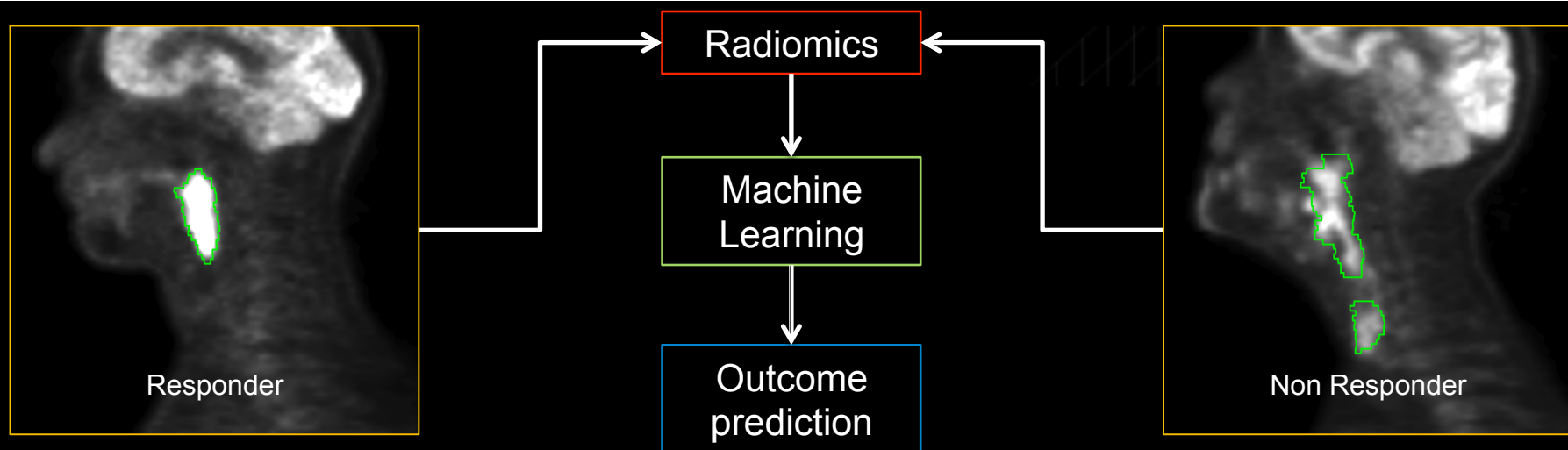


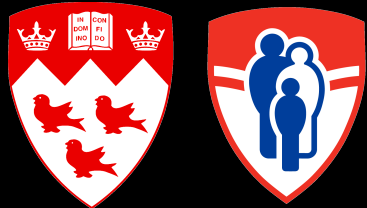


Martin Vallières*, Mathieu Hatt, Dimitris Visvikis, Issam El Naqa

*Post-doctoral fellow at LaTIM, INSERM UMR 1101, IBSAM, UBO, UBL, Brest, France

Enhancement of multimodality texture-based prediction models via optimization of PET and MR image acquisition protocols: a proof of concept

McGill University

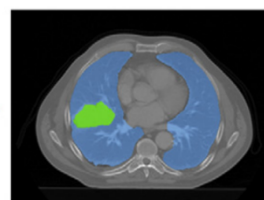


RADIOMICS: MAJOR OBJECTIVE

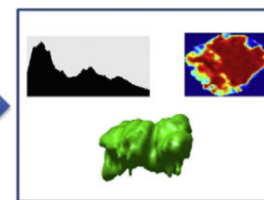
**RADIOMICS
WORLD**



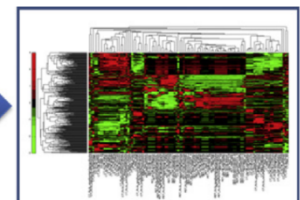
Imaging



Segmentation



Features extraction



Analysis

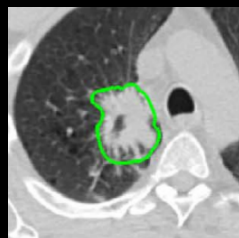
(Lambin et al., Eur J Cancer 48, 2012)



