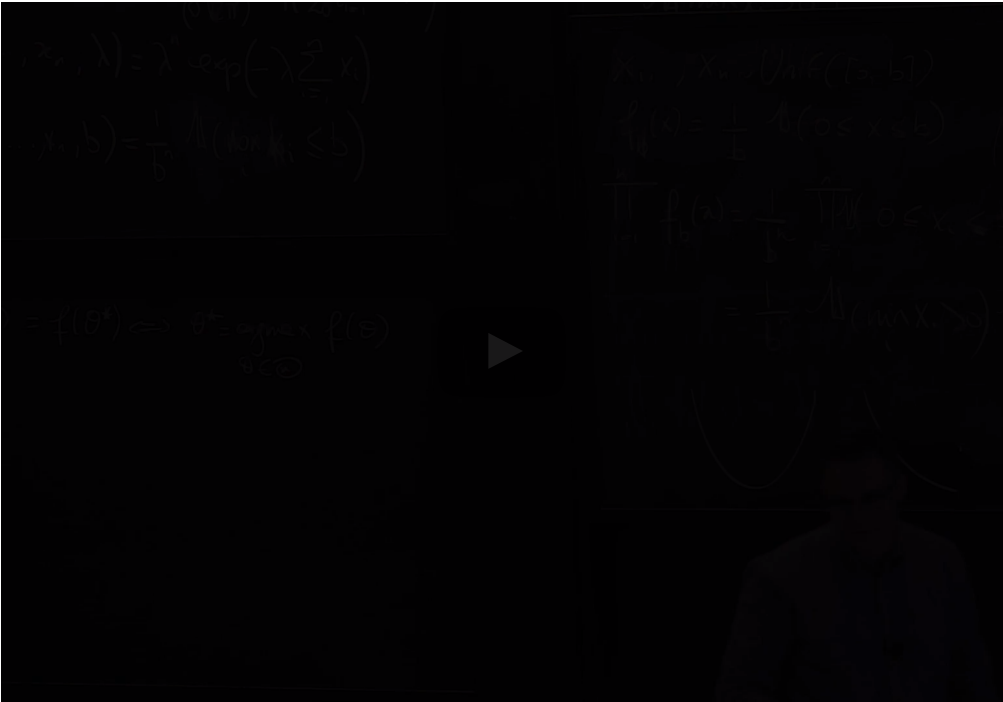


# Concavity in 1 dimension



which is just minus

it's the same as concavity from minus the function.

OK, so a convex function-- so if this is boom, boom, and boom, just think of what would happen if you put a negative sign.

So this is convex, this is convex, and this is convex.

OK.

▶ 7:09 / 7:09

▶ 1.50x

🔊

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CC

“

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## Video

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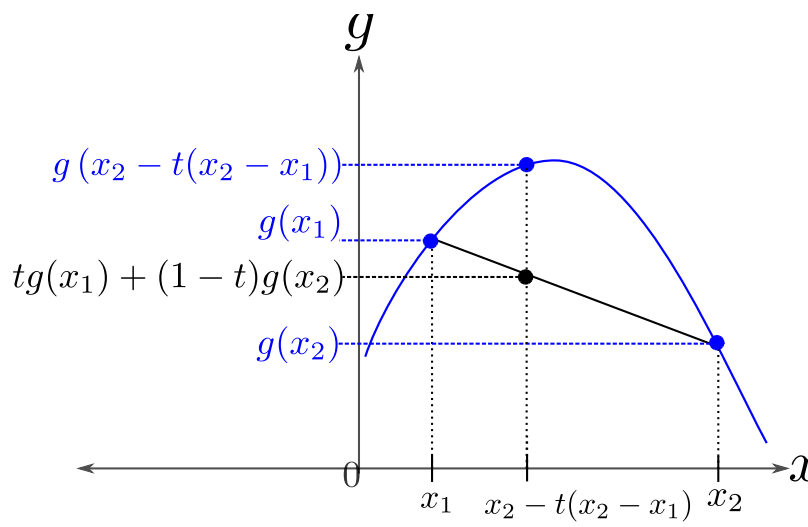
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A function  $g : I \rightarrow \mathbb{R}$  is **concave** (or concave down) on an interval  $I$ , if for all pairs of real numbers  $x_1 < x_2 \in I$

$$g(tx_1 + (1 - t)x_2) \geq tg(x_1) + (1 - t)g(x_2) \quad \text{for all } 0 < t < 1.$$

Geometrically, this means that for  $x_1 < x < x_2$ , the graph of  $g$  is **above** the secant line connecting the two points  $(x_1, g(x_1))$  and  $(x_2, g(x_2))$ .



