

# A Machine-Learning Approach to Differentiating Benign and Malignant Peripheral Nerve Sheath Tumors

A MULTICENTER STUDY

Michael Zhang, MD

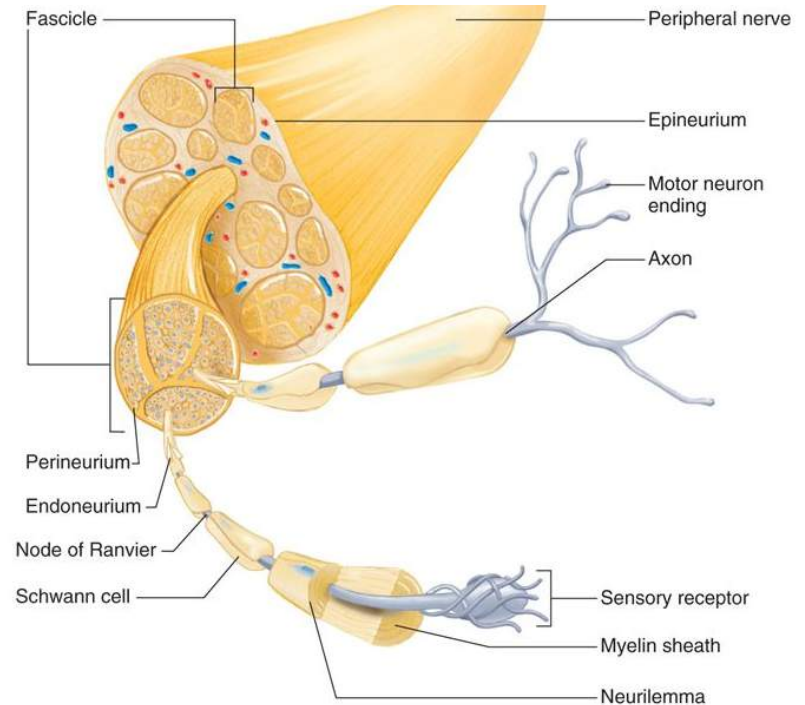
Mentors: K. Yeom, S. Napel

10.21.20

# Outline

- **Background:**
  - › **Benign and Malignant PNST**
  - › **Current Clinical Workflow**
- **Methods:**
- **Results**
- **Future Directions**

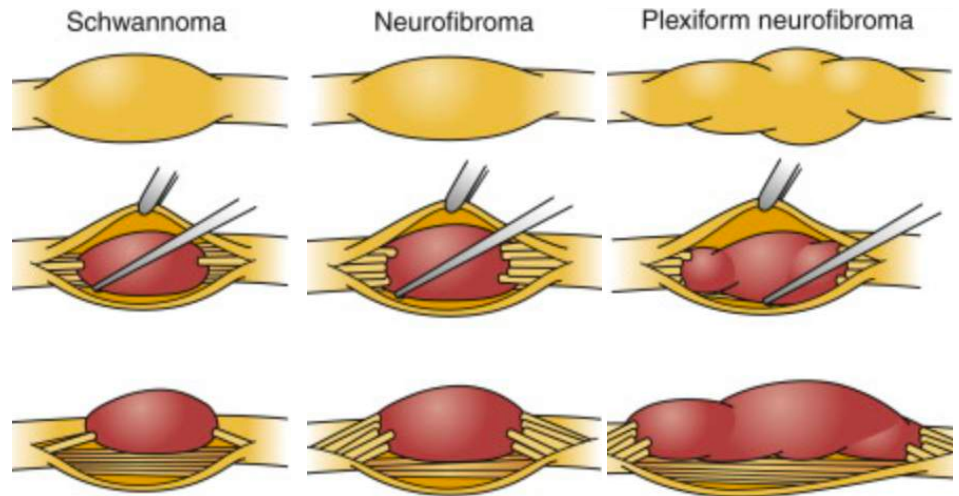
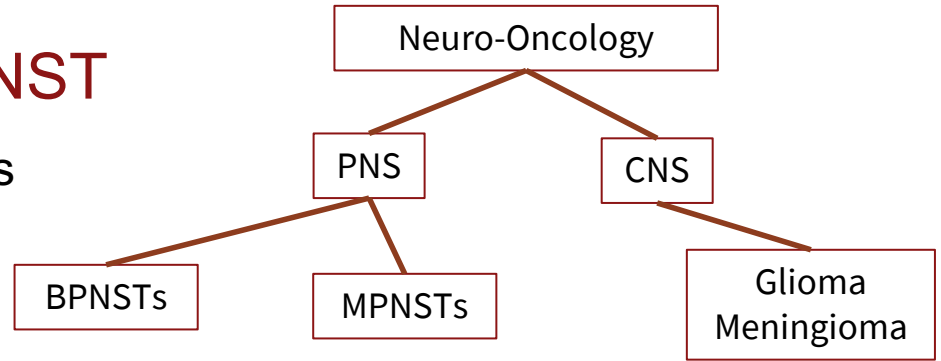
**Goal:** Develop a binary classifier using MRI information, providing early diagnosis of Benign and Malignant peripheral nerve sheath tumors (**PNSTs**)



