

The Solvability Complexity Index of Sampling-based Hilbert Transform Approximations

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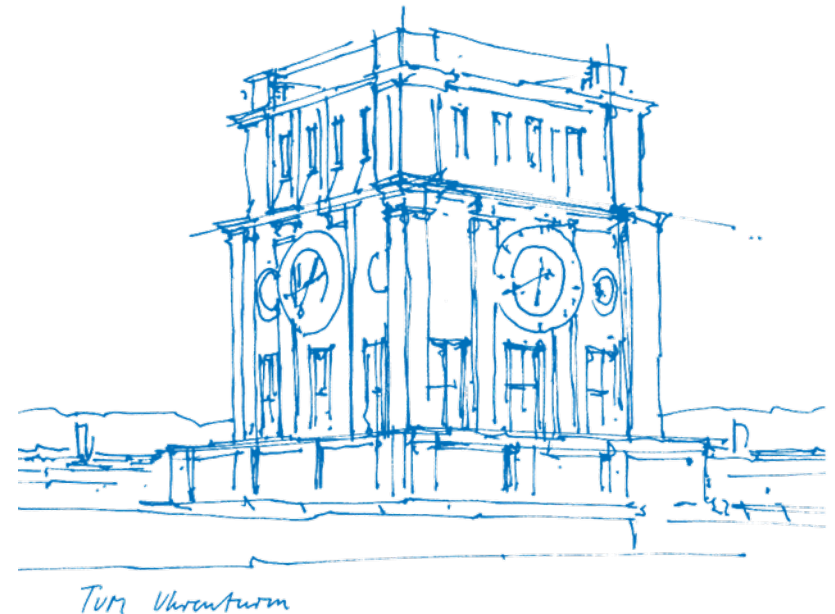
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Outline of this Talk

1. The Solvability Complexity Index and Towers of Algorithms
2. The Hilbert Transform and Sampling-based Hilbert Transform Approximations
3. Apply the the SCI Framework to the Hilbert Transform
4. Conlusions

The Solvability Complexity Index

The Solvability Complexity Index

- ▶ The **Solvability Complexity Index (SCI)** is a new measure to characterize the complexity of certain computational problems.
 - 📄 J. Ben-Artzi, A. C. Hansen, O. Nevanlinna, and M. Seidel, “Can everything be computed? - On the Solvability Complexity Index and towers of algorithms.” *preprint: arXiv:1508.03280v1*, Aug. 2015.
 - 📄 J. Ben-Artzi, A. C. Hansen, O. Nevanlinna, and M. Seidel, “New barriers in complexity theory: On the solvability complexity index and the towers of algorithms,” *Comptes Rendus Mathematique*, vol. 353, no. 10, pp. 931–936, 2015.
- ▶ Solving computational problems by **towers of algorithms**.
- ▶ The SCI is the **minimal height of such a tower**.
- ▶ The SCI counts the **minimum number of limits** necessary for computing a desired quantity.

The Solvability Complexity Index – Polynomials

- ▷ Let \mathcal{P}_N set of polynomials of degree $N \in \mathbb{N}$.
- ▷ Can we find an algorithm to compute the roots of every $p \in \mathcal{P}_N$ using only **finitely** many arithmetic operations?
 - **Yes!** for $N \leq 4$
 - **No!** for $N \geq 5$

- ▷ What if we allow for a limiting process?

Does there exists a sequence of operations $\{\Gamma_n : \mathcal{P}_N \rightarrow \mathbb{C}\}$ where each Γ_n involves only finitely many arithmetic operations such that

$$\lim_{n \rightarrow \infty} \Gamma_n(p) = z_0$$

is a zero of p ?

- **Yes!** for all $N \in \mathbb{N}$
- **Note:** Newton's method will not work.

An there is no „purely iterative generally convergent algorithm“ for root finding.

 S. Smale, “The fundamental theorem of algebra and complexity theory,” *Bull. Amer. Math. Soc.*, vol. 4, 1981.

 P. Doyle and C. McMullen, “Solving the quintic by iteration,” *Acta Math.*, vol. 163, pp. 151–180, 1989.

The Solvability Complexity Index

▷ **There are important computational problems which can not be solved by one limiting process!**



- „purely iterative generally convergent algorithm“ for root finding.
- ...
- calculation of the Hilbert transform from samples.

▷ What if we allow for a several limiting processes?

Does there exist a sequence $\{\Gamma_{n_k, \dots, n_1}\}$ of operations depending on several indices $n_k, \dots, n_1 \in \mathbb{N}$ such that

$$\lim_{n_1 \rightarrow \infty} \dots \lim_{n_k \rightarrow \infty} \Gamma_{n_k, \dots, n_1}(p) = \Xi(p)$$

converges to the desired result $\Xi(p)$?

