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**Course:** Linear Algebra

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**Lab Instructions - session 2**

**Introduction to Python**

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# Lab Instructions - session 2

## Linear Combination, Span, Basis

### How to drawing 3D vectors?

To draw 3D objects first add these three lines after importing matplotlib:

```
from mpl_toolkits.mplot3d import Axes3D

fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
```

A vector can be plotted either as a *point* or an *arrow*. To plot a set of 3D points you can use the **scatter** function.

#### plot1.py

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

fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')

# plot multiple points
u = np.array([1, 2, 3])
v = np.array([2, 0, -2])
w = np.array([-1, -1, -1])

xs = [u[0], v[0], w[0]]
ys = [u[1], v[1], w[1]]
zs = [u[2], v[2], w[2]]

ax.scatter(xs, ys, zs)
plt.show()
```

To plot an arrow you may use the **quiver** function:



### Plot2.py

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

fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')

# plot multiple points
u = np.array([1,2,3])
v = np.array([2, 0, -2])
w = np.array([-1, -1, -1])

xs = [u[0], v[0], w[0]]
ys = [u[1], v[1], w[1]]
zs = [u[2], v[2], w[2]]

# base of the vectors set to the origin
tail_x = [0,0,0]
tail_y = [0,0,0]
tail_z = [0,0,0]

ax.set_xlim(-3,3)
ax.set_ylim(-3,3)
ax.set_zlim(-3,3)

ax.quiver(tail_x, tail_y, tail_z, xs, ys, zs, color='r')
plt.show()
```

- Rotate the plot to view it from different angles. Do you think  $\mathbf{u}$ ,  $\mathbf{v}$  and  $\mathbf{w}$  are linearly dependent? If yes, how can you write one of them as a linear combination of the others?

## Linear combination/span

The following code generates 2 random scalars  $\mathbf{a}$  and  $\mathbf{b}$  using the `numpy.random.rand` function and plots the linear combination  $\mathbf{w} = \mathbf{a} \mathbf{u} + \mathbf{b} \mathbf{v}$  of the vectors  $\mathbf{u}$  and  $\mathbf{v}$ .

### Plot3.py

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



```
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')

# plot multiple points
u = np.array([1,2,3])
v = np.array([2, 0, -2])

xs = [u[0], v[0]]
ys = [u[1], v[1]]
zs = [u[2], v[2]]

# base of the vectors set to the origin
tail_x = [0,0]
tail_y = [0,0]
tail_z = [0,0]

ax.set_xlim(-3,3)
ax.set_ylim(-3,3)
ax.set_zlim(-3,3)

ax.quiver(tail_x, tail_y, tail_z, xs, ys, zs, color='r')

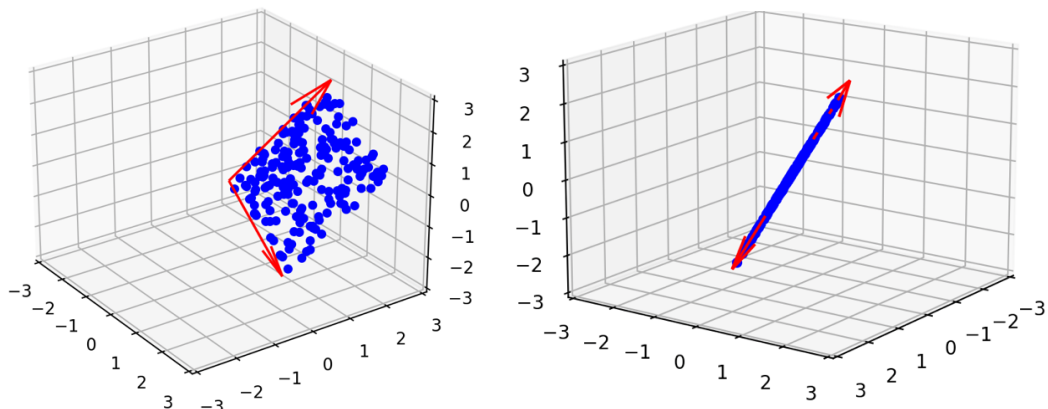
a,b = np.random.rand(2)
w = a * u + b * v
ax.scatter(w[0], w[1], w[2], color='b')

plt.show()
```

- Change the code to repeat plotting the linear combination  $\mathbf{w}$  200 times. This can be done by putting the following three lines in a loop:

```
a,b = np.random.rand(2)
w = a * u + b * v
ax.scatter(w[0], w[1], w[2], color='b')
```

Rotate the plot to view it from different angles and then you may see something like this:





- Notice that the plotted points are in **span(u,v)**. Rotate the plot to see this. Why is the shape of the scatter like that? Notice that the function **numpy.random.rand** generates random samples in the interval **[0,1)**.
- Replace the function **numpy.random.rand** with **numpy.random.randn**. What happens? and why?

