

# GSVD Progress

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## 1 Week of June 21, 2019

### 1.1 Quick exit from GSVD main routine

Given that  $A \in \mathbb{R}^{m \times n}$  and  $B \in \mathbb{R}^{p \times n}$ ,  $\text{rank}(B) = l$  and  $\text{rank}([A; B]) = k + l = r$ , if  $m = k$ , we can exit from main routine once preprocessing is done. Justification is the following:

In 3.1, after preprocessing, we obtain  $R_1$  and  $R_2$ . Let's take a close look at how we compute  $R_1$  and  $R_2$  progressively.

Initially,  $A^{(0)} = A$ ,  $B^{(0)} = B$ .

**Step 1:**  $URV$  decomposition of  $B$ .

$$\begin{matrix} m \\ p \end{matrix} \begin{bmatrix} A^{(1)} \\ B^{(1)} \end{bmatrix} = \begin{matrix} m \\ l \\ p-l \end{matrix} \begin{bmatrix} A_{11}^{(1)} & A_{12}^{(1)} \\ 0 & B_{12}^{(1)} \\ 0 & 0 \end{bmatrix}$$

**Step 2:**  $URV$  decomposition of  $A_{11}^{(1)}$ , if  $A_{11}^{(1)}$  is not empty (0 col).

$$\begin{matrix} m \\ p \end{matrix} \begin{bmatrix} A^{(2)} \\ B^{(2)} \end{bmatrix} = \begin{matrix} s \\ m-s \\ l \\ p-l \end{matrix} \begin{bmatrix} 0 & A_{12}^{(2)} & A_{13}^{(2)} \\ 0 & 0 & A_{23}^{(2)} \\ 0 & 0 & B_{12}^{(1)} \\ 0 & 0 & 0 \end{bmatrix},$$

where  $s$  is the rank of  $A_{11}^{(1)}$ .

**Step 3:**  $URV$  decomposition of  $A_{23}^{(2)}$ , such that  $QA_{23}^{(3)} = A_{23}^{(2)}$ .

$$\begin{matrix} m \\ p \end{matrix} \begin{bmatrix} A^{(3)} \\ B^{(3)} \end{bmatrix} = \begin{matrix} s \\ m-s \\ l \\ p-l \end{matrix} \begin{bmatrix} 0 & A_{12}^{(3)} & A_{13}^{(3)} \\ 0 & 0 & A_{23}^{(3)} \\ 0 & 0 & B_{12}^{(3)} \\ 0 & 0 & 0 \end{bmatrix} = \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} R_1 \\ R_2 \end{bmatrix}$$

As is shown at **Step 2**, if  $A_{11}^{(1)}$  is full (row) rank, namely, its rank is  $m$ , then  $A_{23}^{(2)}$  will no longer exist. Thus, **Step 3** is not needed anymore. Further, we have:

$$\begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ R_1 \\ R_2 \end{bmatrix} = \begin{matrix} m \\ l \\ p-l \end{matrix} \begin{bmatrix} n-m-l & m & l \\ 0 & A_{12}^{(2)} & A_{13}^{(2)} \\ 0 & 0 & B_{13}^{(2)} \\ 0 & 0 & 0 \end{bmatrix} = \begin{matrix} m+l \\ p-l \end{matrix} \begin{bmatrix} n-m-l & m+l \\ 0 & R \\ 0 & 0 \end{bmatrix},$$

which itself is an upper triangular matrix.

Therefore, at this step, we find the exact GSVD of  $A, B$  with rank determination,

$$\begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ A \\ B \end{bmatrix} = \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} m & p \\ U_1 & 0 \\ 0 & V_1 \end{bmatrix} \begin{matrix} m+l \\ p-l \end{matrix} \begin{bmatrix} m+l \\ I \\ 0 \end{bmatrix} \begin{matrix} m+l \\ n \end{matrix} \begin{bmatrix} n-m-l & m+l \\ 0 & R \end{bmatrix} \begin{matrix} n \\ 1 \end{matrix} \begin{bmatrix} Q_1^T \end{bmatrix}$$

## 1.2 QR: Givens rotations VS Householder's method

Test matrices  $A$  and  $B$ :

```
julia> A = [1.0 2 1 0; 2 3 1 1; 3 4 1 2; 4 5 1 3; 5 6 1 4]
5×4 Array{Float64,2}:
 1.0  2.0  1.0  0.0
 2.0  3.0  1.0  1.0
 3.0  4.0  1.0  2.0
 4.0  5.0  1.0  3.0
 5.0  6.0  1.0  4.0

julia> B = [6.0 7 1 5; 7 1 -6 13; -4 8 9 -2]
3×4 Array{Float64,2}:
 6.0  7.0  1.0  5.0
 7.0  1.0 -6.0 13.0
-4.0  8.0  9.0 -2.0
```

