} x f(x) dx$$

$$f(x) = \begin{cases} 0 & x \in (-\infty, 118] \\ 0.06 & x \in (118, 119] \\ 0 & x \in (119, \infty) \end{cases}$$

$$\begin{aligned} 0.06 \int_{118}^{119} x dx &= \\ &= 0.06 \left[\frac{x^2}{2} \right]_{118}^{119} = \\ &= 0.06 \frac{119^2 - 118^2}{2} = \\ &= 0.06 \times 118.5 \end{aligned}$$

expected value of a random variable

Height of students



$$\int_{-\infty}^{\infty} x f(x) dx$$

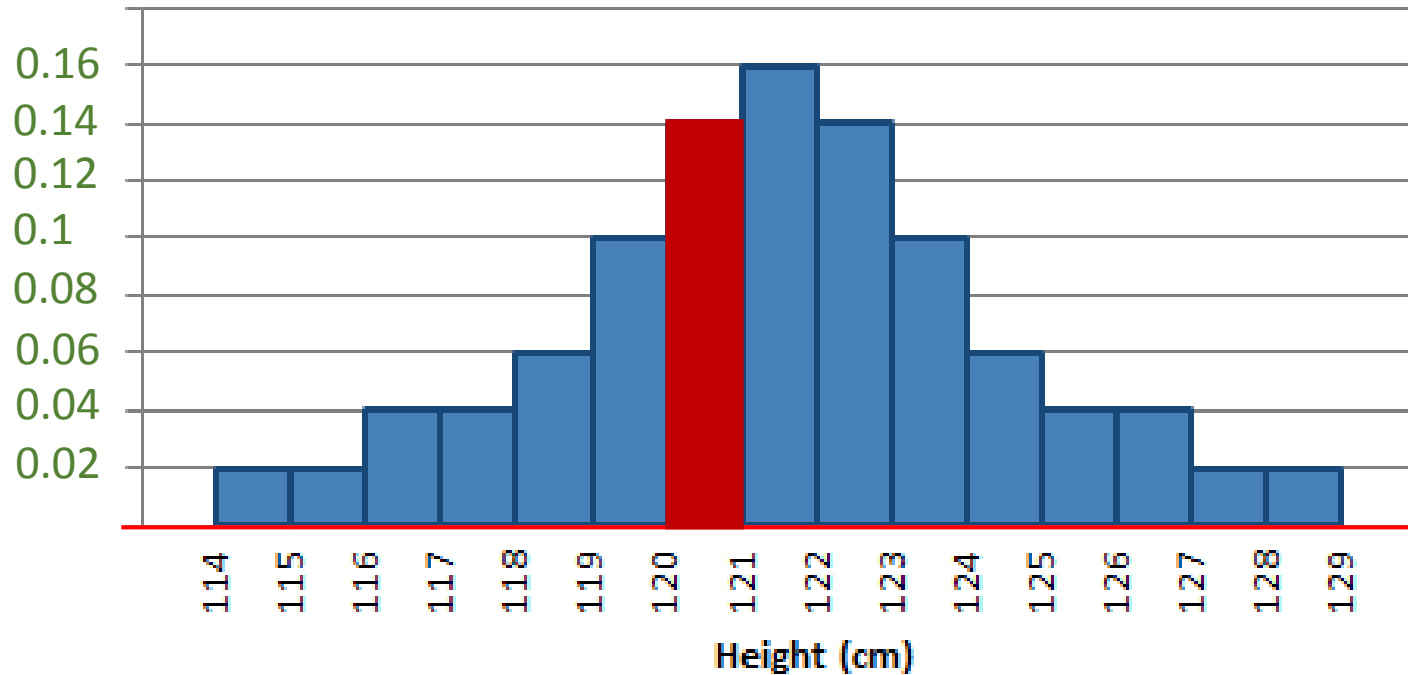
$$f(x) = \begin{cases} 0 & x \in (-\infty, 119] \\ 0.1 & x \in (119, 120] \\ 0 & x \in (120, \infty) \end{cases}$$

$$\begin{aligned} 0.1 \int_{119}^{120} x dx &= \\ &= 0.1 \left[\frac{x^2}{2} \right]_{119}^{120} = \\ &= 0.1 \frac{120^2 - 119^2}{2} = \\ &= 0.1 \times 119.5 \end{aligned}$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5$$

expected value of a random variable

Height of students



$$\int_{-\infty}^{\infty} x f(x) dx$$

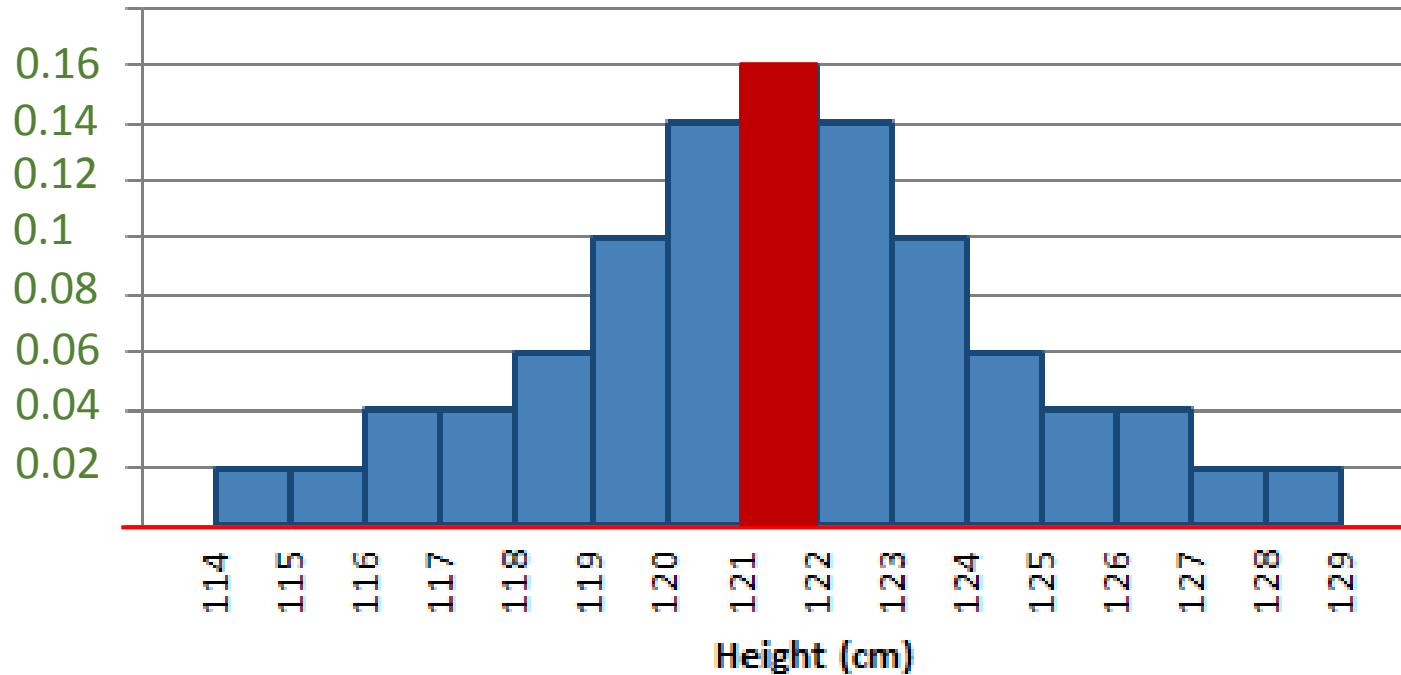
$$f(x) = \begin{cases} 0 & x \in (-\infty, 120] \\ 0.14 & x \in (120, 121] \\ 0 & x \in (121, \infty) \end{cases}$$

$$\begin{aligned} 0.14 \int_{120}^{121} x dx &= \\ &= 0.14 \left[\frac{x^2}{2} \right]_{120}^{121} = \\ &= 0.14 \frac{121^2 - 120^2}{2} = \\ &= 0.14 \times 120.5 \end{aligned}$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5$$

