

## Linearization of systems

### One dimensional linear approximation

Consider a differentiable function  $f$  of one variable.

$$f : \mathbb{R} \rightarrow \mathbb{R}$$

Say we are interested in  $f$  in a small, open neighborhood about a particular point  $t_0$ .  $t_0$  is often called the fixed point of the system. Lets call this open neighborhood  $U$ . In this case, we can construct a linear approximation of  $f$  about the neighborhood  $U$ . Recall that,

$$\frac{df}{dt}(t_0) \approx \frac{f(t) - f(t_0)}{t - t_0} \text{ for } t \in U$$

We can use the above to construct a linear approximation of  $f$ . Let  $f_l$  denote the linear approximation of  $f$  about  $U$ .

$$f_l(t) = \frac{df}{dt}(t_0)(t - t_0) + f(t_0) \quad (1)$$

Strictly speaking,  $f_l$  is an affine approximation unless  $t_0 = 0$ , but the process of obtaining  $f_l$  is colloquially called the linearization of  $f$ .

For example, consider  $f(t) = t^2$ . We will set the fixed point of the system to be  $t_0 = 1$ . Then,

$$\begin{aligned} f_l(t) &= \frac{df}{dt}(t_0)(t - t_0) + f(t_0) \\ &= 2(t - 1) + 1 \\ &= 2t - 1 \end{aligned}$$

Let  $\varepsilon = 10^{-2}$ . Consider the open neighborhood  $U = (1 - \varepsilon, 1 + \varepsilon)$ . Lets plot  $f$  and  $f_l$  when their respective domains are restricted to  $U$ . This is seen in Figure 1.

### Linearization of a system

Consider a continuous, non-linear system with state  $\vec{x}(t)$  (which is  $n$  dimensional) and representation  $\vec{u}(t)$  (which is  $m$  dimensional) of the form,

$$\frac{d\vec{x}}{dt}(t) = f(\vec{x}(t), \vec{u}(t))$$

To clarify and establish notation,  $f$  is a function that takes in  $\vec{x}(t)$  and  $\vec{u}(t)$  and outputs an  $n$  dimensional vector.  $f_k(\vec{x}(t), \vec{u}(t))$  refers to the function that is the  $k^{th}$  coordinate of the output of  $f(\vec{x}(t), \vec{u}(t))$ .

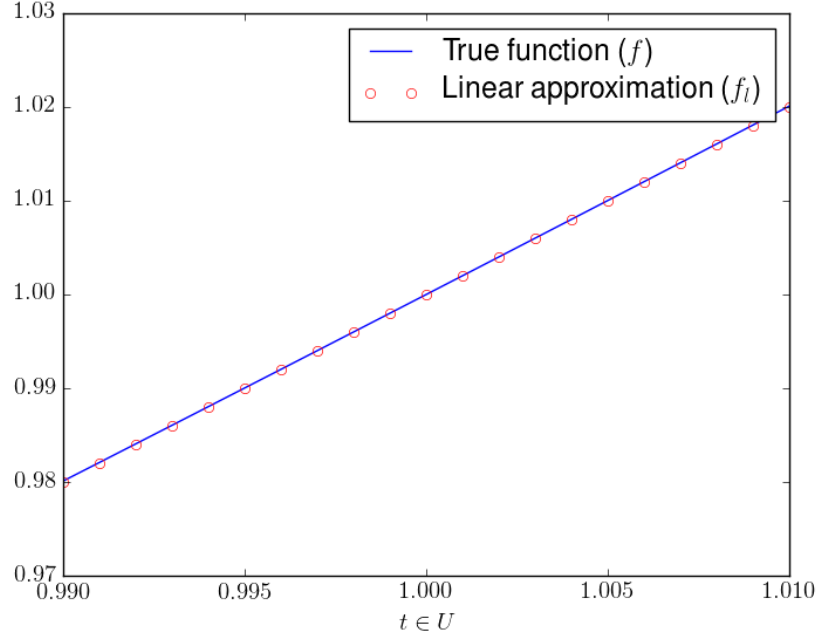


Figure 1: Comparing the linear approximation of a function to the original function

Let  $\vec{x}_0(t)$  be the desired state trajectory and  $\vec{u}_0(t)$  be the desired input. Let  $\delta\vec{x}(t)$  be a small perturbation about the state trajectory and  $\delta\vec{u}(t)$  be the small perturbation about the input trajectory. In equation form,

$$x(t) = x_0(t) + \delta\vec{x}(t) \text{ and } u(t) = u_0(t) + \delta\vec{u}(t)$$

Analogous to (1), we will construct an affine estimate about a neighborhood around the fixed points.

$$f(\vec{x}_0(t) + \delta\vec{x}(t), \vec{u}_0(t) + \delta\vec{u}(t)) \approx f(\vec{x}_0(t), \vec{u}_0(t)) + \frac{df}{dx}(x_0(t), u_0(t))\delta\vec{x}(t) + \frac{df}{du}(x_0, u_0)\delta\vec{u}(t) \quad (2)$$

