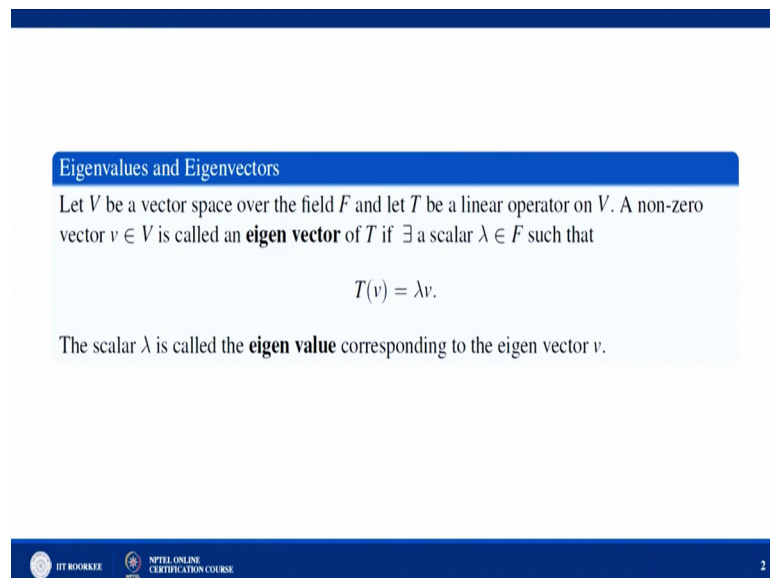


Matrix Analysis with Applications
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Lecture – 13
Eigenvalues and Eigenvectors

Hello friends, welcome to lecture series on Matrix Analysis with Applications. Now, today's lectures based on Eigen values and Eigenvectors. That if we have a given matrix A , then how can you find this eigen values in the corresponding eigenvectors.

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The slide content is as follows:

Eigenvalues and Eigenvectors

Let V be a vector space over the field F and let T be a linear operator on V . A non-zero vector $v \in V$ is called an **eigen vector** of T if \exists a scalar $\lambda \in F$ such that

$$T(v) = \lambda v.$$

The scalar λ is called the **eigen value** corresponding to the eigen vector v .

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So, first of all what eigen values and eigenvectors are. So, let V be a vector space over the field F , and T be a linear operator on V that is T is a linear transformation from V to V . A non-zero vector v belongs to V is called an eigen vector of T , if there exists a scalar λ belongs to F such that $T v$ equal to λv . So, if we have a linear operator T , and we have a non-zero vector v in V , and there exist a λ belongs to field such that $T v$ equals to λv , then we say that λ is eigen value, and the corresponding eigen vector is v .

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A square matrix A may be viewed as a linear operator T , defined by

$$T(x) = Ax, \text{ where } x \in V.$$

Definition

Let $A = (a_{ij})$ be a square matrix of order n . Consider the homogeneous system of equations

$$Ax = \lambda x$$

The values of λ for which system of linear equations has non-trivial solutions are called **eigen values** or the characteristic values and the corresponding non-zero vectors x are called the **eigen vectors** or the characteristic vectors of A .

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Now, a square matrix A may be viewed as a linear operator T , and is defined as $T x$ equals to $A x$, where x belongs to V ok. So, if you have a linear operator the linear operator, then A matrix can be viewed as a linear operator also. Now, now let us define a matrix A , which is a square matrix of order n . Then if we consider a system of homogenous equation $A x$ equals to λx .

Then the value of λ for which the system of linear equations has non-trivial solution are called eigen values or the characteristic values of A , and the corresponding non-zero vectors x are called eigen vectors or the characteristic vectors of A . Since, since every matrix can be viewed as a linear operator T , so in the previous definition we have replace this T by a matrix A here. And this x is in V , this is a scalar λ in field ok.

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$$\begin{aligned} A_{n \times n} \quad & AX = \lambda X, \quad X \neq 0 \\ & \lambda \rightarrow \text{eigen value of } A \\ & X \rightarrow \text{corresponding e-vector of } A \\ \\ AX = \lambda X \quad & \Rightarrow AX - \lambda X = 0 \\ & \Rightarrow (A - \lambda I)X = 0, \quad X \neq 0 \\ & n(A - \lambda I) \geq 1 \quad \dim(A - \lambda I) = n \\ & r(A - \lambda I) + n(A - \lambda I) = n \\ & \Rightarrow r(A - \lambda I) = n - n(A - \lambda I) \\ & \leq n - 1 \\ & \Rightarrow |A - \lambda I| = 0 \rightarrow \text{characteristic} \\ & \text{polynomial of } A \end{aligned}$$

So, what I want to say basically, suppose you have a matrix A , which is of order n by n . Then if $AX = \lambda X$, where X is not equal to 0. Then this λ is called eigen value of A and X is called corresponding eigen vector of A , or eigen vector corresponding to λ .

