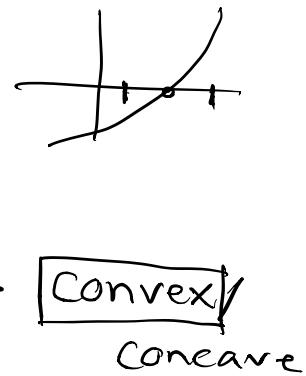


Recap:

root-finding:

- Bisection

- Newton-Raphson
- Secant Method



$f$  is concave     $(-f)$  is convex

Convex optimization

## Agenda

- Taylor Expansion

$\sin(\alpha)$  or  $\sqrt{\alpha}, \dots$

- Reverse Engineering (soft intro)

Curve fitting

## Taylor Expansion

$f(x)$        $f$  is "very" smooth

you can take infinitely many derivatives w/o any issue

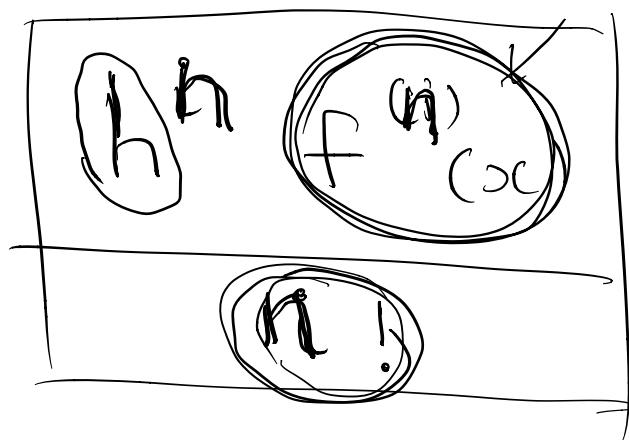
$$f \in C^{\infty}$$

[Science]      vs.      [Art]

$$f(x+h) = f(x) + h \frac{f'(x)}{1!} + h^2 \frac{f''(x)}{2!} + h^3 \frac{f'''(x)}{3!} + h^4 \frac{f^{(4)}(x)}{4!} + \dots$$

expansion around  $x$

$$f(x+h) = \sum_{n=0}^{\infty} \frac{h^n}{n!} f^{(n)}(x)$$



Example  $\boxed{\sin(1.5) = ?}$

$$f(x+h) = f(x) + \dots$$

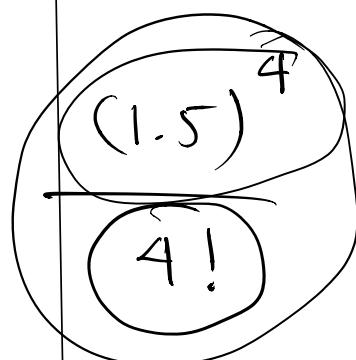
$$\boxed{f^{(n)}(x)}$$

$$x = 0$$

$$x = 1$$

$$h = 1.5$$

$$h = 0.5$$



$$f^{(n)}(x)$$

$$x \rightarrow 0+1.5$$

$$\sin(1.5) = \underline{\sin(0)} + 1.5 \cos(0) + \underline{\frac{(1.5)^2}{2} (-\sin(0))}$$

$$f(x) = \sin x \quad 0$$

$$f'(x) = \cos x \quad 1$$

$$f^{(2)}(x) = -\sin x \quad 0$$

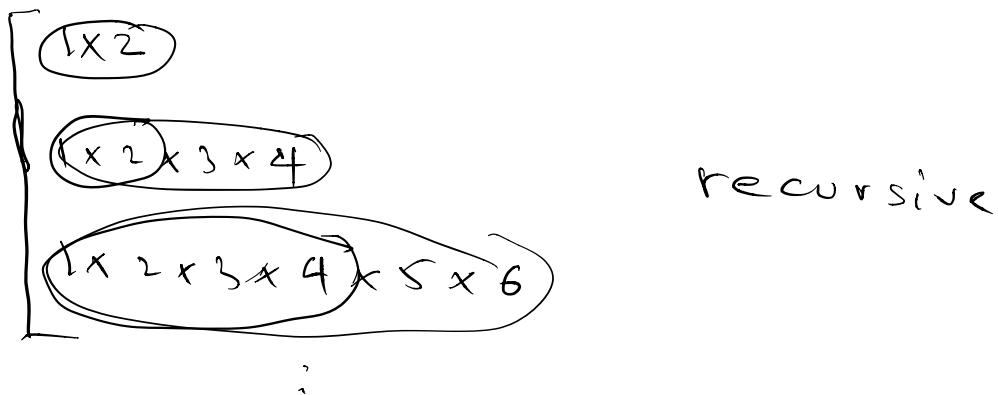
$$f^{(3)}(x) = -\cos x \quad -1$$

$$f^{(4)}(x) = \sin x \quad 0$$

$$-\frac{(1.5)^3}{(3!)} \cos(\alpha) + \cancel{\frac{(1.5)^4}{(4!)}} \sin(\alpha) + \dots$$

+ ...

$$10! = 1 \times 2 \times 3 \times 4 \times \dots \times 10$$



Example 2  $\sqrt{5} = ?$

$$f(x+h) = f(x) + \frac{h}{1!} f'(x) + \frac{h^2}{2!} f''(x)$$

$$+ \frac{h^3}{3!} f^{(3)}(x) + \frac{h^4}{4!} \boxed{f^{(4)}(x)}$$

demon should go much faster than

numer

$$f^{(n)}(x)$$

vs.

