# Homework #6 Solutions

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# 1 For submission to Thayer Anderson

**Problem 1.1.** Akin to the tensor product of functionals described in class, you can also define a tensor product of vectors:  $V \otimes W$  is the vector space populated by formal sums of elements of the form  $v \otimes w$ , subject to the relations

$$(v_1 + v_2) \otimes w = v_1 \otimes w + v_2 \otimes w,$$

$$(kv) \otimes w = k(v \otimes w) = v \otimes (kw),$$

$$v \otimes (w_1 + w_2) = v \otimes w_1 + v \otimes w_2.$$

(The tensor described in class is thus this definition of tensor, applied to the dual space  $\mathbb{R}^*$ .) Similarly, you can also build a wedge product of vectors  $v_1, \ldots, v_k \in V$  as a particular kind of tensor:

$$v_1 \wedge \dots \wedge v_k = \frac{1}{k!} \sum_{\sigma \text{ a permutation of } \{1, \dots, k\}} \operatorname{sign}(\sigma) \cdot (v_{\sigma 1} \otimes \dots \otimes v_{\sigma k}),$$

considered as a vector in  $(\mathbb{R}^n)^{\otimes k}$ . (Again, the wedge product defined in class is thus this definition of wedge product, applied to the dual space  $\mathbb{R}^*$ .)

- 1. Show that a linear map  $V \otimes W \to U$  is identical information to a bilinear function  $V \times W \to U$ .
- 2. Let  $e_1, \ldots, e_n$  be the standard basis for  $\mathbb{R}^n$ , and let  $\varphi_1, \ldots, \varphi_n$  be the dual basis under the standard inner product. Demonstrate the identity

$$(\varphi_{j_1} \wedge \dots \wedge \varphi_{j_k})(e_{j_1} \wedge \dots \wedge e_{j_k}) = 1$$

by considering the input as a vector in  $(\mathbb{R}^n)^{\otimes k}$  and the function as a vector in  $((\mathbb{R}^n)^*)^{\otimes k} = ((\mathbb{R}^n)^{\otimes k})^*$ .

- 3. Remark on the role of the binomial/factorial coefficient in the definition of the wedge product. If that factor were omitted, what would the above pairing evaluate to instead? Why?
- 4. Note that if  $\{v_1, \ldots, v_k\}$  and  $\{w_1, \ldots, w_\ell\}$  are basis of V and W respectively, then  $\{v_i \otimes w_j\}$  forms a basis of  $V \otimes W$ . Conclude more generally that if  $v_1, \ldots, v_k$  is a k-tuple of vectors in  $\mathbb{R}^n$  and  $\psi_1, \ldots, \psi_k$  is a k-tuple of linear functionals on  $\mathbb{R}^n$ , then the following two values agree:

$$(\psi_1 \wedge \dots \wedge \psi_k)(v_1, \dots, v_k) = (\psi_1 \wedge \dots \wedge \psi_k)(v_1 \wedge \dots \wedge v_k).$$

Solution. 1. Given a linear map  $g: V \otimes W \to U$ , define a function  $f: V \times W \to U$  by  $f(u, w) = g(u \otimes w)$ . In the other direction, a function  $g: V \times W \to U$  defines a linear function from the space of all symbols  $v \otimes w$  without any relations imposed to U, and then this function descends to a linear function on the quotient space  $V \otimes W$  because the bilinearity of g causes it to vanish on the subspace of symbols generated by the relations specified above. 2. This is a rather manual computation.

$$(\varphi_{j_1} \wedge \dots \wedge \varphi_{j_k})(e_{j_1} \wedge \dots \wedge e_{j_k}) = \left(\frac{k!}{k!} \sum_{\eta} \operatorname{sign}(\eta)(\varphi_{\eta 1} \otimes \dots \otimes \varphi_{\eta k})\right) \left(\frac{1}{k!} \sum_{\sigma} \operatorname{sign}(\sigma)(v_{\sigma 1} \otimes \dots \otimes v_{\sigma k})\right)$$
$$= \frac{1}{k!} \sum_{\sigma,\eta} (\varphi_{\eta 1} \otimes \dots \otimes \varphi_{\eta k})(v_{\sigma 1} \otimes \dots \otimes v_{\sigma k}).$$

This sum is nonzero if and only if  $\sigma = \eta$ , in which case is gives  $\frac{1}{k!}$ , and this happens k! many times.

