

Detection of Cervical Cancer from Evoked Tissue Fluorescence Images Using 2- and 3-way Methods

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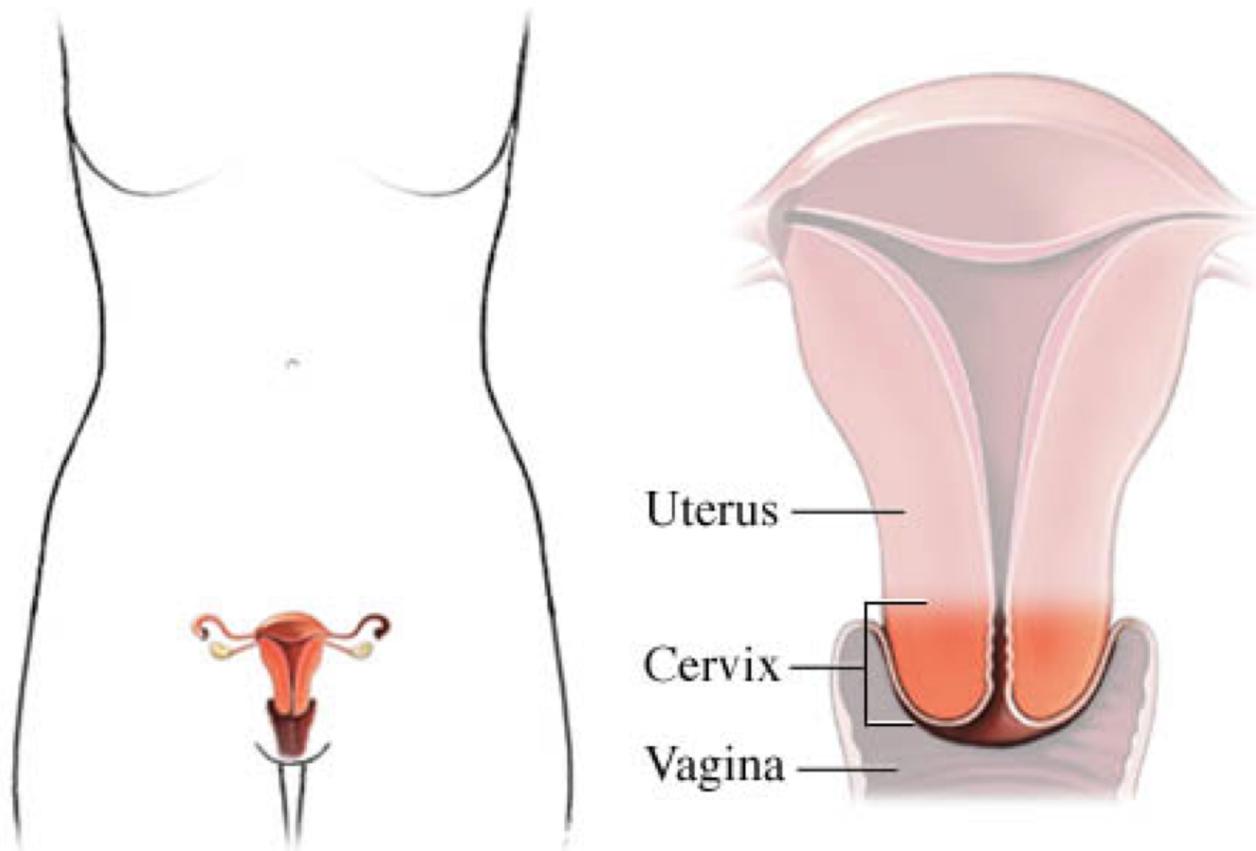
Cervical Cancer

- Pap smears credited with reducing cervical cancer mortality by detecting pre-cancerous cells, but...
- Sensitivity of Pap smears reported as 29-56%
- Abnormal Pap smear-> colposcopic examination, but....
- Colposcopy success depends on interpretation and therefore experience of examiner
- Colposcopic impressions correlate with biopsies as little as 35% of the time

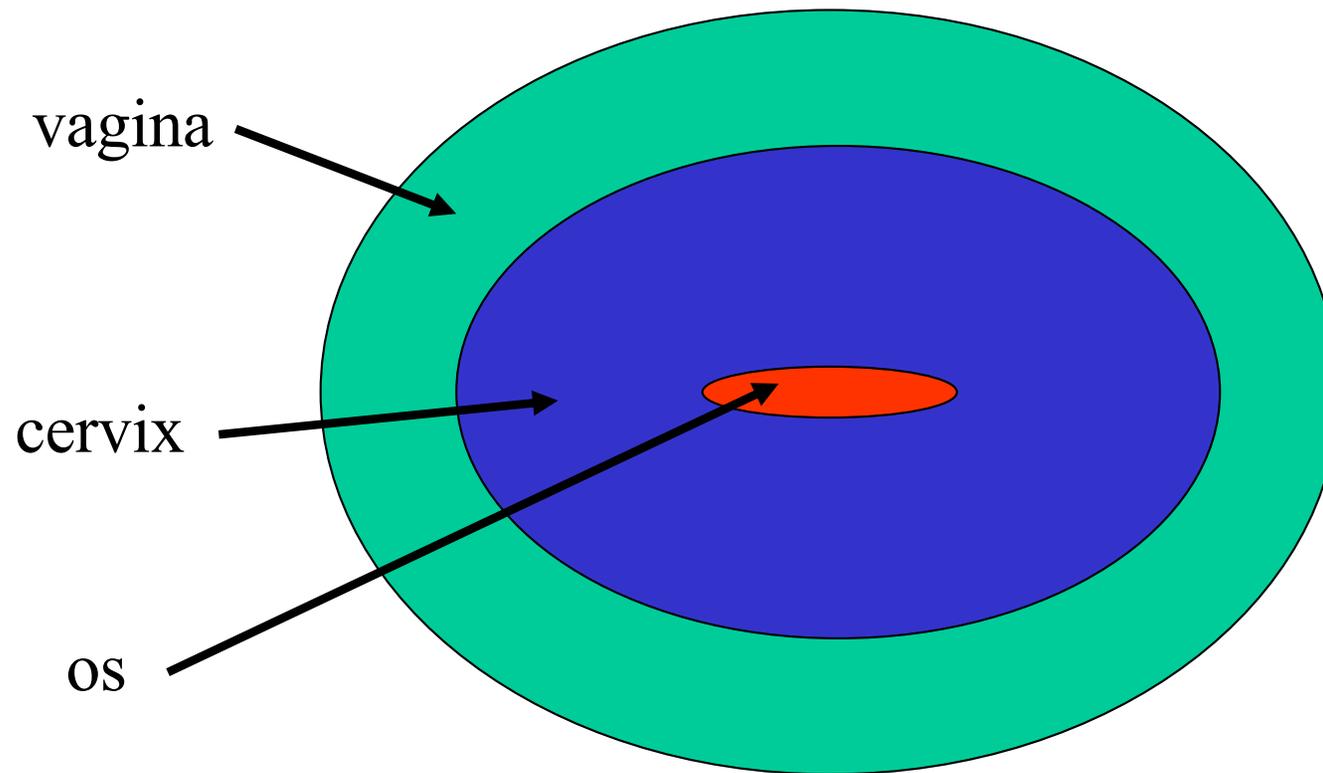
Goal of this work

- Develop a method to classify cervical tissue
 - More sensitive and specific than current methods
 - Doesn't require high level of experience to use
 - Can be easily administered

