

ELEC 405 - Homework 5

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1. Quadratic Forms and Hermitian Matrices

1.

$$A = \frac{1}{9} \begin{bmatrix} -2 & 2 & 8 \\ 2 & 7 & 10 \\ 8 & 10 & 4 \end{bmatrix} \quad (1)$$

2. Let $A = LDL^*$ where L is a lower triangular matrix and D is a diagonal matrix.

$$\begin{aligned} L &= \begin{bmatrix} 1 & 0 & 0 \\ a & 1 & 0 \\ b & c & 1 \end{bmatrix} \\ D &= \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix} \\ A &= LDL^* \\ &= \begin{bmatrix} 1 & 0 & 0 \\ a & 1 & 0 \\ b & c & 1 \end{bmatrix} \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix} \begin{bmatrix} 1 & a & b \\ 0 & 1 & c \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} d_1 & ad_1 & bd_1 \\ ad_1 & a^2d_1 + d_2 & abd_1 + cd_2 \\ bd_1 & abd_1 + cd_2 & b^2d_1 + c^2d_2 + d_3 \end{bmatrix} \end{aligned} \quad (2)$$

$$\begin{aligned}
d_1 &= -\frac{2}{9} & a^2d_1 + d_2 &= \frac{7}{9} \Rightarrow d_2 = 1 \\
ad_1 &= \frac{2}{9} \Rightarrow a = -1 & abd_1 + cd_2 &= \frac{10}{9} \Rightarrow c_2 = 2 \\
bd_1 &= \frac{8}{9} \Rightarrow b = -4 & b^2d_1 + c^2d_2 + d_3 &= \frac{4}{9} \Rightarrow d_3 = 0
\end{aligned} \tag{3}$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ -4 & 2 & 1 \end{bmatrix}, D = \begin{bmatrix} -\frac{2}{9} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \tag{4}$$

There is also another method: first getting LU decomposition of A and then extracting the diagonal elements to a diagonal matrix D , this method is also tried and can be seen in the Appendix.

The inertia of A is $(1, 1, 1)$ because A is star-congruent to D in (4) so they must have the same inertia.

3. No, A is indefinite because it has both negative and positive eigenvalues.

4. The eigenvalues determined by the Julia Notebook (see Appendix) are as follows:

$$-1.0000000000000004, -4.764101178639501e - 17, 1.9999999999999998$$

Dividing by 9 causes overflow in Julia and therefore the second eigenvalue is actually 0.

2. Matrix Factorization Coding Experience...

See the Appendix.

3. Random Vector Generation...

See the Appendix.

4. Positive matrices

Eigenvectors of a positive definite Hermitian matrix A form an orthonormal basis for \mathcal{C}^n . So that we can write x in terms of the eigen basis.

$$\begin{aligned}
x &= \sum_{i=1}^n \beta_i v_i \\
\|x\|^2 &= \sum_{i=1}^n \beta_i \times \beta_i^* \|v_i\|^2 = \sum_{i=1}^n |\beta_i|^2
\end{aligned} \tag{5}$$

$$\begin{aligned}
x^* Ax &= \sum_{i=1}^n \sum_{j=1}^n \beta_i \times \beta_j v_i^* A v_j = \sum_{i=1}^n \sum_{j=1}^n \beta_i \times \beta_j \lambda_i v_i^* v_j \\
&= \sum_{i=1}^n \lambda_i |\beta_i|^2
\end{aligned} \tag{6}$$

$$(6) \geq \lambda_{\min} \times (5) \Rightarrow$$

5. Quadratic approximations of multivariable functions

1. Gradient:

$$\nabla f([x_1 x_2]^T) = e^{-x_1^2 - x_2^2} \begin{bmatrix} -2x_1(4x_1^2 - 5) \\ (2 - 8x_1^2)x_2 \end{bmatrix} \tag{7}$$

Hessian:

$$H([x_1 x_2]^T) = e^{-x_1^2 - x_2^2} \begin{bmatrix} -32x_1^2 + 4(4x_1^2 - 1)x_1^2 + 8 - 2(4x_1^2 - 1) & 4x_1(4x_1^2 - 1)x_2 - 16x_1x_2 \\ 4x_1(4x_1^2 - 1)x_2 - 16x_1x_2 & 4(4x_1^2 - 1)x_2^2 - 2(4x_1^2 - 1) \end{bmatrix} \tag{8}$$

