



National Research  
Foundation of Korea



European Commission

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# Platform Based Deep-Learning Applications as a Solution for Satellite Big Data Analysis : The Case of Rice Paddy Detection in South Korea

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# Overview

- ▷ Background
- ▷ Data Preparation
- ▷ Method
- ▷ Results
- ▷ EOPEN Platform

# From Data to Information

- ▷ Decision makers need “information” not data
  - Proper data processing under researchers’ understanding can change data into information
  - Handling data is being harder as the volume and complexity of data increases
- ▷ Big Data consists of massive satellite images
  - Increasing spatio-temporal resolution
  - Collecting data from wide range of area

# Limitations for Big data Processing

## ▷ Technical barriers

- High performance hardware for Big Data processing
- Infrastructure for downloading/storing massive data

## ▷ Imbalance between Big Data and labeling data

- Data should be labeled to be trained in the model
- Satellite images are acquired in near-real-time over wide areas, while labeling data (land cover map etc.) are not updated rapidly
- Using incorrect labeling data could increase error

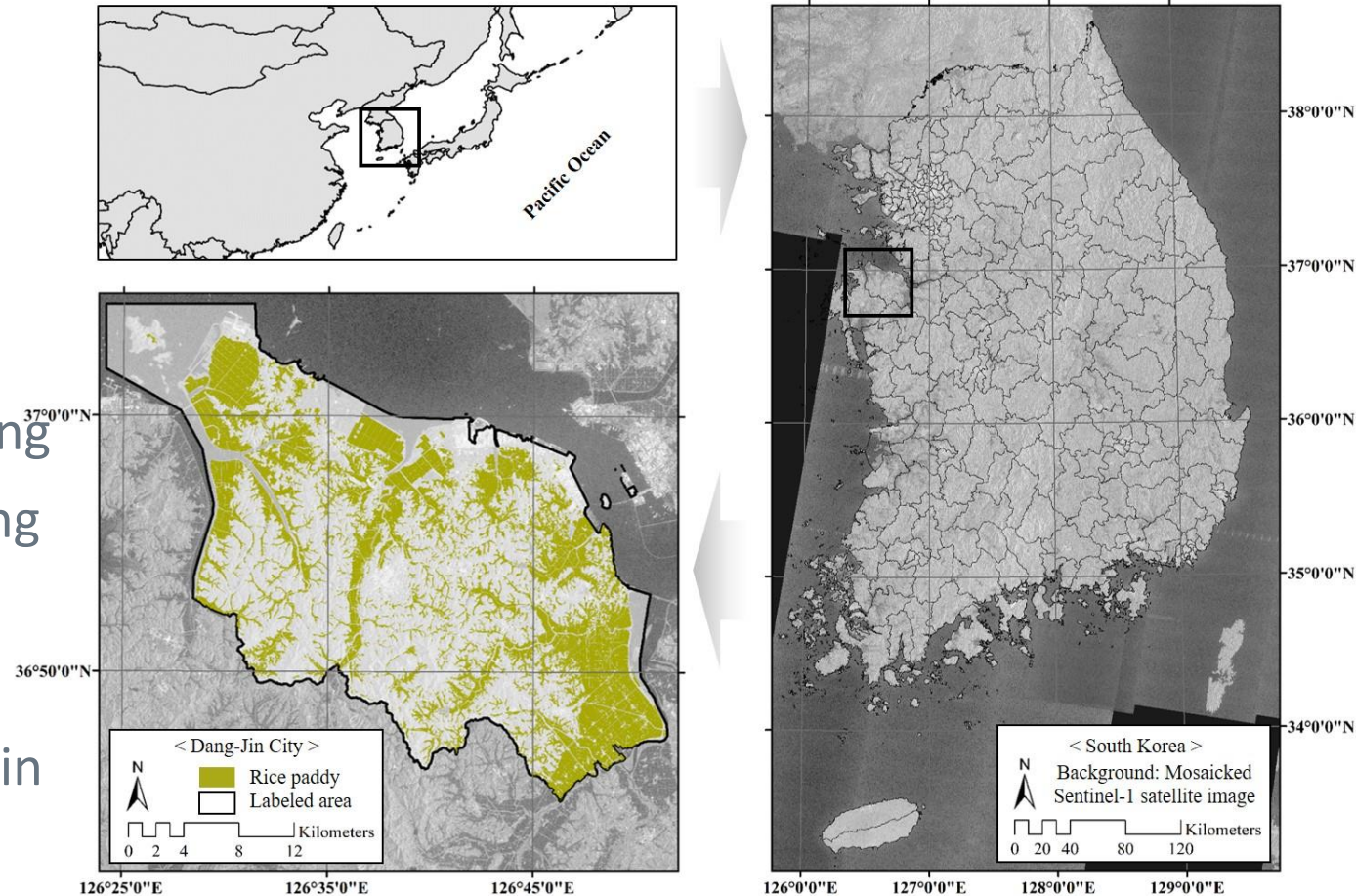
# Solutions for Big Satellite Data Processing

- ▷ Platform based data downloading / storing / processing
  - Sharing high performance server which individual cannot afford
  - Supporting big data processing with the advanced ICT solutions
- ▷ Deep-learning analysis with Data science techniques
  - Extract information from big data with neural networks which provides complex, non-linear modeling
  - Applying techniques to overcome the shortage of labeling data



# Rice Paddy Detection in South Korea

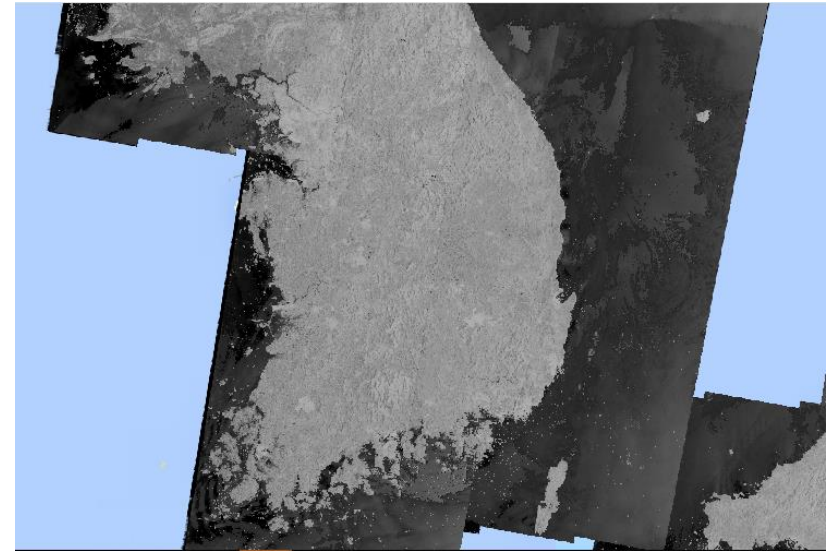
- Ideal testbed for suggested Big Data processing solution
- Fragmented rice paddy in Asia & Europe needs high-resolution monitoring
- Non-linear modeling of deep-learning is effective for exploiting rice phenology appears in time series dataset



