

Compressed Modes for Material Interface Problems

Russel Caflisch

IPAM

Mathematics Department, UCLA

math changes everything.

Collaborators

- Edward Chou (UCLA)
- Stan Osher (UCLA)
- Vidvuds Ozolins (UCLA Materials Sci)
- Omer Tekin (UCLA→Google)

Outline

- Compressed Sensing
- Phase Transitions
- Deposition

- Problem statement
 - Donoho 2006, Candes, Romber & Tao 2006
 - Find x that is m -sparse and solves $Ax = f$
 - Assuming that an m -sparse solution exists

- Standard methods

$$\min \|x\|_0 \text{ subject to constraint } Ax = f$$

- note $\|x\|_0 = \#\{i : x_i \neq 0\}$

- Compressed sensing

$$\min \|x\|_1 \text{ subject to constraint } Ax = f$$

- note $\|x\|_1 = \sum_{i=1}^N |x_i|$

How many measurements are required?

- For $m \ll N$, find m -sparse solution $x \in R^N$ of
$$Ax = f \in R^n \quad A \text{ is } n \times N$$
- Standard methods require: $n = N$
 - #(equations)=#(unknowns), NP hard = intractable
- Compressed sensing: $n = m (\log N)$
 - $n \ll N$. Many fewer equations than unknowns!
 - Solution is exact with high probability!
 - Reduced isometry property (RIP)
 - convex programming, tractable and fast

Why Does L^1 Promote Sparsity?

- Compressed sensing

$$\min \|x\|_1 \text{ subject to constraint } Ax = f$$

- Simplified problems: Find x solving

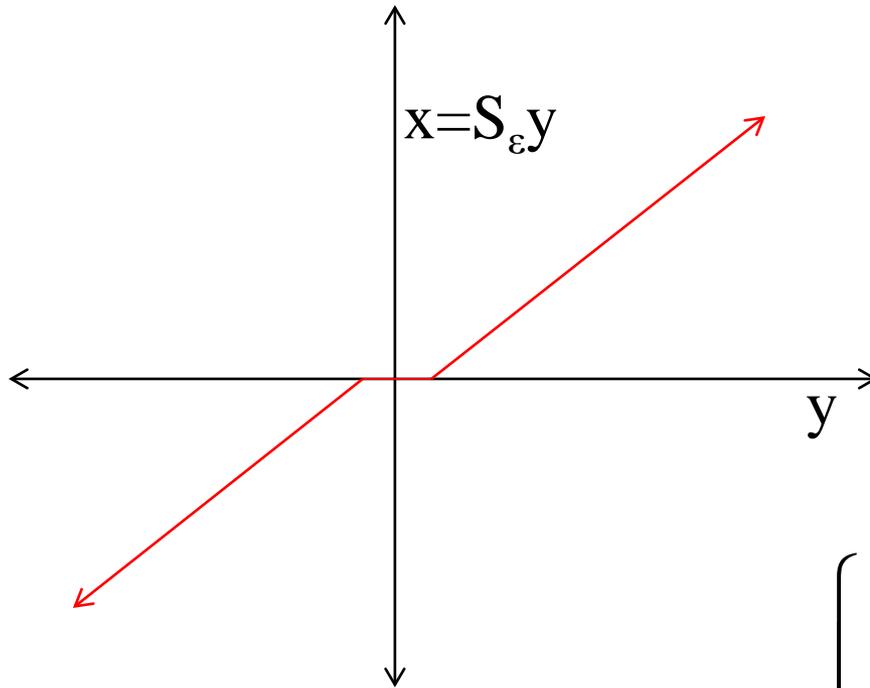
- $\min (x - y)^2 + \varepsilon |x|$ for given y in \mathbb{R}

- Solution

$$x = S_\varepsilon y = \begin{cases} y - \varepsilon & \text{if } y > \varepsilon \\ 0 & \text{if } |y| < \varepsilon \\ y + \varepsilon & \text{if } y < -\varepsilon \end{cases}$$

- Operator S_ε is “soft-thresholding”

