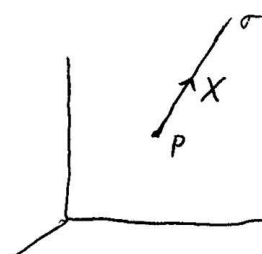


Chapter 5. The Second Fundamental Form

Directional Derivatives in \mathbb{R}^3 .

Let $f : U \subset \mathbb{R}^3 \rightarrow \mathbb{R}$ be a smooth function defined on an open subset of \mathbb{R}^3 . Fix $p \in U$ and $X \in T_p \mathbb{R}^3$. The *directional derivative* of f at p in the direction X , denoted $D_X f$ is defined as follows. Let $\sigma : \mathbb{R} \rightarrow \mathbb{R}^3$ be the parameterized straight line, $\sigma(t) = p + tX$. Note $\sigma(0) = p$ and $\sigma'(0) = X$. Then,

$$\begin{aligned} D_X f &= \frac{d}{dt} f \circ \sigma(t) \Big|_{t=0} \\ &= \frac{d}{dt} f(p + tX) \Big|_{t=0} \\ &= \left(\lim_{t \rightarrow 0} \frac{f(p + tX) - f(p)}{t} \right). \end{aligned}$$


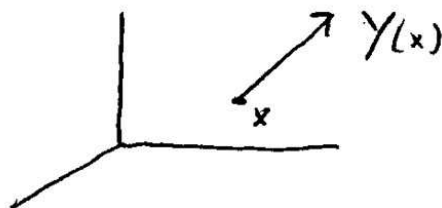
Fact: The directional derivative is given by the following formula,

$$\begin{aligned} D_X f &= X \cdot \nabla f(p) \\ &= (X^1, X^2, X^3) \cdot \left(\frac{\partial f}{\partial x}(p), \frac{\partial f}{\partial y}(p), \frac{\partial f}{\partial z}(p) \right) \\ &= X^1 \frac{\partial f}{\partial x^1}(p) + X^2 \frac{\partial f}{\partial x^2}(p) + X^3 \frac{\partial f}{\partial x^3}(p) \\ &= \sum_{i=1}^3 X^i \frac{\partial f}{\partial x^i}(p). \end{aligned}$$

Proof Chain rule!

Vector Fields on \mathbb{R}^3 . A vector field on \mathbb{R}^3 is a rule which assigns to each point of \mathbb{R}^3 a vector at the point,

$$x \in \mathbb{R}^3 \rightarrow Y(x) \in T_x \mathbb{R}^3$$



Analytically, a vector field is described by a mapping of the form,

$$\begin{aligned} Y : U &\subset \mathbb{R}^3 \rightarrow \mathbb{R}^3, \\ Y(x) &= (Y^1(x), Y^2(x), Y^3(x)) \in T_x \mathbb{R}^3. \end{aligned}$$

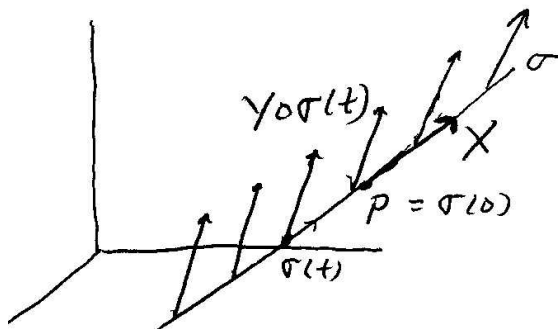
Components of Y : $Y^i : U \rightarrow \mathbb{R}, i = 1, 2, 3$.

Ex. $Y : \mathbb{R}^3 \rightarrow \mathbb{R}^3, Y(x, y, z) = (y + z, z + x, x + y)$. E.g., $Y(1, 2, 3) = (5, 4, 3)$, etc. $Y^1 = y + z, Y^2 = z + x$, and $Y^3 = x + y$.

The directional derivative of a vector field is defined in a manner similar to the directional derivative of a function: Fix $p \in U, X \in T_p \mathbb{R}^3$. Let $\sigma : \mathbb{R} \rightarrow \mathbb{R}^3$ be the parameterized line $\sigma(t) = p + tX$ ($\sigma(0) = p, \sigma'(0) = X$). Then $t \rightarrow Y \circ \sigma(t)$ is a vector field along σ in the sense of the definition in Chapter 2. Then, the directional derivative of Y in the direction X at p , is defined as,

$$D_X Y = \frac{d}{dt} Y \circ \sigma(t)|_{t=0}$$

I.e., to compute $D_X Y$, restrict Y to σ to obtain a vector valued function of t , and differentiate with respect to t .



In terms of components, $Y = (Y^1, Y^2, Y^3)$,

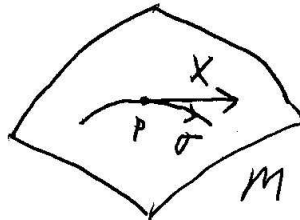
$$\begin{aligned} D_X Y &= \frac{d}{dt} (Y^1 \circ \sigma(t), Y^2 \circ \sigma(t), Y^3 \circ \sigma(t))|_{t=0} \\ &= \left(\frac{d}{dt} Y^1 \circ \sigma(t)|_{t=0}, \frac{d}{dt} Y^2 \circ \sigma(t)|_{t=0}, \frac{d}{dt} Y^3 \circ \sigma(t)|_{t=0} \right) \\ &= (D_X Y^1, D_X Y^2, D_X Y^3). \end{aligned}$$

Directional derivatives on surfaces.

Let M be a surface, and let $f : M \rightarrow \mathbb{R}$ be a smooth function on M . Recall, this means that $\hat{f} = f \circ \mathbf{x}$ is smooth for all proper patches $\mathbf{x} : U \rightarrow M$ in M .

Def. For $p \in M$, $X \in T_p M$, the directional derivative of f at p in the direction X , denoted $\nabla_X f$, is defined as follows. Let $\sigma : (-\epsilon, \epsilon) \rightarrow M \subset \mathbb{R}^3$ be any smooth curve in M such that $\sigma(0) = p$ and $\sigma'(0) = X$. Then,

$$\nabla_X f = \left. \frac{d}{dt} f \circ \sigma(t) \right|_{t=0}$$



I.e., to compute $\nabla_X f$, restrict f to σ and differentiate with respect to parameter t .

Proposition. *The directional derivative is well-defined, i.e. independent of the particular choice of σ .*

Proof. Let $\mathbf{x} : U \rightarrow M$ be a proper patch containing p . Express σ in terms of coordinates in the usual manner,

$$\sigma(t) = \mathbf{x}(u^1(t), u^2(t)).$$

By the chain rule,

$$\frac{d\sigma}{dt} = \sum \frac{du_i}{dt} \mathbf{x}_i \quad \left(\mathbf{x}_i = \frac{\partial \mathbf{x}}{\partial u_i} \right)$$

$X \in T_p M \Rightarrow X = \sum X^i \mathbf{x}_i$. The initial condition, $\frac{d\sigma}{dt}(0) = X$ then implies

$$\frac{du^i}{dt}(0) = X^i, \quad i = 1, 2.$$

Now,

$$\begin{aligned} f \circ \sigma(t) &= f(\sigma(t)) = f(\mathbf{x}(u^1(t), u^2(t))) \\ &= f \circ \mathbf{x}(u^1(t), u^2(t)) \\ &= \hat{f}(u^1(t), u^2(t)). \end{aligned}$$

Hence, by the chain rule,

$$\begin{aligned}\frac{d}{dt}f \circ \sigma(t) &= \frac{\partial \hat{f}}{\partial u^1} \frac{du^1}{dt} + \frac{\partial \hat{f}}{\partial u^2} \frac{du^2}{dt} \\ &= \sum_i \frac{\partial \hat{f}}{\partial u^i} \frac{du^i}{dt} = \sum_i \frac{du^i}{dt} \frac{\partial \hat{f}}{\partial u^i}.\end{aligned}$$

Therefore,

$$\begin{aligned}\nabla_X f &= \frac{d}{dt}f \circ \sigma(t)|_{t=0} \\ &= \sum_i \frac{du^i}{dt}(0) \frac{\partial \hat{f}}{\partial u^i}(u^1, u^2), \quad (p = \mathbf{x}(u^1, u^2)) \\ \nabla_X f &= \sum_i X^i \frac{\partial \hat{f}}{\partial u^i}(u^1, u^2),\end{aligned}$$

or simply,

$$\begin{aligned}\nabla_X f &= \sum X^i \frac{\partial \hat{f}}{\partial u^i} \\ &= X^1 \frac{\partial \hat{f}}{\partial u^1} + X^2 \frac{\partial \hat{f}}{\partial u^2}. \quad (*)\end{aligned}$$

Ex. Let $X = \mathbf{x}_1$. Since $\mathbf{x}_1 = 1 \cdot \mathbf{x}_1 + 0 \cdot \mathbf{x}_2$, $X^1 = 1$ and $X^2 = 0$. Hence the above equation implies, $\nabla_{\mathbf{x}_1} f = \frac{\partial \hat{f}}{\partial u^1}$. Similarly, $\nabla_{\mathbf{x}_2} f = \frac{\partial \hat{f}}{\partial u^2}$. I.e.,

$$\nabla_{\mathbf{x}_i} f = \frac{\partial \hat{f}}{\partial u^i}, \quad i = 1, 2.$$

The following proposition summarizes some basic properties of directional derivatives in surfaces.

Proposition

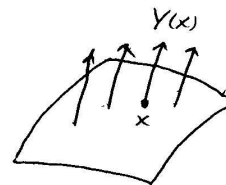
- (1) $\nabla_{(aX+bY)} f = a\nabla_X f + b\nabla_Y f$
- (2) $\nabla_X(f+g) = \nabla_X f + \nabla_X g$
- (3) $\nabla_X fg = (\nabla_X f)g + f(\nabla_X g)$

Exercise 5.1. Prove this proposition.

Vector fields along a surface.

A vector field along a surface M is a rule which assigns to each point of M a vector at that point,

$$x \in M \rightarrow Y(x) \in T_x \mathbb{R}^3$$



N.B. $Y(x)$ need not be tangent to M .

