

For what follows, we're going to consider the set of real numbers to be the universe of discourse.

CONVEX SETS¹

A **convex combination** is a linear combination of points where all coefficients are non-negative and sum to one.

Consider points (possibly vectors) \mathbf{x} , \mathbf{y} , and \mathbf{z} . A general convex combination, which can be denoted \mathbf{w} , is

$$\mathbf{w} = k_1\mathbf{x} + k_2\mathbf{y} + k_3\mathbf{z}$$

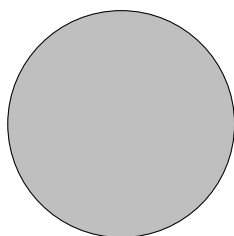
where $k_1 + k_2 + k_3 = 1$ and $k_i \geq 0, i = 1, 2, 3$.

The convex combination we are going to use most is:

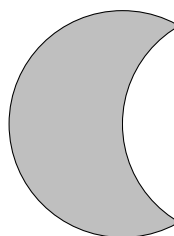
$$\alpha\mathbf{x} + (1 - \alpha)\mathbf{y} \quad \alpha \in [0, 1]$$

Think of it like a weighted average between two points (or vectors), where α determines the weight. The convex combinations made by all possible values of α will be a line between the two points.

$A \subseteq \mathbb{R}^n$ is a **convex set** iff $\alpha\mathbf{x} + (1 - \alpha)\mathbf{y} \in A \quad \forall \mathbf{x}, \mathbf{y} \in A, \alpha \in [0, 1]$



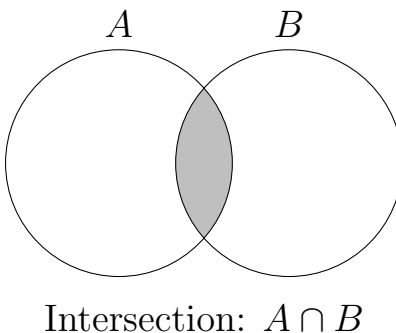
A Convex Set



A Non-Convex Set

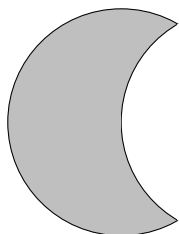
¹Prepared by Sarah Robinson

If A and B are both convex sets in \mathbb{R}^n , then $A \cap B$ is a convex set.

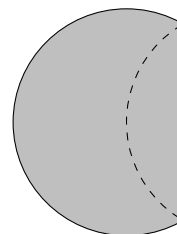


Is $A \cup B$ a convex set?

The **convex hull** of set $B \subseteq \mathbb{R}^n$ is the smallest convex set containing B (the set of all convex combinations of points in B).



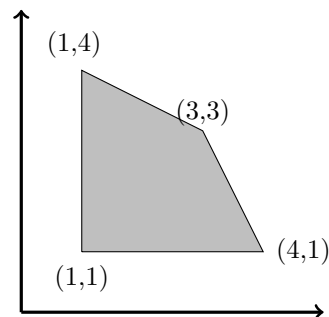
A Non-Convex Set



The Convex Hull

Example: A two-player prisoners' dilemma from game theory and the convex hull of the payoff profiles:

	C	D
C	(3, 3)	(1, 4)
D	(4, 1)	(1, 1)



Example: Consider set S :

$$S = \{x \mid x \in \mathbb{R} \wedge -1 \leq x \leq 1\}$$

Show that S is a convex set.

- $A \subseteq \mathbb{R}^n$ is a **convex set** iff $\alpha \mathbf{x} + (1 - \alpha) \mathbf{y} \in A \quad \forall \mathbf{x}, \mathbf{y} \in A, \alpha \in [0, 1]$

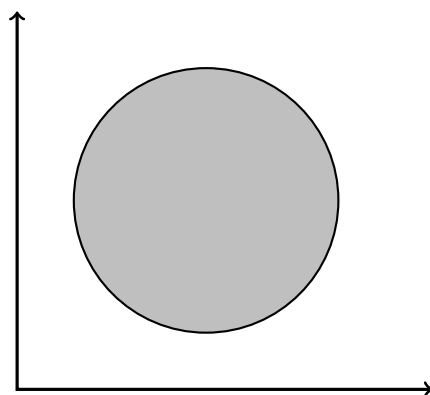
To Show:

Proof:

CLOSED SETS

A set $A \subseteq \mathbb{R}^n$ is **closed** iff for every sequence $\{\mathbf{x}_n\}_{n=1}^{\infty}$ such that $\mathbf{x}_n \in A$ for all n and $\mathbf{x}_n \rightarrow \mathbf{x}$, it is also the case that $\mathbf{x} \in A$

- \approx set A also includes its boundaries



A Closed Set in \mathbb{R}^2

A set is an **open set** if and only if its complement is a closed set.

