

Al-powered Real-time Data-driven Framework: Tool for Flood Early Warning in Small Stream Tae Sung Cheong¹[†], Sangman Jeong²

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Introduction

What is the Small Stream? Feed into larger water bodies like the Local and the National rivers or lakes Fast flow velocity due to steep slope (mean of 0.06)



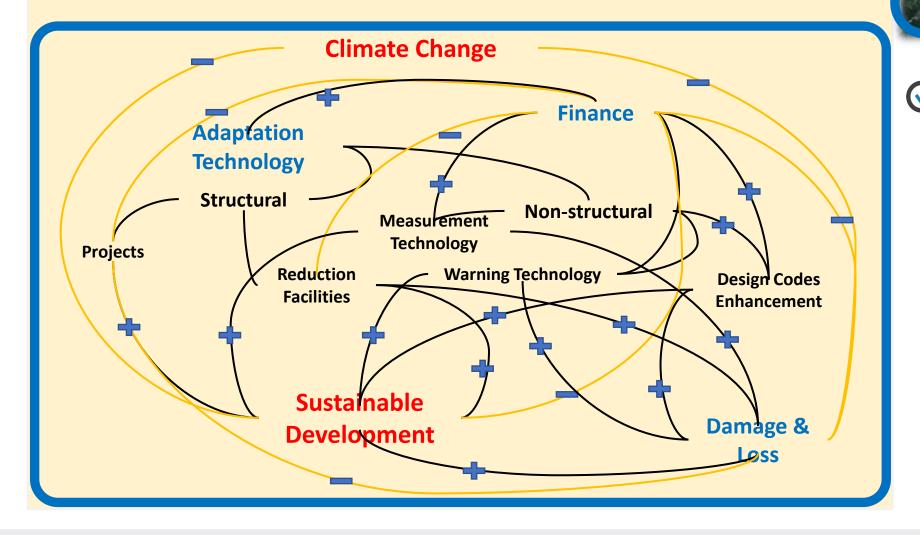
Development & Evaluation

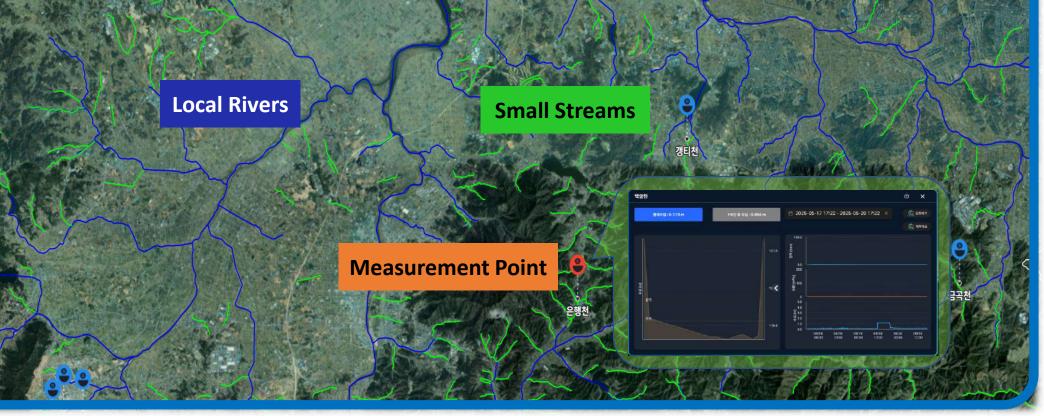
• Develop the nomograph and rating curve using measured data collected from nine small streams

80 a) Daemi	30 b) Bekam	15 c) Jumsil	a) Daemi	b) Bekam	c) Jumsil
°= / °	20 22 22 22 22	(j) 10	1	5 🖉	Ê 0.5
20 Dischar	10	Dischar		0 /	ten /
					۰Ľ
0 10 20 30 40 50 Rainfall (mm/h)	0 20 40 60 Rainfall (mm/h)	0 10 20 30 40 Rainfall (mm/h)	0 20 40 60 80 Discharge (m ³ /s)	0 10 20 30 Discharge (m ³ /s)	0 5 Disc
80 d) Gwangdong	6 e) Balmac	15 f) Songnam	1.5 d) Gwangdong	0.6 e) Balmac	1 f) Songna

