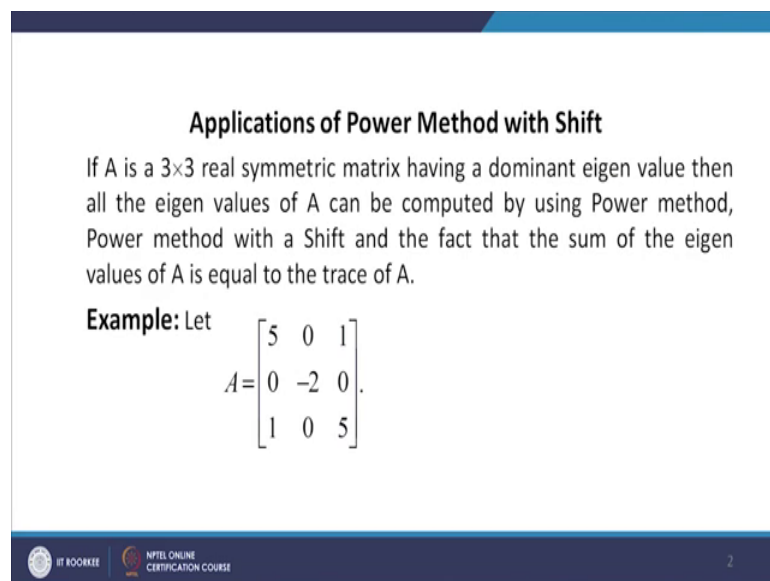


Numerical Linear Algebra
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Lecture – 58
Applications of Power Method with Shift

Hello friends. So, I welcome you to this lecture on applications of power method with shift.

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Applications of Power Method with Shift

If A is a 3×3 real symmetric matrix having a dominant eigen value then all the eigen values of A can be computed by using Power method, Power method with a Shift and the fact that the sum of the eigen values of A is equal to the trace of A.

Example: Let

$$A = \begin{bmatrix} 5 & 0 & 1 \\ 0 & -2 & 0 \\ 1 & 0 & 5 \end{bmatrix}$$

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If A is 3 by 3 real symmetric matrix having a dominant Eigenvalue then all the Eigenvalues of A can be computed by using power method, power method be the shift and the fact that this sum of the Eigen values of a square matrix equals the trace of A. Trace of A means sum of the diagonal entries of A.

So, when the matrix A is given, we can easily find the sum of the diagonal entries of the matrix like here, we have in this example A equal to 5 0 1 0 minus 2 0 1 0 5. So, the diagonal entries are 5 minus 2 5 and therefore, their total is 8 ok.

So, we can say that the Eigenvalues of A will have sum equal to 8 ok. So, sum of the Eigenvalues of A is equal to 8. So, by if is A is 3 by 3 real symmetric matrix having a dominant Eigenvalue, then we will be able to compute all the Eigenvalues. see as such Eigen dominant Eigenvalue can be computed by using the power method. So, by using

the power method, we will find out the dominant Eigenvalue of A, then by power method with a shift we shall be able to find the sub dominant Eigenvalue of A, and then the third Eigenvalue can be found by using the fact that the sum of the Eigenvalues of A is equal to the trace of A.

So, let us take this example.

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First, we shall use the Power method to find the dominant eigen value.

Let $y_0 = [1 \ 1 \ 1]^T$.

k=1: $z_1 = Ay_0 = [6 \ -2 \ 6]^T$, $\max(z_1) = 6$
 $y_1 = [1.0000 \ -0.3333 \ 1.0000]^T$.

k=2: $z_2 = Ay_1 = [6 \ 0.6666 \ 6]^T$, $\max(z_2) = 6$
 $y_2 = [1 \ 0.1111 \ 1]^T$.

k=3: $z_3 = Ay_2 = [6 \ -0.2222 \ 6]^T$, $\max(z_3) = 6$
 $y_3 = [1 \ -0.0370 \ 1]^T$.

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So, first we shall use the power method to find the dominant Eigenvalue. So, y is equal to $[1 \ 1 \ 1]^T$ let us take this $y = [1 \ 1 \ 1]^T$ vector, then in the first iteration we will get $z = [6 \ -2 \ 6]^T$ is equal to Ay .

So, z_1 comes out to be the column vector $[6 \ -2 \ 6]^T$, and the maximum value of z_1 is equal to 6 here as you can see. So, y_1 is equal to z_1 divided by maximum of z_1 . So, one we divide z_1 by maximum of z_1 , we get y_1 equal to $[1.0000 \ -0.3333 \ 1.0000]^T$.

In the second iteration k equal to 2, z_2 is equal to Ay_1 and so, Ay_1 is equal to $[6 \ 0.6666 \ 6]^T$ and this column vector. So, maximum of z_2 is equal to 6 obvious; obviously. So, y_2 is z_2 divided by maximum of z_2 gives you $[1 \ 0.1111 \ 1]^T$.

And in the third iteration what do we get? z_3 equal to Ay_2 . z_3 equal to Ay_2 means when we multiply the matrix A by y_2 we get $[6 \ -0.2222 \ 6]^T$ and maximum of z_3 is equal to 6. then y_3 comes out to be $[1 \ -0.0370 \ 1]^T$.

