

# Kap. 5b

## Anwendungen:

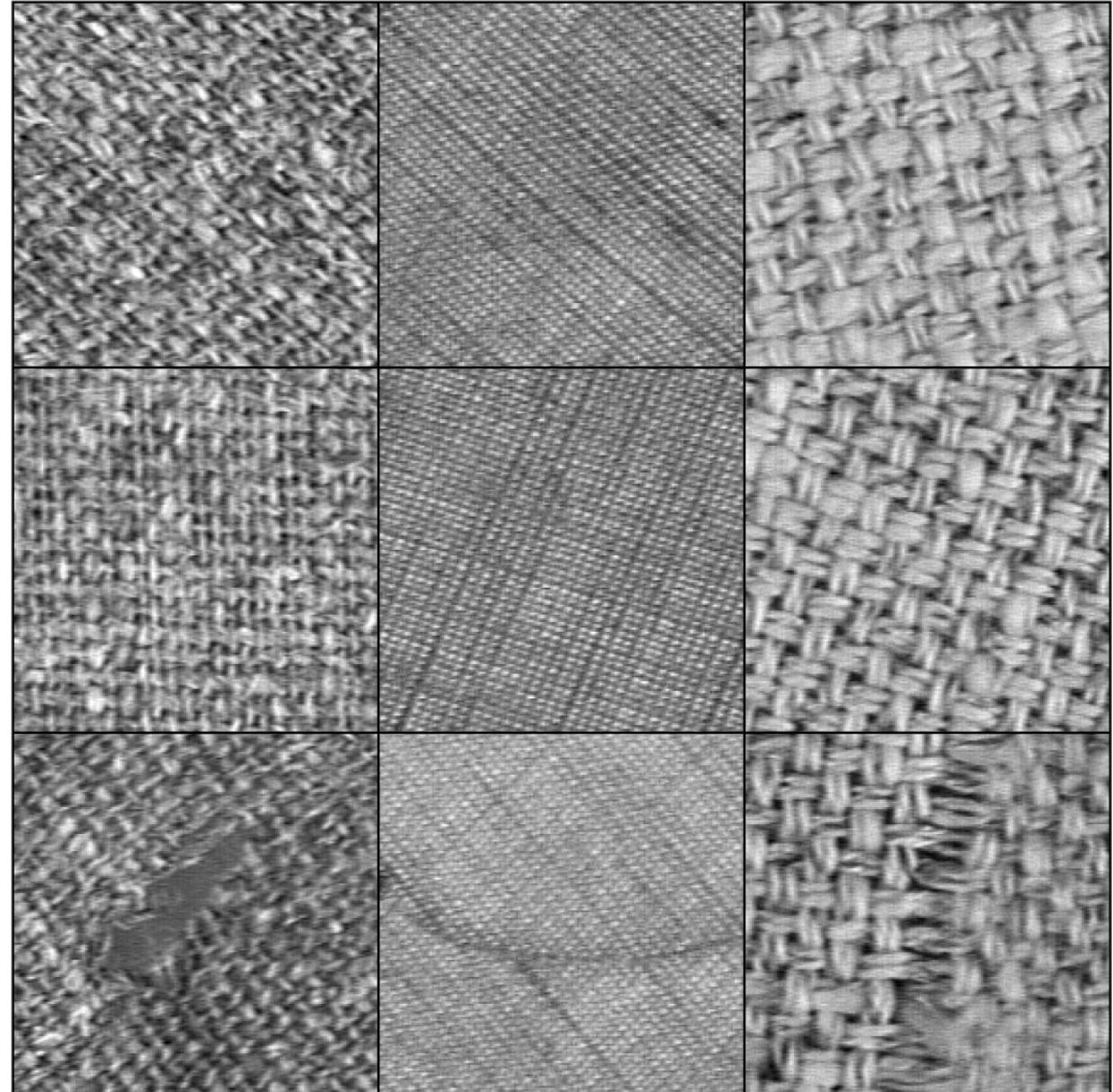
### Invarianten über Gruppenmittel

- Sichtprüfungsaufgaben (Textilien)
- Bildsuchmaschinen (query by example) SIMBA und MICHELscope
- Automatische Klassifikation von Blütenpollen
- Suchen in Proteindatenbanken
- Suchen nach Wasserzeichen in alten Drucken

# Complexity of Search tasks

- Finding an identical object (pixel by pixel) is trivial !
- Challenge for the future:  
Searching in Equivalence classes (up to a very high semantic level) can be very complex and demanding
- Data Bases without an intelligent access and retrieval mechanism is a graveyard for Bits and Bytes!

# Visual inspection of textiles with anisotropic texture



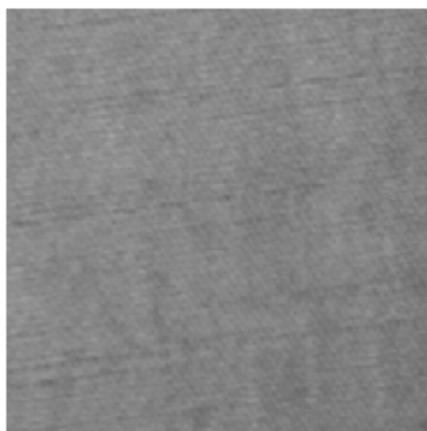
Textildatenbank TILDA

<http://www.informatik.uni-freiburg.de/~lmb/tilda> (ca. 3200 images)

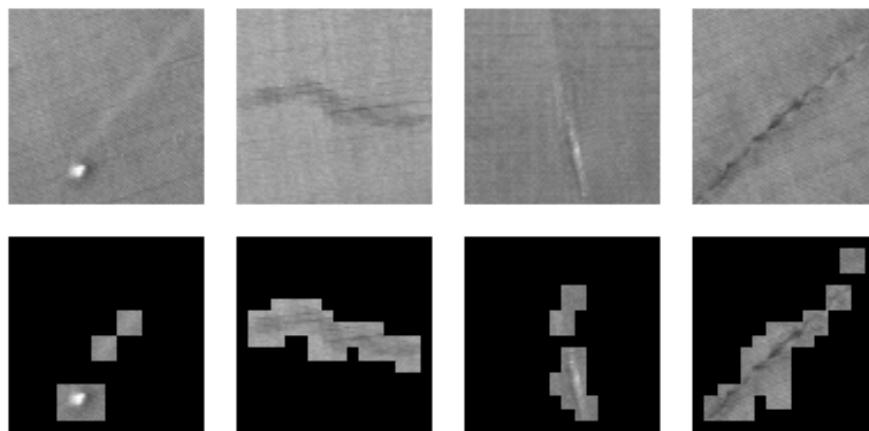
# Texturdefektdetektion bei Textilstoffen

