

Short-term Wind Forecasting using Off-site Observations and Numerical Weather Prediction

Dr. Kristin Larson

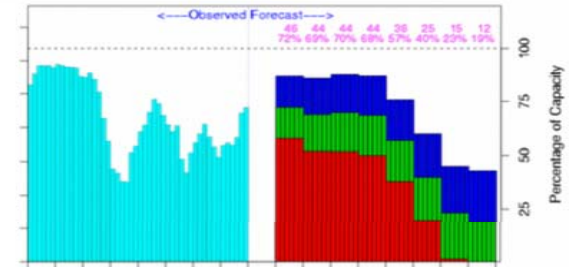
3TIER Environmental Forecast Group

Outline

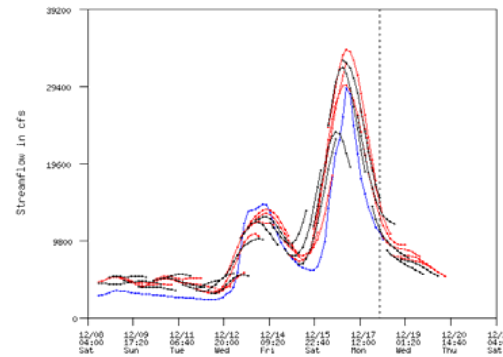
- Overview of 3TIER
- Short-range Forecasts using the **R**egime-switching **S**pace-**T**ime (**RST**) method
- Short-range Forecasts using **N**umerical **W**eather **P**rediction (**NWP**) and the challenge method
- Conclusions

- Founded in 1999
- Forecast processes based on latest *proven* scientific techniques
- Close relationship with leading research and operational communities
- Focused on delivery of *environmental forecast products* for renewable energy resources at multiple time scales

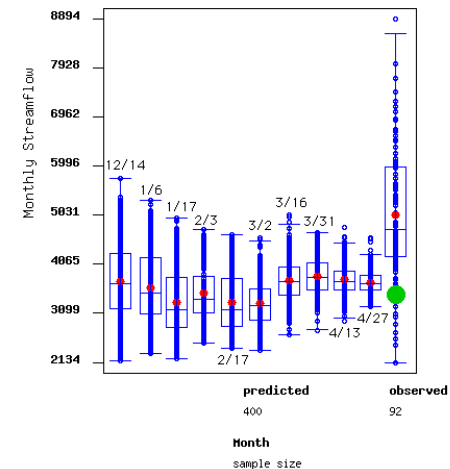
Wind Energy:



Hydro Operations:

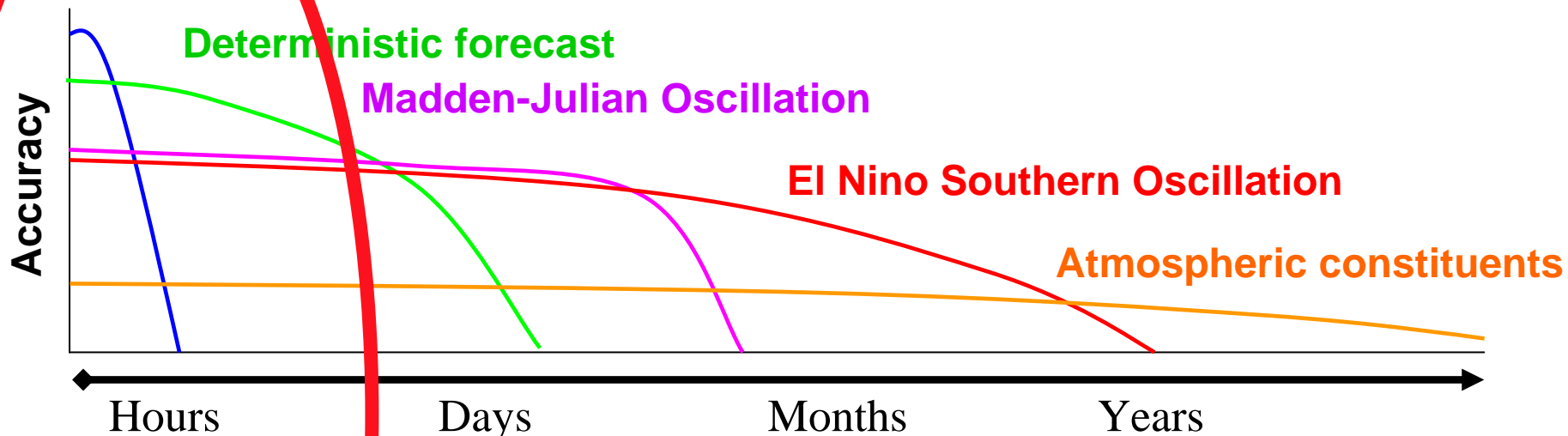


Seasonal Forecasts:

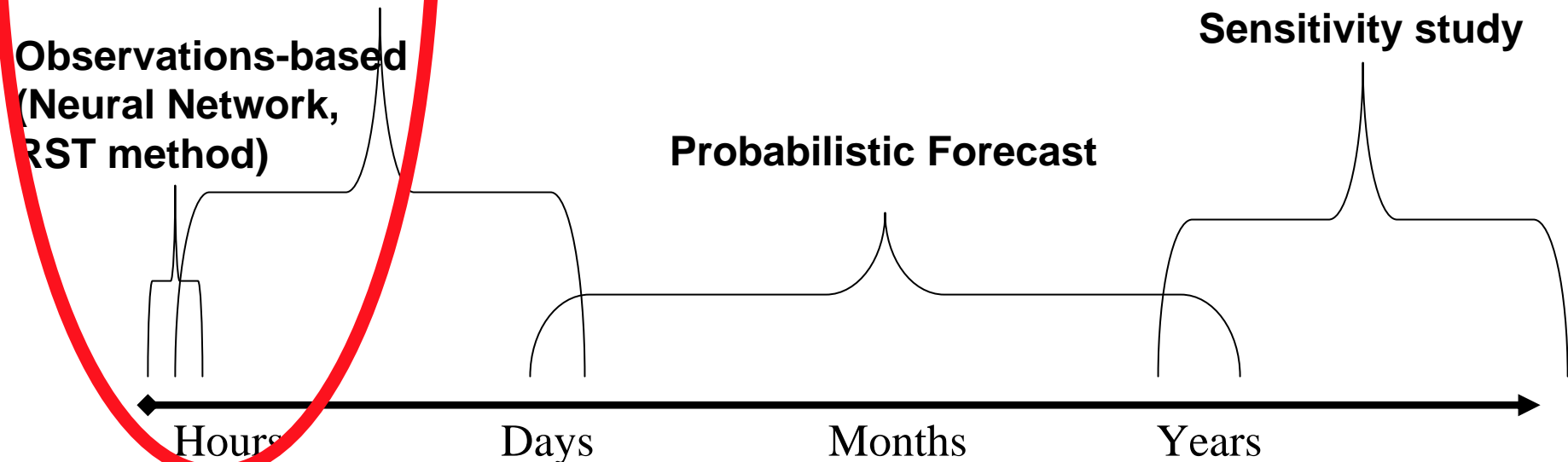


Atmospheric Predictability

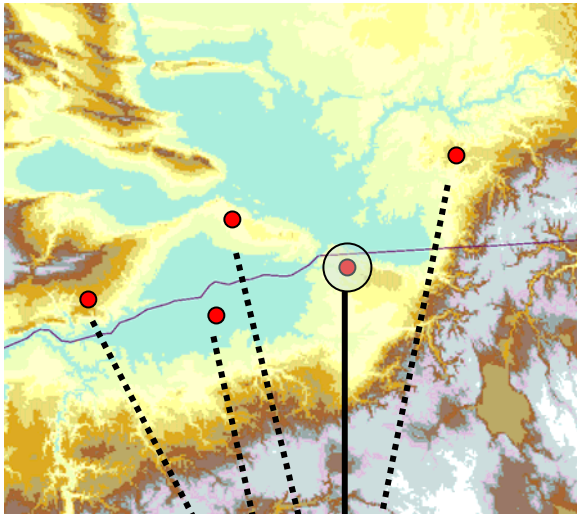
Observations



Mesoscale NWP model



Using Off-Site Observations



Space-time
Statistics

DEVELOPMENT OF NEXT-GENERATION WIND FORECASTING TECHNOLOGIES

Project Description: *Identify and develop statistical space-time methods and algorithms to improve short-term energy forecast accuracy at wind energy sites.*

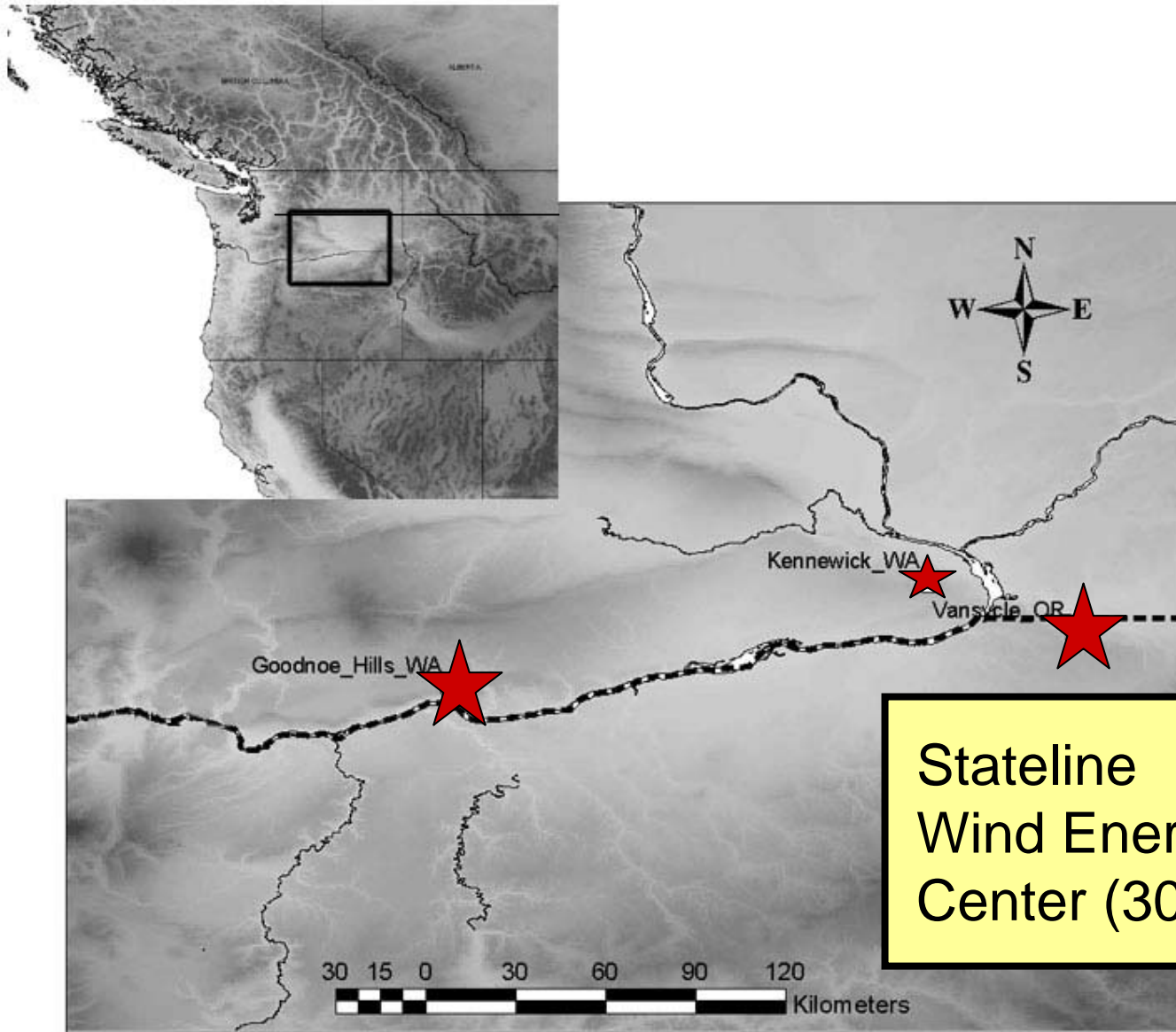
Washington Technology Center
PPM, Energy Inc.

