

Locally Adaptive Automatic Algorithms

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Abstract

Keywords:

1. The Basic Problem

Let \mathcal{F} be a separable Banach space of real-valued input functions with domain \mathcal{X} , and let \mathcal{F} have a semi-norm $|\cdot|_{\mathcal{F}}$. Let \mathcal{H} be a separable Banach space of output functions with norm $\|\cdot\|_{\mathcal{H}}$, and let S be a solution operator, $S : \mathcal{F} \rightarrow \mathcal{H}$. We have three examples in mind for this setting. Let a, b be two fixed real numbers with $a < b$, and $\mathcal{X} = [a, b]$. Here are three problems:

$$\begin{array}{lll} \text{Integration (INT)} & S : f \mapsto \int_a^b f(x) \, dx, & S : \mathcal{W}^{4,1}[a, b] \rightarrow \mathbb{R}, \\ \text{Approximation (APP)} & S : f \mapsto f, & S : \mathcal{W}^{3,\infty}[a, b] \rightarrow \mathcal{L}_{\infty}[a, b], \\ \text{Optimization (OPT)} & S : f \mapsto \min_{a \leq x \leq b} f(x) & S : \mathcal{W}^{3,\infty}[a, b] \rightarrow \mathbb{R}. \end{array}$$

The Sobolev and Lebesgue spaces and their (semi-)norms are defined as follows. For all real numbers α, β with $\alpha < \beta$,

$$\begin{aligned} \mathcal{W}^{k,p} &:= \mathcal{W}^{k,p}[a, b], & \mathcal{W}^{k,p}[\alpha, \beta] &:= \{f \in C[\alpha, \beta] : \|f^{(k)}\|_{p, [\alpha, \beta]} < \infty\}, \\ \mathcal{L}_p &:= \mathcal{L}_p[a, b], & \mathcal{L}_p[\alpha, \beta] &:= \mathcal{W}^{0,p}[\alpha, \beta], \\ \|f\|_{\mathcal{W}^{k,p}[\alpha, \beta]} &:= \|f^{(k)}\|_{p, [\alpha, \beta]}, & \|f\|_{\mathcal{L}_p[\alpha, \beta]} &:= \|f\|_{p, [\alpha, \beta]}, \\ \|f\|_{p, [\alpha, \beta]} &:= \begin{cases} \left[\int_{\alpha}^{\beta} |f(x)|^p \, dx \right]^{1/p}, & 1 \leq p < \infty, \\ \max_{\alpha \leq x \leq \beta} |f(x)|, & p = \infty. \end{cases} \end{aligned}$$

Although these are problems involving univariate functions, one could extend these to multivariate problems, and the general analysis would still apply.

2. Solving the Problems on Partitions

For any $\mathcal{Y} \subset \mathcal{X}$, let $f_{\mathcal{Y}}$ denote f restricted to the set \mathcal{Y} , i.e.,

$$f_{\mathcal{Y}} : \mathcal{Y} \rightarrow \mathbb{R}, \quad f_{\mathcal{Y}} : x \mapsto f(x).$$

Moreover, let $\mathcal{F}_{\mathcal{Y}}$ denote the space of functions in \mathcal{F} restricted to the set \mathcal{Y} , i.e., $\mathcal{F}_{\mathcal{Y}} = \{f_{\mathcal{Y}} : f \in \mathcal{F}\}$.

It is also assumed that there exists, \mathcal{T} , some sets of measurable subsets of \mathcal{X} for which one can define norms, solution operators, and approximation operators. It is assumed that

- For each $\mathcal{Y} \in \mathcal{T}$, the subspace $\mathcal{F}_{\mathcal{Y}}$ has a semi-norm, $|\cdot|_{\mathcal{F}_{\mathcal{Y}}}$ satisfying $|f_{\mathcal{Y}}|_{\mathcal{F}_{\mathcal{Y}}} \leq |f|_{\mathcal{F}}$. For simplicity of notation, we let $|f|_{\mathcal{F}_{\mathcal{Y}}} = |f_{\mathcal{Y}}|_{\mathcal{F}_{\mathcal{Y}}}$.
- For each $\mathcal{Y} \in \mathcal{T}$ there exists a solution operator $S(\cdot; \mathcal{Y}) : \mathcal{F} \rightarrow \mathcal{H}$, for which $S(f, \mathcal{Y})$ is actually only a function of $f_{\mathcal{Y}}$.

