Low-order reconstruction operators on polyhedral meshes: Application to Compatible Discrete Operator schemes

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Abstract

We study low-order reconstruction operators on polyhedral meshes, providing a unified framework for degrees of freedom attached to vertices, edges, faces, and cells. We present two equivalent sets of design properties and draw links with the literature. In particular, the two-level construction based on a $P_0$-consistent and a stabilization part provides a systematic way of designing these operators. We present a simple example of piecewise constant reconstruction in each mesh cell, relying on geometric identities to fulfill the design properties on polyhedral meshes. Finally, we use these reconstruction operators to define a Hodge inner product and build Compatible Discrete Operator schemes, and we test the influence of the reconstruction operators in terms of accuracy and computational efficiency on an anisotropic diffusion problem.

Keywords: reconstruction operator, polyhedral meshes, compatible discretizations

1. Introduction

Reconstruction (or lifting) operators map degrees of freedom (DoFs) to functions living in a finite-dimensional space. DoFs are generally attached to some geometric entities of an underlying three-dimensional mesh, e.g., vertices, edges, faces, and cells. Reconstruction operators provide a right inverse of the de Rham (or reduction) operators which classically map fields (referred to as potentials, circulations, fluxes, and densities – or $k$-forms with $k \in \{0, 1, 2, 3\}$ in the language of differential geometry) to DoFs attached to vertices, edges, faces, and cells, respectively. A reconstruction operator composed with the corresponding de Rham operator yields an interpolation operator. The reconstruction operator is said to be of low-order when this interpolation operator leaves cell-wise constant fields invariant.

Reconstruction operators are found in many applications. Our main focus here is the construction of discrete Hodge operators in the context of the numerical approximation of partial differential equations (PDEs). The discrete Hodge operator is the cornerstone of many compatible discretization schemes aiming at preserving properties of the PDE at the discrete level; see, e.g., [1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11] and references therein. Many of these discretizations draw links between vector calculus, differential geometry, and algebraic topology. Reconstruction operators also constitute a powerful tool to analyze numerical schemes and to derive improved error estimates in different norms; see for instance [4, 12, 13]. One recent example is provided by the Compatible Discrete Operator (CDO) schemes for diffusive PDEs and the Stokes equations [3, 14]. To some extent, CDO schemes can be seen as an extension of Discrete Exterior Calculus (DEC) schemes [6, 15] to polyhedral meshes. We observe that reconstruction operators on polyhedral meshes have a broader applicability than the construction of Hodge operators. Examples of alternative usages include the reconstruction of vector fields for postprocessing and imaging purposes.

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Reconstruction operators have been devised in the Finite Element (FE) literature for specific shapes of mesh cells (tetrahedron, hexahedron, pyramid...). The most famous examples are Whitney reconstruction functions in simplices [16]. These reconstruction functions are built using the Courant hat functions for potentials, the (lowest-order) Nédélec shape functions for circulations, and the (lowest-order) Raviart–Thomas–Nédélec shape functions for fluxes. A typical way to extend the reconstruction of potentials to polyhedral meshes is to use the concept of generalized barycentric coordinates; see [17, 8, 15, 20, 21] and references therein.

A generic way of building reconstruction operators for any type of DoFs on polyhedral meshes has been proposed in [22, 23, 24]. The reconstruction operators are built locally in each mesh cell in such a way that suitable matching conditions are satisfied at mesh interfaces. Specifically, reconstructed potentials are continuous across interfaces, the tangential component of circulations is continuous, and so is the normal component of fluxes. Such matching conditions ensure the conformity of the reconstruction, in the sense that the operator maps to the appropriate Sobolev space such as $H^1(\Omega)$, $H(\text{curl}; \Omega)$, or $H(\text{div}; \Omega)$, where $\Omega$ is the computational domain discretized by the three-dimensional polyhedral mesh. The conformity of the reconstruction then plays a central role in the analysis of the numerical scheme.

An alternative viewpoint, not aiming at conformity, has been developed in the context of other discretization methods such as, e.g., the Hybrid Finite Volume (HFV) scheme [25], the Discrete Geometric Approach (DGA) [26], and, more recently, the CDO schemes [3, 4, 14], the generalized Crouzeix–Raviart method [13], and the Hybrid High-Order (HHO) methods [27, 28] (which also include the possibility to increase the approximation order). For the low-order schemes, the reconstruction operators typically map onto piecewise constant functions on a submesh (thereby discarding local conformity), while their composition with the de Rham operator remains single-valued. In this context, the analysis of the numerical schemes generally hinges on a novel property of the reconstruction, to which we refer as dual consistency.

Our contribution is twofold. Firstly, we devise low-order reconstruction operators on polyhedral meshes within a generic framework for DoFs attached to vertices, edges, faces, and cells. This framework provides a systematic construction principle relying on only one user-defined design parameter. Secondly, we identify a small set of design principles that reconstruction operators have to verify so that the resulting discrete Hodge operator satisfies $P_0$-consistency and stability properties, which in turn ensure the convergence of the numerical scheme.

The novelty is that the proposed framework unifies two design strategies (hereafter called one-level and two-level) and encompasses several low-order nonconforming reconstruction operators already considered in the literature. Specifically, we establish the equivalence between the one-level design strategy of the reconstruction operator as considered in [3, 26] and the two-level design strategy as considered in [23, 25, 30]. This second strategy decomposes the reconstruction operator into the sum of a consistent part and a stabilization part. The consistent part is fixed, while the stabilization part depends on the user-defined parameter related to the weighting of a least-squares penalty. Additionally, the reconstruction operators built with the value $\frac{1}{d}$ ($d$ is the space dimension) for the user-defined parameter correspond to those proposed in [26] for DoFs attached to edges and faces, while the values $\frac{1}{\sqrt{d}}$ and 1 lead, respectively, to the reconstruction operator for the gradient devised in [25] and the generalized Crouzeix–Raviart functions of [13].

As an illustration, we present piecewise constant reconstruction operators in each mesh cell for all types of DoFs and study the impact of the stabilization parameter in terms of accuracy and computational cost for the numerical approximation of anisotropic diffusion problems on polyhedral meshes.

This paper is organized as follows. In Section 2, we introduce the different geometric entities. In Section 3, we briefly present the CDO framework. A detailed presentation of this subject can be found in [3]. In Section 4, we state the design properties of reconstruction operators on polyhedral cells, and show that the one- and two-level design principles are equivalent. In Section 5, we design a family of reconstruction operators which are piecewise constant on each mesh cell and which fulfill the design properties stated in Section 4. Finally, in Section 6, we present an application to CDO schemes for the approximation of anisotropic diffusion problems on polyhedral meshes.
2. Geometric objects

2.1. Mesh and geometric entities

The starting point is a discretization of the geometric domain $\Omega \subset \mathbb{R}^3$ by a (primal) mesh $M := \{V, E, F, C\}$ where $V$ collects vertices (or 0-cells), $E$ edges (or 1-cells), $F$ faces (or 2-cells), and $C$ cells (or 3-cells). A generic element of $V$ (resp. $E$, $F$, $C$) is a vertex denoted by $v$ (resp. an edge $e$, a face $f$, a cell $c$); see Figure 1. The mesh $M$ has the structure of a cellular complex, in the sense that the boundary of a $k$-cell in $M$, $1 \leq k \leq 3$, can be decomposed into $(k-1)$-cells belonging to $M$ (see [23]).

