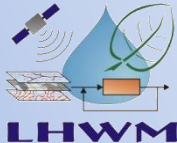


FLOOD MOIST

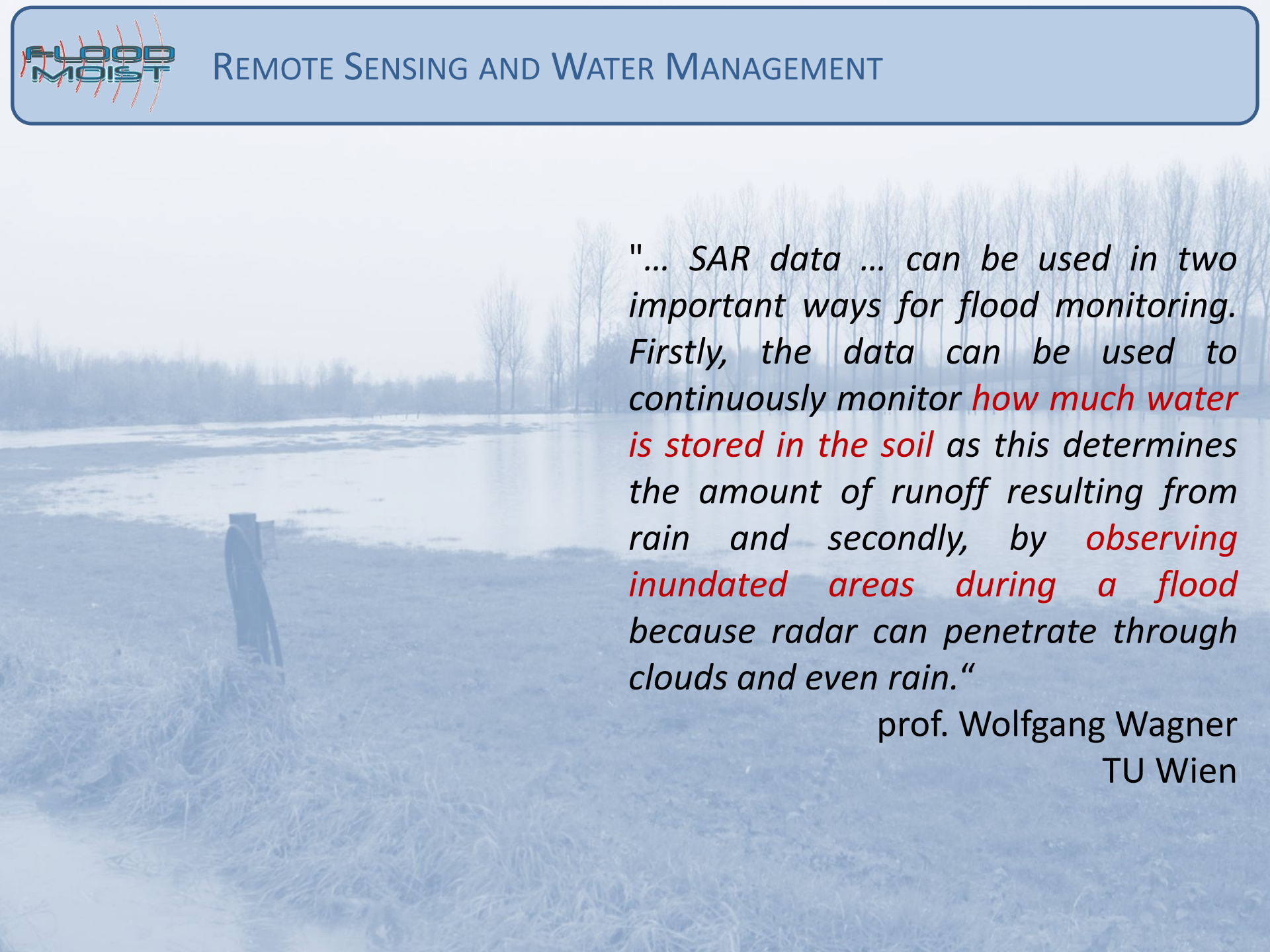
Flood mapping and soil moisture retrieval for improved water management



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PAUL BATES
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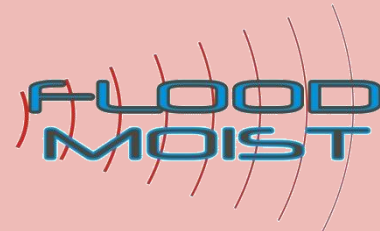
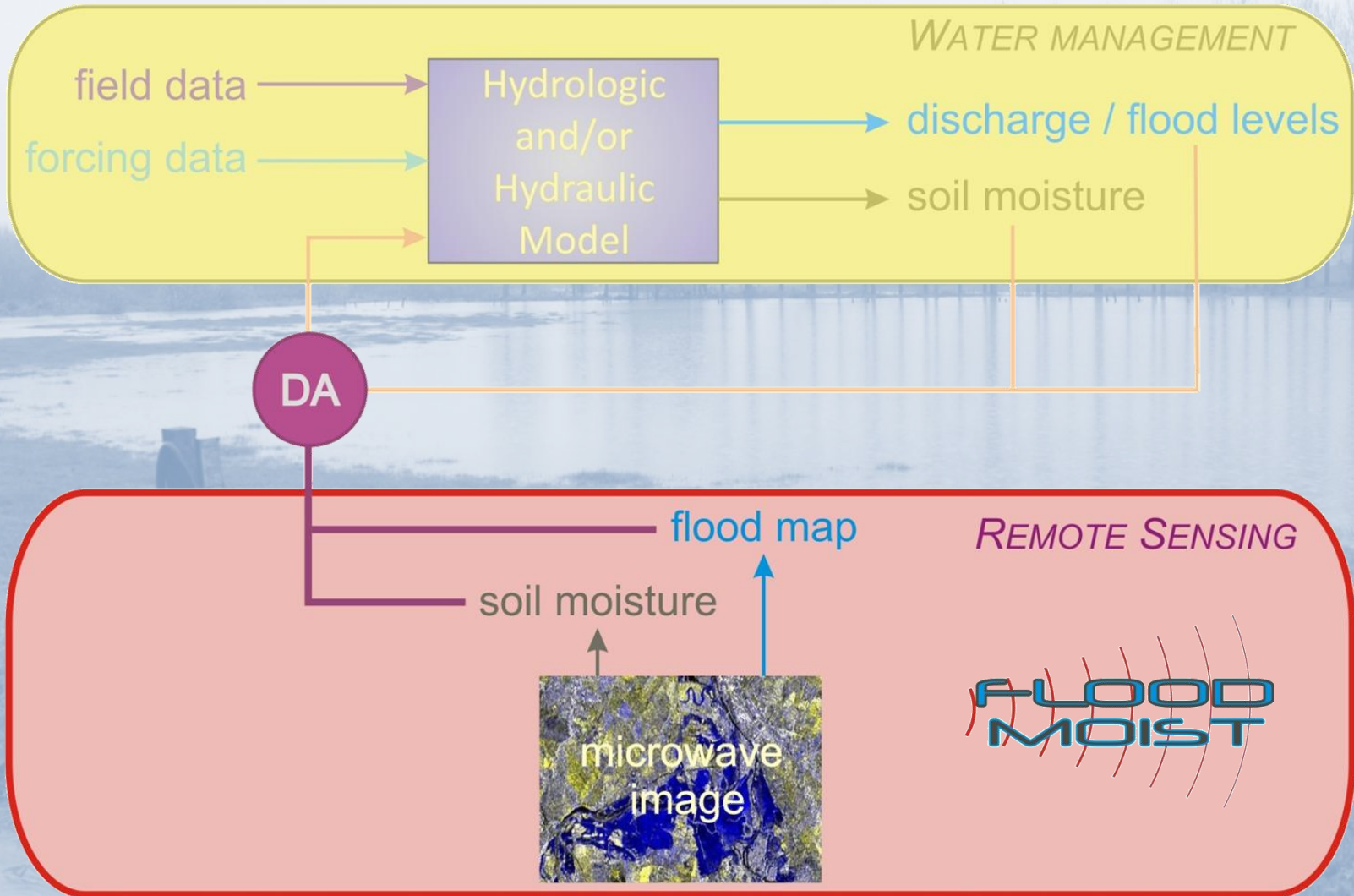
A wide-angle photograph of a flooded agricultural field. The water is calm and reflects the overcast sky. In the foreground, there is a grassy bank with a wooden post. The background shows a line of bare trees under a grey, cloudy sky.

*"... SAR data ... can be used in two important ways for flood monitoring. Firstly, the data can be used to continuously monitor **how much water is stored in the soil** as this determines the amount of runoff resulting from rain and secondly, by **observing inundated areas during a flood** because radar can penetrate through clouds and even rain."*

prof. Wolfgang Wagner
TU Wien



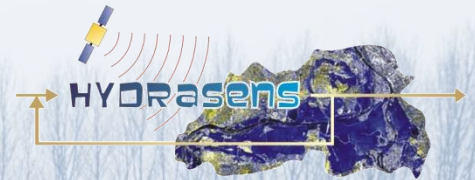
REMOTE SENSING AND WATER MANAGEMENT



HYDRASENS

■ 3 main themes:

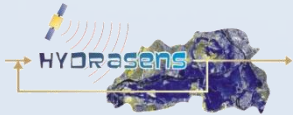
- SAR-remote sensing of soil moisture and flooding
- High-resolution soil moisture using GPR
- Data-assimilation in hydrologic and hydraulic models



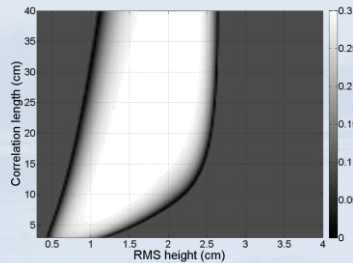
■ Open research questions identified:

- Use of coarse scale remote sensing for soil moisture retrieval
- Use of uncertainty in flood mapping
- Joint assimilation of soil moisture and flooding in a coupled hydrologic-hydraulic model

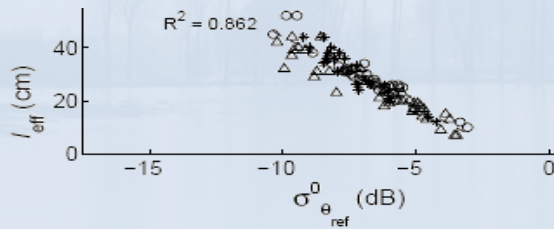
Objective 1: Retrieve soil moisture from microwave remote sensing



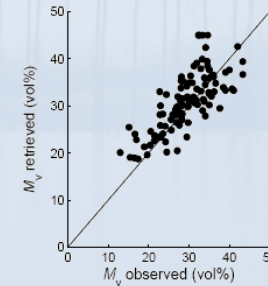
Problem: Roughness parameters measurements: prone to errors
Solution: Method of effective roughness parameters



1. calibrate roughness



2. model roughness parameter



3. retrieve soil moisture

Parameterization of the Land Parameter Retrieval Model for L-Band Observations Using the NAFE'05 Data Set

Richard A. M. de Jeu, Member, IEEE, Thomas R. H. Holmes, Rocco Panciera, and Jeffrey P. Walker

Abstract—The Land Parameter Retrieval Model (LPRM) has been successfully applied to retrieve soil moisture from space-borne passive microwave observations at C-, X-, or Ku-band and high incidence angles (50°–55°). However, LPRM had never

Consequently, they designed an experiment to estimate soil moisture from aircraft observations with L-, Ku-, and Ka-band microwave radiometers. They discovered the strongest relationship between microwave observations and soil moisture at

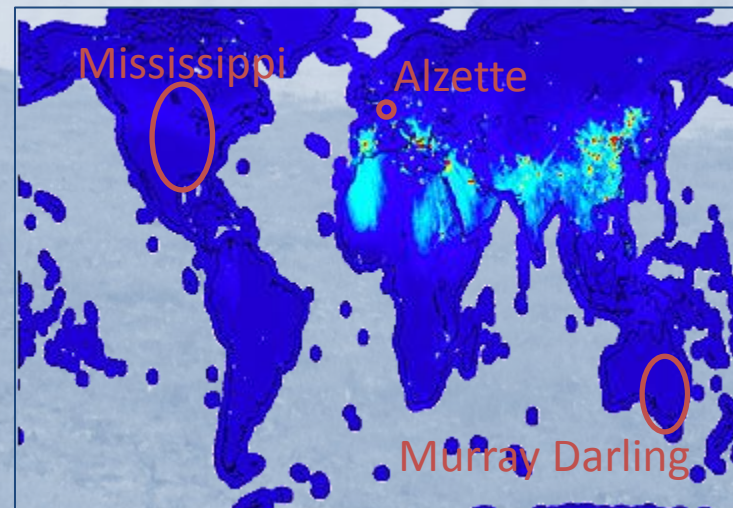
describes the parameterization and performance of LPRM using aircraft and ground data from the National Airborne Field Experiment 2005. This experiment was undertaken in November 2005 in the Goulburn River catchment, which is located in southeastern Australia. It was found that model convergence could only be achieved with a temporally dynamic roughness. The roughness was parameterized according to incidence angle and soil moisture. These findings were integrated in LPRM, resulting in one uniform parameterization for all sites. The parameterized LPRM correlated well with field observations at 5-cm depth ($r = 0.93$)

Use effective roughness parameter for soil moisture retrieval from radiometry?

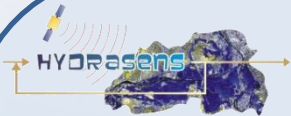
Objective 1: Retrieve soil moisture from microwave remote sensing

Science questions:

1. How robust is method of effective roughness parameters for SAR?
2. Is a similar retrieval approach valid for radiometer data (SMOS)?

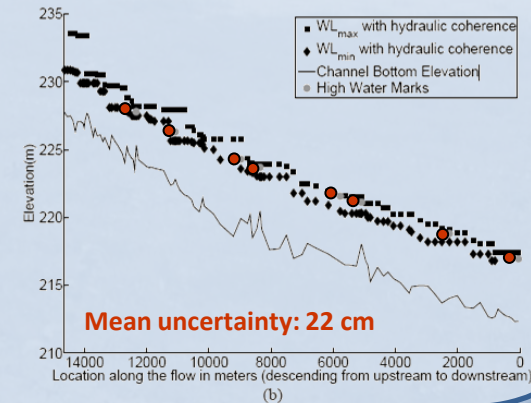
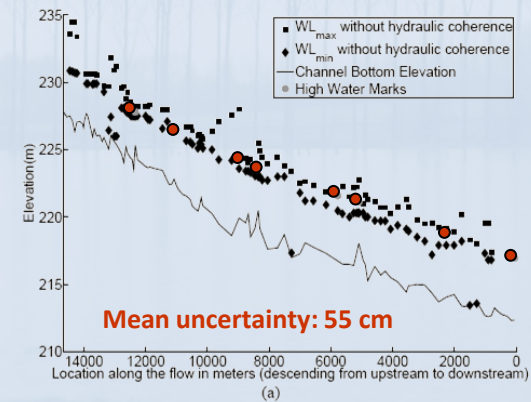
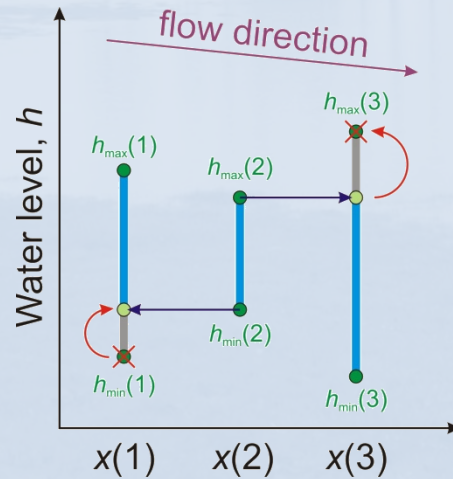
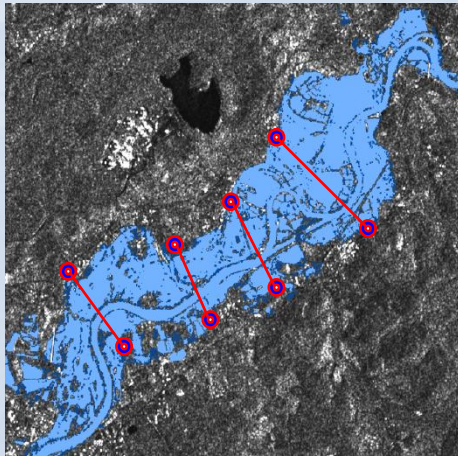


