# HOMEWORK 2

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### Problem 1.

To verify the two matrices are inverses of each other, we can multiply and show the product is  $I_{n\times n}$  identity matrix.

```
i.e. (I+m_ke_k^T)(I-m_ke_k^T)=I (I+m_ke_k^T)(I-m_ke_k^T)=I^2+Ime_k^T-Ime_k^T+(me_k^T)^2\quad \#by\ distribution =I^2+(me_k^T)^2\quad \#by\ cancelling\ like\ terms =I^2\quad \#note\ that\ (me_k^T)^2=(me_k^T)(me_k^T)=0 =I\quad \#note\ the\ identity\ matrix\ multiplied\ by\ itself\ equal\ sitself\ so\ I^2=I L_kM_K=M_kL_k=I Thus, L_k is the inverse of matrix M_k
```

### Problem 2.

```
x = [[0.09 \ 0.38 \ -1.29 \ 0.06 \ 0.05 \ 0.04 \ 0.29 \ -0.49 \ 0.58]]
[[ 0.05
  0.03 -0.04
                     0.
                            0.
                                  0.
         0.01 -0.04 0.16
        0.01 -0.02 -0.07 0.07 0.
-0.01 0. -0.07 0.07 0.
  0.02
 [-0.01 -0.01
                                  0.06 -0.
              0.04 -0.03 -0.05
 [-0.02 0.01
                                 0.08 -0.05 -0.
 [-0.02 0.01 0.01 0.06 -0.04 -0.
         0.01 0.01 0.06 -0.04 -0. -0.01 -0.01
0.16 0.27 0.78 -1.56 1.41 -2.09 -0.37
        -0.48 -1.31 1.46 -4.33
 [-0.
         0.
               0.
                     0.
                                 -0.54 0.61
               -0.
                    -0.
               0.
                                  1.
                                        -1.16
                                               0.03
                           -0.
                    -0.
                     -0.
                           -0.
                                                     0.13]
 [-0.
                                        -0.
```

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```
np.set_printoptions(suppress=True)

def LU(A,b):
    s = A.shape(0)
    t = A.shape(1)
    if s! = t:
        print 'Invalid: Matrix is not square"
    return

p = A.shape(0)
    M = np.ndarray( shape = (p,p,p) )
    L = np.ndarray( shape = (p,p,p) )
    L = np.ndarray( shape = (p,p,p) )

Lfinal = np.identity(p)

Ufinal = np.identity(p)

Will [i,i]=np.float(1)/A[i,i]

for i in range(0,p-1):
    M[i] = np.identity(p)
    Will [i,i]=np.float(1)/A[i,i]

A=M[i].dot(A)
    L[i] = np.linalg.inv(M[i])
    Lfinal = Lfinal.dot(L[i])

Lfinal = np.linalg.inv(K]

print "\n v = ", Ufinal.dot(Lfinal.dot(b))
    print "\n v = ", Ufinal.dot(Lfinal.dot(b))

print "\n v = ", Ufinal.dot(Lfinal.dot(b))

def main():

A = np.matrix([[21. ,32. ,14. ,8. ,6. ,9. ,11. ,3. ,5 ],
    [17. ,2. ,8. ,14. ,55. ,23. ,19. ,1. ,6 ],
    [41. ,23. ,13. ,5. ,11. ,22. ,26. ,7. ,9 ],
    [12. ,11. ,5. ,8. ,3. ,15. ,7. ,25. ,19 ],
    [14. ,7. ,3. ,5. ,11. ,23. ,8. ,7. ,9 ],
    [2. ,8. ,5. ,7. ,1. ,13. ,23. ,11. ,77 ],
    [11. ,7. ,9. ,5. ,3. ,8. ,7. ,9 ],
    [2. ,8. ,5. ,7. ,1. ,13. ,23. ,11. ,77 ],
    [11. ,7. ,9. ,5. ,3. ,8. ,7. ,9 ],
    [23. ,1. ,5. ,19. ,11. ,7. ,9. ,4. ,6 ],
    [31. ,5. ,12. ,7. ,13. ,17. ,24. ,3. ,11]])
    b = np.array([2. ,5. ,7. ,1. ,6. ,9. ,4. ,8. ,3.])
    LU(A,b)

if _name == "__main_":
    main()
```

# Problem 3.

```
From c\|x\|_a \leq \|x\|_b \leq d\|x\|_a we can say c'\|x\|_a \leq \|x\|_b \leq d'\|x\|_a and therfore, \frac{\|Ax\|_a}{\|x\|_a} \leq \frac{d'\|Ax\|_b}{c'\|x\|_b} From these statements, \frac{c'\|Ax\|_b}{d'\|x\|_b} \leq \frac{\|Ax\|_a}{\|x\|_a} \leq \frac{d'\|Ax\|_b}{c'\|x\|_b} Moving forword we use the definition of \|A\|, \|A\|_a = \max_{x \neq 0} \frac{\|Ax\|_a}{\|x\|_a} \leq \max_{x \neq 0} \frac{d'\|Ax\|_b}{c'\|x\|_b} and \|A\|_a = \max_{x \neq 0} \frac{\|Ax\|_a}{\|x\|_a} \geq \max_{x \neq 0} \frac{c'\|Ax\|_b}{d'\|x\|_b} Thus, c'*\max_{x \neq 0} \frac{\|Ax\|_a}{\|x\|_a} \leq \max_{x \neq 0} \frac{\|Ax\|_b}{\|x\|_b} \leq d'*\max_{x \neq 0} \frac{\|Ax\|_a}{\|x\|_a} \Longrightarrow c'\|A\|_a \leq \|A\|_b \leq d'\|A\|_a
```

Therefore, the induced matrix norms are, in fact, also equavalent.

### Problem 4.

HOMEWORK 2 3

After 3 repeats, no more improvement is observable.

\*full code in 4.py\*

```
def solution(A, b, x):
    n=A.shape[0]
    bCopy = deepcopy(b)
    gaussian(np.matrix.copy(A),bCopy,x)

r = np.zeros(n)
    thisX = np.zeros(n)
    newX = np.zeros(n)
    for i in range(0,n):
        thisX[i] = x[i]
    print "x = ", x, "\n\n"

for temp in range(0,3): #no change observed after 3
    aCopy = np.matrix.copy(A)
    Ax = aCopy.dot(thisX)
    for i in range(0, n):
        r[i] = np.float32(b[i] - Ax[0,i])

    print "r = ", r
    z = np.zeros(n)
    gaussian(aCopy, r, z)

    for add in range(0, n):
        newX[add] = thisX[add] + z[add]
    print "Improved: ",newX

    print "\n\n"
    for i in range(0, n):
        thisX[i] = newX[i]

def main():
    A = np.matrix( [[21.0, 67.0, 88.0, 73.0],
        [76.0, 63.0, 7.0, 20.0],
        [0.0, 85.0, 56.0, 54.0],
        [19.3, 43.0, 30.2, 29.4]])

b = np.array([141.0, 109.0, 218.0, 93.7])
    x = np.zeros(4)
    solution(A, b, x)

if __name__ == "__main__":
    main()
```

## Problem 5.

As k increases from 1-10, the value of  $\varepsilon$  decreases and the computed solution becomes less accurate.

HOMEWORK 2

5