

# Verification of flood wave arrival time predictions using remote sensing-derived water levels

Stefania Grimaldi\*, Ashley Wright, Jeffrey Walker, and Valentijn Pauwels.

\*Stefania.Grimaldi@monash.edu



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# Rationale, study aims, and hypothesis

- 2-Dimensional hydraulic models are essential tools for floodplain inundation modelling.
- Remote sensing-derived flood extent and level (at the wet/dry interface) have been increasingly used for the calibration and validation of hydraulic flood forecasting models.
- Due to their uncertainty and discrete temporal coverage, remote sensing-derived data have been so far seen as a complement to field data.

**This study presents a novel remote sensing (RS)-based methodology for the calibration of 2-Dimensional hydraulic flood forecasting models.**

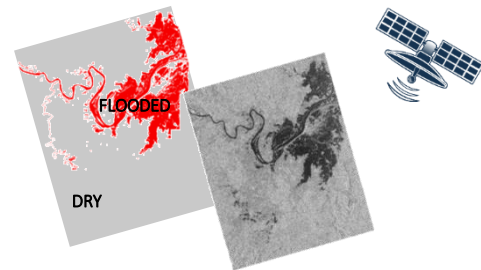
The methodology was designed to meet the following criteria:

- 1) **only RS** data are employed for model calibration;
- 2) discrimination between underprediction and overprediction of flood wave **arrival time**;
- 3) limited **computational time**;
- 4) the selected parameter configuration has to be **robust** for different events.

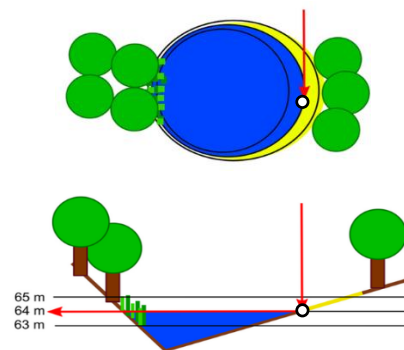
Hypothesis for the application of the methodology:

- roughness is the main parameter affecting flood dynamics.

## REMOTE SENSING-DERIVED (a) FLOOD EXTENT



## (b) WET/DRY BOUNDARY POINTS



## RS-based calibration: workflow

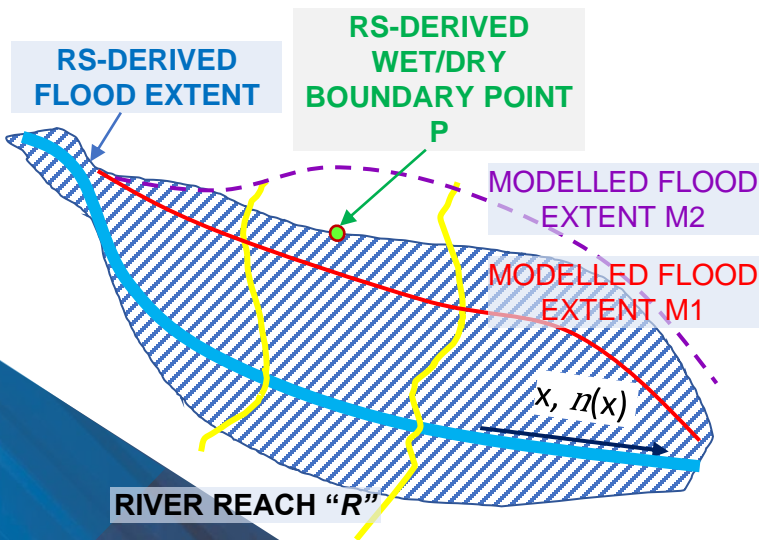
- ## 2 Computation of the performance metrics.

*E.g.:  $n=0.015\text{-}0.035\text{ m}^{1/3}\text{ s}^{-1}$  ( $\Delta n=0.0025\text{ m}^{1/3}\text{ s}^{-1}$ )*

2b) RS-DERIVED FLOOD EXTENT: binary performance metrics (literature).

- ### 3 Computation of novel set of river roughness values.

Steps 2 and 3 are repeated until there is no significant change in the computed river roughness values.



The Space-Time-Score (STS) quantifies:

- **how far** is the observed point P from the modelled flood extent (how far outside in M1/inside in M2) ;
- **how long it takes** for model M1 to reach P/for how long P was wet in model M2 → **VERIFICATION OF THE FLOOD WAVE ARRIVAL TIME**

STS<0 overestimation & early arrival time ( $STS_o, n_o$ )STS>0 underestimation & late arrival time ( $STS_u, n_u$ )

The STS allows comparing the performances of different model realizations for each river reach “ $R$ ”.

