

ME 529 - Stochastics

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Markov Processes

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Markov Processes

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Outline

- **Markov Processes**
- **Important Properties**
- **Chapman-Kolmogorov-Equation**
- **Fokker-Planck Equation**
- **Fokker-Planck Equation for Vector Markov Processes**
- **Fokker-Planck Equation for Ito's Equation**

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Markov Processes

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A stochastic process $X(t)$ is called a **Markov process** if for every n and any $t_1 < t_2 < \dots < t_n$, its conditional probability satisfies the following condition

$$P\{X(t_n) \leq x_n | X(t_{n-1}), X(t_{n-2}), \dots, X(t_1)\} = P\{X(t_n) \leq x_n | X(t_{n-1})\}$$

or

$$P\{X(t_n) \leq x_n | X(t) \text{ for all } t \leq t_{n-1}\} = P\{X(t_n) \leq x_n | X(t_{n-1})\}$$

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Important Properties

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- i) **$X(t)$ is also Markov in reverse:**
 $P\{X(t_1) \leq x_1 | X(t) \text{ for all } t \geq t_2\} = P\{X(t_1) \leq x_1 | X(t_2)\}$
- ii) **The future is independent of the past under the given condition of the present.**
- iii) **If for any $t_1 < t_2$, $X(t_2) - X(t_1)$ is independent of $X(t)$ for $t \leq t_1$, the $X(t)$ is a Markov process.**
Thus, independent increment processes (Poisson, Wiener-Levy) are Markov processes.

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Important Properties

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iv) $E\{X(t_n) | X(t_{n-1}), \dots, X(t_1)\} = E\{X(t_n) | X(t_{n-1})\}$

v) **X(t) is associated with a 1st order equation,**

$$\frac{dX}{dt} - \beta(X, t) = j(t) \quad j(t) = \frac{dW}{dt}$$

with solution $X(t) = X(t_0) + \int_{t_0}^t \beta(X(\tau), \tau) d\tau + \int_{t_0}^t j(\tau) d\tau$

vi) **Conditional probability density satisfies**

$$f_X(x_n, t_n | x_{n-1}, t_{n-1}; \dots; x_2, t_2; x_1, t_1) = f_X(x_n, t_n | x_{n-1}, t_{n-1})$$

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Important Properties

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vi) **The following relation holds**

$$f_X(x_1, t_1; x_2, t_2; x_3, t_3) = f_X(x_3, t_3 | x_2, t_2) f_X(x_2, t_2 | x_1, t_1) f_X(x_1, t_1)$$

$$f_X(x_1, t_1; \dots; x_n, t_n) = f_X(x_n, t_n | x_{n-1}, t_{n-1}) \dots f_X(x_2, t_2 | x_1, t_1) f_X(x_1, t_1)$$

If $X(0) = x_o$ then $f_X(x_1, t_1) = f_X(x_1, t_1 / x_o, 0)$

A Markov Process is fully specified if:

- a) Given 1st order density and transition probability density;
- b) 2nd order density;
- c) Transition density and $X(0)$

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Chapman-Kolmogorov-Smoluchowski Equation

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For Continuous Random Process

$$\begin{aligned} f(x, t | x_0, t_0) &= \int_{-\infty}^{+\infty} f(x, t; x_1, t_1 | x_0, t_0) dx_1 \\ &= \int_{-\infty}^{+\infty} f(x, t | x_1, t_1; x_0, t_0) f(x_1, t_1 | x_0, t_0) dx_1 \end{aligned}$$

For Markov Processes

$$f(x, t | x_1, t_1; x_0, t_0) = f(x, t | x_1, t_1)$$

➡ $f(x, t | x_0, t_0) = \int_{-\infty}^{+\infty} f(x, t | x_1, t_1) f(x_1, t_1 | x_0, t_0) dx_1$

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Fokker-Planck Equation

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$$\frac{\partial f}{\partial t} = - \frac{\partial}{\partial x} [\alpha_1(x, t) f] + \frac{1}{2} \frac{\partial^2}{\partial x^2} [\alpha_{11}(x, t) f]$$

$$\alpha_1(x, t) = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{dX(t) | X(t) = x\}$$

$$\alpha_{11}(x, t) = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{[dX(t)]^2 | X(t) = x\}$$

Kolmogorov Equation

$$\frac{\partial f}{\partial t_0} + \alpha_1(x_0, t_0) \frac{\partial f}{\partial x_0} + \frac{\alpha_{11}(x_0, t_0)}{2} \frac{\partial^2 f}{\partial x_0^2} = 0$$

$$f = f(x, t | x_0, t_0)$$

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Example: Wiener-Levy Process

