



Hydroinformatics in Urban Environment

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11 May 2014

UNIVERSITY OF
EXETER



THE SUNDAY TIMES
UNIVERSITY OF THE YEAR 2012-13
UNIVERSITY OF EXETER

Centre for Water Systems

- Established in 1998
- 30+ members (9 academic staff, 8 post-docs, 20+ PhDs/EngDs, 1 administrator)
- Current projects (~£4M):
 - 5 EPSRC (UK Research Council)
 - 7 EU projects (FP7/STREP/ITN)
 - IDC: STREAM (12 x EngD)
 - CDT: WISE (20 x PhD over 5 years)
 - 3+ Knowledge Transfer Partnerships



Centre for Water Systems

- Part of the multidisciplinary College of Engineering, Mathematics & Physical Sciences
- Research interests across the urban water cycle, with particular emphasis on:
 - **hydroinformatics**
 - **urban water management**
- Consultancy – wide range of projects & partners
- MSc in Water Management

Outline

- RAPIDS and CADDIES Projects(Intro)
- Case Studies
 - RAPIDS
 - Crossness, Portsmouth and Dorchester
 - CADDIES
 - Two EA Benchmark Test Cases
 - Two Real Test Case (up to 14 million cells)
 - Flooding from mains/sewers
- Summary

RAPIDS Project

RAdar Pluvial flooding Identification for Drainage System

- **Two sub-projects**
 - UKWIR – RTM Project (2011-12)
 - Real-time Machine Learning Approach to Near-term Assessment of Risk of Flooding in Urban Areas
 - EA – Bacti Project (2012-13)
 - Early Warning System for Prediction of Bacterial Concentration Exceedance in Tidal Waters

RAPIDS Team



Prof Dragan Savic



Prof Slobodan Djordjevic



Dr Edward Keedwell



Mr Andrew Duncan

Acknowledgments

• UKWIR – RTM Project

- HR Wallingford
 - Richard Allitt Associates
 - Halcrow
 - Mouchel
 - University of Exeter
 - UKWIR
- Coordination
 - Dorchester case study
 - Crossness case study
 - Portsmouth case study
 - ANN models
 - Funding

• Bacti Project

- Environment Agency SW
 - University of Exeter
 - South West Water
- Requirements + Data +
 - Decision tree models
 - ANN models
 - Advice



Overview

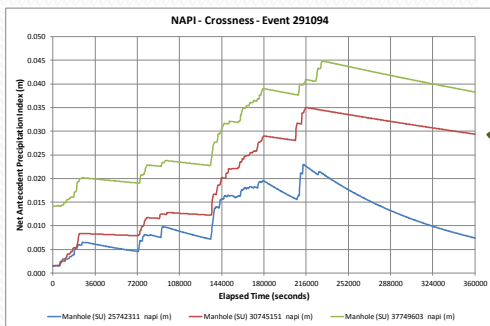
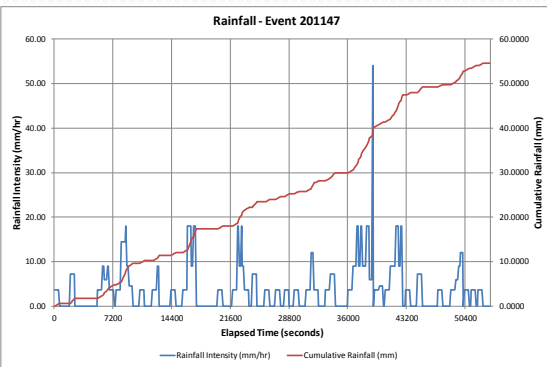
1. UKWIR RTM: Machine learning models of urban flooding – 3 case studies
2. RAPIDS: Artificial Neural Network (ANN) model
3. Bacti: Adapting RAPIDS to predict bathing water quality
4. Rainfall prediction
5. Future plans



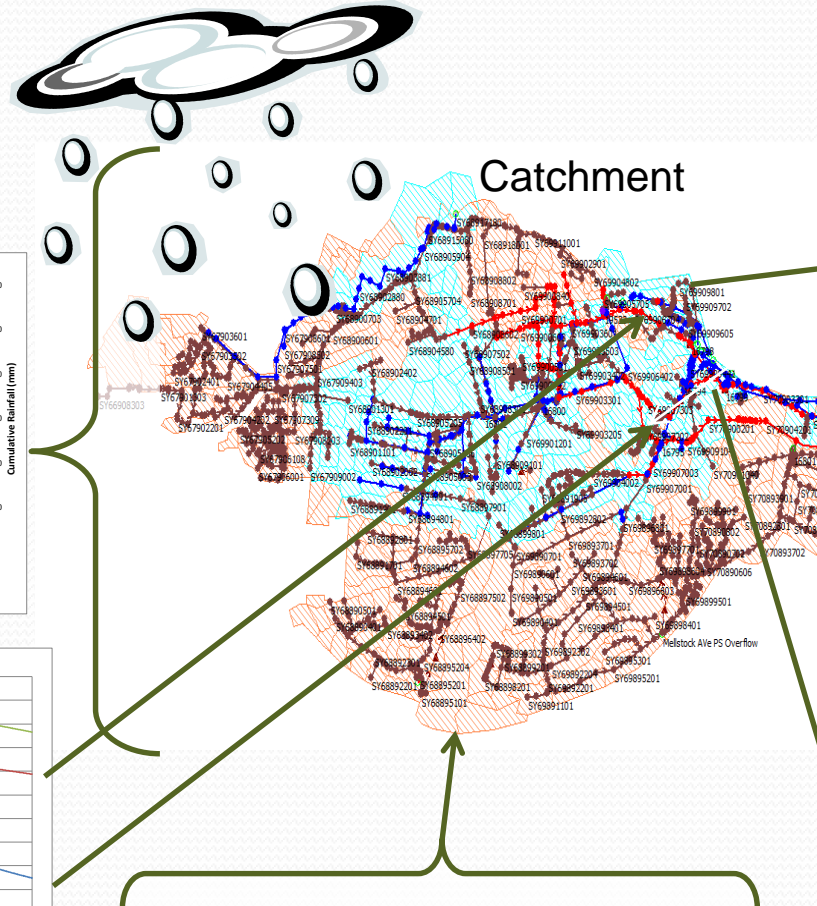
UKWIR RTM: machine learning models of urban flooding

- Scenario

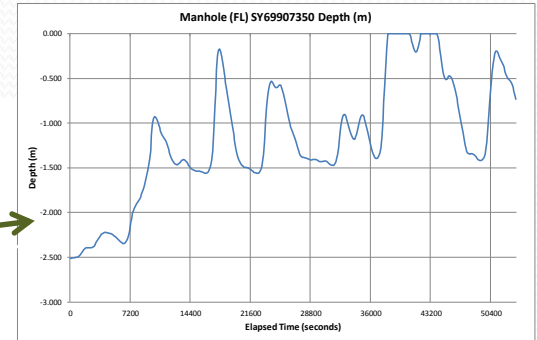
Rainfall Event



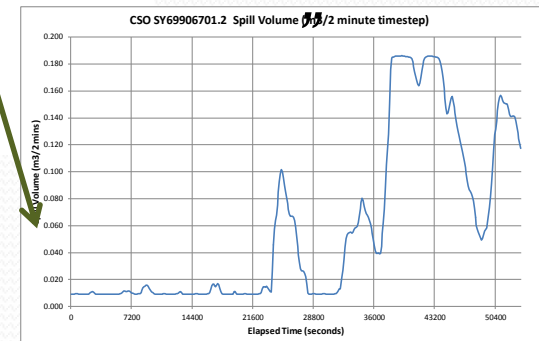
NAPI



Other Inputs: Tides, Pump States etc.



Manhole Depth Hydrograph
"Targets"



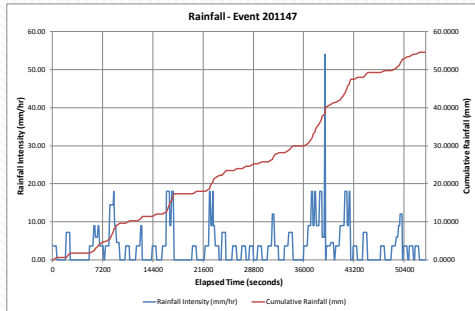
CSO Spill Volume Hydrograph

RAPIDS – Architecture

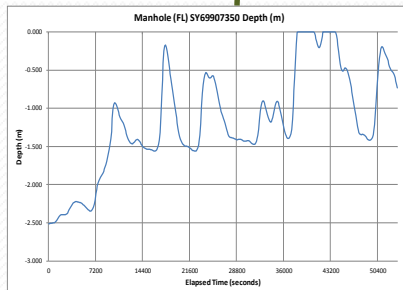
RAPIDS Program Setup Configuration Sheet

Note: Only fill in cells highlighted in light blue (other cells may contain formulae)

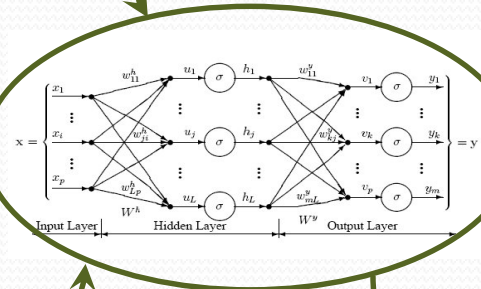
Catchment & Event Description	Value	Units
Training Event Data File	Stage2\data\Crossness_Halcrow_data\Train\Stage_2_Crossness_Train_Surcharged_MH_depthAD.xlsx	
Test Event Data File	Stage2\data\Crossness_Halcrow_data\Test\Stage_2_Crossness_Test_Surcharged_MHs_depthAD.xlsx	
Catchment Name	Crossness	
Rainfall Type	Spatial	
Training Event Spatial Rainfall File	Stage2\data\Crossness_Halcrow_data\Train\Stage_2_Crossness_Train_Spatial_rainfall_dataAD.xlsx	
Test Event Spatial Rainfall File	Stage2\data\Crossness_Halcrow_data\Test\Stage_2_Crossness_Test_Spatial_rainfall_dataAD.xlsx	
Columns of Spatial Rainfall Data	23	
Project Stage	2	
Summary Worksheet Name	Info	
Total Events (Test + Training)	50	
Target Header Range	A1:A53	
Target Data Range	A4:A51204	
Node Label Range	D2:A52	
Columns of Input Data in Target Sheet	3	
NAPI Value Used as ANN Input	FALSE	
Separate NAPI data files?	TRUE	
Training Event NAPI File	Stage2\data\Crossness_Halcrow_data\Train\Stage_2_Crossness_Train_NAPI_AD.xlsx	
Test Event NAPI File	Stage2\data\Crossness_Halcrow_data\Test\Stage_2_Crossness_Test_NAPI_AD.xlsx	
Columns of NAPI Data	40	



Input & Target Data Workbooks



ANN (MATLAB)



Trained Network Structures

ANN Manhole (SU) 29634201 Depth (m) with Target Hydrograph and 10% Limits of Amplitude

30h NSEC = 0.872

% Error of ANN Peak Amplitude = 10.6%

Time Error of ANN Peak (mins) = 4

HydroMAT Analysis

Crossness Design Event 50200 ANN Nash Sutcliffe Efficiency Coefficient (NSE) All Nodes (Prediction @ 0 mins)

Nash Sutcliffe Efficiency Coefficient (NSE) All Nodes

6 Hr NSEC for Depth (m) 6 Hr NSEC for 30 Hr NSEC for Depth (m) 30 Hr NSEC for

Confusion Matrix for Flood Depth Categories

Depth Category \ Target		Depth Category of ANN Prediction		
		A	B	C
Depth Category \ Target	A	1	0	0
	B	0	16	1
	C	0	0	21

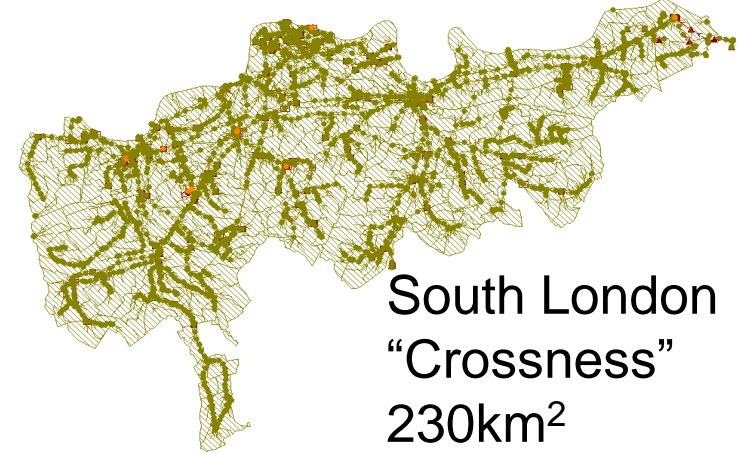
Flood Depth Category Key:

- A = Below Soffit
- B = Between Soffit and Basement Flood level
- C = Above Basement Flood Level