**! CLINICAL
WORLD !**



New patient



**PREDICTION
MODEL**
 $f(\text{Texture1}, \text{Texture2}, \text{etc.})$



**PERSONALIZED
TREATMENT
(Precision
medicine)**

MOTIVATION OF STUDY

FDG-PET/MR texture-based model for lung mets prediction in soft-tissue sarcoma (STS) cancer

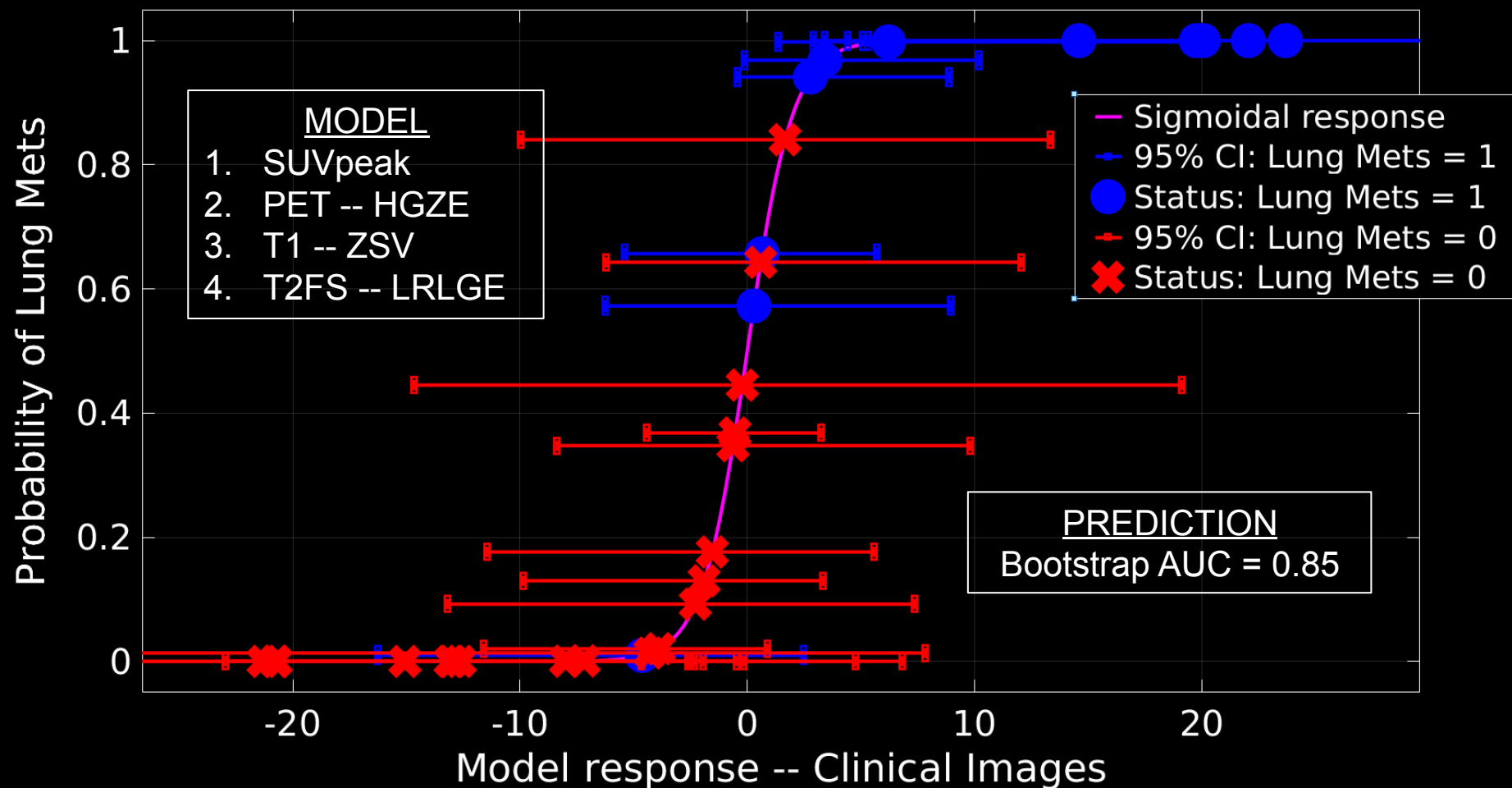


IMAGE ACQUISITION INFLUENCES TEXTURES

PET ACQUISITION PARAMETERS

- 2D versus 3D modes
- Span
- Time of acquisition

Galavis PE *et al* 2010 *Acta Oncol.* **49** 1012

MRI ACQUISITION PARAMETERS

- Number of acquisitions (NAs)
- Repetition time (TR)
- Echo time (TE)
- Sampling bandwidth (SBW)

