

# IMPACT BASED ALERTING SYSTEM USING AI-BASED RADAR MAPS FROM SATELLITE OBSERVATIONS

*AOMSUC-13*

**Doyi Kim<sup>1</sup>**, Yeji Choi<sup>1</sup>, Yong-Jae Moon<sup>2</sup>, and Hyun-Jin Jeong<sup>2</sup>

1. SI Analytics
2. Kyung-Hee University

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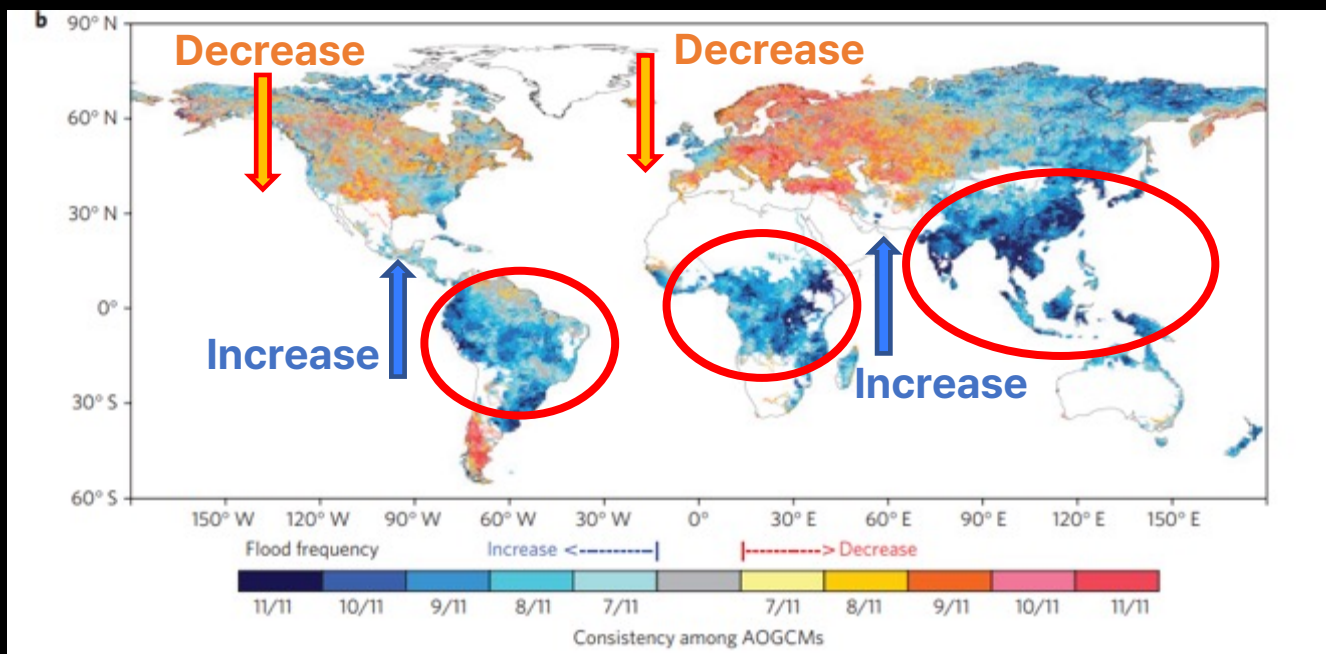
1. Background
2. Data and Method
3. Results and Conclusion

Chapter 01

# Background

# Background

- Weather radar detects and quantifies precipitation and severe weather conditions
- Radar system covers densely populated areas, but still insufficient to cover some regions and oceans



Flood frequency change [1]



A map of weather radar coverage [2]

[1] Hirabayashi, Yukiko, et al. "Global flood risk under climate change." *Nature climate change* 3.9 (2013): 816-821..

[2] Saltikoff, Elena, et al. "An overview of using weather radar for climatological studies: successes, challenges, and potential." *Bulletin of the American Meteorological Society* 100.9 (2019): 1739-1752.



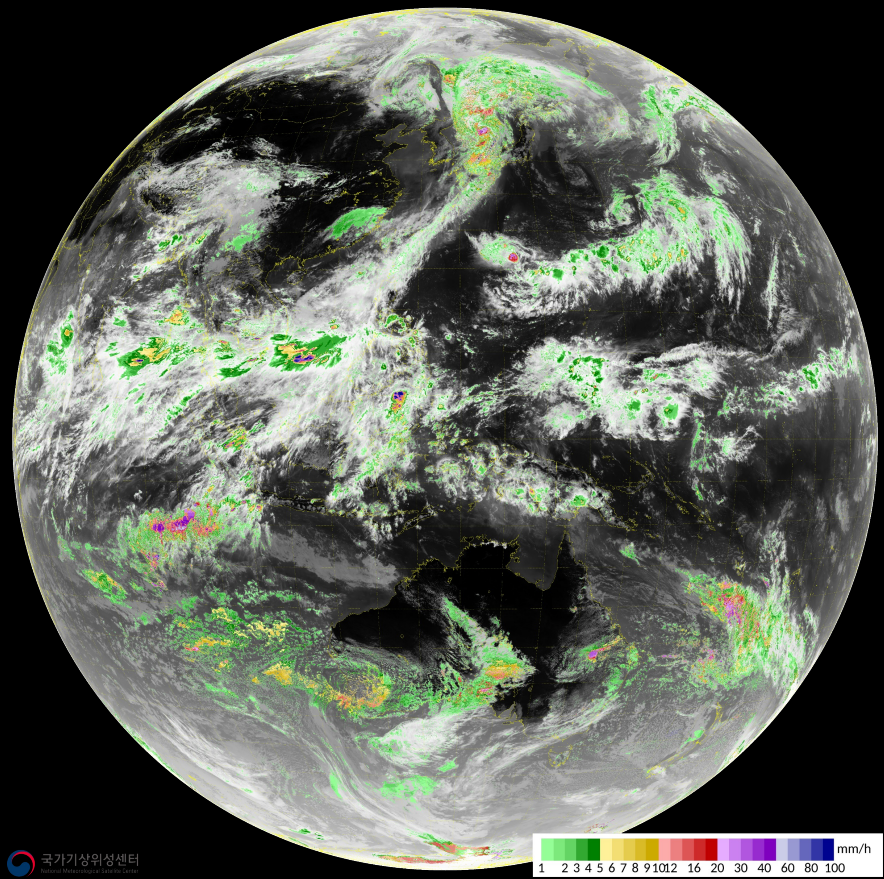
# Background

## Why Satellite?

- Continuous and Wide observation area, High-resolution imagery, Spectral diversity

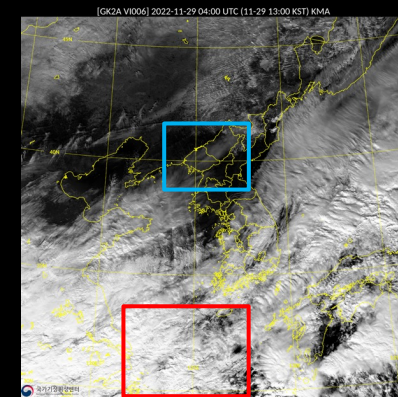


# GEO-KOMPSAT-2A



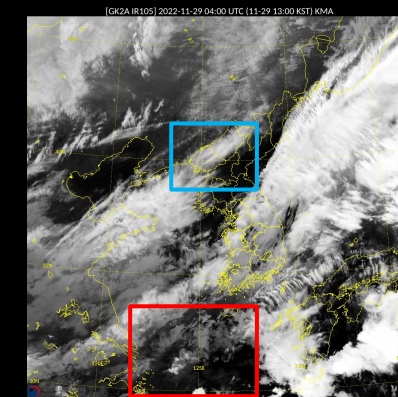
	Day time				All time											
<b>GK2A</b>	VIS0.4	VIS0.5	VIS0.6	VIS0.8	NIR1.3	NIR1.6	SWIR3.8	WV6.3	WV6.9	WV7.3	IR8.7	IR9.6	IR10.5	IR11.2	IR12.3	IR13.3

