[10 marks]

Let $x, y \in \mathcal{R}$. Recall that $f(x), f(y) \in \mathcal{R}_b(t, s)$ denote the floating-point representations of x and y, respectively, where $f(x) = x(1 - \delta_x)$, $f(y) = y(1 - \delta_y)$, and δ_x , δ_y quantify the relative roundoff errors in the respective representations.

In lecture, we showed that a typical computer estimates the product of x and y as

$$f(f(x) \cdot f(y)) = (x \cdot y)(1 - \delta)$$

where $|\delta| \le 3$ eps. Using similar techniques, derive a tight error bound for computer division.

$$\frac{1}{f(y)} = \frac{x \cdot (1 - \delta_x)}{y \cdot (1 - \delta_y)} \cdot (1 - \delta_{xy}) = \frac{x}{y} \cdot \frac{(1 - \delta_x)(1 - \delta_{xy})}{1 - \delta_y}$$

$$= \frac{x}{y} \cdot \frac{1 - \delta_x - \delta_{xy} + \delta_x \cdot \delta_{xy}}{1 - \delta_y}$$

$$= \begin{bmatrix} \text{since } \delta < EPS \implies \delta_x \cdot \delta_{xy} - \text{insignificant, also} \\ 1 - \delta_y = 1 \end{bmatrix}$$

$$= \frac{x}{y} \cdot (1 - \delta_x - \delta_{xy}) = \frac{x}{y} \cdot (1 - \delta_{xx}) \cdot \frac{1}{y} \cdot$$

[10 marks]

Let \hat{x} be a computed solution to Ax = b, $A \in \mathbb{R}^{n \times n}$. The following bound for the relative error in \hat{x} was derived in class:

 $\frac{1}{\operatorname{cond}(A)} \cdot \frac{||\underline{x}||}{||\underline{x}||} \leq \frac{||x - \hat{x}||}{||x||} \leq \operatorname{cond}(A) \frac{||r||}{||b||},$

where $r = b - A\hat{x}$. Starting with the equations Ax = b and $A\hat{x} = b - r$, derive a *lower* bound for $||x - \hat{x}||/||x||$. What do these bounds tell us about the reliability of \hat{x} ?

$$A \stackrel{\times}{\times} = \stackrel{\longrightarrow}{b} - \stackrel{\longrightarrow}{\times} = A \stackrel{\times}{\times} - \stackrel{\longrightarrow}{\times} = A \stackrel{\longrightarrow}{\times} - \stackrel{\longrightarrow}{\times} - \stackrel{\longrightarrow}{\times} = A \stackrel{\longrightarrow}{\times} - \stackrel{\longrightarrow}{\times$$

the reliability of \hat{x} depends are on the size of cond (A), if cond (A) is large a small \hat{x} is not reliable; (even if the residual |rell is small); however if cond (A) is small (close to 1) \hat{x} is a reliable solution;

[15 marks]

Consider the linear system Ax = b where

$$A = \left[\begin{array}{ccc} 3 & 5 & 9 \\ 4 & 4 & 4 \\ 1 & 5 & 5 \end{array} \right], \quad b = \left[\begin{array}{c} 40 \\ 24 \\ 26 \end{array} \right].$$

a. Compute the PA = LU factorization of A. Use exact arithmetic. Show all intermediate calculations, including Gauss transforms and permutation matrices.

1 no need to compute interchange zours; 2 -maximum privat; -> $P_2 = I_3$; $L_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -\frac{1}{2} & 1 \end{bmatrix}$; $L_2 L_2 \cdot P_1 \cdot A = \begin{bmatrix} 4 & 4 & 4 \\ 0 & 2 & 6 \\ 0 & 0 & -2 \end{bmatrix}$

$$= \sum_{i=1}^{n} L_{i} \cdot L_{i} \cdot P_{i} \cdot A = U \implies P_{i} \cdot A = L_{i} \cdot L_{i} \cdot U$$

b. Use the factorization computed in (a) to solve the system.

We the factorization computed in (a) to solve the system.

A
$$\cdot \times = b$$

PA $\times = Pb$

PA $\times = Pb$

Wight place of the system.

A $\cdot \times = b$

PA $\times = Pb$

Wight place of the system.

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c. Why is Gaussian Elimination usually implemented as in this question (i.e., PA = LU is computed separately, and then the factorization is used to solve Ax = b?

Because it can be reused for any b in Ax=b; making the calculation of solutions for systems using the same matrix A, more efficient; (because the factorioation is the most expensive part of the finding the solution to Ax=b);

[5 marks]

Consider the iterative improvement algorithm discussed in tutorial:

```
Solve Ax = b for initial approximation \hat{x}_0.

for i = 0, 1, \ldots until convergence

compute r_i = b - A\hat{x}_i

solve Az_i = r_i

update \hat{x}_{i+1} = \hat{x}_i + z_i

end for
```

After the first iteration of this algorithm,

$$\begin{array}{ll} \hat{x}_1 & = & \hat{x}_0 + z_0 \\ & = & \hat{x}_0 + A^{-1}r_0 \\ & = & \hat{x}_0 + A^{-1}(b - A\hat{x}_0) \\ & = & \hat{x}_0 + A^{-1}b - A^{-1}A\hat{x}_0 \end{array} \quad \text{This alone is not a fallacy.}$$

$$= & \hat{x}_0 + x - \hat{x}_0 \\ & = & x \end{array}$$

Apparently the algorithm converges to the true solution x in just one iteration! What is the fallacy in this argument?

2

[15 marks]

Consider the functions f(x) = 1 - 1/(2x) and g(x) = 2x(1-x).

a. How many roots does f have? Are the roots of f fixed-points of g? Are there more fixed points of g than roots of f? Justify your answers.

for descriptions 2000 product, in fix for the point of g; $f(x) = \frac{2x-1}{2x} \implies x \in \mathbb{R} \setminus \{0\}, f(x) = 0 \text{ if } x = \frac{1}{2};$ graph x = f(x) = x = 22222 + 2222 + 2

· yes, ex, g(0) = 0;

b. Using an appropriate theorem proven in lecture, determine the region of local convergence of the fixed-point iteration $x_{k+1} = g(x_k)$, $k = 0, 1, \ldots$, with g(x) as defined above. In other words, find the interval on the x-axis for which the iteration is guaranteed to converge.

uning the fixed point theorem. g'(x) = -4x + 2, |g'(x)| < 1, $\forall x \in (\frac{1}{4}, \frac{3}{4})$; $g(x) \in (\frac{1}{4}, \frac{3}{4})$, $\forall x \in (\frac{1}{4}, \frac{3}{4})$;

the fixed-point iteration will converge on $(\frac{1}{4}, \frac{3}{4})$;