C Real-time measuring	& monitoring in the Platform
Water depth, flow	수문자료 조회 사·조 89955 · 사·전·구 2844 · 소의법 (2015년:201 시작 144 (금 2024-09-19 ×) 88 144 (금 2024-09-23 ×) (요 요프 이 다운요드
velocity and discharge	두고난 세려년 60년 8년년
	원호 원수권 전계 포스간전역 유전 원고운속 부대용속 유왕 전유수관 원호 원수권식 (김) (#*) (m) (m44) (#*/4) (#*/4)

- Be called a brook, creek or rivulet depending on region
- **Why Small Stream Management is Needed?** Flood damages is increasing with localized rainfalls due to climate change
- River floods arise from the interaction of streams with widely differing temporal scales and can increase if small streams are not properly managed





What is the Real-time Data-driven Framework? • A web tool following a user-centered design approach, consolidating hazards to help decision making and resilience planning Real-time rainfall, flow velocity, depth, discharge data and CCTV image monitoring

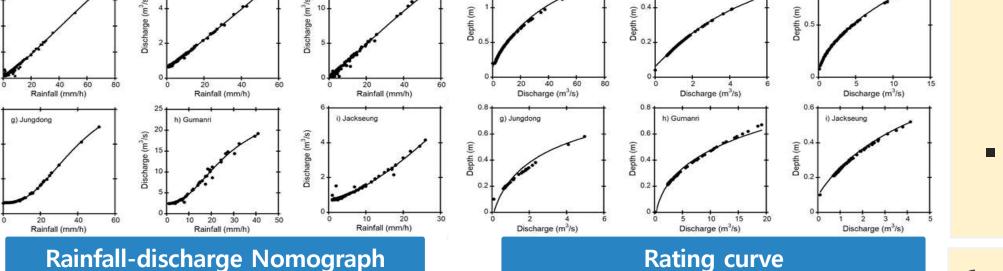
Data-driven rainfall-discharge nomograph to predict discharges • Data-driven depth-discharge rating curve to predict depths Real-time monitoring data based flood vulnerability prediction model

Materials

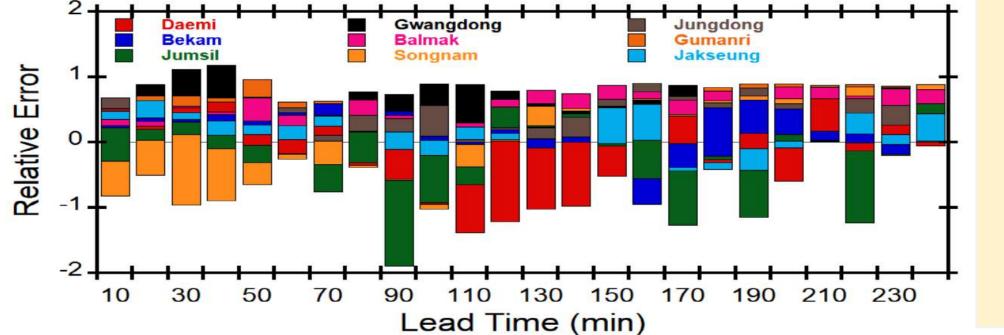
• MOIS establish the Small Stream Smart Measurement & Management System (SMMS) by 2027 in 2,200 small stream, Korea Select nine small streams for test-bed to develop and evaluate flood early warning framework







