



vEGU21

Improvement of the early detection and quantitative risk prediction method with the three-dimensional wind field from multiple-doppler radar analysis

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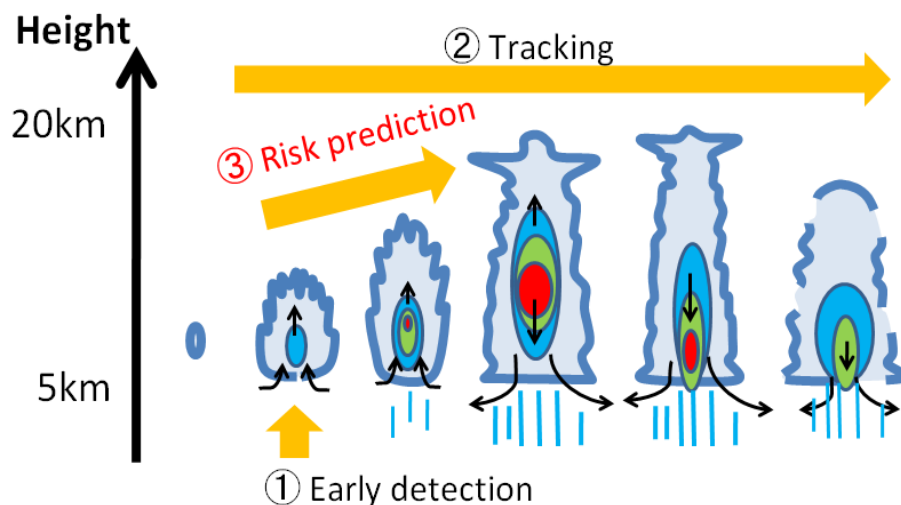
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- Objective**
- Study Area and Dataset**
- Improvement of the quantitative risk prediction**
- Conclusion and Future work**

Background

□ Background

- Nakakita et al.(2013) developed a **prediction system** for localized heavy rainfall by using **pseudo vorticity** in a first echo aloft in 3-D volume scanning data of X-MP
- The current system could predict the Guerrilla heavy rainfall whether peak rainfall intensity would exceed **50 mm/hr or not in 30 min**; only **two risk categories**
- Advanced warning at least 10 minutes before provided by early detection is critical to **saving lives** -> **the early detection and quantitative risk prediction method**



Nakakita et al.(2013)

