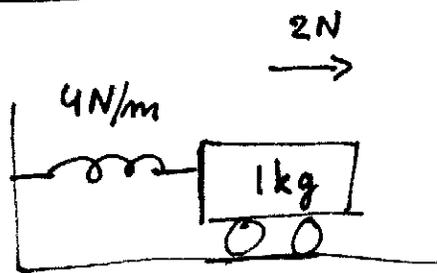


Example



position $x(t)$
velocity $\dot{x}(t)$
acceleration $\ddot{x}(t)$
friction $-k\dot{x}$

forces: $2 - 4x - k\dot{x}$

$$\ddot{x} = F/m = 2 - 4x - k\dot{x} \quad \text{or} \quad \ddot{x} + k\dot{x} + 4x = 2$$

with $x(0) = \dot{x}(0) = 0$ (take $k = 5$)

Homogeneous problem: $\ddot{x} + 5\dot{x} + 4x = 0$

substitute $x = ce^{dt} \rightarrow ce^{dt}(d^2 + 5d + 4) = 0$

$$(d+4)(d+1) = 0 \rightarrow d_1 = -4, d_2 = -1$$

general solution homogeneous system: $x = c_1 e^{-4t} + c_2 e^{-t}$

particular solution of inhomogeneous system: $x = \frac{1}{2}$

$$x = at^2 + bt + c \rightarrow \dot{x} = 2at + b \rightarrow \ddot{x} = 2a$$

$$\ddot{x} + 5\dot{x} + 4x \rightarrow 2a + 10at + 5b + 4at^2 + 4bt + 4c$$

$$= 4at^2 + (10a + 4b)t + 2a + 5b + 4c$$

$$\rightarrow \begin{cases} 4a = 0 & \rightarrow a = 0 \\ 10a + 4b = 0 & \rightarrow b = 0 \\ 2a + 5b + 4c = 2 & \rightarrow 4c = 2 \rightarrow c = \frac{1}{2} \end{cases}$$

(guessing was faster here)

solution: $x(t) = c_1 e^{-4t} + c_2 e^{-t} + \frac{1}{2}$

fit boundary conditions

$$\begin{aligned} x(0) &= c_1 + c_2 + \frac{1}{2} = 0 \\ \dot{x}(0) &= -4c_1 - c_2 = 0 \end{aligned} \quad \rightarrow \quad \begin{cases} c_1 + c_2 = -\frac{1}{2} \\ -4c_1 - c_2 = 0 \end{cases} \quad \rightarrow \quad \begin{cases} c_1 + c_2 = -\frac{1}{2} \\ -3c_1 = -\frac{1}{2} \end{cases}$$

$$c_1 = \frac{1}{6}$$

$$c_2 = -\frac{2}{3}$$

$$x(t) = \frac{1}{6}e^{-4t} - \frac{2}{3}e^{-t} + \frac{1}{2} \quad (x = \frac{1}{2} \text{ steady state})$$

Transform problem into system of first order equations:

$$\begin{cases} x_1 = x \\ x_2 = \dot{x}_1 = \dot{x}_2 \end{cases} \quad \rightarrow \quad \frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_2 \\ 2 - 4x_1 - 5x_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -4 & -5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 0 \\ 2 \end{pmatrix}$$

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -4 & -5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 0 \\ 2 \end{pmatrix}$$

2nd order ODE \rightarrow system of first order ODEs.

Always possible! \rightarrow Std software for systems of first order ODEs

(not always best approach)

$$\frac{d^{\vec{n}} \vec{y}}{dt^{\vec{n}}} = f(t, \dot{y}, \ddot{y}, y^{(3)}, \dots, y^{(n-1)}) \quad \rightarrow$$

$$\frac{d}{dt} \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} y_2 \\ y_3 \\ \vdots \\ f(t, y^{(1)}, y^{(2)}, \dots, y^{(n-1)}) \end{pmatrix}$$

$$\frac{d}{dt} x(t) = A x(t), \quad A \text{ does not depend on } t$$

Try simple solution $x(t) = g(t) v \rightarrow$

$$\dot{x} = \dot{g} v = A g v = g A v \Leftrightarrow$$

$$\dot{g}/g v = A v$$

Although lhs depends in principle on t , this cannot be the case as rhs does not depend on t . \rightarrow

