

# Gamma-Ray Tomography for Field Application

Chemical Reaction Engineering Laboratory (CREL)

Z. Kuzeljevic, R. Varma, E. H. Stitt, M. Al-Dahhan, M. Dudukovic

Department of Energy, Environmental, and Chemical Engineering (WUSTL)



## Introduction

❖ In field application one is not interested in detailed flow description; basic idea of presence/absence of flow mal-distribution often will suffice

❖ Challenges: restriction of access to the vessel; the need to adjust CT setup to the column external piping and instrumentation; need to carry out studies at significant heights on a large diameter column with a high attenuation media; reduction in scan time; need for portable, easily to installed and removed system

❖  $\gamma$  -ray tomography, which is suitable for opaque, high attenuation media and steady state flow conditions seems to be most appropriate technique for field application (Darwood et al., 2003)

## Focus of This Work

❖ Achieve: - reduction in scan time

- guidelines for the best data collection methodology

❖ Key Questions to Answer:

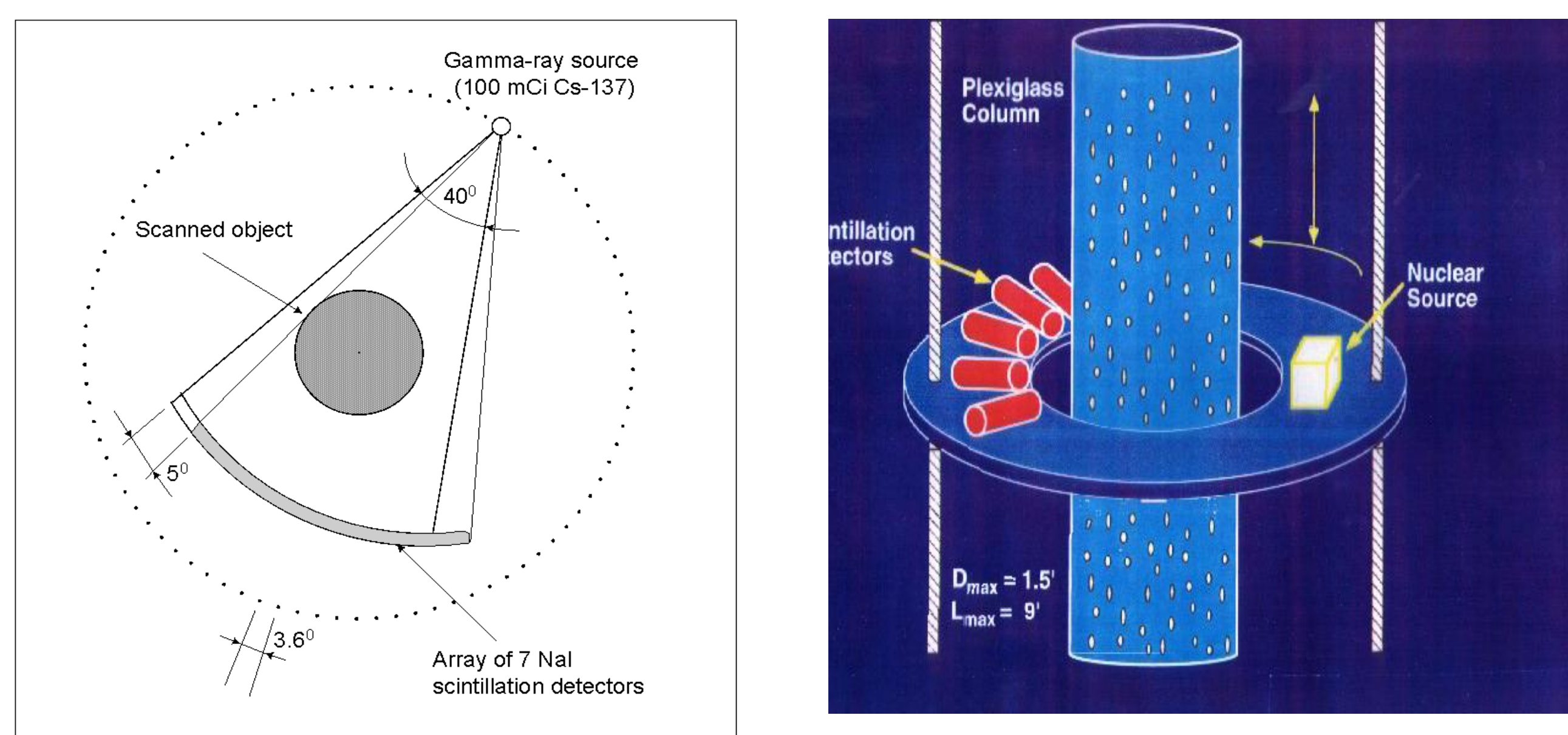
- What is the correlation between image quality and the amount of data used in the process of reconstruction?

