



Identifying Teleconnection Patterns from Point Correlation Maps using Self Organizing Maps

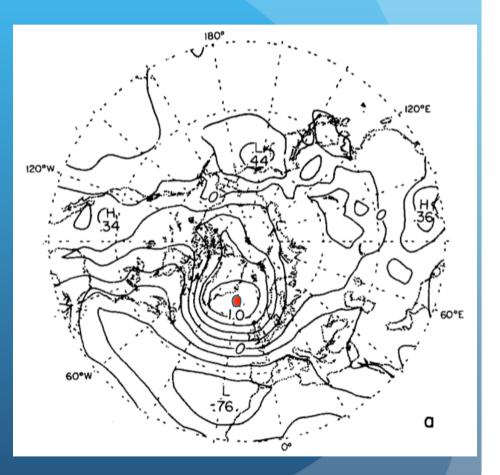
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Correlation Maps

- Wallace & Gutzler 1981
- Correlate a grid point with every other grid point on the map for all grid points





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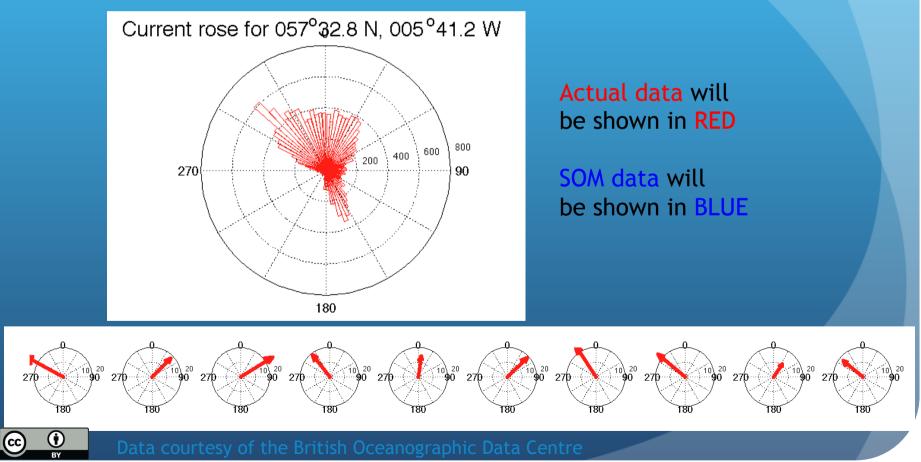
Self Organizing Maps (SOMs)

• What is a SOM?

- An unsupervised non-linear neural network
- Finds representative patterns in the data
- Results are arranged topologically
 - Similar results are close together, different results are far apart
- Examples of SOMs in teleconnections
 - Leloup et al 2008 ENSO, Johnson et al 2008 NAO
- Simple example

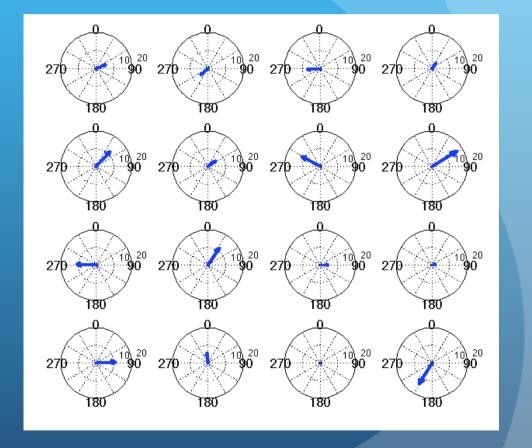


• Current data from a moored buoy in Loch Shieldaig, Scotland



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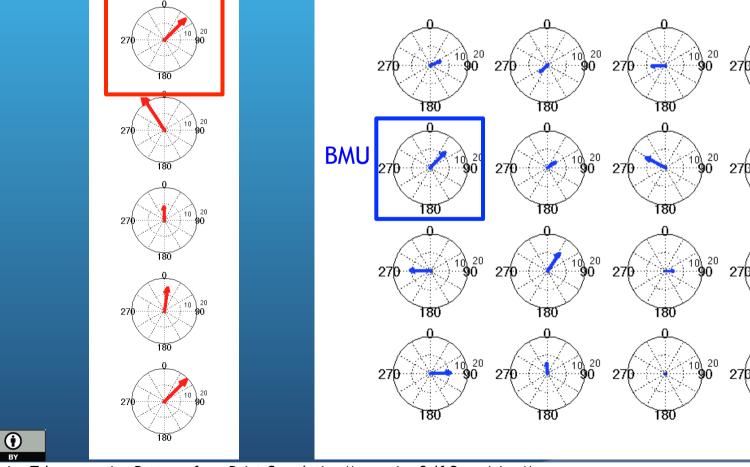
- Initialization
 - How many patterns?
 - Starting patterns





