

# Texture-based classification of forest types using high resolution aerial photographs

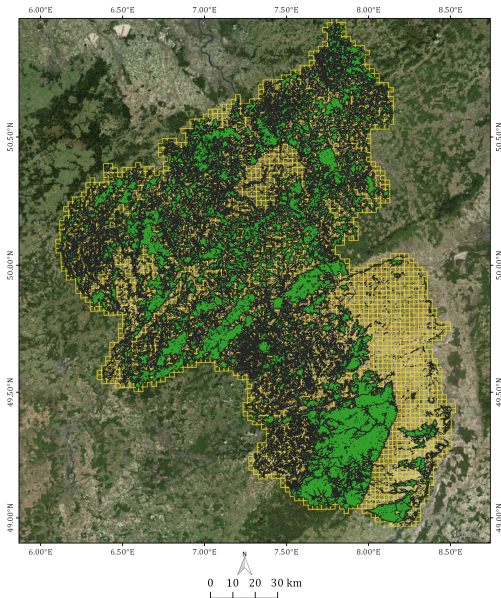
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University Trier  
Environmental Remote Sensing and Geoinformatics

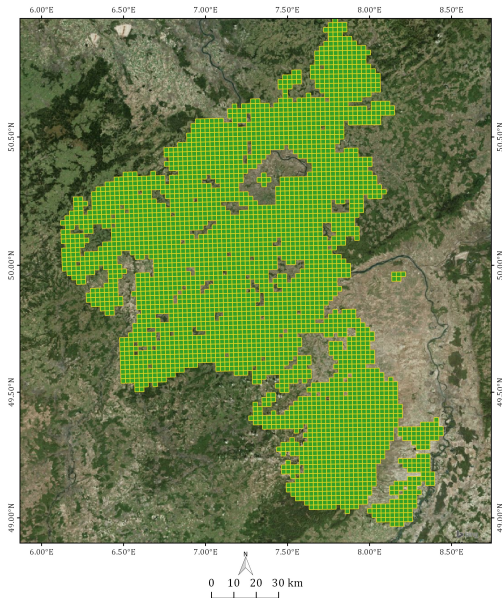
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4. *gemeinsame Jahrestagung der Arbeitskreise der Deutschen Gesellschaft für Geographie (DGfG) e.V. & Auswertung der Deutschen Gesellschaft für Photogrammetrie, Fernerkundung und Geoinformation (DGPF) e.V.*

*24./25. September 2015  
Geozentrum der Universität Bonn*



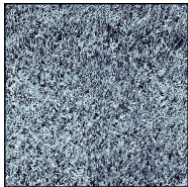
**Coverage of high resolution aerial images in RLP (5266 tiles)**



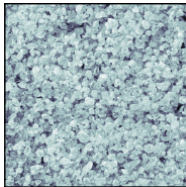
**Coverage of tiles with at least  
50% forest cover (3645 tiles)**

# Natural textures

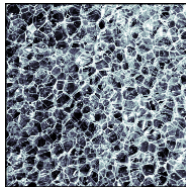
Grass (D9)



Beach sand (D29)



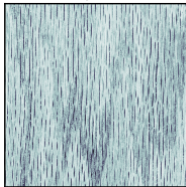
Plastic bubbles (D112)



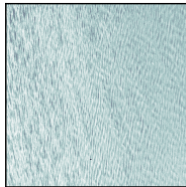
Straw (D15)



Wood grain (D68)

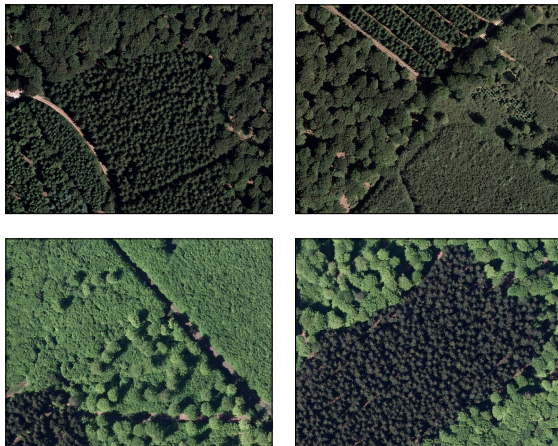


Water (D38)

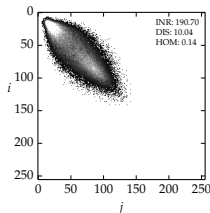
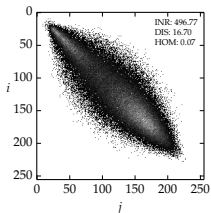
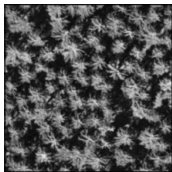
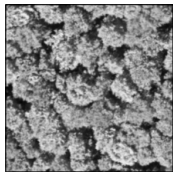


