



HENDRIX

COLLEGE

Multivariable Calculus Notes

MATH 230

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1.1 Parametric Equations

1.1.1 Introduction

Most of your calculus experience has been single variable, so that the functions under consideration were typically $f : \mathbb{R} \rightarrow \mathbb{R}$. Our course is divided into roughly 3 sections:

- Parametric Equations/Functions: Functions of the form $f : \mathbb{R} \rightarrow \mathbb{R}^n$ (Chapters 1 - 3)
- Scalar Functions: Functions of the form $f : \mathbb{R}^n \rightarrow \mathbb{R}$ (Chapters 4 - 5)
- Vector Fields: Functions of the form $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ (Chapter 6)

1.1.2 Parametric Equations

A *parametric equation* (or, *sometimes parametric function* or *vector-valued function*) is a function of the form $f : \mathbb{R} \rightarrow \mathbb{R}^n$. We will typically consider $n = 2$ or $n = 3$ and call the input variable the parameter, usually denoted by t . We write them as

$$f(t) = \begin{cases} x(t) \\ y(t) \end{cases} \quad \text{or} \quad f(t) = \begin{cases} x(t) \\ y(t) \\ z(t) \end{cases}.$$

A *parametric curve* is the set of points $(x(t), y(t))$ in \mathbb{R}^2 or $(x(t), y(t), z(t))$ in \mathbb{R}^3 traced out. Note that in general, the curve may not be a function for y in terms of x , but is a function of the parameter t .

1.1.3 Graphing Parametric Curves in the Second Dimension

Elimination of the Parameter

In some cases, we can explicitly solve for t in terms of one of x or y . When this is possible, you can write $y(x)$ or $x(y)$ and use your “regular” algebraic knowledge. We call this process *eliminating the parameter*.

Using Technology

- Your TI-84 can graph this if you switch to **par** mode.
- Likewise, GeoGebra can do this, using the **curve** function.
 - In general, the syntax is: `curve(x(t), y(t), t, min, max)`



1.1.4 The Cycloid

A wheel of radius a is rolling along a flat road at a constant velocity. The curve generated by a point along the edge of the wheel traces out a shape called a *cycloid*. Let t represent the angle - in radians!!!! - rotated through, and that the point of interest starts at the origin. Before we find the equations for the point, let's find the location of the center of the circle:

$$f_{\text{center}}(t) = \begin{cases} x(t) = at \\ y(t) = a \end{cases}$$

Then, relative to the center, our point along the edge has equations

$$f(t) = \begin{cases} x(t) = -a \sin t \\ y(t) = -a \cos t \end{cases}$$

Thus, our point has parametric equations

$$f(t) = \begin{cases} x(t) = a(t - \sin t) \\ y(t) = a(1 - \cos t) \end{cases}$$

1.1.5 Final Notes

Next time, we'll start asking Calculus-y questions: What are the velocities in the x , y , and total directions? What total distance does it travel? What is the area of the region under one period of the cycloid?

- The syllabus has a number of practice problems to work on. These are not required, and not to be turned in, but are for you to work before class next time.
- We will talk about them at the start of the next class. You should try them beforehand.
- The most common reason for a lack of success in this class is not spending time working problems on your own.



1.2 Calculus of Parametric Curves

For this section, we will have a parametric curve in \mathbb{R}^2 , defined by $f(t) = \begin{cases} x(t) \\ y(t) \end{cases}$. In many cases, the curve does not describe y as a function of x . However, we can still carry over many ideas from single variable calculus.

1.2.1 Slope for a Parametric Curve

Given a point t_0 , the *slope of the curve* in the xy -plane is given by

$$\left. \frac{dy}{dx} \right|_{t=t_0} = \left. \frac{dy/dt}{dx/dt} \right|_{t=t_0}.$$

Note that this is undefined when $x'(t_0) = 0$.

The *tangent line* at t_0 is given by

$$y = \left(\left. \frac{dy}{dx} \right|_{t=t_0} \right) (x - x(t_0)) + y(t_0).$$

1.2.2 Second Derivative

The value of the second derivative for the curve at t_0 is given by

$$\left. \frac{d^2y}{dx^2} \right|_{t=t_0} = \left(\frac{\frac{d}{dt}(dy/dx)}{dx/dt} \right) \Big|_{t=t_0}.$$

Note the benefit of Leibnitz notation for each of these two derivatives!

1.2.3 Area Under a Curve

Suppose that a parametric curve is non-self intersecting. Then, the signed area of the region between the curve and the x -axis on the t interval $[t_a, t_b]$ is given by

$$A = \int_{t_a}^{t_b} y(t) \frac{dx}{dt} dt.$$

1.2.4 Arc Length

The *arc length* of a parametric curve over the t interval $[t_a, t_b]$ is given by

$$s = \int_{t_a}^{t_b} \sqrt{\left(\frac{dx}{dt} \right)^2 + \left(\frac{dy}{dt} \right)^2} dt.$$



1.2.5 Surface Area

The *surface area* of the region obtained by rotating a non-self intersecting parametric curve is given by

$$S = \int_{t_a}^{t_b} 2\pi y(t) \sqrt{\left(\frac{dx}{dt}\right)^2 + \left(\frac{dy}{dt}\right)^2} dt.$$

1.2.6 The Cycloid

We can apply each of the above to the cycloid:

- *Derivative:* $\frac{dy}{dx} = \frac{dy}{dt} = \frac{\sin t}{1 - \cos t}$. Note that the slope is then independent of the radius of the wheel and that the slope is undefined at each of $t = \dots, -4\pi, -2\pi, 0, 2\pi, 4\pi, \dots$
- *Cartesian Equation:* With radius of 3 and when $t = \frac{\pi}{3}$, the point is found by solving for $x(\frac{\pi}{3})$ and $y(\frac{\pi}{3})$:

$$\begin{aligned} x\left(\frac{\pi}{3}\right) &= 3\left(\frac{\pi}{3} - \sin\left(\frac{\pi}{3}\right)\right) = \pi - \frac{3\sqrt{3}}{2} \\ y\left(\frac{\pi}{3}\right) &= 3\left(1 - \cos\left(\frac{\pi}{3}\right)\right) = \frac{3}{2} \\ (x, y) &= \left(\pi - \frac{3\sqrt{3}}{2}, \frac{3}{2}\right) \end{aligned}$$

Plugging in our t value into our derivative, we get a slope of

$$\frac{\sin(\pi/3)}{1 - \cos(\pi/3)} = \frac{\sqrt{3}/2}{1/2} = \sqrt{3}.$$

Now, we can write the equation of the tangent line as

$$y = \sqrt{3}\left(x - \pi + \frac{3\sqrt{3}}{2}\right) + \frac{3}{2}.$$



- *Concavity:* $\frac{d^2y}{dx^2} = \frac{d}{dt} \left(\frac{dy}{dx} \right) = \frac{d}{dt} \left(\frac{\sin t}{1 - \cos t} \right).$

$$\begin{aligned}
 \frac{d^2y}{dx^2} &= \frac{d/dt(dy/dx)}{dx/dt} \\
 &= \frac{\frac{d}{dt} \left(\frac{\sin t}{1 - \cos t} \right)}{a - a \cos t} \\
 &= \frac{\frac{\cos t(1 - \cos t) - \sin t \sin t}{(1 - \cos t)^2}}{a - a \cos t} \\
 &= \frac{\cos t - \cos^2 t - \sin^2 t}{(1 - \cos t)^2 a (1 - \cos t)} \\
 &= \frac{\cos t - 1}{a(1 - \cos t)^2} \\
 &= -\frac{1}{a(1 - \cos t)^2} \\
 &= -\frac{a}{a^2(1 - \cos t)^2} \\
 &= -\frac{a}{y^2}
 \end{aligned}$$

After some work, we find that $\frac{d^2y}{dx^2} = -\frac{a}{y^2}$, which shows that the cycloid is always concave down.

