

# Lecture Summary

3.4 Bivariate Distributions

3.5 Marginal Distributions

3.6 Conditional Distributions

4.7 Conditional Expectation

# Joint/Conditional/Marginal PDFs

Joint, Marginal and Conditional Densities

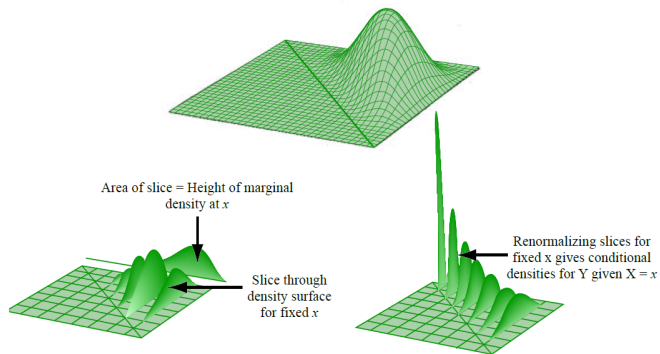


Image by MIT OpenCourseWare, adapted from *Probability*, by J. Pittman, 1999.

# Joint/Conditional/Marginal PDFs

## Example

Suppose that the joint pdf of  $X$  and  $Y$  is

$$f(x, y) = \begin{cases} \frac{21}{4}x^2y, & x^2 \leq y \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

Find the marginal pdfs  $f_X(x)$ ,  $f_Y(y)$ .

# Conditional Expectation

- ▶ Conditional distributions are distributions, so they have expectations and variances.
- ▶  $E(Y|X = x) = \sum_y yP(y|X = x)$  is the conditional expectation of  $Y$  if you know  $X = x$  (a number).
- ▶  $E(Y|X) = h(X)$  is a function of  $X$ . For every possible value  $x$  of  $X$ ,  $E(Y|X)$  takes the value  $E(Y|X = x)$ . So  $E(Y|X)$  is a random variable.
- ▶ Law of total expectation/Law of iterated expectations:

$$E[E(Y|X)] = E(Y)$$

## Conditional Variance

- ▶  $\text{Var}(Y|X = x) = E[(Y - E(Y|X = x))^2|X = x]$  is the variance of  $Y$  given  $X = x$  (a number).
- ▶  $\text{Var}(Y|X)$  is a function of  $Y$  that takes the value  $\text{Var}(Y|X = x)$  for every possible value  $x$  of  $X$ . So  $\text{Var}(Y|X)$  is a random variable.
- ▶ Law of total variance:

$$\text{Var}(Y) = E[\text{Var}(Y|X)] + \text{Var}(E(Y|X))$$

# Practice Exercises

3.4 4

3.5 7,8

3.6 2, 12

4.7 7