Rice paddy in Dang-Jin, Level-3 land cover map

# Sentinel – 1 Acquired from Google Earth Engine

- ▷ Active sensing (available for all-weather)
  - Advantage on acquiring time series data
- ▷ Sensitive to textural information
  - Extreme contrast of reflectance according to rice existence
- ▷ Preprocessed products are downloadable
  1. Apply orbit file
  2. GRD border noise removal
  3. Thermal noise removal
  4. Radiometric calibration
  5. Terrain correction (ortho rectification)

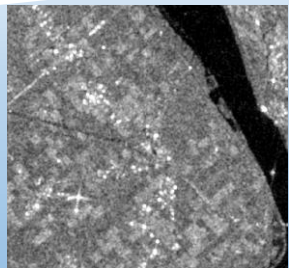
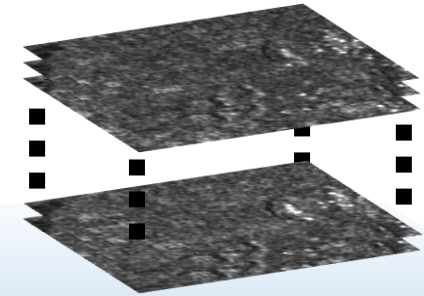


Modified Code removing outlier (Border line)

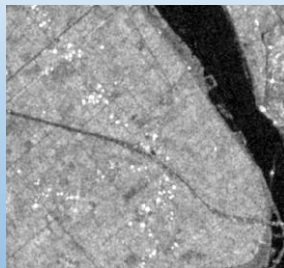


# Time Series Dataset

- ▷ Monthly data are acquired with mean value
- ▷ A year of time series data set with rice paddy labeling is learned in deep-learning model



Mar



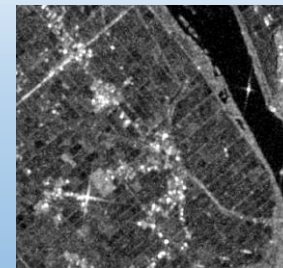
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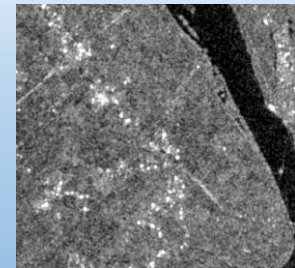
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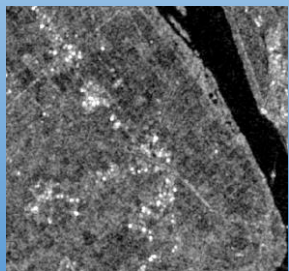
May - 2



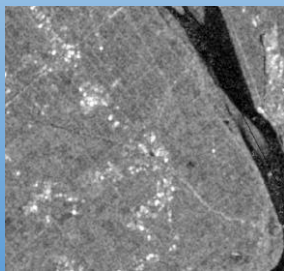
Jun - 1



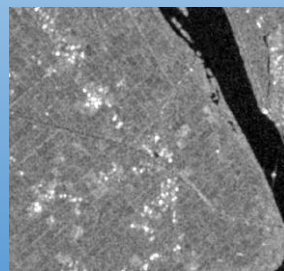
Jun - 2



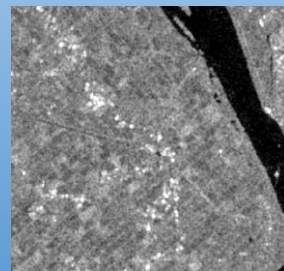
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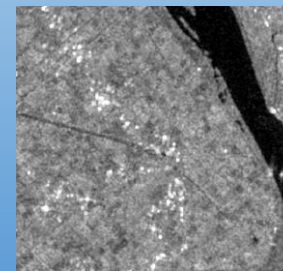
Aug



Sep



Oct



Nov



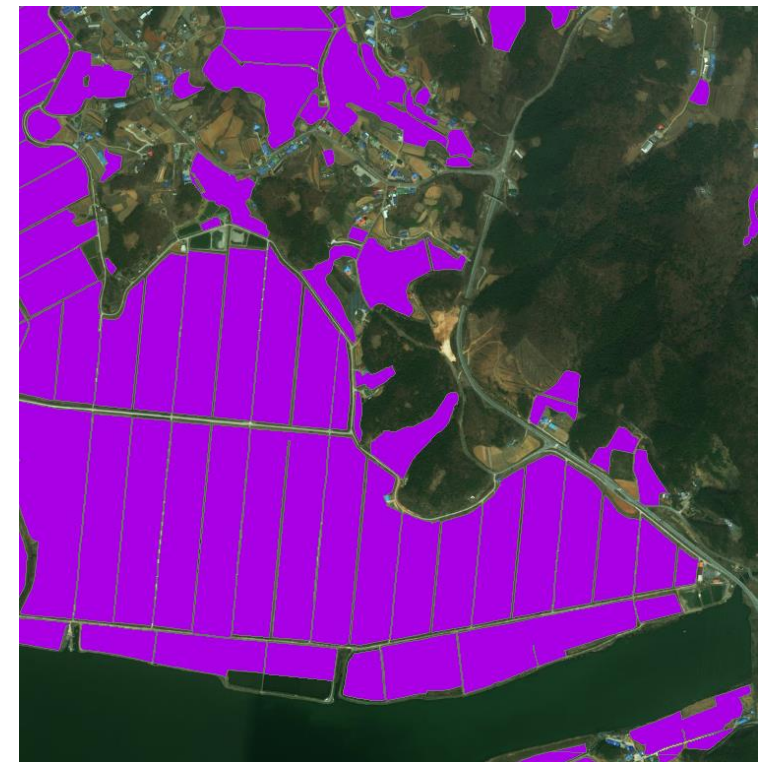
Labeling data



# Labeling Data: level-3 land cover map

- ▷ Rice paddy mapped in parcel level
- ▷ Only on-demand regional update

Level	Scale (resolution)	Format	Class	Materials	Update
Level – 1	1 : 50,000 (30m)	Raster	7	Landsat TM	Every 10 years (whole nation)
Level – 2	1 : 25,000 (5m)	Shape	22	Landsat TM / SPOT5 / KOMPSAT-2	About every 4 years (regionally, province level)
Level – 3	1 : 5,000 (1m)	Shape	41	KOMPSAT-2 / IKONOS / Aerial photo	Every years (regionally, city level)

