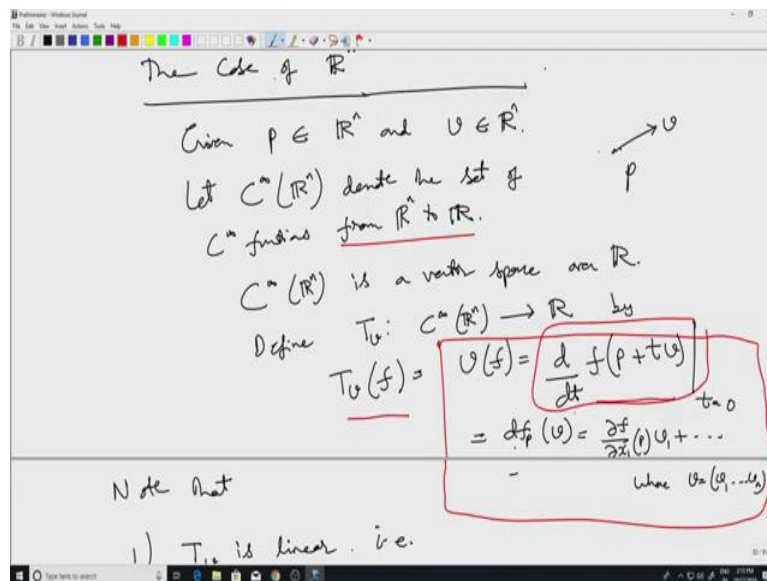


An Introduction to Smooth Manifolds
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Lecture 14
Tangent Spaces

Hello, and welcome to the 14th lecture in this series. So I am going to talk about tangent vectors and tangent spaces, eventually on manifolds, but to begin with. Let us look at the case of \mathbb{R}^n as a manifold.

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So given a point in \mathbb{R}^n , we it is natural to think of a tangent vector at \mathbb{R}^n at that point as any direction with the starting at that point. But, so here, let me redraw this picture. So given a point P , tangent vector would be just schematically one would draw it as an arrow based at P . But really, v is just any element of \mathbb{R}^n . Of course, the key point here is that this point P is being thought of \mathbb{R}^n has the point is that \mathbb{R}^n has two kinds of structures.

One is as a manifold, or even more basically a topological space. If you think of \mathbb{R}^n in that way, then an element of \mathbb{R}^n would be called a point on the other hand \mathbb{R}^n is also a vector space and from that point of view an element in \mathbb{R}^n would be a vector. So, in this setting I have a point P and a vector v and where v is what we would called a tangent victory at P .

But this is not a I mean this is not a formal definition of a tangent vector in fact, we formally define a tangent vector what we do is the following. So, let us as before let us $C^\infty(\mathbb{R}^n)$ denote the set of C^∞ functions from \mathbb{R}^n to \mathbb{R} from \mathbb{R}^n to \mathbb{R} and $C^\infty(\mathbb{R}^n)$ is a vector

space over \mathbb{R} so given a need to C^∞ functions I can take scalar multiples and add them up.

And then we define corresponding to every V , which is now thought of, which is still an element of \mathbb{R}^n , but now thought of as a vector, as an element of a vector space. I define T_v from $C^\infty(\mathbb{R}^n)$ to \mathbb{R} as essentially the directional derivative of f along the direction v at the point P . So, which we have denoted in my earlier lectures by V of f . And by definition, that is equal to the usual derivative of f when I restricted to the straight line that starts at the point P in the direction v namely P plus T_v .

So when I restrict f to that I get a function of one variable, namely T and I take the derivative of that by chain rule. That is also equal to, but not yet, I mean the we saw that this is equal to the derivative the differential of f as a linear transformation acting on v . And that is in fact, equal to the differential is just given it in by partial derivatives. So therefore, I have the last expression here.