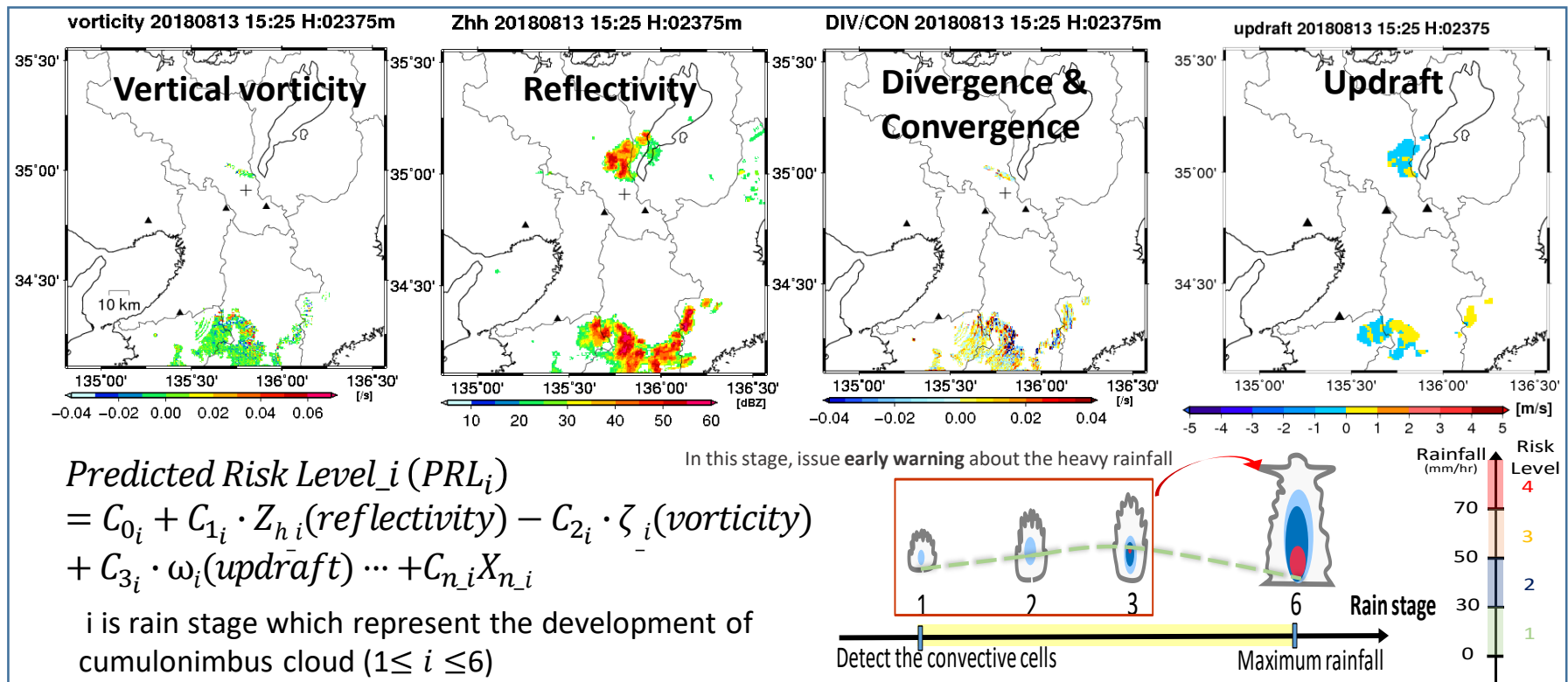


Flash Flood occurred during 30 min caused by Guerrilla heavy rainfall

Objective

Objective

- If the the early detection and quantitative risk prediction method was considered for each rain stage, the method estimate more fitted risk level
- By using the multiple-doppler radar analysis, we can find the reliable wind field reflecting natural environment



Study Area and Dataset

Study Area and Dataset

➤ Kinki region in Japan

- The **four X-band polarimetric RAdar Network (XRAIN)** are used, named as Rokko, Katsuragi, Juubusan, and Tanokuchi
- The three-dimensional volume scan data has the resolution of **250 and 500 m** in the horizontal and vertical directions, respectively
- From August 2013 to August 2018, 12 GHR events are selected

3D observation time	5min
Observation range	80km
No. elevation angles	12
Azimuth	300

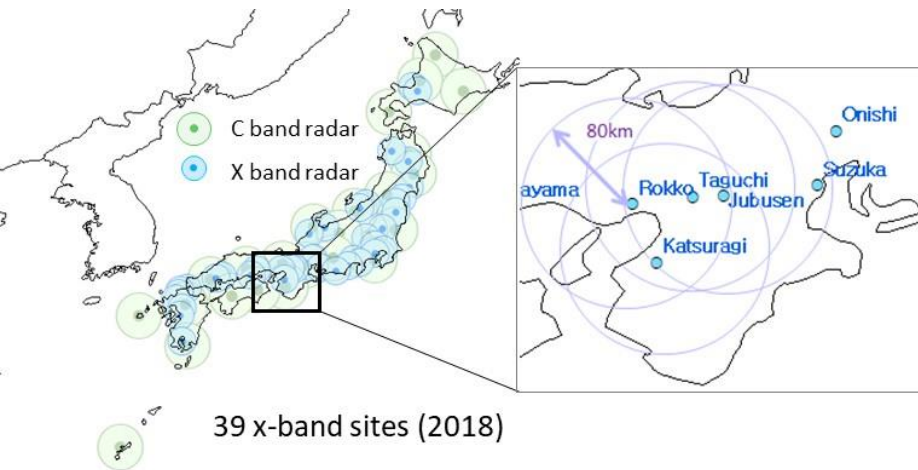


Table 2 List of Guerrilla heavy rainfall events which selected 10 events for calibration and 2 events for verification

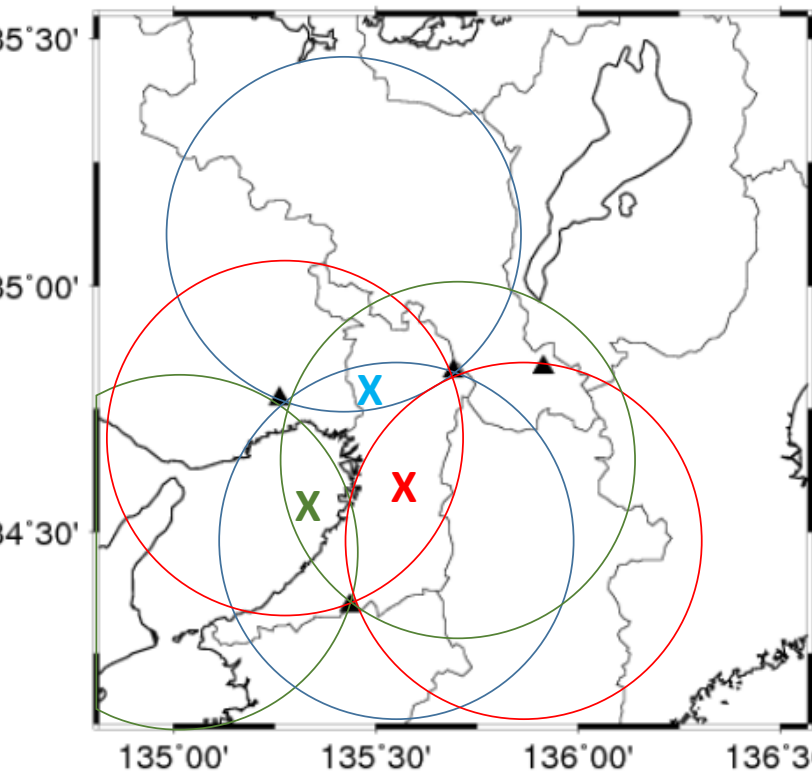
No.	Date	Longitude	Latitude	Time of Early detection	Time of maximum rainfall	Maximum rainfall intensity (mm/h)
1	2013-08-06	135.981	34.469	16:10	16:30	53.3
2	2013-08-07	135.747	34.930	16:50	17:05	191.1
3	2014-08-23	136.159	34.266	11:40	11:45	50.5
4	2014-08-23	135.584	34.240	11:40	11:50	47.7
5	2015-08-07	135.872	34.609	17:20	17:35	160.0
6	2015-08-29	135.752	34.177	14:35	14:40	145.2
7	2016-08-03	135.818	34.573	18:35	18:50	119.9
8	2016-08-25	135.660	34.267	13:15	13:30	117.3
9	2017-08-04	135.611	34.510	16:25	16:35	60.9
10	2018-08-13	135.353	34.992	15:30	15:50	94.2
1	2018-08-13	135.353	34.992	15:30	15:50	94.2
2	2018-08-28	135.900	34.772	12:05	12:15	82.7

