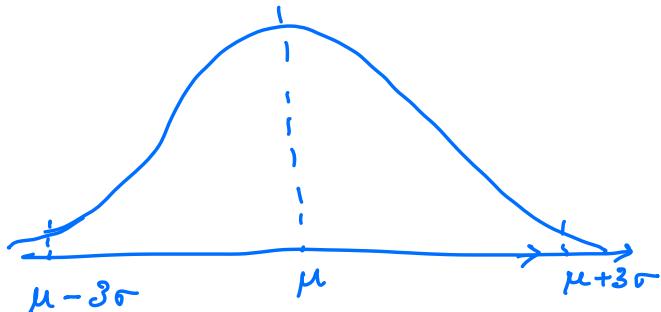


Last class :-  
 Discrete r.v  
 Continuous r.v

Gaussian / Normal random variable

$$f_x(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/2\sigma^2}; -\infty < x < +\infty$$

$$X \sim N(\mu, \sigma^2)$$



Property:

$$\text{If } X \sim N(\mu, \sigma^2)$$

$$Y = aX + b$$

$$Y \sim N(a\mu + b, a^2\sigma^2)$$

$$\begin{aligned} \text{Proof: } F_Y(x) &= P(Y \leq x) \\ &= P(ax + b \leq x) \\ &= P\left(X \leq \frac{x-b}{a}\right) \\ &= F_X\left(\frac{x-b}{a}\right) \end{aligned}$$

$$\begin{aligned} &= \int_{-\infty}^{\frac{x-b}{a}} f_X(x) dx \\ &= \int_{-\infty}^{\frac{x-b}{a}} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/2\sigma^2} dx \end{aligned}$$

$$\begin{aligned} v &= ax + b \\ &= \int_{-\infty}^{\frac{v-b}{a}} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/2\sigma^2} \frac{dv}{a} \end{aligned}$$

$$\begin{aligned}
 &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(\alpha\sigma)^2}} e^{-(v-b-\alpha\mu)^2/2(\alpha\sigma)^2} dv \\
 &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi(\alpha\sigma)^2}} e^{-(v-(\alpha\mu+b))/2(\alpha\sigma)^2} dv \\
 Y \sim N(\alpha\mu+b, \alpha^2\sigma^2)
 \end{aligned}$$

Expectation of random variable.

Discrete random variable

$$\mathbb{E}(X) = \sum x p(x)$$

"Weighting the possible values of the r.v by the probability of taking them"

Example: Expectation of Bernoulli r.v.

$$\begin{aligned}
 P_X(1) &= p & P_X(0) &= 1-p \\
 \mathbb{E}(X) &= 1 \cdot p + 0 \cdot (1-p) \\
 &= p
 \end{aligned}$$

Exercise: (i)  $X \sim \text{Binomial}(n, p)$

$$\begin{aligned}
 \mathbb{E}[X] &= np. & \text{Chap 5} \\
 \text{(iC)} \quad X &\sim \text{Geometric}(p) & \text{or PP}
 \end{aligned}$$

$$\begin{aligned}
 \mathbb{E}(X) &= \frac{1}{p} \\
 \text{(iC)} \quad X &\sim \text{Poisson}(\lambda)
 \end{aligned}$$

$$\mathbb{E}(X) = \frac{1}{\lambda}$$

Example: Contestant in quiz

2 questions  $V_1$  &  $V_2$   
 $\downarrow$        $\downarrow$   
200\$      100\$

Contestant is 60% certain of the answer of  $V_1$   
80% " " " " " of  $V_2$

Attempt the qns in any order

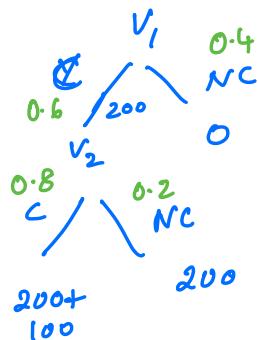
Allowed to move to the 2<sup>nd</sup> qn only if  
the first qn is correct.

What order should he attempt to maximize  
the expected reward

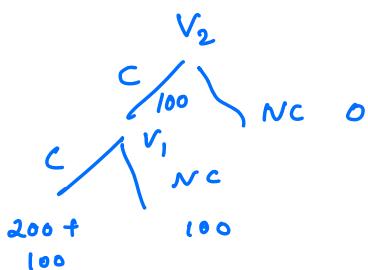
$(V_1, V_2)$

$$P = \begin{cases} 0 & 0.4 \\ 200 & 0.6 \times 0.2 \\ 200+100 & 0.6 \times 0.8 \end{cases}$$

$$E(V_1, V_2) = 0 \times 0.4 + 200 \times 0.12 + 300 \times 0.48 = 168$$



$$(V_2, V_1) = \begin{cases} 0 & 0.2 \\ 100 & 0.32 \\ 300 & 0.48 \end{cases}$$



$$E(V_2, V_1) = 0 \times 0.2 + 100 \times 0.32 + 300 \times 0.48 = 156 // (176)$$

