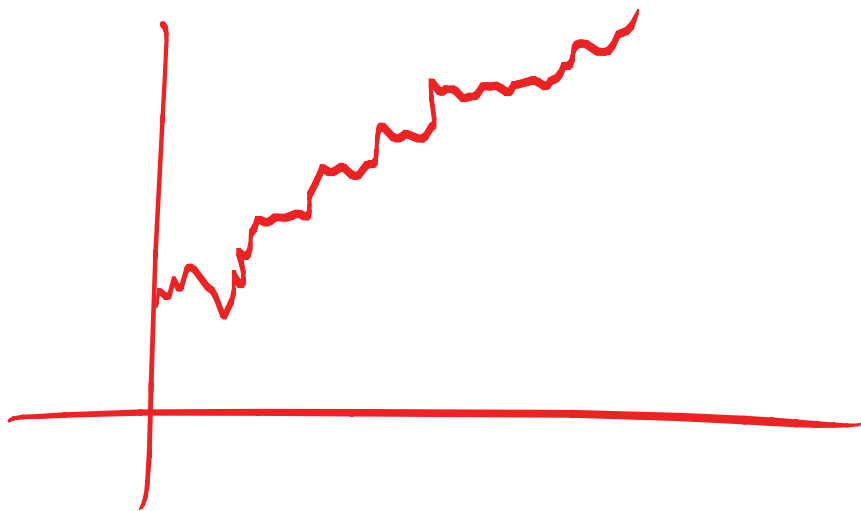



$$dS = \mu S dt + \sigma S dW \quad \text{cts}$$

$$\delta S = \mu S \delta t + \sigma S \phi \sqrt{\delta t} \quad \text{discrete}$$



