## CS 111 (S19): Homework 3

## Due by 6:00pm Monday, April 22

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- 1. How many arithmetic operations (total of additions, subtractions, multiplications, divisions) are required to do each of the following? (You can omit lower-order terms in n.)
  - **1a.** Compute the sum of two *n*-vectors?

n

**1b.** Compute the product of an n-by-n matrix with an n-vector?

$$n*(n+(n-1)) = 2n^2 - n$$

**1c.** Compute the product of two n-by-n matrices?

$$n^2(n + (n-1)) = 2n^3 - n^2$$

**1d.** Solve an *n*-by-*n* upper triangular linear system Ux = y?

$$2\sum_{i=1}^{n-1} i + n = (n-1)n + n = n^2$$

**2.** Suppose A and B are n-by-n matrices, with A nonsingular, and c is an n-vector. Describe the steps you would use to efficiently compute the product  $A^{-1}Bc$ . Describe a different, less efficient, sequence of steps.

Efficient: realizing that  $AA^{-1}Bc = Bc$ , then  $A^{-1}Bc$  is the solution to Ax = Bc, so we can do gaussian elimination to solve x, which is the answer. In this way, we can avoid computing inverse, which take roughtly 3 time more steps than gaussian elimination when matrix is large.

In efficient: inverse A, then calculate  $A^{-1}Bc$ , which take roughly 3 time more steps than previous one. **3.** Suppose that A is a square, nonsingular, nonsymmetric matrix, b is an n-vector, and that you have called

(using the routine from the lecture files). Now suppose you want to solve the system  $A^Tx = b$  (not Ax = b) for x. Show how to do this using calls to Lsolve() and Usolve(), without modifying either of those routines or calling LUfactor() again. Test your method in numpy on a randomly generated 6-by-6 matrix (see np.random.rand()).

```
Get Permutation matrix P from p

PA = LU

A = P^T LU

A^T = U^T L^T P

A^T x = U^T L^T P x = b

let Px = y

A^T x = U^T L^T y = b

therefore, U^T is a lower triangle matrix and L^T is a upper triangle matrix then, y = U solve(L.T, L solve(U.T, b[p])

and x = P.Ty

def getPermutation(p): Penp.zeros((len(p),len(p))).astype('int64') k = 0 for i in p: P[k,i] = 1; k = k+1
```

return P

here, lower p is permuting array, and upper P is permutation matrix that get returned

4. Do problem 2.3 on pages 32–33 of the NCM book, showing the numpy code you use and its output. Note: To understand intuitively what the problem means by "assume that joint 1 is rigidly fixed both horizontally and vertically and that joint 8 is fixed vertically," think of the truss as a (2-dimensional) drawbridge across a river, with the left end being a hinge and the right end lying on the ground.

```
In [79]: a = 1.0/math.sqrt(2)
In [80]: b=np.array([0,10,0,0,0,0,0,15,0,20,0,0,0])
In [82]: A = \text{np.array}([[0,1,0,0,0,-1,0,0,0,0,0,0],
          [0,0,1,0,0,0,0,0,0,0,0,0,0],
          [a,0,0,-1,-a,0,0,0,0,0,0,0,0]
          [a,0,1,0,a,0,0,0,0,0,0,0,0],
          [0,0,0,1,0,0,0,-1,0,0,0,0,0]
          [0,0,0,0,0,0,1,0,0,0,0,0,0]
          [0,0,0,0,a,1,0,0,-a,-1,0,0,0],
          [0,0,0,0,a,0,1,0,a,0,0,0,0],
          [0,0,0,0,0,0,0,0,0,1,0,0,-1],
          [0,0,0,0,0,0,0,0,0,0,1,0,0],
          [0,0,0,0,0,0,0,1,a,0,0,-a,0],
          [0,0,0,0,0,0,0,0,a,0,1,a,0],
          [0,0,0,0,0,0,0,0,0,0,0,a,1]
          ])
In [84]: f=npla.solve(A,b)
In [85]: f=np.round(f,3)
In [86]: f
Out[86]: array([-28.284,
                           20.
                                     10.
                                           , -30.
                                                       14.142,
                                                                 20.
                 -30.
                            7.071,
                                     25.
                                              20.
                                                      -35.355,
                                                                 25.
                                                                       ])
```

## **5.** Consider the linear system

$$\left(\begin{array}{cc} \alpha & 1 \\ 1 & 1 \end{array}\right) \left(\begin{array}{c} x_0 \\ x_1 \end{array}\right) = \left(\begin{array}{c} 3\alpha + 2 \\ 3 \end{array}\right),$$

for some  $\alpha < 1$ . Clearly the solution is  $(x_0, x_1)^T = (1, 2)^T$ . For each value of  $\alpha = 10^{-4}, 10^{-8}, 10^{-16}, 10^{-20}$ , solve this system using the routines LUfactor(), Lsolve(), and Usolve() from LUsolve.ipynb in the lecture files. For each  $\alpha$ , do this twice, first with pivoting = True in LUfactor() and then with pivoting = False. Show your numpy code and its output. Comment on your results.

```
In [69]: def p5(power):
             a = 10 ** power
             A = np.array([[a,1],[1,1]])
             b=np.array([a+2,3])
             LU1 = LUfactor(A, True)
             LU2 = LUfactor(A, False)
             y1 = Lsolve(LU1[0], b[LU1[2]])
             y2 = Lsolve(LU2[0], b[LU2[2]])
             x1 = Usolve(LU1[1], y1)
             x2 = Usolve(LU2[1], y2)
             print("when alpha = 10 to the", power, ":\n", "with pivoting
In [70]: for power in [-4, -8, -16, -20]:
             p5(power)
         when alpha = 10 to the -4:
          with pivoting x = [1. 2.]
          without pivoting x = [1. 2.]
         when alpha = 10 to the -8:
          with pivoting x = [1. 2.]
          without pivoting x = [0.999999999 2.
         when alpha = 10 to the -16:
          with pivoting x = [1. 2.]
          without pivoting x=[4.4408921 2.
                                                   ]
         when alpha = 10 to the -20:
          with pivoting x = [1. 2.]
          without pivoting x = [0. 2.]
In [ ]: #when pivoting is true, the result is more accurate no matter
         #how alpha change since it enhance numerical stability.
         #However, when pivoting is false, the answer start to fluctuate
         #as alpha get smaller because numerical stability is no ensured
```

by selete the pivot with maxium value in the column, we can enhance stability

**6.** Recall that a symmetric matrix A is *positive definite* (SPD for short) if and only if  $x^T A x > 0$  for every nonzero vector x.

**6a.** Find a 2-by-2 matrix A that (1) is symmetric, (2) is not singular, and (3) has all its elements greater than zero, but (4) is not SPD. Show a nonzero vector x such that  $x^T A x < 0$ .

$$A = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}, x = (1, -1), \Rightarrow x^T A x = -2$$

**6b.** Let B be a nonsingular matrix, of any size, not necessarily symmetric. Prove that the matrix  $A = B^T B$  is SPD.

symmetric:  $A^T = (B^T B)^T = B^T (B^T)^T = B^T B = A$  positive definite:  $x^T A x = x^T B^T B x = (Bx)^T (Bx) = y^T y$ , which is a inner product of a non-zero vector y (since B is nonsingular and x is non-zero), always larger than 0.