The Cervix



Images of the Cervix



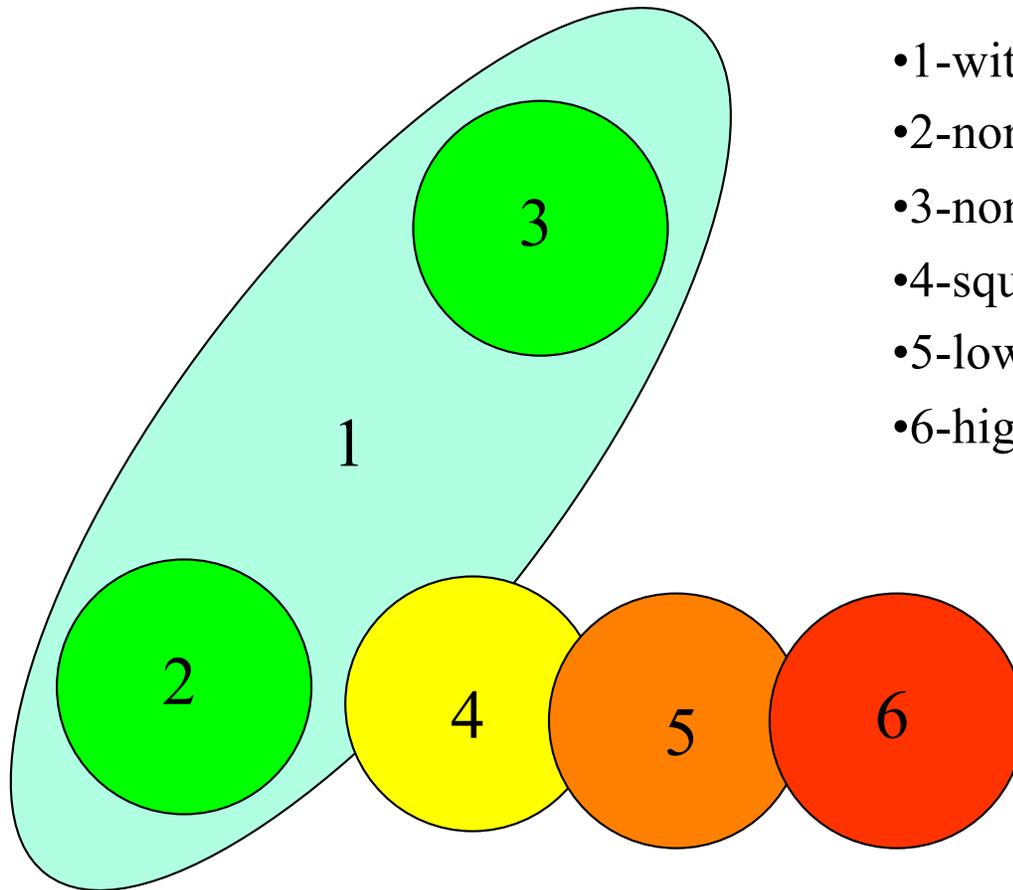
The Data

- Colposcopic Images
 - Interpreted by experts
 - Areas of tissue type identified
- Biopsies
 - Tissue type confirmed using staining and microscopy
 - Areas identified on images
 - “Gold Standard”
- Evoked Tissue Fluorescence Images
 - Excitation Emission Fluorescence Images

Classes of Tissue

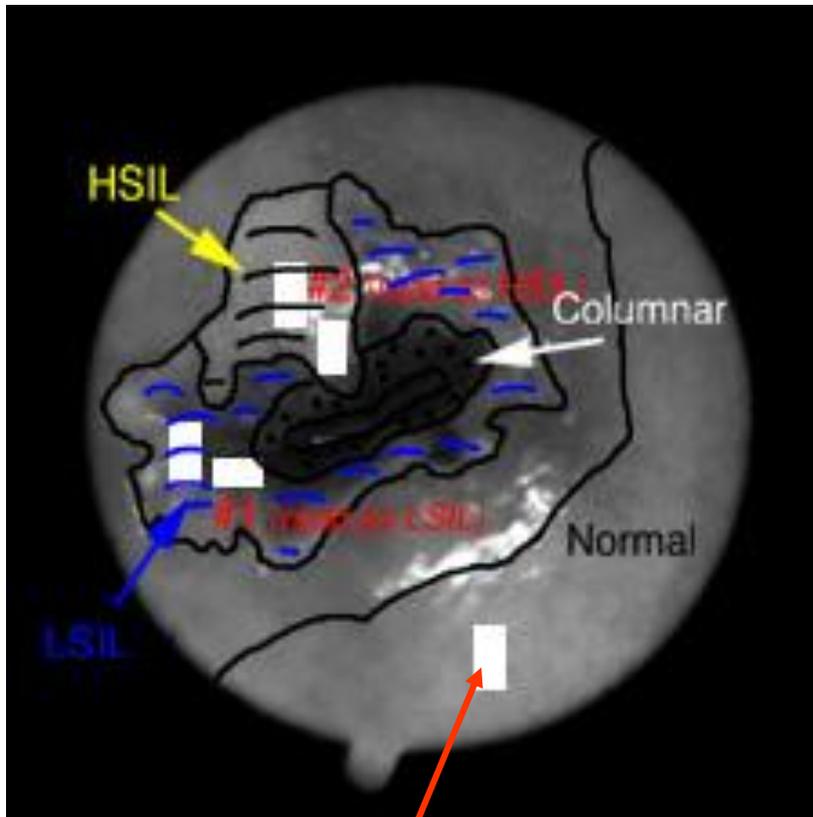
- 1-within normal limits
- 2-normal squamous
- 3-normal columnar
- 4-squamous metaplasia
- 5-low SIL (Squamous Intraepithelial Lesion)
- 6-high SIL

