

different types of matrices

Np.diag((1,2))-->diag matrix np.identity((2/3)) ..>2,3 indicates order of matrix 2*2 A=A^T -->sym transpose.np.transpose(a) Np.tril(A) >>lower tri matrix np.triu(A) >>upper tri matrix

```
In [20]: #write a program to generate identity,diagonal, symmetric and others
import numpy as np
a =np.diag((1,2,3))
a
```

```
Out[20]: array([[1, 0, 0],
                 [0, 2, 0],
                 [0, 0, 3]])
```

```
In [21]: b= np.identity((3))
b
```

```
Out[21]: array([[1., 0., 0.],
                 [0., 1., 0.],
                 [0., 0., 1.]])
```

```
In [23]: c=np.array([[1,2],[2,1]])
d =c.transpose()
if c.all()==d.all(): #checking each and every elements and elements of both matrices are equal
    print('true')
else:
    print('false')
```

```
true
```

```
In [24]: np.tril(a) #applying for sq matrices only
```

```
Out[24]: array([[1, 0, 0],
                 [0, 2, 0],
                 [0, 0, 3]])
```

```
In [25]: a=np.array([[1,2,3],[2,3,4],[4,6,7]])
```

```
Out[25]: array([[1, 0, 0],  
                 [2, 3, 0],  
                 [4, 6, 7]])
```

```
In [26]: np.tril(a)
```

```
Out[26]: array([[1, 0, 0],  
                 [2, 3, 0],  
                 [4, 6, 7]])
```

```
In [27]: np.triu(a)
```

```
Out[27]: array([[1, 2, 3],  
                 [0, 3, 4],  
                 [0, 0, 7]])
```

Determinant

```
np.linalg.det(a)
```

```
In [29]: a=np.array([[1,1,2],[2,1,2],[3,1,1]])  
np.linalg.det(a)
```

```
Out[29]: 1.0000000000000002
```

```
In [30]: #inverse of a matrix  
  
a=np.array([[1,1],[2,3]])  
np.linalg.inv(a)
```

```
Out[30]: array([[ 3., -1.],  
                 [-2.,  1.]])
```

```
In [33]: a=np.array([[1,2,3],[2,3,4],[4,6,7]])  
a  
np.linalg.inv(a)
```

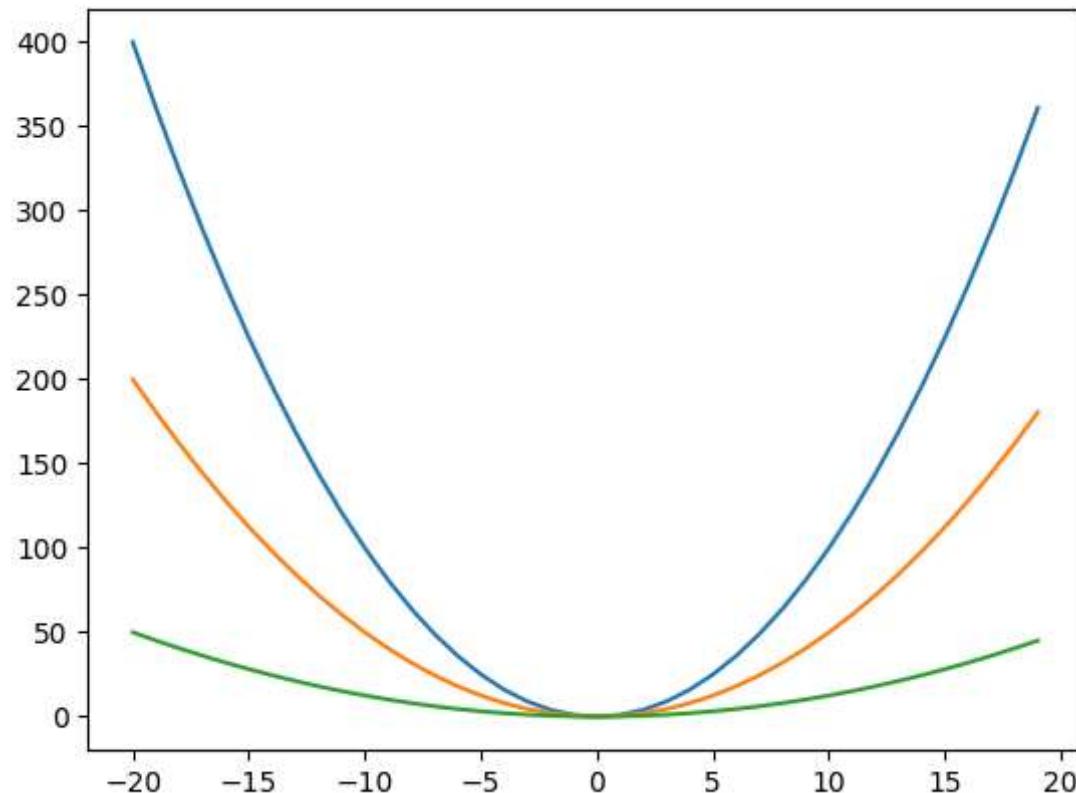
```
Out[33]: array([[-3.,  4., -1.],  
                  [ 2., -5.,  2.],  
                  [ 0.,  2., -1.]])
```

Linear equations

```
In [36]: import matplotlib.pyplot as mp
x=np.arange(-20,20)
y=x**2
print(y)
mp.plot(x,y) #y=x^2
mp.plot(x,y/2) #y=x^2/2
mp.plot(x,y/8)
```

```
[400 361 324 289 256 225 196 169 144 121 100 81 64 49 36 25 16 9
 4   1   0   1   4   9   16  25  36  49  64  81 100 121 144 169 196 225
256 289 324 361]
```

```
Out[36]: [<matplotlib.lines.Line2D at 0x1b762f83850>]
```

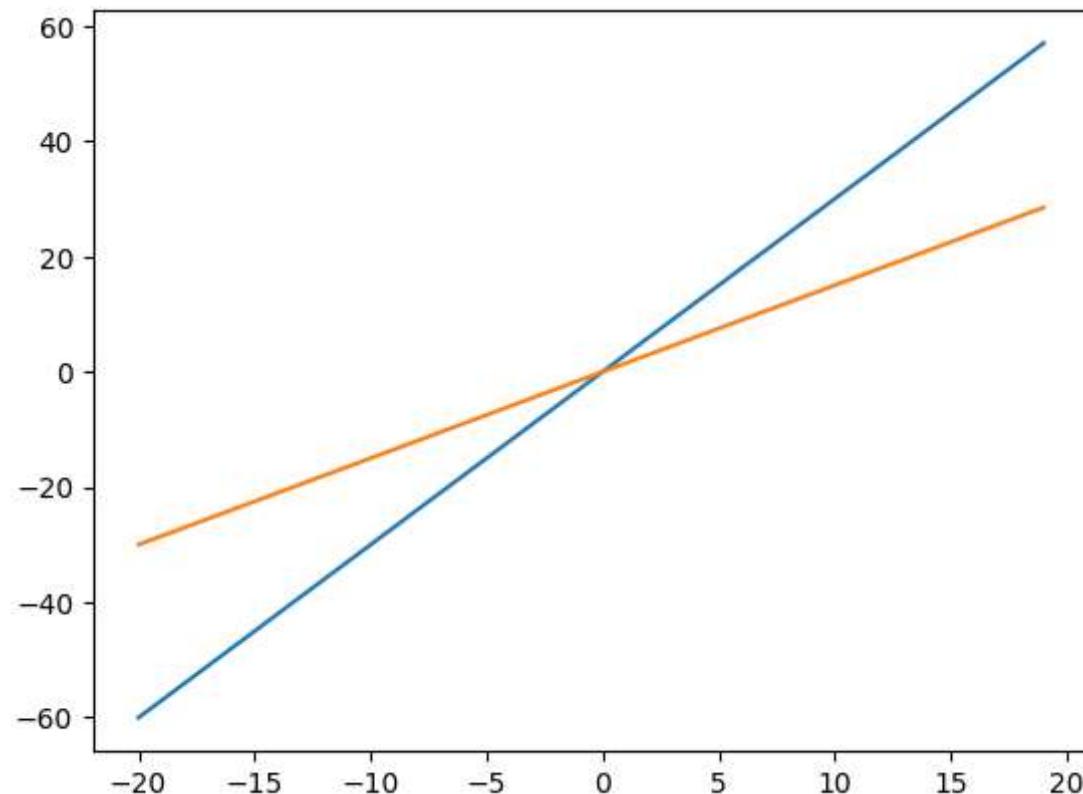


```
In [40]:
```

```
x=np.arange(-20,20)
y= 3*x
print(y)
mp.plot(x,y)
mp.plot(x,y/2)
```

```
[ -60 -57 -54 -51 -48 -45 -42 -39 -36 -33 -30 -27 -24 -21 -18 -15 -12 -9
 -6 -3  0   3   6   9   12  15  18  21  24  27  30  33  36  39  42  45
 48  51  54  57]
```

```
Out[40]: [<matplotlib.lines.Line2D at 0x1b76308be20>]
```

