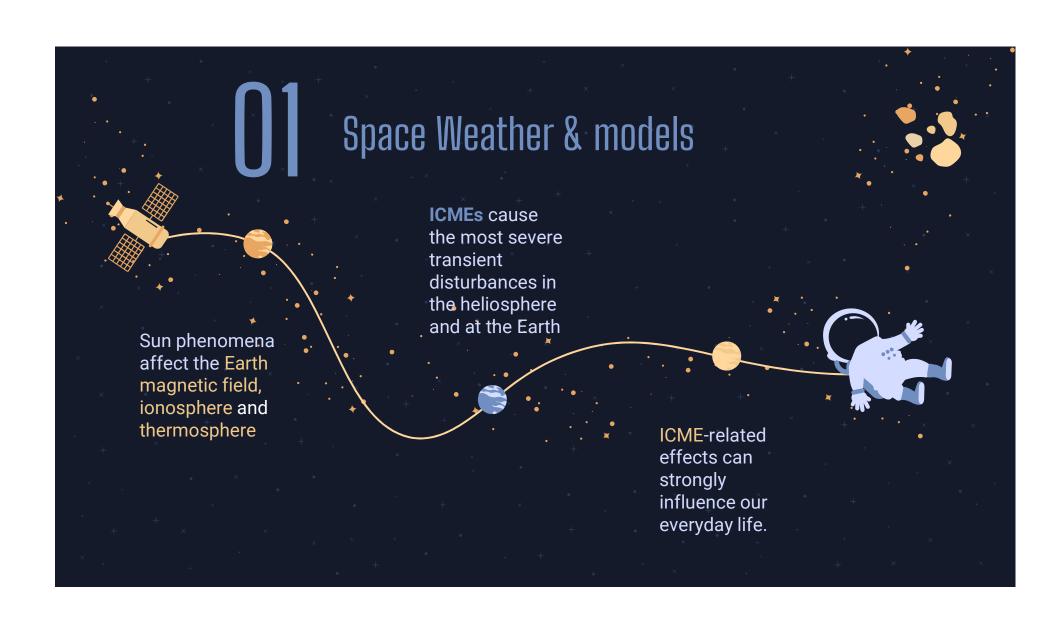
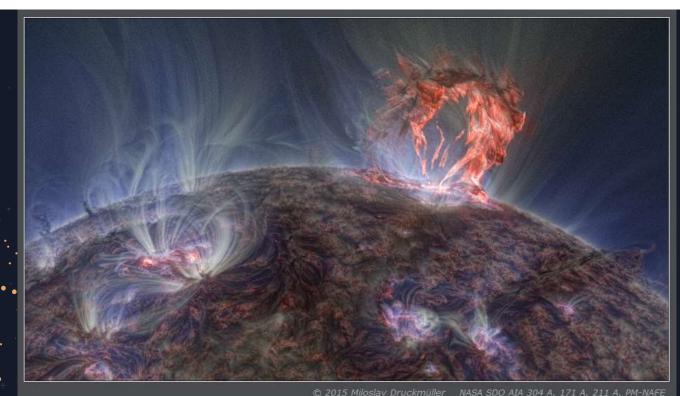




- SW and models
- SW events forecasting using Empirical Models
- **03** Examples
- 14 Conclusions & Useful Info





M1.7 flare, April 16, 2012

Modelling in Geophysics

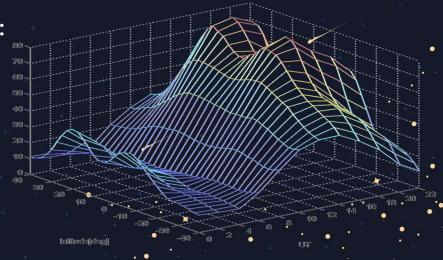
Models are invariably simpler than the real Earth phenomenon.

Any model is a mathematical expression, representing some real processes, that depends on the assumptions that are made.

Two main types of geophysical models:

Physics-based and

Empirical models.



Physics-based & Empirical models

Physics based models apply the laws of classical physics.

- They are deterministic.
- Explanation and prediction of natural phenomena are based on the mathematical representations of physical laws.

Empirical models are descriptive.

- They are based on data.
- They•do not rely on the use of physics.



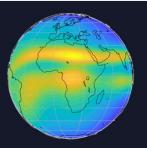
Neither the physics-based nor the empirical approach ignores the other.

Because:

Physicists rely on observations to develop and validate physical models and estimating key quantities like initial and boundary conditions.

The construction of statistical/empirical models is guided by physics that determines the variables and the data sets to be analyzed.

Empirical Models or profilers



- ✓ Based on a analytical description of the ionosphere with functions derived from experimental data.
- ✓ Model systematic ionospheric variation from historical data.
- ✓ Data sources are ionosondes, topside sounders, incoherent scatter radars, rockets and satellites.
- ✓ Mainly used for assessment and prediction purposes.
- ✓ Easy to use.
- ✓ Describe ionospheric climate.
- ✓ Give realistic representation of the ionosphere in the areas sufficiently covered by observations.

International Reference Ionosphere (IRI)

The IRI is an international project sponsored by the Committee on Space Research (COSPAR) and the International Union of Radio Science (URSI). These organizations formed a Working Group in the late sixties to produce an standard empirical model of the ionosphere, based on all available data sources. Several improved editions of the model have been released.

Input

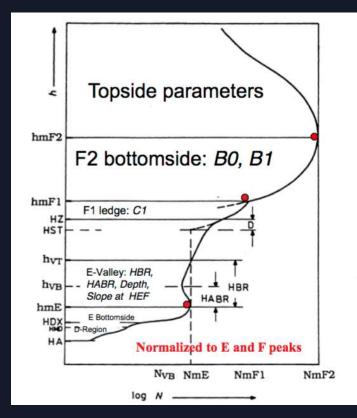
Year, month, day, hour, geographic or geomagnetic coordinates, various optional input.

Output

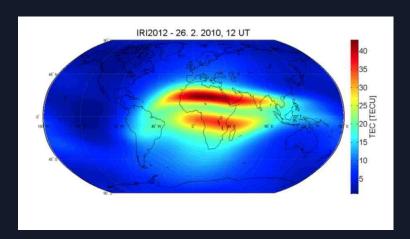
Electron concentration, electron temperature, ion temperature, ion composition (0⁺, H⁺, He⁺, NO⁺, O⁺₂), ion drift, ionopsheric electron content (TEC), F1 and spread-F probability

IRI Web http://irimodel.org/

International Reference Ionosphere (IRI)



Buildup of the IRI electron density profile and its separation into different regions.



From: "Mitigation of Ionospheric Threats to GNSS: an Appraisal of the Scientific and Technological Outputs of the TRANSMIT Project", Chapter 3, p. 166, InTech 2014

NeQuick

NeQuick is a 3D and time dependent ionospheric electron density model developed at the Abdus Salam International Centre for Theoretical Physics (ICTP), Trieste, Italy and at the University of Graz, Austria. It is a quick-run model particularly tailored for trans-ionospheric applications that allows to calculate the electron density at any given location in the ionosphere and thus the total electron content (TEC) along any ground-to-satellite ray-path by means of numerical integration.

Input

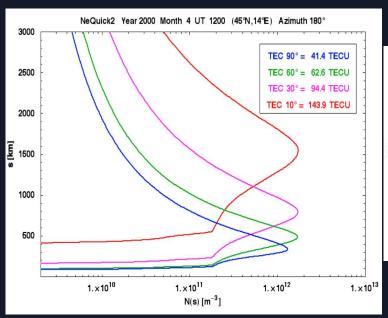
Year, month, day, time, geographic coordinates of lower and higher endpoint, R12 or daily F10.7 solar flux.

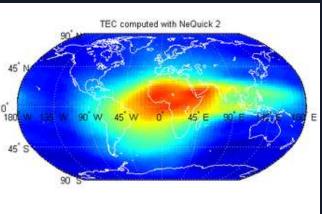
Output

electron density along the path and slant TEC

NeQuick2WEB – http://t-ict4d.ictp.it/nequick2/nequick-2-web-model

NeQuick







02

SW events forecasting using Empirical Models



The impact of Space Weather: from "climate" to "weather"

- Like the lower atmosphere the ionosphere exhibits both a "climate" and a "weather" variability. The ionospheric "climate" has been successfully represented by models of different types.
- The ionospheric weather variability is mostly controlled by the "Space Weather conditions".
- The big challenge of ionospheric modelling is to take into account the impact of varying Space Weather conditions to reproduce the observations.

Two approaches

Systemic approach: coupled physics-based models

Data
Assimilation or ingestion in models



We need to know that...

It is mandatory to have well established experimental databases that can be used to test and validate the existing models in order to generate the improvements needed.

No model is able to reproduce by itself in a satisfactory way both the "climate" and the "weather" of the Earth ionosphere.

Data Assimilation

DA techniques have successfully been applied by meteorologists to improve operational weather forecasts.

Such techniques have also been introduced into ionospheric research and application.

This was possible because of the increasing availability of experimental data even in real time (solar data, ionospheric ground and space-based GNSS data, ionosonde data and radar data, RO data). These models and schemes are of different complexity and rely on different kinds of data (GAIM, IDA3D, etc).





Data Ingestion



Minimization Algorithm | VTECexp – VTECmod |



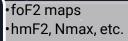
'effective' grids

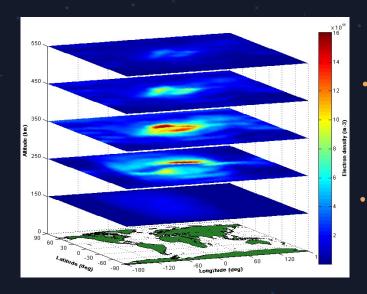
(IG, Az)

3D electron density of the ionosphere that reproduces the starting source data •

·Calculate TEC along

any ray path

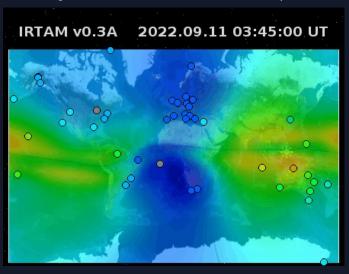




- Is one of the earliest and most simplistic approaches to data assimilation.
- The model states are directly replaced with the observations.

IRTAM

- IRI Real-Time Assimilative Modelling (IRTAM) system assimilates digisonde data from the Global Ionospheric Radio Observatory (GIRO) network into the IRI model.
- The IRTAM approach is based on the ITU-R models for the F2 peak plasma frequency foF2 and the propagation factor M(3000)F2 that are being used in IRI.
- IRTAM uses the CCIR set of functions to describe the global and spatial variation of the difference between the digisonde measurement and the IRI prediction of foF2.





The ICTP ingestion technique

Effective F10.7 (Az)
input values that
minimizes the difference
between an
experimental and the
corresponding
NeQuick2 modeled
TEC are calculated

Applying this concept to all vertical TEC values of a global experimental vertical TEC map a global grid map of Az is obtained

The Az grid is used as input for NeQuick2, providing a 3D global representation of the electron density of the ionosphere

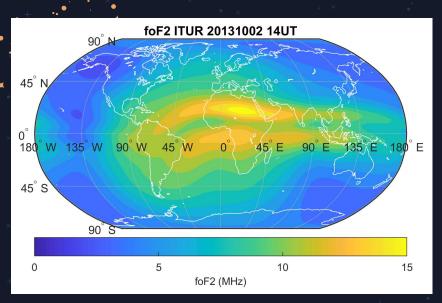
It can therefore be used to retrieve foF2 values where needed

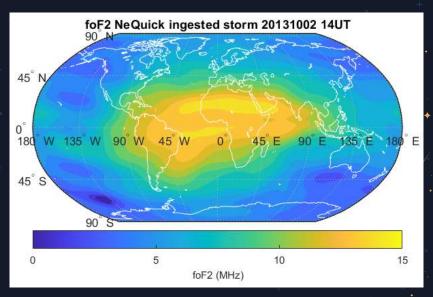
From: Nava, B., S. M. Radicella, and F. Azpilicueta (2011), Data ingestion into NeQuick 2, Radio Sci., 46, RS0D17, doi:10.1029/2010RS004635.).

03

Examples





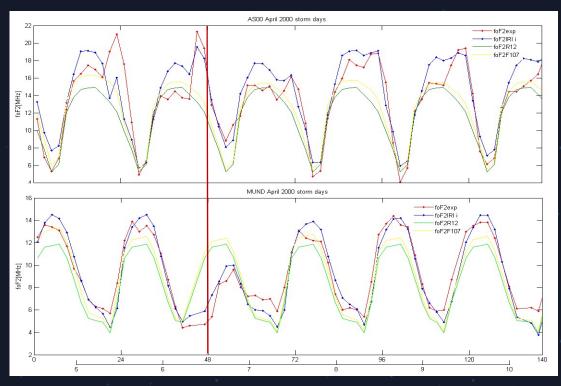


ITU-R foF2 map computed with daily F10.7 for 2nd October 2013 (left) and global map of foF2 for 2nd October 2013 obtained after the CODE GIM VTEC ingestion into the NeQuick model (right).