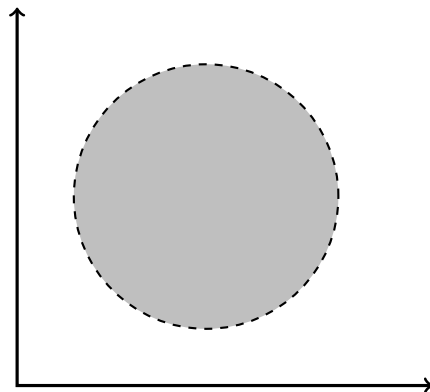
The following sets in \mathbb{R}^n are open sets:

- The empty set \emptyset
- The entire space \mathbb{R}^n
- The union of any number of open sets
- The intersection of any finite number of open sets

The following sets in \mathbb{R}^n are closed sets:

- The empty set \emptyset
- The entire space \mathbb{R}^n
- The union of any finite number of closed sets
- The intersection of any number of closed sets

We could also define open sets using the notion of an epsilon-neighborhood (a ball with radius ε). A set A is open if and only if for all $\mathbf{x} \in A$, there exists some $\varepsilon > 0$ such that the ε -ball centered at \mathbf{x} is contained in A .



An Open Set in \mathbb{R}^2

For any point in an open set, we can always draw a tiny circle around the point that lies entirely within the set. I bring up this definition because ε -balls will come up in other contexts.

Example: Consider set S :

$$S = \{(x, y) \mid (x, y) \in \mathbb{R}^2 \wedge x^2 + y^2 \leq 1\}$$

Show that S is closed.

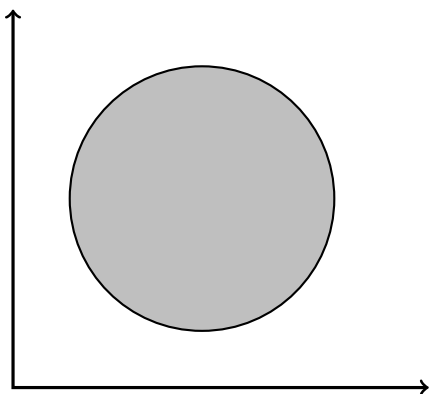
- $A \subseteq \mathbb{R}^n$ is **closed** iff for every sequence $\{\mathbf{x}_n\}_{n=1}^{\infty}$ such that $\mathbf{x}_n \in A$ for all n and $\mathbf{x}_n \rightarrow \mathbf{x}$, it is also the case that $\mathbf{x} \in A$
- Theorem 1: If $a_n \rightarrow a$ and $b_n \rightarrow b$, then $a_n + b_n \rightarrow a + b$ and $a_n b_n \rightarrow ab$
- Theorem 2: If $a_n \rightarrow a$, then $a_n \leq b$ for all n implies $a \leq b$.

To Show:

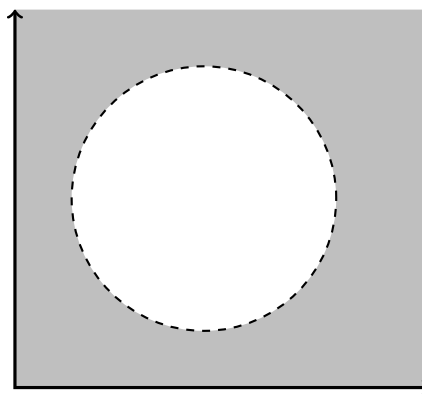
Proof:

BOUNDED SETS

A set $A \subseteq \mathbb{R}^n$ is **bounded** if and only if there exists an M and a point $\mathbf{c} \in \mathbb{R}^n$ such that the M -ball centered at \mathbf{c} contains all of A .