15 metrics

UKWIR RTM: 3 case study cities – focus of study

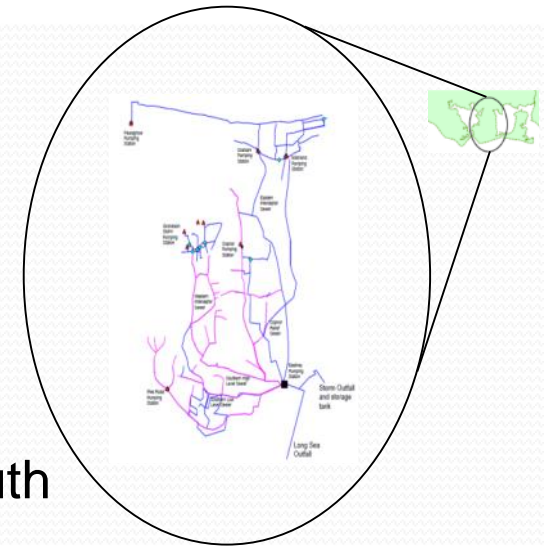
- **Crossness:**
 - Spatially variable rainfall
 - Up to 23 raingauges
- **Portsmouth:**
 - Predictive pump starting
 - Flood mitigation strategy
- **Dorchester:**
 - Relevance of soil moisture (NAPI) as model input



Dorchester
6km²

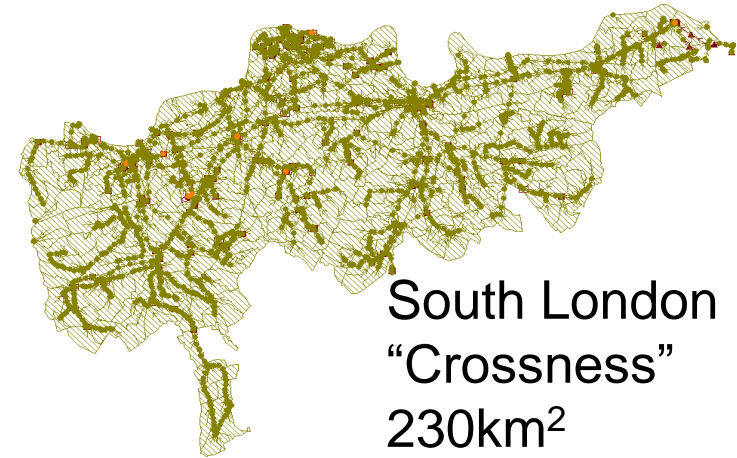


Portsmouth
29km²



UKWIR RTM: 3 case study cities - modelling

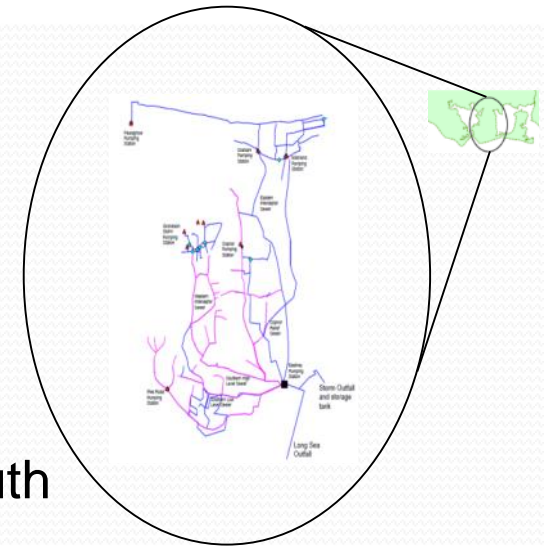
- **ANN Parameters:**
 - Input units (timesteps)
 - Hidden units
 - Optimisation strategies
- **Catchment Parameters:**
 - Times of concentration (ToC)
 - Hydrograph profiles
 - Rainfall events



Dorchester
6km²



Portsmouth
29km²



Results: Volume of data

Measurement Points:	Crossness	Portsmouth	Dorchester
Number of flooding manholes	20	17	20
Number of surcharged manholes	20	6	20
Number of CSOs	19	10	10

Measurement Parameters:

- Flooding volume
- Flooding depth
- CSO volume
- CSO depth
- Surcharged manhole depth

Number of Hydrographs:

Stage 1 = 952

Stage 2 = 1190

Total = 2142 !!!

Results: ANN model performance metrics

Nash-Sutcliffe Efficiency Coefficient (NSEC)
$$E = 1 - \frac{\sum_{t=1}^T (Q_o^t - Q_m^t)^2}{\sum_{t=1}^T (Q_o^t - \overline{Q_o})^2}$$

where Q_o is observed discharge, and Q_m is modelled discharge. Q_o^t is observed discharge at time t .

- Nash–Sutcliffe values can range from $-\infty$ to 1.
- A value of 1 corresponds to a perfect match between Target and ANN model results.
- A value of 0 indicates that the ANN model predictions are as accurate as the mean of the Target data.
- A value less than zero occurs when the mean of the Target data is a better predictor than the ANN model.

Criteria used for results evaluation

0.9 to 1.0	GOOD
0.5 to 0.9	ACCEPTABLE
<0.5	POOR

Results: Hydrograph goodness of fit (NSEC) results summary (~2000 results)

NSEC results		Portsmouth	Crossness	Dorchester (I)	Dorchester (II)
STAGE 1	CSO depth				
	CSO volume				
	Manhole level				
	Flood volume				
STAGE 2	CSO depth				
	CSO volume				
	Manhole level				
	Flood volume				
		Single timeseries rainfall	Spatial rainfall	Zero NAPI	Time-varying NAPI

- Stage 1 (design rainfall) results better than stage 2 (time-series rainfall) – but still largely acceptable (e.g. Portsmouth)
- ANN not capturing spatial rainfall input/response (e.g. Crossness)
- NAPI signal has minor influence on ANN performance (e.g. Dorchester)

Introduction

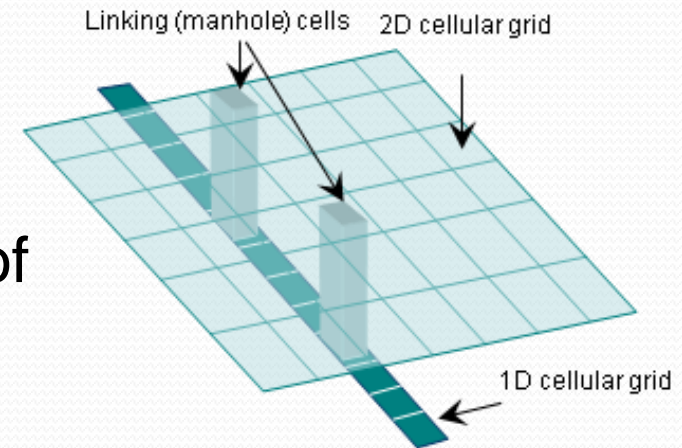
- Cellular Automata for 2D flood modelling
- 2 CA models
- Results comparison

EA Cases	Torquay Cases	Large Case
Multiple Models Multiple Hardware	InfoWorks ICM 3.0	InfoWorks ICM 3.0
UIM		UIM

UIM: in-house physical based
non-inertial urban inundation model

CADDIES Project

- £500k project funded by the UK EPSRC and industry (2010-2013)
 - Rapid, simplified dual-drainage modelling algorithms
 - Realistically capture the nature of flood dynamics over large urban areas



EPSRC

Engineering and Physical Sciences
Research Council



Halcrow



mouchel

City of Bradford MDC
www.bradford.gov.uk

UKWIR

SOUTH WEST WATER



United Utilities

UNIVERSITY OF
EXETER
Centre for Water Systems

TORBAY
COUNCIL

raa
Richard Allitt Associates Ltd
CONSULTING ENGINEERS

NORTHUMBRIAN
WATER

YorkshireWater

CADDIES Team



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Dr Edward Keedwell



Dr Albert Chen



Dr Bidur Ghimire



Dr Michele Guidolin



Dr Rebecca Austin



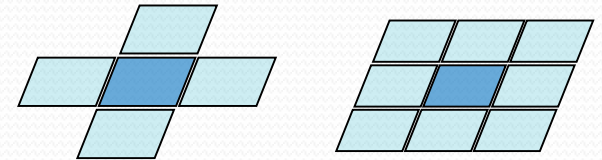
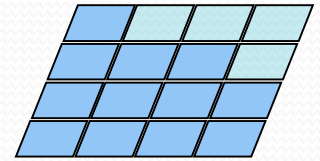
Mr Mike Gibson



Models

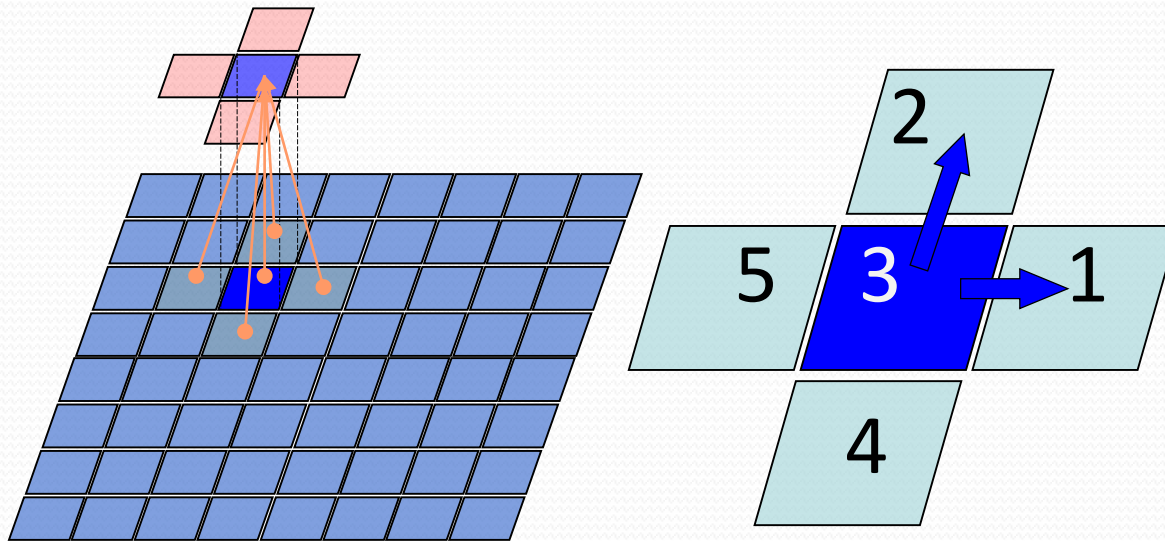
Cellular Automata

1. Discrete space
2. Cell states: discrete, continuous
3. Neighbourhood type
4. Local rules (deterministic and uniform)
5. Independent cell state updating (parallel)



CADDIES 2D Models

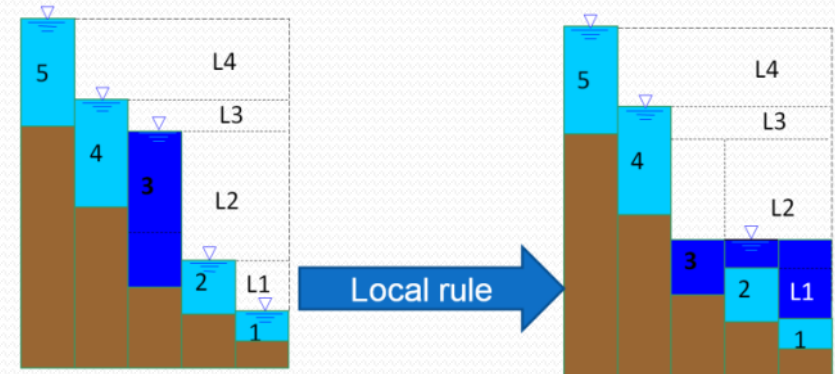
- The 2D CA models describes the surface flow using discretised cell states



CADDIES 2D Models

- **CA2D: First model (2011/2012)**

- Ranking technique to compute the volume of water transferred
- Expensive ranking algorithm
- Oscillation problems



Formulation of a Fast 2D Urban Pluvial Flood Model Using a Cellular Automata Approach. *J. Hydroinformatics* (2013)

