

# Evaluation of AMPS Forecasts for Varied Synoptic Regimes

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# What are SOMs?

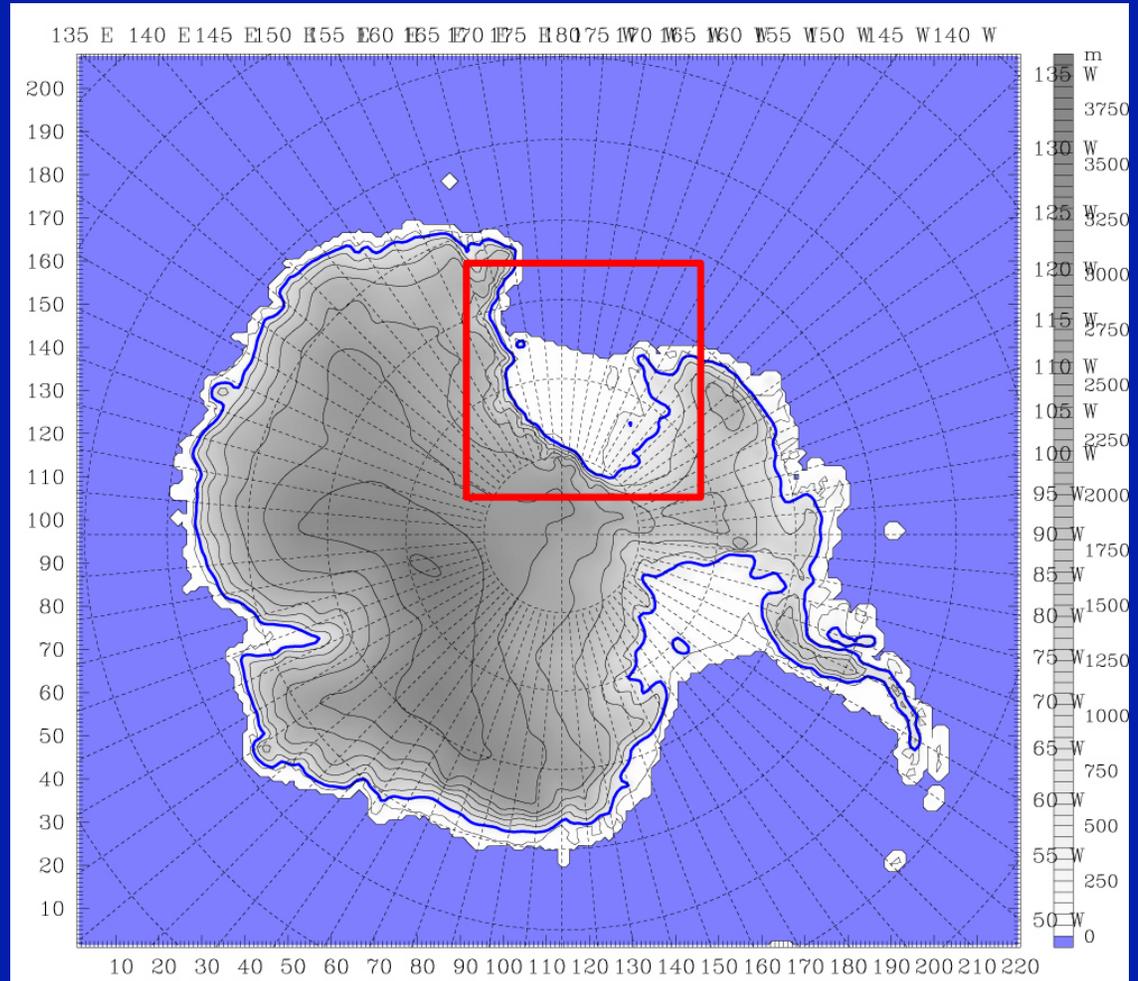
- SOM - Self-Organizing Map
- SOM technique uses an unsupervised learning algorithm
- Clusters data into a user selected number of nodes
- SOM algorithm defines nodes that are representative of the data in the training set
- SOMs are in use across a wide range of disciplines
  - Climate applications of SOMs
    - Hewitson and Crane (2002) *Climate Research*
    - Cassano et al. (2006) *Climate Dynamics*
    - Cassano et al. (2006) *International Journal of Climatology*
    - Lynch et al. (2006) *International Journal of Climatology*

# Application of SOM Analysis to AMPS Data

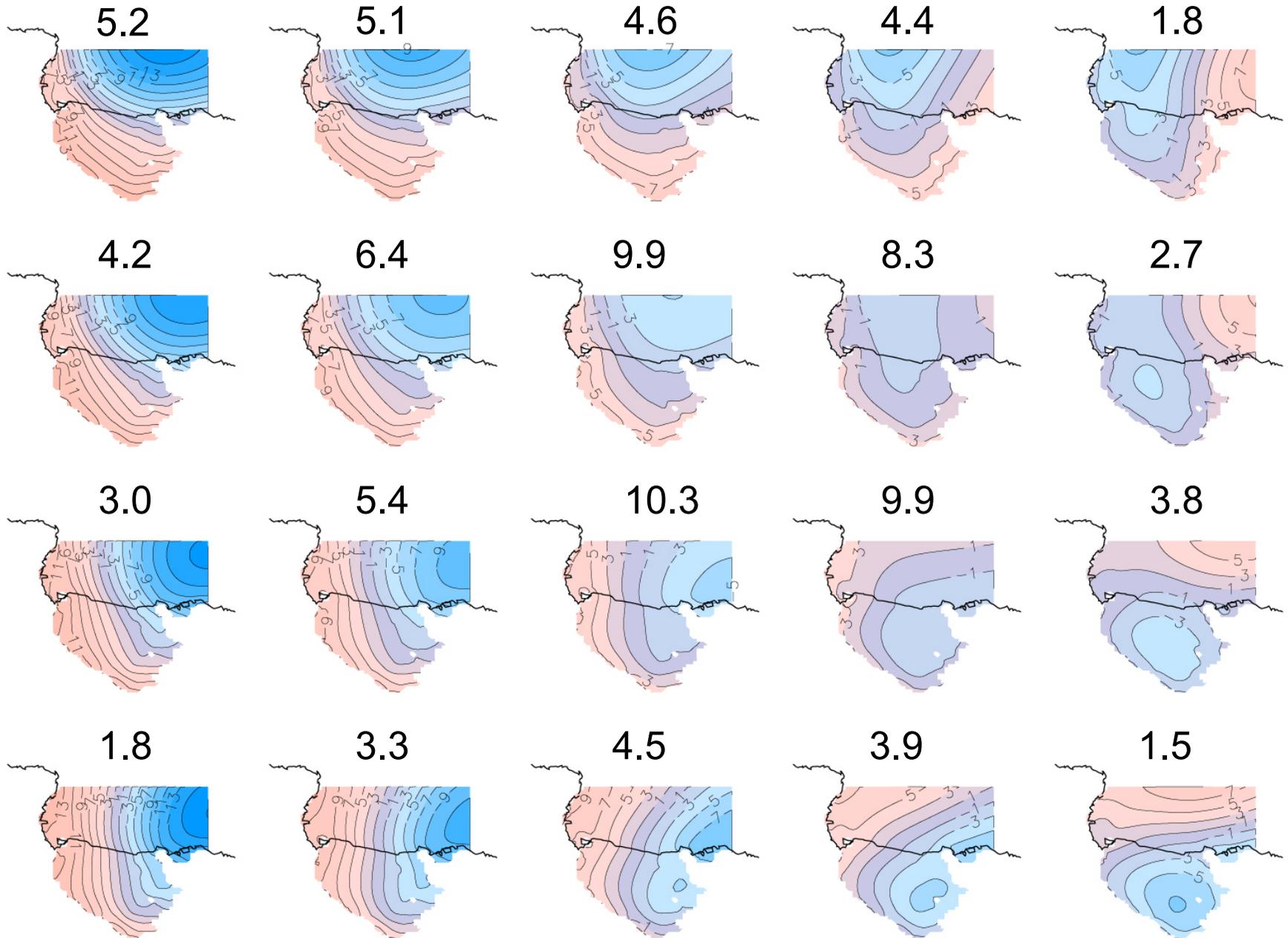
- Train SOM with AMPS SLP data
  - Result is a synoptic pattern classification
- Calculate frequency of occurrence of synoptic patterns
  - Annual and seasonal
  - As a function of forecast duration (0, 12, 24, 36, 60h forecasts)
- Misprediction of AMPS synoptic patterns
- Model validation statistics for specific synoptic patterns

