

Linearna algebra

u MATLAB-u i R-u

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Računalna matematika
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NAPOMENA: Ova prezentacija dodatak je sadržaju pripadajućih “.m”
i “.r” datoteka.

Pregled

- 1 Uvod
- 2 Matrice i vektori
 - Kreiranje matrica i vektora
 - Indeksiranje i odabiranje elemenata matrice
- 3 Osnovne operacije matrične algebre
- 4 Rang, slika i jezgra
- 5 Sustavi linearnih jednažbi
- 6 Skalarni produkt
- 7 Vektorski produkt
- 8 Vektorske i matrične norme
- 9 Svojstvene vrijednosti
- 10 Singularne vrijednosti

Linearna algebra

UVOD

Linearna algebra je dio matematike koji se bavi vektorskim prostorima i linearnim mapiranjima (funkcijama).

Zašto linearna algebra (kao tema u ovom kolegiju)?

- Zauzima centralno mjesto među matemačkim disciplinama.

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- Zauzima centralno mjesto među matemačkim disciplinama.
- Nužno za razumjevanje i osnovni “alat” za “ostatak” matematike.
- Ključna uloga u gotovo svim znanostima koje koriste matematiku.

Linearna algebra

UVOD

Primjeri primjena

- inženjerstvo, tehničke i prirodne znanosti
 - teorija grafova, automatska regulacija, mehanika kontinuuma (...konačni elementi...), elektrotehnika, kemija, obada slika/signala, kodiranje...
- ekonomija, biologija,... općenito obrada podataka...
- optimizacija (inženjerstvo, financije, ekonomija,...)
- ... i još puno toga... gotovo svugdje gdje ima i matematike

Linearna algebra

UVOD

Linearnost

Neka su U, V linearni prostori (nad istim poljem F). Preslikavanje $f : U \rightarrow V$ je *linearno* ako ima ova svojstva

- Aditivnost $f(x + y) = f(x) + f(y), \quad \forall x, y \in U$
- Homogenost $f(\alpha x) = \alpha f(x), \quad \forall \alpha \in F \text{ i } \forall x \in U,$

$$Ax = y$$

$A \in \mathbb{R}^{n \times m}$ Linearno preslikavanje

$x \in \mathbb{R}^m$ Element (vektor, točka) linearnog prostora

$y \in \mathbb{R}^n$ Element (vektor, točka) linearnog prostora

Linearna algebra

UVOD

Mnogi nelinearni problemi (modeli) se aproksimijaju linearnim (modelima).

$$f(x_0 + \Delta x) = f(x_0) + D(x_0)\Delta x + \dots (\text{Taylorov red})$$

Neke “dobre strane” linearnosti

- superpozicija
- “lokalno je globalno” (npr. kod stabilnosti sustava)
- skalabilnost, numerika

Koliko je linearnost važna vidi se i iz toga koliko se često koristi pojam *nelinearno* (npr. nelinearni dinamički sustavi).

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Matrice i vektori

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \dots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} \in \mathbb{R}^{n \times m}$$

$$x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n \quad \text{vektor stupac}$$

$$y = (y_1 \quad y_2 \quad \dots \quad y_m) \in \mathbb{R}^{1 \times m} \quad \text{vektor redak}$$

Matrice i vektori

Kreiranje matrica i vektora u Matlabu i R-u

Description	MATLAB	R
Enter a row vector $\vec{v} =$ $\begin{bmatrix} 1 & 2 & 3 & 4 \end{bmatrix}$	<code>v=[1 2 3 4]</code>	<code>v=c(1,2,3,4)</code> or alternatively <code>v=scan()</code> then enter "1 2 3 4" and press Enter twice (the blank line terminates input)
Enter a column vector $\begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$	<code>[1; 2; 3; 4]</code>	<code>c(1,2,3,4)</code> (R does not distinguish between row and column vectors.)
Enter a matrix $\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$	<code>[1 2 3 ; 4 5 6]</code>	To enter values by row: <code>matrix(c(1,2,3,4,5,6), nrow=2,</code> <code>byrow=TRUE)</code> To enter values by column: <code>matrix(c(1,4,2,5,3,6),</code> <code>nrow=2)</code>

Napomena: za više detalja i primjere vidjeti pripadajuće ".m" i ".r" datoteke i podloge za vježbe.

