

# Stochastic Structural Dynamics

## Lecture-23

### **Markov Vector Approach-3**

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## General: $n$ -dimensional Ito SDE

$$dX(t) = f[t, X(t)]dt + G[t, X(t)]dB(t); t \geq 0; X(0) = X_0$$

$$X(t), f[t, X(t)] \sim n \times 1$$

$$G[t, X(t)] \sim n \times m$$

$$dB(t) \sim m \times 1$$

$$\langle dB(t) \rangle = 0; \langle \Delta B_i(t) \Delta B_j(t + \tau) \rangle = 2D_{ij}\delta(\tau)$$

$$\alpha_j = f_j[t, \tilde{x}]; j = 1, 2, \dots, n$$

$$\alpha_{ij} = 2 \left[ G D G^t \right]_{ij}(\tilde{x}); i, j = 1, 2, \dots, m$$

$$\frac{\partial p}{\partial t} = - \sum_{j=1}^n \frac{\partial}{\partial x_j} [\alpha_j p] + \frac{1}{2} \sum_{j=1}^m \sum_{k=1}^m \frac{\partial^2}{\partial x_j \partial x_k} [\alpha_{jk} p]$$

$$p(\tilde{x}; 0 | \tilde{x}_0; 0) = \prod_{i=1}^n \delta(x_i - x_{i0}) + \text{BCS}$$

## Remarks

- The application of Markov vector approach requires
- The excitations to be modeled as white noises or as outputs dynamical systems to white noise inputs.

The equations to be represented in the state space form.

# Further questions

- How to solve the FPK equations?
  - A few selected examples for which exact solutions are possible
- Can we derive equations governing the moments?

**Recall :** Lagrange's method for solving linear PDE-s.

Consider the PDE of the form

$$P(x, y, z) \frac{\partial z}{\partial x} + Q(x, y, z) \frac{\partial z}{\partial y} = R(x, y, z) \cdots (1)$$

To obtain an integral of the above equation we consider the auxiliary equation

$$\frac{dx}{P} = \frac{dy}{Q} = \frac{dz}{R}$$

Let two independent solutions of this equation be written as

$u(x, y, z) = a$  &  $v(x, y, z) = b$  where  $a$  and  $b$  are constants.

Then  $\phi(u, v) = 0$  is a solution of (1).

Alternatively,  $u = f(v)$  is also a solution.

## Illustration

$$xz \frac{\partial z}{\partial x} + yz \frac{\partial z}{\partial y} = xy$$

Auxiliary equation:  $\frac{dx}{xz} = \frac{dy}{yz} = \frac{dz}{xy}$

Consider  $\frac{dx}{xz} = \frac{dy}{yz} \Rightarrow \frac{y}{x} = a$

Consider  $\frac{dx}{xz} = \frac{dz}{xy} \Rightarrow z^2 - xy = b$

General solution:  $\phi\left(\frac{y}{x}, z^2 - xy\right) = 0$

## Example : Linear MDOF systems

$$M\ddot{X} + C\dot{X} + KX = \Gamma W(t); t \geq 0; X(0) = X_0; \dot{X}(0) = \dot{X}_0$$

$$X(t) \sim N \times 1; \Gamma \sim n \times m; W(t) \sim m \times 1$$

$$\langle W(t) \rangle = 0; \langle W(t)W^t(t+\tau) \rangle = [2D_{ij}]_{m \times m} \delta(\tau)$$

$$\ddot{X} + M^{-1}C\dot{X} + M^{-1}KX = M^{-1}\Gamma W(t)$$

$$Y = \begin{Bmatrix} Y_I \\ Y_{II} \end{Bmatrix} = \begin{Bmatrix} X \\ \dot{X} \end{Bmatrix}$$

$$dY_I = Y_{II} dt$$

$$dY_{II} = -M^{-1}CY_{II} - M^{-1}KY_I + M^{-1}\Gamma dB(t)$$

$$dY(t) = -PYdt + QdB(t) t \geq 0; Y(0) = Y_0$$

$$P = \begin{bmatrix} 0 & -I \\ M^{-1}K & M^{-1}C \end{bmatrix}; Q = \begin{bmatrix} 0 \\ M^{-1}\Gamma \end{bmatrix}$$

$$\underbrace{\begin{Bmatrix} dY_I(t) \\ dY_{II}(t) \end{Bmatrix}}_{2N \times 1} = \underbrace{\begin{bmatrix} 0 & -I \\ M^{-1}K & M^{-1}C \end{bmatrix}}_{2N \times 2N} \underbrace{\begin{Bmatrix} Y_I(t) \\ Y_{II}(t) \end{Bmatrix}}_{2N \times 1} dt + \underbrace{\begin{bmatrix} 0 \\ M^{-1}\Gamma \end{bmatrix}}_{2N \times m} dB(t)$$

$$dY(t) = -PYdt + QdB(t) \quad t \geq 0; \quad Y(0) = Y_0$$

Consider the eigenvalue problem

$$P\phi = \lambda\phi$$

Let  $\Phi$  be the  $2N \times 2N$  matrix of eigenvectors and  $\Lambda$  be the  $2N \times 2N$  diagonal matrix of complete set of eigenvalues of  $P$ .

$$\Rightarrow P\Phi = \Phi\Lambda \Rightarrow \Phi^{-1}P\Phi = \Lambda$$

Introduce the transformation  $Y(t) = \Phi Z(t)$

$$Y(t) = \Phi Z(t)$$

$$\Phi dZ(t) = -P\Phi Z(t)dt + QdB(t)$$

$$\Rightarrow dZ(t) = -\Phi^{-1}P\Phi Z(t)dt + \Phi^{-1}QdB(t)$$

$$\Rightarrow dZ(t) = -\Lambda Z(t)dt + GdB(t)$$

$$\Rightarrow \{\alpha_j\} = -\Lambda z; \quad [\alpha_{ij}] = [GDG^t]$$

$$\frac{\partial p}{\partial t} = \sum_{j=1}^{2N} \frac{\partial}{\partial z_j} [\lambda_j z_j p] + \frac{1}{2} \sum_{j=1}^m \sum_{k=1}^m \alpha_{jk} \frac{\partial^2 p}{\partial z_j \partial z_k};$$

$$p \equiv p(\tilde{z}; t | \tilde{z}_0; 0)$$

$$\text{ICS: } p(z; 0 | \tilde{z}_0; 0) = \prod_{i=1}^{2N} \delta(z_i - z_{i0})$$

$$BCS : \lim_{z_i \rightarrow \pm\infty} p(\tilde{z}; t | \tilde{z}_0; 0) \rightarrow 0 \forall i = 1, 2, \dots, 2N$$

$$\frac{\partial p}{\partial t} = \sum_{j=1}^{2N} \frac{\partial}{\partial z_j} [\lambda_j z_j p] + \frac{1}{2} \sum_{j=1}^m \sum_{k=1}^m \alpha_{jk} \frac{\partial^2 p}{\partial z_j \partial z_k}$$