Soft Thresholding



$$S_\epsilon y = \begin{cases} y - \epsilon & \text{if } y > \epsilon \\ 0 & \text{if } |y| < \epsilon \\ y + \epsilon & \text{if } y < -\epsilon \end{cases}$$

Subgradient

- Euler-Lagrange eqtn for $\min (x - y)^2 + \varepsilon |x|$

$$2(x - y) + \varepsilon p(x) = 0$$

- Subgradient

$$p(x) = \partial_x |x| = \begin{cases} 1 & \text{if } x > 0 \\ [-1, 1] & \text{if } x = 0 \\ -1 & \text{if } x < 0 \end{cases}$$

- Write $\text{sgn}(x) = p(x)$

COMPRESSION FOR PHASE TRANSITIONS

Phase Transitions

- Cahn-Hilliard equations

$$\partial_t c = -\partial_x^2 (\partial_c J[c])$$

$$J_0[c] = \int \frac{1}{2} \gamma c_x^2 + \frac{1}{4} (c^2 - 1)^2 dx$$

- Pure phase regions $c=-1$ and $c=1$
- Steady eqtn $\partial_c J[c]=0$ has solution

$$c(x) = \tanh(x / \sqrt{2\gamma})$$

- Goals of compression

- $c=-1$ and $c=1$ reached at finite distance
- Similar to classification problem in machine learning

Compression for Cahn-Hilliard

- Variational quantity for Cahn-Hilliard

$$J_0[c] = \int \frac{1}{2} \gamma c_x^2 + \frac{1}{4} (c^2 - 1)^2 dx$$

- Variational quantity with compression

$$J[c] = \int \frac{1}{2} \gamma c_x^2 + \frac{1}{4} (c^2 - 1)^2 + \varepsilon |c^2 - 1| dx$$