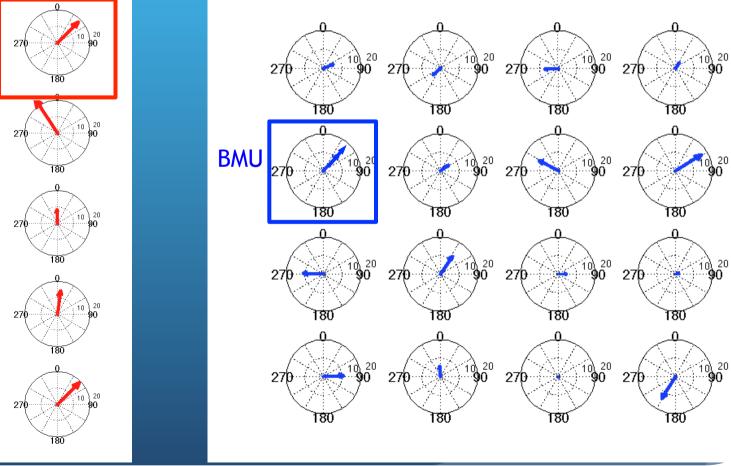
2. Locate SOM pattern most similar to data pattern

- 1. **Present each data pattern to SOM**
- Locate BMU
 Update BMU
 - Update BMU learning rate
 - Update neighbors neighborhood function
 - Learning rate and neighborhood function reduce over time



3. Update BMU to more closely resemble the data pattern

- Present each data pattern to SOM
 Locate BMU
 - Update BMU learning rate
 - Update neighbors neighborhood function
 - Learning rate and neighborhood function reduce over time



3.

 $(\mathbf{\hat{f}})$

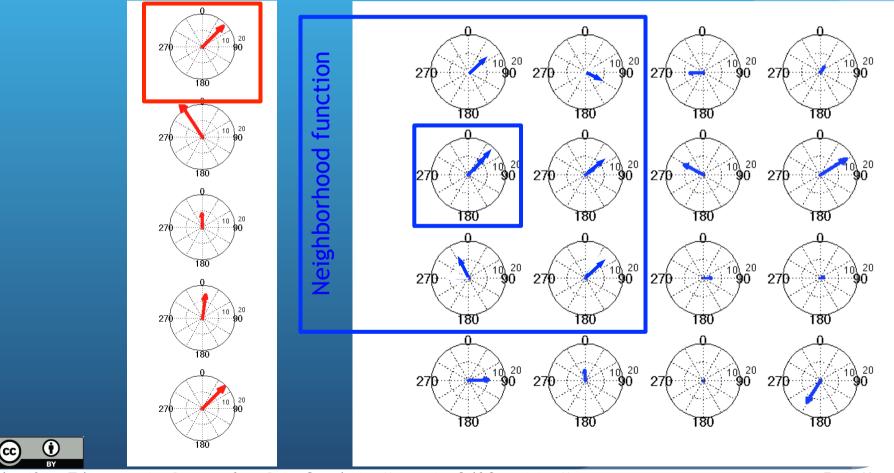
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4. Update neighboring SOM patterns

- Present each data pattern to SOM
 Locate BMU
 - Update BMU learning rate

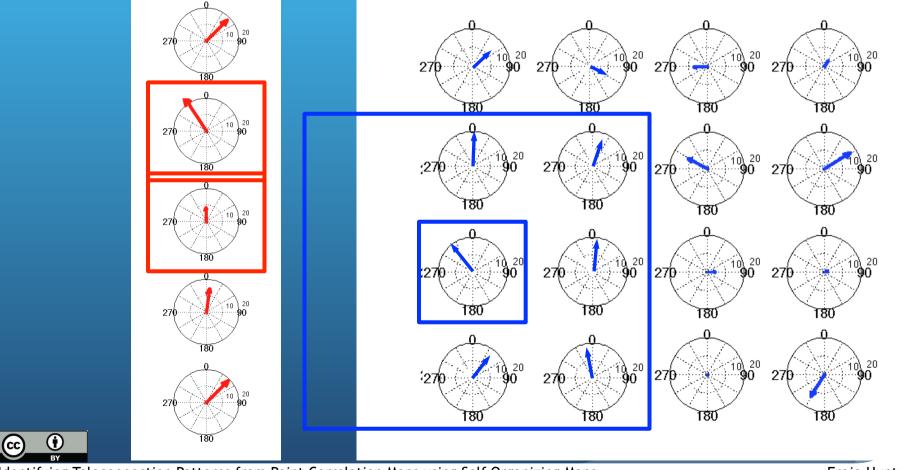
3.

- 4. Update neighbors neighborhood function
- Learning rate and neighborhood function reduce over time



1. Iteratively present each data pattern to SOM

- 1. Present each data pattern to SOM
- 2. Locate BMU
- 3. Update BMU learning rate
- 4. Update neighbors neighborhood function
 - Learning rate and neighborhood function reduce over time

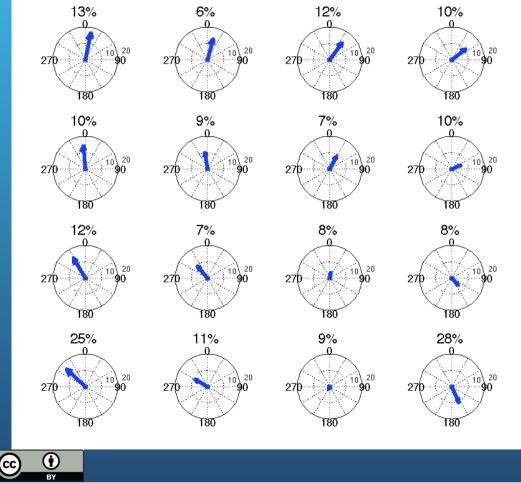


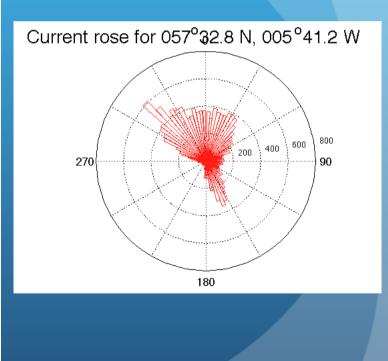
• Comparison

- Compare original data patterns with SOM patterns
- For each data pattern find its BMU
- Add up number of times each SOM pattern is BMU to get 'hits'
 - Frequency of occurrence



Percentage frequency occurrence of each SOM pattern in the original data





Identifying Teleconnection Patterns from Point Correlation Maps using Self Organizing Maps

Correlation Map SOMs

- Gridded data set
- Point correlation maps for each grid point
- nx by ny correlation maps
- Present correlation maps to SOM rather than raw data
- Advantages:
 - Correlation maps already highlight related regions
 - SOM summarizes patterns
 - No requirement for orthogonality



Idealized Self Organizing Maps

- Rectangular domain
- Simple north-south oscillation
- Plus east-west oscillation in northern half
- Add noise

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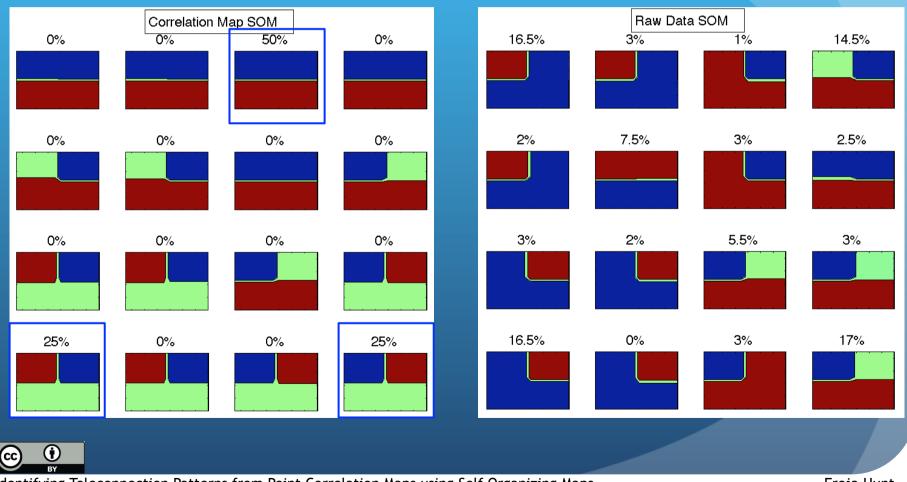
- Construct point correlation maps for each grid point
- Present to 4 x 4 SOM
- Also SOM from raw data



Idealized SOM

N-S oscillation, + E-W oscillation, no noise

Red = positive Blue = negative Green = zero

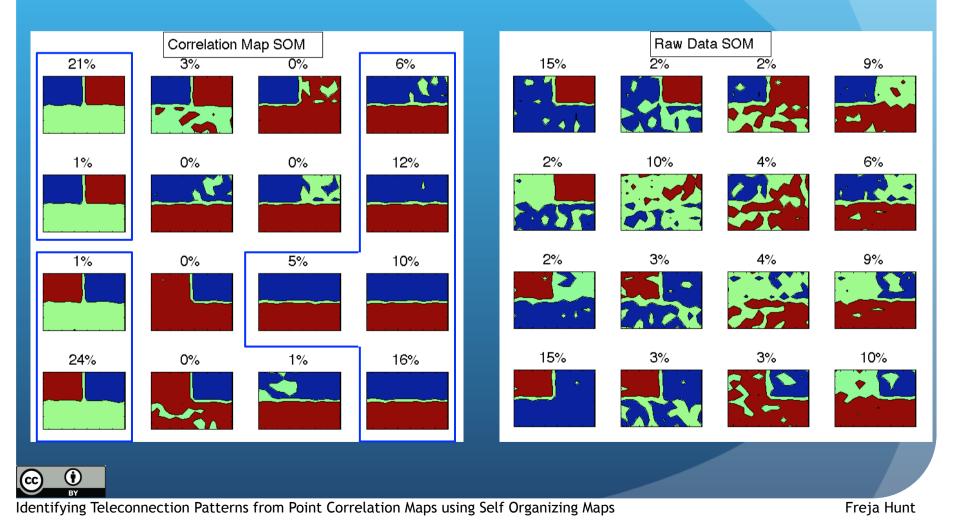


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Idealized SOM

Red = positive Blue = negative Green = zero

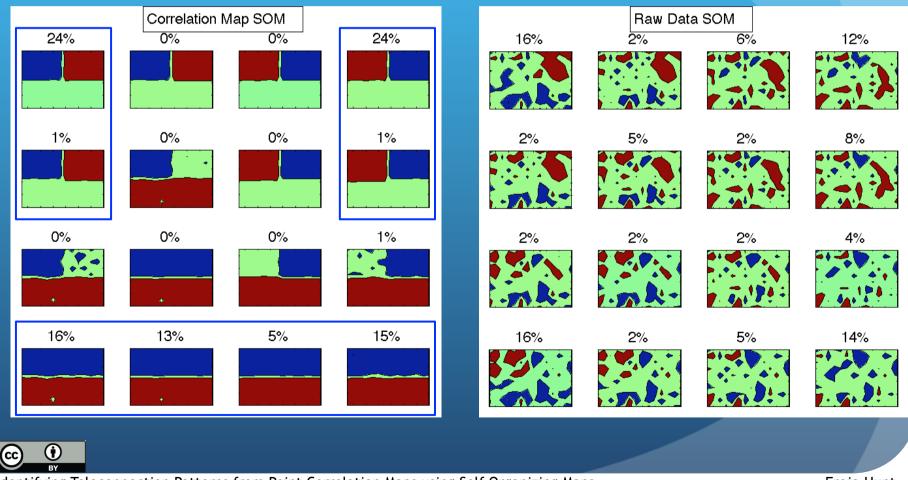
N-S oscillation, + E-W oscillation, + white noise