Improvement of the quantitative risk prediction

Three-dimensional wind field from multiple-doppler radar analysis

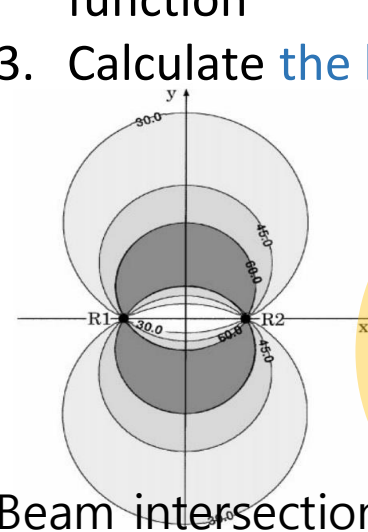
- The Vorticity, Divergence&Convergence is needed to calculate **wind vector**. Also, the vertical wind represents the **updraft** and **downdraft**

One of the method to calculate the 3D wind vector : **multiple-Doppler radar analysis**

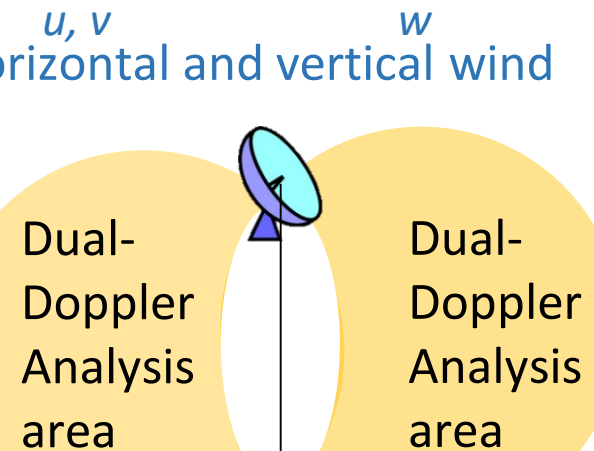


protat et al. (1999) & Shimizu and Maesaka (2006)

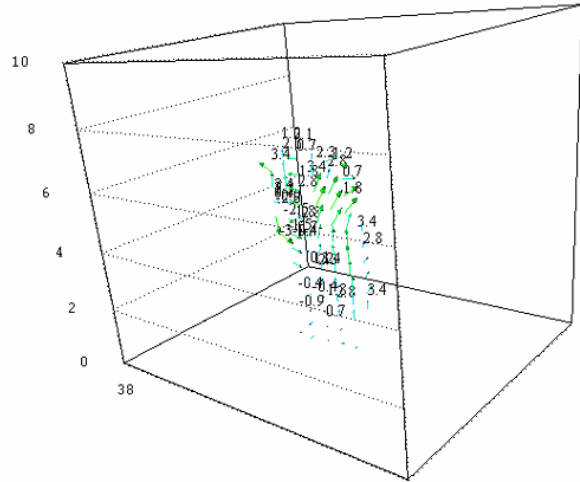
1. Use radial velocity of two radar in CAPPI data
2. Use 3-D variational method to optimize cost function
3. Calculate **the horizontal and vertical wind**



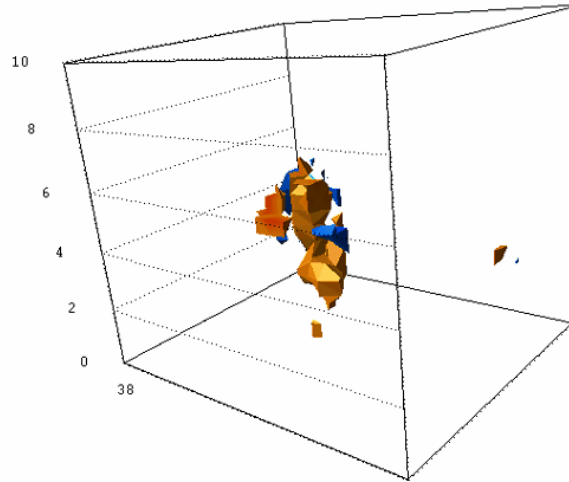
Beam intersection angle (θ) $> 20^\circ$