- Euler-Lagrange for steady minimizer

$$-\gamma c_{xx} + c(c^2 - 1) + 2\varepsilon c \operatorname{sgn}(c^2 - 1) = 0$$

- Look for solution that is odd in x and reaches ± 1 at $x = \pm L$

$$c(x) = -1 \quad x \leq -L$$

$$|c(x)| < 1 \quad |x| < L$$

$$c(x) = 1 \quad x \geq L$$

- ODE and BCs

$$-\gamma c_{xx} + c(c^2 - 1) - 2\varepsilon c = 0 \quad |x| < L$$

$$c(\pm L) = \pm 1, \quad c_x(\pm L) = 0$$

Numerical Method for Compressed Cahn-Hilliard

- Numerical Method: Split Bregman

$$\phi(c) = \varepsilon(c^2 - 1)$$

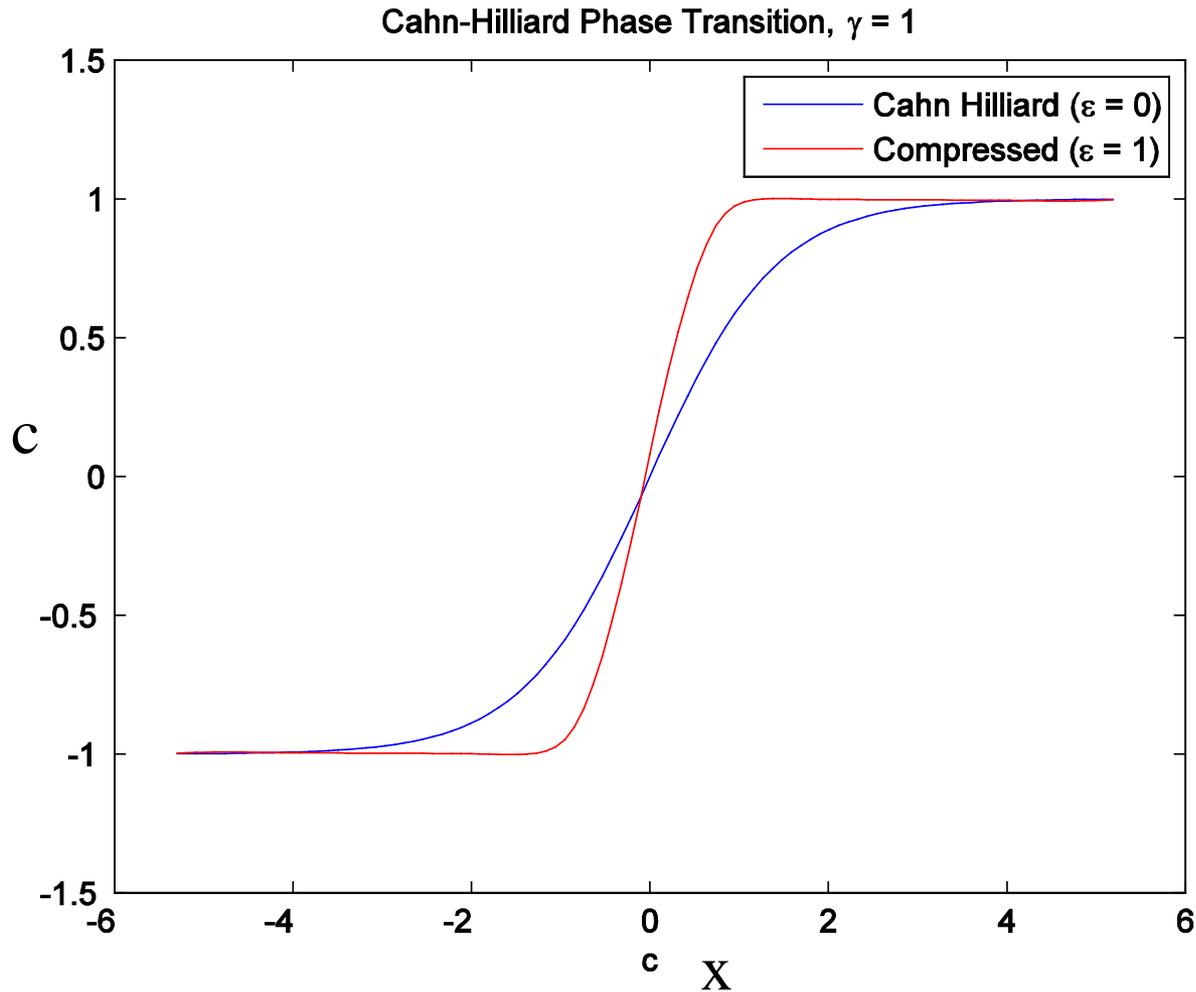
$$c^{(k+1)} = \min_u J_0[c] + \frac{\lambda}{2} \int d^{(k)} - \phi(c) - b^{(k)} dx$$

$$d^{(k+1)} = \min_d \|d\|_1 + \frac{\lambda}{2} \int d^{(k)} - \phi(c^{(k+1)}) - b^{(k)} dx$$

$$b^{(k+1)} = b^{(k)} + \phi(c^{(k+1)}) - d^{(k+1)}$$

- Use gradient descent to update c , soft-thresholding to update d
- λ is a fixed parameter, influences convergence but not solution

Numerical Solution for Compressed Cahn-Hilliard



COMPRESSION FOR DEPOSITION

Simplified Deposition Model

- Deposition model

- $u(x)$ = height of deposited material

- Diffusion with source $s(x)$ and removal rate λ

$$\partial_t u = \partial_x^2 u - \lambda u + s \quad -\pi < x < \pi$$

- Steady state problem

$$u = Gs \quad G = (-\partial_x^2 + \lambda)^{-1}$$

- Desired height profile $f(x) = \cos(x)$

- Choose $s(x)$ so that $\int_{-\pi}^{\pi} (u - f(x))^2 dx \leq \sigma^2$

- Goals of compression

- Choose s which is smooth and has small support

- Variational quantity

- $J[s] = \int_{-\pi}^{\pi} \frac{1}{2} (u - f(x))^2 + \varepsilon |s| + \frac{1}{2} \gamma^2 s_x^2 dx$ with $u = Gs$

$$= \int_{-\pi}^{\pi} \frac{1}{2} (Gs - f(x))^2 + \varepsilon |s| + \frac{1}{2} \gamma^2 s_x^2 dx$$

- Euler-Lagrange eqtn

$$G^2 s - Gf + \varepsilon \operatorname{sgn}(s) - \gamma^2 s_{xx} = 0$$

- Solution s doesn't have compact support (?)

- Modified variational quantity

$$J[s] = \int_{-\pi}^{\pi} \frac{1}{2} (Gs - f(x))^2 + \varepsilon G^2 |s| + \frac{1}{2} \gamma^2 (Gs_x)^2 dx$$

- Euler-Lagrange eqtn

$$G^2 s - Gf + \varepsilon G^2 \operatorname{sgn}(s) - \gamma^2 G^2 s_{xx} = 0$$

$$\longrightarrow s - G^{-1} f + \varepsilon \operatorname{sgn}(s) - \gamma^2 s_{xx} = 0$$

- Variational quantity

- **Original** $J[s] = \int_{-\pi}^{\pi} \frac{1}{2} (Gs - f(x))^2 + \varepsilon |s| + \frac{1}{2} \gamma^2 s_x^2 dx$

- **Modified** $J_m[s] = \int_{-\pi}^{\pi} \frac{1}{2} (Gs - f(x))^2 + \varepsilon G^2 |s| + \frac{1}{2} \gamma^2 (Gs_x)^2 dx$

$$\min J_m[s] \Rightarrow s - g + \varepsilon \operatorname{sgn}(s) - \gamma^2 s_{xx} = 0 \quad g = G^{-1} f$$

- **Equivalent** $J_{\varepsilon, \gamma}[s] = \int_{-\pi}^{\pi} \frac{1}{2} (s - g)^2 + \varepsilon |s| + \frac{1}{2} \gamma^2 s_x^2 dx$

- Thresholding

- For $\gamma=0$, $\min J_{\varepsilon, 0}[s] \Rightarrow s = S_{\varepsilon} g$ **soft-thresholding**

- For $\gamma>0$, $\min J_{\varepsilon, \gamma}[s] \Rightarrow s = S_{\varepsilon, \gamma} g$ **“smooth-thresholding”**

- Equation for s

- $s - G^{-1}f + \varepsilon \operatorname{sgn}(s) - \gamma^2 s_{xx} = 0$ periodic on $|x| < \pi$, $f = \cos(x)$

- Look for solution s to be nonnegative and even for $|x| < \pi/2$

- with $s(x) > 0 \quad |x| < L$

- $s(x) = 0 \quad L < |x| < \pi/2$

- Antisymmetric around $\pi/2$

- Equation for s on $|x| < L$ (using $s > 0$)

$$s - G^{-1}f + \varepsilon - \gamma^2 s_{xx} = 0 \qquad s(L) = s_x(L) = 0$$

- **Solution** $s(x) = a + b \cos(x) + c \cos(x/\gamma)$

- $a = -\varepsilon$ and $a + b \cos(L) - c\gamma^2 \cos(L/\gamma) = 0$

- $b = \frac{1 + \lambda}{1 + \gamma^2}$ and $-b \sin(L) + c\gamma \sin(L/\gamma) = 0$