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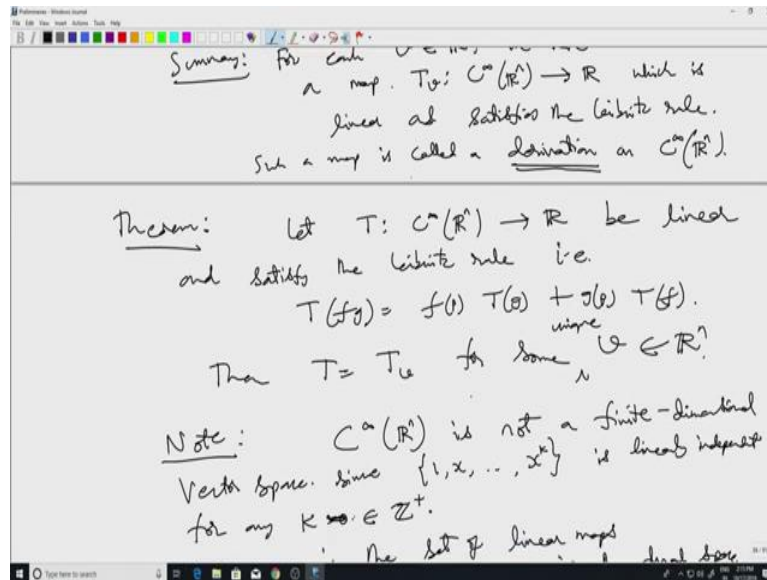
$C^\infty(\mathbb{R}^n)$ is
 Define $T_v: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ by
 $T_v(f) = \left. \frac{d}{dt} f(p+tv) \right|_{t=0}$
 $= df_p(v) = \frac{\partial f}{\partial x_1}(p) v_1 + \dots$
 where $v = (v_1, \dots, v_n)$
 Note that $T_v(f) = V(f)$
 1) T_v is linear, i.e.
 $T_v(c_1 f_1 + c_2 f_2) = c_1 T_v(f_1) + c_2 T_v(f_2)$
 $\forall c_1, c_2 \in \mathbb{R}, f_1, f_2 \in C^\infty(\mathbb{R}^n)$
 2) T_v satisfies the Leibniz rule:
 $T_v(fg) = f(p) T_v(g) + g(p) T_v(f)$

Okay, so the main thing is that I mean, we will need the other expressions but what we have to keep in mind is this T of f is T_v acting on f is the directional derivative of f along the direction v at the point P . Now, if you think of T_v in this fashion as a linear transformation from $C^\infty(\mathbb{R}^n)$ as a map from $C^\infty(\mathbb{R}^n)$ to \mathbb{R} then T_v is linear.

In other words, T_v of $C_1 f_1$ plus $C_2 f_2$ is $C_1 T_v f_1$ etcetera. And this follows immediately from this, this thing here the name that the definition of the directional derivative will immediately give the first linearity and that will also tell us that T_v is a, as T_v satisfies the

Leibnitz rule, the product rule for functions in other words, the directional derivative of a product is given in the expected way like this, what I have here.

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So, in summary, for each v in \mathbb{R}^n we have a map $T_v: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ which is linear and satisfies the Leibnitz rule let us call such a map as a derivation on $C^\infty(\mathbb{R}^n)$. So, linearity and the Leibnitz rule anything satisfying that will we will call it derivation. So, the main idea here is that every instead of talking about v in \mathbb{R}^n we would like to talk about this T_v . So, in other words, we would like to say that corresponding to every derivation is essentially of the form T_v and for a unique v .

So, we would like to regard first, we I mean the, all this property shows that corresponding to every vector I get a derivation, but we would like the converse as well. So that is the theorem here. Let T from $C^\infty(\mathbb{R}^n)$ to \mathbb{R} here be linear and satisfy the Leibnitz rule which I have again written down then the claim is that T corresponds to the directional T is actually equal to the directional derivative of exactly one vector v in \mathbb{R}^n . So this sets up a bijective correspondence between v , between \mathbb{R}^n and derivations on $C^\infty(\mathbb{R}^n)$.

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a map $T: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ linear and satisfies the Leibnitz rule. Such a map is called a derivation on $C^\infty(\mathbb{R}^n)$.

Theorem: Let $T: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ be linear and satisfies the Leibnitz rule i.e.

$$T(fg) = f(p) T(g) + g(p) T(f).$$

Then $T = T_p$ for some $p \in \mathbb{R}^n$.