expected value of a random variable

Height of students



$$\int_{-\infty}^{\infty} x f(x) dx$$

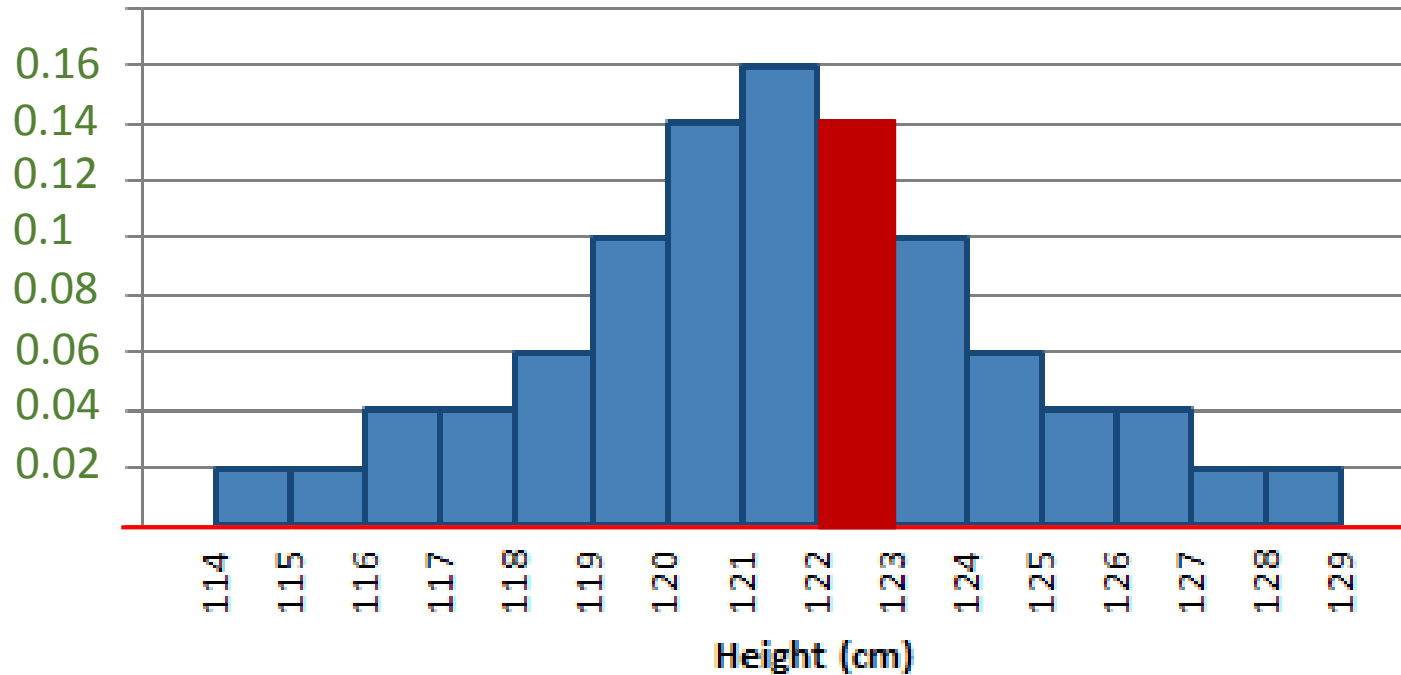
$$f(x) = \begin{cases} 0 & x \in (-\infty, 121] \\ 0.16 & x \in (121, 122] \\ 0 & x \in (122, \infty) \end{cases}$$

$$0.16 \int_{121}^{122} x dx = 0.16 \times 121.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5$$

expected value of a random variable

Height of students



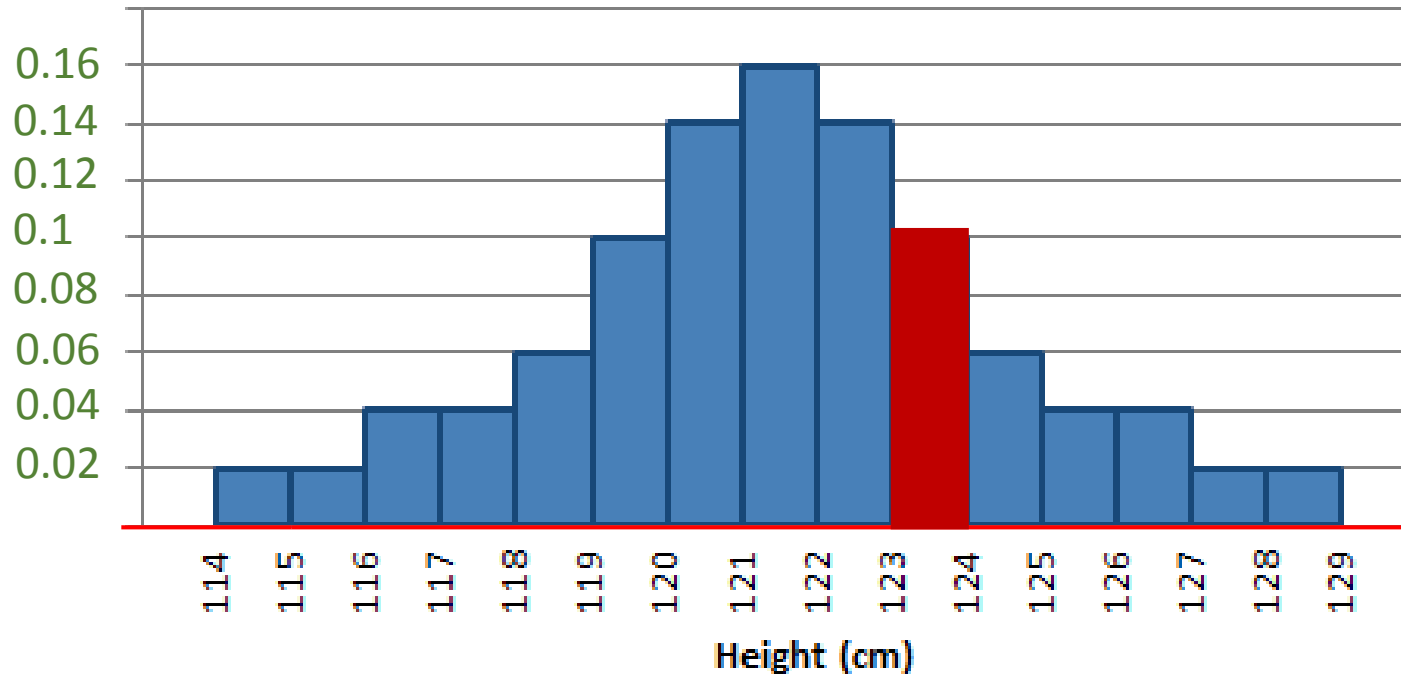
$$\int_{-\infty}^{\infty} x f(x) dx$$

$$0.14 \int_{122}^{123} x dx =$$
$$= 0.14 \times 122.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 +$$
$$+ 0.14 \times 122.5$$

expected value of a random variable

Height of students



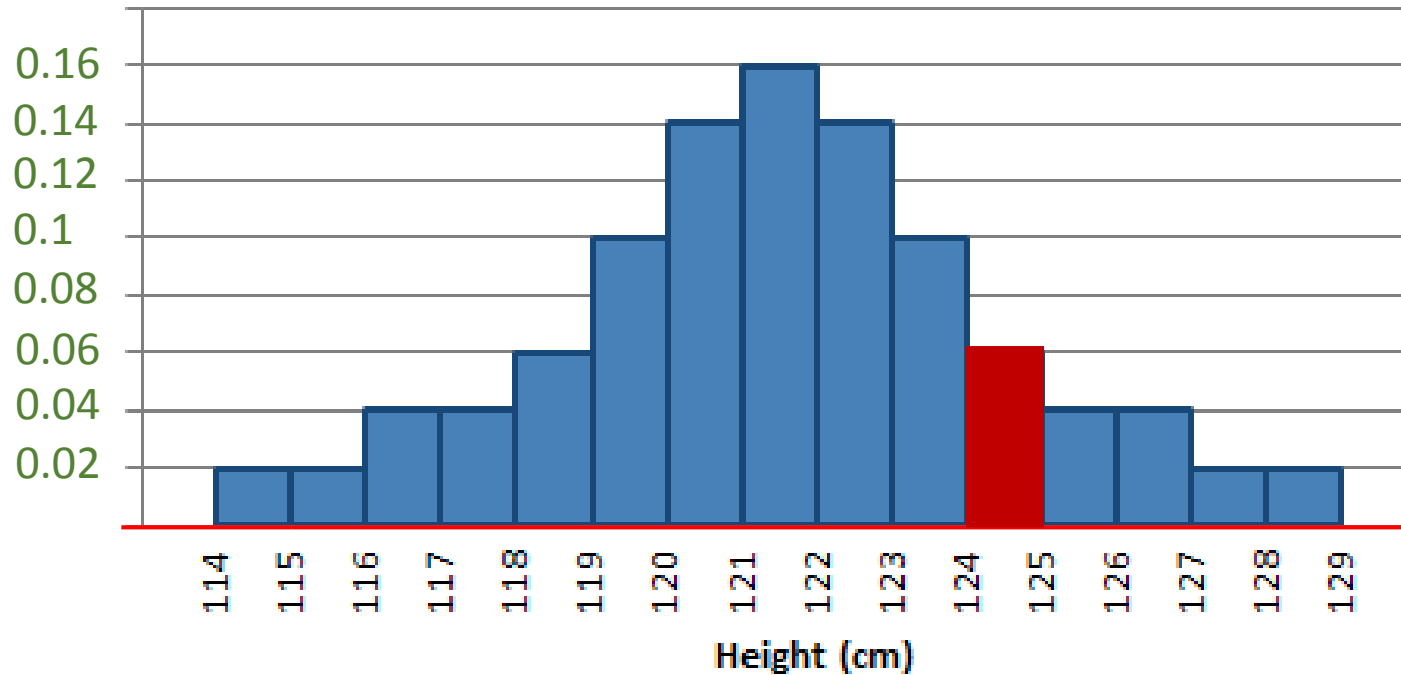
$$\int_{-\infty}^{\infty} x f(x) dx$$

$$0.1 \int_{123}^{124} x dx =$$
$$= 0.1 \times 123.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 +$$
$$+ 0.14 \times 122.5 + 0.1 \times 123.5$$