Idealized SOM

Red = positive Blue = negative Green = zero

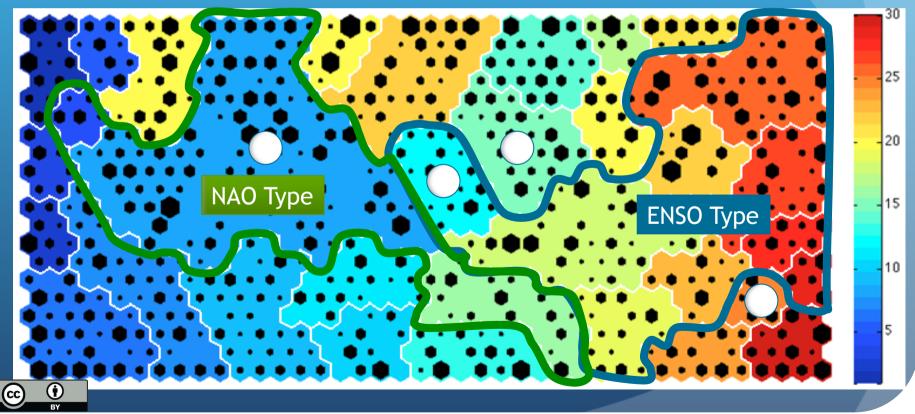
N-S oscillation, + E-W oscillation, + random walk



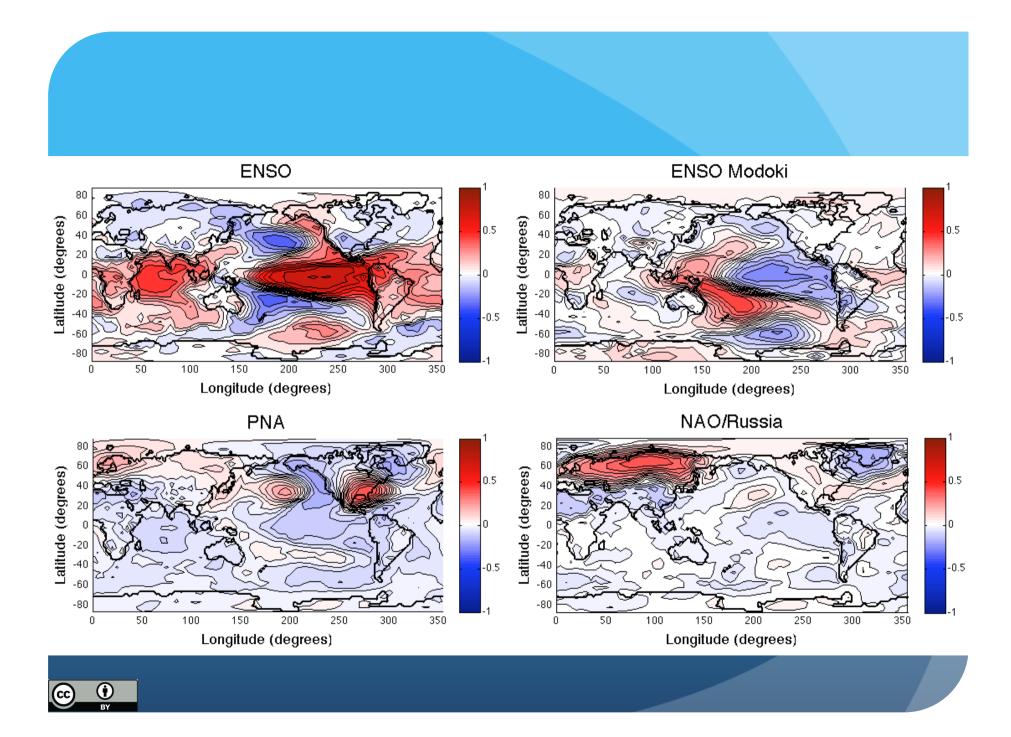
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Temperature SOM

- 20 x 40 SOM NCEP/NCAR monthly 2m temperature anomalies
- 1.1948 to 11.2008



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Conclusions

- Correlation maps + SOMs effectively identify and summarize teleconnections
- Advantage over raw data as relationships already defined
- Advantage over EOFs as no orthogonality
- Flexible method use comparison stage in many different ways to get different insights into large datasets
- Validating model behavior

