## Math 340: Lec 14 Big Ideas Journal (Discrete joint, marginal, and conditional distributions)

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## Definition of distributions

**Definition 1** (joint distribution). Let X and Y be two discrete random variables. Their joint distribution is

$$\mathbb{P}(X = k, Y = j) \text{ for } k \in R(X), j \in R(Y)$$

Remark. The joint distribution can be considered a mass function with domain in the 2D-plane.

**Remark.** If X, Y are independent,  $\mathbb{P}(X = k, Y = j) = \mathbb{P}(X = k)\mathbb{P}(Y = j)$ , and we say that the joint probability distribution has a product structure.

**Definition 2** (marginal distribution). Let X be a discrete random variable. The marginal distribution of X is  $\mathbb{P}(X = k)$ .

**Remark.** Even if a set of three random variables have the same marginal distributions, they may have different joint distributions if one of the random variables is dependent on another.

**Definition 3** (Conditional distribution). Let X and Y be discrete random variables. A conditional distribution looks like  $\mathbb{P}(X = k | Y = j)$ . The conditional distribution is given by:

$$\mathbb{P}(X = k | Y = j) = \frac{\mathbb{P}(X = k, Y = j)}{\mathbb{P}(Y = j)}$$

## Moving between distributions

**Remark.** We can find the marginal distribution  $\mathbb{P}(X=k|Y=j)$  given the joint probability distribution  $\mathbb{P}(X=k,Y=j)$  by normalizing the joint probabilities of X and Y for a given Y=j. We can use the same process to find  $\mathbb{P}(Y=j|X=k)$ .

**Remark.** We can relate the joint and marginal distributions using the partition rule. In particular, if we know the joint distributions of X and some some other r.v. Y, we can recover the marginal distribution of X (and/or Y). But if we know the marginal distribution of X, we cannot recover the joint distribution of X and Y without add'l info.

**Example** (Calculating expected value of the marginal distribution using the conditional). By the partition rule,

$$\mathbb{E}[X] = \sum_{j \in R(Y)} \mathbb{E}[X|Y = j)\mathbb{P}(Y = j)$$

This lets us calculate  $\mathbb{E}[X]$  when it's difficult to think directly about X but easier to think about X|Y. And note

$$\mathbb{E}[X|Y=j] = \sum_{k \in R(X)} k \mathbb{P}(X=k|Y=j)$$