

Numerical behavior of GMRES for singular EP and GP systems

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Generalized Minimal RESidual method [Saad, Schultz 1986]

Given the system $Ax = b$, the initial guess x_0 and residual $r_0 = b - Ax_0$, GMRES computes x_k over $x_0 + \mathcal{K}_k(A, r_0)$, where

$$\mathcal{K}_k(A, r_0) = \text{span}\{r_0, Ar_0, \dots, A^{k-1}r_0\}$$

is the k -th Krylov subspace, such that x_k minimizes $\|b - Ax_k\|$.

Let $AQ_k = Q_{k+1}H_{k+1,k}$ be the Arnoldi decomposition, where the columns of $Q_k \in \mathbb{R}^{n \times k}$ form an orthonormal basis of \mathcal{K}_k and $H_{k+1,k} \in \mathbb{R}^{(k+1) \times k}$ is an extended Hessenberg matrix.

Then the iterate $x_k = x_0 + Q_k y_k$ satisfies

$$\begin{aligned}\|b - Ax_k\| &= \|r_0 - A Q_k y_k\| = \| \|r_0\| e_1 - H_{k+1,k} y_k\| \\ &= \min_{y \in \mathbb{R}^k} \| \|r_0\| e_1 - H_{k+1,k} y\|, \quad e_1 = [1, 0, \dots, 0]^\top.\end{aligned}$$

Numerical behavior of GMRES

The restriction $A|_{\mathcal{K}_k}$ of A to the Krylov subspace $\mathcal{K}_k \subseteq \mathbb{R}^n$ plays an important role. Its condition number is defined as

$$\kappa(A|_{\mathcal{K}_k}) = \frac{\max_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \|Az\| / \|z\|}{\min_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \|Az\| / \|z\|}.$$

The identity $\kappa(A|_{\mathcal{K}_k}) = \kappa(H_{k+1,k})$ ($\equiv \|H_{k+1,k}\| \|H_{k+1,k}^\dagger\|$) follows from

$$\{\max, \min\}_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} = \{\max, \min\}_{w \in \mathbb{R}^k \setminus \{\mathbf{0}\}} \frac{\|AQ_k w\|}{\|Q_k w\|} = \{\max, \min\}_{w \in \mathbb{R}^k \setminus \{\mathbf{0}\}} \frac{\|H_{k+1,k} w\|}{\|w\|}.$$

The accuracy of x_k computed in finite precision arithmetic is affected by the conditioning of $H_{k+1,k}$.

GMRES applied to nonsingular systems

From $AQ_k = Q_{k+1}H_{k+1,k}$ and $Q_{k+1}^T Q_{k+1} = I_{k+1}$ it follows that

$$\|H_{k+1,k}\| = \|AQ_k\| \leq \|A\| \|Q_k\| = \|A\|$$

and for nonsingular A we have

$$\sigma_k(H_{k+1,k}) = \sigma_k(AQ_k) \geq \sigma_n(A)\sigma_k(Q_k) \geq \sigma_n(A) > 0.$$

Then the solution $\mathbf{y}_k = H_{k+1,k}^\dagger \|\mathbf{r}_0\| \mathbf{e}_1$ of the upper Hessenberg least squares problem satisfies the bounds

$$\|\mathbf{y}_k\| \leq \frac{\|\mathbf{r}_0\|}{\sigma_k(H_{k+1,k})} \leq \frac{\|\mathbf{r}_0\|}{\sigma_n(A)}.$$

GMRES on singular systems - example of a breakdown

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \in \mathcal{R}(A),$$

$$\mathcal{R}(A) = \mathcal{N}(A) = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}, \quad \mathcal{R}(A^\top) = \mathcal{N}(A^\top) = \text{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\},$$

$$\mathbf{x}_0 = 0, \quad \mathbf{r}_0 = \mathbf{b} - A\mathbf{x}_0 \in \mathcal{N}(A)$$

If $\mathbf{b} \in \mathcal{R}(A)$ and $\mathbf{r}_0 \in \mathcal{R}(A) \cap \mathcal{N}(A) \neq \{\mathbf{0}\}$, then GMRES breaks down at step 1 without determining a solution of $A\mathbf{x} = \mathbf{b}$.

