

Reliability in a Gaussian Environment

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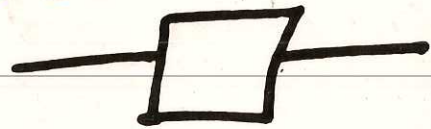
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THE HAZARD RATE

• DEF'N. $h(t) = P_n [t < T \leq t+dt | T > t] / dt$
 $= \frac{f(t)}{1-F(t)}$



$$\therefore F(t) = 1 - e^{-\int_0^t h(t') dt'}$$

• RELIABILITY = $P_n [T > t]$

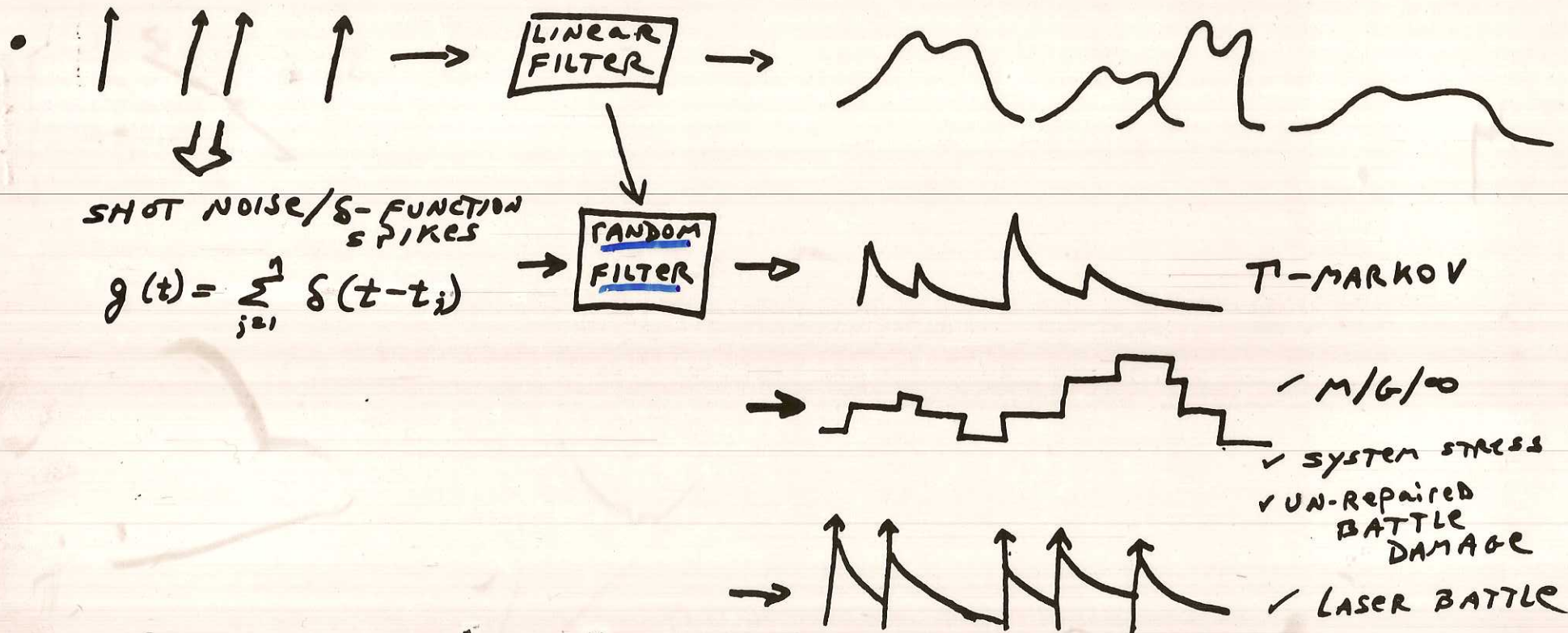
$$R(t) = 1 - F(t) = e^{-\int_0^t h(t') dt'}$$

$$\rightarrow h(t) = -\frac{d}{dt} \ln R(t)$$

• PRESUMABLY AVERAGED OVER

- MFG VARIABLES
- HOW IT'S OPERATED
- ENVIRONMENTAL COND'S
- ⋮

FILTERED SHOT NOISE PROCESSES



•
$$h(t') = \lambda(t') + \sum_{j=1}^n r_j(t'-t_j)$$

↑ random response function

e.g. if failures \propto unrepaired battle damage

$$r_j(t'-t_j) = \kappa \beta_j \{ \Theta(t'-t_j) - \Theta(t'-t_j-\tau_j) \} \sim \begin{array}{c} \left[\leftarrow \tau_j \rightarrow \right] \beta_j \\ \uparrow \\ t_j \end{array}$$

