





Modelling of Ionospheric Perturbation due to variation in Interplanetary Magnetic Field

Dr. Rajat Acharya

Indian Space Research Organization (ISRO)

10.12.19 ICG-14, Bengaluru



Introduction:

- Space Weather affect the ionospheric plasma density
- Variations in interplanetary magnetic field of the solar wind, interacts with the magnetosphere of the earth
- Electric field induced readily gets penetrated to the equatorial region and causes variations in the transport component of the equatorial ionospheric plasma
- The consequent TEC variation, in turn, affect satellite navigation in this region



Observation and prior research

- There exist a definite systematic relationship between the IMF variation and additional induced electric field
- A model can be established between the causal parameters and the effect

Aim

- Study the effects of solar wind and IMF on ionosphere under normal and enhanced solar activity events
- Establish a relation between the IMF and the occurrences of measurable ionospheric TEC variation





Prompt Penetration Electric Field:

- Abrupt and rapid inversion in IMF Bz
- Associated IEF Ey is derived
- Over shielding and under shielding process occurs
- Excess electric field penetrates from the poles to the equatorial region
 - Excess zonal field drives excess plasma up
 - Uplifted excess plasma adds to excess TEC











Data and pre processing:

- Data obtained from the **OMNIWeb** repository of SPDF https://spdf.gsfc.nasa.gov/pub/data/omni/hig h res omni/
- data at 5 minutes interval used
- $-dE_v/dt$ data used as input
- Variation in dB₂/dt >12 nT/5min considered as event
- $-dE_v/dt$ values where the gradient is less is clipped to zero
- A temporal array of 1hour is used





Neural Network Architecture:

- Neural Network of 1 hidden layer of 50 neurons designed
- Sigmoid non-linearity used
- Back propagation used for learning /weight adjustment
- Least mean error used as reference





Approach and Training:

- Model is attempted for data only for the daytime
- dBz/dt, sub solar location, local time are taken as input
- Network trained for d $\Delta TEC/dt$ output
- LME based training with back propagation used
- Training done with >5000 samples
- Training r = 0.95 obtained
- Very good fit was obtained during training





Validation Results:

- Data for date used as input
- Testing done with >3000 samplers
- r = 0.84 obtained during validation
- Fair accuracy observed in temporal variation







Results:

- $d \Delta TEC/dt$ compared with actual measurements on data of an arbitrary storm day
- d Δ TEC/dt integrated to obtain the variation of excess TEC, i.e. Δ TEC with time of day
- Arbitrary threshold of 10 TEC taken as alert level
- Alert generated when the derived excess TEC exceeded 10 TEC
- When tested with many other days, there were considerable amount of false alarm and missed event
- This needs improvement



Essence of Work done:

- SW variable dBz/dt is used as the driving parameter for equatorial TEC variations
- Neural network based model developed between dEy/dt and d Δ TEC/dt
- Storm data from more than 4 years (22 storm events; 8000 samples) used
- Excess TEC (Δ TEC) is obtained from derived values
- Alert generated for excess of 10 TEC over quiet day variation
- IMF at L1 point takes about an hour more to impact on the magnetosphere and cause the ionospheric variations
- The measurements data done at the L1 point can be transmitted to earth in few seconds
- Therefore, about an hour of lead time can be obtained from the impact predicted from the derivation of this data and the actual impact



Final Outcome:

- Model to identify the ionospheric perturbation due to IMF variation
- Can alert the user in advance and add to integrity for single frequency GNSS receivers
- Provide further insight into solar and heliospheric influence on ionosphere

Limitations and Issues

- Need to obtain PP event which are isolated from any DDEF
- Isolated events are sparsely occurring: Hard to distinguish penetration from neutral wind effect

Future plans

- Improve the obtained model with other sophisticated training algorithms
- Add more auxiliary data which factors the variation to get better accuracy















