

Name:

MWF10 Ryan Reich

MWF10 Stewart Wilcox

MWF11 Oliver knill

MWF11 Ryan Reich

MWF11 Stewart Wilcox

MWF12 Juliana Belding

TTH10 Wushi Goldring

TTH1130 Alex Subotic

- Start by writing your name in the above box and check your section in the box to the left.
- Try to answer each question on the same page as the question is asked. If needed, use the back or the next empty page for work. If you need additional paper, write your name on it.
- Do not detach pages from this exam packet or un-staple the packet.
- Please write neatly and except for problems 1-3, give details. Answers which are illegible for the grader can not be given credit.
- No notes, books, calculators, computers, or other electronic aids can be allowed.
- You have 90 minutes time to complete your work.

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| 1      |  | 20  |
| 2      |  | 10  |
| 3      |  | 10  |
| 4      |  | 10  |
| 5      |  | 10  |
| 6      |  | 10  |
| 7      |  | 10  |
| 8      |  | 10  |
| 9      |  | 10  |
| Total: |  | 100 |

Problem 1) (20 points) True or False? No justifications are needed.

- 1) 

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 A  $4 \times 4$  matrix whose entries are all 4 has determinant  $4^4$ .

**Solution:**

The determinant is zero because the matrix has identical rows.

- 2) 

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 $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  is an orthogonal matrix.

**Solution:**

It is an orthogonal **projection**, not an orthogonal **transformation**.

- 3) 

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 Every  $2 \times 2$  matrix is diagonalizable over the complex numbers.

**Solution:**

The shear is a counter example. It can not be diagonalized even using complex numbers. There is an eigenvector to the eigenvalue 1 with geometric multiplicity 1.

- 4) 

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 If  $A$  is a projection onto a linear subspace  $V$ , then  $\text{im}(A) = V$  and  $\ker(A) = V^\perp$ .

**Solution:**

The image is clearly  $V$ . Everything perpendicular to the image is mapped to the origin. One can see second part also using the formula  $\ker(A) = (\text{im}(A^T))^\perp$  and the fact that  $A = A^T$  for a projection.

- 5) 

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 If  $A$  is a  $3 \times 3$  matrix representing reflection about a line in  $\mathbb{R}^3$ , then  $A$  is symmetric.

**Solution:**

The projection  $P$  onto a line is symmetric. Because the reflection  $R$  satisfies  $R = 2P - I$ , also the reflection  $R = 2P - I$  is symmetric.

- 6) 

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 If  $A$  is a matrix with orthonormal columns, then  $\vec{x} = A^T \vec{b}$  must be a least-squares solution of the system  $A\vec{x} = \vec{b}$ .

**Solution:**

Because  $A^T A$  is the identity matrix  $1_n$  where  $n$  is the number of columns of  $A$ , the formula  $(A^T A)^{-1} A^T \vec{b}$  simplifies to  $A^T \vec{b}$ .

- 7) 

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 If  $A$  represents the projection onto a line in  $\mathbb{R}^2$ , then  $A$  is similar to  $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ .

**Solution:**

Yes, a projection is diagonalizable. It has one eigenvalue 1 and one eigenvalue 0.

- 8) 

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 If  $A$  is a square matrix such that  $A\vec{v} \cdot A\vec{w} = 0$ , whenever  $\vec{v} \cdot \vec{w} = 0$ , then  $A$  is an orthogonal matrix.

**Solution:**

Take the transformation which scales by a factor 2. It preserves angles but not lengths and is not an orthogonal transformation. Nevertheless, it preserves right angles.

- 9) 

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 If  $A$  is a  $2 \times 2$  matrix with the eigenvalues 0 and 1, then  $A$  is the matrix of an orthogonal projection.

**Solution:**

It is a projection but not necessarily an orthogonal projection.

- 10) 

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 If  $A$  and  $B$  are similar, then they have the same eigenvectors.

**Solution:**

They have the same **eigenvalues**, but not necessarily the same eigenvectors. The shear for example has the eigenvector  $e_1$ , while its transpose has the eigenvector  $e_2$ .

- 11) 

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 If  $A$  and  $B$  are both diagonalizable, then  $AB$  is diagonalizable.

