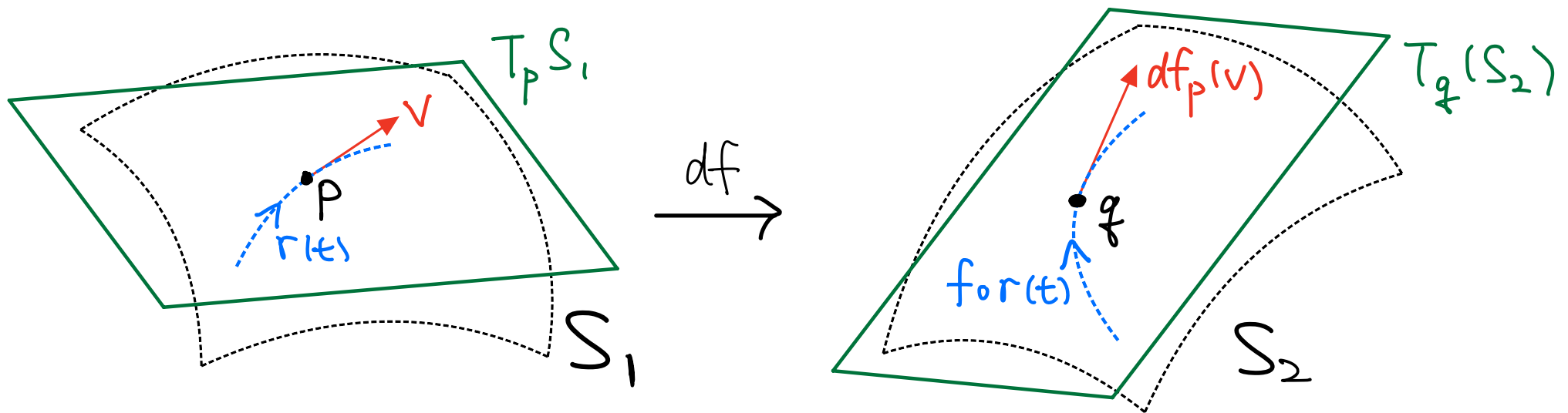


Definition (Differential) Suppose $f: S_1 \rightarrow S_2$. Let $p \in S_1$, $q = f(p) \in S_2$

Define $df_p: T_p S_1 \rightarrow T_q S_2$ as follow:

For $v \in T_p S_1$, let $\vec{r}(t)$ be a curve on S_1 with $\vec{r}(0) = p$, $\vec{r}'(0) = v$

Define $df_p(v) = (f \circ \vec{r})'(0) \in T_q S_2$



Theorem $df_p: T_p S_1 \rightarrow T_q S_2$ is linear: For any $\alpha, \beta \in \mathbb{R}$

$$df_p(\alpha X_u + \beta X_v) = \alpha df_p(X_u) + \beta df_p(X_v)$$

Definition 3.4.5 (Differential of Gauss map). Let S be a regular surface in $\mathbb{R}^3$ with regular parametrization $\mathbf{x}(u, v)$. For each $p \in S$, define $d\mathbf{n}_p : T_p S \rightarrow T_p S$ called the **differential of Gauss map** by

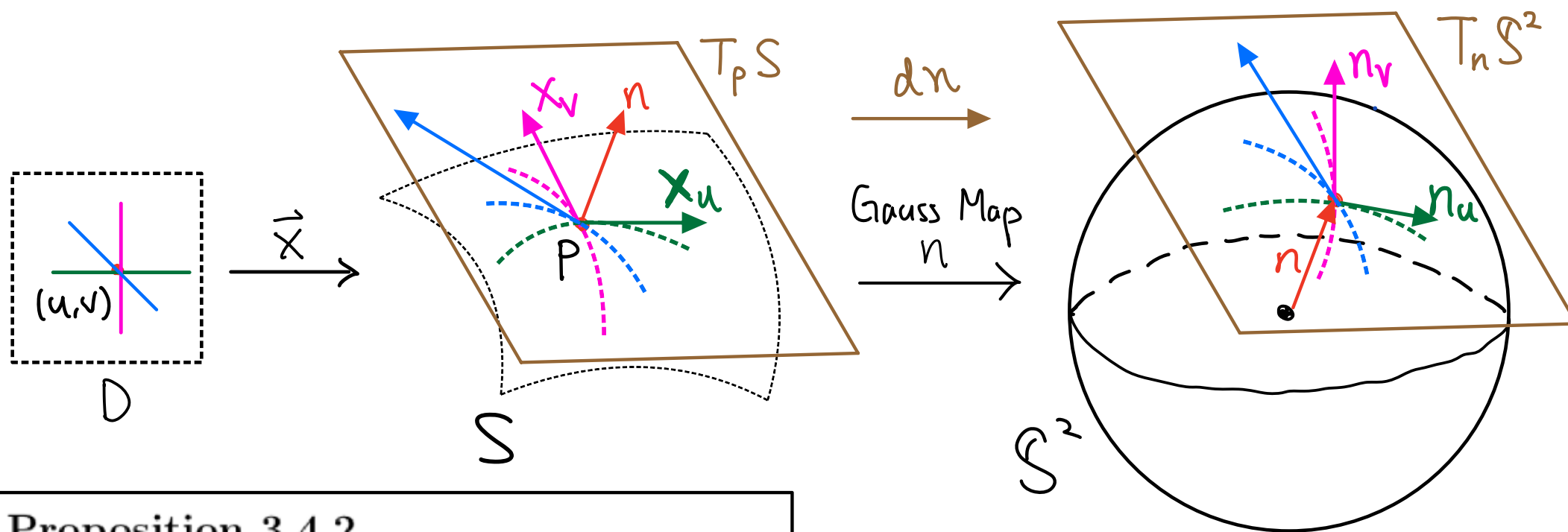
$$d\mathbf{n}_p(\alpha \mathbf{x}_u + \beta \mathbf{x}_v) = \alpha \mathbf{n}_u + \beta \mathbf{n}_v$$

for any real numbers $\alpha, \beta \in \mathbb{R}$.

Rmk

$$d\mathbf{n}_p(\mathbf{x}_u) = \mathbf{n}_u$$

$$d\mathbf{n}_p(\mathbf{x}_v) = \mathbf{n}_v$$



Proposition 3.4.2.

