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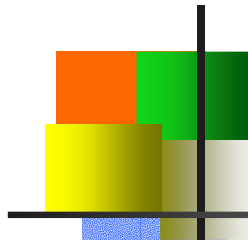
## Spatial information extraction for urban areas based on hyperspectral data



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4 September 2003 – Bruges, Belgium

CASI-SWIR WORKSHOP



# Spatial information extraction for urban areas based on hyperspectral data



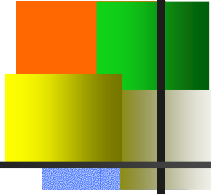
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- aim : to extract all different kinds of information over urban and peri-urban environment using hyperspectral data.
- 2 kinds of information : geometric and thematic
- in conjunction with the Stereo-project “Improving spatial information extraction for local and regional authorities using very-high-resolution data”. SPIDER



# Flight Planning



- 3 images
  - 2 in “metric” mode
    - Altitude 440 m
    - Resolution 0.52 m
    - 10 bands
  - 80 % overlap between 2 metric images
  - 1 in hyperspectral
    - Altitude 970 m
    - Resolution 1.34 m
    - 48 bands



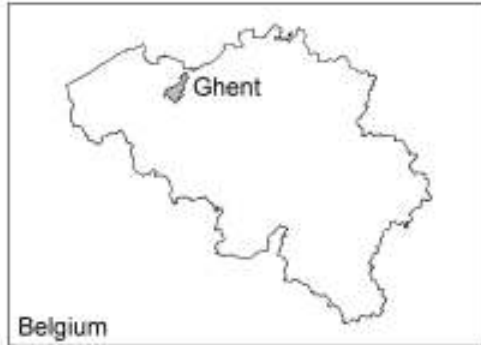
# Flight Planning



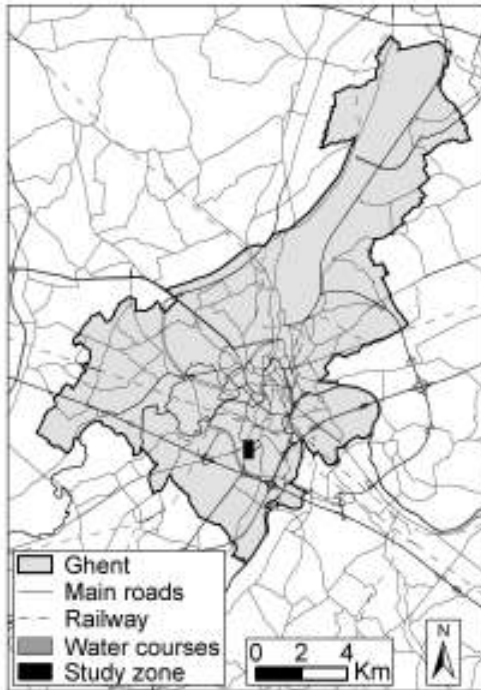
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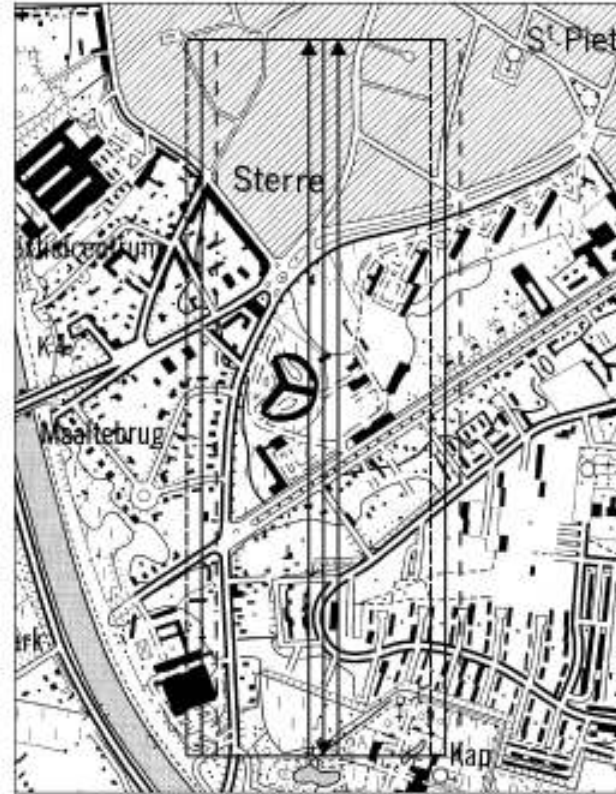


