



CBEB 2024

XXIX CONGRESSO BRASILEIRO
DE ENGENHARIA BIOMÉDICA

Deep Learning for Medical Image Analysis: tips, tricks and traps

Profa. Dra. Leticia Rittner

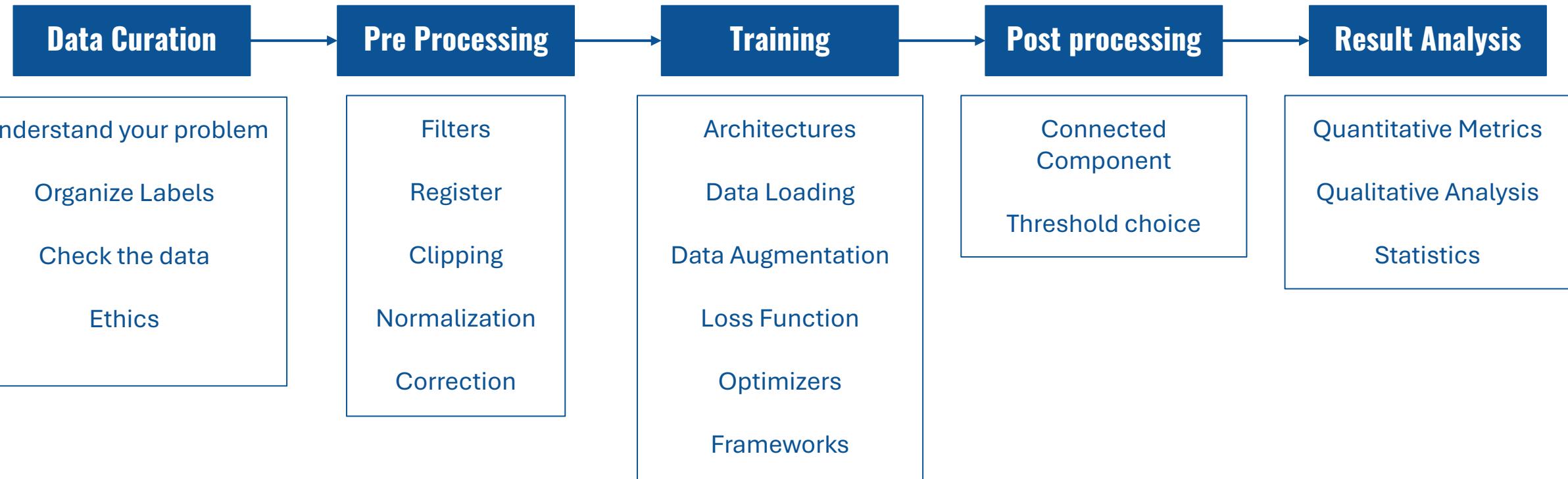
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Faculdade de Engenharia Elétrica e de Computação - Unicamp



Pipeline típico de análise



Pipeline típico de análise



Análise Quantitativa

Análise Qualitativa

Análise Estatística

Pipeline típico de análise



Análise Quantitativa

Análise Qualitativa

Análise Estatística

Análise quantitativa: que métrica escolher?

