

STAT 3375Q: Introduction to Mathematical Statistics I

Lecture 18: Functions of Random Variables (Univariate)

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 - ▶ The CDF Method
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 - ▶ The MGF Method

Midterm 2 Solutions

Problem 1

Suppose X and Y are independent Gaussian random variables. That is, $X \sim \mathcal{N}(1, 4)$ and $Y \sim \mathcal{N}(0, 7)$.

- a Find $\text{Cov}(X, Y)$.

Solution:

Since X and Y are independent, $\text{Cov}(X, Y) = 0$. □

Problem 1

Suppose X and Y are independent Gaussian random variables. That is, $X \sim \mathcal{N}(1, 4)$ and $Y \sim \mathcal{N}(0, 7)$.

- ⓑ Find $E(X^2 Y^2)$.

Solution:

- ▶ Since X and Y are independent, we can split the expected value as follows:

$$E(X^2 Y^2) = E(X^2)E(Y^2).$$

- ▶ Solving for $E(X^2)$, we have

$$\begin{aligned} E(X^2) &= V(X) + \{E(X)\}^2 && \text{def'n of variance} \\ &= 4 + 1^2 && X \sim \mathcal{N}(1, 4) \\ &= 5. \end{aligned}$$

- ▶ Solving for $E(Y^2)$, we have

$$\begin{aligned} E(Y^2) &= V(Y) + \{E(Y)\}^2 && \text{def'n of variance} \\ &= 7 + 0^2 && Y \sim \mathcal{N}(0, 7) \\ &= 7. \end{aligned}$$

- ▶ Therefore, $E(X^2 Y^2) = E(X^2)E(Y^2) = (5)(7) = 35$. □

Problem 1

Suppose X and Y are independent Gaussian random variables. That is, $X \sim \mathcal{N}(1, 4)$ and $Y \sim \mathcal{N}(0, 7)$.

© Find $E(3X - 2Y)$.

Solution:

By the linearity of expectation, we have

$$\begin{aligned} E(3X - 2Y) &= 3E(X) - 2E(Y) \\ &= 3(1) - 2(0) \quad X \sim \mathcal{N}(1, 4) \text{ and } Y \sim \mathcal{N}(0, 7) \\ &= 3. \end{aligned}$$



Problem 1

Suppose X and Y are independent Gaussian random variables. That is, $X \sim \mathcal{N}(1, 4)$ and $Y \sim \mathcal{N}(0, 7)$.

d Find $V(3X - 2Y)$.

Solution:

$$V(3X - 2Y) = V(3X) + V(-2Y)$$

Variance of the sum of independent RVs: $V(X + Y) = V(X) + V(Y)$

$$= 3^2 V(X) + (-2)^2 V(Y) \quad \text{Variance of a linear transform: } V(aX + b) = a^2 V(X)$$

$$= 9(4) + 4(7) \quad X \sim \mathcal{N}(1, 4) \text{ and } Y \sim \mathcal{N}(0, 7)$$

$$= 64.$$



Problem 1

Suppose X and Y are independent Gaussian random variables. That is, $X \sim \mathcal{N}(1, 4)$ and $Y \sim \mathcal{N}(0, 7)$.

- e Find $P(-3 \leq 3X - 2Y \leq 5)$. Hint: Sum of 2 Gaussian RVs is a Gaussian RV.

Solution:

- ▶ Let $W = 3X - 2Y$.
- ▶ From part c) and d), we know that $W \sim \mathcal{N}(\mu = 3, \sigma^2 = 64)$.

$$\begin{aligned}P(-3 \leq W \leq 5) &= P\left(\frac{-3 - \mu}{\sigma} \leq \frac{W - \mu}{\sigma} \leq \frac{5 - \mu}{\sigma}\right) && \text{standardization} \\&= P\left(\frac{-3 - 3}{\sqrt{64}} \leq \frac{W - 3}{\sqrt{64}} \leq \frac{5 - 3}{\sqrt{64}}\right) \\&= P\left(-\frac{6}{8} \leq Z \leq \frac{2}{8}\right) \\&= \Phi\left(\frac{1}{4}\right) - \Phi\left(-\frac{3}{4}\right) && \text{probability = area under the standard normal curve} \\&= 0.59871 - 0.22663 && \text{Z-table values} \\&= 0.3721. \quad \square\end{aligned}$$

Problem 2

Consider a random variable X with the PDF

$$f(x) = A + Bx^2, \quad 0 \leq x \leq 2.$$

If $E(X) = 1/2$, find A and B .

Solution:

- ▶ There are two unknowns, A and B . We will need two linear equations to find their values.
- ▶ Since $f(x)$ is a valid density, it must integrate to 1.

$$\begin{aligned} 1 &= \int_{-\infty}^{\infty} f(x) dx \\ &= \int_0^2 A + Bx^2 dx \\ &= Ax + B \frac{x^3}{3} \Big|_0^2 \\ &= 2A + \frac{8}{3}B. \end{aligned}$$

(cont'd next slide...)

Problem 2

Consider a random variable X with the PDF

$$f(x) = A + Bx^2, \quad 0 \leq x \leq 2.$$

If $E(X) = 1/2$, find A and B .

Solution:

- ▶ Also, we need $E(X) = 1/2$. This means

$$\begin{aligned} \frac{1}{2} &= \int_{-\infty}^{\infty} xf(x)dx \\ &= \int_0^2 x(A + Bx^2)dx \\ &= A \frac{x^2}{2} + B \frac{x^4}{4} \Big|_0^2 \\ &= 2A + 4B. \end{aligned}$$

(cont'd next slide...)

Problem 2

Consider a random variable X with the PDF

$$f(x) = A + Bx^2, \quad 0 \leq x \leq 2.$$

If $E(X) = 1/2$, find A and B .

Solution:

- ▶ Solving the following system of linear equations, we have

$$\begin{cases} 2A + \frac{8}{3}B = 1 \\ 2A + 4B = \frac{1}{2} \end{cases}$$

$$\Rightarrow \frac{4}{3}B = -\frac{1}{2} \quad \text{subtracting the 1st eqn from the 2nd}$$

$$\Rightarrow B = -\frac{3}{8}$$

$$\Rightarrow 2A + \frac{8}{3} \left(-\frac{3}{8} \right) = 1 \quad \text{substituting the value of } B \text{ to the 1st eqn}$$

$$\Rightarrow A = 1.$$

Thus, we have the following PDF:

$$f(x) = 1 - \frac{3}{8}x^2, \quad 0 \leq x \leq 2. \quad \square$$

Problem 3

Suppose that the completion time in hours T for the STAT 3375Q final exam follows a distribution with density

$$f(t) = \frac{2}{27}(t^2 + t), \quad 0 \leq t \leq 3.$$

What is the probability that a randomly chosen student finishes the exam during the first 30 minutes.

