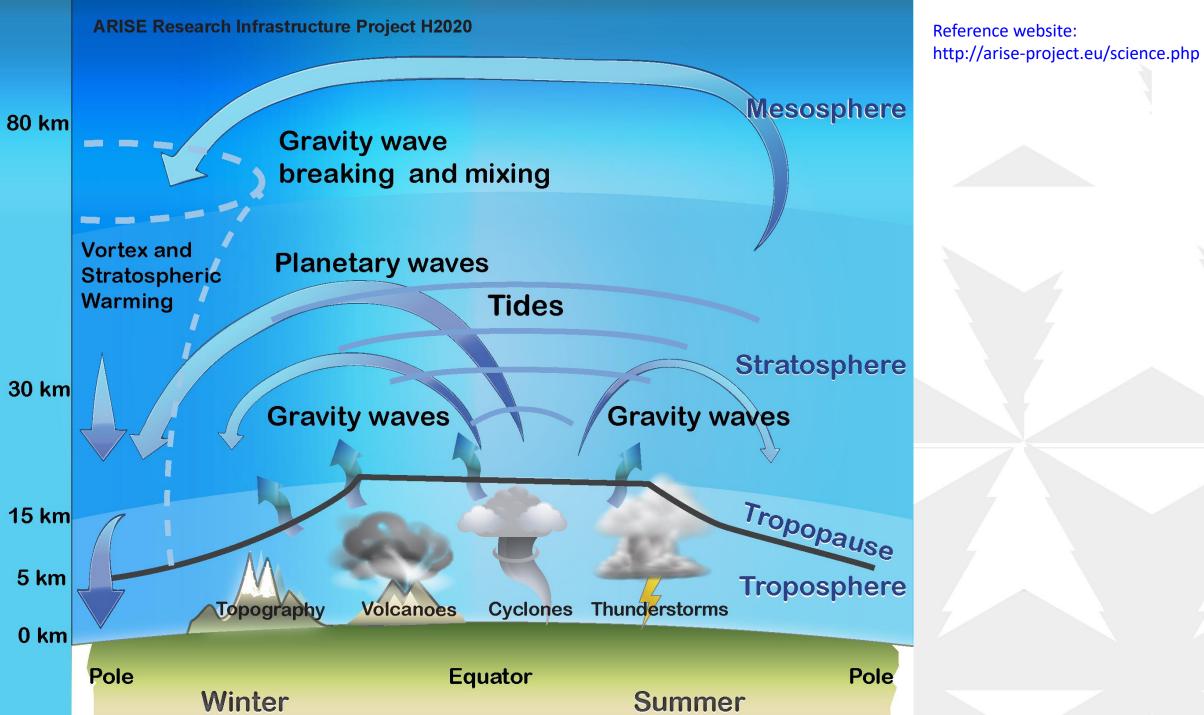
Operational application of deep learning model for auto-detection of Atmospheric Gravity Waves (AGW) over the Asia-Pacific region

12th Asia-Oceania Meteorological Satellite Users' Conference (AOMSUC-12) 11-18 November 2022

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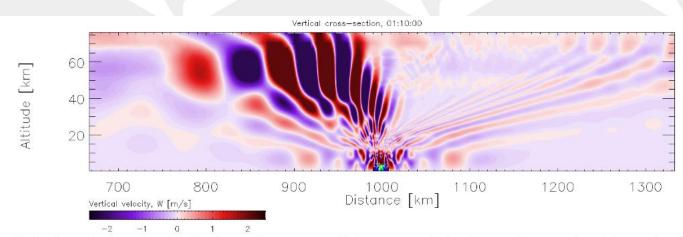
Weather features inducing AGW

- 1. Jet stream
- 2. Orography
- 3. Convection

Note: More pilot reports received from (1) and (2)

General properties of AGW

- Horizontal wavelength (several 10 to 1,000 km);
- Vertical wavelength (several km);
- At mid-latitudes, periods vary from several minutes to about 1 day;
- AGW can propagate vertically and transport momentum and kinetic energy upwards.



Small scale atmospheric waves, usually referred to as gravity waves, are an efficient transport mechanism of energy and momentum through the atmosphere. They propagate upward from their sources (flow over topography, thunderstorms, jet adjustment, etc.) to the middle and upper atmosphere. At a given altitude, gravity waves manifest as concentric rings. Depending on the horizontal wind shear, they can dissipate at different altitudes and force the atmospheric circulation of stratosphere and mesosphere.

