

CSE 381

Tousif Sir Class-01

09.02.14

Probability models  $\Rightarrow$  Conditional expectation  
Markov chain

~~Ref:~~ Probability models  
Sheldon Ross  
10th ed.

Probability is the study of something

- U "outcome uncertain"
- "quantifying uncertainty"

Applications: "Simulation"

"Algorithm analysis"

sp sample space : (S.)

Set of possible outcomes

1. Toss of a ~~coin~~ coin,  $S = \{\text{head, tail}\}$

2. Roll of a dice,  $S = \{1, 2, 3, 4, 5, 6\}$

3. Pick a card from deck,  $S = \left\{ \begin{array}{l} \text{hearts diamonds} \\ (H, 2) (D, 2) \end{array} \right\}^{52}_{2 \text{ colors}}$

13 rank!

2, 3, ... 10

J, K, Q, A

4. Level-term of a BOBT stud.

$$S = \{(1,1), (1,2) \sim (3,2, \dots)\} : 18$$

5. Lifetime of a bulb,  $S = [0, \infty)$

$$\begin{array}{ll} a \leq x \leq b & [a, b] \Rightarrow \text{close} \\ a < x < b & (a, b) \Rightarrow \text{open} \end{array}$$

6. # of students in a class,

$$S = [0, 60]$$

Event: E  
subset of sample space,

1. Head appears,  $E = \{\text{head}\}$  ④

2. 6 comes in dice,  $E = \{6\}$  1/6

$$E = \{\text{greater than } 3\}$$

$$= \{4, 5, 6\} = \frac{3}{6} = \frac{1}{2}$$

3. E {clubs} = {2, 10, J, K, Q, A} : 13

$$E \{\text{aces}\} = \{A, D, C, S\}$$

⑤ 4.  $E = \{ \geq 10 \text{ hours} \} = [10, \infty)$

$$= \{ \leq 10 \text{ hours} \} = [0, 10]$$

⑥  $E = \{4^{\text{th}} \text{ level}\}$

$$= \{(4,1), (4,2)\}$$

tuple.

Event E, F

$$E = \{1, 2, 3\}$$

$$F = \{4, 5\}$$

$$E \cup F = \{1, 2, 3, 4\}$$

$$E \cap F = \{3\}$$

↓

$$EF = \{3\}$$

$$E = \{1, 2\} \quad EF = \emptyset.$$

$$F = \{4, 5\} \Rightarrow E \text{ & } F \text{ are mutually exclusive}$$

### Probability %

Probability, p, of an event

$P(E)$  = prob that E occurs

p is a function from sample space to

$$[0, 1]$$

$$p: |S| \rightarrow [0, 1]$$

$P(E) = 0$   
why?  
etc  
stays?

## Probability

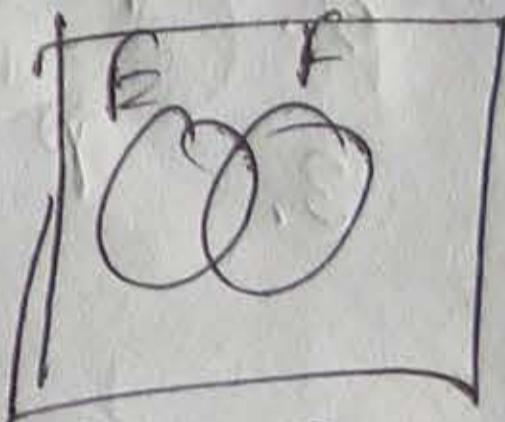
$$\text{① } 0 \leq P(E) \leq 1$$

$$\text{② } P(S) = 1$$

③ If  $E_1, E_2, E_3, \dots, E_n$  which are mutually exclusive

$$P\left(\bigcup_{i=1}^n E_i\right) = \sum_{i=1}^n P(E_i) \quad P(E \cup F) = P(E) + P(F)$$

$$P(E) = \frac{\text{sample favouring Event } E}{\text{Total sample}}$$



$$P(E \cup F) = P(E) + P(F) - P(E \cap F)$$

## Probability Models

$S$  = Sample Space, {outcomes}

$E$  = event, subset of  $S$

$P(E)$  = Probability that event  $E$  happened

$$1. \quad 0 \leq P(E) \leq 1$$

$$2. \quad P(S) = 1$$

$$3. \quad P(E \cup F) = P(E) + P(F)$$

when  $E, F$  mutually exclusive

$$P(E \cap F) = \phi, \quad (EF = \phi)$$

$P(E \cap F) \rightarrow$  probability that both  $E \& F$  happen

## Conditional Probability

$P(E|F)$  = Prob that  $E$  happen given  $F$  has been occurred

$$P(E) \leq P(E|F)$$

1.  $P(E) = P(E|F)$  i.e.,  $E \& F$  ~~are~~ independent

By definition,

$$P(E|F) = \frac{P(EP)}{P(F)}$$

$$\Rightarrow P(EP) = P(B|F) * P(F)$$

$$\Rightarrow P(P|B) * P(E)$$

Q.W  
-E1

If  $E$  &  $F$  are Ind.,

$$P(E) = P(E|R)$$

then

$$P(EF) = P(E) * P(F) \quad [① \text{ } P(E|R) \text{ } \text{so value same}]$$

But

$$P(ER) = P(E) * P(F) \text{ but } E \& F \text{ Ind. } \text{at}^2 \text{ same}$$

#

$$P(EF) = P(E|R), P(F) = P(R|E)P(E)$$

$$\therefore P(E|R) = \frac{P(ER)}{P(F)}$$

$$P(E|R) = \frac{P(R|E)P(E)}{P(F)} \quad \text{Bayes Rule}$$

Example

Ben taken computer or chemistry.