University of Washington

STATISTICS

Regime-switching Space-Time (RST) Algorithm

- Model formulation is parsimonious, yet **takes account of all the salient features of wind speeds:** alternating atmospheric regimes, temporal and spatial autocorrelation, diurnal and seasonal non-stationarity, and conditional heteroscedasticity
- **Space-time algorithm:** use of geographically dispersed meteorological observations in the vicinity of the wind farm
- **Regime-switching:** identification of westerly and easterly regime
- **Fully probabilistic:** provides valid statements of forecast uncertainty



Stateline
Wind Energy
Center (300MW)

Forecast Comparison

- Data is from Mar 2003 to Nov 2003
- RST method based on training for the 45 previous days
- Percent improvement of the root mean square (RMS) error over persistence is displayed for each month

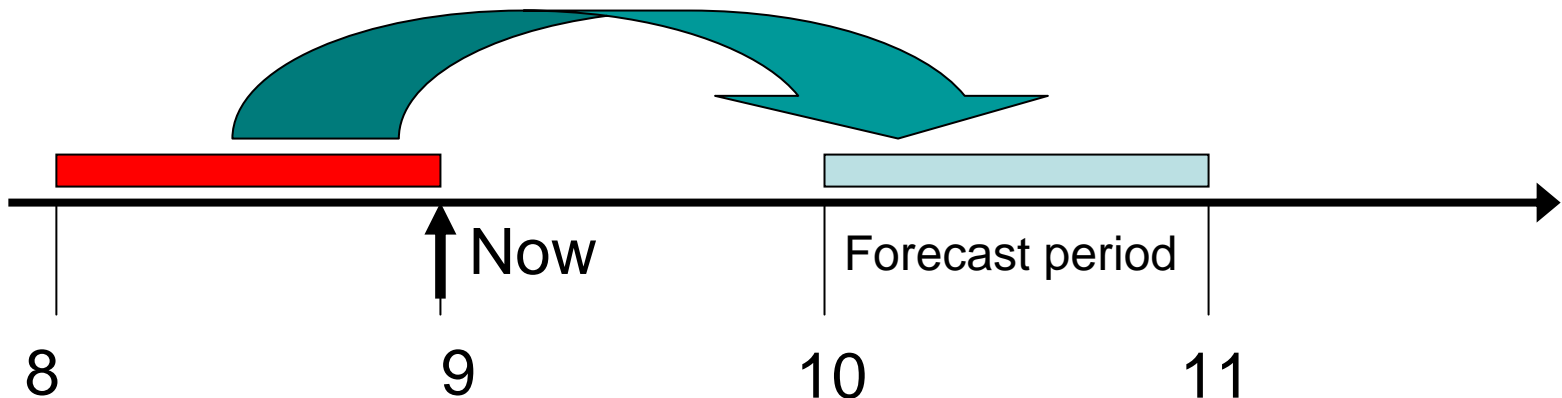
RST method compared to:

- persistence
- new reference (combination of persistence and climatology)
- time series (auto-regressive model)

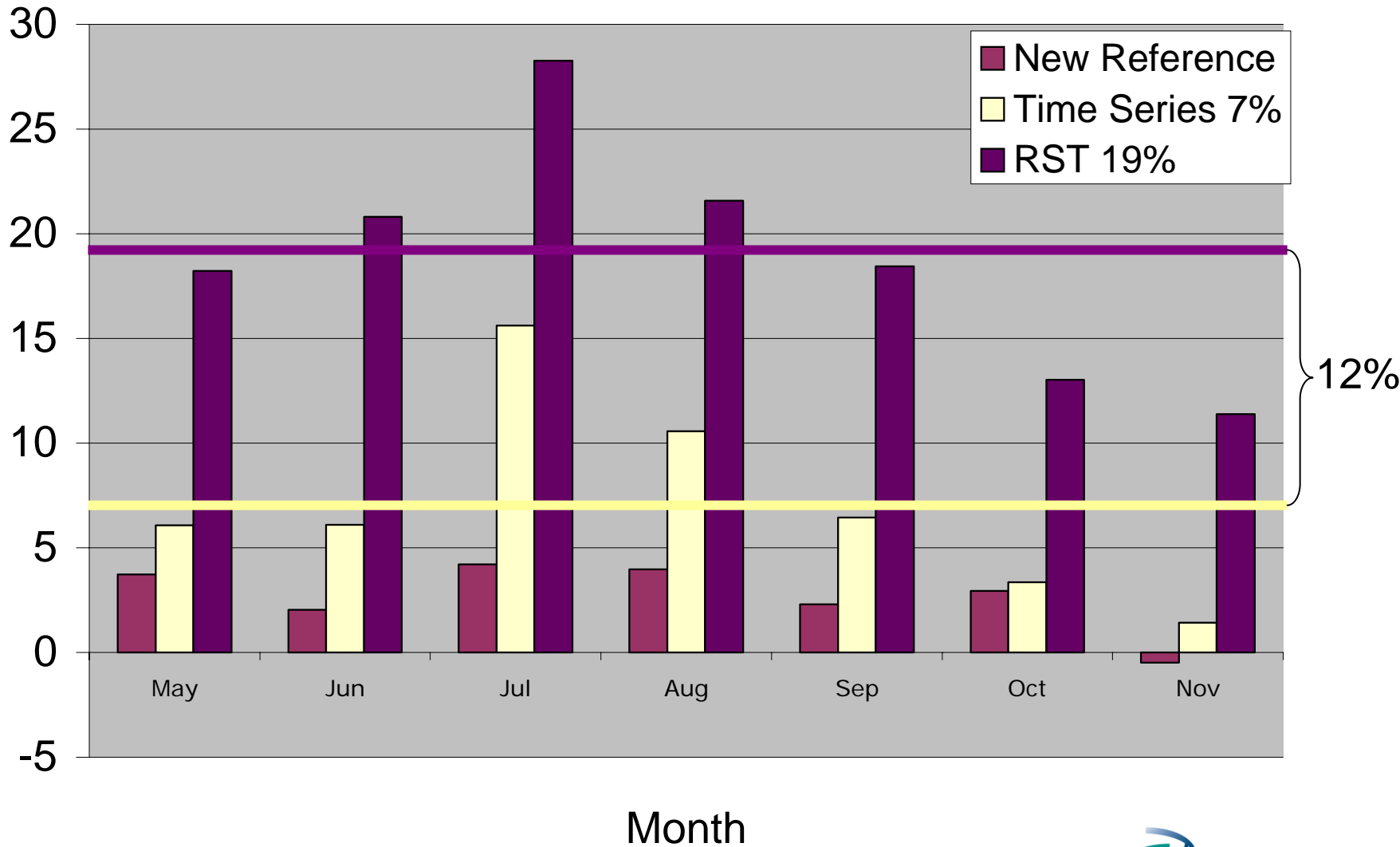
An Example Application of a Persistence Forecast

By 9:00 a.m. a decision-maker must make a prediction for the power to be produced between 10 a.m. and 11 a.m.

A **persistence** forecast uses the power produced between 8 a.m. and 9 a.m. (the last hourly power value) as the forecast. This is a benchmark for measuring improvement.



