

Electrodynamics Notes

Phillips Academy Andover

PRO601 Physics IP: Advanced Electrodynamics with Jeffrey Shi
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Introduction

This document contains my notes for my Spring Term Physics IP (Independent Project) entitled *Advanced Electrodynamics*, taken during my last term at Phillips Academy Andover. Jeffrey Shi was my partner for the group IP and our mentor was Ms. Fei Yao. Our main reference text is *Introduction to Electrodynamics*, 4th edition by Griffiths.

I have also included some of our solutions to selected problems from the text. The numbering of these problems follows the book, rather than this document. We tried to tackle every problem marked with a bullet • or an exclamation point !. We also tackled any other problems that we found interesting.

1 Vector Analysis

§1.1 Vector Algebra

§1.1.1 Vector Operations

§1.1.2 Vector Algebra: Component Form

These first two subsections reviews the basics of vectors; for the purpose of these notes, I will assume the reader already has a strong basic understanding of vectors, so they will be left empty.

§1.1.3 Triple Products

The output of a cross product $\vec{b} \times \vec{c}$ is a vector, so it can be dotted or crossed with another vector \vec{a} to form a triple product.

The **scalar triple product** $\vec{a} \cdot (\vec{b} \times \vec{c})$ gives the signed volume of the parallelepiped generated by $\vec{a}, \vec{b}, \vec{c}$. Therefore,

$$\text{Claim 1.1.1 — } \vec{a} \cdot (\vec{b} \times \vec{c}) = \vec{b} \cdot (\vec{c} \times \vec{a}) = \vec{c} \cdot (\vec{a} \times \vec{b}).$$

Notice that we preserved the cyclic order of the vectors in this case; if we flip the order (e.g. $\vec{a} \cdot (\vec{c} \times \vec{b})$), then we obtain the same value with opposite sign. In component form, we have

$$\vec{a} \cdot (\vec{b} \times \vec{c}) = \det \begin{bmatrix} a_x & a_y & a_z \\ b_x & b_y & b_z \\ c_x & c_y & c_z \end{bmatrix},$$

The **vector triple product** $\vec{a} \times (\vec{b} \times \vec{c})$ can be simplified by the so-called “*bac – cab* rule”:

$$\text{Claim 1.1.2 — } \vec{a} \times (\vec{b} \times \vec{c}) = \vec{b}(\vec{a} \cdot \vec{c}) - \vec{c}(\vec{a} \cdot \vec{b}).$$

Note that cross-products are not associative, so the order of parentheses matters. By repeated application of this fact, it is never necessary for an expression to contain more than one cross product in any term. For example,

$$(\vec{a} \times \vec{b}) \cdot (\vec{c} \times \vec{d}) = (\vec{b} \times (\vec{c} \times \vec{d})) \cdot \vec{a} = (\vec{c}(\vec{b} \cdot \vec{d}) - \vec{d}(\vec{b} \cdot \vec{c})) \cdot \vec{a} = (\vec{a} \cdot \vec{c})(\vec{b} \cdot \vec{d}) - (\vec{a} \cdot \vec{d})(\vec{b} \cdot \vec{c}),$$

where in the first equality we utilized the scalar triple product identity $\vec{a} \cdot (\vec{b} \times \vec{c}) = \vec{b} \cdot (\vec{c} \times \vec{a})$.

§1.1.4 Position, Displacement, and Separation Vectors

Definition 1.1.3. Given a coordinate system with origin \mathcal{O} , the vector from \mathcal{O} to some point is the **position vector**

$$\vec{r} := x\hat{x} + y\hat{y} + z\hat{z},$$

where (x, y, z) are the Cartesian coordinates of the point.

Note that the distance from the point to the origin is

$$r = |\vec{r}| = \sqrt{x^2 + y^2 + z^2}$$

and the unit vector in the direction of \vec{r} is

$$\hat{r} = \frac{\vec{r}}{r} = \frac{x\hat{x} + y\hat{y} + z\hat{z}}{\sqrt{x^2 + y^2 + z^2}}.$$

Definition 1.1.4. The **infinitesimal displacement vector** from (x, y, z) to $(x + dx, y + dy, z + dz)$ is denoted by

$$d\vec{\ell} = dx\hat{x} + dy\hat{y} + dz\hat{z}.$$

Oftentimes in electrodynamics, we encounter problems involving two points:

- A **source point** \vec{r}' , where an electric charge is located,
- A **field point** \vec{r} , where we are calculating the electric or magnetic field.

