CAHMDA-DAFOH Joint Workshop, 8-12 Sep. 2014, Austin, Texas, USA Session 7: Advancing Data Assimilation Science for Operational Hydrology (11 Sep. 2014)



Particle Filtering for Hydraulic Models: Probabilistic Urban Inundation Modeling and Assimilation-based H-Q Relationships

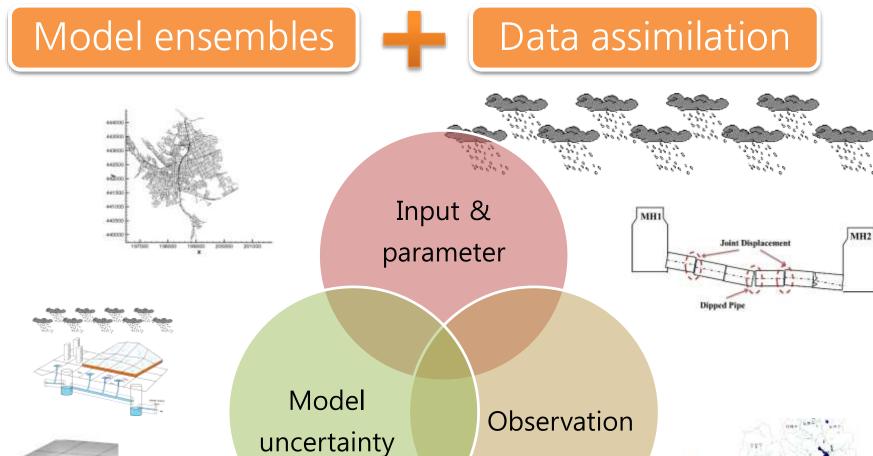
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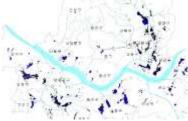


Probabilistic Inundation Modeling using Urban Flood Model and PF



Introduction – Probabilistic urban flood model





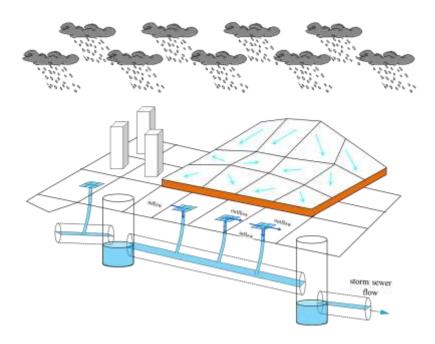


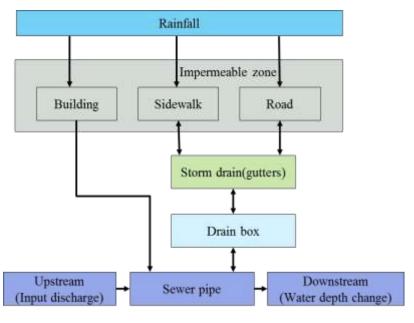
Methodology – Integrated urban flood model

Integrated urban flood model (DPRI, Kyoto Univ.)

- 2-D inundation model on the ground surface
- 1-D network model of sewer pipes
- Combined by a sub-model to exchange storm water between

the ground surface and the sewerage system







Methodology – Integrated urban flood model

Governing equations Ground surface

 $\frac{\partial h}{\partial t} + \frac{\partial M}{\partial t} + \frac{\partial N}{\partial t} = r_e - q_{drain} - q_{sew}$

drain box

$$\frac{\partial A}{\partial t} + \frac{\partial Q}{\partial x} = q_{drain}$$

$$\frac{\partial M}{\partial t} + \frac{\partial (uM)}{\partial x} + \frac{\partial (vM)}{\partial y} = -gh\frac{\partial H}{\partial x} - \frac{gn^2 M\sqrt{u^2 + v^2}}{h^{4/3}}$$
$$\frac{\partial N}{\partial t} + \frac{\partial (uN)}{\partial x} + \frac{\partial (vN)}{\partial y} = -gh\frac{\partial H}{\partial y} - \frac{gn^2 N\sqrt{u^2 + v^2}}{h^{4/3}}$$

* FDM & leap-frog methods are adopted

- *h* : water depth
- H : water elevation
- *u* : x-direction water velocity
- v: y-direction water velocity
- *M* : uh(x-direction flux)
- N : vh(y-direction flux)
- r_e : effective rainfall

- q_{drain} : unit area drainage discharge between ground and drain box
- q_{sew} : unit area drainage discharge between ground and sewer pipe
- *g* : acceleration of gravity
- *n* : Manning's roughness coefficient
- *A* : box bottom area
- *Q* : discharge
- q_{drain} : unit area drainage discharge between ground and drain box



Governing equations (sewer pipe)