# Benign and Malignant PNST

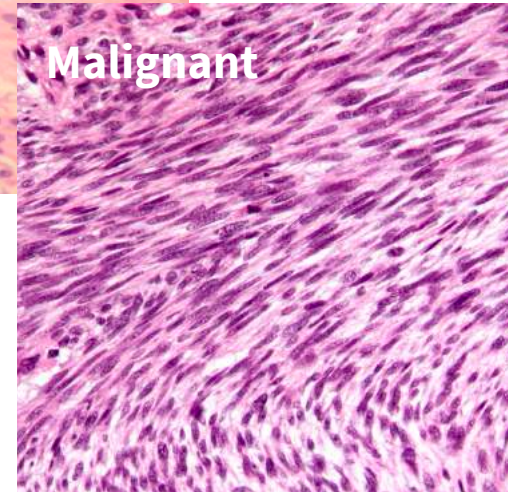
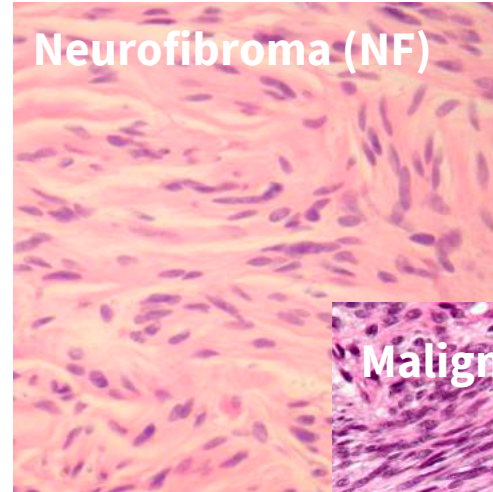
Peripheral Nerve Sheath Tumors  
(outside the brain and spine)

- Benign (BPNSTs)
  - › Schwannoma
  - › Neurofibroma
  - › Perineuroma
  - › Hybrids
  - › Ganglioneuroma
  
- Malignant (MPNSTs)



# MPNSTs – Can't Miss Diagnosis

- Natural History
  - › **5-year survival 30-50%**
  - › 50% of all MPNSTs occur in Neurofibromatosis (NF1)
  - › 5-10% of NF1 patients will develop MPNSTs
- Pathophysiology
  - › **Malignant Transformation**
  - › Neurofibroma → Malignancy
  - › Invasion and metastasis → morbidity & **surgical difficulty**

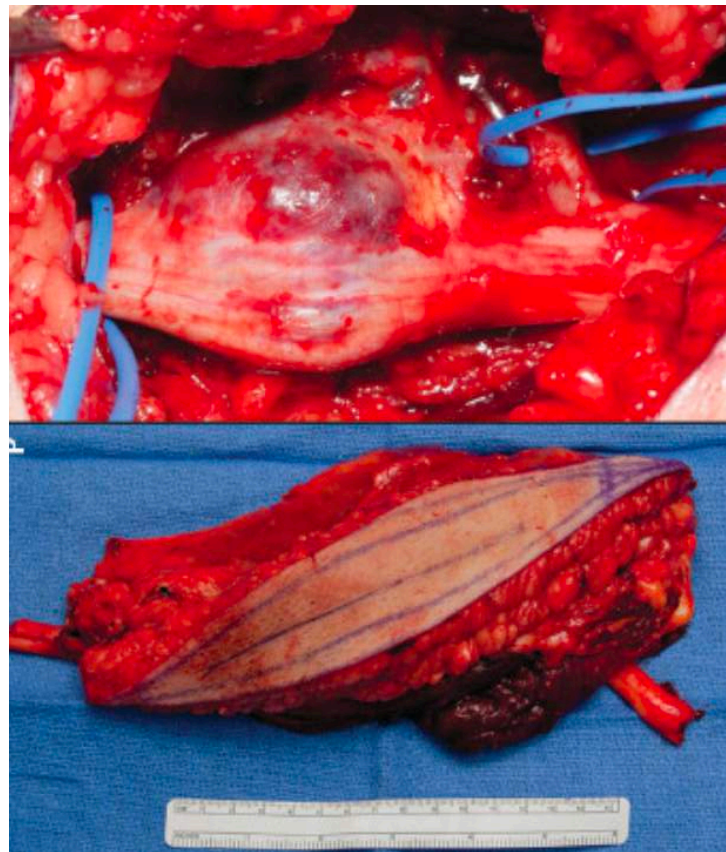


## Surgical Morbidity – Complex

**MPNST, Surgical Goal:** complete resection – **total resection is curative** – with minimal disruption of nerve function

### Surgical Challenges

- Greater nerve fiber invasion
- Neighboring tissue invasion
- Needs wide excisional margins
- Must avoid seeding
- Expect repeat surgery



## Surgical Morbidity – Simple

**BPNSTs** are surgically simpler

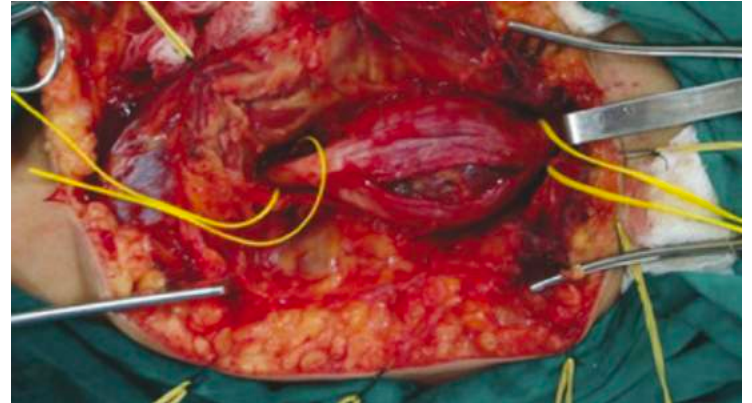
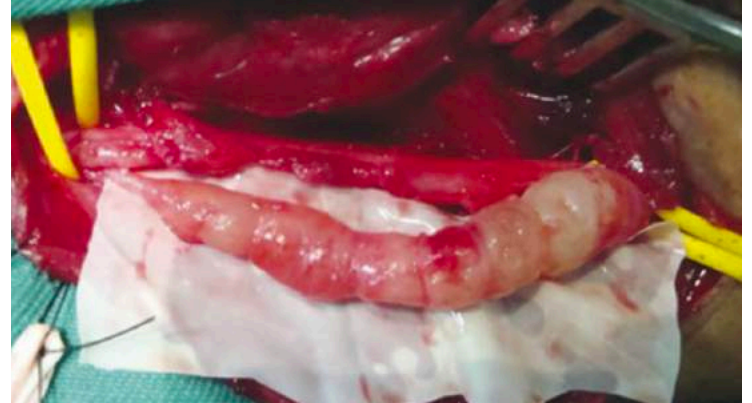
- Local involvement
- Low recurrence potential

