DATA ASSIMILATION OF RADIATION BELT ELECTRONS USING MULTI-POINT OBSERVATIONS AND THE VERB CODE: LESSONS LEARNED AND FUTURE DEVELOPMENTS ADAM C KELLERMAN, Y. Y. SHPRITS, D. A. KONDRASHOV, T. PODLADCHIKOVA, A. Y. DROZDOV, H. ZHU

Kalman Filter and Errors

Forecast

Update

- State vector, PSD $(c/(cm.MeV))^3$ X:
- Model matrix (VERB code) M:
- **P**: State error covariance matrix
- Q: Model covariance matrix.
- PSD measurements y:
- K: Kalman Gain
 - Measurement error

Forecast Step:

R:

Update Step

T7

T7

$$X_{f} = M_{t}X_{t-1|}$$
$$P_{f} = M_{t}P_{t-1}M_{t}^{T} + Q_{t}$$

$$X_{a} = X_{f} + K_{t} (y_{t} - X_{f})$$
$$K_{t} = P_{f} (P_{f} + R_{t})^{-1}$$

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$$P_a = (I - K_t)P_f$$

Operator Splitting in 2D

$$X_{t}^{f} = M_{t-1} X_{t-1}^{f}$$

"Standard" Kalman Filter

$$X_{t}^{f} = M_{t-1\alpha}M_{t-1E}X_{t-1}^{f}$$

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"Operator-Splitting" Kalman Filter



VERB 2.0

In this study, we use the Versatile Electron Radiation Belt (VERB) code

Diffusion code that solves the Fokker-Planck equation for electron phase space density (PSD)

$$\begin{aligned} \frac{\partial f}{\partial t} &= L^2 \frac{\partial}{\partial L} \Big|_{\mu,J} \frac{1}{L^2} D_{LL} \frac{\partial f}{\partial L} \Big|_{\mu,J} \\ &+ \frac{1}{p^2} \frac{\partial}{\partial p} \Big|_{\alpha_0,L} p^2 D_{pp} \frac{\partial f}{\partial p} \Big|_{\alpha_0,L} + \\ &+ \frac{1}{T(\alpha_0) \sin(2\alpha_0)} \frac{\partial}{\partial \alpha_0} \Big|_{p,L} T(\alpha_0) \sin(2\alpha_0) D_{\alpha_0 \alpha_0} \frac{\partial f}{\partial \alpha_0} \Big|_{p,L}. \end{aligned}$$



Shprits et al, 2013, Nature Physics

Forecast Example – Computed 1 MeV flux in T89





Some important lessons learned and future developments

- 1. Operator splitting may be used to approximate a full 3D Kalman filter approach.
- Statistical errors do not result in significant differences in model forecast performance in future we require an activity dependent approach, and perhaps more attention to spatial dependence.
- With the current framework we can not determine innovation, and interpretation of the 3 sets of covariance matrices is difficult – we are limited to improving the model through statistical error analysis and physical understanding
- 4. The method is however suitable for fast operation and performs quite well in an operational framework

Future:

- 1. Introduce 2D Kalman filtering in pitch angle and energy this should allow us to improve our wave-particle interaction models
- 2. Implement more sophisticated ways of accounting for errors and test

Error Analysis – Models and Data

We have computed errors in electron PSD using several B-field models and a PSD matching technique



Error Analysis – Models and Data

Old Diffusion Coefficients See *Subbotin et al.*, [2010] JGR

New Diffusion Coefficients Hiss - *Spasojevic et al,* [2015] and Orlova et al, [2016]



Validation metrics

$$PE = 1 - \frac{\sum_{i=1}^{N} (m_i - p_i)^2}{\sum_{i=1}^{N} (m_i - \langle m_i \rangle)^2}.$$

$$SS = \frac{PE_{Model} - PE_{Persist}}{1 - PE_{Persist}}$$
$$= \frac{\sum_{i=1}^{N} (m_i - m_{i-1})^2 - \sum_{i=1}^{N} (m_i - p_i)^2}{\sum_{i=1}^{N} (m_i - m_{i-1})^2}.$$

$$FS = \frac{PE_{Model}}{PE_{Persist}} = \frac{\sum_{i=1}^{N} (m_i - \langle m_i \rangle)^2 - \sum_{i=1}^{N} (m_i - p_i)^2}{\sum_{i=1}^{N} (m_i - \langle m_i \rangle)^2 - \sum_{i=1}^{N} (m_i - m_{i-1})^2}.$$

Forecast Performance

Old Diffusion Coefficients

New Diffusion Coefficients

PSD in a 15 min interval, one day ahead



PSD in a 15 min interval, two days ahead







Performance Analysis – one day



Equal errors

Statistical errors

Performance Analysis – one day



Old Dxx – Equal errors

New Dxx – Equal errors

Performance Analysis – one day



New Dxx – Equal errors

New Dxx – statistical errors

Performance Analysis – two day





New Dxx – Equal errors

New Dxx – statistical errors