- What is the optimum way to obtain the data in order to minimize the introduced error?

## Gamma-ray tomography

❖ Process of imaging of cross-sectional area (i.e. obtaining phase distribution) consists of scanning and reconstruction

❖ During a scan, the column is placed between gamma-ray source and an arrangement of scintillation detectors

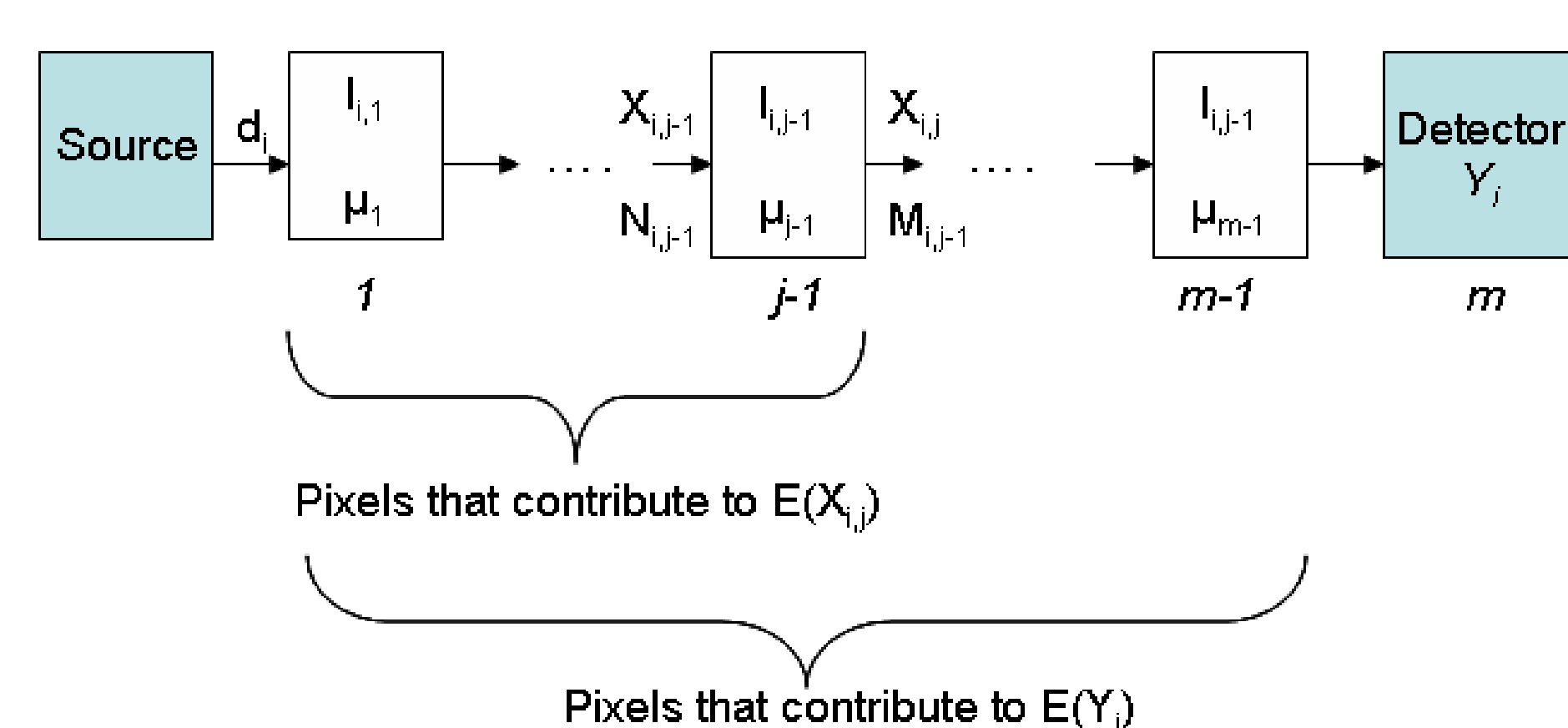


❖ During the scan, the source makes a full circle around the scanned object in step increment of  $3.6^\circ$  and the term **view** is used to indicate fixed source position

❖ **Projections** are different positions of an assembly of detectors towards the source - in any given view, detectors take 25 different positions, in increments of  $3.6^\circ$

