

# Bayesian inversion for 3D dental X-ray imaging

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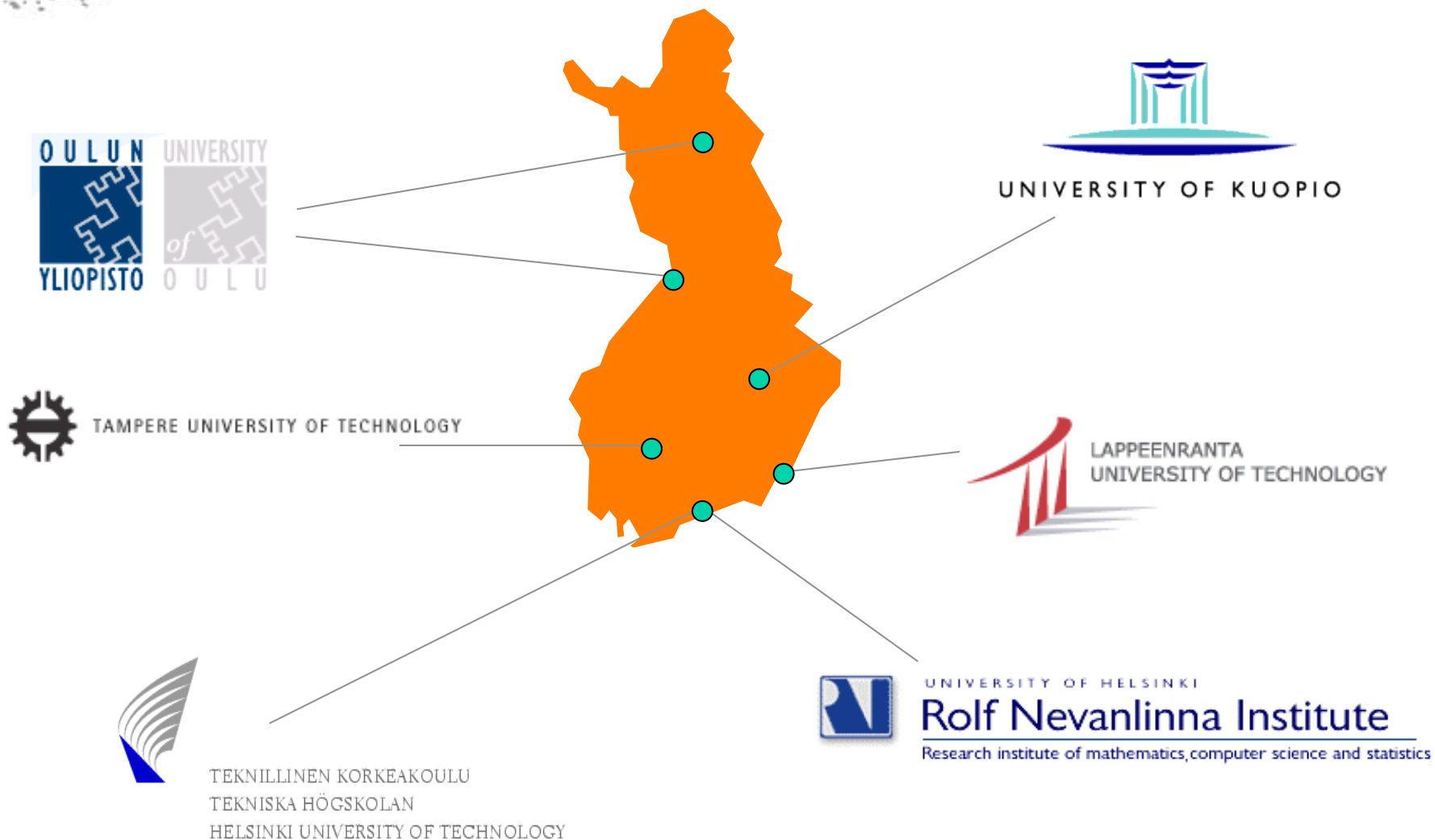
**University of Helsinki**

China-Finland-Germany workshop on inverse problems

March 4, 2010



# Finnish Centre of Excellence in Inverse Problems Research



<http://math.tkk.fi/inverse-coe/>

# A series of projects started in 2001 aiming for a new type of low-dose 3D imaging

The goal was a mathematical algorithm with

**Input:** small number of digital X-ray images taken with any X-ray device

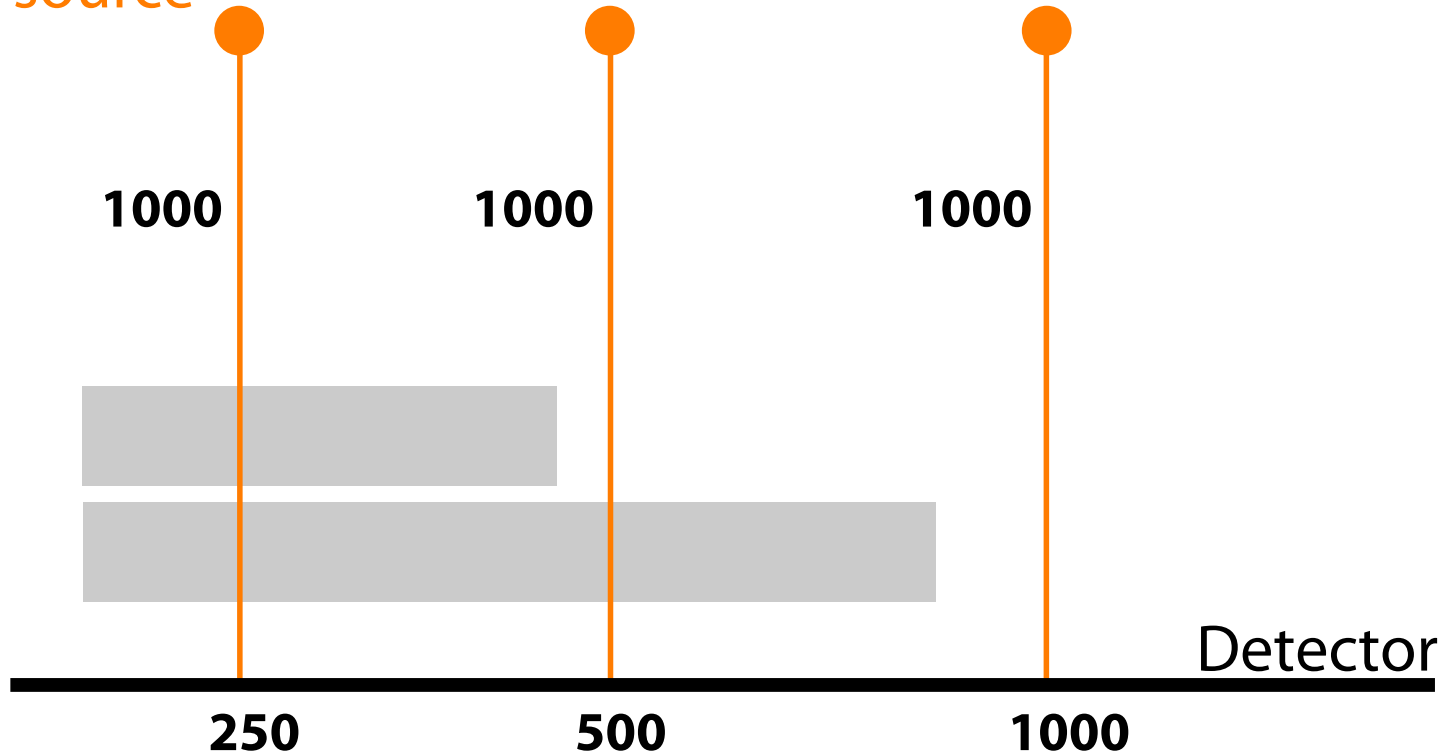
**Output:** three-dimensional reconstruction with quality good enough for the clinical task at hand