Solution:

$$\begin{aligned} P\left(T \leq \frac{1}{2}\right) &= \int_0^{\frac{1}{2}} \frac{2}{27}(t^2 + t) dt \quad \text{probability = area under density curve} \\ &= \frac{2}{27} \left(\frac{t^3}{3} + \frac{t^2}{2} \right) \Big|_0^{\frac{1}{2}} = \frac{2}{27} \left\{ \frac{1}{3} \left(\frac{1}{2} \right)^3 + \frac{1}{2} \left(\frac{1}{2} \right)^2 \right\} \\ &= \frac{2}{27} \left(\frac{1}{24} + \frac{1}{8} \right) = \frac{2}{27} \left(\frac{4}{24} \right) \\ &= \frac{1}{81}. \quad \square \end{aligned}$$

Problem 4

Given that X has MGF

$$m(t) = \frac{1}{6}e^{-2t} + \frac{1}{3}e^{-t} + \frac{1}{4}e^t + \frac{1}{4}e^{2t},$$

what is the probability that X is even.

Solution:

Matching the MGF above to the MGF formula

$m(t) = E(e^{tX}) = \sum_y e^{tx} p(x)$, we know that the given MGF corresponds to a discrete random variable with PMF:

$$p(x) = \begin{cases} \frac{1}{6}, & \text{if } x = -2, \\ \frac{1}{3}, & \text{if } x = -1, \\ \frac{1}{4}, & \text{if } x = 1, \\ \frac{1}{4}, & \text{if } x = 2. \end{cases}$$

Therefore,

$$P(X \text{ is even}) = P(X = -2) + P(X = 2) = \frac{1}{6} + \frac{1}{4} = \frac{5}{12}. \quad \square$$

Problem 5

Suppose X and Y are continuous random variables with joint PDF

$$f(x, y) = \begin{cases} 4xy, & \text{if } 0 \leq x \leq 1; 0 \leq y \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

- a Find the marginal PDF of X , $f(x)$, and Y , $f(y)$.

Solution:

$$\begin{aligned} f(x) &= \int_{-\infty}^{\infty} f(x, y) dy \\ &= \int_0^1 4xy dy \\ &= 4x \frac{y^2}{2} \Big|_0^1 \\ &= 2x, \quad 0 \leq x \leq 1. \end{aligned}$$

Similarly, $f(y) = 2y$, $0 \leq y \leq 1$.



Problem 5

Suppose X and Y are continuous random variables with joint PDF

$$f(x, y) = \begin{cases} 4xy, & \text{if } 0 \leq x \leq 1; 0 \leq y \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

b Find the conditional PDF of Y given X , $f(y|x)$.

Solution:

$$\begin{aligned} f(y|x) &= \frac{f(x, y)}{f(x)} \\ &= \frac{4xy}{2x} \\ &= 2y, \quad 0 \leq y \leq 1. \end{aligned}$$



Problem 5

Suppose X and Y are continuous random variables with joint PDF

$$f(x, y) = \begin{cases} 4xy, & \text{if } 0 \leq x \leq 1; 0 \leq y \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

Ⓒ Find $P(Y \leq 3/4 | X = 1/2)$.

Solution:

$$\begin{aligned} P(Y \leq 3/4 | X = 1/2) &= \int_0^{3/4} f(y | x = 1/2) dy \\ &= \int_0^{3/4} 2y dy \quad \text{Using the conditional PDF in part b)} \\ &= y^2 \Big|_0^{3/4} = \frac{9}{16}. \end{aligned}$$



Problem 5

Suppose X and Y are continuous random variables with joint PDF

$$f(x, y) = \begin{cases} 4xy, & \text{if } 0 \leq x \leq 1; 0 \leq y \leq 1, \\ 0, & \text{elsewhere.} \end{cases}$$

d Find $E(Y|X = x)$.

Solution:

$$\begin{aligned} E(Y|X = x) &= \int_{-\infty}^{\infty} yf(y|x)dy && \text{def'n of conditional expectation} \\ &= \int_0^1 y(2y)dy = \left. \frac{2y^3}{3} \right|_0^1 = \frac{2}{3}. \end{aligned}$$



Problem 6

Let X be a random variable with MGF

$$m(t) = \begin{cases} \frac{e^t - e^{-t}}{2t}, & \text{if } t \neq 0 \\ 1, & \text{if } t = 0. \end{cases}$$

- a Give the distribution of X .

Solution:

Matching the MGF above with known MGF formulas, we know that $X \sim U(-1, 1)$, where $\theta_1 = -1$ and $\theta_2 = 1$. □

Problem 6

Let X be a random variable with MGF

$$m(t) = \begin{cases} \frac{e^t - e^{-t}}{2t}, & \text{if } t \neq 0 \\ 1 & \text{if } t = 0. \end{cases}$$

• Compute $E(X)$ and $V(X)$.

Solution:

Using the mean and variance formula of a uniform RV, we have

$$\begin{aligned} E(X) &= \frac{\theta_1 + \theta_2}{2} = \frac{-1 + 1}{2} = 0 \\ V(X) &= \frac{(\theta_2 - \theta_1)^2}{12} = \frac{\{1 - (-1)\}^2}{12} = \frac{4}{12} = \frac{1}{3}. \end{aligned}$$



Problem 7

Let X and Y be random variables such that

$$E(X) = 1, \quad E(X^2) = 3, \quad E(XY) = -4, \quad E(Y) = 2, \quad V(Y) = 25.$$

a Find $E(2X + Y)$.

Solution:

$$\begin{aligned} E(2X + Y) &= 2E(X) + E(Y) && \text{linearity of expectation} \\ &= 2(1) + 2 && \text{given} \\ &= 4. \end{aligned}$$



Problem 7

Let X and Y be random variables such that

$$E(X) = 1, \quad E(X^2) = 3, \quad E(XY) = -4, \quad E(Y) = 2, \quad V(Y) = 25.$$

b Find $E\{X(2X + Y)\}$.

Solution:

$$\begin{aligned} E\{X(2X + Y)\} &= E(2X^2 + XY) \\ &= 2E(X^2) + E(XY) && \text{linearity of expectation} \\ &= 2(3) + (-4) && \text{given} \\ &= 2. \end{aligned}$$



Problem 7

Let X and Y be random variables such that

$$E(X) = 1, \quad E(X^2) = 3, \quad E(XY) = -4, \quad E(Y) = 2, \quad V(Y) = 25.$$

© Find $\text{Cov}(X, 2X + Y)$.

