

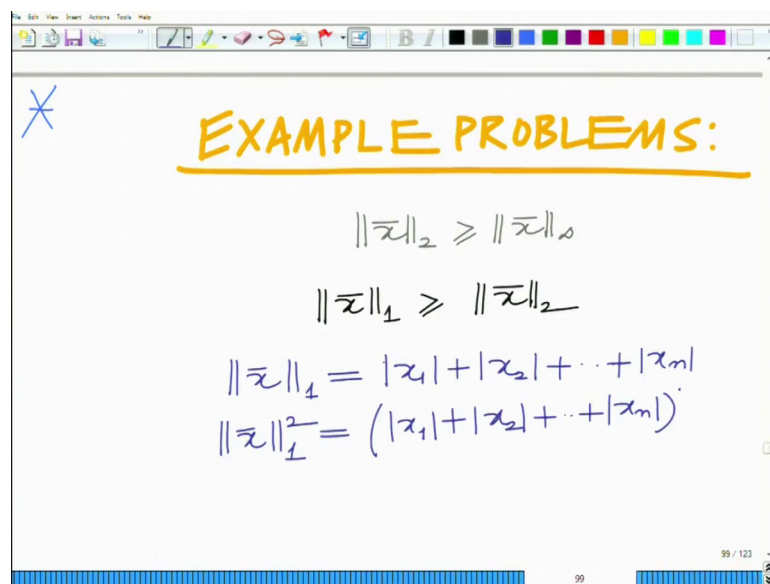
Applied Optimization for Wireless, Machine Learning, Big Data
Prof. Aditya K. Jagannatham
Department of electrical engineering
Indian Institute of Technology, Kanpur

Lecture – 21

Example Problems: Property of Norms, Problems on Convex Sets

Hello welcome to another module in this massive open online course. So, we are looking at example problems related to matrices and convex sets. Let us continue our discussion.

(Refer Slide Time: 00:24)



EXAMPLE PROBLEMS:

$$\|x\|_2 \geq \|x\|_\infty$$
$$\|x\|_1 \geq \|x\|_2$$
$$\|x\|_1 = |x_1| + |x_2| + \dots + |x_m|$$
$$\|x\|_1^2 = (|x_1| + |x_2| + \dots + |x_m|)^2$$

So, you are looking at example problems in particular in the previous module we started looking at the properties of norms, correct in and in particular we have seen that for instance the l 2 norm is greater than the l infinity norm.

Now, in the same way and one can also show that the l 1 norm for a vector is greater than or equal to the l 2 norm ok. And this can be shown simply as follows. If you look at the l 1 norm that is simply for an n dimensional vector, the sum of the absolute values of the magnitudes that is what we have seen that is the l 1 norm ok. And if you look at the square of the l 1 norm the square of this quantity that is simply magnitude x 1 plus magnitude x 2 plus magnitude x n whole square.

(Refer Slide Time: 01:53)

The whiteboard shows the following derivation:

$$\begin{aligned} \|\bar{x}\|_1 &= |x_1| + |x_2| + \dots + |x_n| \\ \|\bar{x}\|_1^2 &= (|x_1| + |x_2| + \dots + |x_n|)^2 \\ &= |x_1|^2 + |x_2|^2 + \dots + |x_n|^2 \\ &\quad + \underbrace{\sum_i \sum_j |x_i| |x_j|}_{\geq 0} \\ &\geq |x_1|^2 + |x_2|^2 + \dots + |x_n|^2 \\ &= \|\bar{x}\|_2^2 \end{aligned}$$

Which you can expand as follows which is basically well it is magnitude x 1 square plus magnitude x 2 square plus so, on magnitude x n square plus summation over all combinations of i comma j the product magnitude x i into magnitude x j.

Now, this quantity this products cross products, the sum of all cross products, this is greater than or equal to 0. Because the magnitudes are positive implies that this is greater than or equal to magnitude x 1 square plus magnitude x 2 square plus magnitude x n square and this is nothing, but the 2 norm square norm x 2 bar square.

(Refer Slide Time: 02:50)

The whiteboard shows the following derivation:

$$\begin{aligned} &\geq |x_1|^2 + |x_2|^2 + \dots + |x_n|^2 \\ &= \|\bar{x}\|_2^2 \\ \Rightarrow \|\bar{x}\|_1^2 &\geq \|\bar{x}\|_2^2 \\ \Rightarrow \boxed{\|\bar{x}\|_1 \geq \|\bar{x}\|_2} \end{aligned}$$

So, we basically have what we have is that norm \bar{x} the 1 norm square is greater than or equal to the 2 norms square and therefore, this implies both these quantities are positive the square of 1 is greater than the square of the other this means the 1 norm is greater than or equal to the 2 norm.