HA!

$$f(x) = \sqrt{x} = x^{\frac{1}{2}}$$

$$f'(x) = \frac{1}{2} x^{-\frac{1}{2}} = \frac{1}{2} \cdot \frac{1}{x^{\frac{1}{2}}}$$

pattern!

$$f''(x) = \frac{-1}{4} x^{-\frac{3}{2}} = \frac{-1}{4} \cdot \frac{1}{x^{\frac{3}{2}}} = \frac{-1}{2^2} \cdot \frac{1}{x^{\frac{3}{2}}}$$

$$f^{(3)}(x) = \frac{+3}{8} x^{-\frac{5}{2}} = \frac{+3}{8} \cdot \frac{1}{x^{\frac{5}{2}}} = \frac{3}{2^3} \cdot \frac{1}{x^{\frac{5}{2}}}$$

$$f^{(4)}(x) = \frac{-3 \times 5}{16} x^{-\frac{7}{2}}$$

$$f^{(k)}(x) = \frac{(-1)^{k+1}}{2^k} \cdot \prod_{j=1}^k (2j-3) \cdot \frac{1}{x^{\frac{(2k-1)}{2}}}$$

$$f^{(k)}(x) = \frac{(-1)^{k+1}}{2^k} \cdot \prod_{j=1}^k (2j-3) \cdot \frac{1}{x^{\frac{(2k-1)}{2}}}$$

$k^{\text{th}}$  derivative of  $x^{\frac{1}{2}}$

$$\sum_{j=1}^k j$$

$$1+2+\dots+k$$

$$\prod_{j=1}^k j$$

$$1 \times 2 \times 3 \times \dots \times k$$

$$\frac{1}{\prod_{j=1}^{2k} (2j-3)} \times (-1)^{k+1}$$

2k

$$\sqrt{5} = \sqrt{4} + 1$$

$$\sqrt{2} = \sqrt{4-2}$$

$$\sqrt{3} = \sqrt{4-1}$$

$$\sqrt{4} = \sqrt{4-0}$$

$$\sqrt{5} = \sqrt{4+1}$$

$$\sqrt{5} = \sqrt{1+4}$$

$$\sqrt{5} = \sqrt{1+4}$$

⋮  
⋮  
⋮

$$f^{(k)}(4) = \frac{(-1)^{k+1} \prod_{j=1}^k (2j-3)}{2^k}$$

$$\sqrt{5} = \sqrt{4+1}$$

$$= \sqrt{4}$$

$$+ \left(\frac{1}{1}\right) \times \left(\frac{1}{2}\right) \times \left(\frac{1}{2}\right)$$

$$+ \frac{1^2}{2!} \times \frac{\cancel{1}}{2^2} \times \frac{1}{2^3}$$

$$+ \frac{1^3}{3!} \times \frac{\cancel{1} \times 3}{2^3} \times \frac{1}{2^5}$$

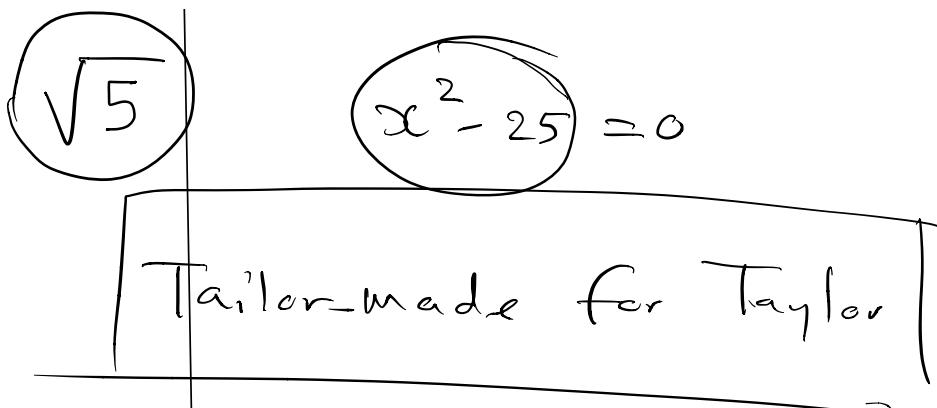
why 4?

$$+ \frac{1^4}{4!} \times \frac{\cancel{1} \times 3 \times 5}{2^4} \times \frac{1}{2^7}$$

$$+ \frac{1^5}{5!} \times \frac{\cancel{1} \times 3 \times 5 \times 7}{2^5} \times \frac{1}{2^9}$$