**Solution:**

Not necessarily. there are  $2 \times 2$  matrices  $A, B$  for which the statement is not true: example:

$$A = \begin{bmatrix} 2 & 3 \\ 0 & 3 \end{bmatrix} \text{ and } B = \begin{bmatrix} 3 & 2 \\ 0 & 2 \end{bmatrix} \text{ are both diagonalizable, but } AB = \begin{bmatrix} 6 & 10 \\ 0 & 6 \end{bmatrix} \text{ is not.}$$

- 12) 

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 If  $\lambda$  is an eigenvalue of  $A$  and  $\mu$  is an eigenvalue of  $B$ , then  $\lambda\mu$  is an eigenvalue of  $AB$ .

**Solution:**

This is even not true for all diagonal  $2 \times 2$  matrices with different eigenvalues. Take  $A = \begin{bmatrix} 3 & 0 \\ 0 & 4 \end{bmatrix}$  and  $B = \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$ . The number 15 is a product of eigenvalues of  $A$  and  $B$  but not an eigenvalue of  $AB$ .

- 13) 

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 If  $S^{-1}AS$  is a diagonal matrix, then the columns of  $S$  must be eigenvectors of  $A$ .

**Solution:**

Yes,  $S$  is then the conjugation to the diagonal matrix and  $S^{-1}ASe_i = e_i$  means  $ASe_i = Se_i$  so that  $Se_i$  is an eigenvector. But  $Se_i$  is the  $i$ 'th column of  $S$ .

- 14) 

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 If  $\lambda$  is an eigenvalue of a  $2 \times 2$  matrix  $A$ , then  $-\lambda$  is an eigenvalue of  $A$ .

**Solution:**

Take a diagonal matrix with eigenvalues 2, 3.

- 15) 

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 If  $A$  is a  $5 \times 5$  matrix, then  $\det(5A) = 25\det(A)$ .

**Solution:**

We have  $\det(5A) = 5^5\det(A)$ .

- 16) 

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 Any two  $2 \times 2$  matrices whose eigenvectors are equal must be similar.

**Solution:**

Two different diagonal matrices have the same eigenvectors but are not similar

- 17) 

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 The standard basis vectors of  $\mathbf{R}^n$  are the eigenvectors of every diagonal  $n \times n$  matrix.

**Solution:**

Obvious, in the  $k$ 'th row, we have a multiple of the  $k$ 'th basis vector.

- 18) 

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 If  $A^2 = A$ , then every eigenvalue  $\lambda$  of  $A$  is either  $\lambda = 1$  or  $\lambda = 0$ .

**Solution:**

If  $A$  has an eigenvalue  $\lambda$ , then for the corresponding eigenvector  $\vec{v}$  we have  $A^2\vec{v} = \lambda^2\vec{v}$  and  $A\vec{v} = \lambda\vec{v}$ . So,  $\lambda^2 = \lambda$ .

- 19) 

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 If  $A$  is a symmetric matrix with characteristic polynomial  $(1 - \lambda)^2(2 - \lambda)$ , then the 1-eigenspace of  $A$  is the orthogonal complement of the 2-eigenspace of  $A$ .

**Solution:**

Eigenvectors of a symmetric matrix are perpendicular

- 20) 

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 A  $2 \times 2$  matrix with characteristic polynomial  $f_A(\lambda) = (2 - \lambda)^2$  is similar to either  $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  or  $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ .

**Solution:**

It would be consequence of the Jordan normal form theorem, treated in class if the characteristic polynomial were  $(1 - \lambda)^2$ . But the characteristic polynomial here does not match.

Problem 2) (10 points)

Check the boxes, where the two matrices  $A$  and  $B$  are similar. No explanations are necessary for this problem. Each pair counts 2 points.

a)  $\square$   $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$   $B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

b)  $\square$   $A = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 2 & 1 \\ 0 & 0 & 3 \end{bmatrix}$   $B = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$

c)  $\square$   $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{bmatrix}$   $B = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

d)  $\square$   $A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$   $B = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 2 \end{bmatrix}$

e)  $\square$   $A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$   $B = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$

**Solution:**

Only b) and c) are similar. a) are different Jordan normal forms, b) are similar because  $A$  is diagonalizable where it is  $B$ , c) can be conjugated by a permutation matrix, d) have different eigenvalues, for e) one can see  $A^2 = 0$ , while  $B^2 \neq 0$ .

