

System for Live Virtual-Endoscopic Guidance of Bronchoscopy

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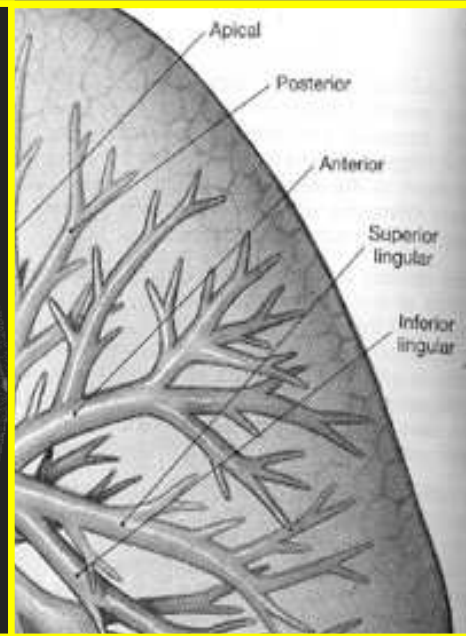
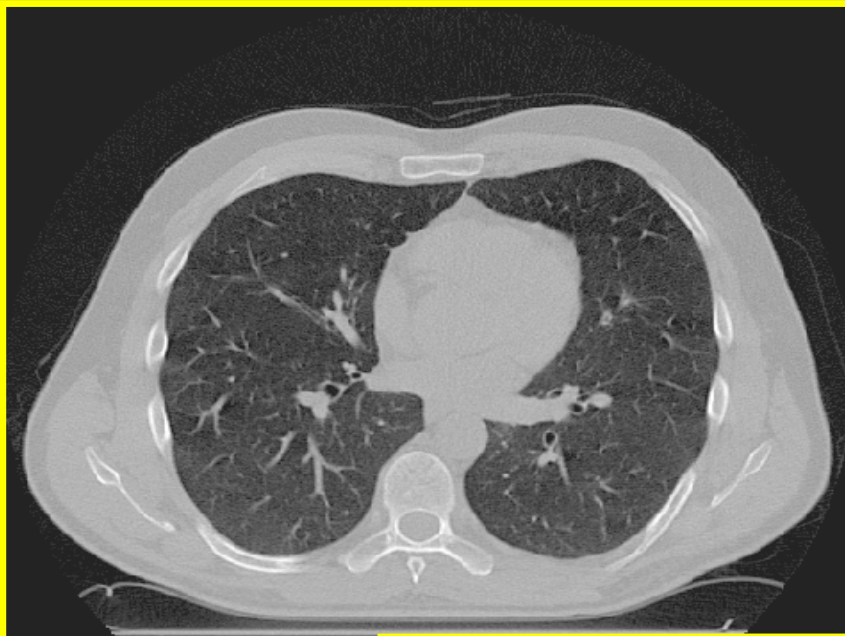
Vision for Human-Computer Interaction (V4HCI) Workshop – CVPR 2005, San Diego, CA, 21 June 2005.

Lung Cancer

- Lung Cancer: #1 cancer killer, 30% of all cancer deaths, 1.5 million deaths world-wide, < 15% 5-year survival rate (nearly the worst of cancer types)
- To diagnose and treat lung cancer,
 - 1) 3D CT-image assessment – preplanning, noninvasive
 - 2) Bronchoscopy – invasive

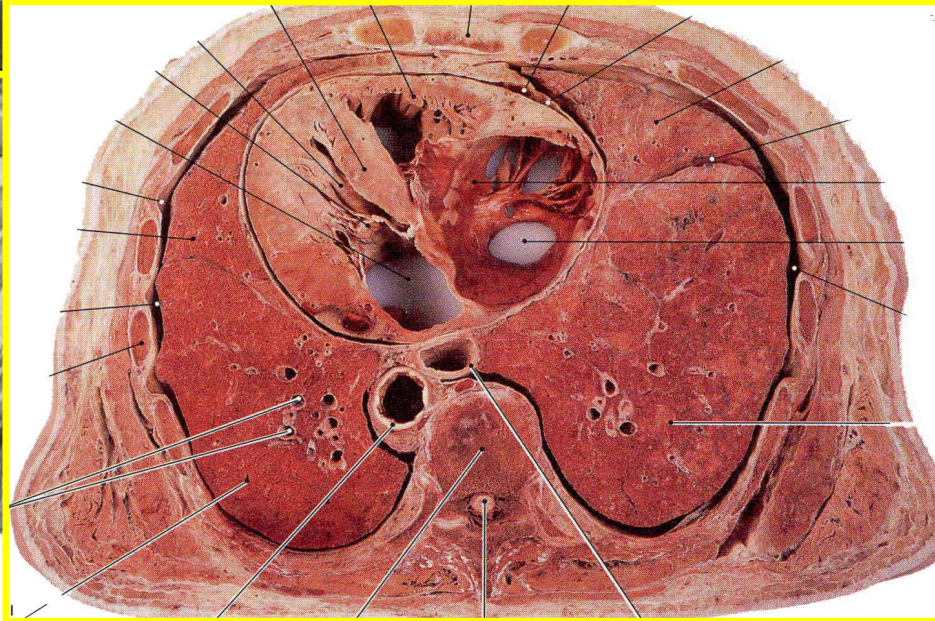
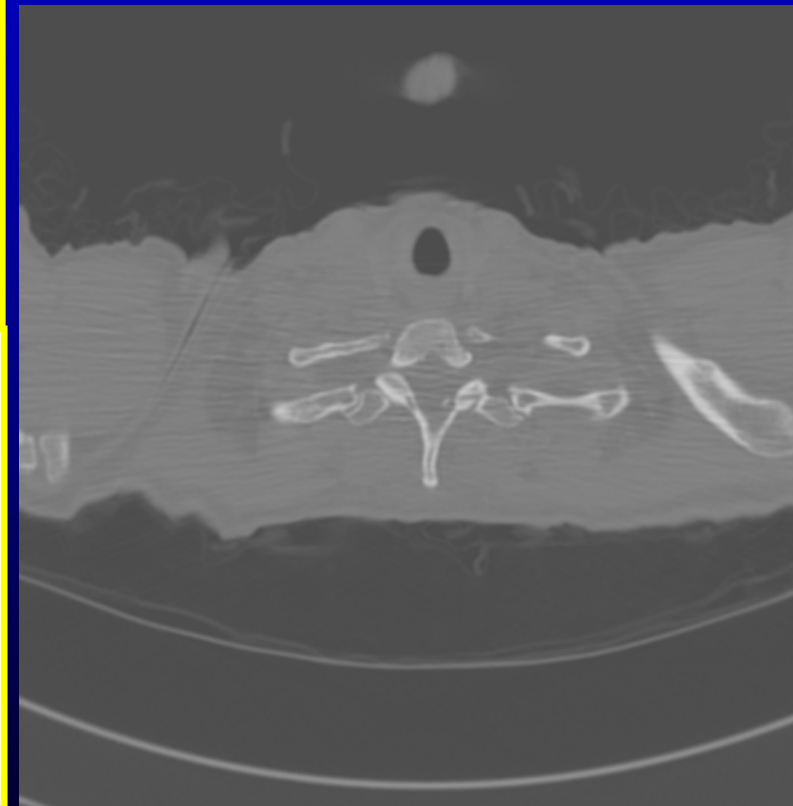
→ Procedure is LITTLE HELP if diagnosis/treatment are poor

