

2.)

Backward Error Analysis of Bordered algo.

we know def of bordered algorithm that

$$\left( \begin{array}{c|c} A_{TL} & A_{TR} \\ \hline A_{BL} & A_{BR} \end{array} \right) = \left( \begin{array}{c|c} L \setminus U_{TL} & \hat{A}_{TR} \\ \hline \hat{A}_{BL} & \hat{A}_{BR} \end{array} \right) \Delta \quad L_{TL} U_{TL} = \hat{A}_{TL}$$

For backward error analysis, can be computed as:

$$\left( \begin{array}{c|c} [\tilde{L} \setminus \tilde{U}]_{TL} = [LU(A_{TL})] & \tilde{U}_{TR} = 0 \\ \hline \tilde{L}_{BL} = 0 & [\tilde{L} \setminus \tilde{U}]_{BR} = 0 \end{array} \right) \quad \begin{array}{l} \text{computed factors } \tilde{L} \text{ and } \tilde{U} \\ \text{satisfy} \\ \tilde{L}\tilde{U} = A + \delta A \end{array}$$

we will use a proof by induction.

a.) Base case:  $n=1$

IF  $A^{1 \times 1}$ , then  $A$  is just a real valued scalar.

From 6.2.2, we defined error from storing a real number as

$$| \delta A | \leq E_{mach} |A|,$$

where Additionally, we will define

$$\gamma_n \text{ as } \gamma_n = \frac{n E_{mach}}{1 - n E_{mach}} \quad \begin{array}{l} \text{(Thm 6.3.2.3)} \\ \text{for } \forall n \geq 1 \text{ \& } n E_{mach} < 1 \end{array}$$

For base case  $n=1$ ,  $\gamma_n = \frac{E_{mach}}{1 - E_{mach}} \approx E_{mach}$

Thus:  $| \delta A | \leq \gamma_n | \tilde{L} | | \tilde{U} |$

$\Rightarrow | \delta A | \leq \gamma_n$  holds true for  $n=1$

( $| \tilde{L} | | \tilde{U} | \leq 1$  due to  $\Delta$  inequality)



## 2) Inductive Step

Assume that for  $A_{00} \in \mathbb{R}^{n \times n}$ , bordered algo factorization computes  $\tilde{L}$  and  $\tilde{U}$  where  $A_{00} + \Delta A_{00} = \tilde{L}\tilde{U}$  w/  $\|A\| \leq \gamma_n \|\tilde{L}\| \|\tilde{U}\|$  result holds for  $n-1$ .

Apply bordered algo to  $A$ :

Define  $A = \begin{pmatrix} A_{00} & A_{01} & A_{02} \\ A_{10}^T & A_{11} & A_{12}^T \\ A_{20} & A_{21} & A_{22} \end{pmatrix}$   $L = \begin{pmatrix} L_{00} & 0 \\ L_{10}^T & L_{11} \end{pmatrix}$   $U = \begin{pmatrix} U_{00} & U_{01} \\ 0 & U_{11} \end{pmatrix}$

At the top of the loop, after  $A$

$$\begin{pmatrix} A_{00} & a_{01} & a_{02} \\ a_{10}^T & a_{11} & a_{12}^T \\ A_{20} & a_{21} & A_{22} \end{pmatrix} \rightarrow \begin{pmatrix} L_{00} u_{01} & \hat{a}_{02} \\ \hat{x}_{10}^T & v_{11} & \hat{a}_{12}^T \\ \hat{a}_{20} & \hat{a}_{21} & \hat{a}_{22} \end{pmatrix} \sim \begin{pmatrix} L_{00} & 0 \\ L_{10}^T & 1 \end{pmatrix} \begin{pmatrix} U_{00} & u_{01} \\ 0 & v_{11} \end{pmatrix} = \begin{pmatrix} \hat{A}_{00} & \hat{a}_{01} \\ \hat{a}_{10} & \hat{a}_{11} \end{pmatrix}$$

We want to prove

Matrix Mult  $\begin{pmatrix} L_{00} & 0 \\ L_{10}^T & 1 \end{pmatrix} \begin{pmatrix} U_{00} & u_{01} \\ 0 & v_{11} \end{pmatrix} = \begin{pmatrix} L_{00} U_{00} & L_{00} u_{01} \\ L_{10}^T U_{00} & L_{10}^T u_{01} + v_{11} \end{pmatrix}$  want to prove each element is bounded.  
Cholesky factorization  $L_{00} U_{00} = A_{00}$

From assumption above  $A_{00} + \Delta A_{00} = \tilde{L}\tilde{U}$  w/  $\|A\| \leq \gamma_n \|\tilde{L}\| \|\tilde{U}\|$