Objective 2: Include uncertainty in flood mapping from SAR

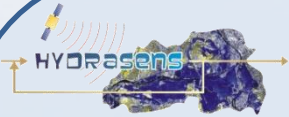


Problem: Floodplain delineation prone to errors and uncertainty

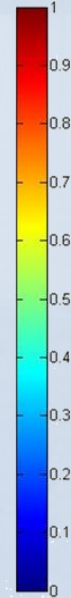
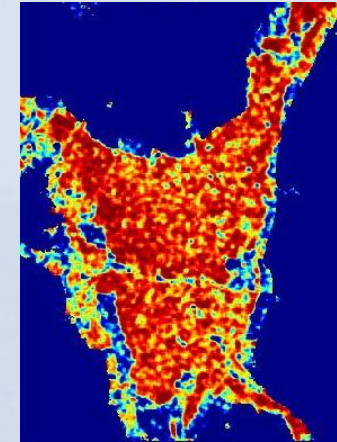
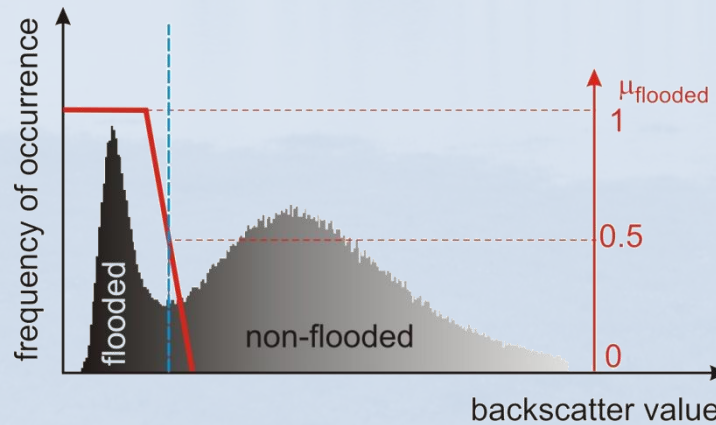
Solution: 1. Make use of DEM



Objective 2: Include uncertainty in flood mapping from SAR



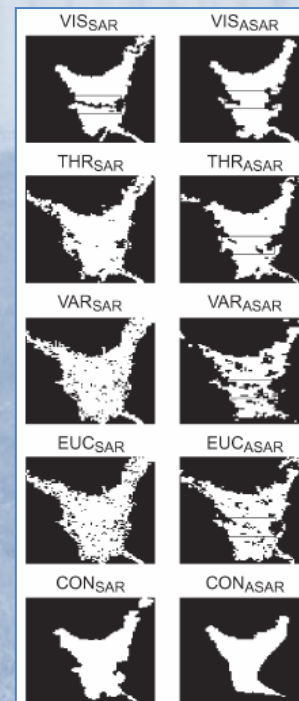
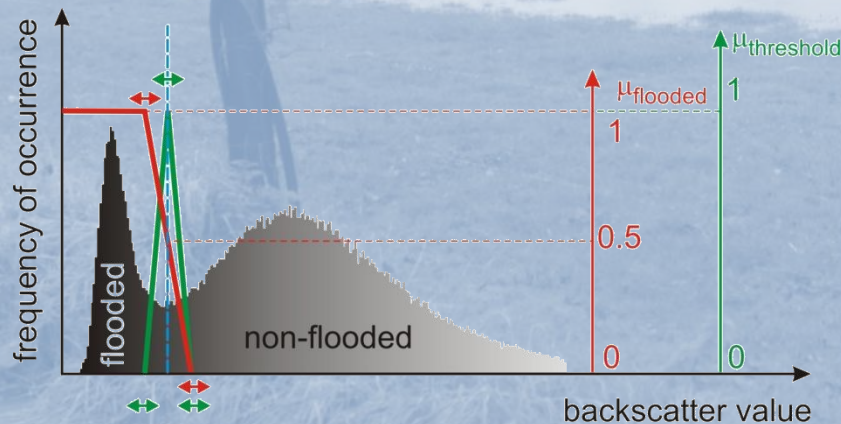
Problem: Floodplain delineation prone to errors and uncertainty
Solution: 2. Produce 'fuzzy' flood maps



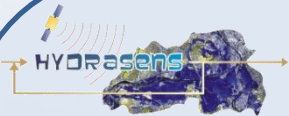
Objective 2: Include uncertainty in flood mapping from SAR

Science questions:

1. How to estimate uncertainty in a SAR-based flood map/flood edge height map?
2. How to merge flood maps from different algorithms?
3. How to fuse uncertain remote sensing-derived flood edges with a DEM?

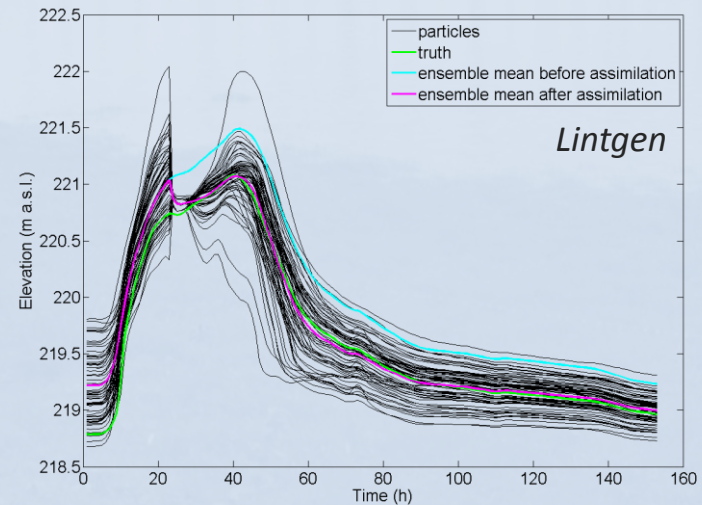
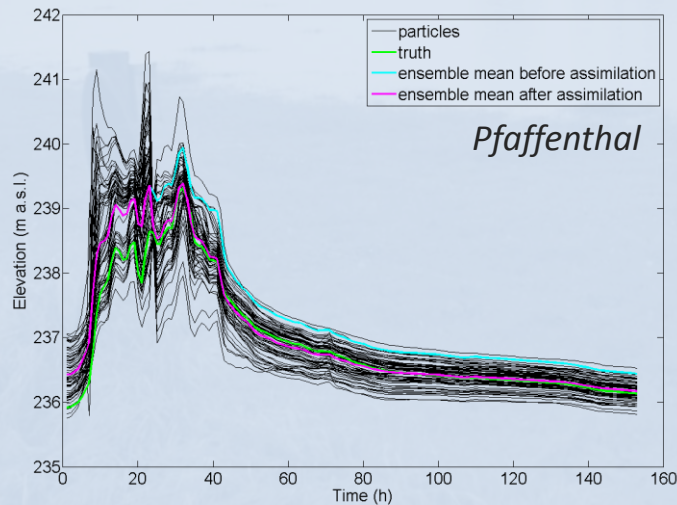


Objective 3: Joint assimilation of the remote sensing products



Problem: use remotely sensed soil moisture and flood maps for updating hydrologic and hydraulic modelling

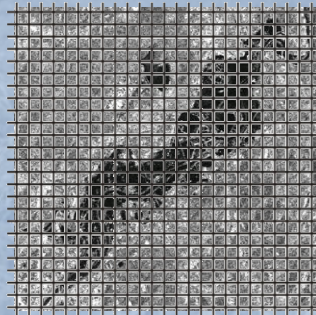
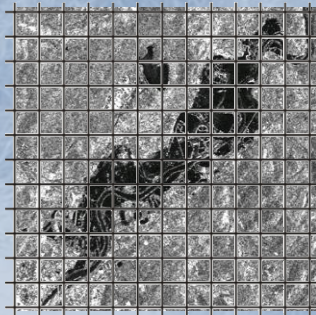
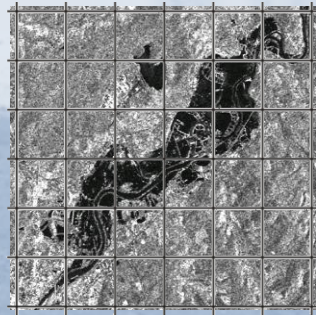
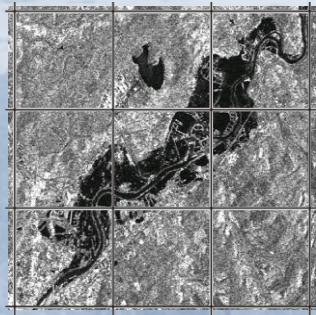
Solution: 1. development of coupled hydrologic-hydraulic model
2. development of particle filter-based data assimilation scheme