-  H. Boche and V. Pohl, “On the calculation of the Hilbert transform from interpolated data,” *IEEE Trans. Inf. Theory*, vol. 54, no. 5, pp. 2358–2366, May 2008.
-  H. Boche and V. Pohl, “Calculating the Hilbert transform on spaces with energy concentration: Convergence and divergence regions,” *IEEE Trans. Inf. Theory*, vol. 65, no. 1, pp. 586–603, Jan. 2019.

The SCI Framework

The SCI framework, is based on the following basic objects

Ω a **primary set**, is the set of objects on which we perform the computations

\mathcal{M} a (pseudo) **metric space**

Ξ a **problem function**, is the mapping $\Xi : \Omega \rightarrow \mathcal{M}$ which we want to compute for our objects in Ω

Λ is the **evaluation set**, that is a set of (measurement) functionals $\lambda : \Omega \rightarrow \mathbb{C}$

Definition (Computational problem)

A collection $\{\Xi, \Omega, \mathcal{M}, \Lambda\}$ is said to be a *computational problem*.

Example ($\{\mathcal{C}(\mathbb{T}), \mathbb{R}, \|\cdot\|_\infty, \mathbb{T}\}$ – „Sampling-based computational problem“)

- $\Omega = \mathcal{C}(\mathbb{T})$, the Banach space of continuous functions on a compact interval $\mathbb{T} \subset \mathbb{R}$
- $\mathcal{M} = \mathbb{R}$ the real numbers.
- $\Xi : \mathcal{C}(\mathbb{T}) \rightarrow \mathbb{R}$ is given by $\Xi(f) = \|f\|_\infty = \max_{t \in \mathbb{T}} |f(t)|$
- $\Lambda = \{\lambda_\tau : \tau \in \mathbb{T}\}$ consists of all point evaluations

$$\lambda_\tau(f) = f(\tau) \quad \text{for all } f \in \mathcal{C}(\mathbb{T}).$$

Fundamental Algorithm

- ▶ Let $\{\Xi, \Omega, \mathcal{M}, \Lambda\}$ be a given computational problem.
- ▶ The problem of calculating $\Xi(f)$ will be done by a so called **tower of algorithms**.
- ▶ The next definition characterizes the algorithms at the lowest (the fundamental) level of the tower.

Definition (Fundamental Algorithm)

Let $\{\Xi, \Omega, \mathcal{M}, \Lambda\}$ be a computational problem. A mapping $\Gamma : \Omega \rightarrow \mathcal{M}$ is said to be a *fundamental algorithm* if for every $f \in \Omega$ the following conditions are satisfied:

- (i) There exists a **finite** subset $\Lambda_\Gamma(f) \subset \Lambda$ of evaluations.
- (ii) The action of Γ on f depends only on the values $\{\lambda(f) : \lambda \in \Lambda_\Gamma(f)\}$.
- (iii) For every $g \in \Omega$, with $\lambda(g) = \lambda(f)$ for all $\lambda \in \Lambda_\Gamma(f)$, holds $\Lambda_\Gamma(g) = \Lambda_\Gamma(f)$.

We say that Γ is a fundamental algorithm of type

- L** for "*linear*", if Γ is linear.
- G** for "*general*", if there is no further restriction on Γ .

Remark:

- The calculation of $\Gamma(f)$ is based on **finitely many measurements** of f .
- The evaluation set $\lambda_\Gamma(f)$ may be chosen **adaptively**, dependent on the actual object f .
- Condition (iii): well definiteness and consistence. It limits adaptivity.

Tower of Algorithms

Definition (Tower of Algorithms)

A *tower of algorithms* of type α and of height k for $\{\Xi, \Omega, \mathcal{M}, \Lambda\}$ is a family of sequences of operators

$$\begin{aligned}\Gamma_{n_k} &: \Omega \rightarrow \mathcal{M}, \\ \Gamma_{n_k, n_{k-1}} &: \Omega \rightarrow \mathcal{M}, \\ &\vdots \\ \Gamma_{n_k, n_{k-1}, \dots, n_1} &: \Omega \rightarrow \mathcal{M},\end{aligned}$$

where $n_k, \dots, n_1 \in \mathbb{N}$ and where the operators $\Gamma_{n_k, n_{k-1}, \dots, n_1}$ at the lowest level of the tower are fundamental algorithms of type α . Moreover, for every $f \in \Omega$ holds

$$\begin{aligned}\Xi(f) &= \lim_{n_k \rightarrow \infty} \Gamma_{n_k}(f) \\ \Gamma_{n_k}(f) &= \lim_{n_{k-1} \rightarrow \infty} \Gamma_{n_k, n_{k-1}}(f) \\ &\vdots \\ \Gamma_{n_k, \dots, n_2}(f) &= \lim_{n_1 \rightarrow \infty} \Gamma_{n_k, \dots, n_2, n_1}(f)\end{aligned}$$

where the convergence is always in the metric of \mathcal{M} .

