

# *Big data in health care: Why? How?*



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U.H. Maastricht

# The 4 « P's » of Precision medicine

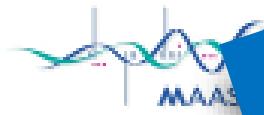
« P » for Personalized

« P » for Preventive

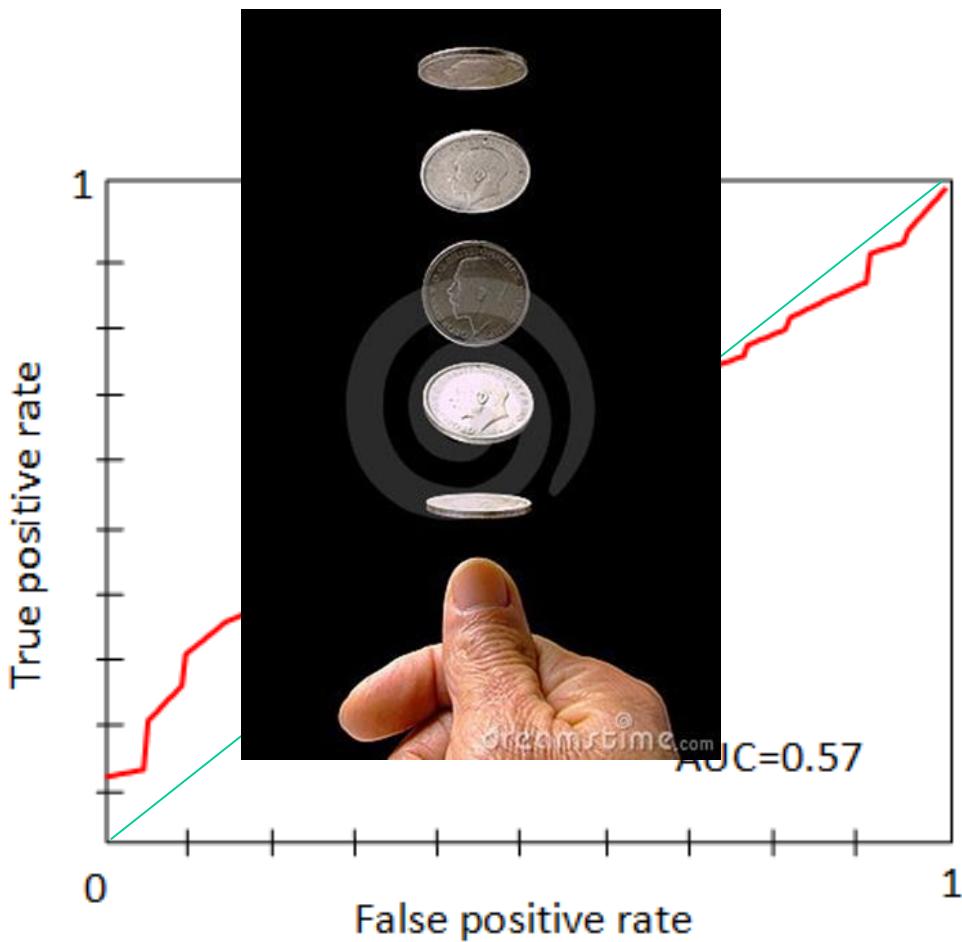
« P » for Predictive

« P » for Participative

And « P » for Particle therapy



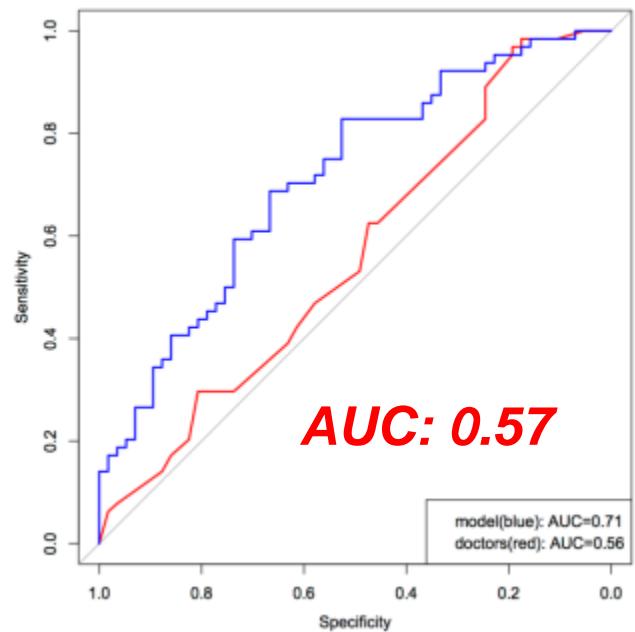
# Prediction by RadOnc's? Two years survival of inoperable NSCLC? Curative or palliative?



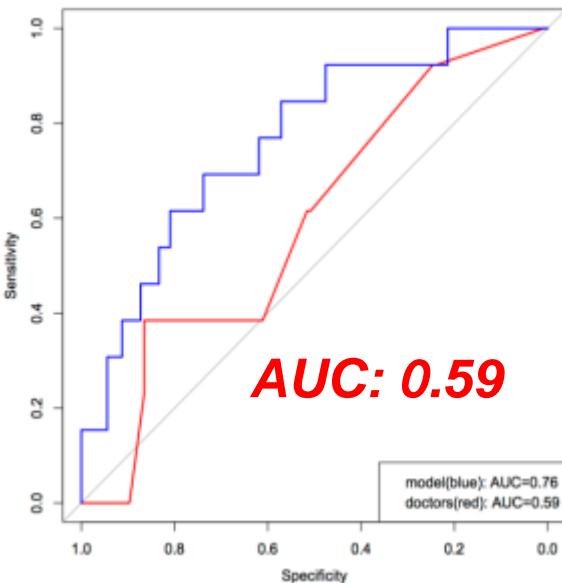
- Non Small Cell Lung Cancer
- 2 year survival
- 30 patients
- 8 MDs
- *Retrospective*
- ***AUC: 0.57 (AUC 0.5 = random)***

