



META-MODELING STRATEGY FOR DATA-DRIVEN FORECASTING

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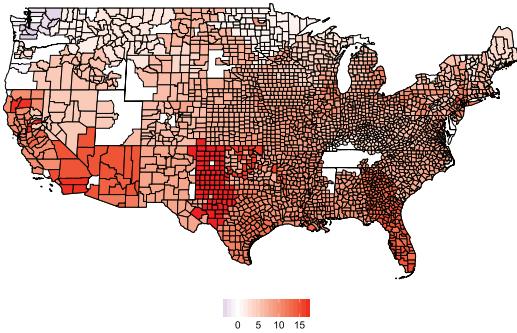
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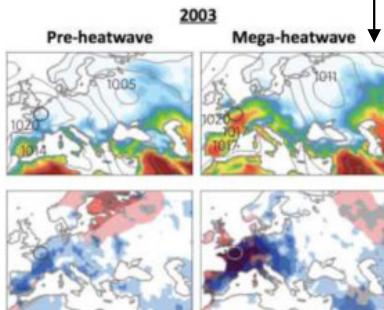
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WEATHER FORECASTING AND CLIMATE CHANGE



Increase in peak power demand,
Auffhammer et al. (2016)

Climate mitigation
requires forecasting

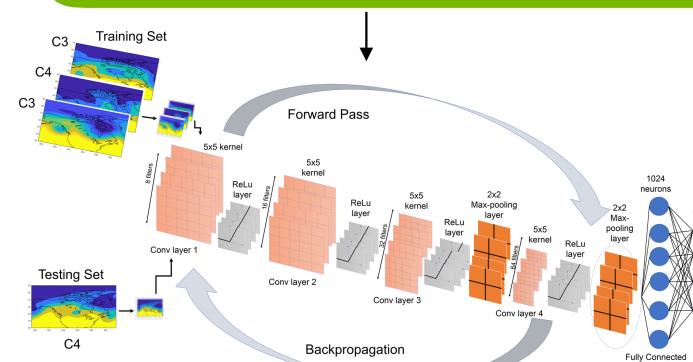


Extreme event forecasting,
Sillmann et al. (2017)

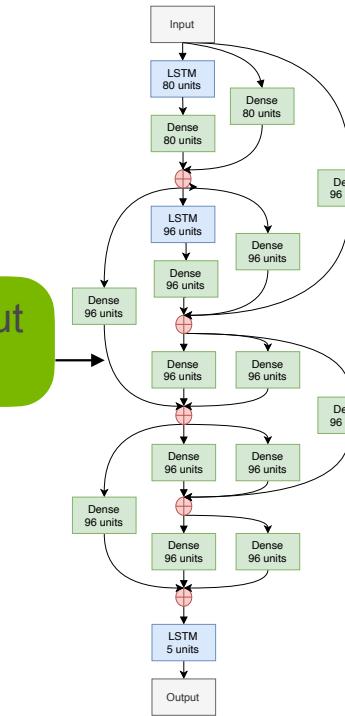
...but weather is
becoming less
predictable, Scher
and Messori (2019)



ML can provide data-driven forecasts, but
can be expensive to train/deploy



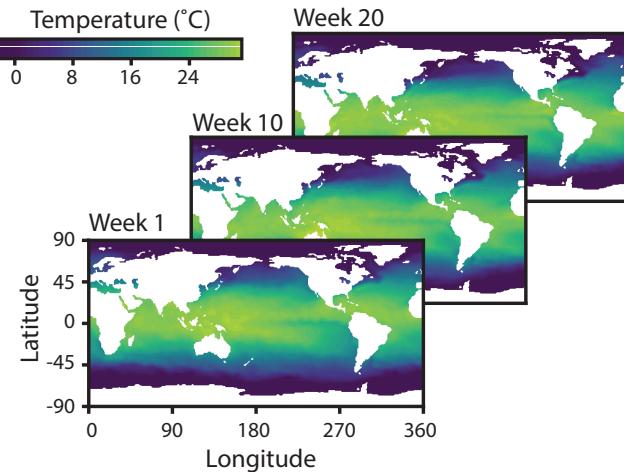
Convolutional neural networks
Chattopadhyay et al. (2020)



Automated architecture,
Maulik et al. (2020)

