

# Analysing Change Continuously

CMEE Maths Week

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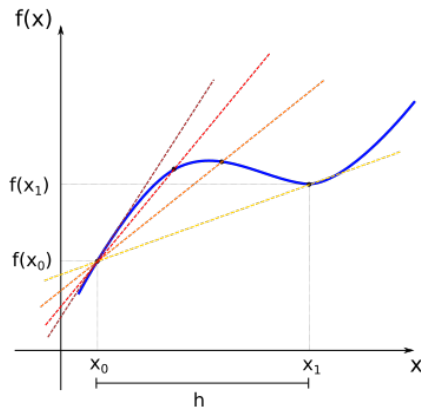
# The Average Rate of Change

## 1 The Derivative

Let's imagine a **continuous** function  $f(x)$ , and define its average rate of change between two points  $x_0$  and  $x_1$  as:

$$\frac{\Delta f}{\Delta x} = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

This quantity is the slope of the line that passes through  $(x_0, f(x_0))$  and  $(x_1, f(x_1))$ , known as the **secant line**.



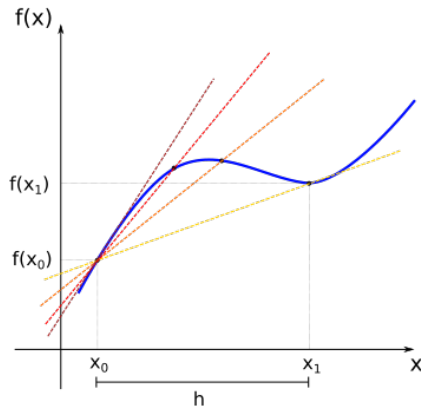
# Instantaneous Rate of Change

## 1 The Derivative

Now, what happens when  $x_1$  inches closer to  $x_0$ . How does the **secant line** change and what can be said about its slope?

$$\lim_{x_1 \rightarrow x_0} \frac{f(x_1) - f(x_0)}{x_1 - x_0} = L$$

As  $x_1 \rightarrow x_0$  the secant becomes the **tangent** of  $f(x)$  at  $x_0$ , and its slope,  $L$  is better known as  $f'(x_0)$ .



# The Derivative

## 1 The Derivative

$f'(x_0)$  is denominated the **derivative** of  $f(x)$  at  $x_0$ . The derivative can be more generally defined for any  $x$  in  $f$ 's domain as:

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h}$$

This is mostly true when  $f(x)$  is **continuous** at  $x$ , a condition we have yet to explain. However, continuous functions are not necessarily **differentiable**, which is something to keep in mind.

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# Continuity

## 2 Continuity

Formally, a function is **continuous** on an interval  $[x_0, x_1]$  if, for every  $a \in [x_0, x_1]$ :

$$\lim_{x \rightarrow a} f(x) = f(a)$$

That is to say, that the limit exists and is equal to the function evaluated at  $a$ . When this limit either does not exist or is different to  $f(a)$ , we say that  $a$  is a **discontinuity** of  $f$ .

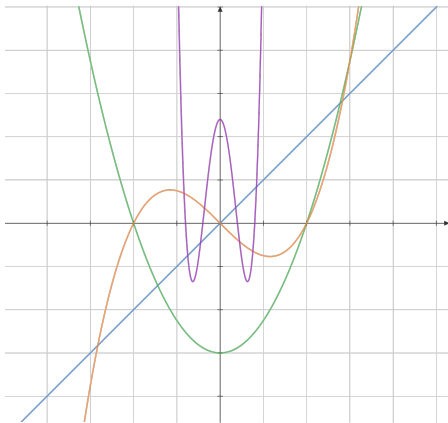
$$\lim_{x \rightarrow a} f(x) \neq f(a) \quad \text{or} \quad \lim_{x \rightarrow a} f(x) \rightarrow \pm\infty$$

Let's look at a few examples to illustrate what a continuous function looks like.

# Continuous Example: Polynomials

## 2 Continuity

All polynomials are continuous in  $\mathbb{R}$ , and as we'll later see, they are infinitely differentiable everywhere in their domain, to abbreviate we say they are class  $C^\infty$ .

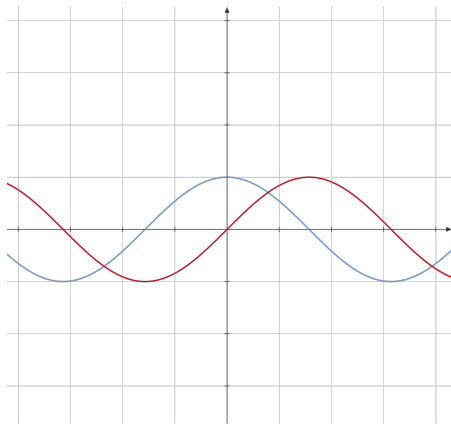




# Continuous Example: $\sin$ and $\cos$

## 2 Continuity

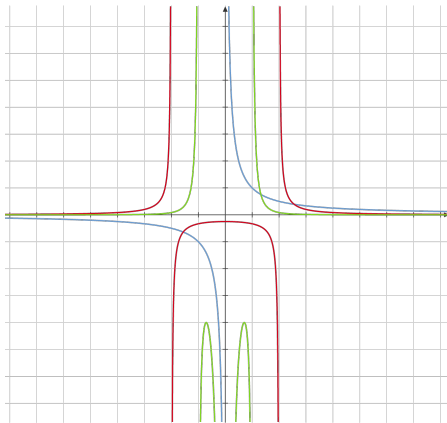
Both  $\sin(x)$  and  $\cos(x)$  are continuous in  $\mathbb{R}$  and class  $C^\infty$ . We'll see why this is the case later.