## Segmentierung der Defekte

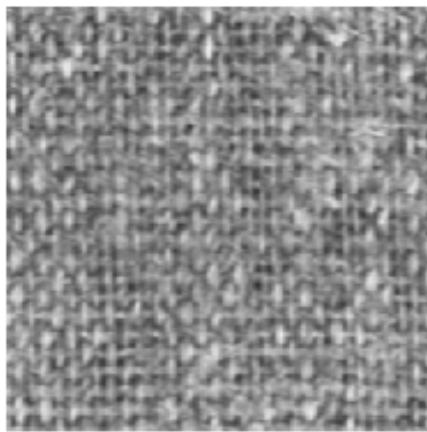
Texturklasse c1r1



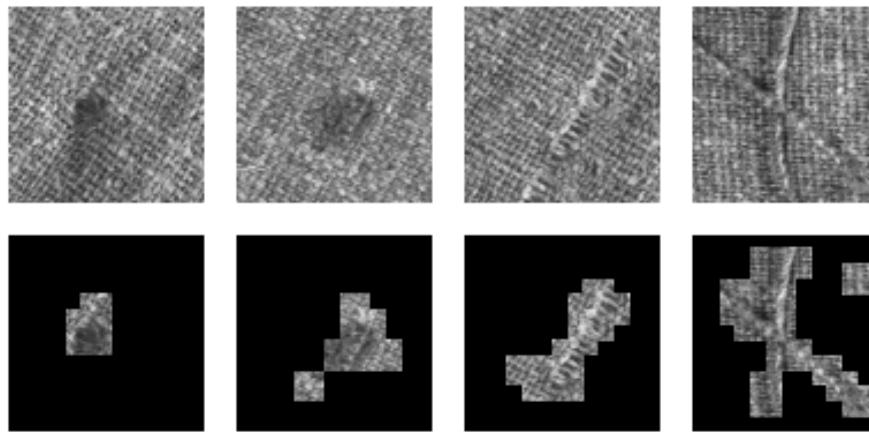
Fehlerklassen  
E1 E2 E3 E4



Texturklasse c2r2



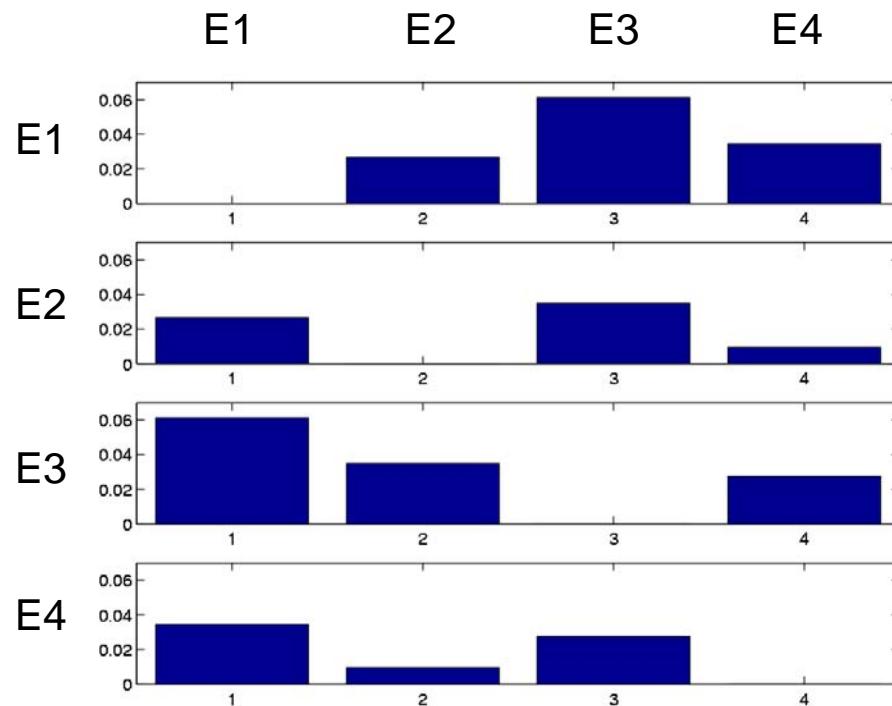
Fehlerklassen  
E1 E2 E3 E4



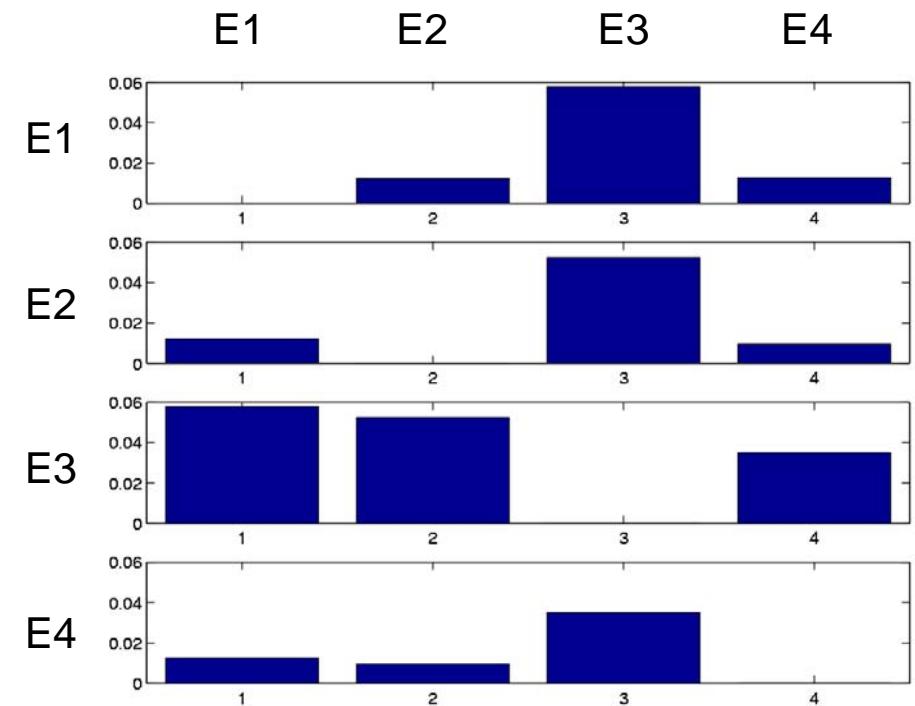
# Texturdefektdetektion bei Textilstoffen

## Diskriminierung der Defektklassen

Klassendistanzen Texturklasse c1r1

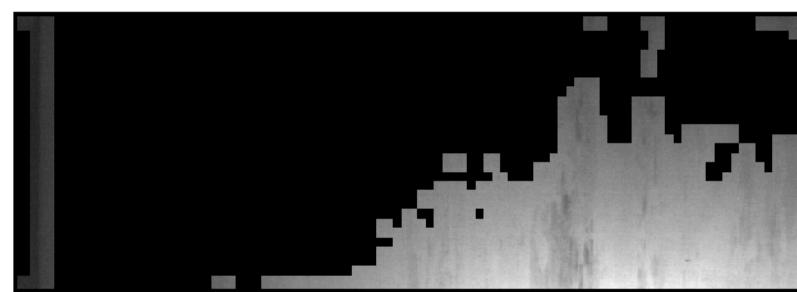
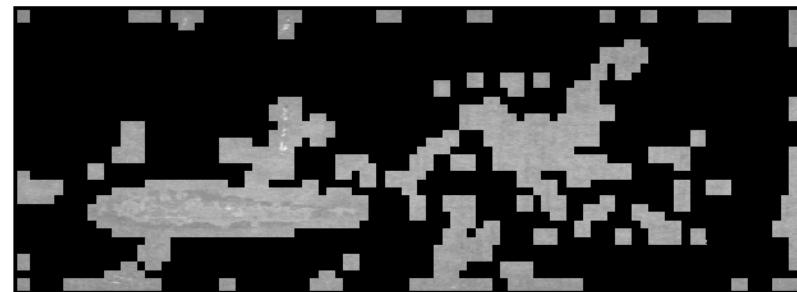
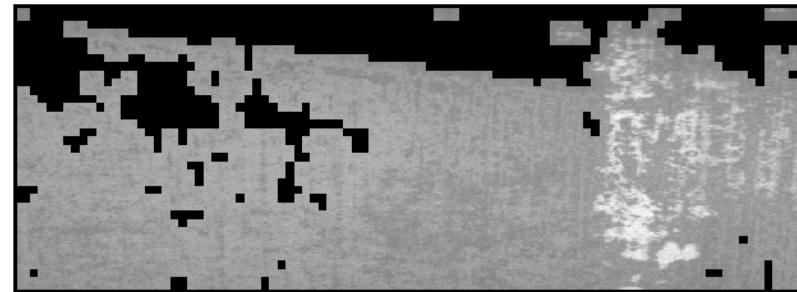
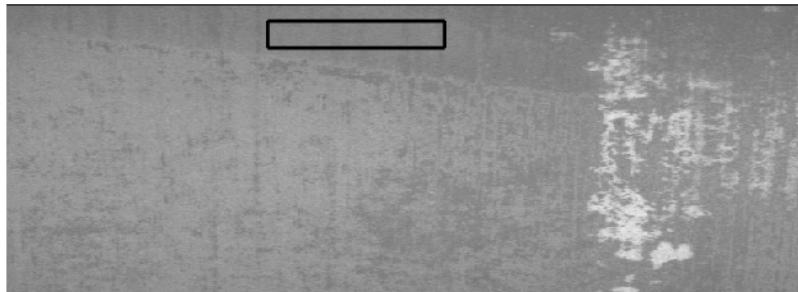


