

Formulas

- Inner product, norm, angle.
 - Relation between inner product, norm, and angle: $a^T b = \|a\| \|b\| \cos \angle(a, b)$.
 - Average value of elements of an n -vector: $\text{avg}(a) = (\mathbf{1}^T a)/n$.
 - Root-mean-square value of an n -vector: $\text{rms}(a) = \|a\|/\sqrt{n}$.
 - Standard deviation of an n -vector: $\text{std}(a) = \|a - \text{avg}(a)\mathbf{1}\|/\sqrt{n}$.
 - Correlation coefficient of two non-constant n -vectors a and b :

$$\rho = \frac{(a - \text{avg}(a)\mathbf{1})^T (b - \text{avg}(b)\mathbf{1})}{\|a - \text{avg}(a)\mathbf{1}\| \|b - \text{avg}(b)\mathbf{1}\|}$$

- Complexity of basic matrix and vector operations (α is a scalar, x and y are n -vectors, A is an $m \times n$ matrix, B is an $n \times p$ matrix).
 - Inner product $x^T y$: $2n - 1$ flops ($\approx 2n$ flops for large n).
 - Vector addition $x + y$: n flops.
 - Scalar-vector multiplication αx : n flops.
 - Scalar-matrix multiplication αA : mn flops.
 - Matrix-vector multiplication Ax : $m(2n - 1)$ flops ($\approx 2mn$ flops for large n).
 - Matrix-matrix multiplication AB : $mp(2n - 1)$ flops ($\approx 2mpn$ flops for large n).
- Pseudo-inverses.
 - Pseudo-inverse of left-invertible matrix A : $A^\dagger = (A^T A)^{-1} A^T$.
 - Pseudo-inverse of right-invertible matrix A : $A^\dagger = A^T (A A^T)^{-1}$.
- Complexity of forward or back substitution with triangular $n \times n$ matrix: n^2 flops.
- Complexity of matrix factorizations.
 - QR factorization of $m \times n$ matrix: $2mn^2$ flops.
 - LU factorization of $n \times n$ matrix: $(2/3)n^3$ flops.

Problem 1. Show that every 2×2 rotation matrix can be written as a product of two reflectors: for every θ , there exist u, v such that

$$\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} = (I - 2uu^T)(I - 2vv^T), \quad \|u\| = \|v\| = 1.$$

Without loss of generality, one can choose $v = (1, 0)$, so it remains to find u . You may find the trigonometric identities $\cos \theta = \cos^2(\theta/2) - \sin^2(\theta/2)$ and $\sin \theta = 2 \cos(\theta/2) \sin(\theta/2)$ useful.

Answer for problem 1.

$$\text{Let } v = (1, 0), \quad vv^T = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \quad (u_1 \ u_2) \begin{pmatrix} 1 \\ 0 \end{pmatrix} = u_1$$

$$(I - 2uu^T)(I - 2vv^T) = I - 2vv^T - 2uu^T + 4uu^Tvv^T$$

$$= I - 2vv^T - 2uu^T + 4(u^T v)uv^T \quad \text{Let } u = \begin{pmatrix} u_1 \\ u_2 \end{pmatrix}$$

$$= \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} - 2uu^T + 4u_1uv^T$$

$$u^T v = u_1$$

$$= \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} - 2 \begin{pmatrix} u_1^2 & u_1 u_2 \\ u_1 u_2 & u_2^2 \end{pmatrix} + 4u_1 \begin{pmatrix} u_1 & 0 \\ u_2 & 0 \end{pmatrix} \begin{pmatrix} u_1 & u_2 \\ u_2 & 0 \end{pmatrix}$$

$$\begin{pmatrix} u_1 \\ u_2 \end{pmatrix} \begin{pmatrix} 1 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} - 2 \begin{pmatrix} u_1^2 & u_1 u_2 \\ u_1 u_2 & u_2^2 \end{pmatrix} + 4 \begin{pmatrix} u_1^2 & 0 \\ u_1 u_2 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} + \begin{pmatrix} 2u_1^2 & -2u_1 u_2 \\ 2u_1 u_2 & -2u_2^2 \end{pmatrix}$$

$$= \begin{pmatrix} -1 + 2u_1^2 & -2u_1 u_2 \\ 2u_1 u_2 & 1 - 2u_2^2 \end{pmatrix}$$

