

Functions of One Random Variable

Example 1:

X	$p_{X}(x)$		$y = x^2$	$p_{\mathbf{Y}}(y) = p_{\mathbf{X}}(\sqrt{y})$			
1	0.2		1	0.2			
2	0.4	$Y = X^2$	4	0.4			
3	0.3		9	0.3			
4	0.1		16	0.1			
Neighto Kern track of changes in support							

Example 2:

<u> </u>	$p_{\rm X}(x)$		у	$p_{\mathrm{Y}}(y)$
-2	0.2		0	$p_{\rm X}(0) = 0.4$
0	0.4	$Y = X^2$	4	$p_{\rm X}(-2) + p_{\rm X}(2) = 0.5$
2	0.3		9	$p_{\rm X}(3) = 0.1$
3	0.1			•

Example 3:

$$X \sim \text{Poisson}(\lambda)$$
: $p_X(x) = \frac{\lambda^x \cdot e^{-\lambda}}{x!}, \quad x = 0, 1, 2, 3, 4, ...$

$$Y = X^2$$
 \Rightarrow $p_Y(y) = \frac{\lambda^{\sqrt{y}} \cdot e^{-\lambda}}{\left(\sqrt{y}\right)!}, \quad y = 0, 1, 4, 9, 16, \dots$

Let
$$Y = g(X)$$
.

What is the probability distribution of Y?

Cumulative Distribution Function approach:

$$\mathsf{F}_{\mathsf{Y}}(y) = \mathsf{P}(\mathsf{Y} \leq y) = \mathsf{P}(\mathsf{g}(\mathsf{X}) \leq y) = \int\limits_{\left\{x:\, g(x) \leq y\right\}} f_{\mathsf{X}}(x) dx = \dots$$

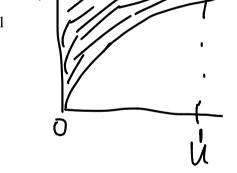
Moment-Generating Function approach:

$$M_{Y}(t) = E(e^{Y \cdot t}) = E(e^{g(X) \cdot t}) = \int_{-\infty}^{\infty} e^{g(x) \cdot t} f_{X}(x) dx = \dots$$
3. Change of Variables Approach

1. Let U be a Uniform (0, 1) random variable:

$$f_{\mathbf{U}}(u) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$
 $F_{\mathbf{U}}(u) = \begin{cases} 0 & u < 0 \\ u & 0 \le u < 1 \\ 1 & u \ge 1 \end{cases}$

Consider $Y = U^2$. What is the probability distribution of Y?



$$F_{\mathbf{V}}(y) = P(Y \le y) = P(U^2 \le y)$$

$$F_{Y}(y) = 0$$

$$0 \le y < 1$$

$$y \ge 1$$

$$P(U^{2} \le y) = P(U \le \sqrt{y}) = \sqrt{y}$$

$$F_{Y}(y) = 0.$$

$$F_{Y}(y) = \sqrt{y}.$$

$$F_{Y}(y) = \sqrt{y}.$$

$$F_{Y}(y) = 1.$$

$$f_{\mathbf{Y}}(y) = \mathbf{F}_{\mathbf{Y}}'(y) = \begin{cases} 1/(2\sqrt{y}) & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

The pating

North (Nange of Variables)

$$P(X \leq X) = F_X(X)$$
 $F_X(X) = f_X(X)$
 $F_X(X)$

fg(y)= fx(6'y))d(6'(y)) for 2
This holds for both cases

Theorem

$$X$$
 – continuous r.v. with p.d.f. $f_X(x)$.

$$Y = g(X)$$

g(x) – one-to-one, differentiable

$$\frac{dx}{dy} = \frac{d[g^{-1}(y)]}{dy}$$

$$\Rightarrow f_{\mathbf{Y}}(y) = f_{\mathbf{X}}(\mathbf{g}^{-1}(y)) \left| \frac{dx}{dy} \right|$$