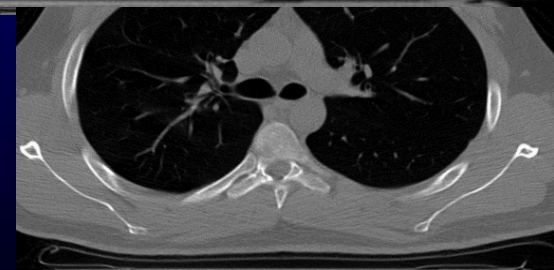
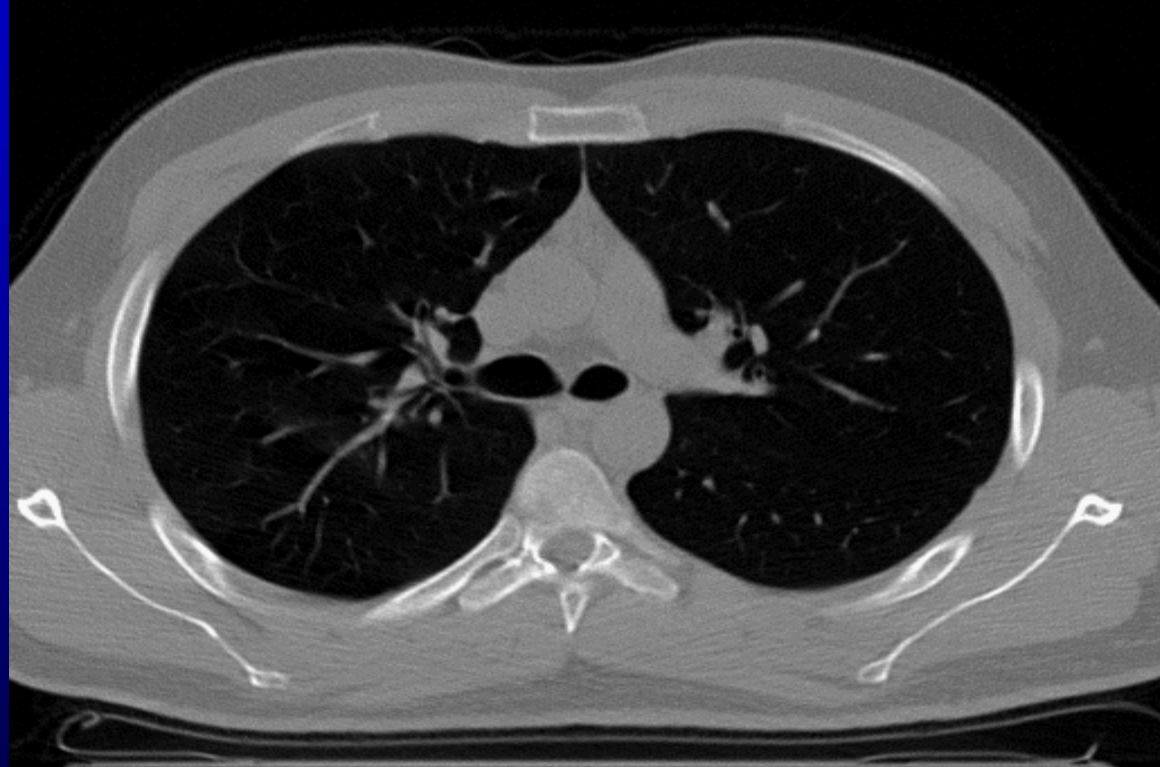
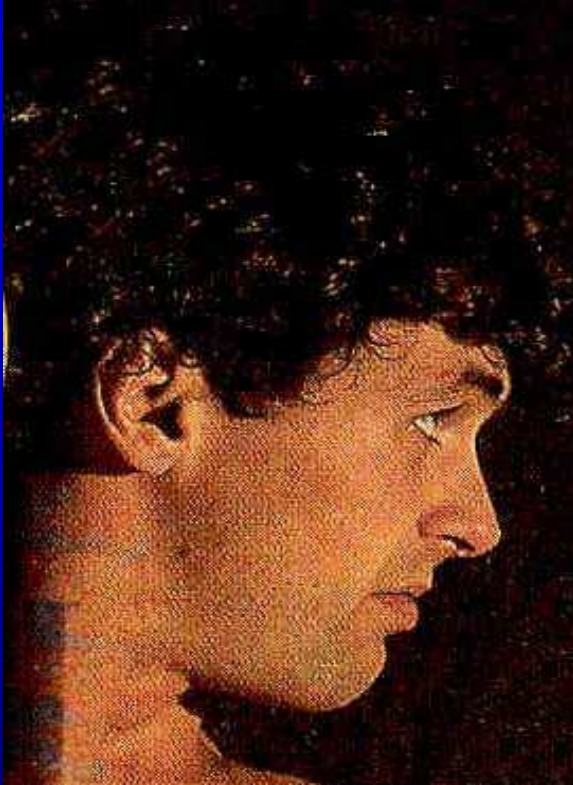
3D CT Chest Images



Typical chest scan $V(x,y,z)$:

1. 500 512X512 slices $V(x,y,.)$
2. 0.5mm sampling interval





3D Mental Reconstruction

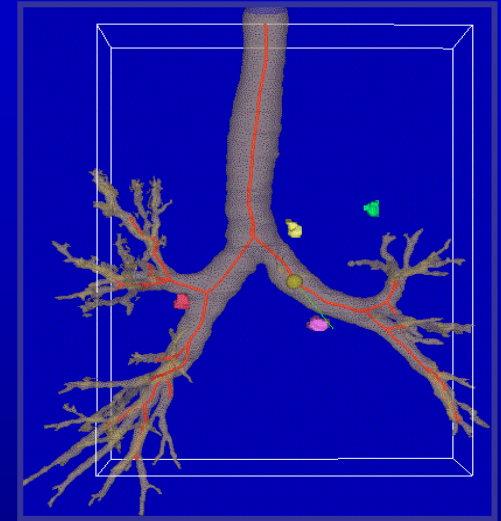
→ How physicians assess CT scans now

Visualization Techniques – see “inside” 3D Images

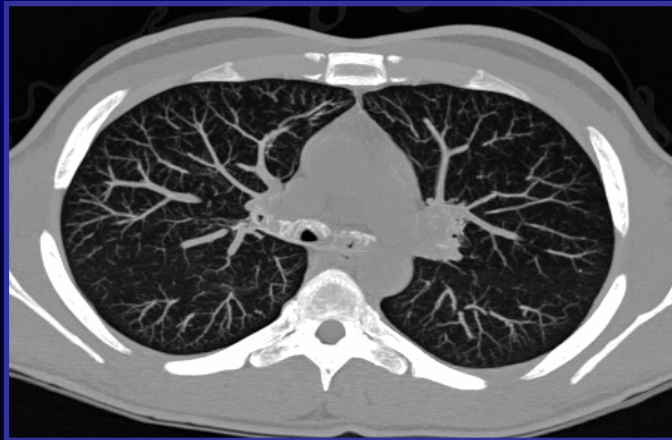
multi-planar reconstruction²



volume/surface rendering⁴

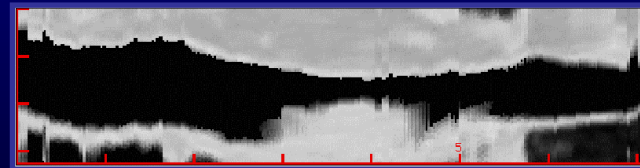


STS-MIP
sliding-thin-slab maximum
intensity projection⁶

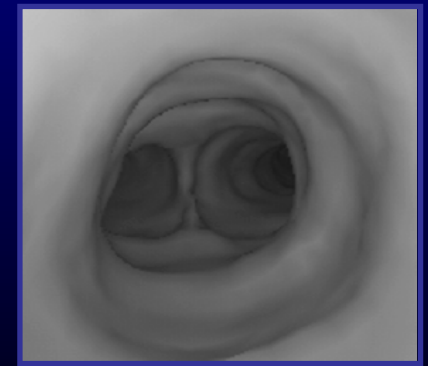


projection imaging¹

virtual endoscopic rendering⁵



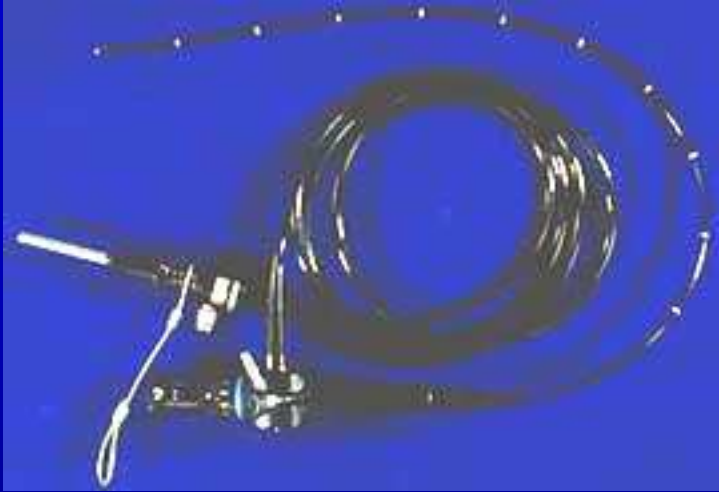
curved-section reformatting³



¹{Hohne87,Napel92} ²{Robb1988,Remy96,McGuinness97} ³{Robb1988,Hara96,Ramaswamy99}
⁴{Ney90,Drebin88,Tiede90} ⁵{Vining94,Ramaswamy99, Helferty01} ⁶{Napel, 92}

Bronchoscopy

→ For “live” procedures



video from
bronchoscope

$$I_V(x,y)$$

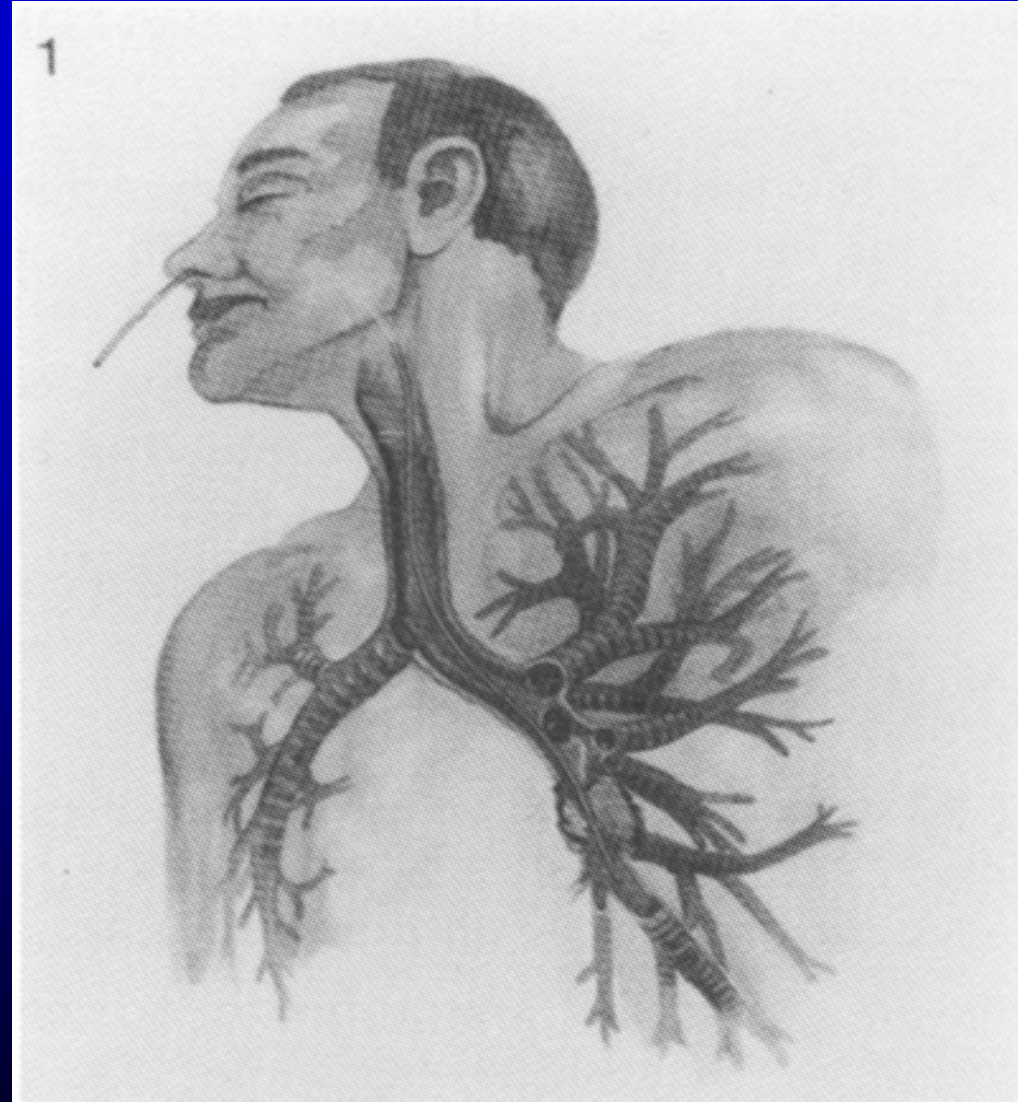


Figure 19.4, Wang/Mehta '95

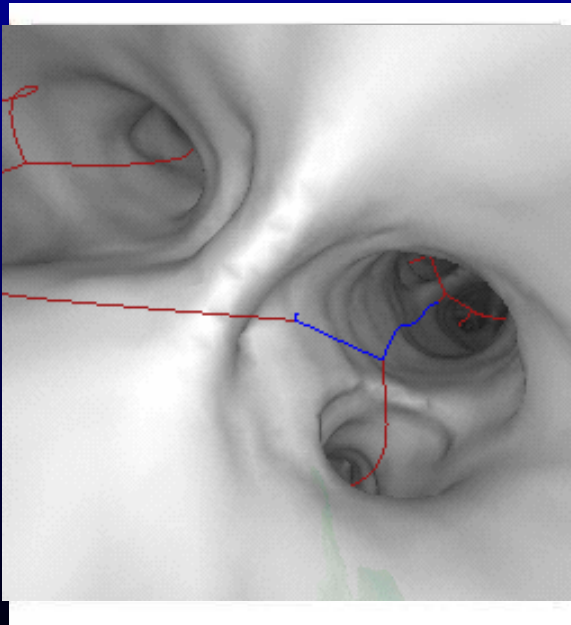
Difficulties with Bronchoscopy

1. Physician skill varies greatly!
2. Low biopsy yield. Many “missed” cancers.
3. Biopsy sites are beyond airway walls – biopsies are done blindly!