## Reconstruction Algorithms

❖ Attenuation - Beer-Lamber's law:  $I = I_0 \cdot \exp(-\mu \cdot l)$



## Expectation-Maximization Algorithm

$$E(Y_i) = E(X_{i,j}) \exp\left(-\sum_{k=j}^{m-1} l_{i,k} \mu_k\right) \quad E(X_{i,j}) = d_i \exp\left(-\sum_{k=1}^{j-1} l_{i,k} \mu_k\right)$$

$$f(X_{i,j} | Y_i) = \exp\left[-\lambda \left( \frac{X_{i,j} - E(Y_i)}{\sqrt{X_{i,j} - Y_i}} \right)^2\right]$$

$$E(X_{i,j} | Y_i) = Y_i + E(X_{i,j}) - E(Y_i)$$

## Alternating-Minimization Algorithm

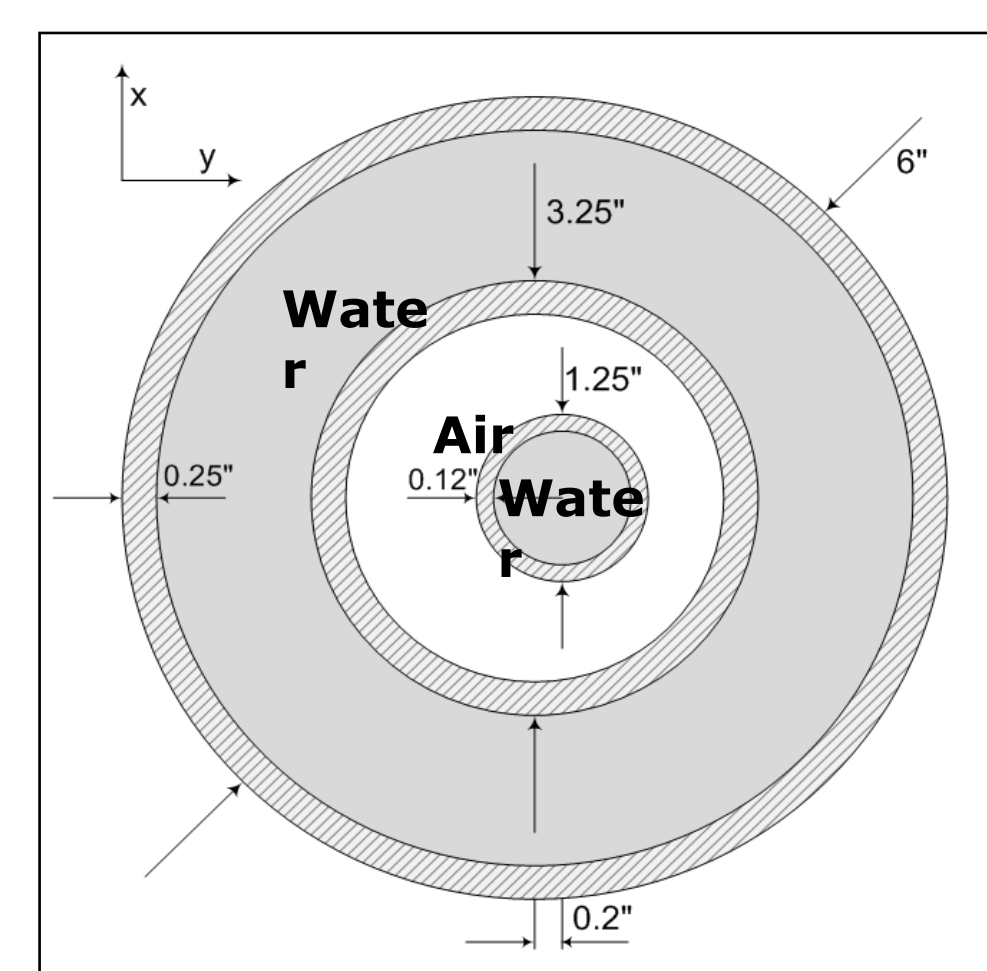
$$l(Y | q) = \sum_i \left[ Y_i \ln\left(\frac{Y_i}{q_i}\right) - \left( Y_i - q_i \right) \right]$$