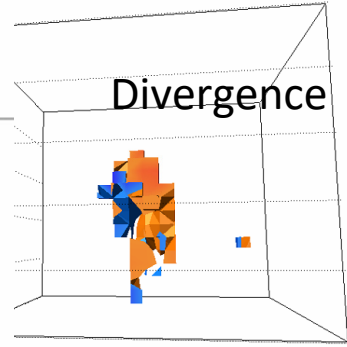
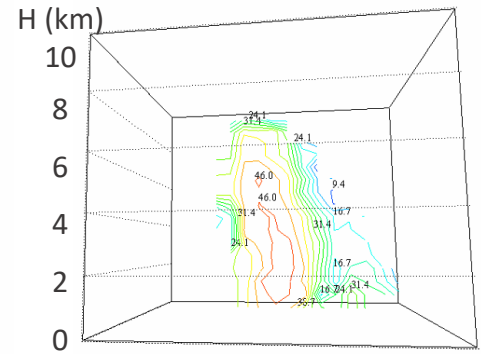
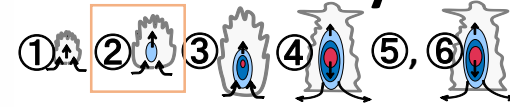
Three-dimensional wind field from multiple-doppler radar analysis



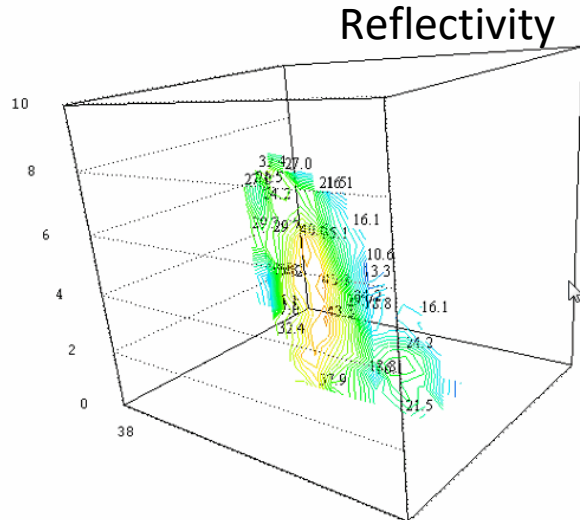
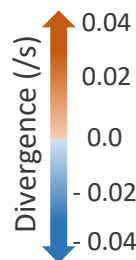
Wind direction



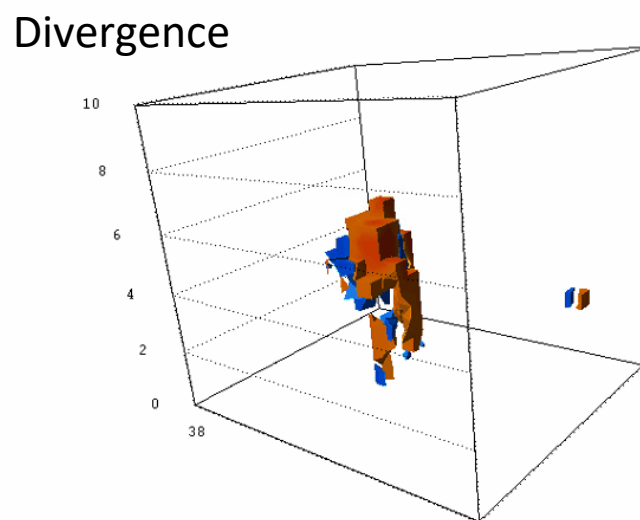
Vorticity



Divergence



Reflectivity



Divergence

$$\nabla \cdot \mathbf{V} = \frac{\partial u}{\partial x} + \frac{\partial v}{\partial y}$$