Now, we are having $AX = \lambda X$, so this can be written as $AX - \lambda X = 0$. Now, this can be written as $(A - \lambda I)X = 0$, where I is an identity matrix. Now, what we are having basically, we are having $(A - \lambda I)X = 0$ means, we are having system of homogeneous linear equations you know, because right side is 0.

Now, now since X is not equal to 0 as we already defined a definition X is not equal to 0. So, this means if we are taking this as a linear operator T , if we take $(A - \lambda I)$ as a linear operator T , then $TX = 0$, that means, that means X is not equal to 0, that means, the nullity the null space. The nullity of this matrix or this operator is greater than or equal to 1 ok, because X is not equal to 0 means this means nullity is not 0, and nullity is not 0 means, it is greater than or equal to 1.

Now, what is the dimension of $(A - \lambda I)$ of course, dimension of $(A - \lambda I)$ is n . So, rank by rank nullity theorem, the rank of $(A - \lambda I)$ plus nullity of $(A - \lambda I)$ must be equal to dimension of matrix or

operator, which is n . So, this implies rank of $A - \lambda I$ must be equal to n minus nullity of $A - \lambda I$.

And since, nullity is greater than equal to 1, this may this is less than equal to $n - 1$. And since, rank is less than equal to $n - 1$ less than equal to $n - 1$, this means in the echelon form of this matrix at least 1 0, 1 row is there, which is having all 0 elements, and that means, this implies determinant of $A - \lambda I$ must be 0. Now, this equation, determinant of $A - \lambda I$ equal to 0 is called a characteristic polynomial, this will give a characteristic polynomial in λ characteristic polynomial of λ [FL]. Now, now this is a matrix of order and cross n . So, determinant, so this determinant will give a polynomial of degree n in λ .

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If the homogeneous system $(A - \lambda I)x = 0$ has a non-trivial solution, then the rank of the coefficient matrix is less than n , i.e. coefficient matrix $(A - \lambda I)$ must be singular. Therefore,

$$\Rightarrow \begin{vmatrix} |A - \lambda I| = 0 \\ a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \vdots & \vdots & \dots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} - \lambda \end{vmatrix} = 0$$

$$\Rightarrow P_n(\lambda) = |A - \lambda I| = (-1)^n [\lambda^n - c_1 \lambda^{n-1} + c_2 \lambda^{n-2} - \dots + (-1)^n c_n] = 0 \quad (1)$$

This equation is called the characteristic equation of the matrix A . The roots $\lambda_1, \lambda_2, \dots, \lambda_n$ of the polynomial $P_n(\lambda)$ are called the **eigen values**.

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So, so what we are having now? So, now if we if we visualize the determinant of $A - \lambda I$ equal to 0, that means, that means this determinant is equal to 0 you see.

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$$\begin{aligned}
 |A - \lambda I| = 0 &\Rightarrow \left| \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & \dots & 1 \end{pmatrix} \right| = 0 \\
 &\Rightarrow \left| \begin{pmatrix} a_{11} - \lambda & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} - \lambda & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} - \lambda \end{pmatrix} \right| = 0 \\
 &\Rightarrow (-1)^n [\lambda^n - c_1 \lambda^{n-1} + c_2 \lambda^{n-2} - \dots + (-1)^n c_n] = 0 \\
 &\underbrace{(a_{11} - \lambda)} \left| \begin{matrix} a_{22} - \lambda & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n2} & \dots & \dots & a_{nn} - \lambda \end{matrix} \right| - a_{12} \underbrace{\left[\begin{matrix} \lambda^{n-2} & \dots & \dots \end{matrix} \right]}_{\lambda^{n-2}} \\
 &\underbrace{(a_{11} - \lambda)} \left[\begin{matrix} (a_{22} - \lambda) & \dots & \dots & \dots \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \ddots & \vdots \end{matrix} \right] - a_{23} \underbrace{\left[\begin{matrix} \lambda^{n-3} & \dots & \dots & \dots \end{matrix} \right]}_{\lambda^{n-3}} \dots
 \end{aligned}$$

If we are taking determinant of A minus lambda I equal to 0, so it implies that determinant of A is this matrix you see, a 1 1, a 1 2 if you are taking this matrix as a and a and 1 up to a n n minus lambda times I is identity matrix. So, when you take determinant put it equal to 0, so this is nothing but determinant of you see, only the only this lambda will be subtracted from the diagonal elements, because all other entries are 0. So, this will be determinant of a 1 1 minus lambda a 1 2 and so on up to a 1 n, then a 2 1, a 2 2 minus lambda, then a 2 n and so on a n 1, a n 2, and so on up to a n n minus lambda, this determinant is equal to 0.

So, this will give us a polynomial in lambda, and that polynomial is called characteristic polynomial ok. So, this is basically gives minus 1 raised to power n lambda raised to power n minus say c 1 lambda raised to power n minus 1 plus c 2 lambda raised to power n minus 2 minus and so on plus minus 1 raised to power n into c n is equal to 0 ok. Because when you when you expect this in terms of lambda, so you will get a polynomial in of degree n in lambda.