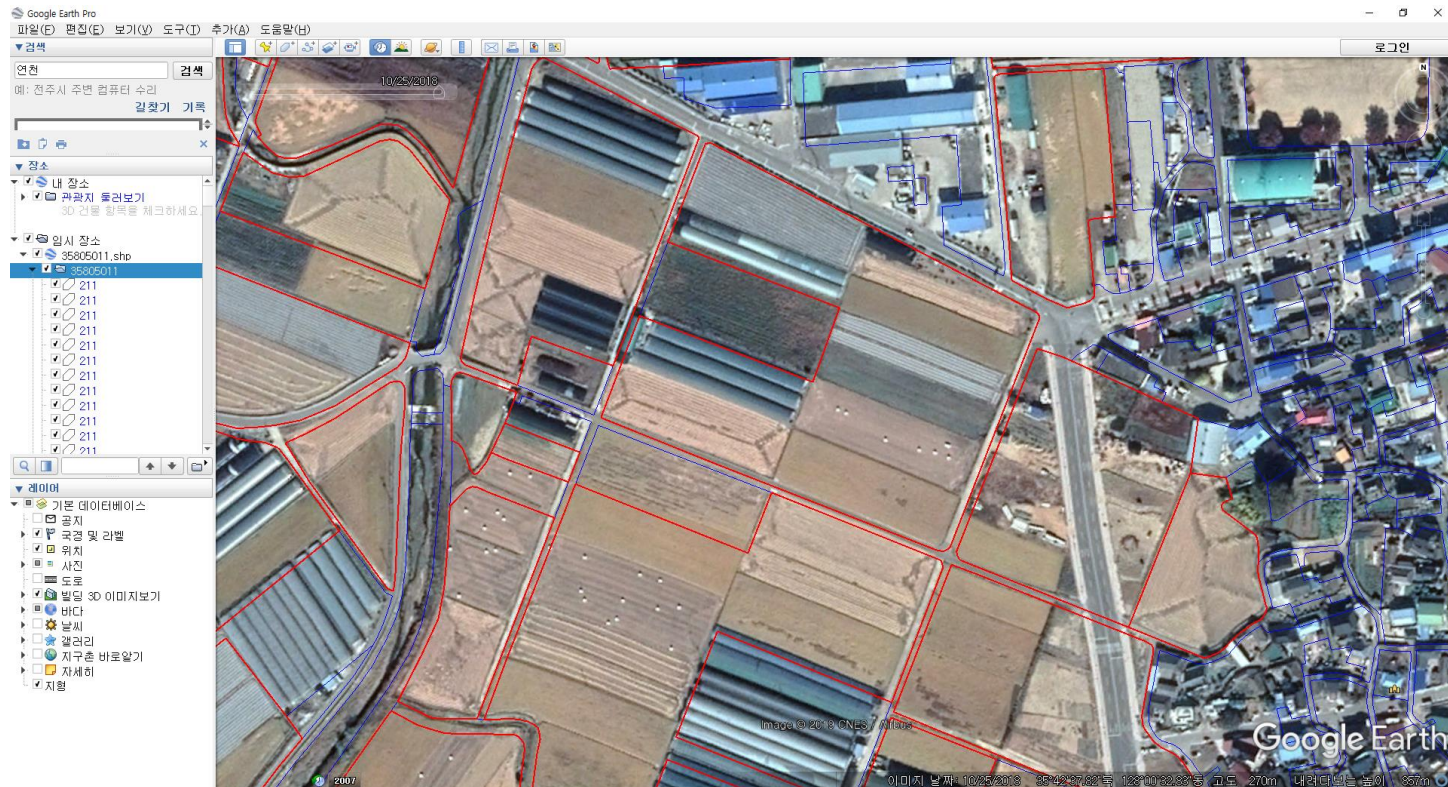


Level-3 land cover map

 Rice Paddy

# Updating Labeling Data: Google Earth

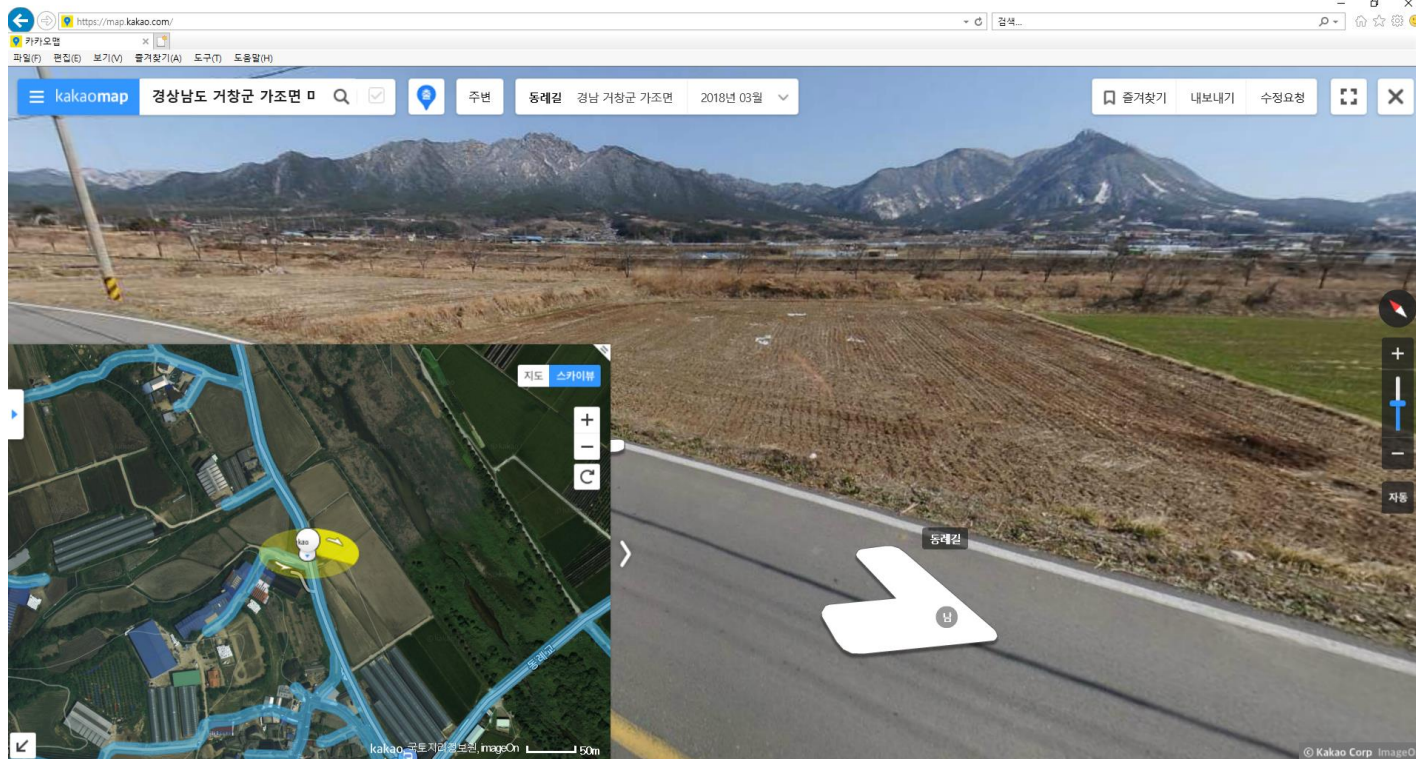
▷ Land cover map (paddy area) is updated with on-screen digitizing