Solution:

$$\text{Cov}(X, 2X + Y) = E\{X(2X + Y)\} - E(X)E(2X + Y)$$

def'n of covariance: $\text{Cov}(X, Y) = E(XY) - E(X)E(Y)$

$$= 2 - (1)(4) \quad \text{answers from part a) and b)}$$

$$= -2.$$



Problem 7

Let X and Y be random variables such that

$$E(X) = 1, \quad E(X^2) = 3, \quad E(XY) = -4, \quad E(Y) = 2, \quad V(Y) = 25.$$

d Find $V(2X + Y)$.

Solution:

$$V(2X + Y) = 2^2V(X) + V(Y) + 2\text{Cov}(2X, Y)$$

Variance of the sum: $V(X + Y) = V(X) + V(Y) + 2\text{Cov}(X, Y)$

$$= 4V(X) + V(Y) + 2(2)\text{Cov}(X, Y)$$

Covariance of linear transform: $\text{Cov}(aX + b, cY + d) = ac\text{Cov}(X, Y)$. Here $a = 2, c = 1$.

$$= 4[E(X^2) - \{E(X)\}^2] + 25 + 4\{E(XY) - E(X)E(Y)\}$$

def'n of variance and covariance

$$= 4(3 - 1^2) + 25 + 4(-4 - (1)(2))$$

$$= 9.$$

Problem 7

Let X and Y be random variables such that

$$E(X) = 1, \quad E(X^2) = 3, \quad E(XY) = -4, \quad E(Y) = 2, \quad V(Y) = 25.$$

e Find $\text{Corr}(X, 2X + Y)$.

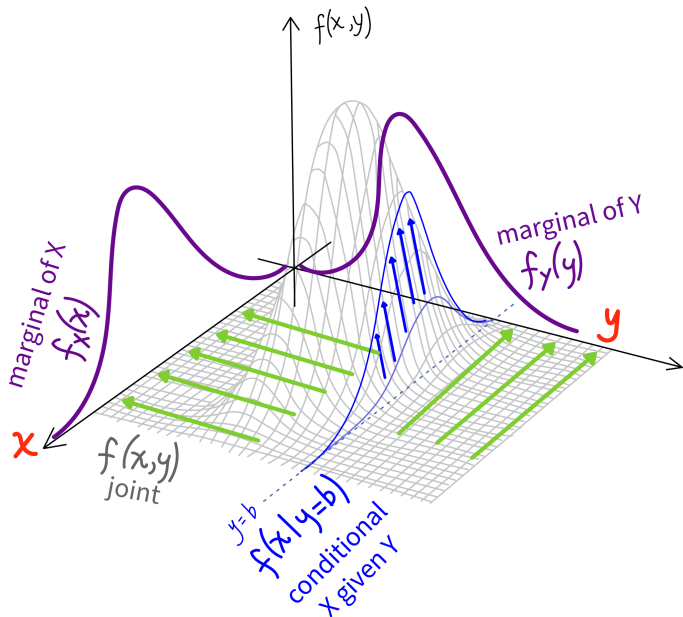
Solution:

$$\begin{aligned}\text{Corr}(X, 2X + Y) &= \frac{\text{Cov}(X, 2X + Y)}{\sqrt{V(X)V(2X + Y)}} && \text{def'n of correlation} \\ &= \frac{-2}{\sqrt{(2)(9)}} && \text{answers from part c) and d)} \\ &= \frac{-2}{\sqrt{18}} = -0.47.\end{aligned}$$



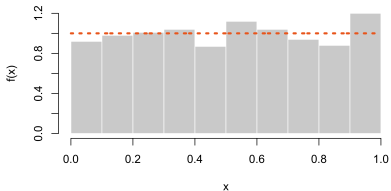
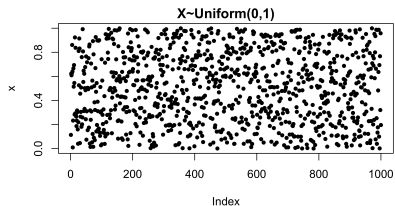
Previously...

Univariate & Multivariate

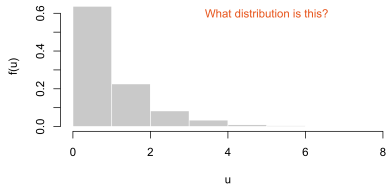
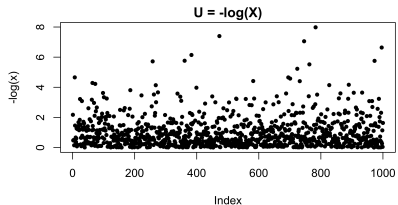


Next Few Topics: $X \rightarrow h(X)$

Original RV



Transformed RV



Functions of Random Variables (Univariate)

Functions of Random Variables (Univariate)

- ▶ Suppose we have a random variable X with PDF $f_X(x)$.
- ▶ Define a new random variable $U = h(X)$, where h is a (one-to-one) monotone function.
 - ▶ $h(x) = e^x$
 - ▶ $h(x) = \ln(x)$
 - ▶ $h(x) = \sqrt{x}$
 - ▶ $h(x) = x^2$
- ▶ What is the PDF of U ?
 - ▶ CDF Method
 - ▶ Jacobian Method (PDF-to-PDF Method or change of variable)
 - ▶ MGF Method

The CDF Method

The CDF Method

Theorem: The CDF Method

Let X be a random variable with CDF $F_X(x)$. Define $U = h(X)$ where h is a monotone function. Let the domain of X and codomain of U be $\mathcal{X} = \{x : f_X(x) > 0\}$ and $\mathcal{U} = \{u : u = h(x) \text{ for some } x \in \mathcal{X}\}$, respectively.

- 1 If h is an increasing function on \mathcal{X} , the CDF of U is given by

$$F_U(u) = F_X(h^{-1}(u)), \quad \text{for all } u \in \mathcal{U}.$$

- 2 If h is a decreasing function on \mathcal{X} , the CDF of U is given by

$$F_U(u) = 1 - F_X(h^{-1}(u)), \quad \text{for all } u \in \mathcal{U}.$$

The density of U , $f_U(u)$ can be obtained by differentiation as follows:

$$f_U(u) = \frac{d}{du} F_U(u).$$

The CDF Method

Proof:

$$\begin{aligned}F_U(u) &= P(U \leq u) && \text{CDF def'n} \\ &= P\{h(X) \leq u\} && U = h(X).\end{aligned}$$

If h is an increasing function:

$$F_U(u) = P[h^{-1}\{h(X)\} \leq h^{-1}(u)] \quad \text{applying inverse transformation}$$

does not change the inequality since h^{-1} is also increasing.

