

Least square fit of a line/polynomial to input/output data

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30th Nov, 2010

Outline

- 1 Scilab
- 2 Least squares
- 3 Scilab commands

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- Today: best fit: line and polynomial : **reglin command**

Linear fit

Given n samples of (x, z) pairs:

x_i and z_i for $i = 1, \dots, n$, we **expect** following equation is satisfied

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x_i and z_i fall on some line with slope a_1 and 'z-intercept'= a_0 .

The '**line fit**' problem:

Find these constants a_1 and a_0 .

'**Best**' fit?

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Real situation:

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Find constants a_d, \dots, a_1 and a_0 such that the 'total **square error**'

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where x and z are vectors with same number of **columns**.

$$\sum_i (z_i - a_1 x_i - a_0)^2$$

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(**Smaller** sig means **better** fit.)

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If x has more than one rows:

components in $a1 =$ number of rows of x

(number of independent variables.)

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Use noisy data to estimate a_0 and a_1 :

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(P is not overwritten. A new matrix B is defined.)

Exercises:

Generate data, add noise, and estimate the parameters back:

One independent variable (first).

$$a_0 = 3;$$

$$a_1 = 6;$$

(actual)

$$x = 1:10;$$

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y = a1a*x + a0a + dev*(rand(x)-0.5)
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Use this x and y to estimate a_0 and a_1

Compare a_0 & a_1 with actual a_{0a} & a_{1a} .

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For sig to compare with dev, note that sig is standard deviation (most relevant for **normal** distribution, and we used **uniform** distribution).

Fit 2nd order polynomial

Find a_2 , a_1 and a_0 such that

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

Suppose $a_0 = 3$; $a_1 = 6$; $a_2 = 2$; (actual)

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y = a1a*x + a2a*x.^2 + a0a + dev*(rand(x)-0.5)
```

```
x = 1:10; x2 = x.^2;
```

```
X=[x;x2];
```

```
reglin(X,y)
```

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r = read_csv('lsquare_data.csv',ascii(9)); // read csv data into r
r = strsubst(r,',',''); // string substitute
r = evstr(r); // convert string to numerical values
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xval = r(1,:); //first row (time)
yval = r(2,:); //second row (displacement)
```

Displacement under gravity

Constant gravity: $g = 9.8 \text{ m/s}^2$ **downwards**.

Initial velocity: v_0

Initial displacement: x_0

$$x(t) = x_0 + v_0 t - \frac{1}{2} g t^2$$

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