Visible (VIS)



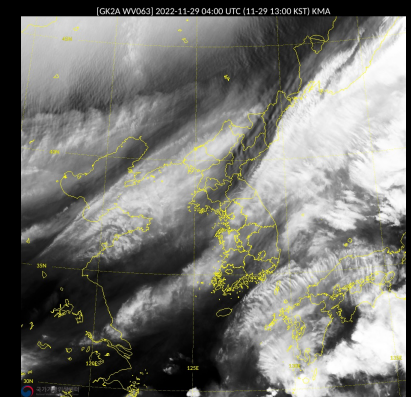
Low level clouds (Red box)

Infrared (IR)



High level cloud (Blue box)

Water vapor (WV)

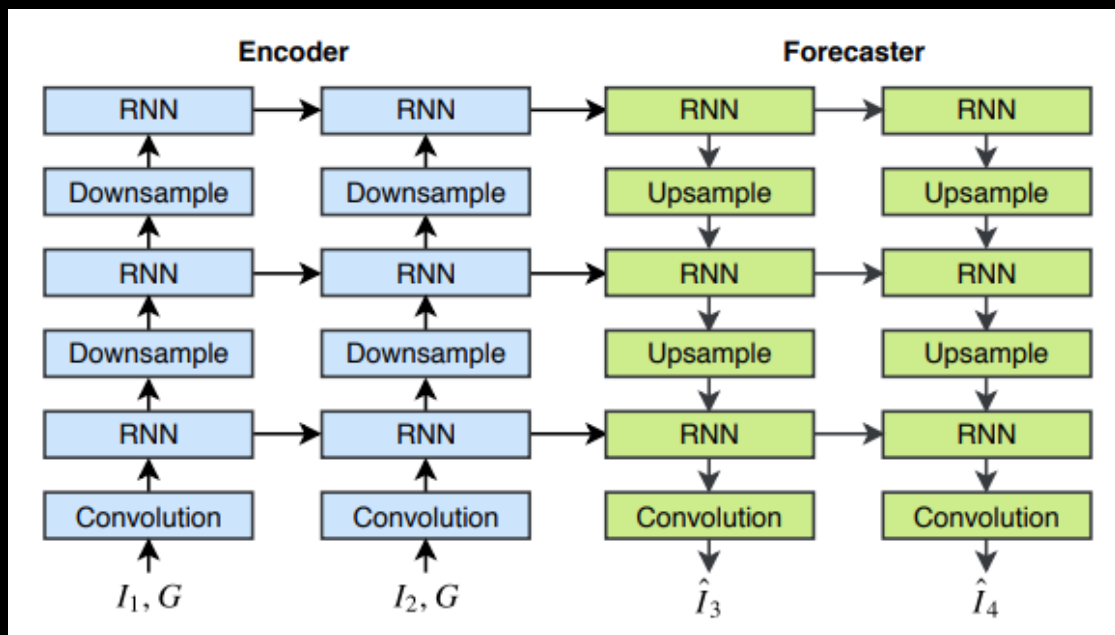


Mid level water vapor

# Background

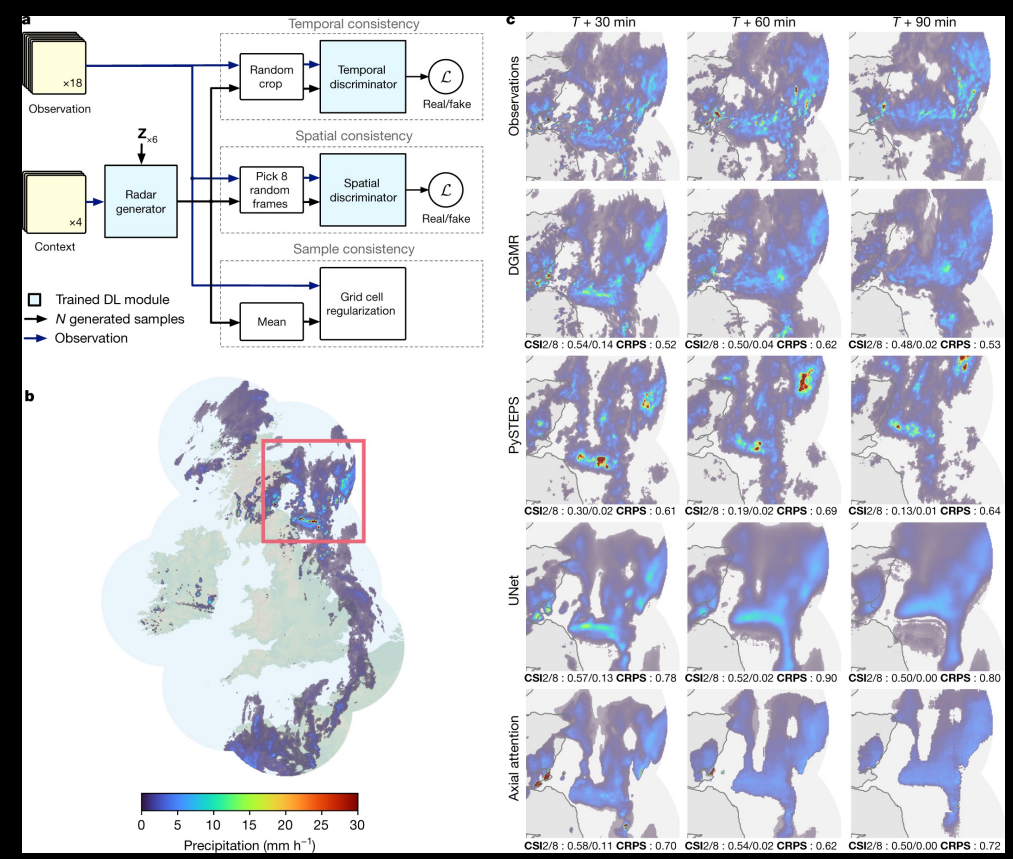
## Deep learning-based weather forecasting model

- Video Frame Prediction



Trajectory GRU, [3]

- Deep generative model of Rain



DGMR, [4]

[3] Shi, Xingjian, et al. "Deep learning for precipitation nowcasting: A benchmark and a new model." *Advances in neural information processing systems* 30 (2017).

[4] Ravuri, Suman, et al. "Skilful precipitation nowcasting using deep generative models of radar." *Nature* 597.7878 (2021): 672-677.



