Reflection seismic Method - 2D

- Zero-offset model
- Wavelets
 - Minimum, maximum, mixed phase
- Convolutional model
- Subsurface sampling
- Reading:
 - Sheriff and Geldart, Chapters 6, 8

Acoustic Impedance What we image in reflection sections

- At near-vertical incidence:
 - P-to-S-wave conversions are negligible;
 - P-wave reflection and transmission amplitudes are sensitive to acoustic impedance (Z=ρV) contrasts:



GEOL483.3

Zero-Offset Section the objective of pre-migration processing



Reflection imaging

- Multi-offset data are transformed into a zero-offset section:
 - Statics place sources and receivers on a flat reference (datum) surface;
 - Deconvolution compresses the wavelet into a "spike" and attenuates "short-period" multiples;
 - Filtering attenuates noise and other multiples.
- Migration transforms the zerooffset section into a depth image

GEOL483.3

Wavelets

 Impedance contrasts are assumed to be sharp, yet the wavelet always imposes its signature on the record



Standard polarity convention



Minimum-, maximum-, and zero-phase wavelets *Key facts*

- Consider a wavelet consisting of two spikes: w=(1,a):
 - For |a| < 1, it is called minimum-phase;</p>
 - For |a| > 1, it is maximum-phase;
 - Note that its z-transform is W(z)=1+az, and 1/W(z) represents a convergent series near z=0. This means that there exists a filter that could convert the wavelet into a spike.
- A convolution of all minimum- (maximum-) phase wavelets is also a minimum- (maximum-) phase wavelet:

$$W(z) = \prod_{i=0}^{N} \left(1 + a_i z \right)$$

- When minimum- and maximum-phase factors are intermixed in the convolution, the wavelet is called *mixed-phase*.
- Minimum- (maximum-) phase wavelets have the fastest (slowest) rate of energy build-up with time
- Minimum-phase wavelets are associated with causal processes.

Ricker wavelet





Convolution

 Convolution of two series, u_i, and w_i, denoted u*w, is:

$$(u*w)_i = \sum_k u_k w_{i-k}$$

For each *i*, the result is dot product of *u* and shifted and "reflected", or "folded" (i..e., running backwards) *w*

In integral form:

$$u(t) * w(t) = \int_{-\infty}^{+\infty} u(\tau) w(t-\tau) d\tau$$

As multiplication, it is symmetric (commutative):

u * w = w * u

Note that to multiply two polynomials, with coefficients u_k and w_k, we would use exactly the first formula above. Therefore, in Z or frequency domains, convolution becomes simple multiplication of polynomials (show this!):

$$u * w \Leftrightarrow U(z)W(z) \Leftrightarrow U(f)W(f)$$

 This property <u>facilitates efficient digital</u> <u>filtering</u>.

Cross-Correlation

• Cross-correlation of two series, u_i , and w_i , is: $ccorr(u*w) = \sum u_i w_{i+1}$

$$ccorr(u*w)_i = \sum_k u_k w_{i+k}$$

Unlike in convolution, no "folding" of w

In integral form:
$$+\infty$$

 $ccorr(u(t), w(t)) = \int_{-\infty}^{+\infty} u(\tau) w(t+\tau) d\tau$

 It is anti-symmetric in the following sense (show this!):

ccorr(u,w)(t) = ccorr(w,u)(-t)

 In Z or frequency domains, crosscorrelation is:

$$ccorr(u,w)(z) = \overline{U(z)}W(z)$$

 Cross-correlation is used as a measure of similarity between time series.

Autocorrelation

 Cross-correlation of a time function with itself is called Autocorrelation:

$$ccorr(u, u)_i = \sum_k u_k u_{i+k}$$

- In integral form: $\int_{-\infty}^{+\infty} u(\tau) u(t+\tau) d\tau$
- It always is an even function (show this!):

$$Auto_u(t) = Auto_u(-t)$$

 In Z or frequency domains, autocorrelation is:

Always real value -Energy Spectrum

$$Auto_u(z) = \overline{U(z)}U(z) = |U(z)|^2$$

- Therefore, autocorrelation is also the Fourier transform of the energy spectrum of the signal
 - It is independent of the phase spectrum!
- Autocorrelation is used as a measure of selfsimilarity within a time series.

Practical convolution and cross-correlation

Usually performed via Fast Fourier Transform



Convolutional model basic idea

- Reflection seismic trace is a convolution of the source wavelet with the Earth's 'reflectivity series'
- The reflectivity series includes:
 - primary reflections;
 - multiples.



Convolutional model general



Convolutional model simplified for practical reflection imaging

$$u = r * w + n_{random}$$

 Assumptions and practical approximations:



 Consequently, the autocorrelation function of the wavelet can be estimated from the data:

$$acorr(w) \approx acorr(u)$$

GEOL483.3

Convolutional model



From Yilmaz, 1987

Subsurface sampling

- Seismic surreys are designed with some knowledge of geology and with specific targets in mind:
 - Limiting factors: velocities, depths, frequencies (thin beds), dips.
- Maximum allowable geophone spacing in order to record reflections from dipping interfaces $\lambda = \lambda = V$

The same, in terms of moveout dt/dx(sin θ = tan (moveout)):

Geophone Spacing
$$_{max} < \frac{1}{2 f_{max} \frac{dt}{dx}}$$

More conservatively,
this factor
is usually taken = 4
Apparent wavelength = true wavelength / sin(dip)
 θ
 $Wavefront$

Voxel (Elementary cell of seismic volume)

- "Voxel" is determined by the spatial and time sampling of the data
 - For a typical time sampling of 2 ms (3 m two-way at 3000 m/s), it is typically 3 by 15 m² in 2D;
 - 3 by 15 by 25 m³ in 3D.
- For a properly designed survey, voxel represents the smallest potentially resolvable volume
 - Note that the Fresnel zone limitation is partially removed by *migration* where sufficiently broad reflection aperture is available.
 - Migration is essentially summation of the amplitudes over the Fresnel zones that collapses them laterally.
 - Migration is particularly important and successful in 3D.