Brodatz library of natural textures [Weber, 1997]

# Texture of forest types



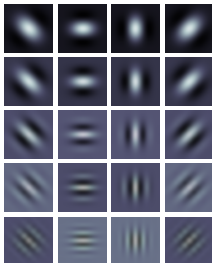
# GLCM



## Gray level co-occurrence texture features

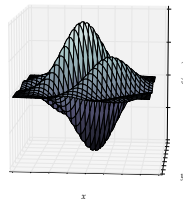
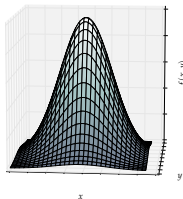
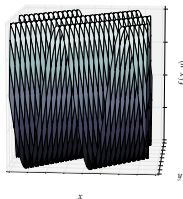
- spatial domain
- second-order statistics
- GLCM ( $P(i, j | \theta, \delta)$ )
- scalar texture descriptors

# Gabor filter banks



## Gabor filter implementations

- frequency domain
- windowed Fourier transform
- filter banks ( $\Psi(x, y, f_l, \theta_k)$ )
- Gabor energy feature



# Research questions

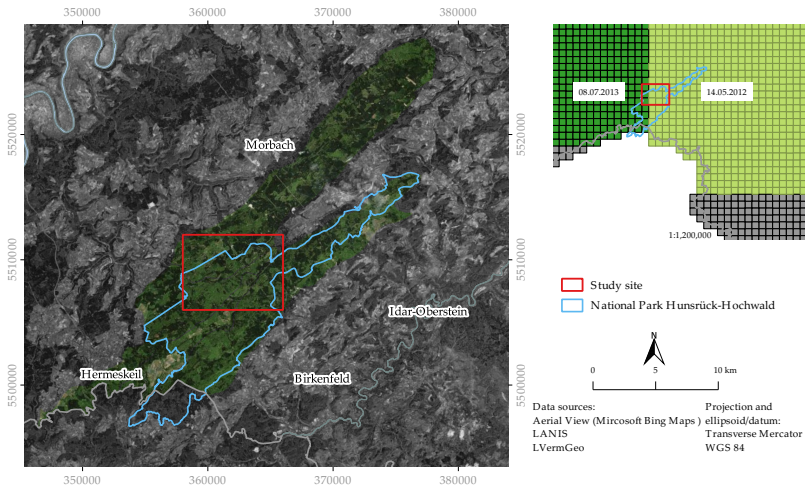
main question

Is a texture-based classification of forest types possible?

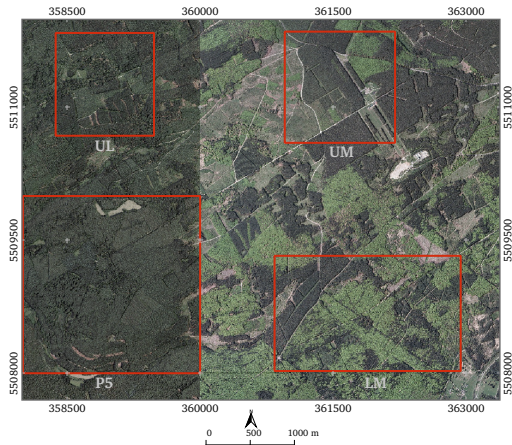
secondary question

Which combination of parameters performs best?





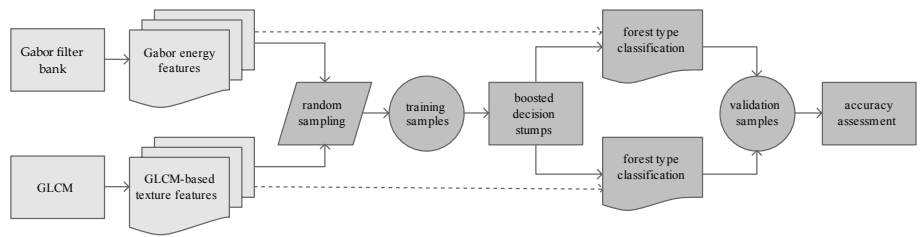
# Setup



## Texture processing

- R - G - B - NIR - Y
- window size
- spatial resolution
- orientation

# Workflow

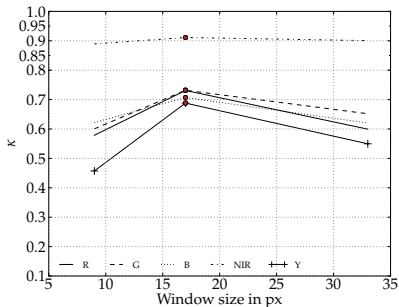
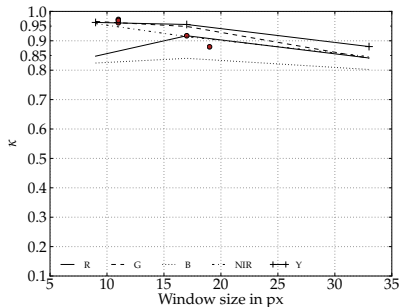


# Classification accuracies

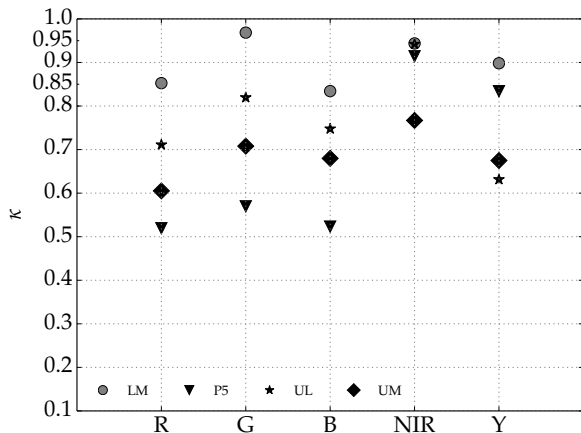
A selection of the highest achieved classification accuracies across all conducted texture-based classifications.

Setup	Subset	$\kappa$	$\sigma$
GLCM <sub>Y-11</sub>	LM	0.97	0.01
GLCM <sub>G-9</sub>	LM	0.97	0.02
D1 <sub>G</sub>	LM	0.97	0.03
D2 <sub>NIR</sub>	LM	0.95	0.02
D1 <sub>NIR</sub>	UL	0.94	0.04
GLCM <sub>G</sub>	LM <sub>2m</sub>	0.93	0.02
D1 <sub>NIR</sub>	P5	0.91	0.03
GLCM <sub>NIR-17</sub>	P5	0.91	0.07
GLCM <sub>NIR-9</sub>	UL	0.92	0.05
GLCM <sub>Y-17</sub>	UM	0.86	0.08

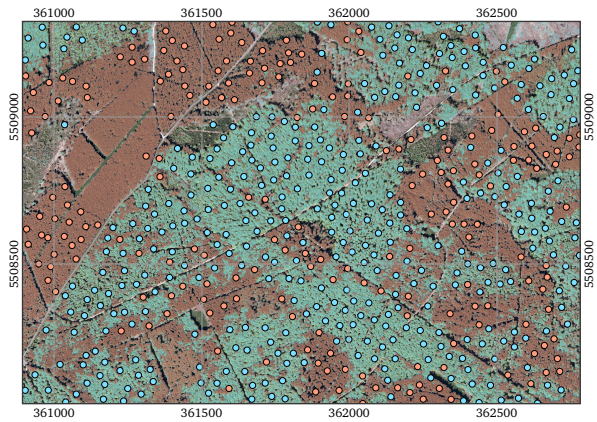
## GLCM-based classification (subset LM and P5)



