HW04

November 3, 2015

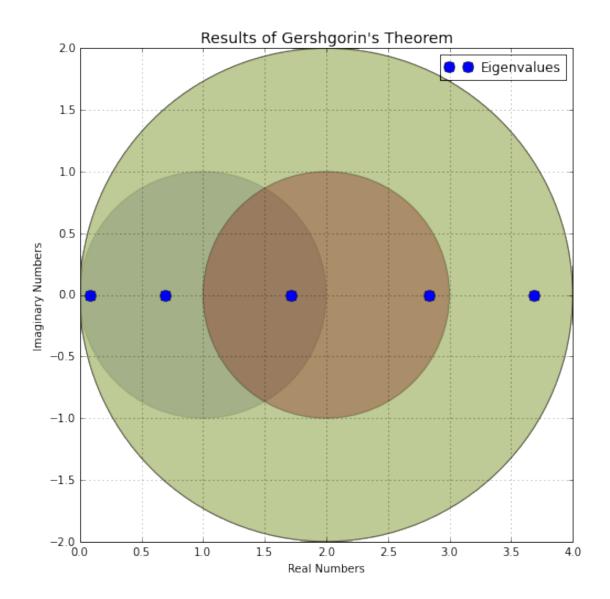
0.1 ME 500 - Assignment 4 - Brandon Lampe

