Basic linear algebra recap. Convergence rates. Line Search

Seminar

Optimization for ML. Faculty of Computer Science. HSE University





• Naive matmul $\mathcal{O}(n^3)$, naive matvec $\mathcal{O}(n^2)$



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- All matrices have SVD

$$A = U\Sigma V^T$$

Lecture reminder

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- $\bullet \ \operatorname{tr}(ABCD) = \operatorname{tr}(DABC) = \operatorname{tr}(CDAB) = \operatorname{tr}(BCDA) \ \text{for any matrices ABCD if the multiplication defined}.$
- $\langle \hat{A}, B \rangle = \operatorname{tr}(A^T \hat{B})$

 $f \to \min_{x,y,z}$

Convergence rate

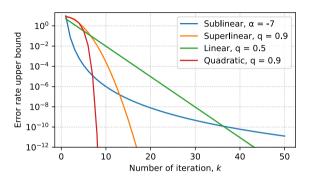


Figure 1: Illustration of different convergence rates

• Linear (geometricm, exponential) convergence:

$$r_k \le Cq^k, \quad 0 < q < 1, C > 0$$

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Convergence rate

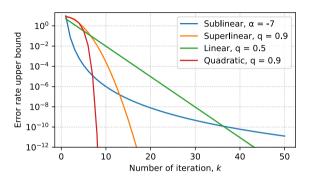


Figure 1: Illustration of different convergence rates

• Linear (geometricm, exponential) convergence:

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 Any convergent sequence, that is slower (faster) than any linearly convergent sequence has sublinear (superlinear) convergence

 $f \to \min_{x \in X}$

Lecture reminder

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Let $\{r_k\}_{k=m}^{\infty}$ be a sequence of non-negative numbers, converging to zero, and let

$$q = \lim_{k \to \infty} \sup_{k} \ r_k^{1/k}$$

• If $0 \le q < 1$, then $\{r_k\}_{k=m}^{\infty}$ has linear convergence with constant q.

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- The case q > 1 is impossible.

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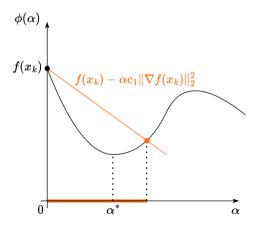
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- If q does not exist, but $q=\lim_{k\to\infty}\sup_k\frac{r_{k+1}}{r_k}<1$, then $\{r_k\}_{k=m}^\infty$ has linear convergence with a constant not exceeding q.
- If $\lim_{k\to\infty}\inf_k\frac{r_{k+1}}{r_k}=1$, then $\{r_k\}_{k=m}^\infty$ has sublinear convergence.
- The case $\lim_{k \to \infty} \inf_k \frac{r_{k+1}}{r_k} > 1$ is impossible.
- In all other cases (i.e., when $\lim_{k\to\infty}\inf_k\frac{r_{k+1}}{r_k}<1\leq\lim_{k\to\infty}\sup_k\frac{r_{k+1}}{r_k}$) we cannot claim anything concrete about the convergence rate $\{r_k\}_{k=m}^\infty$.

Line search

Typical line search problem is finding the good value α of the stepsize:

$$x_{k+1} = x_k - \alpha \nabla f(x_k)$$



Lecture reminder

• Solution localization methods:

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- Solution localization methods:
 - Dichotomy search method

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 - Curvature conditions
 - The idea behind backtracking line search

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Suppose, you have the following expression

$$b = A_1 A_2 A_3 x,$$

where the $A_1,A_2,A_3\in\mathbb{R}^{3\times 3}$ - random square dense matrices and $x\in\mathbb{R}^n$ - vector. You need to compute b.

Which one way is the best to do it?

1. $A_1A_2A_3x$ (from left to right)

Check the simple **\$\rightarrow\$**code snippet after all.

Problems

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- 1. $A_1A_2A_3x$ (from left to right)
- 2. $(A_1(A_2(A_3x)))$ (from right to left)

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- 3. It does not matter

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- 2. $(A_1(A_2(A_3x)))$ (from right to left)
- 3. It does not matter
- 4. The results of the first two options will not be the same.

Check the simple **code** snippet after all.



Problem 2. Connection between Frobenius norm and singular values.

Let $A \in \mathbb{R}^{m \times n}$, and let $q := \min\{m, n\}$. Show that

$$||A||_F^2 = \sum_{i=1}^q \sigma_i^2(A),$$

where $\sigma_1(A) \ge \ldots \ge \sigma_q(A) \ge 0$ are the singular values of matrix A. Hint: use the connection between Frobenius norm and scalar product and SVD.

Problem 3. Known your inner product.

Simplify the following expression:

$$\sum_{i=1}^{n} \langle S^{-1} a_i, a_i \rangle,$$

where $S = \sum_{i=1}^n a_i a_i^T, a_i \in \mathbb{R}^n, \det(S) \neq 0$

•
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- $r_k = 0.707^k$ $r_k = 0.707^{2^k}$



Problem 5. One test is simpler, than another.

$$r_k = \frac{1}{k^k}$$

Problem 6. Quadratic convergence.

Show, that the following sequence does not have a quadratic convergence.

$$r_k = \frac{1}{3^{k^2}}$$