This means that if $a < b$, then $h^{-1}(a) < h^{-1}(b)$.

$$= P\{X \leq h^{-1}(u)\} \quad \text{Def'n of inverse: } f \text{ \& } g \text{ are inverses iff } f(g(x)) = x.$$

$$= F_X\{h^{-1}(u)\}. \quad \text{CDF def'n} \quad \square$$

The CDF Method

Proof:

$$\begin{aligned}F_U(u) &= P(U \leq u) && \text{CDF def'n} \\ &= P\{h(X) \leq u\} && U = h(X).\end{aligned}$$

If h is a decreasing function:

$$F_U(u) = P[h^{-1}\{h(X)\} > h^{-1}(u)] \quad \text{applying inverse transformation}$$

NEED TO CHANGE the inequality since h^{-1} is also decreasing.

This means that if $a < b$, then $h^{-1}(a) > h^{-1}(b)$.

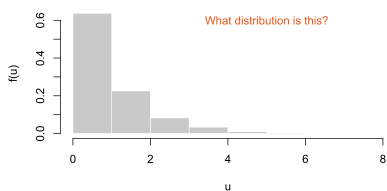
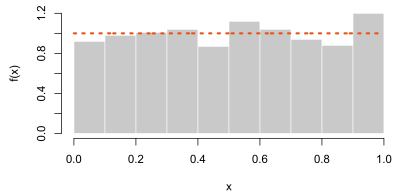
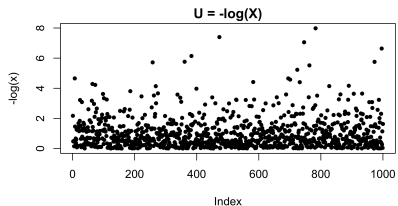
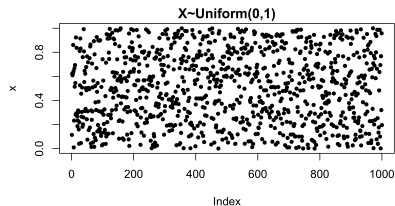
$$\begin{aligned}&= P\{X > h^{-1}(u)\} && \text{Def'n of inverse: } f \text{ \& } g \text{ are inverses iff } f(g(x)) = x. \\ &= 1 - P\{X \leq h^{-1}(u)\} && \text{complement} \\ &= 1 - F_X\{h^{-1}(u)\}. && \text{CDF def'n} \quad \square\end{aligned}$$

The CDF Method

Example 1:

Find the CDF of $U = h(X) = -\log X$, $X \sim \mathcal{U}(0, 1)$.

Visualizing the problem...



The CDF Method

Example 1:

Find the CDF of $U = h(X) = -\log X$, $X \sim \mathcal{U}(0, 1)$.

Solution:

- ▶ Domain of X : $x \in (0, 1)$
- ▶ Codomain of U : $u = -\log(x) \Rightarrow u \in (0, \infty)$.
- ▶ CDF of a uniform RV X over $[0, 1]$: $F_X(x) = \begin{cases} 0, & \text{if } x < 0, \\ x, & \text{if } 0 \leq x \leq 1, \\ 1, & \text{if } x > 1. \end{cases}$

$$\begin{aligned} F_U(u) &= P(U \leq u) && \text{CDF def'n} \\ &= P(-\log X \leq u) && \text{given} \\ &= P(\log X > -u) && \text{multiply } -1 \text{ to both sides} \\ &= P(X > e^{-u}) && \text{take exponential of both sides} \\ &= 1 - P(X \leq e^{-u}) && \text{complement} \\ &= 1 - F_X(e^{-u}) && \text{CDF def'n} \\ &= 1 - e^{-u}. && \text{CDF of } X, \text{ 2nd case since } 0 \leq e^{-u} \leq 1 \quad \square \end{aligned}$$

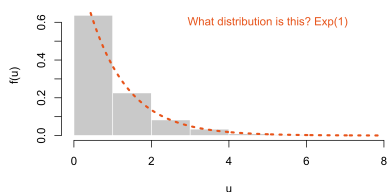
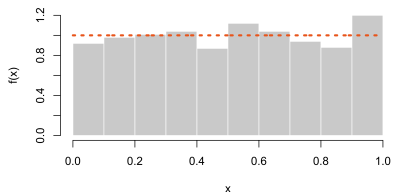
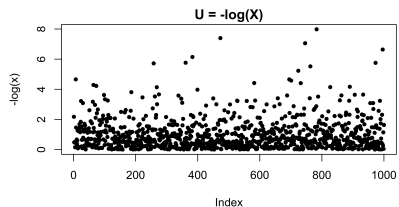
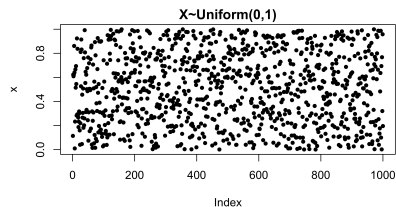
Note that this is the CDF of an exponential RV with $\beta = 1$.

The CDF Method

Example 1:

Find the CDF of $U = h(X) = -\log X$, $X \sim \mathcal{U}(0, 1)$.

Visualizing the problem...



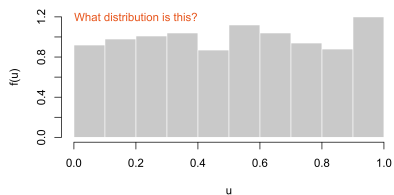
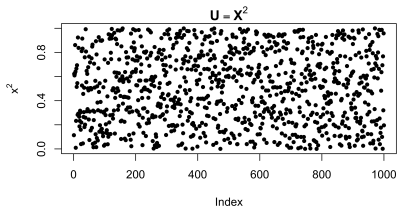
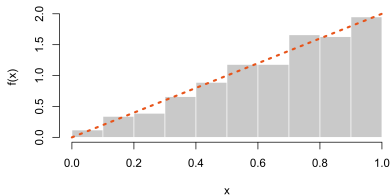
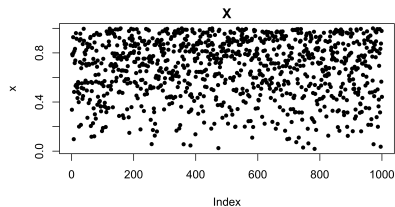
The CDF Method

Example 2:

Let X have the PDF $f_X(x) = 2x$, $0 \leq x \leq 1$.

Let $U = X^2$. Find the PDF of U .

Visualizing the problem...



The CDF Method

Example 2:

Let X have the PDF $f_X(x) = 2x$, $0 \leq x \leq 1$.

Let $U = X^2$. Find the PDF of U .