2. See the Appendix.
3. See the Appendix.

Appendices

Question 1

```
In [1]: using LinearAlgebra
        using Plots
```

```
In [2]: A = [-2 2 8; 2 7 10; 8 10 4]
        A
```

```
Out[2]: 3x3 Array{Int64,2}:
         -2  2  8
          2  7 10
          8 10  4
```

Method 1

```
In [3]: l(a, b, c) = [1 0 0; a 1 0; b c 1]
        d(a, b, c) = [a 0 0; 0 b 0; 0 0 c]
```

```
Out[3]: d (generic function with 1 method)
```

```
In [4]: L = l(-1, -4, 2)
        D = d(-2/9, 1, 0)
        L * D * L'
```

```
Out[4]: 3x3 Array{Float64,2}:
         -0.222222  0.222222  0.888889
          0.222222  0.777778  1.11111
          0.888889  1.11111  0.444444
```

```
In [5]: A/9
```

```
Out[5]: 3x3 Array{Float64,2}:
         -0.222222  0.222222  0.888889
          0.222222  0.777778  1.11111
          0.888889  1.11111  0.444444
```

Method 2

Use LU Decomposition to construct LDU decomposition

```
In [6]: E1 = [-9/2 0 0; 1 1 0; 4 0 1]
        1/9*E1*A
```

```
Out[6]: 3x3 Array{Float64,2}:
         1.0  -1.0  -4.0
          0.0   1.0   2.0
          0.0   2.0   4.0
```

```
In [7]: E2 = [1 0 0; 0 1 0; 0 -2 1]
        U = E2*E1*A
        U /= 9
```

```
Out[7]: 3x3 Array{Float64,2}:
         1.0  -1.0  -4.0
          0.0   1.0   2.0
          0.0   0.0   0.0
```

```
In [8]: inv(E2*E1)
```

```
Out[8]: 3x3 Array{Float64,2}:
 -0.222222  0.0  0.0
  0.222222  1.0  0.0
  0.888889  2.0  1.0
```

$$L = (E2 \times E1)^{-1}$$

Then we need to extract the first diagonal of L into a diagonal matrix

```
In [9]: L0 = [-2/9  0.0  0.0;
              2/9  1.0  0.0;
              8/9  2.0  1.0]
```

```
Out[9]: 3x3 Array{Float64,2}:
 -0.222222  0.0  0.0
  0.222222  1.0  0.0
  0.888889  2.0  1.0
```

Now we need to extract the $-\frac{2}{9}$ term to D , which will multiply the first column with $-\frac{2}{9}$ when $L \times D$ is applied

```
In [10]: L = [1  0.0  0.0;
              -1  1.0  0.0;
              -4  2.0  1.0]
```

```
Out[10]: 3x3 Array{Float64,2}:
 1.0  0.0  0.0
 -1.0 1.0  0.0
 -4.0 2.0  1.0
```

```
In [11]: D = [-2/9  0  0;
              0  1  0;
              0  0  0]
```

```
Out[11]: 3x3 Array{Float64,2}:
 -0.222222  0.0  0.0
  0.0       1.0  0.0
  0.0       0.0  0.0
```

```
In [12]: L * D * L'
```

```
Out[12]: 3x3 Array{Float64,2}:
 -0.222222  0.222222  0.888889
  0.222222  0.777778  1.11111
  0.888889  1.11111  0.444444
```

```
In [13]: A/9
```

```
Out[13]: 3x3 Array{Float64,2}:
 -0.222222  0.222222  0.888889
  0.222222  0.777778  1.11111
  0.888889  1.11111  0.444444
```

inertia

```
In [14]: eig = eigvals(A/9)
```

```
Out[14]: 3-element Array{Float64,1}:
 -1.0000000000000004
 -4.764101178639501e-17
 1.9999999999999998
```

```
In [15]: E = eigvecs(A/9)
```

```
Out[15]: 3x3 Array{Float64,2}:  
  -0.666667  0.666667 -0.333333  
  -0.333333 -0.666667 -0.666667  
   0.666667  0.333333 -0.666667
```

```
In [62]: e3 = E[:, 3]
```

```
Out[62]: 3-element Array{Float64,1}:  
 -0.333333333333333326  
 -0.6666666666666664  
 -0.6666666666666669
```

```
In [63]: eig[3]*e3
```

```
Out[63]: 3-element Array{Float64,1}:  
 -0.6666666666666664  
 -1.3333333333333326  
 -1.3333333333333335
```

```
In [64]: 1/9*A*e3
```

```
Out[64]: 3-element Array{Float64,1}:  
 -0.6666666666666667  
 -1.3333333333333335  
 -1.3333333333333333
```

