

Honors Mathematics III

RC 7

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The Normal Vector of a Curve

Definition. The **unit normal vector** $N : \mathcal{C} \rightarrow \mathbb{R}$ of a smooth C^2 -curve with parametrization $\gamma : I \rightarrow V$ is

$$N \circ \gamma(t) := \frac{(T \circ \gamma)'(t)}{\|(T \circ \gamma)'(t)\|}, \quad t \in \text{int } I$$

Note. The unit normal vector does not depend on γ on

- ▶ magnitude and
- ▶ orientation.

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Curve Length

2.3.25. **Theorem.** $\mathcal{C} \subset V$ is a smooth and **open** curve with parametrization $\gamma : [a, b] \rightarrow \mathcal{C}$. Then \mathcal{C} is rectifiable iff

$$\int_a^b \|\gamma'(t)\| dt < \infty$$

and the **curve length** is

$$\ell(\mathcal{C}) = \int_a^b \|\gamma'(t)\| dt$$

which is independent of γ .

Curve Length

The *length function* is defined as

$$(\ell \circ \gamma)(t) = \int_a^t \|\gamma'(\tau)\| d\tau$$

The curve length gives the *natural parametrization* of an oriented curve \mathcal{C} .

$$\gamma = \ell : I \rightarrow \mathcal{C}, \quad \text{int } I = (0, \ell(\mathcal{C}))$$

Note. Then we also obtain

$$\|\gamma'(t)\| = \frac{d\ell \circ \gamma(t)}{dt}$$

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Curvature

Definition. The *curvature* of a smooth C^2 -curve $\mathcal{C} \subset V$ is

$$\kappa : \mathcal{C} \rightarrow \mathbb{R}, \quad \kappa \circ \ell^{-1}(s) := \left\| \frac{d}{ds}(T \circ \ell^{-1}(s)) \right\|$$

where T is the unit tangent vector and $\ell^{-1} : I \rightarrow \mathcal{C}$ is the curve length parametrization of \mathcal{C} . For parametrization that is *not* the length parametrization,

$$\kappa \circ \gamma(t) = \frac{\|(T \circ \gamma)'(t)\|}{\|\gamma'(t)\|}$$

Note. The curvature κ does not depend on the orientation of \mathcal{C} .

Curvature in \mathbb{R}^3

2.3.31. Lemma. Let $\mathcal{C} \subset \mathbb{R}^3$ be a smooth C^2 -curve with parametrization $\gamma : I \rightarrow \mathcal{C}$, then

$$\kappa \circ \gamma(t) = \frac{\|\gamma'(t) \times \gamma''(t)\|}{\|\gamma'(t)\|^3}$$

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The Directional Derivative

Definition. Let $\Omega \subset \mathbb{R}^n$ be an open set, $f : \Omega \rightarrow \mathbb{R}$ continuous and $h \in \mathbb{R}^n$, $\|h\| = 1$ be a unit vector. Then the **directional derivative** $D_h f$ in the direction h is defined by

$$D_h f|_x := \left. \frac{d}{dt} f(x + th) \right|_{t=0}$$

if the right-hand side exists.

Note.

- ▶ The directional derivative is the derivative of f at x along the line segment joining x and $x + h$. It gives the slope of the tangent line of f at x in the direction of h .
- ▶ The directional derivative is a number.

The Directional Derivative

Results.

- ▶ The tangent line of $f : \mathbb{R}^n \rightarrow \mathbb{R}$ at x in the direction h :

$$t_{f,x;h}(s) = \begin{pmatrix} x + sh \\ f(x) + D_h f|_x s \end{pmatrix}$$

- ▶ If f is differentiable and the line segment is parametrized by $\gamma(t) = x + th$,

$$D_h f|_x = Df|_x h = \langle \nabla f(x), h \rangle$$

The Directional Derivative

Example. Find the directional derivative of the function

$$f(x, y) = \ln(x^2 + y^2)^{1/2}$$

at $(1, 1)$ along the direction $(2, 1)$.