Products of GE Healthcare in 2001:



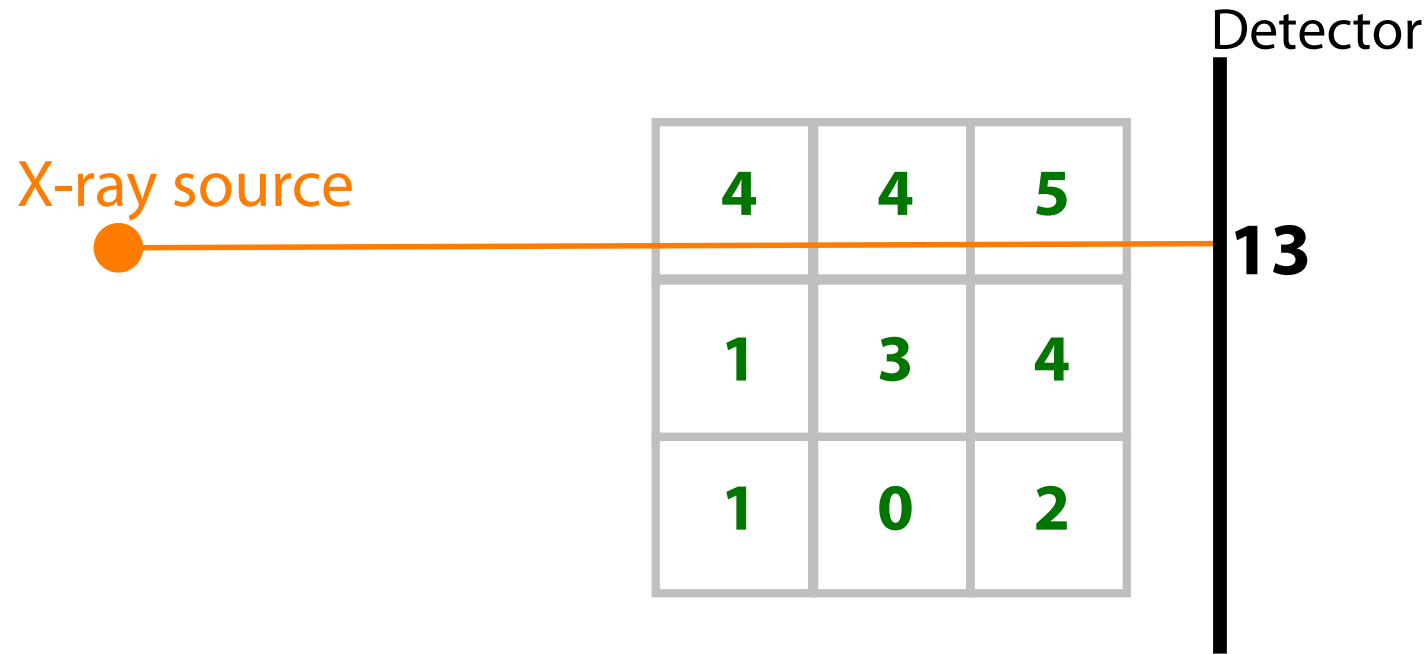
# X-ray images as measurements

X-ray source

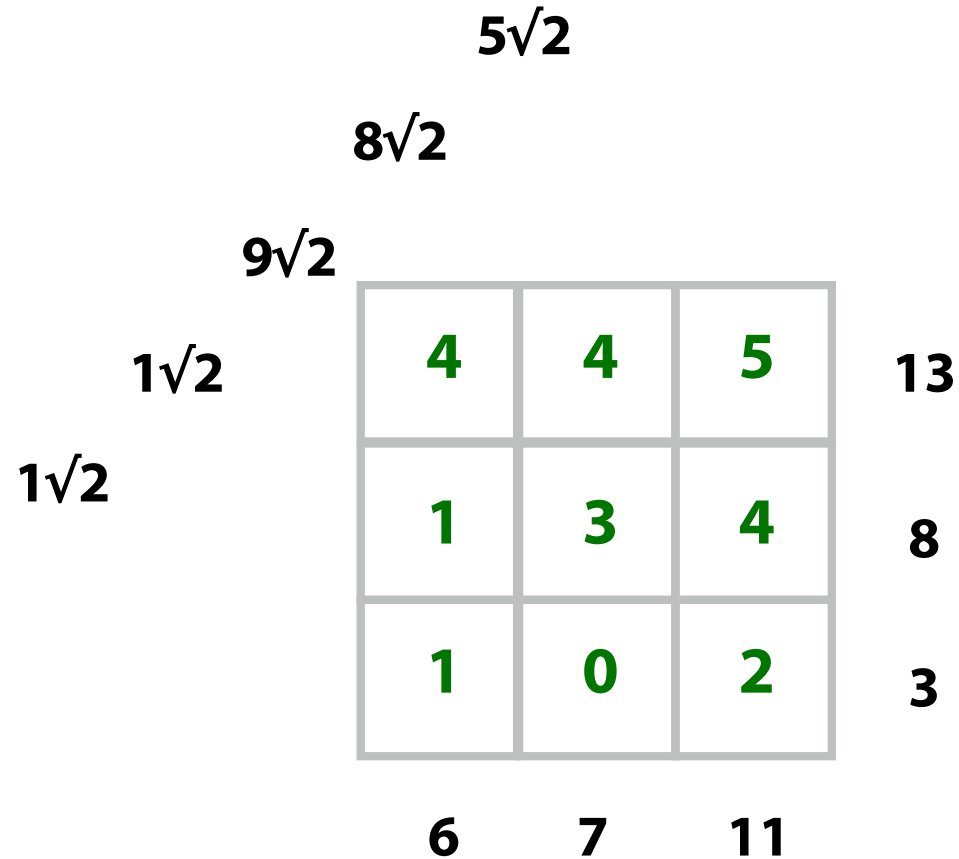


Logarithm	<b>5,5</b>	<b>6,2</b>	<b>6,9</b>
Density	<b>1,4</b>	<b>0,7</b>	<b>0,0</b>

