

## Math Stat 465

### Some notes on expectation with joint distributions

Suppose that  $X$  and  $Y$  are **discrete** random variables with joint pmf  $p(x, y)$ . Then for any function  $h$  of  $X$  and  $Y$ ,

$$E(h(X, Y)) = \sum_x \sum_y h(x, y)p(x, y). \quad (1)$$

In particular,  $E(X) = \sum_x \sum_y xp(x, y)$  and similarly for  $E(Y)$ . However, it is often easier to calculate the marginal pmfs  $p_X$  and  $p_Y$  first and then to use the original rules for expected value:

$$\text{E1. } E(X) = \sum_x xp_X(x)$$

$$\text{E2. } E(Y) = \sum_y yp_Y(y).$$

For **continuous** random variables, equation (1) and rules E1 and E2 hold with the following changes: pmfs are changed to pdfs, and sums are changed to integrals.

For **both the discrete and continuous cases**, the following rules apply:

$$\text{E3. } E(g(X, Y) + h(X, Y)) = E(g(X, Y)) + E(h(X, Y)).$$

$$\text{E4. } E(cg(X, Y)) = cE(g(X, Y)).$$

$$\text{E5. } \text{If } X \text{ and } Y \text{ are independent, then } E(XY) = E(X)E(Y).$$