Solution:

- ▶ Domain of X : $x \in (0, 1)$
- ▶ Codomain of U : $u = x^2 \Rightarrow u \in (0, 1)$.

- ▶ CDF of X : $F_X(x) = \int_0^x 2t dt = \begin{cases} 0, & \text{if } x < 0, \\ x^2, & \text{if } 0 \leq x \leq 1, \\ 1, & \text{if } x > 1. \end{cases}$

$$\begin{aligned} F_U(u) &= P(U \leq u) && \text{CDF def'n} \\ &= P(X^2 \leq u) && \text{given} \\ &= P(X \leq \sqrt{u}) && \text{isolate } X \\ &= u. && \text{CDF of } X, \text{ 2nd case since } 0 \leq \sqrt{u} \leq 1 \end{aligned}$$

$$\begin{aligned} f_U(u) &= \frac{d}{du} F_U(u) \\ &= 1. \quad \square \end{aligned}$$

Note that this is the PDF of a $U(0, 1)$ RV.

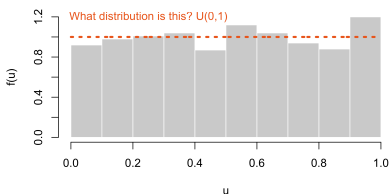
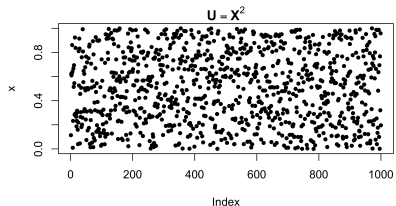
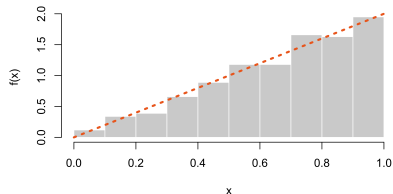
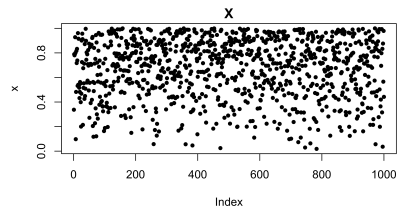
The CDF Method

Example 2:

Let X have the PDF $f_X(x) = 2x$, $0 \leq x \leq 1$.

Let $U = X^2$. Find the PDF of U .

Visualizing the problem...



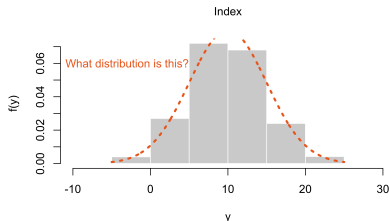
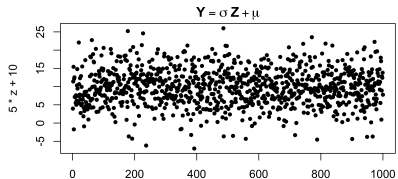
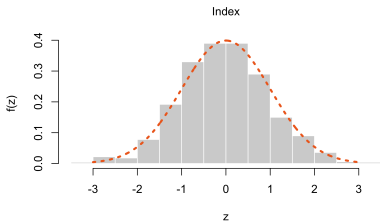
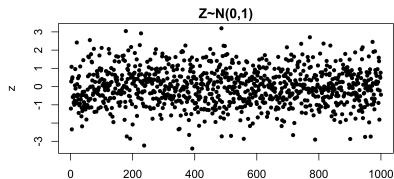
The CDF Method

Example 3:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = \sigma Z + \mu$. Find the PDF of Y .

Visualizing the problem...



The CDF Method

Example 3:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = \sigma Z + \mu$. Find the PDF of Y .

Solution:

- ▶ Domain of Z : $(-\infty, \infty)$
- ▶ Codomain of Y : $(-\infty, \infty)$

$$\begin{aligned} F_Y(y) &= P(Y \leq y) && \text{CDF def'n} \\ &= P(\sigma Z + \mu \leq y) && \text{transformation} \\ &= P\left(Z \leq \frac{y - \mu}{\sigma}\right) && \text{isolate the original RV} \\ &= \Phi\left(\frac{y - \mu}{\sigma}\right). && \text{CDF of the original RV: Standard Normal CDF } \Phi(z) \end{aligned}$$

$$\begin{aligned} f_Y(y) &= \frac{d}{dy} \{F_Y(y)\} = \frac{d}{dy} \left\{ \Phi\left(\frac{y - \mu}{\sigma}\right) \right\} \\ &= \frac{1}{\sigma} \left\{ \phi\left(\frac{y - \mu}{\sigma}\right) \right\} && \text{derivative of CDF } \Phi(z) \text{ is PDF } \phi(z); \text{ chain rule} \\ &= \frac{1}{\sigma} \frac{1}{\sqrt{2\pi}} e^{-\frac{\left(\frac{y - \mu}{\sigma}\right)^2}{2}} = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(y - \mu)^2}{2\sigma^2}}. \quad \square \end{aligned}$$

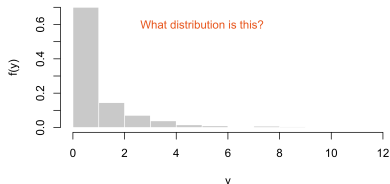
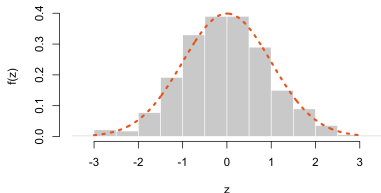
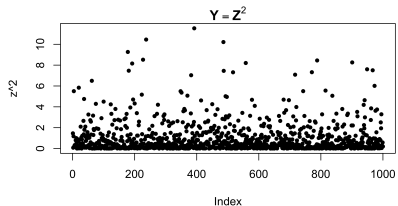
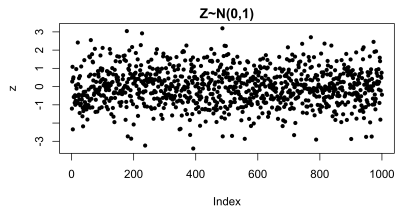
The CDF Method

Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Visualizing the problem...