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k=4: $z_4 = Ay_3 = [6 \ 0.0740 \ 6]^T$, $\max(z_4) = 6$
 $y_4 = [1 \ 0.0123 \ 1]^T$.

Proceeding like this, we get

k=10: $z_{10} = [6 \ 0.0002 \ 6]^T$, $\max(z_{10}) = 6$
 $y_{10} = [1 \ 0.0000 \ 1]^T$.

Hence, the dominant eigen value of A is $\lambda_1 = 6$ and the dominant eigen vector $v_1 = [1 \ 0 \ 1]^T$.

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K equal to 4 when we carry out the fourth iteration, we get z_4 equal to $A y_3$ and we get $[6 \ 0.0740 \ 6]^T$ transpose. Maximum of z_4 comes out to be 6. y_4 is equal to $[1 \ 0.0123 \ 1]^T$ transpose. When we proceed in this manner what do we get in the tenth iteration? Tenth iteration gives you z_{10} equal to $[6 \ 0.0002 \ 6]^T$ and so, maximum of z_{10} equal to 6 while y_{10} is equal to $[1 \ 0.0000 \ 1]^T$. So, what do we get? We can say that the maximum of z_k . Maximum of z_k as k goes to infinity tends to λ_1 , the dominant Eigen value of A. So, maximum of z_{10} here is equal to 6. So, when k is sufficiently large we see that λ_1 is equal to 6.

The dominant Eigenvalue is equal to 6, and the dominant Eigenvector v_1 is equal to approximately y_{10} ; y_{10} is equal to $[1 \ 0 \ 1]^T$. So, the dominant Eigenvector v_1 is given by this y_{10} which is $[1 \ 0 \ 1]^T$.

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Now, we define $B = A - \lambda_1 I = A - 6I$.

Then
$$B = \begin{bmatrix} -1 & 0 & 1 \\ 0 & -8 & 0 \\ 1 & 0 & -1 \end{bmatrix}.$$

Again, let $y_0 = [1 \ 1 \ 1]^T$.

k=1: $z_1 = By_0 = [0 \ -8 \ 0]^T$, $\max(z_1) = -8$
and hence $y_1 = [0 \ 1 \ 0]^T$.

k=2: $z_2 = By_1 = [0 \ -8 \ 0]^T$, $\max(z_2) = -8$
hence $y_2 = [0 \ 1 \ 0]^T$.

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Now, let us define the. So, we have got the dominant Eigen value of the matrix A, it is equal to 6. Let us get another Eigenvalue of the matrix A by using power method with the shift. So, let us use B equal to, now here the rho we choose as equal to lambda 1. The lambda 1 value we have got that is 6.

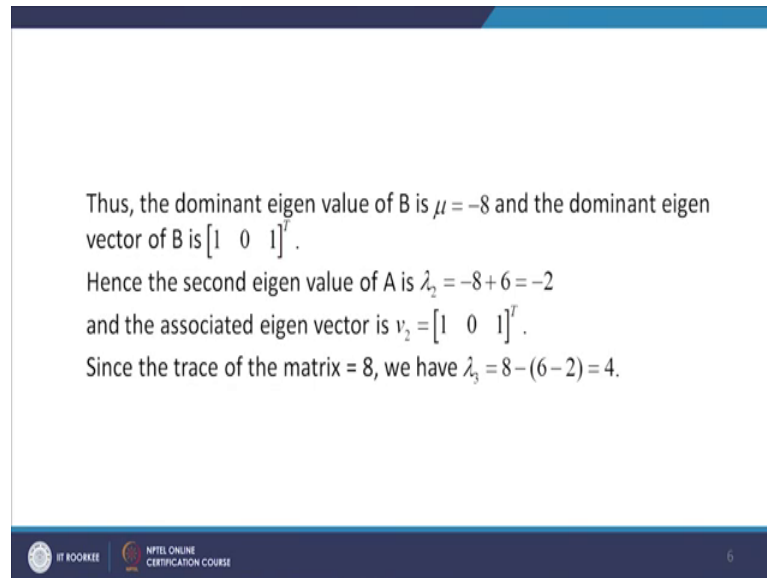
So, we choose that as the value of rho. So, let us define B equal to a minus lambda 1 into I. So, we get A minus 6 I and A minus 6 I means from the diagonal matrix from the diagonal entries of the matrix A. We subtract 6, so we get the matrix Bs minus 1 0 1 0 minus 8 0 and 1 0 minus 1.

Again, we start with y naught equal to 111, we in the first iteration we get z 1 equal to be y naught equal to 0 minus 8 0 transpose. So, maximum of z naught z 1 maximum z 1 comes out to be minus 8 and hence by 1 which is equal to z 1 over maximum of z 1 comes out to be 010 transpose. In the second iteration z 2 is equal to By 1 z 2 equal to B y gives you 0 minus 8 0 transpose the same Eigenvector as z 1.

You can see same vector as z 1 z 2 is same vector as z 1. So, maximum z 2 2 is also minus 8 and. So, y 2 is also same as y 1 ok. So, y 2 is 010 transpose.