# Prospective trial (n=154): Models always significantly better than Rad Onc & TNM

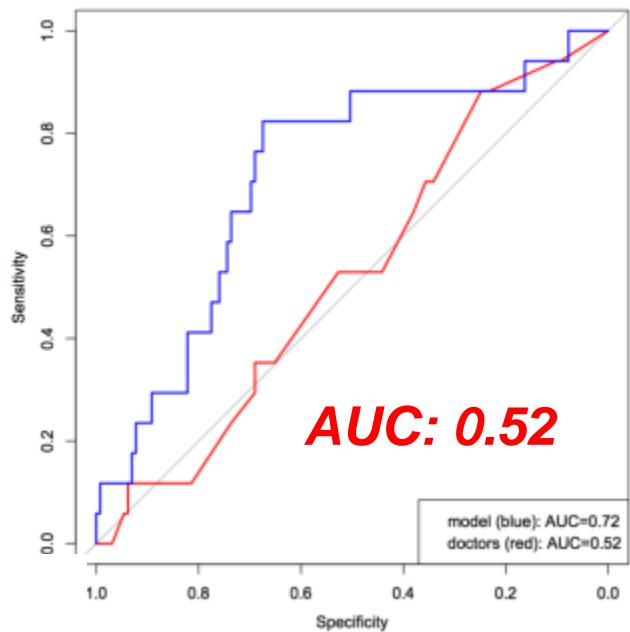
Death at 2 years



Dyspnea

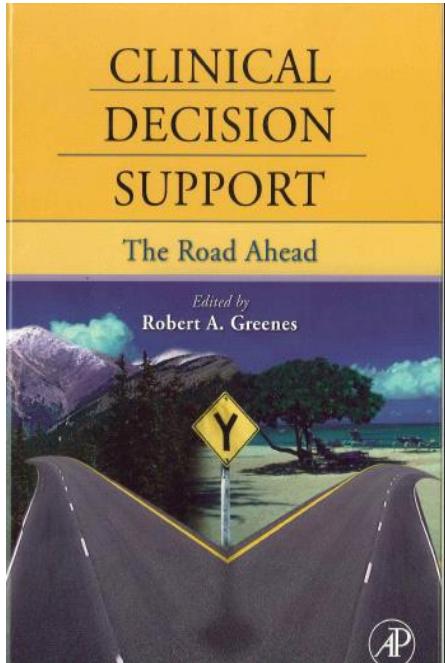


Dysphagia





**It is *unethical* to ask Medical Doctors to make, on their own, complex decisions, as it is to use only DVH.**



We need a validated “decision support system” (“a meta TPS”)

# Two type of clinical research

## Conventional Clinical Research

High data quality

Low data quantity

Controlled

- Selected patients
- “EORTC-RTOG grade” QA/Protocol

## Rapid Learning Health Care

Low data quality

High data quantity

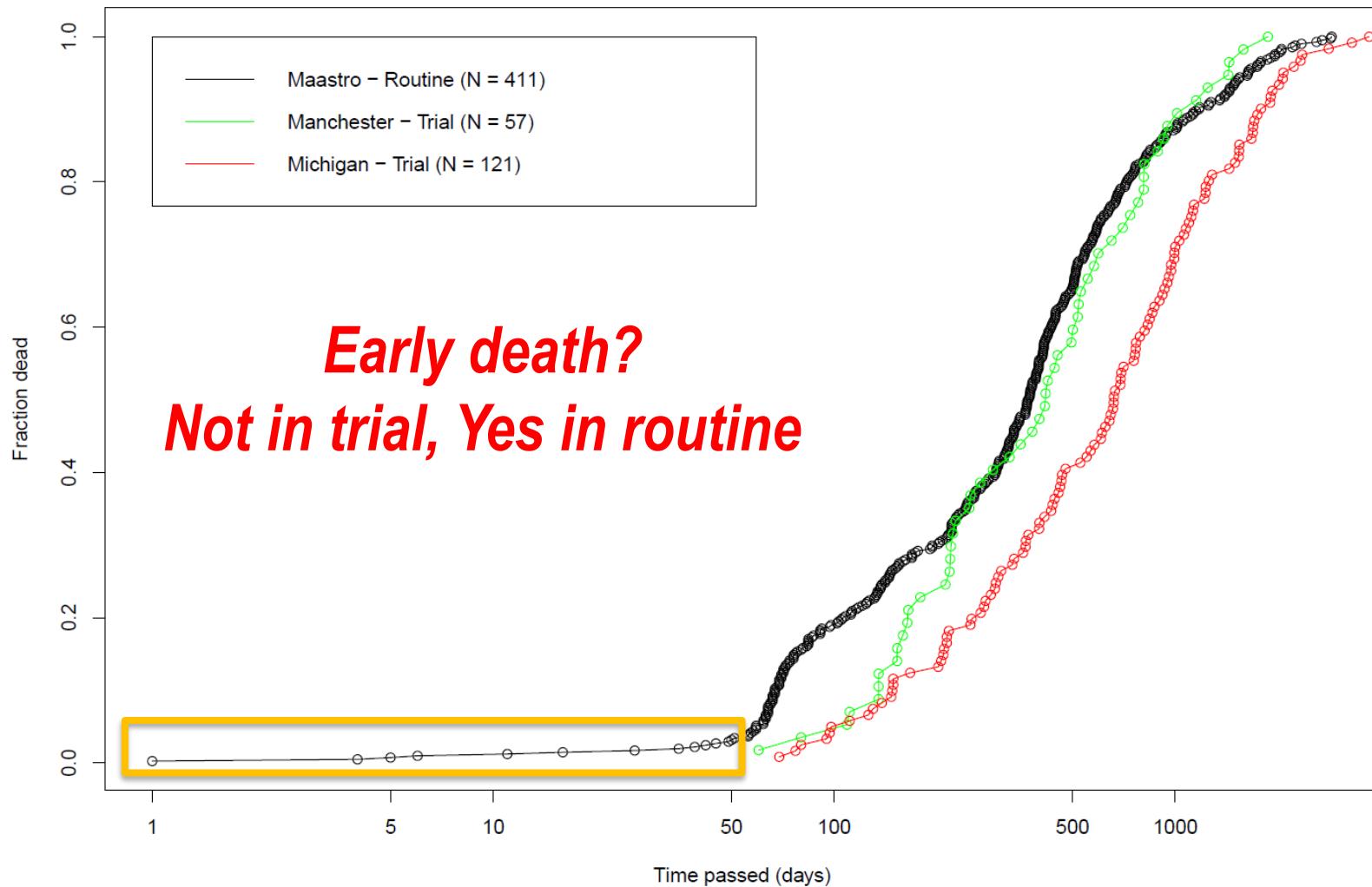
Reality

- Unselected patients
- “Clinical grade” QA/Protocol



# Example: Early death

Overall survival in clinical trials vs routine



"Data! Data! Data!" he cried  
impatiently.

"I can't make bricks without clay."

Sherlock Holmes - The Adventure of the Copper  
Beeches by Sir Arthur Conan Doyle

# Watch the animation

- Distributed learning: <https://youtu.be/nQpqMluHyOk>
- Eurocat: <https://youtu.be/ZDJFOxpwqEA>
- How to participate?  
<http://www.eurocat.info/membership.html>