$$d[S] = \mu S dt + \sigma S dW$$

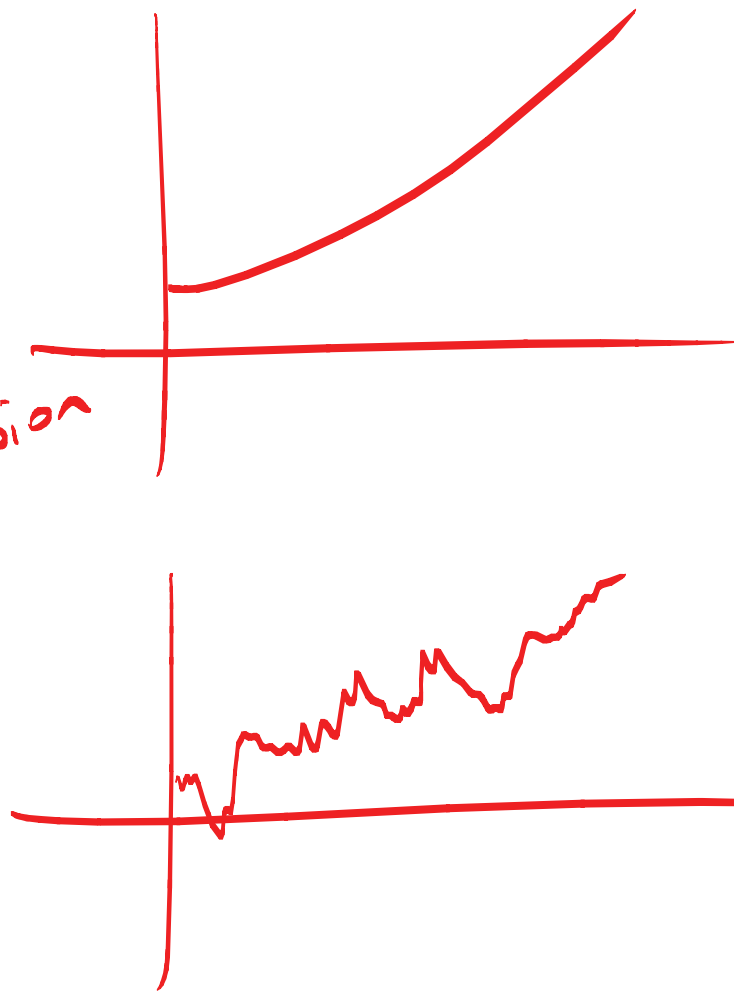

 stochastic process
 state variable

$$dr = \frac{\sigma}{\rho} dW$$

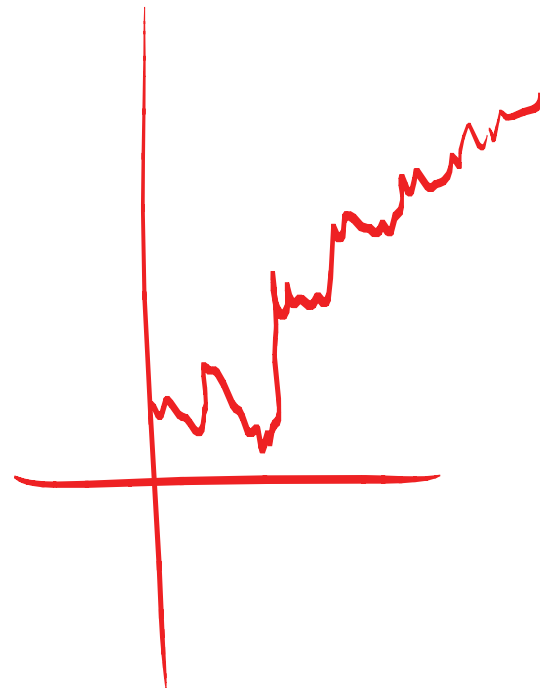
$a dt$
determ.
drift

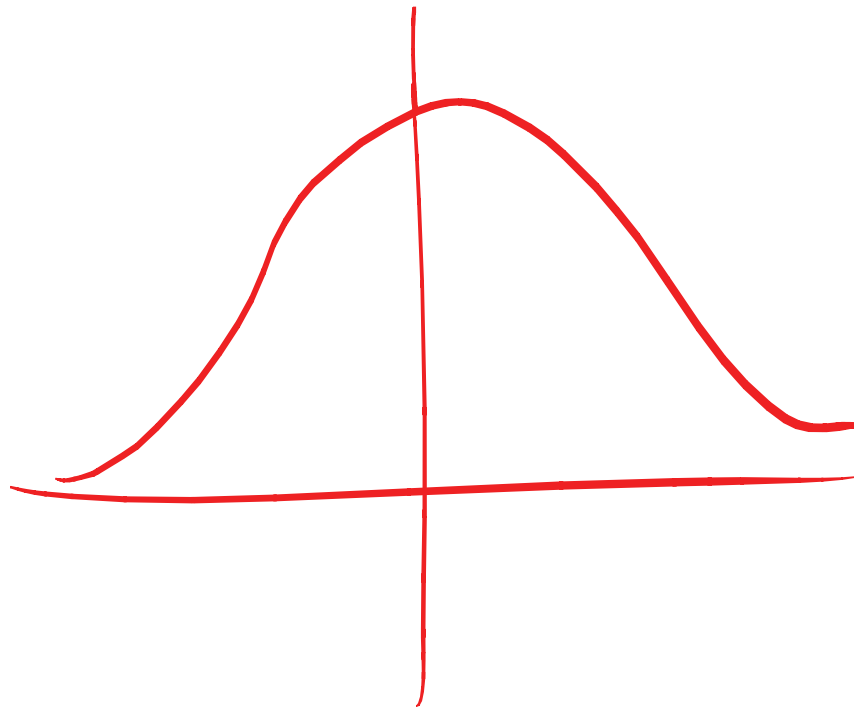
$b dx$
random
part.