- $T_n S^2 = T_p S$.
- $\mathbf{n}_u, \mathbf{n}_v \in T_p S \Rightarrow \begin{cases} \mathbf{n}_u = a\mathbf{x}_u + b\mathbf{x}_v \\ \mathbf{n}_v = c\mathbf{x}_u + d\mathbf{x}_v \end{cases}$

Next :

Express a, b, c, d
in terms of I, II

Express a, b, c, d in terms of I, II

$$\begin{cases} n_u = aX_u + bX_v \dots (1) \\ n_v = cX_u + dX_v \dots (2) \end{cases}$$

$$\begin{aligned} (1) \Rightarrow \langle X_u, n_u \rangle &= \langle X_u, aX_u + bX_v \rangle \\ &= a \langle X_u, X_u \rangle + b \langle X_u, X_v \rangle \\ &= aE + bF \end{aligned}$$

$$\begin{aligned} \langle X_v, n_u \rangle &= \langle X_v, aX_u + bX_v \rangle \\ &= aF + bG \end{aligned}$$

Similarly

$$\begin{aligned} (2) \Rightarrow \langle X_u, n_v \rangle &= cE + dF \\ \langle X_v, n_v \rangle &= cF + dG \end{aligned}$$

$$\begin{aligned} -II &= \begin{bmatrix} \langle X_u, n_u \rangle & \langle X_v, n_u \rangle \\ \langle X_u, n_v \rangle & \langle X_v, n_v \rangle \end{bmatrix} \\ &= \begin{bmatrix} aE + bF & aF + bG \\ cE + dF & cF + dG \end{bmatrix} \\ &= \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} E & F \\ F & G \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} I \end{aligned}$$

$$\begin{aligned} \begin{bmatrix} a & b \\ c & d \end{bmatrix} &= -(II)(I^{-1}) \\ &= - \begin{pmatrix} e & f \\ f & g \end{pmatrix} \begin{pmatrix} E & F \\ F & G \end{pmatrix}^{-1} \\ &= - \begin{pmatrix} e & f \\ f & g \end{pmatrix} \frac{1}{EG - F^2} \begin{pmatrix} G & -F \\ -F & E \end{pmatrix} \\ &= - \frac{1}{EG - F^2} \begin{pmatrix} eG - fF & fE - eF \\ fG - gF & gE - fF \end{pmatrix} \end{aligned}$$

Rmk $\det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = (-1)^2 \frac{\det II}{\det I} = K$

Proposition 3.4.9. *The matrix representation of $d\mathbf{n}_p$ with respect to basis $\mathbf{x}_u, \mathbf{x}_v$ is*

$$-(II)(I^{-1}) = -\frac{1}{EG - F^2} \begin{pmatrix} eG - fF & fE - eF \\ fG - gF & gE - fF \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

It means

$$\begin{aligned} d\mathbf{n}_p(\mathbf{x}_u) &= a\mathbf{x}_u + b\mathbf{x}_v \\ d\mathbf{n}_p(\mathbf{x}_v) &= c\mathbf{x}_u + d\mathbf{x}_v \end{aligned} \quad \text{where} \quad \begin{pmatrix} a & b \\ c & d \end{pmatrix} = -(II)I^{-1}$$

Theorem 3.4.7 (Self-adjointness of differential of Gauss map). *The differential of Gauss map $d\mathbf{n}_p : T_pS \rightarrow T_pS$ is self-adjoint. In other words, for any $\mathbf{u}, \mathbf{v} \in T_pS$, we have*

$$\langle d\mathbf{n}_p(\mathbf{u}), \mathbf{v} \rangle = \langle \mathbf{u}, d\mathbf{n}_p(\mathbf{v}) \rangle.$$

PF Let $u = \alpha x_u + \beta x_v$ $v = \gamma x_u + \delta x_v$

$$\begin{aligned}\langle dn_p(u), v \rangle &= \langle \alpha dn_p(x_u) + \beta dn_p(x_v), \gamma x_u + \delta x_v \rangle \\ &= \alpha \gamma \langle dn_p(x_u), x_u \rangle + \alpha \delta \langle dn_p(x_u), x_v \rangle \\ &\quad + \beta \gamma \langle dn_p(x_v), x_u \rangle + \beta \delta \langle dn_p(x_v), x_v \rangle\end{aligned}$$

$$\begin{aligned}\langle u, dn_p(v) \rangle &= \alpha \gamma \langle x_u, dn_p(x_u) \rangle + \alpha \delta \langle x_u, dn_p(x_v) \rangle \\ &\quad + \beta \gamma \langle x_v, dn_p(x_u) \rangle + \beta \delta \langle x_v, dn_p(x_v) \rangle\end{aligned}$$

Only need to check if $\langle dn_p(x_u), x_v \rangle = \langle x_u, dn_p(x_v) \rangle$:

$$\begin{aligned}\langle dn_p(x_u), x_v \rangle &= \langle n_u, x_v \rangle \\ &= \langle n, x_{vu} \rangle \\ &= \langle n, x_{uv} \rangle \\ &= \langle n_v, x_u \rangle \\ &= \langle x_u, n_v \rangle \\ &= \langle x_u, dn_p(x_v) \rangle\end{aligned} \quad \begin{aligned}&\leftarrow \langle n, x_v \rangle \equiv 0 \\ &\Rightarrow \frac{\partial}{\partial u} \langle n, x_v \rangle \equiv 0 \\ &\Rightarrow \langle n_u, x_v \rangle + \langle n, x_{vu} \rangle \equiv 0\end{aligned}$$

1.5 Eigenvalues, eigenvectors and diagonalization

Definition 1.5.1 (Eigenvalues and eigenvectors). Let A be an $n \times n$ matrix. If λ is a complex number³ and ξ is a non-zero⁴ complex vector such that

$$A\xi = \lambda\xi,$$

then we say that λ is an **eigenvalue** of A and ξ is an **eigenvector** of A associated with λ .