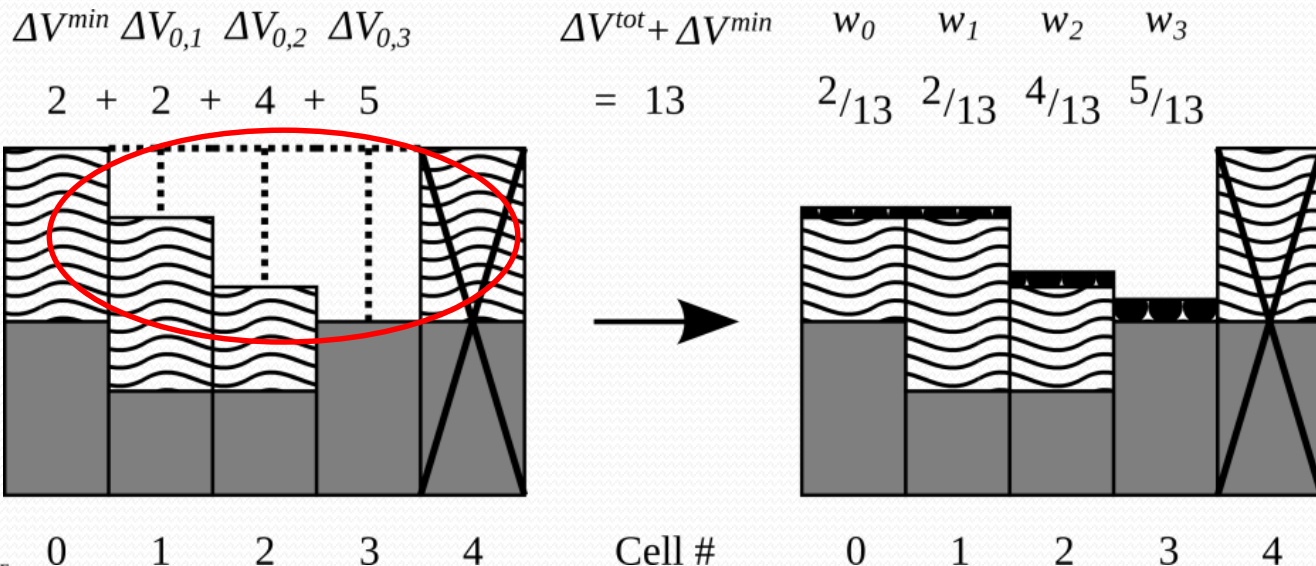
CADDIES 2D Models

- **Weighted CA2D: Improved model (2013)**
 - Quicker weight-based system to compute the volume of water transferred
 - Manning's equation applied to limit flux
 - Quicker with same accuracy of CA2D
 - Journal paper (under review):

CADDIES: a Streamlined, Weighted Cellular Automata 2D Inundation Model for Rapid Flood Analysis, submitted to *J. Env. Mod. & Soft.*

WCA2D Methodology

- For each neighbour cell:
 - Compute a weight that depends on the difference in water volume with the main cell

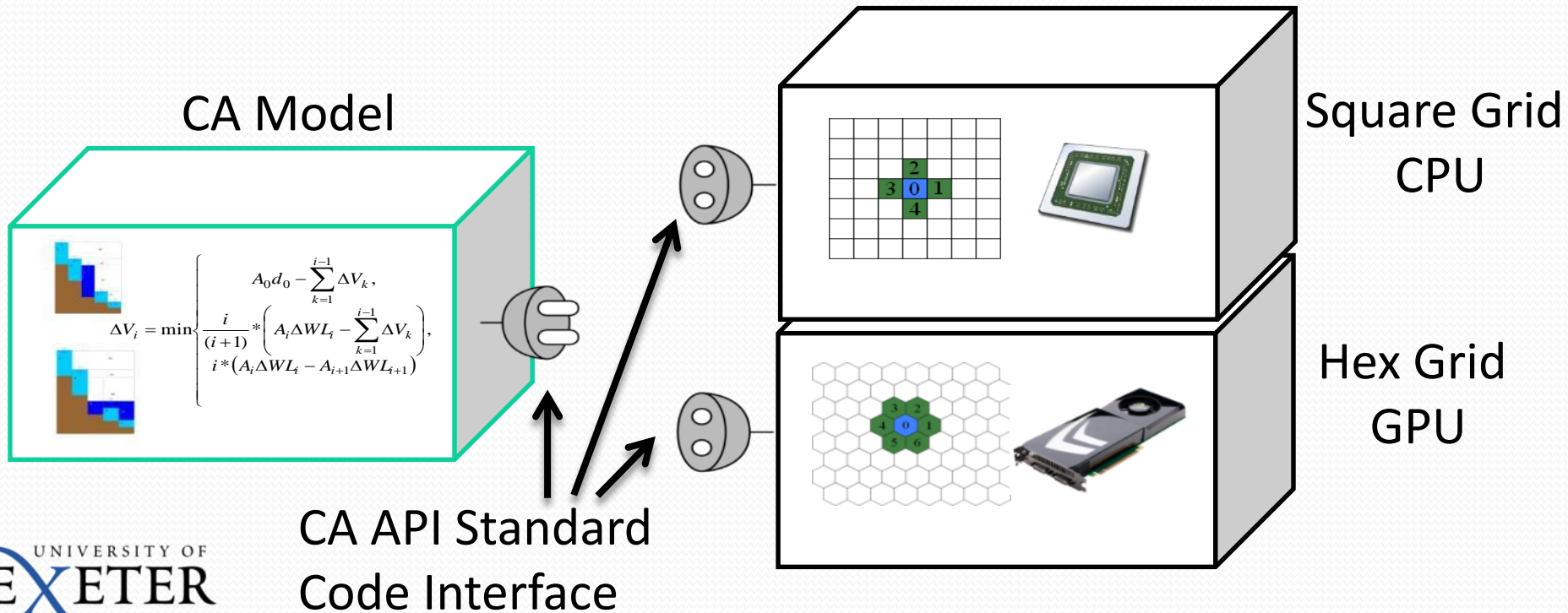


WCA2D Methodology

- The volume transferred between cells is capped by the Manning's formula
- The calculation is only applied to the neighbour cell with the largest weight to save computational cost
- Calculated once per cell

CADDIES Software Platform

- Integrates the numerical models with modern computing techniques



Results

EA Benchmarks

Benchmarking model

Urban Inundation Model (UIM), a physically based non-inertial 2D model based on shallow water equations

$$\frac{\partial d}{\partial t} + \frac{\partial ud}{\partial x} + \frac{\partial vd}{\partial y} = q$$

$$\frac{\partial(d+z)}{\partial x} + \frac{n^2 u \sqrt{u^2 + v^2}}{d^{\frac{4}{3}}} = 0$$

$$\frac{\partial(d+z)}{\partial y} + \frac{n^2 v \sqrt{u^2 + v^2}}{d^{\frac{4}{3}}} = 0$$

where,

q : lateral source term

u : velocity in x-direction

v : velocity in y-direction

d : flow depth

z : bed elevation

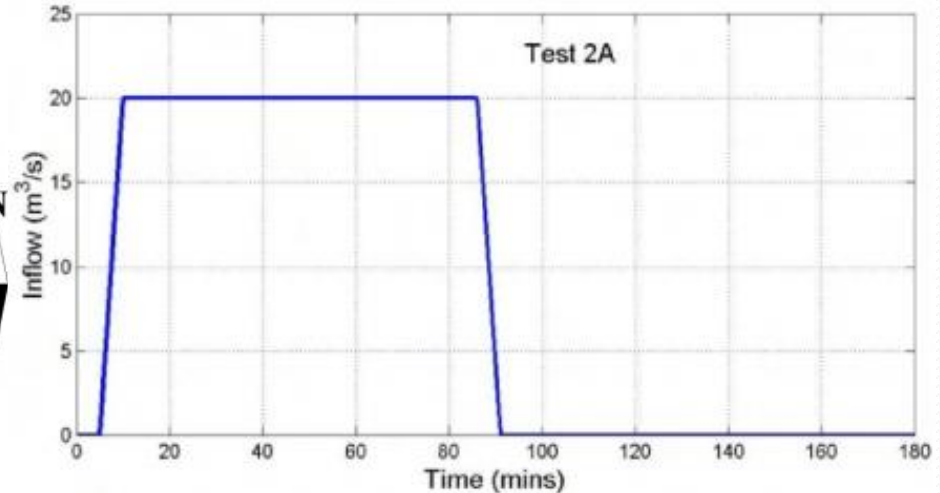
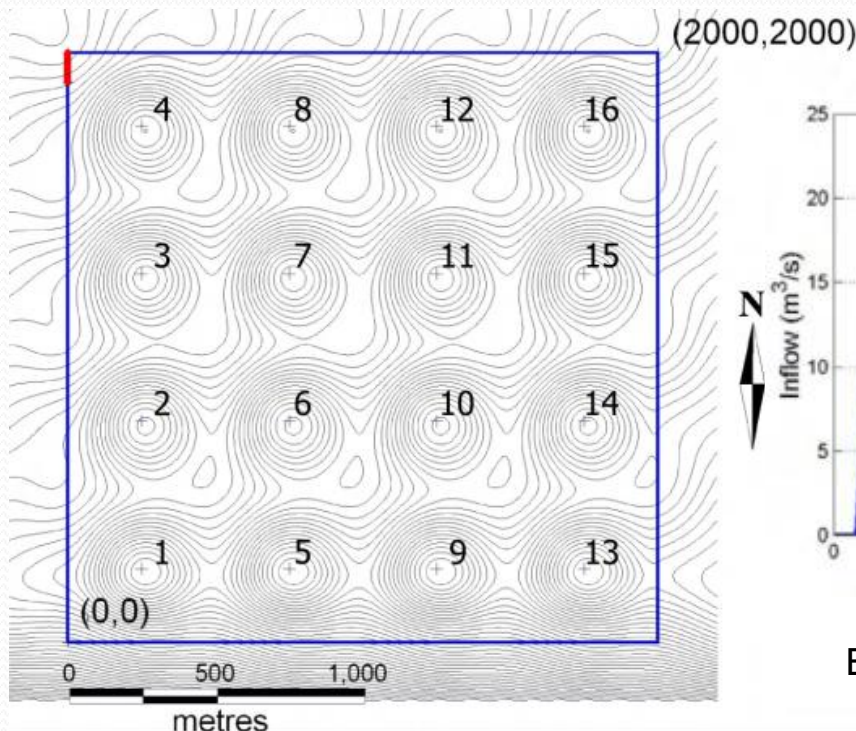
n : Manning's Roughness

x, y, t : space and time coordinates

EA Benchmarks Test cases

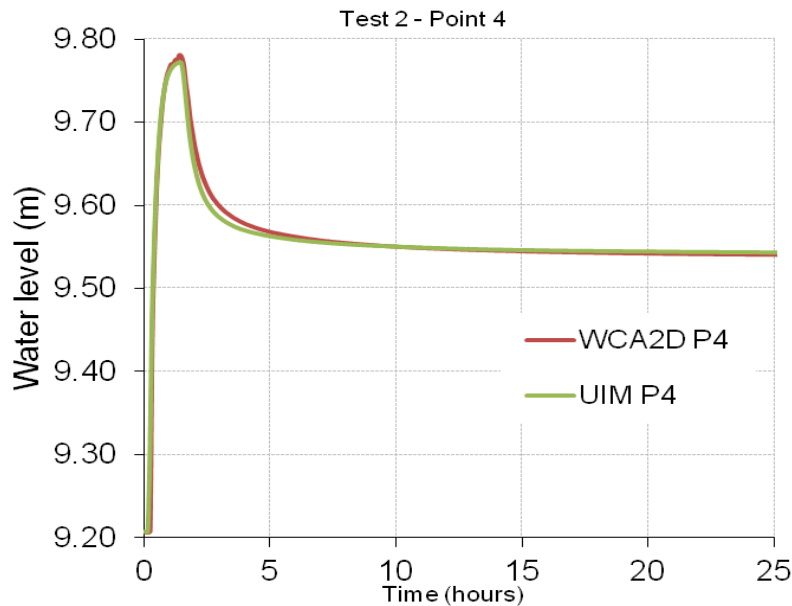
EA Benchmarking TEST2

Terrain (plan) gently sloping (NW to SE) area with 4x4 matrix of ~0.5m deep depressions