- *Area:* The area of one period of the cycloid $A = 3\pi a^2$, after some work:

$$\begin{aligned}
 A &= \int_{t_a}^{t_b} y(t)x'(t)dt \\
 &= \int_0^{2\pi} (a - a \cos t)(a - a \cos t)dt \\
 &= a^2 \int_0^{2\pi} (1 - 2 \cos t + \cos^2 t)dt \\
 &= a^2 \left(t + \frac{t}{2} + \frac{1}{4} \sin(2t) \right) \Big|_0^{2\pi} \\
 &= a^2 \left[\left(2\pi + \frac{2\pi}{2} + \frac{1}{4} \sin(2\pi) \right) - \left(0 + \frac{0}{2} + \frac{1}{4} \sin(0) \right) \right] \\
 &= a^2 [2\pi + \pi] \\
 &= 3\pi a^2.
 \end{aligned}$$



- *Arc Length*: The arc length of one period of the cycloid is $s = 8a$, again after some work:

$$\begin{aligned}
 s &= \int_{t_1}^{t_2} \sqrt{\left(\frac{dx}{dt}\right)^2 + \left(\frac{dy}{dt}\right)^2} dt \\
 &= \int_0^{2\pi} \sqrt{(a - a \cos t)^2 + (a \sin t)^2} dt \\
 &= a \int_0^{2\pi} \sqrt{1 - 2 \cos t + \cos^2 t + \sin^2 t} dt \\
 &= a \int_0^{2\pi} \sqrt{2 - 2 \cos t} dt \\
 &= \sqrt{2}a \int_0^{2\pi} \sqrt{1 - \cos t} dt \\
 &= \sqrt{2}a \int_0^{2\pi} \sqrt{2 \sin^2 \left(\frac{t}{2}\right)} dt \\
 &= \sqrt{2}a \cdot \sqrt{2} \int_0^{2\pi} \sin \left(\frac{t}{2}\right) dt \\
 &= 2a \left(-2 \cos \left(\frac{t}{2}\right) \right) \Big|_0^{2\pi} \\
 &= 8a.
 \end{aligned}$$

- *Surface Area*: The surface area of the solid obtained by rotating one period of the cycloid around the x -axis is $S = \frac{64\pi a^2}{3}$, after a lot of tedious work.

$$S = \int_0^{2\pi} 2\pi y(t) \sqrt{\left(\frac{dx}{dt}\right)^2 + \left(\frac{dy}{dt}\right)^2} dt$$

2.1 Vectors in the Plane

2.1.1 Notation

In print, we write vectors in bold like: \mathbf{v} , \mathbf{w} , \mathbf{u} , \dots . In handwriting, we often write vectors with an arrow over the top: \vec{v} , \vec{w} , \vec{u} , \dots .

2.1.2 Vectors

A *vector* is a quantity with both *magnitude* (size, length, strength, \dots) and *direction*. Given two points in the plane $P = (x_1, y_1)$ and $Q = (x_2, y_2)$, the vector from P to Q , denoted $\overrightarrow{PQ} = \mathbf{PQ} = \langle x_2 - x_1, y_2 - y_1 \rangle$.

We can also simply state components (known as *component form*): $\mathbf{v} = \langle x, y \rangle$.

In \mathbb{R}^2 the *standard unit vectors* are $\hat{i} = \mathbf{i} = \langle 1, 0 \rangle$ and $\hat{j} = \mathbf{j} = \langle 0, 1 \rangle$. This allows us to write $\mathbf{v} = \langle 2, 3 \rangle = 2\mathbf{i} + 3\mathbf{j}$, for example.

In \mathbb{R}^3 , we have three stand unit vectors, $\hat{i} = \mathbf{i} = \langle 1, 0, 0 \rangle$, $\hat{j} = \mathbf{j} = \langle 0, 1, 0 \rangle$, and $\hat{k} = \mathbf{k} = \langle 0, 0, 1 \rangle$.

It is a picky detail, but $\mathbf{i} \in \mathbb{R}^2 \neq \mathbf{i} \in \mathbb{R}^3$.

The *zero vector*, denoted $\mathbf{0}$, is $\mathbf{0} = \langle 0, 0 \rangle$. Note that $\mathbf{0} \neq 0$.

A *scalar* is a real number (or a magnitude), without direction.

If c is a scalar and $\mathbf{v} = \langle x, y \rangle$, then

$$c\mathbf{v} = c\langle x, y \rangle = \langle cx, cy \rangle.$$

This operation is called *scalar multiplication*. Scalar multiplication changes the magnitude of a vector, but not its direction.

Note that the individual components of a vector are themselves *scalars*. You need to keep track of which is which.

If $\mathbf{v} = \langle x_1, y_1 \rangle$ and $\mathbf{w} = \langle x_2, y_2 \rangle$, then the *vector sum*

$$\mathbf{v} + \mathbf{w} = \langle x_1 + x_2, y_1 + y_2 \rangle.$$



That is, we add component wise.

If $\mathbf{v} = \langle x_1, y_1 \rangle$, then the *magnitude* of \mathbf{v} is given by

$$\|\mathbf{v}\| = \sqrt{x_1^2 + y_1^2}.$$

This is really just the Pythagorean theorem.

2.2 Vectors in Space

In \mathbb{R}^3 , we have three axes, x , y , and z , which follow the *right-hand rule*: point the fingers of the right hand in the direction of the positive x -axis, curl them towards the positive y -axis, and the thumb points in the direction of the positive z -axis.

Since the distance formula in \mathbb{R}^3 is $d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2 + (z_2 - z_1)^2}$, then $\mathbf{u} = \langle x, y, z \rangle$ we have $\|\mathbf{u}\| = \sqrt{x^2 + y^2 + z^2}$.

To *normalize* a vector, we divide by its magnitude: $\mathbf{v} = \langle x, y, z \rangle$, then $\mathbf{u} = \frac{1}{\|\mathbf{v}\|} \mathbf{v} = \left\langle \frac{x}{\|\mathbf{v}\|}, \frac{y}{\|\mathbf{v}\|}, \frac{z}{\|\mathbf{v}\|} \right\rangle$. This gives us a *unit vector* in the direction of \mathbf{v} .

Thus, a *unit vector* is a vector \mathbf{u} such that $\|\mathbf{u}\| = 1$.

2.2.1 Vector Properties

Suppose that each of \mathbf{u} , \mathbf{v} , and \mathbf{w} are vectors and r and s are scalars. Then the following properties hold:

- *Additive Commutativity*: $\mathbf{v} + \mathbf{w} = \mathbf{w} + \mathbf{v}$.
- *Additive Associativity*: $\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$.
- *Additive Identity*: $\mathbf{v} + \mathbf{0} = \mathbf{v}$.
- *Additive Inverse*: $-\mathbf{v} = (-1)\mathbf{v}$ and $\mathbf{v} + (-\mathbf{v}) = \mathbf{0}$.
- *Scalar Associativity*: $r(s\mathbf{u}) = (rs)\mathbf{u}$.
- *Scalars Distributive over Vectors*: $r(\mathbf{u} + \mathbf{v}) = r\mathbf{u} + r\mathbf{v}$.
- *Vectors Distributive over Scalars*: $(r + s)\mathbf{u} = r\mathbf{u} + s\mathbf{u}$.
- *Multiplicative Identity*: $1\mathbf{u} = \mathbf{u}$.
- *Zero Scalar*: $0\mathbf{u} = \mathbf{0}$.



2.3 The Dot Product

Suppose $\mathbf{u} = \langle u_1, u_2, \dots, u_n \rangle$ and $\mathbf{v} = \langle v_1, v_2, \dots, v_n \rangle$ are vectors in \mathbb{R}^n . Then the *dot product* of \mathbf{u} and \mathbf{v} is given by

$$\mathbf{u} \cdot \mathbf{v} = u_1v_1 + u_2v_2 + \cdots + u_nv_n.$$

That is, we multiply the corresponding components and sum the results.

It should be clear that $\mathbf{u} \cdot \mathbf{v}$ results in a scalar. The dot product is a special type of inner product.

Think of the dot product as a way to measure how much of one vector points in the same direction as another.

2.3.1 Properties of the Dot Product

Let $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{R}^n$ and c be a scalar. Then the following properties hold:

- *Commutativity*: $\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$.
- *Distributive Property*: $\mathbf{u} \cdot (\mathbf{v} + \mathbf{w}) = \mathbf{u} \cdot \mathbf{v} + \mathbf{u} \cdot \mathbf{w}$.
- *Scalar Associativity*: $(c\mathbf{u}) \cdot \mathbf{v} = c(\mathbf{u} \cdot \mathbf{v}) = \mathbf{u} \cdot (c\mathbf{v})$.
- *Self-Product*: $\mathbf{u} \cdot \mathbf{u} = \|\mathbf{u}\|^2$.
- *Magnitude*: $\|\mathbf{v}\| = \sqrt{\mathbf{v} \cdot \mathbf{v}}$.
- *Angle*: $\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos(\theta)$, where $0 \leq \theta \leq \pi$ is the angle between \mathbf{u} and \mathbf{v} . (Law of Cosines.)
- *Orthogonality*: $\mathbf{u} \cdot \mathbf{v} = 0$ if and only if \mathbf{u} and \mathbf{v} are orthogonal.

2.3.2 Projections

The *projection* of \mathbf{u} onto \mathbf{v} is given by

$$\text{proj}_{\mathbf{v}} \mathbf{u} = \left(\frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{v}\|^2} \right) \mathbf{v}.$$

This is a vector parallel to \mathbf{v} , which has length equal to the amount of \mathbf{u} which points in the same direction as \mathbf{v} .

Think of a projection as a measure of how much of one vector points in the same direction as another.



2.3.3 Work

If a constant force \mathbf{F} moved an object from P to Q , the *work* done is given by

$$W = \mathbf{F} \cdot \overrightarrow{PQ}.$$

Thus, if that force acts at an angle θ to the line of motion, the work is:

$$W = (\|\mathbf{F}\|) \|\mathbf{PQ}\| \cos(\theta).$$

Later this semester, we will learn how to compensate for a non-constant force, and over a non-linear path.