The Directional Derivative

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$$f(x, y) = \ln(x^2 + y^2)^{1/2}$$

at $(1, 1)$ along the direction $(2, 1)$.

Solution. f is smooth in $(0, \infty) \times (0, \infty)$.

$$\nabla f(x, y) = \begin{pmatrix} \frac{x}{x^2 + y^2} \\ \frac{y}{x^2 + y^2} \end{pmatrix}$$

Then

$$Df_h|_{(1,1)} = \langle \nabla f|_{(1,1)}, h \rangle = \frac{3\sqrt{5}}{10}$$

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The Normal Derivative in \mathbb{R}^2

Definition. Let $\Omega \subset \mathbb{R}^2$ be an open set, $f : \Omega \rightarrow \mathbb{R}$ and \mathcal{C} a simple smooth C^2 -curve in Ω . Let $p \in \mathcal{C}$ and $N(p)$ denote the normal vector at p . Then

$$\left. \frac{\partial f}{\partial n} \right|_p := D_{N(p)} f|_p$$

is called the **normal derivative of f at p** with respect to the curve \mathcal{C} .

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The Gradient

The **gradient** $\nabla f(x)$ is the transpose of the Jacobian.

- ▶ $\nabla f(x)$ points in the direction of the greatest directional derivative of f at x .

$$D_h f(x) = \langle \nabla f(x), h \rangle = |\nabla f(x)| \cos \angle(\nabla f(x), h)$$

- ▶ $\nabla f(x)$ is perpendicular to the contour line of f at x .

$$\langle \nabla f(x), h_0 \rangle = 0$$

The Tangent Plane

Definition. Let $\Omega \subset \mathbb{R}^n$ be open and $f : \Omega \rightarrow \mathbb{R}$ differentiable at $x_0 \in \Omega$. Then the equation

$$x_{n+1} = Tf(x; x_0), \quad x = (x_1, \dots, x_n) \in \mathbb{R}^n$$

defines the **tangent plane** to the graph $\Gamma(f) \in \mathbb{R}^n \times \mathbb{R}$ of f at the point $(x_0, f(x_0)) \in \mathbb{R}^{n+1}$.

The Tangent Plane

Results.

- ▶ The tangent plane in \mathbb{R}^3 is found by

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x_0 \\ y_0 \\ f(x_0, y_0) \end{pmatrix} + s \begin{pmatrix} 1 \\ 0 \\ \frac{\partial f}{\partial x}(x_0, y_0) \end{pmatrix} + t \begin{pmatrix} 0 \\ 1 \\ \frac{\partial f}{\partial y}(x_0, y_0) \end{pmatrix}$$

- ▶ The vectors

$$t_1 := \begin{pmatrix} 1 \\ 0 \\ \frac{\partial f}{\partial x}(x_0, y_0) \end{pmatrix}, \quad t_2 := \begin{pmatrix} 0 \\ 1 \\ \frac{\partial f}{\partial y}(x_0, y_0) \end{pmatrix}$$

give the **tangent vectors** to the graph $\Gamma(f)$ at $(x_0, y_0, f(x_0, y_0))$ with normal vector given by

$$n = t_1 \times t_2 = \left(-\frac{\partial f}{\partial x}(x_0, y_0), -\frac{\partial f}{\partial y}(x_0, y_0), 1 \right)^T$$

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The Second Derivative

Definition. Let X, V be finite-dimensional normed vector spaces and $\Omega \subset X$ an open set. A function $f : \Omega \rightarrow V$ is said to be **twice differentiable** at $x \in \Omega$ if

- ▶ f is differentiable in an open ball $B_\varepsilon(x)$ around x and
- ▶ the function $Df : B_\varepsilon(x) \rightarrow \mathcal{L}(X, V)$ is differentiable at x .