Let $A$ be any set among $V, E, F,$ or $C$. If $x$ is a geometric entity of $M$ of dimension larger than that of the elements of $A$, we denote by $A_x$ the subset defined by

\[ A_x := \{ a \in A \mid a \subset \partial x \}, \quad (1) \]

otherwise,

\[ A_x := \{ a \in A \mid x \subset \partial a \}. \quad (2) \]

For instance, $E_e := \{ e \in E \mid e \subset \partial c \}$ collects the edges of $c$ and $C_e := \{ c \in C \mid e \subset \partial c \}$ collects the cells of which $e$ is an edge. In what follows, design properties are stated on each cell $c \in C$. Therefore, the sets $V_c, E_c, and F_c$ play a key role (note that $C_c = \{ c \}$).

We often denote by $X$ any set such as $V, E, F,$ or $C$ and by $x$ any geometric entity such as $v, e, f,$ or $c$. The cardinality of the set $X$ is denoted by $\#X$.

**Definition 1** (measure). $|x|$ represents the measure of the entity $x$. For a vertex $v \in V$, $|v| = 1$ by convention, $|e|$ is the length of the edge $e$, $|f|$ is the area of the face $f$, and $|c|$ is the volume of the cell $c$.

**Definition 2** (barycenter). The barycenters of an edge $e \in E$ and of a face $f \in F$ are defined, respectively, as follows:

\[ \mathbf{e} := \frac{1}{|e|} \int_e \mathbf{x} \quad \text{and} \quad \mathbf{f} := \frac{1}{|f|} \int_f \mathbf{x}. \quad (3) \]

To each edge $e \in E$, we assign a fixed unit tangent vector $\mathbf{e}_e$ and to each face $f$ a fixed unit normal vector $\mathbf{n}_f$. Moreover, we define for all edges $e \in E$ and all faces $f \in F$ the vectors

\[ \mathbf{e} := \int_e \mathbf{e}_e, \quad \mathbf{f} := \int_f \mathbf{n}_f. \quad (4) \]
Mesh assumption. We assume that all primal faces are planar and that each face \( f \in F \) is star-shaped with respect to its barycenter. Moreover, we assume that each cell \( c \in C \) is star-shaped with respect to a point \( x_c \in c \) (not necessarily the barycenter of \( c \)). In what follows, we denote by \((\text{MB})\) this set of assumptions.

2.2. Geometric maps

For the remaining part of the paper, we consider an arbitrary cell \( c \in C \) and state definitions and properties for this cell.

**Definition 3** (primal geometric map). We introduce a primal geometric map \( g_{v_c} : X_c \to \mathbb{E}_X \), where \( \mathbb{E}_X \) corresponds to \( \mathbb{R} \) if \( X \in \{V, C\} \) and to \( \mathbb{R}^3 \) if \( X \in \{E, F\} \), so that

\[
\begin{align*}
g_{v_c}(v) & := 1, & \forall v \in V_c, \quad (5a) \\
g_{e_c}(e) & := \xi, & \forall e \in E_c, \quad (5b) \\
g_{e_c}(f) & := \tau, & \forall f \in F_c, \quad (5c) \\
g_{c_c}(c) & := |c|. \quad (5d)
\end{align*}
\]

**Definition 4** (simplex). For all \( 1 \leq k \leq 3 \), given \( (k+1) \) points \( \{x_0, \ldots, x_k\} \), \( s(0, \ldots, k) \) denotes the convex hull of these points (yielding, up to degenerate cases, a segment for \( k = 1 \), a triangle for \( k = 2 \), and a tetrahedron for \( k = 3 \)); see Figure 2.

![Figure 2: Example of a prismatic cell. Left: the elementary tetrahedron \( s(v, c, f, \tau) \) is highlighted; Middle: the elementary triangle \( s(e, f, \tau) \) is highlighted; Right: the elementary segment \( s(f, \tau) \) is highlighted.](image)

**Definition 5** (dual geometric map). We introduce a dual geometric map \( \tilde{g}_{v_c} : X_c \to \mathbb{E}_X \) defined as follows:

\[
\begin{align*}
\tilde{g}_{v_c}(v) & := \sum_{e \in E_c \cap E} \sum_{f \in F_c} |s(v, e, f, \tau)|, & \forall v \in V_c, \quad (6a) \\
\tilde{g}_{e_c}(e) & := \sum_{f \in F_c} |s(e, f, \tau)|/\mathcal{L}_{s(e, f, \tau)}, & \forall e \in E_c, \quad (6b) \\
\tilde{g}_{e_c}(f) & := |s(f, \tau)|/\mathcal{L}_{s(f, \tau)}, & \forall f \in F_c, \quad (6c) \\
\tilde{g}_{c_c}(c) & := 1, \quad (6d)
\end{align*}
\]

where \( \mathcal{L}_{s(e, f, \tau)} \) is the unit normal vector to the triangle \( s(e, f, \tau) \) oriented according to \( \tau_e \) for all faces \( f \in F_c \) and \( \mathcal{L}_{s(f, \tau)} \) is the unit tangent vector to the segment \( s(f, \tau) \) oriented according to \( \tau_f \); see Figures 2 and 3.

**Remark 1** (dual mesh). The quantities specified in Definition 5 naturally appear when one considers a barycentric dual mesh. Namely, \( \tilde{g}_{v_c}(v) \) is the volume in \( c \) of the dual cell associated with the vertex \( v \in V_c \), \( \tilde{g}_{e_c}(e) \) is the area in \( c \) of the dual face associated with the edge \( e \in E_c \), and \( \tilde{g}_{e_c}(f) \) is the vector length in \( c \) of the dual edge associated with the face \( f \in F_c \).
Proposition 1 (magic formula). Assume (MB). Then, the following identity holds:

$$\sum_{x \in X_c} \tilde{g}_{X_c}(x) \otimes g_{X_c}(x) = |c| \mathbb{I}_{X_c}.$$  \hspace{1cm} (7)

If $X \in \{V, C\}$, $\mathbb{I}_{X_c}$ is equal to 1 and $\otimes$ is simply a multiplication. If $X \in \{E, F\}$, $\mathbb{I}_{X_c}$ is the $3 \times 3$ identity tensor and $\otimes$ is the tensor product.

Proof. The case $X \in \{V, C\}$ is straightforward. The proof for the case $X \in \{E, F\}$ is given in [3, Proposition 5.24]; see also [31]. \qed

The exact representation of constant fields by the reconstruction operators devised in this paper hinges on the identity (7).

3. Overview of the CDO framework

A detailed presentation of the CDO framework can be found in [3]. In what follows, we only present the main ideas.

3.1. Degrees of freedom

Following the seminal ideas of Tonti [32] and Bossavit [33], DoFs are defined using de Rham maps, and their localization results from the physical nature of the fields. For a cell $c \in C$, the (local) finite dimensional space of DoFs related to discrete potentials is denoted by $V_c$ and collects the values of scalar fields at vertices,

$$R_{V_c}(p)|_v := p(x_v), \quad \forall v \in V_c,$$  \hspace{1cm} (8a)

that related to discrete circulations is denoted by $E_c$ and collects the integrals of the tangential component of vector fields along edges,

$$R_{E_c}(g)|_e := \int_e g \cdot \tau_e, \quad \forall e \in E_c,$$  \hspace{1cm} (8b)

that related to discrete fluxes is denoted by $F_c$ and collects the integrals of the normal component of vector fields across faces,

$$R_{F_c}(\phi)|_f := \int_f \phi \cdot \nu_f, \quad \forall f \in F_c,$$  \hspace{1cm} (8c)

and that related to discrete densities is denoted by $C_c$ and collects the integral of scalar fields over the cell,

$$R_{C_c}(s)|_c := \int_c s.$$  \hspace{1cm} (8d)
Let \( \mathcal{X}_c \in \{ \mathcal{V}_c, \mathcal{E}_c, \mathcal{F}_c, \mathcal{C}_c \} \). De Rham maps \( \text{R}_{\mathcal{X}_c} : S_{\mathcal{X}_c}(c) \rightarrow \mathcal{X}_c \) act on sufficiently smooth fields so that DoFs are single valued. The domain of the de Rham maps can be taken to be, for instance, \( S_{\mathcal{V}_c}(c) = H^{1+\delta}(c) \), \( S_{\mathcal{E}_c}(c) = H^{2+\delta}(c) \), and \( S_{\mathcal{C}_c}(c) = L^2(c) \) with \( \delta > 0 \). Moreover, \( a \in \mathcal{X}_c \) can be viewed as an array of size \( \#\mathcal{X}_c \) since \( \mathcal{X}_c \) is isomorphic to \( \mathbb{R}^{\#\mathcal{X}_c} \). The value of the DoF attached to the entity \( x \in \mathcal{X}_c \) is denoted by \( a_x \in \mathbb{R} \).

**Remark 2** (link with algebraic topology). Elements of \( \mathcal{V}_c \) (resp. \( \mathcal{E}_c, \mathcal{F}_c, \mathcal{C}_c \)) are 0-cochains (resp. 1-, 2-, 3-cochains).

### 3.2. Reconstruction operators

**Definition 6** (local reconstruction operator). Let \( c \in \mathbb{C} \). The local reconstruction operator \( \text{L}_{\mathcal{X}_c} : \mathcal{X}_c \rightarrow P_{\mathcal{X}_c}(c) \) is defined in terms of a family of \( \#\mathcal{X}_c \) linearly independent reconstruction functions \( \{\ell_{x,c}\}_{x \in \mathcal{X}_c} \) spanning the finite-dimensional space \( P_{\mathcal{X}_c}(c) \), called the approximation space, so that the reconstructed field \( \text{L}_{\mathcal{X}_c}(a) \) is defined by

\[
\text{L}_{\mathcal{X}_c}(a)(\bar{x}) := \sum_{x \in \mathcal{X}_c} a_x \ell_{x,c}(\bar{x}), \quad \forall a \in \mathcal{X}_c, \forall \bar{x} \in c.
\]

The reconstruction functions \( \ell_{x,c} \) take values in \( \mathbb{E}_X \) (scalar-valued for potential and density reconstructions, vector-valued for circulation and flux reconstructions). Whenever needed, we underline vector-valued functions and the corresponding reconstruction operators. The finite-dimensional space \( P_{\mathcal{X}_c}(c) \) is for instance spanned by piecewise \( \mathbb{E}_X \)-valued polynomials. We assume that the functions in \( P_{\mathcal{X}_c}(c) \) are in the domain of the local de Rham map \( \text{R}_{\mathcal{X}_c} \), i.e. \( P_{\mathcal{X}_c}(c) \subset S_{\mathcal{X}_c}(c) \).

### 3.3. Discrete Hodge operators

The name “Hodge operator” stems from a concept of differential geometry called the Hodge-star operator [see §14 Chapter 14, for instance]. The Hodge operator embeds a metric (usually induced by a phenomenological parameter) and connects spaces in duality \((k\text{-forms and } (d-k)\text{-forms where } d \text{ is the space dimension and } k \text{ an integer such that } 0 \leq k \leq d)\). So, there are four distinct Hodge operators in a three-dimensional space. As its continuous analogue, a discrete Hodge operator is a metric operator since its definition relies on geometric quantities (lengths, areas, volumes . . . ) and on the evaluation of a material property.

**Hodge inner product.** A discrete Hodge operator can be classically associated with a bilinear form which we call Hodge inner product in what follows.

**Definition 7** (local Hodge inner product). Let \( \mathcal{X}_c \in \{ \mathcal{V}_c, \mathcal{E}_c, \mathcal{F}_c, \mathcal{C}_c \} \). Let \( \alpha \) denote a material property assumed to be cellwise constant, taking values in \( \mathbb{E}_X \otimes \mathbb{E}_X \) (scalar-valued for potentials and densities and tensor-valued for circulations and fluxes), and symmetric positive definite. A local reconstruction operator \( \text{L}_{\mathcal{X}_c} \) or, equivalently, a set of local reconstruction functions \( \{\ell_{x,c}\}_{x \in \mathcal{X}_c} \) defines a local Hodge inner product as follows:

\[
H_{\alpha}^{\mathcal{X}_c}(a_1, a_2) := \int_{c} \text{L}_{\mathcal{X}_c}(a_1) \cdot \alpha \cdot \text{L}_{\mathcal{X}_c}(a_2), \quad \forall a_1, a_2 \in \mathcal{X}_c.
\]  (9)

As previously noticed by Bossavit with the concept of Galerkin Hodge based on Whitney reconstruction functions, the algebraic realization of the Hodge inner product defined by (9) is the mass matrix of the local reconstruction functions weighted by the material property \( \alpha \). Thus, the link between the CDO approach and the FE approach appears naturally since FE shape functions can be used to build reconstruction operators.
**Design strategies.** Discrete Hodge operators are the cornerstone of the CDO approach. Well-posedness, convergence and error estimates hinge on the properties satisfied by this operator [see §4 §14 for elliptic and Stokes problems respectively]. In the CDO framework, the crucial point is thus the design of the discrete Hodge operator or, equivalently, the Hodge inner product. Each definition leads to a different scheme.

In the specific case of Cartesian or Delaunay–Voronoi meshes and an isotropic material property, a diagonal discrete Hodge operator can be built as in the DEC [35] or covolume [36] schemes. In more general situations, it is possible to design a discrete Hodge operator whose algebraic realization is a sparse and symmetric positive definite (SPD) matrix. There are two main design strategies. Either one directly sets the entries of the matrix as in Mimetic Finite Difference (MFD) schemes [12, 37] or one relies on Definition 7 using reconstruction functions. In this paper, we focus on this second strategy.

**Local design properties.** Since a (global) Hodge inner product results from a cellwise assembly process, the design properties are stated locally, i.e. in each mesh cell. The design of the (local) Hodge inner product hinges on the two following properties:

(\textbf{H1}) \textit{Stability}. There is a real number \( \eta_\alpha > 0 \) possibly depending on \( \alpha \) but uniform with respect to \( c \) such that for all \( a \in \mathcal{X}_c \),
\[
\eta_\alpha \|a\|_{\mathcal{X}_c}^2 \leq H^X_\alpha(a, a) \leq \eta_\alpha^{-1} \|a\|_{\mathcal{X}_c}^2,
\]
where \( \|a\|_{\mathcal{X}_c} := \sum_{x \in \mathcal{X}_c} |p_{x,c}| \left( \frac{|a|}{|x|} \right)^2 \) and \( p_{x,c} \) is a subvolume related to a partition of the cell (cf. Section 5.1 and Figure 4). For analysis purposes, \( p_{x,c} \) may be replaced by any equivalent volume (e.g. \( |c| \)) assuming some reasonable local mesh regularity. The only consequence is a modification of the value of \( \eta_\alpha \).

(\textbf{H2}) \textit{P}_0-\textit{consistency}. For any constant field \( K \in \mathbb{E}_X \), the following identity holds for all \( a \in \mathcal{X}_c \):
\[
H^X_\alpha(R_{\mathcal{X}_c}(K), a) = K \cdot \alpha \cdot \left( \sum_{x \in \mathcal{X}_c} a_x \tilde{u}_x(x) \right).
\]

4. Design properties of reconstruction operators

The design of reconstruction operators aims at recovering the two local properties (\textbf{H1}) and (\textbf{H2}) of the Hodge inner product. There are two equivalent approaches, hereafter called \textit{one-level} and \textit{two-level} approach. The one-level approach directly requires properties on the reconstruction operators (or functions), while the two-level approach considers a decomposition of the reconstruction operators (or functions) into a consistent and a stabilization part.

4.1. One-level approach

This approach is considered by Codecasa et al. [20] (except for (R1), see [4]). We require that:

(\textbf{R1}) \textit{Stability}. There exists a real number \( \eta_\mathcal{X} > 0 \) uniform with respect to \( c \) such that for all \( a \in \mathcal{X}_c \),
\[
\eta_\mathcal{X} \|a\|_{\mathcal{X}_c}^2 \leq \|L_{\mathcal{X}_c}(a)\|_{L^2(c)}^2 \leq \eta_\mathcal{X}^{-1} \|a\|_{\mathcal{X}_c}^2.
\]

(\textbf{R2}) \textit{Partition of unity}. For any constant field \( K \in \mathbb{E}_X \), the following identity holds:
\[
L_{\mathcal{X}_c} R_{\mathcal{X}_c}(K) = K.
\]

(\textbf{R3}) \textit{Dual consistency}. The mean-value of \( L_{\mathcal{X}_c} \) satisfies the following identity:
\[
\int_{\mathcal{X}_c} L_{\mathcal{X}_c}(a) = \sum_{x \in \mathcal{X}_c} a_x \tilde{u}_x(x), \quad \forall a \in \mathcal{X}_c.
\]

(\textbf{R4}) \textit{Unisolvence}. \( L_{\mathcal{X}_c} \) is a right inverse of \( R_{\mathcal{X}_c} \), i.e.
\[
R_{\mathcal{X}_c} L_{\mathcal{X}_c}(a) = a, \quad \forall a \in \mathcal{X}_c.
\]
Proposition 2. If the Hodge inner product is built using (7), then the properties (R1)–(R3) imply (H1)–(H2).

Proof. The stability property (H1) results from (R1) and (7) together with the positive-definiteness of \( \alpha \). Let \( K \in \mathbb{E}_X \). Recall that \( \alpha \) is constant in \( c \). For all \( a \in \mathcal{X}_c \), (H2) results from
\[
\mathcal{H}_X^c (R_X^c (K), a) = \int_{c} L_X, R_X^c (K) \cdot \alpha \cdot L_X (a) \quad \text{by (7)},
\]
\[
= \int_{c} K \cdot \alpha \cdot L_X (a) \quad \text{by (R2)},
\]
\[
= K \cdot \alpha \cdot \sum_{x \in \mathcal{X}_c} a_x g_{Xc} (x) \quad \text{by (R3)}.
\]

Therefore, every discrete Hodge operator built from (7) with a reconstruction operator verifying the three properties (R1)–(R3) inherits the properties (H1) and (H2), so that the theoretical results derived in [3, Chapter 6] and [4] hold.

Remark 3 (unisolvence). Observe that the unisolvence property (R4) is not needed to satisfy (H1) and (H2).

Local design properties on reconstruction functions. We now rewrite the properties (R2)–(R4) in terms of reconstruction functions for each type of DoFs. We only state the results since the proofs are straightforward.

Proposition 3 (potential reconstruction functions).

\begin{align}
(R2) & \iff \sum_{v \in \mathcal{V}_c} \ell_{v,c} (x) = 1, \quad \forall x \in c, \\
(R3) & \iff \int_{c} \ell_{v,c} = \tilde{g}_{v,c} (v), \quad \forall v \in \mathcal{V}_c, \\
(R4) & \iff \ell_{v,c} (x') = \delta_{v,v'}, \quad \forall v, v' \in \mathcal{V}_c,
\end{align}

where \( \delta_{\cdot, \cdot} \) is the Kronecker symbol.

Proposition 4 (circulation reconstruction functions).

\begin{align}
(R2) & \iff \sum_{e \in \mathcal{E}_c} \ell_{e,c} (x) \otimes e = \text{Id}, \quad \forall x \in c, \\
(R3) & \iff \int_{c} \ell_{e,c} = \tilde{g}_{e,c} (e), \quad \forall e \in \mathcal{E}_c, \\
(R4) & \iff \int_{e'} \ell_{e,c} \cdot \tau_{e'} = \delta_{e,e'}, \quad \forall e, e' \in \mathcal{E}_c.
\end{align}

Proposition 5 (flux reconstruction functions).

\begin{align}
(R2) & \iff \sum_{f \in \mathcal{F}_c} \ell_{f,c} (x) \otimes f = \text{Id}, \quad \forall x \in c, \\
(R3) & \iff \int_{c} \ell_{f,c} = \tilde{g}_{f,c} (f), \quad \forall f \in \mathcal{F}_c, \\
(R4) & \iff \int_{f'} \ell_{f,c} \cdot \nu_{f'} = \delta_{f,f'}, \quad \forall f, f' \in \mathcal{F}_c.
\end{align}
Remark 4 (density reconstruction). \( L_{c} \) is derived from a single reconstruction function \( \ell_{c} \) since \#\( C_{c} \) = 1. From property (R2), we infer that

\[
\ell_{c}(x) = \frac{1}{|c|}, \quad \forall x \in c.
\] (15)

We easily verify that this definition is in agreement with (R3) (since \( \int_{c} L_{c}(a) = a_{c} \int_{c} \ell_{c} = a_{c} \)) and (R4) (since \( \int_{c} \ell_{c} = 1 \)).

In the remaining part of this paper, we focus on the case \( X \in \{ V, E, F \} \), the case \( X = C \) being straightforward.

Remark 5 (physical dimension). Observe that the reconstruction functions \( \ell_{c,e} \) are dimensionless, \( \ell_{x,c} \) scale as the reciprocal of a length, \( \ell_{f,c} \) scale as the reciprocal of a surface, and \( \ell_{c} \) as the reciprocal of a volume.

Remark 6 (P1-consistency). Whenever the linear completeness property

\[
\sum_{\forall x \in V_{c}} e_{c}(x) = x, \quad \forall x \in c,
\] (16)

holds along with (R2), this induces a P1-consistency property i.e., any affine field \( A \) in \( c \) verifies \( L_{V} R_{V}(A) = A \). Indeed, the field \( A \) can be written as \( A(x) := A(x_{c}) + G(\bar{x} - x_{c}) \) with \( G \) constant in \( c \), so that

\[
L_{V} R_{V}(A(x)) = \sum_{\forall x \in V_{c}} A(x_{c}) e_{c}(x) = A(x_{c}) + G(\bar{x} - x_{c}) = A(x). \tag{21}
\]

4.2. Two-level approach

The second approach operates a decomposition of the reconstruction operator \( L_{X_{c}} \) into a consistent part \( C_{X_{c}} \) and a stabilization part \( S_{X_{c}} \), so that

\[
L_{X_{c}} := C_{X_{c}} + S_{X_{c}},
\] (17)

with consistent part \( C_{X_{c}} \) taking a constant value in \( E_{X} \) defined as follows:

\[
C_{X_{c}}(a) := \frac{1}{|c|} \sum_{x \in X_{c}} a_{c} \tilde{g}_{c}(x), \quad \forall a \in X_{c},
\] (18)

and a stabilization part \( S_{X_{c}} : \mathcal{X}_{c} \to P_{X}(c) \) which is the only user-dependent part in the reconstruction operators. Observe that definition (18) implies

\[
C_{X_{c}} R_{X_{c}}(K) = K, \quad \forall K \in E_{X},
\] (19)

owing to (7). Similar decompositions to (17) have been considered in the context of MFD schemes [12, 29], for the reconstruction of gradients in the context of HFV schemes [25] (cf. also [38, Section 3.3]) and of the generalized Crouzeix–Raviart method of [13], and for the reconstruction of gradients and fluxes in the context of HHO schemes [30, 28, 27].

Local design properties. Since the consistent part of the reconstruction operator is defined by (18), the design properties are stated on \( S_{X_{c}} \) for all \( c \in C \). In addition to (R1) and (R4), we require that:

(R2*) For any constant field \( K \in E_{X}, \)

\[
S_{X_{c}} R_{X_{c}}(K) = 0.
\] (20)

(R3*) For all \( a \in X_{c}, \)

\[
\int_{c} S_{X_{c}}(a) = 0.
\] (21)

In terms of reconstruction functions, the translation of (17) is

\[
\ell_{x,c}(x) := \ell_{c}^{C_{X_{c}}}(x) + \ell_{x,c}^{S_{X_{c}}}(x), \quad \forall x \in X_{c}, \forall x \in c,
\] (22)
where \( c_{x,c} \) and \( \ell_{x,c}^t \) are respectively the consistent and stabilization part of the reconstruction function. We infer from \( (18) \) that
\[
\ell_{x,c}^C (x) := \frac{\tilde{g}_{x_c}(x)}{|c|}, \quad \forall x \in X_c, \forall x_c \in c. \tag{23}
\]
Moreover, \((R2^*)\) and \((R3^*)\) are equivalent to
\[
(R2^*) \iff \sum_{x \in X_c} g_{x_c}(x) \cdot \ell_{x,c}^t (x) = 0, \tag{24a}
\]
\[
(R3^*) \iff \int x \ell_{x,c}^t = 0, \quad \forall x \in X_c. \tag{24b}
\]

**Proposition 6** (link between the two sets of properties). Let \( L_{X_c} = C_{X_c} + S_{X_c} \) with \( C_{X_c} \) defined by \( (18) \). Then, \((R2^*)\) is equivalent to \((R2)\) and \((R3^*)\) to \((R3)\).

**Proof.** \((R2)\) readily results from \((R2^*)\) and \( (19) \). Moreover, \((R3^*)\) yields \( \int L_{X_c} (a) = \int C_{X_c} (a) = \sum_{x \in X_c} a_x \tilde{g}_{x_c} (x) \), so that \((R3)\) holds. The converse statement is proven with similar arguments. \( \square \)

A straightforward consequence of Proposition 6 is that every Hodge inner product built using reconstruction operators such that \((R1)\) holds, the consistent part being defined by \( (18) \) and the stabilization part satisfying properties \((R2^*)\) and \((R3^*)\), inherits the properties \((H1)\) and \((H2)\).