Mayerhoefer ME *et al* 2009 *Med. Phys.* **36** 1236

GOAL OF STUDY

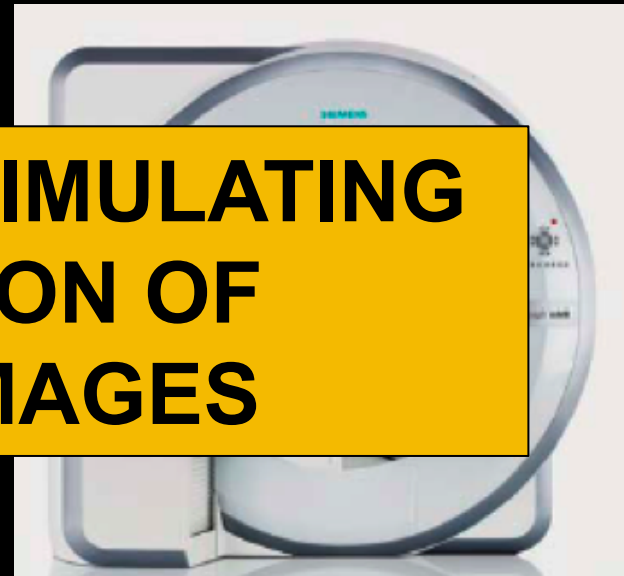
!

ENHANCE TEXTURE-BASED PREDICTION
MODELS THROUGH THE OPTIMIZATION OF
PET AND MR IMAGING ACQUISITION PHYSICS

!