Note: $C^\infty(\mathbb{R}^n)$ is not a finite-dimensional vector space. Since $\{1, x, \dots, x^k\}$ is linearly independent for any $k \in \mathbb{Z}^+$.
 \therefore The set of linear maps $C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$, i.e. the dual space

1) T_p is linear. i.e.

$$T_p(c_1 f_1 + c_2 f_2) = c_1 T_p(f_1) + c_2 T_p(f_2)$$

$\forall c_1, c_2 \in \mathbb{R}, f_1, f_2 \in C^\infty(\mathbb{R}^n)$.

2) T_p satisfies the Leibnitz rule:

$$T_p(fg) = f(p) T_p(g) + g(p) T_p(f)$$

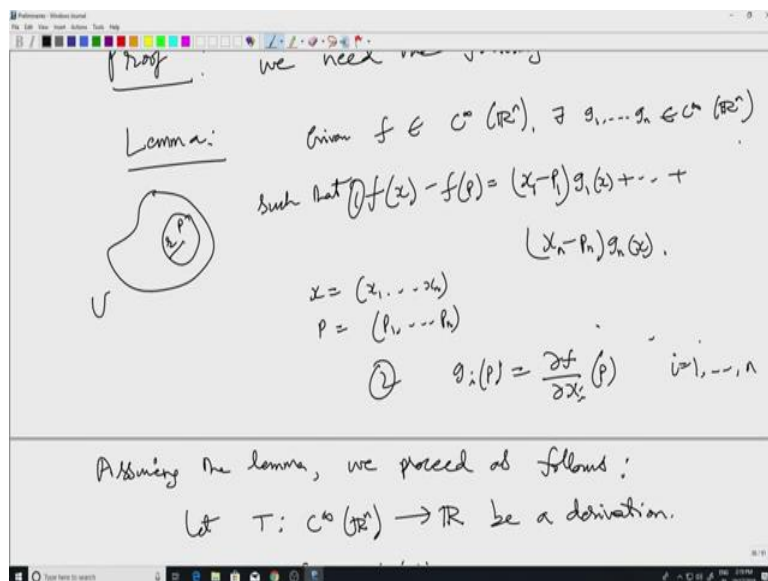
Summary: For each $p \in \mathbb{R}^n$, we have a map $T_p: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ which is linear and satisfies the Leibnitz rule. Such a map is called a derivation on $C^\infty(\mathbb{R}^n)$.

Theorem: Let $T: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ be linear and satisfies the Leibnitz rule i.e.

Now, one thing which we have to note here is that the $C^\infty(\mathbb{R}^n)$ is not a finite dimensional vector space, since we can find linearly independent sets of arbitrarily large size one can just take monomials. And one knows that if it is not a finite dimensional vector space, if any one starts with something which is a vector space which is not finite dimensional then linear maps on such a vector space is also not finite dimensional.

However, with this condition we are looking at not just linear maps but linear maps satisfying this very special property of Leibnitz rule. This severely cuts down on the number of maps. And as it turns out, that is the content of this theorem that the set of derivations is in bijective correspondence with \mathbb{R}^n .

