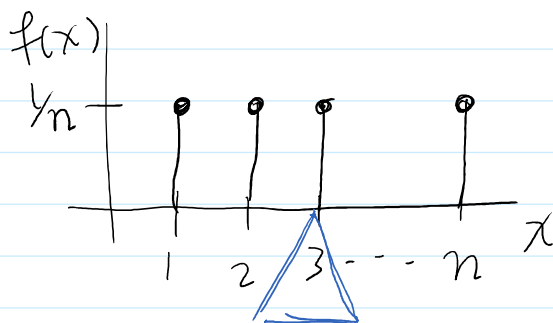
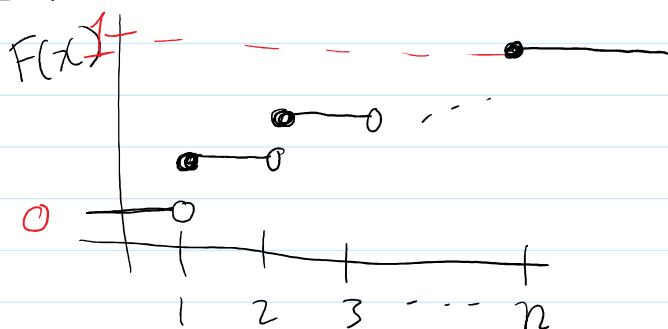


Discrete Uniform

$$X \sim U(\{1, \dots, n\})$$

$$f(x) = \frac{1}{n} \text{ for } x = 1, 2, 3, \dots, n$$

CDF :

$$F(x) = \begin{cases} 0, & x < 1 \\ \lfloor x \rfloor / n & 1 \leq x < n \\ 1 & x \geq n \end{cases}$$

$$\sum_{i=1}^n i = \frac{n(n+1)}{2} *$$

$$\sum_{i=1}^n i^2 = \frac{n(n+1)(2n+1)}{6}$$

Expectation :

$$E[X] = \sum_{x=1}^n x \left(\frac{1}{n} \right) = \frac{1}{n} \sum_{x=1}^n x = \frac{1}{n} \frac{n(n+1)}{2} = \boxed{\frac{n+1}{2}}$$

$$\begin{aligned} E[X^2] &= \sum_{x=1}^n x^2 \left(\frac{1}{n} \right) = \frac{1}{n} \sum_{x=1}^n x^2 = \frac{1}{n} \frac{n(n+1)(2n+1)}{6} \\ &= \frac{(n+1)(2n+1)}{6} \end{aligned}$$

$$= \frac{(n+1)(2n+1)}{6}$$

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

$$= \frac{(n+1)(2n+1)}{6} - \left(\frac{n+1}{2}\right)^2$$

$$= \dots = \boxed{\frac{n^2 - 1}{12}}$$

MGF:

$$M(t) = E[e^{tx}] = \sum_{x=1}^n e^{tx} \left(\frac{1}{n}\right) = \frac{1}{n} \sum_{x=1}^n (e^t)^x$$

Geometric Sum:

$$\sum_{i=0}^{n-1} r^i = \frac{1-r^n}{1-r} \quad \text{for } r \neq 1$$

→ $r = e^t$

$$\frac{1}{n} \sum_{x=1}^n r^x = \frac{1}{n} \sum_{x=0}^{n-1} r^{x+1} = \frac{r}{n} \sum_{x=0}^{n-1} r^x = \frac{r}{n} \frac{1-r^n}{1-r}$$

$$= \frac{e^t (1-(e^t)^n)}{n(1-e^t)}$$

$$= \underline{e^t - e^{t(n+1)}}$$

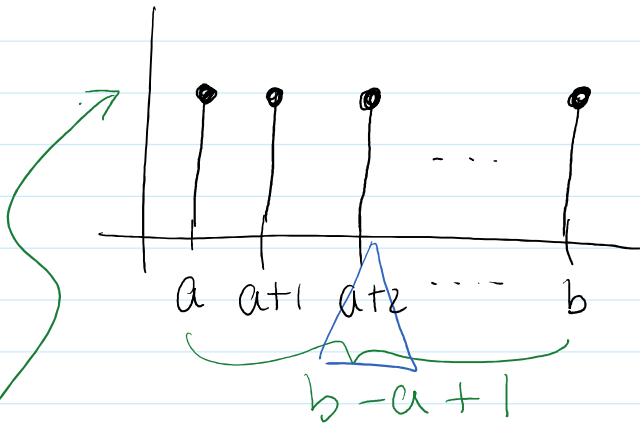
$$= \frac{e^t - e^{t(n+1)}}{n(1-e^t)}$$

Consider $Y \sim U(\{a, \dots, b\})$

let $n = b - a + 1$

then $Y = X + (a-1)$

linear transformation



$$f(y) = \frac{1}{b-a+1} \text{ for } y = a, \dots, b$$

$$\begin{aligned} \mathbb{E}[Y] &= \mathbb{E}[X + (a-1)] = \mathbb{E}[X] + (a-1) \\ &= \frac{n+1}{2} + (a-1) \\ &= \frac{(b-a+1)+1}{2} + (a-1) \\ &= \dots = \frac{a+b}{2} \end{aligned}$$

$$\text{Var}(Y) = \text{Var}(X + a-1)$$

$$\text{Var}(Y) = n^2 - 1$$

$$= \text{Var}(X) = \frac{n^2 - 1}{12} = \frac{(b-a+1)^2 - 1}{12}$$

$$Y = X + a - 1$$

$$M_Y(t) = e^{(a-1)t} M_X(t)$$

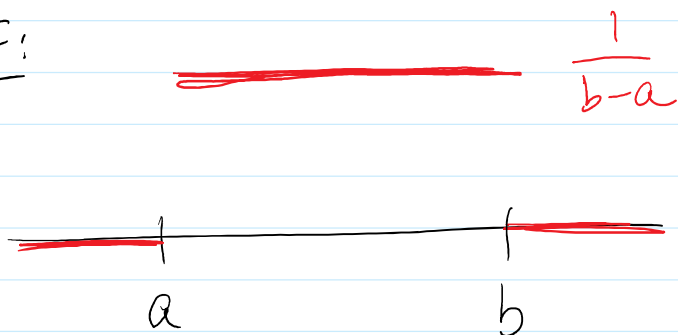
$$= e^{(a-1)t} \frac{e^t - e^{t(n+1)}}{n(1-e^t)}$$

$$= \frac{e^{at} - e^{t(b+1)}}{(b-a+1)(1-e^t)}$$

Continuous Uniform

$$X \sim U(a, b)$$

PDF:



$$f(x) = \frac{1}{b-a} \text{ for } a < x < b$$

CDF:

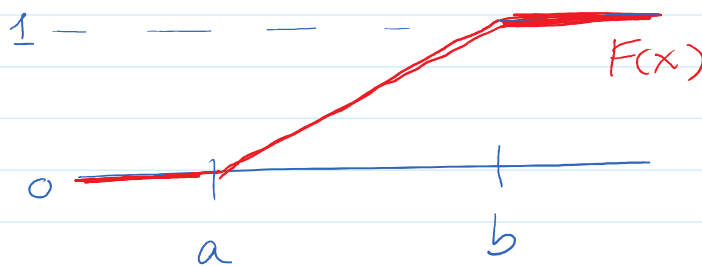
$$F(x) = \int_{-\infty}^x f(t) dt = \int_a^x \frac{1}{b-a} dt = \frac{t}{b-a} \Big|_a^x$$

$$= \frac{x-a}{b-a}$$

if $x < a$

$$F(x) = 0$$

if $x \geq b$ then $F(x) = 1$



Expectation

$$E[X] = \int_{\mathbb{R}} x f(x) dx = \int_a^b x \frac{1}{b-a} dx$$

\mathbb{R}

$\frac{1}{a} \quad b-a$

$$= \frac{x^2}{2(b-a)} \Big|_a^b = \frac{b^2 - a^2}{2(b-a)} = \frac{(b-a)(b+a)}{2(b-a)}$$

$$\boxed{= \frac{a+b}{2}}$$

$$E[X^2] = \int_a^b x^2 \frac{1}{b-a} dx = \frac{x^3}{3(b-a)} \Big|_a^b$$

$$= \frac{b^3 - a^3}{3(b-a)} = \frac{(b-a)(a^2 + ab + b^2)}{3(b-a)}$$

$$= \frac{a^2 + ab + b^2}{3}$$

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

$$= \frac{b^2 + ab + a^2}{3} - \left(\frac{a+b}{2}\right)^2$$

$$= \dots = \boxed{\frac{(b-a)^2}{12}}$$

$$= \frac{b^2 + ab + a^2}{3} - \left(\frac{a+b}{2}\right)^2$$

$$= \dots = \boxed{\frac{(b-a)^2}{12}}$$

MGF:

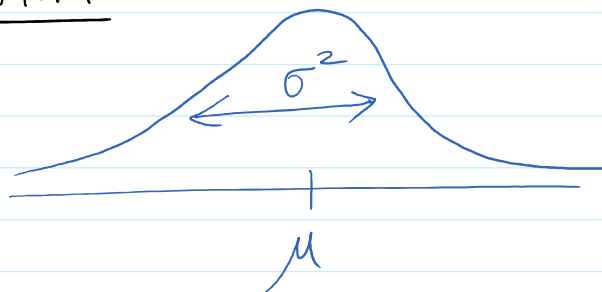
$$M(t) = E[e^{tx}] = \int_a^b e^{tx} \frac{1}{b-a} dx$$

$$= \frac{e^{tx}}{t(b-a)} \Big|_a^b = \boxed{\frac{e^{bt} - e^{at}}{t(b-a)}}$$

Normal / Gaussian Distribution

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

$$\mu \in \mathbb{R}, \sigma^2 > 0$$



PDF:

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right) \quad \forall x \in \mathbb{R}$$

CDF: $F(x) = \int_{-\infty}^x f(t) dt$ no simple closed form for this

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Claim: $E[X] = \mu$ and $\text{Var}(X) = \sigma^2$.

MGF:

$$e^a e^b = e^{a+b}$$

$$M(t) = E[e^{tx}] = \int_{\mathbb{R}} \overbrace{e^{tx}}^{\text{from } e^a e^b} \overbrace{\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2}(x-\mu)^2\right)}^{\text{from } e^a e^b} dx \quad (*)$$

Combine exponents:

$$tx - \frac{1}{2\sigma^2}(x-\mu)^2$$

$$= tx - \frac{1}{2\sigma^2}(x^2 - 2\mu x + \mu^2)$$

$$= -\frac{1}{2\sigma^2} \left(\underbrace{-2\sigma^2 tx}_{\text{from } e^a} + x^2 - \underbrace{2\mu x}_{\text{from } e^b} + \mu^2 \right)$$

looks like
first two
terms of

$$= -\frac{1}{2\sigma^2} \left(\underbrace{x^2 - 2x(\mu + \sigma^2 t)}_{\text{from } e^a} + \mu^2 \right)$$

$$\underbrace{(x - (\mu + \sigma^2 t))^2}_{\text{from } e^b}$$

$$= -\frac{1}{2\sigma^2} \left(\underbrace{x^2 - 2x(\mu + \sigma^2 t) + (\mu + \sigma^2 t)^2}_{(x - (\mu + \sigma^2 t))^2} - (\mu + \sigma^2 t)^2 + \mu^2 \right)$$

$$(x - (\mu + \sigma^2 t))^2$$

$$= -\frac{1}{2\sigma^2} \left(\underbrace{[x - (\mu + \sigma^2 t)]^2}_{\text{PDF of } N(\mu + \sigma^2 t, \sigma^2)} - \underbrace{(\mu + \sigma^2 t)^2 + \mu^2}_{\text{doesn't depend on } x} \right)$$

$$e^{a+b} = e^a e^b$$

$$(*) = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} \exp(\quad) dx$$

$$= \int_{-\infty}^{\infty} \underbrace{\frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{1}{2\sigma^2} [x - (\mu + \sigma^2 t)]^2\right)}_{\text{PDF of } N(\mu + \sigma^2 t, \sigma^2)} \underbrace{\exp\left(-\frac{1}{2\sigma^2} (-(\mu + \sigma^2 t)^2 + \mu^2)\right)}_{\text{doesn't depend on } x} dx$$

← integrates to 1

$$= \exp\left(-\frac{1}{2\sigma^2} (-(\mu + \sigma^2 t)^2 + \mu^2)\right) \cdot 1$$

$$= \dots = \boxed{\exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right) = M(t)}$$

$$E[X] = \left. \frac{dM}{dt} \right|_{t=0} = \underbrace{(\mu + \sigma^2 t)}_{\text{at } t=0} \underbrace{\exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right)}_{\text{at } t=0} \Big|_{t=0}$$

$$= \mu$$

$$\begin{aligned} E[X^2] &= \left. \frac{d^2 M}{dt^2} \right|_{t=0} = \sigma^2 \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right) + (\mu + \sigma^2 t)^2 \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right) \\ &= \sigma^2(1) + (\mu + 0)^2(1) \\ &= \sigma^2 + \mu^2 \end{aligned}$$

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

Theorem: Linear Transformations of Normal RVs

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

$$Y = aX + b$$

then $Y \sim N(\underline{a\mu + b}, \underline{a^2\sigma^2})$

intuition: $E[Y] = E[aX + b] = aE[X] + b = \underline{a\mu + b}$

$$\text{Var}(Y) = \text{Var}(aX + b) = a^2 \text{Var}(X) = \underline{a^2\sigma^2}$$

pf. Recall $M_X(t) = \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right)$

we also have a theorem that says

$$M_Y(t) = e^{bt} M_X(at)$$

$$= e^{bt} \exp\left(\mu at + \frac{\sigma^2 a^2 t^2}{2}\right)$$

$$= \exp\left(\underbrace{(a\mu + b)}_{\text{red wavy}} t + \frac{\underbrace{(a^2 \sigma^2)}_{\text{red wavy}} t^2}{2}\right)$$

↖ this is just the MGF of
a $N(a\mu + b, a^2 \sigma^2)$