2 hour Forecast Improvement over Persistence at Vansycle

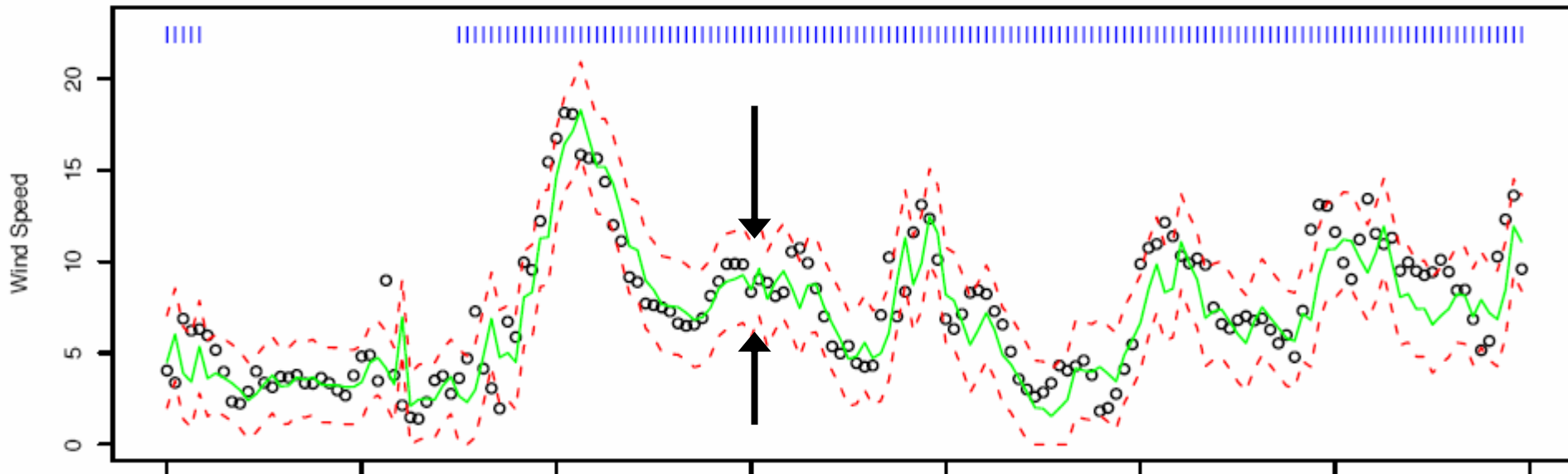


Example of RST method

Black circles = wind speed values
Green line = forecast

Red dotted line =
90% confidence interval
Blue bars = westerly regime

28 June □ 4 July 2003



90 % confidence intervals are on average 5 m/s wide

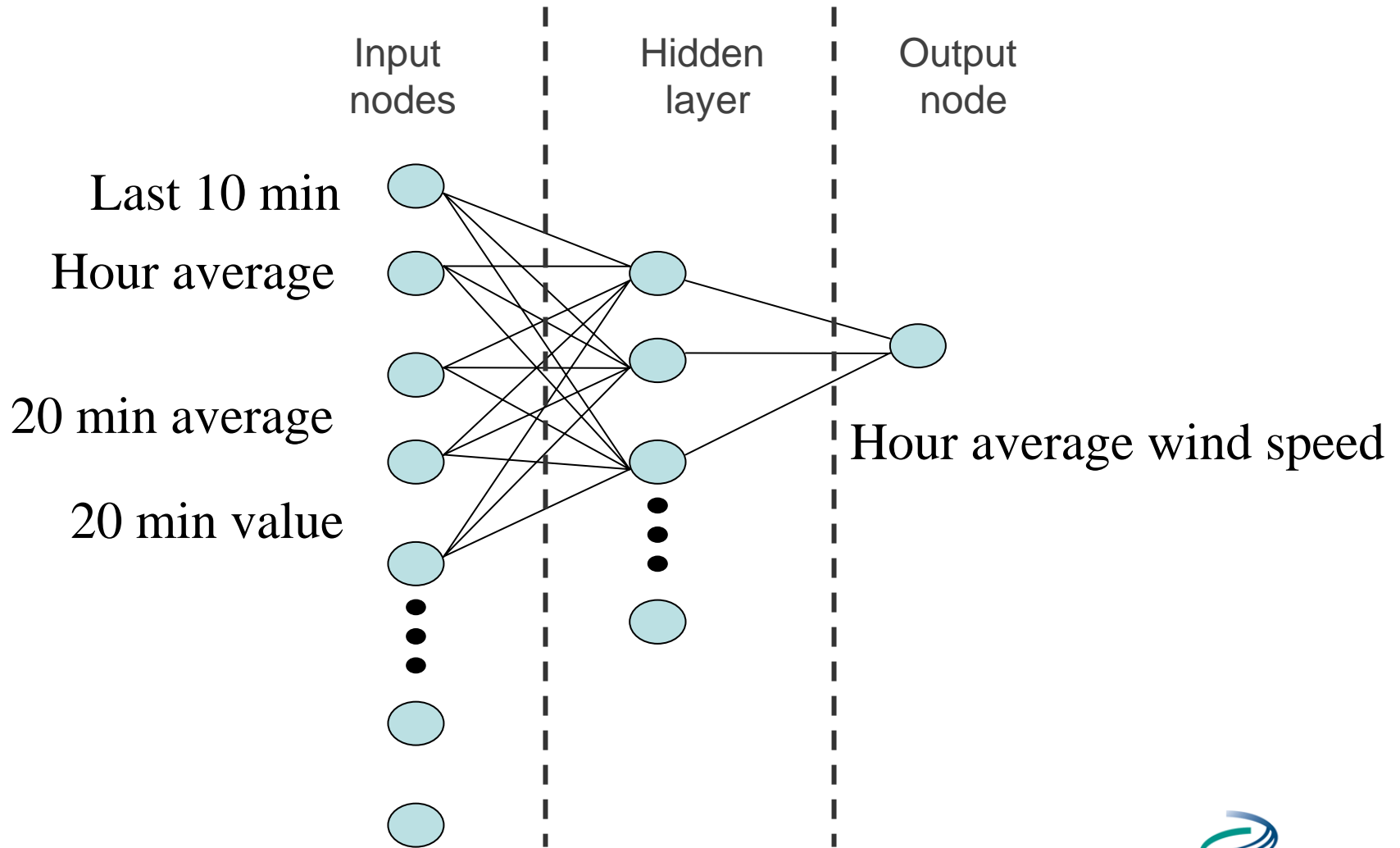
RST Summary

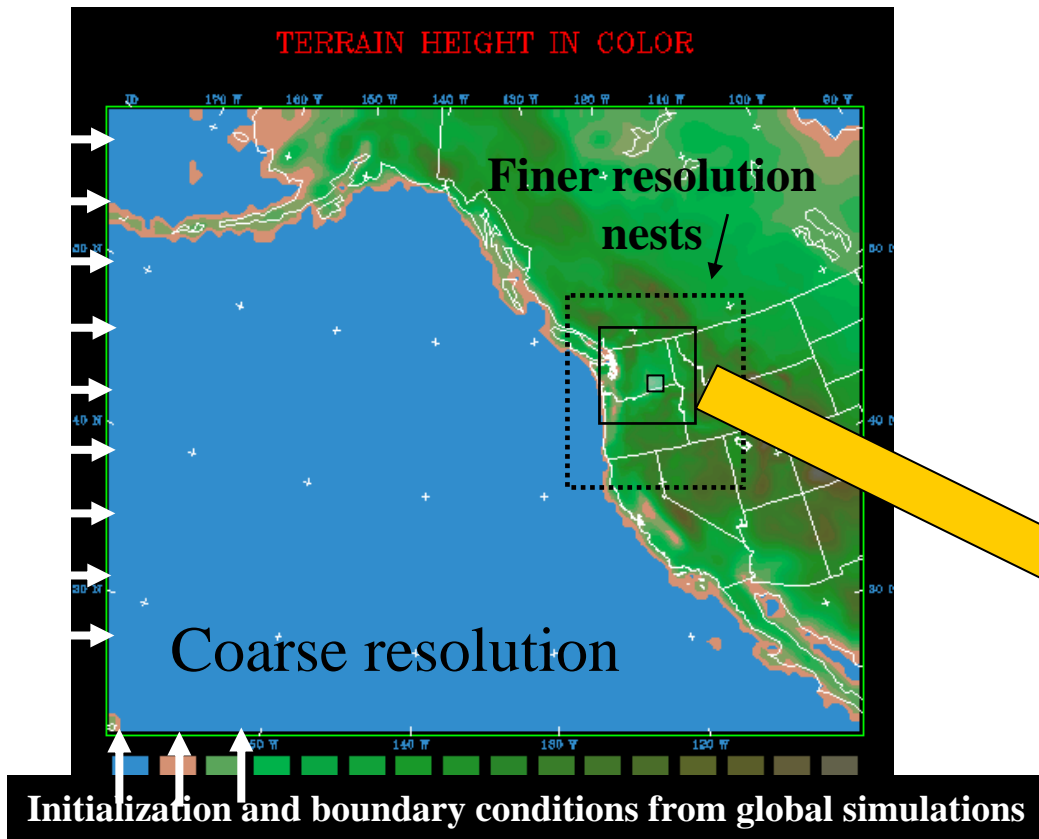