The **second derivative** (if it exists) is a map:

$$D(Df) =: Df^2 : \Omega \rightarrow \mathcal{L}(X, \mathcal{L}(X, V))$$

and is found by

$$Df|_{x+h} = Df|_x + D^2f|_x h + o(h)$$

The Second Derivative

Example. (RC.5.e.g.2.) Calculate the first, second and the third derivatives of the map

$$\Phi : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \text{Mat}(n \times n; \mathbb{R}), \quad \Phi(A) = A^3$$

The Second Derivative

Example. (RC.5.e.g.2.) Calculate the first, second and the third derivatives of the map

$$\Phi : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \text{Mat}(n \times n; \mathbb{R}), \quad \Phi(A) = A^3$$

$$\Phi(A + H) = A^3 + A^2H + AHA + HA^2 + o(H)$$

$$D\Phi|_A H = A^2H + AHA + HA^2$$

then

$$\begin{aligned} D\Phi_{A+J} H &= AH + AHA + HA^2 + AJH + JAH + AHJ + JHA + \\ &\quad + HAJ + HJA + o(J) \end{aligned}$$

$$D^2\Phi|_A [H, J] = AJH + AHJ + JAH + JHA + HAJ + HJA$$

The Second Derivative

The Hessian. For a differentiable potential function $f : \mathbb{R}^n \rightarrow \mathbb{R}$. The derivative is given by the Jacobian:

$$Df|_x = \left(\left. \frac{\partial f}{\partial x_1} \right|_x \quad \cdots \quad \left. \frac{\partial f}{\partial x_n} \right|_x \right), \quad Df|_x \in \mathcal{L}(\mathbb{R}^n, \mathbb{R})$$

and the second derivative is found by **Hessian** where

$$\text{Hess } f(x) = D(\nabla f)|_x = \begin{pmatrix} \left. \frac{\partial^2 f}{\partial x_1 \partial x_1} \right|_x & \left. \frac{\partial^2 f}{\partial x_2 \partial x_1} \right|_x & \cdots & \left. \frac{\partial^2 f}{\partial x_n \partial x_1} \right|_x \\ \vdots & \vdots & & \vdots \\ \left. \frac{\partial^2 f}{\partial x_1 \partial x_n} \right|_x & \left. \frac{\partial^2 f}{\partial x_2 \partial x_n} \right|_x & \cdots & \left. \frac{\partial^2 f}{\partial x_n \partial x_n} \right|_x \end{pmatrix}$$

$$D^2 f|_x h = \text{Hess } f(x) h$$

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Bilinear Maps

- ▶ **The Hessian as a bilinear map.** $D^2f|_x : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}, \quad (h, \tilde{h}) \mapsto \langle \text{Hess}f(x)h, \tilde{h} \rangle.$
- ▶ **The second derivative as a bilinear map.** $L \in \mathcal{X}, \mathcal{L}(\mathcal{X}, \mathcal{V})$, then $Lx_1 \in \mathcal{L}(X, V)$ and $(Lx_1)x_2 = \tilde{L}(x_1, x_2) \in V.$
- ▶ **Multilinear Maps.** Let X, V be finite-dimensional normed vector spaces. The set of multilinear maps from X to V is denoted by

$$\mathcal{L}^{(n)}(X, V) := \{L : X \times \cdots \times X \rightarrow V : L \text{ is linear in each component}\}$$

When $V = \mathbb{R}$, an element of $\mathcal{L}^{(n)}(X, V)$ is called a **multilinear form**.

Bilinear forms on \mathbb{R}^n

- ▶ Every linear map $L \in (\mathbb{R}^n)^*$ has the form $L = \langle z, \cdot \rangle$ for some $z \in \mathbb{R}^n$.
- ▶ Interpret an element $A \in \mathcal{L}(\mathbb{R}^n, \mathcal{L}(\mathbb{R}^n, \mathbb{R}))$ as a linear map, $A : y \mapsto L_y := \langle z_y, \cdot \rangle$, $z_y = A(y)$.
- ▶ A is actually a matrix $A : y \mapsto z_y$.
- ▶ For every $y \in \mathbb{R}^n$ we obtain a linear map $\langle Ay, \cdot \rangle \in \mathcal{L}(\mathbb{R}^n, \mathbb{R})$.
- ▶ Then $L_y x = \langle Ay, x \rangle = L(x, y)$.