And since we are not getting ah any movement further, we are getting the same vector and in the second iteration. So, we stop here and we say that the dominant Eigenvalue of B which is dominant Eigenvalue of B is nu equal to minus 8.

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Thus, the dominant eigen value of B is $\mu = -8$ and the dominant eigen vector of B is $[1 \ 0 \ 1]^T$.

Hence the second eigen value of A is $\lambda_2 = -8 + 6 = -2$ and the associated eigen vector is $v_2 = [1 \ 0 \ 1]^T$.

Since the trace of the matrix = 8, we have $\lambda_3 = 8 - (6 - 2) = 4$.

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Since the, since when we write B as a minus 6 I, the Eigenvalue of comes out to be minus 8 ; that means, to get the Eigenvalue of A we add 6 to this Eigenvalue of B. So; that means, the Eigenvalue of A comes out to be minus 8 plus 6, which is equal to minus 2. So, we have got the dominant Eigenvalue which was lambda 1 equal to 6 and then we have got next Eigenvalue, second Eigenvalue, rather second Eigenvalue of A as minus 2 we have got two Eigenvalues. third Eigenvalue is obtained by using the fact that some of the Eigenvalues of the matrix is equal to its trace and trace was equal to 8 ok.

So, lambda 3 is equal to, are 8 minus some of the Eigenvalues. one Eigenvalue is 6 the Eigenvalue is minus 2. So, 8 minus 6 minus 2 which is equal to 4. So, we have got all the Eigenvalues of the matrix A by using the power method.

We found the dominant Eigenvalue of A by using power method with A shift, where we use the value of rho as the value of the dominant Eigenvalue; that is lambda 1. We get second Eigenvalue of A and the third Eigenvalue of A is found by using the fact that some of Eigenvalues of A equals the trace of A.

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Example: Let


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

We shall find all the eigen values of A.

Let $y_0 = [1 \ 1 \ 1]^T$

k=1: $z_1 = [4 \ -3 \ 1]^T$, $\max(z_1) = 4$
and so $y_1 = [1.0000 \ -0.7500 \ 0.2500]^T$

k=2: $z_2 = [13.0000 \ -12.2500 \ 5.7500]^T$, $\max(z_2) = 13.0000$
hence $y_2 = [1.0000 \ -0.9423 \ 0.4423]^T$



Now, let us take one more example on this applications. So, let A be equal to 8 minus 6 2 minus 6 7 minus 4 2 minus 4 3. So, you can see here, what is the trace 8 plus 7; that is 15 plus 3 that is 18. So, some of the Eigen values of A must be equal to 18.

So, we will find one Eigenvalue by using power method which is the dominant Eigenvalue of A, second Eigenvalue of A will be found by using the power method with shift and the third Eigen value will be found by using the fact that the, some of the Eigenvalues must be equal to trace that is 18.

So, let us first find the dominant Eigenvalue by using power method. So, y naught we start take as 111 transpose. In the first iteration we get z 1 equal to Ay naught Ay naught is equal to 4 minus 3 1 transpose and then maximum of z 1 you can see here is 4 and. So, y 1 is equal to z 1 divided by maximum of z 1.

And therefore, we get 1.0000 minus 0.7 500 0. 2500. In the second iteration z 2 is maximum, z 2 is equal to A into y 1 A when multiplied, when we multiply the matrix A by the vector by 1 we get z 2; that is 13. 0000 minus 12.2500, 5.7500 and maximum value of z 2.

You can see here is 13.0000 and y 2 then is equal to z 2 divided by maximum of z 2. So, we get 1.0000 minus 0.9423 0.4423 transpose k equal to 3. In the third iteration z 3 is

equal to Ay_2 ok. So, we get 14.5385 minus 14.3654 7.0962 transpose and maximum of z_3 is 14.5385 y_3 is z_3 is divided by maximum of z_3 .

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$k=3: z_3 = [14.5385 \quad -14.3654 \quad 7.0962]^T, \max(z_3) = 14.5385$
 and $y_3 = [1.0000 \quad -0.9881 \quad 0.4881]^T$.
 Proceeding similarly, we get
 $k=8: z_8 = [14.9999 \quad -14.9998 \quad 7.4999]^T, \max(z_8) = 14.9999$
 and $y_8 = [1.0000 \quad -1.0000 \quad 0.5000]^T$.
 Hence $\lambda_1 = 15$ and $v_1 = [1 \quad -1 \quad 0.5]^T$.

So, we get 1.0000 minus 0.9881 and 0.4881 transpose, proceeding like this we get in the k th in the eighth iteration k equal to 8 z_8 equal to 14.9999 minus 14.9998 7.4999 transpose and so, maximum of z_8 equal to 14.9999 and y_8 . y_8 which is equal to z_8 divided by maximum of z_8 is equal to 1.0000 minus 1.0000 and 0.5000 . So, we can say that maximum of z_k , maximum of z_k Ah goes to the dominant Eigenvalue as k goes to infinity.

So, we can say that 14.9999 is very close to 15 . We can take λ_1 as 15 and the corresponding Eigenvector is y_8 ; that is 1 minus 1 and 0.5 . So, we denote that Eigenvector of the matrix A v_1 .

