

Effective Waveform Inversion = Nonlinear MVA

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Agenda

MVA and Semblance

Extended Modeling: MVA + WI

Conclusions

Surface-oriented, Depth-oriented Image Volumes

MVA based on prestack depth migration - two major variants. Both produce *image volume* $I(\mathbf{x}, \mathbf{h})$ depending on image point \mathbf{x} , half-offset \mathbf{h} .

(I) Surface oriented: $\mathbf{h} = 0.5(\text{receiver} - \text{source})$, usually computed by diffraction sum (“Kirchhoff common offset migration”); binwise: offset bin $I(\cdot, \mathbf{h})$ depends only on data traces with offset \mathbf{h} .

(II) Depth oriented: $2\mathbf{h} =$ difference between subsurface scattering points, $\mathbf{x} =$ their midpoint. Every point in image volume depends on all data traces. Has diffraction sum rep, but usually computed by one-way (shot profile or DSR) or two-way (RTM) wave extrapolation.

Semblance

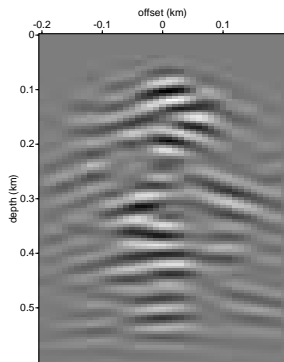
Semblance condition: expresses consistency between data, velocity model in terms of image volume.

(I) **Surface oriented**: velocity-data consistency when $I(\mathbf{x}, \mathbf{h})$ independent of \mathbf{h} (at least in terms of phase), i.e. **image gathers are flat**.

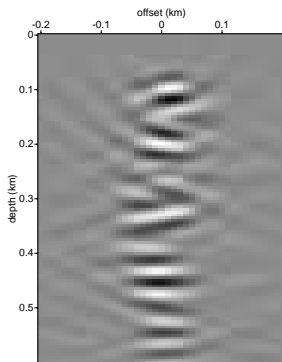
(II) **Depth oriented**: velocity-data consistency when $I(\mathbf{x}, \mathbf{h})$ concentrated near $\mathbf{h} = 0$, i.e. **image gathers are focused** [or flat, when converted to scattering angle].

Main principle of MVA: adjust velocity until image volume satisfies semblance condition.

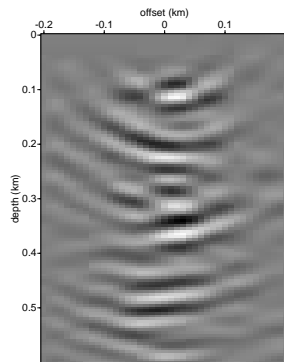
Semblance



OIG, x=1 km: vel 10% low



Offset Image Gather, x=1 km



OIG, x=1 km: vel 10% high

RTM space shift image gathers ($I_D(\mathbf{x}, \mathbf{h})$) from velo model $v + \delta v$,
 $v = \text{const.}$, $\delta v =$ randomly distributed point diffractors. Left to
Right: migration velocity = 90%, 100%, 110% of true velocity.

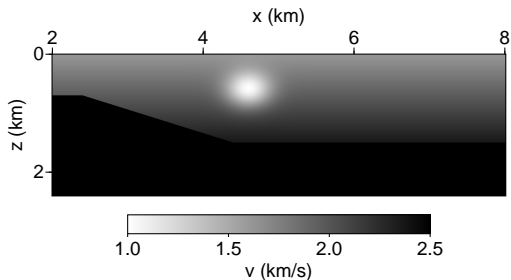
Surface-Oriented vs. Depth-Oriented, MVA

Nolan & S. 97, Stolk & S. 04, deHoop & Brandsberg-Dahl 03: multipathing (multiple rays connecting source, receiver, and image points, caustics) leads to **kinematic artifacts in surface oriented image volume**.

Artifact = coherent event in wrong place, of strength comparable to correct events.