$$\text{Dice Coefficient} = \frac{2 * TP}{FN + (2 * TP) + FP}$$

$$\text{Jaccard Index} = \frac{TP}{TP + FN + FP}$$

$$\text{Sensitivity} = \frac{TP}{TP + FN}$$

$$\text{Precision} = \frac{TP}{TP + FP}$$

TP - true positive

TN - true negative

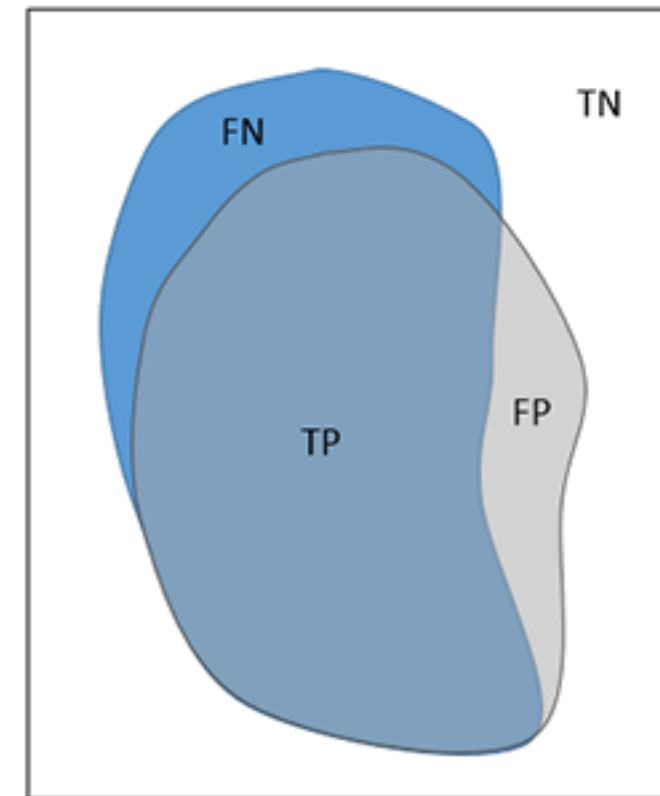
FP - false positive

FN - false negative

Manual Segmentation



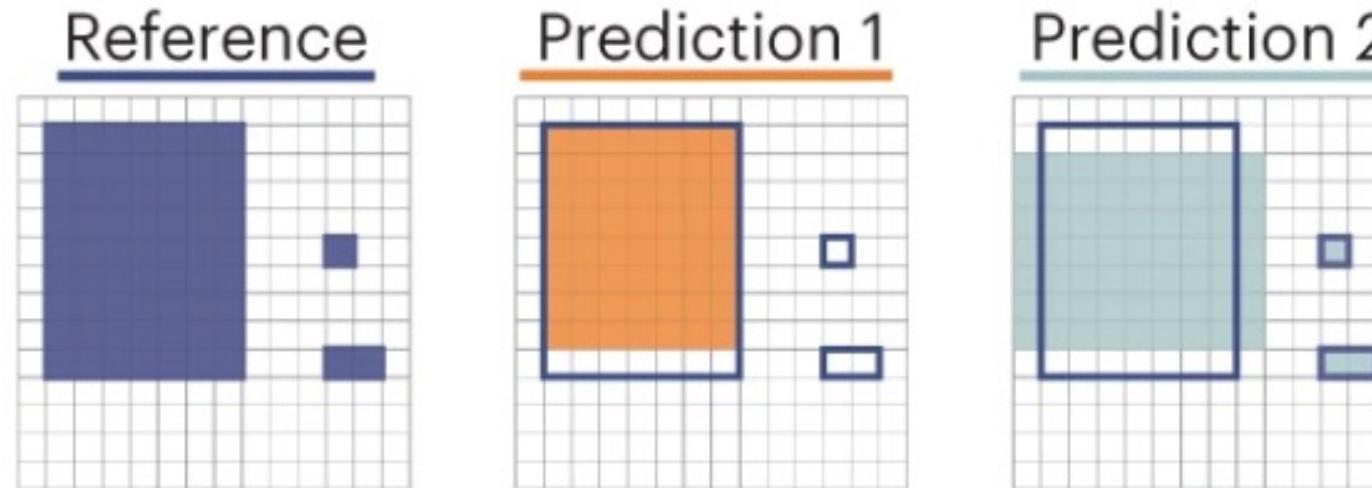
Automated Segmentation



Análise quantitativa: que métrica escolher?