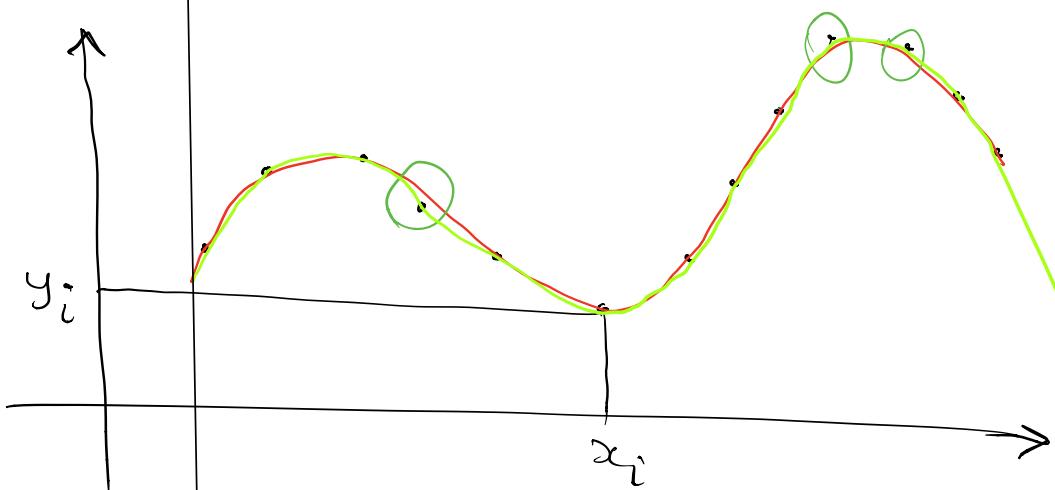
A not

:



## Gentle Introduction to Reverse Engineering

- Curve fitting
- Calibration



$$\{(x_i, y_i)\}_{i=1, 2, \dots, 13}$$

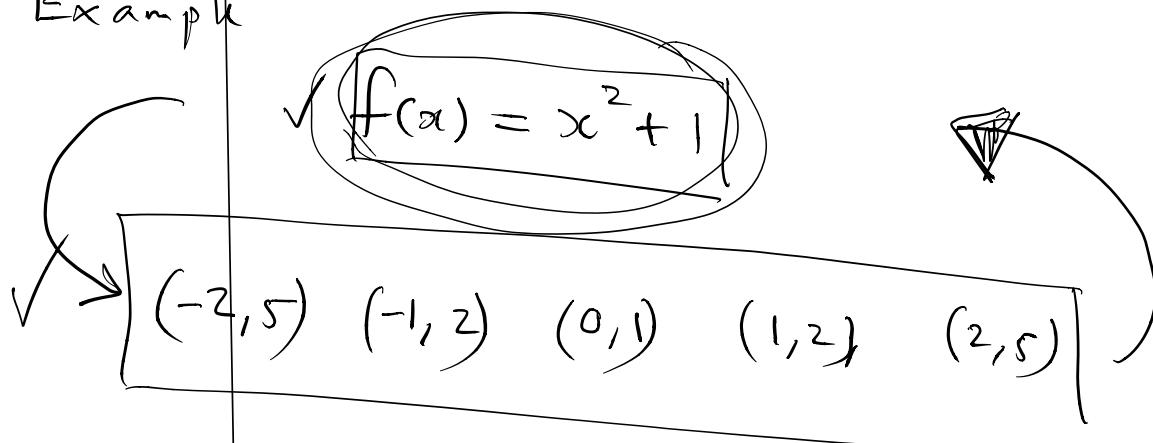
WLOG

$$f(x) = \left| a_0 + a_1 x + a_2 x^2 + \cdots + a_d x^d \right|$$

$a_0, a_1, \dots, a_d$

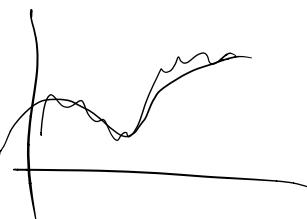
Perfect fit

Example



Real-life

economist-traders



A-prion

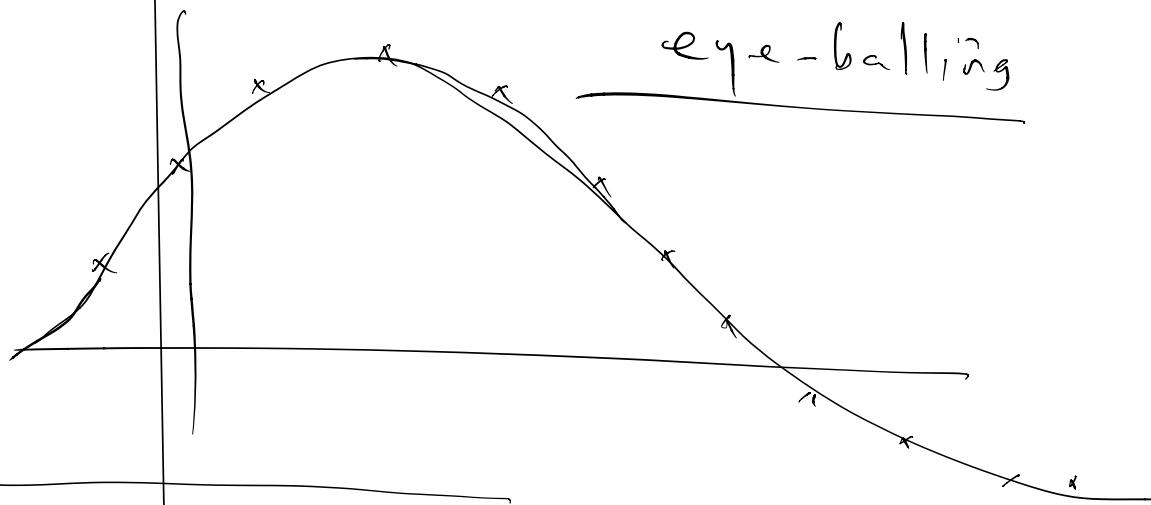
$$f(x) = \left( a_0 \right) + \left( a_3 \right) x^3$$

