

MA423 Theory Assignment 1

Group 4

September 2020

A1.(a)

$$P_k = \left[\begin{array}{c|c} I_{k-1} & \\ \hline & \hat{P}_k \end{array} \right], \hat{P}_k \text{ is a } (n-k+1) \times (n-k+1) \text{ transposition.}$$

$$\mathcal{P}_k = P_{k+1} \cdots P_{n-1}, k = 1, 2 \cdots n-2$$

here P_k is also a transposition in which no row exchanges happen in the first $k-1$ rows. So for any transposition P the matrix $\left[\begin{array}{c|c} I & \\ \hline & P \end{array} \right]$ is also a transposition matrix.

So we can write each $P_j = \left[\begin{array}{c|c} I_k & \\ \hline & \hat{P}_j \end{array} \right]$ for $j = k+1, \dots, n-1$

$$\mathcal{P}_k = P_{k+1} \cdots P_{n-1}, k = 1, 2 \cdots n-2$$

$$\Rightarrow \mathcal{P}_k = \left[\begin{array}{c|c} I_k & \\ \hline & \widehat{P_{k+1}} \end{array} \right] \left[\begin{array}{c|c} I_k & \\ \hline & \widehat{P_{k+2}} \end{array} \right] \cdots \left[\begin{array}{c|c} I_k & \\ \hline & \widehat{P_{n-1}} \end{array} \right]$$

$$\therefore \mathcal{P}_k = \left[\begin{array}{c|c} I_k & \\ \hline & \widehat{P_{k+1} P_{k+2} \cdots P_{n-1}} \end{array} \right]$$

A1.(b)

To prove if $\widetilde{M}_k = \mathcal{P}_k^T M_k \mathcal{P}_k$, then

$$\widetilde{M}_k = I_n - \left[\begin{array}{c} 0 \\ \vdots \\ \widehat{m_{k+1,k}} \\ \vdots \\ \widehat{m_{n,k}} \end{array} \right] e_k^T$$

where,

$$\left[\begin{array}{c} \widehat{m_{k+1,k}} \\ \vdots \\ \widehat{m_{n,k}} \end{array} \right] = \widehat{\mathcal{P}_{n-1}} \cdots \widehat{\mathcal{P}_{k+1}} \left[\begin{array}{c} \widehat{m_{k+1,k}} \\ \vdots \\ \widehat{m_{n,k}} \end{array} \right]$$

we know that , $M_k = I_n - \left[\begin{array}{c} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{array} \right] e_k^T$ so

$$\mathcal{P}_k^T M_k \mathcal{P}_k = \mathcal{P}_k^T I_n \mathcal{P}_k - \mathcal{P}_k^T \left[\begin{array}{c} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{array} \right] \mathcal{P}_k$$

$$\mathcal{P}_k^T M_k \mathcal{P}_k = I_n - \mathcal{P}_k^T \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T \mathcal{P}_k$$

$$\text{so we know that } \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T = \begin{bmatrix} 0 & \cdots & 0 & \cdots & 0 \\ \vdots & & \vdots & \cdots & \vdots \\ 0 & \cdots & m_{k+1,k} & \cdots & 0 \\ \vdots & \cdots & \vdots & \cdots & \vdots \\ 0 & \cdots & m_{n,k} & \cdots & 0 \end{bmatrix}$$

since the last $n - k$ columns are 0 and $A\mathcal{P}_k$ permutes only the columns $k + 1 \cdots n$ of a matrix A

$$\begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T \mathcal{P}_k = \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T$$

$$\mathcal{P}_k^T = \left[\frac{I_k}{\widetilde{P_{k+1} P_{k+2} \cdots P_{n-1}}} \right]^T = \left[\frac{I_k}{\widetilde{P_{n-1}^T P_{n-2}^T \cdots P_{k+1}^T}} \right] = \left[\frac{I_k}{\widetilde{P_{n-1} P_{n-2} \cdots P_{k+1}}} \right]$$

as $P^T = P$ for permutation matrix

$$\therefore \mathcal{P}_k \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T = \left[\frac{I_k}{\widetilde{P_{n-1} P_{n-2} \cdots P_{k+1}}} \right] \begin{bmatrix} 0 \\ \vdots \\ m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix} e_k^T = \widetilde{P_{n-1} P_{n-2} \cdots P_{k+1}} \begin{bmatrix} m_{k+1,k} \\ \vdots \\ m_{n,k} \end{bmatrix}$$