$$\begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} = \begin{pmatrix} \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2} & -2 \cos \frac{\theta}{2} \sin \frac{\theta}{2} \\ 2 \cos \frac{\theta}{2} \sin \frac{\theta}{2} & \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2} \end{pmatrix}$$

$$\text{Let } u_1 = \cos \frac{\theta}{2}, \quad u_2 = \sin \frac{\theta}{2}$$

$$-1 + 2u_1^2 = -1 + 2 \cos^2 \frac{\theta}{2} = -(\cos^2 \frac{\theta}{2} + \sin^2 \frac{\theta}{2}) + 2 \cos^2 \frac{\theta}{2}$$

$$= \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2}$$

$$1 - 2u_2^2 = 1 - 2 \sin^2 \frac{\theta}{2} = (\cos^2 \frac{\theta}{2} + \sin^2 \frac{\theta}{2}) - 2 \sin^2 \frac{\theta}{2}$$

$$= \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2}$$

With $v = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ and $u = \begin{pmatrix} \cos(\theta/2) \\ \sin(\theta/2) \end{pmatrix}$, we get

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$$(I - 2uu^T)(I - 2vv^T) = \begin{pmatrix} \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2} & -2 \cos \frac{\theta}{2} \sin \frac{\theta}{2} \\ 2 \cos \frac{\theta}{2} \sin \frac{\theta}{2} & \cos^2 \frac{\theta}{2} - \sin^2 \frac{\theta}{2} \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

Problem 2. Consider the $n \times n$ matrix $A = aI - \mathbf{1}\mathbf{1}^T$ where a is a real scalar and $\mathbf{1}$ is the n -vector of ones. We assume $n > 1$.

1. For what values of a is A singular?

2. Assuming A is nonsingular, express its inverse as a linear combination of I and $\mathbf{1}\mathbf{1}^T$.

$$\mathbf{1}\mathbf{1}^T = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \begin{pmatrix} 1 & \dots & 1 \end{pmatrix} = \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

Answer for problem 2.

$$A = \begin{pmatrix} a & 0 & \dots & 0 \\ 0 & a & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & a \end{pmatrix} - \begin{pmatrix} 1 & \dots & 1 \\ \vdots & & \vdots \\ 1 & \dots & 1 \end{pmatrix} = \begin{pmatrix} a-1 & -1 & \dots & -1 \\ -1 & a-1 & \dots & -1 \\ \vdots & \vdots & \ddots & \vdots \\ -1 & \dots & -1 & a-1 \end{pmatrix}$$

1) A singular when $Ax = 0$ for $x \neq 0$

$$Ax = a x - \mathbf{1}\mathbf{1}^T x = 0$$

$$a x = \mathbf{1}\mathbf{1}^T x$$

$$\mathbf{1}\mathbf{1}^T x = (\mathbf{1}^T x) \mathbf{1} = (x_1 + \dots + x_n) \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} a x_1 \\ \vdots \\ a x_n \end{pmatrix} = \begin{pmatrix} x_1 + \dots + x_n \\ \vdots \\ x_1 + \dots + x_n \end{pmatrix}$$

If $a=0$, $x = (1, 0, \dots, 0, -1)$ is a solution.

$a x_1 = a x_2 = \dots = a x_n = x_1 + \dots + x_n$ only satisfied if $a=0$ or $x_1 = \dots = x_n$

If $x_1 = x_2 = \dots = x_n$, then $a=n$ solves the equation.

A is singular when $a=0$, or $a=n$



$$2) A^{-1} = cI + d11^T$$

$$A^{-1}A = (cI + d11^T)(aI - 11^T) = I$$

$$= acI + ad11^T - c11^T - d(11^T)(11^T) = I$$

$$= acI + (ad - c)11^T - d11^T11^T \quad 11^T11^T = \begin{pmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{pmatrix} \begin{pmatrix} 1 & \dots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \dots & 1 \end{pmatrix}$$

$$= \begin{pmatrix} n & \dots & n \\ \vdots & \ddots & \vdots \\ n & \dots & n \end{pmatrix}$$

$$= n11^T$$

$$A^{-1}A = acI + (ad - c - nd)11^T = I$$

$$ac = 1 \Rightarrow c = \frac{1}{a}$$

$$ad - c - nd = 0$$

$$ad - \frac{1}{a} - nd = 0$$

$$(a-n)d = \frac{1}{a} \Rightarrow d = \frac{1}{a(a-n)}$$

$$A^{-1} = \frac{1}{a}I + \frac{1}{a(a-n)}11^T$$



Problem 3. Consider an $m \times m$ block matrix A of the form

$$A = \begin{bmatrix} A_{11} & 0 & 0 & \cdots & 0 & 0 \\ 0 & A_{22} & 0 & \cdots & 0 & 0 \\ 0 & 0 & A_{33} & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & A_{m-1,m-1} & 0 \\ A_{m1} & A_{m2} & A_{m3} & \cdots & A_{m,m-1} & A_{mm} \end{bmatrix}$$

