SEPARATION OF DIFFERENT IONOSPHERIC AND MAGNETOSPHERIC CONTRIBUTIONS TO THE GEOMAGNETIC FIELD VARIATIONS

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- Geomagnetically induced currents
- Geomagnetic field & its variations
 - Sq ("solar quiet")
 - Principal component analysis: short review
 - Sq: standard vs PCA
 - Summary
- Future work
 - master student grant



Geomagnetically induced currents



Geomagnetically induced currents

 Quasi-DC GIC flowing through grounded neutral terminals of power transformers located in the northern regions can reach 100 to 300 A. These currents result in deep saturation of magnetic cores, reactive power losses, AC odd and even harmonic generation, reduction of impedance and overheating of windings and magnetic cores





Geomagnetically induced currents

- Mostly affected are regions located at high geomagnetic latitudes: Canada (blackout 13.03.1989, Quebec), northern parts of USA, Scandinavian countries
- GIC-induced transformer failures were observed during intense geomagnetic storms at middle geomagnetic latitudes – South Africa
- Currently studies addressing the local GIC-hazard are done in South Africa, Australia, Kazakhstan, China, Japan, Brazil, Czech Republic, UK, Ireland, Italy and Spain



MAG-GIC

MAG-GIC

Correntes induzidas pelo campo geomagnético no território português

(concurso 02/SAICT/2017)

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A. Pais & F. Pinheiro, REN, 11/10/2018

MAG-GIC



Geomagnetic field: short introduction



Geomagnetic field





Geomagnetic fieldInternal geomagnetic field





Geomagnetic fieldExternal geomagnetic filed





Geomagnetic field variations



COI instruments



COI instruments



Variometer (left) and absolute (right) houses

Q

Geomagnetic field - time

COI geomagnetic field



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Geomagnetic field vs time





MAG-GIC: geomagnetic field variations

- Storm time H variations = relatively smooth daily background variations + fast H variations
- Earth crust conductivity is a f(depth)
- Earth crust = filter for different frequencies of H variations







Daily geomagnetic field variations

(Classification by Chapman, S. and Bartels, J. Geomagnetism, 1940)

- Solar (denoted as "S_{xx}")
- Lunar (denoted as "L_{**}")
- Disturbances (can be denoted as "Dst")



Daily geomagnetic field variations

"solar" = "diurnal"

- "solar quiet" (**Sq**):
 - Sq = μ (daily variations) for 5 most quiet days of a month

- "solar disturbed" (S_D)
 - $S_D = S Dst$
 - S = μ(daily variations) for all days of a month
 - Dst = μ (daily variations) for 5 most disturbed days of the month
 - S_D similar form to Dst





Sq ("solar quiet")

"solar quiet" (Sq)





Sq source: ionospheric currents (90-150 km)

differential solar heating => ionospheric plasma

Geomagnetic filed



Ionospheric Dynamo

the exact location of the centers of the vortices slightly change from day to day => the form of the Sq variations change from day to day



Sq estimation

- Standard procedure: mean daily variations of the 5 most quiet days (QD) of a month
 - not all QD are truly "quiet"
 - actual Sq = "true" Sq + part of S_D + ?
 - final QD definition available after 1-2 yrs
 - 1 curve for all days of a month (with different ionospheric conditions!)

PCAWavelets



Sq estimation

- Principal component analysis
 - no need to select specific days
 - real-time analysis
 - produces reconstruction of a selected mode for the whole analyzed time interval



Principal component analysis: short review



PCA

- Orthogonal transformation to convert a set of some input variables (possibly correlated) to a set of linearly non-correlated output variables (principal components - PC)
- All PC are ranked according to their eigenvalues
- Eigenvalues provide the variance fraction of the original data set variability that is explained by corresponding PC



PCA visualization

- Rotation in multi-dimensional data space:
 - the greatest variance is along the 1st coordinate

output from PCA

-2

0

pc1

2

- second greatest is along the 2nd coordinate
- etc.







- F data matrix (matrix size <u>n x p</u>)
- R = F^t F covariance matrix (size p x p)
- $RC = C\Lambda$

PCA math

- C matrix of eigenvectors (c_i) of R (size p x p)
- Λ diagonal matrix of eigenvalues of R (size $p \times p$)
- Eigenvectors (c_i) empirical orthogonal functions (EOF, vectors size p)
- Principal component: pc_i = F × c_i (vector size n)
- Mode_i pc_i × c_i (matrix size <u>n × p</u>)



PCA is based on eigenvector analysis

Simple

Easy to implement

- Ready to use functions, codes & packages in many programming languages & statistical software
- Used to reduce dimensionality, detect patterns etc.
- Starting point for some more elaborated tools of the data analysis:
 - canonical correlation analysis
 - factor analysis
 - correspondence analysis
 - K-means clustering