Hence proved

A1.(c) We know that

$$U = M_{n-1} P_{n-1} M_{n-2} P_{n-2} \cdots M_k P_k \cdots M_1 P_1 A$$

and

$$\widetilde{M}_k^{-1} = P_{n-1} P_{n-2} \cdots P_{k+1} M_k^{-1} P_{k+1} \cdots P_{n-1}$$

$$\implies \widetilde{M}_{n-1}^{-1} U = M_{n-1}^{-1} M_{n-1} P_{n-1} M_{n-2} P_{n-2} \cdots M_k P_k \cdots M_1 P_1 A$$

$$\implies \widetilde{M}_{n-1}^{-1} U = P_{n-1} M_{n-2} P_{n-2} \cdots M_k P_k \cdots M_1 P_1 A$$

$$\implies \widetilde{M}_{n-2}^{-1} \widetilde{M}_{n-1}^{-1} U = (P_{n-1} M_{n-2}^{-1} P_{n-1}) P_{n-1} M_{n-2} P_{n-2} \cdots M_k P_k \cdots M_1 P_1 A$$

using $P^{-1} = P$ for a permutation matrix

$$\implies \widetilde{M}_{n-2}^{-1} \widetilde{M}_{n-1}^{-1} U = P_{n-1} P_{n-2} M_{n-3} \cdots M_k P_k \cdots M_1 P_1 A$$

\vdots

At the $(n - k)^{\text{th}}$ step, we have

$$\widetilde{M}_{k+1}^{-1} \cdots \widetilde{M}_{n-1}^{-1} U = P_{n-1} \cdots P_{k+1} M_k \cdots M_1 P_1 A = \mathcal{P}_k^T M_k \cdots M_1 P_1 A$$

$$\widetilde{M}_k^{-1} \cdots \widetilde{M}_{n-1}^{-1} U = \mathcal{P}_k^T (M_k^{-1} \mathcal{P}_k \mathcal{P}_k^T M_k) \cdots M_1 P_1 A = \mathcal{P}_k^T M_{k-1} \cdots M_1 P_1 A$$

so we finally get

$$\widetilde{M}_1^{-1} \cdots \widetilde{M}_k^{-1} \cdots \widetilde{M}_{n-1}^{-1} U = P_{n-1} \cdots P_1 A$$

And, since $LU = P_{n-1} \cdots P_1 A$

$$(L - \widetilde{M}_1^{-1} \cdots \widetilde{M}_k^{-1} \cdots \widetilde{M}_{n-1}^{-1}) U = 0$$

$$\therefore L = \widetilde{M}_1^{-1} \cdots \widetilde{M}_k^{-1} \cdots \widetilde{M}_{n-1}^{-1} \text{ (Since } U \text{ is invertible)}$$

now we prove

$$\widetilde{M}_k^{-1} = I_n + \begin{bmatrix} 0 \\ \vdots \\ \widetilde{m}_{k+1,k} \\ \vdots \\ \widetilde{m}_{n,k} \end{bmatrix} e_k^T$$

$$\text{let } x = \begin{bmatrix} 0 \\ \vdots \\ \widetilde{m}_{k+1,k} \\ \vdots \\ \widetilde{m}_{n,k} \end{bmatrix} \text{ which gives}$$

$$\widetilde{M}_k^{-1} \widetilde{M}_k = (I_n + x e_k^T)(I_n - x e_k^T)$$

$$\implies \widetilde{M}_k^{-1} \widetilde{M}_k = I_n + I_n x e_k^T - I_n x e_k^T - x(e_k^T x) e_k^T$$

here $e_k^T x = x_k = 0$ where x_k is the k^{th} element in column vector x

so we get our result $\widetilde{M}_k^{-1} \widetilde{M}_k = I_n$

$$\therefore \widetilde{M}_1^{-1} \cdots \widetilde{M}_{n-1}^{-1} = \left(I_n + \begin{bmatrix} 0 \\ \widetilde{m}_{2,1} \\ \vdots \\ \widetilde{m}_{n,1} \end{bmatrix} e_1^T \right) \cdots \left(I_n + \begin{bmatrix} 0 \\ \vdots \\ \widetilde{m}_{k+1,k} \\ \vdots \\ \widetilde{m}_{n,k} \end{bmatrix} e_k^T \right) \cdots \left(I_n + \begin{bmatrix} 0 \\ \vdots \\ \widetilde{m}_{n-1,n} \end{bmatrix} e_{n-1}^T \right)$$

$$\widetilde{M}_1^{-1} \cdots \widetilde{M}_{n-1}^{-1} = \begin{bmatrix} 1 & & & & & \\ \widetilde{m}_{21} & 1 & & & & \\ \widetilde{m}_{31} & \widetilde{m}_{32} & \ddots & & & \\ \vdots & \vdots & & \ddots & & \\ \widetilde{m}_{k1} & \widetilde{m}_{k2} & \cdots & \cdots & 1 & \\ \vdots & \vdots & & \vdots & & \ddots \\ \widetilde{m}_{n1} & \widetilde{m}_{n2} \cdots & \cdots & \widetilde{m}_{n,k-1} & \cdots \widetilde{m}_{n,n-1} & 1 \end{bmatrix}$$

Q2.(a) Prove $(A^{-1} - uu^T)^{-1} = A^{-1} + \frac{(A^{-1}u)u^T A^{-1}}{1 - u^T A^{-1}u}$

Since A is positive definite, So is A^{-1} .