# Goal

## Accuracy Weather Forecasting without Radar System

**Spatio-temporal limited detection → Geostationary satellite observation**

**Insufficient Radar system → Proxy radar map from satellite images**

Chapter 02

# Data and Method



# Data

## Geo-KOMPSAT-2A (GK2A)

- Korean geostationary satellite
- 2-minutes interval and 0.5 to 2 km spatial resolution
- 16 channels – Including Visible (VIS), Water vapor (WV) and Infrared (IR) channels

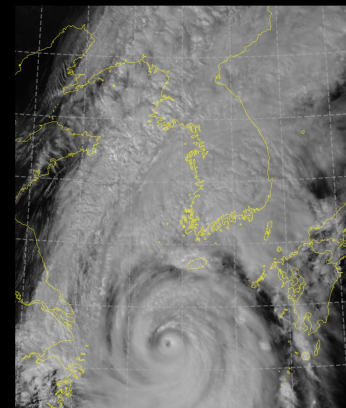
## KMA Weather Radar

- 5-minutes interval and 0.5 km spatial resolution

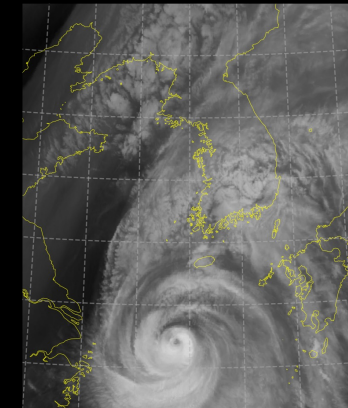
## Test Cases:

- 2023/01/12 2030 UTC (Heavy Rain)
- 2023/05/05 0730 UTC (Heavy Rain)
- 2023/07/14 0010 UTC (Jangma)
- 2022/08/09 1020 UTC (Typhoon Hinnamnor)

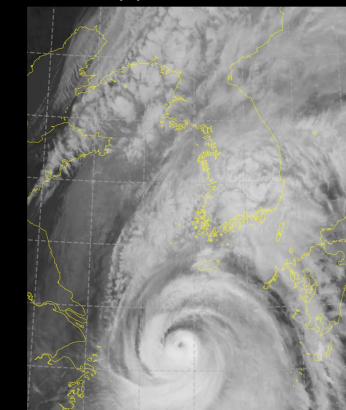
(a) VIS 0.06



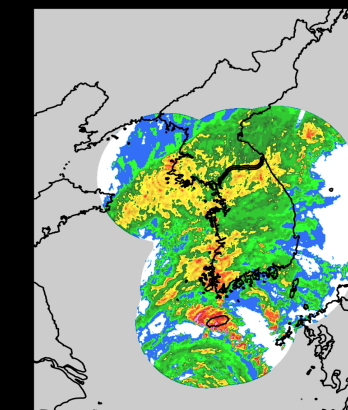
(b) WV 6.04



(c) IR 10.05



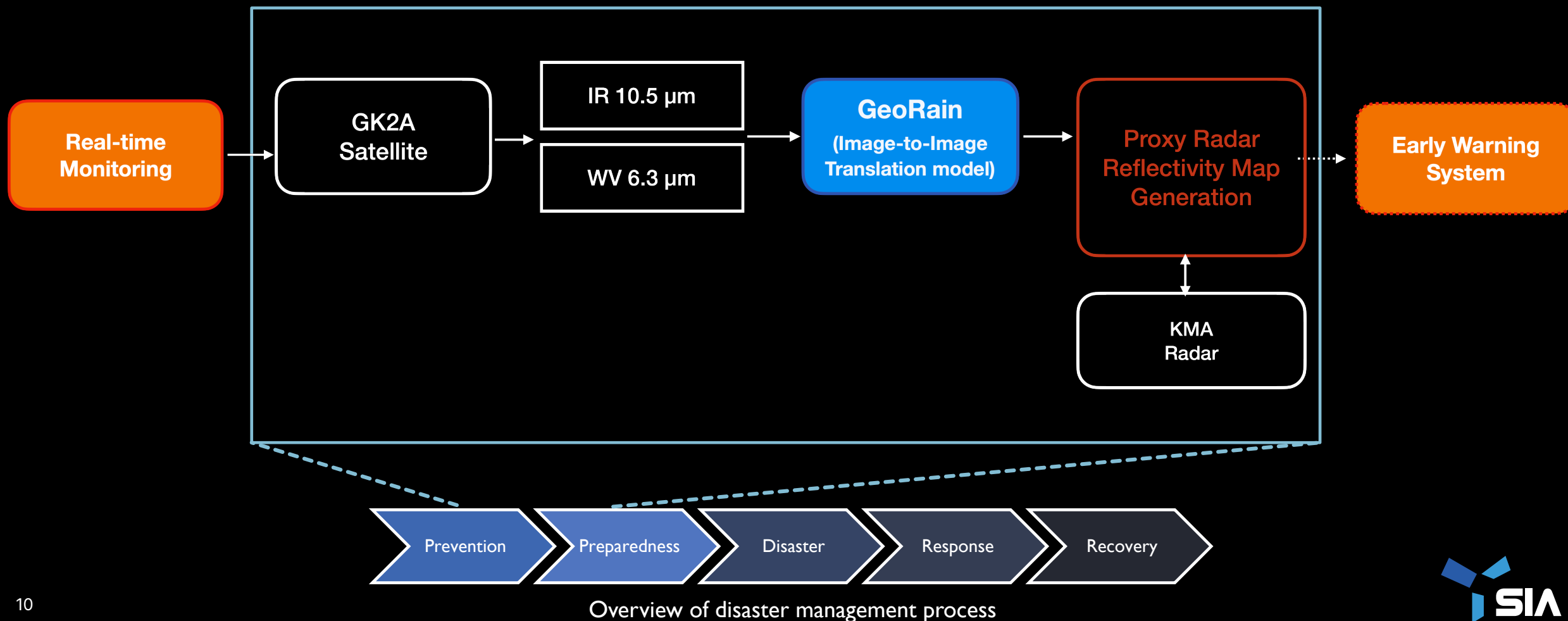
(d) KMA Radar



# SIA's Disaster Monitoring Model

Part of the real-time disaster monitoring model

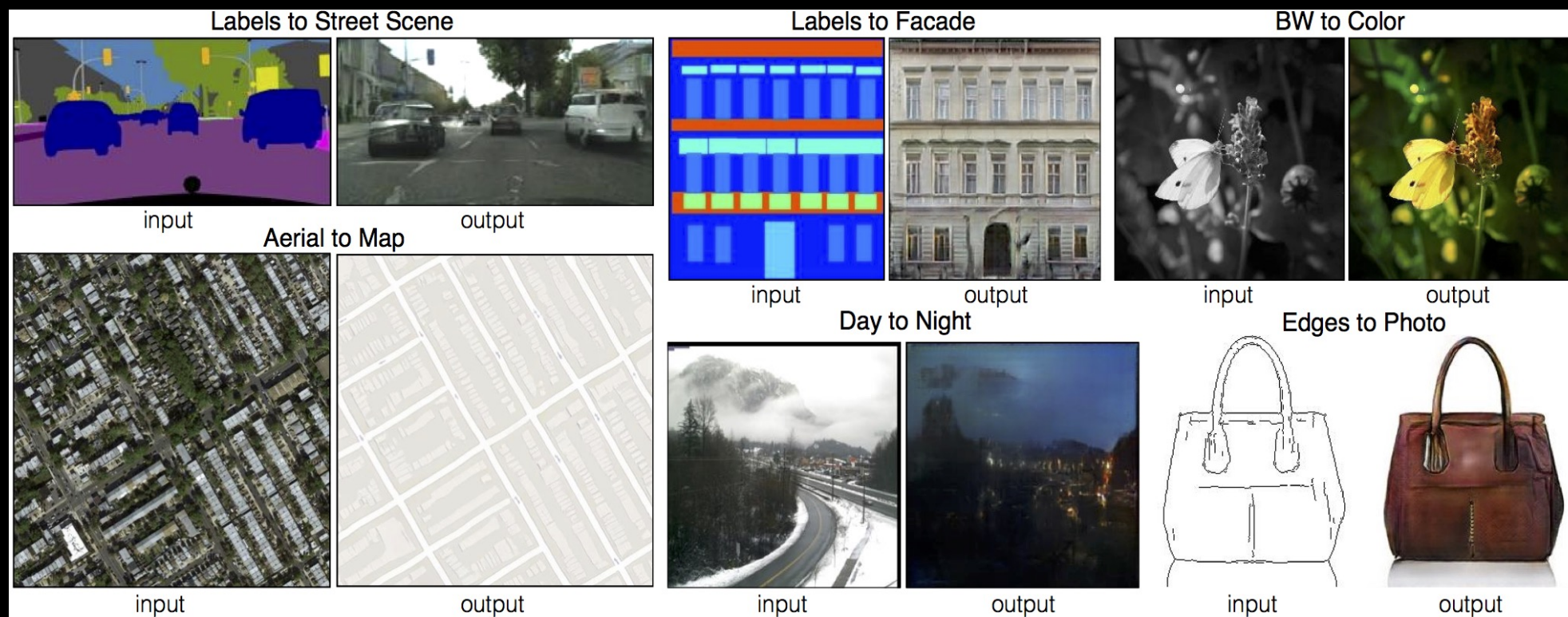
- A proxy radar map is generated to detect and predict heavy rain events



