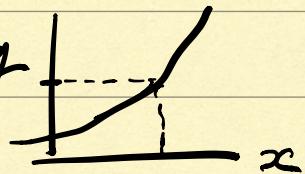


$X \sim p_x(x)$  AND  $Y = g(X)$  THEN

$$p_Y(y) = \sum_k p_X(g^{-1}_k(y)) \cdot \left( \left| \frac{dg}{dx_k} \right| \right)^{-1}$$

WHERE  $\{x_k\}$  IS THE PRE-IMAGE OF  $y$

IF  $g(x)$  IS MONOTONIC

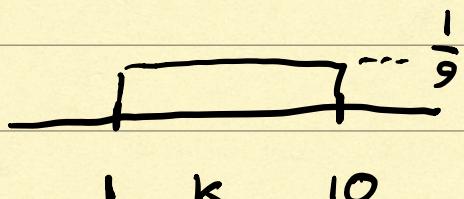


(NOT )

THEN  $p_Y(y) = p_X(g^{-1}(y)) \left| \frac{dg}{dx} \right|^{-1}$

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

EX



$K \sim \text{UNIFORM } [1, 10]$

$K$  - RATE

$p_K(k)$

T - MEAN TIME

$$T = \frac{1}{K}$$

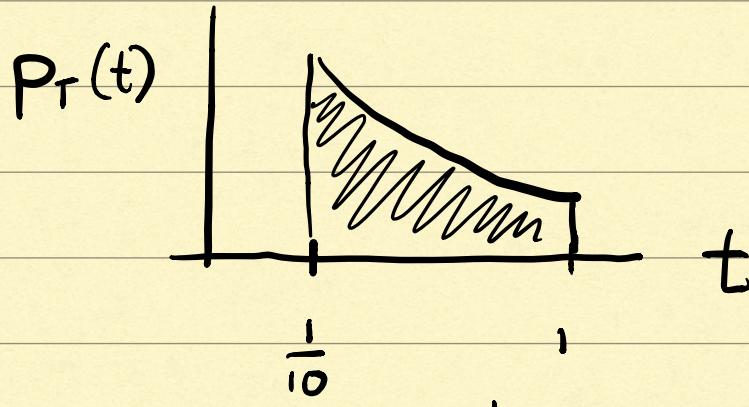
WHAT IS  
 $p_T(t)$ ?

$$g(k) = \frac{1}{k} \quad g^{-1}(t) = \frac{1}{t}$$

$$\frac{dg}{dk} = -\frac{1}{k^2}$$

$$\left. \frac{dg}{dk} \right|_{k=g^{-1}(t)} = -t^2$$

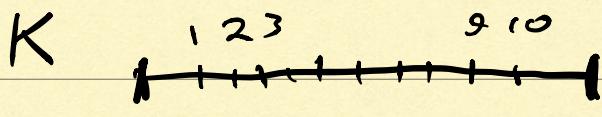
$$P_T(t) = \begin{cases} \frac{1}{9} \cdot t^{-2} & \frac{1}{10} < t < 1 \\ 0 & \text{ELSE} \end{cases}$$

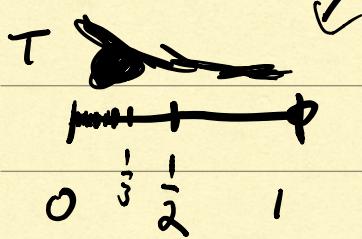


CHECK :

$$\int_{\frac{1}{10}}^1 P_T(t) dt = \int_{\frac{1}{10}}^1 \frac{1}{9} t^{-2} dt = -\frac{1}{9} \frac{1}{t} \Big|_{t=\frac{1}{10}}^{t=1}$$