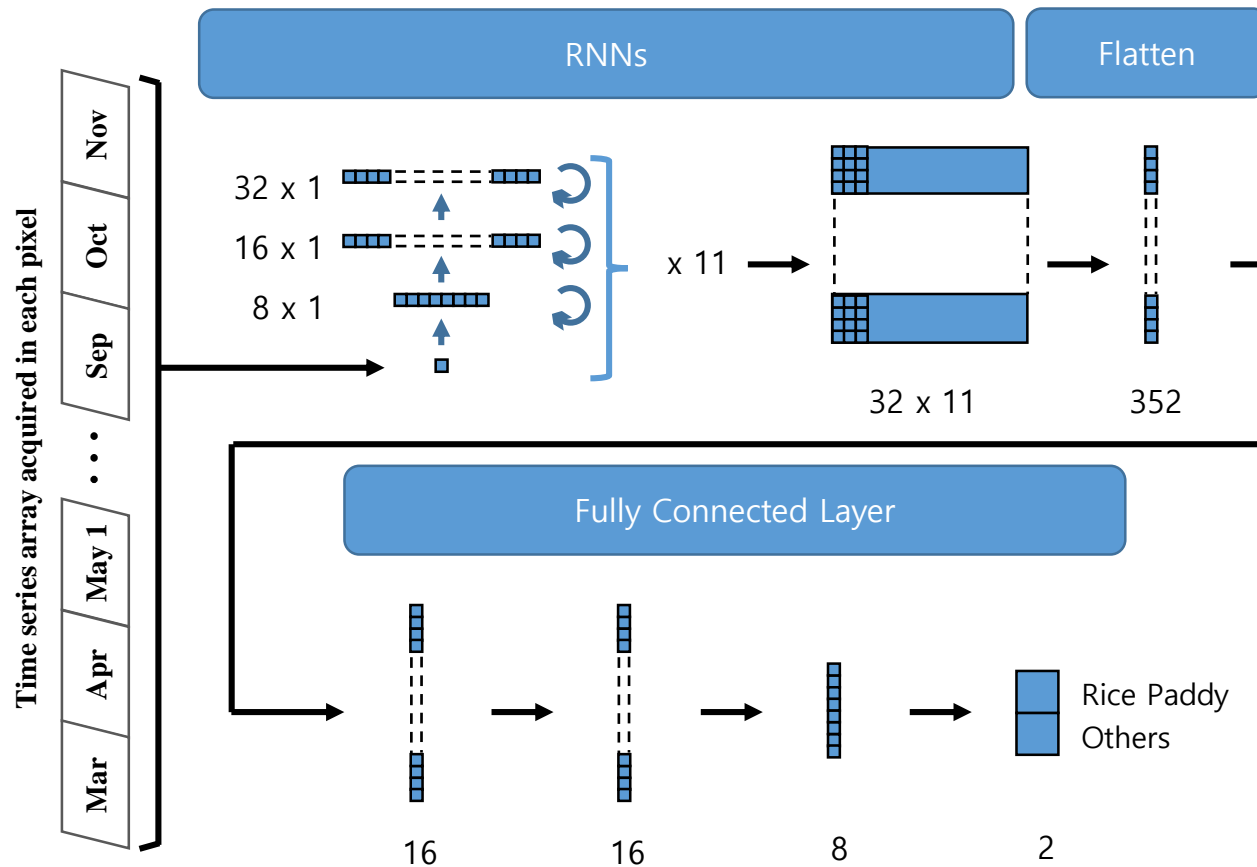
# Updating Labeling Data: Domestic street view service

- ▶ Uncertain areas are explored with domestic street view services for more accurate update



# Recurrent Neural Networks (RNNs)

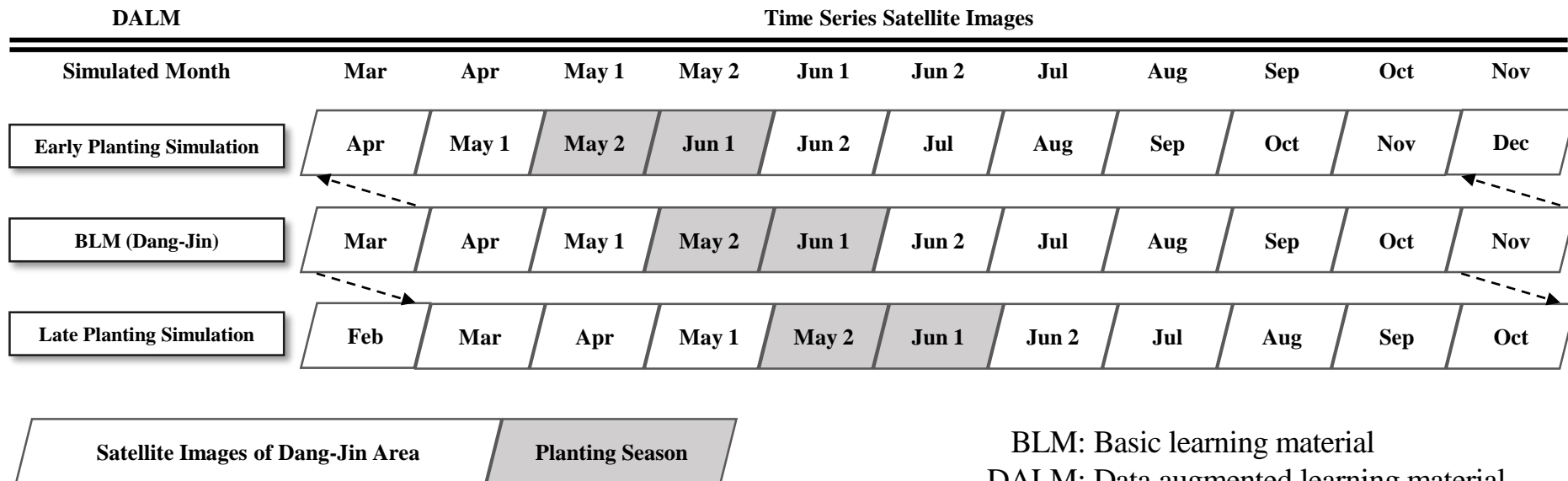
▷ Deep-learning model that is specialized in time-series analysis





