

# Multivariate calculus reminders

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# Multivariate functions

We let  $f : \mathbb{R}^p \rightarrow \mathbb{R}$ .

**Gradient:**  $\nabla f(x) \in \mathbb{R}^p, \quad [\nabla f(x)]_i = \frac{\partial f}{\partial x_i}(x)$



If it exists,  $f$  is  
**differentiable**

**Hessian :**  $\nabla^2 f(x) \in \mathbb{R}^{p \times p}, \quad [\nabla^2 f(x)]_{ij} = \frac{\partial^2 f}{\partial x_i \partial x_j}(x)$

# Taylor expansion

We let  $f : \mathbb{R}^p \rightarrow \mathbb{R}$ ,  $x, \varepsilon \in \mathbb{R}^p$

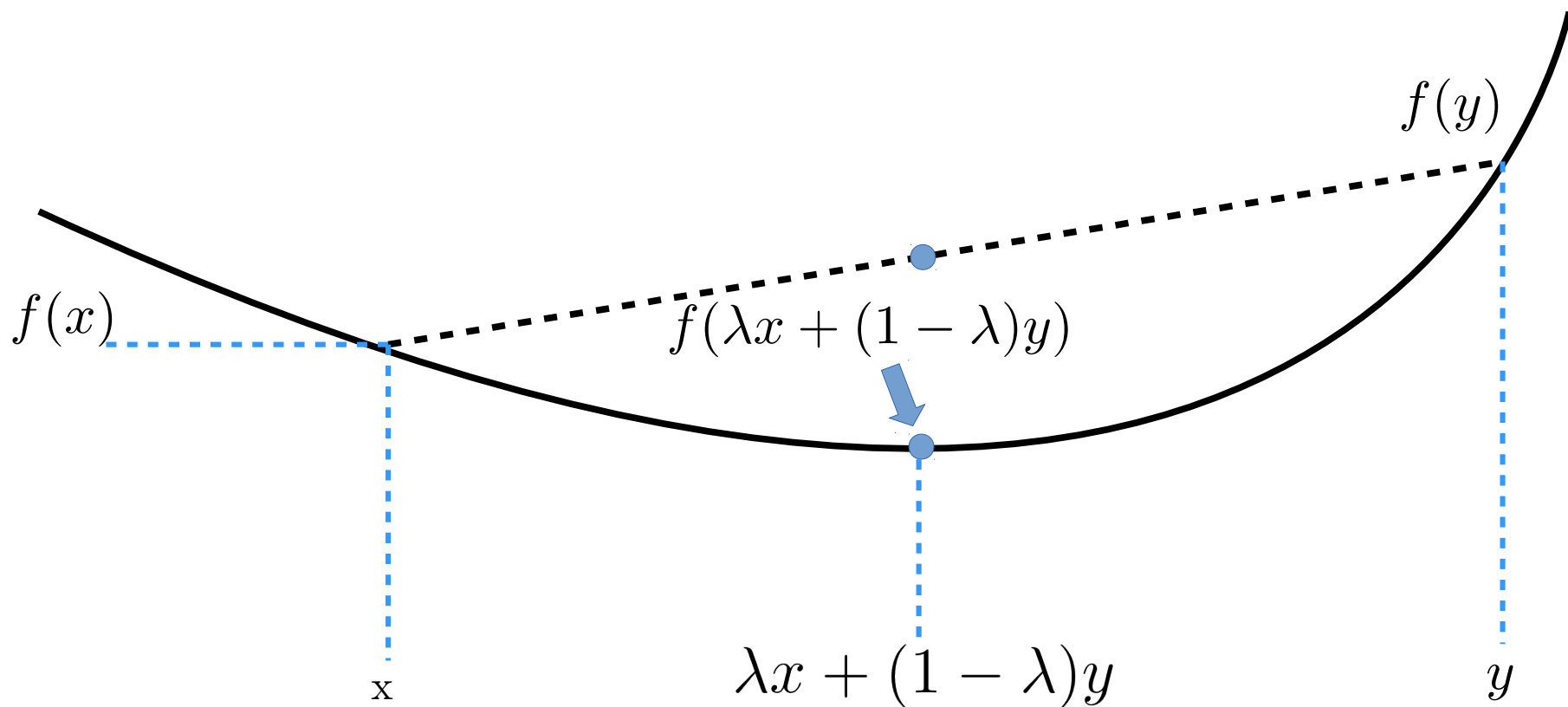
$$f(x + \varepsilon) = f(x) + \langle \nabla f(x), \varepsilon \rangle + \frac{1}{2} \langle \varepsilon, \nabla^2 f(x) \varepsilon \rangle + o(\|\varepsilon\|^2)$$