# Discontinuous Example: Rational Functions

## 2 Continuity

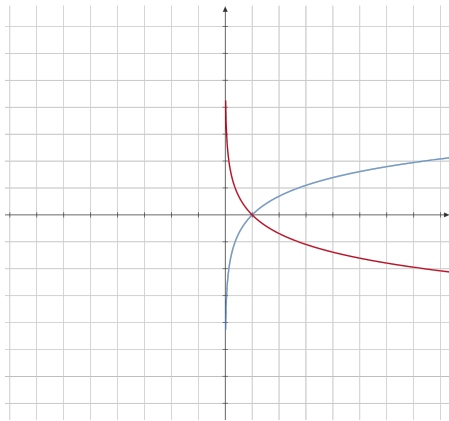
Rational functions are a good example of functions with discontinuities. For the most part they are continuous everywhere except at their vertical asymptotes.



# Discontinuous Example: The Logarithm

## 2 Continuity

Another function which is clearly not continuous is  $\log(x)$ , which diverges as  $x \rightarrow 0$ . Additionally, the logarithm isn't defined for  $x < 0$ .



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# Properties of Limits

## 3 Calculating Derivatives

While we won't go into the formal definition of limits, it is worth going through their properties before moving forward. Given two continuous functions  $f(x)$  and  $g(x)$ , it can be proven that, if  $\lim_{x \rightarrow a} f(x) = L$  and  $\lim_{x \rightarrow a} g(x) = M$ , the following statements are true:

- $\lim_{x \rightarrow a} [f(x) + g(x)] = \lim_{x \rightarrow a} f(x) + \lim_{x \rightarrow a} g(x) = L + M$
- $\lim_{x \rightarrow a} [Cf(x)] = C \lim_{x \rightarrow a} f(x) = CL$ , for any  $C \in \mathbb{R}$
- $\lim_{x \rightarrow a} [f(x)g(x)] = [\lim_{x \rightarrow a} f(x)] [\lim_{x \rightarrow a} g(x)] = LM$
- $\lim_{x \rightarrow a} \left[ \frac{f(x)}{g(x)} \right] = \frac{\lim_{x \rightarrow a} f(x)}{\lim_{x \rightarrow a} g(x)} = \frac{L}{M}$ , so long as  $M \neq 0$
- $\lim_{x \rightarrow a} g(f(x)) = g(\lim_{x \rightarrow a} f(x)) = g(L)$ , if  $g(x)$  is continuous at  $x = L$

Observe that the limit is a **linear operator**.

# Calculating Derivatives

## 3 Calculating Derivatives

We saw the definition of the derivative, now let's look at a simple example: the derivative of  $f(x) = x^2$ .

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} = \lim_{h \rightarrow 0} \frac{(x+h)^2 - x^2}{h} = \lim_{h \rightarrow 0} \frac{x^2 + 2xh + h^2 - x^2}{h}$$

By the properties seen previously, this can be decomposed into:

$$\lim_{h \rightarrow 0} \left( \frac{2xh}{h} + \frac{h^2}{h} \right) = \lim_{h \rightarrow 0} (2x) + \lim_{h \rightarrow 0} (h) = 2x + 0 = 2x$$

Thus  $f'(x) = 2x$ .

# Calculating Derivatives: Power Functions

## 3 Calculating Derivatives

We can generalize our previous example -  $f(x) = x^2$  - to any power function -  $f(x) = x^n$  - with  $n \neq 0$ :

$$f(x) = x^n \Rightarrow f'(x) = nx^{n-1}$$

This follows from the binomial expansion of  $(x + h)^n$ , though it requires a few more steps for rational  $n = \frac{p}{q}$ .

# The Derivative as a Linear Operator

## 3 Calculating Derivatives

From the properties of limits, their linearity in particular, we can make a few observations. Given two differentiable functions  $f(x)$  and  $g(x)$ :

- $\frac{d}{dx} [f(x) + g(x)] = \frac{df}{dx} + \frac{dg}{dx}$  - The derivative preserves function addition.
- For any  $C \in \mathbb{R}$  -  $\frac{d}{dx} [Cf(x)] = C \frac{df}{dx}$  - The derivative preserves scalar multiplication.

That is to say, the derivative is a **linear operator**.



# The Product Rule

## 3 Calculating Derivatives

Say that we have two differentiable functions  $f(x)$  and  $g(x)$ , what can we conclude about the derivative of their product?

$$\frac{d}{dx} [f(x)g(x)] = \lim_{h \rightarrow 0} \frac{f(x+h)g(x+h) - f(x)g(x)}{h}$$

Notice that we can add as many zeros to the above equation without changing its result.

$$\frac{d}{dx} [f(x)g(x)] = \lim_{h \rightarrow 0} \frac{f(x+h)g(x+h) - f(x)g(x) + [f(x+h)g(x) - f(x+h)g(x)]}{h}$$

# The Product Rule

## 3 Calculating Derivatives

Rearranging terms from the previous equation we find that:

$$\frac{d}{dx} [f(x)g(x)] = \lim_{h \rightarrow 0} \frac{f(x+h) [g(x+h) - g(x)] + g(x) [f(x+h) - f(x)]}{h}$$

From the properties of limits, we can are left with:

$$\frac{d}{dx} [f(x)g(x)] = \lim_{h \rightarrow 0} \frac{f(x+h) [g(x+h) - g(x)]}{h} + \lim_{h \rightarrow 0} \frac{g(x) [f(x+h) - f(x)]}{h}$$

Since  $f(x)$  is differentiable  $\lim_{h \rightarrow 0} f(x+h) = f(x)$  and we can safely factor it out from the limit.