Matrice i vektori

Kreiranje matrica i vektora u Matlabu i R-u

Description	MATLAB	R
Build the vector $[2\ 3\ 4\ 5\ 6\ 7]$	<code>2:7</code>	<code>2:7</code>
Build the vector $[7\ 6\ 5\ 4\ 3\ 2]$	<code>7:-1:2</code>	<code>7:2</code>
Build the vector $[2\ 5\ 8\ 11\ 14]$	<code>2:3:14</code>	<code>seq(2,14,3)</code>
Build a vector of length k containing all zeros	<code>zeros(k,1)</code> (for a column vector) or <code>zeros(1,k)</code> (for a row vector)	<code>rep(0,k)</code>
Build a vector of length k containing the value j in all positions	<code>j*ones(k,1)</code> (for a column vector) or <code>j*ones(1,k)</code> (for a row vector)	<code>rep(j,k)</code>
Build an $m \times n$ matrix of zeros	<code>zeros(m,n)</code>	<code>matrix(0,nrow=m,ncol=n)</code> or just <code>matrix(0,m,n)</code>
Build an $m \times n$ matrix containing j in all positions	<code>j*ones(m,n)</code>	<code>matrix(j,nrow=m,ncol=n)</code> or just <code>matrix(j,m,n)</code>
$n \times n$ identity matrix I_n	<code>eye(n)</code>	<code>diag(n)</code>
Build diagonal matrix A using elements of vector v as diagonal entries	<code>diag(v)</code>	<code>diag(v,nrow=length(v))</code> (Note: if you are sure the length of vector v is 2 or more, you can simply say <code>diag(v).</code>)
Extract diagonal elements of matrix A	<code>v=diag(A)</code>	<code>v=diag(A)</code>
“Glue” two matrices $a1$ and $a2$ (with the same number of rows) side-by-side	<code>[a1 a2]</code>	<code>cbind(a1,a2)</code>
“Stack” two matrices $a1$ and $a2$ (with the same number of columns) on top of each other	<code>[a1; a2]</code>	<code>rbind(a1,a2)</code>

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Matrice i vektori

Indeksiranje i odabiranje elemenata matrice

Description	MATLAB	R
Access an element of vector v	<code>v(3)</code>	<code>v[3]</code>
Access an element of matrix A	<code>A(2,3)</code>	<code>A[2,3]</code>
Access an element of matrix A using a single index: indices count down the first column, then down the second column, etc.	<code>A(5)</code>	<code>A[5]</code>
Reverse the order of elements in vector v	<code>v(end:-1:1)</code>	<code>rev(v)</code>
Column 2 of matrix A	<code>A(:,2)</code>	<code>A[,2]</code> Note: that gives the result as a vector. To make the result a $m \times 1$ matrix instead, do <code>A[,2,drop=FALSE]</code>
Row 7 of matrix A	<code>A(7,:)</code>	<code>A[7,]</code> Note: that gives the result as a vector. To make the result a $1 \times n$ matrix instead, do <code>A[7,,drop=FALSE]</code>
All elements of A as a vector, column-by-column	<code>A(:)</code> (gives a column vector)	<code>c(A)</code>
Rows 2–4, columns 6–10 of A (this is a 3×5 matrix)	<code>A(2:4,6:10)</code>	<code>A[2:4,6:10]</code>
A 3×2 matrix consisting of rows 7, 7, and 6 and columns 2 and 1 of A (in that order)	<code>A([7 7 6], [2 1])</code>	<code>A[c(7,7,6),c(2,1)]</code>
Circularly shift the rows of matrix A down by s_1 elements, and right by s_2 elements	<code>circshift(A, [s1 s2])</code>	No simple way, but modulo arithmetic on indices will work: <code>m=dim(A)[1]; n=dim(A)[2]; A[(1:m-s1-1)%m+1, (1:n-s2-1)%n+1]</code>