Analytically vector fields along a surface M are described by mappings.

$$Y : M \rightarrow \mathbb{R}^3$$

$$Y(x) = (Y^1(x), Y^2(x), Y^3(x)) \in T_x \mathbb{R}^3$$

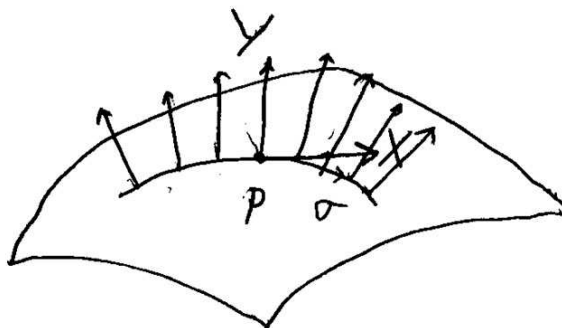
Components of Y : $Y^i : M \rightarrow \mathbb{R}$, $i = 1, 2, 3$. We say Y is *smooth* if its component functions are smooth.

The directional derivative of a vector field along M is defined in a manner similar to the directional derivative of a function defined on M .

Given a vector field along M , $Y : M \rightarrow \mathbb{R}^3$, for $p \in M$, $X \in T_p M$, the directional derivative of Y in the direction X , denoted $\nabla_X Y$, is defined as,

$$\nabla_X Y = \frac{d}{dt} Y \circ \sigma(t) |_{t=0}$$

where $\sigma : (-\epsilon, \epsilon) \rightarrow M$ is a smooth curve in M such that $\sigma(0) = p$ and $\frac{d\sigma}{dt}(0) = X$.



I.e. to compute $\nabla_X Y$, restrict Y to σ to obtain a vector valued function of t - then differentiate with respect to t .

Fact If $Y(x) = (Y^1(x), Y^2(x), Y^3(x))$ then,

$$\nabla_X Y = (\nabla_X Y^1, \nabla_X Y^2, \nabla_X Y^3)$$

Proof. Exercise.

Surface Coordinate Expression. Let $\mathbf{x} : U \rightarrow M$ be a proper patch in M containing p . Let $X \in T_p M$, $X = \sum_i x^i \mathbf{x}_i$. An argument like that for functions on M shows,

$$\nabla_X Y = \sum_{i=1}^2 X^i \frac{\partial \hat{Y}}{\partial u^i}(u^1, u^2), \quad (p = \mathbf{x}(u^1, u^2))$$

where $\hat{Y} = Y \circ \mathbf{x} : U \rightarrow \mathbb{R}^3$ is Y expressed in terms of coordinates.

Exercise 5.2. Derive the expression above for $\nabla_X Y$. In particular, show

$$\nabla_{\mathbf{x}_i} Y = \frac{\partial \hat{Y}}{\partial u^i}, \quad i = 1, 2.$$

Some basic properties are described in the following proposition.

Proposition

- (1) $\nabla_{aX+bY} Z = a\nabla_X Z + b\nabla_Y Z$
- (2) $\nabla_X (Y + Z) = \nabla_X Y + \nabla_X Z$
- (3) $\nabla_X (fY) = (\nabla_X f)Y + f\nabla_X Y$
- (4) $\nabla_X \langle Y, Z \rangle = \langle \nabla_X Y, Z \rangle + \langle Y, \nabla_X Z \rangle$

The Weingarten Map and the 2nd Fundamental Form.

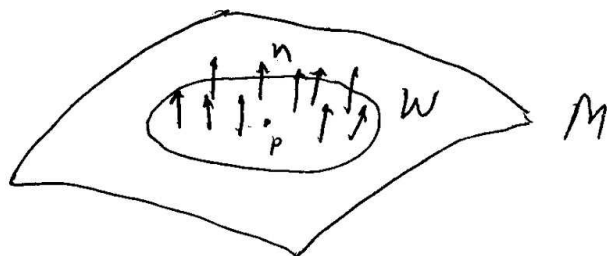
We are interested in studying the *shape* of surfaces in \mathbb{R}^3 . Our approach (essentially due to Gauss) is to study how the unit normal to the surface “wiggles” along the surface.



The objects which describe the shape of M are:

1. The *Weingarten Map*, or *shape operator*. For each $p \in M$ this is a certain linear transformation $L : T_p M \rightarrow T_p M$.
2. The *second fundamental form*. This is a certain bilinear form $\mathcal{L} : T_p M \times T_p M \rightarrow \mathbb{R}$ associated in a natural way with the Weingarten map.

We now describe the Weingarten map. Fix $p \in M$. Let $n : W \rightarrow \mathbb{R}^3$, $p \in W \rightarrow n(p) \in T_p \mathbb{R}^3$, be a smooth *unit normal* vector field defined along a neighborhood W of p .



Remarks

1. n can always be constructed by introducing a proper patch $\mathbf{x} : U \rightarrow M$, $\mathbf{x} = \mathbf{x}(u^1, u^2)$ containing p :

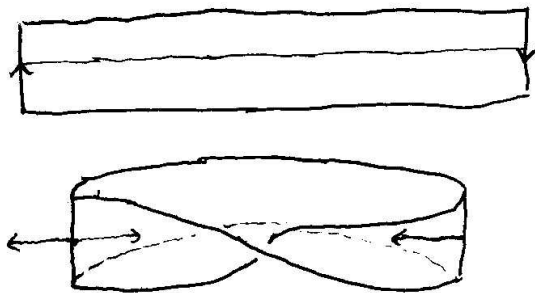
$$\hat{n} = \frac{\mathbf{x}_1 \times \mathbf{x}_2}{|\mathbf{x}_1 \times \mathbf{x}_2|},$$

$\hat{n} : U \rightarrow \mathbb{R}^3$, $\hat{n} = \hat{n}(u^1, u^2)$. Then, $n = \hat{n} \circ \mathbf{x}^{-1} : \mathbf{x}(U) \rightarrow \mathbb{R}^3$ is a smooth unit normal v.f. along $\mathbf{x}(U)$.

2. The choice of n is not quite unique: $n \rightarrow -n$; choice of n is unique “up to sign”

3. A smooth unit normal field n always exists in a neighborhood of any given point p , but it may not be possible to extend n to all of M . This depends on whether or not M is an *orientable* surface.

Ex. Möbius band.



Lemma. Let M be a surface, $p \in M$, and n be a smooth unit normal vector field defined along a neighborhood $W \subset M$ of p . Then for any $X \in T_p M$, $\nabla_X n \in T_p M$.

Proof. It suffices to show that $\nabla_X n$ is perpendicular to n . $|n| = 1 \Rightarrow \langle n, n \rangle = 1 \Rightarrow$

$$\begin{aligned}\nabla_X \langle n, n \rangle &= \nabla_X 1 \\ \langle \nabla_X n, n \rangle + \langle n, \nabla_X n \rangle &= 0 \\ 2\langle \nabla_X n, n \rangle &= 0\end{aligned}$$

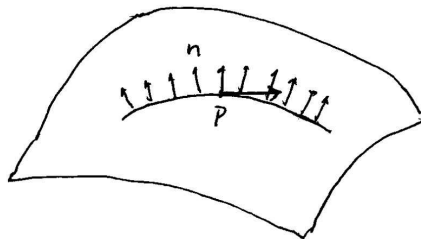
and hence $\nabla_X n \perp n$.

Def. Let M be a surface, $p \in M$, and n be a smooth unit normal v.f. defined along a nbd $W \subset M$ of p . The *Weingarten Map* (or *shape operator*) is the map $L : T_p M \rightarrow T_p M$ defined by,

$$L(X) = -\nabla_X n.$$

Remarks

1. The minus sign is a convention – will explain later.
2. $L(X) = -\nabla_X n = -\frac{d}{dt} n \circ \sigma(t)|_{t=0}$



Lemma: $L : T_p M \rightarrow T_p M$ is a *linear* map, i.e.,

$$L(aX + bY) = aL(X) + bL(Y).$$

for all $X, Y \in T_p M$, $a, b \in \mathbb{R}$.

Proof. Follows from properties of directional derivative,

$$\begin{aligned} L(aX + bY) &= -\nabla_{aX+bY} n \\ &= -[a\nabla_X n + b\nabla_Y n] \\ &= a(-\nabla_X n) + b(-\nabla_Y n) \\ &= aL(X) + bL(Y). \end{aligned}$$

Ex. Let M be a *plane* in \mathbb{R}^3 :

$$M : ax + by + cz = d$$

Determine the Weingarten Map at each point of M . Well,

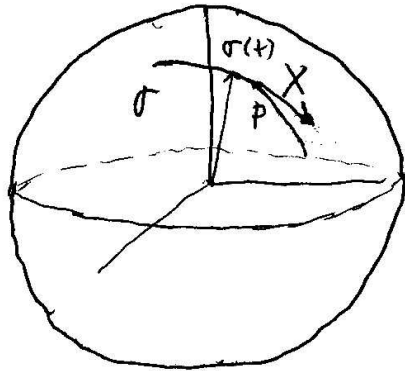
$$n = \frac{(a, b, c)}{\sqrt{a^2 + b^2 + c^2}} = \left(\frac{a}{\lambda}, \frac{b}{\lambda}, \frac{c}{\lambda} \right),$$

where $\lambda = \sqrt{a^2 + b^2 + c^2}$. Hence,

$$\begin{aligned} L(X) &= -\nabla_X n = -\nabla_X \left(\frac{a}{\lambda}, \frac{b}{\lambda}, \frac{c}{\lambda} \right) \\ &= -\left(\nabla_X \frac{a}{\lambda}, \nabla_X \frac{b}{\lambda}, \nabla_X \frac{c}{\lambda} \right) = \mathbf{0}. \end{aligned}$$

Therefore $L(X) = 0 \forall X \in T_p M$, i.e. $L \equiv 0$.

Ex. Let $M = S_r^2$ be the sphere of radius r , let n be the *outward* pointing unit normal. Determine the Weingarten map at each point of M .



