

Matrix Theory
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Courant - Fischer Theorem

(Refer Slide Time: 00:13)

Handwritten notes on a Microsoft Whiteboard showing the derivation of the Courant-Fischer theorem for the $(n-1)$ th eigenvalue. The notes include the following:

$$x^H x = 1$$

$$x \perp w$$

$$\inf_{w \in \mathbb{C}^n} \sup_{\substack{x^H x = 1 \\ x \perp w}} x^H A x \geq \lambda_{n-1}$$

From the above, equality can be attained by setting $w = u_n$.

$$\inf_{w \in \mathbb{C}^n} \sup_{\substack{x^H x = 1 \\ x \perp w}} x^H A x = \lambda_{n-1}$$

Handwritten notes on a Microsoft Whiteboard defining the Courant-Fischer theorem and stating the min-max theorem. The notes include the following:

Thm. [Courant - Fischer]: "min-max thm.":

$A \in \mathbb{C}^{n \times n}$, Herm., EVals $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$

Let k be an integer, $1 \leq k \leq n$. Then,

$$\min_{w_1, \dots, w_{n-k} \in \mathbb{C}^n} \max_{\substack{x \neq 0, x \in \mathbb{C}^n \\ x \perp w_1, \dots, w_{n-k}}} \frac{x^H A x}{x^H x} = \lambda_k$$

and

$$\max_{w_1, \dots, w_{k-1} \in \mathbb{C}^n} \min_{\substack{x \neq 0, x \in \mathbb{C}^n \\ x \perp w_1, \dots, w_{k-1}}} \frac{x^H A x}{x^H x} = \lambda_k$$

Professor Chandra R. Murthy: So, the generalization of this for all eigenvalues is the following theorem, which is a very central theorem in this variational characterization of eigenvalues. This is the theorem by two people Courant and Fischer, this is called the min max theorem.

So, as usual our setup is A is an n cross n Hermitian symmetric matrix and it has eigenvalues λ_1 less than or equal to λ_2 less than or equal to etcetera λ_n and suppose

or let k be an integer $1 \leq k \leq n$, then we have two, two results, the minimum over w_1 through w_{n-k} in C to the n .

So, I am allowed to choose $n-k$ vectors in C to the n over which I am doing this minimization the maximum over $x \neq 0$, x in C to the n , x perpendicular to w_1 through w_{n-k} , $x^H A x$ over $x^H x$ is equal to λ_k and the maximum over w_1 to w_{k-1} , the minimum $x \neq 0$, x in C to the n , x perpendicular to w_1 through w_{k-1} , $x^H A x$ over $x^H x$ is also equal to λ_k .

So, there are two ways to write λ_k as a solution to an optimization problem and in both cases, it is a double optimization there is a min max in one case and a max min in the other case.

And, of course, like I mentioned before if $k=1$ or n in, in this case, when $k=1$, this goes over w_1 through w_{n-1} and x perpendicular to w_1 to w_{n-1} whereas, in this case this maximization step you can drop it because there is no such thing as w_0 , so this constraint does not arise. So, you just have to do the minimum over all $x \neq 0$, x in C to the n $x^H A x$ over $x^H x$ that will be equal to λ_1 . So, this constraint also drops off.

Similarly, when $k=n$, there is no such thing as w_0 here, so, this minimization drops off and similarly, x perpendicular to w_1 through w_0 there is no such thing. So, this constraint also drops off. And so, the maximum over all non-zero x of $x^H A x$ over $x^H x$ is equal to λ_n .

So, when $k=1$ or n in one of the two cases, we will be omitting one of the outer optimizations, so we will prove this result, we will only prove the first, this first part, the other part is almost, exactly the same but you just have to modify the steps a bit.

(Refer Slide Time: 4:53)

and $\max_{\omega_1, \dots, \omega_{k-1} \in \mathbb{C}^n} \min_{\substack{x \neq 0, x \in \mathbb{C}^n \\ x \perp \omega_1, \dots, \omega_{k-1}}} \frac{x^H A x}{x^H x} = \lambda_k.$

Proof: $A = U \Lambda U^H$, U unitary, $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$

Let $1 \leq k \leq n$. If $x \neq 0$

$$\frac{x^H A x}{x^H x} = \frac{(u^H x)^H \Lambda (u^H x)}{(u^H x)^H (u^H x)}$$

$$\{u^H x : x \in \mathbb{C}^n, x \neq 0\} = \{y \in \mathbb{C}^n, y \neq 0\}$$

So, as usual, we will write A as $u \Lambda u^H$, u is unitary and Λ is diagonal of λ_1 through λ_n , then let k be some number which is between 1 and n . Then if x is some vector which is not equal to 0, then $\frac{x^H A x}{x^H x}$ is as usual, $\frac{x^H u \Lambda u^H x}{x^H x}$ divided by $x^H x$.