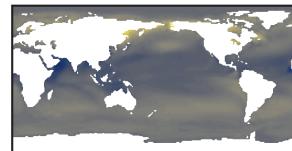
NOAA SEA SURFACE TEMPERATURE

Sea surface temperature measured weekly across 20 years



Proper orthogonal decomposition

Mode 3

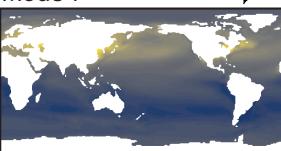


Mode 2

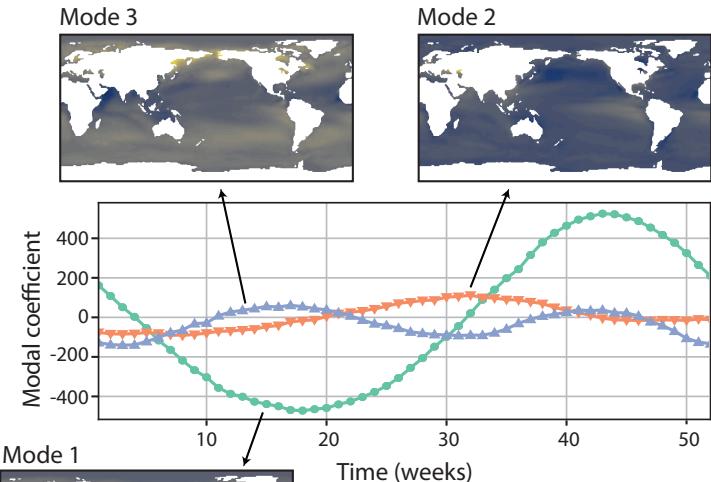


Modal coefficient

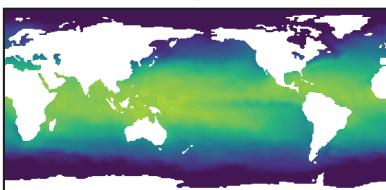
Mode 1



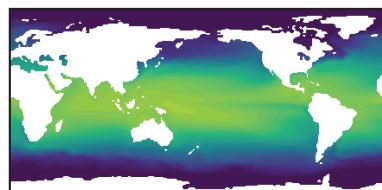
Time (weeks)



Measured temperature field



6 mode POD reconstruction



Make predictions for modal coefficients, then reconstruct to compare against data



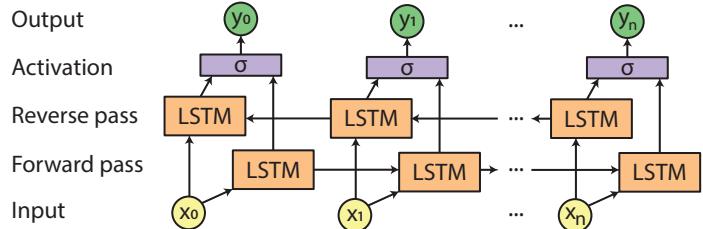
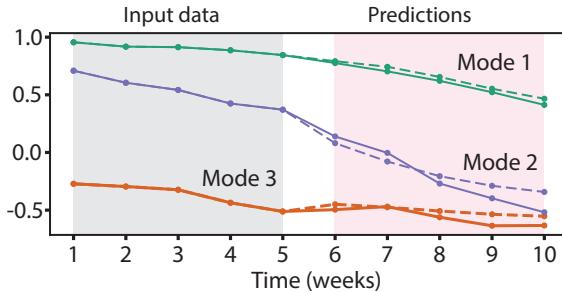
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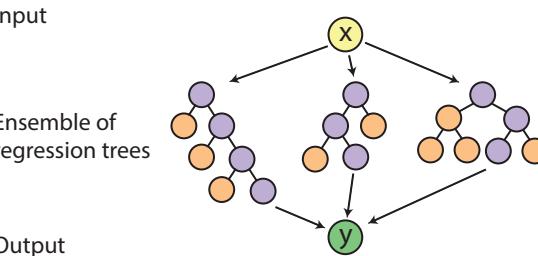
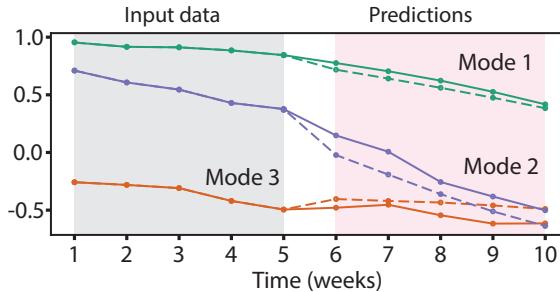
HIGH AND LOW FIDELITY FORECASTS

Prototypical high-fidelity model

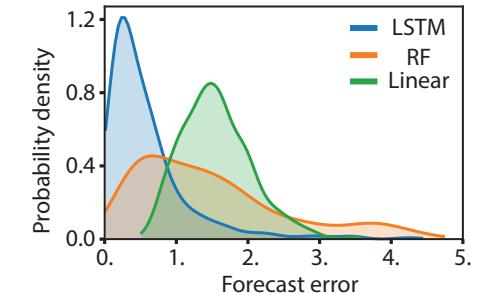
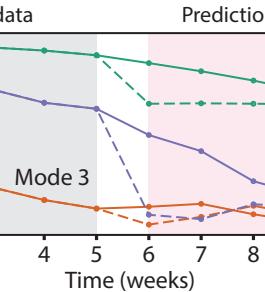
Bi-directional LSTM



