

Least-Squares Regression, General linear least Squares, Nonlinear Regression [1-6]

References:

1. Chapra S.C. and Canale R.P. “Numerical Methods for Engineers”, Sixth Edition, McGraw Hill, International Edition 2010.
2. Chapra S.C. and Canale R. P. “Yazılım ve programlama Uygulamalarıyla Mühendisler için Sayısal Yöntemler” 4.Basımdan Çevirenler: Hasan Heperkan ve Uğur Kesgin 2003.
3. Chapra S.C. “Applied Numerical Methods with MATLAB for engineers and Scientists” Third Edition, McGraw Hill, International Edition 2012.
4. Mathews J.H. and Fink K.D. “Numerical Methods using MATLAB”, Fourth Edition, Pearson P. Hall, International Edition 2004.
5. Fausett L.V. “Applied Numerical Analysis Using MATLAB, Second Edition, Pearson P. Hall, International Edition, 2008.
6. Gilat A. And Subramaniam V. “Numerical Methods, An introduction with Applications Using MATLAB”, Second Edition, John Wiley and Sons. Inc. 2011.

$$y = ae^{\beta x}$$

Take the natural logarithm of both sides.

$$\ln y = \ln \alpha + \beta x$$

This equation is in the form of $Y = a_1 X + a_0$ with $\ln \alpha$ corresponds to a_0 and β corresponds to a_1 .

$$\ln y = Y$$

$$\ln \alpha = a_0$$

$$\beta = a_1$$

$$x = X$$

X (x)	1300	1500	1600	1700	1800	1900
Y (ln y)	4.0943	4.3820	4.6540	5.0752	5.7038	6.1738

The following quantities can be computed.

$$\Sigma X_i Y_i = 1300 * 4.0943 + 1500 * 4.382 + 1600 * 4.654 + 1700 * 5.0752 + 1800 * 5.7038 + 1900 * 6.1738 = 49966.89$$

$$\Sigma X_i^2 = 1300^2 + 1500^2 + 1600^2 + 1700^2 + 1800^2 + 1900^2 = 16240000$$

$$\Sigma X_i = 1300 + 1500 + 1600 + 1700 + 1800 + 1900 = 9800$$

$$\Sigma Y_i = 4.0943 + 4.382 + 4.654 + 5.0752 + 5.7038 + 6.1738 = 30.0831$$

$$\bar{X} = \frac{\Sigma X_i}{n} = \frac{9800}{6} = 1633.3333$$

$$\bar{Y} = \frac{\Sigma Y_i}{n} = \frac{30.0831}{6} = 5.0139$$

Using equations for a_1 and a_0 ;

$$a_1 = \frac{n \Sigma X_i Y_i - \Sigma X_i \Sigma Y_i}{n \Sigma X_i^2 - (\Sigma X_i)^2} = \frac{6 * 49966.89 - 9800 * 30.0831}{6 * 16240000 - 9800^2} = 0.0036$$

$$a_0 = \bar{Y} - a_1 \bar{X} = 5.0139 - 0.0036 * 1633.3333 = -0.8661$$

Thus, the linear least-squares regression yields $Y = 0.0036X - 0.8661$

Substituting $Y = \ln y$ gives $\ln y = 0.0036x - 0.8661$

Thus, the intercept, $\ln \alpha$, equals -0.8661 and therefore by taking the antilogarithm

$$\ln \alpha = -0.8661 \quad \rightarrow \quad \alpha = 0.4206$$

$$\beta = a_1 = 0.0036$$

Consequently, the exponential equation is $y = 0.4206e^{0.0036x}$

linear least-squares regression

$$y = a_1x + a_0$$

x_i	-6	-4	-1	0	3	5	8
y_i	15	12	8	4	1	-5	-14

$$\Sigma x_i = (-6) + (-4) + (-1) + 0 + 3 + 5 + 8 = 5$$

$$\Sigma y_i = 15 + 12 + 8 + 4 + 1 + (-5) + (-14) = 21$$

$$\Sigma x_i^2 = (-6)^2 + (-4)^2 + (-1)^2 + 3^2 + 5^2 + 8^2 = 151$$

$$\Sigma x_i y_i = (-6) * 15 + (-4) * 12 + (-1) * 8 + 0 * 4 + 3 * 1 + 5 * (-5) + 8 * (-14) = -280$$

$$\bar{x} = \frac{\Sigma x_i}{n} = \frac{21}{7} = 3$$

$$\bar{y} = \frac{\Sigma y_i}{n} = \frac{5}{7} = 0.7143$$

Using equations for a_1 and a_0 ;

$$a_1 = \frac{n\Sigma x_i y_i - \Sigma x_i \Sigma y_i}{n\Sigma x_i^2 - (\Sigma x_i)^2} = \frac{7 * (-280) - 5 * 21}{7 * 151 - 5^2} = -2.0010$$

$$a_0 = \bar{y} - a_1 \bar{x} = 3 - (-2.0010) * 0.7143 = 4.4293$$

Therefore the least squares fit is $y = -2.0010x + 4.4293$

$$S_t = \sum (y_i - \bar{y})^2$$

y_i	$(y_i - \bar{y})^2$
15	$(15 - 0.7143)^2 = 204.0812$
12	$(12 - 0.7143)^2 = 127.3670$
8	$(8 - 0.7143)^2 = 53.0814$
4	$(4 - 0.7143)^2 = 10.7958$
1	$(1 - 0.7143)^2 = 0.0816$
-5	$(-5 - 0.7143)^2 = 32.6532$
-14	$(-14 - 0.7143)^2 = 216.5106$