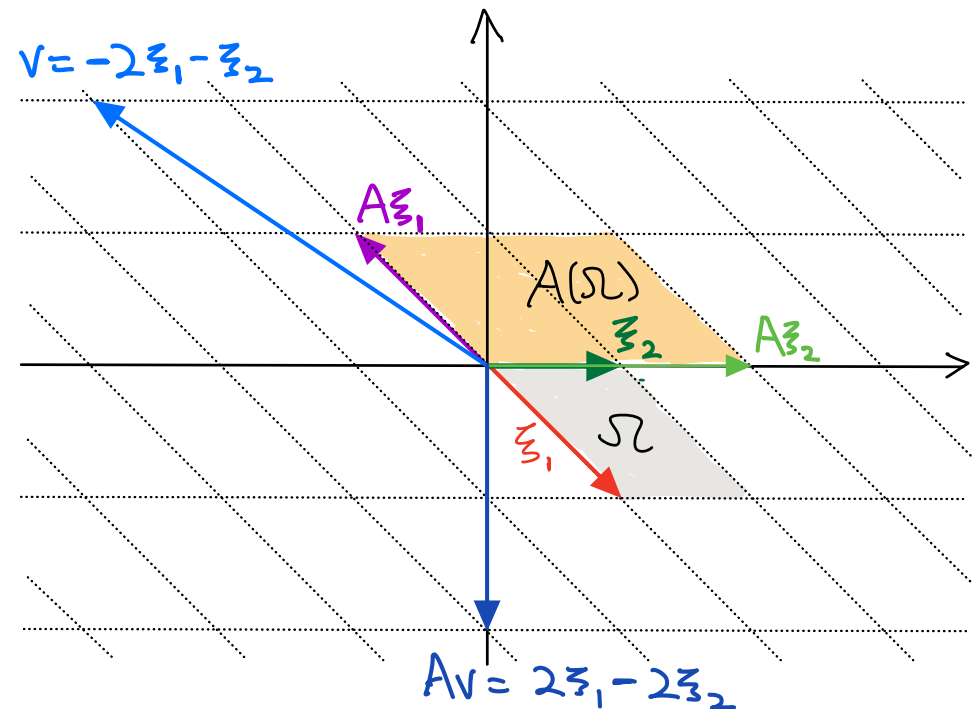
eg $A = \begin{bmatrix} 2 & 3 \\ 0 & -1 \end{bmatrix}$

$$\begin{bmatrix} 2 & 3 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix} = (-1) \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 3 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\xi_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \lambda_1 = -1$$

$$\xi_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \lambda_2 = 2$$



Eigenvectors help to understand Av

Proposition 1.5.3. Let A be an $n \times n$ matrix.

1. A complex number λ is an eigenvalue of A if and only if λ is a root to the characteristic equation $\det(xI - A) = 0$.
2. Let λ be an eigenvalue of A . Then ξ is an eigenvector of A associated with λ if and only if $\xi \neq \mathbf{0}$ and $(\lambda I - A)\xi = \mathbf{0}$.

eg $A = \begin{bmatrix} 1 & 2 \\ -1 & 4 \end{bmatrix}$

$$\det(xI - A) = \begin{vmatrix} x-1 & -2 \\ 1 & x-4 \end{vmatrix}$$

$$= (x-1)(x-4) - (-2)(1)$$

$$= x^2 - 5x + 6$$

$$= (x-2)(x-3)$$

$$\det(xI - A) = 0$$

$$\Rightarrow x=2 \text{ or } 3$$

Eigenvalues: 2, 3

$$\text{For } \lambda_1 = 2, \quad 2I - A = \begin{bmatrix} 2-1 & -2 \\ 1 & 2-4 \end{bmatrix} = \begin{bmatrix} 1 & -2 \\ 1 & -2 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -2 \\ 1 & -2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad x - 2y = 0 \quad \text{Take } \xi_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\text{For } \lambda_2 = 3, \quad 3I - A = \begin{bmatrix} 3-1 & -2 \\ 1 & 3-4 \end{bmatrix} = \begin{bmatrix} 2 & -2 \\ 1 & -1 \end{bmatrix}$$

$$\begin{bmatrix} 2 & -2 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad x - y = 0 \quad \text{Take } \xi_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ is an eigenvector associated with eigenvalue 2

$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is an eigenvector associated with eigenvalue 3

We can define eigenvalue, eigenvector for $d\mathbf{n}_p: T_p S \rightarrow T_{n(p)}(\mathbb{S}^2) = T_p S$ similarly

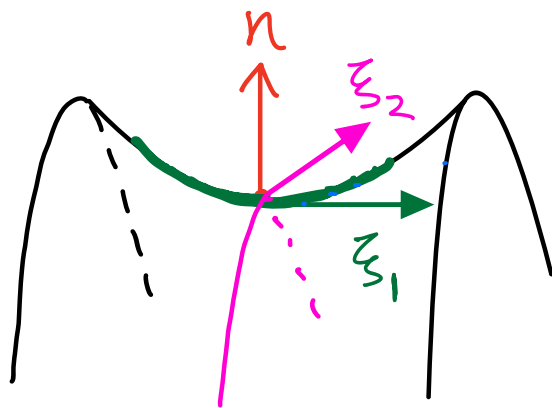
Theorem 3.4.8. Let S be a regular surface in $\mathbb{R}^3$ and $p \in S$. Then there exists principal directions $\xi_1, \xi_2 \in T_p S$ which constitute an orthonormal basis for $T_p S$.

$$\begin{cases} d\mathbf{n}_p(\xi_1) = -\kappa_1 \xi_1 \\ d\mathbf{n}_p(\xi_2) = -\kappa_2 \xi_2 \end{cases}$$

Then we say that κ_1, κ_2 are the **principal curvatures** of S at p , and ξ_1, ξ_2 are the corresponding **principal directions**.

Rmk $d\mathbf{n}_p$ is self-adjoint $\Rightarrow$ Orthonormal basis exists (Linear Algebra: Section 1.4, 1.5)

eg



$$\begin{aligned} d\mathbf{n}_p(\xi_1) &= \overbrace{-\kappa_1}^{<0} \xi_1 & \kappa_1 > 0 \\ d\mathbf{n}_p(\xi_2) &= \underbrace{-\kappa_2}_{>0} \xi_2 & \kappa_2 < 0 \end{aligned}$$

$\kappa > 0$ if the curve bends towards $\vec{n}$

Matrix representation of dn_p

Note $\{\xi_1, \xi_2\}, \{x_u, x_v\}$ are both basis of $T_p S$

$$dn_p(\xi_1) = -k_1 \xi_1 + 0 \xi_2$$

$$dn_p(x_u) = a x_u + b x_v$$

$$dn_p(\xi_2) = 0 \xi_1 + (-k_2) \xi_2$$

$$dn_p(x_v) = c x_u + d x_v$$

Matrix representation of dn_p with respect to these basis

$$\begin{bmatrix} -k_1 & 0 \\ 0 & -k_2 \end{bmatrix}$$

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} = (-II)(I)^{-1} = -\frac{1}{EG-F^2} \begin{pmatrix} eG-fF & fE-eF \\ fG-gF & gE-fF \end{pmatrix}$$