Fix $p \in S_r^2$, and let $X \in T_p S_r^2$. Let $\sigma : (-\varepsilon, \varepsilon) \rightarrow S_r^2$ be a curve in S_r^2 such that $\sigma(0) = p$, $\frac{d\sigma}{dt}(0) = X$

Then,

$$L(X) = -\nabla_X n = -\frac{d}{dt} n \circ \sigma(t)|_{t=0}.$$

But note, $n \circ \sigma(t) = n(\sigma(t)) = \frac{\sigma(t)}{|\sigma(t)|} = \frac{\sigma(t)}{r}$. Hence,

$$\begin{aligned} L(X) &= -\frac{d}{dt} \frac{\sigma(t)}{r} \Big|_{t=0} = -\frac{1}{r} \frac{d\sigma}{dt} \Big|_{t=0} \\ L(X) &= -\frac{1}{r} X. \end{aligned}$$

for all $X \in T_p M$. Hence, $L = -\frac{1}{r} id$, where $id : T_p M \rightarrow T_p M$ is the identity map, $id(X) = X$.

Remark. If we had taken the inward pointing normal then $L = \frac{1}{r} id$.

Def. For each $p \in M$, the *second fundamental form* is the bilinear form $\mathcal{L} = T_p M \times T_p M \rightarrow \mathbb{R}$ defined by,

$$\begin{aligned} \mathcal{L}(X, Y) &= \langle L(X), Y \rangle \\ &= -\langle \nabla_X n, Y \rangle. \end{aligned}$$

\mathcal{L} is indeed *bilinear*, e.g.,

$$\begin{aligned} \mathcal{L}(aX + bY, Z) &= \langle L(aX + bY), Z \rangle \\ &= \langle aL(X) + bL(Y), Z \rangle \\ &= a\langle L(X), Z \rangle + b\langle L(Y), Z \rangle \\ &= a\mathcal{L}(X, Z) + b\mathcal{L}(Y, Z). \end{aligned}$$

Ex. $M = \text{plane}$, $\mathcal{L} \equiv 0$:

$$\mathcal{L}(X, Y) = \langle L(X), Y \rangle = \langle 0, Y \rangle = 0.$$

Ex. The sphere S_r^2 of radius r , $\mathcal{L} : T_p S_r^2 \times T_p S_r^2 \rightarrow \mathbb{R}$,

$$\begin{aligned} \mathcal{L}(X, Y) &= \langle L(X), Y \rangle \\ &= \left\langle -\frac{1}{r} X, Y \right\rangle \\ &= -\frac{1}{r} \langle X, Y \rangle \end{aligned}$$

Hence, $\mathcal{L} = -\frac{1}{r}\langle \cdot, \cdot \rangle$. Multiple of the first fundamental form!

Coordinate expressions

Let $\mathbf{x} : U \rightarrow M$ be a patch containing $p \in M$. Then $\{\mathbf{x}_1, \mathbf{x}_2\}$ is a basis for $T_p M$. We express $L : T_p M \rightarrow T_p M$ and $\mathcal{L} : T_p M \times T_p M \rightarrow \mathbb{R}$ with respect to this basis. Since $L(\mathbf{x}_j) \in T_p M$, we have,

$$\begin{aligned} L(\mathbf{x}_j) &= L^1_j \mathbf{x}_1 + L^2_j \mathbf{x}_2, \quad j = 1, 2 \\ &= \sum_{i=1}^2 L^i_j \mathbf{x}_i. \end{aligned}$$

The numbers L^i_j , $1 \leq i, j \leq 2$, are called the components of L with respect to the coordinate basis $\{\mathbf{x}_1, \mathbf{x}_2\}$. The 2×2 matrix $[L^i_j]$ is the matrix representing the linear map L with respect to the basis $\{\mathbf{x}_1, \mathbf{x}_2\}$.

Exercise 5.3 Let $X \in T_p M$ and let $Y = L(X)$. In terms of components, $X = \sum_j X^j \mathbf{x}_j$ and $Y = \sum_i Y^i \mathbf{x}_i$. Show that

$$Y^i = \sum_j L^i_j X^j, \quad i = 1, 2,$$

which in turn implies the matrix equation,

$$\begin{bmatrix} Y^1 \\ Y^2 \end{bmatrix} = [L^i_j] \begin{bmatrix} X^1 \\ X^2 \end{bmatrix}.$$

This is the Weingarten map expressed as a matrix equation.

Introduce the unit normal field along $W = \mathbf{x}(U)$ with respect to the patch $\mathbf{x} : U \rightarrow M$,

$$\begin{aligned} \hat{n} &= \frac{\mathbf{x}_1 \times \mathbf{x}_2}{|\mathbf{x}_1 \times \mathbf{x}_2|}, \quad \hat{n} = \hat{n}(u^1, u^2), \\ n &= \hat{n} \circ \mathbf{x}^{-1} : W \rightarrow \mathbb{R}. \end{aligned}$$

Then by Exercise 5.2,

$$L(\mathbf{x}_j) = -\nabla_{\mathbf{x}_j} n = -\frac{\partial \hat{n}}{\partial u^j}.$$

Setting $n_j = \frac{\partial \hat{n}}{\partial u^j}$ we have

$$n_j = -L(\mathbf{x}_j)$$

$$n_j = -\sum_i L^i_j \mathbf{x}_i, \quad j = 1, 2 \quad (\text{The Weingarten equations.})$$

These equations can be used to compute the components of the Weingarten map. However, in practice it turns out to be more useful to have a formula for computing the components of the second fundamental form.

Components of \mathcal{L} :

The components of \mathcal{L} with respect to $\{\mathbf{x}_1, \mathbf{x}_2\}$ are defined as,

$$L_{ij} = \mathcal{L}(\mathbf{x}_i, \mathbf{x}_j), \quad 1 \leq i, j \leq 2.$$

By bilinearity, the components completely determine \mathcal{L} ,

$$\begin{aligned} \mathcal{L}(X, Y) &= \mathcal{L}\left(\sum_i X^i \mathbf{x}_i, \sum_j Y^j \mathbf{x}_j\right) \\ &= \sum_{i,j} X^i Y^j \mathcal{L}(\mathbf{x}_i, \mathbf{x}_j) \\ &= \sum_{i,j} L_{ij} X^i Y^j. \end{aligned}$$

The following proposition provides a useful formula for computing the L_{ij} 's.

Proposition. The components L_{ij} of \mathcal{L} are given by,

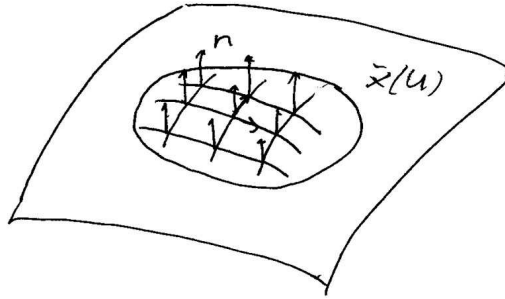
$$L_{ij} = \langle \hat{n}, \mathbf{x}_{ij} \rangle,$$

where $\mathbf{x}_{ij} = \frac{\partial^2 \mathbf{x}}{\partial u^j \partial u^i}$.

Remark. Henceforth we no longer distinguish between n and \hat{n} , i.e., lets agree to drop the “ \wedge ”, then,

$$L_{ij} = \langle n, \mathbf{x}_{ij} \rangle,$$

Proof:



Along $\mathbf{x}(U)$ we have, $\langle n, \frac{\partial \mathbf{x}}{\partial u^j} \rangle = 0$, and hence,

$$\begin{aligned}\frac{\partial}{\partial u^i} \langle n, \frac{\partial \mathbf{x}}{\partial u^j} \rangle &= 0 \\ \langle \frac{\partial n}{\partial u^i}, \frac{\partial \mathbf{x}}{\partial u^j} \rangle + \langle n, \frac{\partial^2 \mathbf{x}}{\partial u^i \partial u^j} \rangle &= 0 \\ \langle \frac{\partial n}{\partial u^i}, \frac{\partial \mathbf{x}}{\partial u^j} \rangle &= -\langle n, \frac{\partial^2 \mathbf{x}}{\partial u^i \partial u^j} \rangle = -\langle n, \frac{\partial^2 \mathbf{x}}{\partial u^j \partial u^i} \rangle,\end{aligned}$$

or, using shorthand notation,

$$\langle n_i, \mathbf{x}_j \rangle = -\langle n, \mathbf{x}_{ij} \rangle.$$

But,

$$\begin{aligned}L_{ij} &= \mathcal{L}(\mathbf{x}_i, \mathbf{x}_j) = \langle L(\mathbf{x}_i), \mathbf{x}_j \rangle \\ &= -\langle n_i, \mathbf{x}_j \rangle,\end{aligned}$$

and hence $L_{ij} = \langle n, \mathbf{x}_{ij} \rangle$. □

Observe,

$$\begin{aligned}L_{ij} &= \langle n, \mathbf{x}_{ij} \rangle \\ &= \langle n, \mathbf{x}_{ji} \rangle \quad (\text{mixed partials equal!}) \\ L_{ij} &= L_{ji}, \quad 1 \leq i, j \leq 2.\end{aligned}$$

In other words, $\mathcal{L}(\mathbf{x}_i, \mathbf{x}_j) = \mathcal{L}(\mathbf{x}_j, \mathbf{x}_i)$.

Proposition. The second fundamental form $\mathcal{L} : T_p M \times T_p M \rightarrow \mathbb{R}$ is symmetric, i.e.

$$\mathcal{L}(X, Y) = \mathcal{L}(Y, X) \quad \forall \quad X, Y \in T_p M.$$

Exercise 5.4 Prove this proposition by showing \mathcal{L} is symmetric iff $L_{ij} = L_{ji}$ for all $1 \leq i, j \leq 2$.

