

# Nowcasting and Community SWIRLS

SWFDP-SeA RFSC Training Desk, Hanoi, Vietnam

16 – 20 December 2019

# Overview

- On Nowcasting
- SWIRLS in Hong Kong
- Community SWIRLS (Com-SWIRLS)
- WMO RSMC for Nowcasting
- Latest Developments

**ON NOWCASTING**

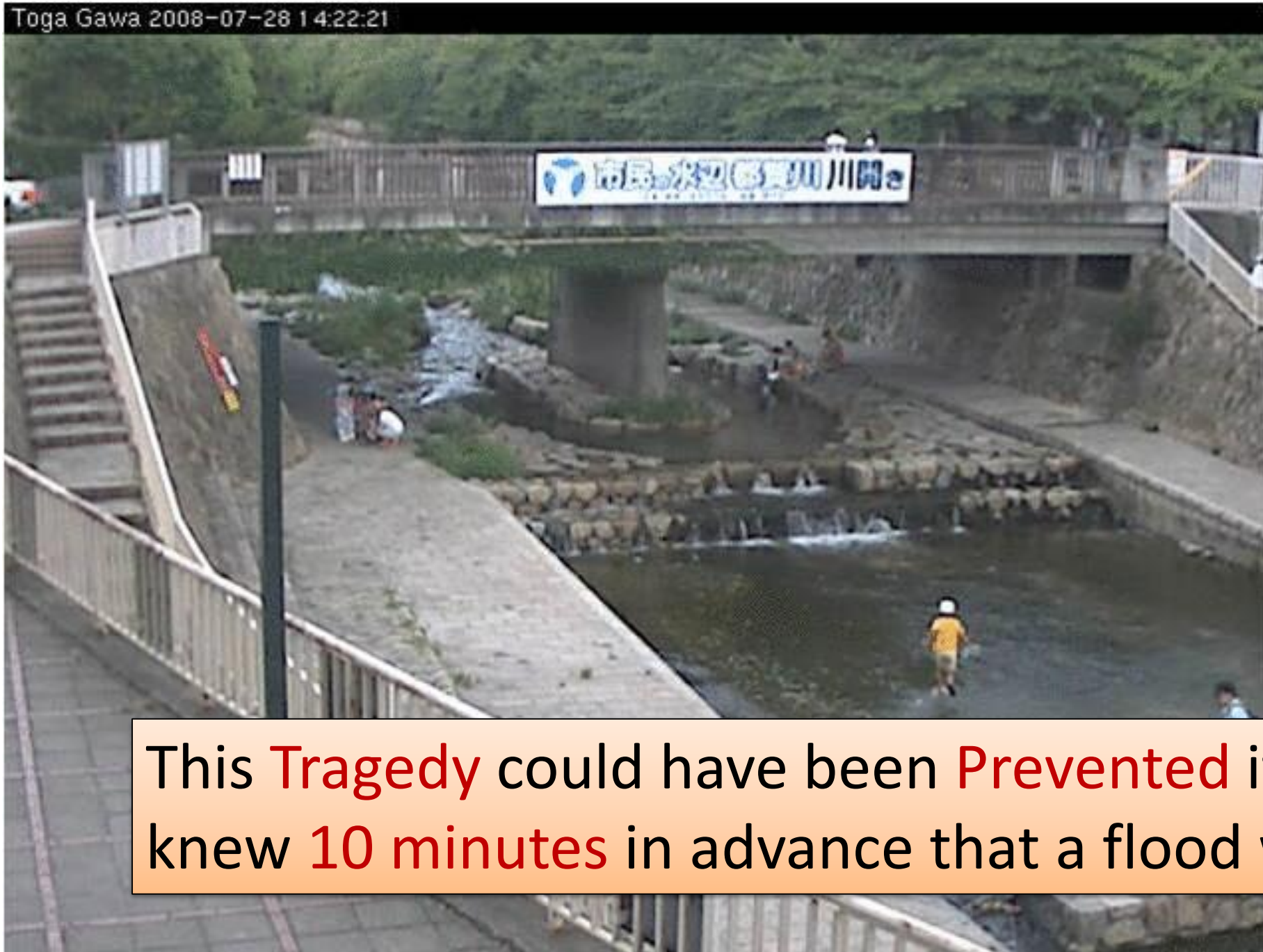
Toga Gawa 2008-07-28 14:22:21

Kobe  
Japan  
July 2008

50 People  
Washed  
Away

5 Died

This **Tragedy** could have been **Prevented** if only they knew **10 minutes** in advance that a flood was coming.





Every year, Storms & Floods kill **Thousands** and wipe away properties of **Tens of Billions of Dollars**.



Myanmar



Macao



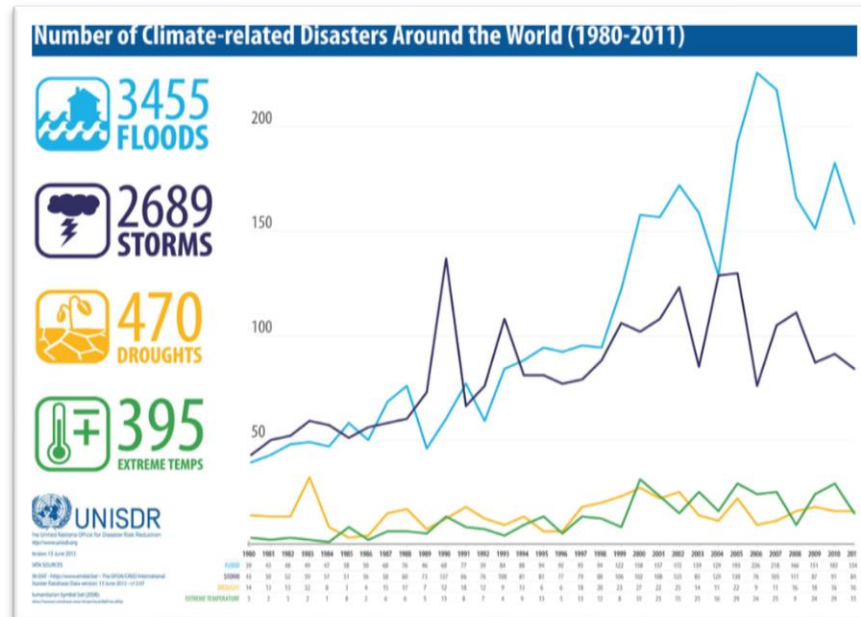
Bangladesh



Sri Lanka



Thailand



Sudden & Extreme  
Weather keeps  
**Increasing** under  
**Climate Change**.

# Scales of Weather Forecasts

## Time Horizon

### 3 Month Forecast

	Temperature Forecast	Rainfall Forecast
Jun – Aug 2019	Normal to above normal	Normal to above normal
Mar – May 2019	Normal to above normal	Normal to above normal
Dec 2018 – Feb 2019	Normal to above normal	Normal to above normal
Sep – Nov 2018	Normal to above normal	Normal to below normal

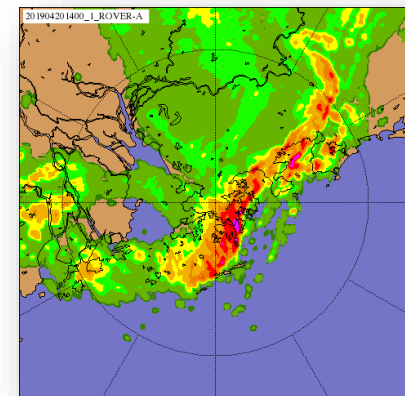
Roughly

### 1 - 9 Day Forecast

8 Nov (FRI)	9 Nov (SAT)	10 Nov (SUN)	11 Nov (MON)
			
21   26 °C 40 - 70 %	20   25 °C 50 - 80 %	21   25 °C 55 - 80 %	22   26 °C 60 - 80 %

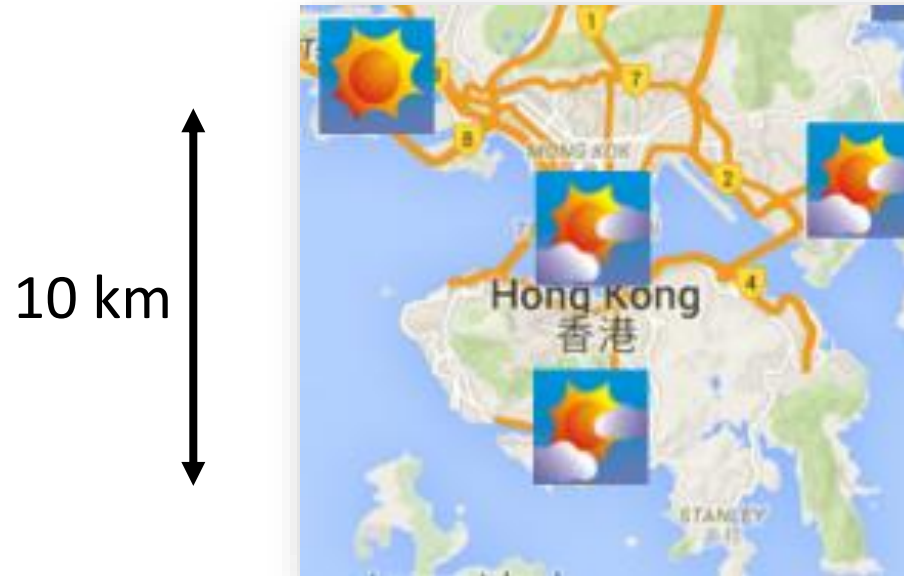
Generally

### 0 – 6 Hour Nowcast



Precisely

# Nowcasting - Resolutions Matter



**Forecast**  
**Range**  
**Interval**

1 – 9 Day  
Hourly



**Nowcast**  
**Range**  
**Interval**

0 – 6 Hour  
6-Minutely



# SWIRLS Nowcast System

SWIRLS: Short-range Warning of Intense Rainstorms in Localized Systems

## Observations



Satellites



Radars



Rain  
Gauges

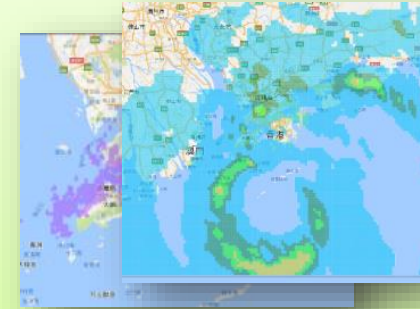


Lightning  
Detectors

## Nowcast Modelling



Computers  
80 TFLOPS CPU/GPU  
1.5 TB RAM  
960 TB Storage



Forecast  
Rainfall,  
Lightning,  
Hail & Gust

## Products and Services



Weather  
Forecasters



Citizens



Government &  
Utilities

# With **SWIRLS** Nowcasting



## **Weather Forecasters**

Issue Weather Warnings and Forecasts, which

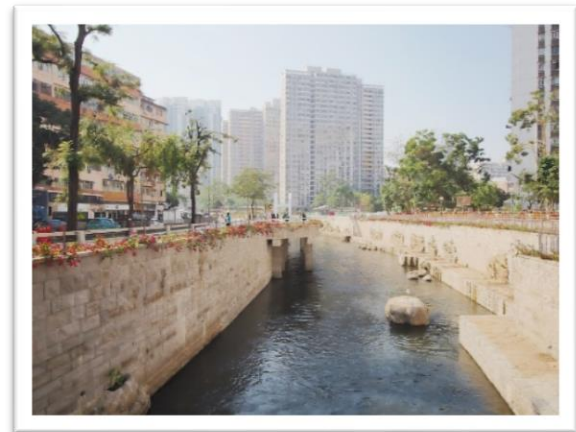
- Save Lives
- Protect Properties



## **Citizens**

Decide whether to:

- Seek Shelter?
- Bring Umbrellas?
- Cancel Hiking?
- Indoors or Outdoors?



## **Government & Utilities**

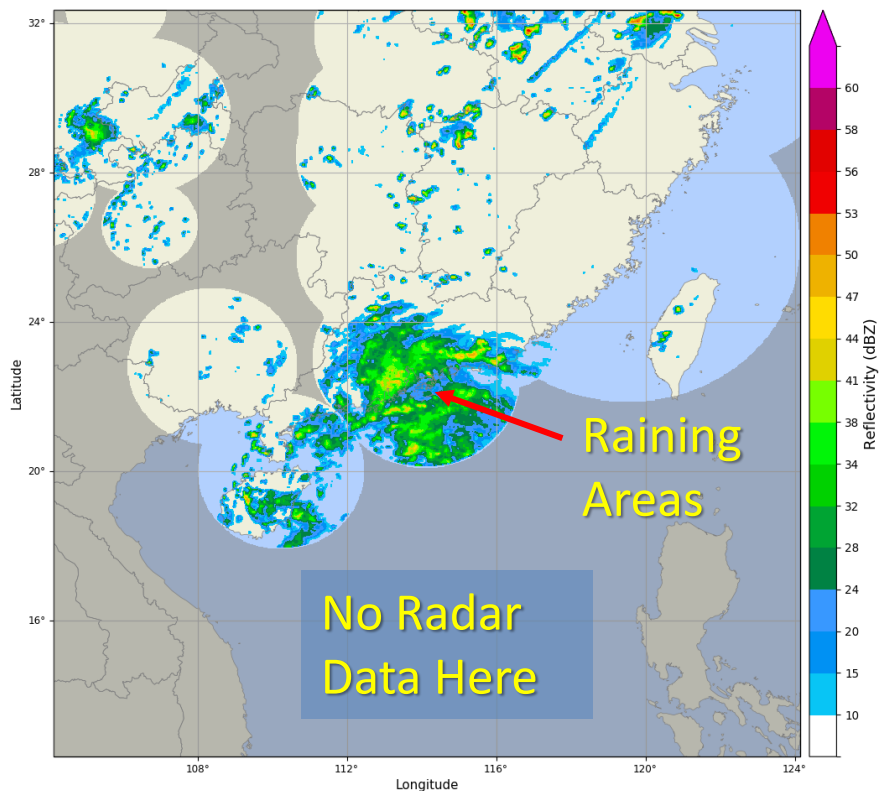
Respond to:

- Improve Ecology of Wetland
- Enhance Productivity of Works
- Mobilize Staff in Advance
- Increase Power Supply Stability
- Suspend and Resume Operations

# 1. Simulate Radar Data from Satellite's using AI Neural Network in SWIRLS

Reflectivity  
CMA Radar / SC

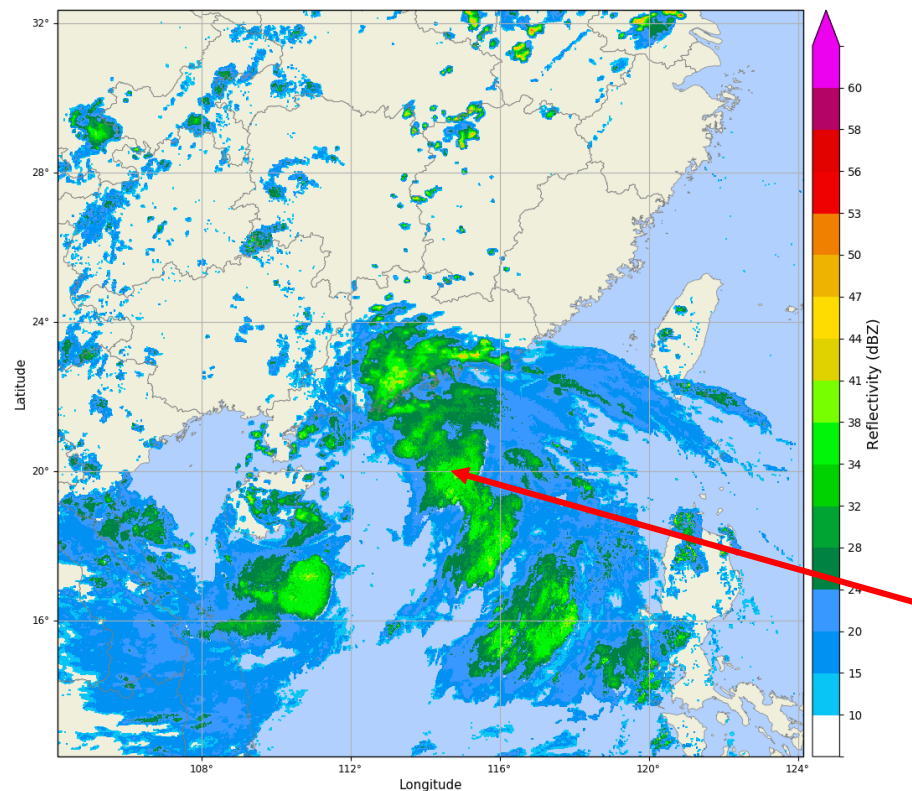
2019-07-31 Based @ 15:00H  
Valid @ 15:06H



Original

Reflectivity  
Multi-Sensor / SC

2019-07-31 Based @ 15:00H  
Valid @ 15:06H

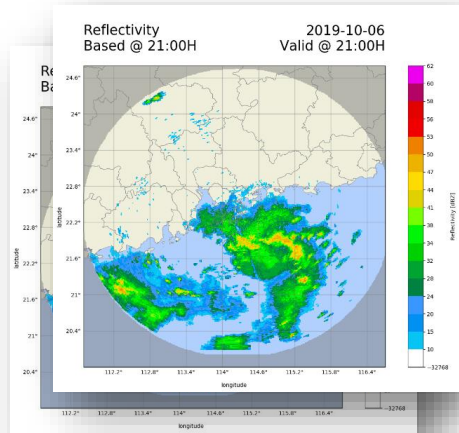


With Our Neural Network

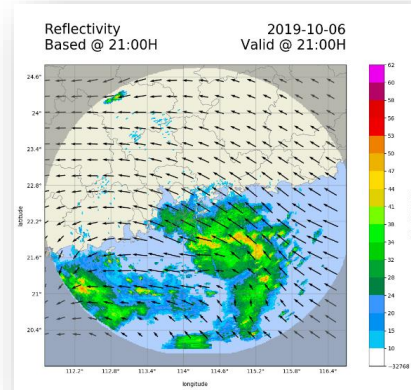
Raining Areas  
Far Away  
Simulated with  
Satellites' Data  
Using Our  
Neural Network



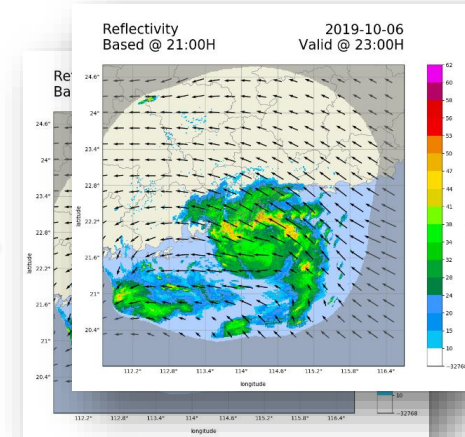
## 2. Nowcast with AI Computer Vision in SWIRLS



Actual  
Raining  
Areas



Calculate  
Speed &  
Direction of  
Raining Areas  
By **AI**  
**Computer Vision**



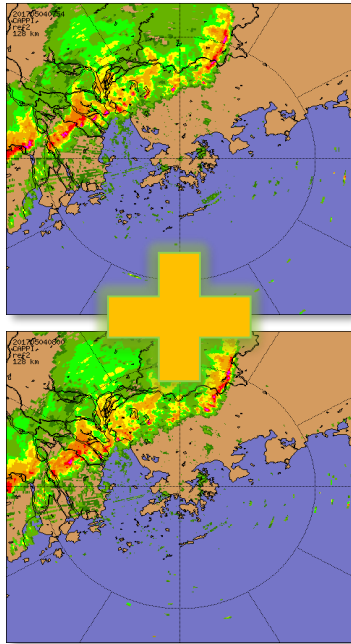
Forecast  
Raining  
Areas  
Using **GPU**



Precisely  
**Where,**  
**When &**  
**How Much**  
Rainfall



# 3. Next-Generation of SWIRLS AI Nowcast Enhanced Accuracy with Deep Learning



Actual Radar  
Images

## Deep Learning Nowcast

### Big Data

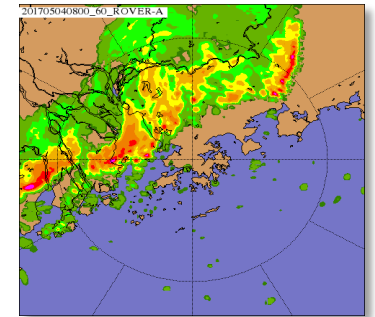
7 Years of Historical Data

140 Billions of Data Points

Trained AI Models to Predict Radar Images

### Edge-Cutting Algorithms:

- Convolutional LSTM (Long-Short-Term-Memory)
- Trajectory GRU (Gated Recurrent Unit)



Forecast  
Radar Images

***Pioneered “Deep Learning Nowcast”***

# Research Papers Related to SWIRLS Nowcast

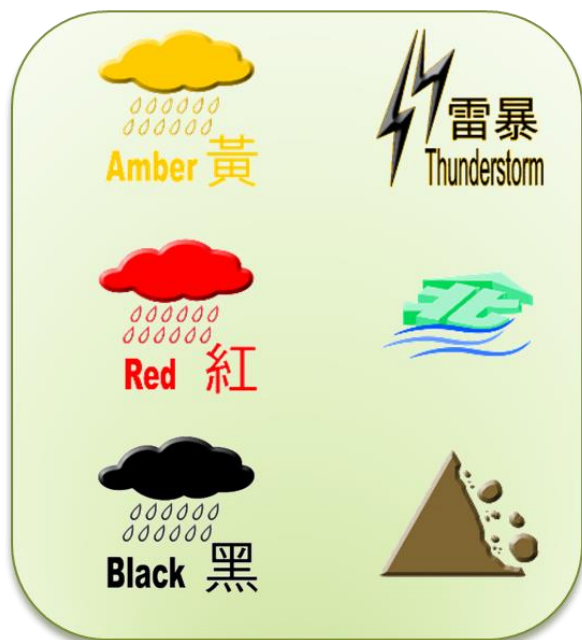
## International / Regional Conference Paper

## Academic Journal Papers

Cited 1,300+ times

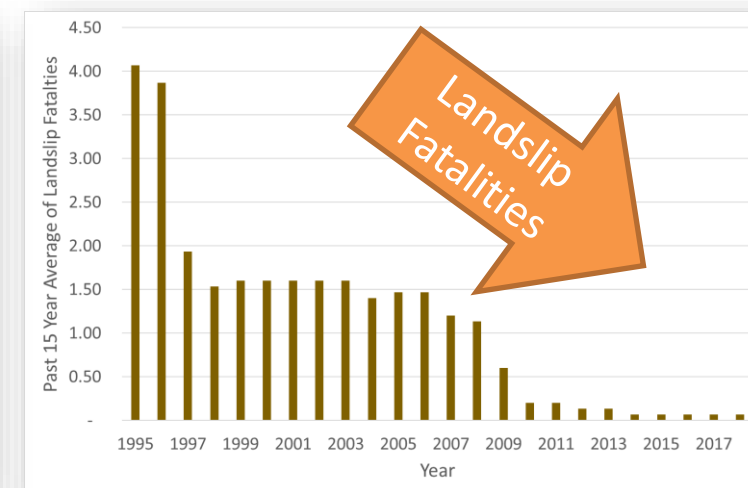
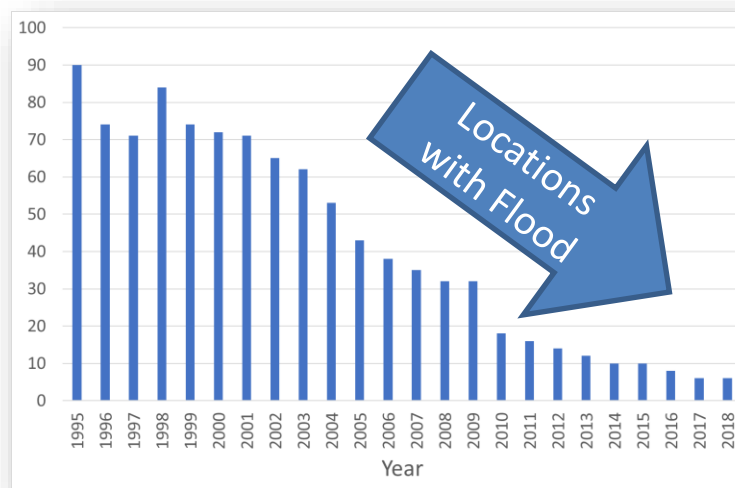


# SWIRLS Nowcast for Weather Forecasters



Guidance on **Rainstorm Warnings** are provided **Half an Hour in Advance**, reducing casualties and damages.

No. of **Floods and Landslides** in Hong Kong decreased over the years





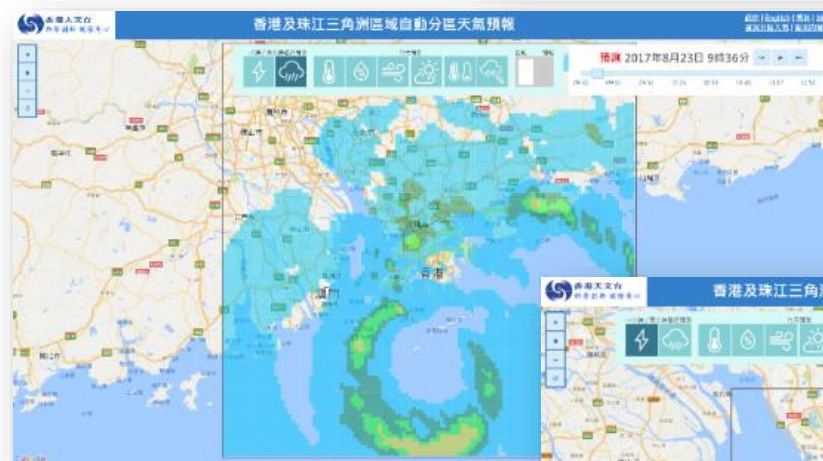
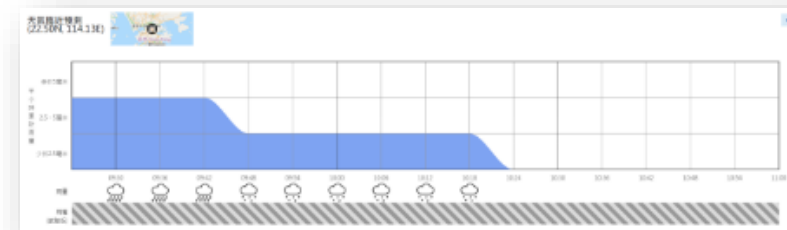
# SWIRLS Nowcast for Citizens

## Rainfall/Lightning Nowcast On Mobile App



Over 1.5 million  
active users

## Rainfall/Lighting Nowcast On Internet Website



Rainfall Nowcast



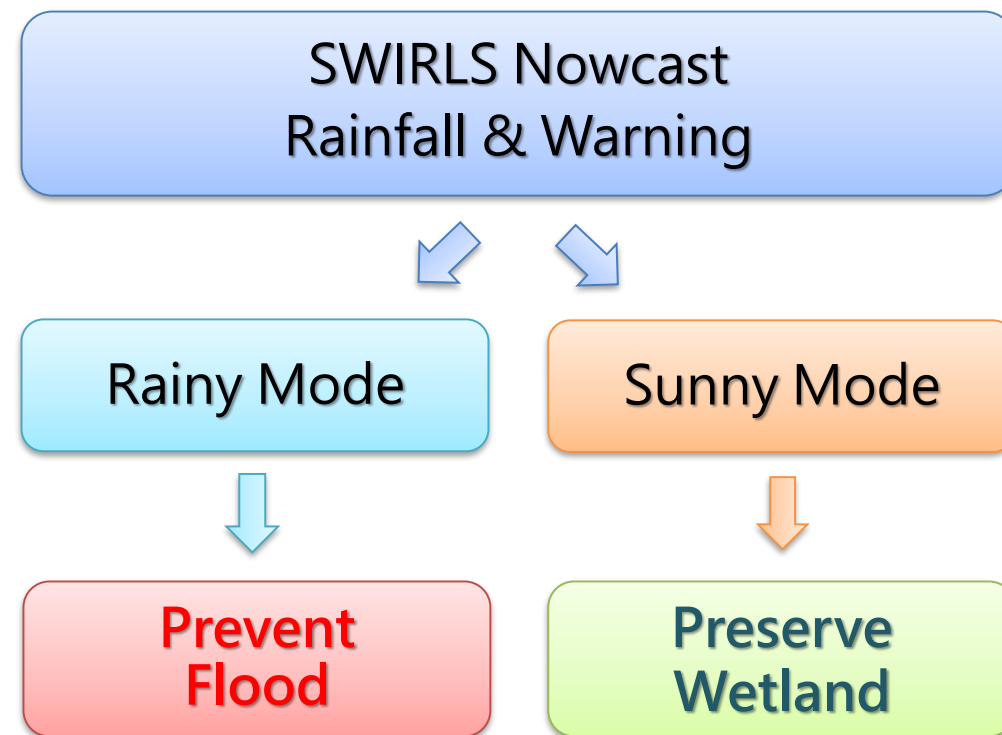
Lightning Nowcast

Over 10 billion visits in 2019

# SWIRLS Rainfall Nowcast for Drainage Services Department



Drainage System in Shuen Wan Wetland



**Greatly Improves Wetland's Ecology**



# SWIRLS Rainfall Nowcast for Drainage Services Department

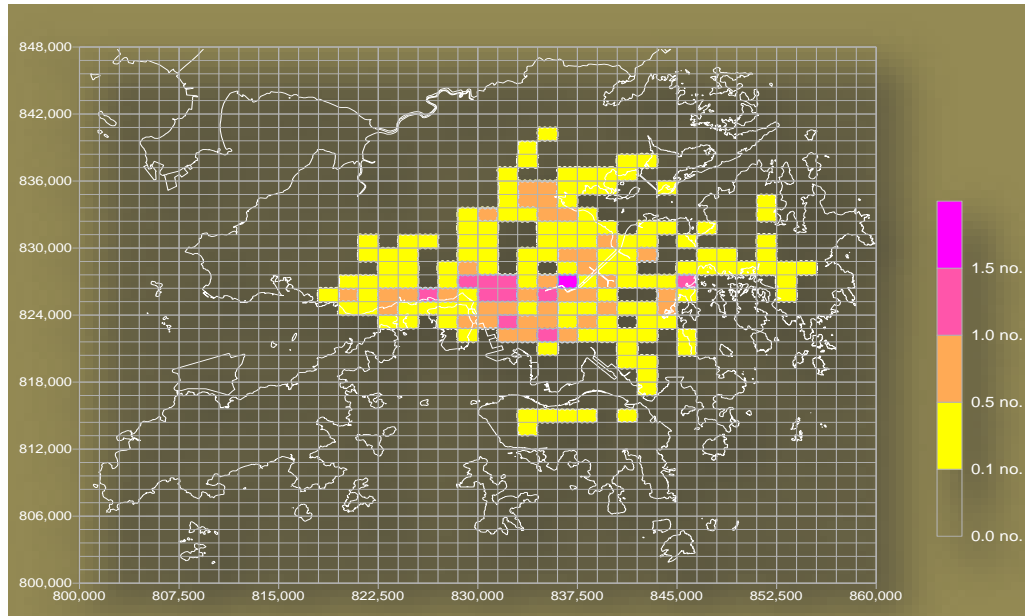


After adopting rainfall nowcast, workers can work until 2 hours before the arrival of rainstorms.