Inappropriate choice of the problem category

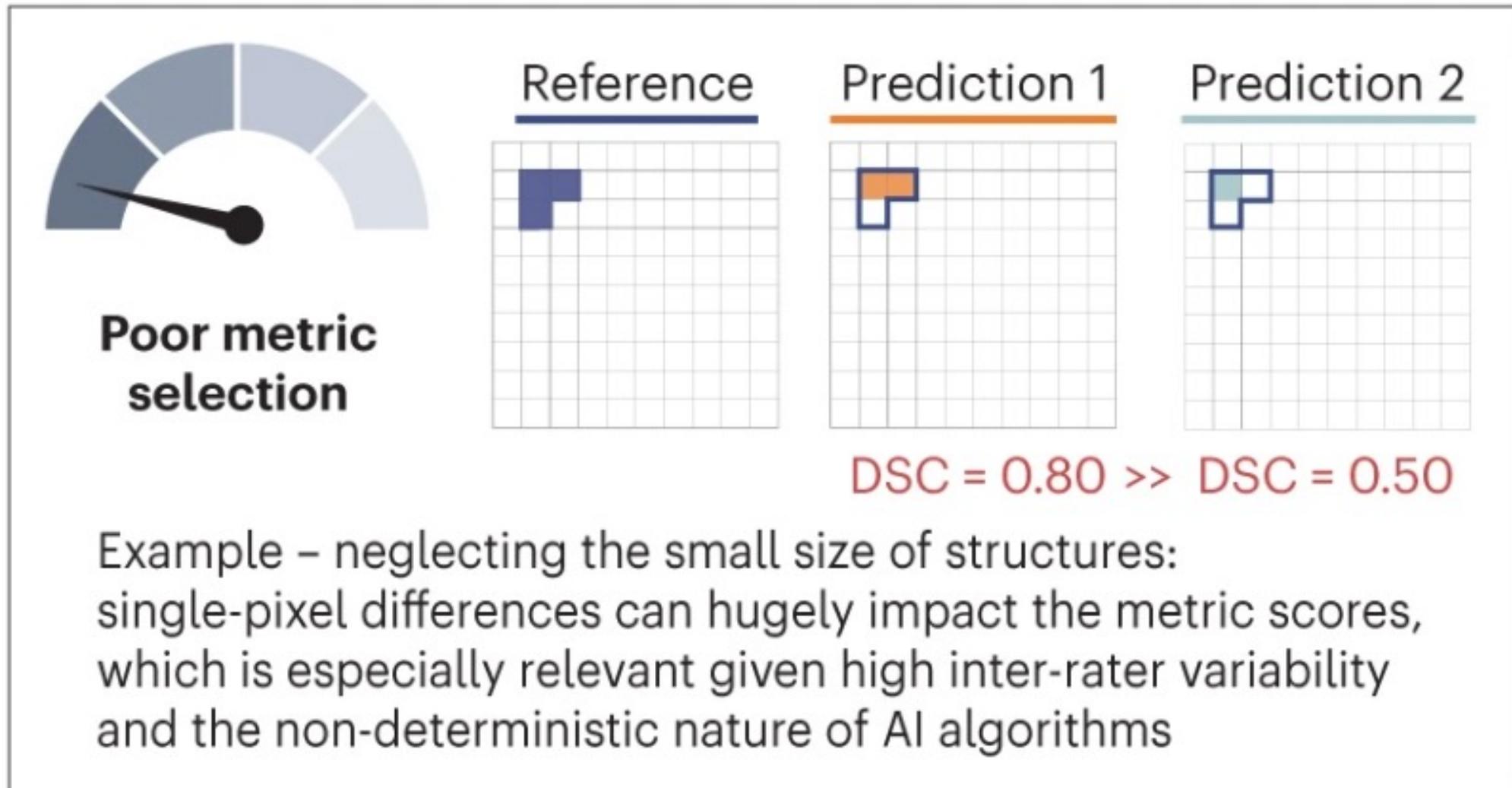


1 object detected ✗ 3 objects detected ✓

DSC = 0.92 >> DSC = 0.79

Example – object detection confused with semantic segmentation:
DSC is strongly biased toward single objects and is therefore not appropriate for measuring the detection of multiple objects

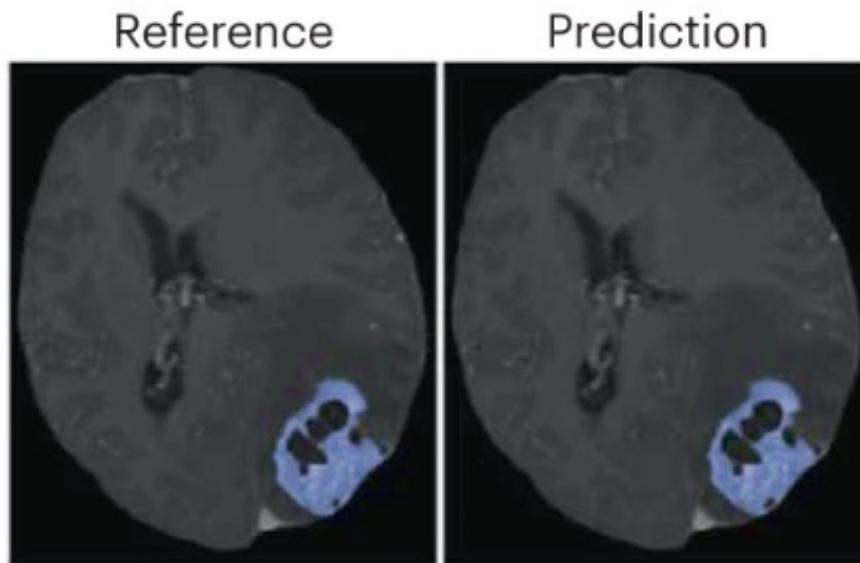
Análise quantitativa: que métrica escolher?



Análise quantitativa: que métrica escolher?

Popular voxel-based metrics fail to capture clinical interest

Magnetic resonance imaging, same patient, different slices



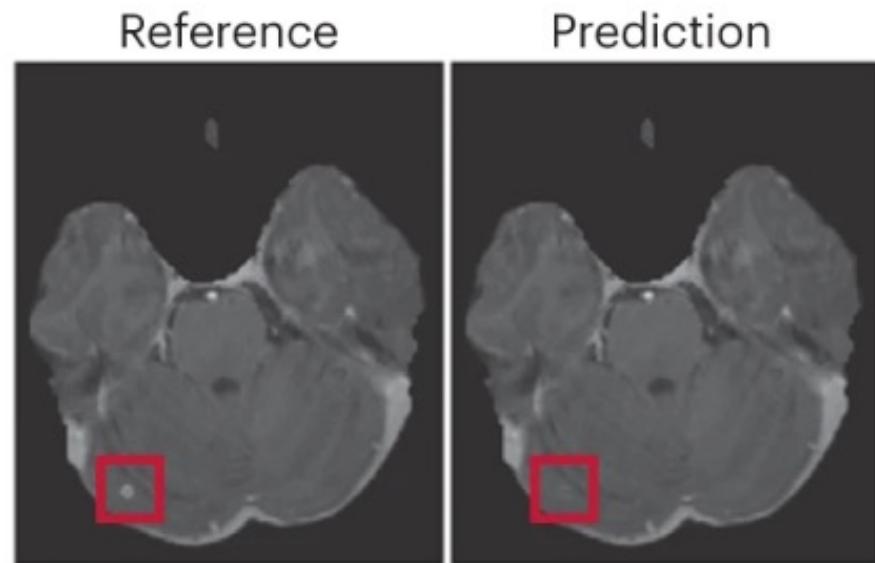
Sensitivity = 0.94
(voxel-level)



Sensitivity = 0.50
(instance-level)