# CAT network (<http://www.eurocat.info/join.html>)

## CORAL: Community in Oncology for Rapid Learning



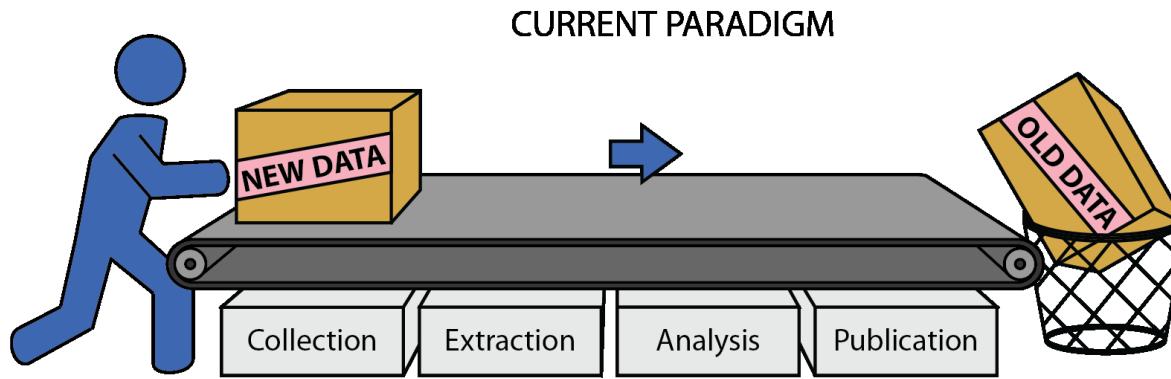
# How to get

## the data?

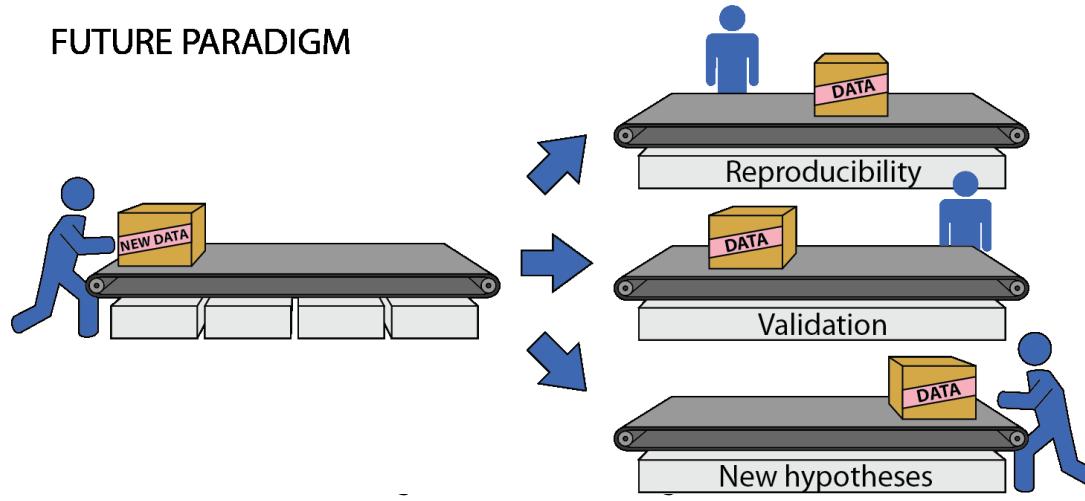
# The open repository approach



# *Utopia: All our published data available opensource*



## FUTURE PARADIGM



Modified from  
Deasy et al.

# Open source data of publications: [www.cancerdata.org](http://www.cancerdata.org)

[HOME](#)[DATA ▾](#)[LINKS ▾](#)[ABOUT ▾](#)

*Sharing medical data for cancer research*

## About CancerData

The *CancerData* site is an effort of the Medical Informatics and Knowledge Engineering team (*MIKE* for short) of Maastro Clinic, Maastricht, The Netherlands. Our activities in the field of medical image analysis and data modelling are visible in a number of projects we are running. Please refer to the [Links](#) for more information.

### Open source driven

*CancerData* is build using Free and Open Source Software (FOSS) only. Refer to [this page](#) for more information on the used software.

In return, we offer tools for image analysis and more. Have a look at the [file manager](#) (ps: allow popups).

### Contact us

Please use the [Contact form](#) for feedback or more information.



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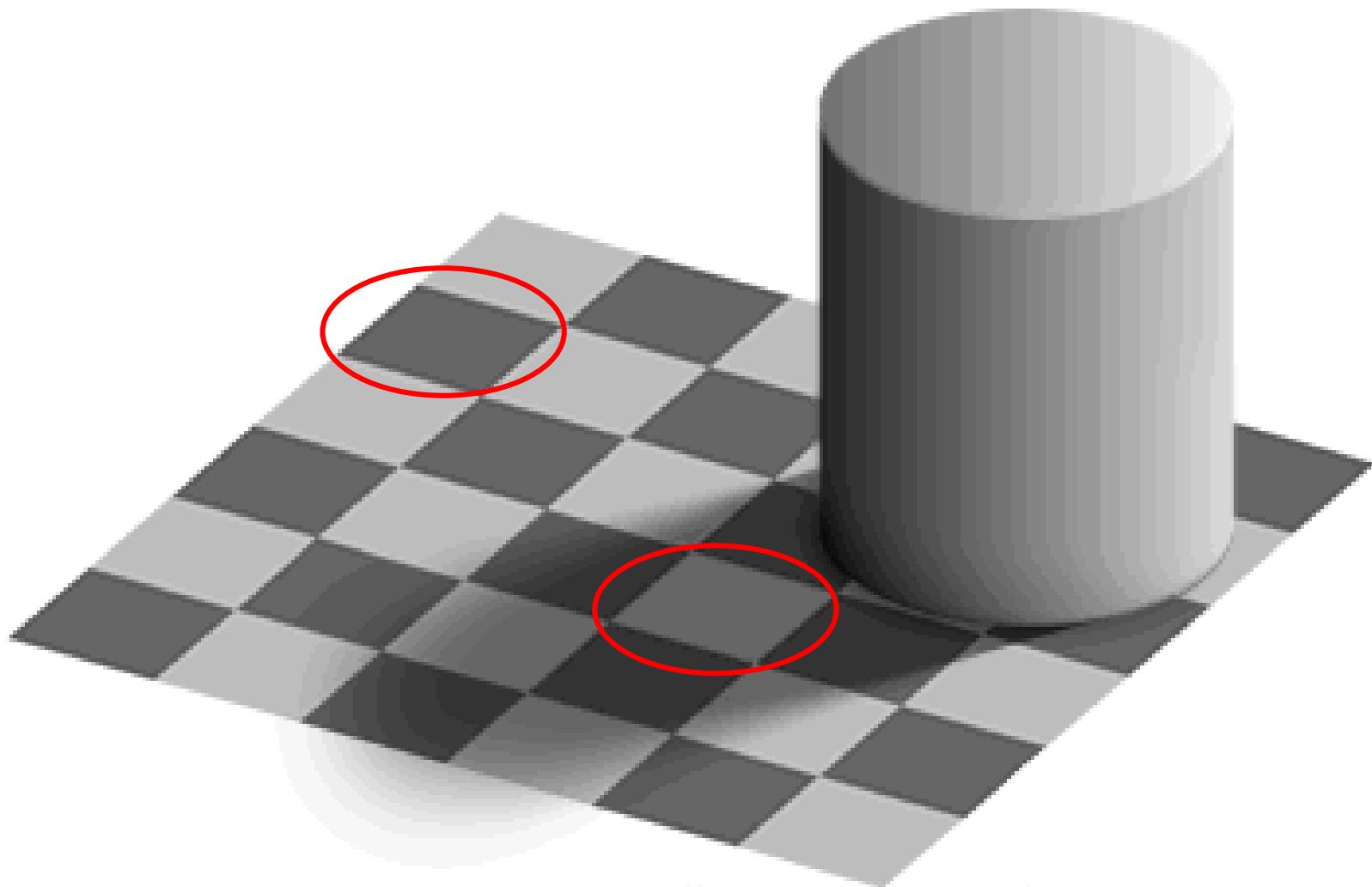
# An examples of new knowledge coming from Rapid Learning Health Care approaches