Solvability Complexity Index

Definition (Solvability Complexity Index)

We say that a computational problem $\{\Xi, \Omega, \mathcal{M}, \Lambda\}$ has a *Solvability Complexity Index*

$$\text{SCI}(\Xi, \Omega, \mathcal{M}, \Lambda)_\alpha = k$$

with respect to towers of algorithms of type α if k is the smallest integer for which there exists a tower of algorithms of type α of height k . If no such tower exists then $\text{SCI}(\Xi, \Omega, \mathcal{M}, \Lambda)_\alpha = \infty$.

▷ We say that $\text{SCI}(\Xi, \Omega, \mathcal{M}, \Lambda)_\alpha = 0$, if $\Xi(f)$ can be calculate by finitely many algebraic operations.

Hilbert Transformation

The Hilbert Transform

- ▶ We consider continuous functions $f \in \mathcal{C}(\mathbb{T})$ on the interval $\mathbb{T} = [-\pi, \pi]$ with $f(-\pi) = f(\pi)$
- ▶ The **Fourier coefficients** of $f \in \mathcal{C}(\mathbb{T})$ are given by

$$c_k(f) = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(t) e^{-ikt} dt, \quad k = 0, \pm 1, \pm 2, \dots$$

- ▶ Assume that the **Fourier series** of f converges to f (in some sense)

$$f(t) = \sum_{k=-\infty}^{\infty} c_k(f) e^{ikt}, \quad t \in \mathbb{T}.$$

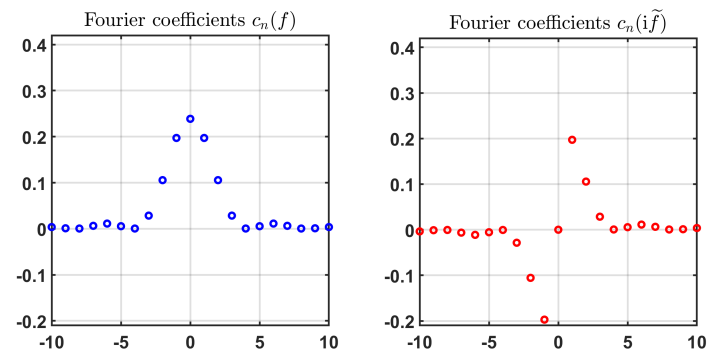
- ▶ The **conjugate function** associated with f is

$$\tilde{f}(t) = (Hf)(t) = -i \sum_{k=-\infty}^{\infty} \operatorname{sgn}(k) c_k(f) e^{ikt}, \quad t \in \mathbb{T} \quad \text{with} \quad \operatorname{sgn}(k) = \begin{cases} 0 & : k = 0 \\ \frac{k}{|k|} & : k \neq 0 \end{cases}$$

- ▶ The **Hilbert transform** is the mapping

$$H : f \mapsto \tilde{f}$$

- ▶ **Applications:** Causality, Kramers–Kronig relation, ...



Signal Spaces for the Hilbert Transform

- ▶ Let $f \in \mathcal{C}(\mathbb{T})$ with Fourier coefficients $\{c_k(f)\}_{k \in \mathbb{Z}}$.
- ▶ For $\beta \geq 0$, we define the seminorm

$$\|f\|_\beta = \left(\sum_{k \in \mathbb{Z}, k \neq 0} |k| (1 + \log |k|)^\beta |c_k(f)|^2 \right)^{1/2}.$$

- ▶ Therewith, we define for any $\beta \geq 0$ the Banach space

$$\mathcal{B}_\beta = \{f \in \mathcal{C}(\mathbb{T}) : \tilde{f} \in \mathcal{C}(\mathbb{T}) \text{ and } \|f\|_\beta < \infty\}$$

with the norm

$$\|f\|_{\mathcal{B}_\beta} = \max \left(\|f\|_{\mathcal{C}(\mathbb{T})}, \|\tilde{f}\|_{\mathcal{C}(\mathbb{T})}, \|f\|_\beta \right).$$