diffusion



$=$





$S(t)$

diffusion $\rightarrow B(s, t) \underline{\underline{dx}}$

SDE in Integral form

$$\textcircled{1} \quad dG = A(s, t) dt + B(s, t) dX_t$$

Intes. over $(0, t)$

$$\int_0^t dG = \int_0^t A d\tau + \int_0^t B dX_\tau$$

$$G_t - G_0 = \int_0^t A d\tau + \int_0^t B dX_\tau$$

$\textcircled{1} = \textcircled{2}$
It's Integral

$$\textcircled{2} \quad G_t = G_0 + \int_0^t A d\tau + \int_0^t B dX_\tau$$

Put $\sigma = 0$ in GBM

$$dS = \mu S dt$$

$$\int \frac{dS}{S} = \mu \int dt$$

$$\ln S = \mu t + C$$

$$S_t = A e^{\mu t}$$

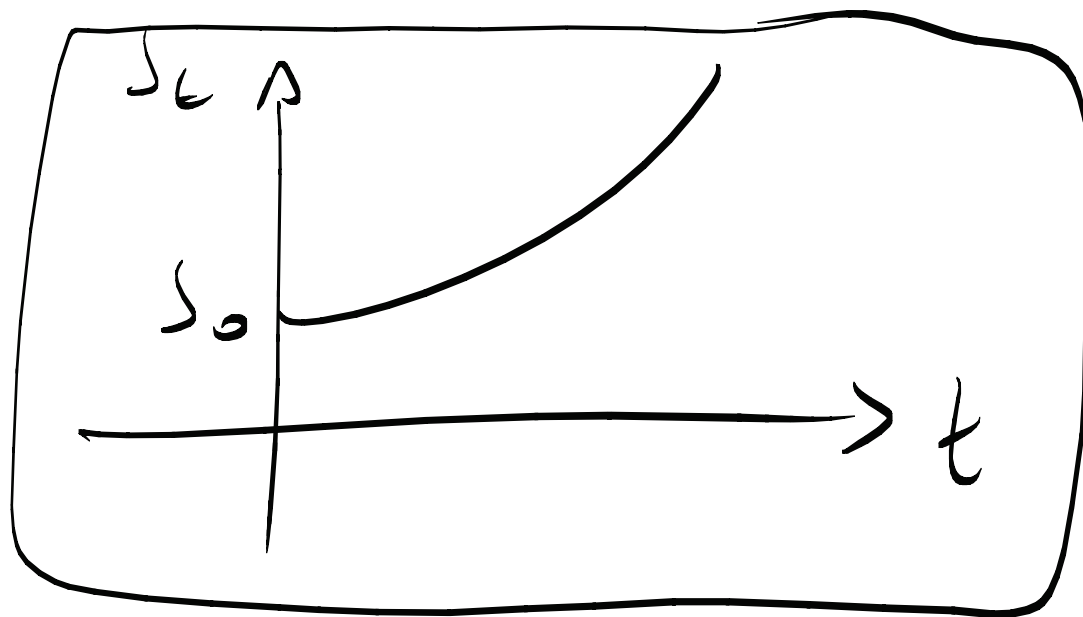
$$\mu \rightarrow r$$

$$\text{at } t=0 \quad S=S_0$$

$$S_t = A e^{rt}$$

$$A = S_0$$

$$S_t = S_0 e^{rt}$$



$$dS = \mu S dt + \sigma S dX$$

$$dS^2 = \cancel{\mu^2 S^2 dt^2} + \underline{\sigma^2 S^2 dt} + \cancel{2\mu\sigma S^2 dt dX}$$

$\swarrow \rightarrow 0$
 $\swarrow \rightarrow o(dt^{3/2})$

$$dS^2 = \sigma^2 S^2 dt$$

Subst in the
r.h.s of dV

Ex: $V = S^2$ S is G.B.M

$V' = 2S$ $V'' = 2$

subst in (1)