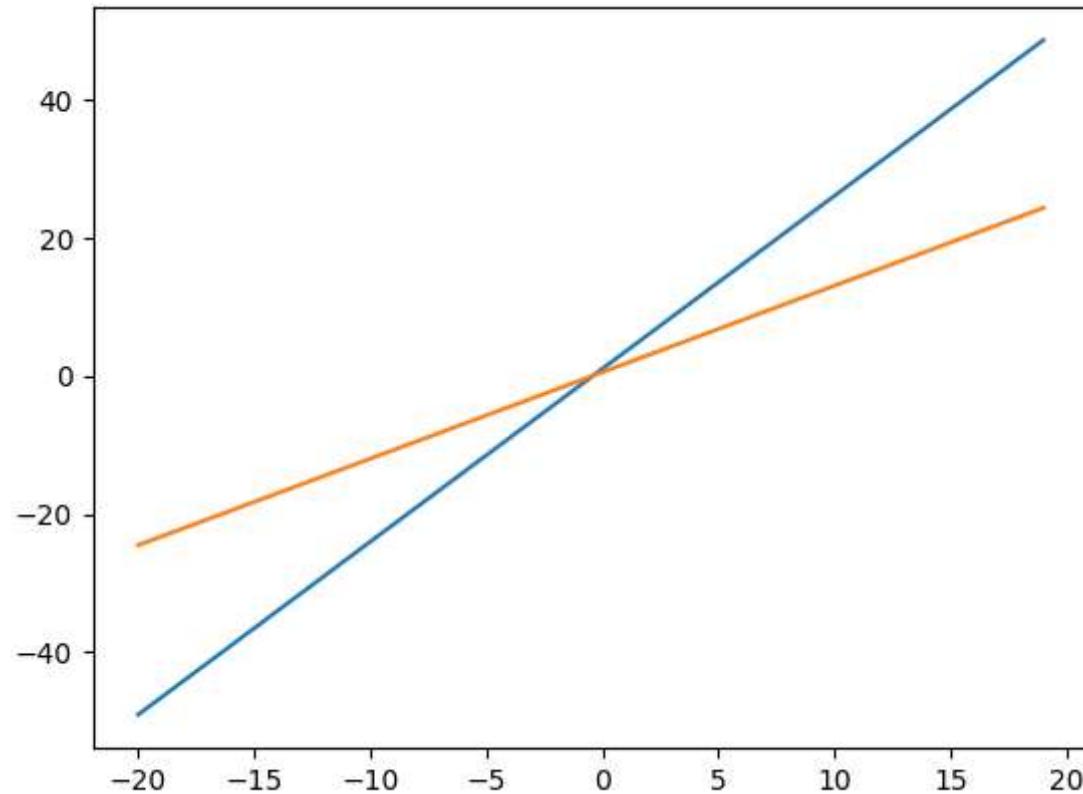


```
In [47]:
```

```
x=np.arange(-20,20)
y2=(2+(5*x))/2
print(y2)
mp.plot(x,y2)
mp.plot(x,y2/2)
```

```
[ -49.   -46.5  -44.   -41.5  -39.   -36.5  -34.   -31.5  -29.   -26.5  -24.   -21.5  
 -19.   -16.5  -14.   -11.5   -9.   -6.5   -4.   -1.5    1.    3.5    6.    8.5  
 11.    13.5   16.   18.5   21.   23.5   26.   28.5   31.   33.5   36.   38.5  
 41.    43.5   46.   48.5]
```

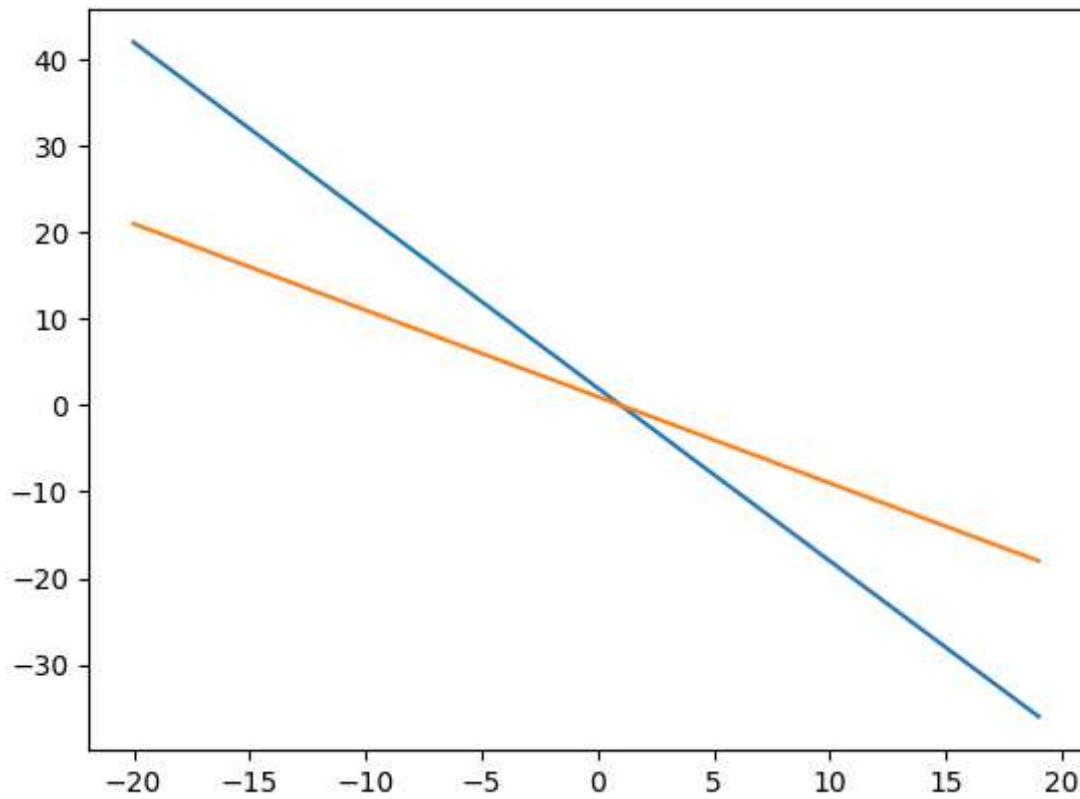
Out[47]: [`<matplotlib.lines.Line2D at 0x1b764321030>`]



```
In [48]: b=np.arange(-20,20)  
c= (4-4*b)/2  
print(c)  
mp.plot(b,c)  
mp.plot(b,c/2)
```

```
[ 42.   40.   38.   36.   34.   32.   30.   28.   26.   24.   22.   20.   18.   16.  
 14.   12.   10.    8.    6.    4.    2.    0.   -2.   -4.   -6.   -8.  -10.  -12.  
 -14.  -16.  -18.  -20.  -22.  -24.  -26.  -28.  -30.  -32.  -34.  -36.]
```