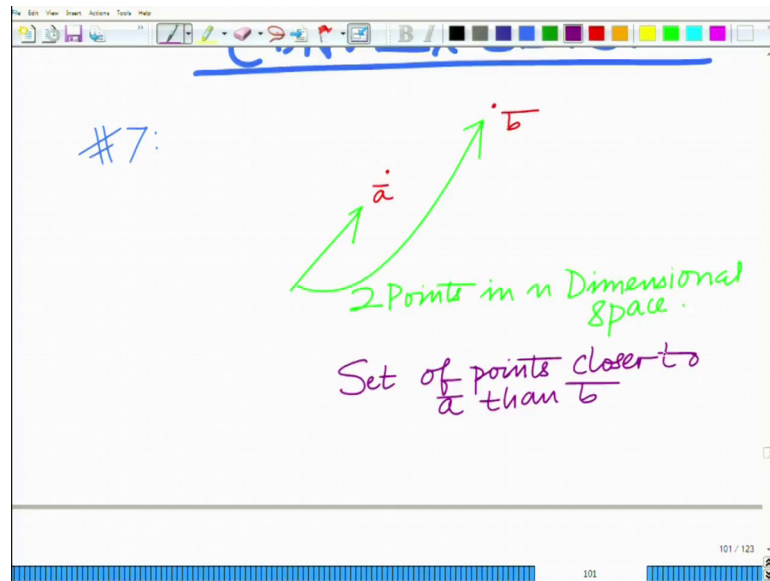
And we are already seen that the 2 norm is greater than the infinity norm. Therefore, putting all these things together, you have the 1 norm is greater than or equal to the 2 norm is greater than or equal to the infinity norm.

(Refer Slide Time: 03:23)

The image shows a whiteboard with handwritten mathematical derivations. At the top, it says $= \|\bar{x}\|_2^2$. Below that, an arrow points to $\|\bar{x}\|_1^2 \geq \|\bar{x}\|_2^2$. A second arrow points to a boxed equation $\|\bar{x}\|_1 \geq \|\bar{x}\|_2$. A third arrow points to a larger boxed equation $\|\bar{x}\|_1 \geq \|\bar{x}\|_2 \geq \|\bar{x}\|_\infty$. The whiteboard interface includes a toolbar at the top and a status bar at the bottom showing '100 / 123'.

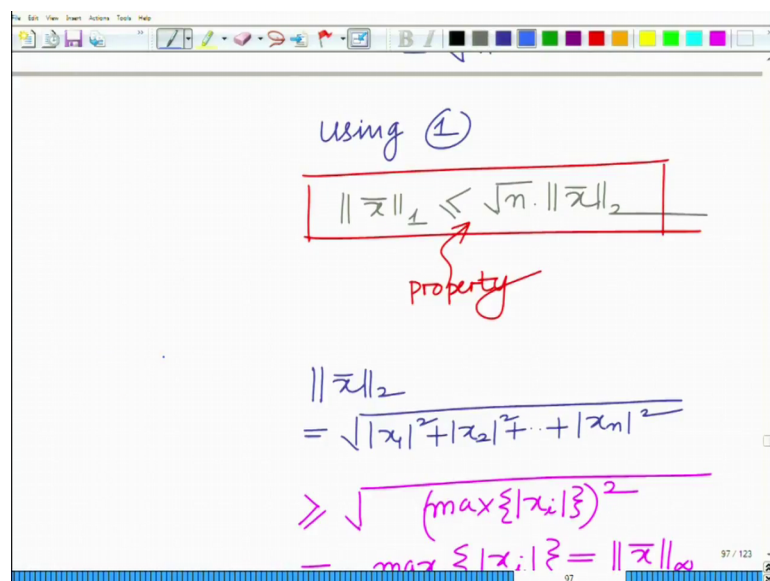
This is the result that we have the 1 norm of a vector is greater than or equal to the 2 norm is greater than or equal to that infinitely norm. So, 1 norm is sum of the magnitude values of the element. The 2 norm is you can think of this as the length the Euclidian the length of the vector in Euclidian space. And the infinity norm is the maximum value of the magnitudes of the different elements of the vector. These are the different norms ok. Let us continue our discussion, now let us move on to look at example problems related to convex sets and their applications alright. So, we have defined convex sets look at their properties. Let us do some example to understand these things better ok.

