

Algorithms of Scientific Computing (Algorithmen des Wissenschaftlichen Rechnens) Spatial Adaptivity (Implementation), Combination Technique

Proposed solution

Exercise 1: One-dimensional Sparse Grids—An Adaptive Implementation

The ipython/python skeleton code is available in the assignment of Sheet 11.

We introduced Archimedes' approach to approximate the integral $F(f, a, b) = \int_a^b f(x) dx$ of a function $f : \mathbb{R} \rightarrow \mathbb{R}$, respectively to approximate the function f itself.

For the one-dimensional case we want to formalize this approach and generalize it in the following ways:

- Let $\phi(x)$ be the “mother of all hat functions” with

$$\phi(x) = \begin{cases} x + 1 & \text{for } -1 \leq x < 0 \\ 1 - x & \text{for } 0 \leq x < 1 \\ 0 & \text{else} \end{cases} \quad (1)$$

- The data structure used to store the hierarchical coefficients is now called *Sparse Grid*.
- A sparse grid is defined by a particular set of interpolation points $x_{l,i}$ and associated ansatz functions $\phi_{l,i}(x)$ with

$$\phi_{l,i}(x) = \phi\left(2^l \cdot \left(x - i \cdot \frac{1}{2^l}\right)\right) = \phi(2^l \cdot x - i), \quad l \in \mathbb{N}^+, i \in \{1, 3, \dots, 2^l - 1\} \quad (2)$$

- Archimedes' approach from the lecture corresponds to a *regular* sparse grid.
- To improve the quality of approximation for arbitrary functions f we introduce spatial adaptivity.

Your task is to implement the missing parts in the members of the *SparseGrid1d* class and turn it into a fully working adaptive implementation of a one-dimensional sparse grid. Import and use the class *GridPoint* and look at the comments in the provided code snippets for some more details.

- a) The constructor `__init__` creates a grid containing all grid points on levels $l \leq \text{minLevel}$. A given function f is then evaluated at those points before *hierarchization* is performed eventually to obtain the hierarchical coefficients.
Implement this behavior.
- b) Implement the member function `computeVolume` that computes an approximation for $F(f, 0, 1)$ using the current sparse grid interpolant.
- c) Implement the member function `refineAdaptively` that takes a certain refinement criterion (see source code) and inserts new grid points accordingly.

Exercise 2: The Combination Technique – A Different View on Sparse Grids

Dealing with hierarchical bases often turns out to be sophisticated. On this worksheet we will therefore see how the so-called *combination technique* finds a sparse grid interpolant, that approximates a function on a number of full grids, each consisting only of a “relatively small” number of grid points.

Let $u_{\underline{l}}$ ($\underline{l} \in \mathbb{N}^2$) for a $u : [0, 1]^2 \rightarrow \mathbb{R}$ the interpolant in $V_{\underline{l}}$ (interpolating piecewise bilinearly at the inner grid points, at the boundary u is assumed to be zero again).

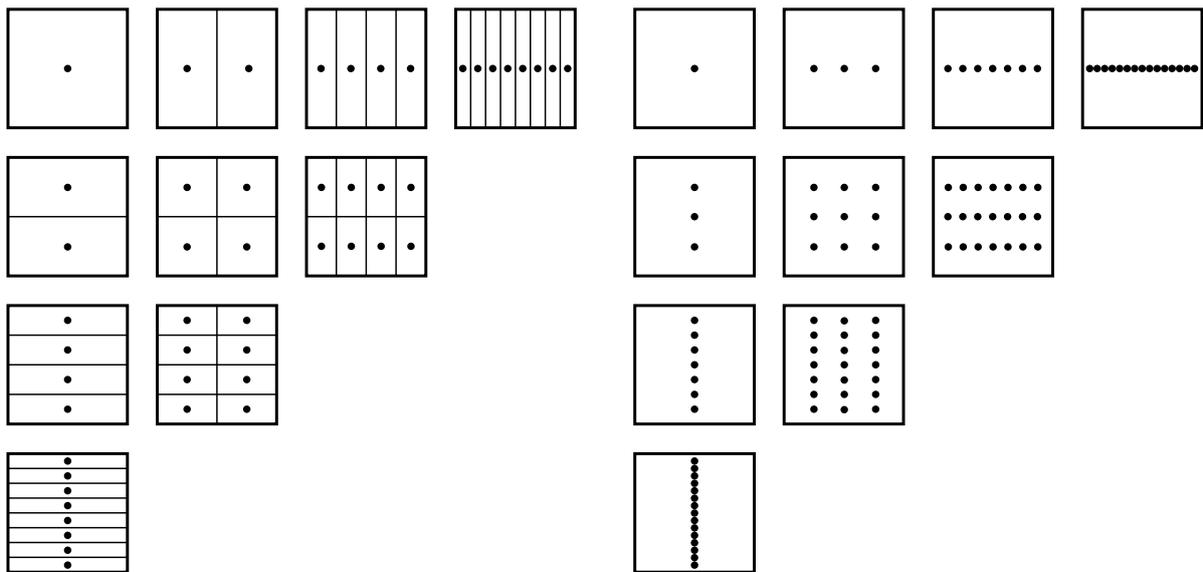


Figure 1: The two parts in the picture show the grid points and supports associated with interpolants $w_{\underline{l}}$ (left) and $u_{\underline{l}}$ (right) up to level 4 for the 2d case.

- (i) $V_{\underline{l}}$ can be decomposed into a set of subspaces $W_{\underline{l}}$. Accordingly, the interpolant $u_{\underline{l}} \in V_{\underline{l}}$ can be written as a sum of $w_{\underline{l}} \in W_{\underline{l}}$.

Spot the grid associated with $u_{(3,2)}$ in the right part of Figure 1. Identify those subspaces in the left part that are needed to reconstruct $u_{(3,2)}$.

We need to “collect” those $w_{\underline{l}}$ with indices bound component-wise by \underline{l} . In subspace scheme (see left hand side of Figure 1) these are the subspaces in the rectangular left upper part relative to \underline{l} . Written as a sum this gives us:

$$u_{\underline{l}} = \sum_{\underline{l}' \leq \underline{l}} w_{\underline{l}'}$$

Note that weights are not needed yet.

(ii) Use the result from (i) to rewrite

$$\sum_{|\underline{l}|_1 = n+1} u_{\underline{l}}, \quad n \in \mathbb{N}$$

for the two-dimensional case as a weighted sum of $w_{\underline{l}}$.

Hint: Look at the subspace scheme in Figure 1 and count the occurrences of each subspace in the sum. What do you notice when comparing $w_{\underline{l}}$ with common level $n = |\underline{l}|_1 + \dim - 1$?

Using the previous result we start with

$$\sum_{|\underline{l}|_1 = n+1} u_{\underline{l}} = \sum_{|\underline{l}|_1 = n+1} \sum_{\underline{l}' \leq \underline{l}} w_{\underline{l}'}$$

For the reorganization part we first count which $w_{\underline{l}'}$ appears how many times in the sums. (remember, this is **2d!**)

$|\underline{l}'|_1 = n + 1$: each $w_{\underline{l}'}$ has one occurrence

$|\underline{l}'|_1 = n$: each $w_{\underline{l}'}$ has two occurrences

⋮

$|\underline{l}'|_1 = k$: each has $n + 2 - k$ occurrences

Technical explanation: Let \underline{l} be of level $n + 1$, i.e. $|\underline{l}|_1 = n + 1$. From \underline{l} construct all possible \underline{l}' of level n with $\underline{l}' \leq \underline{l}$. Those are exactly two, as there are two components (**2d!**) that can be decreased by 1. Applying this scheme down to level 1 then leads to the result above.

Written as an explicit sum formula this is:

$$\sum_{|\underline{l}|_1 = n+1} u_{\underline{l}} = \sum_{|\underline{l}|_1 \leq n+1} (n + 2 - |\underline{l}|_1) w_{\underline{l}}$$

(iii) In the final step use the previous results to give a representation of the sparse grid interpolant

$$u_n^D := \sum_{|\underline{l}|_1 \leq n+1} w_{\underline{l}}$$

as a weighted sum of $w_{\underline{l}}$. Again, count the occurrences of the $w_{\underline{l}}$.

By looking at the subspace scheme rather than by looking at the formulas it becomes clear that the following holds:

$$\sum_{|\underline{l}|_1 \leq n+1} w_{\underline{l}} = \sum_{|\underline{l}|_1 = n+1} u_{\underline{l}} - \sum_{|\underline{l}|_1 = n} u_{\underline{l}}$$

(iv) Assume you are talking to a person who knows how to approximate the volume $F_2(u)$ through the trapezoidal rule (in 2d) with respect to $u_{\underline{l}}$. Give instructions on how to write a program that implements a sparse grid approximation of $F_2(u)$. Remember Archimedes quadrature.

- *First idea: Replace volume $F_2(u)$ by the sparse grid volume approximation $F_2(u_n^D)$.*
- *Second idea: Think of the interpolant as a sum of $u_{\underline{l}}$. We know those $u_{\underline{l}}$ (interpolating u on regular grids) as well as their volumes (trapezoidal rule in 2d).*
- *Together with the weights from the previous part we get*

$$F_2(u) \approx F_2(u_n^D) = \sum_{|\underline{l}|_1=n+1} F_2(u_{\underline{l}}) - \sum_{|\underline{l}|_1=n} F_2(u_{\underline{l}})$$

(v) Compare this method with Archimedes quadrature — what are the (dis-)advantages?

Advantages:

- *Simpler program code (Haven't you tried coding them? Do it! You'll agree...)*
- *It might be possible to reuse an existing program for the trapezoidal rule on common regular grids (advantage is even bigger for more complex applications, e.g. when computing a sparse grid solution for a fluid simulation)*
- *For comprehensive computations the program is more likely and easy to be parallelized as the single grids are processed independently from each other*

Disadvantages:

- *No straight forward approach to include adaptivity, i.e. it's not possible to automatically find the right evaluation points*
- *Recursion is much more beautiful!*