

Final course review

Once more, we review questions from a previous exam to prepare ourselves for an upcoming exam.

1. Suppose we know that A is an m by n matrix of rank r , $A\mathbf{x} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ has

no solution, and $A\mathbf{x} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ has exactly one solution.

- a) What can we say about m , n and r ?

The product $A\mathbf{x}$ is a vector in three dimensions, so $m = 3$.

The fact that $A\mathbf{x} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ has no solution tells us that the column space is not all of \mathbb{R}^3 . In addition, we know that the column space contains $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$, so r is not zero: $1 \leq r < 3$.

The fact that $A\mathbf{x} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ has exactly one solution tells us that the nullspace of A contains only the zero vector and so $n = r$. Hence $1 \leq n < 3$.

- b) Write down an example of a matrix A that fits this description.

The vector $\begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ must be in the column space, so we'll make it a column of A . The simplest way to answer this question is to stop here.

$$A = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}.$$

In this solution, $n = r = 1$ and $m = 3$.

To find a solution in which $n = r = 2$, add a second column. Make sure that $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ is not in the column space:

$$A = \begin{bmatrix} 0 & 0 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

There are many other correct answers to this question.

- c) Cross out all statements that are false about any matrix with the given properties (which are $1 \leq r = n, m = 3$).

- i. $\det A^T A = \det A A^T$
- ii. $A^T A$ is invertible
- iii. $A A^T$ is positive definite

One good approach to this problem is to use our sample matrix to test each statement.

- i. If we leave this part to last, we can quickly answer it (**false**) using what we learn while answering the following two parts.

- ii. The matrix $A^T A$ is invertible if $r = n$; i.e. if the columns of A are independent.

The nullspace of our A contains only the zero vector, so this statement is **true**.

For each of our sample matrices, $A^T A$ equals the identity and so is invertible.

Note that this means $\det A^T A \neq 0$.

- iii. We know that $m = 3$ and $r < 3$, so $A A^T$ will be a 3 by 3 matrix with rank less than 3; it **can't be positive definite**. (It is true that for any matrix A with real valued entries, $A A^T$ is positive semidefinite.)

For our test matrices, $A A^T$ has at least one row that's all zeros, so 0 is an eigenvalue (and is not positive).

Note also that $\det A A^T = 0$ and so statement (i) must be false. (However, if A and B are square matrices then $\det B A = \det A B = \det A \det B$.)

- d) Prove that $A^T \mathbf{y} = \mathbf{c}$ has at least one solution for every right hand side \mathbf{c} , and in fact has infinitely many solutions for every \mathbf{c} .

We know A^T is an n by m matrix with $m = 3$ and rank $r = n < m$. If A^T has full row rank, the equation $A^T \mathbf{y} = \mathbf{c}$ is always solvable. We have n rows and rank $r = n$, so A^T has full row rank. Therefore $A^T \mathbf{y} = \mathbf{c}$ has a solution for every vector \mathbf{c} .

The solvable system $A^T \mathbf{y} = \mathbf{c}$ will have infinitely many solutions if the nullspace of A^T has positive dimension. We know $\dim(N(A^T)) = m - r > 0$, so $A^T \mathbf{y} = \mathbf{c}$ has infinitely many solutions for every \mathbf{c} .

2. Suppose the columns of A are $\mathbf{v}_1, \mathbf{v}_2$ and \mathbf{v}_3 .

- a) Solve $A\mathbf{x} = \mathbf{v}_1 - \mathbf{v}_2 + \mathbf{v}_3$.

This is just the "column method" of multiplying matrices from the first lecture. Choose $\mathbf{x} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$.

- b) True or false: if $\mathbf{v}_1 - \mathbf{v}_2 + \mathbf{v}_3 = \mathbf{0}$, then the solution to (2a) is not unique. Explain your answer.

True. Any scalar multiple of $\mathbf{x} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$ will be a solution.