Output of Givens rotations:

```
julia> test1()
0.000124 seconds (345 allocations: 39.219 KiB)
```

Output of Householder's method:

```
julia> test1()
0.000163 seconds (159 allocations: 20.094 KiB)
```

## 2 Week of May 31, 2019

### 2.1 Quick exit from GSVD main routine

Given that  $A \in \mathbb{R}^{m \times n}$  and  $B \in \mathbb{R}^{p \times n}$ ,  $\text{rank}(B) = l$  and  $\text{rank}([A; B]) = k + l = r$ , if  $m \leq k$ , we can exit from main routine once preprocessing is done. Justification is the following:

In 3.1, after preprocessing, we obtain  $R_1$  and  $R_2$ . Let's take a close look at the structure of  $R_1$  and  $R_2$ , respectively.

$$R_1 = \begin{matrix} & k & & \\ & l & & \\ m-k-l & & n-k-l & k & l \\ & & 0 & R_{12}^{(a)} & R_{13}^{(a)} \\ & & 0 & 0 & R_{23}^{(a)} \\ & & 0 & 0 & 0 \end{matrix} \begin{bmatrix} \\ \\ \\ \\ \\ \end{bmatrix}$$

$$R_2 = \begin{matrix} & & n-k-l & k & l \\ & & & & \\ l & & 0 & 0 & R_{13}^{(b)} \\ p-l & & 0 & 0 & 0 \end{matrix} \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

As is shown, if  $m \leq k$ ,  $R_{23}^{(a)}$  will no longer exist. Further, we have: (The follwing is wrong)

$$\begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ R_1 \\ R_2 \end{bmatrix} = \begin{matrix} m \\ l \\ p-l \end{matrix} \begin{matrix} & & n-k-l & m & k-m & l \\ & & & & & \\ 0 & R & T_1 & T_2 \\ 0 & 0 & 0 & R_{13}^{(b)} \\ 0 & 0 & 0 & 0 \end{matrix} \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

$$\begin{matrix} m & k-m & l \\ l \end{matrix} \begin{bmatrix} R & T_1 & T_2 \\ 0 & 0 & R_{13}^{(b)} \end{bmatrix} = \begin{matrix} m & l \\ l \end{matrix} \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} \begin{matrix} m & k-m & l \\ l \end{matrix} \begin{bmatrix} R & T_1 & T_2 \\ 0 & 0 & R_{13}^{(b)} \end{bmatrix}$$

## 2.2 Reduce space consumption in Givens rotations

still working on it, stay tuned...

## 3 Week of May 6, 2019

### 3.1 Derivation of GSVD from ([A;B]) without rank revealing

$$\begin{aligned} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ A \\ B \end{bmatrix} &\stackrel{\text{preprocessing}}{=} \begin{matrix} m & p \\ p \end{matrix} \begin{bmatrix} U_1 & 0 \\ 0 & V_1 \end{bmatrix} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ R_1 \\ R_2 \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_1^T \end{bmatrix} \\ &\stackrel{\text{split QR}}{=} \begin{matrix} m & p \\ p \end{matrix} \begin{bmatrix} U_1 & 0 \\ 0 & V_1 \end{bmatrix} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ Q_1 \\ Q_2 \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} R_3 \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_1^T \end{bmatrix} \\ &\stackrel{\text{CSD}}{=} \begin{matrix} m & p \\ p \end{matrix} \begin{bmatrix} U_1 & 0 \\ 0 & V_1 \end{bmatrix} \begin{matrix} m & p \\ p \end{matrix} \begin{bmatrix} U_2 & 0 \\ 0 & V_2 \end{bmatrix} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ C \\ S \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_2^T \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} R_3 \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_1^T \end{bmatrix} \\ &= \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} U_1 U_2 & 0 \\ 0 & V_1 V_2 \end{bmatrix} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ C \\ S \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_2^T R_3 \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_1^T \end{bmatrix} \\ &\stackrel{\text{RQ}}{=} \begin{matrix} m & p \\ p \end{matrix} \begin{bmatrix} U & 0 \\ 0 & V \end{bmatrix} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ C \\ S \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} R \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_3 \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q_1^T \end{bmatrix} \\ &= \begin{matrix} m & p \\ p \end{matrix} \begin{bmatrix} U & 0 \\ 0 & V \end{bmatrix} \begin{matrix} m \\ p \end{matrix} \begin{bmatrix} n \\ C \\ S \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} R \end{bmatrix} \begin{matrix} n \\ n \end{matrix} \begin{bmatrix} Q^T \end{bmatrix} \end{aligned}$$