Relationship between L^i_j and L_{ij}

$$\begin{aligned}L_{ij} &= \mathcal{L}(\mathbf{x}_i, \mathbf{x}_j) = \mathcal{L}(\mathbf{x}_j, \mathbf{x}_i) \\ &= \langle L(\mathbf{x}_j), \mathbf{x}_i \rangle = \langle \sum_k L^k_j \mathbf{x}_k, \mathbf{x}_i \rangle \\ &= \sum_k L^k_j \langle \mathbf{x}_k, \mathbf{x}_i \rangle \\ L_{ij} &= \sum_k g_{ik} L^k_j, \quad 1 \leq i, j \leq 2\end{aligned}$$

Classical tensor jargon: L_{ij} obtained from L^k_j by “lowering the index k with the metric”. The equation above implies the matrix equation

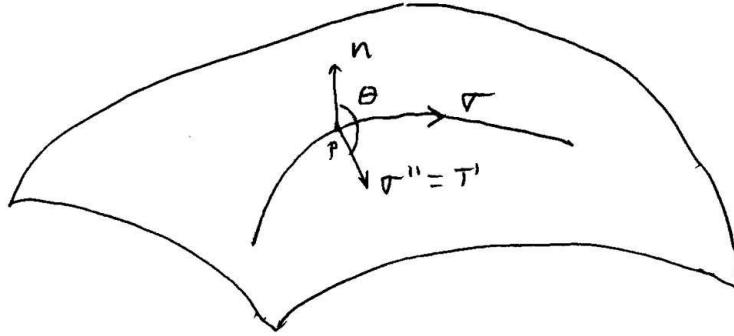
$$[L_{ij}] = [g_{ij}][L^i_j].$$

Geometric Interpretation of the 2nd Fundamental Form

Normal Curvature. Let $s \rightarrow \sigma(s)$ be a unit speed curve lying in a surface M . Let p be a point on σ , and let n be a smooth unit normal v.f. defined in a nbd W of p . The *normal curvature* of σ at p , denoted κ_n , is defined to be the component of the curvature vector $\sigma'' = T'$ along n , i.e.,

$$\begin{aligned} \kappa_n &= \text{normal component of the curvature vector} \\ &= \langle \sigma'', n \rangle \\ &= \langle T', n \rangle \\ &= |T'| |n| \cos \theta \\ &= \kappa \cos \theta, \end{aligned}$$

where θ is the angle between the curvature vector T' and the surface normal n . If $\kappa \neq 0$ then, recall, we can introduce the principal normal N to σ , by the equation, $T' = \kappa N$; in this case θ is the angle between N and n .



Remark: κ_n gives a measure of how much σ is bending in the direction perpendicular to the surface; it neglects the amount of bending tangent to the surface.

Proposition. Let M be a surface, $p \in M$. Let $X \in T_p M$, $|X| = 1$ (i.e. X is a unit tangent vector). Let $s \rightarrow \sigma(s)$ be any unit speed curve in M such that $\sigma(0) = p$ and $\sigma'(0) = X$. Then

$$\begin{aligned} \mathcal{L}(X, X) &= \text{normal curvature of } \sigma \text{ at } p \\ &= \langle \sigma'', n \rangle. \end{aligned}$$

Proof. Along σ ,

$$\begin{aligned}\langle \sigma'(s), n \circ \sigma(s) \rangle &= 0, & \text{for all } s \\ \frac{d}{ds} \langle \sigma', n \circ \sigma \rangle &= 0 \\ \langle \sigma'', n \circ \sigma \rangle + \langle \sigma', \frac{d}{ds} n \circ \sigma \rangle &= 0.\end{aligned}$$

At $s = 0$,

$$\begin{aligned}\langle \sigma'', n \rangle + \langle X, \nabla_X n \rangle &= 0 \\ \langle \sigma'', n \rangle &= -\langle X, \nabla_X n \rangle \\ \kappa_n &= \langle X, L(X) \rangle \\ \kappa_n &= \langle L(X), X \rangle \\ &= \mathcal{L}(X, X).\end{aligned}$$

Remark: the sign convention used in the definition of the Weingarten map ensures that $\mathcal{L}(X, X) = +\kappa_n$ (rather than $-\kappa_n$).

Corollary. *All unit speed curves lying in a surface M which pass through $p \in M$ and have the same unit tangent vector X at p , have the same normal curvature at p . That is, the normal curvature depends only on the tangent direction X .*

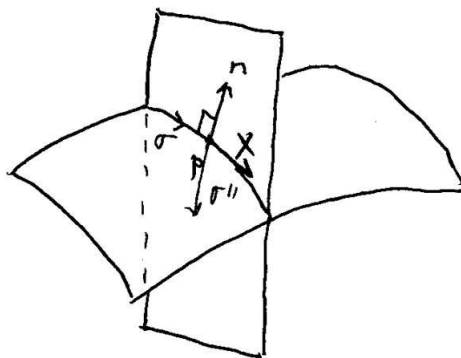
Thus it makes sense to say:

$\mathcal{L}(X, X)$ is the normal curvature in the direction X .

Given a unit tangent vector $X \in T_p M$, there is a distinguished curve in M , called the *normal section* at p in the direction X . Let,

Π = plane through p spanned by n and X .

Π cuts M in a curve σ . Parameterize σ wrt arc length, $s \rightarrow \sigma(s)$, such that $\sigma(0) = p$ and $\frac{d\sigma}{ds}(0) = X$:



By definition, σ is the *normal section* at p in the direction X . By the previous proposition,

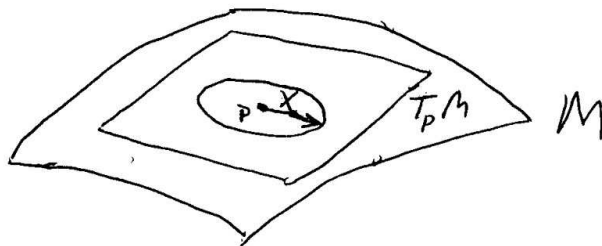
$$\begin{aligned}\mathcal{L}(X, X) &= \text{normal curvature of the normal section } \sigma \\ &= \langle \sigma'', n \rangle = \langle T', n \rangle \\ &= \kappa \cos \theta,\end{aligned}$$

where θ is the angle between n and T' . Since σ lies in Π , T' is tangent to Π , and since T' is also perpendicular to X , it follows that T' is a multiple of n . Hence, $\theta = 0$ or π , which implies that $\mathcal{L}(X, X) = \pm\kappa$.

Thus we conclude that,

$$\mathcal{L}(X, X) = \text{signed curvature of the normal section at } p \text{ in the direction } X.$$

Principal Curvatures.



The set of unit tangent vectors at p , $X \in T_p M$, $|X| = 1$, forms a circle in the tangent plane to M at p . Consider the function from this circle into the reals,

$$\begin{aligned}X &\rightarrow \text{normal curvature in direction } X \\ X &\rightarrow \mathcal{L}(X, X).\end{aligned}$$

The *principal curvatures* of M at p , $\kappa_1 = \kappa_1(p)$ and $\kappa_2 = \kappa_2(p)$, are defined as follows,

$$\begin{aligned}\kappa_1 &= \text{the maximum normal curvature at } p \\ &= \max_{|X|=1} \mathcal{L}(X, X)\end{aligned}$$

$$\begin{aligned}\kappa_2 &= \text{the minimum normal curvature at } p \\ &= \min_{|X|=1} \mathcal{L}(X, X)\end{aligned}$$

This is the *geometric* characterization of principal curvatures. There is also an important *algebraic* characterization.

Some Linear Algebra

Let V be a vector space over the reals, and let $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$ be an inner product on V ; hence V is an inner product space. Let $L : V \rightarrow V$ be a linear transformation. Our main application will be to the case: $V = T_p M$, $\langle \cdot, \cdot \rangle =$ induced metric, and $L =$ Weingarten map.

L is said to be *self adjoint* provided

$$\langle L(v), w \rangle = \langle v, L(w) \rangle \quad \forall v, w \in V.$$

Remark. Let $V = \mathbb{R}^n$, with the usual dot product, and let $L : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a linear map. Let $[L^i_j]$ = matrix representing L with respect to the *standard* basis, $e_1 : (1, 0, \dots, 0)$, etc. Then L is self-adjoint if and only if $[L^i_j]$ is *symmetric* $[L^i_j] = [L^j_i]$.

Proposition. The Weingarten map $L : T_p M \rightarrow T_p M$ is self adjoint, i.e.

$$\langle L(X), Y \rangle = \langle X, L(Y) \rangle \quad \forall X, Y \in T_p M,$$

where $\langle \cdot, \cdot \rangle =$ 1st fundamental form.

Proof. We have,

$$\begin{aligned} \langle L(X), Y \rangle &= \mathcal{L}(X, Y) = \mathcal{L}(Y, X) \\ &= \langle L(Y), X \rangle = \langle X, L(Y) \rangle. \end{aligned}$$

Self adjoint linear transformations have very nice properties, as we now discuss. For this discussion, we restrict attention to 2-dimensional vector spaces, $\dim V = 2$.

A vector $v \in V$, $v \neq 0$, is called an *eigenvector* of L if there is a real number λ such that,

$$L(v) = \lambda v.$$

λ is called an *eigenvalue* of L . The eigenvalues of L can be determined by solving

$$\det(A - \lambda I) = 0 \tag{*}$$

where A is a matrix representing L and $I =$ identity matrix. The equation (*) is a quadratic equation in λ , and hence has at most 2 real roots; it may have no real roots.

Theorem (Fundamental Theorem of Self Adjoint Operators) *Let V be a 2-dimensional inner product space. Let $L : V \rightarrow V$ be a self-adjoint linear map. Then V admits an orthonormal basis consisting of eigenvectors of L . That is, there exists an orthonormal basis $\{e_1, e_2\}$ of V and real numbers λ_1, λ_2 , $\lambda_1 \geq \lambda_2$ such that*

$$L(e_1) = \lambda_1 e_1, \quad L(e_2) = \lambda_2 e_2,$$

i.e., e_1 and e_2 are eigenvectors of L and λ_1, λ_2 are the corresponding eigenvalues. Moreover the eigenvalues are given by

$$\begin{aligned}\lambda_1 &= \max_{|v|=1} \langle L(v), v \rangle \\ \lambda_2 &= \min_{|v|=1} \langle L(v), v \rangle.\end{aligned}$$

Proof. See handout from Do Carmo.

