

The EC H2020 project SWAMI: Space Weather Atmosphere Model and Indices

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Project objectives

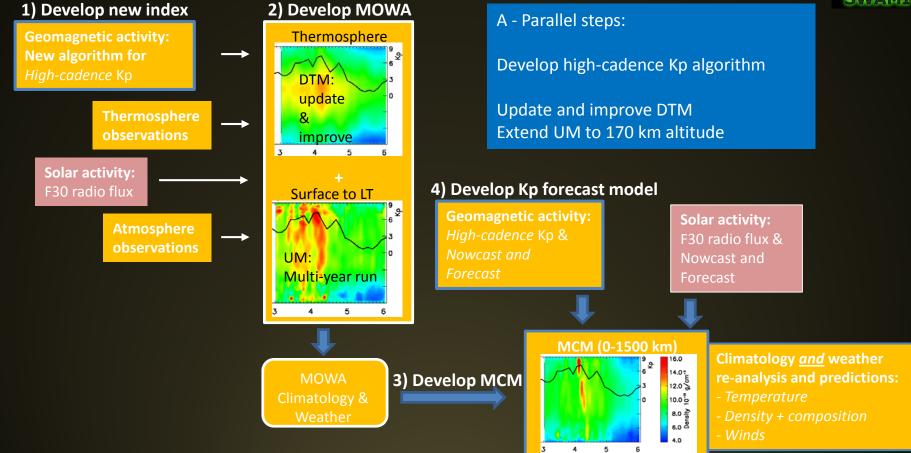


- To develop a model of the whole atmosphere (MOWA) with a science as well as operations-focused approach (MCM). Two existing models of the atmosphere, the UM and the DTM, will be extended and blended to produce this unique new whole atmosphere model, which shall provide estimates of both climatology and space weather variability.

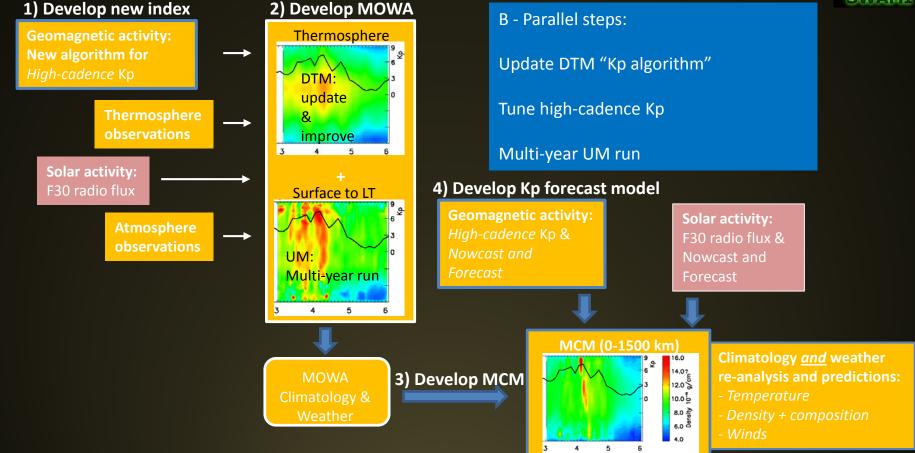
- To provide new high-cadence geomagnetic Kp indices, including its nowcast and predictions to be used in the UM and DTM.

- To develop steps, including provision of software, model output, or data sharing facilities, to transition the improved model system into operations.

