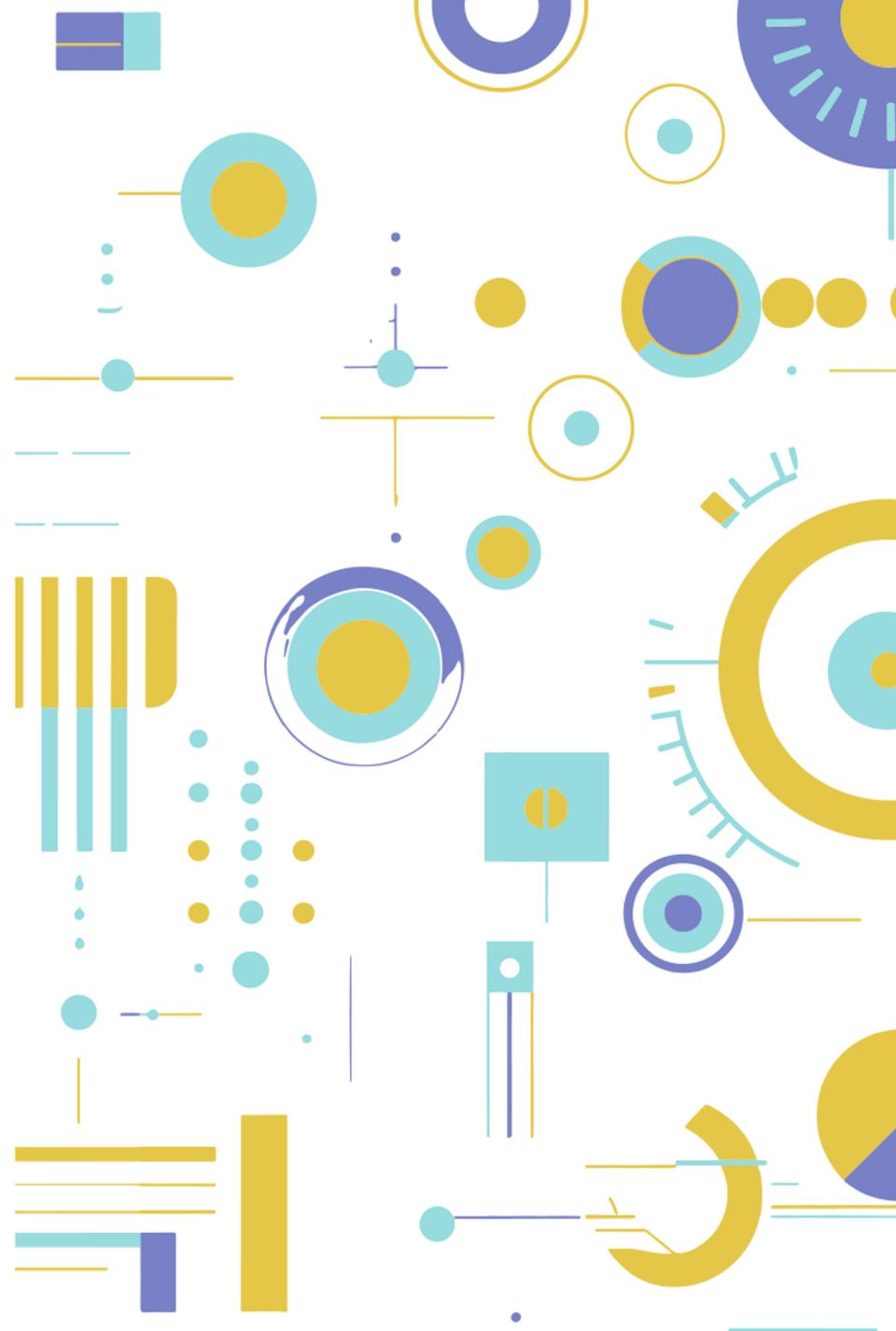


# CBEB 2024

XXIX CONGRESSO BRASILEIRO  
DE ENGENHARIA BIOMÉDICA

Realização





**CBEB 2024**

XXIX CONGRESSO BRASILEIRO  
DE ENGENHARIA BIOMÉDICA



# IA na saúde

Profa. Dra. Leticia Rittner

Dr. Diedre do Carmo

Medical Image Computing Lab. (MICLab)

Faculdade de Engenharia Elétrica e de Computação - Unicamp



# WHY DEEP LEARNING IS SUDDENLY CHANGING YOUR LIFE

Decades-old discoveries are now electrifying the computing industry and will soon transform corporate America.

## Google's New Chip May Be the Future of AI Systems

Google's latest revelation may be the biggest advancement yet in chips that power artificial intelligence systems, but other companies are releasing innovations of their own.

DIGITAL • UGNUT



nature

International weekly journal of science

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 Eren Kocigit  
12 January 2017

CATEGORIES

NATURE | INSIGHT | REVIEW

Deep learning

Yann LeCun, Yoshua Bengio &amp; Geoffrey Hinton

Affiliations | Corresponding author

Nature 521, 436–444 (28 May 2015) | doi:10.1038/nature14539

Received 25 February 2015 | Accepted 01 May 2015 | Published online 27 |

## How Deep Learning is Transforming the Future of Technology?

There is no doubt that Big Data has been one of the most popular topics among marketers and tech enthusiasts for several years.

CRUNCH NETWORK

## Advances in AI and ML are reshaping healthcare

Posted Mar 16, 2017 by Megh Gupta (@guptamegh), Qasim Mohammad (@QasimAMohammad)



WIRED

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## 2016: THE YEAR THAT DEEP LEARNING TOOK OVER THE INTERNET



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NATURE | INSIGHT | REVIEW

Deep learning

Yann LeCun, Yoshua Bengio &amp; Geoffrey Hinton

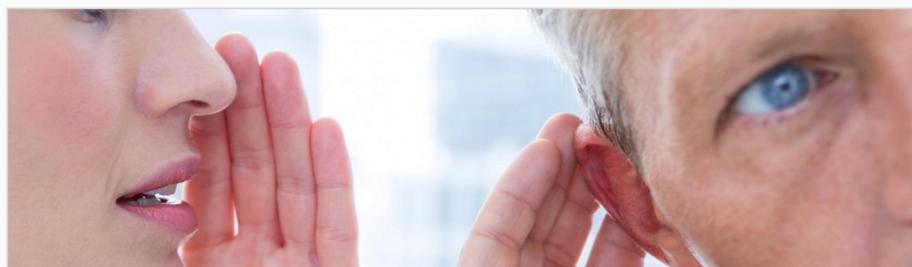
Affiliations | Corresponding author

Nature 521, 436–444 (28 May 2015) | doi:10.1038/nature14539

Received 25 February 2015 | Accepted 01 May 2015 | Published online 27 |

## Microsoft hits a speech recognition milestone with a system just as good as human ears

Posted Oct 18, 2016 by Devin Coldewey



Entrepreneur

TOP 50

HOW TO

MAGAZINE

FRANCHISE 500

EVENTS

## Can Artificial Intelligence Identify Pictures Better than Humans?

ForbesCommunityVoice™ Connecting expert con

Tech / #NewTech

APR 17, 2017 @ 07:00 AM 1,643

## It's Time To Embrace AI's Superior Prediction Powers



Image credit: Shutterstock.com

Porque levou  
tanto tempo?



1958 Perceptron

1986 Backpropagation

awkward silence (AI Winter)

1969  
Perceptron criticized



Convolution Neural Networks for  
Handwritten Recognition

1998

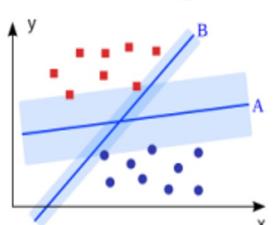


Google Brain Project on  
16k Cores

2012



1995  
SVM reigns

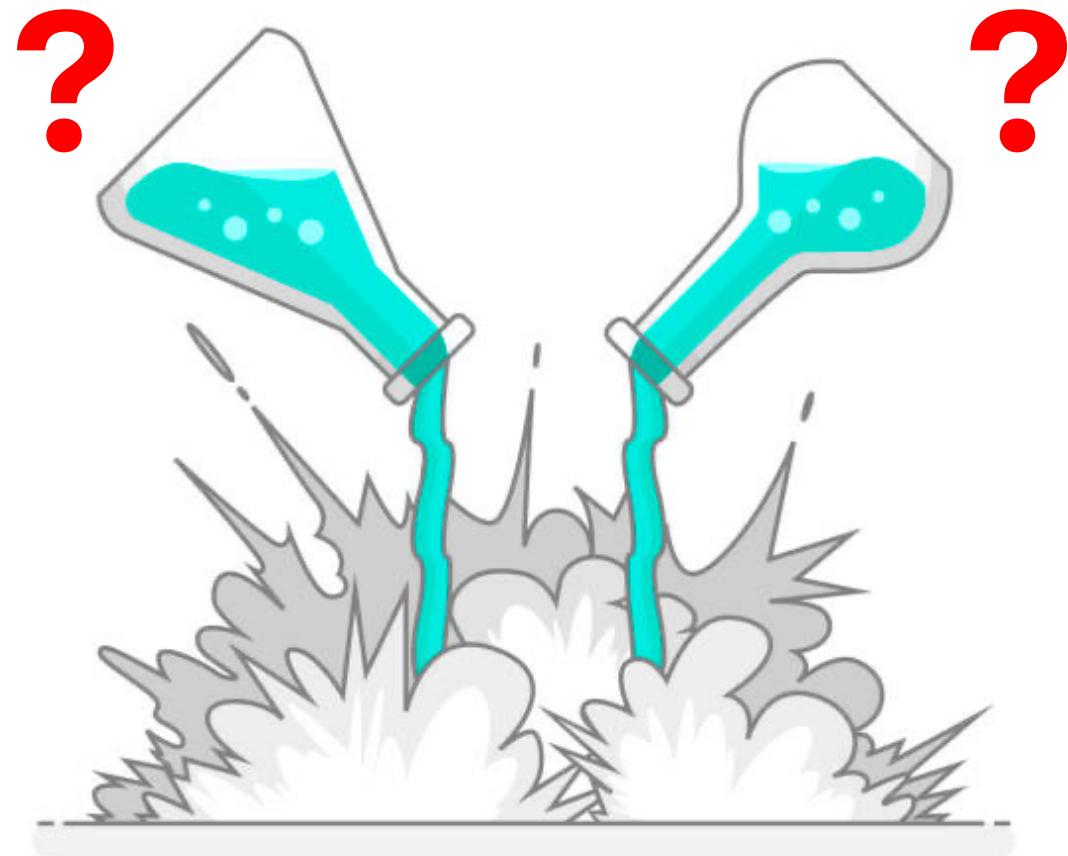


2006  
Restricted  
Boltzmann  
Machine



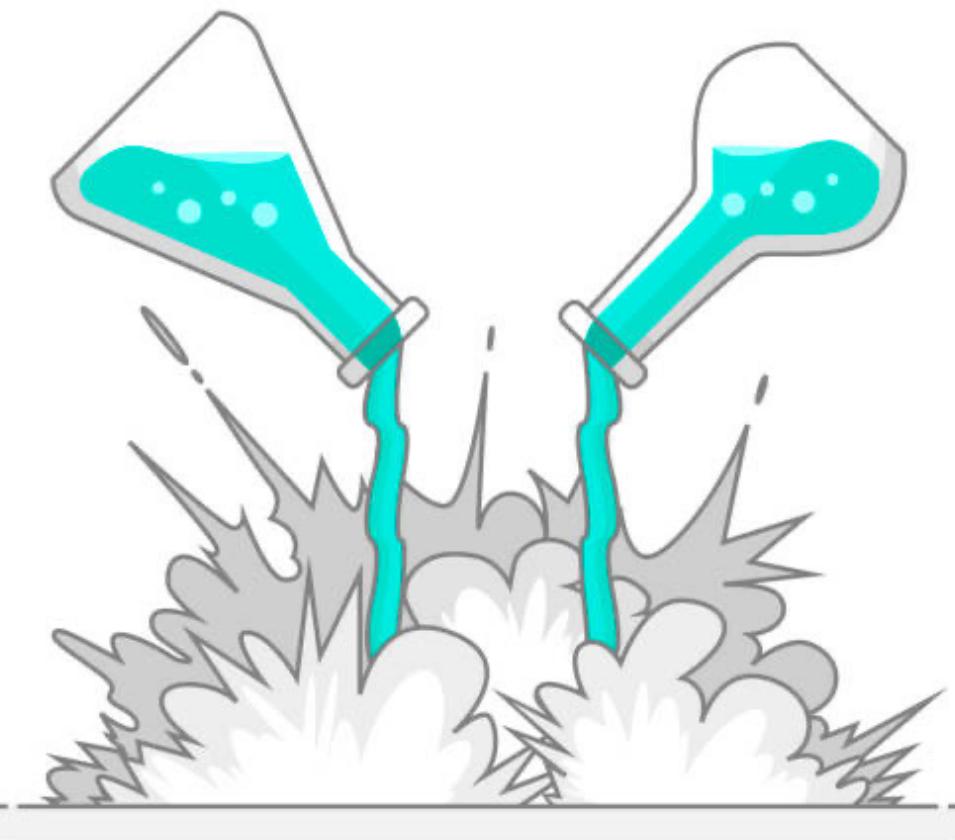
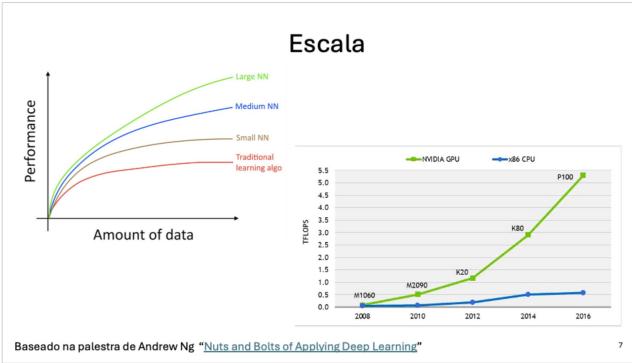
2012  
AlexNet wins  
ImageNet





Aprendizado profundo

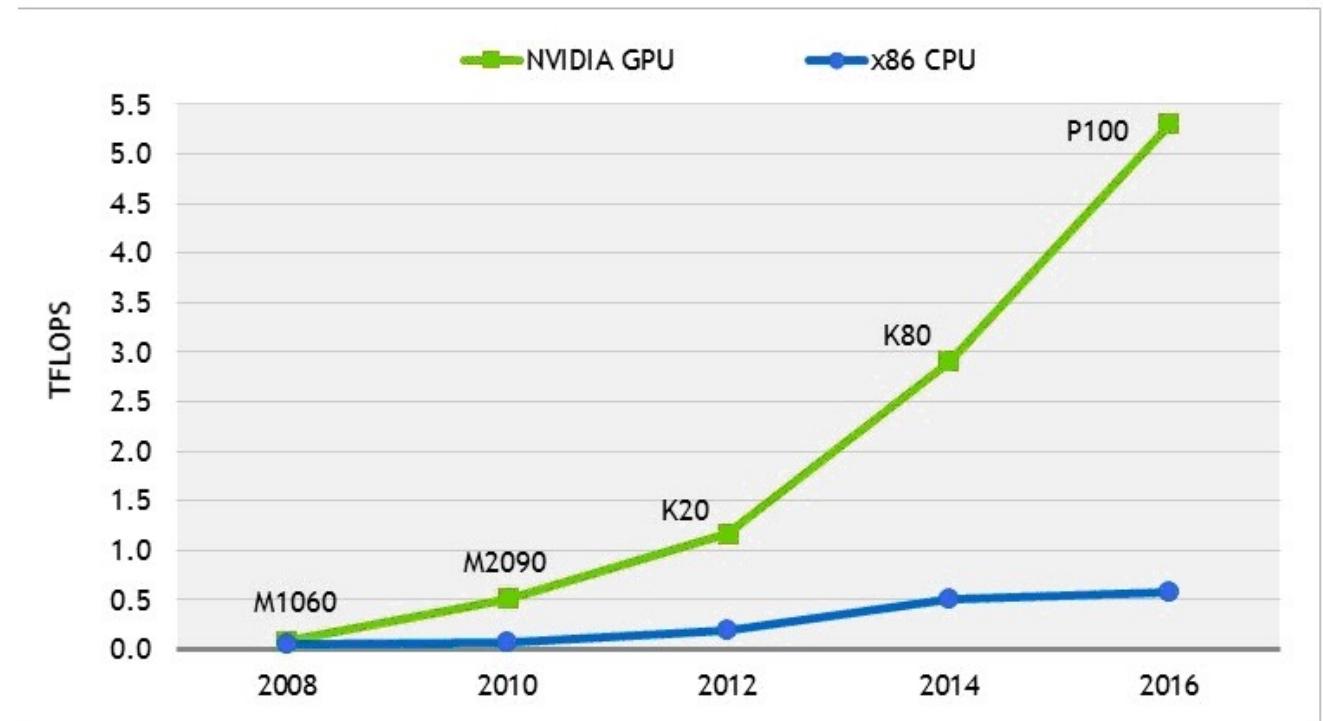
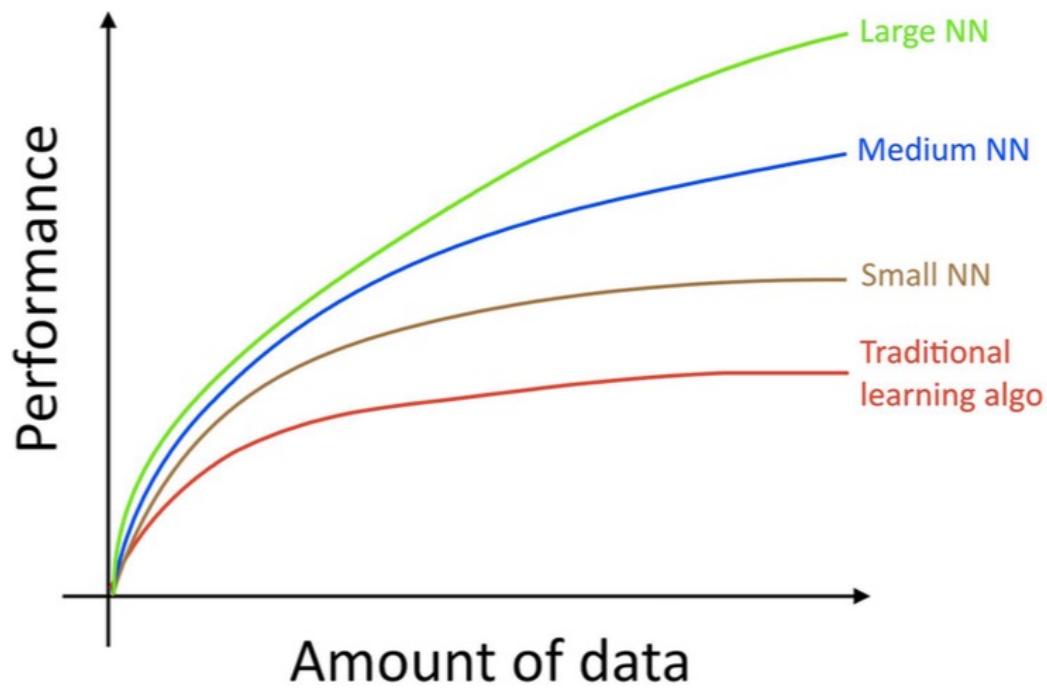




# Aprendizado profundo



# Escala



# Ciência aberta





# Ciência aberta

IP[y]: Notebook spectrogram Last Checkpoint: a few seconds ago (autosaved) IPython (Python 3)

File Edit View Insert Cell Kernel Help

Simple spectral analysis

An illustration of the [Discrete Fourier Transform](#) using windowing, to reveal the frequency content of a sound signal.

$$X_k = \sum_{n=0}^{N-1} x_n e^{-\frac{2\pi i}{N} kn} \quad k = 0, \dots, N - 1$$

We begin by loading a datafile using SciPy's audio file support:

```
scipy.io import wavfile
x, sr = wavfile.read('test_mono.wav')
```

we can easily view its spectral structure using matplotlib's builtin specgram routine:

```
matplotlib inline
matplotlib import pyplot as plt
(ax1, ax2) = plt.subplots(1, 2, figsize=(12, 4))
plot(x); ax1.set_title('Raw audio signal')
specgram(x); ax2.set_title('Spectrogram');
```



Why GitHub? Team Enterprise Explore Marketplace Pricing

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Code

?

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156K

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707

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43

## Machine learning

Machine learning is a way of modeling and interpreting data to respond intelligently.

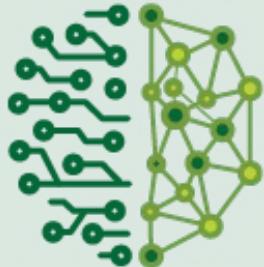
[See topic](#)

320,049 repository results

# AI

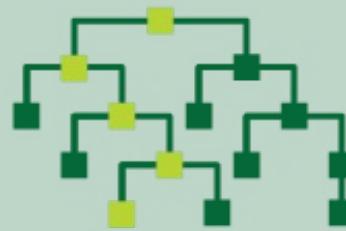
## Artificial Intelligence

Any technique that enables computers to mimic human behavior.



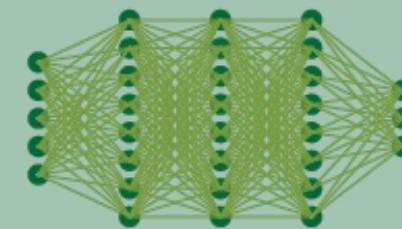
## Machine Learning

The ability to learn without directly being programmed.



## Deep Learning

The learning of underlying features in data using deep neural networks.



# Como as máquinas aprendem?

**Learn From Experience**



**Follow Instructions**



Sunday Telegraph

March 18, 2018

# body + soul

## THE DOCTOR WILL SEE YOU NOW

How robots are changing  
the face of medicine



a Will doctors be re

# **Inteligência artificial eleva em 20% detecção de câncer de mama e 'reduz' trabalho de radiologistas**

Estudo publicado na revista The Lancet Oncology apontou que a inteligência artificial pode tornar a triagem de mamografia mais precisa e eficiente. Os pesquisadores ressaltam, no entanto, que a inteligência artificial não está pronta para ser implementada no 'mundo real'.

# Modelo de aprendizado de máquina prevê AVC no momento da triagem

24/04/2023 às 14:00 • 1 min de leitura

Desenvolvido por pesquisadores das universidades Carnegie Mellon, Florida International e Santa Clara, nos EUA, [um novo modelo de máquina consegue prever](#), com alto grau de precisão, a ocorrência de derrames cerebrais **ainda no momento da triagem dos pacientes.**

A pesquisa considerou o modelo de previsão "com alta sensibilidade e especificidade razoável". Seu grande diferencial é que as variáveis usadas já são coletadas no dia a dia por provedores e pagadores. Se não pode ser considerado um "padrão-ouro" para diagnosticar AVC, o modelo é um complemento útil para sistemas de pontuação existentes, **especialmente em hospitais com pouca disponibilidade de recursos e ferramentas diagnósticas.**

# **Sonho possível? Ministério da Saúde quer adotar inteligência artificial no SUS, mas enfrenta desafios**

— Para pensar no uso da IA, precisamos primeiro completar o processo de digitalização do SUS. Ainda usamos prontuários médicos físicos, por exemplo. Precisamos desses materiais em formato digital, e isso demanda trabalho e investimento, mas como retorno temos otimização dos diagnósticos, dos equipamentos médicos e do próprio sistema de dados do SUS — diz Maranhão.

Em nota, o Ministério da Saúde afirma que o uso da IA tem o potencial de aprimorar os diagnósticos e tratamentos de doenças, o desenvolvimento de novos medicamentos e a realização de cirurgias mais precisas. Essa também é uma realidade em expansão na vigilância epidemiológica, no telecuidado e na utilização de boas evidências traduzidas em algoritmos clínicos", complementa.

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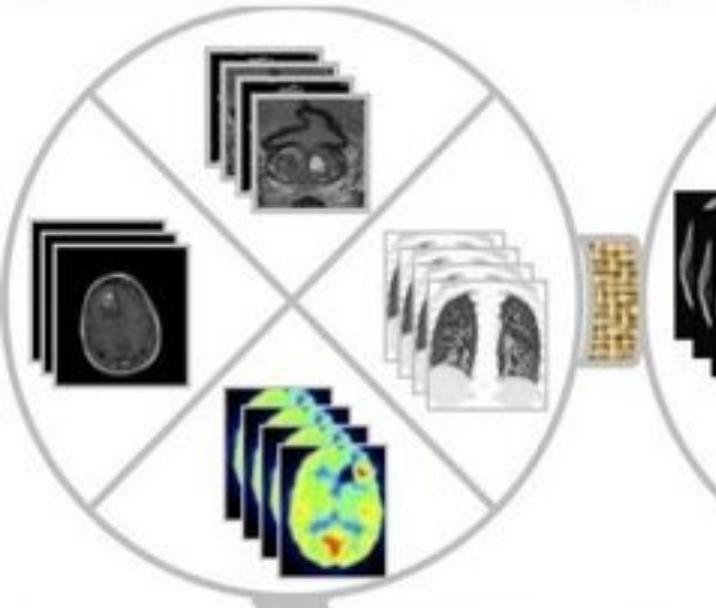

TECNOLOGIA DA INFORMAÇÃO

## A inteligência artificial chega à saúde

Tecnologias dotadas desses recursos já são usadas na área médica, mas desafios precisam ser superados para sua incorporação em maior escala no país

De acordo com o relatório “Artificial intelligence index report 2022”, da Universidade Stanford, foram investidos pelo setor privado em todo o mundo US\$ 11,3 bilhões em pesquisa e inovação com IA para medicina e saúde em 2021, um aumento de 40% em relação ao ano anterior. Nos últimos cinco anos, os recursos somaram US\$ 28,9 bilhões, o que posicionou o segmento como o maior receptor de investimentos privados em IA, superando atividades tradicionais no uso de tecnologias da informação, como o setor financeiro e o varejo. A visão computacional, segmentando imagens de órgãos, lesões ou tumores, foi uma das aplicações que despertaram mais interesse na comunidade médica.

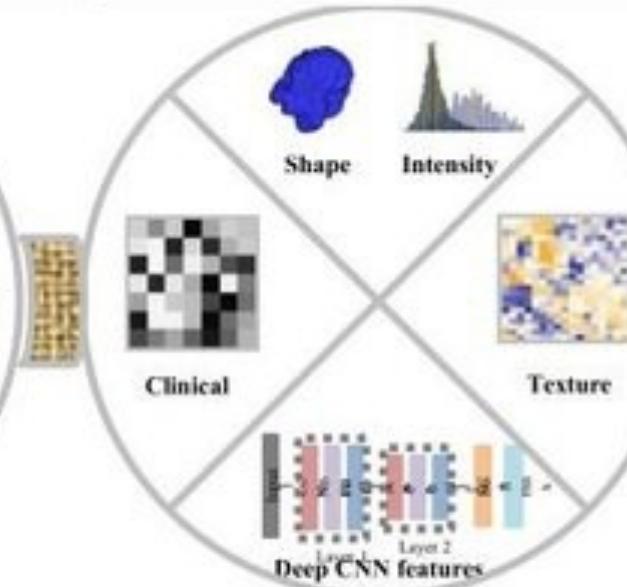
## Aquisição das Imagens



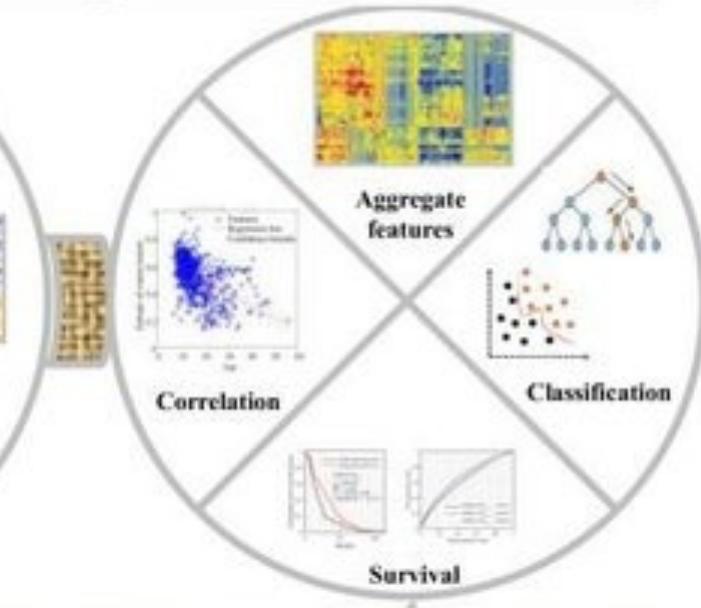
## Segmentação / Seleção de ROI



## Extração de atributos



## Análise / Classificação





Posso usar DL para classificar sujeitos em grupo A e B olhando suas imagens?

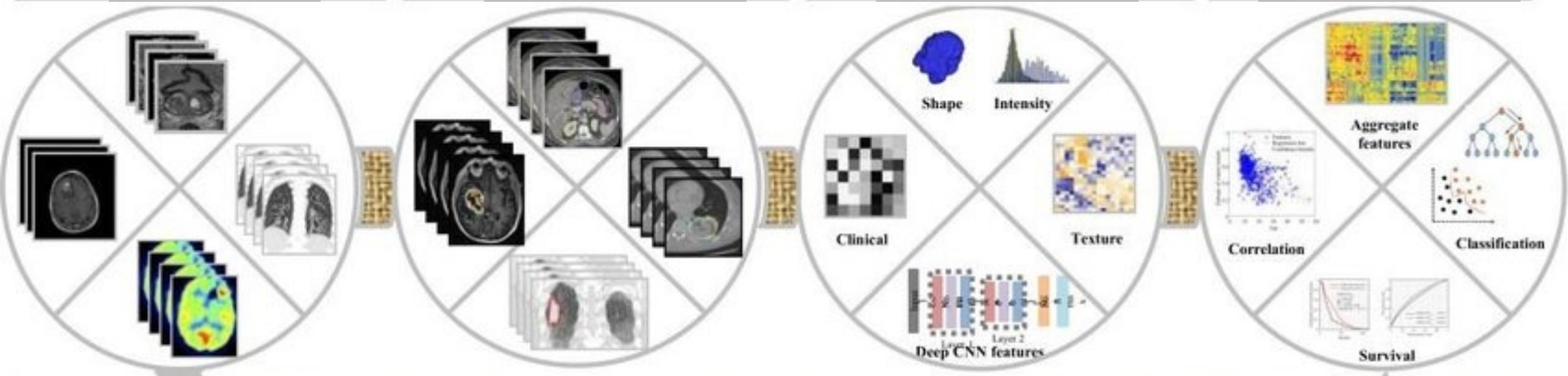


### Aquisição das Imagens

### Segmentação / Seleção de ROI

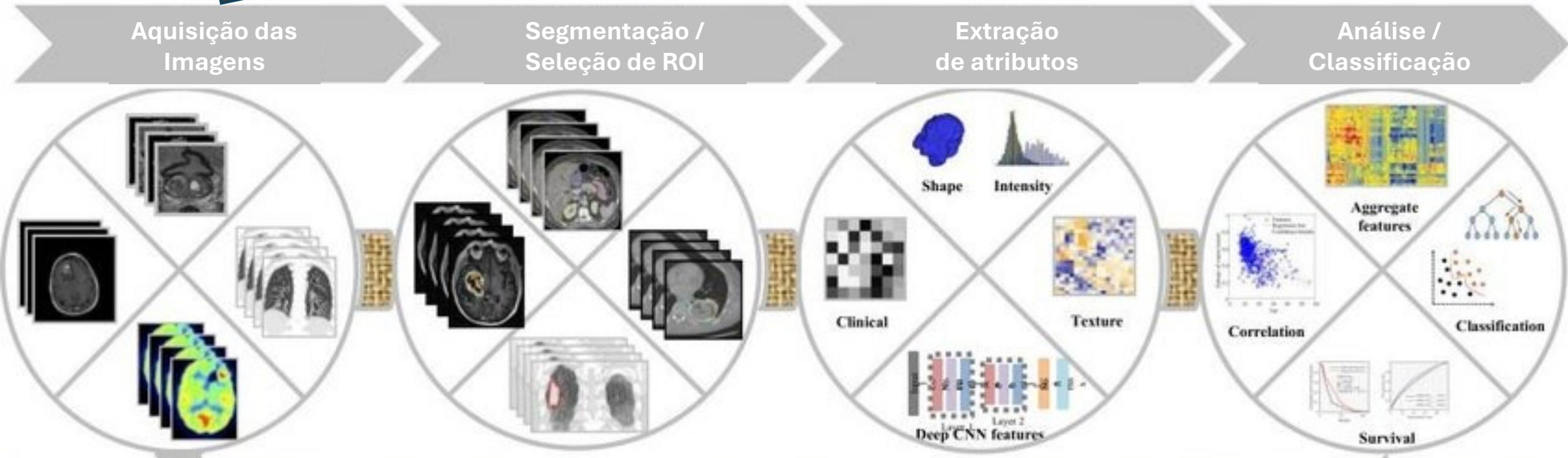
### Extração de atributos

### Análise / Classificação

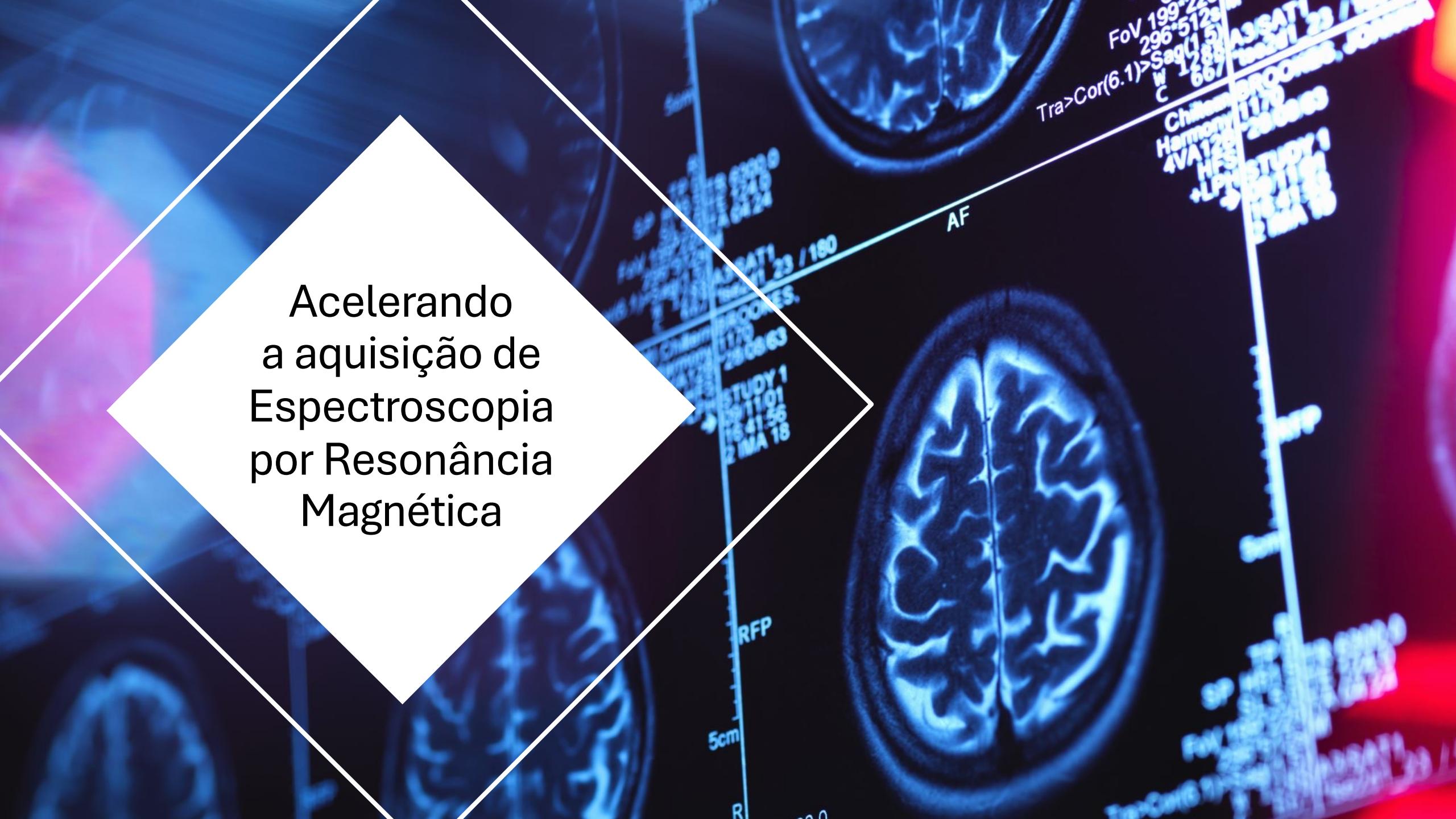


# Diagnóstico

Posso reduzir o tempo (e custo) de aquisição usando DL?



Reconstrução ou Denoising



# Acelerando a aquisição de Espectroscopia por Resonância Magnética

FoV 199°ZE  
296°512SI  
Tra>Cor(6.1)>Sag(1.5)  
W 128x128  
C 667  
Chilman BROOKES.  
Harmon 1140  
AVA128  
HES  
JPT STUDY 1  
JPT STUDY 2  
JPT STUDY 3  
JPT STUDY 4

AF

RFP

5cm

R

# Ressonância Magnética: Imagem vs. Espectroscopia

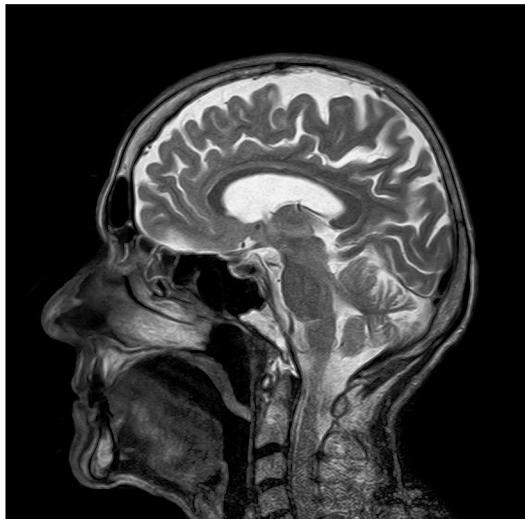
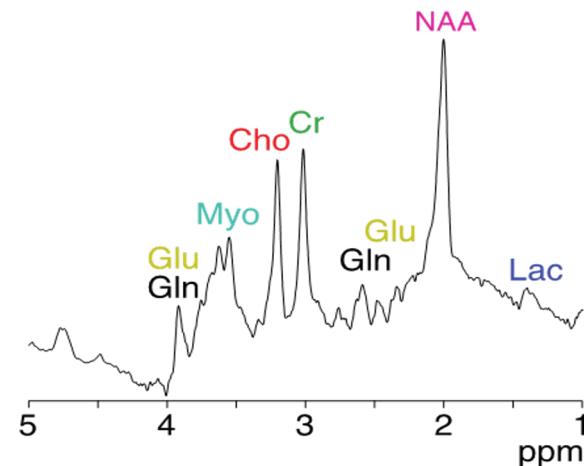


Imagen de RM  
Resonância das moléculas de água



Espectroscopia por RM  
Resonância dos metabólitos

**MRI:** Imagem 2D ou 3D

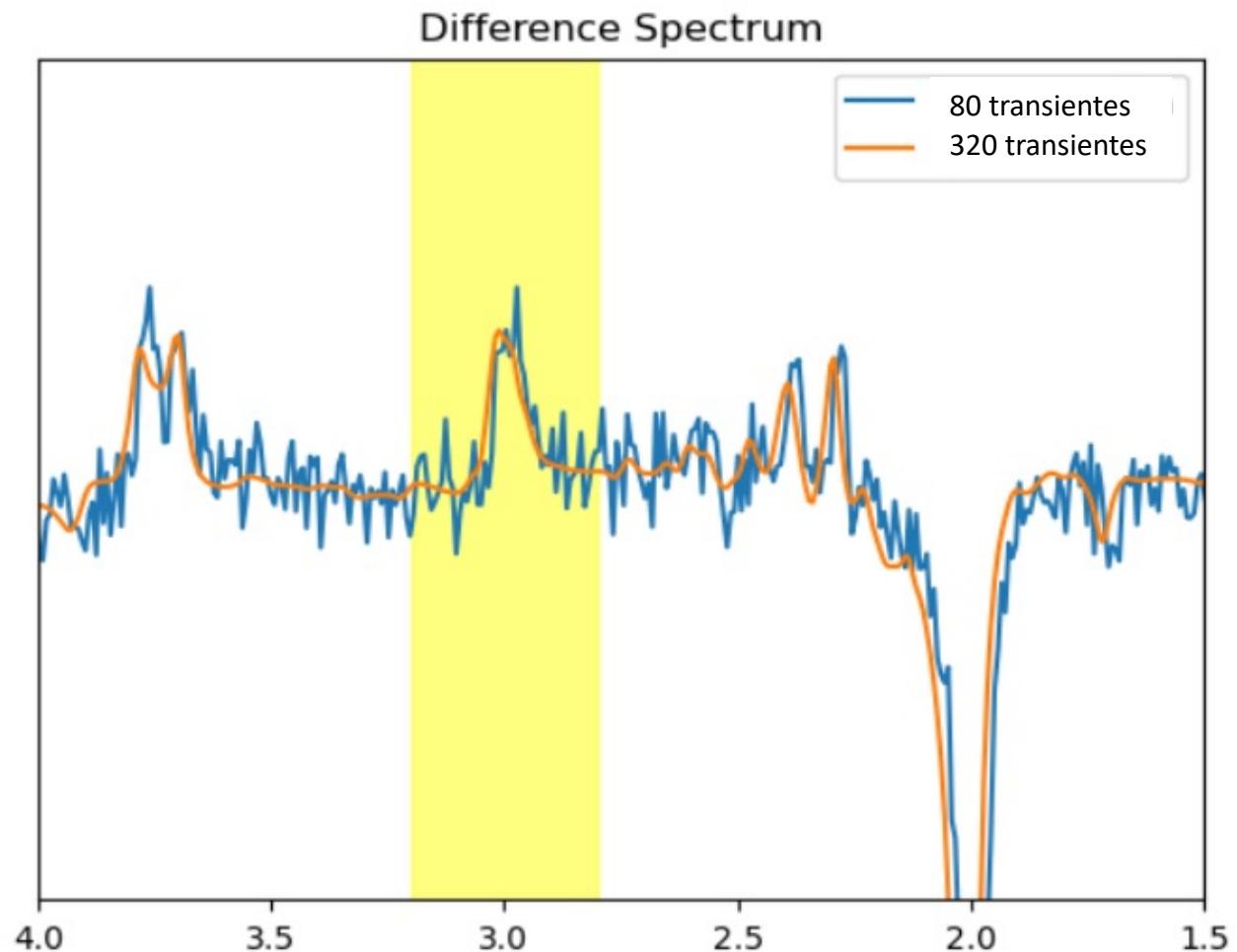
**MRS:** espectro químico

**Vantagens:** MRS is a non-invasive method for metabolites analysis of healthy and pathological brain.

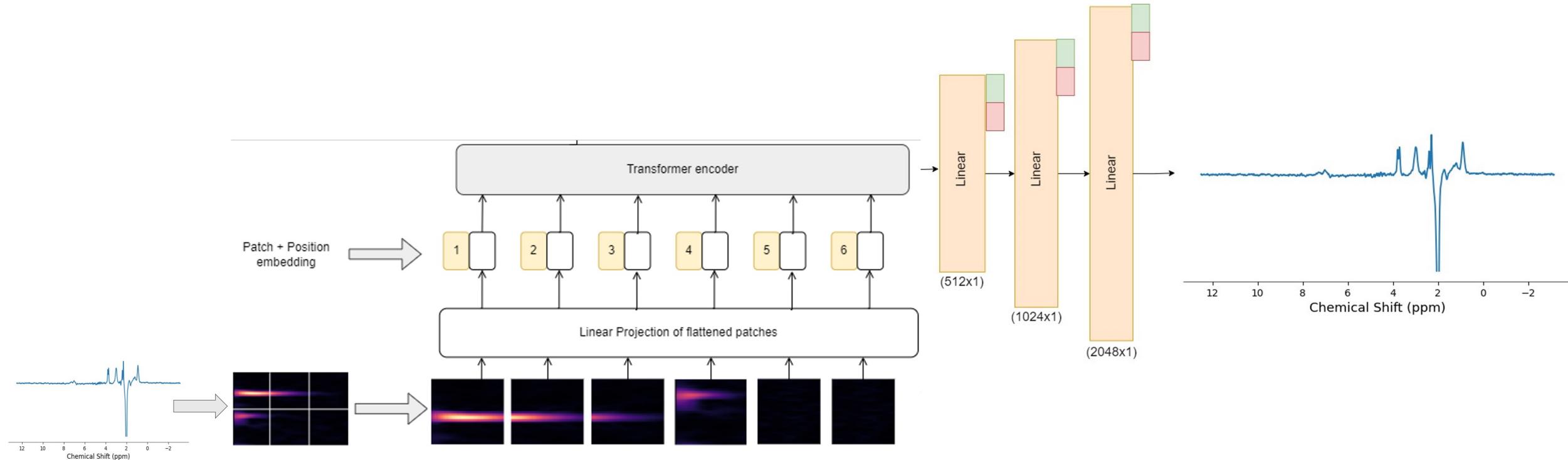
**Challenges:** acquisition time is at least 10-12 minutes (too long, susceptible to motion artifacts).

# Reconstruindo dados de MRS

É possível melhorar a qualidade do espectro final, adquirindo 4 vezes menos dados (transientes)?

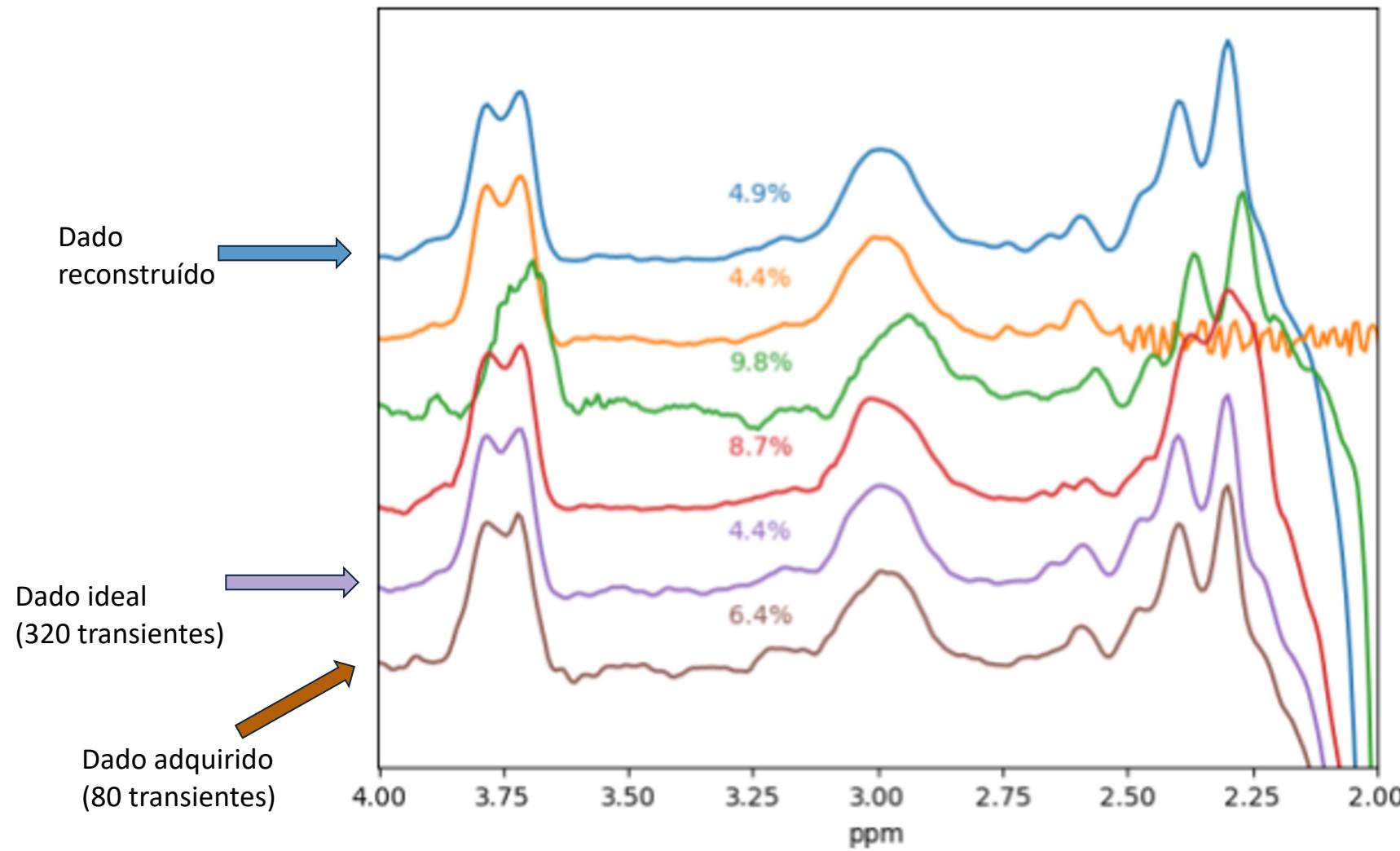


# Arquitetura - Spectro-ViT



<https://github.com/MICLab-Unicamp/Spectro-ViT>

Dias et al., *Magnetic Resonance imaging*, p. 110219 (2024)

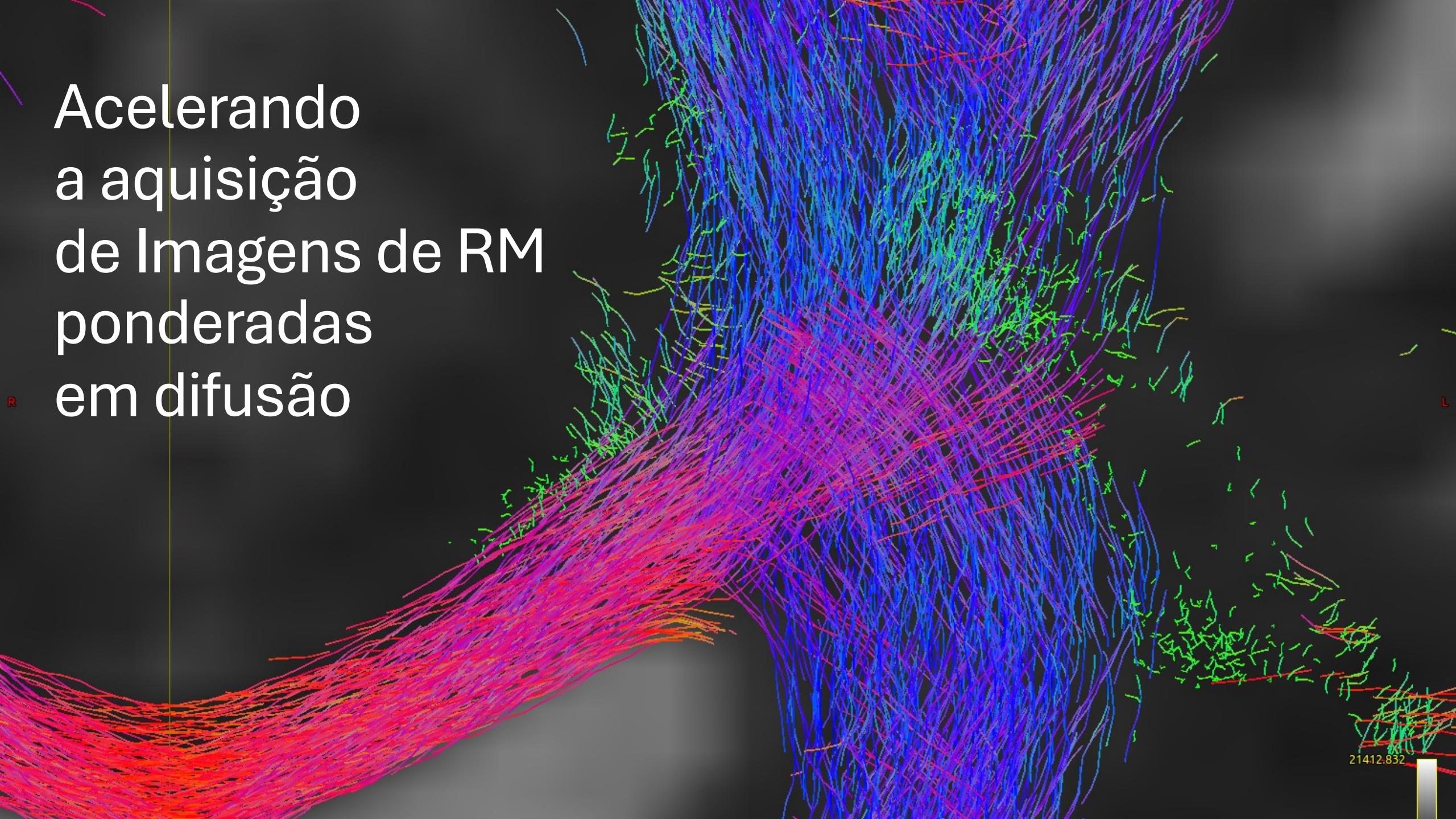


Resultado da  
reconstrução

↓

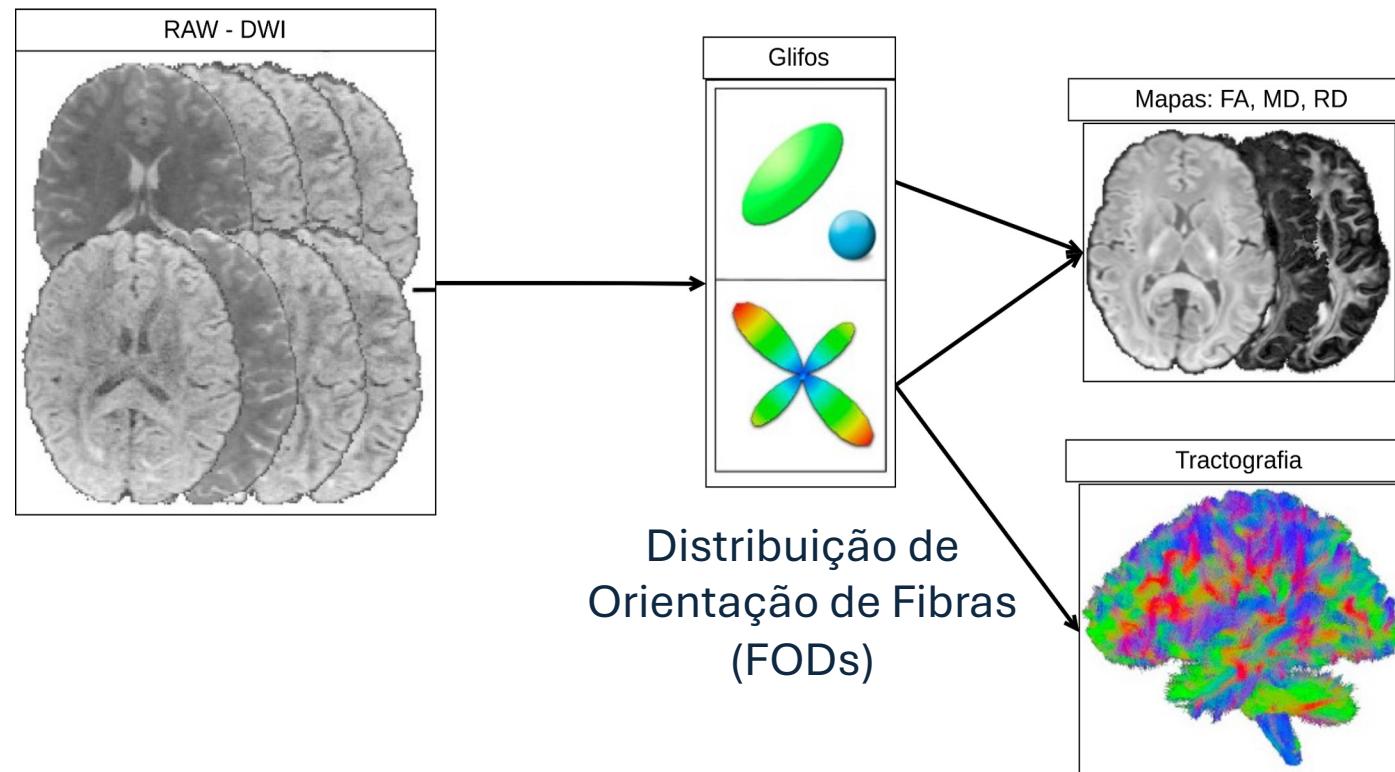
permite uma  
aceleração  
de 4 vezes

# Acelerando a aquisição de Imagens de RM ponderadas em difusão



# Imagen de Difusão por Ressonância Magnética (dMRI)

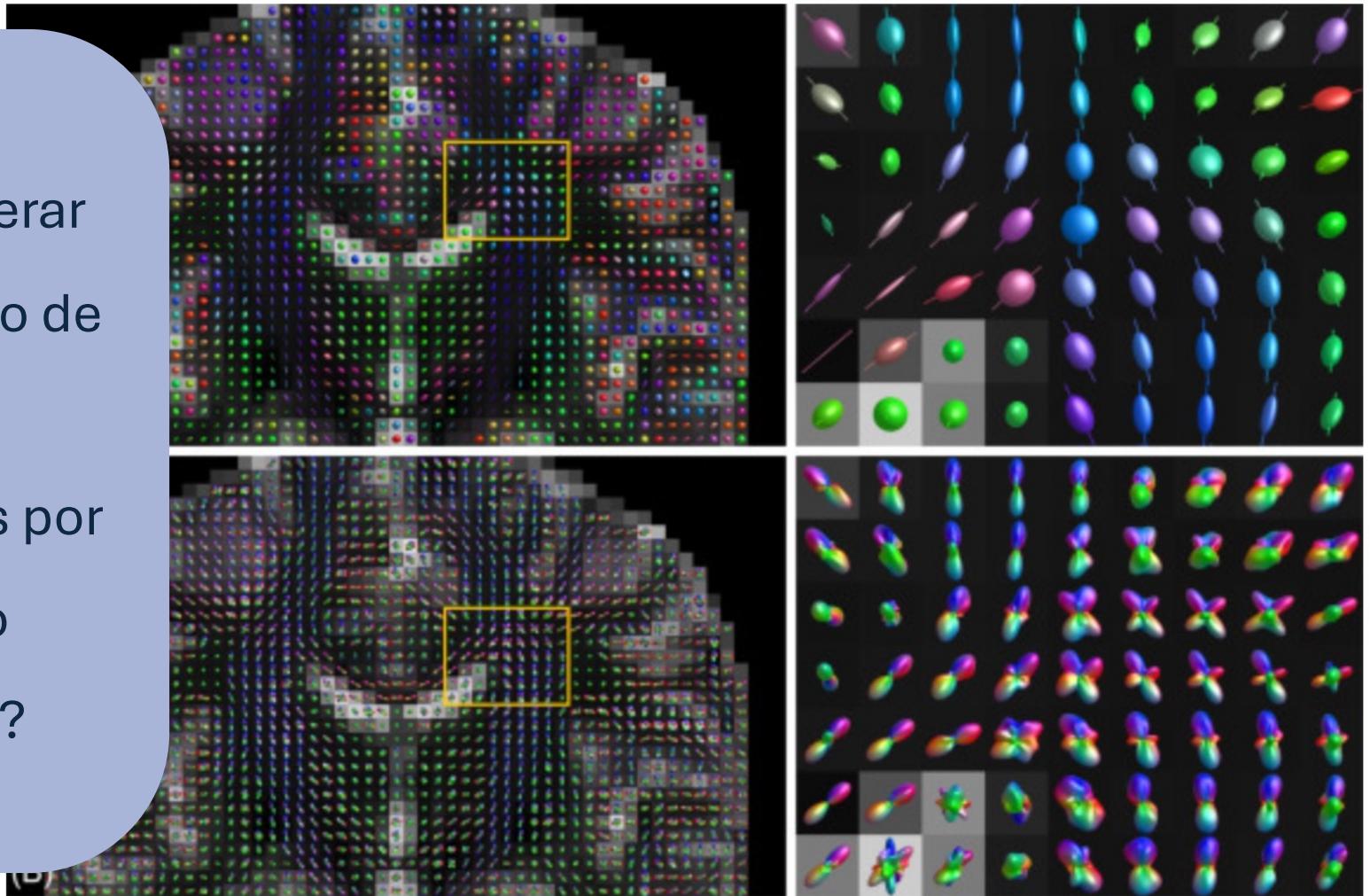
- Mede o movimento das moléculas de água nos tecidos
- Usada para investigar a microestrutura do tecido cerebral
- Exemplos:
  - Planejamento cirúrgico
  - Monitoramento da integridade das fibras neurais



## LARDI Aquisição de 6 minutos

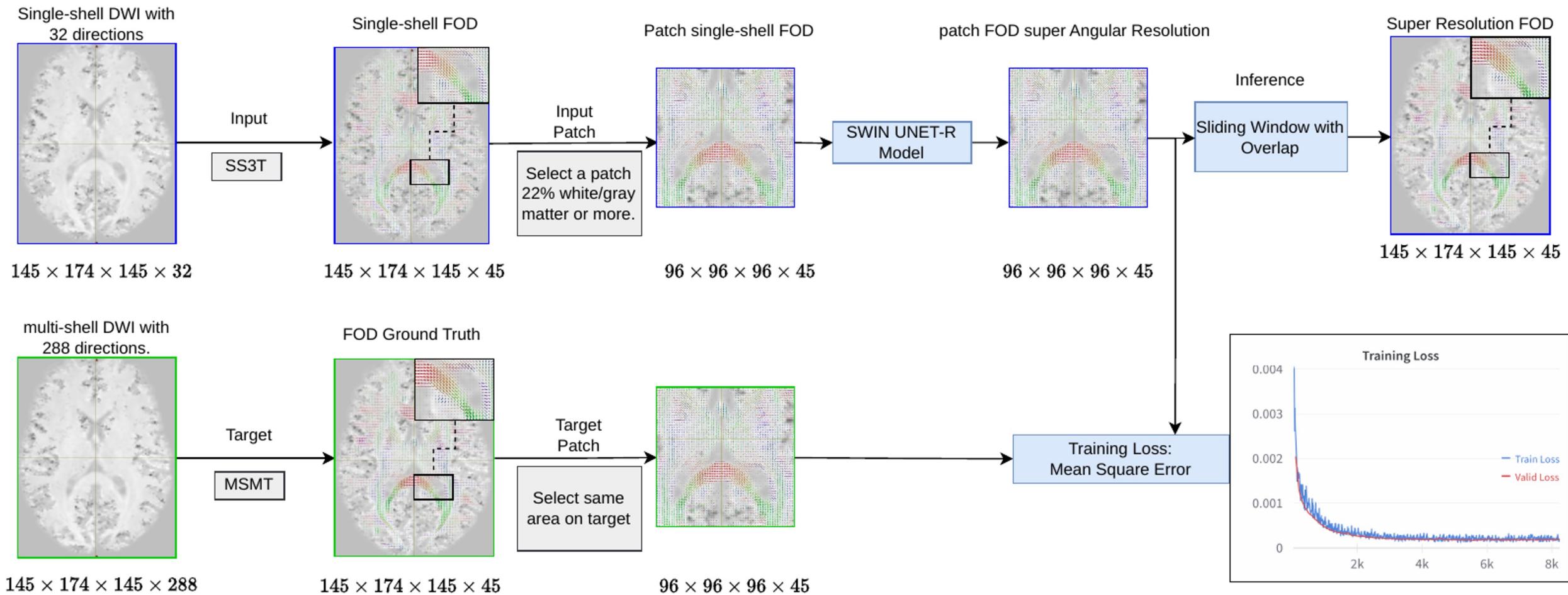
É possível usar DL para gerar Distribuição de Orientação de Fibras (FODs)

equivalente às produzidas por dados **HARDI**, usando somente dados **LARDI**?

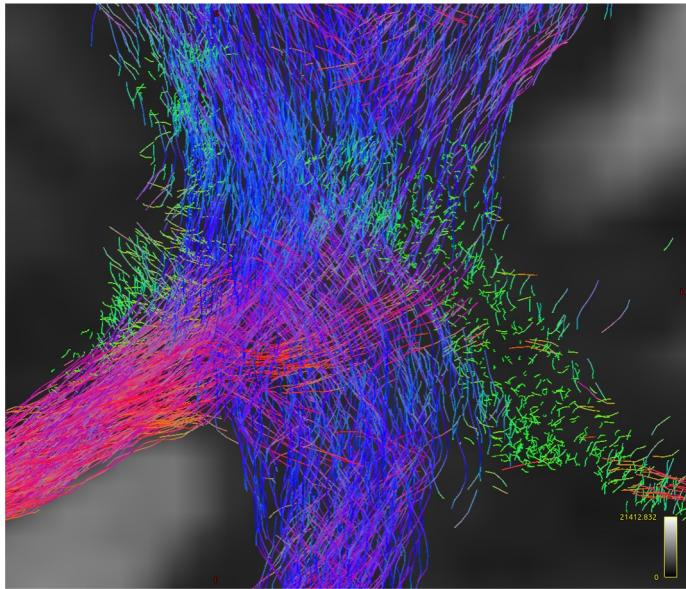
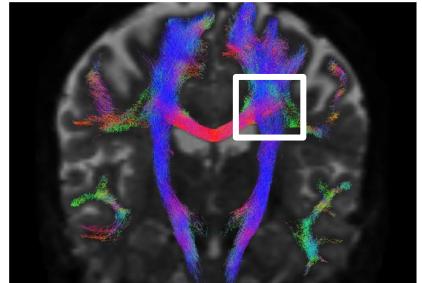


## HARDI Aquisição de 38 minutos

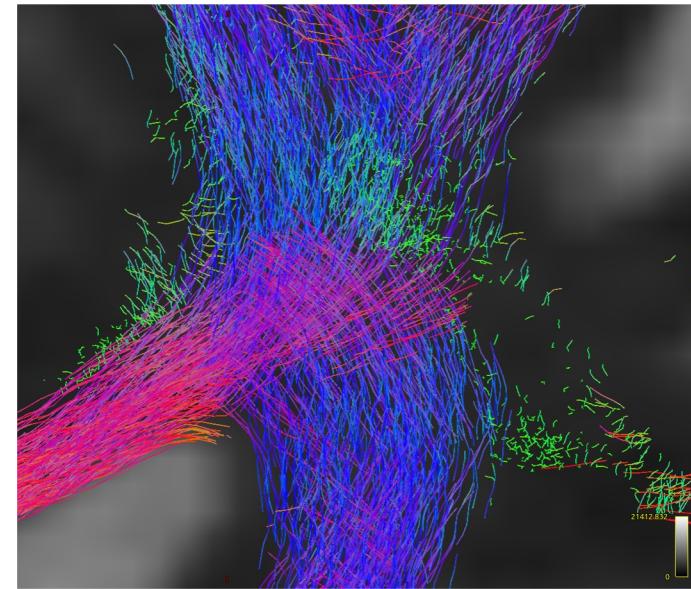
# FOD-Swin-Net: o modelo proposto



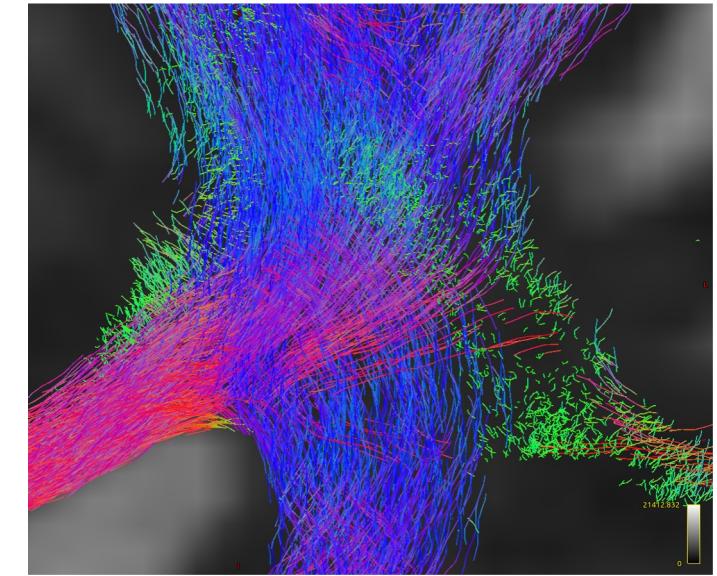
# Impacto na tractografia



LARDI (6 min.)



RECONSTRUÍDO



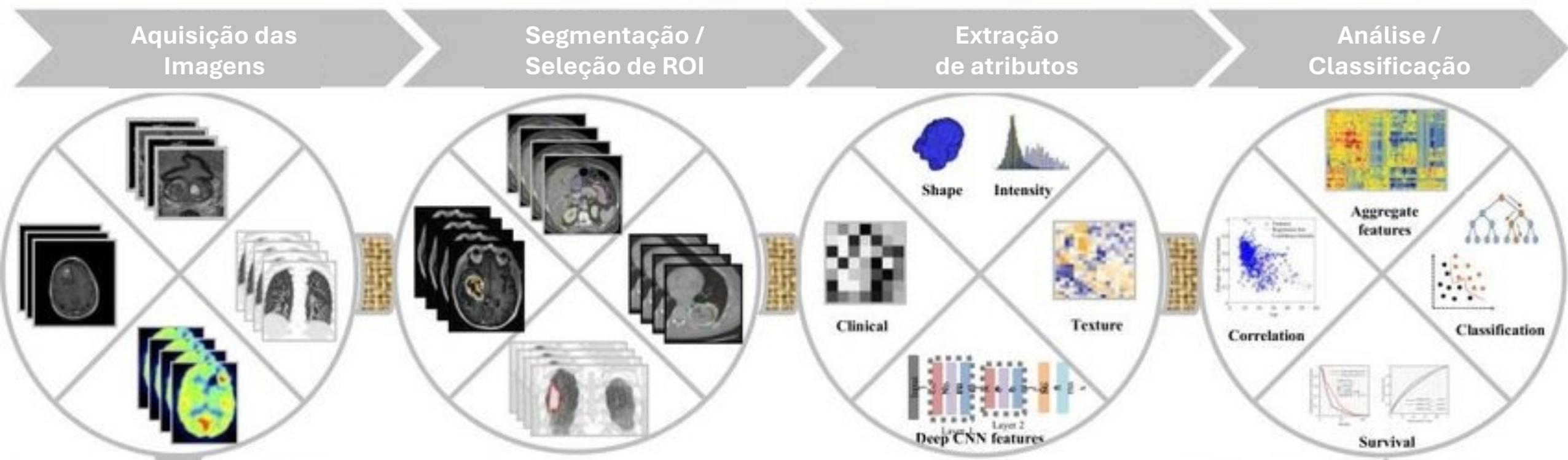
HARDI (38 min.)



<https://github.com/MICLab-Unicamp/FOD-Swin-Net>

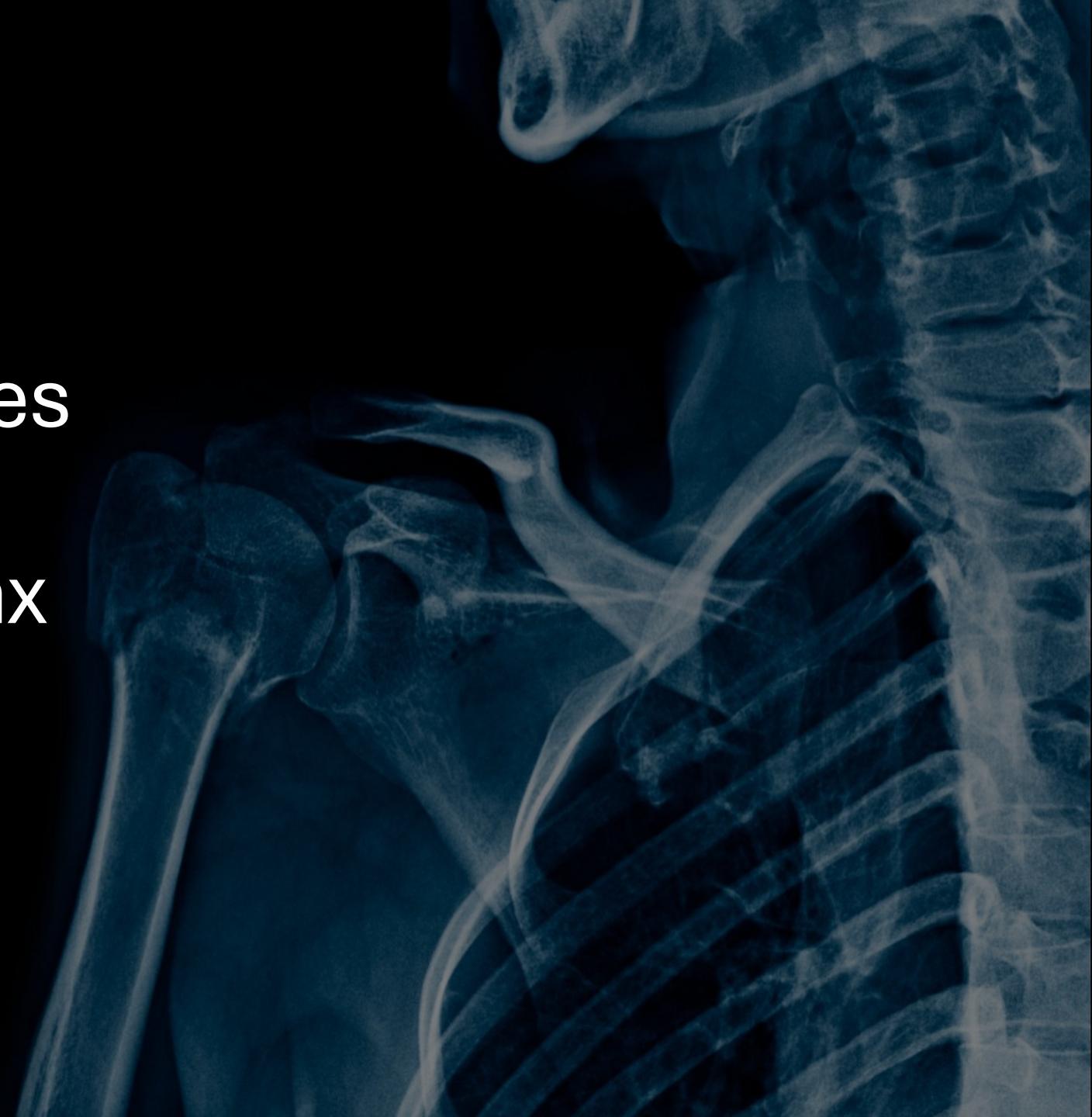
Oliveira M et al., *Proceedings of ISBI* (2024)

Posso pre-processor os dados  
usando DL?



## Seleção de Região de Interesse (ROI)

# Identificando regiões de interesse em Tomografia de tórax



# Estruturas pulmonares em CT

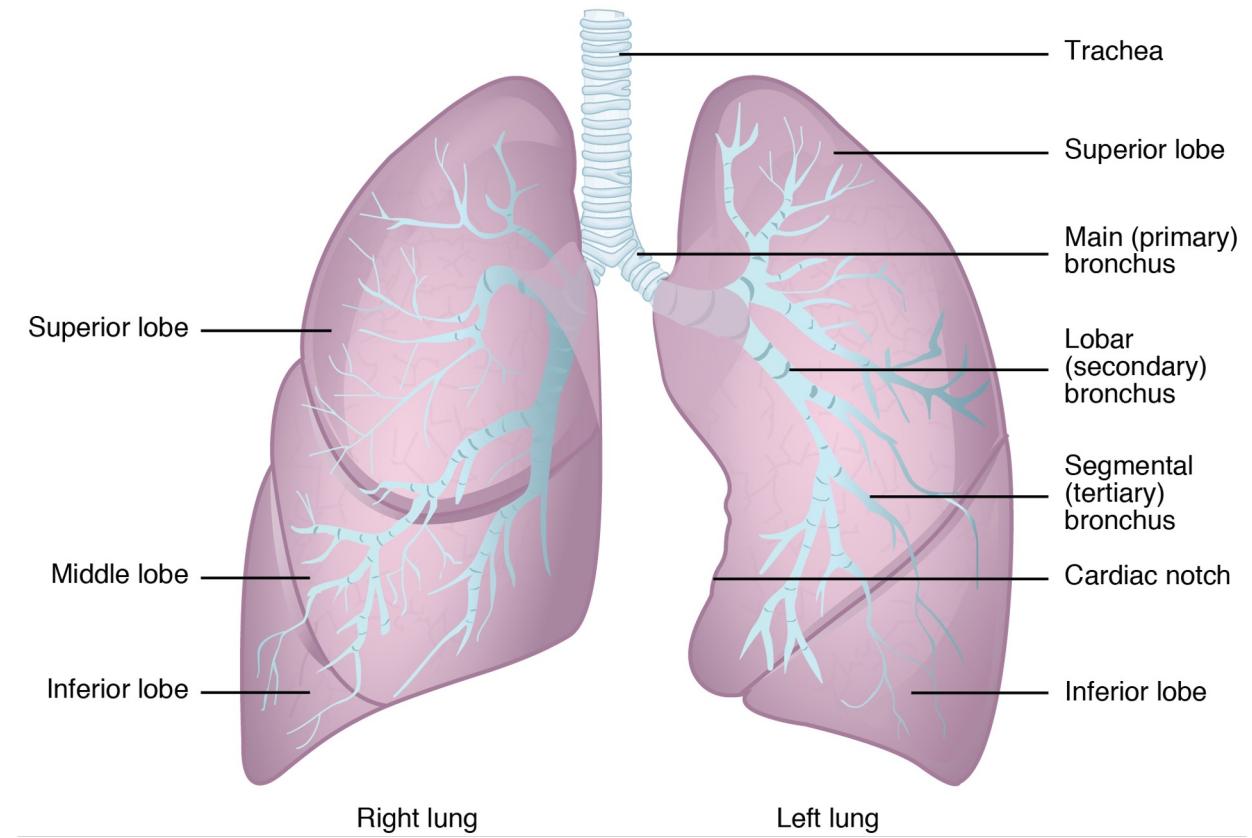
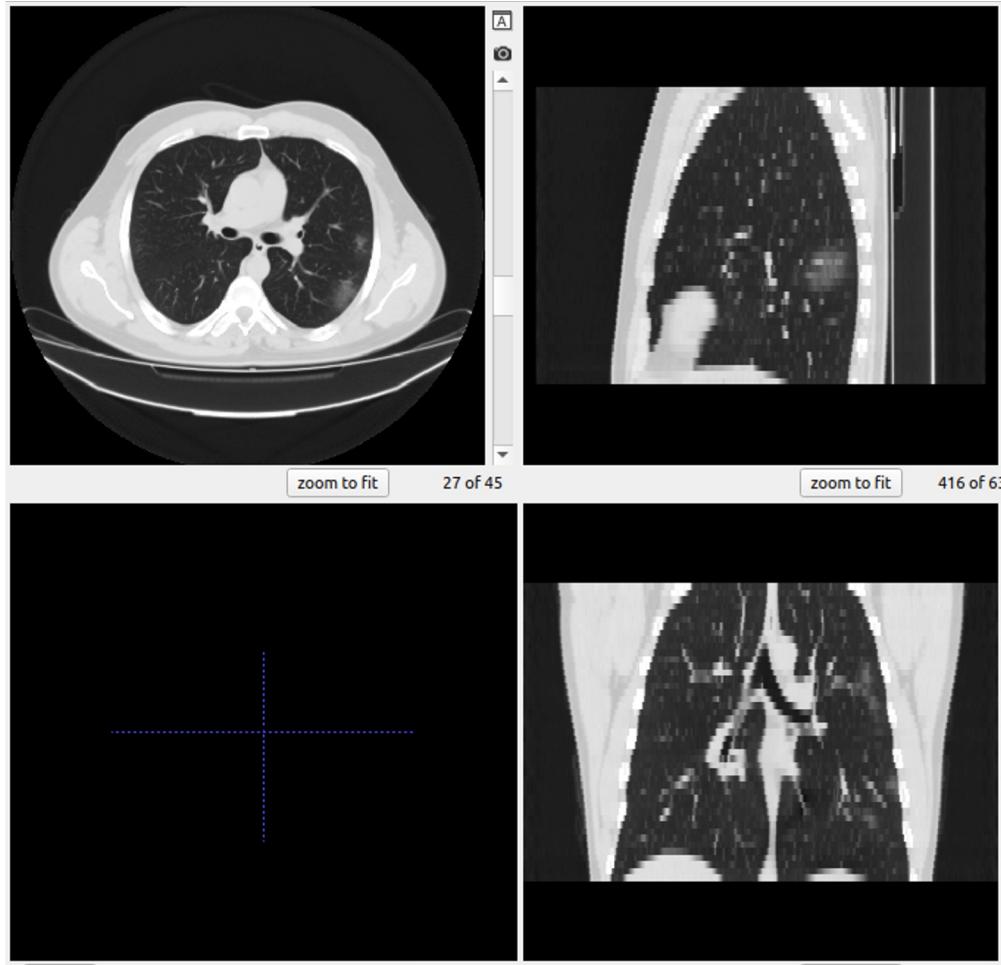
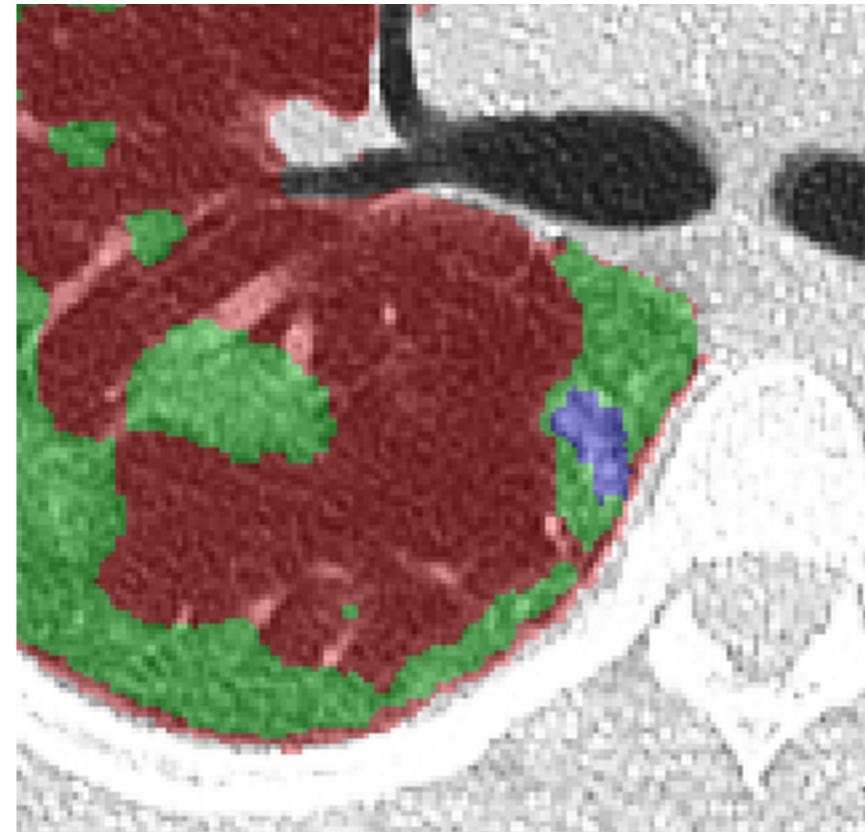
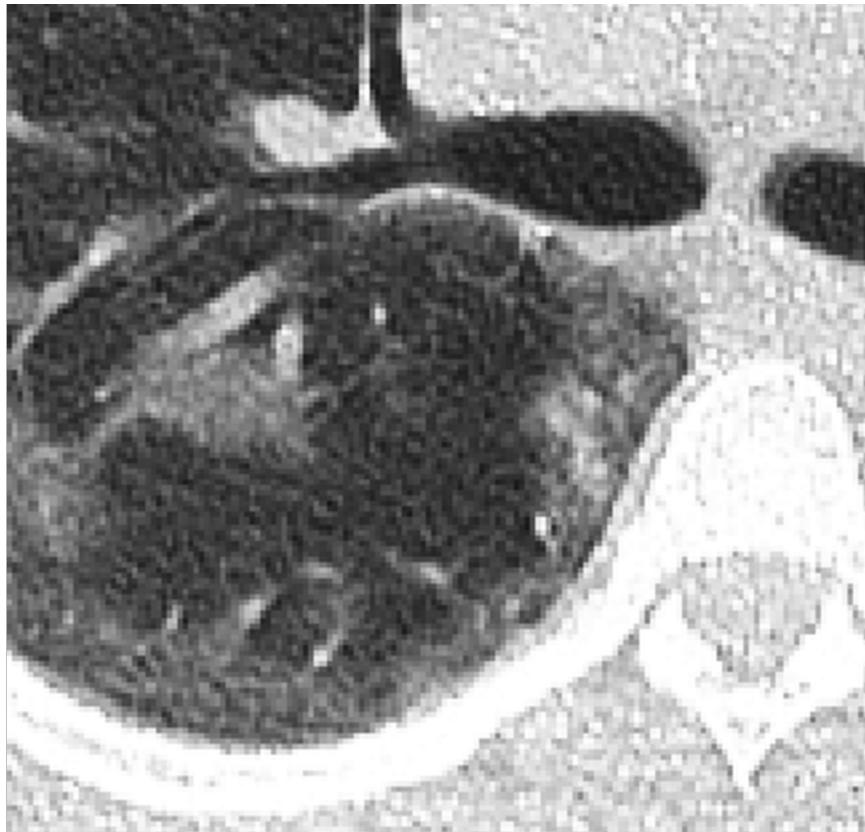


Illustration of the lungs and airways [1].

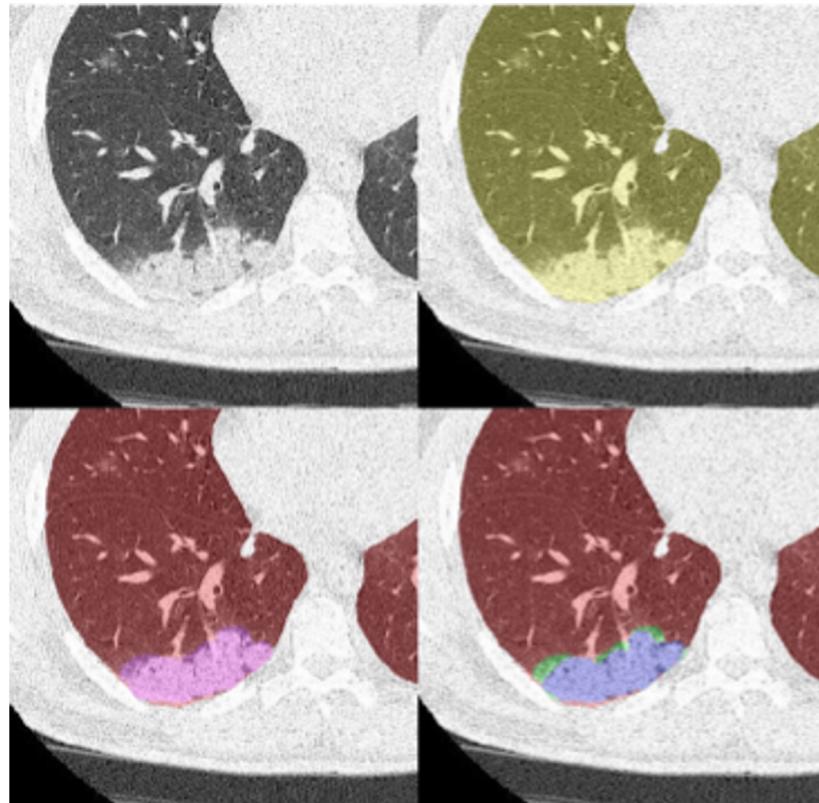
# Achados em pulmão (tecido patológico)

Algumas doenças que afetam o pulmão podem gerar:  
**efeito de vidro fosco (ground-glass opacity - GGO) e consolidações (consolidation).**

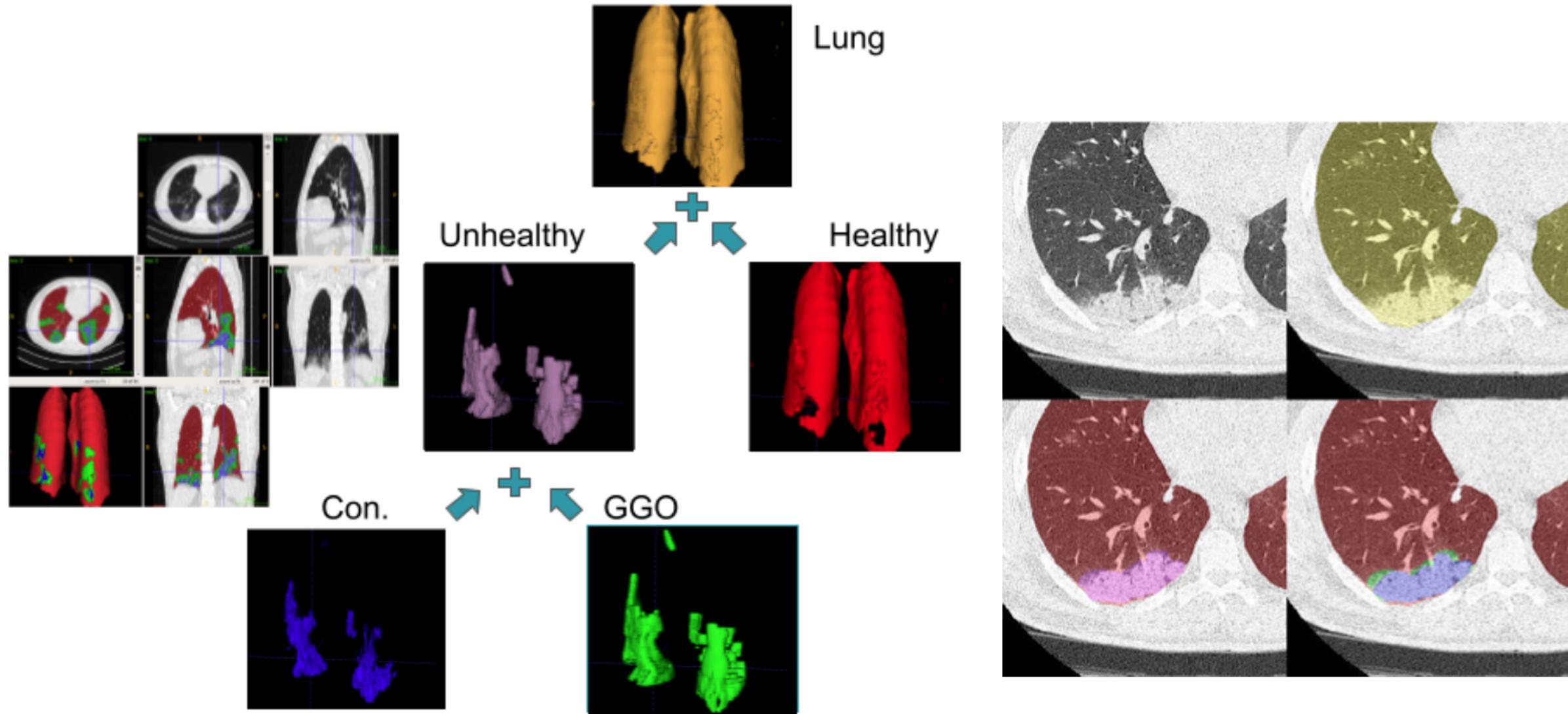


**Problema:** falta de dados anotados (rótulos)!

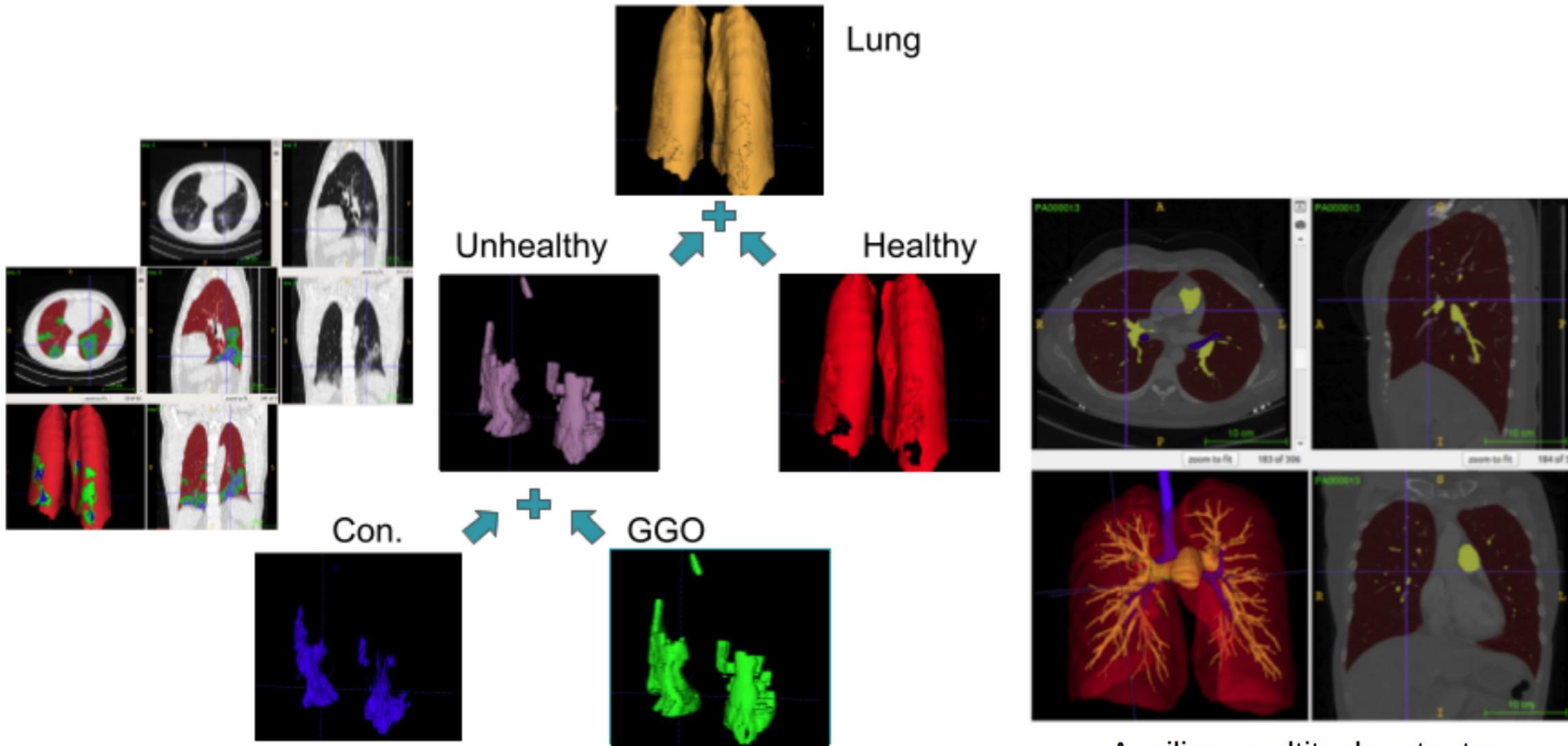
É possível treinar um modelo DL capaz de segmentar diversas regiões de interesse ao mesmo tempo: pulmão, vidro fosco, consolidação, vias aéres, e artérias? Mesmo que meus datasets não contenham todos os rótulos?



# Polymorphic Multitask Learning (PML)



# Polymorphic Multitask Learning (PML)

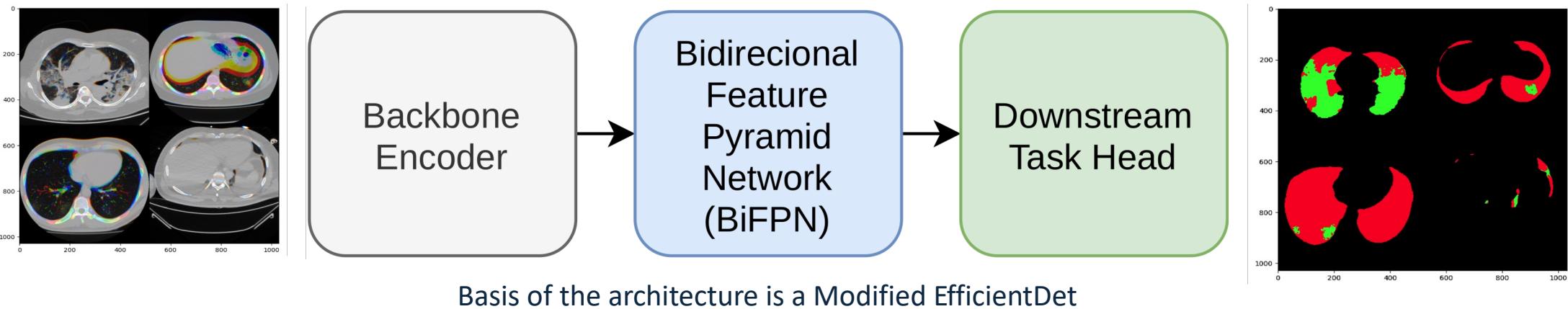


Network output: BG, **Healthy**, GGO, Consolidations

Auxiliary multitask outputs:  
airway, pulmonary artery

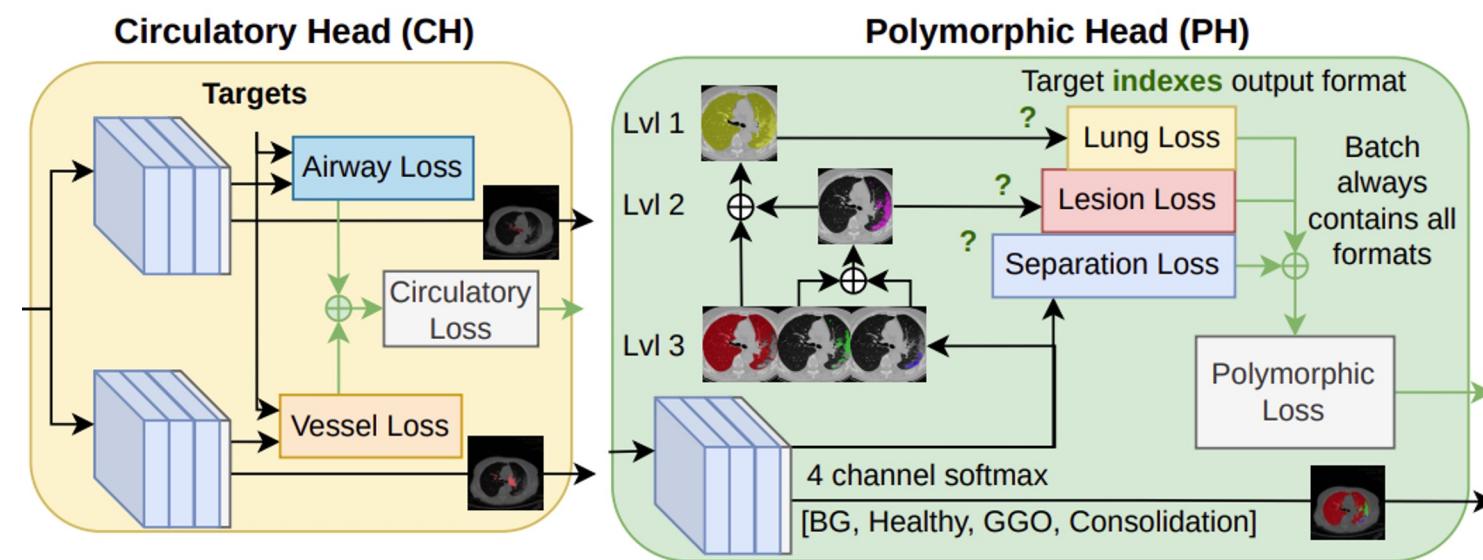
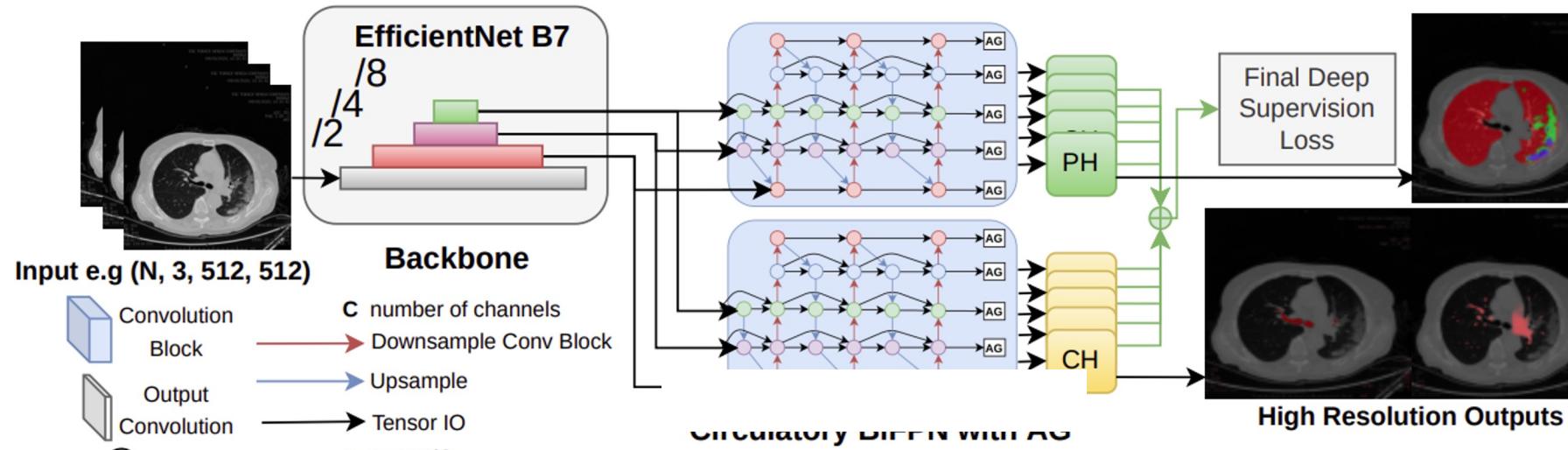
# MEDPSeg

## Modified EfficientDet for Polymorphic Pulmonary Segmentation

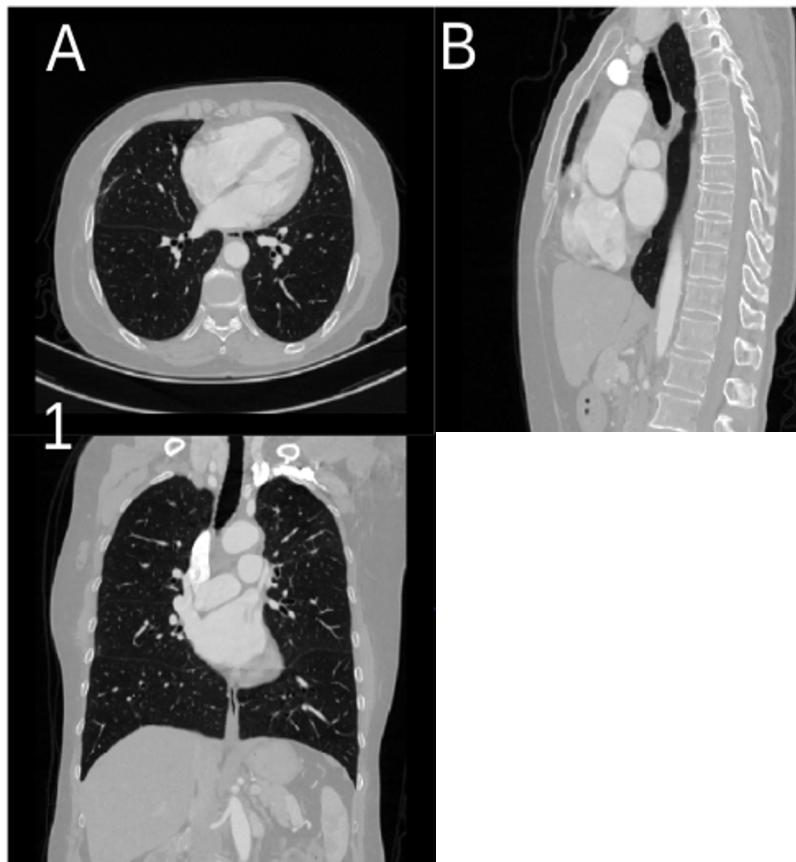


# MEDPSeg

## Modified EfficientDet for Polymorphic Pulmonary Segmentation

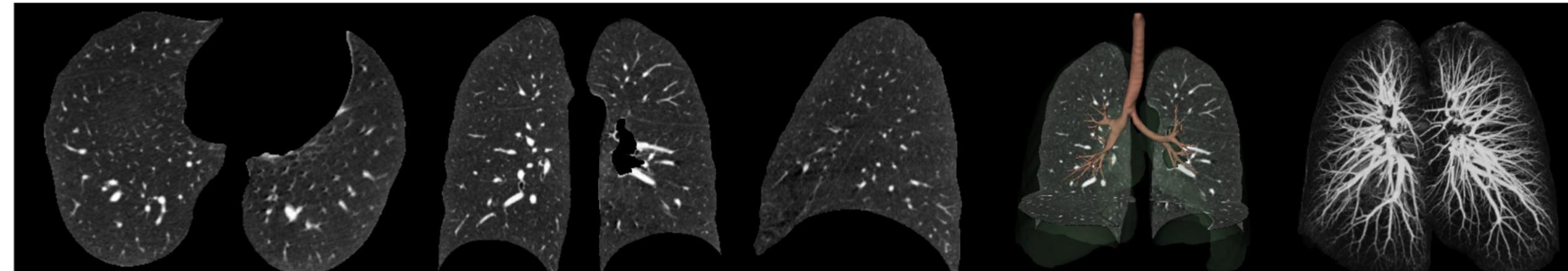


# Visualizando os resultados

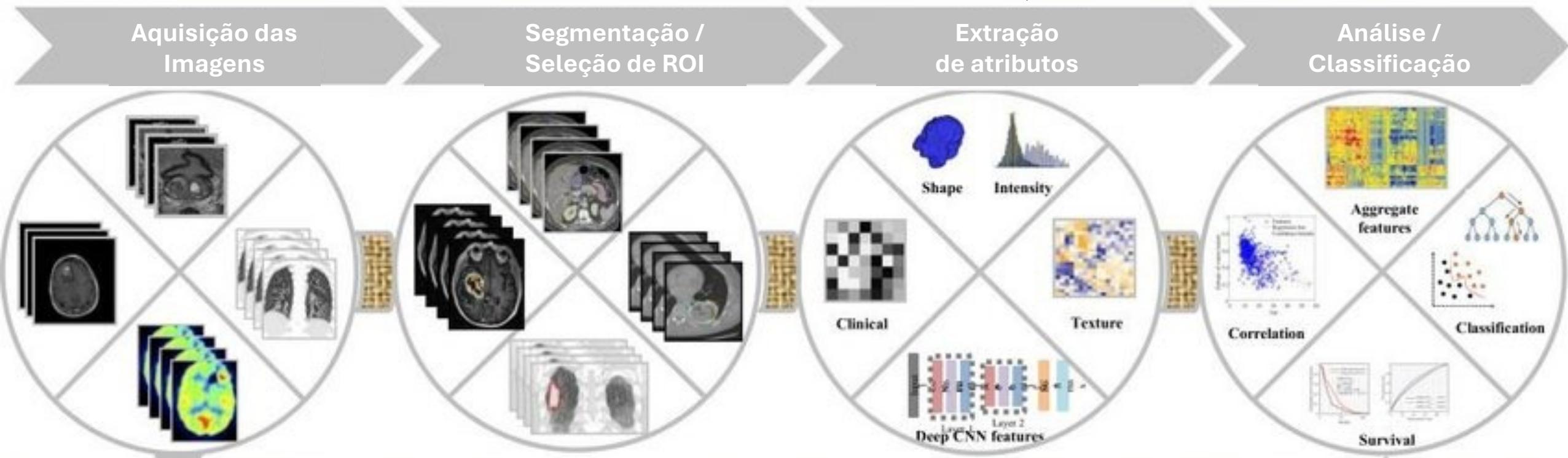


# Quantificando os resultados

0% POI



É possível extrair informações/  
características usando DL?



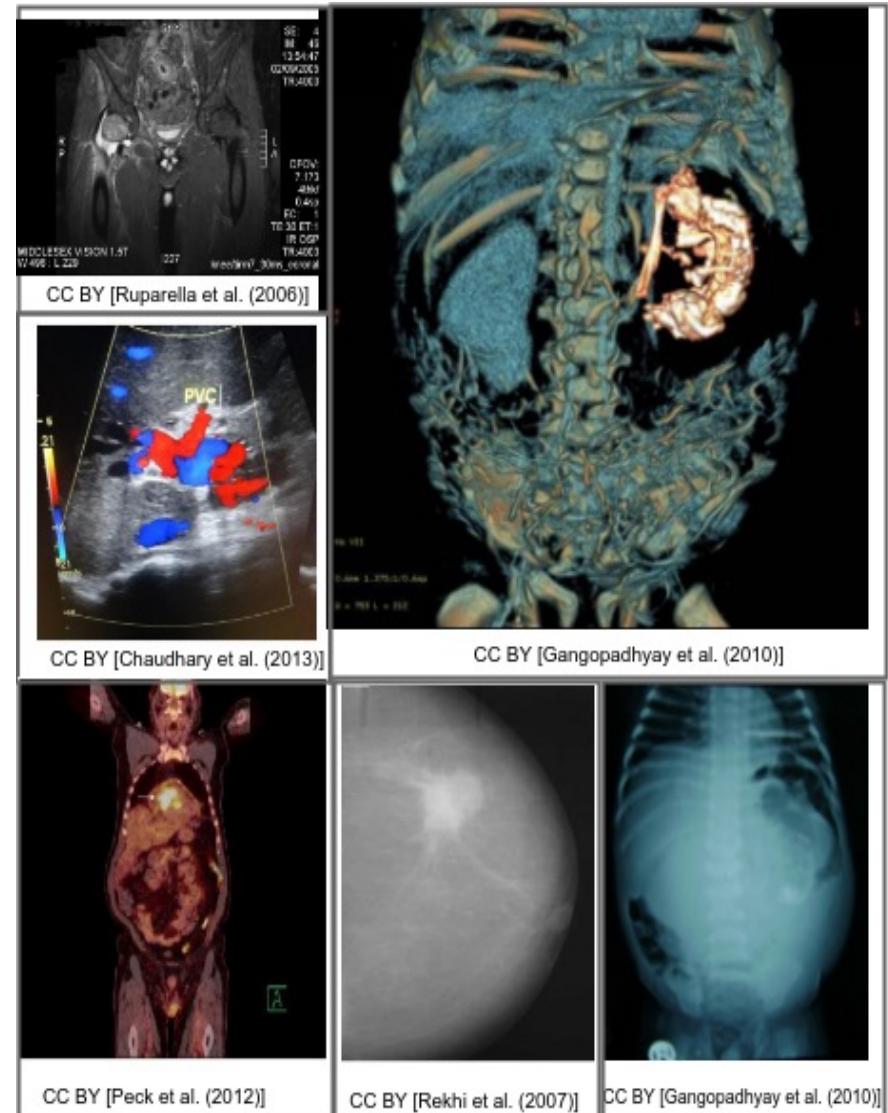
## Extração de características

A close-up photograph of a stethoscope lying on a light-colored wooden surface. The stethoscope has a clear, ribbed chest piece and a white tube. A dark, semi-transparent rectangular overlay covers the top half of the image, containing the title text.