Range-symmetric (EP) systems

A matrix $A \in \mathbb{R}^{n \times n}$ s.t. $\mathcal{R}(A^T) = \mathcal{R}(A)$, is called an EP (equal projection) or range-symmetric matrix.

Theorem ([Brown, Walker, SIMAX 97])

If $\mathcal{R}(A^T) = \mathcal{R}(A)$, then GMRES gives a solution x_* of $\min_{x \in \mathbb{R}^n} \|b - Ax\|$ without breakdown for all $b \in \mathbb{R}^n$ and $\forall x_0 \in \mathbb{R}^n$.

Note that $\mathcal{R}(A^T) = \mathcal{R}(A) \implies \mathcal{N}(A) \cap \mathcal{R}(A) = \{\mathbf{0}\}$.

We say a matrix A such that $\mathcal{N}(A) \cap \mathcal{R}(A) = \{\mathbf{0}\}$ is a group (GP) matrix.

Range-symmetric (EP) system, consistent case $\mathbf{b} \in \mathcal{R}(A)$

Since $\mathcal{K}_k \subseteq \mathcal{R}(A)$, we have

$$\sigma_k(H_{k+1,k}) = \min_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} \geq \min_{z \in \mathcal{R}(A) \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|}$$

and thus $\kappa(H_{k+1,k}) \leq \kappa(A|_{\mathcal{R}(A)})$. If A is EP, then

$$\min_{z \in \mathcal{R}(A) \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} = \min_{z \in \mathcal{R}(A^T) \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} = \sigma_{\min}(A) > 0$$

and

$$\kappa(A|_{\mathcal{R}(A)}) = \frac{\|A\|}{\min_{z \in \mathcal{R}(A^T) \setminus \{\mathbf{0}\}} \|Az\|/\|z\|} = \kappa(A).$$

Consistent EP case \sim nonsingular case, i.e., $\kappa(H_{k+1,k}) \leq \kappa(A)$.

Range-symmetric (EP) system, inconsistent case $\mathbf{b} \notin \mathcal{R}(A)$

Since $\mathcal{R}(A^T) = \mathcal{R}(A) \iff \mathcal{N}(A^T) = \mathcal{N}(A)$, the LS residual $\mathbf{r}_* = \mathbf{b}|_{\mathcal{N}(A^T)}$ satisfies $A^T \mathbf{r}_* = 0$, and so $\mathbf{r}_* \in \mathcal{N}(A^T)$ belongs also to $\mathcal{N}(A)$ with

$$\sigma_k(H_{k+1,k}) = \min_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} \geq \min_{z \in \text{span}\{\mathbf{r}_*\} \cup \mathcal{R}(A) \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} = 0.$$

Since $\mathbf{r}_{k-1} \in \mathcal{K}_k$ satisfies $\mathbf{r}_{k-1} - \mathbf{r}_* \in \mathcal{R}(A)$ and $A\mathbf{r}_* = \mathbf{0}$, we have

$$\sigma_k(H_{k+1,k}) = \min_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} \leq \frac{\|A\mathbf{r}_{k-1}\|}{\|\mathbf{r}_{k-1}\|} = \frac{\|A(\mathbf{r}_{k-1} - \mathbf{r}_*)\|}{\|\mathbf{r}_{k-1}\|} \leq \|A\| \frac{\|\mathbf{r}_{k-1} - \mathbf{r}_*\|}{\|\mathbf{r}_{k-1}\|}.$$

The Hessenberg least squares problem thus becomes very ill-conditioned before a LS solution is reached by GMRES.