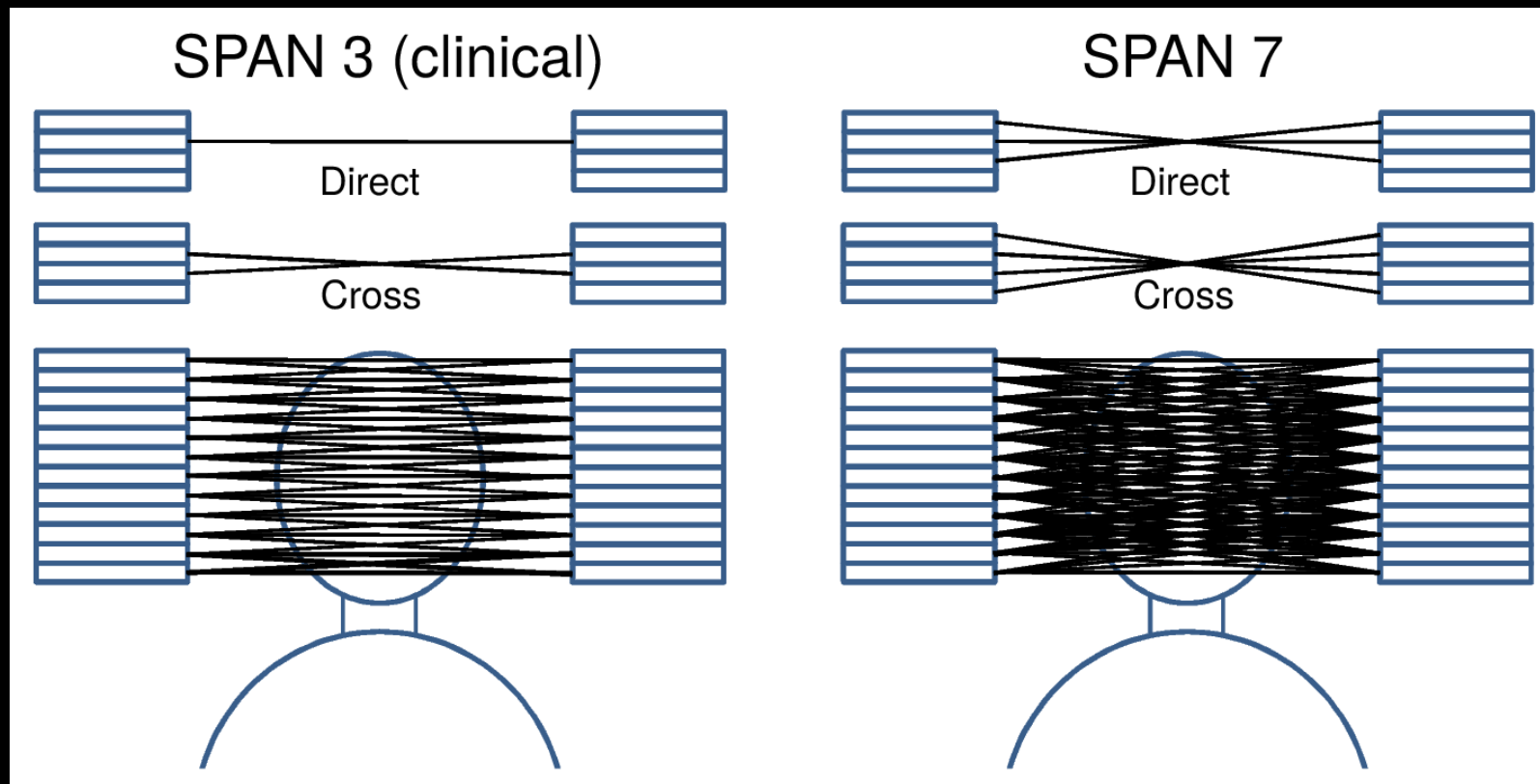


**BY NUMERICALLY SIMULATING
THE ACQUISITION OF
PET AND MR IMAGES**



PET physical parameter: SPAN

Amount of cross-talk allowed between slices



MRI physical parameters: TR, TE



TE_c : Echo time (clinical)
TR_c : Repetition time (clinical)