A Bounded (Closed) Set



A Non-Bounded (Open) Set

To prove a set in $A \subseteq \mathbb{R}^n$ is bounded:

- Pick a radius M
- Pick a center point \mathbf{c}
- Let $\mathbf{x} \in A$ – arbitrary \mathbf{x}
- Show that \mathbf{x} is less than M distance away from \mathbf{c}

To prove a set in $A \subseteq \mathbb{R}^n$ is bounded in the special case where the points furthest away from 0 are along the axes:

- Pick a radius M (use 0 as the center point)
- Let $\mathbf{x} \in A$ – arbitrary \mathbf{x}
- Show that $-M \leq x_i \leq M \quad \forall i = 1, \dots, n$

A set $A \subseteq \mathbb{R}^n$ is **compact** if and only if it is closed and bounded.

Example: Consider set S :

$$S = \{(x, y) \mid (x, y) \in \mathbb{R}^2 \wedge x^2 + y^2 \leq 1\}$$

Show that S is bounded.

To prove a set in $A \subseteq \mathbb{R}^n$ is bounded in the special case where the points furthest away from 0 are along the axes:

- Pick a radius M (use 0 as the center point)
- Let $\mathbf{x} \in A$ – arbitrary \mathbf{x}
- Show that $-M \leq x_i \leq M \quad \forall i = 1, \dots, n$

To Show: S is bounded

Proof:

Let $M = 2$	(by hypothesis)
$\implies x^2 + y^2 \leq 1$	(def. of S)
$\implies (x^2 \leq 1) \wedge (y^2 \leq 1)$	($x^2 \geq 0 \quad \forall x$)
$\implies (-1 \leq x \leq 1) \wedge (-1 \leq y \leq 1)$	(algebra)
$\implies (-2 \leq x \leq 2) \wedge (-2 \leq y \leq 2)$	(algebra)
$\implies (-M \leq x \leq M) \wedge (-M \leq y \leq M)$	(algebra)
$\implies S$ is bounded	(by def. of bounded)

CONTINUITY & DIFFERENTIABILITY OF FUNCTIONS

Let f be a function with domain D and points (or vectors) $\mathbf{x}, \mathbf{y} \in D$.

f is **continuous** at \mathbf{x} iff for every $\varepsilon > 0$ there exists $\delta > 0$ such that $\|\mathbf{x} - \mathbf{y}\| < \delta \implies \|f(\mathbf{x}) - f(\mathbf{y})\| < \varepsilon$.

Recall that for $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$,

$$\|\mathbf{x} - \mathbf{y}\| = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

f is **continuous** at \mathbf{x} iff for every sequence $\mathbf{x}_n \in D$ such that $\mathbf{x}_n \rightarrow \mathbf{x}$, the sequence $f(\mathbf{x}_n) \rightarrow f(\mathbf{x})$.

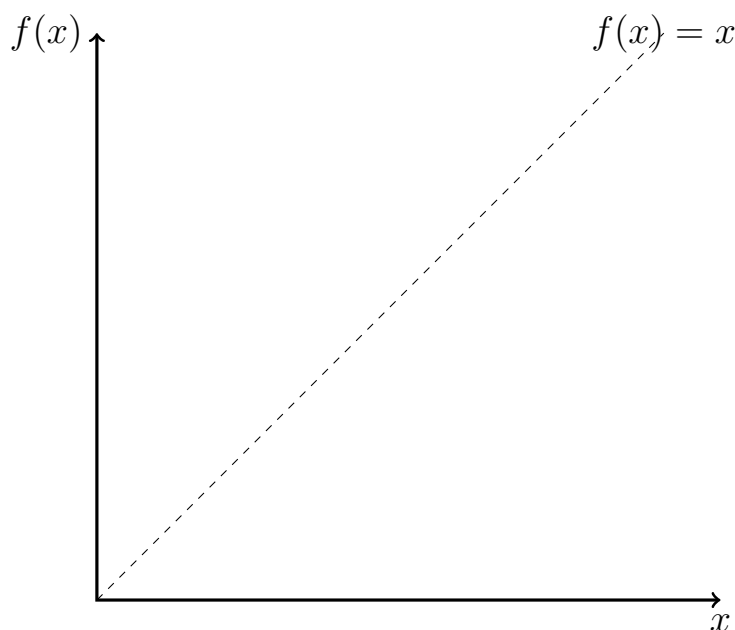
f is a continuous function if it is continuous at every point in its domain.

