

# Statistical Methods for Image Reconstruction

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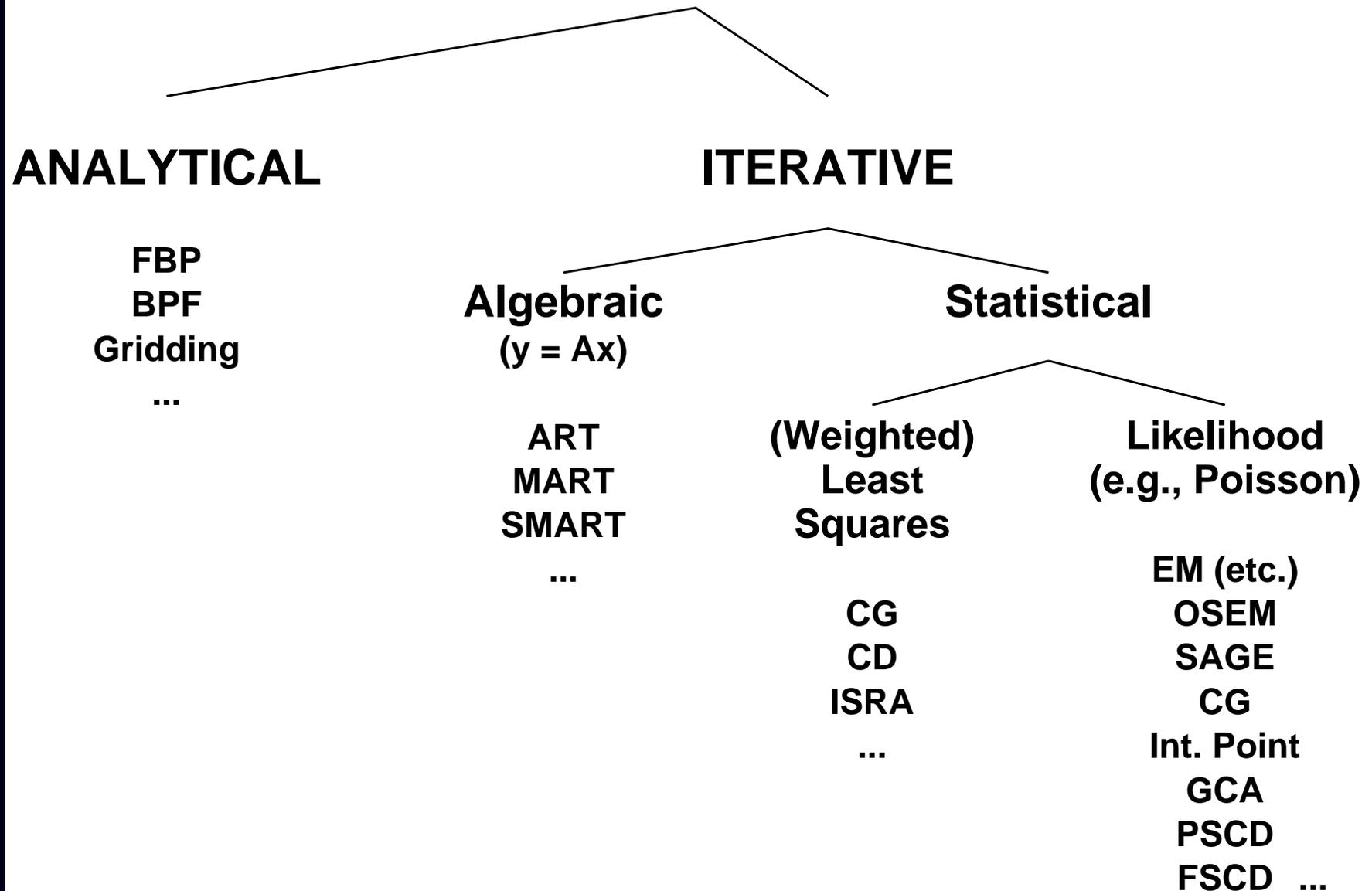
# Image Reconstruction Methods