Construction Works inside Kai Tak River

**Enhances Productivity of Workers**

# SWIRLS Rainfall Nowcast for Geotechnical Engineering Office (GEO)



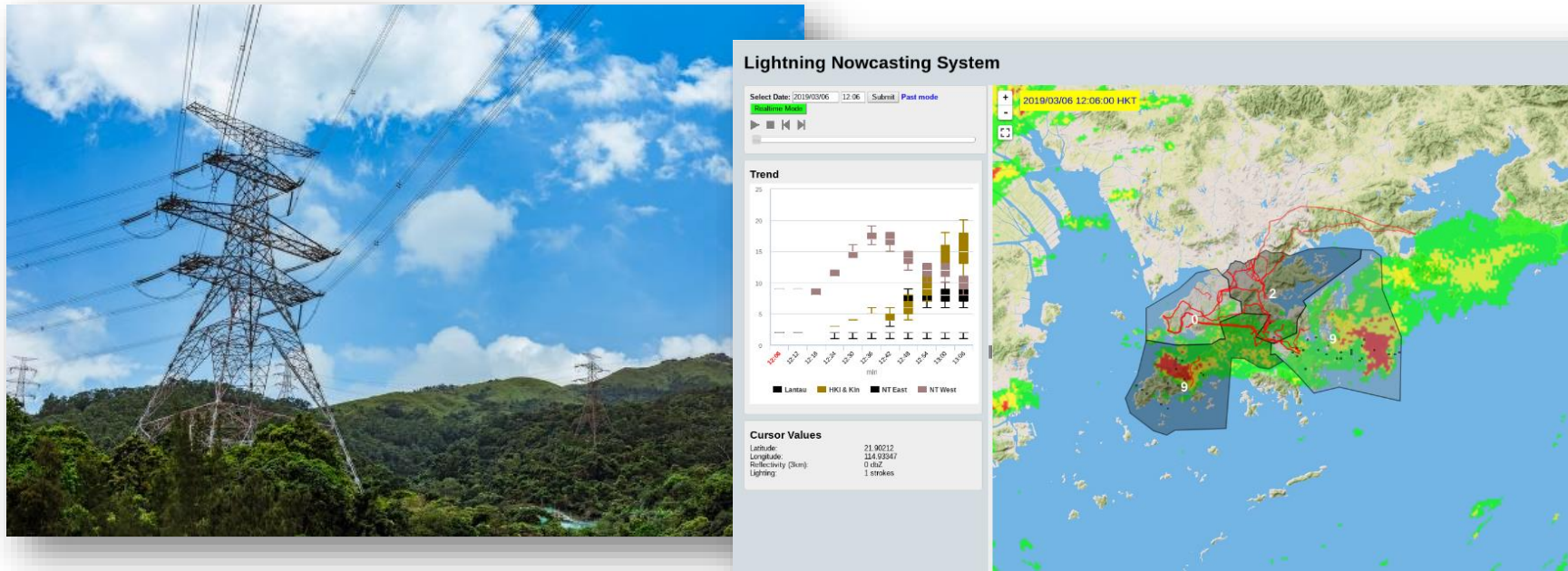
GEO's Landslip Model

Rainfall Nowcast are  
used for modelling  
Number and Locations  
of Landslides

**Supports Decision on Landslip Warning  
Enables Mobilization of Staff in Advance**



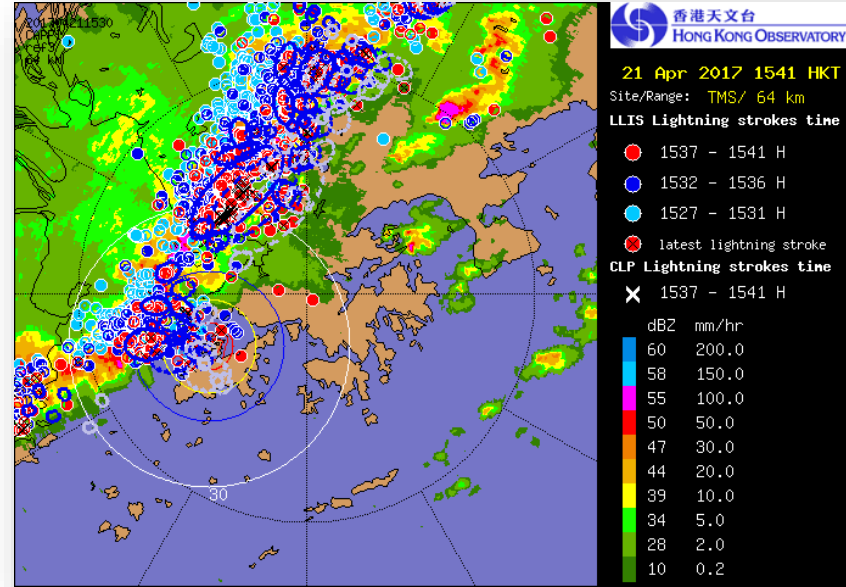
# SWIRLS Lightning Alerts for Power Company



**Power Supply  
Reliability:  
99.999%**

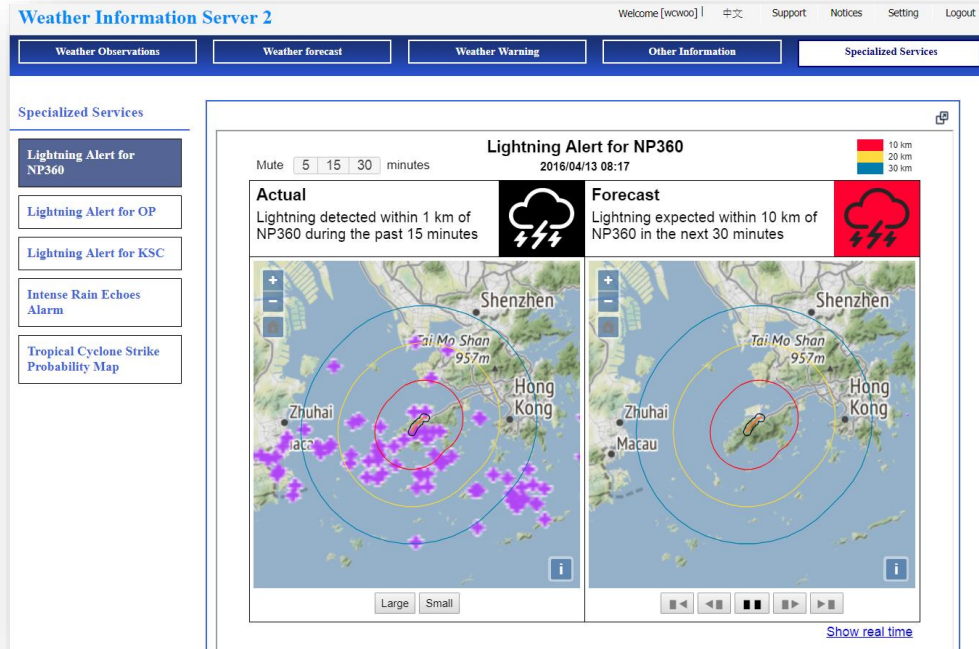
**Supports Decisions on Mobilization  
Enables Rapid Resumption of Power**

# SWIRLS Lightning Alerts for Hong Kong International Airport



Minimizes Risk of Lightning Strikes  
Of Ground Staff in the Airport

# SWIRLS Lightning Alerts for Theme Park, Golf Course & Cable Car

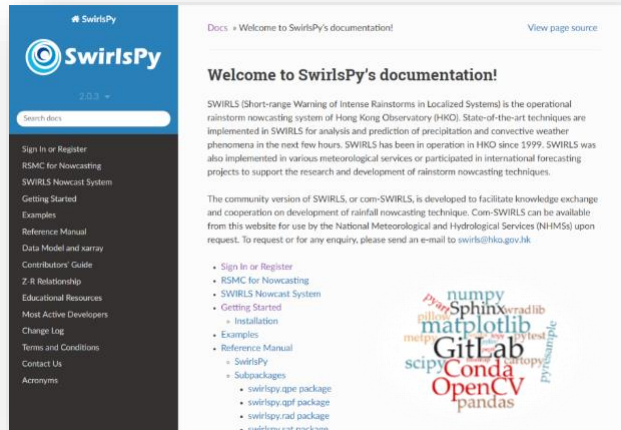


Enable Timely Suspension of Services Before  
Thunderstorms and Early Resumption After



# Technology Transfer for Overseas Weather Services

**Easy to Install,  
Configure and Use**



**Community  
SWIRLS  
Platform  
For Developers**



**Overseas  
Weather  
Services**



**Weather  
Forecasters**



**Citizens**



**Government  
& Utilities**

# Community SWIRLS Nowcast System

## Transferred to 33 Countries/Regions

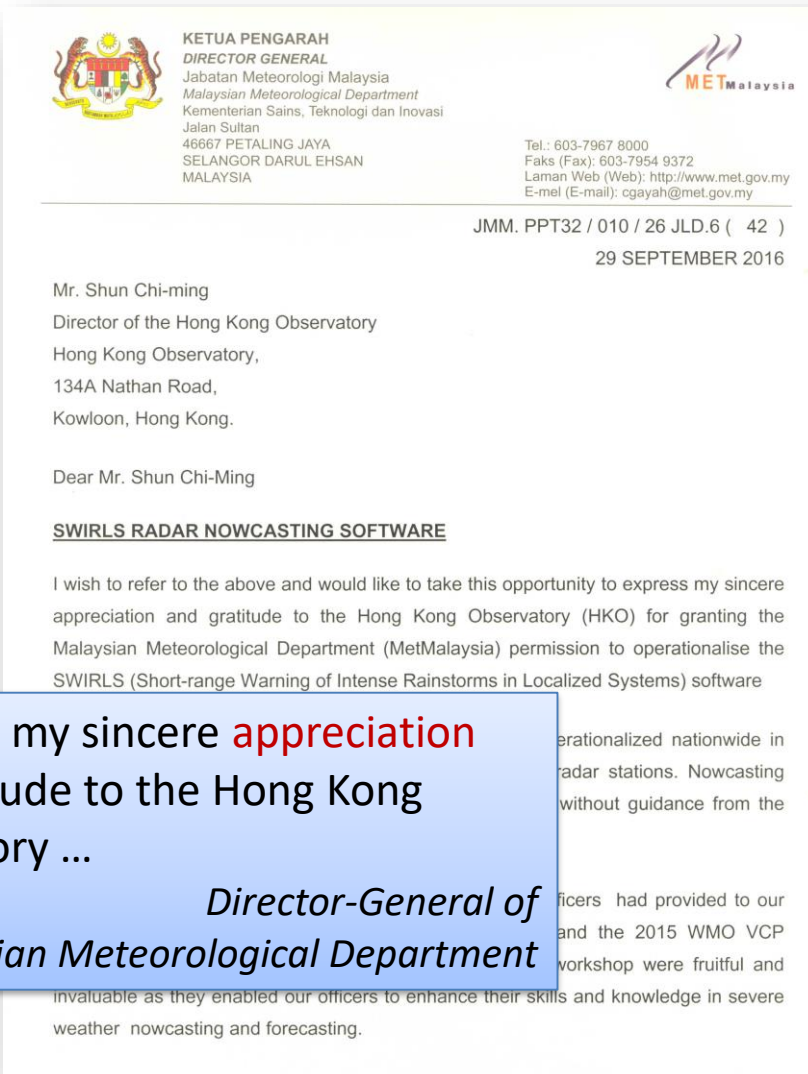


## Asia Pacific

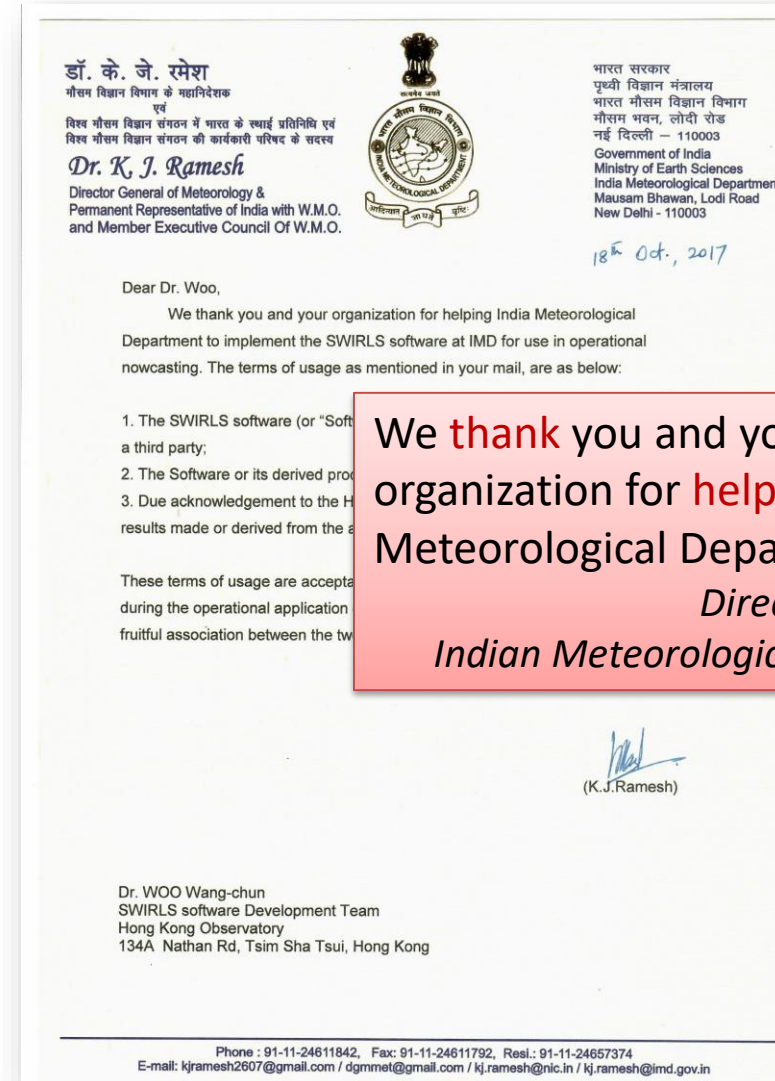
Australia	Macao
Bangladesh	Malaysia
China	Myanmar
Fiji	Philippines
India	Singapore
Indonesia	Sri Lanka
Japan	Thailand
Korea	Vietnam
Laos	

# Appreciation letters from ...

## Malaysia



## India





# Designated as Regional Centre for Nowcasting by World Meteorological Organization (WMO), a United Nations Agency



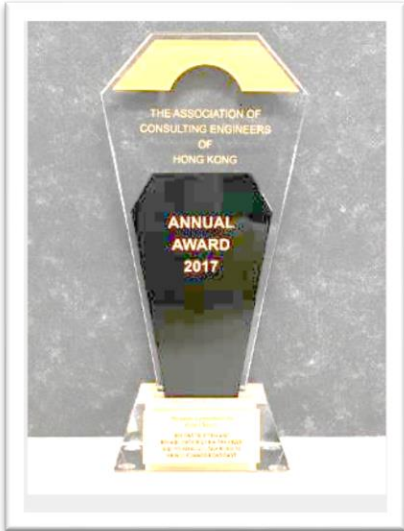
**WMO** is an  
**UN** Organization,  
like WTO, WHO etc.



In Recognition of our **Outstanding Technology**  
and **Contributions**, HKO has been designated by  
UN/WMO as a **Regional Centre** for Nowcasting.  
Opening inaugurated by HKO's Director (left)  
and **Secretary-General of UN/WMO** (right)



# Awards



Annual Award of  
The Association of  
Consulting  
Engineers of Hong  
Kong 2017



Smart Business Grand Award  
Smart Business (Solution for  
Business and Public Sector  
Enterprise) Gold Award  
Hong Kong ICT Awards 2019



Winner Awards at the 19th Asia Pacific Information and  
Communications Technology Alliance (APICTA) Awards  
Public Sector & Government category – Gov't & Citizen Services  
Technology category – Artificial Intelligence  
Ha Long Bay, Nov 2019

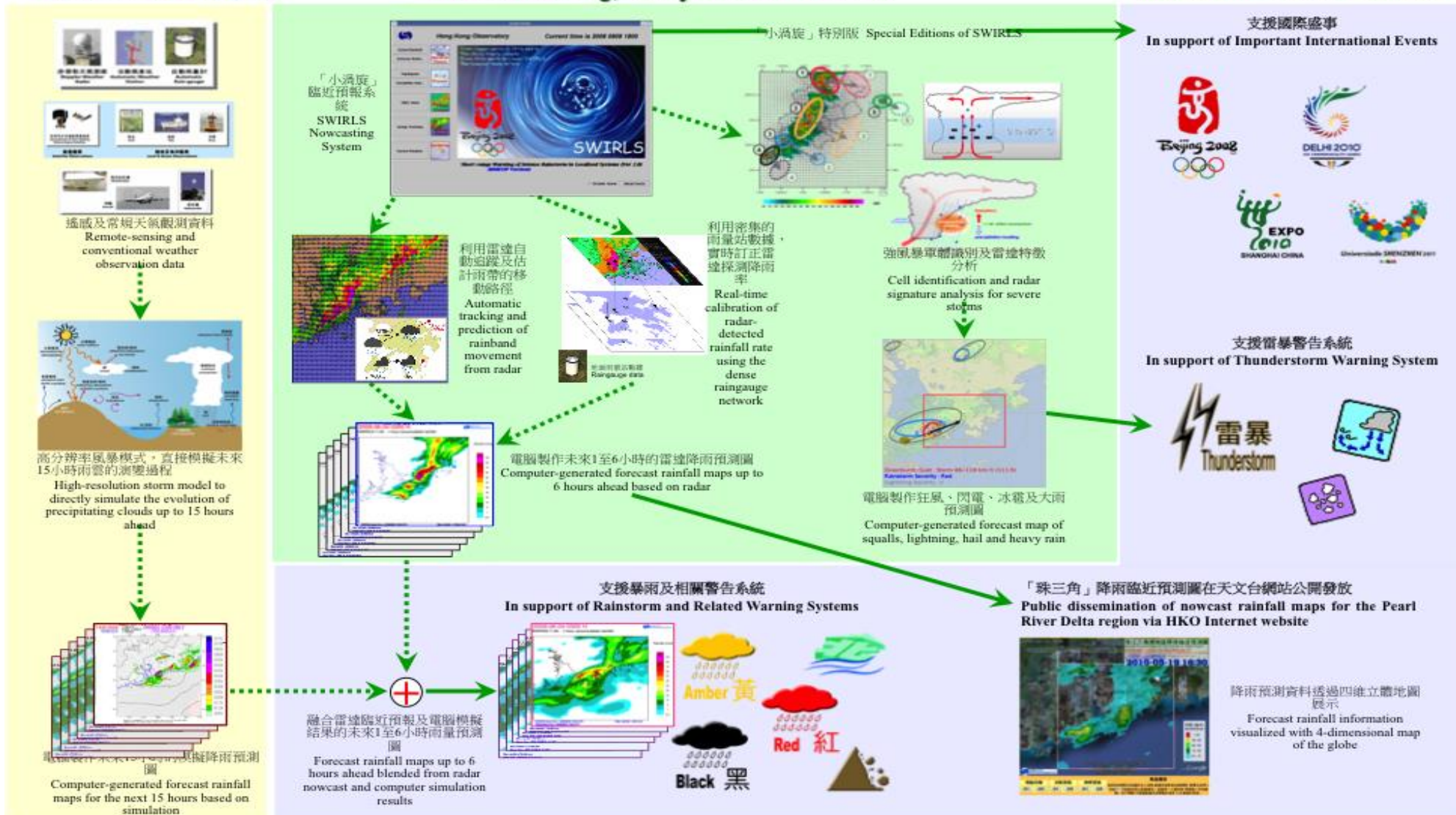
# SWIRLS IN HONG KONG

# SWIRLS – HKO Rainstorm Nowcasting System

電腦模擬大氣物理過程  
Computer Simulation of Physical Processes in the Atmosphere

雷達追蹤、分析及預測  
Radar Tracking, Analysis and Forecast

臨近預報產品及服務  
Nowcast Products & Services





# Warning Signals on Rainstorm, Flooding and Landslide



## Amber Rainstorm Signal

Heavy rain has fallen or is expected to fall generally over Hong Kong, exceeding 30 mm in an hour, and is likely to continue.



## Thunderstorm Warning



## Red Rainstorm Signal

Heavy rain has fallen or is expected to fall generally over Hong Kong, exceeding 50 mm in an hour, and is likely to continue.



## Landslip Warning



## Black Rainstorm Signal

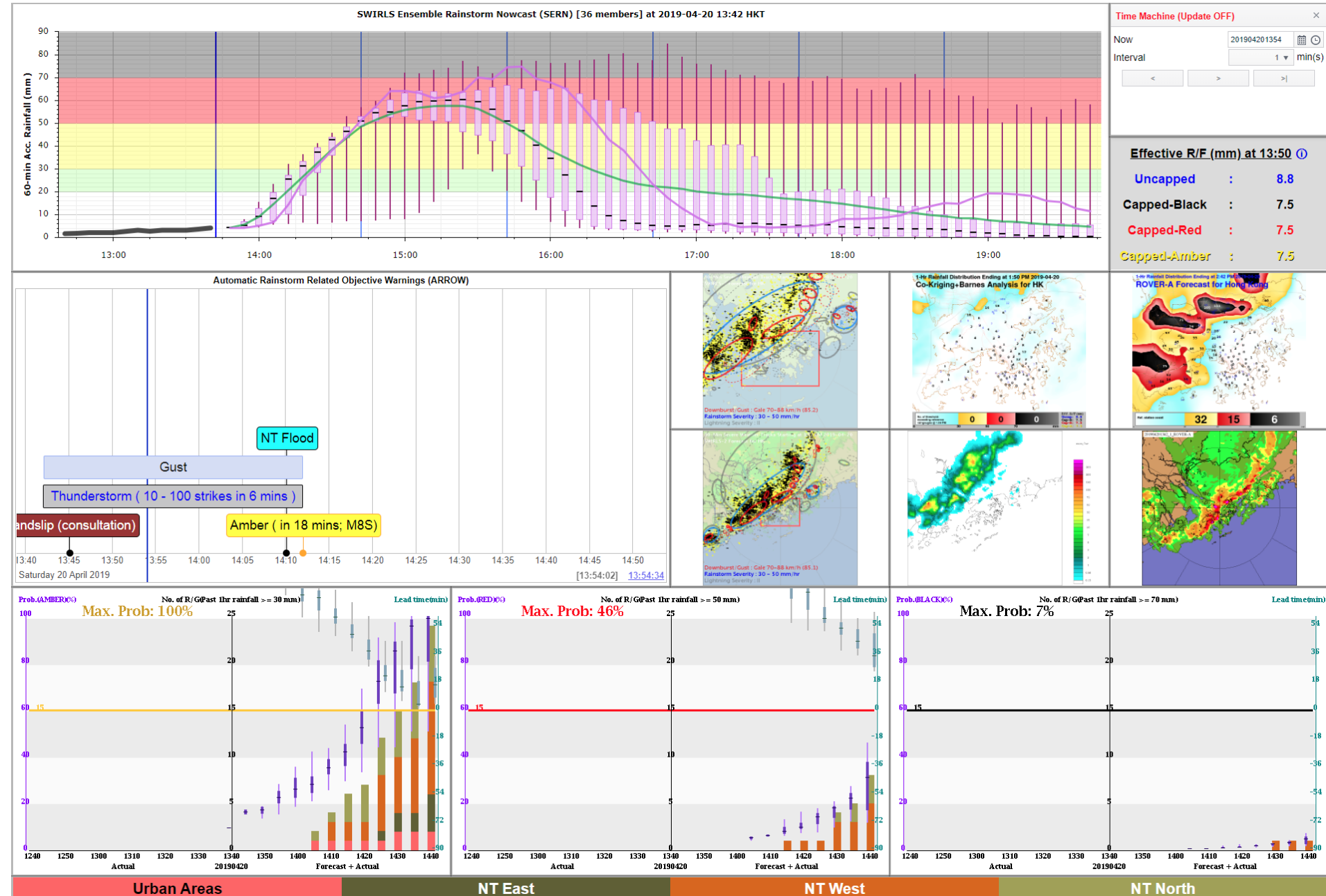
Heavy rain has fallen or is expected to fall generally over Hong Kong, exceeding 70 mm in an hour, and is likely to continue.



## Special Announcement on Flooding in the northern New Territories

# SWIRLS' Core

An integrated support on decision making (alerts, warning, consultation with government departments) during heavy rain situation



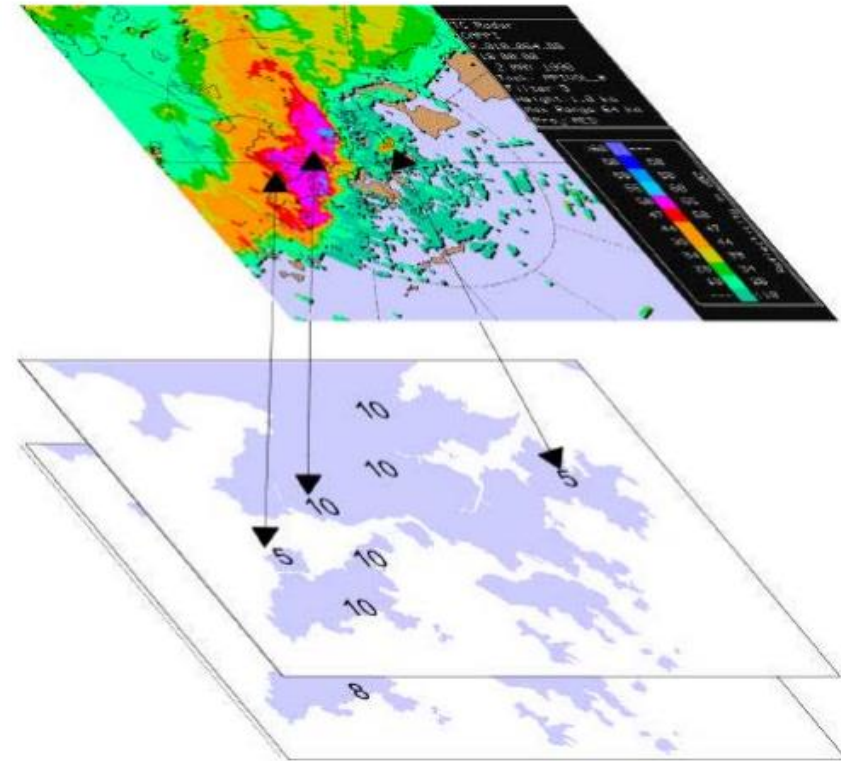
# QPE – Quantitative Precipitation Estimation

- Schematic diagram showing the calibration of radar reflectivity using real-time raingauge measurement.
- Z-R relation for converting reflectivity to rainfall rate

$$Z = aR^b$$

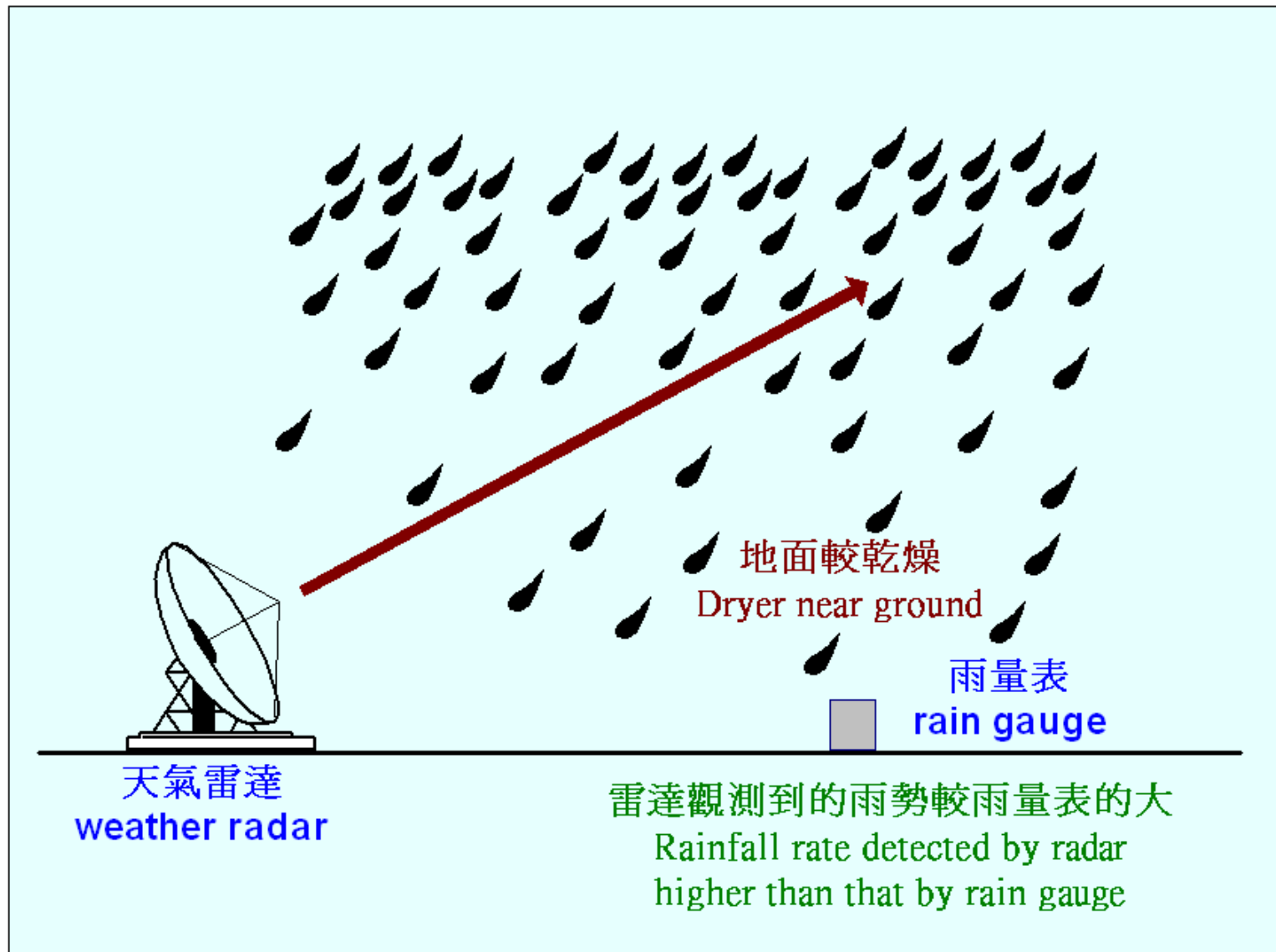
$$dBZ_i = b \text{ dBG}_i + 10\log(a)$$

- Gridded rainfall analysis computed by Barnes analysis or co-kriging algorithm

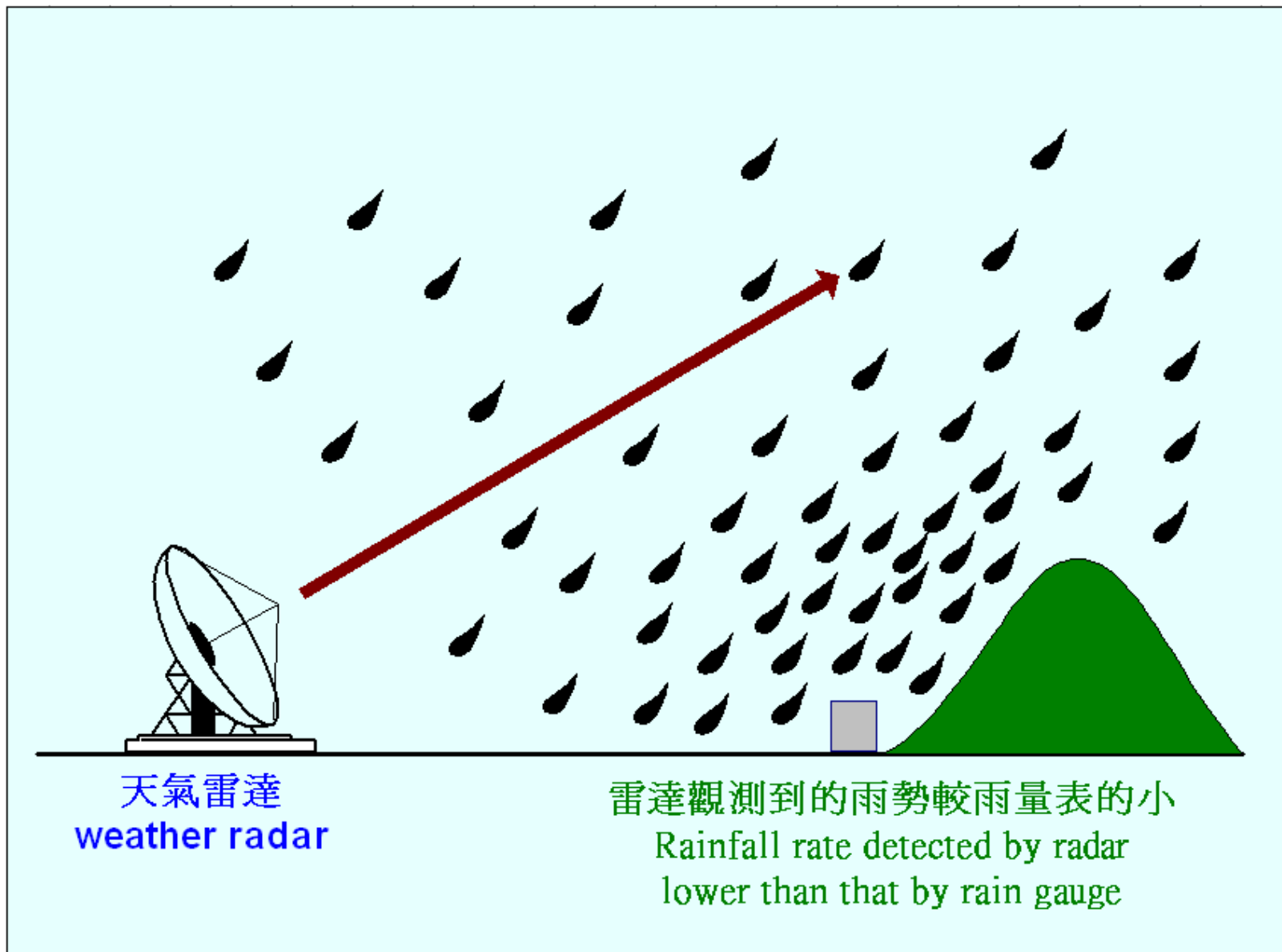




# Limitations of Radar Measurements (1)

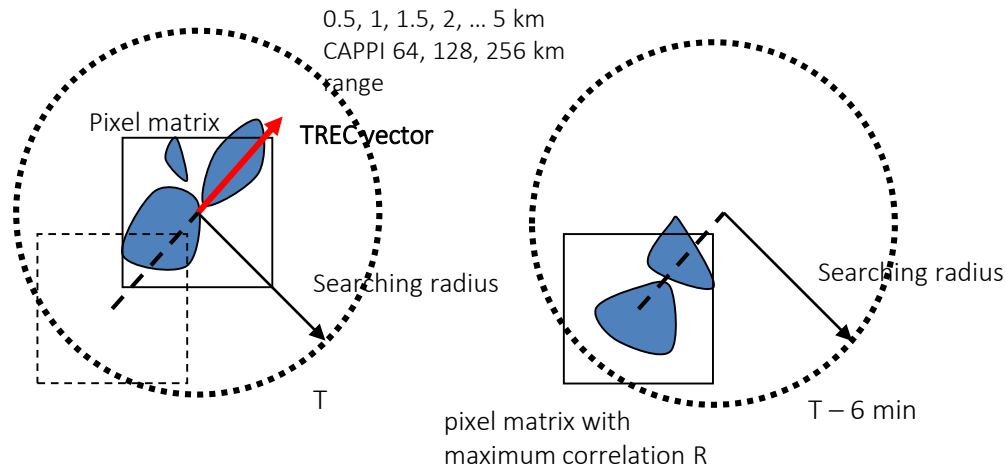


# Limitations of Radar Measurements (1)



# Radar Nowcasting – Echo Tracking

- Maximum Correlation (TREC)



where  $Z_1$  and  $Z_2$  are the reflectivity at  $T+0$  and  $T+6\text{min}$  respectively

$$R = \frac{\sum_k Z_1(k) \times Z_2(k) - \frac{1}{N} \sum_k Z_1(k) \sum_k Z_2(k)}{\left[ \left( \sum_k Z_1^2(k) - N \bar{Z}_1^2 \right) \times \left( \sum_k Z_2^2(k) - N \bar{Z}_2^2 \right) \right]^{1/2}}$$

- Optical Flow

## ROVER – Real-time Optical-flow by Variational method for Echoes of Radar

Given  $I(x,y,t)$  the image brightness at point  $(x,y)$  at time  $t$  and the brightness is constant when pattern moves, the echo motion components  $u(x,y)$  and  $v(x,y)$  can be retrieved via minimization of the cost function:

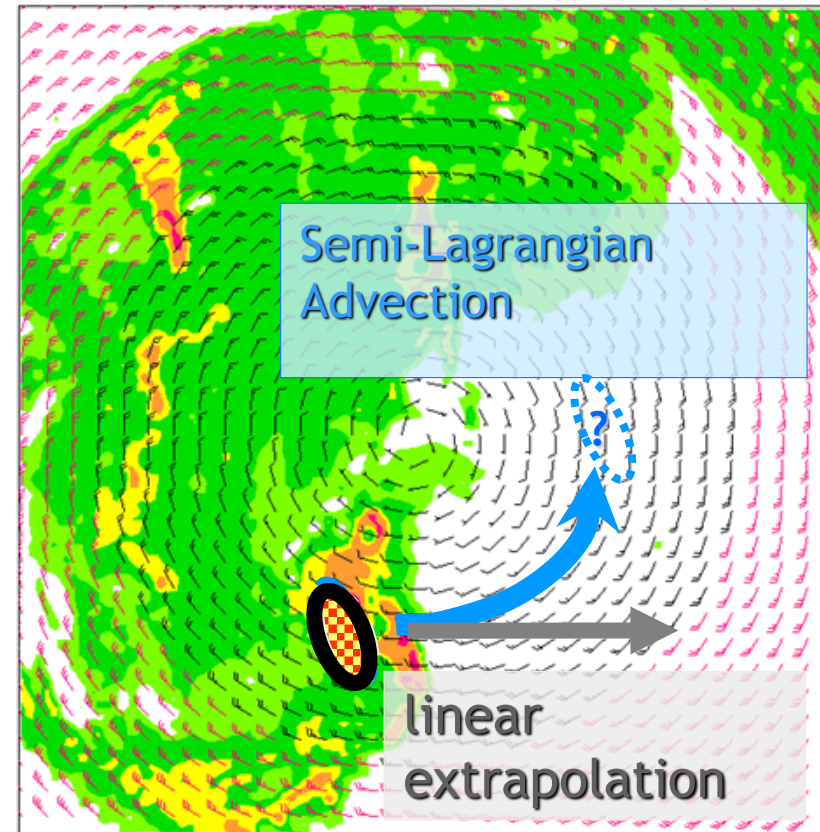
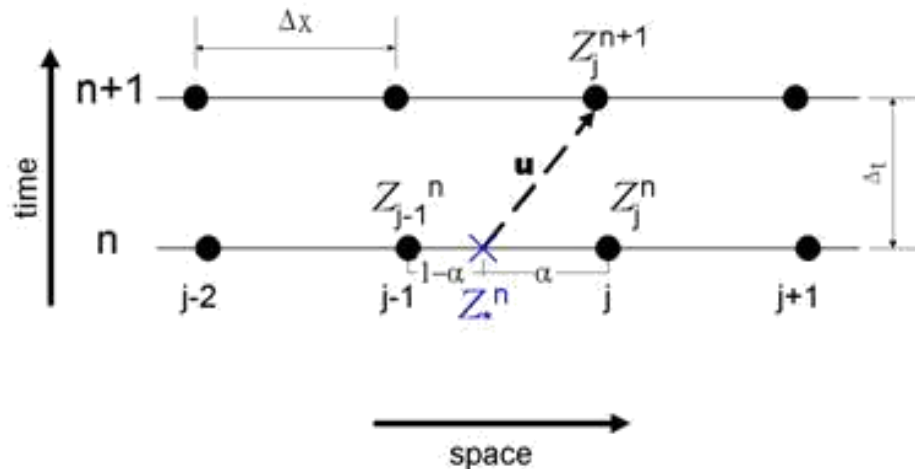
$$J = \iint \left[ \frac{\partial I}{\partial t} + u \frac{\partial I}{\partial x} + v \frac{\partial I}{\partial y} \right]^2 dx dy$$

# QPF - Forecast by Extrapolation

- Semi-Lagrangian Advection

- Robert scheme (3 iterations to find origin point)
- Bi-cubic interpolation
- Flux limiter (local max, min constraint)
- One-way nesting

$$\frac{dZ}{dt} = \frac{\partial Z}{\partial t} + \mathbf{u} \frac{\partial Z}{\partial x} = 0$$





# QPF (Quantitative Precipitation Forecast) For Public Weather Services

HKO Automatic Regional Weather Forecast

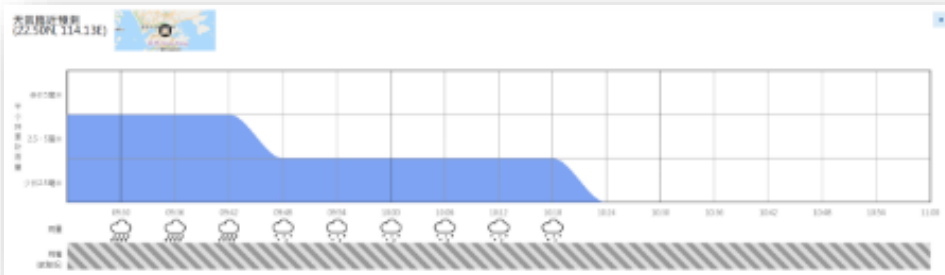
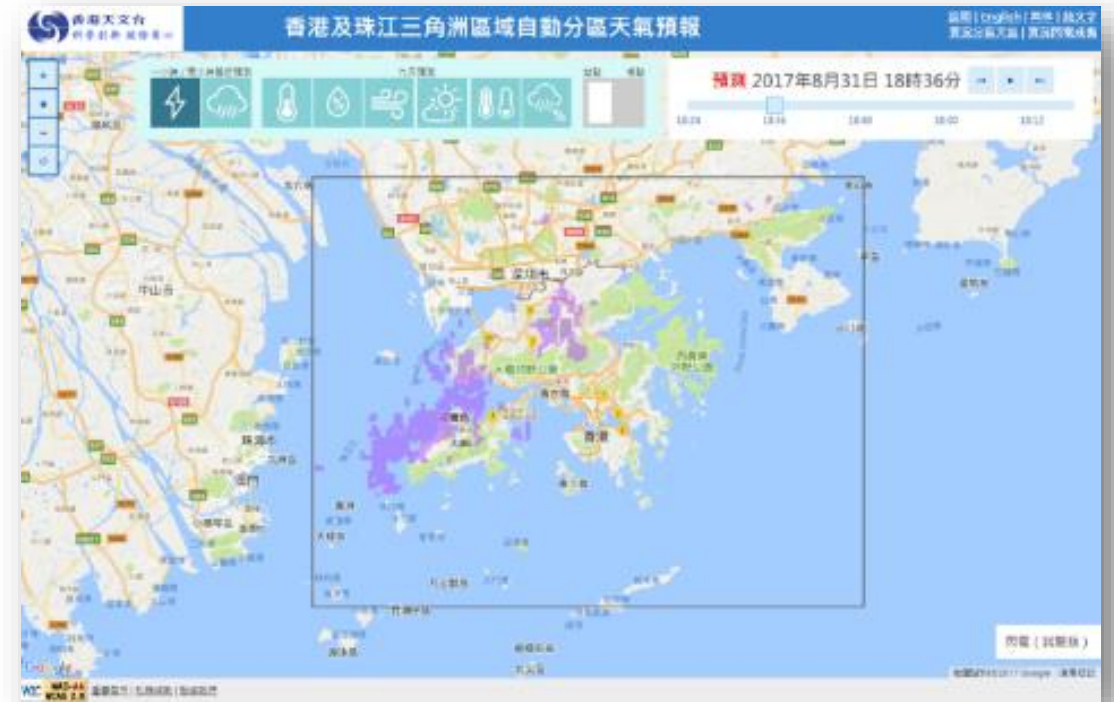
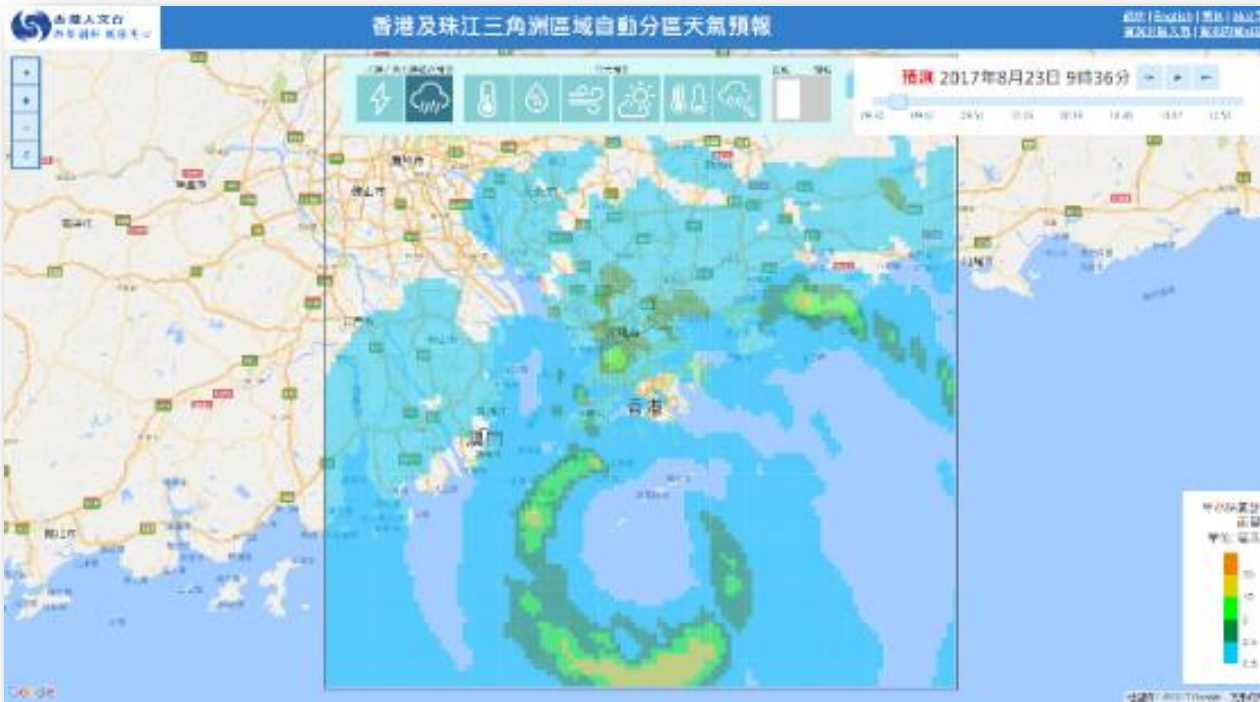
<http://maps.weather.gov.hk/ocf/>

“[MyObservatory](#)” mobile app on iOS and Android



# SWIRLS Nowcast Products to Public

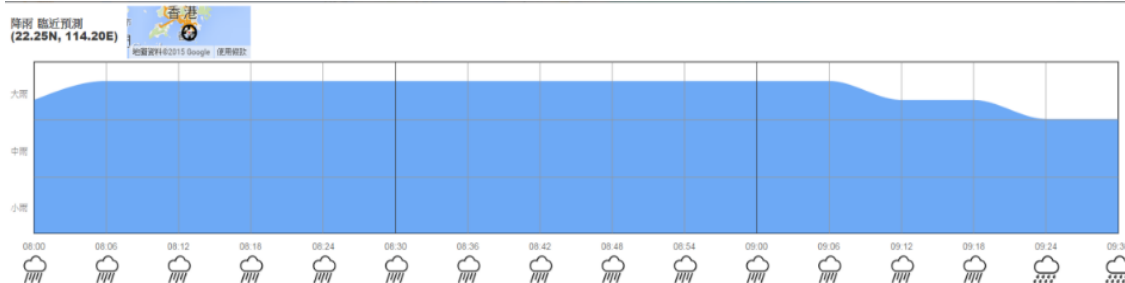
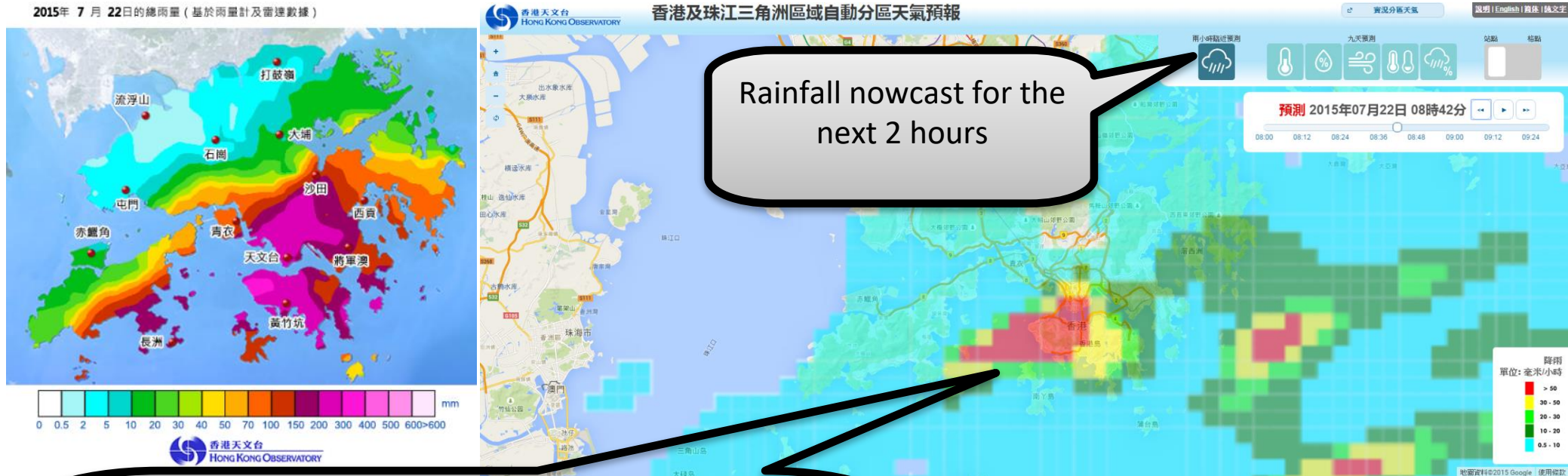
# Rainfall and Lightning Nowcast





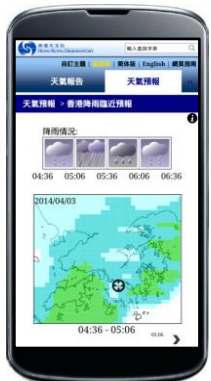
# Seamless Integration of Rainfall Nowcast with Automatic Site-specific Weather Forecast

Rainfall Nowcast in Hong Kong and Pearl River Delta Region in the next 2 hours



2015 年 7 月 22 日，天文台發出黃色暴雨警告接近 12 小時，而山泥傾瀉警告亦生效超過 6 小時。強降雨區在早上一段長時間幾乎停留在香港南部。

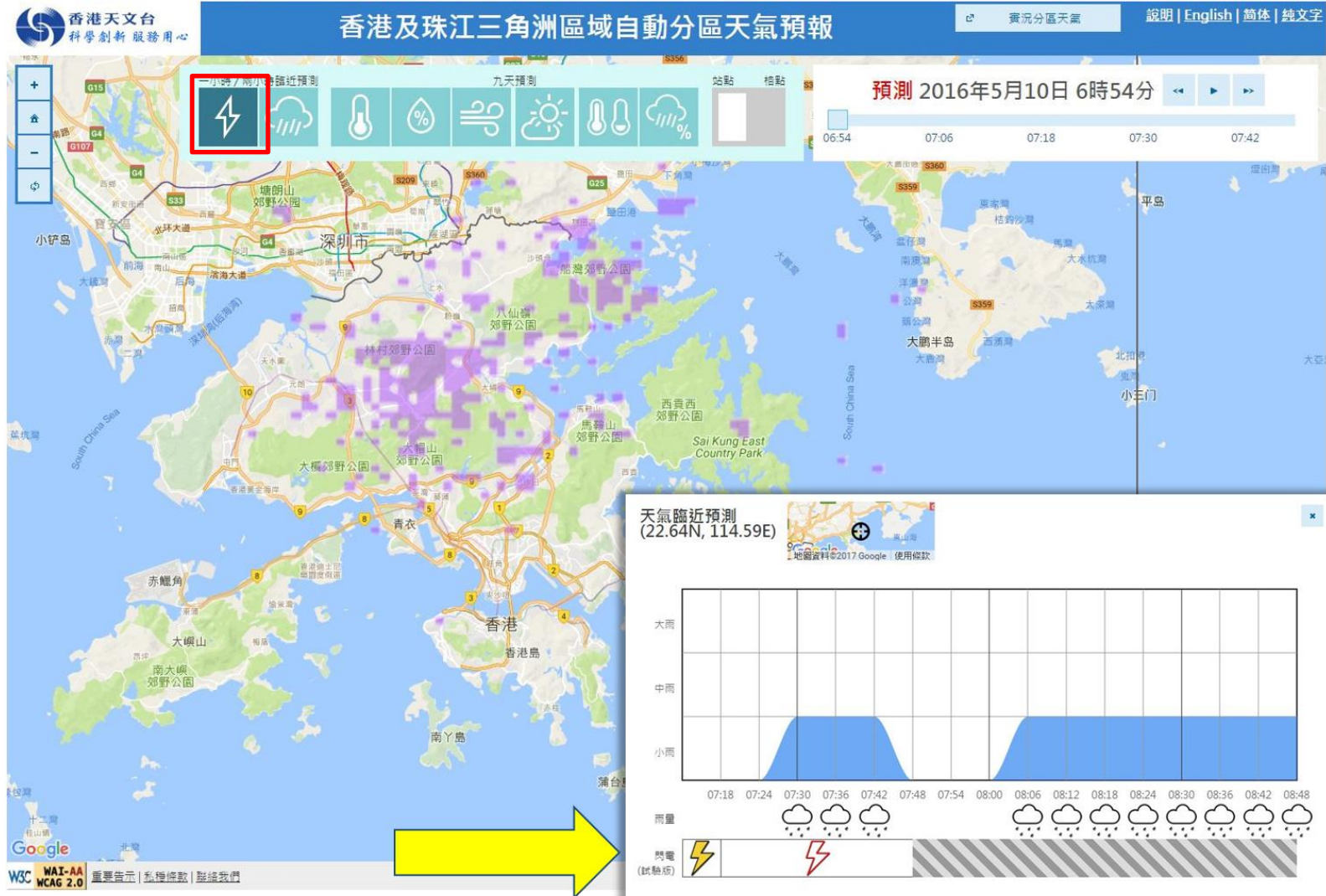
On 22 July 2015, the Amber Rainstorm Warning Signal issued by the Observatory lasted for nearly 12 hours and the Landslip Warning was also in force for more than 6 hours. The area of intense rain remained almost stationary over the southern part of Hong Kong for a prolonged period in the morning









# Lightning Nowcast in Automatic Regional Weather Forecast (ARWF) website

<http://maps.weather.gov.hk/ocf/>



-  Lightning within 10 km in the first 30-minute period
-  Lightning within 15 km in the first 30-minute period
-  Lightning within 10 km in the second 30-minute period
-  Lightning within 15 km in the second 30-minute period

