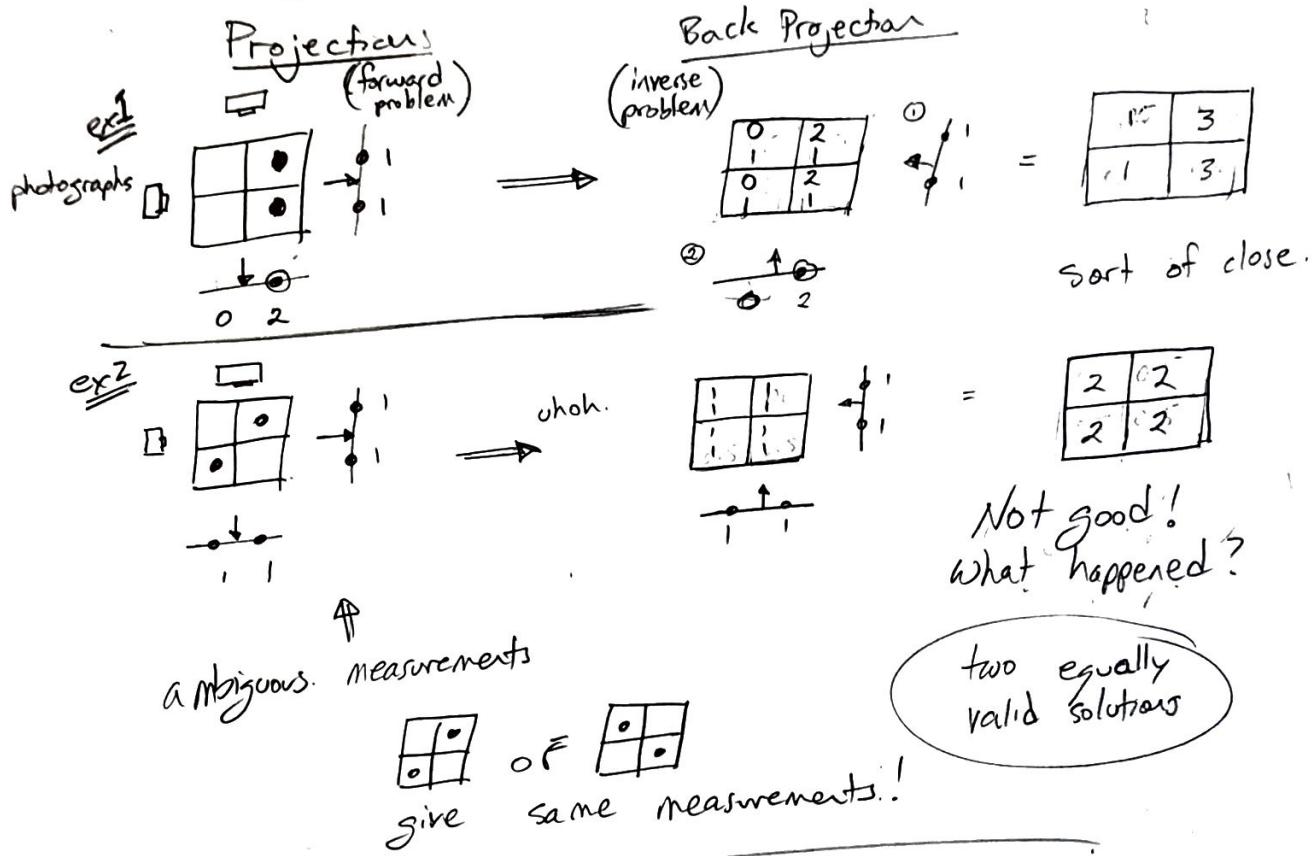


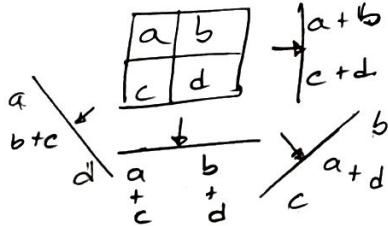
Pictures and CT and, eventually EIT
Electrical Impedance Tomography

Apr 4, 2018 ①

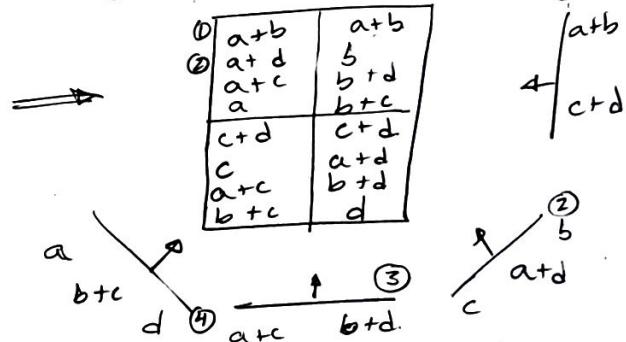


How can we solve the ambiguity?

More angles

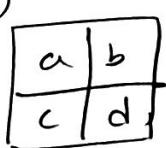


④ back project

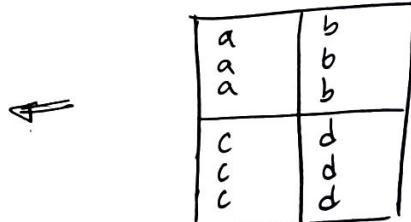


⑤ divide by projection minus one.

~~(4-1)~~



⑥ subtract projection sum: $-(a+b+c+d)$



Will MANY more angles make a perfect picture?

No.: in the limit (continuous data), we still have measurement noise // and imperfect reconstruction algorithms.