A max mark chance

$$P(A \text{ grade}) = \frac{1}{2}$$

$$P(A | CS) = \frac{1}{2}$$

$$P(A \text{ grade}) = \frac{1}{3}$$

$$P(A | ch) = \frac{1}{3}$$

Q. What is the probability that Ben gets A in CHE

Ans

$$P(A \text{ chE}) = P(A \mid \text{chE}) \times \underline{P(\text{chE})}$$

$$= \frac{1}{3} \times \frac{1}{2}$$

$$= \frac{1}{6}$$

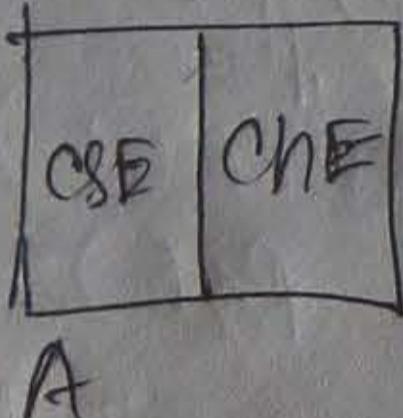
Also,

$$P(A \text{ CSB}) = \frac{1}{2} \times \frac{1}{2}$$

$$P(A \text{ grade}) = P(\text{ACSB}) + P(\text{AchE})$$

$$= P(\text{CSB}) P(A \mid \text{CSB}) + P(\text{chE}) \cdot P(\text{AchE})$$

$$= \frac{1}{2} \cdot \frac{1}{4} + \frac{1}{6}$$



A

$$\times P(F \mid B) = \frac{P(B \mid F) P(F)}{P(B)}$$

$F' = \text{complement of } F$

$$= \frac{P(B \mid F) P(F)}{P(F \mid F) P(F) + P(F \mid F') P(F')}$$

P with respect to  
F while E

F with respect to  
F after E

respect to B who

Q. In MCQ exams  
A student knows answer or makes a guess.

$$P(\text{Know}) = p$$

$$P(\text{Guess}) = 1-p$$

$$P(\text{Correct} | \text{Knows}) = 1$$

correct ans. given that he knows

$$P(C|K^c) = P(C|U) = \frac{1}{m}$$

[no. of options  $m$        $\begin{matrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{matrix}$ ]

# Find Prob that student "knows" given that he gave the "correct ans"

$$P(K|C) = \frac{P(C|K) \cdot P(K)}{P(C)}$$

$$= \frac{1 \cdot p}{P(C|K) \cdot P(K) + P(C|U) \cdot P(U)}$$

$$= \frac{p}{1 \cdot p + \frac{1}{m} (1-p)}$$

if  $p = \frac{1}{2}$  ;

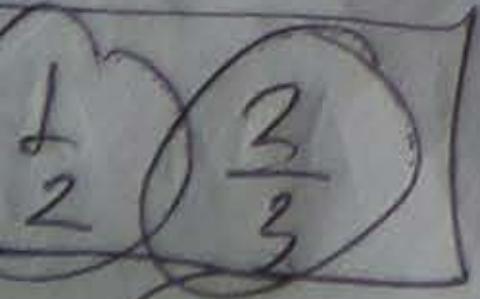
$$P(K|C) = \frac{k_2}{k_2 + \frac{1}{m} \frac{1}{2}} = \frac{m}{m+1}$$

if  $p = \frac{1}{3}$  ;

$$P(K|C) = \frac{k_3}{k_3 + \frac{1}{m} \frac{2}{3}} = \frac{m}{m+2}$$

if  $m=4$  ,

$$P(K|C) = \frac{4}{5}$$



## Random Variable

(30)  
Yousuf SSB-3

22/02/14

Sample space  
event,  $P(\text{event}) =$

$$0 \leq P(E) \leq 1$$

$$\sum_S P(E) = 1$$

$X: \text{Event} \rightarrow \mathbb{R}$

Event event can represent real number

represent not

1. Toss of a coin, {head, tail}

$$\{H, T\} \quad P\{X=H\} = \frac{1}{2}$$

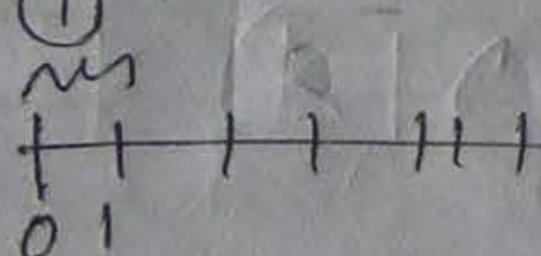
2. Roll of a dice  $\{1, 2, 3, \dots, 6\}$   $P\{X=3\} = \frac{1}{6} / b(3) = \frac{1}{6}$

3. Pick of a card,  $X \in \{1, 2, \dots, 52\}$

4. Life of a bulb  $\leftarrow \infty$

Now, Sample space is nothing but

① go sample space



Real Line/  
Number Line

Random Variable  $\begin{cases} \rightarrow \text{Discrete} \\ \searrow \rightarrow \text{Continuous} \end{cases}$

## Discrete

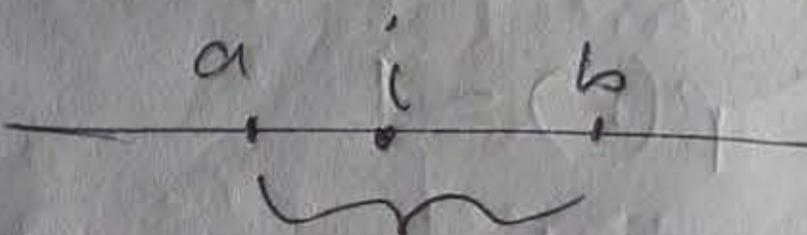
## Continuous

$X$  takes a discrete value

$X$  takes a continuous value

$$\textcircled{1} \quad \phi(i) = P\{X=i\} \quad (\text{from def.}) \quad \textcircled{2} \quad \phi(i) = P\{X=i\} > 0$$

