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Data capture and tomographic reconstruction of phase microobjects

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# Data capture and tomographic reconstruction of phase microobjects

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# Outline

- Introduction: basic algorithms for tomography
- Tomography as the tool for investigation of internal structure of microelements
  - interferometric/holographic tomography
  - elastooptics tomography
- Diffraction tomography of biological microobjects - *tomorrow*







# Phase information in technology and biology



Objective for many research groups: Develop tool for quantitative characterization of 3D and 4D phase techno and bio samples in micro scale

# Why single projection is not sufficient



*Figure 12.1* Examples of sections through three different structures that produce the same 2-D section image: (a) three discrete objects; (b) one object with simple connectivity; (c) one object with multiple connectivity.

# TOMOGRAPHIC METHODS for 3D reconstruction of internal structure of transparent body



CT – computer tomography hard tissue (bones) X rays



MRI – magnetic resonance Imaging, soft tissue

ultrasound tomography Terahertz tomography Optical tomography Diffraction tomography

# Classical tomography



Single projection

Intensity in a single projection :ie

$$P(\alpha, t) = \int f(x, y) dS$$

Fourier transform of intensity

 $S(\phi, \varpi) = \int P(\phi, t) \exp(-j 2\pi \varpi t) dt$ 

Reconstruction of object function

$$f(x,y) = \iint_{0-\alpha}^{\pi \alpha} S(\phi, \varpi) |\varpi| \exp(j 2\pi \omega t) d\varpi d\phi$$

Tomography – the method to obtain information about internal properties in volume (x,y,z)

# Fourier reconstruction method



# Backprojection reconstruction without filtration





# Reconstruction by backprojection method with filtration









# Reconstruction: Backprojection method with filtration



# Typical sources of errors

# Reconstruction from small number of projections



8

32

2% noise

Presence of background in projections

Presence of noises in projections

Noncentric rotation of object

# Typical errors





Radial run-out: 2.5%

5% relative background



# Recent challenges for Optical Tomography for photonics elements

- Provide a convenient tool for 3D material properties determination in novel photonics materials and elements
- Provide experimental data for optimization of novel prototyping and production technologies e.g.
  - deep lithography with protons (DLP),
  - laser ablation and laser writing,
  - hot embossing,
  - injection molding.
- Provide a tool for reliability studies of phase photonics elements (essp. for polimer elements or elements being subjected to radiation, temperature, fatigue)
- Quantities of interest: refractive index, birefringence, residual stresses

# Internal 3D phase objects characterization



# **Diffraction/interference tomography**

### The scheme of standard ODT data aquisition system



Reconstruction of internal structure by filtered back projection algorithm  $O(x, y) = \frac{1}{(2\pi)^2} \int_{0}^{\pi} d\varphi \int_{-\infty}^{\infty} |\mathbf{k}| \widetilde{P}_{\varphi}(\mathbf{k}) \exp\left\{i\mathbf{k}(x\cos\varphi + y\sin\varphi)\right\} d\mathbf{k}$ where  $\widetilde{P}_{\varphi}(\mathbf{k}) = \int_{0}^{\infty} P_{\varphi}(\xi) \exp\left\{-i\mathbf{k}\xi\right\} d\xi$ 

or algebraic tomographic reconstruction

# **Approximations in ODT**

- The captured object projection  $P_{\varphi}$  must well approximate an object integrated phase and amplitude
- Tomographic reconstruction algorithms require linearization of the light interaction with an object
- Strong internal diffraction or refraction causes big errors

Major drawback of ODT applicability in measurement of micro optical elements refractive-index/birefrigence structure is its low dynamic range, i.e. refractive-index structures with small variations can be measured only.