expected value of a random variable

Height of students



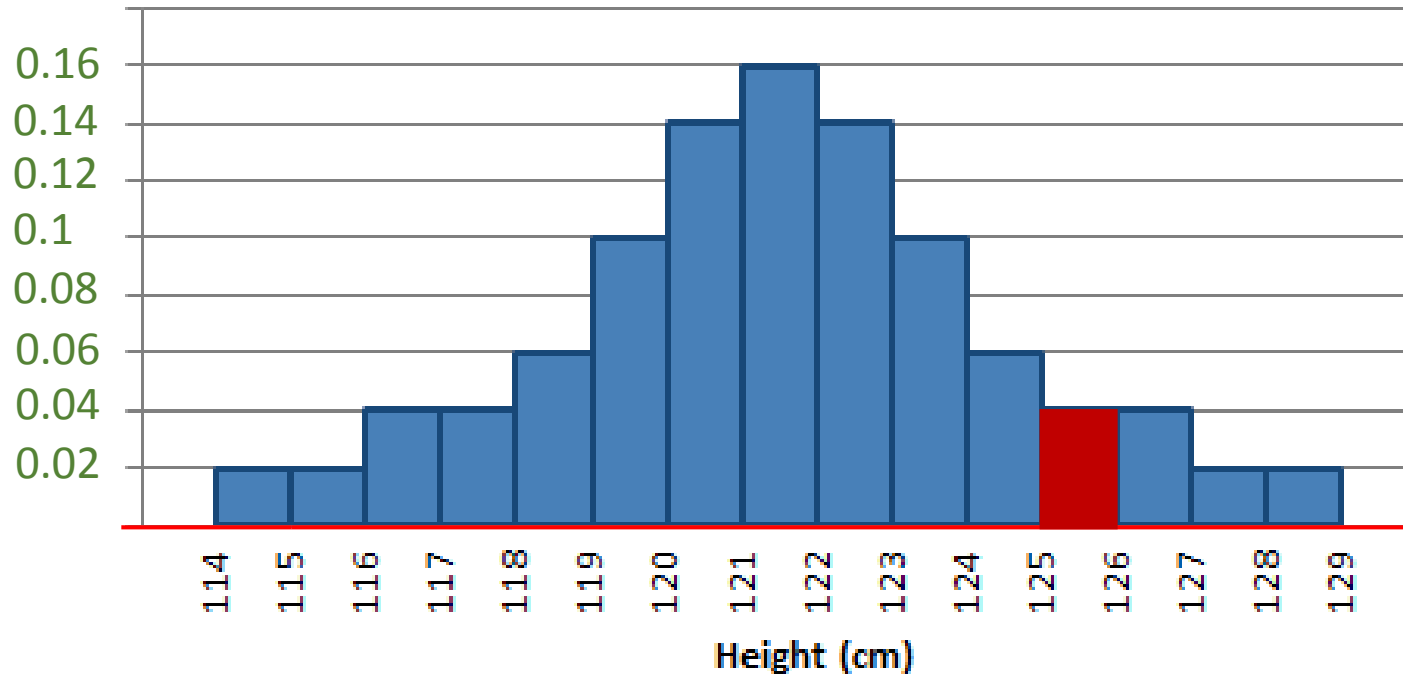
$$\int_{-\infty}^{\infty} x f(x) dx$$

$$0.06 \int_{124}^{125} x dx =$$
$$= 0.06 \times 124.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 +$$
$$+ 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5$$

expected value of a random variable

Height of students



$$\int_{-\infty}^{\infty} x f(x) dx$$

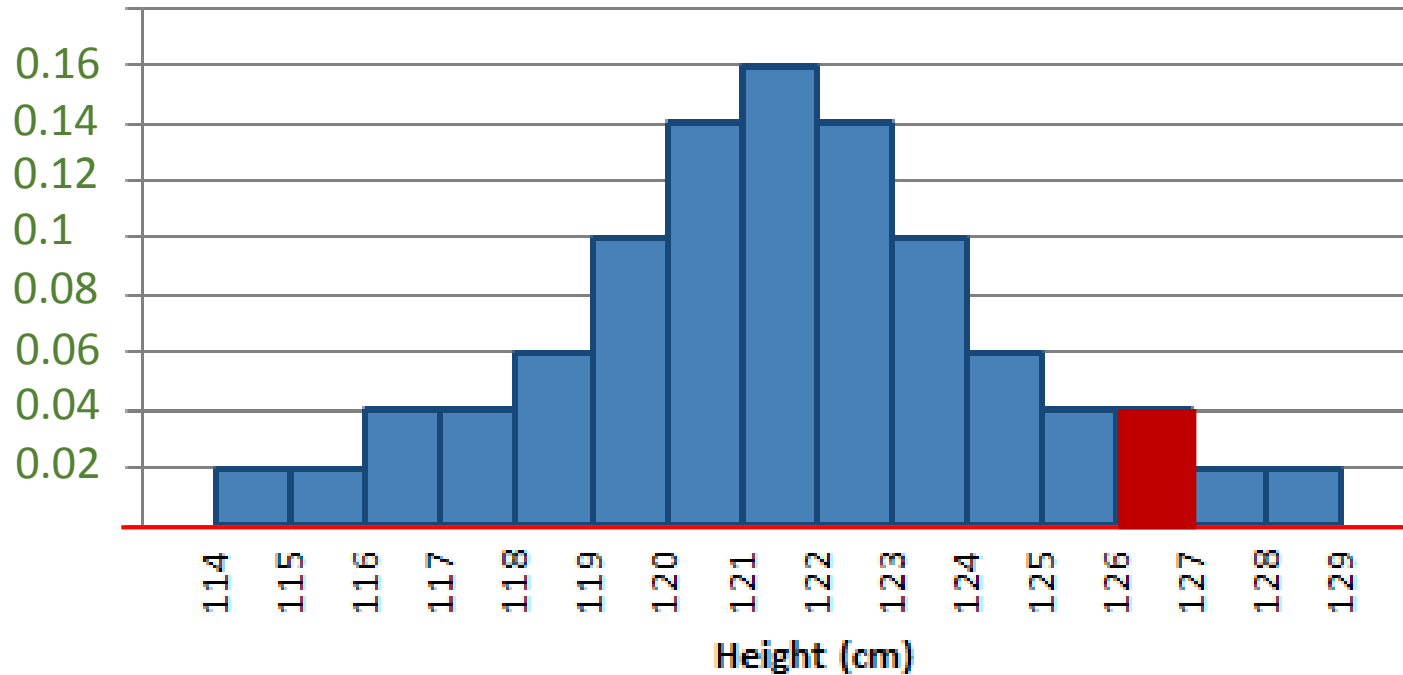
$$0.04 \int_{125}^{126} x dx =$$
$$= 0.04 \times 125.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 +$$
$$+ 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5 + 0.04 \times 125.5$$

expected value of a random variable

$$\int_{-\infty}^{\infty} x f(x) dx$$

Height of students



$$0.04 \int_{126}^{125} x dx =$$

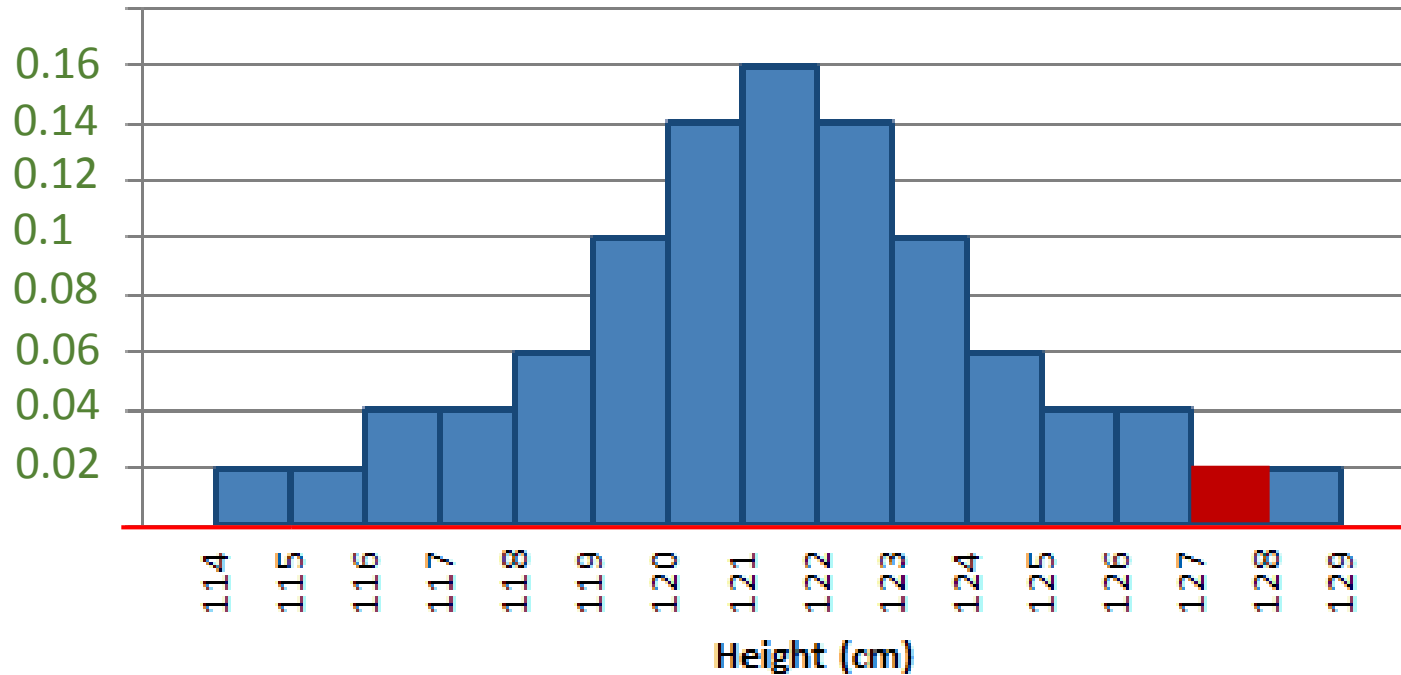
$$= 0.04 \times 126.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 +$$