Fact: The two matrices have same determinant and trace

$$k_1 k_2 = \begin{vmatrix} -k_1 & 0 \\ 0 & -k_2 \end{vmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = (-1)^2 \frac{\det II}{\det I} = K$$

$$-(k_1 + k_2) = a + d = -\frac{gE - 2fF + eG}{EG - F^2}$$

Theorem 3.4.10. *Let S be a regular surface and K be the Gaussian curvature of S . Then for any $p \in S$,*

$$K(p) = \det(d\mathbf{n}_p) = \kappa_1 \kappa_2$$

Definition 3.4.11 (Mean curvature). *Let S be a regular surface and $d\mathbf{n}_p$ be the differential of Gauss map at $p \in S$. The **mean curvature** of S at p is*

$$H = -\frac{1}{2}\text{tr}(d\mathbf{n}_p) = \frac{1}{2}(\kappa_1 + \kappa_2) = \frac{1}{2}\text{tr}((II)(I^{-1})) = \frac{1}{2} \left(\frac{gE - 2fF + eG}{EG - F^2} \right).$$

Definition 3.4.12 (Minimal surface). *Let S be a regular surface in $\mathbb{R}^3$ and H be the mean curvature of S . We say that S is a **minimal surface** if $H = 0$ at every point of S .*

Example 3.4.14. Show that the catenoid parametrized by

$$\mathbf{x}(\theta, v) = (\cosh v \cos \theta, \cosh v \sin \theta, v), \quad 0 < \theta < 2\pi, v \in \mathbb{R},$$

is a minimal surface.

$$\mathbf{x}_\theta = (-\cosh v \sin \theta, \cosh v \cos \theta, 0)$$

$$\mathbf{x}_v = (\sinh v \cos \theta, \sinh v \sin \theta, 1)$$

$$\mathbf{x}_\theta \times \mathbf{x}_v = (\cosh v \cos \theta, \cosh v \sin \theta, -\cosh v \sinh v)$$

$$\|\mathbf{x}_\theta \times \mathbf{x}_v\|^2 = \cosh^2 v + \cosh^2 v \sinh^2 v = \cosh^2 v (1 + \sinh^2 v) = \cosh^4 v$$

$$\mathbf{n} = (\operatorname{sech} v \cos \theta, \operatorname{sech} v \sin \theta, \tanh v)$$

$$\mathbf{x}_{\theta\theta} = (-\cosh v \cos \theta, -\cosh v \sin \theta, 0)$$

$$\mathbf{x}_{\theta v} = (-\sinh v \sin \theta, \sinh v \cos \theta, 0)$$

$$\mathbf{x}_{vv} = (\cosh v \cos \theta, \cosh v \sin \theta, 0).$$

Then the first and second fundamental forms are

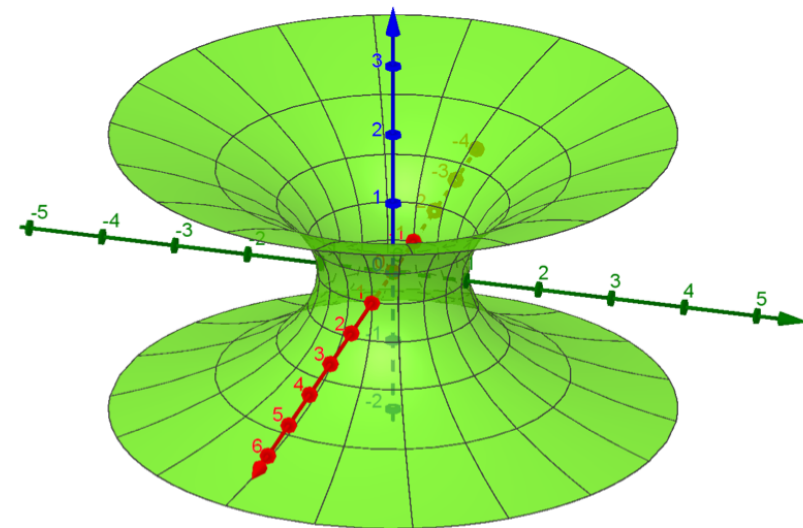
$$I = \begin{pmatrix} E & F \\ F & G \end{pmatrix} = \begin{pmatrix} \cosh^2 v & 0 \\ 0 & \cosh^2 v \end{pmatrix}$$

$$II = \begin{pmatrix} e & f \\ f & g \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

Thus the mean curvature is

$$H = \frac{1}{2} \left(\frac{gE - 2fF + eG}{EG - F^2} \right) = \frac{1}{2} \left(\frac{\cosh^2 v - \cosh^2 v}{\cosh^4 v} \right) = 0.$$

Therefore the catenoid is a minimal surface.

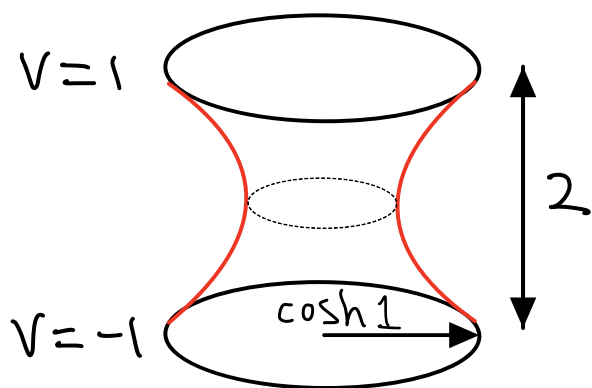


Theorem 3.4.13. Let S be a minimal surface with parametrization $\mathbf{x} : D \rightarrow \mathbb{R}^3$ such that $\mathbf{x}$ can be extended continuously to the boundary. Then S has the minimum surface area among all surfaces with the same boundary of S .

eg Consider the part of the catenoid

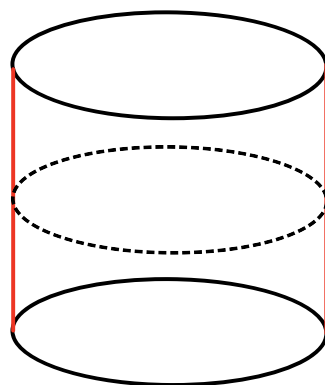
$$\mathbf{x}(\theta, v) = (\cosh v \cos \theta, \cosh v \sin \theta, v), \quad -1 < v < 1, \quad 0 < \theta < 2\pi$$

