# Assignment 4

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**PS1.** Storing and displaying A, X, and Y:

```
A <- matrix(c(-1,1,0,1,3,-2,5,0,1), nrow = 3);A

## [,1] [,2] [,3]
## [1,] -1 1 5
```

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

```
X \leftarrow A %*% t(A);X
```

```
## [,1] [,2] [,3]
## [1,] 27 2 3
## [2,] 2 10 -6
## [3,] 3 -6 5
```

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

```
## [,1] [,2] [,3]
## [1,] 2 2 -5
## [2,] 2 14 3
## [3,] -5 3 26
```

Getting and printing eigenvalues/vectors of X and Y:

```
x.eigen <- eigen(X);x.eigen</pre>
```

```
## $values
## [1] 27.4907376 14.0000000 0.5092624
##
## $vectors
## [,1] [,2] [,3]
## [1,] 0.99086430 0.0000000 0.1348627
## [2,] 0.07480836 -0.8320503 -0.5496326
## [3,] 0.11221254 0.5547002 -0.8244489
```

```
y.eigen <- eigen(Y);y.eigen</pre>
```

```
## $values
## [1] 27.4907376 14.0000000 0.5092624
##
```

```
## $vectors
## [,1] [,2] [,3]
## [1,] -0.1747144 0.2223748 0.9591790
## [2,] 0.1889822 0.9636241 -0.1889822
## [3,] 0.9663129 -0.1482499 0.2103839
```

Getting the singular, left singular, and right singular values of A using svd():

```
a.svd <- svd(A)
a.svd$d
```

```
## [1] 5.2431610 3.7416574 0.7136263
```

#### a.svd\$u

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

## [1,] -0.99086430 -7.771561e-16 0.1348627

## [2,] -0.07480836 8.320503e-01 -0.5496326

## [3,] -0.11221254 -5.547002e-01 -0.8244489
```

## a.svd\$v

```
## [,1] [,2] [,3]
## [1,] 0.1747144 0.2223748 -0.9591790
## [2,] -0.1889822 0.9636241 0.1889822
## [3,] -0.9663129 -0.1482499 -0.2103839
```

We can compare the two sets of singular vectors with the eigenvectors of X and Y. If the comparison works, they should be the same(can be negative).

For X:

# x.eigen\$vectors

```
## [,1] [,2] [,3]
## [1,] 0.99086430 0.0000000 0.1348627
## [2,] 0.07480836 -0.8320503 -0.5496326
## [3,] 0.11221254 0.5547002 -0.8244489
```

#### a.svd\$u

```
## [,1] [,2] [,3]
## [1,] -0.99086430 -7.771561e-16 0.1348627
## [2,] -0.07480836 8.320503e-01 -0.5496326
## [3,] -0.11221254 -5.547002e-01 -0.8244489
```

For Y:

# y.eigen\$vectors

```
## [,1] [,2] [,3]
## [1,] -0.1747144 0.2223748 0.9591790
## [2,] 0.1889822 0.9636241 -0.1889822
## [3,] 0.9663129 -0.1482499 0.2103839
```

#### a.svd\$v

```
## [,1] [,2] [,3]
## [1,] 0.1747144 0.2223748 -0.9591790
## [2,] -0.1889822 0.9636241 0.1889822
## [3,] -0.9663129 -0.1482499 -0.2103839
```

And finally for the singular values:

# x.eigen\$values

```
## [1] 27.4907376 14.0000000 0.5092624
```

### y.eigen\$values

```
## [1] 27.4907376 14.0000000 0.5092624
```

```
a.svd$d^2
```

```
## [1] 27.4907376 14.0000000 0.5092624
```

**PS2** First step is to write a function to find the inverse of a matrix.

```
myinverse <- function(A) {
    # Takes matrix A, and returns the inverse(B).

if (det(A) != 0) { #checking for 0 determinant
    B <- matrix(data = NA,nrow = dim(A)[1], ncol = dim(A)[2])
    for (i in 1:dim(A)[1]) {
        for (j in 1:dim(A)[2]) {
            B[i,j] <- -1^(i+j) * det(A[-i,-j])
        }
    }
    return(t(B) / det(A))
}</pre>
```

Let's test with the matrix from the first problem set.

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

```
## [,1] [,2] [,3]
## [1,] -0.8571429 -0.1428571 -0.7142857
## [2,] 0.4285714 0.5714286 -2.1428571
## [3,] -0.2857143 0.2857143 0.4285714
```

The test for matrix A seems a bit off, but the inverse function itself seems to work fine.