$f$  locally looks like a quadratic function

# Convexity, zero-th order definition

We let  $f : \mathbb{R}^p \rightarrow \mathbb{R}$ .  $f$  is convex if

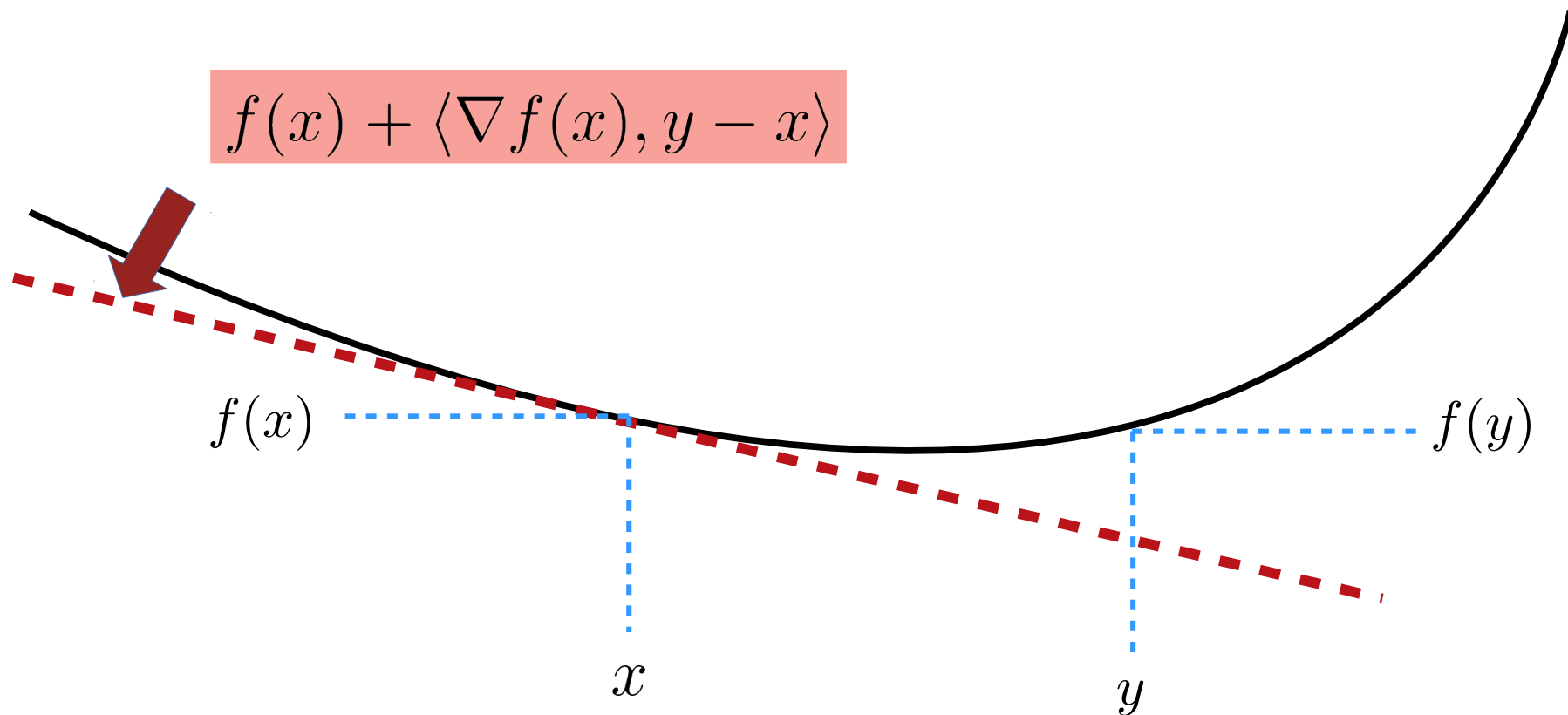
$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y), \quad \forall x, y \in \mathbb{R}^p, \lambda \in [0, 1]$$



