Chapter 4. Continuous Random Variables and Probability **Distributions**

Math 3670 Summer 2024

Georgia Institute of Technology

[Section 1.](#page-1-0) [Probability Density Functions](#page-1-0) Suppose *X* is the depth of a lake at a randomly chosen point on the surface.

Let *M* be the maximum depth (in meters), so that any number in the interval [0*, M*] is a possible value of *X*.

If we "discretize" *X* by measuring depth to the nearest meter, then possible values are nonnegative integers less than or equal to *M*.

The resulting discrete distribution of depth can be pictured using a probability histogram.

If depth is measured much more accurately and the same measurement axis, each rectangle in the resulting probability histogram is much narrower, though the total area of all rectangles is still 1.

We say a random variable *X* is **continuous** if there exists a function $f(x)$ such that

- 1. $f(x) \geq 0$ for all *x*,
- 2. ∫ *[∞] −∞ ^f*(*x*) *dx* ⁼ 1, and
- 3. $\mathbb{P}(a \le X \le b) = \int_a^b f(x) dx$ for all a, b .

The function *f*(*x*) is called the probability density function (PDF) of *X*.

The direction of an imperfection with respect to a reference line on a circular object such as a tire, brake rotor, or flywheel is, in general, subject to uncertainty.

Consider the reference line connecting the valve stem on a tire to the center point. Let *X* be the angle measured clockwise to the location of an imperfection with PDF

$$
f(x) = \begin{cases} \frac{1}{360}, & 0 \le x < 360, \\ 0, & \text{otherwise.} \end{cases}
$$

What is the probability that the angle is between 90*◦* and 180*◦* ?

A continuous RV *X* is said to have a uniform distribution on the interval [*A, B*] if the PDF of *X* is

 $f(x) =$

We denote by *X ∼* Unif(*A, B*).

Let *X* be a continuous RV with PDF

$$
f(x) = \begin{cases} 3e^{-3x}, & x \ge 0 \\ 0, & x < 0. \end{cases}
$$

Find the probability $P(X \le 5)$.

Properties

For a continuous RV *X*,

1. $P(X = c) =$

2. $P(a \leq X \leq b) =$

Let *X* be a continuous RV with PDF

$$
f(x) = \begin{cases} cx(1-x), & 0 \le x \le 1\\ 0, & \text{otherwise.} \end{cases}
$$

- 1. Find the constant $c > 0$.
- 2. Find the probability $\mathbb{P}(X \geq \frac{1}{3})$.

[Section 2.](#page-10-0) [Cumulative Distribution Functions](#page-10-0) [and Expected Values](#page-10-0)

The cumulative distribution function *F*(*x*) for a continuous RV *X* is defined by

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

Let *X ∼* Unif(*A, B*). Find the CDF.

Let *X* be a continuous RV with PDF

$$
f(x) = \begin{cases} \frac{1}{8} + \frac{3x}{8}, & 0 \le x \le 2\\ 0, & \text{otherwise.} \end{cases}
$$

- 1. Find the CDF.
- 2. Find the probability $P(1 \le X \le 1.5)$.

Proposition

If *X* is a continuous RV with PDF *f*(*x*) and CDF *F*(*x*), then at every *x* at which the derivative *F ′* (*x*) exists,

 $F'(x) = f(x)$.

For 0 *≤ p ≤* 1, the (100*p*)-th percentile of the distribution of a continuous RV *X*, denoted by $\eta(p)$, is defined by

$$
p = F(\eta(p)) =
$$

In particular, the 50th percentile is called the median and denoted by $\tilde{\mu}$.

Let *X* be a continuous RV with PDF

$$
f(x) = \begin{cases} \frac{3}{2}(1 - x^2), & 0 \le x \le 1 \\ 0, & \text{otherwise.} \end{cases}
$$

Find the 25th and 50th percentiles.

The expected or mean value of a continuous RV *X* with PDF *f*(*x*) is

$$
\mathbb{E}[X] = \mu_X =
$$

Let *X* be a continuous RV with PDF

$$
f(x) = \begin{cases} \frac{3}{2}(1 - x^2), & 0 \le x \le 1 \\ 0, & \text{otherwise.} \end{cases}
$$

Find E[*X*].

Proposition

If *X* is a continuous RV with PDF *f*(*x*) and *h*(*X*) is a function of *X*, then

 $\mathbb{E}[h(X)] =$

If you break a stick of length 1 at random into two pieces, what is the expected length of the longer piece?