(Refer Slide Time: 04:29)



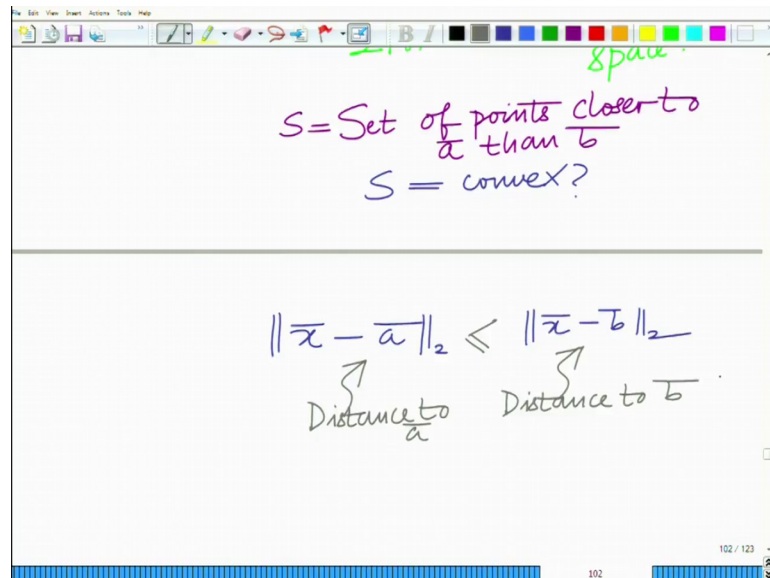
So, what we want to look at next is convex sets, if you want to explore different types of convex sets their definition classification and so, on ok. Now, for instance let us start with the first problem, problem number let us call this as problem number the previous one as problem number we had this, let us call this as problem number. So, we had problem number I think 6. So, let us call this as problem number 5 6. So, let us call this as problem number 7, I think we can start problem number or let us forget this we can start problem number 7 over here.

(Refer Slide Time: 05:16)



Let us call this as problem number 7 and what we want to show is the set of let us say we can consider two vectors or two points in n dimensional space, denoted by the vectors \bar{a} and the vector \bar{b} and we want to find. So, these are two points in n dimensional space ok, these are all points to points in n dimensional space. Now, we want to find the points set of points closer to \bar{a} than \bar{b} set of points that are closer to \bar{a} than \bar{b} .

(Refer Slide Time: 06:34)



And what we want to see is if we call this set S we want to show, we want to find is S convex is this set S of all points which are closer to \bar{a} than \bar{b} that is where our two points in n dimensional space \bar{a} and \bar{b} . It is the set of points \bar{x} which are closer to \bar{a} than \bar{b} is this set of points convex. That is if you look at the set of points, the distance of any point \bar{x} between \bar{a} we know that distance is the 2 norm, norm of \bar{x} minus \bar{a} . And the distance to \bar{b} is norm the 2 norm \bar{x} minus \bar{b} , we want to see what is \bar{x} belongs to this set if this distance that is this distance to \bar{a} is less than or equal to distance to \bar{b} .

(Refer Slide Time: 07:45)

Handwritten mathematical derivation on a whiteboard:

$$\|\bar{x} - \bar{a}\|_2 \leq \|\bar{x} - \bar{b}\|_2$$

\nearrow Distance to \bar{a} \nearrow Distance to \bar{b}

$$\Rightarrow \|\bar{x} - \bar{a}\|^2 \leq \|\bar{x} - \bar{b}\|^2$$

$$\Rightarrow (\bar{x} - \bar{a})^T (\bar{x} - \bar{a}) \leq (\bar{x} - \bar{b})^T (\bar{x} - \bar{b})$$

102 / 123

Now, is this set convex well this implies that norm \bar{x} minus \bar{a} bar whole square less than or equal to norm, \bar{x} bar minus \bar{b} bar whole square. Now, remember the norm square of a vector is a vector transpose times itself, this implies norm \bar{x} bar minus \bar{a} bar transpose norm \bar{x} bar minus \bar{b} bar less than or equal to or norm \bar{x} bar minus \bar{a} bar, vector transpose times itself is less than or equal to norm \bar{x} bar minus \bar{b} bar transpose norm \bar{x} bar minus \bar{b} bar.