### 3.2 Golub & Van Loan's definition, algorithm of GSVD (partially adopted by MatLab)

## 4 Week of April 29, 2019

### 4.1 Ordering of cos and sin

#### 4.1.1 LAPACK 3.6.0

The generalized (or quotient) singular value decomposition of an  $m$ -by- $n$  matrix  $A$  and a  $p$ -by- $n$  matrix  $B$  is given by the pair of factorizations:

$$A = U\Sigma_1[0, R]Q^T \quad \text{and} \quad B = V\Sigma_2[0, R]Q^T$$

$\Sigma_1$  is  $m$ -by- $r$ ,  $\Sigma_2$  is  $p$ -by- $r$ . (The integer  $r$  is the rank of  $\begin{pmatrix} A \\ B \end{pmatrix}$ , and satisfies  $r \leq n$ .) Both are real, non-negative and diagonal, and  $\Sigma_1^T \Sigma_1 + \Sigma_2^T \Sigma_2 = I$ . Write  $\Sigma_1^T \Sigma_1 = \text{diag}(\alpha_1^2, \dots, \alpha_r^2)$  and  $\Sigma_2^T \Sigma_2 = \text{diag}(\beta_1^2, \dots, \beta_r^2)$ , where  $\alpha_i$  and  $\beta_i$  lie in the interval from 0 to 1. The ratios  $\alpha_1/\beta_1, \dots, \alpha_r/\beta_r$  are called the generalized singular values of the pair  $A, B$ . If  $\beta_i = 0$ , then the generalized singular value  $\alpha_i/\beta_i$  is infinite. There is **no finite ordering** of  $\alpha$  and  $\beta$  in GSVD in LAPACK.

### 4.2 Features of my Julia version

#### 4.2.1 Mathematical Definition

The generalized singular value decomposition (GSVD) of an  $m$ -by- $n$  matrix  $A$  and  $p$ -by- $n$  matrix  $B$  is the following, where  $m + p \geq n$ :

$$A = UD_1RQ^T \quad \text{and} \quad B = VD_2RQ^T$$

where  $U, V$  and  $Q$  are orthogonal matrices. Let  $k + l$  be the effective numerical rank of the matrix  $\begin{pmatrix} A \\ B \end{pmatrix}$ , then  $R$  is a  $(k + l)$ -by- $n$  matrix of structure  $[0 \ R_0]$  where  $R_0$  is  $(k + l)$ -by- $(k + l)$  and is nonsingular upper triangular matrix,  $D_1$  and  $D_2$  are  $m$ -by- $(k + l)$  and  $p$ -by- $(k + l)$  non-negative "diagonal" matrices and satisfy  $D_1^T D_1 + D_2^T D_2 = I$ . The nonzero elements of  $D_1$  are in **non-increasing** order while the nonzero elements of  $D_2$  are in **non-decreasing** order.

(TO BE REVEALED) Detailed structure of  $D_1$  and  $D_2$ :

#### 4.2.2 API design

1.  $U, V, Q, \alpha, \beta, R, k, l = \text{gsvd}(A, B, 0)$
2.  $U, V, Q, D_1, D_2, R, k, l = \text{gsvd}(A, B, 1)$  or  $F = \text{gsvd}(A, B)$

Notice that both  $A$  and  $B$  are overwritten.

### 4.3 Issues to be resolved

## 5 Week of April 22, 2019

### 5.1 Ordering of cos and sin

#### 5.1.1 LAPACK 3.6.0

The generalized (or quotient) singular value decomposition of an  $m$ -by- $n$  matrix  $A$  and a  $p$ -by- $n$  matrix  $B$  is given by the pair of factorizations:

$$A = U\Sigma_1[0, R]Q^T \quad \text{and} \quad B = V\Sigma_2[0, R]Q^T$$

$\Sigma_1$  is  $m$ -by- $r$ ,  $\Sigma_2$  is  $p$ -by- $r$ . (The integer  $r$  is the rank of  $\begin{pmatrix} A \\ B \end{pmatrix}$ , and satisfies  $r \leq n$ .) Both are real, non-negative and diagonal, and  $\Sigma_1^T \Sigma_1 + \Sigma_2^T \Sigma_2 = I$ . Write  $\Sigma_1^T \Sigma_1 = \text{diag}(\alpha_1^2, \dots, \alpha_r^2)$  and  $\Sigma_2^T \Sigma_2 = \text{diag}(\beta_1^2, \dots, \beta_r^2)$ , where  $\alpha_i$  and  $\beta_i$  lie in the interval from 0 to 1. The ratios  $\alpha_1/\beta_1, \dots, \alpha_r/\beta_r$  are called the generalized singular values of the pair  $A, B$ . If  $\beta_i = 0$ , then the generalized singular value  $\alpha_i/\beta_i$  is infinite.

