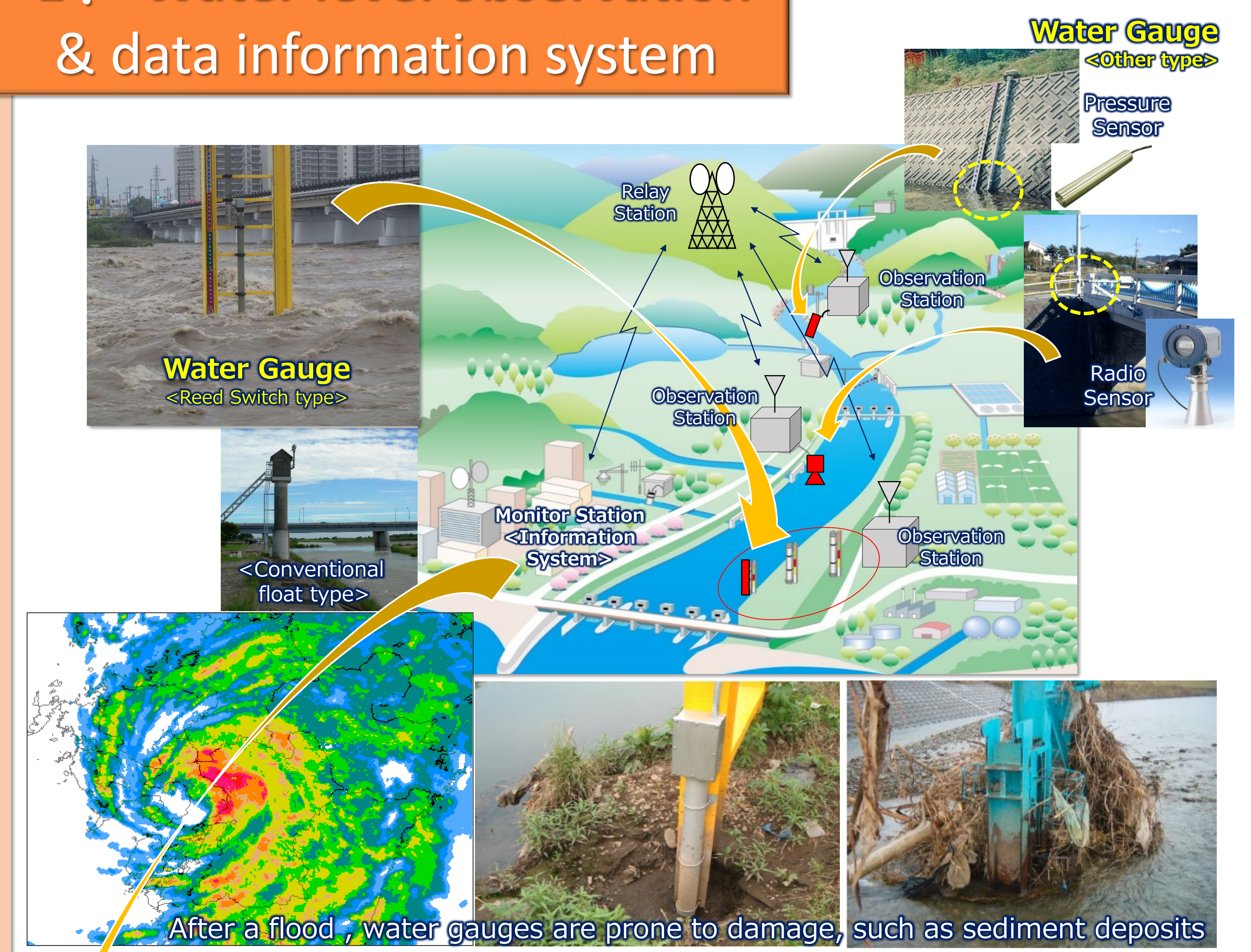


Development and actual use of Real-time anomaly detection system for river water level data

Tadashi TADOKORO / Osamu OKADA / Hisaya SAWANO / Arata SEI
Foundation of River and Basin Integrated Communications (FRICS) , JAPAN

