### MATGEO: 1.1.6.19

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- Problem
- Solution
  - Information Table
  - Method of Solving
  - ullet Finding  $\lambda$  values
  - Verifying Obtained Values
  - λ Verification Code
  - ullet Methods of Finding  $\lambda$

### Problem Statement

The vectors  $\lambda \hat{i} + \hat{j} + 2\hat{k}$ ,  $\hat{i} + \lambda \hat{j} - \hat{k}$  and  $2\hat{i} - \hat{j} + \lambda \hat{k}$  are coplanar if  $\lambda = ?$ 

## Information Table

Variable	Description	Formula
А	A vector in terms of $\lambda$	$A = egin{pmatrix} \lambda \ 1 \ 2 \end{pmatrix}$
В	A vector in terms of $\lambda$	$B = \begin{pmatrix} 1 \\ \lambda \\ -1 \end{pmatrix}$
С	A vector in terms of $\lambda$	$C = \begin{pmatrix} 2 \\ -1 \\ \lambda \end{pmatrix}$
М	It is a matrix comprising of vectors $A, B, C$	M = [A, B, C]
λ	It is the variable with which $A, B, C$ are established	M  = 0

### Method of Solving

The rank of a matrix M is less than 3, then the matrix is coplanar.

$$Rank(M) = 2 \text{ or } 1 \tag{3.1}$$

Equivalently,

$$|M| = 0 \tag{3.2}$$

# Finding $\lambda$ values

$$|M| = \begin{pmatrix} \lambda & 1 & 2 \\ 1 & \lambda & -1 \\ 2 & -1 & \lambda \end{pmatrix} = 0 \tag{3.3}$$

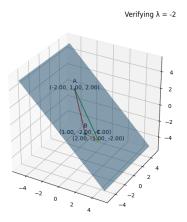
$$\lambda (\lambda^2 - 1) - 1(\lambda + 2) + 2(-1 - 2\lambda) = 0$$
 (3.4)

$$\lambda^3 - 6\lambda - 4 = 0 \tag{3.5}$$

$$(\lambda + 2) \left(\lambda^2 - 2\lambda - 2\right) = 0 \tag{3.6}$$

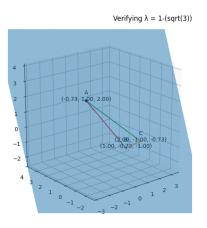
$$\lambda = -2 \text{ or } \lambda = 1 \pm \sqrt{3} \tag{3.7}$$

## Verifying Obtained Values



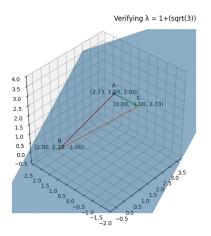
Figure

# Verifying Obtained Values



**Figure** 

# Verifying Obtained Values



Figure

```
import sys #for path to external scripts
sys.path.insert(0, '/home/ysiddhanth/Documents/matgeo/
   codes/CoordGeo') #path to my scripts
import numpy as np
import numpy.linalg as LA
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
from mpl_toolkits.mplot3d import Axes3D
#local imports
from line.funcs import *
from triangle.funcs import *
from conics.funcs import circ_gen
roots = np.loadtxt("lambda.dat")
\#roots are roots[0] = -2; roots[1] and roots[2]
k = roots[2]
A = np.array(([k, 1,2])).reshape(-1,1)
B = np.array(([1,k, -1])).reshape(-1,1)
C = np.array(([2,-1, k])).reshape(-1,1)
```

```
P1 = np.array(([k, 1,2]))
P2 = np.array(([1,k, -1]))
P3 = np.array(([2,-1, k]))
v1 = P2 - P1
v2 = P3 - P1
# Compute the normal vector (cross product of v1 and v2)
normal = np.cross(v1, v2)
# Plane equation: ax + by + cz = d
# where [a, b, c] is the normal vector and d is
   calculated using one of the points
a, b, c = normal
d = np.dot(normal, P1)
# Create a figure and a 3D Axes
fig = plt.figure(figsize=(8, 6))
ax = fig.add_subplot(111, projection='3d')
# Generate grid points for x and y
x = np.linspace(-5, 5, 50)
y = np.linspace(-5, 5, 50)
X, Y = np.meshgrid(x, y)
```

```
# Calculate corresponding z values for each (x, y) pair
    to satisfy the plane equation
Z = (-a*X - b*Y - d) / c
# Plot the plane
ax.plot_surface(X, Y, Z, alpha=0.5)
#Generating all lines
x_BC = line_gen(B,C)
x_AB = line_gen(A,B)
x_AC = line_gen(A,C)
#Plotting all lines
ax.plot(x_BC[0,:],x_BC[1,:], x_BC[2,:],label='^{9}BC')
ax.plot(x_AC[0,:],x_AC[1,:], x_AC[2,:],label={}^{,}AC{}^{,})
ax.plot(x_AB[0,:],x_AB[1,:], x_AB[2,:],label='AB')
# Scatter plot
colors = np.arange(2, 5) # Example colors
tri_coords = np.block([A, B, C]) # VStack A, B, C
ax.scatter(tri_coords[0, :], tri_coords[1, :], tri_coords
    [2, :], c=colors)
vert_labels = ['A', 'B', 'C']
```

```
for i, txt in enumerate(vert_labels):
# Annotate each point with its label and coordinates
ax.text(tri_coords[0, i], tri_coords[1, i], tri_coords[2,
    i], f'{txt}\n({tri_coords[0, i]:.02f}, {tri_coords[1,
    i]:.02f}, {tri_coords[2, i]:.02f})',
fontsize=10, ha='center', va='baseline')
ax.spines['top'].set_color('none')
ax.spines['left'].set_position('zero')
ax.spines['right'].set_color('none')
ax.spines['bottom'].set_position('zero')
# Set limits and aspect ratio to magnify the plane
ax.set_xlim(-4, 4) # Adjust limits based on your data
ax.set_ylim(-4, 4) # Adjust limits based on your data
ax.set_zlim(-4, 4) # Adjust limits based on your data
ax.set_box_aspect([1,1,1]) # Equal aspect ratio for x, y,
    and z axes
plt.grid() # minor
plt.axis('equal')
plt.show()
```

### Methods of Finding $\lambda$

**Method 1(using sympy) :** Use sympy to find the det and solve for lambda.

```
import sympy as sp \lambda = \text{sp.symbols}(`\lambda`) A = \text{sp.Matrix}([[\lambda, 1, 2], [1, \lambda, -1], [2, -1, \lambda]]) \det_A = A.\det() lambda\_solutions = \text{sp.solve}(\det_A, \lambda) numeric\_solutions = [\text{sp.N}(\text{sol}) \text{ for sol in} lambda\_solutions]
```

## Methods of Finding $\lambda$

### Method 2(using the QR algorithm(numpy or C)):

$$|M| = \begin{pmatrix} \lambda & 1 & 2 \\ 1 & \lambda & -1 \\ 2 & -1 & \lambda \end{pmatrix} = \begin{pmatrix} 0 & 1 & 2 \\ 1 & 0 & -1 \\ 2 & -1 & 0 \end{pmatrix} + \lambda I = |A + \lambda I|$$
(3.8)

Thus, the values of  $\lambda$  are the negative of the eigen values of the matrix A.

```
import numpy as np  A = \text{np.array}([[0, 1, 2], [1, 0. -1], [2, -1, 0]])  eigenvalues = np.linalg.eigvals(A)  \lambda - \text{vals} = -\text{eigenvals}
```