# Convexity, first order definition

We let  $f : \mathbb{R}^p \rightarrow \mathbb{R}$ .  $f$  is convex if

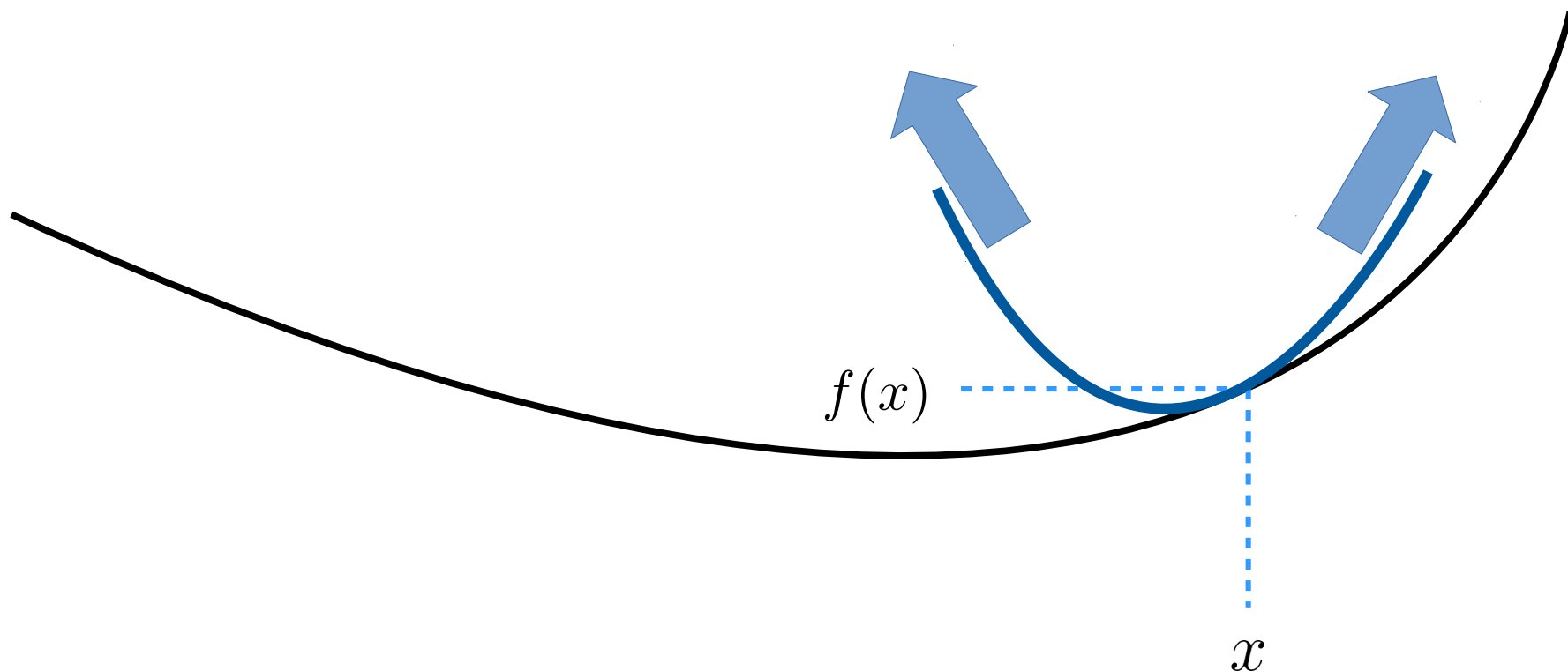
$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle, \quad \forall x, y \in \mathbb{R}^p$$



# Convexity, second order definition

We let  $f : \mathbb{R}^p \rightarrow \mathbb{R}$ .  $f$  is convex if

$$\nabla^2 f(x) \succeq 0, \quad \forall x \in \mathbb{R}^p$$



# Questions

Show that the following functions are convex:

$$\|x\|^2 = \sum_{i=1}^p x_i^2$$

$$\|x\|_1 = \sum_{i=1}^p |x_i|$$

$$F(x) = f(\langle x, y \rangle), \quad \text{for } y \in \mathbb{R}^p \text{ and } f : \mathbb{R} \rightarrow \mathbb{R} \text{ convex}$$

# Questions

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# Smoothness

$f$  is  $L$ -smooth  $\iff \nabla f$  is  $L$ -Lipschitz

**!!  $f$  must be differentiable**

$$\|\nabla f(x) - \nabla f(y)\| \leq L\|x - y\|, \quad \forall x, y$$

## Equivalent formulations:

$$f(y) \leq f(x) + \langle \nabla f(x), y - x \rangle + \frac{L}{2}\|x - y\|^2, \quad \forall x, y \in \mathbb{R}^p$$


$$\nabla^2 f(x) \preceq L \cdot I_p, \quad \forall x \in \mathbb{R}^p$$

**!! If  $\nabla^2 f$  is continuous,  $f$  is  $L$ -smooth on all compact sets**

# Strong convexity

$f$  is  $\mu$  - strongly convex if:

$< 0$  so stronger than convexity !


$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y) - \lambda(1 - \lambda)\frac{\mu}{2}\|x - y\|^2$$

## Equivalent formulations:

$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle + \frac{\mu}{2}\|x - y\|^2, \quad \forall x, y \in \mathbb{R}^p$$

$$\nabla^2 f(x) \succeq \mu \cdot I_p, \quad \forall x \in \mathbb{R}^p$$