 $\frac{\partial A}{\partial t} + \frac{\partial Q}{\partial x} = q_{sew} + q_{drain}$ $\frac{\partial Q}{\partial t} + \frac{\partial (uQ)}{\partial x} = -gA \frac{\partial H_p}{\partial r} - gn^2 \frac{|Q|Q}{R^{4/3}A}$ $h = \begin{cases} f(A) & : A \le A_0 \\ D + \frac{(A - A_0)}{b_s} & : A > A_0 \end{cases}$ $\phi = 2\cos^{-1}(1-2\frac{h}{d})$ $\frac{A}{A_0} = \frac{\phi - \sin \phi}{2\pi}$ $\frac{R}{R_0} = 1 - \frac{\sin \phi}{\phi}$ $B_s = \frac{gA_0}{a^2}$

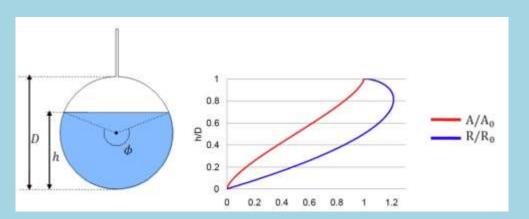


Fig. Hydraulic characteristic curves of the circular sewer pipe

The a is decided as 5.0 m/s in this study

R : hydraulic radius

 q_{sew} : unit area drainage discharge between ground and sewer pipe

 q_{drain} : unit area drainage discharge between sew pipe and drain box

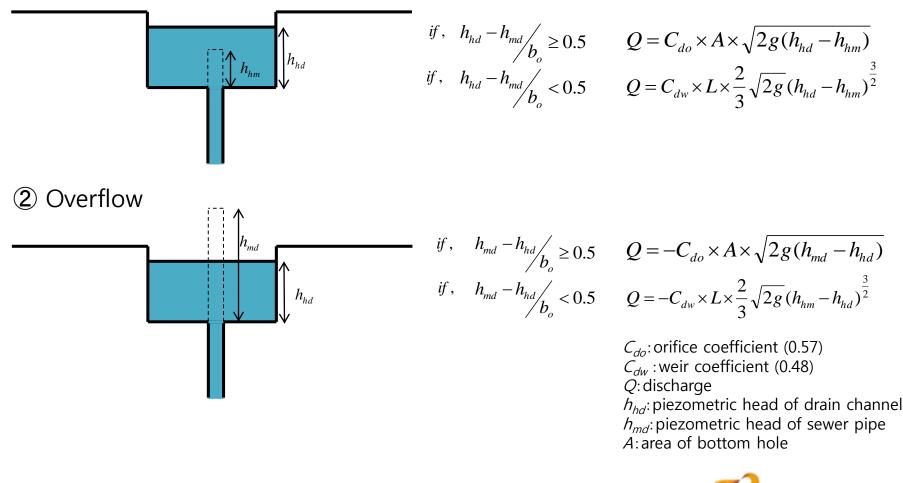
- g : acceleration of gravity
- n : Manning's roughness coefficient
- A : stream cross-section
- Q: discharge
- n : water velocity
- H_p : piezometric head
- h : water depth
- H_z : head losses



Methodology – Integrated urban flood model

Interaction model (surface water - sewer pipe)

1 Inlet



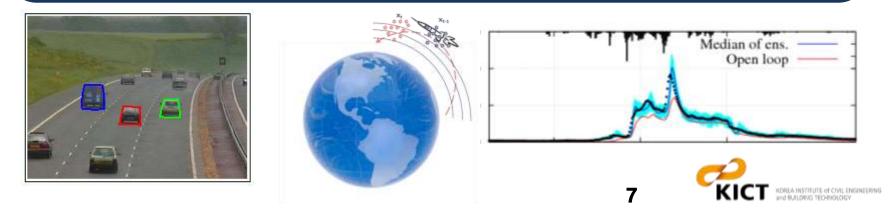
Methodology – Particle filtering

Particle filtering

- Also known as sequential Monte Carlo (SMC)
- Applicable for non-linear, non-Gaussian state-space models (most of hydraulic and hydrologic models)
- Point mass ("particle") representations of probability densities with associated weights
- Expensive computation but easy for parallelization
- Wide-spread applications including image processing, target tracking, and flood

