## CS 370 Fall 2008: Assignment 4

Instructor: Professor Keith Geddes
Lectures: MWF 3:30-4:20 MC 2017
Web Site: UW-ACE
Due: Tue Nov 18, 2008, 5:00 pm, in the Assignment Boxes, 3rd Floor MC

## 1. (LU Factorization)

Consider the following linear system of equations:

$$
\begin{aligned}
1.0 x_{1}+1.0 x_{2}+3.0 x_{3} & =4.0 \\
2.0 x_{1}+1.0 x_{2}-1.0 x_{3}+1.0 x_{4} & =1.0 \\
-1.0 x_{1}+2.0 x_{2}+3.0 x_{3}-1.0 x_{4} & =4.0 \\
3.0 x_{1}-1.0 x_{2}-1.0 x_{3}+2.0 x_{4} & =-3.0
\end{aligned}
$$

Use the floating point number system $F(10,4,-10,10)$ (i.e., base 10 carrying 4 significant digits) in this question.
(a) What is the coefficient matrix $A$ and the right-hand-side vector $b$ so that the linear system is $A x=b ?$
(b) By hand calculation, carry out Gaussian elimination with row pivoting (sometimes called partial pivoting) on the coefficient matrix $A$ to compute the LU factorization of $A$. In other words, find a permutation matrix $P$, unit lower triangular matrix $L$, and upper triangular matrix $U$ such that $P A=L U$.
(c) Solve the linear system (by hand) using the LU factorization determined in part (b).
(d) Solve the new linear system $A x=c$ where the coefficient matrix $A$ is the same as above but with the new right-hand-side vector

$$
c=\left[\begin{array}{c}
-1.8  \tag{1}\\
-5.1 \\
6.3 \\
-8.4
\end{array}\right]
$$

## 2. (Matlab function myInv)

Write a Matlab function $B=\operatorname{my} \operatorname{Inv}(A)$ which takes as input a matrix $A$. The output from myInv depends on the properties of the input matrix, as follows.

- If $A$ is square and not numerically singular then the output $B$ is the inverse of $A$ computed by Gaussian elimination with row pivoting.
- If $A$ is not square then the output $B$ is a $1 \times 2$ matrix giving the number of rows and columns of A .
- If $A$ is square but numerically singular then the output B is a scalar 0. For deciding whether a square input matrix is numerically singular, use the Matlab command cond with any choice of norm.

For the main computation (when the output B is the inverse of A), compute B by solving the matrix equation $A B=I$ for appropriately sized identity matrix $I$. To solve this system of equations, use the Matlab command $[L, U, P]=l u(A)$ and then use Matlab matrix operations with matrices P, I, L, U to compute B. You are not allowed to use the Matlab command inv for this question.

## Testing your function:

Demonstrate that your function works for the following matrices defined in Matlab: ones $(2,4)$, zeros (3,3), and hilb(k) for $k=7: 2: 13$.

To demonstrate each of these six cases, you should do the following steps in a Matlab session:

- Declare a diary file (see help diary).
- Initalize A to one of the matrices to be demonstrated.
- Call myInv.
- Set diary off.

For your submission, print a listing of your function myInv, and print a copy of your diary file that shows the input and result from myInv for each of the six test cases.

## 3. (Efficient matrix-vector operations)

Let $A$ be an $n \times n$ nonsingular matrix, and let $F$ and $G$ be $n \times p$ matrices, where $1 \leq p \ll n$. You are asked to compute the matrix $W=F G^{T} A^{-1}$. One possible algorithm is the following:

- Compute $V=F G^{T}$.
- Noting that $W=V A^{-1}$, we can write $W A=V$, and by transposing both sides, we have $A^{T} W^{T}=V^{T}$.
- Finally, compute $W$ by the Matlab command: $\mathrm{W}=\left(\mathrm{A}^{\prime} \backslash \mathrm{V}^{\prime}\right)^{\prime}$.

The Matlab command $\mathrm{A} \backslash \mathrm{b}$ (where b is a single vector) requires $\frac{2}{3} n^{3}+O\left(n^{2}\right)$ flops.
(a) Show that this algorithm requires $\frac{8}{3} n^{3}+2 p n^{2}+O\left(n^{2}\right)$ flops.
(b) Suggest a better method that requires only $\frac{2}{3} n^{3}+4 p n^{2}+O\left(n^{2}\right)$ flops.