# Basics of refractive index measurements tomographic microinterferometry





 $\lambda\mbox{-source}$  wavelength,  $n_I\mbox{-}$  refractive index of immersion liquid, d-object size corresponding to pixel of camera

# Interferometric tomograph



### **Technical parameters:**

- detection module: high resolution CCD matrix (1376x1035pixels)
- refraction index measurement resolution: Δn=0.0001
- the smallest object dimension 2 μm (diffraction limited)

# Phase microobjects: Technical



a)



obiektyw

#### **Problems:**

- high-numerical aperture (microlenses)
- high-phase difference
- high-phase gradient
- high spatial resolution

### **Known design (theoretical model)**







UNIA EUROPEJSKA

ROZWOJU REGIONALNEGO





**DPW** microlenses



Tapered photonics OF



**INNOWACYJNA** GOSPODARKA NARODOWA STRATEGIA SPÓJNOŚCI



# Interferometric tomography





#### 

# **Experimental results**



### **Multimodal Gradient Profile Fiber Inspection**



# Single mode fiber inspection



Fiber parameters:

- -fiber diameter 120µm
- -core diameter 8µm
- -core refractive index 1,47-cladding refractive index 1.46
- Only central core area was reconstructed propoperly

-refractive index determination error is considerable in core area, source of this error is difraction phenomenon on edge of core and cladding; step of refractive index is equal 0,01

# **Analysis of photonics fibres**













# **Photoelastic tomography**





# **Circular polariscope and its mathematical representation**



where: U - component of light vector perpendicular to polariser axis V - component of light vector parallel to polariser axis

# Integrated retardation calculation by phase shift method

General output intensity equation:

 $i = i_m + i_v \sin 2(\beta - \Psi) \cos 2\Delta - i_v \sin 2(\phi - \psi) \cos 2(\beta - \Psi) \sin 2\Delta$ 

Ψ	β	Output intensity equations	• Isoclinic angle $\Phi = \frac{1}{2} \arctan\left(\frac{i_5 - i_3}{i_4 - i_6}\right)$
0	π/4		
		$i_1 = i_a + i_b \cos \Delta$	
0	3π/4		
		$i_2 = i_a - i_b \cos \Delta$	
0	0		
		$i_3 = i_a - i_b \sin \Phi \sin \Delta$	
π/4	π/4		
		$i_4 = i_a + i_b \cos \Phi \sin \Delta$	
π/2	π/2		<ul> <li>Phase retardation</li> </ul>
		$i_5 = i_a + i_b \sin \Phi \sin \Delta$	$\Delta = -\frac{1}{2}\arctan\frac{(i_5 - i_3)\sin 2\Phi + (i_4 - i_6)\cos 2\Phi}{i_6 - i_6}$
3π/4	3π/4		
		$i_6 = i_a - i_b \cos \Phi \sin \Delta$	$\angle$ $I_1 - I_2$

### **Elastooptics tomograph**





• Expected birefringence B 0.02<B<0.06



# Experiment: birefringence in a single layer



# **3D characterization of phase photonics elements**

### Quantities of interest: n(x,y,z) or $n_0(x,y,z)$ , $n_e(x,y,z)$ , and birefringence $B(x,y,z)^*$



\*) assumption that rotation of anisotropy axes is moderate and birefringence is weak.

### **Experimental and simulation results**

### **Experiment**

### **Simulations**



# Determination of axial stress and refractive index in Panda Fiber

### Microscopic image of sample



- Panda type fiber
- cladding diameter 125 μm, stress members diameter 35 μm
- refractive index of cladding 1.4584 and for matching liquid 1.4582

Determined axial stress in sample



2D field of axial stress Plot of axia

Plot of axial stress profile

Determined refractive index in sample (for horizontal plane



2D distribution of refractive index

Plot of refractive index profile

# Tomography of biological microobjects Introduction - Motivation

- Novel tools for biotechnology
- Quantitative method for cell analysis
- The need for 3D cell analysis
   3D models of tumor cells for anticancer drug testing
- Hot topics



A scanning electron microscope picture of a nerve ending. Source: publications.nigms.nih.gov

