#### XXV. Marginal Probability

#### Discrete Case

- We have an experiment with two random variables,  $Y_1$  and  $Y_2$ .
- The marginal probability functions of  $Y_1$ and  $Y_2$  are

$$p_{1}(y_{1}) := P(Y_{1} = y_{1})$$

$$= \left[\sum_{y_{2}} p(y_{1}, y_{2})\right]$$

$$p_{2}(y_{2}) := P(Y_{2} = y_{2})$$

$$= \left[\sum_{y_{1}} p(y_{1}, y_{2})\right].$$

(For each  $y_1$ , hold it constant and add up over all the  $y_2$ 's. You get an answer that depends on  $y_1$ .)

### Continuous Case

• The marginal density functions of  $Y_1$  and  $Y_2$ are

$$f_{1}(y_{1}) := \int_{y_{2}=-\infty}^{y_{2}=\infty} f(y_{1}, y_{2}) dy_{2}$$

$$f_{2}(y_{2}) := \int_{y_{1}=-\infty}^{y_{1}=\infty} f(y_{1}, y_{2}) dy_{1}$$

Note that it doesn't make sense to say  $P(Y_1 = y_1)$ , because that would be 0. the second line does translate to continuous

distributions.  $f_{2}(y_{2}) = \int_{y_{1}=-\infty}^{y_{1}=\infty} f(y_{1}, y_{2}) dy_{1} = \text{area of a slice.}$ Warning: Note the subscript change:  $f_{1}(y_{1}) = \int_{y_{1}=-\infty}^{y_{1}=-\infty} f(y_{1}, y_{2}) dy_{1} = \text{area of a slice.}$ 

 $\int_{y_2} f \, dy_2, \text{ etc.}$ 

Integrate over whatever your range for  $y_2$  is.

## Example I

Roll two dice, a red die and a blue die. Define the variables:

 $Y_1 := \text{ what shows on red die}$ 

 $Y_2 := total$ 

Compute the marginal probability function  $p_1(y_1)$ .

$$\begin{array}{rcl} Y_1 & := & \text{what shows on red die} \\ Y_2 & := & \text{total} \\ p_1\left(y_1\right) & = & ?? \\ p_1(1) & = & p(1,2) + p(1,3) + \dots + p(1,7) + p(1,8) + \dots + p(1,12) \\ & = & \frac{1}{36} + \dots + \frac{1}{36} + 0 + \dots + 0 \\ & = & \left[\frac{1}{6}\right] \\ p_1(2) & = & p(2,2) + p(2,3) + p(2,4) + \dots + p(2,8) + p(2,9) + \dots + p(2,12) \\ & = & 0 + \frac{1}{36} + \dots + \frac{1}{36} + 0 + \dots + 0 \\ & = & \left[\frac{1}{6}\right] \\ & \vdots \\ p_1(6) & = & \left[\frac{1}{6}\right] \end{array}$$

# Example II

As in Example I, roll two dice, a red die and a blue die. Define the variables:

 $Y_1 := \text{what shows on red die}$  $Y_2 := \text{total}$ 

Compute the marginal probability function  $p_2(y_2)$ .

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$$p_{2}(2) = p(1,2) + p(2,2) + \dots + p(6,2) = \frac{1}{36} + 0 + \dots + 0 = \boxed{\frac{1}{36}}$$

$$p_{2}(3) = p(1,3) + p(2,3) + \dots + p(6,3) = \boxed{\frac{2}{36}}$$

$$p_{2}(4) = \boxed{\frac{3}{36}}$$

$$\vdots$$

$$p_{2}(7) = \boxed{\frac{6}{36}}$$
[They'll biff this line:] 
$$p_{2}(8) = \boxed{\frac{5}{36}}$$

$$\vdots$$

$$p_{2}(12) = \boxed{\frac{1}{36}}$$

# Example III

Consider the joint density function

$$f(y_1, y_2) := 6(1 - y_2), 0 \le y_1 \le y_2 \le 1.$$

Find the marginal density function  $f_1(y_1)$ .

(Graph.)

$$f_1(y_1) = \int_{y_2=y_1}^{y_2=1} 6(1-y_2) dy_2 = \boxed{3y_1^2 - 6y_1 + 3}$$

We'll use this for Example II in the next lecture.

### Example IV

As in Example III, consider the joint density function

$$f(y_1, y_2) := 6(1 - y_2), 0 \le y_1 \le y_2 \le 1.$$

Find the marginal density function  $f_2(y_2)$ .

(Graph.)

$$f_2(y_2) = \int_{y_1=0}^{y_1=y_2} 6(1-y_2) dy_1 = \boxed{6y_2 - 6y_2^2}$$

We'll use these again in the next lecture.

## Example V

Consider the joint density function

$$f(y_1, y_2) := \begin{cases} e^{-y_2}, & 0 \le y_1 \le y_2 < \infty, \\ 0, & \text{elsewhere.} \end{cases}$$

Find the marginal density function  $f_1(y_1)$ .

(Graph.)

$$f_{1}(y_{1}) = \int_{y_{2}=y_{1}}^{y_{2}=\infty} e^{-y_{2}} dy_{2}$$

$$= (-e^{-y_{2}})|_{y_{2}=y_{1}}^{y_{2}=\infty}$$

$$= e^{-y_{1}}$$

We'll use this for Example V in the next lecture.