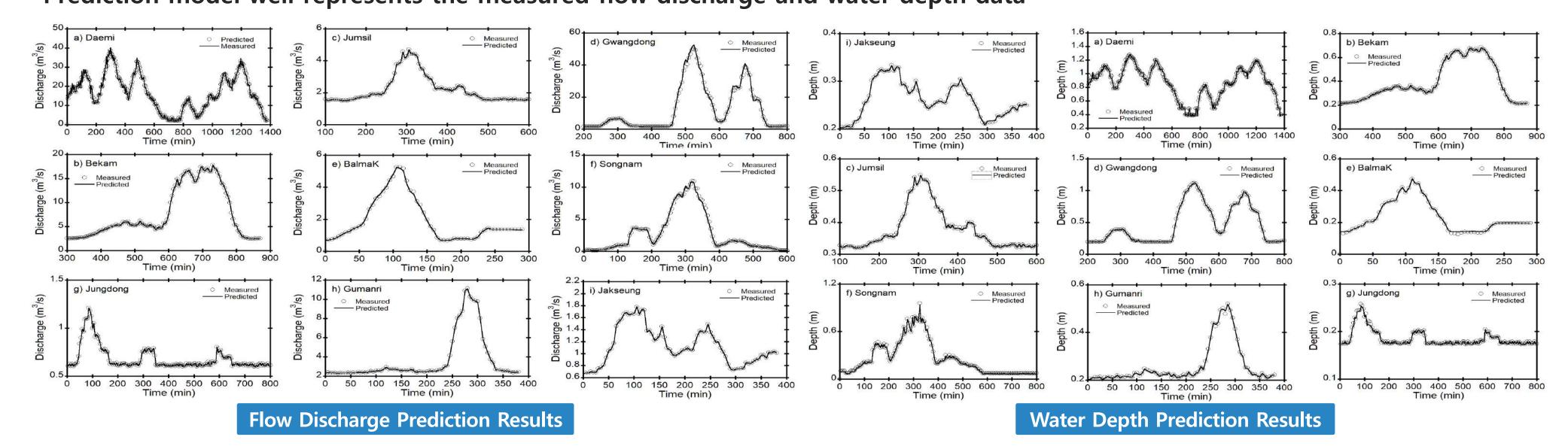
• One hour before predicted rainfall is used to predict depth

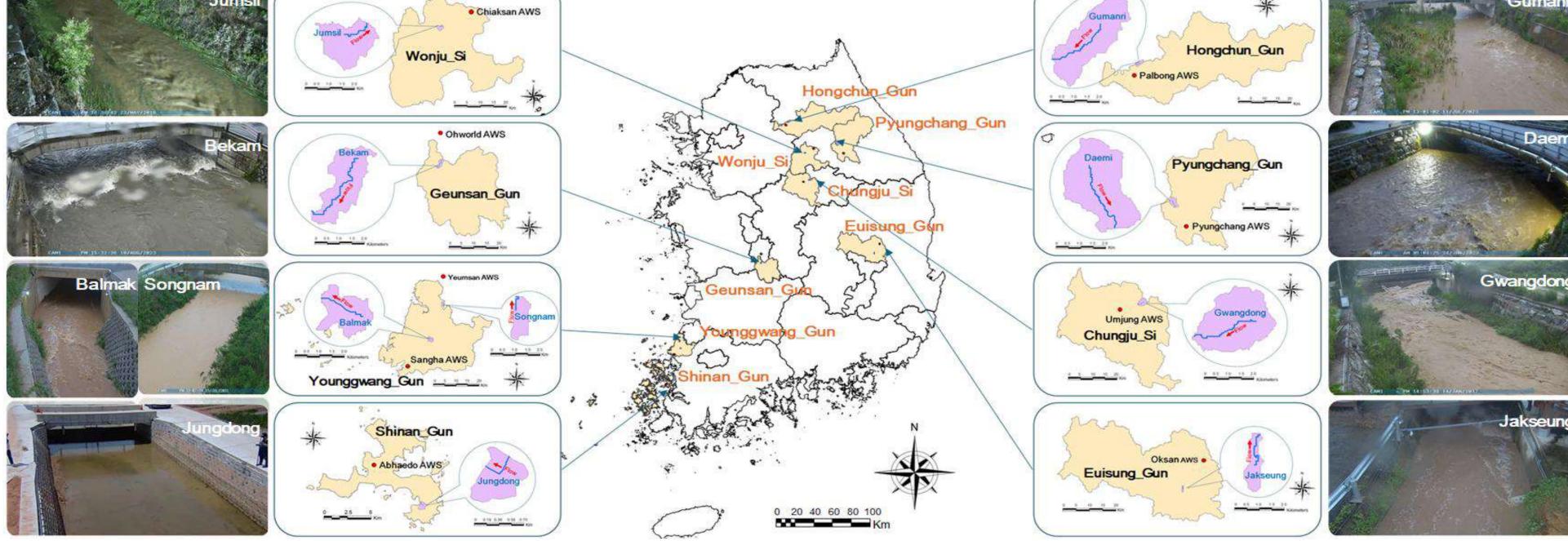




The Platform is capable of automatically generating nomograph and rating equation For development, three years of sufficient data is required

• Evaluate the prediction values of flow discharge and water depth with measured data collected from nine small stream Prediction model well represents the measured flow discharge and water depth data





<Plow images captured from CCTV and location of small stream and the AWS which was the cleanest one>

• Nine Small Stream (SS) has small width, length and steep slope located in river head

