

# Nonlinear Regression

Lecture 51  
Section 7.4

Robb T. Koether

Hampden-Sydney College

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# Objectives

## Objectives

- Derive other least-squares equations.
- Apply the models to the DJIA.

# Quadratic Regression

## Quadratic Regression

**Quadratic regression** requires finding a quadratic function

$$y = ax^2 + bx + c$$

that minimizes the sum of the squared deviations

$$\sum_{i=1}^n ((a + bx_i + cx_i^2) - y_i)^2.$$

The method is the same, but the results are far more complicated.

# Quadratic Regression

## Quadratic Regression

- The equations to be solved for  $a$ ,  $b$ , and  $c$  are

$$na + b \sum x + c \sum x^2 = \sum y,$$

$$a \sum x + b \sum x^2 + c \sum x^3 = \sum xy,$$

$$a \sum x^2 + b \sum x^3 + c \sum x^4 = \sum x^2y.$$

# Quadratic Regression

## Quadratic Regression

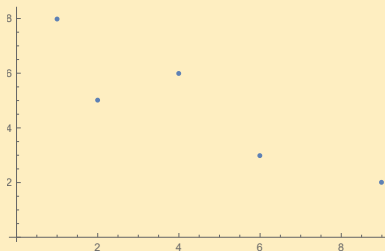
$$a = \frac{(\sum x^2)^2 \sum x^2 y - \sum x \sum x^2 y \sum x^3 + \sum x \sum x^4 \sum xy + (\sum x^2)^2 \sum xy - \sum x^2 \sum x^3 \sum x^2 y + \sum x \sum x^4 \sum y}{(\sum x^2)^3 - 2 \sum x \sum x^2 \sum x^3 + n(\sum x^3)^2 + (\sum x)^2 \sum x^4 - n \sum x^2 \sum x^4},$$
$$b = \frac{n \sum x^2 y \sum x^3 - \sum x \sum x^2 \sum x^2 y + (\sum x^2)^2 \sum xy + \sum x \sum x^4 \sum y - n \sum x^4 \sum xy - \sum x^2 \sum x^3 \sum y}{(\sum x^2)^3 - 2 \sum x \sum x^2 \sum x^3 + n(\sum x^3)^2 + (\sum x)^2 \sum x^4 - n \sum x^2 \sum x^4},$$
$$c = \frac{(\sum x)^2 \sum x^2 y - n \sum x^2 \sum x^2 y - \sum x \sum x^2 \sum xy + n \sum x^3 \sum xy + (\sum x^2)^2 \sum y - \sum x \sum x^3 \sum y}{(\sum x^2)^3 - 2 \sum x \sum x^2 \sum x^3 + n(\sum x^3)^2 + (\sum x)^2 \sum x^4 - n \sum x^2 \sum x^4}.$$

# Quadratic Regression

## Example

Find the least-squares regression quadratic function for the data

$$(1, 8), (2, 5), (4, 6), (6, 3), (9, 2).$$

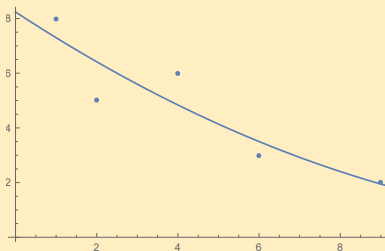


# Quadratic Regression

## Example

Find the least-squares regression quadratic function for the data

$(1, 8), (2, 5), (4, 6), (6, 3), (9, 2)$ .



$$y = 0.0302x^2 - 0.9709x + 8.2387$$

# Logarithmic Regression

## Logarithmic Regression

**Logarithmic regression** requires finding a logarithmic function

$$y = a + b \ln x$$

that minimizes the sum of the squared deviations

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



# Logarithmic Regression

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

This is the same as linear regression with  $x$  replaced by  $\ln x$ .

