

# CURVE FITTING

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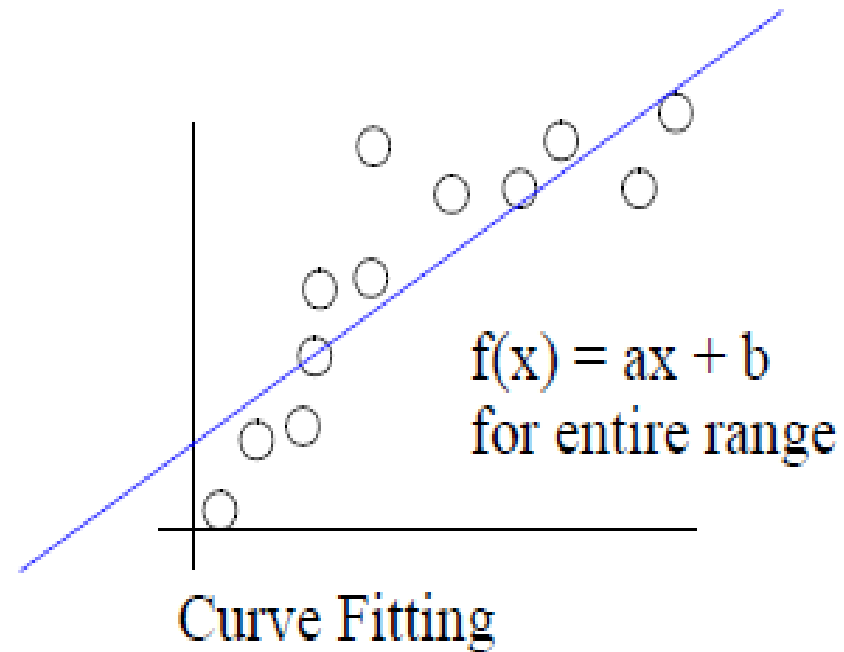
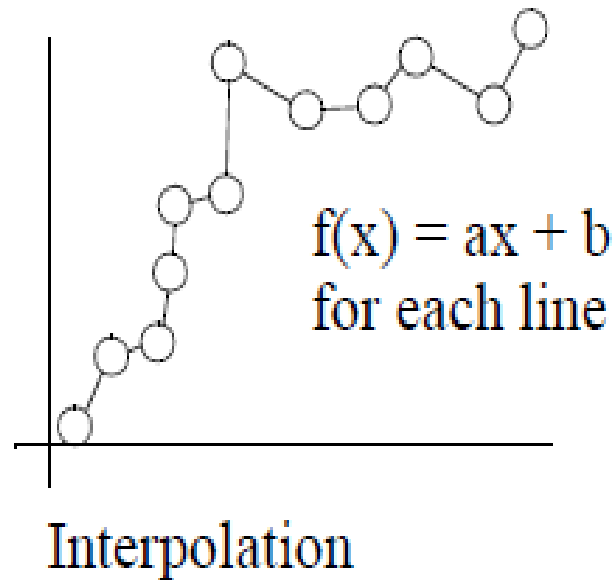
## Introduction

- ❑ Motivation
- ❑ Method of Least Squares
  - ❑ Linear equation  $y=ax+b$
  - ❑ Quadratic equation  $y=a+bx+cx^2$
  - ❑ Exponential  $y=ax^b$  ,  $y=Ce^{ax}$