EP case for $\mathbf{b} \notin \mathcal{R}(A)$

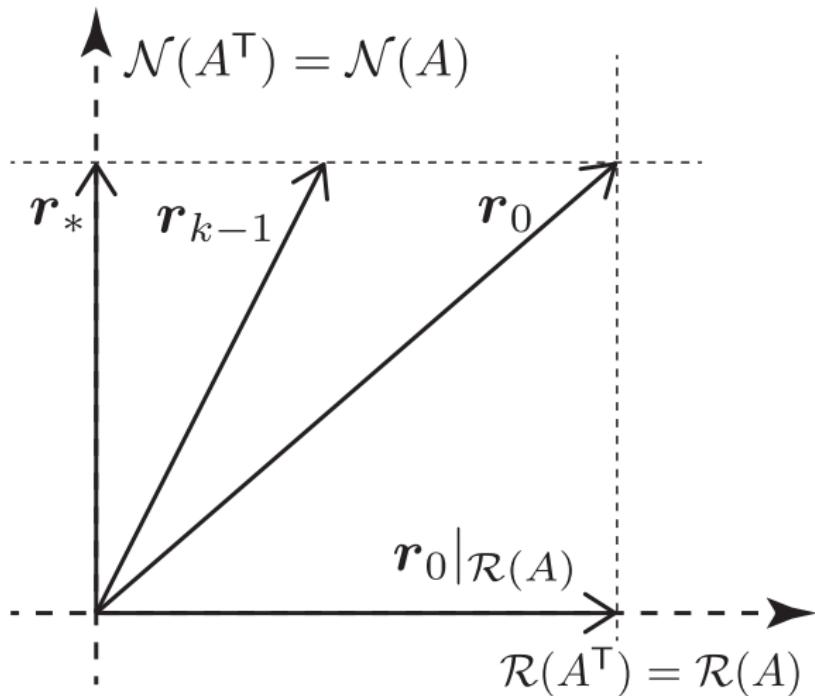


Figure: Geometric illustration of residual vectors.

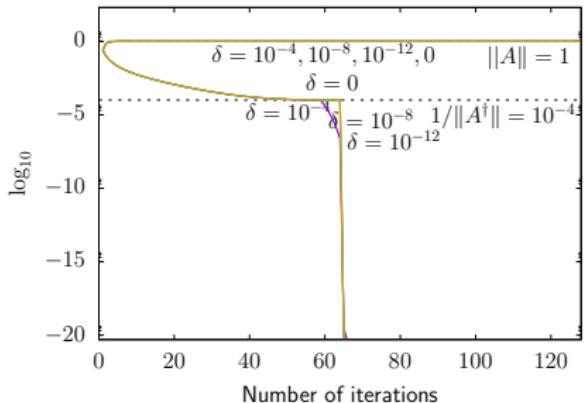
Example: small singular EP system

$$A = \begin{bmatrix} \text{diag}(10^{\frac{0}{127}}, 10^{\frac{-4}{127}}, 10^{\frac{-8}{127}}, \dots, 10^{-4}) & \mathbf{O} \\ \mathbf{O} & \mathbf{O} \end{bmatrix} \in \mathbb{R}^{128 \times 128}, \quad \mathbf{b} = \begin{bmatrix} \boldsymbol{\gamma} \\ \boldsymbol{\delta} \end{bmatrix},$$

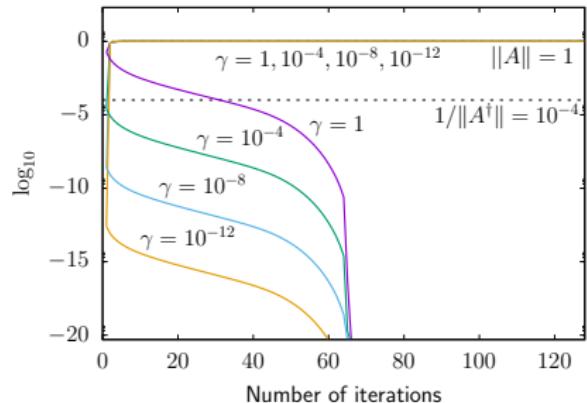
where $\boldsymbol{\gamma} = [\gamma, \dots, \gamma]^T \in \mathbb{R}^{64}$ and $\boldsymbol{\delta} = [\delta, \dots, \delta]^T \in \mathbb{R}^{64}$, i.e., $\kappa(A) = 10^4$, and $\mathbf{b} \notin \mathcal{R}(A) \iff \delta \neq 0$.