Remark

- For every $f \in \mathcal{B}_\beta$ the **Fourier series and the conjugate Fourier series converge uniformly** to f and \tilde{f} .
- The space $\mathcal{B}_{\beta=0}$ corresponds to the usual **Sobolev space** $H^{1/2}(\mathbb{T}) = W^{1/2,2}(\mathbb{T})$.
- The value $\|f\|_{\beta=0}^2$ corresponds to the **Dirichlet energy** of f .
- $\mathcal{B}_{\beta_2} \subset \mathcal{B}_{\beta_1} \subset \mathcal{B}_{\beta=0}$ for all $\beta_2 \geq \beta_1 \geq 0$.

No Sampling-based Hilbert Transform Approximation

Theorem

Let $\{\Gamma_n\}_{n \in \mathbb{N}}$ be a sequence of lower semicontinuous mappings $\Gamma_n : \mathcal{B}_\beta \rightarrow \mathcal{C}(\mathbb{T})$ such that to every $n \in \mathbb{N}$ there exists a finite sampling set $\mathcal{T}_n \subset \mathbb{T}$ such that

$$f_1(\tau) = f_2(\tau) \quad \text{for all } \tau \in \mathcal{T}_n \quad \text{implies} \quad \Gamma_n(f_1) = \Gamma_n(f_2) .$$

Then for every $0 \leq \beta \leq 1$ there exist functions $f \in \mathcal{B}_\beta$ such that

$$\limsup_{n \rightarrow \infty} \|\tilde{f} - \Gamma_n(f)\|_{\mathcal{C}(\mathbb{T})} > 0 .$$

- ▶ The operators $\Gamma_n : \mathcal{B}_\beta \rightarrow \mathcal{C}(\mathbb{T})$ can be non-linear.
- ▶ So the Hilbert transform on \mathcal{B}_β can not be calculated by a single limit (i.e. SCI > 1).
- ▶ So what is the SCI for calculating the Hilbert transform?

 H. Boche and V. Pohl, “Investigations on the approximability and computability of the Hilbert transform with applications,” *Appl. Comput. Harmon. Anal.*, Sep. 2018, in press.

The Solvability Complexity Index for Calculating the Hilbert Transform

The Computational Problem

Now we investigate the Hilbert transform calculation using the SCI farmework.

▷ For $\beta \in [0, 1]$, we consider the computational problems

$$\{H, \mathcal{B}_\beta, \mathcal{C}(\mathbb{T}), \Lambda\} \quad \text{and} \quad \{H, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda\}$$

i.e.

- The **primary set** Ω is \mathcal{B}_β for some $\beta \in [0, 1]$.
- The **metric space** \mathcal{M} is either $\mathcal{C}(\mathbb{T})$ or \mathcal{B}_β .
- The **problem function** Ξ is the Hilbert transform $H: \mathcal{B}_\beta \rightarrow \mathcal{M}$ is the Hilbert transform.
- The **evaluations set** Λ contains the sampling functionals $\lambda_\tau: \mathcal{B}_\beta \rightarrow \mathbb{C}$ with $\tau \in \mathbb{T}$ defined by

$$\lambda_\tau(f) = f(\tau), \quad f \in \mathcal{B}_\beta.$$

▷ So the evaluation set consist now of **point evaluations (i.e. sampling)**.

Sampling-based Fundamental Algorithms

Definition (Sampling-based Fundamental Algorithm)

Let $\{H, \mathcal{B}_\beta, \mathcal{M}, \Lambda\}$ be a sampling-based computational problem for calculating the Hilbert transform. A lower semicontinuous mapping $\Gamma : \mathcal{B}_\beta \rightarrow \mathcal{M}$ is said to be a *sampling-based fundamental algorithm* if

- (i) There exists a finite subset $T_\Gamma \subset \mathbb{T}$
- (ii) For all $f, g \in \mathcal{B}_\beta$

$$\begin{array}{ll} f(\tau) = g(\tau) & \text{for all } \tau \in T_\Gamma \\ \text{implies } [\Gamma(f)](t) = [\Gamma(g)](t) & \text{for all } t \in \mathbb{T}. \end{array}$$

Moreover, we say that Γ is of type

1. *linear (L)*, if Γ is linear.
2. *general (G)*, if there is no further restriction on Γ .

Remark

- ▷ Every fundamental algorithm Γ is **concentrated on a finite sampling set** Λ_Γ .
- ▷ **No adaptivity**, i.e. the sampling sets Λ_Γ depend only on Γ but not on the actual function.

