

Enhancing Oral Diagnostics with Hyperspectral Imaging and Deep Learning:

Mucosa Segmentation and Preliminary Findings on Carcinoma Detection

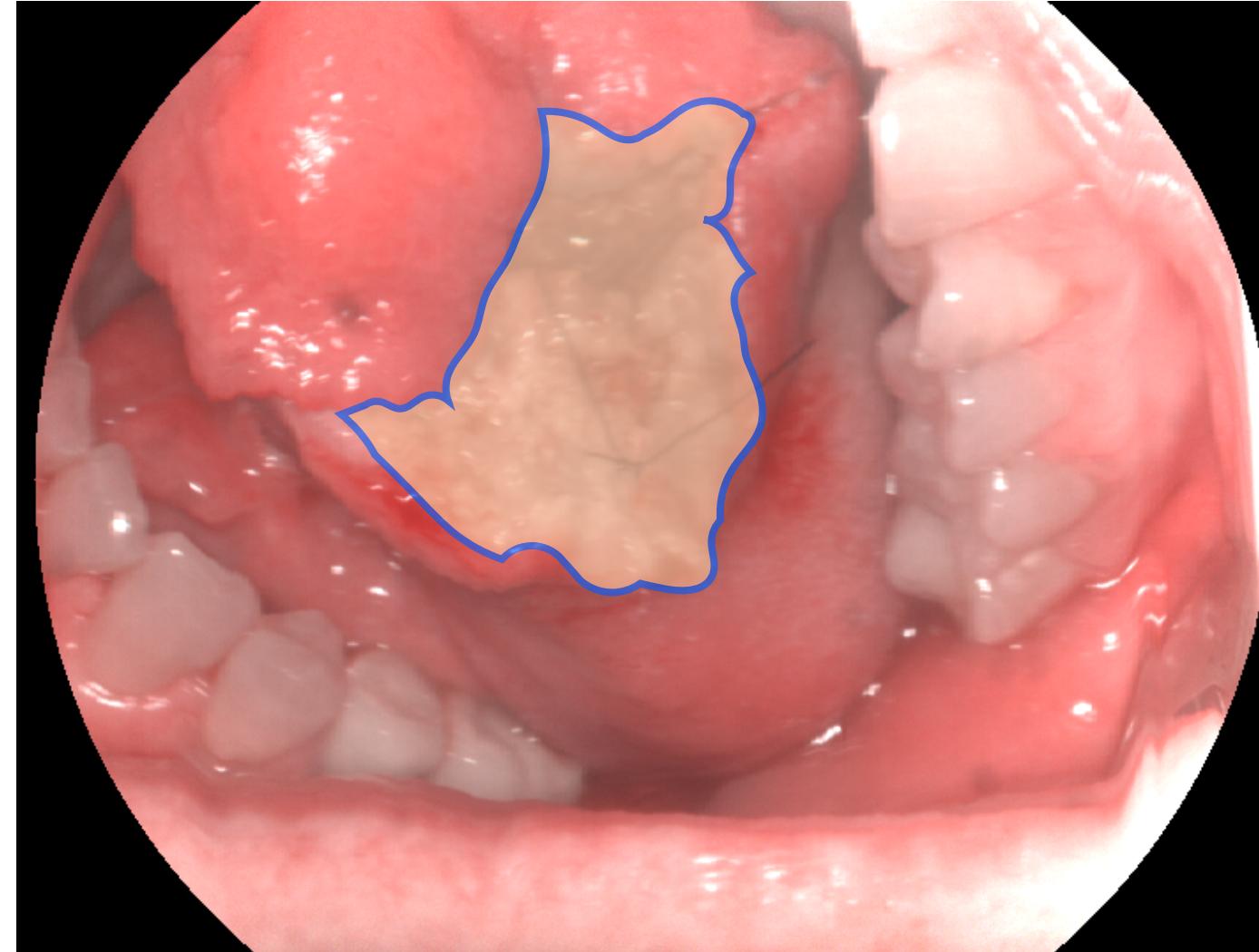
350,000

Purpose of the Research: early detection of carcinoma tumor

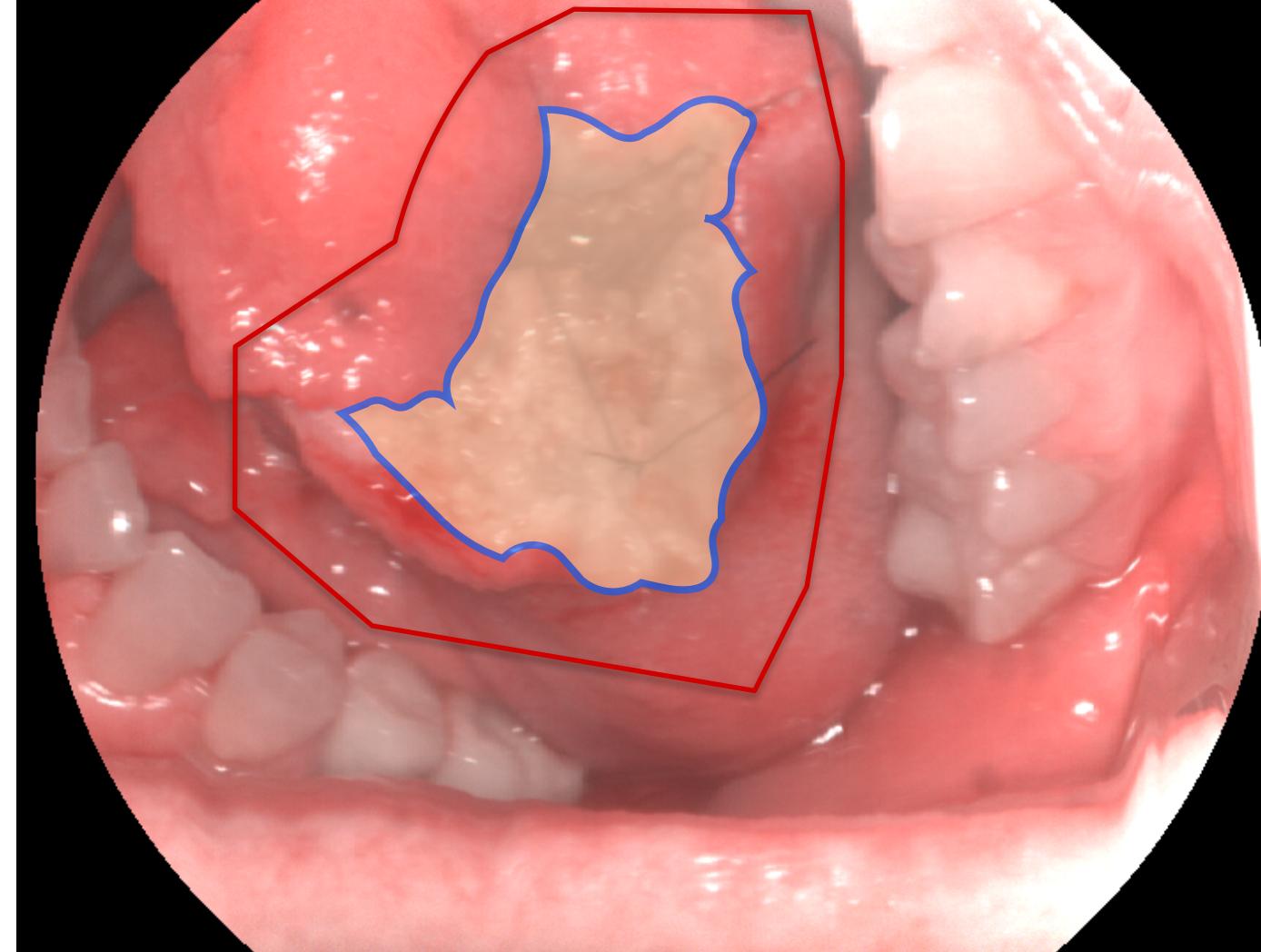
- 350,000 new cases / year
- 60% 5-year survival rate
- 30% Recidivism rate
- 50% of diagnoses in advanced stages
- Surgery as primary therapy is the current standard



Purpose of the Research: early detection of carcinoma tumor

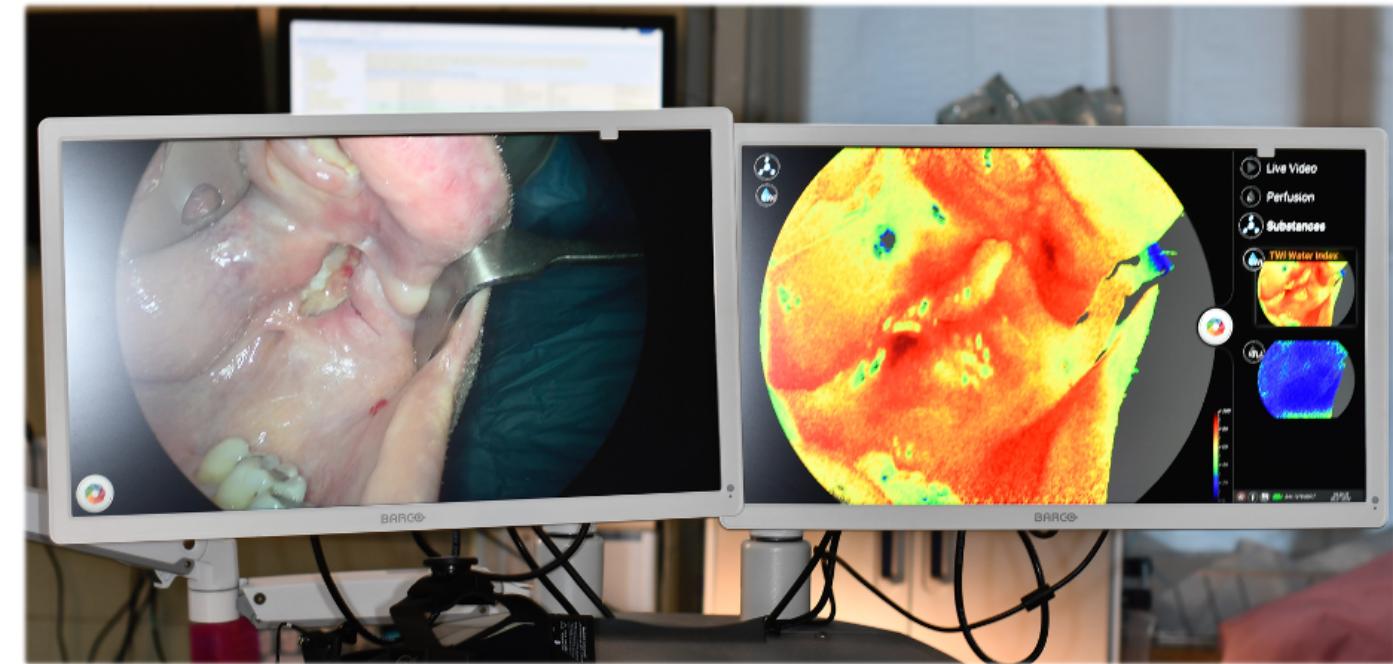
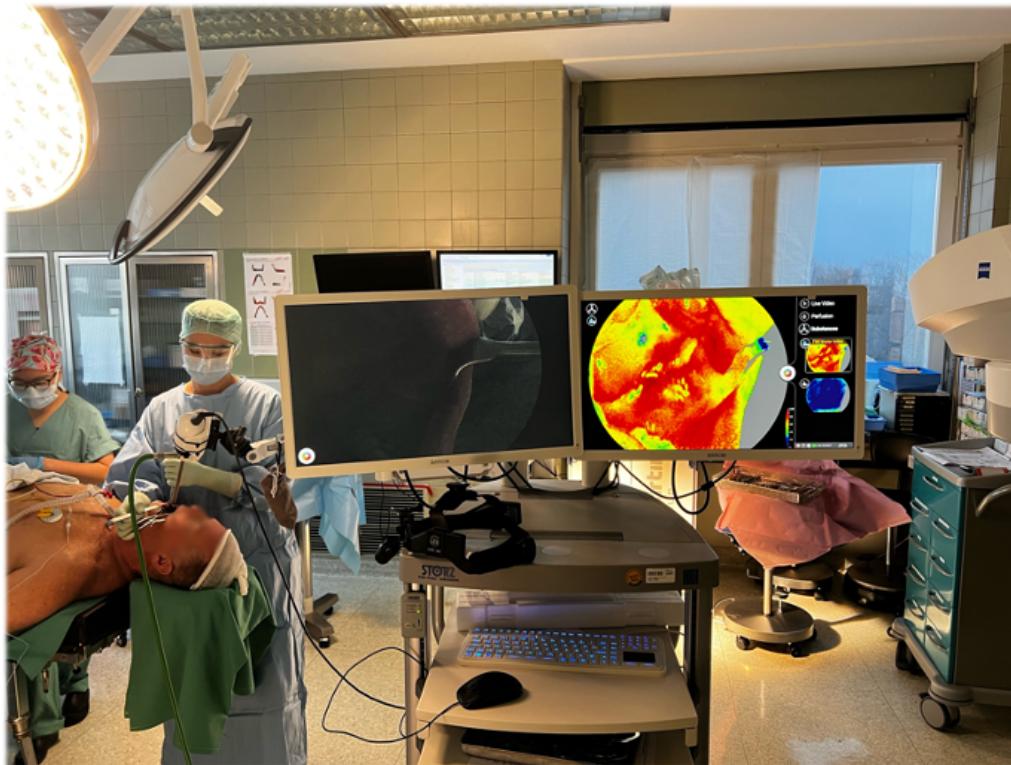


