

# Classification of Confocal Laser Endomicroscopic Images of the Oral Cavity to Distinguish Pathological from Healthy Tissue

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## Structure

- Motivation
- Background
- Data & Methods
- Experiments and Results
- Summary & Conclusion

# Motivation

## Motivation – Cancer of the Oral Cavity

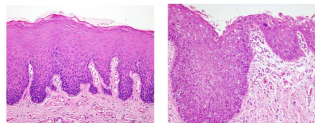
Sixth most common kind of cancer

Problems of diagnosis

- subjectivity of physician
- histological analysis
- surgical resection



Early diagnosis  $\Rightarrow$  difficult!



## Motivation – Cancer of the Oral Cavity

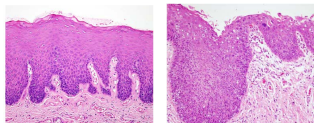
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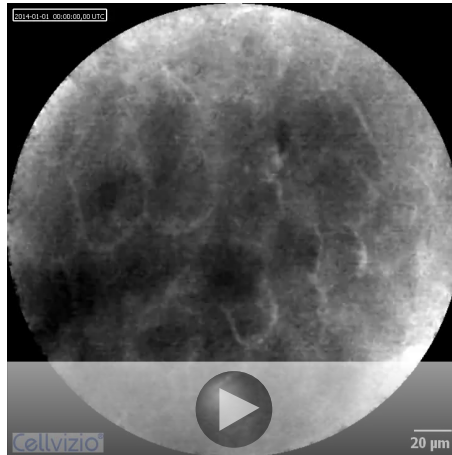


Early diagnosis  $\Rightarrow$  difficult!



$\Rightarrow$  other solutions?

## Motivation – Initial Problem



## Motivation – Objective

Overall: separate pathological from healthy images

Benefits:

- objective method to support the physician
- supports diagnosis & finding of the resection site
- time-saving and less harmful for the patient

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Three problems:

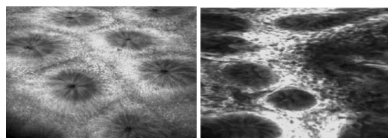
1. creation of image database ✓
2. annotation of images ✓
3. classification of images **today's topic**



## Motivation – State of the Art

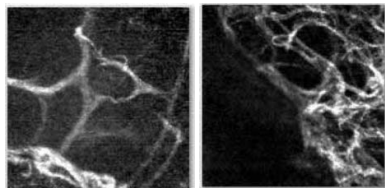
Couceiro et al. [Couceiro, 2012]

- gastrointestinal tract
- arrangement of glands
- Scale Invariant Feature Transform (SIFT)



Désir et al. [Désir, 2012]

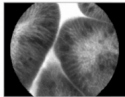
- distal lung
- texture description
- Local Binary Patterns (LBP), SIFT



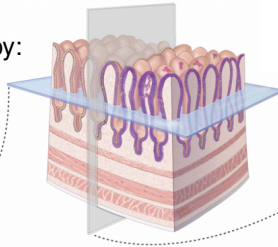
# Background

## Background – Optical Biopsy

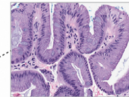
In vivo  
confocal microscopy:  
en face view



Example of  
Optical Biopsy



Conventional  
histology:  
Transvers section

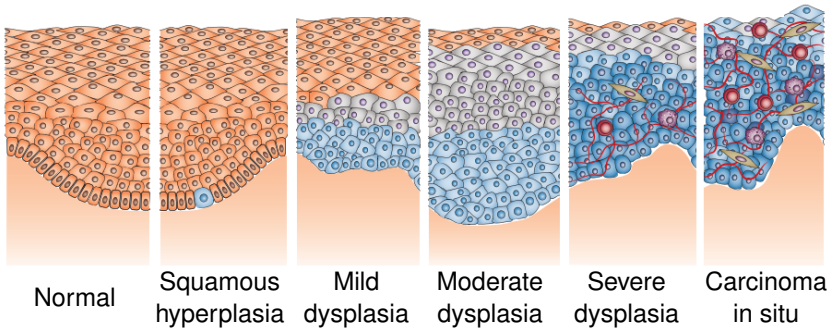


Example of Physical  
Biopsy Image

Confocal LaserEndomicroscopy (CLE) allows real time visualization of  
epithelial layer **in vivo**!

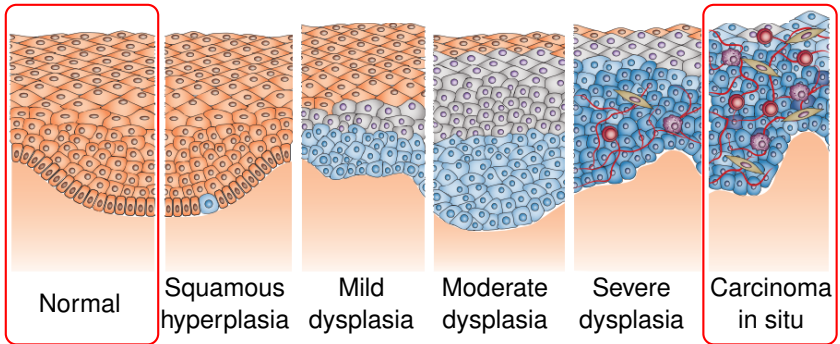
## Background – Carcinogenesis

### Development stages of oral cancer



## Background – Carcinogenesis

### Development stages of oral cancer



# Data & Methods

## Data – Patient & Image Database

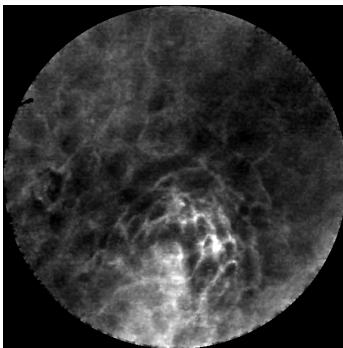
### Patient Data

|              | Control Group | Patient Group  |
|--------------|---------------|----------------|
| Gender (m/f) | 1/-           | 1/1            |
| Age (years)  | 30            | 63.5 $\pm$ 2.1 |

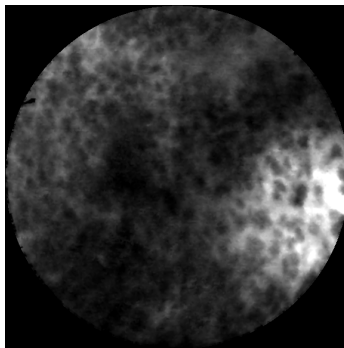
### Image Database

| Location             | Control | Patient 1 | Patient 2 |
|----------------------|---------|-----------|-----------|
| Alveolar Ridge (h/c) | 71/-    | 94/45     | 41/-      |
| Buccal mucosa (h/c)  | -/-     | 32/15     | -/-       |
| Lingual mucosa (h/c) | -/-     | -/-       | 29/27     |

## Data – Image Examples



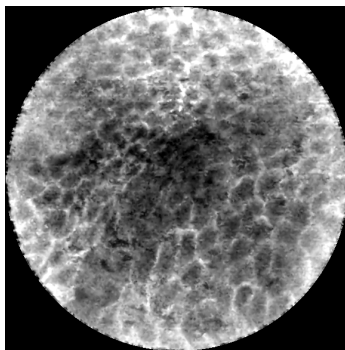
(a) Healthy



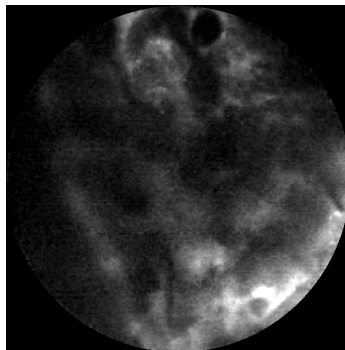
(b) Carcinoma



## Data – Image Examples



(a) Healthy

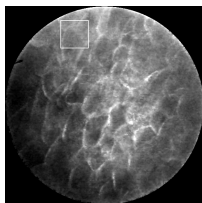


(b) Carcinoma

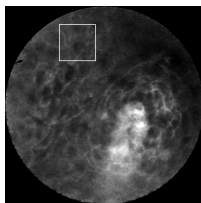
## Methods – Classification Algorithm

### Subdivide images

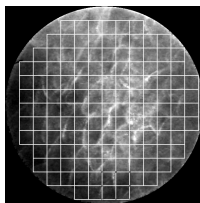
- 110/51 rectangular patches  $\Rightarrow$  precalculated coordinates
- sidelength 80/105 px
- step length  $0.5 \times$  side length  $\Rightarrow$  50 % overlap in x-direction



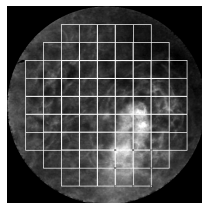
(a)  $1 \times 80$



(b)  $1 \times 105$



(c)  $110 \times 80$

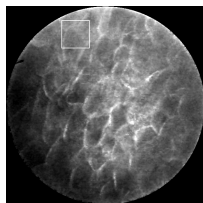


(d)  $51 \times 105$

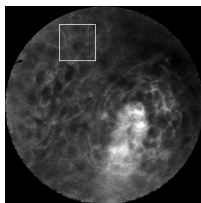
## Methods – Classification Algorithm

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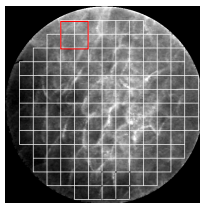
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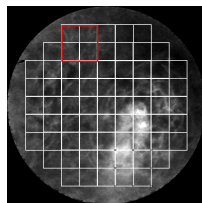
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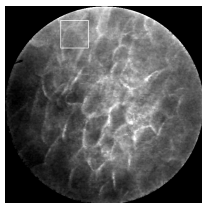


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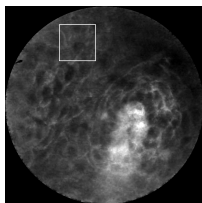
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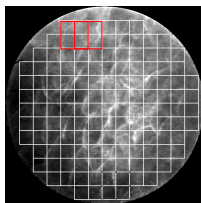
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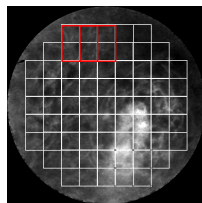
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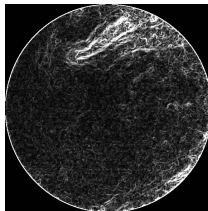
## Methods – Extracted Features

### Histogram features

- frequency of gray level occurrences
- no information of structure
- computation of statistics

### Homogeneity features

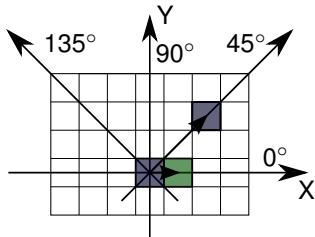
- evaluates gray values
- evaluates edge images
- simple features



## Methods – Extracted Features

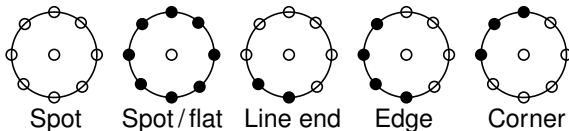
### Grey Level Co-Occurrence Matrices

- frequency of gray values
- geometrical arrangement of gray values
- features by Haralick, GLCM (8/16/32)

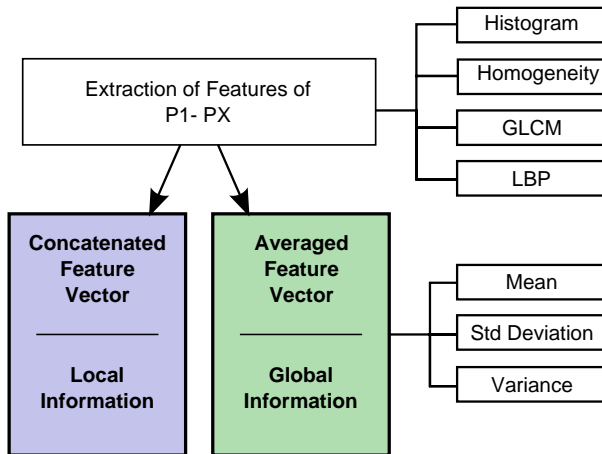


### Local Binary Pattern

- pixel described by binary pattern
- binary patterns describe structures



## Methods – Feature Vectors



## Methods – Classifier & Evaluation

### Classification algorithms

- Support Vector Machine (SVM)
- Random Forest (RF)

