Statistics of Continuous Random Variables

Expected Value / Expectation

$$\mathrm{E}(X) = \int_{-\infty}^{\infty} x f(x) \mathrm{d}x$$

Expected Value and the Mean

Since $\int_{-\infty}^{\infty} f(x) dx = 1$, the expected value of a random variable is equal to the mean of the random variable.

$$E(X) = \int_{-\infty}^{\infty} xf(x)dx = \frac{\int_{-\infty}^{\infty} xf(x)dx}{\int_{-\infty}^{\infty} f(x)dx} = \mu$$

Expected Value of a Function of a Random Variable

$$\mathbf{E}[g(x)] = \int_{-\infty}^{\infty} g(x) f(x) \mathrm{d}x$$

Expected Value of Linear Combinations of Random Variables E(aX + b) = aE(X) + bE(aX + bY) = aE(X) + bE(Y)

Example VCAA 2017 NHT Exam 2 Question 3b

A company supplies schools with whiteboard pens. The total length of time for which a whiteboard pen can be used for writing before it stops working is called its use-time. There are two types of whiteboard pens: Grade A and Grade B. The use-time of Grade B whiteboard pens is described by the probability density function f where x is the use-time in hours

$$f(x) = \begin{cases} \frac{x}{576} (12 - x) \left(e^{\frac{x}{6}} - 1 \right) & 0 \le x \le 12\\ 0 & \text{otherwise} \end{cases}$$

The expected use-time of a Grade B whiteboard pen is

$$E(X) = \int_0^{12} xf(x) dx = \int_0^{12} \frac{x^2}{576} (12 - x) \left(e^{\frac{x}{6}} - 1\right) dx = \frac{9e^2 - 51}{2} \approx 7.75 \text{ hours}$$

Median

The value that separates the probability distribution exactly in half.

That is,
$$\Pr(X \le m) = \frac{1}{2} = \Pr(X \ge m)$$

$$\int_{-\infty}^{m} f(x) dx = \frac{1}{2}, \qquad \int_{m}^{\infty} f(x) dx = \frac{1}{2}, \qquad \text{where } m \text{ is the median}$$

Example VCAA 2014 Exam 1 Question 8a

 $f(x) = \begin{cases} \frac{1}{5}e^{-\frac{x}{5}} & x \ge 0\\ 0 & x < 0 \end{cases}$ A continuous random variable, X, has a probability density function given by *x*. The median of *X* is *m*.

$$\int_0^m \frac{1}{5} e^{-\frac{x}{5}} dx = \left[-e^{-\frac{x}{5}} \right]_0^m = -e^{-\frac{m}{5}} - (-e^0) = -e^{-\frac{m}{5}} + 1 = \frac{1}{2}$$

$$\Rightarrow e^{-\frac{m}{5}} = \frac{1}{2}$$
$$\Rightarrow -\frac{m}{5} = \log_e\left(\frac{1}{2}\right)$$

 $\Rightarrow m = 5 \log_e(2)$

Variance

The variance is the weighted average of the squared deviations from the mean.

$$\operatorname{var}(X) = \sigma^{2} = \operatorname{E}[(X - \mu)^{2}] = \int_{-\infty}^{\infty} (x - \mu)^{2} f(x) dx$$
$$= \int_{-\infty}^{\infty} (x^{2} - 2\mu x + \mu^{2}) f(x) dx = \int_{-\infty}^{\infty} x^{2} f(x) dx + \int_{-\infty}^{\infty} -2\mu x f(x) dx + \int_{-\infty}^{\infty} \mu^{2} f(x) dx$$
$$= \int_{-\infty}^{\infty} x^{2} f(x) dx - 2\mu \int_{-\infty}^{\infty} -x f(x) dx + \mu^{2} \int_{-\infty}^{\infty} f(x) dx = \operatorname{E}(X^{2}) - 2\mu \operatorname{E}(X) + \mu^{2}$$
$$= \operatorname{E}(X^{2}) - 2\mu^{2} + \mu^{2} = \operatorname{E}(X^{2}) - \mu^{2} = \operatorname{E}(X^{2}) - [\operatorname{E}(X)]^{2} = \int_{-\infty}^{\infty} x^{2} f(x) dx - \mu^{2}$$

Variance of Linear Combinations of Random Variables $var(aX + b) = a^2 var(X), \quad var(aX + bY) = a^2 var(X) + b^2 var(Y)$

Standard Deviation

The square root of the variance.

 $sd(X) = \sigma = \sqrt{var(X)}$

Standard Deviation of Linear Combinations of Random Variables sd(aX + b) = a sd(X) $sd(aX + bY) = \sqrt{var(aX + bY)} = \sqrt{a^2var(X) + b^2var(Y)}$

Example VCAA 2017 NHT Exam 2 Question 3c

The use-time of Grade B whiteboard pens is described by the probability density function f where

x is the use-time in hours $f(x) = \begin{cases} \frac{x}{576}(12-x)\left(e^{\frac{x}{6}}-1\right) & 0 \le x \le 12\\ 0 & \text{otherwise} \end{cases}$

The expected use-time of a Grade B whiteboard pen is 7.75 hours

$$\operatorname{var}(X) = \int_0^{12} (x - 7.75)^2 \frac{x}{576} (12 - x) \left(e^{\frac{x}{6}} - 1 \right) \mathrm{d}x \approx 5.3168$$

The standard deviation of the use-time of a Grade B whiteboard pen is $\sqrt{5.3168} \approx 2.31$ hours

Interpreting Statistics

If the standard deviation is large, the data is spread out over a widely from the mean. If the standard deviation is small, the data is clustered close to the mean. $\Pr(\mu - 2\sigma \le X \le \mu + 2\sigma) \approx 0.95$

Chebyshev's Theorem

At least
$$1 - \frac{1}{k^2}$$
 of the distribution lies within k standard deviations of the mean.
 $1 - \frac{1}{k^2} \le \Pr(\mu - k\sigma \le X \le \mu + k\sigma) \le 1, \quad k > 1$

For $k = 2, 0.75 \le \Pr(\mu - 2\sigma \le X \le \mu + 2\sigma) \le 1$