

# Recitation 1

## Gradients and Directional Derivatives

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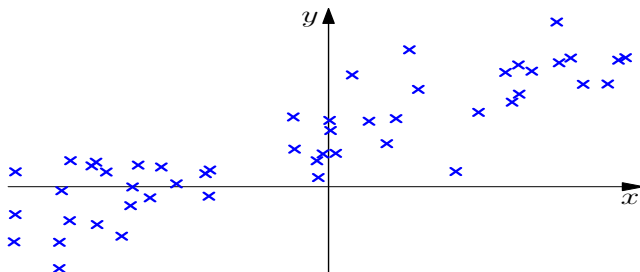
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# Intro Question

## Question

We are given the data set  $(x_1, y_1), \dots, (x_n, y_n)$  where  $x_i \in \mathbb{R}^d$  and  $y_i \in \mathbb{R}$ . We want to fit a linear function to this data by performing empirical risk minimization. More precisely, we are using the hypothesis space  $\mathcal{F} = \{f(x) = w^T x \mid w \in \mathbb{R}^d\}$  and the loss function  $\ell(a, y) = (a - y)^2$ . Given an initial guess  $\tilde{w}$  for the empirical risk minimizing parameter vector, how could we improve our guess?



# Intro Solution

## Solution

- The empirical risk is given by

$$\hat{R}_n(f) = \frac{1}{n} \sum_{i=1}^n \ell(f(x_i), y_i) = \frac{1}{n} \sum_{i=1}^n (w^T x_i - y_i)^2 = \frac{1}{n} \|Xw - y\|_2^2,$$

where  $X \in \mathbb{R}^{n \times d}$  is the matrix whose  $i$ th row is given by  $x_i$ .

- Can improve a non-optimal guess  $\tilde{w}$  by taking a small step in the direction of the negative gradient.

# Single Variable Differentiation

- For  $f : \mathbb{R} \rightarrow \mathbb{R}$  differentiable, the derivative is given by

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

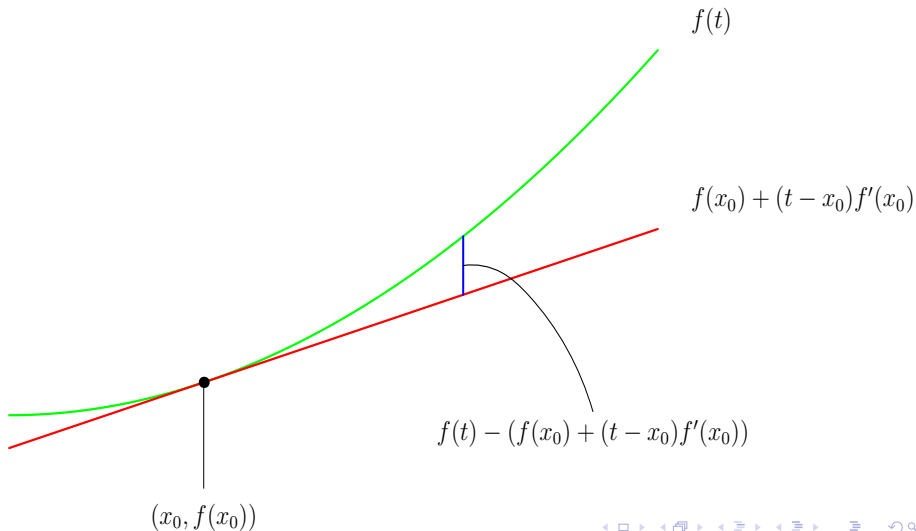
- Can also be written as

$$f(x+h) = f(x) + hf'(x) + o(h) \quad \text{as } h \rightarrow 0,$$

where  $o(h)$  denotes a function  $g(h)$  with  $g(h)/h \rightarrow 0$  as  $h \rightarrow 0$ .

- Points with  $f'(x) = 0$  are called *critical points*.

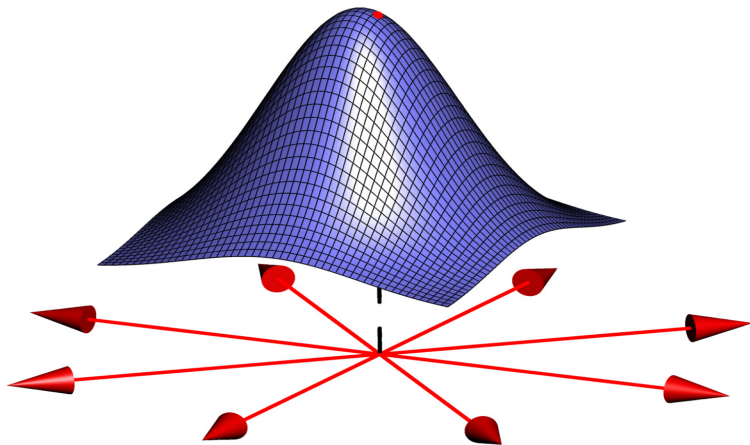
# 1D Linear Approximation By Derivative



# Multivariable Differentiation

- Consider now a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  with inputs of the form  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ .
- Unlike the 1-dimensional case, we cannot assign a single number to the slope at a point since there are many directions we can move in.