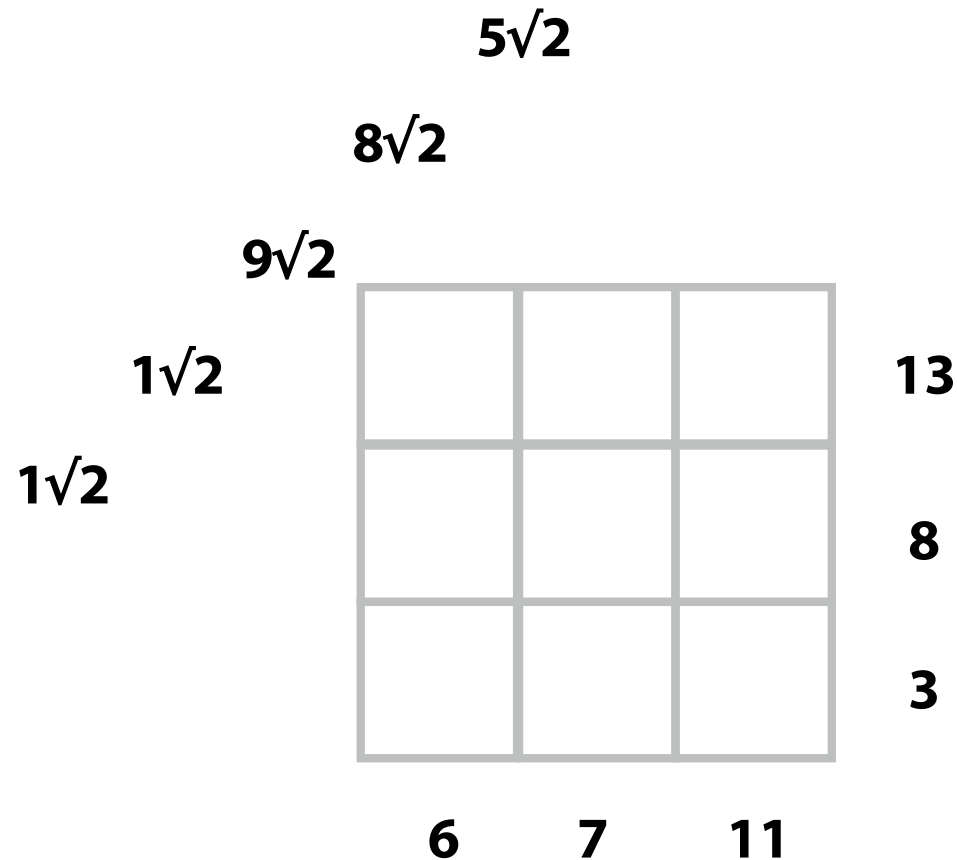
# Every X-ray measures the sum of attenuation through tissue



# Direct problem of tomography is to find the radiographs from given tissue

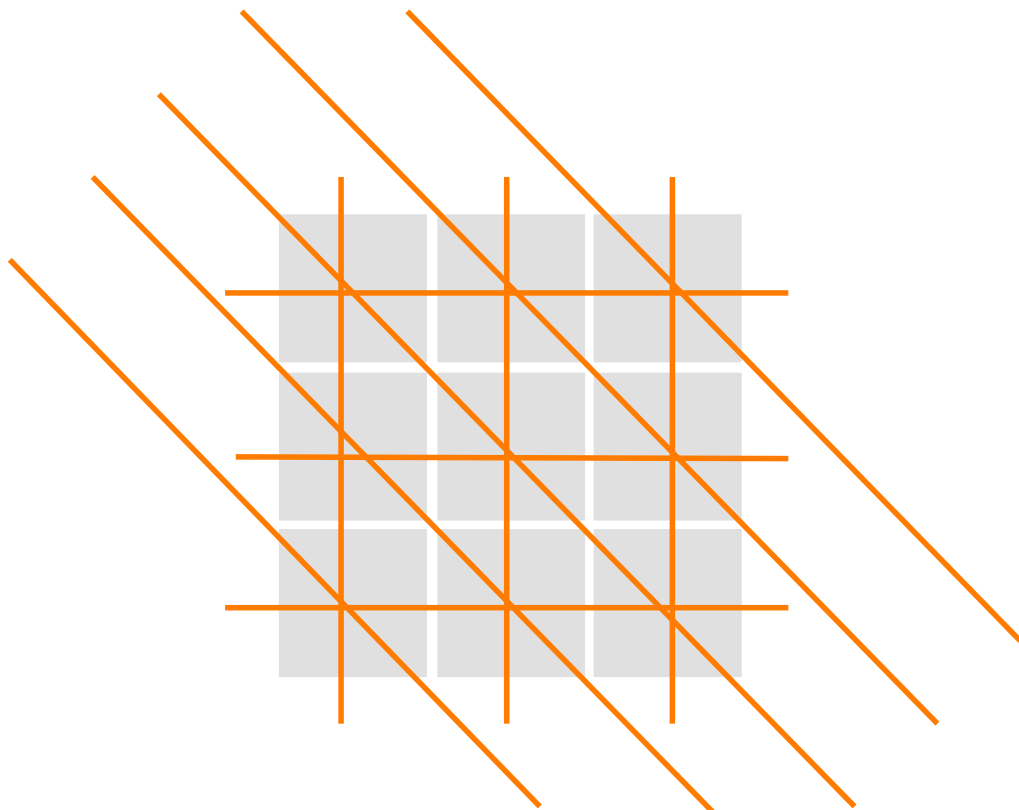


# Inverse problem of tomography is to find the tissue from radiographs

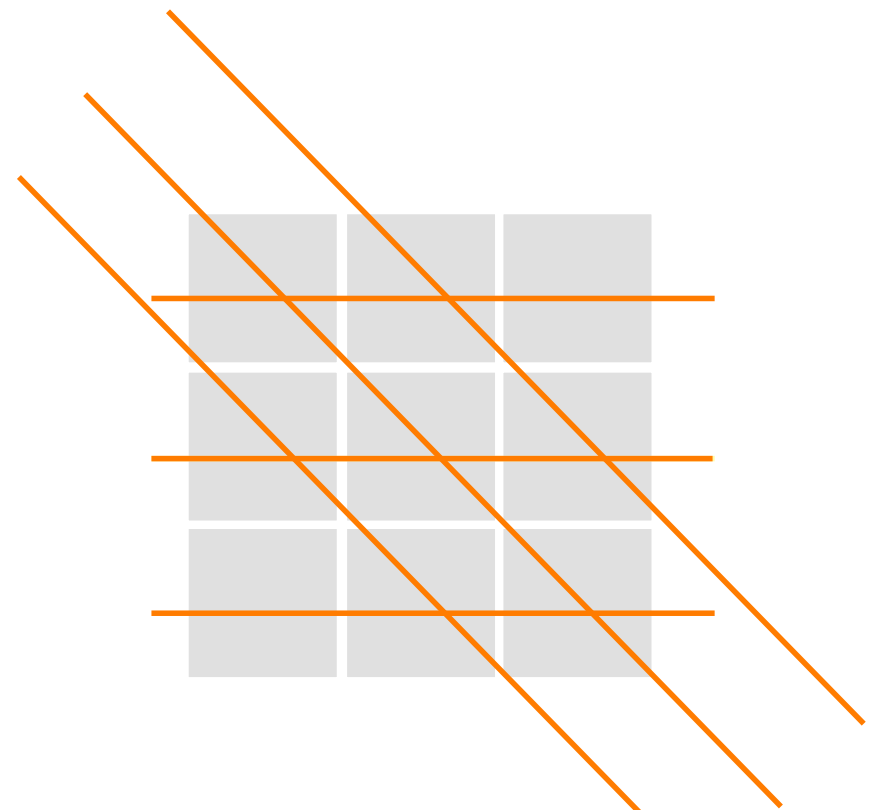


9 unknowns, 11 linear equations

# The limited angle problem is harder than the full angle problem



9 unknowns,  
11 linear equations



9 unknowns,  
6 linear equations



# In limited angle 3D imaging there are many tissues matching the radiographs

			$8\sqrt{2}$	
			$9\sqrt{2}$	
$1\sqrt{2}$				
	4	4	5	13
	1	3	4	8
	1	0	2	3

5	6	2
1	5	2
4	0	-1

9	1	3
1	0	7
3	0	0

*a priori* information is needed!

