

Two-Dimensional Phase Unwrapping Using Neural Networks

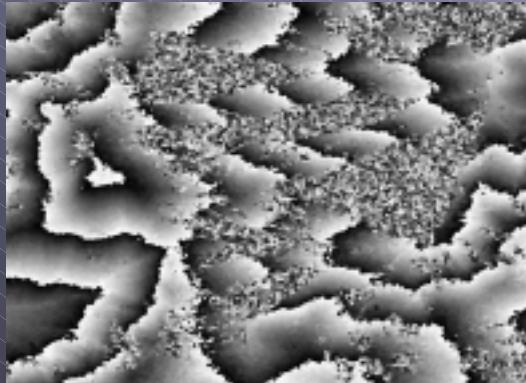
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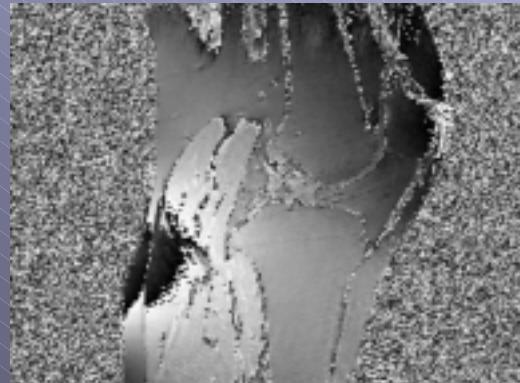
Phase-Based Imaging Methods

Synthetic
Aperture Radar



Mountains

Magnetic
Resonance
Imaging



Knee

Optical Doppler
Tomography



Blood Vessel

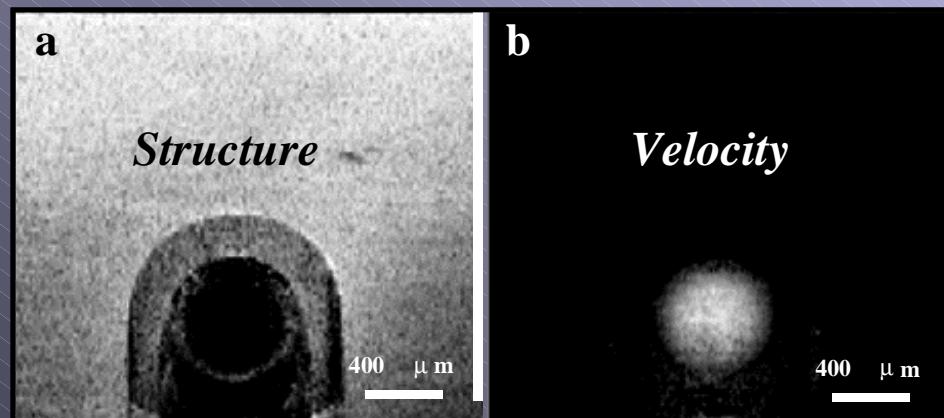
Optical Doppler Tomography (ODT)

- Uses lasers to image skin tissue
- Obtains two images
 - Structure: proportional to amplitude
 - Velocity: proportional to phase