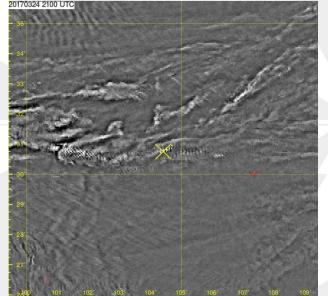
Reference website: http://arise-project.eu/arise-results.php

Identifying AGW from Satellite Images

- Use Himawari-8 (H-8) 6.2μm water vapour channel images;
- Apply Gaussian high-pass filtering^{Ref.} for highlighting AGW, and variations of -1K to +1 K of brightness temperature from the local average are displayed.

 $I_{\rm hp} = {\rm HP}(\sigma)^* I$,

HP(
$$\sigma$$
) = 1 - $\frac{c}{\sigma\sqrt{2\pi}}\sum_{x,y}e^{-(x^2+y^2)/2\sigma^2}$,



Ref.: Wimmers A., S. Griffin, J. Gerth, S. Bachmeier and S. Lindstrom, 2018 "Observations of Gravity Waves with high-pass filtering in the new generation of Geostationary Images and their relation to aircraft turbulence", *Wea. Forecasting*, **33**, Issue 1, 139-144. https://doi.org/10.1175/WAF-D-17-0080.1

Building Deep Learning Model

1. Data collection

1. Architecture selection

1. Model streamlining

Inference

(deployment)

2. Data annotation

2. Model configuration

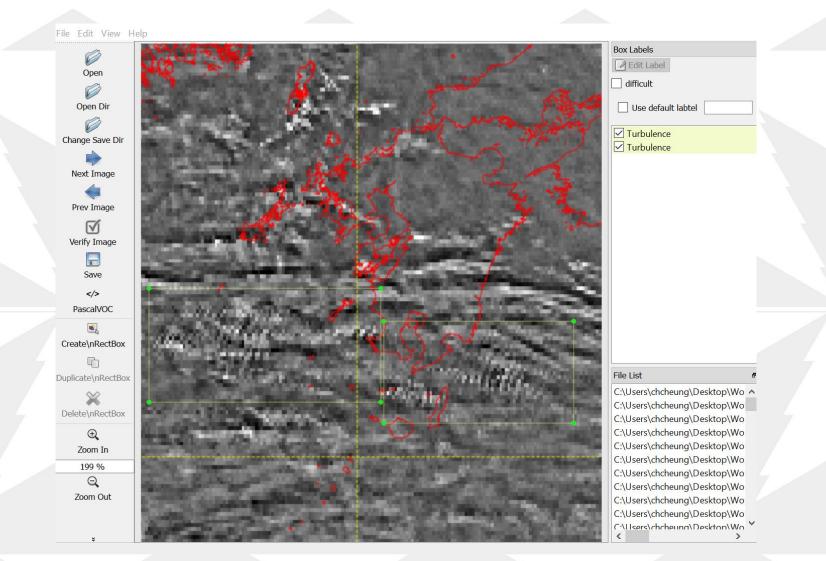
2. TensorRT optimization

- 3. Data augmentation
- 4. Generation of
 - **TensorFlow Records**

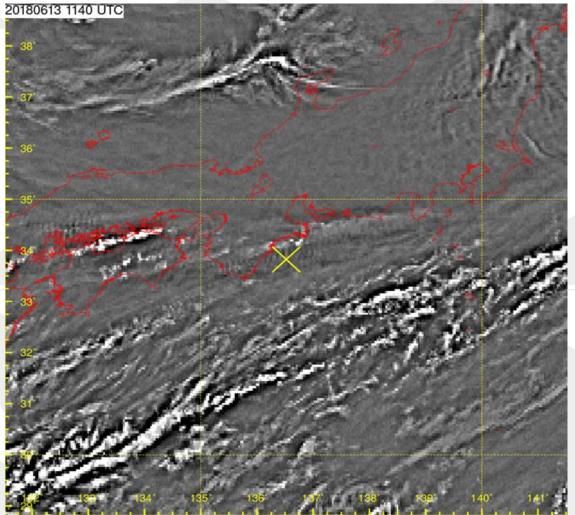
Data Preparation

- Collect over 750 pilot reports from Jan. 2018 to Jun. 2021;
- Generate high-pass filtered images based on T ± 30 min.
 from the observation time (T) of the pilot reports;
- Identify AGW rectangular boxes (Human Truthing);
- Classify AGW intensity.