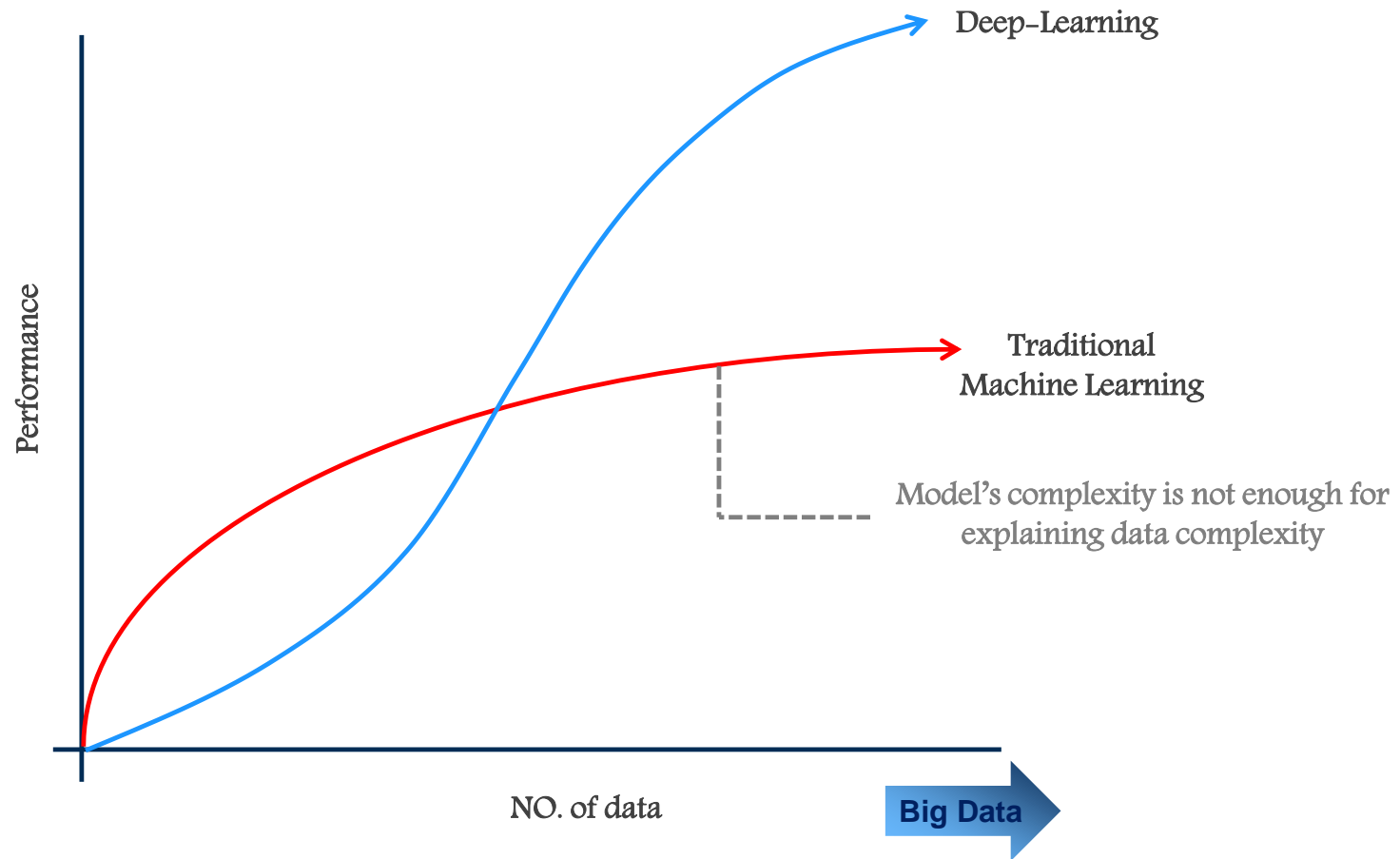
# Data Augmentation

- ▷ Increase the volume of training data
- ▷ Overcome the shortage of labeling data in confined area by simulating rice phenology in other regions
- ▷ Move back and forth the time series of Dang-Jin area