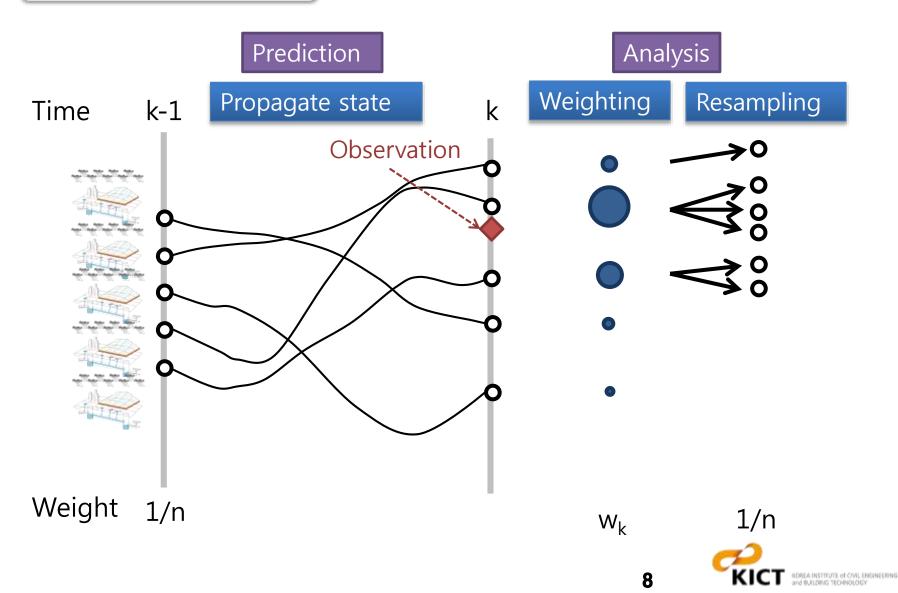
forecasting

- Noh, S. J., Rakovec, O., Weerts, A. H., and Tachikawa, Y.: On noise specification in data assimilation schemes for improved flood forecasting using distributed hydrological models. J. Hydrol. in press, 2014.

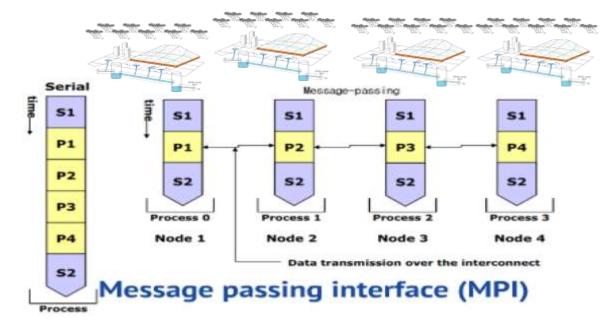


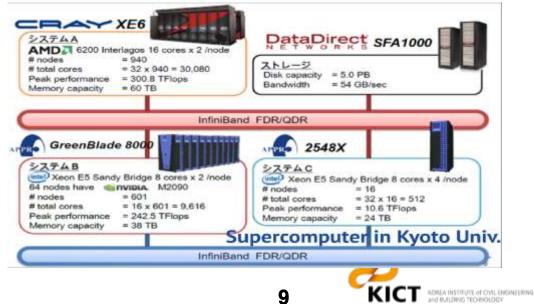
Methodology - Particle filtering

SIR particle filter



※ Methodology - Parallel computing





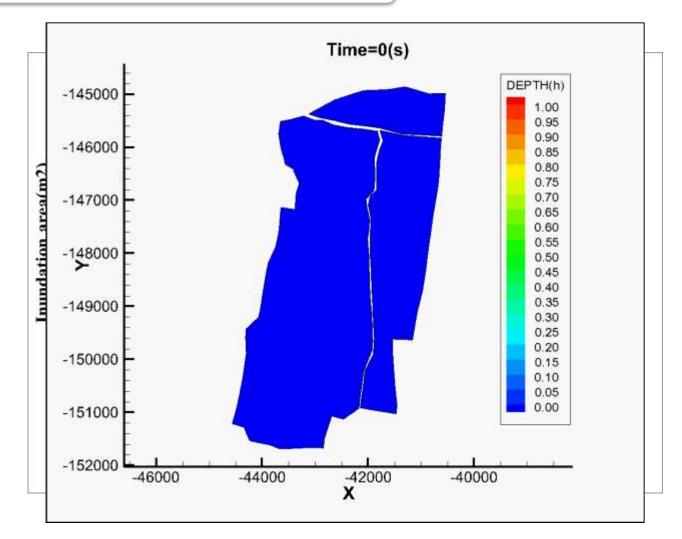
Application – Study area

Urban surface Sewer pipe network <PUMPING STATION: NAKAHAMAHIGASHI : NAKAHAMANISHI 0 BENTEN : NEKOMAGAWA Node : 60,411 Mesh : 117,435 Link: 177,843



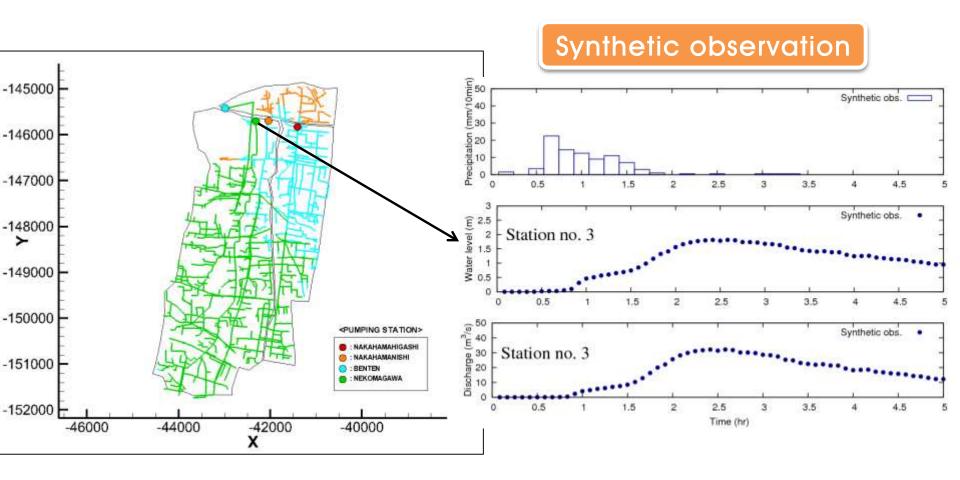
※ Application – Deterministic simulation result