$$\mu_j^{(n+1)} = \mu_j^{(n)} - \frac{1}{Z_j} \ln \frac{d_j'}{q_j} \quad d_j' = \sum_{i \in J_j} l_{i,j} d_i$$

$$q_j = \sum_{i \in J_j} l_{i,j} q_i$$

## Test Object and Error Quantification

❖ To examine relation between amount of data used and image quality, test object has been scanned



❖ The error has been quantified then as the average deviation of the binary gated attenuation coefficient value in the image of the reduced scan from the theoretical values

$$error = 100 \cdot \frac{\sum_i abs(\varepsilon_{i,full} - \varepsilon_{i,red}) / \varepsilon_{i,full}}{N_{pixel}}$$

## Investigation plan

### Amount of data used

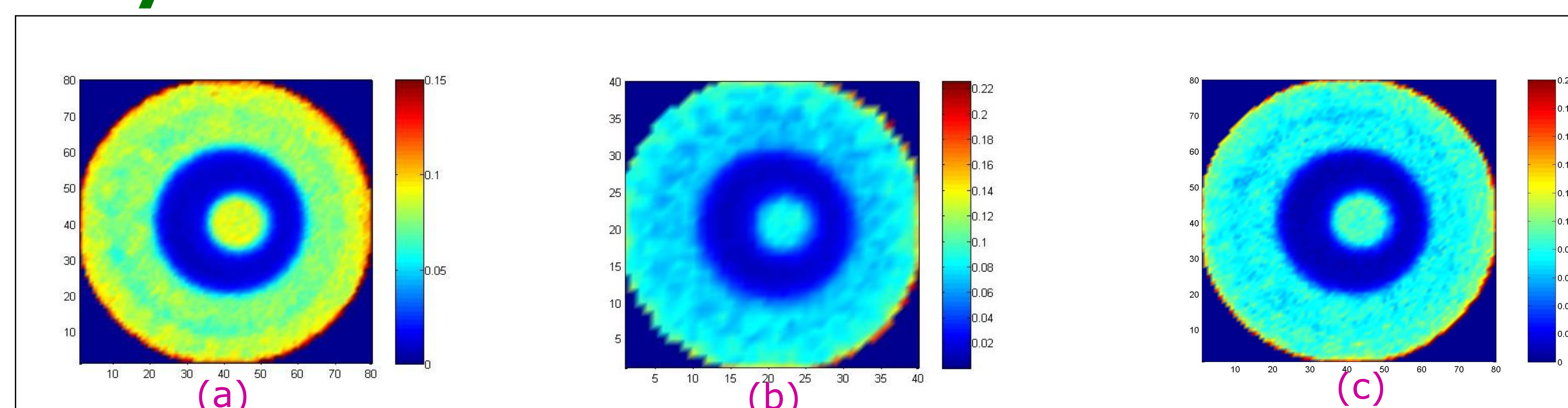
Number of pixels	Data used, % of full set	Views	Projections per view
80x80 60x60 40x40	5	5	25
		10	13
		14	9
		18	7
	10	10	25
		20	13
		28	9
		36	7
	25	25	25
		48	13
		69	9
		89	7
50	50	25	
	96	13	

### Algorithms:

❖ Expectation-Maximization (EM)

❖ Alternating Minimization (AM)