## (Simplified View)

**Analytical**  
(FBP)

**Iterative**  
(OSEM?)

# Image Reconstruction Methods / Algorithms



# Outline

## Part 0: Introduction / Overview

## Part 1: From Physics to Statistics (Emission tomography)

- Assumptions underlying Poisson statistical model
- Emission reconstruction problem statement

## Part 2: Four of Five Choices for Statistical Image Reconstruction

- Object parameterization
- System physical modeling
- Statistical modeling of measurements
- Cost functions and regularization

## Part 3: Fifth Choice: Iterative algorithms

- Classical optimization methods
- Considerations: nonnegativity, convergence rate, ...
- Optimization transfer: EM etc.
- Ordered subsets / block iterative / incremental gradient methods

## Part 4: Performance Analysis

- Spatial resolution properties
- Noise properties
- Detection performance

## Part 5: Miscellaneous topics (?)

- ...

# History

- Iterative method for X-ray CT (Hounsfield, 1968)
- ART for tomography (Gordon, Bender, Herman, JTB, 1970)
- Richardson/Lucy iteration for image restoration (1972, 1974)
- Weighted least squares for 3D SPECT (Goitein, NIM, 1972)
- Proposals to use Poisson likelihood for emission and transmission tomography  
Emission: (Rockmore and Macovski, TNS, 1976)  
Transmission: (Rockmore and Macovski, TNS, 1977)
- First expectation-maximization (EM) algorithms for Poisson model  
Emission: (Shepp and Vardi, TMI, 1982)  
Transmission: (Lange and Carson, JCAT, 1984)
- First regularized (aka Bayesian) Poisson emission reconstruction  
Geman and McClure, ASA, 1985
- Ordered-subsets EM algorithm  
Hudson and Larkin, TMI, 1994
- Commercial introduction of OSEM for PET scanners  
circa 1997

# Why Statistical Methods?

- Object constraints (*e.g.*, nonnegativity, object support)
- Accurate physical models (less bias  $\Rightarrow$  improved quantitative accuracy)  
improved spatial resolution?  
(*e.g.*, nonuniform attenuation in SPECT)
- Appropriate statistical models (less variance  $\Rightarrow$  lower image noise)  
(FBP treats all rays equally)
- Side information (*e.g.*, MRI or CT boundaries)
- Nonstandard geometries (“missing” data)

