nlin.lin Linearization

1 A common method for dealing with a nonlinear system is to linearize it: transform it linearize such that its state equation is linear. A linearized model is typically only valid in some neighborhood of state-space. This neighborhood is selected by choosing an operating point x_0 used in the linearization process. We use two considerations when choosing an operating point:

1. that implied by the name—it should be in a region of state-space in which the state will stay throughout the system's operation-and 2. the validity of the model near the

operating point. Due to the fact that nonlinear systems tend to be more-linear near equilibria, the second consideration frequently suggests we choose

one as an operating point: $x_o = \overline{x}$.

Taylor series expansion

2 A Taylor series expansion of Eq. 1a about an operating point x_o, u_o (for a nonautonomous system) yields polynomial terms that are linear, quadratic, etc. in x and u. If we keep only the linear terms and define new state and input

$$x^* = x - x_o \qquad \text{and} \qquad u^* = u - u_o \qquad \text{(1a)}$$
 we get a linear state equation

 $\frac{\mathrm{d}x^*}{\mathrm{d}t} = Ax^* + Bu^* \tag{1b}$ where the matrix components are given by

 $A_{ij} = \left. \frac{\partial f_i}{\partial x_j} \right|_{\mathbf{x_o},\mathbf{u_o}} \quad \text{and} \quad B_{ij} = \left. \frac{\partial f_i}{\partial \mathbf{u_j}} \right|_{\mathbf{x_o},\mathbf{u_o}}. \quad \text{(1c)}$

These first-derivative matrices are generally called <u>Iacobian</u> matrices.

3 This result also applies to autonomous equations if we drop the $\mathtt{B}\mathfrak{u}^*$ term.

f(x, u) = x2+ 4

Consider a vehicle suspension system that is overloaded such that its springs are exhibiting hardening behavior such that a lumped-parameter constitutive equation for the springs (collectively) is

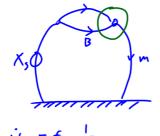
 $f_k = kx_k + \alpha x_k^3$ where f_k is the force, x_k the displacement, and $k, \alpha > 0$ constant parameters of the spring.

A. Develop a (nonlinear) spring-massdamper linear graph model for the vehicle suspension with input position source X_s .

b. Derive a nonlinear state-space model

leave graph model using the

 $\mathbf{x}_{o} = \begin{bmatrix} 1 & 0 \end{bmatrix}^{\top}$ and $\mathbf{u}_{o} = \begin{bmatrix} 0 \end{bmatrix}$ (4) by computing the A, B, and E matrices of



 $\dot{V}_{m} = f_{m} \frac{1}{m}$ $f_{m} = f_{B} + f_{K}$

 $f_B = BV_B$ $X_S = X_k + X_m$ $f_k = kx_k + \epsilon x_K^3$ $X_S - X_m = X_k$

 $\dot{x}_m = v_m$ $V_s = V_B + v_m = \dot{x}_s$ \Longrightarrow $V_B = \dot{x}_s - v_m$

 $\dot{V}_{m} = f_{n} \frac{1}{m} = \frac{1}{m} (f_{B} + f_{K})$ $= \frac{1}{m} (B \vee_B + k \times_K + a \times_K^3)$ $=\frac{1}{m}(\beta \vee_{\beta}+k(x_{5}-x_{m})+a(x_{5}-x_{m})^{2})$ = $\frac{1}{m} (B(\dot{x}_s - v_m) + k(x_s - x_m) + a(x_s - x_m)^3)$

 $\frac{dx}{dt} = f(x, u)$

 $+ \left(\begin{bmatrix} x_m \\ v_m \end{bmatrix}, \begin{bmatrix} x_s \end{bmatrix} \right) = \begin{bmatrix} v_m \\ v_m (B(\dot{x}_s - v_m) + k(x_s - x_m) + a(x_s - x_m)^3) \end{bmatrix}$

 $A_{11} = 0$ $A_{12} = 1$

 $A_{21} = \frac{-k + 3a(x_s - x_m)^2 (-1)}{m} \bigg|_{X_0, u_0} = \frac{-k - 3a(o - 1)^2}{m} = \frac{-k - 3a}{m}$ $A_{22} = -B_{m}$

 $A = \begin{bmatrix} 0 & 1 \\ \frac{-k-3e}{m} & \frac{-B}{m} \end{bmatrix}$

 $\beta_{21} = \frac{k}{m} + \frac{3a(x_s - x_m)^2}{m}\Big|_{X_{o_s} U_{o}} = \frac{k}{m} + \frac{3a(o-1)^2}{m} = \frac{k+3a}{m}$ $B = \begin{bmatrix} 0 \\ \frac{K+3e}{m} \end{bmatrix} \qquad E = \begin{bmatrix} 0 \\ B_n \end{bmatrix}$

 $\dot{x} = A \times + Bu + E \dot{u}$