Partitions of \mathcal{X} are finite subsets of \mathcal{T} such that the following conditions hold:

- There exists a function $\Phi(\cdot; \mathcal{P}) : \mathcal{H}^{|\mathcal{P}|} \rightarrow \mathcal{H}$ that combines the solutions defined on the subsets to reconstruct the true solution:

$$S(f) = \Phi(\mathbf{S}_{\mathcal{P}}(f)), \quad \mathbf{S}_{\mathcal{P}}(f) := \{S(f; \mathcal{Y})\}_{\mathcal{Y} \in \mathcal{P}}, \quad \forall f \in \mathcal{F}.$$

Here $|\mathcal{P}|$ denotes the cardinality of \mathcal{P} .

- There exists a pair of functions $(\tilde{\Phi}, \text{err})(\cdot, \cdot; \mathcal{P}) : \mathcal{H}^{|\mathcal{P}|} \times [0, \infty)^{|\mathcal{P}|} \rightarrow \mathcal{H} \times [0, \infty)$ that combine approximate solutions defined on the subsets with error bounds to reconstruct an approximation to the true solution. If $\|S(f; \mathcal{Y}) - \tilde{S}_{\mathcal{Y}}\|_{\mathcal{H}} \leq \varepsilon_{\mathcal{Y}}$ for all $\mathcal{Y} \in \mathcal{P}$, $\tilde{\mathbf{S}}_{\mathcal{P}} := \{\tilde{S}_{\mathcal{Y}}\}_{\mathcal{Y} \in \mathcal{P}}$, and $\tilde{\varepsilon}_{\mathcal{P}} := \{\varepsilon_{\mathcal{Y}}\}_{\mathcal{Y} \in \mathcal{P}}$, then

$$\|S(f) - \tilde{\Phi}(\tilde{\mathbf{S}}_{\mathcal{P}}, \tilde{\varepsilon}_{\mathcal{P}}; \mathcal{P})\|_{\mathcal{H}} \leq \text{err}(\tilde{\mathbf{S}}_{\mathcal{P}}, \tilde{\varepsilon}_{\mathcal{P}}; \mathcal{P}) \quad \forall f \in \mathcal{F}.$$

Here $|\mathcal{P}|$ denotes the cardinality of \mathcal{P} . Moreover,

$$\text{err}(\tilde{\mathbf{S}}_{\mathcal{P}}, \mathbf{0}; \mathcal{P}) = 0, \quad \tilde{\Phi}(\tilde{\mathbf{S}}_{\mathcal{P}}, \mathbf{0}; \mathcal{P}) = \Phi(\mathbf{S}_{\mathcal{P}}(f)) = S(f).$$

Note that the subsets of \mathcal{X} comprising the partition \mathcal{P} need not have nonempty intersection.

For our three problems, these partitions take the form of subintervals of $[a, b]$:

$$\begin{aligned} \mathcal{T} &= \{[\alpha, \beta] : a \leq \alpha < \beta \leq b\} \\ \mathcal{P} &= \{[t_0, t_1], [t_1, t_2], \dots, [t_{L-1}, t_L]\}, \quad a = t_0 < t_1 < \dots < t_L = b. \end{aligned}$$

The semi-norms and solution operators defined on the elements of \mathcal{T} , and the functions Φ , $\tilde{\Phi}$, and err that combine the solutions on the sets in the partition into the full solution, the approximate, and the upper error bound are the following:

$$\begin{aligned} \text{INT} \quad & \|f_{[\alpha, \beta]}\|_{\mathcal{F}_{[\alpha, \beta]}} = \|f^{(4)}\|_{1, [\alpha, \beta]}, \quad S_{[\alpha, \beta]} : f \mapsto \int_{\alpha}^{\beta} f(x) \, dx, \\ & \Phi(\tilde{\mathbf{S}}_{\mathcal{P}}; \mathcal{P}) = \tilde{\Phi}(\tilde{\mathbf{S}}_{\mathcal{P}}, \varepsilon_{\mathcal{P}}; \mathcal{P}) = \sum_{l=1}^L \tilde{S}_{[t_{l-1}, t_l]}, \\ & \text{err}(\tilde{\mathbf{S}}_{\mathcal{P}}, \varepsilon_{\mathcal{P}}; \mathcal{P}) = \|\varepsilon_{\mathcal{P}}\|_1, \end{aligned}$$