**Schwannoma:**

- Tumor of supporting Schwann cells
- Single nerve fiber
- Tumor displaces uninvolved fascicles

**Neurofibroma:**

- Tumor of the endoneurium
- Single or multiple nerves
- Possible functional implications



# A Need for Early Diagnosis by Imaging

Accurate imaging can aid surgical planning to maximize quality of life

- Resect MPNSTs earlier
- Avoid unnecessary surgery

## Imaging Options

- **MRI:** qualitative and semantics
- **PET:** SUV > 3.5
- **Gold Standard: Surgical Biopsy**

## Wasa et al. - MRI Criteria

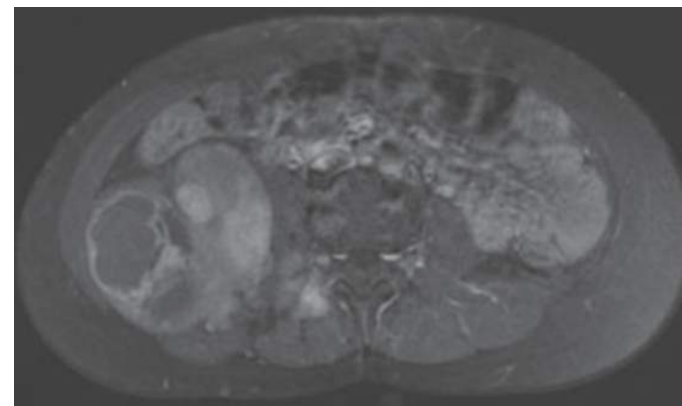
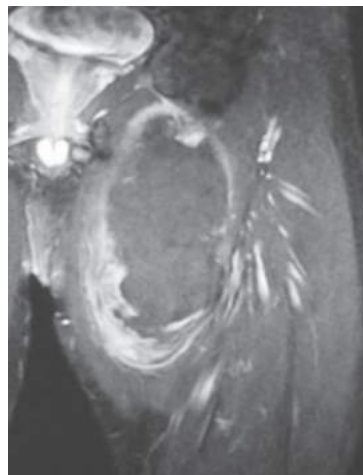
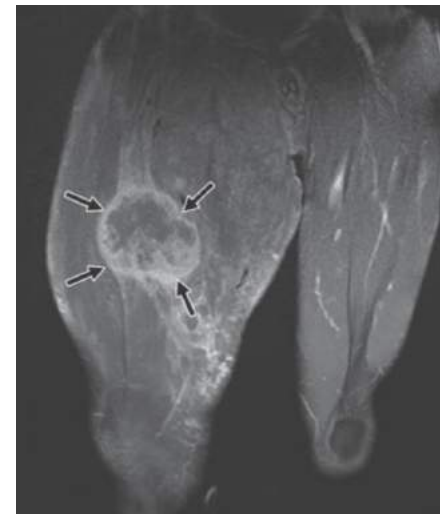
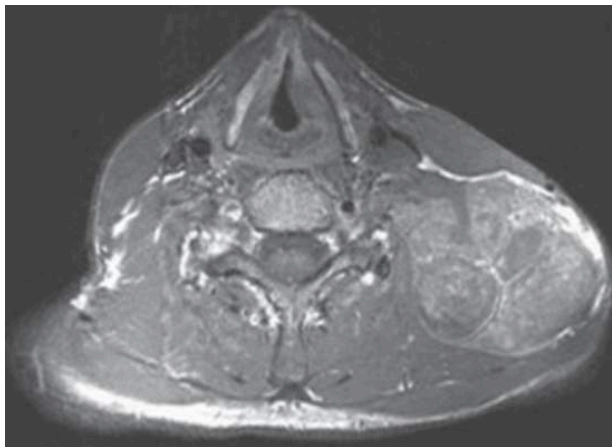
With 2 of 4 – 61% sens, 90% spec

1. Diameter > 5 cm
2. Peripheral tumor enhancement
3. Perilesional edema
4. Intra-tumoral cyst

## Radiographic Ambiguity

**MRI T1 with Gad Fat Sat**  
**commonly available**

- Neck – NF1
- Right Thigh – MPNST
- Left Thigh – MPNST
- Right RP – MPNST





## Additional Tools: PET, Derlin et al.

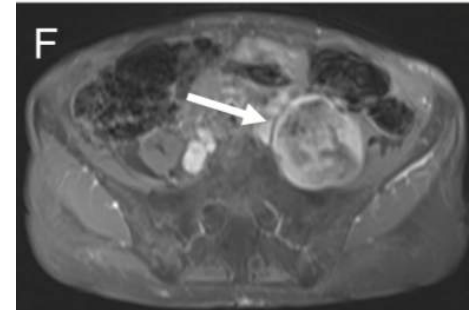
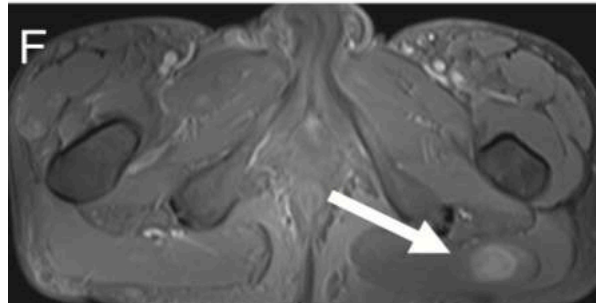
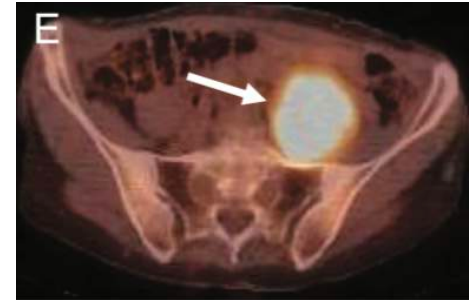
### PET – less available