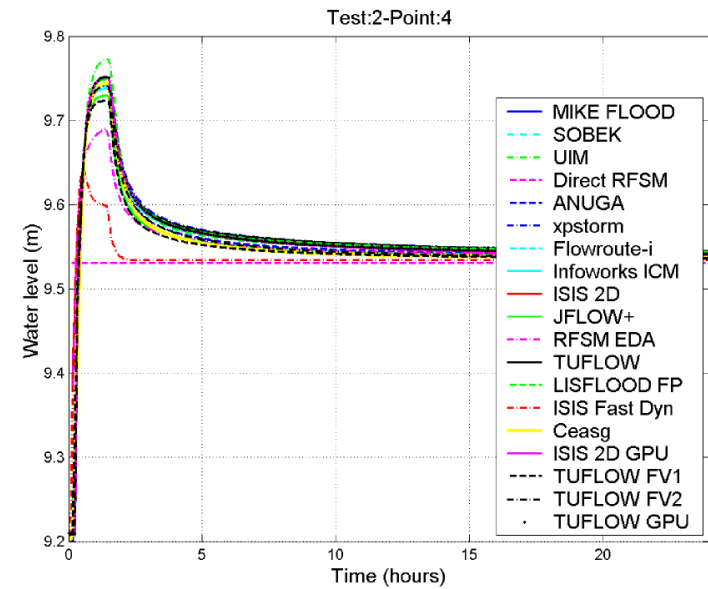


Boundary Condition: Inflow

Results: WCA2D – EAT2

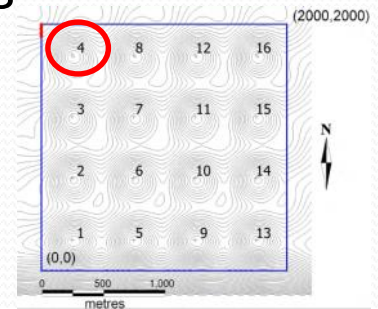


WCA2D And UIM

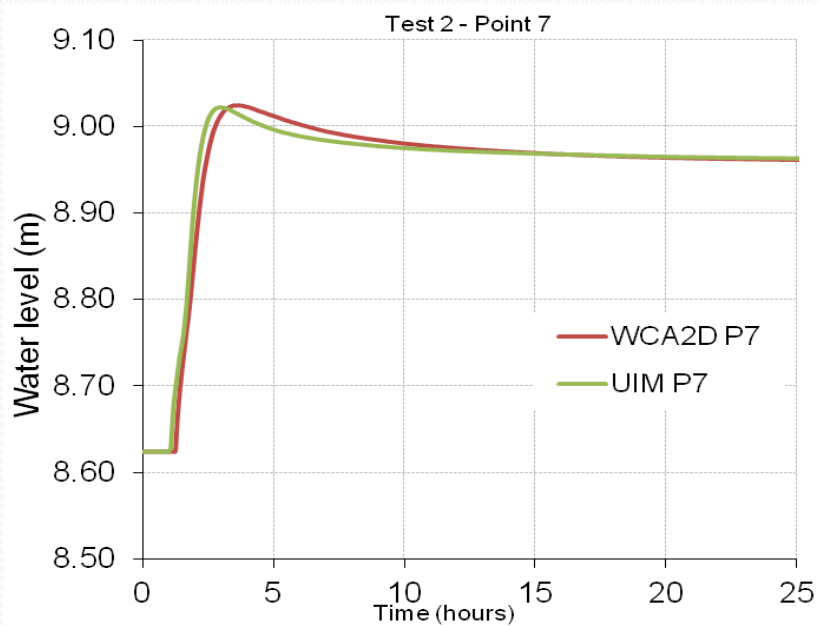


Multiple Models

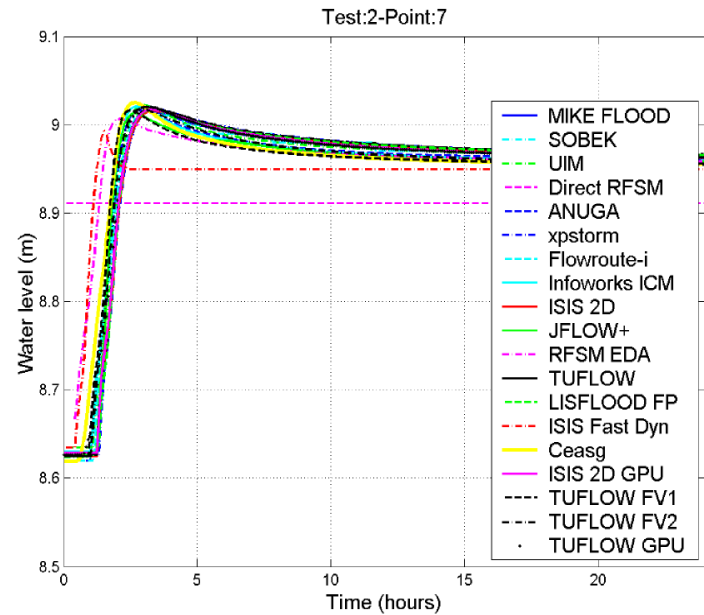
Point 4



Results: WCA2D – EAT2

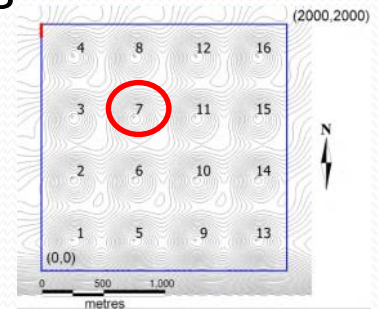


WCA2D And UIM

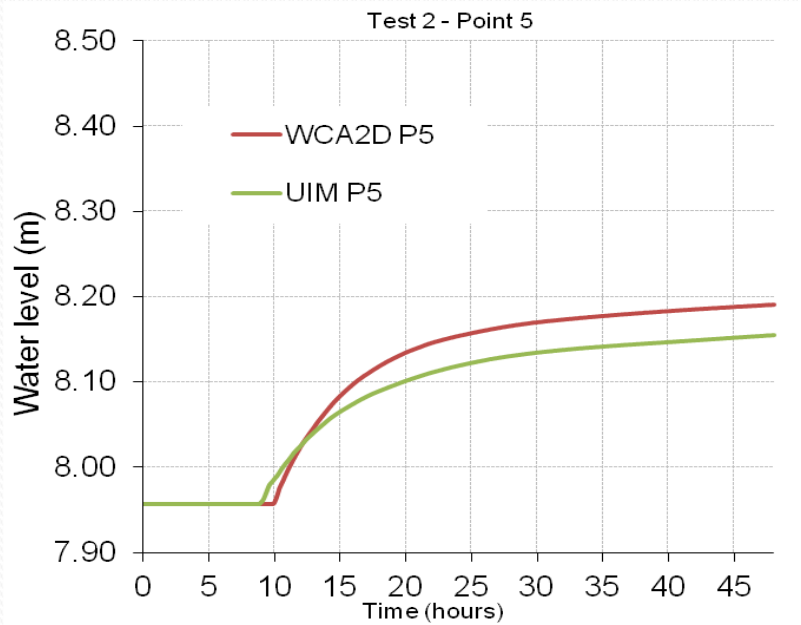


Multiple Models

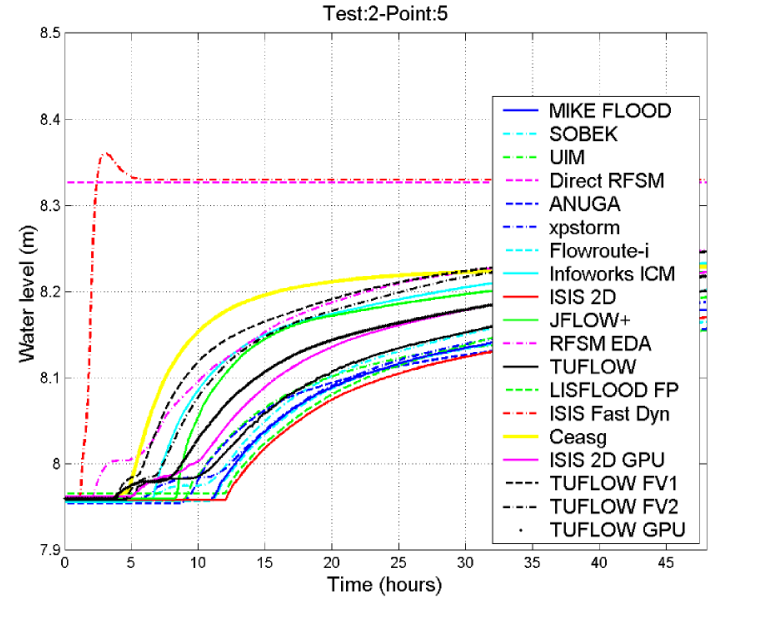
Point 7



Results: WCA2D – EAT2

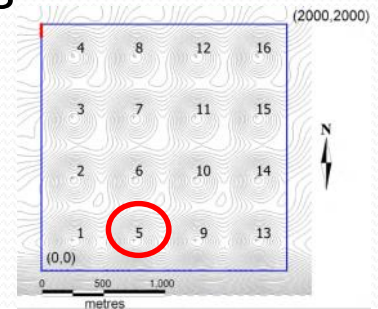


WCA2D And UIM



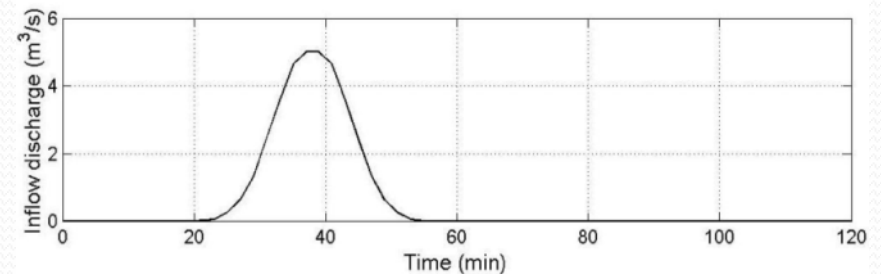
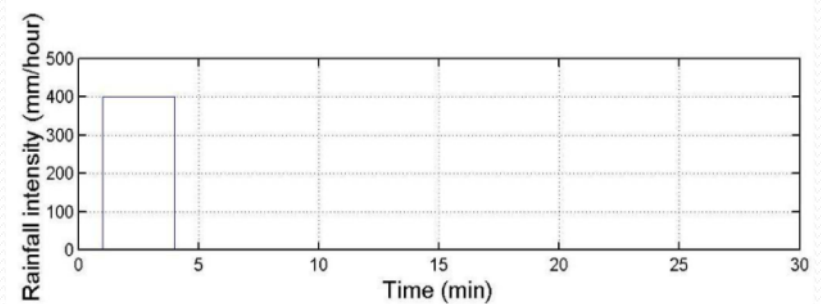
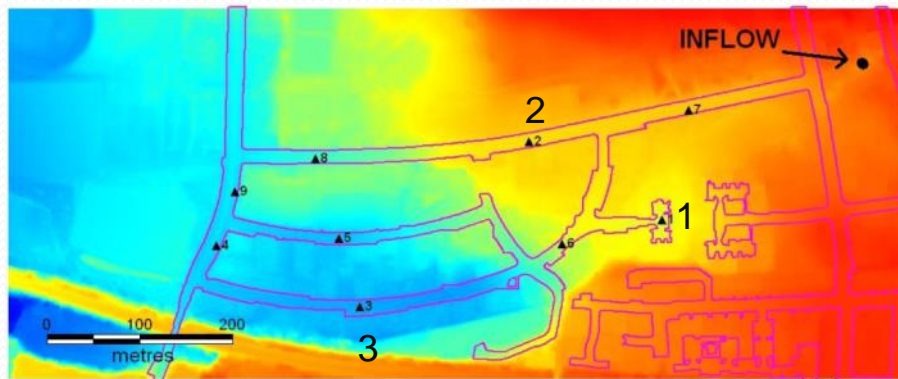
Multiple Models

Point 5



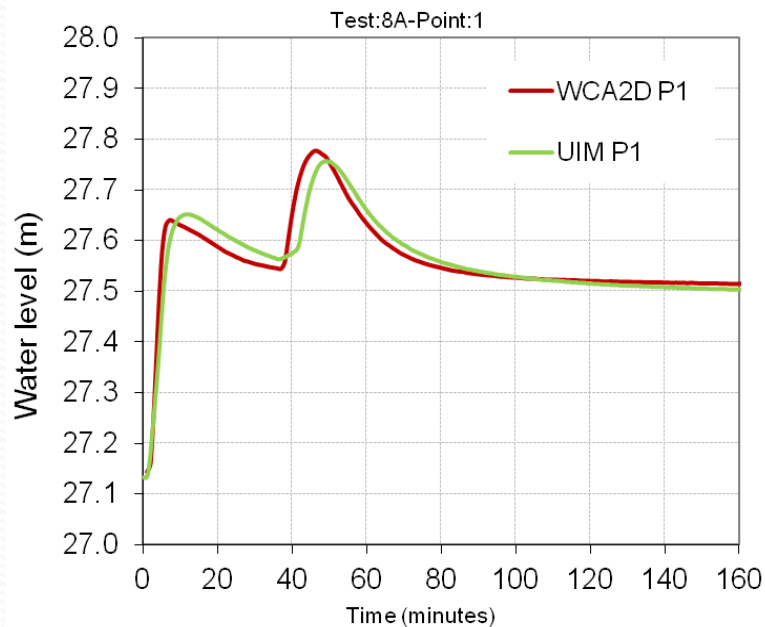
EA Benchmarking TEST8a

Terrain an approximately 0.4 km by 0.96 km urban area in Glasgow, UK

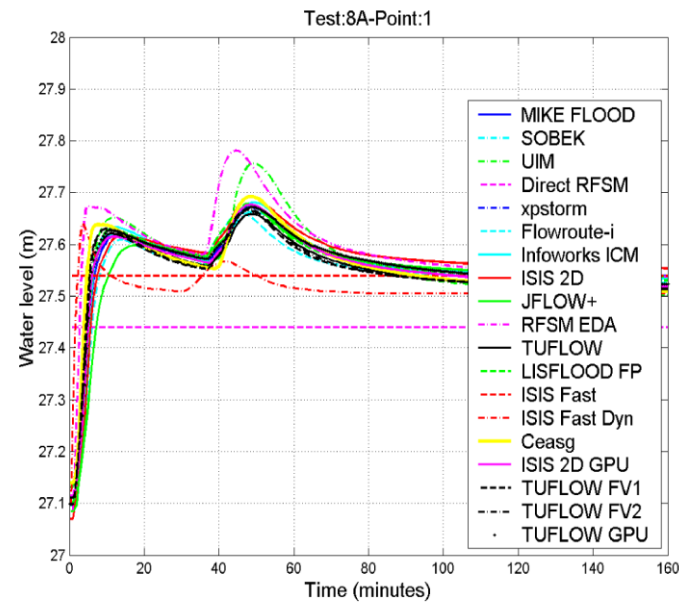


Boundary Condition : Rain (top), Inflow (bottom)