 $g(u) = u^{2} g^{-1}(y) = \sqrt{y} = y^{1/2} du/_{dy} = \frac{1}{2} y^{-1/2}$

$$g(u) = u^2$$

$$g^{-1}(y) = \sqrt{y} = y^{1/2}$$

$$\frac{du}{dy} = \frac{1}{2} y^{-1/2}$$

$$f_{\rm Y}(y) = f_{\rm U}(g^{-1}(y)) \left| \frac{du}{dy} \right| = (1) \left| \frac{1}{2} y^{-1/2} \right| = \frac{1}{2} y^{-1/2}$$

2. Consider a continuous random variable X with p.d.f.

$$f_{\mathbf{X}}(x) = \begin{cases} 2x & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

Find the probability distribution of $Y = \sqrt{X}$. a)

$$f_{\mathbf{X}}(x) = \begin{cases} 2x & 0 < x < \\ 0 & \text{o.w.} \end{cases}$$

$$f_{X}(x) = \begin{cases} 2x & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases} \qquad F_{X}(x) = \begin{cases} 0 & x < 0 \\ x^{2} & 0 \le x < 1 \\ 1 & x \ge 1 \end{cases}$$

$$F_{Y}(y) = P(Y \le y) = P(\sqrt{X} \le y) = 0.$$

$$y \ge 0 \qquad F_{Y}(y) = P(Y \le y) = P(\sqrt{X} \le y) = P(X \le y^{2}) = F_{X}(y^{2}).$$

$$0 \le y < 1 \qquad F_{Y}(y) = F_{X}(y^{2}) = y^{4}.$$

$$F_{Y}(y) = F_{X}(y^{2}) = 1.$$

$$0 \le y < 1$$

$$F_{Y}(y) = F_{X}(y^{2}) = y^{4}$$
.

$$F_{\mathbf{Y}}(y) = F_{\mathbf{X}}(y^2) = 1$$

$$F_{Y}(y) = \begin{cases} 0 & y < 0 \\ y^{4} & 0 \le y < 1 \\ 1 & y \ge 1 \end{cases}$$

$$F_{Y}(y) = \begin{cases} 0 & y < 0 \\ y^{4} & 0 \le y < 1 \\ 1 & y \ge 1 \end{cases} \qquad f_{Y}(y) = F'_{Y}(y) = \begin{cases} 4y^{3} & 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

 $g(x) = \sqrt{x} \qquad g^{-1}(y) = y^{2}$ $f_{Y}(y) = f_{X}(g^{-1}(y)) \left| \frac{dx}{dy} \right| = (2y^{2})(2y) = 4y^{3},$

$$g(x) = \sqrt{x}$$

$$g^{-1}(y) = y^2$$

$$\frac{dx}{dy} = 2y$$

$$f_{Y}(y) = f_{X}(g^{-1}(y)) \left| \frac{dx}{dy} \right| = (2y^{2})(2y) = 4y^{3},$$

$$0 < y < 1$$
.

Find the probability distribution of $W = \frac{1}{V+1}$.

$$\text{Note: } 0 < x < 1 \qquad \Rightarrow \qquad \frac{1}{2} < w < 1$$

$$F_W(w) = P(W \le w) = P(\frac{1}{X+1} \le w) = P(X \ge \frac{1}{w} - 1) = 1 - F_X(\frac{1}{w} - 1)$$

$$= \left(\frac{1 - \sqrt{1 - 1}}{\sqrt{1 - 1}} \right)^{2} = \frac{2}{w} - \frac{1}{w^{2}}, \qquad \frac{1}{2} < w < 1.$$

$$f_{W}(w) = F'_{W}(w) = -\frac{2}{w^{2}} + \frac{2}{w^{3}} = \frac{2 - 2w}{w^{3}}, \qquad \frac{1}{2} < w < 1.$$

$$\frac{1}{2} < w < 1.$$

$$= \frac{1}{w^3}, \qquad \frac{1}{2} < w < 1.$$
OR
$$\frac{2}{w^3} \left(\frac{1 - w}{w^3} \right)$$

$$g(x) = \frac{1}{x+1}$$

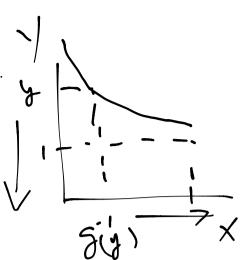
$$g^{-1}(w) = \frac{1}{w} - 1$$

$$g(x) = \frac{1}{x+1}$$
 $g^{-1}(w) = \frac{1}{w} - 1$ $\frac{dx}{dw} = -\frac{1}{w^2}$

$$f_{Y}(y) = f_{X}(g^{-1}(y)) \left| \frac{dx}{dy} \right| = \left[2\left(\frac{1}{w} - 1 \right) \right] \left(\frac{1}{w^{2}} \right) = \frac{2 - 2w}{w^{3}}, \qquad \frac{1}{2} < w < 1.$$

$$f_{\mathbf{X}}(x) = \begin{cases} 6x^5 & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$$

Find the probability distribution of $Y = \frac{1}{X^2}$.



Support of
$$X = \{0 < x < 1\}$$

$$Y = \frac{1}{X^2}$$
 \Rightarrow Support of $Y = \{y > 1\}$

$$\mathcal{Z}$$

$$g(x) = \frac{1}{x^2}$$
 $g^{-1}(y) = \frac{1}{\sqrt{y}} = y^{-1/2}$

$$\frac{dx}{dy} = -\frac{1}{2} y^{-3/2}$$

$$f_{Y}(y) = f_{X}(g^{-1}(y)) \left| \frac{dx}{dy} \right| = (6y^{-5/2})(\frac{1}{2}y^{-3/2}) = 3y^{-4}$$
 $y > 1$.

OR

$$f_{X}(x) = \begin{cases} 6x^{5} & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases} \qquad F_{X}(x) = \begin{cases} 0 & x < 0 \\ x^{6} & 0 \le x < 1 \\ 1 & x > 1 \end{cases}$$

$$F_X(x) = \begin{cases} 0 & x < 0 \\ x^6 & 0 \le x < 1 \\ 1 & x \ge 1 \end{cases}$$

$$F_Y(y) = P(Y \le y) = P(\frac{1}{X^2} \le y) = P(X \ge \frac{1}{\sqrt{y}}) = 1 - F_X(\frac{1}{\sqrt{y}})$$

= 1 - y⁻³, y > 1.

$$f_{Y}(y) = F'_{Y}(y) = 3y^{-4}, y > 1.$$

2 4. Y Consider a continuous random variable X with the p.d.f. $f_X(x) = \frac{24}{x^4}$, x > 2.

a) Let
$$Y = \frac{1}{X}$$
. Find the p.d.f. of Y, $f_Y(y)$.

$$F_{x} = \frac{1}{3} \left[\frac{1}{x^{2}} + \frac{1}{3} + \frac{1}{3} \right]_{x}^{2}$$

$$= -8 \left(\frac{1}{x^{2}} - \frac{1}{8} \right)$$

Support of
$$X = \{x > 2\}$$

 $Y = \frac{1}{X}$ \Rightarrow Support of $Y = \{0 < y < \frac{1}{2}\}$

$$g(x) = \frac{1}{x}$$

$$g^{-1}(y) = \frac{1}{v}$$

$$\frac{dx}{dy} = -\frac{1}{y^2}$$

$$g(x) = \frac{1}{x} \qquad g^{-1}(y) = \frac{1}{y} \qquad \frac{dx}{dy} = -\frac{1}{y^2}$$

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{dx}{dy} \right| = (24y^4)(y^{-2}) = 24y^2, \quad 0 < y < \frac{1}{2}.$$

OR



 $F_X(x) = 1 - \frac{8}{x^3}, \quad x > 2.$

$$F_Y(y) = P(Y \le y) = P(\frac{1}{X} \le y) = P(X \ge \frac{1}{y}) = 1 - F_X(\frac{1}{y}) = 8y^3,$$

$$0 < y < \frac{1}{2}.$$

$$f_{Y}(y) = 24y^{2}, \ 0 < y < \frac{1}{2}.$$

Let $Y = \frac{1}{Y^2}$. Find the p.d.f. of Y, $f_Y(y)$. b)