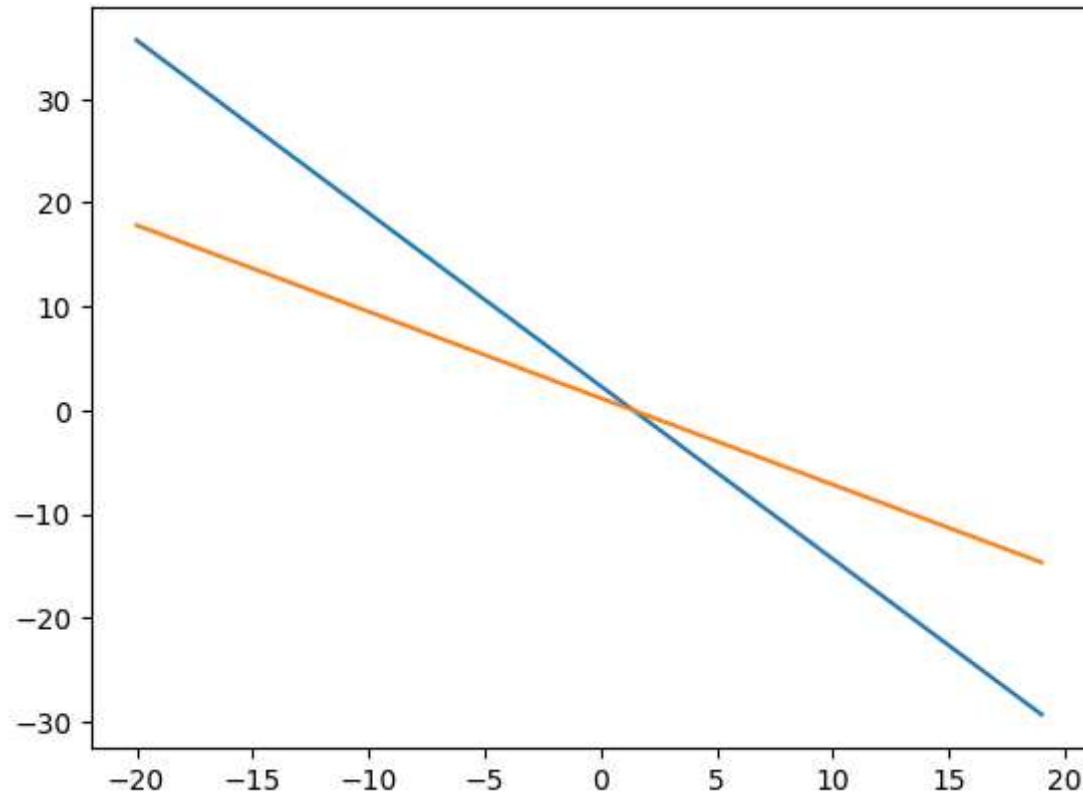
Out[48]: [`<matplotlib.lines.Line2D at 0x1b76438f310>`]



```
In [50]: b1=np.arange(-20,20)
c1= (7-5*b1)/3
print(c1)
mp.plot(b1,c1)
mp.plot(b1,c1/2)
```

```
[ 35.66666667  34.          32.33333333  30.66666667  29.
 27.33333333  25.66666667  24.          22.33333333  20.66666667
 19.          17.33333333  15.66666667  14.          12.33333333
 10.66666667   9.          7.33333333  5.66666667   4.
  2.33333333   0.66666667  -1.          -2.66666667  -4.33333333
  -6.          -7.66666667  -9.33333333  -11.         -12.66666667
 -14.33333333 -16.          -17.66666667  -19.33333333  -21.
 -22.66666667 -24.33333333  -26.          -27.66666667  -29.33333333]
```

```
Out[50]: [<matplotlib.lines.Line2D at 0x1b7644a73a0>]
```



```
In [2]: import numpy as np
```

```
In [4]: a=np.array([[2,1],[3,4]])
x,y = np.linalg.eig(a)
```

```
In [5]: x
```

```
Out[5]: array([1., 5.])
```

```
In [6]: y
```

```
Out[6]: array([[-0.70710678, -0.31622777],
               [ 0.70710678, -0.9486833 ]])
```

Linear transformation

linear transformation >> input vector transformed to output vector

input and output should start at the same point

enlarge/diminishes/rotates the vector on applying transformation on vector>> scaling

rotation can't be applied on matrices due to eigen values..> only scaling is done

```
In [24]: import numpy as np
```

```
In [25]: x=np.array([1,2])
s=3 #scalar value
b=x*s
b
```

```
Out[25]: array([3, 6])
```

Eigen values and eigen vectors

```
In [27]: a=np.array([[2,1],
                 [3,4]])
x,y=np.linalg.eig(a)
```

```
In [28]: x
#displays Eigen Values
```

```
Out[28]: array([1., 5.])
```

```
In [29]: y  
#Displays Eigen Vectors
```

```
Out[29]: array([[-0.70710678, -0.31622777],  
               [ 0.70710678, -0.9486833 ]])
```

Linear Transformation

```
In [30]: #Consider a vector x  
x=np.array([1,2])  
s=3 #here s is a scalar  
b=x*s  
b
```

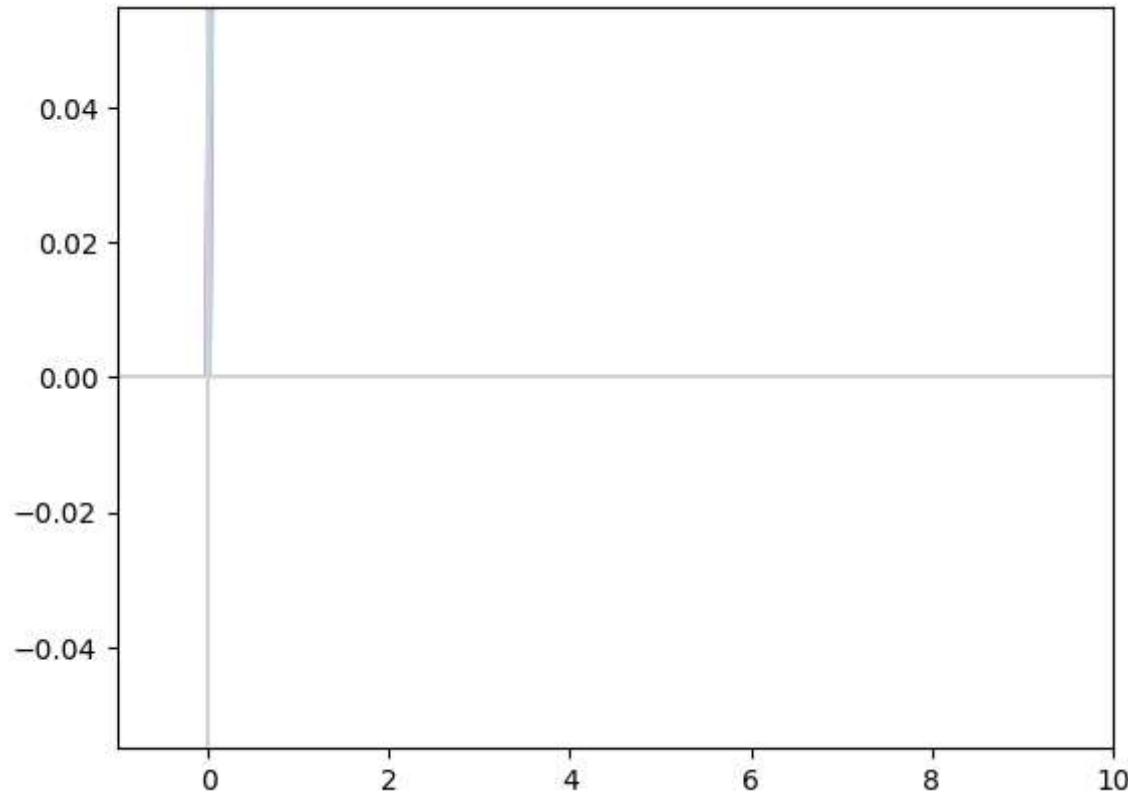
```
Out[30]: array([3, 6])
```

```
In [31]: import matplotlib.pyplot as plt
```

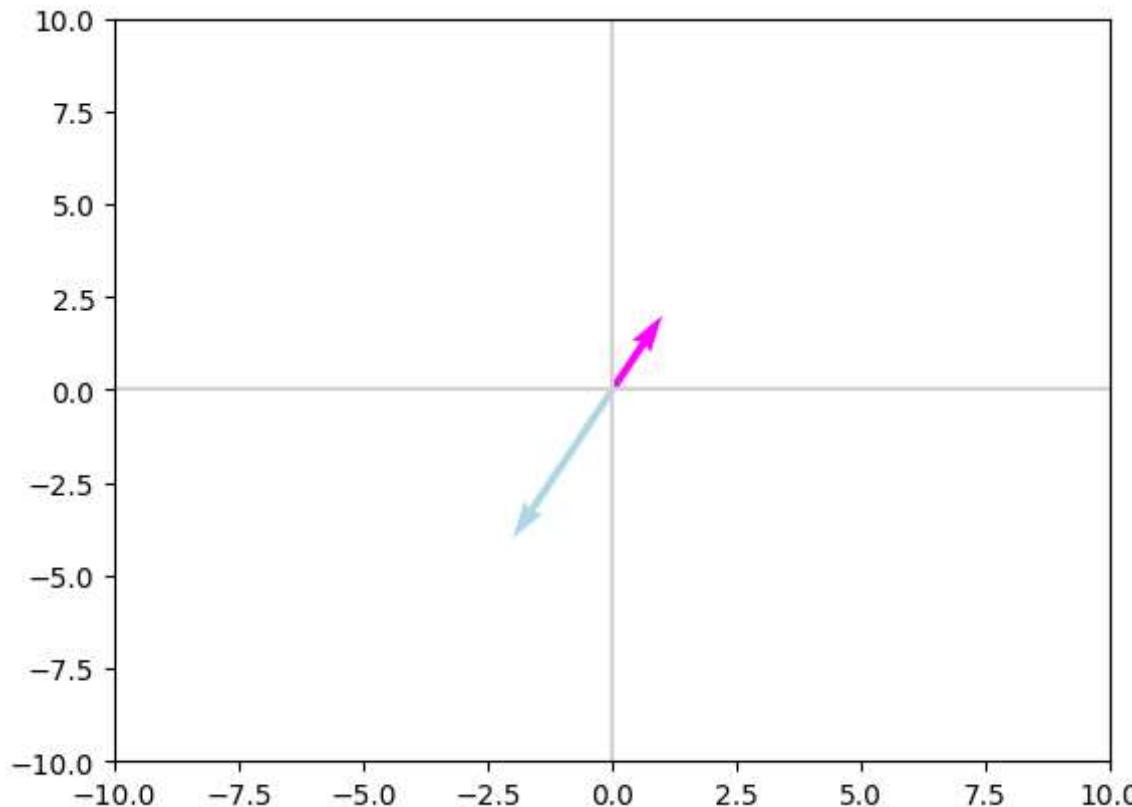
```
In [32]: def plot_vectors(vectors,colors):  
    #Plot one or more vectors in a 2-D plane, specifying  
    plt.axvline(x=0,color='lightgray')  
    plt.axhline(y=0,color='lightgray')  
    for i in range(len(vectors)):  
        x=np.concatenate(([0,0],vectors[i]))  
        plt.quiver([x[0]],[x[1]],[x[2]],[x[3]],  
                  angles='xy', scale_units='xy', scale=1, color=colors[i],)  
    #here [x[0]] and [x[1]] indicate the starting point of a vector,  
    # [x[2]] and [x[3]] indicates the ending point of a vector
```

```
In [33]: plot_vectors([x,b],[ 'magenta','lightblue'])  
plt.xlim(-1,10)  
#_=plt.ylim(-1,10) #here x is the input vector and b is the output vector
```

