

# MAS212 Scientific Computing and Simulation

## #8: Fitting Models to Data

### Key resources:

- Lec 8: <http://sam-dolan.staff.shef.ac.uk/mas212/docs/18.pdf>
- Data sets : <http://sam-dolan.staff.shef.ac.uk/mas212/data/>.

```
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline # include plots in ipython notebook
```

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1. **A straight-line fit.** Download the data set (`ds1.txt`), open with `np.loadtxt()`, and plot:

```
x,y = np.loadtxt('ds1.txt')
plt.plot(x, y, 's')
```

(a) Using `numpy` (or writing your own functions) calculate the means  $\bar{x}$  and  $\bar{y}$ , the variance  $\text{var}(x) = \frac{1}{N} \sum (x_i - \bar{x})^2$  and the covariance,  $\text{covar}(x, y) = \frac{1}{N} \sum (x_i - \bar{x})(y_i - \bar{y})$ . Hence find the parameters of the line of best fit:

$$y = mx + c, \quad m = \frac{\text{covar}(x, y)}{\text{var}(x)}, \quad c = \bar{y} - m\bar{x}.$$

Check your answer by plotting the line of best fit along with the data.

(b) Given that the data set is a just sample of some underlying distribution, with random noise: What do you think might be the ‘true’ values of  $m$  and  $c$  in the underlying model? What is the uncertainty in your best-fit parameters  $m$  and  $c$ ? How could you estimate this uncertainty from the data? Discuss with your neighbour.

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2. **A linear model** (i.e. linear in its parameters).

Load the second data set (`ds2.txt`) and plot. We are going to fit the data to a cubic in  $x$ :

$$f(x, \beta_j) = \sum_{j=0}^3 \beta_j \phi_j(x), \quad \phi_j(x) = x^j.$$

We will solve the **normal equations** for the vector of parameters  $\boldsymbol{\beta} = (\beta_0, \dots, \beta_3)$ :

$$(\mathbf{X}^T \mathbf{X}) \boldsymbol{\beta} = \mathbf{X}^T \mathbf{y}.$$

(a) Calculate the  $N \times 4$  matrix  $\mathbf{X}$  with elements  $X_{ij} = \phi_j(x_i)$ .

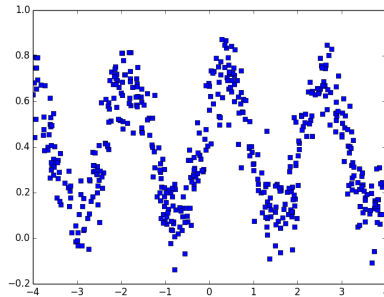
(b) Compute the  $4 \times 4$  matrix  $\mathbf{X}^T \mathbf{X}$  and the vector  $\mathbf{X}^T \mathbf{y}$ . Hence solve the normal equations for  $\boldsymbol{\beta}$  using `np.linalg.solve()`. (You should find  $\beta_0 \approx 1.0$ ,  $\beta_1 \approx 1.5$ ,  $\beta_2 \approx 0$  and  $\beta_3 \approx -0.2$ .)

```
from scipy.optimize import curve_fit
?curve_fit      # to see the help file / docstring.
```

(c) Fit the same model using the function `scipy.optimize.curve_fit()`. Compare the parameter values obtained with those from part (b) – they should be identical.

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**3. A non-linear model.** Load `ds3.txt` and plot.



We could try fitting a non-linear model of the following form:

$$f(x, \beta_j) = \beta_0 + \beta_1 \sin(\beta_2 x + \beta_3)$$

Use `curve_fit()` to find best-fit parameters for this model. Check your answers by plotting the model on top of the data.

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