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| Problem 3) (10 points) |
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Consider the matrix  $A = \begin{bmatrix} 5 & 1 & 1 & 1 & 1 & 1 \\ 1 & 5 & 1 & 1 & 1 & 1 \\ 1 & 1 & 5 & 1 & 1 & 1 \\ 1 & 1 & 1 & 5 & 1 & 1 \\ 1 & 1 & 1 & 1 & 5 & 1 \\ 1 & 1 & 1 & 1 & 1 & 5 \end{bmatrix}$ .

a) Find the kernel of  $B = A - 4I_6$ .

b) Show that  $\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$  is an eigenvector of  $A$ . What is the eigenvalue?

c) Find all the eigenvalues of  $A$  with their algebraic multiplicities.

d) Write down the eigenbasis of  $A$ . What are the geometric multiplicities of the eigenvalues?

e) What is  $\det(A)$ ?

**Solution:**

a) The kernel of  $B = A - 4I = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$  is the kernel of its row reduced

echelon form  $\text{rref}(B) = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$ . which is spanned by the 5 vectors

$$\begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ -1 \end{bmatrix}$$

b) The vector  $\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$  is an eigenvector of  $B$  to  $\lambda = 6$ . So, it is an eigenvector of  $A$  with

eigenvalue  $\lambda = 10$ .

c)  $B = A - 4I_6$  has the eigenvalue 6 with multiplicity 1 and the eigenvalue 0 with multiplicity 5. So  $A$  has the eigenvalue 10 with multiplicity 1 and the eigenvalue 4 with multiplicity 5.

d) The eigenvectors of  $A - 4I_6$  are the same as the eigenvectors of  $A$ . because if  $A\vec{v} = \lambda\vec{v}$ , then  $(A - 4I_6)\vec{v} = (\lambda - 4)\vec{v}$ .

e) The determinant is the product of the eigenvalues which is  $4^5 \cdot 10 = \boxed{10'240}$ .

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| Problem 4) (10 points) |
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Find the function  $y = f(x) = a + b2^x$ , which best fits the data

| x | y |
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| 0 | 1 |
| 1 | 3 |
| 2 | 7 |

**Solution:**

We have to find the least square solution to the system of equations

$$\begin{aligned}a + b &= 1 \\a + 2b &= 3 \\a + 4b &= 7\end{aligned}$$

which is in matrix form written as  $A\vec{x} = \vec{b}$  with

$$A = \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 4 \end{bmatrix}, \vec{b} = \begin{bmatrix} 1 \\ 3 \\ 7 \end{bmatrix}.$$

Now  $A^T\vec{b} = \begin{bmatrix} 11 \\ 35 \end{bmatrix}$  and  $A^T A = \begin{bmatrix} 3 & 7 \\ 7 & 21 \end{bmatrix}$  and  $(A^T A)^{-1} = \begin{bmatrix} 21 & -7 \\ -7 & 3 \end{bmatrix} / 14$  and  $(A^T A)^{-1} A^T \vec{b}$  is  $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$ . The best fit is the function  $\boxed{f(x) = -1 + 2 \cdot 2^x}$ .

Actually, the minimal fit produces an actual solution through the data.

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| Problem 5) (10 points) |
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Let  $A = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 2 & -1 \\ 6 & 0 & -4 \end{bmatrix}$ . You are told that  $\mathcal{B} = \{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$  is an eigenbasis for  $A$ , where:

- $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}$  belongs to the eigenvalue  $-1$ ,
- $\vec{v}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$  belongs to the eigenvalue  $2$ ,
- $\vec{v}_3 = \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}$  belongs to the eigenvalue  $-2$ .



Find an eigenbasis for  $A^T$ .

**Solution:**

There are two possibilities to solve this problem. Since we know the eigenvalues, we just have to find the new eigenvectors to  $A^T$ . They are

- $\vec{v}_1 = \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix}$  belongs to the eigenvalue  $-1$ ,
- $\vec{v}_2 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$  belongs to the eigenvalue  $2$ ,
- $\vec{v}_3 = \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$  belongs to the eigenvalue  $-2$ .

The second solution is to note that  $S^{-1}AS = B$  implies  $S^T A^T (S^{-1})^T = B$  so that we can obtain the eigenvectors as the columns of  $(S^{-1})^T$ , where

$$S = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 1 \\ 2 & 0 & 3 \end{bmatrix}$$

is the matrix containing the eigenvectors of  $A$  as column vectors. We have

$$(S^{-1})^T = \begin{bmatrix} 3 & -1 & -2 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$$

and can read off the eigenvectors of  $A^T$  as the column vectors of this matrix.

Problem 6) (10 points)

a) (4 points) Find the Gram-Schmidt orthogonalization of the basis  $\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 2 \\ 3 \end{bmatrix}$ .

b) (4 points) Find the QR decomposition of the matrix  $A = \begin{bmatrix} 0 & 0 & 2 \\ 0 & 1 & 2 \\ 1 & 2 & 3 \end{bmatrix}$ .

c) (2 points) What is the volume of the parallelepiped spanned by the column vectors of  $A$ ?

**Solution:**

a) The first vector  $\vec{v}_1$  is already normalized so that  $\vec{w}_1 = \vec{v}_1$ .

$$\vec{u}_2 = \vec{v}_2 - (\vec{w}_1 \cdot \vec{v}_2)\vec{w}_1 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \text{ This is already normalized.}$$

$$\vec{u}_3 = \vec{v}_3 - (\vec{w}_1 \cdot \vec{v}_3)\vec{w}_1 - (\vec{u}_2 \cdot \vec{v}_3)\vec{u}_2 = \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix} \text{ which is normalized equal to } \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}.$$

$$\text{b) } Q = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \text{ and } R = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix}.$$

c) The volume is the absolute value of the determinant of  $R$  which is  $\boxed{2}$ . (You could also see directly that the volume is  $|\det(A)| = 2$ .)

**Problem 7) (10 points)**

a) Find the determinant of the matrix

$$\begin{bmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 2 & 3 & 4 \\ 0 & 0 & 2 & 3 & 4 \\ 0 & 0 & 3 & 4 & 5 \\ 3 & -2 & -3 & -4 & 0 \end{bmatrix}$$

b) Find the determinant of the matrix

$$\begin{bmatrix} 1 & 2 & 0 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 3 & 0 & 0 \\ 0 & 0 & 1 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 3 & 4 \\ 0 & 0 & 0 & 0 & 2 & 3 \end{bmatrix}$$

**Solution:**

a) You can use Laplace to make the expansion. But it is easier to see that two rows are identical so that the matrix is not invertible. Therefore  $\boxed{\det(A) = 0}$ .

b) This is a partitioned matrix having 3 matrices of size  $2 \times 2$  in the diagonal. The determinant of  $A$  is  $\det \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \cdot \det \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix} \cdot \det \begin{bmatrix} 3 & 4 \\ 2 & 3 \end{bmatrix} = (-3) \cdot 1 \cdot 1 = \boxed{-3}$ .

**Problem 8) (10 points)**

To support their teams, Harvard and Yale fans both sell T-shirts in November 2004. Each day, the following happens:

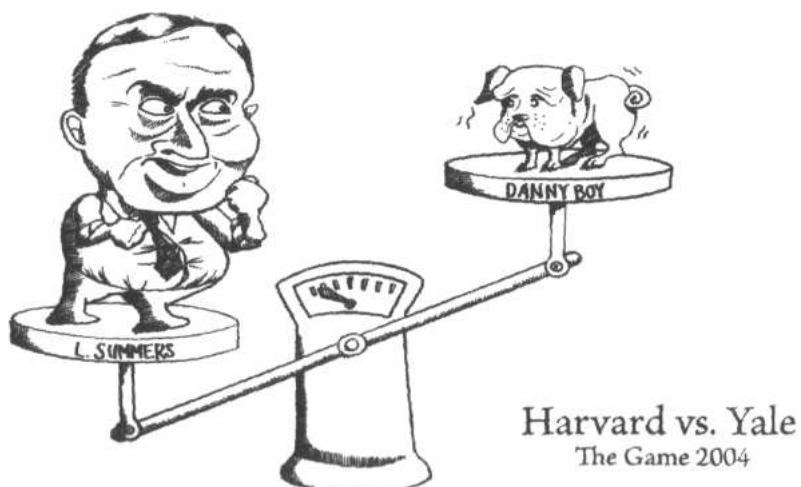
- Every Harvard fan recruits two more Harvard fans and four Yale fans.
- Every Yale fan recruits two more Yale fans and one Harvard fan.

Let  $H(t)$  and  $Y(t)$  be the number of Harvard fans and Yale fans, respectively, on day  $t$ . We can model the above situation with a dynamical system

$$\begin{bmatrix} H(t+1) \\ Y(t+1) \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} H(t) \\ Y(t) \end{bmatrix}.$$

Suppose that  $H(0) = Y(0) = 100$ .

- (5 points) Find a closed formula for  $H(t), Y(t)$ .
- (5 points) What happens with  $H(t)/Y(t)$  for large  $t$ ?



**Solution:**

We have to find the eigenvalues and eigenvectors of the matrix  $A = \begin{bmatrix} 3 & 4 \\ 1 & 3 \end{bmatrix}$ .

The eigenvector to the eigenvalue 5 is  $\vec{v}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ , the eigenvector to the eigenvalue 1 is

$\vec{v}_2 = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$ . Because  $(75\vec{v}_1 + 25\vec{v}_2)$  is the initial condition, we know the solution is

$$A^n \vec{v} = A^n(75\vec{v}_1 + 25\vec{v}_2) = 75 \cdot 5^n \vec{v}_1 + 25 \cdot 1^n \cdot \vec{v}_2 = 75 \cdot 5^n \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix} + 25 \cdot 1^n \cdot \begin{bmatrix} -2 \\ 1 \end{bmatrix}$$

b)

$$\frac{H(t)}{Y(t)} = \frac{75 \cdot 5^n \cdot 2 - 50 \cdot 1^n \cdot 2}{75 \cdot 5^n \cdot 1 + 25 \cdot 1^n \cdot 1} \rightarrow 2$$

c) No, it is not stable. Stability was defined as the property that  $A^n \vec{x} \rightarrow \vec{0}$  for all vectors  $\vec{x}$  and happens if all eigenvalues have absolute value  $|\lambda| < 1$ . We have here two eigenvalues which violate this. Indeed, for most initial conditions,  $|A^n \vec{x}|$  converges to  $\infty$  like for the initial condition  $\vec{x} = \begin{bmatrix} 100 \\ 100 \end{bmatrix}$ .

Problem 9) (10 points)

Let  $V$  be the 3-dimensional subspace of  $\mathbb{R}^4$  spanned by

$$\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 1 \end{bmatrix}, \vec{v}_3 = \begin{bmatrix} 1 \\ 2 \\ 0 \\ 1 \end{bmatrix}.$$

Let  $\{\vec{w}_1, \vec{w}_2, \vec{w}_3\}$  be an orthonormal basis of  $V$  and  $A$  be the matrix  $A = \begin{bmatrix} | & | & | \\ \vec{w}_1 & \vec{w}_2 & \vec{w}_3 \\ | & | & | \end{bmatrix}$ . Find  $\det(AA^T)$  and  $\det(A^T A)$ .

**Hint.** Think before calculating. You don't actually need to compute  $\vec{w}_1, \vec{w}_2, \vec{w}_3$ , nor do you have to multiply matrices to solve the problem.

**Solution:**

Note that neither  $A$  nor  $A^T$  are square matrices so that it does not make sense to talk about  $\det(A)$  and  $\det(A^T)$ .

However, both  $A^T A$  and  $AA^T$  are square matrices for which the determinant is defined. The matrix  $A^T A$  has a different shape than the matrix  $AA^T$ .

As a background for this problem, take the formula for the least square solution  $\vec{x}_* = (A^T A)^{-1} A^T \vec{b}$  of  $A\vec{x} = \vec{b}$  which leads to the general projection formula  $P\vec{b} = A(A^T A)^{-1} A^T \vec{b}$  onto the column space of a matrix  $A$ . In the case, when the column vectors of  $A$  form an orthonormal set, then this formula simplifies to  $P\vec{b} = AA^T \vec{b}$  because  $A^T A$  is then an identity matrix. We have an orthonormal set of vectors as columns of  $A$  so that  $A^T A$  is an identity matrix and  $AA^T$  is a projection from four to three dimensions and not invertible. Here are the solutions:

a)  $AA^T$  is a  $4 \times 4$  matrix which has its image contained in the image of  $A$ . It is the

projection from a four dimensional space to a three dimensional space and therefore not invertible. Therefore  $\boxed{\det(AA^T) = 0}$ .

b)  $A^T A$  is the  $3 \times 3$  identity matrix and has determinant  $\boxed{\det(A^T A) = 1}$ .