# AMPS Data for SOM Analysis

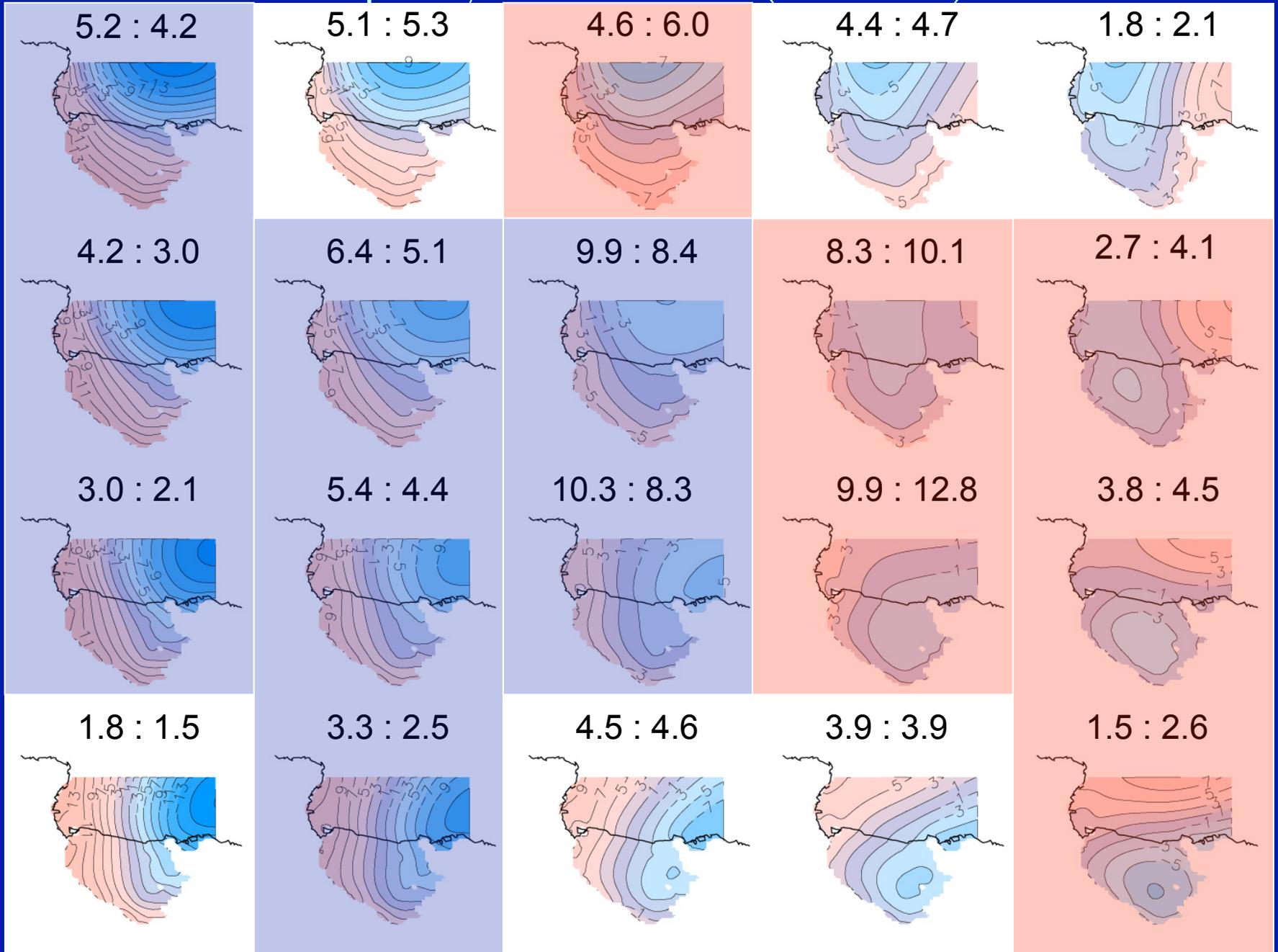
- SLP over Ross Sea sector of AMPS 30 km model domain
- AMPS MM5 simulations from Nov 2001 through Dec 2005
  - 9823 forecast times
- Evaluate forecasts at 12h intervals
  - 000: 0, 3, 6, 9 h
  - 012: 12, 15, 18, 21 h
  - 024: 24, 27, 30, 33 h
  - 036: 36, 39, 42, 45 h
  - 048: 48, 51, 54, 57 h
  - 060: 60, 63, 66, 69 h



# Synoptic Pattern Classification: 000



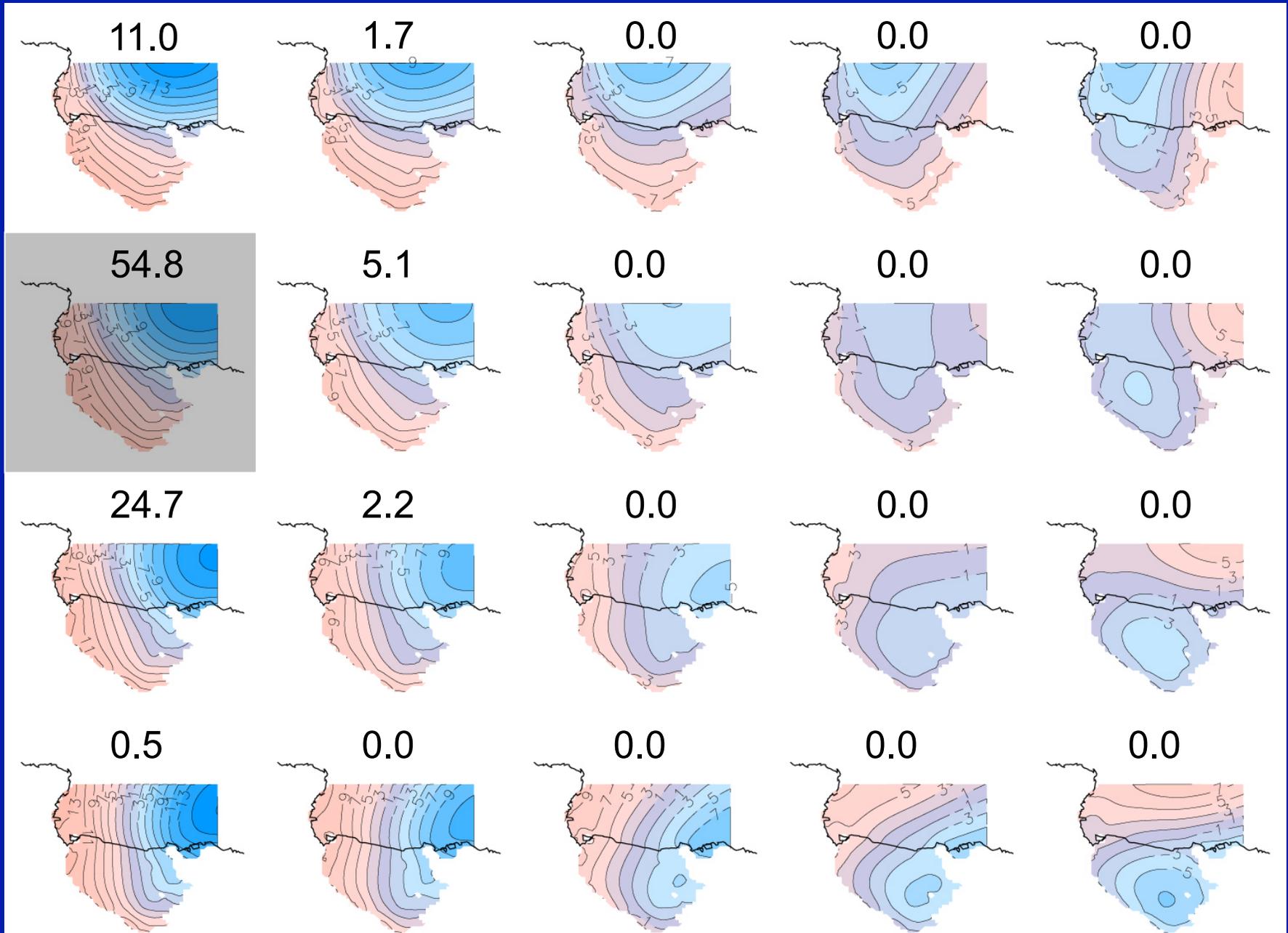
# Frequency of Occurrence (000 : 060)