Catenoid

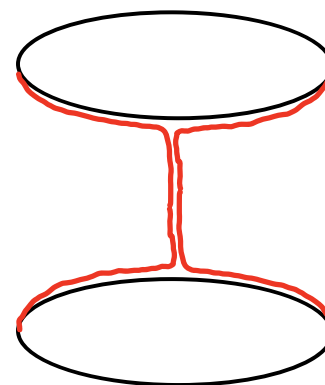


Surface area $2\pi (\cosh^2 1 - 1)$
 ≈ 8.68

Cylinder



$4\pi \cosh 1$
 ≈ 19.39



$\approx 2\pi \cosh^2 1$
 ≈ 14.96

Theorem 3.4.19. *Let S be a regular surface parametrized by $\mathbf{x}(u, v)$ and K be the Gaussian curvature of S .*

1.

$$K = \frac{\det(II)}{\det(I)}$$

where I and II are the first fundamental forms of S .

2.

$$\mathbf{n}_u \times \mathbf{n}_v = K \mathbf{x}_u \times \mathbf{x}_v$$

where $\mathbf{n}$ is the unit normal vector of S .

3.

$$K = \frac{d\sigma}{dA}$$

where A and σ are the signed area function on S and S^2 respectively.

4.

$$K = \kappa_1 \kappa_2$$

where κ_1, κ_2 are the principal curvatures associated with two orthogonal principal directions.

The following theorems relate curvature of a curve on S with κ_1, κ_2, II

Theorem 3.4.15. Let S be a regular surface and $p \in S$ be a point on S . Let C be a regular parametrized curve passing through p . Then we have

$$\kappa \cos \phi = -\langle \mathbf{T}, d\mathbf{n}_p(\mathbf{T}) \rangle$$

where $\mathbf{T}, \kappa$ are the unit tangent vector, signed curvature of C at p respectively, $d\mathbf{n}_p$ is the differential of Gauss map of S at p and ϕ is the angle between the unit normal vector $\mathbf{N}$ of C and the unit normal vector $\mathbf{n}$ of S at p . Furthermore if $\mathbf{T} = \alpha \mathbf{x}_u + \beta \mathbf{x}_v \in T_p S$, then we have

$$\kappa \cos \phi = (\alpha \quad \beta) II \begin{pmatrix} \alpha \\ \beta \end{pmatrix}$$

where II is the second fundamental form.

Definition 3.4.16 (Normal curvature). Let S be a regular surface and p be a point on S . Let $\mathbf{v} \in T_p S$ be a unit vector tangent to the surface S at p . The **normal curvature** of S at p along $\mathbf{v}$ is

$$\kappa_n(\mathbf{v}) = \kappa \cos \phi = -\langle \mathbf{v}, d\mathbf{n}_p(\mathbf{v}) \rangle$$

where κ is the curvature of a curve C which passes through p and has unit tangent vector equals to $\mathbf{v}$, and ϕ is the angle between the unit normal vectors $\mathbf{N}$ and $\mathbf{n}$ of C and S at p respectively.

Theorem 3.4.17.

$$\kappa_n(\mathbf{v}) = \kappa_1 \cos^2 \theta + \kappa_2 \sin^2 \theta.$$

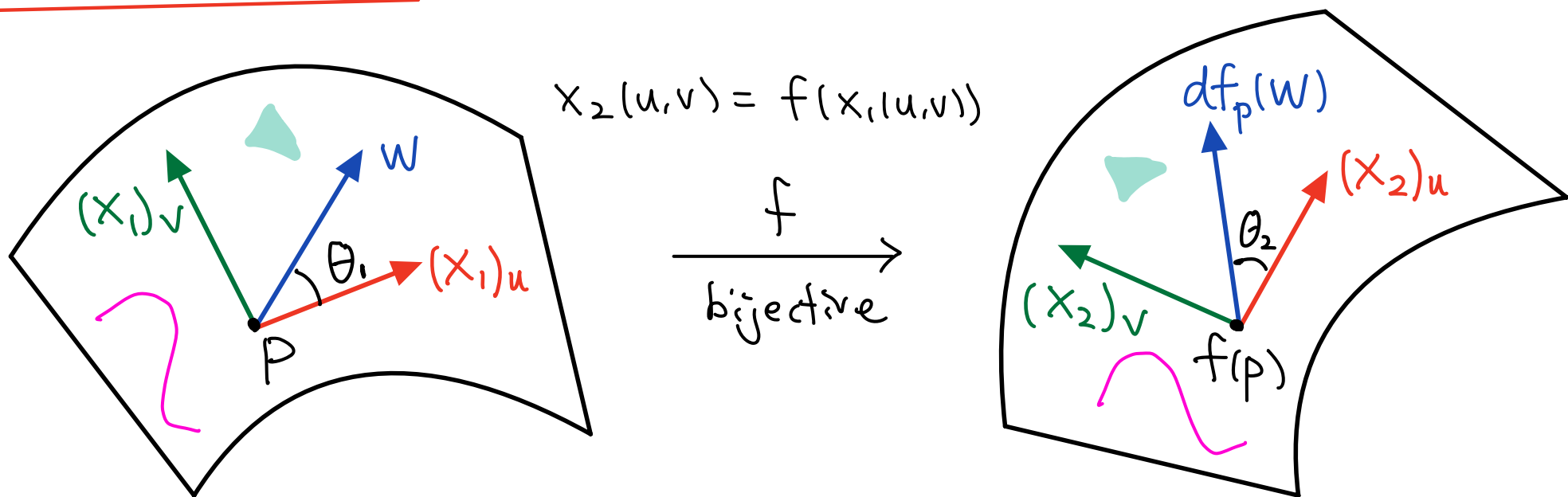
Theorem 3.4.18.

$$\kappa_1 \leq \kappa_n(\mathbf{v}) \leq \kappa_2.$$

3.5 Theorema egregium

Let S_1 be a regular surface and $f : S_1 \rightarrow S_2$ be a differentiable bijective map from S_1 to another regular surface S_2 . Then any regular parametrization $\mathbf{x}_1(u, v)$ of S_1 induces a parametrization of S_2 by $\mathbf{x}_2(u, v) = f \circ \mathbf{x}_1(u, v) = f(\mathbf{x}_1(u, v))$. Furthermore the first fundamental forms $I_1(u, v)$ and $I_2(u, v)$ on S_1 and S_2 with respect to $\mathbf{x}_1(u, v)$ and $\mathbf{x}_2(u, v)$ can both be considered as matrix valued functions of u, v . We say that $f : S_1 \rightarrow S_2$ is an isometry if $I_1(u, v) = I_2(u, v)$ for any u, v .

