# **Estimating the confidence of coherence anomalies** Tengfei Lin, Thang Ha, Kurt J Marfurt, University of Oklahoma; Kevin L Deal, Chevron Company



### . Introduction:

Semblance and other coherence measures are routinely used in seismic In the study, we reexamine the analysis by Douze and Laster (1979) on the

processing such as velocity spectra analysis, seismic edge detection and volumetric dip estimation, and edge-preserving structure-oriented filtering. significance of velocity-based semblance analysis in order to evaluate the significance of coherence anomalies within a noisy background, and the choice of parameters for structure-oriented filtering. These same concepts are readily generalized to eigen-structure type coherence estimates. After that, we applied the significance of coherence to the Footprint Suppression Workflow using edge preserved Structure-Oriented Filtering.

## **2. Theoretical Analysis:**

Taner and Koehler (1969) define the semblance, s, of a collection of J seismic traces u<sub>i</sub> within a 2K+1 sample vertical analysis window to be the ratio of the energy of the average trace to the average energy of the individual traces:

$$s(t) = \frac{\sum_{k=-K}^{+K} \alpha_k \left\{ \left[ \sum_{j=1}^{J} \beta_j u_j(t+k\Delta t) \right]^2 + \left[ \sum_{j=1}^{J} \beta_j u_j^H(t+k\Delta t) \right]^2 \right\}}{\sum_{k=-K}^{+K} \alpha_k \left\{ \sum_{j=1}^{J} \beta_j \left[ u_j^2(t+k\Delta t) + u_j^{H^2}(t+k\Delta t) \right] \right\}}$$

where, denotes the measured amplitude of the *j*th trace at sample *t*, are the weights applied to the kth sample and  $\beta_i$  the weights applied to the *j*th trace. Following Douze and Laster's (1979), we approximate the F-statistic with d1 and d2 degrees of freedom and non-centrally parameter (Blandford,

1974) as:

$$F_{s}(d_{1}, d_{2}, \varepsilon) = (J-1)\sum_{k=-K}^{+K} \alpha_{k} \left\{ \left[ \sum_{j=1}^{J} \beta_{j} u_{j}(t+k\Delta t) \right]^{2} + \left[ \sum_{k=-K}^{+K} \alpha_{k} \left\{ \sum_{j=1}^{J} \beta_{j} \left[ u_{j}^{-2}(t+k\Delta t) + u_{j}^{H^{2}}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{H^{2}}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \beta_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left[ \sum_{j=1}^{J} \alpha_{j} u_{j}^{-2}(t+k\Delta t) + u_{j}^{-2}(t+k\Delta t) \right] - \left$$

 $\sum_{j=1}^{n} \beta_{j} u_{j}^{H} (t + k\Delta t)$  $_{j}(t+k\Delta t) \left| + \left| \sum_{i=1}^{J} \beta_{j} u_{j}^{H}(t+k\Delta t) \right| \right|$ 

**Spectral Analysis** 

Disorder

Dominant frequency: spectral analysis

Similar to the temporal analysis window: to generate a cubic sub volume