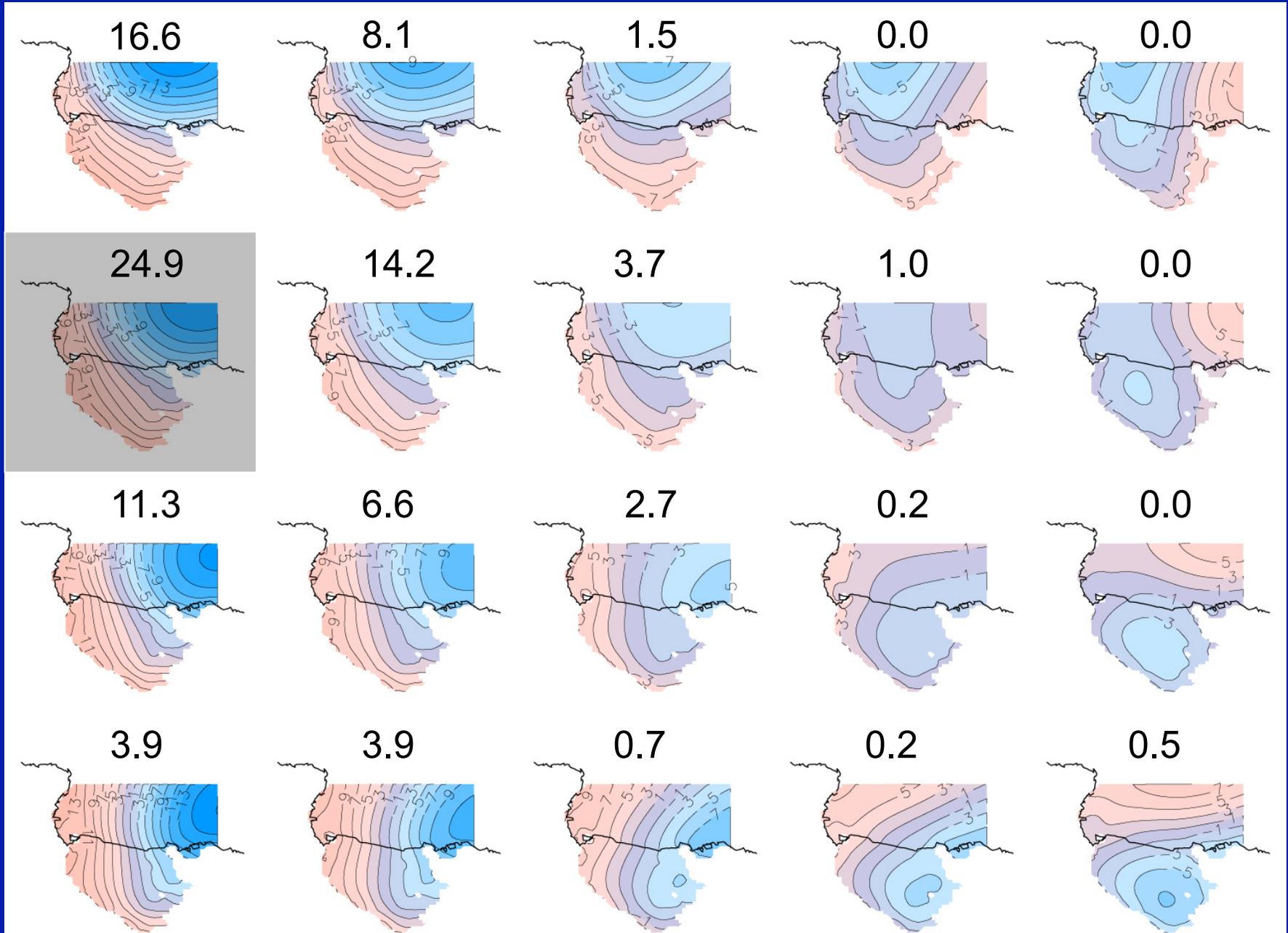
# Misprediction of Synoptic Patterns

- Consider all of the time periods for which the model 000 h forecasts map to a particular node
  - For these time periods determine which nodes the longer duration model forecasts map to
- Calculate:
  - Percent of cases that map to the correct node
  - Mis-mapping of model predictions between nodes

# AMPS 012 Forecasts: Node (1,2)



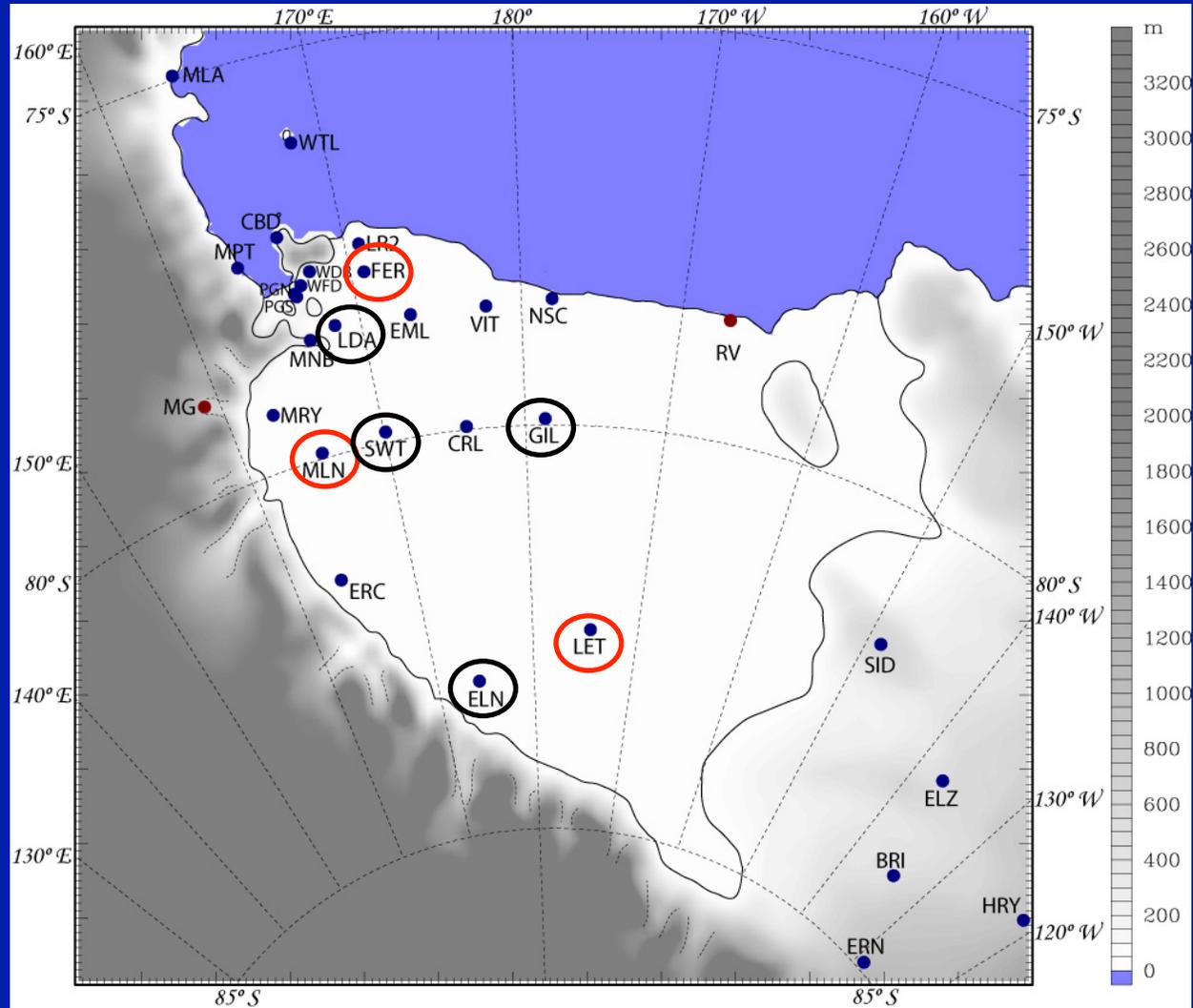
# AMPS 060 Forecasts: Node (1,2)