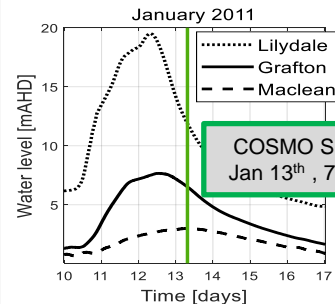
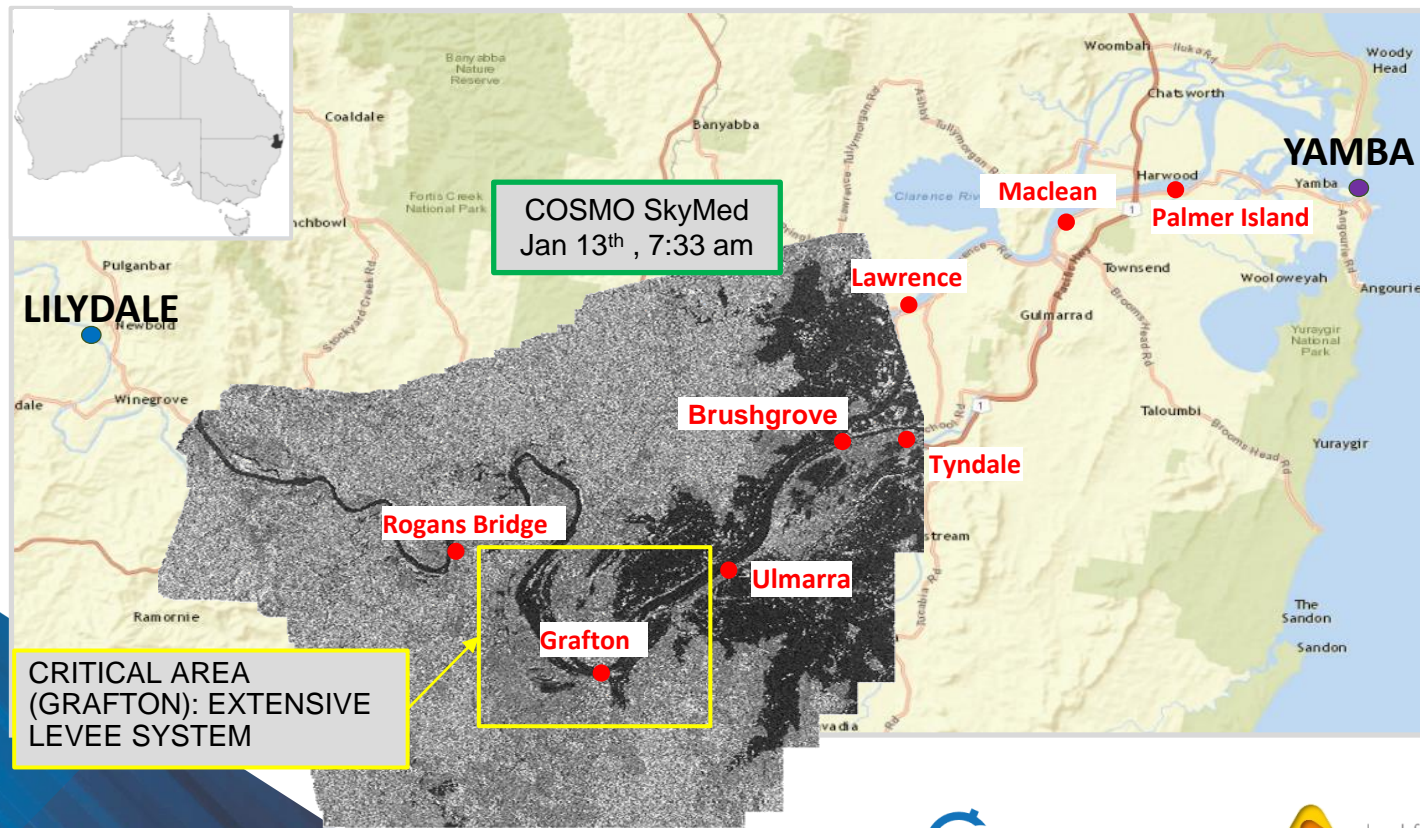
The novel roughness values aims to minimise the discrepancies between model and observations:

$$STC_{*n} - STC_{*n}$$

$$n_{new,"R"} = \frac{STS_u * n_o - STS_o * n_u}{STS_u - STS_o}$$

# Test case 1: Clarence River (Australia), January 2011

Remote Sensing observation: 1 high resolution image, good spatial coverage (63% of the modelled river length); which includes the critical area (Grafton); acquisition time: immediately after the flood peak.