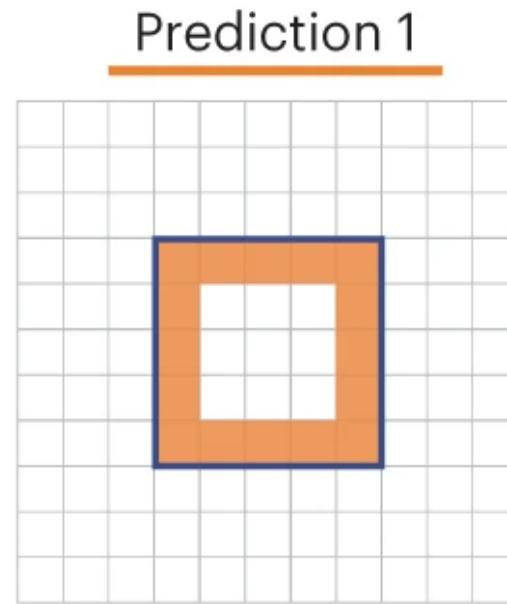
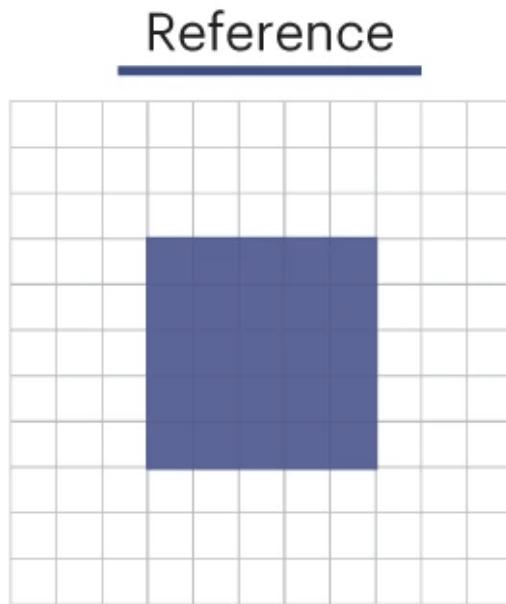
Missed lesion!



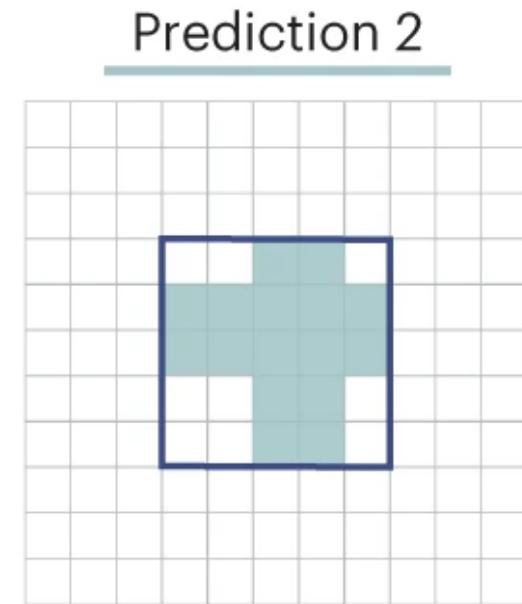
Medical example: brain-tumor segmentation

A near-perfect voxel-level sensitivity hides information on missed lesions

Análise quantitativa: que métrica escolher?



DSC = 0.78



DSC = 0.78

Análise quantitativa: que métrica escolher?

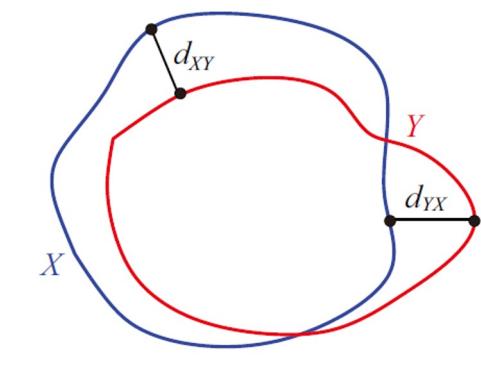
Métricas de segmentação têm vantagens e desvantagens. Sempre use mais de uma métrica, por exemplo, uma de sobreposição e uma de distância de borda.

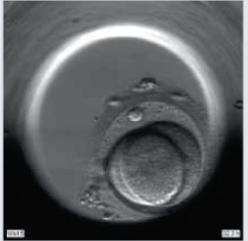
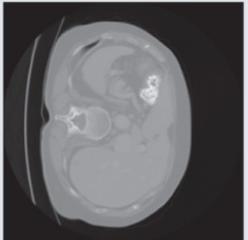
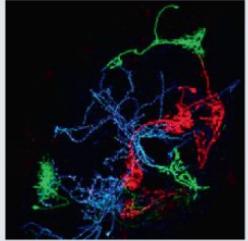
1) *Dice Coefficient*: The *DC* is an overlap measure defined as follows:

$$DC = \frac{2 * |M \cap A|}{|M| + |A|} \quad (4)$$

2) *Hausdorff distance*: computa a distância entre os pontos das 2 bordas: a verdadeira e a predita.

Atenção:
diferentes tarefas
exigem diferentes
métricas!



Problem description	ID	Scenario	Sample input image	Recommended output	Recommendation
Segmentation of large objects	SemS-1	Embryo segmentation from microscopy images			<p>Problem category: Semantic segmentation</p> <p>Overlap-based metric (S6): Dice similarity coefficient (DSC)</p> <p>Boundary-based metric (S7): Normalized surface distance (NSC)</p> <p>Specific property-related metric: Liver segmentation: absolute volume difference</p>
	SemS-2	Liver segmentation in CT images			
Segmentation and distinction of tubular objects	InS-1	Instance segmentation of neurons from the fruit fly in 3D multicolor light microscopy images			<p>Problem category: Instance segmentation</p> <p>Per-class counting metric (S3): F_β score</p> <p>Multi-threshold metric (S4): average precision (AP)</p> <p>Overlap-based metric (S6): Centerline Dice similarity coefficient (cIDice)</p> <p>Boundary-based metric (S7): NSD</p> <p>Localization criterion (S8): Neuron segmentation: mask IoU Instrument segmentation: boundary IoU</p> <p>Assignment strategy (S9): Greedy (by score) matching, set double assignments to FPs</p>
	InS-2	Surgical instrument instance segmentation in colonoscopy videos			

GETTING STARTED

Metric Selection

⌚ Duration 5-10min.

This module is designed to determine proper metrics for your study. Then the questions regarding to your study will begin.

Please select your problem category to start:



Unsure which problem category you are at? Please follow the [Problem Category Selection](#) process first.





Merel Huisman, MD, PhD • 2nd
Radiologist | Clinical Epidemiologist AI
2mo • 

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"A Single Metric is No Metric"

It's time to rethink how we evaluate AI tools in medical imaging. While AUC-ROC and accuracy are intuitive, they often fail to tell the whole story, especially in imbalanced datasets. It really pays off to do your due diligence and understand this matter. Relying solely on these metrics can be misleading and even harmful in clinical settings.

Mosquera et al. recognized this issue and published a great educational study in **European Radiology**. I had the honor to write a comment, in which I break down the issue and put it into context.

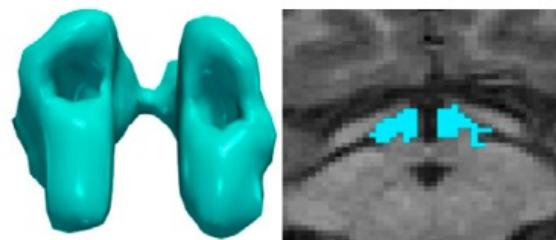
If you remember one thing, let it be this:

"a small imperfection in specificity gets magnified as a high false positive rate in rarer pathologies"

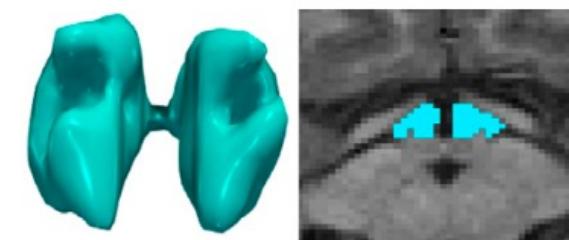
Análise quantitativa: meu resultado é bom?

- Variabilidade *Inter e intra-rater* também são importantes

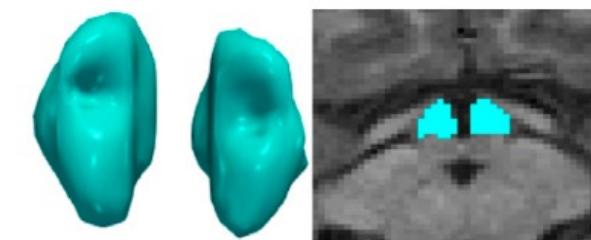
Comparative	DC	VS	AVD	Hausdorff
Rater1 vs Rater2	0.71 ± 0.04	0.86 ± 0.09	0.31 ± 0.04	2.93 ± 0.51
Automated Segmentation vs Rater 1	0.85 ± 0.03	0.95 ± 0.03	0.15 ± 0.03	1.96 ± 0.50
Automated Segmentation vs Rater 2	0.73 ± 0.04	0.86 ± 0.08	0.27 ± 0.04	2.67 ± 0.66



(a) Rater 1



(b) Rater 2



(c) Automated Segmentation

Pipeline típico de análise



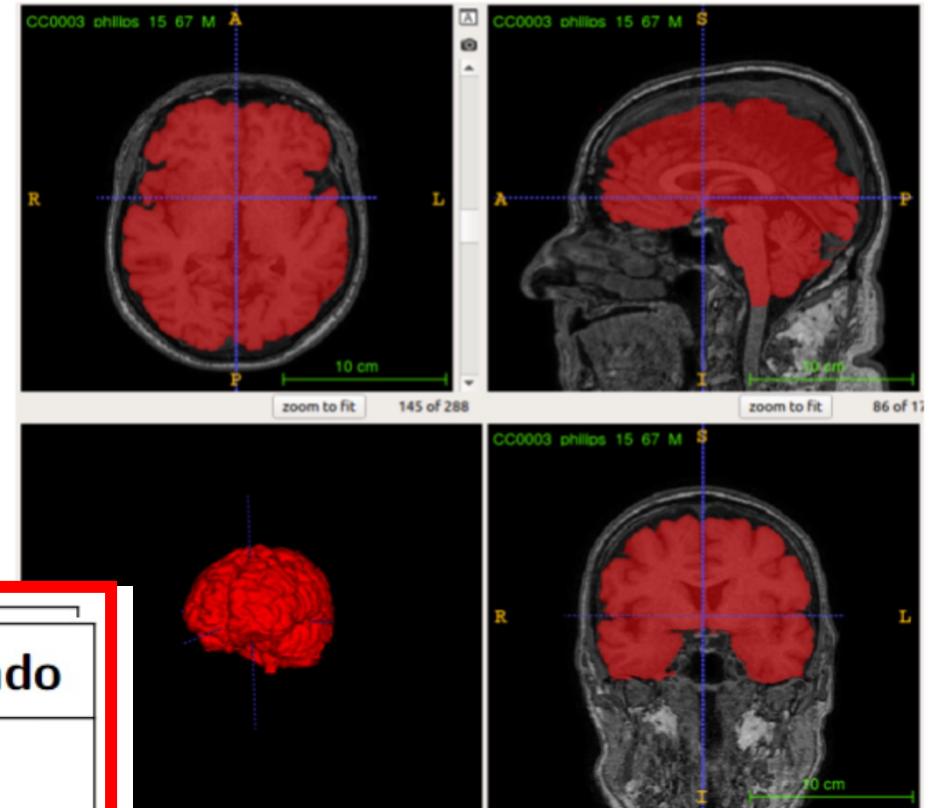
Análise Quantitativa

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Análise Estatística

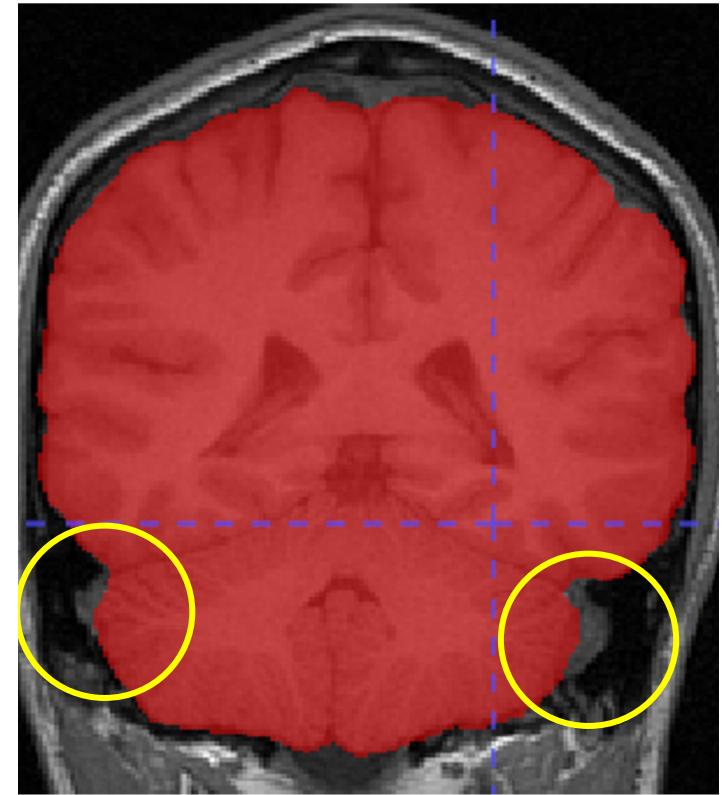
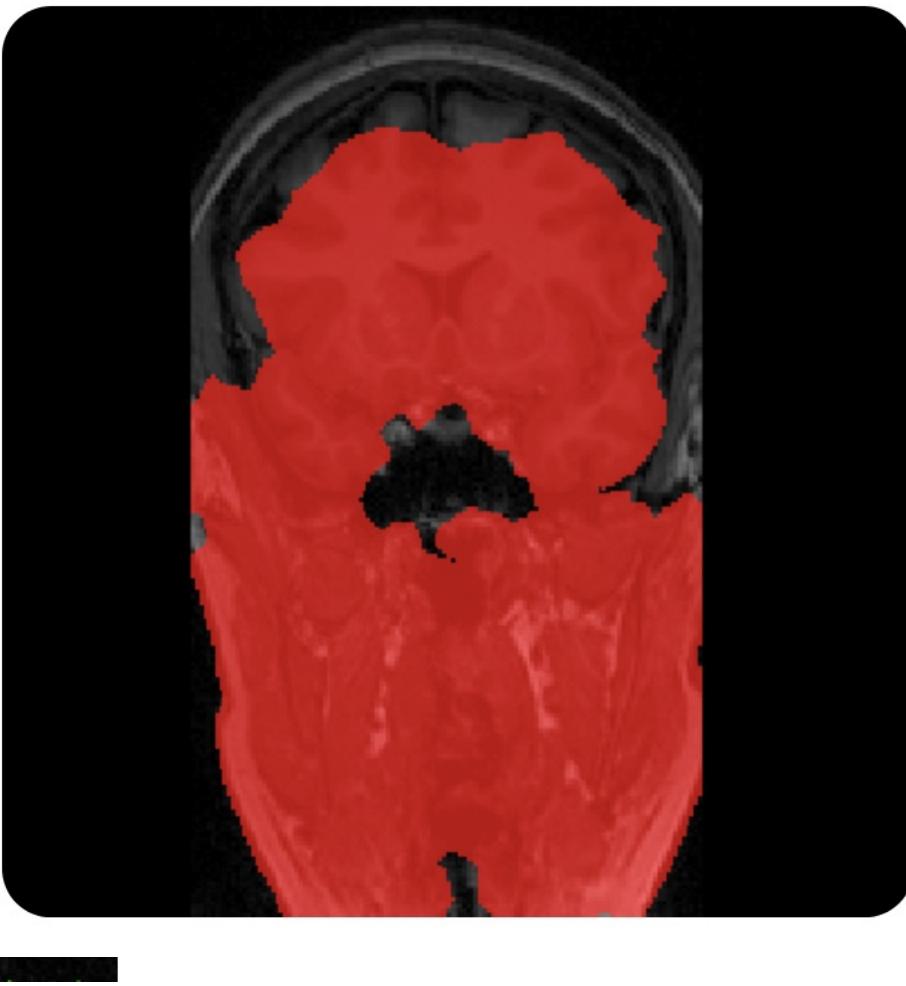
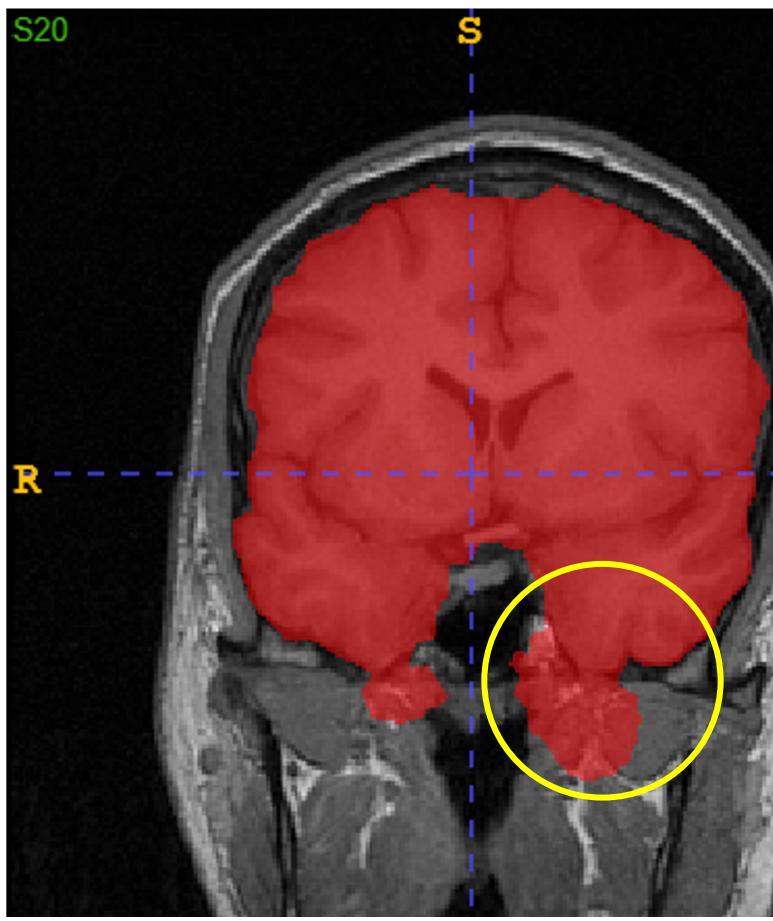
Análise qualitativa