Purpose of the Research: early detection of carcinoma tumor



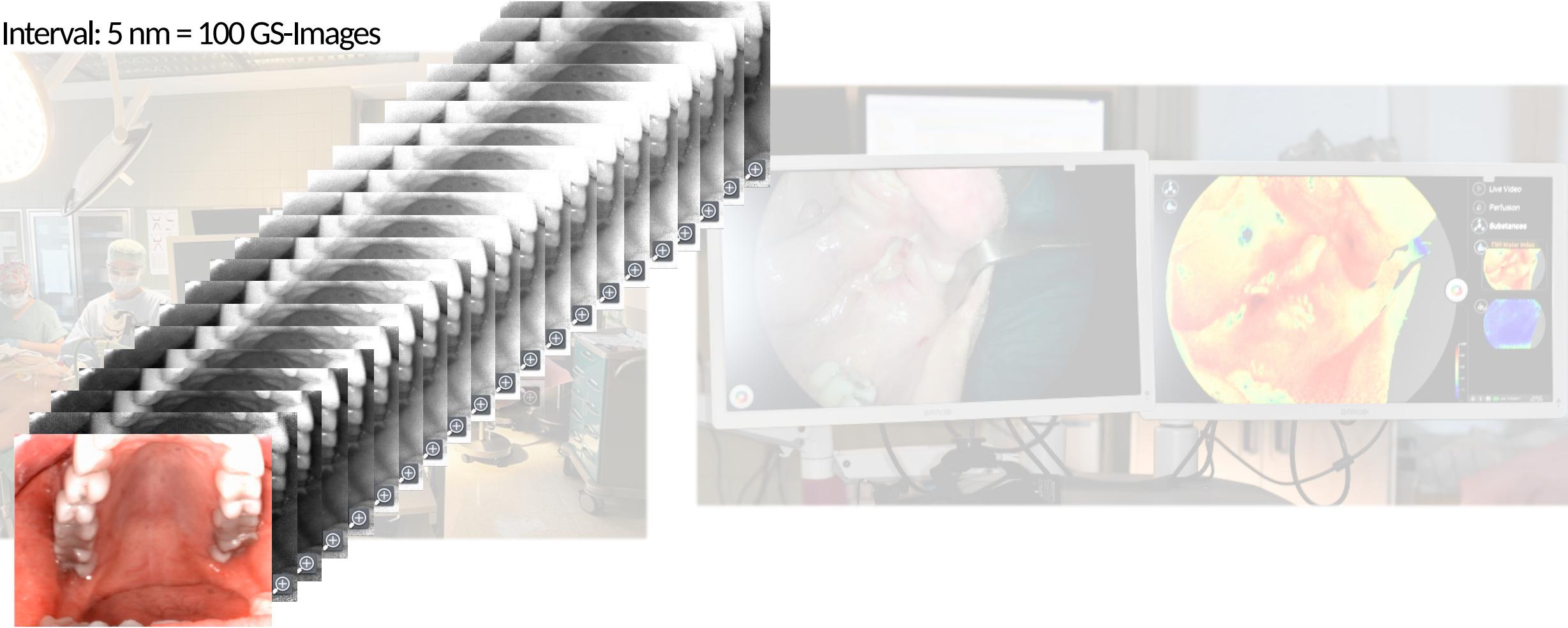
1 cm margin

Purpose of the Research: early detection of carcinoma tumor



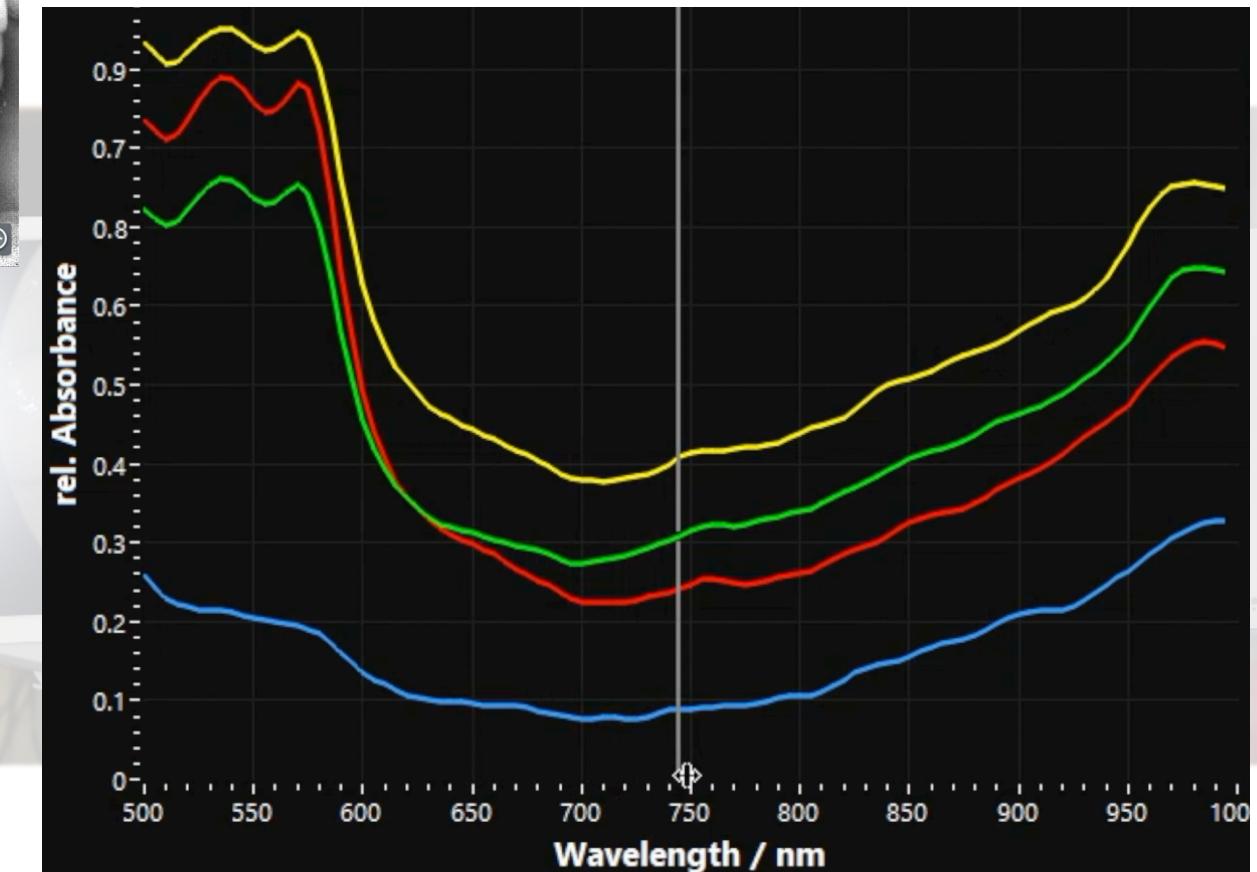
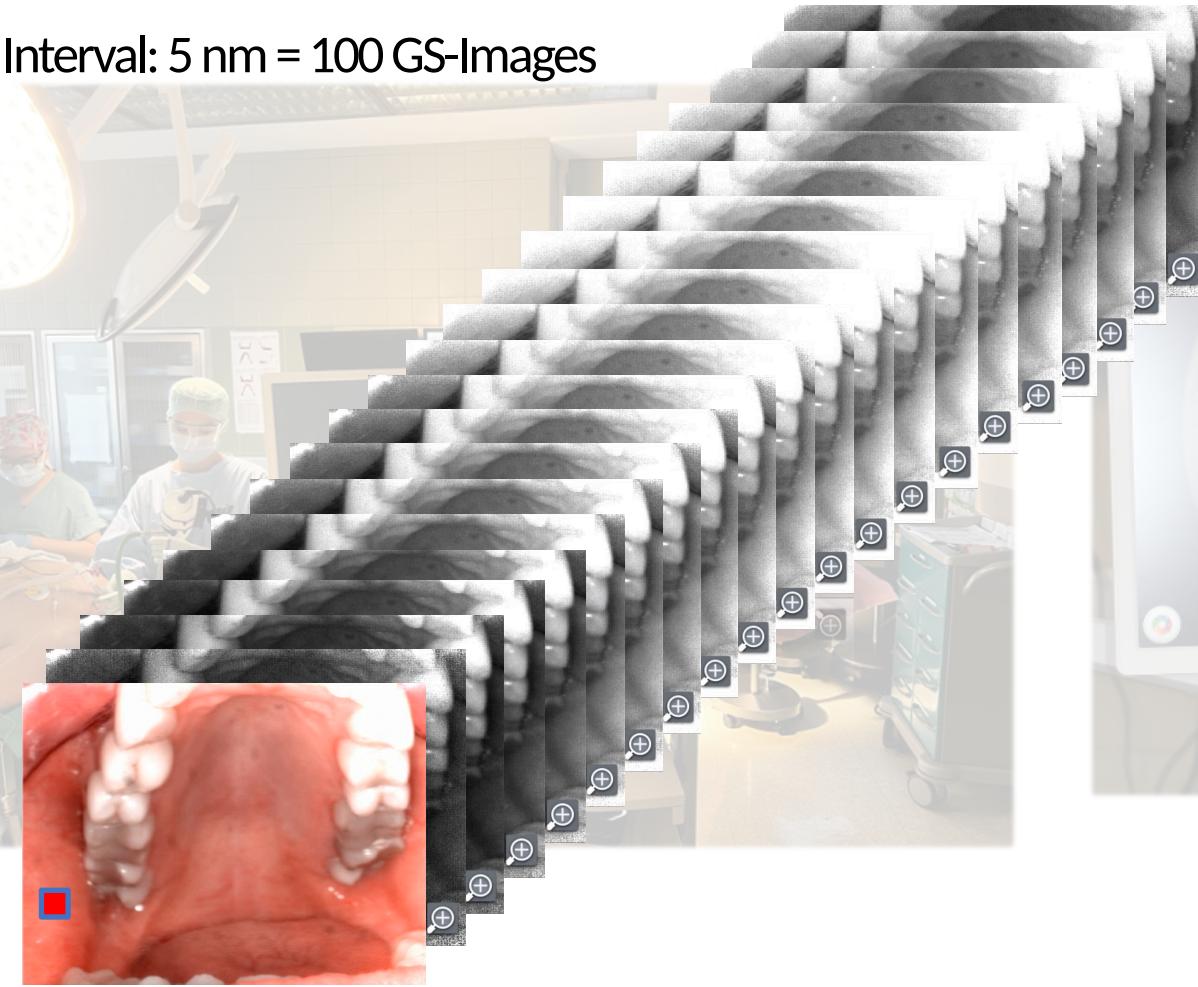
Purpose of the Research: early detection of carcinoma tumor

Interval: 5 nm = 100 GS-Images



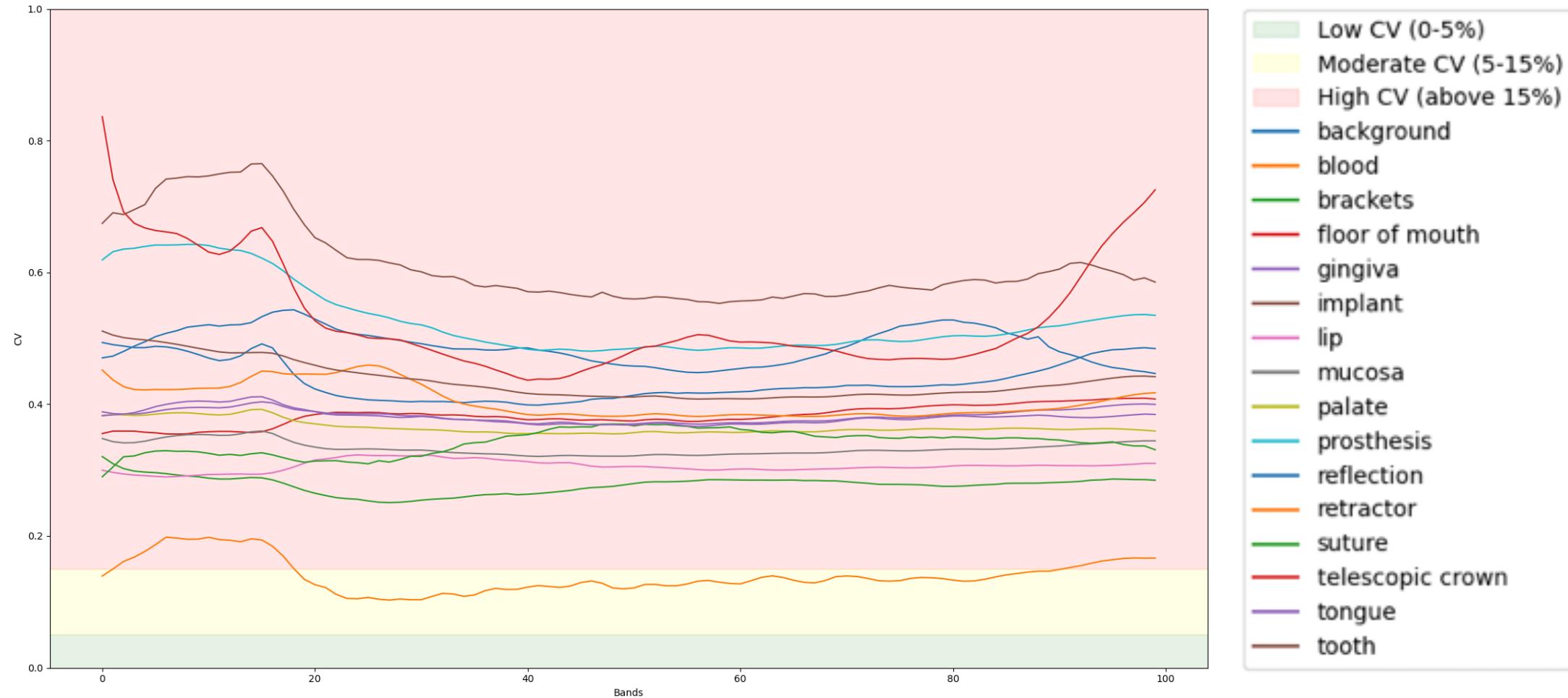
Purpose of the Research: early detection of carcinoma tumor