$$\begin{aligned}
\text{APP} \quad & \|f_{[\alpha, \beta]}\|_{\mathcal{F}_{[\alpha, \beta]}} = \|f^{(3)}\|_{\infty, [\alpha, \beta]}, \quad S_{[\alpha, \beta]} : f \mapsto f \mathbb{1}_{[\alpha, \beta]}, \\
& \Phi(\tilde{\mathbf{S}}_{\mathcal{P}}; \mathcal{P}) = \tilde{\Phi}(\tilde{\mathbf{S}}_{\mathcal{P}}, \varepsilon_{\mathcal{P}}; \mathcal{P}) = \sum_{l=1}^L \tilde{S}_{[t_{l-1}, t_l]}, \\
& \text{err}(\tilde{\mathbf{S}}_{\mathcal{P}}, \varepsilon_{\mathcal{P}}; \mathcal{P}) = \|\varepsilon_{\mathcal{P}}\|_{\infty}, \\
\text{OPT} \quad & \|f_{[\alpha, \beta]}\|_{\mathcal{F}_{[\alpha, \beta]}} = \|f^{(3)}\|_{\infty, [\alpha, \beta]}, \quad S_{[\alpha, \beta]} : f \mapsto \min_{\alpha \leq x \leq \beta} f(x), \\
& \Phi(\tilde{\mathbf{S}}_{\mathcal{P}}; \mathcal{P}) = \min \tilde{\mathbf{S}}_{\mathcal{P}}, \\
& \tilde{\Phi}(\tilde{\mathbf{S}}_{\mathcal{P}}, \varepsilon_{\mathcal{P}}; \mathcal{P}) = \frac{1}{2} \left\{ \min(\tilde{\mathbf{S}}_{\mathcal{P}} + \varepsilon_{\mathcal{P}}) + \min(\tilde{\mathbf{S}}_{\mathcal{P}} - \varepsilon_{\mathcal{P}}) \right\}, \\
& \text{err}(\tilde{\mathbf{S}}_{\mathcal{P}}, \varepsilon_{\mathcal{P}}; \mathcal{P}) = \frac{1}{2} \left\{ \min(\tilde{\mathbf{S}}_{\mathcal{P}} + \varepsilon_{\mathcal{P}}) - \min(\tilde{\mathbf{S}}_{\mathcal{P}} - \varepsilon_{\mathcal{P}}) \right\}.
\end{aligned}$$

Here the operator “min” applied to a vector takes the minimum of the elements.

3. Algorithms

Now we consider numerical algorithms for solving the problems on a subset of the whole domain. Suppose that

- For each $\mathcal{Y} \in \mathcal{T}$ and each $n \in \mathcal{I}$ there exists a non-adaptive approximation operator $A_n(\cdot; \mathcal{Y}) : \mathcal{F} \rightarrow \mathcal{H}$ that uses n function values sampled only in \mathcal{Y} .
- There exists an error bound function $h : \mathcal{I} \times \mathcal{T} \rightarrow [0, \infty)$ such that $h(\cdot, \mathcal{Y})$ is non-increasing, and

$$\|S(f; \mathcal{Y}) - A_n(f; \mathcal{Y})\|_{\mathcal{H}} \leq h(n, \mathcal{Y}) |f|_{\mathcal{F}_{\mathcal{Y}}}, \quad \forall f \in \mathcal{F}.$$

For our three problems, the non-adaptive algorithms A_n are based on piecewise quadratic polynomial approximation of the function.

