



# Forecast of Space Weather Effects with PROGRESS

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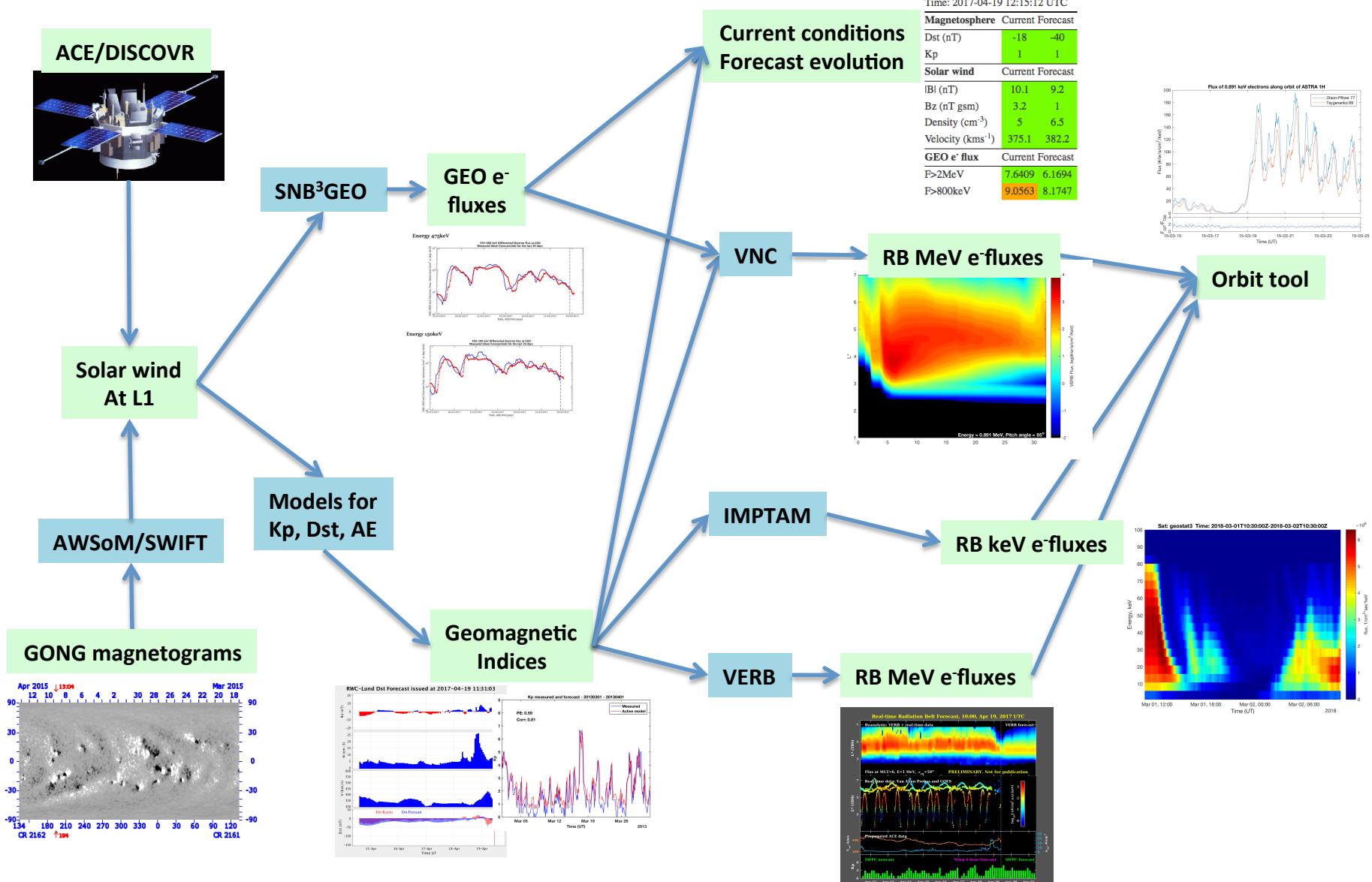
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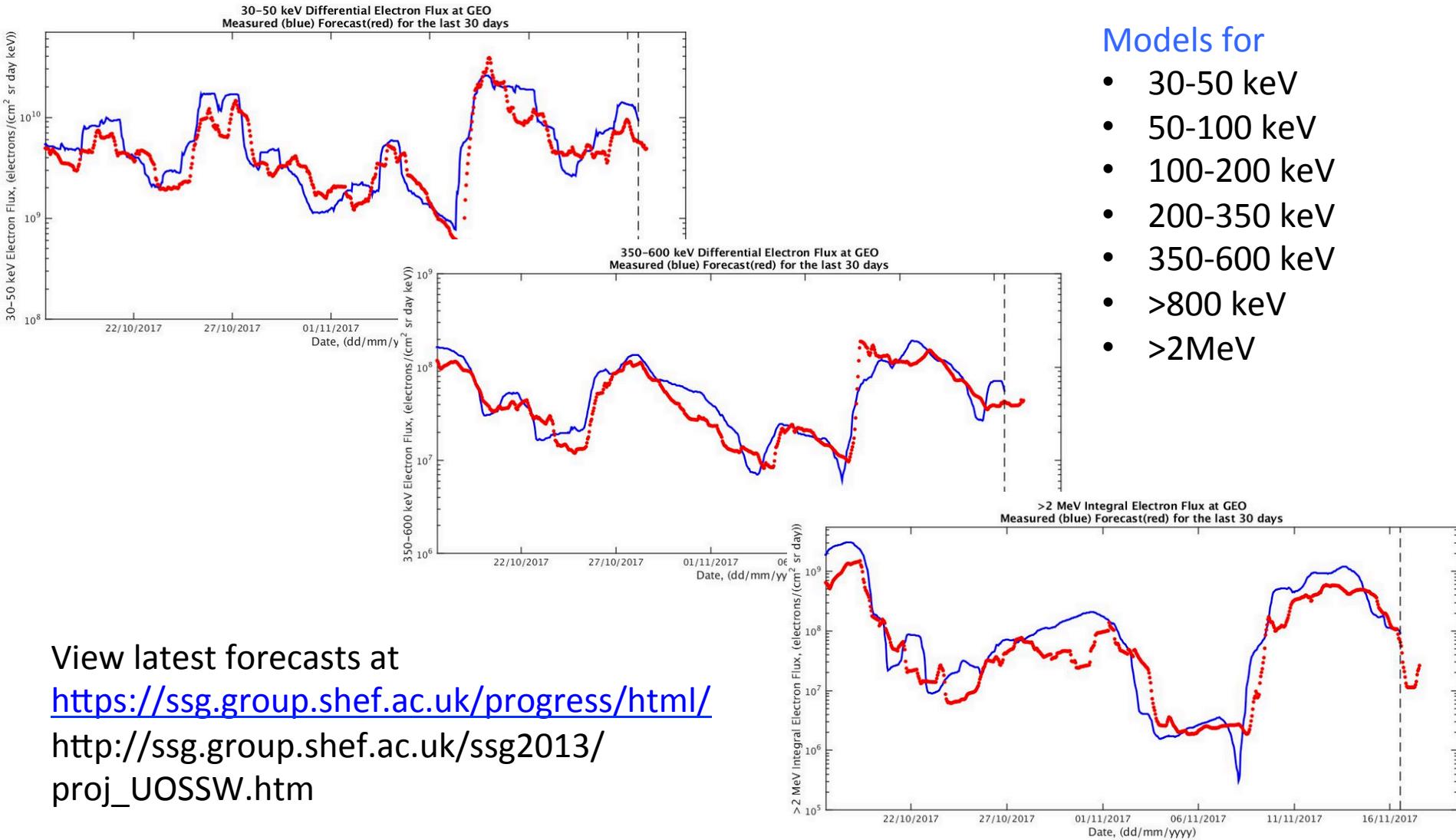
[ssg.group.shef.ac.uk/progress/html](http://ssg.group.shef.ac.uk/progress/html)

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# PROGRESS



# GSO e<sup>-</sup> flux forecasts



# GSO e<sup>-</sup> flux forecasts

## Model comparison

One day ahead forecasted fluxes >2 MeV electrons compared with NOAA REFM

Prediction Efficiency

$$PE = 1 - \frac{1}{N} \sum \frac{(X_i - Y_i)^2}{\text{Var}(X)}$$

Correlation function

$$C_{\log(SNB)} = \frac{1}{N} \sum_{i=1}^N \frac{(\log_{10}(F_{2\text{MeV}}(i)) - \langle \log_{10}(F_{2\text{MeV}}(i)) \rangle)(\log_{10}(F_{SNB}(i)) - \langle \log_{10}(F_{SNB}(i)) \rangle)}{\sqrt{\text{Var}(\log_{10}(F_{2\text{MeV}}))}\sqrt{\text{Var}(\log_{10}(F_{SNB}))}}$$