Each block A_{ij} and all the zero matrices in the definition are matrices of size $n \times n$, so the matrix A itself has size $nm \times nm$. What is the complexity of solving $Ax = b$ in each of the following cases?

1. The matrices A_{ii} are lower triangular and nonsingular.
2. The matrices A_{ii} are orthogonal.
3. The matrices A_{ii} are general nonsingular matrices.

Include in the complexity all terms that are cubic (n^3 , n^2m , nm^2 , m^3) or higher-order. Explain your answers.

Answer for problem 3.

$$Ax = b \quad Ax = \begin{pmatrix} A_{11} & 0 & \cdots & 0 \\ 0 & A_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & A_{m-1,m-1} \\ A_{m1} & \cdots & \cdots & A_{mm} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_{m-1} \\ x_m \end{pmatrix} \quad \text{each } x_i \text{ is an } n\text{-vector.}$$

$$= \begin{pmatrix} A_{11}x_1 \\ A_{22}x_2 \\ \vdots \\ A_{m-1,m-1}x_{m-1} \\ A_{m1}x_1 + \cdots + A_{mm}x_m \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix} \quad \text{Each } b_i \text{ is an } n\text{-vector.}$$

1) $A_{11}x_1 = b_1, \dots, A_{m-1,m-1}x_{m-1} = b_{m-1}$ can be solved by forward substitution

$$\rightarrow (m-1)(n^2) = mn^2 - n^2 \text{ complexity}$$

Then x_1, \dots, x_{m-1} are given, so $A_{m1}x_1 + \dots + A_{m,m-1}x_{m-1}$ is matrix-vector mult. and addition

$$\rightarrow \underbrace{(m-1) \cdot (2n^2)}_{\text{mult}} + \underbrace{(m-2)n}_{\text{add}} = 2mn^2 - 2n^2 + mn - 2n \text{ complexity}$$

Then $A_{mm}x_m = b_m - \sum_{i=1}^{m-1} A_{mi}x_i$ can be solved by forward substitution

$$\rightarrow n^2 \text{ complexity.} \\ + n \text{ vector add}$$

2

Total complexity: $3mn^2$

2) For $1 \leq i \leq m-1$, we solve $A_{ii} x_i = b_i$ by computing $x_i = A_{ii}^{-1} b_i$ as A_{ii} is orthogonal.

$$\rightarrow (m-1)(2n^2) = 2mn^2 - 2n^2 \text{ complexity}$$

Then x_1, \dots, x_{m-1} are given, so $c = b_m - \sum_{i=1}^{m-1} A_{mi} x_i$ is $m-1$ matrix-vector multiplications and $m-1$ vector adds.

$$\rightarrow (m-1)(2n^2) + (m-1)(n) = 2mn^2 - 2n^2 + mn - n \text{ complexity}$$

Then solve $A_{mm} x_m = c$ by computing $x_m = A_{mm}^{-1} c$

$$\rightarrow 2n^2 \text{ complexity.}$$

$$\text{Total complexity: } 4mn^2 \quad \text{2}$$

3) Step 1: Factorize $A_{ii} = P_i L_i V_i$ for $1 \leq i \leq m$

$$\rightarrow m \cdot \frac{2}{3} n^3 = \frac{2}{3} mn^3$$

Step 2: For $1 \leq i \leq m-1$, solve $A_{ii} x_i = b_i$ by $P_i L_i V_i x_i = b_i$

a) Solve $P_i y_i = b_i \rightarrow 0$ flops (permutation only)

b) Solve $L_i z_i = y_i \rightarrow n^2$ flops (forward sub)

c) Solve $V_i x_i = z_i \rightarrow n^2$ flops (back sub)