Deterministic modeling case





Solution - Setup of synthetic experiment



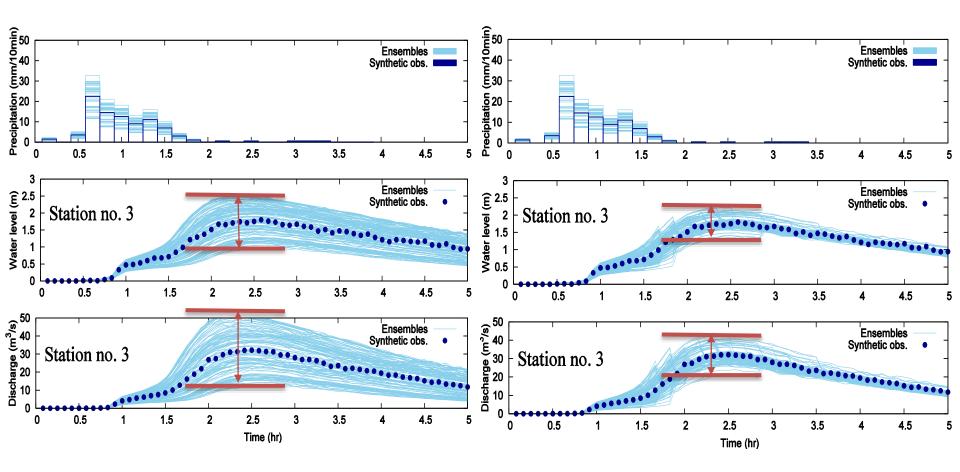
- Water level and discharge at pumping station no. 3 were used as synthetic observation
- Spatial distribution of rainfall was not considered



※ Application – Comparison on H & Q

Without particle filtering

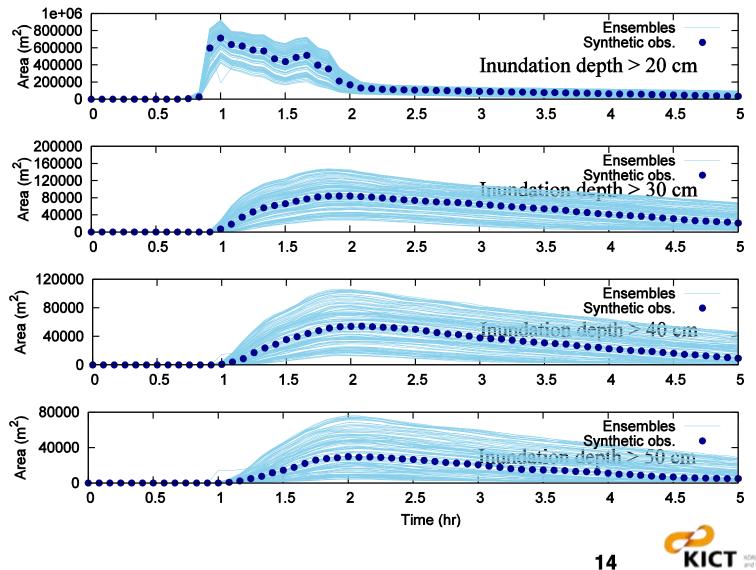
With particle filtering





Application – Comparison on inundation area

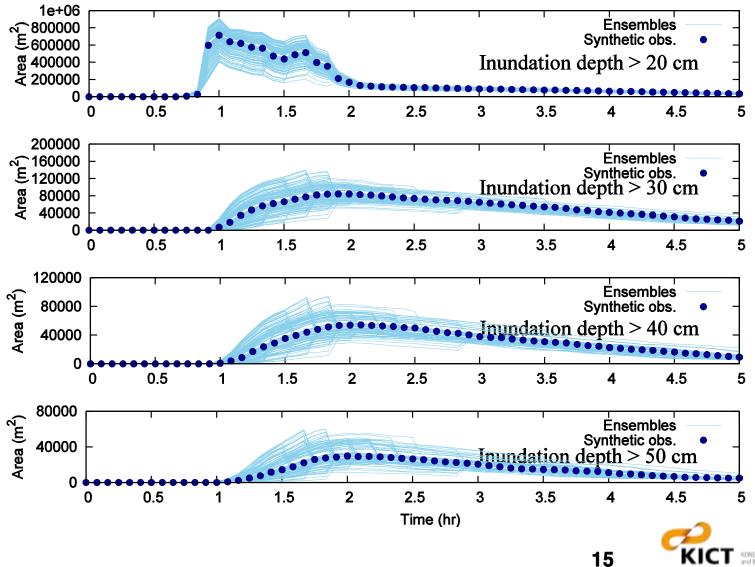
Inundation area – without particle filtering