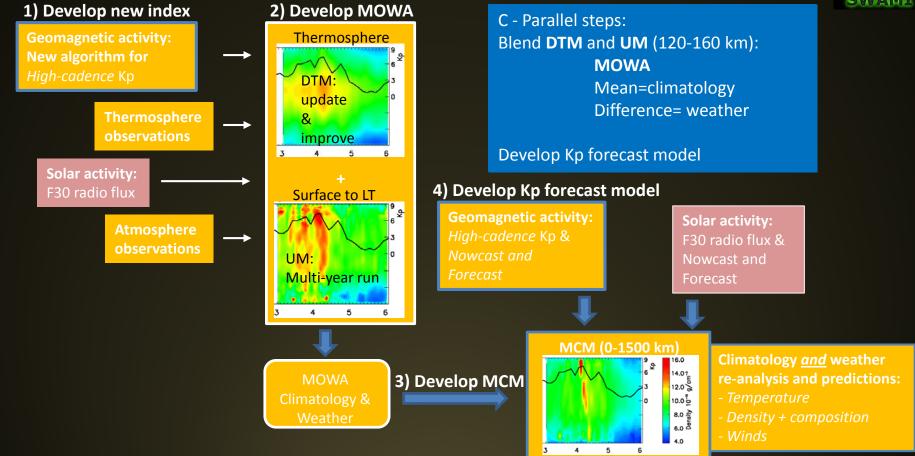




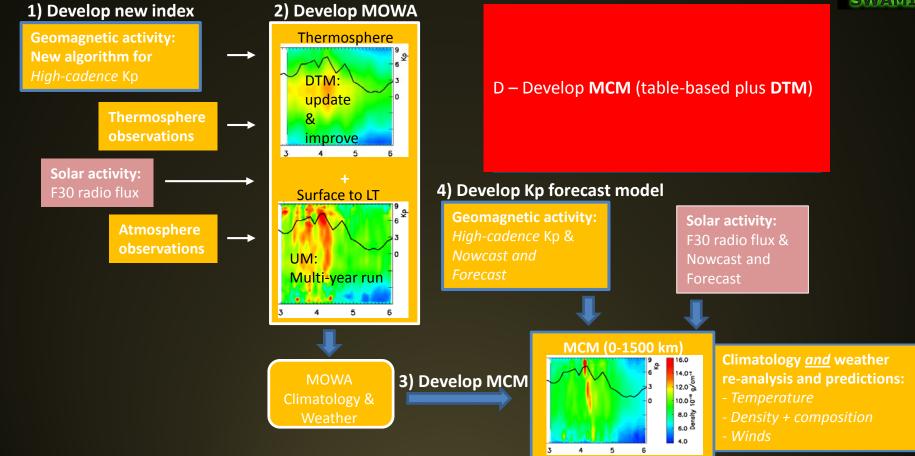












High-Cadence Global Kp Index



New high-cadence Kp index

- Useful for a wide range of space weather services that rely on rapid geomagnetic activity specification
- Will be used to drive the models, enabling accurate phasing of storm events

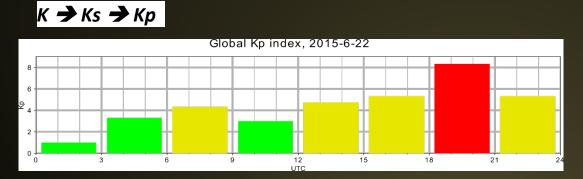
From standard to high-cadence

Adaptation of mature algorithms and implementations for the 3-hour cadence: local K index

Local K and Global Kp Indices







Local K index

- Niemegk station (e.g.): mean values of Bx and By components
- Quiet-day variation pattern removal procedure
- K index: 3-hourly time intervals for the most disturbed field component

Standardized Ks index

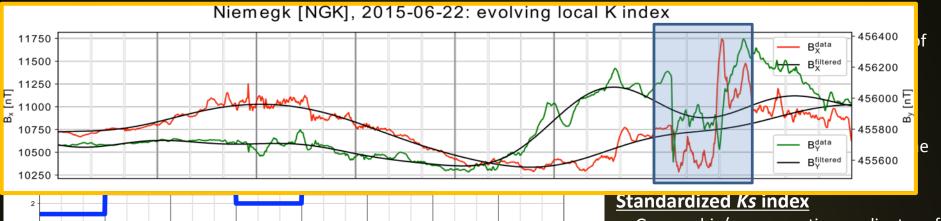
- Geographic/geomagnetic coordinates of stations: an annual cycle of daily variations
- Ks index: conversion tables to eliminate these effects

Global Kp index

$$Kp = \frac{1}{11} \sum_{i=1}^{11} Ks^{i}$$

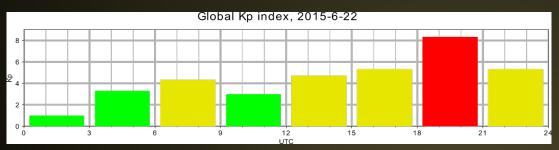
Local K and Global Kp Indices





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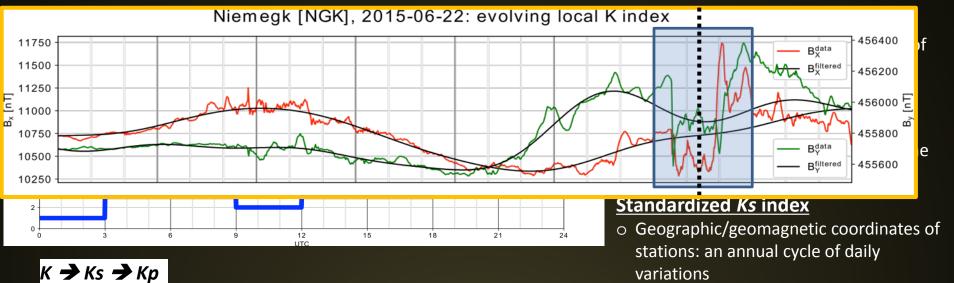
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Global Kp index

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Local K and Global Kp Indices

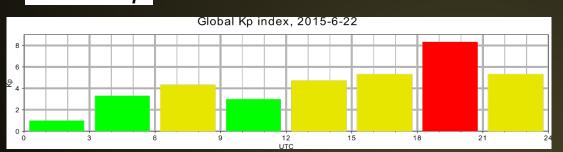




Global Kp index, 2015-6-22 o Ks index: conversion tables to eliminate these effects

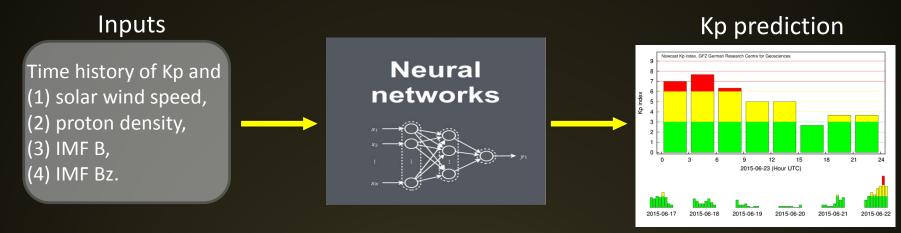
Global Kp index

$$Kp = \frac{1}{11} \sum_{i=1}^{11} Ks^{i}$$



Kp forecast: Methodology



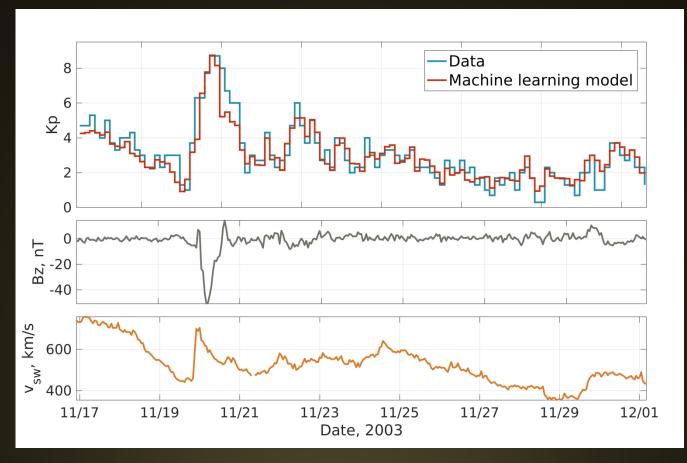


- Methodology: neural network based empirical modelling.
- <u>Data</u>: solar wind and IMF data from ACE (available at OMNIWeb),
 Kp index from GFZ Potsdam, 1993 2017.

Wing, S., J. R. Johnson, J. Jen, C.-I. Meng, D. G. Sibeck, K. Bechtold, J. Freeman, K. Costello, M. Balikhin, and K. Takahashi (2005), Kp forecast models, J. Geophys. Res., 110, A04203, doi:10.1029/2004JA010500
Wintoft, P., Wik, M., Matzka, J. and Shprits, Y. (2017) Forecasting Kp from high and low resolution solar wind and limitations on lead time, Journal of Space Weather and Space Climate, https://doi.org/10.1051/swsc/2017027

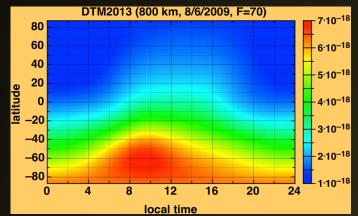
Kp-forecast: Preliminary results

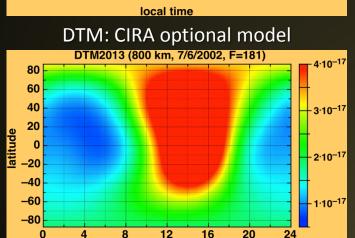




Thermosphere: DTM model







local time

Semi-empirical model:

- Low resolution
- Easy and fast in use
- Relatively accurate
- Climatology

Temperature and constituents (i.e., the winter Helium bulge is present) are modeled:

Concentration at 120 km

Height function

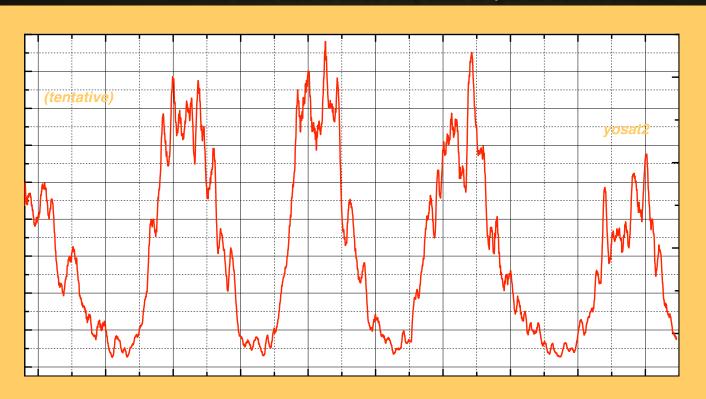
$$\rho(z) = \sum_{i} \frac{m_i}{N_A} c_i (120 \text{ km}) f_i(z) \exp(G_i(L))$$

Spherical harmonics

DTM model: data



NB: No data above 1000 km; data is sparse below 300 km

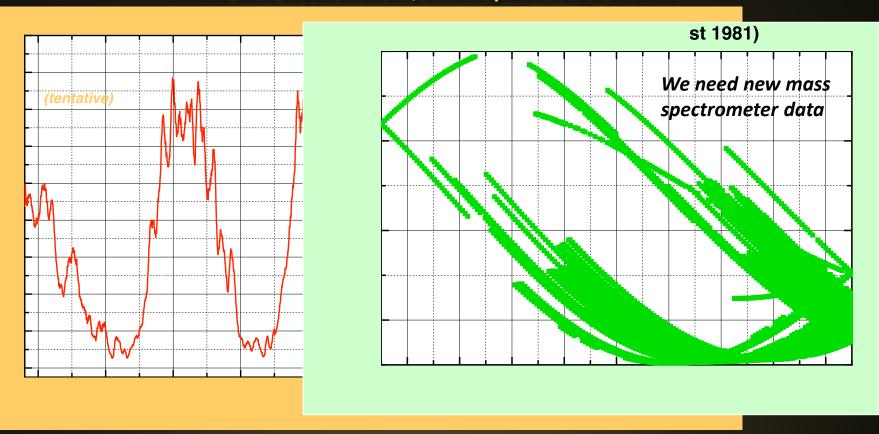


And:
QB50?
Dellingr?
GOLD?
APOD?
GRACE-FO?
...?

