

NAME: SOLUTIONS

1	/30	2	/10	3	/12	4	/18	5	/10	6	/8	7	/12	T	/100
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MATH 430 (Fall 2005) Exam 2, November 3rd

Show all work and give complete explanations for all your answers.

This is a 75 minute exam. It is worth a total of 100 points.

(1) [30 pts]

(a) Let  $\mathbf{u}$  be a non-zero  $n \times 1$  column vector and  $\mathbf{v}$  a non-zero  $m \times 1$  column vector. Prove that  $\mathbf{u}\mathbf{v}^T$  has rank 1.

PROOF 1  $\text{Range}(\mathbf{u}\mathbf{v}^T) = \{ \mathbf{y} \in \mathbb{R}^n / \mathbf{y} = \mathbf{u} \underbrace{\mathbf{v}^T \mathbf{x}}_{1 \times 1} \text{ for some } \mathbf{x} \in \mathbb{R}^m \}$   
 $= \{ \mathbf{y} \in \mathbb{R}^n / \mathbf{y} = (\mathbf{v}^T \mathbf{x}) \mathbf{u} \text{ for some } \mathbf{x} \in \mathbb{R}^m \}$

Since  $\mathbf{v} \neq \mathbf{0} \quad \forall \mathbf{x} \in \mathbb{R}^m$   
 $\exists \mathbf{x} \in \mathbb{R}^m$  with  $\alpha = \mathbf{v}^T \mathbf{x}$ .

So  $\text{Range}(\mathbf{u}\mathbf{v}^T) = \{ \mathbf{y} \in \mathbb{R}^n / \mathbf{y} = \alpha \mathbf{u} \text{ for } \alpha \in \mathbb{R} \}$   
 $= \text{Span}(\mathbf{u})$  is 1D as  $\mathbf{u} \neq \mathbf{0}$

So  $\text{Rk}(\mathbf{u}\mathbf{v}^T) = \dim \text{Range}(\mathbf{u}\mathbf{v}^T) = 1$

PROOF 2 (PTO)

(b) Suppose that  $\mathbf{A}$  and  $\mathbf{B}$  are  $n \times n$  matrices. Prove that  $\text{trace}(\mathbf{AB}) = \text{trace}(\mathbf{BA})$ .

$(\mathbf{AB})_{ij} = A_{ik} B_{kj}$  is  $i$ th row of  $\mathbf{A}$  by  $j$ th col of  $\mathbf{B}$   
 $(-)$  (1)

So  $(\mathbf{AB})_{ii} = A_{ik} B_{ki}$

So  $\text{Trace}(\mathbf{AB}) = \sum_{i=1}^n (\mathbf{AB})_{ii} = \sum_{i=1}^n \sum_{k=1}^n A_{ik} B_{ki}$  Trace(BA)  
 $= \sum_{k=1}^n \sum_{i=1}^n B_{ki} A_{ik} = \sum_{k=1}^n B_{ki} A_{ki} = \sum_{k=1}^n (\mathbf{BA})_{kk}$

1a PROOF 2

$$\vec{u} \vec{v}^T = \begin{pmatrix} u_1 \\ | \\ u_n \end{pmatrix} (v_1 - v_m) = \begin{pmatrix} u_1 \vec{v}^T \\ u_2 \vec{v}^T \\ \vdots \\ u_m \vec{v}^T \end{pmatrix}$$

Since  $\vec{u} \neq 0$  there is a  $j$  with  $u_j \neq 0$ .

Since switching rows of a matrix does not change its rank we can assume  $u_1 \neq 0$ .

Then any other row of  $\vec{u} \vec{v}^T$  is a multiple of the first row as

$$u_k \vec{v}^T = \left( \frac{u_k}{u_1} \right) u_1 \vec{v}^T$$

So doing row operations

$$\text{Row } k = \text{Row } k - \left( \frac{u_k}{u_1} \right) \text{Row } 1$$

yields the matrix

$$\begin{pmatrix} u_1 \vec{v}^T \\ 0 \\ | \\ 0 \end{pmatrix}$$

which has rank 1 since  $u_1 \neq 0$   
and  $\vec{v} \neq 0$ .

