

## Part 5: Unique Solutions and Least Squares for an Overdetermined Matrix

```
In [ ]: import pandas as pd
import numpy as np
```

### A)

Displaying the singular values for matrix  $A$

```
In [ ]: A = np.array([[6, 5, 3],
                      [4, 1, 5],
                      [6, 3, 6],
                      [5, 3, 6],
                      [6, 6, 3]])

u, s, v = np.linalg.svd(A)

s
```

```
Out[ ]: array([17.92667464,  4.69961929,  0.74021274])
```

### B)

Finding the eigenvalues of  $A$ .

```
In [ ]: # a_square = np.matmul(A.T, A)
a_square = A.T @ A
a_square
```

```
Out[ ]: array([[149, 103, 122],
               [103,  80,  74],
               [122,  74, 115]])
```

```
In [ ]: eigenvalues, eigenvectors = np.linalg.eig(a_square)
eigenvalues
```

```
Out[ ]: array([321.36566366,  0.54791491, 22.08642144])
```

The singular value  $\sigma_i$  is related to the eigenvalue  $\lambda_i$  by:

$$\sigma_i = \sqrt{\lambda_i}$$

## C)

Because there are only 3 singular values in  $A$ ,  $rank(A) = 3$

```
In [ ]: np.linalg.matrix_rank(A)
```

```
Out[ ]: 3
```

## D)

Yes, even though the column space and row space do not have the same dimensionality,  $m > n$  and each column vector is linearly independent. This means that our input space is  $\mathbb{R}^3$  and maps onto  $\mathbb{R}^5$  - however they do not span all of  $\mathbb{R}^5$ . The three column vectors are linearly independent meaning  $Span(v_1, v_2, v_3) = 3$ , where  $v_i$  is a column vector in  $A$ . Because the  $Span(v_1, v_2, v_3) = dim(Col(A))$ , if there is a solution in the solution space, it must be unique.

## E)

Finding solution or approximation for



$$b = \begin{bmatrix} -1 \\ 2 \\ 0 \\ 3 \\ 1 \end{bmatrix}$$

```
In [ ]: b = np.array([-1, 2, 0, 3, 1])  
b
```

```
Out[ ]: array([-1,  2,  0,  3,  1])
```

```
In [ ]: A, b
```

```
Out[ ]: (array([[6, 5, 3],  
                [4, 1, 5],  
                [6, 3, 6],  
                [5, 3, 6],  
                [6, 6, 3]]),  
        array([-1,  2,  0,  3,  1]))
```

Demonstration that  $Ax = b$  does not have a solution

```
In [ ]: np.linalg.solve(A, b)
```

```

-----
LinAlgError                                Traceback (most recent call last)
Cell In[8], line 1
----> 1 np.linalg.solve(A, b)

File ~\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.12_qbz5n2kfra8p0\LocalCache\local-packages\Python312\
\site-packages\numpy\linalg\linalg.py:396, in solve(a, b)
    394 a, _ = _makearray(a)
    395 _assert_stacked_2d(a)
--> 396 _assert_stacked_square(a)
    397 b, wrap = _makearray(b)
    398 t, result_t = _commonType(a, b)

File ~\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.12_qbz5n2kfra8p0\LocalCache\local-packages\Python312\
\site-packages\numpy\linalg\linalg.py:213, in _assert_stacked_square(*arrays)
    211 m, n = a.shape[-2:]
    212 if m != n:
--> 213     raise LinAlgError('Last 2 dimensions of the array must be square')

LinAlgError: Last 2 dimensions of the array must be square

```

Finding the value of  $x$  that minimizes  $\|Ax - b\|$ .

```

In [ ]: x, _, _, s = np.linalg.lstsq(A, b, rcond=None)
        x

```

```

Out[ ]: array([-2.06119825,  1.37336076,  1.54641296])

```