



# Short-Term Quantitative Precipitation Forecasting

*Group 16*

# Outline

- Team Members
- Target Problem
- Brief descriptions on the datasets
- Work plan & Methods Used
- Experimental results
- Future work

# Team members



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# Target Problem

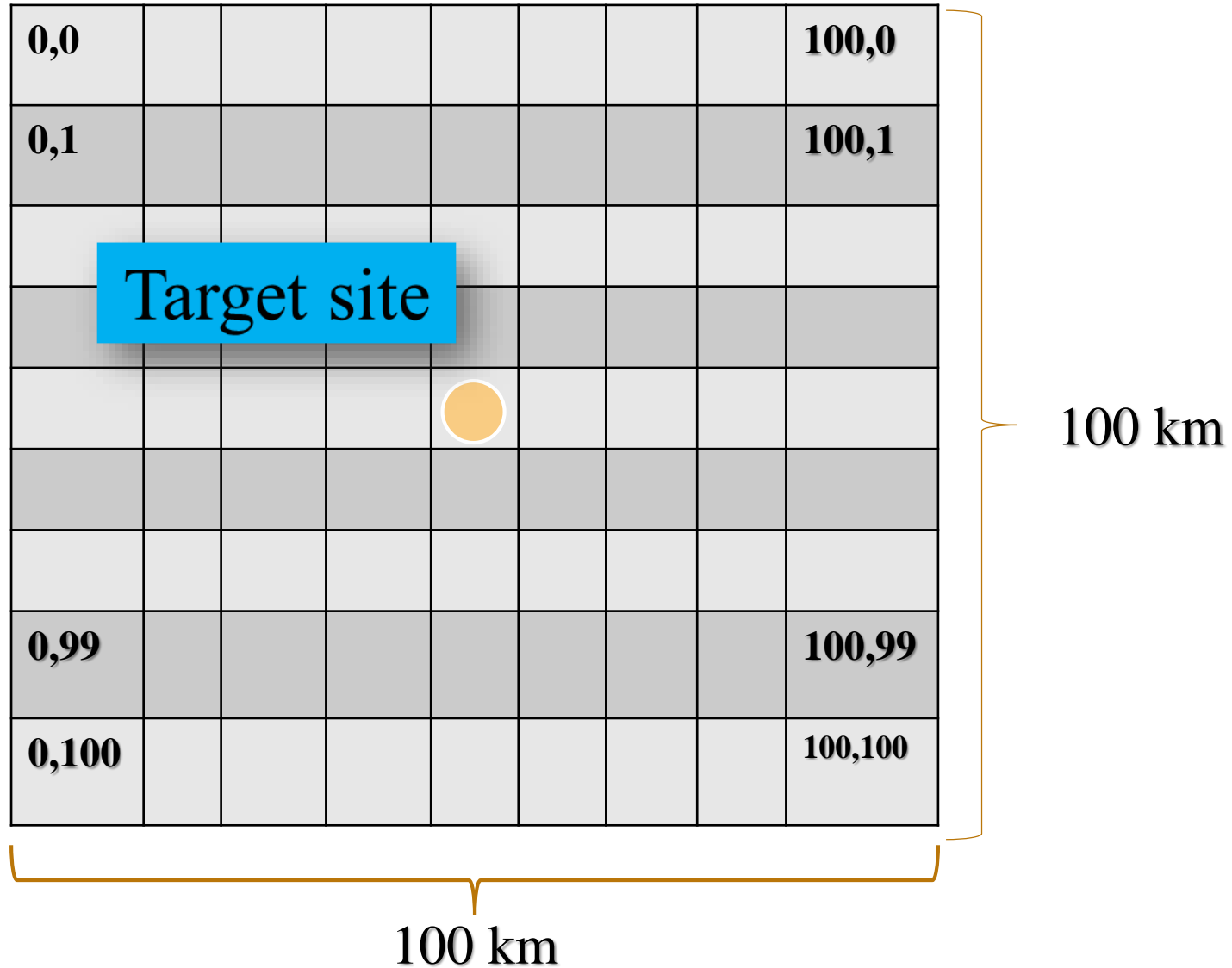
- CIKM AnalytiCup 2017
  - CIKM is called Conference on Information and Knowledge Management.
  - Since its inception in 1992, CIKM has successfully brought together a wide range of R & D personnel in the field of knowledge management, information retrieval and database.
  - CIKM 2017 will be a unique perspective, strategic penetration of knowledge, information and data management of the cross-type research, highlighting the realization of many urban areas and their countries to share the "smart city, smart country" vision of the technology and insights.