Remark on orthogonality of eigenvectors. Let e_1, e_2 be eigenvectors with eigenvalues λ_1, λ_2 . If $\lambda_1 \neq \lambda_2$, then e_1 and e_2 are necessarily orthogonal, as seen by the following,

$$\lambda_1 \langle e_1, e_2 \rangle = \langle L(e_1), e_2 \rangle = \langle e_1, L(e_2) \rangle = \lambda_2 \langle e_1, e_2 \rangle,$$

$\Rightarrow (\lambda_1 - \lambda_2) \langle e_1, e_2 \rangle = 0 \Rightarrow \langle e_1, e_2 \rangle = 0$. On the other hand, if $\lambda_1 = \lambda_2 = \lambda$ then $L(v) = \lambda v$ for all v . Hence *any* o.n. basis is a basis of eigenvectors.

We now apply these facts to the Weingarten map,

$$\begin{aligned}L &: T_p M \rightarrow T_p M, \\ \mathcal{L} &: T_p M \times T_p M \rightarrow \mathbb{R}, \quad \mathcal{L}(X, Y) = \langle L(X), Y \rangle.\end{aligned}$$

Since L is self adjoint, and, by definition,

$$\begin{aligned}\kappa_1 &= \max_{|X|=1} \mathcal{L}(X, X) = \max_{|X|=1} \langle L(X), X \rangle \\ \kappa_2 &= \min_{|X|=1} \mathcal{L}(X, X) = \min_{|X|=1} \langle L(X), X \rangle,\end{aligned}$$

we obtain the following.

Theorem. *The principal curvatures κ_1, κ_2 of M at p are the eigenvalues of the Weingarten map $L : T_p M \rightarrow T_p M$. There exists an orthonormal basis $\{e_1, e_2\}$ of $T_p M$ such that*

$$L(e_1) = \kappa_1 e_1, \quad L(e_2) = \kappa_2 e_2,$$

i.e., e_1, e_2 are eigenvectors of L associated with the eigenvalues κ_1, κ_2 , respectively. The eigenvectors e_1 and e_2 are called **principal directions**.

Observe that,

$$\begin{aligned}\kappa_1 &= \kappa_1 \langle e_1, e_1 \rangle = \langle L(e_1), e_1 \rangle = \mathcal{L}(e_1, e_1) \\ \kappa_2 &= \kappa_2 \langle e_2, e_2 \rangle = \langle L(e_2), e_2 \rangle = \mathcal{L}(e_2, e_2),\end{aligned}$$

i.e., the principal curvature κ_1 is the normal curvature in the principal direction e_1 , and similarly for κ_2 .

Now, let A be the matrix associated to the Weingarten map L with respect to the orthonormal basis $\{e_1, e_2\}$; thus,

$$\begin{aligned} L(e_1) &= \kappa_1 e_1 + 0e_2 \\ L(e_2) &= 0e_1 + \kappa_2 e_2 \end{aligned}$$

which implies,

$$A = \begin{bmatrix} \kappa_1 & 0 \\ 0 & \kappa_2 \end{bmatrix} .$$

Then,

$$\begin{aligned} \det L &= \det A = \kappa_1 \kappa_2 \\ \operatorname{tr} L &= \operatorname{tr} A = \kappa_1 + \kappa_2 . \end{aligned}$$

Definition. The *Gaussian curvature* of M at p , $K = K(p)$, and the *mean curvature* of M at p , $H = H(p)$ are defined as follows,

$$\begin{aligned} K &= \det L = \kappa_1 \kappa_2 \\ H &= \operatorname{tr} L = \kappa_1 + \kappa_2 . \end{aligned}$$

Remarks. The Gaussian curvature is the more important of the two curvatures; it is what is meant by the *curvature* of a surface. A famous discovery by Gauss is that it is intrinsic – in fact can be computed in terms of the g_{ij} 's (This is not obvious!). The *mean curvature* (which has to do with minimal surface theory) is *not* intrinsic. This can be easily seen as follows. Changing the normal $n \rightarrow -n$ changes the sign of the Weingarten map,

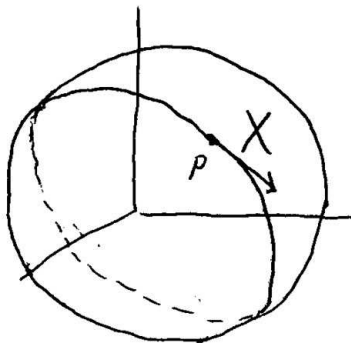
$$L_{-n} = -L_n .$$

This in turn changes the sign of the principal curvatures, hence $H = \kappa_1 + \kappa_2$ changes sign, but $K = \kappa_1 \kappa_2$ does not change sign.

Some Examples

Ex. For $S_r^2 =$ sphere of radius r , compute κ_1, κ_2, K, H (Use outward normal).

Geometrically: $p \in S_r^2, X \in T_p M, |X| = 1,$



$$\begin{aligned}\mathcal{L}(X, X) &= \pm \text{curvature of normal section in direction } X \\ &= -\text{curvature of great circle} \\ &= -\frac{1}{r}.\end{aligned}$$

Therefore

$$\kappa_1 = \max_{|X|=1} \mathcal{L}(X, X) = -\frac{1}{r},$$

$$\kappa_2 = \min_{|X|=1} \mathcal{L}(X, X) = -\frac{1}{r},$$

$$K = \kappa_1 \kappa_2 = \frac{1}{r^2} > 0, \quad H = \kappa_1 + \kappa_2 = -\frac{2}{r}.$$

Algebraically: Find eigenvalues of Weingarten map: $L : T_p M \rightarrow T_p M$. We showed previously,

$$\begin{aligned}L &= -\frac{1}{r} \text{id}, \quad \text{i.e.,} \\ L(X) &= -\frac{1}{r} X \quad \text{for all } X \in T_p M.\end{aligned}$$

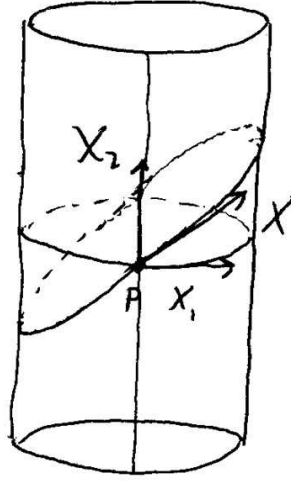
Thus, with respect to *any* orthonormal basis $\{e_1, e_2\}$ of $T_p M$,

$$L(e_i) = -\frac{1}{r} e_i \quad i = 1, 2.$$

Therefore, $\kappa_1 = \kappa_2 = -\frac{1}{r}$, $K = \frac{1}{r^2}$, $H = -\frac{2}{r}$.

Ex. Let M be the cylinder of radius a : $x^2 + y^2 = a^2$. Compute κ_1, κ_2, K, H . (Use the inward pointing normal)

Geometrically:



$$\begin{aligned}\mathcal{L}(X_1, X_1) &= \pm \text{curvature of normal section in direction } X_1 \\ &= + \text{curvature of circle of radius } a \\ &= \frac{1}{a},\end{aligned}$$

$$\begin{aligned}\mathcal{L}(X_2, X_2) &= \pm \text{curvature of normal section in direction } X_2 \\ &= \text{curvature of line} \\ &= 0.\end{aligned}$$

In general, for $X \neq X_1, X_2$,

$$\mathcal{L}(X, X) = \text{curvature of ellipse through } p.$$

The curvature is between 0 and $\frac{1}{a}$, and thus,

$$0 \leq \mathcal{L}(X, X) \leq \frac{1}{a}.$$

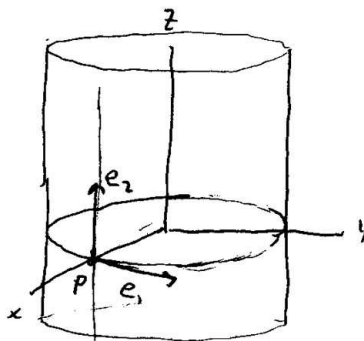
We conclude that,

$$\kappa_1 = \max_{|X|=1} \mathcal{L}(X, X) = \mathcal{L}(X_1, X_1) = \frac{1}{a},$$

$$\kappa_2 = \min_{|X|=1} \mathcal{L}(X, X) = \mathcal{L}(X_2, X_2) = 0.$$

Thus, $K = 0$ (cylinder is flat!) and $H = \frac{1}{a}$.

Algebraically: Determine the eigenvalues of the Weingarten map. By a rotation and translation we may take p to be the point $p = (a, 0, 0)$. Let $e_1, e_2 \in T_p M$ be the tangent vectors $e_1 = (0, 1, 0)$ and $e_2 = (0, 0, 1)$.



To compute $L(e_1)$, consider the circle,

$$\sigma(s) = (a \cos(\frac{s}{a}), a \sin(\frac{s}{a}), 0)$$

Note that $\sigma(0) = p$ and $\sigma'(0) = e_1$. Thus,

$$\begin{aligned} L(e_1) &= -\nabla_{e_1} n \\ &= -\frac{d}{ds} n(\sigma(s))|_{s=0} \end{aligned}$$

. But,

$$\begin{aligned} n(\sigma(s)) &= n(\sigma(s)) = \frac{\sigma(s)}{|\sigma(s)|} = \frac{\sigma(s)}{a} \\ &= -\left(\cos\left(\frac{s}{a}\right), \sin\left(\frac{s}{a}\right), 0\right) \end{aligned}$$

Therefore,

$$\begin{aligned} L(e_1) &= \frac{d}{ds} \left(\cos \left(\frac{s}{a} \right), \sin \left(\frac{s}{a} \right), 0 \right) \Big|_{s=0} \\ &= \frac{1}{a} \left(-\sin \left(\frac{s}{a} \right), \cos \left(\frac{s}{a} \right), 0 \right) \Big|_{s=0} \\ &= \frac{1}{a} (0, 1, 0) \end{aligned}$$

$$L(e_1) = \frac{1}{a} e_1$$

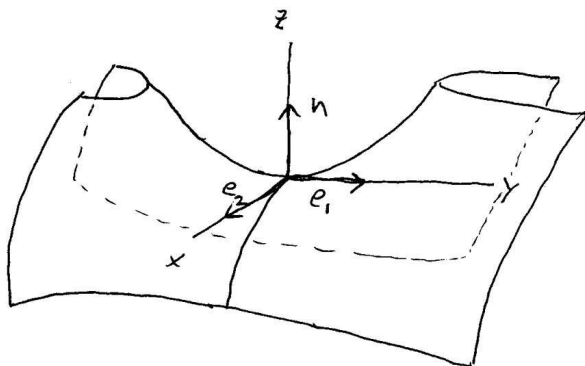
Thus, e_1 is an eigenvector with eigenvalue $\frac{1}{a}$. Similarly (exercise!),

$$L(e_2) = \mathbf{0} = 0 \cdot e_2$$

i.e., e_2 is an eigenvector with eigenvalue 0. (Note; e_2 is tangent to a vertical line in the surface, along which n is constant.)

