

# Lab Instructions - session 3

## Row and Column Space, Linear Maps

### Column Space and Row Space

The following code creates a figure with two subplots. In the left subplot, we plot a bunch of random 3D points in the column space of matrix  $\mathbf{A}$ . The right subplot shows a set of 2D points in the row space of  $\mathbf{A}$ .

plot1.py

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D

# create a 3 x 2 matrix
A = np.array([[1, 2],
              [3, 4],
              [-2, 1]])

fig = plt.figure()

# A 1 by 2 subplot grid, subplot 1 (3D)
ax1 = fig.add_subplot(1,2,1, projection='3d')
ax1.set_title('column space')

for i in range(200):
    # create a random column vector
    u = np.random.randn(2,1)
    # create a point in the column space of A
    v = A @ u
    ax1.scatter(v[0,0], v[1,0], v[2,0], color='b')

# A 1 by 2 subplot grid, subplot 2 (2D)
ax2 = fig.add_subplot(1,2,2)
ax2.set_title('row space')

for i in range(200):
    # create a random row vector
    u = np.random.randn(1,3)
    # create a point in the row space of A
    v = u @ A
    ax2.plot(v[0,0], v[0,1], 'ro')

plt.show()
```

- Rotate the 3D plot. Do all the points lie in a lower-dimensional subspace?

- What is the dimension of the column space? What is the dimension of the row space?

## Task 1 - Practice vectorized coding

You have to write the above without using the `for` loops. To create an  $m$  by  $n$  (normally distributed) random matrix use `np.random.randn(m, n)`. Notice that for a 2 by  $n$  matrix  $A$  containing  $n$  points as its columns, you may plot the points by giving the list of the  $x$ - and  $y$ -coordinates as the first and second argument of the `plot` function respectively:

```
ax.plot(A[0,:], A[1,:], 'o')
```

Similarly, for a 3 by  $n$  matrix containing 3D points, you may use

```
ax.scatter(A[0,:], A[1,:], A[2,:])
```

Likewise, you may plot the points represented as rows of a matrix.

## Task 2

Repeat task 1 for the matrix

```
1, 2  
3, 6  
-2, -4
```

- What are the dimensions of the row and column spaces?

## Task 3

Create a 2 by 3 subplot using `fig.add_subplot(2,3,i, projection='3d')` for plotting the column and row spaces of the following 3 by 3 matrices:

```
A = 1, 2, 1,  
    2, -1, -1,  
    -1, 1, -2  
  
B = 1, 2, -3  
    3, 1, 1  
    2, 1, 0  
  
C = 1, 2, -3  
    3, 6, -9  
    -2, -4, 6
```

The row and column spaces must be plotted in the subplot's first and second rows, respectively. The columns of the subplot correspond to the matrices **A**, **B**, and **C**.

- Rotate the plots. For each matrix, what are the dimensions of the row and the column spaces?
- What can you say about the row and column spaces of a matrix?
- Plot (the points in) the row and column spaces of matrix **B** in the same axes using two different colours. Repeat the same for matrix **C**. Are the row and column spaces of matrices equal in general?

## Linear Transformations

Remember representing the shape of a face as a set of points from the previous lab. Here, we apply a linear transformation to each point.

`face1.py`

```
import matplotlib.pyplot as plt
import numpy as np
from face_data import Face1, edges

def plot_face(plt,X,edges,color='b'):
    "plots a face"
    plt.plot(X[:,0], X[:,1], 'o', color=color)

    for i,j in edges:
        xi,yi = X[i]
        xj,yj = X[j]

        plt.plot((xi,xj), (yi,yj), '-', color=color)

    plt.axis('square')
    plt.xlim(-100,100)
    plt.ylim(-100,100)

th = np.pi/6
A = np.array([[np.cos(th), np.sin(th)],
              [-np.sin(th), np.cos(th)]])