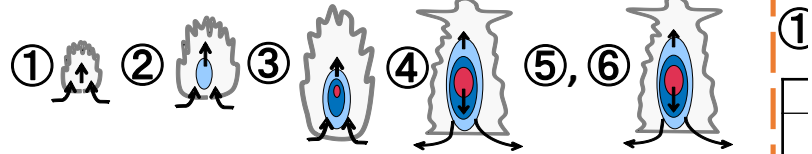
1.49E-2

-0.58E-2

Improvement of the quantitative risk prediction

Single regression $PRL_{all} = r(Z_h, \zeta) \quad S$

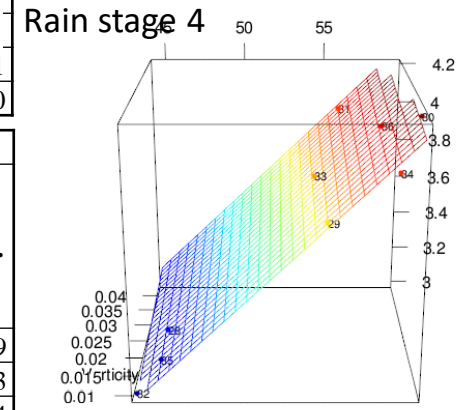
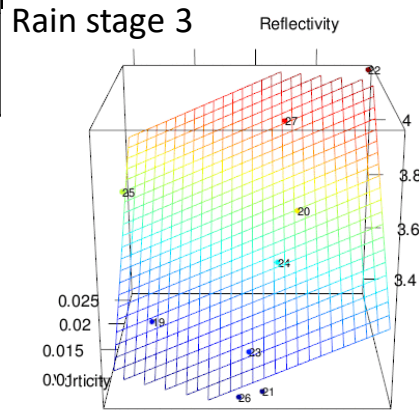
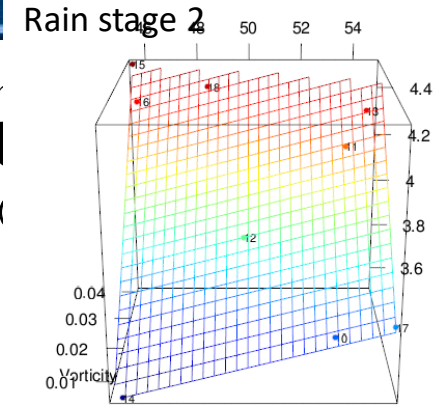
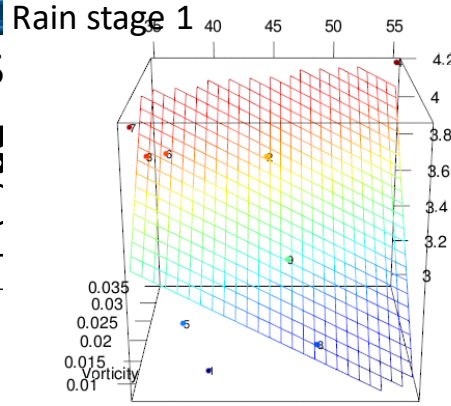
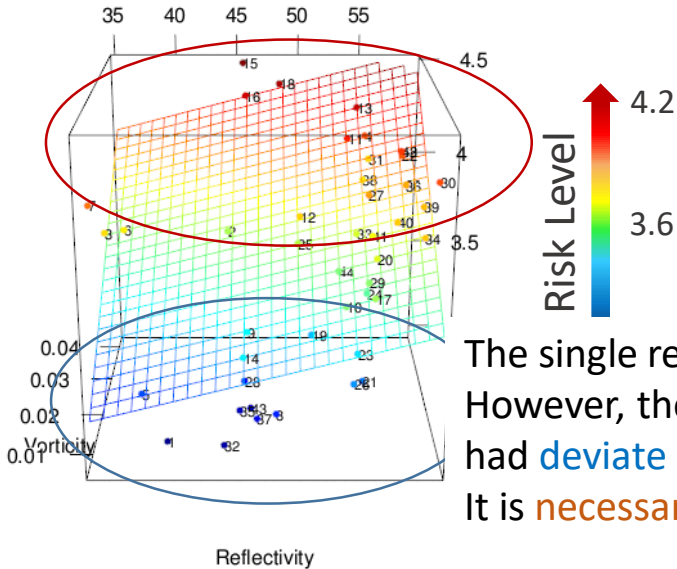
Compare single reg



Coefficients						
Model	Unstandard Coefficients		Standard Coefficients	t	Sig.	
	B	Std.Error	Beta			
1	(Constant)	2.55	0.31		8.21	0.00
	Z _h	0.02	0.01	0.67	4.04	0.00
	vorticity	8.10	3.30	0.41	2.45	0.02

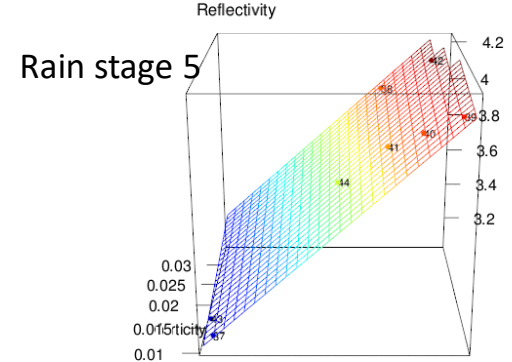
$$PRL = 2.55 + 0.02 \cdot Z_h + 8.10 \cdot \zeta$$

where, PRL : Predicted Risk Level,
Z_h : Reflectivity , ζ : Vorticity



Model	Unstandard Coefficients	Standard Coefficients	t	Sig.
(5), (6)				

The single regression plane fits well the variables. However, the risk level were underestimated and had deviate much than separate regression. It is necessary to make regressions separately.

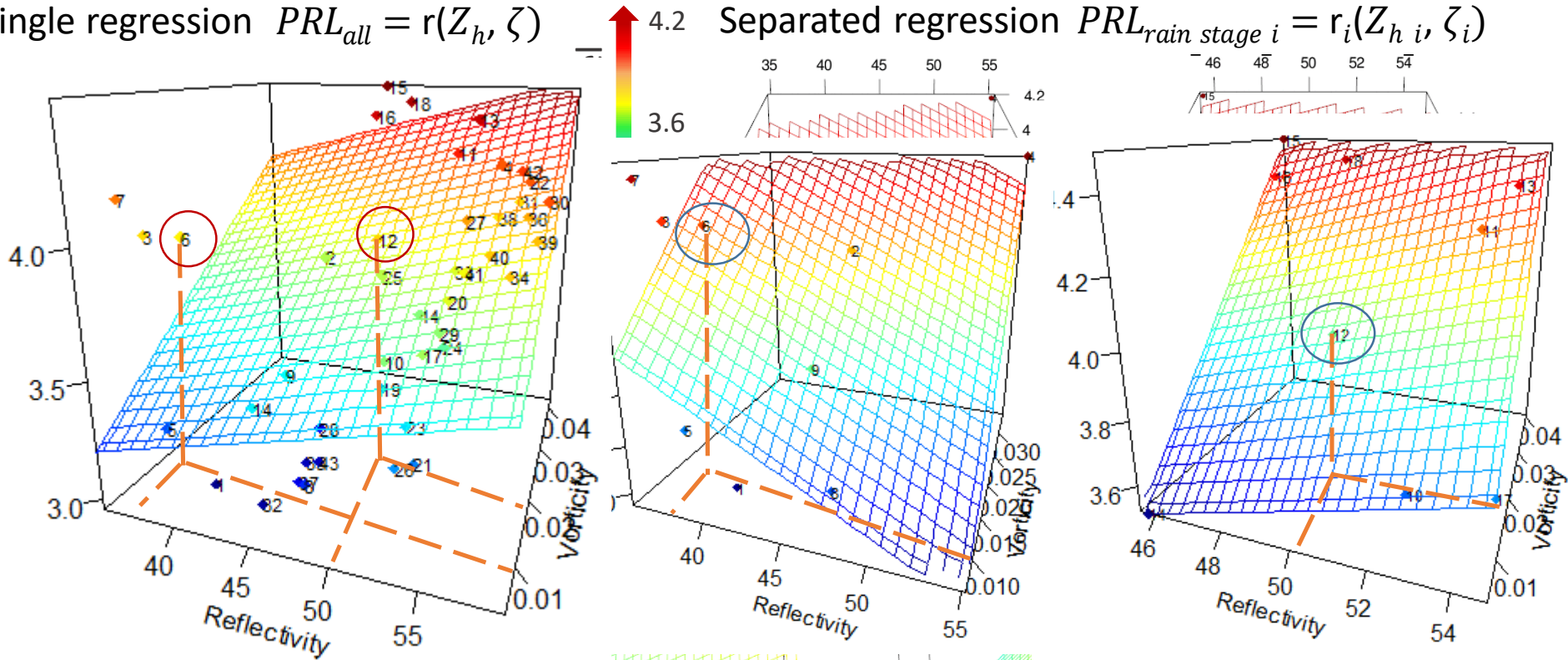


Sig.
.26
.27
.00
Sig.
.30
.00
.25

Improvement of the quantitative risk prediction

Compare single regression & separated regression

Single regression $PRL_{all} = r(Z_h, \zeta)$ Separated regression $PRL_{rain\ stage\ i} = r_i(Z_{h\ i}, \zeta_i)$

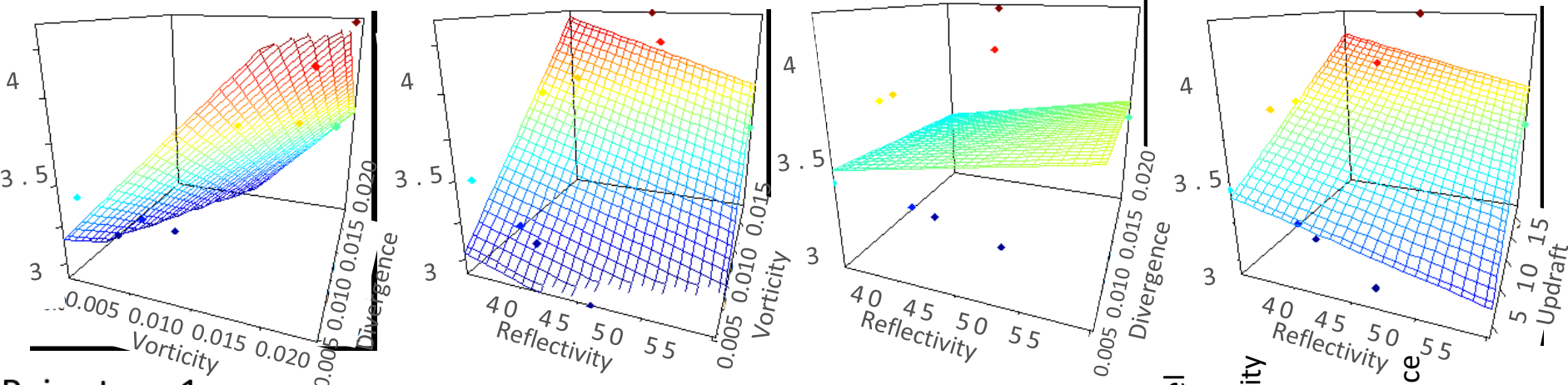


At RS 2, the single regression predict little higher risk level than separated regression
 At RS 1 3 4 5, the single regression predict little smaller risk level than separated regression
 It depends on the value of variables at each rain stage but allowedly the **separated regression** have **small difference** with the **estimated risk level**