At  $x = x_2 - t(x_2 - x_1) = tx_1 + (1 - t)x_2$ , the  $y$ -value of the graph of  $g$  is  $g(x) = g(tx_1 + (1 - t)x_2)$ , while the  $y$ -value of the secant line is  $tg(x_1) + (1 - t)g(x_2)$ .  
If the inequality is strict, i.e. if

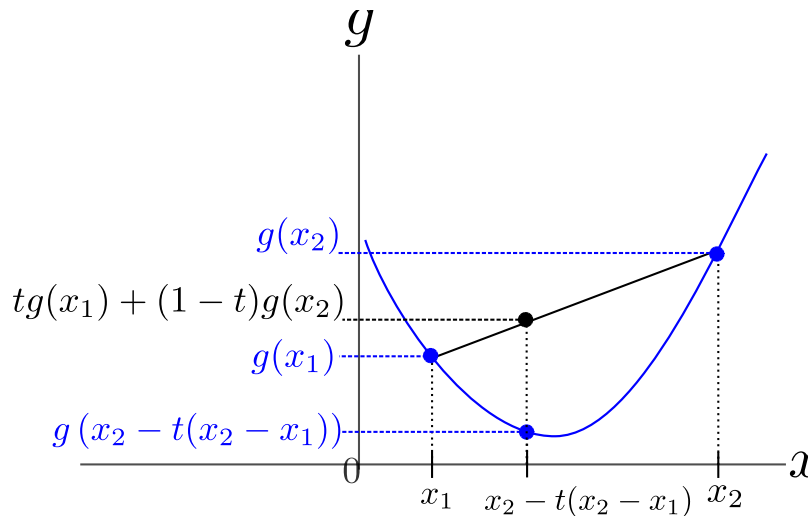
$$g(tx_1 + (1 - t)x_2) > tg(x_1) + (1 - t)g(x_2) \quad \text{for all } 0 < t < 1.$$

then  $g$  is **strictly concave**.

The definition for **(strictly) convex** is analogous. A function  $g : I \rightarrow \mathbb{R}$  is **convex** (or concave up), where  $I$  is an interval, if for all pairs of real numbers  $x_1 < x_2 \in I$

$$g(tx_1 + (1 - t)x_2) \leq tg(x_1) + (1 - t)g(x_2) \quad \text{for all } 0 < t < 1.$$

Geometrically, this means that for  $x_1 < x < x_2$ , the graph of  $g$  is **below** the secant line connecting the two points  $(x_1, g(x_1))$  and  $(x_2, g(x_2))$ .



At  $x = x_2 - t(x_2 - x_1) = tx_1 + (1 - t)x_2$ , the  $y$ -value of the graph of  $g$  is  $g(x) = g(tx_1 + (1 - t)x_2)$ , while the  $y$ -value of the secant line is  $tg(x_1) + (1 - t)g(x_2)$ .

If the inequality is strict, i.e. if

$$g(tx_1 + (1-t)x_2) < tg(x_1) + (1-t)g(x_2) \quad \text{for all } 0 < t < 1.$$

then  $g$  is **strictly convex**.

If in addition  $g$  is twice differentiable in the interval  $I$ , i.e.  $g''(x)$  exists for all  $x \in I$ , then  $g$  is

- **concave** if and only if  $g''(x) \leq 0$  for all  $x \in I$ ;
- **strictly concave** if  $g''(x) < 0$  for all  $x \in I$ ;
- **convex** if and only if  $g''(x) \geq 0$  for all  $x \in I$ ;
- **strictly convex** if  $g''(x) > 0$  for all  $x \in I$ ;

**Note:** In the lecture video and slides, we used these inequality conditions on the second derivative to define concave functions and strictly concave functions *analytically*. The *synthetic* definition above is slightly more general because it does not require differentiability at every point. For example, the function  $x \mapsto x^4$  is strictly convex according to the definition above, but has three vanishing derivatives at the origin  $x = 0$ .

## Discussion

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**Topic:** Unit 3 Methods of Estimation: Lecture 9: Introduction to Maximum Likelihood Estimation / 6. Interlude: Minimizing and Maximizing Functions

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