I. Water level observation & data information system



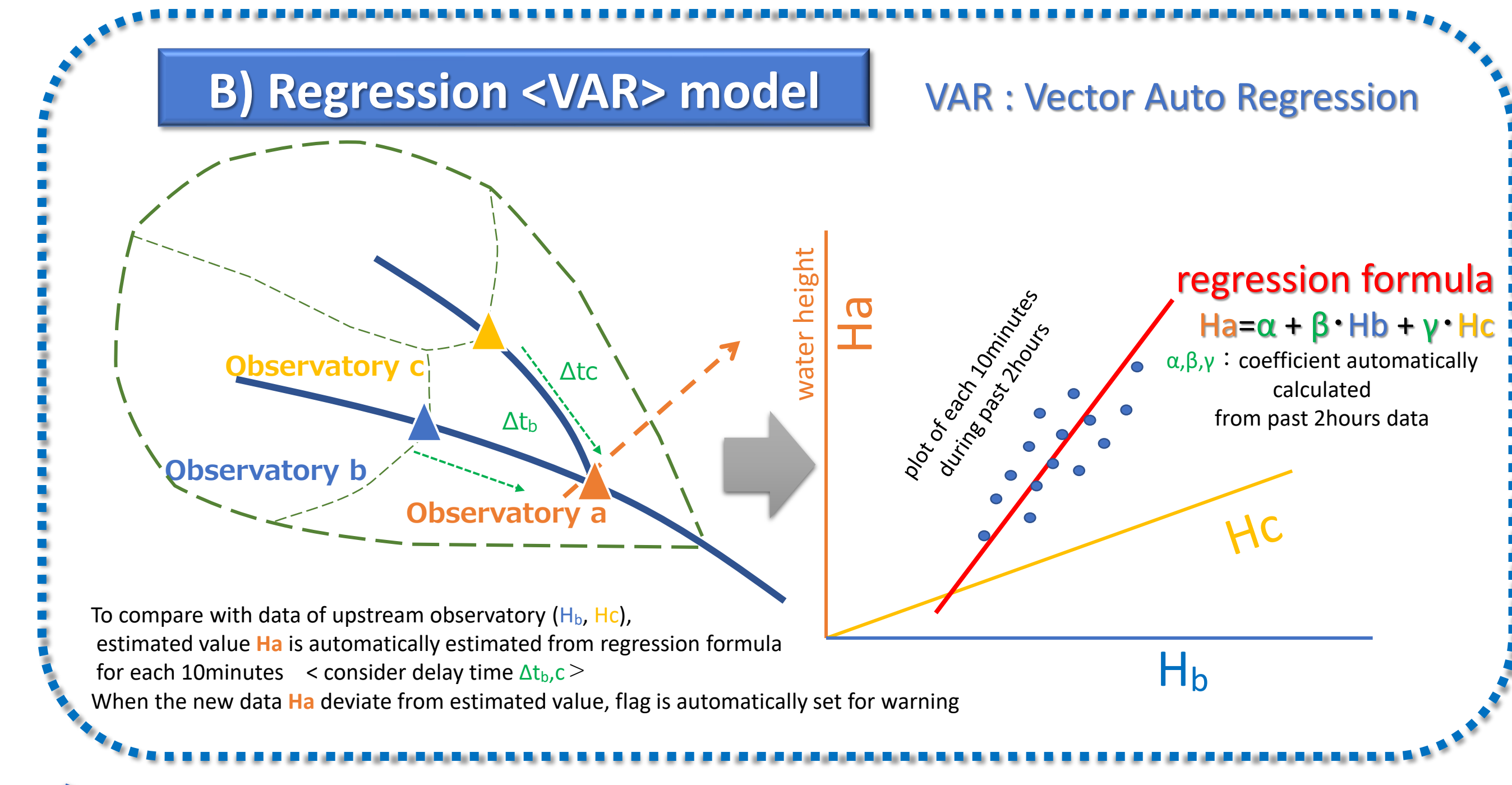
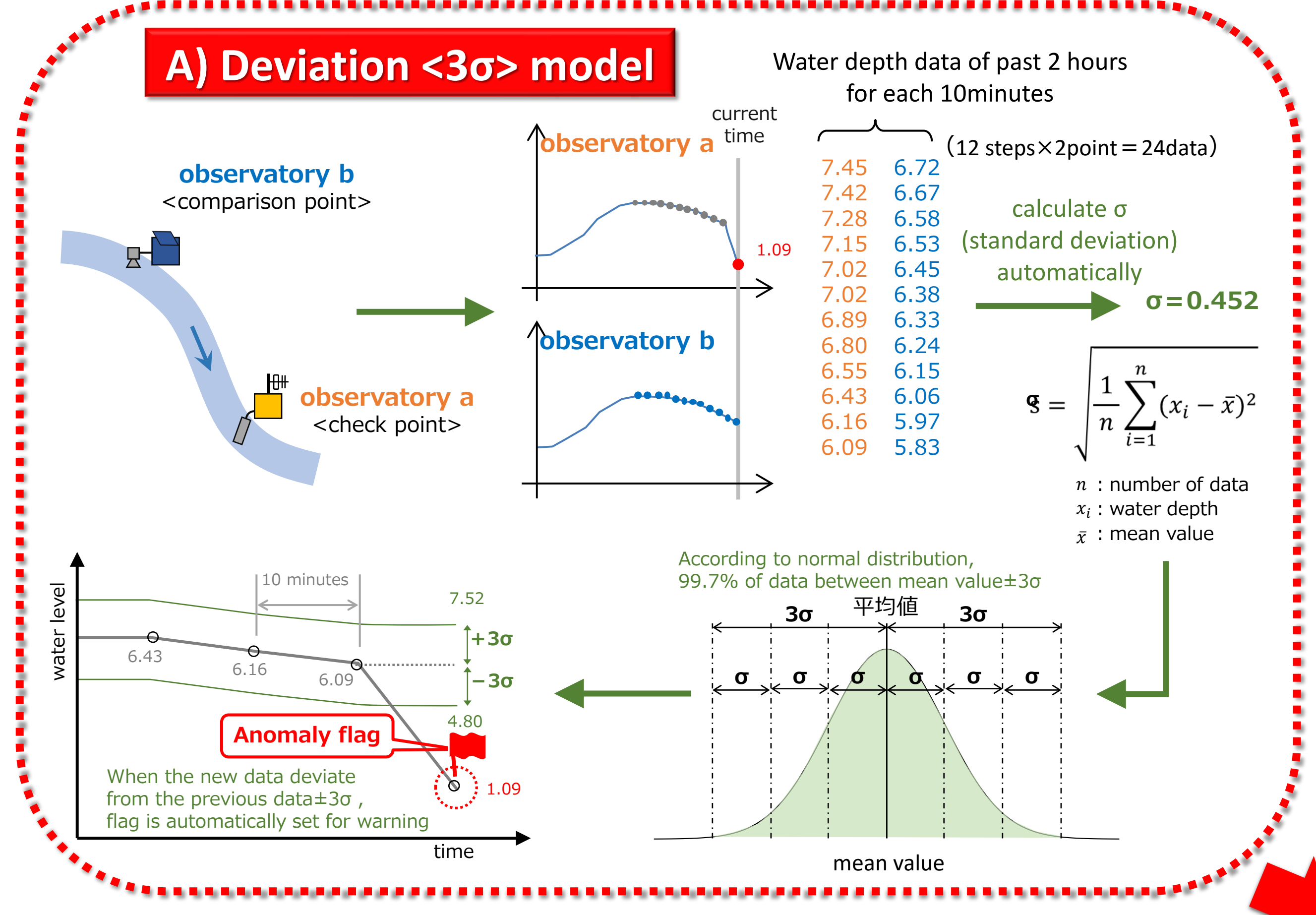
River information web site
https://www.river.go.jp/
< Ministry of Land, Infrastructure, Transport and Tourism>