# GeoRain\_v1.0

## Generative Adversarial Network for rain – GeoRain

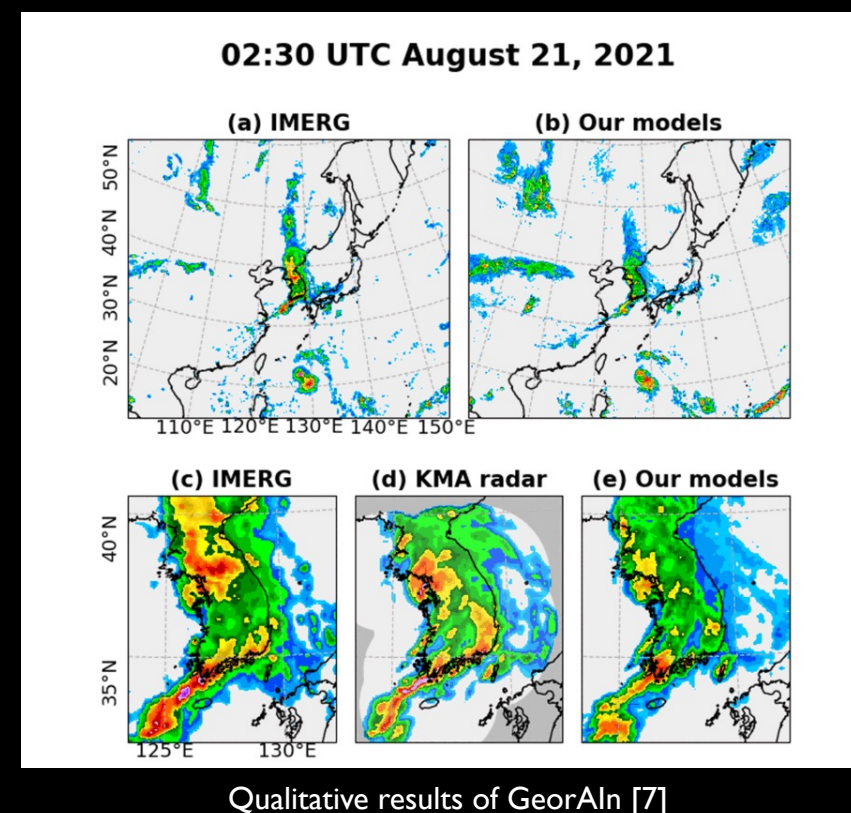
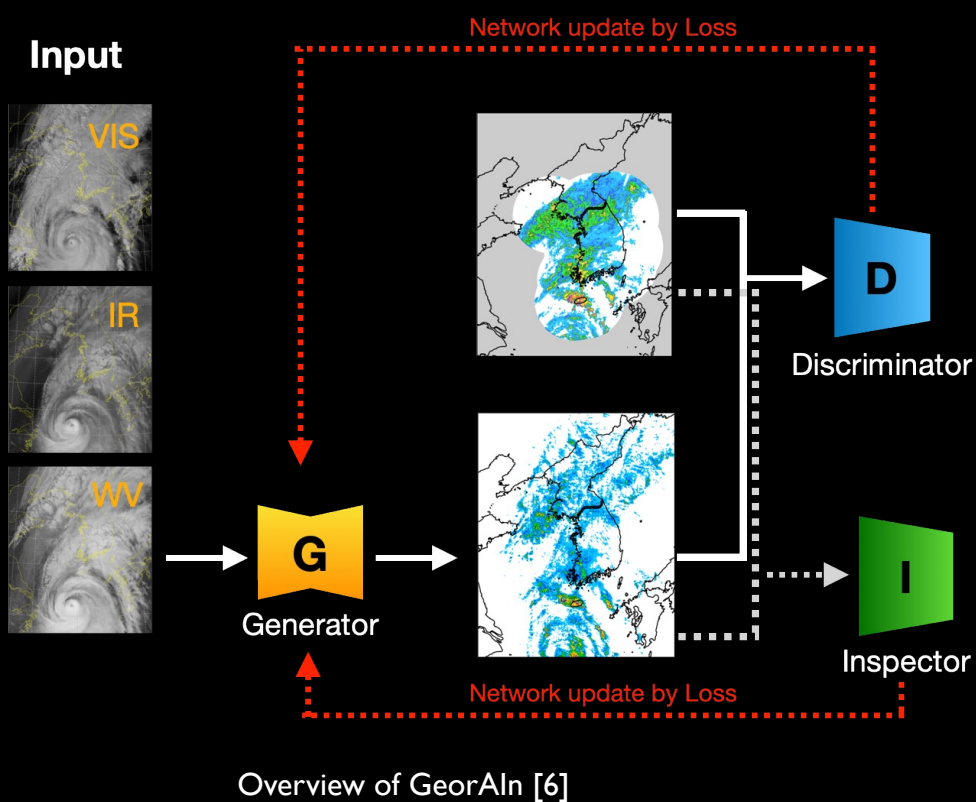
- Isola et al.(2017)[5] suggested a general-purpose solution to resolve image-to-image translation problems using conditional GANs (cGAN)



# GeoRain\_v1.0

## Generative Adversarial Network for rain – GeoRain

- Generate proxy radar reflectivity map using Pix2PixCC model ([6])
- Inspector guides the generated image to be physically consistent with the real image