## *The Radiomics story*

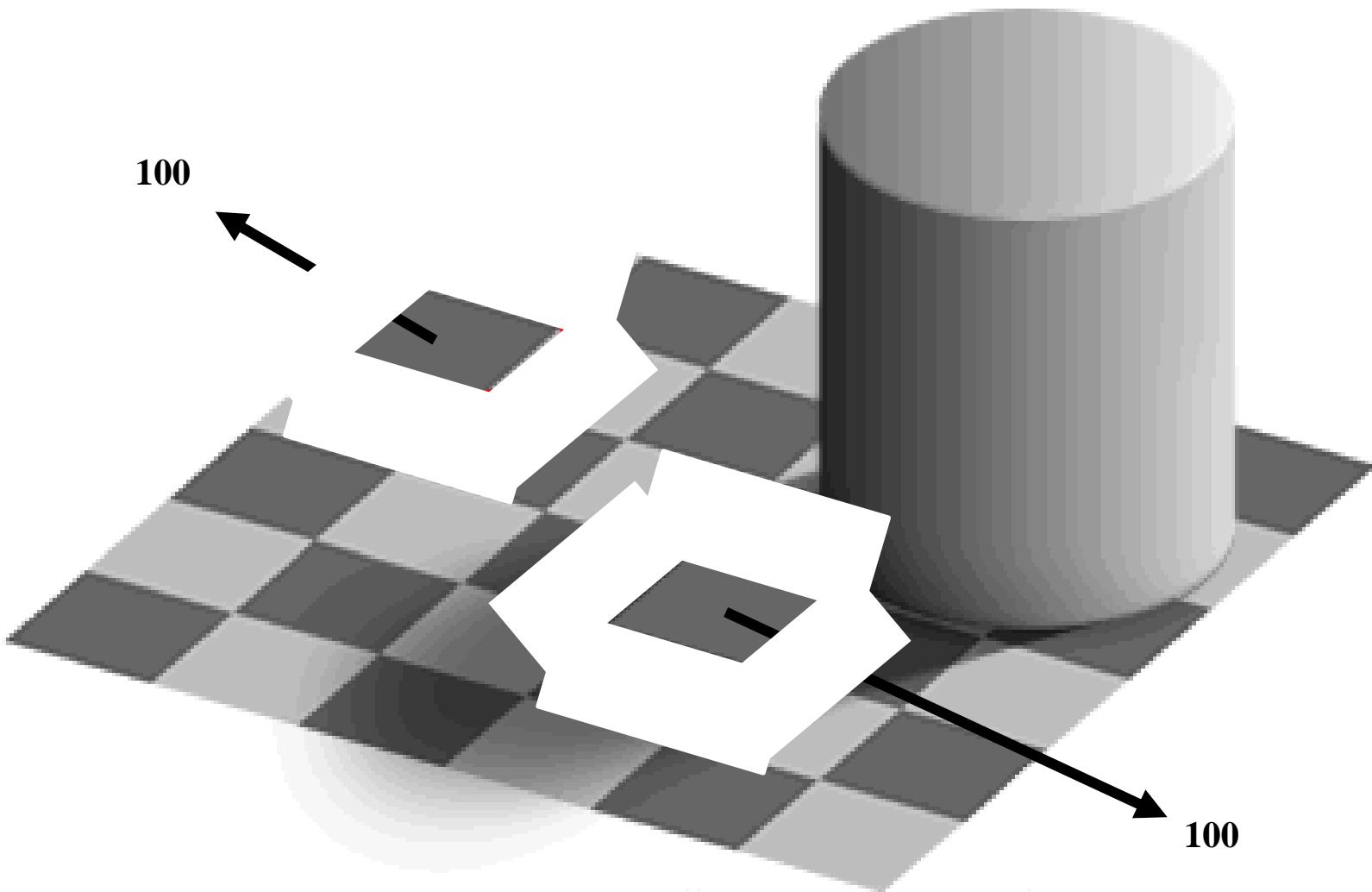


\*Lambin *et al.* Eur J Cancer 2012; Aerts, Lambin *et al.* Nature Commun 2014

# How good is our judgement?



# How good is our judgement?



The human eyes have

*serious limitations*

# Watch the animation

- Radiomics: <https://youtu.be/Tq980GEVP0Y>
- More info on [www.radiomics.org](http://www.radiomics.org)



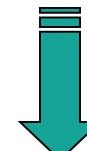
# The Radiomics hypothesis

One can extract more *quantitative* information from standard imaging in contrast with RECIST, WHO...)



## Radiology:

- Implicit knowledge
- Interpretability



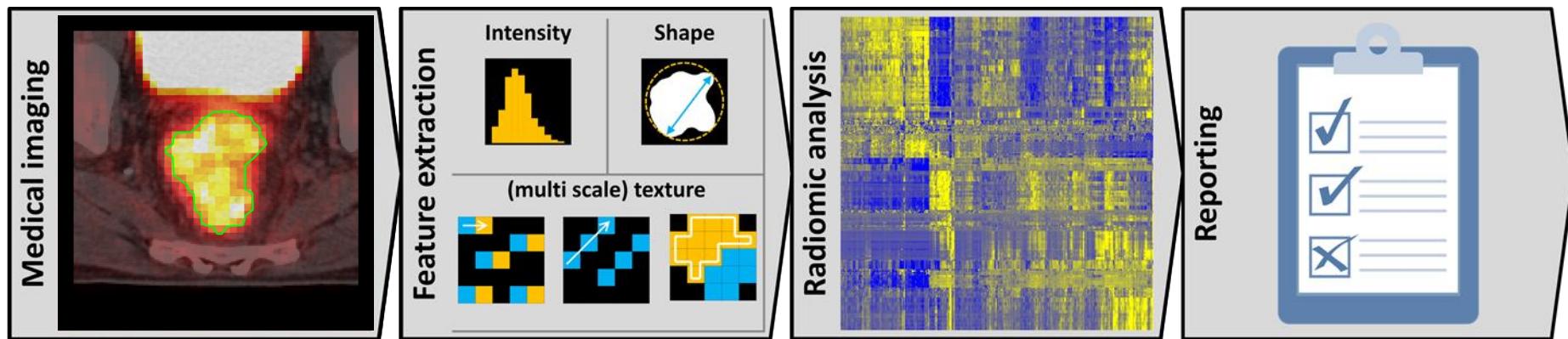
QUANTIFICATION



## RADIOMICS

Extract *quantitative* features from images

# The Radiomics workflow in a nutshell



A high-throughput approach to convert medical images to minable data

The analysis extracts diagnostic, theragnostic, prognostic and predictive Radiomics signatures



