

Numeric and Optimize Method

1. Matrix Decomposition/ Factorization

reference:

- wiki

1.1. LU

reference:

- wiki

formulation:

$$A = LU \quad \langle 1 \rangle$$

- L: lower triangular
- U: upper triangular

1.1.1. Partial Pivoting

formulation:

$$PA = LU \quad \langle 2 \rangle$$

- P: permutation that reorder rows

feature:

- numerically stable

1.1.2. Full Pivoting

formulation:

$$PAQ = LU \quad \langle 3 \rangle$$

- Q : permutation that reorder columns

1.1.3. LDU

+formulation:

$$A = LDU \quad \langle 4 \rangle$$

- D : diagonal
- L, U : unitraingular

1.2. Cholesky

formulation:

$$A = UU^T \quad \langle 5 \rangle$$

- A : symmetric, positive (semi-)defined
- U : upper traingular, (semi-)positive diagonal entries

1.2.1. LDL/LDLT

+formulation:

$$A = LDL^T \quad \langle 6 \rangle$$

- L : lower unitraingular
- D : diagonal

feature:

- +square-root-free

1.3. QR

formulation:

$$A = QR \quad \langle 7 \rangle$$

- Q: orthogonal
- R: upper triangular

compute:

1.3.1. Gram-Schmidt Process

feature:

- low numeric stability
- easy implementation

1.3.2. Householder Reflections

feature

- better numeric stability than Gram-Schmidt Process
- bandwidth heavy
- not parallelizable

1.3.3. Givens Rotations

feature:

- sparse
- parallelizable

2. Equaltion

2.1. Linear

formulation:

$$Ax + b = 0 \qquad \langle 8 \rangle$$

transform:

if $\text{rank}(A) < b$

$$A^{\top}Ax + A^{\top}b = 0 \quad \langle 9 \rangle$$

$$A'x + b' = 0 \quad \langle 10 \rangle$$

$$\rightarrow \langle 8 \rangle$$

2.1.1. Conjugate Gradient

reference:

- wiki
- cornell.edu

2.2. Continuous

formulation:

$$f(x) + b = 0 \quad \langle 11 \rangle$$

2.2.1. Newton

transform:

$$\nabla_{x_k} f(x_k)(x_{k+1} - x_k) + f(x_k) + b = 0 \quad \langle 12 \rangle$$

$$\rightarrow \langle 8 \rangle$$

3. Optimization

formulation:

$$\min_x f(x) \quad \langle 13 \rangle$$

3.1. Quadral

formulation:

$$\min_x \mathbf{x}^\top \mathbf{A} \mathbf{x} + \mathbf{b} \mathbf{x} \quad <14>$$

or

$$\min_x \| \mathbf{A} \mathbf{x} + \mathbf{b} \|_2 \quad <15>$$

transform:

$$\begin{aligned} \mathbf{A} \mathbf{x} + \mathbf{b} &= 0 \\ \rightarrow <8> \end{aligned} \quad <16>$$

or

$$2\mathbf{A}^\top (\mathbf{A} \mathbf{x} + \mathbf{b}) = 0 \quad <17>$$

$$\mathbf{A}^\top \mathbf{A} \mathbf{x} + \mathbf{A}^\top \mathbf{b} = 0 \quad <18>$$

$$\begin{aligned} \mathbf{A}' \mathbf{x} + \mathbf{b}' &= 0 \\ \rightarrow <8> \end{aligned} \quad <19>$$

3.2. Continuous

formulation:

$$\begin{aligned} \min_x f(\mathbf{x}), \\ \nabla_x f(\mathbf{x}_k) = \mathbf{g}, \nabla_x^2 f(\mathbf{x}_k) = \mathbf{H} \end{aligned} \quad <20>$$

3.2.1. Newton

transform:

$$f(\mathbf{x})_k = \mathbf{x}^\top H \mathbf{x}_k + \mathbf{g} \mathbf{x}_k,$$

$$\min_{\mathbf{x}_{k+1}} f(\mathbf{x})_k \quad \langle 21 \rangle$$

$$\rightarrow \langle 15 \rangle$$

3.2.2. Quasi-Newton

reference:

- wiki

3.2.2.1. BFGS

reference:

- wiki

3.2.2.1.1. L-BFGS

reference:

- wiki

3.2.2.2. Compact Representation

reference:

- wiki

4. Constraint

4.1. Equality

formulation:

$$\mathbf{A} \mathbf{x} = \mathbf{b} \quad \langle 22 \rangle$$

kkt:

$$\begin{aligned}\nabla_x f(x) + \nabla_x n(x)\lambda &= 0, \\ n(x) &= 0\end{aligned}\tag{23}$$