Define the conditional characteristic function

$$M[\theta_1, \theta_2, \dots, \theta_{2N}; t | Z(0) = z_0] = M(\tilde{\theta}; t) =$$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} p(\tilde{z}; t | \tilde{z}_0; 0) \exp\left(i \sum_{j=1}^{2N} \theta_j z_j\right) dz_1 dz_2 \cdots dz_{2N}$$

$$= \int_{-\infty}^{\infty} p(\tilde{z}; t | \tilde{z}_0; 0) \exp\left(i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{z}$$

$$\frac{\partial M(\tilde{\theta}; t)}{\partial \theta_k} = \int_{-\infty}^{\infty} z_k p(\tilde{z}; t | \tilde{z}_0; 0) \exp\left(i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{z}$$

$$\begin{aligned}
M(\tilde{\theta}; t) &= \int_{-\infty}^{\infty} p(\tilde{z}; t | \tilde{z}_0; 0) \exp\left(i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{z} \\
&\Rightarrow p(\tilde{z}; t | \tilde{z}_0; 0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} M(\tilde{\theta}; t) \exp\left(-i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{\theta} \\
\frac{\partial M(\tilde{\theta}; t)}{\partial \theta_k} &= \int_{-\infty}^{\infty} i z_k p(\tilde{z}; t | \tilde{z}_0; 0) \exp\left(i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{z} \\
&\Rightarrow z_k p(\tilde{z}; t | \tilde{z}_0; 0) = \frac{1}{2\pi i} \int_{-\infty}^{\infty} \frac{\partial M(\tilde{\theta}; t)}{\partial \theta_k} \exp\left(-i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{\theta} \\
\frac{\partial}{\partial z_k} [z_k p(\tilde{z}; t | \tilde{z}_0; 0)] &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{\partial M(\tilde{\theta}; t)}{\partial \theta_k} (-i \theta_k) \exp\left(-i \sum_{j=1}^{2N} \theta_j z_j\right) d\tilde{\theta}
\end{aligned}$$

$$\frac{\partial p}{\partial t} = \sum_{j=1}^{2N} \frac{\partial}{\partial z_j} \left[ \lambda_j z_j p \right] + \frac{1}{2} \sum_{j=1}^m \sum_{k=1}^m \alpha_{jk} \frac{\partial^2 p}{\partial z_j \partial z_k}$$

$$p(\tilde{z};t \mid \tilde{z}_0;0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} M(\tilde{\theta};t) \exp \left( -i \sum_{l=1}^{2N} \theta_l z_l \right) d\tilde{\theta}$$

$$\frac{\partial p}{\partial t} = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{\partial M(\tilde{\theta};t)}{\partial t} \exp \left( -i \sum_{l=1}^{2N} \theta_l z_l \right) d\tilde{\theta}$$

$$\frac{\partial}{\partial z_j} \left[ z_j p \right] = \frac{1}{2\pi i} \int_{-\infty}^{\infty} \frac{\partial M(\tilde{\theta};t)}{\partial \theta_j} (-i\theta_j) \exp \left( -i \sum_{l=1}^{2N} \theta_l z_l \right) d\tilde{\theta}$$

$$\frac{\partial^2 p}{\partial z_j \partial z_k} = \frac{1}{2\pi} \int_{-\infty}^{\infty} M(\tilde{\theta};t) (-i\theta_j) (-i\theta_k) \exp \left( -i \sum_{l=1}^{2N} \theta_l z_l \right) d\tilde{\theta}$$

$$\frac{\partial p}{\partial t} = \sum_{j=1}^{2N} \frac{\partial}{\partial z_j} [\lambda_j z_j p] + \frac{1}{2} \sum_{j=1}^m \sum_{k=1}^m \alpha_{jk} \frac{\partial^2 p}{\partial z_j \partial z_k}$$

$$\Rightarrow \frac{\partial M}{\partial t} = - \sum_{j=1}^{2N} \lambda_j \theta_j \frac{\partial M}{\partial \theta_j} - \frac{1}{2} \sum_{j=1}^{2N} \sum_{k=1}^{2N} \alpha_{jk} \theta_j \theta_k M$$

$\Rightarrow$

$$\frac{dt}{1} = \frac{d\theta_1}{\lambda_1 \theta_1} = \frac{d\theta_2}{\lambda_2 \theta_2} = \dots = \frac{d\theta_{2N}}{\lambda_{2N} \theta_{2N}} = - \frac{dM}{\frac{1}{2} \sum_{j=1}^{2N} \sum_{k=1}^{2N} \alpha_{jk} \theta_j \theta_k M}$$

$$\frac{dt}{1} = \frac{d\theta_i}{\lambda_i \theta_i}; i = 1, 2, \dots, 2N \Rightarrow \theta_i(t) = \theta_{i0} \exp(\lambda_i t); i = 1, 2, \dots, 2N$$

$$\Rightarrow \{\theta_{i0}\} = [\Omega] \{\theta_i(t)\}$$

$\Omega$  = diagonal matrix with entries  $\exp(-\lambda_i t)$

Consider

$$\frac{d\theta_i}{\lambda_i \theta_i} = -\frac{dM}{\left(\frac{1}{2}\right) \sum_{j=1}^{2N} \sum_{k=1}^{2N} \alpha_{jk} \theta_j \theta_k M} \dots (a) \quad \& \quad \frac{d\theta_l}{\lambda_l \theta_l} = -\frac{dM}{\left(\frac{1}{2}\right) \sum_{j=1}^{2N} \sum_{k=1}^{2N} \alpha_{jk} \theta_j \theta_k M} \dots (b)$$

$$(a) \lambda_i \theta_i \theta_l + (b) \lambda_l \theta_i \theta_l \Rightarrow \theta_l d\theta_i + \theta_i d\theta_l = -\frac{(\lambda_i + \lambda_l) \theta_l \theta_i dM}{\left(\frac{1}{2}\right) \sum_{j=1}^{2N} \sum_{k=1}^{2N} \alpha_{jk} \theta_j \theta_k M}$$

$$\frac{d(\theta_l \theta_i)}{(\lambda_i + \lambda_l)} = -\frac{\theta_l \theta_i dM}{\left(\frac{1}{2}\right) \sum_{j=1}^{2N} \sum_{k=1}^{2N} \alpha_{jk} \theta_j \theta_k M}$$

Multiply both sides by  $\frac{\alpha_{il}}{2}$  and sum over  $i$  and  $l$

$$\frac{dM}{M} = -\sum_{i=1}^{2N} \sum_{l=1}^{2N} \frac{1}{2} \frac{\alpha_{il}}{(\lambda_i + \lambda_l)} d(\theta_l \theta_i)$$

$$\frac{dM}{M} = - \sum_{i=1}^{2N} \sum_{l=1}^{2N} \frac{1}{2} \frac{\alpha_{il}}{(\lambda_i + \lambda_l)} d(\theta_l \theta_i)$$

$$\Rightarrow M(\tilde{\theta}, t) = M_0 \exp \left[ - \sum_{i=1}^{2N} \sum_{l=1}^{2N} \frac{1}{2} \frac{\alpha_{il} \theta_l \theta_i}{(\lambda_i + \lambda_l)} \right]$$

Define  $\Xi = \begin{bmatrix} \alpha_{il} \\ (\lambda_i + \lambda_l) \end{bmatrix} \Rightarrow$

$$M(\tilde{\theta}, t) = M_0 \exp \left[ -\frac{1}{2} \tilde{\theta}^t \Xi \tilde{\theta} \right]$$

We have  $p(\tilde{z}; t | z_0; 0) = \prod_{i=1}^{2N} \delta(z_i - z_{i0}) \& \{\theta_{i0}\} = [\Omega] \{\theta_i(t)\}$

$$M(\tilde{\theta}; 0) = \int_{-\infty}^{\infty} \prod_{i=1}^{2N} \delta(z_i - z_{i0}) \exp \left( i \sum_{j=1}^{2N} \theta_j z_j \right) d\tilde{z} = \exp(i \theta_0^t z_0)$$

$$\Rightarrow M_0 \exp\left[-\frac{1}{2}\tilde{\theta}_0^t \Xi \tilde{\theta}_0\right] = \exp(i\theta_0^t z_0)$$

$$M_0 = \exp\left(i\theta_0^t z_0 + \frac{1}{2}\tilde{\theta}_0^t \Xi \tilde{\theta}_0\right)$$

$$M(\tilde{\theta}, t) = \exp\left[i\theta_0^t z_0 + \frac{1}{2}\tilde{\theta}^t \Omega^t \Xi \Omega \tilde{\theta} - \frac{1}{2}\tilde{\theta}^t \Xi \tilde{\theta}\right]$$

$$= \exp\left[i\theta_0^t z_0 - \frac{1}{2}\tilde{\theta}^t (\Xi - \Omega^t \Xi \Omega)\tilde{\theta}\right]$$

This is the characteristic function of a multivariate Gaussian PDF.