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Now, let

$$B = A - \lambda_1 I = A - 15I = \begin{bmatrix} -7 & -6 & 2 \\ -6 & -8 & -4 \\ 2 & -4 & -12 \end{bmatrix}$$

Let $y_0 = [1 \ 1 \ 1]^T$.

k=1: $z_1 = [-11 \ -18 \ -14]^T$, $\max(z_1) = -18$
 $y_1 = [0.6111 \ 1.0000 \ 0.7778]^T$

k=2: $z_2 = [-8.7222 \ -14.7778 \ -12.1111]^T$, $\max(z_2) = -14.7778$
and $y_2 = [0.5902 \ 1.0000 \ 0.8195]^T$.

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So, one Eigenvalue we have found it is 15. And now, find next second Eigen value of A by using power method with shift. So, we write the matrix B equal to A minus lambda 1 I rho, we have taking as lambda 1 and lambda 1 we have found to be 15.

So, A minus 15 I, so we subtract the diagonal entries of A by 15 and B arrive at the matrix B which is minus 7 minus 6 2 minus 6 minus 8 minus 4 2 minus 4 minus 10 and then we again start with initial vector y naught equal to 111, 111.

In the first iteration k equal to 1 z 1 is equal to Ay naught, gives us minus 11 minus 18 minus 14 transpose maximum of z 1 therefore, is equal to minus 18. So, y 1 is equal to z 1 divided by maximum of z 1 and therefore, we get 0.6111 1.0000 0.77 78 transpose, sowe get y 1.

Now in the second iteration by k equal to 2 z 2 equal to Ay 1 gives us minus 8.7 22 2 minus 14.77 78 minus 12.1111 transpose and. So, maximum of z 2 is minus 14.778 and y 2 which is equal to z 2 over maximum of z 2 gives us 0.59 02 1.0000 0.8195 transpose.

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Proceeding like this, we get
 $k=23$: $z_{23} = [-7.5152 \quad -14.9973 \quad -14.9640]^T$, $\max(z_{23}) = -14.9973$
and $y_{23} = [0.5011 \quad 1.0000 \quad 0.9978]^T$.
Hence, the dominant eigen value of B is -15 and the dominant eigen vector is $[0.5 \quad 1.0 \quad 1.0]^T$.
Thus, the second eigen value of A = -15 + 15 = 0 and the corresponding eigen vector is $[0.5 \quad 1 \quad 1]^T$.
Further, Trace(A) = 18
Hence, third eigen value of A = 3.

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Proceeding like this in the 23 iteration k equal to 23 z_{23} is equal to minus 7; 51 52 minus 14.99e 73 minus 14. 96, 40 transpose. So, maximum of z_{23} is minus 14. 99 73.

And y_{23} is 0; 50 11 1.00 00 0.99 78 transpose. Hence the dominant Eigenvalue of B is minus 15, because this z_{23} is maximum of z_{23} minus 14.9973 is very close to minus 15 and the dominant Eigenvector is 0.5. This is very near to 0.5 and this near this is 1 and this one near to 1. So, we can write 0.5 11 transpose.

So, second Eigen value of A will be found by adding lambda 1 to this Eigenvalue, because we took rho as lambda 1, so we. So, second Eigenvalue of A is this value minus 15 plus 15 and, so we get 0. So, second Eigenvalue is 0.

Now the corresponding Eigenvector we have seen it is 0.5 11.0. Now trace of A is equal to 18. So, one Eigenvalue of we found as 15 the other Eigenvalue we found as 0. So, the third Eigenvalue is; obviously, 3. So, third Eigen value of A is found by the, with the help of trace of A.

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Least Dominant Eigen-pair: Let A be a real $n \times n$ matrix with eigen values $\lambda_1, \lambda_2, \dots, \lambda_n$. The eigen value λ_n of A is called the least dominant eigen value of A if the eigen values of A can be ordered so that

$$|\lambda_1| \geq |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_{n-1}| > |\lambda_n|.$$

Clearly, λ_n is a real eigen value of A with multiplicity one. The associated eigen vector x_n of A is called the least dominant eigen vector of A and the pair (λ_n, x_n) is called the least dominant eigenpair of A .

Suppose that A is also invertible then if $\lambda_1, \lambda_2, \dots, \lambda_n$ are the eigen values of A then $\lambda_1^{-1}, \lambda_2^{-1}, \dots, \lambda_n^{-1}$ are called the eigen values of A^{-1} .

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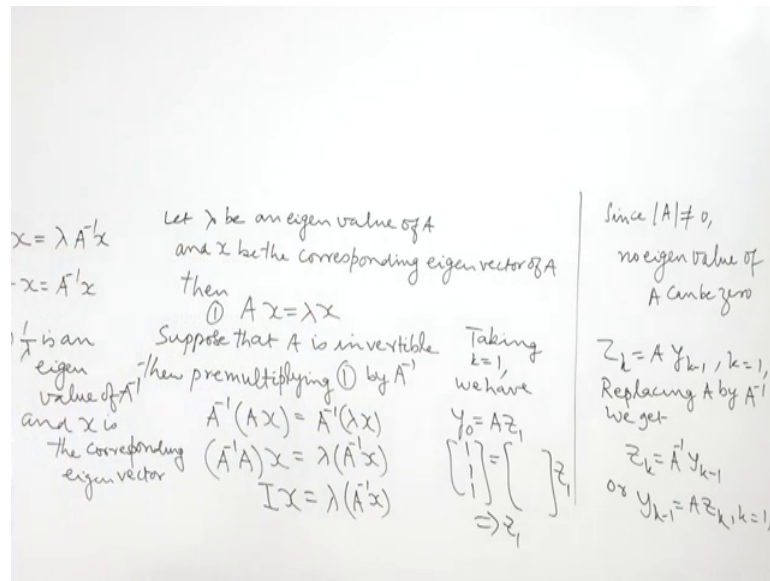
Now, we come to least dominant Eigen, but we seen how to find the dominant Eigen pair of A real 3×3 real $n \times n$ diagonalizable matrix. Now let us see how we can find the least dominant Eigen pair.