Now, now if you carefully observe this determinant you see, when you open from the first element you see, if you take a 1 1 minus lambda, then you take then you delete this row and this row and this column. So, this will be the remaining I mean determinant minus a 1 2.

Now, when you take, when you expand from this element, you delete this column and this row, that means, the two values of two linear factors of lambda are not coming in this in this expansion, so that means, this expansion will contain a maximum lambda raised to power $n - 2$, because when you take, when you expand from this, you are deleting this row and this column. So, two factors of lambda are not coming in are not coming in that expansion, that means, the maximum power of lambda, which is coming from this term is maximum power of lambda, which is coming from this term is lambda raised to power $n - 2$.

Similarly, if you take a 1×3 , so here will be something a 3×3 minus lambda, and when you delete this column and this row, then again two factors of lambda are not coming, so that means, maximum power of lambda, which are coming from this term is lambda raised to power $n - 2$. Similarly from the other terms so, what we can conclude that the remaining terms will contain say α_1 lambda raised to power $n - 1$, and so lambda raised to power $n - 2$ and so on, so that means, the power of lambda raised to power n , and power of lambda raised to power $n - 1$ are coming only from the first determinant.

Now, if you see here, again if you see here, now if you open from this is ok, if you open from this, here is something a 3×3 minus lambda is again here, if you open from this element, again you will delete this column and this row. So, two powers of lambda are not are still not here I mean not coming, so that means, you see if you take a 1×1 minus lambda, if you expand, this determinant then it is a 2×2 minus lambda, and again some determinant plus and minus a 2×3 , and determinant of some other thing, which in which 2 power 2 powers of lambda are not there ok.

So, that means, that means, it will contain maximum lambda raised to power $n - 2$, and minus 3, because this determinant is having lambda raise to power $n - 1$ 1 1 1 factor is already outside. So, this determinant is having lambda raised to power $n - 1$. And two factors are not coming, that means, the maximum power, which this is this element is having is lambda raised to power $n - 3$.

And when you multiply this with lambda, so it will be having lambda raised to power $n - 2$, so that means, what I want to say basically that when you expand similarly, the entire determinant. So, lambda raised to power n and lambda raised to power $n - 1$

are only coming in the product of the diagonal elements. When you similarly try to expand this entire determinant, so lambda raised to power n, and lambda raised to power minus 1 is coming only from product of the determinant diagonal elements ok.

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$$\begin{aligned} \text{coeff of } \lambda^n &= \text{coeff of } \lambda^n \text{ in } [(a_{11}-\lambda)(a_{22}-\lambda) \dots (a_{nn}-\lambda)] \\ &= (-1)^n \\ \text{coeff of } \lambda^{n-1} &= \text{coeff of } \lambda^{n-1} \text{ in } (a_{11}-\lambda)(a_{22}-\lambda) \dots (a_{nn}-\lambda) \\ &= a_{11} + a_{22} + \dots + a_{nn} \end{aligned}$$

$$|A - \lambda I| = (-1)^n [\lambda^n - c_1 \lambda^{n-1} + c_2 \lambda^{n-2} - \dots + (-1)^n c_n] = 0$$

Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the roots of this equation.

$$\lambda_1 + \lambda_2 + \dots + \lambda_n = -(-c_1) = a_{11} + a_{22} + \dots + a_{nn} = \text{trace}(A)$$

$$\left. \begin{array}{l} \lambda = 0 \\ |A| = c_n \end{array} \right\} \begin{array}{l} \text{Product of roots} = \lambda_1 \lambda_2 \dots \lambda_n \\ = \frac{(-1)^n c_n}{(-1)^n} = |A| \\ \Rightarrow \lambda_1 \lambda_2 \dots \lambda_n = |A| \end{array}$$

So, and all other and the lambda raised to power n minus 2 lambda raised to power n minus 3 are from the all from all the components, may be from all the components, all the elements, so that means, that means coefficient of lambda raised to power n is simply coefficient of lambda raise to power n in simply a 1 1 minus lambda a 2 2 minus lambda and so on a n n minus lambda. And it is simply minus 1 raised to power n, so carefully see it is simply minus 1 raised to power n.

And similarly, if you want to see coefficient of lambda raised to power n minus 1, it is simple coefficient of lambda raise to power n minus 1 in again this product a 1 1 minus lambda a 2 2 minus lambda and so on up to a n n minus lambda. So, so what is the coefficient of lambda minus lambda raised to power n minus 1 in this, it is simply a 1 1 plus a 2 2 plus and so on a n n. You can you can simply see, the sum of these elements a 1 1 plus a 2 2 and so on up to a n n will be the coefficient of lambda raised to power n minus 1.

So, what basically now we are having, we are having determinant of A minus lambda I is something like minus 1 raised to power n, and it is lambda raised to power n minus c 1

$\lambda^{n-2} + c_2 \lambda^{n-1} + \dots + c_n$, so this is equal to this.

Now, let so how many roots this equation will be having, this equation is will be having n roots, because the it is of n degree polynomial. So, let λ_1, λ_2 and so on up to λ_n are the roots be the roots of this equation.