- label-free analysis of living cells and tissues (4D)
- cellular biophysics, characterization of physical processes
- vascular and tumor biology
- recognition and monitoring of bacteria colonies
### Introduction







Human egg with coronal cells



Red blood cells



Human egg with coronal cells

Nature Reviews Molecular Cell Biology 5, 427 (June 2004)



### Phase microobjects: biological



Possible:

- high-numerical aperture
- high-phase difference
- high-phase gradient

High spatial resolution required Often high temporal resolution required (for 4D reconstruction)

Most often not known model







### Cells as measurement objects

	Prokaryotes	Eukaryotes		
Typical organisms	Bacteria, archaea	protists, fungi, plants, animals		
Type of nucleus	Nucleoid region, no real nucleus	Nucleus	Nucleus with double membrane	
Typical size	~1-10 μm	~10-100	~10-100 μm	
Refractive index	n=~1,05-1,55			
Refractive index change		∆n<0.1	high-phase gradient	

#### **Properties of bio-samples**

Polarization SensitiveBirefringence ModelsHigh-Phase GradientDepend on Choice of WavelengthsCore of the Cell importantCell boundary importantUse of Born / Rytov Models needed

#### Cell measurement and observation techniques





## Internalization of particles by phagocytosis

Phagocytosis: cellular process in which the cell internalize particles

Example: DHM phase contrast video of Chinese Hamster Ovary (CHO) cells during internalization of SiO<sub>2</sub> micro particles ( $\emptyset$  3.44 µm)



ECBO, Munich 2011

### Cell division monitoring



![](_page_42_Figure_2.jpeg)

6.3, 2012

### General approach to 3D phase reconstruction

![](_page_43_Figure_1.jpeg)

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## Standard registration method in optical diffraction tomography

![](_page_44_Figure_1.jpeg)

system

The simplest reconstruction method : filtered back projection

$$O(\mathbf{x}, \mathbf{y}) = \frac{1}{(2\pi)^2} \int_0^{\pi} d\varphi \qquad \int_{-\infty}^{\infty} |\mathbf{k}| \widetilde{P}_{\varphi}(\mathbf{k}) \exp\{i\mathbf{k}(x\cos\varphi + y\sin\varphi)\} d\mathbf{k}$$
  
where  $\widetilde{P}_{\varphi}(\mathbf{k}) = \int_{-\infty}^{\infty} P_{\varphi}(\xi) \exp\{-i\mathbf{k}\xi\} d\xi$ 

### **Approximations in ODT**

- The captured object projection  $P_{\varphi}$  must well approximate an object integrated phase and amplitude
- Tomographic reconstruction algorithms require linearization of the light interaction with an object
- Strong internal diffraction or refraction causes big errors

Major drawback of ODT applicability in measurement of micro optical elements refractive-index/birefrigence structure is its low dynamic range, i.e. refractive-index structures with small variations can be measured only.

Additional requirement for biosamples reconstruction from the limited angle of projection

![](_page_46_Picture_2.jpeg)

#### Full angle FA

- Fourier slice theorem
- Filtered backprojection
- Filtered backpropagation (FBP)
- Algebraic reconstruction technique (ART + MART)

#### Limited angle LA

- Hybrid backpropagation
- Hybrid backprojection
- Algebraic reconstruction technique (ART + MART)
- Decnovolution-iteration
- Iterative convolution backpropagation
- Iterative constraint algorithm

 Fourier-slice theorem (propagation-slice theorem)