Random forest



Prototypical low-fidelity models



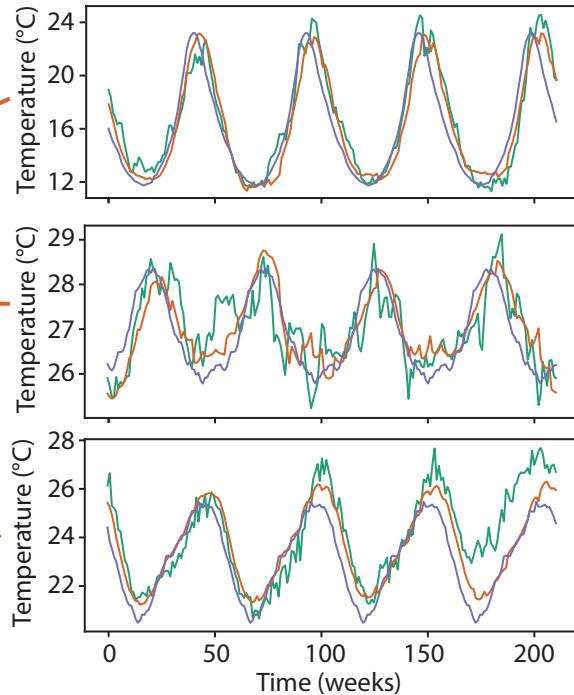
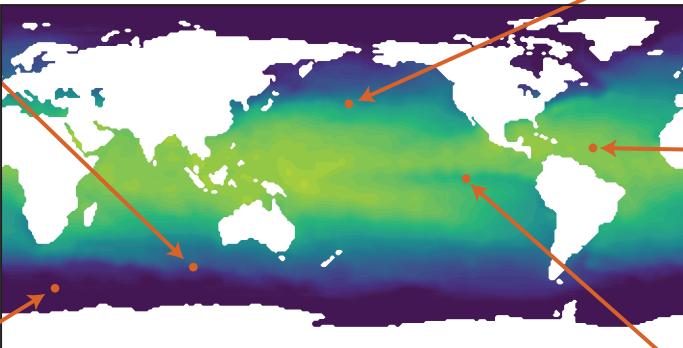
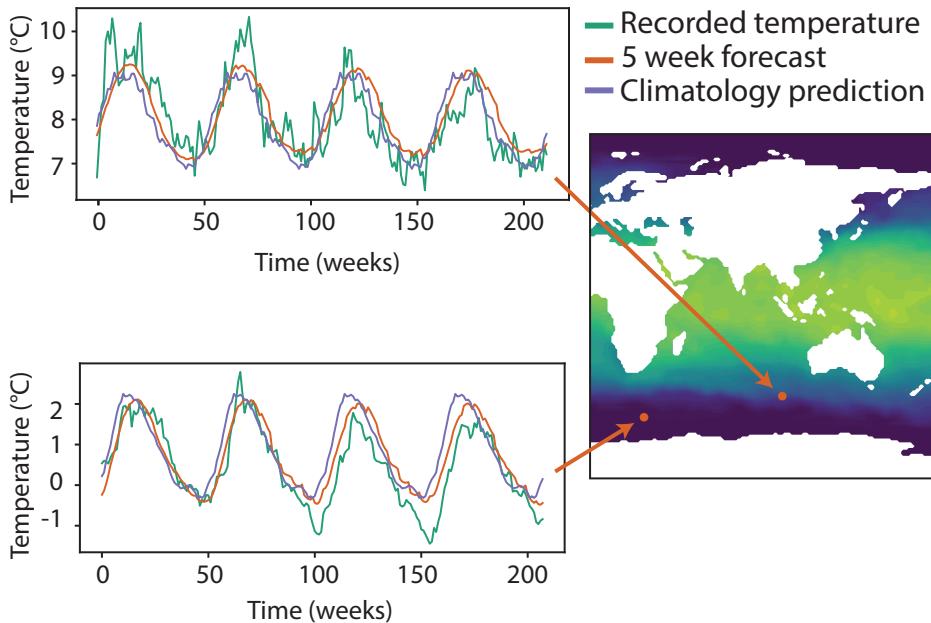
High fidelity performs best on test data



