## Lecture number XX: Probability Recap

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## 1 Preliminaries

A variable X is a random variable if it assumes different values according to a probability distribution. For example, X can denote the outcome of a three sided die throw. The variable X takes the values x = 1, 2, 3 with equal probabilities.

The expectation of X is the sum over the possible values times the probability of the events.

$$\mathbb{E}[X] = \sum_{x=1}^{3} x \Pr(X = x) = 1\frac{1}{3} + 2\frac{1}{3} + 3\frac{1}{3} = 2$$
 (1)

Continuous random variable are described by their distribution function  $\varphi$ .

$$\Pr[Y \in [a, b]] = \int_{a}^{b} \varphi(t)dt.$$

For a function  $\varphi$  to be a valid distribution we must have:

$$\forall t, \varphi(t) \geq 0 \text{ (where it is defined)}$$
 (2)

$$\int_{a}^{b} \varphi(t)dt \qquad \text{is well defined for all } a \text{ and } b$$
 (3)

$$\int_{-\infty}^{-\infty} \varphi(t)dt = 1 \tag{4}$$

For example consider the continuous variable Y taking values in [0,1] uniformly. That means  $\varphi(t) = 1$  if  $t \in [0,1]$  and zero else. This means that the probability of Y being in the interval [t,t+dt] is exactly dt. And so the expectation of Y is:

$$\mathbb{E}[Y] = \int_{t=0}^{1} t\varphi(t)dt = \int_{t=0}^{1} t \cdot dt = \frac{1}{2}t^{2}|_{0}^{1} = 1/2$$
 (5)

**Remark 1.1.** Strictly speaking, distributions are not necessarily continuous or bounded functions. In fact, they can even no be a function at all. For example, the distribution of X above includes three Dirac- $\delta$  functions which are not valid functions.

## 1.1 Dependence and Independence

A variable X is said to be *dependent* on Y if the distribution of X given Y is different than the distribution of X. For example. Assume the variable X takes the value 1 if Y takes a value of less than 1/3 and the values 2 or 3 with equal probably otherwise (1/2 each). Clearly, the probability of X assuming each of its

values is still 1/3. however, if we know that Y is 0.7234 the probability of X assuming the value 1 is zero. Let us calculate the expectation of X given Y as an exercise.

$$\mathbb{E}(X|Y) = \sum_{x=1}^{3} x \Pr(X = x|Y \le 1/3) = 1 \cdot 1 \tag{6}$$

$$\mathbb{E}(X|Y) = \sum_{x=1}^{3} x \Pr(X = x|Y > 1/3) = 1 \cdot 0 + 2\frac{1}{2} + 3\frac{1}{2} = 2.5$$
 (7)

E(X|Y) = 1 for  $y \in [0, 1/3]$  and E(X|Y) = 2.5 for  $y \in (1/3, 1]$ .

Remember:  $\mathbb{E}(X|Y)$  is a function of y!

**Definition 1.1** (Independence). Two variables are said to be Independent if:

$$\forall y, \ \Pr[X = x | Y = y] = \Pr[X = x].$$

They are dependent otherwise.

**Fact 1.1.** If two variables X and Y are Independent the so are f(X) and g(Y) for any functions f and g.

Fact 1.2 (Linearity of expectation 1). For any random variable and any constant  $\alpha$ :

$$\mathbb{E}[\alpha X] = \alpha \mathbb{E}[X] \tag{8}$$

Fact 1.3 (Linearity of expectation 2). For any two random variables

$$\mathbb{E}_{X,Y}[X+Y] = \mathbb{E}[X] + \mathbb{E}[Y] \tag{9}$$

even if they are dependent.

Fact 1.4 (Multiplication of random variables). For any two independent random variables

$$\mathbb{E}_{X,Y}[XY] = \mathbb{E}[X]\mathbb{E}[Y] \tag{10}$$

This does not necessarily hold if they are dependent.

**Definition 1.2** (Variance). For a random variable X we have

$$\operatorname{Var}[X] = \mathbb{E}[(X - \mathbb{E}[X])^2] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2 \tag{11}$$

The standard deviation  $\sigma$  of X is defined to be  $\sigma(X) \equiv \sqrt{\operatorname{Var}[X]}$ .

**Definition 1.3** (Additivity of variances). For any two independent variables X and Y we have

$$Var[X + Y] = Var[X] + Var[Y]$$
(12)

**Fact 1.5** (Markov's inequality). For any positive random variable X:

$$\Pr(X > t) \le \frac{E[X]}{t} \tag{13}$$

Fact 1.6 (Chebyshev's inequality). For any random variable X

$$\Pr[|X - E[X]| > t] \le \frac{\sigma^2(X)}{t^2} \tag{14}$$