

# HW4

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## Problem Set 1

### 1. PROBLEM SET 1

In this problem, we'll verify using R that SVD and Eigenvalues are related as worked out in the weekly module. Given a  $3 \times 2$  matrix  $A$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ -1 & 0 & 4 \end{bmatrix} \quad (1)$$

write code in R to compute  $X = AA^T$  and  $Y = A^T A$ . Then, compute the eigenvalues and eigenvectors of  $X$  and  $Y$  using the built-in commands in R.

Figure 1:

```
X = AA^T
A<- matrix(c(1,2,3,-1,0,4), ncol = 3 , byrow = TRUE )
A
```

```
##      [,1] [,2] [,3]
## [1,]    1    2    3
## [2,]   -1    0    4
```

```
X <- A %*% t(A)
X
```

```
##      [,1] [,2]
## [1,]   14   11
## [2,]   11   17
```

Calculate Eigen Vectors and Values of X

```
eigen_values_X <- eigen(X)$values
eigen_values_X
```

```
## [1] 26.601802  4.398198
```

```
eigen_vectors_X <- eigen(X)$vectors
eigen_vectors_X
```

```
##      [,1]      [,2]
## [1,] 0.6576043 -0.7533635
## [2,] 0.7533635  0.6576043
```

$Y = A^T A$

```
Y <-t(A) %*% A
Y
```

```
##      [,1] [,2] [,3]
## [1,]    2    2   -1
## [2,]    2    4    6
## [3,]   -1    6   25
```

Calculate Eigen Vectors and Values of Y

```
eigen_values_Y <- eigen(Y)$values
eigen_values_Y
```

```
## [1] 2.660180e+01 4.398198e+00 1.058982e-16
```

```
eigen_vectors_Y <- eigen(Y)$vectors
eigen_vectors_Y
```

```
##      [,1]      [,2]      [,3]
## [1,] -0.01856629 -0.6727903  0.7396003
## [2,]  0.25499937 -0.7184510 -0.6471502
## [3,]  0.96676296  0.1765824  0.1849001
```

Then, compute the left-singular, singular values, and right-singular vectors of **A** using the *svd* command. Examine the two sets of singular vectors and show that they are indeed eigenvectors of **X** and **Y**. In addition, the two non-zero eigenvalues (the 3rd value will be very close to zero, if not zero) of both **X** and **Y** are the same and are squares of the non-zero singular values of **A**.

Figure 2:

$svd(A) = U\Sigma V^*$

```
svd_A <- svd(A)
```

Left Vectors (U) = Eigen Vectors of  $AA^T$

```
eigen_vectors_X[,1:2]
```

```
##      [,1]      [,2]
## [1,] 0.6576043 -0.7533635
## [2,] 0.7533635  0.6576043
```

```
svd_A$u
```

```
##      [,1]      [,2]
## [1,] -0.6576043 -0.7533635
## [2,] -0.7533635  0.6576043
```

Right Vectors(V) = Eigen Vectors of  $A^TA$

```
svd_A$v
```

```
##      [,1]      [,2]
## [1,]  0.01856629 -0.6727903
## [2,] -0.25499937 -0.7184510
## [3,] -0.96676296  0.1765824
```

```
eigen_vectors_Y
```

```
##      [,1]      [,2]      [,3]
## [1,] -0.01856629 -0.6727903  0.7396003
## [2,]  0.25499937 -0.7184510 -0.6471502
```

```
## [3,] 0.96676296 0.1765824 0.1849001
```

Singular Values = sqrt of the Eigen Values for either  $A^T A$  or  $AA^T$ , with the extra values of  $A^T A$  being zero or close to zero

```
svd_A$d
```

```
## [1] 5.157693 2.097188
```

```
sqrt(eigen_values_X)
```

```
## [1] 5.157693 2.097188
```

```
sqrt(eigen_values_Y)
```

```
## [1] 5.157693e+00 2.097188e+00 1.029068e-08
```

## Problem Set 2

Using the procedure outlined in section 1 of the weekly handout, write a function to compute the inverse of a well-conditioned full-rank square matrix using co-factors. In order to compute the co-factors, you may use built-in commands to compute the determinant. Your function should have the following signature:

**B = myinverse(A)**

where **A** is a matrix and **B** is its inverse and  $\mathbf{A} \times \mathbf{B} = \mathbf{I}$ . The off-diagonal elements of **I** should be close to zero, if not zero. Likewise, the diagonal elements should be close to 1, if not 1. Small numerical precision errors are acceptable but the function *myinverse* should be correct and must use co-factors and determinant of **A** to compute the inverse.

Please submit PS1 and PS2 in an R-markdown document with your first initial and last name.

Figure 3:

```
myinverse <- function(A) {  
  C <- diag(nrow(A))  
  
  for (i in 1:nrow(A)) {  
    for (j in 1:ncol(A)) {  
      C[i,j] <- (det(A[(-1*i), (-1*j)])) * ((-1)^(i+j))  
    }  
  }  
  T <- diag(nrow(A))  
  
  for (i in 1:nrow(A)) {  
    row <- C[i,]  
    for (j in 1:length(row)) {  
      T[j,i] <- row[j]  
    }  
  }  
}
```

```
inv <- 1/(det(A) ) * T
inv
}
```

```
A <- matrix(c(1, -1, 1, 1,1,1,1,2,4) , ncol = 3 , byrow = TRUE )
```

```
B <- myinverse(A)
```

```
A %*% B
```

```
##           [,1] [,2] [,3]
## [1,]  1.000000e+00    0    0
## [2,] -2.775558e-17    1    0
## [3,] -1.110223e-16    0    1
```