So, if these are the roots, then what is the sum of the roots, you can simply see, the sum of the roots is simply this is the equation. So, some of the roots of this will be simply negative of c_1 . It is minus negative of negative of c_1 , and this is equal to basically this is equal to you see the coefficient of λ^{n-1} is simply this, so this is $a_{11} + a_{22} + \dots + a_{nn}$. And this is called a trace of a matrix, trace of a matrix A .

So, what we have concluded, we have concluded that sum of eigen values is nothing but the trace of the metrics, that means, some of the diagonal elements, the first property. The second property is now this result that this is equal to this whole for every λ for any λ ok, so this whole for λ equal to 0 also. And when you substitute λ equal to 0 we will get determinant of A as equal to c_n as c_n .

Now, if you find product of roots for this equation, product of roots. So, product of roots are $\lambda_1, \lambda_2, \dots, \lambda_n$, and that is simply equal to you see it is c_n raised to power degree of equation last term upon first term, which is nothing but c_n ok, c_n upon λ^{n-1} raised to power n , and that is simply equal to determinant of A . So, this implies product of eigen values is nothing but determinant of A . So, we have noted here, two important properties of eigenvalues, the one property is that the sum of the eigen values is equal to trace of the matrix that is the sum of the diagonal elements, and the second property is the product of the eigen values is equal to the determinant of the matrix A .

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Relations between the roots and the coefficients

Sum of eigen values of $A = \lambda_1 + \lambda_2 + \dots + \lambda_n$
 $= a_{11} + a_{22} + a_{33} + \dots + a_{nn}$
 $= \text{trace}(A)$

Product of eigen values of $A = \lambda_1 \lambda_2 \lambda_3 \dots \lambda_n$
 $= |A|$

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So, the sum of the eigen values simply equal the trace of the matrix. And the product of the eigen values simply the determinant of A.

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Problems

- Let $A = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$. Find the eigen values and the corresponding eigen vectors of A .
- Let $B = \begin{pmatrix} 0 & 0 & 2 \\ 0 & 2 & 0 \\ 2 & 0 & 3 \end{pmatrix}$. If one of the eigen value of B is 4 then find the remaining two eigen values. Also find the corresponding eigen vectors of B .

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Now, let us let us discuss this problem. The first problem is let us consider this matrix A, which is 1 minus 1 1 1, let us find this eigen values in the corresponding eigenvectors.

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$$\begin{aligned}
 A &= \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} & |A - \lambda I| &= 0 \\
 & & \Rightarrow \begin{vmatrix} 1-\lambda & -1 \\ 1 & 1-\lambda \end{vmatrix} &= 0 \\
 & & \Rightarrow (1-\lambda)^2 + 1 &= 0 \\
 & & (1-\lambda)^2 &= -1 \Rightarrow 1-\lambda = \pm i \text{ or } \lambda = 1 \pm i.
 \end{aligned}$$

- e-vector corresponding to $\lambda = 1-i$

$$\begin{aligned}
 (A - \lambda I)X &= 0 \Rightarrow (A - (1-i)I)X = 0 \Rightarrow \begin{pmatrix} i & -1 \\ 1 & i \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\
 X &= \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} & \Rightarrow \begin{pmatrix} 1 & i \\ i & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} &= \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\
 &= \begin{pmatrix} -ix_2 \\ x_2 \end{pmatrix} & R_2 \rightarrow R_2 - iR_1 &\Rightarrow \begin{pmatrix} 1 & i \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \\
 &= x_2 \begin{pmatrix} -i \\ 1 \end{pmatrix} & \Rightarrow x_1 + ix_2 &= 0
 \end{aligned}$$

corresp to $\lambda = 1-i$, the LI e-vector is $(-i \ 1)^T$.

So, the matrix A is simply 1 minus 1 1 1. So, you can simply write A minus lambda I determinant this must be 0 for the characteristic polynomial, or finding the roots of the equation, roots of the latent roots or characteristic roots of this matrix A, so this is nothing but 1 minus lambda minus 1 1 1 minus lambda determinant equal to 0. And this implies on minus lambda whole square plus 1 should be 0, or 1 minus lambda square is equal to minus 1 that implies on minus lambda is equal to plus minus iota, or lambda equal to 1 plus minus iota so that two roots of two eigen values of this equations R 1 plus iota and 1 minus iota, so these are the eigen values.

So, let us find corresponding eigenvectors. So, first you find eigen vector eigen vector corresponding to lambda equals to say 1 minus iota. So, how you will find that you can simply see that A minus lambda I times X must be equal to 0, this we have already seen that this X is a is the eigen vector correspond to this lambda.

So, this implies, now you substitute lambda as lambda as A 1 minus iota, so this is A minus 1 minus iota times iota times identity matrix into x equal to 0. So, this implies, now when you when you take 1 minus lambda here, so it is you can simply take lambda here as 1 minus iota, so it is iota 1, and it is 1 and it is again iota, it is minus 1 it is iota, and x is x 1, x 2 equal to 0 0.