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Application – Comparison on inundation area

Inundation area - with particle filtering

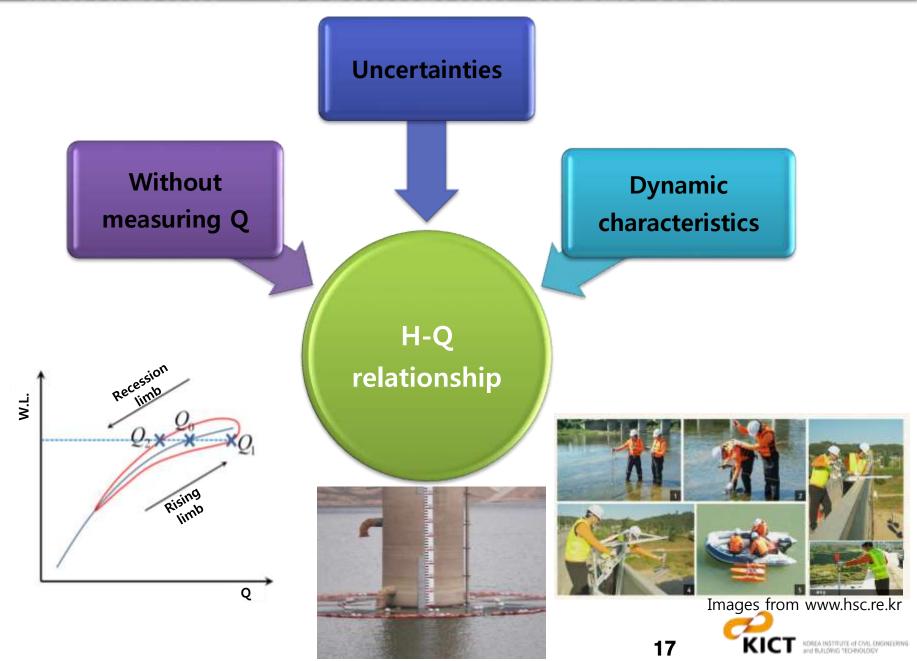


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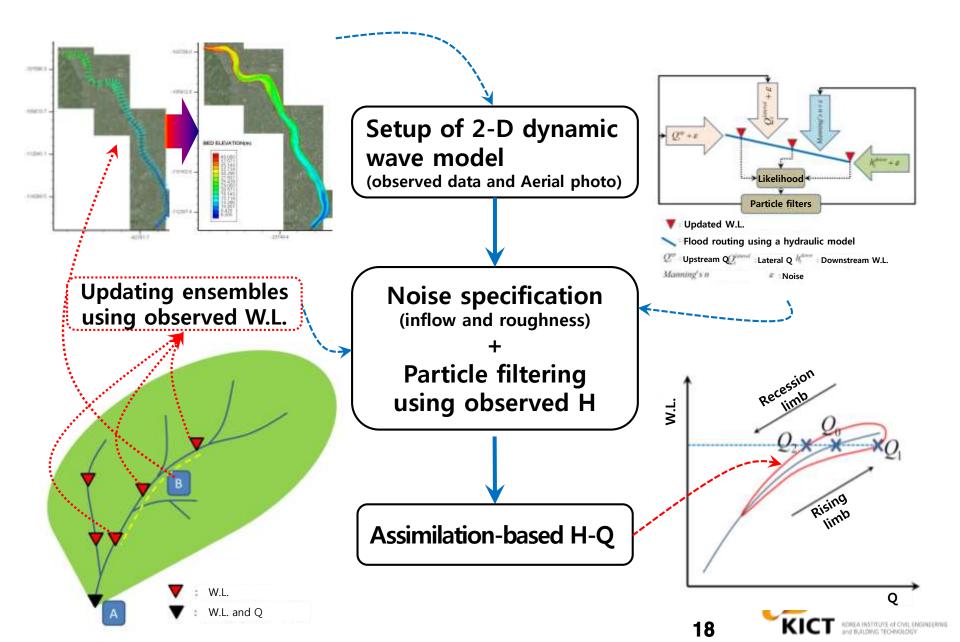
Assimilation-based H-Q Relationships using 2-D Dynamic Wave Model and PF