In the situation when $\vec{r}' \neq \vec{0}$, it is useful to adopt a shorthand notation for the **separation vector** from the source point to the field point; we will use the script letter

$$\vec{r} := \vec{r} - \vec{r}'.$$

§1.1.5 How Vectors Transform

What exactly is a vector? You might have learned it as “a quantity with a magnitude and direction,” but this isn’t exactly satisfactory...

Perhaps a 3-dimensional vector is simply anything with three real components that combine properly under addition. However, what if we considered some silly example like a barrel with N_x pear, N_y apples, and N_z bananas? Is $\vec{N} = N_x\hat{x} + N_y\hat{y} + N_z\hat{z}$ suddenly a vector? Surely not—what is the direction? What is wrong with it?

The answer is that \vec{N} *doesn’t transform properly when you change coordinates*.

In three dimensions, if a x, y, z system is rotated to some $\bar{x}, \bar{y}, \bar{z}$ system, then the coordinates of some vector \vec{a} undergo a matrix transformation

$$\begin{bmatrix} \bar{a}_x \\ \bar{a}_y \\ \bar{a}_z \end{bmatrix} = \begin{bmatrix} R_{xx} & R_{xy} & R_{xz} \\ R_{yx} & R_{yy} & R_{yz} \\ R_{zx} & R_{zy} & R_{zz} \end{bmatrix} \cdot \begin{bmatrix} a_x \\ a_y \\ a_z \end{bmatrix} \quad \text{or} \quad \vec{\bar{a}} = R\vec{a},$$

where R_{ij} are some products of trigonometric functions (or it could simply be any arbitrary linear transformation of \mathbb{R}^3). More compactly, we can write it as

$$\bar{a}_i = \sum_{j=1}^3 R_{ij} a_j,$$

where index 1 stands for x , etc.

The components of \vec{N} clearly don’t transform in this way.

Definition 1.1.5. Formally, a **vector** is any set of three components that transforms in the same manner as a displacement when you change coordinates.

Therefore, displacement is the *model* for the behavior of all vectors.

Definition 1.1.6. A **second-order tensor** is a quantity with *nine* components: $T_{xx}, T_{xy}, T_{xz}, T_{yx}, \dots, T_{zz}$. These transform with *two* factors of R :

$$\bar{T}_{ij} = \sum_{k=1}^3 \sum_{\ell=1}^3 R_{ik} R_{j\ell} T_{k\ell}.$$

You can also rewrite this in matrix form as $\bar{T} = RTR^T$.

In general, an n th-order tensor has n indices and 3^n components (in \mathbb{R}^3), and transforms with n factors of R . Therefore, a scalar is a tensor of order zero (indeed, note that scalars never change under coordinate changes) and a vector is a tensor of order one.

§1.2 Differential Calculus

§1.2.1 “Ordinary” Derivatives

Suppose we have a single-variable function $f(x)$. What does the derivative $\frac{df}{dx}$ do?

The derivative of a function df/dx tells us how rapidly the function f varies when we change x by a tiny amount dx .

In other words, if x is incremented by an infinitesimal amount dx and f changes by an amount df , then the derivative is the proportionality factor:

$$df = \left(\frac{df}{dx} \right) dx.$$

From a geometric perspective, the derivative $\frac{df}{dx}$ gives the *slope* at any point on the graph of f against x .

§1.2.2 Gradient

That’s all well and good for one variable. What about functions of three variables, say temperature $T(x, y, z)$ in \mathbb{R}^3 ?

How do we generalize the notion of the derivative to functions like T ?

We want the derivative to tell us how T changes when we move a tiny distance. However, T may change at a different rate depending on the direction we move in! In a room, nudging z will probably have a much higher affect on the temperature than nudging x or y the same amount.

Fortunately, the situation is not so bad: it turns out that when we zoom in enough, a differentiable function T will always look like a plane, linear in all variables:

$$dT = \left(\frac{\partial T}{\partial x} \right) dx + \left(\frac{\partial T}{\partial y} \right) dy + \left(\frac{\partial T}{\partial z} \right) dz.$$

The proportionality constants are *partial derivatives*, which measure the change in T when you hold all variables constant except for the one in the denominator: therefore, $\frac{\partial T}{\partial x}$ measures the rate of change of T when you move only x by an infinitesimal amount.

This is a rather nice result: we don’t need to compute the rate of change of T among all possible directions; three directions x, y, z suffice.

