

Joint Research Centre (JRC)



Uncertainty propagation in Time-of-Flight Data

IRMM - Institute for Reference Materials and Measurements

Geel - Belgium

<http://irmm.jrc.ec.europa.eu/>

<http://www.jrc.ec.europa.eu/>

- **Definitions**
- **Importance of Covariance Matrix**
- **How to store the matrix**

Mean $\mu_x = \langle x \rangle \quad \mu_y = \langle y \rangle$

Variance $\sigma_x^2 = \langle (x - \mu_x)^2 \rangle = \text{cov}(x, x)$

Covariance $\text{cov}(x, y) = \langle (x - \mu_x)(y - \mu_y) \rangle$

Correlation $\rho(x, y) = \frac{\text{cov}(x, y)}{\sigma_x \sigma_y}$

Linear function

$$f(x_i) = \sum_{i=1}^k a_i x_i$$

Mean

$$\mu_f = \langle f(x_i) \rangle = \sum_{i=1}^k a_i \mu_i$$

Variance

$$\sigma_f^2 = \left\langle \left(f(x_i) - \mu_f \right)^2 \right\rangle = \sum_i \sum_j a_i a_j \text{COV}(x_i, x_j)$$

Taylor expansion (1st order)

$$f(x_i) \cong f(\mu_i) + \sum_i \left. \frac{\partial f}{\partial x_i} \right|_{\mu} (x_i - \mu_i)$$

$$f(x_i) \cong f(\mu_i) + \sum_i g_i (x_i - \mu_i)$$

$$g_i = \left. \frac{\partial f}{\partial x_i} \right|_{\mu}$$

Mean

$$\mu_f = \langle f(x_i) \rangle \cong f(\mu_i)$$

Variance

$$\sigma_f^2 = \left\langle \left(f(x_i) - \mu_f \right)^2 \right\rangle \cong \sum_i \sum_j g_i g_j \text{cov}(x_i, x_j)$$

Linear function

$$f(\vec{x}) = \underline{A} \vec{x}$$

$$a_{ik} = \frac{\partial f_i}{\partial x_k}$$

Mean

$$\vec{\mu}_f = \underline{A} \vec{\mu}_x$$

Variance

$$\underline{V}_f = \underline{A} \underline{V}_x \underline{A}^T \quad \underline{V}_x = \begin{pmatrix} \text{cov}(x_1, x_1) & \text{cov}(x_1, x_2) \\ \text{cov}(x_2, x_1) & \text{cov}(x_2, x_2) \end{pmatrix}$$

Non-linear function (1st order Taylor)

$$f(\vec{x}) \cong f(\vec{\mu}_x) + \underline{G}_x (\vec{x} - \vec{\mu}_x)$$

$$g_{x,ik} = \frac{\partial f_i}{\partial x_k}$$

Mean

$$\vec{\mu}_f \cong f(\vec{\mu}_x)$$

sensitivity matrix
design matrix

Variance

$$\underline{V}_f \cong \underline{G}_x \underline{V}_x \underline{G}_x^T$$

Experiment

Independent observables : $(y_1 \pm \sigma_{y_1})$ $(y_2 \pm \sigma_{y_2})$

Common background component: $(b \pm \sigma_b)$

Data reduction : $z = y - b$

Average of (y_1, y_2) : $\langle y \rangle = \frac{y_1 + y_2}{2}$ $\sigma_{\langle y \rangle}^2 = \frac{\sigma_{y_1}^2 + \sigma_{y_2}^2}{4}$

Background subtraction : $\langle z \rangle = \langle y \rangle - b$ $\sigma_{\langle z \rangle}^2 = \frac{\sigma_{y_1}^2 + \sigma_{y_2}^2}{4} + \sigma_b^2$

$$(y_1, y_2, b) \rightarrow (z_1, z_2) = (y_1 - b, y_2 - b) \rightarrow \langle z \rangle = \frac{z_1 + z_2}{2}$$