GNSS derived TEC ingestion into IRI

Migoya-Orue et al., 2015, Adv. Space Research

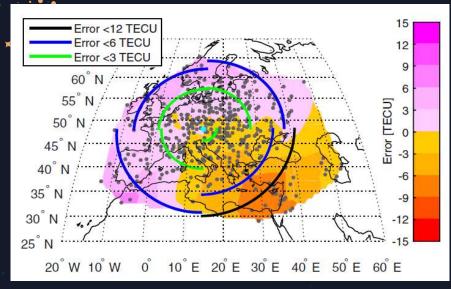


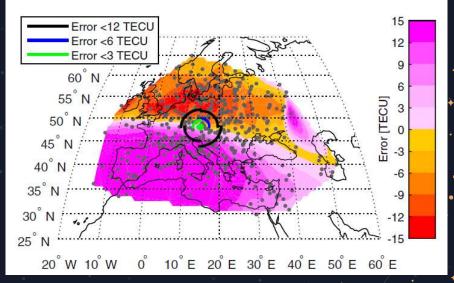
foF2 comparison during storm days April 2000



Locally Adapted NeQuick

Vuković and Kos, 2017, Adv. Space Research



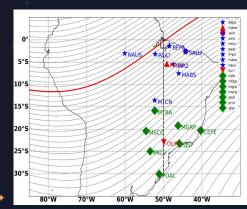


Locally adapted NeQuick maps performance 3 days before (left) and during St. Patrick Storm 2015 (right).

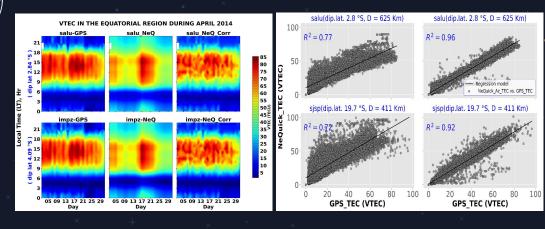
Locally Adapted NeQuick

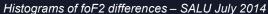


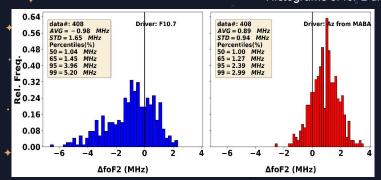
Osanyin et al., in preparation

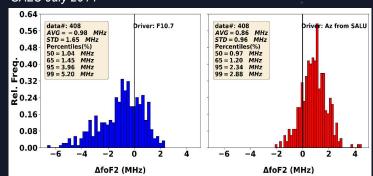


GPS and ionosonde stations used in the study



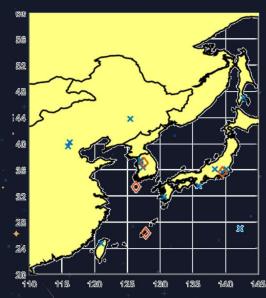






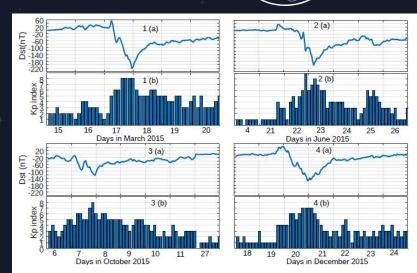
Data Assimilation into NeQuick through KF technique



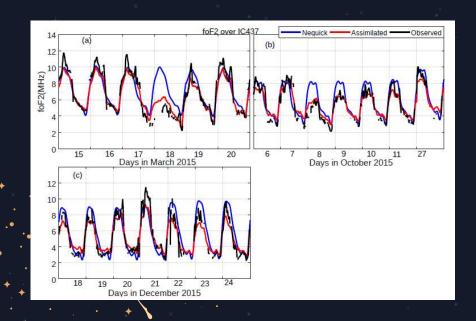


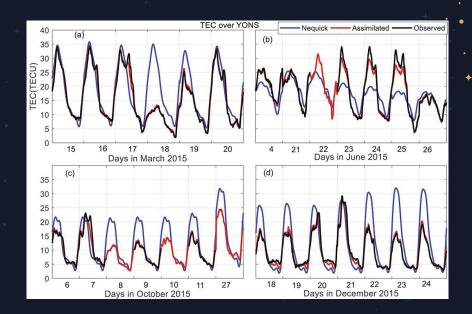
GPS and ionosonde stations used in the study

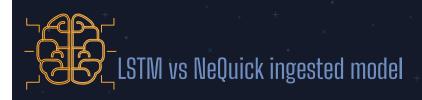
Mungufeni et al., J. Space Weather Space Clim. 2022



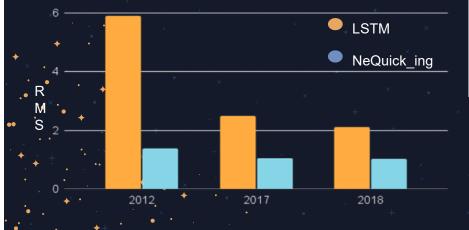
TEC Assimilation into NeQuick through Kalman filtering technique





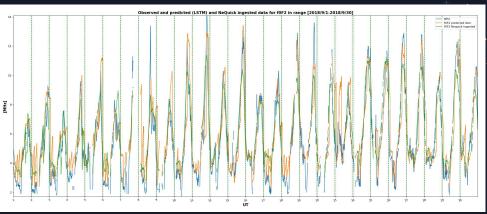


Long-Short Term Memory are a special kind of RNN, capable of learning long-term dependencies.





Namour et al, in preparation



Observed vs LSTM – NeQuick/ing models (Tucumán, September 2018)

- 1 hidden layer, 5 neurons
- Training Data range: 01/01/2017 to 30/08/2017
- Missing data: % 16.1
- Data Split: %70 training (%20 validation), %30 test

04

Conclusions & Useful Info

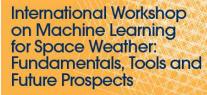
CONCLUSIONS

- The assimilation/ingestion of ionospheric data into empirical models allows to provide global and regional 3D specification of the electron density of the ionosphere and is able to improve the challenge that represents the reproduction of the 'weather' variability of ionospheric parameters during SW events.
- It has been showed examples of the prediction of TEC, Ne and foF2 with NeQuick and IRI by ingesting different ionospheric data series and some comparisons with experimental values and other models.

CONCLUSIONS

- The models performance show a dependence on geomagnetic activity with RMS errors increasing with increasing geomagnetic activity.
- The models performance show also a complex dependence with latitude.
- The models faced a real challenge in their ability to forecast and nowcast local and global ionospheric effects of Space Weather events.
- Statistical and ML techniques application to SW have received a significant boost in the recent years.

ICTP International Workshop on ML for SW



7-11 November 2022 This is a hybrid meeting Buenos Aires, Argentina

Local Organiser: M.G. MOUNA, FACET-UNT / CONICET

ICTP Scientific Contact:

Directors:

This workshop aims to fester Space Weather research through the application of Machine Learning (ML) and adstitical techniques by providing the participants with theoretical and practical training on Space Weather and Machine Learning fundamentals, with handson tutorials.

Description:

becompared and highly coughed Sun-Earth system is constantly, being performed by gound and apose-based instrumentation, which produces a huge amount of daily data. These datasets, in addition the increasing computing capability, are regularly used to produce forecasting models and offiser Space Weather products. In particular, Space Weather data analysis and modeling using filt. Techniques are

The purpose of the workshop is to give theoretical and tailored practical training on Machine Learning fundamentals, is application to Space Weather and future prospects, covering also important topics like Research to Operations (#20), explainable Artificial Intelligence (XAI) and trustworthiness and efficies.

Topics:

- Space Weather fundamentals.
 Space Weather Gaps and applications that can be tackled with Machine Learning.
 Machine Learning Basic Concepts and Tools.

- Macanne searing issue Concepts and toos.
 Deep Learning and current fire-notified to Space Weather and their main challenging the chiques applied to Space Weather and their main challenge. XAI, tustworthines and ethics in ML.
 Open source tools for ML (Python, sckill-learn, Keras, etc.).

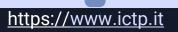
How to apply:



Deadlines: 4 September 2022 30 September 2022







Thanks!

yenca@ictp.it





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