

Computation of SVD.

Recall for any matrix
- $A = U \Sigma V^T$

$$\Sigma = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_r, \underbrace{0, \dots, 0}_{n-r})$$

$$\sigma_1 \geq \sigma_2 \geq \sigma_3 \geq \dots \geq \sigma_r > \sigma_{r+1} = \sigma_{r+2} = \dots = \sigma_n = 0$$

How to compute SVD.

Simple way is through eigenvalue computation of symmetric matrices

$$A^T A, \quad A A^T$$

$$A^T A = V \Sigma U^T U \Sigma V^T = V \Sigma^2 V^T$$

$\sqrt{\lambda(A^T A)}$ → singular values.

$$A A^T = U \Sigma^2 U^T$$

Eigenvectors of $A^T A$ give V .

and of $A A^T$ gives U .

But what if $\sigma_r = 0.001$
in squaring i.e. $A^T A$

$$\lambda_r = (0.001)^2 = 10^{-6} \approx 0$$

Loss of information on squaring

Direct computation of SVD.

$A, B \in \mathbb{R}^{n \times m}$ are orthogonally equivalent

if $\exists P \in \mathbb{R}^{n \times n}$ $Q \in \mathbb{R}^{m \times m}$ s.t.

$$PAQ = B$$

By orthogonal equivalence transformation

it is possible to reduce any matrix to

bidagonal form.

B in bidagonal form if $b_{ij} = 0$ $\forall j > i$ or $i < j-1$

$$\begin{bmatrix} * & * & 0 & \dots & 0 & 0 \\ 0 & * & * & \dots & 0 & 0 \\ 0 & 0 & * & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & & * & * \\ 0 & 0 & 0 & & 0 & * \\ 0 & 0 & 0 & & 0 & 0 \\ 0 & 0 & 0 & & 0 & 0 \end{bmatrix}$$

$$\begin{aligned} i > j \text{ or} \\ i < j-1 \end{aligned}$$

only diagonal & superdiagonal are non-zero.

$$i-j > 0$$

Let $A \in \mathbb{R}^{n \times m}$ with $n \geq m$.

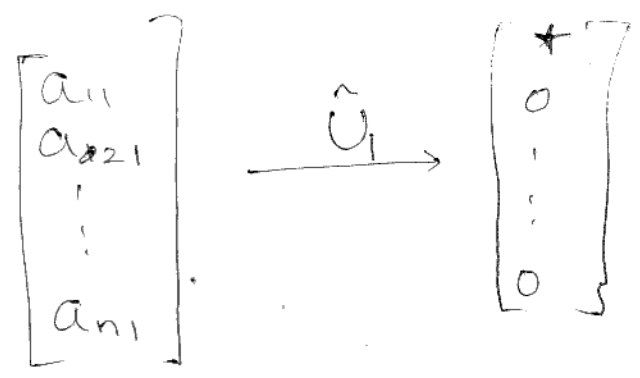
$\exists \hat{U} \in \mathbb{R}^{n \times n}$, $\hat{V} \in \mathbb{R}^{m \times m}$ orthogonal

both are products of finite no. of reflectors ~~and~~ s.t.

$$A = \hat{U} B \hat{V}^T \text{ where } B \text{ is in bidiagonal form.}$$

Givens - Kahan algorithm

Step 1: Create zeros in first column & rows of A.



$\hat{U}_1 A$ first column has zeros except at (1,1) place.