# Target Problem

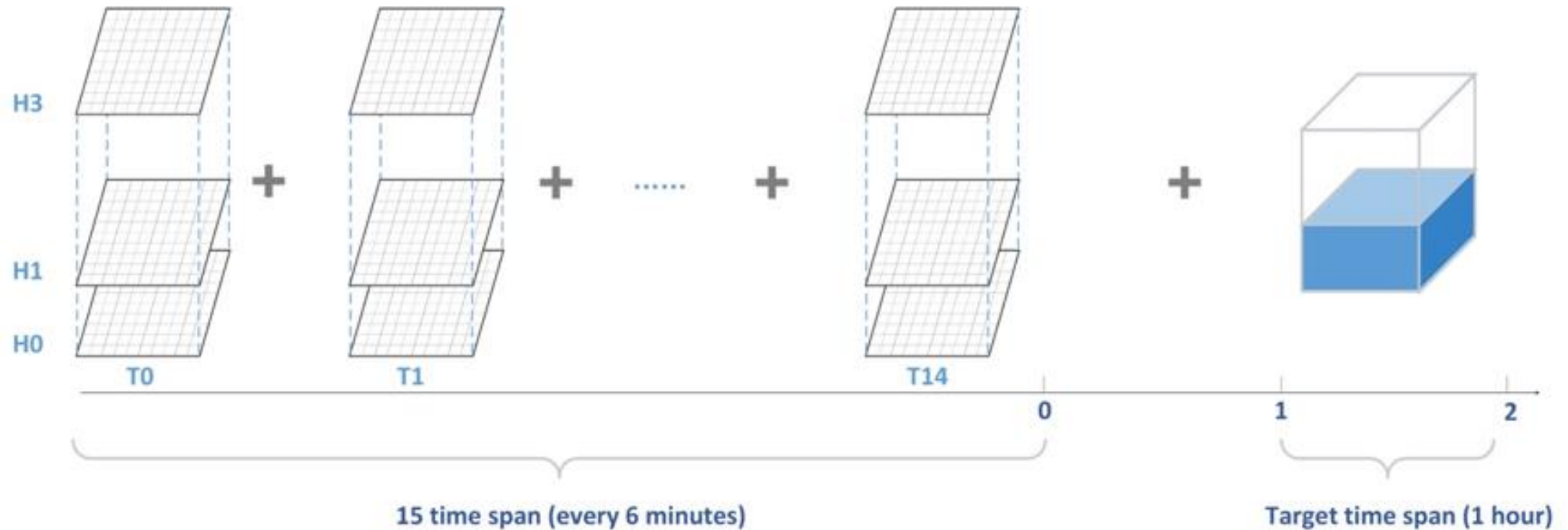
- CIKM AnalytiCup 2017
  - It is an open, notarized big data open competition. For the academic community, it is an exciting data challenge.
  - The short-term precipitation forecast jointly conducted by Shenzhen Meteorological Bureau and Alibaba, aims to significantly improve the accuracy of short-term precipitation forecasts based on radar echo extrapolation data.



# Brief descriptions of datasets



# Brief descriptions of datasets(cont.)



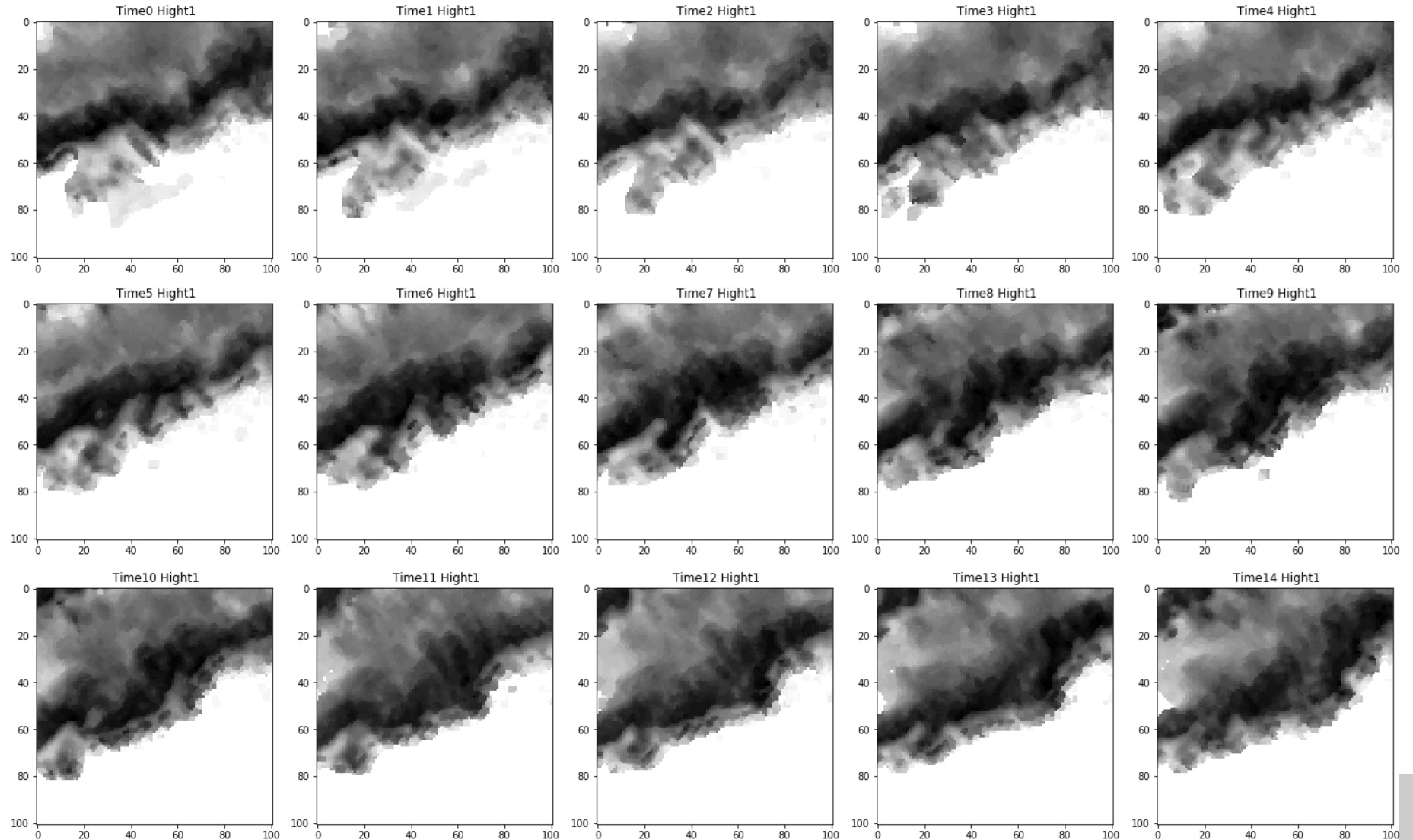
## Brief descriptions of datasets(cont.)