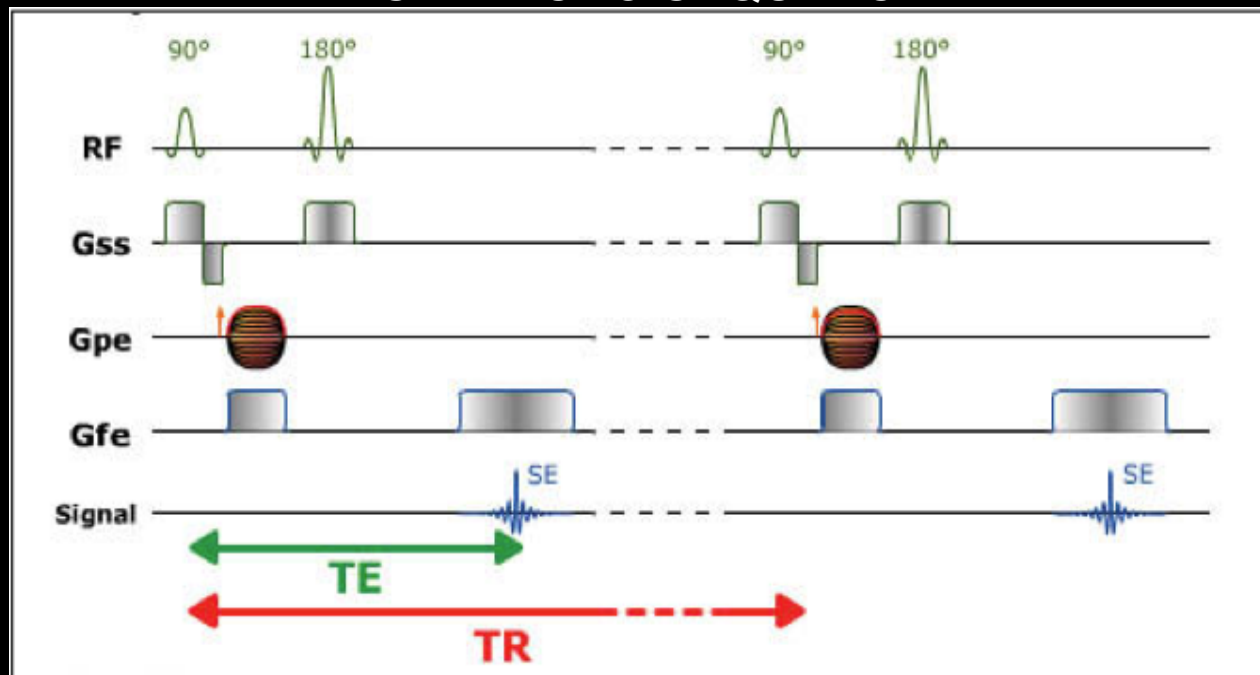
T1-weighted image

- TE_c << T2
- TR_c ~ T1

T2-weighted image

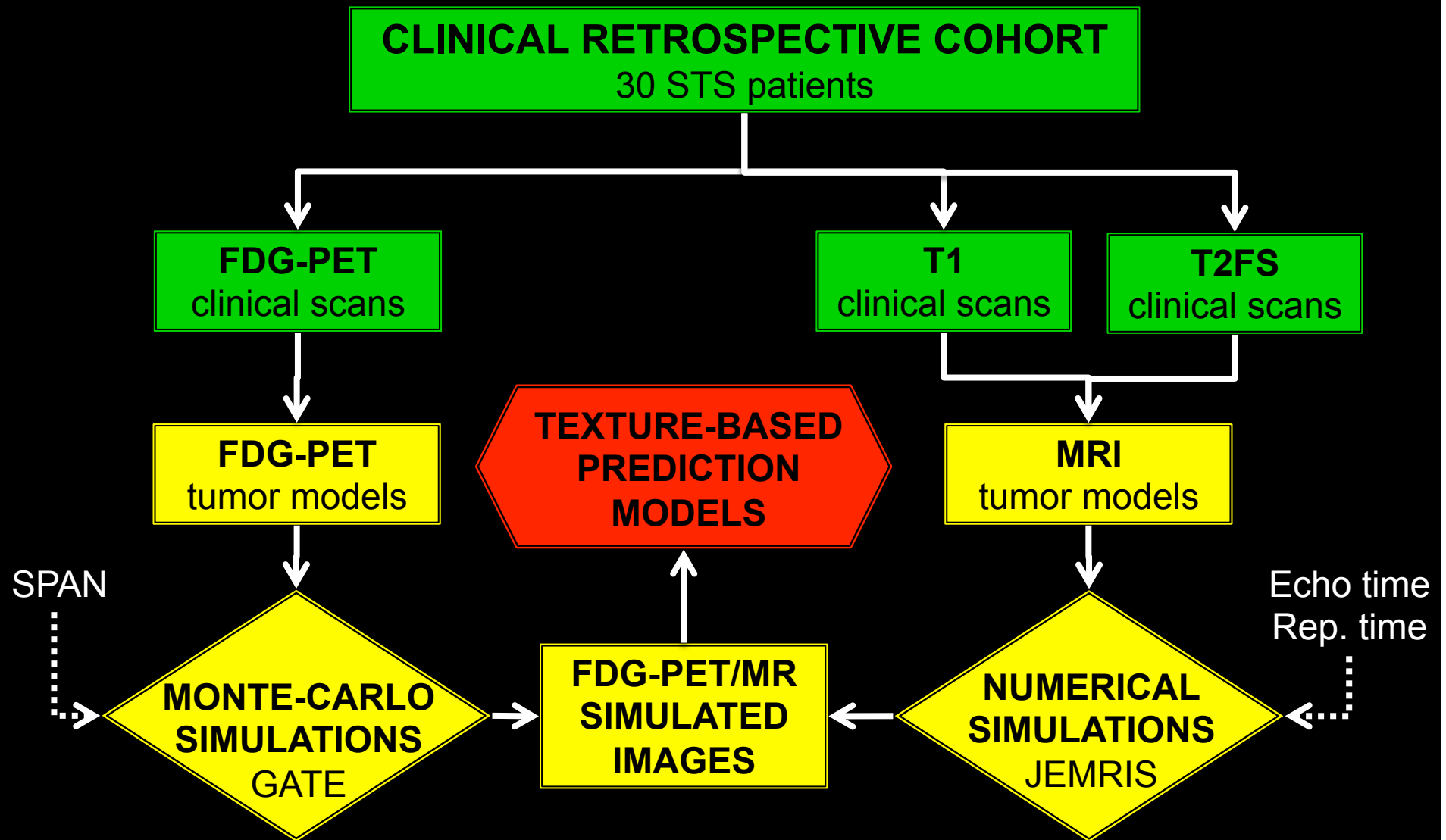
- TE_c ~ T2
- TR_c >> T1

SPIN-ECHO SEQUENCE



(<http://www.imaios.com/en/e-Courses/e-MRI/MRI-Sequences/Spin-echo>)

METHODOLOGY



EXPERIMENTS

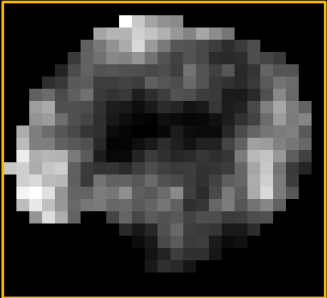
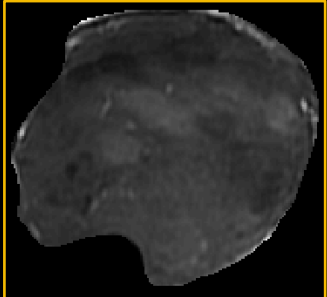