↑  
PMF (Probability Mass function)



$$\textcircled{3} \quad \sum_{i \in X} \phi(i) = 1$$

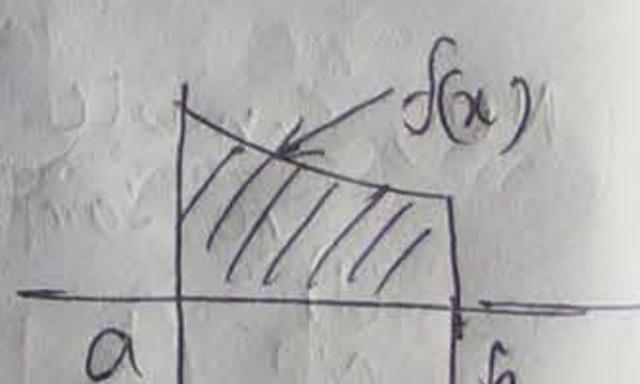
$$\textcircled{1} \quad \textcircled{2} \quad P(X \in I) = P(a < x < b)$$

$$= \int_a^b f(x) dx$$

↑  
probability density function

mass

density



$$\textcircled{3} \quad \int_{-\infty}^{\infty} f(x) dx = 1$$

## # Discrete Random Variable

1. Bernoulli:  $X = \{1, 0\}$  Result of a "trial".  
 (p)

Bernoulli(p)

$$X = \begin{cases} 1, & \text{if success} \\ 0, & \text{if fail} \end{cases}$$

mass.  $P(1) = P\{X=1\} = p$  success probability

$P(0) = P\{X=0\} = (1-p)$  failure "

2. Binomial: # of success in n trials.  $[0, n]$

$$(n, p) \quad P(i) = P(X=i) = \binom{n}{i} p^i (1-p)^{n-i}$$

bin(n, p)

Verify:  $\sum_{i=0}^n P(i) = 1.$

3. Geometric: # of trials to have first success.

$$X = [1, \infty)$$

$$\text{Geo}(p) \quad P(i) = P\{X=i\} = P(\text{i-1 fail, } i\text{th success}) \\ = (1-p)^{i-1} \cdot p = p(1-p)^{i-1}$$

$$x \ x \ x \ x \ x \checkmark \times \text{ith} \\ (i-1)$$

Verify:  $\sum_{i=1}^{\infty} P(i) = 1$

(301)

Ex: There are 10 Ques Yousuf Sir - 4  
 (# optimally)

23/02/14

Q1 Prob that a std gets 4 correct answer by guessing

Sol:  $X \rightarrow$  # correct ans out of 10.

$$X \sim \text{bin}(n, p) \quad n = 10$$

$$P(X=4) = \text{succ. prob.} = \frac{1}{4}$$

$$P(X=4) = \binom{10}{4} \left(\frac{1}{4}\right)^4 \left(\frac{3}{4}\right)^6$$

Q2 Prob that he gets more than 8%.

$$P(X > 5) = P(X=6) + \dots + P(X=10)$$

Poisson is a limiting case of binomial, n

rate is very large, p is small

$$X \sim \text{bin}(n, p) \quad \text{AP: } np = \lambda, \text{ constant}$$

$$\begin{aligned} P(i) &= P(X=i) = \binom{n}{i} p^i (1-p)^{n-i} \\ &= \frac{\ln}{i!} \frac{n!}{(n-i)!} \cdot \frac{i!}{n!} \left(1 - \frac{2}{n}\right)^{n-i} \quad [p = \frac{2}{n}] \end{aligned}$$

$$\begin{aligned} &\rightarrow \frac{\ln}{n!} \frac{n!}{i!} \frac{(1 - \frac{2}{n})^n}{\left(1 - \frac{2}{n}\right)^i} \\ &\quad \dots i \text{ terms} \end{aligned}$$

$$= \frac{n(n-1)(n-2)\dots(n-i+1)}{n^i} \cdot \frac{\lambda^i}{i!} \cdot \frac{\left(1-\frac{\lambda}{n}\right)^n}{\left(1-\frac{\lambda}{n}\right)^i}$$

$n \rightarrow \infty$

$$= e^{-\lambda} \frac{\lambda^i}{i!}$$

$$\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n = e^{-\lambda}$$

$$P(i) = P(X=i) = e^{-\lambda} \frac{\lambda^i}{i!}$$

sum does not < 270  
self study.

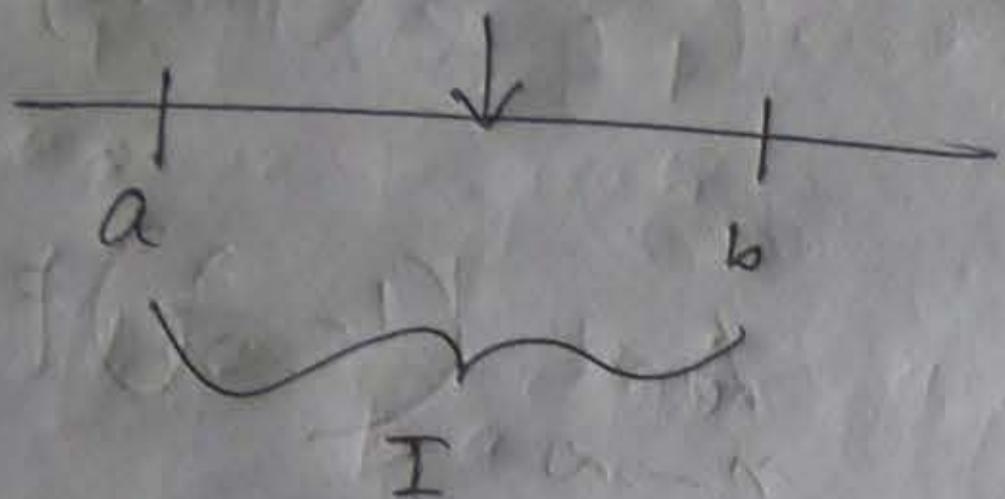
Ex: The # errors/spelling mistakes in a textbook page in poisson distribution with  $\lambda = 10$

Q: Bob there is at least one error

Sol:  $X \rightarrow \# \text{ of error}$   $X \sim \text{Poisson}(\lambda)$

$$\begin{aligned} P(X > 0) &= 1 - P(X=0) \\ &= 1 - e^{-\lambda} \frac{10^0}{0!} \\ &= 1 - e^{-10} \end{aligned}$$