Let New ^{1st} row of $\hat{U}_1 A$ $[\hat{a}_{11} \quad \hat{a}_{12} \quad \dots \quad \hat{a}_{1m}]$
(i.e. 1st row of $\hat{U}_1 A$)

Now we can only operate on $[\hat{a}_{12} \quad \dots \quad \hat{a}_{1m}]$ without disturbing zeros in first column.

$$\hat{V}_1 = \left[\begin{array}{c|ccc} 1 & 0 & \dots & 0 \\ \hline 0 & & & \\ \vdots & & \checkmark & \\ 0 & & \checkmark & \end{array} \right]$$

where $\checkmark V_1$ is s.t.

$$[\hat{a}_{12} \dots \hat{a}_{1m}] \checkmark V_1 = [* 0 \dots 0]$$

Then.

$$\hat{U}_1 A \hat{V}_1 = \left[\begin{array}{c|ccc} * & * & 0 & \dots & 0 \\ \hline 0 & & & & \\ \vdots & & \hat{A} & & \\ \vdots & & & & \\ 0 & & & & \end{array} \right]$$

Second step.

$$\hat{U}_2 \hat{U}_1 A \hat{V}_1 \checkmark V_2 = \left[\begin{array}{cc|ccc} * & * & 0 & 0 & \dots & 0 \\ \hline 0 & * & * & 0 & \dots & 0 \\ \hline 0 & 0 & & & & \\ \vdots & \vdots & & \tilde{A} & & \\ \vdots & \vdots & & & & \\ 0 & 0 & & & & \end{array} \right]$$

and so on. \hat{U}

(5)

$$\hat{U}_m \dots \hat{U}_2 \hat{U}_1 A \hat{V}_1 \hat{V}_2 \dots \hat{V}_{m-2}$$

$$= \begin{bmatrix} * & * & & & \\ & * & * & & \\ & & \ddots & & \\ & & & * & \\ & & & & * \end{bmatrix} = B$$

$$\hat{U} A \hat{V} = B$$

Now we can start directly with bidiagonal form.

Let $B = \begin{bmatrix} \tilde{B} \\ 0 \end{bmatrix} \rightarrow \mathbb{R}^{m \times m}$.
 \rightarrow doesn't matter now.

Let $B = \begin{bmatrix} \beta_1 & \gamma_1 & & & \\ & \beta_2 & \gamma_2 & & \\ & & \ddots & \ddots & \\ & & & \beta_{m-1} & \gamma_{m-1} \\ & & & & \beta_m \end{bmatrix}$

B is properly bidiagonal if $\beta_i \neq 0$ & $\gamma_i \neq 0$

if not then $B = \begin{bmatrix} B_1 & 0 \\ 0 & B_2 \end{bmatrix}$ where B_1 & B_2 are properly bidiagonal

SVD problem of BB^T and B^TB is eig. val. problem for.

BB^T and B^TB are tridiagonal.

$$\text{Let } C = \begin{bmatrix} 0 & B^T \\ B & 0 \end{bmatrix} \quad C^2 = \begin{bmatrix} B^TB & 0 \\ 0 & BB^T \end{bmatrix}$$

C is non-singular.

$$\begin{bmatrix} B & 0 \\ B^T & 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

all eig. values of C are

non-zero, and occur in $\pm\sigma$ pairs.

~~u, v~~ $\begin{bmatrix} u \\ v \end{bmatrix}$ is an eigenvector of C with σ as eigenvalue. then $\begin{bmatrix} u \\ -v \end{bmatrix}$ is vect with $-\sigma$ as eigenvalue.

if σ is singular value of B with u, v as singular vectors

$$\begin{aligned} Bu &= \sigma v \\ B^T v &= \sigma u \\ -Bu &= -\sigma v \\ Bv &= \sigma u \end{aligned}$$

then $Bu = \sigma v$
 $B^T v = \sigma u$

and then $\begin{bmatrix} u \\ v \end{bmatrix}$ is eigenvector of C with σ as eigenvalue, also $\begin{bmatrix} u \\ -v \end{bmatrix}$ is an eigenvector with $-\sigma$ as eigenvalue.

$$C \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} B^T v \\ Bu \end{bmatrix} = \sigma \begin{bmatrix} u \\ v \end{bmatrix}$$

$$C \begin{bmatrix} u \\ -v \end{bmatrix} = \begin{bmatrix} -B^T v \\ Bu \end{bmatrix} = -\sigma \begin{bmatrix} u \\ -v \end{bmatrix}$$