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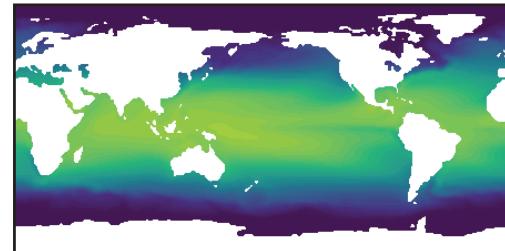
FORECASTING

Method outperforms climatology baseline: The average temperature for that time of year in that location



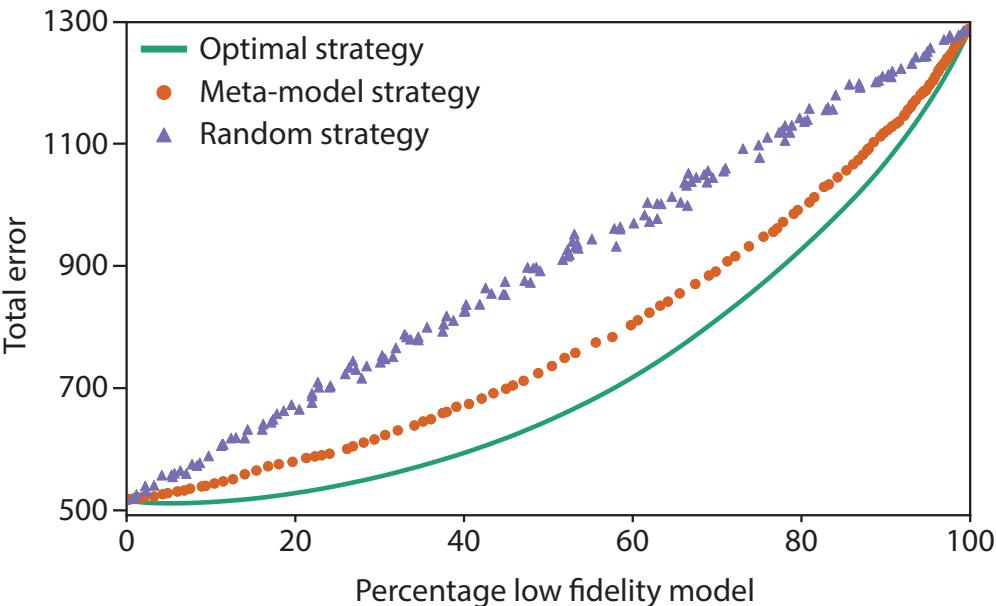
META-MODELING STRATEGY

High fidelity function evaluations are expensive. How can we avoid them, without compromising accuracy?



Algorithm for model selection:

- Take input data and perform low-fidelity forecasts
- Calculate difference between forecasts
- If this exceeds threshold evaluate high-fidelity forecast
- Else use low fidelity forecast

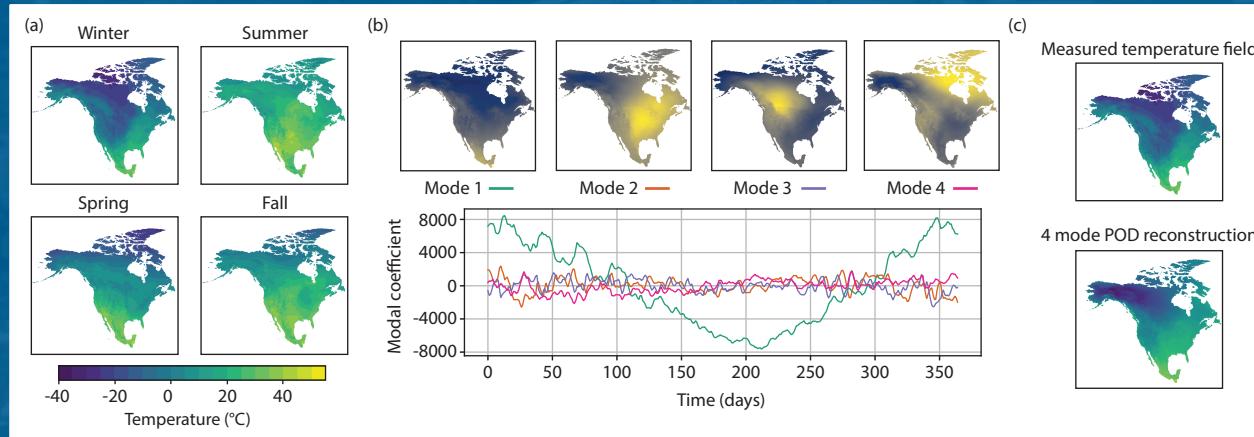


Random strategy chooses to evaluate high or low fidelity models at random

SUMMARY:

- Climate mitigation requires accurate weather predictions machine learning methods can play an important role.
- A combination of low and high-fidelity models can make accurate predictions fast and make results interpretable

Also applied to DayMet dataset



For more details
see workshop
manuscript