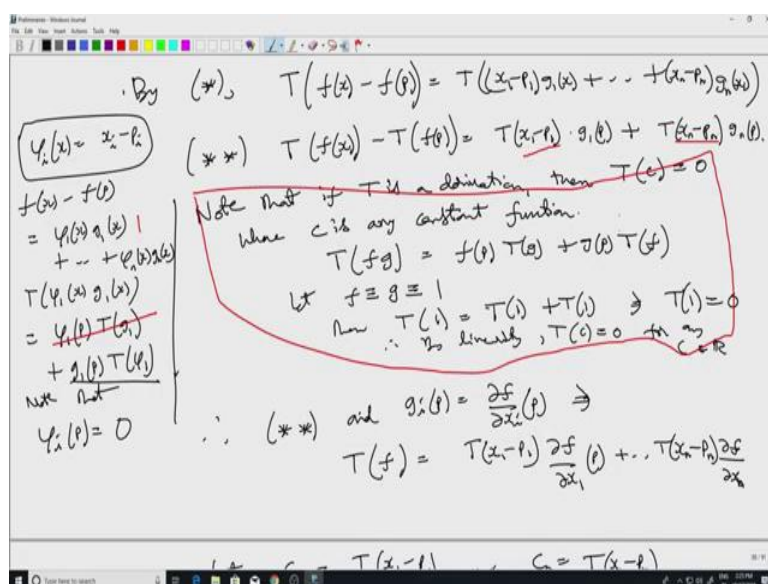
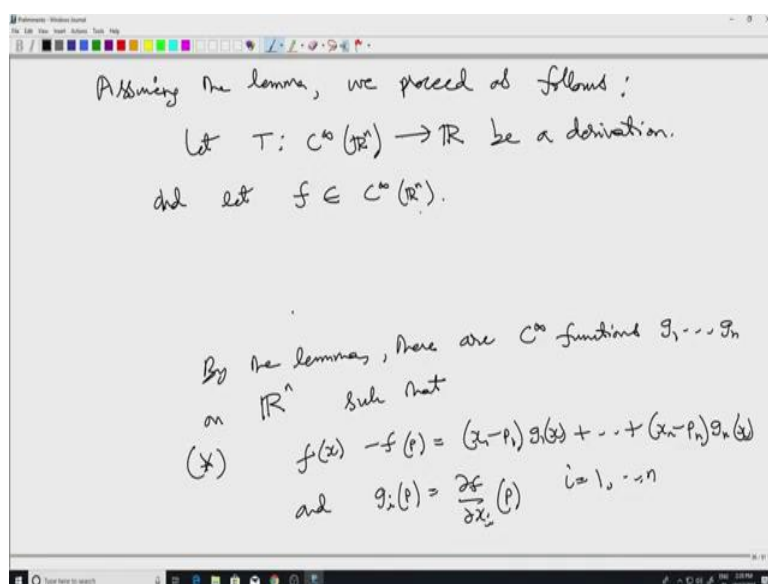
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So, let us prove this, the proof is essentially in just on the following crucial lemma. So the lemma is the following. Given any C^∞ function on \mathbb{R}^n , we can find n C^∞ functions g_1, g_2, \dots, g_n , such that $f(x) - f(p)$ has this form $(x_1 - p_1)g_1(x) + \dots + (x_n - p_n)g_n(x)$. Here of course, as usual this x_i denote the coordinates of x and p_i denote the coordinates of p , such that, this well, there is another such that two things happen.

One is this, the second thing is $g_i(p) = \frac{\partial f}{\partial x_i}(p)$ for all i . So, this is sort of like a Taylor's expansion for f around the point p . It is not really the usual notion of Taylor expansion, but something analogous to that. And this is the main property. So, let see why, if we have this lemma, we will get the result that every derivation is actually a directional derivative. So assuming the lemma we proceed as follows. So I will prove the lemma, once we are done with the proof of main theorem.

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So, we start with a derivation and a function in C^∞ , an element of $C^\infty \mathbb{R}^n$. So, we apply the lemma to this function f , and we get n C^∞ functions g_1, g_2, \dots, g_n , in \mathbb{R}^n such that we have this expansion and the values of the g_i at P_i are given by this. So, let us call this star this thing that this equation star and bi star. Now I am going to just act T on both sides, T of the left side equals to T of the right side, here the what we have to this f each of these terms here, this $x_1 - p_1$ times g_1 etcetera there it is a product of two functions.

So I can this define a function $\varphi_i(x)$ as $x_i - p_i$. If I define such a function, then $f(x) - f(p)$ as the form $\varphi_1(x)g_1(x) + \dots + \varphi_n(x)g_n(x)$ act on both sides. So here I have not in this what I have written down already I have not to use the φ_i notation, or just written it as $x_1 - p_1$ etcetera. So now using linearity, I will get two terms on the left side,

where this f of p is regarded as a constant function. It is a constant map. So this minus this, and here I will get n terms when I use the sum. Then I will be using Leibnitz rule on each of these n terms. As I wrote on the left side, I have a product so and then I can use Leibnitz rule t of $\phi_1 x$ for instance the first term will look like this.

