

Fourier method for inverse source problem using correlation of passive measurements

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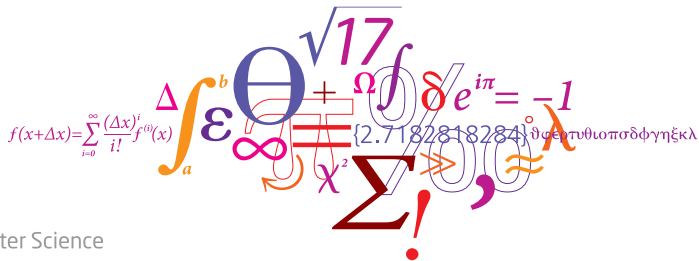
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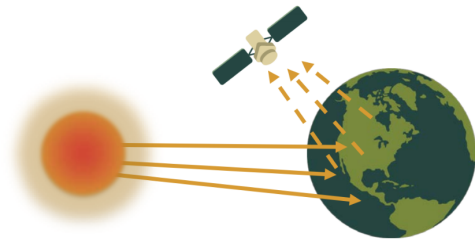
Department of Applied Mathematics and Computer Science



Outline

- Introduction
 - Motivation/inspiration
 - The model
- The inverse source problem in 1d
- Numerical experiments
- Conclusion and future work

Passive Sensors



Active Sensors

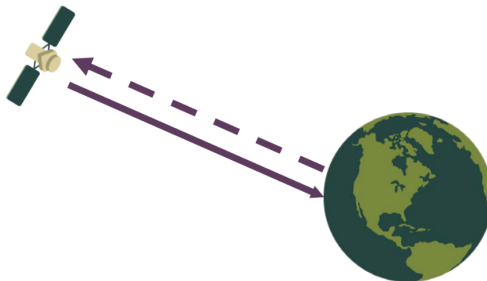


Image credit: NASA

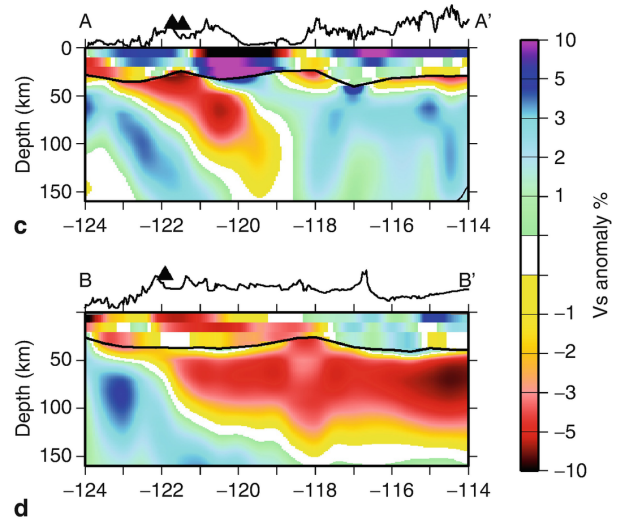
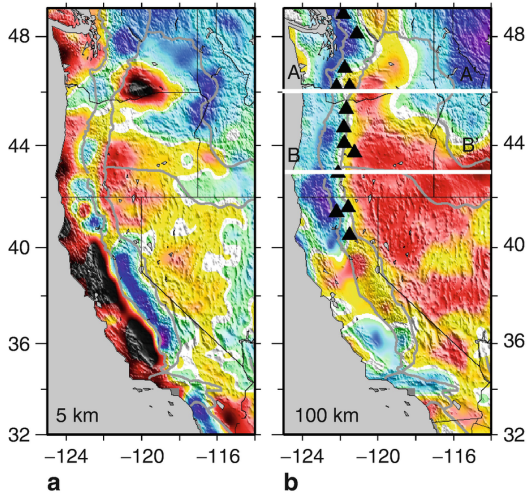
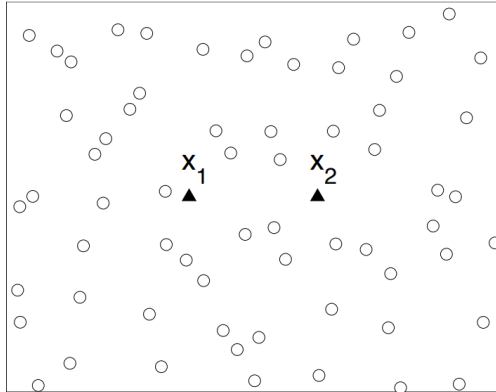
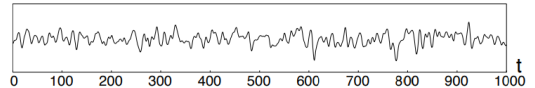
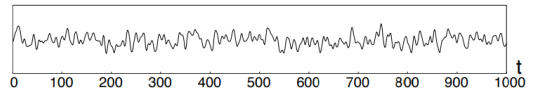
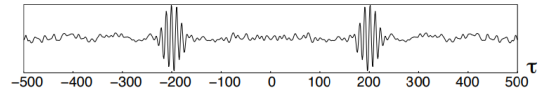


