funNyström: Randomized low-rank approximation of monotone matrix functions

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Problem statement

Given

- 1. **LARGE** $A \succeq 0$ (numerically low rank);
- 2. Operator monotone f s.t. f(0) = 0 $(A \succeq B \Rightarrow f(A) \succeq f(B))$.

Find **LOW RANK** \widehat{B} such that

$$\hat{\boldsymbol{B}} \approx f(\boldsymbol{A})$$

Operator monotone functions:

$$x, \quad \sqrt{x}, \quad x^r \text{ for } r \in [0,1], \quad \log(1+x), \quad \frac{x}{x+\mu} \text{ for } \mu > 0, \dots$$

sums, compositions, positive scalings, ...

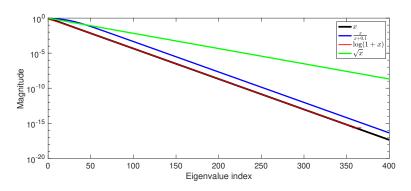
Problem statement

Why monotone and f(0) = 0? (Comment on operator monotonicity later)

f is continuous (implied by operator monotonicity)

 $\Rightarrow f({m A})$ is (numerically) low rank if ${m A}$ is

 \Rightarrow Low rank approximation makes sense



Applications

1. Trace estimation:

$$\operatorname{tr}(\widehat{\boldsymbol{B}}) \approx \operatorname{tr}(f(\boldsymbol{A}))$$

- Nuclear norm estimation: \sqrt{x} ;
- Statistical learning: $\log(1+x)$;
- UQ: $\log(1+x), \frac{x}{x+\mu}$.
- 2. Fast matvecs with $f(\mathbf{A})$:

$$\widehat{\boldsymbol{B}}\boldsymbol{x} \approx f(\boldsymbol{A})\boldsymbol{x}$$

- Sampling from elliptical distributions: \sqrt{x} .
- 3. Diagonal estimation:

$$\mathsf{diag}(\widehat{\boldsymbol{B}}) \approx \mathsf{diag}(f(\boldsymbol{A}))$$

- Ridge leverage scores: $\frac{x}{x+\mu}$.

Low rank approximation of matrix functions - First ideas

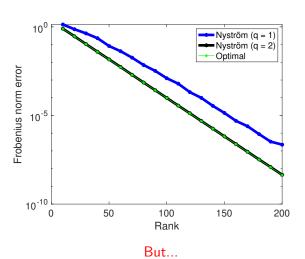
Method 1: Optimal approach via eig/svd $O(n^3)$ **Method 2:** Construct low rank approximation via matvecs.

- Randomized SVD [Halko/Martinsson/Tropp'11]
- Nyström approximation $(f(A) \succeq 0)$ [Gittens/Mahoney'13, Tropp/Yurtsever/Udell/Cevher'17]
- 1. Sample random $n \times (k + p)$ matrix Ω ;
- 2. $\mathbf{Q} = \operatorname{orth}(f(\mathbf{A})^{q-1}\mathbf{\Omega})$
- 3. Return $\hat{B} = f(A)Q(Q^Tf(A)Q)^{\dagger}(f(A)Q)^T$.

Nyström costs q(k+p) matvecs with $f(\mathbf{A})!$

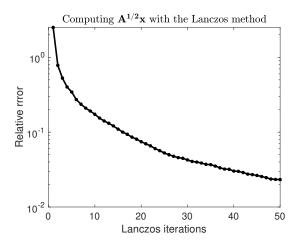
Low rank approximation of matrix functions - First ideas

Nyström is very good!



Low rank approximation of matrix functions - First ideas

Computing $f(A)\Omega$ is expensive (compared to $A\Omega$)!



(Rational Krylov, and other methods, are also 'expensive'.)

Low rank approximation of matrix functions - Better ideas?

We want: obtain low rank approximation using much fewer matvecs.

Back to the setting...