X = Face1 @ A
plot_face(plt, X, edges, color='b')
plt.show()
```

- Why does the above rotates the face counterclockwise, while the matrix **A** corresponds to a 30 degrees clockwise rotation ( $-30^\circ$ )?

## Task 4 - Linear Transformations

- A. Animate the face to rotate around the origin by varying  $\theta$  from 0 to  $2\pi$ . Use what you learned from the previous lab.
- B. Apply a scaling transformation:

$$A = \begin{bmatrix} \alpha & 0 \\ 0 & \alpha \end{bmatrix}$$

- Animate by varying  $\alpha$  from  $3/4$  to  $4/3$ .
- What happens when alpha is negative?

- C. Apply a non-uniform scaling transformation:

$$A = \begin{bmatrix} \alpha & 0 \\ 0 & \beta \end{bmatrix}$$

- Animate by varying  $\alpha$  from  $3/4$  to  $4/3$  and taking  $\beta = 1/\alpha$ .

- D. Shear the face (horizontally) by applying the transformation

$$A = \begin{bmatrix} 1 & 0 \\ s & 1 \end{bmatrix}$$

- Animate by varying  $s$  from  $-0.7$  to  $0.7$ .
- The matrix  $A$  above represents a vertical shear. Why does it perform a horizontal shear here?

## Measuring the execution time

To measure the execution time of an operation or a piece of code put it inside a function and pass it to `timeit.timeit`:

`measure_time.py`

```
import numpy as np
import timeit

n = 1000

A = np.random.rand(n,n)

def f():
    np.linalg.inv(A)
```

```
t1 = timeit.timeit(f, number=1)
t2 = timeit.timeit(f, number=100)/100

print(t1)
print(t2)
```

- The execution time of what operation is measured?
- Which measurement is more reliable?  $t_1$  or  $t_2$ ?

This can be done in a more compact way using the `lambda` functions:

```
timeit.timeit(lambda : np.linalg.inv(A), number=100)/100
```

## Diagonal matrices

Execute the following and see the result.

### diagonal.py

```
import numpy as np

D1 = np.diag([2,3,4])
D2 = np.diag([10,20,30,40])

print('D1=\n', D1)
print('D2=\n', D2)

A = np.array([[1,1,1,1],
              [1,2,2,2],
              [1,2,3,4]])

print('A=\n', A)
print('D1@A=\n', D1 @ A)

print('A@D2=\n', A @ D2)
```

- What is the effect of multiplying a diagonal matrix to the left and right?

## Scaling rows and columns using broadcasting

This is an alternative to scaling the rows of a matrix using the concept of Broadcasting you learned in Lab 1.

### scale\_rows.py

```
import numpy as np
```

```
d1 = np.array([2,3,4]).reshape((3,1))

A = np.array([[1,1,1,1],
              [1,2,2,2],
              [1,2,3,4]])

print('d1=\n', d1)
print('A=\n', A)

print('d1.shape=\n', d1.shape)
print('A.shape=\n', A.shape)

print('d1 * A=\n', d1 * A)
```

- Write an equivalent code to scale columns of a matrix with numbers [10,20,30,40]. Is reshaping `np.array([10,20,30,40])` to shape (1,4) necessary for scaling columns? Why? (refer to the Broadcasting rules)
- Measure the execution time of `d1*A` and `D1@A` using `timeit` Which one is faster? Why?

## Task 5

Compare the execution time of `d1*A` with `D1@A` for random matrices `d1` and `A` and `D1=diag(d1.ravel())`, where `A` is 100 by 200. Which one is faster? (Do not count the time of creating `D1` when computing `D1@A`.)

## Task 6- Practice vectorized code

Consider the following:

`task2.py`

```
import numpy as np

m,n,p = 100,50, 2000

A = np.random.rand(m,n,p)
s = np.random.rand(p)

for i in range(p):
    A[:, :, i] *= s[i]
```

- In the above, replace the `for` loop with a single command.
- Compare the execution time of your code with the for loop using `timeit`.