# The Product Rule

## 3 Calculating Derivatives

Finally, from the last slide it should be obvious that:

$$\frac{d}{dx} [f(x)g(x)] = f(x)\frac{dg}{dx} + g(x)\frac{df}{dx}$$

This last relationship is called the **Product Rule**, and apart from being an important result in differential calculus, it is particularly useful for calculating tricky derivatives.

# The Chain Rule

## 3 Calculating Derivatives

Now that we know what the derivative of the product of functions looks like, we may ask what happens with the composition of functions. With a fair bit of work, it can be shown that, given two differentiable functions  $f(x)$  and  $g(x)$ :

$$\frac{d}{dx}(g \circ f)(x) = f'(x) [g' \circ f(x)] = \frac{dg}{df} \frac{df}{dx}$$

This is known as the **Chain Rule** and it can be generalized to however many compositions. Along with the product rule, the chain rule greatly simplifies more complex derivatives.

# The Quotient Rule

## 3 Calculating Derivatives

An application of both the product and the chain rule is the derivative of the quotient of two functions. Given two differentiable functions  $f(x)$  and  $g(x)$ , with  $g(x) \neq 0$ :

$$\frac{d}{dx} \left[ \frac{f(x)}{g(x)} \right] = \frac{f'(x)g(x) - g'(x)f(x)}{g^2(x)}$$

This is the **Quotient Rule** and while it is a direct consequence of the product and chain rules, it serves as a shortcut for differentiating rational functions.

# The Inverse Function Rule

## 3 Calculating Derivatives

Assuming we know the derivative of a function  $f(x)$ , we can use what we've learned so far to find the derivative of its inverse function. So given  $f(x)$  and its inverse  $g(x)$  such that  $g \circ f(x) = x$ , we can see that if  $f$  and  $g$  are both differentiable:

$$\frac{d}{dx}[g \circ f(x)] = f'(x)[g' \circ f(x)] = 1 \quad \Rightarrow \quad g' \circ f(x) = \frac{1}{f'(x)}$$

Making a substitution in both sides of the equation  $x = g(x)$  we find that the derivative of the inverse is given by:

$$g'(x) = \frac{1}{f'(g(x))} \quad \text{or} \quad \frac{d}{dx}f^{-1}(x) = \frac{1}{f' \circ f^{-1}(x)}$$

# The Derivative of the Exponential

## 3 Calculating Derivatives

The function  $f(x) = e^x$ , has an interesting derivative worth working through explicitly.

$$\frac{d}{dx}e^x = \lim_{h \rightarrow 0} \frac{e^{x+h} - e^x}{h} = \lim_{h \rightarrow 0} \frac{e^x e^h - e^x}{h} = \lim_{h \rightarrow 0} \frac{e^x(e^h - 1)}{h}$$

Since  $e^x$  doesn't depend on  $h$ , we can factor to out from the limit:

$$\frac{d}{dx}e^x = e^x \lim_{h \rightarrow 0} \frac{e^h - 1}{h}$$

Using the definition of  $e^h = \lim_{n \rightarrow \infty} (1 + \frac{h}{n})^n$  it can be proven that the above limit is equals to one:

$$\lim_{h \rightarrow 0} \frac{\lim_{n \rightarrow \infty} (1 + \frac{h}{n})^n - 1}{h} = 1 \quad \Rightarrow \quad \frac{d}{dx}e^x = e^x$$

# The Derivative of the Logarithm

## 3 Calculating Derivatives

Now that we've seen that  $\frac{d}{dx}e^x = e^x$ , we can use the inverse function rule to find the derivative of the logarithm:

$$\frac{d}{dx} \ln(x) = \frac{1}{e^{\ln(x)}} = \frac{1}{x}$$



# Elementary Derivatives

## 3 Calculating Derivatives

Using the rules and properties we've learned so far, we can find the derivatives of the elementary functions:

- $f(x) = a^x$  then  $f'(x) = \ln(a)a^x$ , for  $a > 0$ .
- $f(x) = \log_a(x)$  then  $f'(x) = \frac{1}{\ln(a)x}$ .
- $f(x) = \sin(x)$  then  $f'(x) = \cos(x)$ .
- $f(x) = \cos(x)$  then  $f'(x) = -\sin(x)$ .

These are only a handful of derivatives, but along with the product and chain rules, they pave the road to most of derivatives you'll find.

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# Higher Order Derivatives

## 4 Higher Order Derivatives

Another thing worth asking is whether you can differentiate a function more than once and, unsurprisingly, the answer depends on the function. In general, if the derivative of a function  $f'(x)$  is well behaved or **differentiable**, we can safely assume its derivative exists:

$$\frac{d}{dx}f'(x) = \frac{d^2f}{dx^2} = f''(x)$$

We call this the second derivative of  $f(x)$  with respect to  $x$ , and we say that  $f(x)$  is **at least** twice differentiable.