**uncertainties on y_1 and y_2 not correlated
e.g. due to counting statistics**

$$V_y = \begin{bmatrix} \sigma_{y_1}^2 & 0 \\ 0 & \sigma_{y_2}^2 \end{bmatrix}$$

**uncertainty on background
(b not correlated with y_1 and y_2)**

$$V_b = \sigma_b^2$$

$$(z_1, z_2) \Rightarrow \langle z \rangle$$

$$(y_1, y_2, b) \rightarrow (z_1, z_2) = (y_1 - b, y_2 - b)$$

uncorrelated uncertainties due to counting statistics : $\underline{V}_y = \begin{bmatrix} \sigma_{y_1}^2 & 0 \\ 0 & \sigma_{y_2}^2 \end{bmatrix}$

uncertainty on background (b not correlated with y_1 and y_2): $V_b = \sigma_b^2$

$$\underline{V}_z = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} \sigma_{y_1}^2 & 0 & 0 \\ 0 & \sigma_{y_2}^2 & 0 \\ 0 & 0 & \sigma_b^2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & -1 \end{bmatrix} = \begin{bmatrix} \sigma_{y_1}^2 + \sigma_b^2 & \sigma_b^2 \\ \sigma_b^2 & \sigma_{y_2}^2 + \sigma_b^2 \end{bmatrix}$$

$$\underline{V}_z = \underline{A} \underline{V}_{y_1, y_2, b} \underline{A}^T$$

$$(z_1, z_2) \rightarrow \langle z \rangle = \frac{z_1 + z_2}{2}$$

$$V_{\langle z \rangle} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} \sigma_{y_1}^2 + \sigma_b^2 & \sigma_b^2 \\ \sigma_b^2 & \sigma_{y_2}^2 + \sigma_b^2 \end{bmatrix} \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{bmatrix} = \begin{bmatrix} \frac{\sigma_{y_1}^2 + \sigma_{y_2}^2}{4} + \sigma_b^2 \end{bmatrix}$$

$$\underline{V}_{\langle z \rangle} = \underline{A} \underline{V}_{z_1, z_2} \underline{A}^T$$

