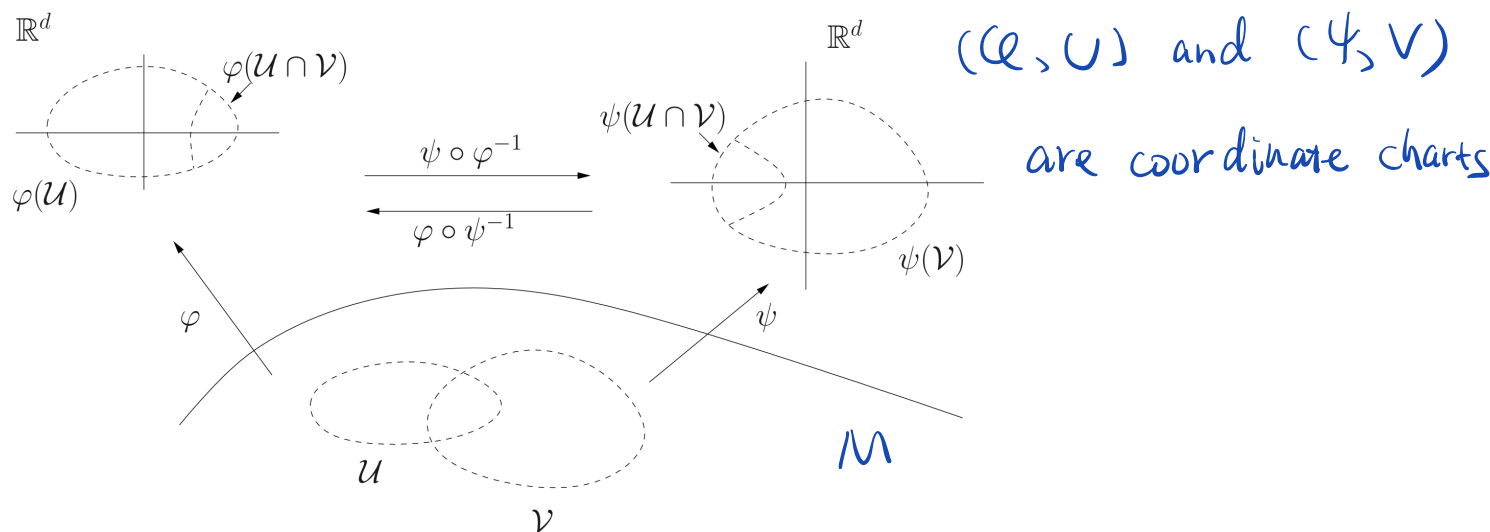


- Recall that the generic definition of a smooth manifold M can be very abstract



- For simplicity, we have consider a manifold that can be described as a level set of some $h(x)$

1) E is d -dim linear space

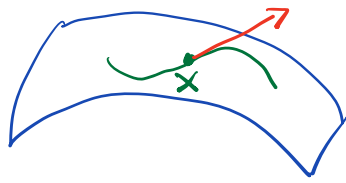
2) $M \subseteq E$ and $\forall x \in E, h(x) = 0 \Leftrightarrow x \in M$

3) $\text{Rank}[Dh(x)] = k, \forall x \in M$

Then $\left\{ \begin{array}{l} M \text{ is an embedded submanifold of dim } d-k \\ T_x M = \text{Ker}[Dh(x)] \text{ is the tangent space at } x. \\ T_x M \text{ is a linear space of dim } d-k. \end{array} \right.$

- The general definition of tangent space of a manifold $M \subseteq E$

$$T_x M = \{ C'(0) \mid C: (-\epsilon, \epsilon) \rightarrow M \text{ is smooth \& } C(0) = x \}$$



$c(t)$ is a curve on M

$c'(t)$ is tangent to the curve

Theorem If $M \subseteq \mathbb{E}^n$ is describable by $h(x) = 0$, then
 $\text{Ker}[Dh(x)] = T_x M$