\*Lambin *et al.* Eur J Cancer 2012; Aerts, Lambin *et al.* Nature Commun 2014

ARTICLE

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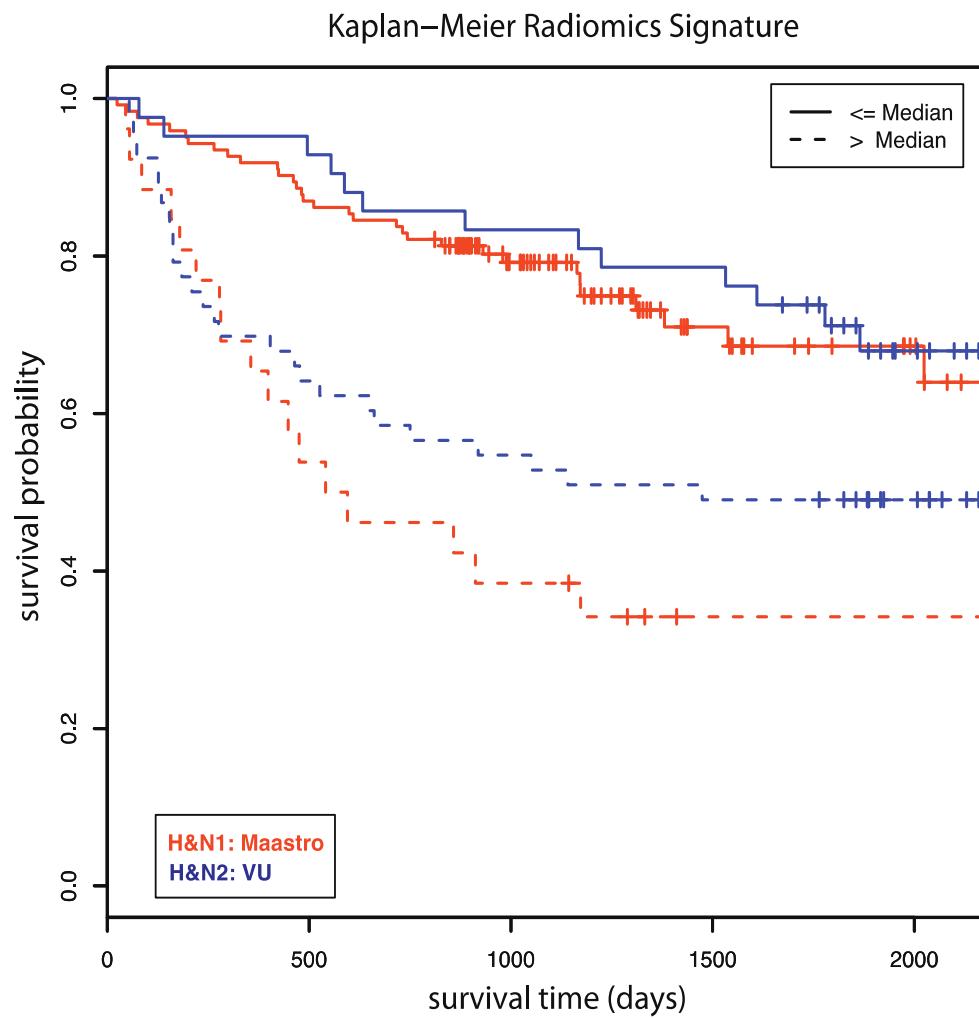
DOI: [10.1038/ncomms5006](https://doi.org/10.1038/ncomms5006)

OPEN

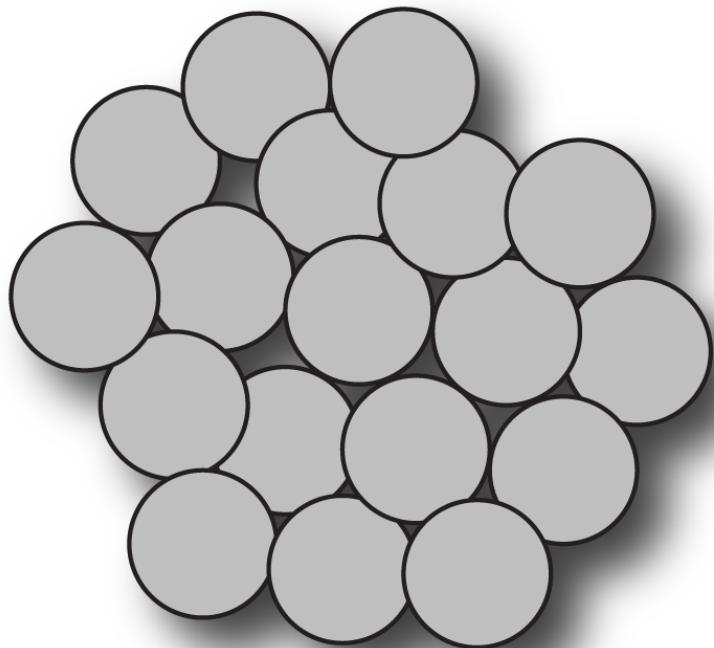
# Decoding tumour phenotype by noninvasive imaging using a quantitative radiomics approach

Hugo J.W.L. Aerts<sup>1,2,3,4,\*</sup>, Emmanuel Rios Velazquez<sup>1,2,\*</sup>, Ralph T.H. Leijenaar<sup>1</sup>, Chintan Parmar<sup>1,2</sup>, Patrick Grossmann<sup>2</sup>, Sara Cavalho<sup>1</sup>, Johan Bussink<sup>5</sup>, René Monshouwer<sup>5</sup>, Benjamin Haibe-Kains<sup>6</sup>, Derek Rietveld<sup>7</sup>, Frank Hoebers<sup>1</sup>, Michelle M. Rietbergen<sup>8</sup>, C. René Leemans<sup>8</sup>, Andre Dekker<sup>1</sup>, John Quackenbush<sup>4</sup>, Robert J. Gillies<sup>9</sup> & Philippe Lambin<sup>1</sup>

