

Next Brand Name cont. r.v.

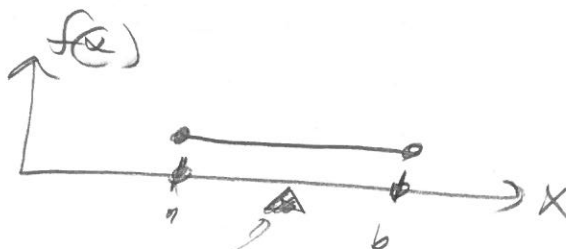
$$X \sim U(a, b) := \frac{1}{b-a} \quad \text{supp}(X) = [a, b]$$

Uniform

Param space:
 $a < b \quad a, b \in \mathbb{R}$

Cont. analogue of discrete uniform
 all #'s between a, b are equally likely

Is it a r.v.? $\int_{x \in \text{supp}(X)} f(x) dx = 1$?



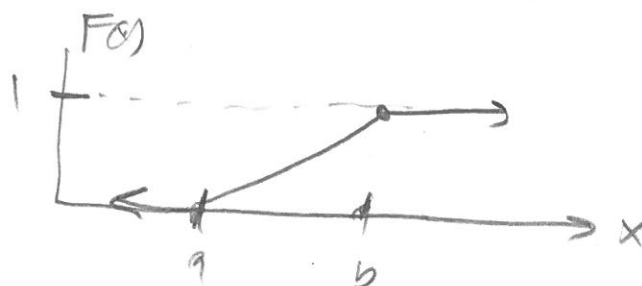
$$\int_a^b \frac{1}{b-a} dx = \frac{1}{b-a} [x]_a^b = \frac{b-a}{b-a} = 1 \checkmark$$

$$F(x) = \int f(x) dx = \frac{x}{b-a} + C$$

Solve for $C \dots F(a) = 0$

$$\Rightarrow \frac{a}{b-a} + C = 0 \Rightarrow C = -\frac{a}{b-a}$$

$$\Rightarrow F(x) = \frac{x}{b-a} - \frac{a}{b-a} = \frac{x-a}{b-a} \checkmark$$



$X \sim U(0, 1)$ "the std uniform"

give me a random # between 0, 1, most important to CS. (Math, Random(); Java)

$f(x) = 1, F(x) = x$ super simple!

$$m := E[X] = \int_{\text{supp}(X)} x f(x) dx = \int_a^b x \frac{1}{b-a} dx = \frac{1}{b-a} \left(\frac{x^2}{2} \right)_a^b = \frac{1}{b-a} \left(\frac{b^2 - a^2}{2} \right) = \frac{(b-a)(b+a)}{2(b-a)} = \frac{b+a}{2} \checkmark$$

$$\begin{aligned} \text{Var}(X) &= \int (x-m)^2 f(x) dx = E[X^2] - m^2 = \int_a^b x^2 f(x) dx - \left(\frac{b+a}{2} \right)^2 \\ &= \frac{1}{b-a} \left(\frac{x^3}{3} \right)_a^b - \left(\frac{b+a}{2} \right)^2 \\ &= \frac{b^3 - a^3}{3(b-a)} - \frac{b^2 + 2ab + a^2}{4} = \frac{b^2 - ab + a^2}{12} \end{aligned}$$

$$\frac{b^3 - a^3}{b - a} = ?$$

polynomial long

division procedure:

$$\begin{array}{r} b^2 + a^2 + ab \\ b - a \overline{) b^3 - a^3} \\ \underline{b^3 - a^3} \\ 0 \end{array}$$

good skill to have!!

$$\begin{array}{r} -a^3 + ab^2 \\ (1a^2) - a^3 + ab^2 \\ \underline{-a^3 + ab^2} \\ ab^2 - a^2b \\ \underline{ab^2 - a^2b} \\ 0 \end{array}$$

Check this...

$$(b-a)(b^2 + a^2 + ab) = b^3 - ab^2 + ab^2 - a^3 + ab^2 - a^2b + a^2b - a^3 = b^3 - a^3$$

$$\frac{b^2 + a^2 + ab}{3} - \frac{b^2 + 2ab + a^2}{4}$$

$$= \frac{4b^2 + 4a^2 + 4ab - 3b^2 - 6ab - 3a^2}{12}$$

$$= \frac{b^2 + a^2 - 2ab}{12} = \frac{(b-a)^2}{12} = \frac{b-a}{12} \cdot \frac{b-a}{1} = \frac{b-a}{12} \cdot \frac{b-a}{1} = \frac{(b-a)^2}{12}$$

Let's find the v.v. of std. normal

$$Z \sim N(0,1) := \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \quad \text{Supp}(Z) = \mathbb{R}$$

Is this a PDF? $\int_{-\infty}^{\infty} > 0$ \checkmark

everywhere!!!