Interval: 5 nm = 100 GS-Images

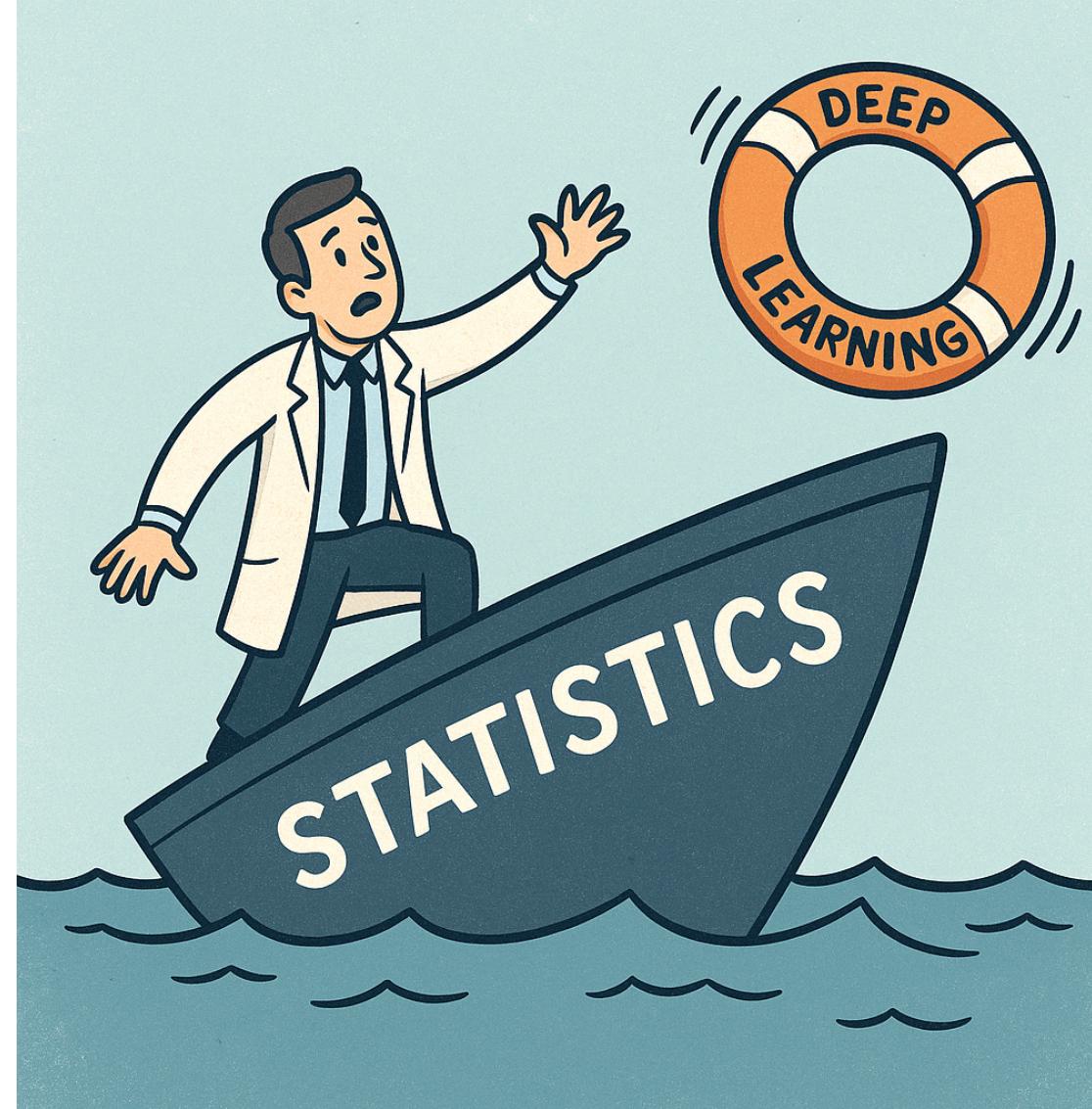


Acquisition Analysis

- High coefficient of variation (CV)
- Class identification based only on its signature is impractical
- Excluding reliance on ad hoc solutions

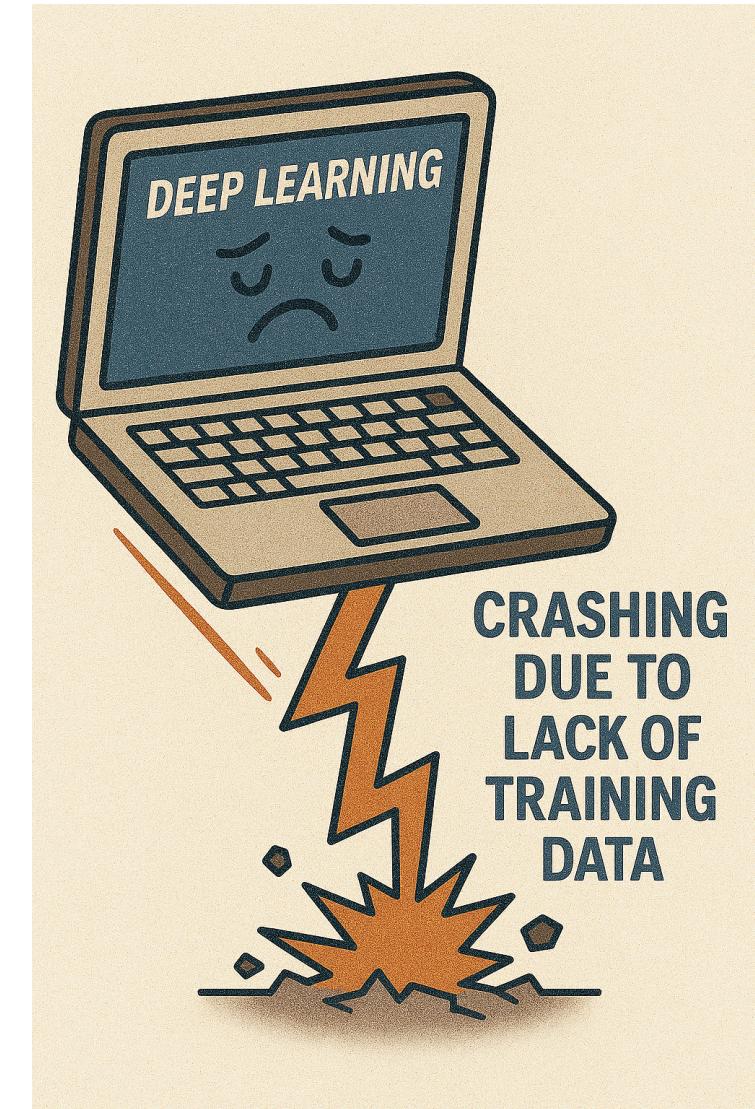


Deep Learning our only chance?



We had only a few samples

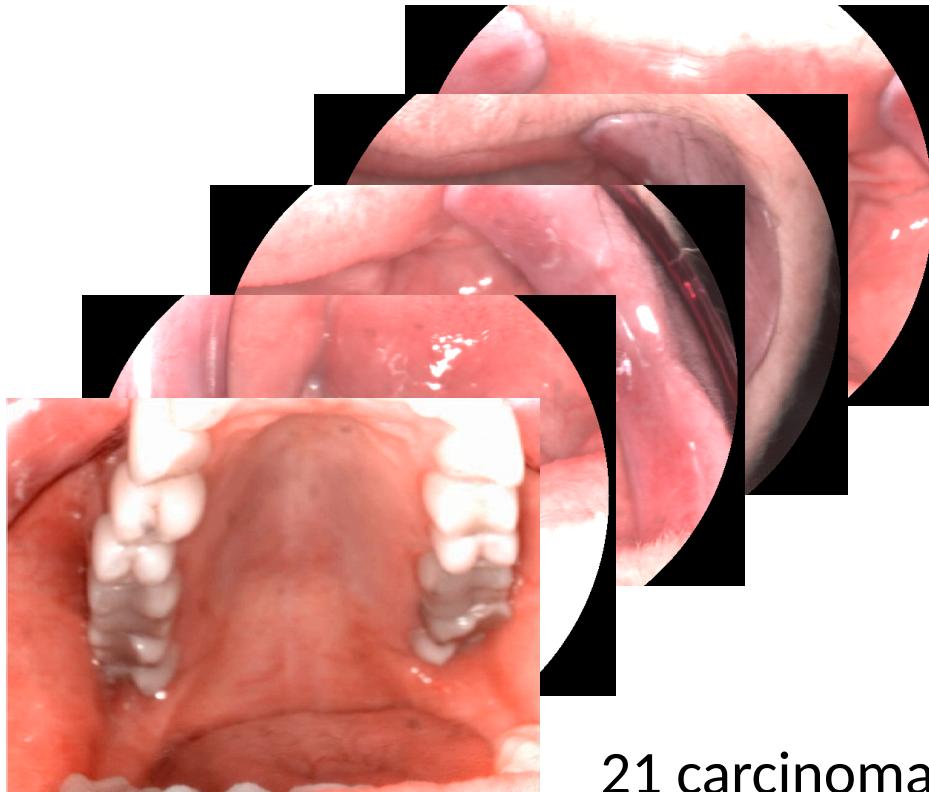
Only 21 cases with carcinoma tumor for more than 752 patients without tumor



Multiple datasets...One possibility?

1

In-Vivo Oral Cavity Scan



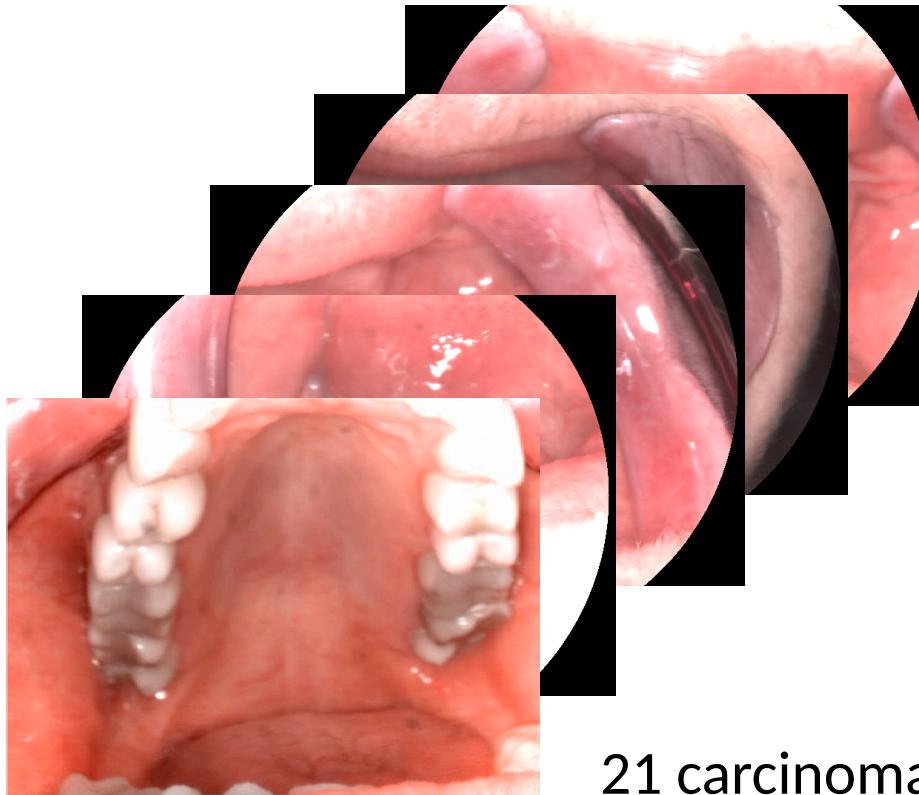
21 carcinoma-positive

752 non-carcinoma

Multiple datasets...One possibility?

