

## Tools for Stat Theory HW7

## 1

Prove Lemma 12.1.1: Let  $Y$  represent a matrix in a linear space  $V$ , let  $U$  and  $W$  represent subspaces of  $V$ , and take  $\{X_1, \dots, X_s\}$  to be a set of matrices that spans  $U$  and  $\{Z_1, \dots, Z_t\}$  to be a set that spans  $W$ . Then,  $Y \perp U$  if and only if  $Y \cdot X = 0$  for  $i = 1, \dots, s$ ; that is,  $Y$  is orthogonal to  $U$  if and only if  $Y$  is orthogonal to each of the matrices  $X_1, \dots, X_s$ . And, similarly,  $U \perp W$  if and only if  $X \cdot Z = 0$  for  $i = 1, \dots, s$  and  $j = 1, \dots, t$ ; that is,  $U$  is orthogonal to  $W$  if and only if each of the matrices  $X_1, \dots, X_s$  is orthogonal to each of the matrices  $Z_1, \dots, Z_t$ .

Since  $\{X_1, \dots, X_s\}$  span  $U$ , then  $U$  can be represented as a linear combination of  $\{X_1, \dots, X_s\}$ :  $U = a_1X_1 + \dots + a_sX_s$ .

1) Prove  $Y$  is orthogonal to  $U$  if  $Y$  is orthogonal to each of the matrices  $X_1, \dots, X_s$

If  $Y$  is orthogonal to each of the matrices  $X_1, \dots, X_s$ , then

$$Y \cdot a_1X_1 + \dots + Y \cdot a_sX_s = 0$$

. By the property of matrix multiplication, we can convert it to the following form

$$Y \cdot (a_1X_1 + \dots + a_sX_s) = 0$$

Since  $U$  can be represented as a linear combination  $U = a_1X_1 + \dots + a_sX_s$ , then if  $Y$  is orthogonal to all linear combinations of  $\{X_1, \dots, X_s\}$ , then  $Y$  is orthogonal to  $U$ .

2) Prove  $Y$  is orthogonal to each of the matrices  $X_1, \dots, X_s$  if  $Y$  is orthogonal to  $U$  -0.5

If  $Y$  orthogonal to  $U$ , then  $Y$  orthogonal to  $\{X_1, \dots, X_s\}$  by definition because  $\{X_1, \dots, X_s\} \subset U$ .

## 2

For the example of Section 12.2c, recompute the projection of  $y$  on  $U$  by taking  $X$  to be the  $3 \times 2$  matrix

$$\begin{pmatrix} 0 & -2 \\ 3 & 2 \\ 6 & 4 \end{pmatrix}$$

....

and carrying out the two steps:

- (1) compute the solution to the normal solutions  $X'Xb = X'y$ ;
- (2) postmultiply  $X$  by the solution you computed in Step (1).

1) Compute solution to normal solutions

$$X'y = \begin{bmatrix} 6 & 3 & 0 \\ 4 & 2 & -2 \end{bmatrix} \begin{bmatrix} 3 \\ -\frac{38}{5} \\ \frac{74}{5} \end{bmatrix}$$

$$b = X^{-1}(X')^{-1}X'y$$

```
X <- matrix(c(0,3,6, -2,2,4), nrow=3)
X_prime <- t(X)
y <- matrix(c(3, -38/5, 74/5), ncol =1)
```

```
XtX <- t(X) %*% X
Xty <- t(X) %*% y

b <- solve(XtX, Xty)
```

```
print("SOLUTION:")
```

```
## [1] "SOLUTION:"
```

```
print(b)
```

```
##           [,1]
## [1,]  2.466667
## [2,] -1.500000
```

```
print("XtXb:")
```

```
## [1] "XtXb:"
```

```
print(XtX %*% b)
```

```
##           [,1]
## [1,]    66
## [2,]    38
```

```
print("Xty")
```

```
## [1] "Xty"
```

```
print(Xty)
```

```
##           [,1]
## [1,]    66
## [2,]    38
```

2) Postmultiply  $X$  by  $b$

```
print(X %*% b)
```

```
##      [,1]
## [1,]  3.0
## [2,]  4.4
## [3,]  8.8
```

### 3

Prove Theorem 13.2.6: If an  $n \times n$  matrix  $B = \{b_{ij}\}$  is formed from an  $n \times n$  matrix  $A = \{a_{ij}\}$  by interchanging two rows or two columns of  $A$ , then  $|B| = -|A|$ .

