Improving CFS Precipitation and 2m Temperature Anomaly Outlooks from Week-1 to Week-6 with Machine Learning

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Outline

- Motivation
- NN Basic
- Preliminary Results
- Summary

Motivation

Demand for S2S P & T2m Fcst Steadily Increasing Problem: Low Forecast Skill Post-Processing:

Data Sets ${ (f_1, f_2,, f_n)_p, O_p }_{p=1,2,....N}$ Where $f_1, f_2,, f_n -- predictors: 1999-2019 daily BC CFSv2$ Week 1-6 P & T2m fcsts, O_p -- predictands: 1999-2019 daily Week 1-6 P & T2m Obs

Mapping: O = M(F)

Traditional method: MLR -- MOS

Can AI or Machine Learning beat BC CFS?



8 Years Cross-Validation



1999 – 2019 daily Week 1-6 forecasts = 6 x 7670 daily samples over North America



Forecast Week-1 P Daily RMSE (mm) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst

Forecast Week-1 T2m Daily RMSE (°C) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst



North American Domain

Time Series of Forecast Week-1 T2m Daily Spatial Anomaly Correlation



North American Domain



Forecast Week-1 P & T2m Daily Spatial Anomaly Correlation (2012-2019)

Forecast Week-2 P Daily RMSE (mm) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst

Forecast Week—2 T2m Daily RMSE (°C) & AC (2012—2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst



Time Series of Forecast Week-2 P Daily Spatial Anomaly Correlation





North American Domain



Forecast Week-2 P & T2m Daily Spatial Anomaly Correlation (2012-2019)

North American

Forecast Week-3 P Daily RMSE (mm) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst

Forecast Week-3 T2m Daily RMSE (°C) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst



Time Series of Forecast Week-3 P Daily Spatial Anomaly Correlation

North American Domain



Time Series of Forecast Week-3 T2m Daily Spatial Anomaly Correlation

North American Domain



Forecast Week—3 P & T2m Daily Spatial Anomaly Correlation (2012—2019)

Forecast Week-4 P Daily RMSE (mm) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst

Forecast Week-4 T2m Daily RMSE (°C) & AC (2012-2019)



CFS: Bias corrected NNv: Yearly Cross-Validation NNd: Dependent Fcst



Time Series of Forecast Week-4 P Daily Spatial Anomaly Correlation

North American Domain



Time Series of Forecast Week-4 T2m Daily Spatial Anomaly Correlation

North American Domain



Forecast Week—4 P & T2m Daily Spatial Anomaly Correlation (2012—2019)







Time Series of Forecast Week-1 P Daily Spatial Anomaly Correlation

North American Domain



Time Series of Forecast Week-1 T2m Daily Spatial Anomaly Correlation

North American Domain



Forecast Week 1-6 Spatial Anomaly Correlations (2012-2019)



Observed and Forecast Week 5 Prop Anomalies (mm) 06Feb2017



Week-5 Prcp Forecasts

Observed and Forecast Week 5 Prcp Anomalies (mm) 12Mar2017



Observed and Forecast Week 5 Prop Anomalies (mm) 25Nov2018















Observed and Forecast Week 5 T2m Anomalies (*C) on 23Mar2017

Week-5 T2m Forecasts







Observed and Forecast Week 6 Prop Anomalies (mm) 13Aug2017





Week-6 Prcp Forecasts

Observed and Forecast Week 6 Prcp Anomalies (mm) 07Apr2019



Observed and Forecast Week 6 Prcp Anomalies (mm) 24Apr2018



Observed and Forecast Week 6 T2m Anomalies (*C) on 15Mar2017



Observed and Forecast Week 6 T2m Anomalies (*C) on 06Sep2017

ved and Forecast Week 6 T2m Anomalies (°C) on 240ct2017



Observed and Forecast Week 6 T2m Anomalies (*C) on 11Jan2018





Observed and Forecast Week 6 T2m Anomalies (*C) on 05Apr2018







Week-6 2m Temperature Forecasts



Observed and Forecast Week 6 T2m Anomalies (*C) on 03Jan2019 -21 -12 -15 -12 -19 -06 -03 03 16 09 12 15 1.8 21

Observed and Forecast Week 6 T2m Anomalies (*C) on 07Jan2019











Summary

1. NN advantages

Flexible nonlinear tool & Easy to handle BIG DATA

2. Unique & beneficial NN architectures: account for

Non-Linear Impact, Pattern Relationship, Co-Variability

3. NN Significantly Improves CFS Week 1-6 P & T2m Fcsts

more sophisticated info hidden behind multiple dimensional big data can be extracted by NN

4. NN can perform more complicated corrections

by reversing incorrect forecast patterns, hardly done by traditional MLR

5. Forecast skills in Week 3-6 ranges have similar tendency

good forecasts in Week 3 tend to be good up to Week 6