

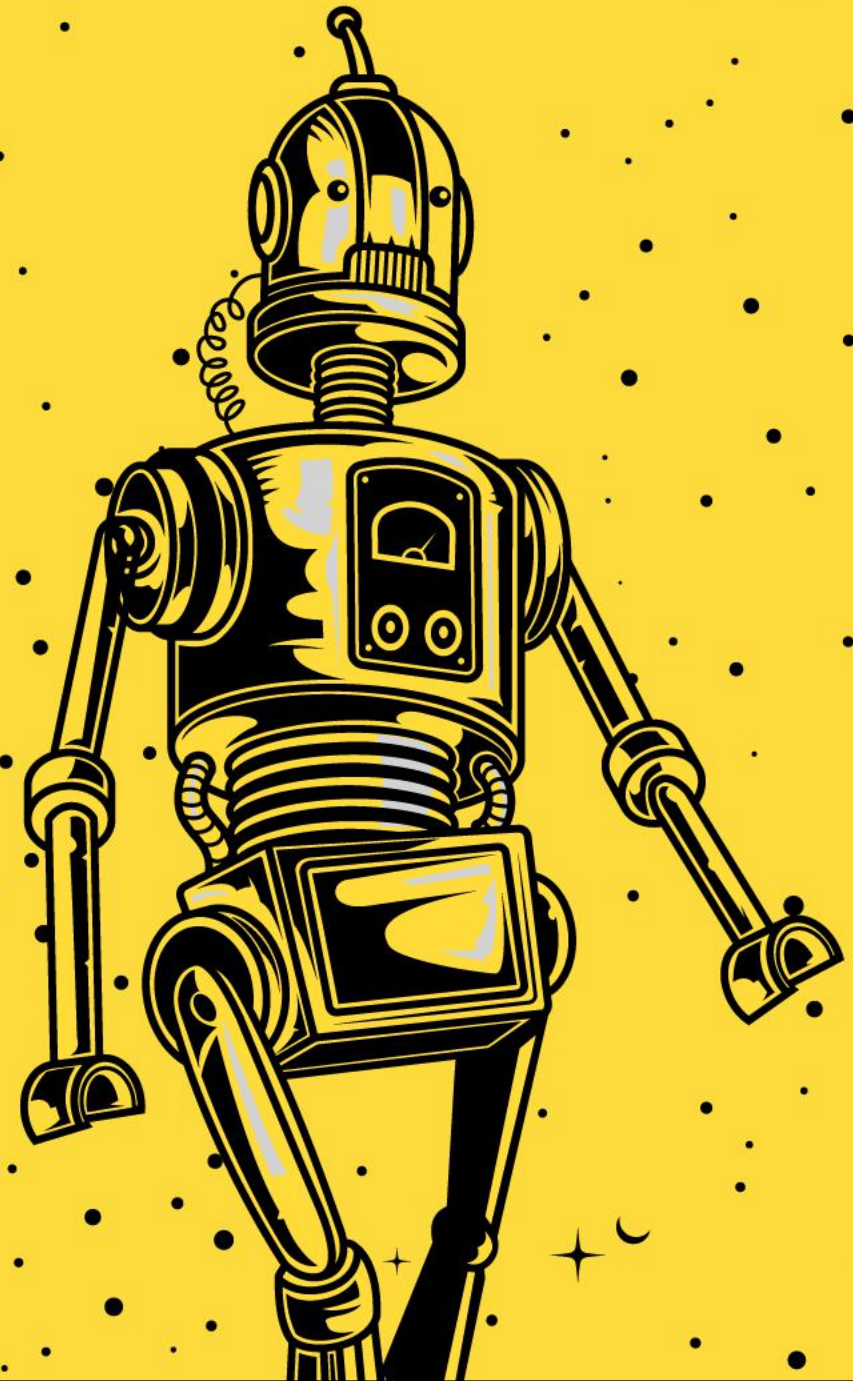
ITdays

#NewPerspectives

10 - 11

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CREIC - Cluj Innovation Park



Enhancing the performance of weather nowcasting by use of machine learning techniques applied on weather data

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Weamyl project

- **Collaboration between:**
 - **Babeş-Bolyai University (BBU)**
 - **Romanian National Meteorological Administration (ANM)**
 - **Norwegian Meteorological Institute (MET)**
- **Major goal: provide an efficient, seamless nowcasting platform which will be integrated with national warning systems**
- **The platform will:**
 - **provide precise nowcasting using Machine Learning methods**
 - **managing large amounts of meteorological data using Big Data approaches**