The CDF Method

Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Domain of Z : $(-\infty, \infty)$
- ▶ Codomain of Y : $(0, \infty)$

$$\begin{aligned}F_Y(y) &= P(Y \leq y) && \text{CDF def'n} \\&= P(Z^2 \leq y) && \text{transformation} \\&= P(-\sqrt{y} \leq Z \leq \sqrt{y}) && \text{isolate the original RV} \\&= P(Z \leq \sqrt{y}) - P(Z \leq -\sqrt{y}) \\&= \Phi(\sqrt{y}) - \Phi(-\sqrt{y}). && \text{CDF of the original RV: Standard Normal CDF } \Phi(z)\end{aligned}$$

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The CDF Method

Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Domain of Z : $(-\infty, \infty)$
- ▶ Codomain of Y : $(0, \infty)$

$$\begin{aligned}f_Y(y) &= \frac{d}{dy} \{F_Y(y)\} = \frac{d}{dy} \{\Phi(\sqrt{y})\} - \frac{d}{dy} \{\Phi(-\sqrt{y})\} \\&= \frac{1}{2}y^{-1/2} \{\phi(\sqrt{y})\} - \left[-\frac{1}{2}y^{-1/2} \{\phi(-\sqrt{y})\} \right] \\&\quad \text{derivative of CDF } \Phi(z) \text{ is PDF } \phi(z); \text{ chain rule} \\&= \frac{1}{2\sqrt{y}} \frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}} + \frac{1}{2\sqrt{y}} \frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}} \\&= \frac{1}{\sqrt{2\pi y}} e^{-\frac{y}{2}}. \quad \square\end{aligned}$$

This is the $\chi^2(1)$ PDF.

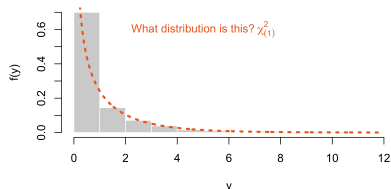
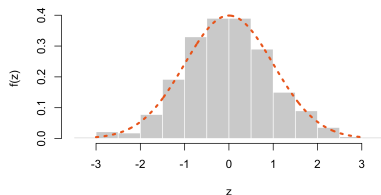
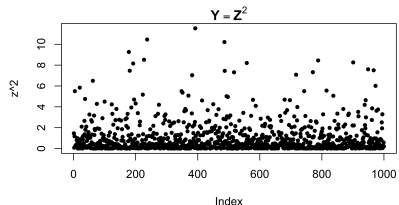
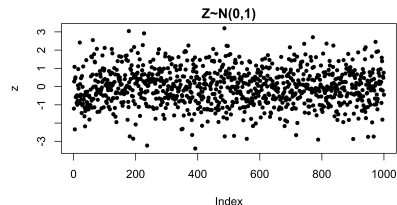
The CDF Method

Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Visualizing the problem...



The Jacobian Transformation Method

The Jacobian Transformation Method

Theorem: The PDF-to-PDF Method (monotone)

Let X be a random variable with PDF $f_X(x)$. Define $U = h(X)$ where h is a differentiable *monotone* function for all values where $f_X(x) > 0$, such that the equation $u = h(x)$ can be solved for x to yield $x = h^{-1}(u)$. Then, the PDF of U has the form:

$$f_U(u) = f_X\{h^{-1}(u)\} \left| \frac{dh^{-1}(u)}{du} \right|,$$

where $|\cdot|$ is the absolute value function.

- ▶ $\frac{dh^{-1}(u)}{du}$ is called the **Jacobian** of the transformation h and accounts for the stretching of the interval caused by the transformation.
- ▶ The theory behind this method is more difficult than the CDF method but it is often easier to compute in practice.
- ▶ **Advantage:** direct computation of the PDF of U without the middle step of finding the CDF of U .

The Jacobian Transformation Method

Theorem: The PDF-to-PDF Method (non-monotone)

Let X be a random variable with PDF $f_X(x)$. Define $U = h(X)$ where h is a differentiable *non-monotone* function. The PDF of U has the form:

$$f_U(u) = \sum_{k=1}^{n(y)} f_X\{h_k^{-1}(u)\} \left| \frac{dh_k^{-1}(u)}{du} \right|,$$

where $n(y)$ is the number of invertible intervals, $h_k^{-1}(u)$ is the inverse transformation on the interval k , and $|\cdot|$ is the absolute value function.

- ▶ non-monotone transformations can be dealt with by **splitting** the transformation into intervals which are **locally monotone**...

Monotonic vs. Non-Monotonic Transformations

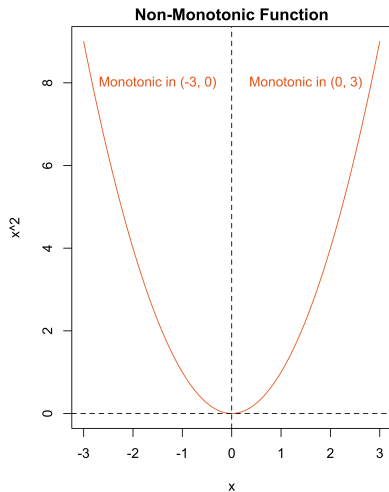
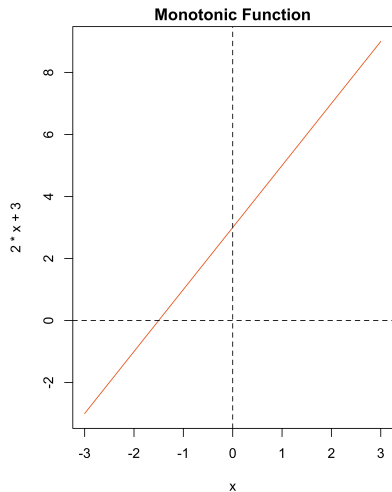
Definition: Monotonic Function

A *monotonic* function is a function whose first derivative does not change signs. Thus, it is always decreasing or always increasing, or always constant, but not more than one of these.

Definition: Non-Monotonic Function

A *non-monotonic* function is a function whose first derivative changes signs. Thus, it is increasing or decreasing for some time and shows opposite behavior at a different location.

Monotonic vs. Non-Monotonic Transformations



The Jacobian Transformation Method: How it Works

- ▶ We need the following to obtain the new PDF:
 - ▶ original PDF: $f_X(x)$
 - ▶ transformation function: $u = h(x)$
 - ▶ inverse of the transformation: $x = h^{-1}(u)$
 - ▶ Jacobian: $\frac{dh^{-1}(u)}{du}$

The Jacobian Transformation Method

Re-doing Example 2:

Let X have the PDF $f_X(x) = 2x$, $0 \leq x \leq 1$.

Let $U = X^2$. Find the PDF of U .

Solution:

- ▶ Domain of X : $x \in (0, 1)$ given
- ▶ Codomain of U : $u \in (0, 1)$
- ▶ Transformation: $u = h(x) = x^2$ given
- ▶ Deriving the inverse of the transformation, $h^{-1}(u)$:
(write x in terms of u)

$$u = x^2$$
$$\sqrt{u} = x \quad \text{only take the positive root since } x \in (0, 1)$$

Therefore, the inverse function is $h^{-1}(u) = \sqrt{u}$

- ▶ Jacobian: $\frac{dh^{-1}(u)}{du} = \frac{d}{du}(\sqrt{u}) = \frac{d}{du}(u^{1/2}) = \frac{1}{2}u^{1/2-1} = \frac{1}{2}u^{-1/2}$.