Notice that we have used the notation  $\frac{df}{dx}$  and  $\frac{df}{du}$ , which might be strange to consider since  $\vec{x}(t)$  is a vector and not a single variable. This is the multidimensional generalization of the derivative, which is constructed as follows.

$$\frac{df}{dx}(x, u) = \begin{bmatrix} \frac{df_1}{dx_1}(x, u) & \frac{df_1}{dx_2}(x, u) & \dots & \frac{df_1}{dx_n}(x, u) \\ \frac{df_2}{dx_1}(x, u) & \frac{df_2}{dx_2}(x, u) & \dots & \frac{df_2}{dx_n}(x, u) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{df_n}{dx_1}(x, u) & \frac{df_n}{dx_2}(x, u) & \dots & \frac{df_n}{dx_n}(x, u) \end{bmatrix} \text{ and } \frac{df}{du} = \begin{bmatrix} \frac{df_1}{du_1}(x, u) & \frac{df_1}{du_2}(x, u) & \dots & \frac{df_1}{du_m}(x, u) \\ \frac{df_2}{du_1}(x, u) & \frac{df_2}{du_2}(x, u) & \dots & \frac{df_2}{du_m}(x, u) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{df_n}{du_1}(x, u) & \frac{df_n}{du_2}(x, u) & \dots & \frac{df_n}{du_m}(x, u) \end{bmatrix}$$

Note the dimensions of the matrices. It must be notated that, when calculating  $\frac{df}{dx}$ ,  $\vec{u}$  is considered constant. Similarly, when calculating  $\frac{df}{du}$ ,  $\vec{x}$  is considered constant.

To continue from (2), note that,

$$\frac{dx}{dt}(t) = \frac{dx_0(t)}{dt} + \frac{d\delta x}{dt}(t) \text{ and } \frac{dx_0}{dt}(t) = f(x_0(t), u_0(t))$$

Plugging this back into (2), we get,

$$\cancel{\frac{dx_0}{dt}(t)} + \frac{d\delta x}{dt}(t) \approx \cancel{\frac{dx_0}{dt}(t)} + \frac{df}{dx}(x_0(t), u_0(t))\delta\vec{x}(t) + \frac{df}{du}(x_0, u_0)\delta\vec{u}(t)$$

Thus, we get linearized version of our system.

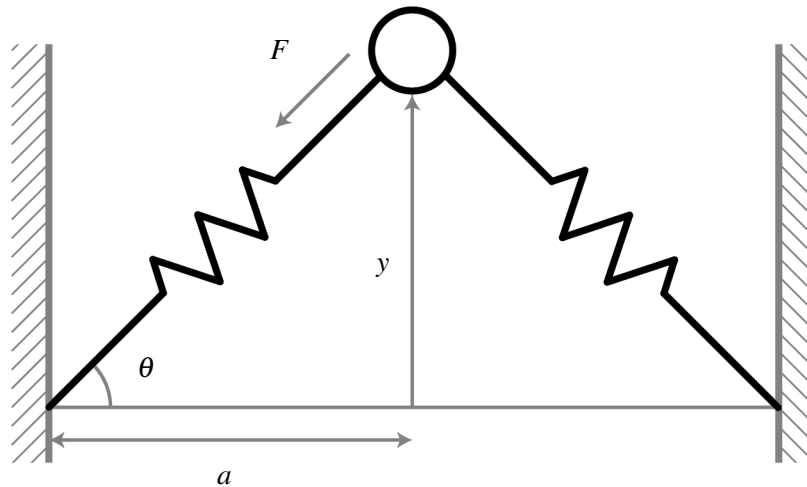
$$\frac{d\delta x}{dt}(t) \approx \frac{df}{dx}(x_0(t), u_0(t))\delta\vec{x}(t) + \frac{df}{du}(x_0, u_0)\delta\vec{u}(t) \quad (3)$$

Note that, unlike the one dimensional linearization example, we are **linearizing with respect to  $\vec{x}$  and  $\vec{u}$** . Also, observe that our state variables are now the perturbations  $\delta\vec{x}(t)$  and  $\delta\vec{u}(t)$ .

## Questions

### 1. Linearization

Consider a mass attached to two springs:



We assume that each spring is linear with spring constant  $k$  and resting length  $X_0$ . We want to build a state space model that describes how the displacement  $y$  of the mass from the spring base evolves.

- Find the force  $F$  applied by each spring.
- Use Newton's law to write an equation for  $\ddot{y}$  in terms of  $y$ .
- Write this model in state space form  $\dot{x} = f(x)$ .
- Find the equilibrium of the model assuming that  $X_0 < a$ .
- Linearize your model about the equilibrium.
- Compute the eigenvalues of your linearized model. Is this equilibrium stable?

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