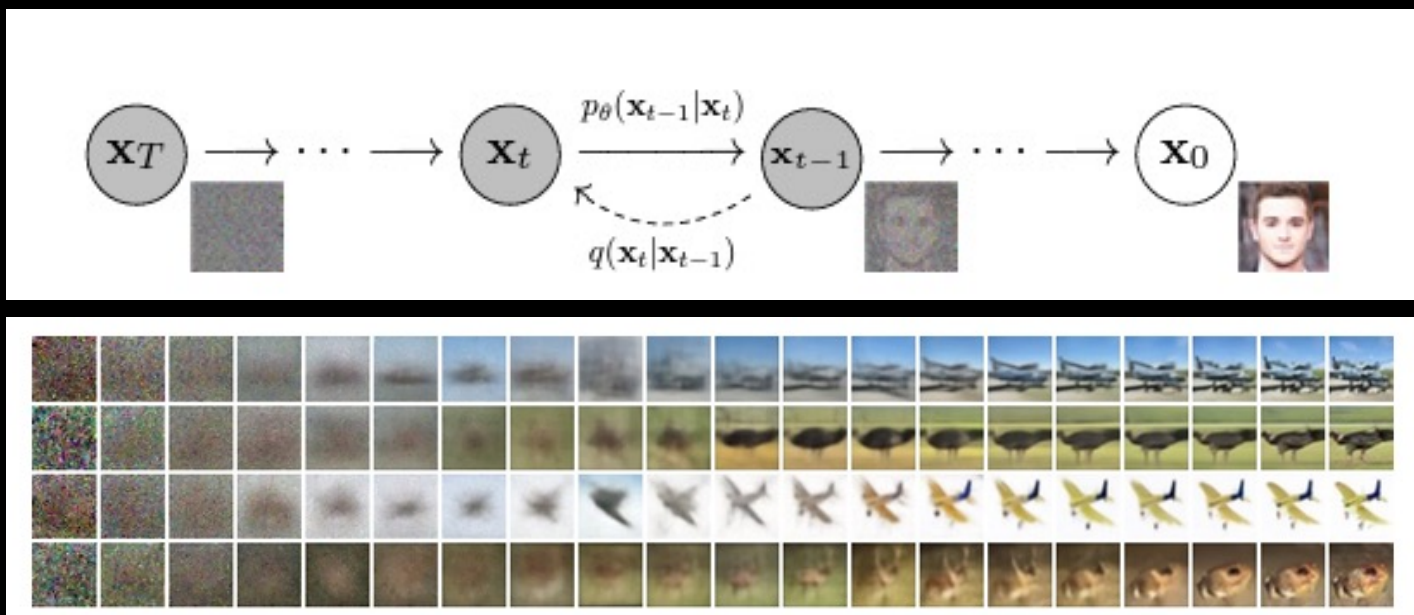




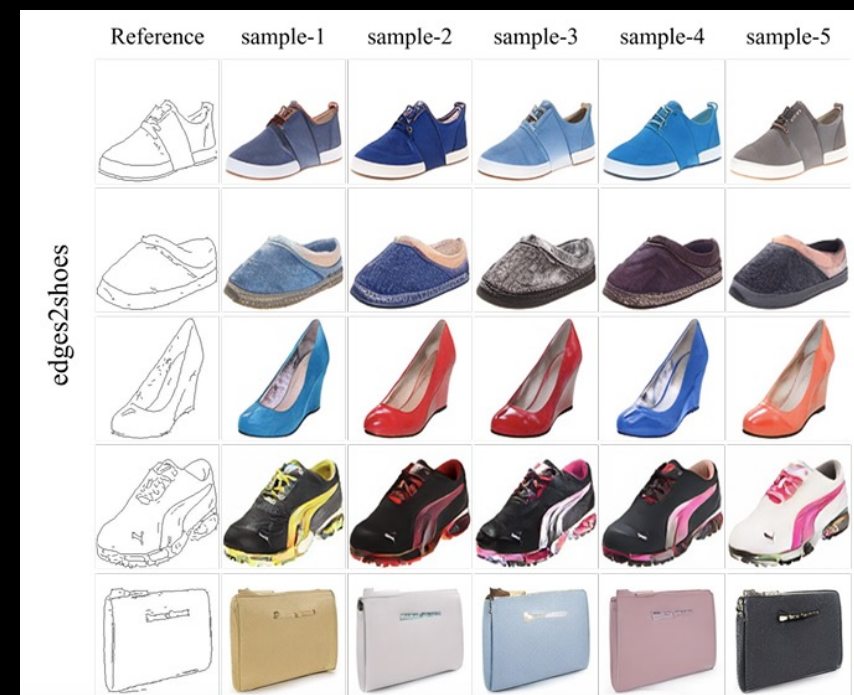
# GeoRain\_v2.0

## Diffusion-based rain forecasting model

- Generate proxy radar reflectivity map using diffusion model (BBDM, [8])
- High Sample diversity and model stability than GAN –based model



Diffusion model structure and results [9]

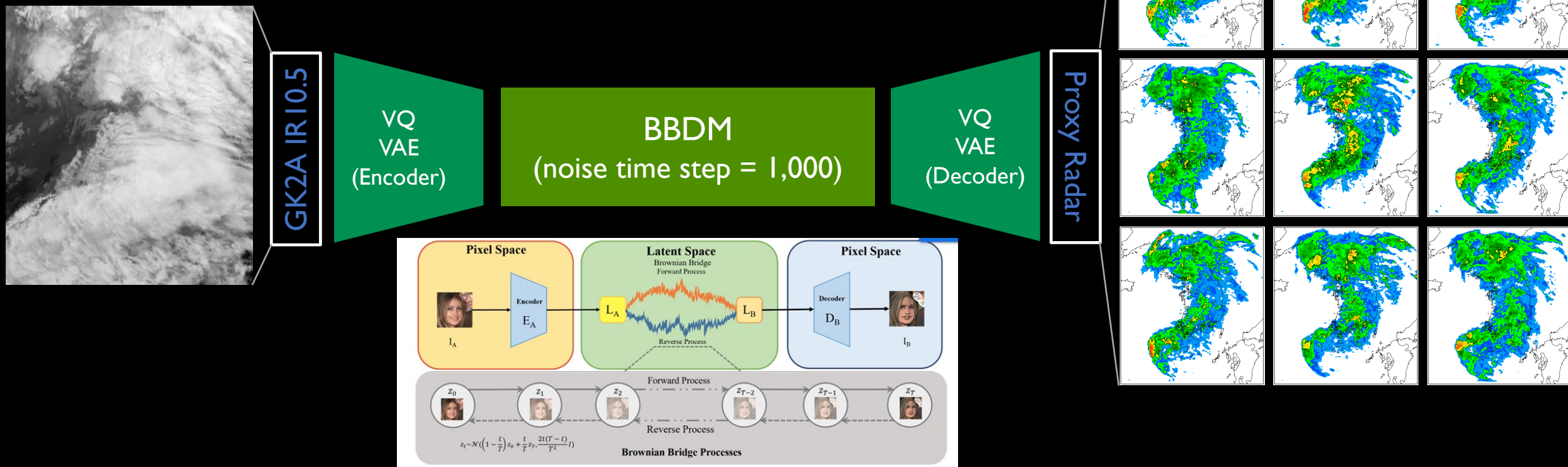


Diverse samples from BBDM [9]

# GeoRain\_v2.0

## Diffusion-based rain forecasting model

- Generate proxy radar reflectivity map using diffusion model (BBDM, [8])



Architecture of BBDM [8]