	SMMS		Basin Area	Channel Length	Planned	Planned	Width	Clana		AWS		
Small Stream	Latitude	Longitude	(km²)	(km)	Discharge (m ³ /s)	Elevation (El.m)	(m)	Slope	n	Name	Distance from SS(km)	
Daemi	37.46586036	128.3204602	12.8	4.48	226	529.9	22.4	0.033	0.033	Pyungchang	11.8	
Bekam	36.18913770	127.3887096	3.44	3.51	50	119.9	13.5	0.014	0.035	Ohworld	11.4	
Jumsil	37.39137582	127.9318512	2.59	1.29	57	105.1	12.6	0.019	0.030	Chiaksab	10.8	
Gwangdong	37.09191389	127.9675028	6.36	2.95	96	105.8	11.6	0.048	0.030	Eumjung	6.04	
Balmak	35.37031182	126.4892379	0.59	0.53	14	7.700	6.80	0.028	0.035	Sangha	8.00	
Songnam	35.27335980	126.4481607	1.61	1.49	45	5.800	18.5	0.008	0.030	Eumsan	10.1	
Jungdong	34.83371571	126.3464317	0.50	0.60	13	17.30	15.0	0.004	0.030	Abhaedo	6.81	
Gumanri	37.72040949	127.7123560	5.00	2.69	108	86.47	24.0	0.026	0.035	Palbong	3.94	
Jakseung	36.30444983	128.7472070	0.94	1.50	13	145.5	15.0	0.039	0.030	Oksan	11.6	

Comparison of RMSE and determinant coefficient

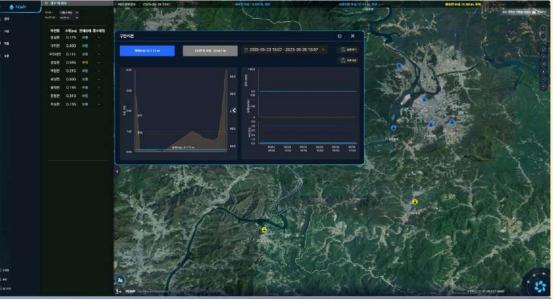
•					Over the warning criteria
Cmall Stream	Discharg	ge (m³/s)	Dep	th (m)	
Small Stream	RMSE	R ²	RMSE	R ²	Divided the warning
Daemi	0.521	0.996	0.006	0.998	criteria level into two
Bekam	0.009	0.992	0.001	0.994	
Jumsil	0.001	0.988	0.001	0.920	stages such as the
Gwangdong	0.433	0.997	0.033	0.997	caution warning criteria
Balmak	0.036	0.996	0.002	0.984	
Songnam	0.077	0.998	0.005	0.999	level (CWCL) and the
Jungdong	0.001	0.936	0.002	0.996	severe warning criteria
Gumanri	0.007	0.994	0.002	0.997	
Jakseung	0.003	0.974	0.001	0.999	level (SWCL)

stages such as the AWCL caution warning criteria level (CWCL) and the CWCL severe warning criteria level (SWCL) Time (h)