# Differentiability Classes

## 4 Higher Order Derivatives

Now, we can take the differentiation process a bit further and define the  $n^{th}$  derivative of  $f(x)$  with respect to  $x$ :

$$f^{(n)} = \frac{d^n f}{dx^n}$$

This is only valid as long as the lower order derivatives  $\left[f^{(0)}(x), f^{(1)}(x), \dots, f^{(n-1)}\right]$  are also differentiable. If a function can be differentiated up to  $n$  times, and  $f^{(n)}$  is continuous, we say it is of **differentiability class**  $C^n$ . If a function can be differentiated infinitely many times, we call it class  $C^\infty$  or **smooth**.

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# Multivariable Functions

## 5 Multivariable Functions

Let's consider a function  $z = f(x, y)$  and note that it defines a surface on the three-dimensional  $xyz$ -coordinate space. If  $f(x, y)$  is differentiable with respect to both  $x$  and  $y$ , we can define its derivatives with respect to each variable:

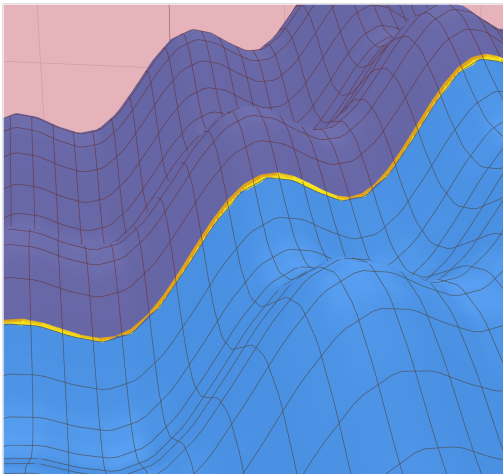
$$\frac{\partial}{\partial x}f(x, y) = \frac{\partial f}{\partial x} \quad \text{and} \quad \frac{\partial}{\partial y}f(x, y) = \frac{\partial f}{\partial y}$$

These are the **partial derivatives** of  $f(x, y)$  with respect to  $x$  and  $y$  respectively. Partial derivatives tell us how a function varies along either of the directions defined by  $x$  or  $y$ .

# A Geometric View

## 5 Multivariable Functions

An intuitive way of viewing partial derivatives is by imagining a cross section of the surface defined by  $z = f(x, y)$ . For example, the partial derivative  $\frac{\partial f}{\partial x}$  at  $y = 0$ , describes the rate of change of the curve at the intersection with  $y = 0$ .



# Calculating Partial Derivatives

## 5 Multivariable Functions

Calculating partial derivatives is, for the most part, identical to normal derivatives, you simply set the variable not under consideration as a constant. So for example, if -  
 $f(x, y) = x^2y$ :

$$\frac{\partial f}{\partial x} = 2xy \quad \text{and} \quad \frac{\partial f}{\partial y} = x^2$$

This is a very simple example, but the principle holds for arbitrarily complex  $f$ , so long as it is differentiable in each variable.



# Higher Order Partialials

## 5 Multivariable Functions

If you've been wondering whether higher order derivatives are possible for multivariate functions, the answer is yes, but it's a bit different. If  $f(x, y)$  has continuous and differentiable partial derivatives, the following derivatives exist:

$$\begin{aligned}\frac{\partial^2 f}{\partial x^2} &= f_{xx}(x, y) & \frac{\partial^2 f}{\partial y^2} &= f_{yy}(x, y) \\ \frac{\partial^2 f}{\partial x \partial y} &= \left( \frac{\partial f}{\partial x} \right) \left( \frac{\partial f}{\partial y} \right) = f_{xy}(x, y) & \frac{\partial^2 f}{\partial y \partial x} &= \left( \frac{\partial f}{\partial y} \right) \left( \frac{\partial f}{\partial x} \right) = f_{yx}(x, y)\end{aligned}$$

Moreover, since both first order partials are differentiable, it can be proven that  $f_{xy} = f_{yx}$ . This result is known as **Clairaut's Theorem**.

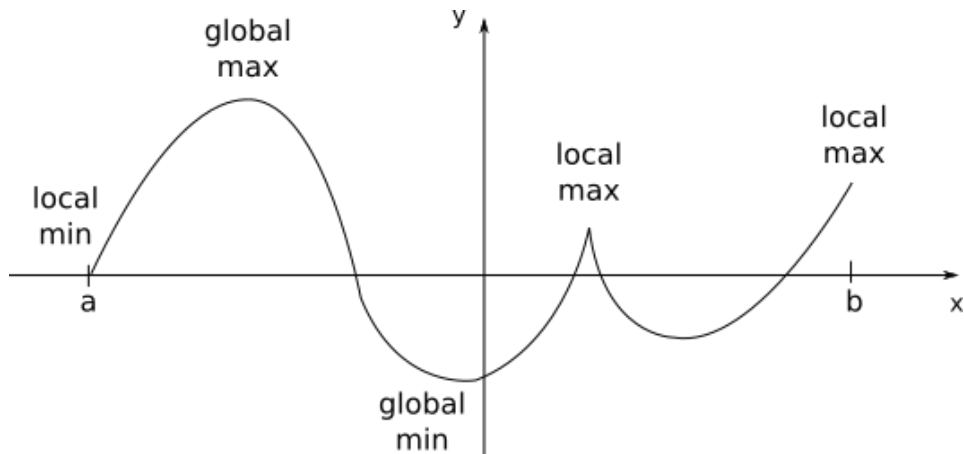
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# Finding the Optima

## 6 Optimization



# Classifying Extreme Values

## 6 Optimization

A key result in calculus is the **Extreme Value Theorem**: given a function  $f(x)$ , that is continuous on the interval  $x \in [a, b]$ , we can assure that  $f(x)$  attains a **maximum** and a **minimum** value at least once. In other words, there exist  $c$  and  $d$  in  $[a, b]$  such that:

$$f(d) \leq f(x) \leq f(c) \quad \forall x \in [a, b]$$

We call  $f(c)$  the **global maximum** of  $f(x)$  on the interval  $[a, b]$  and  $f(d)$  its **global minimum**.