Chapter 03

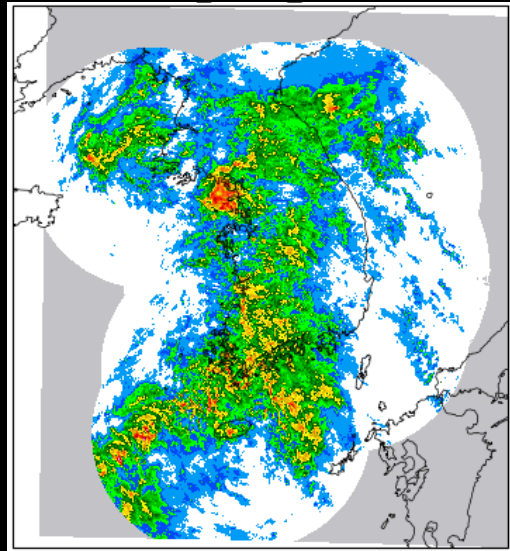
# Results

# Results: Case-1

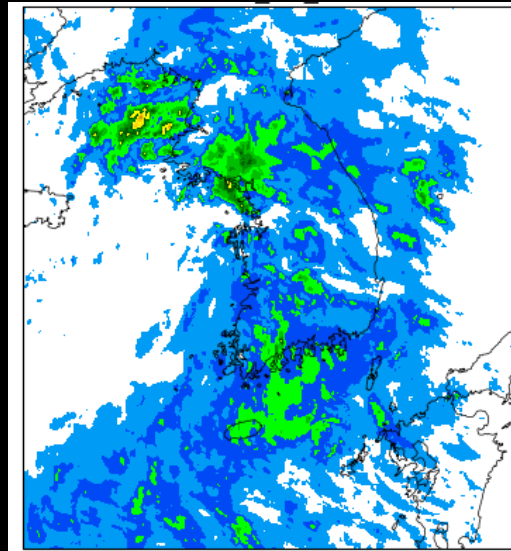
1) 2023/01/12 2030 UTC – Moderate Rain



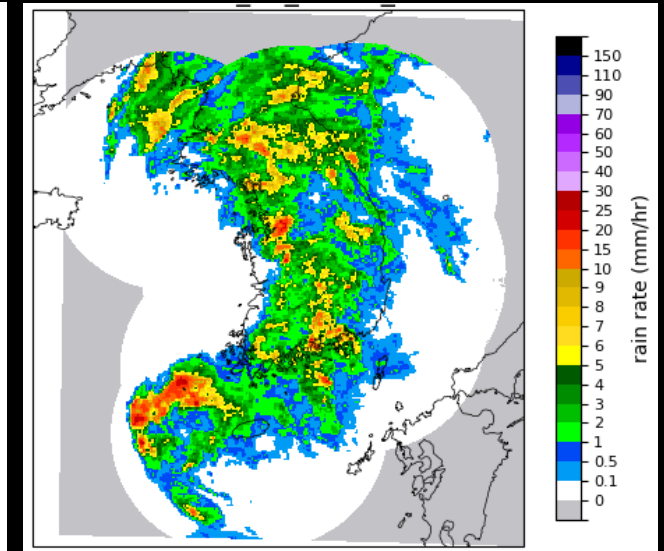
(a) GK2A IR 10.5



(b) KMA\_radar



(c) GeoRain\_v1.0

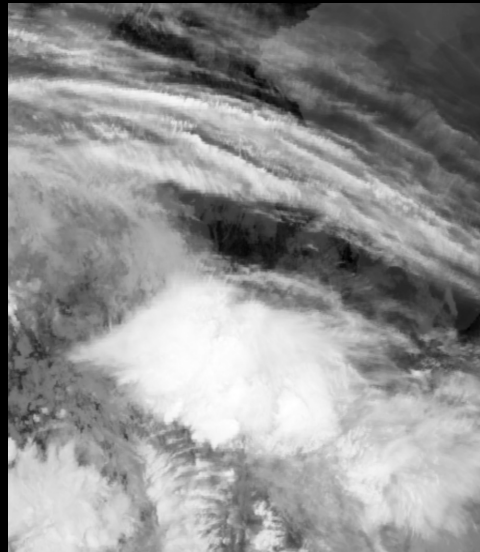


(d) GeoRain\_v2.0

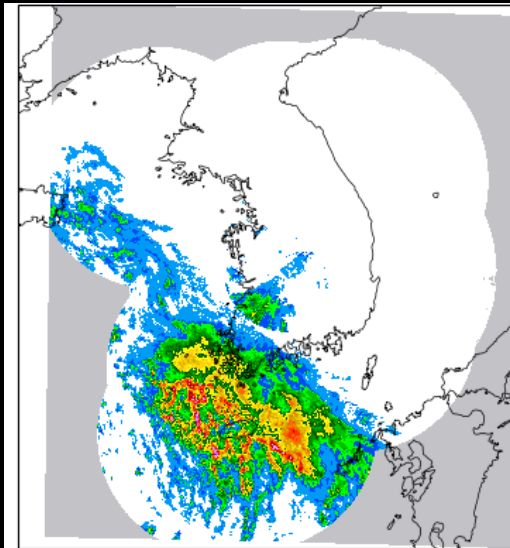


# Results: Case-2

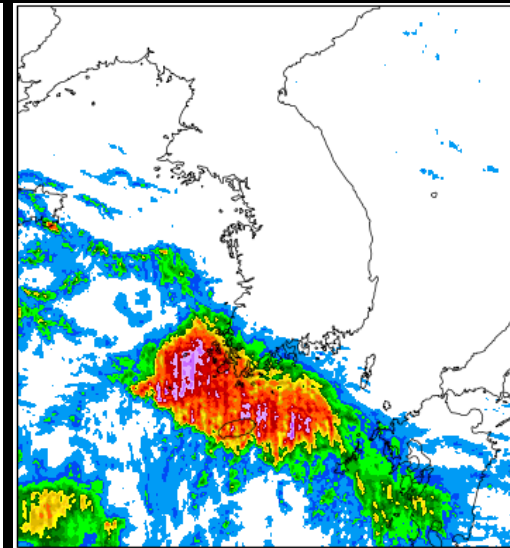
2) 2023/05/04 0100 UTC – Heavy Rain



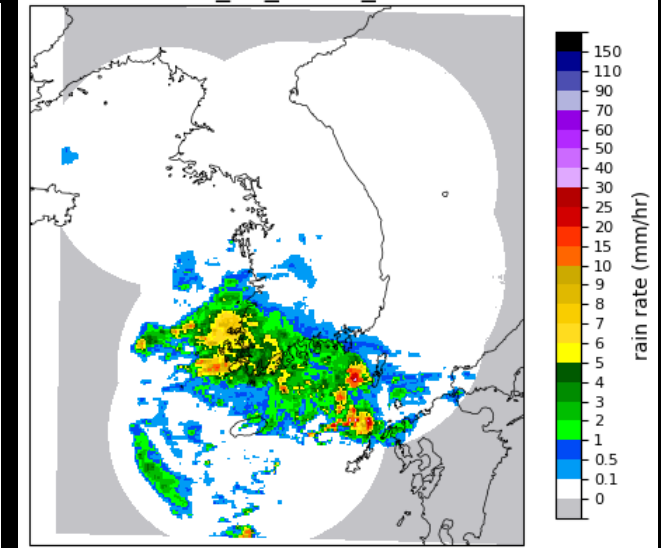
(a) GK2A IR 10.5



(b) KMA\_radar