(Refer Slide Time: 08:29)

Handwritten mathematical derivation on a whiteboard:

$$\Rightarrow \|\bar{x} - \bar{a}\|^2 \leq \|\bar{x} - \bar{b}\|^2$$

$$\Rightarrow (\bar{x} - \bar{a})^T (\bar{x} - \bar{a}) \leq (\bar{x} - \bar{b})^T (\bar{x} - \bar{b})$$

$$\bar{a}^T \bar{x} = (\bar{a}^T \bar{x})^T = \bar{x}^T \bar{a}$$

$$\Rightarrow \bar{x}^T \bar{x} - 2\bar{a}^T \bar{x} + \bar{a}^T \bar{a} \leq \bar{x}^T \bar{x} - 2\bar{b}^T \bar{x} + \bar{b}^T \bar{b}$$

$$\Rightarrow 2(\bar{b} - \bar{a})^T \bar{x} \leq \|\bar{b}\|^2 - \|\bar{a}\|^2$$

102 / 123

This implies that if you split this or if you expand this will be $\bar{x}^T \bar{x}$ minus $\bar{a}^T \bar{x}$ plus $\bar{x}^T \bar{a}$. Now, since they are scalar $\bar{a}^T \bar{x}$ and $\bar{x}^T \bar{a}$ are the same. So, this will be minus 2 $\bar{x}^T \bar{a}$, plus $\bar{a}^T \bar{a}$ less than or equal to again the same thing similar $\bar{x}^T \bar{x}$ minus 2 $\bar{x}^T \bar{a}$ plus $\bar{a}^T \bar{a}$. I can also write it as $\bar{b}^T \bar{x}$ minus $\bar{a}^T \bar{x}$ plus $\bar{x}^T \bar{a}$ minus $\bar{x}^T \bar{a}$ plus $\bar{b}^T \bar{x}$ plus $\bar{a}^T \bar{a}$ minus $\bar{a}^T \bar{a}$.

And now you can see $\bar{x}^T \bar{x}$ minus $\bar{x}^T \bar{x}$ cancels this implies that bring the $\bar{b}^T \bar{x}$ over the other side this implies, let us write this as $\bar{c}^T \bar{x}$. Both of them are same $\bar{c}^T \bar{x}$ note that $\bar{c}^T \bar{x}$ equals, because it is a number $\bar{c}^T \bar{x}$ transpose which is equal to $\bar{x}^T \bar{c}$ because it is a number, it is a scalar quantity.

(Refer Slide Time: 09:59)

The image shows a whiteboard with the following handwritten content:

$$\Rightarrow \frac{(\bar{b} - \bar{a})^T \bar{x}}{\bar{c}} \leq \frac{\frac{1}{2}(\|\bar{b}\|^2 - \|\bar{a}\|^2)}{d}$$

$$\bar{c} = \bar{b} - \bar{a}$$

$$d = \frac{1}{2}(\|\bar{b}\|^2 - \|\bar{a}\|^2)$$

$$\Rightarrow \boxed{\bar{c}^T \bar{x} \leq d}$$

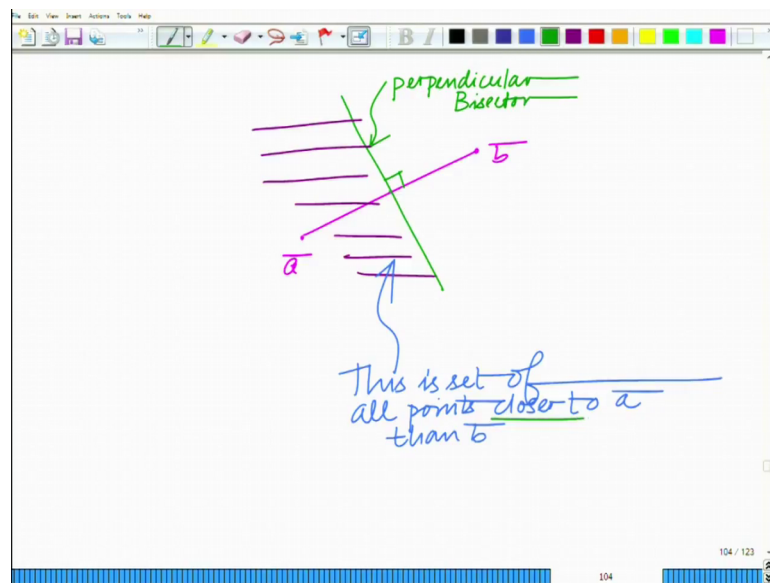
Represents a halfspace
 \Rightarrow convex.

So, this implies $\bar{b}^T \bar{x}$ minus $\bar{a}^T \bar{x}$ times $\bar{x}^T \bar{x}$ less than or equal to well $\bar{b}^T \bar{x}$ minus $\bar{a}^T \bar{x}$ plus $\bar{x}^T \bar{a}$ minus $\bar{x}^T \bar{a}$ plus $\bar{b}^T \bar{x}$ plus $\bar{a}^T \bar{a}$ minus $\bar{a}^T \bar{a}$ which implies that $\bar{b}^T \bar{x}$ minus $\bar{a}^T \bar{x}$ less than or equal to half norm \bar{b} square minus norm \bar{a} square. And therefore, if you call this as if you call this vector as \bar{c} $\bar{b} - \bar{a}$, if you denoted by \bar{c} and if you denote this quantity by d that is your \bar{c} equals $\bar{b} - \bar{a}$ and this constant d equals half, norm \bar{b} square minus norm \bar{a} square.