Objective 3: Joint assimilation of the remote sensing products

Science questions:

1. How to optimize the assimilation if observations and model have different scale?
2. How to account for uncertain water elevation measurements in a hydraulic model?
3. How to jointly assimilate soil moisture and water elevation?



Alzette River Basin (Luxembourg)

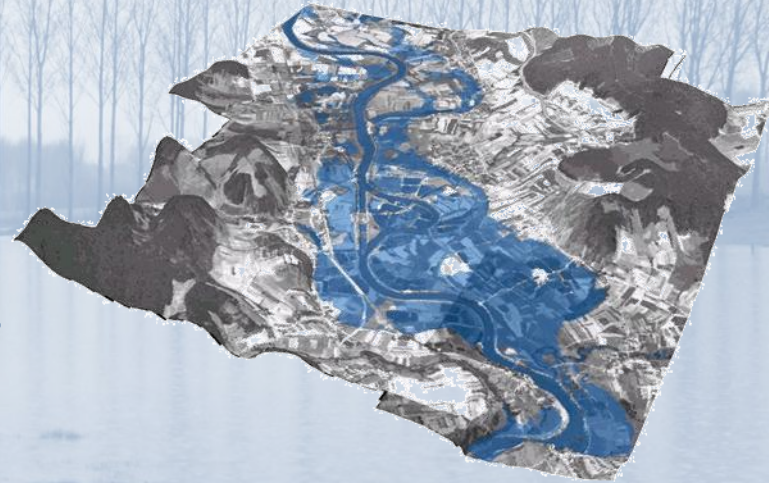


In situ datasets
 soil moisture
 flood maps and stages
 LiDAR data

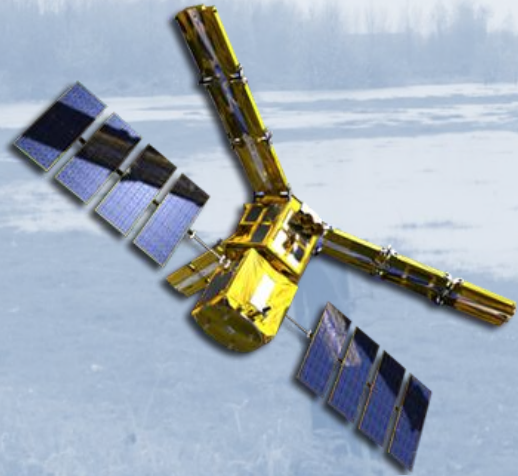
Remote sensing datasets
 ENVISAT, RADARSAT,
 TerraSAR-X, SMOS, ...

Hydrologic and hydraulic models
 CLM, HEC-RAS, LISFLOOD-FP, ...

Data assimilation schemes
 EnKF, particle filter



Mississippi and Murray-Darling basin

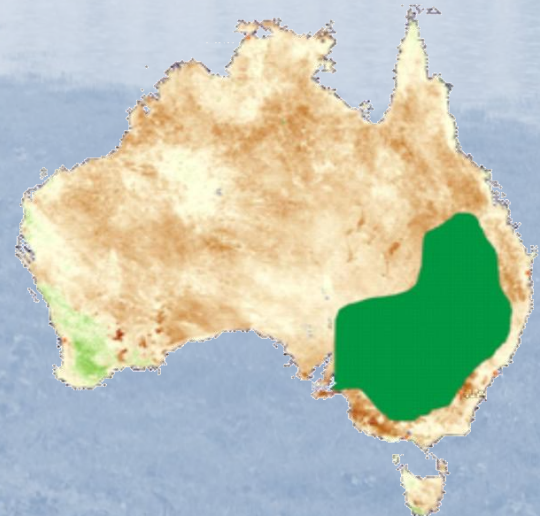
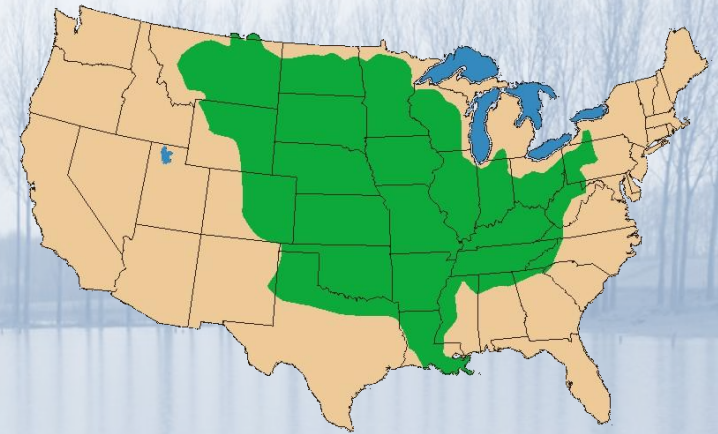