$$\rightarrow (m-1)(2n^2) = 2mn^2 - 2n^2 \text{ flops}$$

Step 3: Compute $c = b_m - \sum_{i=1}^{m-1} A_{mi} x_i$

$$\rightarrow (m-1)(2n^2) + (m-1)(n) = 2mn^2 - 2n^2 + mn - n$$

Step 4: Solve $A_{mm} x_m = c$ by $P_m L_m V_m x_m = c$

a) Solve $P_m y_m = c \rightarrow 0$ flops (permutation)

b) Solve $L_m z_m = y_m \rightarrow n^2$ flops (forward sub)

c) Solve $V_m x_m = z_m \rightarrow n^2$ flops (back sub)

$$\rightarrow 2n^2 \text{ flops}$$

$$\text{Total complexity: } \frac{2}{3} mn^3 + 4mn^2 \quad \text{2}$$

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Problem 4. Let A be a nonsingular $n \times n$ matrix. Explain how each of the following problems can be solved with a single LU factorization of A . Describe the steps in your algorithms and give the complexity of each step, including all quadratic and cubic terms in the flop count. If you know several methods, give the most efficient one.

1. Find the $n \times n$ matrix X that satisfies

$$AXA = uv^T$$

where u and v are given n -vectors.

2. Find the n -vectors x, y that satisfy

$$\begin{bmatrix} 0 & A \\ A^T & B \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} b \\ c \end{bmatrix}.$$

The $n \times n$ matrix B , and the n -vectors b, c are given.

3. Find the n -vectors x, y that satisfy

$$\begin{bmatrix} A & A \\ -A & A \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} b \\ c \end{bmatrix}.$$

The n -vectors b, c are given.

Answer for problem 4.

1) $AXA = uv^T$ $X = A^{-1}u v^T A^{-1}$ Let $y = A^{-1}u$ $X = y v^T A^{-1}$
 $X^T = (A^T)^{-1} v y^T$ Let $z = (A^T)^{-1} v$
 $= z y^T$
 $X = y z^T$

Step 1: Factorize $A = PLU \rightarrow \frac{2}{3}n^3$ flops
 2: Solve $PLUy = u \rightarrow 2n^2$ flops
 a) Solve $Py_1 = u \rightarrow 0$ flops (permutation)
 b) Solve $Ly_2 = y_1 \rightarrow n^2$ flops (fwd sub)
 c) Solve $Uy = y_2 \rightarrow n^2$ flops (back sub)
 3: Solve $V^T L^T P^T z = v \rightarrow 2n^2$ flops
 a) Solve $V^T z_1 = v \rightarrow n^2$ flops (fwd sub)
 b) Solve $L^T z_2 = z_1 \rightarrow n^2$ flops (back sub)
 c) Solve $P^T z = z_2 \rightarrow 0$ flops (permutation)
 4: Compute $X = y z^T \rightarrow n^2$ flops

Total complexity: $\frac{2}{3}n^3 + 5n^2$ flops

$$2) \begin{pmatrix} 0 & A \\ A^T & B \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} b \\ c \end{pmatrix} \rightarrow \begin{aligned} Ay &= b \\ A^T x + By &= c \end{aligned}$$

Step 1: Factorize $A = PLU \rightarrow \frac{2}{3}n^3$ flops
 $A^T = U^T L^T P^T$

- 2: Solve $PLUy = b \rightarrow 2n^2$ flops
 a) Solve $Py_1 = b \rightarrow 0$ flops (permutation)
 b) Solve $Ly_2 = y_1 \rightarrow n^2$ flops (forward sub)
 c) Solve $Uy = y_2 \rightarrow n^2$ flops (back sub).

3) Compute $By = z \rightarrow 2n^2$ flops

4) Compute $d = c - z \rightarrow n$ flops

5) Solve $U^T L^T P^T x = d \rightarrow 2n^2$ flops

a) Solve $U^T x_1 = d \rightarrow n^2$ flops (forward sub)

b) Solve $L^T x_2 = x_1 \rightarrow n^2$ flops (back sub)

c) Solve $P^T x = x_2 \rightarrow 0$ flops (permutation)

Total complexity: $\frac{2}{3}n^3 + 6n^2$ flops

$$3) \begin{pmatrix} A & A \\ -A & A \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} b \\ c \end{pmatrix} \quad \begin{aligned} Ax + Ay &= b & A(x+y) &= b & 2Ay &= b+c \\ -Ax + Ay &= c & A(-x+y) &= c & 2Ax &= b-c \end{aligned}$$

