



EMPIRICAL ORTHOGONAL FUNCTION (EOF) ANALYSIS & THERMOSPHERIC NITRIC OXIDE FLUX

SIERRA FLYNN¹, DELORES KNIPP¹, TOMOKO
MATSUO¹, MARTIN MLYNCZAK², LINDA HUNT³

¹CU BOULDER AEROSPACE ENGINEERING SCIENCES
(AES), ²NASA LANGLEY RESEARCH CENTER, ³SCIENCE
SYSTEMS AND APPLICATIONS, INC.

CURRENT CHALLENGES & GOALS IN GEOSPACE SYSTEMS SCIENCE

Understanding data

- Large data sets
- Multivariate
- Highly correlated
- Various sources
- Sparse in space and time

Presenting data

- Physical insight
- Straightforward
- Raising awareness
- Aesthetically pleasing
- Inspiring interests

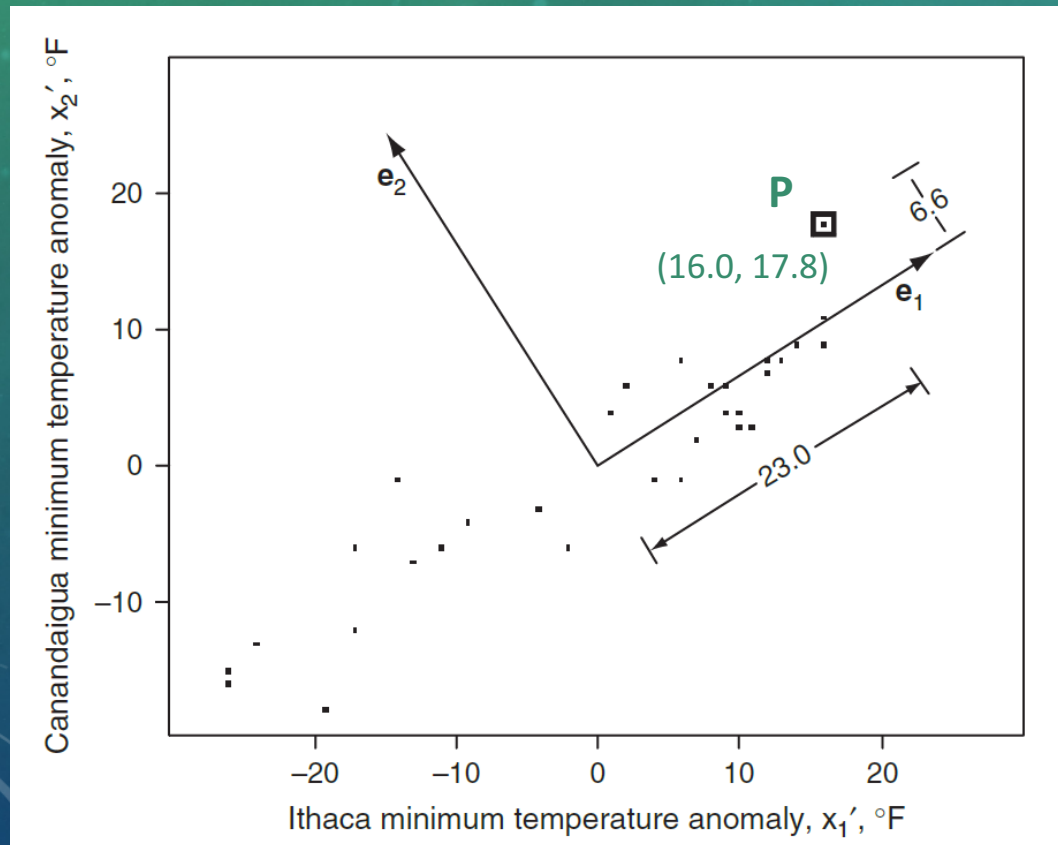
Preparing data

- Compressing data
- Removing noise
- Encouraging collaboration
- Learning through data mining
- Applying predictive modeling

PROPOSED METHOD

EMPIRICAL ORTHOGONAL FUNCTION (EOF) ANALYSIS

- Based on eigenvector-eigenvalue decomposition, e.g.:



(Wilks, 2006)

- Remove the mean of this 2-D data set of minimum temperatures.
- Then, it can be represented as an orthonormal pair of **eigenvectors**:
 - e_1 points in the direction of the most variation (96.8%!)
 - **Eigenvalues** represent the % variance explained by their eigenvector
- Point P can then be defined as a linear combination of these orthonormal eigenvectors:

$$P = 23.0e_1 + 6.6e_2 \approx 23.0e_1$$

PROPOSED METHOD

EMPIRICAL ORTHOGONAL FUNCTION (EOF) ANALYSIS

- EOF Analysis: based on eigenvector-eigenvalue decomposition, but multivariate and time-dependent
- Reduce data set into its mean and its variability using an empirical, non-linear regression-based technique
- Describe the variability empirically using n orthogonal **eigenmodes** (spatial, time-invariant maps, or **EOFs**) and their **eigenvalues** (variability in time, or **EOF coefficient**):

$$\mathbf{y}'(\theta, \varphi, t) \approx \Psi \alpha = \sum_{i=1}^n \alpha_i(t) \Psi_i(\theta, \varphi)$$

Residual after
mean subtracted

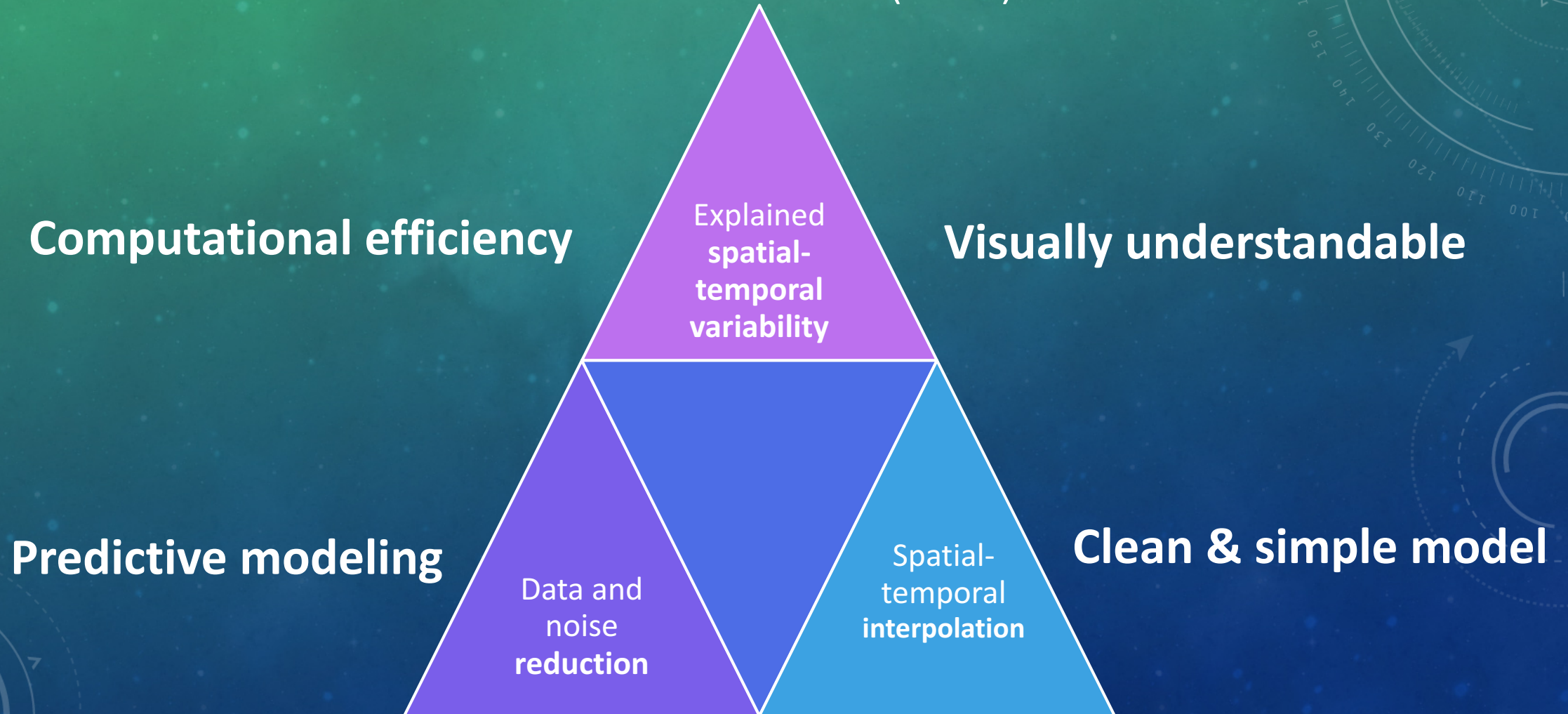
Time-varying
amplitude of EOF_{*i*}

EOF_{*i*} in spatial
coordinates

(Matsuo & Forbes, 2010)

ACCOMPLISHMENTS

EMPIRICAL ORTHOGONAL FUNCTION (EOF) ANALYSIS

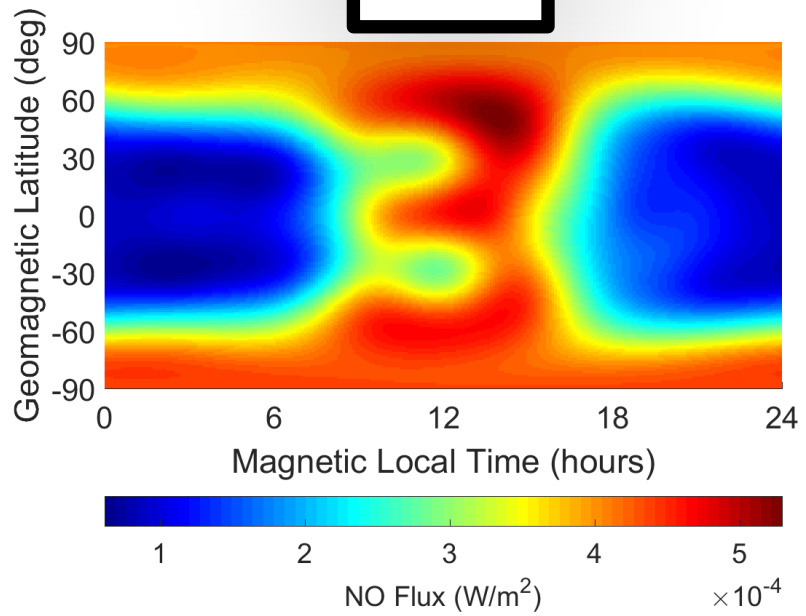


APPLICATIONS: EOFs

THERMOSPHERIC NITRIC OXIDE FLUX

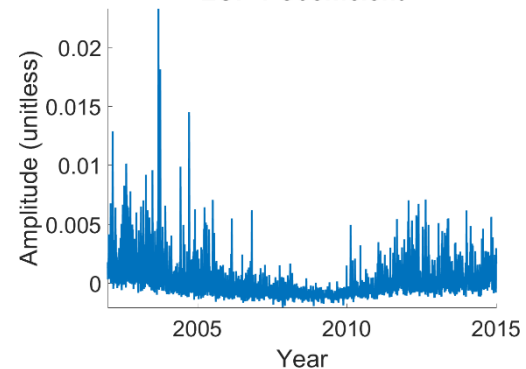
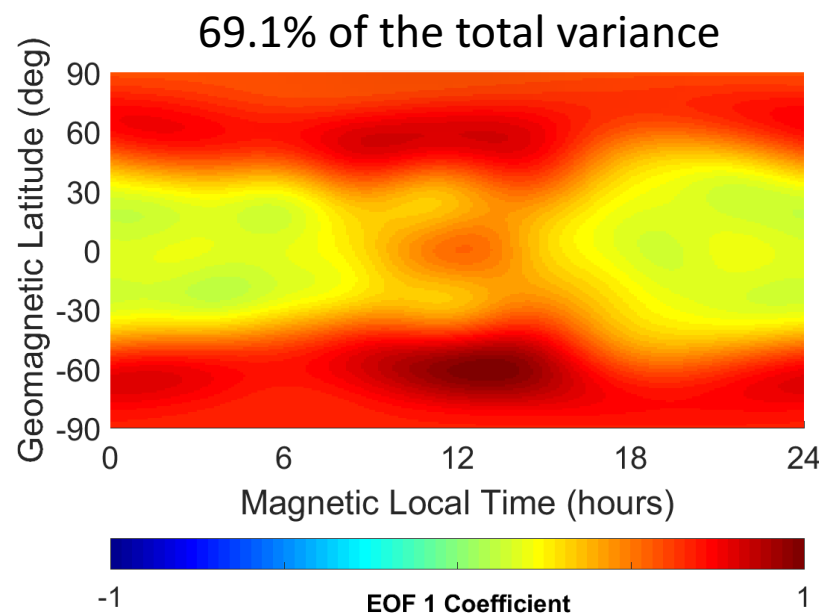
$$y'(\theta, \varphi, t) \approx \sum_{i=1}^n \alpha_i(t) \Psi_i(\theta, \varphi)$$