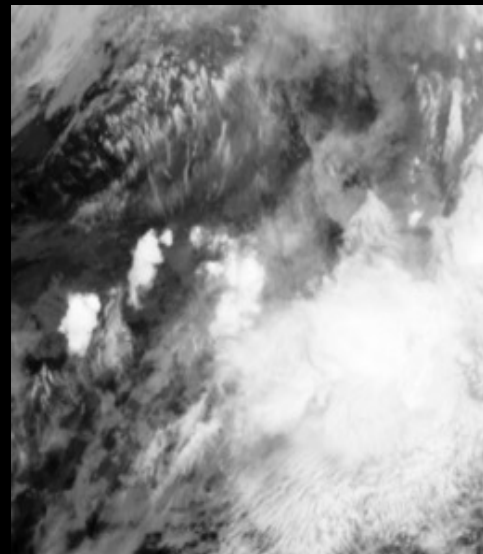
(c) GeoRain\_v1.0



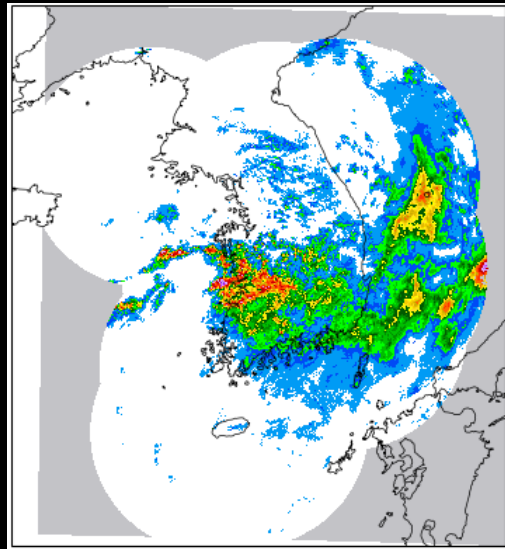
(d) GeoRain\_v2.0

# Results: Case-3

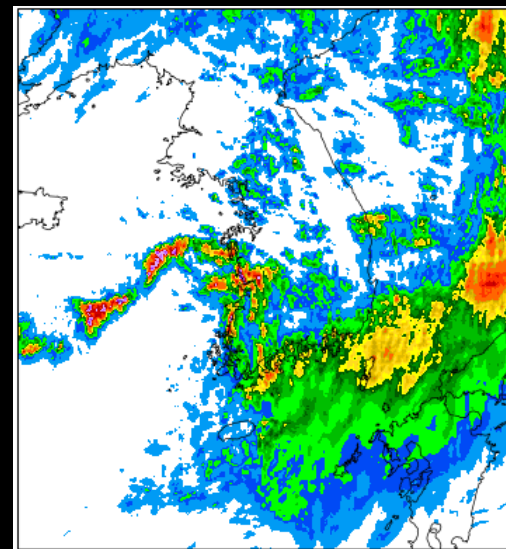
3) 2023/07/14 0010 UTC – Heavy Rain(Jangma)



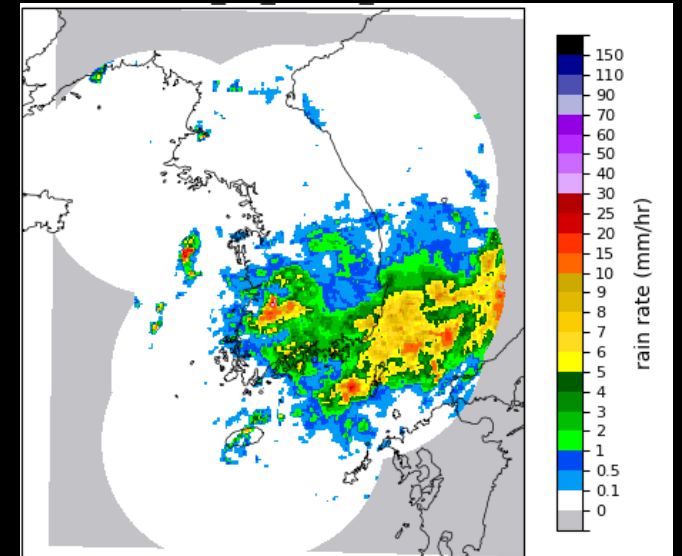
(a) GK2A IR 10.5



(b) KMA\_radar



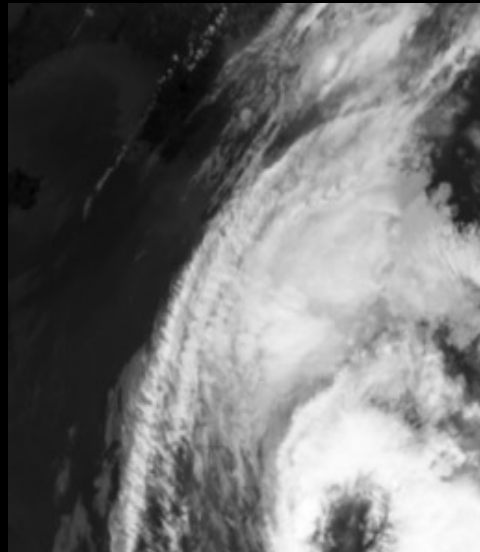
(c) GeoRain\_v1.0



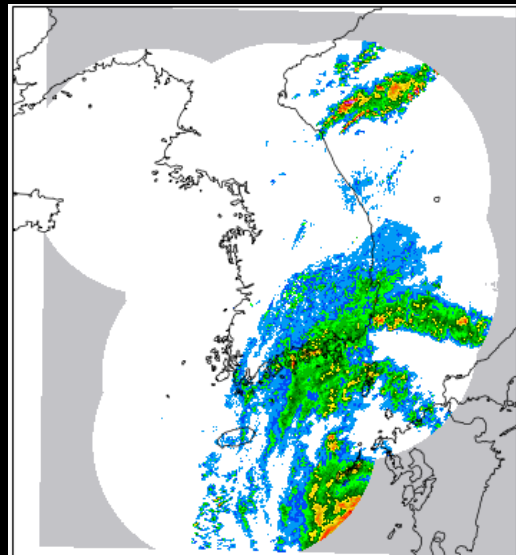
(d) GeoRain\_v2.0

# Results: Case-4

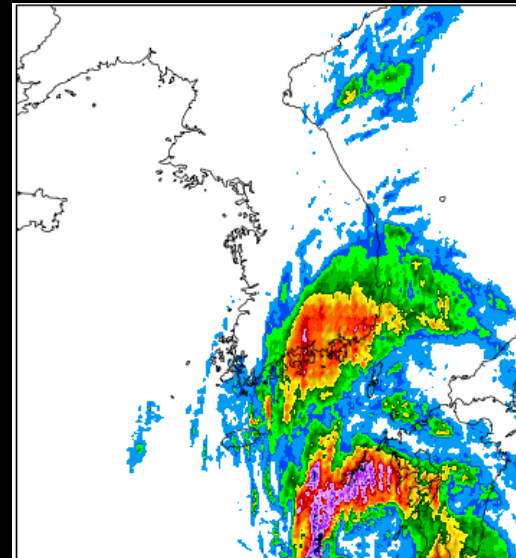
4) 2023/08/09 0120 UTC – Typhoon Hinnamnor