#### SUVmax $\geq 3.5$

- 100% Sensitive
- 54.5% Specific
- 47.4% PPV

### MRI comparison

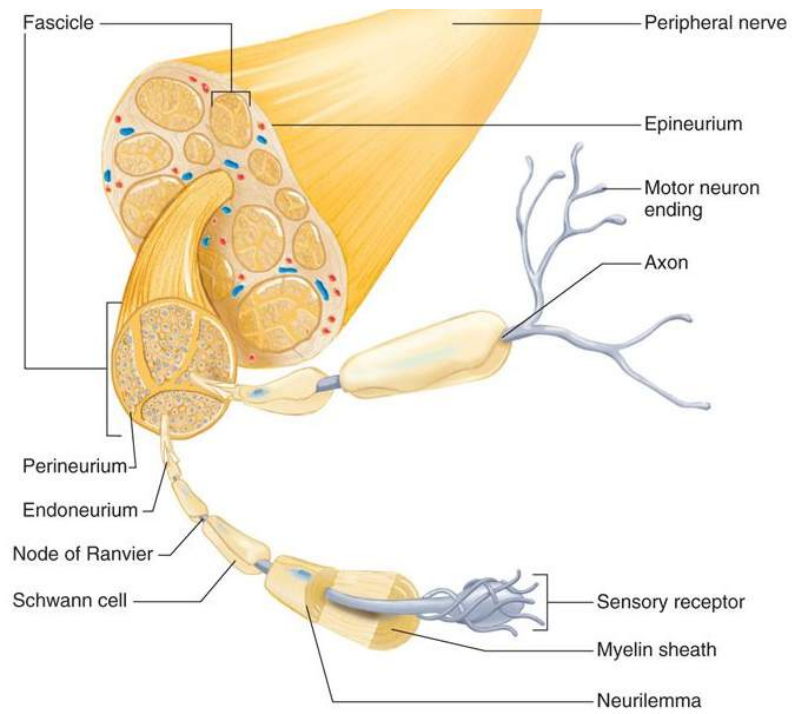
- 66.7% sensitive
- 90% specific
- 75% PPV



# Outline

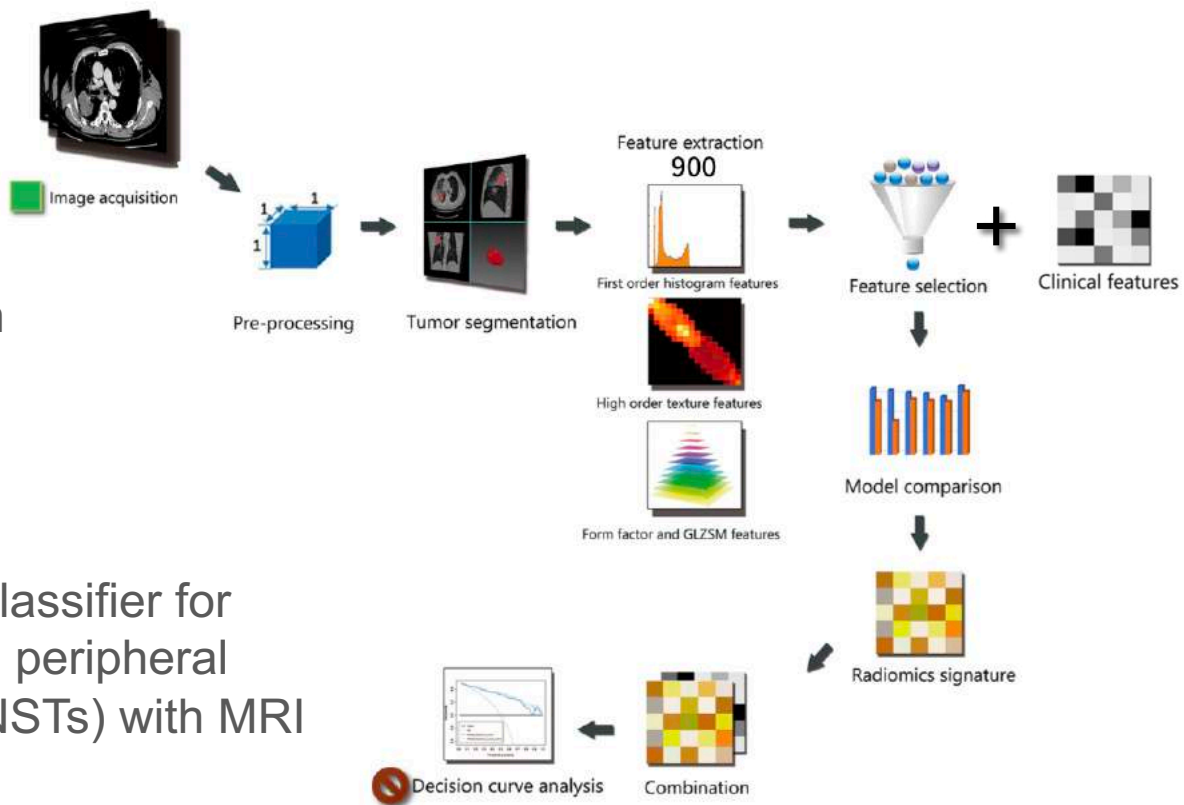
**Goal:** Develop a binary classifier (MPNST vs BPNST) with MRI

- Background:
- **Methods:**
  - › Segmentation
  - › Clinical data collection
  - › Feature
    - Extraction
    - Selection
    - Optimization
  - › Prediction Analysis
- Results
- Future Directions



# Workflow

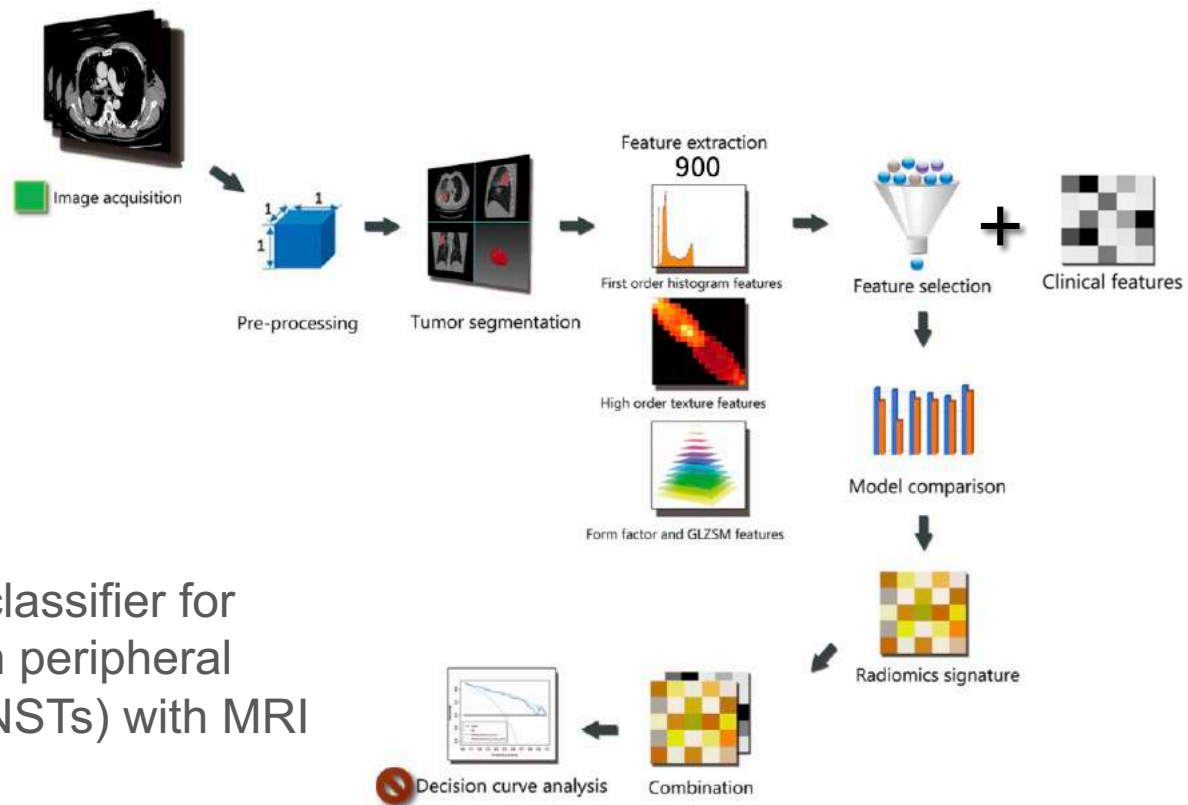
- **Input:** MRI T1-gad with Fat Suppression
  - › Most available
- **Output:** Classification probability and label
  - › MPNST or BPNST



**Goal:** develop a binary classifier for Malignant versus Benign peripheral nerve sheath tumors (PNSTs) with MRI

# Radiomics Workflow

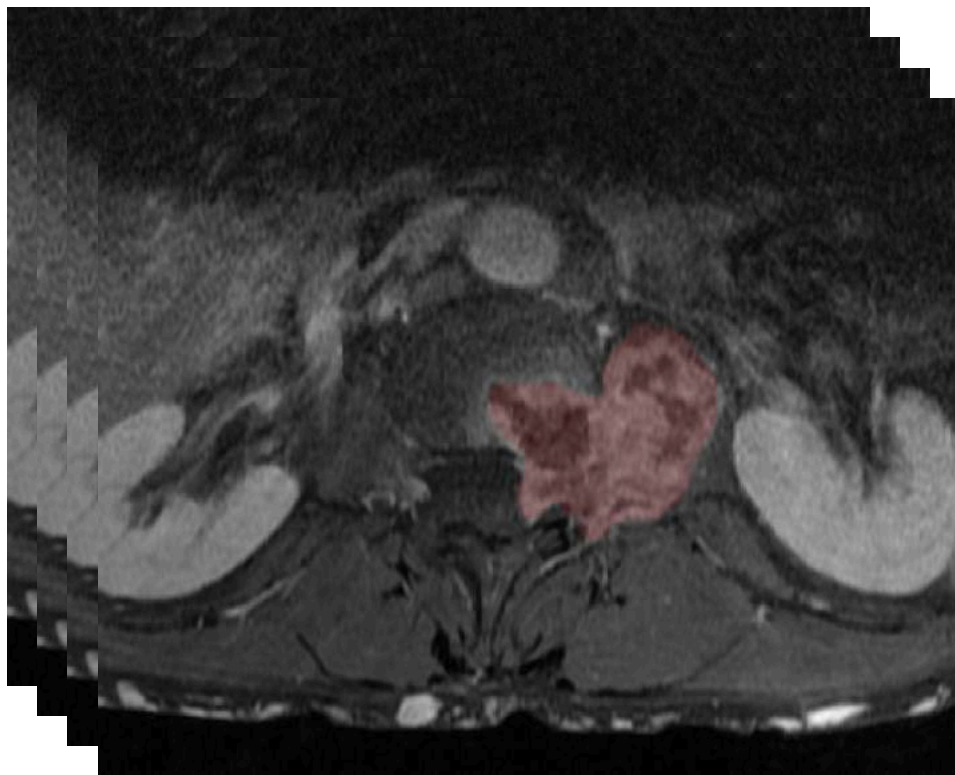
- Image segmentation
- Clinical data collection
- Feature
  - › Extraction
  - › Selection
  - › Optimization
- Prediction Analysis