2.4 The Cross Product

Suppose that $\mathbf{u}, \mathbf{v} \in \mathbb{R}^3$. Then, the *cross product* of \mathbf{u} and \mathbf{v} , denoted by $\mathbf{u} \times \mathbf{v}$, is the unique right-hand rule vector orthogonal to each of \mathbf{u} and \mathbf{v} whose magnitude is equal to the area of the parallelogram spanned by \mathbf{u} and \mathbf{v} .

Let $\mathbf{u} = \langle u_1, u_2, u_3 \rangle$ and $\mathbf{v} = \langle v_1, v_2, v_3 \rangle$. Then,

$$\mathbf{u} \times \mathbf{v} = \langle u_2v_3 - u_3v_2, u_3v_1 - u_1v_3, u_1v_2 - u_2v_1 \rangle.$$

NOTE: You will never multiply an v_1 -coordinate by an u_1 -coordinate. This is true for all v_n and u_n coordinates.

You can show by working the algebra that $\mathbf{u} \cdot (\mathbf{u} \times \mathbf{v}) = 0$ and $\mathbf{v} \cdot (\mathbf{u} \times \mathbf{v}) = 0$.

With determinants, you can do this in one step:

$$\mathbf{u} \times \mathbf{v} = \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ u_1 & u_2 & u_3 \\ v_1 & v_2 & v_3 \end{vmatrix}$$

Oddly, we can only define a cross-product in \mathbb{R} , \mathbb{R}^3 , and \mathbb{R}^7 , while the dot product is *always* defined.

Example

$$\begin{aligned} \mathbf{u} \times \mathbf{v} &= \langle 2, 1, 4 \rangle \cdot \langle 1, -3, 1 \rangle \\ &= \langle (1)(1) - 4(-3), 4(1) - 2(1), 2(-3) - 1(1) \rangle \\ &= \langle 13, 2, -7 \rangle. \end{aligned}$$

2.4.1 Properties of the Cross Product

Let $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{R}^3$ and c be a scalar. Then the following properties hold:

- *Anticommutativity*: $\mathbf{u} \times \mathbf{v} = -\mathbf{v} \times \mathbf{u}$.
- *Distributive Property*: $\mathbf{u} \times (\mathbf{v} + \mathbf{w}) = \mathbf{u} \times \mathbf{v} + \mathbf{u} \times \mathbf{w}$.
- *Scalar Associativity*: $(c\mathbf{u}) \times \mathbf{v} = c(\mathbf{u} \times \mathbf{v}) = \mathbf{u} \times (c\mathbf{v})$.
- *Zero*: $\mathbf{u} \times \mathbf{u} = \mathbf{0}$.
- *Nilpotence*: $\mathbf{u} \times \mathbf{v} = \mathbf{0}$ if and only if \mathbf{u} and \mathbf{v} are parallel.
- *Scalar Triple Product*: $\mathbf{u} \cdot (\mathbf{v} \times \mathbf{w}) = (\mathbf{u} \times \mathbf{v}) \cdot \mathbf{w}$.
- *Angle*: $\|\mathbf{u} \times \mathbf{v}\| = \|\mathbf{u}\| \|\mathbf{v}\| \sin(\theta)$, where $0 \leq \theta \leq \pi$ is the angle between \mathbf{u} and \mathbf{v} .



2.4.2 Standard Unit Vectors and the Cross Product

$$\mathbf{i} \times \mathbf{i} = \mathbf{j} \times \mathbf{j} = \mathbf{k} \times \mathbf{k} = \mathbf{0}.$$

- $\mathbf{i} \times \mathbf{j} = \mathbf{k}$
- $\mathbf{j} \times \mathbf{i} = -\mathbf{k}$
- $\mathbf{j} \times \mathbf{k} = \mathbf{i}$
- $\mathbf{k} \times \mathbf{j} = -\mathbf{i}$
- $\mathbf{k} \times \mathbf{i} = \mathbf{j}$
- $\mathbf{i} \times \mathbf{k} = -\mathbf{j}$

2.4.3 Torque

Torque, denoted by τ , measures the tendency to produce a rotation about an axis.

If \mathbf{r} is a radial vector from an axis to a force and \mathbf{F} is the force, then the torque induced on the axis by the force is given by:

$$\boldsymbol{\tau} = \mathbf{r} \times \mathbf{F} \quad \text{or} \quad \|\boldsymbol{\tau}\| = \|\mathbf{r}\| \|\mathbf{F}\| \sin(\theta).$$



2.5 Equations of Lines and Planes

For the vector equation, parametric equation, and the symmetric equation, use these points for the examples: $(3, 5, 1)$ and $(9, 1, 2)$.

2.5.1 Lines

Lines in Two Dimensions

A line in \mathbb{R}^2 which contains the point (x_0, y_0) and is parallel to the vector $\mathbf{v} = \langle a, b \rangle$ has parametric form

$$f(t) = \begin{cases} x(t) = x_0 + ta \\ y(t) = y_0 + tb \end{cases}.$$

Lines in Three Dimensions

In \mathbb{R}^3 , we have more options for the form of a line. Suppose that our line contains the point $\mathbf{r}_0 = \langle x_0, y_0, z_0 \rangle$ and is parallel to the vector $\mathbf{v} = \langle a, b, c \rangle$. Then the following properties hold:

Vector Equation

The *vector equation* of a line is given by $\mathbf{r}(t) = \mathbf{r}_0 + t\mathbf{v}$.

Example: Find all 3 equations of lines

From our example, $\mathbf{v} = \langle 6, -4, 1 \rangle$ and $\mathbf{r}_0 = \langle 3, 5, 1 \rangle$.

Vector equation: $\mathbf{r}(t) = \langle 3, 5, 1 \rangle + t \langle 6, -4, 1 \rangle$.

Parametric Equation

The *parametric equation* of a line is given by

$$f(t) = \begin{cases} x(t) = x_0 + ta \\ y(t) = y_0 + tb \\ z(t) = z_0 + tc \end{cases}.$$

From our example, we would get $x(t) = 3 + 6t$, $y(t) = 5 - 4t$, and $z(t) = 1 + t$.

Symmetric Equation

For the following formula, we get a , b and c from subtracting the x , y , and z components of the direction vector from the point vector.



As long as each of $a, b, c \neq 0$, the symmetric equation is

$$\frac{x - x_0}{a} = \frac{y - y_0}{b} = \frac{z - z_0}{c}.$$

(Notice that in two dimensions, this is just the equation of the line: $(\frac{b}{a})(x - x_0) + y_0 = y$, when solved for y .)

From our example, we would get

$$\frac{x - 3}{9 - 3} = \frac{y - 5}{1 - 5} = \frac{z - 1}{1 - 2} \implies \frac{x - 3}{6} = \frac{y - 5}{-4} = -z + 1.$$

Line Segment

Suppose that $P = (x_0, y_0, z_0)$ and $Q = (x_1, y_1, z_1)$. The line segment from P to Q is given by

$$\mathbf{r}(t) = (1 - t)\mathbf{p} + t\mathbf{q},$$

where $\mathbf{p} = \langle x_0, y_0, z_0 \rangle$, $\mathbf{q} = \langle x_1, y_1, z_1 \rangle$, and $0 \leq t \leq 1$.

The parametric equations for this segment are

$$f(t) = \begin{cases} x(t) = x_0 + t(x_1 - x_0) \\ y(t) = y_0 + t(y_1 - y_0) \\ z(t) = z_0 + t(z_1 - z_0) \end{cases}.$$

Distance Between Point and Line

The distance from a point M to a line which contains the point P and has direction vector \mathbf{v} is given by

$$d = \frac{\|\mathbf{PM} \times \mathbf{v}\|}{\|\mathbf{v}\|}.$$

Notice that you are free to choose any point on the line you'd like!

Relationships Between Lines

- *Equal*: Same direction vector, share a point.
- *Parallel*: Same direction vector, do not share a point.
- *Intersecting*: Different direction vectors, share a point.
- *Skew*: Different direction vectors, do not share a point.

2.5.2 Planes

A plane can be defined by:



- any three non-colinear points,
- any two intersection points,
- a line and a point not on the line, or
- given two orthogonal vectors with a common starting point: “spin” one vector in place; notice the other sweeps out a circle, which can be extended to a plane. * In notes *

Of particular importance for a plane is a *normal vector*. A vector \mathbf{n} is a normal vector provided it is orthogonal to \overrightarrow{PQ} for any two points P and Q which are in the plane.

2.5.3 Equations of a Plane

Like lines, we have three equations of a plane. Let P and Q be points in the plane and $n = \langle a, b, c \rangle$.

Vector Equation

The *vector equation* of a plane is $n \cdot \overrightarrow{PQ} = 0$. Note that this is an implicit definition (i.e. it is not useful for directly writing down an equation, but is the fundamental idea of why this all works)!

