

- 1) ~~Since SPP matrices are considered SPD matrices, we can use the Cholesky factorization for computing the Cholesky factorization of a symmetric ~~matrix~~ positive definite <sup>CSPD</sup> ~~matrix~~ matrix A, we need to partition A and L as follows:~~

$$A = \begin{pmatrix} A_{00} & a_{01} \\ a_{10}^T & \alpha_{11} \end{pmatrix} \quad L = \begin{pmatrix} L_{00} & 0 \\ l_{10}^T & \lambda_{11} \end{pmatrix}$$

Where  $A_{00}$  is a  $p \times p$  submatrix,  $a_{01}$  is a  $p \times q$  submatrix,  $a_{10}^T$  is a  $q \times p$  submatrix, and  $\alpha_{11}$  is a scalar. Also,  $L_{00}$  is a  $p \times p$  lower triangular submatrix,  $l_{10}^T$  is a  $q \times p$  submatrix, and  $\lambda_{11}$  is a scalar. Here, we assume that A is partitioned into  $p$  and  $q$  such that  $(p+q=n)$  and  $A_{00}$  and  $L_{00}$  are also SPP matrices,

The Cholesky factorization of A is:  $A = LL^T$

Substituting the partitioned form of A and L, this results:

$$\begin{pmatrix} A_{00} & a_{01} \\ a_{10}^T & \alpha_{11} \end{pmatrix} = \begin{pmatrix} L_{00} & 0 \\ l_{10}^T & \lambda_{11} \end{pmatrix} \begin{pmatrix} L_{00}^T & l_{10} \\ 0 & \lambda_{11} \end{pmatrix}$$

\*: part of A that is not updated yet

Multiplying the right hand side, we get:

$$A_{00} = L_{00} L_{00}^T$$

$$a_{10}^T = l_{10}^T L_{00}^T$$

$$\alpha_{11} = l_{10}^T l_{10} + \lambda_{11}$$

~~This is the algorithm~~  
This gives us:

$$L_{00} = \text{Chol}(A)_{00} \quad *$$

$$l_{10}^T = a_{10}^T L_{00}^{-T} \quad \lambda_{11} = \alpha_{11} - l_{10}^T l_{10}$$

The equalities above results in this algorithm:

1. Partition  $A \rightarrow \left( \begin{array}{c|c} A_{00} & * \\ \hline a_{10}^T & \alpha_{11} \end{array} \right)$
2. Assume that  $A_{00} := L_{00} = \text{Chol}(A_{00})$  has already been computed on previous iterations of the ~~for~~ loop in the algorithm.
3. Overwrite  $a_{10}^T := z_{10}^T = a_{10}^T L_{00}^{-T}$
4. Overwrite  $\alpha_{11} := \sqrt{\alpha_{11} - z_{10}^T z_{10}}$

Algorithm  $A := \text{Chol}(A)$

Partition  $A \rightarrow \left( \begin{array}{c|c} A_{TL} & * \\ \hline A_{BL} & A_{BR} \end{array} \right)$

where  $A_{TL}$  is  $0 \times 0$

While  $m(A_{TL}) < m(A)$  do:

repartition  $\left( \begin{array}{c|c} A_{TL} & * \\ \hline A_{BL} & A_{BR} \end{array} \right) \rightarrow \left( \begin{array}{c|c|c} A_{00} & * & * \\ \hline a_{10}^T & \alpha_{11} & * \\ \hline A_{20} & a_{21} & A_{22} \end{array} \right)$

where  $\alpha_{11}$  is  $|x|$

(Bordered Algorithm)

$$a_{10}^T := a_{10}^T \text{TRIL}(A_{00})^{-T}$$

$$\alpha_{11} := \alpha_{11} - a_{10}^T a_{10}$$

$$\alpha_{11} := \sqrt{\alpha_{11}}$$

Continue with

$$\left( \begin{array}{c|c} A_{TL} & * \\ \hline A_{BL} & A_{BR} \end{array} \right) \leftarrow \left( \begin{array}{c|c|c} A_{00} & * & * \\ \hline a_{10}^T & \alpha_{11} & * \\ \hline A_{20} & a_{21} & A_{22} \end{array} \right)$$

end while

1b) We will prove the Cholesky factorization theorem by showing that the Cholesky factorization is well-defined for a matrix  $A$  that is SPD. For this proof, we will use some lemmas.