We can also use it to define another concept. The equation is reminiscent of a dot product:

$$dT = \left(\frac{\partial T}{\partial x} \hat{x} + \frac{\partial T}{\partial y} \hat{y} + \frac{\partial T}{\partial z} \hat{z} \right) \cdot (dx \hat{x} + dy \hat{y} + dz \hat{z}).$$

We give the first vector in this dot product a name:

Definition 1.2.1. The **gradient** of a scalar function T is the vector quantity

$$\nabla T := \frac{\partial T}{\partial x} \hat{x} + \frac{\partial T}{\partial y} \hat{y} + \frac{\partial T}{\partial z} \hat{z}.$$

Then, $dT = \nabla T \cdot d\vec{\ell}$. Wait... what just happened? This means that if we move by a small vector $d\vec{\ell}$, then the change in T is given by taking the dot product of this small shift vector by the gradient of T at that point.

Let's dissect this a little more: we can expand the dot product to get

$$dT = \nabla T \cdot d\vec{\ell} = |\nabla T| |d\vec{\ell}| \cos \theta,$$

where θ is the angle between this new gradient vector we just defined and the small shift $d\vec{\ell}$ that we take. Suppose we *fix* the magnitude of $d\vec{\ell}$, so we're always stepping the same tiny amount. Then, if we search various directions by varying θ , notice the *maximum* change in T occurs when $\theta = 0$.

Focusing on this case, we extract the following intuition on what exactly the gradient is:

The gradient ∇T points in the direction of maximum increase of the function T . Moreover, its magnitude $|\nabla T|$ gives the rate of increase of T along this maximal direction.

If we point $d\vec{\ell}$ in the opposite direction of the gradient ($\theta = 180^\circ$), then the rate of change is precisely the negative, which makes sense. Also note if we point perpendicular to the gradient ($\theta = 90^\circ$), T doesn't change along this direction at all.

Here's a nice picture to keep in your mind: suppose you're on a large, bumpy mountain. We can characterize the surface of this mountain by the *two*-variable function $h(x, y)$ which gives the height of the mountain above any point (x, y) in the plane. Standing at any point (x, y) , you can look around and find the direction of steepest ascent—that is the *direction* of the gradient. If you calculate the slope of the mountain in that direction (rise over run)—that is the *magnitude* of the gradient.

As a person standing on the mountain, if you zoom in enough, you can disregard any large scale bumps in the mountain: it simply looks like a plane; this is what allows the dot product formula $dT = \nabla T \cdot d\vec{\ell}$ to work (with this picture in your mind, reason through the $\theta = 180^\circ$ and $\theta = 90^\circ$ cases again. This is analogous to the one-variable case, where taking the derivative assumes that when you zoom far in enough on f , you get something that looks like a straight line.

Definition 1.2.2. We call points where $\nabla T = \vec{0}$ **stationary points**.

These can be maximums (summits), minimums (valleys), saddle points (a pass), or “shoulders.”

Problems

Problem 1.13 Let \vec{z} be the separation vector from a fixed point (x', y', z') to the point (x, y, z) , and let z be its length. Show that

- (a) $\nabla(z^2) = 2\vec{z}$.
- (b) $\nabla(1/z) = -\vec{z}/z^2$.
- (c) What is the *general* formula for $\nabla(z^n)$?

Proof. We skip to part (c), as (a) and (b) are simply special cases $n = 2$ and $n = -1$, respectively. First, note that $\vec{z} = (x - x')\hat{x} + (y - y')\hat{y} + (z - z')\hat{z}$. Now,

$$\nabla(z^n) = \frac{\partial z^n}{\partial x}\hat{x} + \frac{\partial z^n}{\partial y}\hat{y} + \frac{\partial z^n}{\partial z}\hat{z} = \sum \frac{\partial z^n}{\partial x}\hat{x},$$

where the sum is over x, y, z . Now, note that

$$\frac{\partial z^n}{\partial x} = n z^{n-1} \cdot \frac{\partial z}{\partial x}.$$

Noting that $z = \sqrt{(x - x')^2 + (y - y')^2 + (z - z')^2}$, we have that

$$\frac{\partial z}{\partial x} = \frac{1}{2} \cdot \frac{1}{z} \cdot (2x - 2x') = \frac{x - x'}{z}.$$

Therefore,

$$\nabla(z^n) = n z^{n-1} \cdot \sum \frac{(x - x')\hat{x}}{z} = n z^{n-1} \cdot \frac{\vec{z}}{z} = \boxed{n z^{n-1} \hat{z}}.$$

□

§1.2.3 The Del Operator

Definition 1.2.3. We define the **del operator** as the vector operator

$$\nabla := \hat{x} \frac{\partial}{\partial x} + \hat{y} \frac{\partial}{\partial y} + \hat{z} \frac{\partial}{\partial z}.$$

The del operator ∇ is not really a vector; it's more of an instruction. When given a scalar function T to act upon, it produces the gradient ∇T .