• ALL \Rightarrow non-exp'l lifetimes

GENERAL STOCHASTIC HAZARD PROCESS

- For j -th comp. $h_j(t) = \lambda_j(t) + g_j(t)$

$\left\{ \begin{array}{l} \text{STOCH. PART} \\ \text{wearout/env. factors} \\ \text{- indep. } \forall \text{ comp's.} \end{array} \right.$

- CONDITIONED ON REALIZATION OF $h(t)$

$$R_h(t) = e^{-\int_0^t h(t') dt'}$$

- FOR 1-comp. \rightarrow SIMPLY AVE. OVER SAMPLE PATHS

$$R(t) = E_h [R_h(t)] = E \left\{ e^{-\int_0^t h(t') dt'} \right\}$$

- Define effective hazard rate

$$\hat{h}(t) = -\frac{d}{dt} \ln R(t) \quad (\neq E \{ h(t) \})$$

- e.g. 2-comp's in \parallel :



$$R_h(t) = 1 - (1 - R_{h_1}(t))(1 - R_{h_2}(t)) = R_{h_1}(t) + R_{h_2}(t) - R_{h_1} \cdot R_{h_2}$$

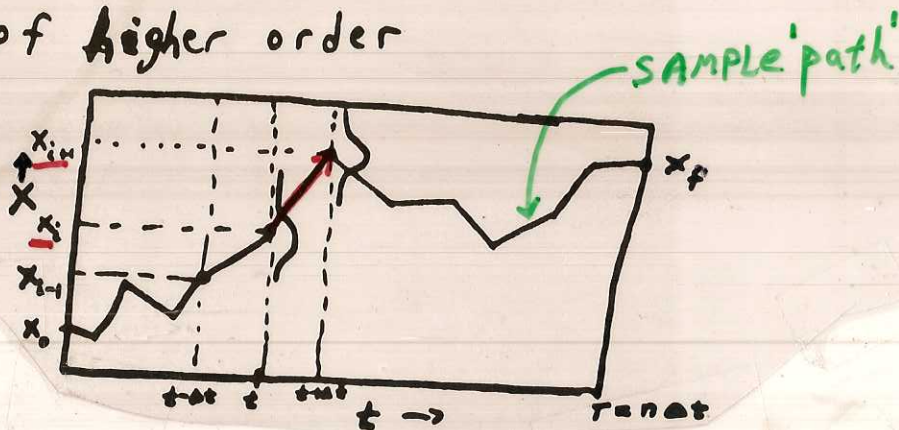
$$\therefore R(t) = R_1 + R_2 - \underline{E \left\{ e^{-\int_0^t (h_1 + h_2) dt'} \right\}} \neq R_1 + R_2 - R_1 \cdot R_2$$

SHORT-TIME LIMIT OF SOLUTION: SDE

$$\hat{P}_{rd}[x_i + \Delta x, t + \Delta t | x_i, t] = \frac{[1 + \hat{\alpha} g(x_i) \Delta t]}{\sqrt{2\pi \sigma^2 \Delta t}} \exp\left\{-\frac{(\Delta x + g(x_i) \Delta t)^2}{2\sigma^2 \Delta t}\right\}$$

- pdf of \mathbb{X}_{i+1}
- Dropped terms in Δt of higher order
- $0 \leq \hat{\alpha} < 1$

Probability of a 'path'



- really a prob. density functn.

- in limit as $\Delta t \rightarrow 0 \Rightarrow$ prob. density functional of a path

$$= \prod_i \hat{P}_{rd}[\underbrace{x_i + \Delta x}_{x_{i+1}}, t_i + \Delta t | x_i, t_i]$$

- ✓ USED MARKOV PROPERTY
- ✓ CONDITIONAL ON x_0

A Transformation on Path Integrals

$$\Delta y + \overset{\text{Environment}}{f(y)} \Delta t = \sqrt{\overset{\text{Wiener process}}{g(y)}} \Delta w \quad \text{Stochastic Differential Equation (SDE)}$$

-- Multiplicative noise term of form $\sqrt{g(x)} \Delta w$ can be transformed with appropriate stochastic calculus

-- stochastic differential equation: $\Delta x + \overset{\text{"force"}}{g(x)} \Delta t = \Delta w$

-- \exists ambiguity wrt $'g(x)'$ even though considering additive noise term

-- $\Delta x = x_{i+1} - x_i$; $\Delta w = w_{i+1} - w_i$; $w =$ Wiener process

-- Possible "discretizations": $'g(x)' = (1 - \hat{\alpha}) g(x_i) + \hat{\alpha} g(x_{i+1})$

-- $\hat{\alpha} = 1/2 \Rightarrow$ Stratonovich SDE; $\hat{\alpha} = 0 \Rightarrow$ Ito SDE

-- since $\sigma(x) = \text{const.}$ (possibly after a transformation)

Fokker-Planck eq'n is same [but may be relevant for path integral]

Short-Time Propagator for Paths

- require short time conditional pdf for
Chapman-Kolmogorov eq./path integral limit --> GAUSSIAN for $\Delta t \rightarrow 0$

- Expand SDE in Taylor series about x_i :

$$\Delta x + [g(x_i) + \hat{\alpha} g'(x_i)\Delta x + o(\Delta x)]\Delta t = \Delta w$$

$$\Delta x [1 + \hat{\alpha} g'(x_i)\Delta t] + g(x_i)\Delta t = \Delta w$$

$$\Rightarrow \Delta x = -g(x_i)\Delta t / [1 + \hat{\alpha} g'(x_i)\Delta t] + \Delta w / [1 + \hat{\alpha} g'(x_i)\Delta t]$$

- Hence, conditional on x_i , for short times, x_{i+1} will have

$$\text{mean } E(x_{i+1}) \approx x_i + g(x_i)\Delta t / [1 + \hat{\alpha} g'(x_i)\Delta t]$$

- and Var(x_{i+1}) $\approx \sigma^2 \Delta t / [1 + \hat{\alpha} g'(x_i)\Delta t]^2$

THE TRANSFORMATION FORMULA

- Express Expectations of FUNCTIONALS - over sample paths of SDE
- IN TERMS OF EXPECTATIONS OVER SAMPLE PATHS OF WIENER (BROWNIAN MOTION) PROCESS

$$E_{x_0, x_f} \left\{ F[x(\cdot)] \right\}_{\text{SDE}} = E_{x_0, x_f} \left\{ F[x(\cdot)] e^{\hat{\alpha} \int_0^t g(x) dt - \int_{x_0}^{x_f} \frac{g(x) dx}{\sigma^2} - \int_0^t \frac{g^2(x) dt}{2\sigma^2}} \right\}_{\text{W}}$$

- choose $\hat{\alpha} = \frac{1}{2}$ for convenience \Rightarrow Stratonovich (ordinary) calculus (NOT really necessary)

- Let $G'(x) \equiv g(x)$

$$E_{x_0, x_f} \left\{ F[x(\cdot)] \right\}_{\text{SDE}} = e^{\frac{G(x_0) - G(x_f)}{\sigma^2}} E \left\{ F[x(\cdot)] e^{\frac{1}{2} \int_0^t [g(x) - \frac{g^2}{\sigma^2}] dt} \right\}_{\text{W}}$$

APPLICATIONS OF THE TRANSFORMATION

Let $F[x_0] = 1$, $g(x) = \Lambda x$ (ORNSTEIN-UHLENBECK PROCESS)

$$\text{LHS} = \hat{P}_{rd}[x_f, t | x_0, 0] = \frac{\Lambda}{\pi \sigma^2 (1 - e^{-2\Lambda t})} \exp \left\{ \frac{-\Lambda (x_f - x_0 e^{-\Lambda t})^2}{\sigma^2 (1 - e^{-2\Lambda t})} \right\}$$

$$\neq \text{RHS} = e^{\frac{x_f^2 - x_0^2}{2\sigma^2} + \frac{\Lambda}{2} t} * E_{w, \mathcal{F}} \left\{ e^{-\frac{\Lambda}{2\sigma^2} \int_0^t x^2(s) ds} \right\}$$

\Rightarrow a little algebra + translate into cosh & sinh

Gives a now classic formula of Cameron/Martin
(KAC FUNCTIONAL)

• Now, identify hazard rate function as $\propto x^2(t)$, $\mathcal{X} \sim$ Wiener Process, w

$$\therefore R(t) = \int_{\mathcal{F}} E_{w, \mathcal{F}} \left\{ e^{-\alpha \int_0^t x^2(s) ds} \right\} = E \left\{ e^{-\int_0^t h(s) ds} \right\}$$

APPLICATIONS TO ORNSTEIN-UHLENBECK PROCESS

- $h(t) = \alpha X^2(t)$, $X \sim$ O.U. Process

$$E_{x_0, x_f} \left\{ e^{-\int_0^t x^2(s) ds} \right\}_{\text{O.U.}} = e^{\frac{x_f^2 - x_0^2}{2\alpha} + \frac{\Delta}{2} t} E_{x_0, x_f} \left\{ e^{-\left[\frac{\Delta}{2\alpha} + \alpha\right] \int_0^t x^2(s) ds} \right\}_{\underline{w}}$$

- $E\{\cdot\}$ on RHS is now same as above

- set $x_0 = 0$ & $\int dx_f$ to obtain $R(t)$

- OFFSET O.U.: Constant Force/VOLTAGE $L \frac{di}{dt} + Ri = f(t) + V_0$

i.e. $\Delta x + \Lambda (x - \hat{x}) \Delta t = \Delta w$

✓ use $\hat{x} = \frac{1}{2} \sqrt{\text{Var}(\text{O.U.})}$ to see effect of \hat{x} in SIMULATION

AVERAGES OVER PATHS

- CALCULATE expected values of function $F[\{x_i\}] = F[x_0, x_1, x_2, \dots, x_n]$

$$E\{F(\{x_i\})\} = \int dx_1 \dots \int dx_{n-1} \prod_i \hat{P}_{rd}[x_{i+1}, t_{i+1} | x_i, t_i] F(\{x_i\}_{i=0}^n)$$

- TAKE the limit \rightarrow continuous paths: $x_0 \rightarrow x_f$
 - insert our short-time pdf
 - $F[\cdot] \rightarrow$ functional of paths

$\begin{matrix} \xrightarrow{\Delta t \rightarrow 0} \\ \xrightarrow{n \rightarrow \infty} \end{matrix}$

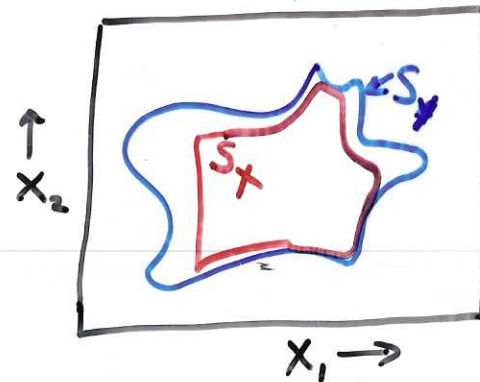
$$\int \mathcal{D}x(\cdot) F[x(\cdot)] \exp\left\{ \hat{\alpha} \int_0^t g(x) dt - \int_0^t \frac{\dot{x}^2}{2\sigma^2} dt - \int_0^t \frac{g(x) dx}{\sigma^2} - \int_0^t \frac{1}{2\sigma^2} dt \right\}$$

$$\lim_{\substack{\Delta t \rightarrow 0 \\ n \rightarrow \infty}} \prod_i \frac{dx_i}{\sqrt{\Delta t \sigma^2}}$$

Usual Wiener measure

ANALOGUE OF OUR PROCEDURE

- $\vec{y} \sim \text{pdf } g_{\vec{Y}}(\vec{y}) : \text{Support } S_Y$
 $\vec{x} \sim \text{pdf } f_{\vec{X}}(\vec{x}) : \text{Support } S_X$ $S_X \subseteq S_Y$



- $$E_{\vec{x}} \{ \hat{F}(\vec{X}) \} = \int_{S_X} \hat{F}(\vec{x}) f(\vec{x}) d^k x$$

$$= \int_{S_X} \hat{F}(\vec{x}) \frac{f_{\vec{x}}(\vec{x})}{g_{\vec{y}}(\vec{x})} \cdot g_{\vec{y}}(\vec{x}) d^k x$$

$$= E_{\vec{y}} \left\{ \hat{F}(\vec{Y}) \frac{f_{\vec{x}}(\vec{Y})}{g_{\vec{y}}(\vec{Y})} \cdot \mathbb{1}_{[\vec{Y} \in S_X]} \right\} \quad (1)$$

- If $S_X = S_Y, \Rightarrow$

$$= E_{\vec{y}} \left\{ \hat{F}(\vec{Y}) \frac{f_{\vec{x}}(\vec{Y})}{g_{\vec{y}}(\vec{Y})} \right\} \quad (2)$$

⇒ Don't need to calculate JACOBIAN if $g_{\vec{y}}(\vec{y})$ known, Eq.(2) helpful.

- If $\vec{x} \leftrightarrow \vec{y}$ (1-1), using Eq.(1) may be helpful.
 (because $S_X \neq S_Y$)

THE (WIENER PROCESS)²

- $h(t) = \lambda(t) + \alpha \underline{X^2(t)}$ where $X(t) \sim$ Wiener Process

- Require:

$$C_{X^2}[-i\gamma(\cdot)] = E \left\{ e^{\alpha \int_0^t \gamma(t') X^2(t') dt'} \right\}$$

- CAMERON & MARTIN (BULL. AM. MATH. SOC. 1945
TRANS " " " 1949)

— EVALUATED USING LINEAR TRANSFORMATIONS OF SPACE OF SAMPLE PATHS, $X(\cdot)$, ONTO ITSELF

-- related to sol'n of certain STURM-LIOUVILLE D.E.'s

$$f''(x) + \alpha \gamma(x) f(x) = 0 \quad \exists f_1(0) = f_1'(x) = 0$$

$$\Rightarrow C_{X^2}[-i\gamma(\cdot)] = \left[\frac{f_{\alpha}(t)}{f_{\alpha}(0)} \right]^{1/2}$$

where $-\infty < \alpha < \infty$

$f_{\alpha}(\cdot)$ depends on $\gamma(\cdot)$

- For $\gamma = -1$ C & M showed

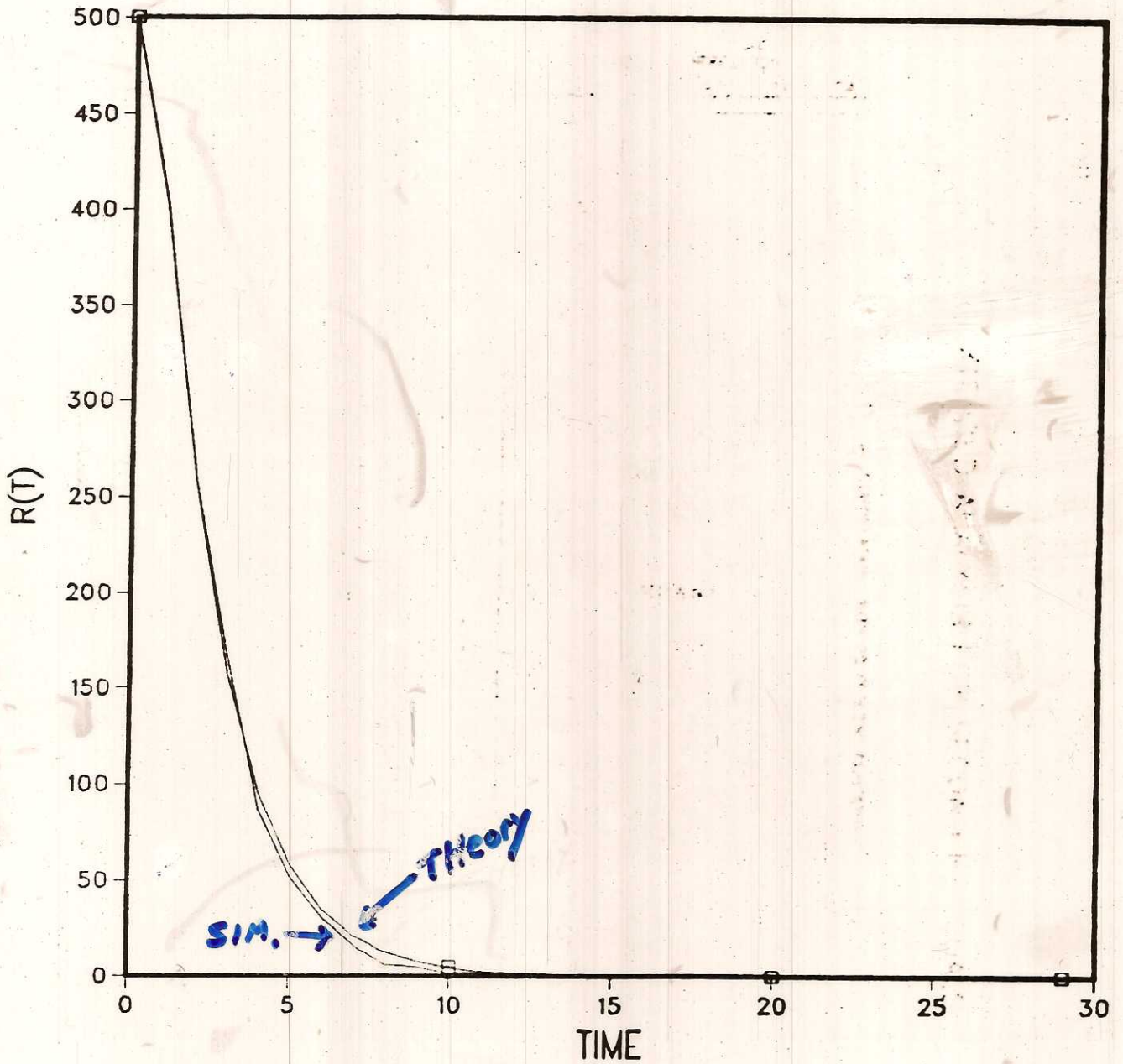
$$C_{X^2}[i] = \left[\frac{1}{\cosh(\sqrt{\alpha} t)} \right]^{1/2}$$

- $\left[R(t) = e^{-m\lambda t} \frac{1}{\sqrt{\cosh(\sqrt{m\alpha} t)}} \right]$

For m -comp's in series

RELIABILITY

(WIENER PROCESS)²



$$h(t) = \alpha x^2$$

$$X \leftarrow X + \sqrt{2 \cdot D \cdot dt} * N$$

$$\text{IF } (U \leq \alpha x^2 dt) \Rightarrow \text{FAIL}$$

$$dt = .01$$

$$\sigma^2 = 2D = 1/2$$

$$\alpha = 1.$$

STATIONARY-GAUSSIAN ENVIRONMENT

- ORNSTEIN-UHLENBECK PROCESS

- $\dot{x} + \lambda x = \eta(t)$

e.g. $L \frac{d^2 i}{dt^2} + R i = v(t)$

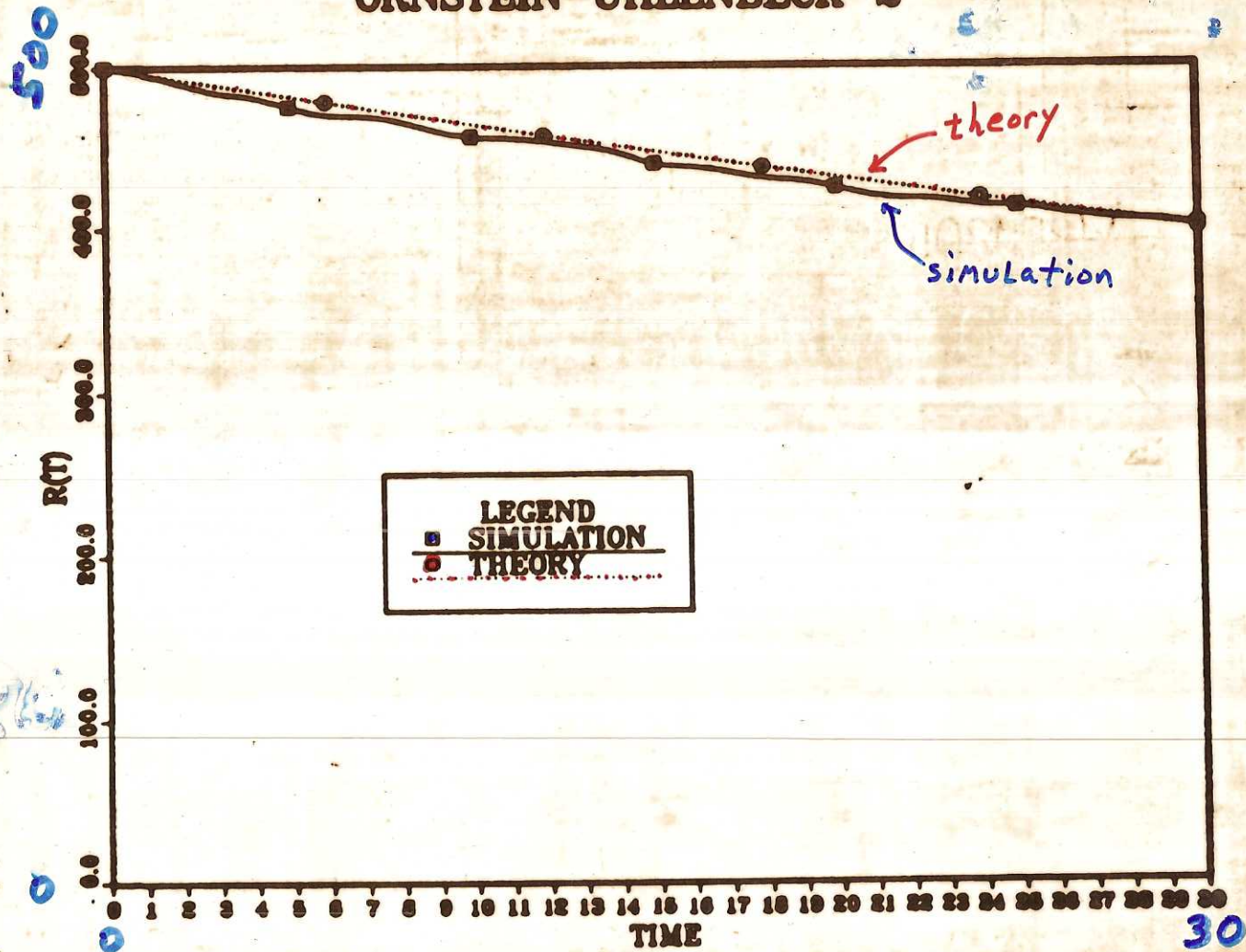
- $P(x, t | x_0, t_0) = \left[\frac{\lambda}{\pi \sigma^2 (1 - e^{-2\lambda t})} \right]^{1/2} \exp \left\{ -\frac{\lambda}{\sigma^2} \frac{(x - x_0 e^{-\lambda t})^2}{(1 - e^{-2\lambda t})} \right\}$

- HAZARD rate, $h(t) = \alpha x^2(t)$

- $R(t) = E \left\{ e^{-\alpha \int_0^t x^2 dt} \right\}_{x_0=0}$

$$R(t) = \frac{e^{\frac{\lambda}{2} t} (\lambda^2 + 2\sigma^2 \alpha)^{1/4}}{\left[\lambda \sinh(\sqrt{\lambda^2 + 2\sigma^2 \alpha} t) + \sqrt{\lambda^2 + 2\sigma^2 \alpha} \cosh(\sqrt{\lambda^2 + 2\sigma^2 \alpha} t) \right]^{1/2}}$$