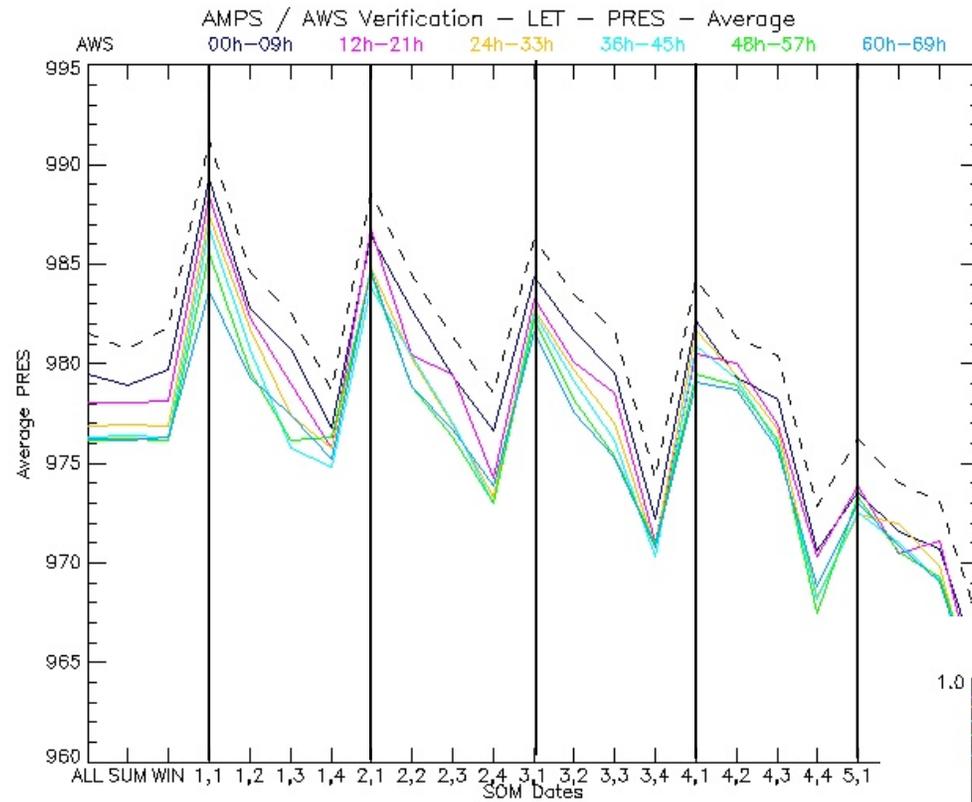


# Model Errors for Synoptic Patterns

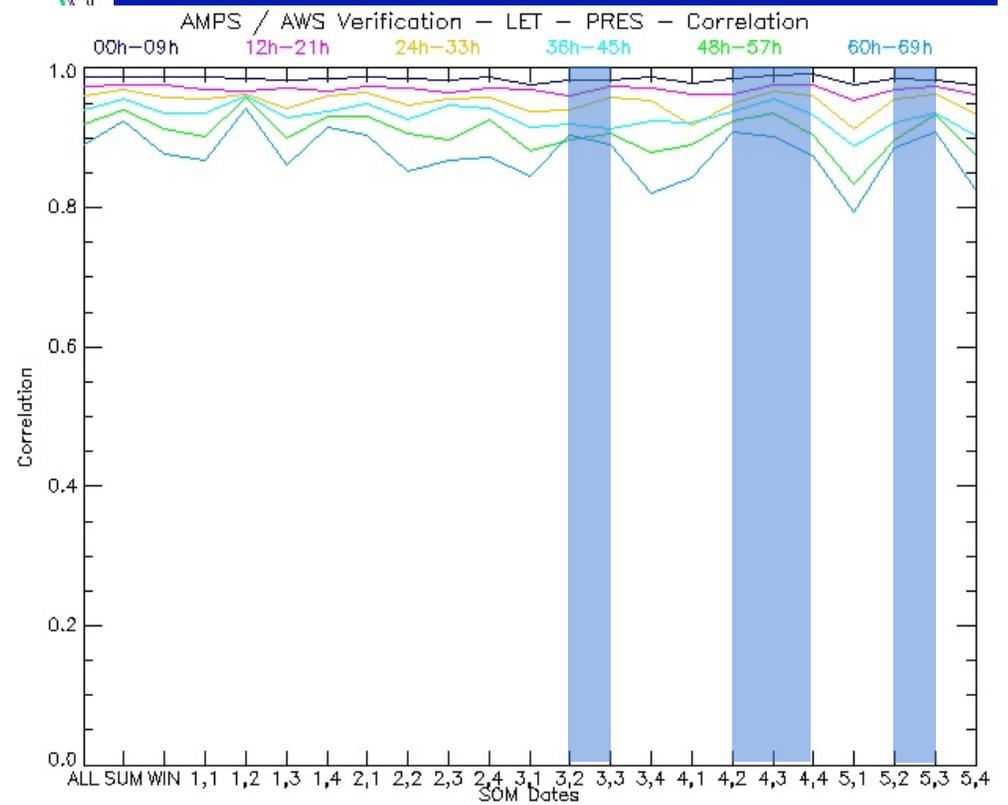
- Determine how observations (or model state) varies as a function of SOM identified nodes
- Compare model predictions to AWS observations
- Calculate model validation statistics for all time periods that map to each node
- Look for model errors that vary from node to node

# AWS Sites Used for SOM Analysis

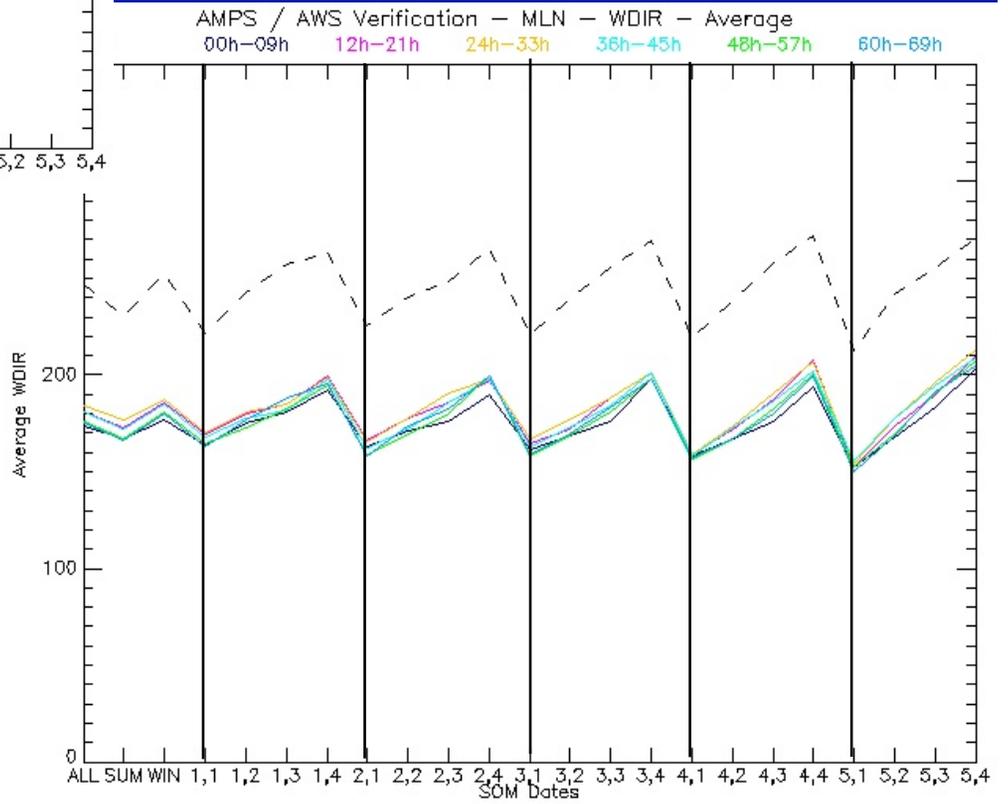
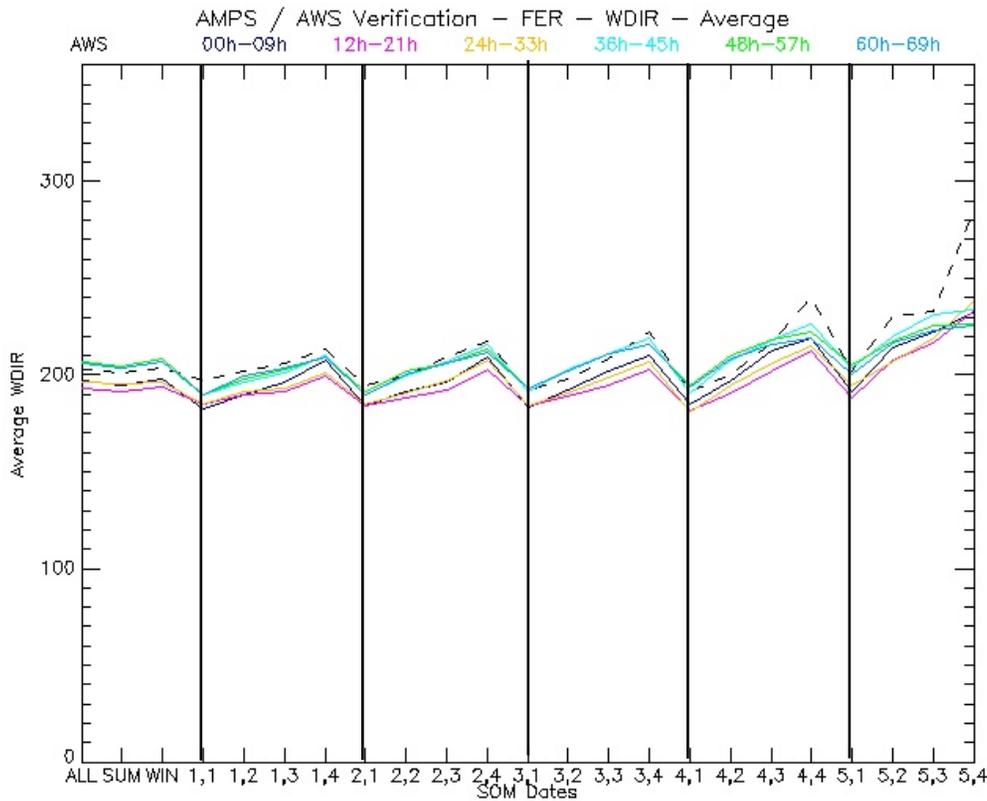