Proof: ① $\forall v \in T_x M, \exists$ a curve $C: (-\epsilon, \epsilon) \rightarrow M$

$$\text{s.t. } \begin{cases} C(0) = x \\ C'(0) = v \end{cases}$$

Since $C(t) \in M, \forall t$, we have $h(C(t)) = 0$

$$\Rightarrow 0 = \frac{d}{dt} h(C(t)) = \langle \text{grad } h(C(t)), C'(t) \rangle = Dh(C(t)) [C'(t)]$$

$$\left(Dh(C(t)) [C'(t)] = \lim_{s \rightarrow 0} \frac{h(C(t) + s C'(t)) - h(C(t))}{s} \right)$$

$$\Rightarrow v = C'(0) \in \text{Ker}[Dh(x)] \Rightarrow T_x M \subseteq \text{Ker}[Dh(x)]$$

② We can show $T_x M$ contains a linear subspace of dimension $n-k$ (skipped, see book B6)

$$\Rightarrow T_x M = \text{Ker}[Dh(x)]$$

- Definition/Notation

M is a manifold

$T_x M$ is the tangent space of M at x

Any element in $T_x M$ is a tangent vector at x

Tangent Bundle TM

$$TM = \{ \underbrace{V_x}_{\downarrow} \in T_x M : \forall x \in M \} = \bigcup_{x \in M} T_x M$$

subscript x denotes it's a tangent vector at x

For convenience, we sometimes ignore the subscript.

A vector field on M is a map

$$V : M \longrightarrow TM$$

s.t. $V(x) \in T_x M$ for any $x \in M$

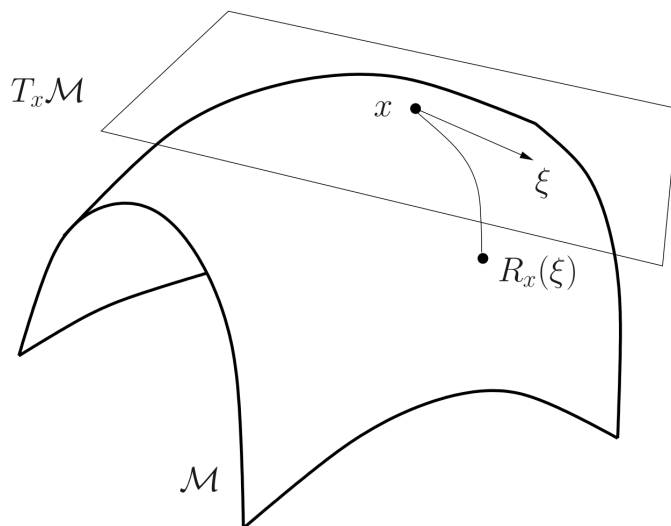
A Retraction on M is a smooth map

$$R : TM \longrightarrow M$$

$$V_x \longmapsto R_x(V)$$

Such that each curve $c(t) = R_x(tV)$ satisfies

$$\begin{cases} c(0) = x \\ c'(0) = V \end{cases}$$



$$R : TM \longrightarrow M$$

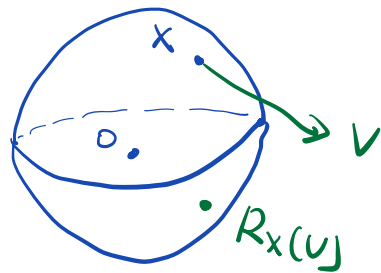
$$V_x \longmapsto R_x(V_x)$$

$$c(t) = R_x(t\xi)$$

$$c(0) = x$$

$$c'(0) = \xi$$

Examples of Retraction on S^{n-1}



$$\begin{aligned} \textcircled{1} \quad R_x(v) &= \frac{x+v}{\|x+v\|} = \frac{x+v}{\sqrt{\|x+v\|^2}} \\ &= \frac{x+v}{\sqrt{1+\|v\|^2}} \end{aligned}$$

$$\textcircled{2} \quad R_x(v) = \cos(\|v\|)x + \frac{\sin(\|v\|)}{\|v\|}v$$

$c(t) = R_x(tv)$ is the great circle passing x with $c'(0) = v$.

Def A metric is a choice of inner product $\langle \cdot, \cdot \rangle_x$ for $T_x M$ at each $x \in M$.

Def A metric is Riemannian if $\langle \cdot, \cdot \rangle_x$ is smooth w.r.t. x i.e. $\langle V(x), W(x) \rangle_x$ is smooth for smooth vector fields V, W .