-3.331e-16+0.j]

```
In [3]: from scipy import linalg as LA
        from scipy.sparse import diags as diags
        import numpy as np
        import scipy as sp
        from matplotlib import pyplot as plt
        import sys
        sys.path.append('/Users/Lampe/PyScripts')
        import blfunc as bl
        import ipdb
        np.set_printoptions(precision=3, suppress=False) # precision for numpy operations
        %precision 3
        %matplotlib inline
0.2 Problem 3
In [4]: A = diags([-1,2,-1],[-1,0,1], shape=(5,5)).toarray()
        A[0,0] = 1
       print A
[[ 1. -1. 0. 0. 0.]
 [-1. 2. -1. 0. 0.]
 [ 0. -1. 2. -1. 0.]
 [ 0. 0. -1. 2. -1.]
 [ 0. 0. 0. -1. 2.]]
0.2.1 3 (a)
Finde the eigenvalues and eigenvectors of [A]
In [5]: eig, Mo = LA.eig(A)
        eig_min = np.real(min(eig))
        eig_max = np.real(max(eig))
In [6]: eig_diag = eig * np.eye(5)
  check to ensure [A] - \lambda_1[I] = 0
In [7]: print (A - eig[0]*np.eye(5)).dot(Mo[:,0])
[ 9.992e-16+0.j
                   1.554e-15+0.j
                                   5.551e-16+0.j
                                                   1.665e-15+0.j
```

```
(i)
In [8]: print np.real(eig_diag)
[[ 3.683 0.
                       0.
                             0.
                0.
         2.831 0.
                       0.
ГО.
                             0.
[ 0.
                0.081 0.
                             0.
                                  1
         0.
ГО.
         0.
                0.
                      1.715 0.
                                  ٦
                             0.69 11
Γ0.
         0.
                0.
                       0.
In [33]: print Mo
[[-0.17 -0.326 -0.597 0.456 -0.549]
[ 0.456  0.597 -0.549 -0.326 -0.17 ]
[-0.597 -0.17 -0.456 -0.549 0.326]
 [ 0.549 -0.456 -0.326 0.17
                             0.597]
[-0.326 0.549 -0.17
                      0.597 0.456]]
(ii)
In [34]: MoT = np.transpose(Mo)
        print Mo.dot(MoT)
[[ 1.000e+00
              5.013e-16
                         8.546e-17 -3.753e-16 -2.776e-17]
[ 5.013e-16
             1.000e+00 4.905e-16 -2.285e-16
                                                1.943e-16]
[ 8.546e-17
              4.905e-16
                         1.000e+00 -3.905e-16
                                                 3.053e-16]
[ -3.753e-16 -2.285e-16 -3.905e-16
                                     1.000e+00 -5.551e-17]
[ -2.776e-17
              1.943e-16 3.053e-16 -5.551e-17
                                                 1.000e+00]]
(iii)
In [35]: A_star = Mo.dot(eig_diag).dot(MoT)
        print A_star
[[ 1.000e+00+0.j -1.000e+00+0.j 7.494e-16+0.j -1.110e-15+0.j
   -8.327e-17+0.j]
 [ -1.000e+00+0.j
                  2.000e+00+0.j -1.000e+00+0.j -8.743e-16+0.j
   1.270e-15+0.j]
 [ 7.494e-16+0.j -1.000e+00+0.j
                                  2.000e+00+0.j -1.000e+00+0.j
   1.457e-15+0.j]
 [ -1.055e-15+0.j -8.743e-16+0.j -1.000e+00+0.j
                                                 2.000e+00+0.j
  -1.000e+00+0.j]
 [ -1.943e-16+0.j
                   1.270e-15+0.j 1.402e-15+0.j -1.000e+00+0.j
   2.000e+00+0.j]]
(iv)
In [36]: A_star_inv = Mo.dot(LA.inv(eig_diag)).dot(MoT)
        print A_star_inv
[[5.+0.j 4.+0.j 3.+0.j 2.+0.j 1.+0.j]
[4.+0.j 4.+0.j 3.+0.j 2.+0.j 1.+0.j]
[3.+0.j 3.+0.j 3.+0.j 2.+0.j 1.+0.j]
[2.+0.j 2.+0.j 2.+0.j 2.+0.j 1.+0.j]
 [ 1.+0.j 1.+0.j 1.+0.j 1.+0.j 1.+0.j]
```

```
(v)
In [37]: print A_star_inv.dot(A_star)
[[ 1.000e+00+0.j -3.469e-17+0.j
                                    6.509e-15+0.j -1.166e-14+0.j
    7.772e-15+0.j]
                                    6.287e-15+0.j -1.177e-14+0.j
 [ -9.437e-16+0.j
                    1.000e+00+0.j
   8.660e-15+0.j]
                                    1.000e+00+0.j -9.437e-15+0.j
 [ -1.388e-15+0.j
                   2.186e-15+0.j
   7.772e-15+0.j]
                                    2.734e-15+0.j 1.000e+00+0.j
 [ -8.049e-16+0.j -2.567e-16+0.j
    4.663e-15+0.j]
                                    1.735e-15+0.j -3.664e-15+0.j
 [ -7.216e-16+0.j 9.506e-16+0.j
    1.000e+00+0.j]]
0.2.2 3 (b)
Gersgorin's theorem to obtain bounds on eigenvalues of [A]
In [38]: nrow = A.shape[0]
        ncol = A.shape[1]
         center = np.diagonal(A)
         radius = np.zeros(ncol)
         for i in xrange(nrow):
            for j in xrange(ncol):
                 if i != j:
                     radius[i] = radius[i] + np.abs(A[i,j])
         fig_gersh, ax = plt.subplots(figsize = (8,8))
         ax.plot(eig, np.zeros(nrow), 'o', markersize = 10, label="Eigenvalues")
         ax.legend(loc=0); # upper left corner
         ax.set_xlabel('Real Numbers')
         ax.set_ylabel('Imaginary Numbers')
         ax.set_title('Results of Gershgorin\'s Theorem', fontsize = 14)
         ax.grid(b = True, which = 'minor')
         ax.grid(b = True, which = 'major')
         ax.set_ylim(-2, 2)
         ax.set_xlim(0,4)
         bl.circles(x=center, y=np.zeros(nrow), s=radius, c=np.arange(nrow), ax=ax, alpha=0.3)
         fig_name = 'plot_3b.pdf'
         path = '/Users/Lampe/Documents/UNM_Courses/ME-500/HW04/'
         fig_gersh.savefig(path + fig_name)
         # show()
```



0.2.3 3 (c)

Obtain the Rayleigh quotient for $\langle v \rangle = \langle 1, 2, 3, 2, 1 \rangle$

Out[40]: 0.263

0.2.4 3 (d)