# Predict survival in Lung and Head & neck cancer better then TNM



# ...in other words, visualize



Good

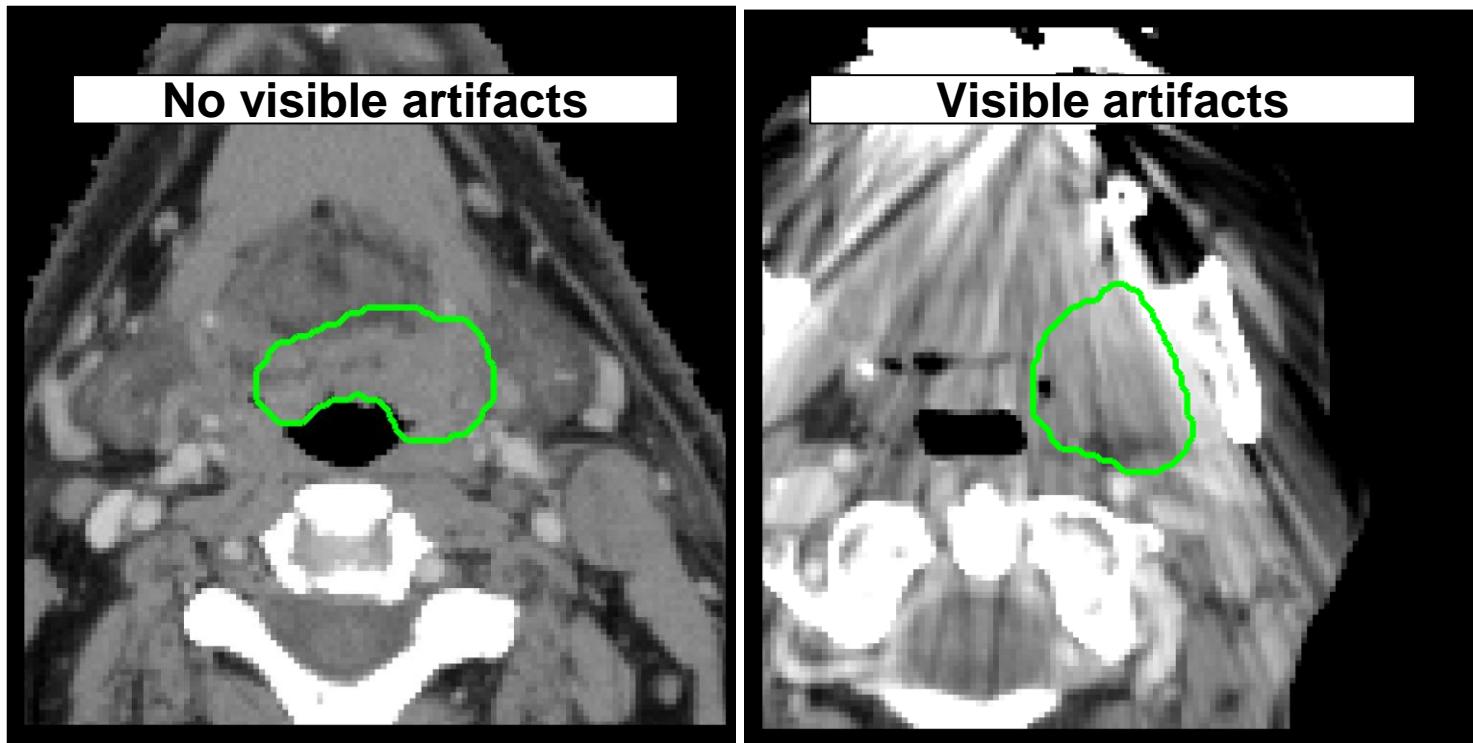


Bad  
Higher proliferation

# What about

## North americans?

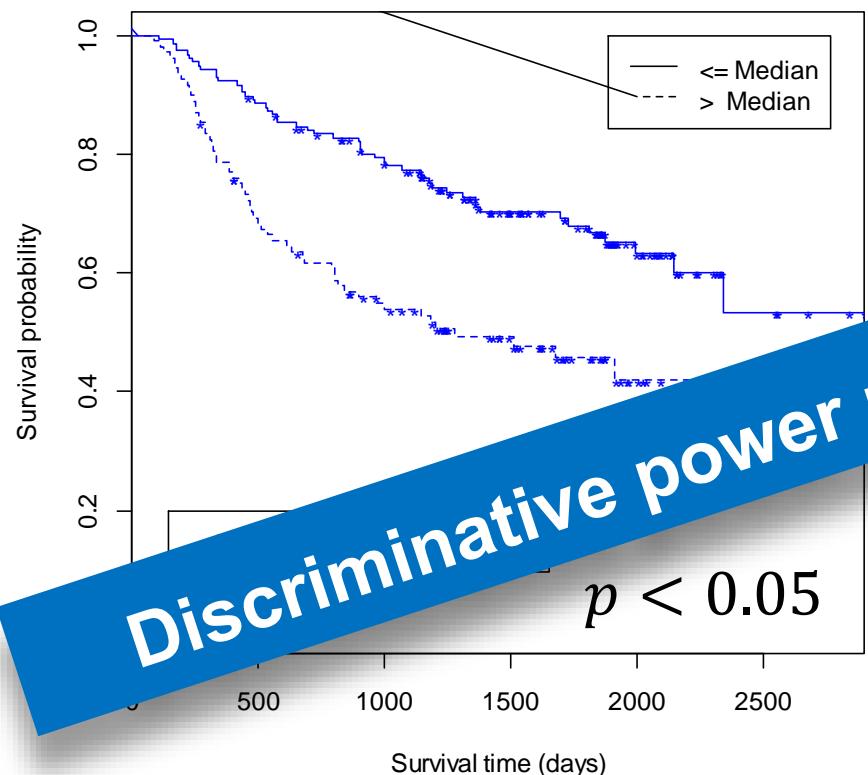
# ...and what about CT artifacts?



The signature was built on Lung cancer data → CT artifacts uncommon (e.g. pacemakers)

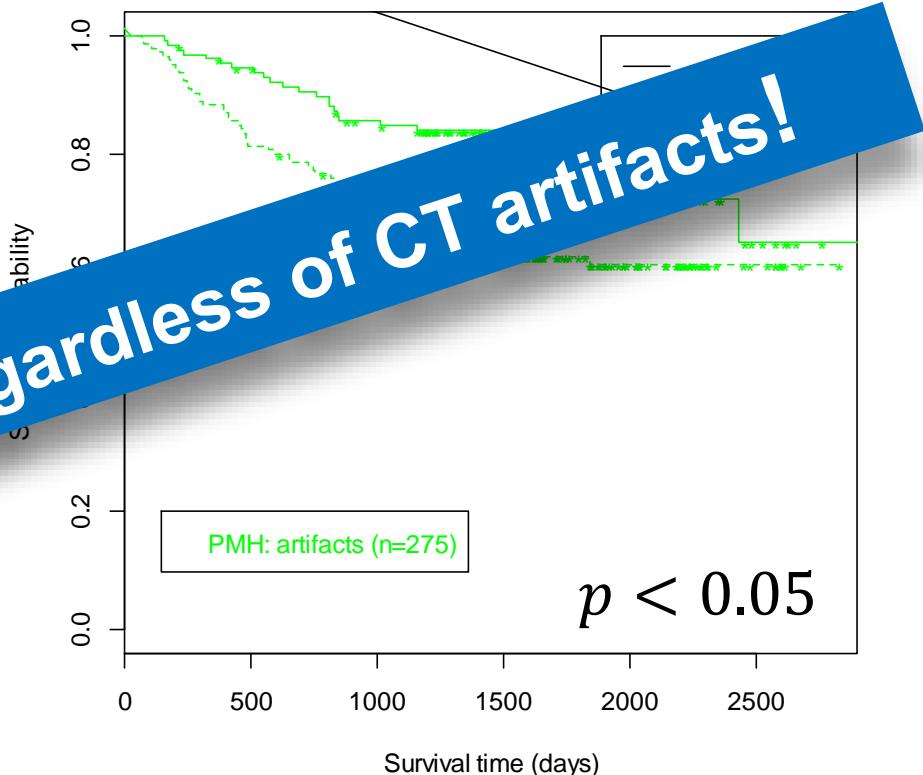
# Is the signature robust? what about CT artifacts in H&N?