# Logarithmic Regression

## Logarithmic Regression

The equations to be solved are

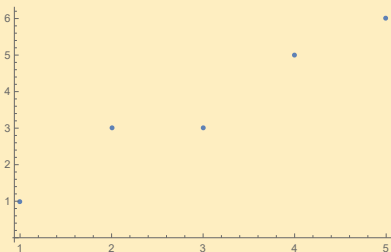
$$na + b \sum \ln x_i = \sum y_i,$$
$$a \sum \ln x_i + b \sum (\ln x_i)^2 = \sum (\ln x_i) y_i.$$

# Logarithmic Regression

## Example

Find the least-squares regression logarithmic function for the data

$$(1, 1), (2, 3), (3, 3), (4, 5), (5, 6).$$

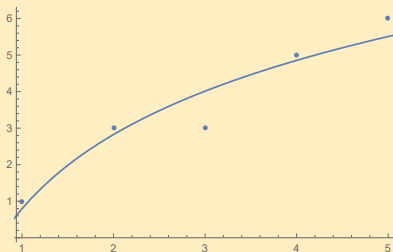


# Logarithmic Regression

## Example

Find the least-squares regression logarithmic function for the data

$(1, 1), (2, 3), (3, 3), (4, 5), (5, 6)$ .



$$y = 0.797 + 2.927 \ln x$$

# Power Regression

## Power Regression

**Power regression** requires finding a logarithmic function

$$y = ax^b$$

that minimizes the sum of the squared deviations

$$\sum_{i=1}^n (ax_i^b - y_i)^2.$$

# Power Regression

## Power Regression

The trick is to apply logarithms and convert the equation into

$$\ln y = \ln a + b \ln x$$

and use logarithmic regression.

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Again, this is linear regression

$$Y = A + bX$$

with  $y$  replaced by  $\ln y$ ,  $x$  replaced by  $\ln x$ , and  $a$  replaced by  $\ln a$ .

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The solution gives  $A$  for  $\ln a$ , so  $a = e^A$ .

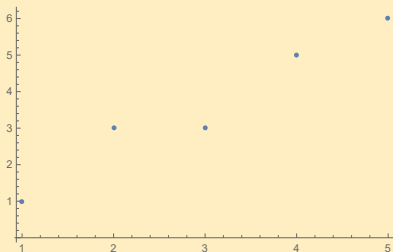


# Power Regression

## Example

Find the least-squares regression power function for the data

$$(1, 1), (2, 3), (3, 3), (4, 5), (5, 6).$$

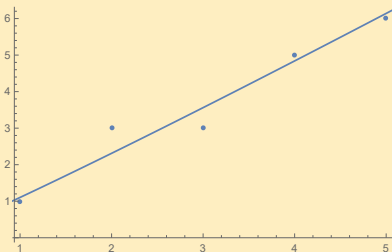


# Power Regression

## Example

Find the least-squares regression power function for the data

$(1, 1), (2, 3), (3, 3), (4, 5), (5, 6)$ .



$$y = 1.1036x^{1.0665}$$

# Exponential Regression

## Exponential Regression

**Exponential regression** requires finding a logarithmic function

$$y = ab^x$$

that minimizes the sum of the squared deviations

$$\sum_{i=1}^n (ab^{x_i} - y_i)^2.$$

# Exponential Regression

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Once again, this is linear regression

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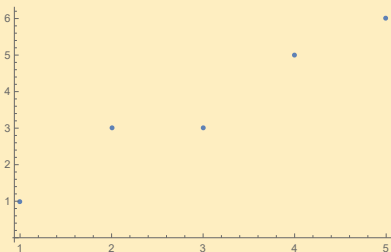
with  $x$  replaced by  $\ln x$ ,  $a$  replaced by  $\ln a$ , and  $b$  replaced by  $\ln b$ .  
The solution gives  $A$  for  $\ln a$  and  $B$  for  $\ln b$ , so  $a = e^A$  and  $b = e^B$ .

# Exponential Regression

## Example

Find the least-squares regression exponential function for the data

$$(1, 1), (2, 3), (3, 3), (4, 5), (5, 6).$$

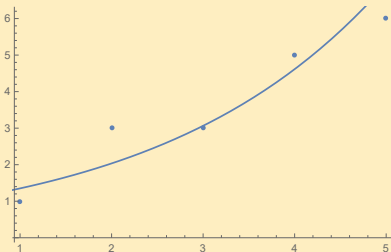


# Exponential Regression

## Example

Find the least-squares regression exponential function for the data

$(1, 1), (2, 3), (3, 3), (4, 5), (5, 6)$ .