- Statistical forecast method that is based on **optimally-placed off-site observations**
- Meteorological **knowledge** is essential
- **Probabilistic** forecast method that effectively leverages the conditional heteroscedastic properties of wind speed

Short-range Forecasts using Off-site Observations and Numerical Weather Prediction (NWP)

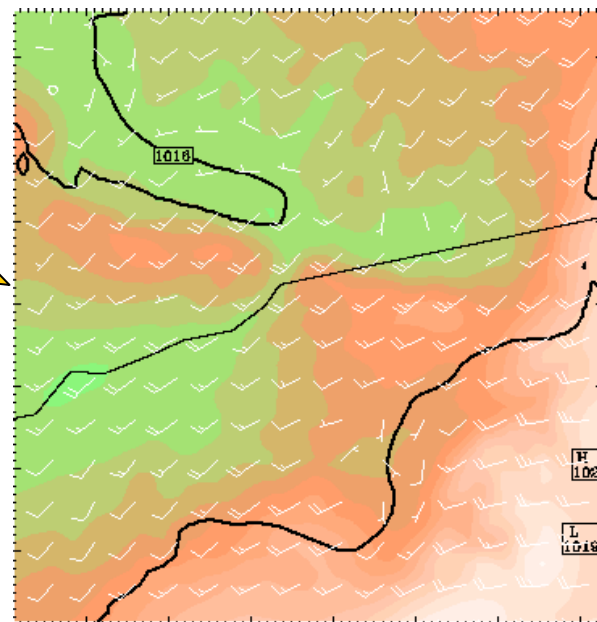
Forecasts use **Neural Networks**
and the challenge method

