

$$X = (X_1, \dots, X_d) \stackrel{d}{\sim} N(\mu, \Sigma)$$

Y_1, \dots, Y_d are jointly normal if
 $Y = (Y_1, \dots, Y_d)$ is MVN

Def Expln

$X \sim N(0, 1)$ Let $W \stackrel{d}{\sim}$ Below condition

$$\Rightarrow P(W=1) = \frac{1}{2} \quad P(W=-1) = \frac{1}{2}$$

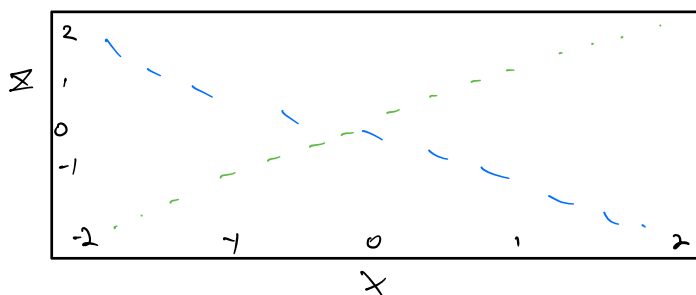
$$\Rightarrow W \perp X$$

$$\Rightarrow Z = WX$$

$Z \sim N(0, 1)$ (compute mgf of Z using

$$E(E(e^{tZ} | W)) = E(e^{tZ})$$

is mv Normal? $(X, Z) \sim N(\mu, \Sigma)$



Show a vector $a^T \begin{pmatrix} X \\ Z \end{pmatrix}$ is not normally dist'd

Take $a = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ then

$$a^T \begin{pmatrix} X \\ Z \end{pmatrix} = X + Z = X + WX = (1+W)X.$$

How to show $X+Z$ is NOT normal?

$$\begin{aligned} P(X+Z=0) &= P(W=-1) = \frac{1}{2} \\ P(X+Z > 0) &= P(W=1, X > 0) \\ &\leq P(X > 0) = \frac{1}{2} \end{aligned}$$

$\therefore X+Z$ is not normal RV.

Let $\begin{pmatrix} X \\ Z \end{pmatrix}$ be MVN but has no density as \circ exist

How to Construct Multivariate normal.

THM:

Theorem 4.6. Let $Z_1, Z_2 \stackrel{iid}{\sim} \mathcal{N}(0,1)$. Then, the bivariate random variable (X, Y) defined by

$$X \doteq \sigma_X Z_1 + \mu_X, \quad Y \doteq \sigma_Y(\rho Z_1 + \sqrt{1-\rho^2} Z_2) + \mu_Y,$$

follows a $\mathcal{N}_2(\mu, \Sigma)$ distribution, where

$$\mu = (\mu_X, \mu_Y), \quad \Sigma = \begin{pmatrix} \sigma_X^2 & \rho\sigma_X\sigma_Y \\ \rho\sigma_X\sigma_Y & \sigma_Y^2 \end{pmatrix}.$$

Theorem 4.7. Suppose that (X, Y) are jointly normal. Then X and Y are independent if and only if $\text{Cov}(X, Y) = 0$.

Brownian Bridge

Pf: if $X \perp Y$ $\text{Cov}(X, Y) =$

$$\begin{aligned} &= E(XY) - E(X)E(Y) \\ &\xrightarrow{X \perp Y} E(X)E(Y) - E(X)E(Y) \\ &= 0 \end{aligned}$$

Suppose $\text{Cov}(X, Y) = 0$

Denote mean vector (X, Y) by (μ_x, μ_y)

Sigma $\rightarrow \Sigma$ by $\begin{pmatrix} \sigma_x^2 & 0 \\ 0 & \sigma_y^2 \end{pmatrix}$

$z_1, z_2 \stackrel{\text{iid}}{\sim} N(0, 1)$ define

$$\bar{X} = \sigma_x z_1 + \mu_x$$

$$\bar{Y} = \sigma_y z_2 + \mu_y$$

$$\text{THM} \Rightarrow (\bar{X}, \bar{Y}) \stackrel{d}{=} (X, Y)$$

$$\bar{X} \perp \bar{Y} \Rightarrow X \perp Y$$

Proposition 4.8. Let $X \sim \mathcal{N}_d(\mu, \Sigma)$, and let $A \in \mathbb{R}^{n \times d}$ be a deterministic matrix. Then, $AX \sim \mathcal{N}_n(A\mu, A\Sigma A^T)$.

\Rightarrow Linear Transformation of MVN is also MVN.

Definition 4.9. An \mathbb{R} -valued stochastic process $\{W_t\} = \{W_t\}_{t \geq 0}$ is said to be a **standard Brownian motion (SBM)** or **Wiener process** if:

- (1) The **increments** of $\{W_t\}$ are stationary and independent.
- (2) For each $t \geq 0$, $W_t \sim \mathcal{N}(0, t)$.
- (3) $\mathbb{P}(W_t \text{ is continuous at all } t \geq 0) = 1$. Continuous sample path property

has Continuous State Space, Continuous Everywhere but not differentiable
PP State space = discrete

Is there a Stochastic Process satisfy all 3 properties?
There is no process $\{X_t\}$ satisfy all 3 properties.

Ⓐ $\{X_t\}$ iid RVs

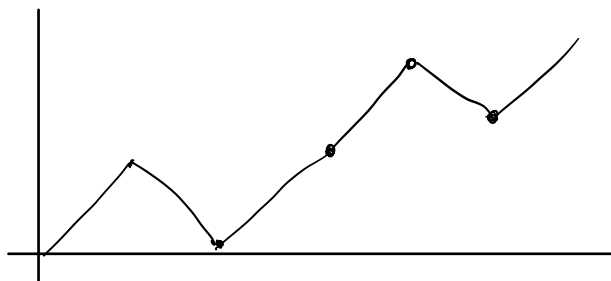
Ⓑ $\text{Var}(X_t) > 0$

Ⓒ $\mathbb{P}(X_t \text{ is a Cont @ all } t > 0) = 1$

Simple Random Walk w/ $\{X_i\}$ iid $\mathbb{P}(X_i = 1) = \mathbb{P}(X_i = -1) = \frac{1}{2}$

(Discrete)

(Continuous)



$$CLT: S_n \approx N(0, n) \quad \hookrightarrow \text{var}$$

$$\frac{S_n}{\sqrt{n}} \approx N(0, 1)$$

if n is large

Discrete

$$S_n = \frac{1}{\sqrt{n}} \sum_{i=1}^n \xi_i$$

continuous

$$X_t = \text{connecting } S_m^n$$

n jumps in $[0, 1]$

Each w/ size $= \frac{1}{\sqrt{n}}$