ORNSTEIN-UHLENBECK²



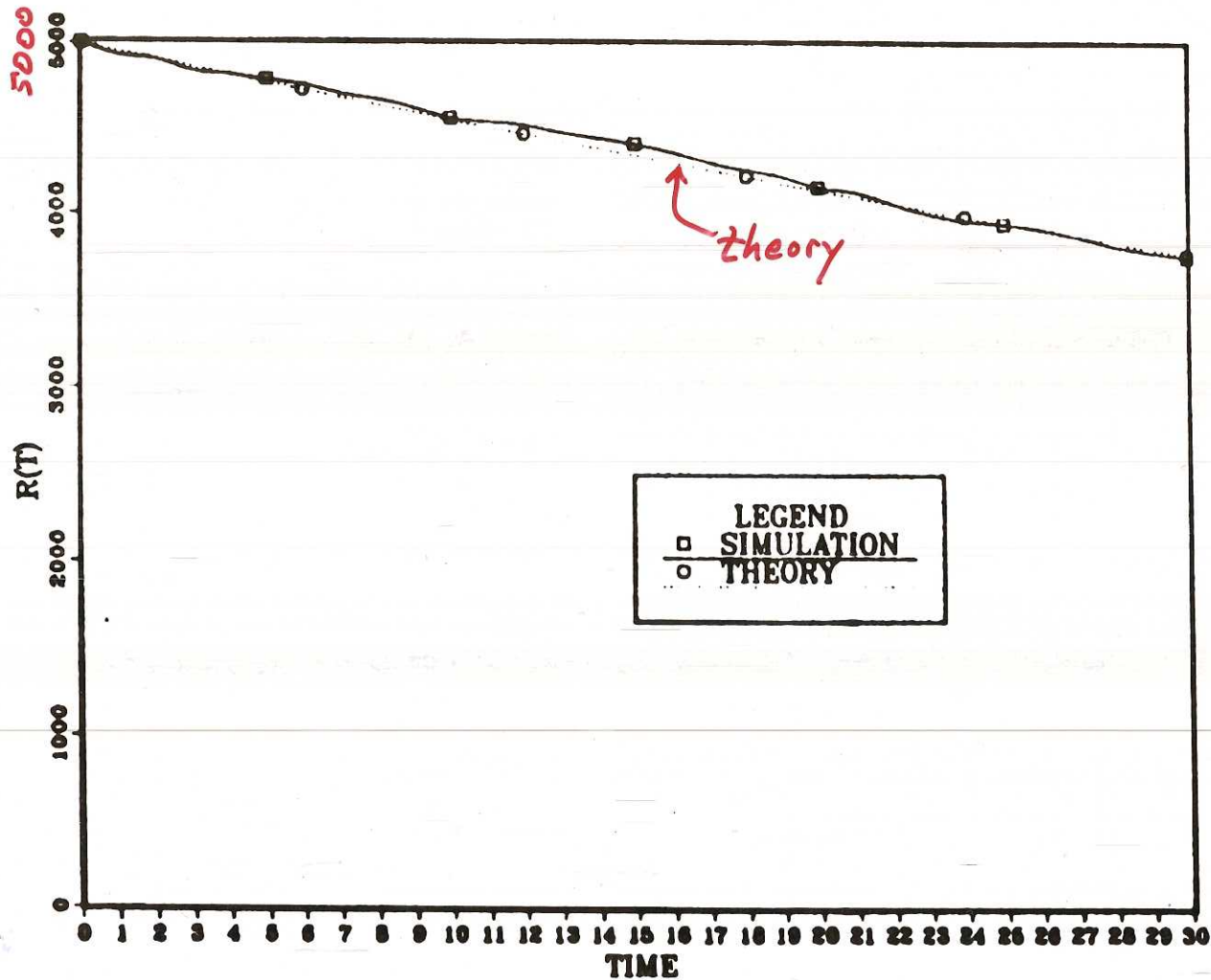
$$X \leftarrow X e^{-\lambda dt} + \sqrt{2 \cdot D \cdot dt} N$$

$$dt = .003$$

$$\sigma^2 = \frac{1}{2} = 2D$$

$$\lambda = 33.3$$

OFFSET ORNSTEIN-UHLENBECK**2



$$\Delta X + \lambda (X - \hat{x}) \Delta t = \Delta w$$

$$\hat{x} = \frac{1}{2} \sqrt{\text{Var}(O.U.)}$$