And further, if I look at all vectors such that x is over all non-zero x , is the same as, so I will write it this way, $u^H x$, $x \in \mathbb{C}^n$, $x \neq 0$ is the same as the set of vectors $y \in \mathbb{C}^n$, $y \neq 0$. In other words, I am thinking of $u^H x$ as y and if I want to optimize this over all $x \neq 0$, I can as well optimize it over all $y \neq 0$.

(Refer Slide Time: 7:00)

$$\begin{aligned} & \{u^H x : x \in \mathbb{C}^n, x \neq 0\} = \{y \in \mathbb{C}^n, y \neq 0\} \\ & \text{If } w_1, \dots, w_{n-k} \in \mathbb{C}^n \text{ are given,} \\ & \sup_{\substack{x \neq 0 \\ x \perp w_1, \dots, w_{n-k}}} \frac{x^H A x}{x^H x} = \sup_{\substack{y \neq 0 \\ y \perp u^H w_1, \dots, u^H w_{n-k}}} \frac{y^H \wedge y}{y^H y} = \sum_{i=1}^n \lambda_i |y_i|^2 \\ & = \sup_{\substack{y^H y = 1 \\ y \perp u^H w_1, \dots, u^H w_{n-k}}} \sum_{i=1}^n \lambda_i |y_i|^2 \end{aligned}$$

So, if w_1 through w_{n-k} in \mathbb{C}^n are given, then the sup of $x \neq 0$ perpendicular to w_1 through w_{n-k} of $x^H A x / x^H x$ is equal to the supremum over $y \neq 0$, y perpendicular to $u^H w_1$ through $u^H w_{n-k}$ of $y^H \wedge y / y^H y$. Since y is $u^H A x$, I can write this as y is perpendicular to $u^H w_1$ through $u^H w_{n-k}$ of $y^H \wedge y / y^H y$.

And as before, I will expand this out. And this numerator is equal to $\sum_{i=1}^n \lambda_i |y_i|^2$. And further, I can impose a constraint that $y^H y = 1$ and optimize this over all $y^H y = 1$. So, just for the sake of completeness, let me write this step. This is equal to supremum over $y^H y = 1$, y perpendicular to $u^H w_1$ through $u^H w_{n-k}$ of $\sum_{i=1}^n \lambda_i |y_i|^2$.

(Refer Slide Time: 8:52)

$$\begin{aligned}
 & y \perp u^H w_1 \dots u^H w_{n-k} \\
 & \Rightarrow \sup_{y^H y = 1} \sum_{i=k}^n \lambda_i |y_i|^2 \\
 & y \perp u^H w_1 \dots u^H w_{n-k} \\
 & y_1 = y_2 = \dots = y_{k-1} = 0 \\
 & \Rightarrow \sup_{|y_k|^2 + |y_{k+1}|^2 + \dots + |y_n|^2 = 1} \sum_{i=k}^n \lambda_i |y_i|^2 \\
 & y \perp u^H w_1 \dots u^H w_{n-k} \\
 & \Rightarrow \lambda_k
 \end{aligned}$$

Now, I will do my brilliant thing from the previous discussion. And I will say that this is greater than or equal to the supremum over y Hermitian y equals 1, y perpendicular to u Hermitian w_1 up to u Hermitian w_{n-k} . And I will further set y_1 to y_{k-1} minus 1, y_k minus 1 equal to 0 of the summation i equal to 1 to n , y_i mod λ_i mod y_i square.

But, since I have set all these guys equal to 0, I can go i equal to k to n λ_i mod y_i square and this in turn is equal to the supremum since the first $k-1$ terms are equal to 0, this constraint $y^H y$ equals 1 reduces to y_k square plus y_{k+1} square plus etcetera up to y_n square equals 1.

And y should still remain perpendicular to u Hermitian w_1 up to u Hermitian w_{n-k} of $\sum_{i=k}^n \lambda_i |y_i|^2$. And of course, as before this is a convex combination of λ_k λ_{k+1} up to λ_n . And this convex combination is at least equal to the smallest value here.

Another way to think about it is I will replace all these λ_i 's with λ_k , then I am only decreasing the value. And $\sum_{i=k}^n |y_i|^2$ equals 1 because of this constraint here. And so, there is nothing left to optimize. And so, I can say that this is greater than or equal to λ_k .