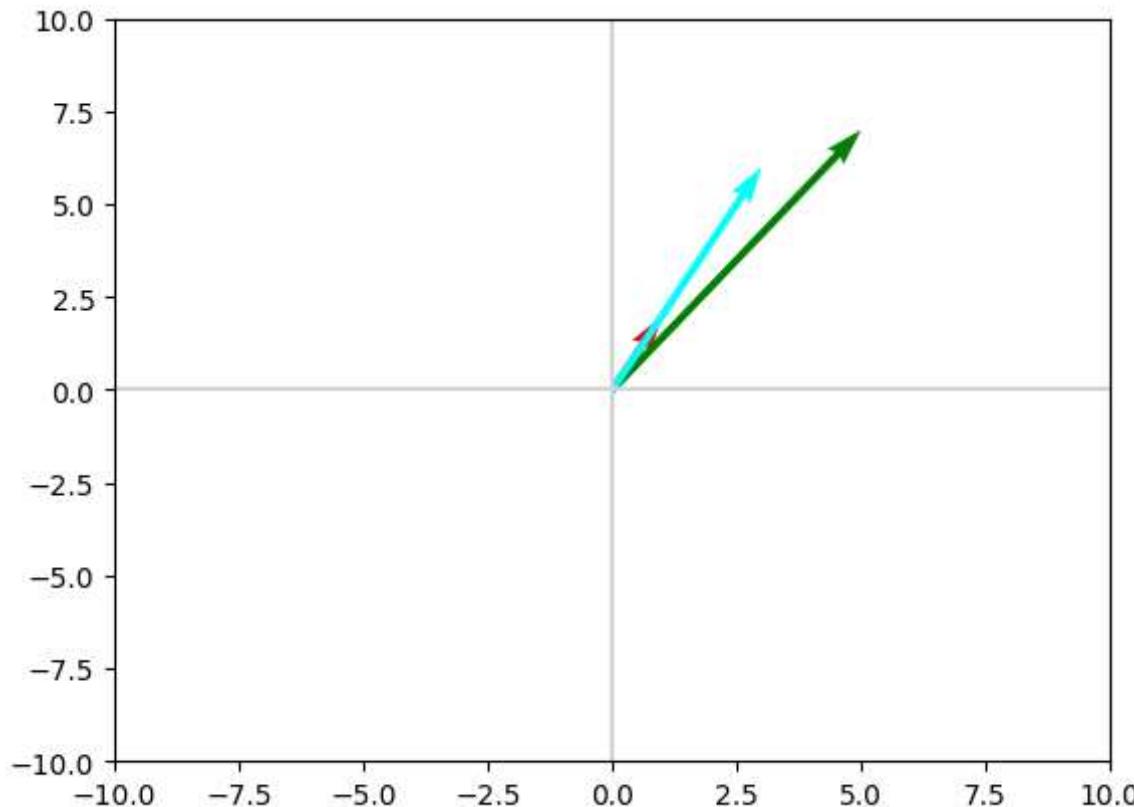
```
Out[33]: (-1.0, 10.0)
```



```
In [34]: # Directly printing without another scalar
s=-2
plot_vectors([x,x*s],['magenta','lightblue'])
plt.xlim(-10,10)
_=plt.ylim(-10,10)
```



```
In [35]: #initialising another input vector as y
#x and y are the input vectors with b as the output vector
y=np.array([5,7])
plot_vectors([x,y,b],[ 'red','green','aqua'])
plt.xlim(-10,10)
_=plt.ylim(-10,10)
```



Plotting Ellipse

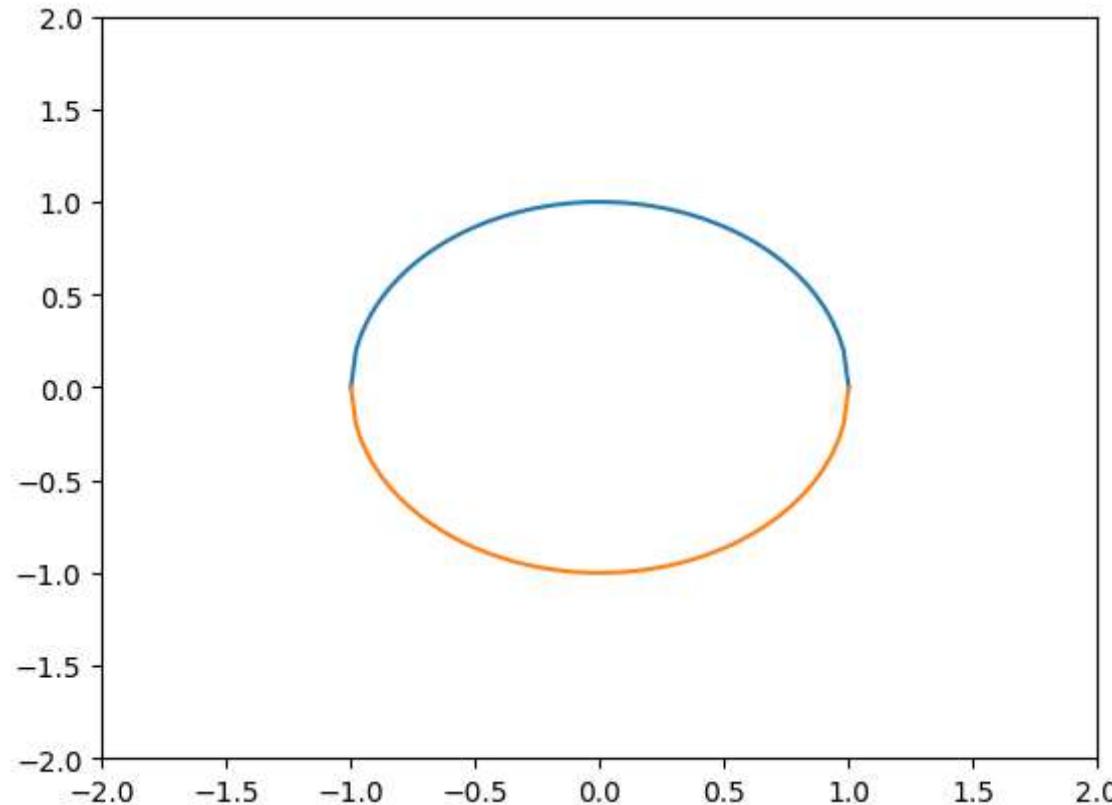
```
In [36]: import numpy as np  
x=np.linspace(-1,1,100)  
#Circle equation: x^2+y^2=100  
#In np.linspace(-1,1,100), we enter the co-efficients of x^2, y^2 and radius^2  
#floating values between -1 to 1 are infinite we have taken 100 values in that
```

```
In [37]: y1=np.sqrt(1-np.square(x))
```

```
In [38]: y2= -1*y1
```

```
In [39]: import matplotlib.pyplot as plt  
plt.plot(x,y1)
```

```
plt.plot(x,y2)
plt.xlim([-2,2])
plt.ylim([-2,2])
plt.show()
```