Nonetheless, ∇ can mimic the behavior of an ordinary vector in virtually every way, if we translate “multiply” as “act upon.” It is great notational simplification, even if it's a bit of an abuse. Indeed, we can mimic all three types of vector products:

1. On a scalar function T , ∇ can act to give the gradient ∇T ;
2. On a vector function \vec{v} , via the dot product, to give the **divergence** $\nabla \cdot \vec{v}$;
3. On a vector function \vec{v} , via the cross product, to give the **curl** $\nabla \times \vec{v}$.

The following sections will analyze the divergence and curl.

§1.2.4 The Divergence

Definition 1.2.4. The **divergence** of a vector function \vec{v} is the scalar quantity

$$\nabla \cdot \vec{v} = \frac{\partial v_x}{\partial x} + \frac{\partial v_y}{\partial y} + \frac{\partial v_z}{\partial z}.$$

The name *divergence* is well-chosen: $\nabla \cdot \vec{v}$ is a measure of how much the vector field \vec{v} spreads out (i.e. diverges) from a given point.

Imagined a lake with streams and eddies where \vec{v} represented the velocity of the water at any given point. If you drop some particles into the lake at any given point, if they spread out, that's a point of positive divergence, while if they collect together, that's a point of negative divergence. The points of positive divergence are sources, while the points of negative divergence are sinks. (This analogy is not perfect, since water is generally incompressible, i.e. has zero divergence; however, you can imagine there are literal underwater faucets or drains adding or removing water from certain places.)

The divergence of a vector field \vec{v} measures how much the vectors “spread out” or “collect in” at any given point.

§1.2.5 The Curl

Definition 1.2.5. The **curl** of a vector function \vec{v} is the vector quantity

$$\begin{aligned}\nabla \times \vec{v} &= \det \begin{bmatrix} \hat{x} & \hat{y} & \hat{z} \\ \partial/\partial x & \partial/\partial y & \partial/\partial z \\ v_x & v_y & v_z \end{bmatrix} \\ &= \left(\frac{\partial v_z}{\partial y} - \frac{\partial v_y}{\partial z} \right) \hat{x} + \left(\frac{\partial v_x}{\partial z} - \frac{\partial v_z}{\partial x} \right) \hat{y} + \left(\frac{\partial v_y}{\partial x} - \frac{\partial v_x}{\partial y} \right) \hat{z}.\end{aligned}$$

The name *curl* is also well-chosen: $\nabla \times \vec{v}$ measures how much the vector field \vec{v} rotates (i.e. curls) around a given point. The direction of the vector $\nabla \times \vec{v}$ is given by the right hand rule: if you curl your fingers in the same direction as the vectors are curling, your thumb points in the direction of $\nabla \times \vec{v}$.

In the lake analogy, if you drop a toothpick, if it starts to rotate, then there is nonzero curl at that point. If it rotates counterclockwise, the curl points upward (positive z), while if it rotates clockwise, the curl points downward (negative z).

The curl of a vector field \vec{v} measure how much the vectors “swirl” around any given point, and also give the direction of this rotation by the right hand rule.

§1.2.6 Product Rules

The vector derivatives of sums and scalar products are rather simple; they’re just what you would expect.

Proposition 1.2.6 (Vector derivative sum rules)

We have that

- $\nabla(f + g) = \nabla f + \nabla g,$
- $\nabla \cdot (\vec{a} + \vec{b}) = (\nabla \cdot \vec{a}) + (\nabla \cdot \vec{b}),$
- $\nabla \times (\vec{a} + \vec{b}) = (\nabla \times \vec{a}) + (\nabla \times \vec{b}).$

Proposition 1.2.7 (Vector derivative scalar multiplication rules)

We have that

- $\nabla(kf) = k\nabla f,$
- $\nabla \cdot (k\vec{a}) = k(\nabla \cdot \vec{a}),$
- $\nabla \times (k\vec{a}) = k(\nabla \times \vec{a}).$

The product rules, however, are not so simple. There are two ways to construct a scalar function as a product of two functions: either by multiplying to scalar functions

to get fg or by taking the dot product of two vector functions to get $\vec{a} \cdot \vec{b}$. In addition, there are two ways to construct a vector function as a product of two functions: either by multiplying a scalar function by a vector function to get $f\vec{a}$, or by taking the cross product of two vector functions to get $\vec{a} \times \vec{b}$.