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The Jacobian Transformation Method

Re-doing Example 2:

Let X have the PDF $f_X(x) = 2x$, $0 \leq x \leq 1$.

Let $U = X^2$. Find the PDF of U .

Solution:

- ▶ Transformation: $h(x) = x^2$
- ▶ Inverse: $h^{-1}(u) = \sqrt{u}$
- ▶ Jacobian: $\frac{dh^{-1}(u)}{du} = \frac{1}{2}u^{-1/2}$
- ▶ Deriving the PDF of U using the formula in the theorem:

$$\begin{aligned}f_U(u) &= f_X\{h^{-1}(u)\} \left| \frac{dh^{-1}(u)}{du} \right| \\&= 2(\sqrt{u}) \left| \frac{1}{2}u^{-1/2} \right| \\&= 1. \quad \text{This is the } U(0, 1) \text{ PDF.}\end{aligned}$$

We arrived at the same PDF as the CDF method's.

The Jacobian Transformation Method

Re-doing Example 3:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = \sigma Z + \mu$. Find the PDF of Y .

Solution:

- ▶ Domain of Z : $(-\infty, \infty)$ given
- ▶ Codomain of Y : $(-\infty, \infty)$
- ▶ Transformation: $y = h(z) = \sigma z + \mu$ given
- ▶ Deriving the inverse of the transformation, $h^{-1}(y)$:
(write z in terms of y)

$$\begin{aligned}y &= \sigma z + \mu \\ \frac{y - \mu}{\sigma} &= z\end{aligned}$$

Therefore, the inverse function is $h^{-1}(y) = \frac{y - \mu}{\sigma}$

- ▶ Jacobian: $\frac{dh^{-1}(y)}{dy} = \frac{d}{dy} \left(\frac{y - \mu}{\sigma} \right) = \frac{1}{\sigma}$.

(cont'd next slide...)

The Jacobian Transformation Method

Re-doing Example 3:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = \sigma Z + \mu$. Find the PDF of Y .

Solution:

- ▶ Transformation: $h(z) = \sigma z + \mu$
- ▶ Inverse: $h^{-1}(y) = \frac{y-\mu}{\sigma}$
- ▶ Jacobian: $\frac{dh^{-1}(y)}{dy} = \frac{1}{\sigma}$
- ▶ Deriving the PDF of Y using the formula in the theorem:

$$\begin{aligned} f_Y(y) &= f_Z\{h^{-1}(y)\} \left| \frac{dh^{-1}(y)}{dy} \right| \\ &= \frac{1}{\sqrt{2\pi}} e^{-\frac{\left(\frac{y-\mu}{\sigma}\right)^2}{2}} \left| \frac{1}{\sigma} \right| \\ &= \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y-\mu)^2}{2\sigma^2}}. \quad \text{This is the Gaussian PDF.} \quad \square \end{aligned}$$

We arrived at the same PDF as the CDF method's.

The Jacobian Transformation Method

Re-doing Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Domain of Z : $(-\infty, \infty)$ given
- ▶ Codomain of Y : $(0, \infty)$
- ▶ Transformation: $y = h(z) = z^2$ is monotone on $(-\infty, 0)$ and $(0, \infty)$ given
- ▶ Deriving the inverse of the transformation, $h^{-1}(y)$: (write z in terms of y)

$$y = z^2$$
$$\pm\sqrt{y} = z$$

Therefore, the inverse function is $h_1^{-1}(y) = \sqrt{y}$ if $z \geq 0$ and $h_2^{-1}(y) = -\sqrt{y}$ if $z < 0$.

- ▶ Jacobian: $\frac{dh_1^{-1}(y)}{dy} = \frac{d}{dy}(\sqrt{y}) = \frac{1}{2}y^{-1/2}$ if $z \geq 0$ and
- $$\frac{dh_2^{-1}(y)}{dy} = \frac{d}{dy}(-\sqrt{y}) = -\frac{1}{2}y^{-1/2} \text{ if } z < 0.$$

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The Jacobian Transformation Method

Re-doing Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Transformation: $h(z) = z^2$
- ▶ Inverse: $h_1^{-1}(y) = \sqrt{y}$ if $z \geq 0$ and $h_2^{-1}(y) = -\sqrt{y}$ if $z < 0$
- ▶ Jacobian: $\frac{dh_1^{-1}(y)}{dy} = \frac{1}{2}y^{-1/2}$ if $z \geq 0$ and $\frac{dh_2^{-1}(y)}{dy} = -\frac{1}{2}y^{-1/2}$ if $z < 0$
- ▶ Deriving the PDF of Y using the formula in the theorem (non-monotone):

$$\begin{aligned}f_Y(y) &= f_Z\{h_1^{-1}(y)\} \left| \frac{dh_1^{-1}(y)}{dy} \right| + f_Z\{h_2^{-1}(y)\} \left| \frac{dh_2^{-1}(y)}{dy} \right| \\&= \frac{1}{\sqrt{2\pi}} e^{-\frac{(\sqrt{y})^2}{2}} \left| \frac{1}{2}y^{-1/2} \right| + \frac{1}{\sqrt{2\pi}} e^{-\frac{(-\sqrt{y})^2}{2}} \left| -\frac{1}{2}y^{-1/2} \right| \\&\quad \text{formula for non-monotone functions} \\&= \frac{1}{\sqrt{2\pi y}} e^{-\frac{y}{2}}. \quad \text{This is the } \chi^2(1) \text{ PDF. } \quad \square\end{aligned}$$

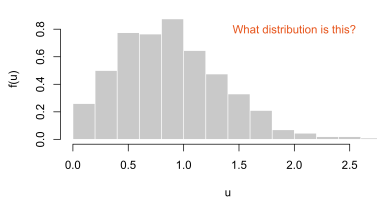
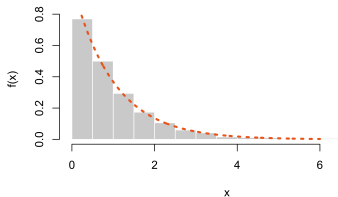
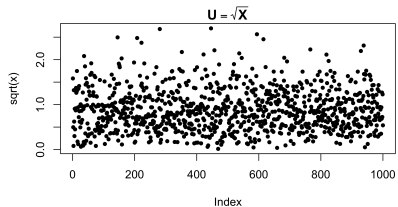
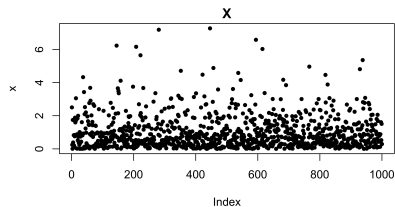
We arrived at the same PDF as the CDF method's.

