COSPAR Capacity Building
Sinaia, 5 June 2007
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## Basic Analysis Techniques \& Multi-Spacecraft Data

This lecture:
$\triangleright$ Introduction: single-spacecraft vs. multi-spacecraft
$\triangleright$ Single S/C data: minimum variance analysis
$\triangleright$ Multi S/C data: reciprocal vectors

Computer session: select your assignment(s)

## 1 Introduction: Magnetospheric structure



Boundary layers, sheet structures:

- bow shock,
$\triangleright$ magnetopause,
$\triangleright$ current sheets.
Orientation, motion, shape, ... of the surface ?

Baumjohann/Treumann/Mayr-Ihbe

## Magnetospheric structure (cont'd)


Magnetospheric currents:

- magnetopause current,
$\triangleright$ neutral sheet current,
$\triangleright$ field-aligned currents:
Orientation of current sheets, current direction and density, ...?

Baumjohann/Treumann/Mayr-Ihbe

## Single spacecraft vs. multi-spacecraft

Single-spacecraft data:
$\triangleright$ one-dim. track in 3+1 dim. space-time
$\rightarrow$ ambiguity,
$\triangleright$ additional information required
$\triangleright$ popular approach: minimum variance analysis.

Multi-spacecraft data:
$\triangleright$ several trajectories in 3+1 dim. space-time,
$\triangleright$ analysis of spatiotemporal processes,
$\triangleright$ additional information still useful,
$\downarrow$ various approaches.

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## 2 Single-Spacecraft Data: MinVar Approach

To resolve space-time ambiguity,
$\triangleright$ additional information is needed,

- choose an appropriate physical model,
$\triangleright$ use physical constraints to estimate model parameters.

| Phenomena | Models | Parameters |
| :---: | :---: | :---: |
| Boundaries | Planar structures | Normal vector $\hat{\mathbf{n}}$ |
| FACs | Current sheets | Orientation, $\mathbf{j}_{\\|}$ |
| Waves | Plane waves | Prop. vector $\mathbf{k}$ |

Key technique: minimum variance analysis

## Minimum variance principle

Physical model for boundary analysis: planar 1D structure with normal vector $\hat{n}$, use physical constraint

$$
0=\nabla \cdot \mathbf{B} \stackrel{1 \mathrm{D}}{=} \hat{\mathbf{n}} \cdot \nabla(\mathbf{B} \cdot \hat{\mathbf{n}}) \quad \Rightarrow \quad \mathbf{B} \cdot \hat{\mathbf{n}}=\mathrm{const}
$$

In practice: model only approximately valid.

Least squares approach: as an estimator for the sheet normal direction take the direction $\hat{\mathbf{n}}$ where data set $\left\{\mathbf{B}^{(\ell)}\right\}_{\ell=1, \ldots, L}$ shows the smallest quadratic variation (variance):

$$
\left.\left.\langle |(\mathbf{B}-\langle\mathbf{B}\rangle) \cdot \hat{\mathbf{n}}\right|^{2}\right\rangle \equiv \frac{1}{L} \sum_{\ell}\left|\left(\mathbf{B}^{(\ell)}-\langle\mathbf{B}\rangle\right) \cdot \hat{\mathbf{n}}\right|^{2}=\operatorname{Min}!
$$

## Minimum variance principle (cont'd)

Result: $\hat{\mathbf{n}}$ is the normalized eigenvector associated with the smallest (positive) eigenvalue $\lambda_{3}$ of the (co)variance matrix:

$$
\mathrm{M}=\left(\begin{array}{ccc}
\operatorname{cov}\left(B_{x}, B_{x}\right) & \operatorname{cov}\left(B_{x}, B_{y}\right) & \operatorname{cov}\left(B_{x}, B_{z}\right) \\
\operatorname{cov}\left(B_{y}, B_{x}\right) & \operatorname{cov}\left(B_{y}, B_{y}\right) & \operatorname{cov}\left(B_{y}, B_{z}\right) \\
\operatorname{cov}\left(B_{z}, B_{x}\right) & \operatorname{cov}\left(B_{z}, B_{y}\right) & \operatorname{cov}\left(B_{z}, B_{z}\right)
\end{array}\right)
$$

Assessment of model quality: eigenvalue ratios $\lambda_{3} / \lambda_{2}$ and $\lambda_{3} / \lambda_{1}$ $\rightarrow$ should be small.

Hodogram plots: project magnetic field components onto eigenvectors of M , then plot ( $B_{1} \mathrm{vs} . B_{2}$ ) and ( $B_{1} \mathrm{vs}$. $B_{3}$ ).

Details: see chapter 8 of the ISSI Cluster data analysis book.

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## Sheet current estimation using FREJA data

Orbit: 2178, Date: 20-MAR-1993, Time: 03:08:00 - 03:22:00


Current density in sheets: $j_{\|}=\frac{\Delta_{\perp} B}{\mu_{0} \mathbf{v} \cdot \hat{\mathbf{n}} \Delta t}$.