$$= \frac{1}{9} (10-1) = 1 \quad \checkmark$$



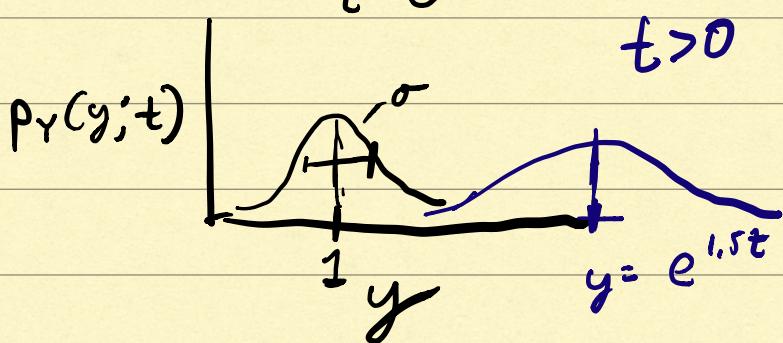
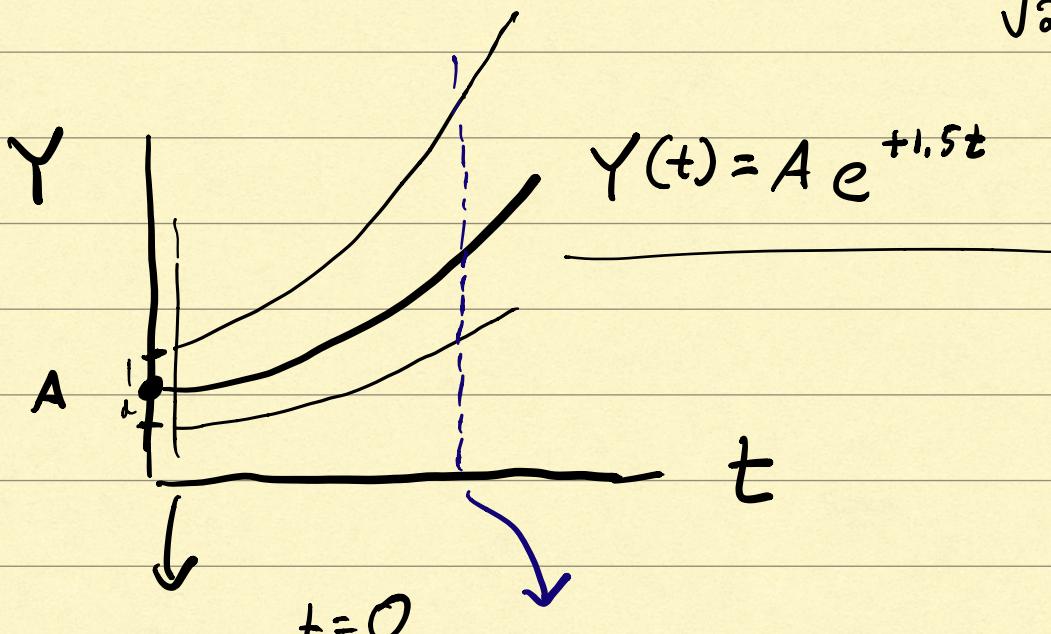