We have an equation of the form $c \bar{x} \leq d$ and, if you look at this nothing, but a this is nothing, but a equation of the half space $c \bar{x} = d$ that is a hyper plane. And $c \bar{x} \leq d$ some constant d that is the half space.

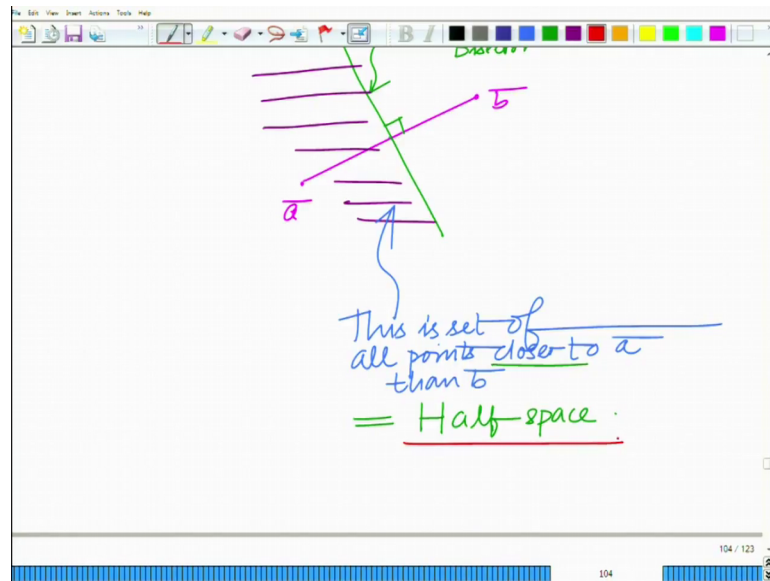
So, this set of all points \bar{x} which are closer to a , then b represents half space and therefore, the set is convex that completes the proof. So, if you look at this set it represents a half space, implies this is convex.

(Refer Slide Time: 11:46)



And interestingly if you look at that if you are wondering what that set is, if you look at these points a and b and the line joining this points a and b . And if you look at the perpendicular bisector of this ok, if you look at the perpendicular bisector of the line joining a and b . All the points that are closer to a than b lie on basically one side of this perpendicular bisector and this is the half space that you are talking about essentially this is the set of all points closer to a , this is a set of all points that are closer to a than b .

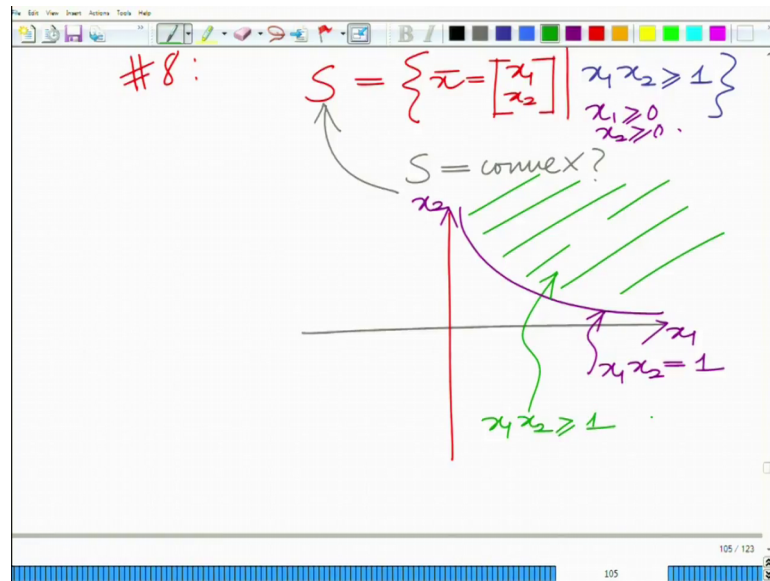
(Refer Slide Time: 12:55)



And therefore, this is equal to this is indeed you can see clearly this is a half space is a set of all points. It is we look at the perpendicular bisector of the line joining a bar and b bar. This is the high half space that lies towards a bar this is a half space that is the perpendicular bisector, divides a bar divides it into two half spaces the half space, that includes a bar that is the set of all points which are closer to a bar than b bar, which is indeed the half space and therefore, it is indeed convex ok

So, that basically complete the proof of course, we have proved it analytically and that should leave no question and this is basically an insight into the convex set ok.