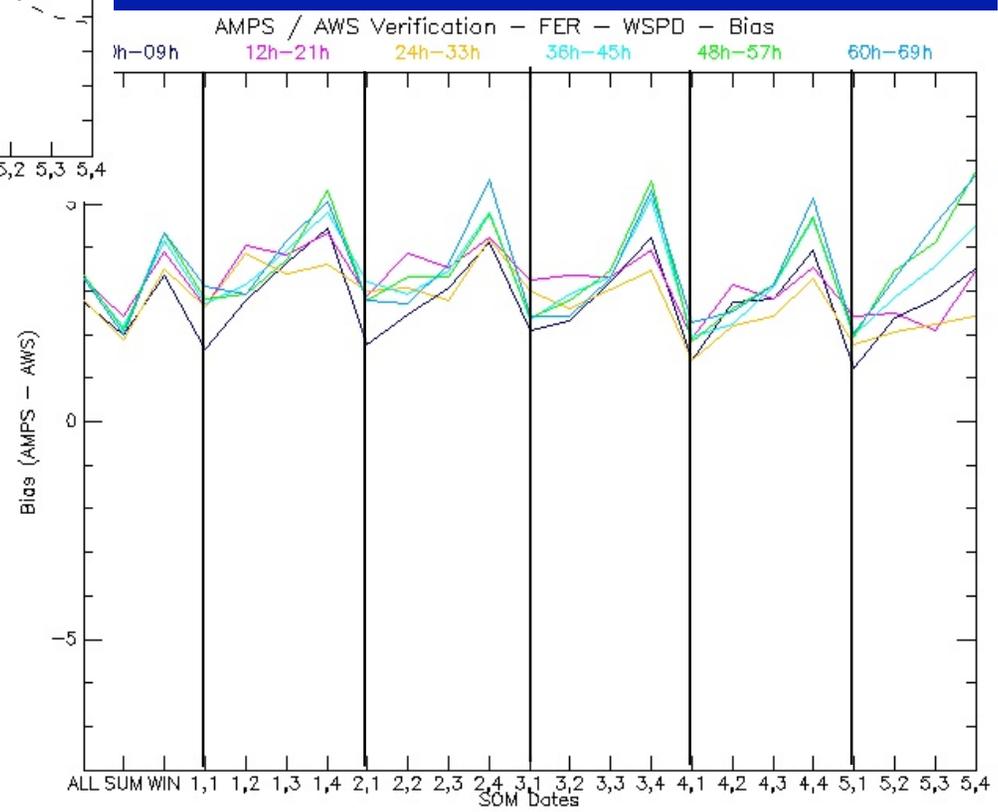
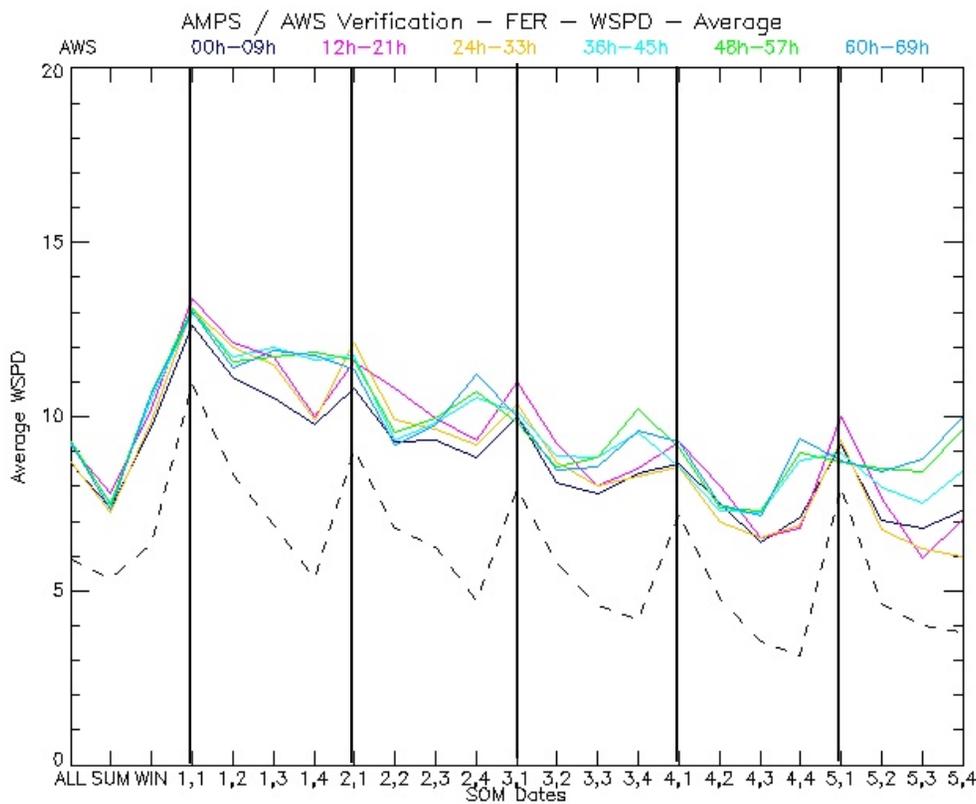
# Lettau: Pressure



# Ferrell and Marilyn: Wind Direction

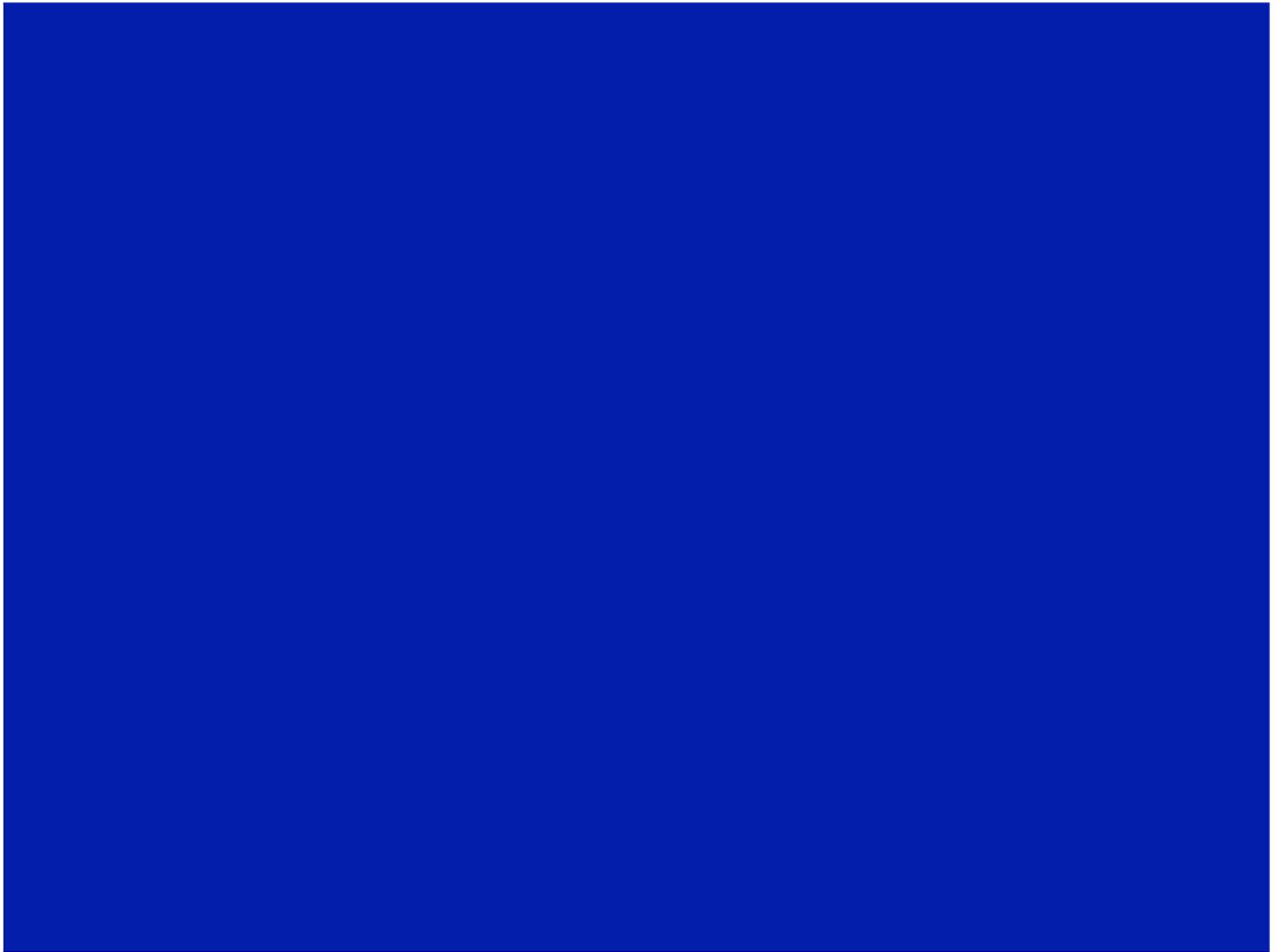


# Ferrell: Wind Speed



# Conclusions / Future Work

- The use of SOMs provides an alternate method of evaluating model performance
  - Identify synoptic patterns which are over or underpredicted
  - Determine model tendency for misprediction of certain synoptic types
  - Provide information on model errors related to specific synoptic patterns
- Manuscript for *Weather and Forecasting*
- Attribution of model errors to circulation and non-circulation related components
  - Ex: model precipitation



# Outline

- What are SOMs?
- Application of SOMs for model evaluation studies
- Application of SOM Analysis to AMPS data
- Conclusions / Future Work

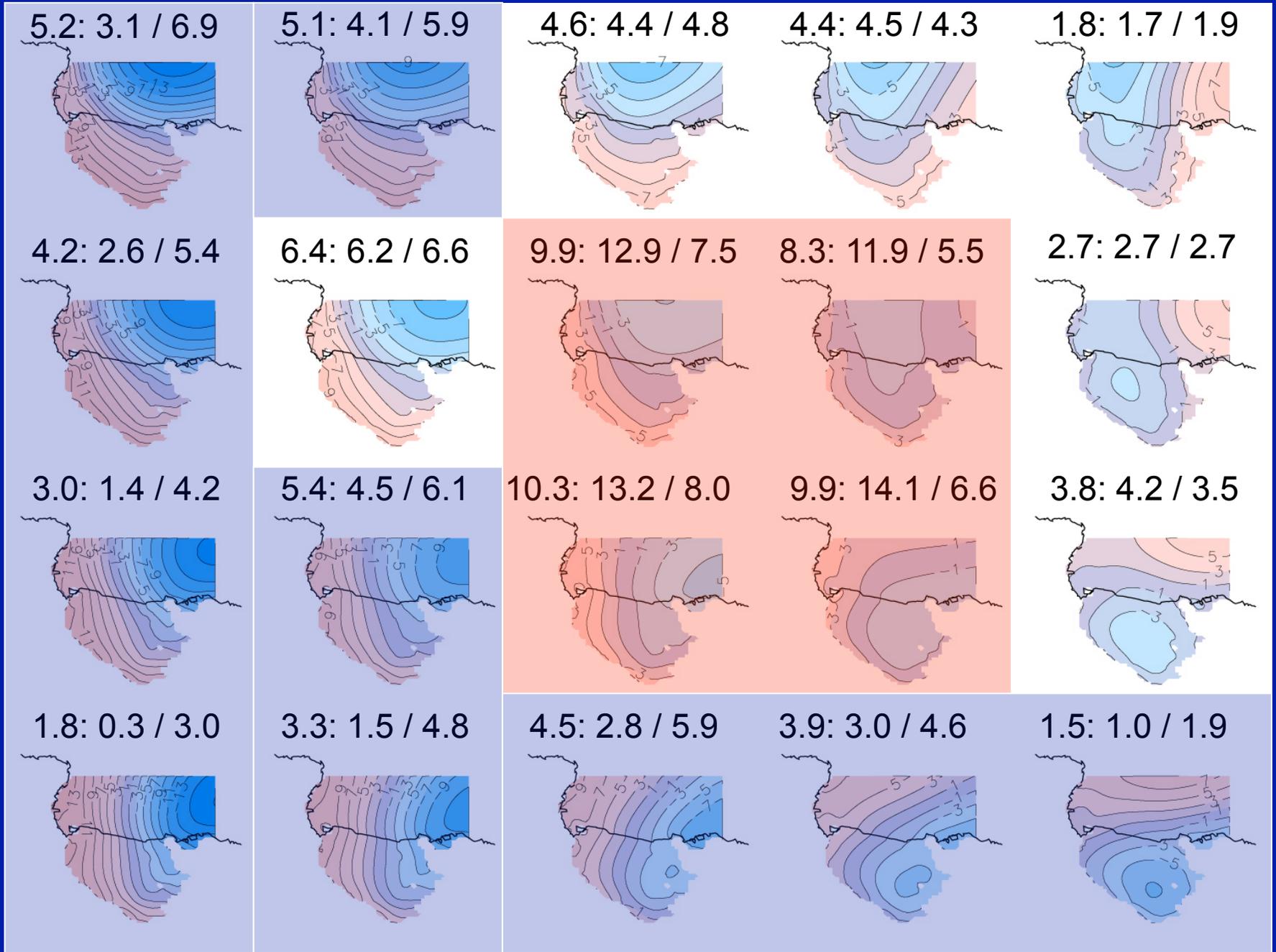
# “Typical” Model Evaluation Strategy

- Compare modeled and observed fields directly
  - Time series of observed and modeled variables
  - Model validation statistics (bias, RMSE, correlation, etc.)
- Case Study Evaluations
- Compare model data with observational analyses
  - Ex. Difference of monthly or seasonal mean sea-level pressure

# “Typical” Model Evaluation Strategy

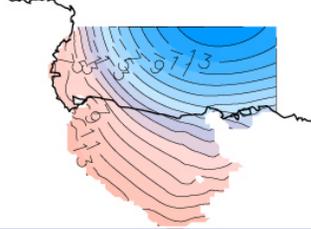
- Advantages
  - Simple techniques with easy interpretation
  - Highlights differences between models and analyses and also inter-model differences
- Disadvantages
  - Neglects differences in synoptic events
    - These events are the items of interest for operational weather forecasting applications
    - Similar seasonal mean SLP may mask differences in simulated synoptic climatology
  - Can be difficult to gain physical insight into the source of model errors

# Seasonal Frequency of Occurrence (Annual: Summer / Winter)

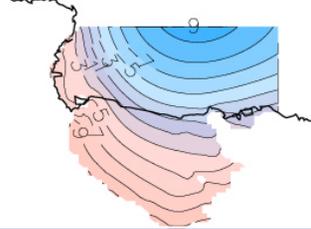


# Frequency of Occurrence (000 : 012)

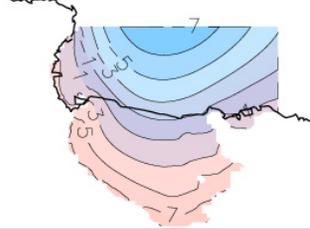
5.2 : 5.3



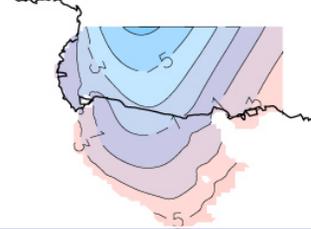
5.1 : 4.6



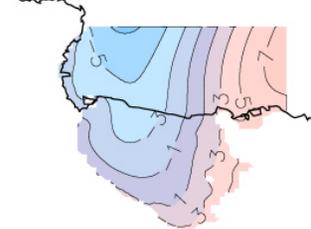
4.6 : 4.8



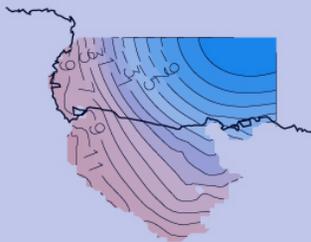
4.4 : 4.0



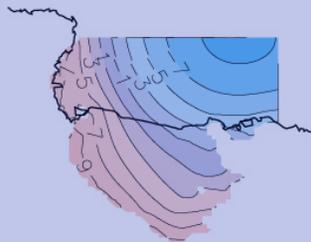
1.8 : 1.8



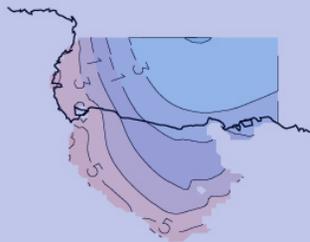
4.2 : 3.6



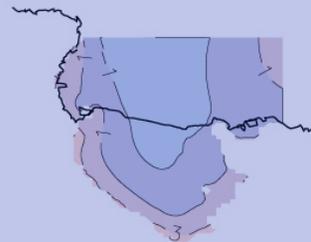
6.4 : 5.6



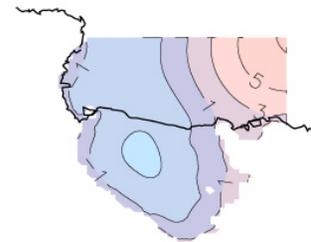
9.9 : 8.5



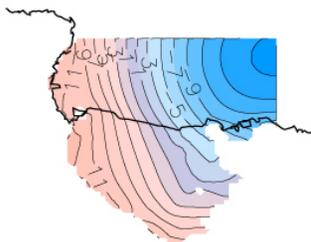
8.3 : 7.4



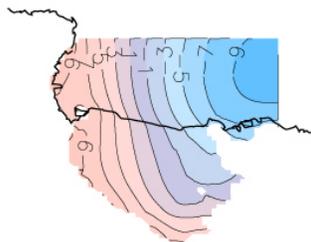
2.7 : 2.7