4.1.1. Linear Equality

formulation:

$$\begin{aligned}Nx + m &= 0, \\ \text{rank}(N^\top) &< \dim(x)\end{aligned}\tag{24}$$

or

$$\begin{aligned}x &= N\lambda, \\ \text{rank}(N) &< \dim(x)\end{aligned}\tag{25}$$

4.1.1.1. Qaldral Optimization (Linear Least Squares)

reference:

- [wiki](#)

transform:

$$\begin{aligned}\langle 14 \rangle, \langle 24 \rangle, \langle 23 \rangle &\Rightarrow \\ N\lambda + 2Ax + b &= 0\end{aligned}\tag{26}$$

$$\begin{aligned}\langle 24 \rangle, \langle 26 \rangle &\Leftrightarrow \\ \begin{pmatrix} 2A & N \\ N & 0 \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix} + \begin{pmatrix} b \\ m \end{pmatrix} &= 0\end{aligned}\tag{27}$$

$$\begin{aligned}A'x' + b' &= 0 \\ &\rightarrow \langle 8 \rangle\end{aligned}\tag{28}$$

or

$$\begin{aligned} & \langle 15 \rangle, \langle 25 \rangle \Rightarrow \\ \min_{\lambda} & \| A N \lambda - b \|_2 \end{aligned} \quad \langle 29 \rangle$$

$$\begin{aligned} \min_{x'} & \| A' x' - b \|_2 \\ & \rightarrow \langle 14 \rangle \end{aligned} \quad \langle 30 \rangle$$

4.2. InEquality

formulation:

$$n(x) \geq 0 \quad \langle 31 \rangle$$

kkt:

$$\begin{aligned} \nabla_x f(x) + \nabla_x n(x) \lambda &= 0, \\ \lambda n(x) &= 0, \\ \lambda &\geq 0 \end{aligned} \quad \langle 32 \rangle$$

4.2.1. Interior Point (Barrier)

reference:

- wiki

4.2.1.1. Path Following

transform:

$$\begin{aligned} \min_{x_{t_k}=x} & f(x) + t_k n'(x), t_{k+1} = \theta t_k \\ & \rightarrow \langle 13 \rangle \end{aligned} \quad \langle 33 \rangle$$

4.2.1.2. Primal Dual

reference:

- cmu.edu

compare:

- little slower than Path Following but more accurate.

transform:

$$\begin{aligned} & \langle 33 \rangle, \\ n'(x) &:= \| -\log(n(x)) \|_1 \end{aligned} \quad \langle 34 \rangle$$

$$\min_{x_{t_k}=x} f(x) + t_k \| -\log(n(x)) \|_1, \quad \langle 35 \rangle$$

$$\lambda \odot n(x) = t_k \mathbf{1}$$

$$\begin{aligned} & \langle 35 \rangle \Rightarrow \\ \nabla_x f(x) + t_k \nabla_x n(x) (-n(x)^{\odot -1}) &= 0 \end{aligned} \quad \langle 36 \rangle$$

$$\nabla_x f(x) + t_k \nabla_x n(x) (-t_k^{-1} \lambda) = 0 \quad \langle 37 \rangle$$

$$\nabla_x f(x) - \nabla_x n(x) \lambda = 0 \quad \langle 38 \rangle$$

$$\begin{aligned} & \langle 38 \rangle, \langle 35 \rangle \Rightarrow \\ \nabla_x f(x) - \nabla_x n(x) \lambda &= 0, \end{aligned} \quad \langle 39 \rangle$$

$$\lambda \odot n(x) = t_k \mathbf{1}$$

$$\rightarrow \langle 11 \rangle$$

or

$$\begin{aligned}
&\langle 32 \rangle \rightarrow \\
&\mathbf{x}_k := \mathbf{x}, \\
&\nabla_{\mathbf{x}} f(\mathbf{x}) + \nabla_{\mathbf{x}} \mathbf{n}(\mathbf{x}) \boldsymbol{\lambda} = 0, \\
&\boldsymbol{\lambda} \mathbf{n}(\mathbf{x}) = t_k, \\
&t_{k+1} = \theta t_k \\
&\rightarrow \langle 11 \rangle
\end{aligned}
\tag{40}$$

4.2.2. Linear InEquality

formulation:

$$N\mathbf{x} + \mathbf{m} \geq 0 \tag{41}$$