(Refer Slide Time: 11:08)

$$\geq \lambda_k$$

$$\Rightarrow \sup_{\substack{x \neq 0 \\ x \perp w_1, \dots, w_{n-k}}} \frac{x^H A x}{x^H x} \geq \lambda_k \quad \text{true for arbitrary } w_1, \dots, w_{n-k}$$

Prev. result above shows that

$$\Rightarrow \inf_{w_1, \dots, w_{n-k}} \sup_{\substack{x \neq 0 \\ x \perp w_1, \dots, w_{n-k}}} \frac{x^H A x}{x^H x} = \lambda_k \quad \square$$

tending this argument,

$$\min_{\substack{x \neq 0 \\ x \perp u_1, u_2, \dots, u_{k-1}}} \frac{x^H A x}{x^H x} = \min_{\substack{x^H x = 1 \\ x \perp u_1, \dots, u_{k-1}}} x^H A x = \lambda_k \quad k = 2, 3, \dots$$

Similarly,

$$\max_{\substack{x \neq 0 \\ x \perp u_n, u_{n-1}, \dots, u_{n-k+1}}} \frac{x^H A x}{x^H x} = \max_{\substack{x^H x = 1 \\ x \perp u_n, u_{n-1}, \dots, u_{n-k+1}}} x^H A x = \lambda_{n-k} \quad k = 1, 2, \dots$$

Let $w \in \mathbb{C}^n$ be a given vector. Then,

$$\sup_{x \perp w} \frac{x^H A x}{x^H x} = \sup_{x \perp U} \frac{x^H U U^H x}{x^H x} = \sup_{x \perp U} \frac{\sum_{i=1}^n \lambda_i |(U^H x)_i|^2}{x^H x}$$

So, what we have shown then is that, the supremum over all x not equal to 0, x perpendicular to w_1 through w_{n-k} , $x^H A x$ divided by $x^H x$ is greater than or equal to λ_k . And this is true for arbitrary w_1 through w_{n-k} . But, again the previous result above shows that which is I am referring to this one here, scrolling up. So, the minimum over, so, as the first step I need to change k to $n-k$, then what will happen is this u_n I would be doing.

So, if I replace k by $n-k$, I would be going from u_n , u_{n-1} up to u_1 , I have replaced k with $n-k$ and I would have, I am counting down, I have to count down up to u_{k+1} .

Student: Sir $k+1$. Yes sir.

Professor Chandra R. Murthy: Okay, and this, if x is perpendicular to all of these vectors, then what I will get here will be let me write this with a different color so that it is not confusing later on, u_1, u_2, \dots, u_{n-k} plus 1 will give me I have replaced k with $n - k$, so, I will get λk here.

Now, what I will do is, I will call this vector w_1 , this vector w_2 , etcetera, then this will become w . So, if, so this will, so how many vectors do I have here? I have, I am going down from n to $k + 1$. So, there is.

Student: $n - k$.

Professor Chandra R. Murthy: And, there are exactly $n - k$ vectors here and so, I have w_1 through w_{n-k} . So basically if I set w_1 equals u_1 , w_2 equals u_2 , and w_{n-k} equals u_{n-k} plus 1 the maximum of this subject to x perpendicular to all these vectors will be equal to λk . So, the largest value this can take for that specific choice of w_1 through w_{n-k} , w_1, w_2 up to w_{n-k} is equal to λk .

So, therefore, just go back and think about it. So, this result shows that equality, equality here is valid when w_i equals u_i plus 1, that implies that the infimum over all w_1 through w_{n-k} of the supremum x not equal to 0, x perpendicular to w_1 through w_{n-k} , x Hermitian Ax over x Hermitian x is equal to λk , which completes the proof. So, I do not want to go further ahead, the next theorem is a is another very, very crucial theorem, which is called Weyl's theorem. We will discuss that the next class.

But what I strongly suggest is, you know, you guys should definitely go over the proof on your own and make sure you understand every step of this proof, because the ideas in this proof will use again and again and again to prove many, many more results. And these arguments are slightly tricky to convey orally, especially in this online mode.

And since I cannot see you guys, I do not know if you are able to follow the proof as I explained it or not. But based on whatever I said, if you now go back and look at the proof on your own, you will be able to fill in the steps. And if you are not, then please stop me at the beginning of the next class and ask me which step you were not able to follow. And we can go over the argument again.

But the arguments we made today are crucial, we are going to reuse them in many proofs going forward. And at that time, if this proof is not completely clear to you, you would not

follow many of the proofs that we are going to discuss in the following classes. So please spend some time on it. So, we will stop here for today and we will continue in the next class.