

## Key ideas for topic 14: Joint Distributions June 8

**Joint distribution:**

For the discrete case of two variables  $x, y$  the following are joint probabilities

$$\begin{aligned}\mathbf{P}(X = x, Y = x) &= P(x, y) \\ \mathbf{P}((X, Y) \in B) &= \sum_{(x, y) \in B} P(x, y),\end{aligned}$$

with

$$P(x, y) \geq 0, \quad \sum_x \sum_y P(x, y) = 1.$$

For the continuous case of two variables  $x, y$  the following is a joint probability density

$$\mathbf{P}(X \in dx, Y \in dy) = f(x, y) dx dy$$

with

$$f(x, y) \geq 0, \quad \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1,$$

and

$$\mathbf{P}((X, Y) \in B) = \int_{(x, y) \in B} f(x, y) dx dy.$$

**Independence:**

For discrete distributions:

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

For continuous distributions:

$$f(x, y) = f_X(x)f_Y(y).$$

**Sta-230: Probability****Summer 2020 I**

Key ideas for topic 15: Conditional and marginal distributions June 9

**Joint and conditional probability for discrete random variables:**

For the discrete case of two variables  $x, y$  the following relation between joint and conditional probabilities hold

$$\begin{aligned}P(X = x, Y = y) &= P(X = x)P(Y = y | X = x) \\P(X = x | Y = y) &= \frac{P(X = x, Y = y)}{P(Y = y)} \\P(Y = y | X = x) &= \frac{P(X = x, Y = y)}{P(X = x)}.\end{aligned}$$

**Bayes rule**

$$P(X = x | Y = y) = \frac{P(Y = y | X = x)P(X = x)}{P(Y = y)}.$$

**Conditional probability and expectation:**

$$\begin{aligned}P(Y \in B | X = x) &= \sum_{y \in B} P(Y = y | X = x) \\ \mathbb{E}(Y | X = x) &= \sum_y yP(Y = y | X = x).\end{aligned}$$

**Marginals:**

$$\begin{aligned}P(B) &= \sum_x P(B | X = x)P(X = x) \\ \mathbb{E}[Y] &= \sum_x \mathbb{E}(Y | X = x)P(X = x).\end{aligned}$$

**Joint and conditional probability for continuous random variables:**

For the continuous case of two variables  $x, y$  the following relation between joint and conditional probabilities hold

$$\begin{aligned}f(x, y) &= f_X(x)f(y | X = x) \\f(y | X = x) &= \frac{f(x, y)}{f_X(x)} \\f(x | Y = y) &= \frac{f(x, y)}{f_Y(y)}.\end{aligned}$$

**Bayes rule**

$$f(X | Y = y) = \frac{f(y | X = x)f_X(x)}{f_Y(y)}.$$

**Conditional probability and expectation:**

$$\begin{aligned}P(Y \in B | X = x) &= \int_{y \in B} f_Y(y | X = x) dy \\ \mathbb{E}(Y | X = x) &= \int_y y f(y | X = x) dy.\end{aligned}$$

**Marginals:**

$$\begin{aligned}P(B) &= \int P(B | X = x)f_X(x)dx \\ \mathbb{E}[Y] &= \int \mathbb{E}(Y | X = x)f_X(x)dx.\end{aligned}$$

## Key ideas for topic 16: Covariance and Correlation June 10

**Covariance:**

The covariance of two random variables  $X$  and  $Y$  is

$$\text{Cov}(X, Y) = \mathbb{E}[(X - \mu_X)(Y - \mu_Y)], \quad \mu_X = \mathbb{E}[X], \quad \mu_Y = \mathbb{E}[Y],$$

and

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y).$$

**Correlation:**

The correlation between two random variables  $X$  and  $Y$  is

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}.$$

**Cauchy-Schwarz-Bunyakovsky inequality:**

$$|\mathbb{E}(XY)| \leq \sqrt{\mathbb{E}[X^2]\mathbb{E}[Y^2]}$$

Cauchy-Schwarz tells the correlation is always between  $-1$  and  $1$ .

**Sta-230: Probability****Summer 2020 I**

Key ideas for topic 17: Tower rules for expectation and variance June 12

**Adam's law:**

$$\mathbb{E}[X] = \mathbb{E}[\mathbb{E}[X|Y]].$$

**The conditional variance:**

$$\text{Var}(X|Y) = \mathbb{E}[(Y - \mathbb{E}[Y|X])^2|X]$$

is defined analogously as the usual variance, just everything conditional on  $X$ .**Eves' law:**

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

Eve's law is a special case of more general phenomenon

$$\mathbb{E}[g(X, Y)] = \mathbb{E}[\mathbb{E}[g(X, Y)|Y]]$$