Results: WCA2D – EAT8a

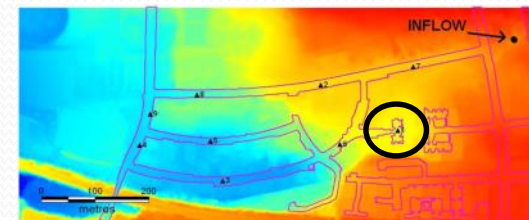


WCA2D And UIM

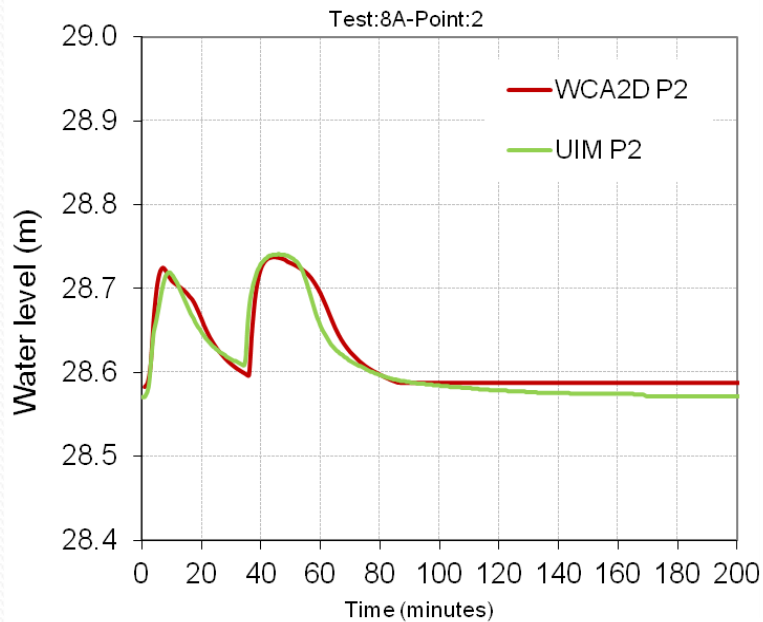


Multiple Models

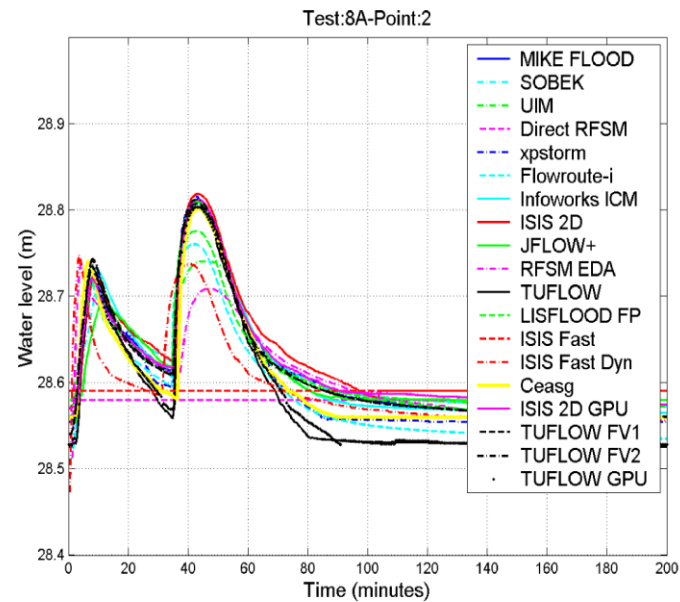
Point 1



Results: WCA2D – EAT8a

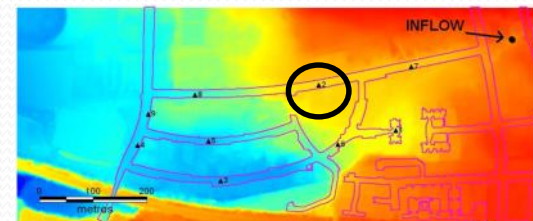


WCA2D And UIM

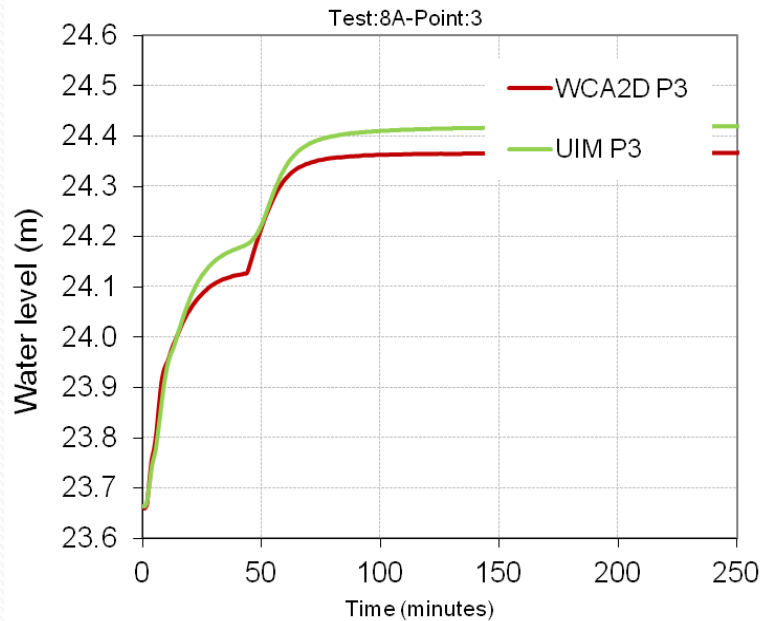


Multiple Models

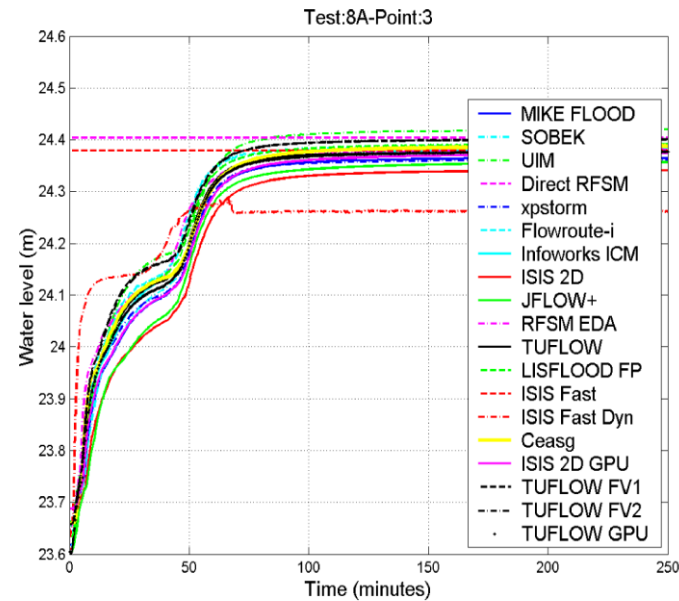
Point 2



Results: WCA2D – EAT8a

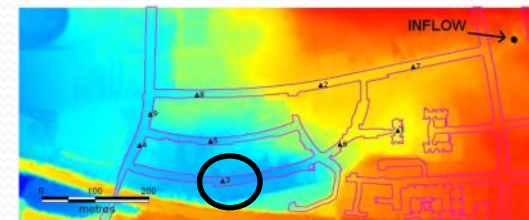


WCA2D And UIM



Multiple Models

Point 3



Computation Time

- EA report contains run times
- Achieved using different hardware
- Table shows the minimum, median and first quartile run time obtained by all models

	Run Time	
	EAT2	EAT8a
Minimum	2 s	66.0 s
Median	12.1 s	297.5 s
1 st Quartile	9.6 s	88.5 s
WCA2D GPU	3.84 s	33.9 s

Results

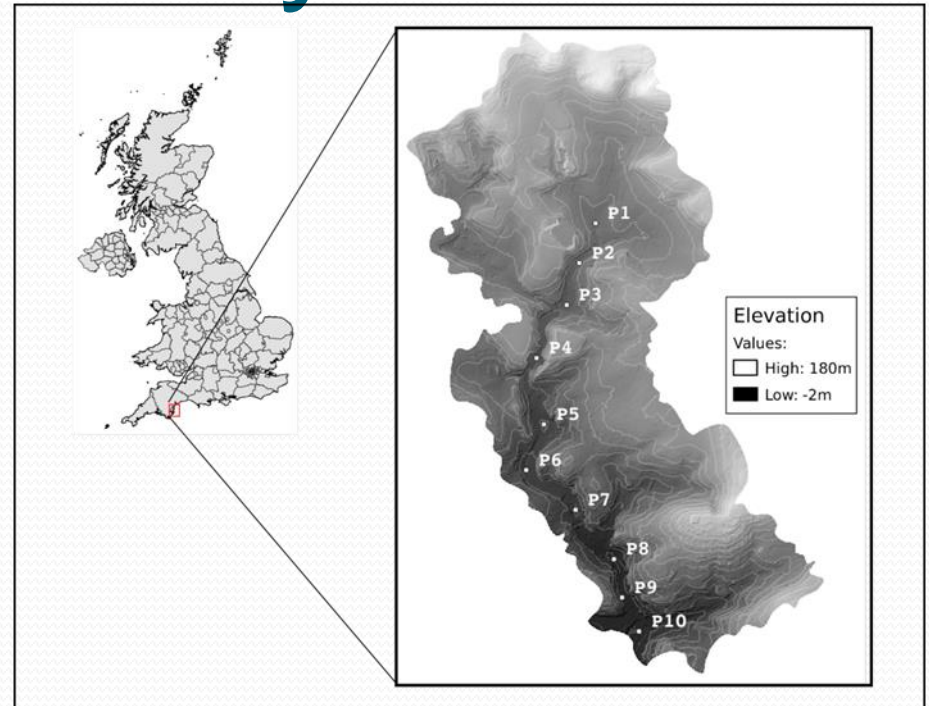
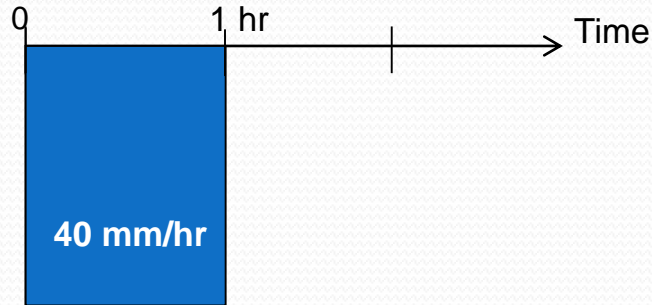
Torquay test cases

Torquay case study

Boundary Condition

Open boundaries

Rainfall: 40 mm/hr

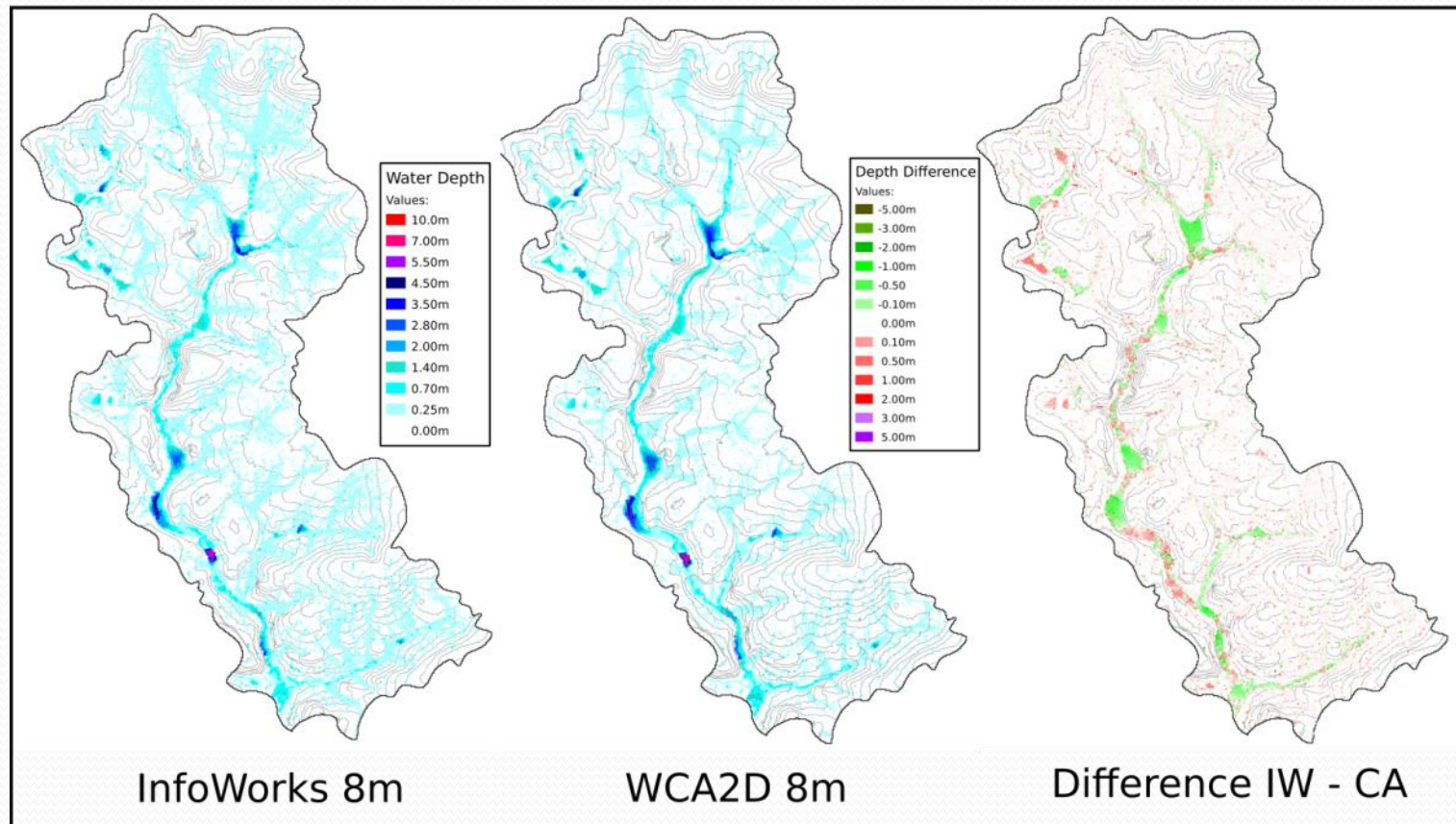


- 8m resolution: ~120,000 cells
- 4m resolution: ~500,000 cells
- 2m resolution: ~2,000,000 cells