- 1. $A \succeq 0$;
- 2. f is (operator) monotone and f(0) = 0.

Lemma: Let A_k be rank-k truncated SVD. Then...

$$f(A_k)$$
 is a best rank- k approximation to $f(A)$!

Idea: Compute Nyström approximation \widehat{A} of A and use approximation

$$f(\widehat{\boldsymbol{A}}) \approx f(\boldsymbol{A}).$$

Bypasses the need for matrix-vector products with f(A)!

Similar idea in trace estimation for $f(x) = x, \log(1+x), \frac{x}{x+1}$ [Saibaba/Alexanderian/Ipsen'17, Herman/Alexanderian/Saibaba'20].

funNyström

1. Obtain eigenvalue decomposition of Nyström approximation

$$\widehat{\boldsymbol{A}} = \boldsymbol{A}^q \boldsymbol{\Omega} (\boldsymbol{\Omega}^T \boldsymbol{A}^{2q-1} \boldsymbol{\Omega}) \boldsymbol{\Omega}^T \boldsymbol{A}^q = \widehat{\boldsymbol{U}} \widehat{\boldsymbol{\Lambda}} \widehat{\boldsymbol{U}}^T.$$

2. Return low-rank approximation of f(A)

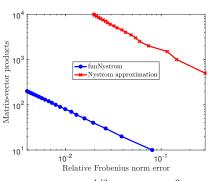
$$f(\widehat{\boldsymbol{A}}) = \widehat{\boldsymbol{U}}f(\widehat{\boldsymbol{\Lambda}})\widehat{\boldsymbol{U}}^T.$$

Potential benefits:

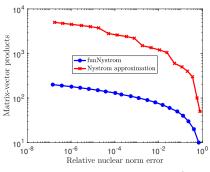
- No approximations of f(A)x \Rightarrow funNyström is much cheaper than Nyström on f(A).
- It can even be more accurate!

Numerical results

How many matvecs do we save?



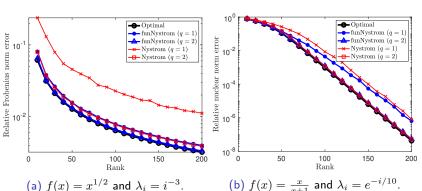
(a)
$$f(x) = x^{1/2}$$
 and $\lambda_i = i^{-3}$.



(b)
$$f(x) = \frac{x}{x+1}$$
 and $\lambda_i = 10e^{-i/10}$.

Numerical results

What if f(A)x is very cheap?



(b) $f(x) = \frac{x}{x+1}$ and $\lambda_i = e^{-i/10}$.

Theoretical results

Let
$$\gamma = \lambda_{k+1}/\lambda_k$$
 and $q \geq 2$

$$\|\mathbb{E}\|f(\mathbf{A}) - f(\widehat{\mathbf{A}})\|_F^2 \le \left(1 + \gamma^{2(q-3/2)} \frac{5k}{p-1}\right) \|f(\mathbf{\Lambda}_2)\|_F^2$$

Assumption $q \ge 2$ can be removed at the cost of a weaker bound.

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If $q \ge 1$

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If $q \geq 1$

$$\mathbb{E}\|f(\mathbf{A}) - f(\widehat{\mathbf{A}})\|_{2} \le \mathbb{E}f(\|\mathbf{A} - \widehat{\mathbf{A}}\|_{2}) \le f(\mathbb{E}\|\mathbf{A} - \widehat{\mathbf{A}}\|_{2}) \le \|f(\mathbf{\Lambda}_{2})\|_{2} + \|f(\mathbf{\Lambda}_{2}$$

Other remarks

Application to trace estimation

 $funNystr\"om + Hutch++ \Rightarrow low rank approx. phase cheaper$

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Operator monotonicity?

- Empirically, the bounds do not hold for any arbitrary monotone functions.
- $f(x) = x^3$ is an example...
- ... but funNyström still good provided you set q=3!