For the integration problem this corresponds to Simpson’s rule. The number of possible function values is the odd integers $\mathcal{I} = \{2i - 1 : i \in \mathbb{N}\}$

$$x_i = \alpha + (\beta - \alpha) \frac{i - 1}{n - 1}, \quad i = 1, \dots, n, \quad (1a)$$

$$\begin{aligned}
A_n(f, [\alpha, \beta]) = \frac{\beta - \alpha}{3(n - 1)} [f(x_1) + 4f(x_2) + 2f(x_3) + 4f(x_4) + \dots \\
+ 4f(x_{n-1}) + f(x_n)]. \quad (1b)
\end{aligned}$$

From ??? it is known that discretization error of Simpson’s rule is

$$\left| \int_{\alpha}^{\beta} f(x) dx - A_n(f, [\alpha, \beta]) \right| \leq \|f^{(4)}\|_{1, [\alpha, \beta]} h(n, [\alpha, \beta]), \quad (1c)$$

$$h(n, [\alpha, \beta]) = ??? \quad (1d)$$

For approximation
For optimization

4. Functions in Cones

The challenge in applying the error bounds for the non-adaptive integration, function recovery, and optimization algorithms is that $|f|_{\mathcal{F}_Y}$ is not known a priori. Our approach is to assume that the input functions lie inside cones. For partitions \mathcal{P} suppose that

- For each $\mathcal{Y} \in \mathcal{T}$ there exists a semi-norm $|\cdot|_{\mathcal{G}_Y}$ defined on the space \mathcal{F}_Y that is weaker than $|\cdot|_{\mathcal{F}_Y}$.
- For a fixed function non-increasing $\tau : (0, 1) \rightarrow (0, \infty)$, define the cone

$$\mathcal{C}_\tau = \{f \in \mathcal{F} : |f|_{\mathcal{F}_Y} \leq \tau(\text{vol}(\mathcal{Y})) |f|_{\mathcal{G}_Y}\}, \quad (2)$$

where $\text{vol}(\mathcal{Y})$ denotes relative volume of \mathcal{Y} , i.e., the Lebesgue measure \mathcal{Y} divided by the Lebesgue measure of \mathcal{X} .

- For each $\mathcal{Y} \in \mathcal{T}$ and each $n \in \mathcal{I}$ there exists a non-adaptive approximation operator $G_n(\cdot; \mathcal{Y}) : \mathcal{F} \rightarrow \mathcal{H}$ that uses n function values sampled only in \mathcal{Y} .
- There exists error bound functions $g_\pm : \mathcal{I} \times \mathcal{T} \rightarrow [0, \infty)$ such that $g(\cdot, \mathcal{Y})$ is non-increasing, and

$$-g_-(n, \mathcal{Y}) |f|_{\mathcal{F}_Y} \leq |f|_{\mathcal{G}_Y} - G_n(f; \mathcal{Y}) \leq g_+(n, \mathcal{Y}) |f|_{\mathcal{F}_Y}, \quad \forall f \in \mathcal{F}.$$

Invoking the definition of the cone implies a two sided bound for $|f|_{\mathcal{G}_Y}$ and $|f|_{\mathcal{F}_Y}$ in terms of $G_n(f; \mathcal{Y})$:

$$-\tau(\text{vol}(\mathcal{Y}))g_-(n, \mathcal{Y}) |f|_{\mathcal{G}_Y} \leq |f|_{\mathcal{G}_Y} - G_n(f; \mathcal{Y}) \leq \tau(\text{vol}(\mathcal{Y}))g_+(n, \mathcal{Y}) |f|_{\mathcal{G}_Y}, \quad \forall f \in \mathcal{C}_\tau.$$

$$\begin{aligned} \frac{G_n(f; \mathcal{Y})}{1 + \tau(\text{vol}(\mathcal{Y}))g_-(n, \mathcal{Y})} &\leq |f|_{\mathcal{G}_Y} \leq \frac{G_n(f; \mathcal{Y})}{1 - \tau(\text{vol}(\mathcal{Y}))g_+(n, \mathcal{Y})}, \quad \forall f \in \mathcal{C}_\tau, \\ \frac{\tau(\text{vol}(\mathcal{Y}))G_n(f; \mathcal{Y})}{1 + \tau(\text{vol}(\mathcal{Y}))g_-(n, \mathcal{Y})} &\leq |f|_{\mathcal{F}_Y} \leq \frac{\tau(\text{vol}(\mathcal{Y}))G_n(f; \mathcal{Y})}{1 - \tau(\text{vol}(\mathcal{Y}))g_+(n, \mathcal{Y})}, \quad \forall f \in \mathcal{C}_\tau. \end{aligned}$$

Algorithm 1. Given