### Chapter 1

# Steady-state stability analysis of power system models

In this Chapter, we conduct stability analysis based on the approximate linearization of power system models. The structure of this Chapter is as follows. First, in Section ??, we derive a linear approximation model for the power system model described by a system of ordinary differential equations using Kron reduction of the generator buses. Then, in Section ??, we explain the method for numerically analyzing the stability of the derived linear approximation model. We also confirm through numerical simulation that the stability of the linear approximation model depends not only on the physical constants of the generators, loads, and transmission lines, but also on the selection of the steady-state power flow. Additionally, in Section ??, we explore advanced topics and demonstrate how the stability of the linear approximation model can be analyzed using the concept of passivity in dynamic systems.

#### COFFEE BREAK

#### **Derivation of the approximate linear system:**

Consider the nonlinear system:

$$\dot{x}(t) = f(x(t)) + Bu(t)$$

where f(0) = 0. The function f(x) can be expressed near the origin by a Taylor expansion as:

$$f(x) = f(0) + \frac{\partial f}{\partial x}(0)x + \text{Second or higher order term}$$

Here, f(x) and x are expressed as  $f_i(x)$  and  $x_i$ , respectively, and  $\frac{\partial f}{\partial x}(x)$  is the *Jacobian matrix* with the (i, j) element given by  $\frac{\partial f_i}{\partial x_j}(x)$ . By using this Jacobian matrix, we define:

$$A := \frac{\partial f}{\partial x}(0)$$

Then, when the magnitudes of the state x(t) and input u(t) are sufficiently small, the behavior of the nonlinear system can be approximated by the behavior of the

linear system obtained by neglecting terms of degree 2 or higher in the function f:

$$\dot{x}^{\rm lin}(t) = Ax^{\rm lin}(t) + Bu^{\rm lin}(t)$$

Note that even if u(t) and  $u^{\text{lin}}(t)$  are the same, the state x(t) of the nonlinear system and the state  $x^{\text{lin}}(t)$  of the approximate linear system may not be exactly the same.

#### 1 Stability analysis based on linear approximation

#### 1.1 Approximate linearization of the power system model

In this section, we derive an approximate linear model for the power system model where each bus has a generator connected, which is equivalent to the Kron-reduced differential equation system model discussed in Section ??. We derive the approximate linear model for the steady-state flow state. The differential equation system model is given by:

$$\begin{cases} \dot{\delta}_{i} = \omega_{0} \Delta \omega_{i} \\ M_{i} \Delta \dot{\omega}_{i} = -D_{i} \Delta \omega_{i} - f_{i} (\delta, E) + P_{\text{mech}i} \\ \tau_{i} \dot{E}_{i} = -\frac{X_{i}}{X_{i}'} E_{i} + (X_{i} - X_{i}') g_{i} (\delta, E) + V_{\text{field}i} \end{cases}$$
 (1)

However,  $\delta$  and E are vectors obtained by vertically arranging  $\delta_i$  and  $E_i$ , respectively. The nonlinear terms representing the interactions between generators are expressed as follows:

$$f_{i}(\delta, E) := -E_{i} \sum_{j=1}^{N} E_{j} \left( B_{ij}^{\text{red}} \sin \delta_{ij} - G_{ij}^{\text{red}} \cos \delta_{ij} \right),$$

$$g_{i}(\delta, E) := -\sum_{j=1}^{N} E_{j} \left( B_{ij}^{\text{red}} \cos \delta_{ij} + G_{ij}^{\text{red}} \sin \delta_{ij} \right)$$

$$(2)$$

In addition,  $\delta_{ij} := \delta_i - \delta_j$  is defined. Note that, due to the properties of reduced admittance, the reduced conductance and reduced susceptance satisfy the symmetry condition:

$$G_{ij}^{\mathrm{red}} = G_{ji}^{\mathrm{red}}, \qquad B_{ij}^{\mathrm{red}} = B_{ji}^{\mathrm{red}}, \qquad \forall (i,j) \in I_{\mathrm{G}} \times I_{\mathrm{G}}$$

To obtain the partial derivatives of these nonlinear functions with respect to each variable, we define:

$$\begin{aligned} k_{ij}(\delta_{ij}) &:= -B_{ij}^{\text{red}} \cos \delta_{ij} - G_{ij}^{\text{red}} \sin \delta_{ij}, \\ h_{ij}(\delta_{ij}) &:= -B_{ij}^{\text{red}} \sin \delta_{ij} + G_{ij}^{\text{red}} \cos \delta_{ij} \end{aligned} \tag{3}$$

Then, for  $f_i$ , we obtain:

$$\frac{\partial f_{i}}{\partial \delta_{i}} = E_{i} \sum_{j=1, j \neq i}^{N} E_{j} k_{ij}(\delta_{ij}), \quad \frac{\partial f_{i}}{\partial E_{i}} = 2E_{i} h_{ii}(\delta_{ii}) + \sum_{j=1, j \neq i}^{N} E_{j} h_{ij}(\delta_{ij}), 
\frac{\partial f_{i}}{\partial \delta_{j}} = -E_{i} E_{j} k_{ij}(\delta_{ij}), \quad \frac{\partial f_{i}}{\partial E_{j}} = E_{i} h_{ij}(\delta_{ij})$$
(4)

where  $j \neq i$ .

Similarly, we can obtain the partial derivatives of  $g_i$  as follows:

$$\frac{\partial g_{i}}{\partial \delta_{i}} = -\sum_{j=1, j \neq i}^{N} E_{j} h_{ij}(\delta_{ij}), \quad \frac{\partial g_{i}}{\partial E_{i}} = k_{ii}(\delta_{ii}), 
\frac{\partial g_{i}}{\partial \delta_{j}} = E_{j} h_{ij}(\delta_{ij}), \qquad \frac{\partial g_{i}}{\partial E_{j}} = k_{ij}(\delta_{ij})$$
(5)

We denote the steady-state values of the internal state of generator i as  $(\delta_i^{\star}, E_i^{\star})$  and the steady-state values of external inputs as  $(P_{\text{mech}i}^{\star}, V_{\text{field}i}^{\star})$  for the differential equation system in equation ??. Furthermore, we use symbols without the subscript i to represent the vector of these values for all  $i \in I_G$ . For example,  $\delta^{\star}$  denotes the vector  $(\delta_i^{\star})_{i \in I_G}$ . With these steady-state values, we can write the following system of equations:

$$\begin{cases}
0 = -f_i \left( \delta^{\star}, E^{\star} \right) + P_{\text{mech}i}^{\star} \\
0 = -\frac{X_i}{X_i'} E_i^{\star} + \left( X_i - X_i' \right) g_i \left( \delta^{\star}, E^{\star} \right) + V_{\text{field}i}^{\star}
\end{cases} \qquad i \in I_G \tag{6}$$

Here, note that we assume the steady-state value of the frequency deviation  $\Delta\omega_i$  in Eq. ?? is zero for all  $i \in I_G$ . The validity of Eq. ?? corresponds to setting the steady-state values of the external input  $(P_{\text{mech}}^{\star}, V_{\text{field}}^{\star})$  to appropriate values that achieve supply-demand balance. By linearizing the system around this steady state, the approximate linear model is obtained as:

$$\begin{bmatrix} \dot{\delta}^{\text{lin}} \\ M\Delta\dot{\omega}^{\text{lin}} \\ \tau \dot{E}^{\text{lin}} \end{bmatrix} = \begin{bmatrix} 0 & \omega_0 I & 0 \\ -L & -D & -C \\ B & 0 & A \end{bmatrix} \begin{bmatrix} \delta^{\text{lin}} \\ \Delta\omega^{\text{lin}} \\ E^{\text{lin}} \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ I & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} P_{\text{mech}}^{\text{lin}} \\ V_{\text{field}}^{\text{lin}} \end{bmatrix}$$
(7)

Note that the state and input variables with the subscript "lin" are vectors consisting of small deviations from the corresponding variables with their steady-state values as the reference. Also,

$$M := \operatorname{diag}(M_i)_{i \in I_G}, \qquad D := \operatorname{diag}(D_i)_{i \in I_G}, \qquad \tau := \operatorname{diag}(\tau_i)_{i \in I_G}$$

are diagonal matrices where  $\text{diag}(\cdot)$  is an operator that creates a diagonal matrix from a vector.

Furthermore, for the functions  $k_{ij}$  and  $h_{ij}$  defined in Equation ??, the (i, j) element of the matrices  $\hat{L}$ ,  $\hat{A}$ ,  $\hat{B}$ , and  $\hat{C}$ , defined as:

$$\hat{L}_{ij} := \begin{cases} E_{i}^{\star} \sum_{j=1, j \neq i}^{N} E_{j}^{\star} k_{ij} (\delta_{ij}^{\star}), & i = j \\ -E_{i}^{\star} E_{j}^{\star} k_{ij} (\delta_{ij}^{\star}), & i \neq j \end{cases}$$

$$\hat{A}_{ij} := \begin{cases} k_{ii} (\delta_{ii}^{\star}) - \frac{X_{i}}{X_{i}^{\prime} (X_{i} - X_{i}^{\prime})}, & i = j \\ k_{ij} (\delta_{ij}^{\star}), & i \neq j \end{cases}$$

$$\hat{B}_{ij} := \begin{cases} -\sum_{j=1, j \neq i}^{N} E_{j}^{\star} h_{ij} (\delta_{ij}^{\star}), & i = j \\ E_{j}^{\star} h_{ij} (\delta_{ij}^{\star}), & i \neq j \end{cases}$$

$$\hat{C}_{ij} := \begin{cases} \sum_{j=1, j \neq i}^{N} E_{j}^{\star} h_{ij} (\delta_{ij}^{\star}), & i = j \\ E_{i}^{\star} h_{ij} (\delta_{ij}^{\star}), & i \neq j \end{cases}$$

The matrices L, A, B, and C are then defined as follows:

$$\begin{split} L &:= \hat{L}, \\ A &:= \operatorname{diag} \left( X_i - X_i' \right)_{i \in I_G} \hat{A}, \\ B &:= \operatorname{diag} \left( X_i - X_i' \right)_{i \in I_G} \hat{B}, \\ C &:= \operatorname{diag} \left( 2E_i^{\star} h_{ii} \left( \delta_{ii}^{\star i} \right) \right)_{i \in I_G} + \hat{C} \end{split} \tag{8}$$

Note that  $\delta_{ij}^{\star} := \delta_i^{\star} - \delta_j^{\star}$ . It should be noted that the system matrix (L, A, B, C) is a function of the steady-state values  $(\delta^{\star}, E^{\star})$ . The block diagram of this approximate linear model is shown in Figure ??. Here,  $P^{\text{lin}}$  represents the approximately linearized active power supplied by the generators. Note that generally  $X_i > X_i'$  for all i.

In power system engineering, the value obtained by differentiating the generator's active power with respect to the rotor angle at the steady-state is called the **synchronizing power coefficient** [?, Section 8.4]. That is, the matrix L in the approximate linear model given by equation ?? corresponds to the synchronizing power coefficient. However, in power system engineering, it is common to define the synchronizing power coefficient using the one-machine infinite-bus system model explained in Section ??, so it is a scalar value rather than a matrix.

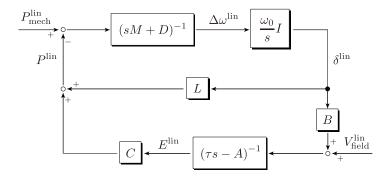


Fig. 1 Block Diagram of Approximate Linear Model

#### 1.2 Stability analysis of approximate linear models

#### 1.2.1 Stability of approximate linear models

In this section, we consider numerically analyzing the stability of the approximate linear model. Whether the approximate linear model of Equation ?? is stable or not is characterized by whether the internal states of the generator groups return to the steady state satisfying the simultaneous equations of Equation ?? in the event of a small disturbance in the power system, such as temporary minor fluctuations in mechanical input or excitation input of the generators, impedance values of the loads, current or voltage values of the transmission lines, etc., from the reference values at the steady state. In power system engineering, stability against such small fluctuations is called **small signal stability**.

It should be noted that the stability of the approximate linear model of Equation  $\ref{eq:table:eq:ta$ 

#### 1.2.2 Stability analysis based on eigenvalues of the system matrix

For the approximate linear model in Equation ??, if we appropriately choose the steady-state values  $(\delta^*, E^*)$  of the internal states as parameters, then the system matrix (L, A, B, C) in Equation ?? and the steady-state values  $(P_{\text{mech}}^*, V_{\text{field}}^*)$  of the

external inputs satisfying Equation ?? are determined dependently. Here, we consider setting

$$P_{\mathrm{mech}i}(t) = P_{\mathrm{mech}i}^{\star}, \qquad V_{\mathrm{field}i}(t) = V_{\mathrm{field}i}^{\star}, \qquad \forall t \ge 0$$

for all  $i \in IG$  in the nonlinear differential equation system model in Equation ??. We then assess the stability of the system using the eigenvalues of the system matrix.