#Transformation of an ellipse

In [43]:

```
def transformation(x,y):
    return 9*x+4*y,4*x+3*y
#[9,4],[4,3] are the values of 2*2 matrix
```

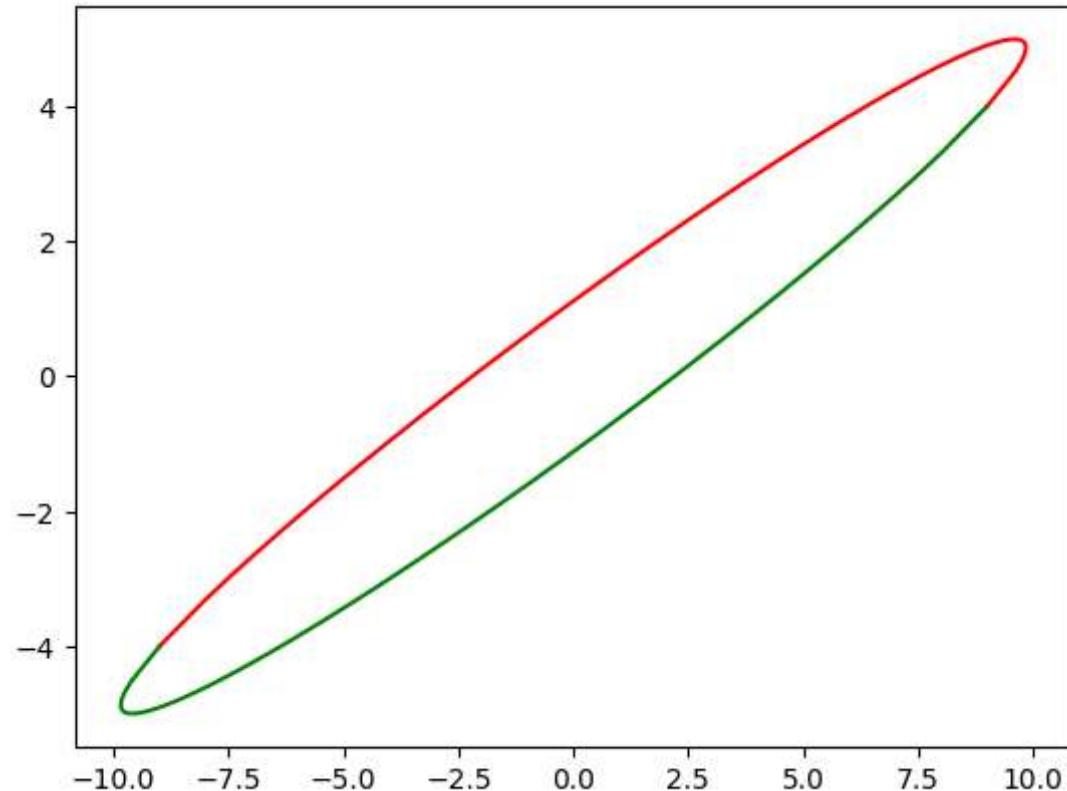
In [46]:

```
#we take new values of transformation into x_new1,y_new1,x_new2,y_new2 variables
```

```
x_new1,y_new1=transformation(x,y1)
x_new2,y_new2=transformation(x,y2)
```

```
In [49]: plt.plot(x_new1,y_new1,'r')
plt.plot(x_new2,y_new2,'g')
```

```
Out[49]: [<matplotlib.lines.Line2D at 0x2d7460a5f60>]
```



```
In [ ]: ##Here, circle is transformed into an ellipse
```

```
In [50]: eigen,eigenvector=np.linalg.eig(np.array([[9,4],[4,3]]))
```

```
In [51]: eigen
```

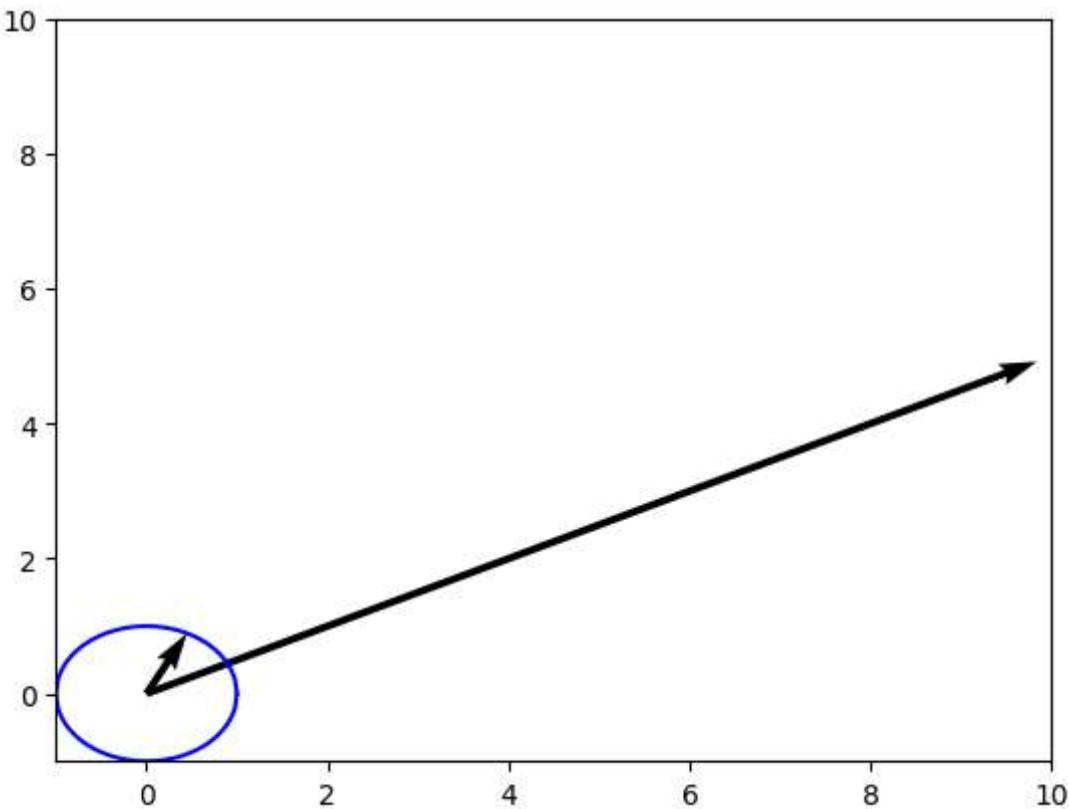
```
Out[51]: array([11.,  1.])
```

```
In [52]: eigenvector
```

```
Out[52]: array([[ 0.89442719, -0.4472136 ],
   [ 0.4472136 ,  0.89442719]])
```

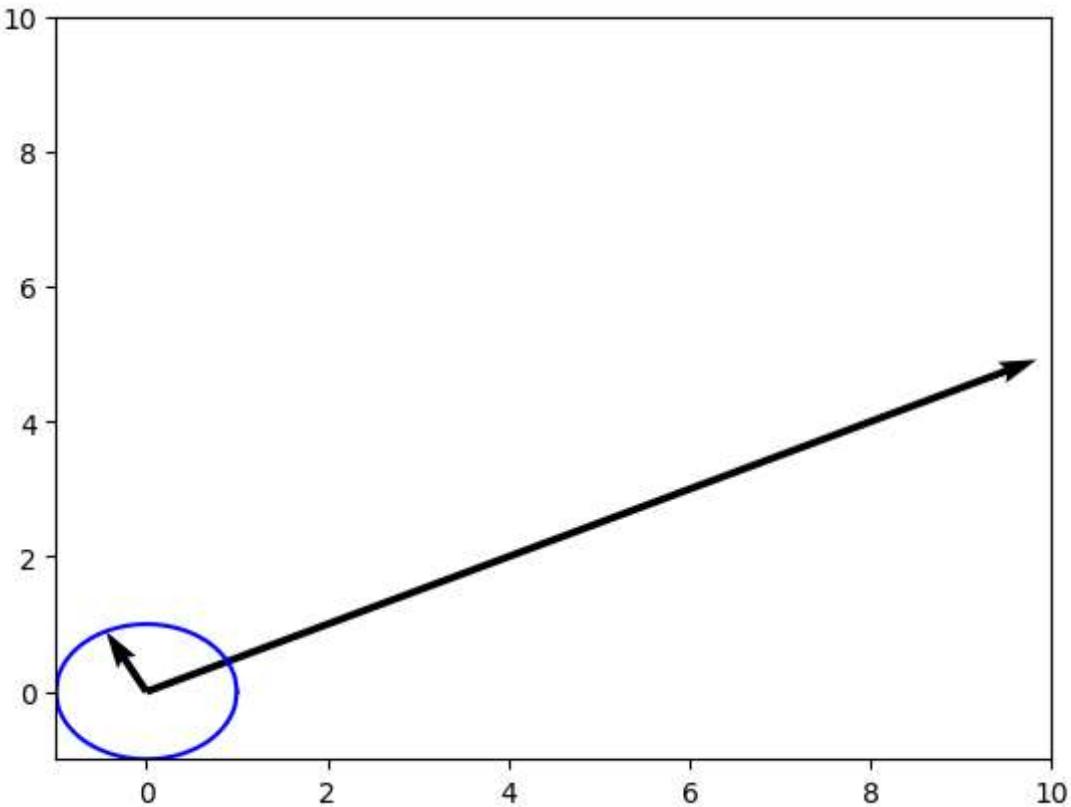
```
In [53]: soa=np.array([[0,0,eigen[0]*eigenvector[0][0],eigen[0]*eigenvector[1][0]]])
sol=np.array([[0,0,eigen[1]*eigenvector[1][0],eigen[1]*eigenvector[1][1]]])
#[]
X,Y,U,V=zip(*soa)
X1,Y1,U1,V1=zip(*sol)
plt.plot(x,y1,'b')
plt.plot(x,y2,'b')
plt.quiver(X,Y,U,V,
            angles='xy',scale_units='xy', scale=1)
plt.quiver(X1,Y1,U1,V1,
            angles='xy',scale_units='xy', scale=1)
plt.xlim([-1,10])
plt.ylim([-1,10])
```

```
Out[53]: (-1.0, 10.0)
```



