

## NonLinear Regression Model Comparison - with Data linearization vs. without Data linearization

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### Introduction

This worksheet illustrates finding the constants of nonlinear regression models without linearization. Three common nonlinear models are illustrated -

- 1) **Exponential:**  $y = ae^{bx}$
- 2) **Power:**  $y = ax^b$
- 3) **Saturation:**  $y = \frac{(ax)}{(b+x)}$

where  $a$  and  $b$  are constants of the model.

Given  $n$  data points  $(x_1, y_1), (x_2, y_2), (x_3, y_3), \dots, (x_n, y_n)$ , you can best fit one of the nonlinear models to the data. To learn more about nonlinear regression models see the [Nonlinear Regression](#) model worksheet.

### Section 1: Input Data

Below are the input parameters to begin the simulation. This is the only section that requires user input. The user can change the highlighted values and Mathcad will return the coefficients of the nonlinear regression model that are derived with data linearization and without linearization.

**NOTE:** Before evaluating the worksheet, the user must enter initial guesses of the coefficients of the model  $a$  and  $b$  for the nonlinear regression without data linearization procedure. For reasonable initial guesses, use the solution from the [Nonlinear model with data linearization](#) worksheet. For convergence use initial guesses for  $a$  and  $b$  close to the values of  $a$  and  $b$  obtained by using data linearization.

**NOTE:** The origin has been set to 1 to redefine that starting index of all arrays. The user SHOULD NOT change this value.

ORIGIN := 1

### Input Parameters:

- Array of  $x$  values,  $\mathbf{X}$

$$\mathbf{X} := \begin{pmatrix} 10 \\ 16 \\ 25 \\ 40 \\ 60 \end{pmatrix}$$

- Array of  $y$  values,  $\mathbf{Y}$

$$\mathbf{Y} := \begin{pmatrix} 94 \\ 118 \\ 147 \\ 180 \\ 230 \end{pmatrix}$$

- For *exponential* model call model type to be "**Exponential**"  
For *power* model assign the model type variable as "**Power**"  
For *saturation growth* model, assign the model type variable to be "**Growth**"

ModelType := "Exponential"

- Insert your initial guesses for  $a$  and  $b$  here. Reasonable initial guesses for  $a$  and  $b$  be obtained from data linearization models.

Initial Guess values for  $a$

Ainit := 86.9

Initial Guess values for  $b$

Binit := 0.0159

**NOTE:** In Section 3, *Solve Block* is used to find the constants of the regression model without data linearization. You may need to use a different method to solve if you are unable to get a solution. Change the algorithm by right clicking on the *Find* command and checking a different nonlinear method in the popup menu.

- Number of data points,  $n$

$n := \text{rows}(X)$

$n = 5$

## Section 2: Nonlinear model with data linearization procedure

In the procedure below, the data is first linearized so that least squares regression method for a linear model can be used. Once the coefficients of the linear model are determined, the constants of the nonlinear regression model  $a$  and  $b$  can be calculated.

Linearizing the data is a useful technique to estimate the parameters of a nonlinear model because it does not require iterative methods to solve for the model constants. Note that data linearization is only done for mathematical convenience. For details, see the [Nonlinear Regression](#) worksheet.

```

datelinearized(x, y, n) :=
  z ← 0
  q ← 0
  sumq ← 0
  sumz ← 0
  sumqz ← 0
  sumqq ← 0
  for i ∈ 1 .. n
    if ModelType = "Exponential"
      zi ← ln(yi)
      qi ← xi
    if ModelType = "Power"
      zi ← ln(yi)
      qi ← ln(xi)
    if ModelType = "Growth"
      zi ← 1/yi
      qi ← 1/xi
    sumq ← sumq + qi
    sumz ← sumz + zi
    sumqz ← sumqz + qi·zi
    sumqq ← sumqq + (qi)2
  n·sumqz - sumq·sumz