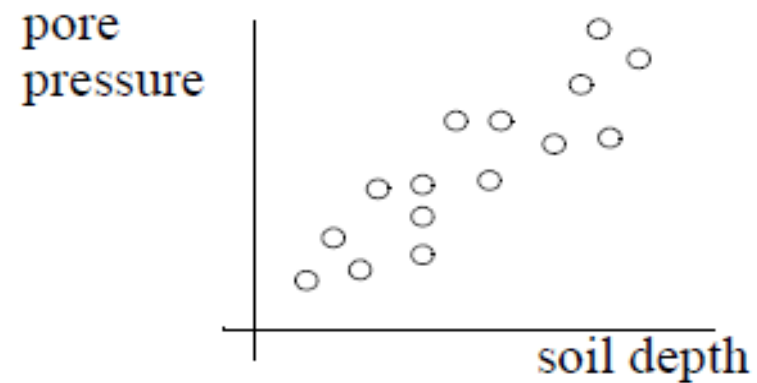
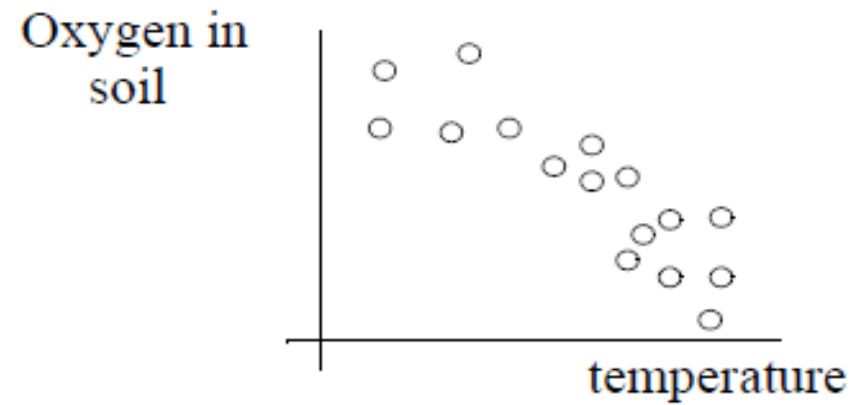
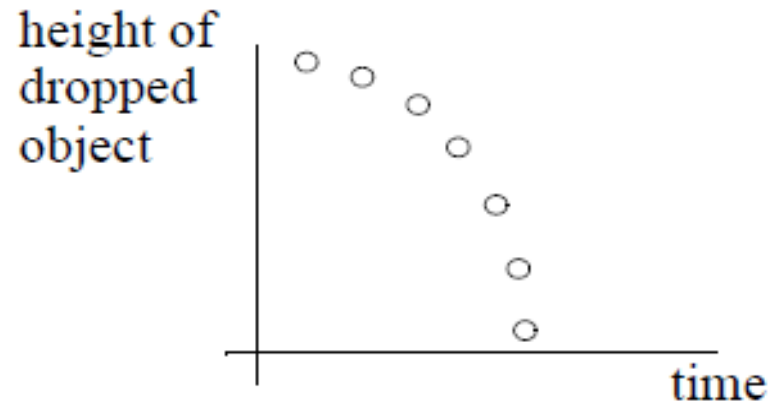
## Curve fitting – Motivation

For root finding, we used a given function to identify where it crossed zero where does ?

**Curve fitting** - capturing the trend in the data by assigning a single function across the entire range. The example below uses a straight line function



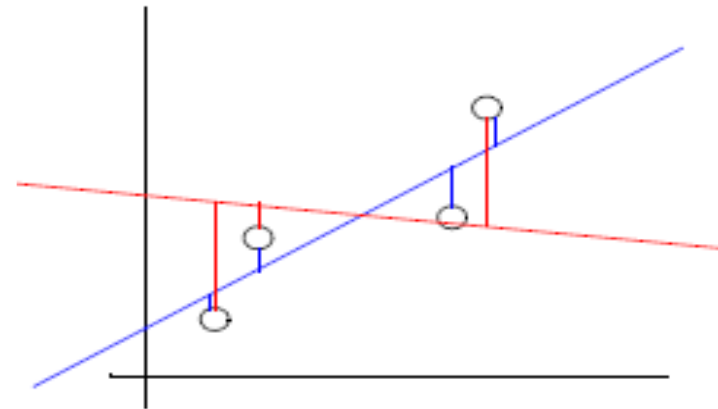
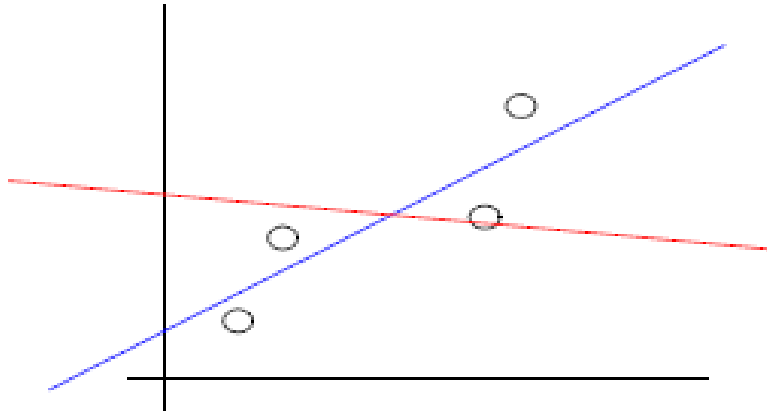
examples of data sets that we can fit a function to