Virtual Endoscopy (Bronchoscopy)

- Input a high-resolution 3D CT chest image
 - *virtual copy of chest anatomy*
- Use computer to explore virtual anatomy
 - permits unlimited “exploration”
 - no risk to patient



Endoluminal
Rendering
 $I_{CT}(x,y)$
(inside airways)

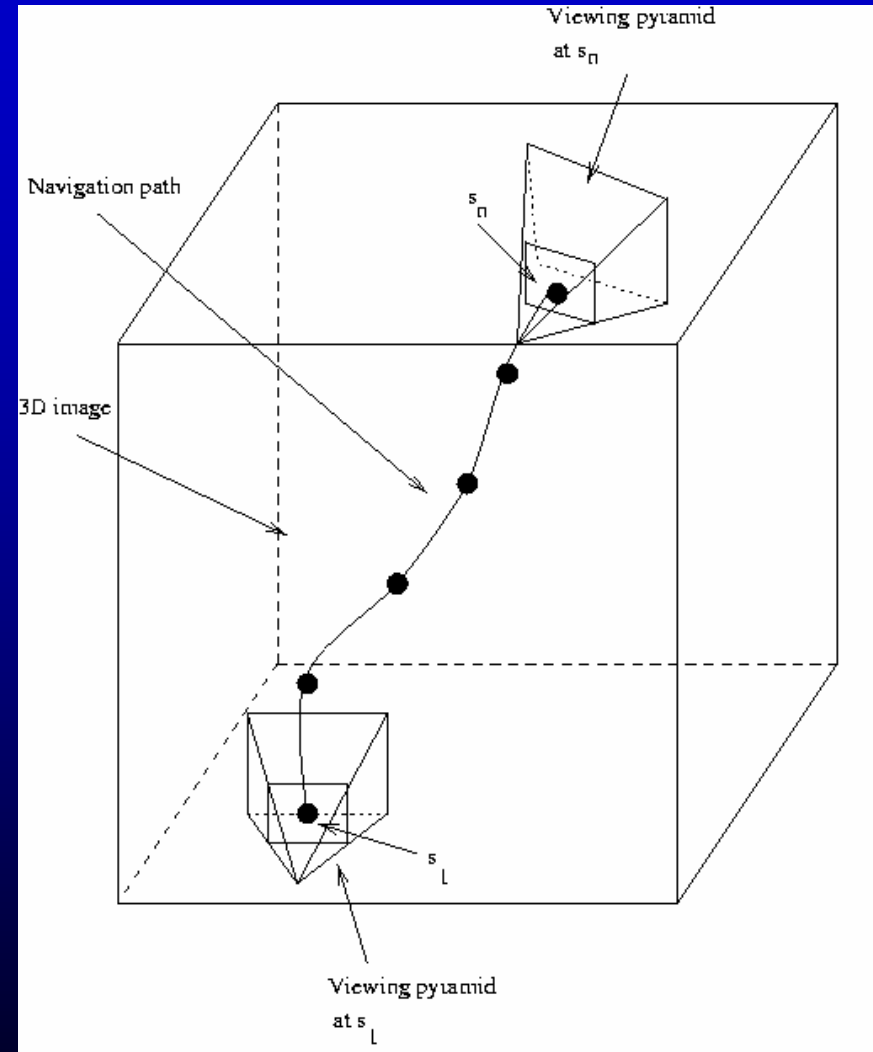
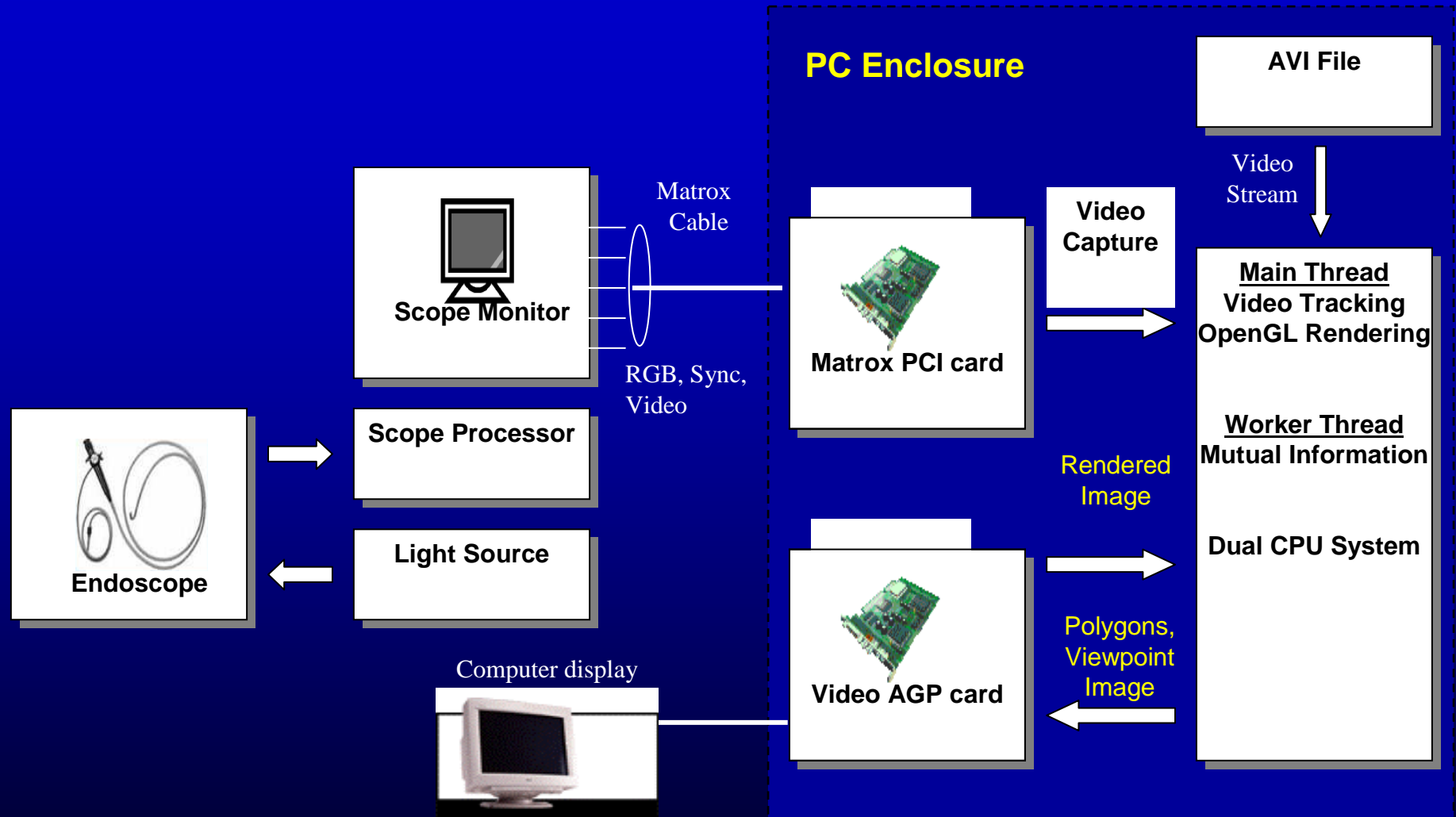


Image-Guided Bronchoscopy Systems

Show potential, but recently proposed systems have limitations:

- CT-Image-based
 - McAdams *et al.* (*AJR* 1998) and Hopper *et al.* (*Radiology* 2001)
 - Bricault *et al.* (*IEEE-TMI* 1998)
 - Mori *et al.* (*SPIE Med. Imaging* 2001, 2002)
 - Electromagnetic Device attached to scope
 - Schwarz *et al.* (*Respiration* 2003)
- *Our system*: reduce skill variation, easy to use, reduce “blindness”

Our System: Hardware



Software written in Visual C++.

