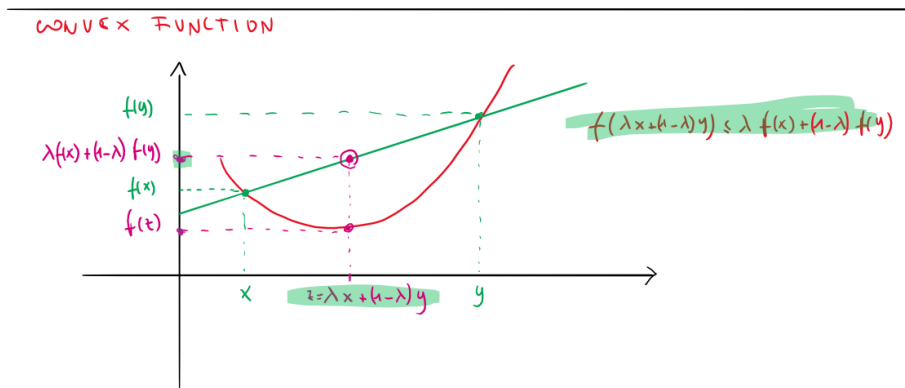


convex function = any function that has a shape like a bowl

$$f(tx + (1 - t)y) \leq tf(x) + (1 - t)f(y) \quad \text{for all } x, y \in \mathcal{X}, t \in [0, 1]. \quad - (1)$$

Notice the equal in the “ \leq ” sign!

From Rinaldi’s whiteboard:



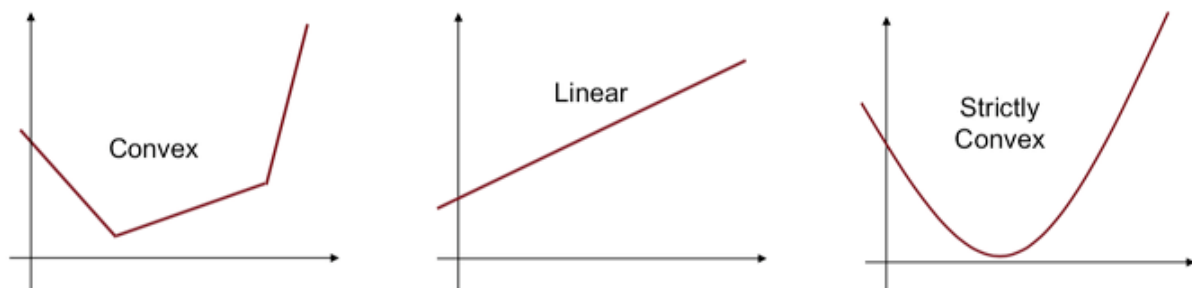
Interpretation: For any two points, we draw a line between them, and for any point on that line, the value of the function in this point is below the point on the line.

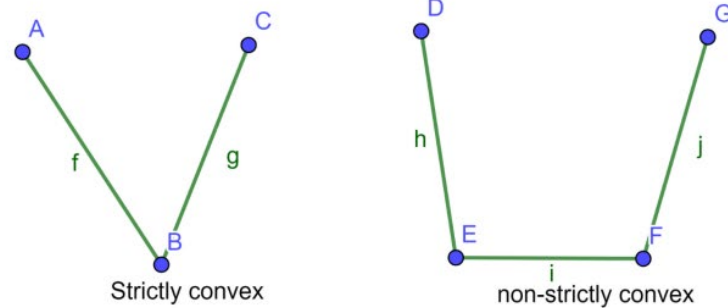
strictly convex function = has only one minimum (local = global)

$$f(tx + (1 - t)y) < tf(x) + (1 - t)f(y) \quad \text{for all } x \neq y \in \mathcal{X}, t \in (0, 1).$$

Here we don’t have the equal in the sign!