So it would be $f_1 P$ of g_1, g_1 plus ϕ_2 , well, no $g_1 P$ t of ϕ_1 now the thing about this ϕ_i functions is that, note that, ϕ_i at the point P ϕ_i after all $\phi_i x$ is x_i minus p_i just the difference of the i th coordinates. So, when I plug in x equals p , then I just get 0. So, that implies that this term does not even make an appearance. So what I am left with is just this $g_1 P$ times T of ϕ_1 and that is what I have written here t of x_1 minus p_1 times $g_1 P$ etcetera. Now, here on the left side, we have in order to get rid of the second term T of constant we have to make a small observation, that if you have any derivation then T of the constant function C is 0 where C is any constant function and this is seen, this follows immediately from the Leibnitz rule, so I start with a Leibnitz rule in general.

And let us take f and g to be the constant function 1. That would imply that T of 1 here, I will get 1 here I get one so T of 1 is two times T of 1 therefore T of 1 0. Of course, if T of the constant function one is 0 T have any other constant function is 0 by linearity. Now, so the, the small observation here, this thing here will imply that the second term T of fP is goes away. So T of f would be equal to all this right hand side. We can also simplify the right hand side a bit more namely, we know exactly what $g_1 P, g_2 P$ are, the lemmas tells us that $g_i P$ is the $\text{del } f$ by $\text{del } x_i$ at P . Therefore, I can write T of f is T of x_1 minus p_1 $\text{del } f$ by $\text{del } x_1$ etcetera.

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Handwritten notes on a digital whiteboard:

Top section:

$$+ g_2(p) T(v_2) \quad \text{Note Note}$$

$$g_2(p) = 0 \quad \therefore (**) \quad \text{and } g_2(p) = \frac{\partial f}{\partial x_2}(p) \Rightarrow$$

$$T(f) = T(x_1-p) \frac{\partial f}{\partial x_1}(p) + \dots + T(x_n-p) \frac{\partial f}{\partial x_n}(p)$$

Bottom section:

$$\text{Let } c_1 := T(x_1-p), \dots, c_n := T(x_n-p)$$

$$\text{Then } T(f) = c_1 \frac{\partial f}{\partial x_1}(p) + \dots + c_n \frac{\partial f}{\partial x_n}(p)$$

$$\Rightarrow T = c_1 \frac{\partial}{\partial x_1} \Big|_p + \dots + c_n \frac{\partial}{\partial x_n} \Big|_p$$

$$= T_v$$

where $v = (c_1, \dots, c_n)$.

Handwritten notes on a digital whiteboard:

Given $p \in \mathbb{R}^n$ and $v \in \mathbb{R}^n$.

Let $C^\infty(\mathbb{R}^n)$ denote the set of C^∞ functions from \mathbb{R}^n to \mathbb{R} .

$C^\infty(\mathbb{R}^n)$ is a vector space over \mathbb{R} .

Define $T_v: C^\infty(\mathbb{R}^n) \rightarrow \mathbb{R}$ by

$$T_v(f) = v(f) = \frac{d}{dt} f(p + tv) \Big|_{t=0}$$

$$= df_p(v) = \frac{\partial f}{\partial x_1}(p) v_1 + \dots + \frac{\partial f}{\partial x_n}(p) v_n$$

where $v = (v_1, \dots, v_n)$.

Note that $T_v(f) = v(f)$.

1) T_v is linear, i.e.

$$T_v(c_1 f_1 + c_2 f_2) = c_1 T_v(f_1) + c_2 T_v(f_2)$$

Let know, given the point P, these are just some numbers that we get and of course, given the point P and the derivation T I get some numbers. C 1 equals this Cn equals this. So, let us define so what I am doing here is that I am defining this C 1 C 2 etcetera. Once I have that, then I just rewrite it as T of f is C 1 del f by del x 1. So, what we have done is we have essentially shown that we started with an arbitrary derivation. We have just shown that it is a linear combination of partial derivatives purely from the algebraic properties of linearity and Leibnitz rule.