Scalar Equation

If (x_0, y_0, z_0) is any point in the plane, the *scalar equation* of the plane is given by

$$\begin{aligned}\langle x - x_0, y - y_0, z - z_0 \rangle \cdot \langle a, b, c \rangle &= 0 \\ a(x - x_0) + b(y - y_0) + c(z - z_0) &= 0\end{aligned}$$

General Form

The *general form* of the equation of a plane is given by $ax + by + cz + d = 0$, where $d = -ax_0 - by_0 - cz_0$.

Distance Between Point and Plane

- Equal: Share a common point, have parallel normal vectors
- Parallel: Do not share a common point, do have parallel normal vectors
- Intersecting: If their normal vectors are not parallel, the two planes intersect in a line.
 - You can use algebra to find a point in common – i.e. solve both equations for the planes
 - Find the line’s direction vector by taking the cross product of the planes’ normal vectors.



2.5.4 Examples

Use the points $P(3, 5, 1)$, $Q(9, 1, 2)$, and $R(0, 2, 5)$.

Example 1

Find the scalar equation of the plane containing P , Q , and R .

We know $\mathbf{PQ} = \langle 6, -4, 1 \rangle$ and $\mathbf{PR} = \langle -3, -3, 4 \rangle$. Then, $\mathbf{n} = \mathbf{PQ} \times \mathbf{PR} = \langle 13, -27, -30 \rangle$.

Thus, the equation of the plane is $13(x - 3) - 27(y - 5) - 30(z - 1) = 0$.

To check, plug in the points: $13(3) - 27(5) - 30(1) = 0$, $13(9) - 27(1) - 30(2) = 0$, and $13(0) - 27(2) - 30(5) = 0$.

Distance from point to line:

Suppose M is a point and P is any point on some line l . Refer to notes for graph. Let $d = \|\mathbf{PM}\| \sin \theta$. Let \mathbf{v} be a directional vector of l .

$$\|\mathbf{PM} \times \mathbf{v}\| = \|\mathbf{PM}\| \|\mathbf{v}\| \sin \theta. \text{ This gives us } d = \frac{\|\mathbf{PM} \times \mathbf{v}\|}{\|\mathbf{v}\|}.$$

Distance from point to plane:

Let P be the point with \mathbf{n} as the norm vector. We start with $d = \|\text{proj}_{\mathbf{n}} \mathbf{PQ}\|$. Then, $d = \frac{\|\mathbf{PQ} \times \mathbf{n}\|}{\|\mathbf{n}\| \cdot \|\mathbf{n}\|} \|\mathbf{n}\|$. Look in the book for the rest of this.



2.6 Quadratic Surfaces

2.6.1 Spheres

A *sphere*, centered at (x_0, y_0, z_0) with radius r , is given by the equation

$$(x - x_0)^2 + (y - y_0)^2 + (z - z_0)^2 = r^2.$$

2.6.2 Cylinder

A *cylinder* is a surface in \mathbb{R}^3 which consists of all lines that are parallel to a given line and pass through a given plane curve. The lines that make up a cylinder are called *rulings*. The *trace* of a cylinder is the cross section generated by intersecting the cylinder with a coordinate plane.

2.6.3 Quadratic Surfaces

A *quadratic surface* is a surface in \mathbb{R}^3 whose equation can be written as

$$Ax^2 + By^2 + Cz^2 + Dxy + Exz + Fyz + Gx + Hy + Iz + J = 0.$$

By a change of axes (rotations) and origin (translations), we can rewrite these always as one of

$$Ax^2 + By^2 + Cz^2 = 1, \quad \text{or} \quad Ax^2 + By^2 + Iz = 0$$

While cylinders have rulings made of parallel lines, quadratic surfaces do not (at least, not in general). However, their traces are always conic sections: lines, parabolas, circles, ellipses, or hyperbolas.

3.1 Vector Valued Functions and Space Curves

3.1.1 Vector-Valued Functions

Recall that a *function* f from a domain D to codomain E is a rule which assigns a single element of E to each element of D .

If each of $f_1, f_2, \dots, f_n: \mathbb{R} \rightarrow \mathbb{R}$ is a function we can then define the *vector-valued function* $\mathbf{r}: \mathbb{R} \rightarrow \mathbb{R}^n$ by

$$\mathbf{r}(t) = \langle f_1(t), f_2(t), \dots, f_n(t) \rangle$$

- When $n = 2$, we might write $\mathbf{r} = \langle f(t), g(t) \rangle = f(t)\hat{i} + g(t)\hat{j}$,
- and when $n = 3$, we might write $\mathbf{r} = \langle f(t), g(t), h(t) \rangle = f(t)\hat{i} + g(t)\hat{j} + h(t)\hat{k}$.

3.1.2 Curves

A *plane curve* is the set of points satisfying the parameterized curve $\mathbf{r}: \mathbb{R} \rightarrow \mathbb{R}^2$ over a given domain and for a given function \mathbf{r} .

A *space curve* is the set of points satisfying the parameterized curve $\mathbf{r}: \mathbb{R} \rightarrow \mathbb{R}^3$ over a given domain and for a given function \mathbf{r} .

We will call these *vector parameterizations* of the curve.

3.1.3 Limits

Formal Definition of a Limit

Suppose that $\mathbf{r}: \mathbb{R} \rightarrow \mathbb{R}^n$ is a vector-valued function and that $a \in \mathbb{R}$, though perhaps into in the domain of \mathbf{r} .

If there exists a vector $\mathbf{L} \in \mathbb{R}^n$ such that for each choice of $\epsilon > 0$ there exists $\delta > 0$ such that whenever $t \in \mathbb{R}$, $t \neq a$ and $|t - a| < \delta$, then

$$|\mathbf{r}(t) - \mathbf{L}| < \epsilon,$$

then we say that \mathbf{r} has *limit* \mathbf{L} as t approaches a , and we write

$$\lim_{t \rightarrow a} \mathbf{r}(t) = \mathbf{L}.$$



That is, if we want the output from \mathbf{r} to be close to \mathbf{L} , we can always choose inputs close to a to make that occur.

Limit Properties

Suppose that $f, g, h: \mathbb{R} \rightarrow \mathbb{R}$ are functions and that each of $\lim_{t \rightarrow a} f(t)$, $\lim_{t \rightarrow a} g(t)$, and $\lim_{t \rightarrow a} h(t)$ exist. Then, if $\mathbf{r}(t) = f(t)\mathbf{i} + g(t)\mathbf{j} + h(t)\mathbf{k}$, then $\lim_{t \rightarrow a} \mathbf{r}(t)$ exists as well, and has value

$$\lim_{t \rightarrow a} \mathbf{r}(t) = \left(\lim_{t \rightarrow a} f(t) \right) \mathbf{i} + \left(\lim_{t \rightarrow a} g(t) \right) \mathbf{j} + \left(\lim_{t \rightarrow a} h(t) \right) \mathbf{k}.$$

That is, we can essentially do limits component-wise.

Also, there is nothing special here about \mathbb{R}^3 . This works equally well for \mathbb{R}^n for any n .

3.1.4 Continuity

Suppose that $\mathbf{r}: \mathbb{R} \rightarrow \mathbb{R}^n$ is a vector-valued function and that $a \in \mathbb{R}$. The statement that \mathbf{r} is *continuous* at $t = a$ means that

- $\lim_{t \rightarrow a} \mathbf{r}(t)$ exists,
- $\mathbf{r}(a)$ is defined, and
- $\mathbf{r}(a) = \lim_{t \rightarrow a} \mathbf{r}(t)$.

This is of course essentially the same as our Calculus I definition of continuity, but now in \mathbb{R}^n .

Like that definition, informally, it means that the curve generated by the function is in “one piece.”



3.2 Calculus of Vector Valued Functions

Last time, we saw that limits and continuity for vector valued functions work component-wise. This is mostly the case for derivatives and integrals as well.

3.2.1 Derivatives

Suppose that $\mathbf{r}: \mathbb{R} \rightarrow \mathbb{R}^n$ is a vector valued function. We define the *derivative* of \mathbf{r} as

$$\mathbf{r}'(t) = \lim_{h \rightarrow 0} \frac{\mathbf{r}(t+h) - \mathbf{r}(t)}{h},$$

when this limit exists. We can of course also define the left- or right- hand derivatives as needed.