Our System: Work Flow

Data Sources



Data Processing

Stage 1: 3D CT Assessment

- 1) Segment 3D Airway Tree
- 2) Calculate Centerline Paths
- 3) Define Target ROI biopsy sites
- 4) Compute polygon data

→ Case Study

Stage 2: Live Bronchoscopy

For each ROI:

- 1) Present virtual ROI site to physician
- 2) Physician moves scope "close" to site
- 3) Do CT-Video registration
- 4) Repeat steps (1-3) until ROI reached

Stage 1: 3D CT Assessment (Briefly)

1. Segment Airway tree

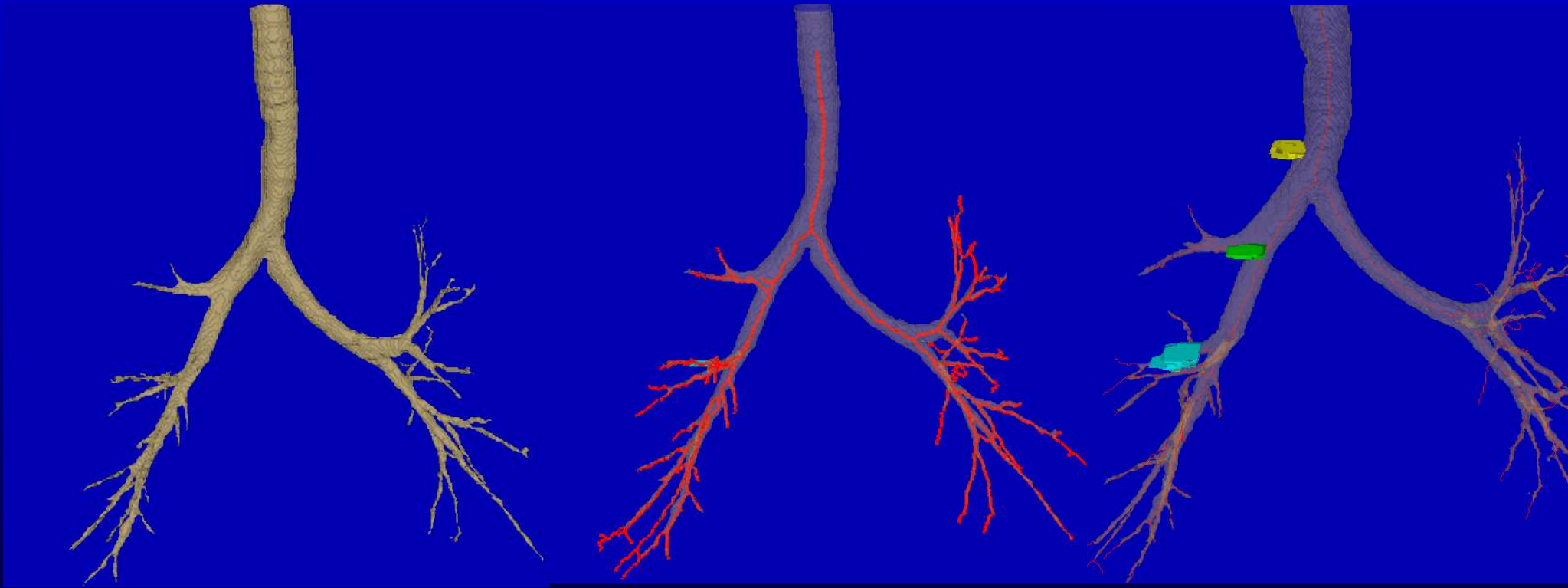
(Kiraly *et al.*, *Acad. Rad.* 10/02)

2. Extract centerlines

(Kiraly *et al.*, *IEEE-TMI* 11/04)

3. Define ROIs

(e.g., suspect cancer)



4. Compute tree-surface polygon data (Marching Cubes – vtk)

→ CASE STUDY to help guide bronchoscopy

Stage 2: Bronchoscopy - Key Step: CT-Video Registration

Register

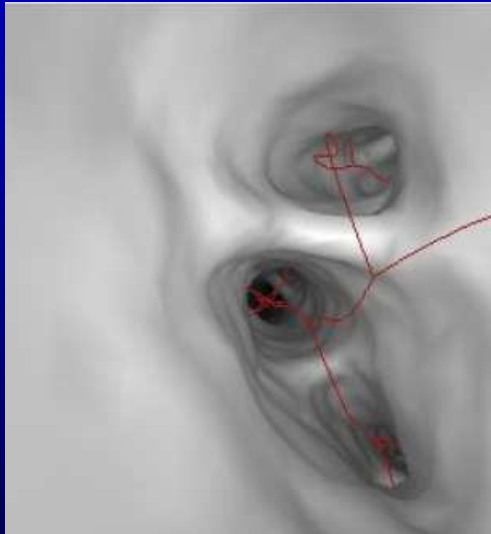
To the

Virtual 3D CT World

Real Endoscopic Video World

$I_{CT}^X(x, y)$ (Image Source 1)

$I_V^F(x, y)$ (Image Source 2)



→ Maximize normalized mutual information to get

$I_{CT}^{X_o}(x, y)$

CT-Video Registration: 1) Match viewpoints of two cameras

Both image sources, I_V and I_{CT} , are cameras.

6-parameter vector representing camera viewpoint $\chi = (X, Y, Z, \alpha, \beta, \gamma)$

3D point $\mathbf{p} = (X_p, Y_p, Z_p)$ mapped to camera point (X_c, Y_c)

through the standard transformation

$$\begin{bmatrix} X_c \\ Y_c \\ Z_c \end{bmatrix} = \mathbf{R}(\alpha, \beta, \gamma) \begin{bmatrix} X_p - X \\ Y_p - Y \\ Z_p - Z \end{bmatrix}$$

The final camera screen point is given by (x, y) where

$$x = \frac{fX_c}{Z_c}, \quad y = \frac{fY_c}{Z_c}$$

Make FOVs of both Cameras equal

To facilitate registration, make both cameras I_V and I_{CT} have the *same* FOV.

To do this, use an endoscope calibration technique (Helferty et al., *IEEE-TMI* 7/01).

Measure the bronchoscope's focal length (done off-line):

$$f = \frac{(x_r - x_l)Z_m}{(X_r - X_l)}$$

Then, the angle subtended by the scope's FOV is

$$\theta_{\text{FOV}} = 2 \tan^{-1} \left(\frac{x_r - x_l}{2f} \right)$$

Use *same* value for endoluminal renderings, I_{CT} .

Normalized Mutual Information

Mutual Information (MI) – used for registering two different image sources:

a) Grimson *et al.* (*IEEE-TMI* 4/96)

b) Studholme *et al.* (*Patt. Recog.* 1/99) → normalized MI (NMI)

Normalized Mutual Information

Normalized mutual information (NMI):

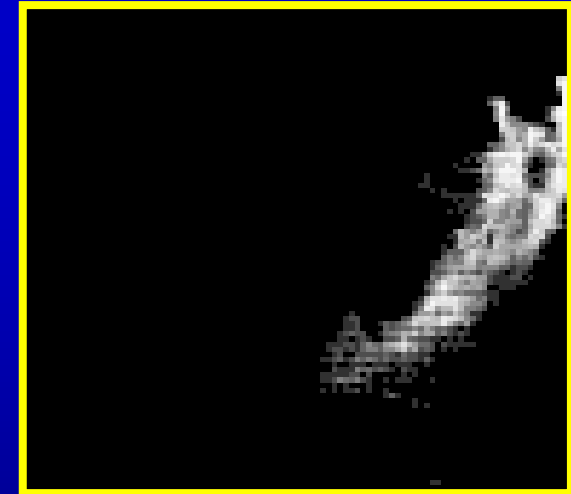
$$S_{NMI}(I_V, I_{CT}) = \frac{h(V) + h(CT)}{h(V, CT)}$$

where

$$h(V) = - \sum_{k=0}^{M-1} \sum_{l=0}^{M-1} p_{V,CT}(k, l) \log p_V(k)$$

$$h(CT) = - \sum_{k=0}^{M-1} \sum_{l=0}^{M-1} p_{V,CT}(k, l) \log p_{CT}(l)$$

$$h(V, CT) = - \sum_{k=0}^{M-1} \sum_{l=0}^{M-1} p_{V,CT}(k, l) \log p_{V,CT}(k, l)$$