# Classifying Extreme Values

## 6 Optimization

Now, it is clear that the **global extrema** will largely depend on the interval chosen. However, this allows us to develop a more general concept: **local extrema**. Assuming  $f(x)$  is continuous, we say that  $x_0$  is a **local maximum or minimum** if, for an arbitrarily small  $c$ :

$$f(x_0) \geq f(x) \quad \text{or} \quad f(x_0) \leq f(x), \quad \forall x \in (x_0 - c, x_0 + c)$$

So as we shrink the interval -  $(x_0 - c, x_0 + c)$  - around  $x_0$ ,  $f(x_0)$  remains the largest or smallest value on the interval.

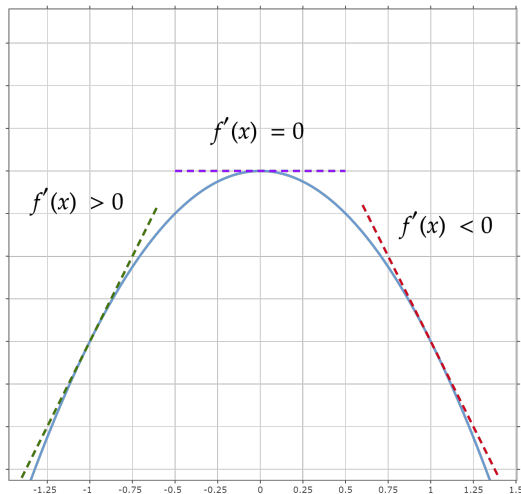
# Finding Extreme Values

## 6 Optimization

An important result of local extrema is that if  $x_0$  is a local extremum of  $f(x)$  and  $f(x)$  is differentiable at  $x_0$ , then

$$f'(x_0) = 0$$

we call  $x_0$  a **critical point** of  $f(x)$ .



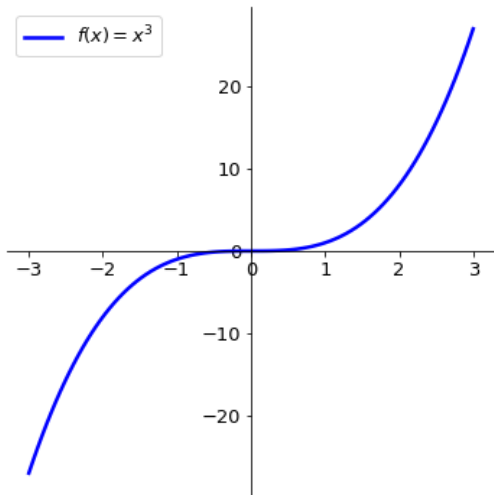
# Inflection Points

## 6 Optimization

Not all **critical points** are local extrema, they are only potential candidates.  
For example:  $f(x) = x^3$ .

$$f'(x) = 3x^2 \Rightarrow f'(0) = 0$$

$x_0 = 0$  is neither a local max nor a local min.  
Instead  $x_0$  is known as an **inflection point**.



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# Monotonicity

## 7 Monotonicity and Concavity

We'll step back a bit and imagine a function  $f(x)$  defined on the interval  $[a, b]$ . Given any two points  $x_1$  and  $x_2$  on the interval  $[a, b]$  such that  $x_2 > x_1$ , we can make the following definitions:

- If  $f(x_2) > f(x_1) \Rightarrow f(x)$  is **increasing** on the interval  $[a, b]$ .
- If  $f(x_2) < f(x_1) \Rightarrow f(x)$  is **decreasing** on the interval  $[a, b]$ .

Since we have a strict inequality, we can add that  $f(x)$  is **monotonous**. Otherwise if  $f(x_2) \geq f(x_1)$  or  $f(x_2) \leq f(x_1)$ , we would say  $f(x)$  was **non-decreasing** or **non-increasing**. Moreover this generalization of increasing/decreasing functions can be stated in terms of the derivatives.

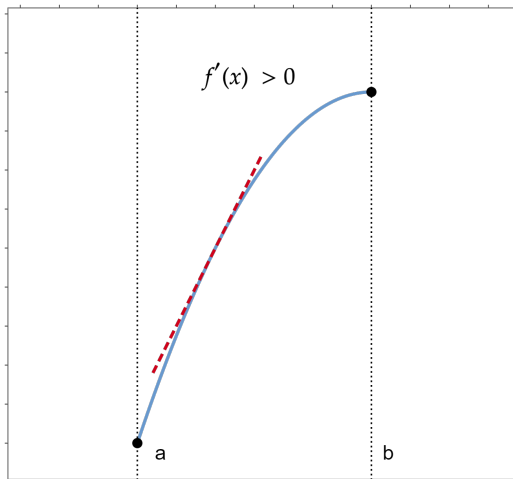
# The Derivative of Increasing Functions

## 7 Monotonicity and Concavity

Given function  $f(x)$  that is differentiable on the interval  $[a, b]$  if:

$$f'(x) > 0 \quad \forall x \in [a, b]$$

Then  $f(x)$  is **increasing** on the interval  $[a, b]$ .