Klassendistanzen Texturklasse c2r2

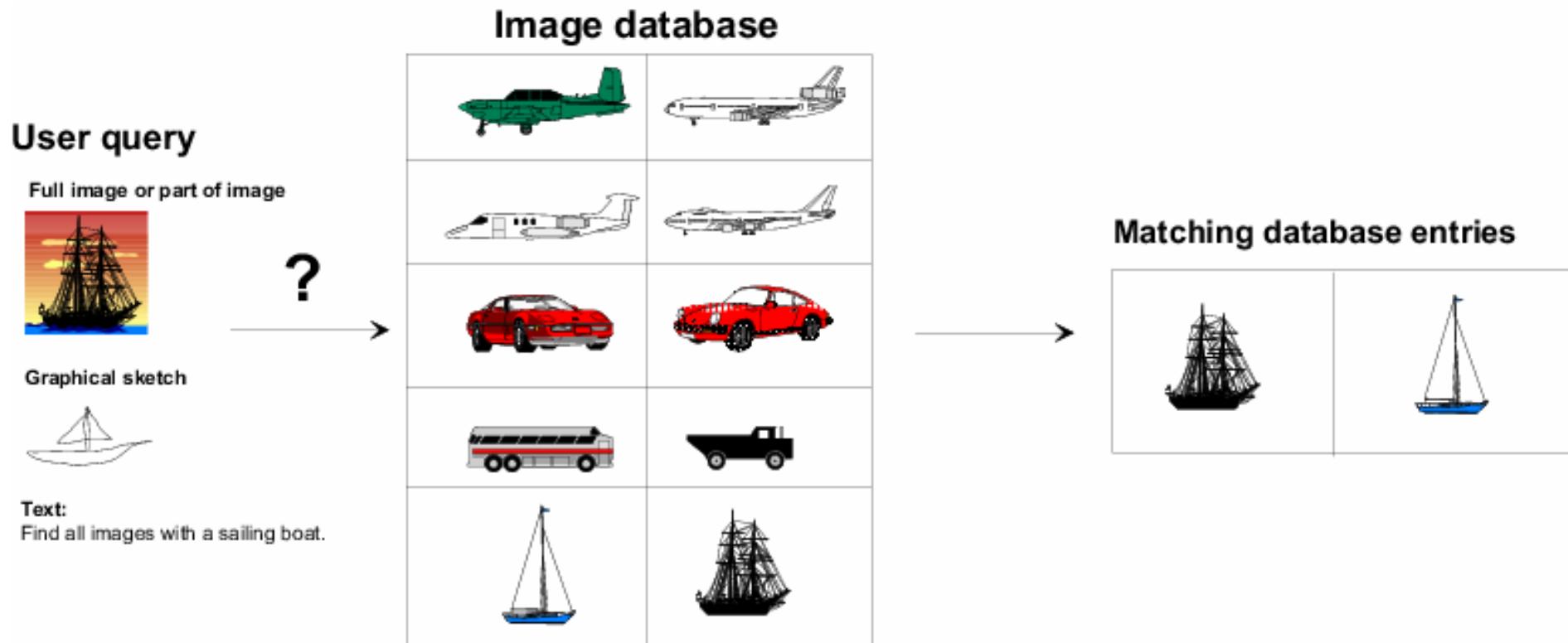


# Texturdefektdetektion auf Stahloberflächen

## Segmentierung der Defekte



# Content based image retrieval



<http://simba.informatik.uni-freiburg.de/>



Query image



0.978



0.828



0.792



0.686



0.674



0.649



0.630



0.595



Query image



0.98



0.82



0.81



0.79



0.78



0.78



0.77



0.76



Query image



0.99



0.79



0.77



0.68



0.68



0.66



0.66



0.66



Query image



0.98



0.82



0.79



0.77



0.74



0.73



0.71



0.71



Query image



0.98



0.79



0.77



0.68



0.68



0.65



0.65



0.63



Query image



0.98



0.80



0.78



0.74



0.70



0.60



0.59



0.57

# SIMBA – Search IMages By Appearance

COLOR only



Search template



96827



63939.4



62383.7



60840.3



59997.4



59316.6



59061



58323.8

# SIMBA – Search IMages By Appearance

RELATIONAL only



Search template



87948.7



85491.7



85163



84896.4



84603.2



84379.5



84065.4



83475.7

# SIMBA – Search IMages By Appearance

COLOR only



Search template



96689.4



81479.3



77883.8



71489.1



70400.4



67864.2



66298.7



66249.5

# SIMBA – Search IMages By Appearance

## COLOR+RELATIONAL



Search template



92393.05



81203.6



77685.85



75091.55



71376.85



70761.3



69292.6



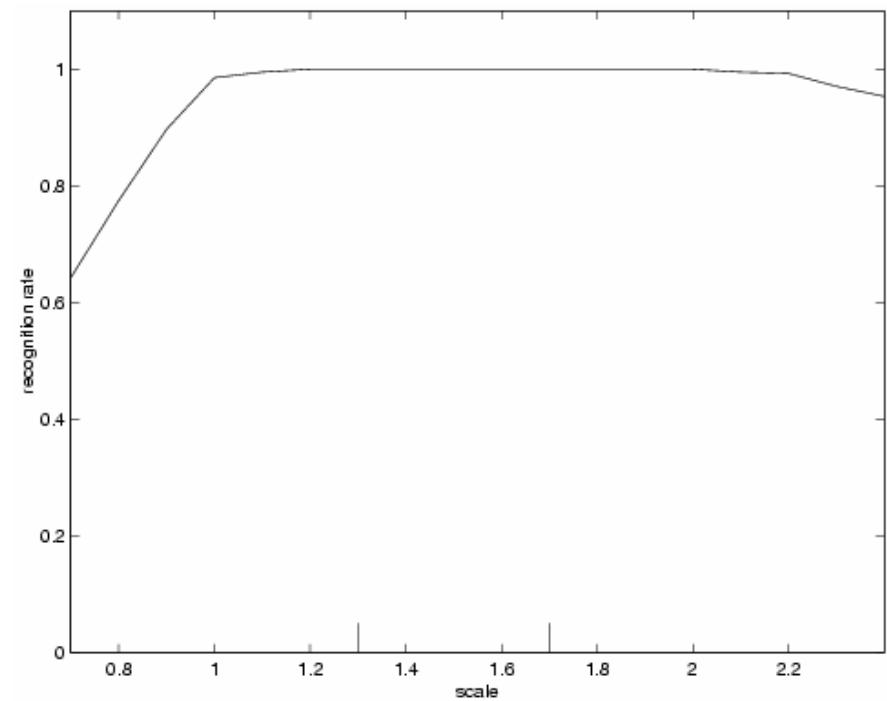
68030.95

# Experiments



# Robustness to scaling

- Database as before,  
grayvalue features only
- Kernels  $\mathbf{M}(1,0)\mathbf{M}(0,2)$ ,  
 $\mathbf{M}(2,0)\mathbf{M}(0,4)$
- Images reduced/zoomed  
(bilinear interpolation) to  
scales from 0.7 to 2.4
- Database contains scales  
1.3 and 1.7 only