$a_0, a_3$

$(x_0, y_0), (x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

find  $a_0 \& a_3 ?$

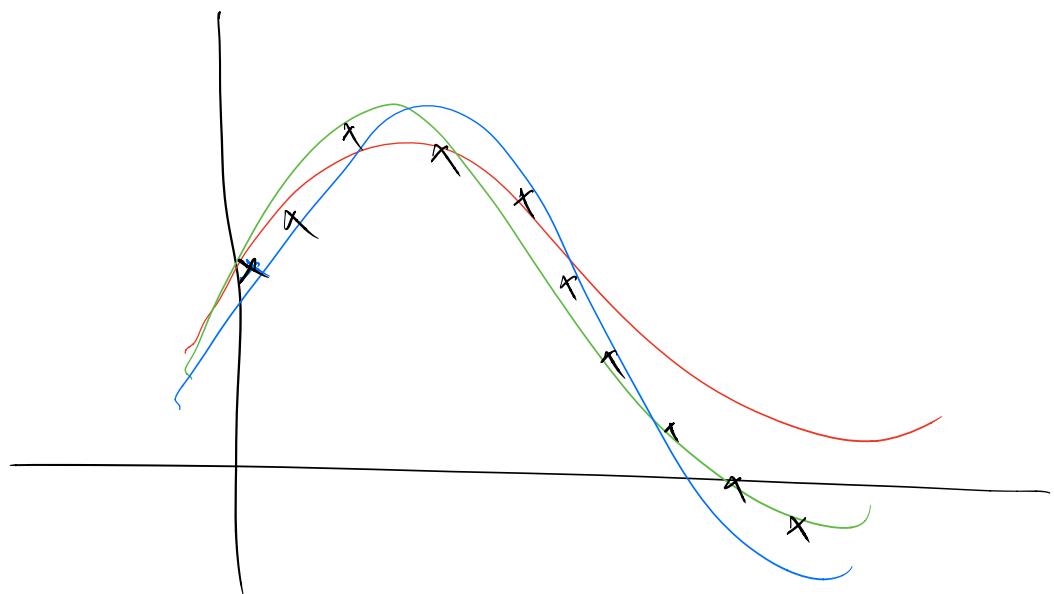
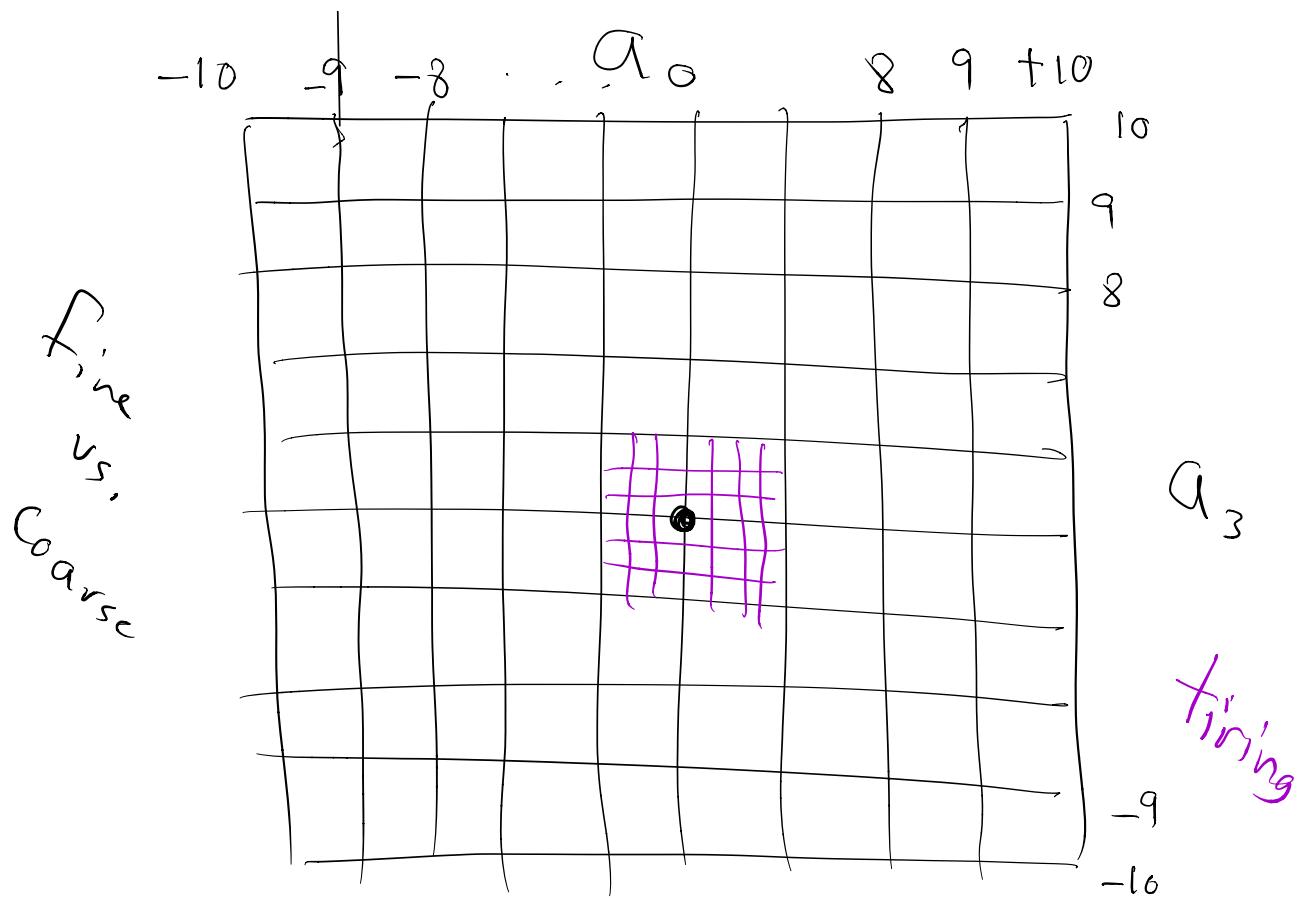
Curse of dimensionality