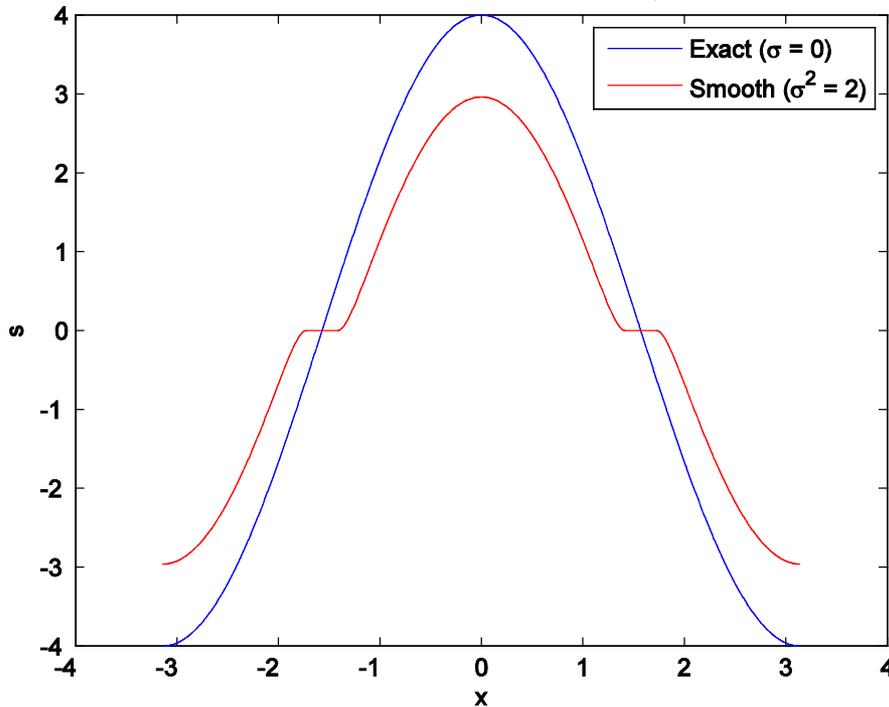
2 eqtns for c, L

Numerical Solution for Compressed Deposition

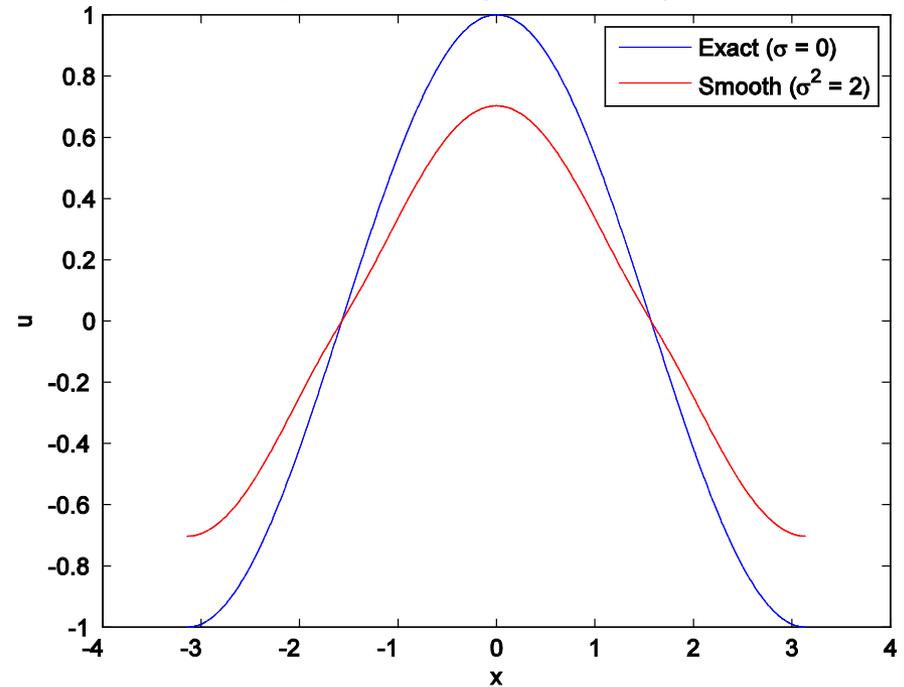
- The analytic solution, for σ large enough, takes the form