“optimal” $p_{V,CT}$

and $p_V(k) = \sum_{l=0}^{M-1} p_{V,CT}(k, l)$ is a histogram (marginal density)

CT-Video Registration – Optimization Problem

Given a fixed video frame $I_V^F(x, y)$ and starting CT view $I_{CT}^{\chi_i}(x, y)$

Search for the optimal CT rendering $I_{CT}^{\chi_o}(x, y)$ subject to

$$\chi_o = \arg \left\{ \max_{\chi \in N_{\chi_i}} [S_{NMI} (I_{CT}^{\chi}(x, y), I_V^F(x, y))] \right\}$$

where viewpoint $\chi = (X, Y, Z, \alpha, \beta, \gamma)$ is varied over neighborhood N_{χ_i} about χ_i

Optimization algorithms used: Simplex and simulated annealing

System Results

Three sets of results are presented:

A. Phantom Test

controlled test, free of subject motion

B. Animal Studies

controlled in vivo (live) tests

C. Human Lung-Cancer Patients

real clinical circumstances

A. Phantom Test

Goal: Compare biopsy accuracy under controlled stationary circumstances using
(1) the standard CT-film approach versus (2) image-guided bronchoscopy.

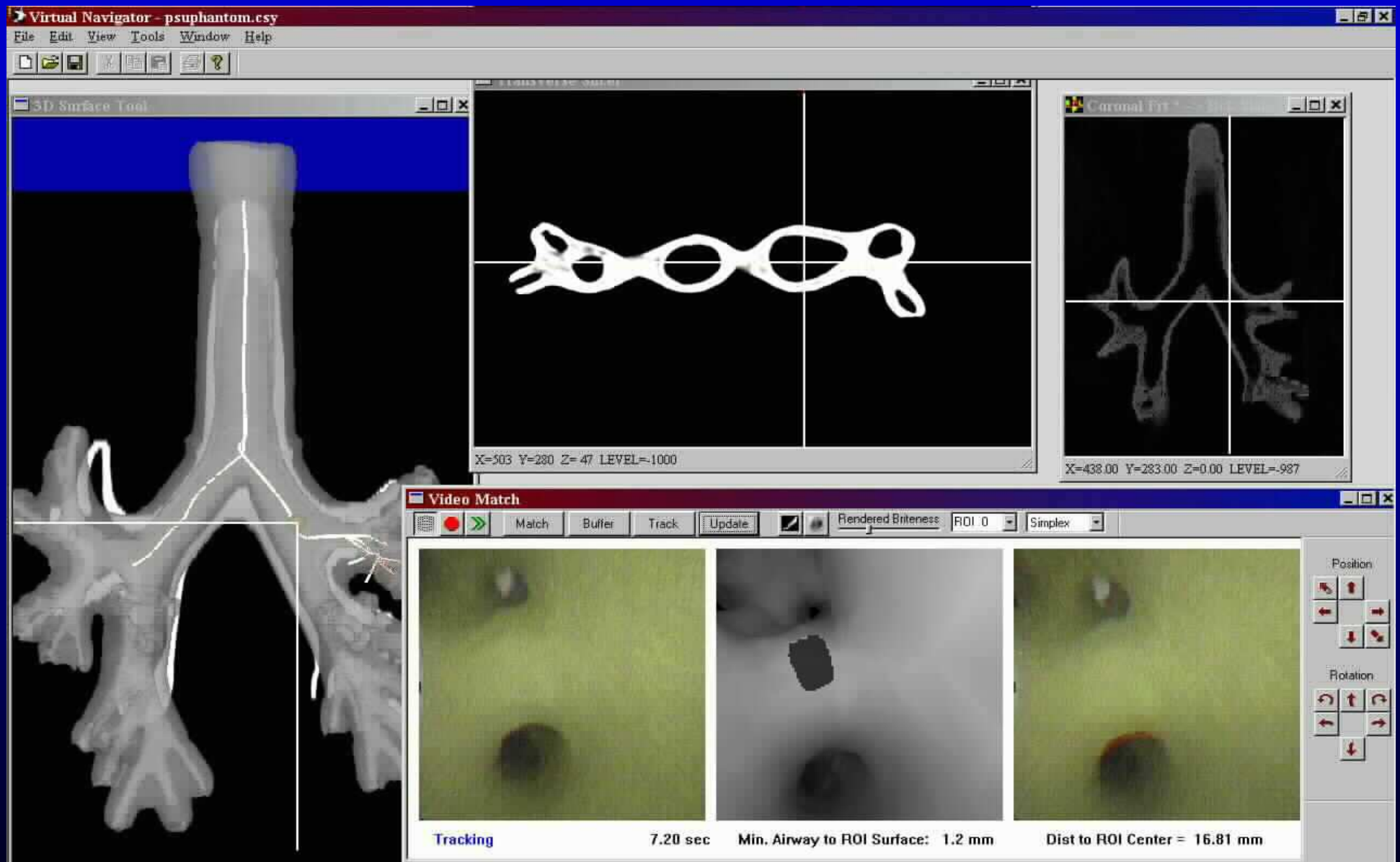
Experimental Set-up:



Rubber phantom - human airway tree model used for training new physicians.

CT Film - standard form of CT data.

Computer Set-up during Image-Guided Phantom "Biopsy"



Phantom Accuracy Results (6 physicians tested)

Film biopsy accuracy: **5.53mm** Std Dev: **4.36mm**

Guided biopsy accuracy: **1.58mm** Std Dev: **1.57mm**

Physician	film accuracy (mm)	guided accuracy (mm)
1	5.80	1.38
2	2.73	1.33
3	4.00	1.49
4	8.87	1.60
5	8.62	2.45
6	3.19	1.24

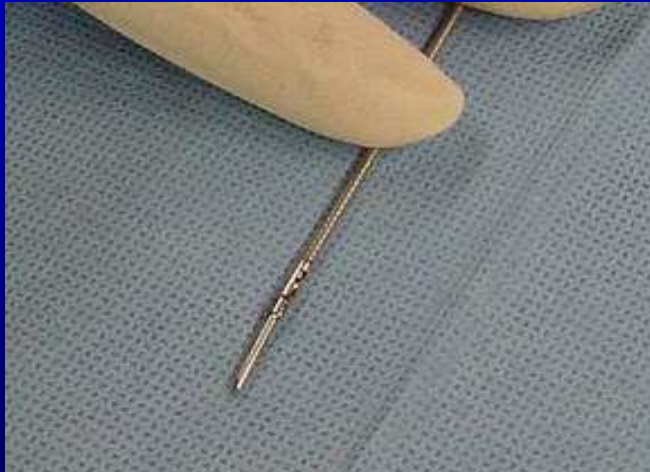
→ ALL physicians improved greatly with guidance

→ ALL performed nearly the **SAME** with guidance!

B. Animal Studies

Goals: Test the performance of the image-guided system under controlled in vivo circumstances (breathing and heart motion present).