Examples for strictly convex:





Every strictly convex function is convex, but the reverse is not true. (see pictures above)

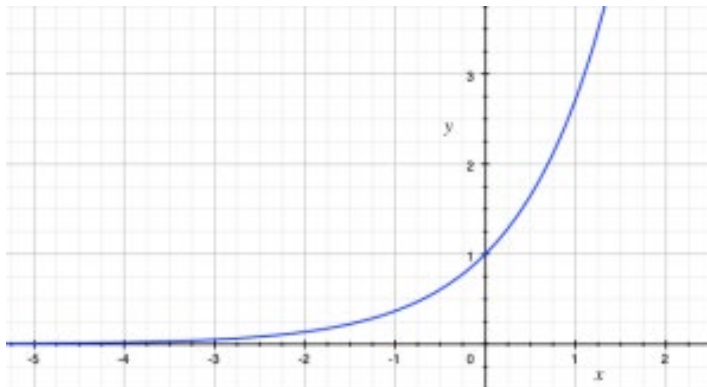
Strongly convex function = how “convex” or “curved” a convex function is.

It can be seen as a kind of “parameterized strict convexity”.

$$f(tx + (1 - t)y) \leq tf(x) + (1 - t)f(y) - \frac{1}{2}mt(1 - t)\|x - y\|_2^2$$

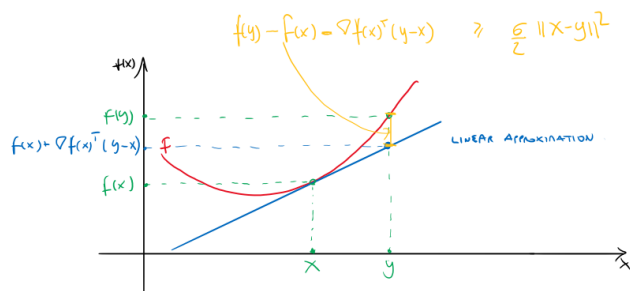
for all $x, y \in \mathcal{X}, t \in [0, 1]$. — (4)

You can think of the parameter m as measuring how “curved” the function is: the larger m is the more curved f is. That is why $f(x) = \exp(x)$ is not strongly convex: as x goes to infinity, the curve becomes flatter and flatter ($m \rightarrow 0$).



From Rinaldi's whiteboard: (sigma is the same as m above)

1ST ORDER STRONG CONVEXITY



Between two strongly convex functions, the function with the larger m is more strongly convex.
 => this function's curvature grows more rapidly

Extras for strong convexity:

From Rinaldi's slides (here lambda refers to t in above's equation):

Finally, by taking limit $\lambda \rightarrow 0$, we can write

$$f(x) \geq f(x^*) + \frac{\sigma}{2} \|x - x^*\|^2,$$

Strong Convexity

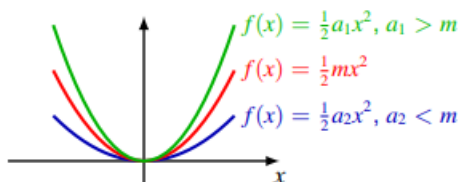
A function f is **strongly convex** with parameter $m > 0$, or simply **m -strongly convex**, if

$$\tilde{f}(x) = f(x) - \frac{m}{2} \|x\|^2$$

is convex.

Note. $f(x) = \frac{m}{2} \|x\|^2 + \tilde{f}(x)$, i.e. f is $\frac{m}{2} \|x\|^2$ plus an extra convex term. Informally, " m -strongly convex" means at least as "convex" as $\frac{m}{2} \|x\|^2$.

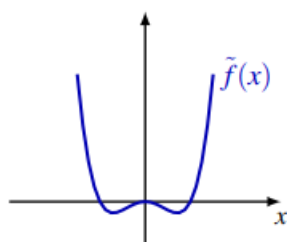
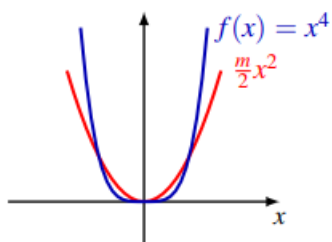
Example. $f(x) = \frac{a}{2} \|x\|^2$ is m -strongly convex iff $a \geq m$



Example. $f(x) = x^4$ is **not** m -strongly convex for any $m > 0$, as $\tilde{f}(x) = x^4 - \frac{m}{2}x^2$ is not convex,

$$\tilde{f}''(x) = 12x^2 - m < 0$$

for $|x| < \sqrt{m/12}$.



Useful link:

[\(Strictly/strongly\) convex functions | Statistical Odds & Ends \(wordpress.com\)](#)