3.2.2 Properties of Derivatives

Suppose that $f, \mathbf{r}, \mathbf{u}$ are differentiable and c is a scalar. Then, we have the following properties:

- $\frac{d}{dt}(c\mathbf{r}) = c\frac{d}{dt}\mathbf{r}$
- $\frac{d}{dt}(\mathbf{r} \pm \mathbf{u}) = \frac{d}{dt}\mathbf{r} \pm \frac{d}{dt}\mathbf{u}$
- $\frac{d}{dt}(f(t)\mathbf{r}(t)) = f'(t)\mathbf{r}(t) + f(t)\mathbf{r}'(t)$
- $\frac{d}{dt}(\mathbf{r} \cdot \mathbf{u}) = \mathbf{r}' \cdot \mathbf{u} + \mathbf{r} \cdot \mathbf{u}'$
- $\frac{d}{dt}(\mathbf{r} \times \mathbf{u}) = \mathbf{r}' \times \mathbf{u} + \mathbf{r} \times \mathbf{u}'$
- $\frac{d}{dt}(\mathbf{r}(f(t))) = \mathbf{r}'(f(t))f'(t)$

From the fourth above, note that if $\mathbf{r} \cdot \mathbf{r}$ is constant, then \mathbf{r} is orthogonal to \mathbf{r}' .

3.2.3 Tangent Vectors

Suppose that $\mathbf{r}(t)$ is a differentiable vector-valued function and t_0 is the domain. This means that $\mathbf{r}(t_0)$ is tangent to the curve generated by $\mathbf{r}(t)$ at $t = t_0$. In particular, it is a tangent vector in the sense that if we interpret $\mathbf{r}(t)$ to refer to the position of some particle as a function of time, then $\mathbf{r}'(t_0)$ is the velocity vector, and $\|\mathbf{r}'(t_0)\|$ describes the speed of the particle.

Unit Tangent Vector

Of importance is the *principle unit tangent vector*, \mathbf{T} , which is defined by

$$\mathbf{T}(t) = \frac{\mathbf{r}'(t)}{\|\mathbf{r}'(t)\|}.$$



Note that this is only defined when $\mathbf{r}'(t) \neq \mathbf{0}$.

The unit tangent vector is simply a vector, of length 1, which points in the tangent direction of the curve. It is useful for describing the direction of motion of a particle along a curve.

3.2.4 Integrals

Like derivatives, we do integrals component-wise:

$$\begin{aligned}\int \mathbf{r}(t) dt &= \int \langle f_1(t), f_2(t), \dots, f_n(t) \rangle dt \\ &= \left\langle \int f_1(t) dt, \int f_2(t) dt, \dots, \int f_n(t) dt \right\rangle,\end{aligned}$$

where since these are indefinite, each produce a constant $C = \langle C_1, C_2, \dots, C_n \rangle$. Definite integrals can be defined in the same way.



3.3 Arc Length and Curvature

3.3.1 The Unit Normal Vector

For a curve C defined by \mathbf{r} in \mathbb{R}^3 , we have the *unit normal vector*, $\mathbf{N}(t)$, which is defined by

$$\mathbf{N}(t) = \frac{\mathbf{T}'(t)}{\|\mathbf{T}'(t)\|}.$$

The normal vector points orthogonally to the tangent vector; it points in the direction the curve is turning.

We show that \mathbf{T} and \mathbf{N} are orthogonal:

$$\begin{aligned} \mathbf{T} \cdot \mathbf{N} &= \mathbf{T}(t) \cdot \frac{\mathbf{T}'(t)}{\|\mathbf{T}'(t)\|} \\ &= \frac{1}{\|\mathbf{T}'(t)\|} \mathbf{T}(t) \cdot \mathbf{T}'(t) \\ &= \frac{1}{\|\mathbf{T}'(t)\|} \cdot 0 \\ &= 0. \end{aligned}$$

The last step is true since \mathbf{T} is a *unit* vector, and so that $\|\mathbf{T}\|$, and from our rules for differentiation, if a vector \mathbf{T} has a constant magnitude, then $\mathbf{T} \cdot \mathbf{T}' = 0$.

3.3.2 The Binormal Vector

The *binormal vector*, denoted by \mathbf{B} , is defined by

$$\mathbf{B}(t) = \mathbf{T}(t) \times \mathbf{N}(t).$$

Since \mathbf{T} and \mathbf{N} are each unit vectors which are orthogonal, by its definition, \mathbf{B} is also a unit vector which is orthogonal to both \mathbf{T} and \mathbf{N} .

You can think of \mathbf{T} and \mathbf{N} as defining the instantaneous plane of motion of a particle. Then, \mathbf{B} is the normal to that plane. If \mathbf{B} is constant, then the particle stays in a single plane, otherwise, $\frac{d\mathbf{B}}{dt}$ measures your “twisting” or torsion of motion.

3.3.3 Arc Length

If $\mathbf{r}(t)$ defines a smooth curve in \mathbb{R}^n , then the *arc length* of the curve from $t = a$ to $t = b$ is given by

$$\int_a^b \|\mathbf{r}'(t)\| dt.$$

That is, we are just adding up the “speed” of the curve to find its length (i.e., its distance).



Arc Length Parameterization

We can define the *arc length parameterization* of a curve C by:

- Define the arc length $s(t) = \int_a^t \|\mathbf{r}'(t)\| dt$.
- Solving, if possible, the resulting expression for t as a function of s .
- Rewriting $\mathbf{r}(t) = \mathbf{r}(t(s)) = \mathbf{r}$, so that the curve is written as a function of its length, from a given starting point.
- This is useful since there are many possible parameterizations of a given curve, but only a single arc length parameterization.

As you may recall from Calculus II, arc length can only be explicitly worked out for carefully selected problems. The same applies here.

3.3.4 Curvature

We define the *curvature*, denoted by κ , for a smooth curve given by $\mathbf{r}(s)$ as

$$\kappa = \left\| \frac{d\mathbf{T}}{ds} \right\|.$$

that is, how fast is the unit tangent vector changing, relative to the length of the curve itself. We use s because we want our answer to be independent of the parameterization.

However, though this formula is the easiest to reason with, it is in practice not useful in most cases. You would have to find the arc-length parameterization, and then find the unit tangent vector from that.

3.3.5 Calculating Curvature

- For all \mathbf{r} : $\kappa = \frac{\|\mathbf{T}'(t)\|}{\|\mathbf{r}'(t)\|}$.
- For \mathbb{R}^3 : $\kappa = \frac{\|\mathbf{r}'(t) \times \mathbf{r}''(t)\|}{\|\mathbf{r}'(t)\|^3}$.
- If $y = f(x)$: $\kappa = \frac{|y''(x)|}{[1+(y'(x)^2)]^{3/2}}$

3.3.6 Curvature of a Circle

We use each of the three equations to determine the curvature of a circle of radius a .



Parameterized Circle

A circle of radius a is parameterized by $\mathbf{r}(t) = a \cos(t)\mathbf{i} + a \sin(t)\mathbf{j}$. Thus, $\mathbf{r}'(5) = -a \sin(5)\mathbf{i} + a \cos(5)\mathbf{j}$, so that $\|\mathbf{r}'(t)\| = a$. So, using the definition above, we get:

$$\mathbf{T}(t) = \frac{\mathbf{r}'(t)}{\|\mathbf{r}'(t)\|} = \frac{-a \sin(t)\mathbf{i} + a \cos(t)\mathbf{j}}{a} = -\sin(t)\mathbf{i} + \cos(t)\mathbf{j}$$

and therefore

$$\mathbf{T}'(t) = -\cos(t)\mathbf{i} - \sin(t)\mathbf{j}$$

so that $\|\mathbf{T}(t)\| = 1$. Finally, we see that

$$\kappa = \frac{\|\mathbf{T}'(t)\|}{\|\mathbf{r}'(t)\|} = \frac{1}{a}.$$

Therefore, a circle has constant curvature, and in fact the curvature is just the reciprocal of the radius. This should make some intuitive sense, as the radius gets smaller, the curvature will increase and vice-versa. Notice also that the curvature of a line is 0 – in some sense, a line is a circle, but with infinite radius!

Curvature of a Circle in Three Dimensions

We can also use the second formula, again with a circle, noting that a circle of radius a is parameterized by $\mathbf{r} = a \cos(t)\mathbf{i} + a \sin(t)\mathbf{j} + 0\mathbf{k}$ so that $\mathbf{r}'(t) = -a \sin(t)\mathbf{i} + a \cos(t)\mathbf{j} + 0\mathbf{k}$ so that $\|\mathbf{r}'(t)\| = a$. Thus, $\mathbf{r}'' = -a \cos(t)\mathbf{i} - a \sin(t)\mathbf{j}$, so that

$$\mathbf{r}'(t) \times \mathbf{r}''(t) = \begin{vmatrix} \mathbf{i} & \mathbf{j} & \mathbf{k} \\ -a \sin(t) & a \cos(t) & 0 \\ -a \cos(t) & -a \sin(t) & 0 \end{vmatrix} = a^2 \mathbf{k}.$$

Then,

$$\kappa = \frac{\|\mathbf{r}'(t) \times \mathbf{r}''(t)\|}{\|\mathbf{r}'(t)\|^3} = \frac{a^2}{a^3} = \frac{1}{a}.$$



Curvature as a Function

We can also write (the top half) of a circle as $y(x) = \sqrt{a^2 - x^2}$. Then, $y' = \frac{-x}{\sqrt{a^2 - x^2}}$ and

$$\begin{aligned} y'' &= \frac{-\sqrt{a^2 - x^2} - \frac{1}{2}(a^2 - x^2)^{-1/2}(-2x)(-x)}{a^2 - x^2} \\ &= \frac{-\sqrt{a^2 - x^2} - x^2/\sqrt{a^2 - x^2}}{a^2 - x^2} \\ &= \frac{-(a^2 - x^2) - x^2}{(a^2 - x^2)^{3/2}} \\ &= \frac{-a^2}{(a^2 - x^2)^{3/2}}. \end{aligned}$$

Thus,

$$\kappa = \frac{|y''(x)|}{[1 + (y'(x)^2)]^{3/2}} = \frac{1}{a},$$

after some algebra.

