

ORIGIN = 1

Simple Linear Regression

Linear Regression and the so-called "General Linear Model" represent a class of methods that seek to relate values of an observed "dependent" random variable (Y) that is Normally distributed to one or more "independent" (or predictor) variables (X) using a linear function analogous to a linear transformation - i.e., using only translation and change of scale. We typically employ "linear coefficients" α & β (not to be confused with the probability of types I & II errors in statistical tests) to describe translation (α) and change of scale (β). Thus a function such as $Y = 5 + 23X$ qualifies as a linear function ($\alpha=5, \beta=23$). Note, however, that with the use of an appropriate non-linear transformations of the data, many non-linear functions can be treated by general linear methods also. For instance, transforming variable X to X^2 allows one to model $Y = X^2$ ($\alpha=0, \beta=1$) and taking logs allows one to model the famous allometric equation: $Y = aX^b$ as $\ln(Y) = \ln(a) + b(\ln(X))$ ($\alpha=\ln(a), \beta=b$).

ZAR := READPRN("c:/DATA/Biostatistics/ZarEX17.1R.txt")

X := ZAR^{<2>} ^Zar Example 17.1

Y := ZAR^{<3>}

X_{bar} := mean(X) X_{bar} = 10

Y_{bar} := mean(Y) Y_{bar} = 3.4154

s := $\sqrt{\text{Var}(Y)}$ s = 1.2799 s² = 1.6381

n := length(Y) n = 13

i := 1..n

X =	Y =	ZAR =
3	1.4	
4	1.5	
5	2.2	
6	2.4	
8	3.1	
9	3.2	
10	3.2	
11	3.9	
12	4.1	
14	4.7	
15	4.5	
16	5.2	
17	5	

	1	2	3
1	1	3	1.4
2	2	4	1.5
3	3	5	2.2
4	4	6	2.4
5	5	8	3.1
6	6	9	3.2
7	7	10	3.2
8	8	11	3.9
9	9	12	4.1
10	10	14	4.7
11	11	15	4.5
12	12	16	5.2
13	13	17	5

Assumptions:

- Standard Linear Regression depends on specifying in advance which variable is to be considered 'dependent' and which 'independent'. This decision matters as changing roles for Y & X usually produces a different result.
- $Y_1, Y_2, Y_3, \dots, Y_n$ (dependent variable) is a random sample $\sim N(\mu, \sigma^2)$.
- $X_1, X_2, X_3, \dots, X_n$ (independent fixed values) with each value of X_i matched to Y_i

Model:

$Y = \alpha + \beta X + \epsilon$

where: α is the y **intercept** of the regression line (translation)
 β is the **slope** of the regression line (scaling coefficient)
 ϵ is the error factor in prediction of Y given that it is a random variable distributed as $N(0, \sigma^2)$.

Least Squares Estimation of the Regression Line:

Sums of Squares and Cross Products corrected for mean location:

$$L_{XX} := \sum_i (X_i - X_{\text{bar}})^2 \qquad L_{XX} = 262 \qquad < \text{corrected Sum of squares of X}$$

$$L_{YY} := \sum_i (Y_i - Y_{\text{bar}})^2 \qquad L_{YY} = 19.6569 \qquad < \text{corrected Sum of squares of Y}$$

$$L_{XY} := \sum_i (X_i - X_{\text{bar}}) \cdot (Y_i - Y_{\text{bar}}) \qquad L_{XY} = 70.8 \qquad < \text{corrected Sum of cross products}$$

Estimated Regression Coefficients for $Y = \alpha + \beta X$:

$$b := \frac{L_{xy}}{L_{xx}} \qquad b = 0.2702 \qquad < \mathbf{b} \text{ is the sample estimate of } \beta$$

$$a := Y_{\text{bar}} - b \cdot X_{\text{bar}} \qquad a = 0.7131 \qquad < \mathbf{a} \text{ is the sample estimate of } \alpha$$

Estimated values of Y (Y_{hat}):

$$Y_{\text{hat}_i} := a + b \cdot X_i \qquad < \text{using estimated coefficients and each value of the independent variable to estimate dependent value points on the Regression line.}$$