→ Good recognition from scale 1 to 2.2, i.e. scale 4.8 in area

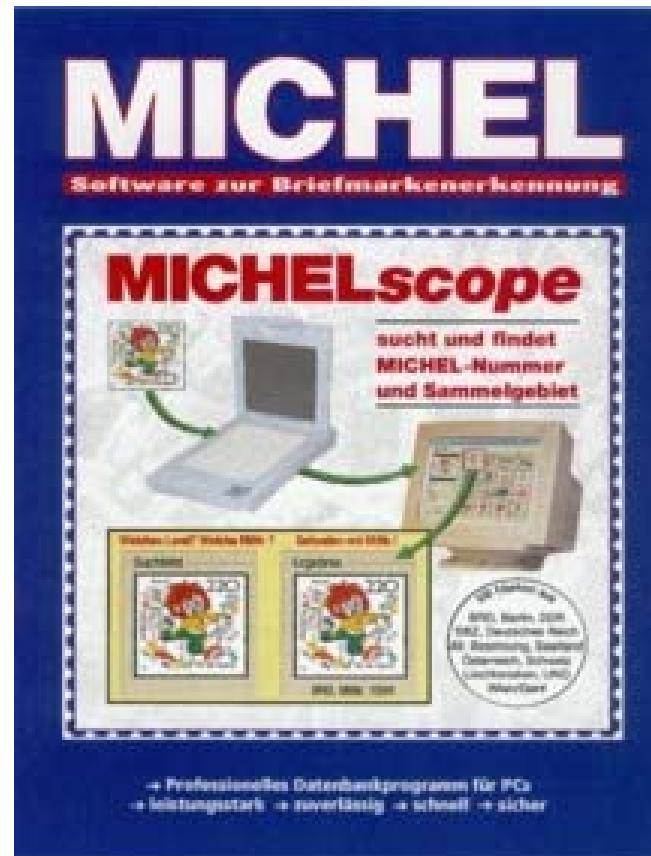
# SIMBA

## Searching Images by Appearance

<http://simba.informatik.uni-freiburg.de/>

# MICHELscope

A search engine for philatelists



## Suchbild



## Ergebnisse



Brd0644



Brd0638



Brd0732



Brd0689



Brd0729



Brd0636



Brd0727



Brd0637



Brd0645



Brd0640



Brd0642



Brd0635



Brd0691



Brd0730



Brd0639



Brd0643

## Suchmethode

Farbe

Textur

Motiv

Größe

Seitenverhältnis

## Datenbank

BRD

Suche starten

## Suchbild



## Ergebnisse



Brd0644



Brd0637



Brd0689



Brd0636



Brd1138



Brd0489



Brd0727



Brd0502



Brd0494



Brd0854



Brd0193



Brd0492



Brd0191



Brd0635



Brd0168



Brd0857

## Suchmethode

Farbe



Textur



Motiv

Größe

Seitenverhältnis

## Datenbank

BRD

Suche starten

## Suchbild



## Ergebnisse



Brd0523



Brd0485



Brd0538



Brd0232



Brd0408



Brd0409



Brd0486



Brd0447



Brd0540



Brd0164



Brd0449



Brd0266



Brd0386



Brd0173



Brd1610



Brd0487

## Suchmethode

Farbe



Textur

Motiv

Größe

Seitenverhältnis

## Datenbank

BRD

Suche starten

[Datei](#) [Einstellungen](#) [Hilfe](#)