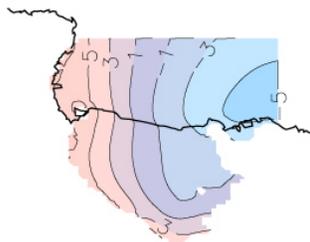
3.0 : 3.2



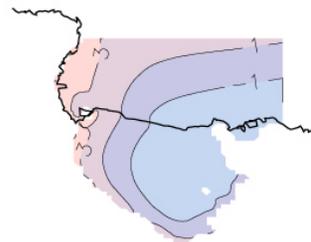
5.4 : 5.1



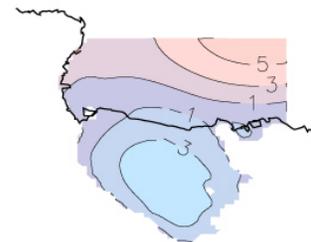
10.3 : 10.5



9.9 : 10.4



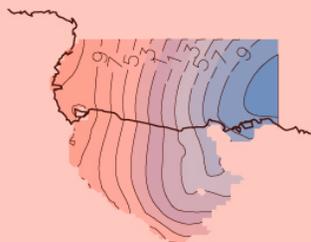
3.8 : 4.0



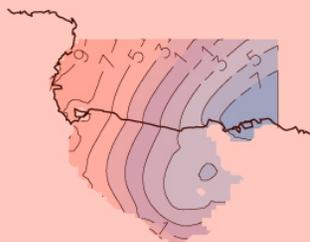
1.8 : 2.4



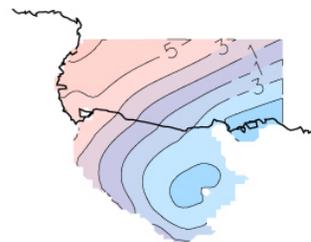
3.3 : 4.0



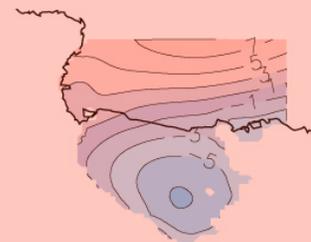
4.5 : 5.7



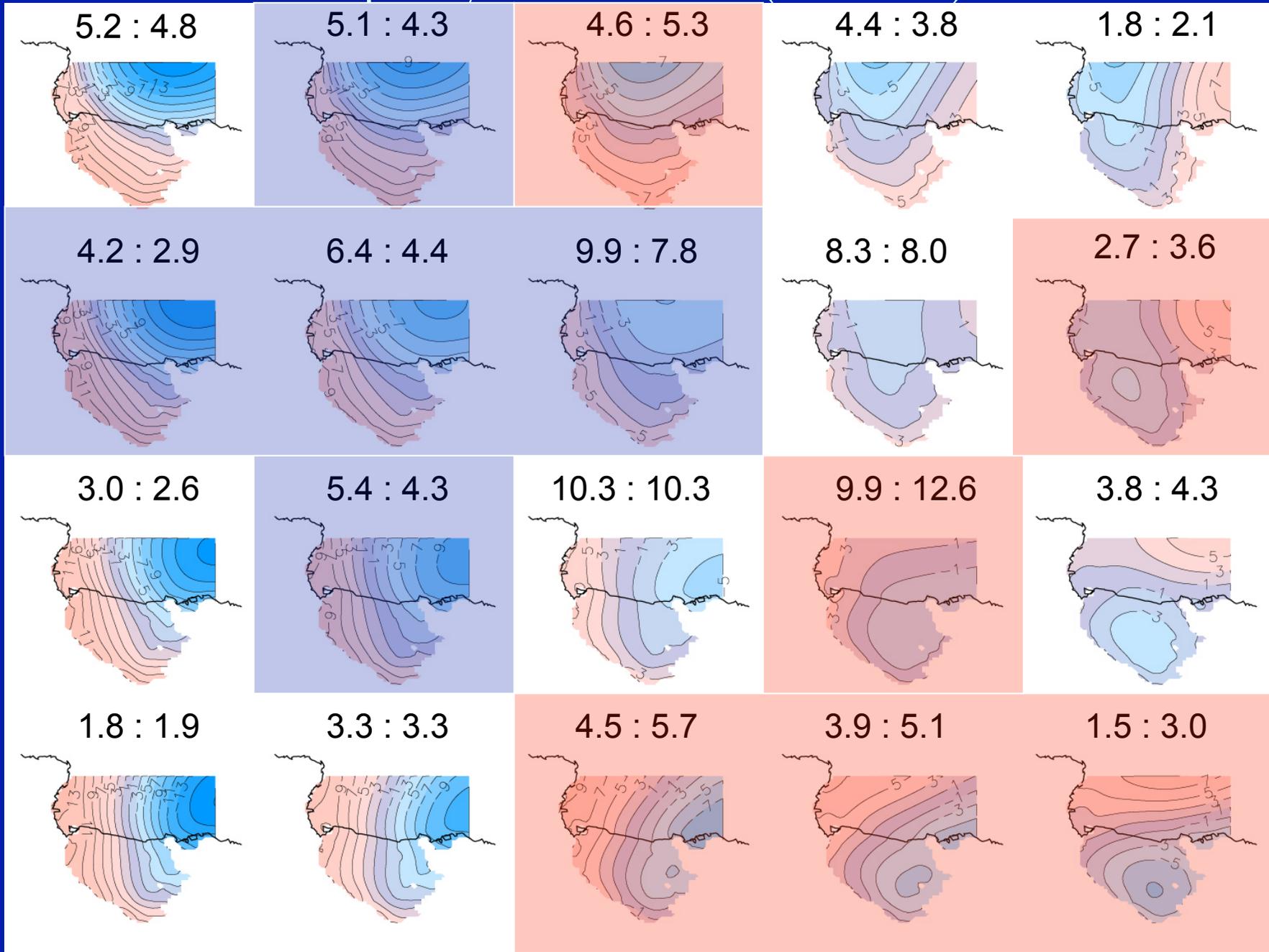
3.9 : 4.3



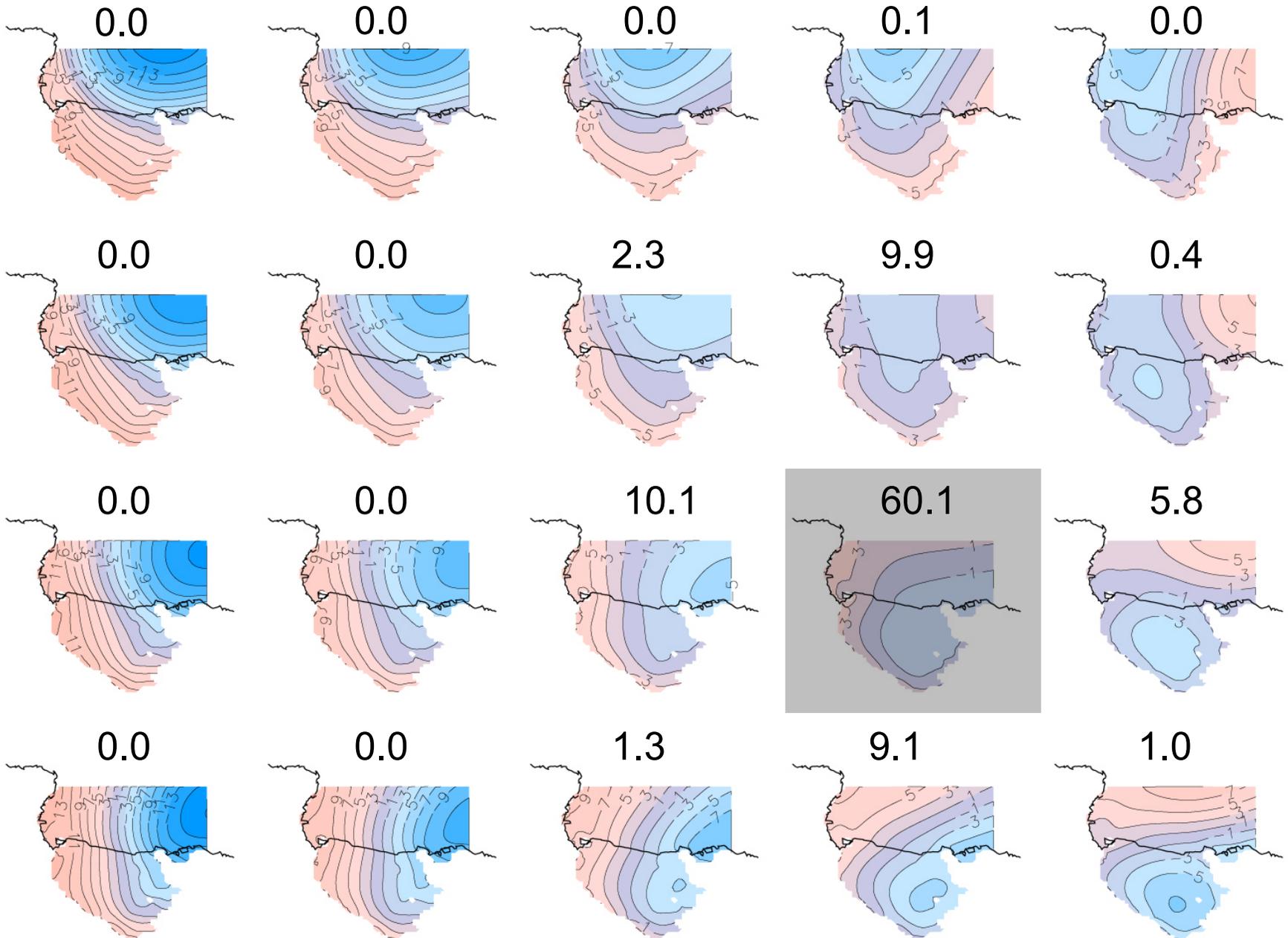
1.5 : 2.1



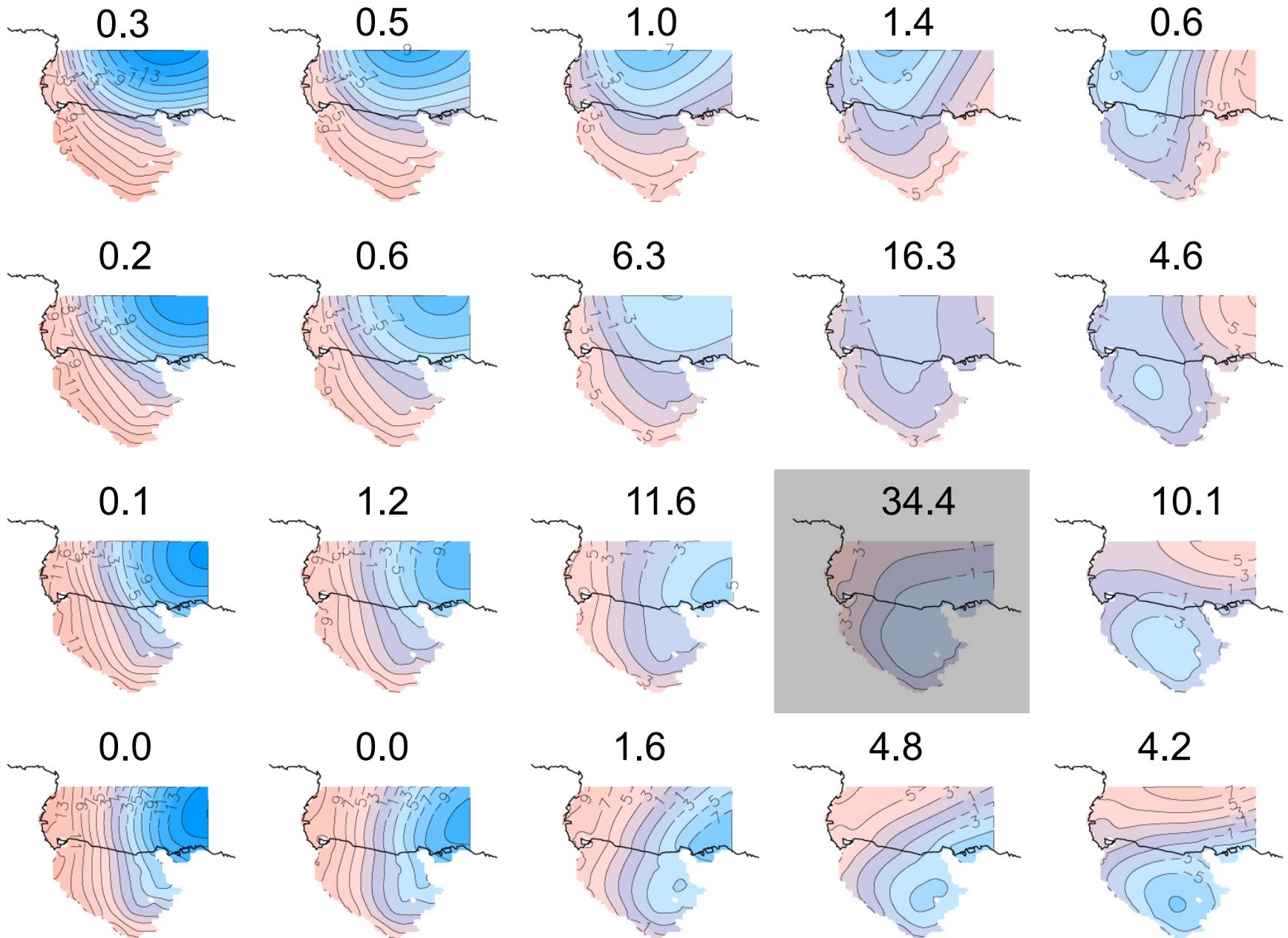
# Frequency of Occurrence (000 : 036)



# AMPS 012 Forecasts: Node (4,3)



# AMPS 060 Forecasts: Node (4,3)



# Marilyn: Wind Speed