Experimental Set-up:

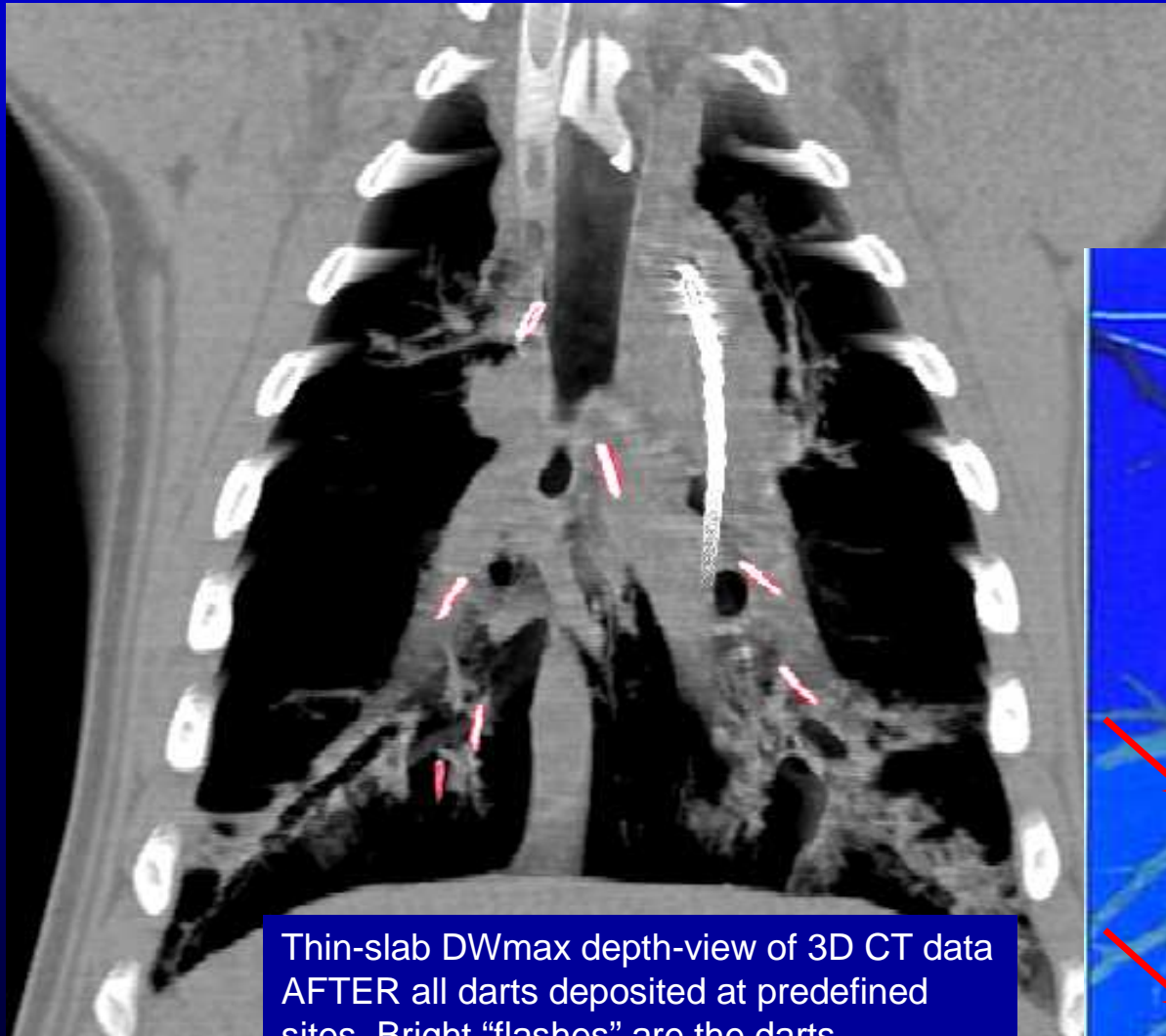


biopsy dart



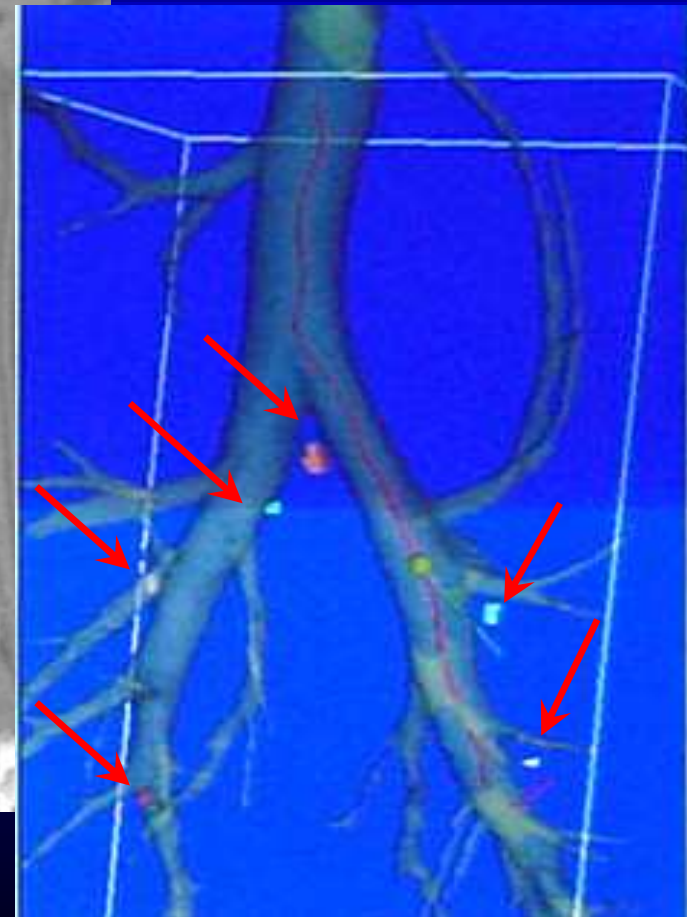
Computer system during animal test (done in EBCT scanner suite).

Composite View after All Real Biopsies Performed



Thin-slab DWmax depth-view of 3D CT data AFTER all darts deposited at predefined sites. Bright "flashes" are the darts.

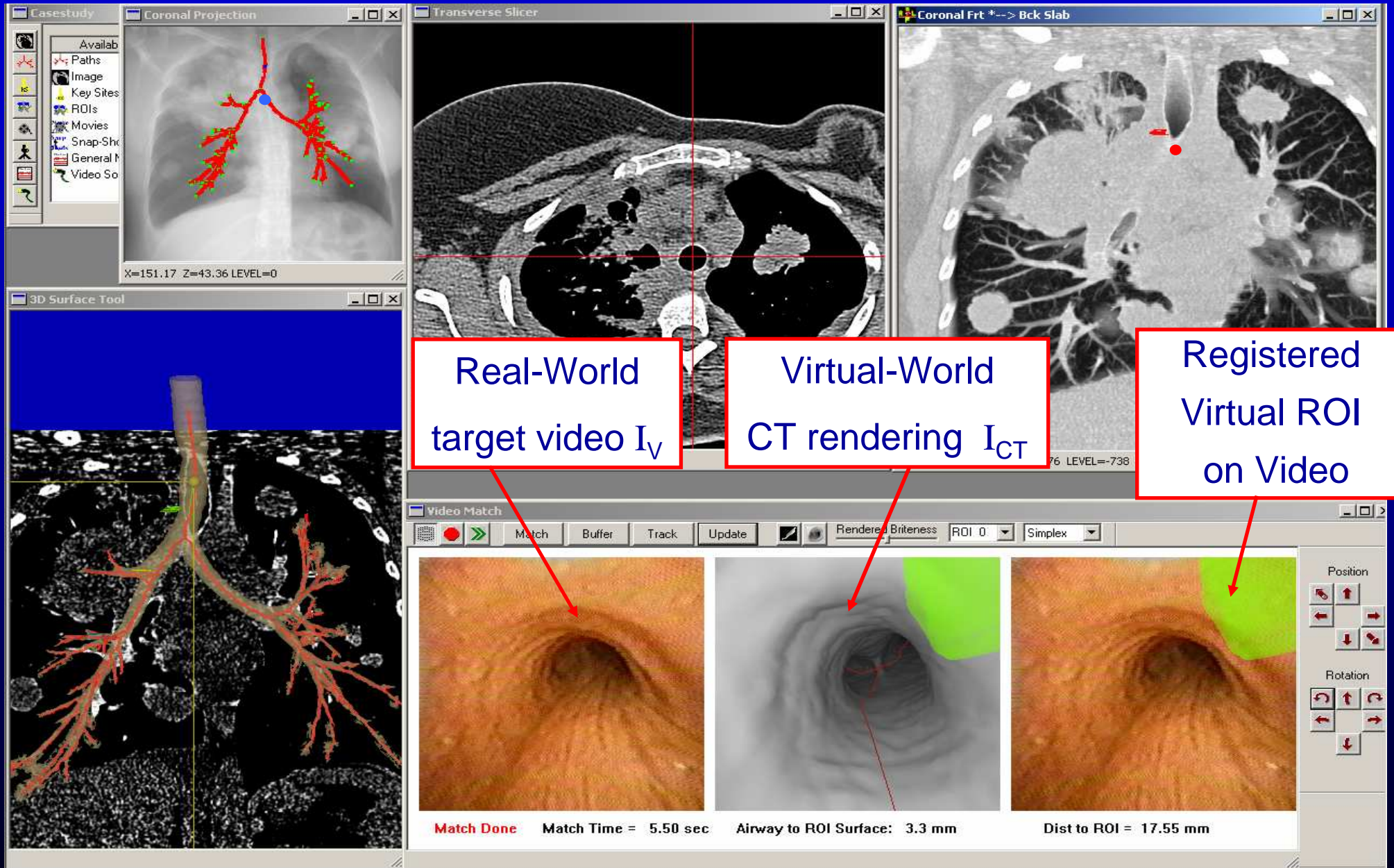
Rendered view of preplanned biopsy Sites