# Location-specific Rainfall and Lightning Nowcasts

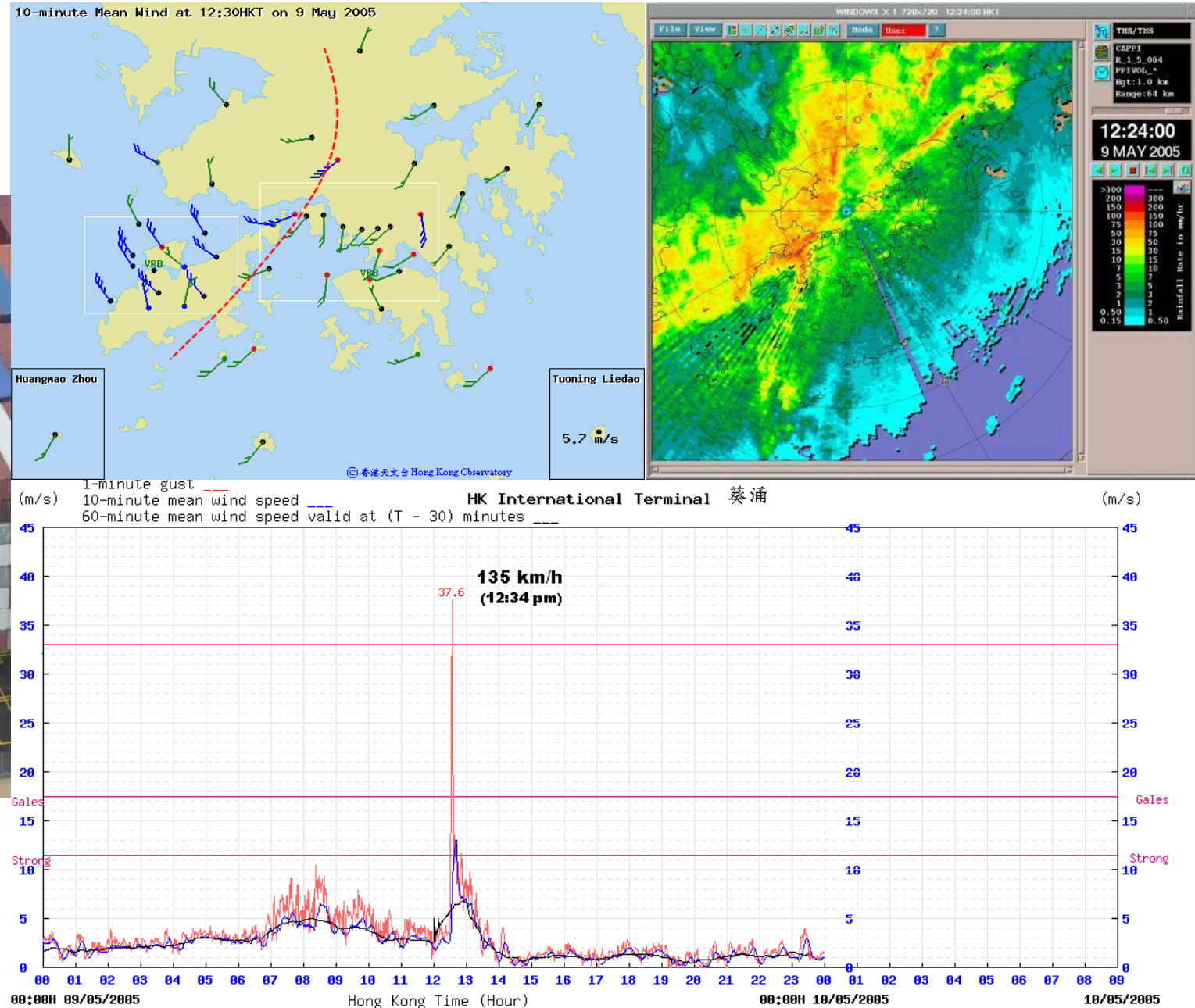


# Severe Weather Nowcast in SWIRLS



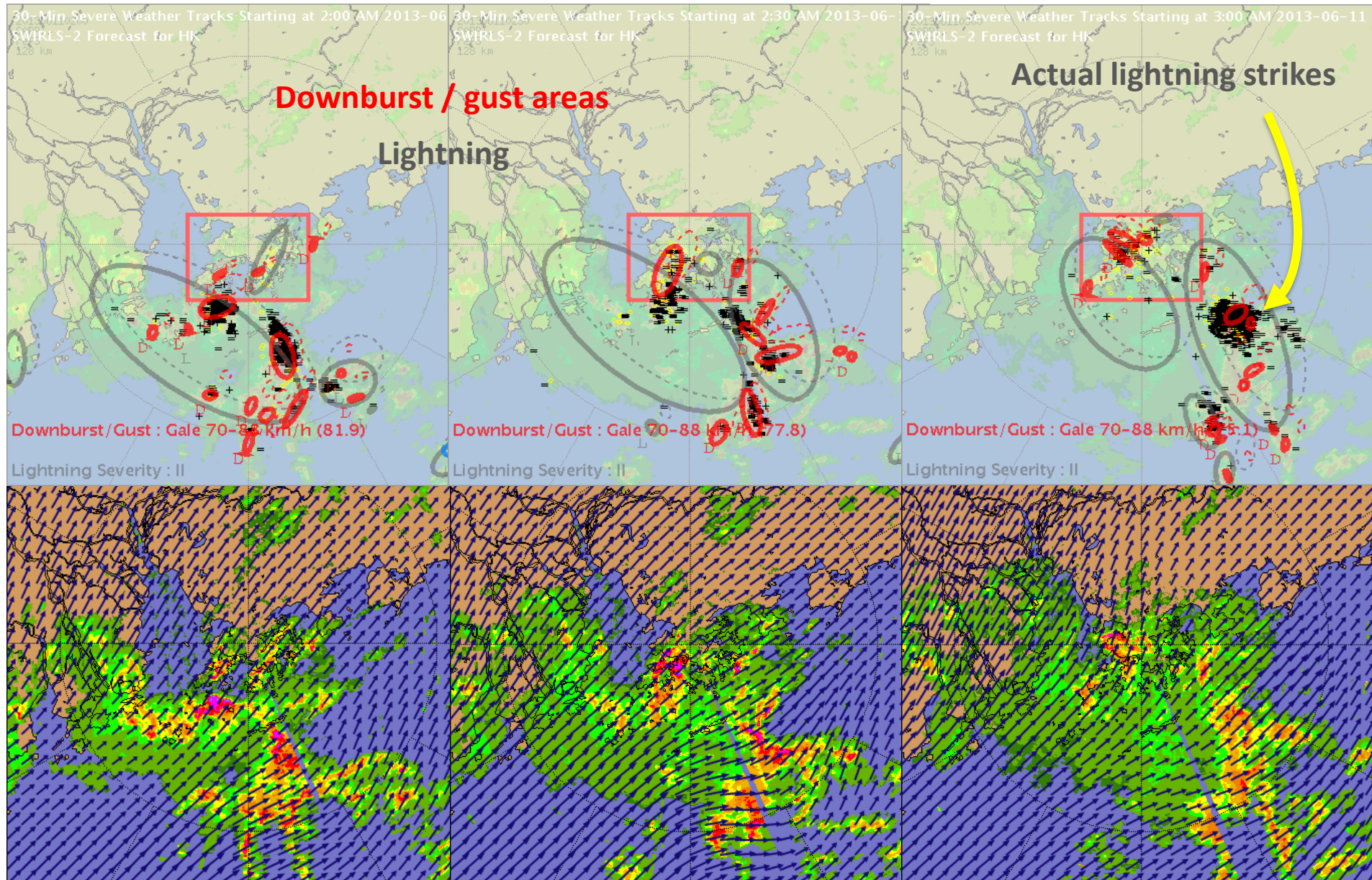


Strong gusts associated with thunderstorms blew down stacks of containers on 9 May 2005



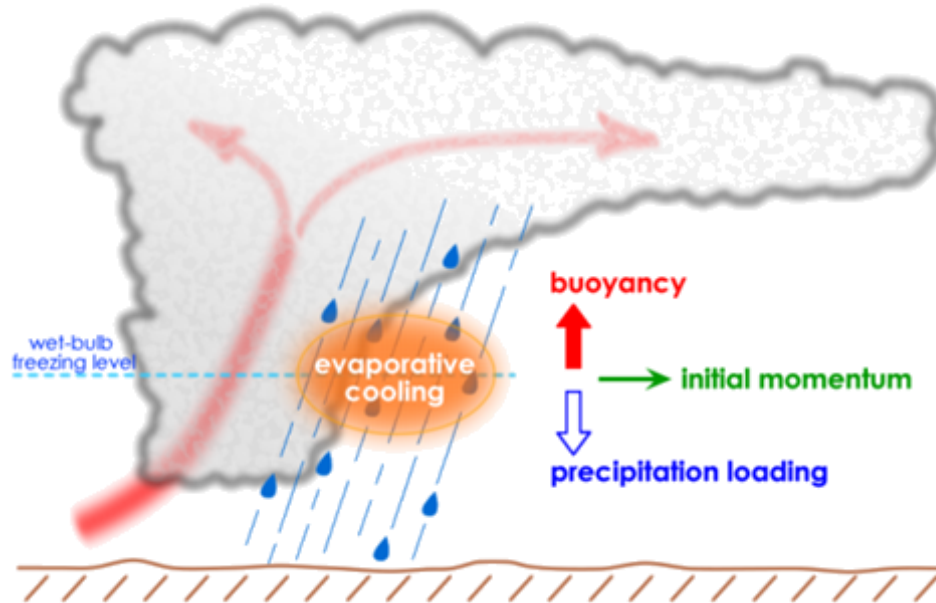


# Severe Weather Nowcast

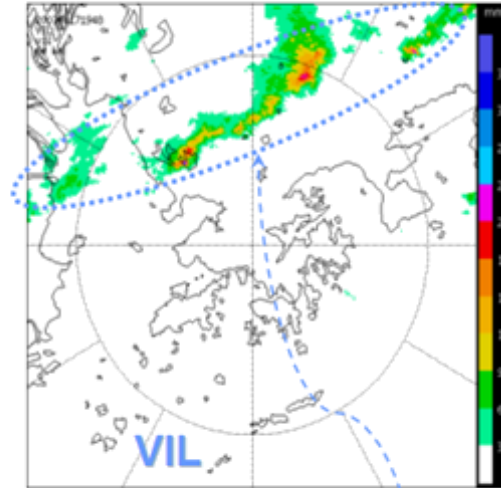


# Conceptual Model of Downburst

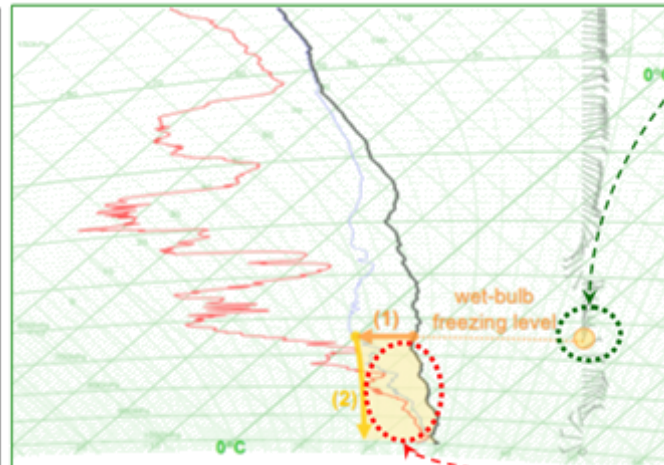
Conceptual model of convective downdraft due to raindrops evaporative cooling of air parcel in the rain shaft



Vertically integrated liquid (VIL) water data derived from radar reflectivity data.



Tephigram with a positive Downdraft Convective Available Potential Energy (DCAPE, the yellowish shaded area). Process (1) isobaric cooling by evaporation of raindrops; Process (2) pseudo-adiabatic descent if favoured.



$$v_s^2 = U_{\text{BUOY}} + v_H^2 + U_{\text{LOAD}}$$

buoyancy

initial mom.

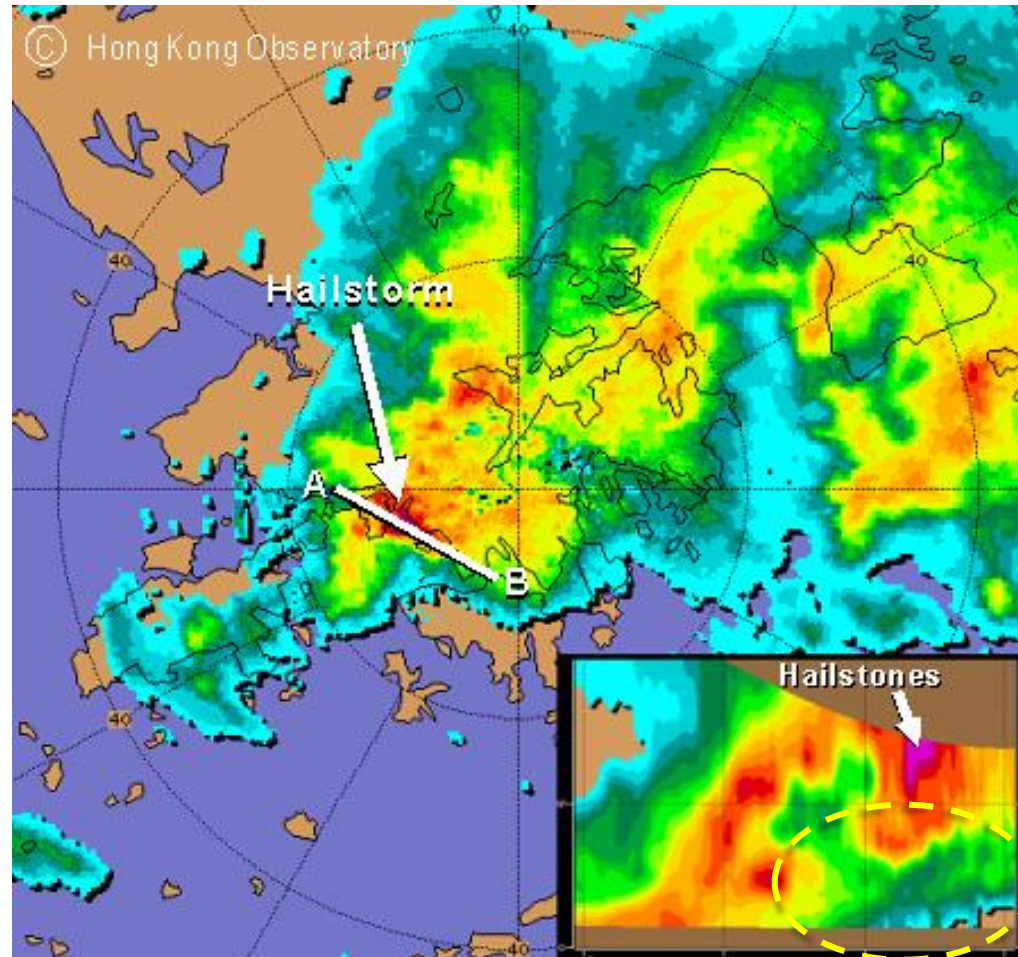
precip. loading

$$U_{\text{BUOY}} = \min \left\{ \int_{p_s}^{p_t} R_d [T_p^{(\text{ambient})} - T_p^{(\text{parcel})}] d \ln p, v_{\text{max}} \right\}$$

$$U_{\text{LOAD}} = -2 \int_0^H L dz = 2g \int_0^H r_\ell dz = 20.3 \times \text{VIL}$$



# Hail

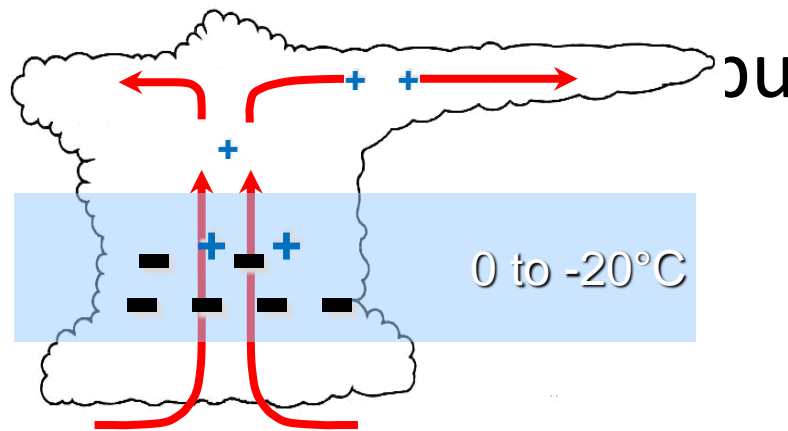


60-dBZ TOPS > 3 km

0-2km VIL < 5 mm

# Lightning Nowcast

- +/- ve charges carried by ice and graupel respectively



(a) charge separated vertically by updraft

(b) physics reflected in radar observations:

Table II – Summary of the conceptual model for lightning initiation.

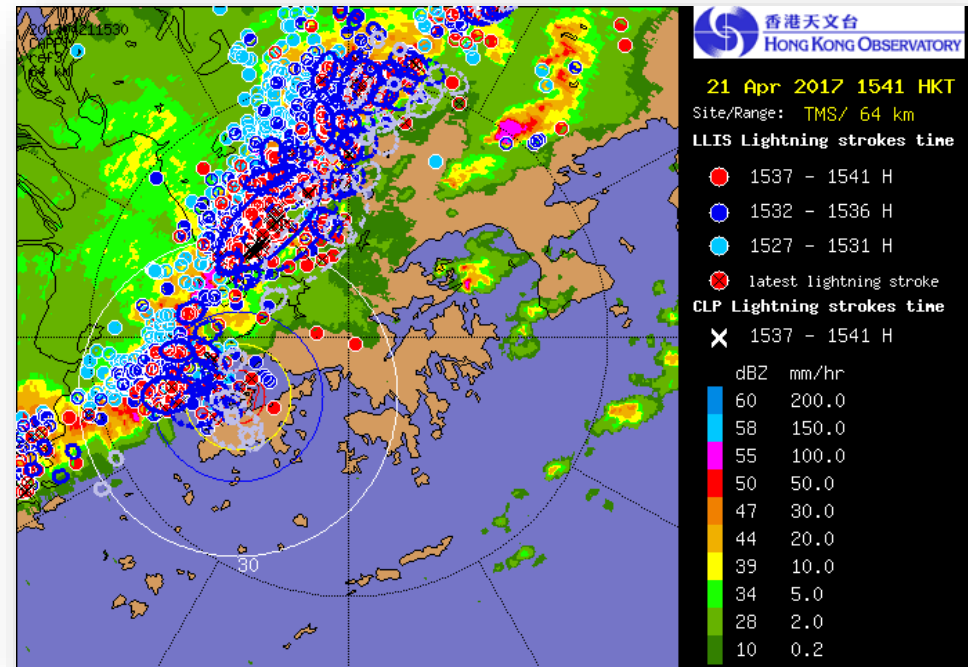
Isothermal Layers	(i) Shallow Cu			(ii) Towering Cu			(iii) mature Cb			(iv) decaying Cb		
	D	H	E	D	H	E	D	H	E	D	H	E
below -40°C							↑	*	ρ	↑	*	ρ
-20 to -40°C				↑	*	ρ	↑	*	ρ	↑	*	
-10 to -20°C	↑	*		↑↓	*△	σ	↑↓	*△	σ		*	
0 to -10°C	↑	⬇		↑	*⬇		↑↓	*△⬇	σ		*	
above 0°C	↑	⬇		↑	⬇		↑↓	△⬇	σ	↓	△	
near surface	↔←			↔←	⬇		↔→	⬇	⬇	↔→	⬇	⬇

Note : Headings D, H and E stand for vertical dynamics, hydrometeors and electric charges respectively. Other symbols are explained in the main text of Section 2.

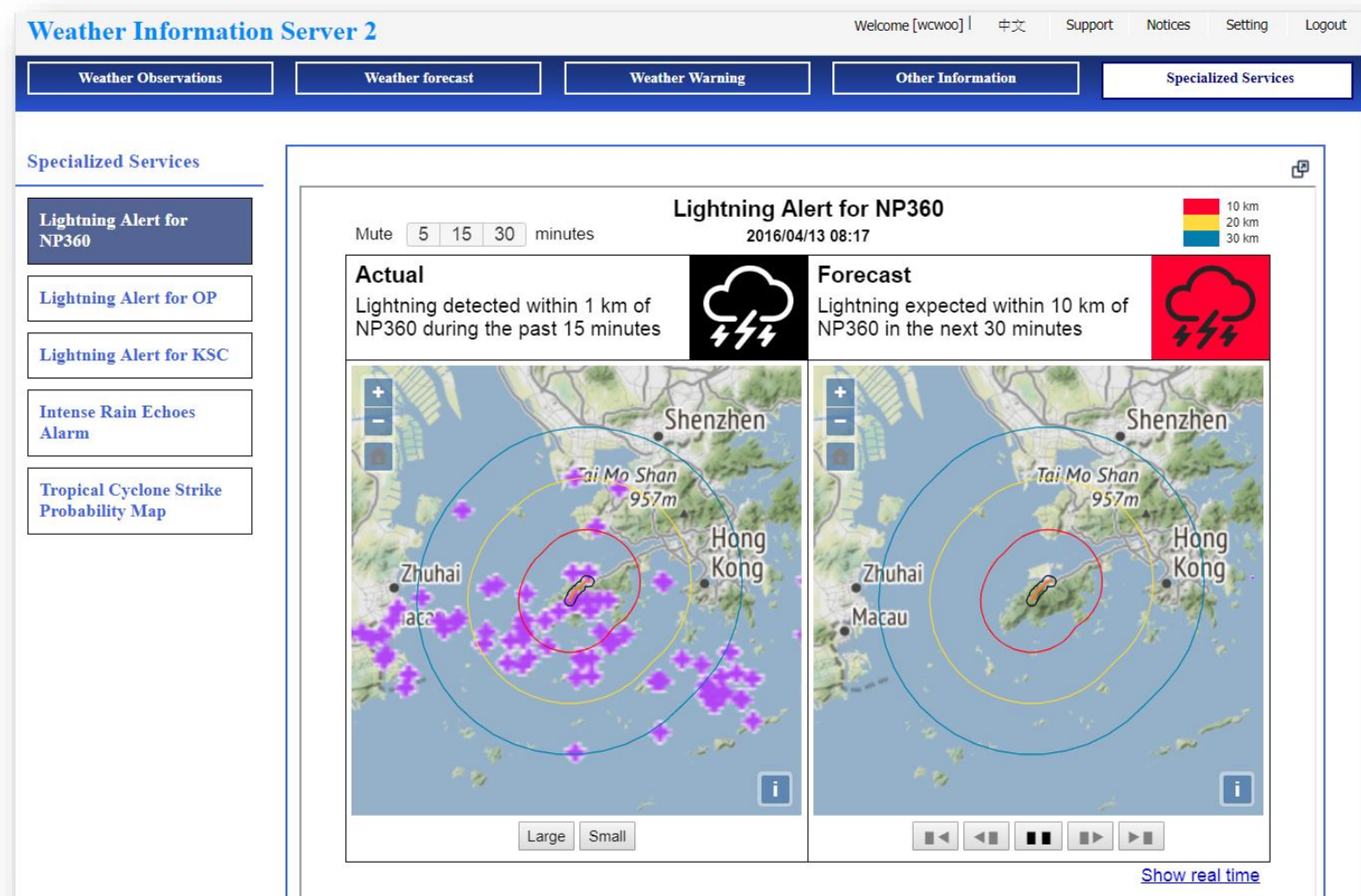


# SWIRLS Lightning Alerts for Hong Kong International Airport

Minimizes Risk of Lightning Strikes  
to ground staff in the Airport



# Lightning Alert and Nowcasting Services for Public Facilities



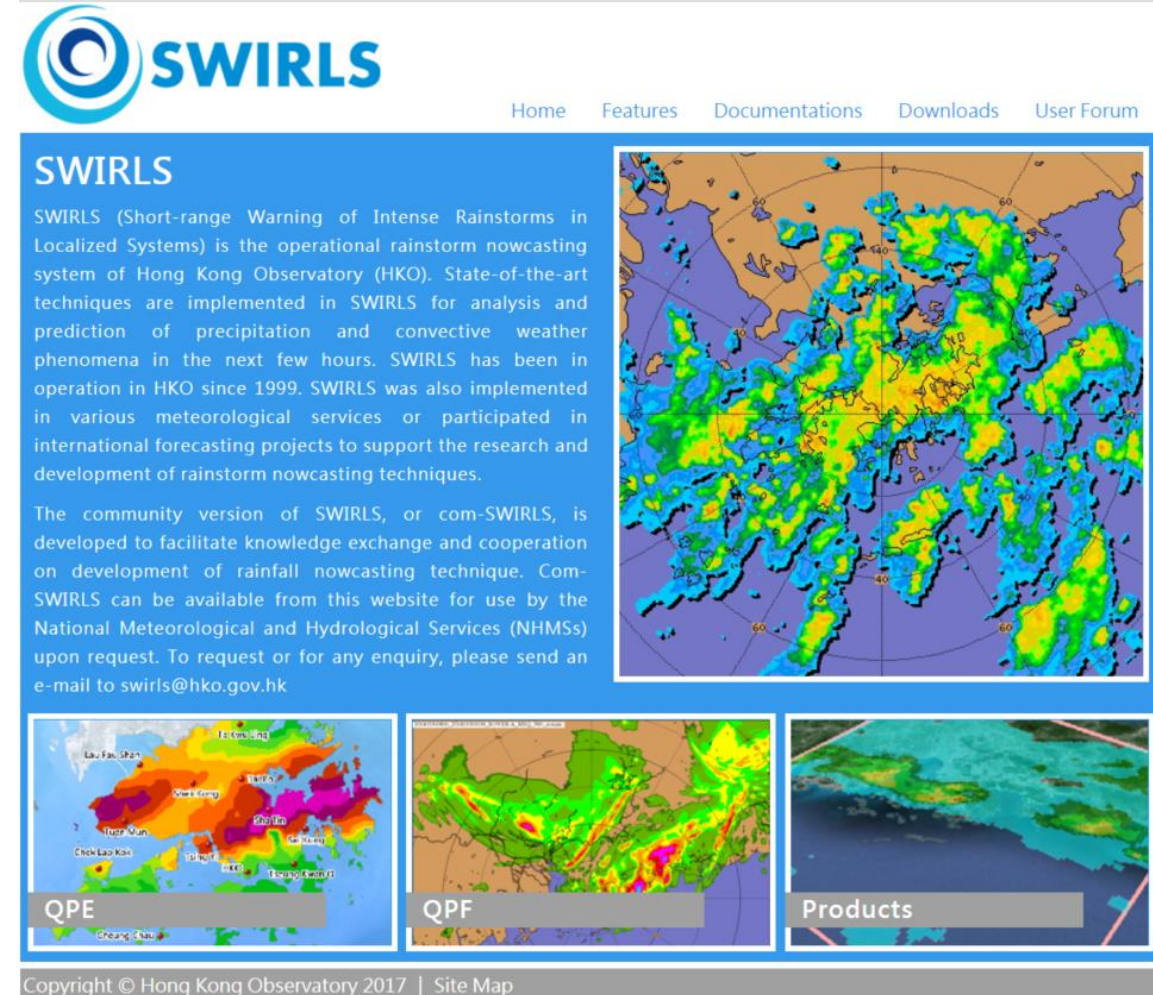
# COMMUNITY SWIRLS (COM-SWIRLS)



# Com-SWIRLS

Website: <http://swirls.hko.gov.hk>

- Capacity building of rainfall nowcasting in NMHSs, promote knowledge exchange and collaboration on research development of rainfall nowcasting techniques
- Com-SWIRLS is developed based on the operational version of SWIRLS, featuring:
  - Quantitative precipitation estimate (QPE), quantitative precipitation forecast (QPF) and graphics utilities to generate rainfall / reflectivity nowcast products
  - Portable code running on common Linux distributions
  - Configurable for implementation in different forecast domains
  - Modular design for easier code changes, and integrate new modules from community users



The screenshot shows the SWIRLS website interface. At the top is the SWIRLS logo and a navigation bar with links: Home, Features, Documentations, Downloads, and User Forum. The main content area has a blue header with the SWIRLS logo and a large map of Hong Kong showing rainfall intensity. Below the header, there is a text block describing SWIRLS as an operational rainstorm nowcasting system of HKO, followed by a paragraph about the community version (com-SWIRLS). At the bottom, there are three smaller maps labeled QPE, QPF, and Products, each showing different types of rainfall data.

**SWIRLS**

SWIRLS (Short-range Warning of Intense Rainstorms in Localized Systems) is the operational rainstorm nowcasting system of Hong Kong Observatory (HKO). State-of-the-art techniques are implemented in SWIRLS for analysis and prediction of precipitation and convective weather phenomena in the next few hours. SWIRLS has been in operation in HKO since 1999. SWIRLS was also implemented in various meteorological services or participated in international forecasting projects to support the research and development of rainstorm nowcasting techniques.

The community version of SWIRLS, or com-SWIRLS, is developed to facilitate knowledge exchange and cooperation on development of rainfall nowcasting technique. Com-SWIRLS can be available from this website for use by the National Meteorological and Hydrological Services (NMHSs) upon request. To request or for any enquiry, please send an e-mail to [swirls@hko.gov.hk](mailto:swirls@hko.gov.hk)

**QPE** **QPF** **Products**

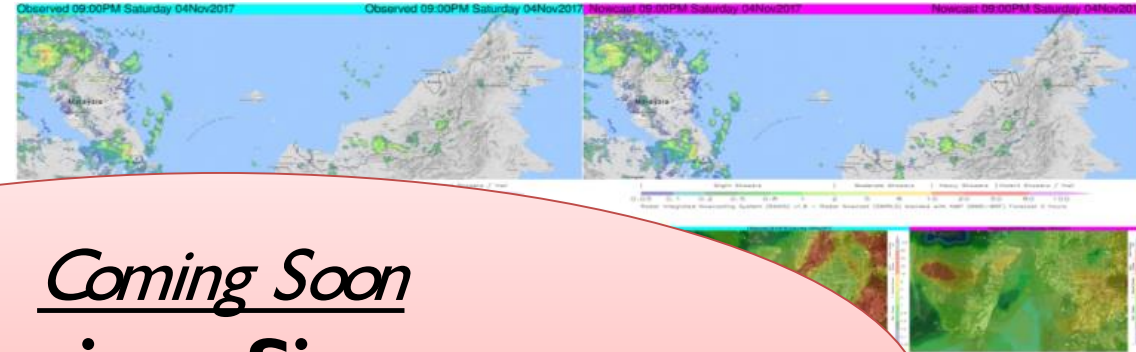
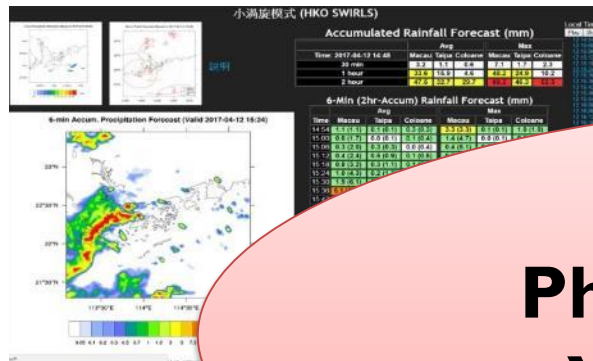
Copyright © Hong Kong Observatory 2017 | Site Map



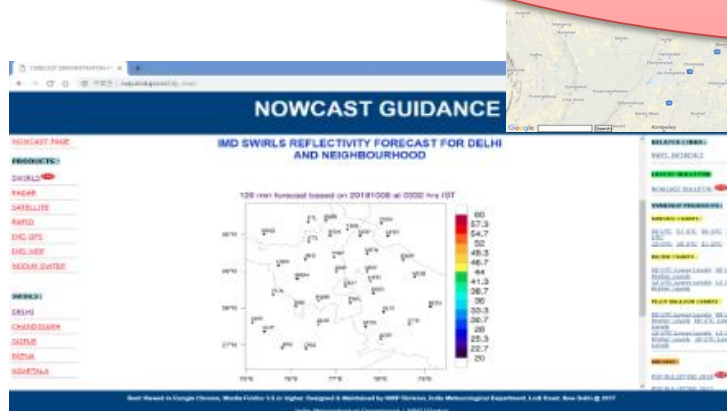
# Community SWIRLS Overseas

Malaysia

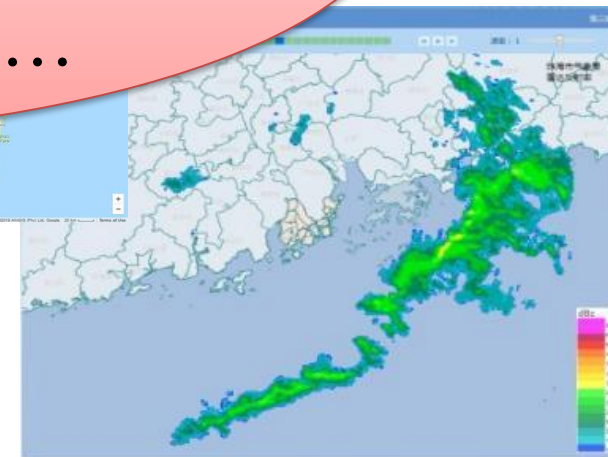
Macao



Coming Soon  
Philippines, Singapore,  
Vietnam, Myanmar,  
Thailand, Indonesia ...