Compute Norms

```
In [41]: p1 = np.linalg.norm(A, ord = 1)
         p2 = np.linalg.norm(A, ord = 2)
         pInf = np.linalg.norm(A, ord = np.inf)
         print '%.3e' %p1
         print '%.3e' %p2
         print '%.3e' %pInf
4.000e+00
3.683e+00
4.000e+00
In [42]: print '%.3e' %np.real(eig_min)
         print '%.3e' %np.real(eig_max)
8.101e-02
3.683e+00
0.2.5 3 (e)
The condition number of [A]
In [43]: cond = eig_max / eig_min
         print '%.3e' %cond
4.546e+01
0.2.6 3 (f)
Obtain a 1, 2, and 3-mode solution
In [44]: x_ex = np.array([1,2,3,4,5])
         b_ex = A.dot(x_ex)
         print b_ex
[-1. 0. 0. 0. 6.]
  approximations for \{x^{ex}\} using all 5 modes of [A]
In [45]: nrow = eig.shape[0]
         eig_vect_arr = np.zeros((nrow,nrow))
         eig_val_arr = np.zeros(nrow)
         A_inv_ap = np.zeros((nrow,nrow))
         x_ap = np.zeros((nrow,nrow))
         error = np.zeros(nrow)
         for i in xrange(nrow):
             A_inv_ap = 1./np.real(eig[i]) * np.outer(Mo[:,i],Mo[:,i]) + A_inv_ap
             x_ap[:,i] = A_inv_ap.dot(b_ex)
             error[i] = np.linalg.norm(x_ap[:,i] - x_ex,2)/np.linalg.norm(x_ex,2)
         print x_ap
         print error
[[ 0.082 -0.334 2.778 3.609 1.
 [-0.221 0.542 3.402 2.808 2.
                                    1
 [ 0.29  0.072  2.449  1.449  3.
                                    1
 [-0.266 -0.848 0.852 1.161 4.
                                    ٦
[ 0.158  0.859  1.745  2.833  5.
                                    ]]
[ 9.979e-01 9.829e-01 6.867e-01
                                       6.413e-01 7.400e-15]
```

