

- Paper Title: Iterative Hard Thresholding for Compressed Sensing
- Author List: Thomas Blumensath and Mike E. Davies
- Venue: University of Southampton
- Year of Publication: 2008

1. Iterative Hard Thresholding algorithm (IHT_s)

$$\mathbf{x} = \boldsymbol{\phi} \mathbf{y} + \mathbf{e}$$

where \mathbf{e} is the observation noise and $\boldsymbol{\phi}$ is the sampling matrix.

Let $y^{[i]}$ be the value of desired vector y during the i^{th} iteration of the algorithm.

Pseudo-Code:

- $y^{[0]} = 0$
- Update y as $y^{[n+1]} = H_s \left(y^{[n]} + \boldsymbol{\phi}^T (\mathbf{x} - \boldsymbol{\phi} y^{[n]}) \right)$
- $H_s(\mathbf{a})$ is the non-linear operator that sets all but the largest (in magnitude) s elements of \mathbf{a} to zero. If there is no unique such set, a set can be selected either randomly or based on a predefined ordering of the elements

2. Key Theorem (stated as it is for the most part)

- A matrix $\boldsymbol{\phi}$ satisfies the Restricted Isometry Condition (RIP) if

$$(1 - \delta_s) \|\mathbf{y}\|_2^2 \leq \|\boldsymbol{\phi} \mathbf{y}\|_2^2 \leq (1 + \delta_s) \|\mathbf{y}\|_2^2$$

- Instead of using the above restricted isometry property, we will use a re-scaled matrix $\boldsymbol{\phi} = \frac{\boldsymbol{\phi}'}{1 + \delta_s}$, which satisfies the following non-symmetric isometry property, which is equivalent to the RIP defined above.

$$(1 - \beta_s) \|\mathbf{y}\|_2^2 \leq \|\boldsymbol{\phi} \mathbf{y}\|_2^2 \leq \|\mathbf{y}\|_2^2, \quad \beta_s = 1 - \frac{1 - \delta_s}{1 + \delta_s}$$

for all s -sparse y . We will say that for a matrix $\boldsymbol{\phi}$ the RIP holds for sparsity s , if $\beta_s < 1$.

- Theorem:* Given a noisy observation $\mathbf{x} = \boldsymbol{\phi} \mathbf{y} + \mathbf{e}$, where y is an arbitrary vector. Let y_s be an approximation to y with no-more than s non-zero elements for which $\|\mathbf{y} - y_s\|_2$ is minimal. If $\boldsymbol{\phi}$ has restricted isometry property with $\beta_{3s} < \frac{1}{8}$, then, at iteration k , IHT_s will recover an approximation $y^{[k]}$ satisfying

$$\|\mathbf{y} - y_s\|_2 \leq 2^{-k} \|\mathbf{y}_s\|_2 + 5\epsilon_s^{\sim}$$

$$\text{where } \epsilon_s^{\sim} = \|\mathbf{y} - y_s\|_2 + \frac{1}{\sqrt{s}} \|\mathbf{y} - y_s\|_1 + \|\mathbf{e}\|_2$$

- Furthermore, after at most

$$k^* = \lceil \log_2 \left(\frac{\|\mathbf{y}_s\|_2}{\epsilon_s^{\sim}} \right) \rceil$$

iterations, IHT_s estimates y with accuracy

a. $\|y - y_{k^*}\|_2 \leq 6 \left(\|y - y_s\|_2 + \frac{1}{\sqrt{s}} \|y - y_s\|_1 + \|e\|_2 \right)$