The variance of a continuous random variable *X* with PDF *f*(*x*) is

 $Var(X) =$

The standard deviation (SD) of *X* is

 $\sigma_X =$

Proposition

 $\mathsf{Var}(\mathsf{X}) =$ $Var(aX + b) =$ Let *X* be a continuous RV with PDF

$$
f(x) = \begin{cases} x, & 0 \le x \le 1 \\ 2 - x, & 1 \le x \le 2 \\ 0, & \text{otherwise.} \end{cases}
$$

Find the CDF and draw the graph.

[Section 3.](#page-24-0) [The Normal Distribution](#page-24-0)

A continuous RV *X* is said to have a normal distribution with parameters $\mu \in \mathbb{R}$ and *σ >* 0 if the PDF of *X* is

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

We denote by $X \sim N(\mu, \sigma^2)$. Note that $\mathbb{E}[X] = \mu$ and $\textsf{Var}(X) = \sigma^2$.

The Normal Distribution

The normal distribution with parameters $\mu = 0, \sigma = 1$ is called the standard normal distribution.

Usually, it is denoted by *Z ∼ N*(0*,* 1).

The PDF is

$$
\mathit{f}(z;0,1)=
$$

and the CDF is

$$
\Phi(x) =
$$

For *Z ∼ N*(0*,* 1), find

- 1. $P(Z \leq 1.25)$
- 2. $P(Z > 1.25)$
- 3. ^P(*^Z ≤ −*1*.*25)
- 4. ^P(*−.*³⁸ *[≤] ^Z [≤]* ¹*.*25)

Find the 75th percentile of the standard normal distribution.

 z_α will denote the value on the *z* axis for which α of the area under the *z* curve lies to the right of z_α .


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Proposition
If X ∼ N(µ, σ2
), then aX + b is also normal and
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aX + *b ∼*

In particular,

The time that it takes a driver to react to the brake lights on a decelerating vehicle is critical in helping to avoid rear-end collisions.

The article "Fast-Rise Brake Lamp as a Collision-Prevention Device" (Ergonomics, 1993: 391–395) suggests that reaction time for an in-traffic response to a brake signal from standard brake lights can be modeled with a normal distribution having mean value 1.25 sec and standard deviation of 46 sec.

What is the probability that reaction time is between 1.00 sec and 1.75 sec?

(4.3-32) Suppose the force acting on a column that helps to support a building is a normally distributed random variable *X* with mean value 15.0 kips and standard deviation 1.25 kips.

Find $P(X \le 15)$ and $P(14 \le X \le 18)$.

IQ in a particular population (as measured by a standard test) is known to be approximately normally distributed with $\mu = 100$ and $\sigma = 15$.

What is the probability that a randomly selected individual has an IQ of at least 125?

Proposition

Let *X ∼* Bin(*n, p*).

If the binomial probability histogram is not too skewed, then *X* has approximately a normal distribution with $\mu = np$ and $\sigma^2 = np(1-p)$.

In practice, the approximation is adequate if

$$
np \ge 10, \qquad n(1-p) \ge 10,
$$

since there is then enough symmetry in the underlying binomial distribution.

Suppose that 25% of all students at a large public university receive financial aid.

Let *X* be the number of students in a random sample of size 50 who receive financial aid, so that $p = .25$.

What is the probability that at most 10 students receive aid?

(4.3-55) Suppose only 75% of all drivers in a certain state regularly wear a seat belt. A random sample of 500 drivers is selected.

What is the probability that

- 1. Between 360 and 400 (inclusive) of the drivers in the sample regularly wear a seat belt?
- 2. Fewer than 400 of those in the sample regularly wear a seat belt?

[Section 4.](#page-38-0) [The Exponential and Gamma](#page-38-0) **[Distributions](#page-38-0)**

A random variable *X* is said to have an exponential distribution with parameter *λ >* 0 if the PDF of *X* is

$$
f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \ge 0 \\ 0, & x < 0. \end{cases}
$$

We denote by $X \sim \text{Exp}(\lambda)$.

Proposition For $X \sim \text{Exp}(\lambda)$, $\mathbb{E}[X] =$ $\mathsf{Var}(\mathsf{X}) =$ $F(X) =$

The article "Probabilistic Fatigue Evaluation of Riveted Railway Bridges" (J. of Bridge Engr., 2008: 237–244) suggested the exponential distribution with mean value 6 MPa as a model for the distribution of stress range in certain bridge connections.

Let's assume that this is in fact the true model.

Find the probability that stress range is at most 10 MPa.

Proposition

Suppose that the number of events occurring in any time interval of length *t* has a Poisson distribution with parameter *αt*.

Further assume that numbers of occurrences in nonoverlapping intervals are independent of one another.

Then the distribution of elapsed time between the occurrence of two successive events is exponential with parameter $\lambda = \alpha$.