Surface



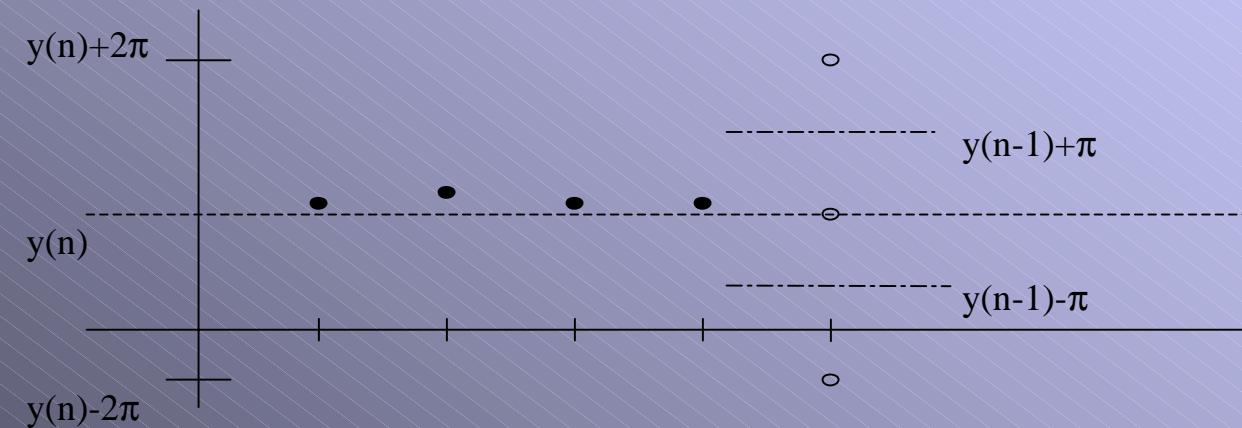
Depth



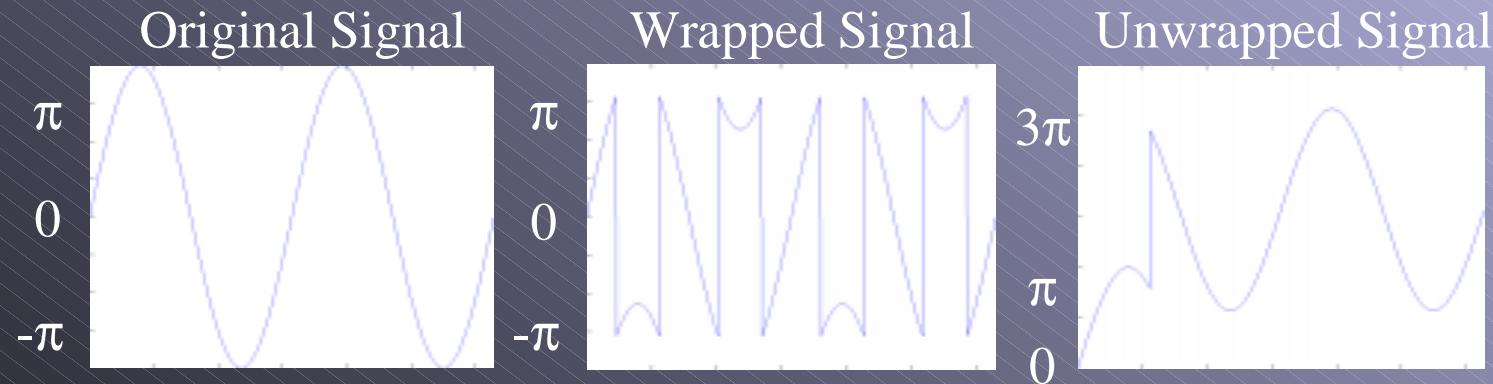
OCT images of
polystyrene microspheres

1-D Phase Unwrapping

- Use information from neighboring pixels



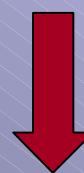
- Errors propagate



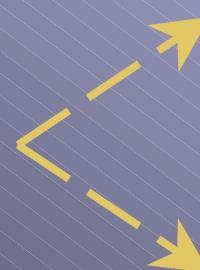
2-D Phase Unwrapping

- Possible to use second dimension to recover
- Residues: conflicts in different dimensions

0.0	0.6π
1.6π	1.2π



0.0	0.6π
-0.4π	1.2π



Two of the
possible orderings

0.0	0.6π
1.6π	1.2π

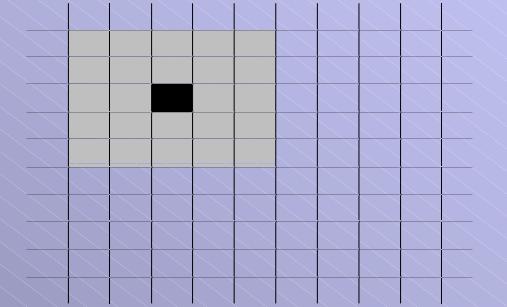


Why Neural Networks?

