



Probability & Statistics,

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Definition:

A random variable is continuous if it can assume any value in interval of real numbers. Or

Range of X is an interval or union of intervals.



Definition:

Let *X* be a continuous random variable. A function *f* such that

$$1. f(x) \ge 0 \text{ for all } x$$

$$2. \int_{-\infty}^{\infty} f(x) dx = 1$$

3.
$$P[a \le X \le b] = \int_a^b f(x) dx$$

is called a density for X.

Note: $P(a \le X \le b) = P(a < X \le b) = P(a \le X < b) = P(a < X < b)$.







Necessary and sufficient conditions for function to be a continuous density:

$$1. f(x) \ge 0$$

$$2. \int_{-\infty}^{\infty} f(x) dx = 1.$$



Cumulative distribution function

Thus, for any value *x*,

$$F(x) = P(X \le x)$$

is the area under the probability density function over the interval $-\infty$ to *x*. Mathematically,

$$F(x) = \int_{-\infty}^{x} f(t) dt.$$

The probability that the random variable will take on a value on the interval from *a* to *b* is given by

$$P(a \le X \le b) = F(b) - F(a)$$



According to the fundamental theorem of integral calculus it follows that

 $\frac{dF(x)}{dx} = f(x)$ wherever this derivative exists.

⇔ *F* is non-decreasing function, $F(-\infty) = 0$ and $F(\infty) = 1$.



Mean of a continuous density:

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

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

kth moment:

MGF:

$$M_k = E(X^k) = \int_{-\infty}^{\infty} x^k f(x) dx. \qquad M_X(t) = E(e^{tX}) = \int_{-\infty}^{\infty} e^{tx} f(x) dx.$$



Variance of a continuous density

$$\sigma^{2} = \int_{-\infty}^{\infty} (x - \mu)^{2} f(x) dx$$
$$= \int_{-\infty}^{\infty} x^{2} f(x) dx - \mu^{2}$$
$$= E(X^{2}) - \mu^{2}.$$

 σ is referred to as the **standard deviation**



Example 1: If the probability density of a random variable is given by

$$f(x) = \begin{cases} x & \text{for } 0 < x < 1\\ 2 - x & \text{for } 1 \le x < 2\\ 0 & \text{elsewhere} \end{cases}$$

find the probabilities that a random variable having this probability density will take on a value

- (a) between 0.2 and 0.8;
- (b) between 0.6 and 1.2.

Solution:

(a)
$$P(0.2 \le X \le 0.8) = \int_{0.2}^{0.8} f(x) dx = \int_{0.2}^{0.8} x dx = \frac{x^2}{2} \Big|_{0.2}^{0.8} = 0.30.$$



(b)
$$P(0.6 \le X \le 1.2) = \int_{0.6}^{1.2} f(x) dx = \int_{0.6}^{1.0} x dx + \int_{1.0}^{1.2} (2-x) dx$$

$$= \frac{x^2}{2} \Big|_{0.6}^{1.0} + \left(2x - \frac{x^2}{2}\right) \Big|_{1.0}^{1.2} = 0.50$$

Example 2: With reference to the preceding example, find the corresponding distribution function and use it to determine the probabilities that a random variable having this distribution function will take on a value

- (a) greater than 1.8;
- (b) between 0.4 and 1.6.



Solution:

$$F(x) = \int_{-\infty}^{x} f(t)dt$$
If $x \le 0$, $F(x) = 0$.
If $0 < x < 1$, $F(x) = \int_{0}^{x} tdt = \frac{x^{2}}{2}$
If $1 \le x < 2$, $F(x) = \int_{0}^{1} tdt + \int_{1}^{x} (2-t)dt = 2x - \frac{x^{2}}{2} - 1$
If $x \ge 2$, $F(x) = \int_{0}^{1} tdt + \int_{1}^{2} (2-t)dt = 1$.

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Example 3: Find μ and σ^2 for the probability density of previous example.

Solution:

$$\mu = \int_{-\infty}^{\infty} xf(x)dx = \int_{0}^{1} x^{2}dx + \int_{1}^{2} x(2-x)dx = 1$$
$$E(X^{2}) = \int_{-\infty}^{\infty} x^{2}f(x)dx = \int_{0}^{1} x^{3}dx + \int_{1}^{2} x^{2}(2-x)dx = \frac{7}{6}$$
$$\sigma^{2} = \frac{7}{6} - 1 = \frac{1}{6}.$$