# Linear curve fitting (linear regression)

Given the general form of a straight line

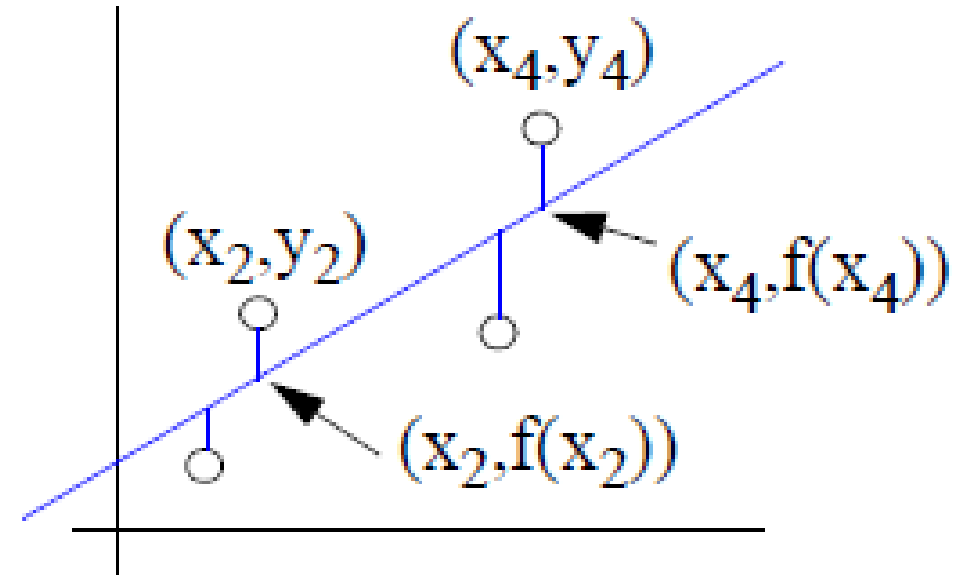
$$f(x) = ax + b$$



# Quantifying errors in a curve fit

## Assumption:

- (1) positive or negative error have the same value  
(data point is above or below the line)
- (2) Weight greater errors more heavily  
we can do both of these things by squaring  
the distance denote data values as  
 $(x, y) \implies$  denote points on the  
fitted line as  $(x, f(x))$  sum the error at  
the four data points



$$\begin{aligned}
 err &= \sum_{i=1}^n d_i^2 = (y_1 - f(x_1))^2 + (y_2 - f(x_2))^2 + \dots (y_n - f(x_n))^2 \\
 &= (y_1 - (ax_1 + b))^2 + (y_2 - (ax_2 + b))^2 + \dots + (y_n - (ax_n + b))^2 \\
 &= \sum_{i=1}^n (y_i - (ax_i + b))^2
 \end{aligned}$$