 $F_r\{p(r,\alpha)\}(R,\alpha) = F_2\{f(x,y)\}(R\cos\alpha,R\sin\alpha)$ 

-f(x,y) – two-dimmensional object function

$$-p(r,\alpha)$$
 – projection of  $f(x,y)$ 

 $F_r\{p(r,\alpha)\} = \int_{-\infty}^{\infty} p(r,\alpha) e^{-2\pi rR} dr$  $F_2\{f(x,y)\} = \int_{-\infty}^{\infty} f(x,y) e^{-2\pi i (xX+yY)} dxdy$ 

incident

plane wave

- Filtered backprojection
  - weak scattering
  - no diffraction
  - straight-line propagation
  - $-\Delta n < 0.1$  (gradient index)
  - *∆n* < 0.03 (step index)

![](_page_49_Figure_7.jpeg)

Hybrid algorithm

T. C. Wedberg, J. J. Stamnes, and W. Singer, Appl. Opt. 34 6575-6581 (1995) C. Esmersoy and D. Miller, Geophysics 54 921-926 (1989)

 numerically backpropagation of total field to the center of the reconstruction area using inverse diffraction before the filtered backprojection algorithm is used

![](_page_50_Figure_1.jpeg)

![](_page_50_Figure_2.jpeg)

$$f_R(\mathbf{r}) = -\frac{s_0^2}{\pi a_S} \int d\hat{\mathbf{u}} \, \cos^2 \alpha \, \mathcal{H} \mathcal{G} q(\hat{\mathbf{u}}, \, t = \tau_S - s_0 \hat{\mathbf{u}} \, \cdot \, \mathbf{r})$$

• Hybrid algorithm

T. C. Wedberg, J. J. Stamnes, and W. Singer, Appl. Opt. 34 6575-6581 (1995) C. Esmersoy and D. Miller, Geophysics 54 921-926 (1989)

 numerically backpropagation of total field to the centre of the reconstruction area using inverse diffraction before the FBP algorithm is used

![](_page_51_Figure_1.jpeg)

- not sensitive to the size of the sample or the total phase delay
- sensitive to the gradient of refractive index scattering field considered

$$n_{\delta} >> \left( \nabla \varphi^{(s)} \frac{\lambda}{2\pi} \right)^2$$
, with  $\varphi^{(s)} = \ln \left( \frac{U(\vec{R})}{U^{(I)}(\vec{R})} \right)$ 

Y. Sung, W. Choi, C. Fang-Yen, K. Badizadegan, R. R. Dasari, and M. S. Feld, Opt. Express 17 266-277 (2009)

## Illustration of backpropagation algorithm

![](_page_52_Picture_1.jpeg)

• Algebraic reconstruction technique (ART)

$$\boldsymbol{\psi} = \boldsymbol{W} \boldsymbol{O} \quad \boldsymbol{\longleftarrow} \quad \boldsymbol{O}^{q+1} = \boldsymbol{O}^{q} + \frac{\boldsymbol{\psi}_{i} - \langle \boldsymbol{w}_{i}, \boldsymbol{O}^{q} \rangle}{\left(\sum_{i=1}^{MN} w_{i,j}\right)^{2}} \boldsymbol{w}_{i}, \quad \sum_{j=1}^{MN} w_{i,j} \neq 0$$

- $\langle w_i, O^q \rangle$  indicates the inner product of vectors  $w_i$  and  $O^q$ , q indicates the iteration number,  $w_i$  is the *i*-th row of the projection matrix, and  $\psi_i$ is the corresponding measured ray sum. If  $\sum_{i=1}^{n} w_{i,i} = 0$ , O is left unchanged.
- Multiplicative algebraic reconstruction technique (MART)

$$O_{j}^{q+1} = R_{j}^{q}O_{j}^{q} \quad \text{for } j = 1, \dots, MN \qquad \begin{bmatrix} R_{j}^{q} = 1 - \lambda w_{i,j}^{*} \left( 1 - \frac{\psi_{i}}{\langle \mathbf{w}_{i}, \mathbf{O}^{q} \rangle} \right), \quad \langle \mathbf{w}_{i}, \mathbf{O}^{q} \rangle \neq 0 \\ R_{j}^{q} = 1, \qquad \text{otherwise.} \end{bmatrix}$$

- The normalized weight  $w^*_{ij}$  is equal to  $w_{ij}/w_{max}$ , where  $w_{max}$  is the largest element of the projection matrix W and  $\lambda$  is a relaxation parameter. The ray sum number *i* is given by  $(q \mod P) + 1$ .