$$+ 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5 + 0.04 \times 125.5 + 0.04 \times 126.5$$

expected value of a random variable

Height of students



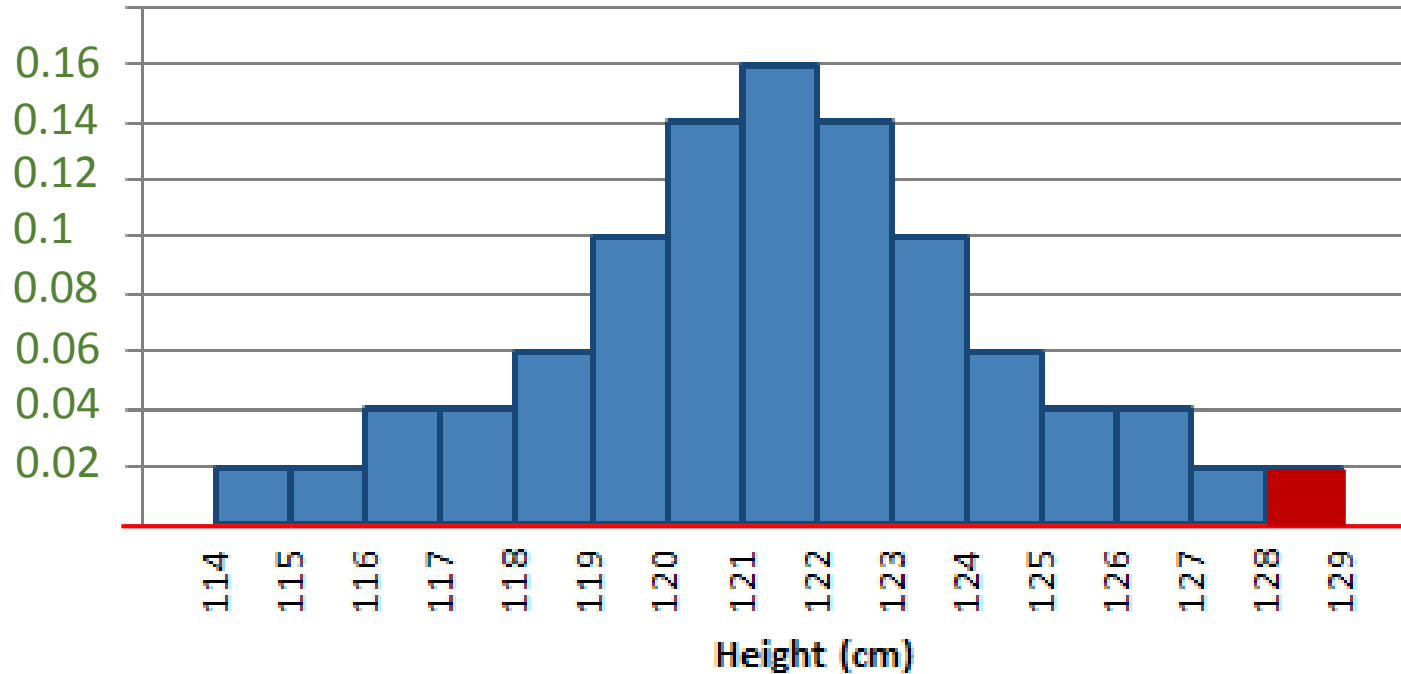
$$\int_{-\infty}^{\infty} x f(x) dx$$

$$0.02 \int_{127}^{128} x dx = \\ = 0.02 \times 127.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 + \\ + 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5 + 0.04 \times 125.5 + 0.04 \times 126.5 + 0.02 \times 127.5$$

expected value of a random variable

Height of students



$$\int_{-\infty}^{\infty} x f(x) dx$$

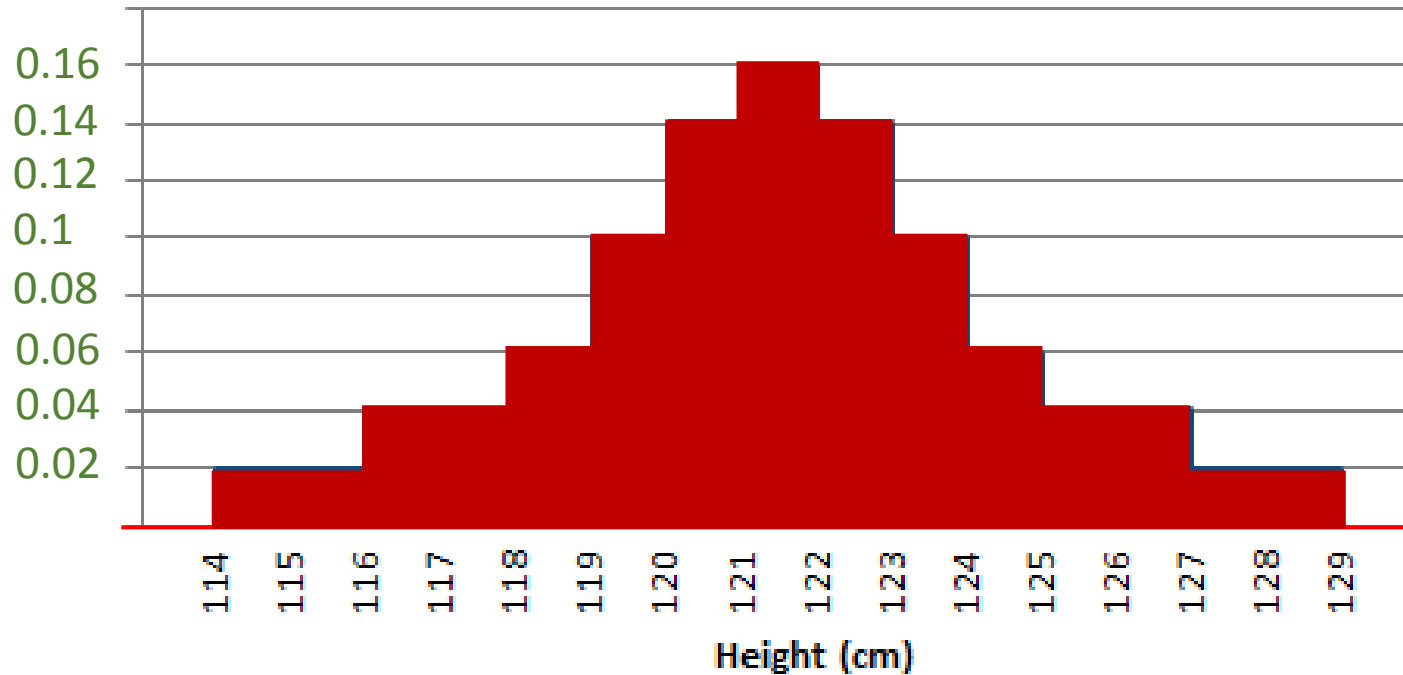
$$0.02 \int_{128}^{129} x dx = \\ = 0.02 \times 128.5$$

$$0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 + \\ + 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5 + 0.04 \times 125.5 + 0.04 \times 126.5 + 0.02 \times 127.5 + 0.02 \times 128.5$$

expected value of a random variable

$$E[X] = \int_{-\infty}^{\infty} xf(x) dx$$

Height of students

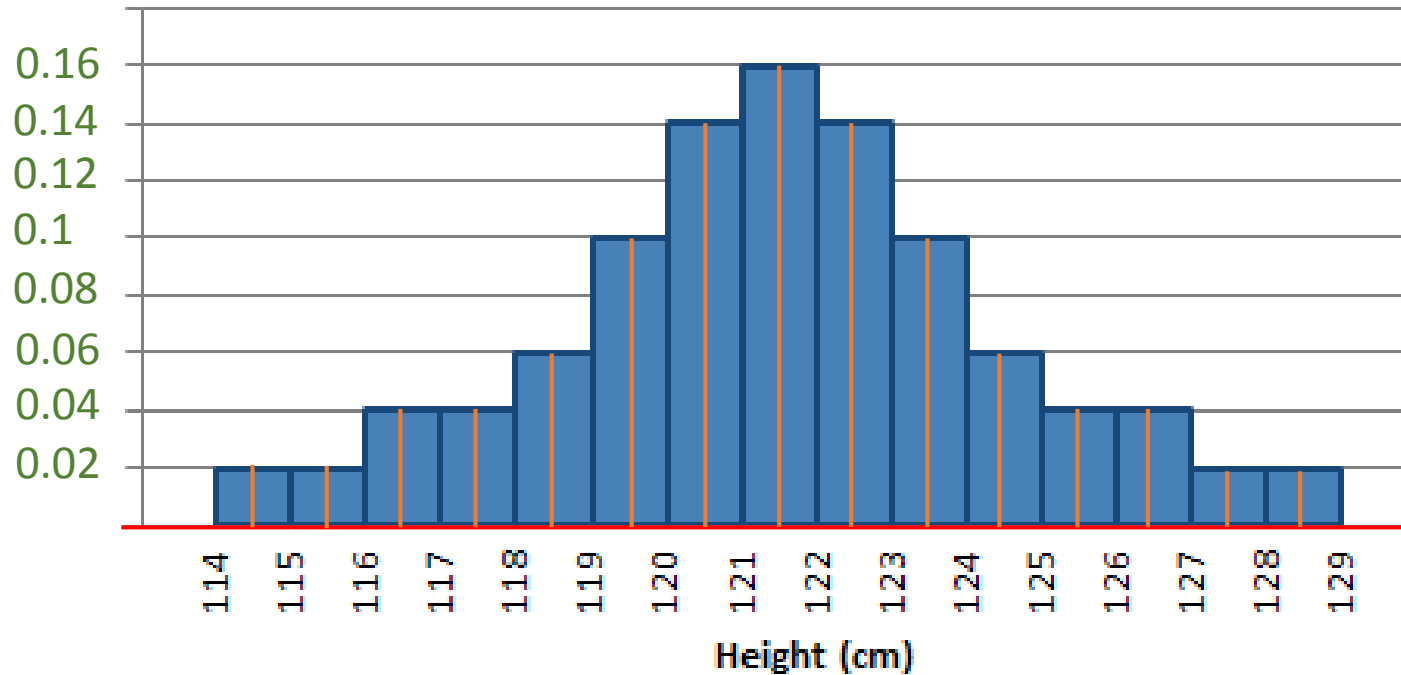


$$E[X] = 0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 + 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5 + 0.04 \times 125.5 + 0.04 \times 126.5 + 0.02 \times 127.5 + 0.02 \times 128.5 = 121.5$$