Osnovne operacije matrične algebre

Description	MATLAB	R
Vector dot product $\vec{x} \cdot \vec{y} = \vec{x}^T \vec{y}$	<code>dot(x,y)</code>	<code>sum(x*y)</code>
Vector cross product $\vec{x} \times \vec{y}$	<code>cross(x,y)</code>	Not in base R, but e.g. the <code>xprod</code> function from the RSEIS package will do it (see item 331 for how to install/load packages)
Matrix multiplication AB	<code>A * B</code>	<code>A %% B</code>
Element-by-element multiplication of A and B	<code>A .* B</code>	<code>A * B</code>
Transpose of a matrix, A^T	<code>A'</code> (This is actually the complex conjugate (i.e. Hermitian) transpose; use <code>A.'</code> for the non-conjugate transpose if you like; they are equivalent for real matrices.)	<code>t(A)</code> for transpose, or <code>Conj(t(A))</code> for conjugate (Hermitian) transpose
Solve $A\vec{x} = \vec{b}$	<code>A\b</code> Warning: if there is no solution, MATLAB gives you a least-squares “best fit.” If there are many solutions, MATLAB just gives you one of them.	<code>solve(A,b)</code> Warning: this only works with square invertible matrices.
Reduced echelon form of A	<code>rref(A)</code>	R does not have a function to do this
Determinant of A	<code>det(A)</code>	<code>det(A)</code>
Inverse of A	<code>inv(A)</code>	<code>solve(A)</code>
Trace of A	<code>trace(A)</code>	<code>sum(diag(A))</code>
Compute AB^{-1}	<code>A/B</code>	<code>A %% solve(B)</code>
Element-by-element division of A and B	<code>A ./ B</code>	<code>A / B</code>
Compute $A^{-1}B$	<code>A\B</code>	<code>solve(A,B)</code>

Sustavi linearnih jednačbi

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1m}x_m = b_1,$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2m}x_m = b_2,$$

$$\vdots \qquad \qquad \qquad \vdots \qquad \qquad \vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nm}x_m = b_n$$

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \dots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}$$

$$Ax = b$$

Slika, jezgra i rang

$$A \in \mathbb{R}^{m \times n}$$

Slika (*eng. image (space), range*)

$$\text{Im}(A) = \text{Range}(A) = \mathcal{R}(A) = \{ Ax \mid x \in \mathbb{R}^n \} \subseteq \mathbb{R}^m$$

Jezgra (*eng. kernel (space), null space*)

$$\text{Ker}(A) = \text{Null}(A) = \mathcal{N}(A) = \{ x \in \mathbb{R}^n \mid Ax = 0 \}$$

Rang matrice

$$\text{Rang}(A) = \dim(A)$$

Slika

$$A \in \mathbb{R}^{m \times n}$$

Slika (*eng. image (space), range*)

$$\text{Im}(A) = \text{Range}(A) = \mathcal{R}(A) = \{ Ax \mid x \in \mathbb{R}^n \} \subseteq \mathbb{R}^m$$

$\mathcal{R}(A)$ se može interpretirati kao:

- prostor koji rasprostiru stupci matrice A
- skup vektora y za koje jednadžba $Ax = y$ ima rješenje

MATLAB: Traženje ortonormalne baze prostora slike matrice A :
`orth(A)`

Jezgra

$$A \in \mathbb{R}^{m \times n}$$

Jezgra (eng. *kernel (space), null space*)

$$\text{Ker}(A) = \text{Null}(A) = \mathcal{N}(A) = \{x \in \mathbb{R}^n \mid Ax = 0\}$$

- $\mathcal{N}(A)$ je skup vektora u \mathbb{R}^n okomitih na redke matrice A
- Ako je $y = Ax$ i $z \in \mathcal{N}(A)$, tada je $y = A(x + z)$
- Ako je $y = Ax$ i $y = A\tilde{x}$, tada je $\tilde{x} = x + z$ za neki $z \in \mathcal{N}(A)$

MATLAB: Traženje ortonormalne baze prostora jezgre matrice A :
`null(A)`

Rang matrice

$$A \in \mathbb{R}^{m \times n}$$

Rang matrice

$$\text{Rang}(A) = \dim(A)$$

- $\text{Rang}(A) + \dim(\mathcal{N}(A)) = n$
- $\text{Rang}(A)$ je maksimalan broj linearno nezavisnih stupaca (= broj nezavisnih redaka) u A
- $\text{Rang}(A) = \text{Rang}(A^T)$

MATLAB: `rank(A)`, `rank(A,tol)` (Oprez: numerika)

R: `qr(A)$rank`

Sustavi linearnih jednačbi

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$$\vdots \qquad \qquad \qquad \vdots \qquad \qquad \vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nm}x_m = b_n$$

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \dots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}$$

$$Ax = b$$

Sustavi linearnih jednažbi

Description	MATLAB	R
Solve $A\vec{x} = \vec{b}$	<code>A\b</code> Warning: if there is no solution, MATLAB gives you a least-squares “best fit.” If there are many solutions, MATLAB just gives you one of them.	<code>solve(A,b)</code> Warning: this only works with square invertible matrices.