✓ Grid Search

Brute-Force

$a_0 \& a_3$



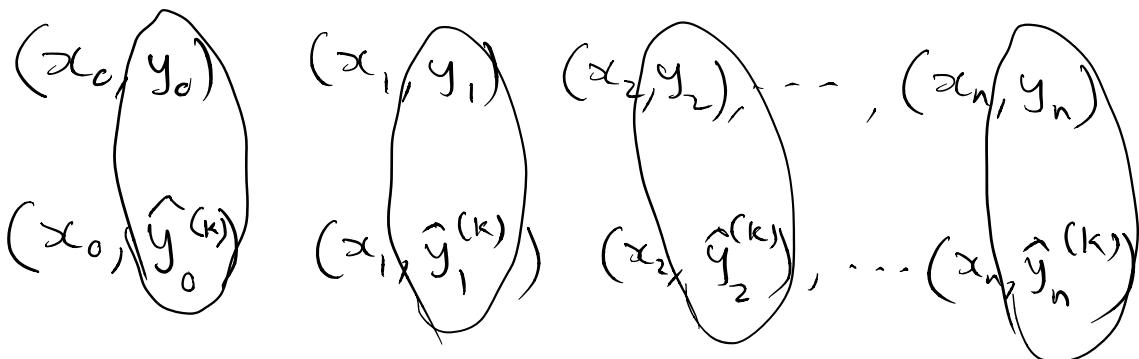
- Measure of goodness  
tightness

... , ... , ...

- Loss function

- Penalty function

$(x_i, y_i)$



distance

$$\|y_0 - \hat{y}_0^{(k)}\| + \|y_1 - \hat{y}_1^{(k)}\| + \dots + \|y_n - \hat{y}_n^{(k)}\|$$

$$(y_0 - \hat{y}_0^{(k)})^2 + (y_1 - \hat{y}_1^{(k)})^2 + \dots + (y_n - \hat{y}_n^{(k)})^2$$

$$L_2 = \sqrt{\sum_{i=0}^n (y_j - \hat{y}_j)^2}$$
 RMS

$L_2 \rightarrow 0$

$$x = (1, 4, 7);$$

$$y = x^T z = (1, 16, 49);$$

$$\begin{array}{r} 17 \\ 49 \\ \hline 66 \end{array}$$

$$\text{sum}(y) = 66$$

Pros.

Bisection

To find some  
starting point!  
Yet could be  
super expensive

Cons.

- Knowing the range a-priori is a MUST.
- Curse of dimensionality

$$a_0 + a_1 x + a_2 x^2 + a_3 x^3 + a_4 x^4 + a_5 x^5$$

$$100 \times 100 = 10,000$$

$$100^6 = 1,000,000,000,000$$

$$(10^6) = 1,000,000$$

Very expensive computationally  
as number of parameters  
grows.

"Curse of Dimensionality"

