

# Maximum A Posteriori Reconstruction Algorithms

## Key Points:

- The Prior term creates complications that require approximations or additional steps.
- The algorithm can be applied to only certain forms of the prior.
- MAP reconstructions are smoother than their ML counterparts.
- Convergence can be controlled by the weighting parameter  $\beta$  (Lalush and Tsui, 1992)
- As the value of  $\beta$  is decreased, the degree of smoothing is reduced, and vice versa.
- Type of Priors:
  - Quadratic priors (Fessler and Rogers, 1996)
  - Linearly increasing priors (Green, 1990; Hebert and Leahy, 1992; Lalush and Tsui, 1992, 1993; Mumcuoglu et al., 1994)

## Approaches:

1. One Step Late (OSL) Approach
2. Generalized EM (GEM) Approach
3. PWLS algorithm Approach
4. MAP conjugate gradient (MAP-CG) Approach
5. Poisson-based Pre-conditioned Conjugate Gradient (PCG) Approach

## 1. One Step Late (OSL) Approach

References: (Green 1990)

- The brief logic is to evaluate the derivative term at the previous image estimate
- Results have been shown to converge to the MAP solution for only certain forms of the prior (Lange, 1990)

Ref:

“Implementation and evaluation of a 3D one-step late reconstruction algorithm for 3D positron emission tomography brain studies using median root prior - V. Bettinardi, E. Pagani, M.C. Gilardi, S. Alenius, K. Thielemans, M. Teras, F. Fazio”

Algorithm implementation

Notation is as follows:

$\lambda^n$	Image at the $n$ th iteration
$b, b'$	Voxel index, $1 \leq b \leq B$
$d$	Detector pair index, $1 \leq d \leq D$
$p_{db}$	Probability matrix: probability that annihilation photon pairs emitted from voxel $b$ are detected in the detector pair $d$
$n_d$	Measured counts at detector pair $d$
$\lambda_d$	Mean counts at detector pair $d$
$\lambda_b$	Mean counts at voxel $b$
LOR	Line of response: tube which connects a pair of opposite detectors in coincidence

The OSEM algorithm is:

$$\lambda_b^{n+1} = \frac{\lambda_b^n}{\sum_{d \in D_k} p_{db}} \sum_{d \in D_k} \frac{n_d p_{db}}{\sum_{b' \in B} \lambda_{b'}^n p_{db'}} \quad (1)$$

where  $D_k$  represents a partition of the projection space into  $m$  subsets,  $n = 0, 1, 2, \dots$  is the sub-iteration number and  $k = n \bmod m$ . A cycle of  $m$  sub-iterations constitutes a complete iteration.

Consider the OSL algorithm in its general form:

$$\lambda_b^{n+1} = \frac{\lambda_b^n}{\sum_{d \in D} p_{db} + \beta \frac{\partial}{\partial \lambda_b} U(\lambda, b) |_{\lambda=\lambda^n}} \sum_{d \in D} \frac{n_d p_{db}}{\sum_{b' \in B} \lambda_{b'}^n p_{db'}} \quad (2)$$

where  $U(\lambda, b)$  is a potential function while  $\beta$  is the weight factor.

The MRP algorithm can be formulated using the median as a penalty reference:

$$\beta \frac{\partial}{\partial \lambda_b} U(\lambda, b) |_{\lambda=\lambda^n} = \beta \frac{\lambda_b^n - M_b(\lambda^n)}{M_b(\lambda^n)} \quad (3)$$

where  $M_b$  is the median in a  $3 \times 3 \times 3$  mask width of neighbourhood voxels centered at voxel  $b$ . The penalty is set if the voxel  $\lambda_b$  is different from  $M_b$ .

## **2. Generalized EM (GEM) Approach**

References: (Hebert and Gopal, 1992)

- Brief Logic:
  - Sequentially update pixels
  - Verify that each update increases the posterior density
  - Make sure that the convergence is to a maximum of the Posterior density

## **3. PWLS algorithm Approach**

References: (Fessler, 1994)

- Brief Logic:
  - Utilize a Co-ordinate Descent Approach
  - Uses a quadratic prior to solving explicitly for the optimal step size  $t$
- The algorithm is efficient (Usually requires approximately 10–15 iterations).
- But requires a matrix-based approach and restricts priors to only quadratic forms.

## **4. MAP conjugate gradient (MAP-CG) Approach**

References: (Lalush and Tsui, 1995)

- More general priors and projector-based models can be accommodated.
- The algorithm also requires approximately 10–15 iterations.
- Complex to code as needed to perform a local linear fit of the prior term in the step size calculation and has no nonnegativity constraint.

## **5. Poisson-based Pre-conditioned Conjugate Gradient (PCG) Approach**

References: (Mumcuoglu et al., 1994)

- Brief Logic:
  - Line searching to optimize the step size.
- Does not have the convergence problems of the OSL method.