D. Verhoeven, Appl. Opt. 32 3736-3754 (1993)

Deconvolution-iteration - Fourier-transform-iteration Detection cone scheme for filling in missing of at P Vertical edge Fourier components **Known Fourier Estimated Fourier** spectrum of the FFT components of the object object Estimated A priori FFT<sup>-1</sup> object knowledge density

K. C. Tam and V. Perez-Mendez, J. Opt. Soc. Am. 71 582-592 (1981)

### **Reconstruction from limited AP**

![](_page_55_Picture_1.jpeg)

![](_page_55_Picture_2.jpeg)

- Iterative convolution backprojection
  - *a priori* knowledge about
    the image and line-integral
    data constraints

![](_page_56_Figure_3.jpeg)

![](_page_56_Figure_4.jpeg)

B. P. Medoff, W. R. Brody, M. Nassi, and A. Macovski, J. Opt. Soc. Am. 73 1493-1500 (1983)

- Iterative constraint algorithm
  - based on iterative convolution backpropagation and deconvolution-Fourier iteration
  - used modified Fourier diffraction theorem  $\hat{F}(K_x, K_y, K_z) = \frac{ik_z}{\pi} \hat{U}^{(S)}(k_x, k_y; z^+ = 0)$   $\hat{F}(K_x, K_y, K_z) = \frac{i(K_z + k_{z0})}{\pi} \hat{U}^{(S)}(K_x + k_{x0}, K_y + k_{y0}; z^+ = 0)$
  - Born approximation  $\hat{U}^{(S)}(K_x + k_{x0}, K_y + k_{y0}; \theta) = \iint (U(x, y; \theta) - U_{bg}(x, y; \theta)) / U_{bg}(x, y; \theta) e^{-iK_x x - iK_y y} dx dy$
  - $\frac{\text{Rytov approximation}}{\hat{U}_{Rytov}^{(S)}(K_x + k_{x0}, K_y + k_{y0}; \theta)} = \iint \ln(U(x, y; \theta) / U_{bg}(x, y; \theta)) e^{-iK_x x iK_y y} dxdy$

- Iterative constraint algorithm
  - a) slice image before application of the constraint algorithm
  - b) same slice image as in (a) after application of the non-negative constraint
  - c) same slice image as in (b) after 100 iterations
  - d) amplitude distribution in K<sub>x</sub>-K<sub>y</sub> plane before application of the constraint algorithm

![](_page_58_Figure_6.jpeg)

- e) 3D Fourier transform of tomogram after non-negative constraint
- f) 3D Fourier transform of tomogram after 100 iterations

Y. Sung, W. Choi, C. Fang-Yen, K. Badizadegan, R. R. Dasari, and M. S. Feld, Opt. Express 17 266-277 (2009)

• Biological objects measurements:

![](_page_59_Figure_2.jpeg)

### Digital Holographic Shearing Microscope

![](_page_60_Figure_1.jpeg)

![](_page_60_Picture_2.jpeg)

B. Kemper, A. Vollmer, C. Rommel, J. Schnekenburger, G. von Bally, J. Biomed. Opt. 16, 026014 (2011)

### Holographic Microscope

DHM Tomographic measurement procedure

![](_page_61_Figure_2.jpeg)

To convert DHM into tomograph we need to add specimen rotation to capture several projections

![](_page_62_Figure_1.jpeg)

### Specimen rotation

• Cell cultivation tools (Petri dishes) limit the angle of observation

![](_page_63_Picture_2.jpeg)