## Key Results



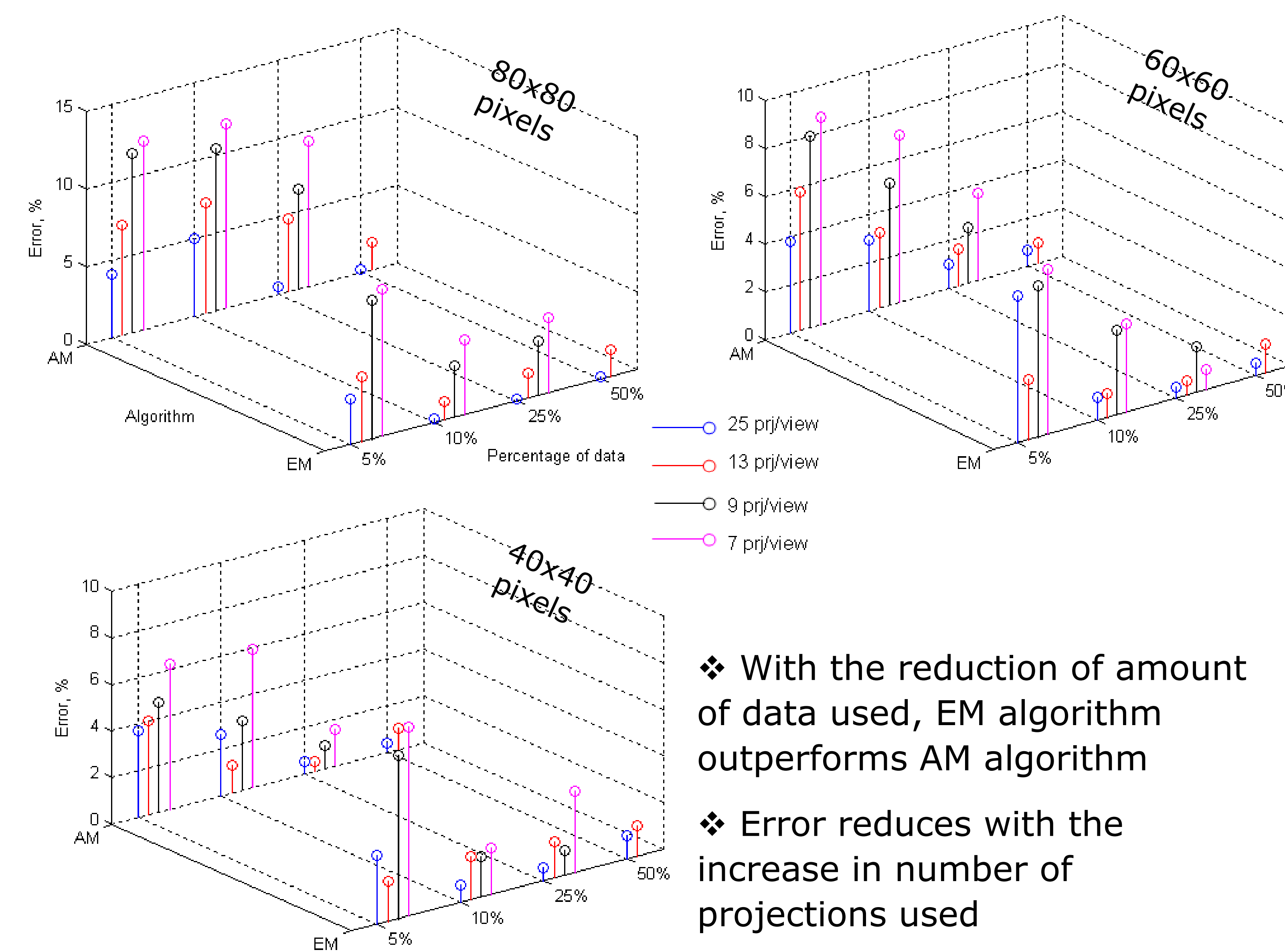
### Image obtained with:

(a) full data set and 80x80 pixels

(b) 25% of full scan data used and 40x40 pixels

(c) 50% of full scan data used and 40x40 pixels

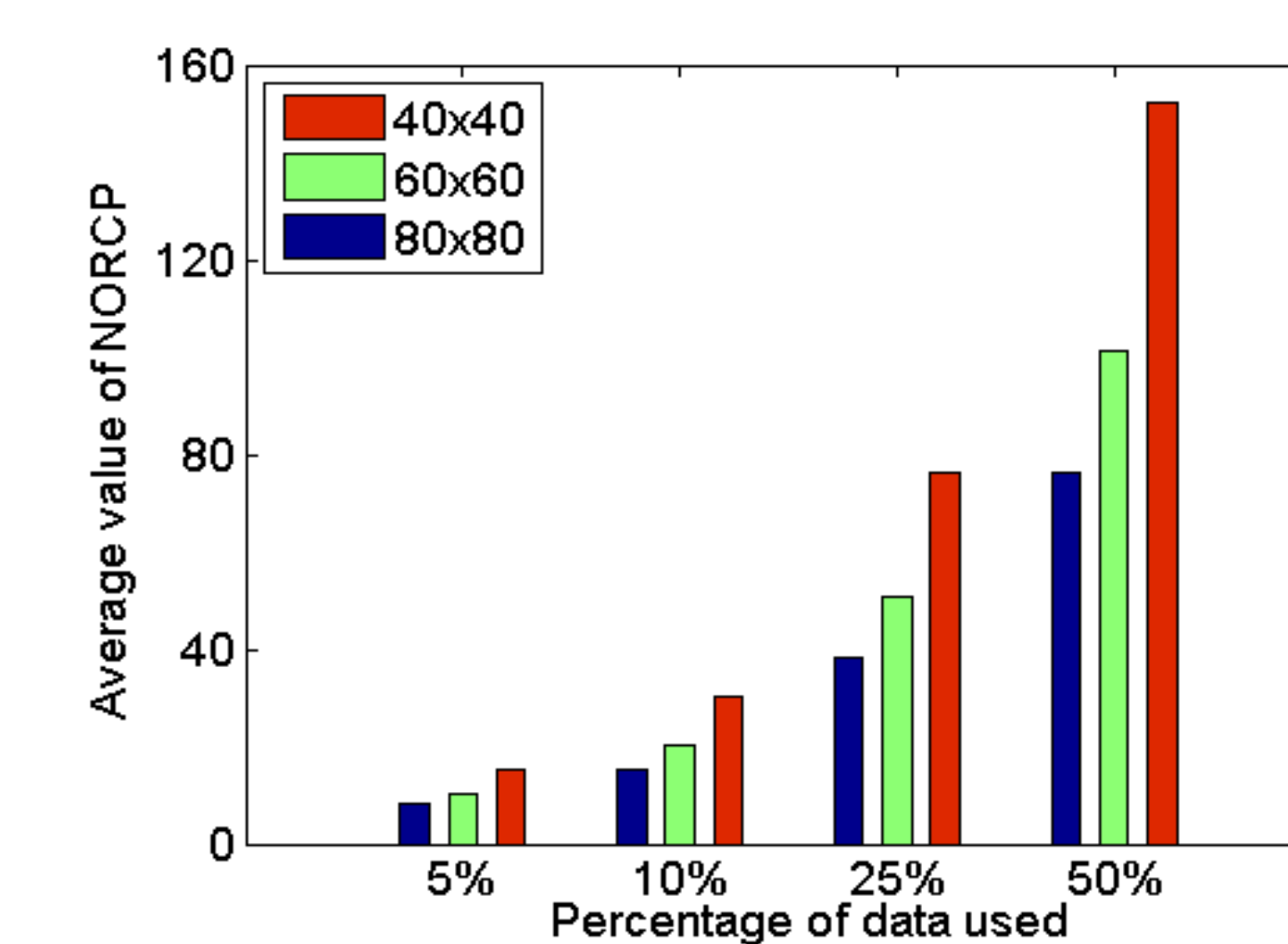
## Error in econstruction



❖ With the reduction of amount of data used, EM algorithm outperforms AM algorithm