Let $f : D \rightarrow \mathbb{R}$ be a continuous, real-valued function where D is non-empty, compact subset of \mathbb{R}^n . Then there exists a vector $\underline{\mathbf{x}} \in D$ and a vector $\overline{\mathbf{x}} \in D$ such that

$$\forall \mathbf{x} \in A, f(\underline{\mathbf{x}}) \leq f(\mathbf{x}) \leq f(\overline{\mathbf{x}})$$

That is, a continuous function $f(\mathbf{x})$ attains a maximum and a minimum on every compact set. (Weierstrass Extreme Value Theorem).

Let $D \subseteq \mathbb{R}^n$ be a non-empty compact, convex set. Let $f : D \rightarrow D$ be a continuous function. Then there exists at least one fixed point of f in D , that is, there exists $\mathbf{x}^* \in D$ such that $f(\mathbf{x}^*) = \mathbf{x}^*$. (Brouwer Fixed Point Theorem).



For $D = [0, 1]$ then any continuous $f : D \rightarrow D$ must cross the 45-degree line.

Let f be a function defined on an interval $(a, b) \subseteq \mathbb{R}$ and let $x \in (a, b)$. Then f is **differentiable** at x if and only if the limit of

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

exists and is finite.

If this is the case, then the limit is called the **derivative** of f at x and is denoted $f'(x)$ or $\frac{df(x)}{dx}$.

For a multivariate functions $f(\mathbf{x})$ where $\mathbf{x} \in \mathbb{R}^n$, the **partial derivative** of f with respect to x_i is given by:

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

$f'(x)$ is a function of x . Often, we want to discuss the value of the derivative at a particular point c :

$$f'(c) \qquad \left. \frac{df}{dx} \right|_c$$

LEVEL SETS

We are focusing on real-valued functions:

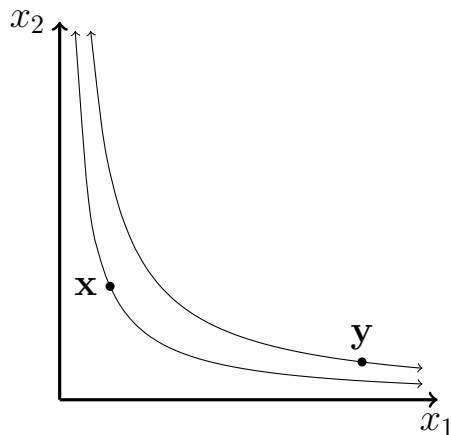
- $f : \mathbb{R} \rightarrow \mathbb{R}$ (univariate)
- $f : \mathbb{R}^n \rightarrow \mathbb{R}$ (multivariate)

Let f be a real valued function such that $f : D \rightarrow \mathbb{R}$ where $D \subseteq \mathbb{R}^n$. Then $L(\mathbf{x}_0)$ is a **level set** relative to \mathbf{x}_0 if and only if

$$L(\mathbf{x}_0) = \{\mathbf{x} \mid \mathbf{x} \in D \wedge f(\mathbf{x}) = f(\mathbf{x}_0)\}$$

Indifference curves are level sets. Consider the utility function:

$$u(x_1, x_2) = x_1^{1/2} x_2^{1/2}$$



$$\bullet \mathbf{x} = (1, 4) \text{ and } u(\mathbf{x}) = 2$$

$$\bullet \mathbf{y} = (32, \frac{1}{2}) \text{ and } u(\mathbf{y}) = 4$$

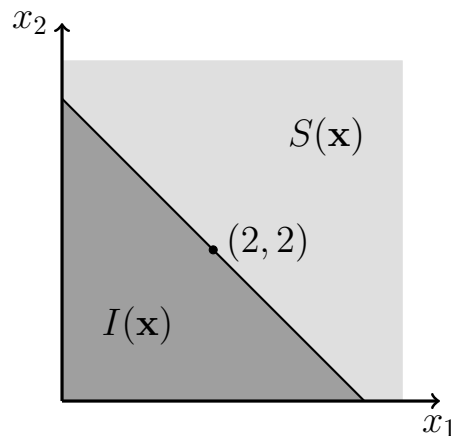
All the points on the curve running through \mathbf{x} give a utility of 2, while all those on the curve running through \mathbf{y} provide a utility of 4.

We can also define superior and inferior sets:

- $S(\mathbf{x}_0) = \{\mathbf{x} \mid \mathbf{x} \in D \wedge f(\mathbf{x}) \geq f(\mathbf{x}_0)\}$ is the **superior set** relative to \mathbf{x}_0
- $I(\mathbf{x}_0) = \{\mathbf{x} \mid \mathbf{x} \in D \wedge f(\mathbf{x}) \leq f(\mathbf{x}_0)\}$ is the **inferior set** relative to \mathbf{x}_0

If the weak inequalities are replaced with strict inequalities, then the sets are the **strictly superior set** and **strictly inferior set**, respectively.