So, this implies, now you can you can apply some elementary row operation this in this matrix. You simply first interchange these two rows, it is 1 iota. Iota minus 1 x 1 x 2

remain as it is, interchange these two also 0 0. Now, this implies you can make 0 here, with the help of this by applying the elementary row operation R_2 , this means R_2 minus i times R_1 .

When you apply this elementary row operations here, it is 1 i it is 0 it is again 0, and it is x_1 x_2 0 0. So, this implies x_1 plus i x_2 equal to 0. So, this give infinitely many solutions of x_1 and x_2 ok. You can substitute any value of x_1 , you can find corresponding value of x_2 such that x_1, x_2 is not equal to 0 of course.

So, 1 such value is 1 such x_1 is you can simply take you see x is what x is x_1, x_2 . You can take x_1 as minus i x_2 from here, and it is x_2 , so it is x_2 times minus i and 1, where x_2 is any real number.

So, so if you are talking about number of linearly independent eigen vector corresponding to λ equal to $1 - i$, so it is 1. So, you can pick any 1 linearly independent eigen vector from this, because because it gives infinitely many eigenvectors. You can pick out any one eigenvector, so we can said that vector is linearly independent eigen vector corresponding to λ equals to $1 - i$. So, we can say that corresponding to corresponding to λ equal to $1 - i$, the linearly independent eigen vector is say $-i \ 1$ transpose.

Now, similarly you will take λ is equal to $1 + i$, and we can you find out the corresponding linearly independent eigen vector on the same lines. Now, let us consider second problem. Here the matrix B is $\begin{bmatrix} 0 & 0 & 2 & 0 \\ 2 & 0 & 0 & 0 \end{bmatrix}$ and this matrix. It is given towards at the one of the eigen value of this matrix B is 4, then we have to find the remaining to eigenvalues, and also the corresponding eigenvectors of B .

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$$\begin{aligned}
 A &= \begin{pmatrix} 0 & 0 & 2 \\ 0 & 2 & 0 \\ 2 & 0 & 3 \end{pmatrix}_{3 \times 3} & \lambda_1 = 4, \lambda_2, \lambda_3 \\
 & & \lambda_1 + \lambda_2 + \lambda_3 = 0 + 2 + 3 = 5 \\
 & & \Rightarrow \lambda_2 + \lambda_3 = 1 \\
 & & \lambda_1 \lambda_2 \lambda_3 = |A| = -8 \\
 & & \Rightarrow \lambda_2 \lambda_3 = -2 \\
 & & \lambda_2 = 2, \lambda_3 = -1 \\
 & & \lambda_1 = 4, \lambda_2 = 2, \lambda_3 = -1 \\
 \text{E-vector corresp to } \lambda = 4. & \\
 (A - \lambda I)X = 0 & \Rightarrow (A - 4I)X = 0 \\
 \Rightarrow \begin{pmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 2 & 0 & -1 \end{pmatrix} X = 0 & \Rightarrow \begin{pmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 0 & 0 & 0 \end{pmatrix} X = 0 \quad R_3 \rightarrow R_3 + \frac{1}{2}R_1 \\
 & \Rightarrow \begin{aligned} -4x_1 + 0x_2 + 2x_3 &= 0, \\ -2x_2 &= 0 \end{aligned} \quad X = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \\
 & \Rightarrow x_2 = 0, \quad x_3 = 2x_1 \\
 X &= \begin{pmatrix} x_1 \\ 0 \\ 2x_1 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}
 \end{aligned}$$

So, let us discuss this problem now. Here the matrix A is simply 0 0 2 0 2 0 and 2 0 3. Now, one lambda is 4, it is given to us say lambda 1 is 4. Now, we know that the trace is equal to the sum of the eigen values. So, let us suppose the other two eigen values are lambda 2, lambda 3, it is of order 3 cross 3, so it will be having three eigen values. So, some of the eigen values is equal to trace of the matrix that is 5 0 plus 2 plus 3. Now, lambda 1 is 4, so this implies lambda 2 plus lambda 3 is 1.

Now, the product of the eigenvalues, we know that product of eigen value is simply determinant of A. Now, the determinant of A, simply when you when you simply open this matrix, I mean determinant, then you simply get it minus 8. Now it is 4, so this implies lambda 2, lambda 3 will be minus 2. So, solving these two equations, we can easily find lambda 2 and lambda 3. So, it is clearly as we as we are seeing it is lambda 2 is 2, and lambda 3 is minus 1. So, sum is 1 and the product is minus 2. So, the other two eigen values are 2 n minus 1.

So, now we can say that this matrix are the eigen values 4, 2, and minus 1. Now, we have to find out the corresponding eigenvectors. Let us suppose, we have to find out the eigen vector corresponding to lambda equal to 4. Similarly, we can find eigen vector cross point lambda equal to 2, and lambda equal to minus 1.

So, let us find eigen vector corresponding to lambda equal to 4. How you will find that this is A minus lambda I X equal to 0 again, so this implies A minus 4 I into X equal to 0.