We conclude that, $\kappa_1 = \frac{1}{a}$, $\kappa_2 = 0$, $K = 0$, $H = \frac{1}{a}$.

Ex. Consider the saddle surface, $M: z = y^2 - x^2$, Compute κ_1 , κ_2 , K , H at $p = (0, 0, 0)$.



$$\begin{aligned} \mathcal{L}(e_1, e_1) &= \pm \text{curvature of normal section in direction of } e_1 \\ &= + \text{curvature of } z = y^2 \end{aligned}$$

The curvature is given by,

$$\kappa = \frac{\left| \frac{d^2 z}{dy^2} \right|}{\left[1 + \left(\frac{dz}{dy} \right)^2 \right]^{3/2}} = 2$$

and so, $\mathcal{L}(e_1, e_1) = 2$. Similarly, $\mathcal{L}(e_2, e_2) = -2$. Observe,

$$\mathcal{L}(e_2, e_2) \leq \mathcal{L}(X, X) \leq \mathcal{L}(e_1, e_1)$$

Therefore, $\kappa_1 = 2$, $\kappa_2 = -2$, $K = -4$, and $H = 0$ at $(0, 0, 0)$.

Exercise 5.5. For the saddle surface M above, consider the Weingarten map $L : T_p M \rightarrow T_p M$ at $p = (0, 0, 0)$. Compute $L(e_1)$ and $L(e_2)$ directly from the definition of the Weingarten map to show,

$$L(e_1) = 2e_1 \text{ and } L(e_2) = -2e_2.$$

Hence, -2 and 2 are the eigenvalues of L , which means $\kappa_1 = 2$ and $\kappa_2 = -2$.

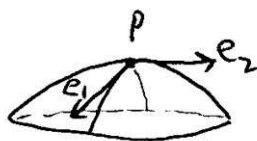
Remark. We have computed the quantities κ_1 , κ_2 , K , and H of the saddle surface only at a single point. To compute these quantities at all points, we will need to develop better computational tools.

Significance of the sign of Gaussian Curvature

We have,

$$K = \det L = \kappa_1 \kappa_2.$$

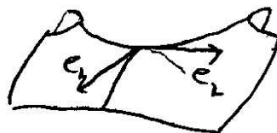
1. $K > 0 \iff \kappa_1$ and κ_2 have the same sign \iff the normal sections in the principal directions e_1, e_2 both bend in the same direction,



$K > 0$ at p .

Ex. $z = ax^2 + by^2$, a, b have the same sign (elliptic paraboloid). At $p = (0, 0, 0)$, $K = 4ab > 0$.

2. $K < 0 \iff \kappa_1$ and κ_2 have opposite signs \iff normal sections in principle directions e_1 and e_2 bend in opposite directions,



$K < 0$ at p .

Ex. $z = ax^2 + by^2$, a, b have opposite sign (hyperbolic paraboloid). At $p = (0, 0, 0)$, $K = 4ab < 0$.

Thus, roughly speaking,

$K > 0$ at $p \Rightarrow$ surface is “bowl-shaped” near p

$K < 0$ at $p \Rightarrow$ surface is “saddle-shaped” near p

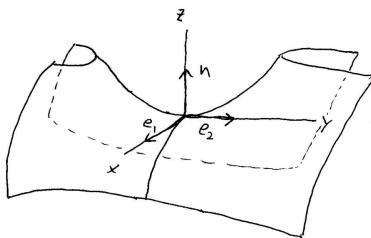
This rough observation can be made more precise, as we now show. Let M be a surface, $p \in M$. Let e_1, e_2 be principal directions at p . Choose e_1, e_2 so that $\{e_1, e_2, n\}$ is a positively oriented orthonormal basis.

By a translation and rotation of the surface, we can assume, (see the figure),

(1) $p = (0, 0, 0)$

(2) $e_1 = (1, 0, 0), \quad e_2 = (0, 1, 0), \quad n = (0, 0, 1)$ at p

(3) Near $p = (0, 0, 0)$, the surface can be described by an equation of form, $z = f(x, y)$, where $f : U \subset \mathbb{R}^2 \rightarrow \mathbb{R}$ is smooth and $f(0, 0) = 0$.



Claim:

$$z = \frac{1}{2}\kappa_1 x^2 + \frac{1}{2}\kappa_2 y^2 + \text{higher order terms}$$

Proof. Consider the Taylor series about $(0, 0)$ for functions of two variables,

$$z = f(0, 0) + f_x(0, 0)x + f_y(0, 0)y + \frac{1}{2}f_{xx}(0, 0)x^2 + f_{xy}(0, 0)xy + \frac{1}{2}f_{yy}(0, 0)y^2 + \text{higher order terms}.$$

We must compute 1st and 2nd order partial derivatives of f at $(0, 0)$. Introduce the Monge patch,

$$\begin{aligned} x &= u \\ \mathbf{x}: \quad y &= v \\ z &= f(u, v) \end{aligned}$$

i.e. $\mathbf{x}(u, v) = (u, v, f(u, v))$.

We have,

$$\begin{aligned}
\mathbf{x}_1 &= \mathbf{x}_u = (1, 0, f_u), \\
\mathbf{x}_2 &= \mathbf{x}_v = (0, 1, f_v), \\
n &= \frac{\mathbf{x}_1 \times \mathbf{x}_2}{|\mathbf{x}_1 \times \mathbf{x}_2|} = \frac{\mathbf{x}_u \times \mathbf{x}_v}{|\mathbf{x}_u \times \mathbf{x}_v|} \\
&= \frac{(-f_u, -f_v, 1)}{\sqrt{1 + f_u^2 + f_v^2}}
\end{aligned}$$

At $(u, v) = (0, 0)$: $n = (0, 0, 1) \Rightarrow f_u = f_v = 0, \Rightarrow \mathbf{x}_1 = (1, 0, 0) = e_1$ and $\mathbf{x}_2 = (0, 1, 0) = e_2$.

Recall, the components of the 2nd fundamental form $L_{ij} = \mathcal{L}(\mathbf{x}_i, \mathbf{x}_j)$ may be computed from the formula,

$$L_{ij} = \langle n, \mathbf{x}_{ij} \rangle, \quad \mathbf{x}_{ij} = \frac{\partial^2 \mathbf{x}}{\partial u^j \partial u^i}.$$

In particular, $L_{11} = \langle n, \mathbf{x}_{11} \rangle$, where $\mathbf{x}_{11} = \mathbf{x}_{uu} = (0, 0, f_{uu})$.

At $(u, v) = (0, 0)$: $L_{11} = \langle n, \mathbf{x}_{11} \rangle = (0, 0, 1) \cdot (0, 0, f_{uu}(0, 0)) = f_{uu}(0, 0)$.
Therefore, $f_{uu}(0, 0) = L_{11} = \mathcal{L}(\mathbf{x}_1, \mathbf{x}_1) = \mathcal{L}(e_1, e_1) = \kappa_1$. Similarly,

$$f_{vv}(0, 0) = \mathcal{L}(e_2, e_2) = \kappa_2$$

$$f_{uv}(0, 0) = \mathcal{L}(e_1, e_2) = \langle L(e_1), e_2 \rangle = \lambda_1(e_1, e_2) = 0.$$

Thus, setting $x = u, y = v$, we have shown,

$$\begin{aligned}
f_x(0, 0) &= f_y(0, 0) = 0 \\
f_{xx}(0, 0) &= \kappa_1, \quad f_{yy}(0, 0) = \kappa_2, \quad f_{xy}(0, 0) = 0,
\end{aligned}$$

which, substituting in the Taylor expansion, implies,

$$z = \frac{1}{2}\kappa_1 x^2 + \frac{1}{2}\kappa_2 y^2 + \text{higher order terms}.$$

Computational Formula for Gaussian Curvature.

We have,

$$\begin{aligned}
K &= \text{Gaussian curvature} \\
&= \det L = \det[L^i_j].
\end{aligned}$$

From the equation at the top of p. 14,

$$\begin{aligned}[L_{ij}] &= [g_{ij}][L^i_j], \\ \det[L_{ij}] &= \det[g_{ij}] \det[L^i_j] \\ &= \det[g_{ij}] \cdot K\end{aligned}$$

Hence,

$$K = \frac{\det[L_{ij}]}{\det[g_{ij}]}, \quad g_{ij} = \langle \mathbf{x}_i, \mathbf{x}_j \rangle, \quad L_{ij} = \langle n, \mathbf{x}_{ij} \rangle.$$

Further,

$$\begin{aligned}\det[L_{ij}] &= \det \begin{bmatrix} L_{11} & L_{12} \\ L_{21} & L_{22} \end{bmatrix} \\ &= L_{11}L_{22} - L_{12}^2,\end{aligned}$$

since $L_{12} = L_{21}$, and similarly,

$$\det[g_{ij}] = g_{11}g_{22} - g_{12}^2.$$

Thus,

$$K = \frac{L_{11}L_{22} - L_{12}^2}{g_{11}g_{22} - g_{12}^2}.$$

Ex. Compute the Gaussian curvature of the saddle surface $z = y^2 - x^2$.

Introduce the Monge patch, $\mathbf{x}(u, v) = (u, v, v^2 - u^2)$.