expected value of a random variable

$$E[g(X)] = \sum_x f(x)g(x)$$

Height of students



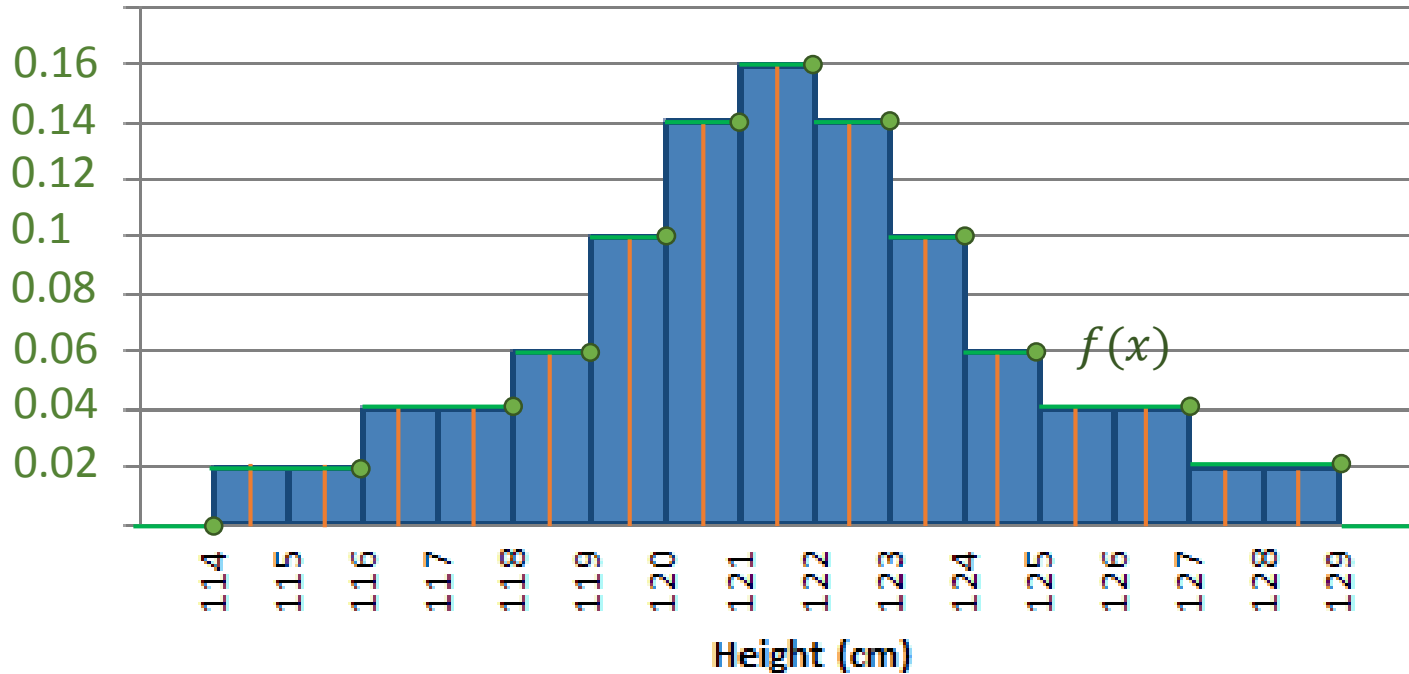
$$E[X] = 0.02 \times 114.5 + 0.02 \times 115.5 + 0.04 \times 116.5 + 0.04 \times 117.5 + 0.06 \times 118.5 + 0.1 \times 119.5 + 0.14 \times 120.5 + 0.16 \times 121.5 + 0.14 \times 122.5 + 0.1 \times 123.5 + 0.06 \times 124.5 + 0.04 \times 125.5 + 0.04 \times 126.5 + 0.02 \times 127.5 + 0.02 \times 128.5 = 121.5$$

expected value of a random variable

$$E[g(X)] = \sum_x f(x)g(x)$$

$$g(x) = x - \frac{1}{2}$$

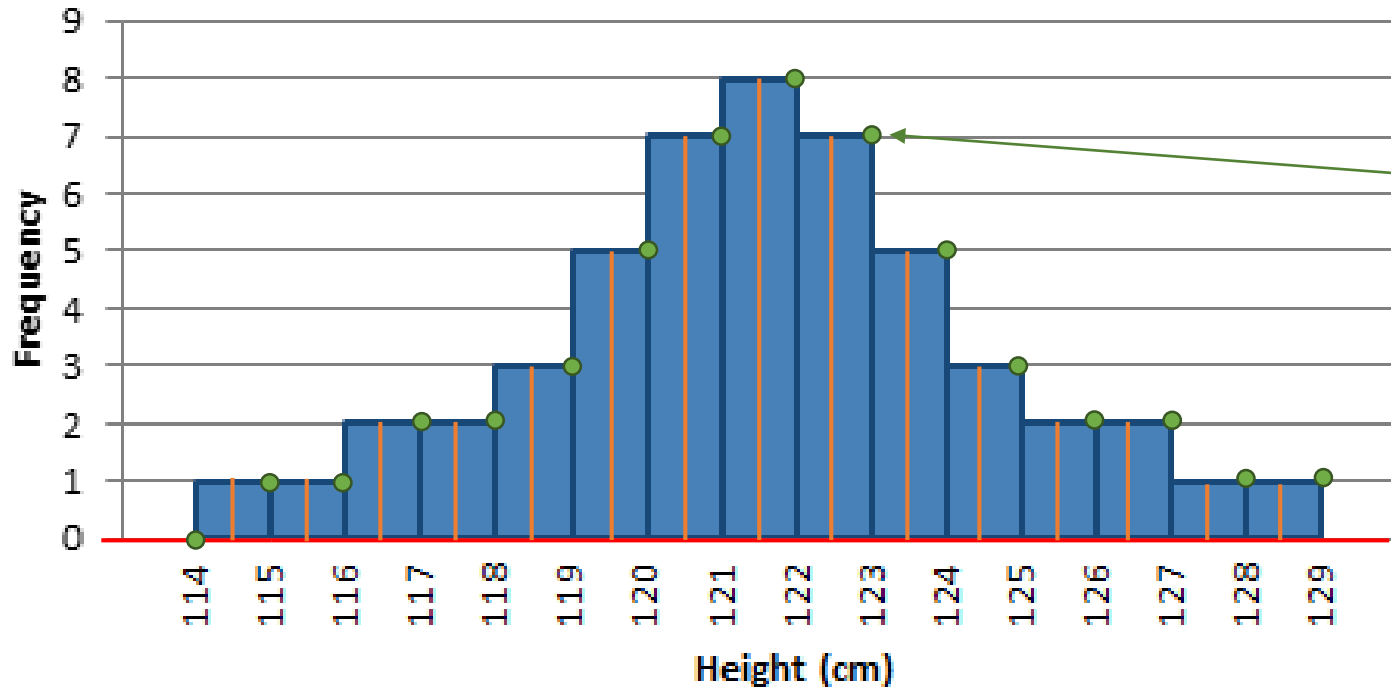
Height of students



$$E[X] = \frac{1}{50} \times 114.5 + \frac{1}{50} \times 115.5 + \frac{2}{50} \times 116.5 + \frac{2}{50} \times 117.5 + \frac{3}{50} \times 118.5 + \frac{5}{50} \times 119.5 + \frac{7}{50} \times 120.5 + \frac{8}{50} \times 121.5 + \frac{7}{50} \times 122.5 + \frac{5}{50} \times 123.5 + \frac{3}{50} \times 124.5 + \frac{2}{50} \times 125.5 + \frac{2}{50} \times 126.5 + \frac{1}{50} \times 127.5 + \frac{1}{50} \times 128.5 = 121.5$$