# Multiple Possible Directions for $f : \mathbb{R}^2 \rightarrow \mathbb{R}$



# Directional Derivative

## Definition

Let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ . The directional derivative  $f'(x; u)$  of  $f$  at  $x \in \mathbb{R}^n$  in the direction  $u \in \mathbb{R}^n$  is given by

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

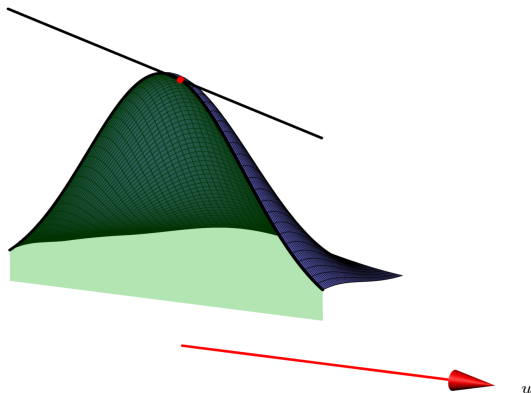
- By fixing a direction  $u$  we turned our multidimensional problem into a 1-dimensional problem.
- Similar to 1-d we have

$$f(x + hu) = f(x) + hf'(x; u) + o(h).$$

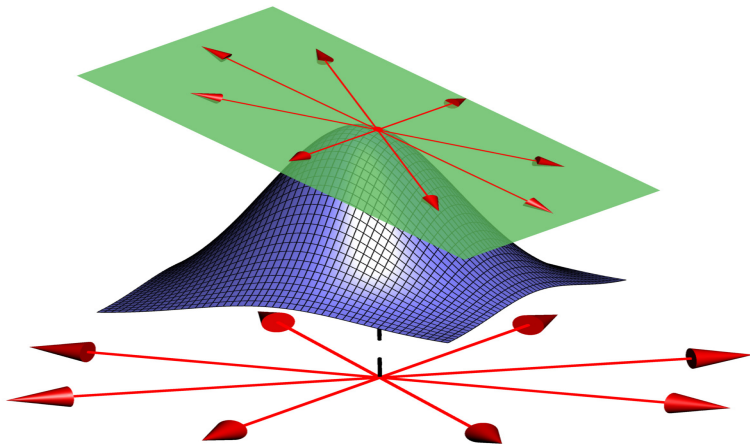
- We say that  $u$  is a *descent direction* of  $f$  at  $x$  if  $f'(x; u) < 0$ .



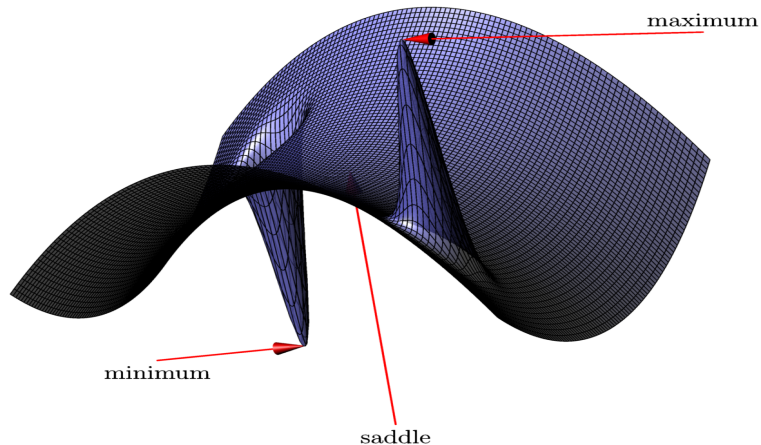
# Directional Derivative as a Slope of a Slice



# Tangent Plane for $f : \mathbb{R}^2 \rightarrow \mathbb{R}$



# Critical Points of $f : \mathbb{R}^2 \rightarrow \mathbb{R}$



# Computing Gradients

## Question

For each of the following functions, compute the gradient.

- ①  $f : \mathbb{R}^3 \rightarrow \mathbb{R}$  is given by

$$f(x_1, x_2, x_3) = \log(1 + e^{x_1 + 2x_2 + 3x_3}).$$

- ②  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is given by

$$f(x) = \|Ax - y\|_2^2 = (Ax - y)^T (Ax - y) = x^T A^T A x - 2y^T A x + y^T y,$$

for some  $A \in \mathbb{R}^{m \times n}$  and  $y \in \mathbb{R}^m$ .