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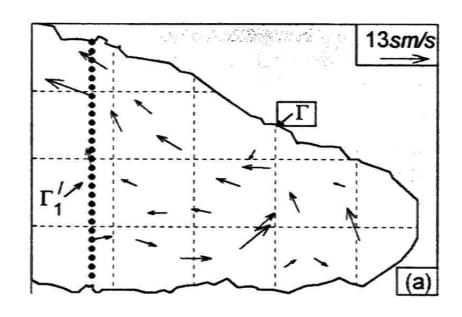
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Reconstruction of Ocean Currents From Sparse and Noisy Data

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Can we get the velocity signal from sparse and noisy data?Black Sea



 How can we assimilate sparse and noisy velocity data into numerical model?

Flow Decomposition

2 D Flow (Helmholtz)

$$u = -\partial \psi / \partial y + \partial \phi / \partial x$$
$$v = \partial \psi / \partial x + \partial \phi / \partial y$$

 3D Flow (Toroidal & Poloidal): Very popular in astrophysics

$$\mathbf{u} = \mathbf{r} \times \nabla A_1 + \mathbf{r} \mathbf{A}_2 + \nabla A_3$$

3D Incompressible Flow

If Incompressible

$$\nabla \cdot \mathbf{u} = 0$$

We have

$$\mathbf{u} = \nabla \times (\mathbf{r}\Psi) + \nabla \times \nabla \times (\mathbf{r}\Phi).$$

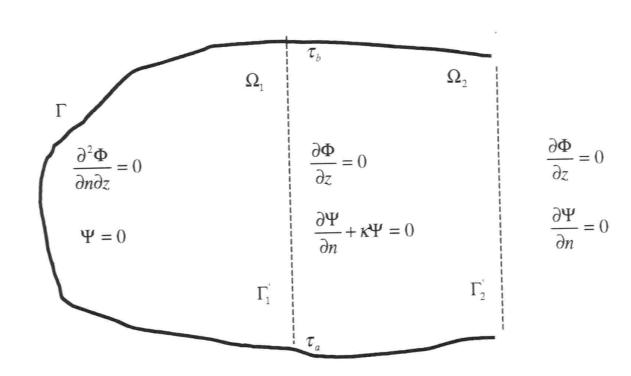
Flow Decomposition

$$u = \frac{\partial \Psi}{\partial y} + \frac{\partial^2 \Phi}{\partial x \partial z}, \qquad v = -\frac{\partial \Psi}{\partial x} + \frac{\partial^2 \Phi}{\partial y \partial z},$$

$$\nabla^2 \Psi = -\zeta$$
, ζ is relative vorticity

$$\nabla^2 \Phi = - W$$

Boundary Conditions



Basis Functions

$$\Psi(x,y,z,t^\circ) = \sum_{k=1}^\infty a_k(z,t^\circ) \Psi_k(x,y,z,\kappa^\circ),$$

$$rac{\partial \Phi(x,y,z,t^\circ)}{\partial z} = \sum_{m=1}^\infty b_m(z,t^\circ) \Phi_m(x,y,z),$$

Determination of Basis Functions

- Poisson Equations
- Γ Rigid Boundary
- Γ' Open Boundary
- $\{\lambda_k\}$, $\{\mu_m\}$ are Eigenvalues.

Basis Functions are predetermined.

$$\Delta \Psi_{\mathbf{k}} = -\lambda_{\mathbf{k}} \Psi_{\mathbf{k}},$$

$$\triangle \Phi_m = -\mu_m \Phi_m,$$

$$\Psi_k|_{\Gamma} = 0, \quad \frac{\partial \Phi_m}{\partial n}|_{\Gamma} = 0,$$

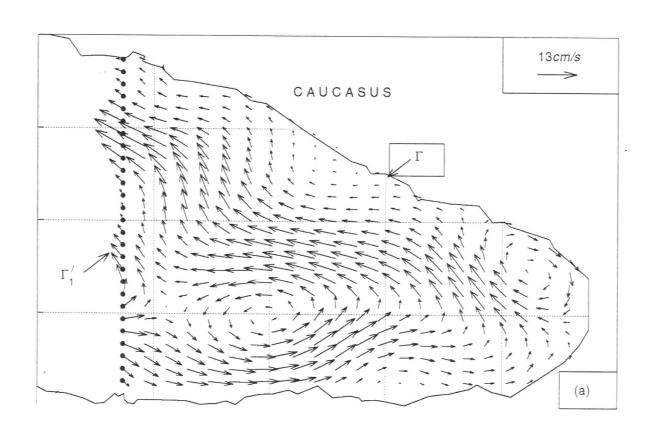
$$\left[\frac{\partial \Psi_k}{\partial n} + \kappa(\tau)\Psi_k\right]|_{\Gamma'} = 0, \quad \Phi_m|_{\Gamma'} = 0.$$

Flow Reconstruction

$$u_{KM} = \sum_{k=1}^{K} a_{k}(z, t^{\circ}) \frac{\partial \Psi_{k}(x, y, z, \kappa^{\circ})}{\partial y} + \sum_{m=1}^{M} b_{m}(z, t^{\circ}) \frac{\partial \Phi_{m}(x, y, z)}{\partial x},$$

$$v_{KM} = -\sum_{k=1}^{K} a_{k}(z, t^{\circ}) \frac{\partial \Psi_{k}(x, y, z, \kappa^{\circ})}{\partial x} + \sum_{m=1}^{M} b_{m}(z, t^{\circ}) \frac{\partial \Phi_{m}(x, y, z)}{\partial y}$$

Reconstructed Circulation



Conclusions

- Reconstruction is a useful tool for processing real-time velocity data with short duration and limited-area sampling.
- The scheme can handle highly noisy data.
- The scheme is model independent.
- The scheme can be used for assimilating sparse velocity data