Kaplan-Meier Radiomic Signature



Without artifacts (N=267)  
C-index: 0.634

Kaplan-Meier Radiomic Signature



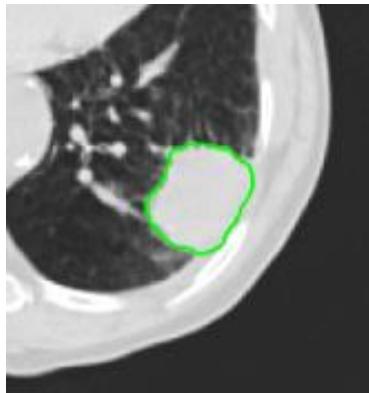
With artifacts (N=275)  
C-index: 0.647

# New

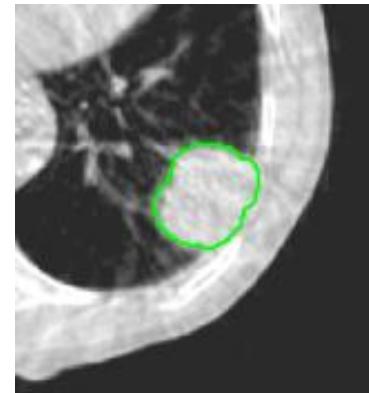
## Would it work with

### cone beam CT?

Conventional CT



vs. kV cone-beam CT



2016 :

“Deep learning”-based software has  
beaten the champion of the “Go game”

Example with the Atari game

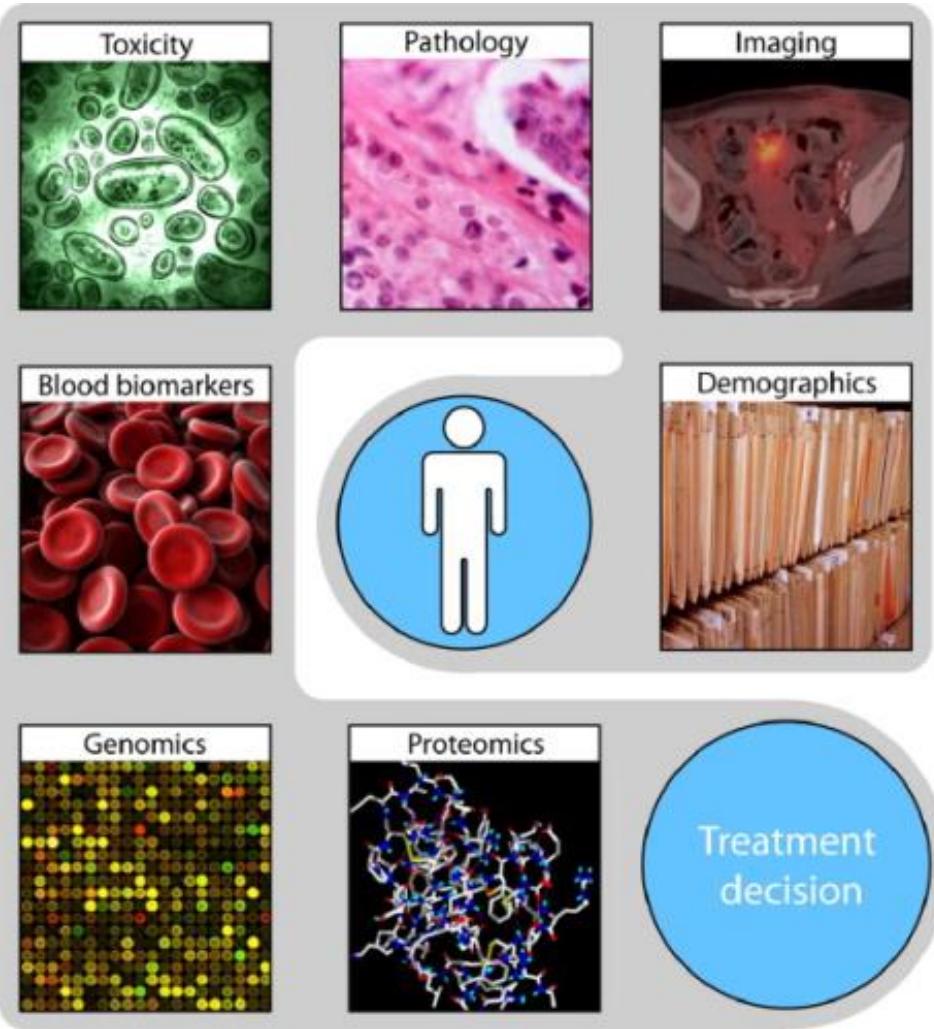
Give me an

example?