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Schwarz's Theorem

2.5.5. Schwarz's Theorem. Let X, V be normed vector spaces and $\Omega \subset X$ an open set. Let $f \in C^2(\Omega, V)$. Then $D^2f|_x \in \mathcal{L}^{(2)}(X \times X, V)$ is symmetric for all $x \in \Omega$.

$$D^2f(u, v) = D^2f(v, u) \quad \text{for all } u, v \in X$$

This implies that if f is twice continuously differentiable, the Hessian of f at x is symmetric.

$$\langle \text{Hess } f(x)y, z \rangle = \langle \text{Hess } f(x)z, y \rangle, \quad \frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}$$

Note. This will be the case if all second-order partial derivatives are continuous.

Schwarz's Theorem

Sufficiency of twice-differentiability. The Schwarz's theorem will hold if all the second-order partial derivatives are continuous.

Example. The symmetry may be broken if the function fails to have differentiable partial derivatives.

$$f(x, y) = \begin{cases} \frac{xy(x^2 - y^2)}{x^2 + y^2} & \text{for } (x, y) \neq (0, 0) \\ 0 & \text{for } (x, y) = (0, 0) \end{cases}$$

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The second partial derivatives are not continuous at $(0, 0)$

$$\frac{\partial^2}{\partial x \partial y} f \Big|_{(0,0)} = \lim_{h \rightarrow 0} \frac{\partial_y f|_{(h,0)} - \partial_y f|_{(0,0)}}{h} = 1$$

$$\frac{\partial^2}{\partial y \partial x} f \Big|_{(0,0)} = \lim_{h \rightarrow 0} \frac{\partial_x f|_{(0,h)} - \partial_x f|_{(0,0)}}{h} = -1$$

Schwarz's Theorem

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Example. The symmetry may be broken if the function fails to have differentiable partial derivatives.

$$f(x, y) = \begin{cases} \frac{xy(x^2 - y^2)}{x^2 + y^2} & \text{for } (x, y) \neq (0, 0) \\ 0 & \text{for } (x, y) = (0, 0) \end{cases}$$

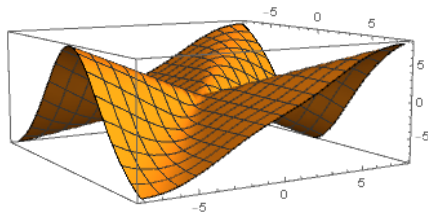


Figure: The First Partial Derivative $\partial_x f$

Exercises

Exercise 1. Calculate the length of the curve between $t = 0$ and $t = \frac{\pi}{4}$ parametrized by

$$\gamma(t) = \begin{pmatrix} t \\ \ln \cos t \end{pmatrix}$$

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Solution.

$$\begin{aligned} \ell &= \int_0^{\pi/4} \sqrt{1 + \frac{\sin^2 t}{\cos^2 t}} dt \\ &= \int_0^{\pi/4} \frac{1}{\cos t} dt \\ &= \int_0^{\sqrt{2}/2} \frac{1}{1-x^2} dx \\ &= \frac{1}{2} (\ln(1+x) - \ln(1-x)) \Big|_0^{\sqrt{2}/2} = \frac{1}{2} \ln(3+2\sqrt{2}) \end{aligned}$$

Exercises.

Exercise 2. Find the curvature of the curve where the points (x, y) are parametrized with $t \in (0, \infty)$ by

$$x(t) = \int_0^t \frac{\cos u}{\sqrt{u}} du$$

$$y(t) = \int_0^t \frac{\sin u}{\sqrt{u}} du$$

Exercises.