Ex: ① $V = S^n$
② $V = e^S$

$$dV = \left(\mu S \cdot 2S + \frac{1}{2} \sigma^2 S^2 \cdot 2 \right) dt + \sigma S \cdot 2S dx$$

$$dV = \underbrace{(2\mu S^2 + \sigma^2 S^2)}_{\text{drift}} dt + \underbrace{2\sigma S^2}_{\text{diffusion}} dx$$

$$dG = A(s, t) dt + B(s, t) dx$$

$$dG^2 = B^2 dt$$

$$\begin{aligned} E[dG] &= E[A(s, t) dt] + E[B dx] \\ &= A dt \end{aligned}$$

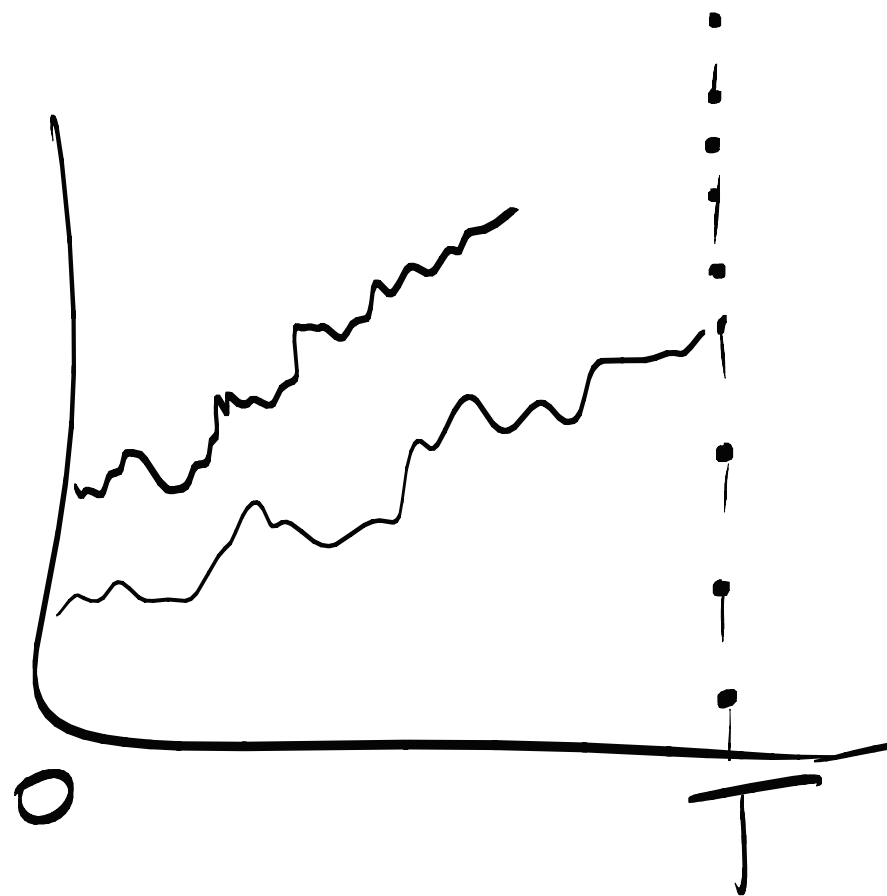
$$V[dG] = V[A dt] + V[B dx]$$

\swarrow
 $\rightarrow 0$

$B^2 dt$

$$dV = \int \left[\cancel{\mu} \cancel{s} \left(\frac{1}{\cancel{s}} \right) + \frac{1}{2} \cancel{\sigma^2} \cancel{s} \left(-\frac{1}{\cancel{s^2}} \right) \right] dt + \cancel{\sigma} \cancel{s} \frac{1}{\cancel{s}} dx$$