# We write the reconstruction problem in matrix form and assume Gaussian noise

			m3		
	m2				
m1		x1	x4	x7	m4
		x2	x5	x8	m5
		x3	x6	x9	m6

$$x = [x_1, x_2, \dots, x_9]^T$$
$$m = [m_1, m_2, \dots, m_6]^T$$

$$Ax = m$$

Measurement  $m = Ax + e$  with Gaussian noise  $\varepsilon$  of standard deviation  $\sigma$  leads to the following likelihood distribution:

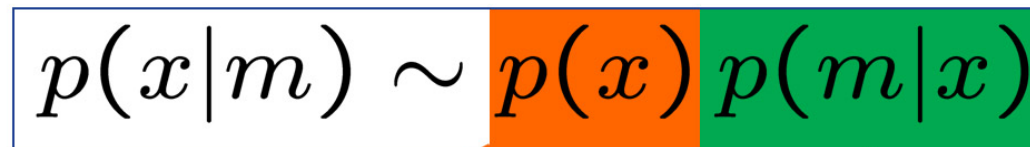
$$p(m|x) = p_\varepsilon(Ax - m) \sim \exp\left(-\frac{1}{2\sigma^2} \|Ax - m\|_2^2\right)$$

# Bayes formula combines measured data and *a priori* information together

We reconstruct the most probable 3D tissue in light of

1. Available radiographs and
2. Physiological *a priori* information

Bayes formula gives the *posterior distribution*  $p(x|m)$ :

$$p(x|m) \sim p(x) p(m|x)$$
The equation is displayed within a white rectangular box with a blue border. The term p(x) is enclosed in an orange box, and p(m|x) is enclosed in a green box. An orange arrow points from the orange box to the label 'Prior distribution, or tissue model' below. A green arrow points from the green box to the label 'Likelihood distribution, or measurement model' below.

**Prior distribution,  
or tissue model**

**Likelihood distribution,  
or measurement model**

We recover  $x$  as a point estimate from  $p(x|m)$

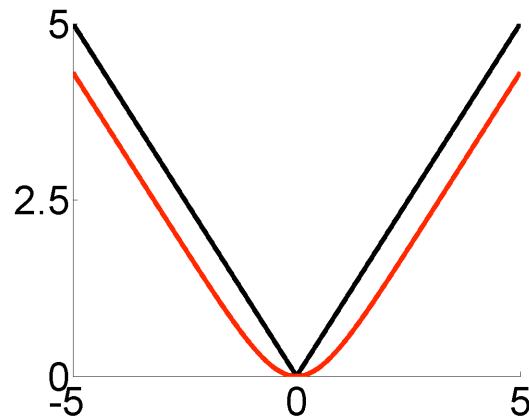
# We build a prior distribution for dental tissue using total variation prior

Positivity constraint:

$$p_+(x) = \begin{cases} 0 & \text{if } x_j < 0 \text{ for some } j \\ 1 & \text{otherwise} \end{cases}$$

Approximate total variation penalty:

$$p_{\text{TV}}(x) = \exp\left(\alpha \sum_{\text{neighbors}} |x_\ell - x_k|^\beta\right)$$



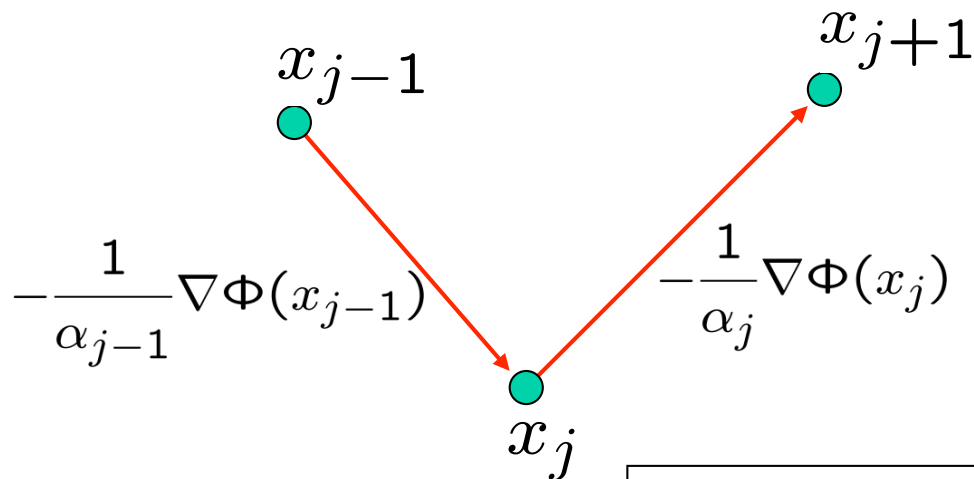
$$|t|_\beta = \frac{1}{\beta} \log(\cosh(\beta t))$$

# Computation of the MAP estimate

Large scale optimization problem:

$$x_{\text{MAP}} = \arg \min_{x_j \geq 0} \left\{ \frac{1}{2\sigma^2} \|Ax - m\|_2^2 + \alpha \sum_{\text{neighbors}} |x_\ell - x_k|^\beta \right\}$$

We use the gradient method of Barzilai & Borwein, which is a modification of Euler's steepest descent method.