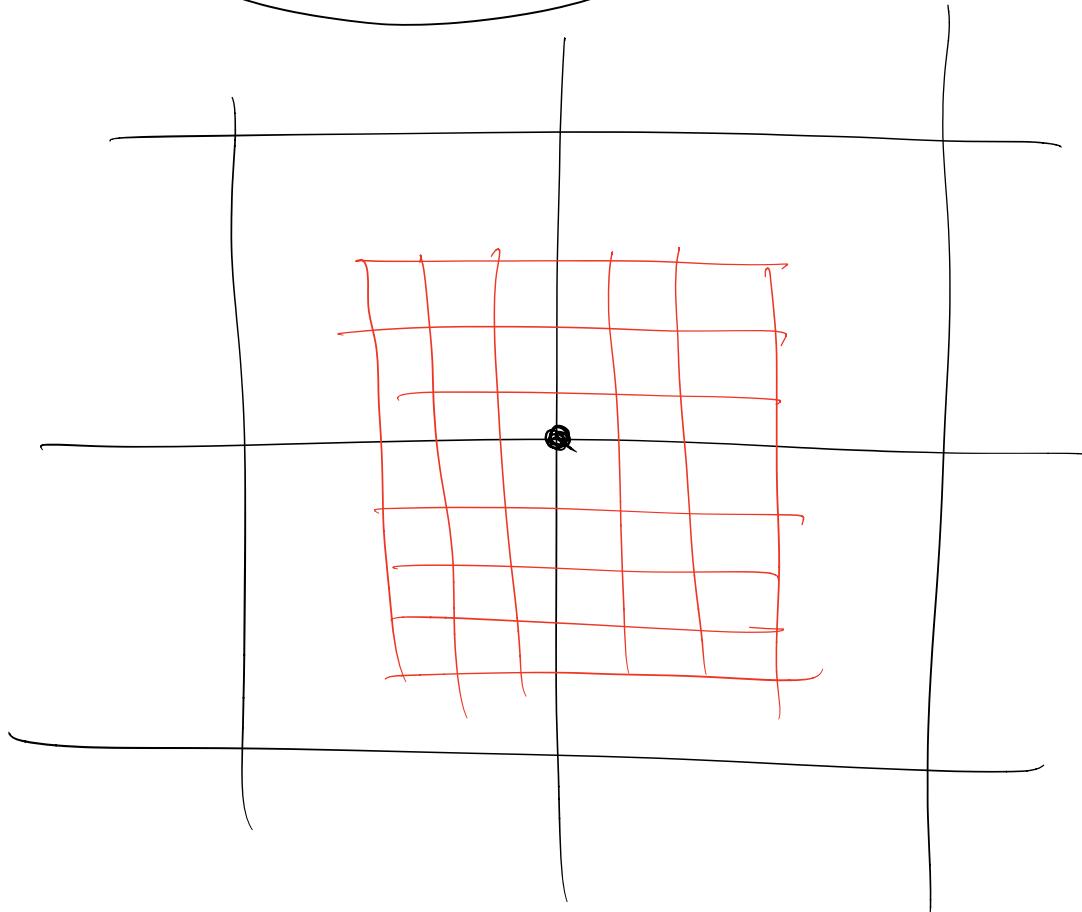
2004

28