Def Let $g_x(\cdot, \cdot) = \langle \cdot, \cdot \rangle_x$ be a Riemannian metric (M, g) is a Riemannian manifold.

Example: $\langle \cdot, \cdot \rangle$ is the inner product in \mathbb{E}
 M is an embedded submanifold in \mathbb{E}
Then $\langle \cdot, \cdot \rangle$ is a Riemannian metric of M .

Example: \mathbb{E} is also a manifold

Differential of a map

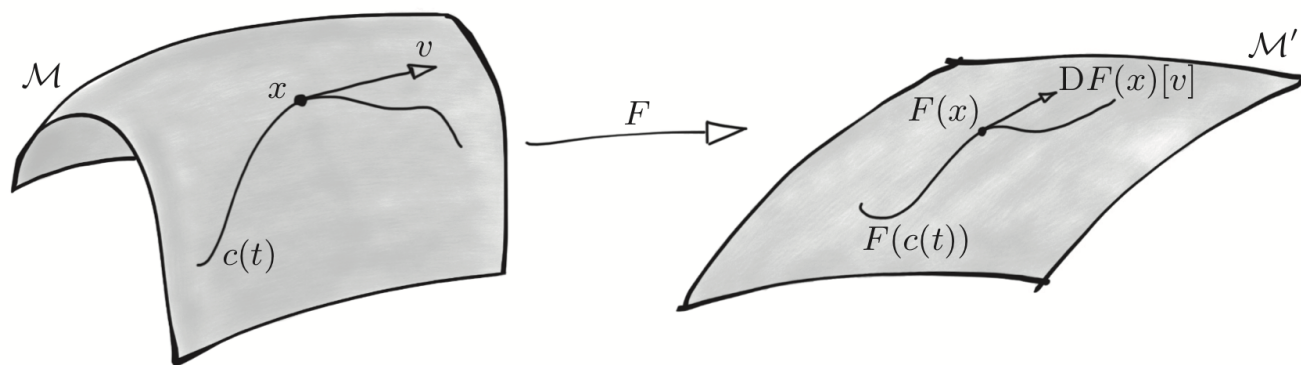
- $F: M \rightarrow M'$ is smooth function where M, M' are manifolds

The differential is the linear map

$$DF(x): T_x M \rightarrow T_{F(x)} M' \quad \text{defined by}$$

$$DF(x)[v] = \left. \frac{d}{dt} F(c(t)) \right|_{t=0} = (F \circ c)'(0)$$

where $c(t)$ is a smooth curve on M $\begin{cases} c(0) = x \\ c'(0) = v \end{cases}$



Facts: ① $DF(x)[v]$ is unique, regardless of choice of $c(t)$
② $DF(x)$ is a linear map

- Let $M \subset \mathcal{E}$ and $M' \subset \mathcal{E}'$ be embedded submanifolds in \mathcal{E} and \mathcal{E}'

Then a smooth map $F: M \rightarrow M'$ has a smooth

extension $\bar{F}: \mathcal{E} \rightarrow \mathcal{E}'$

Example: $\min_{x \in S^m} f(x) = x^T A x$, $f: S^{n-1} \rightarrow \mathbb{R}$ can be regarded as
 $\bar{f}: \mathbb{R}^n \rightarrow \mathbb{R}$

Notice that $F(c(t)) = \bar{F}(c(t))$ for a curve $c(t)$ in M .

$$\frac{d}{dt} \bar{F}(c(t)) \Big|_{t=0} = D\bar{F}(c(0)) [c'(0)] = D\bar{F}(x_0) [v]$$

||

$$\frac{d}{dt} F(c(t)) \Big|_{t=0} = DF(x_0) [v]$$

For a smooth map $\bar{F}: \mathcal{E} \rightarrow \tilde{F}$

its differential at $x_0 \in U$ is $D\bar{F}(x_0): \mathcal{E} \rightarrow \tilde{F}$ defined by

$$\forall u \in \mathcal{E}, D\bar{F}(x_0)[u] = \frac{d}{dt} \bar{F}[x_0 + tu] \Big|_{t=0} = \lim_{t \rightarrow 0} \frac{\bar{F}(x_0 + tu) - \bar{F}(x_0)}{t}$$