Weather nowcasting

- Nowcasting: weather forecasts for the next **0 to 6 hours**
- According to World Meteorological Organization (WMO), severe weather is responsible for damage and loss of life
- Good **nowcasting** \Rightarrow issuing of relevant early warnings
- Predicting short-term weather is a *difficult* task for meteorologists
- **Complexity**: huge amount of data which has to be processed and analyzed

Current systems in use

- **Some nowcasting systems used today based on radar data extrapolation:**
 - **TITAN by Dixon and Wiener**
 - useful for single cell tracking
 - **FAST by Jung and Lee**
 - uses fuzzy logic
 - **AROME**
 - small scale numerical prediction model with AROME-NWC for predicting in the range of 0-6 hours
- **Limitations:**
 - Specific for a geographic region - not adaptable
 - Problems with storm segments with irregular shapes or with variable speed

Data used and predicted

- Radar Data:
 - Data provided on a wide area by ground based stations
 - Provides data such as: *Reflectivity*(particle size) and *Velocity*(particle velocity) among others
- Satellite data
 - Satellite imagery provides information on atmospheric and ground conditions including areas not covered by weather radars
 - Broad range of data collected: *surface reflectance, moisture content, ozone concentration* etc.
- Also considering using other data:
 - temperature
 - geographical features
 - other

Not all data is the same

- Radar data gathered can be different in different regions
 - different types of radar can gather different types of data
 - ANM Romania uses single polarization radars
 - MET Norway uses dual polarization radars
 - different data preparation (data preprocessing, data cleaning, data correction etc.)

⇒ difficulty in making one model that fits all cases

- We need to create *separate* prediction models for *separate* regions

Machine learning techniques

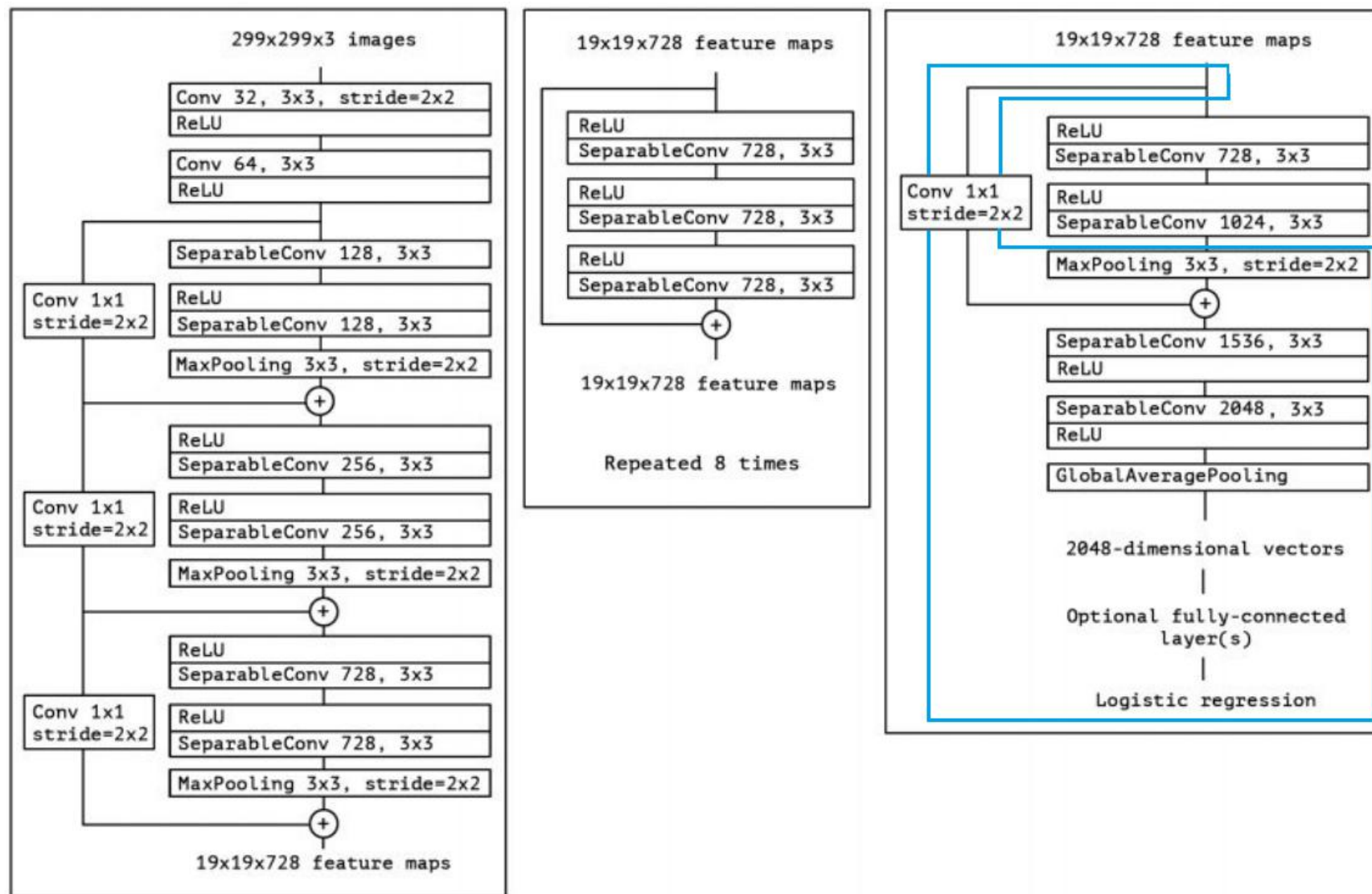
- Main machine learning method used: *Deep Neural Networks*
- Currently our prediction model is based on:
Convolutional Deep Neural Networks
- Some other methods used during our research:
 - self organizing maps
 - autoencoders
 - relational association rules mining

NowcastX model idea

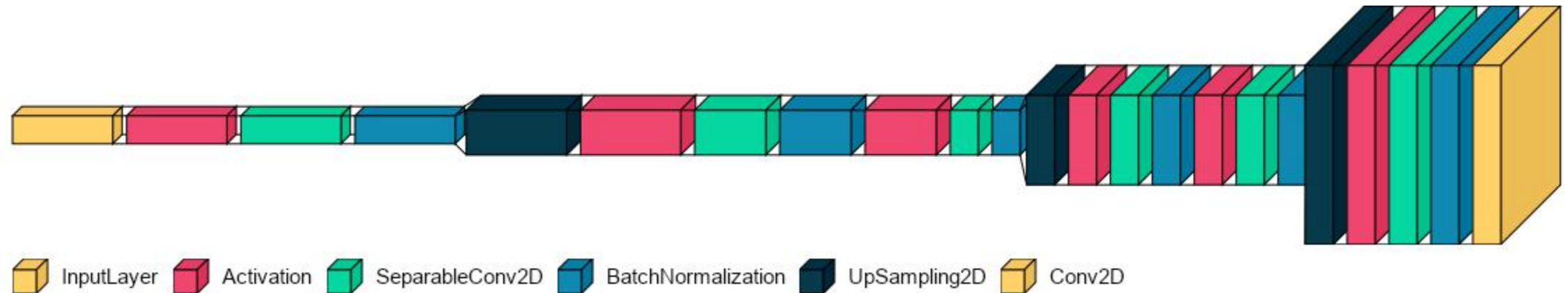
- From the data at one time step predict the data at the next time step
- Adapt data to be view as images:
 - satellite data is already gathered in from of images
 - radar data can be structured into a data model that can be viewed as an image where each channel is a radar product
- Take advantage of powerful Computer Vision neural network models:
The **Xception** Model

Xception model

Figure: The original Xception model, with a mark (the blue lines) on the part that we consider for replacement



NowcastX model decoder



- **Figure:** The decoder that we used in our model replacing the end part of the Xception model.

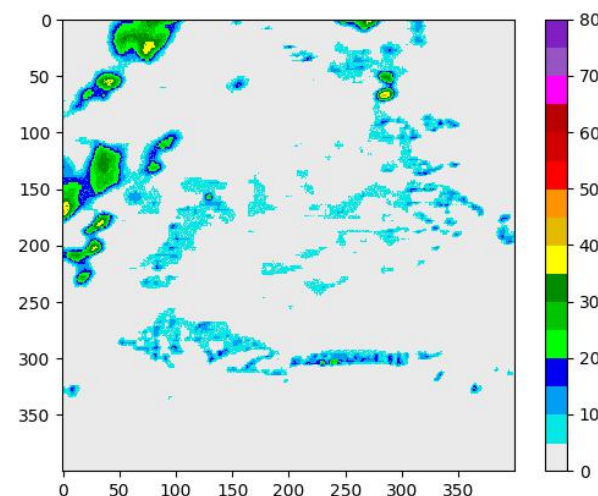
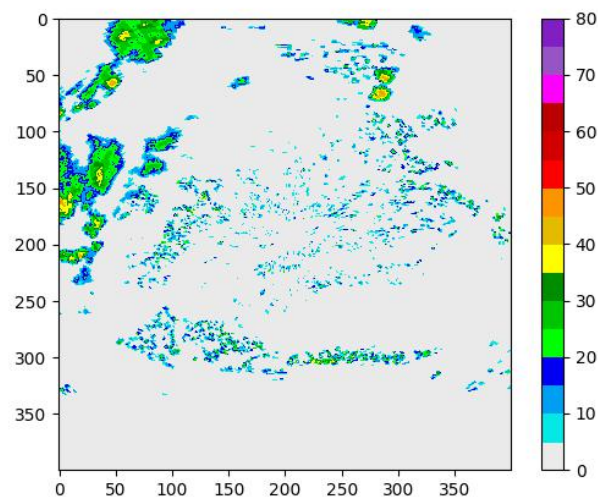
NowcastX results - radar data

Figures: examples of predictions and their comparisons to ground truth for two radar products

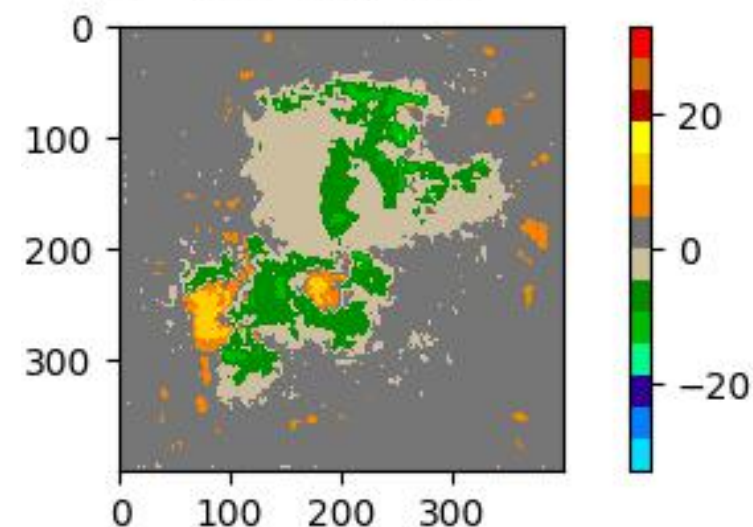
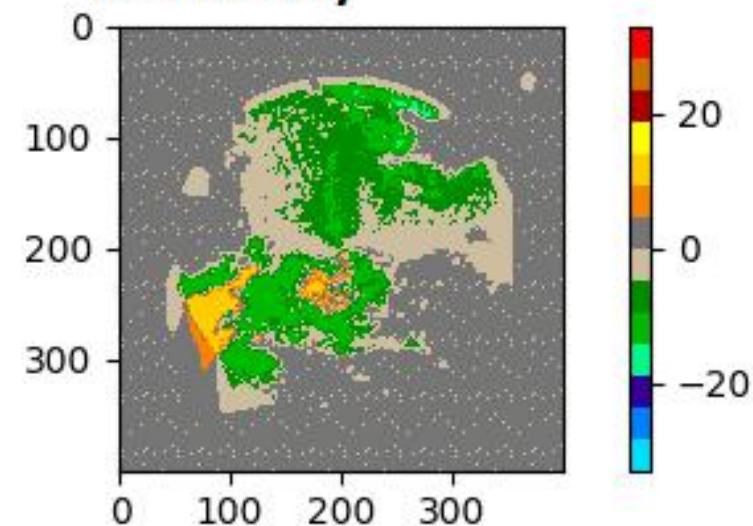
radar data prediction video:

<https://weamyl.met.no/models/>

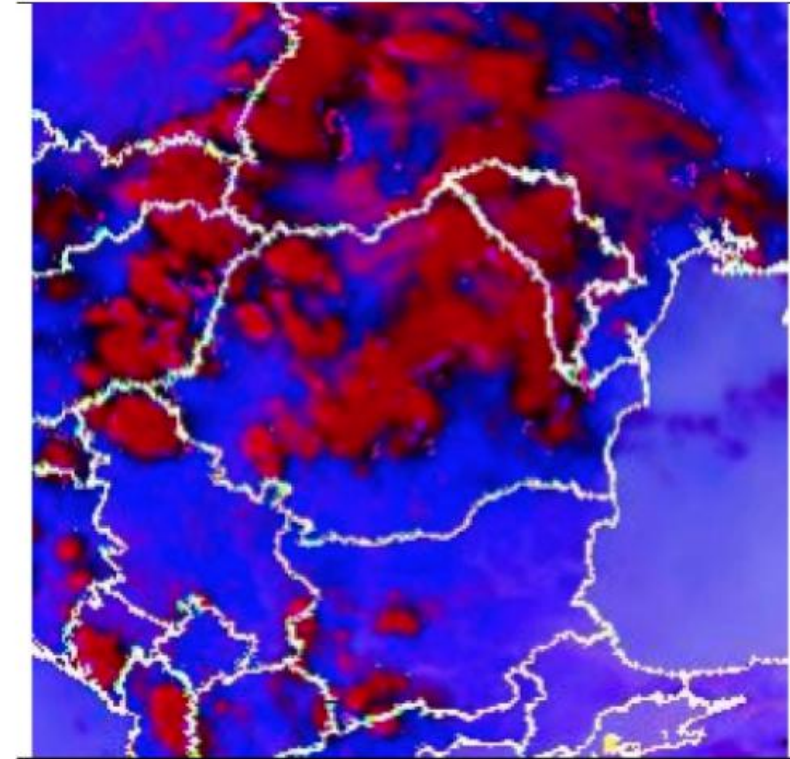
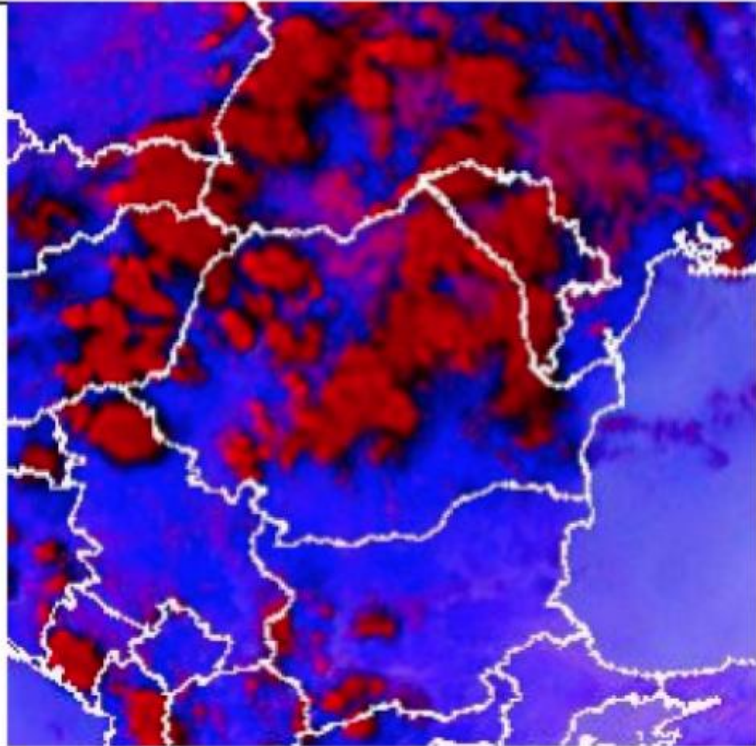
Ground truth and prediction
for Reflectivity



Ground truth and prediction
for Velocity



NowcastX results - satellite data



Figures: grund truth (left) and prediction (right) for a satellite image.