Support of
$$X = \{x > 2\}$$

$$Y = \frac{1}{X^2}$$
 \Rightarrow Support of $Y = \{0 < y < \frac{1}{4}\}$

$$g(x) = \frac{1}{x^2}$$
 $g^{-1}(y) = \frac{1}{\sqrt{y}} = y^{-1/2}$ $\frac{dx}{dy} = -\frac{1}{2}y^{-3/2}$

$$f_{\rm Y}(y) = f_{\rm X}({\rm g}^{-1}(y)) \left| \frac{dx}{dy} \right| = (24y^2) (\frac{1}{2} y^{-3/2}) = 12 y^{1/2} = 12 \sqrt{y},$$

$$0 < y < \frac{1}{4}.$$

OR

$$F_X(x) = 1 - \frac{8}{x^3}, \quad x > 2.$$

$$F_{Y}(y) = P(Y \le y) = P(\frac{1}{X^{2}} \le y) = P(X \ge \frac{1}{\sqrt{y}})$$

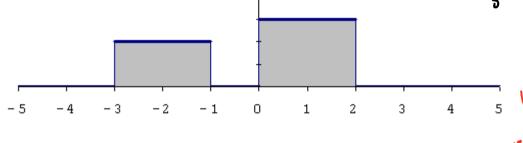
$$= 1 - F_{X}(\frac{1}{\sqrt{y}}) = 8y^{3/2}, \qquad 0 < y < \frac{1}{4}.$$

$$f_{Y}(y) = 12y^{1/2} = 12\sqrt{y}, \ 0 < y < 1/4.$$

continuous random variable X with p.d.f.
$$f_X(x) = \begin{cases} 0.2 & -3 < x < -1 \\ 0.3 & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

$$= \begin{cases} 0.2 & -3 < x < -1 \\ 0.3 & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

$$= \begin{cases} 2 & \text{if } x = \frac{1}{5} (x + 3) \\ -3 & \text{if } x = \frac{1}{5} (x + 3) \\ -3 & \text{if } x = \frac{1}{5} (x + 3) \end{cases}$$



y < 0

Find the probability distribution of
$$Y = X^2$$
.
 $y < 0$ $P(X^2 \le y) = 0$ $F_Y(y) = 0$.

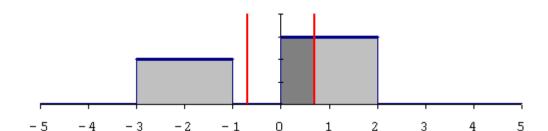
$$y \ge 0 \qquad \qquad F_Y(y) = P(Y \le y) = P(X^2 \le y) = P(-\sqrt{y} \le X \le \sqrt{y})$$

$$F_{y} = P(-f_{y} \le x \le f_{y}) = F_{x}(f_{y}) - F_{x}(-f_{y})$$

$$= f_{x}(f_{y}) - f_{x}(f_{y})$$

$$= f_{x}$$

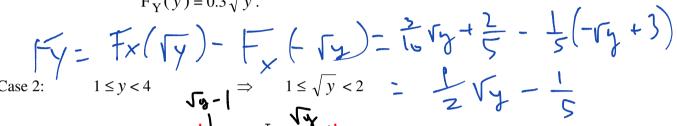
Case 1:

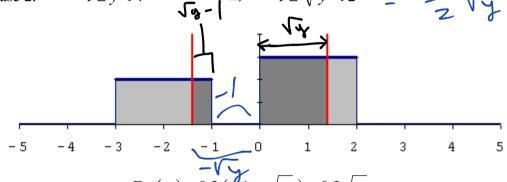


$$F_{Y}(y) = 0.3\sqrt{y}.$$

 $y \ge 9$

Case 4:





$$F_Y(y) = 0.2(-1 + \sqrt{y}) + 0.3\sqrt{y}$$
.