So, let A be a real $n \times n$ matrix with the Eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$. The Eigenvalue λ_n of A is called the least dominant, least dominant Eigenvalue of A . if the Eigenvalues of A can be ordered, so that $|\lambda_1| \geq |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_{n-1}| > |\lambda_n|$. So, $|\lambda_n|$ is the least dominant Eigenvalue and again it is 1 algebraic multiplicity is 1 and it must be real.

So, clearly λ_n is a real Eigen of A . this multiplicity 1, the reasoning is the same as in the case of dominant Eigen value of the matrix A . Now the associated Eigenvector x_n of A is called the least dominant Eigenvector of A and the pair (λ_n, x_n) is called the least dominant Eigen pair of A .

Now, suppose that if the matrix A is also invertible. Let us assume that the matrix A is also invertible then if $\lambda_1, \lambda_2, \dots, \lambda_n$ of the Eigenvalues of A , then $1/\lambda_1, 1/\lambda_2, \dots, 1/\lambda_n$ are the Eigen values of A^{-1} . So, you can see it is very easily.

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Let us say, let lambda be an Eigenvalue of A and x be the corresponding Eigenvector of A, then we have the matrix equation, then we have the matrix equation Ax equal to lambda x.

Now, suppose that A is invertible. Invertible means it is, its inverse exists or it is a non singular matrix, then we can pre multiply this equation 1, then pre multiplying equation 1 by A inverse what do we get?

Since matrix multiplication is associative, I can write it as lambda is a scalar, I can write it as lambda times A inverse x. So, A inverse A is the identity matrix. So, we have identity into x and from here, what do we get? Identity into x is equal to x. So, x equal to lambda time A inverse x ok.

Now, A is invertible matrix. Since A is invertible determinant of A is nonzero, since determinant of A is nonzero, no Eigenvalue of A can be 0 Eigen value, no Eigenvalue of A can be 0. So, if lambda is an Eigenvalue lambda cannot be 0, I can divide by lambda ok. So, I get 1 by lambda into x equal to A inverse x.

So, from here we, if you compare this equation with this equation we see that 1 by lambda is an Eigenvalue of A inverse and x is the corresponding Eigenvector. So, if lambda is an Eigenvalue of A and x is the corresponding Eigenvector, then 1 by lambda is an Eigenvalue of A inverse and x is the corresponding Eigen vector.

So, this implies $1/\lambda$ is an Eigenvalue of A^{-1} . So, if the matrix is invertible and $\lambda_1, \lambda_2, \dots, \lambda_n$ are its Eigenvalues. Then $1/\lambda_1, 1/\lambda_2, \dots, 1/\lambda_n$ are the Eigenvalues of A^{-1} .

So, what we will do? We will determine the dominant Eigenvalue of A^{-1} . Dominant Eigenvalue of A^{-1} will give us the least dominant Eigenvalue of A , because if the least dominant Eigenvalue of A is λ_n , then $1/\lambda_n$ will become the dominant Eigenvalue of A^{-1} .

So, we will determine the dominant Eigen value of A^{-1} ; that is we will determine $1/\lambda_n$ and then from that we will find the Eigenvalue of least dominant Eigenvalue of A ; that is λ_n . So, let us, let. So, hence if λ_n is the least dominant Eigen value of A then $1/\lambda_n$ is the dominant Eigenvalue of A^{-1} with the associated Eigenvector x_n .

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Hence if λ_n is the least dominant eigen value of A then λ_n^{-1} is the dominant eigen value of A^{-1} with the associated eigen vector x_n . Thus, by using power method algorithm to A^{-1} we compute the eigenpair (λ_n^{-1}, x_n) . Hence, we easily obtain (λ_n, x_n) .

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So, by using power method algorithm to A^{-1} we can compute the Eigen pair $1/\lambda_n$ to the power minus 1 x_n and then from that we can easily obtain $\lambda_n x_n$ ok.

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Theorem(Power method for least dominant eigenpair):

Let A be a real, diagonalizable, invertible $n \times n$ matrix with a least dominant eigen value. Further, let the eigen values $\lambda_1, \lambda_2, \dots, \lambda_n$ of A be ordered so that

$$|\lambda_1| \geq |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_{n-1}| > |\lambda_n|.$$

Suppose $y_0 \in \mathbb{R}^n$ is any vector (initial guess) such that if

$$y_0 = \alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_n x_n$$

then $\alpha_1 \neq 0$, x_1, x_2, \dots, x_n being the eigen vectors of A corresponding to the eigen values $\lambda_1, \lambda_2, \dots, \lambda_n$ of A and x_n is real since λ_n is real.