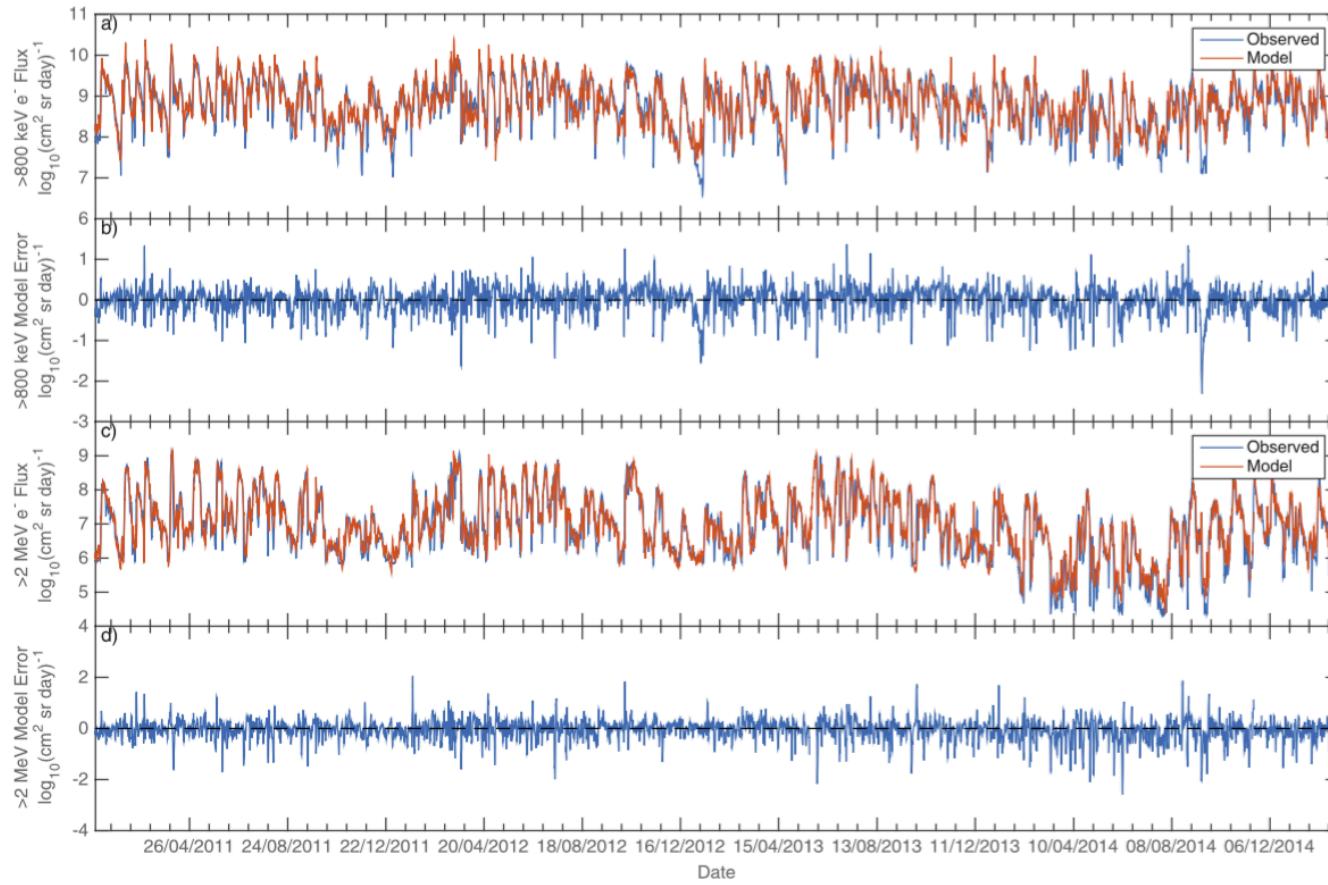
Model	e <sup>-</sup> Flux		Log10(e <sup>-</sup> Flux)	
	PE	Corr	PE	Corr
REFM	-1.31	0.73	0.70	0.85
SNB <sup>3</sup> GEO	0.63	0.82	0.77	0.89

Balikhin, M. A., et al. (2016), Comparative analysis of NOAA REFM and SNB3GEO tools for the forecast of the fluxes of high-energy electrons at GEO, *Space Weather*, 14, 22–31, doi:10.1002/2015SW001303.

# GSO e<sup>-</sup> flux forecasts

- Initially, models run once per day to generate forecasts
- Re-developed to use 24 hour running averages with a time resolution of 1 hour

Energy	PE	$\rho$
>800 keV	72.1%	85.1%
>2 MeV	82.3%	90.9%



# GSO e<sup>-</sup> flux forecasts

Use of 24 hr running averages loses any MLT structure

Set of MLT models currently under developed  
Main problem – lack of continuous  
measurements in different MLT sectors

Current models based on GOES 13/15 data  
Only at specific MLT locations for a short time

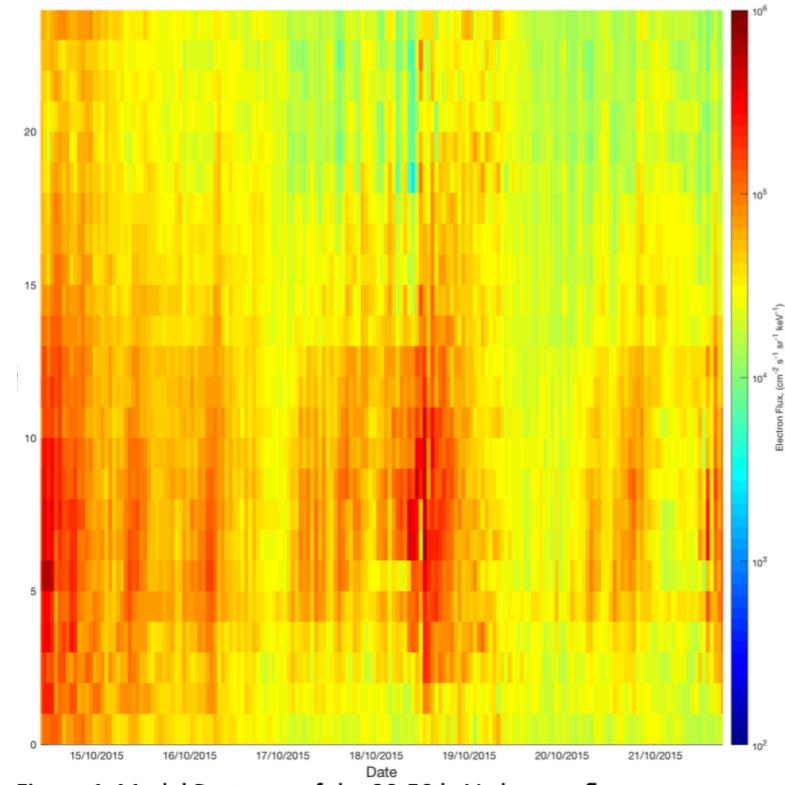
Use 1 hr MLT bins

Training period: 1 Jan 2011 – 1 Mar 2013

Test period: 2 Mar 2013 – 31 Dec 2017

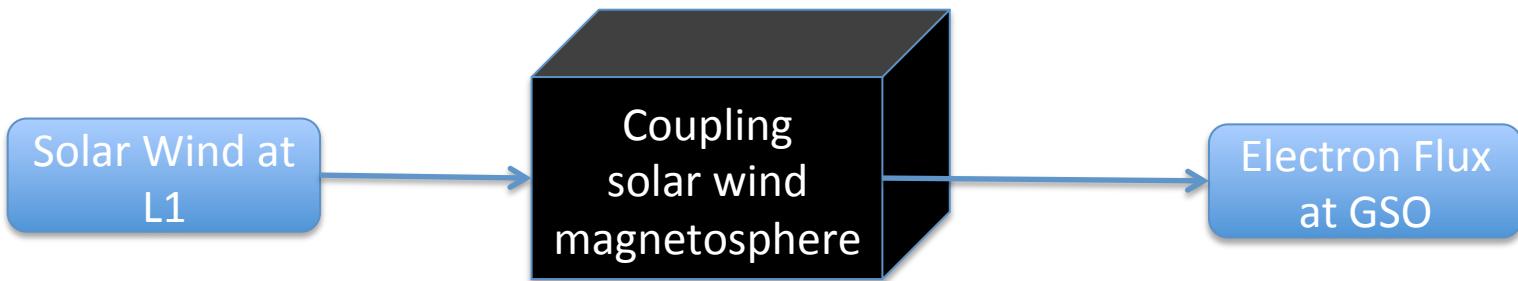
Model performance

PE: 47-75%, Correlation 51-79%



Past cast fluxes of 30-50 keV  
electrons

# Electron Fluxes at GSO



Energy	Term 1	%ERR	Term 2	% ERR
90 keV	$V(t)$	97.0	$V^2(t)$	2.7
127.5 keV	$V(t)$	74.8	$V(t-1)$	22.2
172.5 keV	$V(t-1)$	65.7	$V(t)$	31.6
270 keV	$V(t-1)$	97.5	$V^2(t-1)$	2.3
407.5 keV	$V(t-1)$	84.1	$V(t-2)$	13.7
625 keV	$V(t-1)$	75.9	$V(t-2)$	22.3
925 keV	$V(t-2)$	96.2	$N(t)$	0.3
1.3 MeV	$V^2(t-2)$	76.5	$nV(t-1)$	2.2
2.0 MeV	$N(t-1)$	53.7	$nV(t-1)$	13.6
1.8-3.5 MeV	$N(t-1)$	51.5	$N^2(t-1)$	15.1

Boynton, R. J., et al., (2013), The analysis of electron fluxes at geosynchronous orbit employing a NARMAX approach, *J. Geophys. Res. Space Physics*, 118, 1500–1513, doi: [10.1002/jgra.50192](https://doi.org/10.1002/jgra.50192).

# Geomagnetic Indices

## AIM: Forecasts of K<sub>p</sub>, Dst, and AE

Methodologies – data driven

- Neural Network –IRF Lund, Sweden
- NARMAX – U. Sheffield, UK

## Model inputs – solar wind parameters at L1

- Measurements from ACE/DSCOVR
- L1 forecasts from AWSOM/SWIFT

# Geomagnetic Indices

## Forecasts available

IRF Lund Dst, and Kp

Methodology: Neural Nets

Plots

<http://lund.irf.se/forecast/dst/dst.png>,

<http://lund.irf.se/forecast/kp/kp.png>

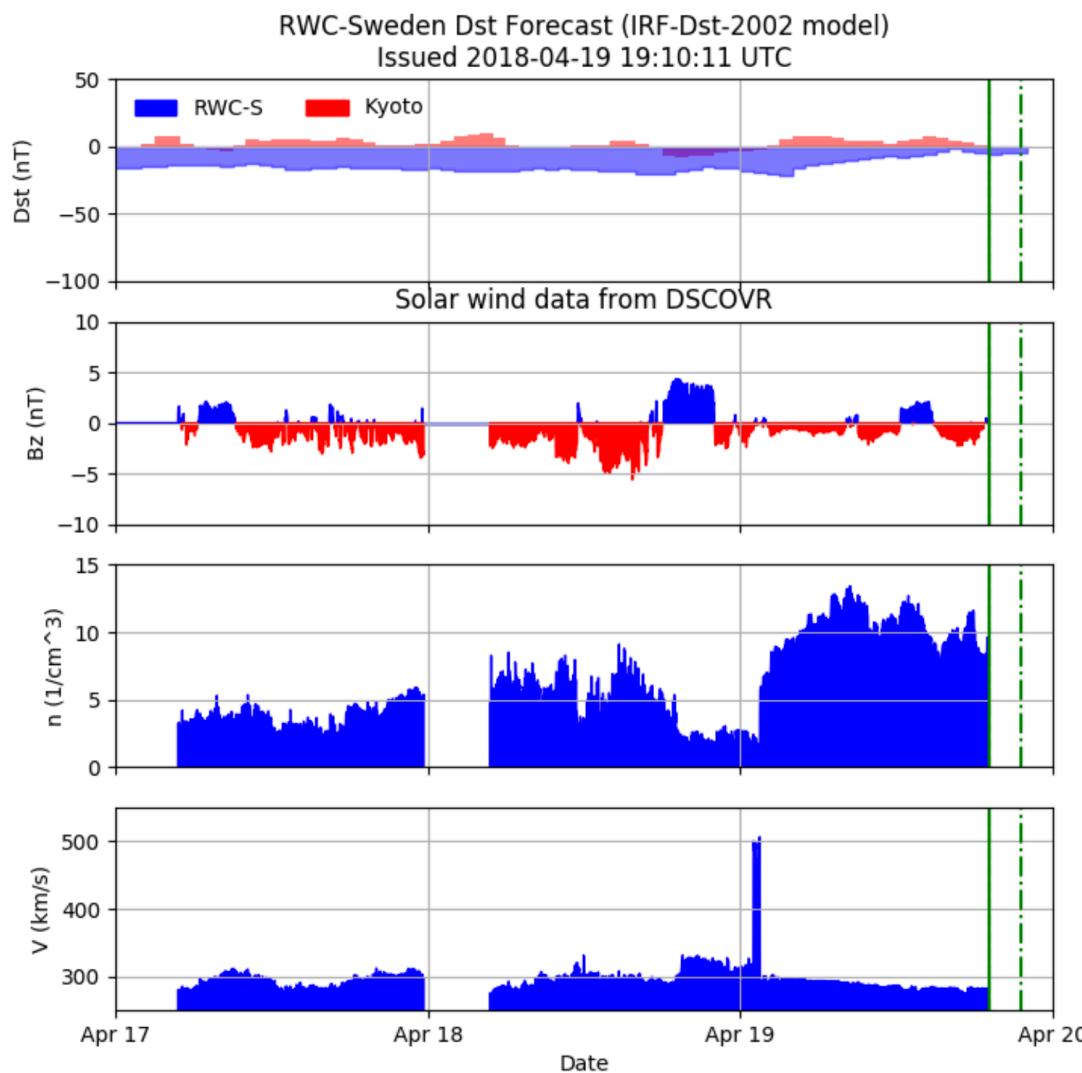
Data (Kp available now, Dst, AE available soon)

RESTful data server created

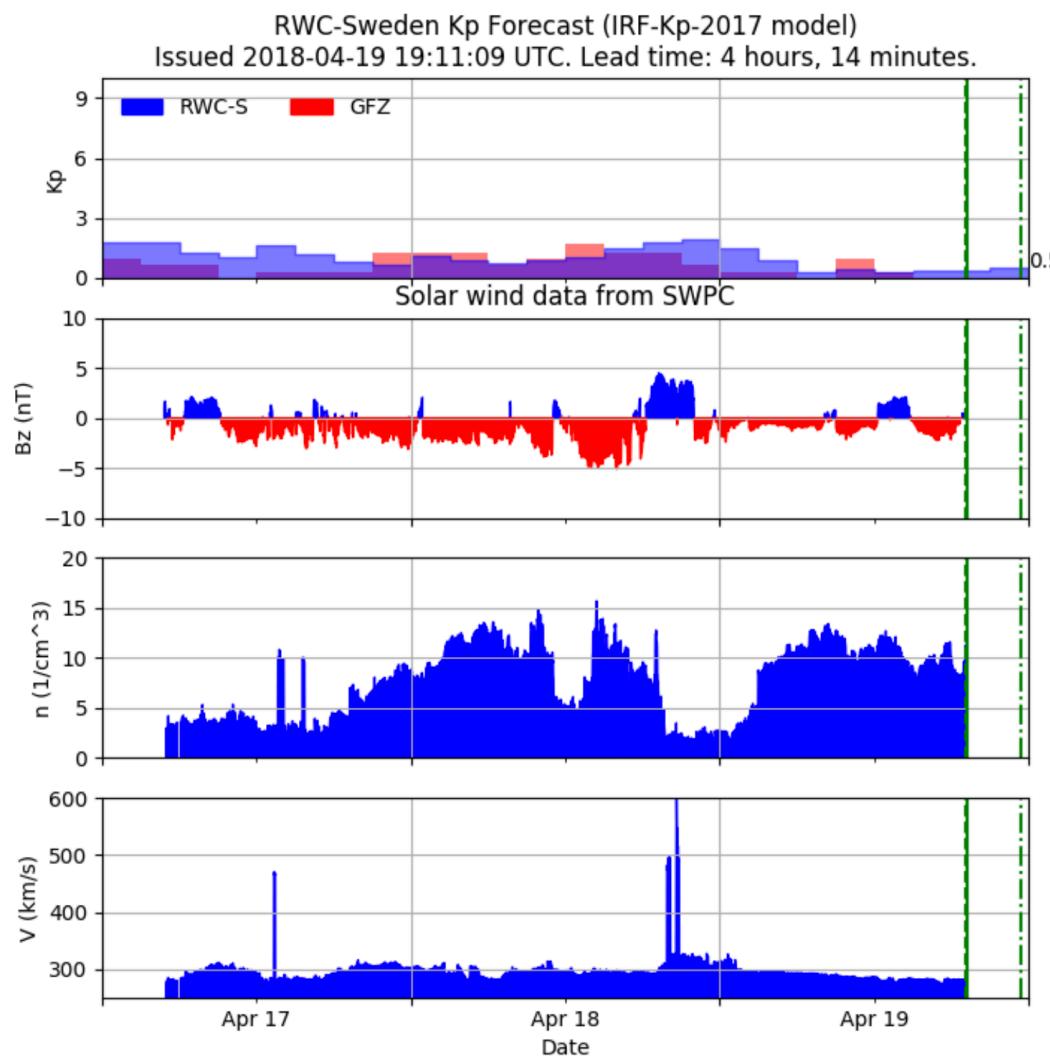
e.g. latest 10 Kp predictions:

[http://lund.irf.se/progress/rest/datasets/irfkp2017/latest?  
limit=10](http://lund.irf.se/progress/rest/datasets/irfkp2017/latest?limit=10)

# IRF Lund Dst



# IRF Lund Kp



# Geomagnetic Indices

## Forecasts available

U. Sheffield Kp – Plots

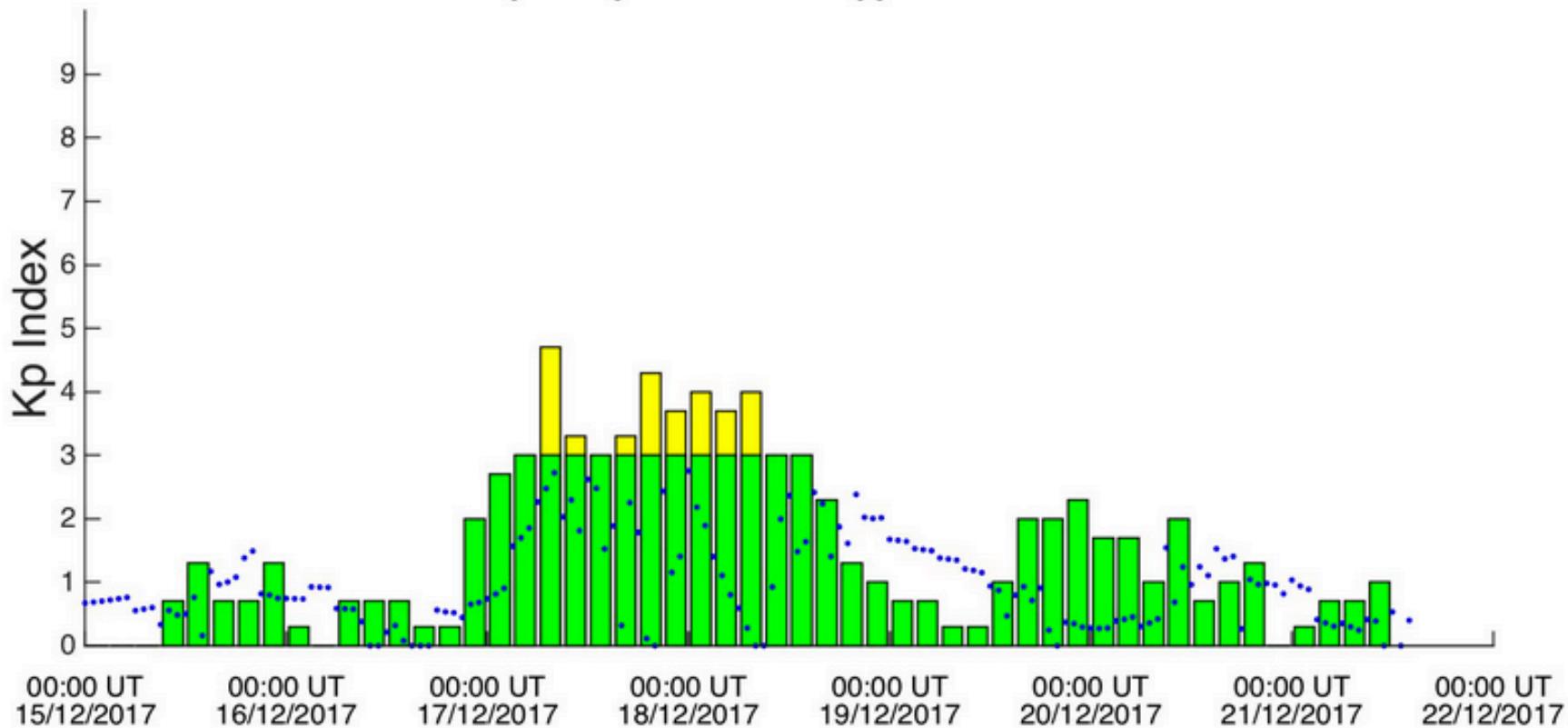
Methodology: NARMAX

[https://ssg.group.shef.ac.uk/USSW2/Kp/fKp\\_1d.jpg](https://ssg.group.shef.ac.uk/USSW2/Kp/fKp_1d.jpg)

# U. Shef Kp

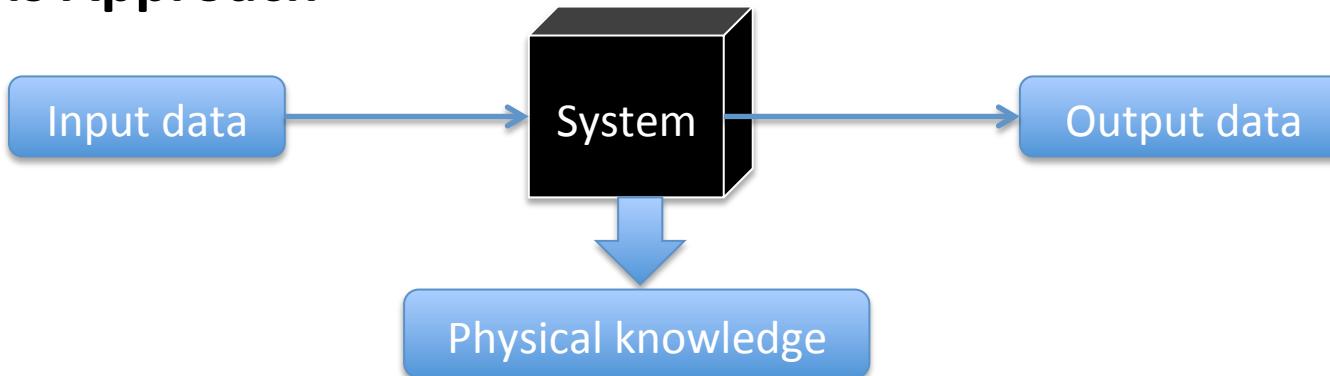
1 Week

Measured (bars) Forecast (.) since 15-Dec-2017



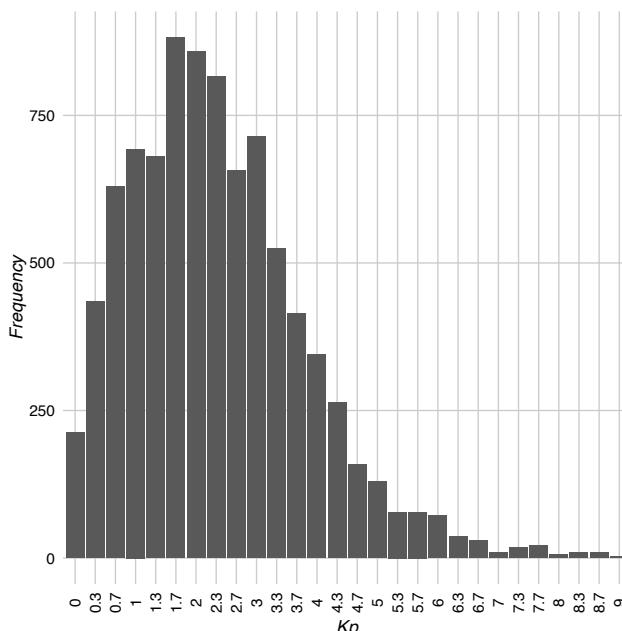
# Modeling Methodologies

## Systems Approach


$$y(k) = F[y(k-1), \dots, y(k-ny)], \quad \text{System outputs}$$
$$u(k), \dots u(k-nu), \quad \text{System inputs}$$
$$e(k-1), \dots e(k-ne)] \quad \text{Noise/errors}$$

$F[]$  is a nonlinear function (polynomial, B-spline, radial basis function)

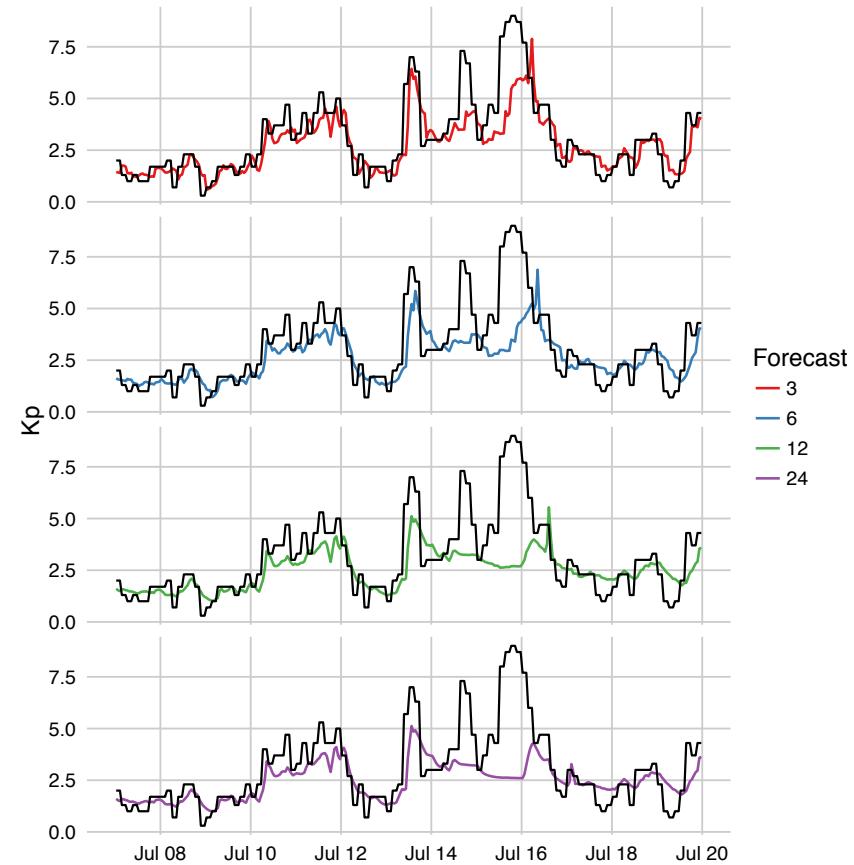
# Problem in training data



Kp values are not evenly distributed

Low to mid range values modeled OK

Forecasts of peak Kp values missed



# Model Parameters

## Input parameters

- Kp – GFZ Potsdam
- Solar wind – OMNI data set 5 min

Name	Description
Kp	Kp index
V	solar wind speed/velocity (flow speed) [km/s]
Bs	IMF southward component
P	Solar wind pressure [nPa]
N	Solar wind density [per cc]
VBs	Product of Velocity and Bs
$\sqrt{p}$	Square root pressure

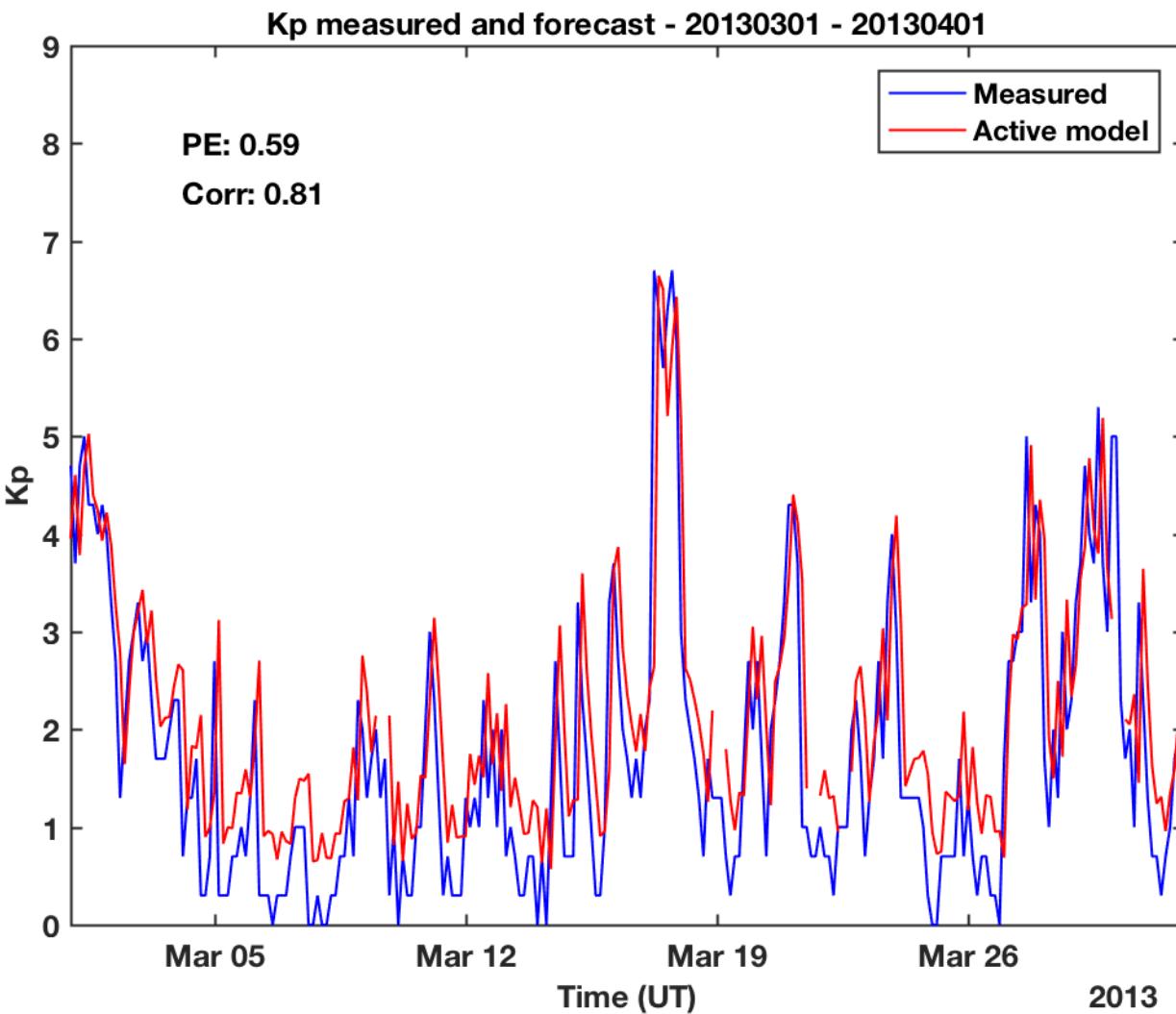
# U. Shef Kp model

## New model

- Target large Kp values
- Training dataset balanced to give more prominence to high values

```
kp= ...
8.8088e-1 * Kp(1) +
7.6502e-1 +
-4.9010e-03 * p(2)*n(2) +
-1.9820e-04 * V(2) * Kp(2) +
1.5981e-02 * p(2) * Kp(2) +
2.3706e-04 * V(1) * VBs(1) +
-3.7429e-03 * Bs(1) * n(1) +
-3.9727e-04 * V(2) * VBs(2) +
5.6176e-02 * Bs(2) * sqrt(p(1)) +
-7.1004e-03 * n(1) * VBs(2);
```