**Proposition 7** (orthogonal decomposition). A reconstruction operator built using \( (17) \) yields a Hodge inner product verifying for all \( a_1, a_2 \in X_c \),
\[
H^X_{\alpha} (a_1, a_2) := \int \sum C_{X_c} (a_1) \cdot \alpha \cdot C_{X_c} (a_2) + \int S_{X_c} (a_1) \cdot \alpha \cdot S_{X_c} (a_2).
\]

**Proof.** This is a consequence of \((R3^*)\) and the fact that \( C_{X_c} \) maps onto constant fields in \( \mathbb{E}_X \). \( \square \)

The consistent part of the Hodge inner product is identical for all choices of the reconstruction operator and is equal, for all \( a_1, a_2 \in X_c \), to
\[
H^X_{\alpha} (a_1, a_2) := \frac{1}{|c|} \sum_{x \in X_c} \sum_{x' \in X_c} a_1 x a_2 x' \tilde{g}_{x_c} (x) \cdot \alpha \cdot \tilde{g}_{x_c} (x').
\]

5. **Piecewise constant reconstruction operators**

The goal of this section is to give an example of reconstruction operators on polyhedral meshes. We reconstruct potential (resp. circulation, flux) fields from DoFs attached to vertices (resp. edges, faces) using a piecewise constant approximation in each mesh cell. \( L_V, L_E, \) and \( L_F \) are nonconforming reconstruction operators which embrace as particular cases, the DGA \( \text{[26]} \) and HFV \( \text{[25]} \) reconstruction operators. This class of reconstruction operators is attractive from an implementation viewpoint since reconstruction operators are explicitly defined, \( i.e. \) they are not the numerical solutions of local problems.

5.1. **Cell partitions**

We first define three partitions of a cell based on the simplicial subdivision introduced in Section 2.

**Definition 8** (partitions of a cell). We set:
\[
P_{V,c} := \bigcup_{v \in E_c, v \in E_c} g(v, e, f), \quad \forall v \in V_c, \tag{25a}
\]
\[
P_{E,c} := \bigcup_{f \in F_c, v \in V_c} g(v, e, f), \quad \forall e \in E_c, \tag{25b}
\]
\[
P_{F,c} := \bigcup_{e \in E_c, v \in V_c} g(v, e, f), \quad \forall f \in F_c. \tag{25c}
\]
The vertex-based partition is denoted by $\mathcal{P}_v := \{v_{x,c}\}_{x \in V_c}$, the edge-based partition by $\mathcal{P}_e := \{e_{x,c}\}_{x \in E_c}$, and the face-based partition by $\mathcal{P}_f := \{f_{x,c}\}_{x \in F_c}$; see Figure 2.

Remark 7 (case $X = C$). Applying the same rationale as in Definition 8 leads to $p_{c,e} := c$.

Remark 8 (link between $p_{c,e}$ and $\tilde{g}_v(v)$). By definition, $|p_{v,c}| = \tilde{g}_v(v)$ for each vertex $v \in V_c$.

We readily verify that $\sum_{x \in X_c} |p_{x,c}| = |c|$ for each cell $c \in C$. Observe also that the following identities hold:

\[ |p_{v,c}| = \tilde{g}_v(v), \quad \forall v \in V_c, \]  
\[ |p_{e,c}| = \frac{1}{d} \tilde{g}_e(e) \cdot \tilde{g}_e(e), \quad \forall e \in E_c, \]  
\[ |p_{f,c}| = \frac{1}{d} \tilde{g}_f(f) \cdot \tilde{g}_f(f), \quad \forall f \in F_c. \]


5.2. Generic definition

Definition 9 (piecewise constant reconstructions). We set $L_{X_c} = C_{X_c} + S_{X_c}$ with $C_{X_c}$ defined by (18) and $S_{X_c} : X_c \rightarrow \mathbb{P}_0(\mathcal{P}_{X_c})$ (the space spanned by $E_X$-valued constant fields in each $p_{x,c}$) defined for all $a \in X_c$ as follows:

\[ S_{X_c}(a) := \tilde{S}_{X_c}(a - R_{X_c} C_{X_c}(a)), \]

where for all $b \in X_c$,

\[ \tilde{S}_{X_c}(b)|_{p_{x,c}} := \beta \frac{\tilde{g}_{x,c}(x)}{|p_{x,c}|} b_x, \quad \forall x \in X_c. \]

$\beta > 0$ is a free-parameter related to the stabilization.

In terms of reconstruction functions, the stabilization part corresponding to Definition 9 is defined as follows:

\[ \ell_{x,c}^{st}|_{p_{x',c}} := \beta \frac{\tilde{g}_{x,c}(x)}{|p_{x',c}|} \left( \delta_{x,x'} - \tilde{g}_{x,c}(x') \otimes \tilde{g}_{x,c}(x') \right). \]

The circulation and flux reconstruction operators proposed in DGA schemes correspond to the choice $\beta = \frac{1}{d}$, while the circulation reconstruction operator proposed in HFV schemes corresponds to the choice $\beta = \frac{1}{\sqrt{d}}$.

Proposition 8. Assume (MB). Then, $S_{X_c}$ specified in Definition 9 verifies properties (R2*) and (R3*).

Proof. (R2*) is a straightforward consequence of (19) and (27). Let us now verify (R3*). Starting from (29), we infer that $\int_e \ell_{x,c}^{st} = 0$ for the last identity.
5.3. Specific definitions

Potential reconstruction operators. \( \mathbf{L}_{V_c} : V_c \to \mathcal{P}_0(\mathcal{P}_{V,c}) \) is defined for all \( p \in V_c \) from the two following contributions:

\[
\mathbf{C}_{V_c}(p) := \frac{1}{|c|} \sum_{v \in V_c} \tilde{g}_{V_c}(v) p_v, \tag{30a}
\]

and, for all \( v' \in V_c \),

\[
\mathbf{S}_{V_c}(p)|_{P_c, v'} = \frac{\beta}{|c|} \sum_{v \in V_c} \tilde{g}_{V_c}(v) (p_{v'} - p_v). \tag{30b}
\]

In terms of reconstruction functions, (30) yields

\[
\ell^{CO}_{V,c} := \frac{\tilde{g}_{V_c}(v)}{|c|}, \quad \forall v \in V_c, \tag{31a}
\]

\[
\ell^{ST}_{V,c}|_{P_c, v'} = \beta \frac{\tilde{g}_{V_c}(v)}{|c|} (\delta_{v, v'} - 1), \quad \forall v, v' \in V_c. \tag{31b}
\]

We observe that the value of these functions is not necessarily continuous across the faces of the partition (induced by \( \mathcal{P}_{V,c} \)) lying inside \( c \), so that, in general, \( \mathbf{L}_{V_c} \) does not map into \( H^1(c) \).

Circulation reconstruction operator. \( \mathbf{L}_{E_c} : E_c \to \mathcal{P}_0(\mathcal{P}_{E,c}) \) is defined for all \( u \in E_c \) from the two following contributions:

\[
\mathbf{C}_{E_c}(u) := \frac{1}{|c|} \sum_{e \in E_c} u_e \tilde{g}_{E_c}(e), \tag{32a}
\]

and, for all \( e' \in E_c \),

\[
\mathbf{S}_{E_c}(u)|_{P_c, e'} = \beta \frac{\tilde{g}_{E_c}(e')}{|c|} (u_{e'} - g_{E_c}(e') \cdot \mathbf{C}_{E_c}(u)). \tag{32b}
\]

In terms of reconstruction functions, (32) yields

\[
\ell^{CO}_{E,c} := \frac{\tilde{g}_{E_c}(e)}{|c|}, \quad \forall e \in E_c, \tag{33a}
\]

and, for all \( e, e' \in E_c \),

\[
\ell^{ST}_{E,c}|_{P_c, e'} = \beta \left( \delta_{e, e'} - \frac{\tilde{g}_{E_c}(e') \otimes g_{E_c}(e')}{|c|} \right) \frac{\tilde{g}_{E_c}(e)}{|c|}. \tag{33b}
\]

We observe that the tangential component of these functions is not necessarily continuous on the edges of the submesh (induced by \( \mathcal{P}_{E,c} \)) lying inside \( c \), so that, in general, \( \mathbf{L}_{E_c} \) does not map into \( H(curl; c) \).

Flux reconstruction operator. \( \mathbf{L}_{F_c} : F_c \to \mathcal{P}_0(\mathcal{P}_{F,c}) \) is defined for all \( \phi \in F_c \) from the two following contributions:

\[
\mathbf{C}_{F_c}(\phi) := \frac{1}{|c|} \sum_{f \in F_c} \phi_{f} \tilde{g}_{F_c}(f), \tag{34a}
\]
and, for all \(f' \in F_c\),
\[
\mathfrak{s}_{F_c}(\phi)|_{P_{f',c}} := \beta \frac{\tilde{g}_{f,c}(f')}{|c|} (\phi_{f'} - g_{f,c}(f') \cdot \mathfrak{s}_{F_c}(\phi)). \tag{34b}
\]
In terms of reconstruction functions, (34) yields
\[
\ell_{\ell,c}^{\mathfrak{s}}(f) := \frac{\tilde{g}_{f,c}(f)}{|c|}, \quad \forall f \in F_c, \tag{35a}
\]
and, for all \(f, f' \in F_c\),
\[
\ell_{\ell,c}^{\mathfrak{s}}|_{P_{f',c}} = \beta \left( \delta_{f,f'} - \frac{\tilde{g}_{f,c}(f') \otimes g_{f,c}(f')}{|c|} \right) \frac{\tilde{g}_{f,c}(f)}{|c|}. \tag{35b}
\]
We observe that the normal component of these functions is not necessarily continuous across the faces of the submesh (induced by \(\mathcal{P}_{F,c}\)) lying inside \(c\), so that, in general, \(\mathbf{L}_{f,c}\) does not map into \(H(\text{div}; c)\).

**Proposition 9** (unisolvence). \(L_{X,c}\) defined from Definition \(\mathcal{B}\) verifies (R4) if and only if \(\beta = 1\) if \(X_c = V_c\) and \(\beta = \frac{1}{d}\) if \(X_c \in \{E_c, F_c\}\).

**Proof.** The case \(V_c\) is readily verified starting from (34). The case \(E_c\) stems from (26b). For all edges \(e \in E_c\), the following identity holds
\[
\int_{\epsilon} \mathcal{L}_{\epsilon,e} \cdot \mathcal{L}_e = \tilde{g}_{\epsilon,e}(e) \cdot \tilde{g}_{\epsilon,e}(e) \left( \frac{1}{|e|} + \frac{\beta}{|p_{\epsilon,e}|} - \frac{\beta g_{\epsilon,e}(e) \cdot \tilde{g}_{\epsilon,e}(e)}{|p_{\epsilon,e}| |e|} \right) = 1 + (\beta d - 1)(1 - \frac{|p_{\epsilon,e}|}{|e|}),
\]
and the right-hand side equals 1 if and only if \(\beta = \frac{1}{d}\). The proof for the case \(F_c\) follows the same lines. \(\square\)

**Remark 9.** The choice \(\beta = \frac{1}{d}\) adopted in HFV schemes has the practical advantage to yield a diagonal discrete Hodge operator when the material property is isotropic and the mesh is superadmissible [see 25, Lemma 2.1]. On the other hand, adapting the arguments of Generalized Crouzeix–Raviart schemes [13, Lemma 8], the choice \(\beta = 1\) allows one to devise a piecewise affine potential reconstruction on the pyramidal submesh \(\bigcup_{c \in C} \mathcal{P}_{F,c}\) with continuous mean values at interfaces of the submesh.

### 6. Application

#### 6.1. CDO schemes for diffusion problems

In this section, we focus on CDO vertex-based schemes for elliptic problems as introduced in [4]. The model problem is
\[
-\text{div}(\underline{\kappa} \text{grad}(p)) = s \quad \text{in} \ \Omega, \tag{36}
\]
where \(p\) is termed the potential, \(\underline{\kappa}\) the conductivity tensor (assumed to be symmetric with eigenvalues uniformly bounded from above and from below away from zero), and \(s\) the source term. We consider Dirichlet boundary conditions. The discrete system is: Find \(p \in V\) such that, for all \(q \in V\),
\[
H^f_{\kappa}(\text{GRAD}(p), \text{GRAD}(q)) = \int_{\Omega} sL_0^V(q). \tag{37}
\]
\(L_0^V\) is defined as the piecewise constant reconstruction detailed in [30] with the choice \(\beta = 1\). The global Hodge inner product is simply defined by collecting the local contributions \(H^f_{\kappa}(u, v) := \sum_{c \in C} H^f_{\kappa}(u, v)\) where \(u_c\) and \(v_c\) are the restriction of the global DoFs to the cell \(c \in C\), i.e. \(u_c, v_c \in E_c\). The discrete gradient operator \(\text{GRAD} : V \rightarrow E\) is defined as follows:
\[
\text{GRAD}(p)|_c = \sum_{v \in V_c} \iota_{v,c} p_v, \quad \forall c \in E, \tag{38}
\]
with the incidence number is such that \(\iota_{v,c} = 1\) if \(v\) points towards \(c\), \(\iota_{v,c} = -1\) otherwise.
6.2. Numerical results

We consider an adaptation of the first test case of the FVCA benchmark [39]. The domain $\Omega$ is the unit cube $[0, 1]^3$, and the exact potential and the conductivity are

$$p(x, y, z) := 1 + \sin(\pi x) \sin(\pi (y + \frac{1}{2})) \sin(\pi (z + \frac{1}{3})), \quad \kappa := \begin{bmatrix} 0.1 & 0.25 & 0 \\ 0.25 & 1 & 0.5 \\ 0 & 0.5 & 10 \end{bmatrix}. \quad (39)$$

The source term and the Dirichlet boundary conditions are set according to (39). Since the global linear system is SPD by construction, it can be efficiently solved using a preconditioned Conjugate Gradient method. Two sequences of three-dimensional polyhedral meshes are tested, each family consisting of successive uniform refinements of an initial mesh. The first mesh sequence, hereafter denoted by $\text{PrG}$, contains prismatic cells with polygonal basis, and the second one, hereafter denoted by $\text{CB}$, checkerboard cells with hanging nodes; see Figure 5. The finest mesh of the $\text{PrG}$ sequence contains approximately 150,000 vertices and 350,000 edges and that of the $\text{CB}$ sequence 250,000 vertices and 700,000 edges.

Figure 5: Two examples of polyhedral meshes. Left: prismatic mesh with polygonal basis; Right: checkerboard mesh with hanging nodes.

Accuracy. We perform a comparative study of the reconstruction operators by computing a discrete error on the potential $\mathcal{E}_V(p)$ and a discrete energy error on the gradient $\mathcal{E}_E(g)$ defined as follows:

$$\mathcal{E}_V(p) := \frac{\|R_V(p) - p\|_V}{\|R_V(p)\|_V}, \quad (40)$$

$$\mathcal{E}_E(g) := \sqrt{\frac{H^{\kappa}_g(R_E(g) - g, R_E(g) - g)}{H^{\kappa}_g(R_E(g), R_E(g))}}, \quad (41)$$

where $\|[a]\|_V := \sum_{c \in C} \sum_{v \in V_c} |p_{x,v}^i| a^2_{x,v}$, $g := \text{grad}(p)$ and $g := \text{GRAD}(p)$. We plot the errors $\mathcal{E}_V(p)$, and $\mathcal{E}_E(g)$ in Figure 6. Four values of the stabilization parameter $\beta$ are considered: an under-penalized value ($\frac{1}{d^3}$) the one used in DGA ($\frac{1}{d}$), the one used in HFV ($\frac{1}{\sqrt{d}}$), and an over-penalized value ($d$). Labels associated with each case are collected in Table 1. We observe that the over-penalized scheme produces a larger error.

<table>
<thead>
<tr>
<th>$\beta$</th>
<th>$\frac{1}{d^3}$</th>
<th>$\frac{1}{d}$</th>
<th>$\frac{1}{\sqrt{d}}$</th>
<th>$d$</th>
</tr>
</thead>
<tbody>
<tr>
<td>PrG</td>
<td>$\bullet$</td>
<td>$\bullet$</td>
<td>$\bullet$</td>
<td>$\bullet$</td>
</tr>
<tr>
<td>CB</td>
<td>$\bullet$</td>
<td>$\bullet$</td>
<td>$\bullet$</td>
<td>$\bullet$</td>
</tr>
</tbody>
</table>

Table 1: Labels associated with each case ($\beta$, mesh sequence).

We also notice a super-convergence in the energy norm for PrG meshes, as already observed in [4].

In Figure 7, we plot the error $\mathcal{E}_E(g)$ for a large set of values of $\beta$ and for the two mesh sequences. Values of $\beta$ around $\frac{1}{\sqrt{d}}$ yield the most accurate results for the present test case.
Figure 6: Discrete error on the potential (left) and discrete error on the gradient (right) for different values of $\beta$ and the two mesh sequences. Dashed (resp. dotted) line indicating second-order (resp. first-order) convergence rates are included.

Figure 7: Error $E_{\mathcal{E}}(g)$ as a function of $\beta$ for each mesh of the PrG sequence (left) and the CB sequence (right).
Cost. In order to compare the efficiency to solve the linear systems produced by the different reconstruction operators, we define the computational cost \( \chi := \text{NNZ} \times n_{\text{ite}} \), where \text{NNZ} is the number of nonzero entries of the matrix to invert and \( n_{\text{ite}} \) is the number of iterations performed by the iterative solver to reduce the Euclidean norm of the residual below a tolerance set to \( 10^{-10} \). \( \chi \) provides a reasonable estimate of the computational cost to solve the linear system since the most costly operation in an iterative solver such as the Conjugate Gradient method is the matrix-vector product. In Figure 8, we plot the computational cost \( \chi \) for a large set of values of \( \beta \) and for the two mesh sequences. The computational cost is higher for the schemes with an over-penalized value of \( \beta \), and it is also slightly higher for an under-penalized value.

Figure 8: Computational cost \( \chi \) as a function of \( \beta \) for each mesh of the PrG sequence (left) and the CB sequence (right).

Preservation of bounds. Finally, we investigate numerically the discrete minimum/maximum principle (DMP). Setting \( p_{\min} := \min_{v \in V} p_v \) and \( p_{\max} := \max_{v \in V} p_v \), we consider that the discrete minimum (resp. maximum) principle is numerically satisfied if \( p_{\min} \geq \min_{\mathbf{x} \in \Omega} p(\mathbf{x}) \) (resp. \( p_{\max} \leq \max_{\mathbf{x} \in \Omega} p(\mathbf{x}) \)). Results are collected in Table 2. \( Y \) indicates that the DMP is satisfied (minimum or maximum) for all the meshes of the sequence and \( N \) indicates that at least one mesh in the sequence does not respect the criterion. Using an under-penalized value of \( \beta \) negatively impacts the DMP.

<table>
<thead>
<tr>
<th>Mesh</th>
<th>( \beta )</th>
<th>( \frac{1}{\sqrt{d}} )</th>
<th>( \frac{1}{d} )</th>
<th>( \frac{1}{d^2} )</th>
<th>( d )</th>
<th>( d^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>PrG</td>
<td>min</td>
<td>N</td>
<td>N</td>
<td>Y</td>
<td>Y</td>
<td>Y</td>
</tr>
<tr>
<td></td>
<td>max</td>
<td>N</td>
<td>N</td>
<td>Y</td>
<td>Y</td>
<td>Y</td>
</tr>
<tr>
<td>CB</td>
<td>min</td>
<td>N</td>
<td>N</td>
<td>N</td>
<td>Y</td>
<td>Y</td>
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<tr>
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<td>N</td>
<td>N</td>
<td>N</td>
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</tr>
</tbody>
</table>

Table 2: Synthesis of results related to the discrete min./max. principle.

7. Conclusion

In this work, we have studied low-order reconstruction operators for polyhedral meshes in a unified framework for degrees of freedom attached to vertices, edges, faces, and cells. These reconstruction operators provide a systematic way of building a Hodge inner product which is a key concept for the compatible numerical approximation of PDEs. We have presented two equivalent sets of design properties. Moreover, a simple example of piecewise constant reconstruction operators depending on a single stabilization parameter
has been detailed, and the influence of this parameter on accuracy and computational costs has been investigated numerically on an anisotropic diffusion problem using CDO vertex-based schemes. Under- and over-penalized values of the stabilization parameter have a negative impact, on the preservation of bounds and on accuracy and costs, respectively. For the problem considered, appropriate choices are values closed to those proposed in DGA and HFV schemes. These conclusions are to be confirmed by further numerical tests.


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