

Normal Distribution

$$Z \sim N(0, 1)$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

$$f_X(x) \propto e^{-x^2/2} \leftarrow \text{must be normalized}$$

\leftarrow no known elementary variable for this func

Integrate the normalized distr,

$$c \int_{-\infty}^{\infty} e^{-x^2/2} dx = 1 \quad \text{no closed form solution}$$

We'll solve a slightly harder problem $\Rightarrow e^{-\alpha x^2/2}$, & set $\alpha=1$ at the end

$$\text{Trick: } I^2 = \int_{-\infty}^{\infty} e^{-\alpha x^2/2} dx \int_{-\infty}^{\infty} e^{-\alpha y^2/2} dy \quad \text{convert to polar coordinates}$$

$$= \int_0^{2\pi} \int_0^{\infty} e^{-\alpha r^2/2} r dr d\theta$$

$$= \frac{2\pi}{\alpha}$$

$$\Rightarrow I = \sqrt{\frac{2\pi}{\alpha}}$$

$$\text{Set } \alpha=1, f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}, x \in \mathbb{R}$$

CDF of $Z \sim N(0, 1)$

$$\Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

Mean: $\mathbb{E}[X] = 0$ by symmetry

$$\text{Variance: } \mathbb{E}[X^2] = \int x^2 e^{-x^2/2} dx$$

$$= \frac{1}{\sqrt{2\pi}} \int x^2 e^{-\alpha x^2/2} dx$$

works like $(-2) \frac{\partial}{\partial \alpha} e^{-\alpha x^2/2}$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (-2) \frac{\partial}{\partial \alpha} [e^{-\alpha x^2/2}] dx$$

$$= -\frac{2}{\sqrt{2\pi}} \frac{\partial}{\partial \alpha} \left[\int_{-\infty}^{\infty} e^{-\alpha x^2/2} dx \right]$$

$$\underbrace{\int_{-\infty}^{\infty} e^{-\alpha x^2/2} dx}_{\sqrt{2\pi}/2}$$

$$= -\frac{2}{\sqrt{2\pi}} \frac{\partial}{\partial \alpha} [\alpha^{-1/2}]$$

$$= \alpha^{-3/2}$$

$$\Phi(x) = 1 - \Phi(-x) \quad \text{symmetric}$$



Change of Variables

Let x be a RV and $g: \mathbb{R} \rightarrow \mathbb{R}$ is one-to-one. $Y = g(X)$

Let h be inverse of g , i.e. $h = g^{-1}$.

$$\text{Then, } f_Y(y) = f_X(h(y)) |h'(y)|$$

Let's show it for g strictly increasing $\Rightarrow h$ also strictly increasing $\Rightarrow h'(y) > 0 \forall y$.

$$F_Y(y) = P(Y \leq y) = P(g(X) \leq y) = P(X \leq h(y)) = F_X(h(y))$$

$$\Rightarrow f_Y(y) = \frac{d}{dy} F_X(h(y)) = f_X(h(y)) h'(y)$$

If g NOT one-to-one,

$$f_Y(y) = \sum_{x: g(x)=y} f_X(x) |h'(y)|$$

Non-standard Gaussian

$$Y = \mu + \sigma Z, \text{ where } Z \sim N(0,1)$$

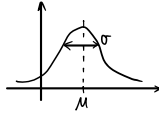
$$f_Z(z) = \frac{1}{\sqrt{2\pi}} e^{-z^2/2}, \quad z \in \mathbb{R}$$

$$F_Y(y) = P(Y \leq y) = P(\mu + \sigma Z \leq y) = P\left(Z \leq \frac{y - \mu}{\sigma}\right) = F_Z\left(\frac{y - \mu}{\sigma}\right)$$

$$\Rightarrow f_Y(y) = f_Z\left(\frac{y - \mu}{\sigma}\right) \frac{1}{\sigma}$$

$$f_Y(y) = \frac{1}{\sqrt{2\pi}} \sigma e^{-\frac{(y - \mu)^2}{2\sigma^2}}$$

$$\left. \begin{array}{l} \mathbb{E}[Y] = \mu \\ \text{Var}(Y) = \sigma^2 \end{array} \right\} Y \sim N(\mu, \sigma^2)$$



$$\text{Ex: } X \sim (\text{annual snowfall}) N(\mu, \sigma^2) \quad \mu = 60", \sigma = 20"$$

$$Q: P(X \geq 80")$$

$$\text{Let } Z = \frac{X - \mu}{\sigma} \text{ (standardize)}$$

$$P(X \geq 80") = P\left(\frac{X - \mu}{\sigma} \geq \frac{80" - \mu}{\sigma}\right)$$

$$= P(Z \geq 1)$$

$$= 1 - \Phi(1)$$

$$= 1 - 0.8413$$

$$= 0.1587$$