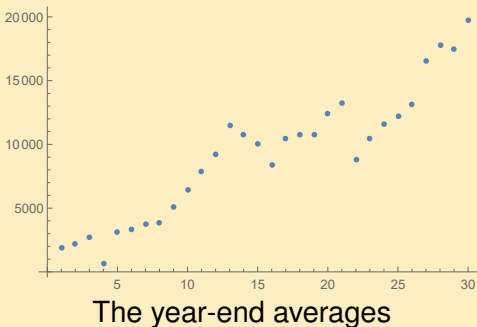
$$y = 0.897(1.506^x)$$



# Example

## DJIA and Exponential Growth

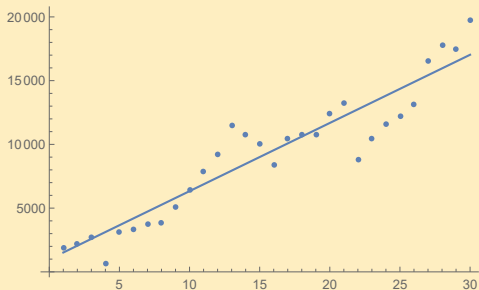
Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years



# Example

## DJIA and Exponential Growth

Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years

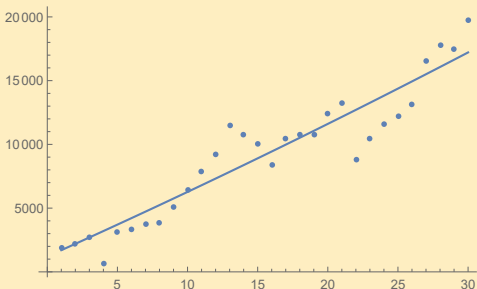


Linear model:  $y = 983.7 + 534.87x$

# Example

## DJIA and Exponential Growth

Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years

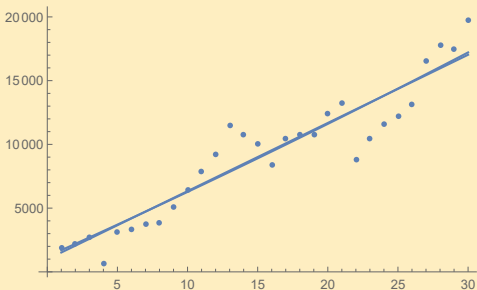


$$\text{Quadratic model: } y = 1.35x^2 + 493.0x + 1206.9$$

# Example

## DJIA and Exponential Growth

Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years

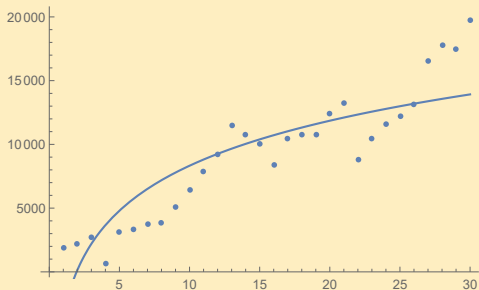


Linear model vs. quadratic model

# Example

## DJIA and Exponential Growth

Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years

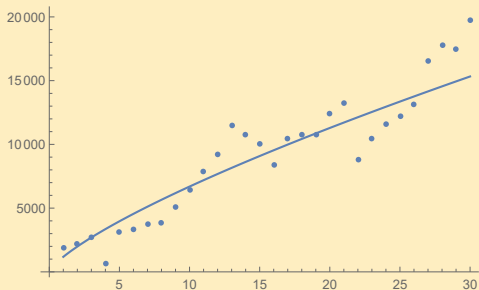


Logarithmic model:  $y = -3395.8 + 5091.2 \ln x$

# Example

## DJIA and Exponential Growth

Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years

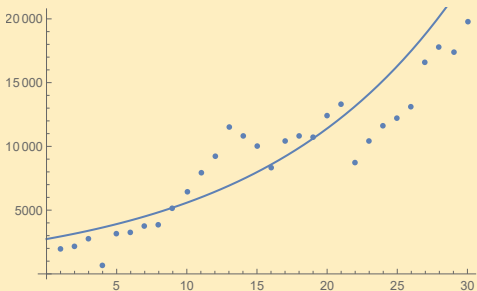


Power model:  $y = 1178.1x^{0.7543}$

# Example

## DJIA and Exponential Growth

Fit the various models to the Dow-Jones Industrials year-end average for the last 30 years



Exponential model:  $y = 2734.3(1.074^x)$