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Recall

$$E\{W(t)\} = 0$$

$$E\{W^2(t)\} = 2Dt$$

$$R_{WW}(t_1, t_2) = \begin{cases} 2Dt_1 & t_2 \geq t_1 \\ 2Dt_2 & t_1 \geq t_2 \end{cases}$$

$$E\{W(t_2) - W(t_1)\} = 0$$

$$E\{[W(t_2) - W(t_1)]^2\} = 2D(t_2 - t_1)$$

$$t_2 > t_1$$

Let

$$t_2 = t + dt$$

$$t_1 = t$$

$$dW = W(t+dt) - W(t)$$

$$\Rightarrow E\{dW\} = 0$$

$$E\{(dW)^2\} = 2Ddt$$

$$\Rightarrow \alpha_1 = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{dW | W\} = 0$$

$$\alpha_{11} = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{(dW)^2 | W\} = 2D$$

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Example: Wiener-Levy Process

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Corresponding Fokker-Planck Equation

$$\frac{\partial f}{\partial t} = -\frac{\partial}{\partial x}(\alpha_1 f) + \frac{1}{2} \frac{\partial^2}{\partial x^2}(\alpha_{11} f) = D \frac{\partial^2 f}{\partial x^2} \quad \Rightarrow \quad \frac{\partial f}{\partial t} = D \frac{\partial^2 f}{\partial W^2}$$

For $f(w, t_o / w_o, t_o) = \delta(w - w_o)$ **solution is**

$$f = \frac{e^{-\frac{(w-w_o)^2}{4D(t-t_o)}}}{\sqrt{4\pi D(t-t_o)}}$$

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Fokker-Planck Equation For Vector-Markov Process

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$$\frac{\partial f}{\partial t} = -\sum_{j=1}^n \frac{\partial}{\partial x_j} [\alpha_j(\mathbf{x}, t)f] + \frac{1}{2} \sum_i \sum_j \frac{\partial^2}{\partial x_i \partial x_j} [\alpha_{ij}(\mathbf{x}, t)f]$$

$$\alpha_j = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{dX_j(t) | \mathbf{X}(t) = \mathbf{x}\}$$

$$\alpha_{ij} = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{dX_i(t)dX_j(t) | \mathbf{X}(t) = \mathbf{x}\}$$

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Ito's Equation

$$\frac{d\mathbf{X}}{dt} = \mathbf{g}(\mathbf{X}, t) + \mathbf{G}(\mathbf{x}, t) \cdot \mathbf{n}(t)$$

or

$$\frac{dX_i}{dt} = g_i(\mathbf{X}, t) + \sum_j G_{ij}(\mathbf{X}, t)n_j(t)$$

$$d\mathbf{X} = \mathbf{g}(\mathbf{X}, t)dt + \mathbf{G}(\mathbf{x}, t) \cdot d\mathbf{W}$$

**Here n and W being vector white noise and
Wiener processes with**

$$E\{n_i\} = E\{dW_i\} = 0$$

$$E\{n_i(t+\tau)n_j(t)\} = 2D_{ij}\delta(\tau)$$

$$E\{dW_i dW_j\} = 2D_{ij}dt$$

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Fokker-Planck Equation

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Fokker-Planck Equation for Ito's Equation

$$a_j = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{dX_j | \mathbf{X} = \mathbf{x}\} = g_j(x, t) \quad a_{ij} = \lim_{dt \rightarrow 0} \frac{1}{dt} E\{dX_i dX_j | \mathbf{X} = \mathbf{x}\}$$

$$dX_i dX_j = g_i g_j (dt)^2 + g_i \sum_k G_{jk} dW_k dt + g_j \sum_k G_{ik} dW_k dt + \sum_k \sum_\ell G_{ik} G_{j\ell} dW_k dW_\ell$$

$$\alpha_{ij} = 2 \sum_k \sum_\ell G_{ik} G_{j\ell} D_{k\ell} = 2(\mathbf{G} \cdot \mathbf{D} \cdot \mathbf{G}^T)_{ij}$$

$$\frac{\partial f}{\partial t} = - \sum_j \frac{\partial}{\partial x_j} [g_j(\mathbf{x}, t) f] + \sum_i \sum_j \frac{\partial^2}{\partial x_i \partial x_j} [(\mathbf{G} \cdot \mathbf{D} \cdot \mathbf{G}^T)_{ij} f]$$

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Concluding Remarks

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- Important Properties
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Example: 1st Order System

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Consider

$$\frac{dx}{dt} = g(x, t) + G(x, t) \frac{dW}{dt} \quad E\{(dW)^2\} = 2Ddt$$

Fokker-Planck

$$\frac{\partial f}{\partial t} = - \frac{\partial}{\partial x} (gf) + D \frac{\partial^2}{\partial x^2} (G^2 f)$$

Stationary

$$\frac{d}{dx} \left[-gf + D \frac{d}{dx} (G^2 f) \right] = 0 \quad -gf + D \frac{d}{dx} (G^2 f) = c_1 = 0$$

Let

$$G^2 f = F \quad D \frac{dF}{dx} = \frac{g}{G^2} F \quad \frac{dF}{F} = \frac{g}{DG^2} dx$$

$$F = C \exp \left\{ + \int_0^x \frac{g(x_1)}{DG^2(x_1)} dx_1 \right\} \Rightarrow f = \frac{c}{G^2} \exp \left\{ \int_0^x \frac{g(x_1)}{DG^2(x_1)} dx_1 \right\}$$

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Thank you!

Questions?

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