## Animating a plot

Run the following piece of code. What does it do?

### Plot4.py

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

fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')

u = np.array([1, 2, 3])
v = np.array([2.0, 0, -2])

rng = np.linspace(0, 1, 20)

for alpha in rng:
    w = (1-alpha) * u + alpha * v

    ax.set_xlim(-4, 4)
    ax.set_ylim(-4, 4)
    ax.set_zlim(-4, 4)

    ax.quiver(0,0,0, u[0], u[1], u[2], color='r')
    ax.quiver(0,0,0, v[0], v[1], v[2], color='r')

    ax.quiver(0,0,0, w[0], w[1], w[2], color='b')
    ax.scatter(w[0], w[1], w[2], color='b')

plt.show()
```

- Rotate the plot to observe it from different angles.



- Add the following lines at the end of the body of the for loop. What happens?  
`plt.draw()`  
`plt.pause(.1)`  
 \* (if using Jupyter notebook, uncomment `%matplotlib` in the above to see the output correctly.)
- A linear combination  $\mathbf{w} = \mathbf{a} \mathbf{u} + \mathbf{b} \mathbf{v}$  of two vectors  $\mathbf{u}$  and  $\mathbf{v}$  is an **affine** combination if  $\mathbf{a} + \mathbf{b} = 1$ . It is also called a **convex** combination if  $\mathbf{a}, \mathbf{b} \geq 0$  in addition to  $\mathbf{a} + \mathbf{b} = 1$ . Are the vectors  $\mathbf{w}$  created here are affine combinations of  $\mathbf{u}$  and  $\mathbf{v}$ ? Are they also convex combinations?
- Change `np.linspace(0,1,20)` to `np.linspace(-0.5,1.5,20)`.
  - How does the plot change and why?
  - Are all the vectors  $\mathbf{w}$  still affine combinations of  $\mathbf{u}$  and  $\mathbf{v}$ ?
  - What about convex combinations?
- Add `ax.cla()` at the beginning of the **for** loop (cla stands for clear axis). What happens?

## Shape models

A "shape" can be represented as an ordered or unordered set of points. Here, we represent a shape consisting of  $n$  points by an  $n$  by 2 matrix, each row of which represents a point. The following code creates a pair of 2D shapes and plots them:

### shape1.py

```
import numpy as np
import matplotlib.pyplot as plt

n = 11

S1 = np.vstack((-np.cos(np.linspace(0, np.pi, n)),
                -.7*np.sin(np.linspace(0, np.pi, n)))).T

S2 = np.vstack((np.linspace(-1.2, 1.2, n),
                np.zeros(n))).T

print(S1.shape)
print(S2.shape)

plt.plot(S1[:, 0], S1[:, 1], 'bo-')
plt.plot(S2[:, 0], S2[:, 1], 'ro-')
```



```
plt.axis('equal')
plt.xlim(-2, 2)
plt.ylim(-2, 2)

plt.show()
```

- What are the shapes (dimensions) of  $S1$  and  $S2$ ?

Shapes, as defined above, form a vector space (can be scaled and added together).

To look at matrices as vectors, you can **vectorize** them. That is to flatten an  $n \times 2$  shape matrix to form a vector of size  $2n$ . Then perform addition, scaling, or linear combination:

```
s1 = S1.ravel()
s2 = S2.ravel()

s3 = a * s1 + b * s2
S3 = s3.reshape((n,2))
```

But, since matrices are added element-wise, you may simply write:

```
S3 = a * S1 + b * S2
```

- Plot the average shape  $S3 = 0.5 * S1 + 0.5 * S2$ .

## Task 1 - Shape Morphing

Use what you learned in section "Animating a plot" (`plt.draw`, `plt.pause`, `ax.cla`) to animate the shape  $S3$  in the form of  $S3 = (1-\alpha) * S1 + \alpha * S2$ , by letting  $\alpha$  range from 0 to 1 (convex combination). Use `plt.cla()` instead of `ax.cla()`.

- This is called **shape morphing**.
- Vary  **$\alpha$**  from 0 to 1.5 (affine combinations). What happens?
- Try other ranges (e. g. -2 to 2). What's the output?
- Each shape has  $2n$  (here 22) entries. But all the shapes you see are in `span(S1,S2)`, that is, they lie in a 2-dimensional subspace of a 22-dimensional vector space.
- (Point correspondence matter) Change `-np.cos(np.linspace(0,np.pi,n))` to `np.cos(np.linspace(0,np.pi,n))` when defining  $S1$ . What happens? Why?



## Task 2 - Face Model

A face can be represented as a shape model consisting of a set of landmark points. The code below imports three faces Face1, Face2, and Face3 and plots Face1. Plotting a face is done using the function **plot\_face** defined below. The file **face\_data.py** has been provided to you.

### Task2a.py

```
import matplotlib.pyplot as plt
import numpy as np

from face_data import Face1, Face2, Face3, edges

def plot_face(plt,X,edges,color='b'):
    "plots a face"

    plt.plot(X[:,0], X[:,1], 'o', color=color)

    i,j = edges[0] # edge from node i to node j
    xi = X[i,0]
    yi = X[i,1]

    xj = X[j,0]
    yj = X[j,1]

    # draw a line between X[i] and X[j]
    plt.plot((xi,xj), (yi,yj), '-', color=color)

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

plot_face(plt, Face1, edges, color='b')

plt.show()
```

- The list **edge** contains a list of edges, each in the form of **(i,j)**. Print it to see how it looks.
- The function **plot\_face** is supposed to plot the landmark points of the face, plus the edges between them. Currently, it only draws the first edge **edge[0]**. Change it to plot all the edges.





- Using the animation technique you learned above, morph a face shape from **Face1** to **Face2**, from **Face2** to **Face3**, and then from **Face3** back to **Face1**.
- Like before, try varying **alpha** from -0.5 to 1.5 instead of 0 to 1.0 and see what happens.

For  $n$  vectors  $v_1, v_2, \dots, v_n$ , a linear combination  $a_1 v_1 + a_2 v_2 + \dots + a_n v_n$  is called an affine combination if  $a_1 + a_2 + \dots + a_n = 1$ . It is also a convex combination if all the scalars  $a_i$  are nonnegative. Here, we want to find linear combinations of **Face1**, **Face2**, and **Face3** to create **TargetFace1** and **TargetFace2**.

### Task2b.py

```
import matplotlib.pyplot as plt
import numpy as np

from face_data import Face1, Face2, Face3, TargetFace1, TargetFace2,
edges

def plot_face(plt,X,edges,color='b'):
    "plots a face"

    plt.plot(X[:,0], X[:,1], 'o', color=color)

    i,j = edges[0] # edge from node i to node j
    xi = X[i,0]
    yi = X[i,1]

    xj = X[j,0]
    yj = X[j,1]

    # draw a line between X[i] and X[j]
    plt.plot((xi,xj), (yi,yj), '-', color=color)

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

# make a guess
a = 1/3.
b = 1/3.
c = 1/3.

F = a * Face1 + b * Face2 + c * Face3
```



```
plot_face(plt, TargetFace1, edges, color='r')
plot_face(plt, F, edges, color='g')

# change a,b,c until the two plots align

plt.show()
```

- Find a convex combination of Face1, Face2, and Face3 to create TargetFace1. Keep tuning the scalars a, b, and c in the code until the blue and green plots align.
- Find a linear (not necessarily convex) combination to create TargetFace2. Assume a, b, and c are positive. Try to guess them yourself before reading the hint below:
  - $a = 5.4 / 18 = ?$ .
- (Optional) Can you think of a way to find the scalars without trial and error?

## Task 3 - Practice Vectorization

Consider an arbitrary matrix A and a vector u like the following

```
m, n = 20, 10
A = np.random.rand(m, n)
u = np.random.rand(n)
```

We perform the following operation on A and u to create the vector v.

```
v = np.zeros(m)
for i in range(n):
    v += A[:, i] * u[i]
```

- Write an equivalent program without loops in just a single line of code.

```
v = ...
```