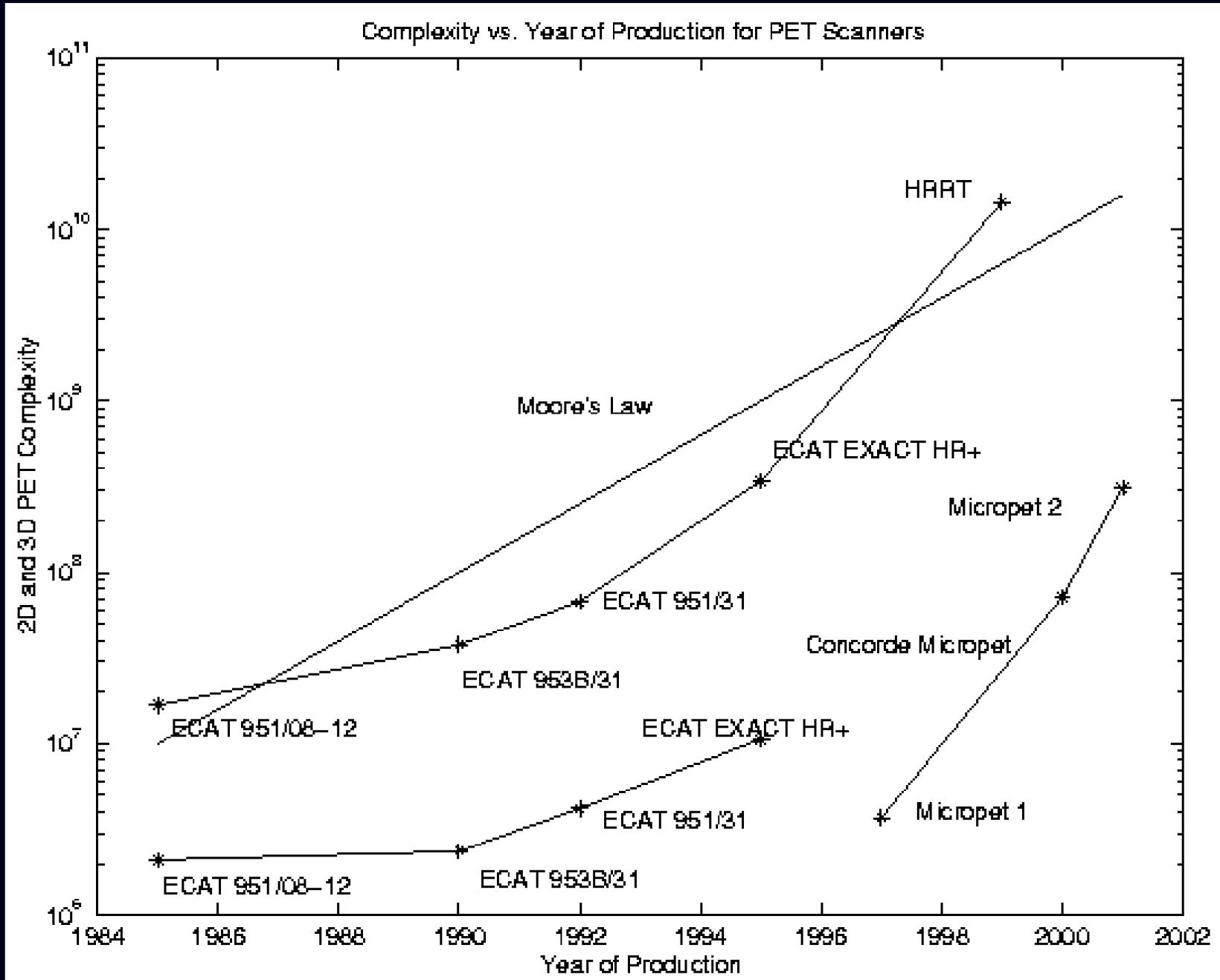
## Disadvantages?