Suppose that calls are received at a 24-hour "suicide hotline" according to a Poisson process with rate a $\alpha = 5$ call per day.

Let *X* be the number of days *X* between successive calls.

What is the probability that more than 2 days elapse between calls?

Memoryless Property

For $X \sim \text{Exp}(\lambda)$,

 $P(X \geq s + t | X \geq s) =$

For $\alpha > 0$, the Gamma function is defined by

$$
\Gamma(\alpha) =
$$

For example,

- 1. $Γ(1) =$
- 2. $Γ(2) =$
- 3. In general, Γ(*n*) =
- 4. Γ(1*/*2) =

A random variable *X* is is said to have a Gamma distribution with parameters *α, β >* 0 if the PDF of *X* is

$$
f(x; \alpha, \beta) = \begin{cases} \frac{1}{\beta^{\alpha} \Gamma(\alpha)} x^{\alpha - 1} e^{-x/\beta}, & x \ge 0 \\ 0, & x < 0. \end{cases}
$$

We denote by $X \sim \text{Gamma}(\alpha, \beta)$.

If $\beta = 1$, *X* is called a standard Gamma random variable.

α is called the shape parameter.

β is called the scale parameter.

The Gamma Distribution

Figure 4.27 (a) Gamma density curves; (b) standard gamma density curves

Proposition

For $X \sim \text{Gamma}(\alpha, \beta)$,

 $\mathbb{E}[X] = \alpha \beta$ $Var(X) = \alpha \beta^2$.

A random variable *X* is said to have a chi-squared distribution with parameter *ν* if $X \sim \text{Gamma}(\nu/2, 2)$.

The parameter *ν* is called the number of degrees of freedom of *X*.

We denote by $X \sim \chi^2(\nu)$.

(4.4-70) If *X ∼* Exp(*λ*), find the 100*p*-th percentile and the median for 0 *< p <* 1.

[Section 6.](#page-51-0) [Probability Plots](#page-51-0)

An investigator will often have obtained a numerical sample x_1, x_2, \cdots, x_n and wish to know whether it is plausible that it came from a population distribution of some particular type (e.g., from a normal distribution).

For one thing, many formal procedures from statistical inference are based on the assumption that the population distribution is of a specified type.

The use of such a procedure is inappropriate if the actual underlying probability distribution differs greatly from the assumed type.

Understanding the underlying distribution can sometimes give insight into the physical mechanisms involved in generating the data.

An effective way to check a distributional assumption is to construct what is called a probability plot.

The essence of such a plot is that if the distribution on which the plot is based is correct, the points in the plot should fall close to a straight line.

If the actual distribution is quite different from the one used to construct the plot, the points will likely depart substantially from a linear pattern.

Order the *n* sample observations from smallest to largest.

Then the *i*-th smallest observation in the list is taken to be the [100(*i −* 0*.*5)*/n*]-th sample percentile.

The sample consisting of *n* = 20 observations on dielectric breakdown voltage of a piece of epoxy resin is

24.46 25.61 26.25 26.42 26.66 27.15 27.31 27.54 27.74 27.94 27.98 28.04 28.28 28.49 28.5 28.87 29.11 29.13 29.5 30.88

Find the sample percentiles.

A plot of the n pairs

([100(*i − .*5)*/n*]-th z percentile, *i*-th smallest observation)

on a two-dimensional coordinate system is called a normal probability plot.

If the sample observations are in fact drawn from a normal distribution with mean value *µ* and standard deviation *σ*, the points should fall close to a straight line with slope *σ* and intercept *µ*.

Thus a plot for which the points fall close to some straight line suggests that the assumption of a normal population distribution is plausible.

Example

The sample consisting of *n* = 20 observations on dielectric breakdown voltage of a piece of epoxy resin is

24.46 25.61 26.25 26.42 26.66 27.15 27.31 27.54 27.74 27.94 27.98 28.04 28.28 28.49 28.5 28.87 29.11 29.13 29.5 30.88

Observation 24.46 25.61 26.25 26.42 26.66 27.15 27.31 27.54 27.74 27.94 z percentile -1.96 -1.44 -1.15 -.93 -.76 -.60 -.45 -.32 -.19 -.06 28.49 28.87 Observation 27.98 28.04 28.28 28.50 29.11 29.13 29.50 30.88 *z* percentile .06 .19 $.32$.45 .60 $.76$.93 1.15 1.44 1.96

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Figure 4.37 Probability plots that suggest a nonnormal distribution: (a) a plot consistent with a heavy-tailed distribution; (b) a plot consistent with a positively skewed distribution