Mean

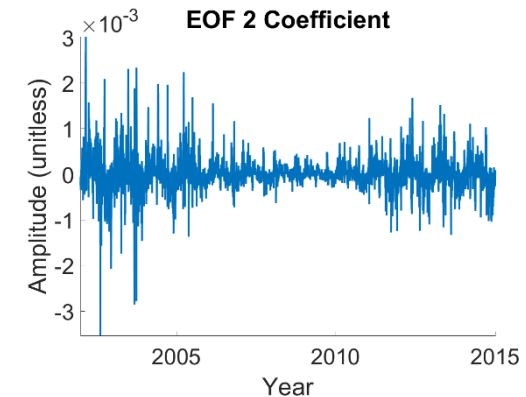
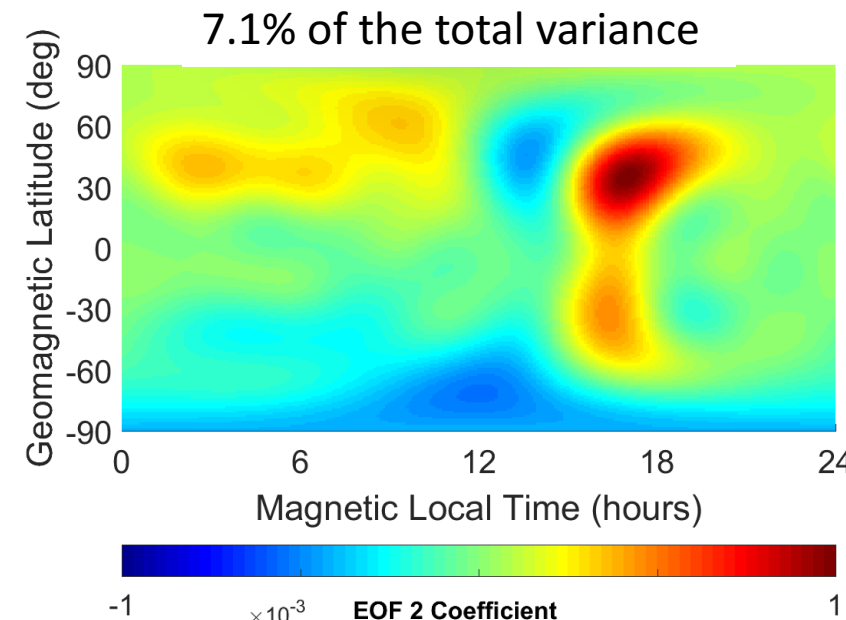


(Does not vary in time)

EOF 1



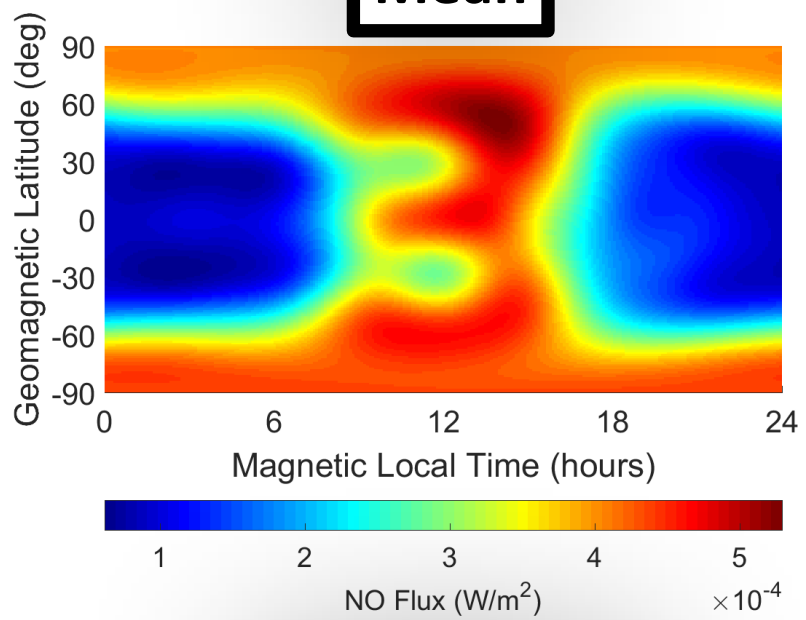
EOF 2



APPLICATIONS: CORRELATION ANALYSIS THERMOSPHERIC NITRIC OXIDE FLUX

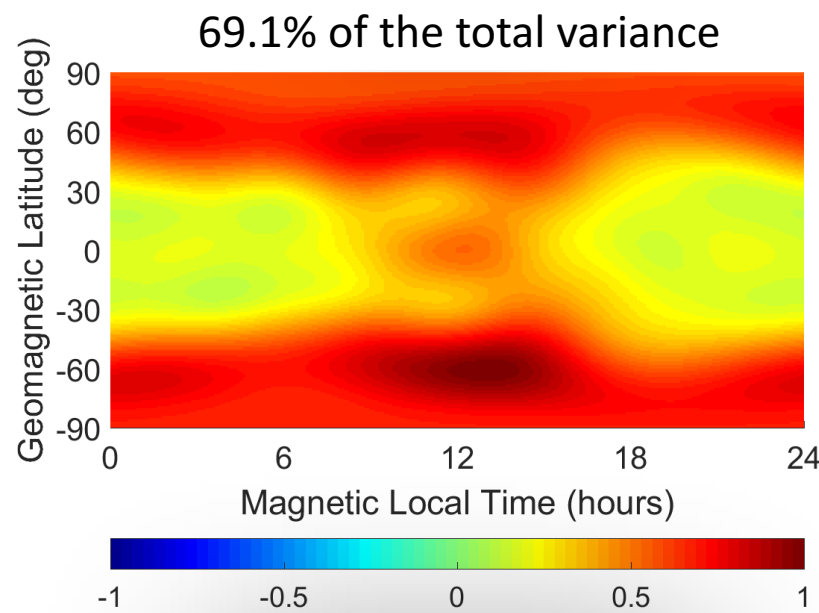
$$y'(\theta, \varphi, t) \approx \sum_{i=1}^n \alpha_i(t) \Psi_i(\theta, \varphi)$$

Mean



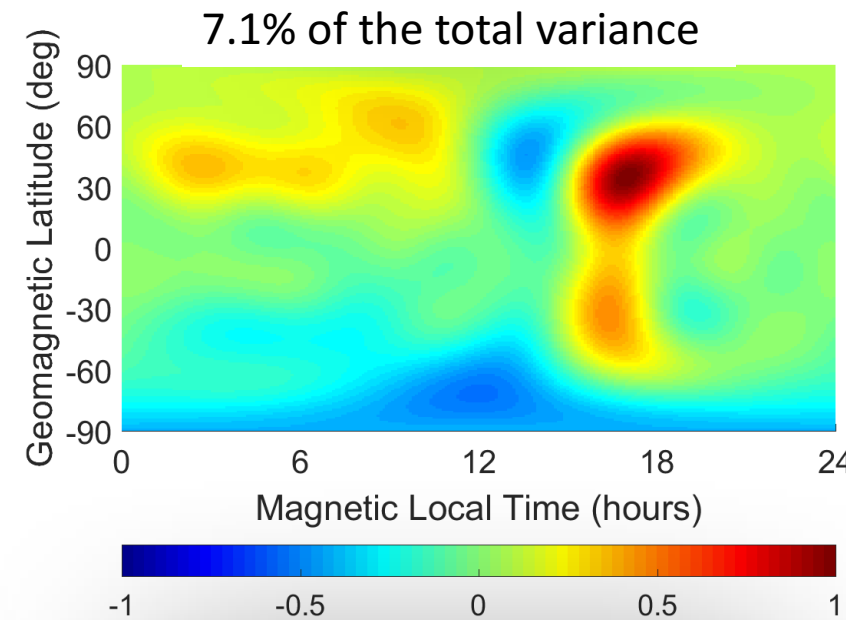
Diurnal & Auroral Drivers

EOF 1



Geomagnetic & Solar Storms
Kp (R = 0.68, 10 hour lag)
F10.7 (R = 0.63, 3 day lag)

EOF 2



Annual & Seasonal Cycles
annual (R = 0.45, peaks @ aphelion)
seasonal (R = 0.17, peaks @ equinox)

CONCLUDING REMARKS

EMPIRICAL ORTHOGONAL FUNCTION (EOF) ANALYSIS

Reconstruct a **global NOF map** that represents **85% of the original covariance** and condenses 13 years of data into a data set that is **6% of its original size!**



Correlation analysis can help determine the **primary geophysical drivers** for each EOF



Further regression modeling can lead to **predictive models**

For more information, see my poster (DATA-01) tomorrow!

REFERENCES

Matsuo, T. & Forbes, J. (2010). Principal modes of thermospheric density variability: Empirical orthogonal function analysis of CHAMP 2001–2008 data. *Journal of Geophysical Research*, 115(A07309). doi:10.1029/2009JA015109.

Wilks, D. S. (2006). *Statistical Methods in the Atmospheric Sciences: Second Edition*. Burlington, MA: Elsevier Inc.