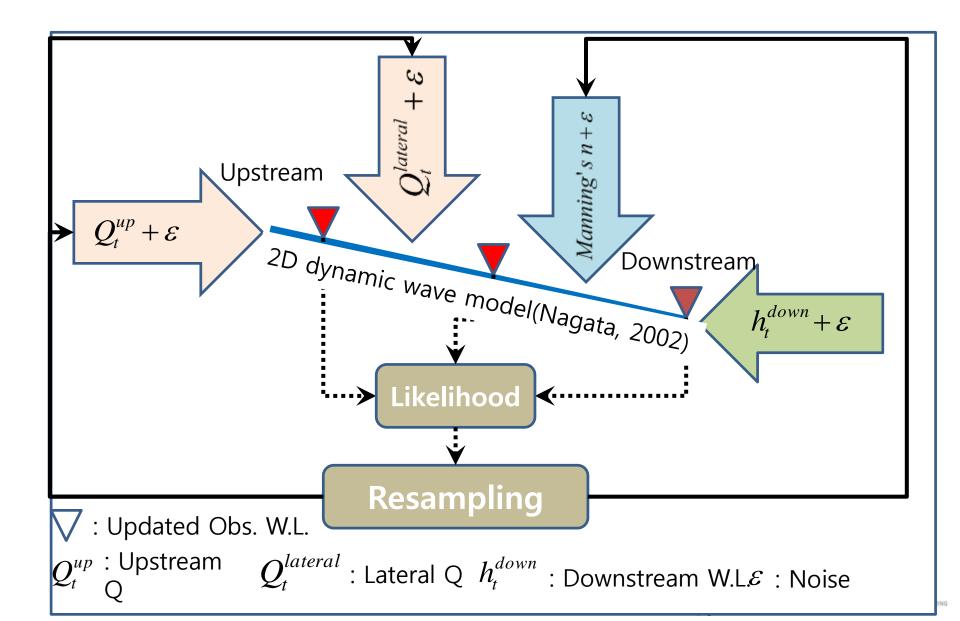
Motivation - Assimilation-based H-Q

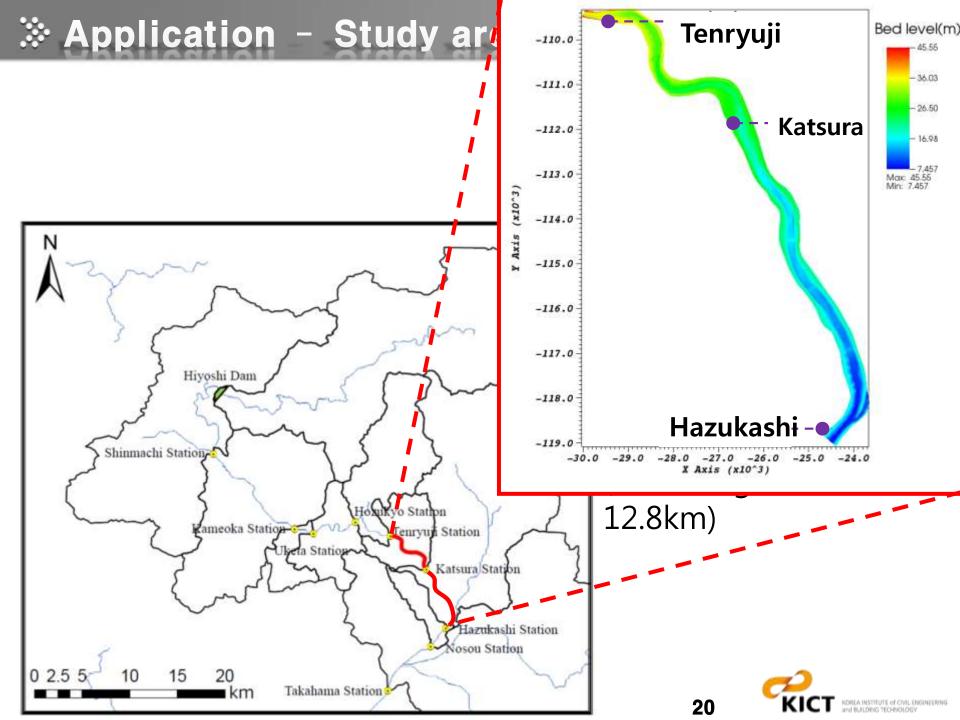


Methodology - DA procedures



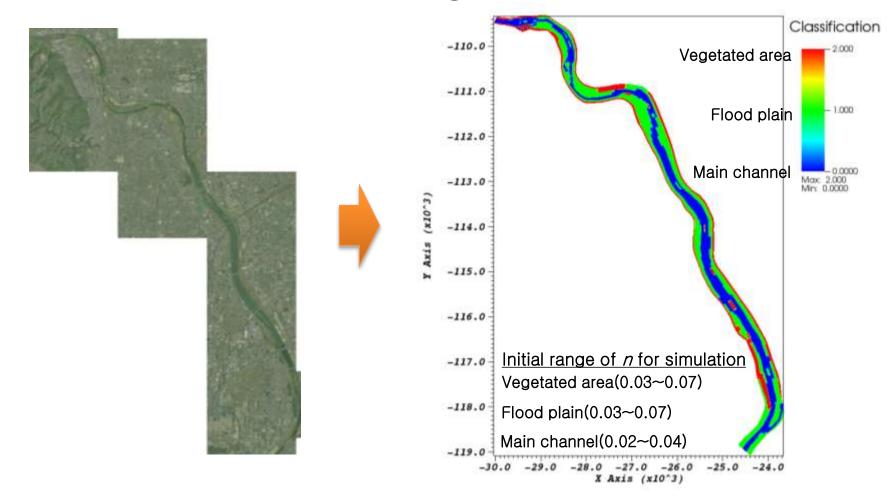
Methodology – Noise specification & PF





Application – Setup of 2–D dynamic wave model

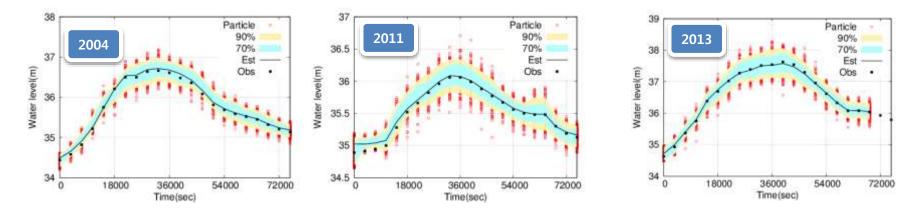
• Classification of Manning's *n*



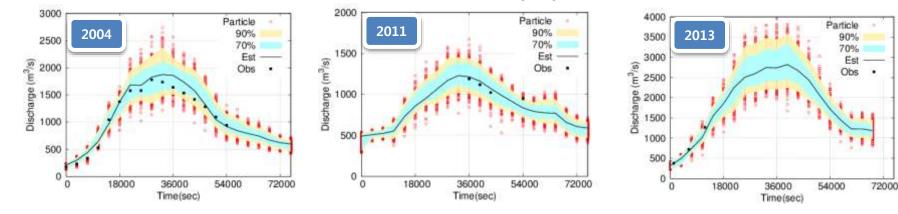


Solution - Estimation of H & Q