Example: Consider the function $u(x_1, x_2) = x_1 + x_2$. The inferior and superior sets, relative to $\mathbf{x} = (2, 2)$ can be illustrated graphically as:



\neq ON AND ABOVE/BELOW

Let $f : D \rightarrow \mathbb{R}$, where $D \subseteq \mathbb{R}^n$ and $\mathbb{R} \subseteq \mathbb{R}$. The the set of points **on and below the graph** of f is defined as:

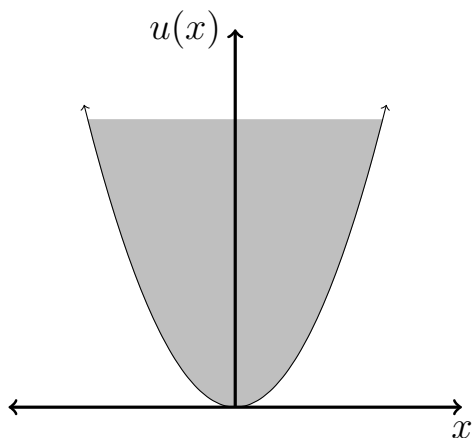
$$A = \{(\mathbf{x}, y) \mid \mathbf{x} \in D \wedge f(\mathbf{x}) \geq y\}$$

Similarly, the set of points **on and above the graph** is defined as:

$$B = \{(\mathbf{x}, y) \mid \mathbf{x} \in D \wedge f(\mathbf{x}) \leq y\}$$

Note that superior/inferior sets are points in the domain, while points relative to graph are *ordered pairs*, $(n + 1)$ -tuples with elements from both the domain *and* the range.

Consider the set of points on and above the graph of the function $u(x) = x^2$.



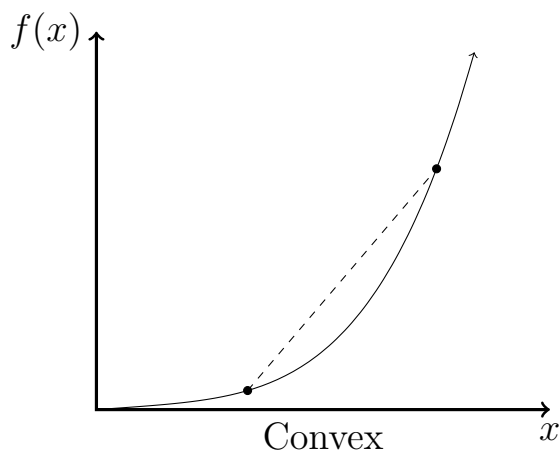
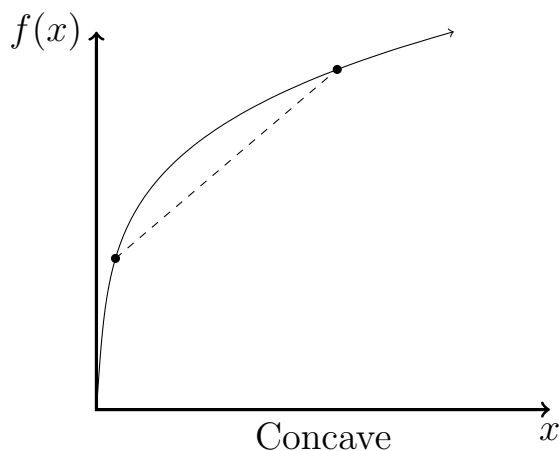
CONCAVITY AND CONVEXITY

Let $f : D \rightarrow \mathbb{R}$ where D is a convex subset of \mathbb{R}^n . A function is **concave** if and only if for all $\mathbf{x}_0, \mathbf{x}_1 \in D$ and $t \in [0, 1]$:

$$f\left(t\mathbf{x}_0 + (1-t)\mathbf{x}_1\right) \geq tf(\mathbf{x}_0) + (1-t)f(\mathbf{x}_1)$$

A function is **convex** if and only if for all $\mathbf{x}_0, \mathbf{x}_1 \in D$ and $t \in [0, 1]$:

$$f\left(t\mathbf{x}_0 + (1-t)\mathbf{x}_1\right) \leq tf(\mathbf{x}_0) + (1-t)f(\mathbf{x}_1)$$



Example: Consider $f : \mathbb{R} \rightarrow \mathbb{R}$ where $f(x) = |x|$. Prove that it is convex.