(c) Let  $T: V \rightarrow V$ , be a linear operator, where  $V$  is a finite dimensional vector space. Using (b), define  $\text{trace}(T)$ .

Let  $B$  be any basis for  $V$ , and let  $[T]_B$  be the matrix of  $T$  in this basis. Define

$$\boxed{\text{Trace}(T) = \text{Trace}([T]_B)} \quad (*)$$

CHECK  $(*)$  is well defined, independent of choice of basis  $B$ .  
If  $B'$  were any other basis for  $V$  then

$$[T]_{B'} = P^{-1} [T]_B P \quad \text{for some invertible } P$$

$$\text{So } \text{Trace}([T]_{B'}) = \text{Trace}(P^{-1} [T]_B P) \stackrel{(b)}{=} \text{Trace}([T]_B P P^{-1}) \\ = \text{Trace}([T]_B). \quad \text{So def } (*) \text{ is indept of}$$

(d) State the three properties that characterize the determinant as a function from the space of  $n \times n$  real matrices to  $\mathbb{R}$ .

choice of basis  $B$ .

(I)  $\det$  depends linearly on 1st row

$$\underline{\text{e}} \quad \text{If } A = \begin{bmatrix} \vec{u} \\ \vdots \\ \vec{v} \\ \vdots \\ \vec{w} \end{bmatrix}, \quad B = \begin{bmatrix} \vec{v} \\ \vdots \\ \vec{u} \\ \vdots \\ \vec{w} \end{bmatrix}, \quad C = \begin{bmatrix} \alpha \vec{u} + \beta \vec{v} \\ \vdots \\ \vec{v} \\ \vdots \\ \vec{w} \end{bmatrix}$$

$$\text{Then } \det(C) = \alpha \det(A) + \beta \det(B)$$

(II)  $\det$  changes sign when two rows of the matrix are swapped

$$\text{(III)} \quad \det(I_{n \times n}) = 1$$

(e) Suppose that  $A$  and  $B$  are  $n \times n$  invertible matrices. Using the definition you gave in (d) to prove that  $\det(AB) = \det(A) \det(B)$ .

Since  $B$  is invertible,  $\det B \neq 0$ .

$$\text{Let } d(A) = \frac{\det(AB)}{\det(B)}$$

If we can show  $d(A)$  satisfies I, II, III of (d)  
Then  $d(A) = \det(A)$  by uniqueness of  $\det$ .

$$\textcircled{\text{III}} \det(I) = \frac{\det(IB)}{\det B} = 1$$

$$\textcircled{\text{II}} (AB)_{i*} = A_{i*} B. \text{ So if we swap two rows of } A \text{ to get } \tilde{A} \text{ we must swap some two rows of } AB \text{ to get } \tilde{A}B. \text{ So } d(\tilde{A}) = \frac{\det(\tilde{A}B)}{\det B} = \frac{\det(AB)}{\det B} = d(A)$$

By property II applied to  $\det(AB)$

(PTO)

(f) Let  $u$  be a length one vector in  $\mathcal{R}^n$ , and let  $R$  be the  $n \times n$  matrix  $R = I_n - 2uu^T$ . Calculate  $\det(R)$  and explain the physical meaning of the linear operator defined by  $R(v) = Rv$ .

$R = I - 2\vec{u}\vec{u}^T$  is a rank 1 update of  $I$ .

$$\begin{aligned} \text{Hence } \det(R) &= 1 - 2 \vec{u}^T \vec{u} \\ &= 1 - 2 \|\vec{u}\|^2 = 1 - 2 \cdot 1 = -1 \end{aligned} \quad \begin{array}{l} \|\vec{u}\|=1 \\ \swarrow \end{array}$$

$R$  is the reflection over the plane through the origin whose normal vector is  $\vec{u}$ .