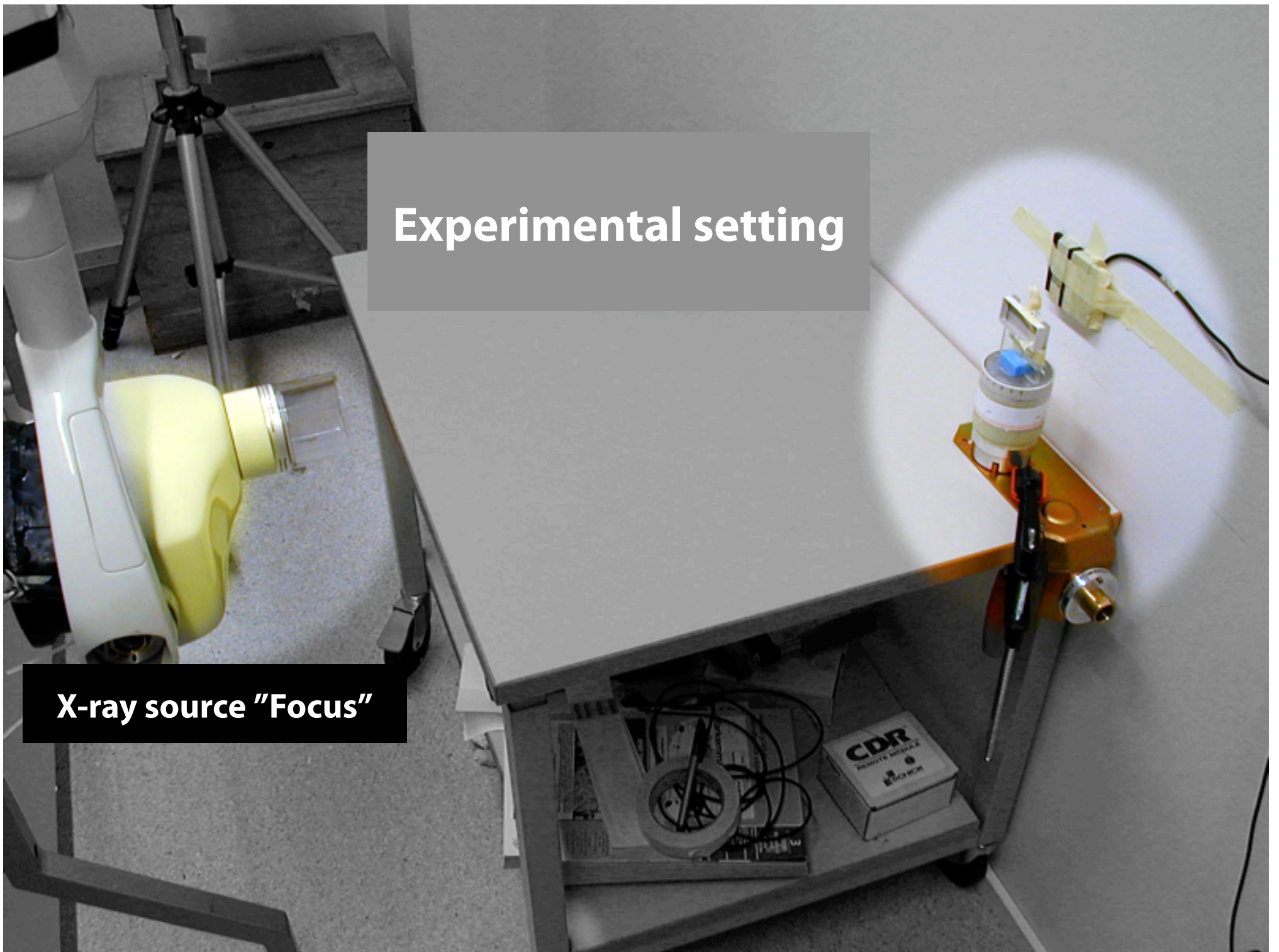


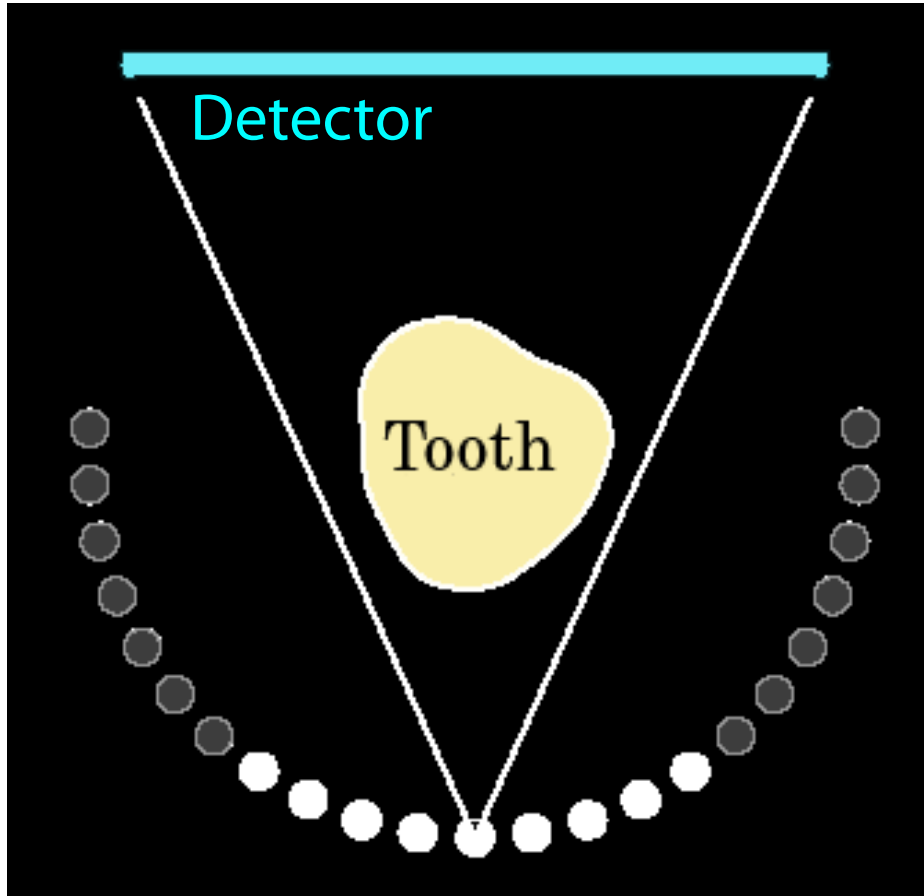
Step size differs from Euler's:

$$\alpha_j = \frac{(x_j - x_{j-1}) \cdot (\nabla\Phi(x_j) - \nabla\Phi(x_{j-1}))}{(x_j - x_{j-1}) \cdot (x_j - x_{j-1})}$$