MATLAB: $A \setminus B$ is the matrix division of A into B , which is roughly the same as $\text{INV}(A)*B$, except it is computed in a different way. If A is an N -by- N matrix and B is a column vector with N components, or a matrix with several such columns, then $X = A \setminus B$ is the solution to the equation $A*X = B$. A warning message is printed if A is badly scaled or nearly singular. $A \setminus \text{EYE}(\text{SIZE}(A))$ produces the inverse of A .

Sustavi linearnih jednačbi

MATLAB: If A is an M -by- N matrix with $M < \text{or} > N$ and B is a column vector with M components, or a matrix with several such columns, then $X = A \backslash B$ is the solution in the least squares sense to the under- or overdetermined system of equations $A * X = B$. The effective rank, K , of A is determined from the QR decomposition with pivoting. A solution X is computed which has at most K nonzero components per column. If $K < N$ this will usually not be the same solution as $\text{PINV}(A) * B$. $A \backslash \text{EYE}(\text{SIZE}(A))$ produces a generalized inverse of A .

Skalani produkt vektora

$$x, y \in \mathbb{R}^n$$

Skalani produkt (*Inner product*)

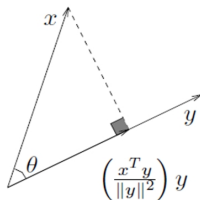
$$\langle x, y \rangle = x_1 y_1 + x_2 y_2 + \dots + x_n y_n = x^\top y$$

Description	MATLAB	R
Vector dot product $\vec{x} \cdot \vec{y} = \vec{x}^\top \vec{y}$	<code>dot(x,y)</code>	<code>sum(x*y)</code>

Cauchy-Schwartz nejednakost i kut između vektora

Za svaki $x, y \in \mathbb{R}^n$ vrijedi

$$|x^\top y| \leq \|x\| \|y\| \quad (\text{Cauchy-Schwartz})$$



Kut između vektora u \mathbb{R}^n :

$$\theta = \angle(x, y) = \cos^{-1} \frac{x^\top y}{\|x\| \|y\|}$$

$$x^\top y = \|x\| \|y\| \cos \theta$$

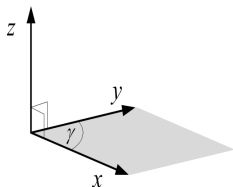
Vektorski produkt

$$x = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix}^T \in \mathbb{R}^3, \quad y = \begin{bmatrix} y_1 & y_2 & y_3 \end{bmatrix}^T \in \mathbb{R}^3$$

Vektorski produkt

$$z = x \times y = \begin{bmatrix} x_2 y_3 - x_3 y_2 \\ x_3 y_1 - x_1 y_3 \\ x_1 y_2 - x_2 y_1 \end{bmatrix}$$

Description	MATLAB	R
Vector cross product $\vec{x} \times \vec{y}$	<code>cross(x,y)</code>	Not in base R, but e.g. the <code>xprod</code> function from the RSEIS package will do it (see item 331 for how to install/load packages)



$$\|z\| = \|x\| \|y\| \sin \gamma$$

Vektorske norme

Norma je sklarna vrijednost koja služi kao mjera veličine vektora ili matrice.

Tzv. p -norma vektora $x \in \mathbb{R}^n(\mathbb{C}^n)$ definarana je

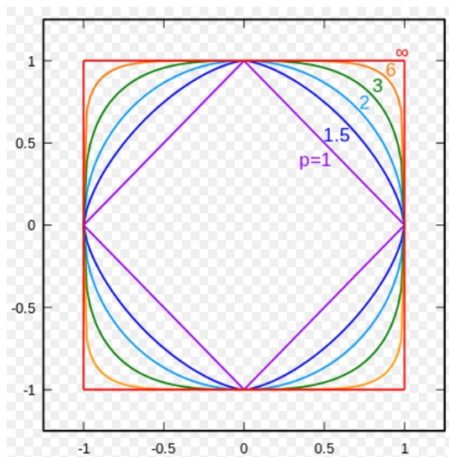
$$\|x\|_p := \left(\sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}}, \quad 1 \leq p < \infty,$$