① The proof of ① is similar to that of ②.

Specifically:

$$\text{Suppose } E = \begin{pmatrix} \vec{u} \\ M \end{pmatrix}, F = \begin{pmatrix} \vec{v} \\ M \end{pmatrix}, G = \begin{pmatrix} \alpha \vec{u} + \beta \vec{v} \\ M \end{pmatrix}$$

Then as  $(EB)_{i*} = E_{i*} B$  we have

$$EB = \begin{pmatrix} \vec{u} B \\ M \end{pmatrix}, FB = \begin{pmatrix} \vec{v} B \\ M \end{pmatrix}, GB = \begin{pmatrix} \alpha (\vec{u} B) + \beta (\vec{v} B) \\ M \end{pmatrix}$$

Since  $\det$  depends linearly on 1st row (I) we have

$$\boxed{\det(GB) = \alpha \det(EB) + \beta \det(FB)}$$

$$\text{So } \frac{\det(GB)}{\det(B)} = \frac{\det(EB)}{\det(B)} + \frac{\det(FB)}{\det(B)}$$

$$d(G) = d(E) + d(F)$$

So Prop I holds for  $d$ . ✓

(2) [10 pts] True or false? If true give a brief justification. If false provide a counterexample.

**FALSE**

(a)  $\det(A+B)\det(A-B) = \det(A^2 - B^2)$ .

Notice that  $(A+B)(A-B) = A^2 + BA - AB - B^2 \neq A^2 - B^2$  unless  $AB = BA$ .

So for a counterexample we need  $A, B$  that do not commute.

EX  $A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, B = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$

$$\left. \begin{aligned} \det(A+B) &= \det \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = -1 \\ \det(A-B) &= \det \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} = +1 \end{aligned} \right\} \det(A+B)\det(A-B) = -1$$

~~det~~ But  $A^2 = B^2 = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$  So  $\det(A^2 - B^2) = 0 \neq -1$ .

(b) Let  $\mathbf{v} = (2, 3)^T$ . In the standard basis  $B$  for  $\mathbb{R}^2$ , the matrix of the projection operator  $P_{\mathbf{v}} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  onto the span of  $\mathbf{v}$  is

$$[P_{\mathbf{v}}]_B = \begin{pmatrix} 4 & 6 \\ 6 & 9 \end{pmatrix}$$

The Projection operator  $P_{\mathbf{v}}$  has the property

that  $P_{\mathbf{v}}(\vec{v}) = \vec{v} \Rightarrow [P_{\vec{v}}]_B \begin{pmatrix} 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 2 \\ 3 \end{pmatrix}$

~~NOT~~ BUT

$$[P_{\vec{v}}] \begin{pmatrix} 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 4 & 6 \\ 6 & 9 \end{pmatrix} \begin{pmatrix} 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 26 \\ 39 \end{pmatrix} \neq \begin{pmatrix} 2 \\ 3 \end{pmatrix}$$

So **FALSE**

In fact  $[P_{\vec{v}}] = \frac{\vec{v} \vec{v}^T}{|\vec{v}|^2} = \frac{1}{13} \begin{pmatrix} 4 & 6 \\ 6 & 9 \end{pmatrix}$

(3) [12 pts] For the linear operator  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  defined by  $T(x, y) = (x - y, 2x + 4y)$ , calculate the matrix,  $[T]_{\mathcal{B}}$ , of  $T$  in the basis  $\mathcal{B} = \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \end{pmatrix} \right\}$ .

$$[T] = [T]_{\mathcal{B}} \text{ satisfies}$$

$$T(\vec{u}_i) = \sum_{j=1}^n [T]_{ji} \vec{u}_j$$

where  $\vec{u}_1, \vec{u}_2$  is the basis  $\mathcal{B}$ .