- Train Data: **15.8GB** (two years)
  - Training sets: 10,000
  - each data about 2MB [**id, label, 60 radar map**] ( $15*4*101*101$ )
- Test Data: **3.09GB** (one year)



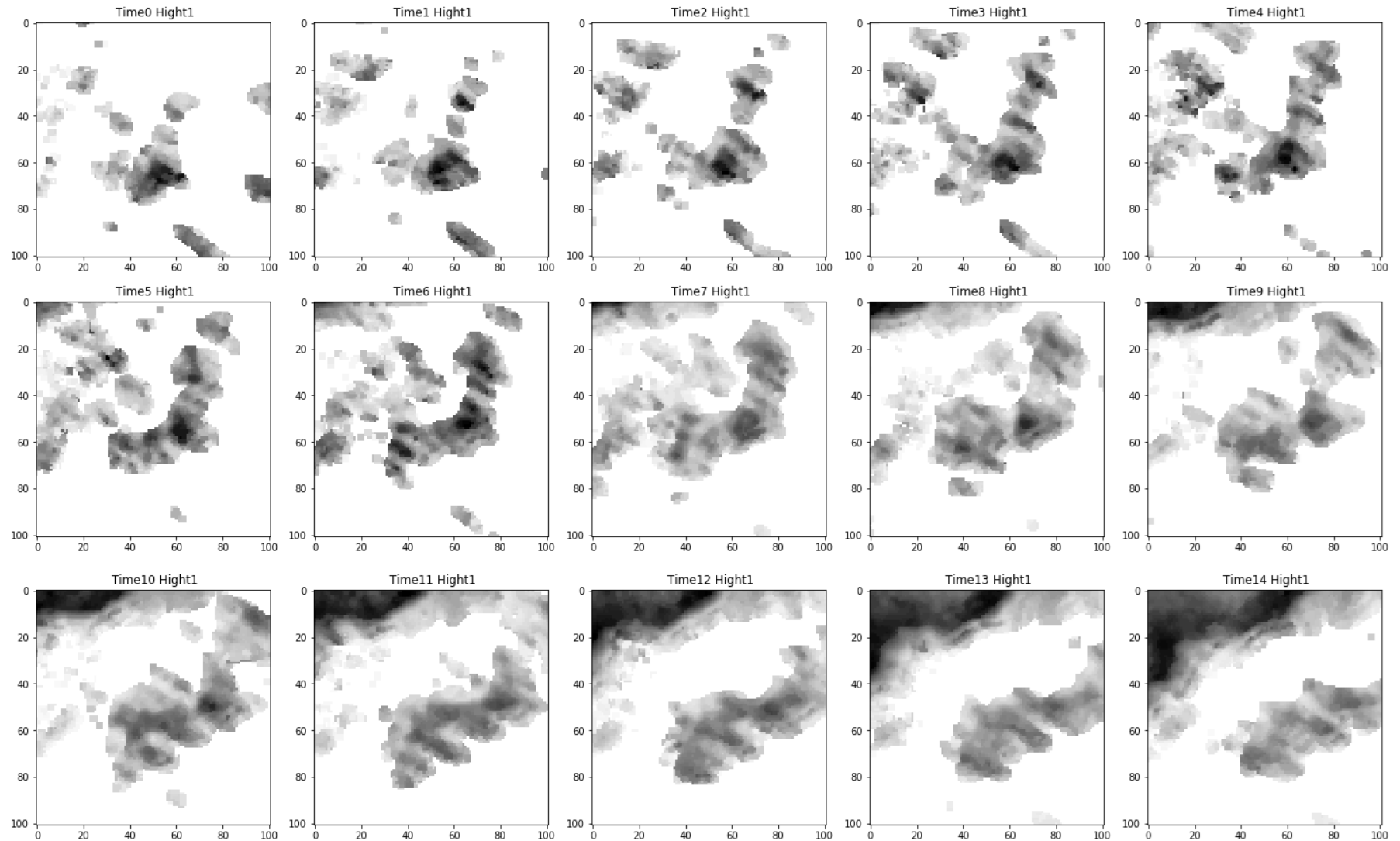
# Plot the radar image

- 15 time span
- height 1
- $Y = 71.6$
  
- No wind



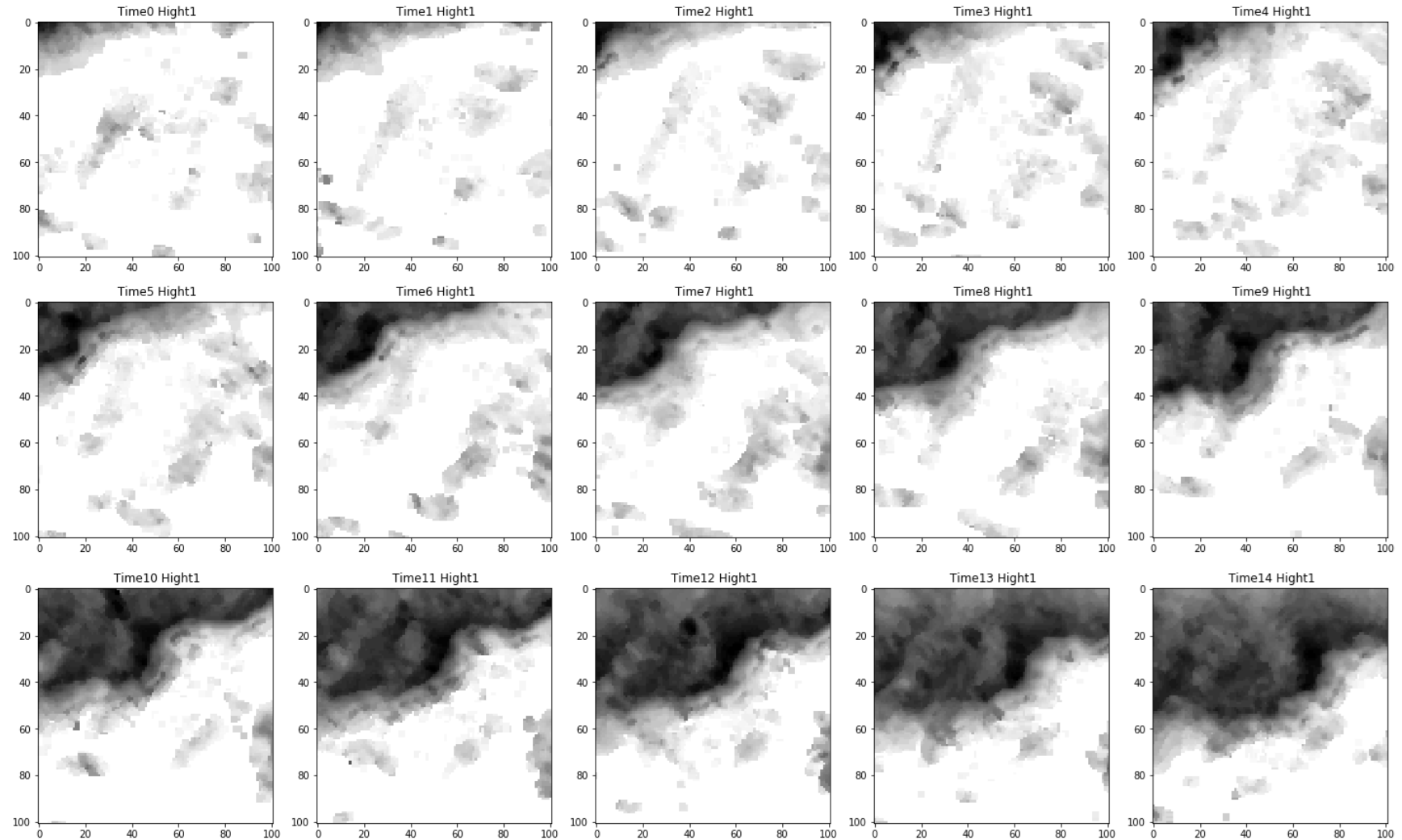
# Plot the radar image(cont.)

- 15 time span
- height 1
- $Y = 29.9$
  
- No wind



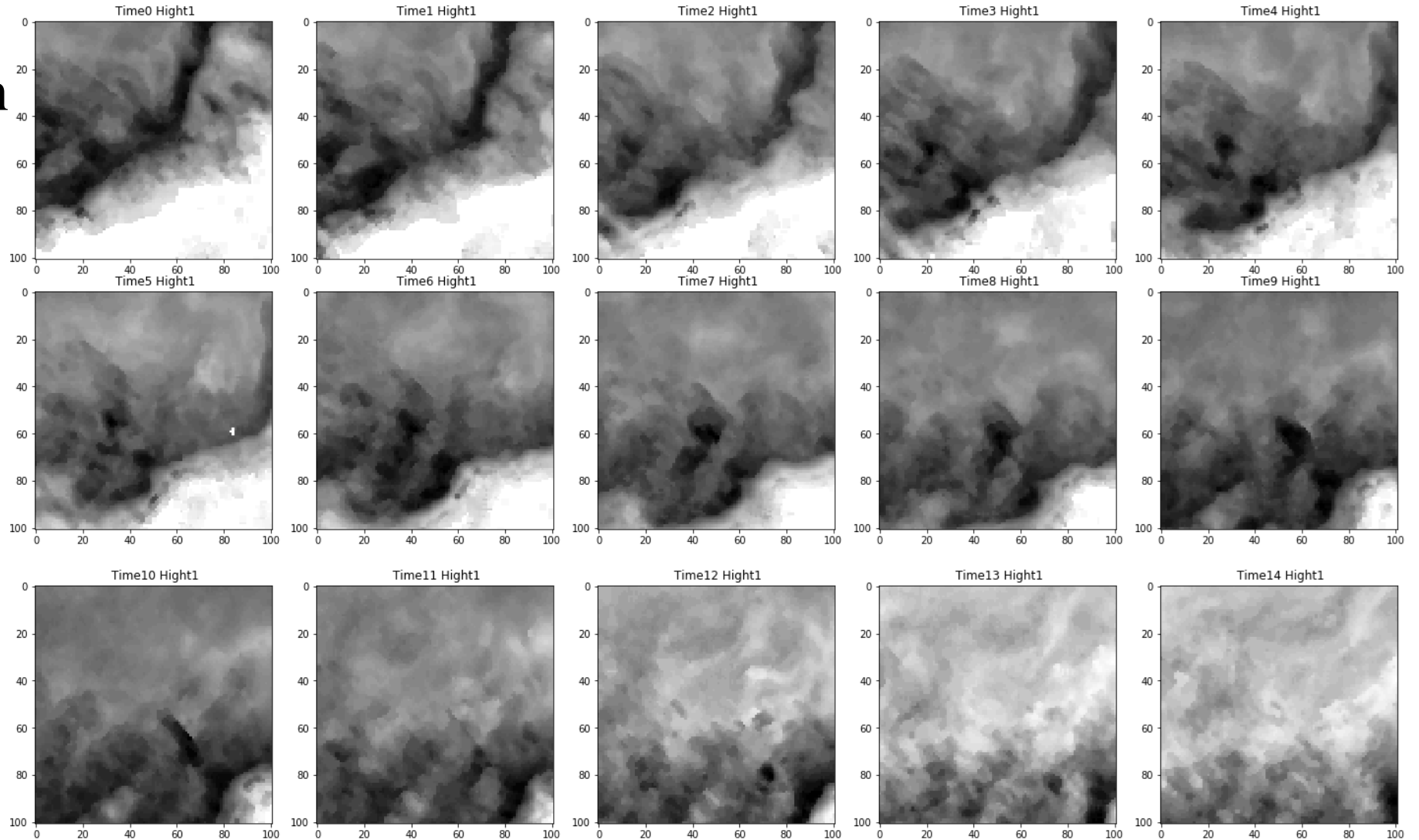
# Plot the radar image(cont.)