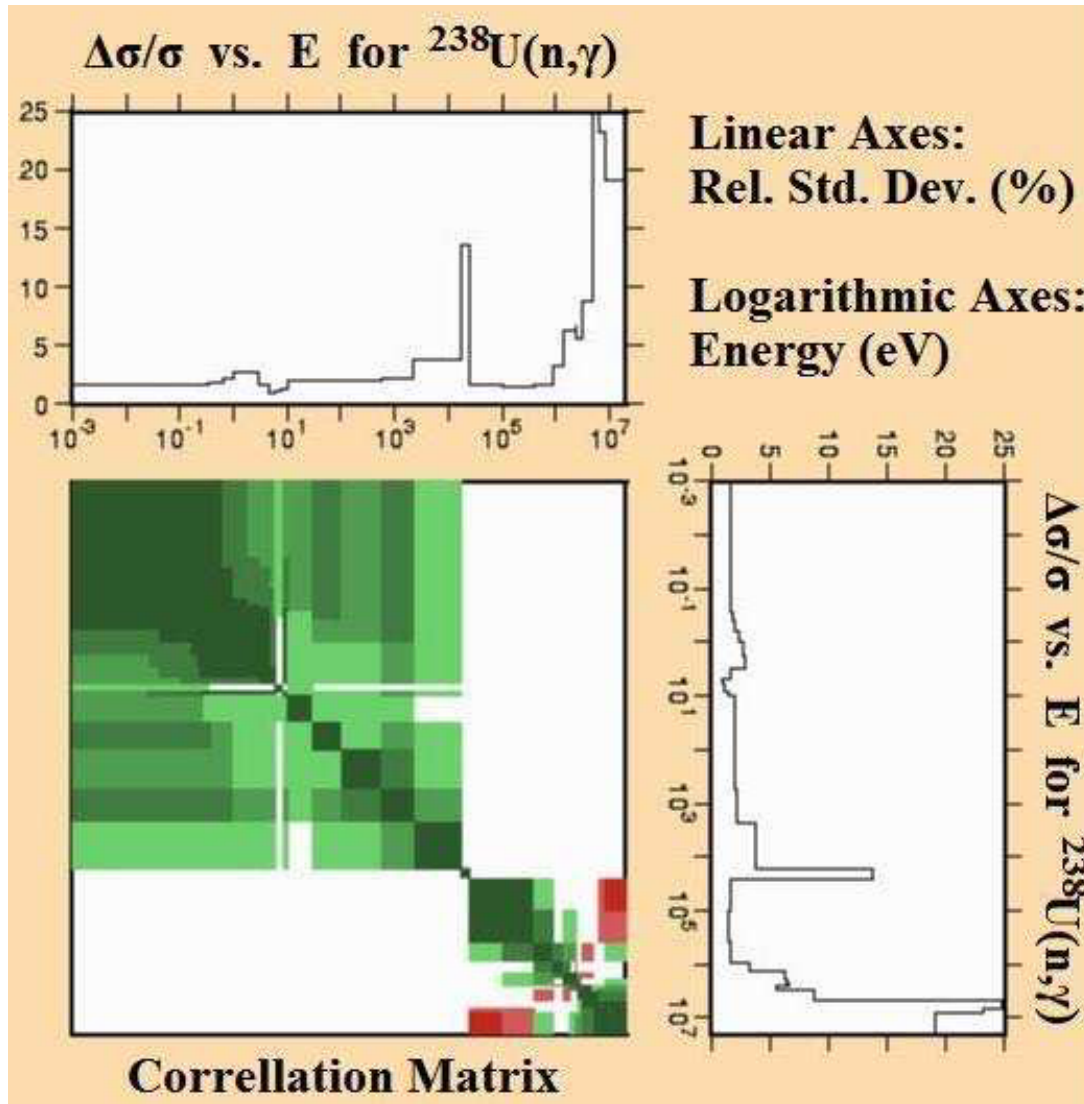
Full covariance

$$\sigma_{\langle z \rangle}^2 = \frac{\sigma_{y_1}^2 + \sigma_{y_2}^2}{4} + \sigma_b^2$$

Only diagonal terms

$$\sigma_{\langle z \rangle}^2 = \frac{\sigma_{y_1}^2 + \sigma_{y_2}^2}{4} + \frac{\sigma_b^2}{2}$$

⇒ Covariance matrix contains information about experiment and data reduction



Covariance Files in Evaluated Data Files

Produced:
During evaluation from data

Retrospective (model)

$^{206}\text{Pb} + n$

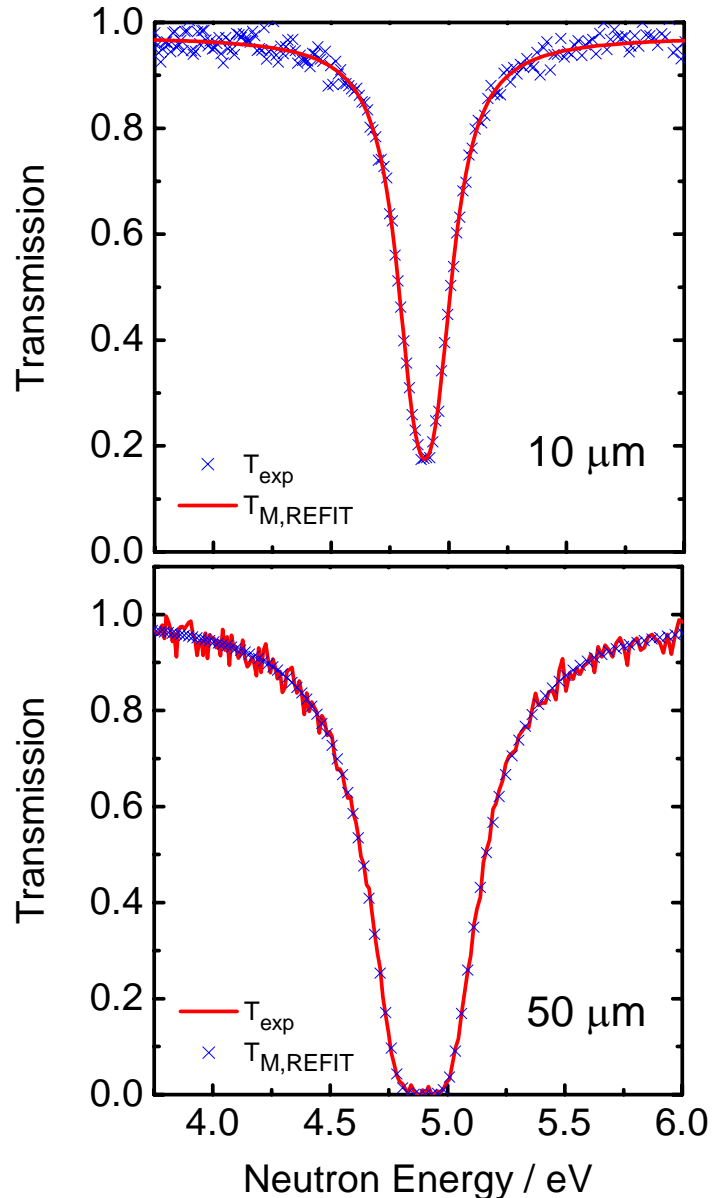
$$C_\theta = (D_\theta^T C_{Y,\text{exp}}^{-1} D_\theta)^{-1}$$