South Africa



India

Zhuhai

# New Com-SWIRLS Collaborative Platform



The screenshot shows the SwirlsPy documentation website. The header includes the SwirlsPy logo and a search bar. The main content area welcomes users to the documentation and provides a list of links for navigation, including 'Sign In or Register', 'RSMC for Nowcasting', 'SWIRLS Nowcast System', 'Getting Started', 'Examples', 'Reference Manual', 'Data Model and xarray', 'Contributors' Guide', 'Z-R Relationship', 'Educational Resources', 'Most Active Developers', 'Change Log', 'Terms and Conditions', 'Contact Us', and 'Acronyms'.



A word cloud featuring various data science and machine learning libraries, including numpy, Sphinx, wradlib, pillow, matplotlib, metpy, h5py, pytest, GitLab, pylint, basemap, cartopy, pyresample, scipy, Conda, OpenCV, and pandas.

**Participants from  
19 Organisations in  
9 Countries**



The screenshot shows a GitHub repository page for the 'swirls' project. The page displays a list of issues, with the first issue titled 'I will check those packages thank you! Yes during installation swirls automatically installs the packages it needed there weren't any errors during installation.' The issue is assigned to 'Wen Jun Cyril Chin' and has a 'Suggest' label. The page also shows a sidebar with navigation links like 'Board', 'Labels', 'Milestones', 'Merge Requests', 'CI / CD', 'Operations', 'Registry', 'Wiki', 'Snippets', and 'Settings'.

# Community SWIRLS Nowcast System



## Users in Asian countries

Australia	Macao
Bangladesh	Malaysia
China	Myanmar
Fiji	Philippines
India	Singapore
Indonesia	Sri Lanka
Japan	Thailand
Korea	Vietnam
Laos	

## Highlights of Com-SWIRLS 2.0+

- Continuously Maintained and Updated
- Installable by Conda, Single Command
- Documentation Website with User Examples
- Readable Codes, Reusable Modules
- Version Control with GitLab
- Technical Support & Discussions by GitLab Issues
- Support Various Radar Data Formats
- Various QPE Interpolation Methods
- Numerous Motion Field and Forecast Algorithms
- Verification Metrics



## Com-SWIRLS 1.x vs 2.x

	Com-SWIRLS 1.x	Com-SWIRLS 2.x
<b>Installation</b>	VM	Conda *
<b>Programming Language</b>	Assorted	Primarily Python, plus C++
<b>Graphics Libraries</b>	NCL, ImageMagick etc.	Open-source Python lib.
<b>Grid Size</b>	480x480 Only	Rectangular grid of any size
<b>Supported Radar Data</b>	1	8
<b>Motion Field Algorithms</b>	1	5
<b>Forecast Algorithms</b>	1	5
<b>QPE Methods</b>	1	14, plus multi-sensor QPE
<b>Verification Metrics</b>	No	14
<b>Documentation</b>	Limited	Full Documentations
<b>Version Control</b>	No	Yes, using GitLab-CE
<b>Software Testing</b>	At initial development only	Upon any changes

\* VM and Docker available on request

# Use of Com-SWIRLS/SwirlsPy

## User Script

- Developed by the NMHS, assisted by HKO
- Tailored to the specific situation, e.g. warning criteria, database, infrastructure, of that NMHS

```
import os
import numpy as np
import pandas as pd
import xarray as xr
import cartopy.feature as cfeature
import matplotlib.pyplot as plt
from matplotlib.colors import BoundaryNorm, ListedColormap
from pyresample import utils

from swirlspy.rad.uf_ph import read_uf_ph
from swirlspy.qpe.utils import locate_file, timestamps_ending
from swirlspy.qpf import rover
from swirlspy.qpf import sla
from swirlspy.utils import standardize_attr, FrameType
from swirlspy.utils.conversion import to_rainfall_depth, acc_rainfall_depth
from swirlspy.core.resample import grid_resample

plt.switch_backend('agg')

THIS_DIR = os.getcwd()
os.chdir(THIS_DIR)

start_time = pd.Timestamp.now()
```

Import and Call  
SwirlsPy Modules

## Shared Modules (SwirlsPy)

- Developed by HKO Staff and Contributing Developers
- Shared with all NMHSs
- Regularly Updated
- Backward Compatible Whenever Possible



# WMO RSMC FOR NOWCASTING

# RSMC for Nowcasting

Opening Ceremony on 2 October 2019 inaugurated by  
DHKO and Professor Petteri Taalas, Secretary-General of  
WMO on 2 October 2018



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WMO has signed an agreement with the Hong Kong Observatory (HKO) to further strengthen meteorological cooperation.

Under a Memorandum of Understanding signed by HKO Director Shun Chi-ming and WMO Secretary-General Petteri Taalas, HKO will support WMO's initiative in establishing the Global Multi-hazard Alert System (GMAS).

In this connection, the HKO has revamped the Severe Weather Information Centre website (SWIC 2.0) and is also

## WMO and Hong Kong Observatory strengthen cooperation

Tags: [WMO](#) [Disaster risk reduction](#) [Forecast](#) [Tropical cyclones](#) [Capacity development](#)

2 Published 2 October 2018

Member: [Hong Kong, China](#)



### Latest WMO News

- Three Ways Forward to Improve Regional Information for Extreme Events: An Early Career Perspective

18 February 2019



## WMO Designated Global Data-processing and Forecasting System Centres

Updated on 24 August 2018



### Legend

- |   |  |   |
|---|--|---|
| ★ World Meteorological Centres* (9)                                     | ● RSMC Nowcasting (3)                  | ⚡ RSMCs Non-Nuclear Emergency Response (2)                          |
| ⊙ Atmospheric Transport Modelling (10)                                  | △ RSMCs TC (6)                         | ⚓ RSMCs Volcano watch services for international air navigation (1) |
| ◇ Global Producing Centres for Long-Range Forecasts (13)                | ⬢ RSMCs Sand/Dust (2)                  | ℝ RSMCs Severe Weather Forecasting (2)                              |
| ⊕ Global Producing Centres for Annual to Decadal Climate Prediction (3) | ▷ Regional Climate Centres (11)        | ⚓ RSMCs marine meteorological services (24)                         |
| ■ RSMCs Geographic (25)   | ⚡ RSMCs Nuclear Emergency Response (9) |   |

\* World Meteorological Centres are also Global Producing Centres for a) Deterministic Numerical Weather Prediction, b) Ensemble Numerical Weather Prediction, and c) Long-Range Forecasts.

### DESIGNATIONS USED

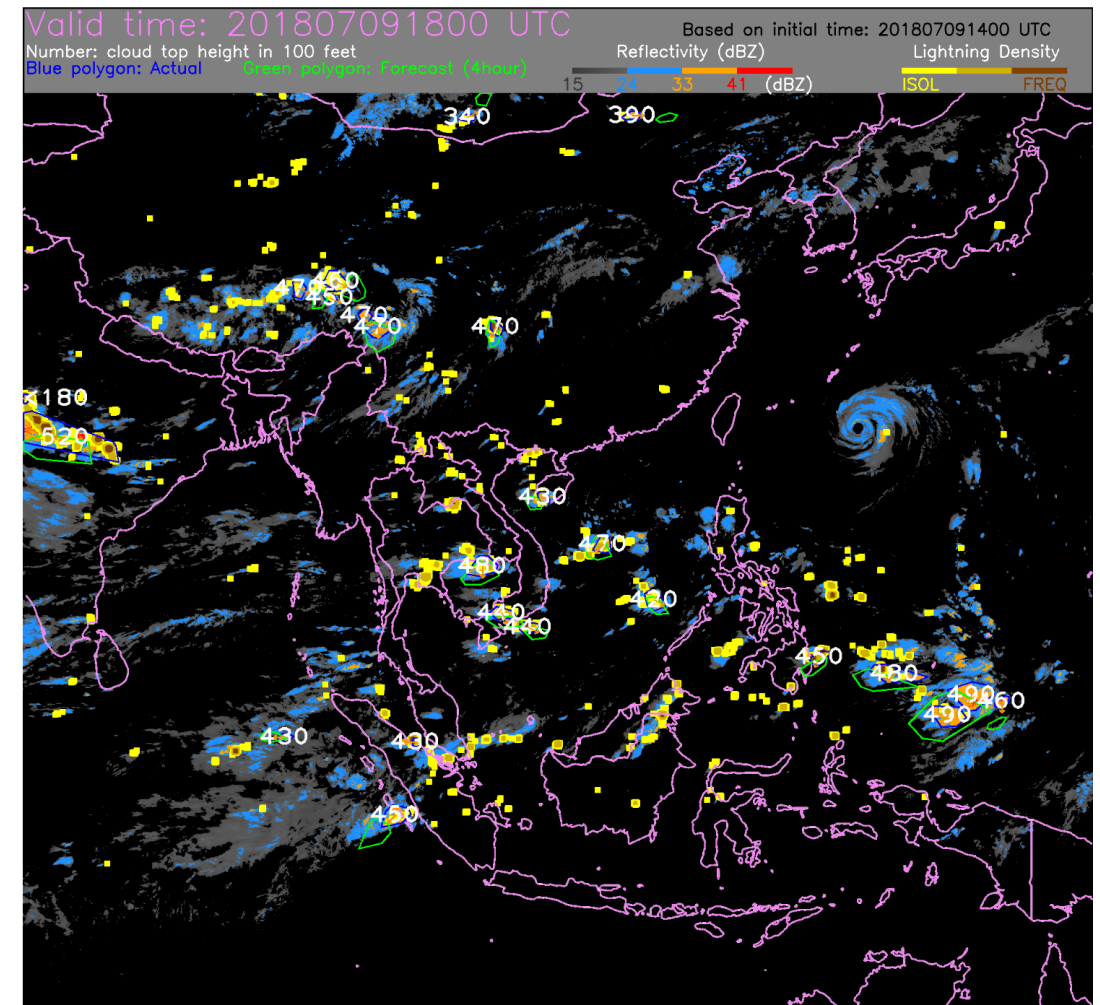
The depiction and use of boundaries, geographic names and related data shown on maps and included in lists, tables, documents, and databases on this web site are not warranted to be error free nor do they necessarily imply official endorsement or acceptance by the WMO.

# WMO RSMC for Nowcasting

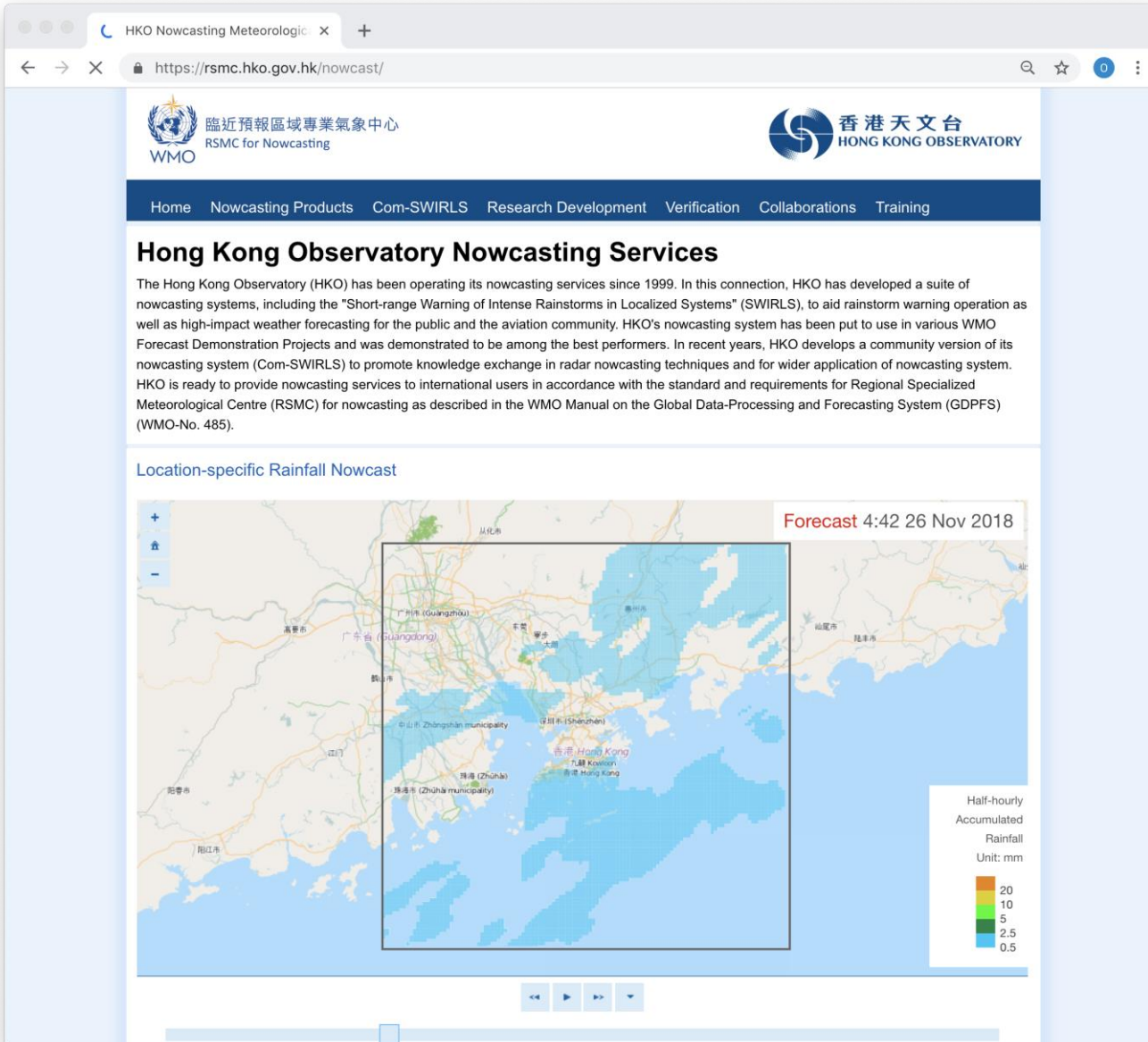
Website: <https://rsmc.hko.gov.hk/nowcast/>

RSMC Hong Kong (since WMO EC-70 in June 2018)

Significant Convection Nowcast over East Asia



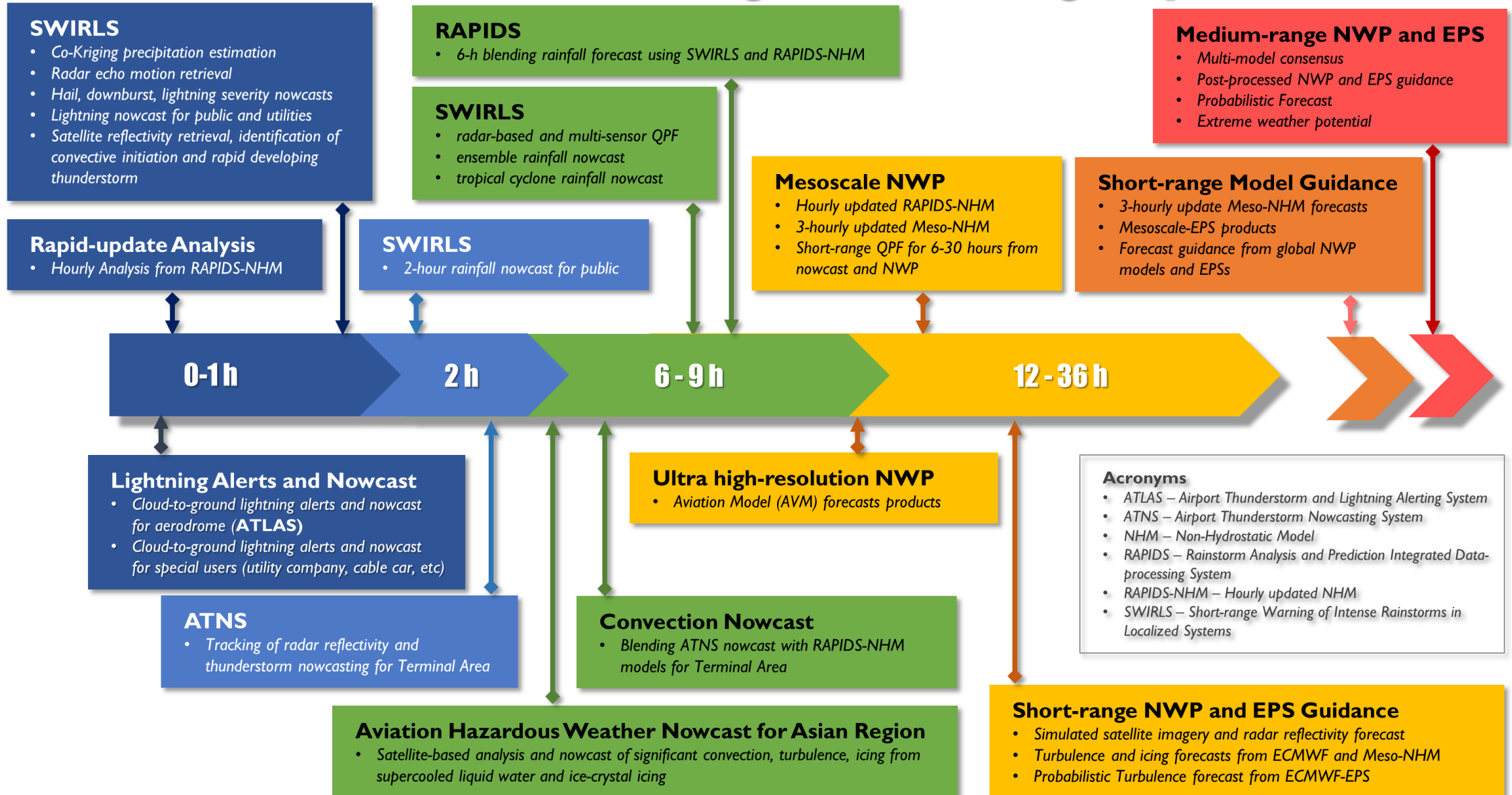
Significant convection nowcast at the fourth hour (in UTC) using retrieved reflectivity (blue  $\geq 24$  dBZ, orange  $\geq 33$  dBZ and red  $\geq 41$  dBZ) from Himawari-8 data and updated every 10 minutes. Lightning counts are marked in yellow (sparse) to brown (dense) with green polygons on areas with high reflectivity or dense lightning.



The screenshot shows the homepage of the Hong Kong Observatory's Nowcasting Services website. The header includes the WMO logo and the text "臨近預報區域專業氣象中心 RSMC for Nowcasting" and "香港天文台 HONG KONG OBSERVATORY". The navigation menu includes links to Home, Nowcasting Products, Com-SWIRLS, Research Development, Verification, Collaborations, and Training. The main content area is titled "Hong Kong Observatory Nowcasting Services" and provides a detailed description of the nowcasting system, including its history, purpose, and the various products it offers. Below the text, there is a section titled "Location-specific Rainfall Nowcast" which features a map of the region around Hong Kong, showing rainfall forecasts for various locations. The map includes a legend for "Half-hourly Accumulated Rainfall" with a color scale from 0.5 to 20 mm. The forecast is for 4:42 on 26 Nov 2018.



# Seamless Nowcast To Medium-range Forecast for High-Impact Weather



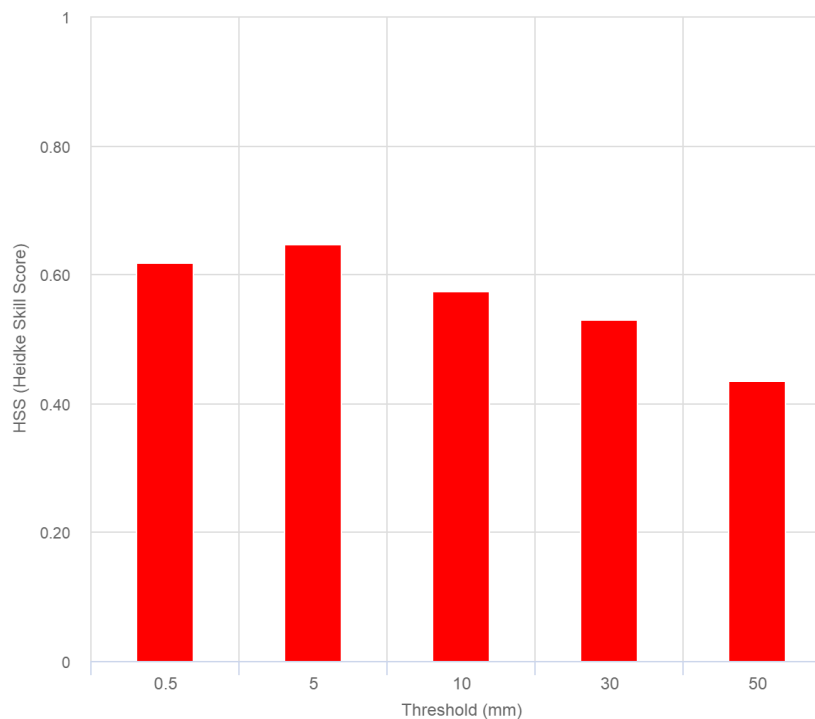


# Verifications

## Heidke Skill Score of I-h nowcast of different thresholds

HSS (Heidke Skill Score) from 2017-01-01 00:00 to 2017-12-31 23:57

Lead Time = 60 min



ROVER

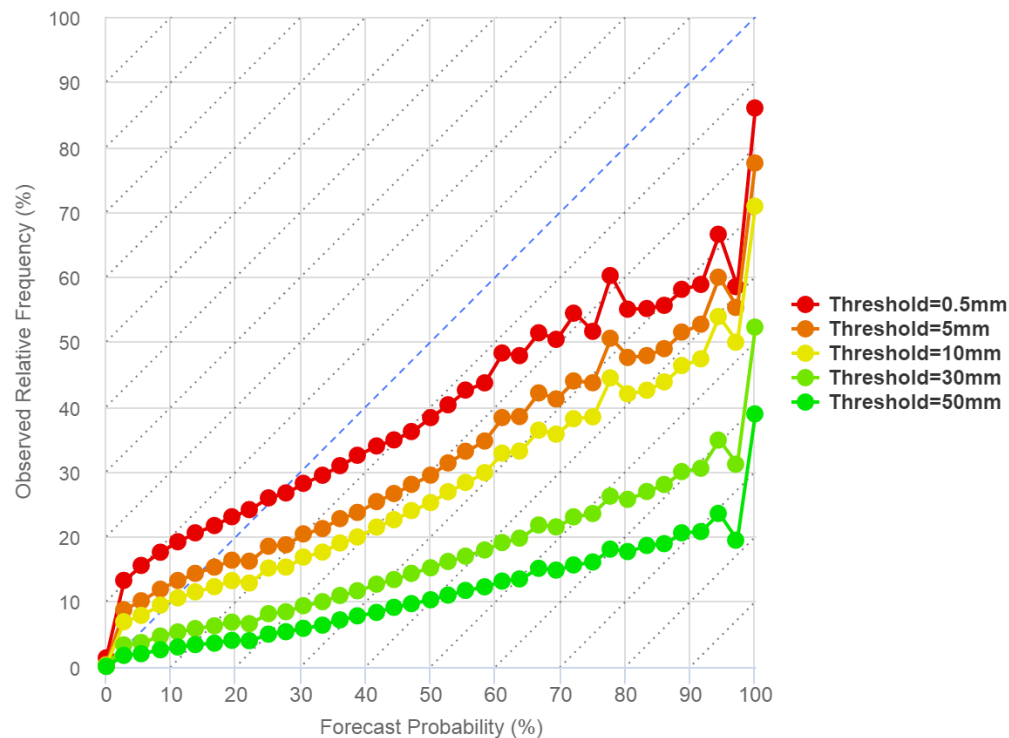
X: All Missing, M: >20% Missing, U: Undefined (divided by zero), 0: Zero

<https://rsmc.hko.gov.hk/nowcast/verificationBenchmark.html>

## Reliability Diagram

Reliability Diagram from 2017-01-01 00:00 to 2017-12-31 23:57

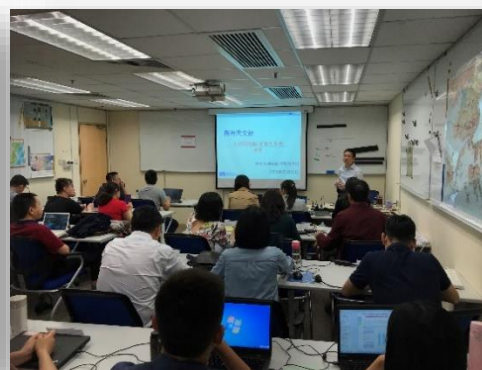
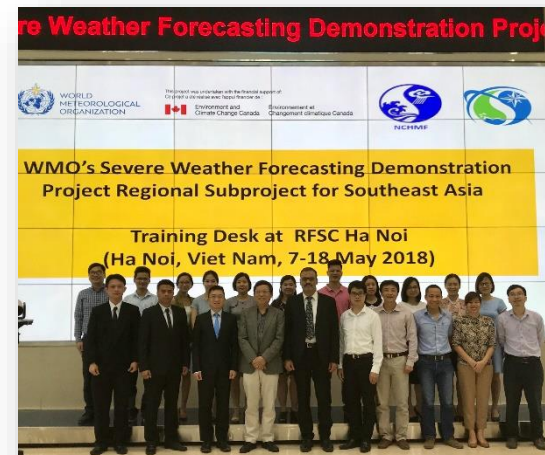
Lead Time = 60 min



## Significant Convection Nowcast (2018)

Forecast time	POD	FAR	CSI
0-1 <sup>st</sup> hr	0.945	0.116	0.841
0-2 <sup>nd</sup> hr	0.893	0.114	0.801
0-3 <sup>rd</sup> hr	0.837	0.117	0.754
0-4 <sup>th</sup> hr	0.788	0.120	0.711
0-5 <sup>th</sup> hr	0.747	0.124	0.676
0-6 <sup>th</sup> hr	0.712	0.128	0.645

# Technology Transfer for Capacity Building on Nowcasting





# Community Version of SWIRLS (Com-SWIRLS)

Presentation ([link](#)) in  
WMO Symposium on Nowcasting and Very-short-range Forecast 2016 (WSN16) in Hong Kong



# LATEST DEVELOPMENTS



# Deep Learning Nowcast

# Deep Learning Nowcast in SWIRLS

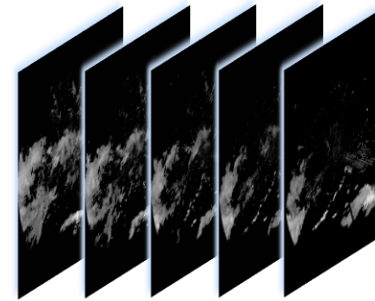
- A Brief History

Year	Event
2014	Initiated collaboration with HKUST
2015	ConvLSTM developed, outperforming ROVER for rain/no-rain (0.5 mm/h), resolution limited to 100x100 only
2016	Model enhancement and adaptation
2017	TrajGRU developed and benchmarked, outperforming ROVER for 30mm/h support arbitrary resolution
2018	Commence operational trial of TrajGRU

- Predicting evolution of weather radar imagery

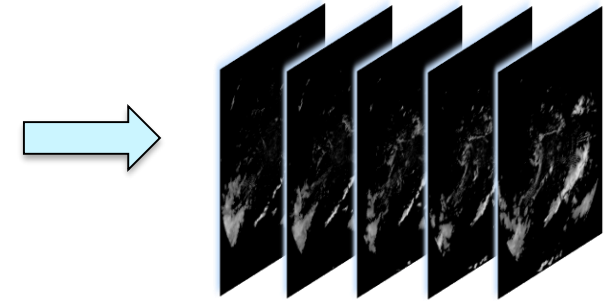
**Input sequence:**

observed radar maps up to current time step



**Output sequence:**

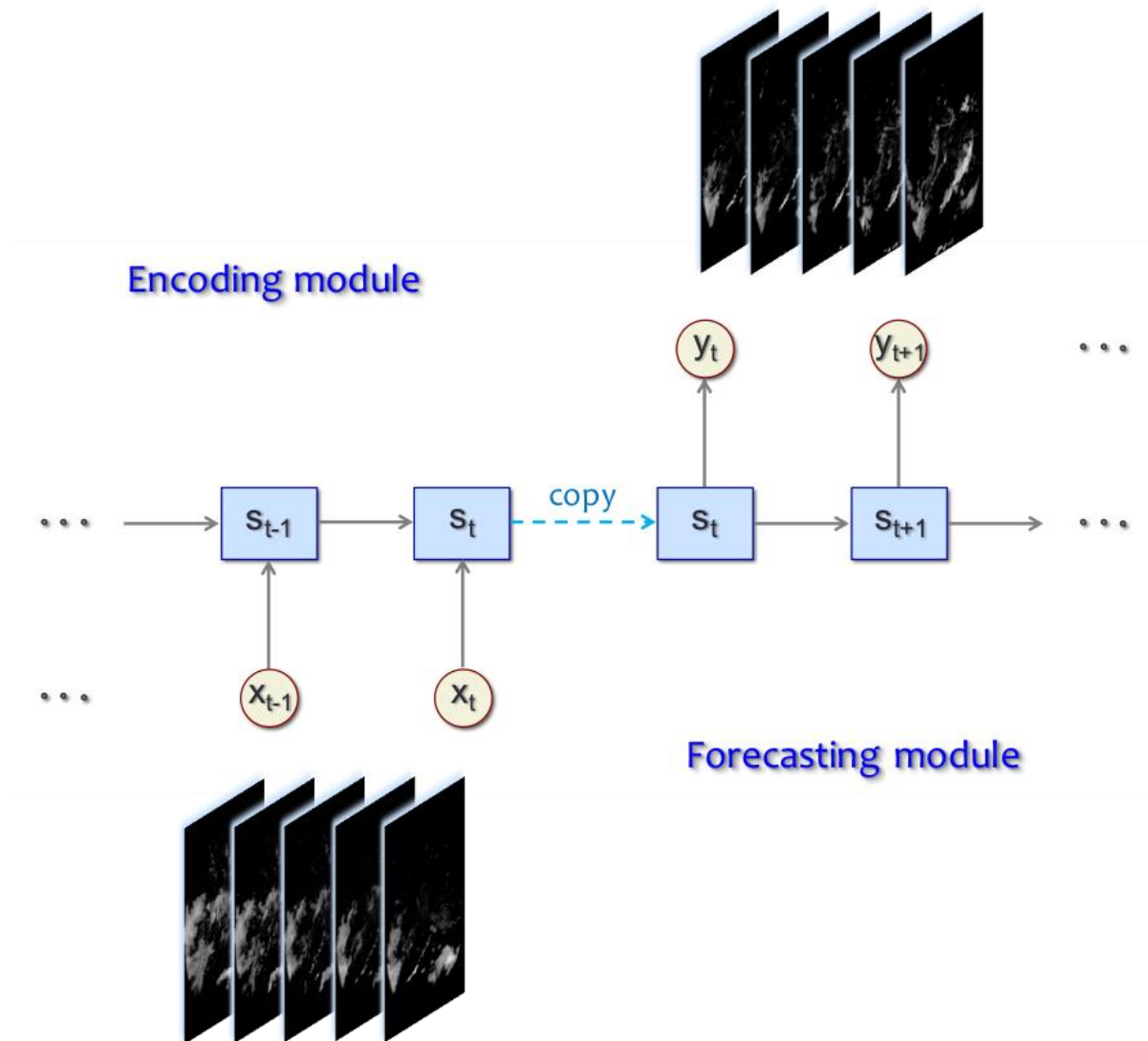
predicted radar maps for future time steps



Mathematically to maximize posterior pdf of echo sequence across K time levels based on previous J time levels of observations

$$\tilde{\mathcal{X}}_{t+1}, \dots, \tilde{\mathcal{X}}_{t+K} = \arg \max_{\mathcal{X}_{t+1}, \dots, \mathcal{X}_{t+K}} p(\mathcal{X}_{t+1}, \dots, \mathcal{X}_{t+K} \mid \hat{\mathcal{X}}_{t-J+1}, \hat{\mathcal{X}}_{t-J+2}, \dots, \hat{\mathcal{X}}_t)$$

# Spatiotemporal encoding-forecasting model



- Convolutional Long Short Term Memory (ConvLSTM) model
  - X. Shi, Z. Chen, H. Wang, D.Y. Yeung, W.K. Wong, and W.C. Woo, 2015: Convolutional LSTM network: A machine learning approach for precipitation nowcasting. NIPS 2015.

