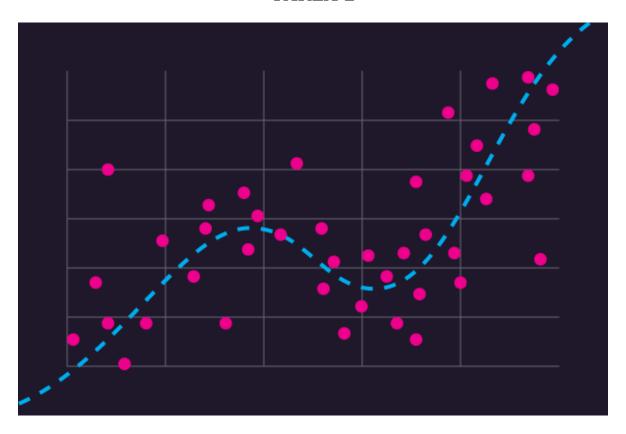


MODELACIÓN Y SIMULACIÓN AVANZADA TAREA 1



Estudiante:

Marcelo Saavedra Alcoba PROFESOR: PhD. Ranjit Das II/2022 a) Derive Standard Least square Regression with the given data

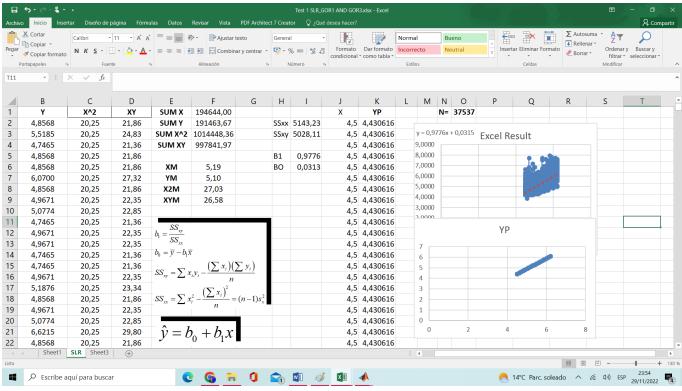


Fig. 1. Simple Lineal Regression with Excel

The same problem was solved from Matlab with the following code:

```
clc; clear all; close all
%T = open('Test 1 SLR_GOR1 AND GOR2.xlsx');
load('valModel.mat');
x=values(:,1);
y=values(:,2);
n=size(x);
x2=x.^2;
xy=x.*y;

SSxy= sum(xy)-(sum(x)*sum(y))/n(1);
SSxx= sum(x2)-(sum(x)^2)/n(1);
b1=SSxy/SSxx;
b0=mean(y)-b1*mean(x);

Yp=b0+b1*x;

plot(x,y,'*')
```

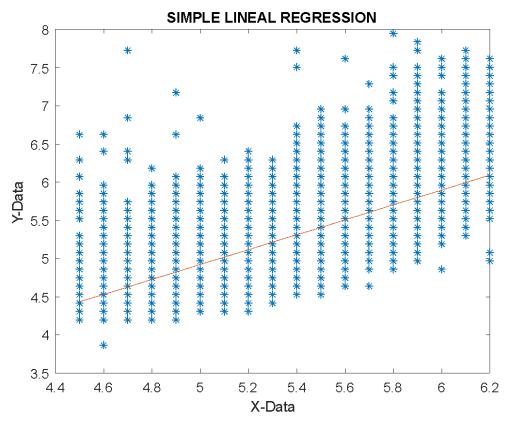


Fig. 2. Simple Lineal Regression with Matlab

In both cases (Excel and matlab) the results were:

$$b_1 = 0.9776$$

$$b_0 = 0.0315$$

$$\hat{y} = b_0 + b_1 x$$

b) Derive Orthogonal Regression with the given data using ita =0.2

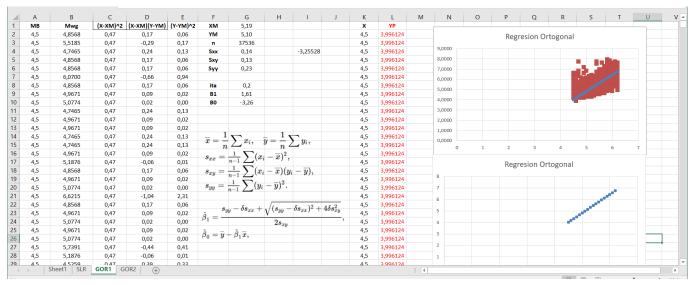


Fig. 3. Orthogonal Regression with Excel

The same problem was solved from Matlab with the following code:

```
clc; clear all; close all
T = open('Test 1 SLR GOR1 AND GOR2.xlsx');
%load('valModel.mat');
x=T.data(:,1);
y=T.data(:,2);;
n=size(x);
xm=mean(x):
ym=mean(y);
% (x-xm)^2
ecux=(x-xm).^2;
% (x-xm) (y-ym)
ecuxy=(x-xm) \cdot (y-ym);
% (y-ym)^2
ecuy=(y-ym).^2;
Sxx = 1/(n(1)-1) * sum(ecux) ;
Sxy = 1/(n(1)-1) * sum(ecuxy);
Syy= 1/(n(1)-1) * sum(ecuy);
ita=0.2;
B1= (Syy - ita*Sxx + sqrt ((Syy-ita*Sxx)^2 + 4*ita*(Sxy^2)))/(2*Sxy);
B0 = ym - B1* xm;
Yp=B0+B1*x;
plot(x,y,'*')
hold on
plot(x, Yp)
```

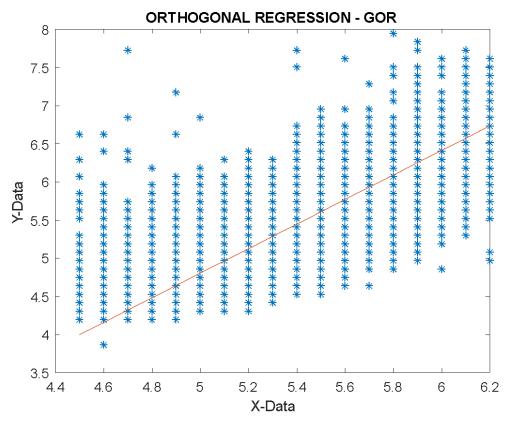


Fig. 4. Orthogonal Regression with Matlab

In both cases (Excel and matlab) the results were::

$$b_1 = 1.6114$$

$$b_0 = -3.2553$$

$$\hat{y} = b_0 + b_1 x$$

c) Derive Modified Orthogonal Regression with the given data using ita = 0.2

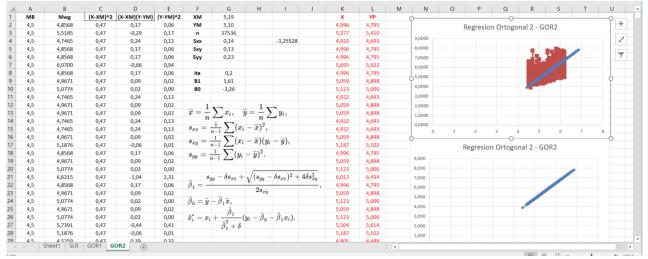


Fig. 5. Modified Orthogonal Regression with Excel

The same problem was solved from Matlab with the following code:

```
clc; clear all; close all
T = open('Test 1 SLR GOR1 AND GOR2.xlsx');
%load('valModel.mat');
x=T.data(:,1);
y=T.data(:,2);;
n=size(x);
xm=mean(x);
ym=mean(y);
% (x-xm)^2
ecux=(x-xm).^2;
% (x-xm) (y-ym)
ecuxy=(x-xm).*(y-ym);
% (y-ym)^2
ecuy=(y-ym).^2;
Sxx = 1/(n(1)-1) * sum(ecux);

Sxy = 1/(n(1)-1) * sum(ecuxy);
Syy= 1/(n(1)-1) * sum(ecuy);
ita=0.2;
B1= (Syy - ita*Sxx + sqrt ((Syy-ita*Sxx)^2 + 4*ita*(Sxy^2)))/(2*Sxy);
B0 = ym - B1* xm;
xi = x+B1/(B1^2+ita) * (y-B0-B1*x);
Yp=B0+B1*xi;
plot(x,y,'*')
hold on
nlativi Vnl
```

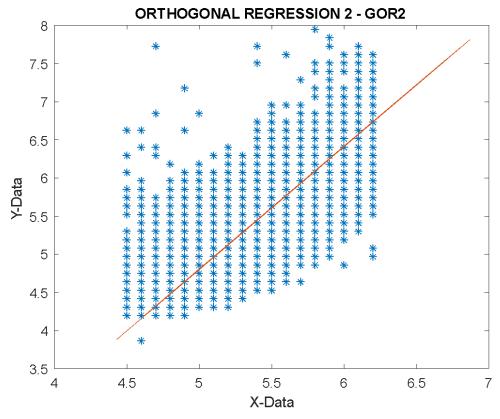


Fig. 6. Orthogonal Regression with Matlab

In both cases (Excel and matlab) the results were::

$$b_1 = 1.6114$$

$$b_0 = -3.2553$$

$$\hat{y} = b_0 + b_1 \hat{x}$$