Similarity of Tissue Types

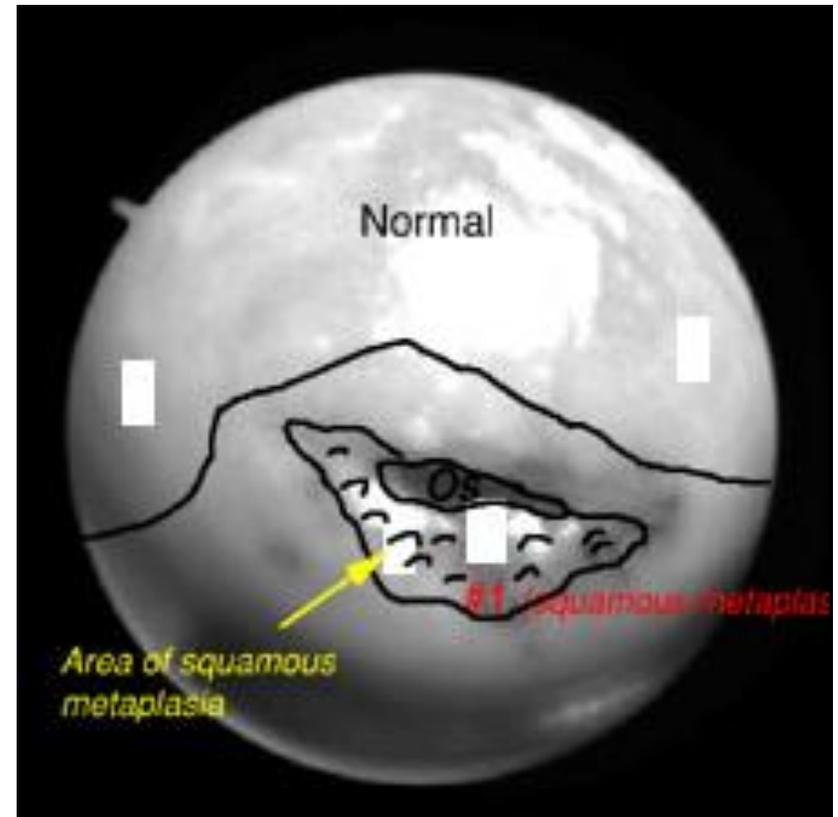


- 1-within normal limits
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Colposcopic Images and Biopsies



Biopsy locations



The ETF Images

- Combinations of
 - 3 excitation wavelengths
 - 9 emission wavelengths
 - 22 combinations measured

		Emission (nm)								
		392	417	452	495	515	580	610	640	670
Excitation	337									
	380									
	460									

Not Measured

Preprocessing Issues

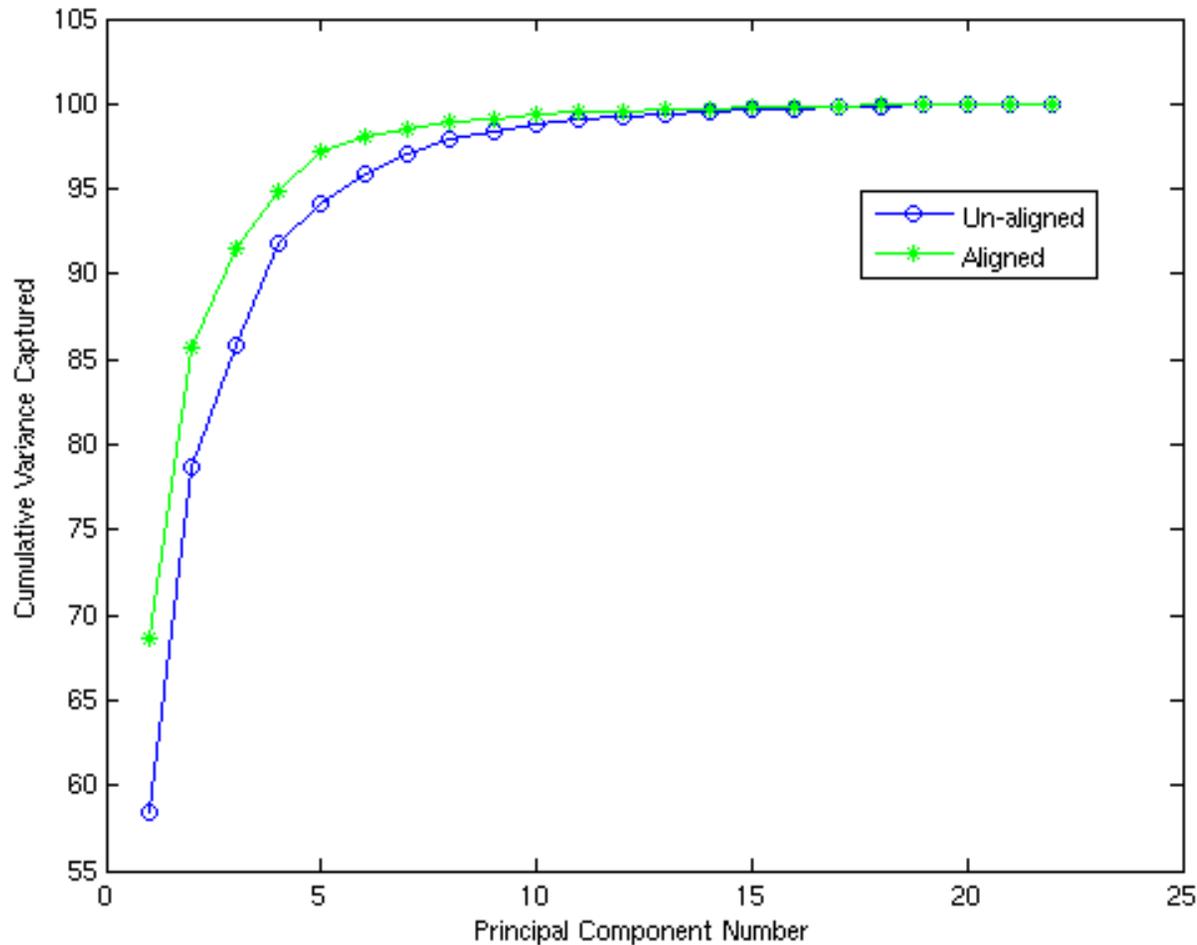
- Image alignment
 - Measurements take about 60s
 - Patient movement an issue
- Patient to patient variability

Image Alignment

- Images at different wavelengths look different
- ...but sub-images should be correlated
 - Should be most correlated when properly aligned
 - Want big PCs to get bigger and small ones to get smaller
- Used Varimax criteria on singular values:

$$\frac{\sum_{i=1}^n s_i^4}{\left(\sum_{i=1}^n s_i\right)^4}$$

Variance Captured Before and After Alignment



Model Development

- Align images
- Center to normal squamous tissue on each patient /
OR center to mean of all tissues
- Pool all patients - center & scale
- **OPTIONALLY:** GLS deweighting based on a
single class