The mean vector and covariance matrix can be evaluated from the characteristic function.

The PDF in the original coordinate system can be obtained by using the transformation  $Y(t) = \Phi Z(t)$ .

## Remarks

- For linear systems, the exact solution can also be obtained using convolution integral approach discussed earlier in this course. The Markov vector approach does not offer any special advantage here.
- The above formulation is also valid when excitations are modeled as filtered white noise excitations.

$$\ddot{x} + 2\eta\omega\dot{x} + \omega^2 x = f(t); t \geq 0; x(0) = x_0; \dot{x}(0) = \dot{x}_0$$

$$\ddot{f} + 2\xi\lambda\dot{f} + \lambda^2 f = w(t); t \geq 0; f(0) = f_0; \dot{f}(0) = \dot{f}_0$$

$$\langle w(t) \rangle = 0; \langle w(t) w(t + \tau) \rangle = 2D\delta(\tau)$$

$$X(t) = \begin{Bmatrix} x(t) & \dot{x}(t) & f(t) & \dot{f}(t) \end{Bmatrix}^T \text{ is Markov}$$

$$dX(t) = -PXdt + QdB(t) t \geq 0; X(0) = X_0$$

**Example:** First order nonlinear systems

$$\dot{x} + \beta(x) = w(t); t \geq 0 \text{ & } x(0) = x_0$$

$$\langle w(t) \rangle = 0; \langle w(t_1) w(t_2) \rangle = 2D\delta(t_1 - t_2)$$

$$dx = -\beta[x(t)]dt + dB(t)$$

$$\frac{\partial p}{\partial t} = \frac{\partial [\beta(x)p]}{\partial x} + D \frac{\partial^2 p}{\partial x^2}$$

**Transient solutions** by eigenfunction expansion.

$$\text{As } t \rightarrow \infty, \frac{\partial p}{\partial t} \rightarrow 0 \Rightarrow$$

$$\text{Stationary solution : } \frac{d[\beta(x)p]}{dx} + D \frac{d^2 p}{dx^2} = 0$$

$$\frac{d[\beta(x)p]}{dx} + D \frac{d^2 p}{dx^2} = 0; \lim_{x \rightarrow \pm\infty} p(x) \rightarrow 0$$

$$D \frac{dp}{dx} = -\beta(x)p$$

$$\Rightarrow p(x) = C \exp \left[ -\frac{1}{D} \int_0^x \beta(s) ds \right]; -\infty < x < \infty$$

Select C such that  $\int_{-\infty}^{\infty} p(x) dx = 1$

Example:  $\beta(x) = ax + bx^3$

$$p(x) = C \exp \left[ -\frac{1}{D} \left( \frac{ax^2}{2} + \frac{bx^4}{4} \right) \right]; -\infty < x < \infty$$

•  $b = 0 \Rightarrow$  pdf is Gaussian, as it should be.

**Example :** sdof system with nonlinear damping and nonlinear stiffness

$$\ddot{x} + \dot{x}f(H) + g(x) = w(t); t \geq 0; x(0) = x_0; \dot{x}(0) = \dot{x}_0.$$

$$\langle w(t) \rangle = 0; \langle w(t)w(t+\tau) \rangle = 2D\delta(\tau)$$

$$H = \frac{\dot{x}^2}{2} + \int_0^x g(u) du = \text{Total energy}$$

$$\begin{Bmatrix} X_1(t) \\ X_2(t) \end{Bmatrix} = \begin{Bmatrix} x(t) \\ \dot{x}(t) \end{Bmatrix}$$

$$dX_1(t) = X_2(t) dt$$

$$dX_2(t) = \left[ -X_2 \cancel{g}(H) - g(X_1) \right] dt + dB(t)$$

$$H = \frac{X_2^2}{2} + \int_0^{X_1} g(u) du$$

$$\frac{\partial p}{\partial t} = -x_2 \frac{\partial p}{\partial x_1} - \frac{\partial}{\partial x_2} \left[ \left\{ -x_2 F(H) - g(x_1) \right\} p \right] + D \frac{\partial^2 p}{\partial x_2^2}$$

$$p \equiv p(x_1, x_2; t | x_0, \dot{x}_0) = tpdf$$

$$p(x_1, x_2; 0 | x_0, \dot{x}_0) = \delta(x_1 - x_0) \delta(x_2 - \dot{x}_0)$$

$$p(\pm\infty, x_2; t | x_0, \dot{x}_0) = p(x_1, \pm\infty; t | x_0, \dot{x}_0) = 0$$

## Steady state

$$-x_2 \frac{\partial p}{\partial x_1} - \frac{\partial}{\partial x_2} \left[ \left\{ -x_2 F(H) - g(x_1) \right\} p \right] + D \frac{\partial^2 p}{\partial x_2^2} = 0$$

$$-x_2 \frac{\partial p}{\partial x_1} - \frac{\partial \left[ -x_2 F(H) p \right]}{\partial x_2} + g(x_1) \frac{\partial p}{\partial x_2} + D \frac{\partial^2 p}{\partial x_2^2} = 0$$

$$-\underline{x_2} \frac{\partial p}{\partial x_1} - \underline{\frac{\partial [-x_2 F(H) p]}{\partial x_2}} + \underline{g(x_1)} \frac{\partial p}{\partial x_2} + D \frac{\partial^2 p}{\partial x_2^2} = 0$$

Look for solutions such that

$$-\underline{x_2} \frac{\partial p}{\partial x_1} + \underline{g(x_1)} \frac{\partial p}{\partial x_2} = 0$$

$$-\underline{\frac{\partial [-x_2 F(H) p]}{\partial x_2}} + \underline{D} \frac{\partial^2 p}{\partial x_2^2} = 0$$

$$\text{Consider } -x_2 \frac{\partial p}{\partial x_1} + g(x_1) \frac{\partial p}{\partial x_2} = 0$$

Solve this equation using Lagrange's method.

$$\text{Auxiliary equation: } \frac{dx_1}{-x_2} = \frac{dx_2}{g(x_1)} = \frac{dp}{0}$$

$$\frac{dx_1}{-x_2} = \frac{dx_2}{g(x_1)} = \frac{dp}{0}$$

$$\frac{dx_1}{-x_2} = \frac{dx_2}{g(x_1)} \Rightarrow \frac{x_2^2}{2} + \int_0^{x_1} g(u) du = a$$

$$\frac{dx_2}{g(x_1)} = \frac{dp}{0} \Rightarrow p = b$$

$$p = \Phi\left(\frac{x_2^2}{2} + \int_0^{x_1} g(u) du\right) = \Phi(H)$$

Now consider the equation  $-\frac{\partial[-x_2 F(H)p]}{\partial x_2} + D \frac{\partial^2 p}{\partial x_2^2} = 0$

$$\frac{\partial}{\partial x_2} \left\{ -x_2 F(H)p + D \frac{\partial p}{\partial x_2} \right\} = 0$$

$$\frac{\partial}{\partial x_2} \left\{ -x_2 F(H) p + D \frac{\partial p}{\partial x_2} \right\} = 0$$

$$\Rightarrow x_2 F(H) \Phi(H) + D \frac{\partial \Phi}{\partial H} x_2 = 0$$

$$\Rightarrow \frac{d\Phi}{dH} = -\frac{1}{D} F(H) \Phi(H)$$

$$\Rightarrow \Phi(H) = \Phi_0 \exp \left[ -\frac{1}{D} \int_0^H F(\xi) d\xi \right]$$