• Manipulate the specimen without optical system modifications

### Specimen rotation -requirements

- Fast and accurate living cells rotation
- Rotation perpendicular to optical axis
- Versatile concept
- Module-based built
- Integration with a microscope system
- Applicable to tomographic setup

### Alternate illumination angle

![](_page_65_Figure_1.jpeg)

## Tomographic phase microscope with varying illumination angle

GM, galvanometer scanning mirror; L1, focal length f=250 mm lens; BF, backfocal plane of the condenser lens; C, condenser lens; S, sample; OL, objective lens; L2, f=200 mm lens; AOM1 and 2, acousto-optic modulators; BS1 and BS2, beam splitters. The frequency-shifted reference laser beam is shown as darkned after AOM

Choi W, Yu CC, Fang-Yen C, Badizadegan K, Dasari RR, Feld MS. Field-based angleresolved light-scattering study of single live cells. Opt. Lett. 2008;33:1596–1598

### Illumination angle

![](_page_66_Figure_1.jpeg)

Isikman, S.O. et al., 2011. Lens-free optical tomographic microscope with a large imaging volume on a chip. *Proceedings of the National Academy of Sciences of the United States of America*, 108(18), pp.7296-301.

### Specimen rotation

![](_page_67_Figure_1.jpeg)

### Mechanical

- Single cell mounted on a rotating micropipette
- •Cells inside a rotating hollow fibre

![](_page_67_Figure_5.jpeg)

![](_page_67_Picture_6.jpeg)

Microscope lens

![](_page_67_Picture_8.jpeg)

- Double trap Optical Cell Rotator
- Two independent optical traps rotation about any desired axis.

![](_page_67_Figure_11.jpeg)

![](_page_67_Figure_12.jpeg)

### Specimen rotation: comparison

#### Mechanical rotation (fiber)

- No additional optical components required
- Fast and uncomplicated software
- Living cells observation

#### **Optical tweezers**

- Only one immersion liquid
- Rotation about any desired axis
- Large number od traps rotating many cells simultaneously
- High precision

- Refractive indices matching (fluids)
- Cells tend to attach to fiber's walls
- Axial runout mechanical adjustment
- Particles can move inside the fiber fluid density related

- High power coherent light source required (100 mW per trap in specimen plane)
- Expensive components
- Risk of cell destruction

### **Optical tweezers**

![](_page_69_Figure_1.jpeg)

- Light refracted through a transparent object imparts momentum to the object to balance its change in direction.
  - At Rayleigh size scale, the electromagnetic field (E) of the light causes an object to act as an induced dipole (p), which is drawn into the brightest part of the beam (the focus) where its energy is minimized.

Dholakia, K. & Reece, P., 2006. Optical micromanipulation takes hold Light can influence the motion of particles , from the size of a single cell forefront of many studies in the natural sciences . , 1(1), pp.18-27.

### **Optical trapping systems**

![](_page_70_Figure_1.jpeg)

 $\tilde{A},$  X.-cheng Y.A.O. & Hang, D.-zhong Z., 2004. Micro-Rotation by Flow-Induced Torque in an Optical Trap. , 11(1), pp.4-6.

#### **Basic optical trapping systems**

- High NA objectives (100x NA 1.3)
- Optical fiber trapping systems

![](_page_70_Figure_6.jpeg)

Ashok, P.C. & Dholakia, K., 2012. Optical trapping for analytical biotechnology. *Current opinion in biotechnology*, 23(1), pp.16-21.

### **Optical cell rotator**

#### **Optical particle rotation**

- Dual beam 1064 nm trap
- Beam shaping using single and dual mode fiber
- Mechanical fiber rotation
- Intermediary solution both mechanical and optical components

![](_page_71_Figure_6.jpeg)

Kreysing, M.K. et al., 2008. The optical cell rotator, Optics express, 16(21), pp.912-914
# **Optical trapping systems**



Curtis, J. & Grier, D., 2003. Structure of Optical Vortices. *Physical Review Letters*, 90(13), pp.13-16.