Neural Networks

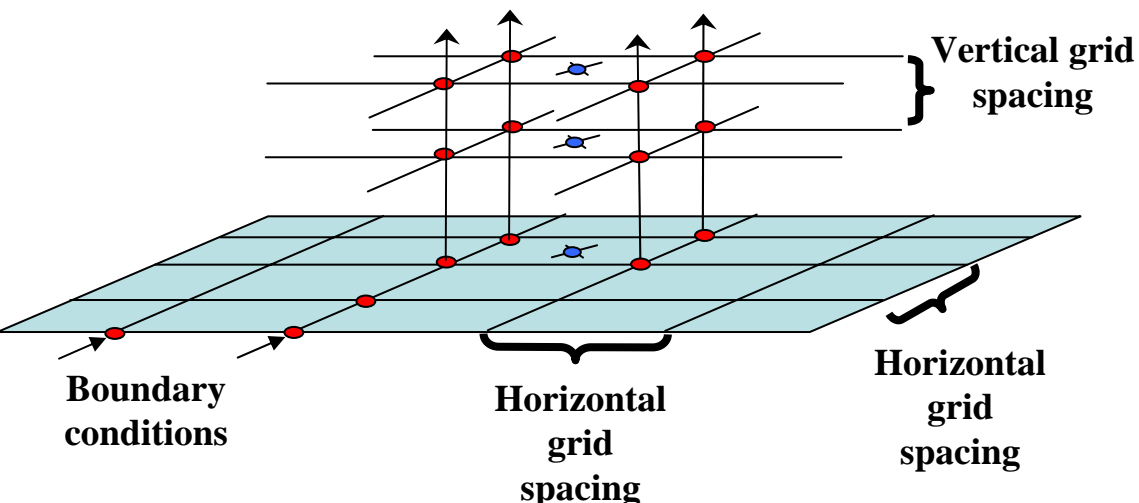




The general layout of a high resolution nested NWP model

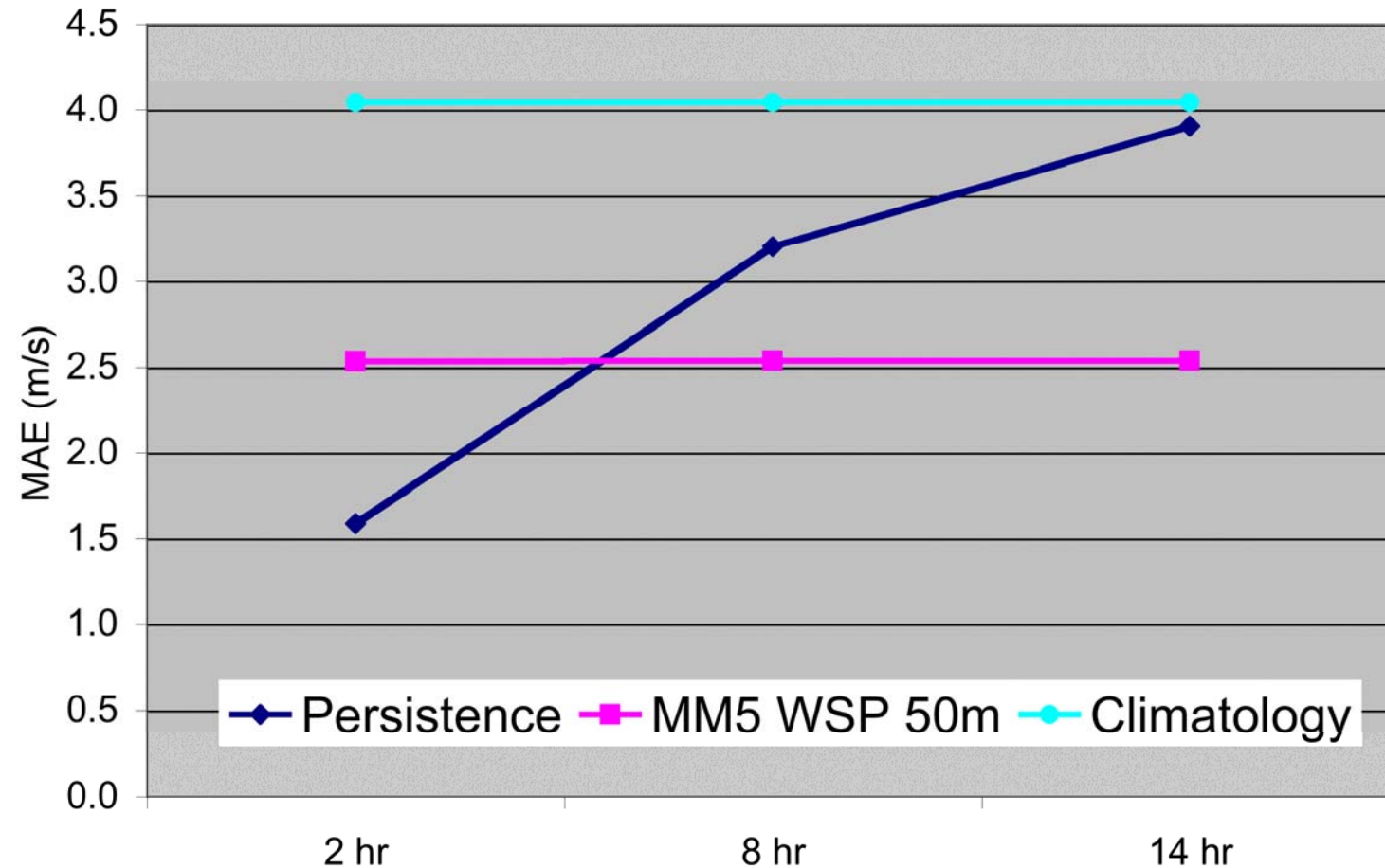


Detail of a 1.67 km 'nest' over Stateline



Skill of NWP forecasts for all months

Mean Absolute Error



NWP better for 8 and 14 hour forecasts

NWP has same error for 2 and 8 and 14 hours

Challenge Method

- Neural network inputs are added sequentially
- The amount/kind of inputs are chosen by the highest forecast accuracy
- Based on Kretzschmar, Eckert, and Cattani, 2004 JAM
 - First choose length of wind speed record (10 min. to 120 min.)
 - Then choose celestial variables to add (time-of-day, julian day, etc.)
 - Then choose meteorological variables (wind direction, temperature, pressure, combinations)
 - Etc.

Forecast Comparison

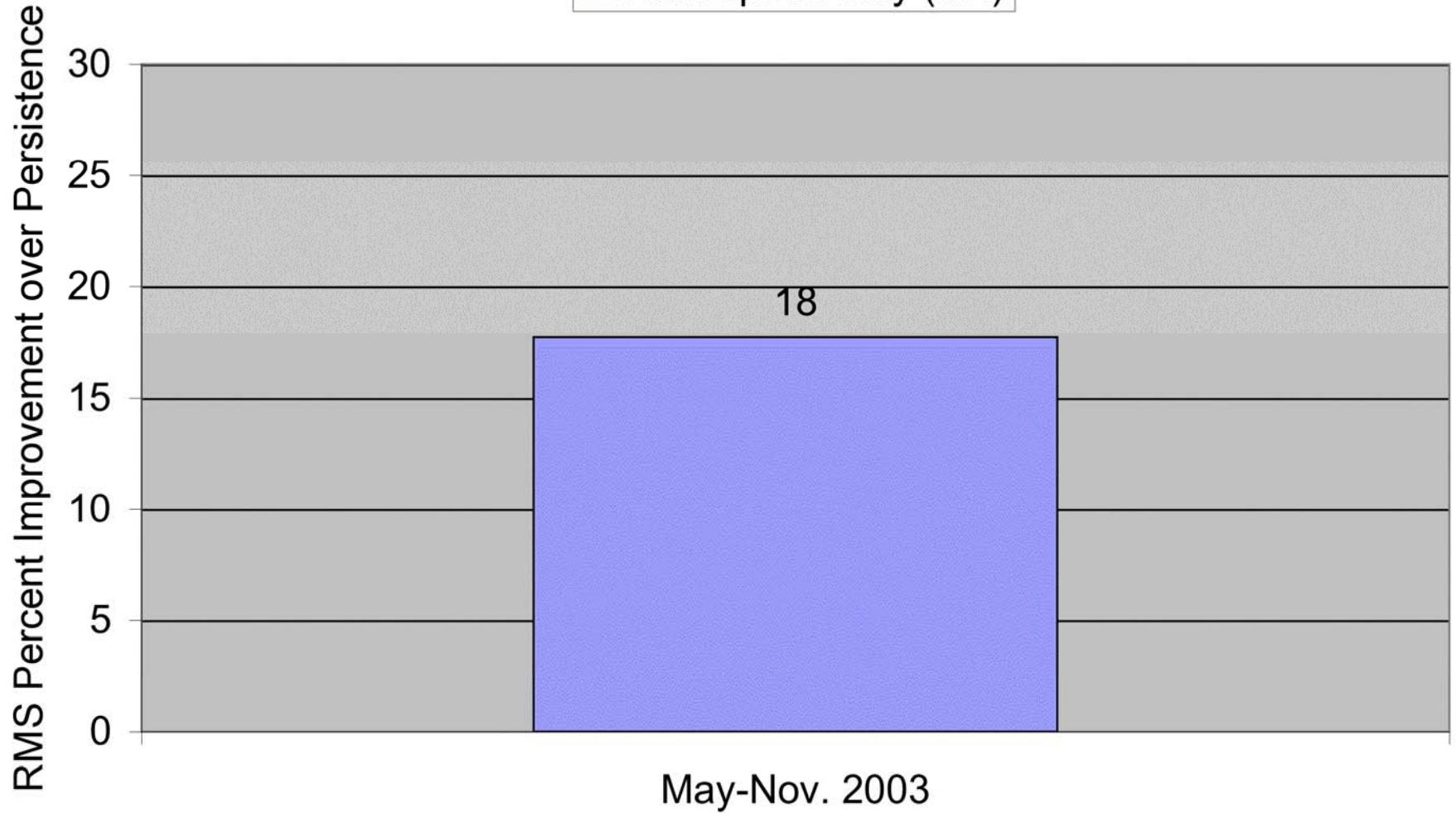
- Data is from Jan. 2003 to Apr. 2004
- Percent improvement of the root mean square (RMS) error over persistence is computed for each month, and the May 2003 to Nov 2003 values are averaged