Exercise 2. Find the curvature of the curve where the points (x, y) are parametrized with $t \in (0, \infty)$ by

$$x(t) = \int_0^t \frac{\cos u}{\sqrt{u}} du$$
$$y(t) = \int_0^t \frac{\sin u}{\sqrt{u}} du$$

Solution.

$$x'(t) = \frac{\cos t}{\sqrt{t}}, \quad y'(t) = \frac{\sin t}{\sqrt{t}}$$

Then

$$\kappa = \frac{\|(T \circ \gamma)'\|}{\|\gamma'(t)\|} = \sqrt{t}$$

Exercises.

Exercise 3. Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be homogeneous of degree of k . That means that f satisfies

$$f(tx) = t^k f(x) \quad \text{for all } x \neq 0 \text{ and } t > 0$$

Show that

$$Df|_x x = k \cdot f(x)$$

Exercises.

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Show that

$$Df|_x x = k \cdot f(x)$$

Solution. We differentiate with respect to t :

$$\frac{d}{dt} f(tx) = Df|_{tx} x, \quad \frac{d}{dt} t^k f(x) = k t^{k-1} f(x)$$

Setting $t = 1$, we obtain

$$Df|_x x = k f(x)$$

Frenet-Serret Formulas

Results. T is the unit tangent vector, N is the unit normal vector, $B = T \times N$ is the binormal unit vector. Then T, N, B form an orthonormal basis in \mathbb{R}^3 .

$$\frac{dT}{ds} = \kappa N, \quad \frac{dN}{ds} = -\kappa T + \tau B, \quad \frac{dB}{ds} = -\tau N$$

and

$$T(t) = \frac{\gamma'(t)}{\|\gamma'(t)\|}$$

$$N(t) = \frac{(T \circ \gamma)'(t)}{\|(T \circ \gamma)'(t)\|}$$

$$B(t) = T \times N$$

Note. $\frac{ds}{dt} = \|\gamma'(t)\|$.

$$\kappa = \frac{\|\gamma'(t) \times \gamma''(t)\|}{\|\gamma'(t)\|^3}$$

$$\tau = \frac{\det(\gamma'(t), \gamma''(t), \gamma'''(t))}{\|\gamma'(t) \times \gamma''(t)\|^2}$$

Exercises

Exercise 4. Find the first and second derivatives of the following functions:

1. $f : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \mathbb{R}, \quad f(A) = \text{tr}(A^2).$
2. $g : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \mathbb{R}, \quad g(A) = (\text{tr} A)^2.$

Exercises

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1. $f : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \mathbb{R}, \quad f(A) = \text{tr}(A^2).$

2. $g : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \mathbb{R}, \quad g(A) = (\text{tr} A)^2.$

Solution.

► First derivatives:

$$Df|_A H = 2\text{tr}(AH), \quad Dg|_A H = 2\text{tr}(A) \cdot \text{tr}(H)$$

► Second derivatives:

$$D^2 f|_A [J, H] = 2\text{tr}(JH), \quad D^2 g|_A [J, H] = 2\text{tr}(J) \cdot \text{tr}(H)$$

Exercises

Exercise 5. Find the second derivative of the determinant of an invertible matrix A .

$$\det : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \mathbb{R}$$

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$$\det : \text{Mat}(n \times n; \mathbb{R}) \rightarrow \mathbb{R}$$

Solution.

1. First derivative of \det : $(D\det|_A H) = \det(A) \cdot \text{tr}(A^{-1}H)$.
2. First derivative of tr : tr .
3. First derivative of $(\cdot)H : (\cdot)H$.
4. First derivative of $(\cdot)^{-1} : D(\cdot)^{-1}|_A H = -A^{-1}HA^{-1}$. (RC.5)

$$D^2\det|_A[J, H] = \det(A) \left(\text{tr}(A^{-1}J) \cdot \text{tr}(A^{-1}H) - \text{tr}(A^{-1}JA^{-1}H) \right)$$

