

Lecture 17: Lie bracket and Lie Algebras

①

M^n smooth, $\mathcal{X}(M) = \left\{ \begin{array}{l} \text{smooth vector} \\ \text{fields on } M \end{array} \right\}$ ← vector space over \mathbb{R}

[Today, give a multiplication, measuring the extent to which the vector fields commute.] ← $\left\{ \begin{array}{l} \text{derivations pg 2} \\ \text{module over } C^\infty(M) \end{array} \right\}$

Lie Algebra: A vector space A with a map

$A \times A \rightarrow A$, denoted $(X, Y) \mapsto [X, Y]$ where

① Bilinear: $\forall X_1, X_2, Y_1, Y_2 \in A, \forall a \in \mathbb{R}$ one has

$$[aX_1 + X_2, Y_1] = a[X_1, Y_1] + [X_2, Y_1]$$

$$[X_1, aY_1 + Y_2] = a[X_1, Y_1] + [X_1, Y_2]$$

② Skew commutative: $[X, Y] = -[Y, X]$

③ $[X, [Y, Z]] + [Y, [Z, X]] + [Z, [X, Y]] = 0$

[Jacobi Identity] $\Leftrightarrow [X, [Y, Z]] - [[X, Y], Z]$

$$= [[Z, X], Y]$$

Ex: $A = \mathbb{R}^3$ with $[v, w] = v \times w$. [On HW!]

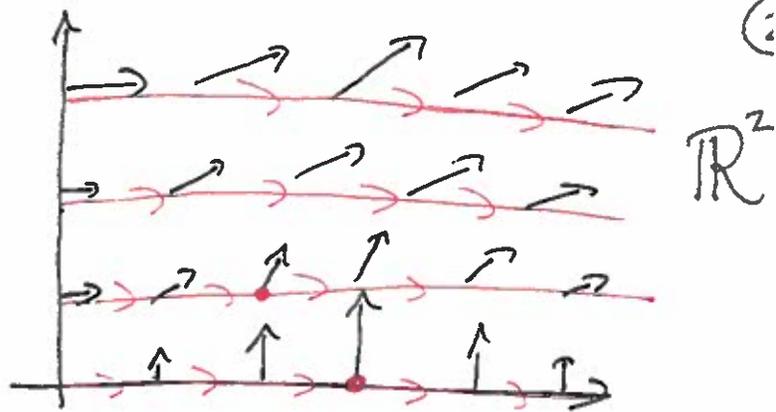
Ex: $A = M_n(\mathbb{R})$ with

$$[X, Y] = XY - YX \quad \text{[Easy check.]}$$

Derivatives of Vector Fields:

$$Y = Y \frac{\partial}{\partial x} + \sin x \frac{\partial}{\partial y}$$

How does Y change as move along x direction?



$$L \frac{\partial}{\partial x} Y = \cos x \frac{\partial}{\partial y} \quad \left. \vphantom{L \frac{\partial}{\partial x} Y} \right\} \text{Another vector field}$$

What about general M ?

Problem: Vectors live in different vector spaces!



$X, Y \in \mathcal{X}(M)$ Let $\Theta: (\mathcal{D} \subseteq \mathbb{R} \times M) \rightarrow M$

be the (partial) flow assoc. to X . That is,

$$\Theta_t(m) = \underbrace{\Theta^{(m)}(t)}_{\text{integral curve for } X \text{ with } \Theta^{(m)}(0) = m} \quad \text{Now define}$$

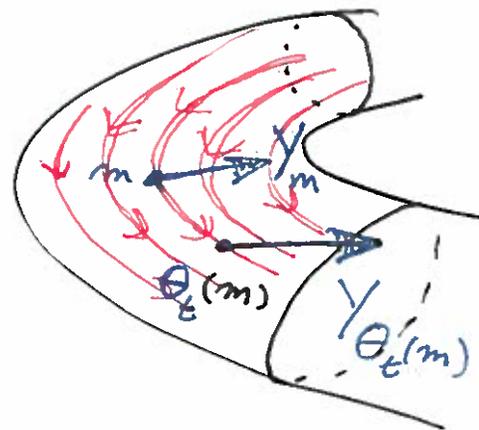
the Lie derivative $L_X Y \in \mathcal{X}(M)$ by

$$(L_X Y)_m = \lim_{t \rightarrow 0} \frac{1}{t} \left(d\theta_{-t}(Y_{\theta_t(m)}) - Y_m \right) \quad (3)$$

$$= \left. \frac{d}{dt} \left(d\theta_{-t}(Y_{\theta_t(m)}) \right) \right|_{t=0}.$$

[Relate back to example.]