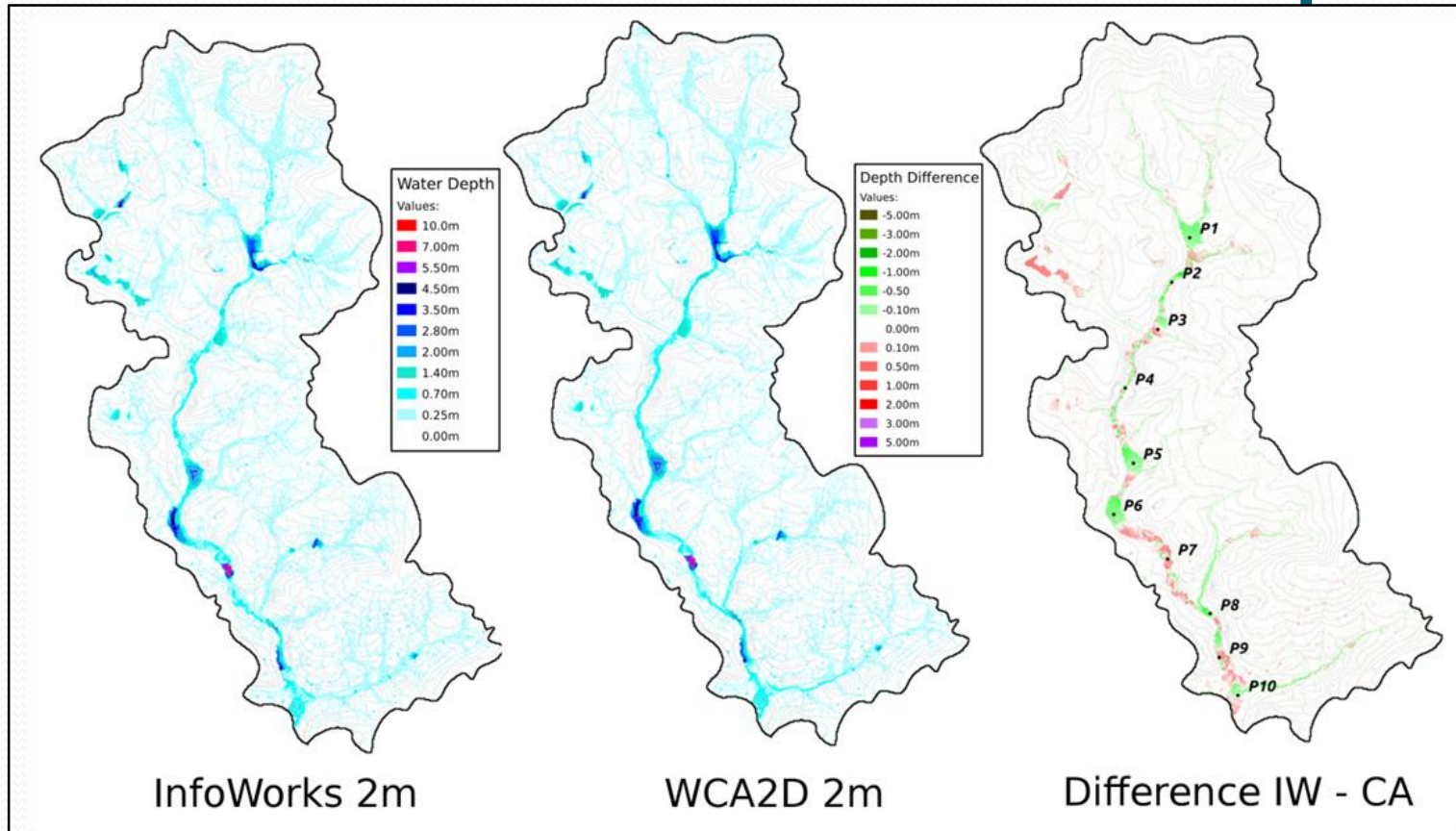
Results analysis

- Compared WCA2D
 - InfoWorks ICM 3.0
- Using three metrics:
 - Maximum absolute error (MAD)
 - Root mean square error (RMSE)
 - R-squared (R^2)

Maximum Inundation Depth



Maximum Inundation Depth

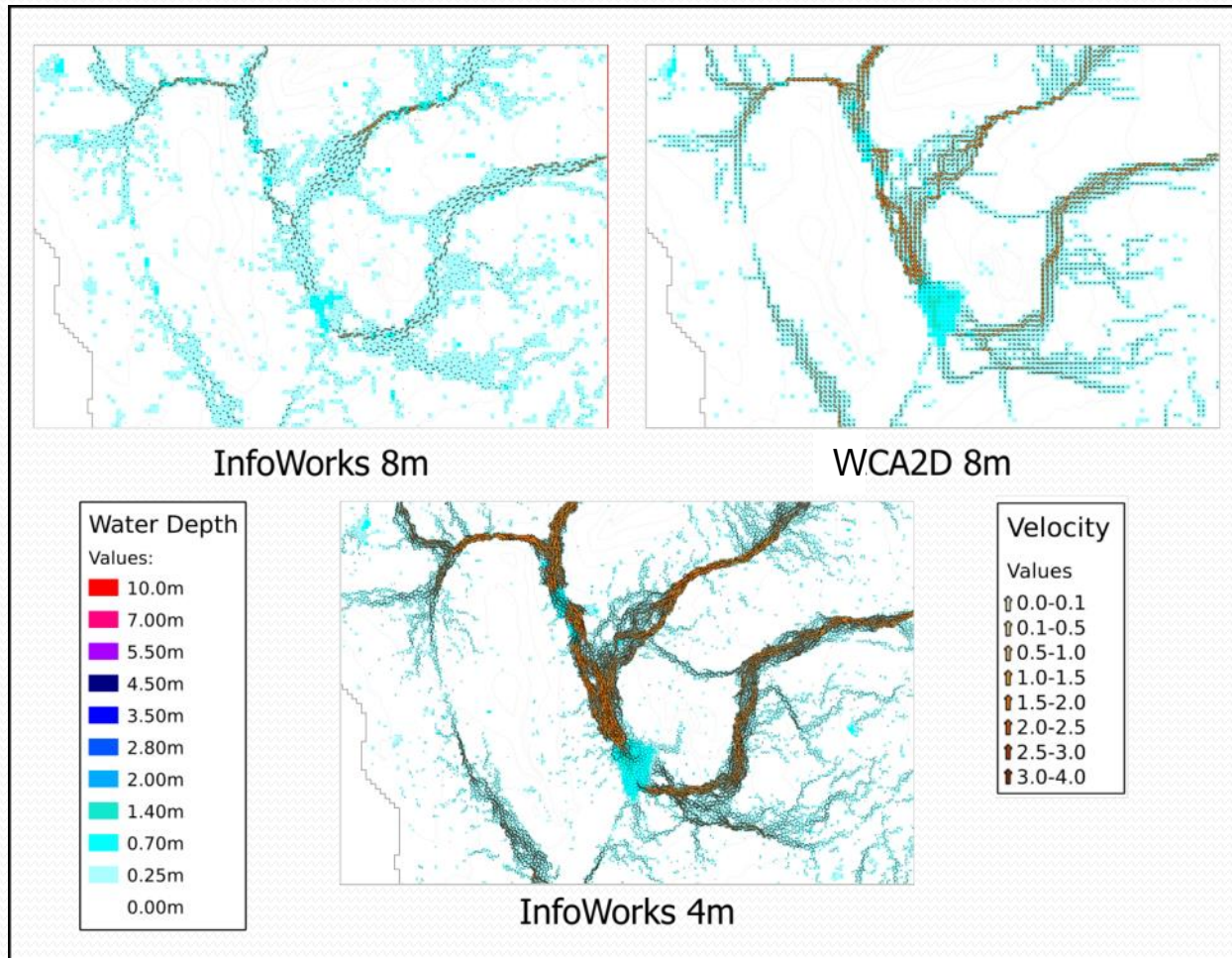


Results vs InfoWorks

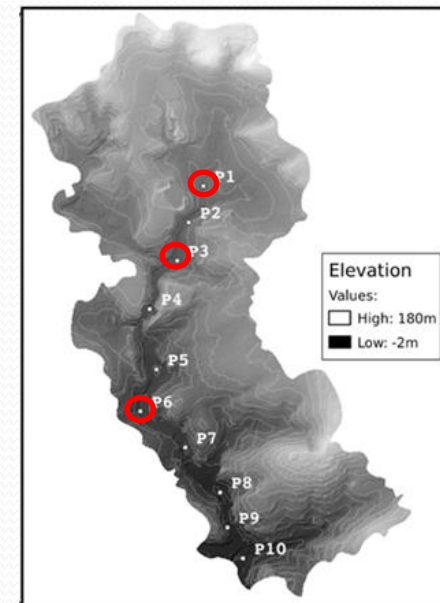
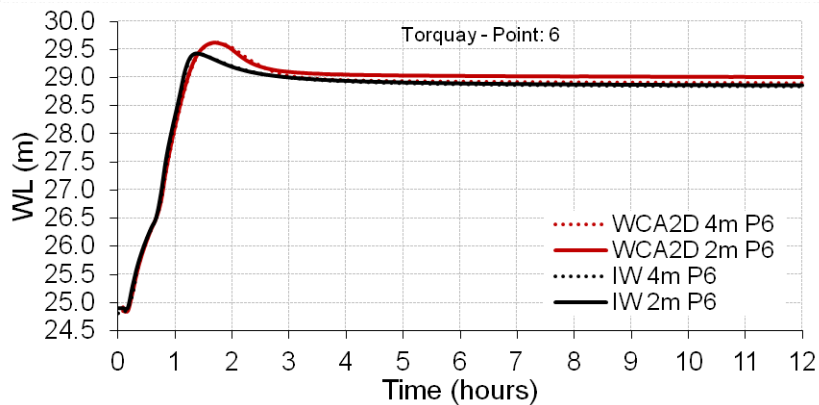
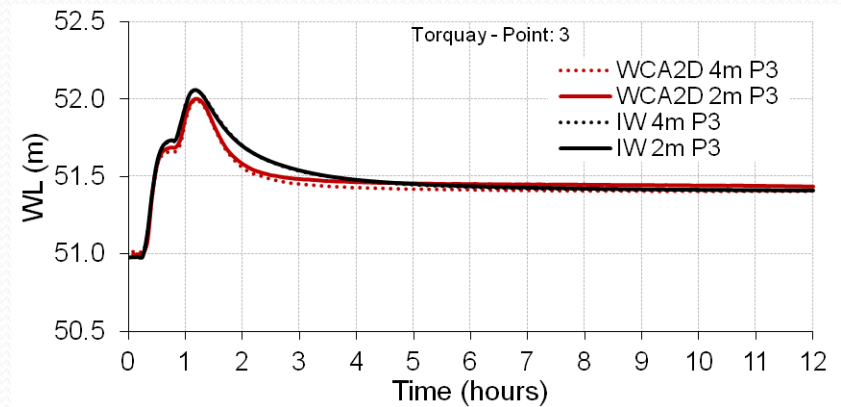
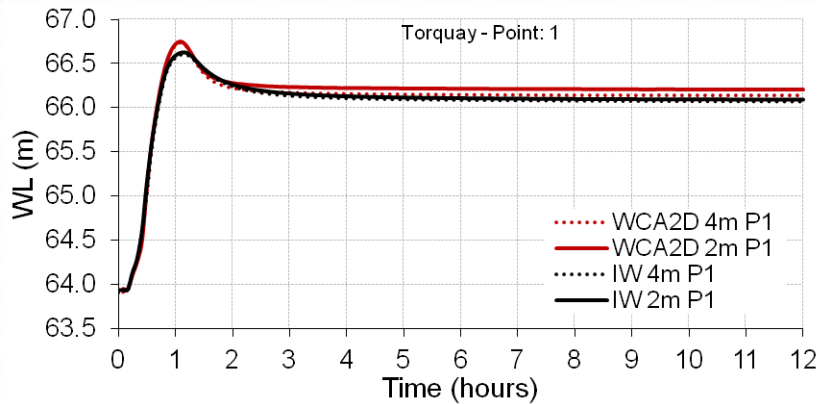
Models comparison time / attribute	IW 8m – WCA2D 8m			IW 4m – WCA2D 4m			IW 2m – WCA2D 2m		
	MAD	RMSE	R2	MAD	RMSE	R2	MAD	RMSE	R2
30 Min.	2.12 m	0.08 m	0.77	2.42 m	0.05 m	0.92	1.52 m	0.04 m	0.96
60 Min.	2.59 m	0.11 m	0.89	2.86 m	0.08 m	0.95	2.80 m	0.06 m	0.97
90 Min.	3.35 m	0.27 m	0.94	5.52 m	0.26 m	0.95	4.73 m	0.25 m	0.95
120 Min.	4.14 m	0.24 m	0.97	3.11 m	0.16 m	0.99	3.39 m	0.12 m	0.99
360 Min.	3.56 m	0.27 m	0.98	3.11 m	0.14 m	0.99	2.75 m	0.10 m	0.99
720 Min.	3.54 m	0.28 m	0.98	3.11 m	0.14 m	0.99	2.73 m	0.11 m	0.99
Max. Depth	4.08 m	0.13 m	0.92	3.12 m	0.09 m	0.97	3.41 m	0.07 m	0.98
Max. Speed	3.42 m/s	0.35 m/s	0.81	3.71 m/s	0.39 m/s	0.83	5.36 m/s	0.44 m/s	0.84