Recall For CT, one approach is filtered back projection (FBP)

FBP

$$f(x, y) = \int_0^{\pi} d\phi \left[\int_{-\infty}^{\infty} dw |w| P(w) e^{2\pi j ws} \right] \quad s = x \cos \phi + y \sin \phi$$

1) FT of projection $p(s, \phi) \xrightarrow{\text{FT}} P(w) e^{j 2\pi ws}$

2) Multiply by frequency filter: ramp $|w|$



- minimize blurring (low freq.)

- maximize contrast (high freq.).

3) Inverse FT

4) Back project

5) Sum over all filtered back projections.

→ complete data, no noise + FBP = original 2D image.

- Error Sources
 - discrete data
 - measurement noise (statistical errors)
 - systematic errors (detector misalignment)
 - * - incomplete data
 - * - scattering, absorption

Another way: Maximum Likelihood, Expectation Maximization.

or Gauss - Newton (Iterative)

$$\begin{aligned} \text{fwd. } M &= Jx \\ J^T M &= (J^T J)x \\ (J^T J)^{-1} J^T M &= (J^T J)^{-1} (J^T J)x \\ (J^T J)^{-1} J^T M &= x \quad \underline{x} \end{aligned}$$

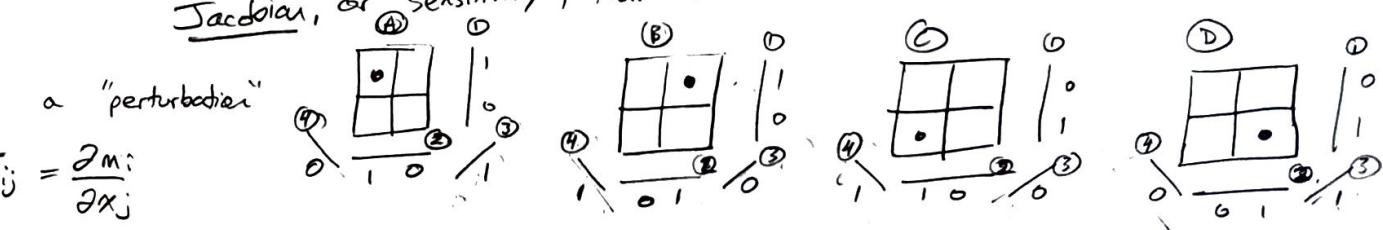
$$x_{n+1} = x_n - (J^T J)^{-1} J^T b_n \quad ; \quad b_n = F(x_n) - M \quad ; \quad x_0$$

↓ next estimate ↑ last estimate ↓ pseudo-inverse $(J^T J)^{-1}$ ↓ error ↓ forward model ↓ measurements ↑ initial guess.

↓ Jacobian

... for PET, SPECT, CT, EIT

Jacobian, or sensitivity, from forward model



$$J = \begin{bmatrix} ④ & ⑤ & ⑥ & ⑦ \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix} \quad \{ ① \} \quad \{ ② \} \quad \{ ③ \} \quad \{ ④ \}$$

ex. $M_A = J x_A = f(x_A)$

for $x_A = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$

*Remember linear algebra?

Try it!

Initial guess: $x_0 = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \emptyset$; Measurements A, solution A

"empty" $x_1 = x_0 - \underbrace{(J^T J)^{-1} J^T b_n}_{x_1 = x_0 - (J^T J)^{-1} [J^T (Jx_0 - m_A)]}$; $m_A = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} = Jx_A$

$$x_1 = x_0 - (J^T J)^{-1} J^T (Jx_0 - m_A) + (J^T J)^{-1} J^T m_A$$

$$x_1 = \underbrace{\alpha (J^T J)^{-1} J^T J x_A}_{x_1 = \alpha \underset{I}{\circ} \underset{x_A}{\circ}}$$

\rightarrow next iteration error = \emptyset , stop!

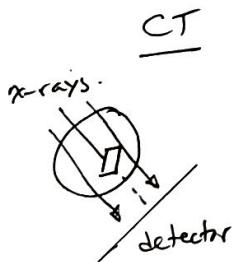
BUT noise + errors means $m_A \neq Jx_A$,
iterations minimize the error

Back projection is $x = J^T m_A$

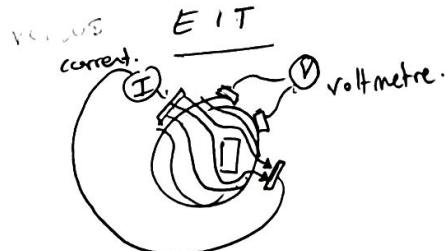
Filter is $J^T J = \begin{bmatrix} 3 & 1 & 1 & 1 \\ 1 & 3 & 1 & 1 \\ 1 & 1 & 3 & 1 \\ 1 & 1 & 1 & 3 \end{bmatrix}$ spatial low pass filter

$$(J^T J)^{-1} \approx \frac{1}{10} \begin{bmatrix} 4 & -1 & -1 & -1 \\ -1 & 4 & -1 & -1 \\ -1 & -1 & 4 & -1 \\ -1 & -1 & -1 & 4 \end{bmatrix}$$
 spatial high pass filter

\star this is type 2 filtered back projection.



\star density



\star conductivity

\star we made it,
on to the pictures!

- individual x-rays
go in a straight line.

- current flows everywhere
- small changes in measurements
can indicate large changes
in conductivity

- measurement noise -

for EIT $x_{n+1} = x_n + (J^T W J + \lambda^2 R)^{-1} (J^T W b_n + \lambda^2 R(x_* - x_n))$

$\underbrace{J^T W J}_{\text{inverse measurement covariance}} + \underbrace{\lambda^2 R}_{\text{regularization.}} \underbrace{(J^T W b_n + \lambda^2 R(x_* - x_n))}_{\text{prior estimate.}}$