

- Today, suppose  $X$  and  $Y$  are two r.v.'s, such that  $X$  is dependent on  $Y$ , so that  $X|Y=y \sim f_{X|Y=y}$

- Now,  $X|Y=y$  is a random variable.

(Ex1)  $Y \sim \text{Binomial}(n, p)$ , "no. of heads in 20 tosses"  
 $X|Y=y \sim \text{Bernoulli}(\frac{y}{n})$ , "you win money, with prob. of heads"

(Ex2)  $H \sim \text{Normal}(\mu, \sigma^2)$ , "height of person"  
 $W|H=h \sim \text{Normal}(2h, \sigma^2)$ , "weight of people, given they have height  $h$ "

(Ex3)  $Y \sim \text{Bernoulli}(\frac{1}{n})$ , "odds of picking fair coin amongst  $n$  unfair coins"  
 $X|Y=y \sim \text{Geometric}((\frac{1}{2})^y (\frac{1}{3})^{1-y})$ , "number of tosses to get one head"

The above demonstrate that something like " $X|Y$ " is a random variable in its own right, w/ the conditioning r.v.  $Y$  specifying a parameter of the distribution, or something we already know

↓  
 Thus, we can compute statistics like mean variance etc. about  $X$ .

Today

① New kind of r.v.'s - "conditional" r.v.'s  
 $X|Y$   $X|Y=y$   $X|A$  (an event)  
 $X|Y=y$   $E[X|Y]$   
 $X|Y=y$

② Inequalities

(1) Markov

(2) Chebyshev

(3) WLLN

(Ques 1) Say,  $Y \sim \text{Binomial}(n, p)$ , "no. of heads in 20 tosses". What is  $E[X|Y]$ ? What does it denote? Is it a random variable or a deterministic value?  
 $X|Y=y \sim \text{Bernoulli}(\frac{y}{n})$ , "you win money, with prob. of heads"  
 (Hint: if  $Z \sim \text{Bernoulli}(\theta)$ ,  $E[Z] = \theta$ )

$$\text{(Formally, } E[X|Y] = \int_{\mathcal{X}} x f_{X|Y}(x) dx)$$

(Ques 2) Say,  $Y \sim \text{Bernoulli}(\frac{1}{n})$ , "odds of picking fair coin amongst  $n$  unfair coins". What is  $E[X|Y]$ ? What does it denote? Is it a random variable or a deterministic value?  
 $X|Y=y \sim \text{Geometric}((\frac{1}{2})^y (\frac{1}{3})^{1-y})$ , "number of tosses to get one head"  
 (Hint: if  $Z \sim \text{Geom}(p)$ ,  $E[Z] = \frac{1}{p}$ )

Another thing we may do is compute  $E[X]$ , i.e. unconditional expectation, from the conditional expectation.

(Ques) Say,  $Y \sim \text{Binomial}(n, p)$ , "no. of heads in 20 tosses". What is  $E[X]$ ? What does it denote? Is it a random variable or a deterministic value?  
 $X|Y=y \sim \text{Bernoulli}(\frac{y}{n})$ , "you win money, with prob. of heads"

$$\begin{aligned} \text{(Formally,)} \\ \text{since } E[X|Y] &= g(Y), \\ E[E[X|Y]] &= E[g(Y)] \\ &= \int_{\mathcal{Y}} g(y) f_Y(y) dy \\ &= \int_{\mathcal{Y}} \int_{\mathcal{X}} x f_{X|Y}(x) dx f_Y(y) dy \\ &= \int_{\mathcal{Y}} \int_{\mathcal{X}} x f_{XY}(x, y) dx dy \\ &= \int_{\mathcal{X}} \int_{\mathcal{Y}} x f_{XY}(x, y) dy dx \\ &= \int_{\mathcal{X}} x f_X(x) dx = E[X] \end{aligned}$$

### Other properties

- (1)  $\mathbb{E}[Xh(Y)|Y] = h(Y)\mathbb{E}[X|Y]$
- (2)  $\mathbb{E}[aX + bY|Z] = \mathbb{E}[aX|Z] + \mathbb{E}[bY|Z]$   
 $= a\mathbb{E}[X|Z] + b\mathbb{E}[Y|Z]$
- (3) If  $X, Y$  independent,  $\mathbb{E}[X|Y] = \mathbb{E}[X]$
- (4) If  $c \in \mathbb{R}$ ,  $\mathbb{E}[c|Y] = c$

If these r.v.'s have calculable expected values, they must have calculable variances too!