Inconsistency is controlled by δ/γ .

Example: GMRES



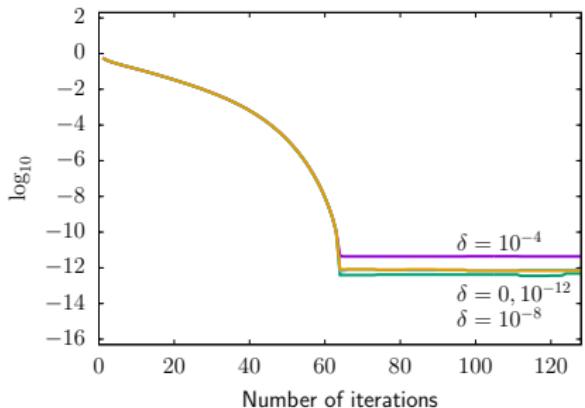
(a) Weakly inconsistent cases $\gamma = 1$.



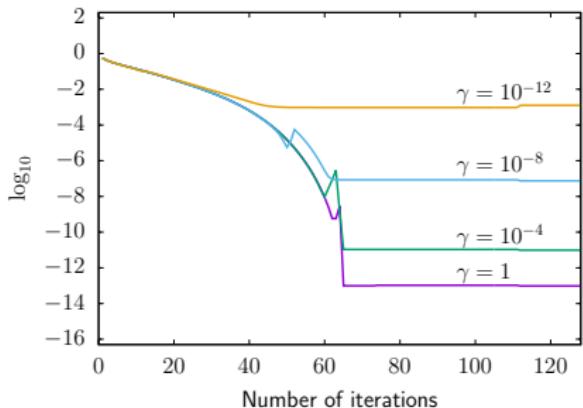
(b) Strongly inconsistent cases $\delta = 1$.

Figure: Extreme singular values of A and $H_{k+1,k}$.

Example: GMRES



(a) Weakly inconsistent cases $\gamma = 1$.



(b) Strongly inconsistent cases $\delta = 1$.

Figure: Relative residual norm $\|A^T \mathbf{r}_k\| / \|A^T \mathbf{b}\|$.

Singular value decomposition of A

$$A = U\Sigma V^T = [U_1, U_2] \begin{bmatrix} \Sigma_1 & 0 \\ 0 & 0 \end{bmatrix} [V_1, V_2]^T,$$

where Σ_1 is diagonal with positive singular values, $U = [U_1, U_2]$ and $V = [V_1, V_2]$ are orthogonal matrices,

$$\mathcal{R}(A) = \mathcal{R}(U_1), \quad \mathcal{R}(A)^\perp = \mathcal{N}(A^T) = \mathcal{R}(U_2),$$

$$\mathcal{N}(A)^\perp = \mathcal{R}(A^T) = \mathcal{R}(V_1), \quad \mathcal{N}(A) = \mathcal{R}(V_2),$$

EP and GP matrices via singular value decomposition

- An EP matrix A , $\mathcal{R}(A^T) = \mathcal{R}(A)$, satisfies

$$\mathcal{R}(V_1) = \mathcal{R}(U_1) \iff V_1^T U_1 \text{ is orthogonal}$$

since $V_1^T U_2 = \mathbf{0}$.

- A GP matrix A , $\mathcal{N}(A) \cap \mathcal{R}(A) = \{\mathbf{0}\}$, satisfies

$$\mathcal{R}(U_1) \cap \mathcal{R}(V_2) = \{\mathbf{0}\} \iff \text{rank}([U_1, V_2]) = n \iff V_1^T U_1 \text{ is nonsingular,}$$

but $V_1^T U_1$ can be ill-conditioned.

The cosines of the principal angles between $\mathcal{R}(A)$ and $\mathcal{R}(A^T)$ are the singular values of $V_1^T U_1$.