# Experimental setting

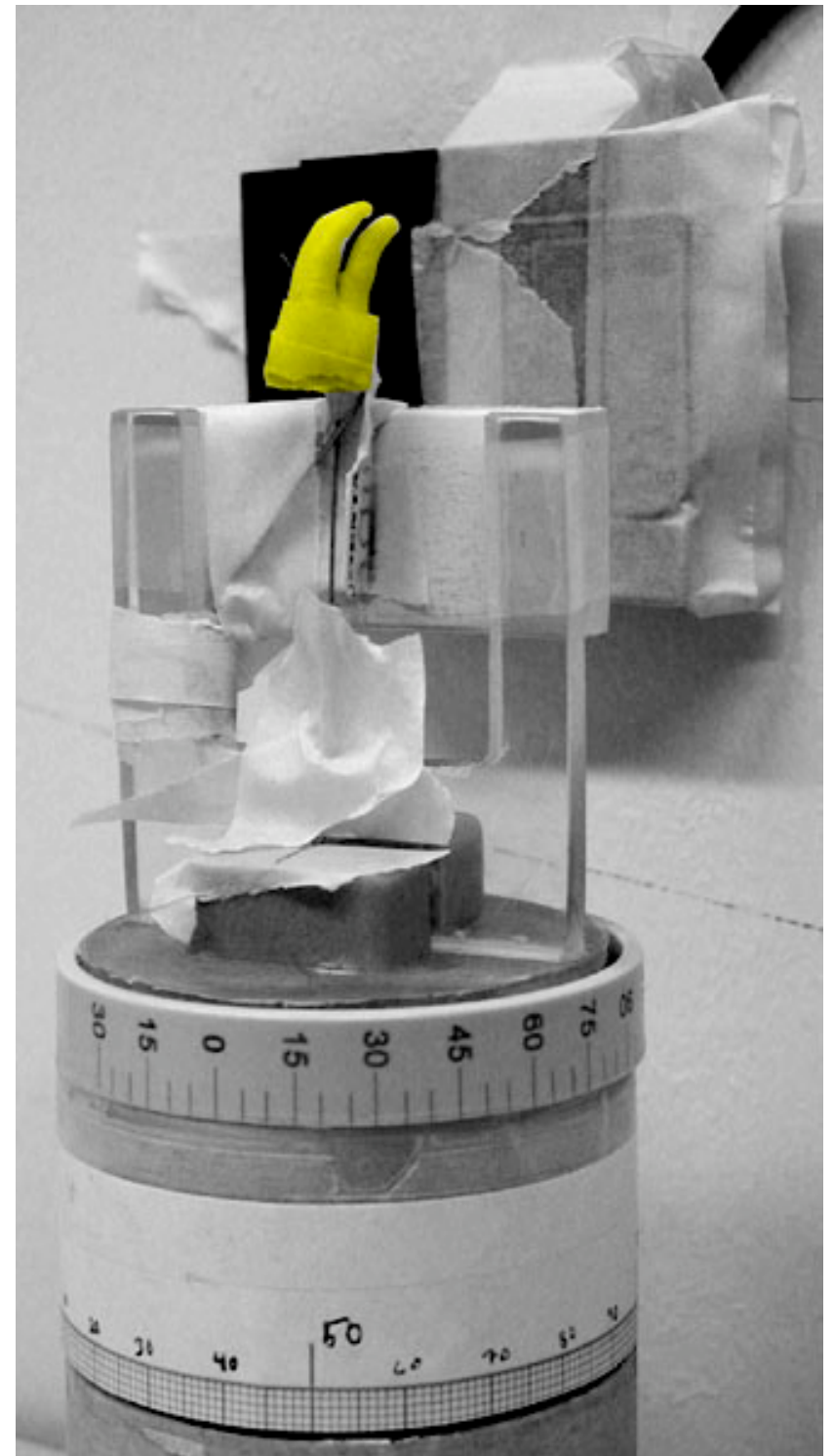
X-ray source "Focus"





X-ray source positions

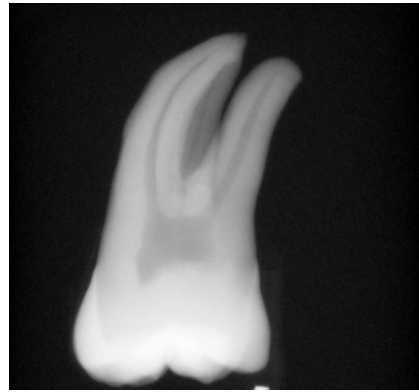
Tooth donated to science  
by Helena Sarlin. Thanks!