Compute metric components g_{ij} :

$$\begin{aligned}\mathbf{x}_u &= (1, 0, -2u), \quad \mathbf{x}_v = (0, 1, 2v), \\ g_{uu} &= \langle \mathbf{x}_u, \mathbf{x}_u \rangle = (1, 0, -2u) \cdot (1, 0, -2u) \\ &= 1 + 4u^2.\end{aligned}$$

Similarly,

$$\begin{aligned}g_{vv} &= \langle \mathbf{x}_v, \mathbf{x}_v \rangle = 1 + 4v^2, \\ g_{uv} &= \langle \mathbf{x}_u, \mathbf{x}_v \rangle = -4uv.\end{aligned}$$

Thus,

$$\begin{aligned}\det[g_{ij}] &= g_{uu}g_{vv} - g_{uv}^2 \\ &= (1 + 4u^2)(1 + 4v^2) - 16u^2v^2 \\ &= 1 + 4u^2 + 4v^2.\end{aligned}$$

Compute the second fundamental form components L_{ij} :

We use, $L_{ij} = \langle n, \mathbf{x}_{ij} \rangle$. We have,

$$n = \frac{\mathbf{x}_u \times \mathbf{x}_v}{|\mathbf{x}_u \times \mathbf{x}_v|} = \frac{(2u, -2v, 1)}{\sqrt{1 + 4u^2 + 4v^2}},$$

and,

$$\mathbf{x}_{uu} = (0, 0, -2), \quad \mathbf{x}_{vv} = (0, 0, 2), \quad \mathbf{x}_{uv} = (0, 0, 0).$$

Then,

$$\begin{aligned} L_{uu} &= \langle n, \mathbf{x}_{uu} \rangle = \frac{-2}{\sqrt{1 + 4u^2 + 4v^2}} \\ L_{vv} &= \langle n, \mathbf{x}_{vv} \rangle = \frac{2}{\sqrt{1 + 4u^2 + 4v^2}} \\ L_{uv} &= \langle n, \mathbf{x}_{uv} \rangle = 0. \end{aligned}$$

Thus,

$$\det[L_{ij}] = L_{uu}L_{vv} - L_{uv}^2 = \frac{-4}{1 + 4u^2 + 4v^2},$$

and therefore,

$$\begin{aligned} K(u, v) &= \frac{\det[L_{ij}]}{\det[g_{ij}]} = \frac{-4}{1 + 4u^2 + 4v^2} \cdot \frac{1}{1 + 4u^2 + 4v^2} \\ K(u, v) &= \frac{-4}{(1 + 4u^2 + 4v^2)^2}. \end{aligned}$$

Hence the saddle surface $z = y^2 - x^2$ has Gaussian curvature function,

$$K(x, y) = \frac{-4}{(1 + 4x^2 + 4y^2)^2}.$$

Observe that $K < 0$ and, $K = \frac{-4}{(1 + 4r^2)^2} \sim \frac{1}{r^4}$, where $r = \sqrt{x^2 + y^2}$ is the distance

from the z -axis. As $r \rightarrow \infty$, $K \rightarrow 0$ rapidly.

Exercise 5.6. Consider the surface M which is the graph of $z = f(x, y)$. Show that the Gaussian curvature $K = K(x, y)$ is given by,

$$K(x, y) = \frac{f_{xx}f_{yy} - f_{xy}^2}{(1 + f_x^2 + f_y^2)^2}$$

where $f_x = \frac{\partial f}{\partial x}$, $f_{xx} = \frac{\partial^2 f}{\partial x^2}$, etc.

Exercise 5.7. Let M be the torus of large radius R and small radius r described in Exercise 3.3. Using the parameterization,

$$\mathbf{x}(t, \theta) = ((R + r \cos t) \cos \theta, (R + r \cos t) \sin \theta, r \sin t)$$

show that the Gaussian curvature $K = K(t, \theta)$ is given by,

$$K = \frac{\cos t}{r(R + r \cos t)}.$$

Where on the torus is the Gaussian curvature negative? Where is it positive?

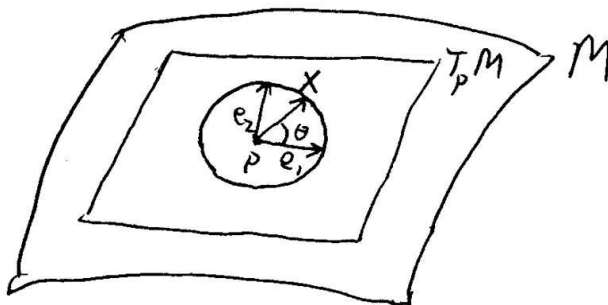
Exercise 5.8. Derive the following expression for the mean curvature H ,

$$H = \frac{g_{11}L_{22} - 2g_{12}L_{12} + g_{22}L_{11}}{g_{11}g_{22} - g_{12}^2}$$

The principal curvatures κ_1 and κ_2 at a point $p \in M$ are the normal curvatures in the principal directions e_1 and e_2 . The normal curvature in any direction X is determined by κ_1 and κ_2 as follows.

If $X \in T_p M$, $|X| = 1$ then X can be expressed as (see the figure),

$$X = \cos \theta e_1 + \sin \theta e_2.$$



Proposition (Euler's formula). The normal curvature in the direction X is given by,

$$\mathcal{L}(X, X) = \kappa_1 \cos^2 \theta + \kappa_2 \sin^2 \theta,$$

where κ_1, κ_2 are the principal curvatures, and θ is the angle between X and the principal direction e_1 .

Proof. Use the shorthand, $c = \cos \theta$, $s = \sin \theta$. Then $X = ce_1 + se_2$, and

$$\begin{aligned} L(X) &= L(ce_1 + se_2) \\ &= cL(e_1) + sL(e_2) \\ &= c\kappa_1 e_1 + s\kappa_2 e_2. \end{aligned}$$

Therefore,

$$\begin{aligned}\mathcal{L}(X, X) &= \langle L(X), X \rangle \\ &= \langle c\kappa_1 e_1 + s\kappa_2 e_2, ce_1 + se_2 \rangle \\ &= c^2\kappa_1 + s^2\kappa_2.\end{aligned}$$

Exercise 5.9. Assuming $\kappa_1 > \kappa_2$, determine where (i.e., for which values of θ) the function,

$$\kappa(\theta) = \kappa_1 \cos^2 \theta + \kappa_2 \sin^2 \theta, \quad 0 \leq \theta \leq 2\pi$$

achieves its maximum and minimum. The answer shows that the principal directions e_1, e_2 are unique, up to sign, in this case.

Gauss Theorema Egregium

The Weingarten map,

$$L(X) = -\nabla_X n$$

is an *extrinsically* defined object - it involves the normal to the surface. There is no reason to suspect that the determinant of L , the Gaussian curvature, is *intrinsic*, i.e. can be computed from measurements taken in the surface. But Gauss carried out some courageous computations and made the extraordinary discovery that, in fact, the Gaussian curvature K is intrinsic - i.e., can be computed from the g_{ij} 's. This is the most important result in the subject - albeit not the prettiest! If this result were not true then the subject of differential geometry, as we know it, would not exist.

We now embark on the same path - courageously carrying out the same computation.

Some notation. Introduce the “inverse” metric components, g^{ij} , $1 \leq i, j \leq 2$, by

$$[g^{ij}] = [g_{ij}]^{-1},$$

i.e. g^{ij} is the i - j th entry of the inverse of the matrix $[g_{ij}]$. Using the formula,

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

we can express g^{ij} explicitly in terms of the g_{ij} , e.g.

$$g^{11} = \frac{g_{22}}{g_{11}g_{22} - g_{12}^2}, \quad \text{etc.}$$

Note, in an orthogonal coordinate system, i.e., a proper patch in which $g_{12} = \langle \mathbf{x}_1, \mathbf{x}_2 \rangle = 0$, we have simply,

$$g^{11} = \frac{1}{g_{11}}, \quad g^{22} = \frac{1}{g_{22}}, \quad g^{12} = g^{21} = 0.$$

By definition of inverse, we have

$$[g_{ij}][g^{ij}] = I$$

where $I = \text{identity matrix} = [\delta_i^j]$, where δ_i^j is the Kronecker delta (cf., Chapter 1),

$$\delta_i^j = \begin{cases} 0 & , i \neq j \\ 1, & i = j, \end{cases}$$

and so,

$$[g_{ij}][g^{ij}] = [\delta_i^j].$$

The product formula for matrices then implies,

$$\sum_k g_{ik} g^{kj} = \delta_i^j$$

or, by the Einstein summation convention,

$$g_{ik} g^{kj} = \delta_i^j.$$

Now, let M be a surface and $\mathbf{x} : U \subset \mathbb{R}^2 \rightarrow M \subset \mathbb{R}^3$ be any proper patch in M . Then,

$$\mathbf{x} = \mathbf{x}(u^1, u^2) = (x(u^1, u^2), y(u^1, u^2), z(u^1, u^2)),$$

$$\mathbf{x}_i = \frac{\partial \mathbf{x}}{\partial u^i} = \left(\frac{\partial x}{\partial u^i}, \frac{\partial y}{\partial u^i}, \frac{\partial z}{\partial u^i} \right),$$

$$\mathbf{x}_{ij} = \frac{\partial^2 \mathbf{x}}{\partial u^j \partial u^i} = \left(\frac{\partial^2 x}{\partial u^j \partial u^i}, \frac{\partial^2 y}{\partial u^j \partial u^i}, \frac{\partial^2 z}{\partial u^j \partial u^i} \right).$$

We seek useful expressions for these second derivatives. At any point $p \in \mathbf{x}(U)$, $\{\mathbf{x}_1, \mathbf{x}_2, n\}$ form a basis for $T_p \mathbb{R}^3$. Since at p , $\mathbf{x}_{ij} \in T_p \mathbb{R}^3$, we can write,

$$\begin{aligned} \mathbf{x}_{ij} &= \Gamma_{ij}^1 \mathbf{x}_1 + \Gamma_{ij}^2 \mathbf{x}_2 + \lambda_{ij} n, \\ \mathbf{x}_{ij} &= \sum_{\ell=1}^2 \Gamma_{ij}^\ell \mathbf{x}_\ell + \lambda_{ij} n. \end{aligned}$$

or, making use of the Einstein summation convention,

$$\mathbf{x}_{ij} = \Gamma_{ij}^\ell \mathbf{x}_\ell + \lambda_{ij} n. \quad (*)$$

We obtain expressions for $\lambda_{ij}, \Gamma_{ij}^\ell$. Dotting (*) with n gives,

$$\begin{aligned}\langle \mathbf{x}_{ij}, n \rangle &= \Gamma_{ij}^\ell \langle \mathbf{x}_\ell, n \rangle + \lambda_{ij} \langle n, n \rangle \\ \Rightarrow \lambda_{ij} &= \langle \mathbf{x}_{ij}, n \rangle = \langle n, \mathbf{x}_{ij} \rangle \\ \lambda_{ij} &= L_{ij} .\end{aligned}$$