## EX HETEROGENITY / PARAMETRIC NOISE

$$\frac{dY}{dt} = (R-1)Y \quad R = 2.5$$

$$Y(0) = A \quad A \sim p_A(a)$$

$$p_A(a) = \frac{1}{\sqrt{2\pi}\sigma_0} e^{-\frac{(a-1)^2}{2\sigma_0^2}}$$



$$Y = g(A) = g(A; t)$$

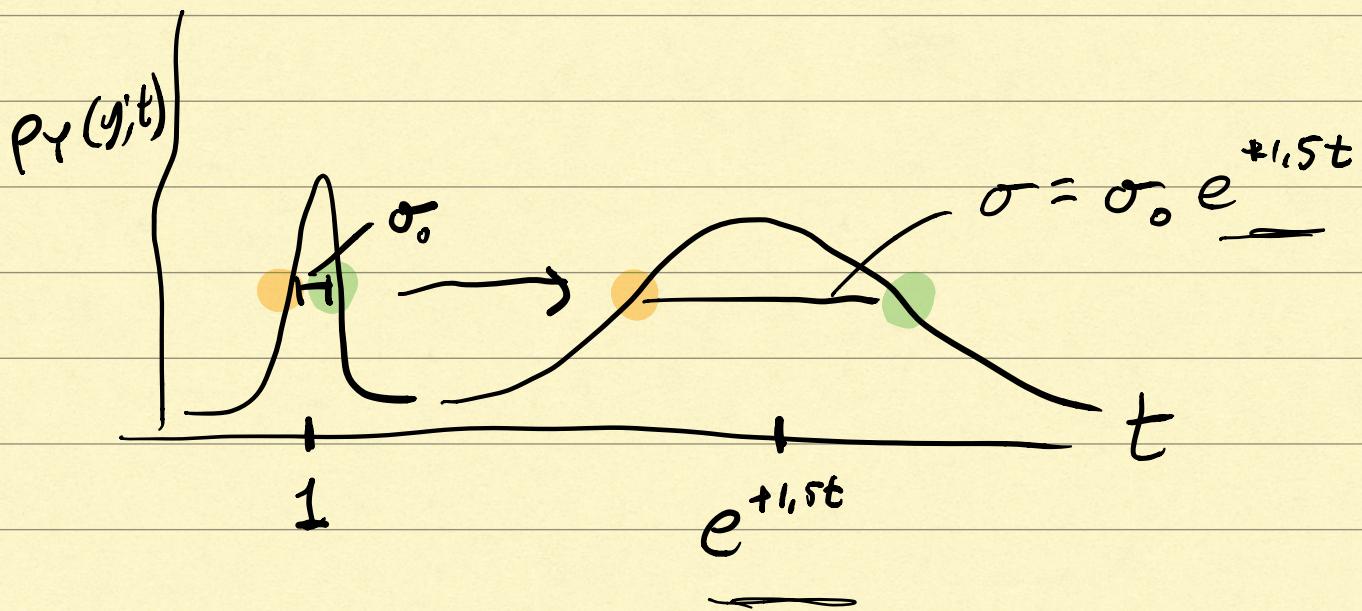
$$g^{-1}(y) = \frac{y}{e^{+1.5t}} = y e^{-1.5t}$$

