Trivariate Spline Collocation Methods for Numerical Solution to 3D Monge-Ampère Equation

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Abstract
We use trivariate spline functions for the numerical solution of the Dirichlet problem of the 3D elliptic Monge-Ampère equation. Mainly we use the spline collocation method introduced in [SIAM J. Numerical Analysis, 2405-2434, 2022] to numerically solve iterative Poisson equations and use an averaged algorithm to ensure the convergence of the iterations. We shall also establish the rate of convergence under a sufficient condition and provide some numerical evidence to show the numerical rates. Then we present many computational results to demonstrate that this approach works very well. In particular, we tested many known convex solutions as well as nonconvex solutions over convex and nonconvex domains and compared them with several existing numerical methods to show the efficiency and effectiveness of our approach.

Keywords Monge-Ampère equation · Collocation method · Iterative constrained minimization · Spline functions

Mathematics Subject Classification 65N30 · 65K10 · 35J96

1 Introduction

We are interested in numerically solving the Monge-Ampère equation with Dirichlet boundary condition:

\[
\begin{align*}
\det(D^2u(x)) &= f(x), & \text{in } \Omega \subset \mathbb{R}^3 \\
u(x) &= g(x), & \text{on } \partial\Omega,
\end{align*}
\]

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where \( x = (x, y, z) \) has 3 independent variables in a bounded domain \( \Omega \subset \mathbb{R}^3 \) and \( D^2 u \) is the Hessian of the function \( u \), more precisely,

\[
\det(D^2 u) = u_{xx}u_{yy}u_{zz} + 2u_{xy}u_{yz}u_{xz} - u_{xx}(u_{yz})^2 - u_{yy}(u_{xz})^2 - u_{zz}(u_{xy})^2.
\]

(3)

This is a first step toward solving the fully nonlinear Monge-Ampère equation

\[
\det(D^2 u(x)) = f(x)/g(\nabla u(x)), \quad x \in \Omega \subset \mathbb{R}^3
\]

(4)

\[
\nabla u(x)|_{\partial \Omega} = \partial W,
\]

(5)

where the boundary condition is called the oblique boundary condition. Such a partial differential equation arises from the optimal transportation problem (cf. e.g., [25] and [53]). More specifically, given a density function \( f \) on the domain \( \Omega \) and another density function \( g \) on a separate domain \( W \), the goal is to find the optimal plan \( T \) which transports \( f \) over \( \Omega \) to \( g \) over \( W \) under the cost functional \( c(x, w) = \frac{1}{2}\|x - w\|^2 \), with \( \int_{\Omega} f(x)dx = \int_{W} g(w)dw \).

It is Y. Brenier who discovered a characterization of the optimal transportation problem.

**Theorem 1** (Brenier, 1988 [11]) Suppose that the transport cost is the quadratic Euclidean distance, \( c(x, y) = \frac{1}{2}\|x - y\|^2 \) and suppose that \( W \) is a convex domain. Then there exists a convex function \( u : \overline{\Omega} \mapsto \mathbb{R} \) satisfying the Monge-Ampère Eq. (4), unique up to a constant, such that the gradient map \( m = \nabla u \) is the unique optimal transport map satisfying the oblique boundary condition \( \nabla u|_{\partial \Omega} = \partial W \).

Although it is hard to determine the oblique boundary condition mentioned above, once we specify a map from the boundary of \( \Omega \) to the boundary of \( W \), the problem (4) becomes a Neumann boundary problem of the Monge-Ampère equation. In particular, if \( u \) is \( C^2 \) function whose gradient \( \nabla u \) transforms \( \Omega \) onto \( W \), we can move the density \( f(x) \) at \( x \in \Omega \) to the location \( \nabla u(x) \in W \) to become the density \( g(\nabla u(x)) \). Such a problem is called the free movement problem which will be addressed at the end of this paper.