- EP: Cosines between $\mathcal{R}(A)$ and $\mathcal{R}(A^T)$ are all zero
- GP: Condition number of $V_1^T U_1$ are the ratio of the cosines

Group inverse and Moore-Penrose inverse of A

$$A^\dagger = [V_1, V_2] \begin{bmatrix} \Sigma_1^{-1} & 0 \\ 0 & 0 \end{bmatrix} [U_1, U_2]^T$$

$$\mathcal{R}(A) + \mathcal{N}(A) = \mathbb{R}^n \iff \text{index}(A) = 1$$

$$A = S \begin{bmatrix} J & 0 \\ 0 & 0 \end{bmatrix} S^{-1}, \quad A^\# = S \begin{bmatrix} J^{-1} & 0 \\ 0 & 0 \end{bmatrix} S^{-1}, \quad \mathcal{R}(A^\#) = \mathcal{R}(A)$$

$$A^\# = A^\dagger \iff \mathcal{R}(A^\top) = \mathcal{R}(A)$$

Convergence of GMRES applied to group (GP) systems

Theorem ([Brown, Walker, SIMAX 97])

If $\mathcal{R}(A) \cap \mathcal{N}(A) = \{\mathbf{0}\}$, then GMRES gives a solution of $A\mathbf{x} = \mathbf{b}$ without breakdown for all $\mathbf{b} \in \mathcal{R}(A)$ and for all $\mathbf{x}_0 \in \mathbb{R}^n$. The solution is $\mathbf{x}_0 + A^\# \mathbf{r}_0 = \mathbf{x}_\# + (I - A^\# A)\mathbf{x}_0$.

$$\mathbf{x}_* = A^\dagger \mathbf{b} \in \mathcal{R}(A^\top), \quad \mathbf{x}_\# = A^\# \mathbf{b} \in \mathcal{R}(A^\top), \quad \mathbf{x}_* = P_{\mathcal{R}(A^\top)} \mathbf{x}_\# = V_1 V_1^\top U_1 U_1^\top \mathbf{x}_\#$$

$$\sigma_{min}(V_1^\top U_1) \|\mathbf{x}_\#\| \leq \|\mathbf{x}_*\| \leq \|\mathbf{x}_\#\|$$

GP case for $\mathbf{b} \in \mathcal{R}(A)$

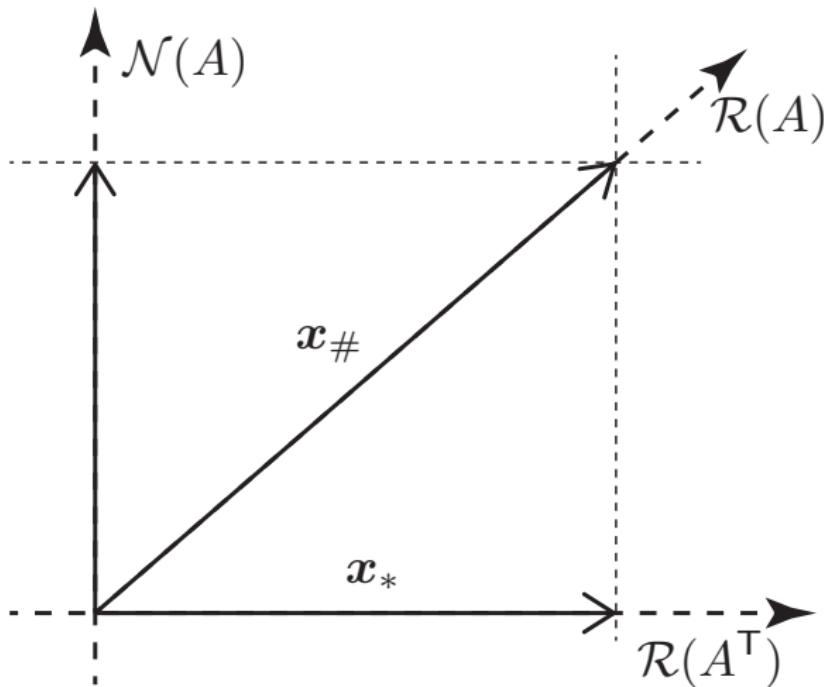


Figure: Geometric illustration of solution vectors.