Expectation of continuous r.v

$$\mathbb{E}(X) = \int_{-\infty}^{\infty} x f_X(x) dx$$

Example: Expectation of an exponential r.v  
 $X \sim \text{Exp}(\lambda)$

$$\begin{aligned} f_X(x) &= \lambda e^{-\lambda x} & x > 0 \\ \mathbb{E}(X) &= \int_0^{\infty} x \lambda e^{-\lambda x} dx \\ &= \lambda \int_0^{\infty} x e^{-\lambda x} dx \\ &= \lambda \left[ x \left. \frac{e^{-\lambda x}}{-\lambda} \right|_0^{\infty} - \int_0^{\infty} \left. \frac{e^{-\lambda x}}{-\lambda} \right. dx \right] \\ &= \int_0^{\infty} e^{-\lambda x} dx \\ &= \left. \frac{e^{-\lambda x}}{-\lambda} \right|_0^{\infty} = \frac{1}{\lambda} // \end{aligned}$$

Exercise: 1.  $X \sim \text{Unif}([\alpha, \beta])$   
 $\mathbb{E}(X)$

2.  $X \sim N(\mu, \sigma^2)$

$$\mathbb{E}(X) = \mu.$$

↓  
"Mean"

Fundamental theorem of expectation

$$\mathbb{E}[g(x)] = \sum g(x) P_x(x) dx.$$

$$\int g(x) f_X(x) dx$$

Example: Let  $X$  be Unit  $[0, 1]$  Calculate  $E[X^3]$

$$\begin{aligned}g(x) &= x^3 \\E[X^3] &= \int_0^1 x^3 \cdot 1 \cdot dx \\&= \frac{x^4}{4} \Big|_0^1 = \frac{1}{4}\end{aligned}$$

Example: If you are  $s$  minutes early to an appointment, then you incur a cost  $c \cdot s$

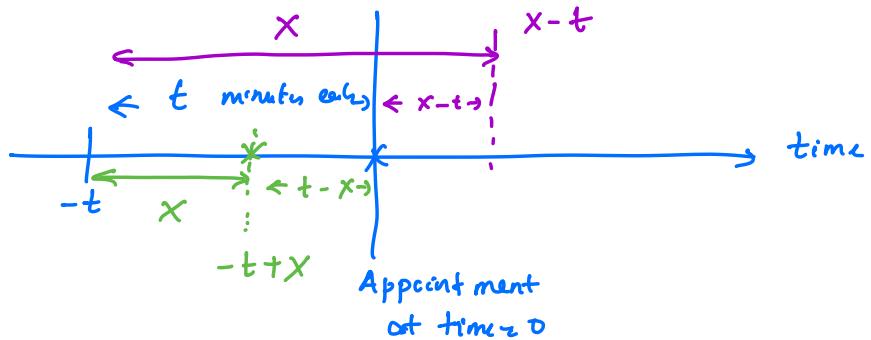
If you are  $s$  minutes late to an appointment, then you incur a cost of  $k \cdot s$

Travel time to this appointment is a random variable with p.d.f  $f$

$$f(t) = 0 \quad t < 0$$

$$f(t) = 0 \quad t > T$$

Determine the time at which you should depart to minimize expected cost?



$$\text{Cost} = \begin{cases} c(t-x) & x \leq t \\ k(x-t) & x > t \end{cases}$$

$$\mathbb{E}[\text{Cost}] = \int_0^t c(t-x) f(x) dx +$$

↑

$$\int_t^+ k(x-t) f(x) dx \quad \sim \quad I_2$$

Fundamental  
Thm of  
Expectation.

$$\begin{aligned} I_1 &= \int_0^t c(t-x) f(x) dx \\ &= c \left[ t \int_0^+ f(x) dx - \int_0^t x f(x) dx \right] \\ &= c \left[ t F(t) - x F(x) \Big|_0^t + \int_0^t F(x) dx \right] \\ &= c \int_0^t F(x) dx \end{aligned}$$

$$I_2 = \int_t^+ k(x-t) f(x) dx$$

[Exercise]

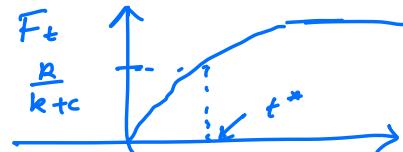
$$\frac{d}{dt} \mathbb{E}[\text{Cost}] = 0$$

$$= c F(t) - k + k F(t)$$

At optimal  $t = t^*$

$$c F(t^*) - k + k F(t^*) = 0$$

$$F(t^*) = \frac{k}{k+c}$$



$$\begin{aligned}
 &= k \left[ \int_t^T x f(x) dx - t \int_t^T f(x) dx \right] \\
 &= k \left[ x F(x) \Big|_t^T - \int_t^T F(x) dx \right. \\
 &\quad \left. - t [F(T) - F(t)] \right] \\
 &= k \left[ T F(T) - t \cancel{F(t)} - \int_t^T F(x) dx \right. \\
 &\quad \left. - t F(T) + t \cancel{F(t)} \right] \\
 &= k \left[ T - t - \int_t^T F(x) dx \right]
 \end{aligned}$$

$$\frac{d}{dt} I_1 = c F(t)$$

$$\frac{d}{dt} I_2 = k [-1 + F(t)]$$