**Goal:** develop a binary classifier for Malignant versus Benign peripheral nerve sheath tumors (PNSTs) with MRI

# Images and Segmentation

- **Imaging Segmentation**
  - › 171 Benign
  - › 95 Malignant
  - › SHC, Uni. Utah, Mayo
- Clinical data collection
- Feature
  - › Extraction
  - › Selection
  - › Optimization
- Prediction Analysis



# Clinical Features

- Chart review
- Red Flag symptoms of MPNSTs/infiltrative tumors
  - › Pain
  - › Rapid growth
  - › Neurological deficits

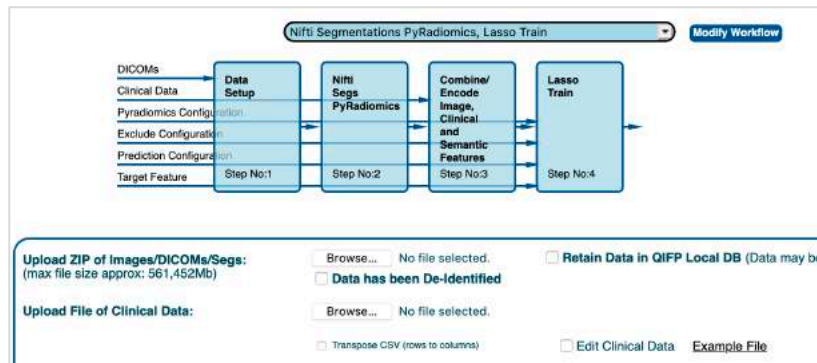
	<b>Benign (N = 171)</b>	<b>Malignant (N = 95)</b>	<b>p-Value</b>
<b>Age, yrs. (SD)</b>	45.5 ( 15.3)	43.3 (18.2)	0.320
<b>Sex</b>			0.042
<b>Male</b>	75 (44%)	54 (57%)	
<b>Female</b>	96 (56%)	41 (43%)	
<b>Spontaneous Pain</b>	41 (24%)	71 (75%)	<0.001
<b>Motor Deficit</b>	45 (26%)	31 (33%)	0.275
<b>NF1</b>	38 (22%)	41 (43%)	<0.001
<b>NF2</b>	10 (6%)	0 (0%)	0.016
<b>Schwannomatosis</b>	5 (3%)	0 (0%)	0.164

# Feature Extraction



- Quantitative Imaging Feature Pipeline (Stanford, Napel)
- Upload NifTI segmentations
- Pyradiomics Package: 900 standardized features

## QIFP: Home Page



# Feature Selection – Preliminary Model

## 1. LASSO

- 70:30 Train Test Split
- R, glmnet-package
- 10x Cross Validation
- 1000 cycles

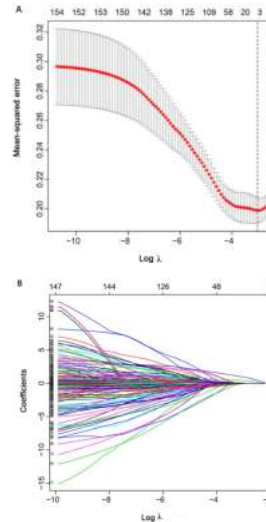
## 2. Preliminary model

- 80% selection
- 21 Features (from 900)**
  - 19 Textural
  - 2 Clinical

## 1. 900 Initial Features

- First Order Statistics (19 features)
- Shape-based (3D) (16 features)
- Shape-based (2D) (10 features)
- Gray Level Cooccurrence Matrix (24 features)
- Gray Level Run Length Matrix (16 features)
- Gray Level Size Zone Matrix (16 features)
- Neighbouring Gray Tone Difference Matrix (5 features)
- Gray Level Dependence Matrix (14 features)

## 2. Error Reduction



## 3. 21 Selected Features

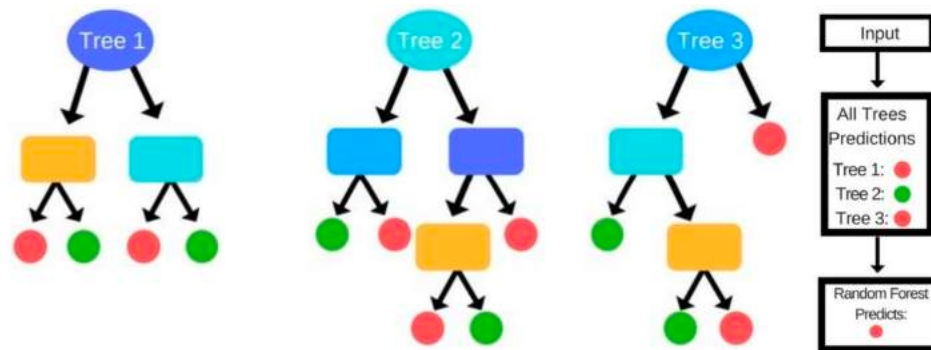
1	Intercept	2.21727218721519
2	log-sigma-5-mm-3D_glrIm_RunLengthNonUniformity	0.387899557771315
3	wavelet-HHL_glcM_MaximumProbability	0.175997193985862
4	wavelet-LHL_glrIm_RunVariance	0.175501773912802
5	original_shape_Maximum2DDiameterRow	0.170282978191518
6	original_glrIm_LongRunEmphasis	0.168658168332157
7	original_shape_Maximum2DDiameterColumn	0.156694553330967
8	log-sigma-5-mm-3D_glszm_GrayLevelNonUniformity	0.131656590446044
9	Energy	0.123258921624558
10	log-sigma-5-mm-3D_firIorder_Skewness	0.107173356877628
11	log-sigma-5-mm-3D_firIorder_Maximum	0.089609832091139
12	wavelet-HLL_glszm_LargeAreaHighGrayLevelEmphasis	0.035471622286746
13	log-sigma-3-mm-3D_firIorder_Energy	0.00410218649987962
14	wavelet-LHL_firIorder_Energy	0.00351860219148743
15	log-sigma-3-mm-3D_firIorder_TotalEnergy	0.0035108393943978
16	original_firIorder_TotalEnergy	6.32893969118343e-15
17	wavelet-LHL_firIorder_TotalEnergy	3.09099168779373e-16