Instead of considering the Neumann or oblique boundary value problem, this paper will focus on the Monge-Ampère equation with a Dirichlet boundary condition. Note that this PDE has been studied for many years. In addition to the mathematical community, the Monge-Ampère equation has also been broadly studied in many applied fields such as elasticity, geometric optics, and image processing. See [14] and [46]. Today such free-form optics are important in illumination applications. For example, they are used in the automotive industry for the construction of headlights that use the full light emitted by the lamp to illuminate the road but at the same time do not glare oncoming traffic [57]. There are multiple ways to solve this inverse reflector problem; brute-force approaches, methods of supporting ellipsoids, simultaneous multiple surfaces approach, and Monge-Ampère approaches. Also, the Monge-Ampère equation finds applications in finance, seismic wave propagation, geostrophic flows, in differential geometry as explained in [17]. In this paper, we shall explain a spline based collocation method to solve the nonlinear PDE (1).

Let us begin recalling some existence, uniqueness, and regularity property of the Monge-Ampère equation (1). When \( f, g \) are sufficiently smooth, the solution of (1) is very smooth explained in the following

**Theorem 2** (Theorem 1 in [19]) Suppose that a bounded domain \( \Omega \in \mathbb{R}^n \) is strictly convex, where \( n \geq 2 \). For any strictly positive right-hand side \( f \in C^\infty(\overline{\Omega}) \) with the boundary condition \( g \) which has an extension \( g \in C^\infty(\overline{\Omega}) \), there exists a unique strictly convex solution \( u \) in \( C^\infty(\Omega) \) satisfying (1).
There are several weaker versions of the existence results with regularity properties in the literature. For example,

**Theorem 3** (Figalli, 2017 [28]) Let $\Omega$ be a uniformly convex domain, $k \geq 2, \alpha \in (0, 1)$, and assume that $\partial \Omega$ is of class $C^{k+2,\alpha}$. Let $f \in C^{k,\alpha}(\bar{\Omega})$ with $f \geq c_0 > 0$. Then for any $g \in C^{k+2,\alpha}(\partial \Omega)$, there exists a unique solution $u \in C^{k+2,\alpha}(\bar{\Omega})$ to the Dirichlet problem (1).

In [4], Awanou introduced another weaker version of the existence theorem:

**Theorem 4** (Awanou, 2013 [4]) Let $\Omega$ be a uniformly convex domain in $\mathbb{R}^n$ with boundary in $C^3$. Suppose $g \in C^3(\bar{\Omega}), \inf f > 0$, and $f \in C^{\alpha}(\bar{\Omega})$ for some $\alpha \in (0, 1)$. Then (1) has a convex solution $u$ which satisfies the a priori estimate

$$\|u\|_{C^{2,\alpha}(\bar{\Omega})} \leq C,$$

where $C$ depends only on $n \geq 2, \alpha, \inf f, \Omega, \|f\|_{C^{\alpha}(\bar{\Omega})}$ and $\|g\|_{C^3}$.

In general, there are at least three different notions of solutions which have been studied in the literature besides the classic solution: one is called Aleksandrov solution, another one is viscosity solution, and the next one is Brainer’s solution, according to the monograph by Villani, 2003, see page 129 in [53]. The theory for the Monge-Ampère equation is deep (cf. [20, 25, 53] and [54]). In particular, the regularity of the solution has been extensively studied (cf. e.g. [18, 22, 55]). In a landmark paper [18], Caffarelli showed that the solution of the Monge-Ampère equation has an interior regularity over $\Omega' \subset \Omega$, i.e. $u \in H^{2,\rho}(\Omega')$ for any open set $\Omega'$ inside $\Omega$. Furthermore, the solution has $H^2$ regularity over the entire domain, as established in [55]:

**Theorem 5** (Wang, 1996 [55]) Let $\Omega$ be a strictly convex domain in $\mathbb{R}^n$. If $\partial \Omega$ and $g$ in the equation (2) are $C^3$ smooth, and $f(x) \in C^{1,1}(\bar{\Omega})$, then the solution $u \in C^{2+\alpha}(\bar{\Omega})$.