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## 3 Multi-Spacecraft Data

More information than in single-spacecraft missions but spatial resolution still much less than temporal resolution.

Different approaches to data analysis and interpretation:
(1) Apply single-spacecraft techniques separately, estimate model parameters for each satellite, and compare.
$\triangleright$ Assess evolution/variation of parameters.
$\triangleright$ Consistency checks.

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Different approaches to data analysis (cont'd)
(2) Build refined models based on the degree of coherency/correlation between measurements at different spacecraft.
$\triangleright$ Use timing information of boundary crossings to perform boundary parameter estimation.
$\triangleright$ Interpret differences in measurements as (linear) spatial variations to estimate spatial derivatives (curl, div, grad).
$\triangleright$ Interpret differences in measurements as phase variations of plane waves to estimate wave parameters (k-filtering $=$ wave telescope technique).

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Boundary analysis: crossing times approach

Boundary velocity $\mathbf{V}=V \hat{\mathbf{n}}$, crossings at $t_{\alpha}, \mathbf{r}_{\alpha}$.

For convenience, choose time and space origin such that

$$
\sum_{\alpha} \mathbf{r}_{\alpha}=0 \quad, \quad \sum_{\alpha} t_{\alpha}=0 .
$$

Now minimize the cost function

$$
\sum_{\alpha}\left[\hat{\mathbf{n}} \cdot \mathbf{r}_{\alpha}-V t_{\alpha}\right]^{2}=\operatorname{Min}!\Leftrightarrow \sum_{\alpha}\left[\mathbf{m} \cdot \mathbf{r}_{\alpha}-t_{\alpha}\right]^{2}=\operatorname{Min}!
$$

w.r.t. $\hat{\mathbf{n}}$ and $V$ or $\mathbf{m}=\hat{\mathbf{n}} / V$ ('slowness').

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## Crossing times approach (cont'd)

Result: $\underbrace{\left(\sum_{\alpha} \mathbf{r}_{\alpha} \mathbf{r}_{\alpha}^{\dagger}\right)}_{=\mathrm{R}} \mathbf{m}=\sum_{\alpha} t_{\alpha} \mathbf{r}_{\alpha} \Rightarrow \mathbf{m}=\mathrm{R}^{-1} \sum_{\alpha} t_{\alpha} \mathbf{r}_{\alpha}$.

Cluster-II mission: four satellites, then

$$
\mathrm{R}^{-1}=\mathrm{K}=\sum_{\alpha} \mathbf{k}_{\alpha} \mathbf{k}_{\alpha}^{\dagger}
$$

where the vectors $\mathbf{k}_{\alpha}$ are the reciprocal vectors.

Explicit formula: $\mathbf{m}=\sum_{\alpha} t_{\alpha} \mathbf{k}_{\alpha} \rightarrow V=1 /|\mathbf{m}|, \hat{\mathbf{n}}=V \mathbf{m}$.

## Reciprocal vectors of the Cluster-II tetrahedron

Cluster-II mission: four satellites, tetrahedral configuration.

Reciprocal base of the tetrahedron: $\left\{\mathbf{k}_{\alpha}\right\}$, defined through

$$
\mathbf{k}_{\alpha}=\frac{\mathbf{r}_{\beta \gamma} \times \mathbf{r}_{\beta \lambda}}{\mathbf{r}_{\beta \alpha} \cdot\left(\mathbf{r}_{\beta \gamma} \times \mathbf{r}_{\beta \lambda}\right)}
$$

where $\mathbf{r}_{\alpha \beta}=\mathbf{r}_{\beta}-\mathbf{r}_{\alpha}$ are relative position vectors, $(\alpha, \beta, \gamma, \lambda)$ must be a permutation of $(0,1,2,3)$.

Very useful for boundary analysis, estimation of spatial derivatives (curl, div, grad), characterization of the geometric quality of the tetrahedron.

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## Current estimation using multi-spacecraft data

Electrical currents from magnetic field data: estimate $\nabla \times \mathbf{B}$.

Linear estimators for spatial derivatives:

$$
\begin{aligned}
\nabla \times \mathbf{B} & \simeq \sum_{\alpha} \mathbf{k}_{\alpha} \times \mathbf{B}_{\alpha} \\
\nabla \cdot \mathbf{B} & \simeq \sum_{\alpha} \mathbf{k}_{\alpha} \cdot \mathbf{B}_{\alpha} \equiv \sum_{\alpha} \mathbf{k}_{\alpha}^{\dagger} \mathbf{B}_{\alpha} \\
\nabla \mathbf{B} & \simeq \sum_{\alpha} \mathbf{k}_{\alpha} \mathbf{B}_{\alpha}^{\dagger}
\end{aligned}
$$

Gradient of scalar field: $\nabla p \simeq \sum_{\alpha} \mathbf{k}_{\alpha} p_{\alpha}$

## Error analysis of spatial derivative estimation

Sources of error:
$\triangleright$ FGM measurement errors $\delta B$,
$\triangleright$ position inaccuracies $\delta r$,
$\triangleright$ nonlinear field variations.

Effect of $\delta B$ on gradient estimation errors, approx. formula:

$$
\left\langle(\delta|D \mathbf{B}|)^{2}\right\rangle=\frac{f}{3} \sum_{\alpha}\left|\mathbf{k}_{\alpha}\right|^{2}(\delta B)^{2}
$$

Here $f=3$ for $\nabla \cdot \mathbf{B}, f=2$ for $\nabla \times \mathbf{B}, f=1$ for $\hat{\mathbf{e}} \cdot \nabla \mathbf{B}$.

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## Computer session assignments



curlBy

curlBx

curlBz


## Estimation of inhomogeneity length scales

Unit vector field: $\hat{\mathbf{B}}=\mathbf{B} / B=\mathbf{B} /|\mathbf{B}|$.
Associated gradient matrix: $\nabla \hat{\mathbf{B}} \simeq \sum_{\alpha} \mathbf{k}_{\alpha} \hat{\mathbf{B}}_{\alpha}^{\dagger}$.
Curvature: $\hat{\mathbf{B}} \cdot \nabla \hat{\mathbf{B}}$, curvature radius

$$
R_{\text {curv }}=|\hat{\mathbf{B}} \cdot \nabla \hat{\mathbf{B}}|^{-1}
$$

Other inhomogeneity length scales measure the convergence of field lines $(\hat{V} \perp \hat{B})$ :

$$
R_{\text {conv }}=|\hat{\mathbf{V}} \cdot \nabla \hat{\mathbf{B}}|^{-1}
$$

## Wave identification using Cluster-II



## Cluster-II mission

Four point measurements allow to study spatiotemporal phenomena: Alfvén waves, surface waves, turbulence ...
[Image credit: ESA]
Spatial coverage too bad to determine k-spectrum directly
$\longrightarrow$ use methods from array signal (e.g. seismic data) processing.

## Wave identification using Cluster-II (cont'd)

Fourier transformed (scalar) data: $b_{\alpha}(\omega)$.
Method can be generalized to vector data $\mathbf{B}_{\alpha}$.

$$
\begin{gathered}
\text { Data vector } \\
\mathbf{b}(\omega)=\left(\begin{array}{c}
\text { CSD matrix } \\
b_{\alpha=1}(\omega) \\
b_{\alpha=2}(\omega) \\
b_{\alpha=3}(\omega) \\
b_{\alpha=4}(\omega)
\end{array}\right) \\
\\
\\
\text { (measurements) }
\end{gathered} \begin{gathered}
\text { (array geometry) }
\end{gathered}
$$

## Wave identification using Cluster-II (cont'd)

High-resolution beamformers, idea: make the sensor array most sensitive in one "looking direction" through assignment of sensor weights.

Other names: wave telescope, k-filtering technique, minimum variance estimators, Capon estimators...

Resulting estimator for the power spectrum:

$$
P(\omega, \mathbf{k})=\left(\hat{\mathbf{h}}^{\dagger} \mathrm{C}^{-1} \hat{\mathbf{h}}\right)^{-1}
$$

## Cluster-II wave identification: analysis procedure

$\triangleright$ compute $\mathrm{C}(\omega)$ and the inverse matrix $C^{-1}$,
$\triangleright$ compute $\operatorname{Tr}\{C(\omega)\}$ (i.e., frequency spectrum) and identify peaks,
$\triangleright$ discretize k-space and compute $\hat{\mathbf{h}}(\mathbf{k})$,
$\triangleright$ construct $P(\omega, \mathbf{k})$ and
 search for peaks.
[Glassmeier et al., 2001]

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## 4 Summary

Single-spacecraft missions: space-time ambiguity, additional information needed, minimum variance analysis.

Multi-spacecraft missions: much better chance to study spatiotemporal phenomena.

Cluster-II: reciprocal vectors useful for the analysis of discontinuities and spatial gradients.

Cluster-II as a wave telescope.

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## Computer session

http://www.faculty.iu-bremen.de/jvogt/cospar/cbw6/intro/
(1) Getting started with IDL
(2) Probability density estimation using the kernel method
(3) Gradient estimation accuracy in model magnetic fields
(4) Gradient estimation in measured magnetic fields
(5) Magnetospheric boundary analysis

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## Review of basic concepts in time series analysis

See textbooks and web resouces on data analysis, e.g.:
http://www.faculty.iu-bremen.de/jvogt/edu/fall06/c210222/space/
(1) Statistical description of data
(2) Correlation and regression
(3) Fourier transformation and spectral analysis
(4) Basic aspects of time series filtering