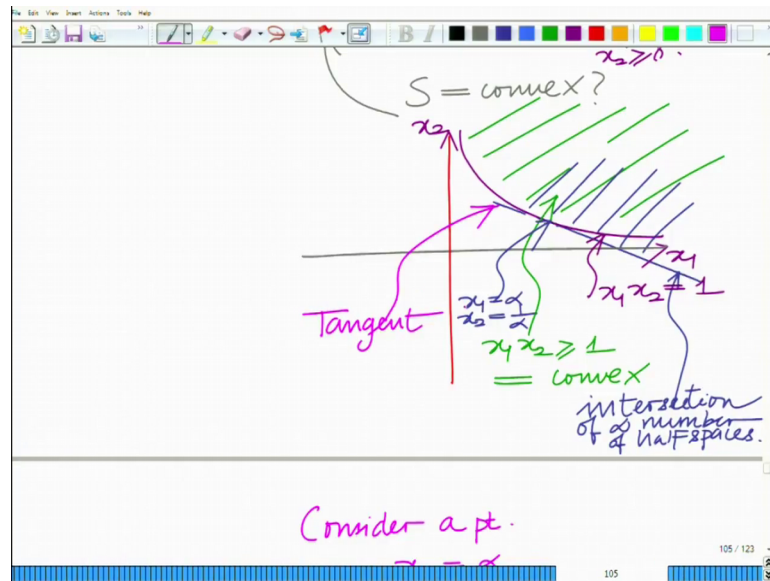
(Refer Slide Time: 13:49)



Let us look at another problem let us look at another interesting set. So, we have S equals the set of vectors \bar{x} . So, set of two dimensional vectors \bar{x} equals $x_1 \times x_2$ such that $x_1 \times x_2$ greater than or equal to 1. And we want to ask the question again is S convex is the set x convex. Now, if you look at this two dimensional plane ok, if you plot x_1 equals x_2 that will be this hyperbola, correct x_1 equals x_2 or $x_1 \times x_2$ equals 1. And of course, we are considering the positive let us also maintain that let us also impose this additional constraint x_1 greater than equal to 0 x_2 greater than or equal to 0 ok.

And now if you look at this now if you look at this so, this is your x_1 this is your x_2 and $x_1 \times x_2$ greater than or equal to 1, that includes this set ok. So, this is a set of all points such that $x_1 \times x_2$ greater equal to 1 and you can see visually see that this is convex.

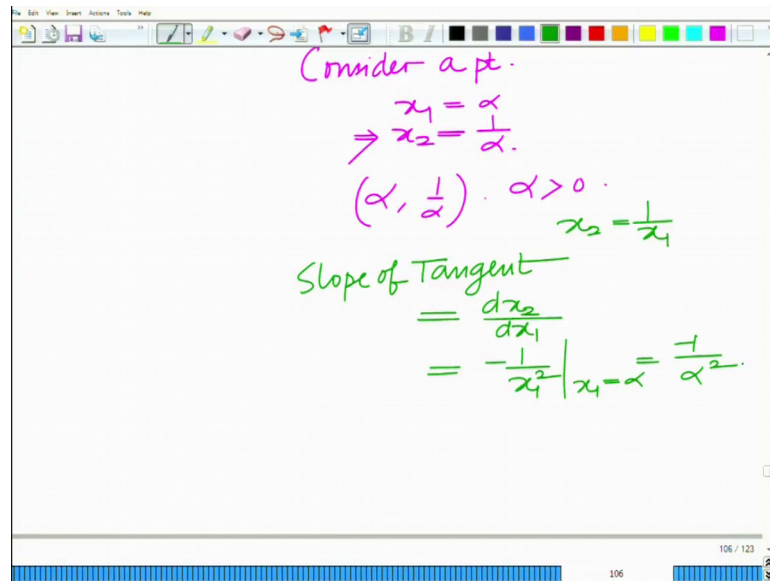
(Refer Slide Time: 15:22)



We are going to take an alternative approach what we are going to show is that this half space this convex set can be represented as a intersection of a infinite number of infinite number of half spaces. And since it is the intersection of half spaces it is essentially a polyhedral, it is well it is an infinite number of half spaces its essentially a convex set, because is the intersection of several convex sets that half spaces basically.

And you can clearly see that if you draw that tangent and, if you look at the corresponding half space ok. And if you draw all tangent all such tangents the infinite number of tangents, you can represent this as the intersection of an infinite number of such half spaces. Now, let us look at any point well where $x_1 = \alpha$ implies $x_2 = 1 - \alpha$ so, $x_1 x_2 = \alpha(1 - \alpha) \leq 1$ so, $x_2 \leq 1 - \alpha$.