# The projection images look like this



0



30



60



90

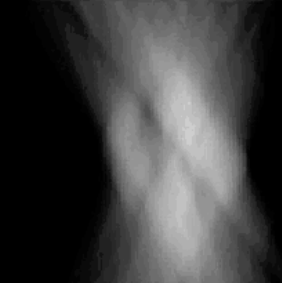


# Horizontal slices:

**truth**  
full angle

**Bayes**  
limited angle

**tomo**  
limited angle

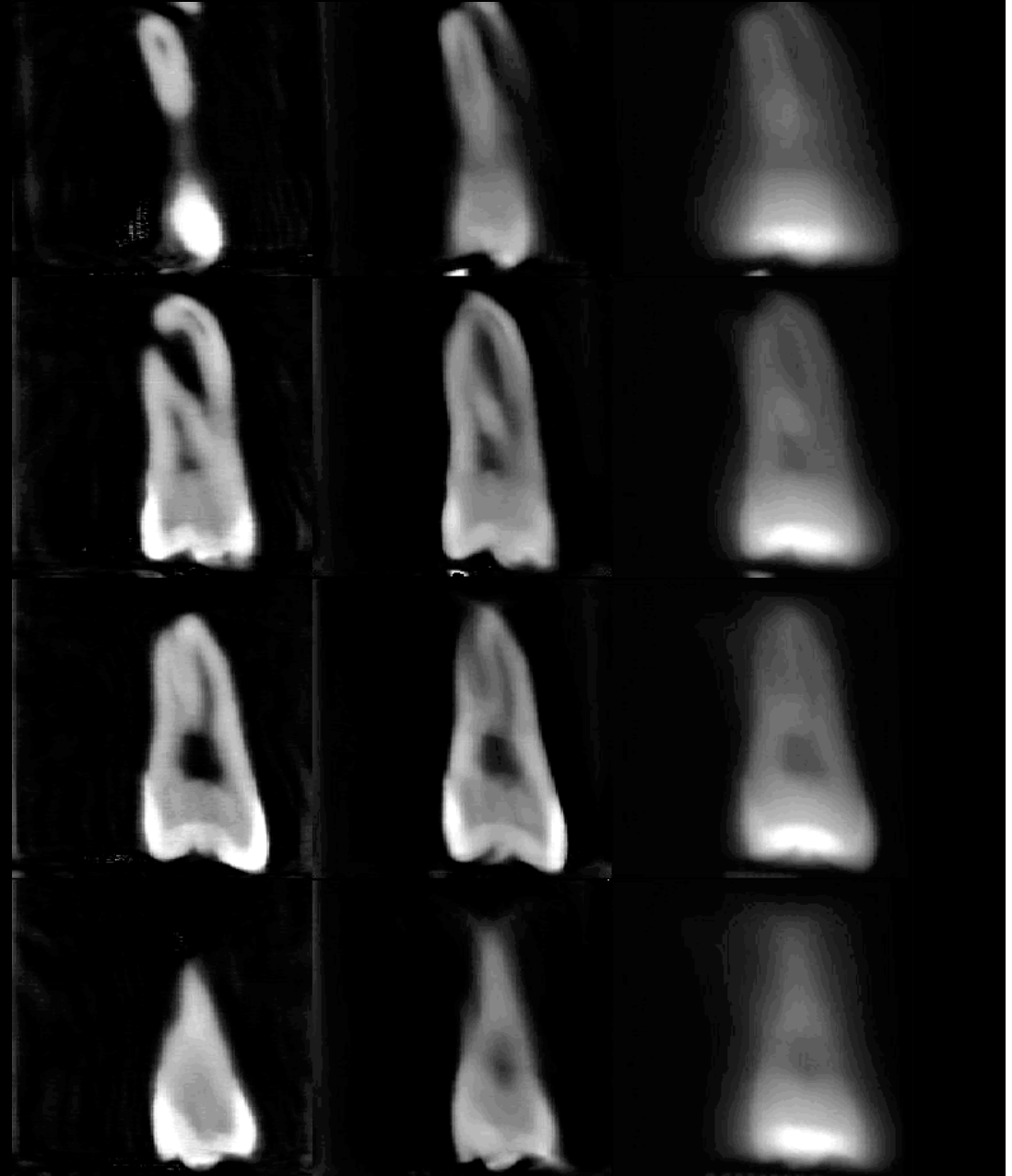


# Vertical slices:

**truth**  
full angle

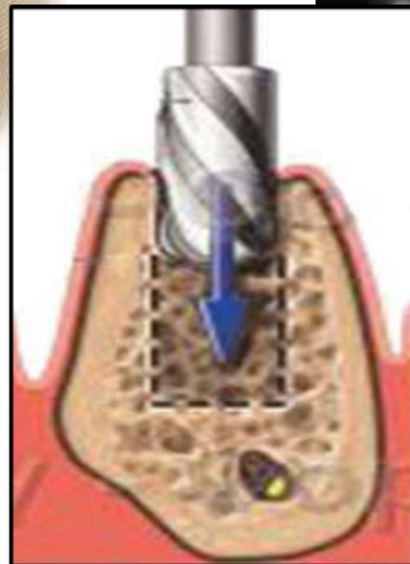
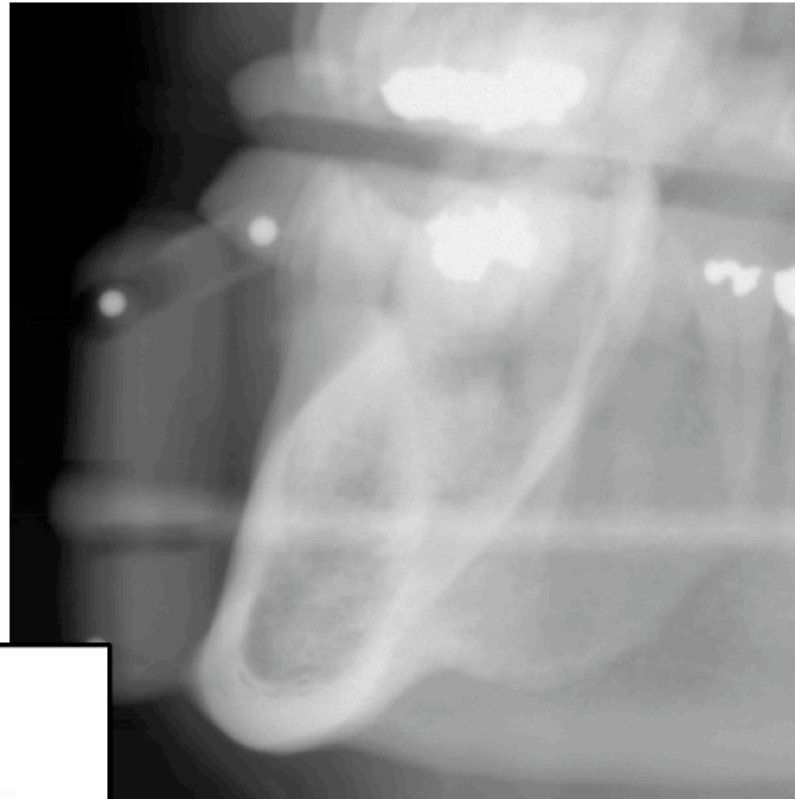
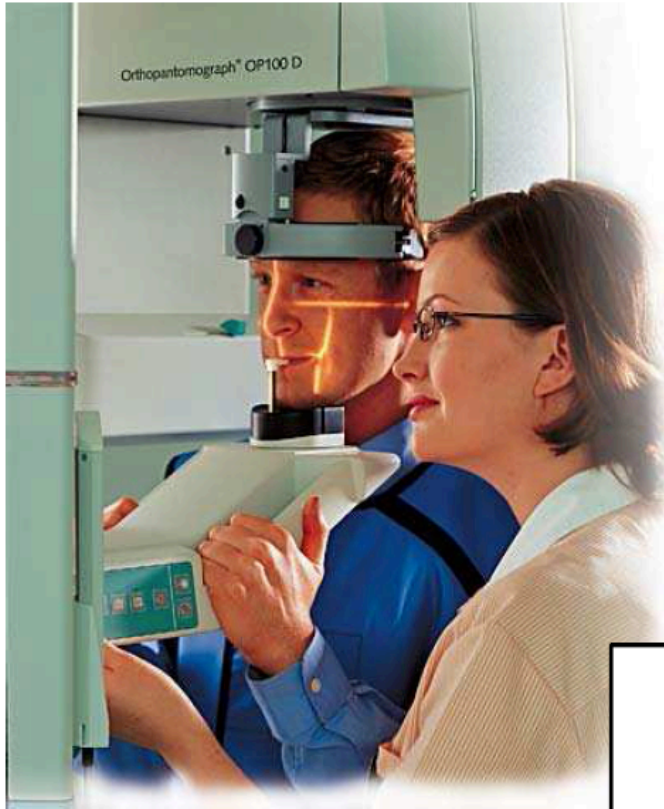
**Bayes**  
limited angle

**tomo**  
limited angle



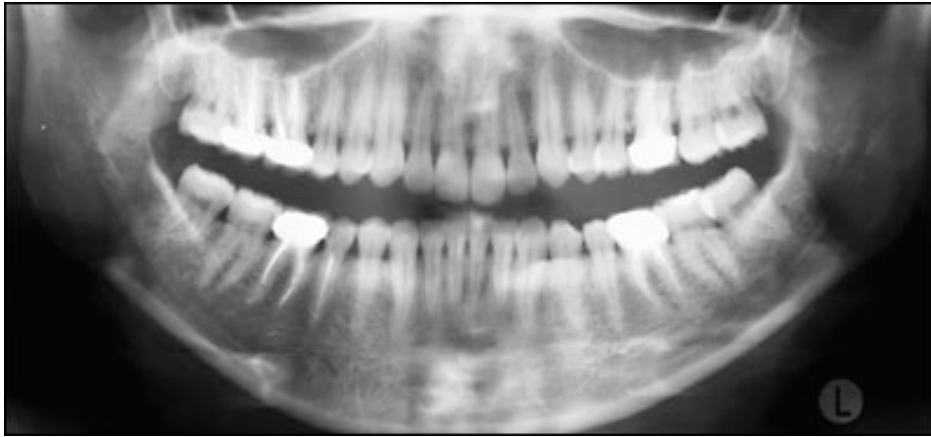
Kolehmainen, S,  
Järvenpää, Kaipio,  
Koistinen, Lassas,  
Pirttilä, Somersalo  
(2003)