So

$$T\left(\begin{pmatrix} 1 \\ 1 \end{pmatrix}\right) = \begin{pmatrix} 0 \\ 6 \end{pmatrix} = [T]_{11} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + [T]_{21} \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix} [T]_{*1}$$

$$T\left(\begin{pmatrix} 2 \\ 1 \end{pmatrix}\right) = \begin{pmatrix} 1 \\ 8 \end{pmatrix} = [T]_{12} \begin{pmatrix} 1 \\ 1 \end{pmatrix} + [T]_{22} \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix} [T]_{*2}$$

NOW to find  $[T]_{*1}$ :

$$\left( \begin{array}{cc|c} 1 & 2 & 0 \\ 1 & 1 & 6 \end{array} \right) \xrightarrow[R2 = -R2]{R2 = R2 - R1} \left( \begin{array}{cc|c} 1 & 2 & 0 \\ 0 & -1 & -6 \end{array} \right) \xrightarrow{R1 = R1 - 2R2} \left( \begin{array}{cc|c} 1 & 0 & 12 \\ 0 & -1 & -6 \end{array} \right)$$

$$\text{So } [T]_{*1} = \begin{pmatrix} 12 \\ -6 \end{pmatrix}$$

And for  $[T]_{*2}$  using same row ops

$$\left( \begin{array}{cc|c} 1 & 2 & 1 \\ 1 & 1 & 8 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 1 & 2 & 1 \\ 0 & -1 & -7 \end{array} \right) \rightarrow \left( \begin{array}{cc|c} 1 & 0 & 15 \\ 0 & -1 & -7 \end{array} \right)$$

$$\text{So } [T]_{\mathcal{B}} = \begin{pmatrix} 12 & 15 \\ -6 & -7 \end{pmatrix}$$

(4) [18 pts] Let  $P$  be the matrix

$$P = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 6 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

(a) Calculate  $\det(P)$  using

(i) Row operations

$$\begin{vmatrix} 1 & 2 & 3 \\ 0 & 6 & 6 \\ 7 & 8 & 9 \end{vmatrix} \xrightarrow{R_3 \rightarrow R_3 - 7R_1} \begin{vmatrix} 1 & 2 & 3 \\ 0 & 6 & 6 \\ 0 & -6 & -12 \end{vmatrix}$$

$$\xrightarrow{R_3 \rightarrow R_3 + R_2} \begin{vmatrix} 1 & 2 & 3 \\ 0 & 6 & 6 \\ 0 & 0 & -6 \end{vmatrix} = 1 \cdot 6 \cdot (-6) = -36$$

(ii) Block determinants based on the blocking

$$P = \begin{pmatrix} A & B \\ C & D \end{pmatrix}, \quad \text{where } A \text{ is } 1 \times 1 \text{ and } D \text{ is } 2 \times 2.$$

Since  $A$  is  $1 \times 1$ ,  $A^{-1} = A = [1]$ . So let's use

$$\det(P) = \det(A) \det(D - CA^{-1}B)$$

$$= 1 \cdot \det(D - CB)$$

$$= \begin{vmatrix} \begin{pmatrix} 6 & 6 \\ 8 & 9 \end{pmatrix} - \begin{pmatrix} 0 \\ 7 \end{pmatrix} \begin{pmatrix} 2 & 3 \end{pmatrix} \end{vmatrix}$$

$$= \begin{vmatrix} \begin{pmatrix} 6 & 6 \\ 8 & 9 \end{pmatrix} - \begin{pmatrix} 0 & 0 \\ 14 & 21 \end{pmatrix} \end{vmatrix} = \begin{vmatrix} 6 & 6 \\ -6 & -12 \end{vmatrix} = -36$$



(iii) A cofactor expansion.