- 3. The fraction is there to handle this double-counting appearing from the double summation.
- 4. Each vector  $v_j$  and each functional  $\varphi_j$  decomposes as a sum of standard basis vectors, and then the (ECP)above results apply.

**Problem 1.2.** For  $f, g: \mathbb{R}^n \to \mathbb{R}$ , demonstrate the *Leibniz rule* 

$$d(f \cdot g) = df \cdot g + f \cdot dg.$$

Solution. This is a totally formal consequence of the Leibniz rule for the full derivative of functions  $\mathbb{R}^n \to \mathbb{R}$ , solved way back in Problem 2.1.1.2. Recall  $D_a(f \cdot g) = g \cdot D_a f + f \cdot D_a g$ . By reading the coordinate functions of the 1-form  $d(f \cdot g)$  out of  $D(f \cdot g)$ , we have

$$d(f \cdot g) = \sum_{j} [D(f \cdot g)]_j \, \mathrm{d}x_j = \sum_{j} \left( \frac{\partial f}{\partial x_j} \cdot g + f \cdot \frac{\partial g}{\partial x_j} \right) \, \mathrm{d}x_j = \, \mathrm{d}f \cdot g + f \cdot \, \mathrm{d}g. \tag{ECP}$$

#### 2 For submission to Davis Lazowski

**Problem 2.1.** Recall the definition of the cross-product in  $\mathbb{R}^3$ : for  $v, w \in \mathbb{R}^3$ ,  $v \times w$  is the vector representing the linear functional

$$\iota \mapsto \det( |u| |v| |w|)$$

1 under the standard inner product. Demonstrate the following truckload of identities:

1.

$e_1 \times e_1 = 0,$	$e_1 \times e_2 = e_3,$	$e_1 \times e_3 = -e_2,$
$e_2 \times e_1 = -e_3,$	$e_2 \times e_2 = 0,$	$e_2 \times e_3 = e_1,$
$e_3 \times e_1 = e_2,$	$e_3 \times e_2 = -e_1,$	$e_3 \times e_3 = 0.$

- 2.  $v \times w = (v_2w_2 v_3w_2)e_1 + (v_3w_1 v_1w_3)e_2 + (v_1w_2 v_2w_1)e_3$ .
- 3.  $\|v \times w\| = \|v\| \cdot \|w\| \cdot \sin \theta$ , where  $\theta$  is the angle formed by v and w as rays intersecting at the origin. Conclude  $\langle v \times w, v \rangle = 0$  and  $\langle v \times w, w \rangle = 0$ .
- 4. The juggling identities:  $\langle v, w \times u \rangle = \langle w, u \times v \rangle = \langle u, v \times w \rangle$ .
- 5. The associative identities:  $v \times (w \times u) = \langle v, u \rangle w \langle v, w \rangle u$  and  $(v \times w) \times u = \langle v, u \rangle w \langle w, u \rangle v$ .
- 6.  $||v \times w|| = \sqrt{||v||^2 \cdot ||w||^2 \langle v, w \rangle^2}$ .

Solution. I will note two lemmas which make all the identities easy.

First, the cross product is linear in both elements. This is clear since  $u \to \det \begin{pmatrix} u & v & w \end{pmatrix}$  is a linear functional.

Second, the cross product is rotation invariant, in the sense that if R is a rotation, then  $R(v \times W) = (Rv) \times (Rw)$ .