$$p(x_1, x_2; t) = \Phi_0 \exp \left[ -\frac{1}{D} \int_0^H F(\xi) d\xi \right]; -\infty < x_1, x_2 < \infty$$

$$H = \frac{x_2^2}{2} + \int_0^{x_1} g(u) du$$

### Case - 1 $F(H) = 2\eta\omega$

$$\ddot{x} + 2\eta\omega\dot{x} + g(x) = w(t); t \geq 0; x(0) = x_0; \dot{x}(0) = \dot{x}_0.$$

$$\langle w(t) \rangle = 0; \langle w(t)w(t+\tau) \rangle = 2D\delta(\tau)$$

$$H = \frac{\dot{x}^2}{2} + \int_0^x g(u) du = \text{Total energy}$$

$$p(x_1, x_2; t) = \Phi_0 \exp \left[ -\frac{1}{D} \int_0^H F(\xi) d\xi \right]; -\infty < x_1, x_2 < \infty$$

$$= \Phi_0 \exp \left[ -\frac{1}{D} (2\eta\omega H) \right]$$

$$p(x, \dot{x}; t) = \Phi_0 \exp \left[ -\frac{2\eta\omega}{D} \left( \frac{\dot{x}^2}{2} + \int_0^x g(u) du \right) \right]; -\infty < x, \dot{x} < \infty$$

Let  $g(x) = \omega^2 x + \alpha x^3 \Rightarrow$

$$\underline{p(x, \dot{x}; t)} = \underline{\underline{\Phi_0}} \exp \left[ -\frac{2\eta\omega}{D} \left( \frac{\dot{x}^2}{2} + \frac{\omega^2 x^2}{2} + \frac{\alpha x^4}{4} \right) \right]; -\infty < x, \dot{x} < \infty$$

$$\Phi_0 \text{ to be selected such that } \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p(x, \dot{x}; t) dxd\dot{x} = 1.$$

## Remarks

- This is an exact solution
- If  $\alpha=0$ , the result for linear sdof system under white noise is recovered.
- It can be verified that  $\underline{p(x_1, x_2; t)} = \underline{p(x_1; t)} \underline{p(x_2; t)}$
- We know that  $X(t)$  and  $\dot{X}(t)$  are uncorrelated and it turns out that  $X(t)$  and  $\dot{X}(t)$  are independent

## Remarks (continued)

- It can be verified that

$$p(x; t) = \Phi_{01} \exp \left[ -\frac{2\eta\omega}{D} \left( \frac{\omega^2 x^2}{2} + \frac{\alpha x^4}{4} \right) \right]; -\infty < x < \infty$$

*Non-Gaussian*

$$p(\dot{x}; t) = \Phi_{02} \exp \left[ -\frac{2\eta\omega}{D} \frac{\dot{x}^2}{2} \right]; -\infty < \dot{x} < \infty$$

*Gaussian*

- The velocity is a gaussian random variable while displacement is a nongaussian random variable.
- The velocity is not a Gaussian random process (if it were so, displacement would also be Gaussian).  
Thus velocity is a non-Gaussian random process with a first order pdf that is Gaussian.

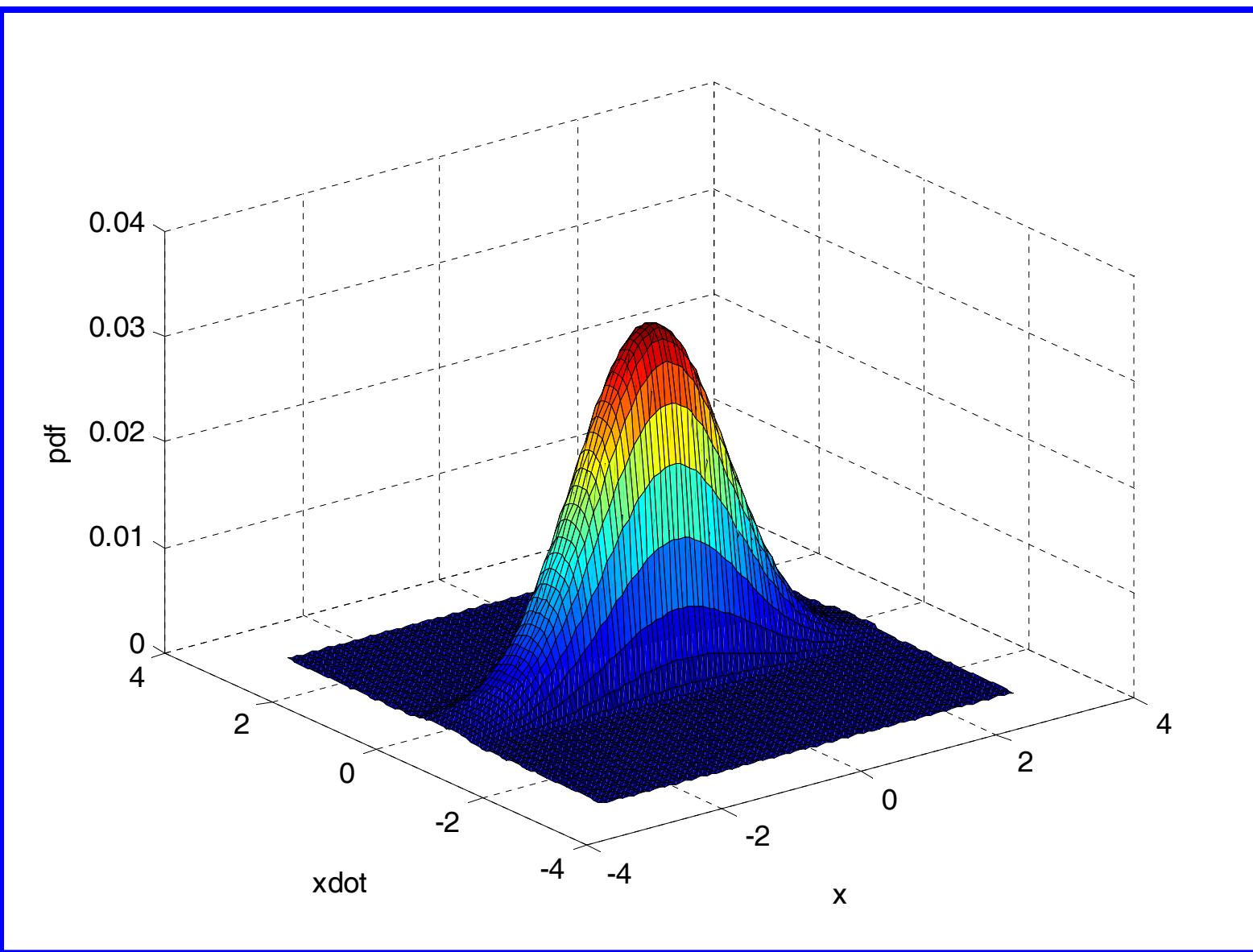
## Remarks (continued)

- $H = \underbrace{\frac{\dot{x}^2}{2}}_{KE} + \underbrace{\frac{\omega^2 x^2}{2}}_{PE} + \frac{\alpha x^4}{4}$  represents the total energy.