You can take the gradient of the two products that output a scalar function, as well as the divergence or curl of the two products that output a vector function, resulting in *six* product rules.

In general, product rules involving a scalar function are easy, and exactly what you expect (in the divergence/curl formulas, the ∇ on the scalar function turns into a gradient). The product rules involving dot and cross products of two vector functions are more complicated, however.

Proposition 1.2.8 (Gradient product rules)

We have that

- $\nabla(fg) = f\nabla g + g\nabla f$
- $\nabla(\vec{a} \cdot \vec{b}) = \vec{a} \times (\nabla \times \vec{b}) + \vec{b} \times (\nabla \times \vec{a}) + (\vec{a} \cdot \nabla)\vec{b} + (\vec{b} \cdot \nabla)\vec{a}$.

Proposition 1.2.9 (Divergence product rules)

We have that

- $\nabla \cdot (f\vec{a}) = f(\nabla \cdot \vec{a}) + \vec{a} \cdot (\nabla f)$,
- $\nabla \cdot (\vec{a} \times \vec{b}) = \vec{b} \cdot (\nabla \times \vec{a}) - \vec{a} \cdot (\nabla \times \vec{b})$.

Proposition 1.2.10 (Curl product rules)

We have that

- $\nabla \times (f\vec{a}) = f(\nabla \times \vec{a}) + \vec{a} \cdot (\nabla f)$,
- $\nabla \times (\vec{a} \times \vec{b}) = (\vec{b} \cdot \nabla)\vec{a} - (\vec{a} \cdot \nabla)\vec{b} + \vec{a}(\nabla \cdot \vec{b}) - \vec{b}(\nabla \cdot \vec{a})$.

§1.2.7 Second Derivatives

We have covered the three possible first derivatives using the del operator. By using the del operator twice, we can construct *five* species of second derivatives.

Since the gradient and curl are vector, we can take the divergence or curl of them. Since the divergence is a scalar, we can only take the gradient of it. Let's consider these possibilities:

Definition 1.2.11. We call the divergence of the gradient of a scalar function T the **Laplacian** of T , denoted $\nabla^2 T$. It is equal to

$$\begin{aligned} \nabla \cdot (\nabla T) &= \left(\hat{x} \frac{\partial}{\partial x} + \hat{y} \frac{\partial}{\partial y} + \hat{z} \frac{\partial}{\partial z} \right) \cdot \left(\hat{x} \frac{\partial T}{\partial x} + \hat{y} \frac{\partial T}{\partial y} + \hat{z} \frac{\partial T}{\partial z} \right) \\ &= \frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} + \frac{\partial^2 T}{\partial z^2}. \end{aligned}$$

Definition 1.2.12. We can also refer to the **Laplacian** of a vector function, which is simply defined as

$$\nabla^2 \vec{v} := (\nabla^2 v_x)\hat{x} + (\nabla^2 v_y)\hat{y} + (\nabla^2 v_z)\hat{z}.$$

The *curl* of the gradient, however, is always equal to zero:

$$\nabla \times (\nabla T) = \vec{0}.$$

A proof of this by components amounts to the fact that $\frac{\partial^2 T}{\partial x \partial y} = \frac{\partial^2 T}{\partial y \partial x}$.

Remark 1.2.13. There is a simple physical explanation for why the curl of the gradient always vanishes. Imagine T represents the height of some mountain; the gradient tells you which direction walks you upwards. If there was nonzero curl of the gradient, this would amount to the gradient vector going around in a closed loop; however, this would mean there is a closed loop where you're always travelling upwards, which is impossible! You can't end up in the same place but be at a higher height.

In essence, we're combining the fundamental theorems of gradients and curls, which will be introduced in the next section. By the fundamental theorem of curl (Stokes' Theorem), the surface integral of the curl of the gradient of T is equal to the line integral of the gradient of T around the boundary of that surface. Then, the fundamental theorem of gradient implies that this loop integral equals 0. In symbols,

$$\iint_S \nabla \times (\nabla T) \cdot d\vec{a} = \oint_{\partial S} \nabla T \cdot d\vec{\ell} = 0$$

since $\partial^2 S = 0$.