# St. Patrick's Day 2013



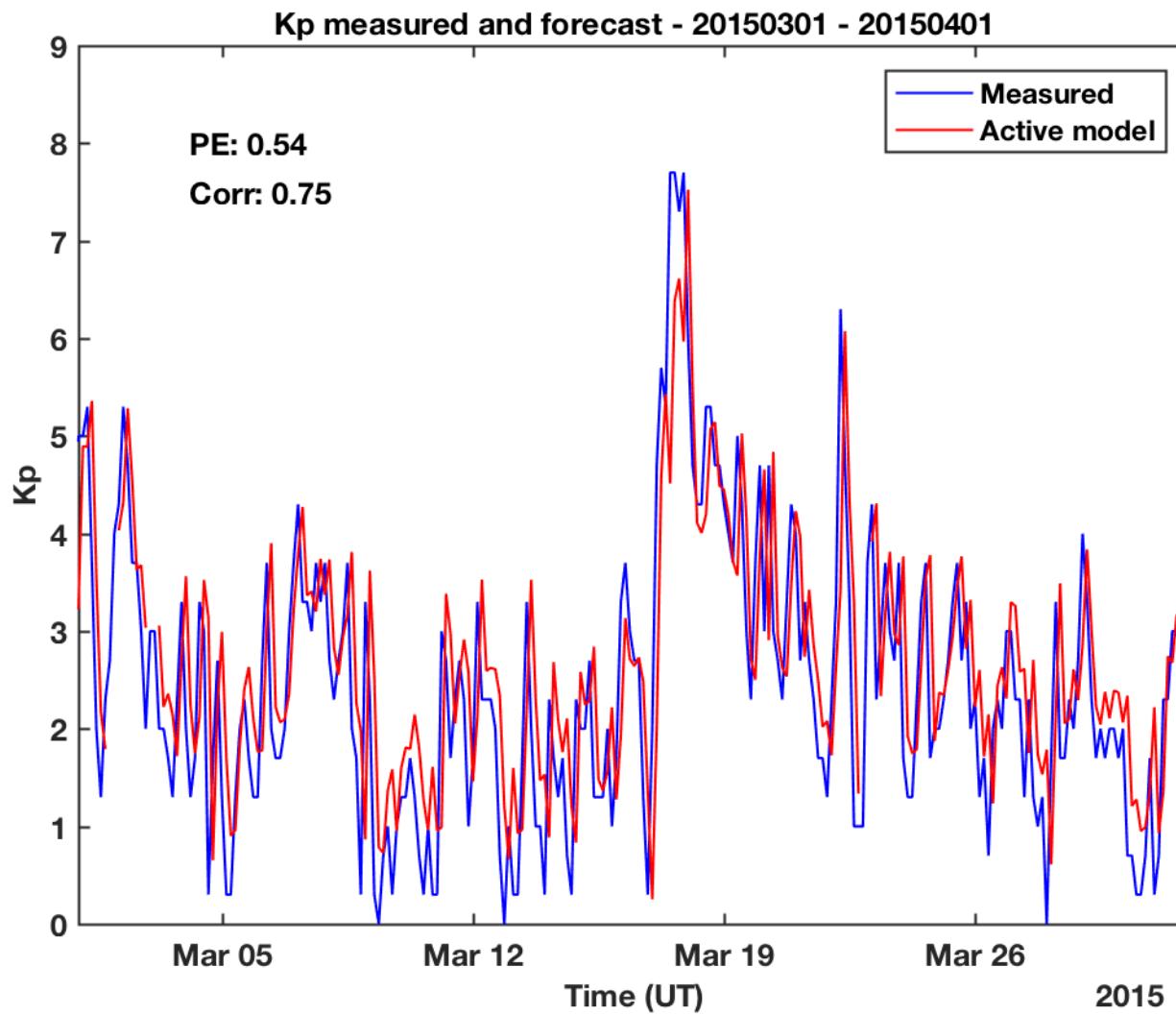
## Auto-regressive models

- Good agreement with measured values
- One time step delay

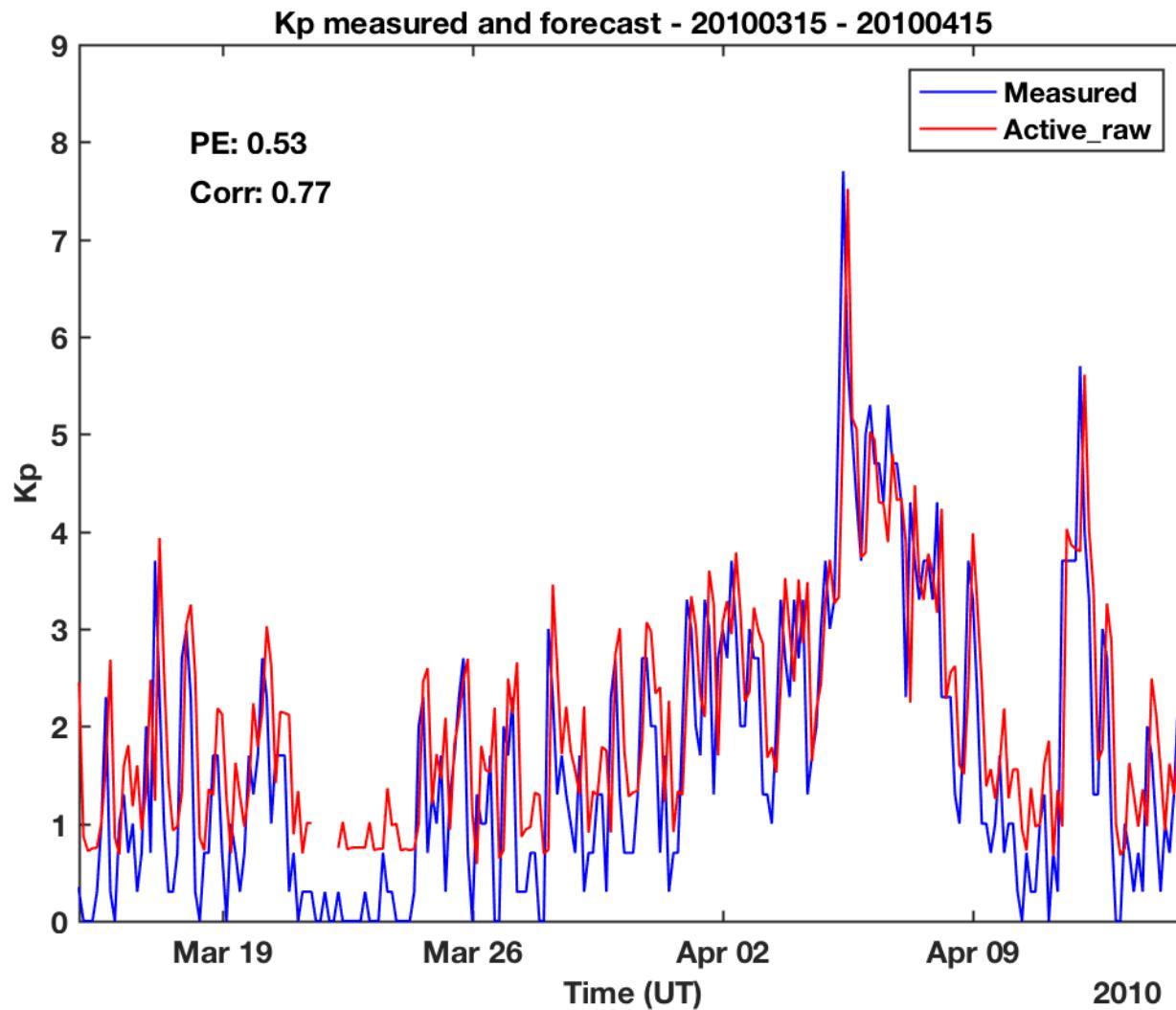
## Non-auto-regressive models

- Not particularly good agreement with measured values
- No time delays

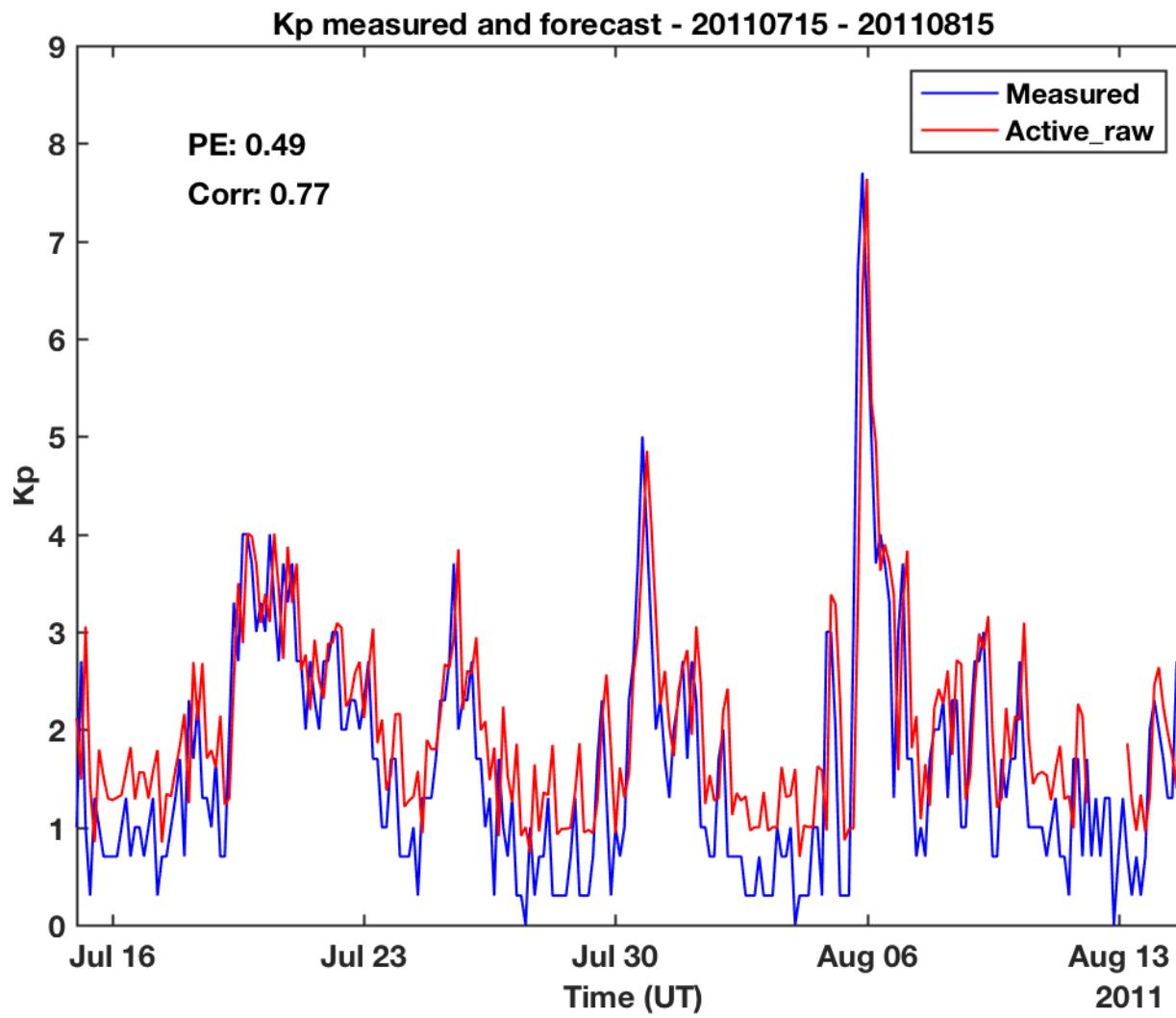
# St. Patrick's Day 2015



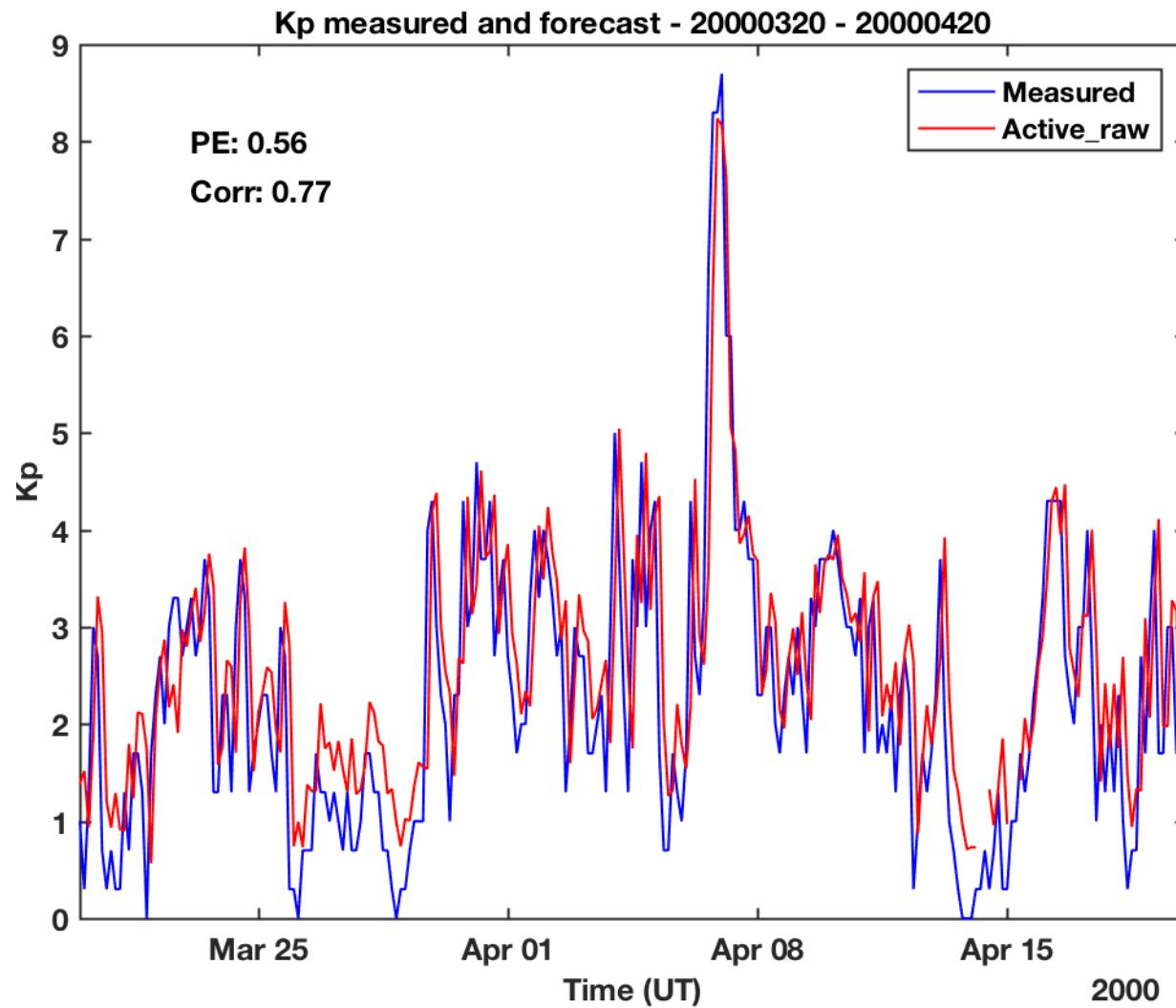
# April 2010



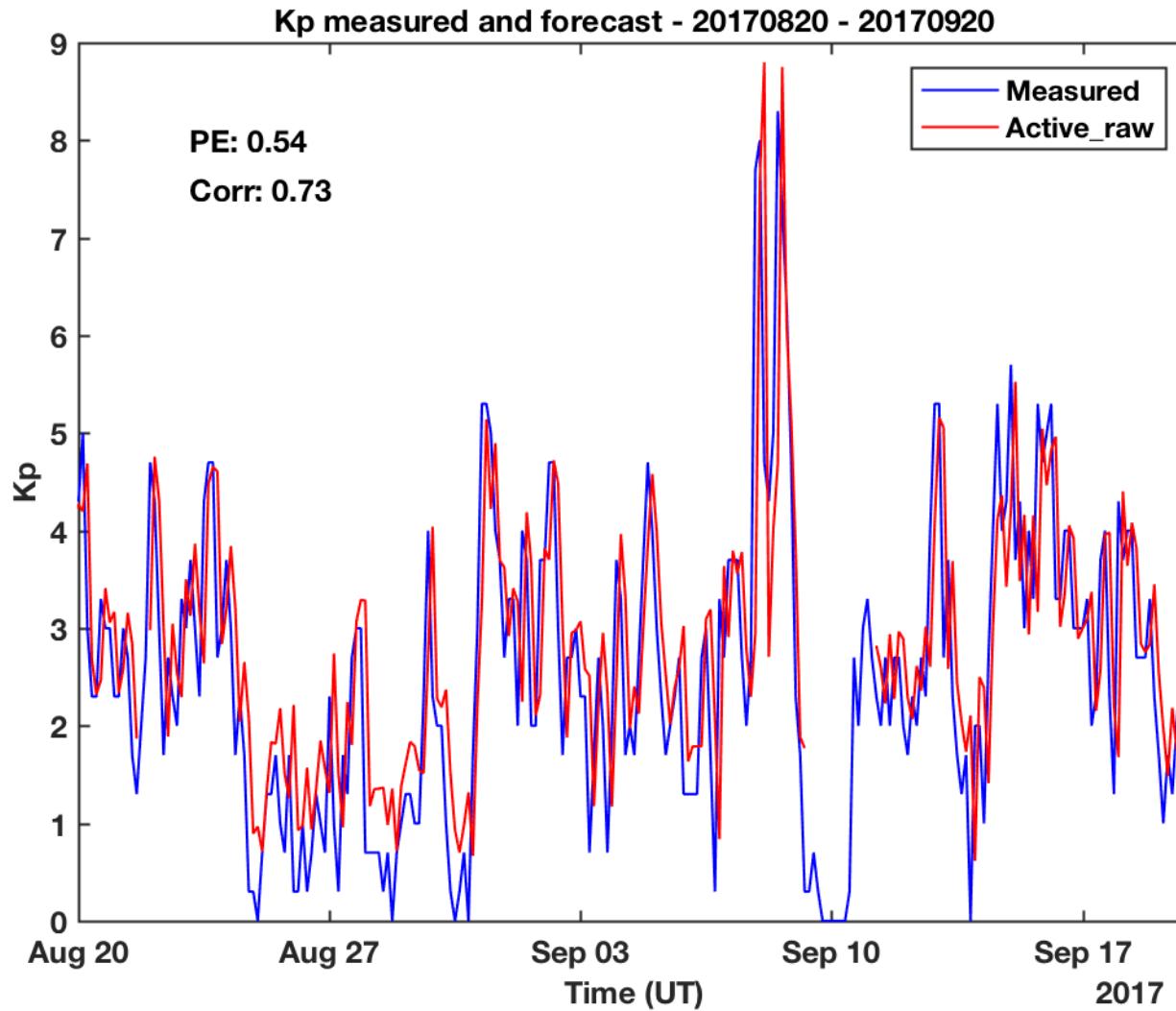
# August 2011



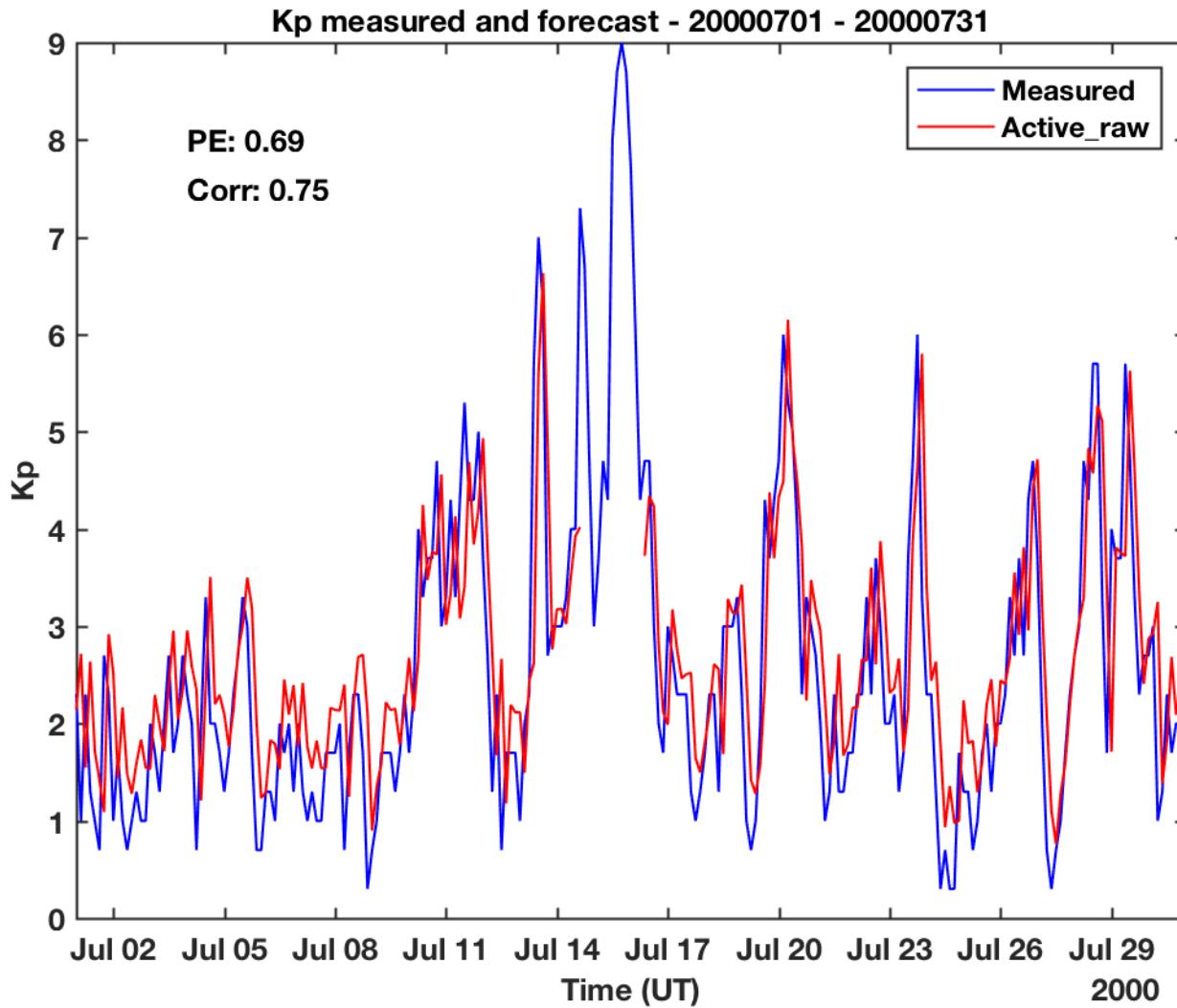
# April 2000



# September 2017

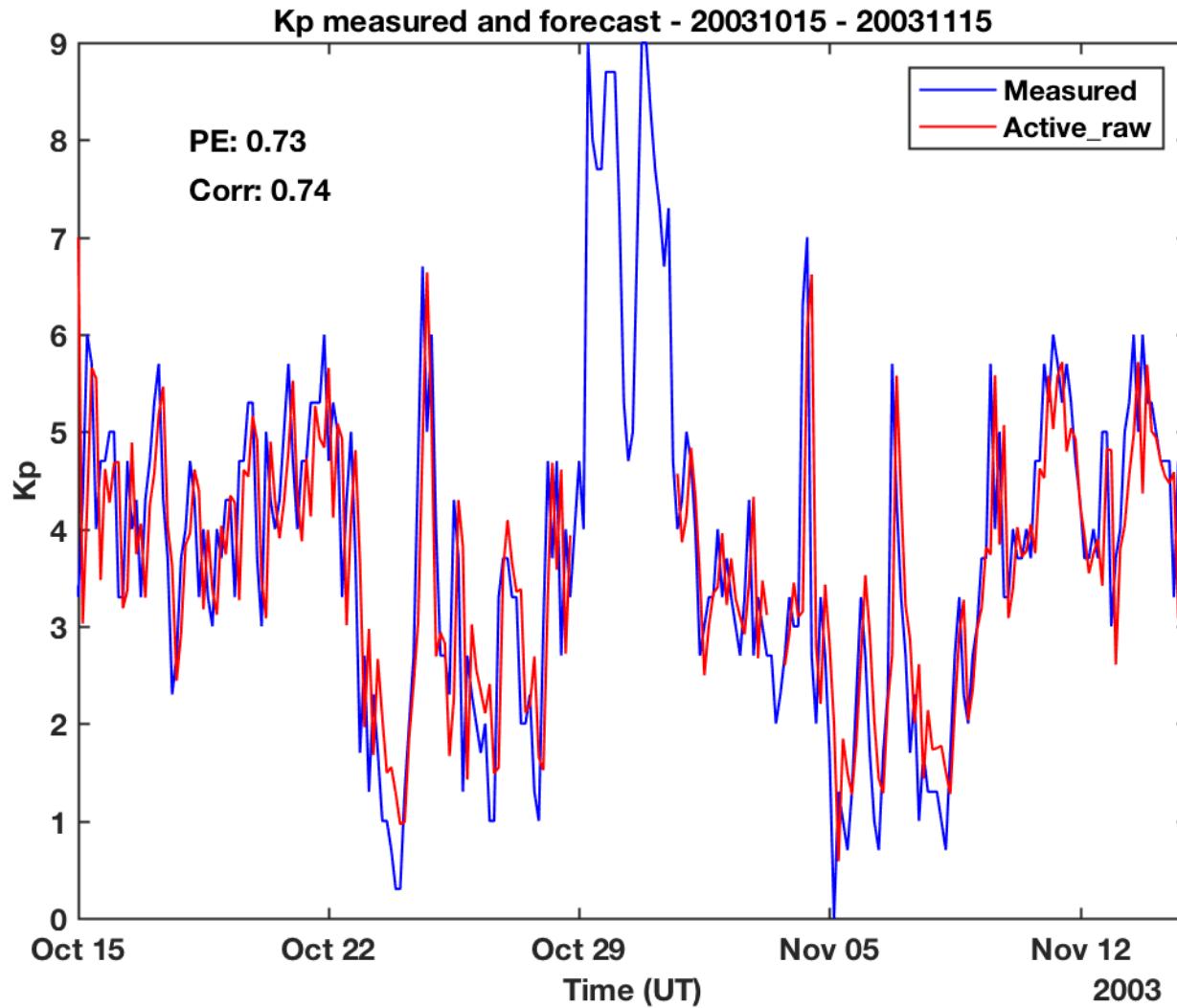


# Bastille Day Storm



Gap in OMNI plasma data, therefore no forecast

# Halloween Storms



# Summary

PROGRESS is a Horizon 2020 funded space weather project.

Main goals are to provide accurate and timely forecasts of

- Solar wind parameters at L1
- The evolution of geomagnetic indices
- The particle environment in the inner magnetosphere

Real time forecasts of interest to

- Scientists modeling these processes
- Satellite operators

All forecast information will be available from the project web site [ssg.group.shef.ac.uk/progress/html](http://ssg.group.shef.ac.uk/progress/html)

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