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So, let us now see the power method for the least dominant Eigen pair. Let A be a real diagonalizable invertible n by n matrix with a least dominant Eigenvalue. Further let the Eigenvalue λ_1 , Eigenvalues $\lambda_1, \lambda_2, \lambda_n$ of A be ordered.

So, that mod of λ_1 is greater than or equal to mod of λ_2 , greater than or equal to mod of λ_3 and so on. Greater than or equal to mod of λ_{n-1} and greater than mod of λ_n and suppose we take the vector B naught in \mathbb{R} to the power n as $\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_n x_n$.

So, an $\alpha_n \times n$, then α_1 is not equal to 0. So, we write the linear combination y naught as a linear combination of x_1, x_2, \dots, x_n in such a way that α_1 is not equal to 0. x_1, x_2, \dots, x_n are the Eigen vectors of A corresponding to the Eigenvalue $\lambda_1, \lambda_2, \lambda_n$ of A and x_n is real. Since λ_n is real you know that λ_n is the least dominant Eigenvalue of A . So, its algebraic multiplicity is 1 and it is real.

So, as the same reasoning with we, which we gave in the case of dominant Eigenvalue that if λ_1 is real and A is real then x_1 is real so, here also when λ_1, λ_n is real and A is real matrix x_n is real. So, x_n is real Eigenvector, because λ_n is real and A matrix is real.

So, now, here the power method algorithm, we shall use for the matrix A inverse. So, what do we do? We write z_k equal to y_{k-1} in the power method algorithm will

replace in the power method algorithm what we had? Z^k was equal to $A^{k-1} z^1$ equal to 123 and so on.

For the power method algorithm for the matrix A we had z^k , will given by this, but now A will be replaced by A^{-1} ok. So, replacing A by A^{-1} . We get z^k equal to and which gives you or y^k equal to $a z^k$ equal to 123 and so on.

Now we have y naught with us. So, take k equal to 1 . Taking k equal to 1 we have y naught equal to $A z^1$. So, we solve this equation y naught is equal to 111 . We choose the initial approximation as 111 . The matrix A is known to us. The matrix A is known to us.

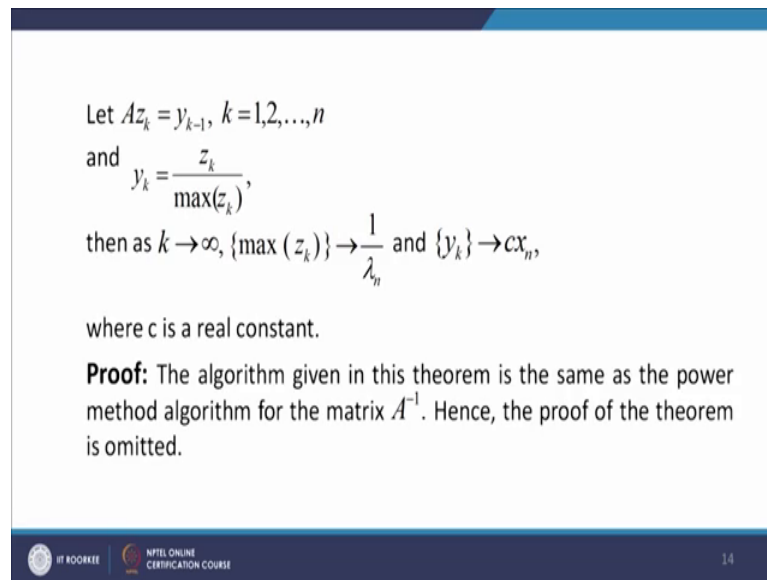
So, we write 111 equal to the matrix, which is given to us in to z^1 and solve this system of equations. Solve this system of equations we can use Gaussian elimination method with partial pivoting and determine the vector z^1 .

So, we get the vector z^1 . This we do, because we do not want to find the inverse of the matrix A and right z^1 . So, because if the, if the A order of the matrix is very large ok. A^{-1} inverse cannot be found easily.

So, we will write this equation in this manner and solve this equation matrix equation to determine the vector z^1 and once z^1 is known, we will determine by 1 by z^1 upon maximum of z^1 .

Then s^k will go to infinity, maximum of z^k will go to $1/\lambda_n$, because we are replacing the matrix A by A^{-1} . So, maximum of z^k will now go to $1/\lambda_n$ and y^k will go to c/λ_n , where c is a real constant. So, the algorithm given is the here.