Improvement of the quantitative risk prediction

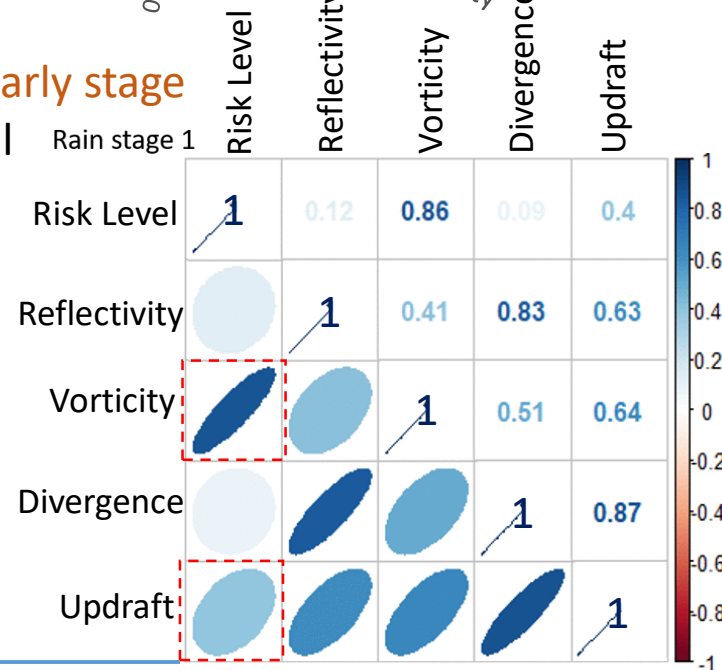
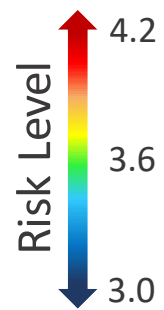
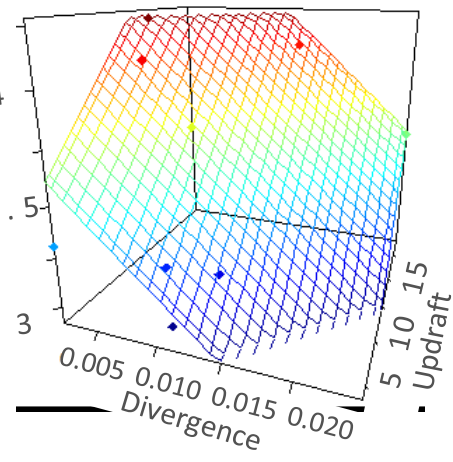
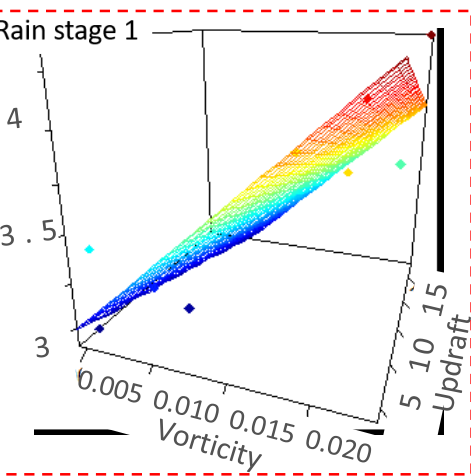
Z_h , ζ : Vorticity of multi-Doppler radar analysis, D : Divergence of multi-Doppler radar analysis, W

The relationship between the risk level and the variables



Rain stage 1

- Vorticity and Updraft do the very important role at the early stage
- The relationship btw vorticity and updraft explain RL well