Image credit: [CRS21]

piece of signal recorded at x_1 piece of signal recorded at x_2 cross correlation $x_1 - x_2$ 

Cross correlation

Consider an array of N passive sensors at position $\{x_j\}_{j=1}^N$, wave fields $u(t, x_j)$ recorded in the time interval $[0, T]$, $T > 0$, then we write the cross correlation matrix for time lag τ as:

$$C_T(\tau, x_j, x_l) = \frac{1}{T} \int_0^T u(t, x_j) u(t + \tau, x_l) dt, \quad j, l = \{1, \dots, N\}, \tau \in \mathbb{R} \quad (1)$$

For space-time stationary and "delta"-correlated noise sources, we get [GP16]:

Green's function proportionality

Denoting by G the Green's function of the scalar wave equation, it is possible to arrive at the following relation:

$$\frac{\partial}{\partial \tau} C_T(\tau, x_1, x_2) \approx - [G(-\tau, x_1, x_2) - G(\tau, x_1, x_2)] \quad (2)$$

The travel time can be inferred from (2) by considering the singular support of the right-hand side.

The setup under consideration

Underlying wave equation

We assume the measured data at sensor positions $\{x_j\}_{j=1}^2$ originates from the following wave equation

$$\begin{aligned} \frac{1}{c^2} \frac{\partial^2 u}{\partial t^2} - u_{xx} &= 0, & x \in \mathbb{R}, & \quad t \in (0, T) \\ u(0, x) &= 0, & \partial_t u(0, x) &= f(x), & x \in \mathbb{R} \end{aligned} \tag{3}$$

where $c > 0$ is the wave speed and f the initial speed.

Solution space

We seek f in the solution set \mathcal{Q}_L defined as

$$\mathcal{Q}_L := \{f \mid f \in C(\mathbb{R}), \text{supp } f \subseteq [0, 1], |f| < L\}$$

for some constant $L > 0$.

Definition (The inverse source problem)

Let $f \in Q_L$, fix $x_1 = \tau = 0$, then the data, at the given values for any combination of $c, T \in \mathbf{R}^+$, is defined as originating from:

$$d(x) := \partial_\tau C_T(0, 0, x) + c \partial_x C_T(0, 0, x) = \frac{1}{2cT} \int_0^T \int_{-ct}^{ct} f(s) ds f(x + ct) dt \quad (4)$$

and the inverse source problem becomes

Given the data, $d(x)$, seen in (4) recover the initial speed, $f \in Q_L$, of the wave equation (3).

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The inverse source problem in 1d

Manipulating the data equation

We can express the data as follows:

$$d(x) = \frac{1}{2cT} \int_0^T [F(ct) - F(-ct)] F'(x + ct) dt, \quad F(q) = \int_0^q f(s) ds$$

as $c > 0$, $T > 0$ with the compact support of f in $[0, 1]$ the term $F(-ct)$ vanishes, reducing the above to:

$$d(x) = \frac{1}{2cT} \int_0^T F(ct) F'(x + ct) dt = \frac{1}{2c^2T} \int_0^{cT} F(q) F'(x + q) dq, \quad q = ct$$

A direct application of integration by parts shows the data is equivalent to

$$d(x) = \frac{F(cT)}{2c^2T} F(x + cT) - \frac{1}{2c^2T} \int_0^{cT} F'(q) F(x + q) dq$$

The inverse source problem in 1d

Fourier domain and mean estimation

Data in the Fourier domain

$$\mathcal{F}[d'(x)](\omega) = \frac{F(cT)}{2c^2T} e^{2\pi i c T \omega} \hat{f}(\omega) - \frac{1}{2c^2T} |\hat{f}(\omega)|^2, \quad cT \geq 1,$$

where \hat{f} is the Fourier transform of f .

