

Solutions to Homework 5

Lecturer: Inderjit Dhillon

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Keywords: *Conditioning, Stability, Error Analysis, Gaussian Elimination*1. $U = A, L = I$ **for** $k = 1 : m - 1$ **do** $\mathbf{l}_{k+1:m,k} = \mathbf{u}_{k+1:m,k} / u_{kk}$ $U_{k+1:m,k:m} = U_{k+1:m,k:m} - \mathbf{l}_{k+1:m,k} U_{k,k:m}$ **end for**

2. (a) Elimination by columns from left to right is equivalent to post-multiplication by an upper-triangular matrix. $m - 1$ such operations will make A a lower-triangular matrix, thus:

$$AU_1U_2\ldots U_{m-1} = L \Rightarrow A = LU_{m-1}^{-1}U_{m-2}^{-1}\ldots U_1^{-1}.$$

Now the inverse of a upper triangular matrix is upper-triangular and multiplication of two upper-triangular matrices is also an upper-triangular matrix. Hence, $A = LU$.

- (b) Gaussian Elimination is equivalent to pre-multiplication with lower-triangular matrices. Thus after rescaling of columns, Gaussian Elimination is equivalent to: $L_{m-1}L_{m-2}\ldots L_1AD = U \Rightarrow A = LUD^{-1}$. Thus, unknowns are rescaled by D^{-1} .
- (c) Elimination by columns from left to right is equivalent to post-multiplication by an upper-triangular matrix. Thus, using part (a), $U = D\tilde{U}$. Thus, $A = LD\tilde{U}$, where \tilde{U} incorporates the additional column operations.

3. (a) We prove the claim using induction. Base case follows directly from the fact that $x_1 \otimes a_1 = x_1a_1(1 + \epsilon)$, $|\epsilon| \leq \epsilon_{\text{machine}}$. Now, $fl(\mathbf{x}^T \mathbf{a}) = x_1 \otimes a_1 \oplus x_2 \otimes a_2 \oplus \ldots x_n \otimes a_n$. Using induction hypothesis, $fl(\mathbf{x}^T \mathbf{a}) = (\sum_{i=1}^{n-1} x_i a_i + e_{n-1}) \oplus x_n \otimes a_n$, $|e_{n-1}| \leq (n-1)\epsilon_{\text{machine}} \sum_{i=1}^{n-1} |x_i||a_i| + O(\epsilon_{\text{machine}}^2)$. Thus, $fl(\mathbf{x}^T \mathbf{a}) = (\sum_{i=1}^{n-1} x_i a_i + e_{n-1} + x_n a_n(1 + \epsilon_1))(1 + \epsilon_2) = \mathbf{x}^T \mathbf{a} + e_{n-1} + x_n a_n \epsilon_1 + \mathbf{x}^T \mathbf{a} \epsilon_2 + e_{n-1} \epsilon_2 + x_n a_n \epsilon_1 \epsilon_2$, where $|\epsilon_1| \leq \epsilon_{\text{machine}}$, $|\epsilon_2| \leq \epsilon_{\text{machine}}$. Thus, $fl(\mathbf{x}^T \mathbf{a}) = \mathbf{x}^T \mathbf{a} + e_n$, where $e_n = e_{n-1} + x_n a_n \epsilon_1 + \mathbf{x}^T \mathbf{a} \epsilon_2 + e_{n-1} \epsilon_2 + x_n a_n \epsilon_1 \epsilon_2$. Hence, $|e_n| \leq |e_{n-1}| + |x_n||a_n|\epsilon_{\text{machine}} + |\mathbf{x}^T \mathbf{a}|\epsilon_{\text{machine}} + O(\epsilon_{\text{machine}}^2) = (n-1) \sum_{i=1}^{n-1} |x_i||a_i|\epsilon_{\text{machine}} + |x_n||a_n|\epsilon_{\text{machine}} + |\mathbf{x}^T \mathbf{a}|\epsilon_{\text{machine}} \leq n|\mathbf{x}^T \mathbf{a}|\epsilon_{\text{machine}}$. Hence proved.

