## Homework 2

## CS 210

| Question | Points | Score |
| :--- | :--- | :--- |
| 1 | 10 |  |
| 2 | 10 |  |
| 3 | 10 |  |
| 4 | 10 |  |
| 5 | 10 |  |
| 6 | 10 |  |
| 7 | 10 |  |
| 8 | 10 |  |
| 9 | 10 |  |
| 10 | 10 |  |
| Total | 100 |  |

## Matrix algebra

1. (Trefethen\&Bau 2.6) If $\mathbf{u}$ and $\mathbf{v}$ are $m$-vectors, the matrix $A=I+\mathbf{u v}^{T}$ is known as a rank-one pertubation of the identity. Show that if $A$ is nonsingular, then its inverse has the form $A^{-1}=I+\alpha \mathbf{u v}^{T}$ for some scalar $\alpha$, and give an expression for $\alpha$. For what $\mathbf{u}$ and $\mathbf{v}$ is $A$ singular? If it is singular, what is $\operatorname{null}(A)$ ?
2. (Heath 2.8) Let $A$ and $B$ be any two $n \times n$ matrices.
(a) Prove that $(A B)^{T}=B^{T} A^{T}$.
(b) If $A$ and $B$ are both non-singular, prove that $(A B)^{-1}=B^{-1} A^{-1}$.

## Vector and matrix norms

3. Let $\mathbf{x} \in \mathbb{R}^{n}$. Two vector norms, $\|\mathbf{x}\|_{a}$ and $\|\mathbf{x}\|_{b}$, are equivalent if $\exists c, d \in \mathbb{R}$ such that

$$
c\|\mathbf{x}\|_{b} \leq\|\mathbf{x}\|_{a} \leq d\|\mathbf{x}\|_{b} .
$$

Matrix norm equivalence is defined analogously to vector norm equivalence, i.e., $\|\cdot\|_{a}$ and $\|\cdot\|_{b}$ are equivalent if $\exists c, d$ s.t. $c\|A\|_{b} \leq\|A\|_{a} \leq d\|A\|_{b}$.
(a) Let $\mathbf{x} \in \mathbb{R}^{n}, A \in \mathbb{R}^{n \times n}$. For each of the following, verify the inequality and give an example of a non-zero vector or matrix for which the bound is achieved (showing that the bound is tight):
i. $\|\mathbf{x}\|_{\infty} \leq\|\mathbf{x}\|_{2}$
ii. $\|\mathbf{x}\|_{2} \leq \sqrt{n}\|\mathbf{x}\|_{\infty}$
iii. $\|A\|_{\infty} \leq \sqrt{n}\|A\|_{2}$
iv. $\|A\|_{2} \leq \sqrt{n}\|A\|_{\infty}$

This shows that $\|\cdot\|_{\infty}$ and $\|\cdot\|_{2}$ are equivalent, and that their induced matrix norms are equivalent.
(b) Prove that the equivalence of two vector norms implies the equivalence of their induced matrix norms.

## Sensitivity and conditioning

4. (Heath 2.58) Suppose that the $n \times n$ matrix $A$ is perfectly well-conditioned, i.e., $\operatorname{cond}(\mathrm{A})=1$. Which of the following matrices would then necessarily share this same property?
(a) $c A$, where $c$ is any nonzero scalar
(b) $D A$, where $D$ is a nonsingular diagonal matrix
(c) $P A$, where $P$ is any permutation matrix
(d) $B A$, where $B$ is any nonsingular matrix
(e) $A^{-1}$, the inverse of $A$
(f) $A^{T}$, the transpose of $A$

## Linear Systems

5. (Heath 2.4a) Show that the following matrix is singular.

$$
A=\left(\begin{array}{lll}
1 & 1 & 0 \\
1 & 2 & 1 \\
1 & 3 & 2
\end{array}\right)
$$

6. For each of the following statements, indicate whether the statement is true or false.
$\mathbf{T} / \mathbf{F}$ If a matrix $A$ is singular, then the number of solutions to the linear system $A \mathbf{x}=\mathbf{b}$ depends on the particular choice of right-hand-side $\mathbf{b}$.
$\mathbf{T} / \mathbf{F}$ If a matrix $A$ is nonsingular, then the number of solutions to the linear system $A \mathbf{x}=\mathbf{b}$ depends on the particular choice of right-hand-side $\mathbf{b}$.
$\mathbf{T} / \mathbf{F}$ If a matrix has a very small determinant, then the matrix is nearly singular.
T/F If any matrix has a zero on its main diagonal, then it is necessarily singular.
7. Can a system of linear equations $A \mathbf{x}=\mathbf{b}$ have exactly two solutions? Explain your answer.

## LU Factorization and Gaussian Eliminiation

8. For each of the following statements, indicate whether the statement is true or false.
$\mathbf{T} / \mathbf{F}$ If a triangular matrix has a zero on its main diagonal, then it is necessarily singular.
$\mathbf{T} / \mathbf{F}$ The product of two upper triangular matrices is upper triangular.
$\mathbf{T} / \mathbf{F}$ If a linear system is well-conditioned, then pivoting is unnecessary in Gaussian elimination.
T/F Once the LU factorization of a matrix has been computed to solve a linear system, then subsequent linear systems with the same matrix but different right-hand-side vectors can be solved without refactoring the matrix.
9. Consider $L U$ factorization with partial pivoting of the matrix $A$ which computes

$$
M_{n-1} P_{n-1} \cdots M_{3} P_{3} M_{2} P_{2} M_{1} P_{1} A=U
$$

where $P_{i}$ is a row permutation matrix interchanging rows $i$ and $j>i$.
(a) Show that the matrix $P_{3} P_{2} M_{1} P_{2}^{-1} P_{3}^{-1}$ has the same structure as the matrix $M_{1}$.
(b) Explain how the above expression is transformed into the form $P A=L U$, where $P$ is a row permutation matrix.

## Cholesky Factorization

10. (Heath 2.37) Suppose that the symmetric $(n+1) \times(n+1)$ matrix

$$
B=\left(\begin{array}{cc}
\alpha & \mathbf{a}^{T} \\
\mathbf{a} & A
\end{array}\right)
$$

is positive definite.
(a) Show that the scalar $\alpha$ must be positive and the $n \times n$ matrix $A$ must be positive definite.
(b) What is the Cholesky factorization of $B$ in terms of $\alpha$, $\mathbf{a}$, and the Cholesky factorization of $A-\frac{1}{\alpha} \mathbf{a a}^{T}$ ?