## Suchbild



## Suchmethode

Farbe



Textur

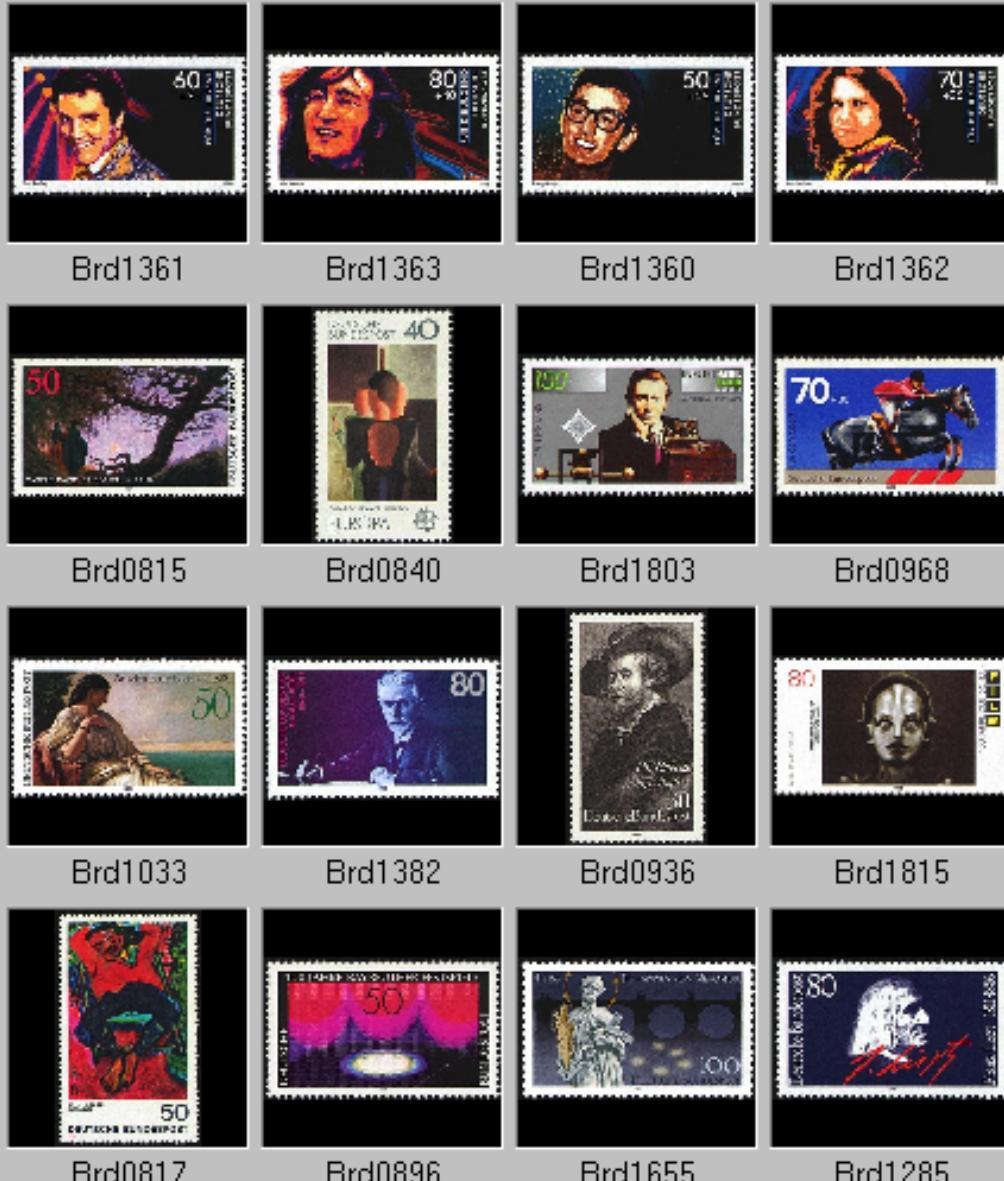
 Motiv Größe Seitenverhältnis

## Datenbank



Suche starten

## Ergebnisse





## Suchbild



## Suchmethode

Farbe



Textur

 Motiv Größe Seitenverhältnis

## Datenbank



Suche starten

## Ergebnisse



Brd1362

Brd1361

Brd1360

Brd1363



Brd0815

Brd1285

Brd1503

Brd1430



Brd1655

Brd1816

Brd1815

Brd0816



Brd0737

Brd1290

Brd0936

Brd1103

## Suchbild



## Ergebnisse



Brd0700

Brd0694

Brd0697

Brd0773



Brd0703

Brd0702

Brd0698

Brd0696



Brd0701

Brd1623

Brd1140

Brd0695



Brd1379

Brd1137

Brd0699

Brd1038

## Suchmethode

- Farbe  Textur
- Motiv
- Größe
- Seitenverhältnis

## Datenbank

BRD

Suche starten



Suchbild



Suchmethode

Farbe



Textur

 Motiv Größe Seitenverhältnis

Datenbank

Deutschland

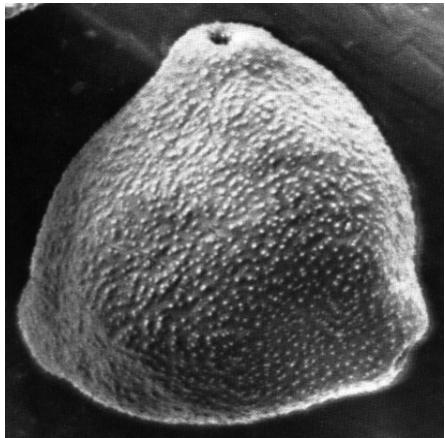
Suche starten

Ergebnisse

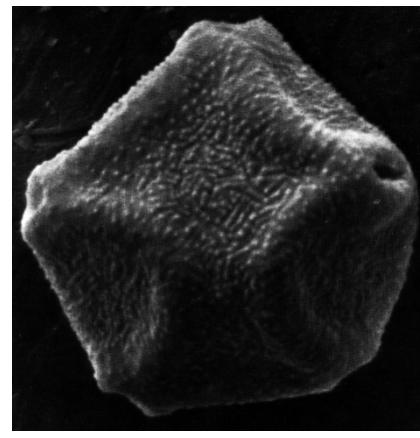


Deutschland Berlin (West) 0042

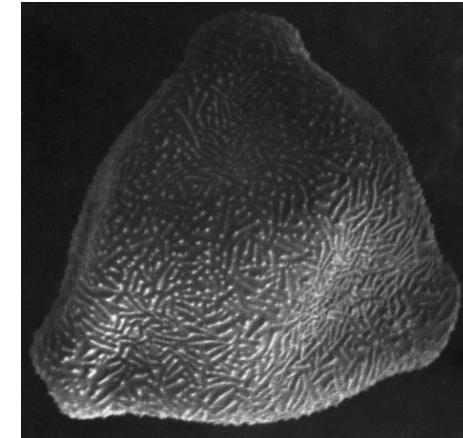
# Projekt zur automatischen Erkennung von Blütenpollen - Elektronenmikroskopische Aufnahmen von Pollen



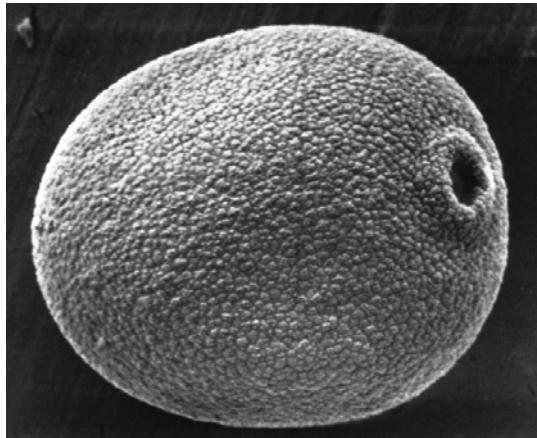
Hasel



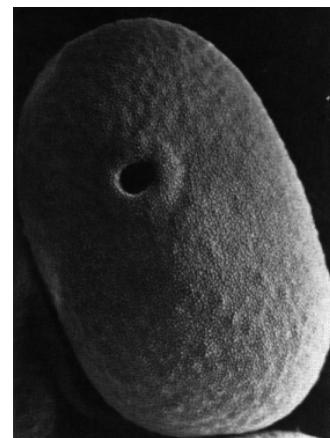
Erle



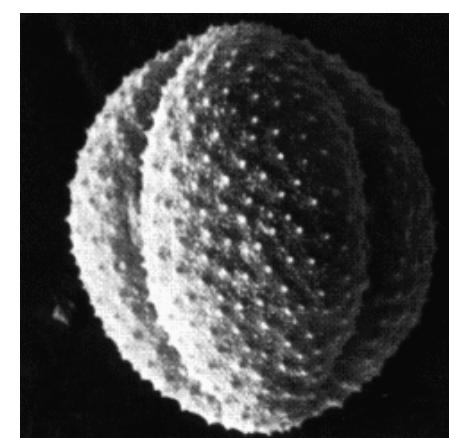
Birke



Gräser



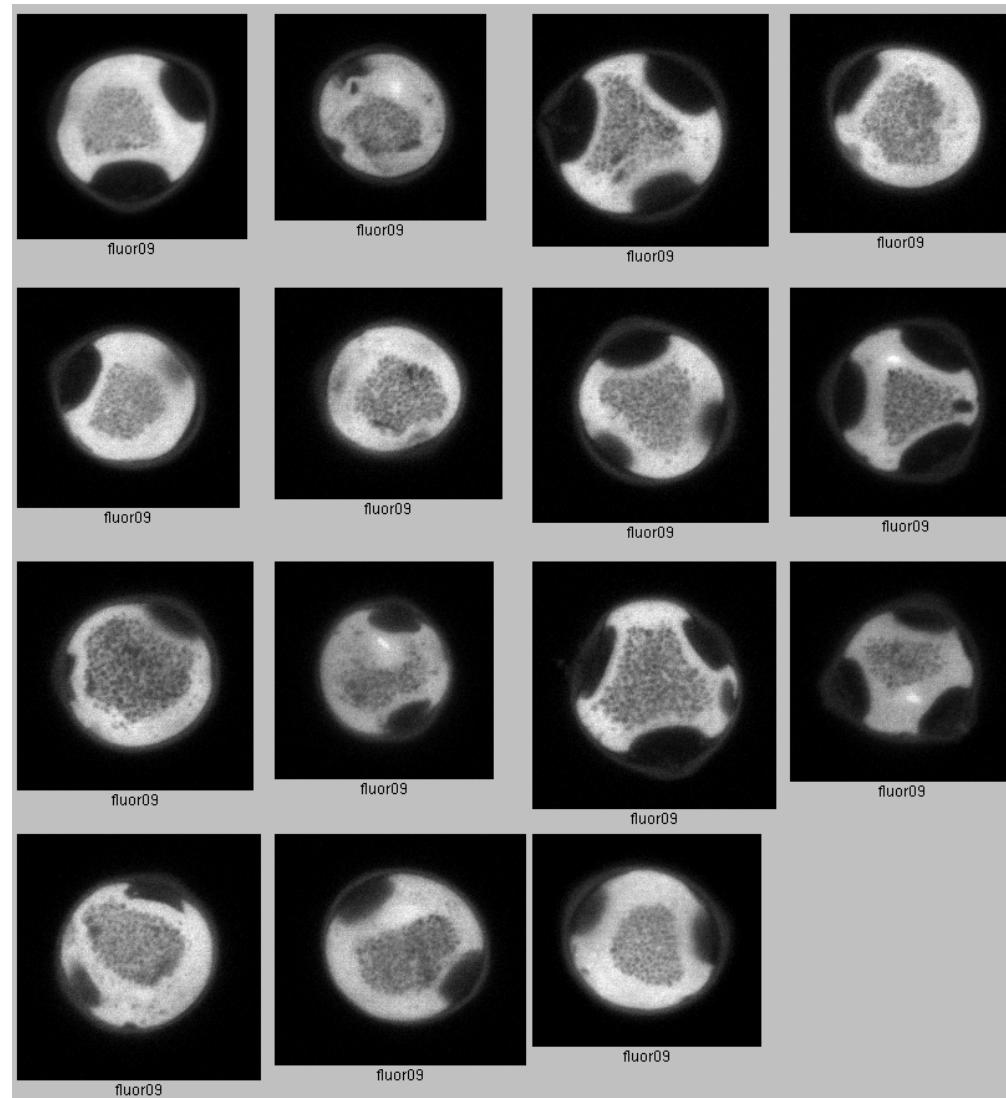
Roggen



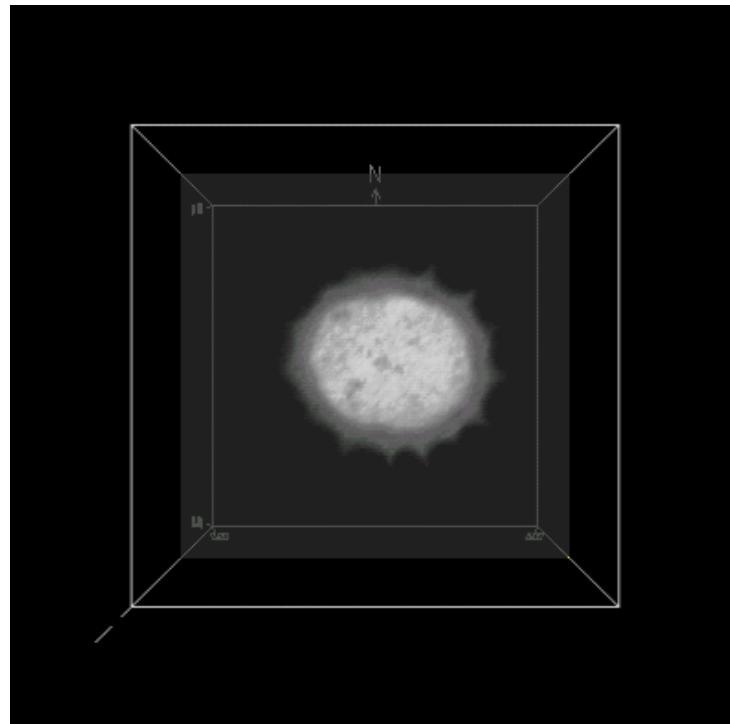
Beifuß

+ 33 further species (not relevant for allergies)