$$\begin{aligned} \bar{F}(x) = x^T A x &\Rightarrow D\bar{F}(x)[v] = 2x^T A v & \begin{cases} f: S^m \rightarrow \mathbb{R} \\ \bar{F}: \mathbb{R}^n \rightarrow \mathbb{R} \end{cases} \\ \Rightarrow Df(x)[v] = 2x^T A v & \end{aligned}$$

$$\forall v \in T_x S^{n-1} = \{v \in \mathbb{R}^n : x^T v = 0\}$$

$$D\bar{F}(x): \mathbb{R}^n \rightarrow \mathbb{R} \quad \text{but} \quad Df(x): T_x S^{n-1} \rightarrow \mathbb{R}$$

Riemannian Gradient

For a smooth $f: M \rightarrow \mathbb{R}$ on a Riemannian manifold (M, g)

the Riemannian gradient of f is the vector field $\text{grad} f$

$$\text{defined by } g_x(\text{grad} f(x), v_x) = Df(x)[v_x], \quad \begin{matrix} \forall x \in M \\ \forall v_x \in T_x M \end{matrix}$$

Remark: $\text{grad} f(x)$ is a tangent vector

For a smooth function $\bar{f} : \mathcal{E} \rightarrow \mathbb{R}$,
the Euclidean gradient $\nabla \bar{f} : \mathcal{E} \rightarrow \mathcal{E}$ is defined by

$$\langle \nabla \bar{f}(x), v \rangle = D\bar{f}(x)[v] \quad \forall x, v \in \mathcal{E}$$

If $f : M \rightarrow \mathbb{R}$ can be extended to

$\bar{f} : \mathcal{E} \rightarrow \mathbb{R}$, then

$\nabla \bar{f}$ is the Euclidean gradient of \bar{f}

$\text{grad } f$ is the Riemannian gradient of f

and $\forall v \in T_x M$, $g_x(\text{grad } f(x), v) = Df(x)[v]$

$$= D\bar{f}(x)[v]$$

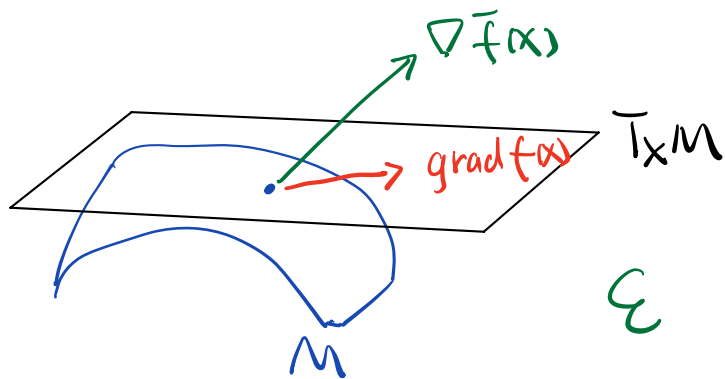
$$= \langle \nabla \bar{f}(x), v \rangle$$

Notice that the metric g_x does not need to be the same as \langle, \rangle .

Riemannian Gradient of a Riemannian submanifold of \mathcal{E}
meaning metric is \langle, \rangle in \mathcal{E}

$$\langle \text{grad } f(x), v \rangle = g_x(\text{grad } f(x), v) = \langle \nabla \bar{f}(x), v \rangle$$

$\Rightarrow \text{grad } f(x)$ is the projection of $\nabla \bar{f}(x)$ onto $T_x M$



$$\langle \nabla \bar{f}(x) - \text{grad} f(x), v \rangle = 0$$

$$\forall v \in T_x M$$

$$\Rightarrow [\nabla \bar{f}(x) - \text{grad} f(x)] \perp T_x M$$



$$[\nabla \bar{f}(x) - w] \perp T_x M$$

\Leftrightarrow Critical point

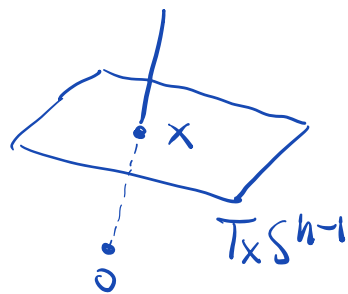
$$w = \underset{v \in T_x M}{\text{argmin}} \|\nabla \bar{f}(x) - v\|^2 = \underset{v \in T_x M}{\text{argmin}} \langle \nabla \bar{f}(x) - v, \nabla \bar{f}(x) - v \rangle$$