Thus eigenvalue of C always occur in pair $\pm \sigma$

FRANCIS ^{like} Algorithm to obtain SVD

$$C = \begin{bmatrix} 0 & B^T \\ B & 0 \end{bmatrix} \quad (\text{GOLUB REINSCH ALGORITHM})$$

$$\tilde{C} = P^T C P = \begin{bmatrix} 0 & \beta_1 & 0 & 0 & & & & & \\ \beta_1 & 0 & \gamma_1 & 0 & & & & & \\ \hline 0 & \gamma_1 & 0 & \beta_2 & 0 & 0 & & & \\ 0 & 0 & \beta_2 & 0 & \gamma_2 & 0 & & & \\ \hline & & 0 & \gamma_2 & 0 & \beta_3 & 0 & 0 & \\ & & 0 & 0 & \beta_3 & 0 & \gamma_3 & 0 & \\ \hline & & & & 0 & \gamma_3 & & & \\ & & & & 0 & 0 & & & \end{bmatrix}$$

Eigenvalues in pairs $\pm \sigma$.

We work with

$$(\tilde{C} - \sigma I)(\tilde{C} + \sigma I)$$

$$= \tilde{C}^2 - \sigma^2 I$$

$$\tilde{C} - sI =$$

$$\begin{bmatrix} -s & \beta_1 & 0 & 0 & & \\ \beta_1 & -s & \gamma_1 & 0 & & \\ 0 & \gamma_1 & -s & \beta_2 & 0 & 0 \\ 0 & 0 & \beta_2 & -s & \gamma_2 & 0 \\ & & 0 & \gamma_2 & -s & \beta_3 \\ & & 0 & 0 & \beta_3 & -s & \gamma_3 & 0 \\ & & & & 0 & \gamma_3 & -s & \beta_4 \\ & & & & & 0 & 0 & \ddots \end{bmatrix}$$

$$\tilde{C} + sI =$$

$$\begin{bmatrix} s & \beta_1 & 0 & 0 & & \\ \beta_1 & s & \gamma_1 & 0 & & \\ 0 & \gamma_1 & s & \beta_2 & 0 & 0 \\ 0 & 0 & \beta_2 & s & \gamma_2 & 0 \\ & & 0 & \gamma_2 & s & \beta_3 \\ & & 0 & 0 & \beta_3 & s \end{bmatrix}$$

$$(\tilde{C} - sI)(\tilde{C} + sI) \text{ first col.}$$

$$= \begin{bmatrix} \beta_1^2 - s^2 \\ 0 \\ \gamma_1 \beta_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

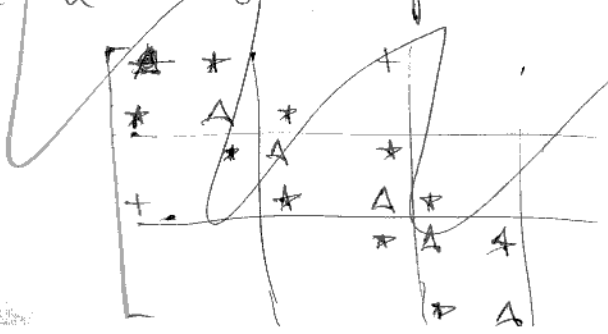
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$$V_0^T \begin{bmatrix} \beta_1^2 - s^2 \\ \gamma_1 \beta_1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} * \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$B \rightarrow B V_0$

(1, 3) ~~column~~ and zeros.

Create a zero in place of $\gamma_1 \beta_1$ using rotator.



① find a rotator that creates zero in place of γ_1, β_1 in \tilde{Q}_0^T . Make it act from right on \tilde{C} .

$$\tilde{Q}_0^T \tilde{C} = \begin{bmatrix} 0 & \beta_1 & 0 & 0 \\ \beta_1 & 0 & \gamma_1 & 0 \\ 0 & \gamma_1 & 0 & \beta_2 \\ 0 & 0 & \beta_2 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \vdots & \vdots \end{bmatrix}$$

Changes only 1, 3 rows.