- Computation time
- Model complexity
- Software complexity

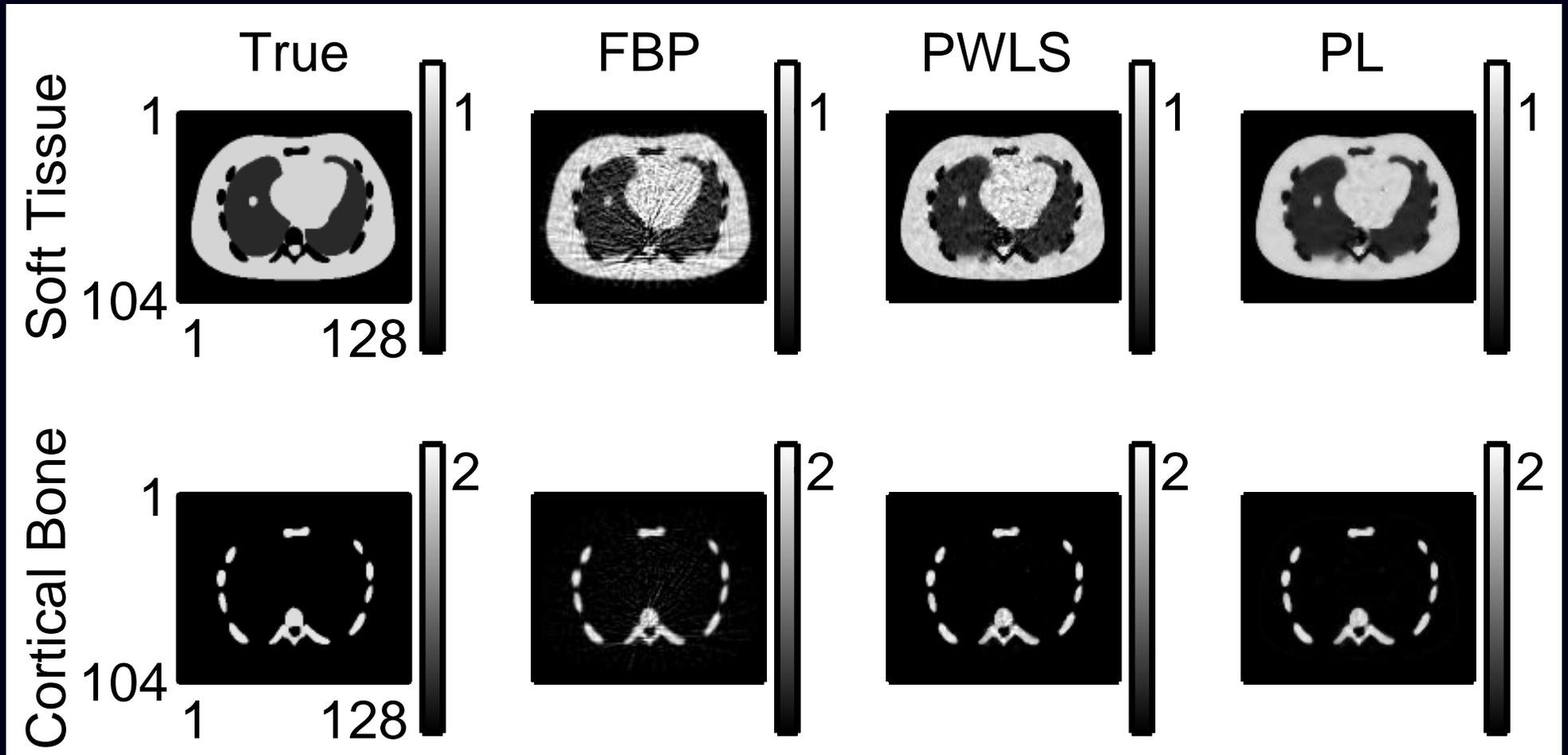
## Analytical methods (a different short course!)

- Idealized mathematical model
  - Usually geometry only, greatly over-simplified physics
  - Continuum measurements
- No statistical model
- Easier analysis of properties (due to linearity)  
*e.g.*, Huesman (1984) FBP ROI variance for kinetic fitting

# What about Moore's Law?



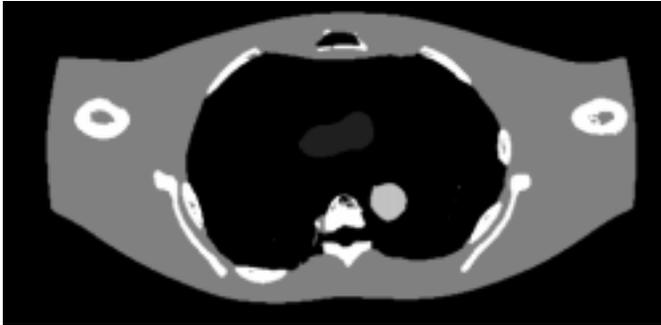
# Benefit Example: Statistical Models



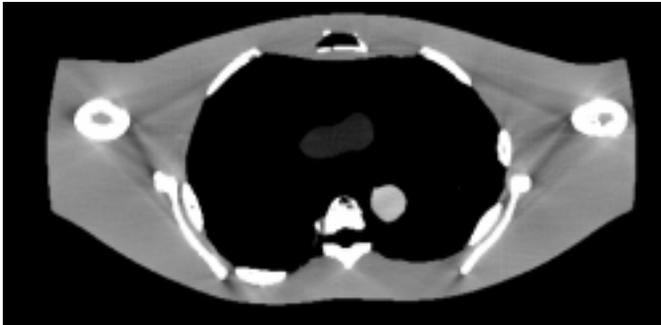
Method	NRMS Error	
	Soft Tissue	Cortical Bone
FBP	22.7%	29.6%
PWLS	13.6%	16.2%
PL	11.8%	15.8%