Sq: PCA vs standard



Data: all Januaries 2007-2014 (COI H - 25 000 nT)

- 2007 2014
- I h series
- interpolated gaps





Methods to calculate Sq

- 2. Principal Component Analysis (Sq_{PCA})
- Input matrix:
 - Rows: hours (1, ..., 24)
 - Columns: days (1 Jan Y1, ..., 31 Jan Y1, 1 Jan Y2, ...31 Jan Y2 etc.)
- Time interval:
 - all Januaries 2007-2014, together & separately



PCA: January - Variance fraction 0.65 January : 0.60 ---- 2007 ---- 2008 0.55 ---- 2009 0.50 ---- 2010 ---- 2011 0.45 ---- 2012 0.40 2013 ---- 2014 0.35 40% λ 0.30 0.45 0.25 0.40 0.20 0.35 0.15 67% 0.10 0.30 0.05 0.25 0.00 2 0 3 7 8 9 10 4 5 6 λ _{0.20} ## of a component 75% 0.15 83% -90% 0.10 95% 99% 0.05 0.00 2 3 5 0 4 6 7 8 9 10 11 12 13 14 15

of a component

ca IS















Sq_{PCA}vs Sq_{QD}: January PC2 & PC2 vs Sq_{QD}





Sq_{QD} vs PCs: correlation coefficients

	2007	2008	2009	2010	2011	2012	2013	2014	all
PC1	0.34	0.42 (0.12)	0.62	0.2	0.94 (<0.01)	0.46	Ο	0.59 (0.02)	0.4 (0.1)
PC2	0.77 (0.05)	0.47	0.55	0.9 (<0.01)	0.25	0.3	0	0.41	0.62 (0.08)
PC3	0.18	0.57 (0.06)	0.31	0.18	0.16	0.63 (0.02)	0.76 (0.02)	0.55 (0.1)	0.18
PC1 + PC2	0.76 (0.02)	0.5 (0.04)	0.71	0.63 (0.07)	0.94 (<0.01)	0.36	0.11	0.65 (0.01)	0.62 (<0.01)

Física

PCA vs QD: preliminary conclusion

- PCA is a good method to extract Sq-type variations
- In most cases (PC1 + PC2) explains 25-88% of the Sq_{QD} variations (r = 0.5 ÷ 0.94)
- For the geomagnetically disturbed time intervals PC₃ is needed





Do Sq_{QD} and Sq_{PCA} represent the "true" ionospheric component?



Ionospheric model vs Sq_{op}

-y2010 Sq COI CM5 Xionoprim + Xiono on 15.03.2010 12 15 18 21 24 ~ OK 40N 5 q, nT sph. January 2013. 12 15 18 21 24 h -v2009 Sq COI — CM5 X_{iono}prim + X_{iono}ind on 15.03.2009 Contamination by Dst 12 24 15-OK 10-10 38N 40N 5 d Sq, nl sph. 42N CM5 Xionoprim + Xiono ind on 15.03.2012 IO -v2008 Sq COI — CM5 X_{iono}prim + X_{iono}ind on 15.03.2008 12 15 18 21 12 10 -CM5. 38N -5 Not OK OK 40N 38N -10 -1040N 21 0 12 15 18 24 3 0 42N Sq, nT 42N CM5, -1021 12 12 15 18 24 0 h Sq COI CM5 Xionoprim + Xiono ind on 15.03.2007 -y2011 Sq COI 12 15 21 24 March, 2007 - 2012 15-Not OK 38N 10-OK 38N 40N 40N Sq, nT 42N Sq, nT 42N _5 -1021 3 12 15 18 24 0 9 12

CM5 is an empirical inverse model derived from pre-Swarm satellite data & ground observatory data from August 2000 to





Sq_{PCA} vs Sq_{QD} : Conclusions

- Sq_{QD} ~ Sq_{PCA}
- Both kind of Sq can be contaminated by the disturbance field
- PCA: the larger time interval, the better separation of the "quiet" and "disturbed" variations
- Ionospheric and magnetospheric models are needed
 - to "prove" that the obtained Sq-type variations are of ionospheric origin
 - to select of a PC that correspond to the "true" Sq
 (= ionospheric contribution)



Future work

- Different input time intervals
 - Different months/seasons
 - Different epoch of solar activity cycle
 - Short time intervals (around storms)
 VS

Long time intervals (several months)

 Ionospheric & magnetospheric models: modelled daily variations

VS Sq_{QD} & Sq_{PCA}



Future work
 + master student grant
 (to start in September 2019)
First-hand experience with

- ⇒PCA
- ⇒ Statistical analysis of observational data
- Models for ionosphere & magnetosphere

Control Con

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Questions?