Persistence compared to:

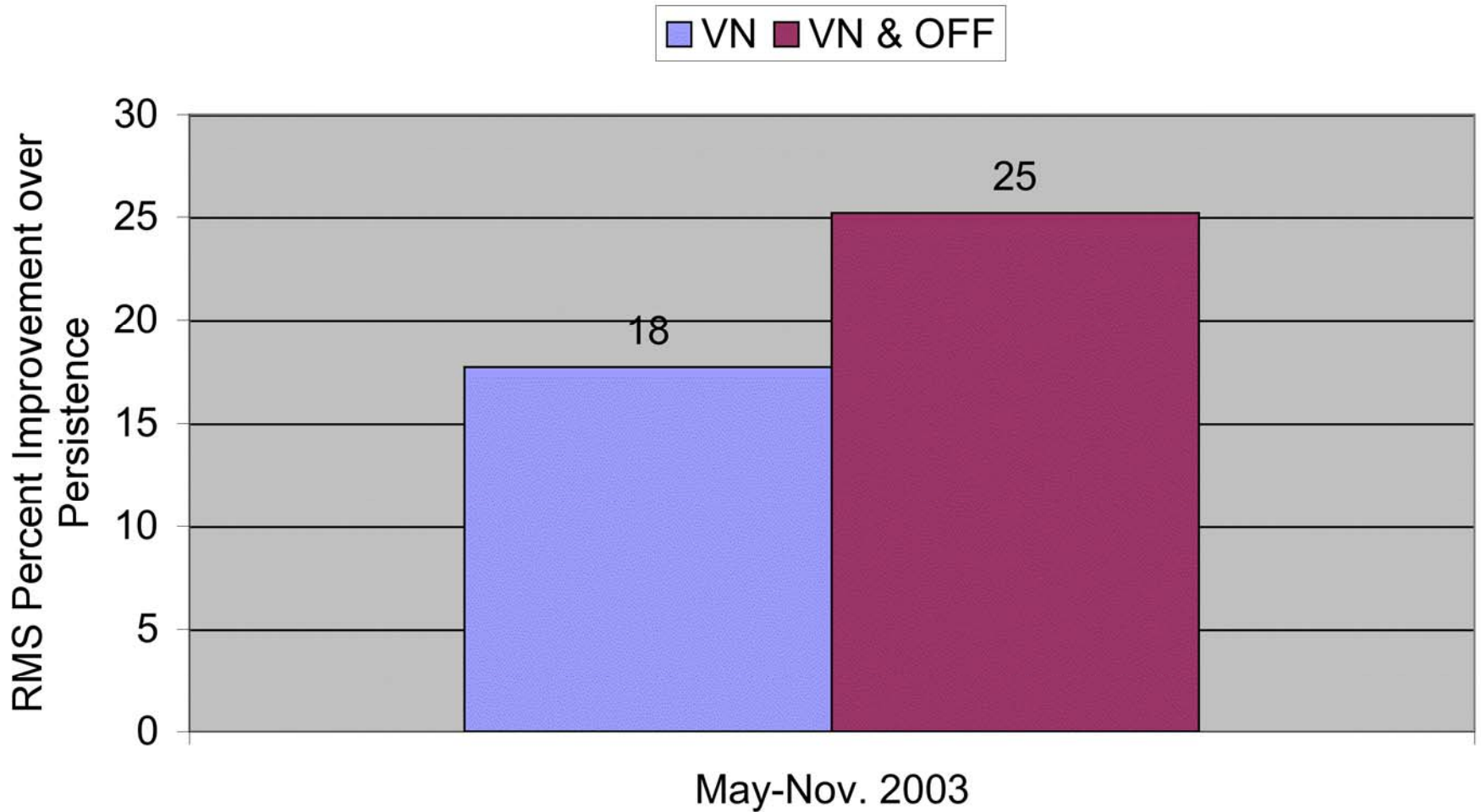
- NN forecast using **wind speed data** only
- NN forecast using **wind speed data** and **off-site observations**
- NN forecast using **wind speed data** and **NWP**
- NN forecast using **wind speed data**, **off-site observations** and **NWP**

Forecast Improvement over Persistence

Wind speed only (VN)