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Let $Az_k = y_{k-1}$, $k = 1, 2, \dots, n$
and $y_k = \frac{z_k}{\max(z_k)}$,
then as $k \rightarrow \infty$, $\{\max(z_k)\} \rightarrow \frac{1}{\lambda_n}$ and $\{y_k\} \rightarrow cx_n$,

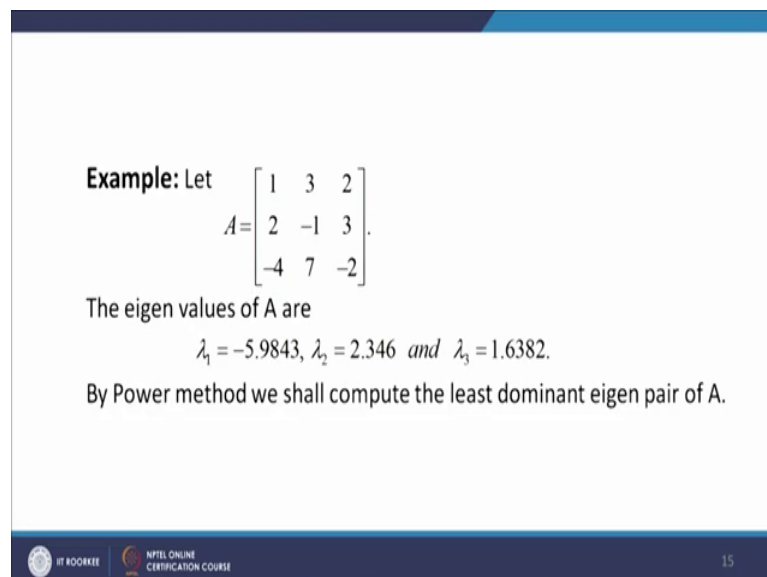
where c is a real constant.

Proof: The algorithm given in this theorem is the same as the power method algorithm for the matrix A^{-1} . Hence, the proof of the theorem is omitted.

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The algorithm is the same as the algorithm for the matrix A inverse. So, the, we do not write the proof. Let us illustrate this method by an example.

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Example: Let $A = \begin{bmatrix} 1 & 3 & 2 \\ 2 & -1 & 3 \\ -4 & 7 & -2 \end{bmatrix}$.

The eigen values of A are $\lambda_1 = -5.9843$, $\lambda_2 = 2.346$ and $\lambda_3 = 1.6382$.

By Power method we shall compute the least dominant eigen pair of A .

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Say, let us take the matrix A to be $\begin{bmatrix} 1 & 3 & 2 \\ 2 & -1 & 3 \\ -4 & 7 & -2 \end{bmatrix}$. Then if we find the Eigenvalues of A , they are -5.9843 , 2.346 and 1.6382 and you can see from here that the least dominant Eigen value of A here is 1.6382 .

This dominant Eigenvalue here is 1.63 82. We are going to find the least dominant Eigen pair of A; that is the Eigenvalue lambda 3 1. 6382 and the Eigenvector x 3 corresponding to lambda 3 by power method.

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Let $y_0 = [1 \ 1 \ 1]^T$.

k=1: $Az_1 = y_0$ gives

$$z_1 = [-0.5217 \ 0.0435 \ 0.6957]^T, \max(z_1) = 0.6957$$

$$y_1 = \frac{z_1}{\max(z_1)} = [-0.7499 \ 0.0625 \ 1.0000]^T$$

k=2: $Az_2 = y_1$

$$\Rightarrow z_2 = [-1.1521 \ -0.3206 \ 0.6820]^T, \max(z_2) = -1.1521$$

$$y_2 = \frac{z_2}{\max(z_2)} = [1.0000 \ 0.2783 \ -0.5820]^T$$

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So, we begin with y naught equal to 111 and then solve the equation $A z_1$ equal to y naught $A z_1$ equal to y naught. When we solve we get the value of z_1 equal to minus 0.5 217 0.0435 0.6957 and you can see here maximum of z_1 is the 0.69 57. So, we divide z_1 by maximum of z_1 to arrive at the vector y_1 and it is 0 minus 0.7499 0.0625 1.00 00 transpose.

Then to find the next approximation vector by 2, we first solve $A z_2$ equal to y_1 . So, we put here k equal to 2, so that $A z_2$ equal to y_1 , we have, y_1 is given, y_1 is already. We have found A is known to us.

So, we can solve this equation $A z_2$ equal to y_1 to arrive at the value of z_2 z_2 is then minus 1. 15 21 minus 0.ah 3206 0; 68 20 and we can then find maximum of z_2 here which is 0, which is, sorry which is minus 1 1521 absolute value. We look at absolutely, if you take absolute values of the components of z_2 then this component is the highest, this is largest.

So, maximum of z_2 is minus 1. 1521. So, to arrive at vector y_2 we divide z_2 y maximum of z_2 and we get this vector 1. 0000 1 0. 2783 minus 0.58 20.

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k=3: $Az_3 = y_2$
 $\Rightarrow z_3 = [0.8672 \quad 0.3010 \quad -0.3851]^T$, $\max(z_3) = 0.8672$
 $y_3 = \frac{z_3}{\max(z_3)} = [1.0000 \quad 0.3471 \quad -0.4440]^T$.

Proceeding similarly, we obtain

k=20: $z_{20} = [0.6107 \quad 0.2530 \quad -0.1848]^T$, $\max(z_{20}) = 0.6107$
and $y_{20} = \frac{z_{20}}{\max(z_{20})} = [1.0000 \quad 0.4142 \quad -0.3025]^T$.

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Then we performed third iteration and in the third iteration we will solve the question Az_3 equal to y_2 . When we take k equal to 3 z_3 equal to $A^{-1}y_2$, so Az_3 equal to y_2 .