Combines 1, 3 rows, thus zero structure in 4th column is destroyed as follows.

$$\begin{bmatrix} 0 & \tilde{\beta}_1 & 0 & \boxed{*} \\ \beta_1 & 0 & \gamma_1 & 0 \\ 0 & \tilde{\gamma}_1 & 0 & \tilde{\beta}_2 \\ 0 & 0 & \beta_2 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \vdots & \vdots \end{bmatrix}$$

non-zero entry created in (1, 4) column

Completing Similarity transformation.

$$\begin{bmatrix} 0 & * & 0 & + \\ * & 0 & * & 0 \\ 0 & * & 0 & * \\ + & 0 & * & 0 \\ 0 & 0 & 0 & * \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Bulge of special form is created.

$(3,1), (4,1), (4,2)$

0	*	0	+		
*	0	*	0		
0	*	0	*		
+	0	*	0	*	
		*	0	*	
		*	0	*	

→ bulge

Q_1 that acts on $(2,4)$ rows.

bulge shifts as follows

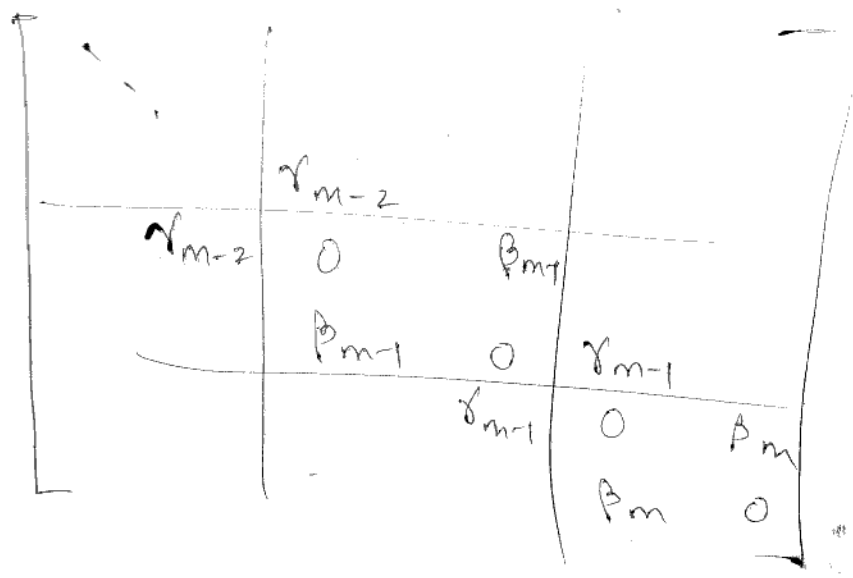
0	*	0	0	0	0
*	0	*	0	+	0
0	*	0	*	0	0
*	0	*	0	*	0
0	+	0	*	0	*
0	0	0	0	*	0

→ bulge

Q_2 acting on $3, 5^{\text{th}}$ rows.