GELINA (transmission + capture) + ORELA (transmission)
Borella et al. PRC 76 (2007) 014605 and Horen et al. PRC 20 (1979) 478

RETROSPECTIVE

Rochman and Koning
NIM A589 (2008) 85

E_R	$g\Gamma_n / \text{eV}$	$g\Gamma_\gamma / \text{eV}$	$g\Gamma_n\Gamma_\gamma / (\Gamma_n + \Gamma_\gamma) / \text{eV}$	$\rho(\Gamma_n, \Gamma_\gamma)$	$\rho(\Gamma_n, \Gamma_\gamma)$
3.36	0.570 ± 0.004	0.146 ± 0.001	0.116 ± 0.001	- 0.15	- 0.16
42.07	1.966 ± 0.163	1.171 ± 0.082	0.734 ± 0.014	- 0.92	- 0.51
66.00	82.210 ± 0.414	1.398 ± 0.018	1.375 ± 0.017	0.03	- 0.00
92.61	32.0 ± 4.0	1.503 ± 0.017	1.436 ± 0.016	- 0.10	- 0.77

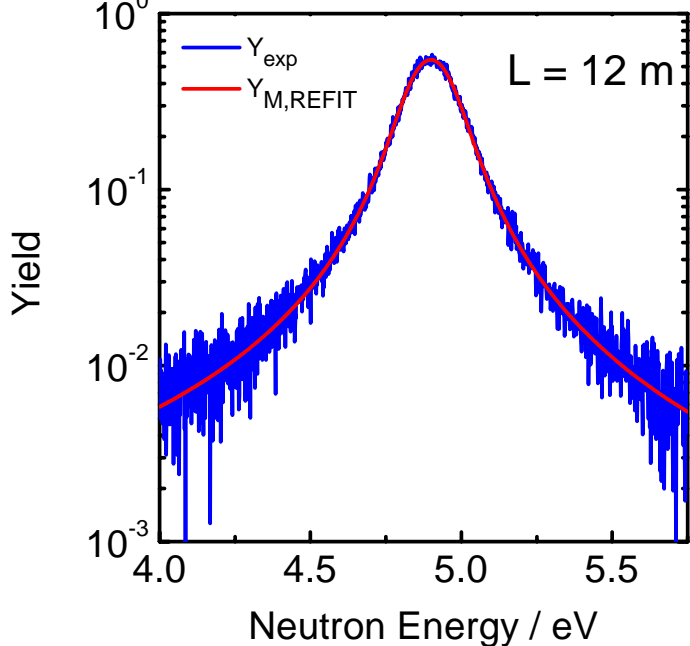
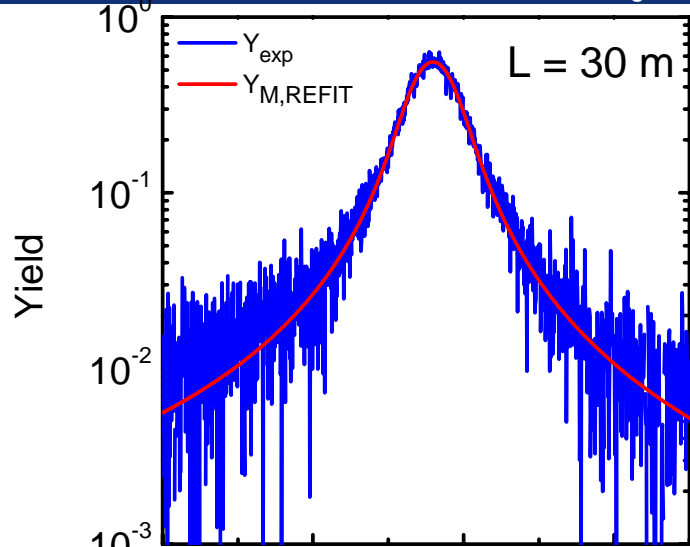