Error is minimum

$$\frac{\partial(err)}{\partial a} = \sum_{i=1}^n -2x_i (y_i - (ax_i + b)) = 0$$

$$\therefore \sum_{i=1}^n x_i y_i - a \sum_{i=1}^n x_i^2 - b \sum_{i=1}^n x_i = 0$$

$$\therefore \sum_{i=1}^n x_i y_i = a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i$$

$$\frac{\partial(err)}{\partial b} = \sum_{i=1}^n -2(y_i - (ax_i + b)) = 0$$

$$\therefore \sum_{i=1}^n y_i - a \sum_{i=1}^n x_i - b \sum_{i=1}^n 1 = 0$$

$$\therefore \sum_{i=1}^n y_i = a \sum_{i=1}^n x_i + nb$$



Solve the equations

$$\sum_{i=1}^n y_i = a \sum_{i=1}^n x_i + nb \quad (1)$$

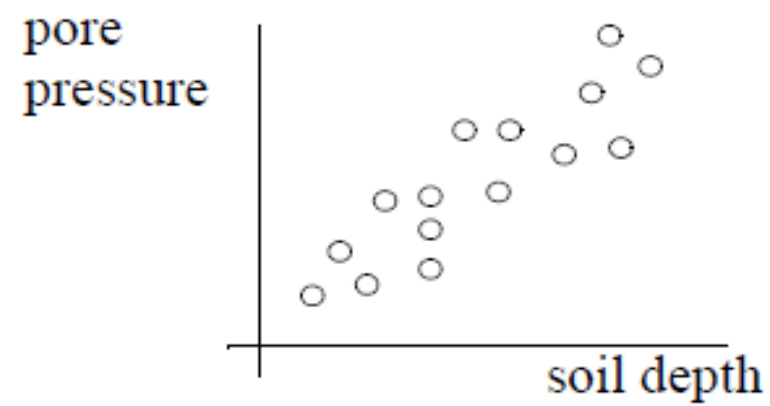
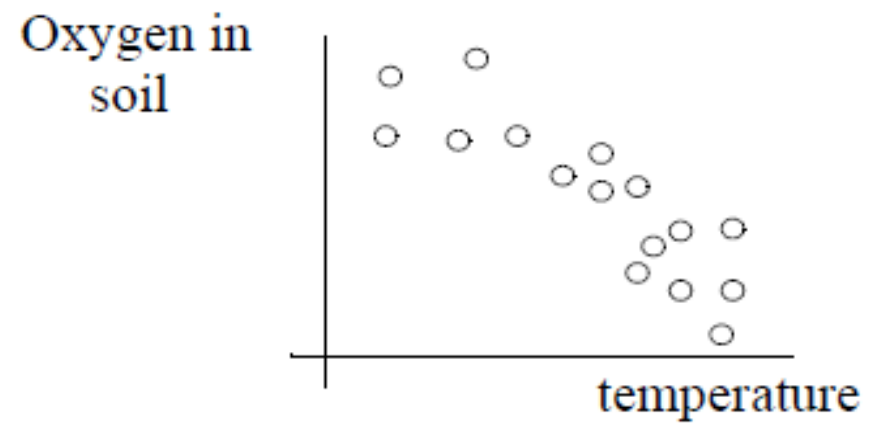
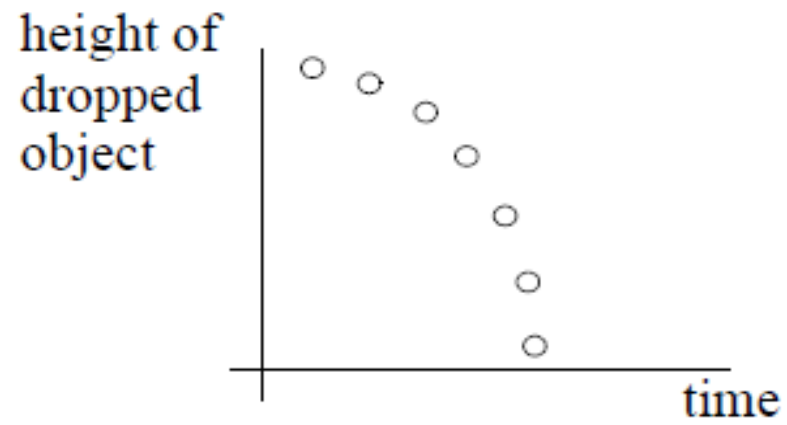
$$\sum_{i=1}^n x_i y_i = a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i \quad (2)$$