expected value of a random variable

Height of students



$$E[g(X)] = \sum_x f(x)g(x)$$

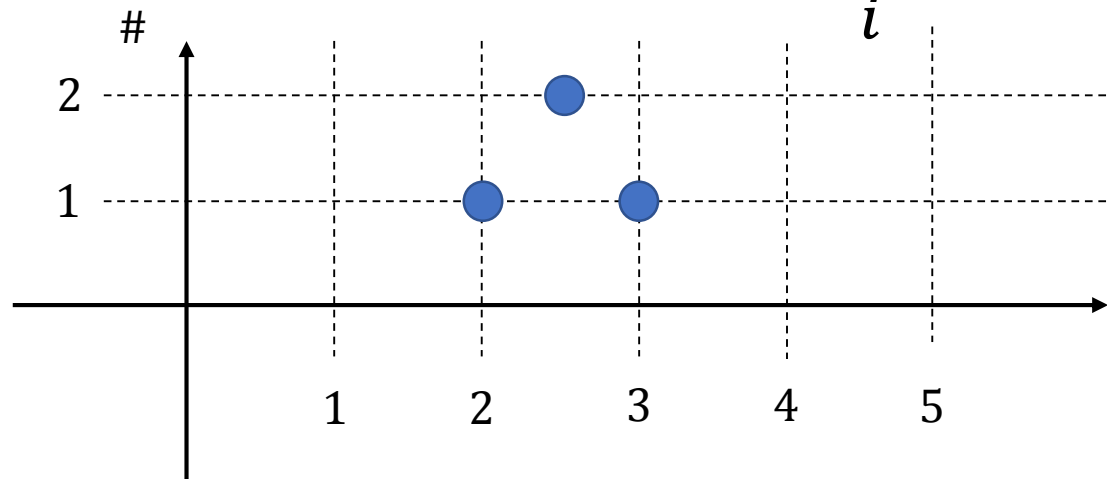
$$g(x) = x - \frac{1}{2}$$

$$f(x) = \frac{n_x}{\sum_j n_j}$$

$$E[X] = \frac{1}{50} (1 \times 114.5 + 1 \times 115.5 + 2 \times 116.5 + 2 \times 117.5 + 3 \times 118.5 + 5 \times 119.5 + 7 \times 120.5 + 8 \times 121.5 + 7 \times 122.5 + 5 \times 123.5 + 3 \times 124.5 + 2 \times 125.5 + 2 \times 126.5 + 1 \times 127.5 + 1 \times 128.5) = 121.5$$

variance of a random variable

$$E[X] = \frac{1}{N} \sum_i n_i x_i$$

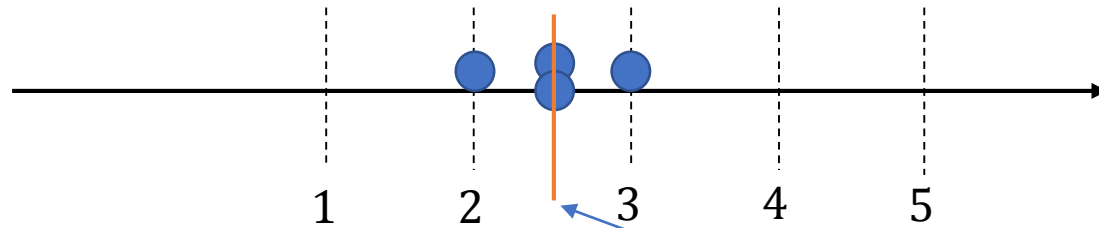


i	n_i	x_i
1	1	2
2	2	2.5
3	1	3

$$E[X] = \frac{1}{4} (2 + 2 \times 2.5 + 3)$$

variance of a random variable

$$E[X] = \frac{1}{N} \sum_i n_i x_i$$

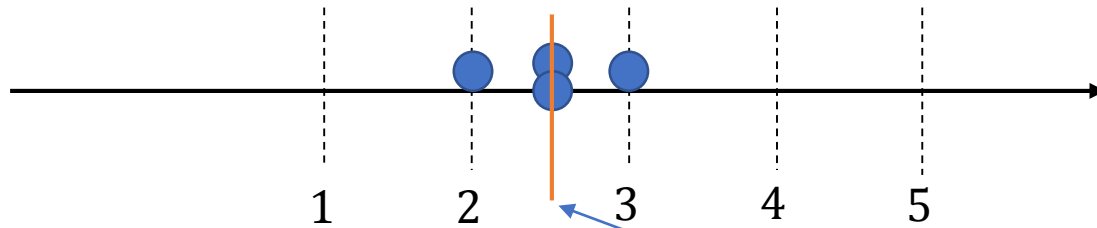


$$E[X] = \frac{1}{4} (2 + 2 \times 2.5 + 3) = 2.5$$

$$E[(X - E[X])] = \frac{1}{4} (2 - 2.5 + 2 \times (2.5 - 2.5) + 3 - 2.5)$$

variance of a random variable

$$E[X] = \frac{1}{N} \sum_i n_i x_i$$

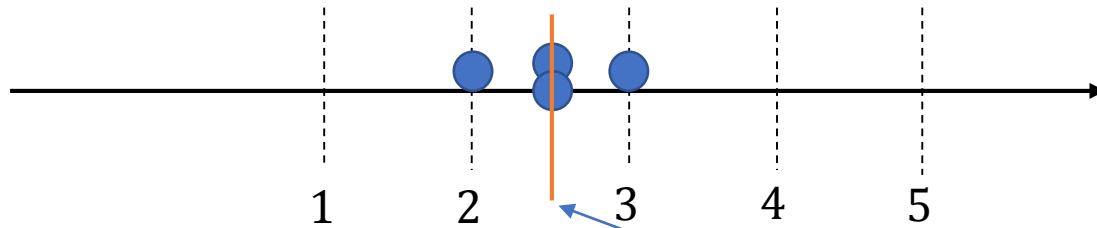


$$E[X] = \frac{1}{4} (2 + 2 \times 2.5 + 3) = 2.5$$

$$E[(X - E[X])^2] = \frac{1}{4} (2 - 2.5 + 2 \times (2.5 - 2.5) + 3 - 2.5)$$

variance of a random variable

$$E[X] = \frac{1}{N} \sum_i n_i x_i$$



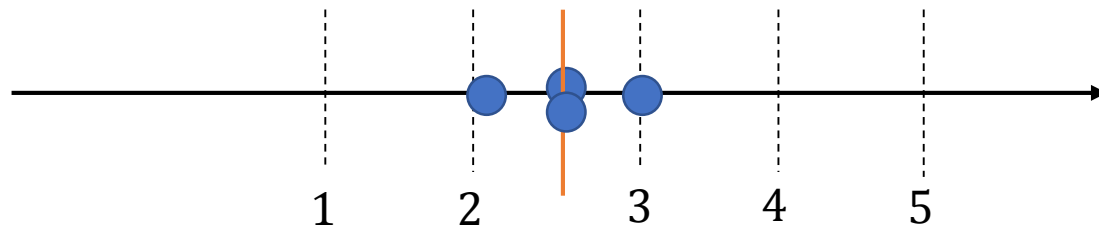
$$E[X] = \frac{1}{4} (2 + 2 \times 2.5 + 3) = 2.5$$

$$E[(X - E[X])^2] = \frac{1}{4} ((2 - 2.5)^2 + 2 \times (2.5 - 2.5)^2 + (3 - 2.5)^2) = 0.125$$

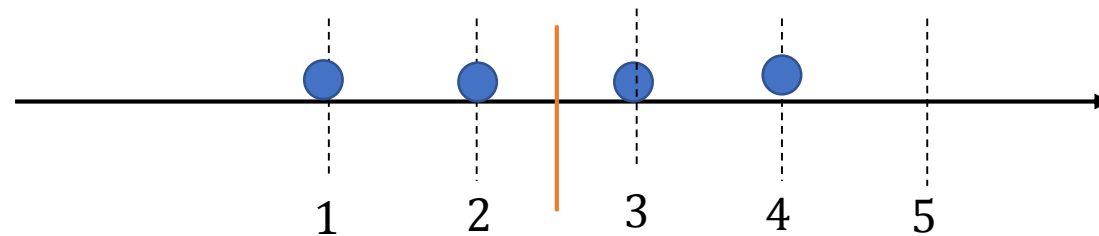
variance of a random variable

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

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



$$\text{Var}(X) = \frac{1}{4} ((2 - 2.5)^2 + 2 \times (2.5 - 2.5)^2 + (3 - 2.5)^2) = 0.125$$

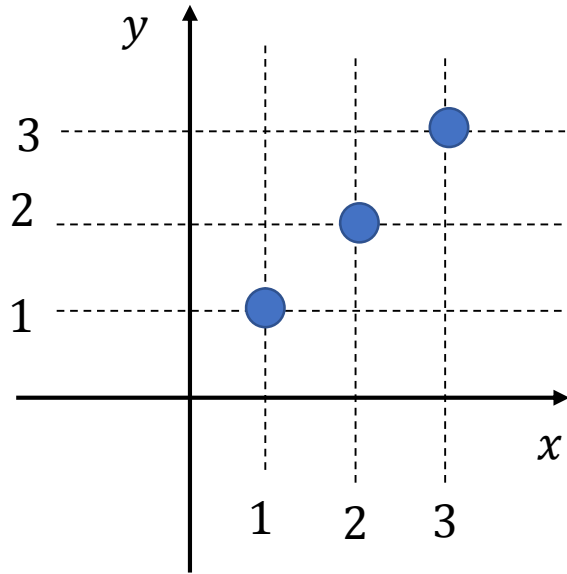


$$\text{Var}(X) = \frac{1}{4} ((1 - 2.5)^2 + (2 - 2.5)^2 + (3 - 2.5)^2 + (4 - 2.5)^2) = 1.25$$