$C_{T,\text{exp}}$: only uncorrelated uncertainties
due to counting statistics

$$\Rightarrow \begin{aligned} \Gamma_n &= (15.06 \pm 0.08) \text{ meV} \\ \Gamma_\gamma &= (121.7 \pm 1.3) \text{ meV} \\ \rho(\Gamma_n, \Gamma_\gamma) &= 0.55 \end{aligned}$$

$$\Rightarrow \begin{aligned} \Gamma_n &= (14.66 \pm 0.30) \text{ meV} \\ \Gamma_\gamma &= (124.8 \pm 3.7) \text{ meV} \\ \rho(\Gamma_n, \Gamma_\gamma) &= -0.96 \end{aligned}$$

RSA by REFIT



$C_{Y,\text{exp}}$: only uncorrelated uncertainties
due to counting statistics

$$\Rightarrow \begin{aligned} \Gamma_n &= (15.31 \pm 0.12) \text{ meV} \\ \Gamma_\gamma &= (118.0 \pm 1.4) \text{ meV} \\ \rho(\Gamma_n, \Gamma_\gamma) &= -0.50 \end{aligned}$$

$$\Rightarrow \begin{aligned} \Gamma_n &= (15.26 \pm 0.15) \text{ meV} \\ \Gamma_\gamma &= (118.9 \pm 1.2) \text{ meV} \\ \rho(\Gamma_n, \Gamma_\gamma) &= -0.63 \end{aligned}$$

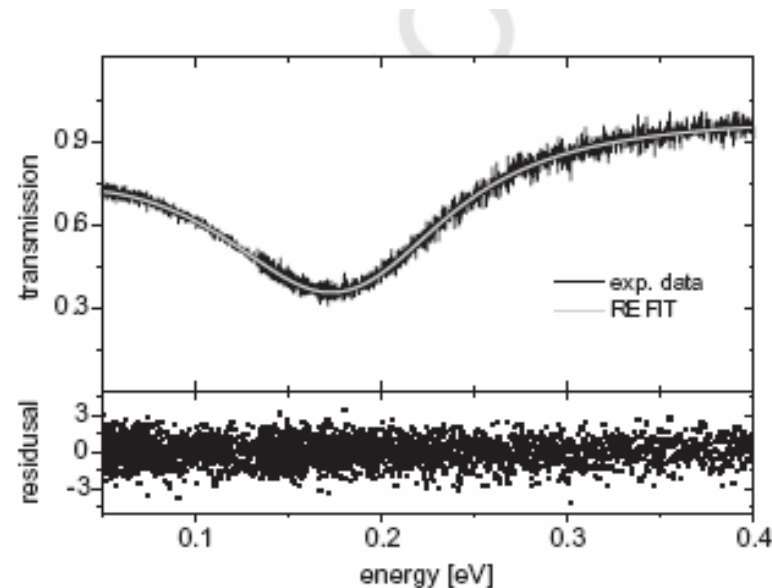
RSA by REFIT

$C_{Z,\text{exp}}$: only uncorrelated uncertainties due to counting statistics

Measurements	Γ_n / meV	Γ_{γ} / meV	$\rho(\Gamma_n, \Gamma_{\gamma})$
T1	15.06 ± 0.08	121.7 ± 1.3	0.55
T2	14.66 ± 0.30	124.8 ± 3.7	- 0.96
C1	15.31 ± 0.12	118.0 ± 1.4	- 0.50
C2	15.26 ± 0.15	118.9 ± 1.2	- 0.63
T1 + C1	15.14 ± 0.07	120.0 ± 1.0	0.06
T1 + C2	15.10 ± 0.07	120.2 ± 0.9	- 0.29
T1 + T2 + C1 + C2	15.14 ± 0.06	119.8 ± 0.7	- 0.47

Id-number	Measurement	Distance	Angle (flight path – moderator)	Target thickness
T1	Transmission	50 m	9°	10 μm
T2	Transmission	50 m	9°	50 μm
C1	Capture	30 m	0°	5 μm
C2	Capture	12 m	18°	5 μm

thickness (at/b)	sample type	flight path length
$1.40 \cdot 10^{-4}$	solution	25 m
$1.36 \cdot 10^{-4}$	foil	25m
$2.24 \cdot 10^{-4}$	foil	25 m