0	*				
*	0				
0	*	0	0	0	+
0	0	*	*	*	0
0	0	0	*	*	*
0	0	+	0	*	0

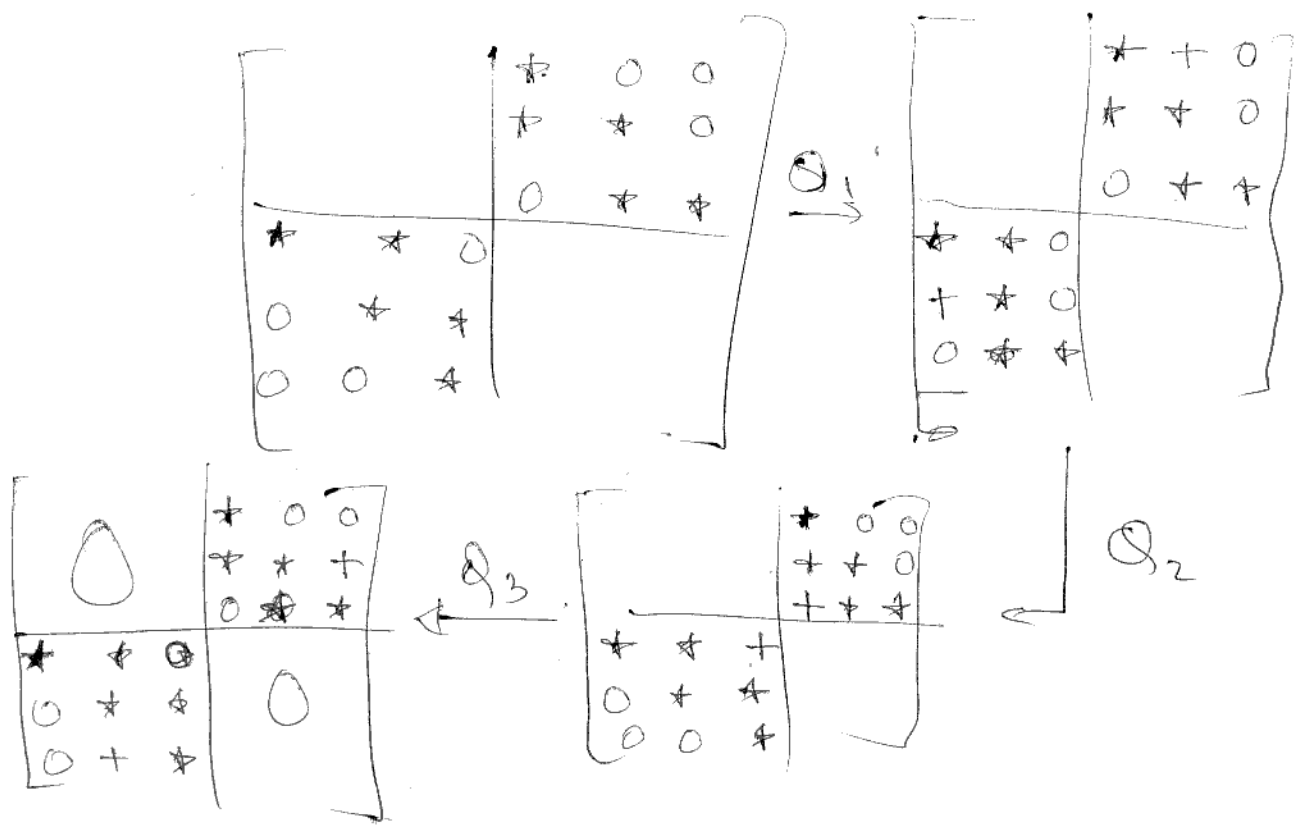
and so on.



$\gamma_{m-1} \rightarrow 0$ then $\pm \beta_m$ eigenvalues.

Recall $\tilde{C} = P^T C P$.

What goes on in original unshuffled mode



$$\begin{bmatrix} 0 & B^T \\ B & 0 \end{bmatrix} \begin{bmatrix} V & V \\ U & -U \end{bmatrix} = \begin{bmatrix} V & V \\ U & -U \end{bmatrix} \begin{bmatrix} \Sigma & 0 \\ 0 & -\Sigma \end{bmatrix}$$

- all collected
orthogonal
matrices
in deshuffled
form.

Transpose
of this
was ~~pre~~ being
premultiplied to

$$\begin{bmatrix} B^T U & -B^T V \\ B V & B U \end{bmatrix} = \begin{bmatrix} U \Sigma & -U \Sigma \\ U \Sigma & -U \Sigma \end{bmatrix} \begin{bmatrix} 0 & B^T \\ B & 0 \end{bmatrix}$$

$$B^T U = V \Sigma$$

$$B V = U \Sigma$$

$$U^T B V = \Sigma$$

$$\boxed{B = U \Sigma V^T}$$

SVD.