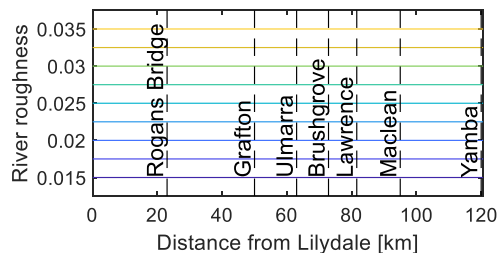


## Verification of the calibrated model:

gauged time series of water level are used as independent validation dataset.

# RESULTS: 2011 flood – schematic of the application of the calibration algorithm

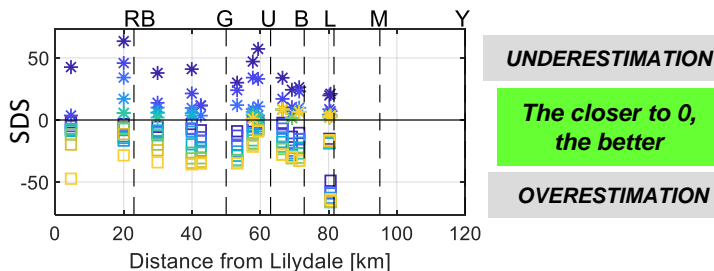
## 1 INITIAL SET OF MODEL REALIZATIONS



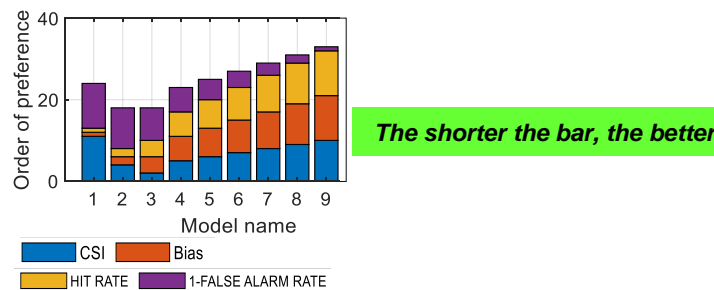
- Model 1:  $n=0.0150 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 2:  $n=0.0175 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 3:  $n=0.0200 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 4:  $n=0.0225 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 5:  $n=0.0250 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 6:  $n=0.0275 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 7:  $n=0.0300 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 8:  $n=0.0325 \text{ m}^{1/3} \text{ s}^{-1}$
- Model 9:  $n=0.0350 \text{ m}^{1/3} \text{ s}^{-1}$

## 2 COMPUTATION OF THE PERFORMANCE METRICS using REMOTE SENSING data

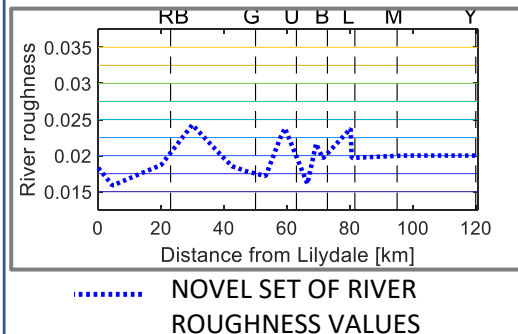
### 2a RS-derived WET/DRY BOUNDARY POINTS



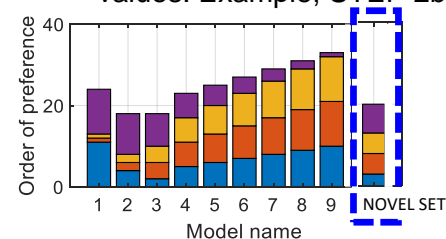
### 2b RS-derived FLOOD EXTENT



## 3 COMPUTATION OF THE NOVEL SET OF RIVER ROUGHNESS VALUES



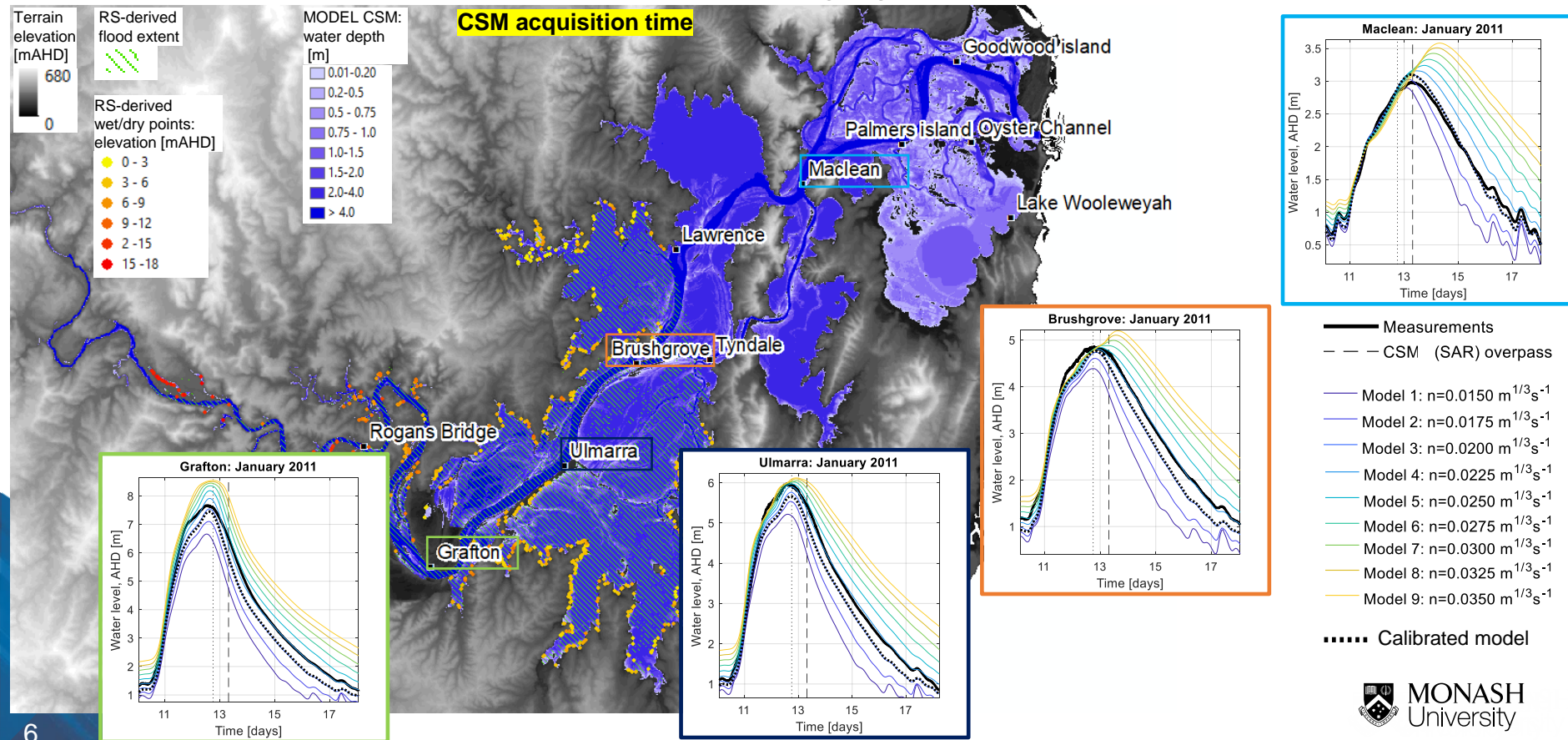
STEP 2 is used to evaluate the performances of the novel set of roughness values. Example, STEP 2b:





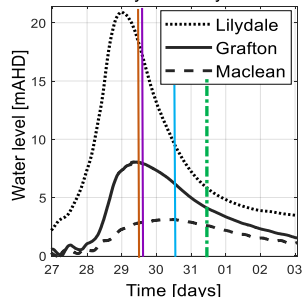
# RESULTS: 2011 flood – one “large” image including the critical area

**Calibrated model:** - comparison between modelled and RS-derived flood extent and wet/dry boundary points;  
- comparison between modelled and gauged water level (independent validation dataset).

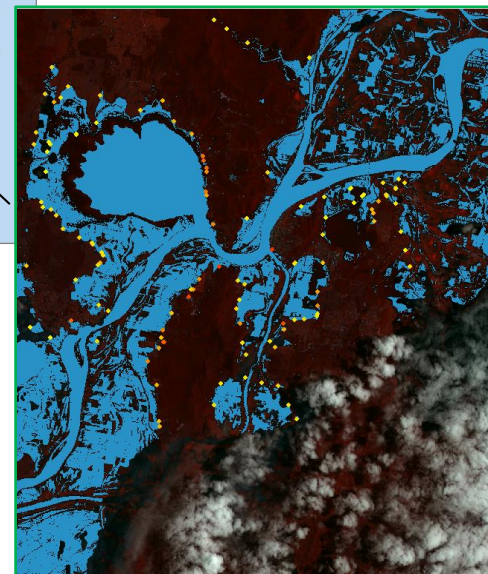
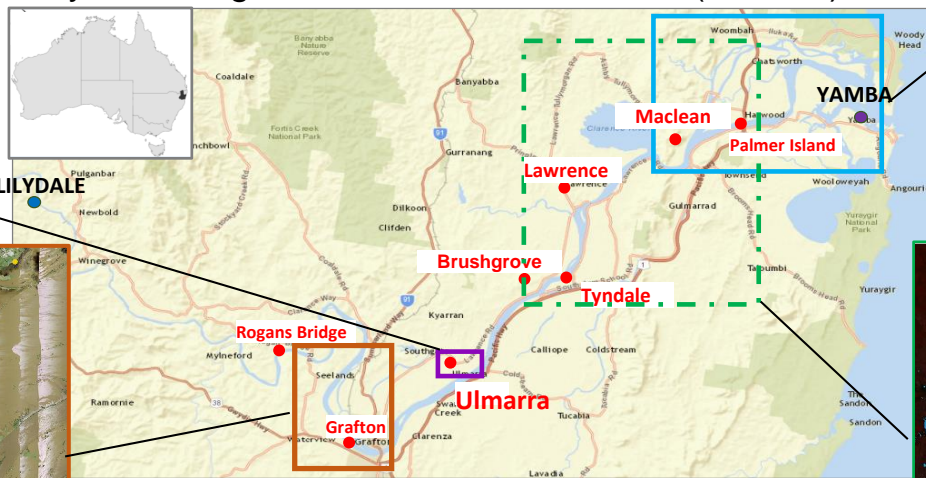


# Test case 2: Clarence River, January-February 2013

January-February 2013



Remote Sensing observations: 4 high resolution images, with limited spatial coverage (up to 18% of the river length), all acquired after the flood peak. Only one image included the critical area (Grafton).



RS-derived wet/dry points: elevation [m AHD]

