5) 
$$\begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix} = A$$

when 2 reows are same in a matix, it value is zero. 
$$1AI = D$$
 $A^3 + A^2 + A + 1AI = D$ 

(A 2 + A + J = 0

and hence over eigen value is zero.

$$A = \begin{bmatrix} 1 & 2 & 3 & 0 \\ 2 & 4 & 3 & 2 \\ 3 & 2 & 1 & 3 \\ 6 & 8 & 7 & 5 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 2 & 3 & 0 \\ 2 & 4 & 3 & 2 \\ 3 & 2 & 1 & 3 \\ 6 & 8 & 7 & 5 \end{bmatrix} \begin{array}{c} R_2 \longrightarrow R_2 - 2R_1 \\ R_3 \longrightarrow R_3 - 3R_1 \\ R_4 \longrightarrow R_4 - 6R_1 \\ 0 & -4 & -11 & 5 \\ 0 & -4 & -11$$

Rank = 3 (no. of non zeros in a read)

3)  $A = \begin{bmatrix} 2 & -1 \end{bmatrix}$  Eigen values and eigen vectore of A - J and  $\begin{bmatrix} -1 & 2 \end{bmatrix}$  A+4I.

$$\begin{bmatrix} 2-\lambda & -1 \\ -1 & 2-\lambda \end{bmatrix}$$

$$\begin{vmatrix} 2-\lambda & 2 \\ -1 & 2-\lambda \end{vmatrix}$$

$$\begin{vmatrix} 2-\lambda & 2 \\ \lambda^2 + 4 - 4\lambda - 1 \\ \lambda^2 - 4\lambda + 3 = 0 \end{vmatrix}$$

The 
$$A^{2} = -3A - A + 3 = 0$$
 $A(A-3) = 1(A-3) = 0$ 
 $A(A-3) = 0$ 
 $A = 1, 3$ 

Por  $A = 1$ 
 $A = 1, 3$ 

Por  $A = 3$ 
 $A = 1 = 1$ 
 $A = 1, 3$ 

Por  $A = 3$ 
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Por  $A = 3$ 
 $A = 1, 3$ 

Por  $A = 3$ 

Por  $A$ 

for second iteration base case (n,y,z) = (2-6167, 19.5617, 6.6703) N= 7.85+0.1(19.5617) +0.2(6.6703) = 7.716 y = -19.3 + 6.1 (7.716) + 0.3 (6.6703) = -2.3671.4 - 6.3(7-716) - 0.2(-2.36) = 6.956for third iteration, base case (n,y, z) = R7.716, -2.36, 6.956)

n= 7.85+0.1(-2-36)+0.2(6.956) = n=3.002

-19.3 +0.1 (3.002) +0.3 (6.956) (3) Jacobs 1046 (6) 25-2.426.

Z= 71.4 - 0.3 (3.002) - 0.2 (-2.455) = Z=7.099 1000

T a b = ca-b)+(b-c)n+(e-a)n3. Rank and numity.

T:W -P2

dimensions of w:-Symmetric 2x2 matrix has 2 independent value:-

The diagonal element and off-diagonal element. a b [b=c] [a] [d]

dim(w)=3

Rank of Ti-That all polynomials in P that can come by applying 7 to 2x2 max Cegimmetricia.

T mapping to all polynomial of degree at 2.

Using Rank - nullity Theorem, we have :-

$$\begin{bmatrix}
3 & 1 & 3 & 2 & 0 \\
2 & -1 & 3 & 0 \\
3 & -5 & 4 & 0
\end{bmatrix}$$

$$\begin{bmatrix}
R_2 \rightarrow R_2 - 2R_1 & 0 & -4 & -1 & 0 \\
R_3 \rightarrow R_3 - 3R_1 & 0 & -14 & -2 & 0 \\
R_4 \rightarrow R_4 - R_1 & 0 & 14 & 2 & 0
\end{bmatrix}$$

( Jacobis method cperyorem 3 îtercatione).

$$9 -4x + y - z = -15$$

$$x -3y +7z = 16$$

$$3x-6y+2z=23$$
 initial value  $x-207=1$ 
 $4x+y-z=-15$ 
 $x-3y+7z=16$ 
 $z-207=1$ 

= 23 #6y-2z

itercation 2:

$$y = -15 + 4x + 2$$

$$z = \frac{16 - \chi + 3y}{7}$$

$$z = \frac{1}{3}$$

$$z = \frac{1}{3}$$

$$z = \frac{3}{3}$$

 $21 = \frac{16-1+3(1)}{7} = 2$ 

$$\frac{9(2)}{3} = \frac{23+6(-10)-2(2)}{3} = -27$$

$$\frac{1}{3}$$

$$\frac{1}{3} = -15+4(-25)+2 = -102$$

$$\frac{1}{7}$$

$$\frac{1}{7} = \frac{16-(-25)+3(-10)=-5}{7}$$

Herafian 3:

$$168 = 23 + 6(-102) - 2(-5) = 187$$

y 3 = -15 + 4(187) - (-5) = 753

 $Z = \frac{16 - 187 + 3(-102)}{7} = = 138$ 

7 = 187, y = 453, 7 = -138

© 7:  $P_2 \rightarrow P_1$  is the are the number mation on not.  $7(a+bx+cx^2) = (a+1) + (b+1)x + (c+1)x^2$ 

Additivity: T (u+u) = T(u) + T(v) for all v, v in P2

Homogeneity: T(cu) = CT(u) for all us in P2 and all scalars in c.