• Estimated and observed H at Tenryuji station

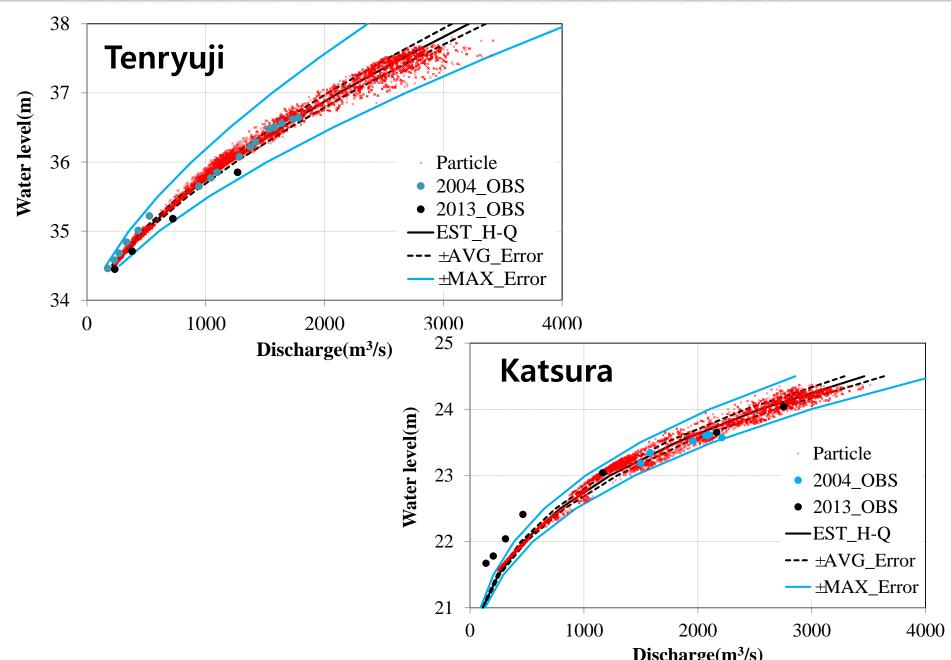


• Estimated and observed Q at Tenryuji station





Application - Assimilation-based H-Q



- Applicability of particle filtering for two hydraulic models were evaluated.
- PF is sound when momentum equilibrium is important in the prediction models
- Assimilation-based estimation using 2-D model and PF could provide reliable H-Q relationships for poorly-gauged or ungauged basins