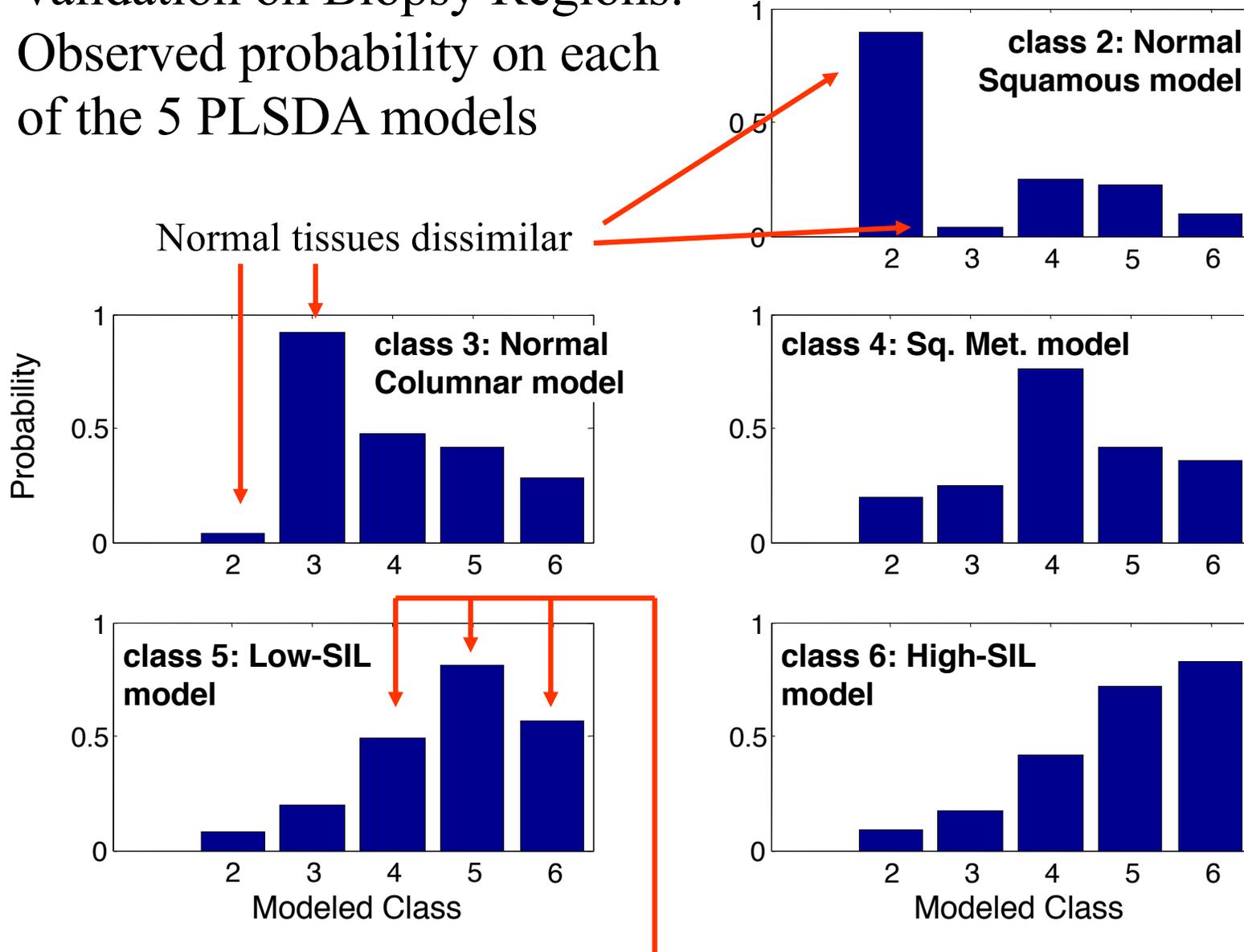
Calibration

- PLS-DA on EACH CLASS
- take predicted y for each class and
 - threshold
 - convert levels of disease to SIL scale

Application

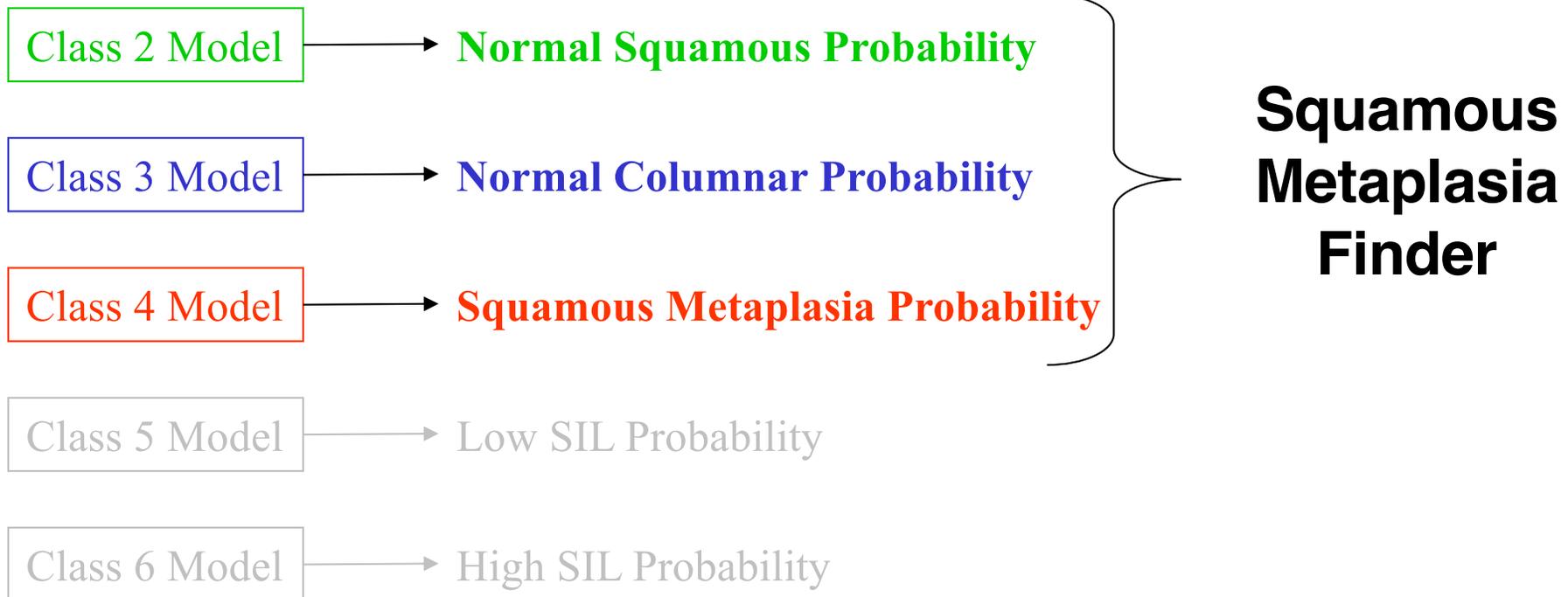
- Align
- Center each patient image to its own mean
(NOTE: high levels of SIL will bias)
- Apply model
- Identify absolute “normal”, repeat (1)-(2) using centering to NORMAL