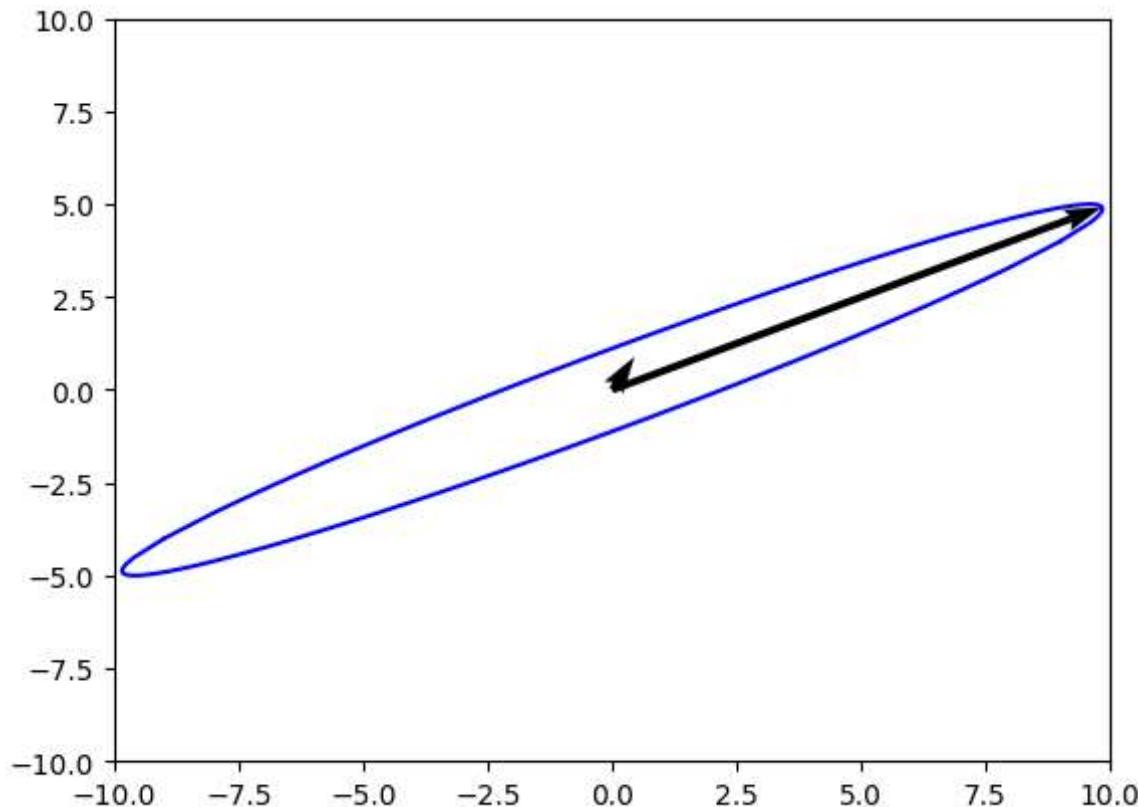
```
In [54]: #Changed the index from eigen[1]*eigenvector[1][0] to eigen[1]*eigenvector[0][1]
#Column Change
soa=np.array([[0,0,eigen[0]*eigenvector[0][0],eigen[0]*eigenvector[1][0]]])
sol=np.array([[0,0,eigen[1]*eigenvector[0][1],eigen[1]*eigenvector[1][1]]])
X,Y,U,V=zip(*soa)
X1,Y1,U1,V1=zip(*sol)
plt.plot(x,y1,'b')
plt.plot(x,y2,'b')
plt.quiver(X,Y,U,V,
            angles='xy',scale_units='xy', scale=1)
plt.quiver(X1,Y1,U1,V1,
            angles='xy',scale_units='xy', scale=1)
plt.xlim([-1,10])
plt.ylim([-1,10])
```

Out[54]: (-1.0, 10.0)



```
In [55]: #Changed the value from x,y1,x,y2 to x_new1,y_new1 and x_new2,y_new2
#Column Change
soa=np.array([[0,0,eigen[0]*eigenvector[0][0],eigen[0]*eigenvector[1][0]]])
sol=np.array([[0,0,eigen[1]*eigenvector[1][0],eigen[1]*eigenvector[1][1]]])
X,Y,U,V=zip(*soa)
X1,Y1,U1,V1=zip(*sol)
plt.plot(x_new1,y_new1,'b')
plt.plot(x_new2,y_new2,'b')
plt.quiver(X,Y,U,V,
           angles='xy',scale_units='xy', scale=1)
plt.quiver(X1,Y1,U1,V1,
           angles='xy',scale_units='xy', scale=1)
plt.xlim([-10,10])
plt.ylim([-10,10])
```

Out[55]: (-10.0, 10.0)



Singular Value Decomposition

```
In [57]: #svd[a] eigenvector of a and a^t d= sq root of eigen values v^t = a^t and a  
#svd returns 3 values  
#svd is applied to reduce the size of images  
#Singular Value Decompostion (SVD): The Singular Value Decompostion (SVD) of a matrix  
#is a factorization of that the matrix into three matrices.  
#U,D,VT  
#U--Eigenvector of AA^T  
#D--Squareroot of eigenvalues  
#V^T--#right singular value, i.e., A^TA  
#Compress an image
```

```
In [ ]: #Compressing an Image
```

```
In [59]: A=np.array([[-1,2],[3,-2],[5,7]])  
A
```

```
Out[59]: array([[-1,  2],  
                 [ 3, -2],  
                 [ 5,  7]])
```

```
In [60]: U,d,VT= np.linalg.svd(A)
```

```
In [61]: U #the column values of u called left singular vector
```

```
Out[61]: array([[ 0.12708324,  0.47409506,  0.87125411],  
                 [ 0.00164602, -0.87847553,  0.47778451],  
                 [ 0.99189069, -0.0592843 , -0.11241989]])
```

```
In [62]: VT# >>> #Column value of right singular vector
```