- A function is convex iff for all $\mathbf{x}_0, \mathbf{x}_1 \in D$ and $t \in [0, 1]$,
$$f\left(t\mathbf{x}_0 + (1-t)\mathbf{x}_1\right) \leq tf(\mathbf{x}_0) + (1-t)f(\mathbf{x}_1)$$
- The absolute value function $|x| = \begin{cases} x & \text{if } x \geq 0 \\ -x & \text{if } x < 0 \end{cases}$
- Theorem 1: $|ab| = |a||b|$
- Theorem 2: The triangle inequality, $|a + b| \leq |a| + |b|$

To show: $f(tx + (1-t)y) \leq tf(x) + (1-t)f(y)$

Proof:

Let $f : D \rightarrow \mathbb{R}$, where $D \subseteq \mathbb{R}^n$ and $\mathbb{R} \subseteq \mathbb{R}$. Then:

f is a concave function \iff the set on and below f is a convex set

f is a convex function \iff the set on and above f is a convex set

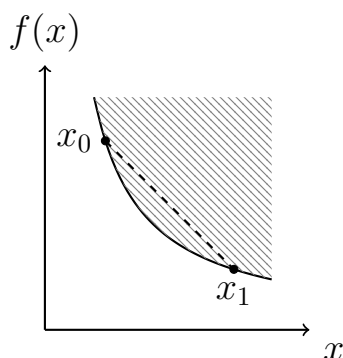
Let $f : D \rightarrow \mathbb{R}$ where D is a convex subset of \mathbb{R}^n . A function is **strictly concave** if and only if for all $\mathbf{x}_0, \mathbf{x}_1 \in D \ni \mathbf{x}_0 \neq \mathbf{x}_1$ and $t \in (0, 1)$:

$$f\left(t\mathbf{x}_0 + (1-t)\mathbf{x}_1\right) > tf(\mathbf{x}_0) + (1-t)f(\mathbf{x}_1)$$

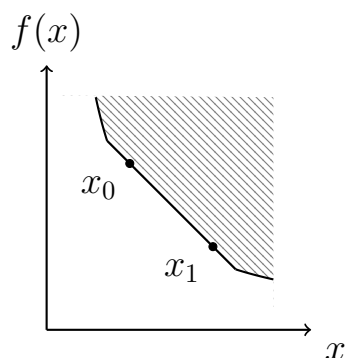
A function is **strictly convex** if and only if for all $\mathbf{x}_0, \mathbf{x}_1 \in D \ni \mathbf{x}_0 \neq \mathbf{x}_1$ and $t \in (0, 1)$:

$$f\left(t\mathbf{x}_0 + (1-t)\mathbf{x}_1\right) < tf(\mathbf{x}_0) + (1-t)f(\mathbf{x}_1)$$

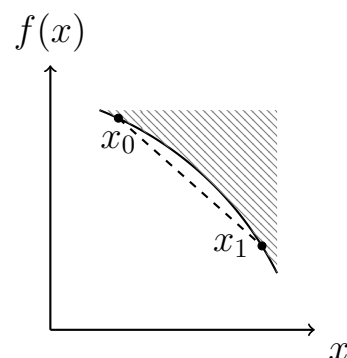
(We've changed the inequality, made sure the two points are distinct, and made t strictly between 0 and 1).



Strictly convex



Convex but not strictly



Strictly concave

Let D be a convex, non-degenerate interval on \mathbb{R} , such that on the interior of D , f is twice continuously differentiable. Then the following statements are equivalent:

1. f is concave
2. $f''(x) \leq 0$ for all non-endpoints $x \in D$.
3. For all $x_0 \in D$, $f(x) \leq f(x_0) + f'(x_0)(x - x_0)$
4. $f''(x) < 0$ for all non-endpoints $x \in D \implies f$ is strictly concave

The following statements are also equivalent:

1. f is convex
2. $f''(x) \geq 0$ for all non-endpoints $x \in D$.
3. For all $x_0 \in D$, $f(x) \geq f(x_0) + f'(x_0)(x - x_0)$
4. $f''(x) > 0$ for all non-endpoints $x \in D \implies f$ is strictly convex