# Surface Charging: Validation Challenges and Suggestions

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## **Surface Charging Challenges**

- Surface charging physics is a current balance phenomenon
- Surface charging threat environments

Environment	Extreme Potential
LEO auroral	-1 to -2 kV
LEO solar array	~90% of string voltage
GEO (GTO)	-10 kV to -20 kV

- Validating that a space weather model and charging code can be used to predict surface charging requires validating two models:
  - Does the space weather model correctly generate the relevant environments?
  - Does a spacecraft charging code generate the correct surface potential (or electric field) using space weather environment model inputs?

$$\begin{split} \frac{dQ}{dt} = & C \frac{dV}{dt} = \frac{d\sigma}{dt} A = \sum_k I_k \\ \frac{dQ}{dt} = & \sum_k I_k = \\ & + I_i(V) & \text{incident ions} \\ & - I_e(V) & \text{incident electrons} \\ & + I_{bs,e}(V) & \text{backscattered electrons} \\ & \pm I_c(V) & \text{conduction currents} \\ & + I_{se}(V) & \text{secondary electrons due to I}_e \\ & + I_{si}(V) & \text{secondary electrons due to I}_i \\ & + I_{ph,e}(V) & \text{photoelectrons} \end{split}$$

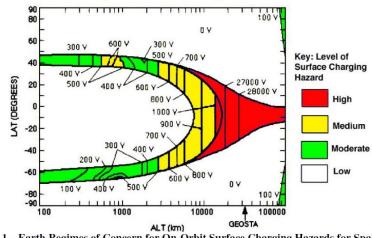
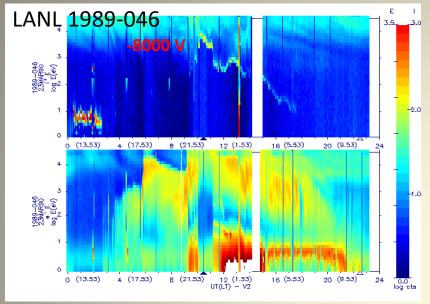


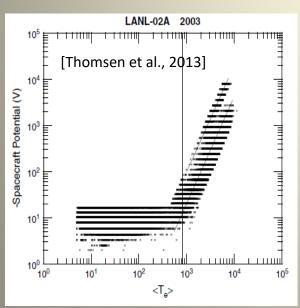
Figure 1—Earth Regimes of Concern for On-Orbit Surface Charging Hazards for Spacecraft Passing Through Indicated Latitude and Altitude (Evans and others (1989))

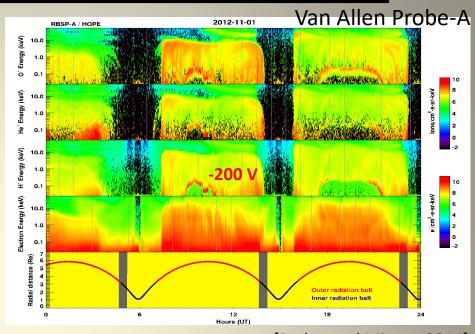


# **GEO** (Outer Radiation Belt) Charging



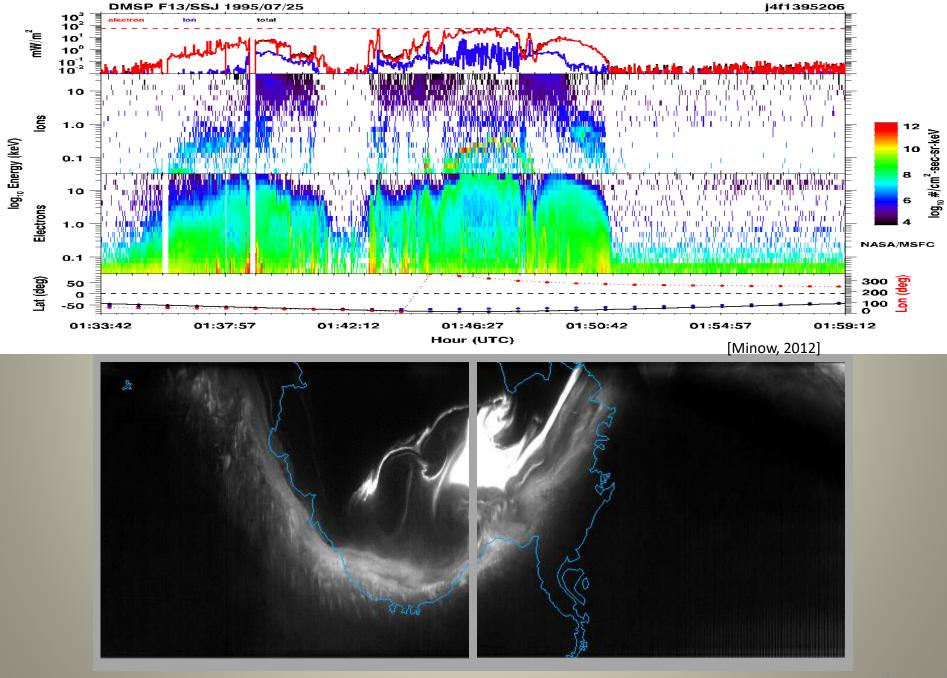
[https://www.mpa.lanl.gov]





[Parker and Minow, 2014]

- GEO charging is driven by the hot thermal plasma during storm environments
- Space weather models need to predict the background
- Charging code inputs are typically Ne, Te parameters for a single or bi-maxwellian fits to electron and ion flux





### Fontheim Distribution

#### **Ambient background**

n=1.0e10 1/m3

Te=0.2 eV

Maxwellian

Jmax = 4.0e-6 A/m<sup>2</sup>

Te = 3.0e3 eV

Gaussian (beam)

 $Jgau = 0.9e-4 \qquad A/m^2$ 

Egau = 10.0e3 eV beam energy

dgau = 4.0e3 eV beam width

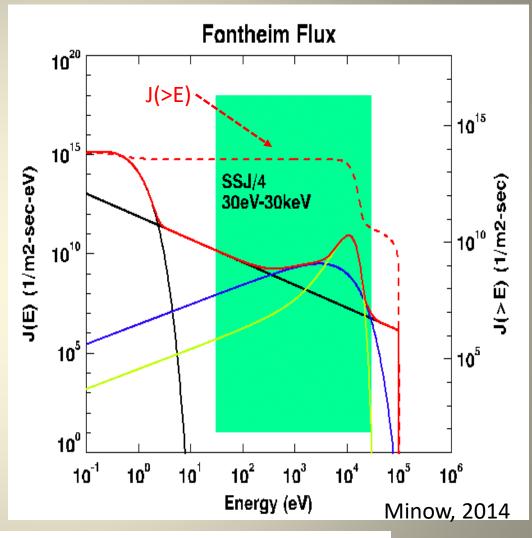
**Power Law** 

 $Jpwr = 3.0e-7 \qquad A/m^2$ 

alpha = 1.15 exponent

E1=50.0 eV, first energy

E2=1.0e5 eV, second energy



$$Flux\left(E\right) = \sqrt{\frac{e}{2\pi\theta m_{_{e}}}} \frac{E}{\theta} n \exp\left(-\frac{E}{\theta}\right) + \pi \zeta_{max} E \exp\left(-\frac{E}{\theta_{max}}\right) + \pi \zeta_{gauss} E \exp\left(-\left(\frac{E_{gauss} - E}{\Delta}\right)^{2}\right) + \pi \zeta_{power} E^{-\alpha}$$



## Surface Charging Suggestions\*

- Initial effort will focus on high priority GEO, MEO, GTO, and LEO polar environments where surface charging can exceed hundreds of volts
- User groups include spacecraft designers, operational situational awareness, anomaly investigations, and impact on science measurements
- Metrics (team is evaluating options):
  - Statistical evaluation using O'Brien "green anomalies" technique
  - Parameters used for inputs to charging models
    - GEO, MEO, GTO: Ne, Te, Ni, Ti or other
    - LEO polar (auroral): Ne,  $E_{beam}$ ,  $\Delta E_{beam}$ , and other Fontheim parameters
    - Flux spectra at different locations
- Environment models (initial focus):
  - Ovation CCMC implementation
  - LANL model (Vania Jordanova)
  - IMPTAM (Natalia Ganjushkina)), run online in near-real time since 2013
  - CIMI (Natalia Buzulukova)
- Spacecraft charging models (secondary effort, but compare with  $\Phi_{\rm s/c}$ )
  - Nascap
  - SPIS
  - SPENVIS, MUSCAT, and other small group charging codes

<sup>\*</sup>Output from the surface charging sessions at the ICCMC-LWS, Cape Canaveral, FL 3-7 April 2017



## Metrics (I): Surface Charging

- The traditional ones
  - RMSE root mean square error
  - Ratio of maximum amplitude(flux)
  - Prediction efficiency
- Above a certain threshold (yes/no prediction)
  - >10 keV flux exceeds 1.5e7 (1/cm<sup>2</sup>/sec/ster) (surface charging threshold)
  - Whether a model prediction predicts a surface charging event during a certain interval with the defined threshold
  - Generation of a contingency table of hit (H), miss (M), false positive (F), and correct negative (N) model predictions
  - Heidke Skill Score (for a large enough surface charging events)



## Metrics (II): Surface Charging

New ones using log ratio (Morley et al., 2018) satisfying the four attributes (see below)

The median symmetric accuracy

$$\varsigma = 100(\exp(M(|log_e(Q_i)|)) - 1)$$

where  $Q_i = \frac{y_i}{x_i}$ : ratio of predicted versus observed,  $y_i$ : model,  $x_i$ : observation

M: median value

The Symmetric Signed Percentage Bias (SSPB):

$$\mathsf{SSPB=}\ 100\ sgn\ \Big(M\left(log_e(Q_i)\right)\Big)\left(\exp\bigl(\left|M\bigl(log_e(Q_i)\bigr)\right|\right)-1)$$

Sgn: signum function; M: median value

- (1) The metrics must be meaningful for data that cover orders of magnitude,
- (2) underprediction and overprediction by the same factor should be penalized equally,
- (3) the metrics should be easy to interpret, and
- (4) the metrics should be robust to the presence of outliers and bad data.