covariance of two random variables

$$\text{Var}(X) = \text{Cov}(X, X)$$

$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$



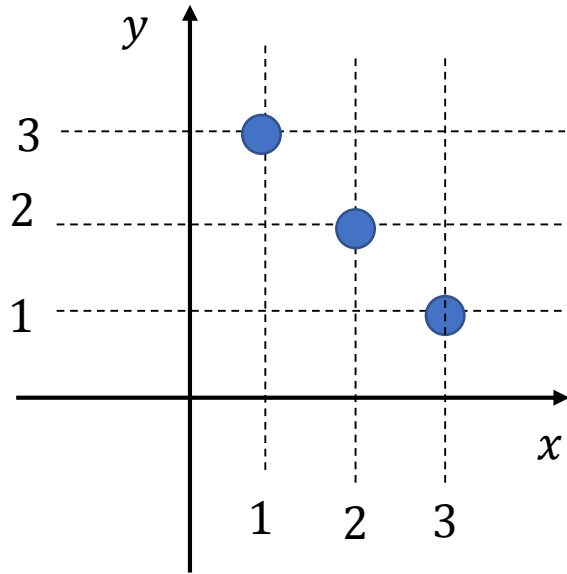
$$\text{Var}(X) = \text{Cov}(X, X) =$$

$$= \frac{1}{3} ((1 - 2)^2 + (2 - 2)^2 + (3 - 2)^2) = +\frac{2}{3}$$

covariance of two random variables

$$\text{Var}(X) = \text{Cov}(X, X)$$

$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$



$$\text{Cov}(X, Y) =$$

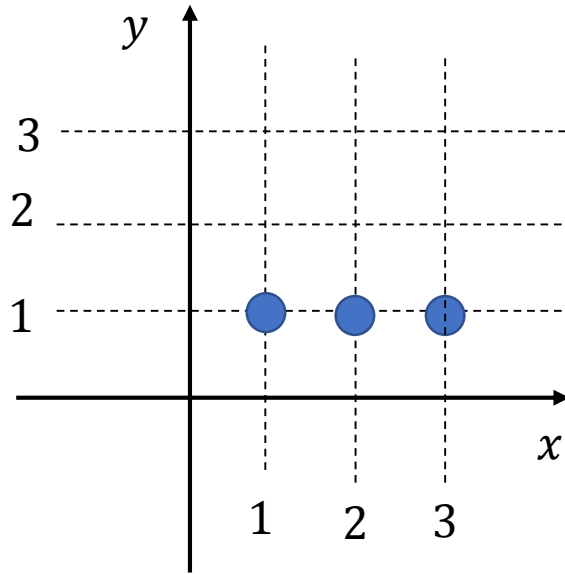
$$= \frac{1}{3} ((1 - 2)(3 - 2) + (2 - 2)^2 + (3 - 2)(1 - 2)) = -\frac{2}{3}$$

- + + -

covariance of two random variables

$$\text{Var}(X) = \text{Cov}(X, X)$$

$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$



$$\text{Cov}(X, Y) =$$

$$= \frac{1}{3} ((1 - 2) + (2 - 2) + (3 - 2))(1 - 1) = 0$$

covariance matrix

$$\mathbf{X} = \begin{bmatrix} X_1 \\ \vdots \\ X_n \end{bmatrix}$$

$$\begin{bmatrix} E[(X_1 - E[X_1])(X_1 - E[X_1])] & E[(X_1 - E[X_1])(X_2 - E[X_2])] & \cdots & E[(X_1 - E[X_1])(X_n - E[X_n])] \\ E[(X_2 - E[X_2])(X_1 - E[X_1])] & E[(X_2 - E[X_2])(X_2 - E[X_2])] & \cdots & E[(X_2 - E[X_2])(X_n - E[X_n])] \\ \vdots & \vdots & \ddots & \vdots \\ E[(X_n - E[X_n])(X_1 - E[X_1])] & E[(X_n - E[X_n])(X_2 - E[X_2])] & \cdots & E[(X_n - E[X_n])(X_n - E[X_n])] \end{bmatrix}$$

covariance matrix

$$\mathbf{x} = \begin{bmatrix} X \\ Y \end{bmatrix}$$

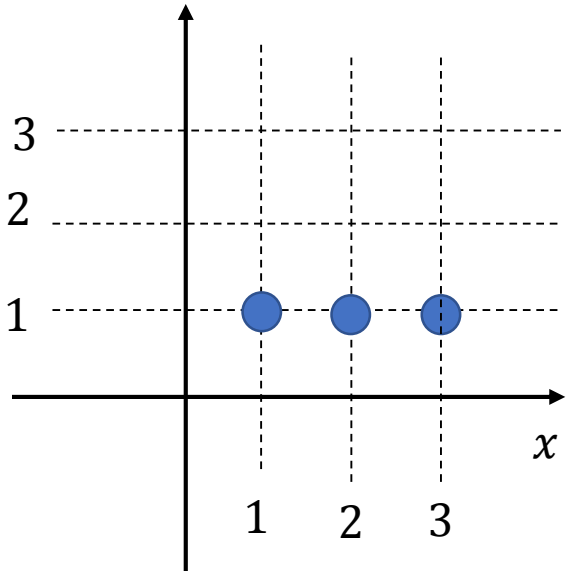
$Cov(X, X)$

$Cov(X, Y)$

$Cov(Y, X)$

$Cov(Y, Y)$

covariance matrix

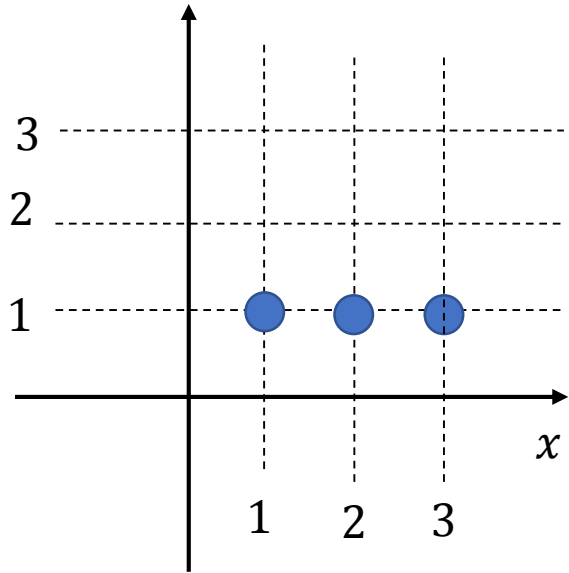


$$\begin{aligned}Cov(X, Y) &= Cov(Y, X) = \\ &= \frac{1}{3} ((1 - 2) + (2 - 2) + (3 - 2))(1 - 1) = 0\end{aligned}$$

$$\begin{aligned}Cov(Y, Y) &= \\ &= \frac{1}{3} ((1 - 1) + (1 - 1) + (1 - 1))(1 - 1) = 0\end{aligned}$$

$$\begin{aligned}Cov(X, X) &= \\ &= \frac{1}{3} ((1 - 2)^2 + (2 - 2)^2 + (3 - 2)^2) = \frac{2}{3}\end{aligned}$$

covariance matrix



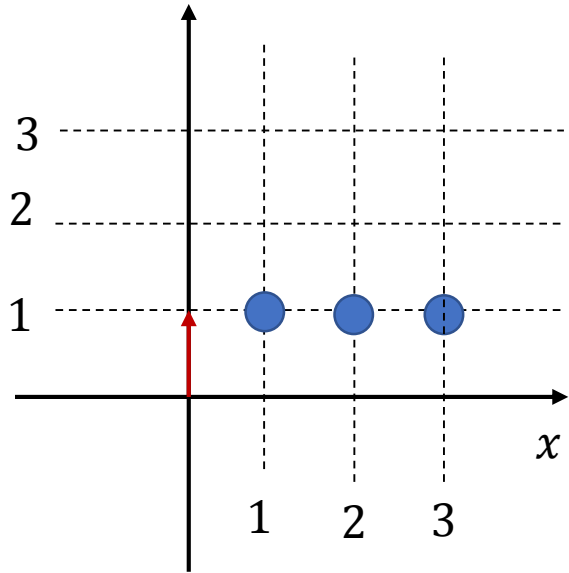
$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = 0$$

$$\text{Cov}(Y, Y) = 0$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} & 0 \\ 0 & 0 \end{bmatrix}$$

covariance matrix



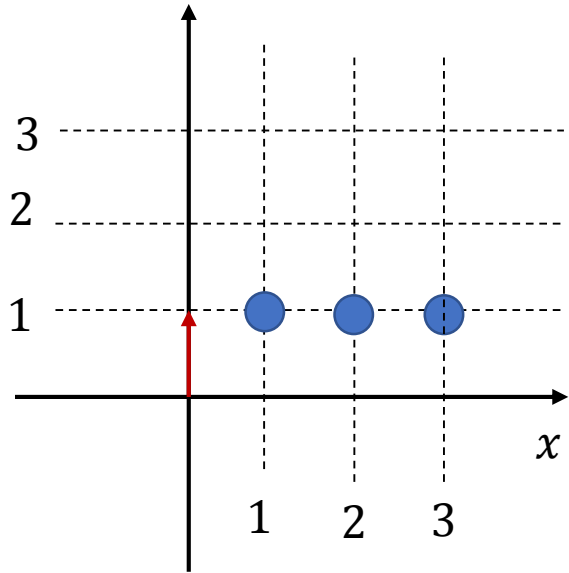
$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = 0$$

$$\text{Cov}(Y, Y) = 0$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

covariance matrix



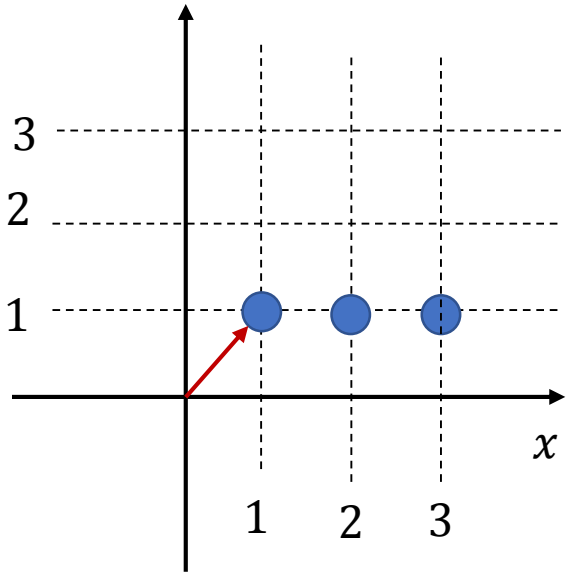
$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = 0$$