Like the curl of the gradient, the *divergence* of the curl is always equal to zero:

$$\nabla \cdot (\nabla \times \vec{v}) = 0.$$

This can also be easily checked through the definitions of divergence and curl.

Remark 1.2.14. There is also an explanation of why this is true by combining the fundamental theorems of divergences and curls (also introduced in the next section). By the fundamental theorem of divergence (Gauss' Theorem), the volume integral of the divergence of the curl of \vec{v} is equal to the surface integral of the curl of \vec{v} over the boundary of the volume. Then, the fundamental theorem of curl (Stokes' Theorem) implies that this closed surface integral equals 0. In symbols,

$$\iiint_V \nabla \cdot (\nabla \times \vec{v}) d\tau = \oiint_{\partial V} (\nabla \times \vec{v}) \cdot d\vec{a} = \oint_{\partial^2 V} \vec{v} \cdot d\vec{\ell} = 0$$

since $\partial^2 V = 0$.

Remark 1.2.15. Remarks 1.2.13 and 1.2.14 both rely on the fact that the boundary operator ∂ satisfies $\partial^2 = 0$: the first uses the fact that the boundary of the boundary of a surface (the boundary of a closed loop) is nothing, and the second uses the fact that the boundary of the boundary of a volume (the boundary of a closed surface) is nothing.

Definition 1.2.16. We simply refer to $\nabla(\nabla \cdot \vec{v})$ as the **gradient of the divergence** of \vec{v} . Note that this is *not* equal to the Laplacian.

This quantity seldom appears in physics, so we don't give it a special name. The curl of the curl is also nothing new:

Definition 1.2.17. We also simply refer to $\nabla \times (\nabla \times \vec{v})$ as the **curl of the curl** of \vec{v} .

It turns out that

$$\nabla \times (\nabla \times \vec{v}) = \nabla(\nabla \cdot \vec{v}) - \nabla^2 \vec{v}.$$

Often, this is actually used to *define* the Laplacian of a vector function $\nabla^2 \vec{v}$, since our previous definition relied on Cartesian coordinates $\hat{x}, \hat{y}, \hat{z}$, which is undesirable.

§1.3 Integral Calculus

§1.3.1 Line, Surface, and Volume Integrals

Definition 1.3.1. A **line integral** is an expression of the form

$$\int_{\vec{a}}^{\vec{b}} \vec{v} \cdot d\vec{\ell},$$

where \vec{v} is some vector function and $d\vec{\ell}$ is the infinitesimal displacement vector. These two vectors are dotted along every point on some prescribed path \mathcal{P} from point \vec{a} to point \vec{b} ; these values are summed by the integral to obtain the result.

If \mathcal{P} is a closed loop, i.e. $\vec{a} = \vec{b}$, then we instead write a loop integral

$$\oint \vec{v} \cdot d\vec{\ell}.$$

Usually, the value of the line integral depends on the path taken from \vec{a} to \vec{b} , but there is a certain class of functions where it actually isn't, for all \vec{a}, \vec{b} , and potential paths \mathcal{P} ! We *force* functions with this property **conservative**.

Definition 1.3.2. A **surface integral** is an expression of the form

$$\iint_{\mathcal{S}} \vec{v} \cdot d\vec{a},$$

where \vec{v} is some vector function and $d\vec{a}$ is an infinitesimal patch of area, with magnitude equal to the size of the patch and direction perpendicular to the surface \mathcal{S} that we're integrating over. While the choice of direction of $d\vec{a}$ is arbitrary for open surfaces, if the surface is closed (forming an inside and an outside), we will designate the outward direction as $d\vec{a}$ and write

$$\oiint_{\mathcal{S}} \vec{v} \cdot d\vec{a}$$

instead.

If \vec{v} describes the flow of a fluid, then $\iint_{\mathcal{S}} \vec{v} \cdot d\vec{a}$ represents the total mass per unit time passing through \mathcal{S} , also called the **flux**. This same term appears when talking about surface integrals of electric and magnetic fields.

As with conservative forces, there also exist a certain class of vector functions where the value of a surface integral is independent of the shape of the particular surface chosen, and instead is determined entirely by the boundary line $\partial\mathcal{S}$. Classifying these functions is also an important task.