The Jacobian Transformation Method

Example 5:

Suppose we have an exponential RV X with PDF $f_X(x) = e^{-x}$, $x \geq 0$. Let $U = \sqrt{X}$. Find the PDF of U .

Visualizing the problem...



The Jacobian Transformation Method

Example 5:

Suppose we have an exponential RV X with PDF $f_X(x) = e^{-x}$, $x \geq 0$.
Let $U = \sqrt{X}$. Find the PDF of U .

Solution:

- ▶ Domain of X : $[0, \infty]$ given
- ▶ Codomain of U : $[0, \infty]$
- ▶ Transformation function: $u = h(x) = \sqrt{x}$ given
- ▶ Deriving the inverse of the transformation, $h^{-1}(u)$:
(write x in terms of u)

$$\begin{aligned}u &= \sqrt{x} \\ u^2 &= x.\end{aligned}$$

Therefore, the inverse function is $h^{-1}(u) = u^2$

- ▶ Jacobian: $\frac{dh^{-1}(u)}{du} = \frac{d}{du}(u^2) = 2u$.

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The Jacobian Transformation Method

Example 5:

Suppose we have an exponential RV X with PDF $f_X(x) = e^{-x}$, $x \geq 0$. Let $U = \sqrt{X}$. Find the PDF of U .

Solution:

- ▶ Transformation function: $h(x) = \sqrt{x}$
- ▶ Inverse: $h^{-1}(u) = u^2$
- ▶ Jacobian: $\frac{dh^{-1}(u)}{du} = 2u$
- ▶ Deriving the PDF of U using the formula in the theorem:

$$\begin{aligned}f_U(u) &= f_X\{h^{-1}(u)\} \left| \frac{dh^{-1}(u)}{du} \right| \\&= e^{-u^2} |2u| \\&= 2ue^{-u^2}. \quad 2u \text{ is always positive since } u \in [0, \infty].\end{aligned}$$

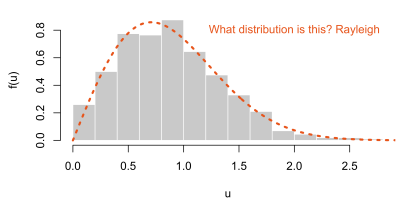
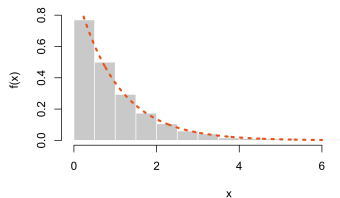
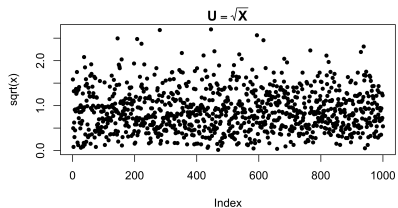
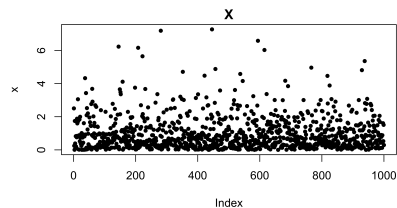
This is the Rayleigh distribution.

The Jacobian Transformation Method

Example 5:

Suppose we have an exponential RV X with PDF $f_X(x) = e^{-x}$, $x \geq 0$. Let $U = \sqrt{X}$. Find the PDF of U .

Visualizing the problem...



The MGF Method

Theorem: Uniqueness Theorem

Let $m_X(t)$ and $m_Y(t)$ denote the MGF of RVs X and Y , respectively. If both MGFs exist and $m_X(t) = m_Y(t)$ for all values of t , then X and Y have the same probability distribution.

The MGF Method: How It Works

To find the distribution of the transformation:

- 1 Derive the MGF of the transformed RV.
- 2 Compare the MGF of the transformed RV to the MGFs of known distributions.
- 3 The distribution of the transformed RV follows the distribution of the matching MGF.

The MGF Method

Re-doing Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Derive the MGF of the transformed RV.

$$\begin{aligned}m_Y(t) = m_{Z^2}(t) &= E(e^{tZ^2}) \quad \text{def'n of MGF} \\ &= \int_{-\infty}^{\infty} e^{tz^2} f(z) dz \quad \text{def'n of expected value} \\ &= \int_{-\infty}^{\infty} e^{tz^2} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \quad Z \text{ is standard normal} \\ &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-z^2\left(\frac{1-2t}{2}\right)} dz = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2\left(\frac{1-2t}{1-2t}\right)}} dz \\ &\quad \text{the integrand resembles a Gaussian PDF with } \mu = 0 \text{ and } \sigma^2 = \frac{1}{1-2t}\end{aligned}$$

(cont'd next slide...)

The MGF Method

Re-doing Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Derive the MGF of the transformed RV. (cont'd)

$$\begin{aligned} m_Y(t) &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2\left(\frac{1}{1-2t}\right)}} dz && \text{Recall the Gaussian PDF: } \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y-\mu)^2}{2\sigma^2}} \\ &= \sqrt{\frac{1}{1-2t}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{\frac{1}{1-2t}}} e^{-\frac{z^2}{2\left(\frac{1}{1-2t}\right)}} dz && \text{multiply a factor of 1} \\ &= \frac{1}{\sqrt{1-2t}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi \frac{1}{1-2t}}} e^{-\frac{z^2}{2\left(\frac{1}{1-2t}\right)}} dz \\ &= \frac{1}{\sqrt{1-2t}} (1) && \text{Gaussian PDF integrates to 1} \end{aligned}$$

(cont'd next slide...)

The MGF Method

Re-doing Example 4:

Let Z have the PDF $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$, $-\infty \leq z \leq \infty$.

Let $Y = Z^2$. Find the PDF of Y .

Solution:

- ▶ Derive the MGF of the transformed RV. (cont'd)

$$m_Y(t) = \frac{1}{(1-2t)^{1/2}} \quad \text{Recall Gamma MGF: } \frac{1}{(1-\beta t)^\alpha}$$

Note that this is the MGF of a Gamma RV with $\alpha = 1/2$ and $\beta = 2$.

Thus, Y is χ^2 with $\nu = 1$ degree of freedom.

Questions?

Homework Exercises: 6.15, 6.20, 6.23, 6.28, 6.46

Solutions will be discussed this Friday by the TA.