This means that in the approximate linear model of Equation ?? the following values are set:

$$P_{\text{mech}}^{\text{lin}}(t) = 0, \qquad V_{\text{field}}^{\text{lin}}(t) = 0, \qquad \forall t \ge 0$$

In the following, under this assumption, we analyze the stability of an autonomous approximate linear model with input set identically to zero, given by:

$$\begin{bmatrix} \delta^{\text{lin}} \\ \Delta \dot{\omega}^{\text{lin}} \\ \dot{E}^{\text{lin}} \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & \omega_0 I & 0 \\ -M^{-1}L - M^{-1}D - M^{-1}C \\ \tau^{-1}B & 0 & \tau^{-1}A \end{bmatrix}}_{\Psi} \begin{bmatrix} \delta^{\text{lin}} \\ \Delta \omega^{\text{lin}} \\ E^{\text{lin}} \end{bmatrix}$$
(9)

Specifically, by examining the sign of the real part of the eigenvalues of the matrix  $\Psi$ , we can determine the stability of this approximate linear model. However, it should be noted that  $\Psi$  generally has at least one zero eigenvalue. In fact, from the structure of the matrices L and B in equation  $\ref{eq:partial}$ , we have:

$$L\mathbb{1} = 0, \qquad B\mathbb{1} = 0 \tag{10}$$

Therefore, for any model parameters, we have:

$$\Psi v = 0, \qquad v := \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

This means that v is an eigenvector of  $\Psi$  corresponding to a zero eigenvalue. If the real parts of all eigenvalues, except for the zero eigenvalue, are negative, then for any initial value, the solution trajectory of Equation ?? satisfies:

$$\lim_{t \to \infty} \delta^{\text{lin}}(t) = c_0 \mathbb{1}, \qquad \lim_{t \to \infty} \Delta \omega^{\text{lin}}(t) = 0, \qquad \lim_{t \to \infty} E^{\text{lin}}(t) = 0$$
 (11)

Here,  $c_0$  is a constant determined by the initial value. Note that the value of  $c_0$  does not make a significant difference in the analytical results. This is because in the differential equation system model of Equation ??, the rotor angle  $\delta_i$  of a generator has meaning only in relation to the difference between the rotor angle  $\delta_j$  of other generators. Specifically, if  $(\delta^*, E^*)$  satisfies the system of equations in equation ?? for a certain  $(P^*_{\text{mech}}, V^*_{\text{field}})$ , then  $(\delta^* + c_0\mathbb{1}, E^*)$  also satisfies the same system of equations. Therefore,  $\delta^*$  and  $\delta^* + c_0\mathbb{1}$  are essentially equivalent steady-state values where all generator rotor angles are rotated by the same amount of  $c_0$ . Equation

?? means the asymptotic convergence of solution trajectories to these essentially equivalent steady-state values.

## 2 stability analysis of approximate linear models using numerical calculations

## 2.1 Implementation of approximate linearization using a group of partitioned modules

In this section, we explain the implementation method for obtaining an approximate linear model numerically. Specifically, we describe how to add the functionality of linearization to the program that has been segmented into module groups as explained in Sections ?? and ??.

In the numerical simulation program of the power system created in Section ??, the following state and output equations are implemented for each device as differential and algebraic equations, respectively:

$$\dot{x}_i = f_i^{(1)}(x_i, V_i, I_i, u_i), \qquad 0 = f_i^{(2)}(x_i, V_i, I_i, u_i)$$

In the following, we derive the approximate linear model in the vicinity of the equilibrium point  $(x_i^*, V_i^*, I_i^*, u_i^*)$  for the device of interest. Specifically, we explain the implementation method of the linear approximation function to the program that has been partitioned into the module group described in Sections ?? and ??.

For the numerical simulation program of the power system created in Section ??, differential equations for the state and algebraic equations for the output are implemented for each device as:

$$\dot{x}_i = f_i^{(1)}(x_i, \pmb{V}_i, \pmb{I}_i, u_i), \qquad 0 = f_i^{(2)}(x_i, \pmb{V}_i, \pmb{I}_i, u_i)$$

We consider the linearization of the functions  $f_i^{(1)}$  and  $f_i^{(2)}$  as follows:

$$f_i^{(1)}(x_i, \boldsymbol{V}_i, \boldsymbol{I}_i, u_i) \approx A_i(x_i - x_i^{\star}) + B_{u_i} u_i$$

$$+ B_{\boldsymbol{V}_i} \begin{bmatrix} \operatorname{Re}[\boldsymbol{V}_i - \boldsymbol{V}^{\star}] \\ \mathrm{i}[\boldsymbol{V}_i - \boldsymbol{V}^{\star}] \end{bmatrix} + B_{\boldsymbol{I}_i} \begin{bmatrix} \operatorname{Re}[\boldsymbol{I}_i - \boldsymbol{I}_i^{\star}] \\ \mathrm{i}[\boldsymbol{I}_i - \boldsymbol{I}_i^{\star}] \end{bmatrix}$$
(12)

$$f_i^{(2)}(x_i, \boldsymbol{V}_i, \boldsymbol{I}_i, u_i) \approx C_i(x_i - x_i^{\star}) + D_{u_i} u_i$$

$$+ D_{\boldsymbol{V}_i} \begin{bmatrix} \operatorname{Re}[\boldsymbol{V}_i - \boldsymbol{V}^{\star}] \\ i[\boldsymbol{V}_i - \boldsymbol{V}^{\star}] \end{bmatrix} + D_{\boldsymbol{I}_i} \begin{bmatrix} \operatorname{Re}[\boldsymbol{I}_i - \boldsymbol{I}_i^{\star}] \\ i[\boldsymbol{I}_i - \boldsymbol{I}_i^{\star}] \end{bmatrix}$$
(13)

A system of simultaneous equations for each machine and algebraic equations for the entire power system can be used to obtain an expression using ordinary differential equations for the approximate linear model by eliminating all  $V_i - V_i^*$  and  $I_i - I_i^*$ , where  $i \in 1, ..., N$ , as follows:

$$I_i - I_i^* = \sum_{j=1}^N Y_{ij} (V_j - V_j^*), \qquad i \in \{1, \dots, N\}$$

Here,  $Y_{ij}$  represents the (i, j)th element of the admittance matrix Y. Let us check the specific implementation method with the following example.

#### **Example 1.1 (Implementation of Approximate Linear Model)**

Equations ?? and ?? depend on the dynamic characteristics of the device, so it is natural to implement the calculation of coefficient matrices such as  $A_i$  and  $B_{u_i}$  in the classes of devices such as generators and loads in the implementation example of Section ??. For example, in the generator model:

$$A_i = \begin{bmatrix} 0 & \omega_0 & 0 \\ 0 & -\frac{D_i}{M_i} & 0 \\ -\frac{1}{\tau_i} (\frac{X_i}{X_i'} - 1) |V_i^{\star}| \sin(\delta_i^{\star} - \angle V_i^{\star}) & 0 & -\frac{X_i}{\tau_i X_i'} \end{bmatrix}$$

$$B_{u_i} = \begin{bmatrix} 0 \\ \frac{1}{M_i} \\ 0 \end{bmatrix}, \qquad B_{V_i} = \begin{bmatrix} 0 & 0 \\ -\frac{\text{Re}[I_i^{\star}]}{M_i} & -\frac{\text{i}[I_i^{\star}]}{M_i} \\ \frac{1}{\tau_i} (\frac{X_i}{X_i'} - 1) \cos \delta_i^{\star} & \frac{1}{\tau_i} (\frac{X_i}{X_i'} - 1) \sin \delta_i^{\star} \end{bmatrix}$$

$$B_{\boldsymbol{I}_i} = \begin{bmatrix} 0 & 0 \\ -\frac{\operatorname{Re}[\boldsymbol{V}_i^\star]}{M_i} & -\frac{\mathrm{i}[\boldsymbol{V}_i^\star]}{M_i} \\ 0 & 0 \end{bmatrix}, \qquad C_i = \begin{bmatrix} E_i^\star \cos \delta_i^\star & 0 & \sin(\delta_i^\star) \\ E_i^\star \sin \delta_i^\star & 0 - \cos(\delta_i^\star) \end{bmatrix}$$

$$D_{u_i} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \qquad D_{V_i} = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}, \qquad D_{I_i} = \begin{bmatrix} -X_i' & 0 \\ 0 & -X_i' \end{bmatrix}$$

If the calculation of these coefficient matrices is added to the generator class as a method named get\_linear\_matrix, the program ?? is obtained.

```
classdef generator < handle

properties
(Same as lines 4-11 in program 3-23)

x_equilibrium
V_equilibrium
I_equilibrium
end

methods
```

```
11
12
  (Same as lines 7 through 21 in program 3-34)
      function x_equilibrium = set_equilibrium(obj, V, I, P, Q)
14
  (Same as lines 10 through 23 of program 3-28)
16
        obj.x_equilibrium = x_equilibrium;
18
        obj.V_equilibrium = V;
19
        obj.I_equilibrium = I;
20
21
22
      function [A, Bu, BV, BI, C, Du, DV, DI] =...
           get_linear_matrix(obj)
24
25
        X = obj.X;
26
        X_prime = obj.X_prime;
27
        D = obj.D;
28
        M = obj.M;
29
        tau = obj.tau;
30
31
        omega0 = obj.omega0;
32
        delta = obj.x_equilibrium(1);
34
        E = obj.x_equilibrium(3);
35
        V = obj.V_equilibrium;
        Vabs = abs(obj.V_equilibrium);
36
        Vangle = angle(obj.V_equilibrium);
37
        I = obj.I_equilibrium;
38
        A = [0, omega0, 0;
40
          0, -D/M, 0;
           -(X/X_prime-1)*Vabs*sin(delta-Vangle)/tau,...
41
          0, -X/X_prime/tau];
42
        Bu = [0; 1/M; 0];
43
        BV = [0, 0;
44
           -real(I)/M, -imag(I)/M;
45
           (X/X_prime-1)*cos(delta)/tau,...
46
           (X/X_prime-1)*sin(delta)/tau];
47
48
        BI = [0, 0;
49
           -real(V)/M, -imag(V)/M;
           0, 0];
50
        C = [E*cos(delta), 0, sin(delta);
51
           E*sin(delta), 0, -cos(delta)];
52
53
        Du = [0; 0];
        DV = [0, -1; 1, 0];
54
        DI = -X_prime*eye(2);
55
56
57
    end
58
59
60 end
```

Program 1.1 generator.m

In Program ??, lines 18 to 20 in set\_equilibrium store information about the equilibrium point used in the calculation of the approximate linear model.

If implemented similarly for the constant impedance load model, the program ?? is obtained.

```
classdef load_impedance < handle
    properties
      I_equilibrium
    methods
  (Same as lines 7 through 18 in program 3-35)
      function x_equilibrium = set_equilibrium(obj, V, I, P, Q)
        x_equilibrium = [];
        obj.z = -V/I;
14
15
        obj.I_equilibrium = I;
16
      function [A, Bu, BV, BI, C, Du, DV, DI] = ...
18
          get_linear_matrix(obj)
19
20
        A = [];
        Bu = zeros(0, 2);
23
        BV = zeros(0, 2);
24
        BI = zeros(0, 2);
        C = zeros(2, 0);
25
        I = obj.I_equilibrium;
26
27
        z = obj.z;
        Du = [real(z)*real(I), imag(z)*imag(I);
28
          real(z)*imag(I), imag(z)*real(I)];
29
        DV = eye(2);
30
        DI = [real(z), -imag(z); imag(z), real(z)];
31
32
33
34
    end
35
```

Program 1.2 load\_impedance.m

By using the class of equipment such as modified generators and loads, the function for obtaining an approximate linear model can be described as shown in Program ??.

```
function sys = get_linear_model(a_component, Y)

A = cell(numel(a_component), 1);
Bu = cell(numel(a_component), 1);
BV = cell(numel(a_component), 1);
BI = cell(numel(a_component), 1);
```

```
C = cell(numel(a_component), 1);
    Du = cell(numel(a_component), 1);
    DV = cell(numel(a_component), 1);
    DI = cell(numel(a_component), 1);
10
    for k = 1:numel(a_component)
      component = a_component{k};
13
      14
        component.get_linear_matrix();
15
16
17
    A = blkdiag(A{:});
    Bu = blkdiag(Bu{:});
19
    BV = blkdiag(BV{:});
20
21
    BI = blkdiag(BI{:});
    C = blkdiag(C{:});
22
    Du = blkdiag(Du{:});
    DV = blkdiag(DV{:});
24
    DI = blkdiag(DI{:});
25
26
27
    Ymat = zeros(size(Y, 1)*2, size(Y, 2)*2);
    Ymat(1:2:end, 1:2:end) = real(Y);
28
    Ymat(2:2:end, 1:2:end) = imag(Y);
Ymat(1:2:end, 2:2:end) = -imag(Y);
29
30
    Ymat(2:2:end, 2:2:end) = real(Y);
31
    nx = size(A, 1);
34
    A11 = A;
36
    A12 = [BV, BI];
    A21 = [C; zeros(size(Ymat, 1), nx)];
37
    A22 = [DV, DI; Ymat, -eye(size(Ymat))];
38
    B1 = Bu;
    B2 = [Du; zeros(size(Ymat, 1), size(Du, 2))];
41
42
43
    Aout = A11 - A12/A22*A21;
45
    Bout = B1 - A12/A22*B2;
    Cout = eye(nx);
46
    Dout = 0;
47
    sys = ss(Aout, Bout, Cout, Dout);
49
```

Program 1.3 get\_linear\_model.m

In lines 12 to 16 of Program ??, the coefficient matrix of the approximate linear model is obtained from each equipment. Additionally, by eliminating the voltage and current phases of all buses from lines 27 to 47, an expression for the approximate linear model's system of ordinary differential equations is obtained.

