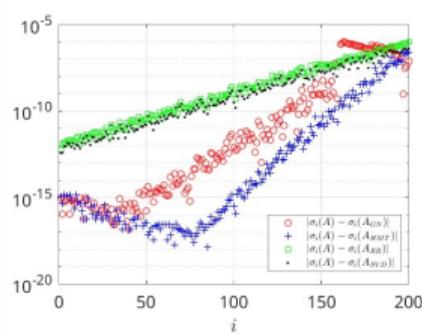


Given $\tilde{U} \in \mathbb{R}^{n \times r}, \tilde{V} \in \mathbb{R}^{n \times r}$ (orthonormal) approximations of the leading singular vectors of $A \in \mathbb{R}^{m \times n}, m \geq n$, extract the first r singular values $\{\sigma_i(A)\}_{i=1}^r$.

- Rayleigh Ritz: $\sigma_i(A) \approx \sigma_i(\tilde{U}^* A \tilde{V})$
- SVD approximation: $\sigma_i(A) \approx \sigma_i(A \tilde{V})$
- Generalized Nyström:

$$\sigma_i(A) \approx \sigma_i(A \tilde{V} (\tilde{U}^* A \tilde{V})^\dagger \tilde{U}^* A)$$
- HMT: $A \tilde{V} = QR, \sigma_i(A) \approx \sigma_i(Q^* A)$



1. Interpret as Perturbation

$$Q_1^* (A - A_{GN, \tilde{V}, \tilde{U}}) Q_2 = \begin{bmatrix} 0 & 0 \\ 0 & \bar{A}_{22} - \bar{A}_{21} \bar{A}_{11}^\dagger \bar{A}_{12} \end{bmatrix}$$

2. Matrix Perturbation Result

$$H := \begin{bmatrix} G_1 & B \\ C & G_2 \end{bmatrix} \quad F := \begin{bmatrix} F_{11} & F_{12} \\ F_{21} & F_{22} \end{bmatrix}$$

Define, for $i = 1, \dots, n$,

$$\tau_i = \left(\frac{\max\{\|B\|_2, \|C\|_2\} + \max\{\|F_{12}\|_2, \|F_{21}\|_2\}}{\min_k |\sigma_i - \sigma_k(G_2)| - 2\|F\|_2} \right).$$

Then, for all i for which $\tau_i > 0$, it holds

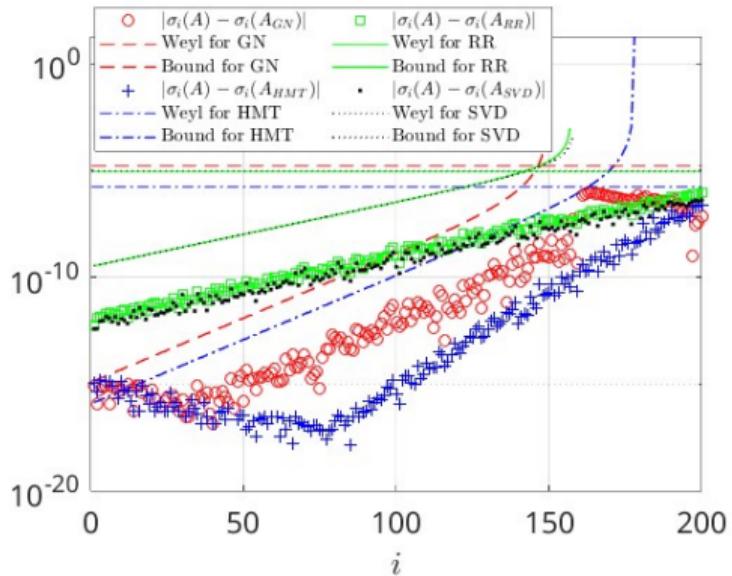
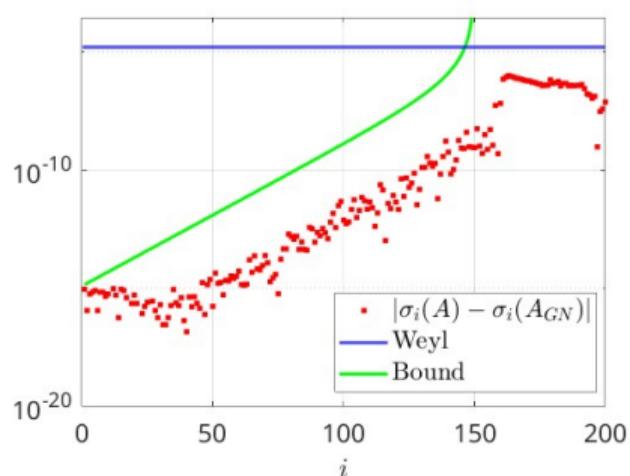
$$|\sigma_i - \hat{\sigma}_i| \leq \|F_{11}\|_2 + 2 \max\{\|F_{12}\|_2, \|F_{21}\|_2\} \tau_i + \|F_{22}\|_2 \tau_i^2,$$

3. Application to Methods

$$\tau_i = \frac{\max\{\|\bar{A}_{12}\|_2, \|\bar{A}_{21}\|_2\}}{\min_k |\sigma_i - \sigma_k(\bar{A}_{22})| - 2\|E_{GN}\|_2}.$$

Then, for all i for which $\tau_i > 0$, it holds

$$|\sigma_i - \sigma_i^{GN}| \leq \left\| \bar{A}_{22} - \bar{A}_{21}\bar{A}_{11}^\dagger\bar{A}_{12} \right\|_2 \tau_i^2$$



- More formal comparison of methods
- Computability
- Oversampling vs No-oversampling?