```

Linearizing the data and making substitutions

Calculating the constants  $a_0$  and  $a_1$  of linear model  $z = a_0 + a_1q$

$$\begin{aligned}
 a1 &\leftarrow \frac{\sum z^2}{n \cdot \sum q^2 - \sum q^2} \\
 a0 &\leftarrow \frac{\sum z}{n} - a1 \cdot \frac{\sum q}{n} \\
 \text{if } \text{ModelType} = \text{"Exponential"} & \\
 \quad \left| \begin{array}{l} a \leftarrow \exp(a0) \\ b \leftarrow a1 \end{array} \right. & \\
 \text{if } \text{ModelType} = \text{"Power"} & \\
 \quad \left| \begin{array}{l} a \leftarrow \exp(a0) \\ b \leftarrow a1 \end{array} \right. & \\
 \text{if } \text{ModelType} = \text{"Growth"} & \\
 \quad \left| \begin{array}{l} a \leftarrow \frac{1}{a0} \\ b \leftarrow \frac{a1}{a0} \end{array} \right. & \\
 \left( \begin{array}{l} a \\ b \end{array} \right) &
 \end{aligned}$$

Using substitutions of linear model constants to find the constants of the original nonlinear model,  $a$  and  $b$ .

Calling the data linearization procedure for the input data set and returning the constants of the nonlinear regression model  $a$  and  $b$ .

$$a := \text{datalinearized}(X, Y, n)_1$$

$$b := \text{datalinearized}(X, Y, n)_2$$

$$g(x) := \begin{cases} a \cdot \exp(b \cdot x) & \text{if } \text{ModelType} = \text{"Exponential"} \\ a \cdot x^b & \text{if } \text{ModelType} = \text{"Power"} \\ \frac{a \cdot x}{b + x} & \text{if } \text{ModelType} = \text{"Growth"} \end{cases}$$

The constants of the  $\text{ModelType} = \text{"Exponential"}$  nonlinear model found with data linearization are:

$$a = 87.8045128$$

$$b = 0.0169529$$

### Section 3: Nonlinear model without data linearization procedure

In the following procedure, the constants of the nonlinear regression model  $a$  and  $b$  are found without linearizing the data. This requires the use of a **Solve Block** which utilizes numerical techniques to converge to a real solution. In this section, the initial guess in from Section 1,  $A_{init}$  and  $B_{init}$ , are used as the starting values.

Assigning the proper regression model:

$$f(x, aa, bb) := \begin{cases} aa \cdot \exp(bb \cdot x) & \text{if ModelType} = \text{"Exponential"} \\ aa \cdot x^{bb} & \text{if ModelType} = \text{"Power"} \\ \frac{aa \cdot x}{bb + x} & \text{if ModelType} = \text{"Growth"} \end{cases}$$

Calculating the sum of the square of the residuals,  $S_r$ :

$$S_r(x, aa, bb, y, n) := \begin{cases} S_r \leftarrow 0 \\ \text{for } i \in 1..n \\ S_r \leftarrow S_r + (y_i - f(x_i, aa, bb))^2 \\ S_r \end{cases}$$

Differentiating  $S_r$  with respect to the constants of the model  $a$  and  $b$  to setup two simultaneous nonlinear equations and two unknowns:

$$q(aa, bb) := \frac{d}{daa} S_r(X, aa, bb, Y, n)$$

$$r(aa, bb) := \frac{d}{dbb} S_r(X, aa, bb, Y, n)$$

Solving the two simultaneous nonlinear equations. We are using a *Solve Block* since we are looking for real solutions.

**NOTE:** If the solution is not close to the values obtained by data linearization, change the algorithm that is used by right clicking on the *Find* command and checking a different nonlinear method in the popup menu.

Guess

$$aa := A_{init}$$

$$bb := B_{init}$$

Given

$$q(aa, bb) = 0$$

$$r(aa, bb) = 0$$

constants\_of\_model := Find(aa, bb)

$$\underline{aa} := \text{constants\_of\_model}_1$$

$$\underline{bb} := \text{constants\_of\_model}_2$$

The constants of the ModelType = "Exponential" model without data linearization are