As a proof, if v and w are not linearly independent, then Rv and Rw aren't either, so  $v \times w = 0$  and  $Rv \times Rw = 0 = R0$ , so this is trivial.

Otherwise, v, w and  $v \times w$  are linearly independent. (To prove this, observe that  $\langle v \times w, v \times w \rangle$  is nonzero.) Therefore, Rv, Rw and  $R(v \times w)$  are linearly independent. So we know that  $Rv \times Rw \propto R(v \times w)$ . So we just need to fix the proportionality constant, and show that it's one.

But since R is a rotation matrix,

$$\det \begin{pmatrix} R(v \times w) & | & Rv & | & Rw \end{pmatrix} = \det \begin{pmatrix} v \times w & | & v & | & w \end{pmatrix}$$

this constant is just one.

Let's use these to prove all the identities.

1. Clearly  $e_j \times e_j = 0$ , since the matrix  $\begin{pmatrix} u & e_j & e_j \end{pmatrix}$  has less than full rank, so has zero determinant. Now suppose we have  $k \neq i \neq j \neq k$ . Then since  $\begin{pmatrix} u & | & e_i & | & e_j \end{pmatrix}$  is only nonzero when u has a  $\hat{e}_k$  component, so we know that  $e_i \times e_j = \alpha_{ij}e_k$ ; we just need to fix  $\alpha_{ij}$ .

For k = 1, i = 2, j = 3, by inputting  $u = e_1$ , since the determinant is just the identity, we know that  $1 = \alpha_{23} \langle e_1, e_1 \rangle = \alpha_{23}$ . So  $e_2 \times e_3 = e_1$ .

Also observe that det  $\begin{pmatrix} u & | & e_i & | & e_j \end{pmatrix}$  swaps sign under transposition of rows, since a row transposition matrix has determinant -1.

Therefore  $\alpha_{ij} = \operatorname{sign}(\sigma)$ , where  $\sigma$  is the permutation sending (k, i, j) to (1, 2, 3). This demonstrates the remaining identities.

- 2. By linearity in both elements, we only need to calculate  $v_i \times w_j$ . By the previous part, this is  $\alpha_{ij}v_iw_je_k$ . Summing all these individual parts, we get the required identity.
- 3. . Since rotation doesn't change absolute value, we might as well rotate so that v and w are on the same plane– i.e.  $Rv = v_1e_1$ , and  $Rw = w_1e_1 + w_2e_2$ .

Then

$$Rv \times Rw = v_1w_2e_3$$

So that

$$\begin{aligned} & \frac{||v \times w||}{||v||||w||} \\ &= \frac{||Rv \times Rw||}{||Rv||||Rw||} \\ &= \frac{|v_1||w_2|}{|v_1|\sqrt{w_1^2 + w_2^2}|} \\ &= \frac{|w_2|}{\sqrt{w_1^2 + w_2^2}} \\ &= \sin \theta \end{aligned}$$

as desired.

4. This is equivalent to the statement that

 $\det (v \mid w \mid u) = \det (w \mid u \mid v) = \det (u \mid v \mid w)$ 

Which is true because each matrix is related by two row transpositions, and the determinant of the matrix which performs two row transpositions is 1 (since it is the product of two transposition matrices of determinant -1.)

5. First, observe that (with i, j, k not necessarily distinct)

$$e_i \times (e_j \times e_k) = \langle e_i, e_k \rangle e_j - \langle e_i, e_j \rangle e_k$$

directly by computing the cross product. Now

$$e_i \times (e_j \times u) = e_i \times (e_j \times (u_1e_1 + u_2e_2 + u_3e_3))$$
  
=  $e_i \times [u_1(e_j \times e_1) + u_2(e_j \times e_2) + u_3(e_j \times e_3)]$   
=  $\sum_{k=1}^3 u_k e_i \times (e_j \times e_k)$   
=  $\sum_{k=1}^3 \langle e_i, u_k e_k \rangle e_j - \langle e_i, e_j \rangle (u_k e_k)$   
=  $\langle e_i, u \rangle e_j - \langle e_i, e_j \rangle u$ 

Now use linearity the exact same way to replace  $e_i$  and  $e_j$  with v and w. Prove the second associative identity the exact same way.