- Fast unwrapping of ODT velocity images
 - 10 100×100 images per second
 - Global optimization methods impractical
 - Fast local methods necessary
- Learn features in ODT velocity images of skin tissue
 - Sparse images
 - All blood flow follows parabolic law
- Goal: Train neural networks to detect phase jumps in ODT phase images

Phase Jump Detector

- Feedforward multilayer perceptron network
- Wrapped values of 5×5 neighborhood of pixels as inputs to network
- Three outputs for the current pixel
 - Positive jump of 2π
 - No jump of 2π
 - Negative jump of 2π
- Non-iterative, pixel parallel computation

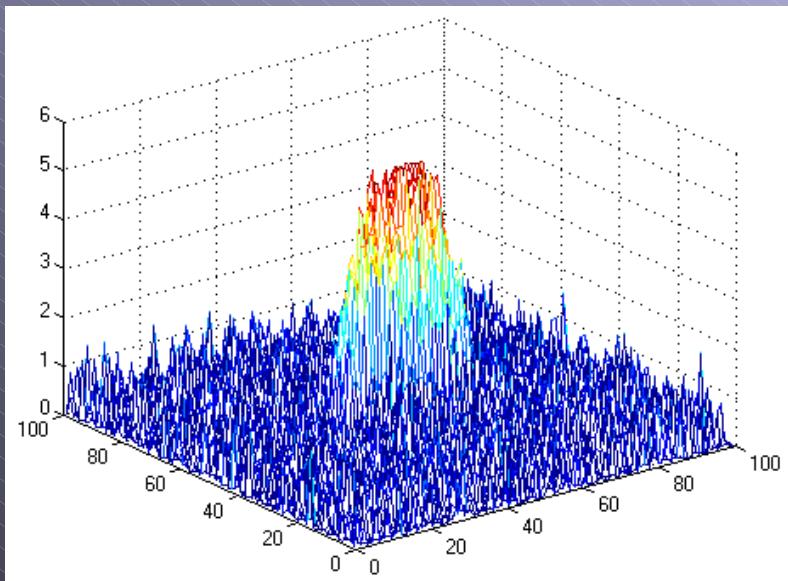


- Neighboring pixels
- Current pixel

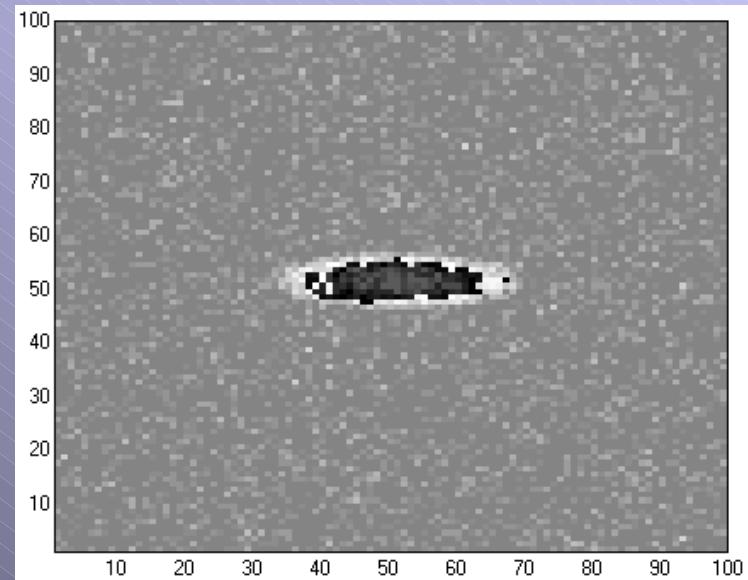
Phase Jump Detector

- Neural network parameters
 - 25 input neurons (5×5 image block)
 - 5 hidden neurons
 - 3 output neurons
- Training of neural network
 - Train using 90 simulated ODT images with conjugate-gradient method
 - Validate with 10 simulated ODT images
 - Simulated images have variety of vessel sizes and shapes
- Tested on 100 simulated ODT images