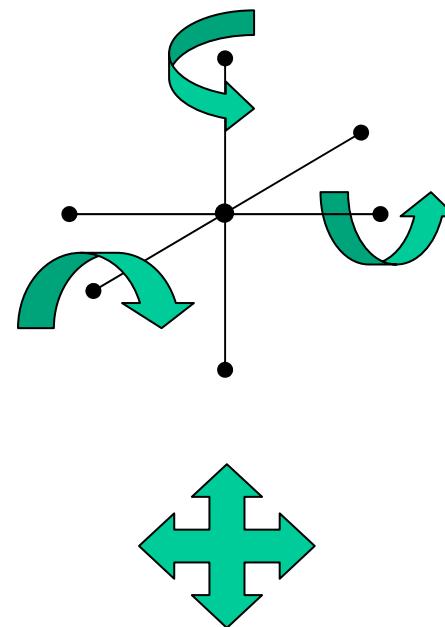
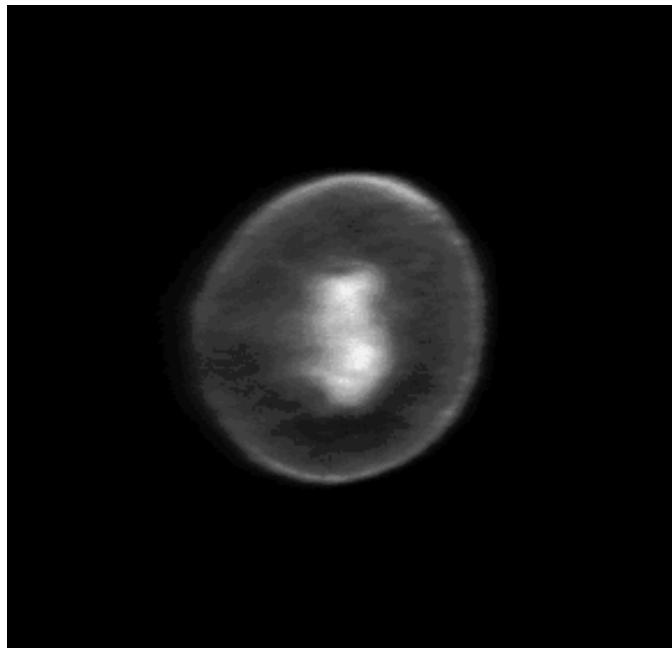
# Haselpollen



# Gänseblümchen/daisy pollen grain



# Eibe/Taxus



Integriere über die 3D Euklidsche Bewegung

# Reference Data Base Description

- 26 pollen species directly sampled from the plants
  - Correctly assigned species
  - Maybe less variation in size and shape compared to airborne pollen (no different sub-species, no regional variation)
  - No deformation or contamination or agglomeration
- Recording 3D volumes of about 15 samples of each species with a Confocal Laser Scanning Microscope
  - No distortions of the data due to imperfect imaging of a usual microscope

# Automated Pollen Recognition in Air Samples by Digital Microscopy

O. Ronneberger, U. Heimann, V. Dietze, E. Schultz

## ► Motivation

- Time-consuming, but still inaccurate visual pollen counting

## ► Demand

- Reliable pollen data for pollen forecast

## ► Approach

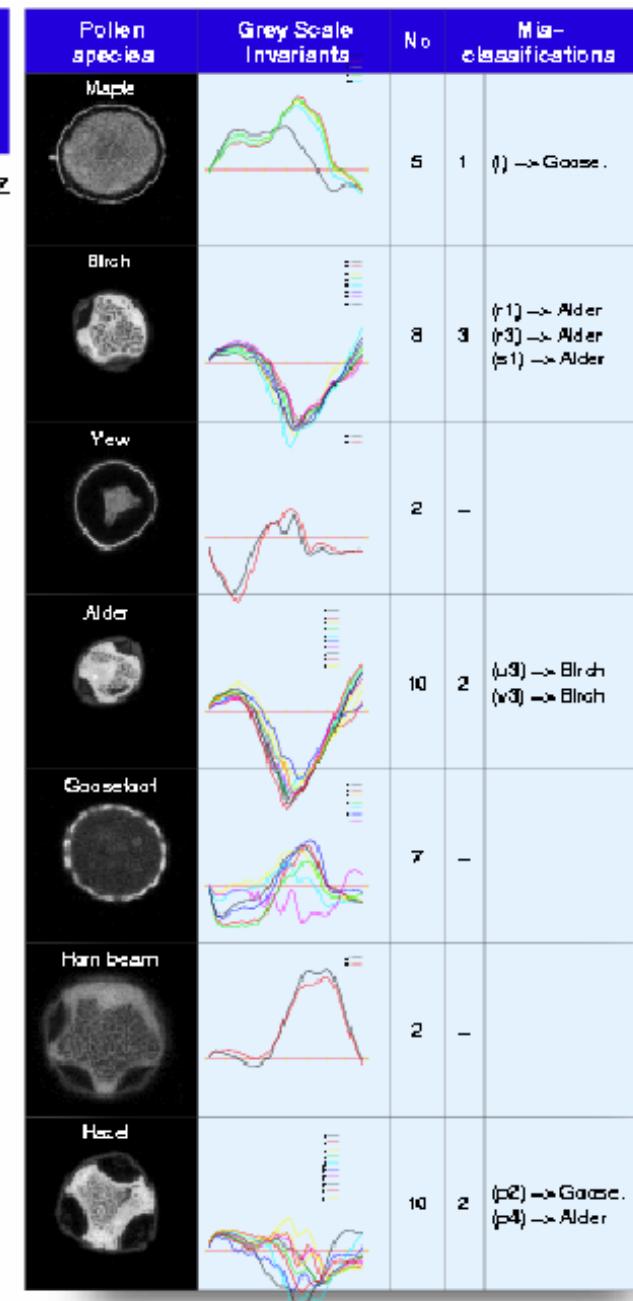
- 3D fluorescence imaging (tomography) instead of 2D conventional translucent microscopy
- Image recognition based on general »grey scale invariants«, instead of traditional object specific features

## ► First results (see table on the right)

- »Grey scale invariants« have characteristic shapes for different pollen species
- Recognition rate of 82% already in a first run

## ► Outlook

- Employment of digital microscopy for automated particle analysis in general.



Collaborating institutions



Deutscher  
Wetter  
Dienst



MeteoSchweiz



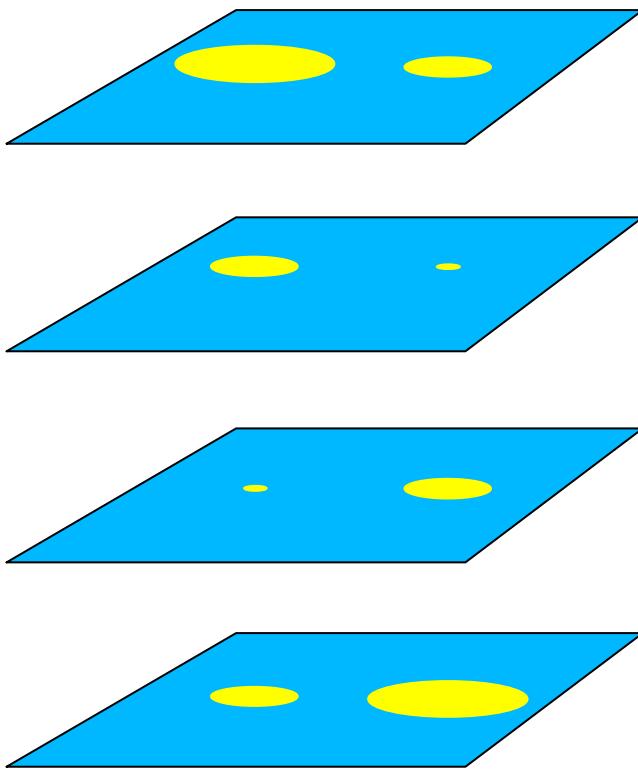
# Classification Results using 3D LSM Data