Belgium



- Ghent
- Main roads
- - Railway
- Water courses
- Study zone

0 2 4 Km



→ Flight lines

**Images**

- metrical mode
- hyperspectral mode
- - metrical mode

0 100 200 Meters

**Ghent**

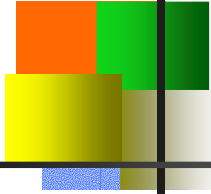
1000 x 200 m



# Field Campaign



movie <http://allserv.rug.ac.be/~dedvrien/casi>



# Data Handling



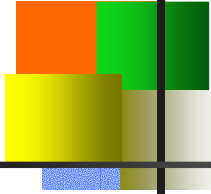
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- Radiometric Correction
  
- Geometrical Correction :
  - Metrical mode : not for terrain
  - Hyperspectral mode : terrain (DSM aerial photo)
  
- Atmospheric Correction



## Goals – geometrical aspects



- Digital Surface Model

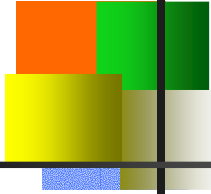
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- Orthophoto





# Photogrammetrical processing



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- Reducing 10 bands to 3 bands
- 2 methods :
  - Grouping bands
  - FCC with individual bands  $\leq$
- 3 bands that react best to the matching algorithm

band	interval	no points
1	467.4nm+/- 17.9nm	74
2	502.9nm+/- 16.1nm	71
3	540.5nm+/- 19.9nm	60
4	581.1nm+/- 19.1nm	78
5	639.0nm+/- 10.6nm	49
6	670.4nm+/- 19.2nm	68
7	725.9nm+/- 34.6nm	86
8	795.9nm+/- 33.8nm	101
9	866.2nm+/- 34.8nm	96
10	926.0nm+/- 23.3nm	106
	RMS = 1/5 pixel	





# Photogrammetrical processing



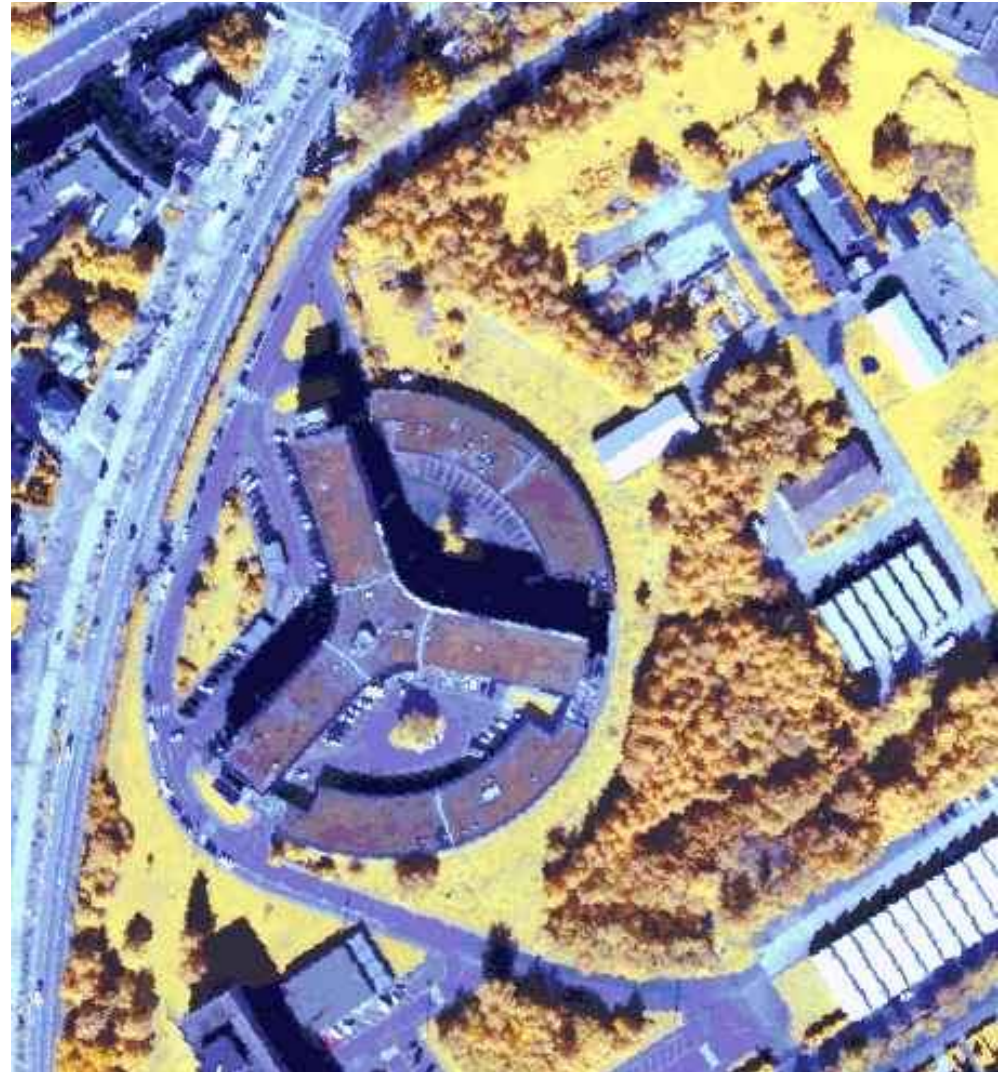
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- FCC
- band 8 – R
- band 9 – G
- band 10 – B

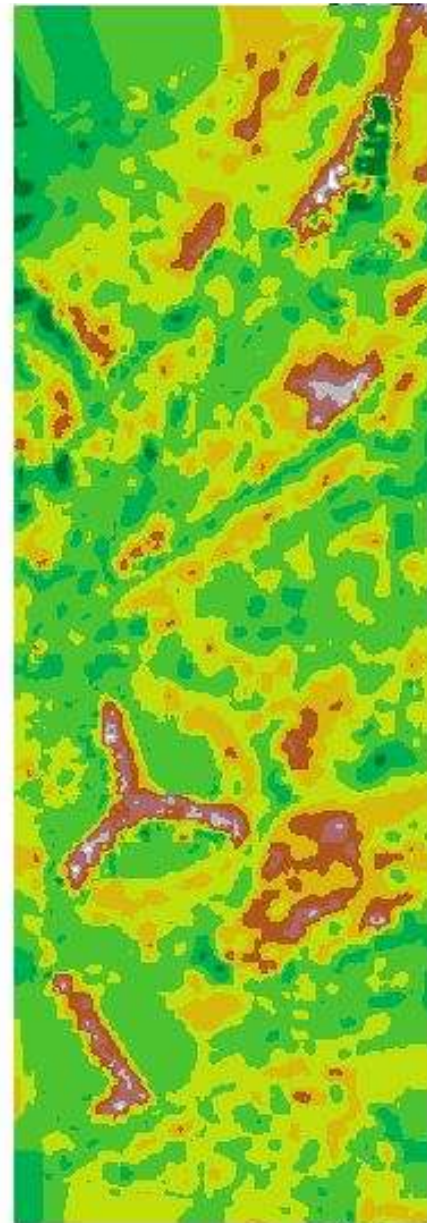
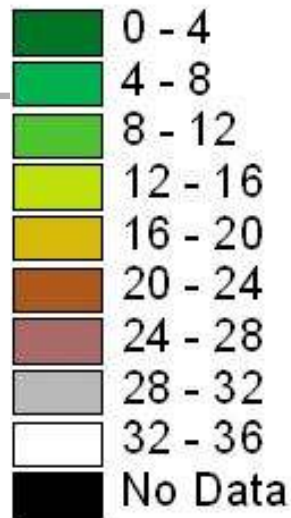
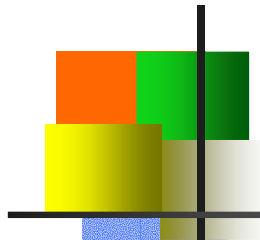




# Photogrammetrical processing



- Relative orientation
  - $\Delta t$  between images : shadows have moved  
15 min = 1.3 m for 20 m building
  - manual : 70 points
- Absolute orientation
  - Ground level : 9 points  
AO with only these points resulted in no Z results in stereomodel
  - Rooftop Level : 4 points
  - Problem :  $B/H = 0.1$  to  $0.2$
- Automatic parallax matching, DEM extraction and orthophotogeneration



Geography  
DEPARTMENT

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## Conclusion



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- better spatial resolution, better result
- larger area
- cross-track stereo overlap with 2 nadir images not ideal for good stereomodel
- suggestion : nadir image + backward image  
nadir image + forward image

...