$$I = \begin{pmatrix} \langle \mathbf{x}_u, \mathbf{x}_u \rangle & \langle \mathbf{x}_u, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_v, \mathbf{x}_u \rangle & \langle \mathbf{x}_v, \mathbf{x}_v \rangle \end{pmatrix}$$



$$I_1 = I_2 \implies \begin{aligned} \|(x_1)_u\| &= \|(x_2)_u\| \\ \|(x_1)_v\| &= \|(x_2)_v\| \\ \|w\| &= \|df_p(w)\| \end{aligned}$$

$\theta_1 = \theta_2$
preserve arclength, area ...

Theorem 3.5.2 (Theorema egregium). Let S_1 and S_2 be two regular surfaces. Suppose S_1 and S_2 are isometric, that is, there exists isometry $f : S_1 \rightarrow S_2$ between S_1 and S_2 . Then for any $p \in S_1$, the Gaussian curvature of S_1 at p is equal to the Gaussian curvature of S_2 at $f(p)$. In other words,

$$K(f(p)) = K(p)$$

for any $p \in S_1$.

Pf It follows from Theorem 3.5.4 :

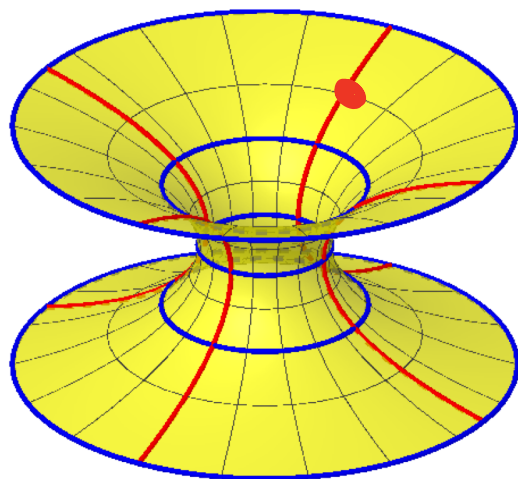
$$K = \frac{1}{4(EG - F^2)^2} \left(\begin{vmatrix} -E_{vv} + 2F_{uv} - G_{uu} & E_u & 2F_u - E_v \\ 2F_v - G_u & E & F \\ G_v & F & G \end{vmatrix} - \begin{vmatrix} 0 & E_v & G_u \\ E_v & E & F \\ G_u & F & G \end{vmatrix} \right)$$

Rmk We defined K using $\vec{n}, I, II$

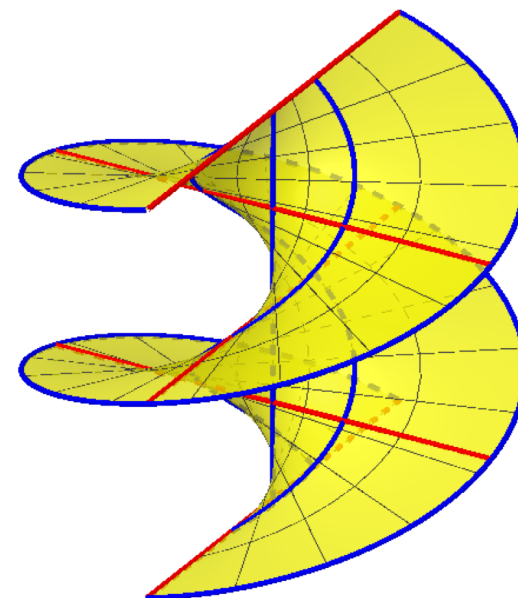
Thm 3.5.2 $\Rightarrow K$ is actually independent of $\vec{n}, II$

Example 3.5.3 (Isometry between catenoid and helicoid)

Catenoid



Helicoid



$$\begin{array}{c} f \\ \xrightarrow{\hspace{1cm}} \\ f(x_1(\theta, v)) \\ = x_2(\theta, v) \end{array}$$

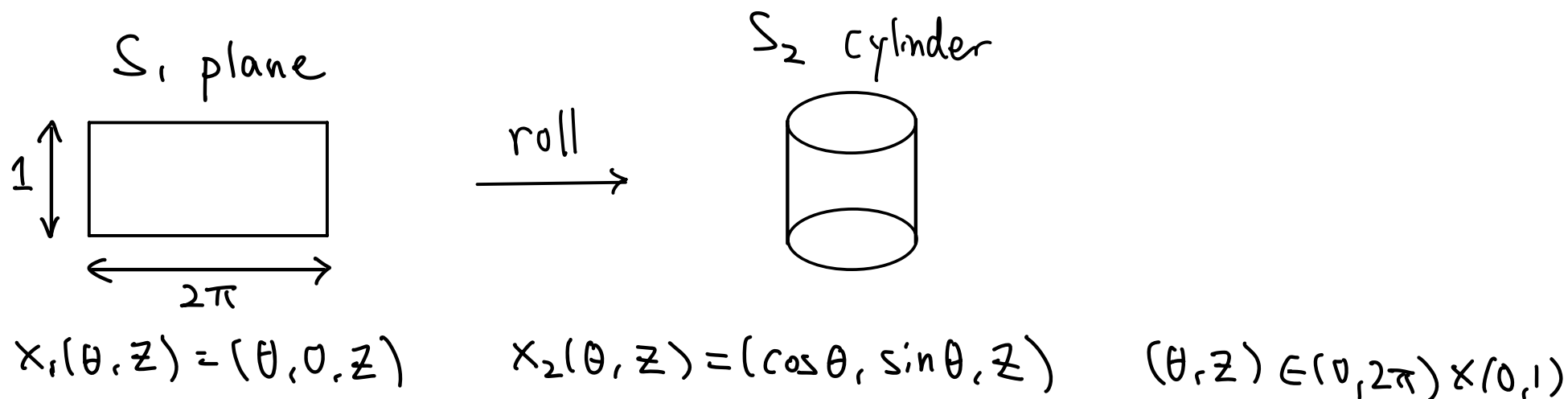
$$\mathbf{x}_1(\theta, v) = (\cosh v \cos \theta, \cosh v \sin \theta, v) \quad (\theta, v) \in (0, 2\pi) \times \mathbb{R}$$

$$\mathbf{x}_2(\theta, v) = (\sinh v \cos \theta, \sinh v \sin \theta, \theta)$$

$$I_1(\theta, v) = I_2(\theta, v) = \begin{pmatrix} \cosh^2 v & 0 \\ 0 & \cosh^2 v \end{pmatrix}$$

$$\Rightarrow f \text{ is an isometry} \Rightarrow K_1(\theta, v) = K_2(\theta, v) = -\frac{1}{\cosh^4 v}$$

