

## Sketching for low-rank matrix approximation

## ► Setting

Sketching is a popular method for low-rank approximation (that we did not cover in the lectures); see e.g. [1]. The basic idea is that instead of working with a large matrix  $A \in \mathbb{R}^{n \times n}$ , we use a *sketch*  $A\Omega$  where  $\Omega$  is a suitable random matrix with  $m \ll n$  columns; an advantage of this representation is that if  $A$  is updated by a low-rank matrix  $B$ , the sketch can be updated by multiplying  $B$  with  $\Omega$  without accessing  $A$  again. The goal of this project is to implement and analyze a sketch-based algorithm for low-rank approximation, taken from [2].

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## ► Tasks

1. Given a matrix  $A \in \mathbb{R}^{n \times m}$  and integers  $r$  and  $\ell \geq 0$ , the following algorithm computes a rank- $r$  approximation  $A \approx BC^T$  for  $B \in \mathbb{R}^{n \times r}$  and  $C \in \mathbb{R}^{m \times r}$ .

**Algorithm 1**


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1: procedure SKETCHING( $A, r, \ell$ )
2:   Draw Gaussian random matrices  $X \in \mathbb{R}^{m \times r}$  and  $Y \in \mathbb{R}^{n \times (r+\ell)}$ 
3:   Compute the products  $AX$  and  $Y^T A$ 
4:   Compute an economy QR factorization  $QR = Y^T AX$ 
5:   Set  $B \leftarrow AXR^\dagger$  and  $C \leftarrow A^T YQ$ 
6: end procedure

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$R^\dagger$  denotes the Moore-Penrose pseudo-inverse of the matrix  $R$ .

Implement this algorithm. What is the asymptotic time complexity of the algorithm in function of  $n, m, r, \ell$  and  $\text{nnz}(A)$ ?

2. Prove that if  $A$  has rank exactly  $r$ , then  $\text{SKETCHING}(A, r, 0)$  returns an exact low-rank factorization of  $A$  with probability 1.

*Hint:* A square Gaussian random matrix is invertible with probability 1.

3. Find and correct the mistake in [2, Lemma 3.1].
4. Consider the three following  $200 \times 200$  matrices:

- The Hilbert matrix  $A_1(i, j) = \frac{1}{i+j-1}$ ;
- A matrix  $A_2$  constructed in the following way:

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d = 0.8.^ (1:n);
[Q, ~] = qr(randn(n));
[V, ~] = qr(randn(n));
A = Q*diag(d)*V;

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- A matrix  $A_3$  constructed similarly as before, with  $d = (1:n).^(-1.5)$ .

Plot the error of the low-rank approximations that you obtain with ranks  $1, \dots, 100$ , for  $\ell = 0, \ell = 3, \ell = \text{floor}(r/2)$ . What do you notice?

5. Explore the effect of randomness in the algorithm: For a fixed instance of matrix  $A_3$ , for  $r = 1, \dots, 100$ , apply the algorithm 1000 times and plot the average error and the 95% confidence interval, in the case  $\ell = 3$  and then in the case  $\ell = \text{floor}(r/2)$ .

## ► References

- [1] Woodruff, David P. Sketching as a tool for numerical linear algebra. arXiv preprint arXiv:1411.4357 (2014). <https://arxiv.org/pdf/1411.4357>
- [2] Nakatsukasa, Yuji. Fast and stable randomized low-rank matrix approximation. arXiv preprint arXiv:2009.11392 (2020). <https://arxiv.org/pdf/2009.11392>