To see this is smooth and that it sat. the Lie algebra axioms, will need to define another way...



$$\mathfrak{X}(M) = \left\{ X: C^\infty(M) \rightarrow C^\infty(M) \text{ which is } \mathbb{R}\text{-linear} \right\}$$

and set $X(fg) = (Xf)g + f(Xg)$

First try: $X, Y \in \mathfrak{X}(M)$ consider $(f \mapsto X(Yf))$

Typically, not a derivation.

Second try: $Z: f \mapsto X(Yf) - Y(Xf)$

This is a derivation:

$$\begin{aligned} Z(fg) &= X(Y(fg)) - Y(X(fg)) \\ &= X((Yf)g + f(Yg)) - Y((Xf)g + f(Xg)) \\ &= (X(Yf))g + (Yf)(Xg) + (Xf)(Yg) + f(X(Yg)) \\ &\quad - (Y(Xf)g + (Xf)(Yg) + (Yf)(Xg) + f(Y(Xg))) \\ &= Z(f)g + fZ(g) \end{aligned}$$

Lie
bracket
↓

Define $[X, Y] = Z$. You can easily check that it makes $\mathfrak{X}(M)$ into a Lie algebra.

Thm: $X, Y \in \mathcal{X}(M)$. Then $L_X Y = [X, Y]$

(5)

[Will prove later. For now, let's investigate $[X, Y]$...]

Lemma: $[fX, Y] = f[X, Y] - (Yf)X$

Pf: Suppose $g \in C^\infty(M)$. Then $\Rightarrow [fX, gY]$
 $= fg[X, Y]$
 $+ f(Xg)Y - g(Yf)X$

$$[fX, Y]g = (fX)(Yg) - Y(fXg)$$
$$= fX(Yg) - (Yf)(Xg) - fY(Xg) \quad \square$$

Lemma: For $m \in M$, the vector $[X, Y]_m$ is determined by $X|_U, Y|_U$ for any open $U \ni m$.

If $F: M \rightarrow N$ is a diffeo, $X, Y \in \mathcal{X}(M)$, then

$$[F_*(X), F_*(Y)] = F_*([X, Y])$$

Point: Can calculate in local coordinates.

Suppose on U in \mathbb{R}^n with coordinates (x_1, \dots, x_n) we have

$$X = \sum X_i \frac{\partial}{\partial x_i}$$

$$X_i, Y_i: U \rightarrow \mathbb{R}$$

$$Y = \sum Y_j \frac{\partial}{\partial x_j}$$

$$\left[\frac{\partial}{\partial x_i}, \frac{\partial}{\partial x_j} \right] f = \frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_j} \right) - \frac{\partial}{\partial x_j} \left(\frac{\partial f}{\partial x_i} \right) \quad (6)$$

$$= 0 \text{ since } f \in C^\infty(\mathbb{R}^n)$$

So $\left[\frac{\partial}{\partial x_i}, \frac{\partial}{\partial x_j} \right] = 0$ vector field.

Hence

in previous lemma

$$[X, Y] = \sum_{i,j} \left[X_i \frac{\partial}{\partial x_i}, Y_j \frac{\partial}{\partial x_j} \right]$$

$$= \sum_{i,j} X_i \frac{\partial Y_j}{\partial x_i} \frac{\partial}{\partial x_j} - Y_j \frac{\partial X_i}{\partial x_j} \frac{\partial}{\partial x_i}$$

$$= \sum_{i=1}^n \left(\sum_{j=1}^n X_i \frac{\partial Y_j}{\partial x_j} - Y_j \frac{\partial X_i}{\partial x_j} \right) \frac{\partial}{\partial x_i}$$