Validation on Biopsy Regions: Observed probability on each of the 5 PLSDA models

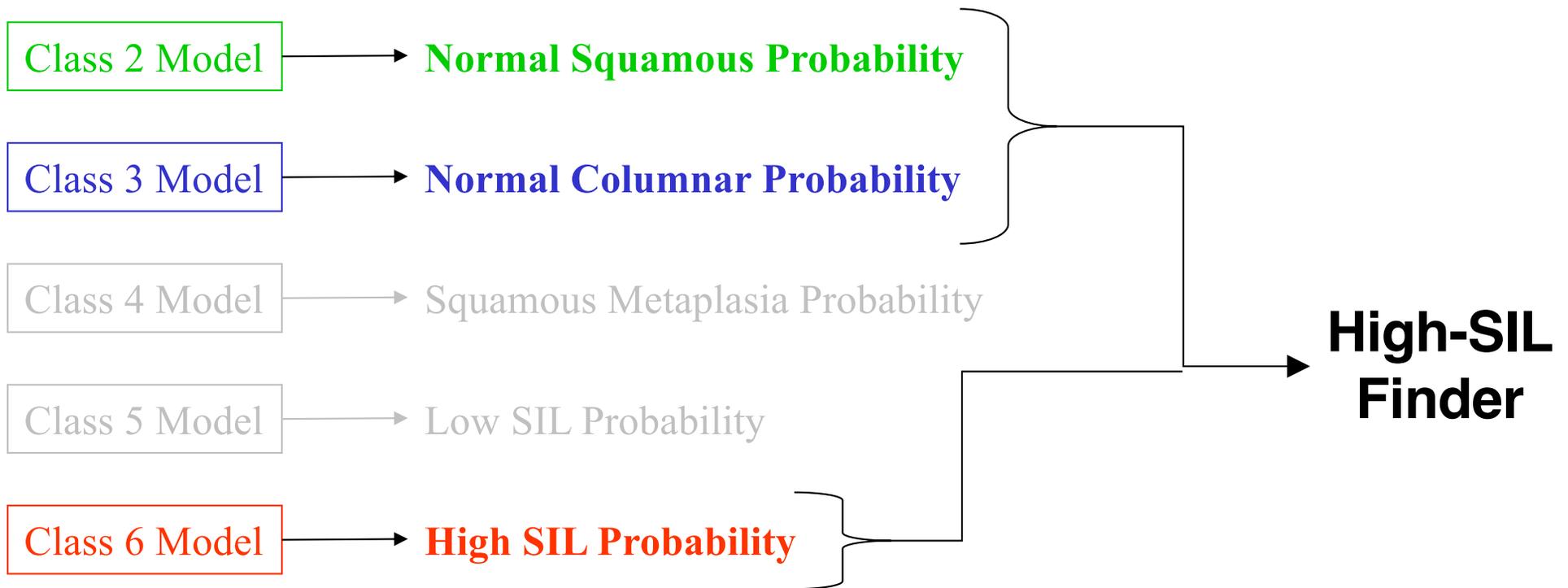


Three disease states very similar!

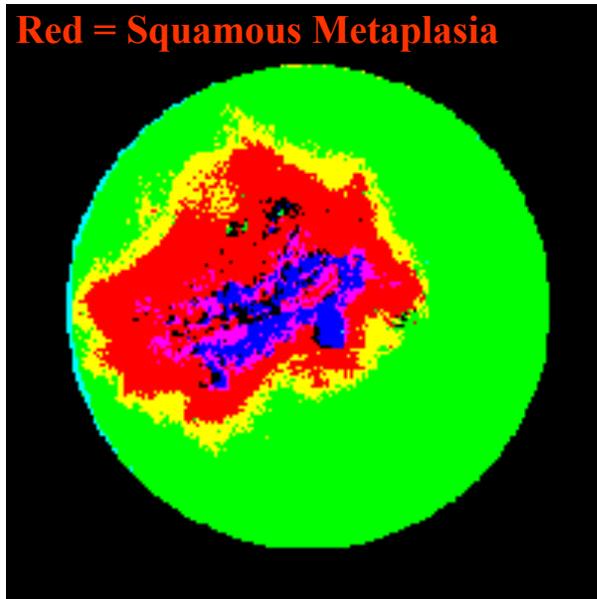
Diagnostic False-Color Images



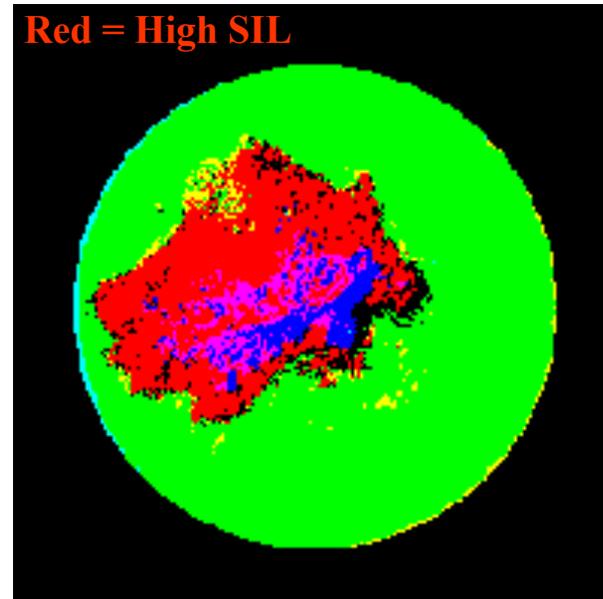
Diagnostic False-Color Images



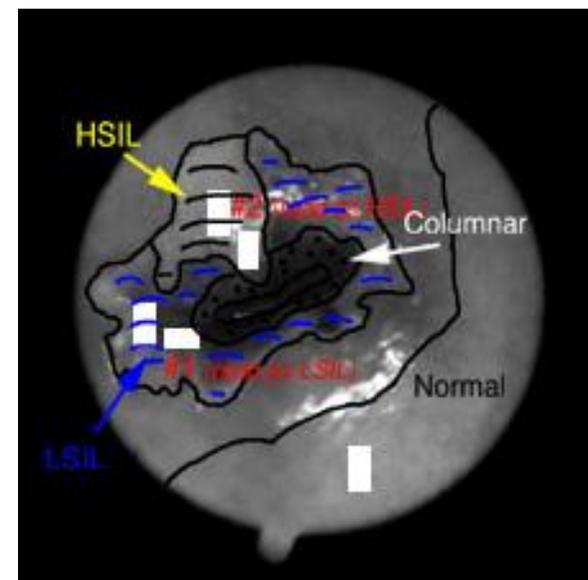
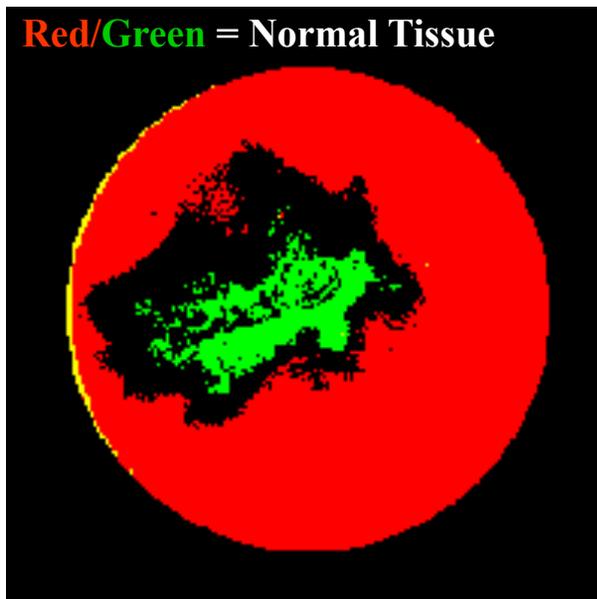
Squamous Met. Finder - Classes: 4,2,3