Método	Autor	Ano	Base	Dice Reportado
BET	Smith, S. M.	2002	Atlas	0.95
CONSNET	Lucena et al.	2019	UNet	0.97
HDBET	Isensee et al.	2019	UNet 3D	0.97
PARIETAL	Valverde et al.	2021	UNet 3D	0.97
ROBEX	Iglesias et al.	2011	Híbrido	0.96
SynthStrip	Hoopes et al.	2022	UNet 3D	0.96



Então o
problema está
resolvido?

Análise qualitativa



Análise qualitativa

- Números não são usualmente suficientes
- Inspeccione os resultados visualmente, especialmente os piores casos
- Você pode também construir mapas de erros

BET

ROBEX

CONSNET

HDBET

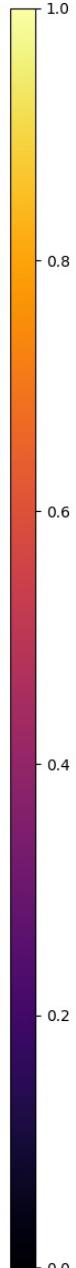
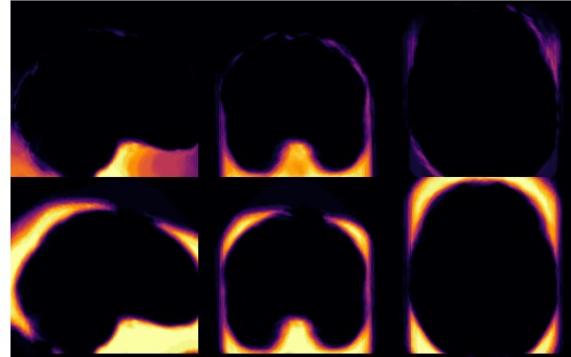
PAEIETAL

SYNTHSTRIP

UNETR

Falsos Positivos

Falsos Negativos



[Submitted on 12 Apr 2023]

Automated computed tomography and magnetic resonance imaging segmentation using deep learning: a beginner's guide

Diedre Carmo, Gustavo Pinheiro, Lívia Rodrigues, Thays Abreu, Roberto Lotufo, Letícia Rittner

Medical image segmentation is an increasingly popular area of research in medical imaging processing and analysis. However, many researchers who are new to the field struggle with basic concepts. This tutorial paper aims to provide an overview of the fundamental concepts of medical imaging, with a focus on Magnetic Resonance and Computerized Tomography. We will also discuss deep learning algorithms, tools, and frameworks used for segmentation tasks, and suggest best practices for method development and image analysis. Our tutorial includes sample tasks using public data, and accompanying code is available on GitHub ([this https URL](https://github.com)). By sharing our insights gained from years of experience in the field and learning from relevant literature, we hope to assist researchers in overcoming the initial challenges they may encounter in this exciting and important area of research.