6. Again, rotate so that  $Rv = v_1e_1$ . and  $Rw = w_1e_1 + w_2e_2$ . Then we want to show

$$||Rv \times Rw|| = |v_1w_2|$$

$$\sqrt{v_1^2(w_1^2 + w_2^2) - (v_1w_1)^2} = |v_1w_2|$$

$$= v, \langle Rv, Rw \rangle = \langle v, w \rangle, \text{ done.}$$
(DL)

which is true. Then since  $||Rv|| = v, \langle Rv, Rw \rangle = \langle v, w \rangle$ , done.

**Problem 2.2.** Recall the polar coordinate transformation  $x(r, \theta) = r \cos \theta$  and  $y(r, \theta) = r \sin \theta$ , defined for  $0 < \theta < 2\pi$  and r > 0. Prove that where  $\theta$  is defined as a function of x and y, we have

$$d\theta = \frac{-y}{x^2 + y^2} \,\mathrm{d}x + \frac{x}{x^2 + y^2} \,\mathrm{d}y.$$

Solution. Recall that we have a lemma saying

$$u^*(dx_i) = \sum_{j=1}^m \frac{\partial u_i}{\partial x_j} dx_j$$

Let u be the map  $(x, y) \to (\theta, r)$ , i.e.  $(x, y) \to (\arctan \frac{y}{x}, \sqrt{x^2 + y^2})$ . Then this lemma applied to dx says

$$d\theta = \frac{\partial}{\partial x} [\arctan(y/x)] dx + \frac{\partial}{\partial y} [\arctan(y/x)] dy$$

Computing the derivatives, we have the requisite

$$d\theta = -\frac{y}{x^2 + y^2}dx + \frac{x}{x^2 + y^2}dy$$
 (DL)

# 3 For submission to Handong Park

**Problem 3.1.** For  $f: \mathbb{R}^n \to \mathbb{R}$ , we define a vector field  $\nabla f$  by the formula

$$\nabla f \colon (a \in \mathbb{R}^n) \mapsto \begin{pmatrix} \left. \frac{\partial f}{\partial x_1} \right|_{x=a} \\ \vdots \\ \left. \frac{\partial f}{\partial x_n} \right|_{x=a} \end{pmatrix} \in T_a \mathbb{R}^n.$$

Recall also the directional derivative from Homework #3: given a tangent vector  $v \in T_a \mathbb{R}^n$ , we set

$$\mathbb{D}_a^v f = \lim_{t \to 0} \frac{f(a+tv) - f(a)}{t}.$$

Conclude  $\mathbb{D}_a^v(f) = \langle v, \nabla f(a) \rangle$ , and hence that  $\nabla f(a)$  is the direction of greatest ascent<sup>1</sup>.

Solution. Our conclusion from Homework #3 was  $\mathbb{D}_a^v f = (D_a f)(v)$ . We know that  $D_a f$  admits expression as a matrix:

$$D_a f = \left(\begin{array}{cc} \frac{\partial f}{\partial x_1} & \cdots & \frac{\partial f}{\partial x_n} \end{array}\right),$$

and hence we have the evaluation formula

$$(D_a f)(v) = \begin{pmatrix} \frac{\partial f}{\partial x_1} & \cdots & \frac{\partial f}{\partial x_n} \end{pmatrix} \cdot \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} = \langle v, \nabla f(a) \rangle.$$

By writing this as an inner product, we can appeal to Cauchy–Schwarz to see that the expression  $\langle v, \nabla f(a) \rangle$ is maximized over unit-length tangent vectors v by  $v = (\nabla f(a))/\|\nabla f(a)\|$ . (ECP)

**Problem 3.2.** Let  $f: U \to \mathbb{R}^n$  be a differentiable function with a differentiable inverse  $f^{-1}: f(U) \to \mathbb{R}^n$ . If every closed form on U is exact, show that the same is true for f(U).