Radar rainfall information

Water level gage <Telemetry system>
Graph of water level & precipitation

- In Japan, river water levels are observed at over 14,000 points
- Real-time data from these stations are integrated into the information system and delivered on web sites
- However, water gauges often produce irregular data, so, we need a precise data verification system at the monitoring station

II. Concept of anomaly detection models



- At first, we devised two types of anomaly detection model to compare with neighboring station data.
 - A) Deviation <3σ> model
 - B) Regression <VAR> model
- When compared with C) Existing difference model (comparison with previous data from the same station), both models showed a significantly better precision ratio.
- A) Deviation <3σ> model recorded highest Precision ratio. Therefore, we adopted this model as a real-time check system for the monitoring offices.

Verification

Result of checking some representative observation data during one rainy season

Precision rate : $P = TP / (TP + FP)$

Classification	Predicted	
	outlier	normal
Actual	outlier	TP
	normal	FP

A) Deviation <3σ> model P = 0.60
B) Regression <VAR> model P = 0.40
C) Existing differential model P = 0.21

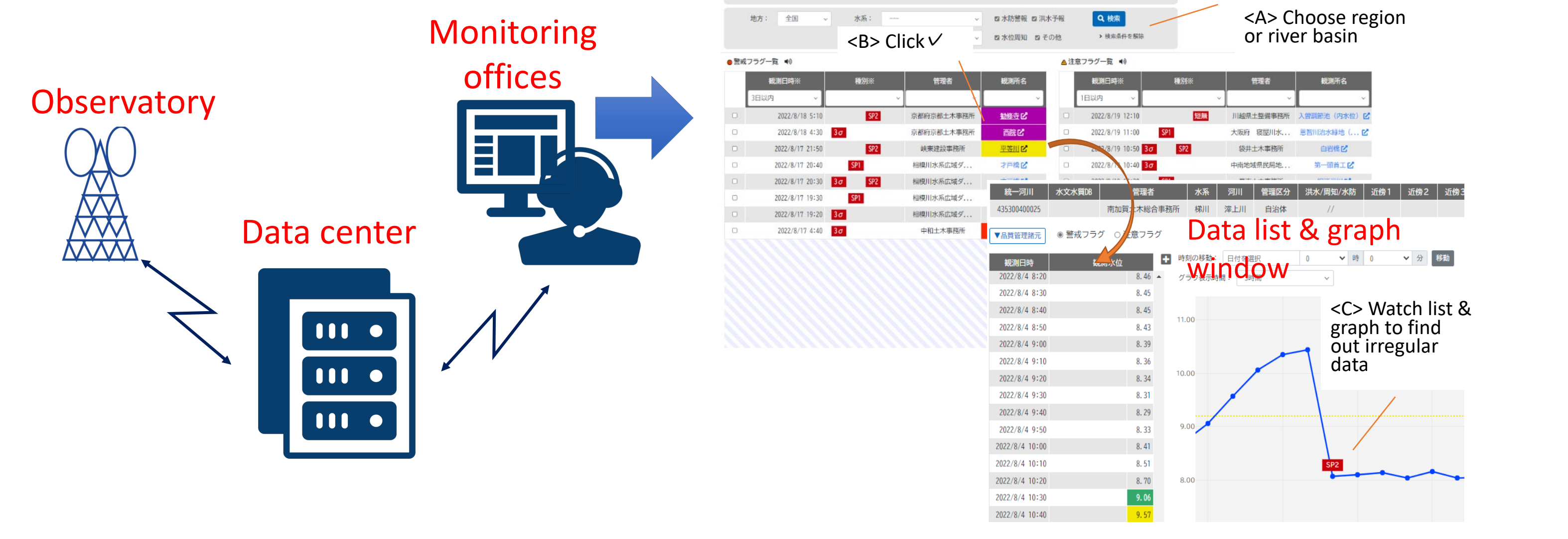
A) Deviation <3σ> model	Predicted	
	outlier	normal
Actual	outlier	6
	normal	4

B) Regression <VAR> model	Predicted	
	outlier	normal
Actual	outlier	6
	normal	9

C) Existing difference model	Predicted	
	outlier	normal
Actual	outlier	6
	normal	22

III. Construction of real-time data check system

- Two years ago, we constructed a real-time anomaly detection system hosted on a cloud server in a data center.
- Today, data inspectors in 24-hour monitoring offices utilize this system to identify and correct irregular data.
- This has significantly enhanced the precision and efficiency of our operations.



Structure of LSTM-Encoder Decoder

a) normal : observed data
b) normal : AI predicted data
a-b) threshold level

a) abnormal : observed data
b) abnormal : AI predicted data
a-b) exceed the threshold level

- We aim to further increase the accuracy by implementing an AI model using deep learning technology, specifically an LSTM-Encoder Decoder.
- Our prototype model has shown promising results, but we need to enhance its accuracy by training it with a larger dataset from across Japan.

We have tested the AI model using over 50 flood data series.

anormal data exceed threshold level

minimize $F = 2 \cdot P \cdot R / (P + R)$

Precision : $P = TP / (TP + FP)$

Recall : $R = TP / (TP + FN)$