Due to these regularity results, we can use $C^2$ smooth trivariate splines to approximate the solution $u$ under the conditions $f \in C^{1,1}(\bar{\Omega}), g \in C^3(\partial \Omega)$ and $\Omega$ being a strictly convex domain. In our computation, we are able to solve the Monge-Ampère equation over domains with uniform positive reach (cf. [31]) which include strictly convex domains as a special case. Additionally, we can use our method to experiment with the solution (1) even when $f$ is not in $C^{1,1}(\bar{\Omega})$.

The numerical solution of the Monge-Ampere equation (MAE) is an active area of research, with many researchers developing different numerical methods and analyzing their theoretical convergence. As mentioned in [12], the MAE poses several challenges for numerical solutions. The first challenge is that the equation is fully nonlinear, which means that geometric solutions or viscosity solutions must be used as weak solutions. The second challenge is the convex constraint, as the equation might not have a unique solution without it.

The popular finite element method is not directly applicable because of the involvement of the Hessian of the solution. This restricts the use of the Finite Element Method (FEM) or general Galerkin projection methods, discontinuous Galerkin method or continuous Galerkin method. However, there are several remedy approaches based on the finite element method such as a mixed finite element method, vanishing moment method, etc. See [3], [5], [9] and [27].

Besides of finite element type methods, there are many finite difference methods, as seen in [7, 12, 13, 43, 44]. Moreover, many interesting approaches are based on the classic finite difference method as demonstrated in [13, 15, 45]. However, these methods have a weakness:
they do not have analytic form of solution over the entire domain. In addition, we can find time marching methods in [4, 6], and least squares relaxation methods in [17, 23, 24].

Let us be more precise on the numerical methods mentioned above. The least square notion of the solution was proposed and studied in [23, 24], and [17]. Especially, this least square approach using a relaxation algorithm of the Gauss-Seidel-type iterations to decouple differential operators in [17]. The approximation relies on mixed low order finite element methods with regularization techniques. Several 3D examples were demonstrated to show the performance of this method. In this paper, we will compare the numerical results from our method to those to in [17] to show that our method produces more accurate results. These comparisons will be presented in the last section.

In [6], a time marching approach is used to solve the Monge-Ampére equation. Given \( \nu > 0 \), the researcher considered the sequence of iterates

\[
-\nu \Delta u_{k+1} = -\nu \Delta u_k + \det D^2 u_k - f, \quad u_{k+1} = g \text{ on } \partial \Omega.
\]

He used the discrete version of Newton’s method in the vanishing moment methodology. And he showed the convergence of the iterative method for solving the nonlinear system. We shall also compare his numerical results with our results in the last section to show that our proposed method is also more accurate.

In [13], the researchers introduced the meshless Generalized Finite Difference Method (GFDM) in both 2D and 3D settings. They tested several examples using the Cascadic iterative algorithm over convex and non-convex domains. We will compare our proposed method with the results from the Cascadic iterative algorithm in the last section to demonstrate that our method is also better.

We now describe our numerical method to solve (1) by using trivariate spline functions over a tetrahedralization of \( \Omega \). See [8, 39, 40, 51] for theoretical properties and numerical implementation of bivariate/triavariate spline functions. In addition, there are several dissertations written to explain how to implement and how to use multivariate splines for the numerical solution of Helmholtz equations, Maxwell equations and 3D surface reconstruction. See [2, 47] and [56]. There are several reasons why we use trivariate splines for the numerical solution of the Monge-Ampére equation. One is that we can use trivariate splines with smoothness \( r \geq 2 \) to approximate the solution \( u \) of the Monge-Ampére equation over an arbitrary convex polyhedral domain. Due to the \( C^2 \) smoothness of spline function, we can calculate the Hessian of the solution, so that we simply use the collocation method instead of the weak formulations in [3, 5], etc.. Many researchers adopted the iterative algorithm called the fixed point algorithm introduced in [12]:

\[
\Delta u_{k+1} = ((\Delta u_k)^n + a(f - \det D^2 u_k))^{1/2} \quad (7)
\]

along with the prescribed Dirichlet boundary conditions with \( a = 2 \) and \( n = 2 \). The researchers in [12] explained that this is a fixed point method as the true solution \( u \) satisfies (7) trivially.

In [6], this iterative algorithm is generalized to the 3D setting with \( n \geq 3 \) for various \( a > 0 \). In particular, the researcher in [6] explained that the iteration (7) is well-defined for \( a \leq n^a \) as \( \det (D^2 u_k) \leq \frac{1}{n^2} (\Delta u)^n \). Numerical results in [6] are demonstrated in the framework of the spline element method with \( a = 2 \) for the 2D case and \( a = 9 \) for the 3D case.

In this paper, we shall use the following iterative method:

\[
\Delta u_{k+1} = ((\Delta u_k)^n + n^n (f - \det D^2 u_k))^{1/2} \quad (8)
\]
to handle the nonlinearity of the Monge-Ampère equation where \( n = 3 \).

However, another requirement of the solution of the Monge-Ampère equation is that \( u \) must be convex in order for the equation to be elliptic. Without this constraint, the equation does not have a unique solution. (For example, taking boundary data \( g = 0 \), if \( u \) is a solution, then \(-u\) is also a solution in \( \mathbb{R}^2 \).) Many numerical methods mentioned above failed to enforce this convexity constraint. The convexity of \( u \) is equivalent to the positive definiteness of the Hessian matrix \( D^2 u \). In terms of the eigenvalues \( \lambda_1 \geq \lambda_2 \geq \lambda_3 \) of \( D^2 u \), we will ensure that three eigenvalues \( \lambda_1(k) \geq \lambda_2(k) \geq \lambda_3(k) \) of the \( k \)th iteration \( u_k \) in a spline space satisfy \( \lambda_1(k) + \lambda_2(k) + \lambda_3(k) \geq 0 \) as well as \( \lambda_1(k)\lambda_2(k)\lambda_3(k) > 0 \), although they are not enough to ensure the convexity of \( k \)th spline solution \( u_k \).

This paper is organized as follows. In Sect. 2, we first explain trivariate splines, domains with uniformly positive reach, and the spline collocation method for the Poisson equation which is the same as the one discussed in [39]. In Sect. 3, we introduce the spline collocation method for the Monge-Ampère equation and its average algorithm, and establish three different versions of convergence results. Finally, in the last Sect. 4, we present numerical results for several 3D examples of smooth and convex solutions, as well as nonsmooth convex solution over convex and nonconvex bounded domains to demonstrate the effectiveness of our proposed method. We compare our results with those of several existing numerical methods to show the accuracy and efficiency of our method. Finally, we shall present some examples for the free movement in 2D and 3D settings to show how the density from one place is moved to another place. This will demonstrate further that our proposed method is versatile enough.

2 Preliminaries

2.1 Trivariate Splines

Let us quickly summarize the essentials of trivariate splines in this section. Given a tetrahedron \( T \), we write \( |T| \) for the length of its longest edge, and \( \rho_T \) for the radius of the largest inscribed ball in \( T \). For any polygonal domain \( \Omega \subset \mathbb{R}^3 \), let \( \Delta := \{ T_1, \ldots, T_n \} \) be a tetrahedralization of \( \Omega \) which is a collection of tetrahedra and \( \mathcal{V} \) be the set of vertices of \( \Delta \). We called a tetrahedralization as a quasi-uniform tetrahedralization if all tetrahedra \( T \) in \( \Delta \) have comparable sizes in the sense that

\[
\frac{|T|}{\rho_T} \leq C < \infty, \quad \text{all tetrahedra } T \in \Delta,
\]

where \( \rho_T \) is the inradius of \( T \). Let \( |\Delta| \) be the length of the longest edge in \( \Delta \).

Next for a tetrahedron \( T = (v_1, v_2, v_3, v_4) \in \Delta \), we define the barycentric coordinates \((b_1, b_2, b_3, b_4)\) of a point \((x, y, z) \in \Omega\) as the solution to the following system of equations

\[
\begin{align*}
\quad b_1 + b_2 + b_3 + b_4 &= 1 \\
\quad b_1v_{1,x} + b_2v_{2,x} + b_3v_{3,x} + b_4v_{4,x} &= x \\
\quad b_1v_{1,y} + b_2v_{2,y} + b_3v_{3,y} + b_4v_{4,y} &= y \\
\quad b_1v_{1,z} + b_2v_{2,z} + b_3v_{3,z} + b_4v_{4,z} &= z,
\end{align*}
\]

where the vertices \( v_i = (v_{i,x}, v_{i,y}, v_{i,z}) \) for \( i = 1, 2, 3, 4 \). \( b_1, \ldots, b_4 \) are nonnegative if \((x, y, z) \in T\). Next we use the barycentric coordinates to define the Bernstein polynomials of degree \( D \):
for a positive constant $C$

Let $D$ be the coefficient vector associated with spline function $s_T$ where the $B$-coefficients $c_T^{i,j,k,\ell}$ are uniquely determined by $s$. Let $c = \{c_T^{i,j,k,\ell}, i + j + k + \ell = D, T \in \Delta\}$ be the coefficient vector associated with spline function $s$.

Moreover, for given $T = (v_1, v_2, v_3, v_4) \in \Delta$, we define the associated set of domain points to be

$$\mathcal{D}_{D,T} := \left\{ \frac{i v_1 + j v_2 + k v_3 + \ell v_4}{D} \right\}_{i+j+k=D}. \tag{9}$$

Let $\mathcal{D}_{D,\Delta} = \bigcup_{T \in \Delta} \mathcal{D}_{D,T}$ be the domain points of tetrahedral $\Delta$ and degree $D$.

We use the discontinuous spline space $S_D^r(\Delta) := \{s \mid T \in \mathcal{P}_D, T \in \Delta\}$ as a base. Then we add the smoothness conditions to define the space $S_D^r := C^r(\Omega) \cap S_D^{-1}(\Delta)$. The smoothness conditions are explained in [40]. Indeed, see Theorem 15.31 in [40]. We use $C^r$ smooth spline functions in $H^2(\Omega)$ with $r \geq 1$ and the degree $D$ of splines sufficiently large satisfying $D \geq 3r + 2$ in $\mathbb{R}^2$ and $D \geq 6r + 3$ in $\mathbb{R}^3$. And we get the following Lemma in [40]

**Lemma 1** For all $u \in W^{m+1,p}(\Omega)$ for some $0 \leq m \leq D$ and $1 \leq p \leq \infty$, there exists a quasi-interpolatory spline $s_u \in S_D^r(\Delta)$ such that

$$\|D^\alpha(u - s_u)\|_{L^p(\Omega)} \leq C|\Delta|^{m+1-|\alpha|} |\alpha| |u|_{m+1,p,\Omega},$$

for all $0 \leq |\alpha| \leq m$, where $|\cdot|_{m+1,p,\Omega}$ is a semi-norm, $C$ is a positive constant independent of $u$ and $|\Delta|$ but is dependent on the geometry of $\Delta$.

In addition, we shall use Markov inequality(cf. [40]):

$$\|\nabla s\|_{\infty,\Omega} \leq C |\Delta|^{-1} \|s\|_{\infty,\Omega}, \forall s \in S_D^r(\Delta) \tag{10}$$

for a positive constant $C$ independent of $s$ and the size $|\Delta|$ of tetrahedralization $\Delta$.

### 2.2 Domains with Uniformly Positive Reach

Let us recall a concept on domains of interest explained in [31].

**Definition 1** Let $K \subseteq \mathbb{R}^n$ be a non-empty set. Let $r_K$ be the supremum of the number $r$ such that every points in

$$P = \{x \in \mathbb{R}^n : \text{dist}(x, K) < r\}$$

has a unique projection in $K$. The set $K$ is said to have a positive reach if $r_K > 0$.

A domain with $C^2$ boundary has a positive reach. Sets of positive reach are much more general than convex sets. Let $B(0, \epsilon)$ be the closed ball centering at 0 with radius $\epsilon > 0$, and let $K^c$ stand for the complement of the set $K \subseteq \mathbb{R}^n$. For any $\epsilon > 0$, the set

$$E_\epsilon(K) := (K^c + B(0, \epsilon))^c \subseteq K$$

is called an $\epsilon$-erosion of $K$. Next we recall the following definition from [31].
Definition 2 A set $K \subseteq \mathbb{R}^n$ is said to have a uniformly positive reach $r_0$ if there exists some $\epsilon_0 > 0$ such that for all $\epsilon \in [0, \epsilon_0]$, $E_\epsilon(K)$ has a positive reach at least $r_0$.

Many examples of domains with positive reach can be found in [31]. And we have the following property about these domains

Lemma 2 If $\Omega \subset \mathbb{R}^n$ is of positive reach $r_0$, then for any $0 < \epsilon < r_0$, the boundary of $\Omega_\epsilon := \Omega + B(0, \epsilon)$ containing $\Omega$ is of $C^{1,1}$. Furthermore, $\Omega_\epsilon$ has a positive reach $\geq r_0 - \epsilon$.

In [31], Gao and Lai proved the following regularity theorem

Theorem 6 Let $\Omega$ be a bounded domain. Suppose the closure of $\Omega$ is of uniformly positive reach $r_\Omega$. For any $f \in L^2(\Omega)$, let $u \in H^1_0(\Omega)$ be the unique weak solution of the Dirichlet problem:

$$\begin{cases} -\Delta u = f & \text{in } \Omega \\ u = 0 & \text{on } \partial\Omega \end{cases}$$

Then $u \in H^2(\Omega)$ in the sense that

$$\sum_{i,j=1}^n \int_\Omega \left( \frac{\partial^2 u}{\partial x_i \partial x_j} \right)^2 \leq C_0 \int_\Omega f^2 \, dx$$

for a positive constant $C_0$ depending only on $r_\Omega$.

### 2.3 A Collocation Method for the Poisson Equation

For convenience, let us start with the Poisson equation

$$\begin{align*}
\Delta u(x) &= f(x) \quad \forall x \in \Omega \subset \mathbb{R}^3 \\
u(x) &= g(x), \quad \forall x \in \partial\Omega \quad (11)
\end{align*}$$

For given $\Delta$, let it be a tetrahedral partition of $\Omega$, we choose a set of domain points $\{\xi_i\}_{i=1,\ldots,N}$ explained in the previous section as collocation points and let $s = \sum_{t \in \Delta} \sum_{|\alpha| = D} c_{t,\alpha} B_{t,\alpha}^\epsilon$ in $S^r_{D'}(\Delta)$ with the coefficient vector $c$ of $s$. Then we want to find the coefficient vector $c$ of spline function satisfying the standard Poisson Eq. (11) at those collocation points

$$\begin{cases} \Delta s(\xi_i) = f(\xi_i), & \xi_i \in \Omega \subset \mathbb{R}^n, \\
s(\xi_i) = g(\xi_i), & \xi_i \in \partial\Omega, \\
\end{cases} \quad (12)$$

where $\{\xi_i\}_{i=1,\ldots,N} \in \mathcal{D}_{D',\Delta}$ are the domain points of $\Delta$ of degree $D' > 0$ as explained in (9) in the previous section, where $D'$ will be different from $D(D' > D)$.