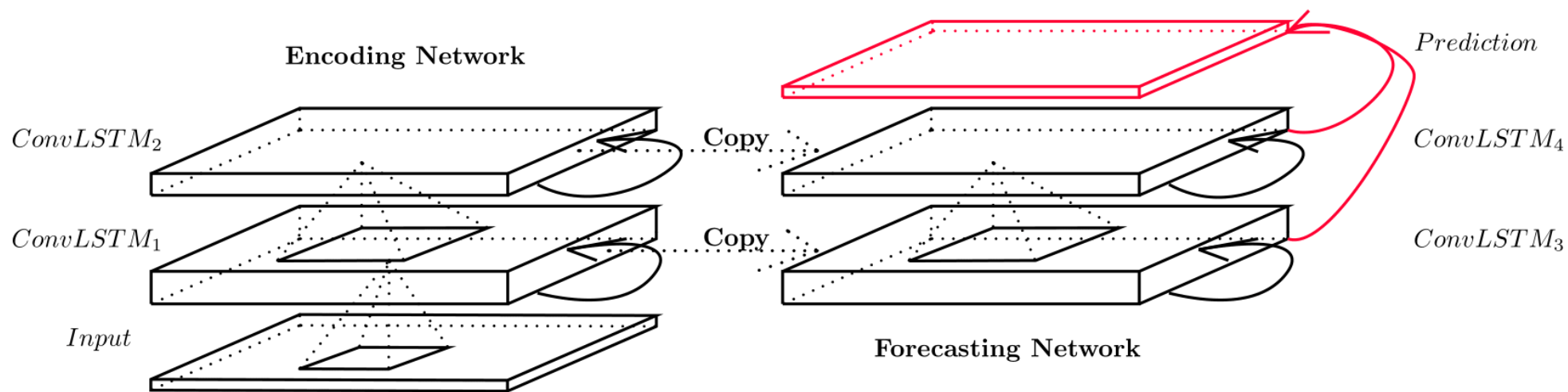
Link: <https://arxiv.org/abs/1506.04214>

- Two key components:
  - **Convolutional** layers
  - **Long short-term memory** (LSTM) cells in **recurrent neural network** (RNN) model



# Encoding and Forecasting Networks in ConvLSTM

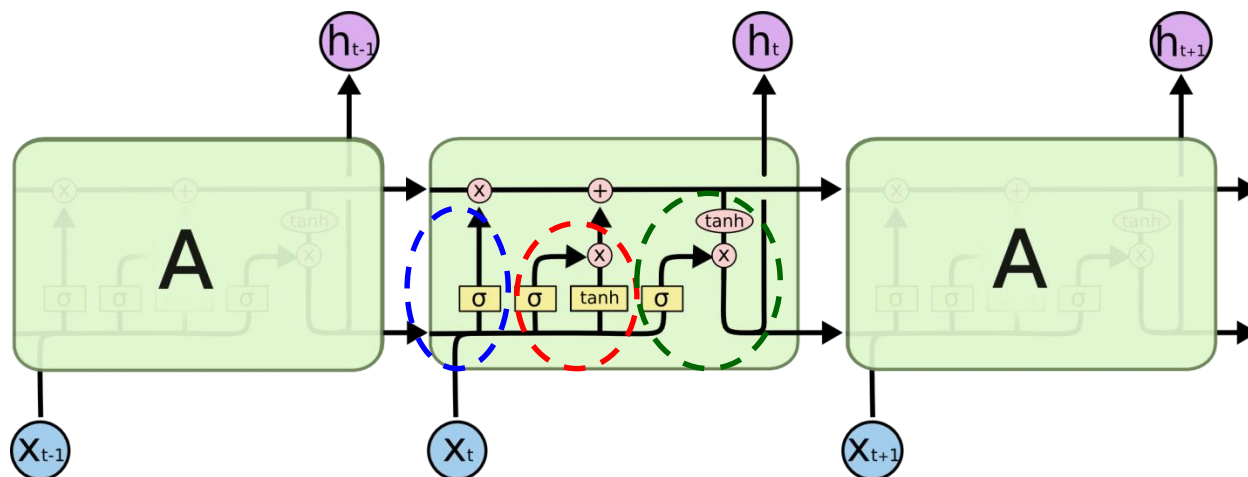
- Last states and cell outputs of encoding network become initial states and cell outputs of forecasting network
- **Encoding network** compresses the input sequence into a hidden state tensor
- **Forecasting network** unfolds the hidden state tensor to make prediction



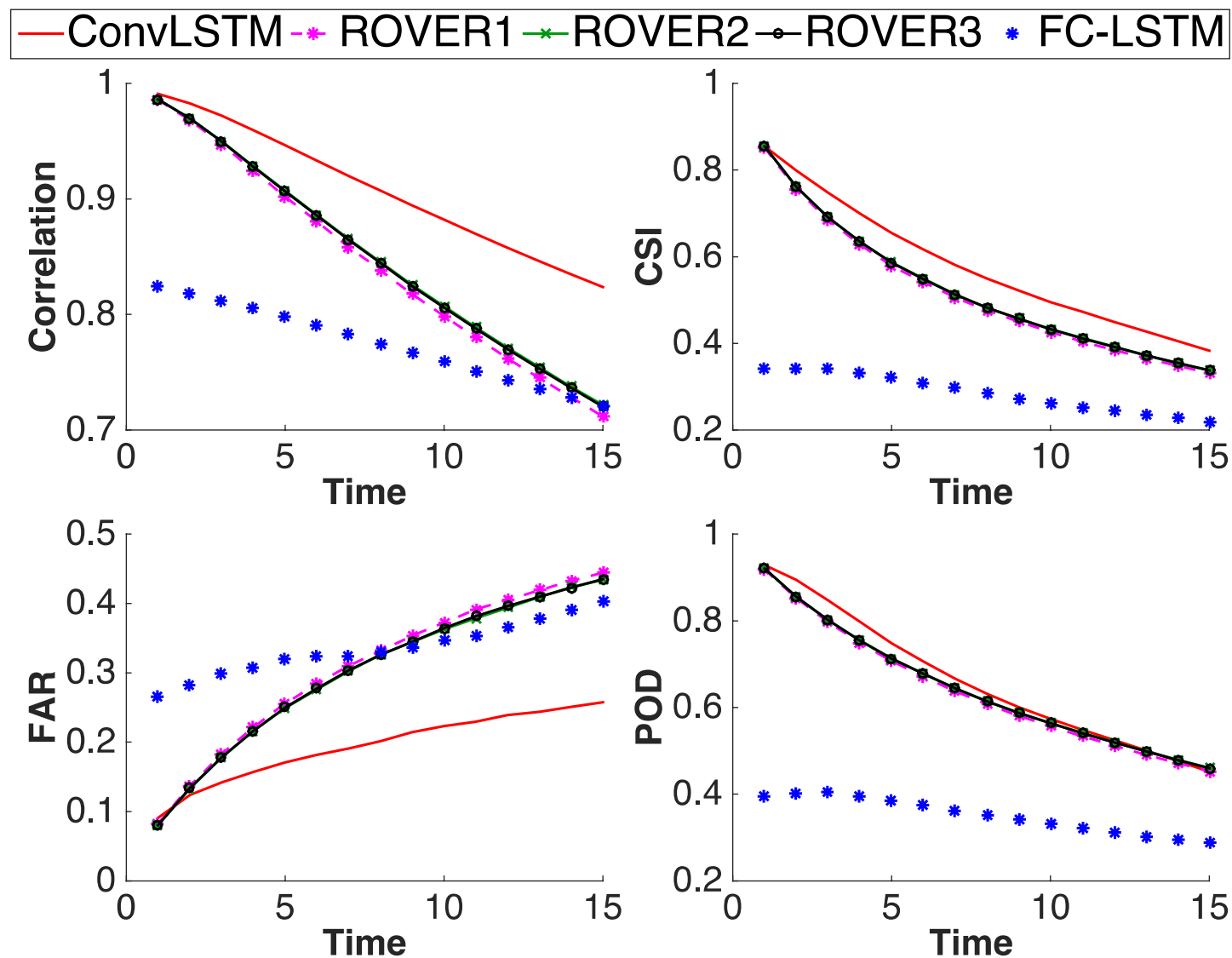
# Governing Equations in ConvLSTM

Inputs  
↓

input gate      $i_t = \sigma(W_{xi} * \mathcal{X}_t + W_{hi} * \mathcal{H}_{t-1} + W_{ci} \circ \mathcal{C}_{t-1} + b_i)$   
forget gate      $f_t = \sigma(W_{xf} * \mathcal{X}_t + W_{hf} * \mathcal{H}_{t-1} + W_{cf} \circ \mathcal{C}_{t-1} + b_f)$   
Cell outputs      $\mathcal{C}_t = f_t \circ \mathcal{C}_{t-1} + i_t \circ \tanh(W_{xc} * \mathcal{X}_t + W_{hc} * \mathcal{H}_{t-1} + b_c)$   
output gate      $o_t = \sigma(W_{xo} * \mathcal{X}_t + W_{ho} * \mathcal{H}_{t-1} + W_{co} \circ \mathcal{C}_t + b_o)$   
Hidden states      $\mathcal{H}_t = o_t \circ \tanh(\mathcal{C}_t)$

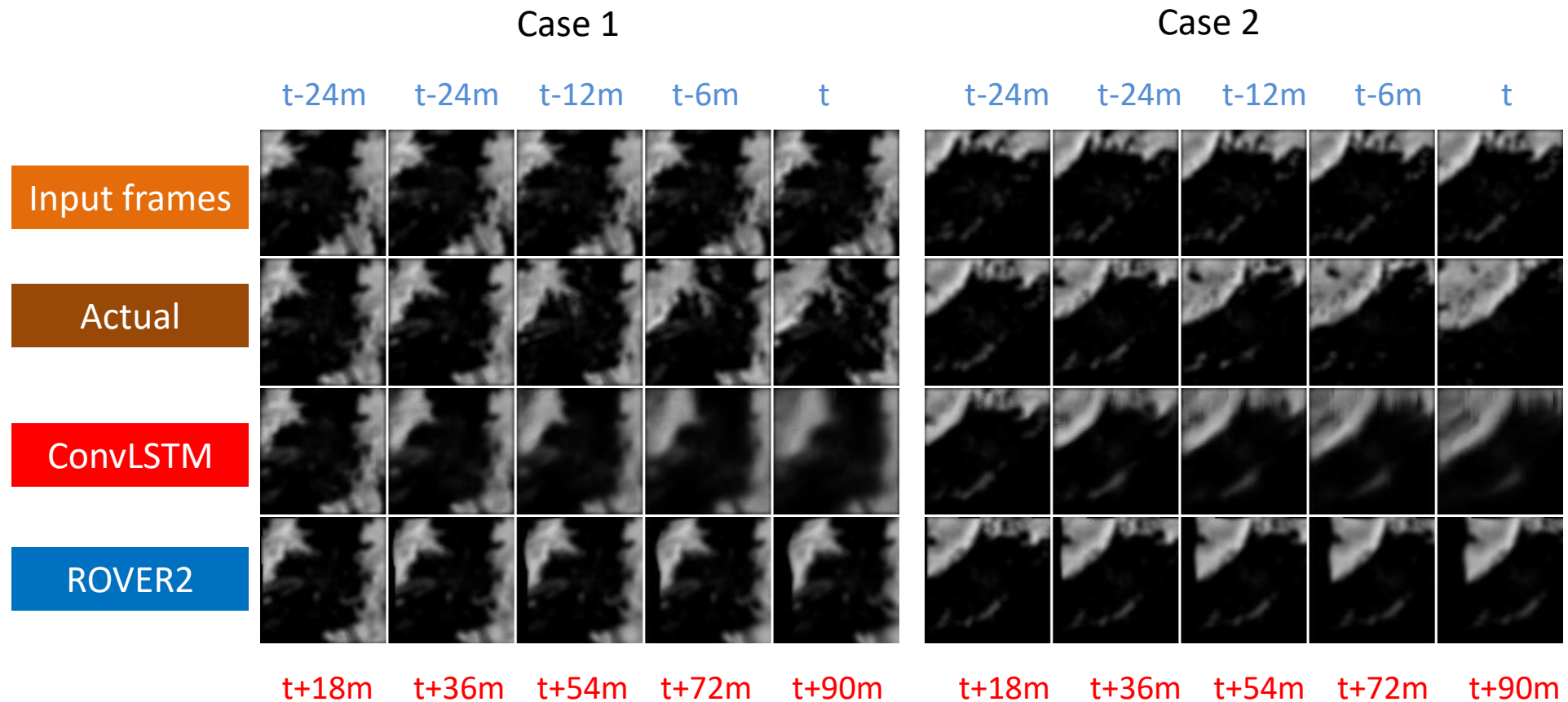


# Forecast performance vs. lead time



# Two squall line cases

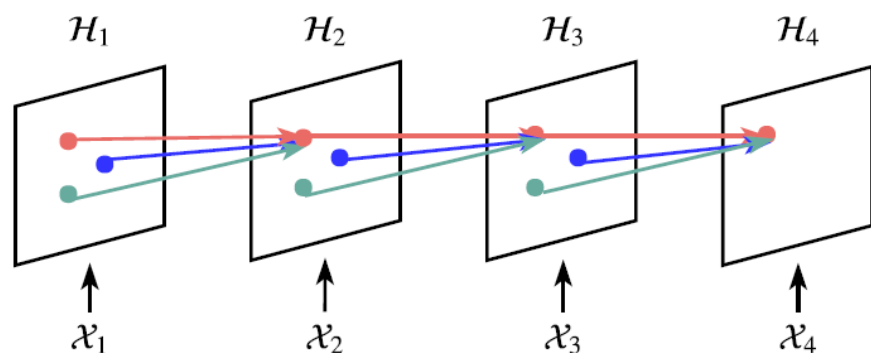
- Radar located at center
- 5 input frames are used and a total of 15 frames (i.e. T+90 min) in forecasts



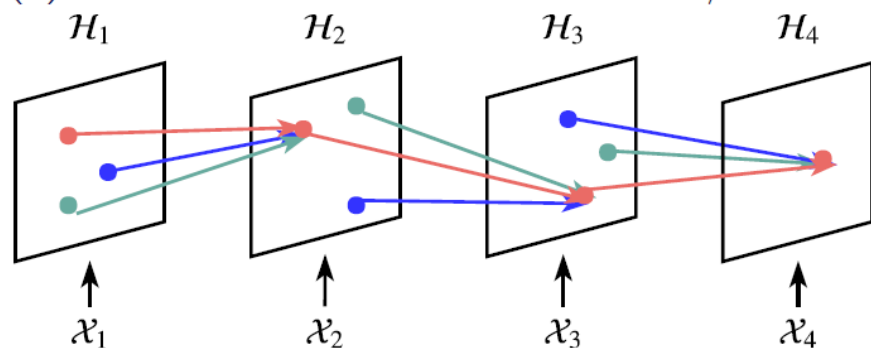


# A New Model for Deep Learning in Nowcast

## Trajectory Gated Recurrent Unit (TrajGRU)



(a) ConvRNN: Links are fixed over time/location.



(b) TrajRNN: Links are dynamically determined.

<https://arxiv.org/pdf/1706.03458.pdf>

### Deep Learning for Precipitation Nowcasting: A Benchmark and A New Model

Xingjian Shi, Zhihan Gao, Leonard Lausen, Hao Wang, Dit-Yan Yeung  
 Department of Computer Science and Engineering  
 Hong Kong University of Science and Technology  
 {xshiab, zgaoag, lelausen, hwangaz, dyyeung}@cse.ust.hk

Wai-kin Wong, Wang-chun Woo  
 Hong Kong Observatory  
 Hong Kong, China  
 {wkwong, wcwoo}@hko.gov.hk

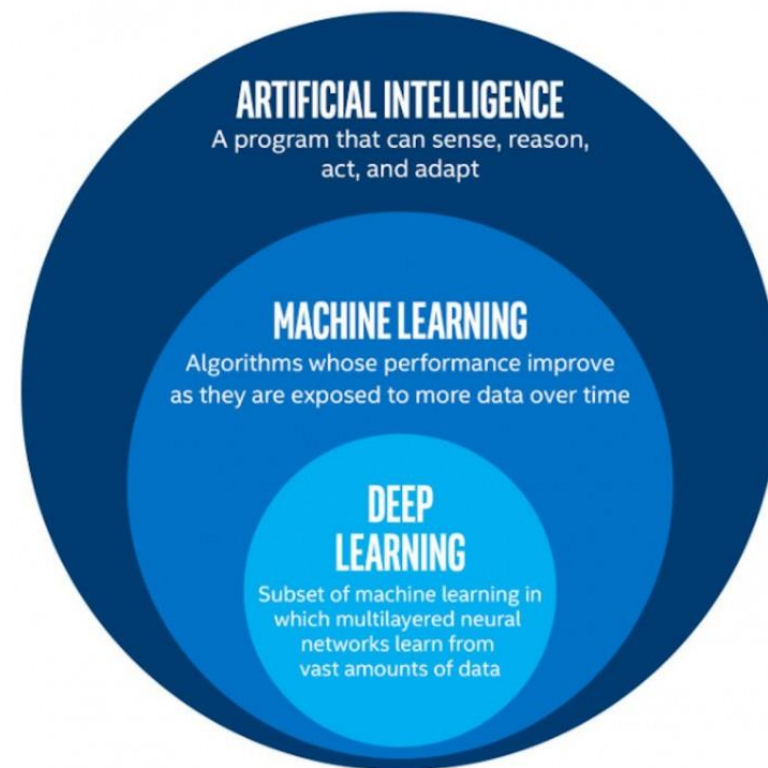
#### Abstract

With the goal of making high-resolution forecasts of regional rainfall, precipitation nowcasting has become an important and fundamental technology underlying various public services ranging from rainstorm warnings to flight safety. Recently, the *Convolutional LSTM* (ConvLSTM) model has been shown to outperform traditional optical flow based methods for precipitation nowcasting, suggesting that deep learning models have a huge potential for solving the problem. However, the convolutional recurrence structure in ConvLSTM-based models is *location-invariant* while natural motion and transformation (e.g., rotation) are *location-variant* in general. Furthermore, since deep-learning-based precipitation nowcasting is a newly emerging area, clear evaluation protocols have not yet been established. To address these problems, we propose both a new model and a benchmark for precipitation nowcasting. Specifically, we go beyond ConvLSTM and propose the *Trajectory GRU* (TrajGRU) model that can actively learn the *location-variant* structure for recurrent connections. Besides, we provide a benchmark that includes a real-world large-scale dataset from the Hong Kong Observatory, a new training loss, and a comprehensive evaluation protocol to facilitate future research and gauge the state of the art.

# Performance

TrajGRU outperformed operational method and other deep-learning techniques, in particular for heavy rain  $r \geq 30$  mm in 1 h, based on 7-years of radar data

Algorithms	$r \geq 0.5$	$r \geq 2$	HSS $\uparrow$ $r \geq 5$	$r \geq 10$	$r \geq 30$	B-MSE $\downarrow$	B-MAE $\downarrow$
Offline Setting							
Last Frame	0.5207	0.4531	0.3582	0.2512	0.1193	15274	28042
ROVER + Linear	0.6038	0.5473	0.4516	0.3301	0.1762	11651	23437
ROVER + Non-linear	0.5896	0.5436	0.4590	0.3318	0.1576	10945	22857
2D CNN	0.6366	0.5809	0.4851	0.3690	0.1885	7332	18091
3D CNN	0.6334	0.5825	0.4862	0.3734	0.2034	7202	17593
ConvGRU-nobal	0.6756	0.6094	0.4981	0.3286	0.1160	9087	19642
ConvGRU	0.6701	0.6104	0.5163	0.4159	0.2893	5951	15000
TrajGRU	<b>0.6731</b>	<b>0.6126</b>	<b>0.5192</b>	<b>0.4207</b>	<b>0.2996</b>	<b>5816</b>	<b>14675</b>
Online Setting							
2D CNN	0.6365	0.5756	0.4790	0.3744	0.2162	6654	17071
3D CNN	0.6355	0.5736	0.4766	0.3733	0.2220	6690	16903
ConvGRU	0.6712	0.6105	0.5183	0.4226	0.2981	5724	14772
TrajGRU	<b>0.6760</b>	<b>0.6164</b>	<b>0.5253</b>	<b>0.4308</b>	<b>0.3111</b>	<b>5589</b>	<b>14465</b>

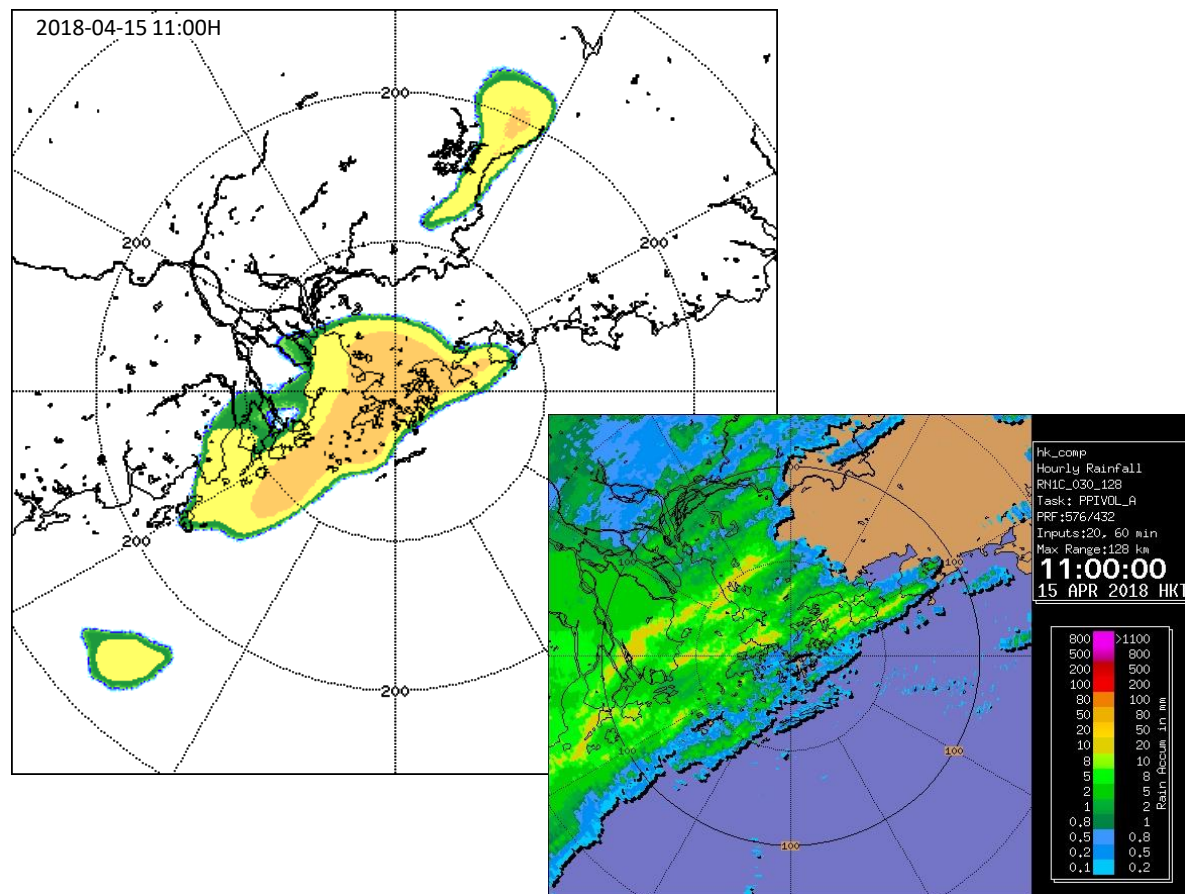


# TrajGRU and HKO-7 Benchmark Dataset

- TrajGRU available to public on GitHub:
  - <https://github.com/sxjscience/HKO-7>
- The 7-years of HKO radar image archive as benchmark for development of deep learning in nowcasting and other weather application from the above Github repository
  - Available to researchers and weather services subject to agreement to the undertaking