Simulated Image



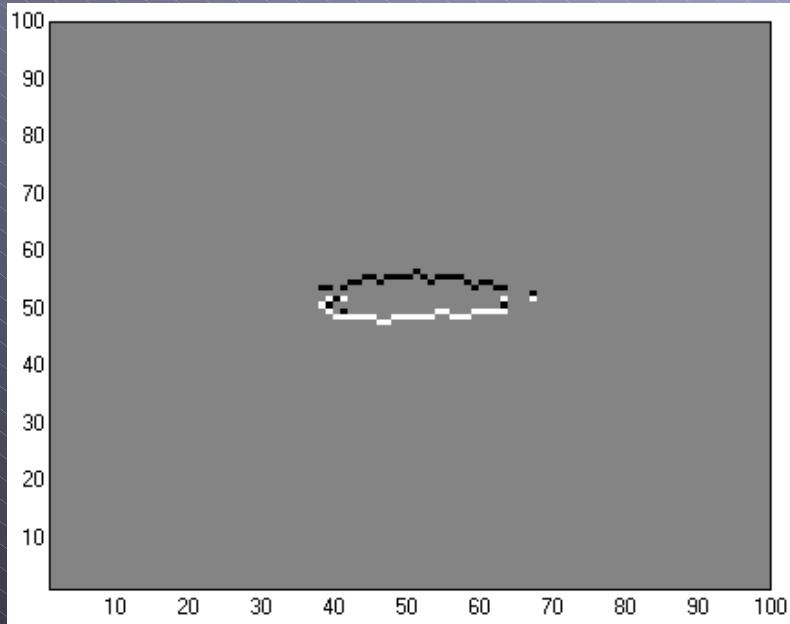
Simulated flow in a blood
vessel without wrapping



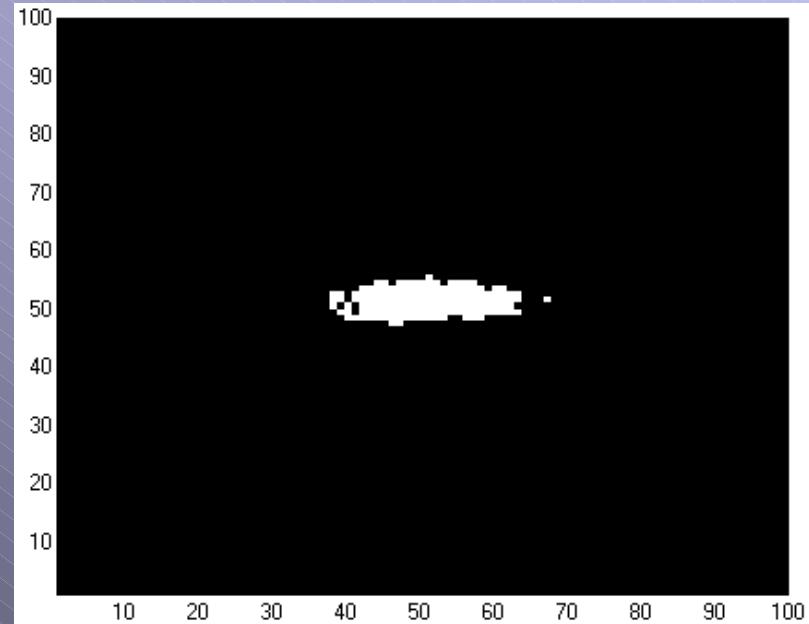
Simulated flow in a blood
vessel after wrapping

Phase Jump Detection

- Unwrap along columns
 - Top pixel is always zero since no movement at edge of skin
 - At positive (negative) jump, add (subtract) 2π to rest of column



Location of positive (black) and negative (white) phase jumps



Pixels in the range $[\pi, 3\pi)$

Confusion Matrix

Classified

		Positive	No Jump	Negative
A c t u a l	Positive	0.89	0.11	0
	No Jump	0.0004	0.9988	0.0008
	Negative	0	0.09	0.91