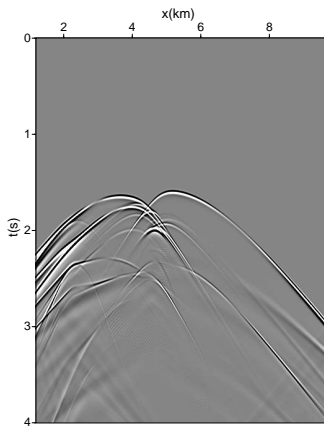
Consequences for velocity analysis: artifacts \Rightarrow semblance condition not satisfied even if velocity is correct!: Nolan and S. 97, Xu SEG 07.

Surface-Oriented vs. Depth-Oriented, MVA



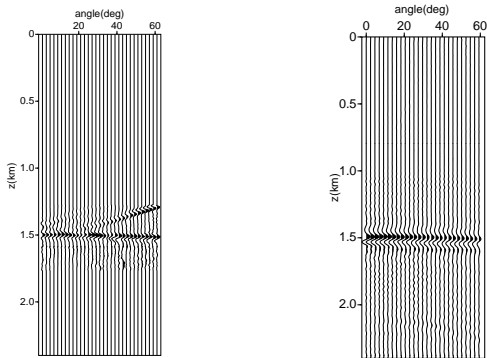
Velocity model after Valhall field, North Sea. Note sloping reflector at left, large low-velocity lens (modeling gas accumulation) in center. Both tend to produce multipathing. (Thanks: M. de Hoop, A. Malcolm)

Surface-Oriented vs. Depth-Oriented, MVA



Typical shot gather over center of model, exhibiting extensive multipathing.

Surface-Oriented vs. Depth-Oriented, MVA



Angle Domain CIGs at same horizontal position from surface-oriented (Kirchhoff) and depth-oriented (DSR) migrated image volumes. **Left:** ADCIG from Kirchhoff migration: kinematic artifacts clearly visible. **Right:** ADCIG from DSR migration: no artifacts!

Surface-Oriented vs. Depth-Oriented, MVA

Stolk & deHoop 01, S. 02, deHoop, Stolk & S. 05: **depth-oriented image volume generally free of artifacts**, even with strong multipathing.

So the two types of image volume are not even kinematically equivalent!

Accounts for perceived superiority of “wave equation migration”.

Suggests: depth-oriented volume possibly better domain for MVA in complex, refracting subsurface.

Surface-Oriented vs. Depth-Oriented, MVA

MVA via Optimization:

- ▶ form measure of deviation of image volume from semblance condition - function of velocity model; all energy not conforming to semblance condition contributes.
- ▶ optimize numerically: gradient = backprojection of semblance-inconsistent energy into velocity update.

Inherently uses all events in data, weighted by strength.

Example: for depth-oriented, minimize $J[v] = \sum |\mathbf{h}/(\mathbf{x}, \mathbf{h})|^2$ - penalizes energy at $\mathbf{h} \neq 0$. **Apparently: no local mins!**

Recent contributions: Shen 03, 05, Li & S. 05, Foss 06, Albertin 06, Khoury 06, Verm 06, Kabir SEG 07, Shen & S TRIP08.

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Extended Modeling

Extended model $\bar{\mathcal{F}} : \bar{\mathcal{M}} \rightarrow \mathcal{D}$, where $\bar{\mathcal{M}}$ is a *bigger model space* = models depending on \mathbf{x} and \mathbf{h} , i.e. $\bar{v}(\mathbf{x}, \mathbf{h})$.

Physical (normal) model identified with extended model: for depth oriented modeling, $v(\mathbf{x}) \mapsto v(\mathbf{x})\delta(\mathbf{h}) = \bar{v}(\mathbf{x}, \mathbf{h})$ (satisfies semblance condition!).

Extension property: $\mathcal{F}[v] = \bar{\mathcal{F}}[\bar{v}]$.

Extended Modeling

Lailly, Tarantola, Claerbout (80's): migration operator (producing image) is *adjoint* or transpose $D\mathcal{F}[v]^T$.

True amplitude migration is (pseudo)inverse $D\mathcal{F}[v]^{-1}$. Same for extended modeling $\bar{\mathcal{F}}[\bar{v}]$:

$$D\bar{\mathcal{F}}[\chi[v]]^T d(\mathbf{x}, \mathbf{h}) = I(\mathbf{x}, \mathbf{h}), \quad D\bar{\mathcal{F}}[\chi[v]]^{-1} d(\mathbf{x}, \mathbf{h}) = \delta\bar{v}(\mathbf{x}, \mathbf{h}).$$

Extended Modeling

(1) MVA (with true amplitude) solves “partially linearized” problem: find reference velocity v and perturbation δv so that $D\mathcal{F}[v]\delta v \simeq d - \mathcal{F}[v]$.

Proof: successful true amplitude MVA produces image volume satisfying imaging condition ($\delta\bar{v}(\mathbf{x}, \mathbf{h}) \simeq \delta v(\mathbf{x})\delta(\mathbf{h})$) and fitting data, that is,

$$\min_{v, \delta\bar{v} \in M \times \bar{M}} \|h\delta\bar{v}\|^2 \text{ subj } D\bar{\mathcal{F}}[v]\delta\bar{v} \simeq d - \bar{\mathcal{F}}[v].$$

(2) Nonlinear MVA, or WI based on semblance:

$$\min_{\bar{v} \in \bar{M}} \|h\bar{v}\|^2 \text{ subj } \bar{\mathcal{F}}[\bar{v}] \simeq d.$$

Extended Modeling

- ▶ \mathcal{F} can be *any modeling* operator - acoustic, elastic, ... - So: **MVA extended to elastic modeling with multiples**, for instance.
- ▶ For surface oriented extension, nonlinear bin-by-bin modeling - cf. **Dong Sun**.
- ▶ For depth-oriented extension, $\bar{\mathcal{F}}$ expresses **action at a distance**: elastic moduli are nonlocal, stress at $\mathbf{x} + \mathbf{h}$ results from strain at $\mathbf{x} - \mathbf{h}$. So Claerbout's semblance principle is actually Cauchy's no-action-at-a-distance hypothesis!
[Thanks: Scott Morton]

Extended Modeling

Example: an acoustic extended model.

$$\rho \frac{\partial \mathbf{v}}{\partial t} = -\nabla p, \quad \frac{\partial p}{\partial t} = -K[\nabla \cdot \mathbf{v}],$$

where K is a bulk modulus *operator* expressing action-at-a-distance.

Key question: how to choose classes of ops K (nonphysical degrees of freedom) so that application of K is em cheap - else time-stepping becomes prohibitive.

Extended Modeling

Must find sparse representation for K !

Possible answer, based on experience with PSDM:

- ▶ K is nonlinear analogue of *prestack image volume*;
- ▶ prestack image volume **sparse in phase space**;
- ▶ sparsity in phase space expressible via **multiresolution frames** - eg. curvelets, see Candes-Donoho, de Hoop-Douma, Herrmann et al., Chauris.

Conjecture: bulk modulus operator K etc. should be sparse in phase space, hence admit low-cost implementation.

Extended Modeling

Second key question: how to update \bar{v} while remaining in the set of data-fitting models \bar{v} with $\bar{\mathcal{F}}[\bar{v}] \simeq d$?

Possible answer, analogous to approach in [Dong Sun](#) project:
based on

Conjecture: suppose source is impulsive, has full bandwidth down to dc. Then \bar{v} uniquely determined by data. [Not even known in enough generality for 1D! cf. [Kirk Blazek's](#) talk.]

If so, then can use **low frequency data components**, missing from field data, as control parameters permitting navigation of feasible set $\{\bar{v} : \bar{\mathcal{F}}[\bar{v}] \simeq d\}$ -**nonlinear substitute for migration velocity = macromodel.**

Extended Modeling

Another important issue: source calibration.



Patrick Lailly, Florence Delprat 01,03: nonlinear inversion (any kind!) *demands* good knowledge of source - but for extremely complex media with intense internal multiples, very difficult to secure!

Contrast: Minkoff & S 97, Winslow 99, Anno et al. 03: successful linearized inversion for source and reflectivity.

What is typical of the Earth?

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Takeaway messages of this talk:

- ▶ “Kirchhoff” and “Wave Equation” prestack migrations have different kinematic properties.
- ▶ MVA solves a “partially linearized” WI problem based on *extended modeling* - nonphysical degrees of freedom.
- ▶ MVA via waveform tomography (“differential semblance”), uses semblance condition and numerical optimization - all events constrain velocity updates, much less tendency towards local minima than least squares WI.
- ▶ Nonlinear extended scattering = framework for uniting MVA and waveform inversion.