# Benefit Example: Physical Models

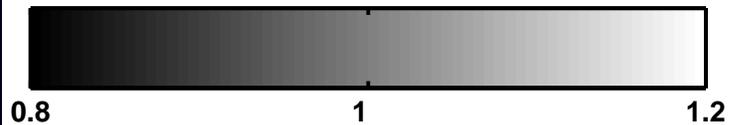
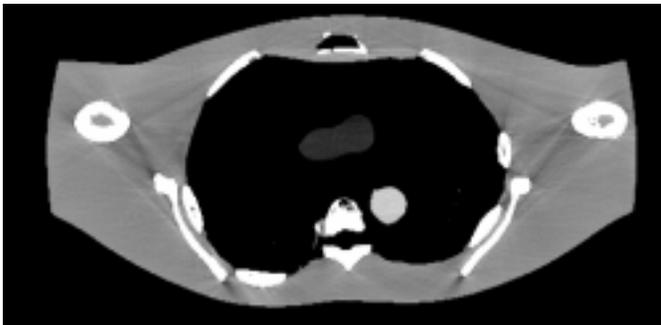
a. True object



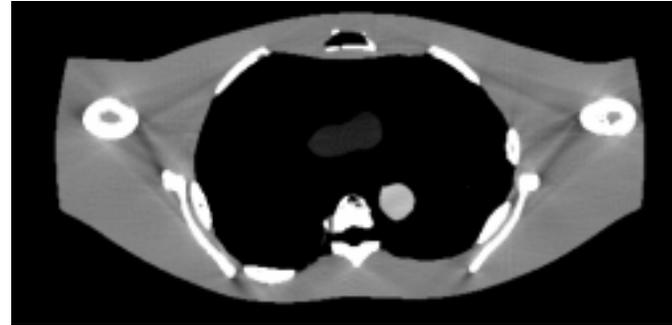
b. Unocorrected FBP



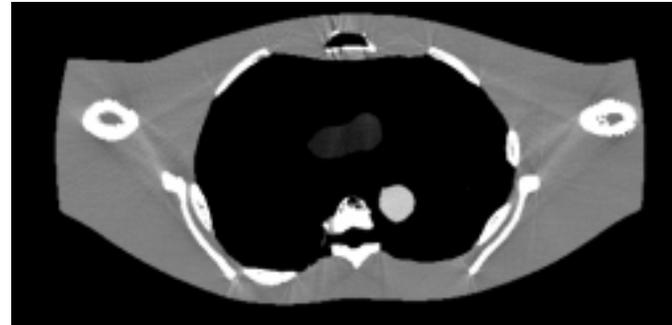
c. Monoenergetic statistical reconstruction



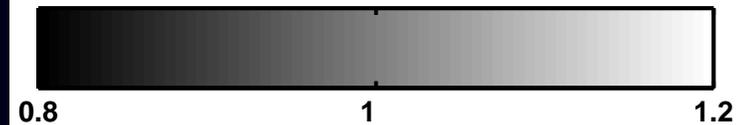
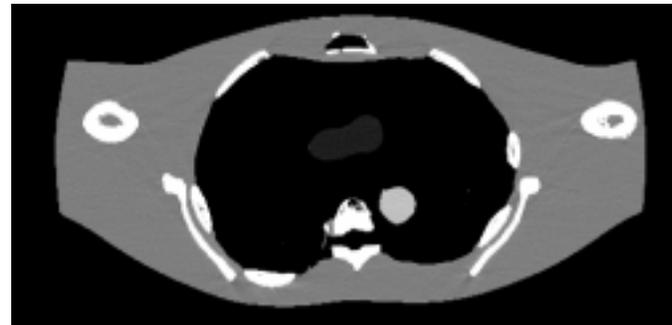
a. Soft-tissue corrected FBP