So, this implies $\begin{bmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 2 & 0 & -1 \end{bmatrix} X = 0$, this is $0 \begin{bmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 2 & 0 & -1 \end{bmatrix} X = 0$, this is $2 \begin{bmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 2 & 0 & -1 \end{bmatrix} X = 0$, and X is X , which is equal to 0 , this implies.

Now, we will try to convert this into its echelon form. So, we will make 0 here with the help of this, this is already 0 , we will make 0 here with the help of this. So, which operation we will apply $R_3 - 2R_1$ plus half of R_2 . So, this is $\begin{bmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$, this is $0 \begin{bmatrix} -4 & 0 & 2 \\ 0 & -2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$, this plus half of this is against 0 . Now, this implies, this implies $-4x_1 + 2x_3 = 0$ plus 0 into x_2 plus 2 into x_3 equal to 0 . If you are taking X as x_1, x_2, x_3 , it substitute it here, and we multiplied these two matrices, then we simply get these equations, and $-2x_2 = 0$. So, this implies $x_2 = 0$, and $x_3 = 2x_1$.



So, what is the corresponding eigen vector, corresponding eigen vector will be you can say, you can see, it is $x_1 = 0$ and $2x_1$, so it is x_1 times $\begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$. So, again there are infinitely many eigenvectors corresponding to $\lambda = 4$, but if you are talking about number of linearly independent eigen vector correspond to $\lambda = 4$, so that is 1 . You can pick any one eigen vector from this space, so we can say that is a linearly independent eigen vector correspond to $\lambda = 4$. Similarly, we can find out eigen vector corresponding to $\lambda = 2$, and $\lambda = -1$.

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Properties

Let λ be an eigen value of A and x be its corresponding eigen vector, then

- αA has eigen value $\alpha\lambda$ and the corresponding eigen vector is x .
- A and A^T have the same eigen values, since a determinant can be extended by rows or columns.
- A^m has eigen value λ^m and the corresponding eigen vector is x for any positive integer m .
- A^{-1} (if exists) has the eigen value $\frac{1}{\lambda}$ and the corresponding eigen vector is x .
- $g(\lambda)$ is an eigen value of $g(A)$, where $g(\lambda)$ is a polynomial of λ and x is an eigen vector of $g(A)$ corresponding to the eigen value $g(\lambda)$.



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Now, if λ is an eigen value of matrix A , and x is a corresponding eigen vector, then we have the following properties. Now, it is very easy to see these properties, you see here.

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$$\left. \begin{array}{l}
 AX = \lambda X, \quad X \neq 0 \\
 (\alpha A)X = (\alpha \lambda)X, \quad \alpha \neq 0 \\
 |A - \lambda I| = 0 \\
 |(A - \lambda I)^T| = |(A - \lambda I)| = 0 \\
 \Rightarrow |A^T - \lambda I| = 0
 \end{array} \right\} \begin{array}{l}
 AX = \lambda X, \quad X \neq 0 \\
 A(AX) = A(\lambda X) \\
 = \lambda(A X) \\
 = \lambda(\lambda X) \\
 A^2 X = \lambda^2 X \\
 A^k X = \lambda^k X \\
 AX = \lambda X, \quad |A| \neq 0 \\
 A^{-1}(AX) = \lambda(A^{-1}X) \\
 \Rightarrow IX = \lambda(A^{-1}X) \\
 \Rightarrow A^{-1}X = \frac{1}{\lambda}X
 \end{array}$$

See if A has a eigen vector eigen value lambda and corresponding eigen vector is X, that means, A X equal to lambda X. Now, if you want to calculate the eigen, eigen values of alpha A, where alpha is a non-zero scalar. Then how we can find this, you simply multiplied these two by alpha, both sides by alpha. So, we can simply see that we can simply see that this alpha A has an eigen value alpha lambda, and the corresponding eigen vector is A corresponding eigen vector is X.

Now, if you want for A square, you see A X equal to lambda X, X not equal to 0. If you multiply both sides by A, it is A into X equals to A into lambda X, which is lambda times A X. And A X is lambda X, so it is lambda X. So, we can say that A square X is equals to lambda square X. So, what we can say that the if A as an eigen value lambda and the corresponding eigen vector is X, then A square as an eigen value lambda square and the corresponding eigen vector is X.

Similarly, similarly we can say that A raised to power K is equals to lambda raised to power K into X, you can simply find similarly find A cube, A raised to power 4 and so on. So, what we have concluded, we have concluded that if A as an eigen value lambda and the corresponding eigen vector is X, then A raised to power K has eigen value lambda raised to power K, and the corresponding eigen vector is X.

Now, next property is A and A transpose, both have same eigen values that is very easy to show, you can see that eigen values of or given by this expression. Now, A minus A this

matrix, the transpose of this matrix is equals to A minus the determinant of this is same as determinant of this, because by changing interchanging rows and columns will not change the value of the determinant. This is equal to 0, it is given to us, and this implies A transpose minus lambda I whole determinant is equal to 0 so that lambda is not changing here, whatever lambda we are we are having here, same lambda we are having here, that means, the eigen values of A and A transpose are same.