$$\frac{dg}{da} = e^{1.5t}$$

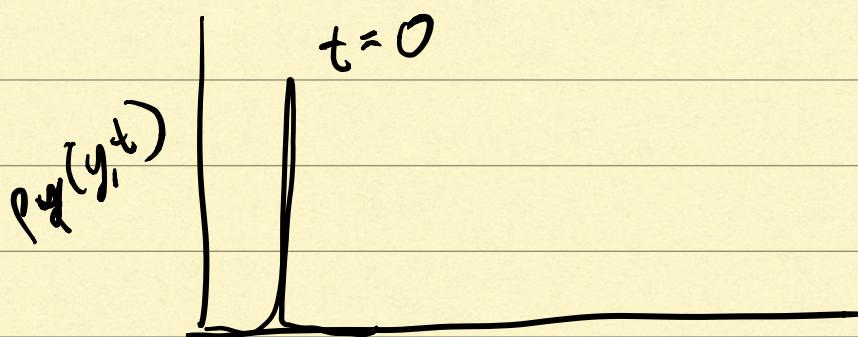
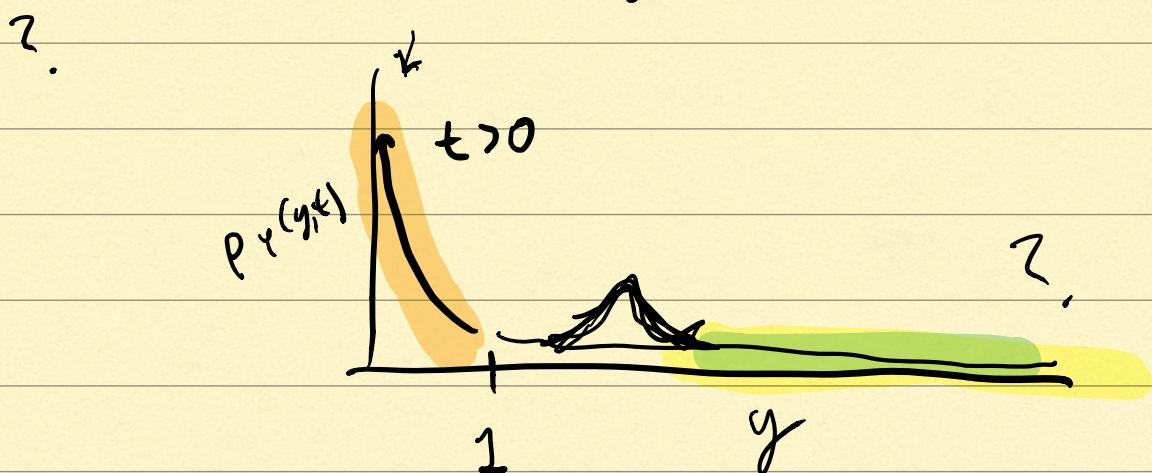
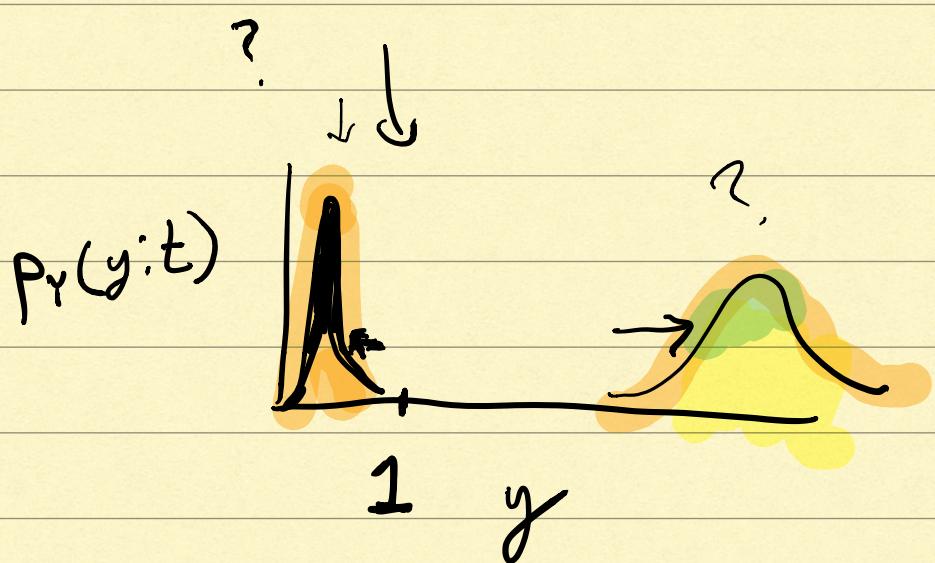
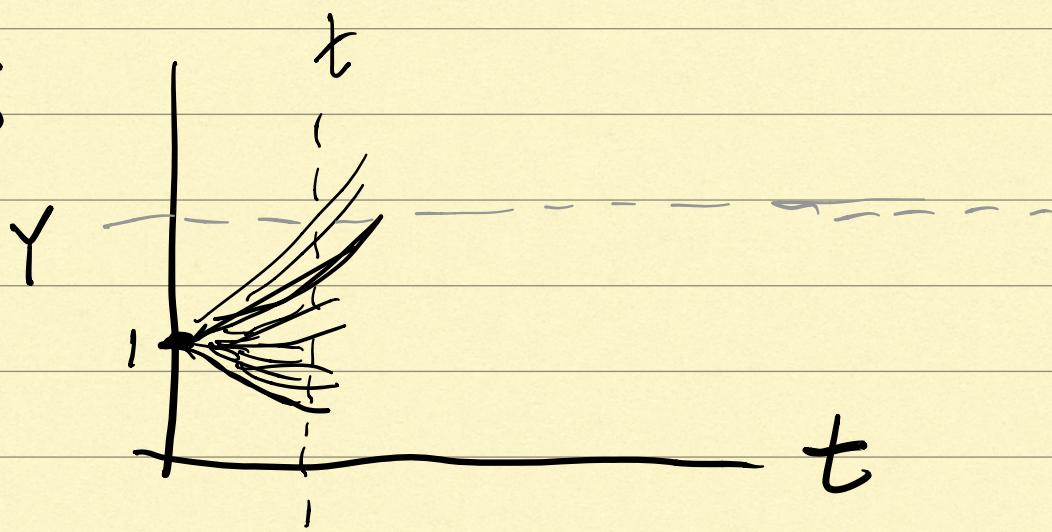
$$\left. \frac{dg}{da} \right|_{a=g^{-1}(y)} = e^{1.5t} \quad p_x(g^{-1}(y)) \left( \left| \frac{dg}{da} \right| \right)^{-1}$$