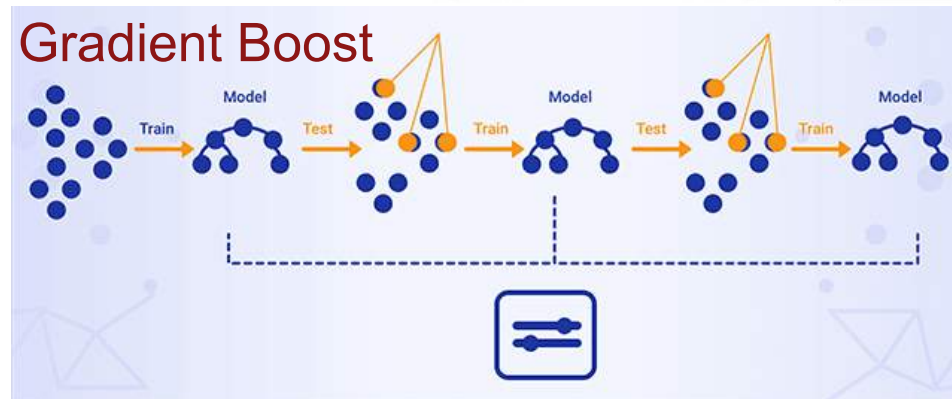
# Feature Selection – Optimized Model

- Imaging Segmentation
- Clinical data collection
- Feature
  - › Extraction
  - › Selection
  - › **Optimization**
    - caret package
    - **21 features, ranked**
- Prediction Analysis

## Random Forest



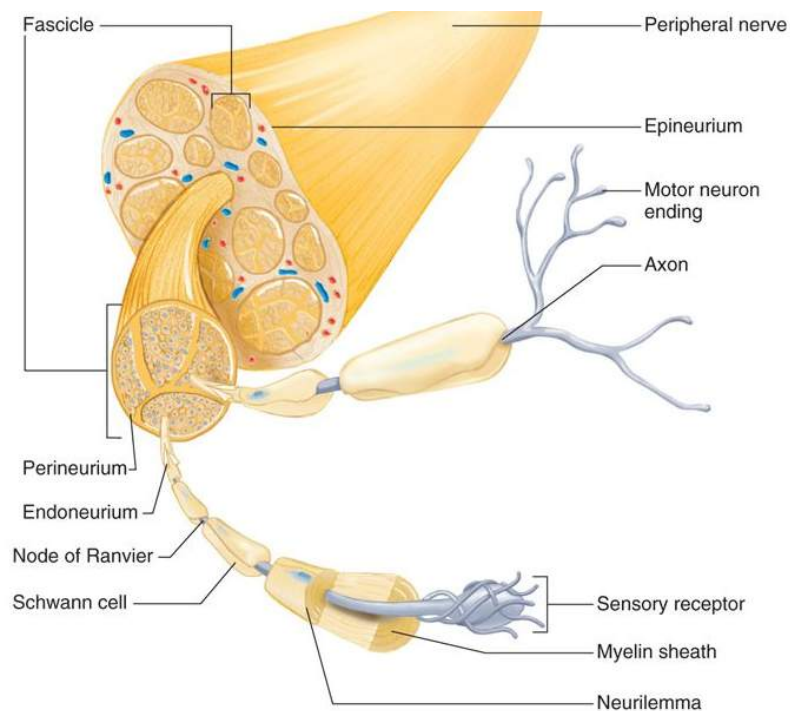
## Gradient Boost



# Outline

**Goal:** Develop a binary classifier (Malignant/Benign) with MRI

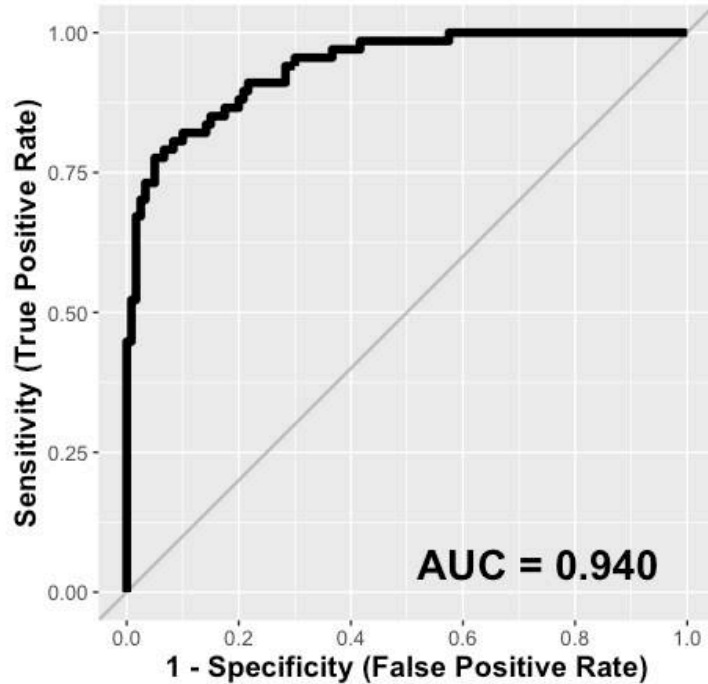
- Background:
- Methods:
- **Results**
  - › 1. Texture + Clinical
  - › 2. Texture Only
  - › 3. Clinical Only
  - › 4. Comparison to Human
- **Future Directions**



