Colorado Center for Astrodynamics Research

University of Colorado Boulder, Colorado

Adaptive technique for high-latitude conductivity covariance refinement

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Overview

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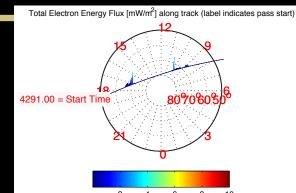
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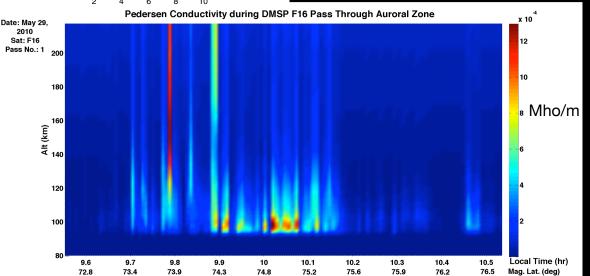
Outline Conductivity calculations

Methodology

Discussion of improvements moving forward

 What we hope to accomplish Devise self-consistent scheme to give





conductivity gradients from satellite observations

Enable 3D studies of the high-latitude ionosphere

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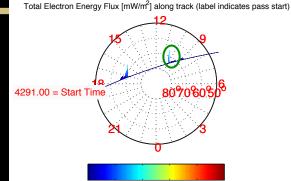
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Outline Conductivity calculations

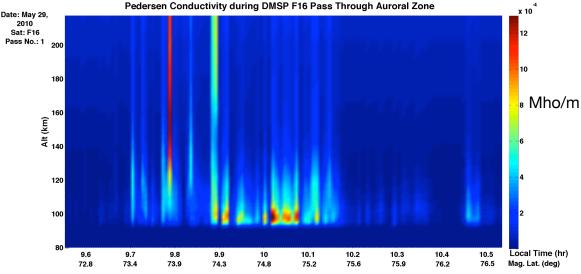
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GLobal AirglOW

al. [1988])

model (Solomon et

Calculating Conductivity

Ionization sources:

- DMSP particle precipitation poleward of 45° mag. lat., separated by hemisphere
- EUVAC model of solar irradiance

Transport and chemistry calculations:

- Two-stream electron transport code (Banks and Nagy [1970], Nagy and Banks [1970], Banks et al. [1974])
- Elastic collisions with O, N₂, and O₂; Inelastic collisions leading to excitation and ionization
- Energy redistribution in 190 bin energy grid (0.25 eV – 49 keV)

$$\sigma_{P} = \frac{q_{e}}{B} \left[N_{O^{+}} \frac{r_{O^{+}}}{1 + r_{O^{+}}^{2}} + N_{O_{2}^{+}} \frac{r_{O_{2}^{+}}}{1 + r_{O_{2}^{+}}^{2}} + N_{NO} \frac{r_{NO}}{1 + r_{NO}^{2}} + N_{e} \frac{r_{e}}{1 + r_{e}^{2}} \right]$$
$$\sigma_{H} = \frac{q_{e}}{B} \left[-N_{O^{+}} \frac{1}{1 + r_{O^{+}}^{2}} - N_{O_{2}^{+}} \frac{1}{1 + r_{O_{2}^{+}}^{2}} - N_{NO} \frac{1}{1 + r_{NO}^{2}} + N_{e} \frac{1}{1 + r_{e}^{2}} \right]$$

Methodology

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Following methodology has been applied to electric fields by *Matsuo et al.* [2002, 2005] and *Cousins et al.* [2013a, 2013b]

Extensive set of observations:

- -> Mean field yields background conductivity field
 - -> EOFs give variation from background field

EOFs used as bases whose coefficients are updated with new observations; background covariance matrix built from these

Background covariance parameterized, estimated at each analysis time step

Parameters estimated using maximum likelihood method in innovation vector space by matching innovation vector to covariance model at each analysis step

$$\mathbf{x}_b = \bar{\Sigma}$$

 $\mathbf{P}_b = \Psi E \left[\alpha \cdot \alpha^T \right] \Psi^T$ $\alpha = \text{EOF coefficients}$

$$\begin{aligned} \mathbf{P}_b &\simeq \mathbf{P}_b(\zeta) \\ \mathrm{diag}(\mathbf{P}_b) &\simeq \zeta_{b1} v^{-\zeta_{b2}} \\ v &= \mathrm{no.~of~bases} \\ & \text{In method of Dee [1994]} \end{aligned}$$

Methodology

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In method of Dee [1994]

New application: Whole process applied at discrete levels of altitude for conductivity

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Address shortcomings in ionospheric modeling (namely 2-D and Maxwellian distribution assumptions)

Outstanding Issues

Already done

Global picture of height-dependent conductivity

Sparsity of observations

Constrained to DMSP paths

Conductivity is not directly observed

Stability of estimation process

- Uniqueness
- Identifiability of parameters

Outstanding issues

Completed

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AMIE – relationship among electrodynamic variables

AMIE NextGen

Inverse procedure to infer maps of

$$\vec{E}, \Phi, \vec{I}_{\perp}, \vec{J}_{\parallel}, \Delta \vec{B}$$

From observations of

 $\vec{E} \\ \vec{I}_{\perp} \\ \vec{J}_{\parallel} \\ \Delta \vec{B}$

IS or HF radar, Satellites

IS radar

Satellite or ground-based magnetometers

<u>linear relationship</u> (for a given $\underline{\sum}$)

$$\begin{split} F(\vec{E}) &= \Phi, \vec{I}_{\perp}, \vec{J}_{\parallel}, \Delta \vec{B} \\ \vec{E} &= -\nabla \Phi \\ \vec{I} &= \underline{\Sigma} \cdot \vec{E} \\ \vec{J}_{\parallel} &= \nabla \cdot \vec{I}_{\perp} \\ \vec{I}_{\perp}, \vec{J}_{\parallel} &\leftrightarrow \Delta \vec{B} _{\text{Biot-Savart's law}} \end{split}$$

Inversion Problem in Condutivity McGranaghan: CEDAR 2014

Slide source: Tomoko Matsuo



Self-consistent procedure to determine conductivity profiles in high-latitude ionosphere

Concluding Remarks

Provides better starting point for electrodynamic calculations

Addressing shortcomings in ionospheric modeling (namely 2-D and Maxwellian distribution assumptions)

Open questions:

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- Global coverage
- Limited observations
- Performance of the inversion procedure