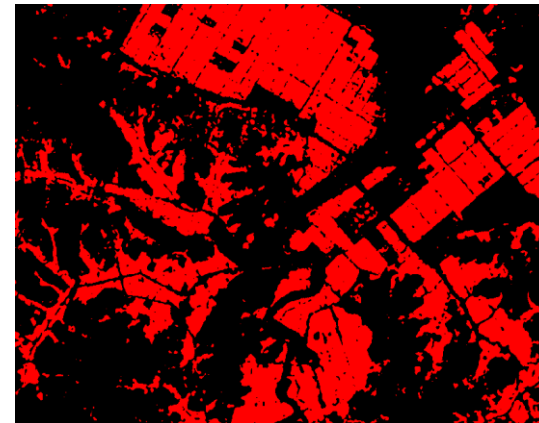
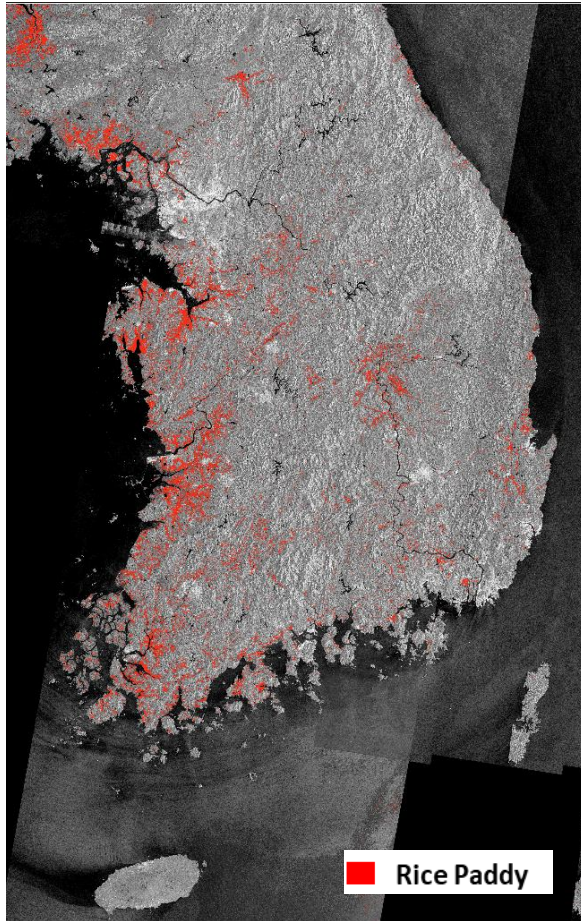