Example ① Regard S^{n-1} as a Riemannian submanifold of \mathbb{R}^n

$$\left. \begin{aligned} f(x) &= \frac{1}{2} x^T A x \text{ defined on } S^{n-1} \\ \bar{f}(x) &= \frac{1}{2} x^T A x \text{ defined on } \mathbb{R}^n \end{aligned} \right\} A^T = A$$

$$\nabla \bar{f}(x) = Ax$$

Projection of u onto $T_x S^{n-1}$



$$P_{T_x S^{n-1}}(u) = x - \frac{\langle x, u \rangle}{\langle x, x \rangle} x, \quad x \in S^{n-1}$$

$$= u - \langle x, u \rangle x$$

$$= (I - xx^T)u$$

$$\begin{bmatrix} x \\ \vdots \\ x \end{bmatrix} \begin{bmatrix} x^T \\ \vdots \\ x^T \end{bmatrix} \begin{bmatrix} u \\ \vdots \\ u \end{bmatrix}$$

$$\text{grad} f(x) = P_{T_x S^{n-1}}(\nabla \bar{f}(x))$$

$$= (I - x x^T) Ax = Ax - (x^T Ax) x$$

Remark: $\text{grad } f(x) = 0 \Leftrightarrow Ax = \underbrace{(x^T Ax)}_{\text{scalar}} x$

So critical point on $S^m \Leftrightarrow$ eigenvectors

② $M = \Delta_+^{n-1} = \{x \in \mathbb{R}^n : x_i > 0, \sum_{i=1}^n x_i = 1\}$

relative interior of the simplex or probability manifold

M is an embedded manifold of \mathbb{R}^n

$$T_x M = \left\{ v \in \mathbb{R}^n : \sum_{i=1}^n v_i = 0 \right\}$$

The Fisher-Rao metric

$$g_x(u, v) = \sum_{i=1}^n \frac{u_i v_i}{x_i}$$

$$f : M \rightarrow \mathbb{R}$$

$$\bar{f} : \mathbb{R}^n \rightarrow \mathbb{R} \quad \nabla \bar{f}(x)$$

$$g_x(\text{grad } f(x), v) = \langle \nabla \bar{f}(x), v \rangle = v^T \cdot \nabla \bar{f}(x)$$

To find an expression

Let $\text{grad } f(x) = u \in T_x M$

$\frac{u}{x}$ denotes the vector $\left[\frac{u_1}{x_1} \quad \dots \quad \frac{u_n}{x_n} \right]^T$

$$v^T \cdot \frac{\text{grad } f(x)}{x} = v^T \cdot \nabla \bar{f}(x) \quad \forall v \in T_x M$$

Given $w \in \mathbb{R}^n$, solve $v^T u = v^T w$, $\forall v \in T_x M$

$$\Rightarrow v^T (u - w) = 0$$

$$\Rightarrow P_{T_x M} (u - w) = 0$$

$$\Rightarrow u - w \parallel \begin{bmatrix} | \\ | \\ | \end{bmatrix}$$

$$\Rightarrow u = w + \lambda \mathbf{1}, \lambda \in \mathbb{R}$$

$$\Rightarrow \frac{\text{grad } f(x)}{x} = \nabla \bar{f}(x) + \lambda \mathbf{1}$$

$$\Rightarrow \text{grad } f(x) = \underline{x} \circ \nabla \bar{f}(x) + \lambda x$$

entry wise product

$$\text{grad } f(x) \in T_x M \Rightarrow \sum_{i=1}^n (x_i [\nabla \bar{f}(x)]_i + \lambda x_i) = 0$$

$$\Rightarrow \lambda = - \sum_{i=1}^n x_i [\nabla \bar{f}(x)]_i$$

$$\Rightarrow \text{grad } f(x) = x \circ \nabla \bar{f}(x) - \langle x, \nabla \bar{f}(x) \rangle x$$