We will first try to prove the theorem using a special case of swapping 2 rows of  $A$  to obtain  $B$ , as follows:

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

$$B = \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \dots & \dots & \dots & \dots \\ b_{n1} & b_{n2} & \dots & b_{nn} \end{bmatrix} = \begin{bmatrix} a_{21} & a_{22} & \dots & a_{2n} \\ a_{11} & a_{12} & \dots & a_{1n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

We can expand  $|A|$  along its first row to obtain its value:  $|A| = \sum_{j=1}^n a_{1j}(-1)^{1+j}|A_{1j}|$ . The property of the cofactor expansion is such that whichever row or column we choose to expand by, the results are the same.

To obtain  $|B|$  we can expand along the 2nd row as well,  $|B| = \sum_{j=1}^n b_{2j}(-1)^{2+j}|B_{2j}| = \sum_{j=1}^n a_{1j}(-1)^{2+j}|A_{1j}|$

In this case, we show that  $|B_{2j}| = |A_{1j}|$  because it's along the same row. Since the cofactor expansions along the other row are the same,  $|B| = -|A|$ .

Now that we show that  $|B| = -|A|$  in the case that  $B$  is just  $A$  with the 1st and 2nd row swapped. To find  $B$  in the general case. We can assume that there is row  $f$  and row  $h$ . Between row  $f$  and row  $h$  there are  $k$  rows. We can swap the  $h$ th row with the rows above it one by one until it reaches the original positions of row  $f$ . Now  $f$  will be at  $k+2$ th row after  $k$  swaps. To swap  $h$  &  $f$  back we need  $k+1$  more swaps so  $(-1)^{2k+1}$ .

### 4

Numerically test: determinant keeps invariant when we expand by every row or every column (with order  $n \geq 6$ , randomly generate integers as matrix elements), i.e.,  $\det(A) = \sum_{j=1}^n a_{ij}(-1)^{i+j} \det(A_{ij})$  where  $A_{ij}$  is the submatrix of  $A$  by deleting the  $i$ th row and  $j$ th column of  $A$ ,  $(-1)^{i+j} \det(A_{ij})$  is the **cofactor of matrix A** or **algebraic cofactor of matrix A**, apply R or Python function to calculate  $\det(A_{ij})$  is available.

```
import numpy as np

def calc_det_row(A, i):
    n = A.shape[0]
    det = 0
    for j in range(0, n):
        A_ij = np.delete(A, i, axis=0)
```

```

    A_ij = np.delete(A_ij, j, axis=1)
    det_A_ij = np.linalg.det(A_ij)
    det += A[i,j] * (-1) ** (i + j + 2) * det_A_ij

    return det

def calc_det_col(A, j):
    n = A.shape[0]
    det = 0
    for i in range(0, n):
        A_ij = np.delete(A, i, axis=0)
        A_ij = np.delete(A_ij, j, axis=1)
        det_A_ij = np.linalg.det(A_ij)
        det += A[i,j] * (-1) ** (i + j + 2) * det_A_ij

    return det

for n in range(6, 9):

    A = np.random.randint(-30, 30, size=(n, n))
    det_A = np.linalg.det(A).round(0) # compute the determinant of A
    print(det_A)

    dets = []
    for i in range(0, n):
        det = calc_det_row(A, i)
        det = det.round(0)
        assert(det_A - det <= 1)
        dets.append(det)
        det = calc_det_col(A, i)
        det = det.round(0)
        assert(det_A - det <= 1)
        dets.append(det)
    print(dets)

```

```

## 66615165.0
## [66615165.0, 66615165.0, 66615165.0, 66615165.0, 66615165.0, 66615165.0, 66615165.0, 66615165.0, 66615165.0]
## -12423391964.0
## [-12423391964.0, -12423391964.0, -12423391964.0, -12423391964.0, -12423391964.0, -12423391964.0, -12423391964.0, -12423391964.0, -12423391964.0]
## 2630165909022.0
## [2630165909022.0, 2630165909022.0, 2630165909022.0, 2630165909022.0, 2630165909022.0, 2630165909022.0, 2630165909022.0, 2630165909022.0, 2630165909022.0]

```

```

print("Determinant is invariant")

```

```

## Determinant is invariant

```