The model-based indications for protons

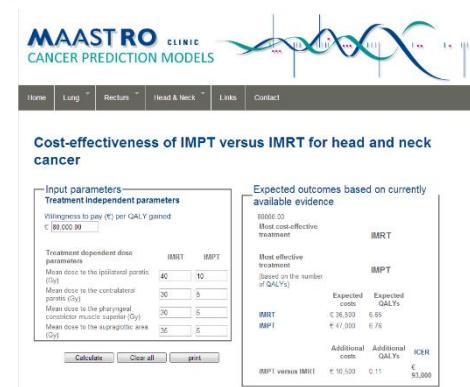
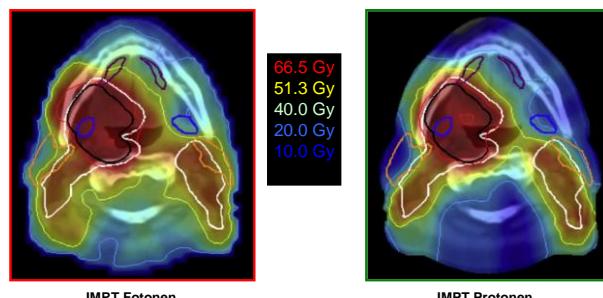
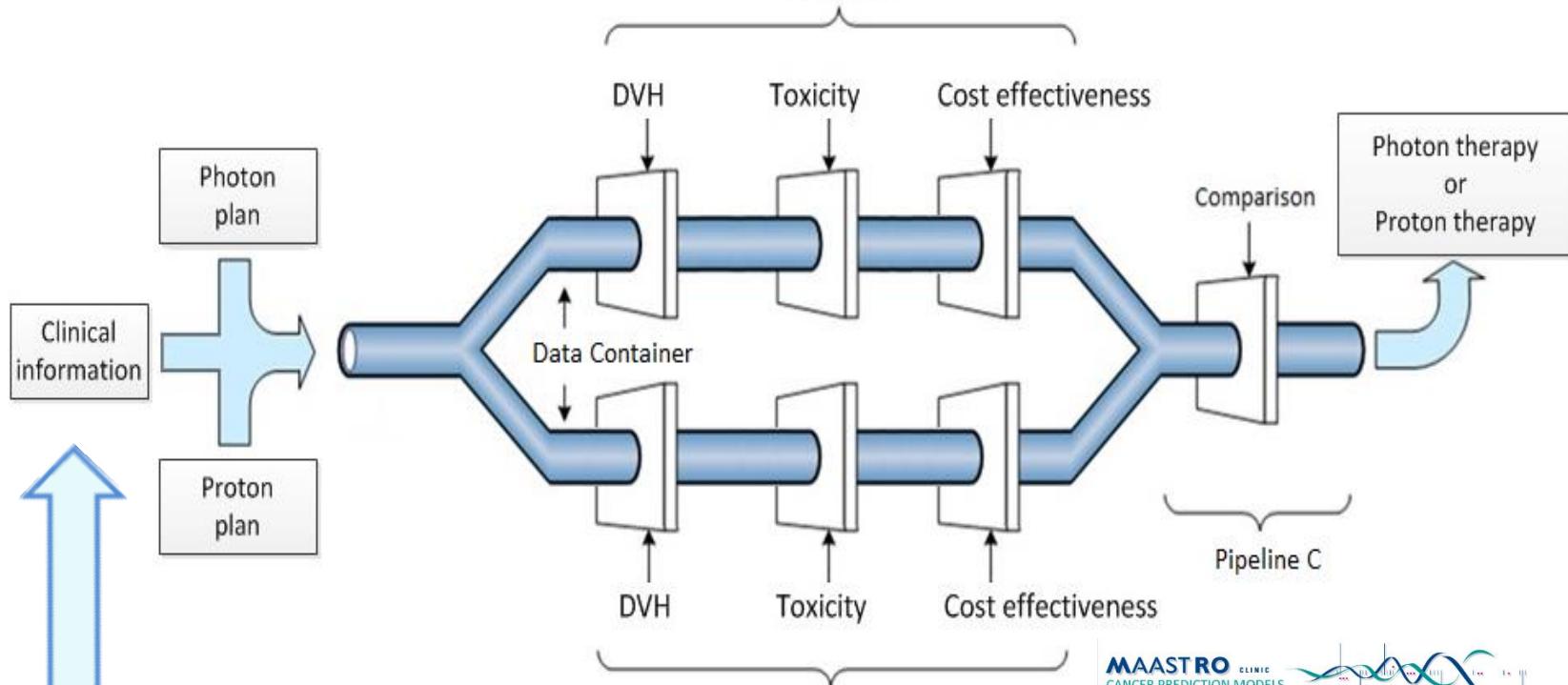


# *Build a machine for precision medicine: “Multifactorial Decision Support System”*



A meta-TPS  
Reuniting  
Physics &  
Biology

# Proton therapy Decision Support System (PRODECIS)

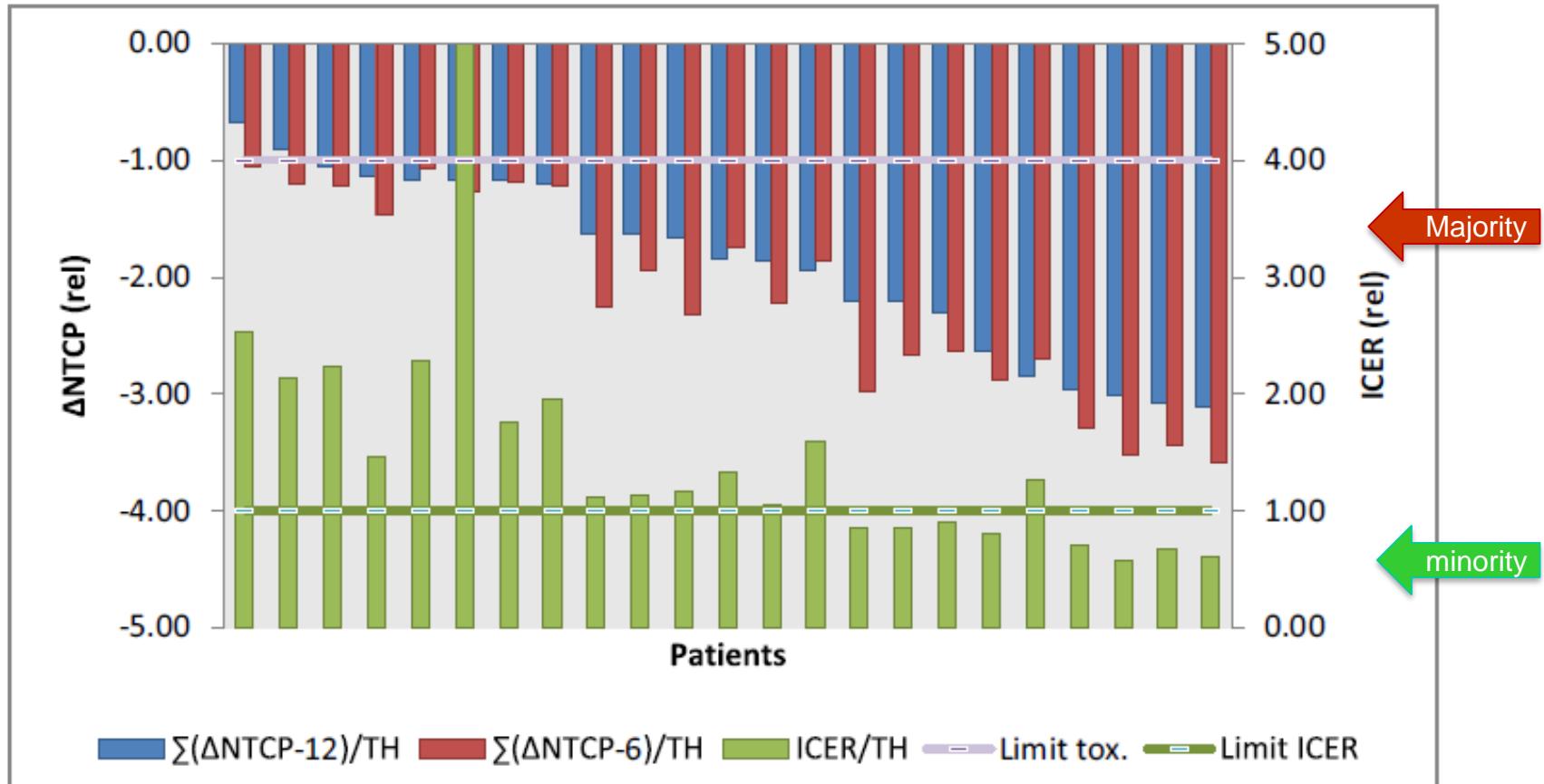


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Cheng Q, Roelofs E, Lambin P et al. Radiother Oncol 2016

# Proton therapy Decision Support System (PRODECIS)



# Conclusions

- *We need a Big Data-based Decision Support System, in particular for particle therapy.*
- *The “fashion” of Big Data induces a change of culture: to gather data.*
- *Radiomics is an approach which allows us to extract clinical and biological information from standard imaging (including cone-beam CT).*

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- Wouter van Elmpt

