

# Recitation 1

## Gradients and Directional Derivatives

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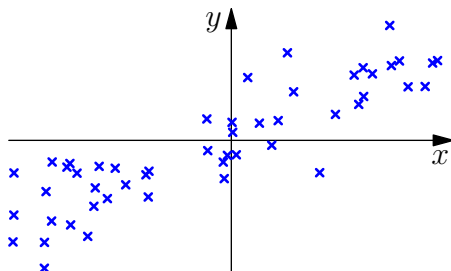
CDS at NYU

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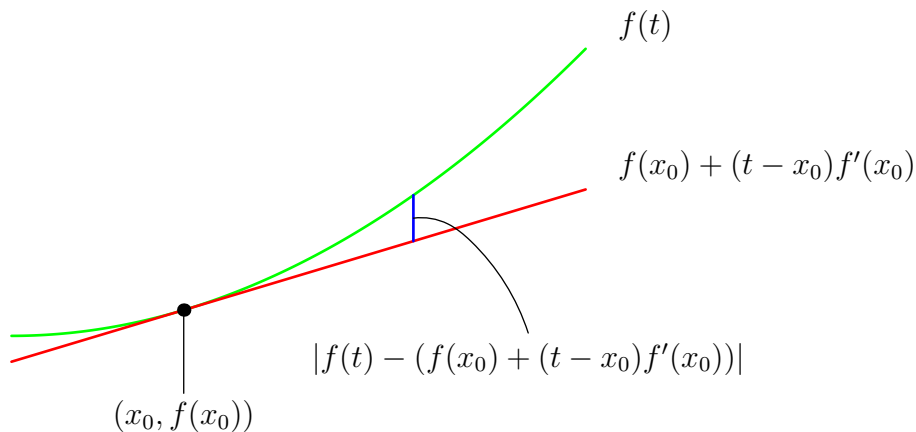
# Intro Question

## Question

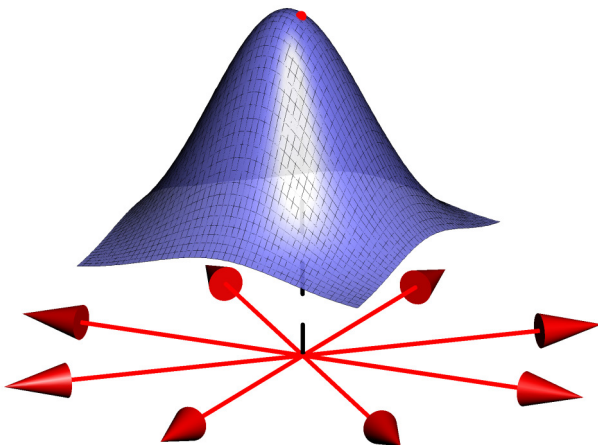
We are given the data set  $(x_1, y_1), \dots, (x_n, y_n)$  where  $x_i \in \mathbb{R}^d$  and  $y_i \in \mathbb{R}$ . We want to fit a linear function to this data by performing empirical risk minimization. More precisely, we are using the hypothesis space  $\mathcal{F} = \{f(x) = w^T x \mid w \in \mathbb{R}^d\}$  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?



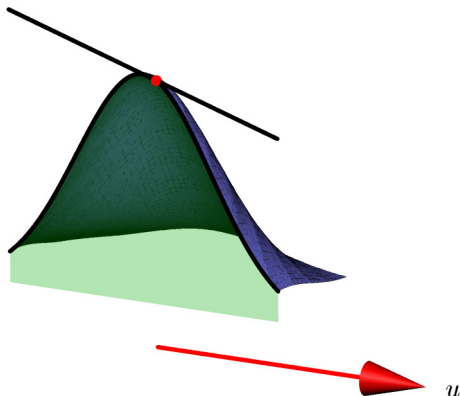
# 1D Linear Approximation By Derivative



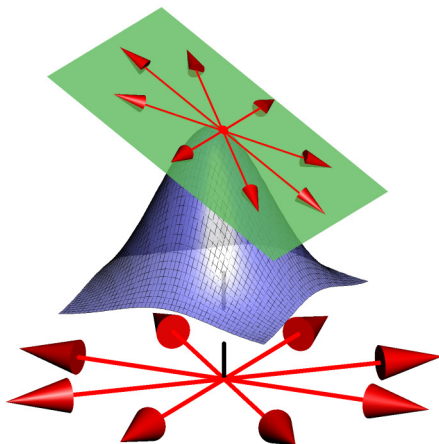
# Multiple Possible Directions for $f : \mathbb{R}^2 \rightarrow \mathbb{R}$



# Directional Derivative as a Slope of a Slice



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



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