- $R^2 > 0.95$ good agreement
- $RMSE < 0.10m$ - max depth at 4m and 2m
- Only water depth at 30 Min. for 8m test case and maximum speed non satisfactory

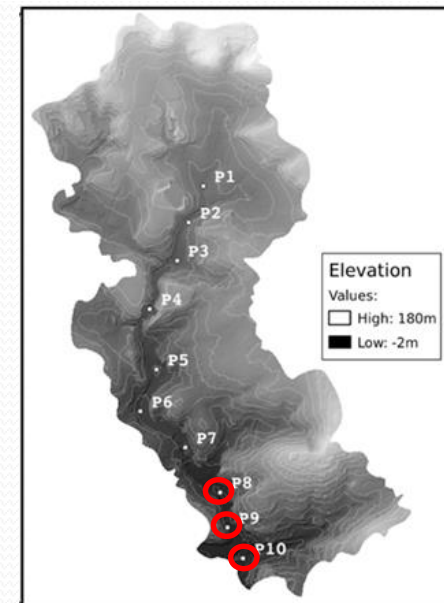
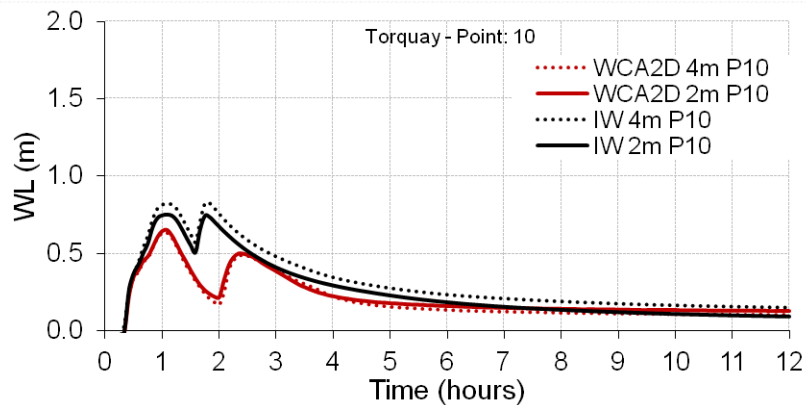
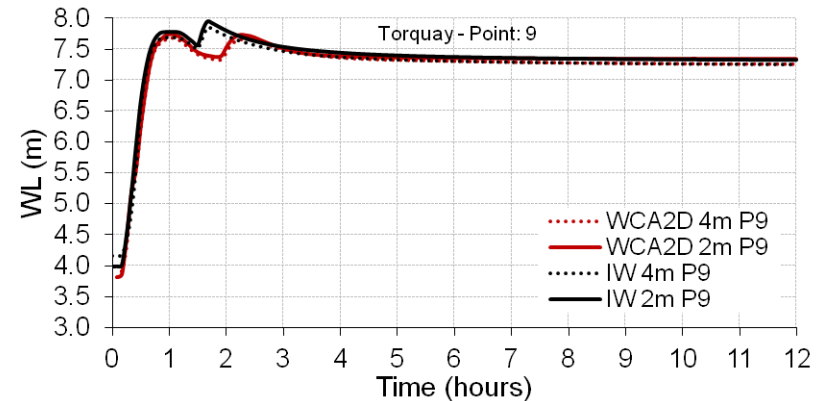
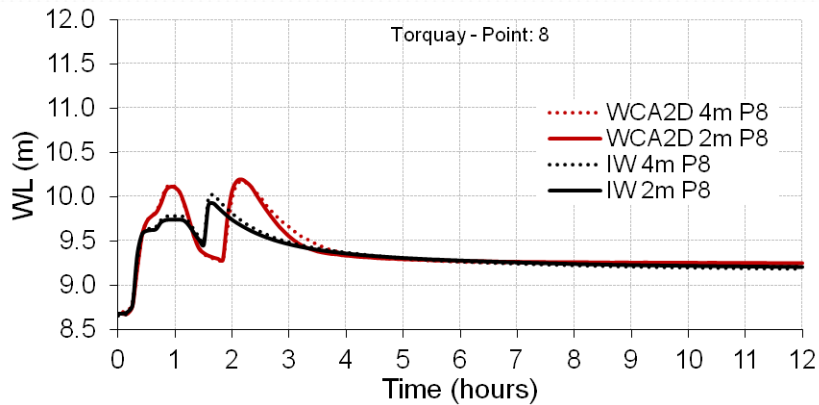
Depth and Velocity at 30 min



Results: Upstream Points



Results: Downstream Points



Computation Time

	WCA2D 8m		IW 8m		WCA2D 4m		IW 4M		WCA2D 2m		IW 2m	
Memory	~12MB		~230MB		~45MB		~900MB		~280MB		~3600MB	
Type	MC	GPU	MC	GPU	MC	GPU	MC	GPU	MC	GPU	MC	GPU
Time (Min)	1.63	0.21	8.82	1.83	16.63	1.99	64.55	9.91	252.93	27.90	600.47	77.28
Sp vs IW	5.41	8.71	---	---	3.80	4.98	---	---	2.37	2.77		

- Run on multi-core CPU and GPU
- From 3x to over 8x faster than InfoWorks
- WCA2D use around 10 times less memory

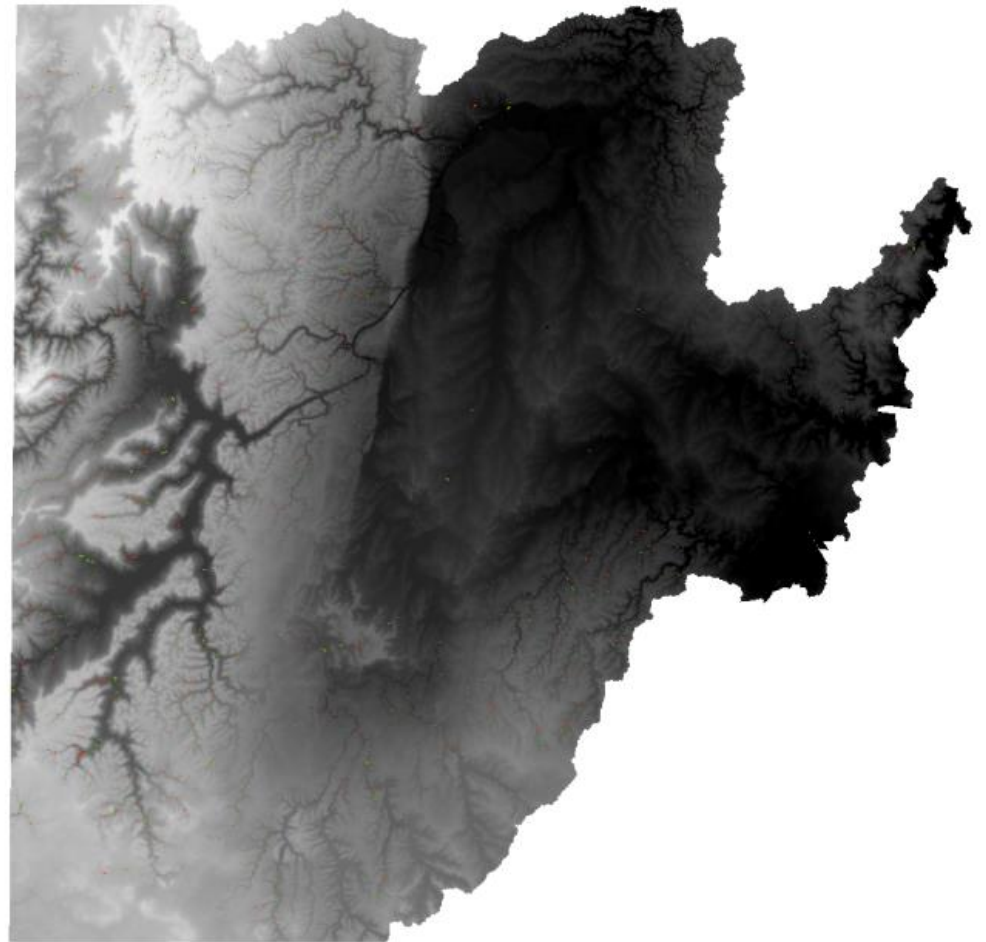
Results

Very Large Test case

Munich RE Data

Sydney, Australia

- 3700 x 3700 cells
- ~14 million cells
- 30m x 30m



Hardware



Name	Machine 1	Machine 2
Processor(s)	Intel i5-2500K 3.60GHz	2 x Intel Westmere 2.80GHz
Number of Cores	4	2 x 6
Memory	4 GB	24GB
Graphics Card	GeForce 550TI 192 Cores 1GB	N/A
Operating System	Windows 7, 64 bits	Linux, 64 bit
Note		1 Node of the Supercomputer

Simulation Information

- WCA2D (Machine 1)
 - 4 cores CPU version
 - GPU version 24 hrs simulation
- UIM (Machine 2)
 - 12 cores CPU version
 - 24 hours simulation

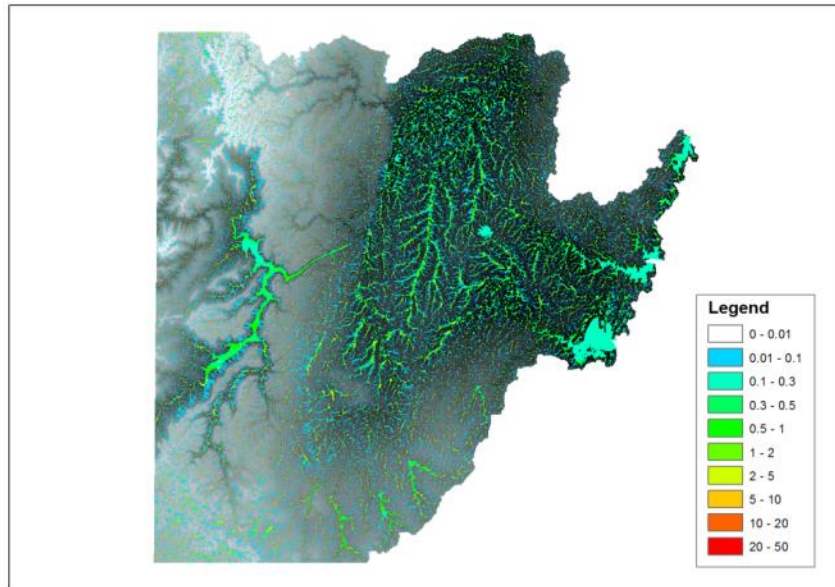
Results Case A and B

Name	WCA2D MC A	WCA2D GPU A	UIM A
Run Time Min.	Tot. (Model) 21 (14)	Tot. (Model) 11 (4)	Tot. 2091
RMSE > 0.01m	0.51m		---
Correlation	0.926		

Name	WCA2D MC B	WCA2D GPU B	UIM B
Run Time	Tot. (Model) 22 (15)	Tot. (Model) 11 (4)	Tot. 1972
RMSE > 0.01m	0.61m		---
Correlation	0.922		