- (b) $(XA)_{ij} = \mathbf{x}_i^T \mathbf{a}_j$, where \mathbf{x}_i^T represents i -th row of X and \mathbf{a}_j represents j -th column of A . Using part (a), $fl((XA)_{ij}) = (XA)_{ij} + e_{ij}$, where $|e_{ij}| \leq n|\mathbf{x}_i^T||\mathbf{a}_j|\epsilon_{\text{machine}} + O(\epsilon_{\text{machine}}^2)$. Thus $\|fl(XA) - XA\|_F \leq \sqrt{\sum_{ij} n^2 \epsilon_{\text{machine}}^2 (|\mathbf{x}_i^T||\mathbf{a}_j|)^2 + O(\epsilon_{\text{machine}}^2)} = n\epsilon_{\text{machine}} \|X\|_F \|A\|_F + O(\epsilon_{\text{machine}}^2) \leq n\epsilon_{\text{machine}} \|X\|_F \|A\|_F + O(\epsilon_{\text{machine}}^2) = n\epsilon_{\text{machine}} \|X\|_F \|A\|_F + O(\epsilon_{\text{machine}}^2)$, as $\|AB\|_F \leq \|A\|_F \|B\|_F$ and $\|A\|_F = \|A\|_F$.

- (c) Using part (b), $fl(XA) = XA + E$, where $\|E\|_F \leq n\|X\|_F \|A\|_F O(\epsilon_{\text{machine}})$. Now $fl(XA) = X(A + X^{-1}E) = X(A + \delta A)$, where $\delta A = X^{-1}E$. Thus, $\|\delta A\|_F \leq \|X^{-1}\|_F \|E\|_F \leq n\|X^{-1}\|_F \|X\|_F \|A\|_F O(\epsilon_{\text{machine}})$. Thus, $\frac{\|\delta A\|_F}{\|A\|_F} \leq n\kappa(X)O(\epsilon_{\text{machine}})$.

4. $(A + \delta A)(\mathbf{x} + \delta \mathbf{x}) = A\mathbf{x} + A\delta \mathbf{x} + \delta A\mathbf{x} + \delta A\delta \mathbf{x} = \mathbf{b} + \delta \mathbf{b}$. Now $A\mathbf{x} = \mathbf{b}$, hence $(A + \delta A)\delta \mathbf{x} = \delta \mathbf{b} - \delta A\mathbf{x}$. Thus, $\|\delta \mathbf{x}\| = \|(A + \delta A)^{-1}(\delta \mathbf{b} - \delta A\mathbf{x})\| \leq \|(A + \delta A)^{-1}\|(\|\delta \mathbf{b}\| + \|\delta A\mathbf{x}\|) \leq \|(I + A^{-1}\delta A)^{-1}\| \|A^{-1}\|(\|\delta \mathbf{b}\| + \|\delta A\mathbf{x}\|)$.

Using Homework 2, Problem 2(c): if $\|A^{-1}\delta A\| \leq 1$, then $\|(I + A^{-1}\delta A)^{-1}\| \leq \frac{1}{1 - \|A^{-1}\delta A\|} \leq \frac{1}{1 - \|A^{-1}\|\|\delta A\|}$. Thus, $\|\delta \mathbf{x}\| \leq \frac{\|A^{-1}\|}{1 - \|A^{-1}\|\|\delta A\|}(\|\delta \mathbf{b}\| + \|\delta A \mathbf{x}\|)$. Now $\mathbf{b} = A\mathbf{x}$, thus $\|\mathbf{b}\| \leq \|A\|\|\mathbf{x}\|$. Hence,

$$\|\delta \mathbf{x}\| \leq \frac{\|A^{-1}\|}{1 - \|A^{-1}\|\|\delta A\|} \left(\frac{\|\delta \mathbf{b}\|\|A\|\|\mathbf{x}\|}{\|\mathbf{b}\|} + \frac{\|\delta A\|\|\mathbf{x}\|\|A\|}{\|A\|} \right),$$

i.e.,

$$\frac{\|\delta \mathbf{x}\|}{\|\mathbf{x}\|} \leq \frac{\kappa(A)}{1 - \kappa(A) \frac{\|\delta A\|}{\|A\|}} \left(\frac{\|\delta A\|}{\|A\|} + \frac{\|\delta \mathbf{b}\|}{\|\mathbf{b}\|} \right).$$