C. Human Studies



Stage 2: Image-Guided Bronchoscopy



(case h005 [UF], mediastinal lymph-node biopsy, in-plane res. = 0.59mm, slice spacing = 0.60mm)

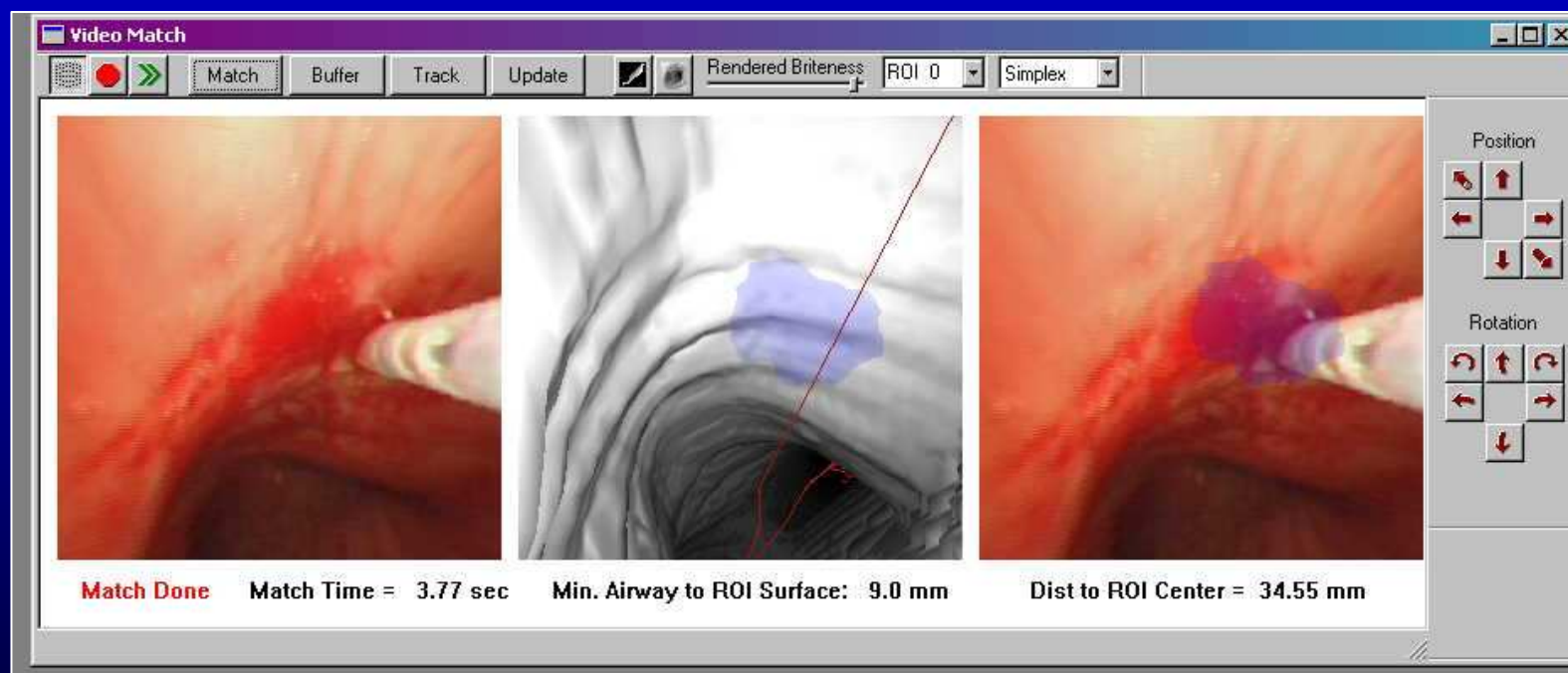
Case p1h013: performing a biopsy

Left view: Real-time bronchoscopic video view; biopsy needle in view

Center: Matching virtual-bronchoscopic view showing preplanned region (green)

Right: Preplanned region mapped onto bronchoscopic view, with biopsy needle in view.

Distance to ROI = scope's current distance from preplanned biopsy site (ROI).



→ 40 lung-cancer patients done to date

Comments on System

- Effective, easy to use
 - A technician – instead of \$\$ physician – performs nearly all operations
- Gives a considerable “augmented reality” view of patient anatomy
 - less physician stress
- Fits seamlessly into the clinical lung-cancer management process.
- Appears to greatly reduce the variation in physician skill level.

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Whitaker Foundation, Olympus Corporation

Thank You!



Bronchoscope Video Camera Model

Following Okatani and Deguchi (*CVIU* 5/97), assume video frame $I(p)$

abides by a Lambertian surface model; i.e.,

$$I(\mathbf{p}) = \sigma \frac{L \cos \theta_s}{\pi R^2}$$

where

$$\mathbf{p} = (X_p, Y_p, Z_p)$$

θ_s = light source-to-surface angle

R = distance from camera to surface point \mathbf{p}

Lung Cancer

- Lung Cancer: #1 cancer killer, 30% of all cancer deaths, 1.5 million deaths world-wide, < 15% 5-year survival rate (nearly the worst of cancer types)
- To diagnose and treat lung cancer,
 - 1) 3D CT-image preplanning – noninvasive
 - 2) Bronchoscopy – invasive
- 500,000 bronchoscopies done each year in U.S. alone
- A test for CT Image-based Lung-Cancer Screening in progress!
 - 10-30 million patient population in U.S. alone!
 - Screening is WORTHLESS if diagnosis/treatment are poor

Normalized Mutual Information

Mutual Information (MI) – used for registering two different image sources:

a) Grimson *et al.* (*IEEE-TMI* 4/96)

b) Studholme *et al.* (*Patt. Recog.* 1/99) → normalized MI (NMI)

We use normalized mutual information (NMI) for registration:

$$S_{NMI}(I_V, I_{CT}) = \frac{h(V) + h(CT)}{h(V, CT)}$$