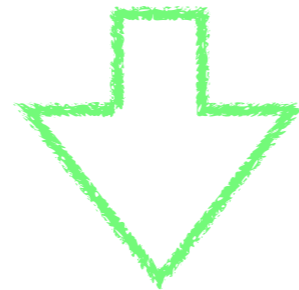


Analisi NMR

Introduction

- The therapy for locally advanced rectal cancers is:
 - neoadjuvant chemo-radiotherapy (CRT)
 - followed by radical surgery



- local pelvic recurrence to rate lower than 10%

However...

- This therapeutic approach is an over-treatment of many patients:
- those who do not respond to the treatment (non-responders)
 - whose early identification (2–3 weeks after the start of neoadjuvant CRT) might help clinicians in referring them to alternative treatments;
- patients with pathological complete response
 - who could benefit from either less invasive surgery (ie, transanal endoscopic microsurgery) or “wait-and-watch” strategy

The purpose

- Recognize non-responders and complete responder patients during the CRT (before surgery)

The dataset

- 75 patients with histologically confirmed colorectal adenocarcinoma and locally advanced tumor stages II (cT3-4, N0, M0) and III (cT1-4, N+, M0)
- Images acquired in three stages:
 - just after diagnosis (pre-CRT)
 - CRT response evaluation at early phase: CRT treatment was day 40 and tumor response was assessed with MRI at day 21
 - post-surgical analysis: 6 to 8 weeks after the end of CRT
- gross specimen was analyzed by pathologist

The dataset

- Three kind of NMR images acquired:
 - T2-weighted
 - diffusion-weighted imaging (DWI) or apparent diffusion coefficient (ADC)
 - and perfusion MRI (pMRI)
- Two acquisition per NMR kind:
 - axial oblique (planes orthogonal to the rectum)
 - coronal oblique (planes parallel to the rectum)
- The images acquired at the same time have a common reference system

The project

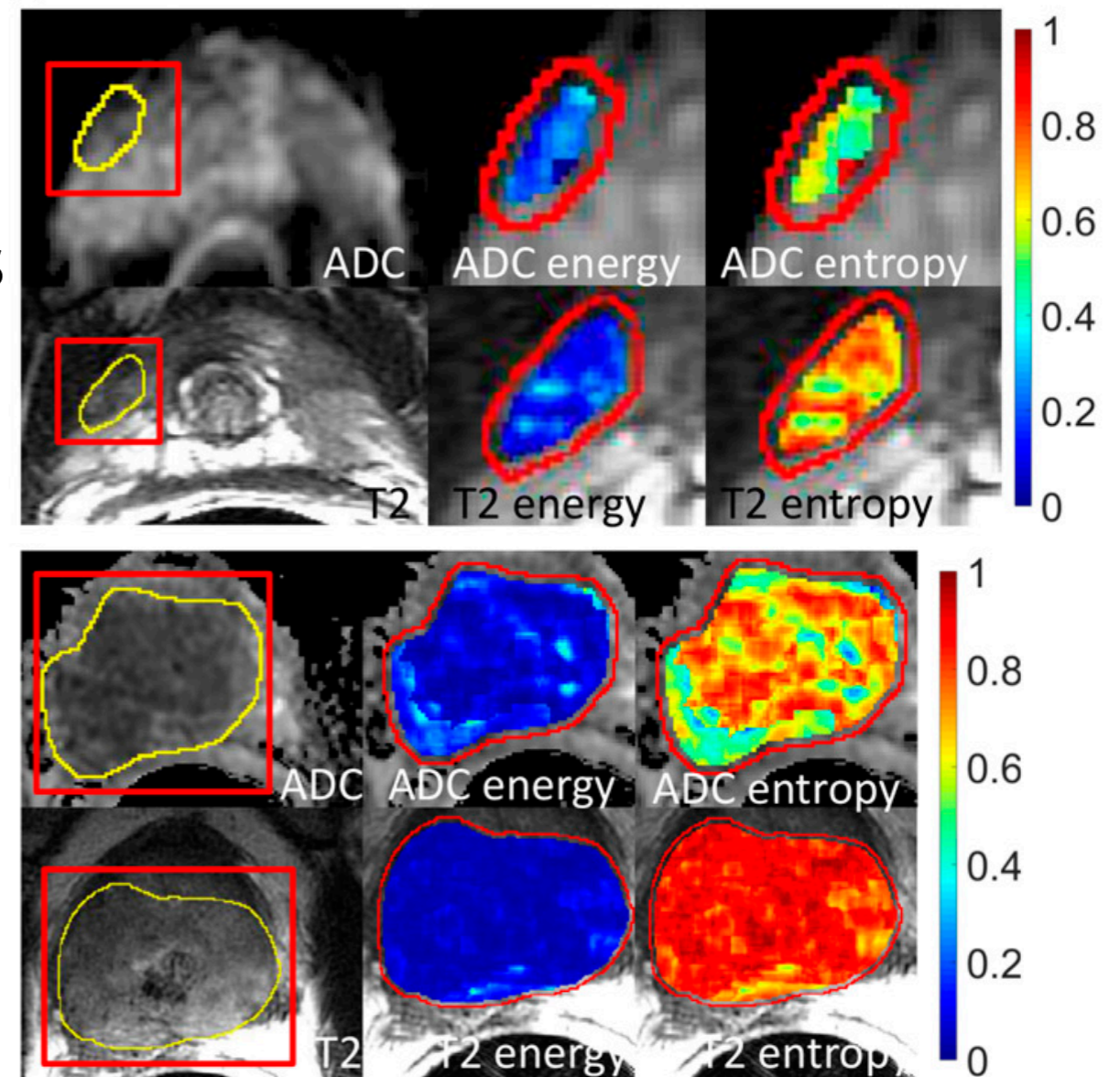
- Analyze the available images and find observables that allow to categorize the patients
- Increase the patient database
- Automatically define the Region Of Interest (ROI)

Possible tools

- Texture analysis
 - Structural methods
 - useful for the synthesis of an image than for its analysis
 - Transform methods
 - Fourier, Gabor or Wavelet transform
 - Statistical methods
- Neural networks
- Deep learning neural networks
 - H. Greenspan, B. van Ginneken, and R. M. Summers, "Guest Editorial Deep Learning in Medical Imaging: Overview and Future Promise of an Exciting New Technique," *IEEE Trans. Med. Imaging*, vol. 35, no. 5, pp. 1153–1159, Dec. 2016.

Application on prostate cancer

- machine learning-based automatic classification of prostate cancer aggressiveness
- combining apparent diffusion coefficient (ADC) and T2-weighted (T2-w) MRI-based texture features
- achieved an accuracy greater than 90%



D. Fehr et al. "Automatic classification of prostate cancer Gleason scores from multiparametric magnetic resonance images," Proc Natl Acad Sci USA, vol. 112, no. 46, pp. E6265–E6273, Nov. 2015.

People involved

- INFN Roma1

- Riccardo Faccini
- Stefano Giagu
- Carlo Mancini
- Riccardo Paramatti
- Roberta Santacesaria
- Alessia Satta
- Elena Solfaroli
- Cecilia Voena



- ISC CNR

- Antonio Scala
- Sapienza Pol. Umberto I.
 - Andrea Laghi
- Azienda Ospedaliera San Camillo-Forlanini
 - Riccardo Ferrari



- ISS ?

- Paolo Del Giudice

- INFN Genova ?

- Andrea Chincarini



SAPIENZA
UNIVERSITÀ DI ROMA