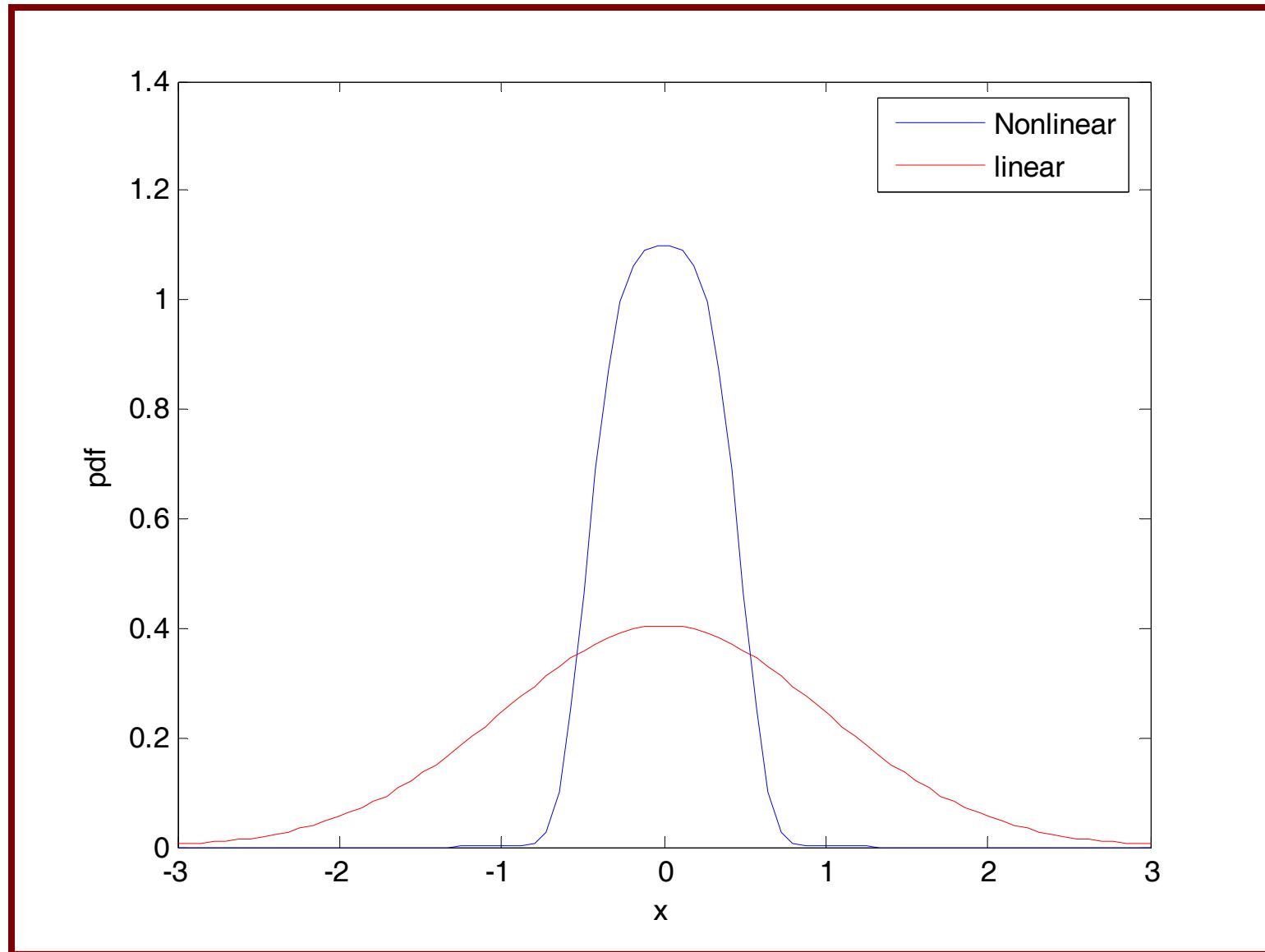
$$p(x_1, x_2; t) = \Phi_0 \exp\left[-\frac{2\eta\omega}{D}\left(\frac{\dot{x}^2}{2} + \frac{\omega^2 x^2}{2} + \frac{\alpha x^4}{4}\right)\right]; -\infty < x, \dot{x} < \infty$$

$\Rightarrow H$  is exponentially distributed

$$p_H(h; t) = \Phi_0 \exp\left(-\frac{2\eta\omega}{D} h\right); 0 < h < \infty$$



# pdf of displacement



## Remarks (continued)

- We have determined

$$p(x, \dot{x}; t) = \Phi_0 \exp \left[ -\frac{2\eta\omega}{D} \left( \frac{\dot{x}^2}{2} + \frac{\omega^2 x^2}{2} + \frac{\alpha x^4}{4} \right) \right]; -\infty < x, \dot{x} < \infty$$

Therefore we are in a position to study

- Number of times a level  $\varsigma$  is crossed in an interval 0 to  $T$

$$N(T) = \int_0^T |\dot{X}(t)| \delta[X(t) - \varsigma] dt$$

- PDF of first passage times

$$P[T_f(\varsigma) > t] = P[N^+(\varsigma, 0, T) = 0]$$

- PDF of extremes of  $X(t)$  over an interval 0 to  $T$

$$P_{X_m}(\varsigma) = P[X_m \leq \varsigma] = P[T_f(\varsigma) > T]$$

$$\begin{aligned}
 \langle n^+(\alpha, t) \rangle &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |\dot{y}| \delta[y - \alpha] p(y, \dot{y}) dy d\dot{y} \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |\dot{y}| \delta[y - \alpha] p(y) p(\dot{y}) dy d\dot{y} \\
 &= p_Y(\alpha) \int_{-\infty}^{\infty} |\dot{y}| p_{\dot{Y}}(\dot{y}) d\dot{y} \quad \checkmark
 \end{aligned}$$

**Case - 2**  $F(H) = -(1 - x^2 - \dot{x}^2)$ ;  $g(x) = x$

$$\ddot{x} - \dot{x}(1 - x^2 - \dot{x}^2) + x = w(t); t \geq 0; x(0) = x_0; \dot{x}(0) = \dot{x}_0.$$

$$\langle w(t) \rangle = 0; \langle w(t) w(t + \tau) \rangle = 2D\delta(\tau)$$

$$H = \frac{\dot{x}^2}{2} + \frac{x^2}{2} = \text{Total energy} \Rightarrow F(H) = -(1 - 2H)$$

$$p(x_1, x_2; t) = \Phi_0 \exp \left[ -\frac{1}{D} \int_0^H F(\xi) d\xi \right]; -\infty < x_1, x_2 < \infty$$

$$= \Phi_0 \exp \left[ \frac{1}{D} \int_0^H (1 - 2\xi) d\xi \right] = \Phi_0 \exp \left[ \frac{1}{D} (H - H^2) \right]$$

$$p(x, \dot{x}; t) = \Phi_0 \exp \left[ \frac{1}{D} \left\{ \left( \frac{\dot{x}^2}{2} + \frac{x^2}{2} \right) - \left( \frac{\dot{x}^2}{2} + \frac{x^2}{2} \right)^2 \right\} \right]; -\infty < x_1, x_2 < \infty$$