$$S_t = \sum (y_i - \bar{y})^2 = 644.5708$$

linear least-squares regression

the coefficients α and β in the function $y = \left(\frac{1}{\alpha x^2 + \beta}\right)^2$

x_i	-1.0	-0.3	0.2	1.0	2.0
y_i	2.8	4.4	4.6	2.7	1.2

The function is linearized by inverting to give

$$y = \left(\frac{1}{\alpha x^2 + \beta} \right)^2$$

$$\sqrt{y} = \frac{1}{\alpha x^2 + \beta} \quad \rightarrow \quad \frac{1}{\sqrt{y}} = \alpha x^2 + \beta \quad \text{or} \quad Y = a_1 X + a_0$$

where

$$\frac{1}{\sqrt{y}} = Y$$

$$\alpha = a_1$$

$$x^2 = X$$

$$\beta = a_0$$

X (x ²)	1	0.09	0.04	1	4
Y ($\frac{1}{\sqrt{y}}$)	0.5976	0.4767	0.4663	0.6086	0.9129

The following quantities can be computed.

$$\Sigma X_i Y_i = 1 * 0.5976 + 0.09 * 0.4767 + 0.04 * 0.4663 + 1 * 0.6086 + 4 * 0.9129 = 4.9194$$

$$\Sigma X_i^2 = 1^2 + 0.09^2 + 0.04^2 + 1^2 + 4^2 = 18.0097$$

$$\Sigma X_i = 1 + 0.09 + 0.04 + 1 + 4 = 6.13$$

$$\Sigma Y_i = 0.5976 + 0.4767 + 0.4663 + 0.6086 + 0.9129 = 3.0621$$

$$\bar{X} = \frac{\sum X_i}{n} = \frac{6.13}{5} = 1.226$$

$$\bar{Y} = \frac{\sum Y_i}{n} = \frac{3.0621}{5} = 0.6124$$

Using equations for a_1 and a_0 ;

$$a_1 = \frac{n\sum X_i Y_i - \sum X_i \sum Y_i}{n\sum X_i^2 - (\sum X_i)^2} = \frac{5 * 4.9194 - 6.13 * 3.0621}{5 * 18.0097 - 6.13^2} = 0.1110$$

$$a_0 = \bar{Y} - a_1 \bar{X} = 0.6124 - 0.1110 * 1.226 = 0.4763$$

Thus, the linear least-squares regression yields

$$y = \left(\frac{1}{0.1110x^2 + 0.4763} \right)^2$$

least-squares regression

$$y = a_2x^2 + a_1x + a_0$$

the coefficients a_0 , a_1 and a_2

$$\Sigma x_i = 0 + 1 + 2 + 3 = 6$$

$$\Sigma y_i = 2.3 + 7 + 14 + 27 = 50.3$$

$$\Sigma x_i^2 = (0)^2 + (1)^2 + (2)^2 + (3)^2 = 14$$

$$\Sigma x_i^3 = (0)^3 + (1)^3 + (2)^3 + (3)^3 = 36$$

$$\Sigma x_i^4 = (0)^4 + (1)^4 + (2)^4 + (3)^4 = 98$$

$$\Sigma x_i y_i = 0 \times 2.3 + 1 \times 7 + 2 \times 14 + 3 \times 27 = 116$$

$$\Sigma x_i^2 y_i = (0)^2 \times 2.3 + (1)^2 \times 7.0 + (2)^2 \times 14 + (3)^2 \times 27 = 306$$

So the matrices become,

$$\begin{bmatrix} 4 & 6 & 14 \\ 6 & 14 & 36 \\ 14 & 36 & 98 \end{bmatrix} \begin{Bmatrix} a_0 \\ a_1 \\ a_2 \end{Bmatrix} = \begin{Bmatrix} 50.3 \\ 116 \\ 306 \end{Bmatrix}$$

Solving the system of linear equations through Gauss-Elimination gives,

$$[6 \ 14 \ 36 \ 116] - \frac{6}{4}[4 \ 6 \ 14 \ 50.3] = [0 \ 5 \ 15 \ 40.55]$$

$$[14 \ 36 \ 98 \ 306] - \frac{14}{4}[4 \ 6 \ 14 \ 50.3] = [0 \ 15 \ 49 \ 129.95]$$

$$[0 \ 15 \ 49 \ 129.95] - \frac{15}{5}[0 \ 5 \ 15 \ 40.55] = [0 \ 0 \ 4 \ 8.3]$$

So, the matrix becomes

$$\begin{bmatrix} 4 & 6 & 14 \\ 0 & 5 & 15 \\ 0 & 0 & 4 \end{bmatrix} \begin{Bmatrix} a_0 \\ a_1 \\ a_2 \end{Bmatrix} = \begin{Bmatrix} 50.3 \\ 40.55 \\ 8.3 \end{Bmatrix} \rightarrow \begin{matrix} a_2 = 2.075 \\ a_1 = 1.885 \\ a_0 = 2.485 \end{matrix}$$

$$2.075 + 1.885x + 2.485x^2$$

$$\bar{y} = \frac{\sum y_i}{n} = \frac{50.3}{4} = 12.575$$

x_i	y_i	$S_t = (y_i - \bar{y})^2$	$S_r = (y_i - a_0 - a_1x_i - a_2x_i^2)^2$
0	2.3	109.7256	0.0342
1	7.0	31.0806	0.3080
2	14	2.0306	0.3080
3	27	208.0806	0.0342
Σ	50.3	350.9174	0.6844

$$r^2 = \frac{S_t - S_r}{S_t} = 0.9980$$