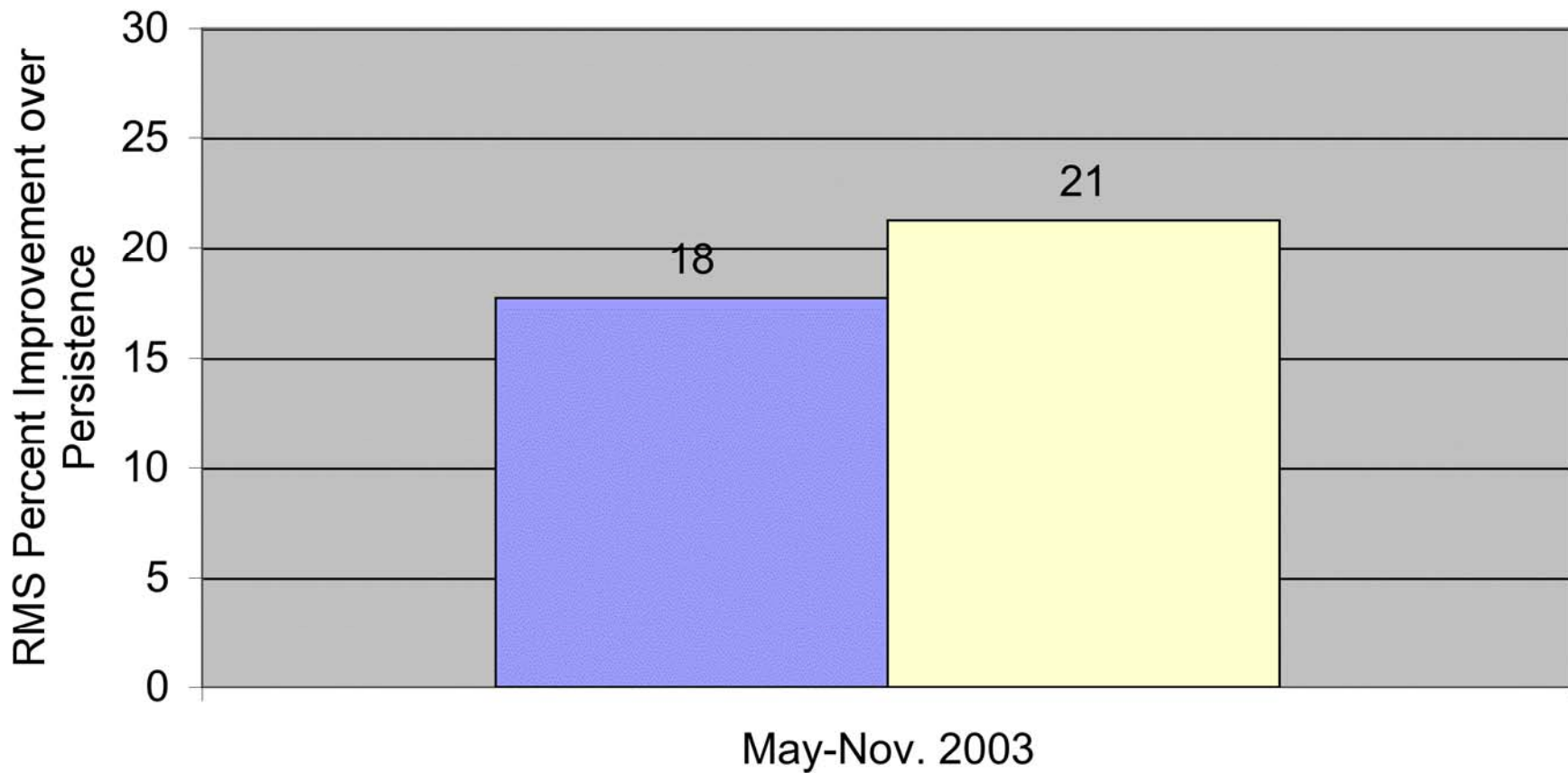


Forecast Improvement over Persistence using wind speed data and off-site observations

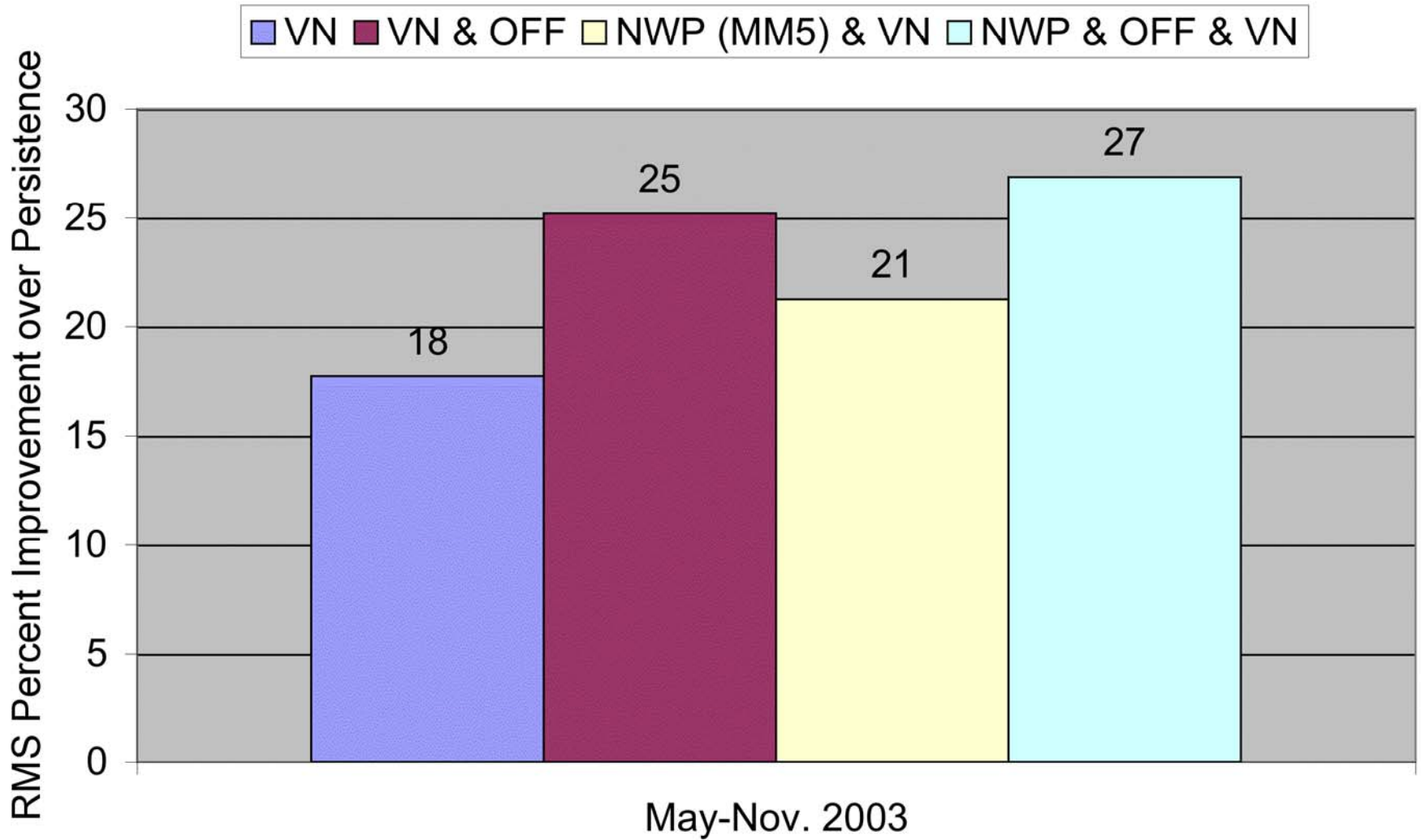


Forecast Improvement over Persistence using wind speed data and NWP

■ VN ■ NWP (MM5) & VN



Forecast Improvement over Persistence



NWP Challenge Method Summary

- Neural Networks with the challenge method provide an improvement over a persistence forecast
- Both off-site observations and NWP alone improve forecast
- Largest forecast improvements occur with both NWP and off-site observations

Conclusions

- Meteorological knowledge improves short term forecasts
 - Regime Switching Space-Time Method
 - NWP- Challenge Method
- Significant improvements over persistence are possible
- The 90% confidence intervals are 5 m/s wide, in this example



www.3tiergroup.com

Thank you for listening