Osculating Circle

In general, of course, curvature is not constant. At a particular point of interest, the curvature finds the reciprocal of the radius of the *osculating circle*. That is, the “tangent circle”—the circle that best matches the curve at the point of interest.



3.4 Motion in Space

3.4.1 Motion

Let $\mathbf{r}(t)$ represent the position of some object as a function of time in \mathbb{R}^3 . Then, as in Calculus I:

- Velocity vector: $\mathbf{v}(t) = \mathbf{r}'(t)$
- Speed: $\|\mathbf{v}(t)\| = \|\mathbf{r}'(t)\|$
- Acceleration vector: $\mathbf{a}(t) = \mathbf{v}'(t) = \mathbf{r}''(t)$

Note that $\mathbf{v}(t) \neq \mathbf{T}(t)$, though they do point in the same direction. More importantly, the relationship between $\mathbf{a}(t)$ and $\mathbf{N}(t)$ is complicated. $\mathbf{N}(t)$ is not acceleration, or even the direction of acceleration, but is the direction of “turning.”

Acceleration

In fact, $\mathbf{a}(t) = v'(t)\mathbf{T}(t) + (v(t))^2\kappa\mathbf{N}(t)$.

Proof. Notice that $\mathbf{v}(t) = v(t)\mathbf{T}(t)$. Then,

$$\begin{aligned}\mathbf{a}(t) &= v'(t)\mathbf{T}(t) + v(t)\mathbf{T}'(t) \\ &= v'(t)\mathbf{T}(t) + v(t)\|\mathbf{T}'(t)\|\mathbf{N}(t) \\ &= v'(t)\mathbf{T}(t) + v(t)\kappa\|\mathbf{r}'(t)\|\mathbf{N}(t) \\ &= v'(t)\mathbf{T}(t) + (v(t))^2\kappa\mathbf{N}(t).\end{aligned}$$

Since \mathbf{T} and \mathbf{T}' are orthogonal, this decomposes \mathbf{a} into two pieces – the first is the change in speed, and the second the change in direction. \square

3.4.2 Projectile Motion

The motion of an object – in 2-dimensions, typically, acted on only by gravity $\mathbf{F}_g = -mg\mathbf{j}$, where $g \approx 9.8 \text{ m/s}^2$ and m is the mass of the object. By Newton’s second law, $\mathbf{F} = m\mathbf{a}$, so we have

$$\mathbf{a}(t) = -g\mathbf{j}.$$

Thus, $\mathbf{v}(t) = -gt\mathbf{j} + \mathbf{v}_0$, where \mathbf{v}_0 is the initial velocity vector, and

$$\mathbf{s}(t) = -\frac{1}{2}gt^2\mathbf{j} + \mathbf{v}_0t + \mathbf{s}_0,$$

where \mathbf{s} is the position, and \mathbf{s}_0 is the initial position vector.

Often, we have an object starting at the origin (so $\mathbf{s}_0 = \mathbf{0}$) and fired at a velocity of v_0 at an angle θ above the horizon. Then,

$$\mathbf{s}(t) = v_0t \cos(\theta)\mathbf{i} + \left(v_0t \sin(\theta) - \frac{1}{2}gt^2\right)\mathbf{j}.$$



What is the horizontal distance travelled by the object before it hits the ground again?

Solving $(v_0 t \sin(\theta) - \frac{1}{2} g t^2) = 0$ gives $t_1 = 0, \frac{2v_0 \sin(\theta)}{g}$. We plug this time into the horizontal component and find

$$s(t_1) = \frac{v_0^2}{g} \sin(2\theta).$$

We note the unsurprising result that this is maximized when $\theta = \frac{\pi}{4}$.

3.4.3 Planetary Motion

There is a handout (on teams) that works through the mathematics of planetary motion—specifically:

- Starting from Newton's Law of Gravity and Newton's Second Law, we determine the position of a planet circling a (fixed) star.
- Describe the conditions under which the planet is bound to the star, or separately, flung off into space!

These are not exam questions, but are the sort of things that this material is used for in the “real world.”

4.1-4.3 Functions of Several Variables

A *scalar function* is a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$. This is in some sense a reverse of a parametric function. Parametric functions take in a single real variable t and output an answer of multiple variable (which we typically interpreted as a vector). A scalar function takes in multiple independent variables (typically interpreted as a point in \mathbb{R}^n) and outputs a single real variable.

Notation and Graphs

When $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, we can write $z = f(x, y)$. For each point (x, y) in the domain of f , we can interpret z as a height, so that the set of points (x, y, z) which satisfy $z = f(x, y)$ form a surface in \mathbb{R}^3 . You can graph this in GeoGebra using [geogebra.com/3d](https://www.geogebra.com/3d).

When $f: \mathbb{R}^3 \rightarrow \mathbb{R}$, we often write $w = f(x, y, z)$. Then, the point (x, y, z) is in the domain and w is still a height, though since the graph lives we are in \mathbb{R}^4 , we cannot draw it in a simple way.

Contour Plots

Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ be a function. The graph is a surface in \mathbb{R}^3 , but often it can be difficult to show this on paper. For a given $z_0 \in \mathbb{R}$ a *level curve* consists of all points $(x, y) \in \mathbb{R}^2$ for which $f(x, y) = z_0$. If we think of the function f as describing a height at a function of two-dimension position, a level curve is the set of all points with a particular given height.

A *contour plot* consists of a collection of level curves – typically, we select some number of equally spaced $\{z_1, z_2, \dots, z_n\}$ and build the level curves for each. This gives a “topographical map” of the function f .

4.1.1 Limits

Formal Definition of $\lim_{(x,y) \rightarrow (a,b)} f(x, y)$

Let $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, $(a, b) \in \mathbb{R}^2$, and $L \in \mathbb{R}$. The statement that f has limit L as (x, y) approaches (a, b) means that for each choice of $\epsilon > 0$, there exists $\delta > 0$ such that for all $(x, y) \neq (a, b)$ within a disk of radius δ , centered at (a, b) , then

$$|f(x, y) - L| < \epsilon.$$

Essentially, this says that the limit exists whenever we can make the output from f as close to L as we’d like by choosing (x, y) sufficiently close to (a, b) .



In Calculus I, you showed that a limit $\lim_{x \rightarrow a} f(x)$ exists if and only if the left and right hand limits both exist and match. Basically, you showed that no matter if you approach a from the left or from the right, you always get the same answer. The same idea applies here, but there is an important complication: We need to check every possible path into the point (a, b) – and rather than just two, there are infinitely many paths!

In general, there is not a simple recipe that will work. You can try various combinations of paths until something goes wrong. For practice set/exam purposes, I will always tell you that the limit does not exist, and ask you to prove it. Showing that a limit does exist is a completely different procedure – just because two paths happen to produce the same answer does not prove that all do.

Problem #4 on Practice Set #3 will ask you to show a limit exists. It provides step-by-step instructions and this is the only time I will ask you to do this.

Limit Laws

The basic rules are all the same. Suppose that $a, b, c \in \mathbb{R}$ and $f, g: \mathbb{R}^2 \rightarrow \mathbb{R}$ and that $\lim_{(x,y) \rightarrow (a,b)} f(x, y) = L$ and $\lim_{(x,y) \rightarrow (a,b)} g(x, y) = M$.

- $\lim_{(x,y) \rightarrow (a,b)} c = c$
- $\lim_{(x,y) \rightarrow (a,b)} [f(x, y) + g(x, y)] = L + M$
- $\lim_{(x,y) \rightarrow (a,b)} x = a$
- $\lim_{(x,y) \rightarrow (a,b)} y = b$
- $\lim_{(x,y) \rightarrow (a,b)} [cf(x, y)] = cL$
- if f has a single, simple algebraic definition and (a, b) is in the domain, then $\lim_{(x,y) \rightarrow (a,b)} f(x, y) = f(a, b)$.

4.1.2 Continuity

A function $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is continuous at $(a, b) \in \mathbb{R}^2$ provided that $\lim_{(x,y) \rightarrow (a,b)} f(x, y)$ exists and is equal in value to $f(a, b)$. This is exactly our usual definition of continuity from Calculus I again – and there is nothing special here about being in two dimensions.