$$\|x\|_\infty := \max_{1 \leq i \leq n} (|x_i|), \quad p = \infty,$$

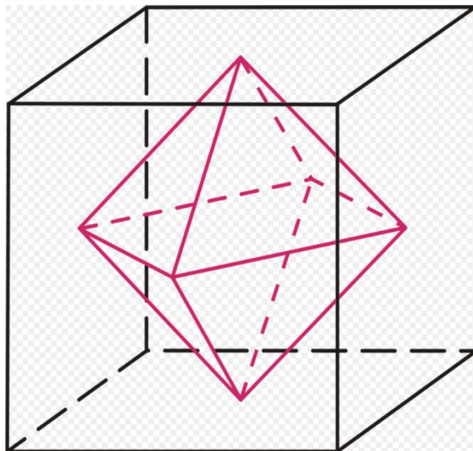
Euklidska norma (2-norma):

$$\|x\|_2 = \sqrt{x^* x} = \sqrt{x_1^2 + x_2^2 + \dots x_n^2}$$

Vektorske norme



Vektorske norme



Vektorske i matrične norme

Inducirane matrične norme

$$\|A\|_p := \max_{x \neq 0} \frac{\|Ax\|_p}{\|x\|_p}$$

Za $A \in \mathbb{R}^{m \times n}$

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}|$$

$$\|A\|_2 = \sqrt{\lambda_{\max}(A^*A)} = \sigma_{\max}(A)$$

$$\|A\|_{\infty} = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$$

Vektorske i matrice norme

Description	MATLAB	R
Vector norms	<code>norm(v,1)</code> for 1-norm $\ \vec{v}\ _1$, <code>norm(v,2)</code> for Euclidean norm $\ \vec{v}\ _2$, <code>norm(v,inf)</code> for infinity-norm $\ \vec{v}\ _\infty$, and <code>norm(v,p)</code> for p -norm $\ \vec{v}\ _p = (\sum v_i ^p)^{1/p}$	R does not have a <code>norm</code> function for vectors; only one for matrices. But the following will work: <code>norm(matrix(v),'1')</code> for 1-norm $\ \vec{v}\ _1$, <code>norm(matrix(v),'i')</code> for infinity-norm $\ \vec{v}\ _\infty$, and <code>sum(abs(v)^p)^(1/p)</code> for p -norm $\ \vec{v}\ _p = (\sum v_i ^p)^{1/p}$
Matrix norms	<code>norm(A,1)</code> for 1-norm $\ A\ _1$, <code>norm(A)</code> for 2-norm $\ A\ _2$, <code>norm(A,inf)</code> for infinity-norm $\ A\ _\infty$, and <code>norm(A,'fro')</code> for Frobenius norm $(\sum_i (A^T A)_{ii})^{1/2}$	<code>norm(A,'1')</code> for 1-norm $\ A\ _1$, <code>max(svd(A)\$d)</code> for 2-norm $\ A\ _2$, <code>norm(A,'i')</code> for infinity-norm $\ A\ _\infty$, and <code>norm(A,'f')</code> for Frobenius norm $(\sum_i (A^T A)_{ii})^{1/2}$

Svojstvene vrijednosti

Neka je $A \in \mathbb{C}^{n \times n}$. Skalar $\lambda \in \mathbb{C}$ zove se *svojstvena vrijednost* matrice A , ako postoji vektor $x \in \mathbb{C}^n$, $x \neq 0$, takav da je

$$Ax = \lambda x.$$

Takav vektor x zove se *svojstveni vektor* od A , koji pripada svojstvenoj vrijednosti λ .

Description	MATLAB	R
Set w to be a vector of eigenvalues of A , and V a matrix containing the corresponding eigenvectors	<code>[V,D]=eig(A)</code> and then <code>w=diag(D)</code> since MATLAB returns the eigenvalues on the diagonal of D	<code>tmp=eigen(A); w=tmp\$values;</code> <code>V=tmp\$vectors</code>

Svojstvene vrijednosti

Primjer: ponašanje dinamičkih sustava
Vibracije Tacoma Narrows mosta



Frekvencija uzbudnih sila vjetra bila je blizu rezonantnoj frekvenciji mosta (svojstvena vrijednost!)