And once we have this, then we are essentially done because this is nothing but in fact, I do not have to write this at all. So at this point, I can just say that this is this thing on the right hand side is exactly Tv of f where v C1, C2, Cn, because remember that the directional

derivative of v of f along the direction v was exactly this expression here, as I have written earlier. So I am talking about this last part here. So this is what I am using.

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Let $c_1 := T(x_1 - t_1), \dots, c_n := T(x_n - t_n)$

Then
$$T(f) = c_1 \frac{\partial f}{\partial x_1}(p) + \dots + c_n \frac{\partial f}{\partial x_n}(p)$$

$$= T_v(f)$$

where $v = (c_1, \dots, c_n)$.

Note that v is unique!

i.e. if $T_v = T_w$, then $v = w$

to see this, note that

$$v_i = T_v(x_i) = T_w(x_i) = w_i \quad \forall i=1, \dots, n$$

Proof of lemma: $f \in C^1(\mathbb{R}^n)$

o.k. let $x \in \mathbb{R}^n$

So, we found a v starting with an arbitrary derivation, we found a v such that $T f$ equals $t v$ of F for all f there was no assumption on f . And of course, it should be noted that this V , as I said earlier, is determined once you know what T is once you given T and a point P . f obviously f should not play a role here. And the next thing is to observe that this v is unique, this is almost immediate from what we have discussed so far.

So, I cannot have $T f$ equals $T v$ as well as equal to $T w$ for two different vectors. The reason is that if $T v$ equals $T w$ for some v and w in \mathbb{R}^n , then v necessarily has to be equal to w . And reason for this is that, if you can recover the coefficients of the vector from the directional

derivative operator just by acting the directional derivative operator along these coordinate functions x_i . So, T_v of x_i again, all I am doing is I am just using this, this way of looking at the directional derivative. So if f equals x_i , only the i th term will survive, all other partial derivatives will be 0. Therefore, I get the coefficient, the i th term will survive, and it will be just equal to C_i . So here, it is v_i equal to $T x_i$ equal to $T w_i$, x_i equal to w_i .

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$$\dots = T_v(f)$$

where $v = (c_1, \dots, c_n)$.

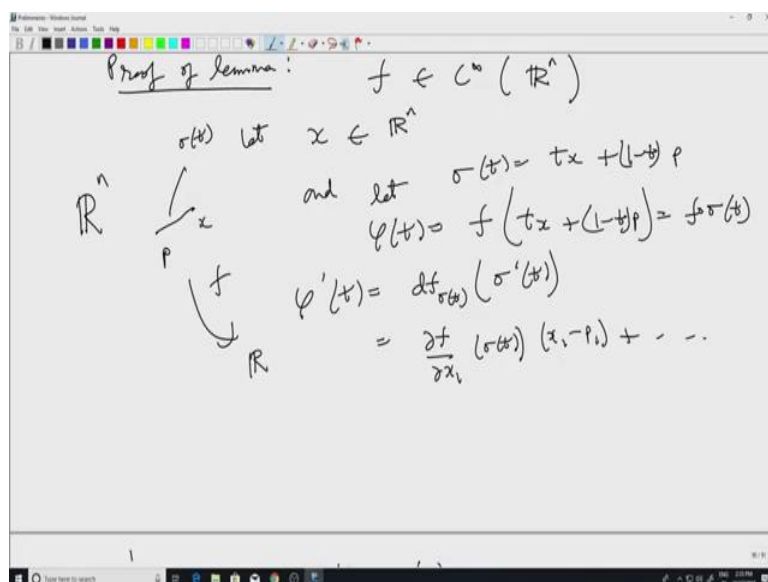
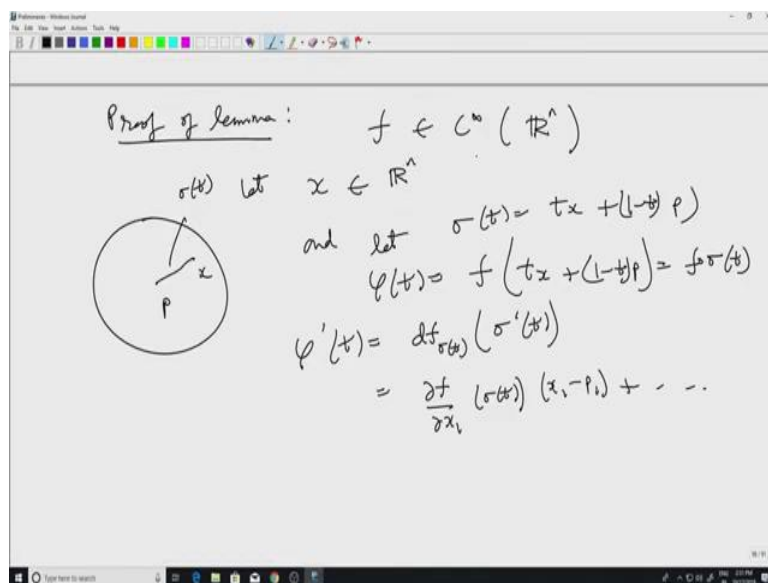
Note that v is unique!
 i.e. if $T_v = T_w$, then $v = w$
 to see this, note that