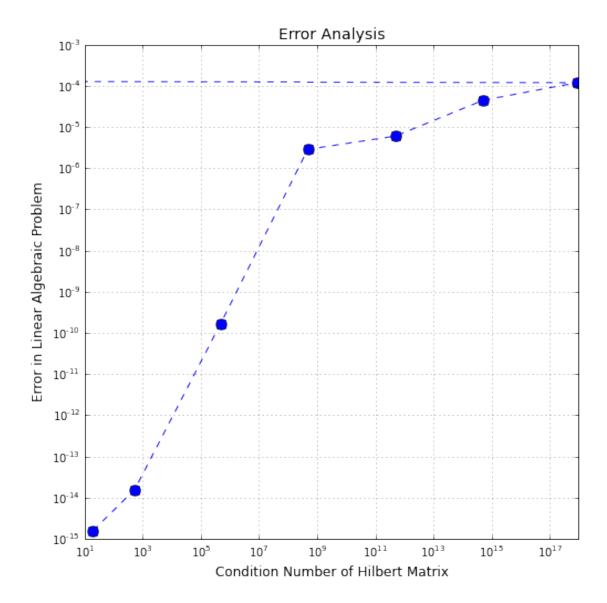
Gram-Schmidt method for obtaining eigenpairs

```
In [46]: #G-S for eigenpairs
         eig_vect_old = np.array([2,2,2,2,5])
         eig_val_old = eig_vect_old.dot(A).dot(eig_vect_old)/(eig_vect_old.dot(eig_vect_old))**0.5
         A_{inv_ap_old} = np.zeros((5,5))
         x_ap_arr = np.zeros((5,5))
         eig_vect_arr = np.zeros((5,5))
         eig_val_arr = np.zeros(5)
         alpha_arr = np.zeros(5)
         error_vect = np.zeros(5)
         neg_terms = np.zeros(5)
         tol = 0.001
         iter_max = 100
         error = 10
         mode_max = x_ex.shape[0]
         for h in xrange(mode_max):
             if h == 0:
                 for i in xrange(iter_max): # obtain lowest eigenpair by reverse iteration
                     x_star = bl.QR_solve(A,eig_vect_old)
                     eig_vect_new = x_star / np.sqrt(x_star.dot(x_star))
                     eig_val_new = (eig_vect_new.dot(A).dot(eig_vect_new))/den
                     error = np.abs(eig_val_new - eig_val_old)/np.abs(eig_val_new)
                     beta = eig_vect_new.dot(b_ex)
                     x_ap = beta/eig_val_new * eig_vect_new
                     eig_vect_old = eig_vect_new
                     eig_val_old = eig_val_new
                     if error <= tol:</pre>
                         eig_vect_arr[:,h] = eig_vect_new
                         eig_val_arr[h] = eig_val_new
                         break
             else:
                 for i in xrange(iter_max):
                     x_hat = bl.QR_solve(A,eig_vect_old) # x hat
                     alpha_arr[h-1] = eig_vect_arr[:,h-1].dot(x_hat)
                     neg_terms = 0
                     for j in xrange(mode_max):
                         neg_terms = neg_terms + alpha_arr[j]*eig_vect_arr[:,j]
         #
                           print neg_terms
                     x_star = x_hat - neg_terms # apply G-S, corrected for previous eigenvectors
                     eig_vect_new = x_star / np.sqrt(x_star.dot(x_star))
                     den = eig_vect_new.dot(eig_vect_new) # D
                     eig_val_new = (eig_vect_new.dot(A).dot(eig_vect_new))/den
                     error = np.abs(eig_val_new - eig_val_old)/np.abs(eig_val_new)
```

```
eig_vect_old = eig_vect_new
                     eig_val_old = eig_val_new
                     if error <= tol:</pre>
                         eig_vect_arr[:,h] = eig_vect_new
                         eig_val_arr[h] = eig_val_new
                         break
            for k in xrange(h+1):
                 beta = eig_vect_arr[:,k].dot(b_ex)
                 x_ap = beta/eig_val_arr[k] * eig_vect_arr[:,k]
                 x_ap_arr[:,h] = x_ap_arr[:,h] + x_ap
            error_vect[h] = np.linalg.norm(A.dot(x_ap_arr[:,h]) - b_ex,2)/np.linalg.norm(b_ex,2)
         print eig_vect_arr
         print eig_val_arr
        print x_ap_arr
        print error_vect
[[ 0.596  0.544  -0.596  -0.556  0.593]
 [ 0.548  0.173  -0.548  -0.172  0.547]
 [ 0.456 -0.32 -0.456 0.325 0.458]
 [ 0.327 -0.598 -0.327 0.593 0.33 ]
 [ 0.17 -0.462 -0.17  0.451  0.173]]
[ 0.004 0.69 0.081 0.69 0.081]
[[ 59.545 56.928 60.057 57.43 60.693]
 [ 54.748 53.914 56.792 55.977 58.988]
 [ 45.527 47.064 49.457 50.995 53.515]
 [ 32.605 35.479 37.192 39.996 41.813]
 [ 17.008 19.23 20.123 22.254 23.207]]
[ 1.603  1.508  1.567  1.663  1.724]
0.2.7 4
Analysis of Hilbert Matrices
In [98]: rng = np.array((2,3,5,7,9,11,13,15))
         analysis = np.zeros((rng.shape[0], 2))
         inc = 0
        for i in rng:
            H = LA.hilbert(i)
            val, Mo = LA.eig(H)
            text = "Size of Hilber Matix: " + str(i)
            print text
            print "Eigenvalues:"
            print np.real(val)
            c = max(np.real(val))/min(np.real(val))
            print 'Condition Number'
            print '%e' %c
            x_ex = np.arange(i)
            b = H.dot(x_ex)
            x_ap = bl.QR_solve(H,b)
```

```
error = np.linalg.norm(H.dot(x_ap) - b)/np.linalg.norm(b)
            print "approximate solution and error"
            out = str(x_ap) + " error: " + str(error)
            print out
            print
            analysis[inc,0] = c
            analysis[inc,1] = error
            inc = inc + 1
Size of Hilber Matix: 2
Eigenvalues:
[ 1.268 0.066]
Condition Number
1.928147e+01
approximate solution and error
[ -6.752e-15
             1.000e+00] error: 1.57039437051e-15
Size of Hilber Matix: 3
Eigenvalues:
[ 1.408  0.122  0.003]
Condition Number
5.240568e+02
approximate solution and error
[ -1.168e-12
             1.000e+00
                          2.000e+00] error: 1.56539504238e-14