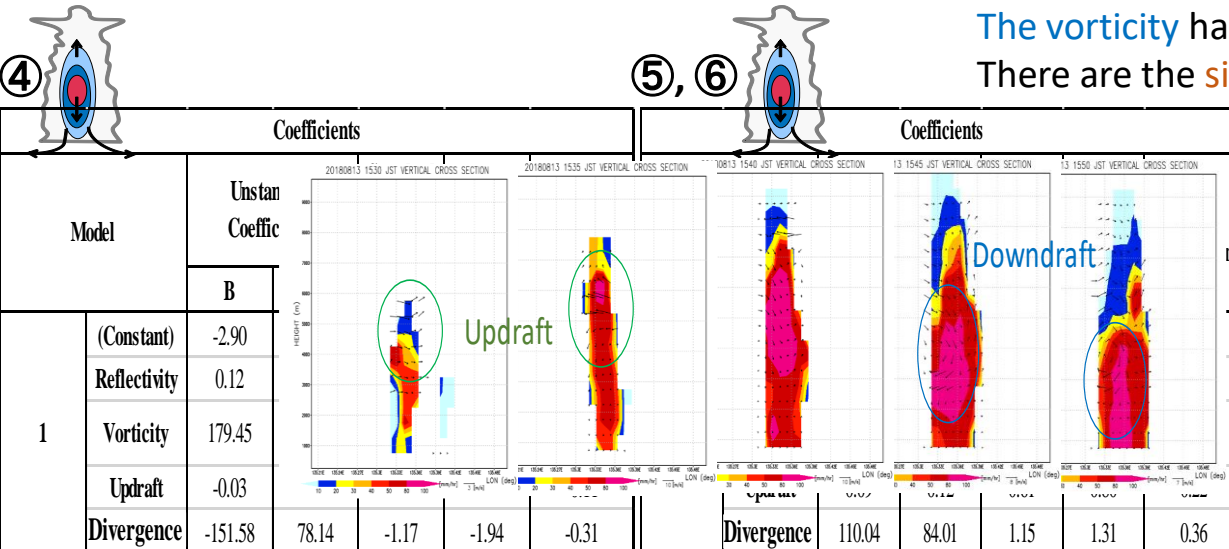


Improvement of the quantitative risk prediction

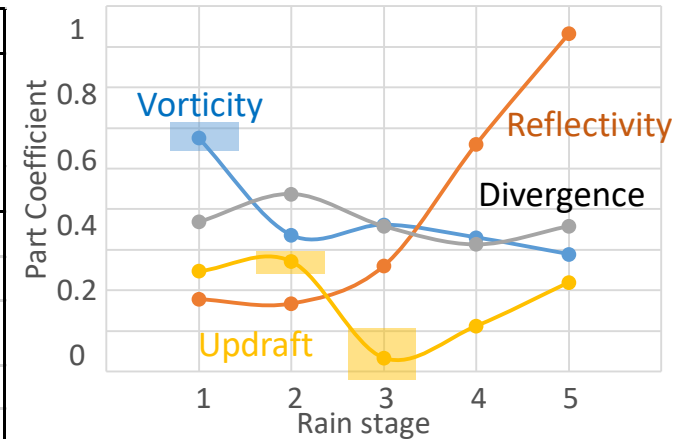
Z_h , ζ : Vorticity of multi-Doppler radar analysis, D : Divergence of multi-Doppler radar analysis, W

The relationship between the risk level and the variables

①							②							③						
Coefficients							Coefficients							Coefficients						
Model	Unstandard Coefficients		Standard Coefficients	t	Part Correlation		Model	Unstandard Coefficients		Standard Coefficients	t	Part Correlation		Model	Unstandard Coefficients		Standard Coefficients	t	Part Correlation	
	B	Std.Error	Beta					B	Std.Error	Beta					B	Std.Error	Beta			
1	(Constant)	2.40	1.25		1.92		1	(Constant)	4.01	1.31		3.07		1	(Constant)	-9.38	12.87		-0.73	
	Reflectivity	0.03	0.04	0.35	0.67	0.18		Reflectivity	-0.02	0.03	-0.17	-0.88	-0.17		Reflectivity	0.23	0.24	0.98	0.96	0.26
	Vorticity	124.44	57.55	0.77	2.16	0.58		Vorticity	100.27	56.49	0.96	1.77	0.34		Vorticity	237.70	177.98	2.79	1.34	0.36
	Updraft	0.09	0.10	0.61	0.93	0.25		Updraft	0.04	0.03	0.45	1.43	0.27		Updraft	0.01	0.08	0.09	0.12	0.03
	Divergence	-231.64	167.33	-1.14	-1.38	-0.37		Divergence	-74.75	32.39	-0.97	-2.31	-0.44		Divergence	-200.09	151.35	-2.63	-1.32	-0.36



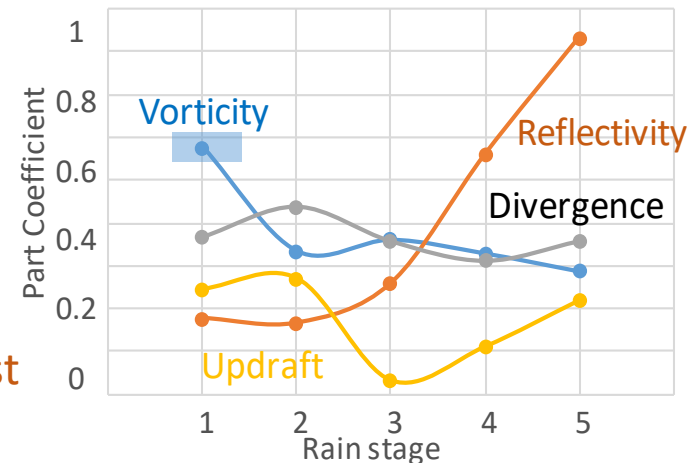
The vorticity has the most effective variable at early stage
 There are the significant change of peak on updraft



Conclusion and Future works

Conclusion

- By using the three-dimensional wind field from multiple-Doppler radar analysis, **the reliable wind field** reflecting natural environment was calculated
- At the early stage, the part coefficients of vertical vorticity have high value -> the **vorticity** has the **most important** in the dependent variable
- The **predicted risk level** by using **separated regression** didn't deviate much than single regression



Future study

- By using the equation, I check the accuracy by using the **ROC curve** on the linear regression equation.
- Then, compare the accuracy between **single and separated regression**
Predicted Risk Level (PRL)

$$= C_0 + C_1 \cdot Z_h(\text{reflectivity}) - C_2 \cdot \zeta(\text{vorticity}) + C_3 \cdot \omega(\text{updraft}) \cdots + C_n X_n$$
- Apply the quantitative risk prediction method to flash flood

A decorative header image showing blue water with ripples from a central point.

Thank you for listening