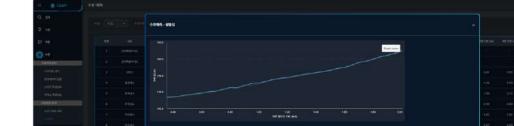
—— Measured Value

------ Extended Value

Solution Flood warning in the Platform



Sharing predicted depth of whole section of SS

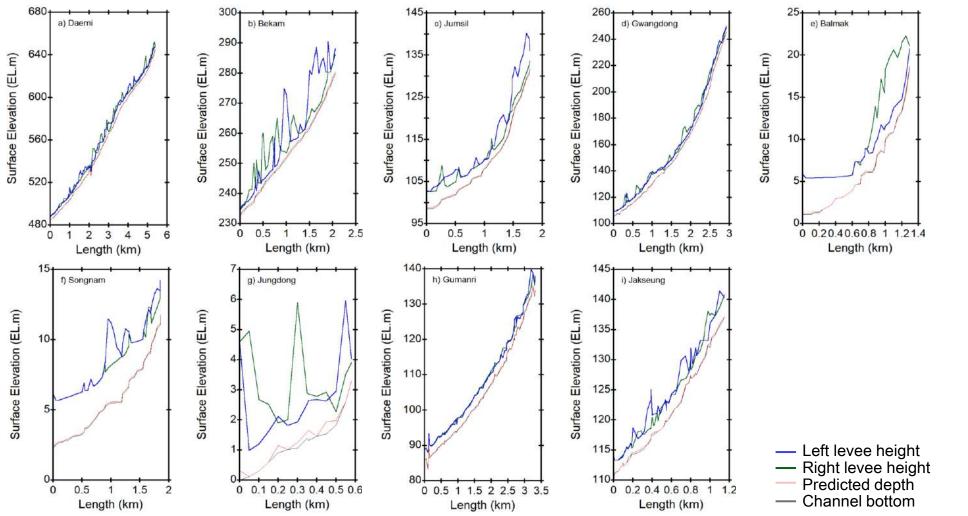


Find predicted and measured water depths with cross-section data Warning level (caution) reached at 50 cm, severe reached at planned depth

Depth prediction & warning for unmeasured sections at **10-minuates intervals**

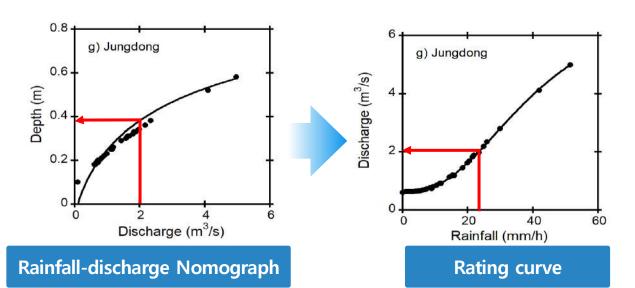
Predict depth on unmeasured section using the Manning Formula in nine small streams

 Determine optimum manning coefficient to minimize errors between predicted and measured depth data



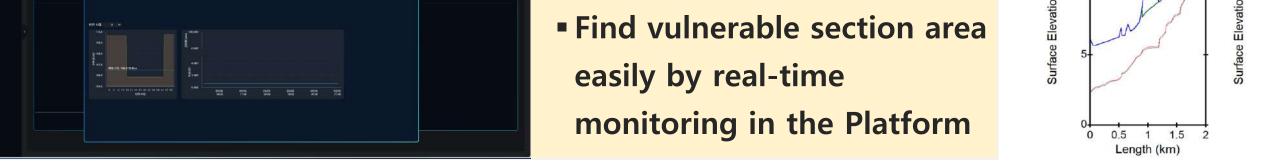
Methods

•Step 1: Develop rainfall-discharge nomograph and rating curve using measured data •Step 2: Predict discharge and depth using predicted rainfall and discharge by the MAPLE and Levenberg-Marquardt method •Step 3: Predict unmeasured section depth using Manning Formula under assumption as discharge is same in SS



Input	Processes					
ollection Measurement Rainfall, Depth, Discharge Data Set	on Measurement Rainfall, Depth, Discharge Data Set					
	Application of Discharge Prediction Process Using the Nomograph					
Prediction Rainfall Data using Motion Method	estimation weight Order Using Levenberg-Marguardt Method to Minimize the Sum of Squared Residuals					
Measurement Discharge Data 🔷	Resampling Predicted Discharges from Ordered Observation					
	+	•				
	Application of Depth Prediction Process Using the Rating Curve	Application of Depth Prediction Process Using the Manning Formula				
	Estimation Weight Order Using Levenberg-Marquardt Method to Minimize the Sum of Squared Residuals	Estimation Manning Coefficient using Non-linear Optimization to Minimize the Sum of Squared Residuals	→	Depth		
Measurement Discharge Data	Resampling Predicted Depth from Ordered Observation		→ III	Deptil		

< Concept diagram of predict discharge and depth of the AI-powered flood early warning framework >



Conclusions

- •Advancements in technology and real-time measurement, including imagery, have significantly improved the availability and accessibility of monitoring data.
- Building on these developments, our research developed a real-time, data-driven framework that integrates a user-centered design approach to co-create a web-based decision-support tool.
- The framework was applied to nine small streams in Korea to assess vulnerabilities and evaluate the effectiveness of early flood warning systems.
- A stakeholder workshop was conducted in Seoul to test and validate the framework, where participants provided positive feedback regarding its usability and potential impact.
- This study exemplifies the effective utilization of real-time monitoring data in developing flood early warning tools, ultimately contributing to reduced flood damage and enhanced resilience planning for small streams in Korea.

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