$$aa = 92.3399879$$

$$bb = 0.0155851$$

**Data Table:**

$$\text{data\_compared} := \begin{pmatrix} a & b \\ aa & bb \end{pmatrix}$$

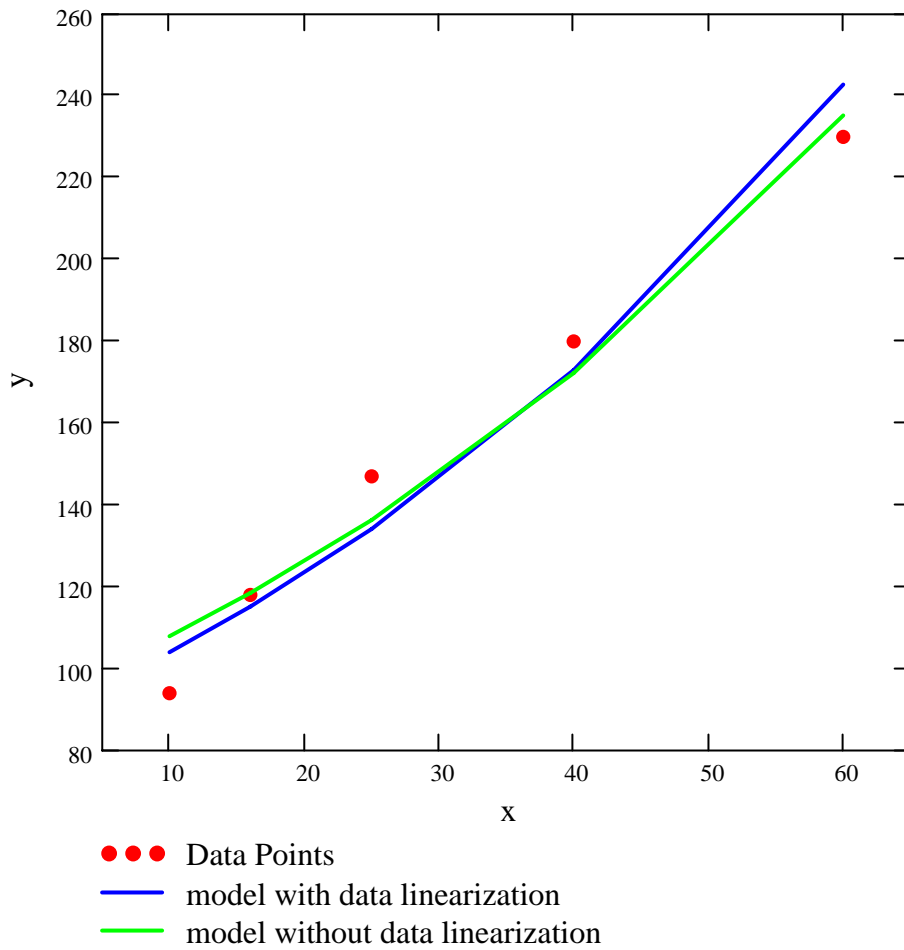
	<b>a</b>		<b>b</b>	
	1	2	1	2
data_compared =	1	87.8045128	0.0169529	0.0155851
	2	92.3399879	0.0155851	0.0169529

**data linearization**  
**without data linearization**

**Section 4: Plot of nonlinear model with data linearization and without data linearization**

Plotting the observed values and both predicted curves.

**Regression model with data linearization and data without linearization, y vs x**



## References

[1]Autar Kaw, *Holistic Numerical Methods Institute*,  
<http://numericalmethods.eng.usf.edu>, See  
[Nonlinear Regression](#)

## Conclusion

Mathcad helped us to compare a nonlinear regression model that was determined by linearizing the data to one that was found without linearizing the data.

Question 1: A functional relationship,  $\rho = k_1 \exp(-k_2 h)$  exists between the mass density of air and the altitude above sea level,  $h$ .

$$i := 1..6$$

Altitude above sea level (km)      Mass Density of air ( $kg/m^3$ )

$h_i :=$	$\rho_i :=$
0.32	1.15
0.5	1.12
0.64	1.10
1.28	1.05
1.35	1.03
1.60	0.95

- Find the constants of the model,  $k_1$  and  $k_2$  using data linearization.
- Find the constants of the model,  $k_1$  and  $k_2$  without data linearization.
- Compare the two models from (a) and (b) on a plot.
- Find the sum of the square of the residuals for both models using the original model  $\rho = k_1 \exp(-k_2 h)$ . Which one is smaller? Is that what you expected?

Question 2: Theoretical considerations assume that the rate of flow,  $Q$  from a fire hose is proportional to some power of the nozzle pressure,  $p$  as given by  $Q = ap^b$ . Determine whether this conjecture is true.



$$i := 1..7$$

Flow rate (*gallons/min*)

Pressure (*psi*)

$Q_i :=$

$p_i :=$

91
120
127
190
240
310
409

9
15
23
40
61
72
90

- Find the constants,  $a$  and  $b$  of the model using data linearization.
  - Find the constants,  $a$  and  $b$  of the model without data linearization.
  - Compare the two models on a plot.
  - Find the sum of the square of the residuals for both the models using the original  $n$   $Q = ap^b$ . Which one is smaller? Is that what you expected?
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