# Curve fitting - higher order polynomials

We started the linear curve fit by choosing a generic form of the straight line  $f(x) = ax + b$

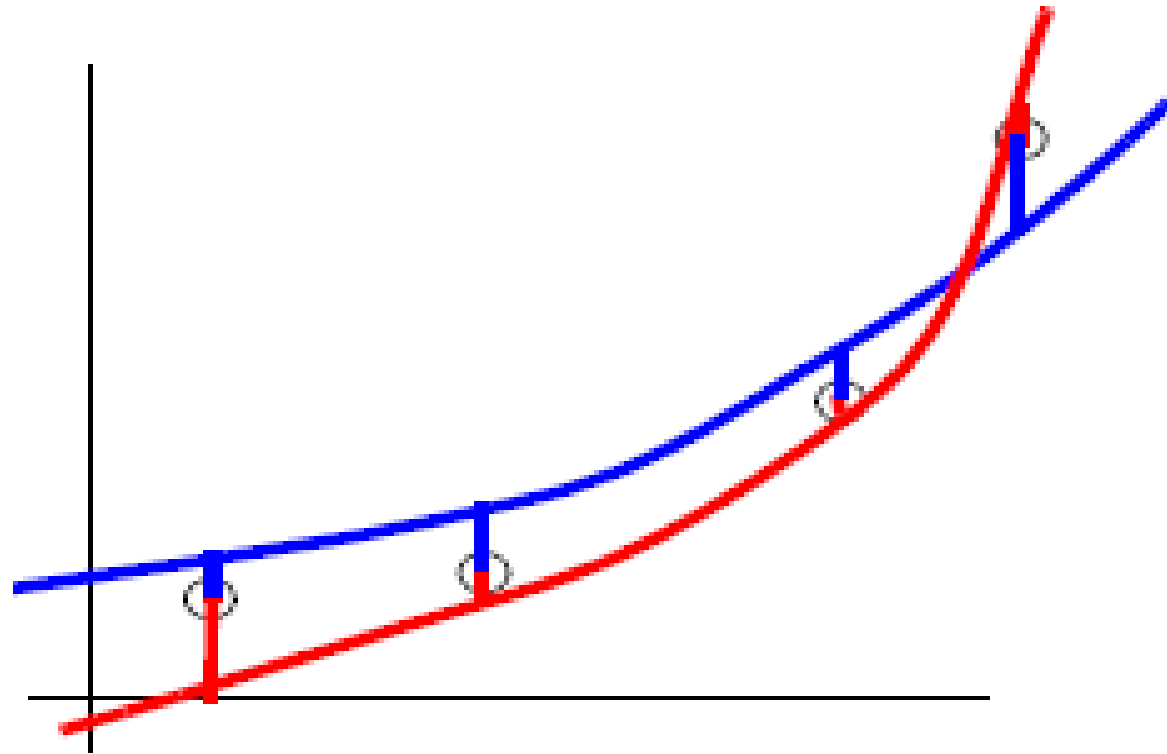
This is just one kind of function. There are an infinite number of generic forms we could choose from for almost any shape we want.

