

WEATHER CLIMATE WATER

CIMO TECHNICAL CONFERENCE ON METEOROLOGICAL AND ENVIRONMENTAL INSTRUMENTS AND METHODS OF OBSERVATION (CIMO TECO-2018)

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Operational Early Warning System in Southern Brazil – Blending Remote and Surface Observations for Precipitation Nowcasting

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Lightning Incidence in Southeastern South America

Observed from Satellite (LIS/OTD)



High Resolution Full Climatology Annual Flash Rate

Global distribution of lightning April 1995-February 2003 from the combined observations of the NASA OTD (4/95-3/00) and LIS (1/98-2/03) instruments (Adapted from Goodman & Cecil 2002)



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Lightning Incidence in Southeastern South America

Observed from Surface Global Lightning Detection Network (GLD360)





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Hydrometeorological Monitoring Networks in Brazil



INMET Conventional Weather Stations (446)



CEMADEN Automatic Rain Gauges (1581)





Regional Hydrometeorological Monitoring Networks in the South of Brazil

(+250 AWS)

"a network of networks"





Weather Radar Coverage in Brazil

Period: 1980 – 2000



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TOTAL = 23 Single Pol





Weather Radar Coverage in Brazil

Period: 2011 – 2018

TOTAL = 25 Single Pol & 20 Dual-Pol

OMM

WMO

SIME





Lightning Detection Networks in Brazil - RINDAT





Lightning Detection Networks in Brazil - RINDAT





Lightning Detection Networks in Brazil - RINDAT



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Lightning Detection Networks in Brazil - RINDAT





Lightning Detection Networks in Brazil – BRASILDAT







Lightning Detection Networks in Brazil – Examples





Regional Hydrometeorological Service - SIMEPAR

- Operational Hydro-Meteorological Monitoring (2 S-Band & X-Band Radars, Lightning Detection, +200 gauges & AWS) and Hydro-Met Forecasting (Nowcasting, Short-Term Forecasting, Hydrology Fcst)
- R&D in Meteorology, Hydrology, and Atmospheric Electricity
- Cooperation with:
 - Reg. Hydromet Centers
 - National Met Services (INMET, DINAC)
 - Natl. NWP Center (CPTEC)
 - Natl. Center for Disaster Monitoring and Alerts (CEMADEN)
 - Civil Protection





Regional Hydrometeorological Service - SIMEPAR

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Energy Production and Transmission

- 85% is Hydro power in Brazil
- 35% is produced in Parana river basins
- Long transmission lines to urban centers

Agriculture

- 1st largest wheat and soy production
- 2nd largest corn and beans production





Dual Pol. Radar QPE and Multi-Sensor QPE (radar+satellite+gauge - SIPREC)





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Flooding in Urban Area

CAS weather radar: S-Band, Dual Polarization

Quantitative Precipitation Estimation

Reflectivity – Rainfall Rate Relationships

Marshall-Palmer: $Z = 200 R^{1.6}$

NEXRAD: $Z = 300 R^{1.5}$

SIMEPAR Distrometric: Z = 288 R^{1.4} (Calheiros et al 2017)

Multi-Sensor QPE: Radar + Satellite + Gauge

 $\nabla^2(R_{radar} + R_{satellite}) = P_{Gauge}$ (Calvetti et al 2015)

Polarimetric – Rainfall Rate Relationships

R (Z,ZDR) = $(1.42 \times 10^{-2}) Z^{0.770} Z_{dr}^{-1.67}$

 $R(Z,KDP) = 44.0 |K_{DP}|^{0.822} sign(K_{DP})$

(Ryzhkov & Zrnic 1996, Gorgucci et al 1999, Ryzhkov et al 2005, Vulpiane & Baldini 2013, among others)

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(Beneti et al, AMS 2018)



Dual Pol. Radar QPE and Multi-Sensor QPE (radar+satellite+gauge - SIPREC) Case: 2016-10-11





Dual Pol. Radar QPE and Multi-Sensor QPE (radar+satellite+gauge - SIPREC) Case: 2016-10-12





Dual Pol. Radar QPE and Multi-Sensor QPE (radar+satellite+gauge - SIPREC)





Multi-Sensor QPE (radar+satellite+gauge + LIGHTNING?)

Evaluating GOES16/GLM vs RINDAT for Convective Storm Identification and QPE





Multi-Sensor QPE (radar+satellite+gauge + LIGHTNING?)

Evaluating EN (PulseRad) vs Dual-Pol S-Band Radar for Convective Storms Identification and QPE





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Comparison of Lightning and Radar Reflectivity



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High Impact Weather Nowcasting Using Machine Learning







High Impact Weather Nowcasting Using Machine Learning

Data from several radar formats are received, processed, converted (to MDV) and a Weather Radar Mosaic is created and used for the Machine Learning Nowcasting Algorithm

Data Set

- Weather Radar: TITAN thunderstorm tracks and histories from mosaic (4 km²)
- RINDAT Lightning: Density maps (4 km²) separated in +CG_s and -CG_s



WMO OMM

SIM





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SUPERVISED MACHINE LEARNING MODEL



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Data from several radar formats are received, processed, converted (to MDV) and a Weather Radar Mosaic is created and used for the Machine Learning Nowcasting Algorithm Results using Linear Regression Model (Huber) and Ensemble Model (Random Forest)

- Test data set (20% of random sampled from original dataset Aug 2016 to Aug 2017)
- Validation data set (from Sept 2017 to Jan 2018)



(Beneti et al, ERAD2018)



Conclusions and Future Work

- Multi-sensor QPE (Radar + Satellite + Gauge) has been in use for over 10 years, in hydrology applications (input for stream flow forecast model) in operational hydropower companies, with better results than using gauge or gauge+radar alone.
- Radar QPE, with dual polarization, can be improved if considered not only a Z-R relation or R(Z,ZDR,KDP) alone, which has been applied to the Brazilian weather radar network, but with particle identification and stratification of the R(Z,ZDR,KDP) relation.
- Although a lightning QPE can be tricky to obtain/use, satellite lightning observations (GLM) and total lightning detection networks have demonstrated that there is good agreement specially to identify convective regions within the storms, which can then be used to improve QPE algorithms.
- Machine learning algorithms applied to thunderstorms forecasting/nowcasting showed better results than TITAN algorithm alone. <u>Random Forest</u> model presented better results for the storm centroid location and ellipses sizes/shape.
- An algorithm for nowcasting precipitation, using machine learning to forecast storm displacement and also multi-sensor QPE data fusion is under development.





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Thank You!

Bedankt!