- 15 time span
- height 1
- $Y = 4.5$
  
- Windy



# Plot the radar image(cont.)

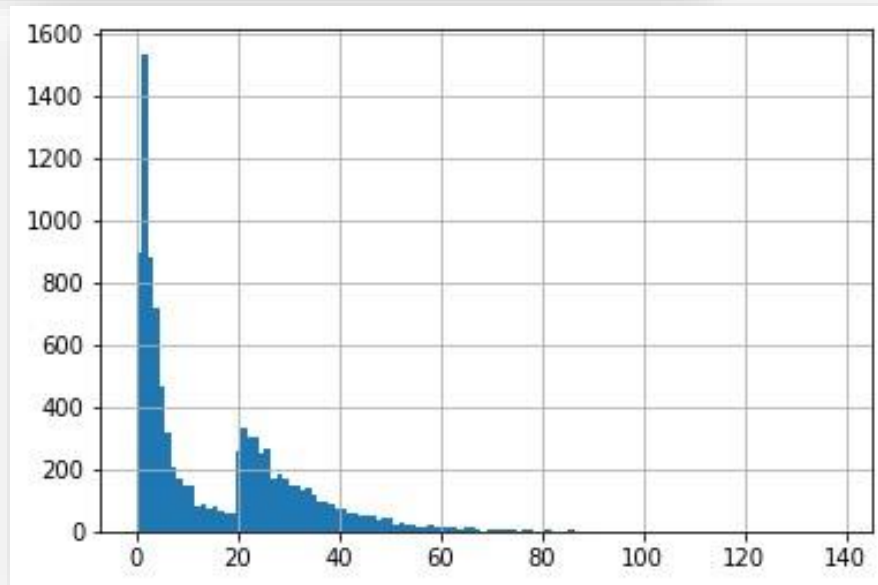
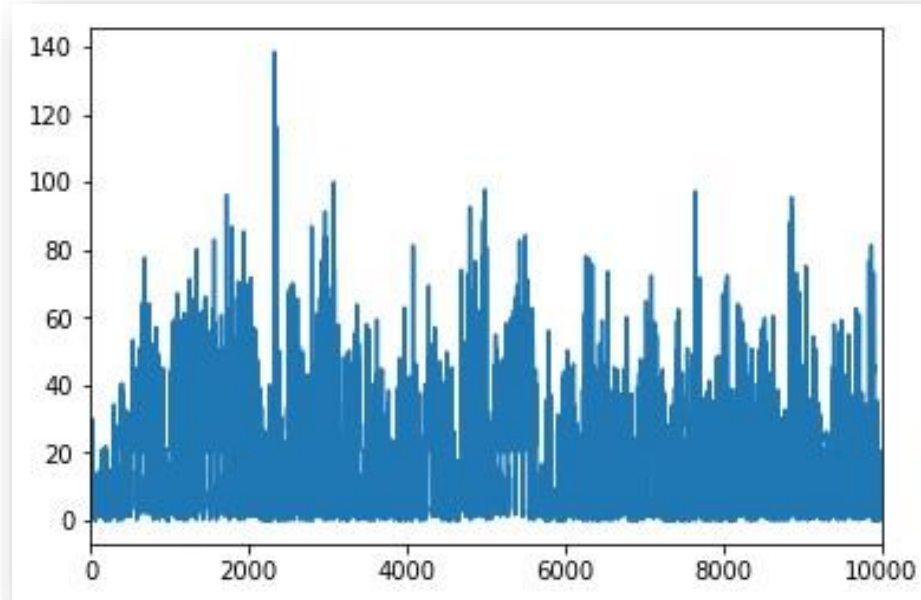
- 15 time span
- height 1
- $Y = 2.3$
  
- Windy



# Plot the label $y$

- Summary Statistics

```
count    10000.000000
mean      15.545400
std       15.855781
min       0.000000
25%       2.400000
50%       8.000000
75%      25.700000
max      138.400000
dtype: float64
```



# Work plan & Methods Used

- Platform and Tools
  - Hadoop & **Spark** & MongoDB (Pig, **MLlib** etc.)
  - **Python** ( scikit-learn pandas numpy etc. )
  - **TensorFlow** & **keras** (training CNN/RNN)
  - etc.

# Work plan & Methods Used(cont.)

- Workflow
  - Basic methods
    - Linear Regression
    - SVM
  - Preprocess Clustering methods
  - Deep Learning
    - Classification
    - Regression

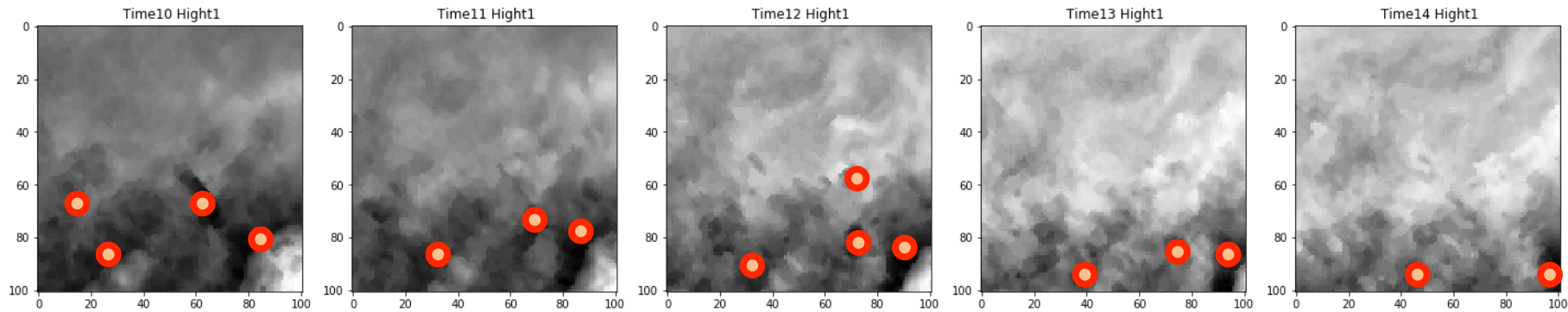
# Work plan & Methods Used(cont.)

- Basic methods(baseline)
  - Linear Regression
    - Only using the target point(50,50) to train the model
    - Only using the mid 3x3 map to train the model
  - Ensemble Linear Regression
    - According different changing rate of images, train the models by different situation.
  - SVM
    - Only using the target point(50,50) to train the model
    - Only using the mid 3x3 map to train the model



# Work plan & Methods Used(cont.)

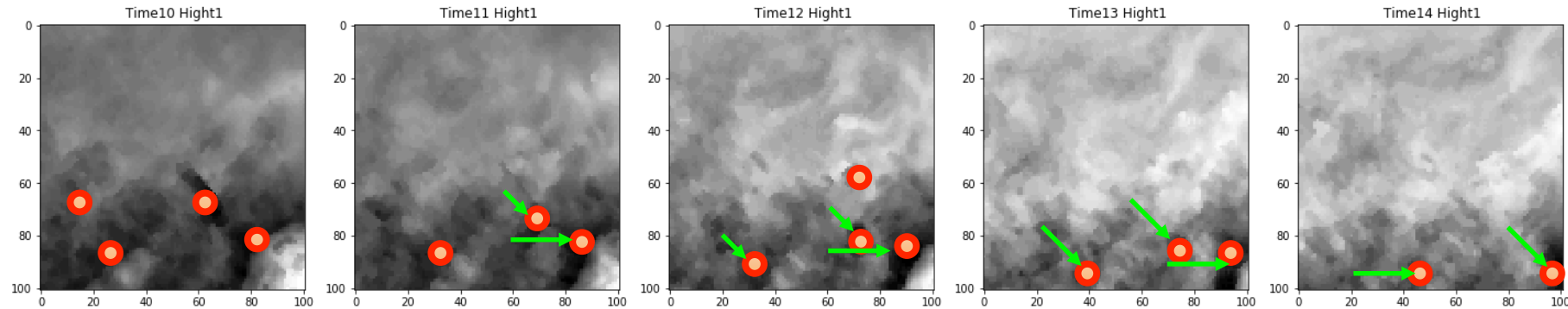
## ■ Preprocess Clustering Method



- Generate Clusters
  - **Find Peak point** - the potential center of cluster
  - **Aggression**: Flood Fill. Find out the real center of clusters.

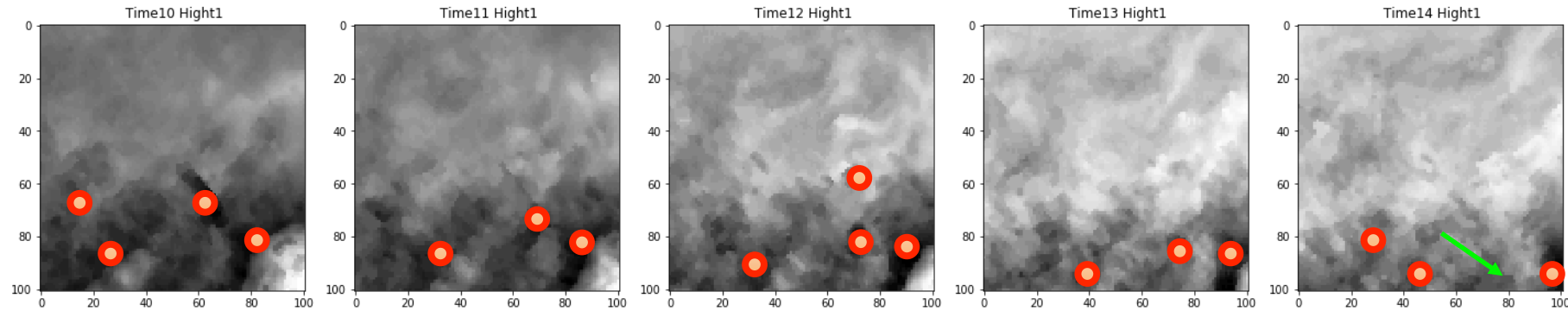
# Work plan & Methods Used(cont.)

## ■ Preprocess Clustering Method



- Calculate Immigration Direction
  - Do **statistics** of the direction of 14 images. Select the most possible one.
  - **Speed**: calculate the speed of both latitude and longitude.

# Work plan & Methods Used(cont.)

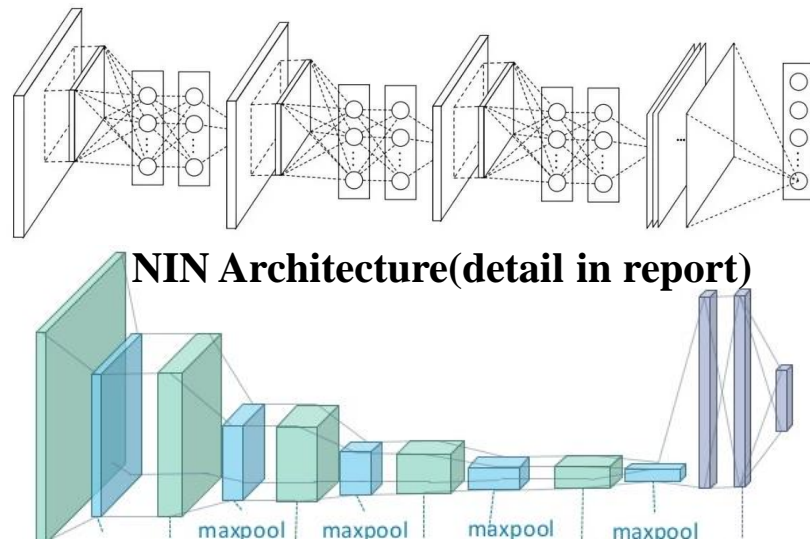


- The Influence of cluster works on center point
  - We only consider the last image
  - **Predict cluster position**: in the next image, the cluster center will move, so we should take speed in our formula.
  - **The bound of cluster**: we should consider not only the radar value of the cluster center, but also the point around center.

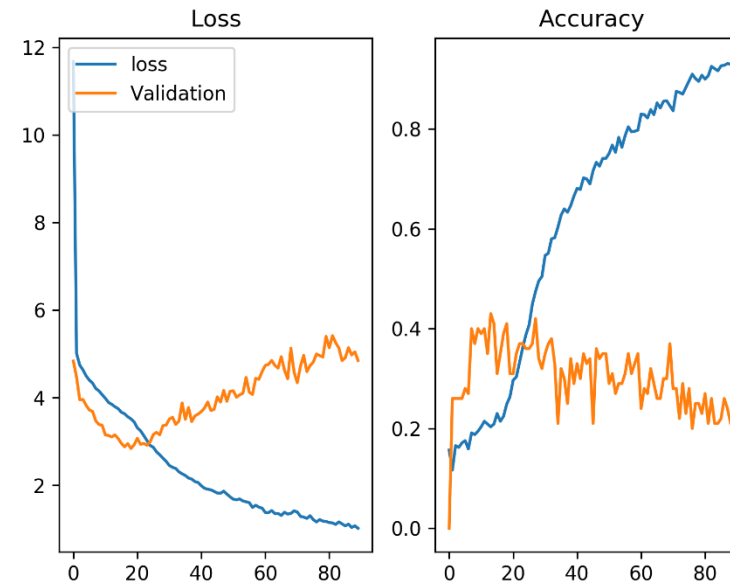
$$y = w_1 \sum_{i=1}^k \frac{v_i}{D^2} + w_2 \sum_{i=1}^k \frac{\sum_{j=1}^9 v_{ij}}{9D^2} - w_2 \sum_{i=1}^k \frac{\sum_{j=1}^{25} v_{ij}}{25D^2}$$

# Work plan & Methods Used(cont.)

- **Deep Learning(Classification)**
  - Assume 100 classes
  - Training acc is good, but test acc is bad(overfitting)
    - 100 classes is too small? Maybe 1000?
    - **Training data is not enough.(the more, the better)**