Hi-SIL Finder - Classes: 6,2,3



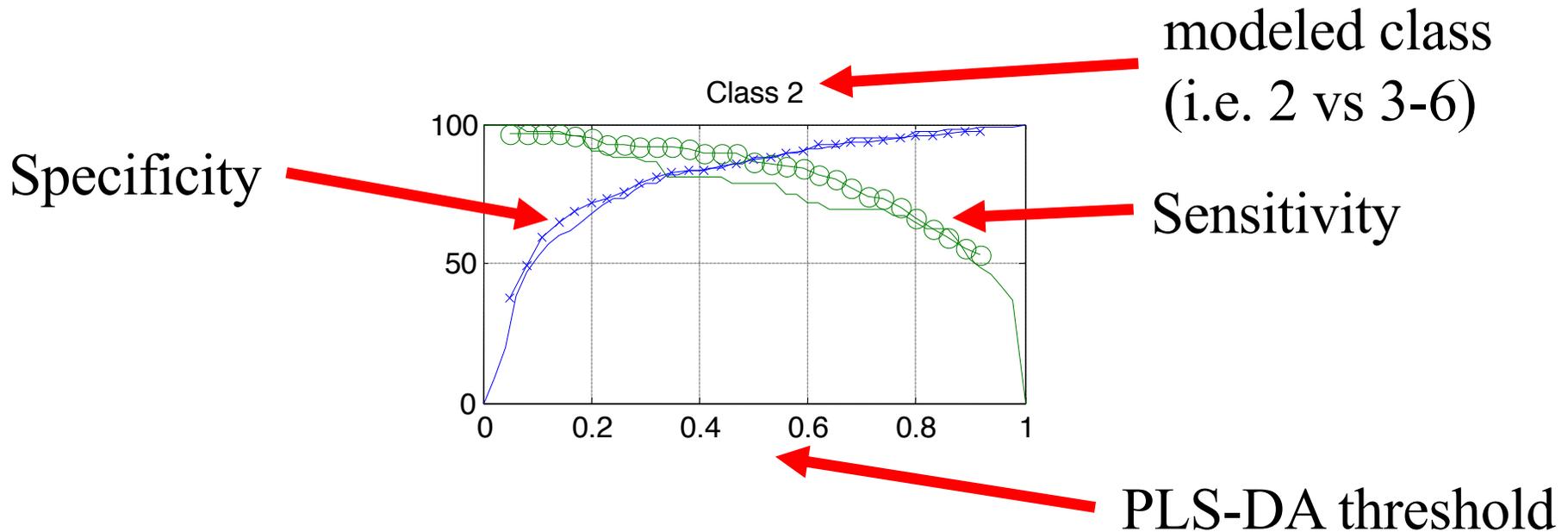
Normal Tissue Finder - Classes: 2,3



aD047

red, green, blue

Key to ROC Plots



sensitivity / specificity (at crossing)

n: = results for class "n" model

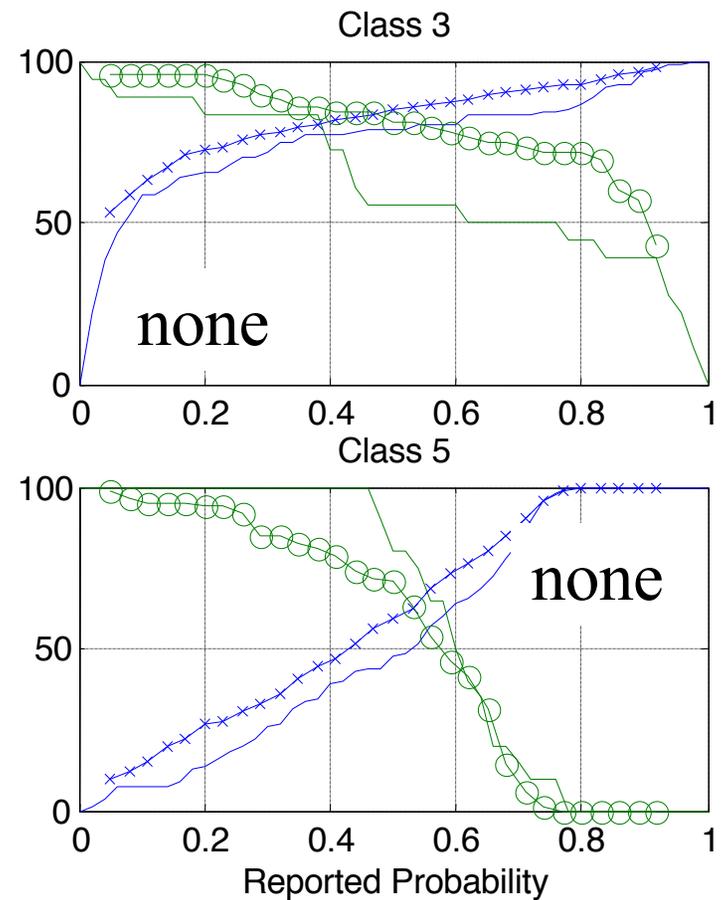
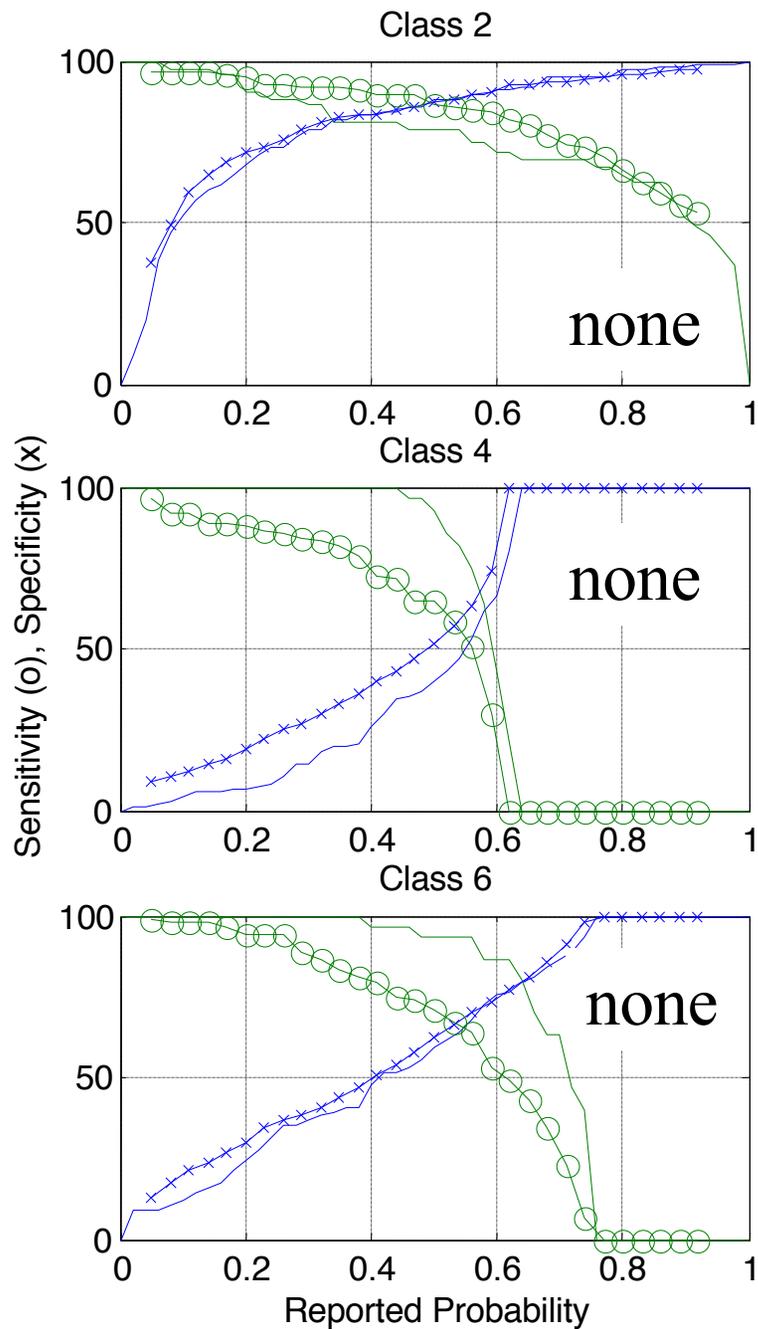
cv = cross-validation results