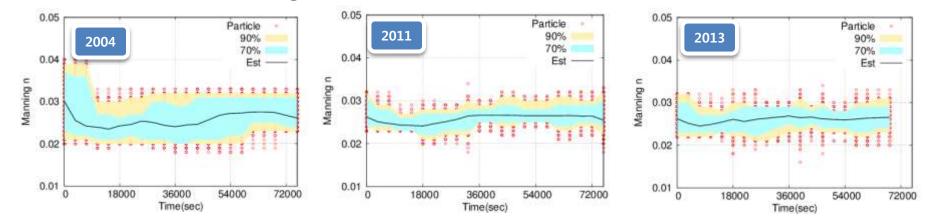
- PF and dimensionality
 - Different ensembles are required for estimation of states and parameters
 - Hybrid DA to reduce dimensionality
 - PF-MLEF, PF-EnKF, ...
- Real-time applications
 - Parallel computing for ensembles
 - Parallel computing within a ensemble
- More attention is needed to improve DA to consider both covariance and dynamics of the system



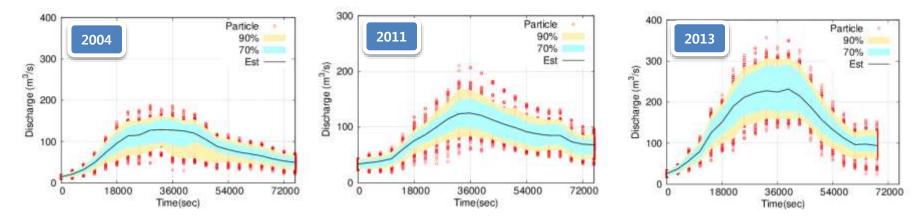
Thank you for your attention! seongjin.noh@gmail.com

※ Application – Estimation of n & lateral flow

• Estimated Manning's n at main channel



• Estimated lateral inflow from Tenryuji to Katsura station

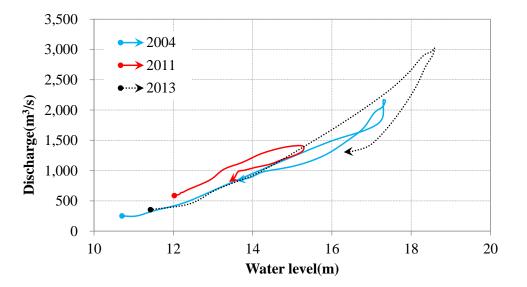


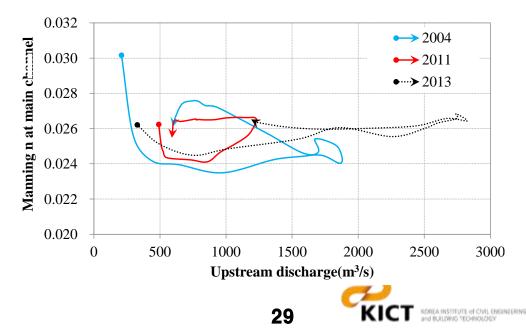


Potential observation for data assimilation









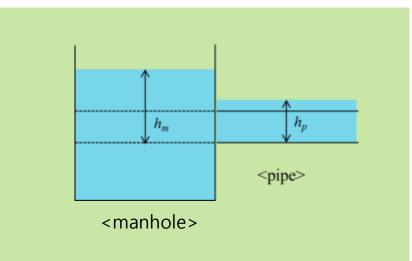
Methodology – Integrated urban flood model

Connection models

• Manhole and sewer pipe

• In the case of
$$(h_m \ge h_p)$$

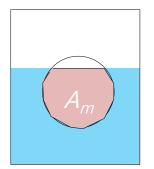
 $Q = 0.35 \times A_m \times \sqrt{2gh_m}$ If, $(h_p / h_m \ge 2/3)$
 $Q = 0.91 \times A_p \times \sqrt{2g(h_m - h_p)}$ If, $(h_p / h_m <= 2/3)$



•	In the case of $(h_p > = h_m)$	
	$Q = -0.35 \times A_p \times \sqrt{2gh_p}$	If, $(h_m / h_p > 2/3)$
	$Q = -0.91 \times A_m \times \sqrt{2g(h_p - h_m)}$	If, $(h_m / h_p <= 2/3)$

 z_m : elevation of manhole z_p : elevation of pipe h_m : water depth of manhole z_p : critical depth of pipe v_m : velocity of manhole v_p : critical velocity of pipeg: gravity acceleration h_p : critical depth of pipe + velocity
head

30



 A_m : calculated by $h_m A_p$: calculated by h_{cp}

Fig. Concept of virtual area of manhole



Introduction

- Urban inundation due to heavy rainfall and climate change is an inevitable problem for many cities around the world and constitutes a severe threat to residential life, property and infrastructure(Mark et al., 2004)
- Therefore, it is important to accurately simulate urban hydrological processes and efficiently predict the potential risks of urban floods (Lee et al., 2009)
- However, it is insufficient to obtain accurate predictions due to various uncertainties coming from input forcing data, model parameters, and observations.









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