$$\dot{g}/g = d \text{ constant}$$

$$\dot{g}/g = d \rightarrow \ln g(t) = dt + c \rightarrow g(t) = c_1 e^{dt}$$

Also $\dot{g}/g v = d v = A v$ (eigenvalue problem)
 $\begin{cases} d \text{ eigenvalue} \\ v \text{ eigenvector} \end{cases} \rightarrow \text{eigenpair}$

$$\text{solution } x(t) = g(t) v = c_1 e^{dt} v$$

This can be done for any eigenpair: $A v_i = d_i v_i$
 \rightarrow solution component $g_i(t) v_i = c_i e^{d_i t} v_i$

$$x(t) = \sum_i c_i e^{d_i t} v_i$$

$x(0) = x_0 = \sum_i c_i v_i \rightarrow$ decomposition of initial solution along eigenvectors.

$$\frac{d}{dt} x(t) = A x(t) = \sum_i c_i d_i e^{d_i t} v_i$$

$$\frac{d}{dt} x(t) = f(x), \text{ where } f: \mathbb{R}^m \rightarrow \mathbb{R}^m$$

(some autonomous dynamical system)

$$\hat{x}: f(\hat{x}) = 0 \rightarrow \frac{d}{dt} \hat{x} = 0 \text{ stationary point?}$$

$$\tilde{x} = \hat{x} + \varepsilon(t) \rightarrow$$

$$\frac{d}{dt} \tilde{x}(t) = \frac{d}{dt} (\hat{x} + \varepsilon(t)) = \dot{\varepsilon}(t) = f(\hat{x} + \varepsilon) =$$

$$f(\hat{x}) + J\varepsilon + O(\|\varepsilon\|^2)$$

$\underset{=0}{f(\hat{x})}$

$$J = \begin{pmatrix} \partial f_1 / \partial x_1 & \dots & \partial f_1 / \partial x_m \\ \vdots & & \vdots \\ \partial f_m / \partial x_1 & \dots & \partial f_m / \partial x_m \end{pmatrix} \rightarrow \text{jacobian of } f \text{ at } \hat{x}$$

So, perturbation from \hat{x} behaves (for small perturbations) as $\dot{\varepsilon}(t) = J\varepsilon(t)$.

Compute general solution as before. Eigenvalues of J play key role. $\varepsilon(t) = \sum_i c_i e^{d_i t} v_i$ ($Jv_i = d_i v_i$)

System is asymptotically stable if $\varepsilon \rightarrow 0$ for $t \rightarrow \infty$ (for small enough $\varepsilon(t)$)

Mode i , $c_i e^{d_i t} v_i$, decays if $\text{Re}(d_i) < 0$, constant magnitude if $\text{Re}(d_i) = 0$ (neutrally stable).

If $\text{Im}(d_i) \neq 0$ we get oscillations.

$$d_j = d_j^R + i d_j^I \rightarrow e^{(d_j^R + i d_j^I)t} = e^{d_j^R t} (\cos d_j^I t + i \sin d_j^I t)$$

Basis

$$A \in \mathbb{C}^{n \times n}$$

v eigenvector, d eigenvalue if

$$Av = dv \quad (v \neq 0)$$

(d, v) eigenpair

$$Av = dv \Leftrightarrow (A - dI)v = 0 \quad \text{for some } v \neq 0$$

So, $A - dI$ singular $\rightarrow \det(A - dI) = 0$

d eigenvalue of A if and only if $\det(A - dI)$ singular

\rightarrow characteristic equation

\rightarrow characteristic polynomial of A (of degree n)

Eigenvalues are roots of char. pol.

$\exists \hat{\lambda}$ is a multiple root of multiplicity k

$$\det(A - dI) = (d - \hat{\lambda})^k P_{n-k}(d)$$

we say $\hat{\lambda}$ has algebraic multiplicity k

If $A \in \mathbb{C}^{n \times n}$ has n distinct eigenvalues, then
A has n independent eigenvectors (proof later)
→ semisimple

If A has fewer than n independent eigenvectors, A
is called defective.

~~Defn~~ Let A have eigenvalue λ of algebraic multiplicity
 k . There can be at most k independent eigenvectors
associated with λ . Fewer possible.

Number of independent eigenvectors associated with
eigenvalue λ is called its geometric multiplicity.

$$\rightarrow \dim \text{Null}(A - \lambda I)$$

Generalization of eigenvector: invariant subspace.

S is invariant subspace of A if

$$\forall x \in S: Ax \in S$$

Each eigenvalue has an associated invariant subspace
whose dimension equals algebraic multiplicity.

Jordan decomposition of $A \in \mathbb{C}^{n \times n}$:

For every $A \in \mathbb{C}^{n \times n}$ there exists nonsingular $X \in \mathbb{C}^{n \times n}$ such that

$$X^{-1}AX = J = \begin{pmatrix} J_1 & & \\ & J_2 & \\ & & \ddots \\ & & & J_k \end{pmatrix}$$

where $J_i = \begin{pmatrix} d_i & & \\ & d_i & \\ & & \ddots \\ & & & d_i \end{pmatrix}^{m_i \times m_i}$

and $m_1 + m_2 + \dots + m_k = n$.

$$A = XJX^{-1}$$

Each block has one corresponding eigenvector \rightarrow k independent eigenvectors for A .

Each block has $m_i - 1$ principal vectors (of grade e_1, e_2, \dots, e_{m_i})

If $m_i = 1$ (for all i), matrix is diagonalizable \rightarrow full set of n eigenvectors.

$$Ax_1 = XJX^{-1}x_1 = XJe_1 = Xe_1d_1 = d_1x_1 \quad (m_i = 5)$$

$$Ax_2 = XJe_2 = X(e_2d_2 + e_1) = d_2x_2 + x_1$$

\vdots

$$Ax_5 = d_1x_5 + x_4$$

$$Ax_6 = d_2x_6$$

etc

Jordan decomposition (continued)

Multiple (Jordan) blocks can have the same eigenvalue: $d_i = d_j$

Sum of sizes of all blocks with same eigenvalue d is algebraic multiplicity of d .

The number of blocks with same eigenvalue d is the geometric multiplicity of d
(because 1 eigenvector per block)

$\begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ has eig. value 2 with multiplicity 2 and 2 eig. vectors $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ and $\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ (for example).

Note, any vector $a \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + b \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ also eig. vector

$\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ has double eigenvalue $d = 1$

Solve for eigenvectors: $\begin{pmatrix} 1-d & 1 \\ 0 & 1-d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ 0 \end{pmatrix}$

$$\rightarrow \begin{pmatrix} y \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \rightarrow y = 0$$

So, $\begin{pmatrix} x \\ 0 \end{pmatrix}$ eigenvector, but no other independent

algebraic multiplicity $\equiv 2$

geometric multiplicity 1 $\rightarrow \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ defective

$$A = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix} \rightarrow \det(A - dI) = \begin{vmatrix} 2-d & -1 \\ -1 & 2-d \end{vmatrix} =$$

$$d^2 - 4d + 4 - 1 = \cancel{(d-2)^2} d^2 - 4d + 3 = 0$$

$$(d-3)(d-1) = 0 \rightarrow d=3 \text{ or } d=1$$

$$A = \begin{pmatrix} 2 & -1 \\ 1 & 2 \end{pmatrix} \rightarrow \det(A - dI) = \begin{vmatrix} 2-d & -1 \\ 1 & 2-d \end{vmatrix} = d^2 - 4d + 4 + 1$$

$$d^2 - 4d + 5 = 0 \rightarrow d_{1,2} = 2 \pm \sqrt{16 - 20}$$

$$\rightarrow d_{1,2} = 2 \pm 2i \quad (\text{complex!})$$

$$A = \begin{pmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{pmatrix} \rightarrow d(A) = d(A_{11}) \cup d(A_{22}) \quad \left(A = \begin{pmatrix} A_{11} & 0 \\ A_{21} & A_{22} \end{pmatrix} \right)$$

\rightarrow diagonal

\rightarrow upper triangular / lower triangular

Since there is not finite formula for roots of polynomial degree ≥ 5 , there cannot be a "direct" algorithm for computing eigenvalues of matrix $\dim \geq 5$.

$$A = \begin{pmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{pmatrix} \rightarrow d(A) = d(A_{11}) \cup d(A_{22})$$

can be shown using determinant.

Alternative:

$$a) \text{ if } A_{11}v_1 = d_1 v_1 \rightarrow \begin{pmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{pmatrix} \begin{pmatrix} v_1 \\ 0 \end{pmatrix} = d_1 \begin{pmatrix} v_1 \\ 0 \end{pmatrix}$$

$$b) \text{ assume } d \in d(A_{22}) \Rightarrow d \notin d(A_{11})$$

$$\text{let } A_{22}v_2 = d v_2 \rightarrow$$

$$\begin{pmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} A_{11}v_1 + A_{12}v_2 \\ d v_2 \end{pmatrix} = d \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

$$A_{11}v_1 + A_{12}v_2 = d v_1 \Leftrightarrow (A_{11} - dI)v_1 = -A_{12}v_2$$

$$\Rightarrow v_1 = -(A_{11} - dI)^{-1} A_{12}v_2$$

\hookrightarrow possible since $d \in d(A_{11})$

vector $\begin{pmatrix} -(A_{11} - dI)^{-1} A_{12}v_2 \\ v_2 \end{pmatrix}$ is eigenvector of A
with eigenvalue d

$$A = \begin{pmatrix} -a_1 & -a_0 \\ 1 & 0 \end{pmatrix} \rightarrow \det(A - dI) = \begin{vmatrix} -a_1 - d & -a_0 \\ 1 & -d \end{vmatrix} =$$

$$\lambda^2 + a_1 d + a_0$$

$$A = \begin{pmatrix} -a_3 & -a_2 & -a_1 & -a_0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \rightarrow \det(A - dI) =$$

$$\begin{vmatrix} -a_3 - d & -a_2 & -a_1 & -a_0 \\ 1 & -d & 0 & 0 \\ 0 & d & -d & 0 \\ 0 & 0 & 1 & -d \end{vmatrix} = (-a_3 - d) \begin{vmatrix} -d & 0 & 0 \\ 1 & -d & 0 \\ 0 & 1 & -d \end{vmatrix}$$

$$= (-a_3 - d) \left[(-d)^3 - \left[(-a_2) \begin{vmatrix} -d & 0 \\ 1 & -d \end{vmatrix} - \begin{vmatrix} -a_1 & -a_0 \\ 1 & -d \end{vmatrix} \right] \right]$$

$$= d^4 + a_3 d^3 - \left[-a_2 d^2 - a_1 d - a_0 \right]$$

$$= d^4 + a_3 d^3 + a_2 d^2 + a_1 d + a_0$$

~~roots~~ of eigenvalues of A are roots of

$$p(d) = a_0 + a_1 d + a_2 d^2 + \dots + a_{m-1} d^{m-1} + d^m$$

Can be done for any polynomial after scaling
to be monic polynomial

$$\begin{aligned} \text{Let } Av_1 &= d_1 v_1 & d_1 &\neq d_2 & (v_1, v_2 \neq 0) \\ Av_2 &= d_2 v_2 \end{aligned}$$

assume $\alpha v_1 + \beta v_2 = 0$ (α, β not ~~both~~ zero)

$$A(\underbrace{\alpha v_1 + \beta v_2}_{=0}) = \alpha d_1 v_1 + \beta d_2 v_2 = 0$$

However, $v_1 = -\beta/\alpha v_2$ (or $v_2 = -\alpha/\beta v_1$)

$$\text{subst. } -\beta d_1 v_2 + \beta d_2 v_2 = \beta(d_2 - d_1)v_2 = 0$$

but $\beta \neq 0, d_2 - d_1 \neq 0, v_2 \neq 0 \rightarrow$ contradict.

v_1, v_2 independent

let $d_3 \neq d_1$ and $d_3 \neq d_2$ by same argument

v_1, v_2 indep and v_1, v_3 indep.

Assume $\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3 = 0$ (where $\alpha_i \neq 0$, see above)

$$A(\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3) = \alpha_1 d_1 v_1 + \alpha_2 d_2 v_2 + \alpha_3 d_3 v_3 = 0$$

$$\begin{pmatrix} v_1 & v_2 & v_3 \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = 0 \quad \text{and} \quad \begin{pmatrix} \alpha_1 d_1 \\ \alpha_2 d_2 \\ \alpha_3 d_3 \end{pmatrix} = 0$$

however $\begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix}, \begin{pmatrix} \alpha_1 d_1 \\ \alpha_2 d_2 \\ \alpha_3 d_3 \end{pmatrix}$ indep $\rightarrow \forall$ rank 1 contradiction

Proof of $A \in \mathbb{C}^{m \times m}$ has n distinct eigen values then it has (at least) n independent eigenvectors.

→ We already showed that each selection of 2 eigenvectors is independent (set).

→ Assume this holds for each selection of $(n-1)$ eigenvectors and show it must hold for n as well.

If set of n vectors dependent, but not any set of $n-1$,

then $\exists c \in \mathbb{C}^m : Vc = 0$ and $c_i \neq 0 \ i=1 \dots n$

$V = [v_1 \ v_2 \ \dots \ v_n]$ matrix with eig. vectors as columns

$$AV = [Av_1 \ \dots \ Av_n] = [v_1 d_1 \ \dots \ v_n d_n] = [v_1 \ \dots \ v_n] \begin{bmatrix} d_1 & & \\ & \ddots & \\ & & d_n \end{bmatrix} = V\Lambda$$

$$Vc = 0 \rightarrow AVc = V\Lambda c = V \begin{pmatrix} d_1 c_1 \\ \vdots \\ d_n c_n \end{pmatrix} = 0$$

$$A^2 Vc = V\Lambda^2 c = V \begin{pmatrix} d_1^2 c_1 \\ \vdots \\ d_n^2 c_n \end{pmatrix} = 0 \quad \text{etc } (A^{n-1} Vc)$$

$$\Rightarrow V \begin{bmatrix} c_1 & d_1 c_1 & d_1^2 c_1 & \dots & c_1 d_1^{n-1} \\ c_2 & d_2 c_2 & d_2^2 c_2 & \dots & c_2 d_2^{n-1} \\ \vdots & \vdots & \vdots & \dots & \vdots \\ c_n & d_n c_n & d_n^2 c_n & \dots & c_n d_n^{n-1} \end{bmatrix} = 0$$

$$\text{but } \begin{bmatrix} c_1 & c_1 d_1 & d_1^2 c_1 & \dots & d_1^{n-1} c_1 \\ c_2 & c_2 d_2 & \dots & \dots & d_2^{n-1} c_2 \\ \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \text{ nonsingular } \rightarrow$$

V must be zero matrix
→ contradiction.

So, we must have $c_i = 0$ for at least one coefficient, but then there is a selection of $n-1$ vectors dependent. (contradict.)

→ $\{v_1 \dots v_n\}$ independent

$A \in \mathbb{C}^{m \times m}$, n independent eigenvectors

$$\begin{cases} Av_1 = v_1 d_1 \\ Av_2 = v_2 d_2 \\ \vdots \end{cases} \rightarrow A[v_1 \ v_2 \ \dots \ v_m] = [v_1 \ v_2 \ \dots \ v_m] \begin{pmatrix} d_1 & & \\ & d_2 & \\ & & \ddots \\ & & & d_m \end{pmatrix}$$

in matrix notation $AV = V\Lambda$ (eigendecomposition of A)

since all v_i independent V nonsingular

$$A = V\Lambda V^{-1} \text{ and } V^{-1}AV = \Lambda \rightarrow \underline{A \text{ diagonalizable}}$$

$$Av_i = V\Lambda \underbrace{V^{-1}v_i}_{e_i} = [v_1 \ v_2 \ \dots \ v_m] \begin{bmatrix} d_1 & & \\ & d_2 & \\ & & \ddots \\ & & & d_m \end{bmatrix} e_i = v_i d_i$$

since $Ve_i = v_i$

(it works!)

Consider again $\frac{d}{dt} x(t) = Ax(t) = V\Lambda V^{-1}x(t)$

$$\eta(t) = V^{-1}x(t) \text{ (and hence } x(t) = V\eta(t) \rightarrow$$

$$V \frac{d}{dt} \eta(t) = V\Lambda \eta(t) \xrightarrow{V \text{ constant}} \frac{d}{dt} \eta(t) = \Lambda \eta(t)$$

\rightarrow decoupled system of ODEs : $\dot{\eta}_i(t) = d_i \eta_i(t)$

$$\text{solution } \eta_i(t) = \eta_i(t_0) \cdot e^{d_i(t-t_0)} \quad (\text{for } i = 1 \dots n)$$

$$\hookrightarrow x(t) = V\eta(t) = \sum_{i=1}^n v_i \eta_i(t_0) \cdot e^{d_i(t-t_0)}$$

$$x_0 = x(t_0) \text{ given } \rightarrow x_0 = V\eta(t_0) \rightarrow \eta(t_0) = V^{-1}x_0$$

(x_0 so initial condition defines $\eta(t_0)$)