NowcastX model variations

- Research in progress
- The model needs to be adapted and perfected
- Techniques we use:
 - adapting the training loss:
e.g. give more weight to non-zero values or to higher values
 - train and predict separately for each product
- Too much fine-tuning may lead to overfitting for one dataset or for one testing measure

Evaluation measures

- Regression measures: Mean Squared Error (MSE) and Root Mean Squared Error (RMSE)

- $MSE = \frac{1}{n} \sum_{i=1}^n (real_i - predicted_i)^2$

- $RMSE = \sqrt{MSE}$

- Non-Zero RMSE

- Classification measures: Critical Success Index (CSI)

- $CSI = \frac{True\ Positive}{True\ Positives + False\ Positives + False\ Negatives}$

- Expert feedback

Difficulty comparing to related work

- In order to assess the model comparison to state of the art is needed
- Difficulties in comparisons with other experiments:
 - Different experiment design (which products used, and how, regression vs classification etc.)
 - Different nowcasting predictions (e.g. predicting precipitations instead of reflectivity)
 - Different dataset over different regions (using different kinds of radars etc.)
- Often the dataset or model are not accessible/open source \Rightarrow direct comparisons are impossible

Conclusions and Future Work

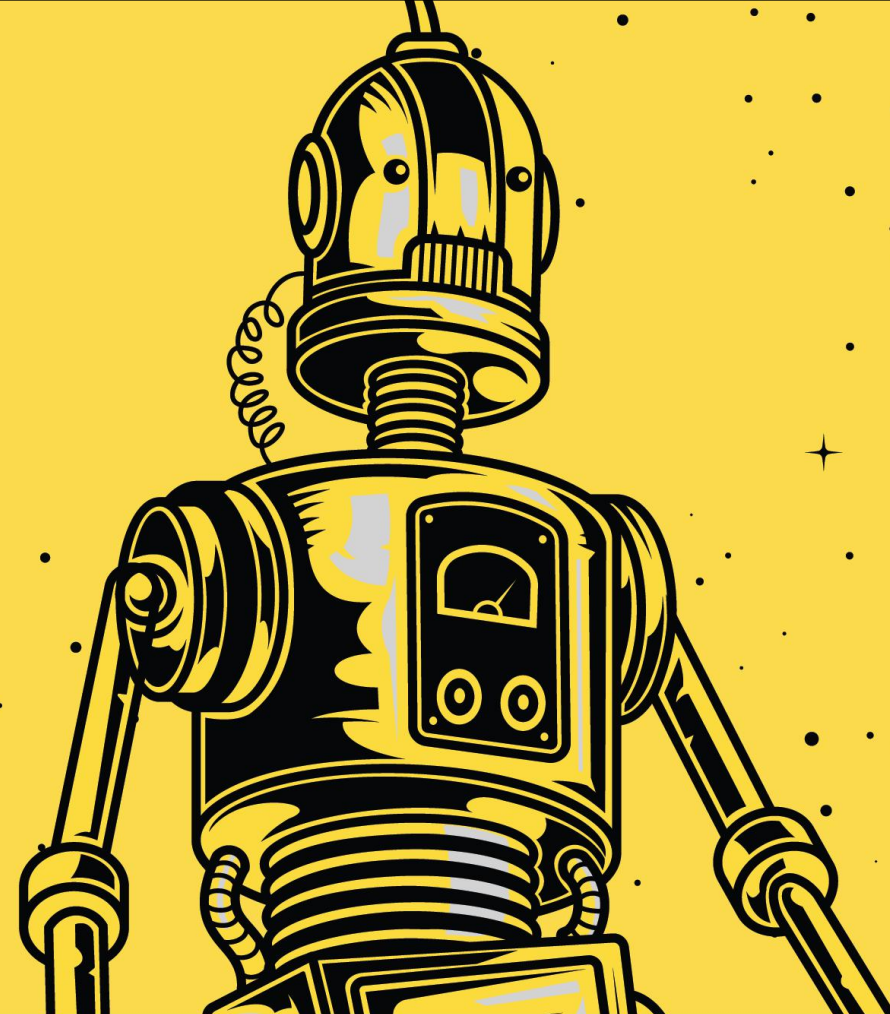
- **The WeaMyL project:** a collaboration project between UBB, MET and ANM to develop a machine learning-based nowcasting platform to be integrated in national warning systems
- **The NowcastX model:** our current weather nowcasting model based on Convolutional Deep Neural Networks
- **Future work:**
 - Perfect the NowcastX model
 - Predict further into the future
 - Create a model based on Convolutional LongShort Term Memory Networks (ConvLSTM)

IT days

#NewPerspectives



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