## # Continuous Random Variable



prob. density function

$$P(X \in I) = P(a < X < b) = \int_a^b f(x) dx$$

$$F(a) = P(X \leq a) = P(-\infty \leq X \leq a)$$

$$\downarrow \text{CDF} \Rightarrow F(a) = \int_{-\infty}^a f(x) dx$$

Cumulative distribution function

$$f(a) = \int_{-\infty}^a f(x) dx$$

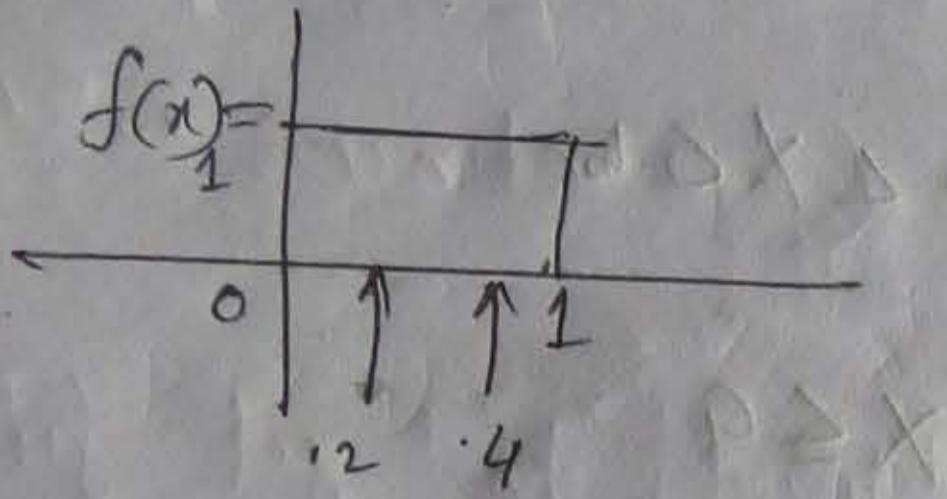
$$F(u) = \int_{-\infty}^u f(t) dt \Rightarrow \int_{-\infty}^u f(u) du$$

$$F(y) = \int_{-\infty}^y f(x) dx$$

✓ Uniform(0, 1)

$$0 < X < 1$$

density  $f(x) = 1, ; 0 < x < 1$   
 $= 0 ; \text{else}$



$$\begin{aligned} P(0.2 < X < 0.4) &= \int_{0.2}^{0.4} f(x) dx = \int_{0.2}^{0.4} 1 dx \\ &= x \Big|_{0.2}^{0.4} = 0.4 - 0.2 = 0.2 \end{aligned}$$

CDF,

$$F(x) = \int_{-\infty}^x f(t) dt$$

$$\Rightarrow \int_0^x 1 dt = t \Big|_0^x = x$$

$$\therefore \text{CDF}; F(x) = x$$

$$F(0.7) = P(X \leq 0.7) = 0.7$$

$U(a, b)$



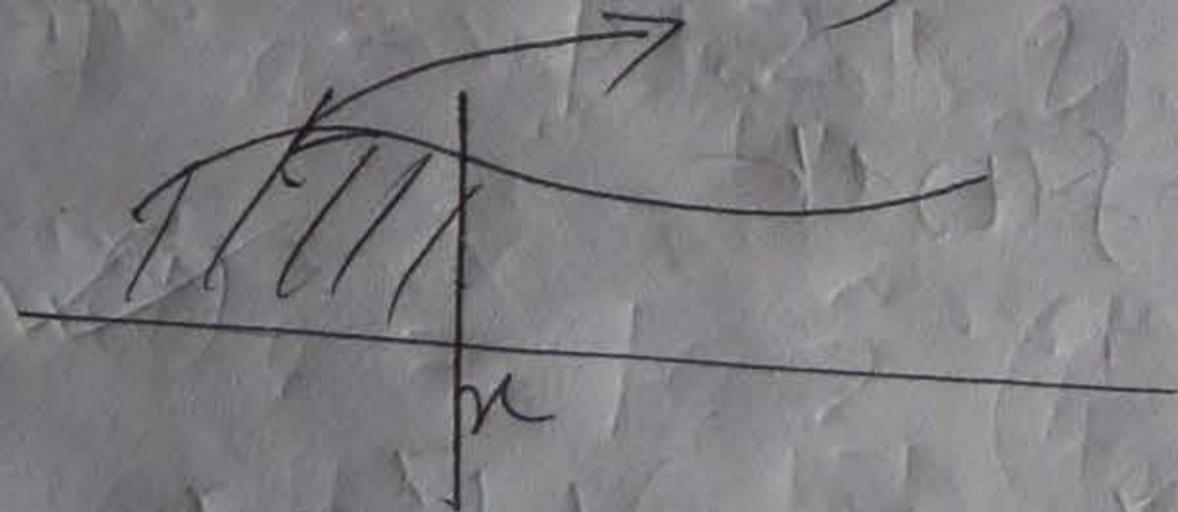
pdf,  $f(x) = \frac{1}{b-a}$ ;  $a < x < b$

cdf,  $F(x) = \frac{x-a}{b-a}$ ;  $a < x < b$

$$\begin{aligned} &= 0 & x \leq a \\ &= 1 & x \geq b \end{aligned}$$

$F(x)$  is monotonically increasing

$$F(x) = P(X \leq x)$$

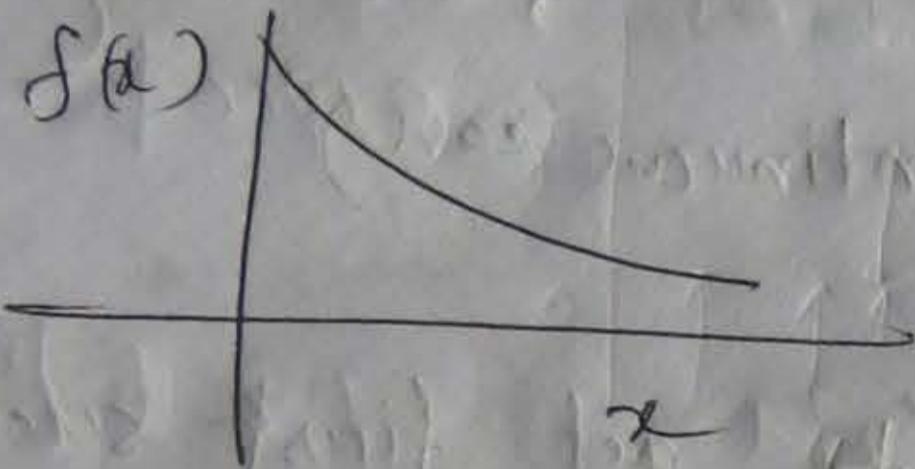


) area always  
increasing for  
increasing  $x$ .