Summary

- ▶ Tangent line at a point $\gamma(t_0) : \{\gamma(t_0) + \gamma'(t_0)t, t \in \mathbb{R}\}$
- ▶ Unit tangent vector: $T \circ \gamma(t) = \frac{\gamma'(t)}{\|\gamma'(t)\|}$.
- ▶ Unit normal vector: $N \circ \gamma(t) = \frac{(T \circ \gamma)'(t)}{\|(T \circ \gamma)'(t)\|}$.
- ▶ Open curve length: $\ell(\mathcal{C}) = \int_a^b \|\gamma'(t)\| dt$.
- ▶ Curve length function: $\ell \circ \gamma(t) = \int_a^t \|\gamma'(\tau)\| d\tau$.
- ▶ $\|\gamma'(t)\| = \frac{d(\ell \circ \gamma)(t)}{dt}$.
- ▶ Curvature: $\kappa \circ \gamma(t) = \kappa \circ \ell^{-1}(s)|_{s=\ell \circ \gamma(t)} = \frac{\|(T \circ \gamma)'(t)\|}{\|\gamma'(t)\|}$.
- ▶ Curvature in \mathbb{R}^3 :
$$\kappa \circ \gamma(t) = \kappa \circ \ell^{-1}(s)|_{s=\ell \circ \gamma(t)} = \frac{\|\gamma'(t) \times \gamma''(t)\|}{\|\gamma'(t)\|^3}.$$

Summary

- ▶ Directional derivative $D_h f$ along h : $D_h f|_x = \left. \frac{d}{dt} f(x + th) \right|_{t=0}$.
- ▶ Directional derivative of smooth functions: $D_h f|_x = Df|_x h = \langle \nabla f(x), h \rangle$.
- ▶ The tangent line of $f : \mathbb{R}^n \rightarrow \mathbb{R}$ at x along h : $t_{f,x;h}(x) = \begin{pmatrix} x + sh \\ f(x) + D_h f|_x s \end{pmatrix}$.
- ▶ The normal derivative in \mathbb{R}^2 : $\left. \frac{\partial f}{\partial n} \right|_p = D_{N(p)} f|_p$.
- ▶ Best linear approximation of f at x_0 : $Tf(\cdot; x_0) = f(x_0) + Df|_{x_0}(\cdot - x_0)$.
- ▶ The tangent plane to the graph $\Gamma(f)$ at $(x_0, f(x_0))$: $x_{n+1} = Tf(x; x_0)$ with tangent vectors and normal vector:

$$t_1 = (1, 0, \left. \frac{\partial f}{\partial x} \right|_{x_0})^T, \quad t_2 = (0, 1, \left. \frac{\partial f}{\partial y} \right|_{x_0})^T, \quad n = (-\left. \frac{\partial f}{\partial x} \right|_{x_0}, -\left. \frac{\partial f}{\partial y} \right|_{x_0}, 1)^T$$

Summary

- *Frenet-Serret formulas.*

$$B = T \times N, \quad \frac{dT}{ds} = \kappa N, \quad \frac{dB}{ds} = -\tau N, \quad \frac{dN}{ds} = -\kappa T + \tau B$$



$$\frac{d}{dt} \begin{pmatrix} T \\ N \\ B \end{pmatrix} = \|\gamma'(t)\| \begin{pmatrix} 0 & \kappa & 0 \\ -\kappa & 0 & \tau \\ 0 & -\tau & 0 \end{pmatrix} \begin{pmatrix} T \\ N \\ B \end{pmatrix}$$



$$\begin{aligned} \gamma'' &= \ell'' T + \kappa(\ell')^2 N, & \gamma' \times \gamma'' &= \kappa(\ell')^2 B \\ \gamma''' &= (\ell''' - \kappa^2(\ell')^3) T + (3\kappa\ell'\ell'' + \kappa'(\ell')^2) N + \kappa\tau(\ell')^3 B \end{aligned}$$

- $\tau = \frac{\langle \gamma' \times \gamma'', \gamma''' \rangle}{\|\gamma' \times \gamma''\|^2} = \frac{\det(\gamma', \gamma'', \gamma''')}{\|\gamma' \times \gamma''\|^2}.$

Thanks for your attention!