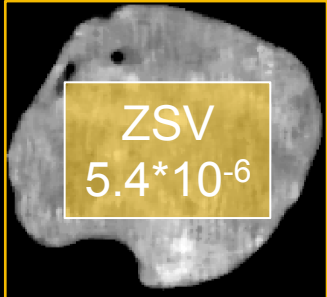
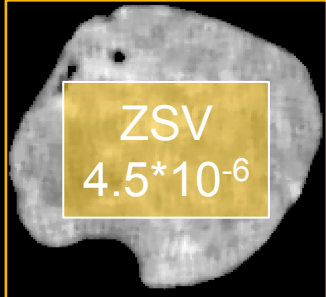
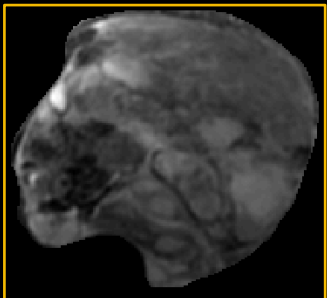
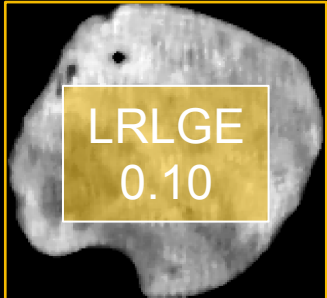
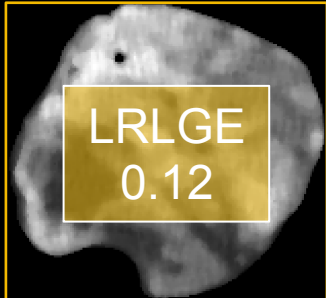
IMAGING SIMULATIONS

- PET
 - SPAN: 3,5,7,9,11,13,15,17
 - Clinical SPAN: 3
- T1
 - TR: $(1/3, 1/2, 1, 2, 3) * TR_c$
 - TR_c of (492 ± 81) ms
- T2FS
 - TE: $(1/3, 1/2, 1, 2, 3) * TE_c$
 - TE_c of (77 ± 12) ms