The SCI of the Hilbert Transform

Theorem

Let $\beta \in [0, 1]$ be arbitrary, then we have

$$\text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_L = \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{C}(\mathbb{T}), \Lambda)_G = 2.$$

Remark

▷ We also get

$$\begin{aligned} 2 = \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{C}(\mathbb{T}), \Lambda)_G &\leq \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{C}(\mathbb{T}), \Lambda)_L \leq \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_L = 2 \\ 2 = \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{C}(\mathbb{T}), \Lambda)_G &\leq \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_G \leq \text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_L = 2 \end{aligned}$$

▷ *Non-linear methods give no improvement over linear approximation methods (as far as the SCI is concerned).*

Proof

- ▷ Our previous already implied that $\text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_G > 1$.
- ▷ It remains to show that $\text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_L \leq 2$.

Tower of Algorithms for the Hilbert Transform

$$(Hf)(t) = -i \sum_{k=-\infty}^{\infty} \operatorname{sgn}(k) c_k(f) e^{ikt}, \quad t \in \mathbb{T} \quad \text{with} \quad c_k(f) = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(t) e^{-ikt} dt.$$

► Let $n_1, n_2 \in \mathbb{N}$ be fixed. For $f \in \mathcal{B}_\beta$, we consider

$$c_k(f, n_1) = \frac{1}{n_1} \sum_{l=0}^{n_1-1} f\left(l \frac{2\pi}{n_1}\right) e^{-i \frac{2\pi}{n_1} lk}, \quad \text{for all } |k| \leq n_2.$$

$c_k(f, n_1)$ is the approximation of the k th Fourier coefficient of f by a Riemann sum based on n_1 equally spaced sampling points and we have

$$\lim_{n_1 \rightarrow \infty} c_k(f, n_1) = c_k(f) \quad \text{for all } |k| \leq n_2.$$

► Define the operators in the lower level of the tower by

$$(\Gamma_{n_2, n_1} f)(t) = -i \sum_{k=-n_2}^{n_2} \operatorname{sgn}(k) c_k(f, n_1) e^{ikt}, \quad t \in \mathbb{T},$$

► Define the operators in the upper level of the tower by

$$(\Gamma_{n_2} f)(t) = -i \sum_{k=-n_2}^{n_2} \operatorname{sgn}(k) c_k(f) e^{ikt}, \quad t \in \mathbb{T}.$$

Discussion

- ▶ Tower of algorithm

$$(\Gamma_{n_2} f)(t) = -i \sum_{k=-n_2}^{n_2} \text{sgn}(k) c_k(f) e^{ikt}$$

$$(\Gamma_{n_2, n_1} f)(t) = -i \sum_{k=-n_2}^{n_2} \text{sgn}(k) c_k(f, n_1) e^{ikt} \quad \text{with} \quad c_k(f, n_1) = \frac{1}{n_1} \sum_{l=0}^{n_1-1} f\left(l \frac{2\pi}{n_1}\right) e^{-i \frac{2\pi}{n_1} l k}.$$

- ▶ The **first layer** calculates the Fourier coefficients $c_k(f)$ of f from the samples $\left\{ f\left(l \frac{2\pi}{n_1}\right) \right\}$ of f by the first limit

$$\lim_{n_1 \rightarrow \infty} c_k(f, n_1) = c_k(f).$$

- ▶ The **second layer** calculates $\tilde{f} = Hf$ by the conjugate Fourier series (second limit).

- ▶ How to get a tower of algorithm of height 1?

⇒ Other evaluation set Λ (i.e. other measurement functionals)

$$\lambda_k : f \mapsto c_k(f), \quad k = 0, \pm 1, \pm 2, \dots$$

⇒ Smaller primary set $\Omega \subset \mathcal{B}_\beta$.

Conclusions

Conclusions and Discussion

- ▷ The SCI for calculating the Hilbert transform Hf from samples of $f \in \mathcal{B}_\beta$ is equal to 2.
 - One needs 2 limits to calculate $\tilde{f} = H(f)$.
 - The SCI = 2 for linear and general towers of algorithms.
 - This statement holds primary sets \mathcal{B}_β with $\beta \in [0, 1]$

- ▷ For $\beta > 1$ we (probably) get

$$\text{SCI}(\mathbf{H}, \mathcal{B}_\beta, \mathcal{B}_\beta, \Lambda)_L = 1 .$$

- ▷ Similar results for other sampling based problems:
 - Recovering functions from samples on the space of uniformly convergent Fourier series.
 - Solving the Dirichlet problem based on samples of the boundary function.
 - Calculating the Spectral factorization.