$$d(\log s) = \left(\mu - \frac{1}{2} \sigma^2 \right) dt + \sigma dX$$



$$dr = - \underbrace{\gamma(r - \bar{r})}_u dt + \sigma dx$$

$$u = r - \bar{r}$$

$$du = dr$$

Ornstein
Uhlenbeck process

$$du = -\gamma u dt + \sigma dx$$

$$d(u e^{rt}) = e^{rt} du + u r e^{rt} dt$$

$$= e^{rt} [-\cancel{ru dt} + \sigma dX] + \cancel{ru e^{rt} dt}$$

$$\int_0^t d(u e^{rs}) = \int_0^t \sigma e^{rs} dX$$

$$\frac{d}{dt}(e^{rt}) = r e^{rt}$$

$$de^{rt} = r e^{rt} dt$$

$$u_s e^{rs} \Big|_0^t = \sigma \int_0^t e^{rs} dX_s$$

$$u_t e^{rt} - u_0 = \sigma \int_0^t e^{rs} dX_s$$

$$U_t = U_0 e^{-rt} + \sigma \int_0^t e^{r(s-t)} dX_s$$

$$\int_0^t e^{r(s-t)} dX_s \equiv \int V d\bar{u}$$

$$V = e^{r(s-t)}$$

$$dV = r e^{r(s-t)} ds$$

$$dX_s \equiv d\bar{u}$$

$$\bar{u} = X_s$$

$$X_s e^{\gamma(s-t)} \Big|_0^t - \gamma \int_0^t X_s e^{\gamma(s-t)} ds$$

$$X_t - \underbrace{X_0}_{\text{L}_0} e^{\gamma(t)}$$

o

$$X_t - \gamma \int_0^t e^{\gamma(s-t)} X_s ds$$

Unsteady \rightarrow t -dep

Steady \rightarrow t -indep.

Steady state solⁿ: $\hookrightarrow t \rightarrow \infty$

solⁿ is time indep.

$$\frac{1}{2} \sigma^2 \frac{d^2}{dr^2} (p) - \gamma \frac{d}{dr} (\bar{r} - r) p = 0$$

Integ. through

$$\frac{1}{2} \sigma^2 \int \frac{d^2}{dr^2} (p) = \gamma \int \frac{d}{dr} (\bar{r} - r)$$

$$\frac{1}{2} \sigma^2 \frac{d}{dr} p = -\gamma (r - \bar{r}) p + C$$

a) $r \rightarrow \infty$

$$\left. \begin{array}{l} p \rightarrow 0 \\ p' \rightarrow 0 \end{array} \right\}$$

$$\therefore C = 0$$

$$\frac{1}{2} \sigma^2 \frac{dp}{dr} = -\gamma (r - \bar{r}) p$$

$$\frac{dp}{dr} = -\frac{2\gamma}{\sigma^2} (r - \bar{r}) p \int \frac{d}{dr} \frac{1}{2} (r - \bar{r})^2$$

$$\frac{dp}{p} = -\frac{2\gamma}{\sigma^2} \int (r - \bar{r}) dr$$

$$\int_{-\infty}^{\infty} e^{-x^2} dx = \sqrt{\pi}$$

$$dS = S_{i+1} - S_i = \mu S_i \delta t + \sigma S_i dX$$

$$S(t) = S_i$$

next time step
now

$$S_{i+1} = S_i + \mu S_i \delta t + \sigma S_i \phi \sqrt{\delta t}$$

$$S_{i+1} = S_i (1 + \mu \delta t + \sigma \phi \sqrt{\delta t})$$

Normal Distⁿ

pdf

$$p(x) = \frac{dF}{dx} \leftarrow \text{CDF}$$

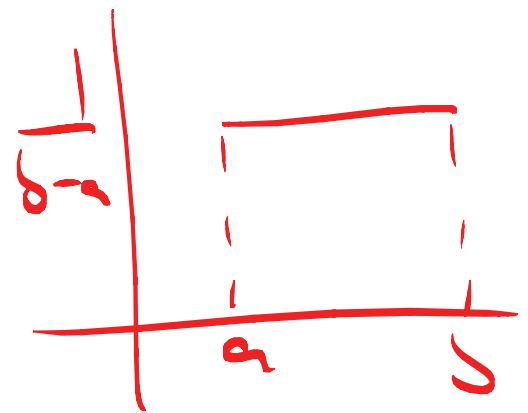
$$F(x) = \int_{-\infty}^x p(x) dx$$

$$N(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\boxed{x}} e^{-\frac{1}{2}\phi^2} d\phi = P[X \leq x]$$

Standard Normal.