The next is the next is if you say that $AX = \lambda X$, and suppose determinant of A is not equal to 0, that means, A is invertible. Then you can take A inverse both the sides, then a inverse of A X will be equals to lambda times A inverse of X. And this implies, identity into x equals to lambda into A inverse of X. And this implies, A inverse of X will be $\frac{1}{\lambda} X$, this is this is this is one or identity here. So, we can say that if A as an eigen value lambda and the corresponding eigen vector is X, then A inverse as an eigen value $\frac{1}{\lambda}$ and the corresponding eigen vector is X.

So, these are few of the properties of eigenvalues, which is stated here. And if $g(\lambda)$ is an eigen value of $g(A)$, where $g(\lambda)$ is a polynomial of lambda, where x and x is an eigen vector of $g(A)$ corresponding to eigen value $g(\lambda)$. That that is very easy to show again, because the third property itself is states that if λ is a polynomial of lambda, then $g(\lambda)$ is an eigen value of $g(A)$.

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Problems

- ❶ If A is a singular matrix of order 2 having trace 3, then find the eigen values of $(A^T)^2$.
- ❷ Let A be 4×4 matrix with eigen values 1, -1, 2, -2. Find the value of the determinant of the matrix $B = 2A + A^{-1} - I$.
- ❸ If A is a matrix of order n such that $A^k = 0$ (where $k \in \mathbb{N}$). Find all eigen values of A.



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Now, let us let us discuss these few problems quickly. You see, if A is a singular matrix of order 2 and having trace 3. If it is singular matrix means, determinant is equal to 0, determinant is nothing but product of eigenvalues, product of eigen values equal to 0 means at least one of the eigen value is 0. Now, it is order 2, that means, it has it is having only two eigenvalues, say lambda 1, and lambda 2.

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$$\begin{array}{l}
 A_{2 \times 2} \quad \lambda_1, \lambda_2 \quad \lambda_1 \lambda_2 = 0, \quad \lambda_1 + \lambda_2 = 3 \\
 \lambda_1 = 0 \quad 0 + \lambda_2 = 3 \\
 \lambda_2 = 3 \\
 \\
 A \rightarrow 0, 3 \\
 A^T \rightarrow 0, 3 \\
 (A^T)^2 \rightarrow 0, 9 \\
 \\
 A \rightarrow \begin{array}{cccc} 1 & -1 & 2 & -2 \\ \downarrow & \downarrow & \downarrow & \downarrow \\ B \rightarrow 2 & -4 & \frac{7}{2} & -\frac{11}{2} \end{array} \quad B = 2A + A^{-1} - I \\
 A \rightarrow \lambda \\
 B \rightarrow 2\lambda + \frac{1}{\lambda} - 1 \quad \rightarrow \begin{array}{l} -4 - \frac{1}{2} - 1 \\ -2 - 1 - 1 \\ 4 + \frac{1}{2} - 1 = 4 - \frac{1}{2} \end{array}
 \end{array}$$

So, so we can say A is a order 2 cross 2 having two eigenvalues, lambda 1, and lambda 2. Then product of eigen values is equal to 0, because matrices singular, and trace is 3, trace is 3. So, here from here, we can say that one of the eigen value is 0 say lambda 1 is 0. If lambda 1 is 0, so lambda 2 will be 3. So, the eigen values are 0 and 3 of A; eigen value of A transpose will be again 0 and 3, and eigen value of A transpose square will be 0 and 9 ok.

The next one is if A is a 4 cross 4 cross 4 matrix of eigen values 1 minus 1 2 minus 2. Then the eigen values of this how we can find, you see eigen values of A are 1 minus 1 2 minus 2, and B is given as it is 2 A it is 2 A plus A inverse minus I. So, it is some polynomial in I mean, it is some polynomial in I mean B, I mean A. So, the eigen value of B will be nothing but if A as an eigen value lambda, then the eigen value of B will be nothing but 2 lambda plus 1 by lambda minus 1 2 has 2 lambda eigen value A inverse as 1 by lambda eigen value and this eigen value is 1.

So, simply, simply you can substitute here, cross point to 1, where having 2 plus 1 plus 1 minus 1 is 2 for B. Cross point to minus 1 if it substitute minus 1, it is minus 1 minus 1 that is minus 4. Cross point to 2, it is 1 plus 1 by 2 minus 1 that is that is 4 minus 1 by 2 that is 7 by 2. Cross point to minus 2, if I talking about minus 2, because minus 2, it is minus 4 minus 1 by 2 minus 1 that is minus 4 minus 3 by 2 that is minus 11 by 2. So, these are the eigen values corresponding to B.

So, the third problem is simple, if A is A matrix of order n, and such that A raised to power k is 0, where k belongs to a natural number. Then all this eigen values are 0, because if one of because if any of the eigen value is not equal to 0 of A, then A raised to power k will not be 0 ok. So, therefore, all the eigen values of a must be 0.