Next, we prove  $L_{00} u_{01} = a_{01}$

We know  $\tilde{L}_{00} \tilde{U}_{01} + \Delta \tilde{L}_{00} \tilde{U}_{01} = a_{01}$  (Corollary 6.4.1.4)

There exists a matrix notation st  $(L + \Delta L)\hat{x} = y$  where  $\|\Delta L\| \leq \gamma_n \|L\|$  applying to eqn above

$$\tilde{L}_{00} \tilde{U}_{01} = a_{01} - \Delta \tilde{L}_{00} \tilde{U}_{01}$$

Thus  $E_{01} = -\Delta \tilde{L}_{00} \tilde{U}_{01} = \hat{a}_{01}$

Hence  $\|E_{01}\| \leq \gamma_n \|\tilde{L}_{00} \tilde{U}_{01}\| \leq \gamma_n \|\tilde{L}_{00}\| \|\tilde{U}_{01}\|$



Next, we prove

$$\hat{L}_{10}^T U_{00} = a_{10}^T$$

Similar to above, we show

$$\hat{L}_{10}^T U_{00} = a_{10}^T - \hat{L}_{10}^T \hat{U}_{00}$$

Corollary 6.3.3.2.

$$E_{10}^T = -\hat{L}_{10}^T U_{00} = \delta a_{10}^T \quad \text{so} \quad |E_{10}^T| \leq \gamma_n |\hat{L}_{10}^T| |U_{00}|$$

Finally, we must bound  $\hat{L}_{10}^T U_{01} + v_{11} = \alpha_{11}$

$$\text{Rewrite } \alpha_{11} = \hat{L}_{10}^T \hat{U}_{01} + \delta a_{11} = \hat{v}_{11}$$

We will use an external source to help. In Science of Deriving Stable Analyzers by Paolo Bientinesi and Robert Van Gueun, thm 6.4 states

Consider assignments  $\mu = \alpha - (x^T y)$  and  $v = (\alpha - (x^T y)) / \lambda$ .

$$\text{Then, } |\alpha - x^T y - \lambda v| \leq \gamma_n (|x|^T |y| + |v| |\lambda|) = \gamma_n |v|^T |z|$$

$$\text{where } v = \left(\frac{x}{\lambda}\right) \text{ and } z = \left(\frac{y}{\lambda}\right)$$

Thus, if  $\lambda = 1$

$$E_{11} = \delta a_{11} = \hat{v}_{11} - a_{11} + \hat{L}_{10}^T \hat{U}_{01}$$

apply lemma:

$$|E_{11}| \leq \gamma_n \left( \frac{|\hat{L}_{10}^T|}{1} \right) \left| \left( \frac{\hat{U}_{01}}{\delta a_{11}} \right) \right|$$

Therefore,

Thus if all four elements are bounded, now add backward error

$$A + \delta A = \tilde{L} \tilde{U}$$

$$\begin{pmatrix} A_{00} & a_{01} \\ \hat{L}_{10}^T & \alpha_{11} \end{pmatrix} + \begin{pmatrix} \delta A_{00} & \delta a_{01} \\ \delta \hat{L}_{10}^T & \delta \alpha_{11} \end{pmatrix} = \begin{pmatrix} \tilde{L}_{00} & \tilde{a} \\ \tilde{L}_{10}^T & 1 \end{pmatrix} \begin{pmatrix} \tilde{U}_{00} & \tilde{u}_{01} \\ \tilde{v} & \tilde{v}_{00} \end{pmatrix}$$

$$\left| \begin{pmatrix} \delta A_{00} & \delta a_{01} \\ \delta \hat{L}_{10}^T & \delta \alpha_{11} \end{pmatrix} \right| \leq \left( \frac{\gamma_n |\tilde{L}_{00}| |\tilde{U}_{00}|}{\gamma_n |\tilde{L}_{10}^T| |\tilde{U}_{00}|} \frac{\gamma_n |\tilde{L}_{00}| |\tilde{u}_{01}|}{\gamma_n |\tilde{L}_{10}^T| |\tilde{U}_{00}|} \right)$$

$$= \gamma_n \left( \left| \begin{pmatrix} \tilde{L}_{00} & \tilde{a} \\ \tilde{L}_{10}^T & 1 \end{pmatrix} \right| \left| \begin{pmatrix} \tilde{U}_{00} & \tilde{u}_{01} \\ \tilde{v} & \tilde{v}_{00} \end{pmatrix} \right| \right) = \text{Right hand side}$$

backward analysis for dot product

factor & separate