$$v_i = T_v(x_i) = T_w(x_i) = w_i \quad \forall i=1, \dots, n$$

Proof of lemma: $f \in C^0(\mathbb{R}^n)$
 o/f let $x \in \mathbb{R}^n$

In the previous thing I had put v equal to $C_1 C_2 C_n$, here I am using the coordinates of v to v_1, v_2, v_n , and I mean it is not that important. The point is, the coordinates of the vector can be recovered by knowing the directional derivative operator, all it so that proves what we wanted. So, that we have set up a correspondence between derivations, bijective correspondence between the derivations and elements of \mathbb{R}^n .

(Refer Slide Time: 22:00)



So, but that still leaves the lemma, which I will prove now, so I will us start with C infinity function the goal is to find g_1, g_2, g_n , such that and of course, given the point P, so goal is to find g_1, g_2, g_n with the appropriate equations holding. So, to define this g_i functions to start with any point x in \mathbb{R}^n and then what I will do is look at the straight line sigma T joining so there is no need for this bracket. Let sigma t be the straight line joining x and P. So it is going to start when t is 0, I will start at p when t is 1, I will end at x .

Let me define in another, so this is a map from, sigma t is a map from the real line to \mathbb{R}^n . Now I am going to compose this sigma t with f . So in as, in other words, restrict, f was already a function from here to \mathbb{R} . So this entire thing is \mathbb{R}^n . I am going to restrict f to this line. So f of sigma t, which has this form. If I take the derivative of phi I will just get chain

rule tells me that there is a composition of two functions. So chain rule tells me that it is derivative of f at $\sigma(t)$ acting on $\sigma'(t)$. And, again, writing it out in coordinates will tell me that it has this form.

(Refer Slide Time: 24:23)

The top screenshot shows the following derivations:

$$\int_0^1 \varphi'(t) dt = \varphi(1) - \varphi(0)$$

$$= f(x) - f(p)$$

$$R.H.S. = (x_1 - p_1) \int_0^1 \frac{\partial f}{\partial x_1}(\sigma(t)) dt + \dots - f(x_1 - p_1) \left(\frac{\partial f}{\partial x_1}(\sigma(t)) \right)$$

Below this, a diagram shows a line segment from p to x in \mathbb{R}^n . The path is defined as $\sigma(t) = tx + (1-t)p$. The derivative of the function along this path is given by:

$$g_1(x) = \int_0^1 \frac{\partial f}{\partial x_1}(\sigma(t)) dt = \int_0^1 \frac{\partial f}{\partial x_1}(tx + (1-t)p) dt$$

$$g_1(p) = \frac{\partial f}{\partial x_1}(p)$$

The bottom screenshot shows the "Proof of lemma:" for $f \in C^1(\mathbb{R}^n)$. It defines $\sigma(t) = tx + (1-t)p$ and $\varphi(t) = f(\sigma(t))$. The derivative of φ is then calculated using the chain rule:

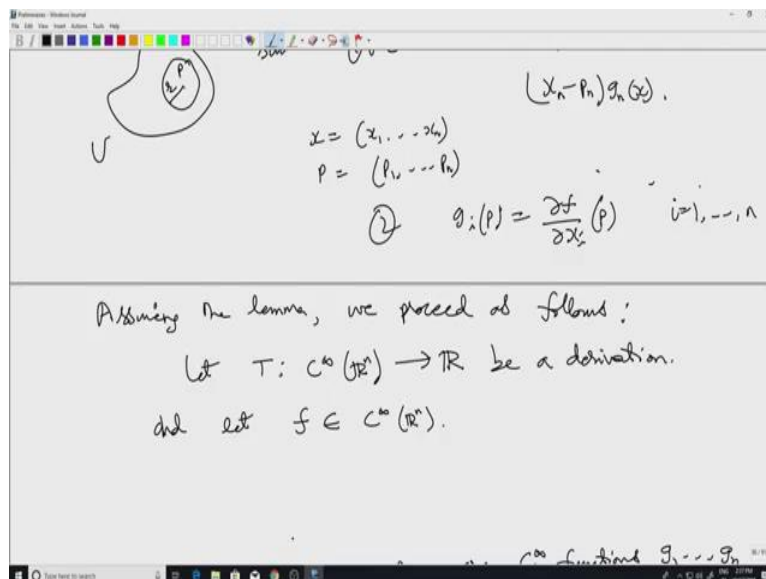
$$\varphi'(t) = \frac{d}{dt} f(\sigma(t)) = \frac{\partial f}{\partial x_1}(\sigma(t)) (x_1 - p_1) + \dots$$

Now let us integrate, now that I know derivative of φ , let us use the fundamental theorem of calculus to and integrate out this. So when I integrate this between 0 and 1, I will get a φ of 1 minus φ of 0. But as we observed, when t is 1, $\sigma(t)$ is x . No, $\sigma(t)$ is p . The line is ending at x . So and when t is 0, I get p , so I just get this f of x minus f of p . On the other hand, no, this is not the LHS, this is the RHS, the right hand side is given by, I just integrate out all the n terms here. And note that this, the each term is a product of two things. One is this partial derivative evaluated at $\sigma(t)$, the other term is x_i minus p_i .

Well, if the $x_i - p_i$ do not depend on T . So I can put it outside the integral sign like this and just integrate out this partial derivative. And I do this for all n of them. And this is what I call $g_1 x$, $g_2 x$, etcetera. These things here, it is not apparent from the way it is written that where the dependence on x is coming from. But it is clear if you remember that σ_t is the straight line joining P and x , or rather more explicitly, σ_t is in fact, given by this formula, $tx + (1-t)P$.

So here the dependence on x is clear. So that is what I am plugging in, in this. So here it is 0 to 1 $\frac{df}{dx_i}$ at $tx + (1-t)P$ dt . And finally, the last statement about g_i was that g_i at P equals $\frac{df}{dx_i}$ at P . And that is clear because when I plug in x equals P here, then this I just get, well, $tx + (1-t)P$ cancel out and there is a p here. So, I just get the $\frac{df}{dx_i}$ at P . So, the dependence on T goes away. So, I can put it outside the integral sign and integrate 1 , integral consider 0 to 1 dt , which is 1 . So I end up with this.

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U (Diagram of a curve with point P)
 $x = (x_1, \dots, x_n)$
 $p = (p_1, \dots, p_n)$
 $g_i(p) = \frac{\partial f}{\partial x_i}(p) \quad i=1, \dots, n$
 $(x_i - p_i) g_i(x)$
 Assuming the lemma, we proceed as follows:
 Let $T: C^0(\mathbb{R}^n) \rightarrow \mathbb{R}$ be a derivation.
 and let $f \in C^0(\mathbb{R}^n)$.

So, that proves the lemma and therefore we have the whole, the theorem as well that every derivation is actually the directional derivative of a unique V in \mathbb{R}^n . And we will use this to this is going to be the motivating, sort of motivation behind defining tangent vectors as derivations on manifolds. So, we will stop here in the next class. I will see how all this can be transferred to a manifold. Okay, thank you.