1

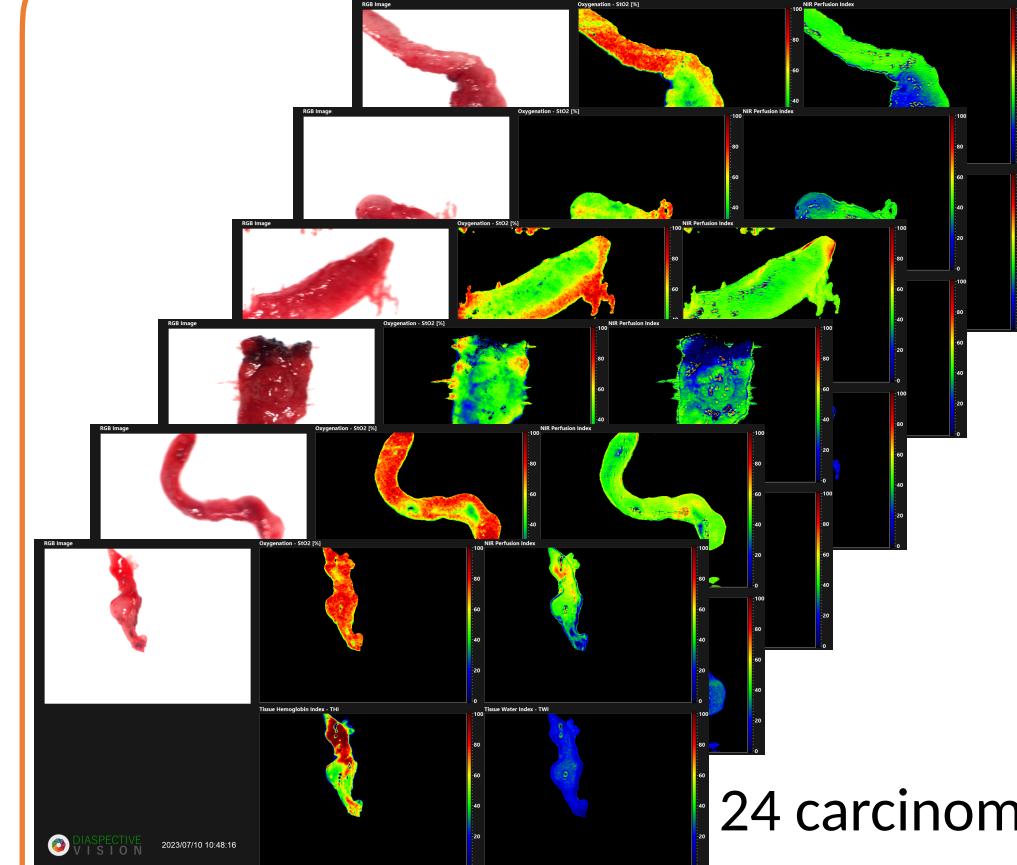
In-Vivo Oral Cavity Scan



21 carcinoma-positive
752 non-carcinoma

2

Ex-Vivo Frozen Section

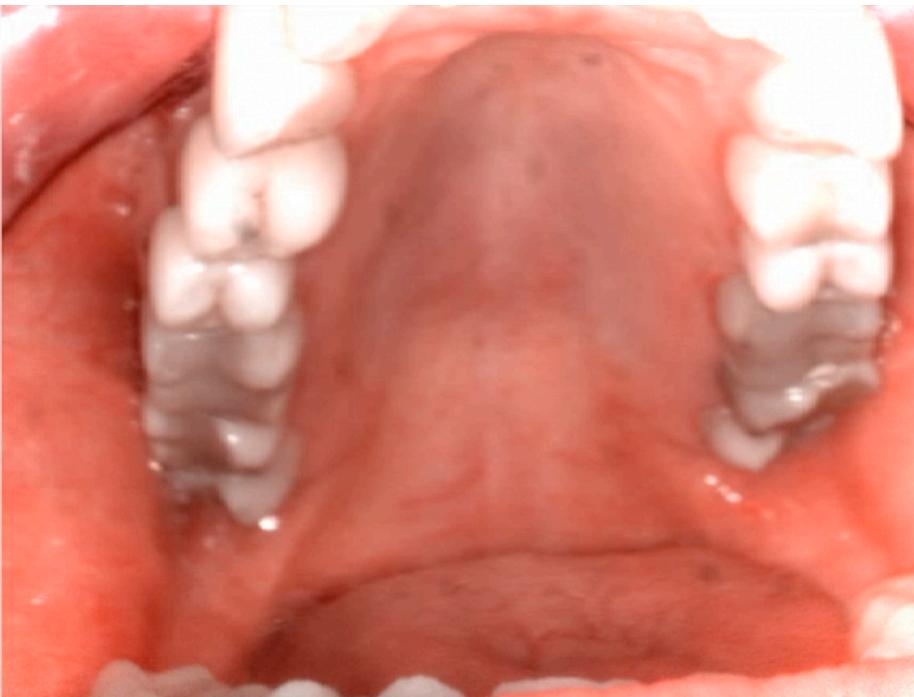


24 carcinoma-positive
200 non-carcinoma

Multiple datasets...One possibility?

1

In-Vivo Oral Cavity Scan



2

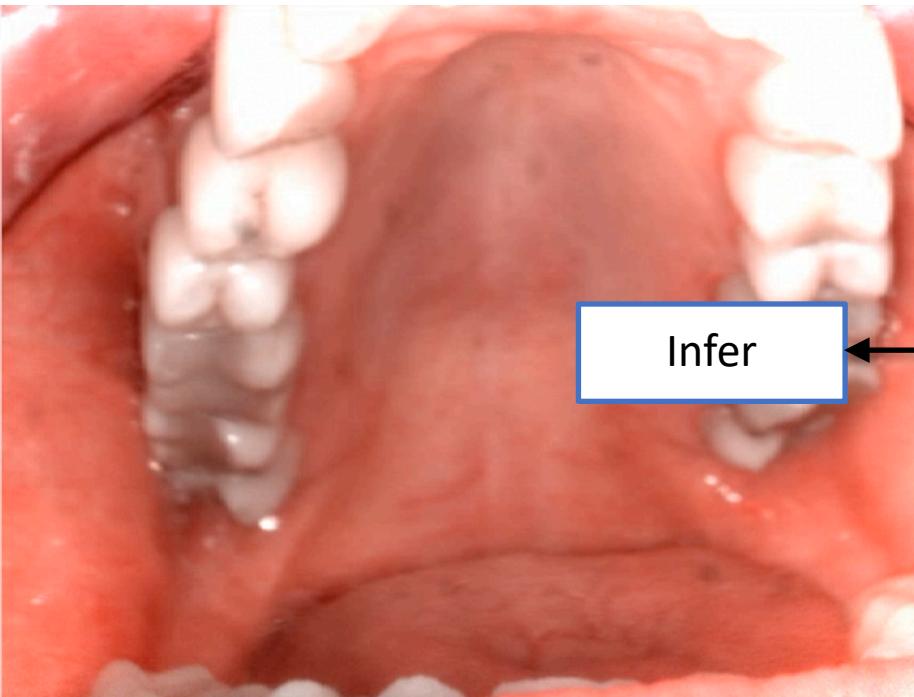
Ex-Vivo Frozen Section



Multiple datasets...One possibility?

1

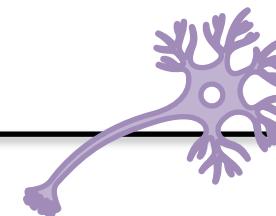
In-Vivo Oral Cavity Scan



Infer

2

Ex-Vivo Frozen Section



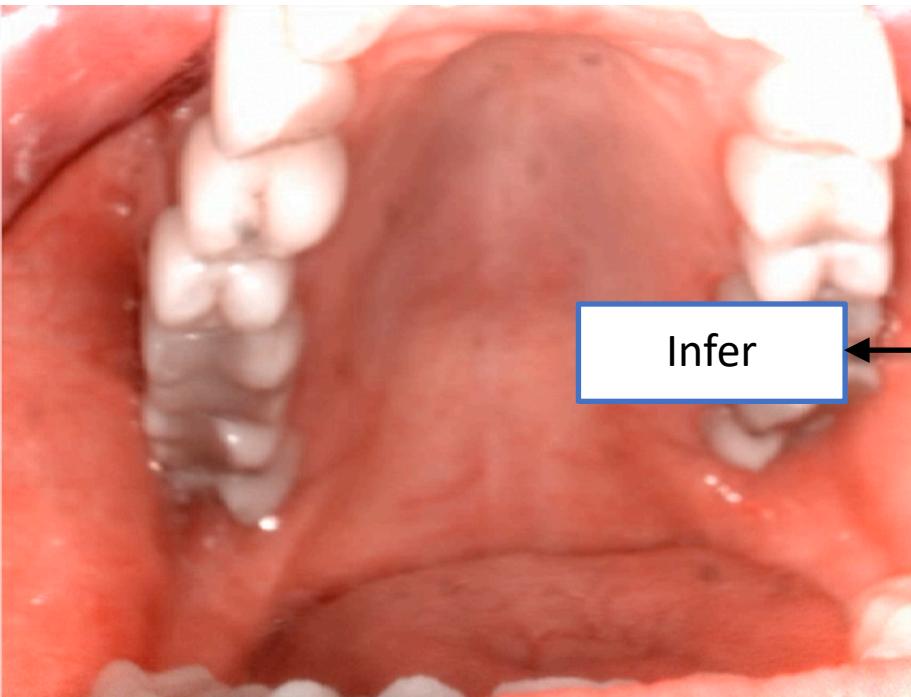
Train



Multiple datasets...One possibility?

1

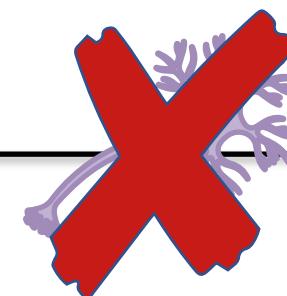
In-Vivo Oral Cavity Scan



Infer

2

Ex-Vivo Frozen Section

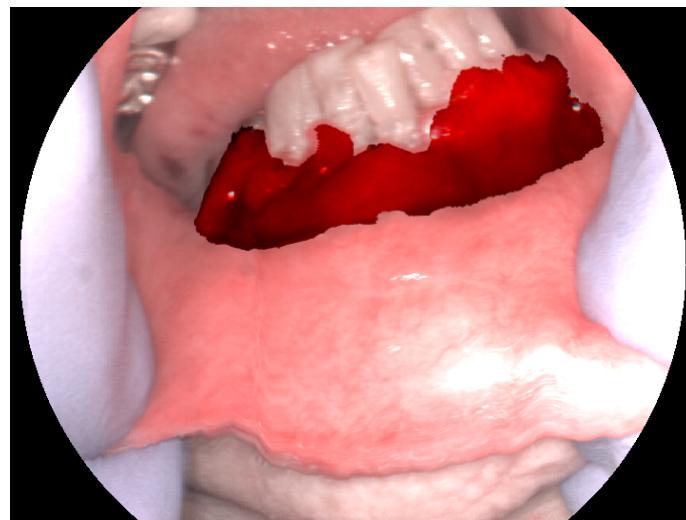
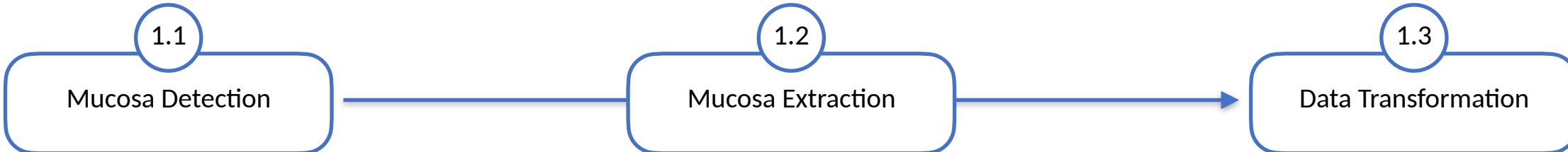