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Theorem
 Let A be a square matrix of order n having k distinct eigen values $\lambda_1, \lambda_2, \dots, \lambda_k$. Let v_i be an eigen vector corresponding to the eigen value $\lambda_i, i = 1, 2, \dots, k$. Then the set $\{v_1, v_2, \dots, v_k\}$ is LI.

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So, this is a result that if A is A square matrix of order n having k distinct eigen values lambda 1 up to lambda k. Let v i be the eigen vector corresponding to the eigen value lambda i, i from 1 to k. Then the set this is linearly independent.

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$$\begin{aligned}
 & \lambda_1, \lambda_2, \dots, \lambda_k, \quad \lambda_i \neq \lambda_j, \quad i \neq j \\
 S = & \{v_1, v_2, \dots, v_k\} \rightarrow \text{L.I.} \\
 & v_i \\
 A v_i = & \lambda_i v_i \quad \forall i
 \end{aligned}$$

$$\begin{aligned}
 & \{\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_{r+1} v_{r+1}\} \\
 & \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_{r+1} v_{r+1} = 0 \quad (1) \\
 \Rightarrow & A \alpha_1 v_1 + A \alpha_2 v_2 + \dots + A \alpha_{r+1} v_{r+1} = 0 \\
 & \alpha_1 \lambda_1 v_1 + \alpha_2 \lambda_2 v_2 + \dots + \alpha_{r+1} \lambda_{r+1} v_{r+1} \\
 & = 0 \quad (2) \\
 \lambda_{r+1} \times & (1) - (2) \Rightarrow \\
 & \alpha_1 (\lambda_{r+1} - \lambda_1) v_1 + \dots + \alpha_r (\lambda_{r+1} - \lambda_r) v_r = 0 \\
 \Rightarrow & \alpha_i (\lambda_{r+1} - \lambda_i) = 0 \quad \forall i \\
 \Rightarrow & \alpha_i = 0 \quad \forall i, \quad i=1, 2, \dots, r \\
 \Rightarrow & \alpha_{r+1} v_{r+1} = 0 \\
 v_{r+1} \neq & 0 \Rightarrow \alpha_{r+1} = 0
 \end{aligned}$$

So, let us try to prove this result. You see, lambda 1, lambda 2 up to lambda k are distinct eigen values, that means, that means, lambda i is not equal to lambda j for i not equal to j number 1. Now, the set we are considering this v 1, v 2 up to v k, we have to showed at this set is linearly independent. v 1 is a eigen vector linearly independent eigen vector correspond to lambda 1, v 2 is a linearly independent eigen vector correspond to lambda 2 and so on. So, we will prove this by the method of induction ok.

Let us for n equal to 1 for k equal to 1, we are having v 1, and we v 1 a single set is I mean, it is it is linearly independent. I mean singleton set is always linearly independent, singleton non-zero vector, and v 1 is a non-zero vector is always linearly independent. Now, we will assume that it is true for k equal to r I mean this result hold for k equal to r, that means, the set of k r vectors are linearly independent. And we will try to showed that this also holds for k equal to r plus 1.

So, take the linear combination of these vectors now, to in order to showed that these are linearly independent. Now, you multiply both sides by matrix A, so we can simply take A into alpha 1 v 1 plus A into alpha 2 v 2 and so on A into alpha r plus 1 v r plus 1 equal to 0.

Now, A into v i is lambda i into v i for all i, because lambda i is the eigen value and the corresponding eigen vector is v i. So, we can simply write lambda 1 A v 1 is a lambda 1 v 1, similarly alpha 2 lambda 2 v 2 and so on up to alpha r plus 1 lambda r plus 1 v r plus

1. Now, say this equation 1, and this is equation 2. You can multiply 1 by λ^{r+1} into 1 and subtract with 2, so what we will obtain, we will obtain $\lambda \alpha_1$ into $\lambda^{r+1} v_1$ minus λv_1 and so on up to $\alpha_r \lambda^{r+1} v_r$ minus $\lambda^r v_r$ equal to 0.

Now, we have already assumed that a set of r vectors are linearly independent. So, this is some linear combination of this linearly independent vector, which is equal to 0. So, this implies, α_i is $\lambda^{r+1} v_i$ minus λv_i equal to 0 for all i . But, eigen values are distinct, so this is not equal to 0. So, this implies, α_i equal to 0 for all i for all i means for i from 1 to r .

And when you substitute α_i from 1 to r here with all 0, then you will get you will get $\lambda^{r+1} v_r$ minus λv_r equal to 0. And since, v_r is not equal to 0 is eigenvector, so this implies, λ^{r+1} equal to λ . So, we have shown that all α_i 's from i from 1 to r plus 1 equal to 0 that mean, this set of vectors are linearly independent. So, in this lecture, we have seen that that what eigen values and eigenvectors are, and some of the important properties of eigen values and eigenvectors.

So, thank you.