Solution. Let  $\omega$  be a closed form defined on f(U). Then  $(f^{-1})^*\omega$  is a form on U, and because exterior differentiation commutes with pullback we have

$$d((f^{-1})^*\omega) = (f^{-1})^*d\omega = (f^{-1})^*0 = 0$$

shows that  $(f^{-1})^*\omega$  is closed as well. Since on U every closed form is exact, there exists some  $\xi$  with  $d\xi = (f^{-1})^*\omega$ . We claim that  $f^*\xi$  witnesses  $\omega$  as an exact form:

$$df^*\xi = f^*d\xi = f^*(f^{-1})^*\omega = (f^{-1} \circ f)^*\omega = \mathrm{id}^*\omega = \omega.$$
(ECP)

## 4 For submission to Rohil Prasad

**Problem 4.1.** Let  $c: [0,1] \to (\mathbb{R}^n)^n$  be a 1-parameter continuous family of families of n vectors in  $\mathbb{R}^n$ , and suppose that  $c(t) = \{c_1(t), \ldots, c_n(t)\}$  is a basis of  $\mathbb{R}^n$  for each  $0 \le t \le 1$ . Show that the orientation of each basis must be the same, i.e. the value

$$[c_1(t),\ldots,c_n(t)] = \operatorname{sign}(\det(c_1(t)|\ldots|c_n(t)))$$

is constant even as t varies.

Solution. Postcompose c with the determinant map to get a continuous map  $\varphi: [0,1] \to \mathbb{R}$ .

Since c(t) is a basis of  $\mathbb{R}^n$  for every t, this determinant must be nonzero for every t.

Now assume for the sake of contradiction that the orientation switches. This implies that there exists  $s, t \in [0, 1]$  such that one of  $\varphi(s), \varphi(t)$  is less than 0 and one is greater than 0.

Applying the intermediate value theorem to  $\varphi$ , this tells us that there is some  $a \in [s, t]$  such that  $\varphi(a) = 0$ . However, we assumed in the problem statement that  $\varphi$  is nonzero everywhere, so we arrive at a contradiction and the orientation remains constant. (RP)

<sup>&</sup>lt;sup>1</sup>Or "direction of fastest change", if you prefer.

**Problem 4.2.** In class, we "proved" by example that any quadratic form  $Q: \mathbb{R}^n \to \mathbb{R}$  can be written in the form

$$Q = a_1^2 + \dots + a_k^2 - b_1^2 - \dots b_l^2$$

for a family of linearly independent linear functionals  $a_*$  and  $b_*$ . Complete our discussion by turning our examples into an honest proof. (Don't worry about the invariance of the signature; just work on this existence half.)

Solution. We can do this by induction on n.

For n = 1, Q(x) is simply equal to  $Cx^2$  for some constant C.

Now take a quadratic form  $Q(x_1, \ldots, x_n)$ . We can write it as  $Cx_n^2 + x_n \cdot A(x_1, \ldots, x_{n-1}) + B(x_1, \ldots, x_{n-1})$ . Here C is a constant, A is a linear functional, and B is a quadratic form.

Then we can set u to be the linear functional  $\sqrt{C}x_n + A(x_1, \dots, x_{n-1})/2\sqrt{C}$ . It follows that we can write  $Q = u^2 + B(x_1, \dots, x_{n-1}) - \frac{1}{4C}A(x_1, \dots, x_{n-1})^2$ . The term  $B(x_1, \dots, x_{n-1}) - \frac{1}{4C}A(x_1, \dots, x_{n-1})^2$  is a quadratic form in n-1 variables, so we can apply our induction hypothesis and obtain the desired formula for Q. (RP)