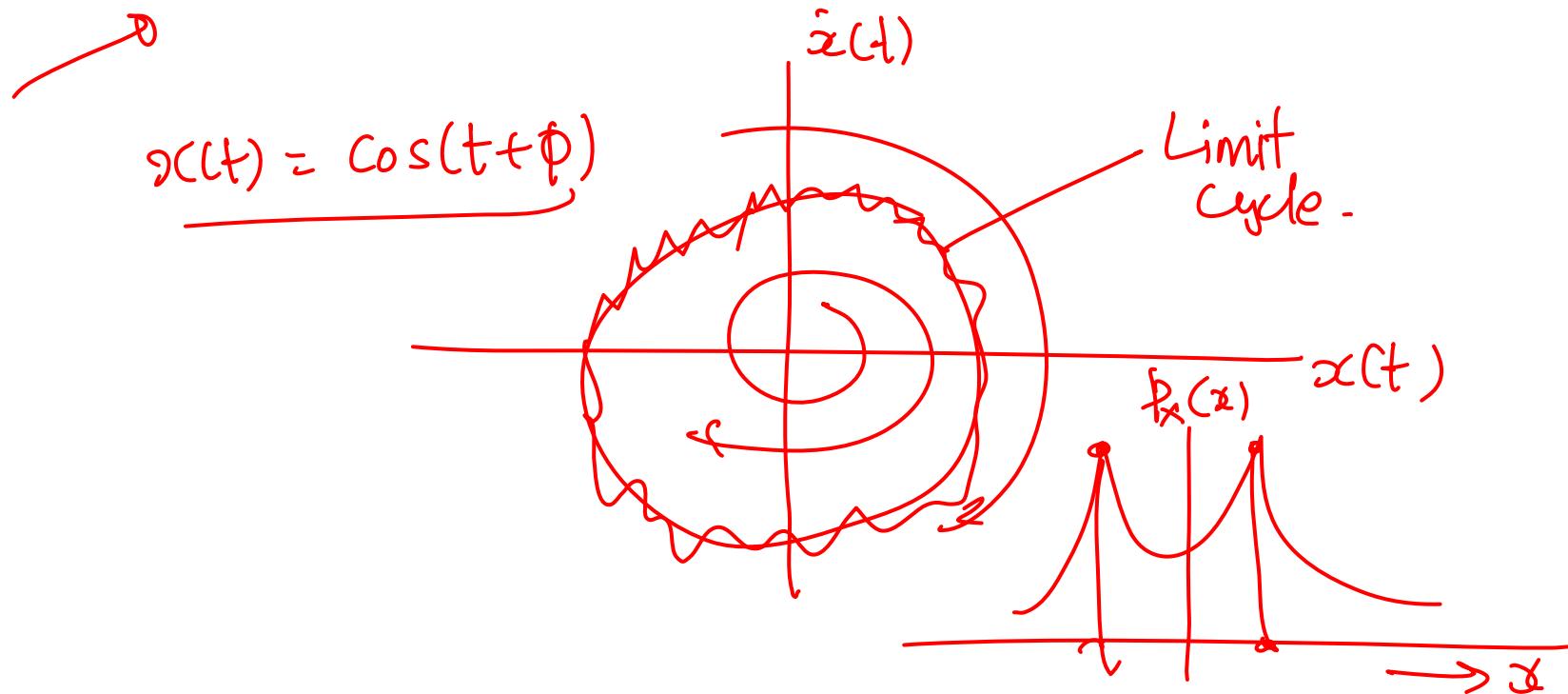
$$p(x, \dot{x}; t) = \Phi_0 \exp \left[ \frac{1}{D} \left\{ \left( \frac{\dot{x}^2}{2} + \frac{x^2}{2} \right) - \left( \frac{\dot{x}^2}{2} + \frac{x^2}{2} \right)^2 \right\} \right]; -\infty < x, \dot{x} < \infty$$