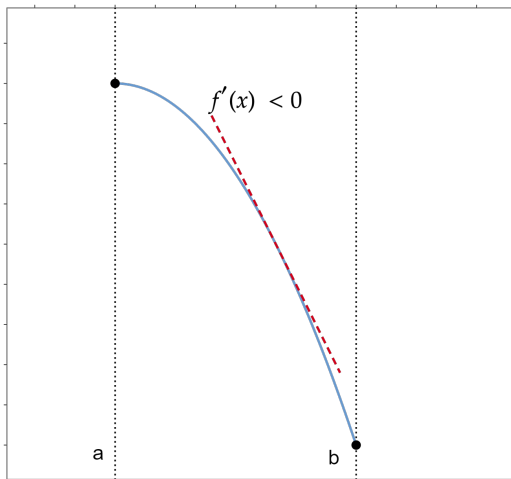
# The Derivative of Decreasing Functions

## 7 Monotonicity and Concavity

Given function  $f(x)$  that is differentiable on the interval  $[a, b]$  if:

$$f'(x) < 0 \quad \forall x \in [a, b]$$

Then  $f(x)$  is **decreasing** on the interval  $[a, b]$ .



# Concavity

## 7 Monotonicity and Concavity

Using derivatives, can generalize what we had seen on concavity to arbitrarily defined functions. Again, let's imagine  $f(x)$  differentiable on  $[a, b]$  and add that its derivative,  $f'(x)$ , is also differentiable on the same interval. Then, we find the following statements are true:

- If  $f'(x)$  is **increasing** for  $[a, b] \Rightarrow f(x)$  is **concave up**.
- If  $f'(x)$  is **decreasing** for  $[a, b] \Rightarrow f(x)$  is **concave down**.

This is another way of saying that the slope of the tangent lines of  $f(x)$  either increase or decrease with  $x$ .

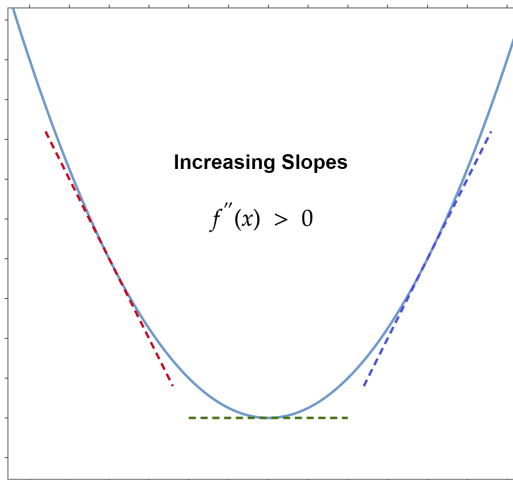
# Concave Up

## 7 Monotonicity and Concavity

Given function  $f(x)$  that is twice differentiable on the interval  $[a, b]$  if:

$$f''(x) > 0 \quad \forall x \in [a, b]$$

Then  $f(x)$  is **concave up** on the interval  $[a, b]$ .



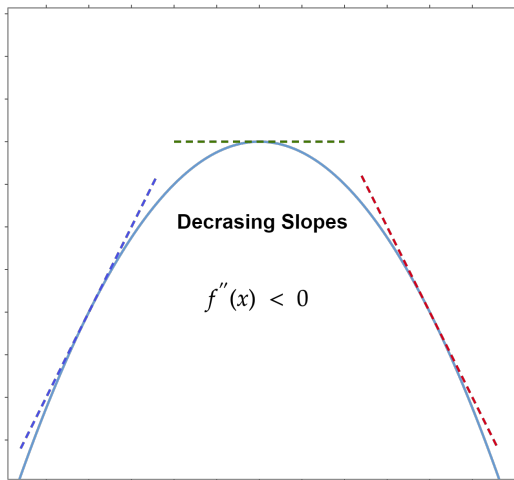
# Concave Down

## 7 Monotonicity and Concavity

Given function  $f(x)$  that is twice differentiable on the interval  $[a, b]$  if:

$$f''(x) < 0 \quad \forall x \in [a, b]$$

Then  $f(x)$  is **concave down** on the interval  $[a, b]$ .



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# Finding Extrema

## 8 Finding Extrema

These last few results provide a simple framework for finding local maxima and minima of a function. Given a twice differentiable function  $f(x)$ , defined on the interval  $[a, b]$ :

1. Find all critical points of  $f(x)$  -  $c \in [a, b]$  such that  $f'(c) = 0$ .
2. Compute  $f''(x)$  for each critical point  $c$ .
3. Classify  $c$  according to local concavity.

The last step requires a bit of intuition on what concavity implies for  $c$ .



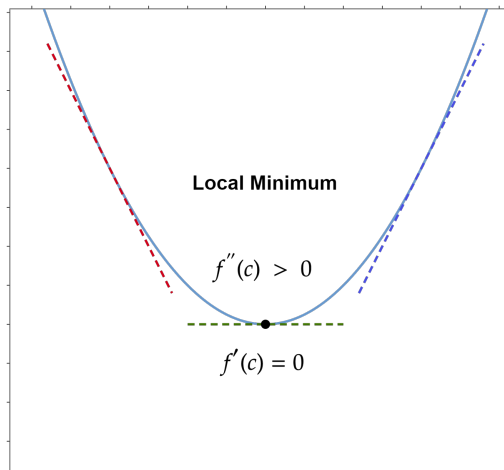
# Concavity of Local Minima

## 8 Finding Extrema

Given function  $f(x)$  that is twice differentiable on the interval  $[a, b]$ , if for  $c \in [a, b]$ :

$$f'(c) = 0 \quad \text{and} \quad f''(c) > 0$$

Then  $c$  is a **local minimum** of  $f(x)$ .