Only uncorrelated uncertainties

Parameter	p / meV	$\rho(p_i, p_j)$
E_R	178.7 ± 0.1	1.00 0.43 0.79
Γ_n	0.640 ± 0.001	1.00 0.43
Γ_γ	113.5 ± 0.2	1.00

Full Covariance Matrix

Parameter	p / meV	$\rho(p_i, p_j)$
E_R	178.7 ± 0.1	1.00 0.53 0.28
Γ_n	0.640 ± 0.004	1.00 0.25
Γ_γ	113.5 ± 0.2	1.00

$$\mathbf{Z}_1 = F(\mathbf{a}_1, \mathbf{Y}_1)$$

$\mathbf{Z}_1, \mathbf{Y}_1$: dimension n
 \mathbf{a}_1 : dimension k_1

$$\mathbf{V}_{\mathbf{Z}_1} = \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{a}_1} \right) \mathbf{V}_{\mathbf{a}_1} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{a}_1} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Y}_1} \right) \mathbf{V}_{\mathbf{Y}_1} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Y}_1} \right)^T$$

$\mathbf{V}_{\mathbf{a}_1}$ covariance matrix
 (symmetric & positive definite)

$\mathbf{V}_{\mathbf{Y}_1}$ only diagonal terms : $\mathbf{D}_{\mathbf{Y}_1} = \mathbf{V}_{\mathbf{Y}_1}$

$\mathbf{V}_{\mathbf{a}_1} = \mathbf{L}_{\mathbf{a}_1} \mathbf{L}_{\mathbf{a}_1}^T$ (Cholesky decomposition)

$\left(\frac{\partial \mathbf{Z}_1}{\partial \mathbf{Y}_1} \right)$ only diagonal terms

$$\mathbf{S}_{\mathbf{a}_1} = \left(\frac{\partial \mathbf{Z}_1}{\partial \mathbf{a}_1} \right) \mathbf{L}_{\mathbf{a}_1}$$

$$\mathbf{D}_{\mathbf{Z}_1} = \left(\frac{\partial \mathbf{Z}_1}{\partial \mathbf{Y}_1} \right) \mathbf{D}_{\mathbf{Y}_1} \left(\frac{\partial \mathbf{Z}_1}{\partial \mathbf{Y}_1} \right)^T$$

$\mathbf{s}_{\mathbf{Y}_1 u}^2$

$$\mathbf{V}_{\mathbf{Z}_1} = \mathbf{S}_{\mathbf{a}_1} \mathbf{S}_{\mathbf{a}_1}^T + \mathbf{D}_{\mathbf{Z}_1}$$

dimension: $n \times k_1$

n values (diagonal)

$$\mathbf{Z} = F(\mathbf{b}, \mathbf{Z}_1, \mathbf{Z}_2)$$

Z, Z_1, Z_2 : dimension n
 \mathbf{b} : dimension k_b

$$\mathbf{V}_b = \mathbf{L}_b \mathbf{L}_b^T$$

$$\mathbf{S}_b = \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{b}} \right) \mathbf{L}_b$$

$$\mathbf{V}_{Z_1} = \mathbf{S}_{a_1} \mathbf{S}_{a_1}^T + \mathbf{D}_{Z_1}$$

$\dim \mathbf{S}_{a_1} = n \times k_1$

$$\mathbf{V}_{Z_2} = \mathbf{S}_{a_2} \mathbf{S}_{a_2}^T + \mathbf{D}_{Z_2}$$

$\dim \mathbf{S}_{a_2} = n \times k_2$

$$\mathbf{V}_Z = \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{b}} \right) \mathbf{V}_b \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{b}} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right) \mathbf{V}_{Z_1} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right) \mathbf{V}_{Z_2} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right)^T$$

$$\mathbf{V}_Z = \mathbf{S}_b \mathbf{S}_b^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right) \mathbf{S}_{a_1} \mathbf{S}_{a_1}^T \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right) \mathbf{S}_{a_2} \mathbf{S}_{a_2}^T \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right) \mathbf{D}_{Z_1} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right) \mathbf{D}_{Z_2} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right)^T$$