# 1. Clinical + Texture: AUCs

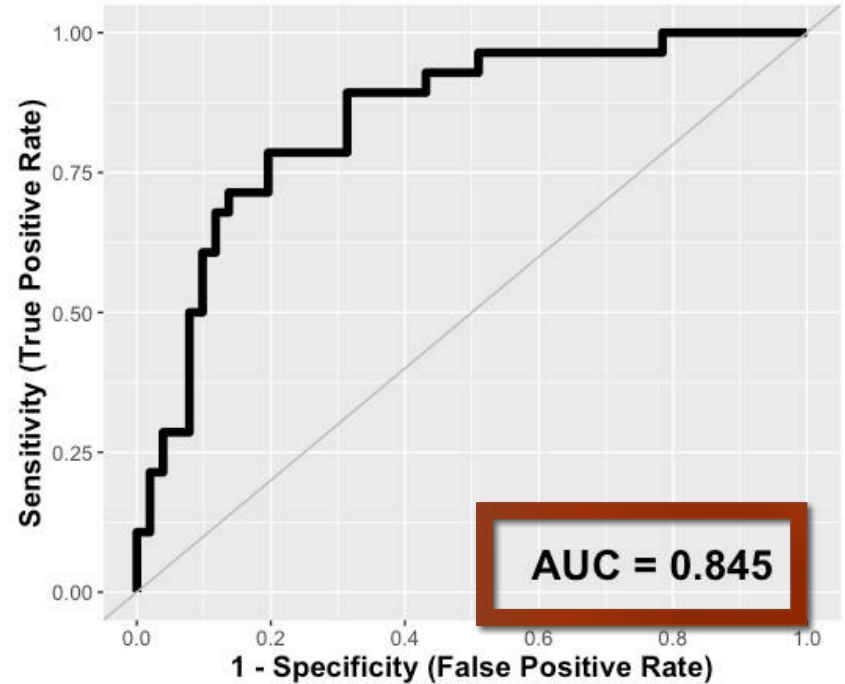
A

Receiver Operating Curve for Training Set



B

Receiver Operating Curve for Test Set



## 2. Comparison to Human Readers

- Two of each training level
- Two trials
  - › 1. Imaging + Clinical
  - › 2. Imaging only
- Imaging:
  - › T1W, post-gad, fat-sat +
  - › T2W or proton-density images, when available

	Sensitivity	Specificity	PPV	NPV
Medical Student	0.518	0.775	0.538	0.752
PN Surgery Fellow	0.625	0.814	0.648	0.798
PN Surgeon Radiologist	0.821	0.667	0.575	0.872
Overall	0.684	0.742	0.589	0.823
			AUC	0.704

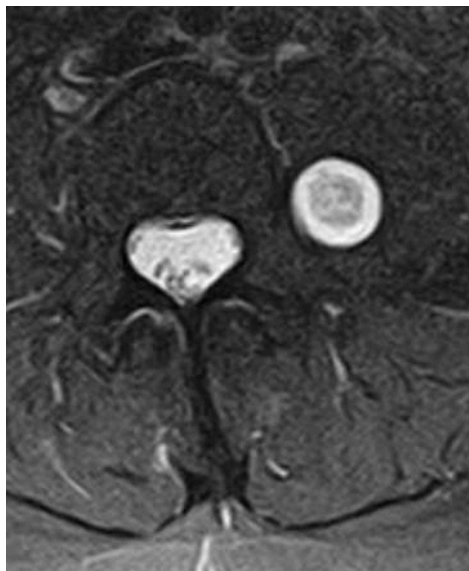
	Sensitivity	Specificity	PPV	NPV
Medical Student	0.661	0.725	0.569	0.796
PN Surgery Fellow	0.554	0.775	0.577	0.759
PN Surgeon Radiologist	0.821	0.686	0.590	0.875
Overall	0.704	0.723	0.582	0.826
			AUC	0.702

## Conclusions and Limitations

- Radiomics Classifier  
Outperforms Human readers
- Radiomics
  - › Sensitivity 0.676
  - › Specificity 0.882
- Humans
  - › Sensitivity 0.839
  - › Specificity 0.686
- Imaging Availability
  - › Axial view only
  - › T1W only
  - › Acquisition technique: fat saturation, contrast quality
- Patient heterogeneity: brachial plexus, arm, wrist
- Segmentation vs Semantics

# Semantics not captured by segmentations

BPNST:  
Neurofibroma  
-Target sign



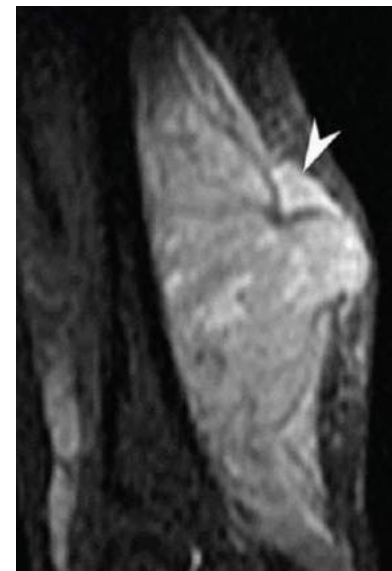
BPNST:  
Schwannoma  
-Split fat sign



MPNST  
-Absent split  
fat sign



MPNST  
-Perilesional  
edema



# Acknowledgements

## Radiology

- Dr. Kristen Yeom
- Dr. Sam Gambhir
- Dr. Heike Daldrup-Link

## PNST Team

- Dr. Elizabeth Tong
- Dr. Thomas Wilson
- Dr. Mark Mahan
- Lydia Tam
- Edward Lee

## QIFP Team

- Professor Sandy Napel
- Sarah Mattonen
- Dev Gude

## Stanford Neurosurgery

- Dr. Gary K. Steinberg
- Dr. Gerald Grant
- Dr. Gordon Li



**STANFORD**  
SCHOOL OF MEDICINE