Catenoid and Helicoid are both minimal surface and thus have mean curvature identically zero. However, the mean curvature of two isometric surfaces may not be identical. For example, a cylindrical surface and a plane are isometric but a cylindrical surface has nonzero mean curvature while that of a plane is zero.



$$I_1 = I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \Rightarrow f \text{ is isometry} \Rightarrow \text{same } K(\theta, z)$$

Rmk Different mean curvature $H_1 = 0$ $H_2 = \frac{1}{2}$ (or $-\frac{1}{2}$)

Theorem 3.5.4. *Let $\mathbf{x}(u, v)$ be a regular parametrized surface. Then*

$$K = \frac{1}{4(EG - F^2)^2} \left(\begin{vmatrix} -E_{vv} + 2F_{uv} - G_{uu} & E_u & 2F_u - E_v \\ 2F_v - G_u & E & F \\ G_v & F & G \end{vmatrix} - \begin{vmatrix} 0 & E_v & G_u \\ E_v & E & F \\ G_u & F & G \end{vmatrix} \right).$$

In particular, if $F = 0$ is identically zero, then

$$K = -\frac{1}{2\sqrt{EG}} \left[\left(\frac{E_v}{\sqrt{EG}} \right)_v + \left(\frac{G_u}{\sqrt{EG}} \right)_u \right].$$

Proof. Since $\det(I) = EG - F^2 = \|\mathbf{x}_u \times \mathbf{x}_v\|^2$ and $\|\mathbf{x}_u \times \mathbf{x}_v\|\mathbf{n} = \mathbf{x}_u \times \mathbf{x}_v$,

$$\begin{aligned} & K(EG - F^2)^2 \\ &= \det(I) \det(II) \\ &= \|\mathbf{x}_u \times \mathbf{x}_v\|^2 \begin{vmatrix} \langle \mathbf{x}_{uu}, \mathbf{n} \rangle & \langle \mathbf{x}_{uv}, \mathbf{n} \rangle \\ \langle \mathbf{x}_{vu}, \mathbf{n} \rangle & \langle \mathbf{x}_{vv}, \mathbf{n} \rangle \end{vmatrix} \\ &= \begin{vmatrix} \langle \mathbf{x}_{uu}, \|\mathbf{x}_u \times \mathbf{x}_v\| \mathbf{n} \rangle & \langle \mathbf{x}_{uv}, \|\mathbf{x}_u \times \mathbf{x}_v\| \mathbf{n} \rangle \\ \langle \mathbf{x}_{vu}, \|\mathbf{x}_u \times \mathbf{x}_v\| \mathbf{n} \rangle & \langle \mathbf{x}_{vv}, \|\mathbf{x}_u \times \mathbf{x}_v\| \mathbf{n} \rangle \end{vmatrix} \\ &= \begin{vmatrix} \langle \mathbf{x}_{uu}, \mathbf{x}_u \times \mathbf{x}_v \rangle & \langle \mathbf{x}_{uv}, \mathbf{x}_u \times \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vu}, \mathbf{x}_u \times \mathbf{x}_v \rangle & \langle \mathbf{x}_{vv}, \mathbf{x}_u \times \mathbf{x}_v \rangle \end{vmatrix} \\ &= \begin{vmatrix} \langle \mathbf{x}_{uu}, \mathbf{x}_{vv} \rangle - \langle \mathbf{x}_{uv}, \mathbf{x}_{uv} \rangle & \langle \mathbf{x}_{uu}, \mathbf{x}_u \rangle & \langle \mathbf{x}_{uu}, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vv}, \mathbf{x}_u \rangle & \langle \mathbf{x}_u, \mathbf{x}_u \rangle & \langle \mathbf{x}_u, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vv}, \mathbf{x}_v \rangle & \langle \mathbf{x}_v, \mathbf{x}_u \rangle & \langle \mathbf{x}_v, \mathbf{x}_v \rangle \end{vmatrix} \\ &\quad - \begin{vmatrix} 0 & \langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle & \langle \mathbf{x}_{uv}, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vu}, \mathbf{x}_u \rangle & \langle \mathbf{x}_u, \mathbf{x}_u \rangle & \langle \mathbf{x}_u, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vu}, \mathbf{x}_v \rangle & \langle \mathbf{x}_v, \mathbf{x}_u \rangle & \langle \mathbf{x}_v, \mathbf{x}_v \rangle \end{vmatrix}. \quad (\text{Proposition 1.3.17}) \end{aligned}$$

Observe that by product rule (Proposition [1.3.34](#)),

$$\begin{aligned}
\begin{pmatrix} E_u & F_u \\ F_u & G_u \end{pmatrix} &= \frac{\partial}{\partial u} \begin{pmatrix} E & F \\ F & G \end{pmatrix} \\
&= \frac{\partial}{\partial u} \begin{pmatrix} \langle \mathbf{x}_u, \mathbf{x}_u \rangle & \langle \mathbf{x}_u, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_v, \mathbf{x}_u \rangle & \langle \mathbf{x}_v, \mathbf{x}_v \rangle \end{pmatrix} \\
&= \begin{pmatrix} \langle \mathbf{x}_{uu}, \mathbf{x}_u \rangle & \langle \mathbf{x}_{uu}, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vu}, \mathbf{x}_u \rangle & \langle \mathbf{x}_{vu}, \mathbf{x}_v \rangle \end{pmatrix} + \begin{pmatrix} \langle \mathbf{x}_u, \mathbf{x}_{uu} \rangle & \langle \mathbf{x}_u, \mathbf{x}_{vu} \rangle \\ \langle \mathbf{x}_v, \mathbf{x}_{uu} \rangle & \langle \mathbf{x}_v, \mathbf{x}_{vu} \rangle \end{pmatrix} \\
&= \begin{pmatrix} 2\langle \mathbf{x}_{uu}, \mathbf{x}_u \rangle & \langle \mathbf{x}_{uu}, \mathbf{x}_v \rangle + \langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle \\ \langle \mathbf{x}_{uu}, \mathbf{x}_v \rangle + \langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle & 2\langle \mathbf{x}_{vu}, \mathbf{x}_v \rangle \end{pmatrix}.
\end{aligned}$$