Comments: Equal contribution from Diedre Carmo, Gustavo Pinheiro, and Lívia Rodrigues

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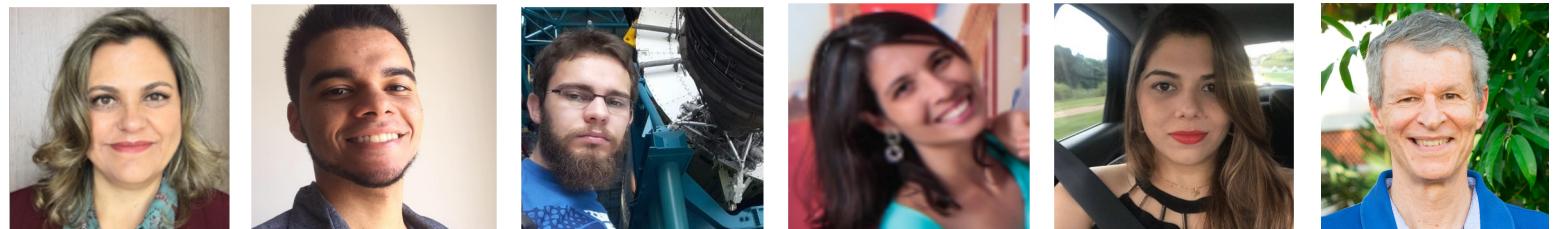
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References & Citations

- NASA ADS
- Google Scholar
- Semantic Scholar



[1] Carmo, D., Pinheiro, G., Rodrigues, L., Abreu, T., Lotufo, R., & Rittner, L. (2023). Automated computed tomography and magnetic resonance imaging segmentation using deep learning: a beginner's guide. *arXiv preprint arXiv:2304.05901*.



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>

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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



MICLab

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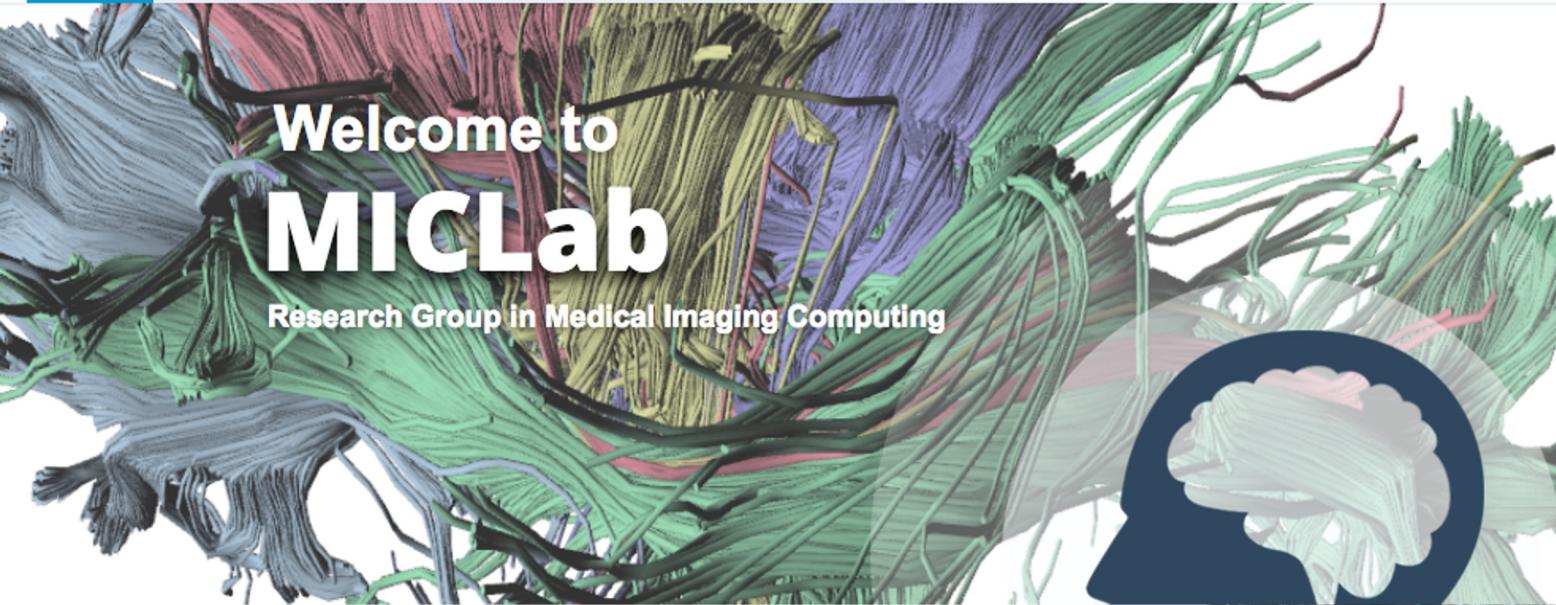
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