$$p_y(y; t) = \frac{1}{\sqrt{2\pi}\sigma_0} \exp\left(-\frac{(ye^{-1.5t} - 1)^2}{2\sigma_0^2}\right) e^{-1.5t}$$

$$= \frac{1}{\sqrt{2\pi}\sigma_0 e^{+1.5t}} \exp\left(-\frac{(y - e^{+1.5t})^2}{2(\sigma_0 e^{+1.5t})^2}\right)$$



PS 6



1

y

$$p_Y(y; t) = p_R \left( \frac{\ln(y)}{t} + 1 \right) \cdot \frac{1}{t y}$$

check  $\int_0^\infty p_Y(y; t) dy = 1$

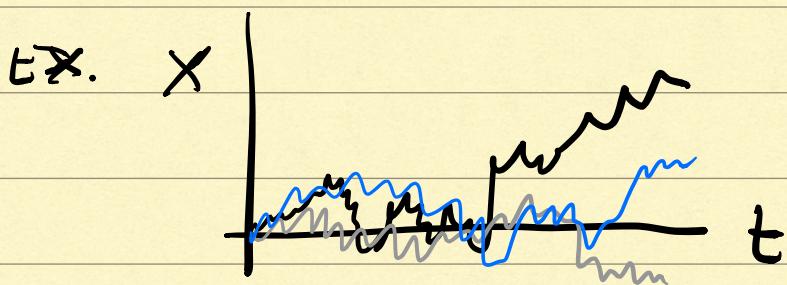
$t \gg 1$  ?

$$\mathbb{P}(Y > 1) = \int_1^\infty p_Y(y; t) dy$$

$$\mathbb{P}(Y < 1) = \int_0^1 p_Y(y; t) dy$$

## STOCHASTIC DIFFERENTIAL EQUATIONS

$X(t)$  - continuous-time, continuous-state space stochastic process

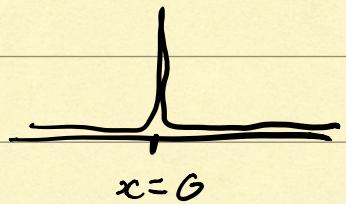


DIFFUSION  $X(t)$

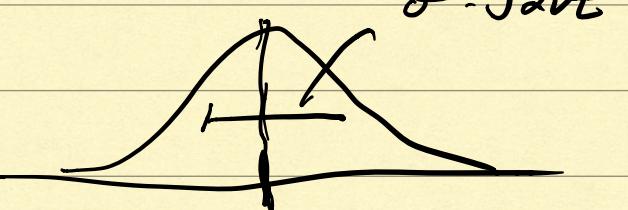
$P_x(x; t)$  satisfies

$$\frac{\partial p_x}{\partial t} = D \frac{\partial^2 p_x}{\partial x^2} \quad t=0$$