(leave-one-out Classification)

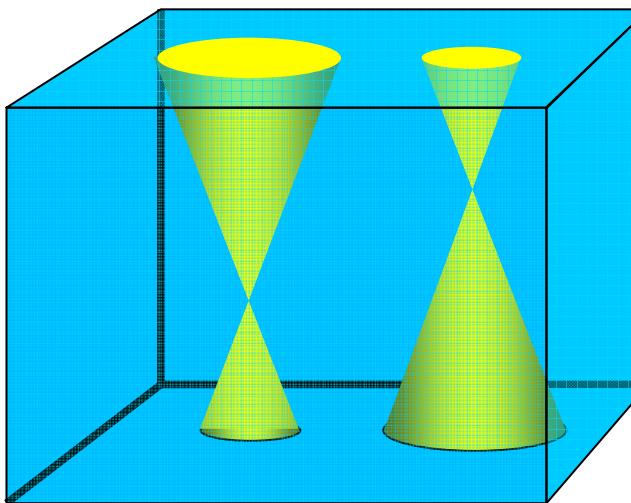
	Correct	Wrong classifications
<i>Artemisia</i> :	13	1 -> <i>Compositae</i> , 1 -> <i>Platanus</i>
<i>Alnus</i> :	15	-
<i>Alnus viridis</i> :	12	-
<i>Betula</i> :	13	2 -> <i>Plantago</i>
<i>Corylus</i> :	13	1 -> <i>Alnus</i>
<i>Gramineae/Poaceae</i> :	15	-
<i>Secale</i> :	11	3 -> <i>Fagus</i> , 1 -> <i>Tilia</i>
Allergological irrelevant*:	282	2 -> <i>Gramineae</i>
Total:	97.4%	2.6%

\* *Acer, Carpinus, Chenopodium, Compositae, Cruciferae, Fagus, Quercus, Aesculus, Juglans, Fraxinus, Plantago, Platanus, Rumex, Populus, Salix, Taxus, Tilia, Ulmus, Urtica*

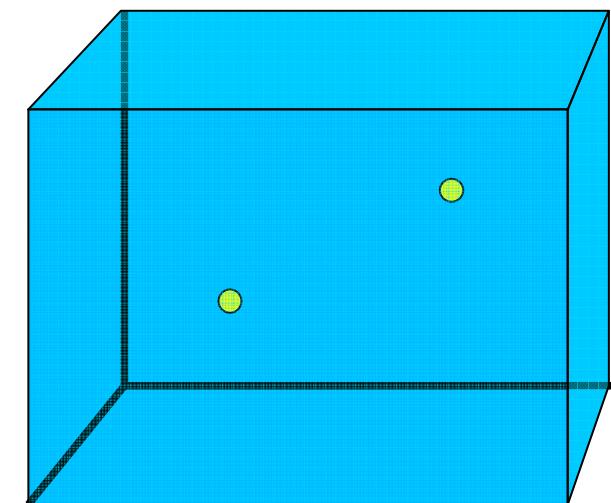
# Gewinnung von tomographischen 3D-Daten mit einem normalen Fluoreszenz-Mikroskop



Aufnahme eines Bildstapels mit einem normalen Mikroskop

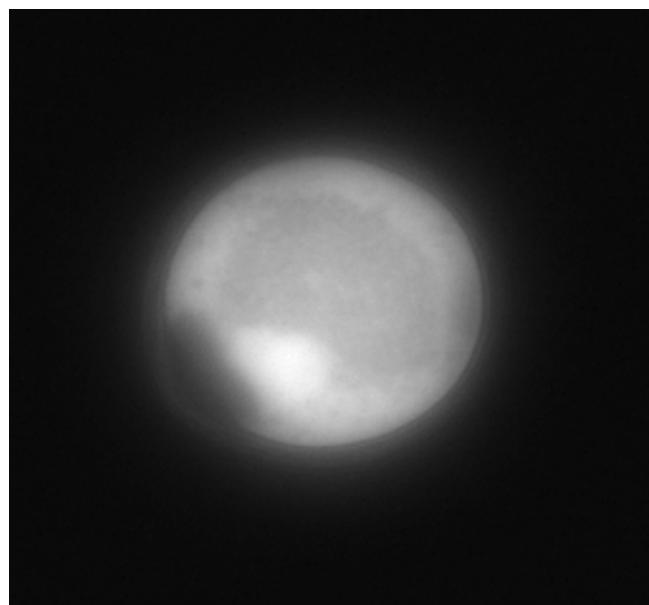


Zusammensetzen der Bilder im Computer zu einem Volumen

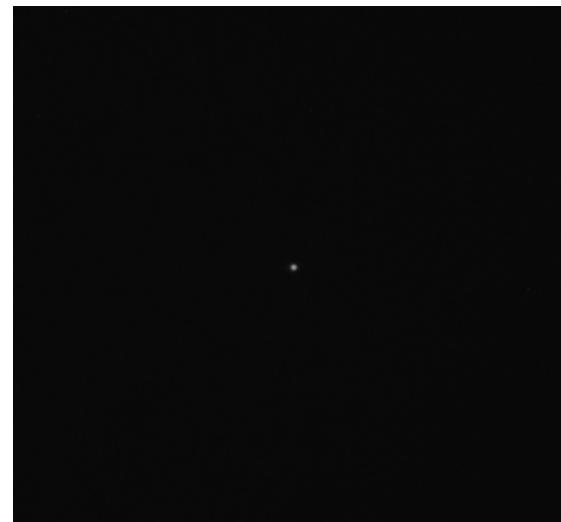


Nach Entfaltung mit der Punktbildfunktion sind alle unscharfen Bereiche entfernt

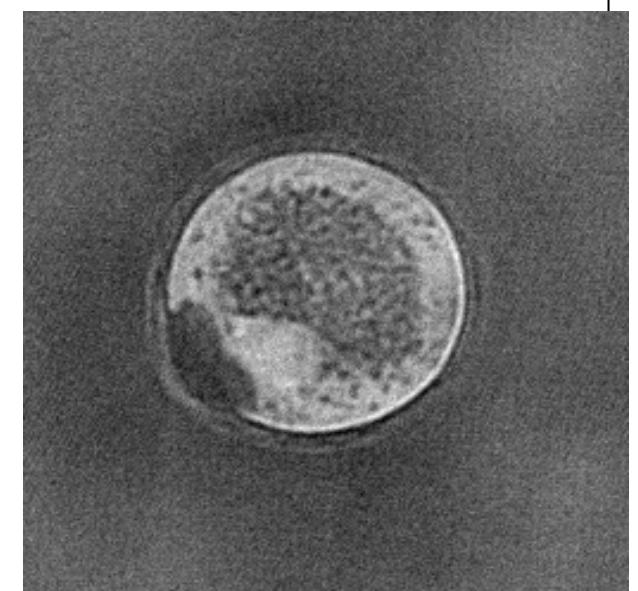
# Deconvolution of a Pollen (Hasel/Corylus)



volume



point-spread  
function

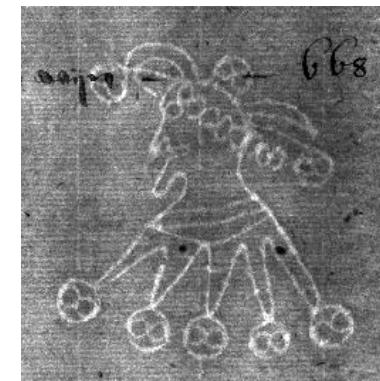
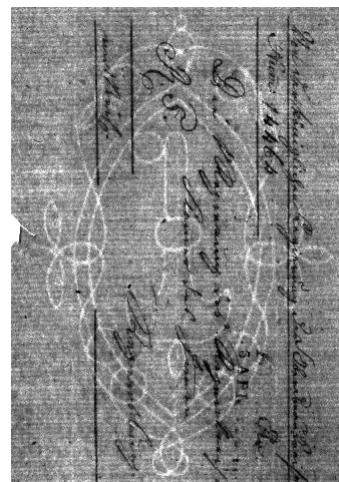