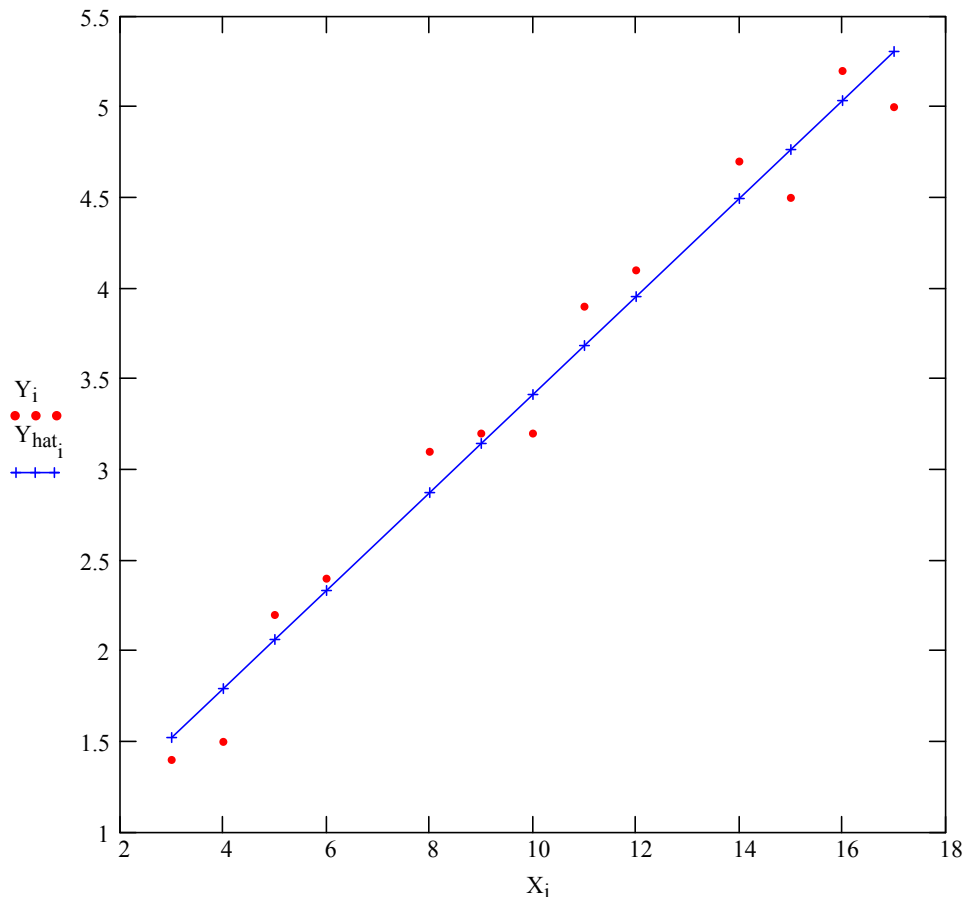
$Y_{\text{hat}} =$	1.5238
	1.794
	2.0642
	2.3345
	2.8749
	3.1452
	3.4154
	3.6856
	3.9558
	4.4963
	4.7665
	5.0368
	5.307

Residuals:

$$e_i := Y_{\text{hat}_i} - Y_i \qquad < \text{deviation of each value } Y_i \text{ from Regression line} = Y_{\text{hat}_i}$$

$e =$	0.1238
	0.294
	-0.1358
	-0.0655
	-0.2251
	-0.0548
	0.2154
	-0.2144
	-0.1442
	-0.2037
	0.2665
	-0.1632
	0.307

Plot of Values:



Prototype in R:

```
#SIMPLE LINEAR REGRESSION
```

```
#ZAR EXAMPLE 17.1
```

```
ZAR=read.table("c:/DATA/Biostatistics/ZarEX17.1R.txt")
```

```
ZAR
```

```
attach(ZAR)
```

```
X=ageX
```

```
Y=wingY
```

```
#CREATING LINEAR MODEL LM:
```

```
LM=lm(Y~X)
```

```
LM
```

```
Yhat=fitted(LM)
```

```
e=residuals(LM)
```

```
RESULTS=data.frame(X,Y,Yhat,e)
```

```
RESULTS
```

```
> RESULTS
```

	X	Y	Yhat	e
1	3	1.4	1.523782	-0.12378156
2	4	1.5	1.794011	-0.29401057
3	5	2.2	2.064240	0.13576042
4	6	2.4	2.334469	0.06553142
5	8	3.1	2.874927	0.22507340
6	9	3.2	3.145156	0.05484439
7	10	3.2	3.415385	-0.21538462
8	11	3.9	3.685614	0.21438638
9	12	4.1	3.955843	0.14415737
10	14	4.7	4.496301	0.20369935
11	15	4.5	4.766530	-0.26652965
12	16	5.2	5.036759	0.16324134
13	17	5.0	5.306988	-0.30698767

```
#PLOTING REGRESSION LINE & POINTS:
```

```
plot(X,Y)
```

```
abline(LM,col="blue")
```

```
segments(X,Yhat,X,Y,col="red")
```

