

# NEW APPLICATIONS OF ITERATIVE SCALING IN TRANSMISSION TOMOGRAPHY

---

Hudson

Macquarie Univ.

Arghami

Ferdowsi Univ.

Larkin

Macquarie Univ.

- TRANSMISSION:  $E(y) = \mu^0 \exp(-Ax)$

Objective - reconstruct  $x$  by ML.  $J \times 1$

- GIS = SMART APPLIED TO  
A DUAL-SYSTEM.

A key multiplicative form of solution.  
What about  $x$ ?

- ACCELERATION VIA OS  
Subset balance

- OTHER PRACTICAL ISSUES  
Avoiding exponentiation

SMART in multiplicative form:

$$x_j^{k+1} = x_j^k \prod_i \left( \frac{y_i}{\hat{y}_i^k} \right)^{\alpha_{ij}}$$

In transmission : we could apply SMART to  $\tilde{y} = -\log(y/\mu_0)$

(Method A)

This is NOT a ML-solution  
Equations are INCONSISTENT  
Noise amplified at low counts.

The solution retains a MULTIPLICATIVE FORM  
 $\hat{x}_j = x_j^0 \prod_i \hat{s}_i^{\alpha_{ij}}$

## METHOD 1 :

Apply SMART to linear system

for

$$\tilde{y} = -\log(y/\mu^0)$$

{blank scan vals.}

---

- NOT a ML-solution
  - The linear system is INCONSISTENT
  - This method amplifies noise at low counts.
- 

## METHOD 2 : Generalized Iterative Scaling

Apply SMART to a DUAL linear system

for

$$g = A^T y$$

---

- Start point must be  $\mu^0$
  - The linear system is CONSISTENT
  - The method provides a ML solution and thus is efficient.
-

KNOWN RESULTS - Problem  $T$ 

*Only previous work providing algebraic solution of ML*

1. Lange, 1984

An EM algorithm for transmission requires approximate solution (M-step).

2. Darroch & Ratcliffe, 1972

Consider solutions

$$A^T \mu = A^T y,$$

$$\text{Dual Equations} \quad (5)$$

*consistent equations in unknowns  $\mu$  which are restricted to the multiplicative form*

$$\mu_i = \mu_i^0 \prod \eta_j^{a_{ij}}.$$

$$(6) \quad -\log(\text{survival fraction})$$

A unique solution of system (5) of the multiplicative (log-linear) form (6) exists and may be recovered by generalized iterative scaling.

*NB we are solving for ~~unknowns~~ linear equations in ~~unknowns~~  $\mu$  — the ~~very~~ sums of other sums.*

# GENERALIZED ITERATIVE SCALING

- Initial computations:  
 $\hat{g}^k = A^T y$  (backward project)  
 $\mu^0 \rightarrow$  expected blank scan  
 $\eta^0 \rightarrow 1$ ,

Iterations:

$$\hat{g}^k = A^T \mu^k$$

$$\begin{aligned}\mu^{k+1} &= \mu^k - \prod_{j=1}^n \left( \frac{\alpha_{ij}}{\hat{g}_{ij}^k} \right) \alpha_{ij} \\ \eta_{j,k+1} &= \eta_{j,k} \left( \frac{\alpha_{ij}}{\hat{g}_{ij}^k} \right)\end{aligned}$$

**NB**  $i \rightarrow j$ ,  $j \rightarrow i$ :  
variables / rays  
unknowns in linear system  
 $x_j \leftarrow \mu_i$

in DUAL System:

## GIS for TRANSMISSION

The GIS/SMART solution from start-point  $\mu^0$  always retains the multiplicative form

$$\mu_i = \mu_i^0 \prod \eta_j^{a_{ij}}.$$

which respects the tomographic model

$$\mu = \mu^0 \exp(-Ax),$$

with  $\eta_j = \exp(-x_j)$ .

Thus we advocate seeking a solution of the dual system of equation (5) employing GIS, or SV-EM, algorithms to provide the solution  $\mu$  for transmission tomography!

We want  $x$ , not  $\mu$ , though. See Arghami and Hudson.

A further issue, acceleration ... by ordered subsets!

*What of acceleration by subset methods.*

### THEOREM

*For projection matrix  $A$  with row sums equal to 1, accelerated GIS  $\{\mu^m\}$  converges to the unique solution of the Normal equations for which  $KL(\mu, \mu^0)$  is minimized. The solution may be expressed in the product form required for Transmission.*

The product form of solution is retained, . . . a key point for solution in Transmission.

*Which subsets of pixels?*

*. . . 'subset balance'.*

## FURTHER CONSIDERATIONS

SMART / GIS requires exponentiation on every  $a_{ij} > 0$ .  
SV-EM does NOT, AND provides a similar solution.

↑ Solve DUAL systems from start by using  
SV-EM (or accelerate by OS-EM )  
to provide solutions  $\hat{\mu}$ .

Since the solution may not have PRODUCT FORM  
search for the solution  $x$  of  
 $\hat{\mu} = \mu^0 \exp(-Ax)$

$$\hat{\mu}_j = \log \frac{\hat{\mu}}{\hat{\mu}^0} \approx -\log \frac{\hat{\mu}}{\hat{\mu}^0}$$

$\hat{\mu}^0$  should have much  
reduced noise

~~SMART~~  
if it has a night sky property in consistent systems

and if we solve ~~2~~ DUAL equations  $J$  eqns in  $I$  unknowns  
 $(J < I)$

then we expect ~~at most~~ ~~at least~~  
 $J-1$  non-zero fitted  $\mu_i$ .

The remaining rays are estimated to have  $\infty$  attenuat  
~~#~~ cumulative hazard.

Avoiding exponentiation:

- SMART/BIS requires exponentiation on every non-zero  $a_{ij}$   
(problem considered, <sup>comput.</sup> problem by Fessler)
- SV-EM does NOT, <sup>AND</sup> provides a similar solution  
to SMART

⇒ 1. SOLVE DUAL SYSTEM FROM START  $\mu^0$   
USING SV-EM / OS-EM (FAST!!)  $\rightarrow \hat{\mu}$

2. If the solution is not ~~quite~~ of PRODUCT FORM <sup>(quite)</sup>

$$\check{y} = \frac{(A^T y)_j}{(A^T \hat{\mu})_j} \quad o = \sum_j a_{ij} \log \frac{(A^T y)_j}{(A^T \hat{\mu})_j} \quad \forall i$$

Now search for solution  $x$  of  $\hat{\mu} = \mu^0 \exp(-Ax)$ :

a. FBP

b. OS-EM on  $\log \hat{y} = -\log \frac{\hat{\mu}}{\mu^0} = Ax$  (much reduced  
noise  $\hat{y}$   
in system)

~~SMART~~ (requires logs just once).

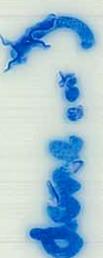
## CONCLUSION

GITS algorithm in Transmission applies a very similar approach to Dual equations that SV-EM algorithm applies to forward equations in Emission.

Acceleration of both algorithms is affected by Ordered Subsets.

With GITS provides convergence

Each iteration reduces a  $\chi^2$ -distance ... faster convergence.

With effective subsets 

GITS would provide a general purpose algorithm perhaps as effective as SV-EM.

PLAN

1. Linear systems
2. Two algorithms for their solution
3. Models for Medical Imaging
4. Known results for consistent and inconsistent linear systems
5. A new approach in Transmission Tomography

Symmetry       $\text{EM} / \text{SMART}$   
 $\text{OSRM} / \text{MAPF}$

MPL / Bayesian