$$\text{Cov}(Y, Y) = 0$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

covariance matrix



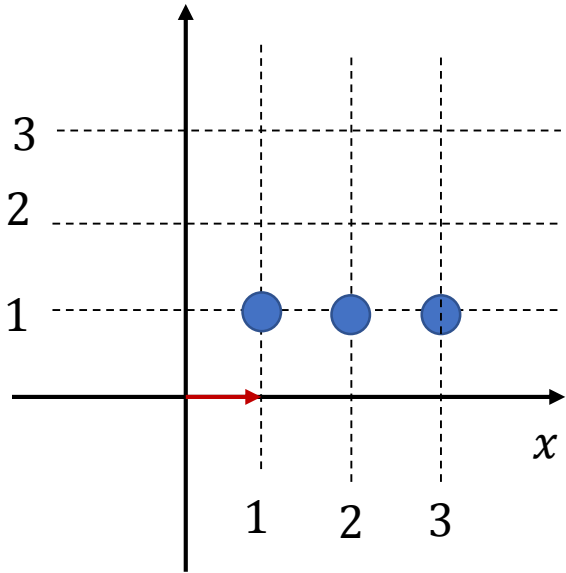
$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = 0$$

$$\text{Cov}(Y, Y) = 0$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} \\ 0 \end{bmatrix}$$

covariance matrix



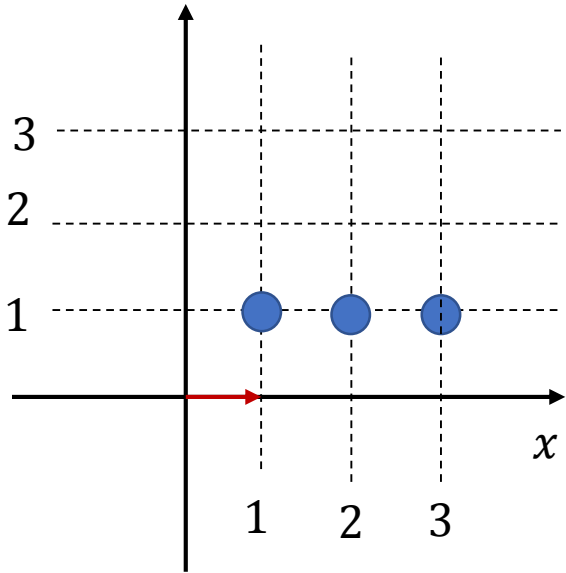
$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = 0$$

$$\text{Cov}(Y, Y) = 0$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} \\ 0 \end{bmatrix}$$

covariance matrix



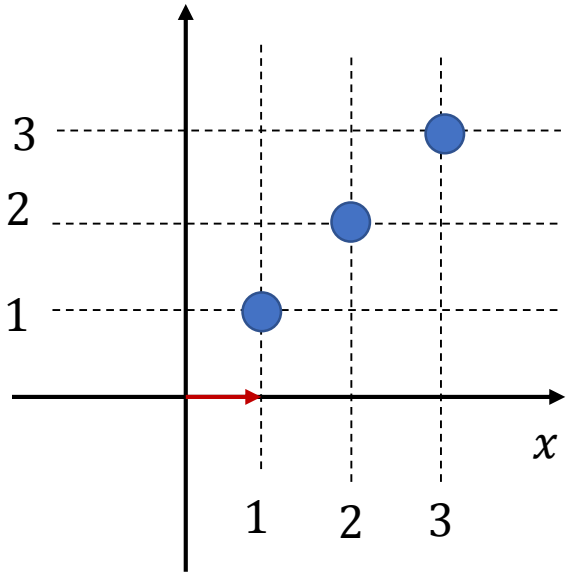
$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = 0$$

$$\text{Cov}(Y, Y) = 0$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{2}{3} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

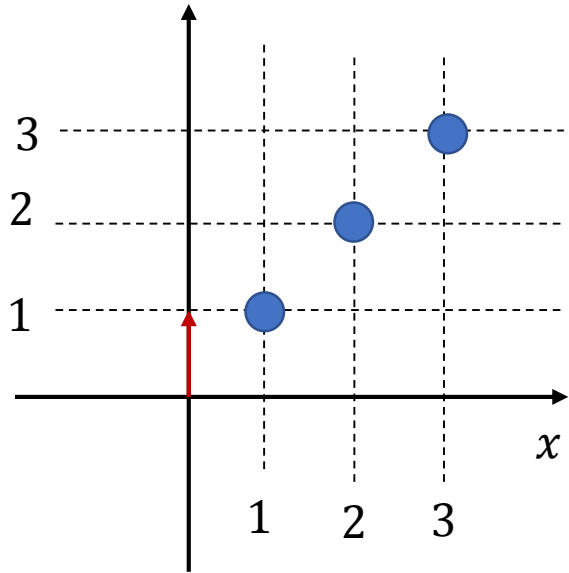
covariance matrix



$$\begin{aligned} \text{Cov}(X, Y) &= \text{Cov}(Y, X) = \frac{2}{3} \\ \text{Cov}(Y, Y) &= \frac{2}{3} \\ \text{Cov}(X, X) &= \frac{2}{3} \end{aligned}$$

$$\begin{bmatrix} \frac{2}{3} & \frac{2}{3} \\ \frac{2}{3} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} \\ \frac{2}{3} \end{bmatrix}$$

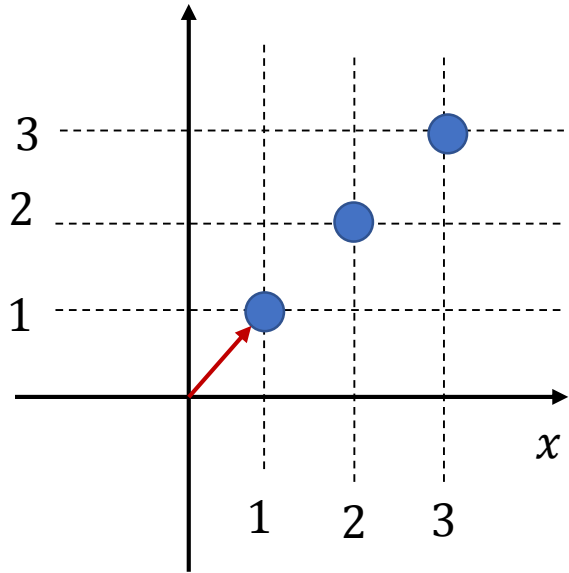
covariance matrix



$$\begin{aligned} \text{Cov}(X, Y) &= \text{Cov}(Y, X) = \frac{2}{3} \\ \text{Cov}(Y, Y) &= \frac{2}{3} \\ \text{Cov}(X, X) &= \frac{2}{3} \end{aligned}$$

$$\begin{bmatrix} \frac{2}{3} & \frac{2}{3} \\ \frac{2}{3} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} \\ \frac{2}{3} \end{bmatrix}$$

covariance matrix



$$\text{Cov}(X, Y) = \text{Cov}(Y, X) = \frac{2}{3}$$

$$\text{Cov}(Y, Y) = \frac{2}{3}$$

$$\text{Cov}(X, X) = \frac{2}{3}$$

$$\begin{bmatrix} \frac{2}{3} & \frac{2}{3} \\ \frac{2}{3} & \frac{2}{3} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{2}{3} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

eigenvector vector of a linear transformation

$$Kv = \lambda v$$

$$\begin{bmatrix} 2 & 2 \\ 3 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{2}{3} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

eigenvector vector of a linear transformation

$$K\mathbf{v} = \lambda\mathbf{v}$$

eigenvector

eigenvalue

$$K - \lambda I = M$$

$$M\mathbf{v} = \mathbf{0}$$

eigenvector vector of a linear transformation

$$K\mathbf{v} = \lambda\mathbf{v}$$

eigenvector

eigenvalue

$$K - \lambda I = M$$

$$M\mathbf{v} = \mathbf{0} \quad \Leftrightarrow \quad |M| = \mathbf{0}$$

determinant

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$K - \lambda I = M$$

$$|M| = \left| \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \right|$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$$\left| \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \right| = \left| \begin{array}{cc} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{array} \right|$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$$\begin{vmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{vmatrix} = 0 \qquad \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$$4 - 4\lambda + \lambda^2 - 1 = 0 \quad \left| \begin{array}{cc} a & b \\ c & d \end{array} \right| = ad - bc$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$$(\lambda - 1)(\lambda - 3) = 0$$

$$\lambda = 1; \lambda = 3$$

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$$\lambda = 1: \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$x \in \mathbb{R}; x \neq 0$

$$\lambda = 1: \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow v_{\lambda=1} = \begin{bmatrix} x \\ -x \end{bmatrix}$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$$\lambda = 3: \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} - \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

eigenvector vector of a linear transformation

$$K = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

$$|K - \lambda I| = 0$$

$x \in \mathbb{R}; x \neq 0$

$$\lambda = 3: \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow v_{\lambda=3} = \begin{bmatrix} x \\ x \end{bmatrix}$$