Since there is a 0 in Col 1 (and Row 2)  
use cofactor expansion based on Col 1  
(or Row 2)

$$\begin{vmatrix} 1 & 2 & 3 \\ 0 & 6 & 6 \\ 7 & 8 & 9 \end{vmatrix} = 1 \begin{vmatrix} 6 & 6 \\ 8 & 9 \end{vmatrix} + 7 \begin{vmatrix} 2 & 3 \\ 6 & 6 \end{vmatrix}$$
$$= 34 - 48 + 7(12 - 18)$$
$$= -36$$

(b) What is  $\det(\mathbf{P}^T \mathbf{P})$ , and why?

$$\det(\mathbf{P}^T \mathbf{P}) = \det(\mathbf{P}^T) \det(\mathbf{P}) \quad \text{by Product Rule}$$
$$= (\det(\mathbf{P}))^2 \quad \text{as } \det(\mathbf{P}^T) = \det(\mathbf{P})$$

(5) [10 pts] Let  $T: V \rightarrow W$  be a linear transformation between finite-dimensional vector spaces  $V$  and  $W$ . Let  $B$  be a basis for  $V$  and let  $B'$  be a basis for  $W$ . Define the matrix  $[T]_{B'B}$  of  $T$  with respect to these two bases, and prove that

$$[T(u)]_{B'} = [T]_{B'B} [u]_B.$$

Let  $\vec{v}_1, \dots, \vec{v}_n$  be a basis for  $V$   
and  $\vec{w}_1, \dots, \vec{w}_m$  a basis for  $W$ .

The matrix  $[T]_{B'B}$  is defined by the equation

$$T(\vec{v}_i) = \sum_{j=1}^m ([T]_{B'B})_{ji} \vec{w}_j \quad \textcircled{1}$$

$([T]_{B'B})_{ji}$  =  $j$ th coefficient of  $T(\vec{v}_i)$  in the basis  $w_1, \dots, w_m$ .

Equivalently  $[T]_{B'B} = ([T(\vec{v}_1)]_{B'} \mid \dots \mid [T(\vec{v}_n)]_{B'})$ .

Suppose  $[\vec{u}]_B = \begin{pmatrix} \alpha_1 \\ \vdots \\ \alpha_n \end{pmatrix}$  so that  $\vec{u} = \sum_{i=1}^n \alpha_i \vec{v}_i$ .

Then

$$T(\vec{u}) = \sum_{i=1}^n \alpha_i T(\vec{v}_i) \quad \text{by linearity of } T$$

$$= \sum_{i=1}^n \alpha_i \sum_{j=1}^m ([T]_{B'B})_{ji} \vec{w}_j \quad \text{by } \textcircled{1}$$

$$= \sum_{j=1}^m \left( \sum_{i=1}^n ([T]_{B'B})_{ji} \alpha_i \right) \vec{w}_j = \sum_{j=1}^m ([T]_{B'B} [\vec{u}]_B)_j \vec{w}_j$$

But

$$T(\vec{u}) = \sum ([T(\vec{u})]_{B'})_j \vec{w}_j$$

So by uniqueness

$$[T]_{B'B} [\vec{u}]_B = [T(\vec{u})]_{B'}$$

(6) [8 pts] Suppose  $A$  is a square matrix whose entries are differentiable functions of a real variable  $t$ , that is,  $A_{ij} = A_{ij}(t)$ . Prove that  $\det A$  is also a differentiable function of  $t$ .

We know

$$\det(A(t)) = \sum_p \sigma(p) A_{1p_1}(t) A_{2p_2}(t) \cdots A_{np_n}(t)$$

where we sum over all  $n!$  permutations

$p$  of  $(1, 2, \dots, n)$  and where  $\sigma(p) = \pm 1$  is the parity of  $p$ .

So  $\det(A(t))$  is a polynomial in the  $A_{ij}(t)$ ,  
i.e.  $\det(A(t))$  is a sum of products of  
differentiable functions and so is  
differentiable by the sum, product  
and Chain Rules for differentiation.