All of the usual algebraic functions (polynomials, root functions, trig functions, exponentials, and logarithms) and their combinations by addition, subtraction, multiplication, division, exponentiation, and composition are continuous everywhere in their domains.

4.1.3 Partial Derivatives

Suppose $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a scalar function. Then, the *partial derivative* of f with respect to x_i is

$$\frac{\partial f}{\partial x_i} = \lim_{h \rightarrow 0} \frac{f(x_1, x_2, \dots, x_i + h, \dots, x_n) - f(x_1, x_2, \dots, x_i, \dots, x_n)}{h},$$



if this limit exists.

That is, we hold all other variables constant and take the derivative as though x_i is the only variable. Just like in Calculus I, the partial derivative is a rate of change – it measures the ratio of how much the output of f changes relative to a small change in x_i .

Notation

Suppose that f is a function and x is one of its variables. Then, we denote the partial derivative of f with respect to x by any of:

- Leibnitz/Condorcet Notation: $\frac{\partial f}{\partial x}$
- Jacobi Notation: f_x
- Euler Operation Notation: $D_x f$

Higher order derivatives are notated as:

- Leibnitz/Condorcet Notation: $\frac{\partial^n f}{\partial x^n}$
- Jacobi Notation: $f_{xxx\dots x}$
- Euler Operation Notation: $D_n f$

4.1.4 Mixed Partial Derivatives

Given a function $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ we can determine the *mixed partial derivative* of f with respect to x and then y . In Leibnitz notation, this would be

$$\frac{\partial}{\partial y} \left(\frac{\partial f}{\partial x} \right) \quad \text{or in Jacobi notation} \quad f_{xy}.$$

Clairaut's Theorem

Suppose that $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is defined on an open disk D containing the point (a, b) . If each of f_{xy} and f_{yx} are continuous on D , then $f_{xy}(a, b) = f_{yx}(a, b)$.

That is, for pretty much any function you'll even encounter, order does not matter for the mixed partial derivatives! This is especially nice, since notationally, $\frac{\partial^2 f}{\partial x \partial y} = f_{yx}$ and $\frac{\partial^2 f}{\partial y \partial x} = f_{xy}$, and who wants to worry about keeping up with that!



4.4 Tangent Planes

4.4.1 Tangent Lines

For a single-variable differentiable function $f: \mathbb{R} \rightarrow \mathbb{R}$, in Calculus I we defined the *tangent line* at $x = x_0$ as the unique line

$$y = f'(x_0)(x - x_0) + f(x_0).$$

This is the line which most closely matches the graph of $y = f(x)$ at the desired point. Functions which have tangent lines are said to be “smooth” or “locally linear.”

4.4.2 Tangent Planes

For functions $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ we have a similar idea. If the surface generated by such a function has no sharp corners or edges, you might see that as you zoom in, the surface becomes flatter and flatter – and will eventually resemble a plane. In fact, we define the *tangent plane* as the unique plane at $(x, y) = (x_0, y_0)$ which satisfies

$$z = f(x_0, y_0) + f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0).$$

But notice that $f(x_0, y_0)$ is just z_0 , the height of the surface at (x_0, y_0) . So, the tangent plane is the unique plane which most closely matches the graph of $z = f(x, y)$ at the desired point.

4.4.3 Linearizations

Recall from Calculus I that if f is differentiable at x_0 and x is close to x_0 then

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0).$$

This is the *linear approximation* of f at x_0 . In the same way, if $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is differentiable at (x_0, y_0) and (x, y) is near (x_0, y_0) , then

$$f(x, y) \approx f(x_0, y_0) + f_x(x_0, y_0)(x - x_0) + f_y(x_0, y_0)(y - y_0).$$

This is the *linearization* of f at (x_0, y_0) .

4.4.4 Examples

$f(x, y) = x \cos(\pi x) \sin(\pi y)$, $(x_0, y_0) = (\frac{1}{3}, \frac{1}{2})$. Our point of interest is $(\frac{1}{3}, \frac{1}{2}), \frac{1}{6}$, because we can just plug in the values of x_0 and y_0 into the function to get the height.

To find the equation of the tangent plane, we need to find the partial derivatives to fill out



the following equation:

$$z = \text{_____} + \text{_____} \left(x - \frac{1}{3} \right) + \text{_____} \left(y - \frac{1}{2} \right).$$

We found z_0 to be $\frac{1}{6}$, and the partial derivatives are

$$f_x(x, y) = \cos(\pi x) \sin(\pi y) - \pi x \sin(\pi x) \sin(\pi y),$$

and

$$f_x \left(\frac{1}{3}, \frac{1}{2} \right) = \frac{1}{2} - \frac{\pi}{3} \left(\frac{\sqrt{3}}{2} \right) (1) = \frac{1}{2} - \frac{\pi\sqrt{3}}{3}.$$

Similarly, we have

$$f_y(x, y) = \pi x \cos(\pi x) \cos(\pi y),$$

and

$$f_y \left(\frac{1}{3}, \frac{1}{2} \right) = 0.$$

Now, we can fill out the rest of our equation:

$$z = \frac{1}{6} + \left(\frac{1}{2} - \frac{\pi\sqrt{3}}{3} \right) \left(x - \frac{1}{3} \right) + 0 \left(y - \frac{1}{2} \right).$$



4.5 The Chain Rule

4.5.1 One Dimensional Chain Rule

Suppose that $f(x)$ and $g(x)$ are both single variable, differentiable functions. Then, if $h(x) = f(g(x))$, h is also differentiable and

$$h'(x) = f'(g(x))g'(x).$$

Alternatively, if we write $u = g(x)$, then

$$h'(x) = f'(u)g'(x) \quad \text{or} \quad \frac{dh}{dx} = \frac{df}{du} \frac{du}{dx}.$$

4.5.2 Higher Dimensional Chain Rules

The Chain Rule – One Independent Variable

Suppose that $x = f(t)$, $y = h(t)$, and $z = f(x, y)$ are all differentiable functions. Notice that z is “really” a function of the single independent variable t . Then,

$$\frac{dz}{dt} = \frac{\partial z}{\partial x} \cdot \frac{dx}{dt} + \frac{\partial z}{\partial y} \cdot \frac{dy}{dt}.$$

The Chain Rule – Two Independent Variables

Suppose that $x = g(u, v)$, $y = h(u, v)$, and $z = f(x, y)$. Then,

$$\frac{\partial z}{\partial u} = \frac{\partial z}{\partial x} \cdot \frac{\partial x}{\partial u} + \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial u},$$

and

$$\frac{\partial z}{\partial v} = \frac{\partial z}{\partial x} \cdot \frac{\partial x}{\partial v} + \frac{\partial z}{\partial y} \cdot \frac{\partial y}{\partial v}.$$

The Generalized Chain Rule

Let $w = f(x_1, x_2, \dots, x_m)$ be a function of m independent variables and each $x_i = x_i(t_1, t_2, \dots, t_n)$ has n independent variables. Then,

$$\frac{\partial w}{\partial t_j} = \frac{\partial w}{\partial x_1} \cdot \frac{\partial x_1}{\partial t_j} + \frac{\partial w}{\partial x_2} \cdot \frac{\partial x_2}{\partial t_j} + \cdots + \frac{\partial w}{\partial x_m} \cdot \frac{\partial x_m}{\partial t_j} = \sum_{i=1}^m \frac{\partial w}{\partial x_i} \cdot \frac{\partial x_i}{\partial t_j}.$$

4.5.3 Implicit Differentiation

Suppose we have an equation involving x and y . We can rewrite this as an equation of the form $z = f(x, y)$, where we let $z = 0$. Then,

$$z = 0, \text{ defines a curve for } x, y$$



4.6 Directional Derivatives and the Gradient

The function $z = f(x, y)$ defines a 2-dimensional surface in \mathbb{R}^3 . The values of $\frac{\partial z}{\partial x}$ and $\frac{\partial z}{\partial y}$ tell us about the rate of change of z relative to the two independent (and orthogonal) directions x and y . What if we want to know about the rate of change in some other direction?

Let v be a vector in the direction we are interested in. Then, the *directional derivative* at the point (x_1, x_2, \dots, x_n) of the function f in the direction of a vector v is given by

$$D_v = \frac{\partial f_{x_1}(x_1, x_2, \dots, x_n)}{\|v\|} v_1 + \frac{\partial f_{x_2}(x_1, x_2, \dots, x_n)}{\|v\|} v_2 + \dots + \frac{\partial f_{x_n}(x_1, x_2, \dots, x_n)}{\|v\|} v_n.$$

4.6.1 The Gradient, ∇f

Suppose that $f: \mathbb{R}^2 \rightarrow \mathbb{R}$. We define the *gradient* of f , denoted by $\nabla f(x, y)$, as

$$\nabla f(x, y) = \frac{\partial f}{\partial x}(x, y)\mathbf{i} + \frac{\partial f}{\partial y}(x, y)\mathbf{j}.$$

We can extend this: If $g: \mathbb{R}^3 \rightarrow \mathbb{R}$, then

$$\nabla g(x, y, z) = \frac{\partial g}{\partial x}(x, y, z)\mathbf{i} + \frac{\partial g}{\partial y}(x, y, z)\mathbf{j} + \frac{\partial g}{\partial z}(x, y, z)\mathbf{k},$$

and so on to higher dimensions.