Case 3:
$$4 \le y < 9$$
 $\Rightarrow 2 \le \sqrt{y} < 3$ $-\frac{1}{x}(y) - \frac{1}{x}(y)$ $-\frac{1}{x}(y)$ $-\frac{1}{x}(y)$

6. Let $\lambda > 0$ and let X be a random variable with the probability density function

$$f(x) = \frac{\lambda}{x^{\lambda+1}},$$
 zero otherwise.

Let $W = \ln X$. What is the probability distribution of W?

a) Determine the probability distribution of W by finding the c.d.f. of W

$$F_{W}(w) = P(W \le w) = P(\ln X \le w).$$

"Hint": Find $F_X(x)$ first.

$$F_X(x) = \int_{1}^{x} \frac{\lambda}{y^{\lambda+1}} dy = 1 - \frac{1}{x^{\lambda}}, \quad x > 1.$$

$$F_{Y}(y) = P(W \le w) = P(X \le e^{w}) = F_{X}(e^{w}) = 1 - e^{-\lambda w}, w > 0.$$

$$f_{\mathbf{W}}(w) = \lambda e^{-\lambda w}, \quad w > 0.$$

- \Rightarrow W has Exponential distribution with mean $1/\lambda$.
- b) Determine the probability distribution of W by finding the m.g.f. of W

$$M_W(t) = E(e^{W \cdot t}) = E(e^{\ln X \cdot t}).$$

$$\mathbf{M}_{\mathbf{W}}(t) = \mathbf{E}(e^{\mathbf{W} \cdot t}) = \mathbf{E}(e^{\ln \mathbf{X} \cdot t}) = \mathbf{E}(\mathbf{X}^t) = \int_{1}^{\infty} \left(x^t \cdot \frac{\lambda}{x^{\lambda + 1}}\right) dx$$

$$= \int_{1}^{\infty} \lambda x^{t-\lambda-1} dx = \frac{\lambda}{\lambda-t} = \frac{1}{1-\frac{1}{\lambda}t}, \qquad t < \lambda.$$

 \Rightarrow W has Exponential distribution with mean $1/\lambda$.

Determine the probability distribution of W by finding the p.d.f. of W, $f_{W}(w)$, c) using the change-of-variable technique.

$$w = \ln(x)$$
 $x = g^{-1}(w) = e^{w}$
$$\frac{dx}{dw} = e^{w}$$

$$x > 1 \implies w > 0$$

$$f_{\mathbf{W}}(w) = f_{\mathbf{X}}(g^{-1}(w)) \left| \frac{dx}{dw} \right| = \frac{\lambda}{\left(e^{w}\right)^{\lambda+1}} \cdot \left| e^{w} \right| = \lambda e^{-\lambda w}, \quad w > 0.$$

W has Exponential distribution with mean $1/\lambda$.

Consider a continuous random variable X, with p.d.f. f and c.d.f. F, where F is strictly increasing on some interval I, F = 0 to the left of I, and F = 1 to the right of I. I may be a bounded interval or an unbounded interval such as the whole real line. $F^{-1}(u)$ is then well defined for 0 < u < 1.

Fact 1:

Let $U \sim \text{Uniform}(0, 1)$, and let $X = F^{-1}(U)$. Then the c.d.f. of X is F.

Proof:
$$P(X \le x) = P(F^{-1}(U) \le x) = P(U \le F(x)) = F(x).$$

is p(U su) = u

Fact 2:

Let U = F(X); then U has a Uniform (0, 1) distribution.

Let
$$U = F(X)$$
; then U has a Uniform $(0, 1)$ distribution.

Proof:
$$P(U \le u) = P(F(X) \le u) = P(X \le F^{-1}(u)) = F(F^{-1}(u)) = u.$$

Let X have a logistic distribution with p.d.f.
$$f(x) = \frac{e^{-x}}{(1+e^{-x})^2}, \quad -\infty < x < \infty.$$

$$= \int_{-\infty}^{\infty} \frac{-x}{(1+e^{-x})^2} dx dx = -\frac{1}{2} \int_{-\infty}^{\infty} dx = -\frac{1}{2}$$

 $=\frac{1}{1+\bar{e}^{\star}}\Big|_{x=1+\bar{e}^{\star}}$

Show that $Y = \frac{1}{1 + \rho^{-X}}$ has a U(0, 1) distribution.

$$F_X(x) = \frac{1}{1 + e^{-x}}, \quad -\infty < x < \infty.$$

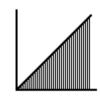
 $Y = F_X(X)$ has a Uniform (0, 1) distribution by Fact 2.