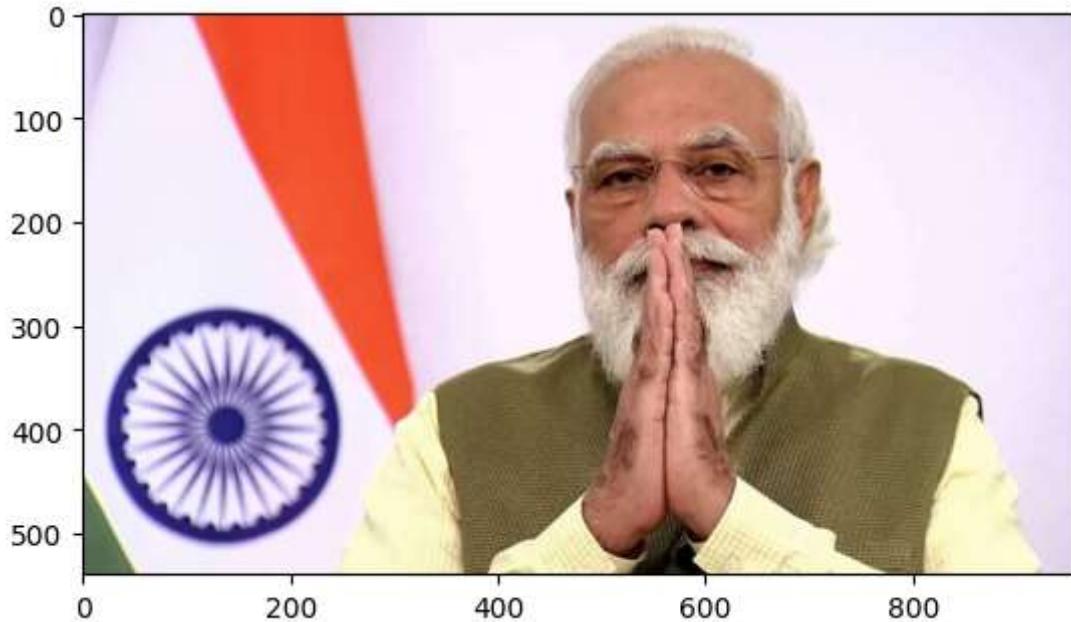
```
Out[62]: array([[ 0.55798885,  0.82984845],  
                 [-0.82984845,  0.55798885]])
```

```
In [64]: d #diag values of singular value
```

```
Out[64]: array([8.66918448, 4.10429538])
```

```
In [65]: from PIL import Image
```

```
In [66]: img = Image.open('modi1.jpg.jpeg')  
_=plt.imshow(img)
```



```
In [67]: imggray = img.convert('LA') #LA indicates color band  
_ = plt.imshow(imggray)
```



```
In [69]: #creating new matrix from image before applying svd
#band=1 rgb
#band =0 for red 1 for green 2 for blue
#size 0 >> no of rows present and size 1>> no.of coloumns
#svd compresses the image
imgmat=np.array(list(imggray.getdata(band=0)),float)
imgmat.shape =(imggray.size[1],imggray.size[0])
imgmat=np.matrix(imgmat)
_=plt.imshow(imgmat, cmap='gray')
```



```
In [70]: U, sigma, V = np.linalg.svd(imgmat)
```

```
In [71]: U
```

```
Out[71]: matrix([[-0.04759306,  0.02120062,  0.03605262, ...,  0.00534043,
   0.01610291,  0.02993521],
 [-0.04759571,  0.02127977,  0.0359307 , ...,  0.0043117 ,
 -0.07943641, -0.03956248],
 [-0.04762426,  0.02141364,  0.03607913, ..., -0.00332636,
  0.05008565,  0.00585326],
 ...,
 [-0.04537899, -0.02159797,  0.08780722, ..., -0.0010821 ,
 -0.00111973, -0.00021762],
 [-0.04534663, -0.02054186,  0.08475936, ...,  0.00020093,
  0.00351981, -0.0012747 ],
 [-0.04530587, -0.01970305,  0.08261453, ...,  0.00023264,
 -0.00297974,  0.0015839 ]])
```

```
In [72]: V
```

```
Out[72]: matrix([[-0.03516809, -0.03520418, -0.03526966, ..., -0.03746736,
   -0.0374089 , -0.03739109],
   [-0.00572139, -0.00620153, -0.00698787, ..., -0.02780694,
   -0.02779575, -0.02773563],
   [-0.06947448, -0.06993877, -0.07049955, ..., -0.01715403,
   -0.01722642, -0.01726522],
   ...,
   [ 0.0045503 , -0.00279934, -0.01781835, ...,  0.79109661,
   -0.05618234, -0.06572806],
   [-0.05452047, -0.01220729,  0.01027495, ..., -0.04645416,
   0.77358226, -0.06230439],
   [-0.02596574,  0.00392267,  0.00405014, ..., -0.02919379,
   -0.02963976,  0.74059022]])
```

```
In [73]: sigma
```

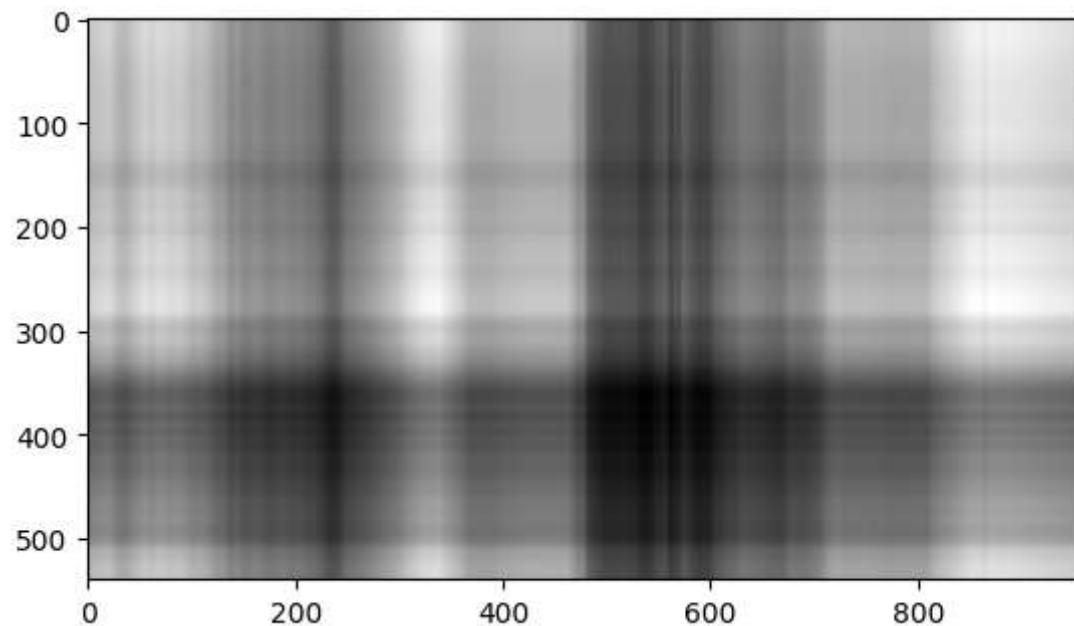
```
Out[73]: array([1.39617756e+05, 2.07387337e+04, 1.44637774e+04, 1.24224845e+04,
 9.82499351e+03, 8.28718025e+03, 7.76586823e+03, 6.08976516e+03,
4.94608675e+03, 4.75129425e+03, 4.10933494e+03, 3.58078398e+03,
3.47809188e+03, 3.22403074e+03, 2.89287851e+03, 2.81016782e+03,
2.61409305e+03, 2.39554512e+03, 2.24189338e+03, 2.20566843e+03,
2.05941918e+03, 1.96053731e+03, 1.83318235e+03, 1.80505193e+03,
1.77128447e+03, 1.73844682e+03, 1.72845085e+03, 1.65298952e+03,
1.64128038e+03, 1.59099438e+03, 1.56440590e+03, 1.45847421e+03,
1.40807315e+03, 1.38456869e+03, 1.37123507e+03, 1.33663498e+03,
1.30466006e+03, 1.26464839e+03, 1.24778185e+03, 1.21453104e+03,
1.19115790e+03, 1.13003035e+03, 1.10675280e+03, 1.07487462e+03,
1.06582225e+03, 1.03942994e+03, 1.01215904e+03, 1.00116796e+03,
9.80452346e+02, 9.66643889e+02, 9.32424096e+02, 9.16660212e+02,
8.99665508e+02, 8.91109672e+02, 8.71135759e+02, 8.58619107e+02,
8.40786412e+02, 8.15021637e+02, 8.01947255e+02, 7.93812423e+02,
7.85572827e+02, 7.71834543e+02, 7.51345953e+02, 7.36410462e+02,
7.22327452e+02, 7.05894336e+02, 6.95608000e+02, 6.87287913e+02,
6.72895643e+02, 6.66884072e+02, 6.59436876e+02, 6.50667686e+02,
6.38679033e+02, 6.31872559e+02, 6.22868440e+02, 6.18990981e+02,
6.05889122e+02, 5.94981766e+02, 5.89062745e+02, 5.78339150e+02,
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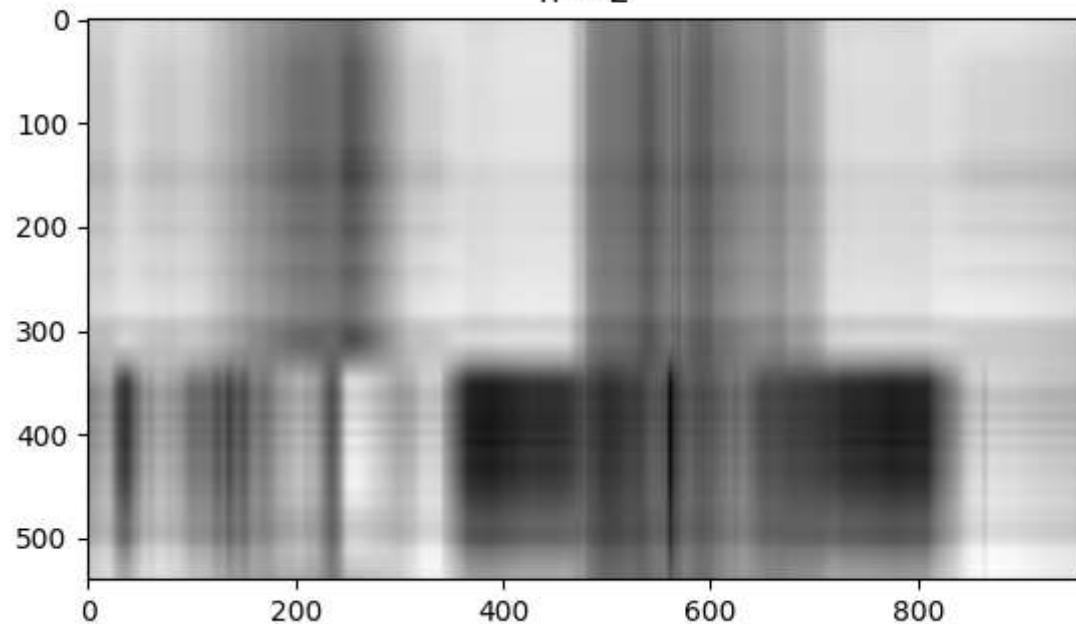
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```