## 4. (Condition Numbers and Errors)

For an $n \times n$ linear system of equations $A x=b$, the sensitivity of the computed solution $\hat{x}$ to changes in the input (or to roundoff errors during computation) is measured by the condition number $\kappa(A)$. Specifically, if $x_{\text {exact }}$ denotes the exact solution and $\hat{x}$ denotes the solution computed by Gaussian elimination with partial pivoting then, as stated in the Course Notes, the relative error is bounded as follows:

$$
\begin{equation*}
\frac{\left\|x_{\text {exact }}-\hat{x}\right\|}{\|\hat{x}\|} \leq \kappa(A) \epsilon_{\text {machine }} . \tag{2}
\end{equation*}
$$

Any valid norm may be used; for this question use the infinity norm.
Write a Matlab script to test the accuracy of the solutions to the following linear systems:

$$
\begin{equation*}
H_{n} x=b_{n} \tag{3}
\end{equation*}
$$

for $\mathrm{n}=2: 2: 12$, where $H_{n}$ denotes the Hilbert matrix of order $n$ (defined in Matlab by the command hilb(n)), and the order- $n$ vector $b_{n}$ is defined by

$$
\begin{equation*}
b_{n}[i]=\frac{i}{i^{2}+1}, \quad i=1, \ldots, n \tag{4}
\end{equation*}
$$

For each $n$, compute $\hat{x}$ using the Matlab operator $\backslash$. Since the inverse matrix $H_{n}^{-1}$ is known (see the Matlab command $\operatorname{invhilb}(\mathrm{n})$ ), compute $x_{\text {exact }}$ and the actual relative error:

$$
\begin{equation*}
\operatorname{err}_{n}=\frac{\left\|x_{\text {exact }}-\hat{x}\right\|}{\|\hat{x}\|} \tag{5}
\end{equation*}
$$

Compare the actual relative error with the error bound stated above in terms of the condition number $\kappa\left(H_{n}\right)$ which can be obtained via the Matlab command cond. Present the results in a table showing, for each $n, \operatorname{err}_{n}$ and the corresponding theoretical error bound.

## 5. (Google Page Rank)

The objective of this question is to develop Matlab code that computes the Page Rank of a set of web pages based on the network adjacency graph. The adjacency graph is represented by the connectivity matrix which is a sparse matrix $G$ where

$$
G_{i j}= \begin{cases}1 & \text { if } \exists \text { a link from } j \text { to } i \\ 0 & \text { otherwise }\end{cases}
$$

(a) Create a Matlab function with the calling prototype

```
function [p, iters] = PageRank(G, alpha)
```

which finds the steady-state solution (eigenvector for $\lambda=1$ ) of the Page Rank problem using the iterative method discussed in class. The output p is a vector containing the node scores, and iters is the number of iterations the method took to converge.
Your method should take advantage of the sparsity of $G$. That is, at no time should your function create a full matrix of size $R \times R$, where $R$ is the number of nodes. (See the discussion on "Practicalities" in Section 5.6 of the Course Notes.) Terminate the iteration once the solution is found to within a tolerance of $10^{-8}$ (i.e., none of the scores changes by more than the tolerance).
(b) Given the small web shown in the figure below, edit the supplied Matlab script Google.m to create the corresponding sparse matrix $G$ as well as the node labels $U$. The node labels can be stored in a cell array, indexed using curly brackets. For example, you can set the second cell to "b" using $U\{2\}=$ 'b'. Run your PageRank function on the network with an $\alpha$-value of 0.85 . Print out and hand in a bar plot showing the node scores (see the command bar). Which node has the highest score? Lowest score? Write the scores on your plot printout.

(c) An adjacency graph represented by connectivity matrix $G$ and a list of URLs $U$ is given in the file cs.mat. (Download a4_skeleton_code.zip and unzip to obtain the two supplied files cs.mat and Google.m.) Run the Google.m script on the network stored in cs.mat. Try different values of $\alpha$ in the range 0 to 1 . Answer the following questions.
i. What do you notice about the relationship between $\alpha$ and the number of iterations?
ii. The list of the top- 10 web pages changes dramatically with extreme values of $\alpha$ (e.g., 0 versus 1 ). What are the relative advantages and disadvantages of choosing $\alpha$ close to 0 or $\alpha$ close to 1 ?