Question 2

Finding Cholesky Factorization of a Matrix

```
In [168]: function cholesky_factorization(A)  
           n = size(A)[1]  
           α = sqrt(Complex(A[1]))  
           if n == 1  
               return α  
           end  
           v = A[2:n, 1]/α  
           return [α (zeros(n-1)+im*zeros(n-1))'; v cholesky_factorization(A[2:n, 2:n] - v*v')]  
       end
```

```
Out[168]: cholesky_factorization (generic function with 1 method)
```

```
In [146]: n = 3  
CL_real = [2 0 0 ; 1 3 0; 2 1 3]  
C_real = CL_real * CL_real'  
eigvals(C_real)
```

```
Out[146]: 3-element Array{Float64,1}:  
  2.586414898264002  
  6.690759634051093  
 18.722825467684906
```

```
In [172]: cholesky_factorization(C_real) == CL_real
```

```
Out[172]: true
```

```
In [139]: n = 3  
CL = [2 0 0; 1+.5im 3 0; 2+im 1-4im 3]  
C = CL * CL'  
eigvals(C)
```

```
Out[139]: 3-element Array{Float64,1}:  
  1.4818827284435265  
  5.751134554403829  
 38.01698271715264
```

Check if the factorization method is correct

```
In [170]: cholesky_factorization(C)
```

```
Out[170]: 3x3 Array{Complex{Float64},2}:
 2.0+0.0im  0.0-0.0im  0.0-0.0im
 1.0+0.5im  3.0+0.0im  0.0-0.0im
 2.0+1.0im  1.0-4.0im  3.0+0.0im
```

```
In [171]: cholesky_factorization(C) == CL
```

```
Out[171]: 3x3 BitArray{2}:
 1  1  1
 1  1  1
 1  1  1
```

```
In [173]: # built-in cholesky
cholesky(C)
```

```
Out[173]: Cholesky{Complex{Float64},Array{Complex{Float64},2}}
U factor:
3x3 UpperTriangular{Complex{Float64},Array{Complex{Float64},2}}:
 2.0+0.0im  1.0-0.5im  2.0-1.0im
      .      3.0+0.0im  1.0+4.0im
      .      .      3.0+0.0im
```

Question 3

```
In [228]: function generate_rand_vec(N, R)
           V = randn(3, N)
           R_L = cholesky_factorization(R)
           vectors = zeros(3, N) + zeros(3, N) * im
           for i in 1:N
               y_i = R_L * (V[:, i])
               # since mean is zero, we don't need to define x_i = y_i + m_x
               vectors[:, i] = y_i
           end
           return vectors
       end
```

```
Out[228]: generate_rand_vec (generic function with 1 method)
```

```
In [231]: R = [28 15+9im 2+21im; 15-9im 48 15-11im; 2-21im 15+11im 30]
```

```
Out[231]: 3x3 Array{Complex{Int64},2}:
 28+0im  15+9im  2+21im
 15-9im  48+0im  15-11im
 2-21im  15+11im  30+0im
```

```
In [232]: N = 100000
vectors = generate_rand_vec(N, R)
```

```
Out[232]: 3x100000 Array{Complex{Float64},2}:
 5.89581+0.0im  -3.40973+0.0im  ...  -7.7226+0.0im
 5.64755-1.89508im  -5.08827+1.09598im  0.309757+2.48226im
 1.54956-2.97109im  -0.485671+0.656249im  0.173862+8.38381im
```

```
In [233]: r = zeros(3, 3) + im*zeros(3, 3)
for j in 1:N
    v = vectors[:, j]
    r += v*v'
end
R_hat = r/N
R_hat
```

```
Out[233]: 3x3 Array{Complex{Float64},2}:
 28.0487+0.0im      15.0907+9.01565im   2.01648+20.9989im
 15.0907-9.01565im   48.317+0.0im       15.064-11.0908im
 2.01648-20.9989im   15.064+11.0908im   30.0612+0.0im
```

- It seems like it is not likely to find \hat{R} without using for loops
- \hat{R} is close to R .
- My random generator seems to be working fine.

Question 5

```
In [39]: f(x, y) = (4*x^2-1)*exp(-x^2-y^2)
f(x) = f(x[1], x[2])

ex(x_1, x_2) = exp(-x_1^2-x_2^2)
ex(x) = ex(x[1], x[2])

∇f(x_1, x_2) = [(10*x_1-8x_1^2)*ex(x_1,x_2); -2*x_2*ex(x_1, x_2)]
∇f(x) = ∇f(x[1], x[2])

H(x_1, x_2) = ex(x_1, x_2)*
  [(-32*x_1^2 + 4*(4 *x_1^2 - 1)*x_1^2 + 8 - 2*(4* x_1^2 - 1)) (4 * x_1 * (4 * x_1^2 - 1) * x_2 - 16
*x_1 *x_2);
  (4 * x_1 * (4 * x_1^2 - 1)*x_2 - 16 *x_1 *x_2) (4 *(4* x_1^2 - 1) *x_2^2 - 2 *(4* x_1^2 - 1))]
H(x) = H(x[1], x[2])

f_q(x0, x) = f(x) + dot(∇f(x0), (x-x0)) + 0.5*dot(H(x0)*(x-x0), (x-x0))
f_q(x) = f_q([0; 0], x)
fq(x_1, x_2) = f_q([x_1; x_2])
```

```
Out[39]: fq (generic function with 1 method)
```

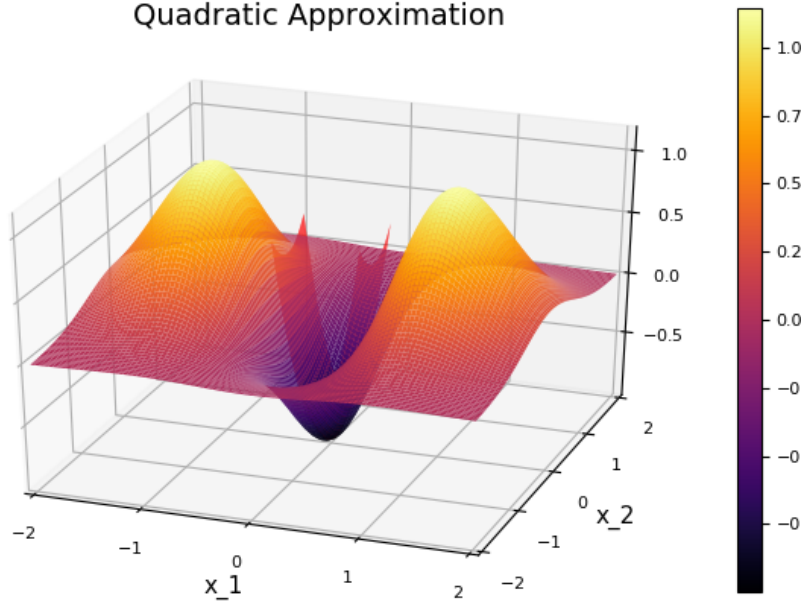


```
In [45]: camera = (20, 30)
pyplot()
x_1 = range(-2, stop=2, length=100)
x_2 = range(-2, stop=2, length=100)
p = plot(x_1, x_2, f, title="Quadratic Approximation",
         st=:surface, camera=camera, xlabel="x_1", ylabel="x_2")

xq_1 = range(-.4, stop=.4, length=100)
xq_2 = range(-.4, stop=.4, length=100)
plot!(p, xq_1, xq_2, fq, st=:surface, camera=camera, color=:red)
```

Out[45]:

Quadratic Approximation



Question 5.c

1. Find an appropriate point.
2. Plot it

```
In [44]: for i in x_1
        for j in x_2
            hess = H(i, j)
            if all(x -> x< 0, eigvals(hess)) && ∇f(i, j) != 0
                println(i, j)
                break
            end
        end
    end
end
```

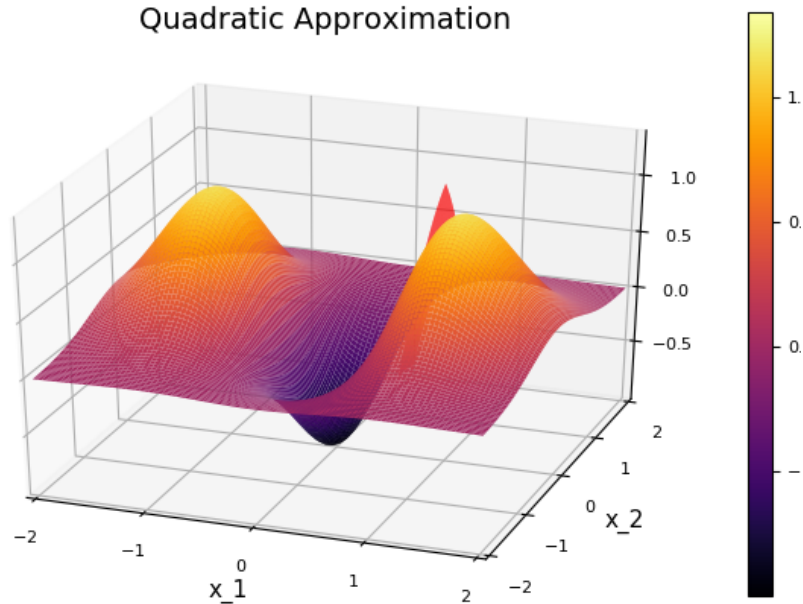
```
-1.5555555555555556-0.2222222222222222
-1.5151515151515151-0.3434343434343434
-1.4747474747474747-0.42424242424242425
-1.4343434343434343-0.5050505050505051
-1.3939393939393939-0.5454545454545454
-1.3535353535353536-0.5858585858585859
-1.3131313131313131-0.6262626262626263
-1.2727272727272727-0.6262626262626263
-1.2323232323232323-0.6666666666666666
-1.1919191919191918-0.6666666666666666
-1.1515151515151516-0.6666666666666666
-1.1111111111111112-0.6666666666666666
-1.0707070707070707-0.6666666666666666
-1.0303030303030303-0.6666666666666666
-0.9898989898989899-0.6666666666666666
-0.9494949494949495-0.6262626262626263
-0.9090909090909091-0.6262626262626263
-0.8686868686868687-0.5858585858585859
-0.8282828282828283-0.5454545454545454
-0.7878787878787878-0.5050505050505051
-0.7474747474747475-0.42424242424242425
-0.7070707070707071-0.3838383838383838
-0.6666666666666666-0.30303030303030304
-0.6262626262626263-0.2222222222222222
-0.5858585858585859-0.1414141414141414
-0.5454545454545454-0.06060606060606061
0.5454545454545454-0.06060606060606061
0.5858585858585859-0.1414141414141414
0.6262626262626263-0.2222222222222222
0.6666666666666666-0.30303030303030304
0.7070707070707071-0.3838383838383838
0.7474747474747475-0.42424242424242425
0.7878787878787878-0.5050505050505051
0.8282828282828283-0.5454545454545454
0.8686868686868687-0.5858585858585859
0.9090909090909091-0.6262626262626263
0.9494949494949495-0.6262626262626263
0.9898989898989899-0.6666666666666666
1.0303030303030303-0.6666666666666666
1.0707070707070707-0.6666666666666666
1.1111111111111112-0.6666666666666666
1.1515151515151516-0.6666666666666666
1.1919191919191918-0.6666666666666666
1.2323232323232323-0.6666666666666666
1.2727272727272727-0.6262626262626263
1.3131313131313131-0.6262626262626263
1.3535353535353536-0.5858585858585859
1.3939393939393939-0.5454545454545454
1.4343434343434343-0.5050505050505051
1.4747474747474747-0.42424242424242425
1.5151515151515151-0.3434343434343434
1.5555555555555556-0.2222222222222222
```

```
In [43]: x0_approximation = [.7; .4]
fq_negative(x_1, x_2) = f_q(x0_approximation, [x_1; x_2])
camera = (20, 30)
pyplot()
x_1 = range(-2, stop=2, length=100)
x_2 = range(-2, stop=2, length=100)
p = plot(x_1, x_2, f, title="Quadratic Approximation",
         st=:surface, camera=camera, xlabel="x_1", ylabel="x_2")

xq_1 = range(x0_approximation[1]-.2, stop=x0_approximation[1]+.2, length=100)
xq_2 = range(x0_approximation[2]-.2, stop=x0_approximation[2]+.2, length=100)
plot!(p, xq_1, xq_2, fq_negative, st=:surface, camera=camera, color=:red)
```

Out[43]:

Quadratic Approximation



In []: