

Bayesian Priors for Sparse Inversion

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Outline

- Sparse inversion of seismograms
- Priors for sparse inversion
- Hyperparameter selection
- Applications
- Discussion and Conclusions

The problem

$$\mathbf{W}\mathbf{x} = \mathbf{y} + \mathbf{n}$$

$$\mathbf{C}\mathbf{x} = \boldsymbol{\xi} + \boldsymbol{\epsilon}$$

The linear programming approach

Minimize $J = \alpha|\mathbf{x}|_1 + |\mathbf{e}|_1$

subject to $\mathbf{W}\mathbf{x} = \mathbf{y} + \mathbf{e}$

and $\xi_l < \mathbf{C}\mathbf{x} < \xi_u$.

Bayes' rule

$$p(m|d) \propto p(m) \times p(d|m)$$

Posterior \propto Prior \times Likelihood

MAP solution

$$J = -\ln[p(m|d)]$$

MAP - Objective Function

$$J = \alpha \underbrace{J_x}_1 + \underbrace{\frac{1}{2} \left\| \frac{1}{\sigma} (\mathbf{W}\mathbf{x} - \mathbf{y}) \right\|^2}_2 + \underbrace{\frac{1}{2} \left\| \mathbf{S}^{-1} (\mathbf{C}\mathbf{x} - \boldsymbol{\xi}) \right\|^2}_3$$

- 1 - The solution must be sparse.
- 2 - The solution must honour the seismic trace.
- 3 - The solution must honour a set of impedance constraints.

Regularization - Sparseness criteria

$$J_p = \frac{1}{p} \sum_i |x_i|^p$$

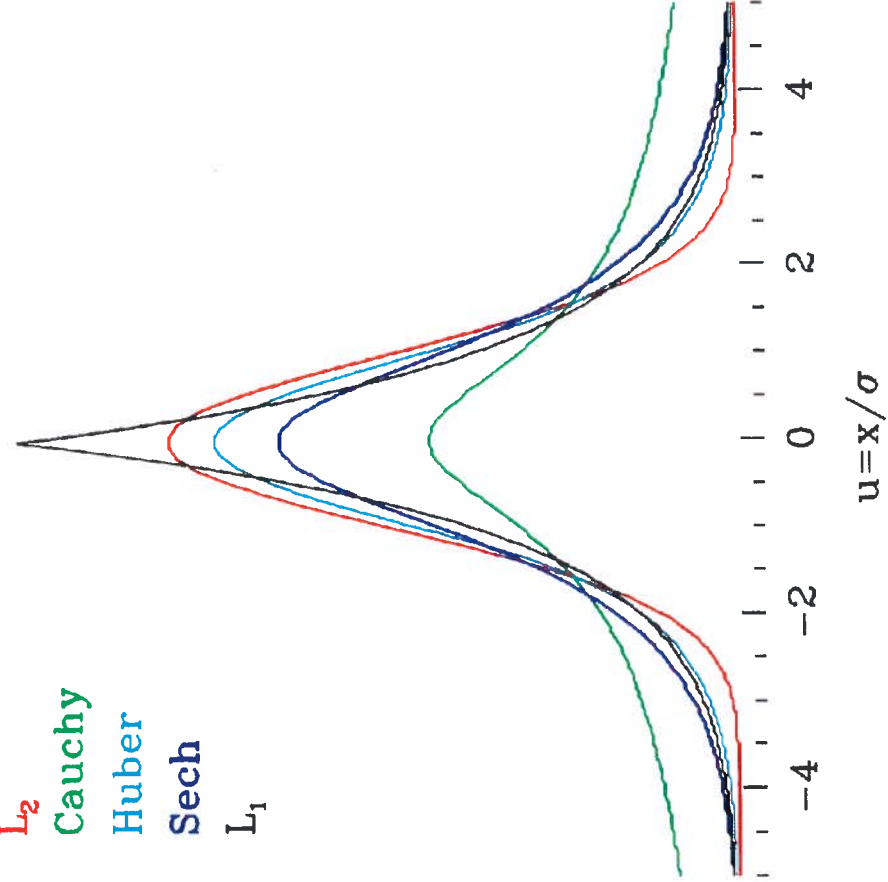
$$J_{Cauchy} = \frac{1}{2} \sum_i \ln\left(1 + \frac{x_i^2}{\sigma_x^2}\right)$$

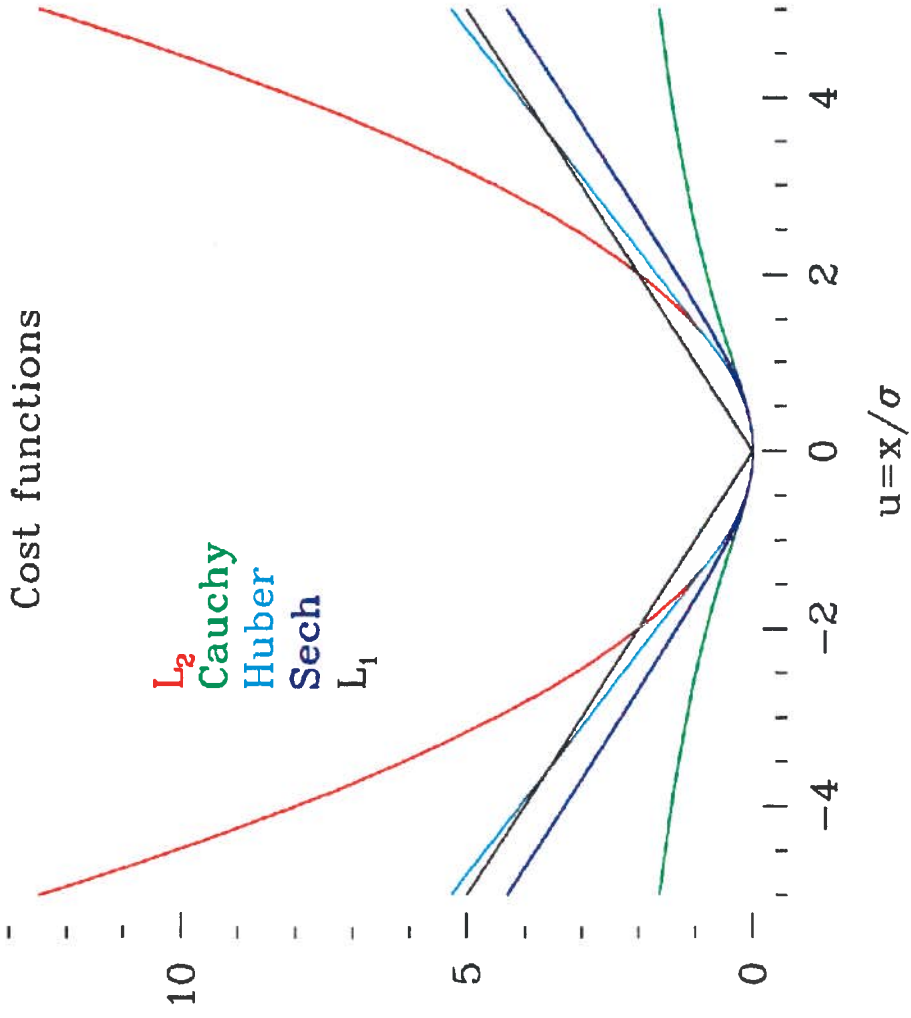
$$J_{Sech} = \sum_i \ln\left(\cosh \frac{x_i}{\sigma_x}\right)$$

$$J_{Huber} = \sum_i \begin{cases} x_i^2/2 & \text{if } |x_i| \leq x_c \\ a|x_i| - x_c^2/2 & \text{if } |x_i| > x_c \end{cases}$$

Priors for sparse inversion

- L_2
- Cauchy
- Huber
- Sech
- L_1





Minimization of J - CG algorithm

1. Given \mathbf{x}_0 compute $\mathbf{g}_0 = \nabla J(\mathbf{x})$ and set $\mathbf{d}_0 = -\mathbf{g}_0$
2. for $k = 0, 1, 2, \dots, n - 1$
3. $\mathbf{x}_{k+1} = \mathbf{x}_k + a\mathbf{d}_k$ where a minimizes $J(\mathbf{x}_k + a\mathbf{d}_k)$ (line search)
4. compute $\mathbf{g}_k = \nabla J(\mathbf{x}_{k+1})$
5. set $\mathbf{d}_{k+1} = -\mathbf{g}_{k+1} + b_k\mathbf{d}_k$ where

$$b_k = (\mathbf{g}_{k+1} - \mathbf{g}_k)^T \mathbf{g}_k / \mathbf{g}_k^T \mathbf{g}_k$$

6. Stop if

$$\frac{|J_{k+1} - J_k|}{1/2(|J_{k+1}| + |J_k|)} < \text{tolerance}$$

Hyperparameter estimation

- Cost function

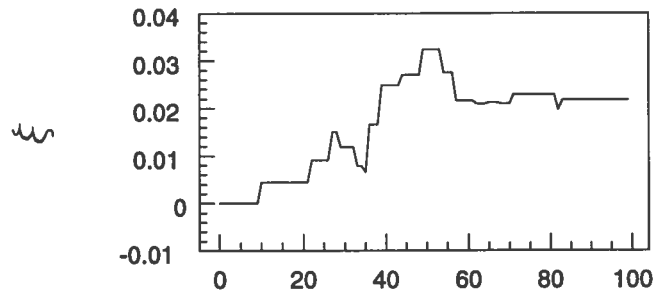
$$J = \alpha J_x + \frac{1}{2} \left\| \frac{1}{\sigma} (\mathbf{W}\mathbf{x} - \mathbf{y}) \right\|^2 + \frac{1}{2} \left\| \mathbf{S}^{-1} (\mathbf{C}\mathbf{x} - \boldsymbol{\xi}) \right\|^2$$

- Discrepancy principle

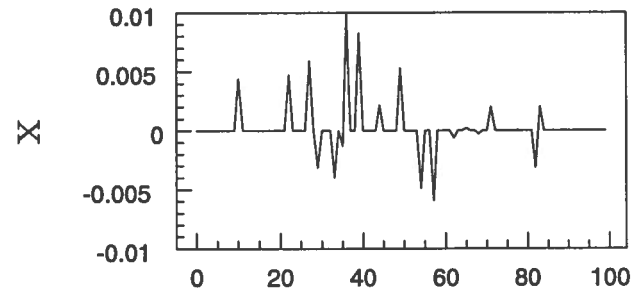
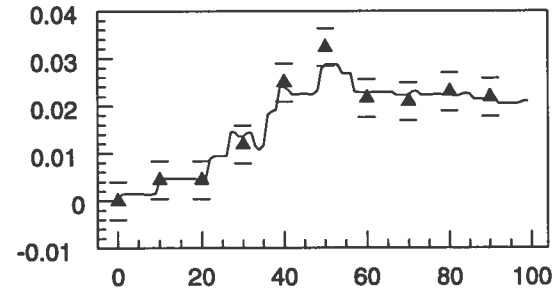
$$\chi^2 = \left\| \frac{1}{\sigma} (\mathbf{W}\mathbf{x} - \mathbf{y}) \right\|^2 + \left\| \mathbf{S}^{-1} (\mathbf{C}\mathbf{x} - \boldsymbol{\xi}) \right\|^2$$

$$E[\chi^2] = n \pm \sqrt{2n}$$

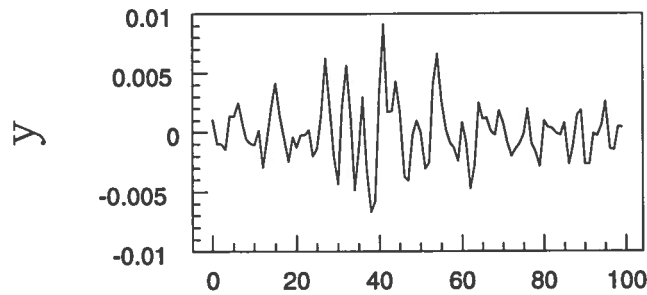
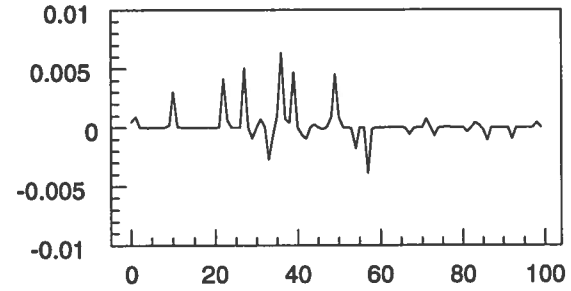
L_1 prior, Gaussian Likelihood



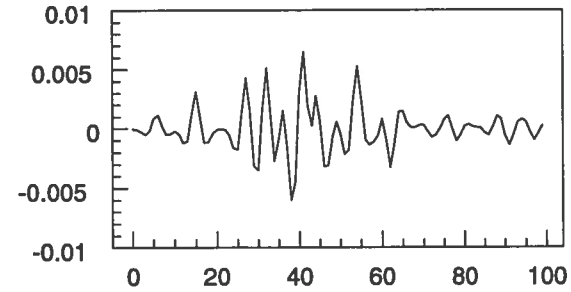
$\langle \tilde{w} \rangle$



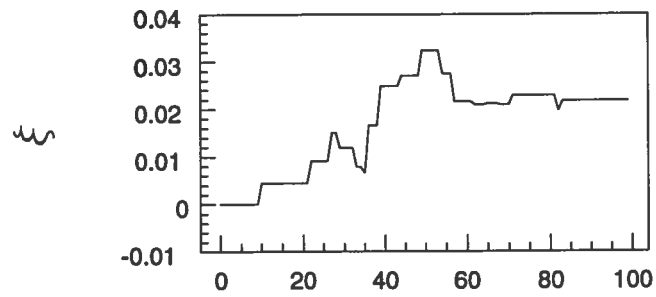
$\langle X \rangle$



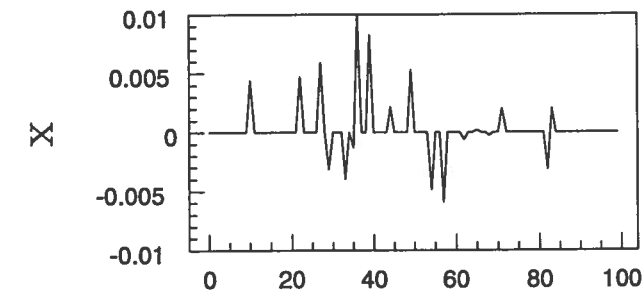
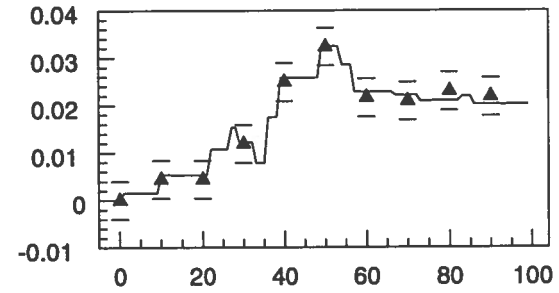
$\langle y \rangle$



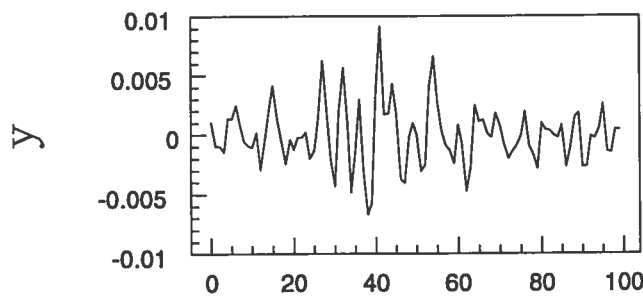
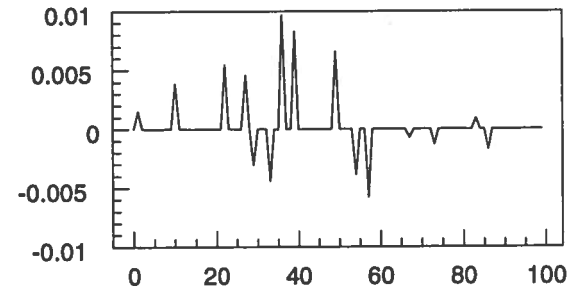
Cauchy prior, Gaussian Likelihood



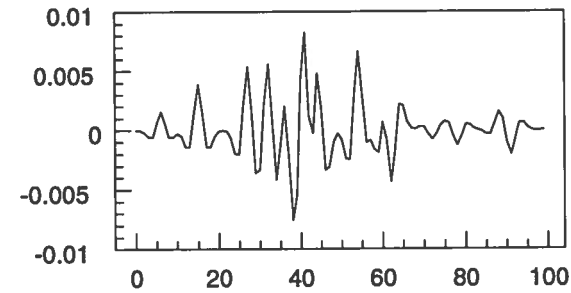
$\langle \tilde{w} \rangle$



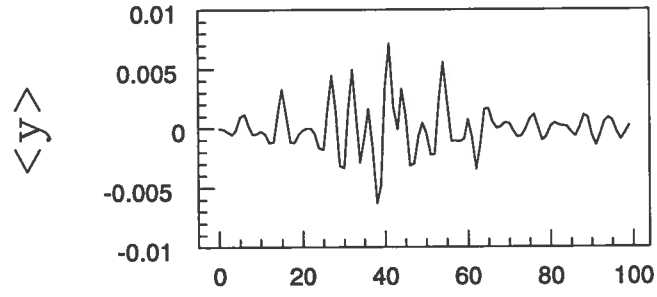
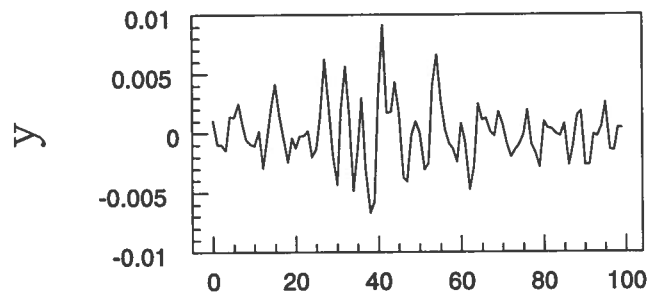
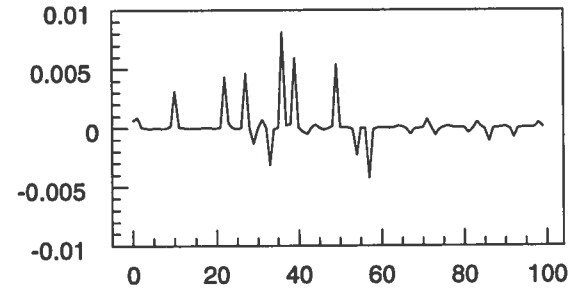
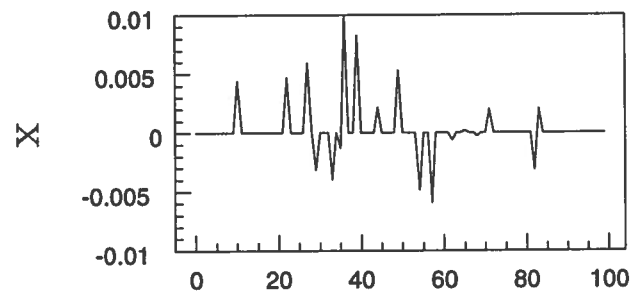
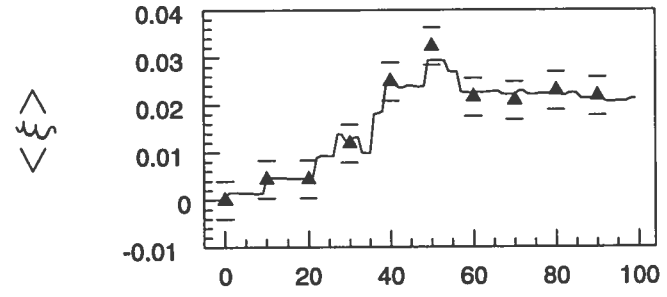
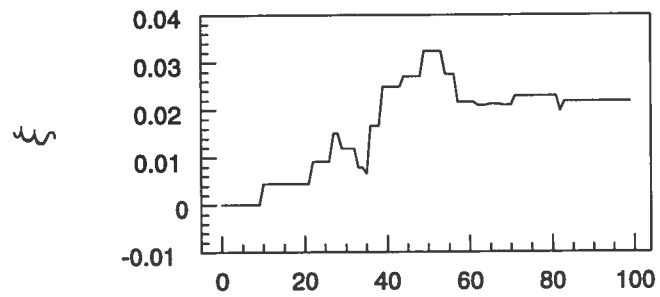
$\langle X \rangle$



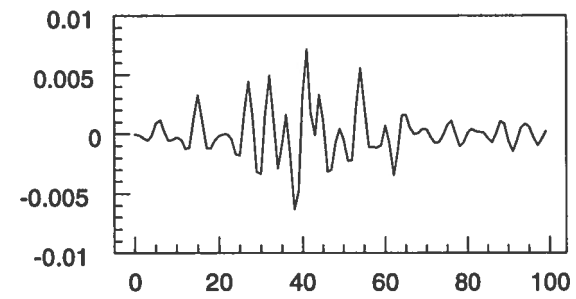
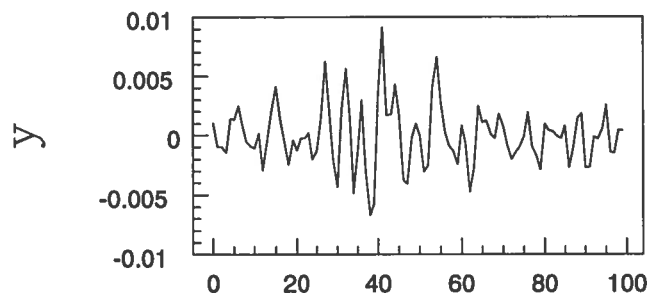
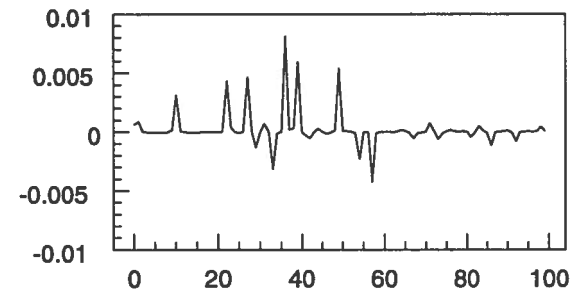
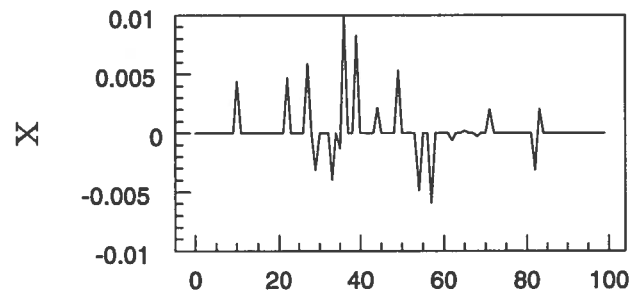
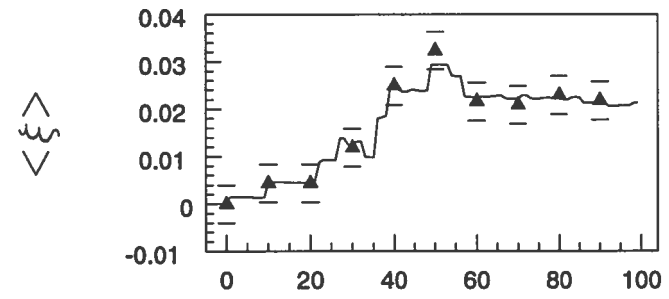
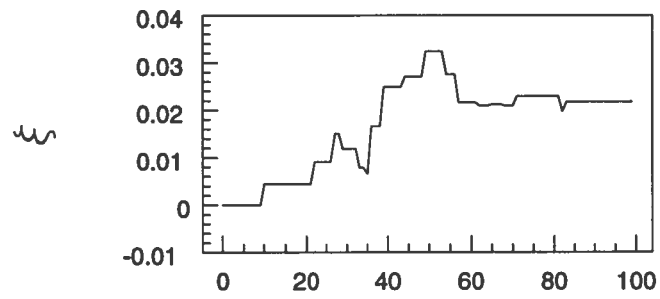
$\langle y \rangle$



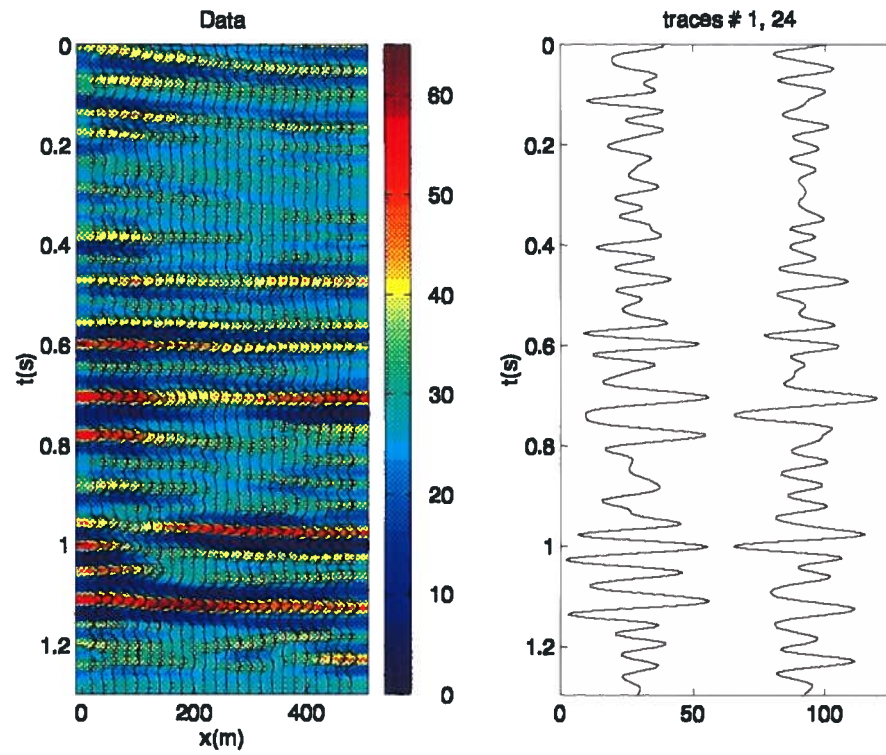
Sech prior, Gaussian Likelihood



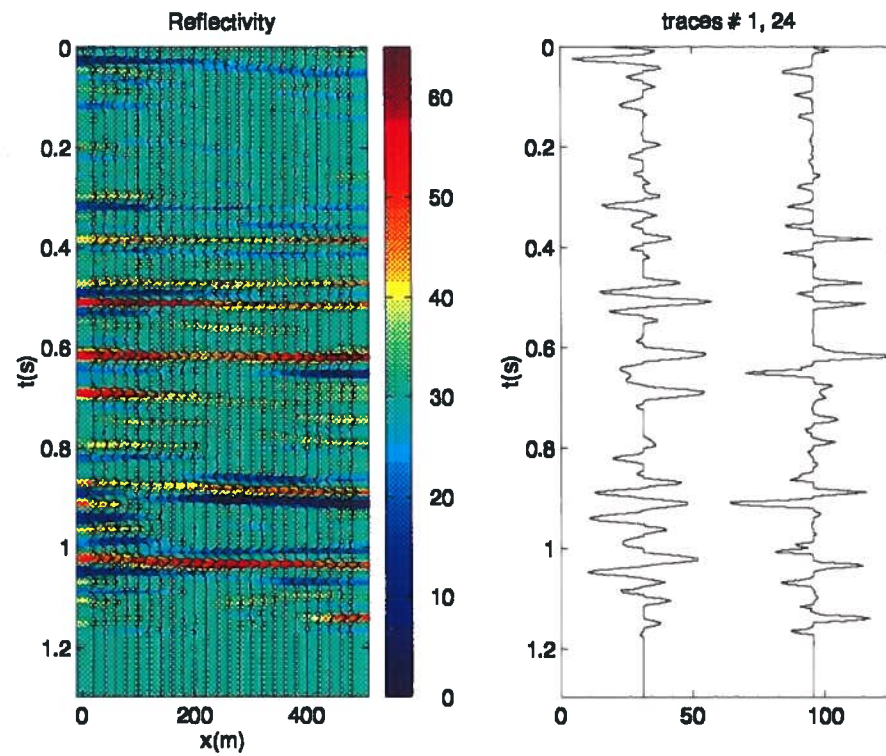
Huber prior, Gaussian Likelihood



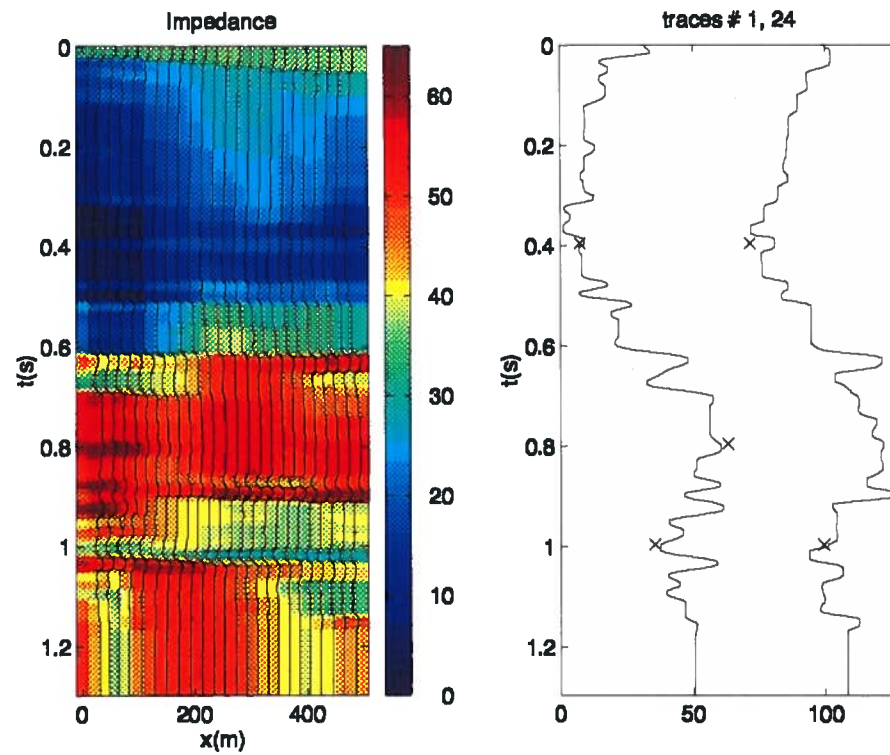
Field data example



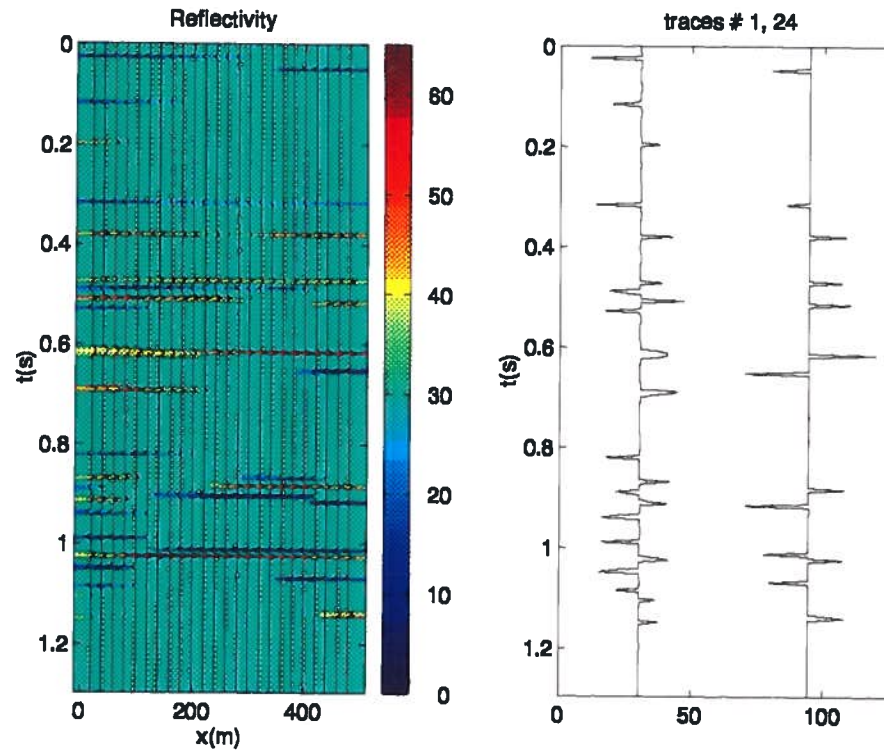
L_1 prior, Gaussian likelihood



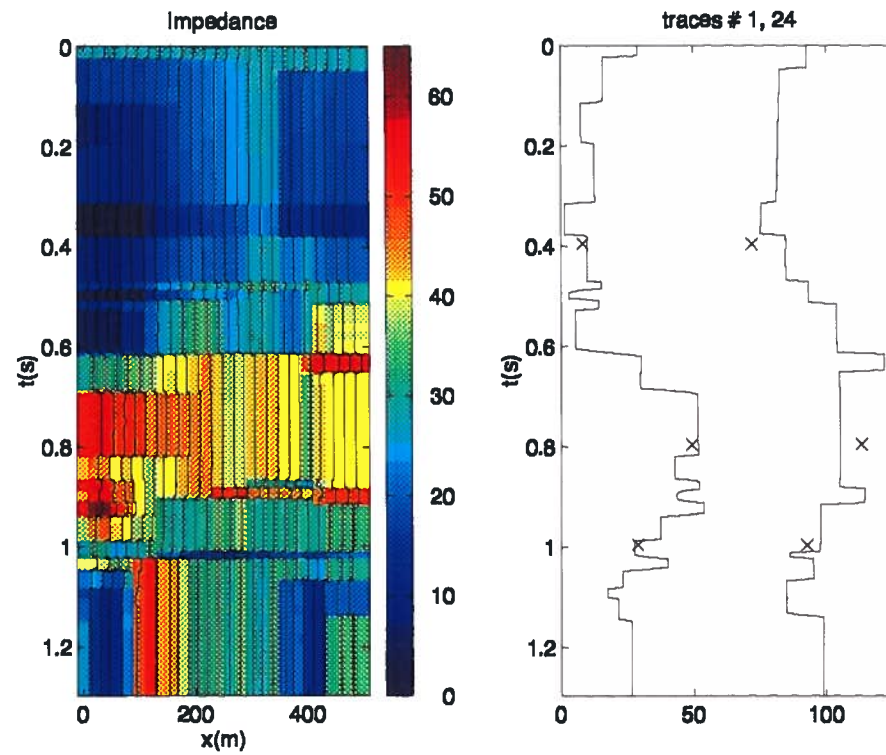
L_1 prior, Gaussian likelihood



Cauchy prior, Gaussian likelihood



Cauchy prior, Gaussian likelihood



Computations time for Bayes (CG) and Linear Programming

	N	NC	Time (s) (*)
Bayes (CG)	100	5	1.8
	300	15	6.3
L-prog. (BR)	100	5	13.3
	300	15	225.

(*) Sparc 5.

CG: Conjugate gradients

BR: Barrodale & Roberts

Conclusions

- Priors for sparse inversion

The *Huber* and the *Sech* criteria tend to treat large amplitudes like the L_1 criterion and small amplitudes like the L_2 criterion.

The Cauchy criterion treats small amplitudes like the L_2 criterion. Large amplitudes are emphasized.

- The algorithm

The cost function derived with Bayes' rule is minimized using a CG procedure. The computational cost of the linear programming approach is avoided.

Human intervention can be minimized using diverse strategies for hyperparameter selection.

Acknowledgment

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- Participant companies

AGIP

ARCO

JNOC

PanCanadian

Pulsonic