Remark: It should be noted that $F(cT) = M_x[f]$ is the mean of f by the definition of F and the assumption of the support of f . From the above relation it is apparent $F(cT)$ is important to know a priori and hence something that is recovered.

Lemma (Mean of the source from data)

Assume access to the data $d(x)$, and the sensor signal $u_0(t) = u(t, 0)$, then the mean $M_x[f]$ can be calculated as:

$$M_x[f] = \frac{2c \text{Leb}(\text{supp } d) M_x[d]}{M_t[u_0]}$$

The inverse source problem in 1d

Zero mean, non-uniqueness. and phase-retrieval

Now, assume $M_x[f] = 0$ and introduce $\widetilde{f}(x) = f(-x)$, then

$$2c^2 T d'(x) = -\left(f * \widetilde{f}\right)(-x)$$

with $*$ the usual convolution operator. Applying the Fourier transform we find:

$$2i\omega c^2 T \hat{d}(\omega) = -\hat{f}(-\omega) \widehat{\widetilde{f}}(-\omega) = -\hat{f}(\omega) \hat{f}(-\omega)$$

since f is real-valued, we have $\hat{f}(-\omega) = \overline{\hat{f}(\omega)}$, the complex conjugate, which yields the phase retrieval problem

$$-2i\omega c^2 T \hat{d}(\omega) = |\hat{f}(\omega)|^2$$

Phase retrieval

Hence, in the case where the mean of the source function is zero, we are faced with the following problem

Reconstruct f from $|\hat{f}|$

Theorem (Akwowicz [Aku57])

Let $f, g \in \mathcal{Q}_L$, and denote by \hat{f} and \hat{g} the Fourier transform. Then \hat{f} and \hat{g} admits a holomorphic extensions, F and G to all of \mathbb{C} . If $|\hat{f}| = |\hat{g}|$ then $\{z_n\} \cup \{\bar{z}_n\} = \{z'_n\} \cup \{\bar{z}'_n\}$ where $\{z_n\} \cup \{\bar{z}_n\}$ and $\{z'_n\} \cup \{\bar{z}'_n\}$ are the sets of zeros of F and G respectively. This implies

$$\hat{g}(z) = \hat{h}(z)e^{i\theta z}\gamma, \quad \theta \in \mathbb{R}, \gamma \in \mathbb{C}, |\gamma| = 1$$

for some function $h \in \mathcal{E}'(\mathbb{R})$ completely determined by $\{z_n\} \cup \{\bar{z}_n\}$.

Lemma

Let $f \in \mathcal{Q}_L$, $M_x[f] = 0$, and assume $cT \geq 1$ the non-linear inverse source problem based on the data seen in (4) is ill-posed. Furthermore, it is a phase retrieval problem.

The inverse source problem in 1d

Non-zero reconstruction theorem

Theorem

Let $f \in Q_L(\mathbb{R})$, $M_x[f] \neq 0$, assume $cT \geq 1$ put $\mathcal{G}(\omega) = 2c^2 T \mathcal{F}[d'(x)](\omega)$ and $\alpha(\omega) = E_{cT}(\omega)$ then the non-linear inverse source problem based on the data seen in (4) can be reconstructed from:

$$\Re \hat{f} = \frac{\pm \sqrt{\kappa^4 - 4\kappa^2 \Re \mathcal{G} - 4(\Im \mathcal{G})^2} |\Re \alpha| + \Re \alpha \kappa^2 + 2\Im \alpha \Im \mathcal{G}}{2\kappa}$$

$$\Im \hat{f} = \frac{-\kappa \Im \alpha \Re \hat{f} - \Im \mathcal{G}}{\kappa \Re \alpha}$$

where $\kappa = F(cT)$. Furthermore, we obtain the following stability estimate

$$\|f_1 - f_2\|_{L^2(\mathbb{R})} \leq \kappa^{-1} C(L) \left(2\kappa^2 + 1\right) |\mathcal{G}_1 - \mathcal{G}_2|_{L^\infty(\mathbb{R})}^{1/2}$$

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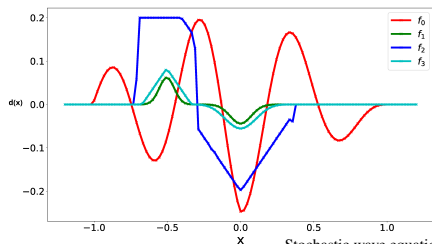
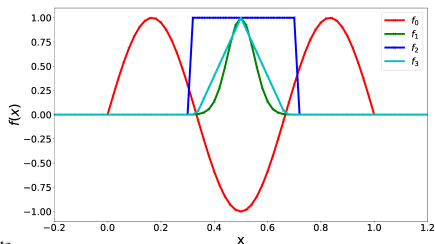
Example of numerical sources

$$f_0(x) = \sin(3\pi x) \mathbb{1}_A,$$

$$f_2(x) = \begin{cases} 1, & 0.3 < x < 0.7 \\ 0, & \text{otherwise} \end{cases},$$

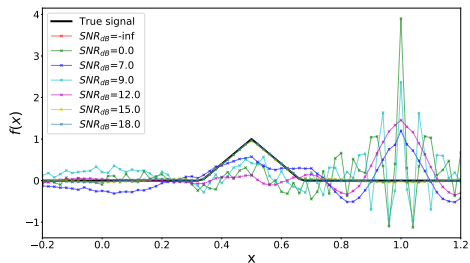
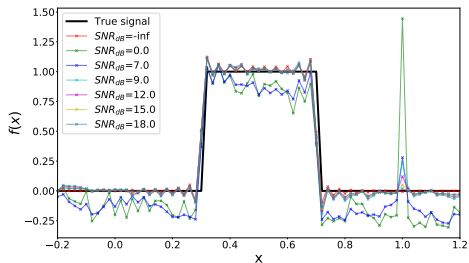
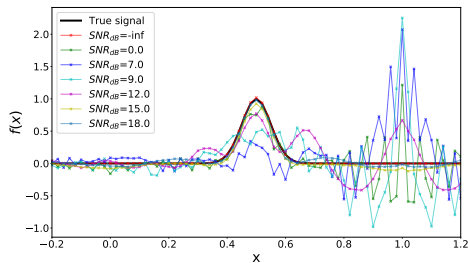
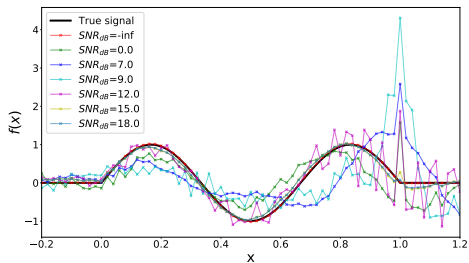
$$f_1(x) = e^{-\frac{1}{2}\left(\frac{x-1/2}{1/20}\right)^2} \mathbb{1}_A,$$

$$f_3(x) = \begin{cases} 0, & x < 1/3 \\ 6(x - 1/3), & x \geq 1/3 \wedge x < 1/2 \\ -6(x - 2/3), & x \geq 1/2 \wedge x < 2/3, \\ 0, & x \geq 2/3 \end{cases}$$



Numerical experiments

Reconstructions



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Conclusion and future work

Conclusion and future work

Conclusion:

- Proved non-uniqueness in the case of a source with mean zero, using results from phase retrieval in one dimension.
- Derived a reconstruction method and stability estimate where the source has a non-zero mean in one dimension.
- Numerical proof of concept using simulated data in one dimension.

Future work

- Extend to higher dimensions (in progress)
- Consider the stochastic setting in terms of initial conditions and time fluctuations (in progress)
- Possible real applications, e.g. within seismology and astronomy.