BLM: Basic learning material  
DALM: Data augmented learning material

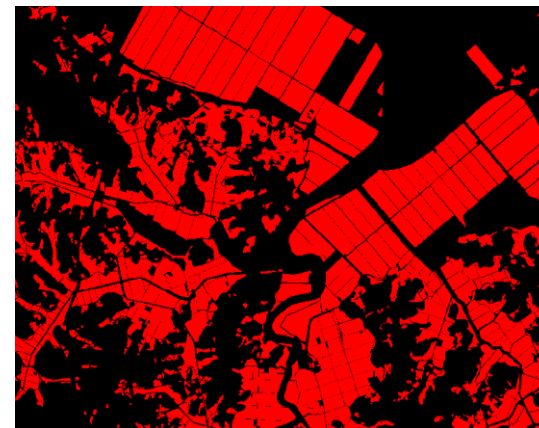
# Deep-learning Application



## Results: Rice paddy detection in South Korea



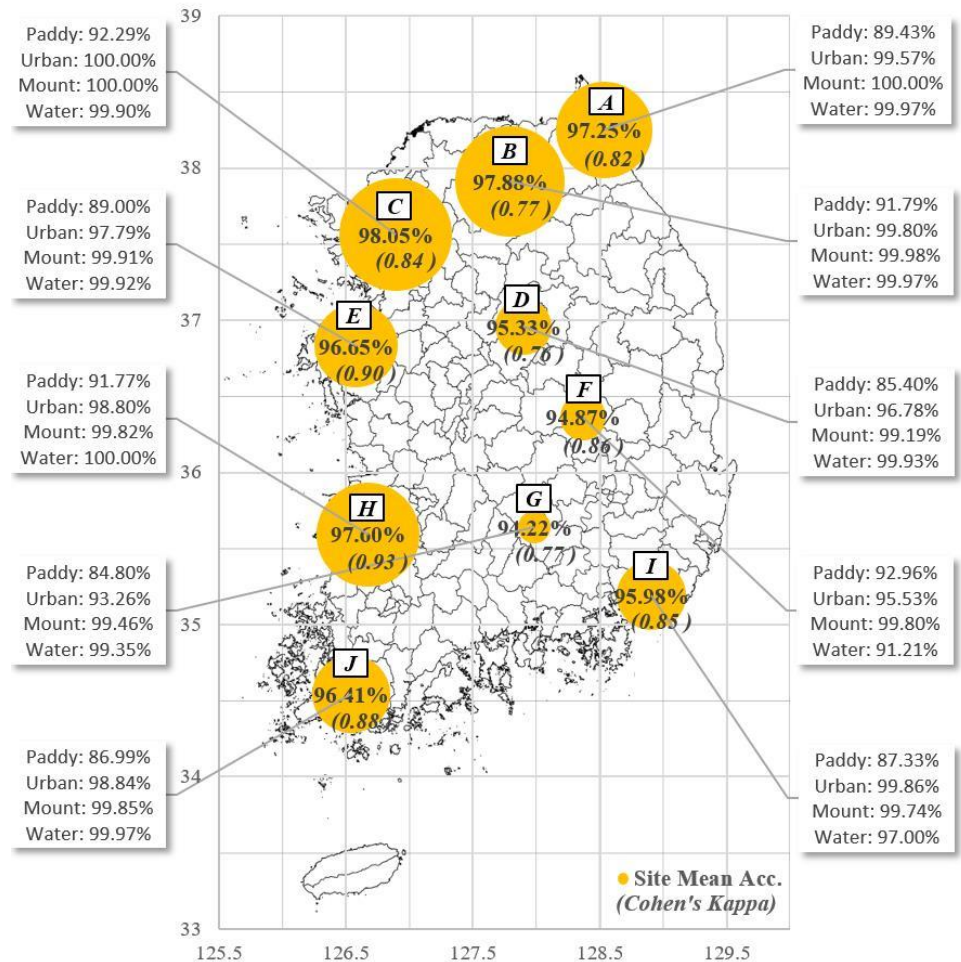
Rice Paddy Detection with RNN



Level-3 Land Cover Map

## Results: Evaluation on 40 sample plots

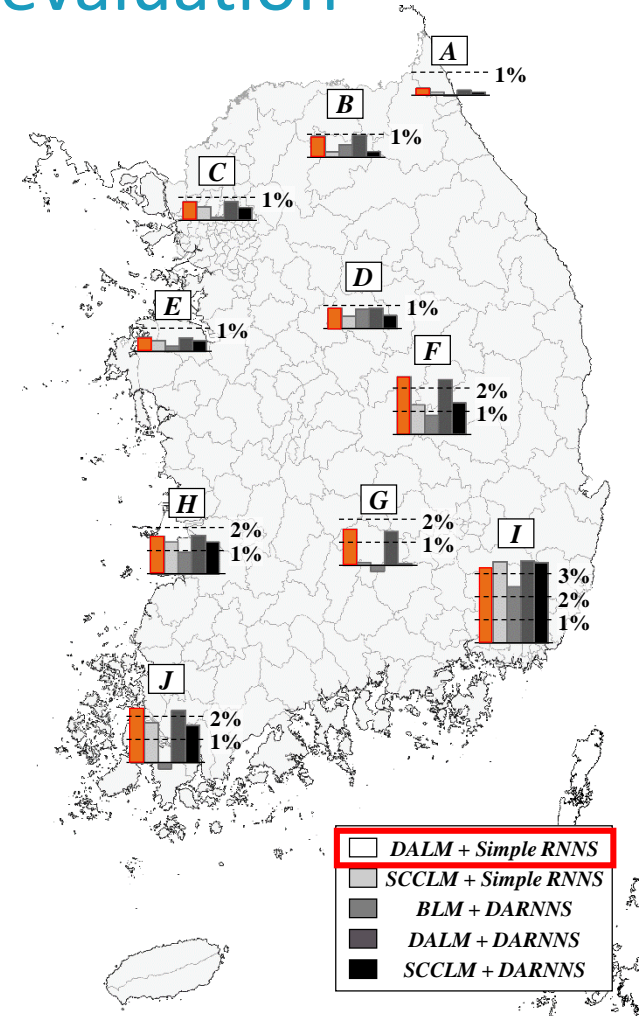
- ▷ 10 sites were selected from the entire nation
- ▷ Each site consists of 4 plots with different landscapes : Agricultural, Urban, Mountain, Water
- ▷ Mean performance in 40 plots : Acc. 96.42% : Cohen's Kappa 0.86





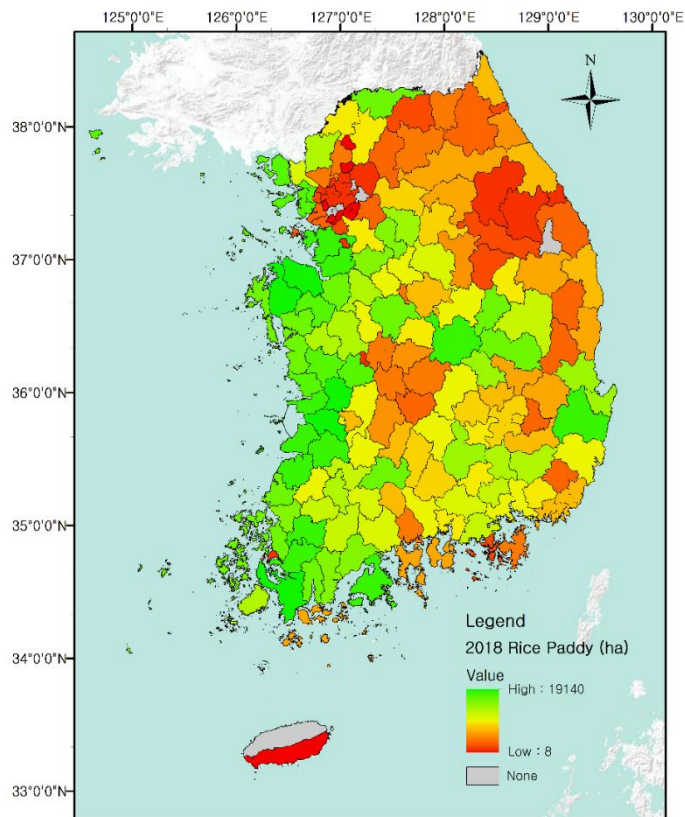
# Results: Technique effectiveness evaluation

- ▷ Accuracy improvement of data augmentation compared with basic learning material
- ▷ Applying data augmentation improved overall accuracy in spite of using confined labeling data in Dang-Jin
- ▷ Area far from Dang-Jin showed greater accuracy improvement

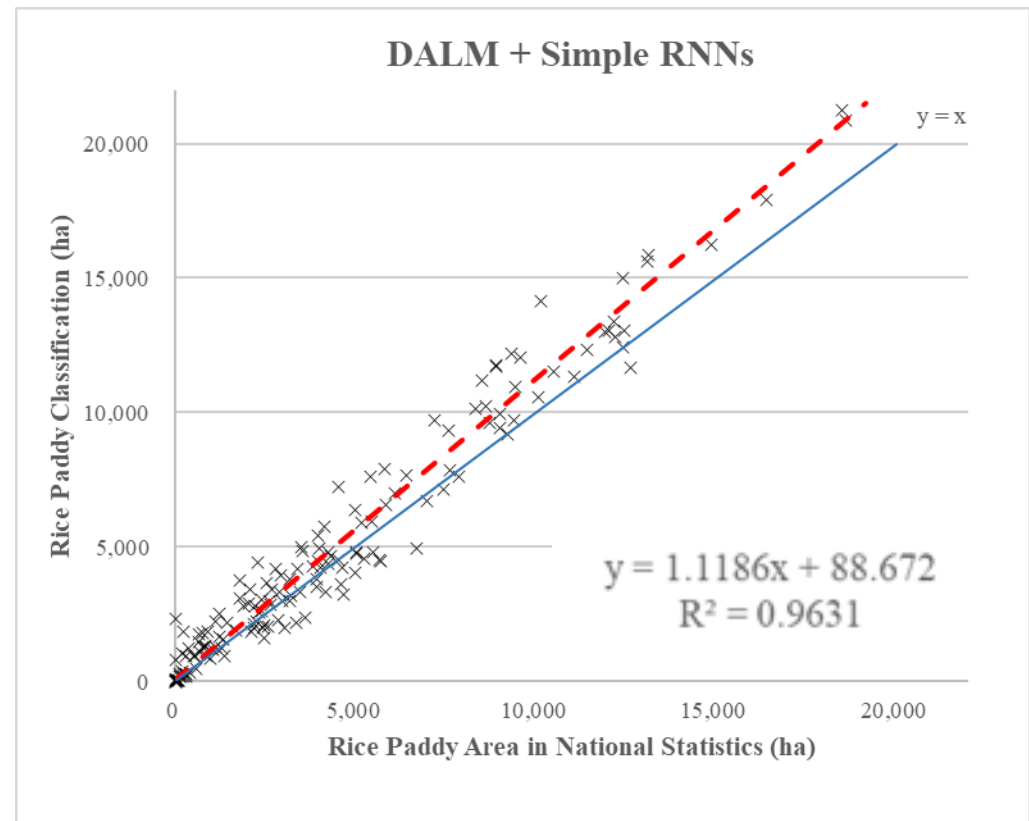


# Results: City-level statistic evaluation

▷ Evaluated in 255 cities



Level-3 Land Cover Map (2015)



