



Absolute Value Information

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(joint work with P. Siedlecki and H. Woźniakowski)

IBC: worst case setting

Solution operator:

$$S : F \rightarrow G$$

F - linear space, G - normed space with $\|\cdot\|$

Approximation:

$$S(f) \sim \mathcal{A}_n(f) = \varphi(\mathbf{y})$$

where $\mathbf{y} = (y_1, y_2, \dots, y_n)$ is *information* about f ,

$$y_i = L_i(f) \quad (\text{nonadaptive})$$

$$y_i = L_i(f; y_1, \dots, y_{i-1}) \quad (\text{adaptive})$$

$L_i(\cdot; y_1, \dots, y_{i-1}) \in \Lambda$ - a class of linear functionals on F

$$\Lambda \subseteq \Lambda^{\text{all}}$$

Worst case error

Algorithm error:

(for a set $\mathcal{F} \subset F$ of problem instances)

$$e^{\text{wor}}(\mathcal{A}_n) = \sup_{f \in \mathcal{F}} \|S(f) - \mathcal{A}_n(f)\|$$

n th minimal error:

(for a class Λ of information functionals)

$$r_n^{\text{wor}}(\Lambda) = \inf_{\mathcal{A}_n} e^{\text{wor}}(\mathcal{A}_n)$$

(information) ε -complexity:

$$\text{comp}^{\text{wor}}(\varepsilon) = \min\{n : r_n^{\text{wor}}(\Lambda) \leq \varepsilon\}$$

Absolute Value Information (AVI)

(Motivated by the problem of *phase retrieval*, I. Daubechies, S. Mallat,)

For a given class Λ of information functionals we define

$$|\Lambda| = \{ |L| : L \in \Lambda \}$$

Then AVI about f is given as $\mathbf{y} = (y_1, y_2, \dots, y_n)$ where

$$y_i = |L_i(f; y_1, \dots, y_{i-1})|, \quad L_i(\cdot; y_1, \dots, y_{i-1}) \in \Lambda$$

We want to compare the powers of Λ and $|\Lambda|$ by

$$\mathbf{r}_n^{\text{wor}}(\Lambda) \quad \text{and} \quad \mathbf{r}_n^{\text{wor}}(|\Lambda|)$$

Approximation of Lipschitz functions

$$F = G = C([0, 1]), \quad S(f) = f, \quad f \in F$$

$$\mathcal{F} = \{f : [0, 1] \rightarrow \mathbb{R} : |f(x) - f(y)| \leq |x - y| \quad \forall x, y\}$$

$$\Lambda = \Lambda^{\text{std}} = \{f \mapsto L_t(f) = f(t) : 0 \leq t \leq 1\}$$

Then

$$r_n^{\text{wor}}(\Lambda) = \frac{1}{2n} \quad \text{but} \quad r_n^{\text{wor}}(|\Lambda|) = +\infty$$

More generally, if

- (i) \mathcal{F} is balanced, and
- (ii) for all $f \in \mathcal{F}$ is $S(-f) = -S(f)$

then for any Λ

$$r_n^{\text{wor}}(|\Lambda|) = r_0^{\text{wor}}(|\Lambda|) = \sup_{f \in \mathcal{F}} \|S(f)\|$$

Zero finding

Solve

$$f(x) = 0$$

for the function class

$$\mathcal{F} = \{f \in C([0,1]) : f(0)f(1) < 0 \quad \& \quad \exists! \eta_f \text{ s.t. } f(\eta_f) = 0\}$$

and $S(f) = \eta_f$. Then

$$\mathbf{r}_n^{\text{wor}}(\Lambda^{\text{std}}) = 2^{-n} \quad \text{and} \quad \mathbf{r}_n^{\text{wor}}(|\Lambda^{\text{std}}|) = 1/2$$

But, for

$$\Lambda_+^{\text{std}} = \{f \mapsto f(x) + f(y) : 0 \leq x, y \leq 1\}$$

we have

$$\mathbf{r}_n^{\text{wor}}(|\Lambda_+^{\text{std}}|) \leq 2^{-\lfloor n/2 \rfloor}$$

Question

What to do with those negative results for $|\Lambda|$?

Modified worst case error

Define

$$e^{\text{mod}}(\mathcal{A}_n) = \sup_{f \in \mathcal{F}} \inf_{|z|=1} \|S(f) - z \mathcal{A}_n(f)\|$$

and correspondingly

$$r_n^{\text{mod}}(\Lambda) \quad \text{and} \quad r_n^{\text{mod}}(|\Lambda|)$$

(Already used for the problem of *phase retrieval*, I. Daubechies, S. Mallat,)

Approximation of Lipschitz functions



We still have

$$r_n^{\text{mod}}(|\Lambda^{\text{std}}|) = 1$$

But, if

$$\Lambda_+^{\text{std}} = \{L_1 + L_2 : L_1, L_2 \in \Lambda^{\text{std}}\}$$

then

$$r_{2n}^{\text{mod}}(|\Lambda_+^{\text{std}}|) = r_n^{\text{wor}}(\Lambda^{\text{std}}) = \frac{1}{2n}$$

Theorem

Suppose that

- (i) S is homogenous,*
- (ii) there exists an algorithm \mathcal{A}_n that uses nonadaptively chosen information functionals from Λ , such that*

$$e^{\text{wor}}(\mathcal{A}_n) \leq C r_n^{\text{wor}}(\Lambda)$$

Then for

$$\Lambda_{\pm} = \{L_1 \pm L_2 : L_1, L_2 \in \Lambda\}$$

we have

$$r_n^{\text{mod}}(|\Lambda_{\pm}|) \leq C r_{\lfloor n/2 \rfloor}^{\text{wor}}(\Lambda)$$

AVI for general IBC problems: lower bound

**Theorem**

Let S be linear and $\mathcal{F} \subset F$ be convex and balanced. Then

$$r_n^{\text{mod}}(|\Lambda|) \geq \frac{1}{4} r_n^{\text{wor}}(\Lambda)$$

AVI for linear problems



Corollary

If S is linear, \mathcal{F} is convex and balanced, then

$$\frac{1}{4} r_n^{\text{wor}}(\Lambda^{\text{all}}) \leq r_n^{\text{mod}}(|\Lambda^{\text{all}}|) \leq 2 r_{\lfloor n/2 \rfloor}^{\text{wor}}(\Lambda^{\text{all}})$$

Conclusions for modified error



- Λ^{std} and $|\Lambda^{\text{std}}|$ are generally not related
- $\Lambda^{\text{all}} \sim |\Lambda^{\text{all}}|$ for linear problems