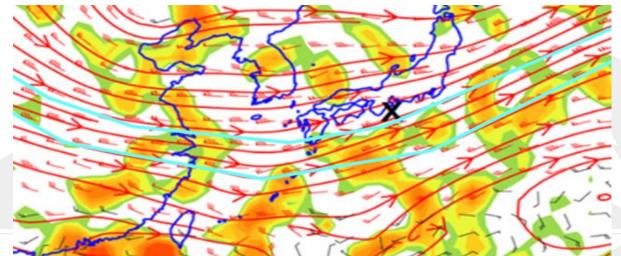
Data Annotation



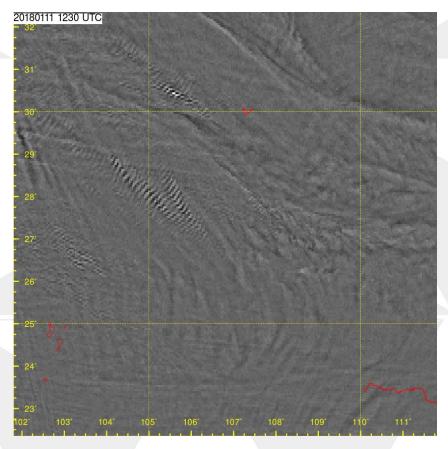
Classification



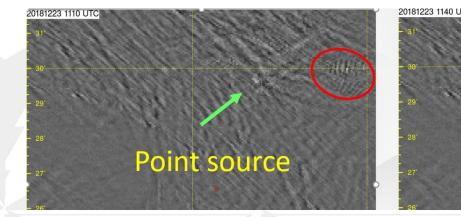
Severe Turbulence



Classification



Significant Severe Turbulence



Merging of gravity waves or waves displaying Herringbone patterns

Curve-shape

wavefront

Point source

Larger contrast in brightness temperature in regions with overlapping wavetrains

AGW events showing Herringbone pattern



Two characteristics coincident with mountain wave turbulence.

- Mountain waves with higher amplitudes (>5K)
- Mountain waves that displayed Herringbone patterns

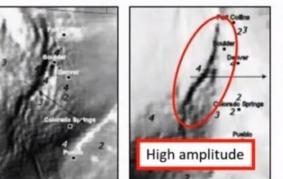


Case observed in Australia

Case observed in Colorado U.S.A.