Step 1: Factorize $A = PLU \rightarrow \frac{2}{3}n^3$ flops

2: Compute $d = \frac{1}{2}(b+c) \rightarrow 2n$ flops

3: Compute $f = \frac{1}{2}(b-c) \rightarrow 2n$ flops

4: Solve $PLUy = d \rightarrow 2n^2$ flops

a) Solve $Py_1 = d \rightarrow 0$ flops (permutation)

b) Solve $Ly_2 = y_1 \rightarrow n^2$ flops (forward sub)

c) Solve $Uy = y_2 \rightarrow n^2$ flops (back sub)

5: Solve $PLUx = f \rightarrow 2n^2$ flops

a) Solve $Px_1 = f \rightarrow 0$ flops (permutation)

b) Solve $Lx_2 = x_1 \rightarrow n^2$ flops (forward sub)

c) Solve $Ux = x_2 \rightarrow n^2$ flops (back sub)

Total complexity: $\frac{2}{3}n^3 + 4n$
 flops



Problem 5. Suppose A and B are left-invertible matrices of the same size, that satisfy

$$AA^T = BB^T.$$

In this problem we show that $B = AQ$ for some orthogonal Q .

1. Show that the matrix $U = A^\dagger B$ is orthogonal.
2. Show that the matrix $V = B^\dagger A$ is orthogonal.
3. Show that U is the inverse of V .
4. Find an orthogonal matrix Q such that $B = AQ$.

Answer for problem 5.

$$1) A^\dagger = (A^T A)^{-1} A^T$$

$$V = (A^T A)^{-1} A^T B \quad V^T = B^T A (A^T A)^{-T} = B^T A (A^T A)^{-1}$$

$$VV^T = ((A^T A)^{-1} A^T B) (B^T A (A^T A)^{-1})$$

$$= (A^T A)^{-1} A^T B B^T A (A^T A)^{-1}$$

$$= (A^T A)^{-1} A^T A A^T A (A^T A)^{-1}$$

$$= (A^T A)^{-1} (A^T A) (A^T A) (A^T A)^{-1}$$

$$\boxed{VV^T = I} \text{ so } V \text{ is orthogonal. } \checkmark$$

$$(A^T A)^{-T} = ((A^T A)^T)^{-1} = (A^T A)^{-1}$$

$$2) B^\dagger = (B^T B)^{-1} B^T$$

$$V = (B^T B)^{-1} B^T A \quad V^T = A^T B (B^T B)^{-T} = A^T B (B^T B)^{-1}$$

$$VV^T = ((B^T B)^{-1} B^T A) (A^T B (B^T B)^{-1})$$

$$= (B^T B)^{-1} B^T A A^T B (B^T B)^{-1}$$

$$= (B^T B)^{-1} B^T B B^T B (B^T B)^{-1}$$

$$\boxed{VV^T = I} \text{ so } V \text{ is orthogonal. } \checkmark$$

3) Show $VV = I$ ✓

$$VV = (A+B)(B^+A) = (A^+A)^{-1}A^+B(B^+B)^{-1}B^+A$$

$$A^+ = A^+BB^+ \quad \text{by } A^+AA^+ = A^+BB^+$$

$$B^+ = B^+AA^+$$

$$VV = (A^+A)^{-1}A^+BB^+B(B^+B)^{-1}B^+A$$

$$\cancel{(A^+A)^{-1}A^+BB^+A}$$

$$= (A^+A)^{-1}A^+AA^+A$$

$$= (A^+A)^{-1}A^+A$$

$$VV = I$$

$$VV = (B^+B)^{-1}B^+A(A^+A)^{-1}A^+B$$

$$= (B^+B)^{-1}B^+AA^+A(A^+A)^{-1}A^+B$$

$$= (B^+B)^{-1}B^+AA^+B$$

$$= (B^+B)^{-1}B^+BB^+B = (B^+B)^{-1}B^+B$$

$$VV = I$$

4) $B = A Q$

$$BB^+ = AA^+ \quad B^+BB^+ = B^+AA^+$$

$$B^+ = B^+AA^+ = (B^+B)^{-1}B^+AA^+$$

$$B = AA^+B(B^+B)^{-1}$$

$$= AV^+$$

$$Q = V^+$$