Let's start with a simple extension to the linear regression concept  
recall the examples of sampled data



# Curve fitting – Quadratic polynomial

Let the general form of second order polynomial  $f(x) = a + bx + cx^2$



# Error - Least squares approach

$$\begin{aligned}err &= \sum_{i=1}^n d_i^2 = (y_1 - f(x_1))^2 + (y_2 - f(x_2))^2 + \dots\dots(y_n - f(x_n))^2 \\&= \left(y_1 - \left(a + bx_1 + cx_1^2\right)\right)^2 + \left(y_2 - \left(a + bx_2 + cx_2^2\right)\right)^2 + \dots\dots + \left(y_n - \left(a + bx_n + cx_n^2\right)\right)^2 \\&= \sum_{i=1}^n \left(y_i - \left(a + bx_i + cx_i^2\right)\right)^2\end{aligned}$$

To minimize the error

$$\frac{\partial(err)}{\partial a} = \sum_{i=1}^n -2\left(y_i - \left(a + bx_i + cx_i^2\right)\right) = 0$$

$$\frac{\partial(err)}{\partial b} = \sum_{i=1}^n -2x_i \left(y_i - \left(a + bx_i + cx_i^2\right)\right) = 0$$

$$\frac{\partial(err)}{\partial c} = \sum_{i=1}^n -2x_i^2 \left(y_i - \left(a + bx_i + cx_i^2\right)\right) = 0$$

Simplify these equations, We get

$$\sum_{i=1}^n y_i = a n + b \sum_{i=1}^n x_i + c \sum_{i=1}^n x_i^2$$

$$\sum_{i=1}^n x_i y_i = a \sum_{i=1}^n x_i + b \sum_{i=1}^n x_i^2 + c \sum_{i=1}^n x_i^3$$

$$\sum_{i=1}^n x_i^2 y_i = a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i^3 + c \sum_{i=1}^n x_i^4$$

## Curve fitting - Other nonlinear fits (exponential)

General exponential equation  $f(x) = C e^{Ax}$

Now, take log on both side, we get

$$\ln y = \ln C + Ax$$

$$Y = b + aX; \quad \text{where } Y = \ln y, X = x, \ln C = b \text{ and } a = \ln A$$

Which is equation of line, the original data in xy- plane mapped into XY-plane. This is called *linearization*. The data  $(x, y)$  transformed as  $(x, \ln y)$

To find the value of  $a$  and  $b$  we will use the equations

$$\sum_{i=1}^n Y_i = a \sum_{i=1}^n X_i + nb \quad (1)$$

$$\sum_{i=1}^n X_i Y_i = a \sum_{i=1}^n X_i^2 + b \sum_{i=1}^n X_i \quad (2)$$

$$A = \text{antilog } a, C = \text{antilog } b$$

**Example:** Fit a straight line using least square method

$x_i$	0	0.5	1	1.5	2	2.5
$y_i$	0	1.5	3	4.5	6	7.5

**Solution :**Solve the equations

Substitute the values from the table, here n=6.

$$\sum_{i=1}^n y_i = a \sum_{i=1}^n x_i + nb$$

$$7.5 = 2.5a + 6b$$

$$18.75 = 6.25a + 2.5b$$

$$\sum_{i=1}^n x_i y_i = a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i$$

$$a = 3.561 \text{ and } b = -0.975$$

Hence, the best fit line is  $y = 3.561x - 0.975$



**Example: Fit a second order polynomial equation to following data**

$x_i$	0	0.5	1.0	1.5	2.0	2.5
$y_i$	0	0.25	1.0	2.25	4.0	6.25

$$\sum_{i=1}^n y_i = a n + b \sum_{i=1}^n x_i + c \sum_{i=1}^n x_i^2$$

$$\sum_{i=1}^n x_i y_i = a \sum_{i=1}^n x_i + b \sum_{i=1}^n x_i^2 + c \sum_{i=1}^n x_i^3$$

$$\sum_{i=1}^n x_i^2 y_i = a \sum_{i=1}^n x_i^2 + b \sum_{i=1}^n x_i^3 + c \sum_{i=1}^n x_i^4$$

$y = x^2$  is required equation which fits the data.

**Example: An experiment gave the following values:**

**Fit an exponential curve**  $y = Ce^{Ax}$

$x$	1	5	7	9
$Y$	10	15	12	21