Resolution of a Gamma Camera: experimental data and analytical formula

Davide Poggiali



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Calendar of seminars

May 14 Introduction to Analytic Tomography

general introduction, Radon tranform, FBP

- May 21 Reconstruction in X-ray computed tomography iterative methods, kernel methods
- May 28 The mathematics behind SPECT/TC reconstruction analytical and iterative hybrid reconstruction
- June 4 Resolution of a Gamma Camera: experimental data and analytical formula

resolution formulas

June 11 An alternative Radon transform for the correction of partial volume effect

a modest proposal..

Resolution and related issues The phantom and its fillings

Analytical Resolution

$$R_c = D\left(1 + \frac{x+c}{L_{eff}}\right)$$

The resolution of the machine depends on the collimator resolution and on the intrinsic resolution (the resolution of the crystal and the electronics).

$$R_s = \sqrt{R_c^2 + R_i^2}$$













Resolution and related issues The phantom and its fillings

The 3-capillaries phantom



We use four different methods for the filling of the phantom:

- only one capillar filled with tracer is scanned (we call this experiment *in air*).
- 2 the central capillar filled with tracer and the rest filled with air (*in plexiglas*).
- **3** the central capillar filled with tracer and the rest filled with water (*in aqua*).

We have the parameters of the machine:

Descript.	L	D	С	t	FWHM	FWHM
	(mm)	(mm)	(mm)	(mm)	@ 0mm	@ 100mm
UHR Irix	58.4	1.78	19	0.152	4.8 mm	6.7 mm

so we can find the analytical resolution $R_s(x) = \sqrt{\alpha x^2 + \beta x + \gamma}$. From the known formula we get

$$\alpha = 0.0010, \ \beta = 0.1468, \ \gamma = 2.4267.$$

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da



A graphic view on the data coming from the 3cap-phantom, an *in water* scan with the central capillary posed at several distances from the collimators. The color refers to the value of the data in the relative pixel.

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da

We can use as mean value of FWHM, the mean value of FWHMs at J different heights

$$mean(FWHM) = \sum_{j \in J} rac{FWHM_j}{|J|}$$

and thus obtaining more accurate results with an estimation of the *absolute error* given by the *standard deviation* (or s.d.) of the results

$$\sigma = \sqrt{\frac{\sum_{j \in J} (FWHM_j - mean(FWHM))^2}{|J|}}$$

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da

The first method is quite simple: let us consider the data at a given heigh j, we call them $(x_i, y_i)_{i=1,...,N}$. We find the maximum for the second value $h = \max(y_i)$ and the relative argument \tilde{x} . We take the two points $z_1 < \tilde{x}$ and $z_2 > \tilde{x}$ which are closer to the half of the maximum and calculate their distance

$$FWHM = |z_1 - z_2|.$$

For this case we have used the following cost

$$C_1(y_1, y_2) = \frac{(y(z_1) - h/2)^2 + (y(z_2) - h/2)^2}{2}$$

This method is very fast, but also very inaccurate.

Now we think our data as they come from a deterministic function $y = f_{\bar{a}}(x)$ plus a small level of noise. Let $(x_i, y_i)_{i=1,...,N}$ be the experimental data at a given height, then we look for

$$\bar{a}^* = \arg\min_{\bar{a}\in\mathbb{R}^n} J(\bar{a}) = \arg\min_{\bar{a}\in\mathbb{R}^n} \|y_i - f_{\bar{a}}(x_i)\|^2$$
.

Since not always f_a depends linearly from the parameters \bar{a} we use an iterative optimization algorithm to estimate \bar{a}^* .

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da

The cost used in with this method is

$$C_2(\bar{a}) = rac{J(\bar{a}^*)}{N}$$

We choose three different functions $f_{\bar{a}}$.

(i) The gaussian

$$f_{\bar{a}}(x) = a_1 e^{-a_2^2 x^2}$$

(ii) The *tempt*

$$f_{\overline{a}}(x) = \max\left((-a_1x + a_2)\chi_{x>0} + (a_1x + a_2)\chi_{x\leq 0}, 0\right)$$

W

(iii) The truncated quadratic

$$f_{\bar{a}}(x) = (a_1 x^2 + a_2)_+ = \max(a_1 x^2 + a_2, 0)$$
 with $a_1 < 0$,

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da



The Gaussian function is the best one.

Now we want to approximate the data $(x_i, y_i)_{i=1,...,N}$ by a function s defined as follows.

- 1 On each subinterval $I_i = [x_i, x_{i+1}], i = 1, ..., N 1, s_{|I_i|} = s_i \in \mathcal{P}^m(\mathbb{R})$ where $\mathcal{P}^m(\mathbb{R})$ is the space of real polynomials of degree $\leq m$.
- 2 $s(x_i) = y_i \ \forall i = 1, \dots, N$, that is f interpolates the data.
- 3 $s_i^{(k)}(x_i) = s_{i+1}^{(k)}(x_i), \quad \forall i = 2, ..., N-1, k \le m-1, m \le N.$ This means that the polynomial pieces are countinuous up to order m-1 in each interior points.

This approach is called *local interpolation* and the functions f are known as (polynomial) *splines* of order m.

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da

As in method 1, we have found two points X_1 and X_2 whose distance from the half of the maximum is minimal. To be more precise we find

$$z_i = \arg\min_{x \in I_i} |s(x) - h/2|$$

with $I_1 = \text{linspace}(x_1, \tilde{x}, 10^4)$ and $I_2 = \text{linspace}(\tilde{x}, x_N, 10^4)$,

 $h = max(y_i)$ and \tilde{x} the argument of the previous maximum, as in Method 1.

The distance of these two points will be a precise estimate of the $\ensuremath{\mathsf{FWHM}}$

$$FWHM = |z_1 - z_2|.$$

The cost is defined as in method 1

$$C_3(z_1, z_2) = \frac{(s(z_1) - h/2)^2 + (s(z_2) - h/2)^2}{2}$$

Method 1: Direct calculation Method 2: Global interpolation Method 3: Local interpolation of the da

From a comparison made on the cost the third method results the most accurate one.

Method	Direct	Gloabal interp	Local interp
Mean cost	$1.5 \cdot 10^{3}$	0.21	0.15
Mean v.c.	9,3%	2.4%	3.0%

Since the collimator resolution is linear with the distance and the Intrinsic Resolution R_i is constant, then the dependence from the distance of the System Resolution can be modelled by a function

$$R_s(x) = \sqrt{\alpha x^2 + \beta x + \gamma}.$$

Now we interpolated the FWHM experimental data using such model and then we compared the fitting curve with the "analytical" one. The fitting curve as been computed as a *weighted least-squares method* (cf. [2]). We considered

$$y^2 = ax^2 + bx + c \,,$$

with y the vector of FWHMs, the vector $(y^2)_j = (y_j^2)$ and x the vector of distances, so that the system is linear on its parameters. We found the parameters p = (a, b, c) by solving the normal equations

$$(V^T \Sigma V) p = V^T \Sigma f$$

where V is the Vandermonde matrix, $f = y^2$ and $\Sigma = diag(1/\sigma_1^2, \ldots, 1/\sigma_n^2)$ the weight matrix, with σ_i the standard deviation (in mm) of the *i*-th resolution value.



Analytical and experimental resolution with different levels of scattering.

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