images from Uhlenbrock et al. 2007

Turbulence conditions for mountain waves

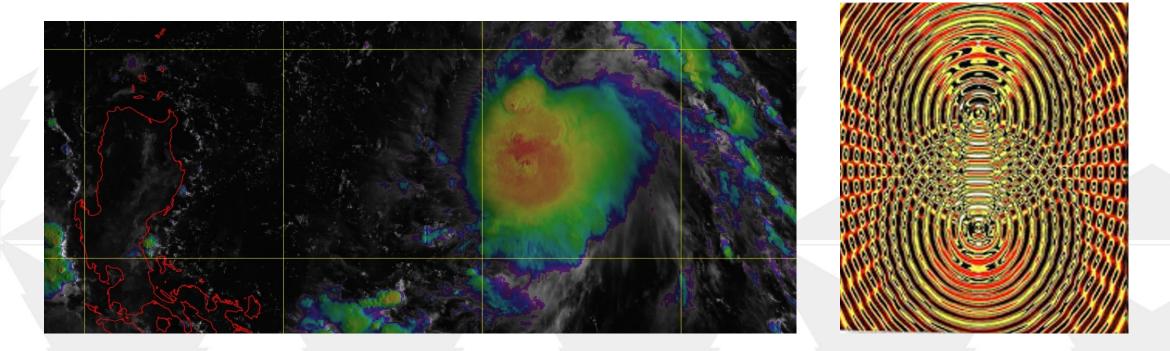




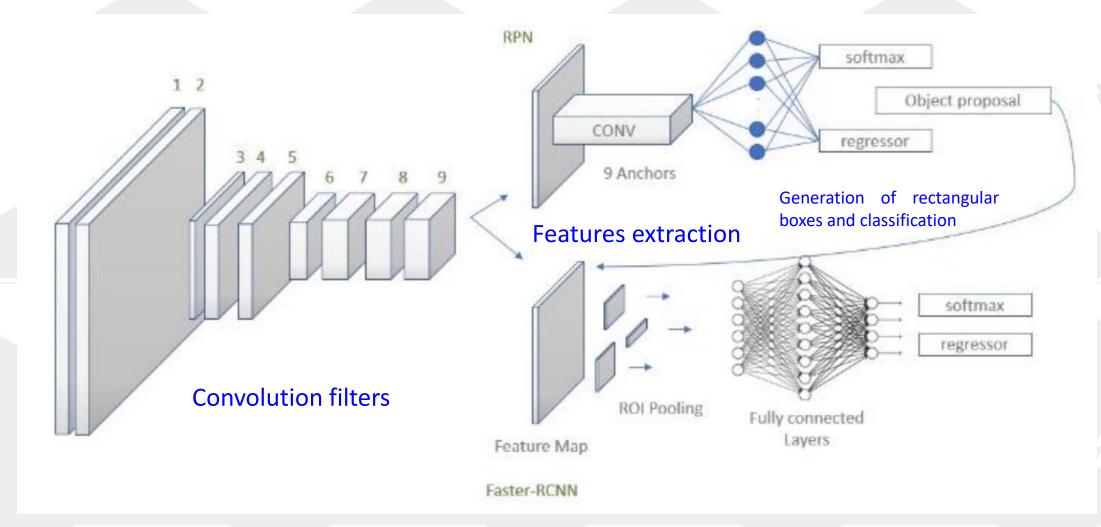
Ref.: Uhlenbrock N.L., K.M. Bedka, W.F. Feltz and S.A. Ackerman, 2007 "Mountain wave signatures in MODIS 6.7μm Imagery and their relation to pilot reports of turbulence", *Wea. Forecasting*, **22**, Issue 3, 662-670. https://doi.org/10.1175/WAF1007.1

Classification

Convection-induced Turbulence

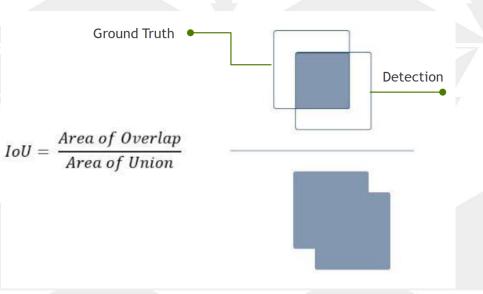


Model Architecture: Faster Region-based Convolutional Neural Network (Faster RCNN)



Model Training and Testing

- 80% of dataset collected for training and 20% for testing (randomly selected);
- Calculate model precision (IoU: Intersection over Union);
- After model training, IoU_{0.5} = 0.7182;



Verification

- Pilot reports in Jul. Dec. 2021 were extracted for verification.
- Reports seem to follow a flight route are removed.
- Reports with severe turbulence at low level say 500 ft are removed
- 38 reports with flight levels at least 15,000 ft are used for verification
- High-pass WV images ± 10 minutes from the reported time was counted.

Results

	Location of severe turbulence within ± 1° of boxes	Location of severe turbulence within ± 2° of boxes
No. of reports = 38	21	33
Percentage (%)	55	87

Post-processing

- Rectangular boxes in the last 30 minutes were also shown to enhance continuity;
- Boxes were filled in colours but transparency was added to enable visualization of high-pass filtered satellite images on the back.