Dotting (*) with \mathbf{x}_k gives,

$$\begin{aligned}\langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle &= \Gamma_{ij}^\ell \langle \mathbf{x}_\ell, \mathbf{x}_k \rangle + \lambda_{ij} \langle n, \mathbf{x}_k \rangle \\ \langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle &= \Gamma_{ij}^\ell g_{\ell k} .\end{aligned}$$

Solving for Γ_{ij}^ℓ ,

$$\begin{aligned}\langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle g^{km} &= \Gamma_{ij}^\ell g_{\ell k} g^{km} \\ &= \Gamma_{ij}^\ell \delta_\ell^m \\ \langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle g^{km} &= \Gamma_{ij}^m\end{aligned}$$

Thus,

$$\Gamma_{ij}^\ell = g^{k\ell} \langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle$$

Claim. The quantity $\langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle$ is given by,

$$\begin{aligned}\langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle &= \frac{1}{2} \left(\frac{\partial g_{ik}}{\partial u^j} + \frac{\partial g_{jk}}{\partial u^i} - \frac{\partial g_{ij}}{\partial u^k} \right) \\ &= \frac{1}{2} (g_{ik,j} + g_{jk,i} - g_{ij,k}) .\end{aligned}$$

Proof of Claim. We use Gauss' trick of permuting indices.

$$\begin{aligned}g_{ij,k} &= \frac{\partial}{\partial u^k} g_{ij} = \frac{\partial}{\partial u^k} \langle \mathbf{x}_i, \mathbf{x}_j \rangle \\ &= \left\langle \frac{\partial \mathbf{x}_i}{\partial u^k}, \mathbf{x}_j \right\rangle + \left\langle \mathbf{x}_i, \frac{\partial \mathbf{x}_j}{\partial u^k} \right\rangle\end{aligned}$$

$$(1) \quad g_{ij,k} = \langle \mathbf{x}_{ik}, \mathbf{x}_j \rangle + \langle \mathbf{x}_i, \mathbf{x}_{jk} \rangle$$

$$(j \leftrightarrow k) \quad (2) \quad g_{ik,j} = \langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle + \langle \mathbf{x}_i, \mathbf{x}_{kj} \rangle$$

$$(i \leftrightarrow j) \quad (3) \quad g_{jk,i} = \langle \mathbf{x}_{ji}, \mathbf{x}_k \rangle + \langle \mathbf{x}_j, \mathbf{x}_{ki} \rangle$$

Then (2) + (3) – (1) gives:

$$g_{ik,j} + g_{jk,i} - g_{ij,k} = 2\langle \mathbf{x}_{ij}, \mathbf{x}_k \rangle$$

Thus,

$$\Gamma_{ij}^\ell = \frac{1}{2} g^{k\ell} (g_{ik,j} + g_{jk,i} - g_{ij,k}).$$

Remark. These are known as the *Christoffel* symbols.

Summarizing, we have,

$$\mathbf{x}_{ij} = \Gamma_{ij}^\ell \mathbf{x}_\ell + L_{ij} n, \quad \text{Gauss Formula}$$

where L_{ij} are the components of the 2nd fundamental form and Γ_{ij}^ℓ are the Christoffel symbols as given above. Let us also recall the *Weingarten* equations (p. 11),

$$n_j = -L^i_j \mathbf{x}_i$$

Remark. The vector fields $\mathbf{x}_1, \mathbf{x}_2, n$, play a role in surface theory roughly analogous to the Frenet frame for curves. The two formulas above for the partial derivatives of $\mathbf{x}_1, \mathbf{x}_2, n$ then play a role roughly analogous to the Frenet formulas.

Now, Gauss takes things one step further and computes the 3rd derivatives,
 $\mathbf{x}_{ijk} = \frac{\partial}{\partial u^k} \mathbf{x}_{ij}$:

$$\begin{aligned} \mathbf{x}_{ijk} &= \frac{\partial}{\partial u^k} (\Gamma_{ij}^\ell \mathbf{x}_\ell + L_{ij} n) = \frac{\partial}{\partial u^k} \Gamma_{ij}^\ell \mathbf{x}_\ell + \frac{\partial}{\partial u^k} L_{ij} n \\ &= \Gamma_{ij,k}^\ell \mathbf{x}_\ell + \Gamma_{ij}^\ell \mathbf{x}_{\ell k} + L_{ij,k} n + L_{ij} n_k \\ &= \Gamma_{ij,k}^\ell \mathbf{x}_\ell + \Gamma_{ij}^\ell (\Gamma_{\ell k}^m \mathbf{x}_m + L_{\ell k} n) + L_{ij,k} n + L_{ij} (-L^{\ell}_k \mathbf{x}_\ell) \\ &= \Gamma_{ij,k}^\ell \mathbf{x}_\ell + \underbrace{\Gamma_{ij}^\ell \Gamma_{\ell k}^m \mathbf{x}_m}_{\Gamma_{ij}^m \Gamma_{mk}^\ell \mathbf{x}_\ell} + \Gamma_{ij}^\ell L_{\ell k} n + L_{ij,k} n - L_{ij} L^{\ell}_k \mathbf{x}_\ell \end{aligned}$$

Thus,

$$\mathbf{x}_{ijk} = (\Gamma_{ij,k}^\ell + \Gamma_{ij}^m \Gamma_{mk}^\ell - L_{ij} L^{\ell}_k) \mathbf{x}_\ell + (L_{ij,k} + \Gamma_{ij}^\ell L_{\ell k}) n,$$

and interchanging indices ($j \leftrightarrow k$),

$$\mathbf{x}_{ikj} = (\Gamma_{ik,j}^\ell + \Gamma_{ik}^m \Gamma_{mj}^\ell - L_{ik} L^{\ell}_j) \mathbf{x}_\ell + (L_{ik,j} + \Gamma_{ik}^\ell L_{\ell j}) n.$$

Now, $\mathbf{x}_{ikj} = \mathbf{x}_{ijk}$ implies

$$\Gamma_{ik,j}^\ell + \Gamma_{ik}^m \Gamma_{mj}^\ell - L_{ik} L^{\ell}_j = \Gamma_{ij,k}^\ell + \Gamma_{ij}^m \Gamma_{mk}^\ell - L_{ij} L^{\ell}_k =$$

or,

$$\underbrace{\Gamma_{ik,j}^\ell - \Gamma_{ij,k}^\ell + \Gamma_{ik}^m \Gamma_{mj}^\ell - \Gamma_{ij}^m \Gamma_{mk}^\ell}_{R_{ijk}^\ell} = L_{ik} L_j^\ell - L_{ij} L_k^\ell.$$

These are the components of the famous *Riemann curvature tensor*. **Observe:** R_{ijk}^ℓ are *intrinsic*, i.e. can be computed from the g_{ij} 's (involve 1st and 2nd derivatives of the g_{ij} 's).

We arrive at,

$$R_{ijk}^\ell = L_{ik} L_j^\ell - L_{ij} L_k^\ell \quad \text{The Gauss Equations.}$$

Gauss' Theorem Egregium. *The Gaussian curvature of a surface is intrinsic, i.e. can be computed in terms of the g_{ij} 's.*

Proof. This follows from the Gauss equations. Multiply both sides by $g_{m\ell}$,

$$g_{m\ell} R_{ijk}^\ell = L_{ik} g_{m\ell} L_j^\ell - L_{ij} g_{m\ell} L_k^\ell.$$

But recall (see p. 13),

$$L_{mj} = g_{m\ell} L_j^\ell.$$

Hence,

$$g_{m\ell} R_{ijk}^\ell = L_{ik} L_{mj} - L_{ij} L_{mk}.$$

Setting $i = k = 1$, $m = j = 2$ we obtain,

$$\begin{aligned} g_{2\ell} R_{121}^\ell &= L_{11} L_{22} - L_{12} L_{21} \\ &= \det[L_{ij}]. \end{aligned}$$

Thus,

$$\begin{aligned} K &= \frac{\det[L_{ij}]}{\det[g_{ij}]} \\ K &= \frac{g_{2\ell} R_{121}^\ell}{g}, \quad g = \det[g_{ij}] \end{aligned}$$

□

Comment. Gauss' Theorema Egregium can be interpreted in a slightly different way in terms of *isometries*. We discuss this point here very *briefly* and very *informally*.

Let M and N be two surfaces. A one-to-one, onto map $f : M \rightarrow N$ that preserves lengths and angles is called an isometry. (This may be understood at the level of tangent vectors: f takes curves to curves, and hence velocity vectors to velocity vectors. f is an isometry $\Leftrightarrow f$ preserves angle between velocity vectors and preserve length of velocity vectors.)

Ex. The process of wrapping a piece of paper into a cylinder is an isometry.

Theorem Gaussian curvature is a *bending invariant*, i.e. is invariant under isometries, by which we mean: if $f : M \rightarrow N$ is an isometry then

$$K_N(f(p)) = K_M(p),$$

i.e., the Gaussian curvature is the same at corresponding points.

Proof f preserves lengths and angles. Hence, in appropriate coordinate systems, the metric components for M and N are the same. By the formula for K above, the Gaussian curvature will be the same at corresponding points.

Application 1. The cylinder has Gaussian curvature $K = 0$ (because a plane has zero Gaussian curvature).

Application 2. No piece of a plane can be bent into a piece of a sphere without distorting lengths (because $K_{\text{plane}} = 0$, $K_{\text{sphere}} = \frac{1}{r^2}$, $r = \text{radius}$).

Theorem (Riemann). Let M be a surface with vanishing Gaussian curvature, $K = 0$. Then each $p \in M$ has a neighborhood which is isometric to an open set in the Euclidean plane.

Exercise 5.10. Although The Gaussian curvature K is a “bending invariant”, show that the principal curvatures κ_1, κ_2 are not. I.e., show that the principle curvatures are not in general invariant under an isometry. (Hint: Consider the bending of a rectangle into a cylinder).