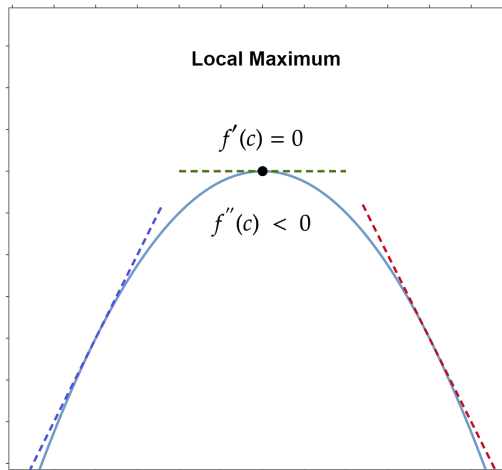
# Concavity of Local Maxima

## 8 Finding Extrema

Given function  $f(x)$  that is twice differentiable on the interval  $[a, b]$ , if for  $c \in [a, b]$ :

$$f'(c) = 0 \quad \text{and} \quad f''(c) < 0$$

Then  $c$  is a **local maximum** of  $f(x)$ .



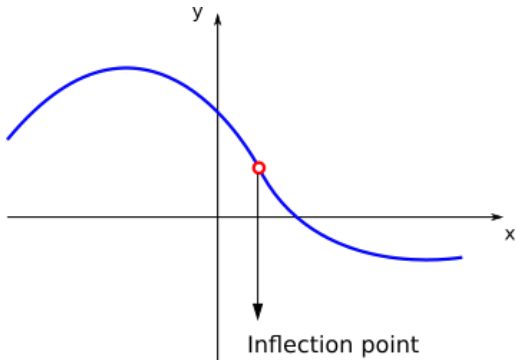
# Concavity of Inflection Points

## 8 Finding Extrema

As you may have already noted, there exists a third class of critical point. This pertains to the case:

$$f'(c) = 0 \quad \text{and} \quad f''(c) = 0$$

Here  $c$  is an **inflection point** of  $f(x)$ . We further observe that inflection points indicate where functions **change concavity**.



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# Series Expansions

## 9 Taylor Series

Now, let's take another look at the definition of the derivative. Given a differentiable function  $f(x)$ :

$$f'(a) = \lim_{x \rightarrow a} \frac{f(a) - f(x)}{a - x}$$

If  $x$  is in a small enough neighborhood around  $a$ , we can make the following approximation.

$$f'(a) \approx \frac{f(a) - f(x)}{a - x} \quad \Rightarrow \quad f(x) \approx f(a) + f'(a)(x - a)$$

So around a very small neighborhood of  $a$ , we can approximate  $f(x)$  as a **linear function**

# The Taylor Theorem

## 9 Taylor Series

Assuming  $f(x)$  can be differentiated  $n$  times and using the **mean value theorem**, it can be proven that:

$$f(x) \approx f(a) + f'(a)(x - a) + \frac{1}{2}f''(a)(x - a)^2 + \cdots + \frac{1}{n!}f^{(n)}(a)(x - a)^n$$

The above statement is known as **Taylor's Theorem**. Taylor's Theorem implies that any arbitrary  $C^n$  function can be approximated as a polynomial of its derivatives around a point  $a$ , this approximation is called the **Taylor polynomial** of  $f(x)$ .

# Building the Taylor Polynomial

## 9 Taylor Series

Although we won't be proving Taylor's theorem, we'll look at the motivation behind it. Assume you have an  $n$ -differentiable function  $f(x)$  that can be approximated around a point  $a$  by an  $n^{\text{th}}$  degree polynomial  $P(x)$ :

$$P(x) = a_0 + a_1(x - a) + a_2(x - a)^2 + \cdots + a_n(x - a)^n$$

Since we want this to be the best approximation possible, we expect the derivatives of  $f(x)$  and  $P(x)$ , and the functions themselves to be the same at  $x = a$ :

$$f(a) = P(a)$$

$$f'(a) = P'(a)$$

$$\vdots$$

$$f^{(n)}(a) = P^{(n)}(a)$$

# Building the Taylor Polynomial

## 9 Taylor Series

Differentiating up to the  $n^{\text{th}}$  order derivative of  $P(x)$ , we find that:

$$P(a) = f(a) = a_0$$

$$P'(a) = f'(a) = a_1$$

$$P''(a) = f''(a) = 2 \cdot 1 \cdot a_2$$

$$\vdots$$

$$P^{(n)}(a) = f^{(n)}(a) = n \cdot (n - 1) \cdots 2 \cdot 1 \cdot a_n$$

Defining  $n! = n \cdot (n - 1) \cdots 2 \cdot 1$  as the **factorial** of  $n$ , we can solve for the coefficients  $a_i$ .



# Building the Taylor Polynomial

## 9 Taylor Series

Solving for the coefficients in our last set of equations, we have:

$$f^{(n)}(a) = n!a_n \quad \Rightarrow \quad a_n = \frac{f^{(n)}(a)}{n!}$$

Hence the polynomial  $P(x)$  is defined by:

$$P(x) = f(a) + f'(a)(x - a) + \frac{1}{2}f''(a)(x - a)^2 + \cdots + \frac{1}{n!}f^{(n)}(a)(x - a)^n$$

$P(x)$  is the  $n^{\text{th}}$  degree **Taylor Polynomial** of  $f(x)$  defined at the point  $x = a$ .