# Classification of hyperspectral data



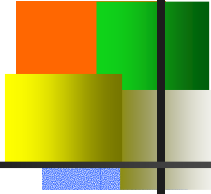
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- Objective:
  - Extract **detailed thematic information** from hyperspectral data for urban areas
  
- Topics:
  - Band reduction
  
  - Different classification strategies
    - Pixel-based
    - Region-based methods
  
  - Postclassification



# Land-cover legend



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LAND COVER CLASS	
grey surface	grey slate and tile
	roofing light roofing dark roofing
	grey gravel
	concrete
	asphalt road
	light road material
red surface	red tile
	exposed red tile not exposed red tile
	red gravel

LAND COVER CLASS	
reflecting surface	metal
	light metal dark metal
vegetation	grass
	shrub and tree
	moss
shadow	shadow vegetation
	shadow non vegetation

# Collection of training data



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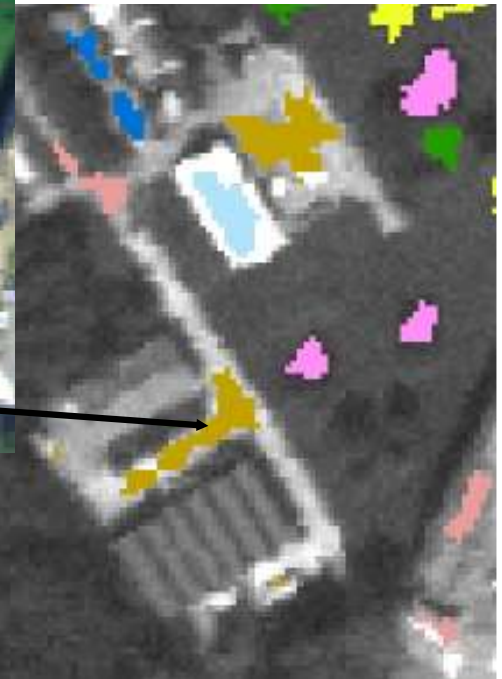


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- Sampling blocks on aerial photos
- Terrain verification
- Add shadow classes
- Ecognition segments as training blocks

concrete



# Collection of validation data



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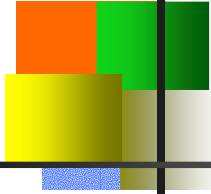
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- CAPI based on **aerial photo** and terrain verification
- **random sampling** in CAPI
- removal of points too close to training sites







# Band reduction



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- **Feature selection**
  - Visually from signature profiles
  - Best combination of bands (Bhattacharyya distance)
  
- **Feature extraction**
  - Principal Components Analysis (PCA)
  - Discriminant Analysis Feature Extraction (DAFE)
  - Decision Boundary Feature Extraction (DBFE)



# Land-cover classification approach



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- Comparison of results obtained with:
  - **Different classifiers**
    - Pixel-based approach
      - Minimum distance classification (MD)
      - Maximum-likelihood classification (ML)
    - Region-based classification
      - ECHO classifier (ML)
      - eCognition (MD)
  - **Different band reduction techniques**
  - **Results:**
    - Kappa's between 0.47 and 0.73

# Pixel-based classification



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True-color composite (17, 10, 2)



ML DBFE (7 features): Kappa 0.73

-  red gravel
-  grey slate and tile
-  red tile
-  concrete
-  grey gravel
-  metal
-  grass
-  shrub&tree
-  moss
-  shadow\_nv
-  shadow\_v
-  asphalt road
-  light road material
-  roofing

# Region-based classification



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True-color composite (17, 10, 2)



ECHO DBFE (7 features): Kappa 0.70

-  red gravel
-  grey slate and tile
-  red tile
-  concrete
-  grey gravel
-  metal
-  grass
-  shrub&tree
-  moss
-  shadow\_nv
-  shadow\_v
-  asphalt road
-  light road material
-  roofing



## Conclusions classification results



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- **Band reduction**
  - Best results DBFE (7 bands)
  
- **Classification**
  - Kappa's are high considering the number of classes (0.73 for 14 classes)
  - Best results for maximum-likelihood
  - Region-based groups too many pixels together
  