The symbol ∇ is called “nabla” or “del.” Sometimes the gradient is written as $\text{grad } f$. The gradient is a vector which points in the direction of largest increase in the value of the function – if interpreted as height, it points in the direction that is the steepest uphill.

Note that the ∇ lives in the domain of the function. That is, if $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, then $\nabla f: \mathbb{R}^2 \rightarrow \mathbb{R}^2$. The gradient is a vector field – it assigns a vector to each point in the domain of the function.

The gradient on a contour plot is the vector which is orthogonal to the level curve at that point. This is because the gradient points in the direction of the greatest increase, and the level curve is the set of points where the function does not change.

Properties of the Gradient

- If $\nabla f(x_0, y_0) = 0$, then $D_u f(x_0, y_0) = 0$ for all u .
- If $\nabla f(x_0, y_0) \neq 0$, then $D_u f(x_0, y_0)$ is maximized when u and $\nabla f(x_0, y_0)$ point in the same direction.
- This maximum value is $\|\nabla f(x_0, y_0)\|$.



- The gradient vector is always orthogonal to any level curve.

Del – Vector Definition

We can think of ∇ as the vector

$$\nabla = \left\langle \frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z} \right\rangle.$$

From chapter 2, we know that when we multiply a vector (like ∇) by a scalar (like f), we simply multiply each coordinate of the vector by the scalar. Thus, we can think of ∇f as a “simple” vector operation. Later, when we have functions which have vector-values themselves, we will talk about the meaning of $\nabla \cdot \mathbf{F}$ and $\nabla \times \mathbf{F}$.

As a final note, we will start to blur a bit the distinction between \mathbb{R}^2 and \mathbb{R}^3 here. We can think of always working within \mathbb{R}^3 , and that \mathbb{R}^2 is the special case when the k component is 0.

4.6.2 Example

Let $f(x, y) = x^2 + y^2 - 2x - 6y + 14$. Find the direction derivative of f at $(4, 1)$, in the $\langle 1, 3 \rangle$ direction.

Solution. We first find the partial derivative of f with respect to x and y :

$$f_x(x, y) = 2x - 2 \quad \text{and} \quad f_y(x, y) = 2y - 6.$$

Then, at the point $(4, 1)$:

$$f_x(4, 1) = 6 \quad \text{and} \quad f_y(4, 1) = -4.$$

Then, we find the unit vector in the direction of $\langle 1, 3 \rangle$:

$$\frac{\langle 6, -4 \rangle \cdot \langle 1, 3 \rangle}{\sqrt{10}} = \frac{-6}{\sqrt{10}}.$$



4.7 Maxima and Minima

For this section, we will be considering the specific case $f: \mathbb{R}^2 \rightarrow \mathbb{R}$.

4.7.1 Critical Points

The statement that (x_0, y_0) is a *critical point* of f means that either:

- both $f_x(x_0, y_0) = 0$ and $f_y(x_0, y_0) = 0$, or
- one or both partials does not exist.

4.7.2 Maxima and Minima

Local Extrema

The function f has a *local maximum* at (x_0, y_0) provided that $f(x_0, y_0) \geq f(x, y)$ for all choices of (x, y) in some disk centered at (x_0, y_0) – that is, in some neighborhood of (x_0, y_0) .

In a similar manner, we can define a local minimum.

We will use the term *extremum* to refer to something that is either a maximum or minimum; the plural is *extrema*.

Note that if there is a local extrema, $\nabla f = \mathbf{0}$. This is because the gradient points in the direction of greatest increase, and if we are at a maximum or minimum, the function does not change.

Fermat's Theorem for Extrema: Suppose that $z = f(x, y)$ and that (x_0, y_0) is a local extremum. Then, (x_0, y_0) is also a critical point.

Second Derivative Test

Calculus I Version: In Calculus I, the sign of the second derivative tells you whether a critical point is a local max/min, or inconclusive: Suppose that $g: \mathbb{R} \rightarrow \mathbb{R}$ is a function of one variable, and x_0 is a critical point.

- if $g''(x_0) > 0$, then x_0 is a local minimum
- if $g''(x_0) < 0$, then x_0 is a local maximum
- if $g''(x_0) = 0$, this test is inconclusive – it could be a max, min, or neither.



Multivariable Calculus Version: We have a similar test for $f: \mathbb{R}^2 \rightarrow \mathbb{R}$, where (x_0, y_0) is a critical point. Define

$$D = f_{xx}(x_0, y_0)f_{yy}(x_0, y_0) - (f_{xy}(x_0, y_0))^2.$$

- if $D > 0$ and $f_{xx}(x_0, y_0) > 0$, then f has a local minimum
- if $D > 0$ and $f_{xx}(x_0, y_0) < 0$, then f has a local maximum
- if $D < 0$, then f has a saddle point
- if $D = 0$, then the test is inconclusive

Absolute Extrema

The function f has an *absolute maximum* at (x_0, y_0) provided that $f(x_0, y_0) \geq f(x, y)$ for all choices of (x, y) in the domain of f . Likewise, we can define an *absolute minimum*.

Extreme Value Theorem: Suppose that f is continuous on a closed, bounded domain D . Then f has both an absolute maximum and absolute minimum.

In Calculus I, we saw that a continuous function, defined on a closed interval, has its absolute extrema either at interior critical points or at the endpoints of the domain interval. This idea transfers over here – but is somewhat more complicated.

If $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is continuous on a closed and bounded domain, then by the Extreme Value Theorem, it has absolute extrema. These occur either at interior critical points of the domain or somewhere along the edge of the domain. The interior critical points are fairly straightforward. However, the edge of the domain is more complicated.

Extrema Finding Algorithm

- Find all critical points inside the region of interest by setting the two first partials equal to zero
- For each boundary, use the relationship between x and y to convert $f(x, y)$ to a new function, say $g(x)$, of a single variable
- This function will have an interval $[a, b]$ as its domain
- Work the Calculus I problem of finding absolute extrema for g on $[a, b]$. This will generate additional points of interests.
 - Absolute extrema occur at:
 1. Interior critical points.
 2. Along the boundary.
- Find the value of $f(x, y)$ for each point of interest – the largest value is the absolute maximum, the smallest is the absolute minimum.



4.7.3 Examples

Let $f(x, y) = x^2 + y^2 - 2x - 6y + 14$.

Solution. We know that the critical points are $(1, 3)$. We can use the second derivative test to determine that this is a local minimum.

$$f_{xx} = 2 \quad f_{yy} = 2 \quad f_{xy} = 0 \quad D = 4 > 0.$$

Let $f(x, y) = x^4 + y^4 - 4xy + 1$.

Solution. Find critical points:

$$f_x = 4x^3 - 4y \quad f_y = 4y^3 - 4x,$$

and set to 0:

$$x^3 = y \quad y^3 = x.$$

Substituting, we see:

$$\begin{aligned} x^9 - x &= 0 \\ x(x^8 - 1) &= 0 \\ x(x^4 + 1)(x^2 + 1)(x + 1)(x - 1) &= 0. \\ x &= 0, \pm 1 \\ \therefore y &= 0, \pm 1. \end{aligned}$$

Now we have our critical points $(0, 0)$, $(1, 1)$, $(-1, -1)$. We can use the second derivative test to determine the maximum or minimums. But, to use this formula, we need the second partials for the formula:

$$f_{xx} = 12x^2 \quad f_{yy} = 12y^2 \quad f_{xy} = -4.$$

For $(0, 0)$:

$$0 \cdot 0 - (4)^2 = -16 < 0.$$

Therefore, $(0, 0)$ is a saddle point. For $(1, 1)$:

$$12 \cdot 12 - 16 = 128 > 0.$$

Therefore, $(1, 1)$ is a local minimum. For the same reasons, $(-1, -1)$ is also a local minimum.



From Calculus I, remember we found the absolute extrema. Thus, let $g(x) = x^2 - 6x + 1$, on $[0, 4]$.

Solution. To find the absolute extrema of this function, get the critical points from the function, and also evaluate the function at the endpoints of the interval.

$$g'(x) = 2x - 6 = 0 \implies x = 3.$$

Then, $g(0) = 1$, $g(3) = -8$, and $g(4) = -7$. Thus, the absolute minimum is (-8) and the absolute maximum is 1 .

With that refresher, let's return to our $f(x, y)$ function and find the absolute extrema. Specifically, find absolute extreme on the triangle with vertices $(0, 0)$, $(0, 1)$, $(\frac{1}{2}, 1)$.

Solution.