References

- [Aku57] Edwin J. Akutowicz. “On the Determination of the Phase of a Fourier Integral, II”. In: *Proceedings of the American Mathematical Society* 8.2 (1957), pp. 234–238.
- [BCJ01] N. Bleistein, J. Cohen, and John Jr. *Mathematics of Multidimensional Seismic Imaging, Migration, and Inversion*. Vol. 13. Jan. 2001. ISBN: 978-1-4612-6514-6.
- [GP09] Josselin Garnier and George Papanicolaou. “Passive Sensor Imaging Using Cross Correlations of Noisy Signals in a Scattering Medium”. In: *SIAM Journal on Imaging Sciences* 2.2 (2009), pp. 396–437.
- [GP16] J. Garnier and G. Papanicolaou. *Passive Imaging with Ambient Noise*. Cambridge Monographs on Applied Mathematics. Cambridge University Press, 2016. ISBN: 9781107135635.
- [Fou+18] Jacques Fournier et al. “Correlation-based imaging of fast moving objects using a sparse network of passive receivers”. In: (2018), pp. 1618–1622.
- [Mos+20] Miguel Moscoso et al. “The Noise Collector for sparse recovery in high dimensions”. In: *Proceedings of the National Academy of Sciences* 117.21 (2020), pp. 11226–11232.
- [CRS21] Michel Campillo, Philippe Roux, and Nikolai M. Shapiro. “Seismic, Ambient Noise Correlation”. In: *Encyclopedia of Solid Earth Geophysics*. Ed. by Harsh K. Gupta. Cham: Springer International Publishing, 2021, pp. 1557–1562. ISBN: 978-3-030-58631-7.
- [Gar21] Josselin Garnier. “Passive Communication with Ambient Noise”. In: *SIAM Journal on Applied Mathematics* 81.3 (2021), pp. 814–833.
- [Tou+21] Rita Touma et al. “A distortion matrix framework for high-resolution passive seismic 3-D imaging: application to the San Jacinto fault zone, California”. In: *Geophysical Journal International* 226.2 (Mar. 2021), pp. 780–794.

Wave equation

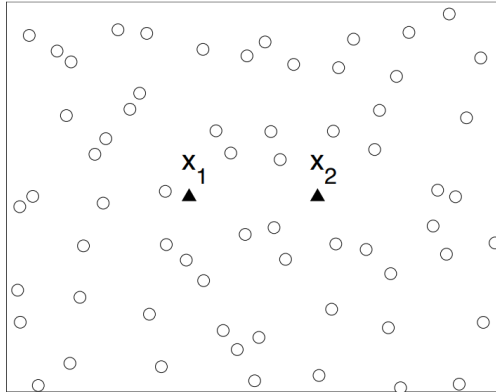
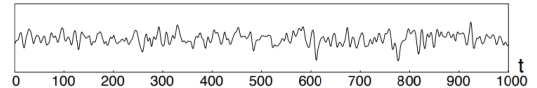
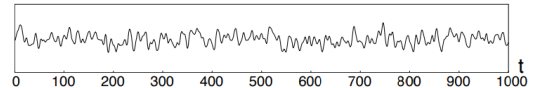
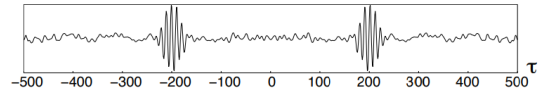
$$\frac{1}{c^2(x)} \frac{\partial^2 u}{\partial t^2} - \Delta_x u = n(t, x), \quad x \in \mathbf{R}^n, \quad t \in \mathbf{R}^+ \quad *$$

Green's function

$$\frac{1}{c^2(x)} \frac{\partial^2 G}{\partial t^2} - \Delta_x G = \delta(t)\delta(x - y), \quad x, y \in \mathbf{R}^n, \quad t \in \mathbf{R}$$
$$G(t, x, y) = 0, \quad \text{for all } t < 0$$

Solution (Causal)

$$u(t, x) = \int_{-\infty}^t \int_{\mathbf{R}^n} G(t - s, x, y) n(s, y) dy ds$$

piece of signal recorded at x_1 piece of signal recorded at x_2 cross correlation $x_1 - x_2$ 

Singular support

Consider any distribution $f \in \mathcal{D}'(\Omega)$ for $\Omega \subseteq \mathbf{R}^n$, then we define singular support as:

$$\text{singsupp}(f) = \{x \in \Omega, f \text{ is } C^\infty \text{ near } x\}^c$$

Derivative and singular support

For any differential operator, D , and distribution, f , that;

$$\text{singsupp}(Df) \subseteq \text{singsupp}(f)$$