### Evaluation methods

- 10-fold crossvalidation
- classification rate (Acc)
- average recall (Rec)

### Software

- CONRAD → image analysis & feature extraction
- Weka → classification tasks



# Experiments and Results

## Experiments – Classification Method

Pathological vs. non-pathological

One patient – same location

- P1 vs. P1 – alveolar ridge
- P1 vs. P1 – buccal
- P2 vs. P2 – lingual

All subjects – all locations

Between subjects – same location

- P1 vs. P2 – alveolar ridge
- P1 vs. P2 & Ctrl – alveolar ridge

⇒ Acc / Rec: 95.8 % / 93.3 %

## Experiments – Classification Method

Pathological vs. non-pathological

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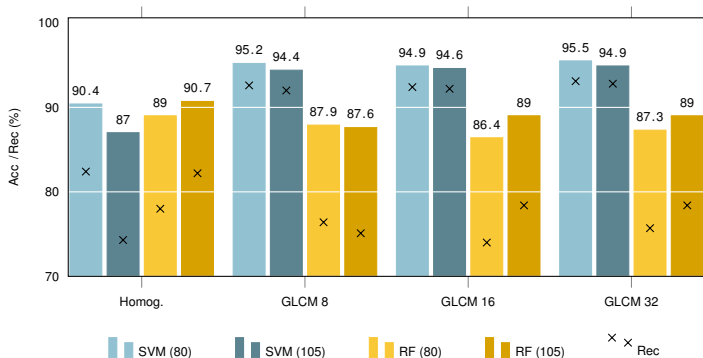
All subjects – all locations

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... best performing features?  
... best performing feature vector?

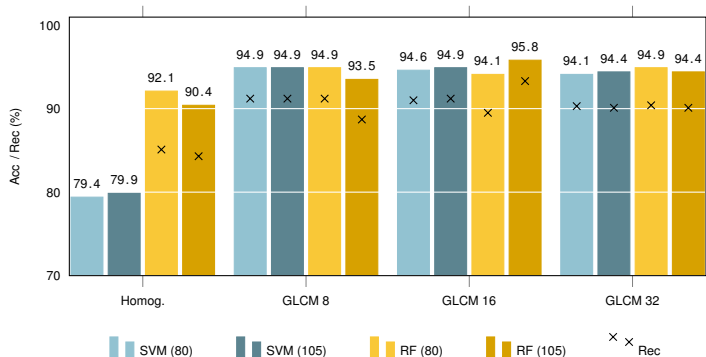
## Results – All Subjects all Locations

### Comparison of feature vector – concatenated feature vector results



## Results – All Subjects all Locations

### Comparison of feature vector – average feature vector results



# Summary & Conclusion

## Summary & Conclusion

Objective: separate pathological from healthy images

Three problems:

1. creation of image database ✓
2. annotation of images ✓
3. classification of images ✓ **95.8 % / 93.3 %**

Benefits:

- objective method to support the physician
- supports finding of the resection site
- time-saving and less harmful for the patient

⇒ monitor progress of cancer?

**Thank you for your attention!**

**Questions?**



## Backup – Hardware

### Cellvizio Gastro-flex UHD

|                                      |                   |
|--------------------------------------|-------------------|
| Imaging rate (frames/s)              | 12.8              |
| Probe diameter (mm)                  | 2.7               |
| Depth of imaging ( $\mu\text{m}$ )   | 55-65             |
| Lateral resolution ( $\mu\text{m}$ ) | 1                 |
| Field of view ( $\mu\text{m}$ )      | $\varnothing$ 240 |
| Image resolution (px)                | $576 \times 576$  |



The End