```
In [74]: #reconstructing image
reconstimg= np.matrix(U[:, :1])*np.diag(sigma[:1])*np.matrix(V[:1, :]) #:1 means 1 column
_= plt.imshow(reconstimg, cmap='gray')
```

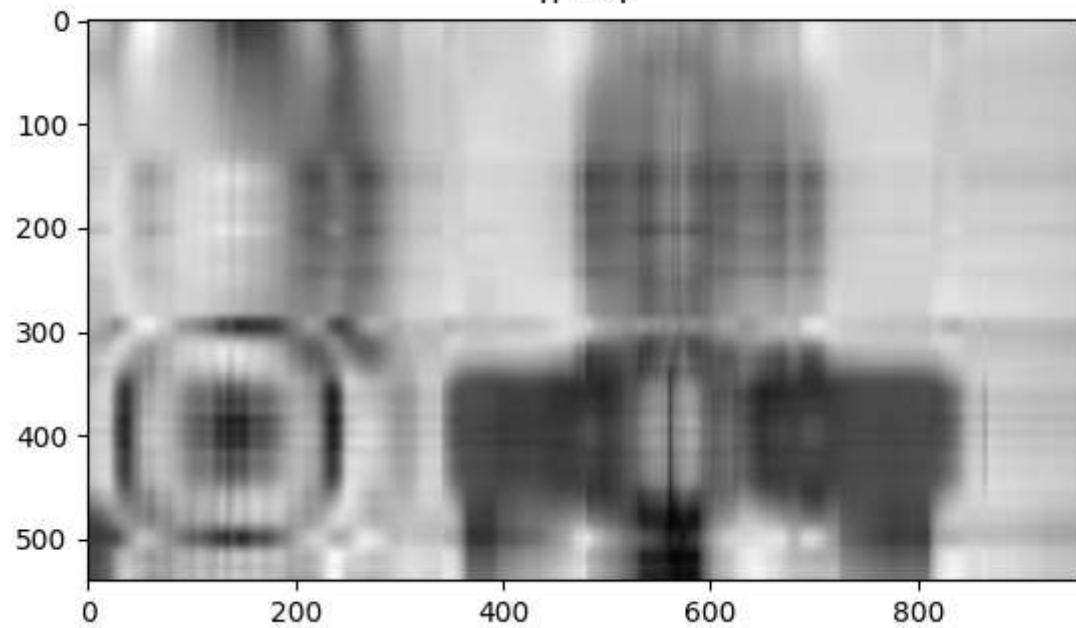


```
In [75]: #more coloumns>> for more clear image>> construct a for Loop
for i in [2,4,8,16,32,64]:
    reconstimg = np.matrix(U[:, :i]) *np.diag(sigma[:i]) * np.matrix(V[:i, :])
    plt.imshow(reconstimg, cmap='gray')
    title = "n = %s" % i
    plt.title(title)
    plt.show()
```

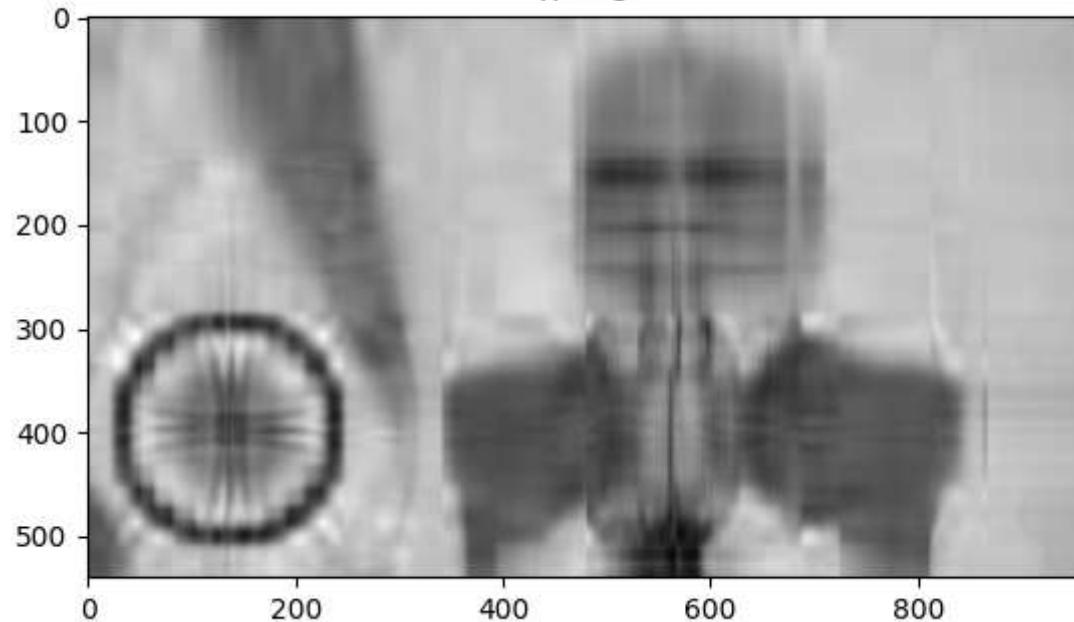
$n = 2$



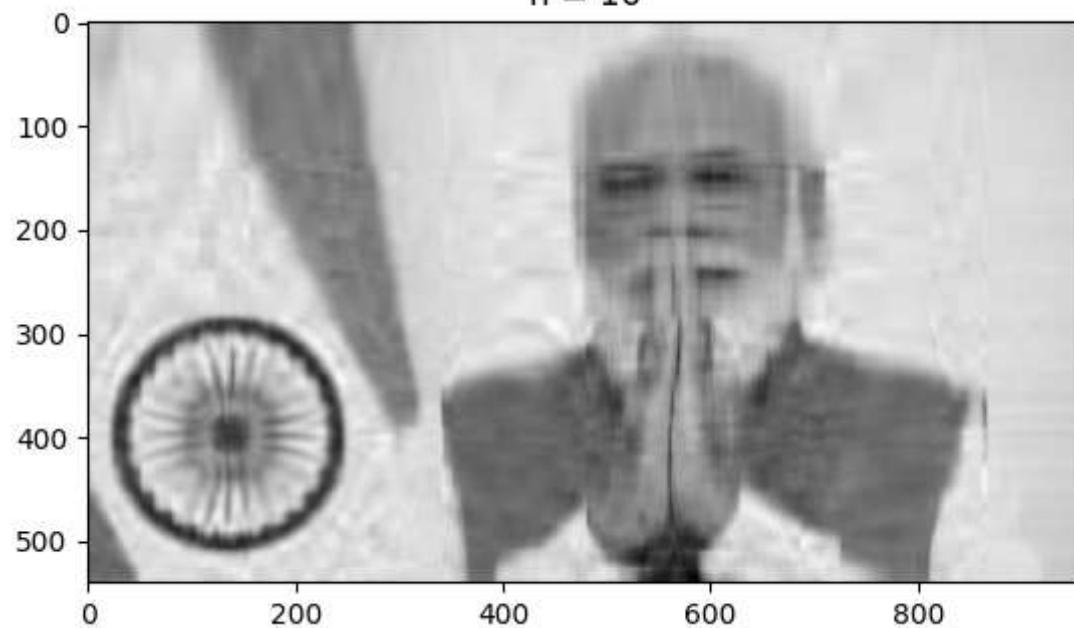
$n = 4$



$n = 8$



$n = 16$



$n = 32$



$n = 64$



In [76]: *#dimensionality redn pca =1 means it generates 1 column
#pca is feature extraction technique*