$$\int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = 1? \quad \int_{\mathbb{R}} e^{-\frac{x^2}{2}} dx = \sqrt{2\pi}$$

$$\text{Let } u = \frac{x^2}{2} \quad \frac{du}{dx} = \frac{1}{\sqrt{2}}$$

$$\int_{\mathbb{R}} e^{-u^2} \sqrt{2} du = \sqrt{2\pi} \quad \int_{\mathbb{R}} e^{-u^2} du = \sqrt{\pi}$$

NOT ON EXAM

Imagine $x^2 + y^2 = u^2$

$$\Rightarrow \iint_{\mathbb{R}^2} e^{-(x^2+y^2)} dx dy = \int_{\mathbb{R}} e^{-x^2} dx \int_{\mathbb{R}} e^{-y^2} dy = \sqrt{\pi} \cdot \sqrt{\pi} = \pi$$

Polar coords...

$$dA = \left| \frac{\partial x}{\partial r} \frac{\partial y}{\partial \theta} - \frac{\partial y}{\partial r} \frac{\partial x}{\partial \theta} \right| dr d\theta = \begin{vmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{vmatrix} = r \cos^2 \theta + r \sin^2 \theta = r$$

$$= \iint_{\theta \in [0, 2\pi]} e^{-r^2} r dr d\theta$$

$\theta \in [0, 2\pi]$ $r \in [0, \infty)$

$$= 2\pi \int_0^{\infty} r e^{-r^2} dr$$

$$\int_0^{\infty} r e^{-r^2} dr = \frac{1}{2} \int_0^{\infty} e^{-u} du = \frac{1}{2}$$

$$= 2\pi \cdot \frac{1}{2} = \pi$$

$$= \pi \int_0^{\infty} e^{-u} du = \pi [-e^{-u}]_0^{\infty} = \pi$$

let $u = -\frac{x^2}{2} \Rightarrow \frac{du}{dx} = -x \Rightarrow dx = -\frac{du}{x}$

$$\mu = E(Z) = \int_{\mathbb{R}} x \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} x e^{-\frac{x^2}{2}} dx = \frac{1}{\sqrt{2\pi}} \int_{u_1}^{u_2} e^u du = -\frac{1}{\sqrt{2\pi}} (e^u)_{u_1}^{u_2} = -\frac{1}{\sqrt{2\pi}} [e^{-\frac{x^2}{2}}]_{-\infty}^{\infty} = -\frac{1}{\sqrt{2\pi}} (0-0) = 0$$

$V_Z(Z) = E(X^2) - \mu^2 = E(X^2) = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} x^2 e^{-\frac{x^2}{2}} dx = 1$

$\Rightarrow SE(Z) = 1 = \sigma$

(Fubini by prob)

Z is called the std. normal and it is a standard v.v.

$F_Z(x) = ?$ Not solvable in closed form!

$F(0) = ?$ 0.5 due to symmetry

A few to remember...

- $P(Z \in [-1, 1]) = F(1) - F(-1) \approx 0.68$
- $P(Z \in [-2, 2]) = F(2) - F(-2) \approx 0.95$
- $P(Z \in [-3, 3]) = F(3) - F(-3) \approx 0.997$

"68-95-99.7 Rule"
or "3σ rule"
"empirical rule"

Remind $X \sim \text{Exp}(\lambda) := \lambda e^{-\lambda x}$

let $Y = 2X$ how is Y distributed? If $X \sim \text{Bin}(n, p)$ $2X$ was not binomial!

$$F_Y(x) = P(Y \leq x) = P(2X \leq x) = P(X \leq \frac{x}{2}) = F_X(\frac{x}{2})$$

$$f_Y(x) = \frac{dF_Y(x)}{dx} = \frac{d}{dx} F_X(\frac{x}{2}) = \frac{1}{2} \frac{d}{du} F_X(u) = \frac{1}{2} f_X(u) = \frac{1}{2} \lambda e^{-\lambda u} = \frac{\lambda}{2} e^{-\lambda \frac{x}{2}} = \text{Exp}(\frac{\lambda}{2})$$

so if $Y = aX$, $X \sim \text{Exp}(\lambda)$, $Y \sim \text{Exp}(\frac{\lambda}{a})$

let $X \sim U(a, b)$, $Y = cX + d \sim U(ca + d, cb + d)$

$$F_Y(x) = P(Y \leq x) = P(cX + d \leq x) = P(X \leq \frac{x-d}{c}) = F_X(\frac{x-d}{c})$$

$$f_Y(x) = \frac{d}{dx} F_Y(x) = \frac{1}{c} \frac{d}{du} F_X(u) = \frac{1}{c} f_X(u) = \frac{1}{c} \frac{1}{b-a}$$

$Z \sim N(0,1)$

From HW & recall this...

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$$X = \sigma Z + \mu$$

$$E(X) = \sigma E(Z) + \mu = \mu$$

$$\text{Var}(X) = \sigma^2 \text{Var}(Z) = \sigma^2 \Rightarrow \text{SE}(X) = \sigma$$

How is X distributed? $\text{Supp}(X) = \mathbb{R}$ still...

$$F_X(x) = P(X \leq x) = P(\sigma Z + \mu \leq x) = P(Z \leq \frac{x-\mu}{\sigma}) = F_Z\left(\frac{x-\mu}{\sigma}\right)$$

$$f_X(x) = \frac{d}{dx} [F_X(x)] = \frac{d}{dx} \left[F_Z\left(\frac{x-\mu}{\sigma}\right) \right] \quad \text{let } u = \frac{x-\mu}{\sigma} \quad \frac{du}{dx} = \frac{1}{\sigma} \Rightarrow dx = \sigma du$$

big density has
smaller loss
↓

$$= \frac{1}{\sigma} \frac{d}{du} F_Z(u) = \frac{1}{\sigma} f_Z(u) = \frac{1}{\sigma} \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2} = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2\sigma^2} (x-\mu)^2}$$

density of general Normal r.v.

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

mean var.

But why is the normal density so important? ???

Wait two lectures.... We need to build some tools to understand the normal density

$$\text{let } L(t) := \int_{\mathbb{R}} e^{-tx} f(x) dx$$

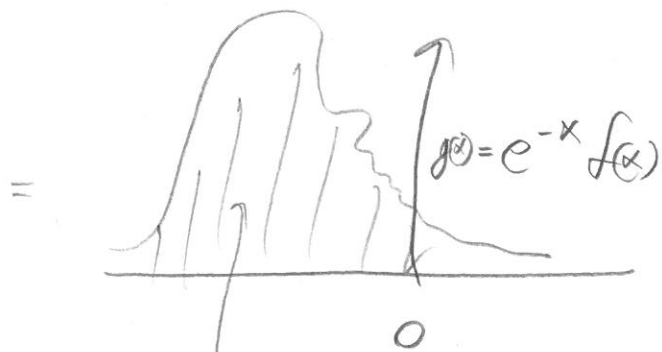
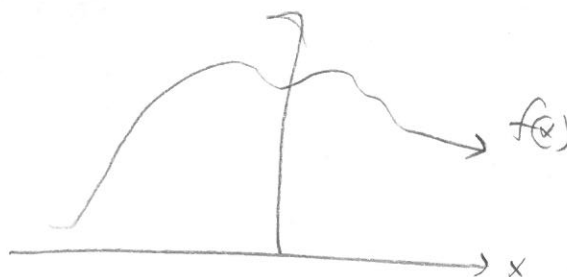
"Bilateral Laplace Transform"
↑
demon

"Integrates x out"

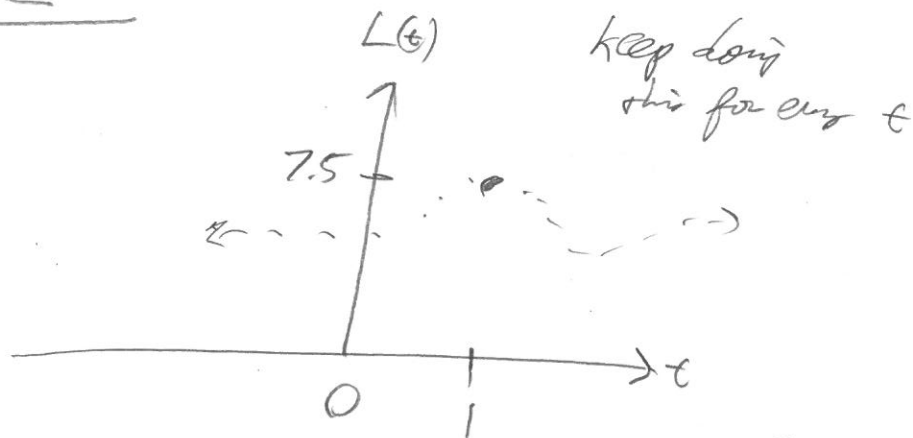
to be left with

a function of t . It's an "operator" on function f .

$$\text{if } f \text{ is } L(f) = \int_{\mathbb{R}} e^{-x} f(x) dx$$



$$\int_0^{\infty} e^{-x} f(x) dx = 7.5$$



turns out $L(t) \Leftrightarrow f(x)$ are 1:1 (if $L(t)$ exists) sometimes $\int = \infty$

If you give me $f(x)$, I give you $L(t)$ which is not shared by any other $f(x)$. " $L(t)$ " " $f(x)$ " " " $L(t)$

Thus $f(x)$ and $L(t)$ are two sides of the same coin.
Looking at the same thing a different way.

Define $M_X(t) := E[e^{tx}] = \int_{x \in \text{supp}(X)} e^{tx} f(x) dx$ if cont. X r.v.,
 \downarrow
 Moment generating function
 (mgf) $= \sum_{x \in \text{supp}(X)} e^{tx} p(x)$ if discrete X r.v.