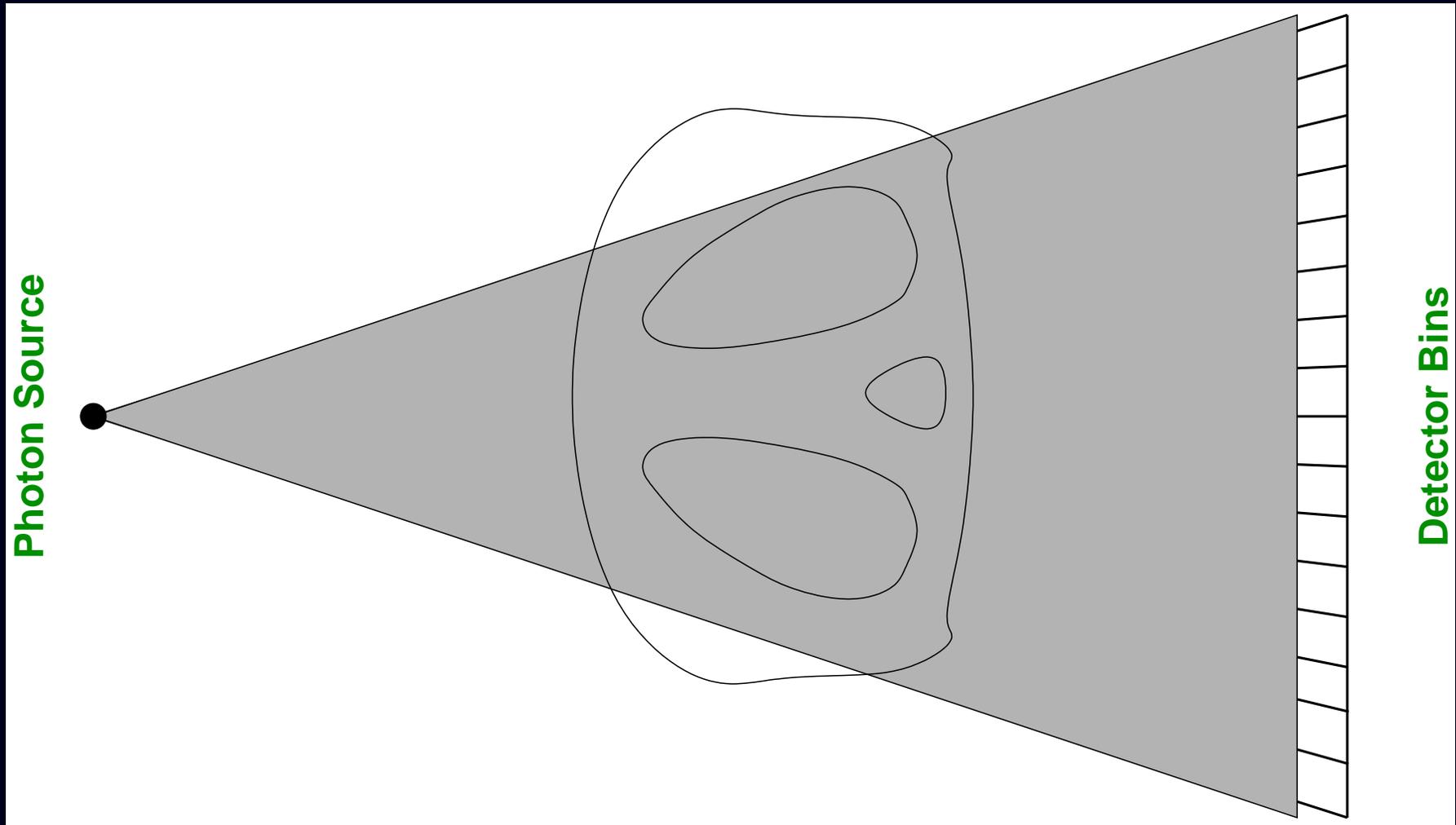
b. JS corrected FBP



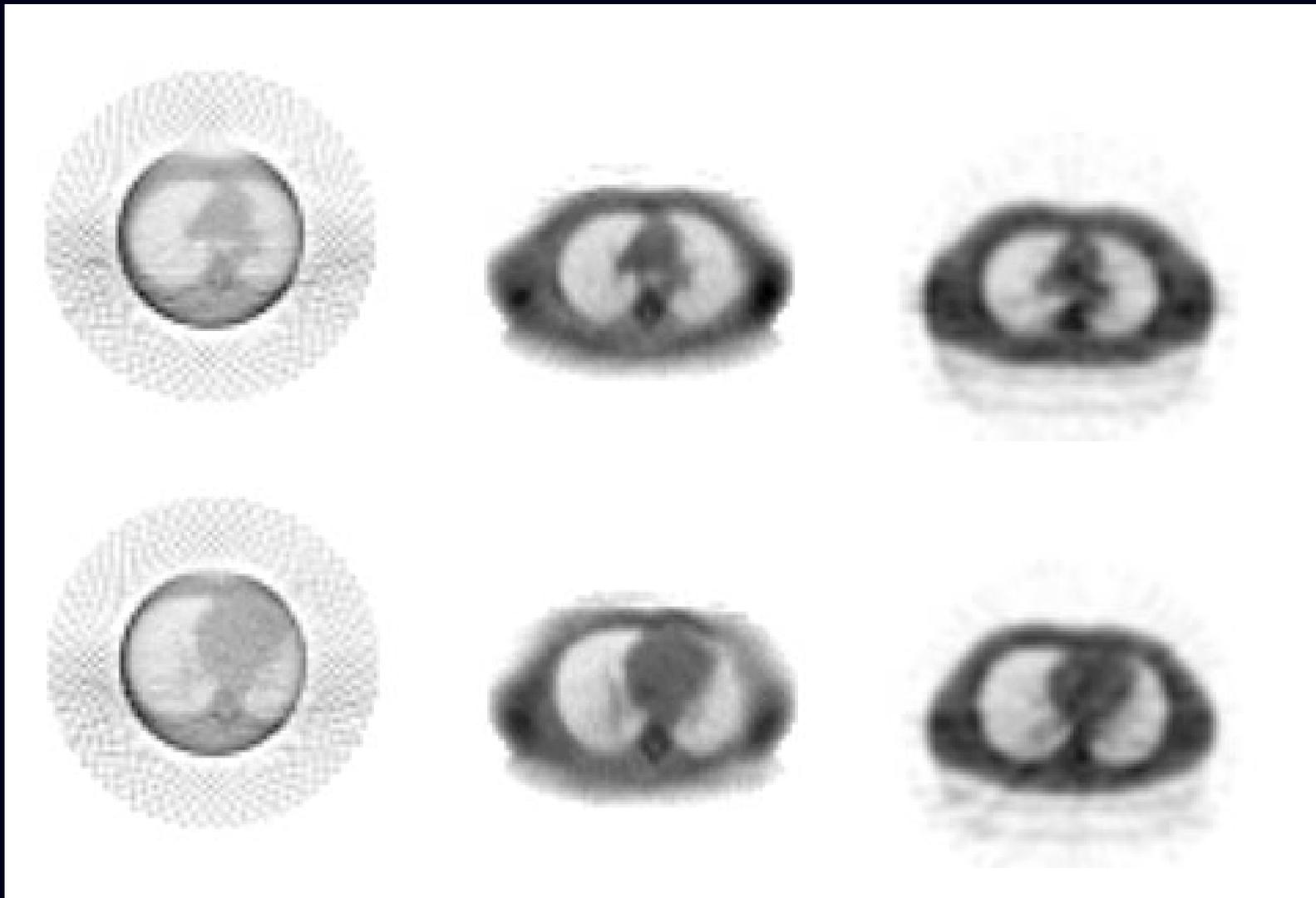
c. Polyenergetic Statistical Reconstruction



# Benefit Example: Nonstandard Geometries



# Truncated Fan-Beam SPECT Transmission Scan



Truncated  
FBP

Truncated  
PWLS

Untruncated  
FBP

# One Final Advertisement: Iterative MR Reconstruction