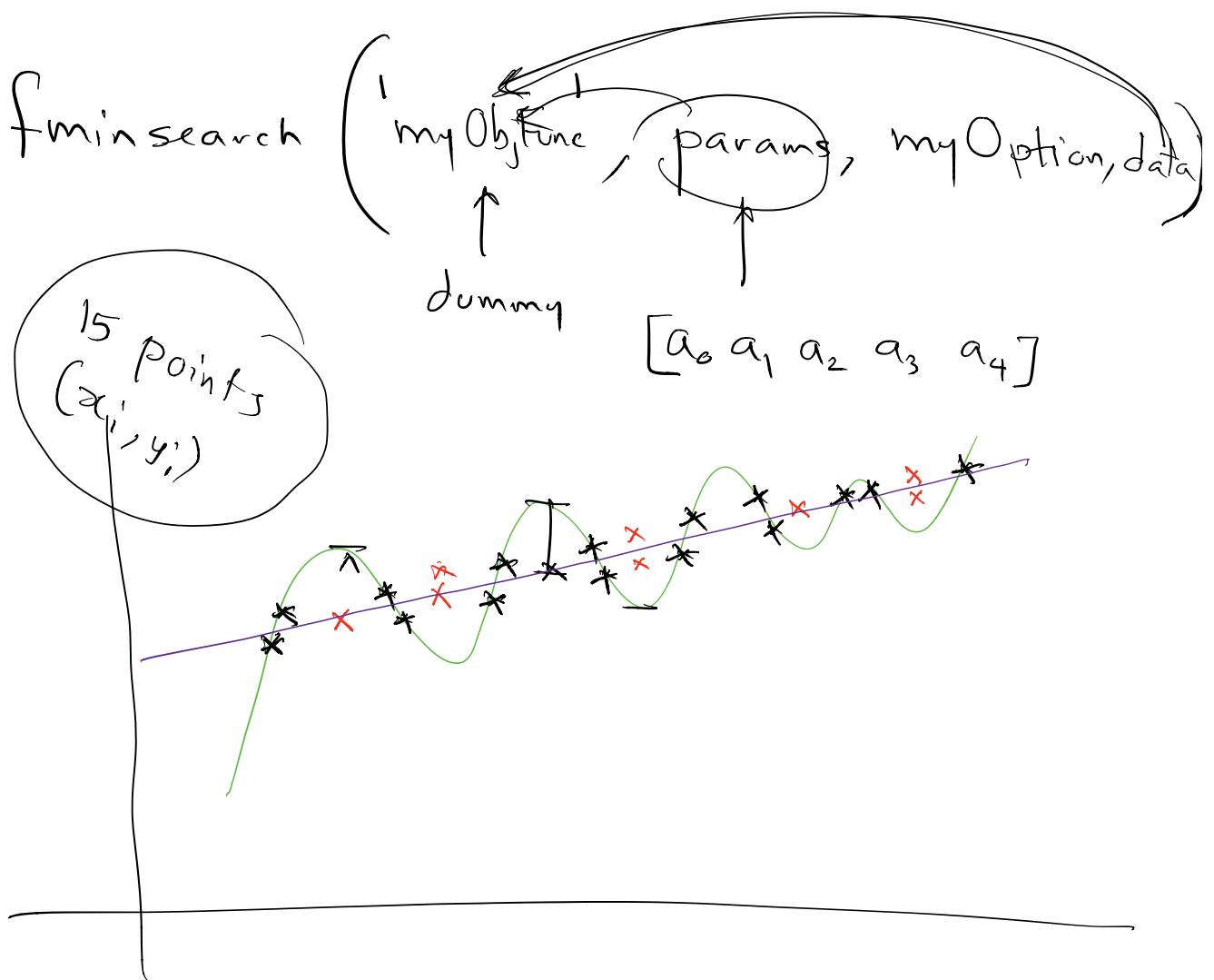
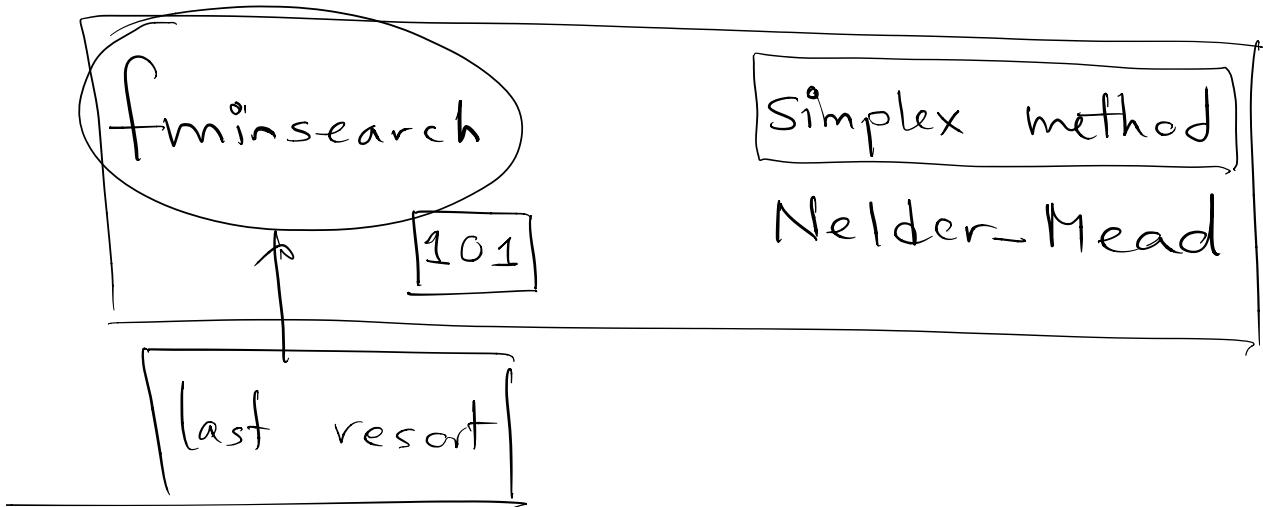
6 Weeks