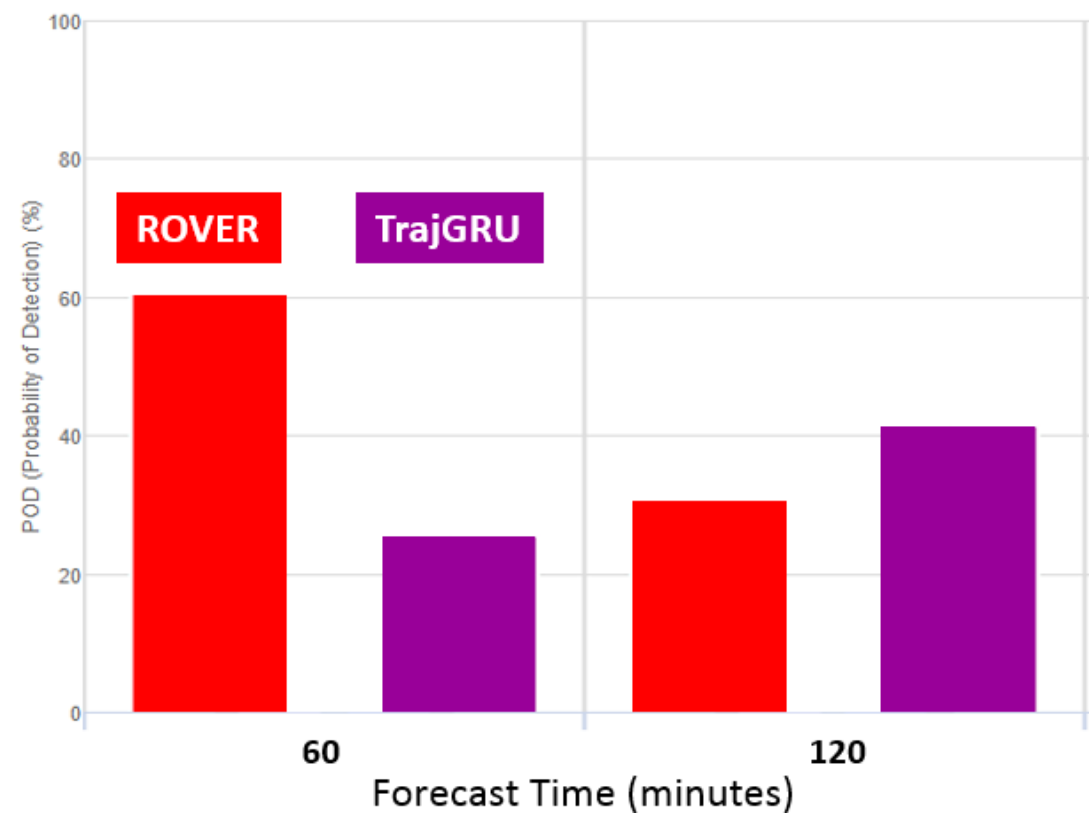
# TrajGRU in real-time trial

T+60 min TrajGRU Nowcast



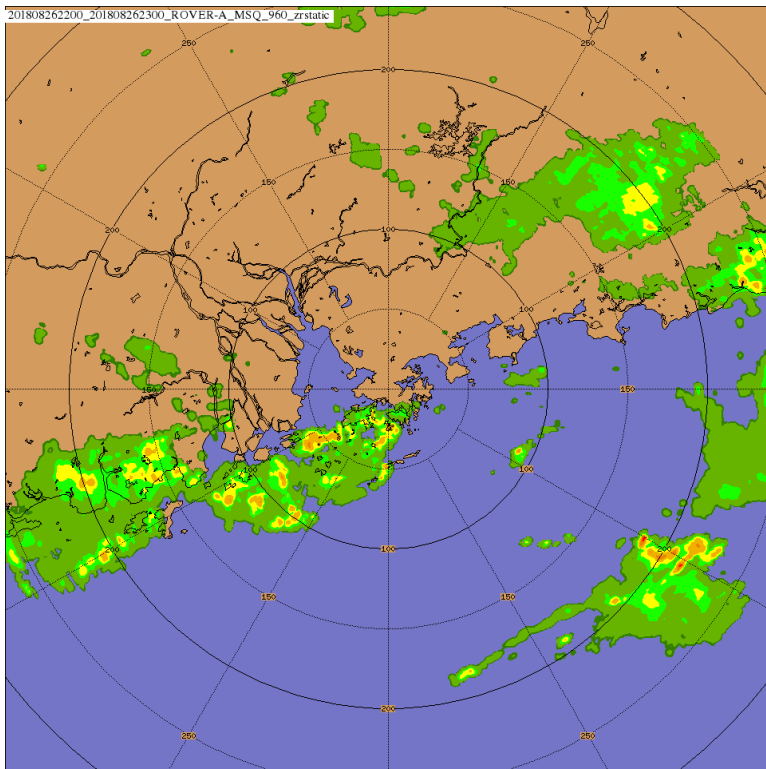
Verification Period: May – Sept 2018

POD ( $\geq 10$  mm/h)

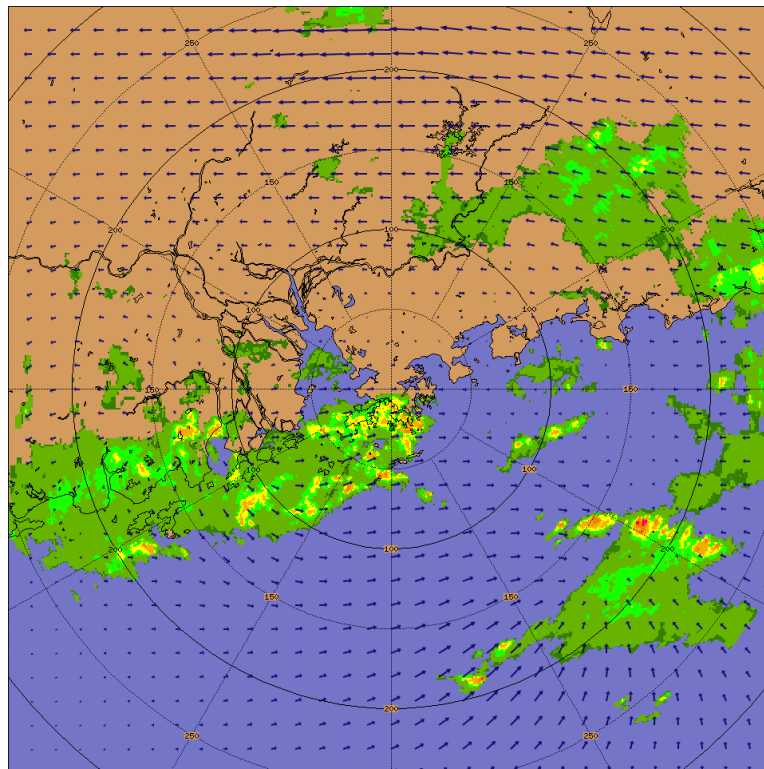




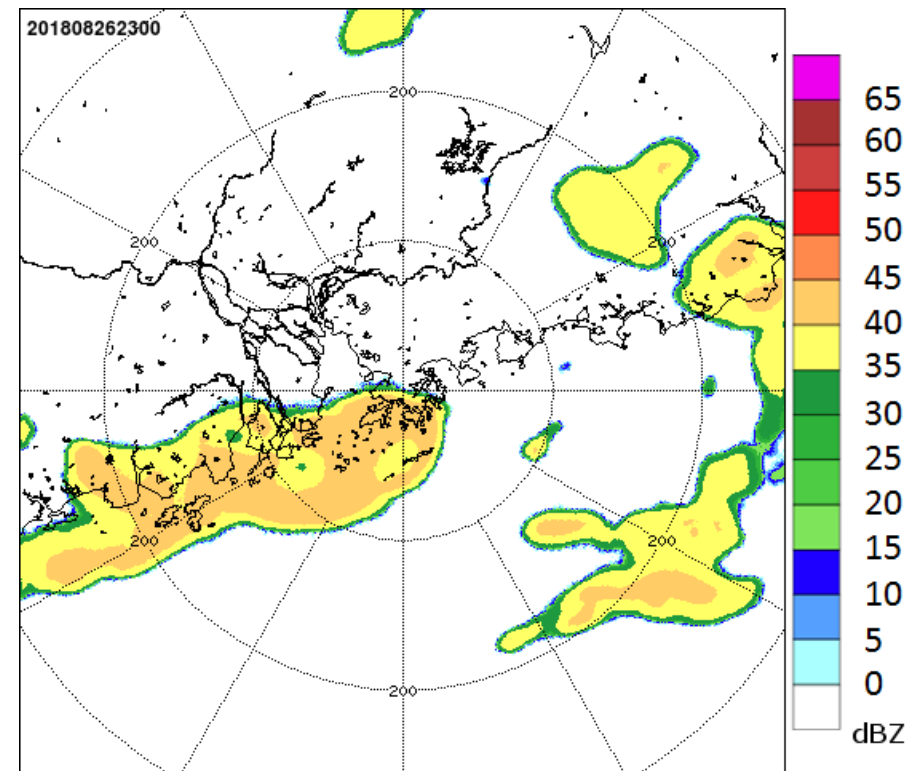
# 60-min Forecast Reflectivity Based at 2018/08/26 23:00H



Optical Flow (ROVER)

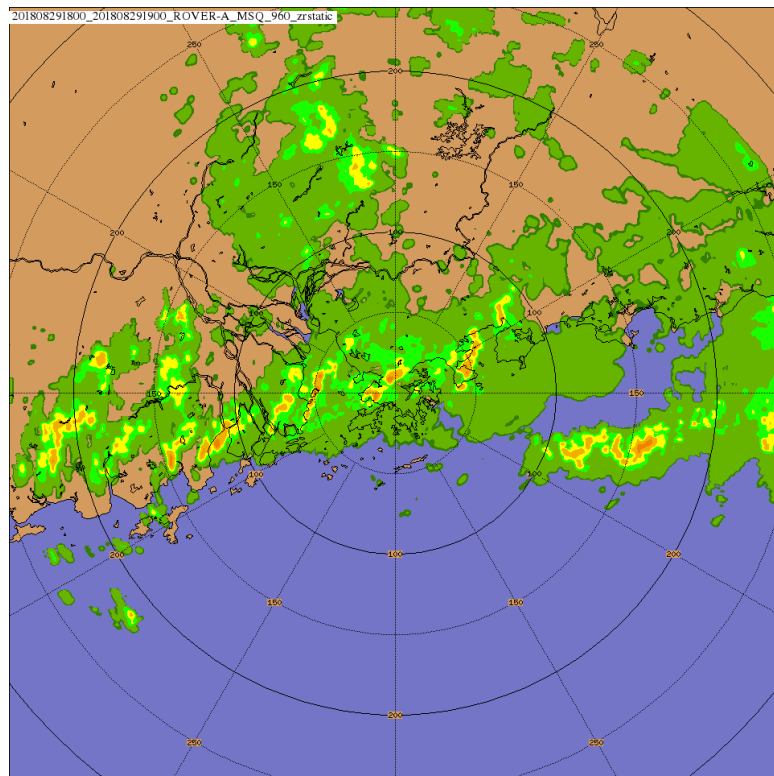


Actual

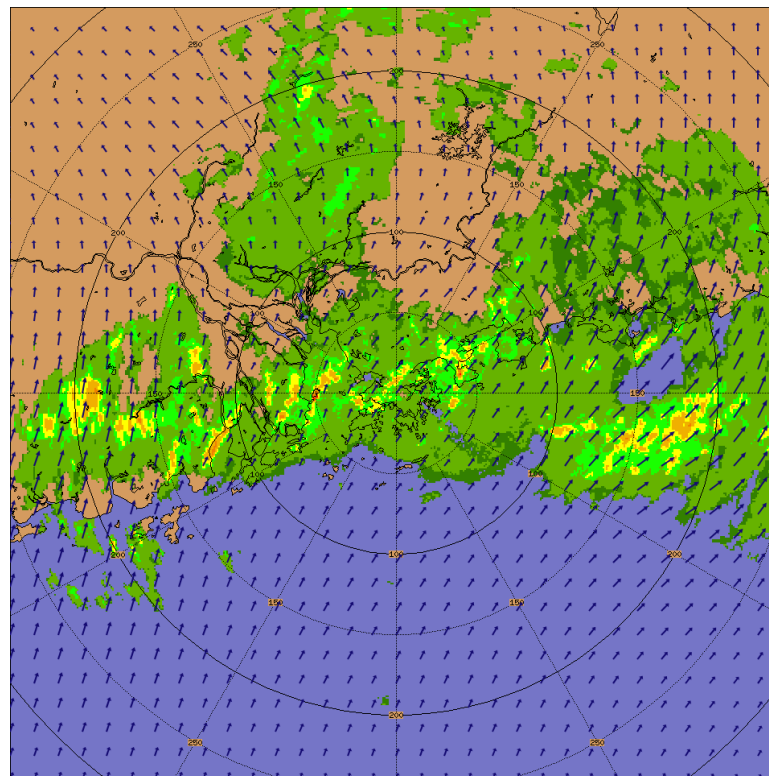


Deep Learn (TrajGRU)

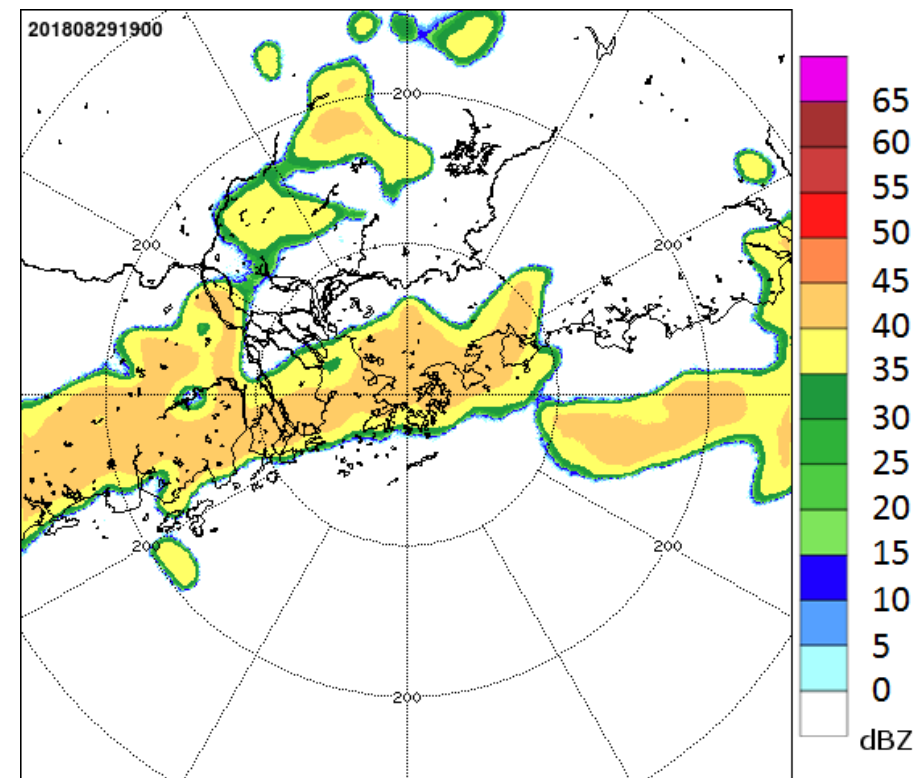
# 60-min Forecast Reflectivity Based at 2018/08/29 18:00



Optical Flow (ROVER)

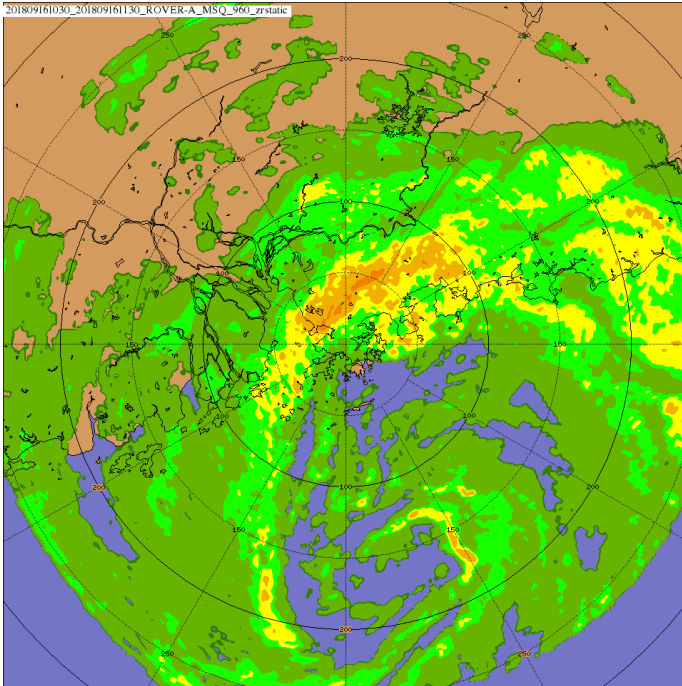


Actual

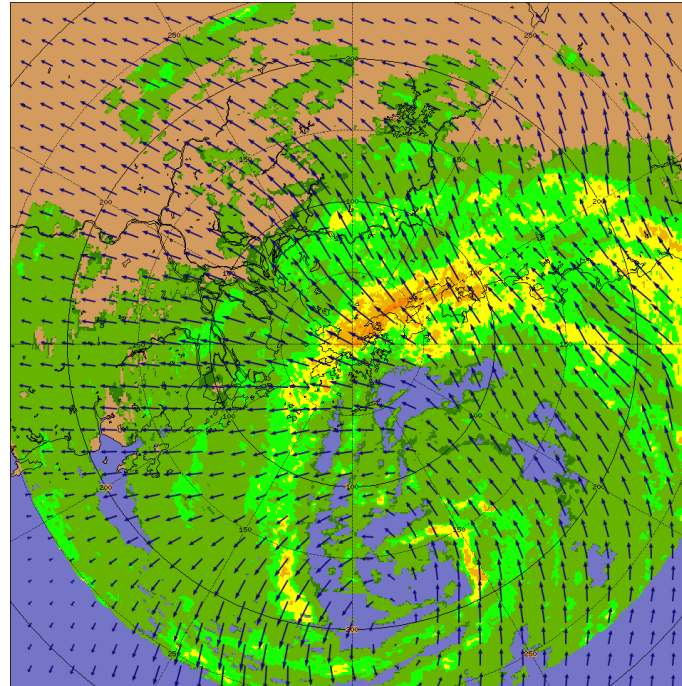


Deep Learn (TrajGRU)

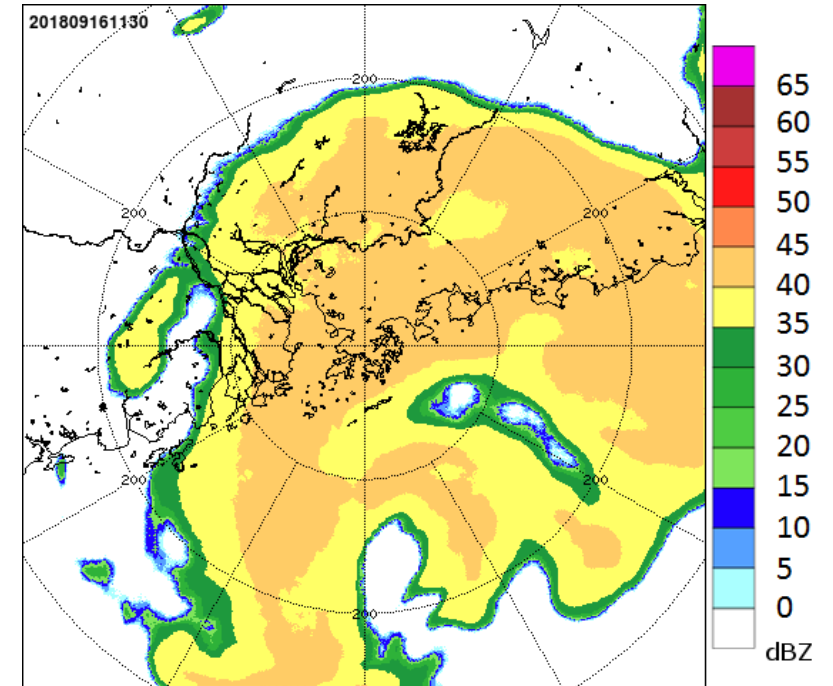
# 60-min Forecast Reflectivity Based at 2018/09/16 10:30



Optical Flow (ROVER)



Actual

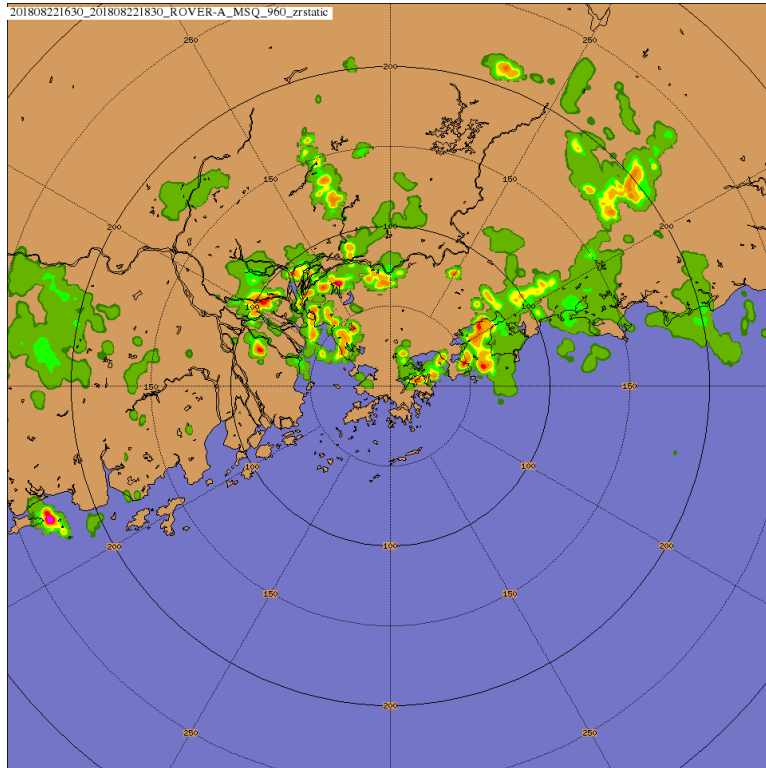


Deep Learn (TrajGRU)

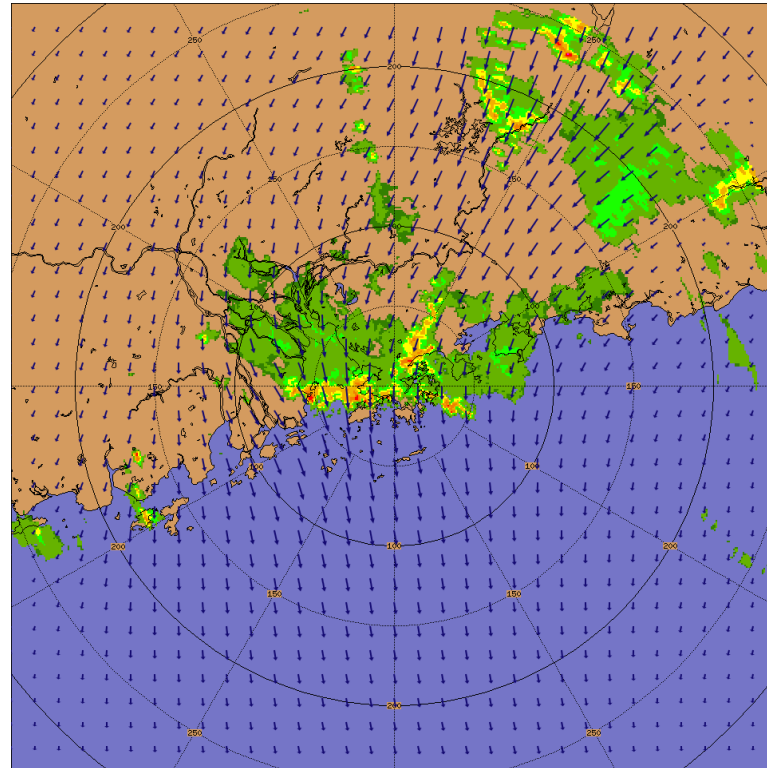


# 120-min Forecast Reflectivity Based at 2018/08/22 16:30

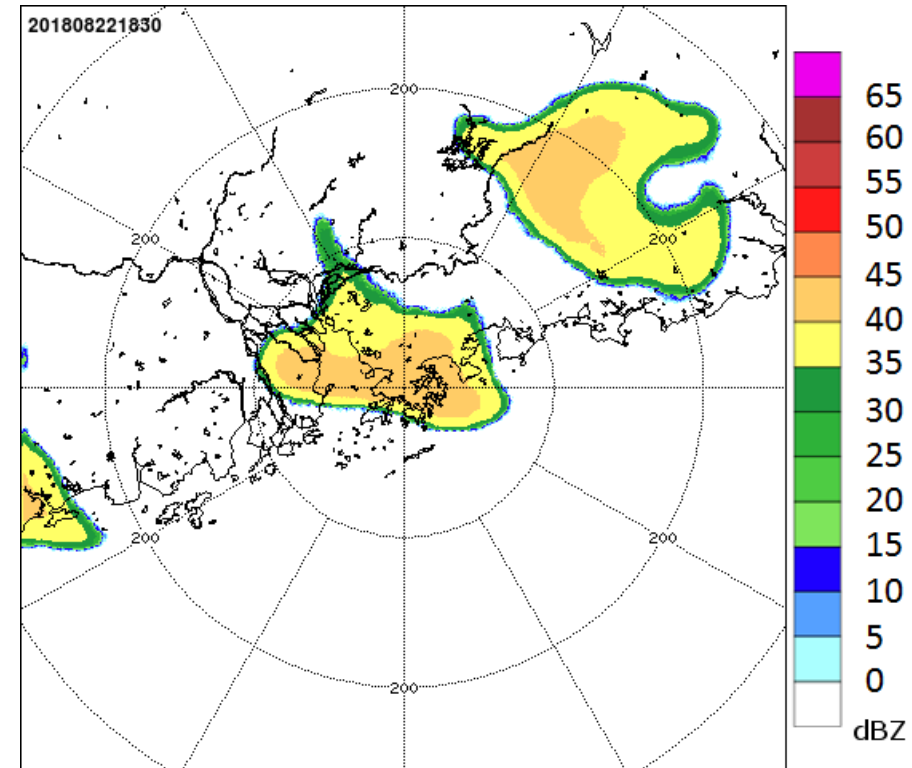
Better capture growth and movement of  
radar echoes



Optical Flow (ROVER)



**Actual**



Deep Learn (TrajGRU)

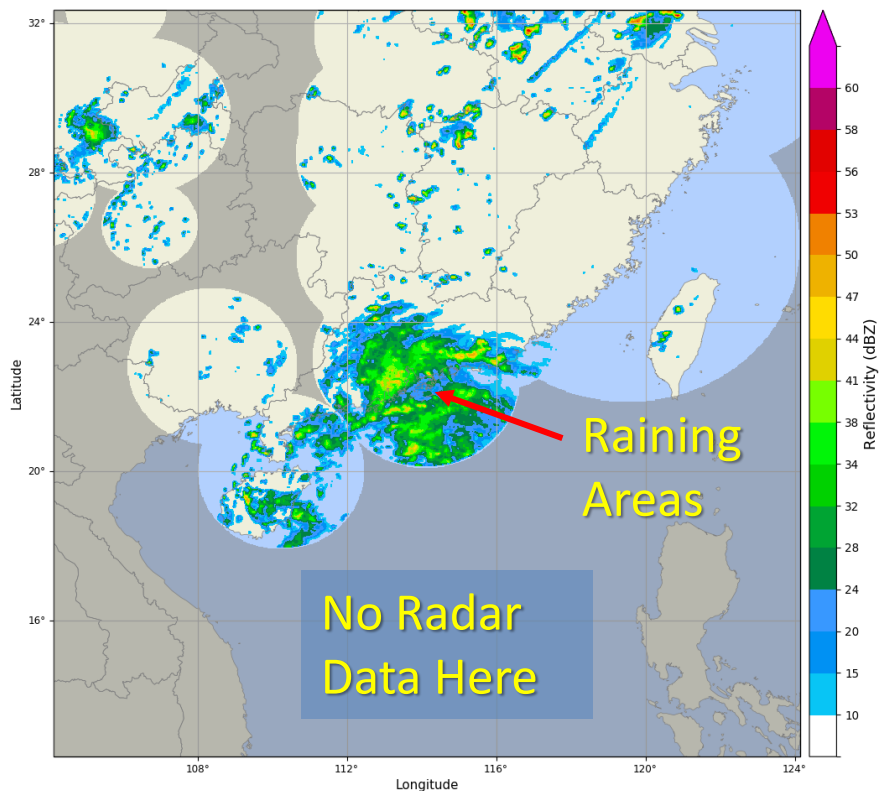


# Satellite Nowcasting

# Retrieval of equivalent radar reflectivity using Himawari-8

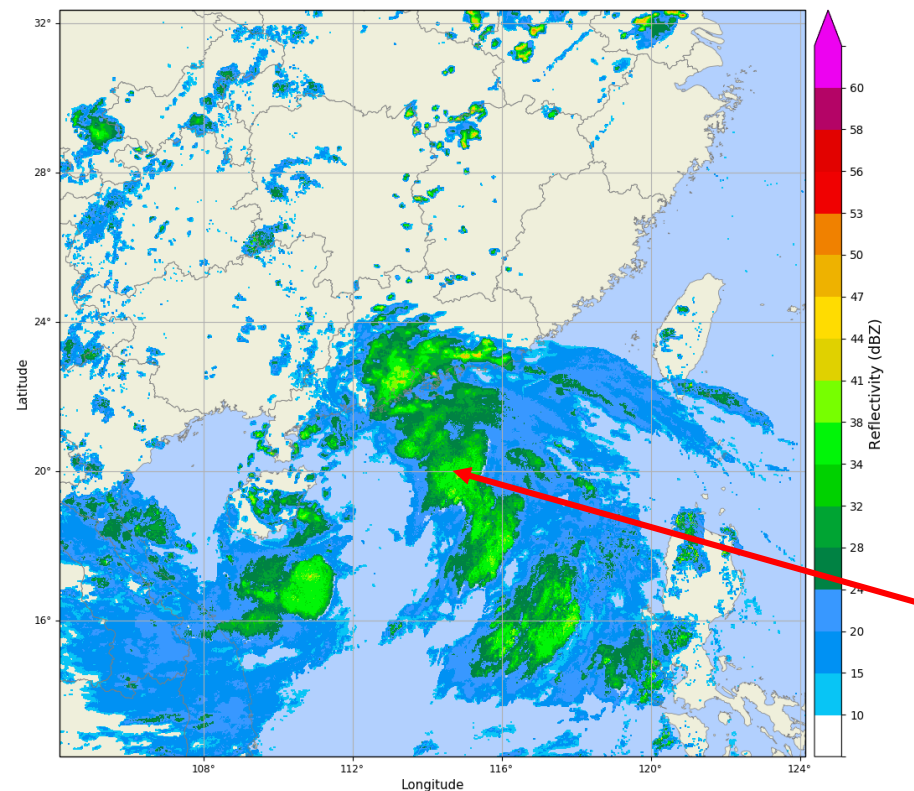
Reflectivity  
CMA Radar / SC

2019-07-31 Based @ 15:00H  
Valid @ 15:06H



Reflectivity  
Multi-Sensor / SC

2019-07-31 Based @ 15:00H  
Valid @ 15:06H



Raining Areas  
Far Away  
Simulated with  
Satellites' Data  
Using Our  
Neural Network