Size of Hilber Matix: 5
Eigenvalues:
              2.085e-01
「 1.567e+00
                          1.141e-02
                                      3.059e-04
                                                  3.288e-06]
Condition Number
4.766073e+05
approximate solution and error
[ -1.353e-06
             1.000e+00
                          2.000e+00
                                                  4.000e+00] error: 1.67635118738e-10
                                      3.000e+00
Size of Hilber Matix: 7
Eigenvalues:
[ 1.661e+00
              2.719e-01
                          2.129e-02 1.009e-03
                                                  2.939e-05 4.857e-07
   3.494e-09]
Condition Number
4.753674e+08
approximate solution and error
[ 1.096e+00 -4.410e+01
                          4.460e+02 -1.751e+03
                                                  3.258e+03 -2.834e+03
  9.455e+02] error: 2.84920789322e-06
Size of Hilber Matix: 9
Eigenvalues:
[ 1.726e+00
              3.216e-01
                          3.104e-02
                                      1.979e-03
                                                  8.758e-05
                                                              2.673e-06
  5.386e-08
              6.461e-10
                          3.500e-12]
Condition Number
4.931537e+11
approximate solution and error
[ 1.602e+00 -6.010e+01
                          5.751e+02 -2.201e+03
                                                  4.080e+03 -3.664e+03
   1.378e+03 -6.298e+01 -9.741e+00] error: 5.95702344812e-06
Size of Hilber Matix: 11
```

```
Eigenvalues:
                          4.031e-02
[ 1.775e+00
              3.624e-01
                                      3.114e-03
                                                 1.774e-04
                                                             7.542e-06
  2.372e-07
              5.368e-09
                          8.283e-11 7.807e-13
                                                 3.399e-15]
Condition Number
5.221964e+14
approximate solution and error
9.610e+00 -3.522e+02
                          3.237e+03 -1.227e+04
                                                 2.260e+04 -2.046e+04
  7.845e+03 -4.589e+02 -6.220e+01 -1.788e+01 -7.197e+00] error: 4.43459592315e-05
Size of Hilber Matix: 13
Eigenvalues:
                                      4.349e-03
[ 1.814e+00
              3.968e-01
                          4.903e-02
                                                 2.952e-04
                                                             1.562e-05
                                                 1.144e-13
  6.466e-07
              2.076e-08
                          5.077e-10
                                      9.141e-12
                                                             8.892e-16
  2.042e-18]
Condition Number
8.883830e+17
approximate solution and error
[ 3.078e+01 -1.174e+03
                         1.144e+04 -4.740e+04 9.904e+04 -1.087e+05
  5.863e+04 -1.129e+04 -3.642e+02 -7.949e+01 -2.795e+01 -1.262e+01
 -6.671e+001 error: 0.000118244924399
Size of Hilber Matix: 15
Eigenvalues:
「 1.846e+00
              4.266e-01
                          5.721e-02
                                      5.640e-03
                                                 4.365e-04
                                                             2.711e-05
                                      4.658e-11
                                                 9.322e-13 1.394e-14
  1.362e-06 5.529e-08 1.803e-09
  1.421e-16 8.051e-18 -1.052e-17]
Condition Number
-1.754630e+17
approximate solution and error
[ 1.188e+02 -4.647e+03
                          4.676e+04 -2.020e+05
                                                 4.444e+05 -5.208e+05
  3.074e+05 -6.961e+04 -1.102e+03 -2.294e+02 -7.887e+01 -3.508e+01
 -1.834e+01 -1.071e+01 -6.791e+00] error: 0.000364001386973
In [99]: fig_Hilb, ax = plt.subplots(figsize = (8,8))
        ax.plot(analysis[:,0],analysis[:,1], 'o--', markersize=10)
        # ax.legend(loc=0); # upper left corner
        ax.set_xlabel('Condition Number of Hilbert Matrix', fontsize = 12)
        ax.set_ylabel('Error in Linear Algebraic Problem', fontsize = 12)
        ax.set_title('Error Analysis' , fontsize = 14)
        ax.grid(b = True, which = 'major')
        ax.grid(b = True, which = 'major')
        \# ax.set\_ylim(-2, 2)
        # ax.set_xlim(0,4)
        ax.set_xscale('log')
        ax.set_yscale('log')
        fig_name = 'plot_4.pdf'
        path = '/Users/Lampe/Documents/UNM_Courses/ME-500/HW04/'
        fig_gersh.savefig(path + fig_name)
        # show()
```



In []: