

# STAT 702/J702

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-Lecture 14-

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## Today

- Homework
- Joint Distributions (continued)



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### Chapter 2 #67: The Weibull has CDF

$$F(x) = 1 - e^{-(x/\alpha)^\beta}$$

for  $x \geq 0$ ,  $\alpha > 0$ , and  $\beta > 0$

a) Find the density function.



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b) Show if  $W$  follows a Weibull, then  $X=(W/\alpha)^\beta$  follows an exponential.



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c) How could Weibull random variables be generated from a uniform random generator?



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Also) Use R to plot the pdf for a few values of alpha and beta to demonstrate how they affect the behavior of the Weibull distribution.



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### Chapter 3 – Joint Distributions

The joint behavior of two random variables  $X$  and  $Y$  is determined by their CDF:

$$F_{XY}(x, y) = P(X \leq x, Y \leq y)$$



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### 3.2 - Discrete R.V.'s

For discrete R.V.'s the joint p.m.f. is

$$p(x, y) = P(X=x, Y=y)$$



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Example) A fair coin is tossed three times. Let  $X$ =number of heads in three tossings and  $Y$ = difference (in absolute values) between the number of heads and number of tails.



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The Marginal p.m.f of X is

$$p_X(x) = \sum_y p(x, y)$$

The Conditional p.m.f. of X is

$$\begin{aligned} p_{X|Y}(x|y) &= P(X=x|Y=y) \\ &= P(X=x, Y=y) / P(Y=y) \\ &= p_{XY}(x, y) / p_Y(y) \end{aligned}$$



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X and Y are independent if

$$F_{XY}(x, y) = F_X(x) F_Y(y)$$

This implies that

$$p_{XY}(x, y) = p_X(x) p_Y(y)$$

It also works for functions g(x) and h(y).



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### 3.3 Continuous R.V.'s

Continuous (X, Y) have joint cdf

$$F_{XY}(x, y) = P(X \leq x, Y \leq y)$$

The joint pdf is

$$f_{XY}(x, y) = \frac{\partial^2}{\partial x \partial y} F_{XY}(x, y)$$



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So

$$F(x, y) = \int_{-\infty}^x \int_{-\infty}^y f(u, v) du dv$$

and

$$P((x, y) \in A) = \iint_A f(x, y) dx dy$$



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The marginal p.d.f.'s are analogous to the marginal p.m.f.'s for discrete variables, but are defined using integrals:

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy, \quad f_Y(y) = \int_{-\infty}^{\infty} f(x, y) dx.$$



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The conditional p.d.f.'s are also analogous:

$$f_{Y|X}(y|x) = \frac{f_{X,Y}(x, y)}{f_X(x)}$$



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Example 1)  $f(x, y) = 2, 0 < x < y < 1$   
and 0 elsewhere




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Example 2)

$$f(x, y) = \lambda^2 \exp(-y\lambda)$$

$$0 \leq x \leq y \leq 1, \lambda > 0$$




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Example 3)

$$f(x, y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp \left\{ -\frac{1}{2(1-\rho^2)} \left[ \frac{(x-\mu_x)^2}{\sigma_x^2} \right. \right.$$

$$\left. \left. - \frac{2\rho(x-\mu_x)(y-\mu_y)}{\sigma_x\sigma_y} + \frac{(y-\mu_y)^2}{\sigma_y^2} \right] \right\},$$

$$-\infty < \mu_x, \mu_y < \infty, \sigma_x, \sigma_y > 0, -1 < \rho < 1$$




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