Current Operational Setting

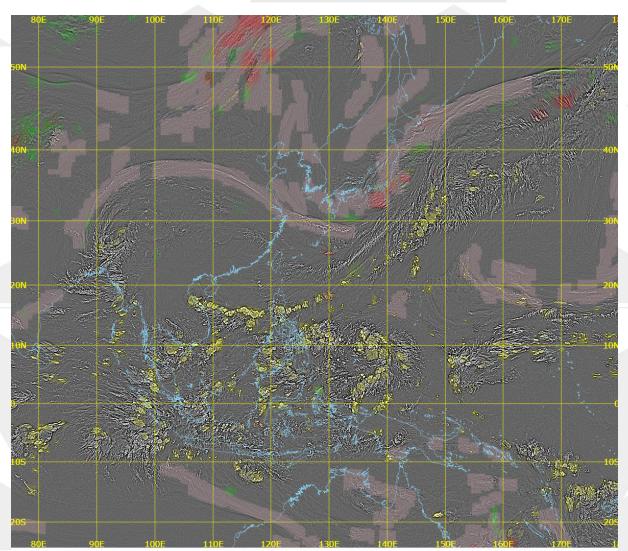
Severe Turbulence

Significant Severe Turbulence

Convection-induced Turbulence

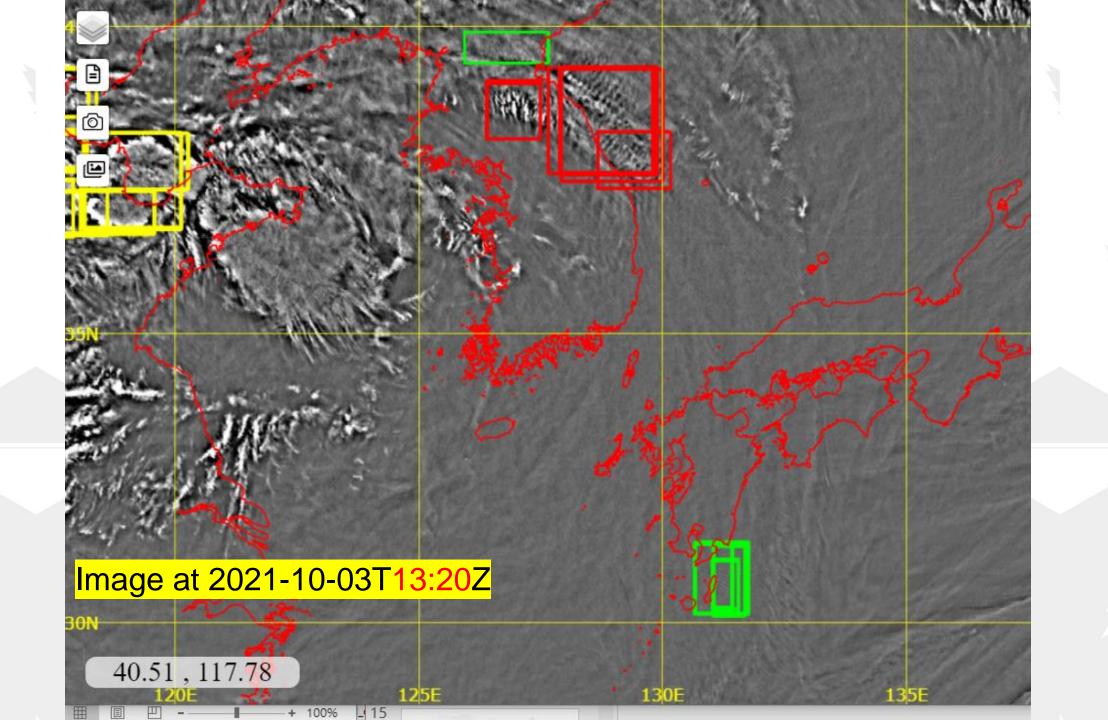
Tropopause Folding Turbulence Detection (TFTD)