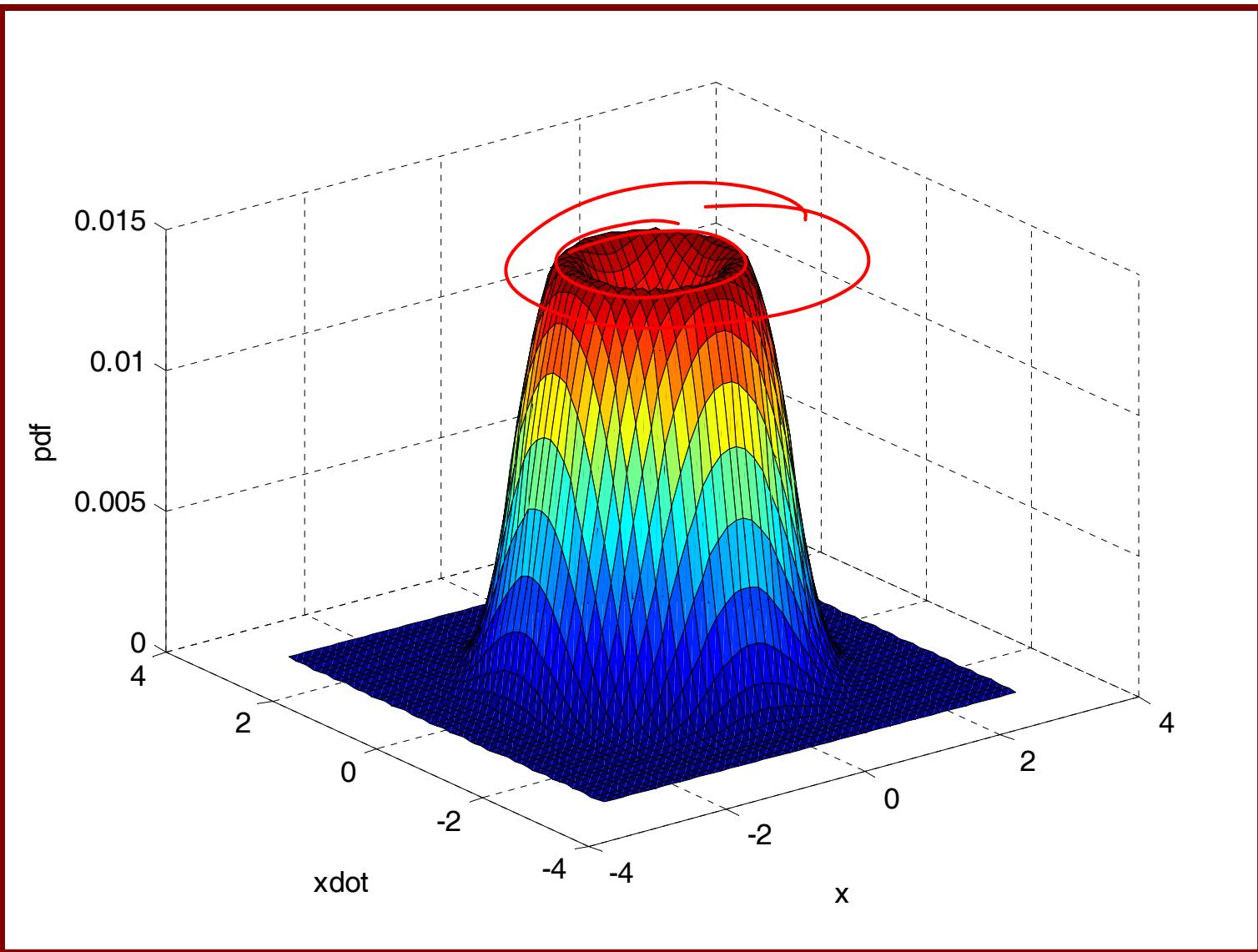
### Remarks

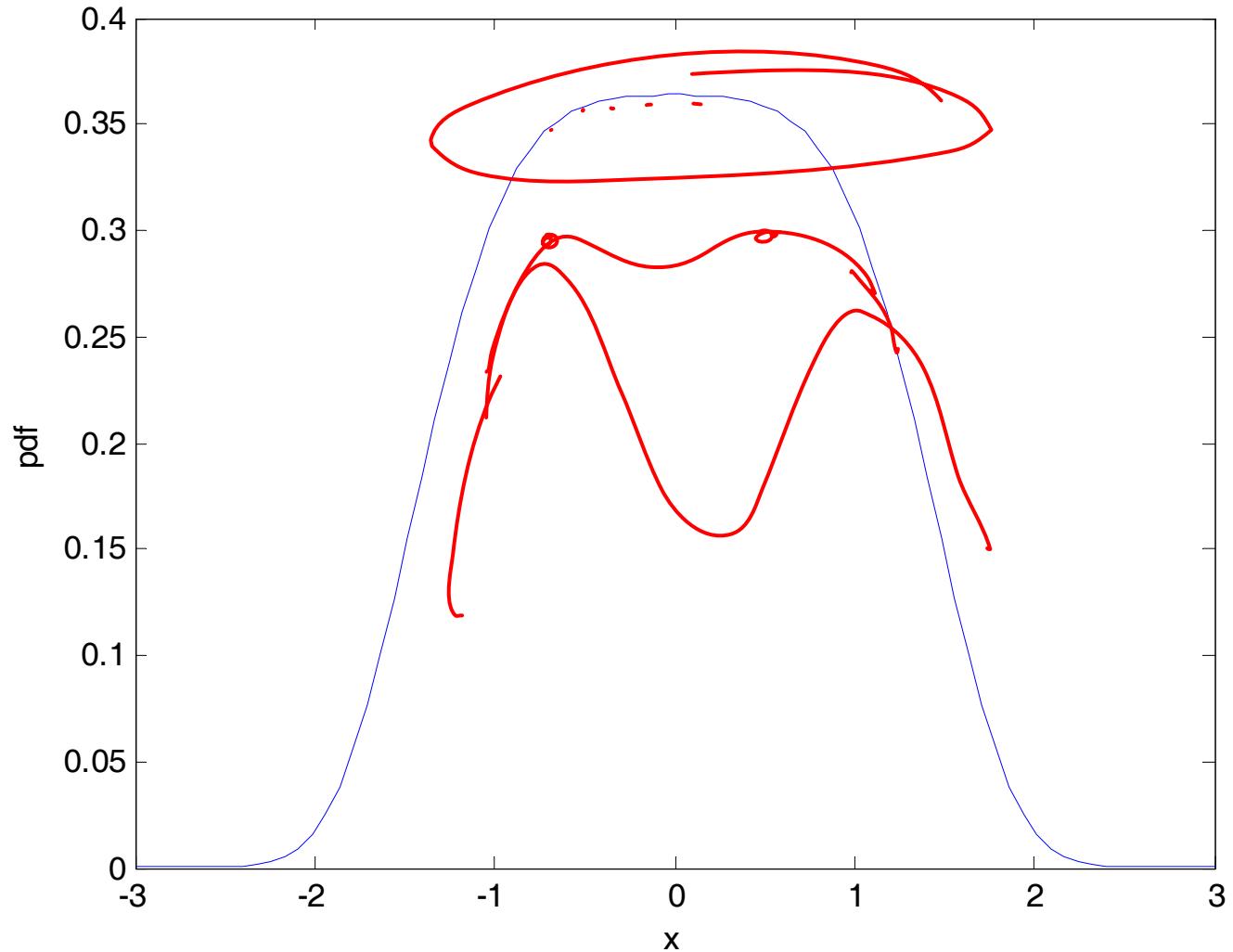
- Displacement and velocity here are non-Gaussian and dependent (although they are uncorrelated)
- In the absence of noise, the system has a stable limit cycle.
- The jpdf has a modal-line in the neighbourhood of the limit cycle.
- The pdf of displacement and velocity are not uni-modal.

# Limit cycle oscillations

$$\ddot{x} - \dot{x} \left( 1 - x^2 - \dot{x}^2 \right) + x = 0; x(0) = x_0; \dot{x}(0) = \dot{x}_0$$







## Example: Nonlinear $n$ -dof systems

$$m_j \ddot{X}_j + m_j \beta_j \dot{X}_j + \frac{\partial U}{\partial X_j} = W_j(t); t \geq 0;$$

$$X_j(0) = x_{j0} \text{ & } \dot{X}_j(0) = \dot{x}_{j0}$$

$$U = \frac{1}{2} X^t K X = \text{Potential energy of the structure}$$

$$\langle W_j(t) \rangle = 0; \langle W_j(t) W_k(t + \tau) \rangle = 2D_{jk} \delta_{jk} \delta(\tau);$$

$$\frac{m_j \beta_j}{2D_{jj}} = \gamma \quad \forall j = 1, 2, \dots, n$$

$$dY_j = Y_{j+n} dt$$

$$dY_{j+n} = \left( -\beta_j Y_{j+n} - \frac{1}{m_j} \frac{\partial U}{\partial Y_j} \right) dt + \frac{1}{m_j} dB_j(t)$$

$$\alpha_j = y_{j+n};$$

$$\alpha_{j+n} = \left( -\beta_j y_{j+n} - \frac{1}{m_j} \frac{\partial U}{\partial y_j} \right)$$

$$\alpha_{ij} = 0 \forall i \neq j; \alpha_{jj} = 0 \forall j < n;$$

$$\alpha_{j+n, j+n} = \frac{2D_{jj}}{m_j^2}$$



**Exercise :** show that the steady state pdf is given by

$$p(\tilde{y}; t) = C \exp \left\{ -\frac{\gamma}{\pi} \left[ \frac{1}{2} \sum_{j=1}^n m_j^2 y_{j+n}^2 + U(y_1, y_2, \dots, y_n) \right] \right\};$$

$$-\infty < y_i < \infty \forall i = 1, 2, \dots, 2n$$

## Moment equations

$$dX(t) = f[X(t), t]dt + G[X(t), t]dB(t); t \geq 0; X(0) = X_0$$

$$X(t), f \sim n \times 1; dB(t) \sim m \times 1; G \sim n \times m$$

$$\frac{\partial p}{\partial t} = - \sum_{j=1}^n \frac{\partial}{\partial x_j} [f(x, t) p] + \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2}{\partial x_i \partial x_j} [(G D G^t) p]$$

Denote

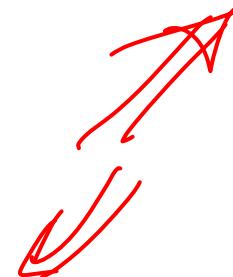
$$m_{k_1, k_2, \dots, k_n}(t) = \langle X_1^{k_1}(t) X_2^{k_2}(t) \cdots X_n^{k_n}(t) \rangle$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} x_1^{k_1} x_2^{k_2} \cdots x_n^{k_n} p(\tilde{x}; t | \tilde{x}_0; 0) p(x_0; 0) d\tilde{x} dx_0$$

$$\dot{m}_{k_1, k_2, \dots, k_n}(t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} x_1^{k_1} x_2^{k_2} \cdots x_n^{k_n} \frac{\partial}{\partial t} p(\tilde{x}; t | \tilde{x}_0; 0) p(x_0; 0) d\tilde{x} dx_0$$

$$\frac{\partial p}{\partial t} = - \sum_{j=1}^n \frac{\partial}{\partial x_j} [f(x, t) p] + \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2}{\partial x_i \partial x_j} [(G D G^t) p]$$

$$\dot{m}_{k_1, k_2, \dots, k_n}(t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} x_1^{k_1} x_2^{k_2} \cdots x_n^{k_n} \frac{\partial}{\partial t} p(\tilde{x}; t | \tilde{x}_0; 0) p(x_0; 0) d\tilde{x} dx_0$$

$$\dot{m}_{k_1, k_2, \dots, k_n}(t) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \cdots \int_{-\infty}^{\infty} x_1^{k_1} x_2^{k_2} \cdots x_n^{k_n}$$


$$- \sum_{j=1}^n \frac{\partial}{\partial x_j} [f(x, t) p] + \sum_{i=1}^n \sum_{j=1}^n \frac{\partial^2}{\partial x_i \partial x_j} [(G D G^t) p] p(x_0; 0) d\tilde{x} dx_0$$

Integrate the terms on the RHS by parts.

## General formulation

Let  $h[X, t]$  be a function which is well behaved

That is,  $\frac{\partial^2 h}{\partial x_i \partial x_j}$  &  $\frac{\partial h}{\partial t}$  exist, continuous, bounded on any finite interval of  $x$  and  $t$ .