Train

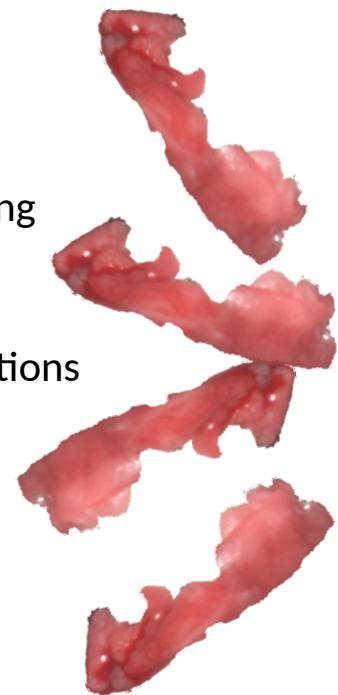


Our strategy

In-Vivo Oral Cavity Scan: Real world scenario

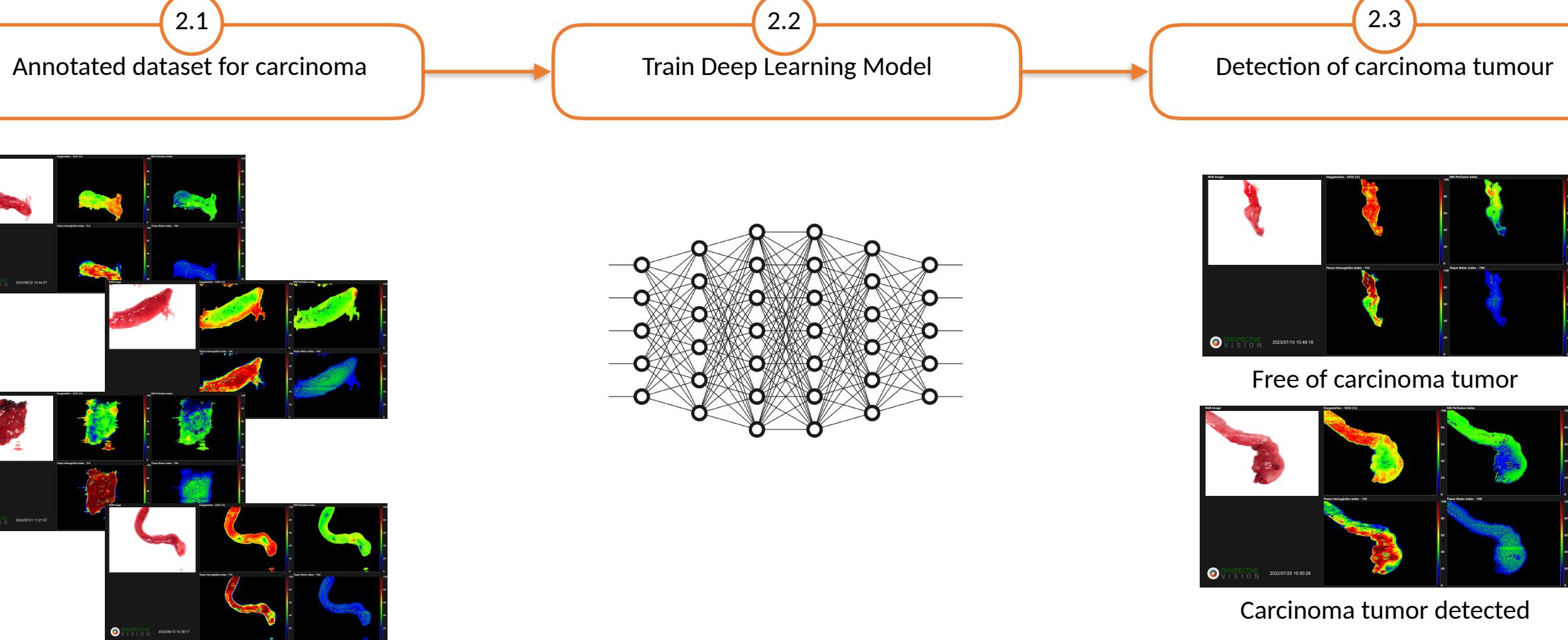


- Random cropping
- Flipping
- Rotation
- Elastic deformations



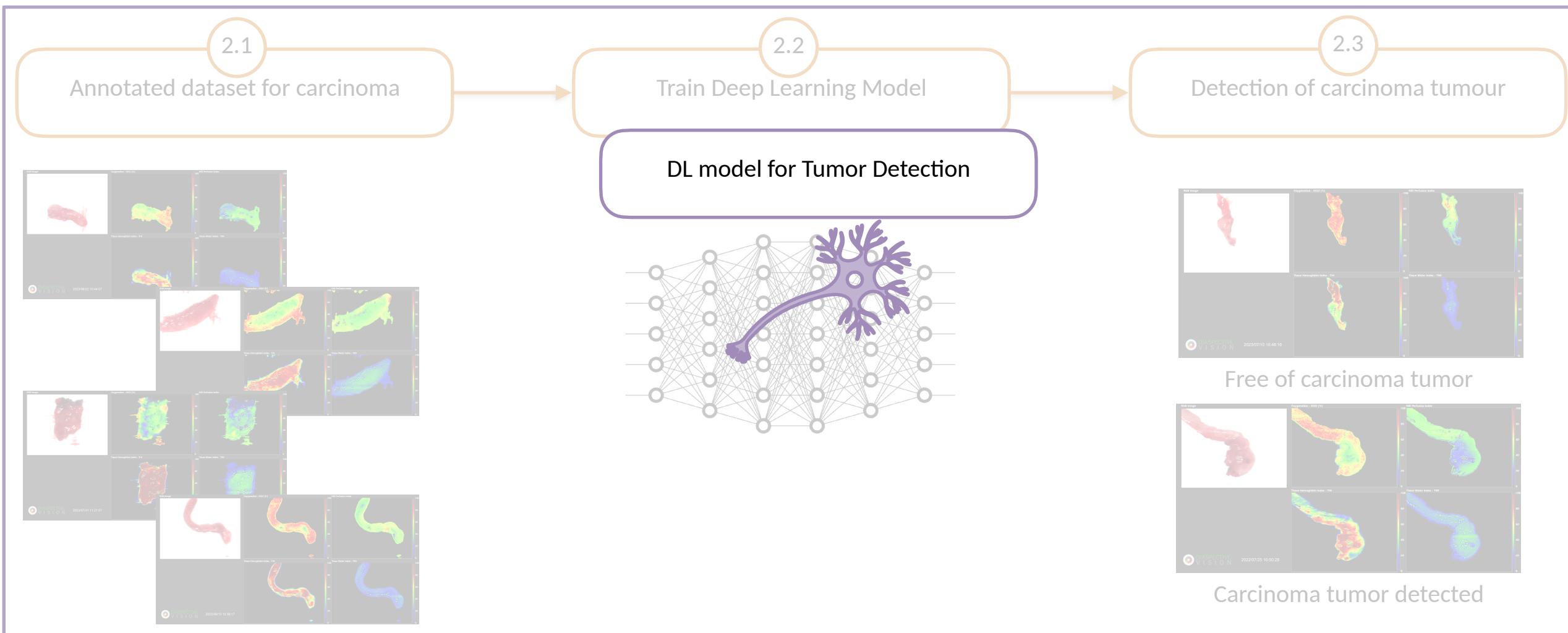
Our strategy

Tumor detection training on Ex-Vivo Frozen Section



Our strategy

Tumor detection training on Ex-Vivo Frozen Section

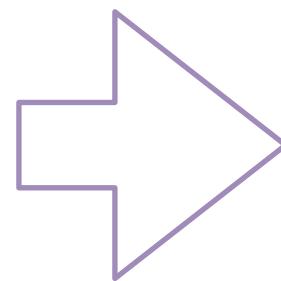
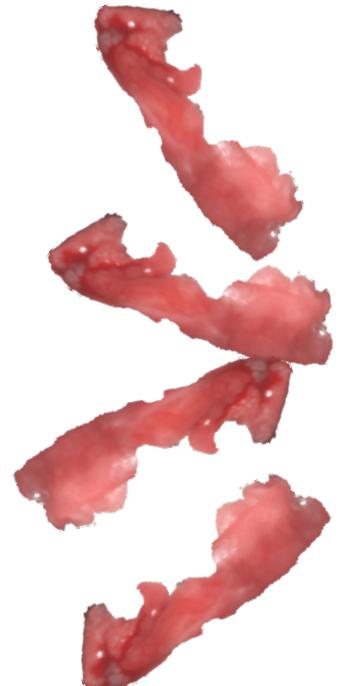
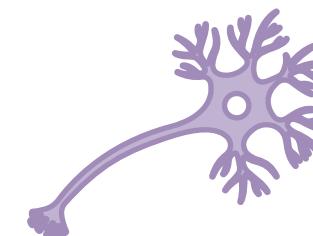


Our strategy

In-Vivo Oral Cavity Scan: Real world scenario

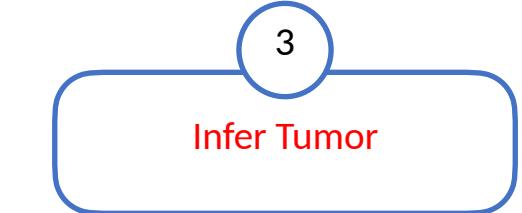
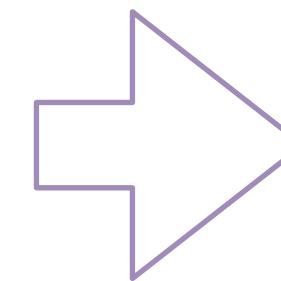
1.3

Data Transformation

Deep Learning model for
Tumor Detection

3

Infer Tumor





1.1 Mucosa segmentation

1.1.1

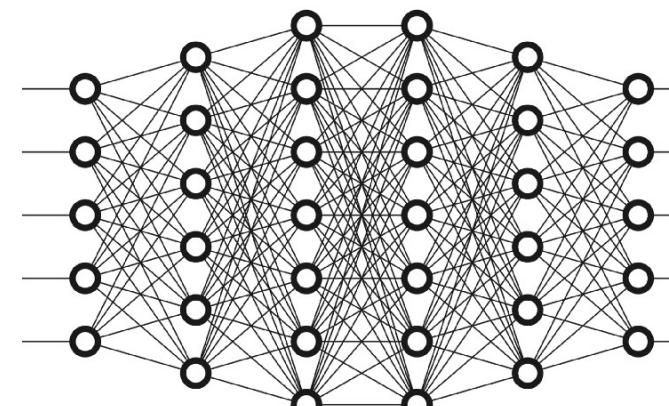
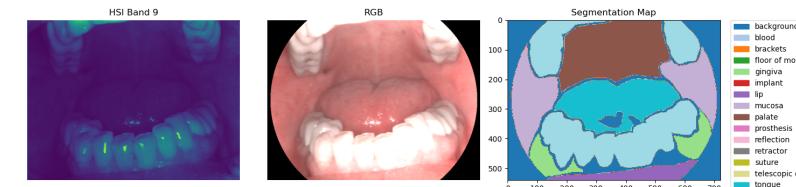
Annotated dataset

1.1.2

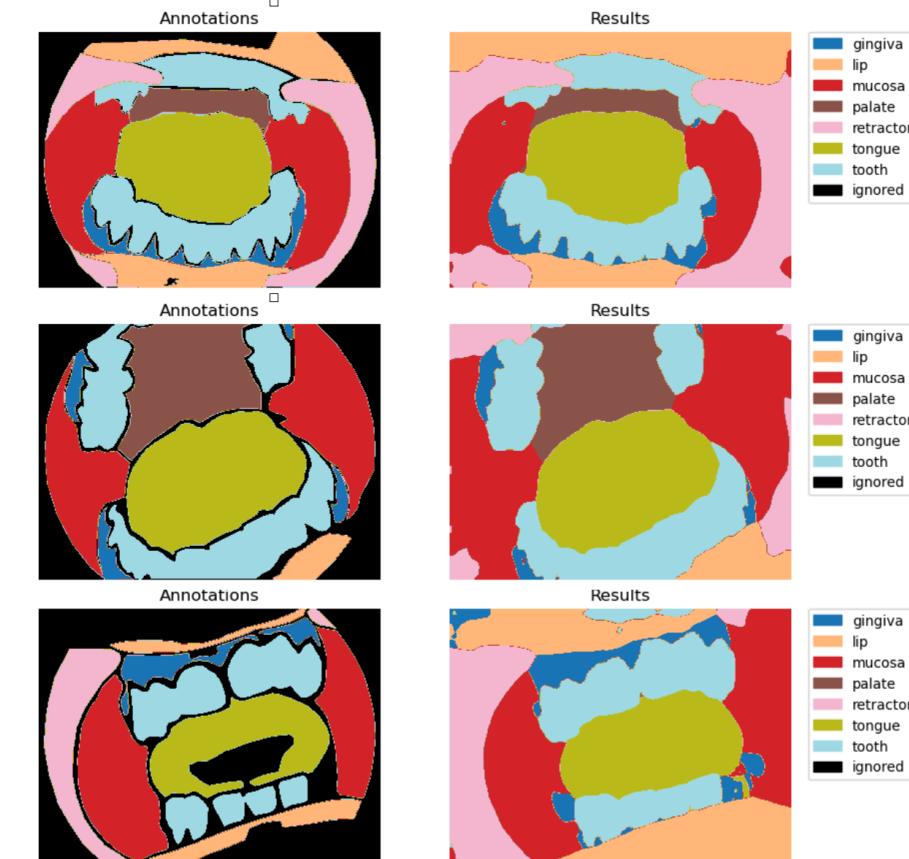
Train Deep Learning Model

1.1.3

Semantic Segmentation



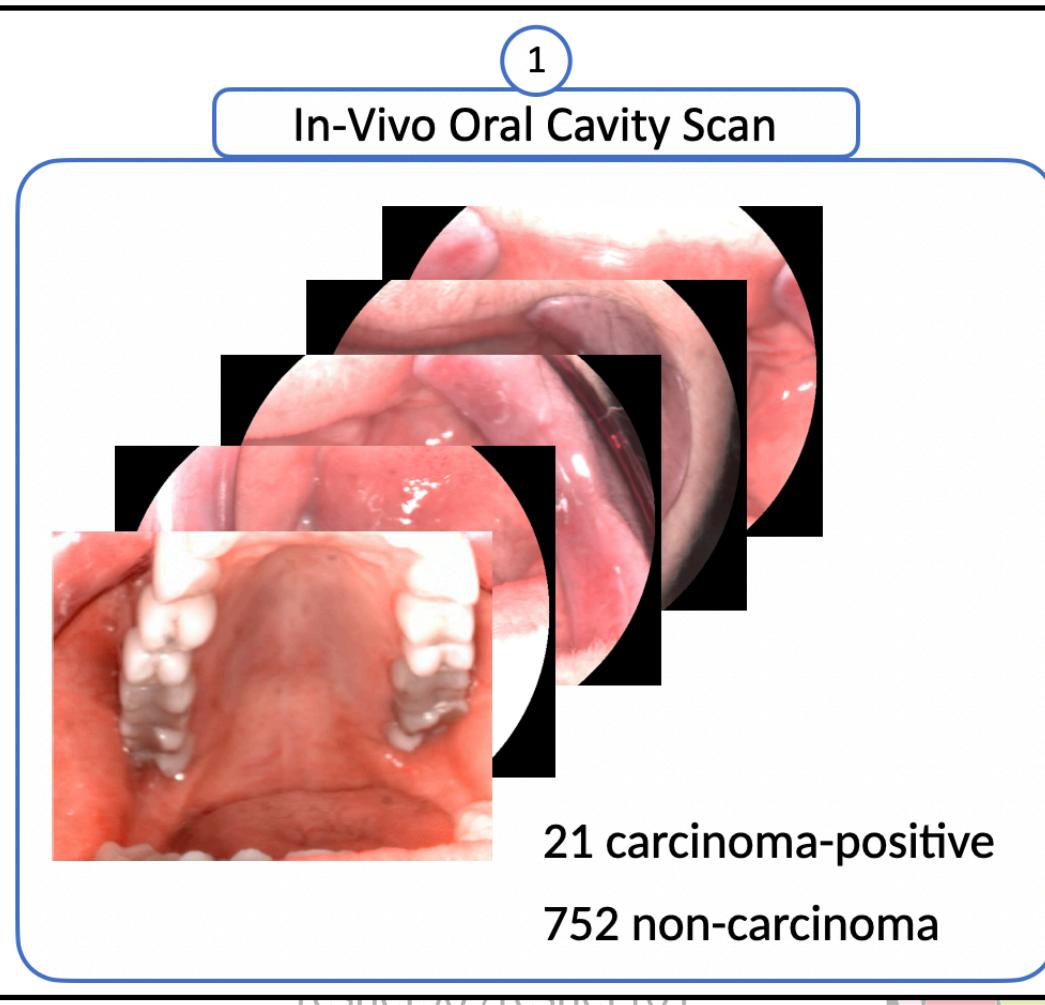
deeplabv3
+
resnet50 / resnet101



1.1.1

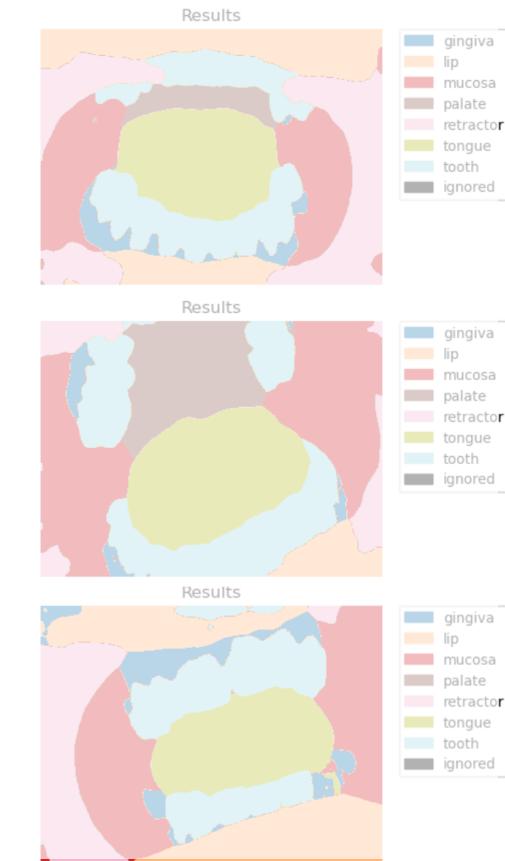
Annotated dataset

1.1 Mucosa segmentation



1.1.3

Semantic Segmentation



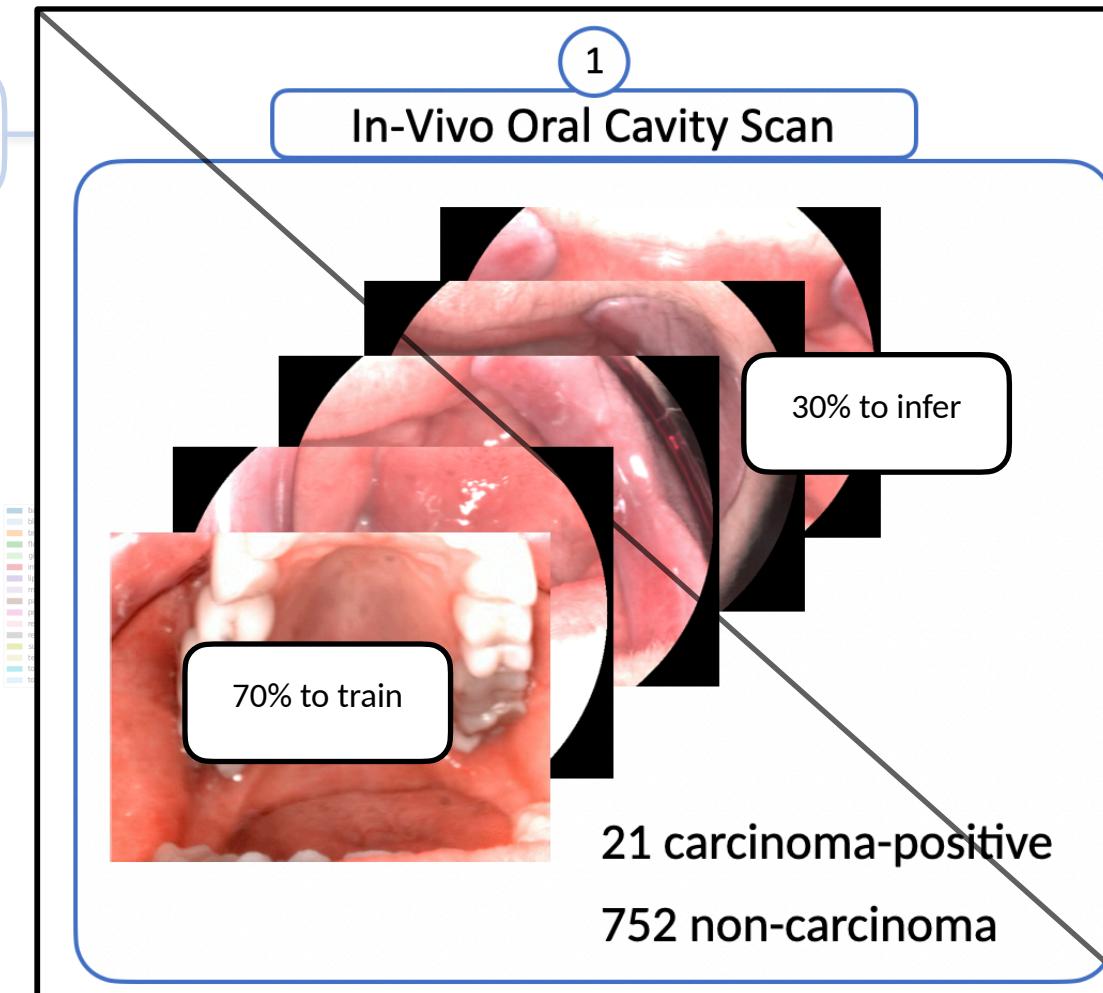
1.1.1

Annotated dataset



1.1 Mucosa segmentation

1.1



1.1.3

Semantic Segmentation



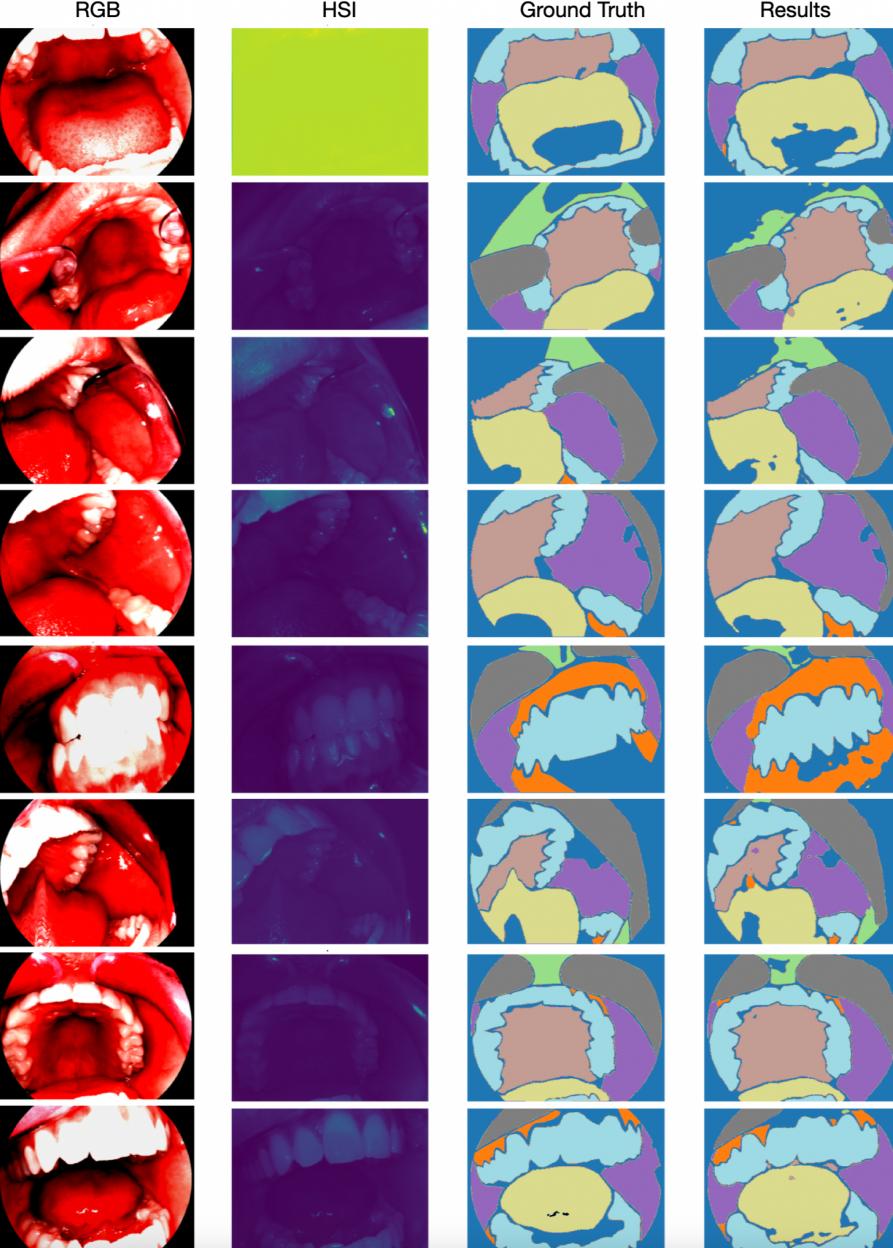
1.1 Mucosa segmentation

Results

DeepLabv3 ResNet-50 / DeepLabv3 ResNet-101

Model	Precision	Recall	F1 Score			
Overall	0.849	0.851	0.861	0.863	0.855	0.857
Palate	0.886	0.874	0.918	0.929	0.901	0.900
Retractor	0.936	0.942	0.940	0.938	0.938	0.940
Mucosa	0.915	0.914	0.914	0.916	0.914	0.915
Tongue	0.878	0.872	0.888	0.892	0.883	0.882
Clutter	0.840	0.844	0.848	0.849	0.844	0.846
Tooth	0.862	0.871	0.937	0.936	0.898	0.902
Lip	0.723	0.730	0.695	0.683	0.709	0.706
Gingiva	0.756	0.757	0.751	0.765	0.754	0.761

- clutter
- gingiva
- lip
- mucosa
- palate
- retractor
- tongue
- tooth
- ignored



2

Carcinoma Detection

2.1

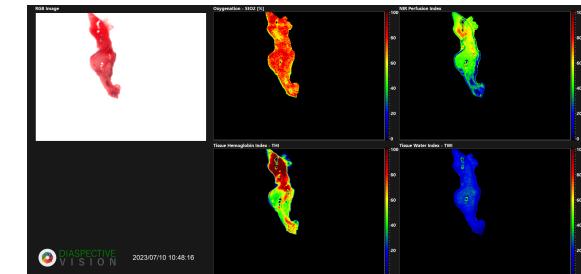
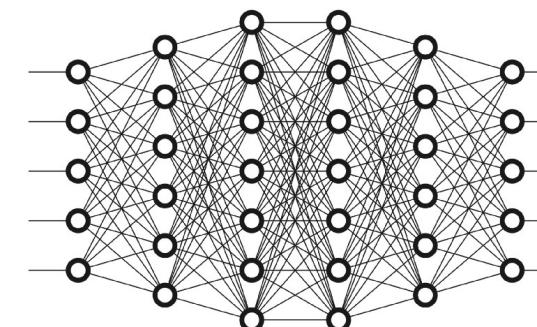
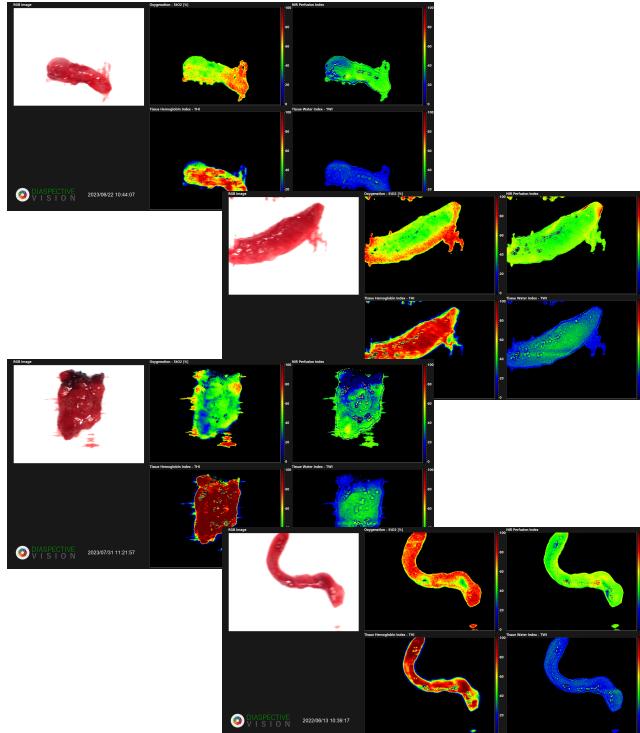
Annotated dataset for carcinoma

2.2

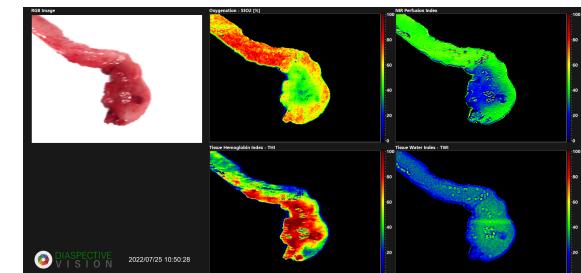
Train Deep Learning Model

2.3

Detection of carcinoma tumour



Free of carcinoma tumor



Carcinoma tumor detected

2

Carcinoma Detection

2.1

Annotated dataset for carcinoma

2.2

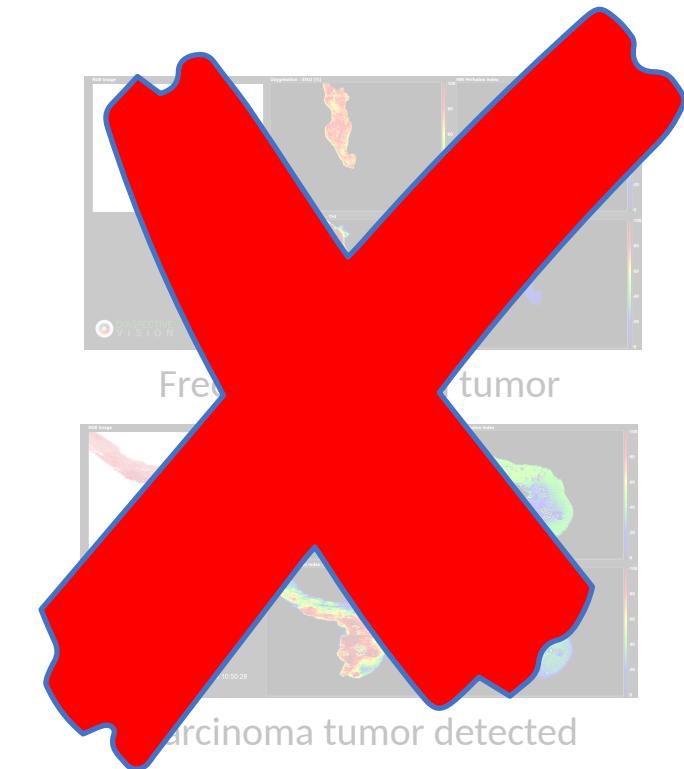
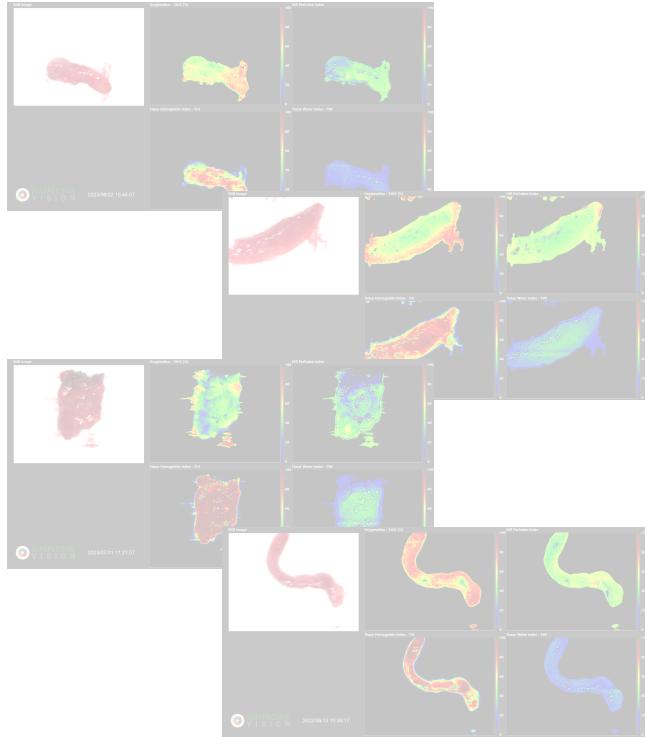
Train Deep Learning Model

2.3

Detection of carcinoma tumour

Models trained and tested

- AlexNet
- ConvNeXt Base
- ConvNeXt Large
- ConvNeXt Small
- ConvNeXt Tiny
- ResNet18
- ResNet34
- ResNet50
- ResNet101
- ResNet152
- ResNeXt50-32x4d
- ResNeXt101-32x8d
- ResNeXt101-64x4d
- VGG11
- VGG11-BN
- VGG13
- VGG13-BN
- VGG16
- VGG16-BN
- VGG19
- VGG19-BN
- Wide ResNet50-2
- Wide ResNet101-2



2

Carcinoma Detection

Annotate

Add Data augmentation

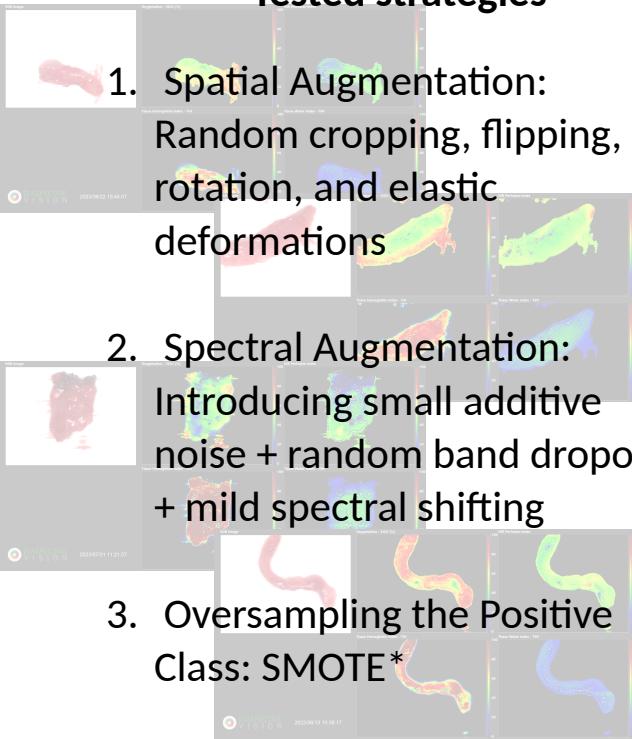
2.2

Train Deep Learning Model

2.3

Detection of carcinoma tumour

Tested strategies



Models trained and tested

- AlexNet
- ConvNeXt Base
- ConvNeXt Large
- ConvNeXt Small
- ConvNeXt Tiny
- ResNet18
- ResNet34
- ResNet50
- ResNet101
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- ResNeXt101-64x4d
- VGG11
- VGG11-BN
- VGG13
- VGG13-BN
- VGG16
- VGG16-BN
- VGG19
- VGG19-BN
- Wide ResNet50-2
- Wide ResNet101-2



2

Carcinoma Detection

2.1

Annotated dataset for carcinoma

2.2

Build new model

Model

2.3

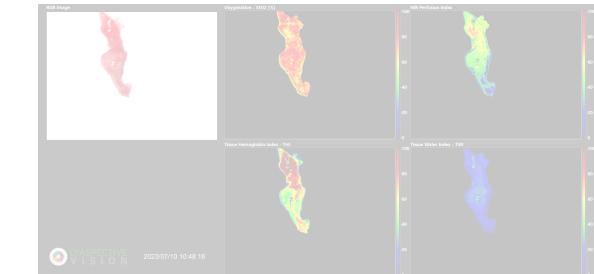
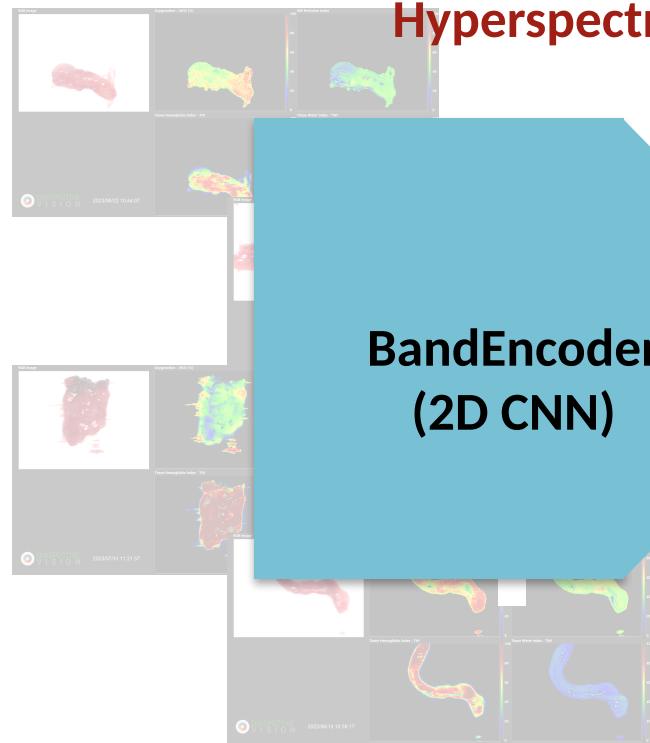
Detection of carcinoma tumour

Hyperspectral 2D+1D CNN for Oral Carcinoma Detection

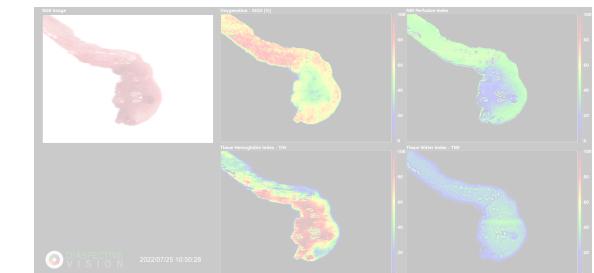
BandEncoder
(2D CNN)

1D Convolution
Across Bands

Global Pooling
& Classification



Free of carcinoma tumor



Carcinoma tumor detected

2

Carcinoma Detection

2.1

Annotated dataset for carcinoma

2.2

Build new model

Model

2.3

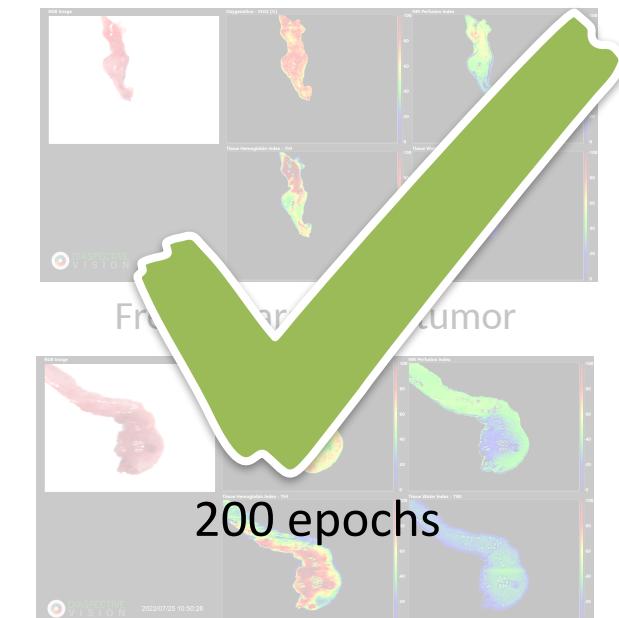
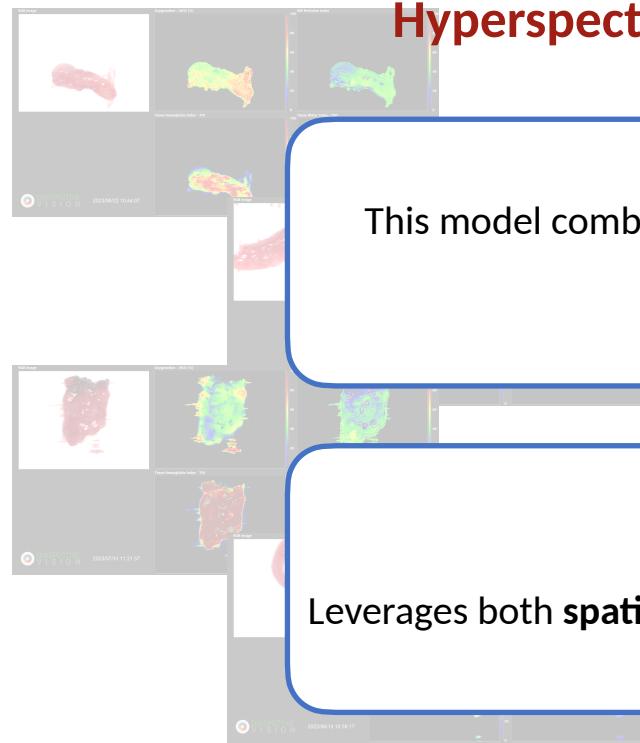
Detection of carcinoma tumour

Hyperspectral 2D+1D CNN for Oral Carcinoma Detection

This model combines **spatial and spectral analysis** for hyperspectral imaging (HSI) classification

Key Strength:

Leverages both **spatial structure** (within each band) and **spectral variation** (across bands) for **early small dataset**.

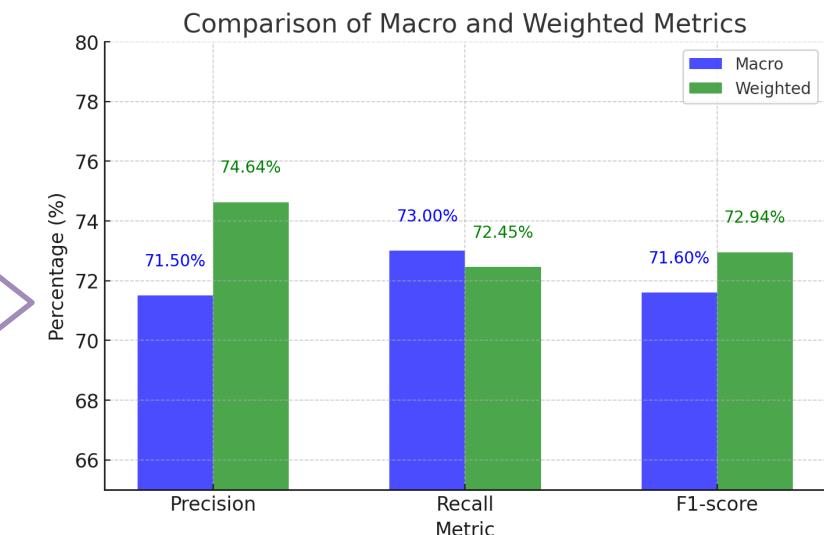
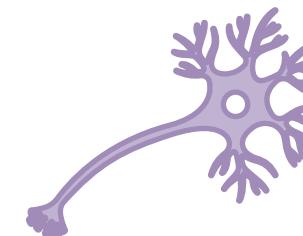


Detection of carcinoma tumor

Results

2.3

Data Transformation

Deep Learning model for Tumor
Detection

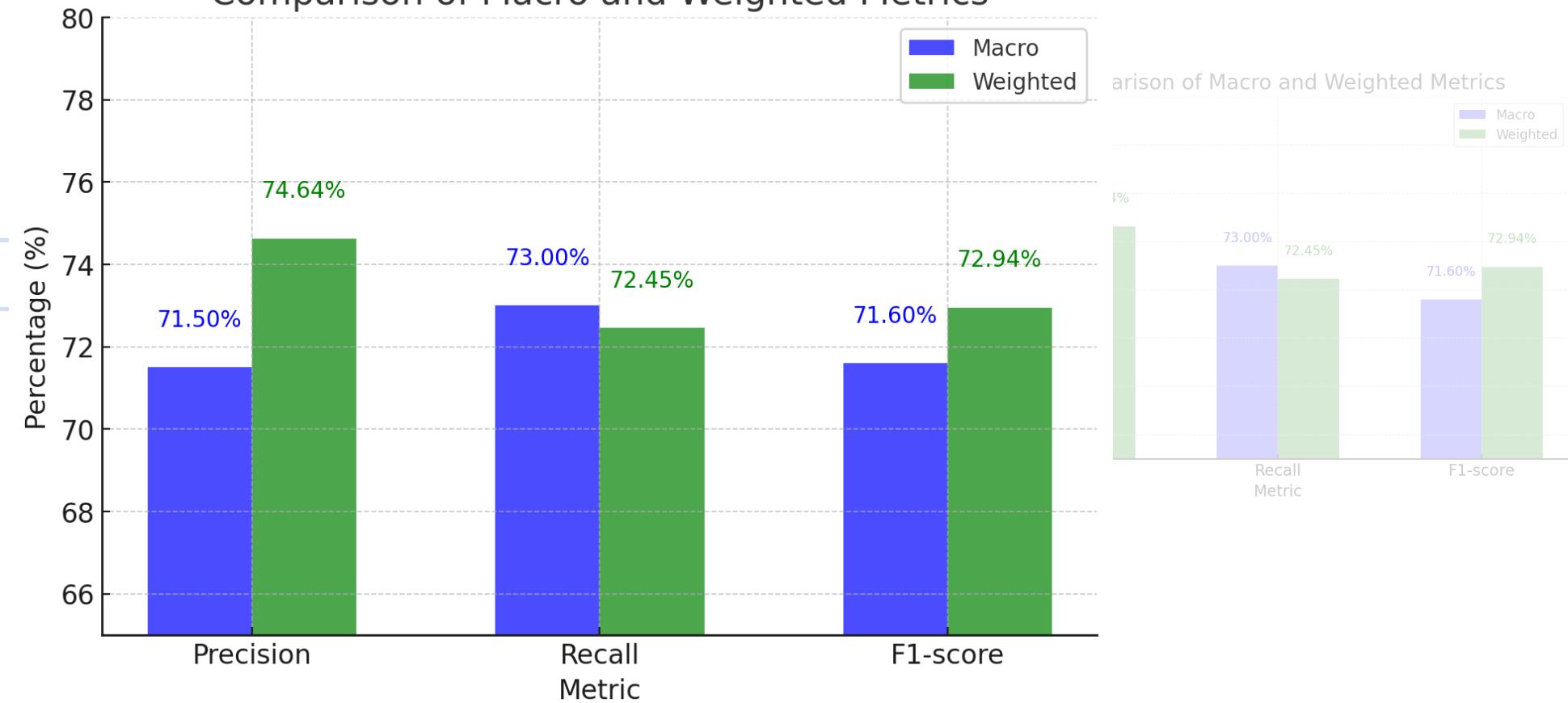
3. Detection of carcinoma tumor Results

2.3

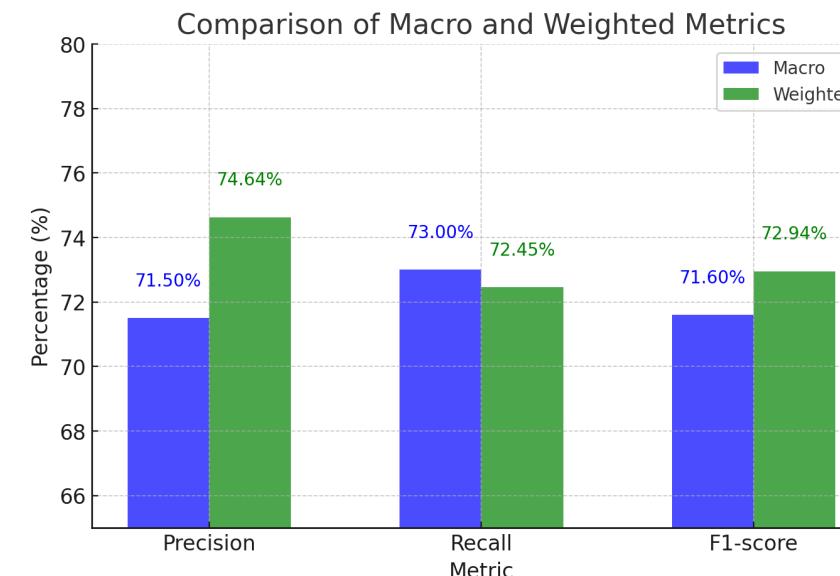
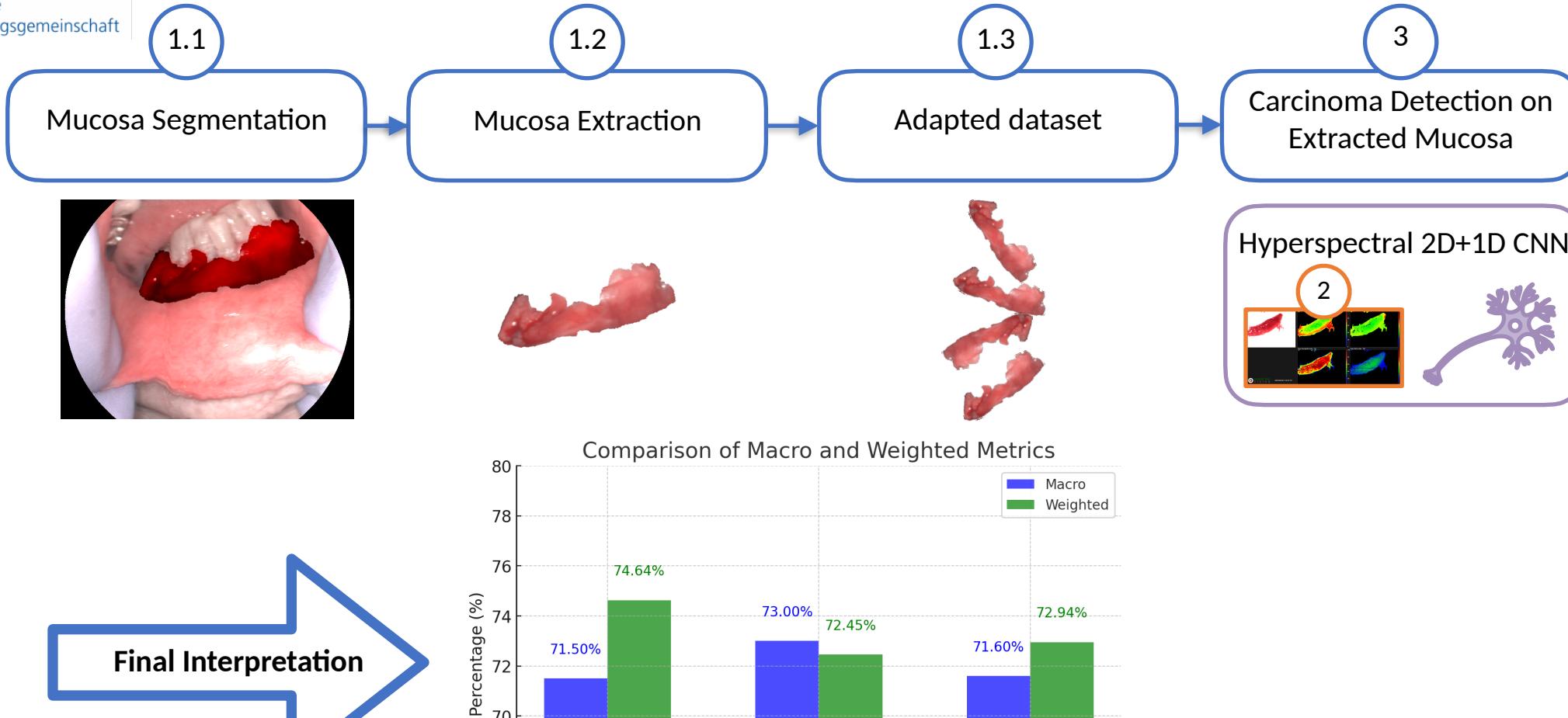
Data Transformation



Comparison of Macro and Weighted Metrics



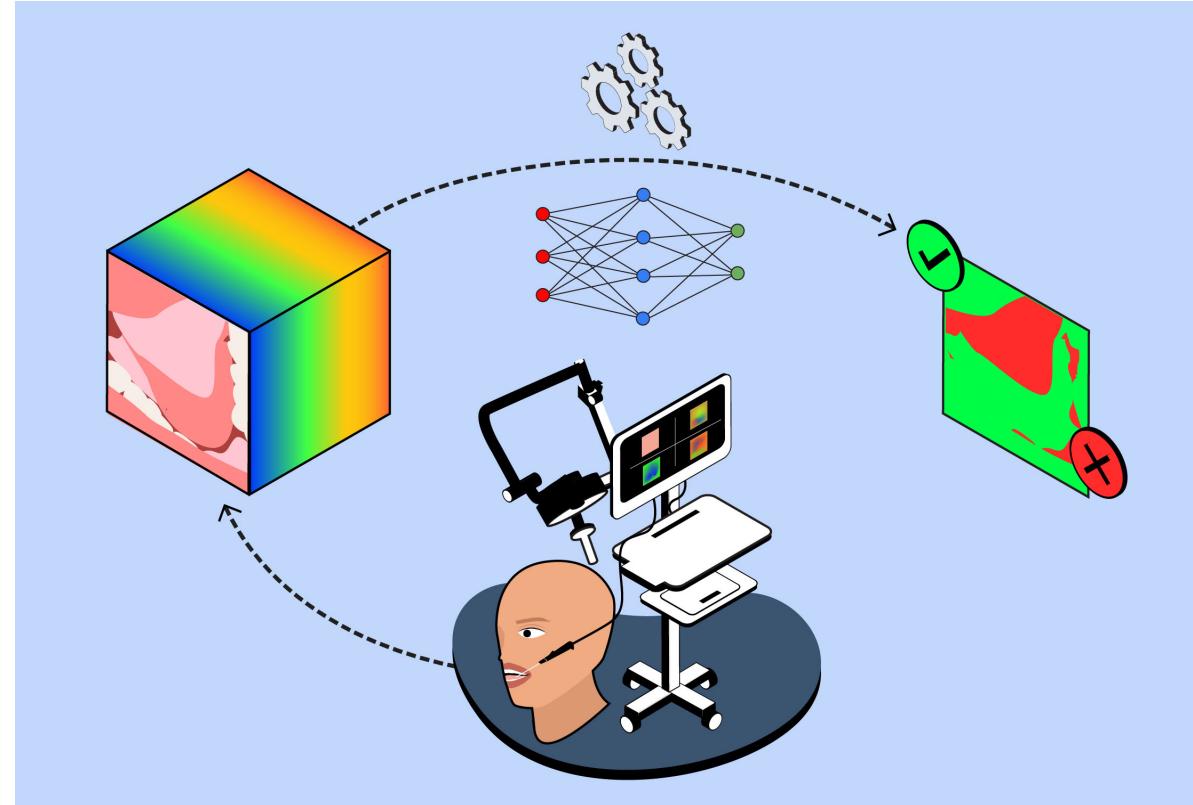
Conclusion



Enhancing Oral Diagnostics with Hyperspectral Imaging and Deep Learning



[https://i3mainz.hs-mainz.de/
projekte/oralhype2-0/](https://i3mainz.hs-mainz.de/projekte/oralhype2-0/)



www.linkedin.com/in/jean-jacques-ponciano

Thank you

jean-jacques.ponciano@hs-mainz.de