val = validation results

	cv	val
2:	87/87	79/87
3:	86/86	78/84
4:	55/55	58/52
5:	64/64	68/55
6:	69/69	82/68

(sens/spec)

- Sensitivity (crossval)
- Sensitivity (validation)
- Specificity (crossval)
- Specificity (validation)

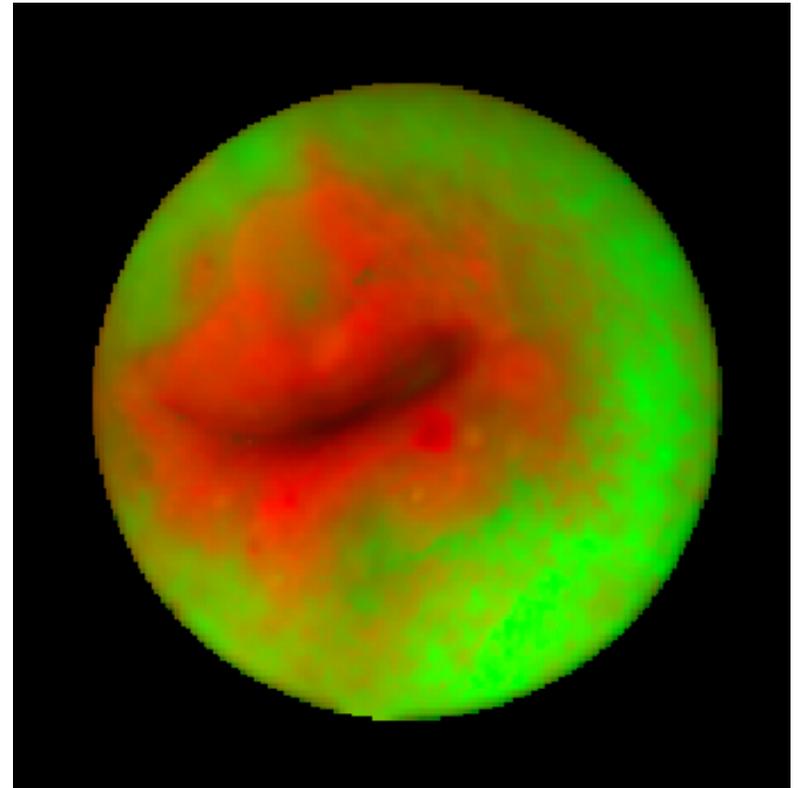
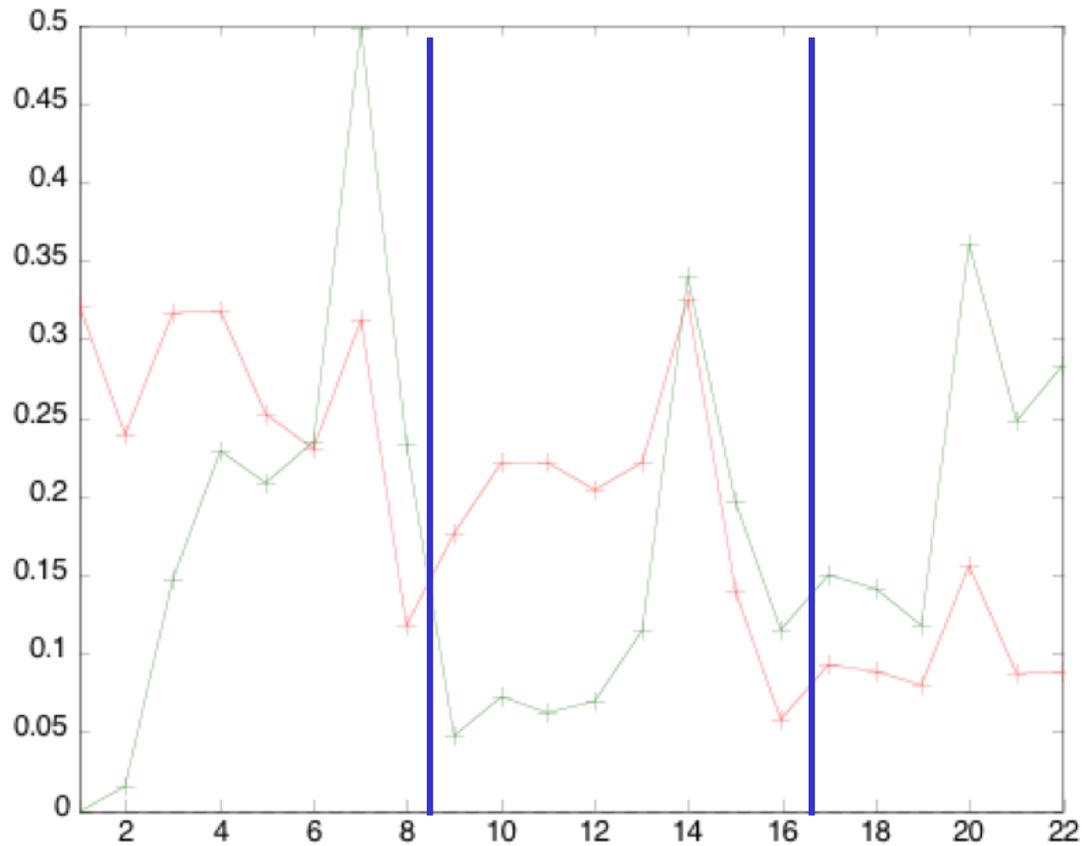