E(E, 4 = ) (0,4, 2), (8,0,0) 3 = 3

①  $u = a + bx + cx^2$  and  $v = a^1 + b^1x + c^1x^2$  in  $P_2$ 

9hu utv = (a+a) + (b+b) x + (c+c') u2

Now, T(u) = (a+1) + (b+1) 24 (c+1) x2 and

7(V) = (a'+1) + (b'+1) + (c'+) x2

9 (u+v) = (a+a'+1) + (b+b'+1)x+(c+c'+1)x2

7 (u) + 9(v) > Ca'+1) + (b+b'+1) + (c+c'+1) x2

7(u) +7(v) = (a+1) + (b+1) + (c+1) x 2+ (a+1) +(b+1) + (c+1)

=: (a+a2+1)+ (b+b'+1)+(c+c'+1) x2

n(a+v) = T(i)++ T(v) sarify additivity.

1 u = atbx + cx2 - P2 , as let L be sealor

Cu = ca + Cb x + ccx2

1(cul = (ca+1) + (cb+1)x+(cc+1)x2

c7(w = c(a+1) + c (b+1) x + c(c+1) x2

Cost (cat1) + (b+1) x + (cc+1) x 2 satisfy homogeneity function of satisfies both additivity and homogeneity so it is a linear transformation.

$$AX = 0$$

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

$$R_2 \rightarrow R_2 - 3R_1 \begin{bmatrix} 1 & 2 & 3 \\ 0 & -5 & -9 \\ -2 & 1 & 3 \end{bmatrix}$$

$$R_3 = R_3 + R_2 \begin{bmatrix} 1 & 2 & 3 \\ 0 & -5 & -9 \\ 0 & 0 & 0 \end{bmatrix}$$

(A) = 2 cinearly dependent, R3 has dimension 3.

Sig unearly independent -> span &3

S is cinearly independent and consist of a vector in Ps

i di ancare transportation

4000) = (a+1) = (b+1)

1 Explain one application of matrix operations in image proceeding with example.

One extensive application of matrix operation in image prevelling image treamformation. It includes task such as scaling, treamlation and restation.

Here are the steps.

De Rotation matrix: - To restate an image, you can we a cot matrix. For a 20 image restation, where typically we = 12 The form of the restation matrix for 20 restation by un angle o.

[cor(0) - sin(0)] sin(0) co(0)]

- (1) Represent pixel coordinates each pixel in the original image can be represented as a vector containing its or and y coordinates.
- (2) Apply notation matrix-multiply each pixel vector by the reotation matrix to get the new coordinates of the reotated pixel.
- (3) Interpolation After reotation, the new pixel positions might not align perfectly withthe grid of pixels. Interpolation methods can be used to determine the intensity values of the pixels in the reotated image based on the intensity values of neighborsing pixels.
- (4) Apply to entire page

for example - [1 2 3]

4 5 6

Rotating 9mg by 90° countex clock wire.

[1 0] After performing matrix multiplication with each 1 0] pixel coordinate, we will get outer restarted image mate

Then we will apply interpolation to get the final restated image.

Continuar transformations play a crucial rule in computer vision talk, particularly in manipulation and analyzing their operation efficiently image. When it comes to restating a 2D image, linear treansformations are employed to achoive this operation efficiently.

In the context of restating a 2D image, a linear treamformation involves applying a treamformation matrix to every pixel in

This treansformation matrix represents the reptation operention for 2D reptation, it looks like

[ coro -sino ]

where or is the angle of reof-ation in readians.

when this treansportmation matrix is applied to the cooredinates of each pixel in the original image. It effectively restates the entires image amound the origin.

Linear treansformations are preferenced for image restation in compiler vision because they preserve important properties such as incommains straight and parallel lines remains parallel expert treansformation. Additionally, linear treansformations can be efficiently implemented using matrix multiplication, making them computationally tac freatable in summary linear treansformations are fundamental in restation 2 Dimages in computer vision. They provide a mathematically and computationally efficient way to provide a mathematically and computationally efficient way to provide a mathematically and computationally efficient way to provide a mathematically images while preserving impositant geometric properties.