PREDICTION MODELS EVALUATION

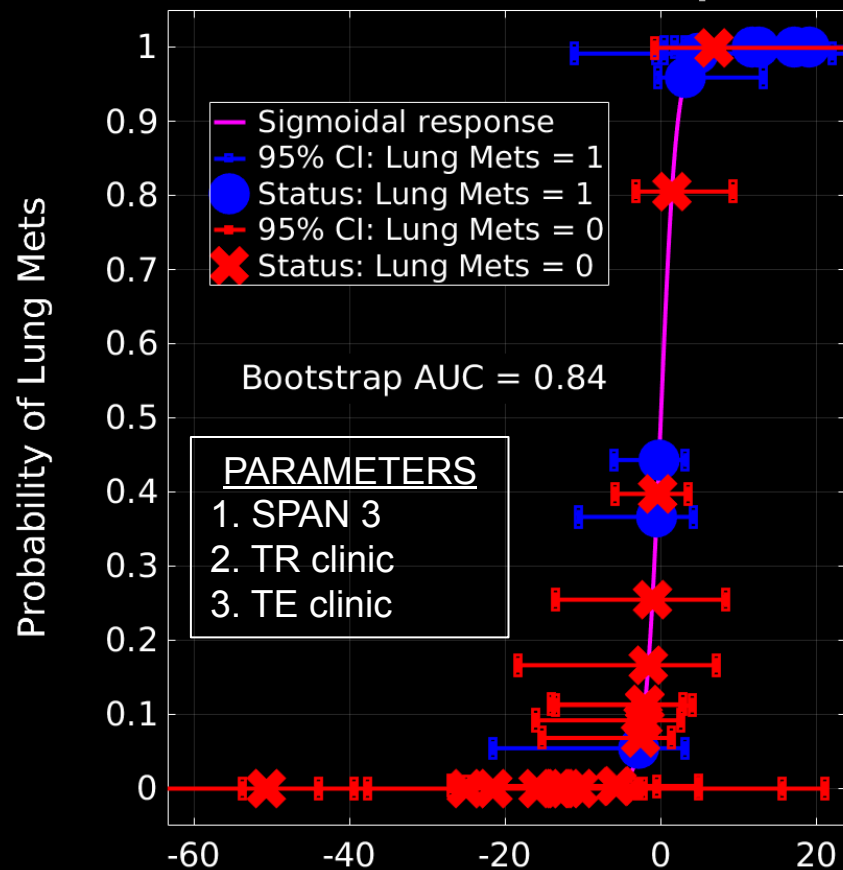
- Variables
 - SUV_{peak}
 - PET -- HGZE
 - T1 -- ZSV
 - T2FS -- LRLGE
- Model construction
 - Logistic regression
- Evaluation
 - 1000 bootstrap samples

SIMULATION OF IMAGE ACQUISITION: EXAMPLE

	CLINICAL → original	SIMULATED → clinical param.	SIMULATED → SPAN 17, $3 \times \{TEc, TRc\}$
PET		 HGZE 1180	 HGZE 1189
T1		 ZSV $5.4 \cdot 10^{-6}$	 ZSV $4.5 \cdot 10^{-6}$
T2FS		 LRLGE 0.10	 LRLGE 0.12