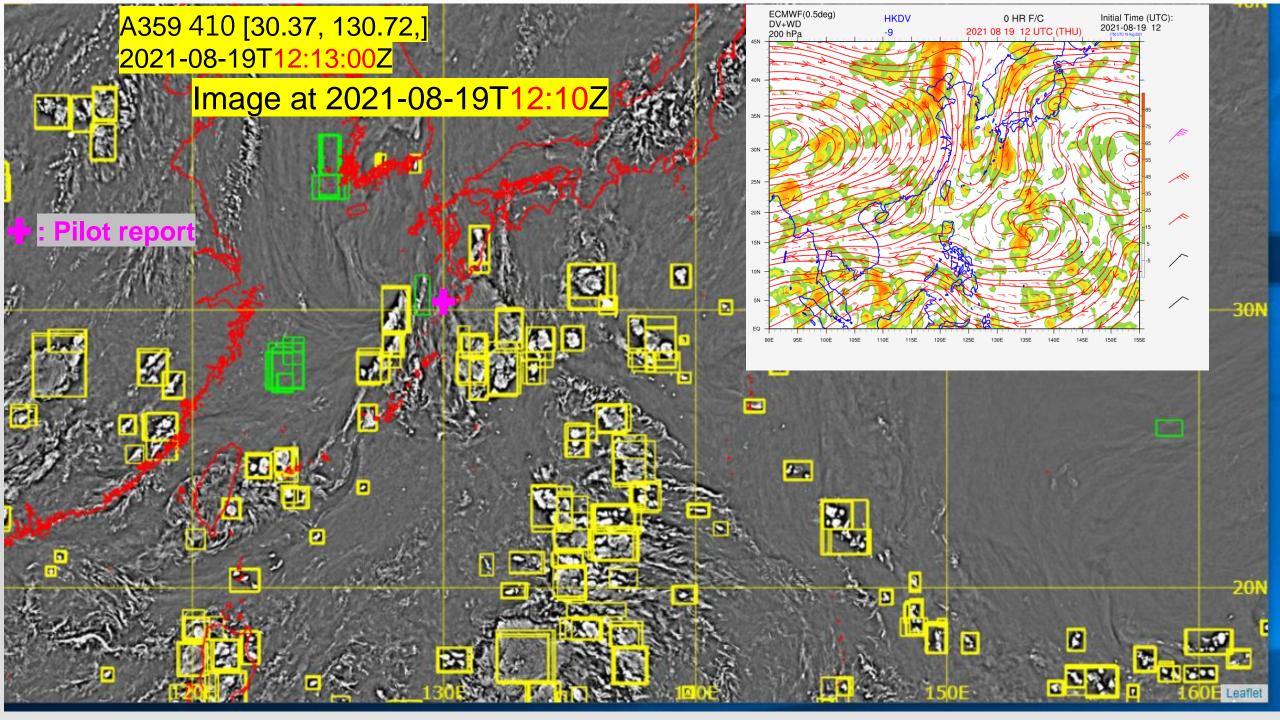
- Model output to WebGIS display;
- GK2ATropopauseFoldingTurbulenceDetection(TFTD)productswere alsodisplayed.