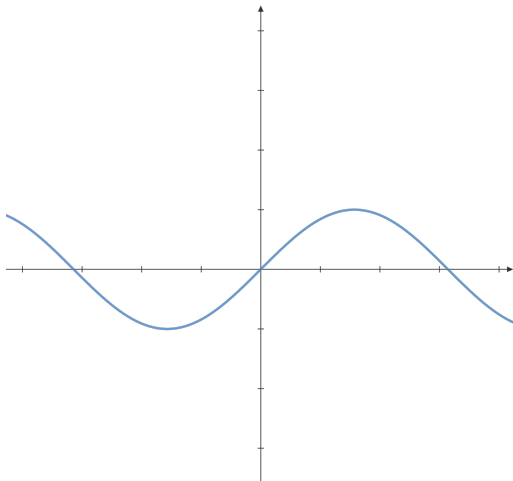
## Example: Taylor Expansion of $\sin(x)$

### 9 Taylor Series

So let's go through an example of a Taylor expansion -  $f(x) = \sin(x)$  - around the point  $x = 0$ .

$$\sin(x) = P(x)$$

First let's note that all of the even order terms  $n = 2k$  of  $P(x)$  are going to be zero since  $\sin(0) = 0$ .



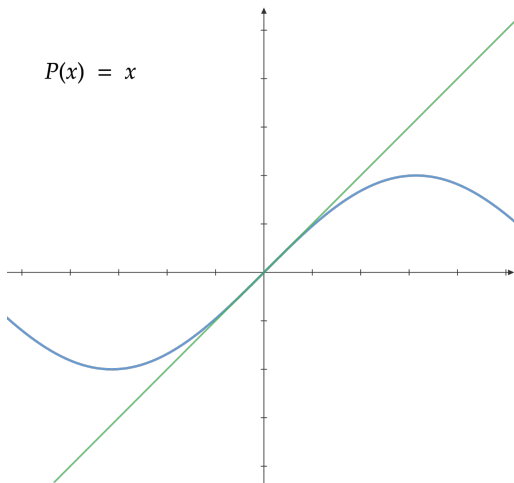
## Example: Taylor Expansion of $\sin(x)$

### 9 Taylor Series

We can jump straight into the 1<sup>st</sup> order expansion.

$$\sin(x) = \sin(0) + \cos(0)(x)^1 = x$$

Turns out  $\sin(x) \approx x$  is a decent approximation for  $x \ll 1$ , this is usually referred to the small angle approximation.



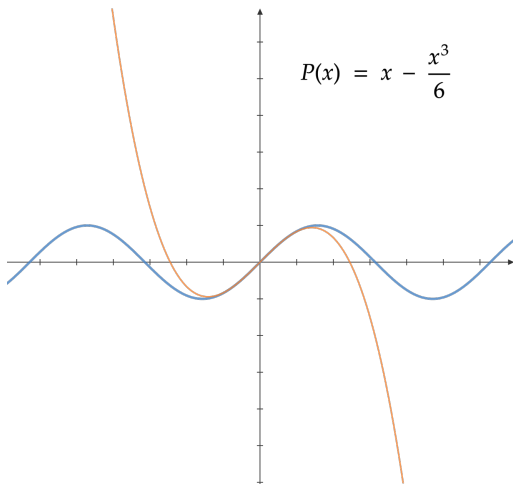
## Example: Taylor Expansion of $\sin(x)$

### 9 Taylor Series

Since even order terms don't change the polynomial, we can skip to the third order polynomial.

$$\sin(x) = x - \frac{\cos(0)}{3!}(x)^3 = x - \frac{x^3}{3!}$$

Here the negative sign appears from  $\frac{d}{dx} \cos(x) = -\sin(x)$ . This will create an alternating  $\pm$  pattern.



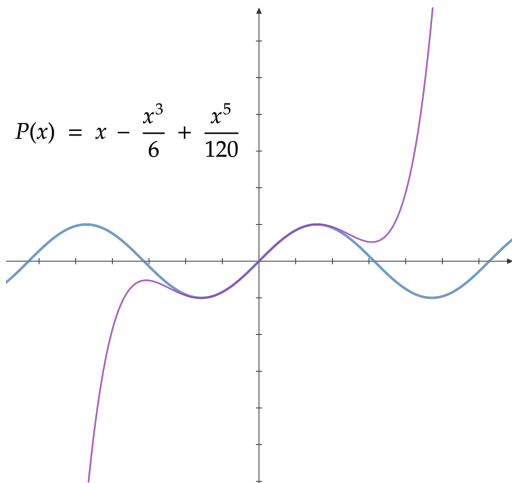
## Example: Taylor Expansion of $\sin(x)$

### 9 Taylor Series

We go on and add the fifth order term:

$$\sin(x) = x - \frac{x^3}{3!} + \frac{x^5}{5!}$$

Notice how adding terms improves the approximation.



## Example: Taylor Expansion of $\sin(x)$

### 9 Taylor Series

And since  $\sin(x)$  is  $C^\infty$  we can build an arbitrarily large polynomial

$$\sin(x) = \sum_{k=0}^n (-1)^k \frac{x^{(2k+1)}}{(2k+1)!}$$

Eventually (as  $n \rightarrow \infty$ )  $P(x)$  will be almost indistinguishable from  $\sin(x)$ . Though it is usually unnecessary to resort to such high order terms.