In situ datasets
soil moisture
flood maps and stages

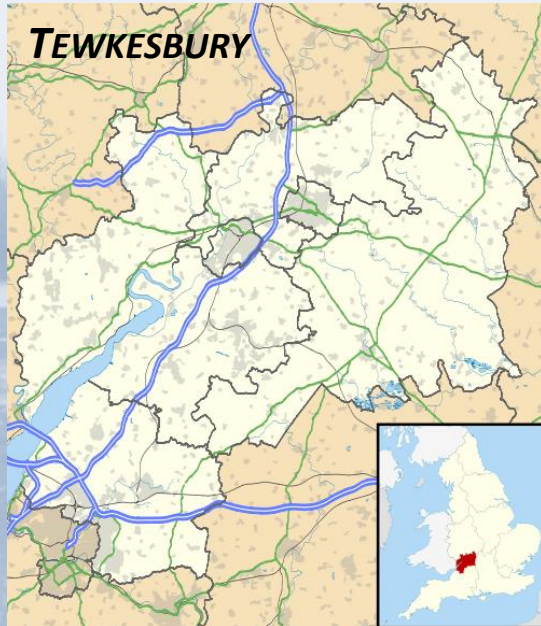
Remote sensing datasets
SMOS

Hydrologic model
VIC, ...

Data assimilation scheme
EnKF



Tewkesbury, Upton and Buscot (UK)



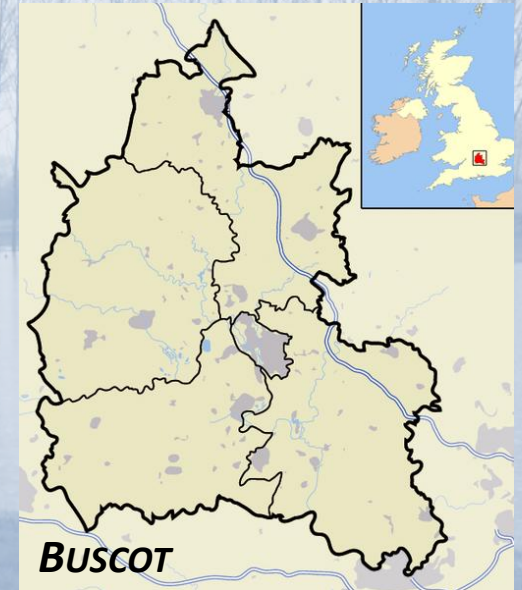
River Severn

In situ datasets
 flood maps
 flood stages
 LiDAR data



River Severn

Remote sensing datasets
 ENVISAT, RADARSAT,
 TerraSAR-X, ...



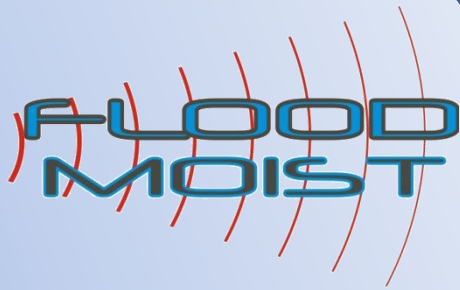
River Thames

Hydrologic and hydraulic
 models
 HBV-Light, HEC-RAS,
 LISFLOOD-FP, ...



RESULTS

... Forthcoming ...



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thank you for your attention

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