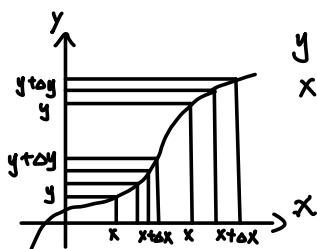


Change of Variable (Transformation)



$$y = h(x)$$

$$x = h^{-1}(y) = g(y)$$

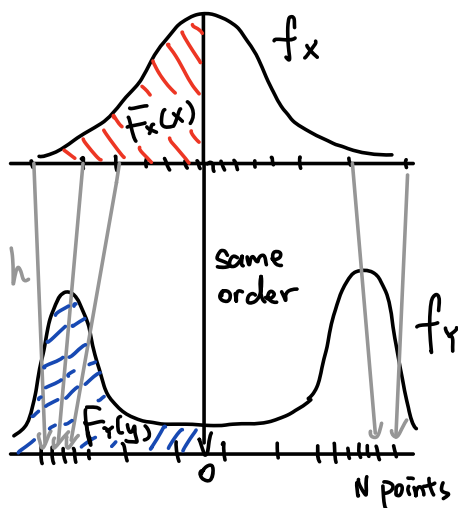
$$X \sim f_X(x) \quad Y \sim f_Y(y)$$

$$P(X \in (x, x+\Delta x)) = P(Y \in (y, y+\Delta y))$$

equivalent statements

$$\Rightarrow f_X(x) \Delta x = f_Y(y) \Delta y$$

$$f_Y(y) = f_X(x) \frac{\Delta x}{\Delta y} = f_X(g(y)) |g'(y)| \quad (x = g(y), \frac{dx}{dy} = g'(y))$$



order preserving

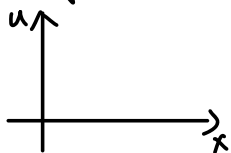
$$P(X \leq x) = P(Y \leq y)$$

$$\boxed{F_X(x) = F_Y(y)}$$

$$y = F_Y^{-1}(F_X(x)) = h(x)$$

Inversion Method :

Change notation

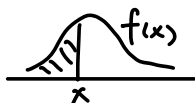
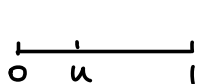


$$U \sim \text{Unif}[0, 1]$$

$$X \sim f(x)$$

$$F(x) = P(X \leq x) \\ = \int_{-\infty}^{\infty} f(x) dx$$

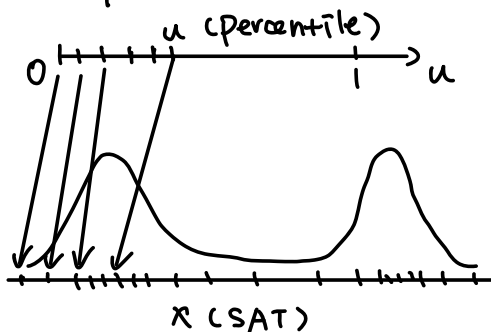
$$\Rightarrow P(U \leq u) = P(X \leq x)$$



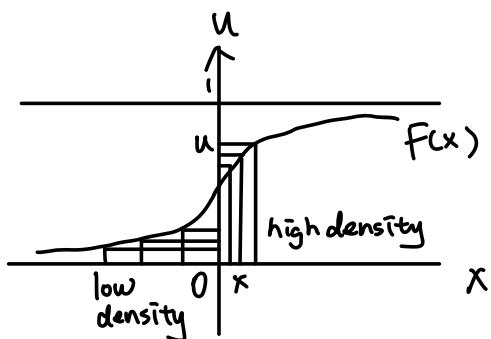
$$u = F(x)$$

$$\Rightarrow x = F^{-1}(u)$$

Ex. percentile $\longleftrightarrow$ SAT score



order preserving



$$\text{density} = \text{slope} = F'(x) = f(x)$$

Goal: generate $X \sim f(x)$

Method:

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

$$\text{generate } U \sim \text{Unif}[0, 1]$$

$$\text{rewrite } X = F^{-1}(U)$$

Back to Normal Distribution

For a R.V. X

$$\mu = E(X) \quad \sigma^2 = \text{Var}(X)$$

$$\sigma = \sqrt{\text{Var}(X)}$$

Standardization:

$$Z = \frac{X - \mu}{\sigma}$$

$$E(Z) = 0$$

$$\text{Var}(Z) = 1$$

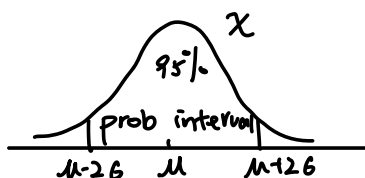
$$\text{If } X = \varepsilon_1 + \varepsilon_2 + \dots + \varepsilon_n \quad \varepsilon_i \sim \text{Ber}\left(\frac{1}{2}\right)$$

$$= \sum_{i=1}^n \varepsilon_i \quad (\varepsilon_i \text{ independent})$$

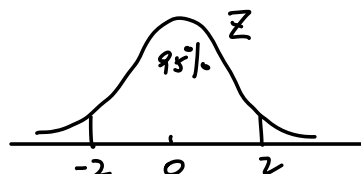
Central Limit Theorem (CLT) (normal approximation)

$$Z = \frac{X - \mu}{\sigma} \sim N(0, 1)$$

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



standardization $\rightarrow$



$$\text{Example: } X \sim \text{Bin}(n=100, p=\frac{1}{2})$$

$$\mu = E(X) = np = 50$$

$$\sigma^2 = \text{Var}(X) = np(1-p) = 25 \quad \sigma = 5$$

$$95\% \text{ prob interval: } P(X \in [40, 60]) = 95\%$$

Start from binomial distribution :

$$P(X \in [40, 60]) = \sum_{k=40}^{60} P(X=k) = \sum_{k=40}^{60} \frac{\binom{n}{k}}{2^n}$$

$$Z = \frac{X - \frac{n}{2}}{\sqrt{n}/2} = \frac{k - \frac{n}{2}}{\sqrt{n}/2} \quad (X = k \in [40, 60])$$

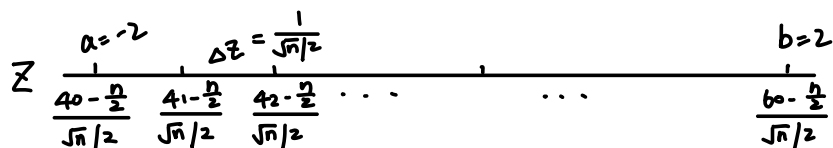
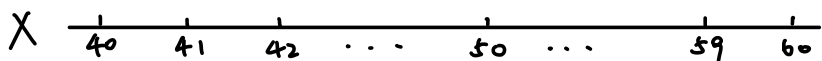
$$\Leftrightarrow k = Z \frac{\sqrt{n}}{2} + \frac{n}{2}$$

$$P(X \in [40, 60]) = \sum_{k=40}^{60} \frac{\binom{n}{k}}{2^n}$$

$$= \sum_{Z \in (a,b)} \frac{\binom{n}{Z \frac{\sqrt{n}}{2} + \frac{n}{2}}}{2^n}$$

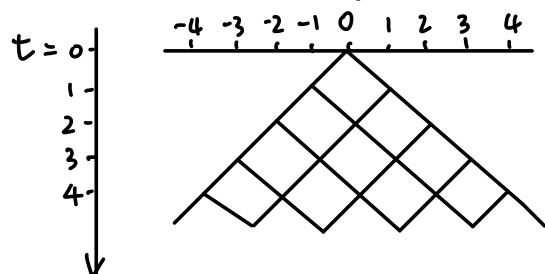
$$= \sum_{Z \in (a,b)} f(Z) \Delta Z \quad \xrightarrow{\quad} \frac{1}{\sqrt{2\pi}} e^{-\frac{Z^2}{2}}$$

$$= \int_a^b f(z) dz$$



Random Walk

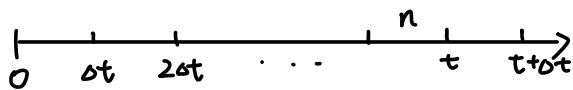
Recall on integer



Similar to Galton Board

Continuous time, Diffusion, Brownian motion

$$\Delta t = \frac{t}{n}$$



$X(0) \quad X(\Delta t) \quad \dots$

$X(t) \quad X(t + \Delta t)$

$$\left(\begin{array}{c} \text{vs. poisson} \\ \text{---} \\ t \quad t + \Delta t \\ \text{p(head)} = \lambda \Delta t \end{array} \right)$$