$$\mathbb{E}\left[\sum_{i=1}^{12} R_i\right] = ? \sum_{i=1}^{12} \mathbb{E}(R_i) = 12 \cdot \frac{1}{2} = 6$$

$$p(x) = \begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & \text{elsewhere} \end{cases}$$



$$\mathbb{E}[\text{Unif}] = \frac{a+b}{2}$$

$$\mathbb{E}(\text{RAND}) = \frac{1}{2}$$

$$\text{Var}(\text{Unif}) = (b-a)^2/12$$

$$\text{Var}(\text{RAND}) = 1/12$$

$$p = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{elsewhere} \end{cases}$$

$$E(\text{RAND}) = \int_0^1 x \, dx = \left. \frac{x^2}{2} \right|_0^1 = \frac{1}{2}$$

$$\begin{aligned} V(\text{RAND}) &= \int_0^1 x^2 p(x) \, dx - \frac{1}{4} \\ &= \left. \frac{x^3}{3} \right|_0^1 - \frac{1}{4} = \frac{1}{12} \end{aligned}$$

X_i i.i.d

$$\sum_{i=1}^n X_i - N\mu$$

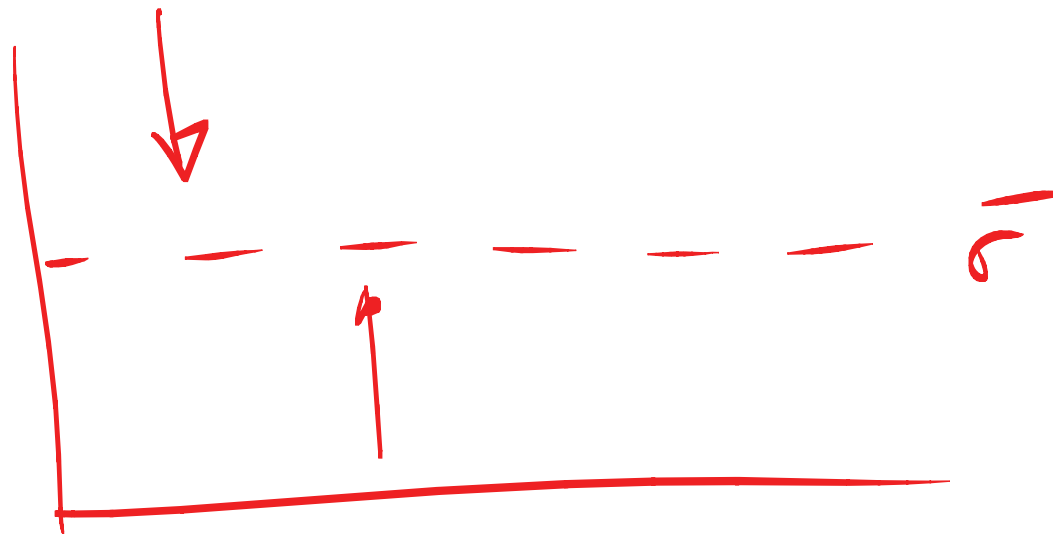
$$\rightarrow \phi \sim N(0, 1)$$

$$\frac{\sum_{i=1}^n \text{RAND} - n/2 \cdot \frac{1}{2}}{\frac{1}{\sqrt{2}} \sqrt{2}}$$