# EOPEN: opEn interOperable Platform for unified access and analysis of Earth observationN data

## Consortium Members



# EOPEN Developer Platform

- ▷ From data download to analysis under the platform environment
- ▷ Perform all processes in this presentation with EOPEN platform

## NOA

Sentinel Image  
Download



Preprocessing  
(calibration)

## KU eGIS/RS

Data  
Augmentation



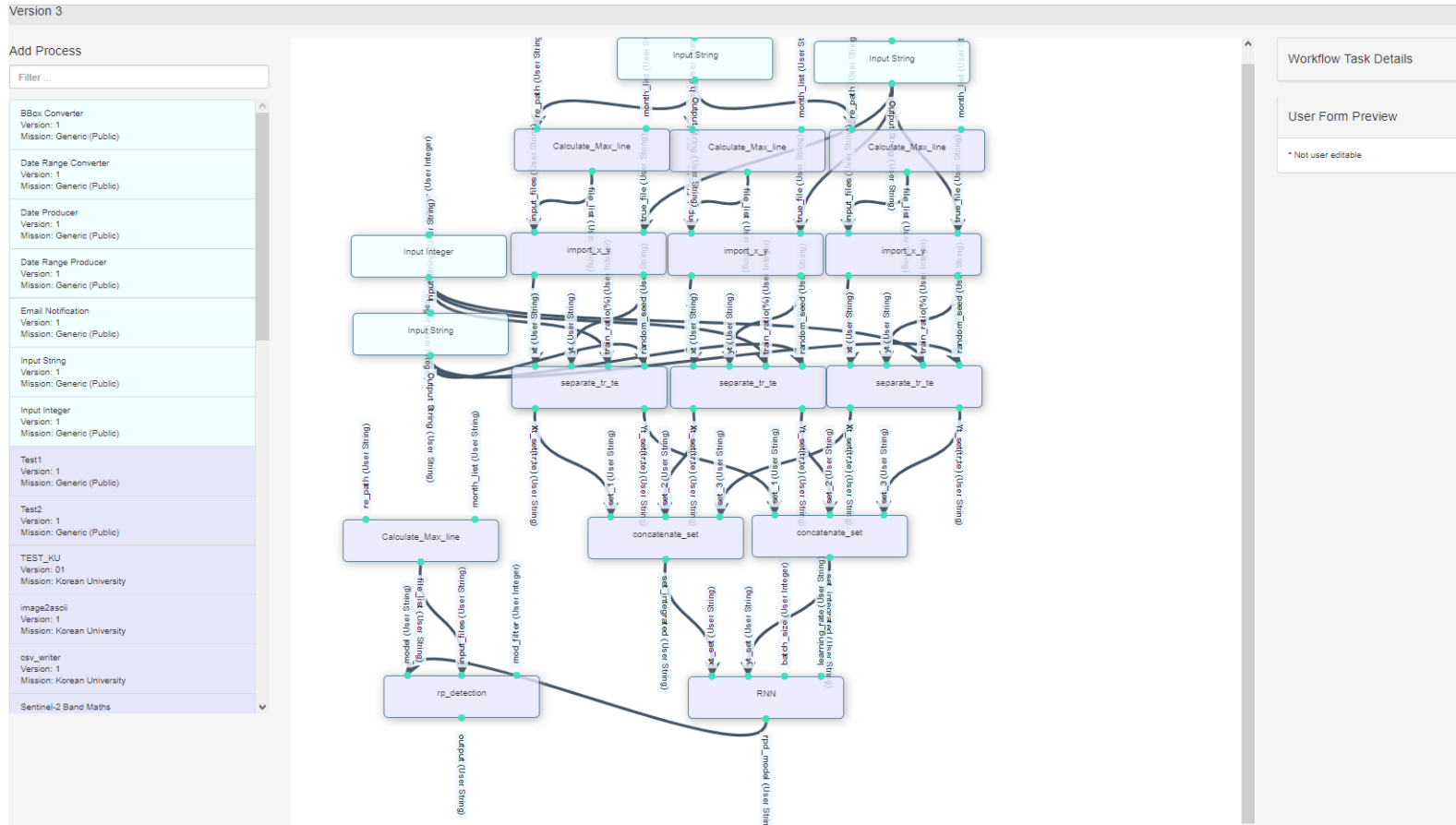
Integration with  
Labeling Data

Deep-learning  
Analysis



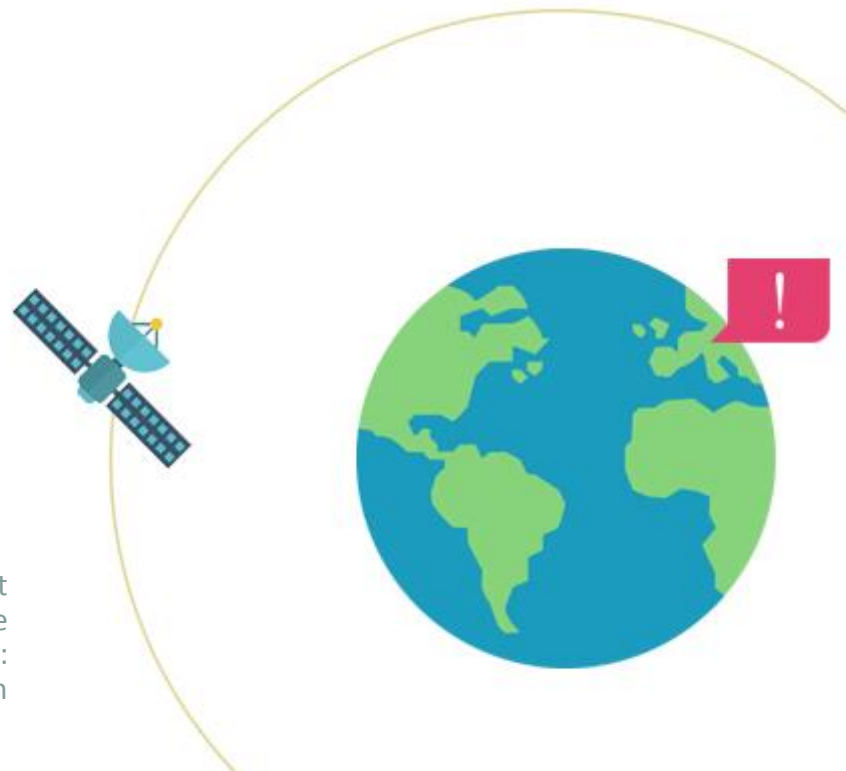
# EOPEN Developer Platform

## ▷ Deep-learning analysis available



# Thanks!

## Any questions?



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