# Gabor energy based classification



# Subset LM



### Validation points

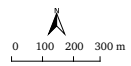
- Deciduous forest
- Coniferous forest

### Classification (GLCM-based)

- Deciduous forest
- Coniferous forest

Accuracy assessment:  
5-fold cross validation  
Kappa

- fold 1 : 0.95
- fold 2 : 0.97
- fold 3 : 0.98
- fold 4 : 0.97
- fold 5 : 0.95



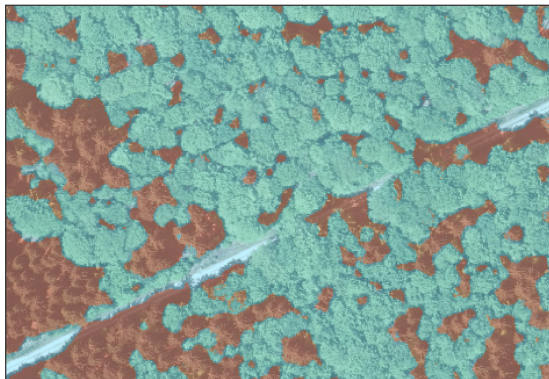
## Subset LM



1:1.500



## Subset LM



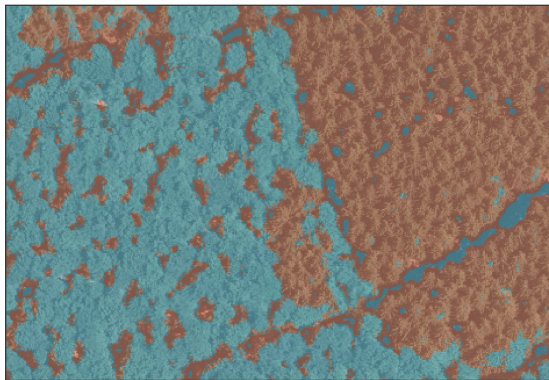
1:1.500

## Subset P5



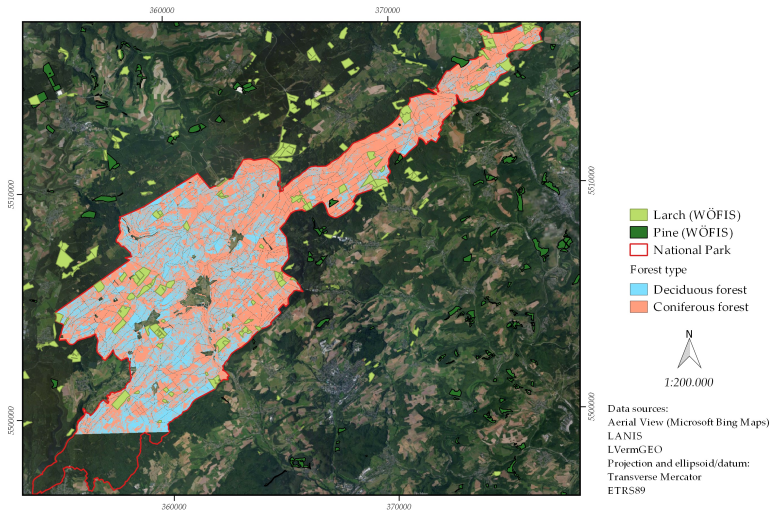
1:1.500

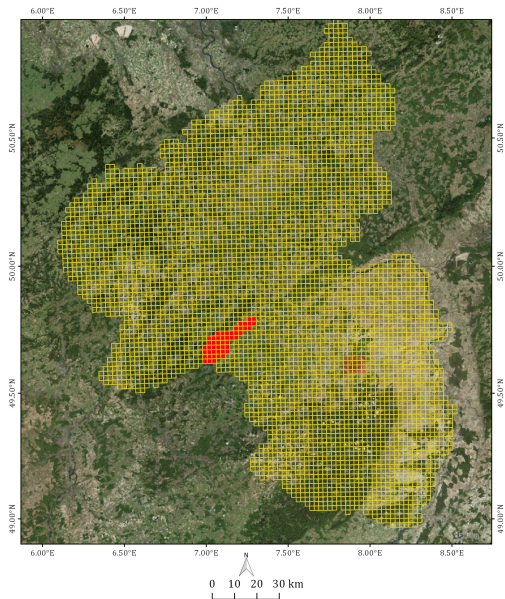
## Subset P5



1:1.500

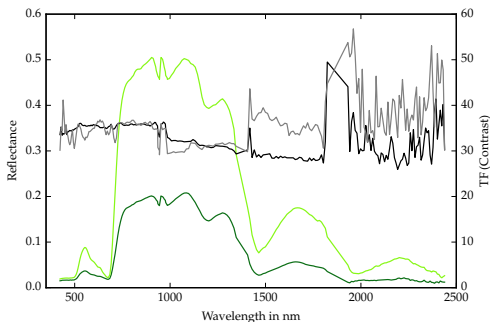
# Hunsrück-Hochwald National Park





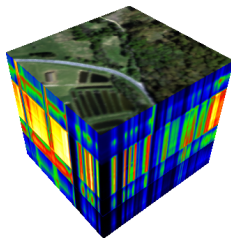
**Coverage of generated forest type maps (48 tiles)**

## Focus areas



### GLCM in 3D

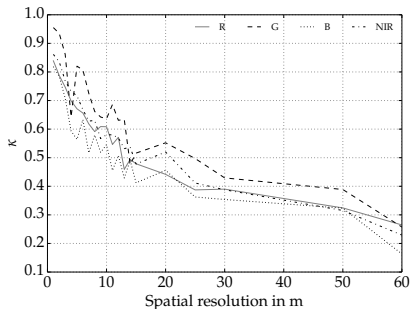
- texture of hyperspectral images
- volumetric texture features
- VGLCM [Tsai et al., 2007]






## Focus areas

### Topics

- spectral and textural information
- change in scale
- texture features based on geomertic properties of the GLCM
- image segmentation



## References

-  Tsai, F., Chang, C. K., Rau, J. Y., Lin, T. H. and Liu, G. R. (2007)  
3D computation of gray level co-occurrence in hyperspectral image cubes.  
*Energy Minimization Methods in Computer Vision and Pattern Recognition*, pp. 429-440.
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The USC-SIPI Image Database: Version 5, Original release: October 1997, Signal and Image Processing Institute.  
*University of Southern California, Department of Electrical Engineering.*
-  Haralick, R. M. (1973)  
Textural features for image classification.  
*IEEE Transactions on Systems, Man and Cybernetics* No. 6, 610 – 621.