Ex: Extend to N RANDC

$$\sqrt{\frac{12}{N}} \left[\sum_{i=1}^N \text{RANDC}_i - \frac{N}{2} \right]$$

γ high



γ low



$$-\gamma(r - \bar{r})$$

$$r > \bar{r} \quad -\gamma(+ve) \quad -ve \text{ trend}$$

$$r < \bar{r} \quad -\gamma(-ve) \quad +ve \text{ trend}$$

In Vasicek put $\sigma = 0$

$$dr = -\gamma(r - \bar{r}) dt$$

$$\int \frac{dr}{r-\bar{r}} = -\gamma \int dt$$

$$\ln(r-\bar{r}) = -\gamma t + A$$

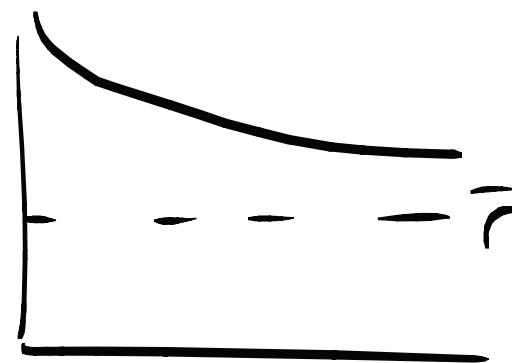
$$r-\bar{r} = C e^{-\gamma t}$$

$$r(t) = \bar{r} + C e^{-\gamma t}$$

γ high



γ low



$$\left. \begin{aligned}
 \varepsilon_1, \varepsilon_2 &\sim N(0, 1) \\
 \mathbb{E}[\varepsilon_1] &= 0 = \mathbb{E}[\varepsilon_2] \\
 \mathbb{E}[\varepsilon_1^2] &= 1 = \mathbb{E}[\varepsilon_2^2] \\
 \mathbb{E}[\varepsilon_1 \varepsilon_2] &= 0
 \end{aligned} \right\} \begin{aligned}
 &\text{We want produce} \\
 &\phi_1, \phi_2 \text{ such that} \\
 &\phi_1, \phi_2 \sim N(0, 1) \\
 &\mathbb{E}[\phi_1 \phi_2] = \rho
 \end{aligned}$$

$$\left. \begin{aligned}
 \phi_1 &= \varepsilon_1 \\
 \phi_2 &= \alpha \varepsilon_1 + \beta \varepsilon_2
 \end{aligned} \right\} \begin{aligned}
 &\text{What values of } \alpha, \beta \\
 &\text{give,} \\
 &\mathbb{E}[\phi_1 \phi_2] = \rho
 \end{aligned}$$

$$\mathbb{E}[\phi, \phi_2] = \rho = \mathbb{E}[\varepsilon_1 (\alpha \varepsilon_1 + \beta \varepsilon_2)]$$

$$= \alpha \underbrace{\mathbb{E}[\varepsilon_1^2]}_1 + \beta \underbrace{\mathbb{E}[\varepsilon_1 \varepsilon_2]}_0 \Rightarrow \boxed{\alpha = \rho}$$

$$\mathbb{E}[\phi_2^2] = 1 = \mathbb{E}[(\alpha \varepsilon_1 + \beta \varepsilon_2)^2]$$

$$= \mathbb{E}[\alpha^2 \varepsilon_1^2 + \beta^2 \varepsilon_2^2 + 2\alpha\beta \varepsilon_1 \varepsilon_2] = 1$$

$$\alpha^2 \cancel{\mathbb{E}(\varepsilon_1^2)} + \beta^2 \cancel{\mathbb{E}(\varepsilon_2^2)} + 2\alpha\beta \cancel{\mathbb{E}(\varepsilon_1 \varepsilon_2)} \leq 0$$

$$\alpha^2 + \beta^2 = 1 \Rightarrow \beta = \sqrt{1 - \alpha^2}$$

\therefore

$$\boxed{\begin{array}{l} \alpha = \rho \\ \beta = \sqrt{1 - \rho^2} \end{array}}$$

$$S_{1,i+1} = S_{1,i} [1 + \mu_1 \delta t + \sigma_1 \sqrt{\delta t} \varepsilon_1]$$

$$S_{2,i+1} = S_{2,i} [1 + \mu_2 \delta t + \sigma_2 \sqrt{\delta t} (\rho \varepsilon_1 + \sqrt{1-\rho^2} \varepsilon_2)]$$

$$\mathbb{E}[dX_1 dX_2] = \rho dt$$

$$\cancel{\phi} \sqrt{ft}$$

$$S_{i+1} = S_i \left[1 + \mu \delta t + \sigma \sqrt{\delta t} * \text{NORMSINV}(\text{RAND()}) \right]$$