#### 5.1.2 MATLAB 2019b

$[U, V, X, C, S] = \text{gsvd}(A, B)$  returns unitary matrices  $U$  and  $V$ , a (usually) square matrix  $X$ , and non-negative diagonal matrices  $C$  and  $S$  so that

$$\begin{aligned} A &= U * C * X' \\ B &= V * S * X' \\ C' * C + S' * S &= I \end{aligned}$$

$A$  and  $B$  must have the same number of columns, but may have different numbers of rows. If  $A$  is  $m$ -by- $p$  and  $B$  is  $n$ -by- $p$ , then  $U$  is  $m$ -by- $m$ ,  $V$  is  $n$ -by- $n$ ,  $X$  is  $p$ -by- $q$ ,  $C$  is  $m$ -by- $q$  and  $S$  is  $n$ -by- $q$ , where  $q = \min(m + n, p)$ .

The nonzero elements of  $S$  are always on its main diagonal. The nonzero elements of  $C$  are on the diagonal  $\text{diag}(C, \max(0, q - m))$ . If  $m \geq q$ , this is the main diagonal of  $C$ .

$\text{sigma} = \text{gsvd}(A, B)$  returns the vector of generalized singular values,  $\text{sqr}t(\text{diag}(C' * C) ./ \text{diag}(S' * S))$ . The vector  $\text{sigma}$  has length  $q$  and is in **non-decreasing** order, where  $\mathbf{q} = \mathbf{min}(\mathbf{m} + \mathbf{n}, \mathbf{p})$ . In other words, The nonzero elements of  $S$  are in **non-increasing** order while the nonzero elements of  $C$  are in **non-decreasing** order.

It's interesting to notice that MATLAB has a compact form of products for  $\text{gsvd}$ .

#### 5.1.3 Julia 1.10

Compute the generalized SVD of  $A$  and  $B$ , returning a GeneralizedSVD factorization object  $F$ , such that  $A = F.U * F.D_1 * F.R_0 * F.Q'$  and  $B = F.V * F.D_2 * F.R_0 * F.Q'$ .

The entries of  $F.D_1$  and  $F.D_2$  are related, as explained in the LAPACK documentation for the generalized SVD and the xGGSVD3 routine which is called underneath (in LAPACK 3.6.0 and newer).

## 5.2 Features of my Julia version

### 5.2.1 Definition

The generalized singular value decomposition (GSVD) of an  $m$ -by- $n$  matrix  $A$  and  $p$ -by- $n$  matrix  $B$  is the following, where  $m + p \geq n$ :

$$A = U D_1 R Q^T \quad \text{and} \quad B = V D_2 R Q^T$$

where  $U, V$  and  $Q$  are orthogonal matrices. Let  $k + l$  be the effective numerical rank of the matrix  $\begin{pmatrix} A \\ B \end{pmatrix}$ , then  $R$  is a  $(k + l)$ -by- $n$  matrix of structure  $[0 \ R_0]$  where  $R_0$  is  $(k + l)$ -by- $(k + l)$  and is nonsingular upper triangular matrix,  $D_1$  and  $D_2$  are  $m$ -by- $(k + l)$  and  $p$ -by- $(k + l)$  non-negative “diagonal” matrices and satisfy  $D_1^T D_1 + D_2^T D_2 = I$ . The nonzero elements of  $D_1$  are in **non-increasing** order while the nonzero elements of  $D_2$  are in **non-decreasing** order.

### 5.2.2 2 APIs

1. LAPACK-style:  $U, V, Q, \alpha, \beta, R, k, l = gsvd(A, B, 0)$
2. Julia-style:  $U, V, Q, D_1, D_2, R, k, l = gsvd(A, B, 1)$

## 5.3 Testing

### 5.3.1 Alan testing

Still suffers some subtlety in the dimension of  $D_1$  and  $D_2$ .

### 5.3.2 Spatial matrices testing

Tested example matrices pair from documentation, and publication.

### 5.3.3 General testing

Tested function correctness, time performance and stability.