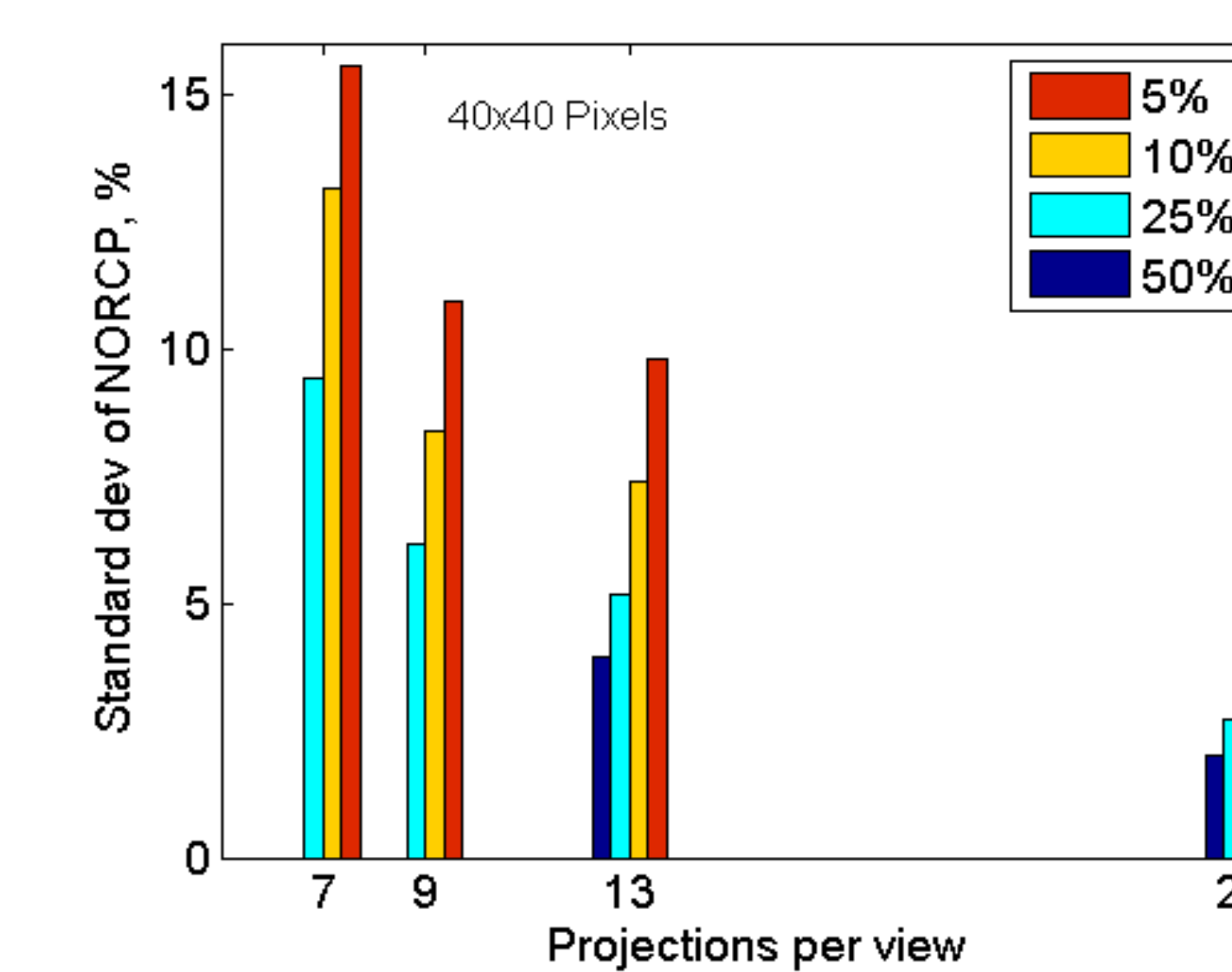
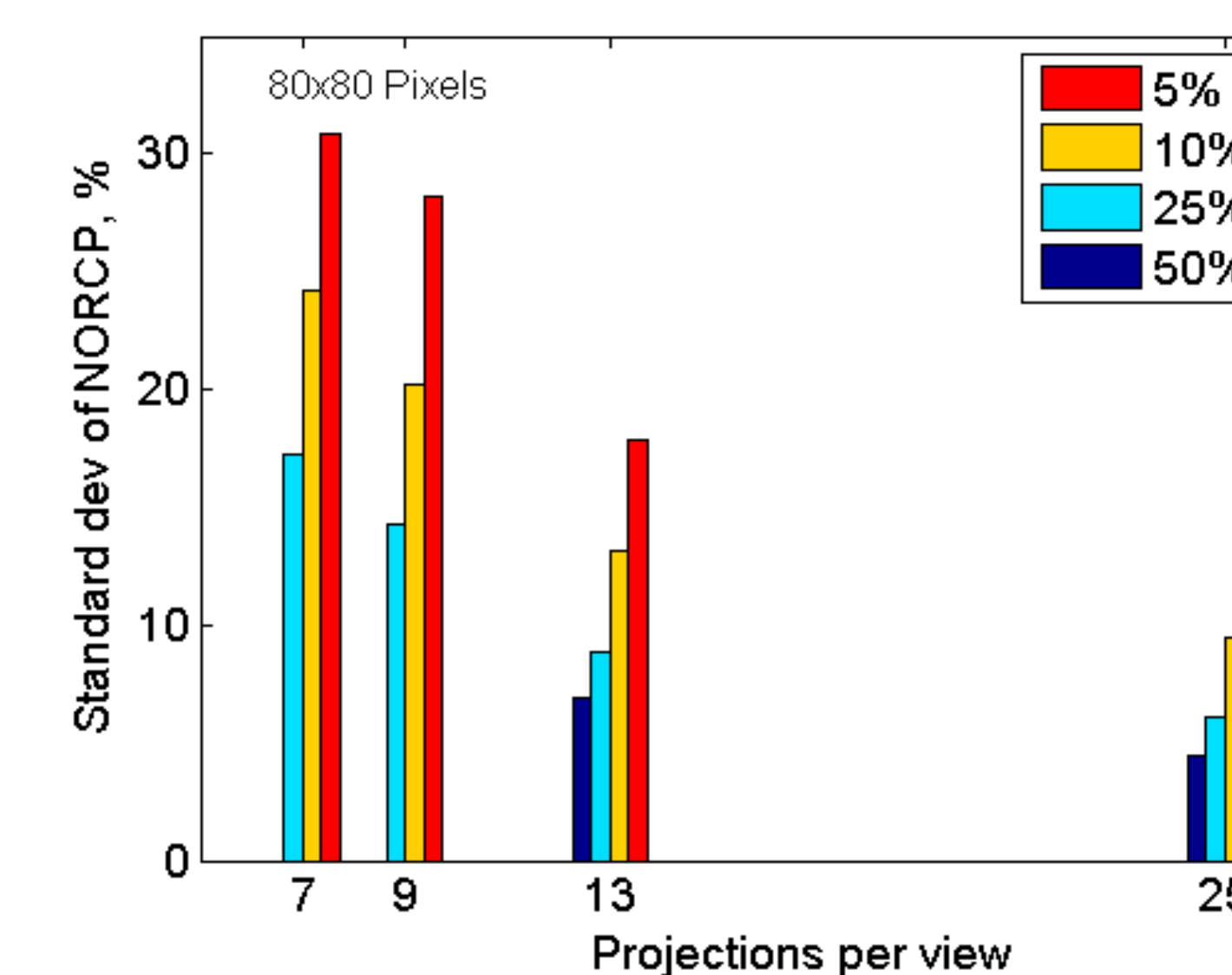
❖ Error reduces with the increase in number of projections used

## Parameters affecting reconstruction error



❖ NORCP - "Number of rays crossing pixel"

❖ Average value of NORCP is a function of pixel size and percentage of data used



❖ Standard deviation of NORCP decreases with the increase in number of projections per view used in reconstruction and with the decrease in number of pixels used in reconstruction image

## Acknowledgements

▪ This work was supported by the National Science Foundation Engineering Research Centers Program, Grant EEC-0310689

▪ Dr. Werner van der Merwe, Prof. Willie Nicol (Pretoria University, South Africa)

▪ TOTAL