IF  $p_x(x, 0) = \delta(x)$



THEN  $p_x(x; t) = \frac{1}{\sqrt{2\pi D t}} e^{-\frac{x^2}{2Dt}}$



IF  $D = 1$

$B(t)$  - Brownian motion

GENERALIZE  $X(t) = \sigma B(t)$

$$\frac{dX}{dt} = \underline{f(x)} + \sigma \underline{w(t)}$$

$$w(t) = \frac{dB}{dt}$$

↓

WHITE NOISE

PROBLEM:  $B(t)$  IS NOT DIFFERENTIABLE!

OLD CALCULUS  $X(t) \approx a_0 + a_1 t + a_2 t^2 + \dots$

NEW CALCULUS  $X(t) \approx a_0 + a_1 t^{\frac{1}{2}} + a_1 t + a_2 t^2$

①  $\underline{dX} = \underline{f(x) dt} + \sigma \underline{dB}$

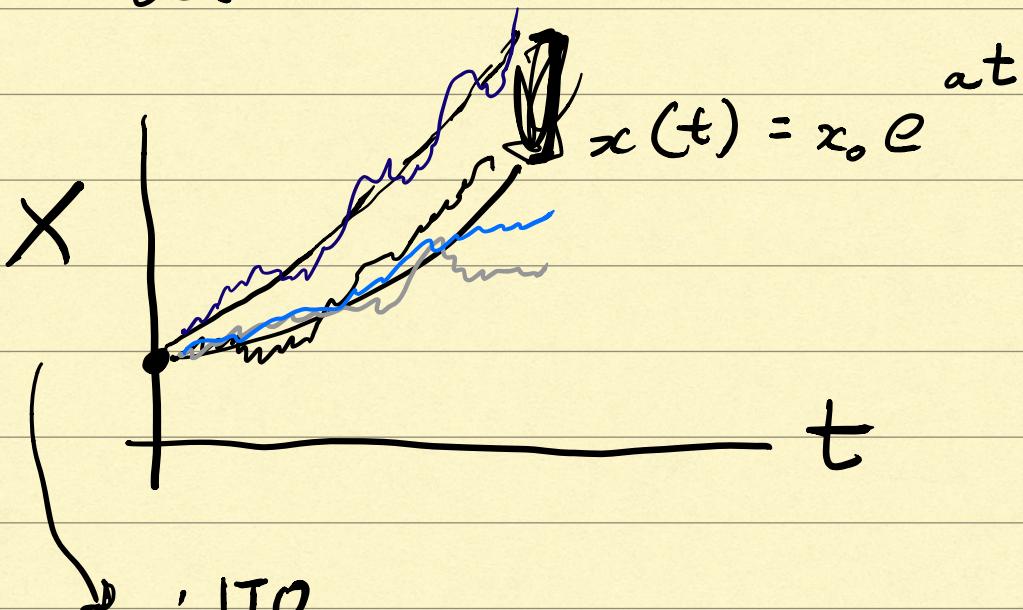
$$\frac{dX}{dt} = f(x) + \sigma \underline{\xi(t)}$$

② IN STOCHASTIC CALCULUS, THE  
NEW CHAIN RULE GETS A  
BONUS TERM.

$$dx = \alpha x dt + \sigma x dB$$

$$\frac{dx}{dt} = \alpha x$$

↓  
BLACK-SCHOLES  
MODEL



- ITO
- THORP
- BLACK
- SCHOLES
- SIMMONS

MFPT

$$-1 = D \frac{\partial^2 T}{\partial x^2}$$