Another way of answering this is to note that A^T has a nontrivial nullspace, and we can always add any vector in the nullspace to a solution \mathbf{x} to get a different solution.

- c) Suppose $\mathbf{v}_1, \mathbf{v}_2$ and \mathbf{v}_3 are orthonormal (forget about (2b)). What combination of \mathbf{v}_1 and \mathbf{v}_2 is closest to \mathbf{v}_3 ?

If we imagine the right triangle out from the origin formed by $a\mathbf{v}_1 + b\mathbf{v}_2$ and \mathbf{v}_3 , the Pythagorean theorem tells us that $0\mathbf{v}_1 + 0\mathbf{v}_2 = \mathbf{0}$ is the closest point to \mathbf{v}_3 in the plane spanned by \mathbf{v}_1 and \mathbf{v}_2 .

3. Suppose we have the Markov matrix

$$A = \begin{bmatrix} .2 & .4 & .3 \\ .4 & .2 & .3 \\ .4 & .4 & .4 \end{bmatrix}.$$

Note that the sum of the first two columns of A equals twice the third column of A .

- a) What are the eigenvalues of A ?

Zero is an eigenvalue because the columns of A are dependent. (A is singular.)

One is an eigenvalue because A is a Markov matrix.

The third eigenvalue is $-.2$ because the trace of A is $.8$. So $\lambda = 0, 1, -.2$.

- b) Let $\mathbf{u}_k = A^k \mathbf{u}(0)$. If $\mathbf{u}(0) = \begin{bmatrix} 0 \\ 10 \\ 0 \end{bmatrix}$, what is $\lim_{k \rightarrow \infty} \mathbf{u}_k$?

We'll start by computing \mathbf{u}_k and then find the steady state. This means finding a general expression of the form:

$$\mathbf{u}_k = c_1 \lambda_1^k \mathbf{x}_1 + c_2 \lambda_2^k \mathbf{x}_2 + c_3 \lambda_3^k \mathbf{x}_3.$$

When we plug in the eigenvalues we found in part (3a), this becomes

$$\mathbf{u}_k = \mathbf{0} + c_2 \mathbf{x}_2 + c_3 (-.2)^k \mathbf{x}_3.$$

We see that as k approaches infinity, $c_2 \mathbf{x}_2$ is the only term that does not go to zero.

The key eigenvector in any Markov process is the one with eigenvalue one.

To find \mathbf{x}_2 , solve $(A - I)\mathbf{x}_2 = \mathbf{0}$:

$$\begin{bmatrix} -.8 & .4 & .3 \\ .4 & -.8 & .3 \\ .4 & .4 & -.6 \end{bmatrix} \mathbf{x}_2 = \mathbf{0}.$$

The best way to solve this might be by elimination. However, because the first two columns look like multiples of 4 and the third column

looks like a multiple of 3, we might get lucky and guess $\mathbf{x}_2 = \begin{bmatrix} 3 \\ 3 \\ 4 \end{bmatrix}$.

This gives us $\mathbf{u}_\infty = c_2 \begin{bmatrix} 3 \\ 3 \\ 4 \end{bmatrix}$. We know that in a Markov process, the

sum of the entries of \mathbf{u}_k is the same for all k . The sum of the entries of

$\mathbf{u}(0)$ is 10, so $c_2 = 1$ and $\mathbf{u}_\infty = \begin{bmatrix} 3 \\ 3 \\ 4 \end{bmatrix}$.

4. Find a two by two matrix that:

a) projects onto the line spanned by $\mathbf{a} = \begin{bmatrix} 4 \\ -3 \end{bmatrix}$.

The formula for this matrix is $P = \frac{\mathbf{a}\mathbf{a}^T}{\mathbf{a}^T\mathbf{a}}$. This gives us

$$P = \begin{bmatrix} 16/25 & -12/25 \\ -12/25 & 9/25 \end{bmatrix}.$$

(To test this answer, we can quickly check that $\det P = 0$.)

b) has eigenvalues $\lambda_1 = 0$ and $\lambda_2 = 3$ and eigenvectors $\mathbf{x}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and

$$\mathbf{x}_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}.$$

Here the formula we need is $A = S\Lambda S^{-1}$.

$$\begin{aligned} A &= \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}^{-1} \\ &= \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} -1/3 & 2/3 \\ 2/3 & -1/3 \end{bmatrix} \\ A &= \begin{bmatrix} 4 & -2 \\ 2 & -1 \end{bmatrix}. \end{aligned}$$

If time permits, we can check this by computing the products $A\mathbf{x}_i$.

c) has real entries and cannot be factored as B^TB for any B .

We know that B^TB will always be symmetric, so any asymmetric matrix has this property. For example, we could choose $\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$.

d) is not symmetric, but has orthogonal eigenvectors.

We know that symmetric matrices have orthogonal eigenvectors, but so do other types of matrices (e.g. **skew symmetric and orthogonal**) when we allow complex eigenvectors.

Two possible answers are:

$$\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \quad (\text{skew symmetric})$$

$$\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \quad (\text{orthogonal}).$$

5. Applying the least squares method to the system

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} c \\ d \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix} = \mathbf{b}$$

gives the best fit vector $\begin{bmatrix} \hat{c} \\ \hat{d} \end{bmatrix} = \begin{bmatrix} 11/3 \\ -1 \end{bmatrix}$.

a) What is the projection \mathbf{p} of $\mathbf{b} = \begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}$ onto the column space of

$$A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix}?$$

We know that $11/3$ times the first column minus 1 times the second column is the closest point P in the column space to $\begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}$, so the answer is

$$A \begin{bmatrix} \hat{c} \\ \hat{d} \end{bmatrix} = \frac{11}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 11/3 \\ 8/3 \\ 5/3 \end{bmatrix}.$$

b) Draw the straight line problem that corresponds to this system.

Plotting the entries of the second column of A against the entries of \mathbf{b} we get the three points shown in Figure 1. The best fit line is $\hat{c} + \hat{d}t$.

c) Find a different vector $\mathbf{b} \neq \mathbf{0} \in \mathbb{R}^3$ so that the least squares solution is

$$\begin{bmatrix} \hat{c} \\ \hat{d} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

We know that $\begin{bmatrix} \hat{c} \\ \hat{d} \end{bmatrix}$ is the projection of \mathbf{b} onto the column space, so to get a zero projection we need to find a vector *orthogonal to the columns*.

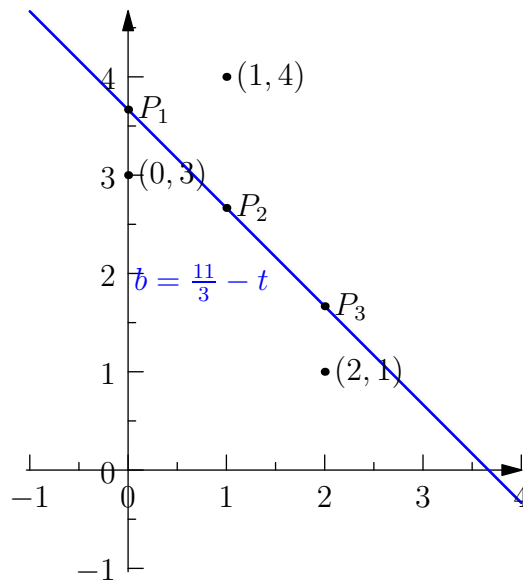


Figure 1: Three data points and their “best fit” line $\frac{11}{3} - t$.

We could get the answer $\mathbf{b} = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$ by inspection, or we could use the cross product of the columns to find a value for \mathbf{b} .

Thank you for taking this course!

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18.06SC Linear Algebra
Fall 2011

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