Z_i	-1	+1
prob	$\frac{1}{2}$	$\frac{1}{2}$

$$X(t) = \epsilon_1 + \epsilon_2 + \dots + \epsilon_n$$

$$= \sum_{i=1}^n \epsilon_i$$

$$= \sum_{i=1}^n Z_i \Delta x$$

According to Central Limit Theorem:

$$X(t) = \sum_{i=1}^n Z_i \Delta x \sim N(0, \sigma^2 t)$$

Q: What is the relationship between x and Δt .

$$E(Z_i) = (-1) \times \frac{1}{2} + (1) \times \frac{1}{2} = 0$$

$$\text{Var}(Z_i) = (-1 - 0)^2 \times \frac{1}{2} + (1 - 0)^2 \times \frac{1}{2} = 1$$

$$E(X(t)) = E\left(\sum_{i=1}^n Z_i \Delta x\right) = \sum_{i=1}^n E(Z_i) \Delta x = 0$$

$$\text{Var}(X(t)) = \sum_{i=1}^n \text{Var}(Z_i \Delta x) = \sum_{i=1}^n \text{Var}(Z_i) \Delta x^2 = n \Delta x^2 = \sigma^2 t$$

$$\Rightarrow \Delta x^2 = \sigma^2 \frac{t}{n} = \sigma^2 \Delta t$$

$$\Delta x = \sigma \sqrt{\Delta t}$$

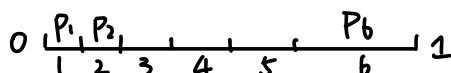
$$\text{Velocity} = \frac{\Delta x}{\Delta t} = \frac{\sigma \sqrt{\Delta t}}{\Delta t} = \frac{\sigma}{\sqrt{\Delta t}} \xrightarrow{\Delta t \rightarrow 0} \infty$$

Part 2: 2 Random Variables

Notation: $X = (X_1, X_2)$

Discrete: $X \sim p(x)$

biased die



$$X = (X_1, X_2)$$

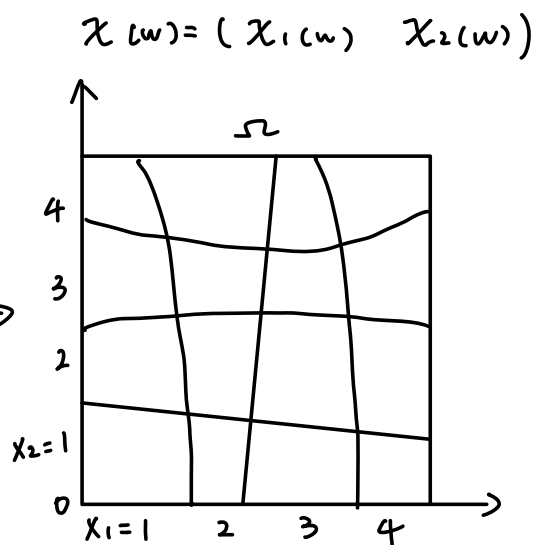
$$p(x) = p(x_1, x_2)$$

$$X = (x_1, x_2)$$

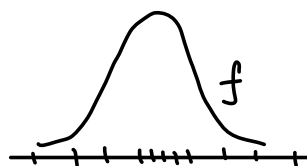
scalar
↓
vector

Ex.

		$x_2 = \text{hair color}$			
		$x_1 = \text{eye color}$			
	color	1	2	3	4
1		p_{11}	p_{12}	p_{13}	p_{14}
2		p_{21}	$\dots$		$\vdots$
3		p_{31}		$\dots$	
4		p_{41}	$\dots$		p_{44}

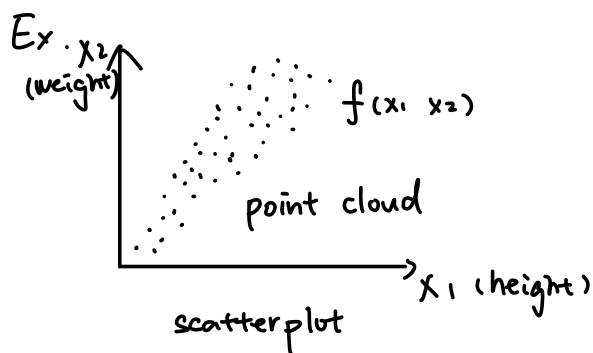


Continuous: $\chi \sim f(x)$



$$\chi = (\chi_1, \chi_2)$$

$$f(x) = f(x_1, x_2)$$



$$3D: f(x) = f(x_1, x_2, x_3)$$