(Refer Slide Time: 16:41)



Consider a pt.
 $x_1 = \alpha$
 $\Rightarrow x_2 = \frac{1}{\alpha}$
 $(\alpha, \frac{1}{\alpha})$. $\alpha > 0$.
 $x_2 = \frac{1}{x_1}$
Slope of Tangent
 $= \frac{dx_2}{dx_1}$
 $= -\frac{1}{x_1^2} \Big|_{x_1 = \alpha} = -\frac{1}{\alpha^2}$

So, we are considering any point consider any point so, let us start with this consider a point x_1 equals alpha this implies x_2 equals 1 over alpha since x_1 into x_2 equals 1 . So, we consider this point alpha comma 1 over alpha with alpha greater than 0 . And now if you look at this quantity dy by dx x_2 by dx_1 ,

That is if you look at the tangent to the curve, if look at the slope of the tangent. So, this is your tangent and the slope of the tangent to the curve equals dx_2 by dx_1 that is at that point the tangent has the same slope as the curve. So, dx_2 by dx_1 which is remember x_2 equals 1 over x_1 which implies dx_2 over dx_1 is minus derivative of x_2 that is 1 over x_1 with respect to x_1 is minus 1 over x whole square.

(Refer Slide Time: 18:22)

The whiteboard shows the following handwritten work:

$$x_2 = \frac{1}{x_1}$$
$$\text{Slope of Tangent} = \frac{dx_2}{dx_1}$$
$$= -\frac{1}{x_1^2} \Big|_{x_1 = \alpha} = -\frac{1}{\alpha^2}$$

There is a purple arrow pointing from the expression $(\alpha, \frac{1}{\alpha})$ to the point $x_1 = \alpha$ in the derivative calculation. The final result $-\frac{1}{\alpha^2}$ is also labeled as the slope of the tangent.

And at this point evaluated at x_1 equals alpha, so, this will be minus 1 over alpha 1 square. And so, this minus 1 over alpha square what is this is basically the slope of the tangent, why is this the slope of the tangent at that point alpha 1 over alpha that point x_1 comma x_2 . If you look at the derivative correct, if you look at the derivative of the curve that itself gives the slope of the tangent ok.

(Refer Slide Time: 18:46)

The whiteboard shows the following handwritten work:

$$\left(\alpha, \frac{1}{\alpha}\right) \text{ slope} = -\frac{1}{\alpha^2}$$
$$\frac{\left(x_2 - \frac{1}{\alpha}\right)}{\left(x_1 - \alpha\right)} = -\frac{1}{\alpha^2}$$
$$\Rightarrow x_2 - \frac{1}{\alpha} = -\frac{1}{\alpha^2} (x_1 - \alpha)$$
$$\Rightarrow \alpha^2 x_2 - \alpha = -x_1 + \alpha$$

There are three green arrows pointing downwards from the last equation, indicating further steps in the derivation.

And therefore, we are in the slope and we have the point the point is 1 over 1 over alpha and the slope equals minus 1 over alpha square. So, the tangent will be point through

alpha 1 over alpha with slope minus 1 over Alpha Square. So, that will be given as well x 2 minus x 2 minus 1 over alpha divided by x 1 minus alpha the slope is nothing, but difference of the y coordinate by difference of the x coordinate.

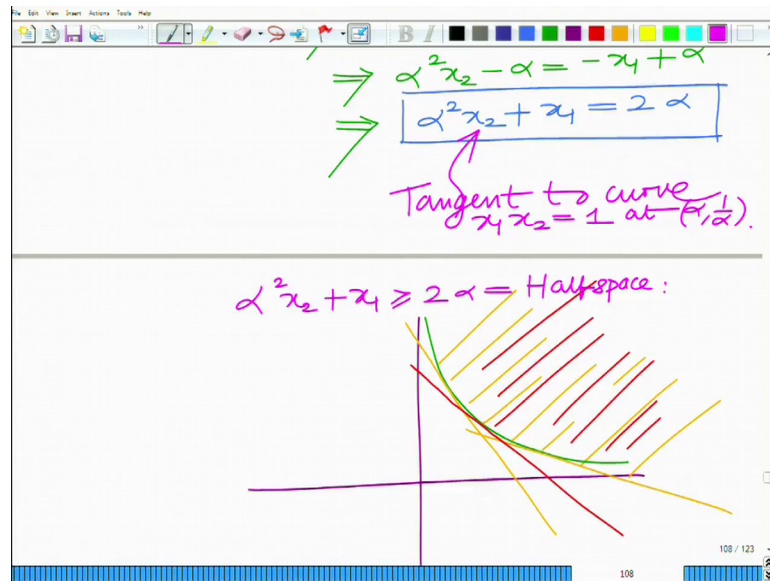
So, x 2 minus 1 over alpha divided by x 1 over alpha divided by x 1 over alpha equals minus 1 over alpha square, this implies x 2 minus 1 over alpha equals minus 1 over alpha square x 1 minus alpha this implies alpha square x 2 alpha square x 2 minus alpha equals well minus x 1 plus alpha this implies, if you can see this implies alpha square x 2 plus x 1 equals 2 alpha.