	cv	val
2:	87/87	79/87
3:	83/83	56/79
4:	57/57	83/46
5:	63/63	77/50
6:	66/66	93/68

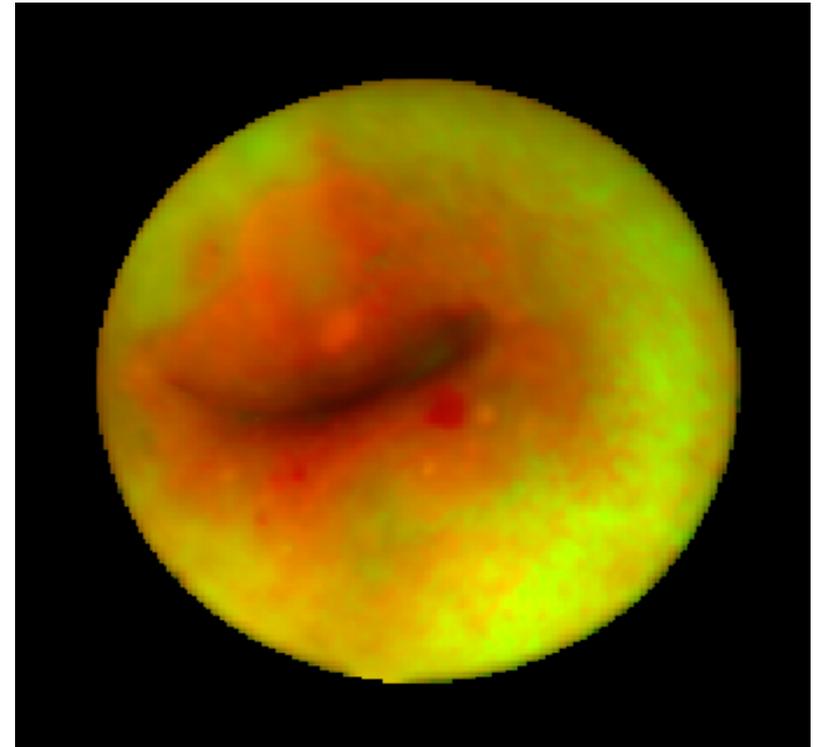
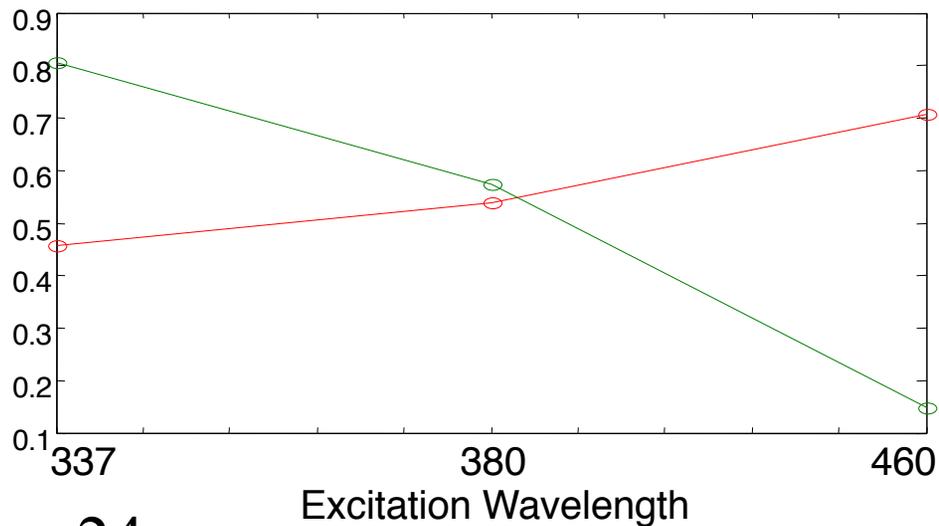
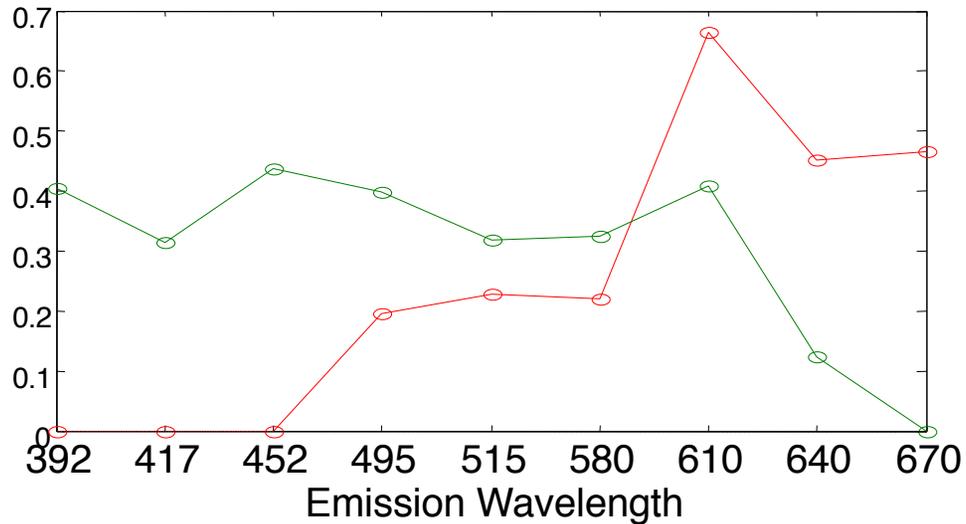
(sens/spec)

- Sensitivity (crossval)
- Sensitivity (validation)
- ×— Specificity (crossval)
- Specificity (validation)

Multivariate Curve Resolution (On pseudo first-order data)



Parallel Factor Analysis (On second-order data)



Summary

- ETF based device very close to clinical usefulness
- Mis-classifications tend to be on progression of disease
- Pre-processing critical
- PLS-DA effective
- Issues
 - Only translational motion considered in alignment
 - Other preprocessing and DA methods to consider