deconvolved volume

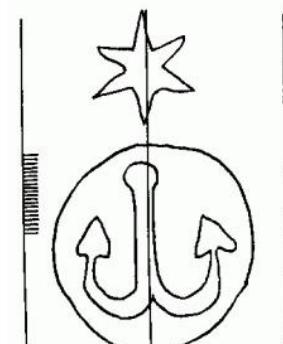
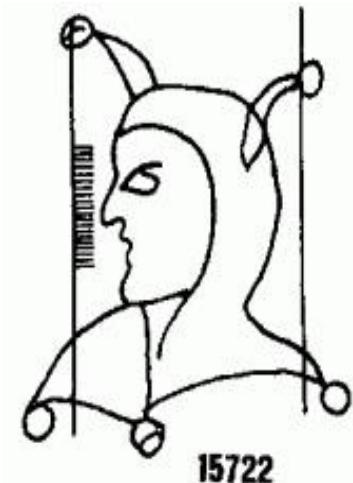
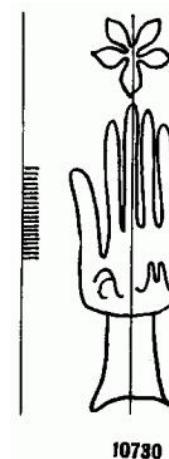
# Search Engine for Watermarks in old paper prints

(Joint project with Basel Paper Mill, Swiss Museum for  
Paper, Writing and Printing)

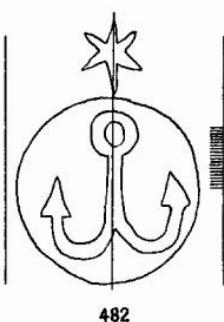
Scanned watermark images  
(samples)



After preprocessing  
(sample images)



Query Image



482



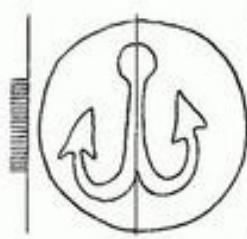
1.00000



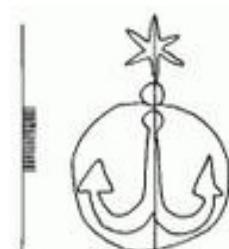
0.98446



0.98229



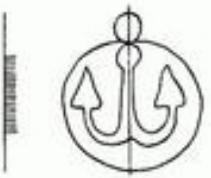
0.98045



0.97986



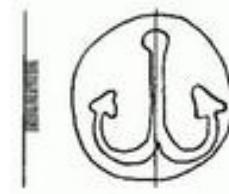
0.97949



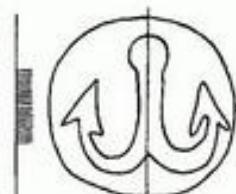
0.97924



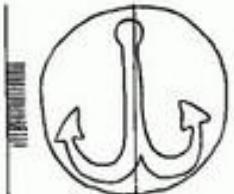
0.97890



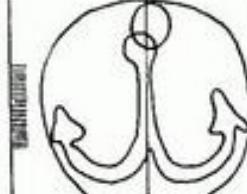
0.97787



0.97580

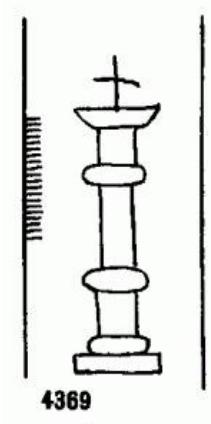


0.97565



0.97326

Query Image



4369	1.00000
4365	0.80769
4364	0.80128
4377	0.78125
4368	0.77926
4371	0.77350
4372	0.76923
4366	0.72436
4363	0.72192
4369	0.71923
4362	0.71795
4370	0.70964

Query Image



<p>13957 1.00000</p>	<p>13959 0.94530</p>	<p>13958 0.93585</p>	<p>13956 0.91158</p>
<p>13959 0.87267</p>	<p>13953 0.86996</p>	<p>13954 0.86776</p>	<p>13960 0.83026</p>
<p>13792 0.76324</p>	<p>13793 0.76137</p>	<p>13851 0.75978</p>	<p>13799 0.75896</p>

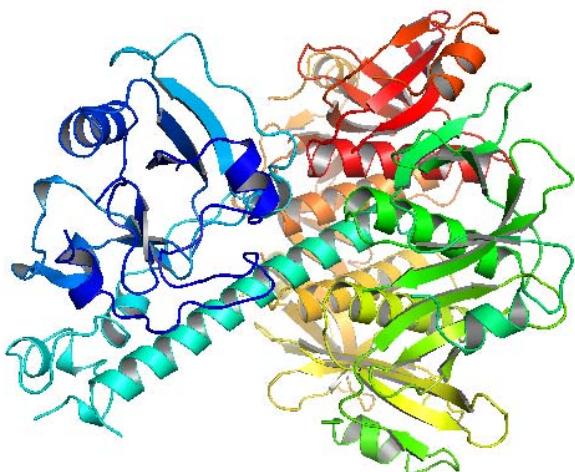
## USED FEATURES:

- We encode inter-relational structure knowledge from feature vectors.
- Line segments are grouped by hierarchical clustering.
- Relative spatial arrangements of the grouped line segments are captured in histograms.
- The resulting histogram is robust against illumination changes, rotation, translation and scaling.

# Development of a Search Engine for Protein Databases

- There is no canonical way of describing the folding structure of proteins
- This search engines develops space-invariant descriptors for the folding structure of proteins which can be used as index for searching in databases

# A Search Engine for Protein Databases based on 3D Folding-Structure



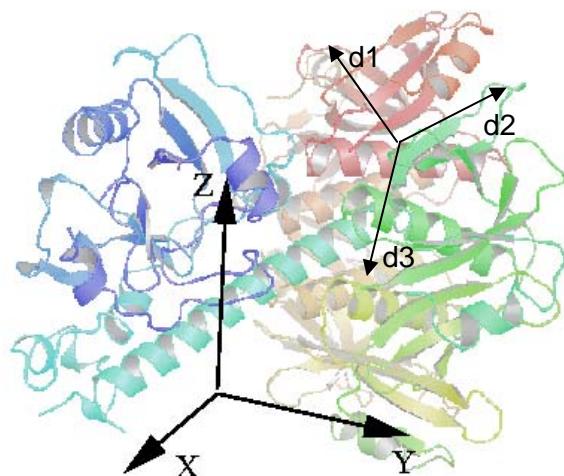
- A Protein is a highly structured 3D object consisting of amino acid chains
- Typical number of amino acids: 100-1000  
→ several thousands of atoms
- In the past proteins were compared solely by its amino acid sequence, but different sequence → same 3D-structure →  
3D-structure                      functional behavior
- Growing amount of 3D-macromolecular structural data (currently > 32000 in Brookhaven's PDB)
- need for efficient search and retrieval techniques
  - drug design
  - evolutionary link detection
- current structural search techniques rely on alignment algorithms  
→ slow (retrieval times >hour or >day)

# A Search Engine for Protein Databases based on 3D Folding-Structure

- Methods known from Content-Based 3D-Shape/Image Retrieval offer numerous fast algorithms
- Search and Classification is feature-based. The features are invariant to the 3D-pose and the numbering of the amino acids.
- *Invariant Protein Features* are numerical quantities that describe global properties of the Protein, for example

F1 = „How often occur two atoms of the Protein within some distance d“

F2 = „Percentage of atoms which are part of a alpha-Helix structure“



## Preliminary Results:

### Classification Accuracy:

- for SCOP-'class' granularity: **99.8%**
- for SCOP-'fold' (class alpha): **97.6%**

**Retrieval times:** less than a second in DB with

more than 32000 proteins (>65000 chains)

# Video

## “Licht ins Dunkel”

[Licht ins Dunkel\VTS\\_01\\_1.avi](#)