Vidjeti impresivan video na Youtubeu (Tacoma Narrows Bridge)

Svojstvene vrijednosti

Primjer: ponašanje dinamičkih sustava
Vibracije Tacoma Narrows mosta



Svojstvene vrijednosti

Primjer: ponašanje dinamičkih sustava

Skalarna linearna diferencijalna jednadžba

$$\frac{dx}{dt} = \lambda x, \quad x(0) = x_0$$

Rješenje?

Svojstvene vrijednosti

Primjer: ponašanje dinamičkih sustava

Skalarna linearna diferencijalna jednadžba

$$\frac{dx}{dt} = \lambda x, \quad x(0) = x_0$$

Rješenje?

$$x(t) = e^{\lambda t} x_0$$

Ravnotežna točka $x_e = 0$ je

- **stabilna** ako je $\lambda \leq 0$
- **asimptotski stabilna** ako je $\lambda < 0$

Primjer: ponašanje dinamičkih sustava

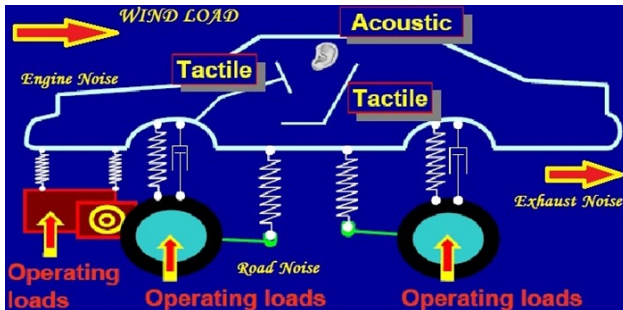
Sustav linearnih jednažbi, za $x(t) \in \mathbb{R}^n$, $A \in \mathbb{R}^{n \times n}$:

$$\frac{dx}{dt} = Ax, \quad x(0) = x_0$$

Ravnotežna točka $x_e = 0$ je **asimptotski stabilna** ako je $\operatorname{Re}(\lambda_i) < 0$ za svaki $i = 1, \dots, n$; gdje λ_i označava i -tu svojstvenu vrijednost matrice A , a Re označava realni dio kompleksnog broja.

Svojstvene vrijednosti

Primjer: konstrukcija automobila. Svojstvene vrijednosti = vibracije, udobnost vožnje



Svojstvene vrijednosti

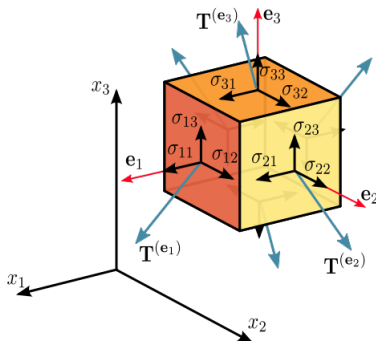
Primjer: provjera konstrukcije na pukotine i deformacije.



Ne treba provjeravati svaki centimetar konstrukcije (neučinkovito); već konstrukciju u cjelini kroz svojstvene vrijednosti (učinkovito).

Svojstvene vrijednosti = vibracije, zvuk. Pukotine \implies promjena svj. vrijednosti

Svojstvene vrijednosti



$$\sigma = \begin{bmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_{22} & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_{33} \end{bmatrix}$$

Svojstvene vrijednosti tenzora naprezanja σ su **glavna naprezanja**.
Pripadajući svojstveni vektori definiraju smjerove glavnih naprezanja.

Singularne vrijednosti

Za svaku matricu $A \in \mathbb{C}^{m \times n}$ postoji *dekompozicija singularnih vrijednosti*

$$A = UDV^*$$

gdje je

$U \in \mathbb{C}^{m \times m}$ unitarna matrica, $U^*U = UU^* = I$

$V \in \mathbb{C}^{n \times n}$ unitarna matrica, $V^*V = VV^* = I$

$D \in \mathbb{C}^{m \times n}$ sa dijagonalnim elementima $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_p \geq 0$,
 $p = \min(m, n)$

Singularne vrijednosti

Za svaku matricu $A \in \mathbb{C}^{m \times n}$ postoji *dekompozicija singularnih vrijednosti*

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gdje je

$D \in \mathbb{C}^{m \times n}$ sa dijagonalnim elementima $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_p \geq 0$