The approximate linear model can be used as follows by using Program ??.

```
(Same as lines 1 through 23 in Program 3-30)
sys = get_linear_model(a_component, Y);
sys = sys(2, 1);
nyquist(sys)
```

Program 1.4 main\_linearization.m

In this example, an approximate linear model is constructed in line 5 with the mechanical input  $P_{\text{mech1}}$  of generator 1 as input and the frequency deviation  $\Delta\omega_1$  of generator 1 as output. In addition, a Nyquist plot is drawn in line 6.

In the mathematical analysis of Section ??, an approximate linear model is derived from a nonlinear system of ordinary differential equations where all buses are Kron reduced. On the other hand, in the numerical implementation of this section, the nonlinear differential-algebraic equation system is first linearized, and then Kron reduction is applied to construct the ordinary differential equation system. It should be noted that this is because in the power system model with Kron reduction, expressions generally involve a mixture of information about equipment, buses, and transmission lines.

To increase the readability and expandability of the program, it is important to modularize each element appropriately, as in the implementation of this section.

#### 2.2 Numerical analysis of small signal stability

Let us perform a stability analysis based on approximate linearization for an actual electrical power system model consisting of three generators.

**Example 1.2** Numerical stability analysis of the linearized model Let us consider an electrical power system model consisting of three generators discussed in the Example ??. The constant of the generators and transmission lines are set to the same value as in the Example ??, and a linear approximation model for Equation ?? is derived with the approximate of the steady value shown in ??. Figure ?? shows the time response When the initial values are set as follows to correspond to Equation ??:

$$\delta^{\text{lin}}(0) = \begin{bmatrix} \frac{\pi}{6} \\ 0 \\ 0 \\ 0 \end{bmatrix}, \quad \Delta\omega^{\text{lin}}(0) = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \quad E^{\text{lin}}(0) = \begin{bmatrix} 0.1 \\ 0 \\ 0 \end{bmatrix}$$
 (14)

The blue, black, and red lines represent generators 1, 2, and 3, respectively. From this figure, we can see that the internal state of the generator group converges asymptotically as given in (??). Moreover, it approximately reproduces the initial value response of the nonlinear model shown in Fig. ??.

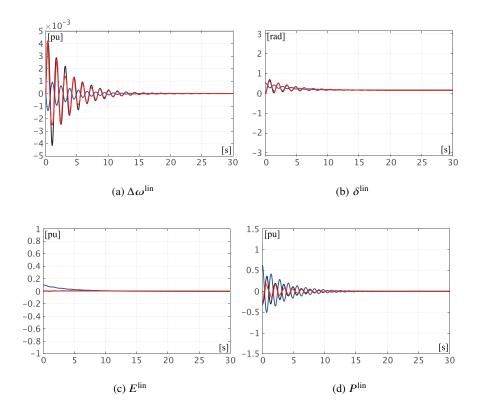


Fig. 2 Initial value response of approximate linear model (Blue: Generator 1, Black: Generator 2, Red: Generator 3)

Next, we parameterize the constants and steady-state values of the generators and transmission lines to analyze the stability of the resulting approximate linear model. For the generator constants, we compare the cases where all the damping coefficients are set to 10 and where they are set to 0.1, i.e., we consider the two cases:

$$(D_1, D_2, D_3) = (10, 10, 10), \qquad (D_1, D_2, D_3) = (0.1, 0.1, 0.1)$$

Other constants are set to the values in Table ??. In addition, the steady-state values of the rotor angle differences are expressed in terms of a parameter  $\theta_1 \in [0, 1]$  as follows:

$$\delta_{12}^{\star} = -\frac{\pi}{2}\theta_1, \qquad \delta_{13}^{\star} = \frac{\pi}{2}\theta_1$$
 (15)

Here,  $\theta_1$  is a parameter that specifies the magnitude of the rotor angle difference in the steady-state. By varying this value, the system matrix in ?? changes. Note that

the steady-state values of the internal voltages are not changed from the values in Table ??.

The admittance matrix is also modified as follows. Using the admittance values  $y_{12}$  and  $y_{23}$  in Equation ??, the admittance matrix of the power system in Equation ?? is constructed. The real part of this admittance matrix, which is the conductance matrix, is denoted as  $G_0$ , and the imaginary part, which is the susceptance matrix, is denoted as  $B_0$ . Specifically,

$$G_0 = \begin{bmatrix} 1.3652 & -1.3652 & 0 \\ -1.3652 & 3.3074 & -1.9422 \\ 0 & -1.9422 & 1.9422 \end{bmatrix},$$

$$B_0 = \begin{bmatrix} -11.6041 & 11.6041 & 0 \\ 11.6041 & -22.1148 & 10.5107 \\ 0 & 10.5107 & -10.5107 \end{bmatrix}$$
(16)

Using the parameter  $\theta_2 \in [0, 5]$ , we express the reference admittance matrix as follows:

$$Y_0(\theta_2) := \theta_2 G_0 + j B_0 \tag{17}$$

Here,  $\theta_2$  is a parameter that specifies the size of the real part (conductance matrix). For comparison, we consider two parameterized admittance matrices:

$$Y = Y_0(\theta_2), \qquad Y = \frac{Y_0(\theta_2)}{100}$$

The changes in the admittance matrix are approximately represented in the linearized model by the changes in the values of the reduced conductor  $B^{\text{red}}ij$  and the reduced susceptance  $G^{\text{red}}ij$  in equation ??. The parameter settings for the comparison are summarized in Table ??.

Let us numerically analyze the stability of the approximate linear model by varying the parameters  $(\theta_1, \theta_2)$  for each case (a)-(d) in Table ??. Specifically, we will check whether the approximate linear model is stable or not by examining the eigenvalues of  $\Psi$  in Equation ?? by varying  $\theta_1$  and  $\theta_2$  on a grid of 100 equidistant points each. The results are shown in Figure ??. The blue area represents the parameter region where the approximate linear model is stable. First, in the case of (a), we see that the approximate linear model is stable regardless of the size of the conductance matrix specified by  $\theta_2$ , as long as  $\theta_1$  is below approximately 0.4, which corresponds to a rotor angle difference of approximately 36° in the steady state. The same result is obtained for case (b), where the generator's damping coefficient is small at 0.1.

Next, we examine the results for cases (c) and (d), where the admittance matrix is multiplied by  $\frac{1}{100}$ . In this case, we find that when  $\theta_2$  is small and the size of the conductance matrix is around 1, the approximate linear model is stable as long as the rotor angle difference in the steady state is below approximately 76°. We also find that as  $\theta_2$  increases to 2 or more, the upper limit of the rotor angle difference for stability of the approximate linear model decreases.

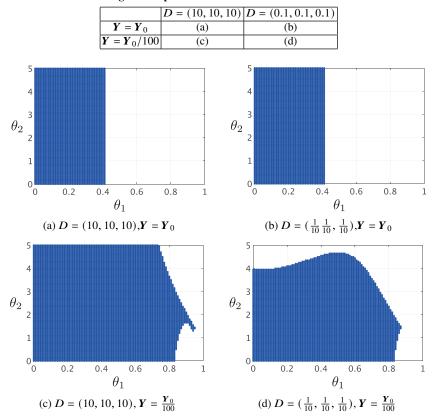


Table 1 Parameter settings to compare

Fig. 3 Area of parameters where the approximate linear model is stable

### 3 Mathematical stability analysis of the linearized model<sup>‡</sup>

### 3.1 Small signal stability of the linearized model $^{\ddagger}$

In this section, we mathematically analyze the stability of the linearized model given in equation  $\ref{eq:table_eq}$ . The stability is characterized by the eigenvalues of the matrix  $\Psi$ . However, as discussed in section  $\ref{eq:table_eq}$ ,  $\Psi$  is not regular and the eigenspace for the zero eigenvalue is given by

$$\mathcal{M} = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\} \tag{18}$$

This eigenspace represents the set of equivalent steady-state values obtained by varying the phase angles of all generators while maintaining their relative values constant. Therefore, it is not a problem which point in the equilibrium set of Equation ?? the state of the approximate linear model converges to. Based on this fact, we give the following definition:

#### - COFFEE BREAK -

**Eigenspace of a square matrix**: For a square matrix A and a given eigenvalue  $\lambda$ , the *eigenspace*  $V_{\lambda}$  is defined as:

$$V_{\lambda} := \ker(\lambda I - A)$$

where ker denotes the kernel of the matrix. If  $v_1, \ldots, v_k$  are all linearly independent eigenvectors corresponding to a specific eigenvalue  $\lambda$ , then:

$$\mathcal{V}_{\lambda} = \operatorname{span}\{v_1, \ldots, v_k\}$$

which means that it is a linear space spanned by the eigenvectors associated with a specific eigenvalue.

#### Definition 1.1 (Small-signal stability of the linear approximation model)

Consider the linearized model given by Equation  $\ref{eq:constraint}$ . For any initial values, if the internal state converges to one of the equilibrium points in the set  $\mathcal{M}$  defined by Equation  $\ref{eq:constraint}$ , the linearized model is said to be **steady-state stable**.

The small-signal stability in Definition ?? means that for any initial condition, Equation ?? holds. Note that the value of  $c_0$  in Equation ?? is arbitrary, so we express its arbitrariness as "converging to one of the equilibrium points in  $\mathcal{M}$ ."

In power system engineering, the term "small-signal stability" is widely used to discuss the stability of a power system against small disturbances using an approximate linear model. However, introducing a mathematical definition like Definition ?? is not common practice.

In the following discussion, we assume that the kernel space of  $\Psi$  in Equation ?? is one-dimensional and that Equation ?? holds:

$$\ker \Psi = \mathcal{M} \tag{19}$$

It is clear from the structure of the matrix  $\Psi$  that  $\mathcal{M}$  is a subset of ker  $\Psi$ , but it should be noted that we are assuming that ker  $\Psi$  is one-dimensional and that the equality holds. If the kernel space were two-dimensional or greater, the invariant eigenspace would be larger than  $\mathcal{M}$ , and the approximate linear model would not be in a steady state stability. Therefore, equation ?? is a necessary condition for the steady-state stability of the linear approximation model. In particular, when A is invertible, and using the definition of  $L_0 := L - CA^{-1}B$ , the necessary condition can be equivalently expressed as:

$$\ker L_0 = \operatorname{span} \{1\} \tag{20}$$

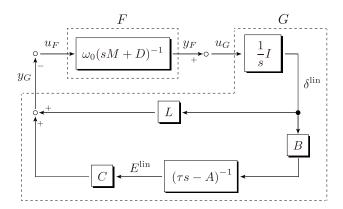


Fig. 4 Feedback system representation of approximate linear models

Note that this matrix  $L_0$  plays an important role in the subsequent analysis.

The relationship between equation  $\ref{eq:thm.1}$  and equation  $\ref{eq:thm.2}$  can be verified as follows. Because the (1,2) block of the matrix  $\Psi$  is invertible, the necessary and sufficient condition for the kernel of  $\Psi$  to be equal to  $\mathcal{M}$  is:

$$\ker \begin{bmatrix} -L - C \\ B & A \end{bmatrix} = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$$

In particular, when A is invertible, we have

$$\begin{bmatrix} -L - C \\ B & A \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = 0 \qquad \Longleftrightarrow \qquad L_0 x = 0, \qquad y = -A^{-1} B x$$

Thus, the necessary condition is equivalent to equation ??.

For the following discussion, we introduce the following fundamental terminology.

**Definition 1.2 (Stability of square matrix)** For a square matrix A, if all the real parts of its eigenvalues are negative, A is called **stable** or **asymptotically stable**.

#### 3.2 Passivity of approximate linear models<sup>‡</sup>

#### 3.2.1 Representation of Approximate Linear Models with Feedback

Let us consider describing the approximate linear model of Equation ?? as a feedback system of two subsystems (Figure ??). The first subsystem is described as a system of differential equations for the deviation of the angular frequency:

$$F: \begin{cases} M\Delta\dot{\omega}^{\text{lin}} = -D\Delta\omega^{\text{lin}} + u_F \\ y_F = \omega_0\Delta\omega^{\text{lin}} \end{cases}$$
 (21)

In this book, we refer to this subsystem as the **mechanical subsystem**. The mechanical subsystem is determined solely by the physical constants of the generator set  $(M_i, D_i)i \in IG$  and the reference angular frequency  $\omega_0$ , and does not depend on the steady-state values of the internal state  $(\delta^*, E^*)$ .

The second subsystem is described as a system of differential equations with respect to the rotor angle and internal voltage as follows:

$$G: \begin{cases} \dot{\delta}^{\text{lin}} = u_G \\ \tau \dot{E}^{\text{lin}} = AE^{\text{lin}} + B\delta^{\text{lin}} \\ y_G = CE^{\text{lin}} + L\delta^{\text{lin}} \end{cases}$$
(22)

We call this subsystem the **electrical subsystem** <sup>1</sup>. The electrical subsystem not only depends on the physical constants of the generator group,  $(\tau_i)i \in IG$ , but also on the steady-state values of the internal states,  $(\delta^*, E^*)$ . In fact, the system matrix (L, A, B, C) in equation **??** is a function of  $(\delta^*, E^*)$ .

If the two subsystems' inputs and outputs are coupled through negative feedback as follows:

$$u_F = -y_G, \qquad u_G = y_F \tag{23}$$

the approximate linear model of Equation ?? is represented. The subsequent analysis of steady-state stability is based on the property called the *passivity* of the mechanical and electrical subsystems. It is well-known that a negative feedback system of a passive subsystem is stable.