Lemma 1:

Let  $A \in \mathbb{R}^{n \times n}$  be an SPD matrix. Then  $\alpha_{11}$  is real and positive.  
 - proof: This is just a special case for lemma 5.4.4.1 in the textbook. Since all SPD matrices are by definition <sup>also</sup> HPD matrices.

Lemma 2:

Let  $A \in \mathbb{R}^{n \times n}$  be an SPD matrix and  $\lambda_{21} = a_{21}/\sqrt{\alpha_{11}}$ . Then,  $A_{22} - \lambda_{21}\lambda_{21}^T$  is an SPD matrix.

- proof: Since  $A$  is symmetric so are  $A_{22}$  and  $A_{22} - \lambda_{21}\lambda_{21}^T$ . Given  $x_1 \neq 0$  be any vector with length  $n-1$ . Let's define  $x$  to be  $\begin{pmatrix} x_0 \\ x_1 \end{pmatrix}$  where  $x_0 = -a_{21}^T x_1 / \alpha_{11}$ . Then, because  $x \neq 0$ ,

$$\begin{aligned}
 & \text{---} \\
 & = \begin{pmatrix} x_0 \\ x_1 \end{pmatrix}^T \begin{pmatrix} \alpha_{11} & a_{21}^T \\ a_{21} & A_{22} \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \end{pmatrix} = x^T A x > 0 \\
 & = \begin{pmatrix} x_0 \\ x_1 \end{pmatrix}^T \begin{pmatrix} \alpha_{11} x_0 + a_{21}^T x_1 \\ a_{21} x_0 + A_{22} x_1 \end{pmatrix} \\
 & = \alpha_{11} x_0^2 + x_0 a_{21}^T x_1 + x_1^T a_{21} x_0 + x_1^T A_{22} x_1 \\
 & = \alpha_{11} \left( \frac{a_{21}^T x_1 x_1^T a_{21}}{\alpha_{11}} \right) - \frac{x_1^T a_{21} a_{21}^T x_1}{\alpha_{11}} - x_1^T a_{21} \frac{a_{21}^T x_1}{\alpha_{11}} + x_1^T A_{22} x_1 \\
 & = x_1^T \left( A_{22} - \frac{a_{21} a_{21}^T}{\alpha_{11}} \right) x_1 \\
 & = x_1^T (A_{22} - \lambda_{21} \lambda_{21}^T) x_1
 \end{aligned}$$

Therefore, we conclude that  $A_{22} - \lambda_{21}\lambda_{21}^T$  is an SPD matrix.

## Proof by induction:

~~Base case~~

Base case:

For  $n=1$ , the result is trivial for a  $1 \times 1$  matrix  $A$ , which is just equal to  $a_{11}$ . In this example, since  $A$  is SPD, then  $a_{11}$  is real and positive. A Cholesky factor is then given as  $\lambda_{11} = \sqrt{a_{11}}$ . This is unique if we know that  $\lambda_{11}$  is positive.

Inductive step:

Let's assume the result is true for an SPD matrix  $A$ , where  $A \in \mathbb{R}^{(n-1) \times (n-1)}$ . We will also show that this is true as well for SPD matrix  $A$ , where  $A \in \mathbb{R}^{n \times n}$ . First, given  $A \in \mathbb{R}^{n \times n}$  be an SPD matrix, partition  $A$  and  $L$  such that:

$$A = \begin{pmatrix} a_{11} & a_{21}^T \\ a_{21} & A_{22} \end{pmatrix} \quad L = \begin{pmatrix} \lambda_{11} & 0 \\ l_{21} & L_{22} \end{pmatrix}$$

and let's set  $\lambda_{11} = \sqrt{a_{11}}$  (which has been shown previously),  $l_{21} = a_{21}/\lambda_{11}$ , and  $L_{22}$  will then be equal to  $\text{chol}(A_{22} - l_{21}l_{21}^T)$ . Therefore,  $L$  is the Cholesky factor of a SPD matrix  $A$ . By the principle of mathematical induction, this proof holds.