 $\sim 1.9.20$ (7th edition) **1.9.19** (6th edition)

> Let X be a nonnegative continuous random variable with p.d.f. f(x) and c.d.f. F(x). Show that

$$E(X) = \int_{0}^{\infty} (1 - F(x)) dx.$$

$$E(X) = \int_{0}^{\infty} x f(x) dx = \int_{0}^{\infty} \left(\int_{0}^{x} dy \right) f(x) dx = \int_{0}^{\infty} \left(\int_{0}^{x} f(x) dy \right) dx$$



$$\int_{0}^{\infty} \left(\int_{0}^{x} f(x) dy \right) dx$$

$$\int_{0}^{\infty} \left(\int_{0}^{x} f(x) dy \right) dx = \int_{0}^{\infty} \left(\int_{y}^{\infty} f(x) dx \right) dy$$

$$\Rightarrow E(X) = \int_{0}^{\infty} \left(\int_{y}^{\infty} f(x) dx \right) dy = \int_{0}^{\infty} P(X > y) dy = \int_{0}^{\infty} (1 - F(y)) dy.$$

Example: Find the expected value of an Exponential (θ) distribution.

For Exponential (
$$\theta$$
), $1 - F(x) = P(X > x) = e^{-x/\theta}$, $t > 0$

$$E(X) = \int_{0}^{\infty} (1 - F(x)) dx = \int_{0}^{\infty} e^{-x/\theta} dx = \theta.$$

Let X be a random variable of the discrete type with pmf p(x) that is positive on the nonnegative integers and is equal to zero elsewhere. Show that

$$E(X) = \sum_{x=0}^{\infty} [1 - F(x)],$$

where F(x) is the cdf of X.

$$1 - F(x) = P(X > x) = p(x+1) + p(x+2) + p(x+3) + p(x+4) + \dots$$

$$1-F(0)$$
 $p(1)+p(2)+p(3)+p(4)+p(5)+p(6)+p(7)+...$

$$1-F(1)$$
 $p(2) + p(3) + p(4) + p(5) + p(6) + p(7) + ...$

$$1-F(2)$$
 $p(3) + p(4) + p(5) + p(6) + p(7) + ...$

$$1-F(3)$$
 $p(4) + p(5) + p(6) + p(7) + ...$

$$1-F(4)$$
 $p(5) + p(6) + p(7) + ...$

... ..

$$\Rightarrow \sum_{x=0}^{\infty} [1 - F(x)] = 1 \times p(1) + 2 \times p(2) + 3 \times p(3) + 4 \times p(4) + \dots = E(X).$$

Example: Find the expected value of a Geometric (p) distribution.

For Geometric
$$(p)$$
, $1 - F(x) = P(X > x) = (1 - p)^x$, $x = 0, 1, 2, ...$

$$E(X) = \sum_{x=0}^{\infty} [1 - F(x)] = \sum_{x=0}^{\infty} [1 - p]^x = \frac{1}{1 - [1 - p]} = \frac{1}{p}.$$

$$\begin{array}{l}
x_{1} exp(\lambda), x_{2} \wedge Qxp(\lambda), x_{3}(x_{2}) \\
y = \frac{x_{1}}{x_{2}} \quad F_{1}(x_{1}) = |-e^{\lambda_{1}x_{1}}| \\
F_{2}(x_{2}) = \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}}} = \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}}} = \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) = P(x_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} = \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{1} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} \\
= \int_{1}^{1} (y_{2} + y_{2}) + \frac{\lambda_{2}e^{\lambda_{2}x_{2}}}{e^{\lambda_{2}x_{2}}} + \frac{\lambda_{2}$$

$$= \frac{1}{\sqrt{2}} \left(\frac{1}{\sqrt{2}} \right) = \frac{$$