




## Math 3607: Exam 3

Due: 11:59PM, Wednesday, November 10, 2021

Please read the statements below and sign your name.

### Disclaimers and Instructions

- You are **not** allowed to use MATLAB commands and functions **other than** the ones discussed in lectures, accompanying live scripts, textbooks, and homework/practice problem solutions.
- You may be requested to explain your code to me, in which case a proper and satisfactory explanation must be provided to receive any credits on relevant parts.
- You are **not** allowed to search online forums or even MathWorks website for this exam.
- You are **not** allowed to collaborate with classmates, unlike for homework.
- If any code is found to be plagiarized from the internet or another person, you will receive a zero on the *entire* exam and will be reported to the COAM.
- Do not carry out computations using *Symbolic Math Toolbox*. Any work done using `sym`, `syms`, `vpa`, and such will receive NO credit.
- **Notation.** Problems marked with  are to be done by hand; those marked with  are to be solved using a computer.
- Answers to analytical questions (ones marked with  ) without supporting work or justification will not receive any credit.

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### Academic Integrity Statements

- All of the work shown on this exam is my own.
- I will not consult with any resources (MathWorks website, online searches, etc.) other than the textbooks, lecture notes, and supplementary resources provided on the course Carmen pages.
- I will not discuss any part of this exam with anyone, online or offline.
- I understand that academic misconduct during an exam at The Ohio State University is very serious and can result in my failing this class or worse.
- I understand that any suspicious activity on my part will be automatically reported to the OSU Committee on Academic Misconduct (COAM) for their review.

Signature \_\_\_\_\_

# 1 EVD of Householder Matrices

[12 points]

 Choose the correct answer for each of the blanks below. You do not need to justify your answers for this problem; there is no partial credits.

Let  $\mathbf{u} \in \mathbb{R}^m$  be a unit vector and define a Householder matrix  $H \in \mathbb{R}^{m \times m}$  by  $H = I - 2\mathbf{u}\mathbf{u}^T$ . Observe that

$$H\mathbf{u} = (I - 2\mathbf{u}\mathbf{u}^T)\mathbf{u} = \mathbf{u} - 2\mathbf{u} = -\mathbf{u}.$$

This implies that

$$\lambda = \boxed{\text{(a)}}$$

is an eigenvalue of  $H$  and  $\mathbf{u}$  is a corresponding eigenvector. Now let  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{m-1}\}$  be an orthonormal basis of the orthogonal complement  $\mathbf{u}^\perp$ , which is an  $(m-1)$ -dimensional subspace of  $\mathbb{R}^m$  consisting of all vectors perpendicular to  $\mathbf{u}$ ; in particular,  $\mathbf{v}_i^T \mathbf{v}_i = \delta_{k,j}$  (Kronecker delta) and

$$\mathbf{u}^T \mathbf{v}_j = \boxed{\text{(b)}}.$$

Consequently, for any  $1 \leq j \leq m-1$ ,

$$H\mathbf{v}_j = (I - 2\mathbf{u}\mathbf{u}^T)\mathbf{v}_j = \boxed{\text{(c)}}.$$

This implies that  $\mathbf{v}_j$  is  $\boxed{\text{(d)}}$  of  $H$  corresponding to the eigenvalue  $\lambda_j = \boxed{\text{(e)}}$ .

Since all eigenvectors are orthogonal and thus linearly  $\boxed{\text{(f)}}$ ,  $H$  has an EVD  $H = VDV^{-1}$ , where

$$V = \left[ \begin{array}{c|c|c|c|c} \mathbf{u} & \mathbf{v}_1 & \mathbf{v}_2 & \cdots & \mathbf{v}_{m-1} \end{array} \right] \quad \text{and} \quad D = \begin{bmatrix} -1 & & & & \\ & 1 & & & \\ & & \ddots & & \\ & & & 1 & \\ & & & & 1 \end{bmatrix}.$$

- (a) ☐ +1  
☐ -1

- (c) ☐  $+\mathbf{v}_j$   
☐  $-\mathbf{v}_j$

- (e) ☐ +1  
☐ -1


- (b) ☐ 0  
☐ 1

- (d) ☐ an eigenvector  
☐ a singular vector

- (f) ☐ dependent  
☐ independent

## 2 Demystifying Recursion using EVD

[18 points]

 Consider the sequence  $(a_n)$  defined recursively by

$$a_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ 2a_{n-1} + a_{n-2} & \text{if } n \geq 2 \end{cases}$$

Find the general formula for the  $n$ th term  $a_n$ . Show all your work.

### 3 Newton's Method Faster Than Quadratic?

[20 points]

 Suppose that  $f$  has a simple root at  $r$  at which  $f''$  vanishes, that is,

$$f(r) = 0, \quad f'(r) \neq 0, \quad f''(r) = 0.$$

Assuming that  $f$  is (at least) three times continuously differentiable near  $r$ , determine the rate of convergence of Newton's method. That is, express the relationship between  $\epsilon_{k+1}$  and  $\epsilon_k$  in the form

$$\epsilon_{k+1} = C\epsilon_k^p + O(\epsilon_k^{p+1}),$$

where  $C$  and  $p$  are to be determined. Recall that  $\epsilon_k = x_k - r$ .






## 4 Air Resistance, Rootfinding, and Lambert W

[25 points]

The function

$$h(t) = -30t + 780(1 - e^{-t/3})$$

models the height of a rocket in the air at time  $t$ , subject to air resistance; see the figure below. Let  $t_{\max}$  be the time at which the rocket reaches its maximum height and let  $t_{\text{ground}}$  be the time at which the rocket hits the ground.

-  Find  $t_{\max}$  analytically. That is, find an exact expression for  $t_{\max}$ . Justify all your steps.
-  Find  $t_{\max}$  numerically, using `fzero`. Then compare the result against the analytical answer obtained in part (a).
-  (Optional/Bonus) Find  $t_{\text{ground}}$  analytically. That is, find an exact expression for  $t_{\text{ground}}$ . Justify all your steps.
-  Find  $t_{\text{ground}}$  numerically, using `fzero`. Then compare the result against the analytical answer obtained in part (c), only if you did it.
-  Produce a well-labeled plot of  $h(x)$  as shown in the figure. Clearly mark the maximum height and the time the rocket hits the ground.

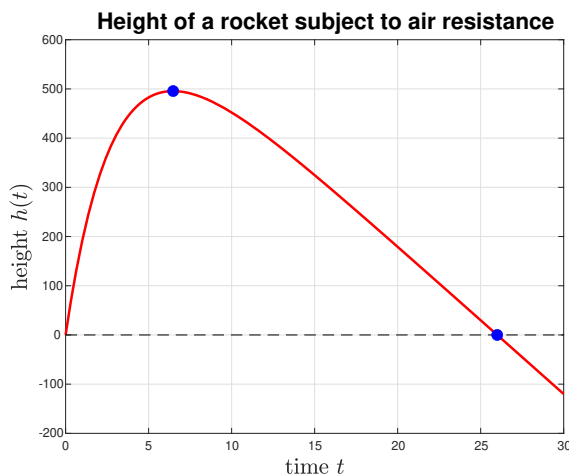


Figure 1: Example output.

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**Hints for (a) and (c).** To find  $t_{\max}$  which maximizes the height  $h(t)$ , use just what you know from Calculus. To find  $t_{\text{ground}}$  analytically, you need to solve an equation of the form

$$At + e^{Bt} = 1 \tag{1}$$


for  $t$ . To do this, introduce a new variable  $u$  by the transformation

$$t = -\frac{1}{B}u + \frac{1}{A}. \tag{2}$$

Substitute this into (1) and solve for  $u$ , using Lambert W function. Then use the transformation (2) to solve for  $t$ .

## 5 Embedding Watermark inside Image

[25 points]

 Let  $A$  and  $W$  be  $m$ -by- $n$  matrices representing two grayscale images of the same pixel dimensions as shown below. The image  $W$  is *embedded* inside the image  $A$  using an SVD-based scheme described below.

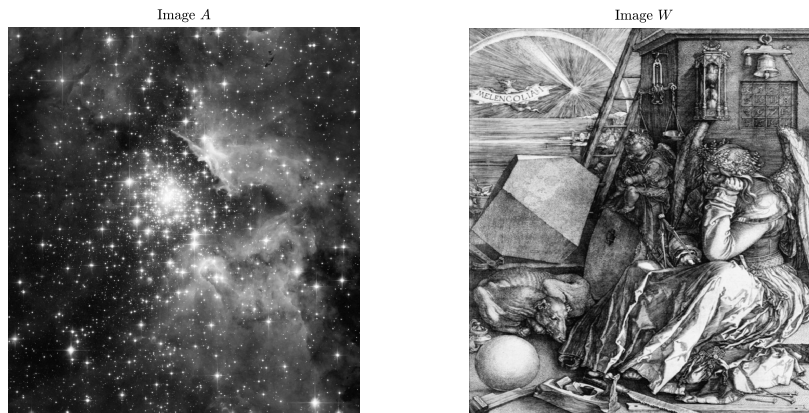


Figure 2: Two original images. The image represented by  $A$  on the left and the image represented by  $W$  on the right.

Let  $A = U\Sigma V^T$  be an SVD of  $A$ . For some small scaling factor  $\alpha > 0$ , construct a new matrix  $\Sigma + \alpha W$  and consider its SVD

$$\Sigma + \alpha W = U_w \Sigma_w V_w^T. \quad (3)$$

Note that both  $\Sigma$  and  $\Sigma_w$  are diagonal matrices and, for a suitably chosen  $\alpha$ ,  $\Sigma \approx \Sigma_w$ . Thus,  $A_w$  obtained by

$$A_w = U \Sigma_w V^T \quad (4)$$

is approximately equal to  $A$  and so represents an image which looks nearly identical to that of  $A$ , while containing information about the image  $W$  as well.

Your mission, should you choose to accept it, is to establish an algorithm to retrieve (an approximation of)  $W$  embedded in  $A_w$  given  $U_w$ ,  $V_w$ ,  $\Sigma$ , and  $\alpha$ .

Begin by loading the data with

```
>> load watermark % download and save 'watermark.mat' first
```

This loads  $A_w$ ,  $U_w$ ,  $V_w$ ,  $S$ , and  $\alpha$ , which correspond to  $A_w$ ,  $U_w$ ,  $V_w$ ,  $\Sigma$ , and  $\alpha$  as described above, respectively. Note that  $A_w$  is a matrix of double-precision floating point numbers whose elements range from 0 to 255.

- Write a MATLAB code to find a matrix  $W_0$ , an *approximation* of  $W$ , out of  $A_w$ .
- Display the images represented by  $A_w$  and  $W_0$  side by side using `subplot`. Do they look similar to their respective original images in Figure 2?

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**Hint for (a).** Begin by computing an SVD of  $A_w$ . This will not give you the same  $\Sigma_w$  as in (4), but will give you something quite close. (This is why the best you can do is to find an estimate of  $W$ , which we call  $W_0$ .) Then substitute this approximate  $\Sigma_w$  into (3) and solve for  $W$ .