## 2. Exponential Random Variable

$$f(x) = \lambda e^{-\lambda x}, x \geq 0$$

Verify,  $\int_0^\infty f(x) dx = 1$



$$F(x) = 1 - e^{-\lambda x}$$

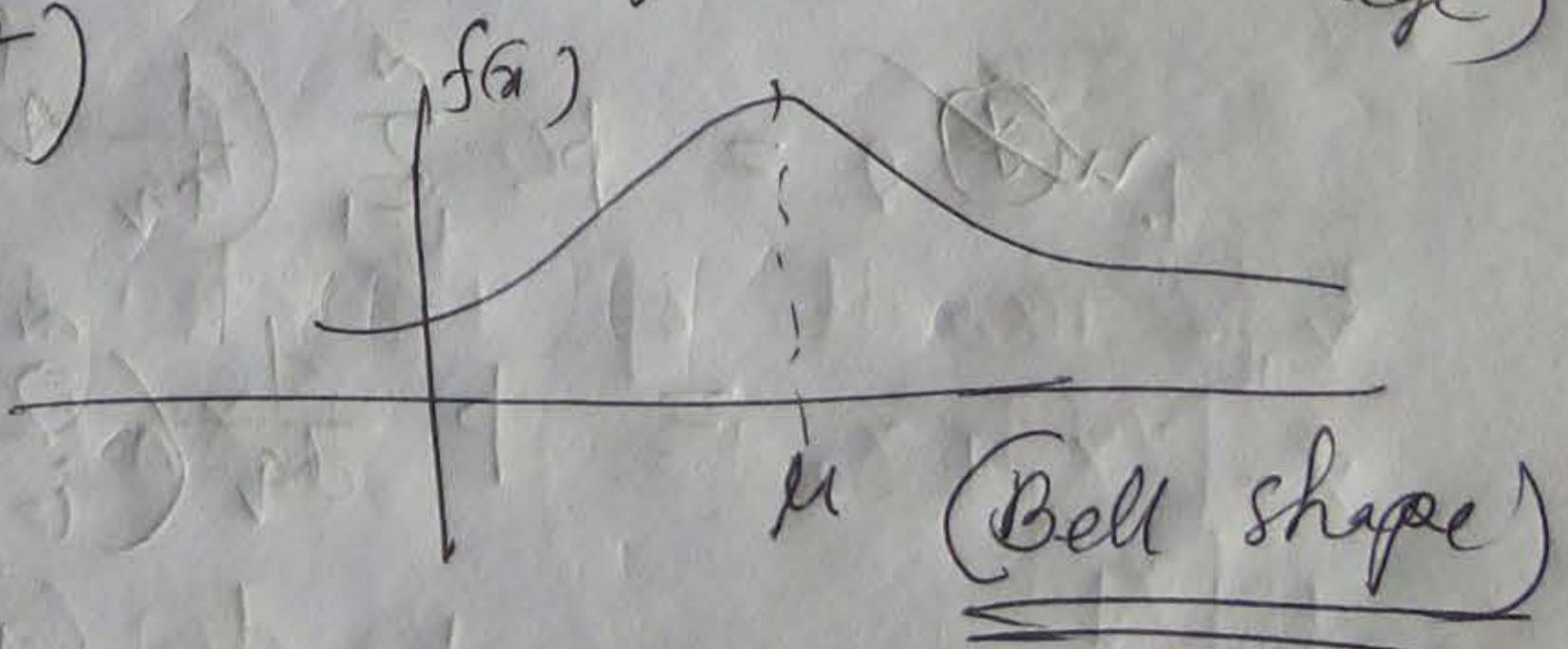
(2)  $\lambda$    
 time measure  
 - const  $\lambda$   
 expmt H21  
 (3)  $\lambda$    
 std G 1st br  
 std  $\lambda$    
 var  $\lambda$    
 std  $\lambda$    
 var  $\lambda$

3. Gamma:  $f(x) = \frac{\lambda^x x^{x-1}}{\Gamma(x)}, x > 0$

4. Normal:  $f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, -\infty < x < \infty$

gaussian  $(\mu, \sigma^2)$

(avg height of tree)  
 Ex. ~~height~~ of people of a certain age



③ + ④ many detail G onto  $\sim$

Expectation

Yusuf SID-5

24/02/14

Expectation of a random variable

mean, average

$$\begin{aligned} E[X] &= \sum_i i p(i) \quad (\text{in discrete case}) \\ &= \int_{-\infty}^{\infty} x f(x) dx \quad (\text{in continuous case}) \end{aligned}$$

expected value

Ex:  $X \sim \text{geom}(p)$ .  $X = \# \text{ trial to get first success}$

$$\begin{aligned} E[X] &= \sum_{i=1}^{\infty} i p(i) \\ &= \sum_{i=1}^{\infty} i p(1-p)^{i-1} \\ &\geq p \sum_{i=1}^{\infty} i q^{i-1} \quad [q = 1-p] \end{aligned}$$

$$\begin{aligned} &= p \sum_{i=1}^{\infty} \left( \frac{d}{dq} (q^i) \right) \\ &= p \frac{d}{dq} \left( \sum_{i=1}^{\infty} q^i \right) \\ &= p \frac{d}{dq} (q + q^2 + \dots) \\ &\geq p \frac{d}{dq} \left[ q (1 + q + q^2 + \dots) - q \right] \\ &\approx p \frac{d}{dq} q \frac{q^2}{1-q} \end{aligned}$$

$$\begin{aligned} &= p \frac{1}{(1-p)^2} \\ &= p \frac{1}{p^2} = \frac{1}{p} \end{aligned}$$

$\boxed{E[X] = \frac{1}{p}}$

a team's winning probability  $\Rightarrow \frac{1}{4}$   
 (in a knockout series)

$\checkmark$      $\checkmark$      $\checkmark$      $\times$      $\frac{3}{4}$

So,  $E[X] = \frac{1}{\frac{3}{4}} = \frac{4}{3}$

Name

Bernoulli

:  $p$

Binomial :  $np$

Geometric :  $\frac{1}{p}$

Poisson :  $\lambda$

Theory (proof) not

LTO 1

$$E[X] = \sum_{i=0}^{\infty} i e^{-\lambda} \frac{\lambda^i}{i!} = \lambda$$

Uniform(0, 1) :  $\frac{1}{2}$

↓ 2nd proof (MN20)

Normal :  $\mu$

$$\text{exponential } (\lambda) : \frac{1}{\lambda}, \quad E[X] = \int_0^\infty x \lambda e^{-\lambda x} dx = \frac{1}{\lambda}.$$

Identity:

$$1. E[x+b] = aE[x] + b$$

$$E[c] = c \quad \text{constant}$$

$$2. E[x_1 + x_2] = E[x_1] + E[x_2]$$

$$3. E[a x_1 + b x_2 + \dots + c x_n] = a E[x_1] + b E[x_2] + \dots + c E[x_n]$$

linearity of expectation

#1 Binomial:

$$E[X] = \sum_{i=0}^n i \binom{n}{i} p^i (1-p)^{n-i} = np$$

$X \rightarrow \# \text{ of trials in } n \text{ trials.}$

1. 0

0.

1.

0

1

$X_i = \text{result of } i\text{th trial.}$

$$X = X_1 + X_2 + \dots + X_n$$

$$E[X] = 1 \cdot p + 0 \cdot (1-p) = p.$$

$$E[X] = E[X_1] + E[X_2] + \dots + E[X_n] = np$$

## Example:

$\downarrow$  (f - N) hat work  $102m^2$  hat per m<sup>2</sup>

$X \sim \#$  people who gets his own hat.

$$E[X] = ?$$

$\bar{x} \quad \bar{x} \quad \bar{x} \quad \bar{x} \quad \dots \quad \bar{x}$

$x_i$  = bernoulli trial  $\begin{cases} 1, & \text{success} \\ 0, & \text{failure} \end{cases}$

$$X = x_1 + x_2 + \dots + x_n$$

$$E[X] = E[x_1] + E[x_2] + \dots + E[x_n]$$

$$E[\bar{x}] = \frac{1}{N} \cdot N = 1$$

$$E[\bar{x}] = \frac{1}{N} + O\left(1 - \frac{1}{N}\right)$$

$E$  (constant) = constant

## Variance

$$\begin{aligned} \text{Var}(X) &= E[(X - E[X])^2] \\ &= E[X^2 - 2X E[X] + E[X]^2] \\ &= E[X^2] - 2 E[X] E[X] + E[X]^2 \end{aligned}$$

[Standard form]

$$\therefore \boxed{\text{Var}(X) = E[X^2] - E[X]^2}$$

$$E[X^2] = \sum_i i^2 p(i)$$

$$= \int x^2 f(x) dx$$

$$E[5X] = 5 E[X]$$

$$E[\ln X] = \int \ln x f(x) dx$$

$$= \sum_i \ln i p(i)$$

~~So~~

$$E[g(X)] = \sum_i g(i) p(i)$$

$$= \int g(x) f(x) dx$$

$$E[X^2]$$

$$E[X^3]$$

$$E[X^4]$$

moments

in product over all  
moment generating function

$$M_1$$

## Moment Generating Function

$$\phi(t) = E[e^{tX}]$$

$$\begin{aligned}\phi(t) &= \int_0^{\infty} e^{tx} \lambda e^{-\lambda x} dx \\ &= \int_0^{\infty} \lambda e^{-(\lambda-t)x} dt \\ &\Rightarrow \frac{\lambda}{\lambda-t}\end{aligned}$$

$$\begin{aligned}\phi'(t) &= E\left[\frac{d}{dt} e^{tx}\right] \\ &= E[X e^{tx}]\end{aligned}$$

$$\phi''(t) = E[X^2 e^{tx}]$$

$$\phi'(0) = E[X]$$

$$\phi''(0) = E[X^2]$$

$$\phi'''(0) = E[X^3]$$

# MGF Moment Generating Function

01/03/14

MGF's property:

$$\Phi(t) \rightarrow E[e^{tx}]$$

$$E(X) = \Phi'(0)$$

$$E(X^2) = \Phi''(0); E(X^3) = \Phi'''(0)$$

①  $\Phi_{X+Y}(t) = \Phi_X(t) \cdot \Phi_Y(t)$ ;  $X$  and  $Y$  are

② MGF uniquely identifies a distribution.

$$Z \rightarrow \Phi_Z(t) = \frac{s}{s-t}$$

we know, if exponent.

$$\text{poisson } (\lambda), \Phi(t) = \frac{\lambda}{\lambda - t}$$

? (check)

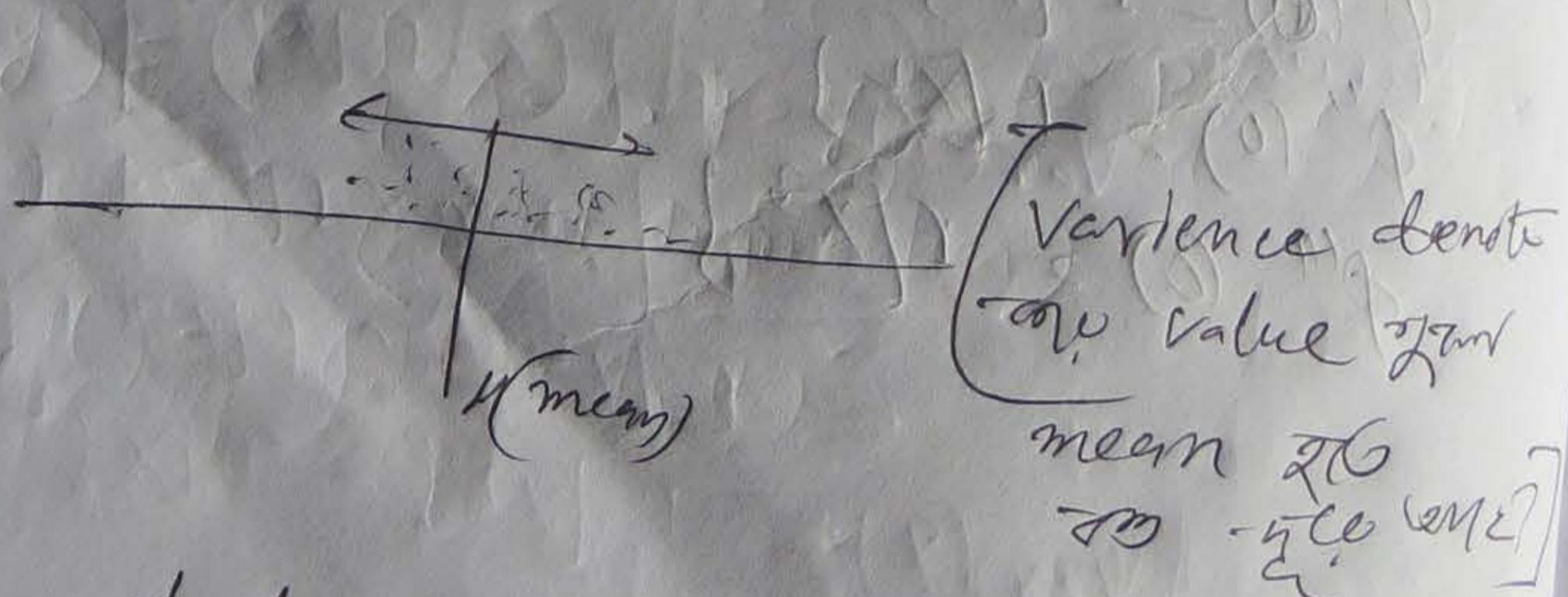
So,  $Z$  is exponential with  $\lambda = s$ .

$$\sigma^2 = \text{Var}(X) = E[(X-\mu)^2] = (E[X^2] - E[X]^2)$$

$\sigma^2$  = Variance

$-1 \leq \text{Corr}(X, Y) \leq 1$

Correlation



Only  $\sigma$  = standard deviation

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

degree  
of  
depender  
ency

$$= E[XY] - (E[X]E[Y])$$

since  $Y > X \text{ const}$   
 Var  $\approx$  formula

$$\text{Cov}(X, X) = \text{Var}(X)$$

( $X, Y$  ~~not~~ same thing ~~not~~ dependent)

\* If  $X$  &  $Y$  are independent,

$$\text{Cov}(X, Y) = 0.$$

$$-1 \leq \text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \cdot \text{Var}(Y)}} \leq +1$$

[correlation]

$$\text{Corr}(X, Y) = +1 \text{ if }$$

if  $X > \mu_X$  then  $Y > \mu_Y$

## property:

$$\textcircled{1} \quad \text{Cov}(X, Y) = \text{Cov}(Y, X)$$

$$\textcircled{2} \quad \text{Cov}(X, X) = \text{Var}(X)$$

~~proof~~  $\textcircled{3} \quad \text{Cov}(X, Y+Z) = \text{Cov}(X, Y) + \text{Cov}(X, Z)$

$$\textcircled{4} \quad \text{Cov}(cX, Y) = c \text{Cov}(X, Y)$$

#  $\text{Var}(X+Y) \neq \text{Var}(X) + \text{Var}(Y)$

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

what we have seen so far

$\Rightarrow$  if  $X$  &  $Y$  are independent

$$\textcircled{1} \quad \text{Cov}(X, Y) = 0.$$

$$\textcircled{2} \quad \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$$

PMF,  $P(x, y) = P\{X=x, Y=y\} = P\{X=x\} \cdot P\{Y=y\}$  So,

$$\textcircled{3} \quad P_{X+Y}(t) = \text{joint distribution}$$

$$\text{Ansatz} - P_{X+Y}(t) = \varphi_X(t) \cdot \varphi_Y(t)$$

Proof

$$\underline{X, \mu = E[X], \sigma^2}$$

$$\mu = 14$$

$$\boxed{P(\text{Score} = 18)} \text{ Ans. } \text{or } m \text{ or }$$

$$\text{Q: } P(\text{Score} \geq 18) \text{ or } m \text{ or } 2m \text{ or }$$

(Markov Inequality):

$X$  is a non-negative RV (Random Variable),

for any  $a > 0$ ,

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

$$\text{So, } P\{X \geq 18\} \leq \frac{14}{18}$$

$$\begin{aligned}\text{Proof: } E[X] &= \int_0^\infty x f(x) dx \\ &= \int_0^a x f(a) dx + \int_a^\infty x f(x) dx \\ &\geq \int_a^\infty x \cdot 0 dx \geq a f(a)\end{aligned}$$

$$= a \int_a^\infty f(x) dx \geq a \cdot P(X \geq a)$$

$\underbrace{\quad}_{P(a \geq a)}$

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

$$P\{X < a\} = 1 - P\{X \geq 18\}$$

# Chebyshov's Inequality

$X$  with mean  $\mu, \sigma^2$  kg

$$P\{|(X-\mu)| \geq k\} \leq \frac{\sigma^2}{k^2}$$

Proof: Let,  $Y = (X-\mu)^2$ ,  $Y$  is non-neg

$$P\{Y \geq k^2\} \leq \frac{E[Y]}{k^2}$$

markov inequality

$$P\{(X-\mu)^2 \geq k^2\} \leq \frac{E(Y)}{k^2}$$

$$\therefore \{ |X-\mu| \geq k \} \leq \frac{\sigma^2}{k^2}$$

Cond

301

2/3/14

## Conditional Probability & expectation

$$P(E|F) = \frac{P(E \cap F)}{P(F)} \text{ for } P(F) > 0.$$

If E & F are independent,

$$P(E|F) = P(E)$$

$$P(E \cap F) = P(E) P(F)$$

## Conditional Mass function

$$p_x(x) = P\{X=x\}$$

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

$$= \frac{P\{X=x, Y=y\}}{P\{Y=y\}}$$

$$= \frac{p(x,y)}{p_y(y)}$$

$$f_{x|y}(x|y) = f(x|y=y)$$

Conditional CDF,  $F_{x|y}(x|y) = P\{X \leq x | Y=y\}$

$$= \sum_{a \leq x} p_{x|y}(a|y)$$

(discrete)

$$f_{X|Y}(x|y) = \int_0^x f_{XY}(t|y) dt$$

(continuous case)

### Conditional Expectation

$$E[X] = \sum_x x p(x)$$

$$E[X|Y=y] = \sum_x x p_{x|y}(x|y)$$

Ex:

$$X = 1, 2$$

$$Y = 1, 2$$

$$P(1,1) = 0.5$$

$$P(2,1) = 0.1$$

$$P(1,2) = 0.1$$

$$P(2,2) = 0.3$$

mass function

$$P_{x|y}(x|y_1) = \frac{P\{X=x, Y=1\}}{P\{Y=1\}}$$

$$\frac{P(1,1)}{P(Y=1)}$$

$$\frac{P(1,1)}{0.6}$$

Now,

$$P_{Y=1} = P\{X=1, Y=1\} + P\{X=2, Y=1\}$$
$$= p(1,1) + p(2,1)$$

$$\begin{aligned} P(Y=1) &= 0.5 + 0.1 \\ &= 0.6 \end{aligned}$$

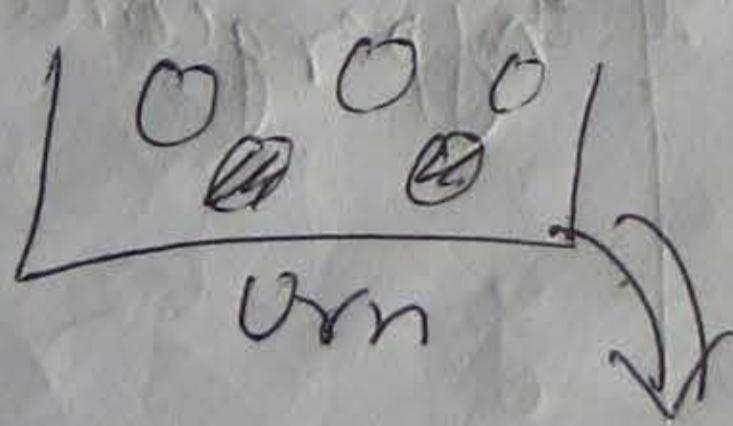
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$$P_{X=1|Y=1} = \frac{P(1,1)}{0.6} = \frac{0.5}{0.6}$$

$$P_{X=2|Y=1} = \frac{P(2,1)}{0.6} = \frac{0.1}{0.6}$$

Ans

Ex



O: red ( $n_1$ )

O: blue ( $n_2$ )

pick  $m$  balls ( $m < n_1 + n_2$ )

$X = \#$  red balls in these  $m$  balls.

Q: What is the distribution of  $X$ ?

$Y = \#$  blue balls in  $m$  balls.

$$P\{X=k \mid X+Y=m\}$$

$$= \frac{P\{X=k, X+Y=m\}}{P\{X+Y=m\}}$$

$$\geq \frac{P\{X=k, Y=m-k\}}{P\{X+Y=m\}}$$

$$= \frac{P\{X=k\} P\{Y=m-k\}}{P\{X+Y=m\}}$$

$$= \frac{\binom{n}{k} p^k (1-p)^{n-k} \binom{n_2}{n-k} (1-p)^{n_2-k} |_{n_2=n-k}}{P\{X+Y=m\} p}$$

$$\geq \frac{\binom{n_1}{k} \binom{n_2}{m-k}}{\binom{n_1+n_2}{m}}$$

hypergeometric

$$\begin{aligned}
 P(X+Y=m) &= \sum_{k=0}^m P(X=k, Y=m-k) \\
 &= \sum_{k=0}^m P(X=k) \underbrace{P(Y=m-k)}_{\text{binomial}}
 \end{aligned}$$

$$\begin{aligned}
 P(X+Y=m) &= \sum_{k=0}^m \binom{n_1}{k} \binom{n_2}{m-k} \\
 &\Rightarrow \binom{n_1+n_2}{m} \quad \text{par pour}
 \end{aligned}$$

$$\begin{aligned}
 P\{X=k \mid X+Y=m\} &\\
 &\downarrow \quad \diagdown \\
 \text{bin}(n, p) \quad \text{bin}(n, p) & \\
 \text{poisson} \quad \text{poisson} & \\
 \exp(\lambda) &
 \end{aligned}$$

$E[X] = \text{value}$

$E[X|Y=y] = \text{value}$

$E[X|Y] = \text{a function of } Y$

Important!

proof

$$\begin{aligned} E[E[X|Y]] &= \sum_y g(y) \cdot p(y) \\ &= \sum_y E[X|Y=y] \cdot p(Y=y) \\ &= \sum_y \sum_x x \cdot P(X=x|Y=y) \cdot p(Y=y) \\ &\quad \cancel{P(X=x|Y=y) \cdot p(Y=y)} \\ &= \sum_y \sum_x x \cdot P\{X=x, Y=y\} \\ &= \sum_y x \sum_y P\{X=x, Y=y\} \\ &\hookrightarrow \sum_x x \cdot P\{X=x\} \\ &= E[X] \end{aligned}$$