

Lecture 8

Office Hours!

Final Exam Date/Time.

12/17 10:30 - 12:30

(See Ed Stem)

Coord Descent = Gradient Descent? $\Rightarrow$ No!

Steepest Descent vs. Gradient Descent!

Does Gradient Descent $\Rightarrow$ following gradient?

$\Rightarrow$ Basically, yes, Technically, Can go to the a better "Relaxed" step than gradient?

Richardson Relx to Pass?

Is Richardson used? vs. Steepest Descent?

Matrix Norms vs. Vector Norms

Vector Norms $\hookrightarrow$ Vector Norm of Each row/col?

When to pick Coord descent vs. gradient descent?

How to compute the rank of a matrix/ Sparse matrix efficiently?

$p = \infty$ is $\checkmark$

$\|x\|_{\infty}$ vs. $\max_{i=1}^n |x_i|$

Properties of the Rayleigh quotient $\frac{x^T A x}{x^T x}$

How to pick Coordinate j in Coord descent?

Gradient Descent vs. Steepest Descent.

Steepest Descent is an instance of gradient descent where

$f(x) = \frac{1}{2} x^T A x - x^T b$

$g(x) = A x - b$

AND $x_{k+1} = x_k - \alpha_k g(x_k)$

α_k chosen to minimize

$f(x_k)$

Gradient Descent Any alg when

$x_{k+1} = x_k - \alpha g_k$

when $g_k \approx g(x_k)$

$f(x) = \frac{1}{2} x^T A x - x^T b \rightarrow$ as usual

$g(x) =$ 1st order approx $\Rightarrow$ a vector

1st-order approx $\Rightarrow$ $g(x) = A x - b$

2nd order approx.

$f(x+p) = f(x) + p^T g(x) + \frac{1}{2} p^T H(x) p + O(\|p\|^3)$

Using 2nd order ideas $\Rightarrow$

Solve $A x = b$ in go $\forall x = A^{-1} b$

New Line method.

Given $x_0 = \text{arb.}$

Find p s.t.

$x_0 + p = A^{-1} b$

if $x_k \neq x^*$ Iterative Refinement

Then $x_k - x^* = A^{-1} (A x_k - b)$

Let $\tilde{A} = A$ Then we are solving $\tilde{A} p = A x_k - b$

Residual.

Coord Descent aka Coordinate Relaxation!

$\min_x f(x) = \frac{1}{2} x^T A x - x^T b$

$\min_{\alpha} f(x_k + \alpha e_j) =$

$x_{k+1} = x_k + \alpha e_j$

Coord descent for all coords 1 to $n \approx$ Work for 1 step of gradient descent.

What if?

- Cyclic / Separation

- greedy

- random

- queue / optimal cycle

Richardson vs. Steepest Descent

Yes! Richardson doesn't need symmetry!

How to compute rank efficiently?

1. What is "rank" computationally?

2. Rank(A) = # of Sing. values $\geq \sigma_{\max} \epsilon$ when $\epsilon =$ floating point Round-off.

How to compute rank efficiently?

1. What is "rank" computationally?

2. Rank(A) = # of Sing. values $\geq \sigma_{\max} \epsilon$ when $\epsilon =$ floating point Round-off.

$\max_{x \neq 0} \frac{x^T A x}{x^T x} = \lambda_{\max}$

if $A \succ 0, A = A^T$

$\min_{x \neq 0} \frac{x^T A x}{x^T x} = \lambda_{\min}$

if $A \succ 0, A = A^T$

if you fix $\|x\|_2 = 1$, look at $x^T A x$

$\lambda_{\max}$