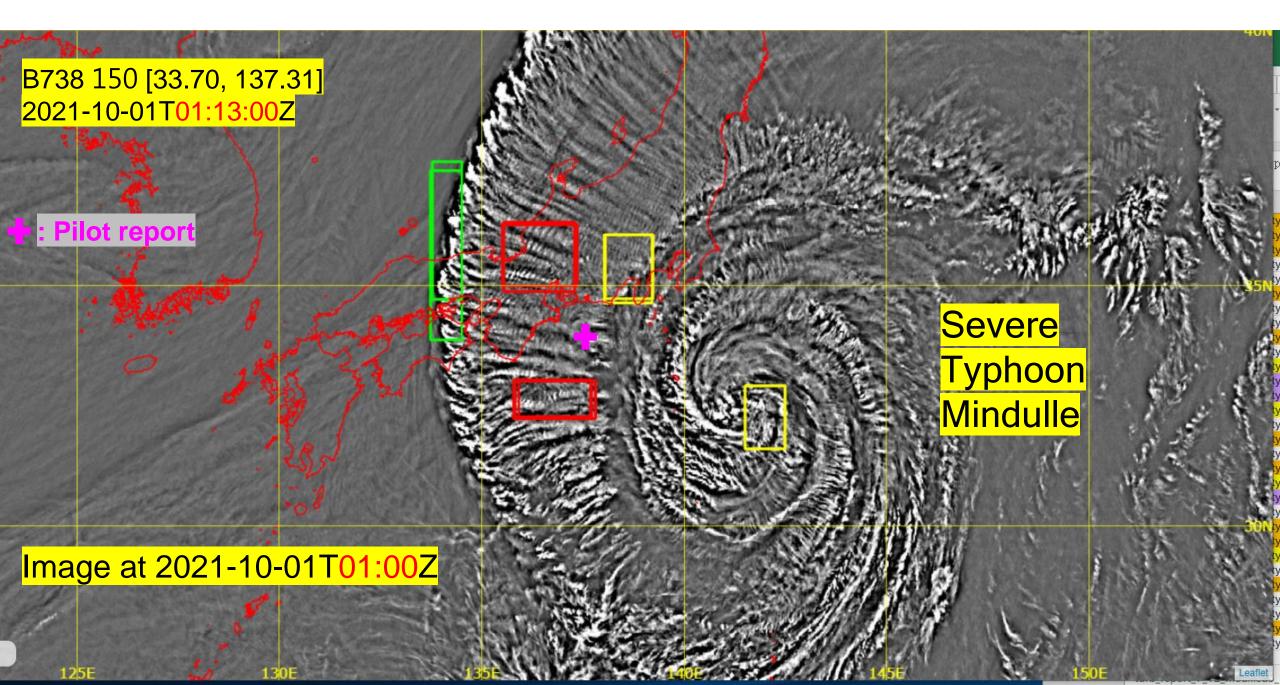


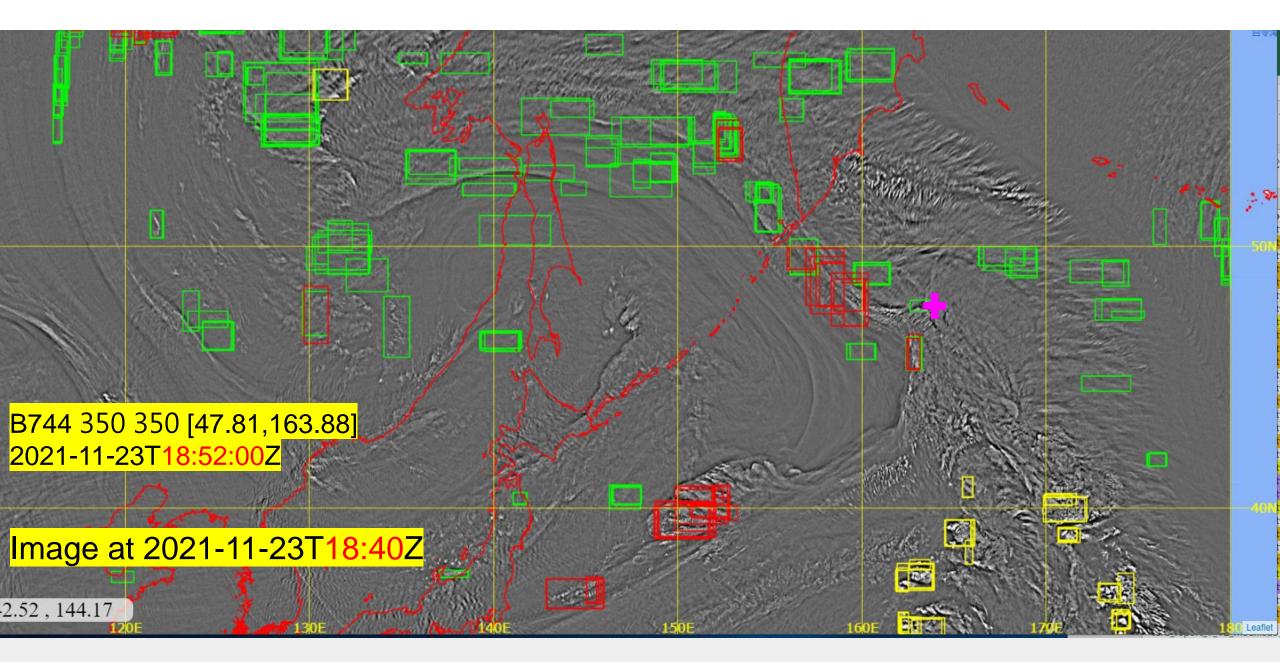
H8 WVFFT-L-u3-tftd 2022/10/25 00:00 UTC

Beautiful Catch

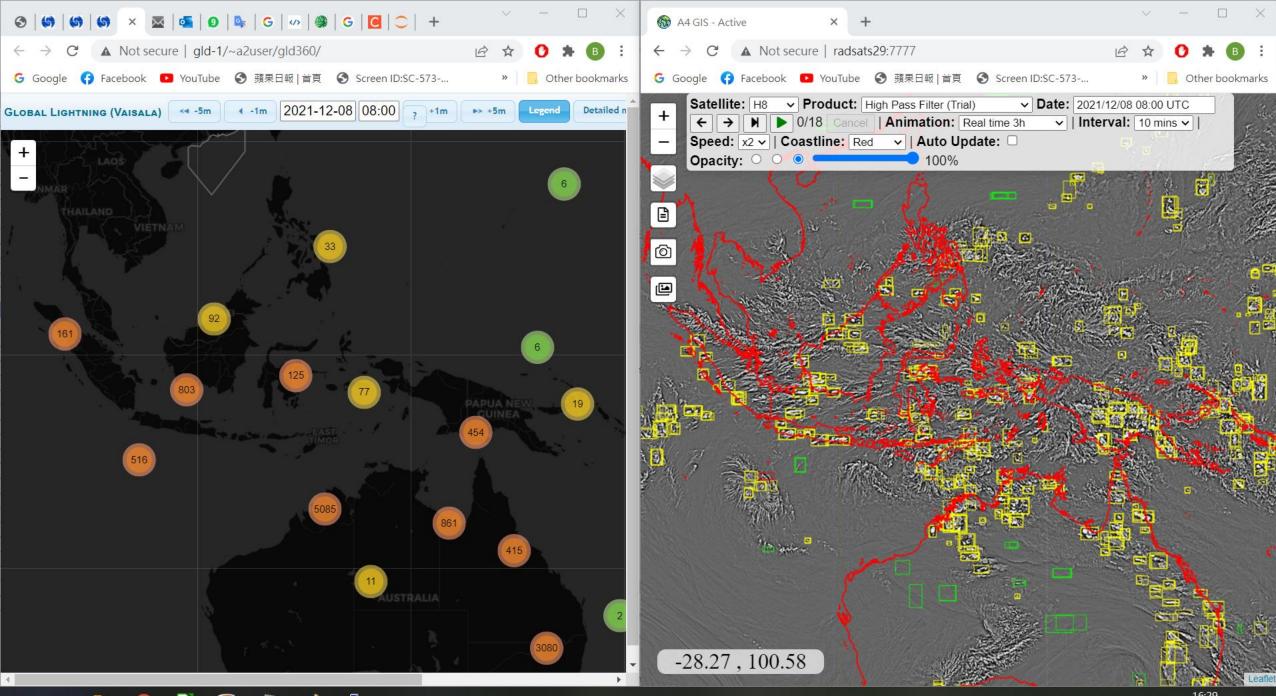


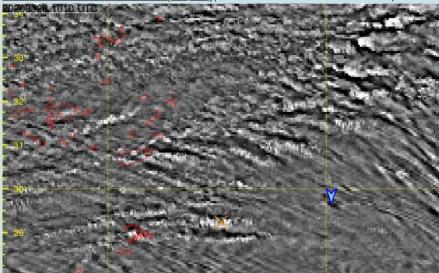






Additional Values



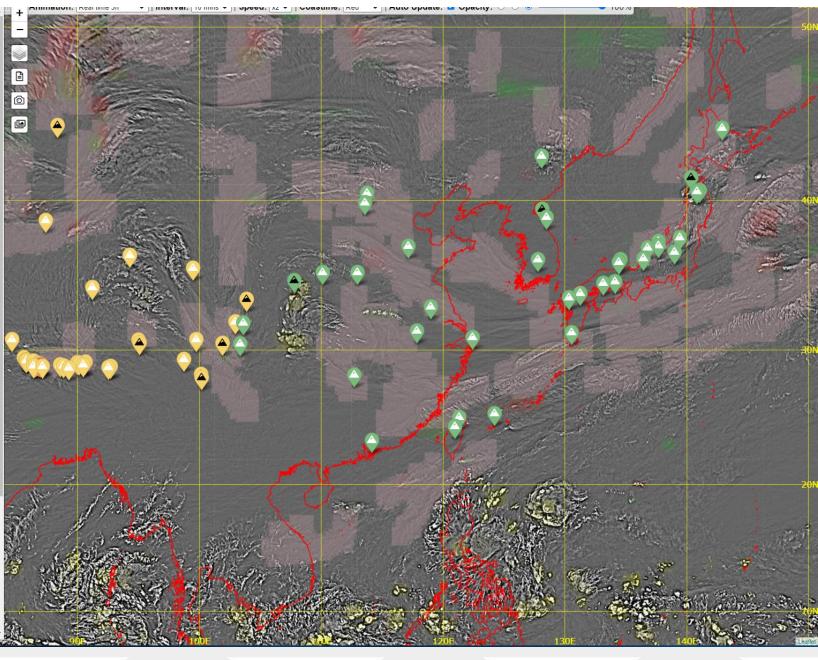


Past Turbulence Case in 2020/03/28 10:00 UTC at [29.23, 92.59] (Position shown as X in the satellite animation below)



×

<u>Mountain Info:</u> Name: Namcha Barwa Height: 7,782 m Lat/Long: 29.62, 95.05 (Please see V in the satellite animation below)



Conclusion and Future Work

- Verification shows reasonably good model skills
- Those yellow boxes possibly reflect lightning locations
- AGW casebook as educational materials
- Identify flight levels affected by AGW events

Thank You !