Consider  $\underline{\delta h} = h(X + \delta X, t + \delta t) - h(X, t)$

Taylor's expansion

$$\delta h = \sum_{j=1}^n \delta X_j \frac{\partial h}{\partial X_j} + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \delta X_i \delta X_j \frac{\partial^2 h}{\partial X_i \partial X_j} + \underline{\delta t \frac{\partial h}{\partial t} + o(\delta X \delta X^t) + o(\delta t)}$$

$$dX(t) = f[X(t), t]dt + G[X(t), t]dB(t); t \geq 0; X(0) = X_0$$

$$X(t), f \sim n \times 1; dB(t) \sim m \times 1; G \sim n \times m$$

$$\langle \delta h | X \rangle = \sum_{j=1}^n f_j(X, t) \frac{\partial h}{\partial X_j} \delta t + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n (GDG^t)_{ij} \frac{\partial^2 h}{\partial X_i \partial X_j} \delta t +$$

$$\delta t \frac{\partial h}{\partial t} + o(\delta t)$$

**Digress :** Let  $\underline{X}$  and  $\underline{Y}$  be two random variables

$$\langle Y | X = x \rangle = \int_{-\infty}^{\infty} yp(y | x) dy$$

$$E[\langle Y | X = x \rangle] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} yp(y | x) p(x) dy = \underline{\underline{\langle Y \rangle}}$$

$$\langle \delta h | X \rangle = \sum_{j=1}^n f_j(X, t) \frac{\partial h}{\partial X_j} \delta t + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n (GDG^t)_{ij} \frac{\partial^2 h}{\partial X_i \partial X_j} \delta t +$$

$$\delta t \left\langle \frac{\partial h}{\partial t} \right\rangle + o(\delta t)$$

$$E[\langle \delta h | X \rangle] = \langle \delta h \rangle$$

$\langle h(x; t) \rangle$

$$\langle \delta h \rangle = \sum_{j=1}^n \left\langle f_j(X, t) \frac{\partial h}{\partial X_j} \right\rangle \delta t + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \left\langle (GDG^t)_{ij} \frac{\partial^2 h}{\partial X_i \partial X_j} \right\rangle \delta t +$$

$$\delta t \left\langle \frac{\partial h}{\partial t} \right\rangle + o(\delta t)$$

Divide by  $\delta t$  and consider the limit  $\delta t \rightarrow 0 \Rightarrow$

$$\frac{d}{dt} \langle h[X(t), t] \rangle = \left\langle \frac{\partial h}{\partial t} \right\rangle + \sum_{j=1}^n \left\langle f_j(X, t) \frac{\partial h}{\partial X_j} \right\rangle + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \left\langle (GDG^t)_{ij} \frac{\partial^2 h}{\partial X_i \partial X_j} \right\rangle$$

## Example

$$\dot{X} + \beta X = w(t); t \geq 0; X(0) = X_0$$

$$dX(t) = -\beta X dt + dB(t)$$

$$f = -\beta x$$

$$\frac{d}{dt} \left\langle h[X(t), t] \right\rangle = \left\langle \frac{\partial h}{\partial t} \right\rangle + \left\langle -\beta X \frac{\partial h}{\partial X_j} \right\rangle + D \left\langle \frac{\partial^2 h}{\partial X^2} \right\rangle$$

$$h = X^k; m_k = \langle X^k \rangle$$

$$\dot{m}_k = -\beta m_k + D k (k-1) m_{k-2}$$

$$\dot{m}_1 = -\beta m_1 \quad \checkmark$$

$$\dot{m}_2 = -\beta m_2 + 2D \quad \checkmark$$

$$\dot{m}_3 = -\beta m_3 + 6Dm_1 \quad \checkmark$$

$$\dot{m}_4 = -\beta m_4 + 12Dm_2 \quad \checkmark$$

⋮

Steady state  $\dot{m}_k = 0$

$$\dot{m}_1 = -\beta m_1 = 0 \Rightarrow \cancel{\underline{m_1 = 0}}$$

$$\dot{m}_2 = -\beta m_2 + 2D = 0 \Rightarrow -\beta m_2 + 2D = 0; m_2 = \frac{2D}{\beta} = \sigma_x^2$$

$$\dot{m}_3 = -\beta m_3 + 6Dm_1 = 0 \Rightarrow \cancel{\underline{m_3 = 0}}$$

$$\dot{m}_4 = -\beta m_4 + 12Dm_2 = 0 \Rightarrow m_4 = \frac{12D}{\beta} = 3\left(\frac{2D}{\beta}\right)^2 = 3\sigma_x^4$$

Remarks

- Moment equations are closed
- Moments in the steady state display Gaussian properties which is as it must be

## Example

$\ddot{X} + 2\eta\omega\dot{X} + \omega^2 X = w(t); t \geq 0; X(0) \& \dot{X}(0)$  specified.

$$dX_1 = X_2 dt$$

$$dX_2 = (-2\eta\omega X_2 - \omega^2 X_1) dt + dB(t)$$

$$\langle h(X_1, X_2) \rangle = \underbrace{\langle X_1^m X_2^n \rangle}_{= m_{mn}}$$

$$\frac{d}{dt} \langle h[X(t), t] \rangle = \left\langle \frac{\partial h}{\partial t} \right\rangle +$$

$$\sum_{j=1}^n \left\langle f_j(X, t) \frac{\partial h}{\partial X_j} \right\rangle + \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \left\langle \left( GDG^t \right)_{ij} \frac{\partial^2 h}{\partial X_i \partial X_j} \right\rangle$$

$$\dot{m}_{10} = \frac{d}{dt} \langle X_1 \rangle = \underline{\underline{m_{01}}} \quad \langle \dot{x} \rangle = \langle x \rangle$$

$$\dot{m}_{01} = \frac{d}{dt} \langle X_2 \rangle = -2\eta\omega m_{01} - \omega^2 m_{10} \quad \langle \dot{x} \rangle$$

$$\dot{m}_{20} = \langle X_2 \dot{X}_1 \rangle = 2m_{11} \quad \langle x^2 \rangle$$

$$\dot{m}_{11} = \langle X_2 X_2 \rangle + \left\langle \left( -2\eta\omega X_2 - \omega^2 X_1 \right) X_1 \right\rangle = \underline{\underline{m_{02}}} - 2\eta\omega m_{11} - \omega^2 m_{20} \quad \langle \dot{x} \dot{x} \rangle$$

$$\dot{m}_{02} = \left\langle 2X_2 \left( -2\eta\omega X_2 - \omega^2 X_1 \right) \right\rangle + D = \underline{\underline{-4\eta\omega m_{02}}} - 2\omega^2 m_{11} + D \quad \langle \dot{x}^2 \rangle$$

## Remark

Moment equations are closed

Steady state

$$\dot{m}_{10} = 0 \Rightarrow m_{01} = \underline{\underline{0}}$$

$$\dot{m}_{01} = 0 \Rightarrow -2\eta\omega m_{01} - \omega^2 m_{10} = 0 \Rightarrow m_{10} = \underline{\underline{0}}$$

$$\dot{m}_{20} = 0 \Rightarrow 2m_{11} = \underline{\underline{0}}$$

$$\dot{m}_{11} = 0 \Rightarrow m_{02} - 2\eta\omega m_{11} - \omega^2 m_{20} = 0 \Rightarrow m_{02} = \omega^2 m_{20}$$

$$m_{02} = 0 \Rightarrow -4\eta\omega m_{02} - 2\omega^2 m_{11} + D = 0 \Rightarrow m_{02} = \frac{D}{4\eta\omega}$$

$$\sigma_x^2 = \frac{D}{4\eta\omega^3}, \sigma_{\dot{X}}^2 = \frac{D}{4\eta\omega} \text{ & } \langle X(t) \dot{X}(t) \rangle = 0$$

These results agree with the exact solutions obtained earlier using convolution integral approach