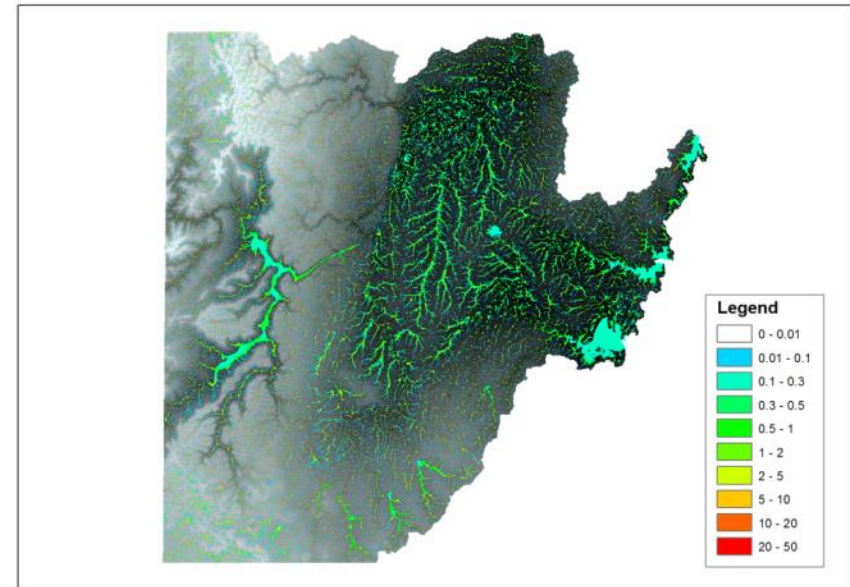
Results Case A

UIM / Case A / Peak Values



(a)

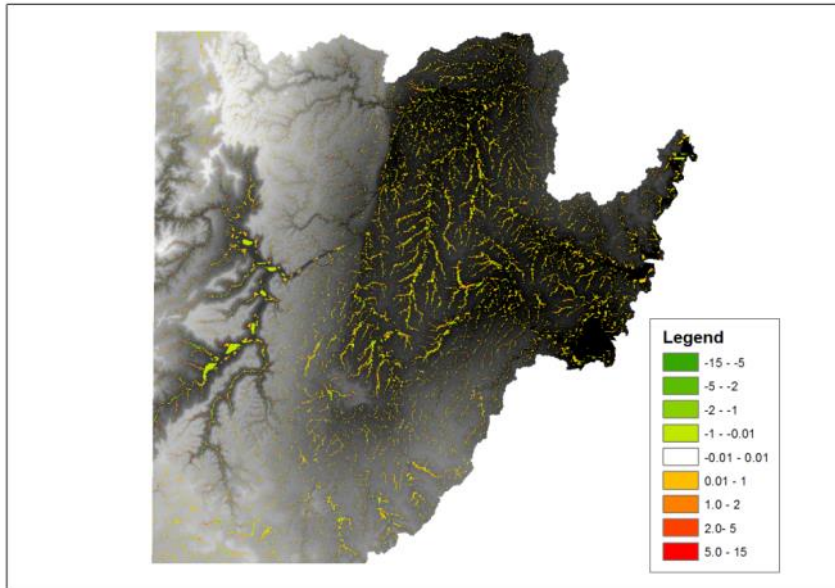
ICA2D / Case A / Peak Values



(b)

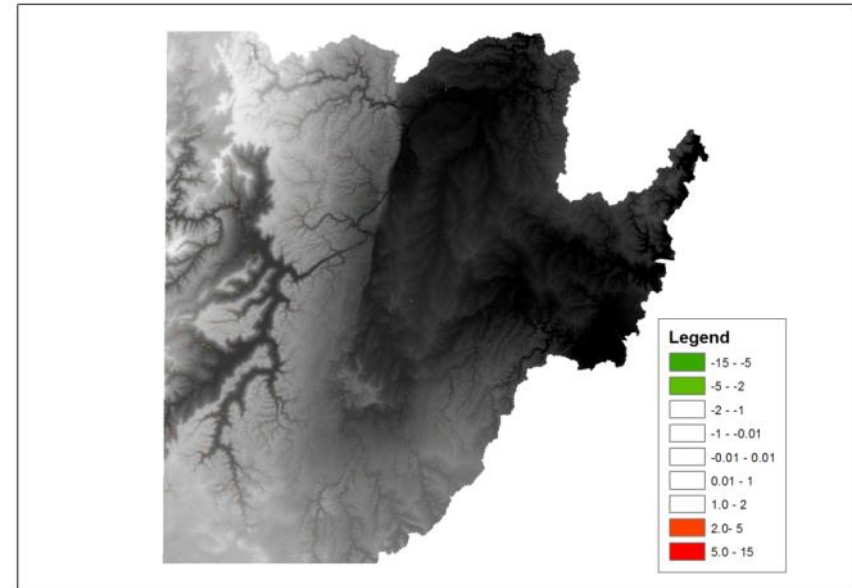
Results Case A Difference

Difference / ICA2D- UIM / Case A / Full Range



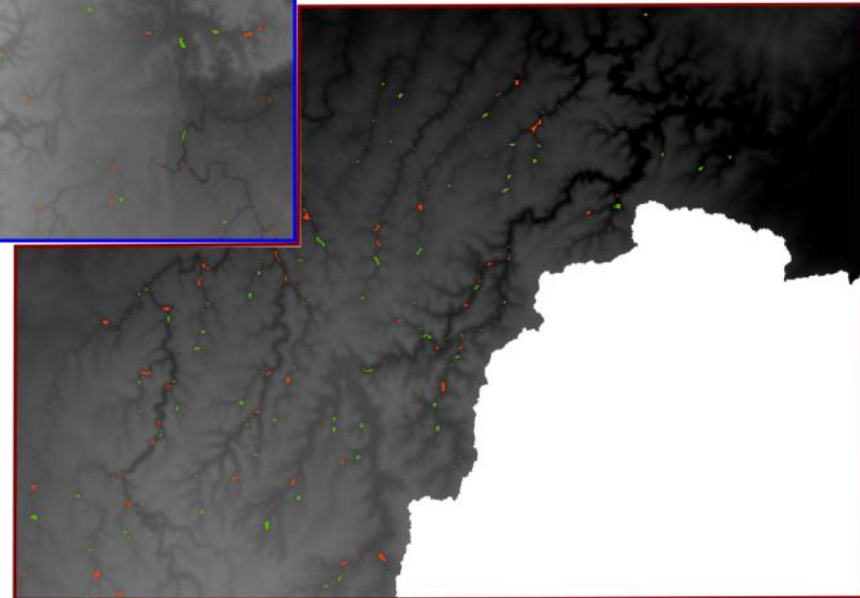
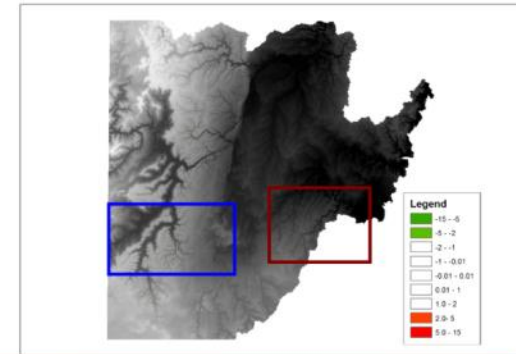
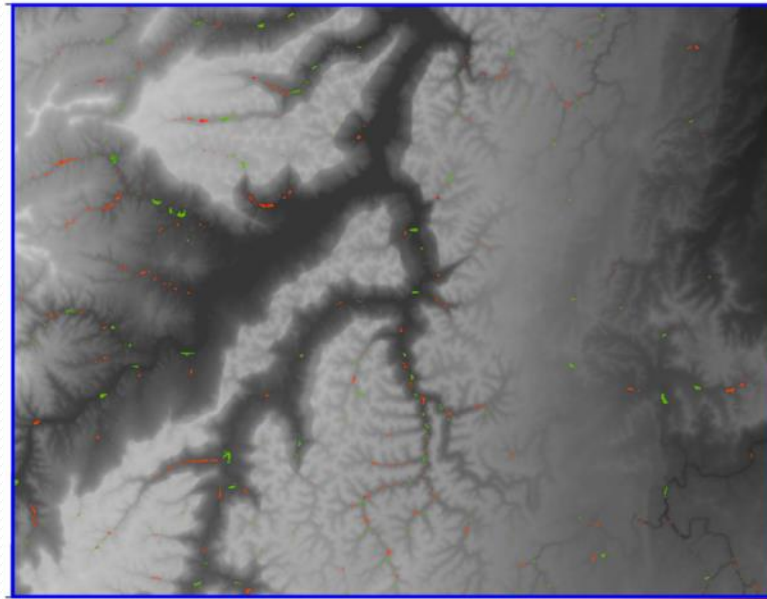
(c)

Difference / ICA2D- UIM / Case A / Over 2M



(d)

Results Case A Zoom



Extra Simulations

- Planned extra comparisons (on Machine 1)
 - InfoWorks ICM 3.0
- ICM was not able to complete a simulation due to memory limitation on the machine
 - When set with 6M triangles
 - Compared to around 10M of WCA2D
- Worked with 2.5M triangles
 - Run time was around 6 hours (4 cores)

1D vs. 2D risk of flooding from pipes

ICS Consultants Ltd.

Ageing assets... Thames Water £4m bill for Herne Hill flood after burst water main - August 2013



Herne Hill was swamped when an 88-year-old main burst



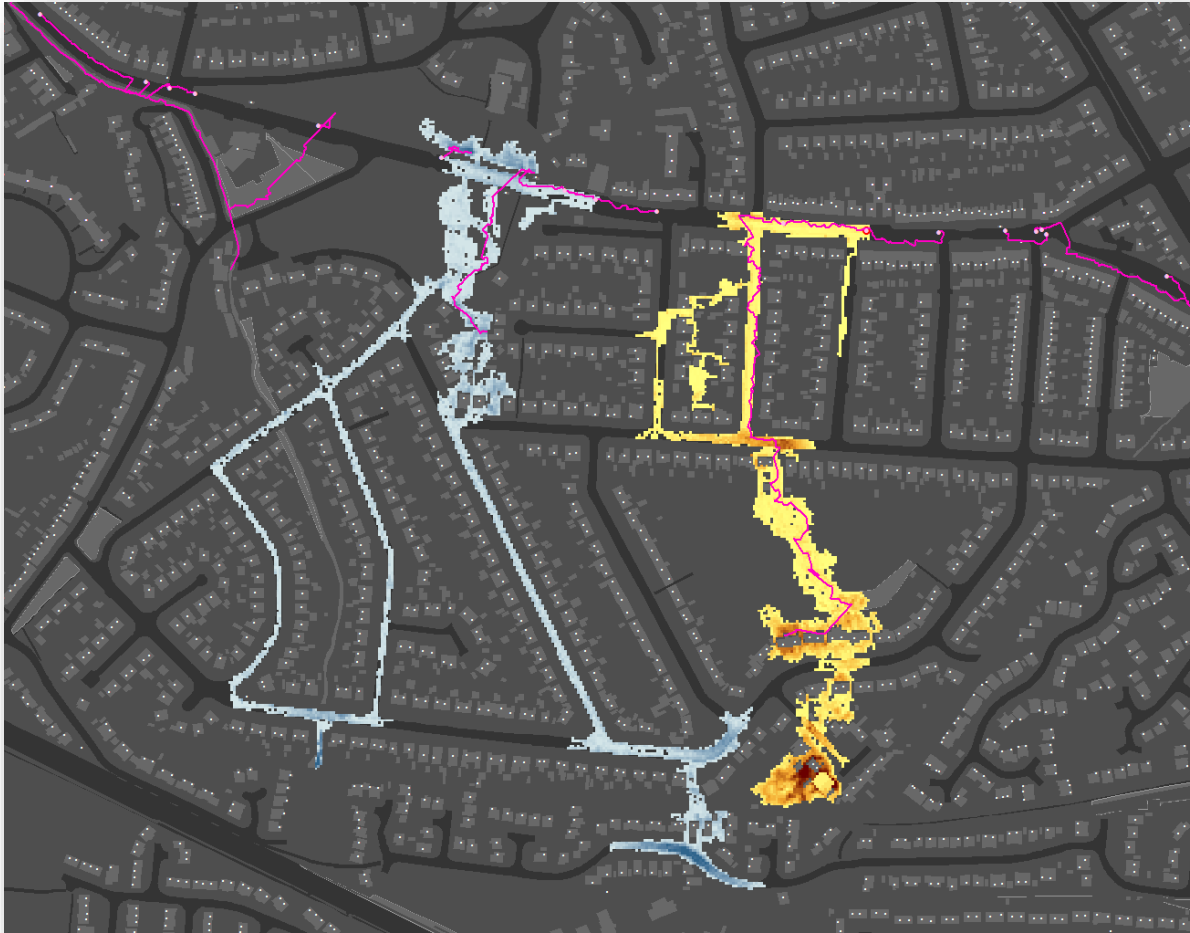
Consequences - 1D Flood Routing



Consequences - 1D versus 2D Flood Risk



Consequences - 1D versus 2D Flood Risk



Summary

- Developed a fast 2D flood model that uses a CA technique
- Rules (not SWE) developed manually, based on problem physics
- Produced results comparable to UIM and InfoWorks
- Runs quicker than UIM and InfoWorks in the examples showed

Future Work

- A new advance version of the model
 - Spatial roughness, infiltration, rainfall
 - Batch execution
 - Further ~30% faster and more accurate
- Automatic generation of transition rules
 - Genetic Programming

Thank You Questions?

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