

Inverse of a Matrix & Singular Value Decomposition

IS605

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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 3×2 matrix

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

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. 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 .

Your code should compute all these vectors and scalars and store them in variables. Please add enough comments in your code to show me how to interpret your steps. <https://www.r-bloggers.com/quick-review-of-matrix-algebra-in-r/>

```
A <- matrix(c(1,2,3,-1,0,4), nrow = 2, byrow = TRUE)
A
```

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

```
X <- A%*%t(A) #Compute X=AA^T
X
```

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

```
Y <- t(A)%*%A #Compute Y=A^TA
Y
```

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

```
# Compute eigenvalues, eigenvectors of X and Y
eignX <- eigen(X)
eignX
```

```
## $values
## [1] 26.601802  4.398198
##
## $vectors
##      [,1]      [,2]
## [1,] 0.6576043 -0.7533635
## [2,] 0.7533635  0.6576043
```

```

eignY <- eigen(Y)
eignY

## $values
## [1] 2.660180e+01 4.398198e+00 1.058982e-16
##
## $vectors
##          [,1]      [,2]      [,3]
## [1,] -0.01856629 -0.6727903  0.7396003
## [2,]  0.25499937 -0.7184510 -0.6471502
## [3,]  0.96676296  0.1765824  0.1849001

# Compute singular vectors
svdA <- svd(A)
(svdA$d)^2

## [1] 26.601802  4.398198

eignX$vectors # Compute left singular vector U to the eigen vectors of X

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

svdA$u

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

round(abs(eignX$vectors)) == round(abs(svdA$u))

##          [,1] [,2]
## [1,] TRUE TRUE
## [2,] TRUE TRUE

eignY$vectors #eigenvectors of Y with the right singular vectors

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

svdA$v

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

round(abs(eignY$vectors[, -3])) == round(abs(svdA$v))

##          [,1] [,2]
## [1,] TRUE TRUE
## [2,] TRUE TRUE
## [3,] TRUE TRUE

eignX$values #compare the eigenvalues and the squares of singular values

## [1] 26.601802  4.398198

```

```
eignY$values

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

svdA$d^2

## [1] 26.601802 4.398198

round(eignX$values) == round(svdA$d^2)

## [1] TRUE TRUE

round(eignY$values[-3]) == round(svdA$d^2)

## [1] TRUE TRUE
```

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: $B = \text{myinverse}(A)$ where A is a matrix and B is its inverse and $AB = I$. 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.

<https://rpubs.com/aaronsc32/inverse-matrices>

```
#Matrix A.
myinverse <- function(A) {

  if (det(A) == 0) { #if the determinant is 0 the matrix is not invertable
    stop('No inverse')
  }

  m <- nrow(A)
  n <- ncol(A)
  C <- diag(0, nrow = m, ncol = n)

  #compute the cofactors
  #- loop thru m and n of matrix A, and remove the current row/col, and calc the determinant
  for(i in 1:m)
  {
    for(j in 1:n)
    {
      C[i, j] <- (-1)^(i+j) * det(A[-i, -j])
    }
  }

  return((t(C) / det(A))) #Inverse of A = t(C)/det(A)
}

set.seed(40)
A <- matrix(sample(10:20, 9, replace=TRUE),byrow=T, nrow =3)

B <- myinverse(A)

all.equal(solve(A), B) # R inverse function

## [1] TRUE
```

```
I <- round(A %*% B, 10); I
```

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