- Holographic Optical Tweezers (HOT) concept
- Large number of traps
- Versatile use
- Beam profile modification
- SLM required

# Optical trapping systems

### Particle rotation in HOT

- Two traps for one object
- 3-D rotation
- 3-D translation



Bingelyte, V. et al., 2003. Optically controlled threedimensional rotation of microscopic objects. *Applied Physics Letters*, 82(5), p.829.



### **Mechanical rotation**





 $\nabla \nu$ 

Fig. 2. Holographic microscope for transmission imaging: NF neutral density filter; PBS polarizing beam splitter; BE beam expander with spatial filter;  $\lambda/2$  half-wave plate; MO microscope objective; FL field lens; M mirror; BS beam splitter; O object wave; R reference wave; MP micropipette; CS coverslip; S specimen; IL immersion liquid. Inset: a detail showing the off-axis geometry at the incidence on the CCD.

Charrière, F. et al., 2006. Living specimen tomography by digital holographic microscopy: morphometry of testate amoeba. *Optics express*, 14(16), pp.7005-13.

## Mechanical rotation

#### **Mechanical particle rotation**

- Hollow core fiber capillary
- Rotational fiber holder
- Inverted microscope
- Immersion liquid dish that would allow a fibre rotation





# Specimen rotation - conclusion



### Mechanical specimen rotation









Gauge	Needle ID [µm]	Capillary OD[µm]	Capillary ID [µm]
25	260	250	140
24	311	300	111/212
23	337	340	128

## Cell culture preparation



### Cell culture preparation

object	PVA coated	inside collagen 0,08%	inside collagen 0,16%	inside Agar 0,15%	inside Glycerin	outside immersion oil	Outside Phosphat Buffered Saline	Inner diameter [µm]	Incubation time	result
HT1080 fibroblasts	+	+				+		128	1h	most cells stick to the wall
HT1080 fibroblasts	+				+	+		128	0h	cells stick to the wall and shrink
Agarose beads 30µm					+	+		111	0h	good results, less diffraction
U937 Human Leukemia	+			+			+	212	0h	good results, some cells are in the middle of the fiber, diffraction not important
HT1080 fibroblasts	+		+				+	212	24h vertical	good results, most cells stick to the wall but some are in the middle of the fiber
HT1080 fibroblasts		+		+		+		111	24h vertical	good results, strong cells, centered

# Cells under study

#### Fibrosarcoma HT1080

- Fibroblastic sarcoma malignant mesenchymal tumor derived from fibroblasts
- Fibroblasts are the most common cells of connective tissue



#### Pancreatic Tumor PaTu-8988T

- Human pancreas tumor
- A suitable model for Adenocarcinoma Pancreas Tumor studies
- The most common pancreatic cancer

#### Human Leukemia U937

- Lymphoma cancer of the lymphocytes
- Are used to study the behaviour and differentiation of monocytes







#### PaTu8988-T cells 0.07 $212/300 \ \mu m$ capillary 0.06 👍 Volume Viewer 1.31 -Thermal LUT Volume I (slow) z-aspect 1.0 Save View 150 Dist -1 0.05 100 P -0.04 50 Depth: 10 0.03 0 0 50 0 ¢ 0.02 50 100 100 Thr.: 28 150 150 0.01 200 200 ¢ Scale: 2.46 Angle of rotation x: 116 z: 98 Markers Axes $\square$ ху yz XZ

### **HT1080 Fibrosarcoma** 212/300 μm capillary









## Future work

- Resolve the matter of callibration and absolute refractive index value determination
- Reduce noise and diffraction artifacts
- Improve the algorithms for bigger phase gradients
- Improve algorithms for rec. with limited angle projections
- Introduce optical manipulation
- Adapt system for dynamic measurements

Final goal – one shot tomographic camera system for dynamic processes such as infections

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