(7) [12 pts] The least squares quadratic fit to  $m$  data points  $(t_1, y_1), (t_2, y_2), \dots, (t_m, y_m)$  in  $\mathcal{R}^2$  is the quadratic function  $y = f(t) = \alpha + \beta t + \gamma t^2$  for which the parameter vector  $(\alpha, \beta, \gamma)$  is the global minimum of the function

$$Q = Q(\alpha, \beta, \gamma) = \sum_{i=1}^m (\alpha + \beta t_i + \gamma t_i^2 - y_i)^2.$$

(a) Let  $\mathbf{x} = (\alpha, \beta, \gamma)^T$ . Find an  $m \times 1$  vector  $\mathbf{y}$  and an  $m \times 3$  matrix  $\mathbf{A}$  so that

$$Q = Q(\mathbf{x}) = \|\mathbf{Ax} - \mathbf{y}\|^2.$$

Let  $\mathbf{y} = \begin{pmatrix} y_1 \\ 1 \\ y_m \end{pmatrix}$   $\mathbf{A} = \begin{pmatrix} 1 & t_1 & t_1^2 \\ 1 & t_2 & t_2^2 \\ 1 & 1 & 1 \\ 1 & t_m & t_m^2 \end{pmatrix}$

Then

$$\begin{matrix} \mathbf{A} & \mathbf{x} \\ m \times 3 & 3 \times 1 \end{matrix} = \begin{pmatrix} \alpha + \beta t_1 + \gamma t_1^2 \\ \vdots \\ \alpha + \beta t_m + \gamma t_m^2 \end{pmatrix}$$

So

$$\begin{aligned} \|\mathbf{Ax} - \mathbf{y}\|^2 &= \sum_{i=1}^m \left( (\mathbf{Ax})_i - y_i \right)^2 \\ &= \sum_{i=1}^m \left( \alpha + \beta t_i + \gamma t_i^2 - y_i \right)^2 \end{aligned}$$

(b) By differentiating  $Q(\mathbf{x})$  with respect to the  $i$ -th coordinate  $x_i$  of  $\mathbf{x}$ , prove that the minimizer of  $Q$  satisfies the normal equations  $A^T A \mathbf{x} = A^T \mathbf{y}$ .

The most elegant proof is:

$$Q(\mathbf{x}) = \|A\mathbf{x} - \mathbf{y}\|^2 = (A\mathbf{x} - \mathbf{y})^T (A\mathbf{x} - \mathbf{y})$$

$$= (\mathbf{x}^T A^T - \mathbf{y}^T) (A\mathbf{x} - \mathbf{y})$$

$$= \mathbf{x}^T A^T A \mathbf{x} - \mathbf{y}^T A \mathbf{x} - \mathbf{x}^T A \mathbf{y} + \mathbf{y}^T \mathbf{y}$$

$$Q(\mathbf{x}) = \mathbf{x}^T A^T A \mathbf{x} - 2\mathbf{y}^T A \mathbf{x} + \mathbf{y}^T \mathbf{y}$$

$$\text{as } (\mathbf{x}^T A \mathbf{y}) \stackrel{1 \times 1}{=} (\mathbf{x}^T A \mathbf{y})^T = \mathbf{y}^T A^T \mathbf{x}$$

So

$$\begin{aligned} 0 = \frac{\partial Q}{\partial x_i} &= \frac{\partial \mathbf{x}^T}{\partial x_i} A^T A \mathbf{x} + \mathbf{x}^T A^T A \frac{\partial \mathbf{x}}{\partial x_i} - 2\mathbf{y}^T A \frac{\partial \mathbf{x}}{\partial x_i} \\ &= \mathbf{e}_i^T A^T A \mathbf{x} + \mathbf{x}^T A^T A \mathbf{e}_i - 2\mathbf{y}^T A \mathbf{e}_i \\ &= 2\mathbf{e}_i^T (A^T A \mathbf{x} - A \mathbf{y}) \\ &= 2(A^T A \mathbf{x} - A \mathbf{y})_{+i} \quad \text{for all } i=1-m \end{aligned}$$

So  $A^T A \mathbf{x} = A \mathbf{y}$  holds

Pledge: I have neither given nor received aid on this exam.

Signature: \_\_\_\_\_