Za $m \neq n$:

$$\Sigma = \begin{bmatrix} \sigma_1 & 0 & \dots & 0 \\ 0 & \sigma_2 & & 0 \\ & & \ddots & \\ 0 & & & \sigma_n \\ 0 & \dots & & 0 \\ \vdots & & & \vdots \\ 0 & \dots & & 0 \end{bmatrix} \quad \text{ili} \quad \Sigma = \begin{bmatrix} \sigma_1 & 0 & \dots & 0 & 0 & \dots & 0 \\ 0 & \sigma_2 & & & & \vdots & \vdots \\ & & \ddots & & & & \\ 0 & \dots & & \sigma_m & 0 & \dots & 0 \end{bmatrix}$$

Singularne vrijednosti

Za svaku matricu $A \in \mathbb{C}^{m \times n}$ postoji *dekompozicija singularnih vrijednosti*

$$A = UDV^*$$

gdje su $U \in \mathbb{C}^{m \times m}$, $V \in \mathbb{C}^{n \times n}$ unitarne matrice i $\Sigma \in \mathbb{C}^{m \times n}$ je dijagonalna matrica.

- brojevi $\sigma_1, \sigma_2, \dots, \sigma_p$ su singularne vrijednosti od A
- $U = (u_1, \dots, u_m)$; vektori $u_i \in \mathbb{C}^m$ su lijevi singularni vektori od A
- $V = (v_1, \dots, v_n)$; vektori $v_i \in \mathbb{C}^n$ su desni singularni vektori od A
- broj singularnih vrijednosti p različitih od nule jednak je rangu matrice A

Singularne vrijednosti

$$A = \underbrace{\begin{bmatrix} u_1 & | & u_2 & | & \dots & | & u_m \end{bmatrix}}_{U \in \mathbb{C}^{m \times m}} \underbrace{\begin{bmatrix} \sigma_1 & 0 & \dots & 0 \\ 0 & \sigma_2 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & & 0 & \sigma_n \\ 0 & & 0 & 0 \\ \vdots & & & \vdots \\ 0 & \dots & \dots & 0 \end{bmatrix}}_{\Sigma \in \mathbb{R}^{m \times n}} \underbrace{\begin{bmatrix} v_1^* \\ v_2^* \\ \vdots \\ v_n^* \end{bmatrix}}_{V^* \in \mathbb{C}^{n \times n}}$$

$$Av_1 = U \Sigma V^* v_1 = U \Sigma \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = U \begin{bmatrix} \sigma_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \sigma_1 u_1$$

$$Av_2 = U \Sigma V^* v_2 = U \Sigma \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} = U \begin{bmatrix} 0 \\ \sigma_2 \\ \vdots \\ 0 \end{bmatrix} = \sigma_2 u_2$$

Singularne vrijednosti

$$A = \underbrace{\begin{bmatrix} u_1 & u_2 & \dots & u_m \end{bmatrix}}_{U \in \mathbb{C}^{m \times m}} \underbrace{\begin{bmatrix} \sigma_1 & 0 & \dots & 0 \\ 0 & \sigma_2 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & & 0 & \sigma_n \\ 0 & & 0 & 0 \\ \vdots & & \vdots & \\ 0 & \dots & \dots & 0 \end{bmatrix}}_{\Sigma \in \mathbb{R}^{m \times n}} \underbrace{\begin{bmatrix} v_1^* \\ v_2^* \\ \vdots \\ v_n^* \end{bmatrix}}_{V^* \in \mathbb{C}^{n \times n}}$$

$$Av_1 = U \Sigma V^* v_1 = U \Sigma \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = U \begin{bmatrix} \sigma_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \sigma_1 u_1$$

$$\|A\| = \max_{\|x\|=1} \frac{\|Ax\|}{\|x\|} = \sigma_{\max}(A) = \sigma_1(A)$$

Najveća singularna vrijednost matrice je ujedno i norma te matrice

Singularne vrijednosti

Description	MATLAB	R
Singular-value decomposition: given $m \times n$ matrix A with rank r , find $m \times r$ matrix P with orthonormal columns, diagonal $r \times r$ matrix S , and $r \times n$ matrix Q^T with orthonormal rows so that $PSQ^T = A$	<code>[P,S,Q]=svd(A,'econ')</code>	<code>tmp=svd(A); U=tmp\$u; V=tmp\$v; S=diag(tmp\$d)</code>