Let $x = A^{-1}u$. Then $x^T = u^T(A^{-1})^T = u^T A^{-1}$

$$\therefore A^{-1} + \frac{(A^{-1}u)u^T A^{-1}}{1 - u^T A^{-1}u} = A^{-1} + \frac{xx^T}{1 - u^T x}$$

$$\begin{aligned} \text{Then, } (A - uu^T)(A^{-1} + \frac{xx^T}{1 - u^T x}) &= I + \frac{Axx^T}{1 - u^T x} - uu^T A^{-1} - \frac{uu^T xx^T}{1 - u^T x} \\ &= I + \frac{Axx^T}{1 - u^T x} - (Ax)(x^T A)A^{-1} - \frac{(Ax)u^T xx^T}{1 - u^T x} \quad (x = A^{-1}u \implies u = Ax \text{ and } u^T = x^T A^T = x^T A) \\ &= I + \frac{Axx^T}{1 - u^T x} - (1 - u^T x) \frac{Axx^T}{1 - u^T x} - (u^T x) \frac{Axx^T}{1 - u^T x} \quad (u^T x \text{ is scalar}) \\ &= I + 0 = I \\ \implies A^{-1} + \frac{(A^{-1}u)u^T A^{-1}}{1 - u^T A^{-1}u} &\text{ is the inverse of } A - uu^T \end{aligned}$$

Q2.(b) Write pseudo code to solve $(A - uu^T)x = b$, using the relation established in part (a) in $\frac{n^3}{3} + O(n^2)$ flops.

Here, $x = (A - uu^T)^{-1}b$, substituting the value of $(A - uu^T)^{-1}$ from part (a)

$$x = A^{-1}b + \frac{(A^{-1}u)(A^{-1}u)^T b}{1 - u^T A^{-1}u}$$

Let $x_1 = A^{-1}b$ and $x_2 = A^{-1}u$

$$\text{Then } x = x_1 + \frac{(x_2)(x_2)^T b}{1 - u^T x_2}$$

We can compute the value of x by solving these two systems. We find the cholesky factor of A and use it to solve for x_1 and x_2 .

Algorithm

1. Find the Cholesky factor G of A . $\frac{n^3}{3} + O(n^2)$ flops
2. Solve $G^T y_1 = b$, where $y_1 = Gx_1$ $O(n^2)$ flops
3. Solve $Gx_1 = y_1$ $O(n^2)$ flops
4. Solve $G^T y_2 = u$, where $y_2 = Gx_2$ $O(n^2)$ flops
5. Solve $Gx_2 = y_2$ $O(n^2)$ flops
6. Using the values of x_1 and x_2 , compute the value of x . $O(n^2)$ flops

Total flop count $= \frac{n^3}{3} + O(n^2)$

Q2.(c)

$A = [a_{i,j}]$ is a positive definite matrix.

Consider $x = \alpha e_i + \beta e_j$, $1 \leq i < j \leq n$, where e_i is the i^{th} column vector of I_n .

Since A is positive definite $x^T A x > 0 \quad \forall x \in R^n \setminus 0$.

$$\implies (\alpha e_i + \beta e_j)^T A (\alpha e_i + \beta e_j) > 0$$

$$\implies \begin{bmatrix} 0 & \cdots & \alpha & \cdots & \beta & \cdots & 0 \end{bmatrix} \begin{bmatrix} a_{1,1} & \cdots & a_{1,n} \\ \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,n} \end{bmatrix} \begin{bmatrix} 0 \\ \vdots \\ \alpha \\ \vdots \\ \beta \\ \vdots \\ 0 \end{bmatrix} > 0.$$

$$\implies \alpha^2 a_{i,i} + \beta^2 a_{j,j} + \alpha\beta(a_{i,j} + a_{j,i}) > 0$$

Since A is symmetric positive definite $a_{i,j} = a_{j,i}$.

$$\implies \alpha^2 a_{i,i} + \beta^2 a_{j,j} + 2\alpha\beta a_{i,j} > 0$$

For $\alpha = a_{i,j}$ & $\beta = -a_{i,i}$

$$a_{i,j}^2 a_{i,i} + a_{i,i}^2 a_{j,j} - 2a_{i,i} a_{i,j}^2 > 0$$

$$\implies a_{i,i}(a_{i,i} a_{j,j} - a_{i,j}^2) > 0$$

Here $a_{i,i} > 0 \quad \forall 1 \leq i \leq n$ as A is positive definite.

$$\therefore a_{i,i} a_{j,j} - a_{i,j}^2 > 0 \quad \forall 1 \leq i < j \leq n.$$