

## Key ideas for topic 11: Continuous distributions May 29

**Cumulative distribution:** The cumulative distribution function (CDF) for a continuous random variable  $X$  is

$$F(z) = \mathbf{P}(X \leq z) = \int_{-\infty}^z p(x)dx,$$

where  $p(z)$  is the probability density function (pdf) for a random variable  $x$

$$p(x) = \lim_{\delta \rightarrow 0} \frac{\mathbf{P}(x - \delta/2 \leq x \leq x + \delta/2)}{\delta}.$$

A standard way of thinking about the pdf is via the cdf

$$p(x) = \frac{dF(x)}{dx}.$$

**Expectation:** The population mean. For a continuous random variable  $X$

$$\mu = \mathbb{E}[X] = \int x p(x)df,$$

Linearity of expectations. For a collection of random variables  $X_1, \dots, X_n$

$$\mathbb{E} \left[ \sum_i X_i \right] = \sum_i \mathbb{E}[X_i].$$

Expectation of a function  $f(s)$  given a random variable  $X$  is

$$\mathbb{E}[f(x)] = \int_x f(x)p(x)dx.$$

**Variance:** The variance of a random variable measures its spread. For a continuous random variable  $X$

$$\sigma^2 = \text{Var}(X) = \int_x (x - \mu)^2 p(x)dx.$$

The standard deviation is  $\sigma = \sqrt{\text{Var}(X)}$ .

**The uniform distribution:** The uniform distribution is perhaps the most simple continuous distribution. A random variable  $X$  follows the uniform distribution on  $[0, 1]$ , denoted as  $X \sim U(0, 1)$ , if  $X$  has density

$$f_X(x) = 1, \quad 0 < x < 1,$$

and 0 otherwise. In general if  $Y = aX + b$ , and  $X$  as above, then  $Y$  follows the uniform distribution  $Y \sim U(b, a + b)$ . This can be asserted by looking at the cumulative distributions of  $X$  and  $Y$ .

Another example of a continuous distribution is the normal distribution that we have seen in binomial approximation.

## Key ideas for topic 12: Exponential and Gamma distributions June 01

**Exponential distribution:** A random variable  $X$  follows the exponential distribution with mean  $1/\lambda$ ,  $\lambda > 0$  if the density of  $X$  is

$$f_X(x) = \lambda e^{-\lambda x}, \quad x \geq 0,$$

and 0 otherwise. The exponential distribution can be seen as the continuous analog of the geometric distribution. The exponential distribution is a special case of the

**Gamma distribution:** A random variable  $X$  follows the gamma distribution with parameters  $k, \lambda$ , ( $k, \lambda > 0$ ) if it has the density

$$f_X(x) = \frac{\lambda^k}{\Gamma(k)} x^{k-1} e^{-\lambda x}, \quad x \geq 0,$$

and 0 otherwise. The gamma function is a generalization of the factorial  $\Gamma(k) = (k-1)!$  for  $k$  a positive integer. In that case the Gamma distribution can be seen as the sum of  $k$  independent exponential random variables, each with parameter  $\lambda$ .

Both of these distributions can be studied via a **Poisson process**  $N_t$ . This is an example of a stochastic process, where for each  $t \geq 0$  we assign a Poisson random variable  $N_t := N_{t-0} \sim \text{Poisson}(\lambda t)$ . In the Poisson process the random variables are connected such that  $N_{[t_4-t_3]}$  and  $N_{[t_2-t_1]}$  are independent whenever  $t_1 < t_2 < t_3 < t_4$ .

The arrival times  $T_k$  of the process follow the Gamma distribution with parameters  $k, \lambda$ , and the inter-arrival times (time between consecutive arrivals) are iid exponential with parameter  $\lambda$ .

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## Key ideas for topic 13: Transformations June 2

**Change of variables:** For a random variable  $X$  with density  $p_X(x)$  and range  $(a, b)$  and  $Y = g(X)$  where  $g$  is strictly increasing or decreasing on  $(a, b)$  with  $X = v(Y)$  ( $v = g^{-1}$ ) the density of  $Y$  is

$$\begin{aligned} p_Y(y) &= p_X(x) \times \left| \frac{dy}{dx} \right|^{-1}, \\ p_Y(y) &= p_X(v(y)) \times v'(y), \end{aligned}$$

over  $u(a) < y < u(b)$ .

This is a consequence of the chain rule applied to the CDFs of  $X$  and  $Y$ .

**Moment generating function (mgf):** For  $X$  a random variable, the mgf is defined as:

$$\psi_X(t) = \mathbb{E}[e^{tX}], \quad t \in (-\infty, \infty).$$

If the mgf is finite in an open interval around  $t = 0$ , it completely determines the law of the random variable  $X$ . For such random variables, The MGF can be used to find moments of  $X$  ( $k$ th moment is the  $k$ th derivative evaluated at  $t = 0$ ) or to find the distribution of sum of iid random variables.