Group (GP) system, consistent case $\mathbf{b} \in \mathcal{R}(A)$

Since $\mathcal{K}_k \subseteq \mathcal{R}(A)$, we have

$$\sigma_k(H_{k+1,k}) = \min_{z \in \mathcal{K}_k \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} \geq \min_{z \in \mathcal{R}(A) \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|}$$

$$\min_{z \in \mathcal{R}(A) \setminus \{\mathbf{0}\}} \frac{\|Az\|}{\|z\|} = \min_{z \in \mathbb{R}^k \setminus \{\mathbf{0}\}} \frac{\|U_1 \Sigma_1 V_1^\top U_1 z\|}{\|U_1 z\|} \geq \sigma_{\min}(A) \sigma_{\min}(V_1^\top U_1)$$

Condition number of the Hessenberg matrix $H_{k+1,k}$ depends on the principal angles between $\mathcal{R}(A)$ and $\mathcal{R}(A^\top)$!

Example: small singular GP system

$$A = \begin{bmatrix} D & I \\ 0 & 0 \end{bmatrix} \in \mathbb{R}^{128 \times 128}, \quad \mathbf{b} = \begin{bmatrix} f \\ \mathbf{0} \end{bmatrix}$$

D is a diagonal matrix whose values of the diagonal entries have the so-called Strakoš distribution [Strakoš, LAA 91]

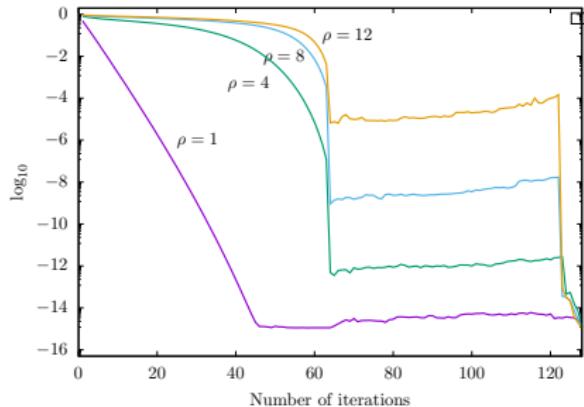
$$d_{1,1} = 1, \quad d_{64,64} = 10^{-\rho}, \quad d_{j,j} = d_{64,64} + \frac{64-j}{63}(d_{1,1} - d_{64,64}) \cdot 0.7^{j-1},$$

$\mathbf{f} = \{f_i\} \in \mathbb{R}^r$ has the entries $f_{r-i+1} = 10^{-(i-1)\rho/63}$. For $\rho \gg 1$,

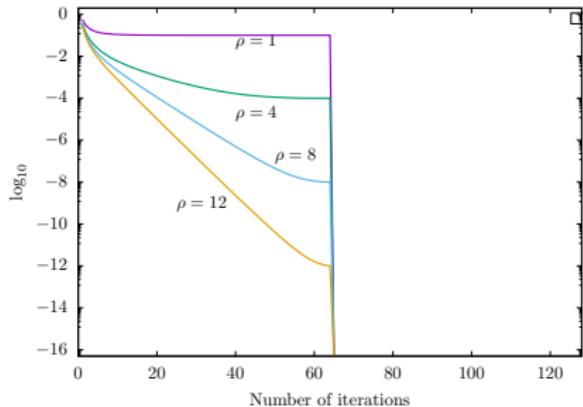
$$\kappa(A) = \sqrt{2/(10^{-\rho} + 1)} \simeq \sqrt{2},$$

$$\kappa(V_1^\top U_1) = 10^\rho \sqrt{(10^{-2\rho} + 1)/2} \simeq 10^\rho / \sqrt{2}.$$

Example: GMRES



(a) Relative residual norm $\|\mathbf{r}_k\|/\|\mathbf{b}\|$.



(b) Smallest singular value of $H_{k+1,k}$.

Figure: Different condition numbers of $V_1^T U_1$.

Thank you for your attention!

Morikuni Keiichi, R: On GMRES for singular EP and GP systems,
SIAM Journal on Matrix Analysis and Applications 39 (2), 2018,
1033-1048.