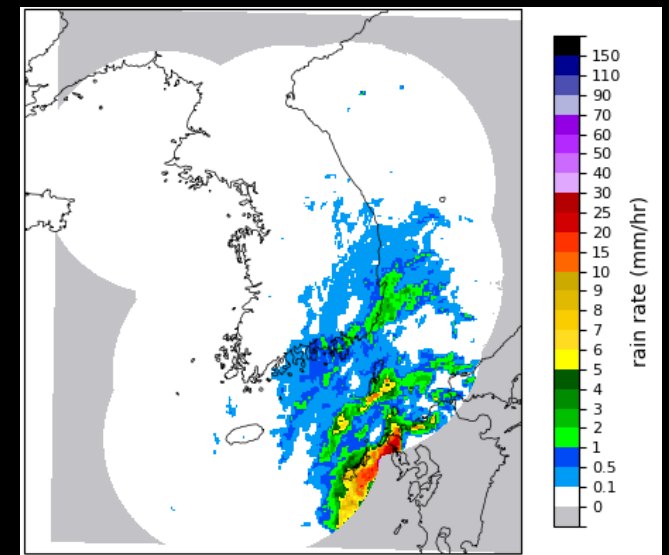
(a) GK2A IR 10.5



(b) KMA\_radar



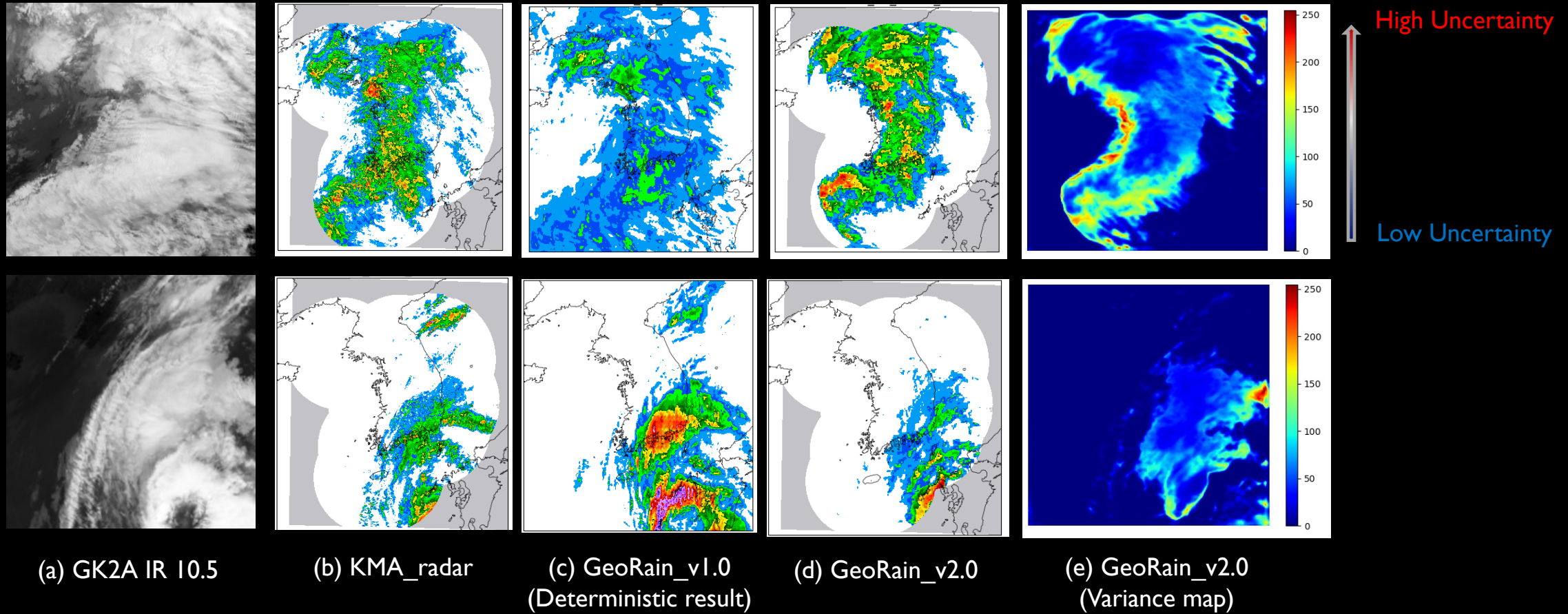
(c) GeoRain\_v1.0



(d) GeoRain\_v2.0

# Results: Stochastic results

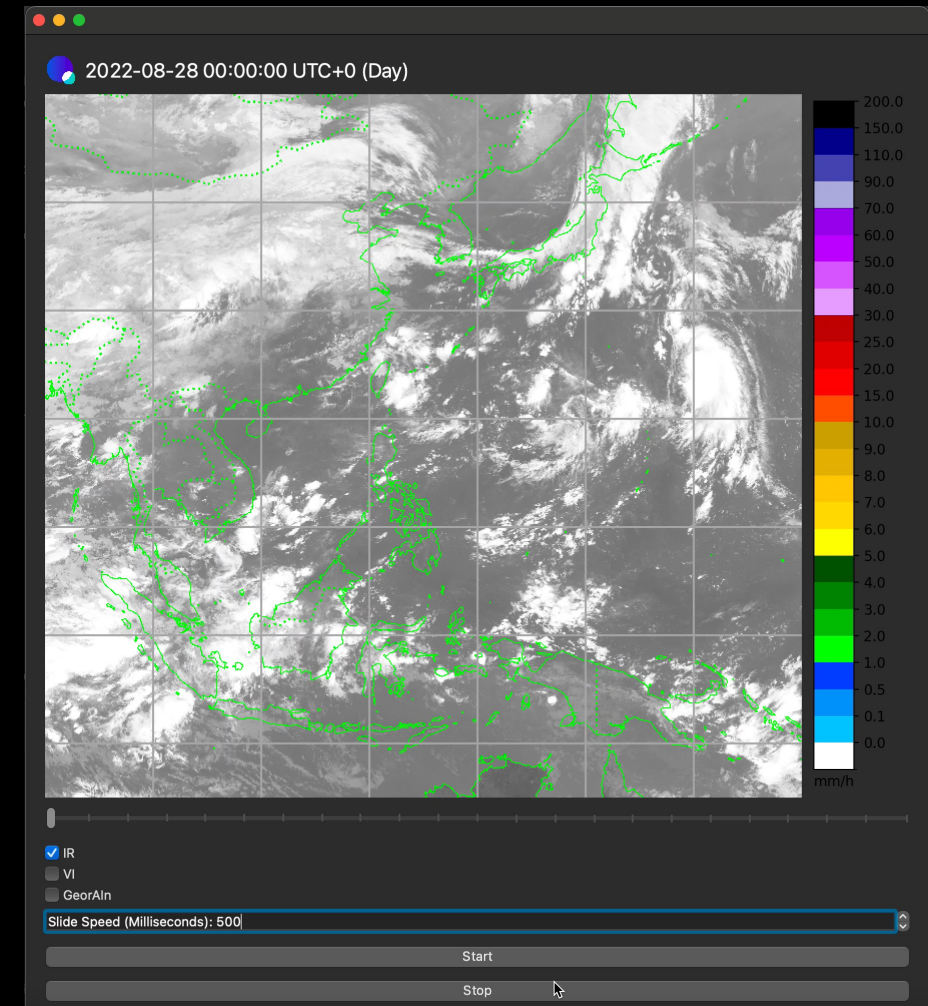
Diffusion model can generate stochastic forecasting





# Conclusions

- We generate proxy rainfall maps by using GeoRain model with geostationary satellite imageries .
- The GeoRain results show our model can predict the accurate timing, location, and intensity of heavy rain areas.
- The GeoRain\_v1.0 shows overestimated results than the GeoRain\_v2.0.
- Diffusion-based GeoRain\_v2.0 model can generate stochastic results that take uncertainty into account.
- We expect our service to help communicate preemptive and precise early warnings.
- We plan to expand our disaster monitoring model to predict future precipitation on a global scale.



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# Thank you for attention

[doyikim@si-analytics.ai](mailto:doyikim@si-analytics.ai)

[www.si-analytics.ai](http://www.si-analytics.ai)

# Appendix

- Training information – GeoRain models

	WR-Net	GeoRain_v1	GeoRain_v2
Training data	GK2A 2020.08-2021.07, 2 min	GK2A 2019.08-2021.07. 10 min	
Base model	TV-L1 algorithm (optical flow)/ U-Net based VGG16 (refinement)	Pix2PixCC model	Diffusion model(BBDM)
Loss function	PSNR, SSIM	LSGAN, FM loss, CC loss	BBDM score loss
Optimizer	Adam	Adam	Adam
Learning Rate	1e-4	0.0002	1e-4