- **Problems**
  - Clutter in pixel-based classifications
  - Shadows



# Postclassification



- Two-step approach:
  - Rule-based classification or grouping of land-cover regions
    - Using:
      - Region-based metrics (area, shape,...)
      - Properties of neighbouring regions
      - Ancillary data, e.g. DSMs, vector maps
  - Post-classification filtering
    - Class specific rules

# Post-classification based on DSM



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ML classification, DBFE (7 features)

- red gravel
- grey slate and tile
- red tile
- concrete
- grey gravel
- metal
- grass
- shrub&tree
- moss
- shadow\_nv
- shadow\_v
- asphalt road
- light road material
- roofing



- < 12 m
- > 12 m

DSM threshold (12 m)

# Post-classification based on DSM



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Intersection ML – DSM threshold

- red gravel (<12m)
- grey slate (<12m)
- red tile (<12m)
- concrete (<12m)
- grey gravel (<12m)
- metal (<12m)
- grass (<12m)
- shrub (<12m)
- moss (<12m)
- shadow\_v (<12m)
- shadow\_nv (<12m)
- asphalt road (<12m)
- light road mat (<12m)
- roofing (<12m)
- red gravel (>12m)
- grey slate (>12m)
- red tile (>12m)
- concrete (>12m)
- grey gravel (>12m)
- metal (>12m)
- grass (>12m)
- trees (>12m)
- moss (>12m)
- shadow\_v (>12m)
- shadow\_nv (>12m)
- asphalt road (>12m)
- light road mat (>12m)
- roofing (>12m)



Postclassification (2 levels)



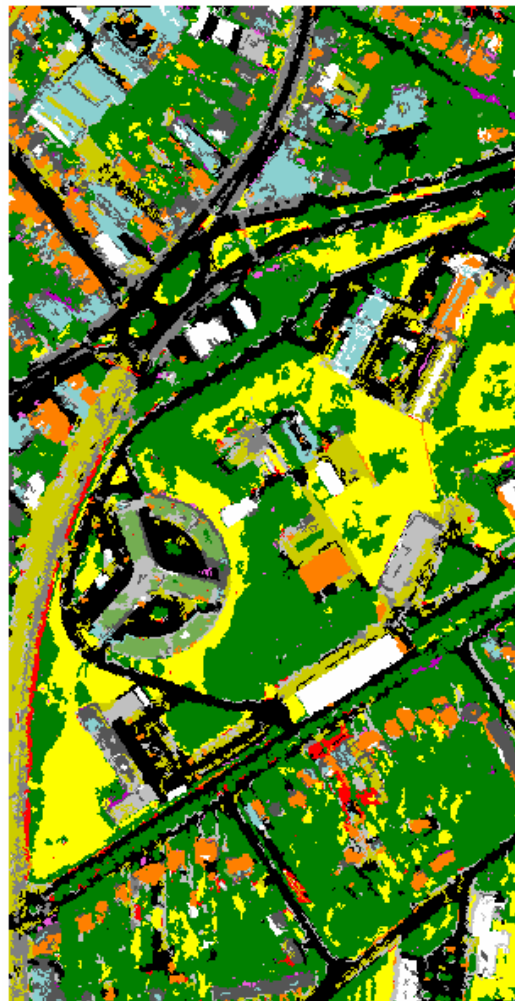
# Post-classification based on DSM



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Postclassification (1 level) Kappa: 0.78

- red gravel
- grey slate and tile
- red tile
- concrete
- grey gravel
- metal
- grass
- shrub&tree
- moss
- shadow\_nv
- shadow\_v
- asphalt road
- light road material
- roofing



ML classification DBFE Kappa: 0.73



# Post-classification based on DSM



## ■ Advantages

- Higher accuracy, Kappa = 0.78
- Strong reduction of clutter, better structure
- Strong reduction in amount of shadow
- Better performance than region-based (ECHO)

## ■ Problems

- Quality of DSM
- DSM threshold
- Additional rules are needed (shape, context larger than neighbourhood)



# General Conclusions



- **Geometrical issues:**
  - Low B/H ratio
  - Nadir + backwards/forwards
  
- **Thematic issues:**
  - Thematically very detailed land-cover maps
  - High accuracy
  - Maximum likelihood + Band reduction
    - Pixel based + DBFE
  - Postclassification
    - Increases accuracy
    - Improves structure
    - Removal of shadow