Indeed:

$$\text{Var}(X|Y=y) = \mathbb{E}\left[\frac{(X - \mathbb{E}[X|Y=y])^2}{Y=y}\right], \text{ Variance of } \underline{X|Y=y}$$

### Law of total variance

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

(Ques) Compute the variance of  $X$ , given

$$Y \sim \text{Binomial}(n, p), \text{ " \# of heads in } n \text{ tosses"}$$

$$X|Y=y \sim \text{Bernoulli}\left(\frac{y}{n}\right), \text{ "you win money, no. of heads with winning prob of } \frac{y}{n} \text{"}$$

### Inequalities

- Finitely many r.v.'s  $\xrightarrow{\text{Markov}}$  Infinite sequence of r.v.'s. WLLN, SLLN, CLT
- Behavior for finite sample size  $\xrightarrow{\text{Chebyshev}}$  Asymptotic behavior

(Ques) Prove Markov's inequality, explicitly stating which assumption is used where:

If  $X$  is non-negative, continuous random variable, then for  $a > 0$ ,

$$\mathbb{P}[X \geq a] \leq \frac{\mathbb{E}X}{a}$$

$$\text{For } a > 0, \mathbb{P}[X \geq a] = \int_a^\infty f_X(x) dx \leq \frac{1}{a} \int_a^\infty x f_X(x) dx \leq \frac{1}{a} \int_0^\infty x f_X(x) dx = \frac{\mathbb{E}X}{a}$$

(Ques) Prove Chebyshev's inequality, explicitly stating which assumption is used where:

For arbitrary r.v.  $X$  with finite mean  $\mu$ ,  $\forall \epsilon > 0$

$$\mathbb{P}[|X - \mu| > \epsilon] \leq \frac{\text{Var}(X)}{\epsilon^2}$$

$$\mathbb{P}[|X - \mu| > \epsilon] = \mathbb{P}[|X - \mu|^2 > \epsilon^2] \leq \frac{\mathbb{E}[|X - \mu|^2]}{\epsilon^2} = \frac{\text{Var}(X)}{\epsilon^2}$$

$$\epsilon^2$$

$$\mathbb{P}[|X - \mu| > \epsilon] = \mathbb{P}[|X - \mu|^2 > \epsilon^2] \leq \frac{\text{Var}(X)}{\epsilon^2}$$

The above theorems are very cool because they give bounds on the distribution of a large class of r.v.'s - they tell us that, just because your mean is finite, there's only so much probability that you can deviate from it.

(Ex) Suppose the mean on a probability exam is 75.

With just this information,

(i) Find an upper bound for probability that someone scored above 90.

(ii) If I told you that 85% of students scored > 90, would that be possible or impossible? Why? What about 75%?

(iii) Suppose you gain some information - specifically, the standard deviation was 5. Answer (i) and (ii) again.

(Ex) Suppose that students at Hopkins take an average of 16 credits / semester, with standard deviation of 1.3.

Compute a bound for the probability that a student is taking  $\geq 15$  and  $\leq 17$  credits.

(Ques) Prove the following version of "The Weak Law of Large Numbers"

Given  $X_1, \dots, X_n \sim \text{iid}$  with finite mean  $\mu$ , let  $\bar{X} = \frac{\sum X_i}{n}$ , "sample mean".  
& finite variance  $\sigma^2$

Then,  $\forall \epsilon > 0$ ,  $\mathbb{P}[|\bar{X} - \mu| > \epsilon] \rightarrow 0$  as  $n \rightarrow \infty$ .

(Hint, consider the r.v.  $\bar{X}$  and apply Chebyshev)