

Straight Line Fit: Matrix Formalism

- It is useful to restate the problem using a matrix formalism [4:Avery 1991]

- This is useful because:

- It is more compact
- It is easily extensible to other linear problems
- It is more useful to formulate an iterative procedure

- With the same assumption as before

- The linear model is given by $\mathbf{f} = \mathbf{A}\mathbf{p}$

$$\mathbf{f} = \begin{pmatrix} f_0 \\ \dots \\ f_N \end{pmatrix} = \begin{pmatrix} a + bz_0 \\ \dots \\ a + bz_N \end{pmatrix} = \begin{pmatrix} 1 & z_0 \\ 1 & \dots \\ 1 & z_N \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \mathbf{A}\mathbf{p}$$

- Measurements and errors are

$$\mathbf{Y} = \begin{pmatrix} y_0 \\ \dots \\ y_N \end{pmatrix} \quad (\mathbf{V})_{ij} = \langle (y_i - \langle y_i \rangle)(y_j - \langle y_j \rangle) \rangle$$

$$(\mathbf{V})_{ij} = \sigma_i^2 \delta_{ij} \quad \text{if uncorrelated}$$

- The χ^2 can be written as

$$\chi^2 = (\mathbf{Y} - \mathbf{A}\mathbf{p})^T \mathbf{W} (\mathbf{Y} - \mathbf{A}\mathbf{p}) \quad \mathbf{W} = \mathbf{V}^{-1}$$

- The minimum χ^2 is obtained by

$$\tilde{\mathbf{p}} = (\mathbf{A}^T \mathbf{W} \mathbf{A})^{-1} \mathbf{A}^T \mathbf{W} \mathbf{Y}$$

- The covariance matrix of the parameters is obtained from covariance matrix \mathbf{V} of the measurements

$$\mathbf{V}_p = (\mathbf{A}^T \mathbf{V}^{-1} \mathbf{A})^{-1}$$

- Please notice ($N+1$ measurements, M parameters)

- Dimensions $\mathbf{A} = (N+1) \times M$
- Dimensions $\mathbf{V} = (N+1) \times (N+1)$
- Dimensions $\mathbf{A}^T \mathbf{W} \mathbf{A} = M \times M$
- Dimensions $\mathbf{A}^T \mathbf{W} = M \times (N+1)$
- Dimensions $\mathbf{V}_p = M \times M$