$$\mathbf{S}_Z = \left(\mathbf{S}_b \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right) \mathbf{S}_{a_1} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right) \mathbf{S}_{a_2} \right)$$

$$\mathbf{D}_Z = \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right) \mathbf{D}_{Z_1} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_1} \right)^T + \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right) \mathbf{D}_{Z_2} \left(\frac{\partial \mathbf{Z}}{\partial \mathbf{Z}_2} \right)^T$$

$$\mathbf{V}_Z = \mathbf{S}_Z \mathbf{S}_Z^T + \mathbf{D}_Z$$

dimension: $n \times k$
 $k = k_b + k_1 + k_2$

n values (diagonal)

Conditions :

- (1) Data reduction starts from spectra subject only to uncorrelated uncertainties
- (2) Additional computations using parameters with well defined covariance matrix
- (3) Channel – channel operations (+ , - , x , ÷) and log, exp, ...

Uncertainty propagation results in:

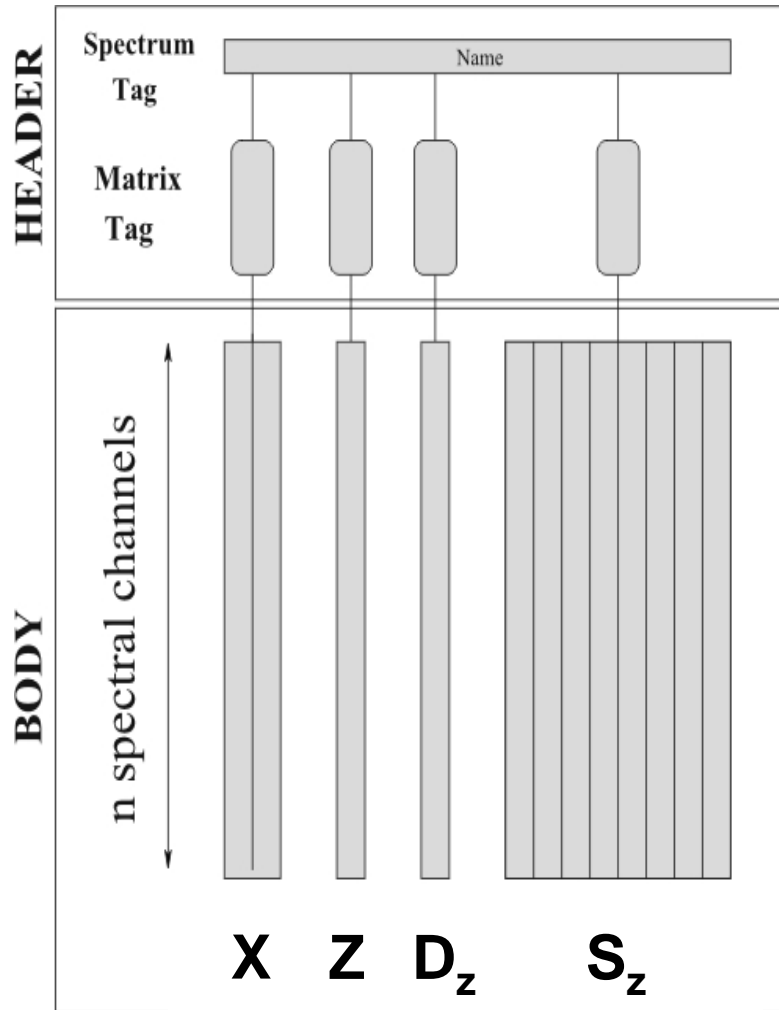
$$V_Z = S_Z S_Z^T + D_Z$$

correlated part
dimension: $n \times k$

uncorrelated part
diagonal : n values

n : length of vector Z

k : number of common sources of uncertainties



Storage

Covariance matrix

n^2 elements (e.g. 32k x 8 bytes

→ 8 Gb !)

AGS representation

$n(k+1)$ elements (32k, 20 corr.

→ 5 Mb)

- **Covariance Matrix are required evaluated data files**
 - Try to produce them from experimental data
- **Requires:**
 - Covariance Matrix from exp. Data
 - Good description of experimental data
- **To reduce storage space**
 - Cholesky decomposition and uncorrelated uncertainties