So, Az_3 equal to y_2 we solve to get z_3 vector. z_3 vector is this $0.8672 \quad 0.3010$ minus 0.3851 transpose and if you find the maximum of vector maximum of z_3 here, then it is 0.8672 .

So, we can find y_3 , y_3 is equal to z_3 divided by maximum of z_3 . So, what do we get? $1.0000 \quad 0.3471$ minus 0.4440 transpose and proceeding similarly in the twentieth iteration z_{20} is equal to $0.6107 \quad 0.2530$ and minus 0.1848 transpose and you can find maximum of z_{20} here maximum of z_{20} is 0.6107 .

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Example: Let

$$A = \begin{bmatrix} 1 & 3 & 2 \\ 2 & -1 & 3 \\ -4 & 7 & -2 \end{bmatrix}$$

The eigen values of A are

$$\lambda_1 = -5.9843, \lambda_2 = 2.346 \text{ and } \lambda_3 = 1.6382.$$

By Power method we shall compute the least dominant eigen pair of A.

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Now, you see, let us go to this the least dominant is 1. 6382 if you find 1 over 1. 63 82, it comes out to be, it comes out to be ah this one 0.610 4.

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We note that $\lambda_3^{-1} = 0.6104$, hence $\max(z_{20}) = 0.6107$ is a good approximation for λ_3^{-1} . Further

$$y_{20} - \max(z_{20})Ay_{20} = 10^{-3} [0.4535 \quad 0.2263 \quad -0.0560]^T$$

is small so that y_{20} is a good approximation for the least dominant eigen vector x_3 for A.

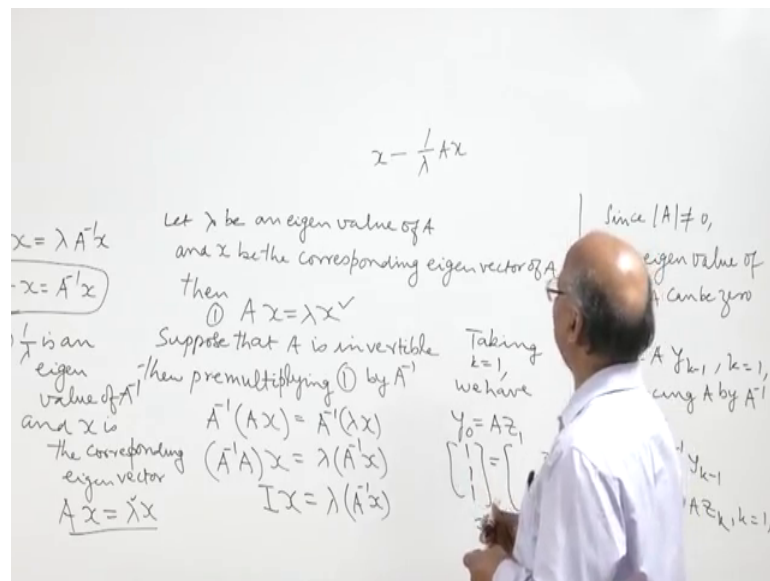
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So, you can see here maximum of z 20 which is 0.6 107 and when we divide z 20 by maximum of z 20, we arrive at y 20. Y 20 is 1.00 00 0.41 42 minus 0. 3025 and transpose of this vector.

So, we note that $\frac{1}{\lambda^3}$ is 0.6104 and maximum of z_{20} , we have found to be equal to 0.6107. So, you can see this maximum of z_{20} which is 0.6107 is a good approximation for $\frac{1}{\lambda^3}$ ok.

Further y_{20} , this y_{20} , let us look at this y_{20} . y_{20} is this one, this y_{20} minus maximum of $A z_{20}$ into $A y_{20}$. Maximum of $A z_{20}$ maximum of $A z_{20}$ is $\frac{1}{\lambda^3}$ ok, This $\frac{1}{\lambda^3}$ by λ^3 .

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So, see what we have, we have the question $\frac{1}{\lambda} - \frac{1}{\lambda} A x$ minus $\frac{1}{\lambda} A x$. This is what we are finding. this difference $x - \frac{1}{\lambda} A x$ is equal to $\frac{1}{\lambda} A x$. So, $x - \frac{1}{\lambda} A x$ must be nearly 0. So, what do we get here; x is y_{20} .

That is $x_{3 y_{20}} - \frac{1}{\lambda^3}$ maximum of z_{20} , then $A x_{y_{20}}$ which is $x_{3 y_{20}}$. So, this difference is 10^{-10} to the power minus 3 $0.45350.2263 - 0.0560$ transpose and you can see that, these components of the column vector are multiplied by 10^{-10} to the power minus 3.

So, this difference of y_{20} with the maximum of z_{20} into $A y_{20}$ is very small and therefore, we can say that y_{20} is a good approximation for the least dominant Eigenvector x_3 of A . So, we can find the least dominant Eigenvalue of the matrix A and the least dominant Eigenvector of the matrix A by using the power method, when we

assume that the matrix A is real diagnosable matrix and also invertible. With this, I would like to conclude my lecture.

Thank you very much for your attention.