# Spectral Information of Himawari-8 AHI

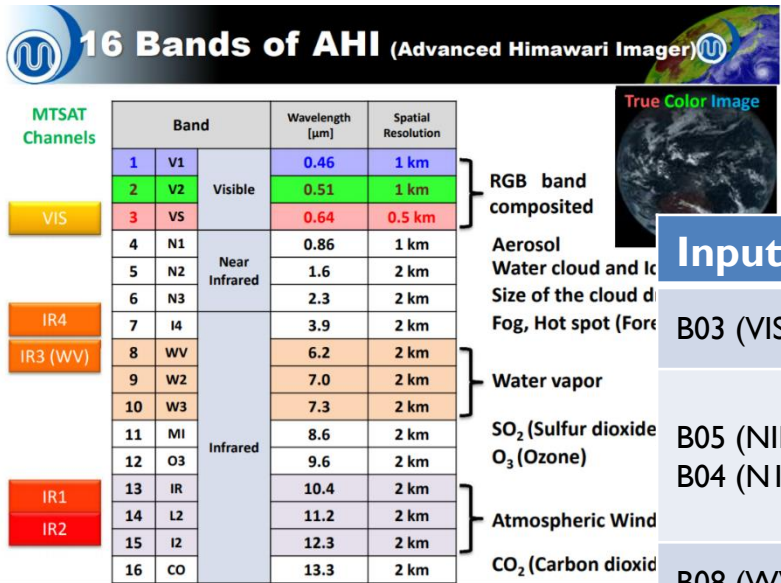
## **16 Bands of AHI** (Advanced Himawari Imager)

MTSAT Channels		Band		Wavelength [μm]	Spatial Resolution	
VIS	1	V1	Visible	0.46	1 km	RGB band composited
	2	V2		0.51	1 km	
	3	VS		0.64	0.5 km	
IR4	4	N1	Near Infrared	0.86	1 km	Aerosol Water cloud and Ice cloud Size of the cloud droplet Fog, Hot spot (Forest fire)
	5	N2		1.6	2 km	
	6	N3		2.3	2 km	
IR3 (WV)	7	I4	Infrared	3.9	2 km	Water vapor
	8	WV		6.2	2 km	
	9	W2		7.0	2 km	
	10	W3		7.3	2 km	
IR1 IR2	11	MI		8.6	2 km	SO <sub>2</sub> (Sulfur dioxide) O <sub>3</sub> (Ozone)
	12	O3		9.6	2 km	
	13	IR		10.4	2 km	Atmospheric Windows CO <sub>2</sub> (Carbon dioxide)
	14	L2		11.2	2 km	
	15	I2		12.3	2 km	
	16	CO		13.3	2 km	

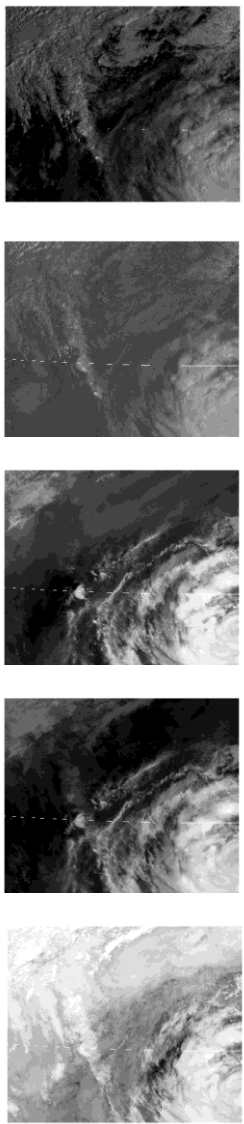


Extracted from  
[“Introduction to Himawari-8”](#),  
 JMA

# Satellite Nowcasting



Input	Description
B03 (VIS0.64)	VIS0.64. Reflectivity of B03 depends on optical thickness. Thick cloud are displayed in white.
B05 (NIR1.6) - B04 (NI 0.86)	Difference between NIR1.6 and NI 0.86. B04 has high reflectivity for snow/ice covered area and clouds, sea surface looks dark. Reflection characteristic of B05 depends on the phase and size of cloud particles. On difference image, thick clouds with large ice particles are displayed in black (dark), low clouds and land/sea surface look whitish (bright)
B08 (WV6.2) - B10 (W3 7.3)	Difference between WV6.2 and W3 7.3. On difference image, thick clouds with high cloud are displayed in white, low clouds and thin Ci are indistinct.
B13 (IR10.4)	IR10.4, Atmospheric window band, available for 24 hours. High-level clouds and developed Cbs appear in white, mid-level cloud appear in bright gray.
B13 (IR10.4) - B15 (I2 12.3)	Difference between IR 10.4 and I2 12.3. Absorption by water vapor of B15 is slightly larger than that of B13. On difference image, thick cloud and low-level cloud contribute to rather grey color, high-level cloud contributes bright color.
Equinox day diff.	Indication of different Season





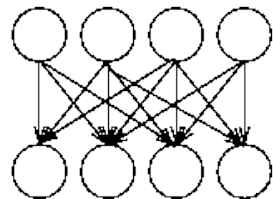
# Multi-layer perceptron artificial neural network (MLPANN)

- Features of MLPANN implemented in HKO:

Neural Network Architecture	Deep neural networks
Training Algorithm	Backpropagation
Learning Strategy	Supervised learning

Supervised learning

Observations (inputs)

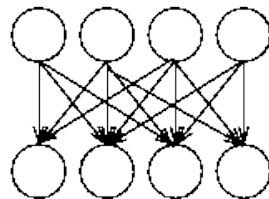


Observations (outputs)

(a)

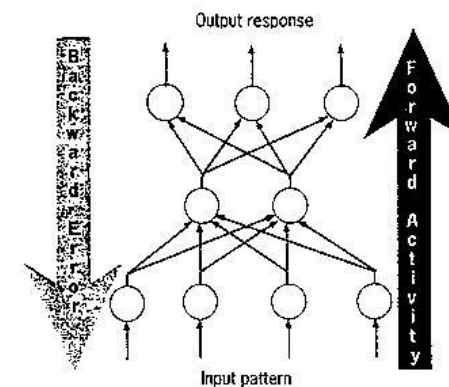
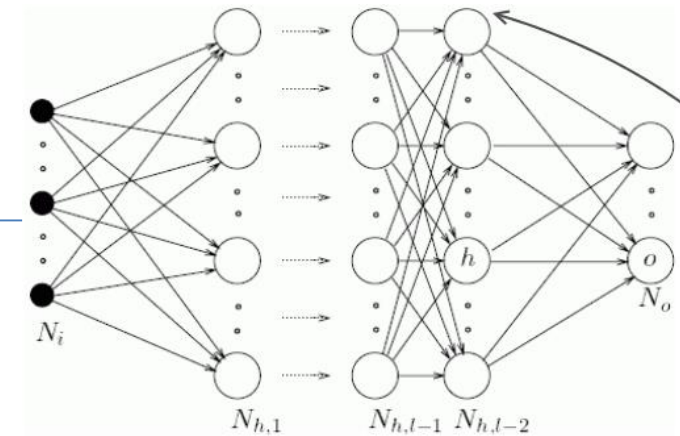
Unsupervised learning

Latent variables



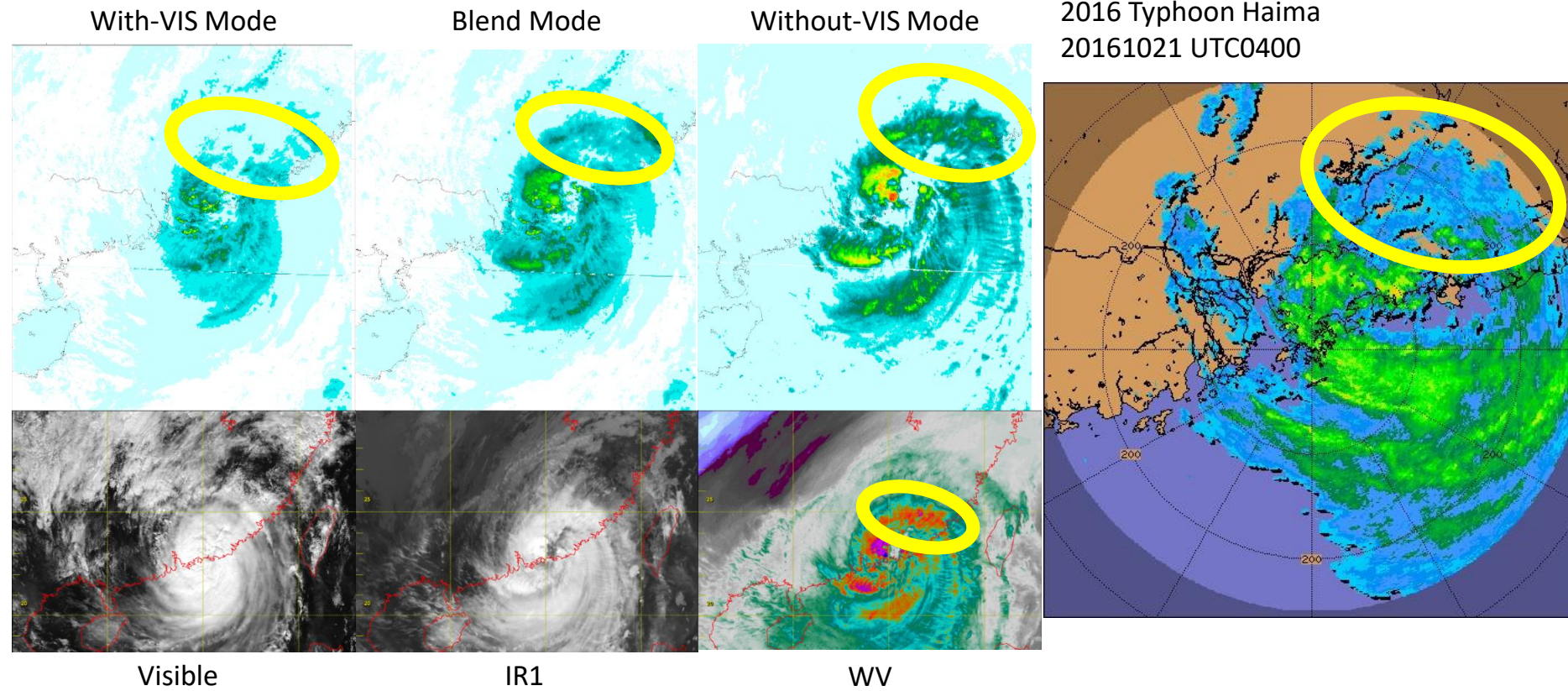
Observations

(b)

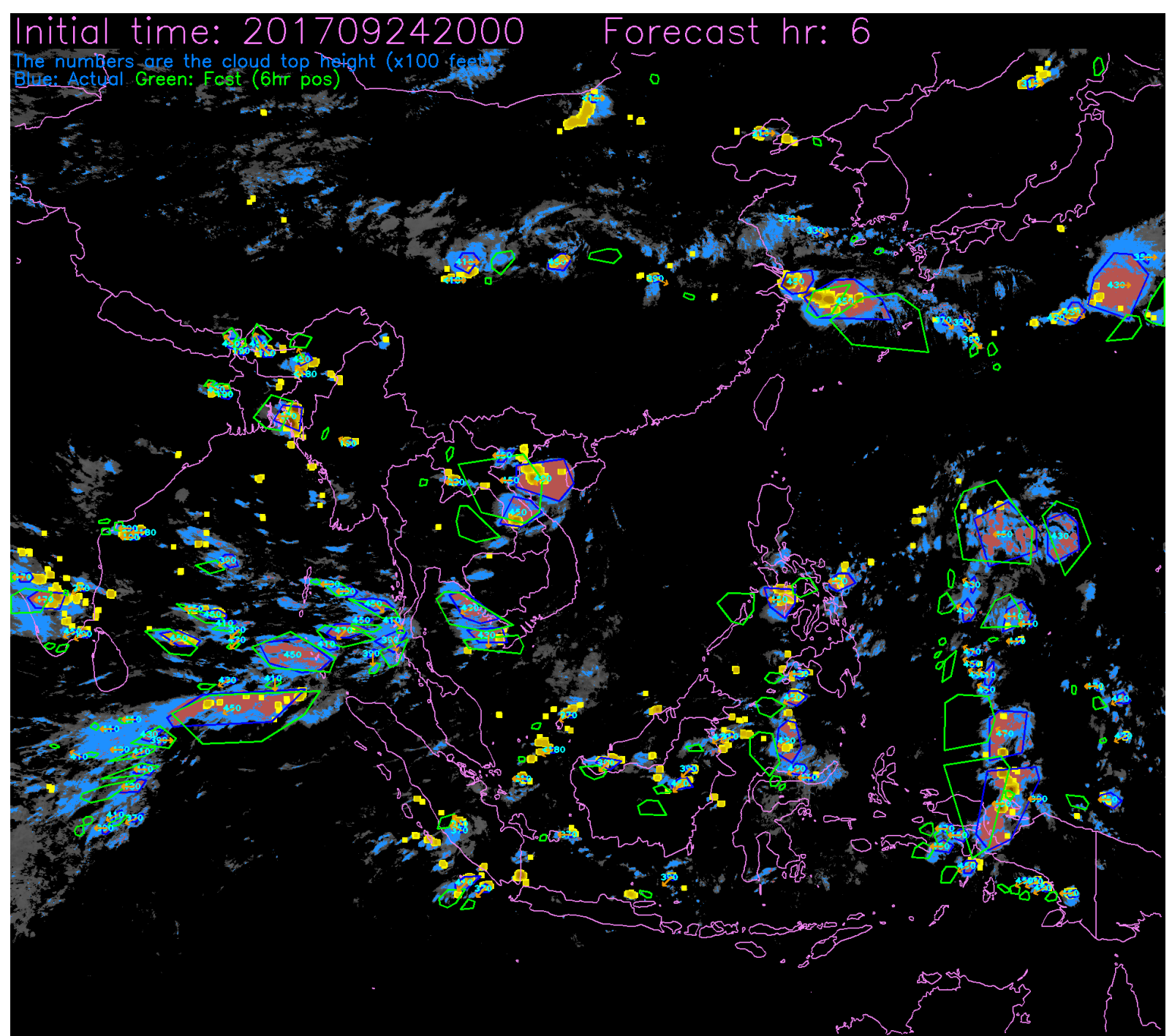


A backpropagation network trains with a two-step procedure. The activity from the input pattern flows forward through the network, and the error signal flows backward to adjust the weights.

# Compare with Available Satellite Product



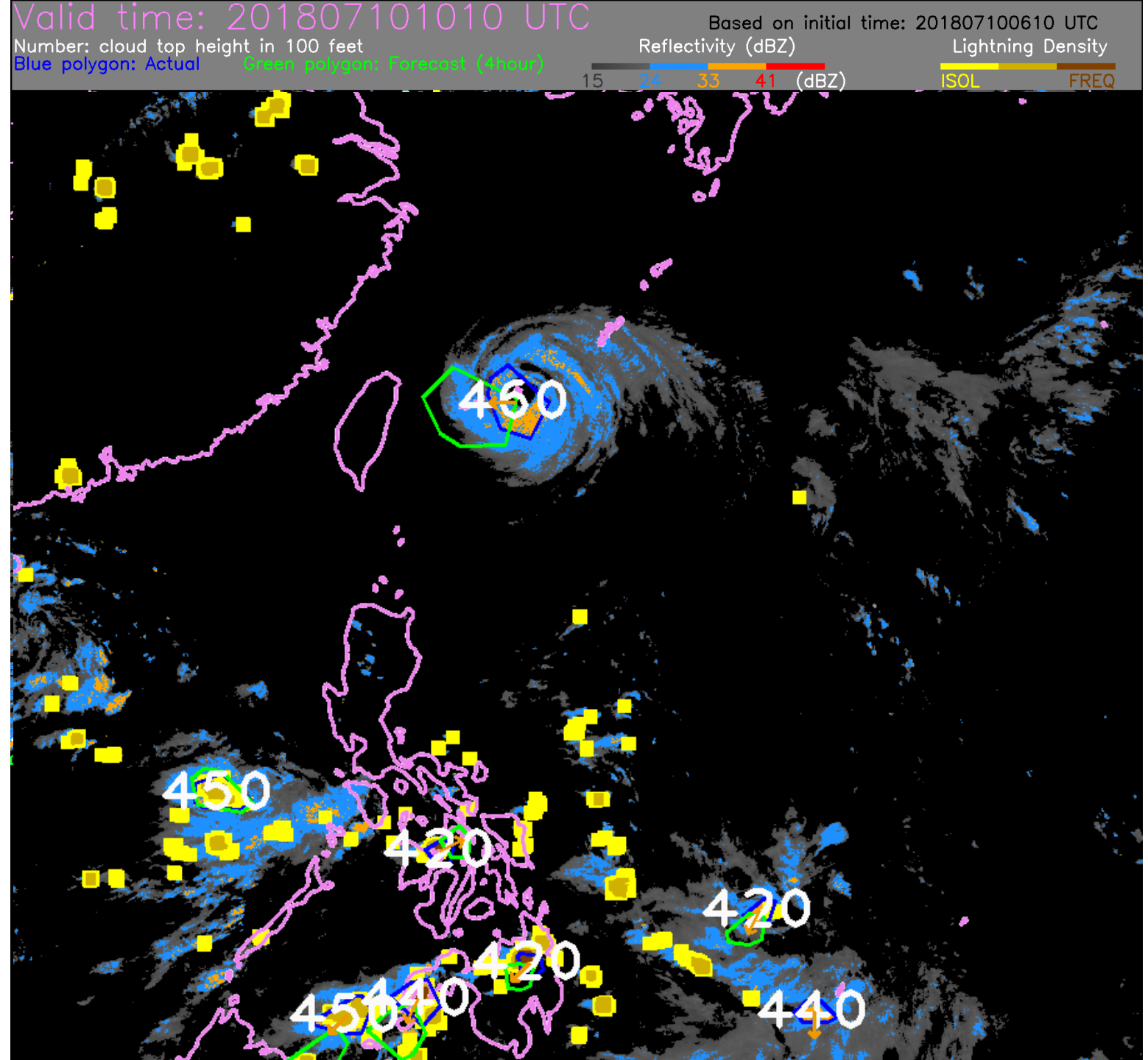
# Significant Convection Nowcast using Himawari-8 retrieved reflectivity available in real-time on HKO's RSMC for Nowcasting website





- Significant Convection Nowcast

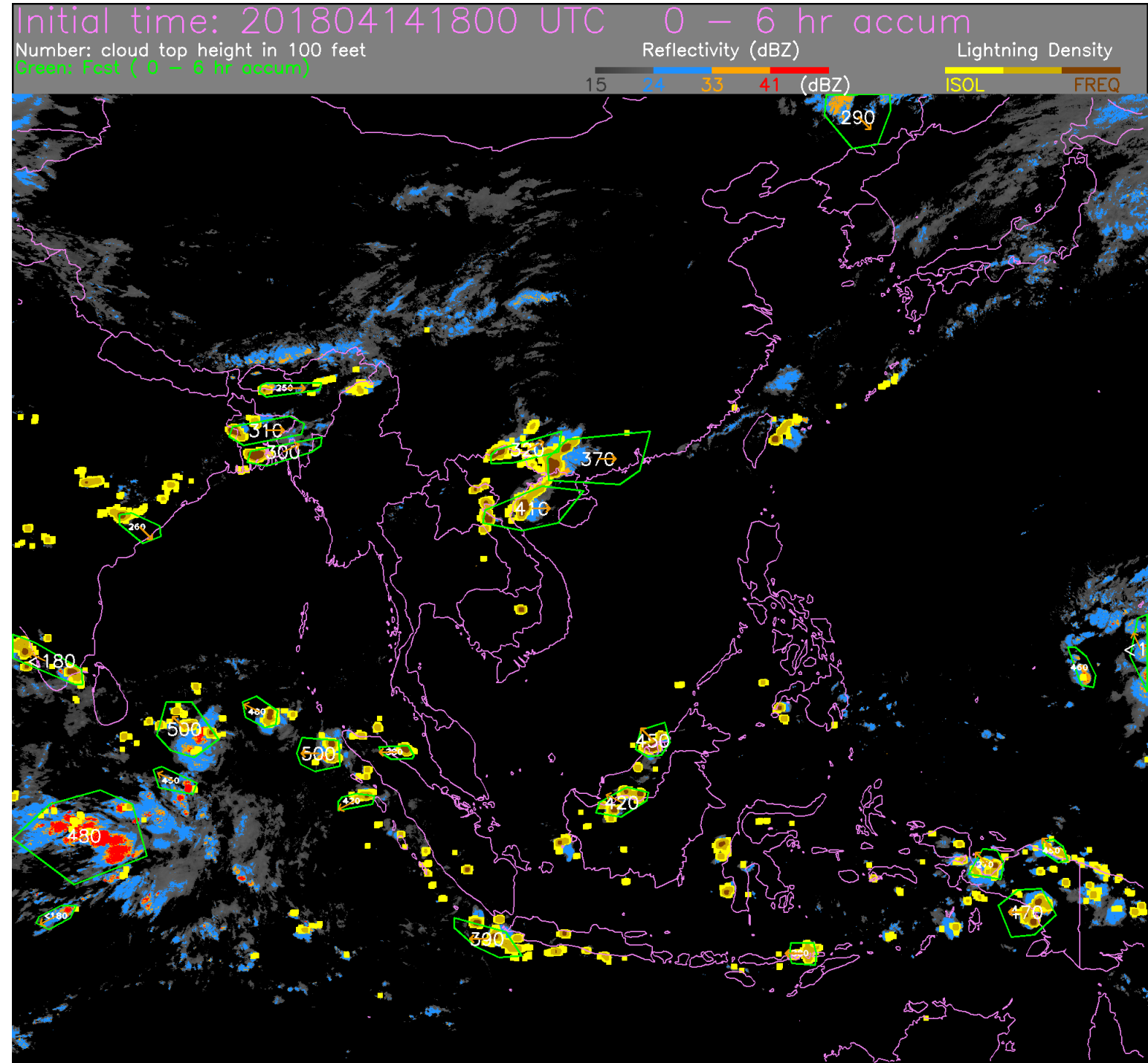
- 0-6 hour nowcast
- Retrieved reflectivity from Himawari-8
- Composite imagery with GDL360 lightning density (yellow/brown)
- Fengyun-4 under development



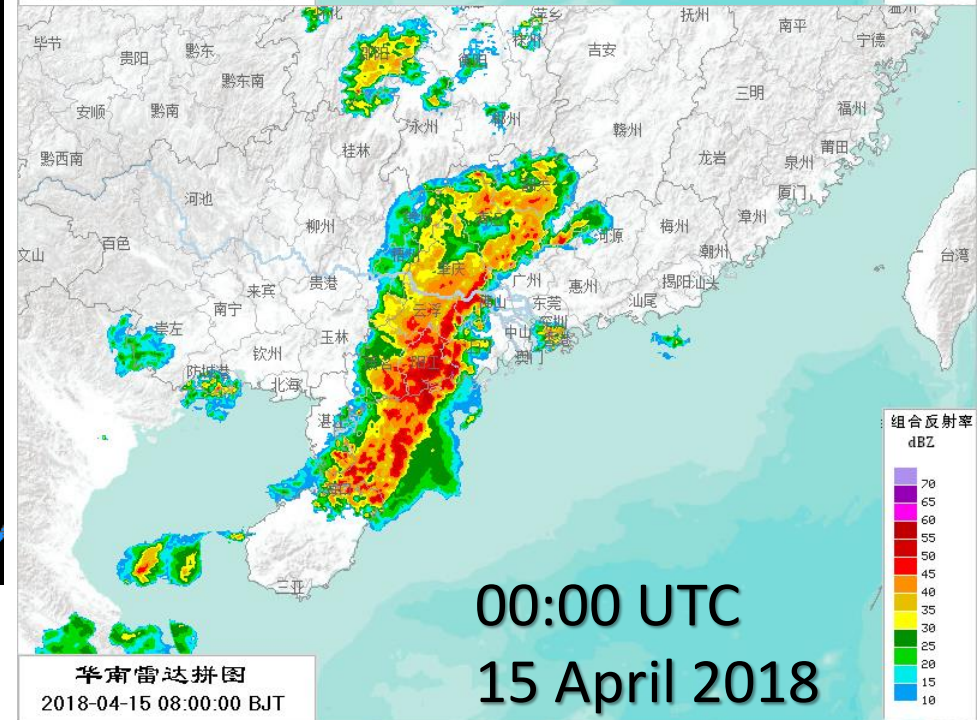
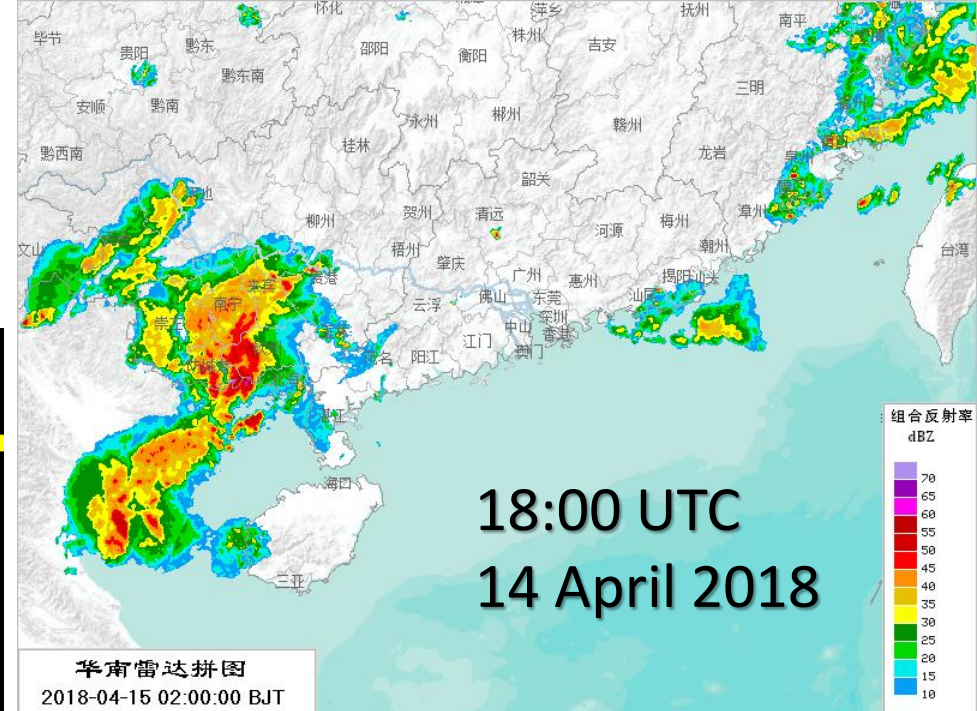
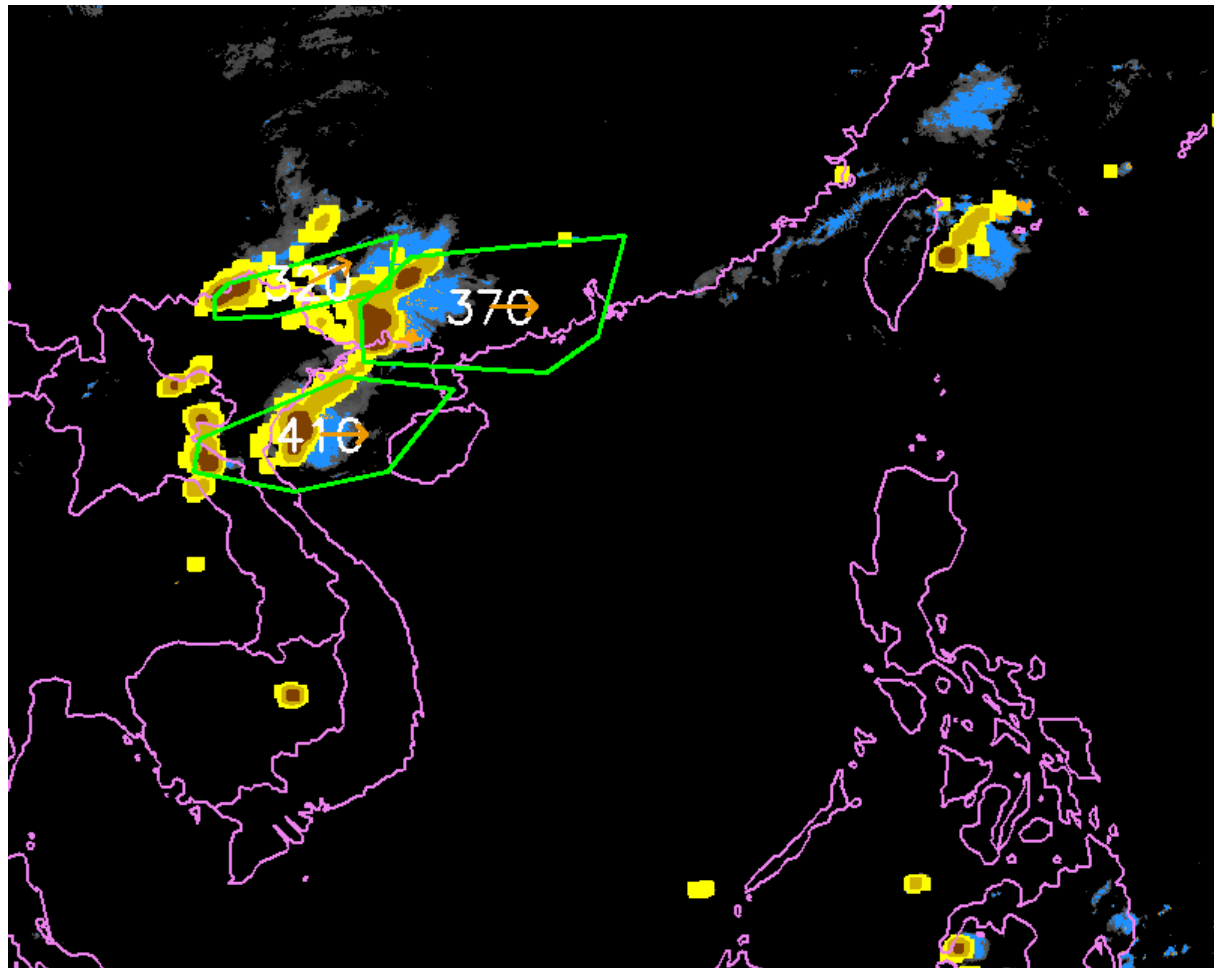


15 Apr 2018

0-6 hr nowcast from  
18:00 UTC



0-6 hr nowcast from 18:00 UTC



Thank You!

**THE END**