#### 3.2.2 Passivity of the mechanical subsystem

The mechanical subsystem F in Equation  $\ref{eq:property}$  has the following strong passivity:

#### **Definition 1.3 (Passivity of linear systems)**

Consider the linear system

$$\Sigma : \begin{cases} \dot{x} = Ax + Bu \\ y = Cx \end{cases}$$
 (24)

Using the symmetric matrix P, define the function:

$$W(x) := \frac{1}{2}x^{\mathsf{T}}Px \tag{25}$$

<sup>&</sup>lt;sup>1</sup> The terms "mechanical subsystem" and "electrical subsystem" introduced here are unique to this book.

For any u, if there exists a semi-positive definite matrix P that satisfies:

$$\frac{d}{dt}W(x(t)) \le u^{\mathsf{T}}(t)y(t), \qquad \forall t \ge 0$$
 (26)

Then, we call  $\Sigma$  passive.

In particular, when there exists a positive definite number  $\rho$  such that in addition to the semi-positive definite matrix, the inequality:

$$\frac{d}{dt}W(x(t)) \le u^{\mathsf{T}}(t)y(t) - \rho \|y(t)\|^2, \qquad \forall t \ge 0$$
(27)

is satisfied, then we call  $\Sigma$  strictly passive.

The function W(x) in Definition ?? is generally called the **storage function**. Moreover, the inequality in Equation ?? describes a type of passivity that is more strictly called **output-strict passivity**, where the energy represented by the storage function W(x) dissipates more quickly in proportion to the square of the output compared to the passivity in Equation ??.

The mechanical subsystem F of Equation?? having strong passivity can be confirmed as follows. First, the subsystem is written as below:

$$F: \begin{cases} \dot{x}_F = A_F x_F + B_F u_F \\ y_F = C_F x_F \end{cases}$$
 (28)

where state  $x_F$  represents  $\Delta \omega^{lin}$ , and the system matrices are:

$$A_F := -M^{-1}D, \qquad B_F := M^{-1}, \qquad C_F := \omega_0 I$$

Additionally, we define the symmetric matrix  $P_F$  as:

$$P_F := \omega_0 M$$

and since M is positive definite,  $P_F$  is also positive definite. Then, the following inequalities hold:

$$A_F^{\mathsf{T}} P_F + P_F A_F \leq -\frac{2 \min \left\{D_i\right\}}{\omega_0} C_F^{\mathsf{T}} C_F, \qquad P_F B_F = C_F^{\mathsf{T}}$$

Therefore, when the storage function is defined as:

$$W_F(x_F) := \frac{1}{2} x_F^{\mathsf{T}} P_F x_F \tag{29}$$

the time derivative along the solution trajectory of F is evaluated as:

$$\frac{d}{dt}W_F(x_F(t)) = \nabla W_F^{\mathsf{T}}(x_F) \frac{dx_F}{dt}$$

$$= (P_F x_F(t))^{\mathsf{T}} (A_F x_F(t) + B_F u_F(t))$$

$$= y_F^{\mathsf{T}}(t) u_F(t) + \frac{1}{2} x_F^{\mathsf{T}}(t) \left( A_F^{\mathsf{T}} P_F + P_F A_F \right) x_F(t)$$

$$\leq y_F^{\mathsf{T}}(t) u_F(t) - \frac{\min\{D_t\}}{\omega_0} \|y_F(t)\|^2$$
(30)

where  $\nabla W_F(x_F)$  is the gradient function obtained by partially differentiating  $W_F(x_F)$  with respect to its elements and arranging them vertically. From this, it can be seen that the machine subsystem F in Equation ?? is strongly passive for any positive definite  $(M_i, D_i)_{i \in I_G}$ . Note that the function  $W_F(x_F)$  represents the mechanical kinetic energy of the power system.

#### 3.2.3 Passivity of the electrical subsystem

Next, we consider the electrical subsystem G in Equation ??. Unlike the mechanical subsystem F, the electrical subsystem G only possesses passivity under limited conditions. While this may seem arbitrary, we consider the case where all reduced conductances in Equation ?? are zero, in other words:

$$G_{ij}^{\text{red}} = 0, \qquad \forall (i, j) \in I_{G} \times I_{G}$$
 (31)

Excluding special cases, the condition in Equation ?? only holds when the conductance of all transmission lines in the power system is zero, or equivalently, the resistance of all transmission lines is zero. In this case, for the functions  $k_{ij}(\delta_{ij})$  and  $h_{ij}(\delta_{ij})$  defined in Equation ??, the following holds:

$$k_{ij}(\delta_{ij}^{\star}) = k_{ji}(\delta_{ii}^{\star}), \qquad h_{ij}(\delta_{ij}^{\star}) = -h_{ji}(\delta_{ii}^{\star}), \qquad h_{ii}(\delta_{ii}^{\star}) = 0$$

Therefore, for the system matrix (L, A, B, C) in Equation ??, it holds that:

$$L = L^{\mathsf{T}}, \qquad \hat{A} = \hat{A}^{\mathsf{T}}, \qquad C = -\hat{B}^{\mathsf{T}}$$
 (32)

In the following, we use the symmetric structure of this special system matrix to analyze the passivity of the electrical subsystem.

First, let us express the electrical subsystem G of Equation ?? as follows:

$$G: \begin{cases} \dot{x}_G = A_G x_G + B_G u_G \\ y_G = C_G x_G \end{cases}$$
 (33)

where the state  $x_G$  is a column vector obtained by concatenating  $\delta^{\text{lin}}$  and  $E^{\text{lin}}$ , and  $\Omega$  is a positive definite diagonal matrix defined as follows:

$$\Omega := \mathsf{diag}\!\left(\sqrt{rac{X_i - X_i'}{ au_i}}
ight)_{i \in I_{\mathsf{G}}}$$

The system matrices are expressed as:

$$A_G := \begin{bmatrix} 0 & 0 \\ \Omega^2 \hat{B} & \Omega^2 \hat{A} \end{bmatrix}, \qquad B_G := \begin{bmatrix} I \\ 0 \end{bmatrix}, \qquad C_G := \begin{bmatrix} L - \hat{B}^{\mathsf{T}} \end{bmatrix}$$

Furthermore, we define the symmetric matrix  $P_G$  as follows:

$$P_G := \begin{bmatrix} L & -\hat{B}^{\mathsf{T}} \\ -\hat{B} & -\hat{A} \end{bmatrix} \tag{34}$$

The following inequalities hold for these matrices:

$$A_G^{\mathsf{T}} P_G + P_G A_G \le 0, \qquad P_G B_G = C_G^{\mathsf{T}} \tag{35}$$

If we calculated the left of the inequality, it can be expressed as follows using a symmetric matrix  $\hat{A}_{\Omega} := \Omega \hat{A}\Omega$ :

$$\frac{A_G^\mathsf{T} P_G + P_G A_G}{2} = \begin{bmatrix} \Omega \hat{B} & 0 \\ 0 & \Omega^{-1} \end{bmatrix}^\mathsf{T} \underbrace{\begin{bmatrix} -I & -\hat{A}_\Omega \\ -\hat{A}_\Omega & -\hat{A}_\Omega^2 \end{bmatrix}}_{V} \begin{bmatrix} \Omega \hat{B} & 0 \\ 0 & \Omega^{-1} \end{bmatrix}$$

Here, the top-left block -I of Y is negative definite, and the Schur complement of Y with respect to -I is 0, which implies that Y is negative semi-definite. Therefore, the matrix inequality in Equation ?? holds.

#### COFFEE BREAK

#### Schur complement:

Let a symmetric matrix M be partitioned as

$$M = \left[ \begin{array}{c} M_{11} & M_{12} \\ M_{12}^\mathsf{T} & M_{22} \end{array} \right]$$

Then, the **Schur complement** of M with respect to  $M_{22}$  is defined as:

$$M/M_{22} := M_{11} - M_{12}M_{22}^{-1}M_{12}^{\mathsf{T}}$$

Similarly, the Schur complement of M with respect to  $M_{11}$  is defined as:

$$M/M_{11} := M_{22} - M_{12}^{\mathsf{T}} M_{11}^{-1} M_{12}$$

If the matrix  $M_{22}$  is positive definite, then M is positive semidefinite if and only if  $M/M_{22}$  is positive semidefinite. The same fact holds for the Schur complement of M with respect to  $M_{11}$  [?]. The same fact holds if we replace positive semidefinite with positive definite.

#### **Properties of semidefinite matrices:**

For any positive semidefinite (or negative semidefinite) matrix  $Y \in \mathbb{R}^{n \times n}$  and any matrix  $X \in \mathbb{R}^{n \times m}$ , the matrix  $X^T Y X$  is positive semidefinite (or negative semidefinite). This can be shown from the fact that

$$v^{\mathsf{T}}Yv \ge 0, \quad \forall v \in \mathbb{R}^n \Longrightarrow (Xw)^{\mathsf{T}}Y(Xw) \ge 0, \quad \forall w \in \mathbb{R}^m$$

By using the relationship given by Equation ??, the time derivative of the storage function  $W_G(x_G)$  along the solution trajectory of G can be evaluated, where  $W_G(x_G)$  is defined by Equation ??, similarly to Equation ??:

$$W_G(x_G) := \frac{1}{2} x_G^\mathsf{T} P_G x_G \tag{36}$$

$$\frac{d}{dt}W_G(x_G(t)) \le y_G^{\mathsf{T}}(t)u_G(t) \tag{37}$$

However, to show the passivity of G,  $P_G$  in equation  $\ref{eq:condition}$  must be positive semi-definite. If the matrix A in equation  $\ref{eq:condition}$  is stable, then it can be shown that:

$$A = S^2 \hat{A} \iff S^{-1} A S = S \hat{A} S$$

where  $S := \operatorname{diag}\left(\sqrt{X_i - X_i'}\right)i \in IG$ . Here,  $\hat{A}$  is negative definite. Under this condition, the necessary and sufficient condition for  $P_G$  in Equation ?? to be positive semi-definite is that the Schur complement of  $-\hat{A}$  is positive semi-definite, in other words:

$$L_0 = L_0^{\mathsf{T}} \ge 0 \tag{38}$$

where  $L_0$  is defined by Equation ?? and can be expressed as  $L_0 = L + \hat{B}^T \hat{A}^{-1} \hat{B}$  using Equation ??. To summarize the above discussion, the following definition is introduced.

**Definition 1.4 (Passive power transmission condition)** For the system matrix (L, A, B, C) of Equation ??, the following three conditions are together called **passive power transmission conditions**. <sup>2</sup>

- (i) Matrix A is stable.
- (ii) As in Equation ??, all reduced conductances are zero.
- (iii) For the matrix  $L_0$  of Equation ??, the matrix inequality of Equation ?? holds.

Each of these conditions may be referred to individually as the passive transmission condition (i), and so on.

<sup>&</sup>lt;sup>2</sup> "Passive power transmission conditions" is a term unique to this book.

Based on the above discussions, we can see that the passive power transmission conditions describe the conditions necessary for the electrical system G of Equation  $\ref{eq:conditions}$  to be passive. Furthermore, these conditions are necessary for the linear approximation model to be statically stable for the passivity of an electrical subsystem and arbitrary physical constant. The details are discussed in Section  $\ref{eq:conditions}$  and Section  $\ref{eq:conditions}$ . Function  $\ref{eq:conditions}$  indicates the electrical potential energy of an electrical power system.

#### 3.3 Analysis of small signal stability based on passivity<sup>‡</sup>

#### 3.3.1 Stability analysis of feedback systems

In the following, under the passive power transmission conditions defined in Definition ??, the small signal stability of the linear approximation model given in Equation ?? is analyzed for electric subsystems that are passive. The stability of their feedback systems is also analyzed.

Since the inequalities in Equations ?? and ?? hold, their sum is given by:

$$\frac{d}{dt} \left\{ W_F \left( x_F(t) \right) + W_G \left( x_G(t) \right) \right\}$$

$$\leq \underbrace{y_F^\mathsf{T}(t) u_F(t) + y_G^\mathsf{T}(t) u_G(t)}_{t} - \underbrace{\frac{\min\{D_i\}}{\omega_0} \|y_F(t)\|^2}_{t}$$

By substituting the feedback coupling equation in Equation ?? into this inequality, the term indicated by "\*\[ \\*\ "\] is cancelled out, and the inequality for the entire feedback system can be expressed as:

$$\frac{d}{dt} \{ W_F(x_F(t)) + W_G(x_G(t)) \} \le -\frac{\min\{D_i\}}{\omega_0} \|y_F(t)\|^2$$
 (39)

In other words, the sum of the functions  $W_F(x_F)$  and  $W_G(x_G)$  is monotonically non-increasing with respect to the time evolution along the feedback system trajectory. Furthermore, since the lower bounds of  $W_F(x_F)$  and  $W_G(x_G)$  are both 0, their sum asymptotically converges to a certain value as time passes sufficiently. This means that the value of the time derivative on the left-hand side of Equation ?? converges to 0. Additionally, since the right-hand side of Equation ?? is negative when  $y_F(t) \neq 0$  and is only 0 when  $y_F(t) = 0$ , the following is obtained:

$$\lim_{t \to \infty} y_F(t) = 0 \tag{40}$$

Furthermore, focusing on the output equation of Equation ??, since the output  $y_F$  is a constant multiple of the internal state  $\Delta\omega^{\text{lin}}$ , it can be understood that for the mechanical subsystem F:

$$y_F(t) = 0, \quad \forall t \ge 0 \qquad \Longrightarrow \qquad \Delta \omega^{\text{lin}}(t) = 0, \quad \forall t \ge 0$$
 (41)

This is a property called **observability** in system control engineering. Therefore, from Equations **??** and **??**, for the approximated linear model of Equation **??**, for any initial value  $(\Delta\omega^{\text{lin}}(0), \delta^{\text{lin}}(0), E^{\text{lin}}(0))$ , the following holds:

$$\lim_{t \to \infty} \Delta \omega^{\text{lin}}(t) = 0 \tag{42}$$

#### COFFEE BREAK

**Observability**: For a linear system  $\Sigma$  described by Equation  $\ref{eq:condition}$ , if the output y(t) is identically zero, then the internal state x(t) is also identically zero, and  $\Sigma$  is said to be **observable** in this case. The necessary and sufficient condition for  $\Sigma$  to be observable is given by Equation  $\ref{eq:condition}$ , where n is the dimension of the state.

$$\ker \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} = \{0\}$$
(43)

In some contexts, a pair of matrices that satisfies Equation  $\ref{eq:context}$  is referred to as an observable pair (C, A).

**Controllability**: For a linear system  $\Sigma$  described by Equation  $\ref{eq:controllability}$ , if there exists an input u(t) such that for all initial states x(0), there exists a time T>0 such that x(T)=0, then  $\Sigma$  is said to be **controllable**. The necessary and sufficient condition for  $\Sigma$  to be controllable is given by Equation  $\ref{eq:controllable}$ , where n is the dimension of the state.

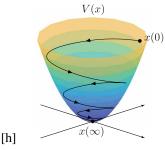
$$\operatorname{im}\left[B A B \cdots A^{n-1} B\right] = \mathbb{R}^n \tag{44}$$

Here, im denotes the **image** of the matrix. In some contexts, a pair of matrices that satisfies Equation ?? is referred to as a controllable pair (A, B)

**Lyapunov function**: Consider the observable linear system  $\Sigma$  described by Equation ??, where the input u(t) is identically zero. Let V(x) be a positive semi-definite function that satisfies  $V(x) \ge 0$  for all x and V(0) = 0. If there exists a positive constant  $\rho$  such that the derivative of V(x) along the solution trajectory of  $\Sigma$  satisfies

$$\frac{d}{dt}V\big(x(t)\big) = \nabla V^\mathsf{T}(x)\frac{dx}{dt}(t) \le -\rho\|y(t)\|^2, \qquad \forall t \ge 0$$

then the solution trajectory x(t) converges asymptotically to zero for any initial state. Such a function V(x) is called a **Lyapunov function**.



The fact that the value of the Lyapunov function decreases monotonically along the solution trajectory of the system can be interpreted as a type of energy dissipation over time (Figure ??). Similar stability analyses based on Lyapunov functions can also be applied to nonlinear systems.

On the one hand, it is not possible to deduce from the above discussion whether the internal state of the electric subsystem G in Equation ?? converges asymptotically to 0. Specifically, from the relation in Equation ?? and the asymptotic convergence in Equation ??, it can be derived that the input and output of the two subsystems satisfy:

$$\lim_{t\to\infty} u_F(t) = 0, \qquad \lim_{t\to\infty} u_G(t) = 0, \qquad \lim_{t\to\infty} y_G(t) = 0$$

However, since the electric subsystem is not observable, it cannot be concluded that its internal state converges asymptotically to zero. Assuming that the electric subsystem is observable, it can be concluded that for any initial values:

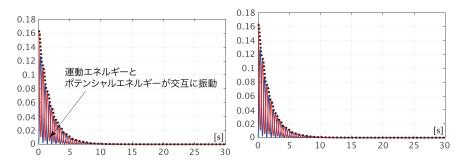
$$\lim_{t \to \infty} \delta^{\text{lin}}(t) = 0, \qquad \lim_{t \to \infty} E^{\text{lin}}(t) = 0$$

However, this implies that  $c_0$  must always be equal to 0 in Equation ??. This fact contradicts the instability of Equation ?? due to the zero eigenvalue of  $\Psi$ . It should be noted that except for some special cases, the electric subsystem is controllable.

**Example 1.3** Time evolution of stored energy Consider the approximated linear model discussed in the first half of Example ??. First, consider the case where the passive transmission condition (ii) is satisfied, i.e., when the conductance of both transmission lines is 0. Specifically, set the admittance values of the transmission lines to:

$$y_{12} = -j11.6041, y_{23} = -j10.5107 (45)$$

This corresponds to setting the parameter  $\theta_2$  to 0 in Equation ??. In this case, the matrix A becomes stable. Also, all the eigenvalues of  $L_0$  in Equation ?? become non-negative. Thus, the passive transmission conditions (i) and (iii) hold.



- (a) If passive transmission condition (ii) is met
- (b) If passive transmission condition (ii) is not met

Fig. 5 Time variation of the accumulation function according to Example ?? (Blue:  $W_F$ , Red:  $W_G$ , Black:  $W_F + W_G$ )

Consider the time response of the initial values in Equation  $\ref{eq:constraint}$  and calculate the time variation of the kinetic energy  $W_F(x_F)$  in Equation  $\ref{eq:constraint}$ ? and the potential energy  $W_G(x_G)$  in Equation  $\ref{eq:constraint}$ ?. The calculation results are shown in Figure  $\ref{eq:constraint}$ ? (a) where the blue and red solid lines represent  $W_F(x_F)$  and  $W_G(x_G)$ , respectively, and the black dashed line represents their sum, which is the total energy of the system. From this figure, it can be seen that while the kinetic and potential energies increase and decrease alternatively, the total energy of the system, which is the sum of these energies, decreases monotonically. The decrease in total energy over time can be interpreted as energy loss due to friction caused by the damping coefficient.

Next, as a reference, let us show the results when the passive power transmission condition (ii) is not satisfied. Specifically, we set  $Y_0(1)$  in Equation ?? to be the admittance matrix Y by setting  $\theta_2$  to 1. This is equivalent to calculating the time variation of the kinetic and potential energies for the initial value response in Figure ??. Note that when the passive power transmission condition (ii) is not satisfied,  $P_G$  in Equation ?? does not become a symmetric matrix, but the potential energy  $W_G(x_G)$  can still be calculated using the definition in Equation ??. The calculation result in Figure ??(b) is almost identical to that in Figure ??(a). This fact suggests that even when the conductance of the transmission line is not zero, the electrical potential energy can be approximately calculated based on the definition in Equation ??.

#### 3.3.2 Basis transformation for separating unobservable state variables

Let us consider deriving an observable subsystem from the electrical subsystem G in Equation  $\ref{eq:continuous}$  by removing the common component of unobservable rotor angles.

Specifically, we apply a basis transformation to the state  $\delta^{lin}$  in Equation ?? to derive a set of differential equations describing only the deviations of the rotor angles. Note that the following basis transformation is always applicable regardless of whether the passive power delivery condition holds or not.

#### COFFEE BREAK -

#### Basis transformation of linear systems:

In the state equation:

$$\dot{x}(t) = Ax(t) + Bu(t)$$

each element  $x_i(t)$  of the *n*-dimensional state vector x(t) can be represented as the component of the expansion with respect to the basis  $e_1, \ldots, e_n$ , such as:

$$x(t) = e_1 x_1(t) + \dots + e_n x_n(t)$$

where  $e_i$  is an *n*-dimensional vector with only the *i*-th element being 1. This basis is called the **standard basis** and represents the time evolution of the "components" of the state vector x(t) in a certain basis. We now consider representing x(t) in another basis  $v_1, \ldots, v_n$  such that

$$x(t) = v_1 \xi_1(t) + \cdots + v_n \xi_n(t)$$

where  $\xi_i(t)$  is the component of the basis vector  $v_i$ . Let us denote the matrix obtained by arranging the vectors  $v_i$  horizontally as V and the vector obtained by arranging  $\xi_i(t)$  vertically as  $\xi(t)$ . Then, the linear transformation  $x(t) = V\xi(t)$  corresponds to this representation. In this case, the state equation is transformed as:

$$\dot{\xi}(t) = V^{-1}AV\xi(t) + V^{-1}Bu(t)$$

where  $V_a$  and  $V_b$  are the matrices obtained by arranging the basis vectors of  $\mathcal{V}_a$  and  $\mathcal{V}_b$  horizontally, respectively, the transformed state equation is obtained as:

$$x(t) = V_a \xi_a(t) + V_b \xi_b(t)$$

where  $W_a$  and  $W_b$  are matrices that satisfy:

$$\begin{bmatrix} \dot{\xi}_a(t) \\ \dot{\xi}_b(t) \end{bmatrix} = \begin{bmatrix} W_a A V_a \ W_a A V_b \\ W_b A V_a \ W_b A W_b \end{bmatrix} \begin{bmatrix} \xi_a(t) \\ \xi_b(t) \end{bmatrix}$$

However,  $V_a$  and  $V_b$  are matrices consisting of the basis vectors of  $\mathcal{V}a$  and  $\mathcal{V}b$ , respectively. Wa and Wb are matrices that satisfy:

$$\begin{bmatrix} W_a \\ W_b \end{bmatrix} = \begin{bmatrix} V_a V_b \end{bmatrix}^{-1} \qquad \Longleftrightarrow \qquad \begin{bmatrix} V_a V_b \end{bmatrix} \begin{bmatrix} W_a \\ W_b \end{bmatrix} = I$$

In this representation,  $\xi_a(t)$  represents the component of x(t) with respect to the subspace span  $\mathcal{V}_a$ . Similarly,  $\xi_b(t)$  represents the component of x(t) with respect to the subspace span  $\mathcal{V}_b$ .

The change of basis explained below can be applied regardless of whether passive power transmission conditions hold.  $\delta^{\text{lin}}$  is expanded using a matrix  $W \in \mathbb{R}^{N \times (N-1)}$ :

$$\delta^{\text{lin}} = W \delta_{\text{e}}^{\text{lin}} + 1 \overline{\delta}_{\text{e}}^{\text{lin}} \tag{46}$$

Here, 1 is a base vector that expresses the common component of  $\delta^{\rm lin}$ , while W is a matrix with base vectors that express other deviation components. In other words, the common component of  $\delta^{\rm lin}$  is  $\overline{\delta}_{\rm e}^{\rm lin}$ , and deviation components are  $\delta_{\rm e}^{\rm lin}$ . The common component  $\overline{\delta}_{\rm e}^{\rm lin}$  is one-dimensional, while deviation components  $\delta_{\rm e}^{\rm lin}$  are (N-1)-dimensional.

Next, let us consider the inverse transformation of Equation ??. Specifically, let us consider a matrix  $W^{\dagger} \in \mathbb{R}^{(N-1) \times N}$ :

$$\delta^{\text{lin}} = \underbrace{\left[ \begin{array}{c} W \ 1 \end{array} \right]}_{T} \left[ \begin{array}{c} \delta_{\text{e}}^{\text{lin}} \\ \overline{\delta}_{\text{e}}^{\text{lin}} \end{array} \right] \quad \Longleftrightarrow \quad \left[ \begin{array}{c} \delta_{\text{e}}^{\text{lin}} \\ \overline{\delta}_{\text{e}}^{\text{lin}} \end{array} \right] = \underbrace{\left[ \begin{array}{c} W^{\dagger} \\ \frac{1}{N} \mathbb{1}^{T} \end{array} \right]}_{T^{-1}} \delta^{\text{lin}}$$

For this inverse transformation to exist, the column vector of W must be orthogonal to  $\mathbb{1}$ . This can be confirmed as follows. From the relationship of inverse transformation:

$$T^{-1}T = \begin{bmatrix} W^{\dagger}W & W^{\dagger}\mathbb{1} \\ \frac{1}{N}\mathbb{1}^{\mathsf{T}}W & \frac{1}{N}\mathbb{1}^{\mathsf{T}}\mathbb{1} \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & 1 \end{bmatrix}$$

must hold. In other words, W and  $W^{\dagger}$  are matrices that satisfy:

$$1 W = 0, W^{\dagger} = I, W^{\dagger} = 0$$

Therefore, from the first equation, we can see that the column vector of W is orthogonal to  $\mathbb{1}$ . W and  $W^{\dagger}$  can be constructed by using an appropriate matrix,  $U \in \mathbb{R}^{N \times (N-1)}$ , that satisfies

$$W = U(U^{\mathsf{T}}U)^{-1}, \qquad W^{\dagger} = U^{\mathsf{T}}$$

At this time, the product  $WW^{\dagger}$  is an orthogonal projection matrix to the orthogonal complement of span{1}; thus, this can be expressed as:

$$WW^{\dagger} = I - \frac{1}{N} \mathbb{1} \mathbb{1}^{\mathsf{T}} \tag{47}$$

Such a pseudo inverse matrix of W,  $W^{\dagger}$ , is called a **Moore-Penrose pseudoinverse**. The concept of orthogonal projection is shown in Figure ??. The black arrow shows the space of span $\{1\}$ , as the orthogonal complement span $\{1\}$ , a plane that

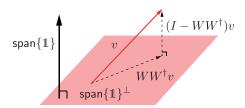


Fig. 6 Conceptual diagram of orthogonal projection

is orthogonal to the space. If the vector v is multiplied by orthogonal projection matrix  $WW^{\dagger}$ ,  $WW^{\dagger}v$  is obtained as an image projected perpendicular to span $\{1\}^{\perp}$  from v.

$$I - WW^{\dagger} = \frac{1}{N} \mathbb{1} \mathbb{1}^{\mathsf{T}}$$

in a complimentary relationship is an orthogonal projection matrix to  $span\{1\}$ .

We apply the above-mentioned change of basis to the electrical subsystem G of Equation  $\ref{eq:condition}$ , irst, if we substitute Equation  $\ref{eq:condition}$  into a differential equation related to  $\delta^{\text{lin}}$ , the following is obtained:

$$W\dot{\delta}_{\rm e}^{\rm lin} + 1\!\!1\!\!1\dot{\delta}_{\rm e}^{\rm lin} = u_G$$

If this differential equation is multiplied by  $W^{\dagger}$  or  $\frac{1}{N}\mathbb{1}^{\mathsf{T}}$  from the left, the following is obtained:

$$\dot{\delta}_{\rm e}^{\rm lin} = W^{\dagger} u_G, \qquad \dot{\overline{\delta}}_{\rm e}^{\rm lin} = \frac{1}{N} \mathbb{1}^{\mathsf{T}} u_G$$

Next, if we pay attention such that the relationship of Equation ?? holds for matrices L and B, the differential equation and output equation related to  $E^{\text{lin}}$  can be rewritten as:

$$\tau \dot{E}^{\rm lin} = A E^{\rm lin} + B W \delta_{\rm e}^{\rm lin}, \qquad y_G = C E^{\rm lin} + L W \delta_{\rm e}^{\rm lin}$$

Therefore, the electrical subsystem with a changed basis is obtained as:

$$G: \begin{cases} & \vec{\delta}_{e}^{\text{lin}} = \frac{1}{N} \mathbb{1}^{\mathsf{T}} u_{G} \\ & \hat{\delta}_{e}^{\text{lin}} = W^{\dagger} u_{G} \\ & \tau \dot{E}^{\text{lin}} = A E^{\text{lin}} + B W \delta_{e}^{\text{lin}} \\ & y_{G} = C E^{\text{lin}} + L W \delta_{e}^{\text{lin}} \end{cases}$$
(48)

What we need to focus on in this system expression is that  $\overline{\delta}_e^{\text{lin}}$  that expresses the common component of  $\delta^{\text{lin}}$  is impacted by the input  $u_G$ , but not the output  $y_G$ . In other words,  $\overline{\delta}_e^{\text{lin}}$  is an unobservable state variable.

other words,  $\overline{\delta}_{\rm e}^{\rm lin}$  is an unobservable state variable.

By removing the differential equation of  $\overline{\delta}_{\rm e}^{\rm lin}$  from Equation ??, (N-1)-dimensional controllable and observable subsystem is obtained as:

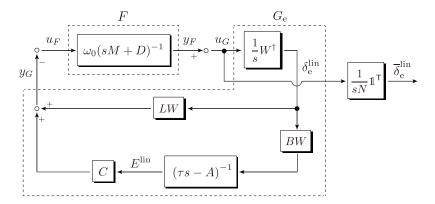


Fig. 7 Basis-transformed approximate linear model

$$G_{e}: \begin{cases} \delta_{e}^{\text{lin}} = W^{\dagger} u_{G} \\ \tau \dot{E}^{\text{lin}} = AE^{\text{lin}} + BW \delta_{e}^{\text{lin}} \\ y_{G} = CE^{\text{lin}} + LW \delta_{e}^{\text{lin}} \end{cases}$$
(49)

Here, please note that from observability of  $G_{\rm e}$ , the following holds.

$$y_G(t) = 0, \quad \forall t \ge 0 \qquad \Longrightarrow \qquad \begin{bmatrix} \delta_e^{\text{lin}}(t) \\ E^{\text{lin}}(t) \end{bmatrix} = 0, \quad \forall t \ge 0$$
 (50)

For the analysis of the small signal stability of the linear approximation model of Equation ??, this fact is important. For reference, Figure ?? shows a block diagram of the linear approximation model with a change of basis.

The set  $(\tau^{-1}A, \tau^{-1}B)$  being controllable while the set  $(C, \tau^{-1}A)$  being observable is the necessary condition for  $G_e$  to be controllable and observable. Below, we assume this controllability and observability in our discussions. Strict proof is not easy, but since the rank of B and C is (N-1) or higher except for special situations, such assumptions do not interfere with a realistic analysis.

#### 3.3.3 Analysis of small signal stability based on passivity

Below, we assume the passive power transmission conditions of definition  $\ref{Ge}$  and show the passivity of  $G_e$  of Equation  $\ref{Ge}$  using the same steps as the electrical system G of Equation  $\ref{Ge}$ . To that end, we used the following expression:

$$G_{e}: \begin{cases} \dot{x}_{G_{e}} = A_{G_{e}} x_{G_{e}} + B_{G_{e}} u_{G} \\ y_{G} = C_{G_{e}} x_{G_{e}} \end{cases}$$
 (51)

 $x_{G_e}$  and  $\delta_e^{\text{lin}}$  are vectors with  $E^{\text{lin}}$ :

$$A_{G_{\mathrm{e}}} := \begin{bmatrix} 0 & 0 \\ \Omega^2 \hat{B} W & \Omega^2 \hat{A} \end{bmatrix}, \quad B_{G_{\mathrm{e}}} := \begin{bmatrix} W^{\dagger} \\ 0 \end{bmatrix}, \quad C_{G_{\mathrm{e}}} := \begin{bmatrix} LW - \hat{B}^{\mathsf{T}} \end{bmatrix}$$

A positive semi-definite matrix  $P_{G_e}$  is defined as:

$$P_{G_{c}} := \begin{bmatrix} W & 0 \\ 0 & I \end{bmatrix}^{\mathsf{T}} \underbrace{\begin{bmatrix} L & -\hat{B}^{\mathsf{T}} \\ -\hat{B} & -\hat{A} \end{bmatrix}}_{P_{G}} \begin{bmatrix} W & 0 \\ 0 & I \end{bmatrix}$$
 (52)

If  $P_G$  of Equation ?? is positive semi-definite,  $P_{G_e}$  is also positive semi-definite.

$$A_{G_e}^{\mathsf{T}} P_{G_e} + P_{G_e} A_{G_e} \le 0, \qquad P_{G_e} B_{G_e} = C_{G_e}^{\mathsf{T}}$$
 (53)

At this time, because of the relationship of Equation ??, we can see that:

$$\frac{A_{G_{e}}^{\mathsf{T}} P_{G_{e}} + P_{G_{e}} A_{G_{e}}}{2} = \begin{bmatrix} \Omega \hat{B} W & 0 \\ 0 & \Omega^{-1} \end{bmatrix}^{\mathsf{T}} \underbrace{\begin{bmatrix} -I & -\hat{A}_{\Omega} \\ -\hat{A}_{\Omega} & -\hat{A}_{\Omega}^{2} \end{bmatrix}}_{V} \begin{bmatrix} \Omega \hat{B} W & 0 \\ 0 & \Omega^{-1} \end{bmatrix}$$

Therefore, the time derivative along the solution trajectory of  $G_e$  of the storage function:

$$W_{G_{e}}(x_{G_{e}}) := \frac{1}{2} x_{G_{e}}^{\mathsf{T}} P_{G_{e}} x_{G_{e}}$$

can be evaluated as:

$$\frac{d}{dt}W_{G_{e}}(x_{G_{e}}(t)) \le y_{G}^{\mathsf{T}}(t)u_{G}(t) \tag{54}$$

In other words,  $G_e$  of Equation ?? is passive. This inequality and inequality of Equation ?? are equivalent, and the following is true for the values of two storage functions:

$$W_G(x_G(t)) = W_{G_0}(x_{G_0}(t)), \quad \forall t \ge 0$$

By considering the observability of  $G_e$  shown by Equation ??, the following is true for the arbitrary initial value of the solution trajectory of the linear approximation model of Equation ??:

$$\lim_{t \to \infty} \Delta \omega^{\text{lin}}(t) = 0, \qquad \lim_{t \to \infty} \begin{bmatrix} \delta_{\text{e}}^{\text{lin}}(t) \\ E^{\text{lin}}(t) \end{bmatrix} = 0$$
 (55)

Therefore, from the relationship of the change of basis of Equation ??, we can see that Equation ?? holds for the arbitrary initial value. In other words, the linear approximation model of Equation ?? is statically stable. Also:

$$c_0 = \lim_{t \to \infty} \overline{\delta}_{\mathrm{e}}^{\mathrm{lin}}(t)$$

and state variables  $\overline{\delta}_e^{\text{lin}}$  follow the differential equation of Equation ??.

Let us theorize the above discussion in the following Theorem.

[Small signal stability of the linear approximation model based on passivity] For the arbitrary steady value  $(\delta^*, E^*)$  that satisfies the passive power transmission conditions of definition ??, the electrical subsystem G of Equation ?? is passive. For the arbitrary positive definite numbers  $(M_i, D_i, \tau_i)_{i \in I_G}$ , the linear approximation model of Equation?? is statically stable.

As discussed in Theorem ??, under the passive power transmission conditions, the linear approximation model is statically stable for combinations of all physical constants  $(M_i, D_i, \tau_i)_{i \in I_G}$ . Analysis based on passivity allows stability independent of model parameters.

## 3.4 Necessary conditions for the linear approximation model to be passive<sup>‡</sup>

#### 3.4.1 Passivity and positive realness

Passivity of a linear system is mathematically equivalent to a property called positive realness of a transfer function. In this Section, based on this equivalence, the necessity of the passive power transmission conditions of definition ?? is discussed from the viewpoint of the passivity of an electrical subsystem.

#### COFFEE BREAK -

**Transfer function**: For a linear system:

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

its transfer function is defined as:

$$Q(s) := C(sI - A)^{-1}B + D$$

When the Laplace transform of the input u(t) is U(s), and the Laplace transform of the output y(t) is Y(s), there is a relationship, Y(s) = Q(s)U(s), with a transfer function. The input/output characteristics of a linear system are characterized by the transfer function.

The transfer function from the input  $u_G$  to output  $y_G$  of the electrical subsystem G of Equation  $\ref{eq:condition}$  is:

$$G(s) := -\frac{1}{s} \underbrace{\left\{ -C(\tau s - A)^{-1}B - L \right\}}_{H(s)}$$

$$(56)$$

Since unobservable state variables are not related to input/output characteristics, please note that the transfer function of  $G_e$  of Equation ?? is also equal to G(s). Below, let us consider a situation where the transfer function H(s) of Equation ?? is stable. The stability of the transfer function is defined as follows:

**Definition 1.5 (Stability of a transfer function)** When the real part of all poles of the transfer function Q(s) is negative, Q(s) is called **stable**.

The pole of a transfer function is the zero point of a denominator polynomial. The fact that H(s) of Equation ?? is stable is equivalent to the real part of all eigenvalues of the matrix  $\tau^{-1}A$  being negative.

**Positive realness** of a transfer function is defined as follows:

**Definition 1.6 (Positive realness of a transfer function)** For a square transfer function Q(s), the following is defined:

$$\Omega_0 := \{ \omega_0 \in \mathbb{R} : \ \mathbf{j}\omega_0 Q(s) \} \tag{57}$$

When the following three conditions are satisfied, Q(s) is called **positive real**.

- The real part of all poles of Q(s) is nonpositive.
- For all  $\omega \in [0, \infty) \setminus \Omega_0$ ,  $Q(j\omega) + Q^{\mathsf{T}}(-j\omega)$  is positive semi-definite.
- When there are poles of a pure imaginary number, their multiplicity is 1, and the following is true for the remaining number:

$$\lim_{s\to \boldsymbol{j}\,\omega_0}(s-\boldsymbol{j}\omega_0)Q(s)=\lim_{s\to \boldsymbol{j}\,\omega_0}\{(s-\boldsymbol{j}\omega_0)Q(s)\}^*\geq 0, \qquad \forall \omega_0\in\Omega_0$$

With definition ??, what is especially important is the first and second conditions. The first condition shows the stability of a transfer function. However, this includes the cases where the real part of the pole is 0. The second condition is related to the positive definite nature of the Hermitian part when a transfer function is evaluated on an imaginary axis.

Specifically, when Q(s) is a scalar; in other words, when input and output are both scalar, two conditions show that the real part of  $Q(j\omega)$  of all  $\omega \in [0, \infty) \setminus \Omega_0$  is non-negative. But please note that when Q(s) is a matrix, it is usually:

$$Q(j\omega) + Q^{\mathsf{T}}(-j\omega) \neq 2 \operatorname{Re} [Q(j\omega)]$$

For Q(s) with a real number coefficient,  $Q^{\mathsf{T}}(-j\omega)$  is equal to  $\{Q(j\omega)\}^*$ . The third condition is exceptions where Q(s) has a pole of a pure imaginary number. For example, as in G(s) of Equation ??, it is used to analyze the transfer functions with a pole at the origin.

#### COFFEE BREAK —

Complex symmetric and complex skew-symmetric parts of a square matrix: Arbitrary square matrix M can be broken down to

$$M = \frac{M + M^*}{2} + \frac{M - M^*}{2}$$

This  $\frac{M+M^*}{2}$  is called the **Hermitian part** of M, while  $\frac{M-M^*}{2}$  is called the **skew Hermitian part** of M.

In control systems engineering, it is known that the passivity in definition  $\ref{fig:partial}$ ?? and positive realness of definition  $\ref{fig:partial}$ ?? are equivalent. In the discussion in this Section, the necessary condition for G(s) of Equation  $\ref{fig:partial}$ ?? to be positive real is that there is a positive definite matrix  $P_{G_e}$  that satisfies Equation  $\ref{fig:partial}$ ? for  $G_e$  of Equation  $\ref{fig:partial}$ ?, which is a controllable and observable realization of state space. This is equivalent to the passivity of  $G_e$  defined by the inequality of Equation  $\ref{fig:partial}$ ? The fact that  $P_{G_e}$  of Equation  $\ref{fig:partial}$ ? is positive definite is shown by Schur complements related to  $-\^{A}$  and  $-\^{A}$ :

 $W^{\mathsf{T}}\left(L + \hat{B}^{\mathsf{T}}\hat{A}^{-1}\hat{B}\right)W = W^{\mathsf{T}}L_{0}W$ 

being both positive definite. However, since  $L_0$  of Equation ?? satisfies ?? and the column vector of W in Equation ?? is orthogonal to  $\mathbb{1}$ ,  $W^TL_0W$  is nonsingular.

### 3.4.2 Necessary conditions for a transfer function of an electrical subsystem to be positive real

As a mathematical preparation to derive necessary conditions, we introduce **negative imaginariness** of a transfer function, which is a similar concept to positive realness [?,?].

**Definition 1.7** (Negative imaginariness of a transfer function) For a square transfer function Q(s) without a pole at the origin, we define  $\Omega_0$  of Equation ??. When the following three conditions are satisfied, Q(s) is called **negative imaginary**.

- The real part of all poles of Q(s) is nonpositive.
- For all  $\omega \in (0, \infty) \setminus \Omega_0$ ,  $j \{Q(j\omega) Q^{\mathsf{T}}(-j\omega)\}$  is positive semi-definite.
- When there is a pole of a pure imaginary number, their multiplicity is 1, and the following holds for the remaining numbers:

$$\lim_{s\to \boldsymbol{j}\,\omega_0}(s-\boldsymbol{j}\omega_0)\boldsymbol{j}Q(s)=\lim_{s\to \boldsymbol{j}\,\omega_0}\{(s-\boldsymbol{j}\omega_0)\boldsymbol{j}Q(s)\}^*\geq 0,\quad\forall\omega_0\in\Omega_0$$

While the positive realness of definition  $\ref{eq:condition}$  was defined by the positive semi-definite nature related to the Hermitian part of a transfer function, the negative imaginariness of definition  $\ref{eq:condition}$  is defined by the positive semi-definite nature of the skew Hermitian part of a Specifically, when Q(s) is scalar, the following is true:

$$\boldsymbol{j}\left\{\boldsymbol{Q}(\boldsymbol{j}\boldsymbol{\omega}) - \boldsymbol{Q}^{\mathsf{T}}(-\boldsymbol{j}\boldsymbol{\omega})\right\} = -2\mathrm{i}[\boldsymbol{Q}(\boldsymbol{j}\boldsymbol{\omega})]$$

thus, the second condition shows that for all  $\omega \in (0, \infty) \setminus \Omega_0$ , the imaginary part of  $Q(j\omega)$  is nonpositive. Definition ?? includes a case where Q(s) has a pole on

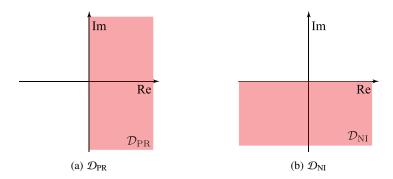


Fig. 8 Positive realms and negative imaginary realms

an imaginary axis to contrast with definition ??, but in the following discussion, we need to only focus on the second condition to consider the negative imaginariness of a stable transfer function. Similar to positive realness, the negative imaginariness of the transfer function is characterized as a possibility of matrix inequality. For more information, please refer to the supplemental title at the end of the Chapter ??.

#### COFFEE BREAK

**Nyquist trajectory**: When the trajectory related to  $\omega \in \mathbb{R}$  of the frequency response function  $Q(j\omega)$  is plotted on a complex plane, it is called the **Nyquist curve**. The Nyquist curve is often used for geometrical analysis related to the stability of a feedback system. This analytical method is called the **Nyquist stability criterion**. When Q(s) is scalar, and the coefficient of the numerator polynomial and denominator polynomial is a real number, the trajectory of  $Q(j\omega)$  against negative  $\omega$  is symmetrical to the trajectory and the real axis against positive  $\omega$ .

The relationship of positive realness and negative imaginariness is explained with Figure ??. If the transfer function Q(s) is scalar, Q(s) being positive real can be understood as the trajectory related to non-negative  $\omega$  of the frequency response function  $Q(j\omega)$  is included in the range  $\mathcal{D}_{PR}$  shown in ??(a).

$$\mathcal{D}_{PR} = i \mathcal{D}_{NI}$$

and  $-\frac{1}{i} = j$  for G(s) and H(s) of Equation ??, the following is derived:

$$G(j\omega) \in \mathcal{D}_{PR}, \quad \forall \omega > 0 \qquad \Longleftrightarrow \qquad H(j\omega) \in \mathcal{D}_{NI}, \quad \forall \omega > 0$$

Therefore, a negative imaginariness analysis of H(s) is equivalent to a positive realness analysis of G(s). To be accurate,  $G(j\omega)$  and  $H(j\omega)$  are complex matrices; thus,  $\mathcal{D}_{PR}$  and  $\mathcal{D}_{NI}$  should be redefined with a set of positive semi-definite matrices.

Therefore, a negative imaginariness analysis of H(s) is equivalent to a positive realness analysis of G(s). Based on this fact, the passive power transmission conditions (ii) and (iii) are necessary conditions for G(s) to be positive real.

[Positive realness of the transfer function of an electrical subsystem] Translated with DeepL For an arbitrary  $(\delta^*, E^*)$  where the transfer function H(s) of Equation ?? becomes stable, a condition necessary for H(s) to be negative imaginary is that the passive power transmission condition (ii) of definition ?? holds. In addition, a condition necessary for the transfer function G(s) of Equation ?? to be positive real is that the passive power transmission conditions (ii) and (iii) hold.

First, we show that for any  $(\delta^*, E^*)$  for which H(s) is stable, if H(s) is negative imaginary, then the passive transmission condition (ii), i.e., the Equation ??, holds. Now

$$\lim_{\omega \to \infty} j \left\{ H(j\omega) - H^{\mathsf{T}}(-j\omega) \right\} = j \left( -L + L^{\mathsf{T}} \right) \ge 0$$

Therefore, L must be symmetric for H(s) to be negative vacuity. Thus, we have  $K_{IJ}(\delta_{IJ}^{\star}) = K_{JI}(\delta_{JI}^{\star})$ . In other words

$$G_{ij}^{\mathrm{red}}\sin\delta_{ij}^{\star}=0, \qquad \forall (i,j)\in I_{\mathrm{G}} imes I_{\mathrm{G}}$$

This implies  $\delta_i^{\star} \neq \delta_j^{\star}$  for (i,j) where  $G_{ij}^{\rm red} = 0$ . Also, when  $\delta_i^{\star} = \delta_j^{\star}$ , continuity regarding parameter variation of eigenvalues for matrices with parameters implies that  $\tau^{-1}A$  for  $\delta^{\star} + \gamma e_i$  is There exists a sufficiently small  $\gamma > 0$  such that it is stable. However,  $e_i$  denotes a vector where only the *i*th *i*-element is 1 and the rest are 0. Therefore, there exists a vector

$$G_{ij}^{\text{red}} = 0, \qquad \forall i \neq j$$
 (58)

Furthermore, if H(s) is negative imaginary, then

$$\lim_{\omega \to 0} j \left\{ H(j\omega) - H^{\mathsf{T}}(-j\omega) \right\} = j \left( -L_0 + L_0^{\mathsf{T}} \right) \ge 0$$

Therefore,  $L_0$  in equation?? must also be symmetric. When the Equation?? holds.

$$C = \operatorname{diag}\left(2E_i^{\star}G_{ii}^{\mathrm{red}}\right) - \hat{B}^{\mathsf{T}}$$

Note that  $L_0$  is given by:

$$L_0 = \underbrace{L + \hat{B}^{\mathsf{T}} \hat{A}^{-1} \hat{B}}_{L_1} - \underbrace{\mathsf{diag}(2E_i^{\mathsf{x}} G_{ii}^{\mathsf{red}}) \hat{A}^{-1} \hat{B}}_{L_2}$$

However,  $\hat{A}$  is a symmetric matrix defined by the Equation ??. On the other hand, for  $L_2$  to be symmetric for any  $E^*$ , it must have  $G_{ii}^{\text{red}} = 0$  for all i. From this, for any  $(\delta^*, E^*)$  for which H(s) is stable, if H(s) is negative imaginary, then the expression ?? holds.

Next, we show that H(s) is negative imaginary for any  $(\delta^*, E^*)$  for which H(s) is stable if the expression?? holds. This requires that L is symmetric and:

$$\tilde{A}^{\mathsf{T}}P + P\tilde{A} \le 0, \qquad P\tilde{A}^{-1}\tilde{B} = C^{\mathsf{T}}$$
 (59)

It is enough to show that there exists a positive definite matrix *P* satisfying. However, we need to show that there exists a positive definite matrix *P* such that

$$\tilde{A} := \tau^{-1}A, \qquad \tilde{B} := \tau^{-1}B$$

,??

$$k_{ij}(\delta_{ij}^{\star}) = k_{ji}(\delta_{ji}^{\star}), \qquad h_{ij}(\delta_{ij}^{\star}) = -h_{ji}(\delta_{ji}^{\star}), \qquad h_{ii}(\delta_{ii}^{\star}) = 0$$

Since L is symmetric, it follows that L is symmetric. Also, since H(s) is stable

$$\tilde{A} = \operatorname{diag}\left(\frac{X_i - X_i'}{\tau_i}\right)\hat{A}$$

Since  $X_i > X_i'$ ,  $\hat{A}$  in Equation ?? is negative definite. Therefore, we can choose  $-\hat{A}$  as the positive definite matrix P satisfying Equation ??, which shows that H(s) is negative-definite.

Next, we show the equivalence for G(s). Since H(s) is stable, the only pole on the imaginary axis of G(s) is the origin, and its degree of overlap is 1. Therefore, G(s) is The necessary and sufficient condition for being positively real is:

$$G(j\omega) + G^{\mathsf{T}}(-j\omega) \ge 0, \qquad \forall \omega \in \mathbb{R} \setminus \{0\}$$
 (60)

is valid, and the following can be established:

$$\lim_{s \to 0} sG(s) = \lim_{s \to 0} \{sG(s)\}^{\mathsf{T}} \ge 0 \tag{61}$$

When the Equation ?? holds, then the Equation ?? holds.

$$G(\boldsymbol{j}\omega) + G^{\mathsf{T}}(-\boldsymbol{j}\omega) = \frac{\boldsymbol{j}}{\omega} \left\{ H(\boldsymbol{j}\omega) - H^{\mathsf{T}}(-\boldsymbol{j}\omega) \right\}, \qquad \forall \omega \in \mathbb{R} \setminus \{0\}$$
 (62)

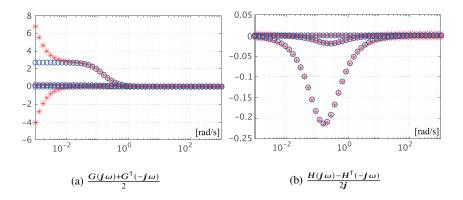
It is shown from H(s) that H(s) is negative imaginary. And,

$$\lim_{s \to 0} sG(s) = L - C\tilde{A}^{-1}\tilde{B} = L - CA^{-1}B$$

Therefore, the semi-positive definiteness of equation  $\ref{eq:condition}$  is equivalent to the passive transmission condition (iii), i.e., the condition of equation  $\ref{eq:condition}$ . Note that when the passive transmission condition (ii) holds, L is symmetric and:

$$C\tilde{A}^{-1}\tilde{B} = CP^{-1}C^{\mathsf{T}}$$

is also symmetric, which also shows the symmetry of the expression ??.



**Fig. 9** G(s)H(s) (Blue: passive transmission condition (ii) is satisfied, Red: not satisfied)

Conversely, if the passive power transmission conditions (ii) or (iii) do not hold, then G(s) is not positively real. The latter is evident from the fact that the condition in equation ?? is equal to the condition in equation ??. Also, when the passive power transmission condition (ii) does not hold, since H(s) is not negative vacuous, there exists a point  $\omega_0 \ge 0$  and a sufficiently small  $\epsilon > 0$  such that:

$$\lambda_{\min} \left[ j \left\{ H(j(\omega_0 + \alpha)) - H^{\mathsf{T}}(-j(\omega_0 + \alpha)) \right\} \right] < 0, \quad \forall \alpha \in (0, \epsilon]$$

where  $\lambda_{\min}$  denotes the smallest eigenvalue. Thus, for all  $\omega \in (\omega_0, \omega_0 + \epsilon]$  that are not 0, the complex symmetric part of  $G(j\omega)$  is not semidefinite.

Let us confirm the result of Theorem ?? with an example.

Example 1.4 (Transmission loss and positive realness of the transfer function of an electrical subsystemTransmission loss and positive realness of the transfer function of an electrical subsystem) For the electrical power system model consisting of three generators discussed in the Example ??, let us examine the positive realness of G(s) and negative imaginariness of H(s). For comparison, we will calculate a case where the passive power transmission condition (ii) is satisfied, and when it is not satisfied. Specifically, similar to the Example ??, we set  $Y_0(0)$  and  $Y_0(1)$  as the admittance matrix Y of the power grid. With the horizontal axis as frequency  $\omega$ , the eigenvalue of the Hermitian part of  $G(j\omega)$  is shown in Figure ??(a), while the imaginary part of the eigenvalue of the skew Hermitian part of  $H(j\omega)$  is shown in Figure  $P(j\omega)$  is shown in Figure  $P(j\omega)$  is shown in Figure  $P(j\omega)$  is satisfied. With this Figure, we can see that when the conductance of a transmission line is not 0, the Hermitian part of  $P(j\omega)$  is positive semi-definite in a low frequency band.

The significance of the passive power transmission condition (iii), which appeared as a condition for  $P_G$  of Equation ?? to be positive semi-definite, can be explained

as follows. In the electrical subsystem G of Equation  $\ref{eq:condition}$ , let us look at the equation of state for the internal voltage:

$$\tau \dot{E}^{\rm lin} = A E^{\rm lin} + B \delta^{\rm lin}$$

With this differential equation, let us consider a limit at which the time constant  $(\tau_i)_{i \in I_G}$  asymptotically approaches 0. This is equivalent to "a limit for which the time it takes for the internal voltage to reach a steady state is sufficiently shorter than the fluctuations in  $\delta^{\text{lin}}$ ". At this time, the following approximation holds:

$$E^{\text{lin}}(t) \simeq -A^{-1}B\delta^{\text{lin}}(t), \qquad \forall t \ge 0$$
 (63)

If A is not stable; in other words, if the passive power transmission condition (i) does not hold, state  $E^{\rm lin}$  dissipates. The method to approximate a differential equation with an algebraic equation using such a difference in the timescale of state variables is called **singular perturbation approximation** in control systems engineering. Actually, the dynamic characteristics of the internal voltage often have smaller time constants compared to the dynamic characteristics of mechanical turbines.

If we assume Equation ?? establishes an equation and substitutes as an output equation of Equation ??, the following low-dimensional approximation of the electrical system is obtained:

$$\hat{G}: \begin{cases} & \hat{\delta}^{\text{lin}} = u_G \\ & y_G = L_0 \hat{\delta}^{\text{lin}} \end{cases}$$
 (64)

To show that it is an approximation, we classified state variables as  $\hat{\delta}^{lin}$ . The entire linear approximation model of Equation ?? is approximated as a differential equation system where the second-order oscillator is combined by this singular perturbation approximation:

$$M\ddot{\delta}^{\text{lin}} + D\dot{\delta}^{\text{lin}} + \omega_0 L_0 \hat{\delta}^{\text{lin}} = 0$$
 (65)

This result shows that the passive power transmission condition (iii) shows the "positive semi-definite nature of a spring constant matrix" when the time constant is small. It can be interpreted as equivalent to a dynamic spring constant of the electrical subsystem G of Equation ??.

**Example 1.5** (Singular perturbation approximation of a linear approximation model) As a reference, Figure ?? shows the time response of a second-order oscillator system of Equation ?? to the linear approximation model discussed in the Example ??. The solid line is the response of the original linear approximation model, while the dashed line is the response of a second-order oscillator system after applying the singular perturbation approximation. Also:

$$\Delta\hat{\omega}^{\mathrm{lin}} := \omega_0^{-1} \dot{\hat{\delta}}^{\mathrm{lin}}, \qquad \hat{E}^{\mathrm{lin}} := -A^{-1}B\hat{\delta}^{\mathrm{lin}}, \qquad \hat{P}^{\mathrm{lin}} := L\hat{\delta}^{\mathrm{lin}} + C\hat{E}^{\mathrm{lin}}$$

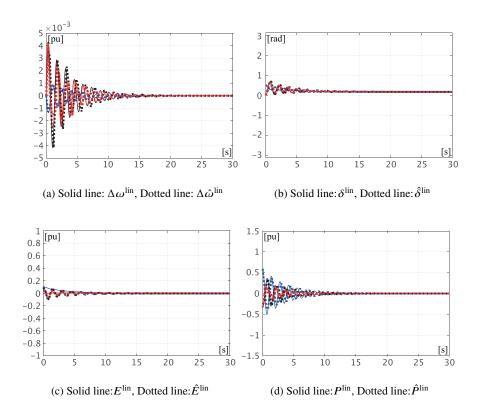


Fig. 10 Time response when low-dimensional approximation is applied (Blue: Generator 1, Black: Generator 2, Red: Generator 3)

The initial value of the linear approximation model is given as follows in response to Equation ??:

$$\hat{\delta}^{\text{lin}}(0) = \begin{bmatrix} \frac{\pi}{6} \\ 0 \\ 0 \end{bmatrix}, \qquad \Delta \hat{\omega}^{\text{lin}}(0) = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Figure ?? shows that the time response of both is largely consistent with the peak value of the oscillations and attenuation rate.

## 3.5 Necessary conditions for the linear approximation model to be statically stable<sup>‡</sup>

Below, we discuss the necessity of the passive power transmission condition (iii) from the viewpoint of small signal stability of the linear approximation model of Equation ??. Specifically, it shows that the passive power transmission condition (iii) is a necessary condition for the linear approximation model to be statically stable regardless of the physical constants of generators. As shown in the discussions of Section ??, the passive power transmission condition (i) is a necessary condition for the linear approximation model to not be unstable for the time constant  $(\tau_i)_{i \in I_G}$  at a sufficiently small limit. When the matrix A is not stable, this can be confirmed from the dissipation of the internal voltage since the singular perturbation approximation of Equation ?? cannot be applied.

When the passive power transmission condition (ii) does not hold, since  $L_0$  is usually not symmetrical, we consider generalization of the passive power transmission condition (iii) so that it can be applied to unsymmetrical  $L_0$ :

$$\mathbf{\Lambda}(L_0) \subseteq [0, \infty) \tag{66}$$

However,  $\Lambda(L_0)$  shows a set of eigenvalues of  $L_0$ . The conditions of Equation ?? show that all eigenvalues of  $L_0$  are "non-negative real numbers". Below, we call this generalized condition the passive power transmission condition (iii) ' of definition ??. When  $L_0$  is symmetrical, the passive power transmission conditions (iii) and (iii)' are equivalent. The following lemma is presented.

[Necessary condition for the small signal stability of a second-order oscillator system] Let us consider a second-order oscillator system of Equation ??. For an arbitrary initial value and arbitrary positive definite number  $(M_i, D_i)_{i \in I_G}$ , a condition necessary for a certain constant  $c_0$  to exist and the following to hold:

$$\lim_{t \to \infty} \hat{\delta}^{\text{lin}}(t) = c_0 \mathbb{1} \tag{67}$$

is that the passive power transmission condition (iii)' holds.

Translated with DeepL If the passive transmission condition (iii)' does not hold, then there exists a certain positive constant  $(M_i, D_i)_{i \in I_G}$  such that the Equation ??. For this purpose, the following two cases will be discussed.

- (a) Among the eigenvalues of  $L_0$ , there exist eigenvalues whose real part is negative or pure imaginary.
- (b) There exist eigenvalues of  $L_0$  whose real part is positive and whose imaginary part is nonzero.

First, let's consider the case (a). In the following, we choose constant matrices as  $M = \omega_0 I$  and  $D = \omega_0 dI$ . In this case, the eigen equation of the equation ?? are

$$\begin{bmatrix} 0 & I \\ -L_0 & -dI \end{bmatrix} \begin{bmatrix} v \\ w \end{bmatrix} = \lambda \begin{bmatrix} v \\ w \end{bmatrix}$$

If w is eliminated from this equation by substitution, the following is obtained:

$$\left(\lambda^2 I + d\lambda I + L_0\right) v = 0$$

This eigenvalue is an eigenvector with v of  $L_0$  and for that eigenvalue  $\kappa$ , and therefore the following is true.

$$\lambda^2 + d\lambda + \kappa = 0 \qquad \Longleftrightarrow \qquad \lambda = \frac{-d \pm \sqrt{d^2 - 4\kappa}}{2} \tag{68}$$

Therefore, in the case of (a), it is sufficient to show that the real part of  $\sqrt{d^2 - 4\kappa}$  is larger than d. In general, for any complex number z

$$Re[z] = \sqrt{Re[z^2] + (i[z])^2}$$

Since it can be expressed as  $z = \sqrt{d^2 - 4\kappa}$ , the following result can be obtained:

$$\operatorname{Re}\left[\sqrt{d^2 - 4\kappa}\right] = \sqrt{d^2 - 4\operatorname{Re}[\kappa] + (\mathrm{i}[z])^2}$$

This value is in case (a), that is, when the real part of  $\kappa$  is negative, or the real part of  $\kappa$  is 0 and the imaginary part of  $\kappa$  is non-zero. Must be greater than d. Therefore, the secondary oscillator system of the equation ?? is unstable.

Next, consider the case of (b). In the following, it is shown that there exists a positive constant d such that the eigenvalue  $\lambda$  of the expression ?? is a pure imaginary number. When  $L_0$  in the execution column has a complex eigenvalue, there is always something with a negative imaginary part. The eigenvalue is expressed as  $\kappa = \alpha + \beta \mathbf{j}$  using  $\alpha > 0$  and  $\beta < 0$ . For this  $\kappa$ , there exists  $\omega \neq 0$  and d > 0 that satisfy:

$$-d + \sqrt{d^2 - 4\kappa} = \omega \mathbf{j}$$

If you transfer -d on the left side and square both sides:

$$-4(\alpha + \beta \mathbf{j}) = 2d\omega \mathbf{j} - \omega^2$$

This equation is satisfied by choosing  $\omega = 2\sqrt{\alpha}$ ,  $d = -\frac{\beta}{\sqrt{\alpha}}$ . Therefore, since the secondary oscillator system has a steady-state vibration solution, the equation ?? does not hold.

Lemma ?? shows that, for a limit where the time constant of the internal voltage is sufficiently small, the passive power transmission condition (iii)' is a necessary condition for the linear approximation model to be statically stable against the arbitrary physical constants of generators. Furthermore, with Theorem ??, when the passive power transmission conditions (i)–(iii) hold, the linear approximation model is stable against arbitrary physical constants. Based on these facts, the conclusion of this Section is summarized in the following Theorem.

[Small signal stability of the linear approximation model] For an arbitrary positive definite number  $(M_i, D_i, \tau_i)_{i \in I_G}$ , a necessary condition for the linear approximation model of Equation ?? to be statically stable is that the passive power transmission conditions (i) and (iii)' of definition ?? hold. Specifically, when the passive power transmission condition (ii) holds, the above-mentioned necessary condition for the small signal stability is that the passive power transmission conditions (i) and (iii)' hold.

We present an analytical example of the small signal stability of the linear approximate model using Theorem ??.

Example 1.6 (Small signal stability analysis based on the passive power transmission conditions) Using Theorem ??, let us analyze the small signal stability of the linear approximation model consisting of three generators discussed in the Example ??. The physical constants of generators are set to the same value as Example ??. Since the passive power transmission condition (i) was satisfied for all parameters, we also plotted the range or parameters where the passive power transmission condition (iii) ' is not satisfied in Figure ??. When the eigenvalue of  $L_0$  of Equation ?? includes those where the real part is negative, it is shown in red. When the eigenvalue of complex numbers is included, it is shown with purple. This shows when the passive power transmission condition (ii) holds for the range on the horizontal axis where  $\theta_2$  is 0.

Theorem ?? shows that the ranges shown with red and purple are "dangerous parameter ranges where the linear approximation model is always unstable with some physical constant settings". When the passive power transmission condition (ii) holds; in other words, for parameters on the horizontal axis where  $\theta_2$  is 0, as long as  $\theta_1$  is set to a value that is not red, the linear approximation model has a small signal stability regardless of the value of these constants.

What we need to pay attention to in the result of Figure  $\ref{eq:total:eq:$ 

In the cases of (a) and (b), there is no blue range. In other words, for the searched parameters, the eigenvalue of  $L_0$  is a real number. Generally, as long as  $\theta_2$  is not 0,  $L_0$  is a non-symmetrical matrix; thus, it is unclear whether only  $L_0$  has a real eigenvalue. On the other hand, in (c) and (d) where the admittance matrix is multiplied by  $\frac{1}{100}$ , if  $\theta_1$  and  $\theta_2$  are relatively large,  $L_0$  has a complex eigenvalue. However, in a realistic setting, it has been confirmed that  $L_0$  often only has real eigenvalues.

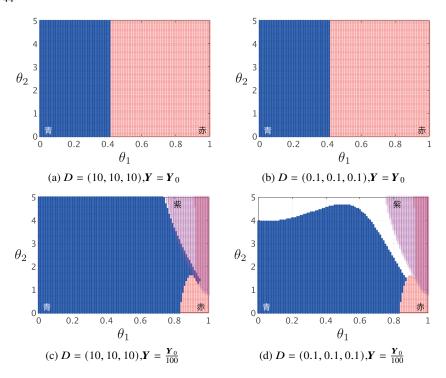


Fig. 11 Regions of parameters for which the approximate linear model is stable

### **Mathematical Supplement**

Translated with DeepL Stable and square transfer function

$$Q(s) = C(sI - A)^{-1}B + D$$

$$\begin{bmatrix} A^\mathsf{T} P + PA & PB - C^\mathsf{T} \\ B^\mathsf{T} P - C & -(D + D^\mathsf{T}) \end{bmatrix} \leq 0$$

[?, Theorem 5.31] [?, Theorem 3], [?],

$$Q(s) = C(sI - A)^{-1}B + D$$

$$A^{\mathsf{T}}P + PA \le 0, \qquad -PA^{-1}B = C^{\mathsf{T}}$$

[**?**, Lemma 7]