# 2D projection radiograph is not enough for dental implant planning



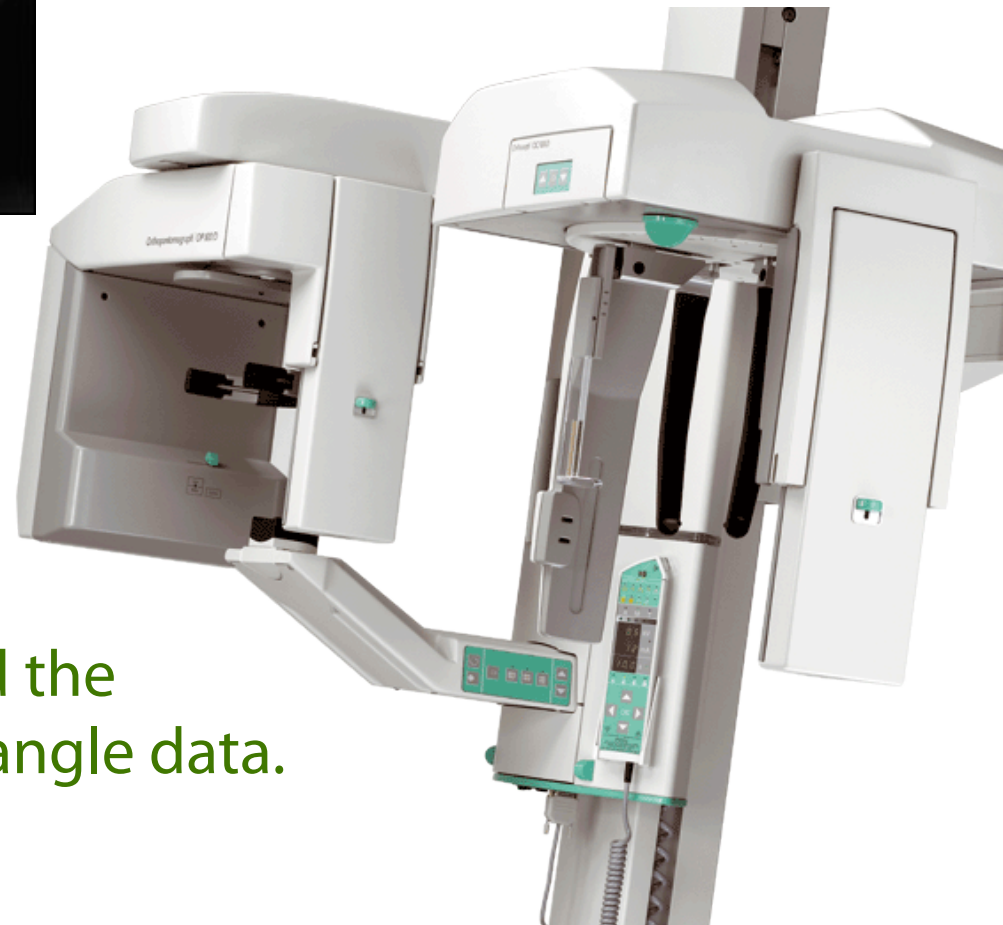
# Panoramic X-ray device rotates around the head and produces a general picture

Panoramic imaging was invented by Yrjö Paatero in 1950's.



Nowadays a panoramic device is standard equipment at every dental clinic around the world.

In our project, we reprogrammed the device so that it collects limited-angle data.

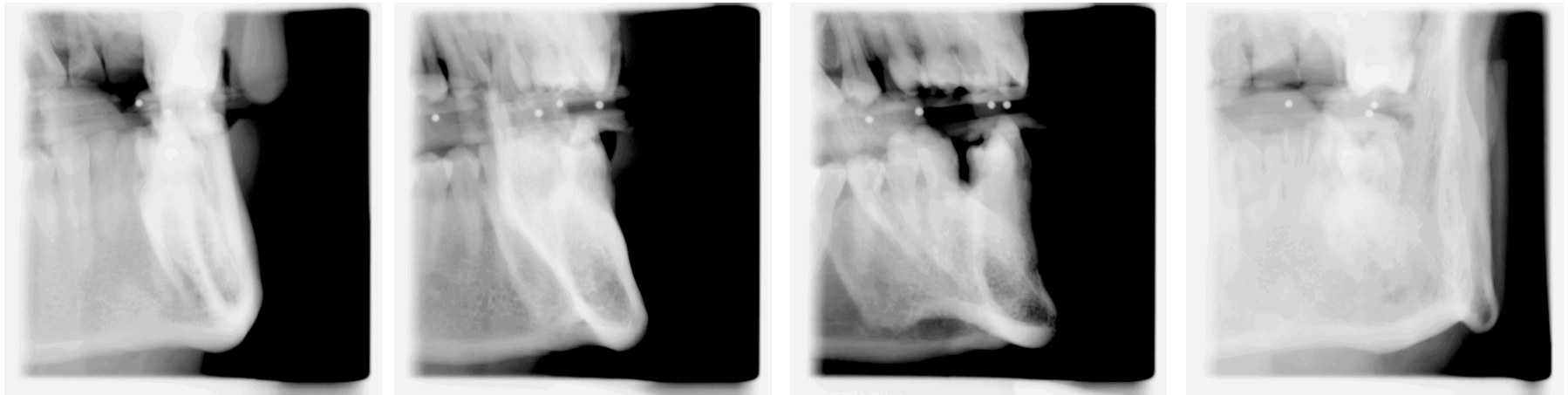
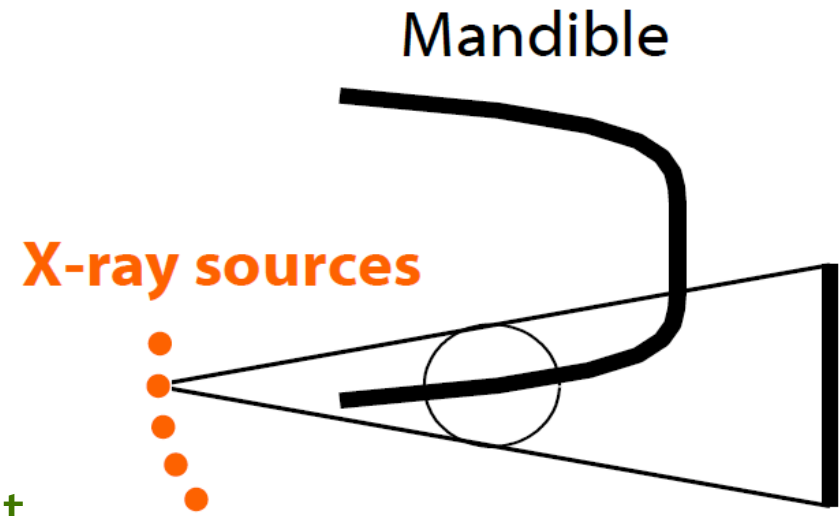


# We consider the following limited angle experiment with the panoramic x-ray device:

**11** projection images of the mandibular area

**40** degrees aperture

**1000 x 1000** pixels per image formed by a scanning movement



# Limited angle reconstruction can be used for locating the mandibular nerve



This is core technology for the VT product of PaloDEX Group

Kolehmainen, Vanne, S, Järvenpää, Kaipio, Lassas and Kalke (2006)

Kolehmainen, Lassas and S (2008)

Cederlund, Kalke and Welander (2009)

Hyvönen, Kalke, Lassas, Setälä, Siltanen (submitted)

# Essential history of the three projects:

**Academic members:** Inverse problems research groups in University of Helsinki, University of Kuopio, Helsinki University of Technology and Tampere University of Technology

## **Industrial members:**

**2001-2002** Instrumentarium Imaging and Invers Ltd

**2003-2004** GE Healthcare Finland

**2005-2007** PaloDEX Group

**Funding** by TEKES and the companies.

**Outcome:** 12 peer-reviewed articles, 3 patents, algorithms for 3 commercial products