Similarly

$$\begin{pmatrix} E_v & F_v \\ F_v & G_v \end{pmatrix} = \begin{pmatrix} 2\langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle & \langle \mathbf{x}_{vv}, \mathbf{x}_u \rangle + \langle \mathbf{x}_{uv}, \mathbf{x}_v \rangle \\ \langle \mathbf{x}_{vv}, \mathbf{x}_u \rangle + \langle \mathbf{x}_{uv}, \mathbf{x}_v \rangle & 2\langle \mathbf{x}_{vv}, \mathbf{x}_v \rangle \end{pmatrix}.$$

Combining the above two equalities, we obtain

$$\begin{cases} \langle \mathbf{x}_{uu}, \mathbf{x}_u \rangle = \frac{E_u}{2}, \\ \langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle = \frac{E_v}{2}, \\ \langle \mathbf{x}_{uv}, \mathbf{x}_v \rangle = \frac{G_u}{2}, \\ \langle \mathbf{x}_{vv}, \mathbf{x}_v \rangle = \frac{G_v}{2}, \\ \langle \mathbf{x}_{uu}, \mathbf{x}_v \rangle = F_u - \langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle = F_u - \frac{E_v}{2}, \\ \langle \mathbf{x}_{vv}, \mathbf{x}_u \rangle = F_v - \langle \mathbf{x}_{uv}, \mathbf{x}_v \rangle = F_v - \frac{G_u}{2}. \end{cases}$$

Moreover by considering the second derivative of $F = \langle \mathbf{x}_u, \mathbf{x}_v \rangle$ with respect to u, v , we have

$$\begin{aligned}
\langle \mathbf{x}_{uu}, \mathbf{x}_{vv} \rangle &= \frac{\partial}{\partial u} \langle \mathbf{x}_{vv}, \mathbf{x}_u \rangle - \langle \mathbf{x}_{vvu}, \mathbf{x}_u \rangle \\
&= \frac{\partial}{\partial u} \left(F_v - \frac{G_u}{2} \right) - \left(\frac{\partial}{\partial v} \langle \mathbf{x}_{uv}, \mathbf{x}_u \rangle - \langle \mathbf{x}_{uv}, \mathbf{x}_{uv} \rangle \right) \\
&= F_{uv} - \frac{G_{uu}}{2} - \left(\frac{\partial}{\partial v} \frac{E_v}{2} - \langle \mathbf{x}_{uv}, \mathbf{x}_{uv} \rangle \right) \\
&= -\frac{E_{vv}}{2} + F_{uv} - \frac{G_{uu}}{2} + \langle \mathbf{x}_{uv}, \mathbf{x}_{uv} \rangle
\end{aligned}$$

which implies

$$\langle \mathbf{x}_{uu}, \mathbf{x}_{vv} \rangle - \langle \mathbf{x}_{uv}, \mathbf{x}_{uv} \rangle = -\frac{E_{vv}}{2} + F_{uv} - \frac{G_{uu}}{2}.$$

Therefore

$$K(EG - F^2)^2 = \begin{vmatrix} -\frac{E_{vv}}{2} + F_{uv} - \frac{G_{uu}}{2} & \frac{E_u}{2} & F_u - \frac{E_v}{2} \\ F_v - \frac{G_u}{2} & E & F \\ \frac{G_v}{2} & F & G \end{vmatrix} - \begin{vmatrix} 0 & \frac{E_v}{2} & \frac{G_u}{2} \\ \frac{E_v}{2} & E & F \\ \frac{G_u}{2} & F & G \end{vmatrix}$$

as desire. If particular, if $F = 0$, then

$$\begin{aligned}
K &= \frac{1}{4E^2G^2} \begin{vmatrix} -E_{vv} - G_{uu} & E_u & -E_v \\ -G_u & E & 0 \\ G_v & 0 & G \end{vmatrix} - \begin{vmatrix} 0 & E_v & G_u \\ E_v & E & 0 \\ G_u & 0 & G \end{vmatrix} \\
&= \frac{1}{4E^2G^2} (-EG E_{vv} - EGG_{uu} + GE_u G_u + EE_v G_v + GE_v^2 + EG_u^2) \\
&= -\frac{E_{vv}}{4EG} - \frac{G_{uu}}{4EG} + \frac{E_u G_u}{4E^2G} + \frac{E_v G_v}{4EG^2} + \frac{E_v^2}{4E^2G} + \frac{G_u^2}{4EG^2}.
\end{aligned}$$

Observe that

$$\begin{cases} \left(\frac{E_v}{\sqrt{EG}} \right)_v = \frac{E_{vv}}{\sqrt{EG}} - \frac{E_v^2}{2E\sqrt{EG}} - \frac{E_v G_v}{2G\sqrt{EG}} \\ \left(\frac{G_u}{\sqrt{EG}} \right)_u = \frac{G_{uu}}{\sqrt{EG}} - \frac{G_u^2}{2G\sqrt{EG}} - \frac{E_u G_u}{2E\sqrt{EG}} \end{cases}$$

Hence

$$\begin{aligned}
&\left(\frac{E_v}{\sqrt{EG}} \right)_v + \left(\frac{G_u}{\sqrt{EG}} \right)_u \\
&= \frac{E_{vv}}{\sqrt{EG}} - \frac{E_v^2}{2E\sqrt{EG}} - \frac{E_v G_v}{2G\sqrt{EG}} + \frac{G_{uu}}{\sqrt{EG}} - \frac{G_u^2}{2G\sqrt{EG}} - \frac{E_u G_u}{2E\sqrt{EG}} \\
&= -2\sqrt{EG} \left(-\frac{E_{vv}}{4EG} + \frac{E_v^2}{4E^2G} + \frac{E_v G_v}{4EG^2} - \frac{G_{uu}}{4EG} + \frac{G_u^2}{4EG^2} + \frac{E_u G_u}{4E^2G} \right) \\
&= -2K\sqrt{EG}
\end{aligned}$$

and the result follows. $\square$