$$s(x) = c_1 + c_2 \cos(x) + c_3 \cosh(kx)$$

Deposition Model Height, $\lambda = 3$, $\varepsilon = 1$, $\gamma = 0.01$

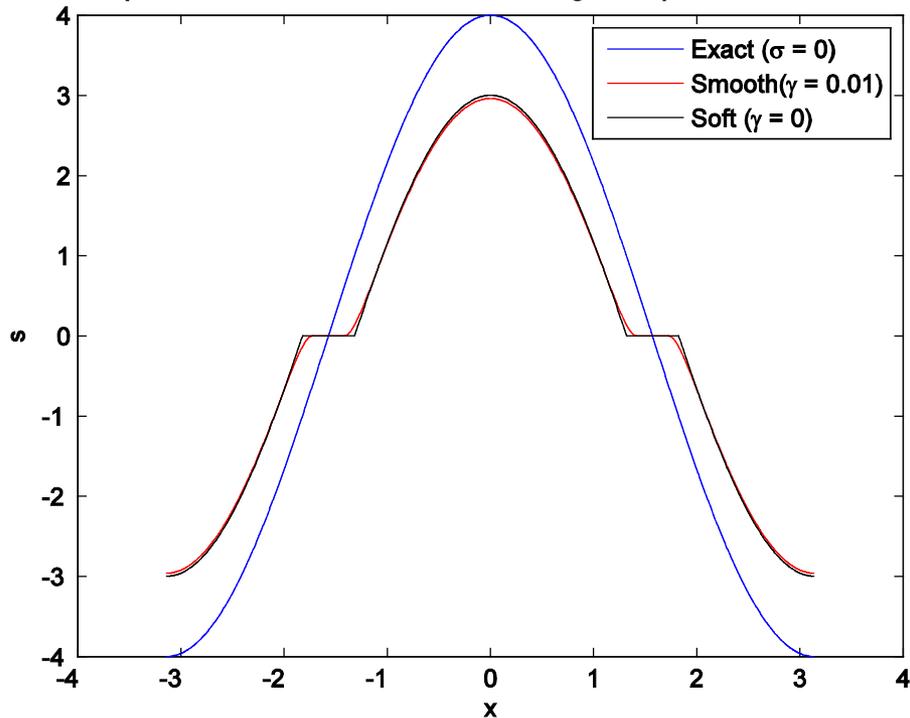


Deposition Model Height, $\lambda = 3$, $\varepsilon = 1$, $\gamma = 0.01$

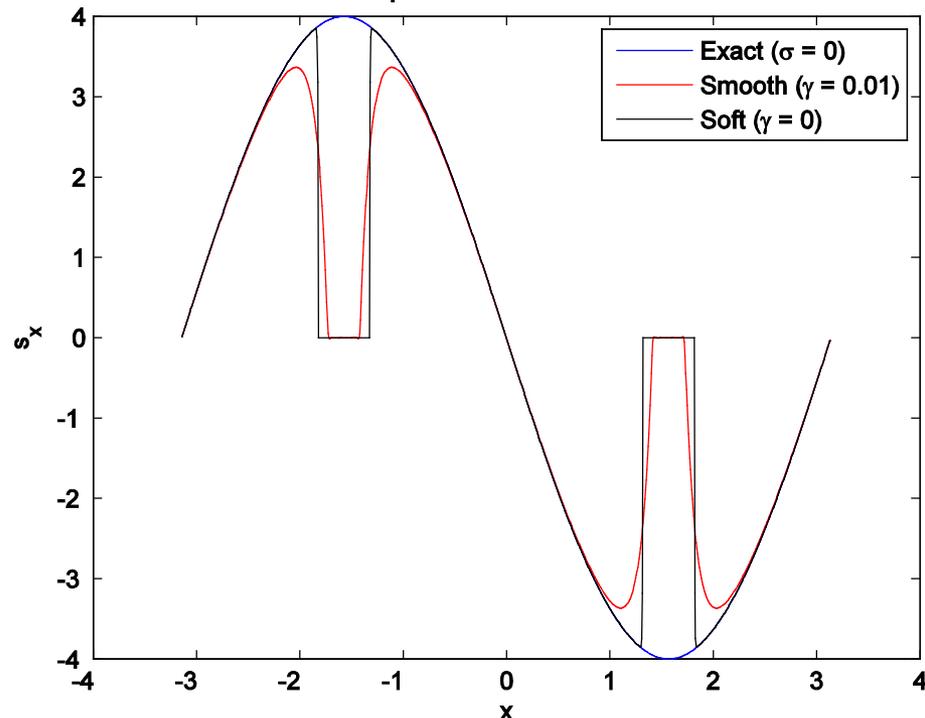


Comparison of Soft and Smooth Thresholding

Comparison of Soft and Smooth Thresholding for Deposition Model Source



Comparison of First Derivative



Ongoing and Future Work

- Compression of Schrodinger equation
 - Wannier modes
 - Application to DFT, including symmetries
- Signal fragmentation
 - Decompose a desired signal into a sum of fragments
 - Fragments have compact support
- Multiscale method
 - Use compression to mediate interaction between micro and macro scales