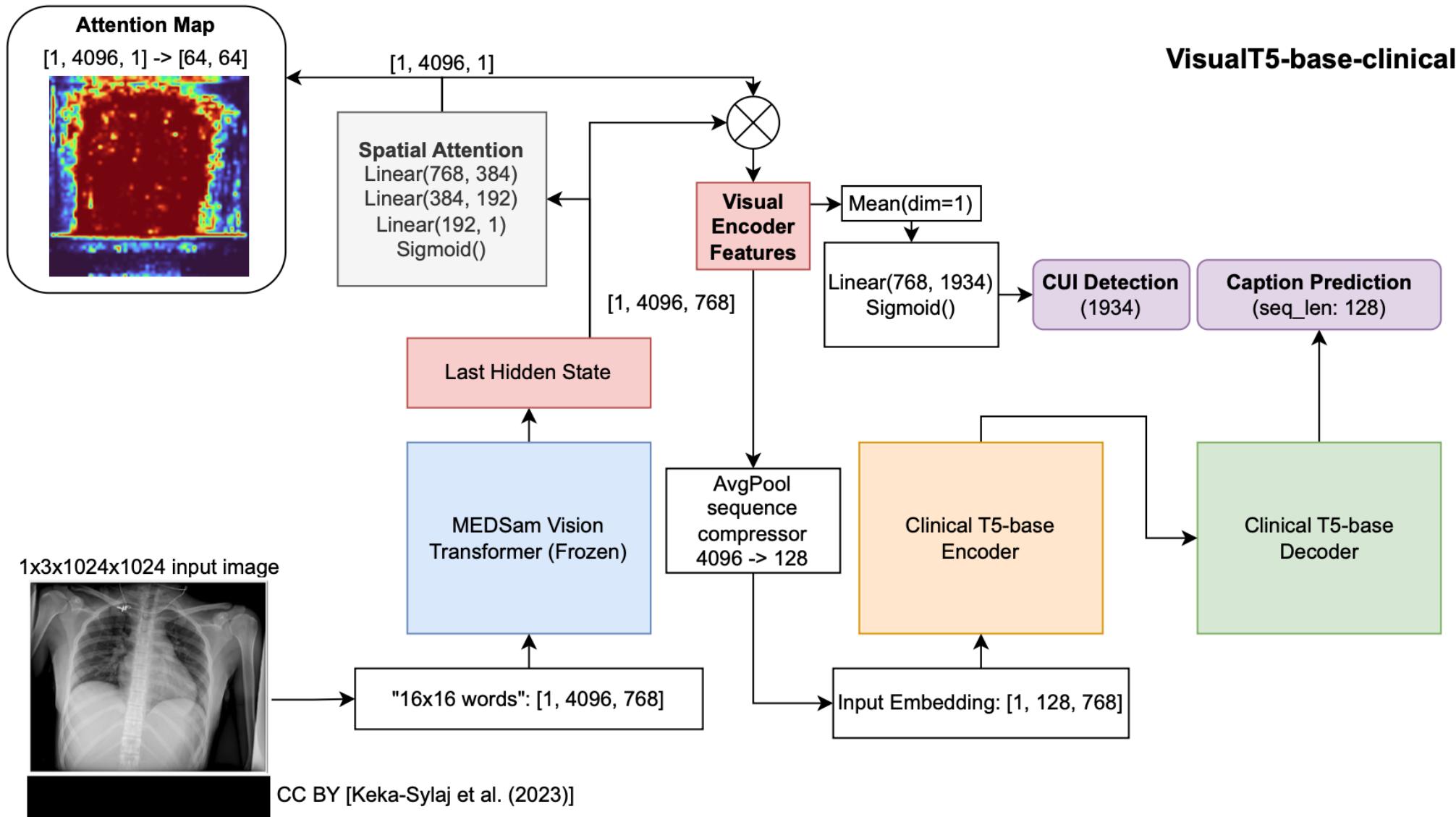
# Gerando legendas para imagens médicas

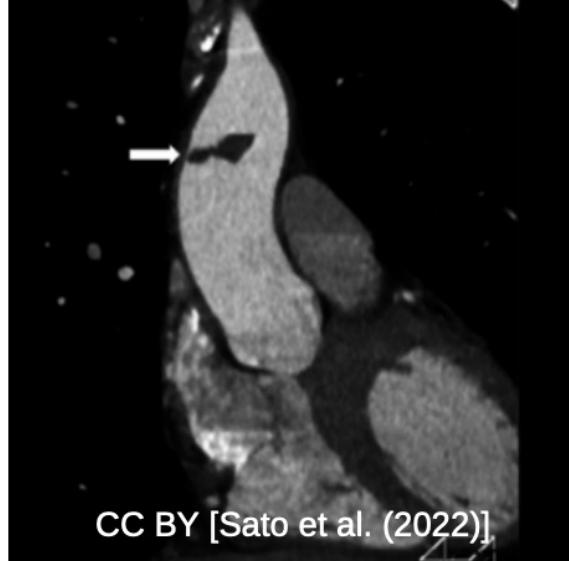
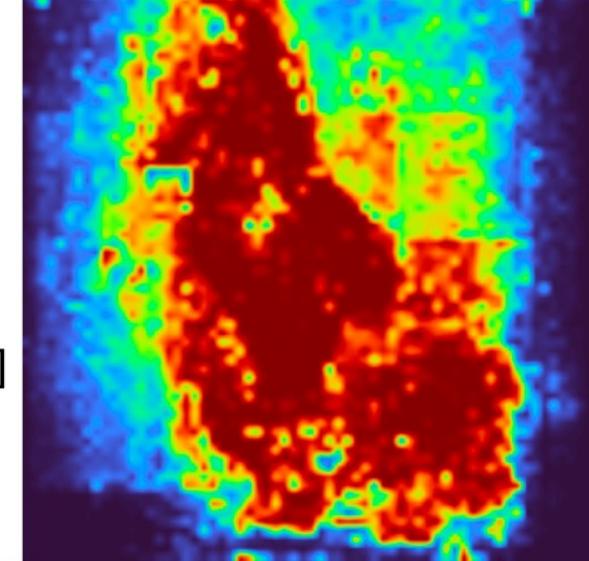
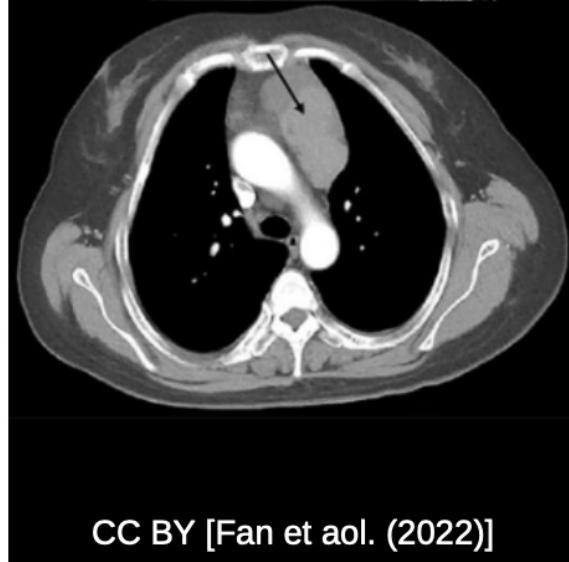
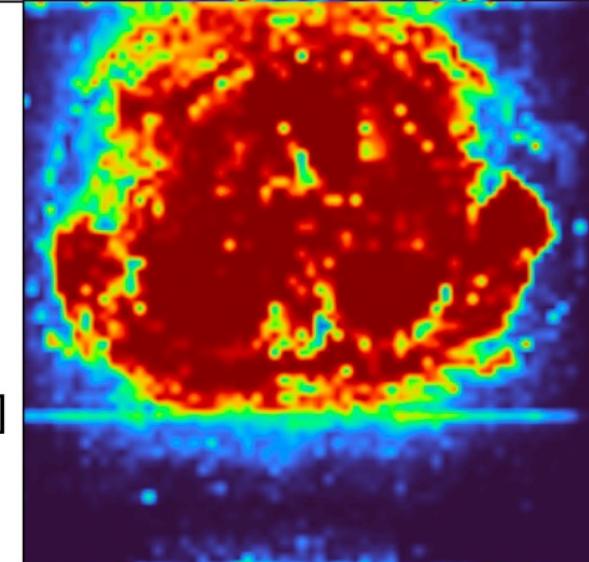
# Legenda para imagens (captioning)

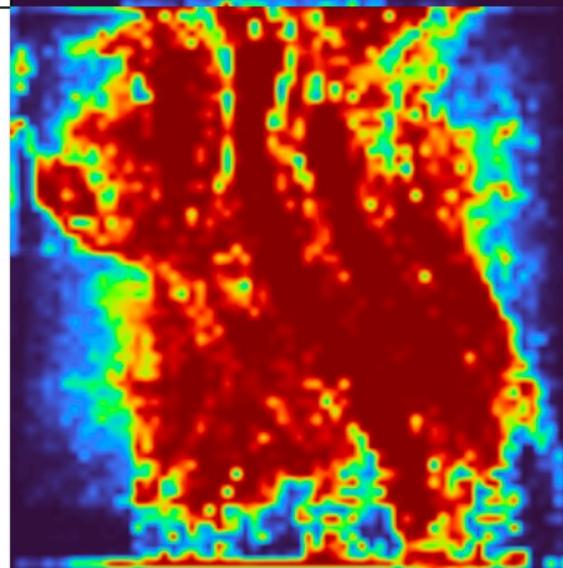
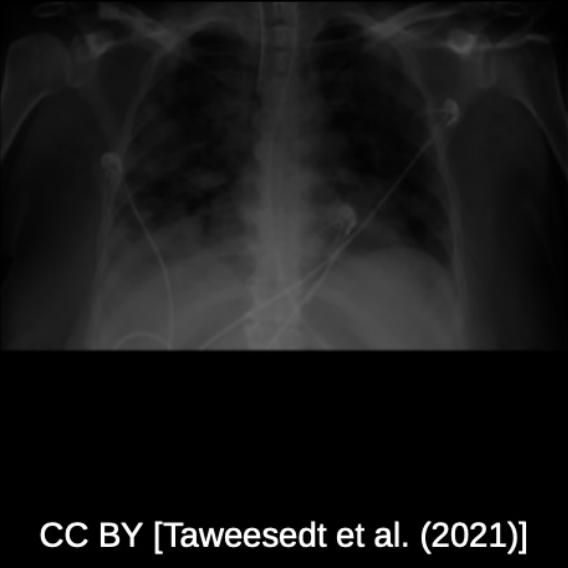
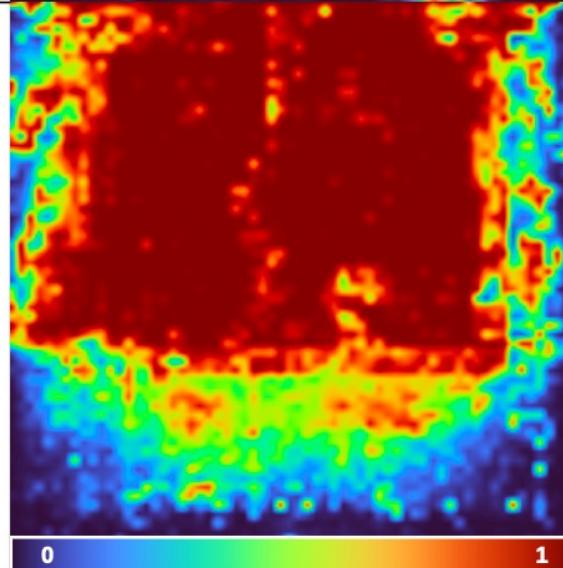
É possível treinar um modelo DL capaz de, ao mesmo tempo, classificar imagens médicas e também gerar uma legenda que as descreva?



# Modelo proposto



Input	Target	Prediction	Spatial Attention
 CC BY [Sato et al. (2022)]	<p>Computed tomography (CT) shows floating thrombosis (white arrow)</p> <p>Concept: ['C0040405', 'C0040053'], ['X-Ray Computed Tomography', 'Thrombosis']</p>	<p>Aortic root CT angiogram showing aortic root dilation.</p> <p>Concept: ['C0040405'], ['X-Ray Computed Tomography']</p>	
 CC BY [Fan et al. (2022)]	<p>Enhanced CT scan of the chest revealed an anterior mediastinal tumor (black arrow).</p> <p>Concept: ['C0025066', 'C0027651', 'C0040405'], ['Mediastinum', 'Neoplasms', 'X-Ray Computed Tomography']</p>	<p>Axial CT scan of the chest showing a large mass in the anterior mediastinum.</p> <p>Concept: ['C0040405'], ['X-Ray Computed Tomography']</p>	

Input	Target	Prediction	Spatial Attention
 CC BY [Trowbridge et al. (2022)]	<p>Early sagittal T2-weighted MRI.</p> <p>Concept: ['C0024485'], ['Magnetic Resonance Imaging']</p>	<p>Sagittal T2-weighted MRI of the cervical spine showing a hyperintense signal in the spinal cord at the C3-C4 level (arrow).</p> <p>Concept: ['C0024485'], ['Magnetic Resonance Imaging']</p>	
 CC BY [Taweesedt et al. (2021)]	<p>The typical chest X-ray finding of a patient with coronavirus disease 2019 infection showing bilateral infiltration.</p> <p>Concept: ['C0332448', 'C1306645', 'C0817096', 'C0009450'], ['Infiltration', 'Plain x-ray', 'Chest', 'Communicable Diseases']</p>	<p>Chest X-ray showing bilateral infiltrates</p> <p>Concept: ['C1306645;C0817096'], ['Plain x-ray;Chest']</p>	 <div style="display: flex; justify-content: space-around; width: 100%;"> <span>0</span> <span>1</span> </div>

# Profa. Letícia Rittner

- Professora Associada – Faculdade de Engenharia Elétrica e de Computação (FEEC)
- Diretora do Medical Image Computing Lab (MICLab)
- Pesquisadora Associada – Brazilian Institute of Neuroscience and Neurotechnology (BRAINN)



# Dr. Diedre do Carmo

- Pesquisador - Neuralmind
- Pesquisador Colaborador - Faculdade de Engenharia Elétrica e de Computação (FEEC)





# Medical Image Computing Lab - MICLab Unicamp

The lab was created under the leadership of Roberto Lotufo and Letícia Rittner to be a focal point for the majority of the students under their supervision

20 followers University of Campinas <https://miclab.fee.unicamp.br>

Overview

Repositories 23

Projects

Packages

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## Popular repositories

### e2dhipseg

Public

This contains official implementation for Extended 2D Volumetric Consensus Hippocampus Segmentation

Python 20 6

### ia636

Public

Python Toolbox for Teaching Image Processing

Python 14 38

### medpseg

Public

MEDPSeg: Official implementation of Modified EfficientDet for Pulmonary Polymorphic Segmentation

Python 9 4

### CONSNet

Public

CONSNet: Convolutional Neural Networks for Skull-stripping in Brain MR Imaging using Consensus-based Silver Standard Masks

Python 17 4

### medseg

Public

Official repository for reproducing COVID and Lung segmentation prediction (old version of MEDPSeg)

Python 10 3

### inCCsight

Public

Python/Dash/Plotly implementation of a data exploration platform for corpus callosum data in DTI

Python 5 2

## People



## Top languages

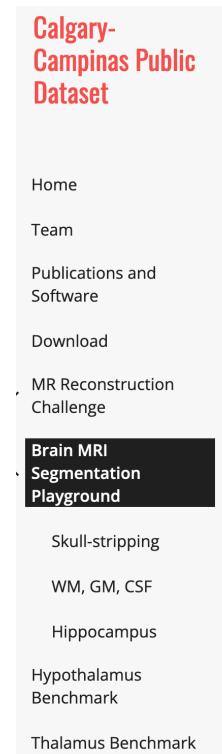
Python Jupyter Notebook

# Public datasets

Se

## Benchmarks for Medical Image Segmentation

- Hypothalamus segmentation
  - <https://codalab.lisn.upsaclay.fr/competitions/7583>
- Thalamus segmentation on dMRI
  - <https://codalab.lisn.upsaclay.fr/competitions/8329>  
(preliminary link)
- Calgary-Campinas Dataset
  - Brain MRI segmentation playground
  - Skull-striping
  - Hippocampus



CodaLab

The screenshot shows the CodaLab competition page for 'Hypothalamus Segmentation'. It features three grayscale brain MRI slices with cyan-colored segmentation masks overlaid. To the right, a summary table provides details about the competition: it is organized by 'liviamarodrigues', the current server time is 'April 15, 2023, 5:07 p.m. UTC', and the status is 'Current'. The first phase ended on 'Sept. 20, 2022, midnight UTC', while the competition ends 'Never'. A button labeled 'Competition Ends' is also visible.

Competition	
	Hypothalamus Segmentation
Organized by liviamarodrigues - Current server time: April 15, 2023, 5:07 p.m. UTC	
<b>▶ Current</b>	End
First phase	Competition Ends
Sept. 20, 2022, midnight UTC	Never

## BRAIN MRI SEGMENTATION PLAYGROUND

### Overview

Accurate brain segmentation is critical for many magnetic resonance imaging (MRI) analysis pipelines. Machine-learning-based brain MR image segmentation methods are among the state-of-the-art techniques for this task. Nevertheless, the segmentations produced by machine learning models often degrade in the presence of expected domain shifts between the test and train sets data distributions. These domain shifts are expected due to several factors, such as scanner hardware and software differences, technology evolution over time, and differences in MRI acquisition parameters. Domain adaptation (DA) methods can make machine learning models more robust to these domain shifts.

In this brain MRI segmentation playground, part of the Calgary-Campinas (CC) dataset initiative, we investigate DA techniques for brain MR image segmentation using the 3 T portion of the CC-359 dataset. These data were collected across sites with scanners from different vendors (Philips, Siemens, and General Electric) and different magnetic fields (1.5 T and 3 T).

Currently, our playground supports the following segmentation tasks:

- ① ■ Skull-Stripping (SS)



MICLab

Home

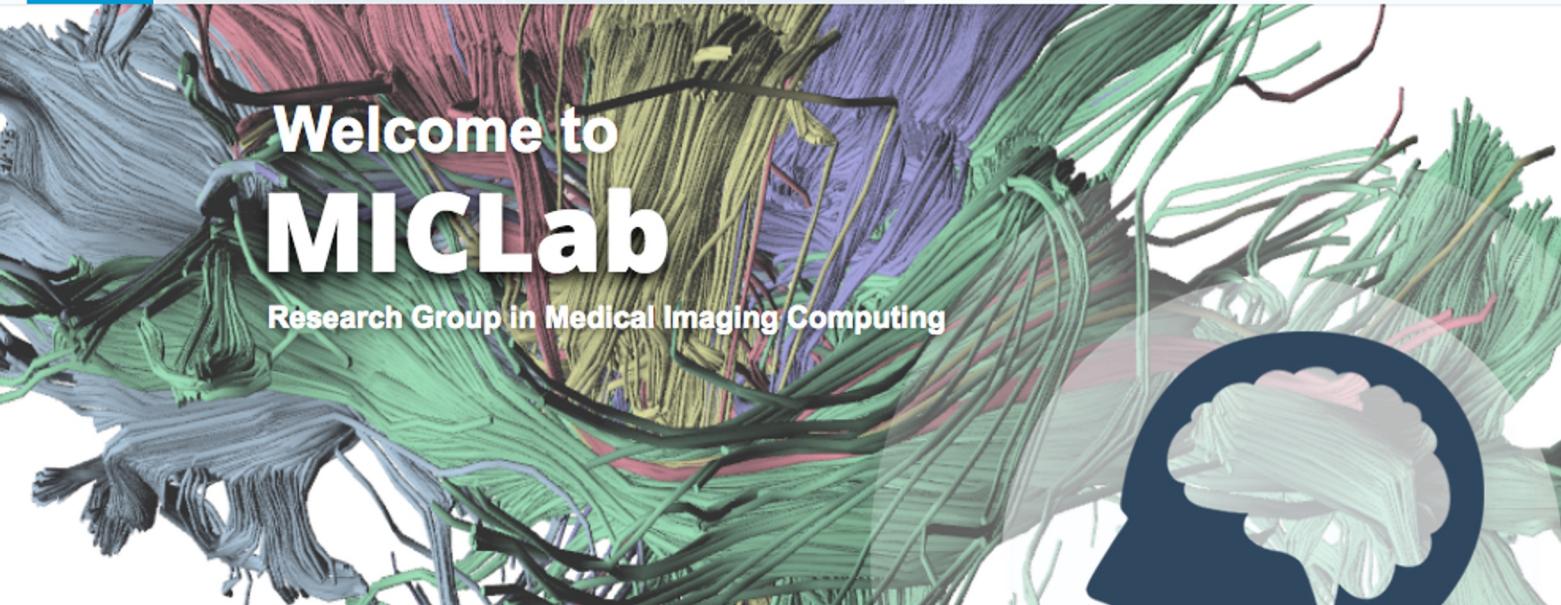
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<http://miclab.fee.unicamp.br/>

<https://github.com/MICLab-Unicamp>

<http://www.lrittner.fee.unicamp.br>



[lrittner@unicamp.br](mailto:lrittner@unicamp.br)

Funding



# Agenda do curso



8h00 – Motivação: o uso de DL na análise de imagens médicas



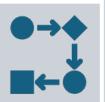
9h00 – *Garbage in, Garbage out*

Diretrizes para o tratamento adequado dos dados  
Parte prática: pré-processamento e uso dos dados



10h00 – Escolhendo o Modelo: o melhor? Depende...

Como funcionam as CNNs? E os Transformers?  
Parte prática: Modelos multimodais envolvendo imagens e texto



11h00 – Avaliando os resultados: do's and dont's

Métricas são mais complicadas do que parecem  
Parte prática: Dice não é tudo