DTM model: data



NB: No data above 1000 km; data is sparse below 300 km

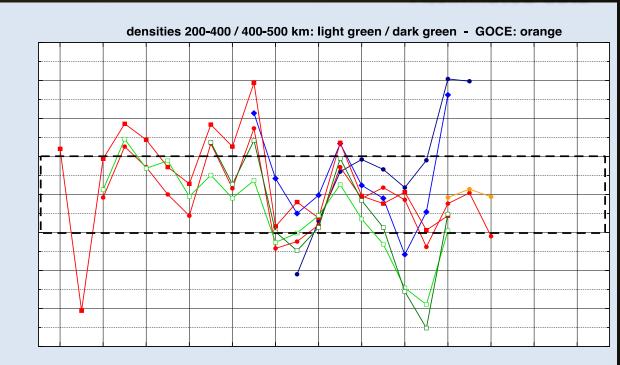


DTM2013: benchmark



DTM2013 *annual mean density ratios* (O/C): all altitudes (satellites)

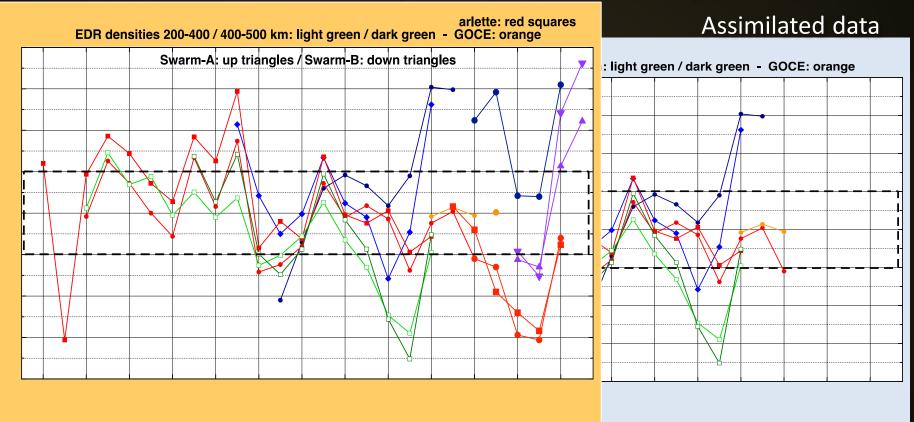
Assimilated data



DTM2013: benchmark



DTM2013 annual mean density ratios (O/C): all altitudes (satellites)

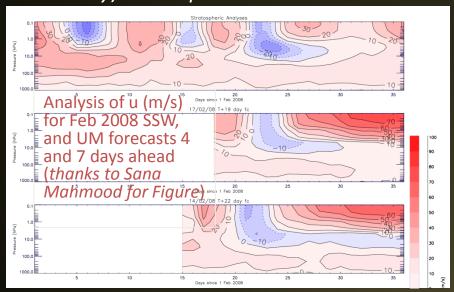


UM is a good candidate to represent MLT



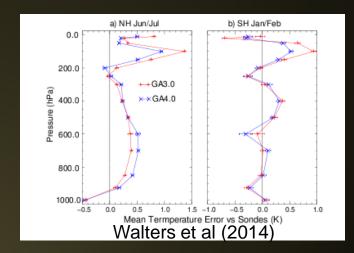
Unified Model (UM) has the distinct advantage of non-hydrostatic, deep atmosphere dynamical formulation (unlike other models: TIEGCM, WAM, WACCM,...)

Comprehensive dynamics and physics; all key waves (and associated atmospheric variability) well represented



Accurate;

Analysis / short-term forecasts have mean errors typically < 1 K. Multi-year climate run (w/o assimilation) have errors =< 5 K



Timeline products