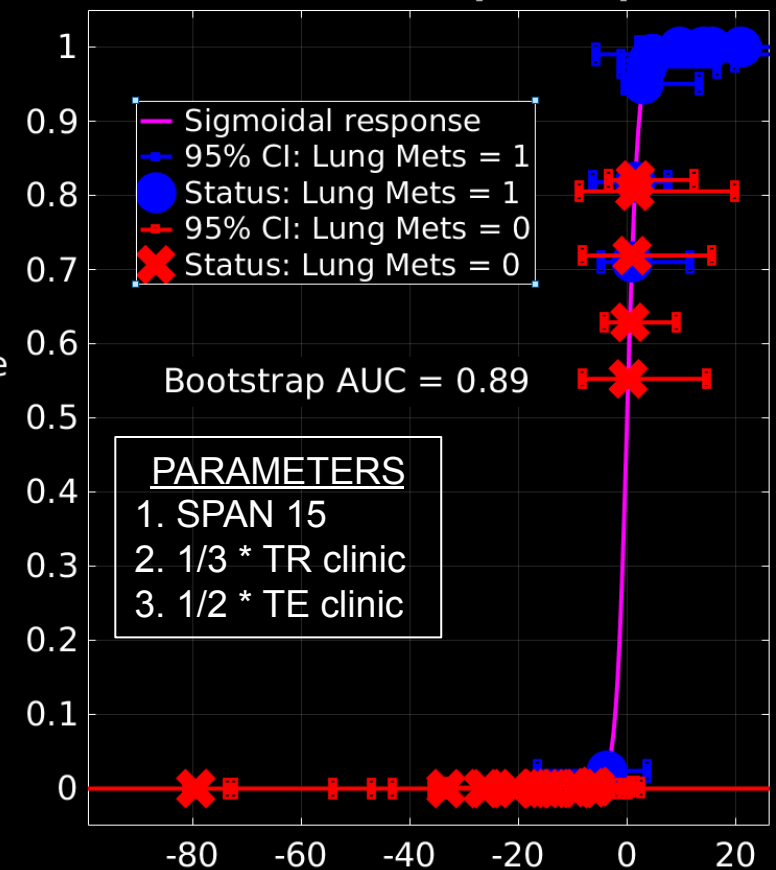
TEXTURE-BASED MODEL OPTIMIZATION

SIM IMAGES -- Clinical param.



★
AUC increase
→
 $p = 0.04$

SIM IMAGES -- Optimal param.

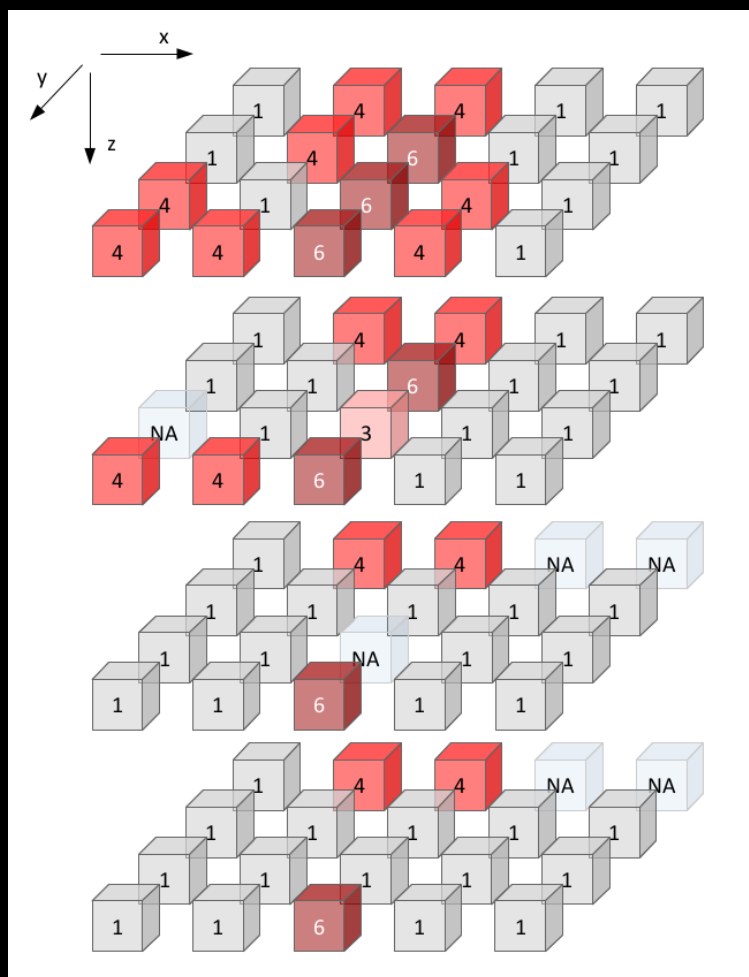


CONCLUSIONS

- Texture-based imaging models could be used to predict cancer outcomes (e.g. lung mets in STS cancer).
- Texture-based prediction models could be enhanced by optimizing FDG-PET and MR imaging acquisitions.
- Prediction performance of our texture-based model improves with:
 - Higher SPAN for PET images
 - Shorter TR for T_1 -weighted images
 - Shorter TE for T_2 -weighted images
- Proof-of-concept study: findings need to be verified on larger cohorts.

STANDARDIZATION OF RADIOMICS

DIGITAL PHANTOM



- Group: 50 scientists from 18 institutions in 8 countries.
- Goal: Standardize and benchmark radiomic feature computation.
- Results: Exhaustive list of features (172) and benchmarks results available online.

Zwanenburg A, Leger S, Vallières M & Löck S. Image biomarker standardisation initiative. *arXiv preprint*, arXiv:1612.07003 (2016).