(Refer Slide Time: 20:04)

The image shows a whiteboard with handwritten mathematical work. At the top, the equation $\frac{(x_2 - \frac{1}{\alpha})}{(x_1 - \alpha)} = -\frac{1}{\alpha^2}$ is written in purple. Below it, three steps are shown with green arrows: $\Rightarrow x_2 - \frac{1}{\alpha} = -\frac{1}{\alpha^2}(x_1 - \alpha)$, $\Rightarrow \alpha^2 x_2 - \alpha = -x_1 + \alpha$, and $\Rightarrow \alpha^2 x_2 + x_1 = 2\alpha$. The final equation is boxed in blue. A pink arrow points from the boxed equation to the text "Tangent to curve $x_1 x_2 = 1$ at $(\alpha, \frac{1}{\alpha})$." The whiteboard interface includes a toolbar at the top and a status bar at the bottom showing "107 / 123".

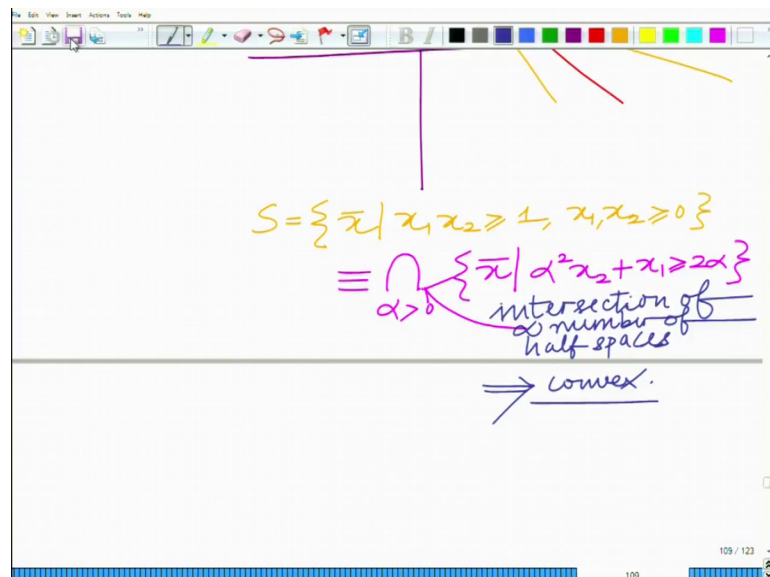
So, this is the tangent to this curve x 1 x 2 equal to 1 at alpha come 1 over alpha tangent ok.

(Refer Slide Time: 20:42)



So, basically what we have is you have you have this curve $x_1 x_2$ equal to 1 and you have this particular tangent you have this half space. Now, you take all such tangents and you take the intersection of these half spaces and what you will see, you will end up getting this original convex set.

(Refer Slide Time: 21:14)



So, set $x_1 x_2$ so, if you look at the set \bar{x} such that $x_1 x_2 \geq 1$, $x_1 \geq 0$, $x_2 \geq 0$. If you look at this set, if you look at this set this set is equivalent to the set, that is this it is equal to

the intersection of all the sets intersection over what intersection over any point α greater than 0 such that \bar{x} such that $\alpha^2 x_2 + x_1$ greater than or equal to 2α .

So this is $\alpha^2 x_2 + x_1$ equal to 2α , this is the tangent which implies $\alpha^2 x_2 + x_1$ greater than equal to 2α this is the half space ok. So, this convex set $x_1 x_2 x_1 x_2$ greater than equal to 1. This intersection of all these half spaces the intersection of an infinite number of half spaces implies, this is convex. So, the set $x_1 x_2$ greater than equal to 1, we represented that as the intersection of an infinite number of half spaces and therefore, indeed it is convex let. So, these are interesting applications interesting problems which demonstrate which show you how to demonstrate the convexity of set and how to visualize the different properties or different aspects of a convex set alright. So, we will stop here and continue in the subsequent modules.

Thank you.