4<sup>28</sup>

Starting point

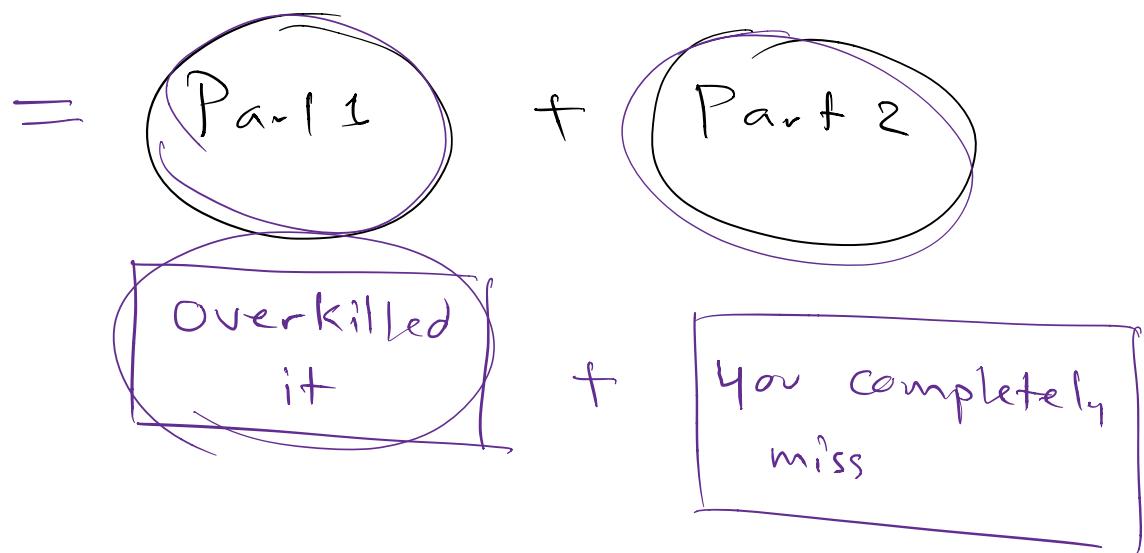


Brute Force  
Grid Search



Over-fitting

data set

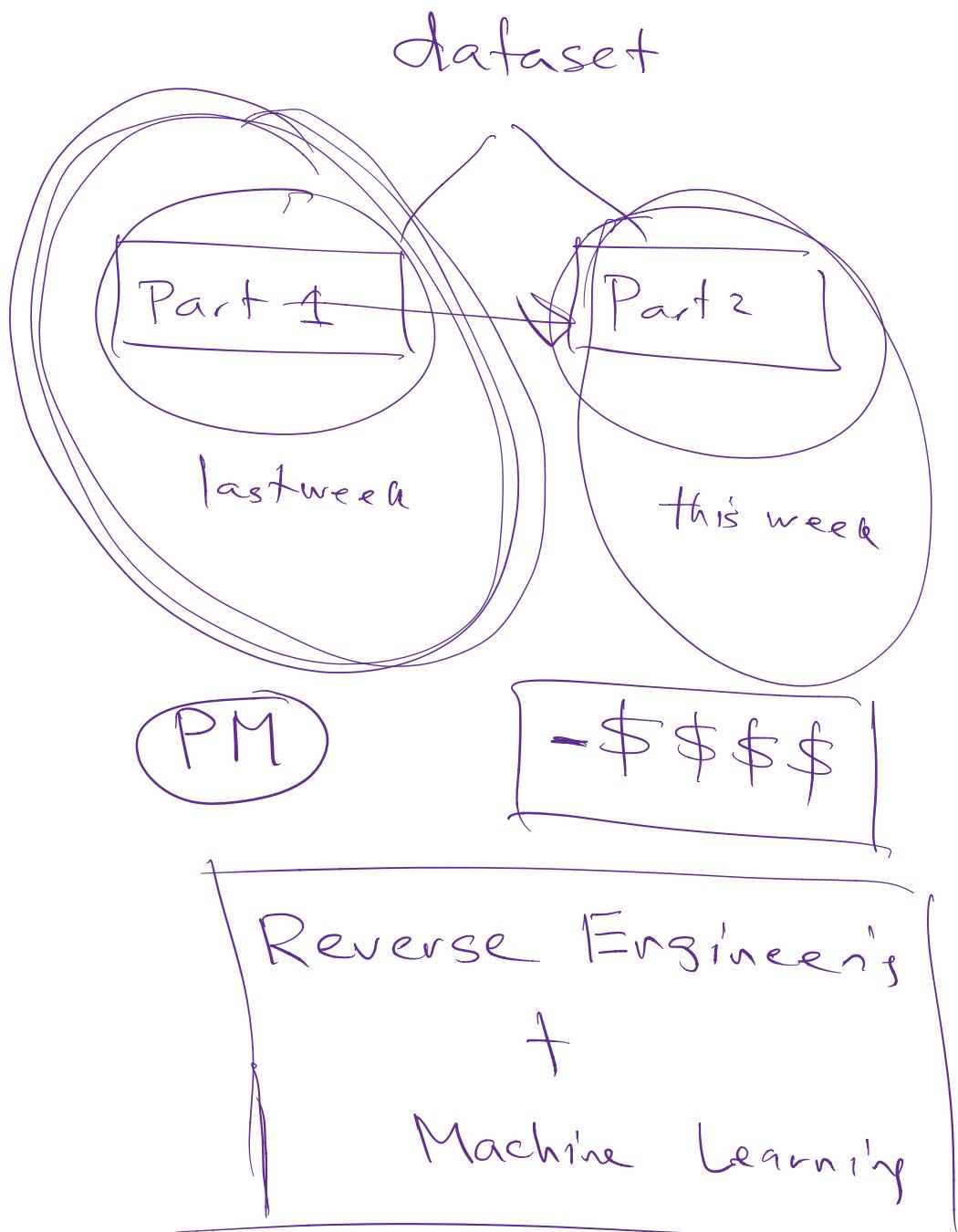


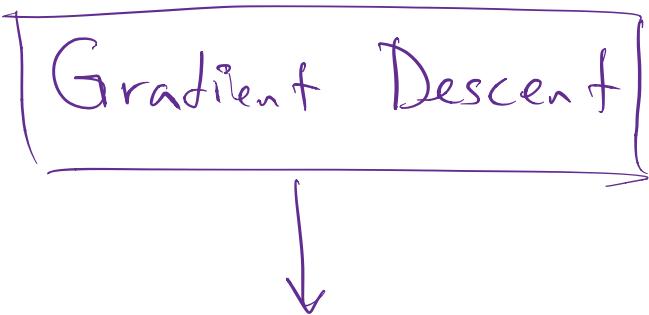
Machine Learning

NLP

The diagram illustrates a machine learning model as a sum of multiple terms. On the left, there is a partial sum  $a_0 + \dots$ . To its right is a plus sign, followed by a large circle containing the term  $a_q x^q$ . Above the circle, there is a single exclamation mark (!) indicating that this term is the focus or result of the entire model.

$$a_0 + \dots + a_q x^q$$





3 lectures

- Batch - Vanilla
- Stochastic
- Mini-Batch