- 0 - 3
- 3 - 6
- 6 - 9
- 9 - 12
- 12 - 15
- 15 - 18

Jan 29<sup>th</sup> h 7:32 AM



Jan 29<sup>th</sup> h 7:43 AM



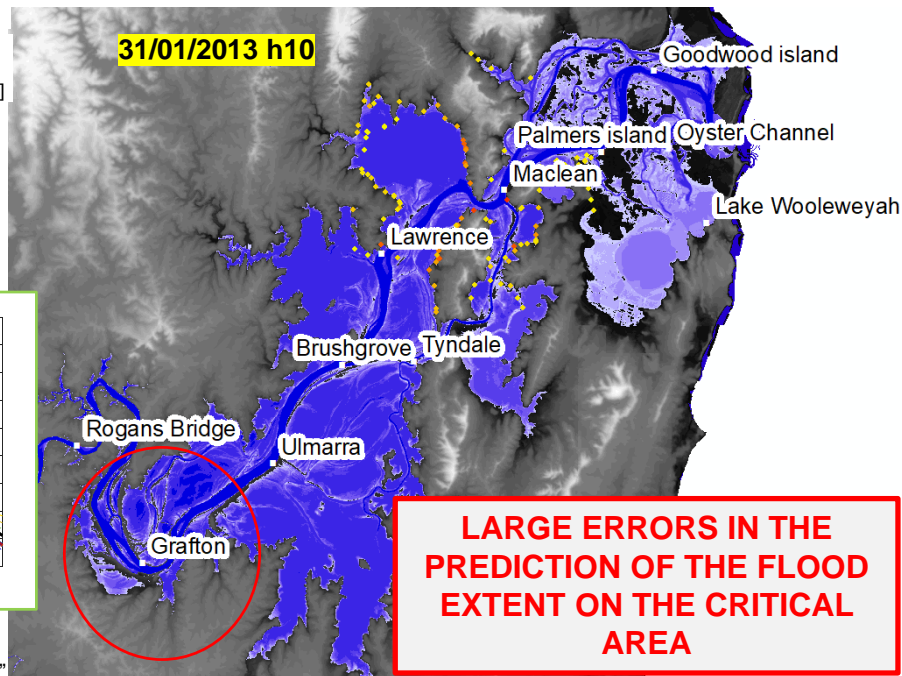
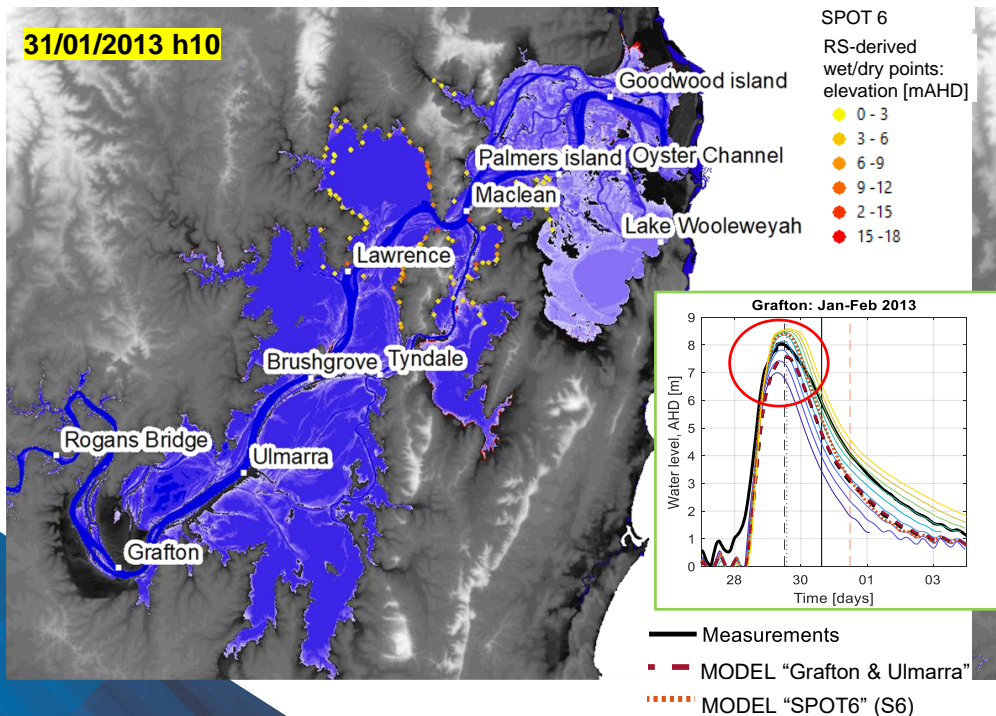
THE LEVEE SYSTEM WAS NEARLY OVERTOPPED



# RESULTS: 2013 flood – analysis of the importance of the acquisition footprint

Calibration using two small images (7.5% of river length) including **the critical area** (Grafton, levees):  
MODEL “Grafton & Ulmarra”

Calibration using a larger image (18%) acquired **downstream of the critical area**: MODEL “SPOT6”  
[figures: wet/dry points from SPOT6, 31/01/2013 h10]





# Summary and future work

- ❖ This study proposed a calibration methodology that makes exclusive use of RS-derived observations and consequently enables model calibration in ungauged catchments.
- ❖ RS observations are used in a two-fold manner: **1) flood extent; 2) wet/dry boundary points.**
- ❖ A novel performance metric (named Space Time Score) was introduced to discriminate between underprediction and overprediction of flood dynamics (in space and time).
- ❖ Differently from a Monte-Carlo approach, this methodology requires a limited number of simulations. Nevertheless there is a potential caveat: the iterations stop when all the available information has been used (but the model could still have poor accuracy).
- ❖ The analysis of a number of scenarios demonstrated the **importance of the footprint of RS acquisitions.**
- ❖ The accuracy of RS-derived observation and terrain data clearly affects the effectiveness of the calibration exercise.
- ❖ Future research include:
  - the analysis of the impact of remote sensing uncertainty on the effectiveness of the calibration methodology.
  - the testing of a large number of case studies to investigate the impacts of different catchment morphologies, flood dynamics, image resolution and accuracy on the results of the calibration methodology.