VGG19 Architecture

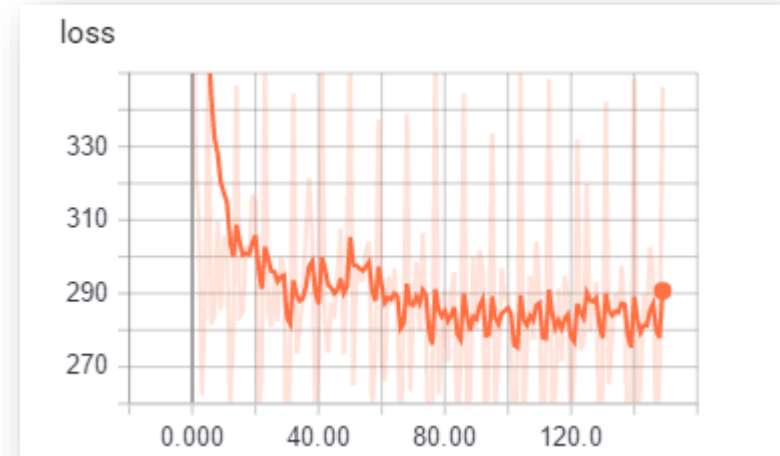


# Work plan & Methods Used(cont.)

- **Deep Learning(Regression)**
  - Modify CNN model to regression (**detail in report**)
    - Add fully connected layer to instead of softmax layer
    - Using Adam(RMSprop) optimizer
  - Need a lot of time to train
    - 12 GB memory & GTX 1060(**Can't load data in one time**)
    - **about 30 hours / 100 epochs**

```
model.add(Flatten())
model.add(Dense(1))

# optimizers should be tested
# sgd + momentum
# others
adam = optimizers.Adam(lr=0.0035, beta_1=0.9, beta_2=0.999, epsilon=1e-08, decay=1e-6)
# sgd = optimizers.SGD(lr=0.005, momentum=0.9, decay=1e-6, nesterov=True)
# rms = optimizers.RMSprop(lr=0.0035, rho=0.9, epsilon=1e-08, decay=1e-6)
model.compile(optimizer=adam, loss='mse')
return model
```



# Experimental results

- Using **RMSE** to judge
- Rank: **118/1307**
- Basic methods
  - Linear Regression(baseline): 14.90
  - SVM: about 16.8
  - Decision Tree: 17.5(worst)
- Preprocess Clustering methods:
  - **Less than 14.44(still improving)**
- Deep Learning(CNN)
  - Classification: no result
  - Regression: **less than 14.50 (still improving)**
  - **Tring to combine RNN**

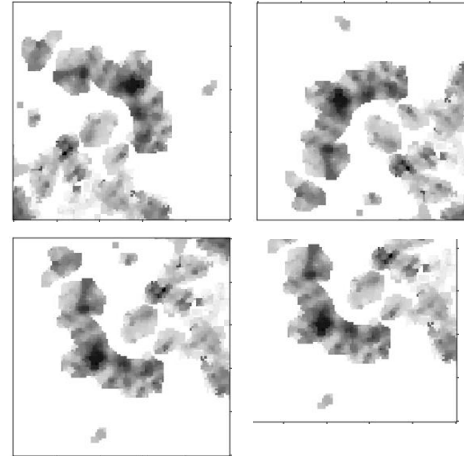
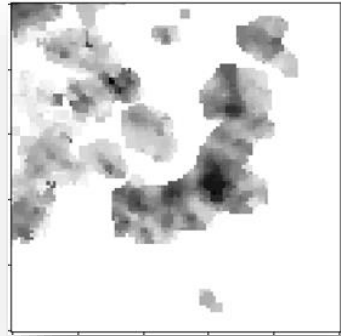
第 1 赛季截止日期	总奖池	参赛队
2017/07/01	\$11000	1307

排名	参赛者	所在组织	RMSE
1	Marmot	清华大学	10.70334
2	AI Rookies VN	AI4U	12.64242
3	不存在的里皮1	华南理工大学	13.11170
4	Is the order a rabbit?	Rabbit House	13.13560
5	怀北村明远湖	中国科学院	13.16495
...			
117	The Former Over Fitt...	北京邮电大学	14.45423
118	<sup>46</sup> 楼上的等着，楼下的...	国立交通大学	14.45430
119	<sup>1</sup> 该我上场表演了	东北大学	14.45531

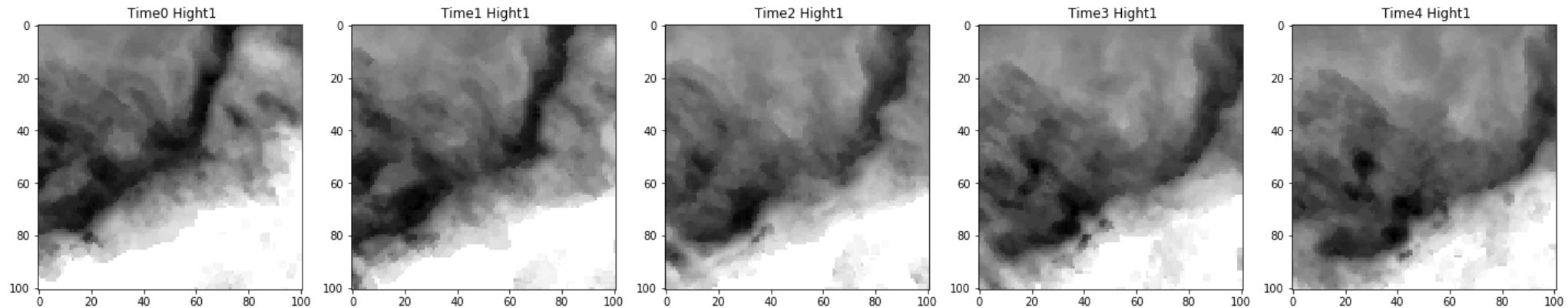
# Future work

- **Generate more data**
  - Rotate or Flip
  - Data augmentation



# Future work(cont.)

- Generate more data
- **RNN(LSTM or other)**
  - There are some correlations between the radar map





## Future work(cont.)

- Generate more data
- RNN(LSTM or other)
- **Ensemble learning**
- **Consider better clustering methods**
- **etc.**

## Future work(cont.)

- Generate more data
  - RNN(LSTM or other)
  - Ensemble learning
  - Consider better clustering methods
  - etc.
- 
- **We are still improving**



Q & A