Using these points, we let $K$ be the following matrix:

$$K := [\Delta(B_{i,j,k,l})^\epsilon(\xi_i)].$$ 

In general, the spline $s$ with coefficients in $c$ is a discontinuous function. In order to make $s \in S^r_{D'}(\Delta)$, its coefficient vector $c$ must satisfy the constraints $Hc = 0$ for the smoothness conditions that the $S^r_{D'}(\Delta)$ functions possess (cf. [40]). Based on the smoothness conditions (cf. Theorem 2.28 or Theorem 15.38 in [40]), we can construct matrices $H$ for the $C^r$ smoothness conditions. Then, our collocation method is to find $c$ which solves the following constrained minimizations:

$$\min_c J(c) = \frac{1}{2} (\|Bc - G\|^2 + \|Hc\|^2) \quad (13)$$
subject to $Kc = f$ \hfill (14)

where $B, G$ are from the boundary condition and $H$ is from the smoothness condition. Note that we need to justify that the minimization has a solution. In general, we do not know if $Kc = f$ has a solution or not. However, we can show that a neighborhood of $Kc = f$, i.e.

$$
N = \{ c : \| Kc - f \| \leq \epsilon \}
$$

is not empty for an appropriate $\epsilon > 0$ when $|\Delta|$ is small enough. Indeed, let us use spline approximation theorem (cf. [40]) to have

**Lemma 3** Suppose that $\Omega$ is a polygonal domain. Suppose that $u \in H^3(\Omega)$. Then there exists a spline function $u_s \in S_D^r(\Delta)$ and a positive constant $\hat{C}$ depending on $D \geq 1$ and $D' > D$ such that

$$
\| \Delta u(x, y, z) - \Delta u_s(x, y, z) \|_{L^\infty(\Omega)} \leq \epsilon_1 \hat{C}
$$

where $\epsilon_1$ is a given tolerance and $\hat{C}$ depends on $\Omega$, $D$, $D'$. We thus consider a nearby problem of the minimization (13), which is:

$$
\min_c \| Bc - G \|^2 + \| Hc \|^2,
$$

subject to $\| Kc - f \|_{L^\infty} \leq \epsilon_1$. \hfill (16)

It is easy to see that the minimizer of (16) clearly approximates the minimizer of (13) if $\epsilon_1 \ll 1$. As the new minimization problem is convex and the feasible set is also convex, the minimization (16) will have a unique solution if the feasible set is non-empty.

We may assume that our numerical solution $u_s$ approximates $u$ on $\partial\Omega$ very well in the sense that $\| u - u_s \| \leq C \epsilon_2$ for a positive constant $C$. Denote $\| u \|_L := \| \Delta u \|_{L^2(\Omega)}$. In [39], Lai and Lee proved the following theorems

**Theorem 7** Suppose $f$ and $g$ are continuous over bound domain $\Omega \subseteq \mathbb{R}^d$ for $d = 2$ or $d = 3$. Suppose that $u \in H^3(\Omega)$. When $\Omega$ is a domain with uniform positive reach, we have the following inequality

$$
\| u - u_s \|_{L^2(\Omega)} \leq C \| u - u_s \|_{L}, \| \nabla (u - u_s) \|_{L^2(\Omega)} \leq C \| u - u_s \|_{L}
$$

and

$$
\sum_{i+j=2} \| \frac{\partial^2}{\partial x^i \partial y^j} u \|_{L^2(\Omega)} \leq C \| u - u_s \|_{L}
$$

for a positive constant $C$ depending on $\Omega$.

And we can obtain the better convergence results if we assume that $|u - u_s|_{\partial\Omega} = 0$, as shown in the following theorem:

**Theorem 8** Suppose that $|u - u_s|_{\partial\Omega} = 0$. Under the assumptions in Theorem 7, we have the following inequality

$$
\| u - u_s \|_{L^2(\Omega)} \leq C |\Delta|^2 (\| u - u_s \|_{L}) \text{ and } \| \nabla (u - u_s) \|_{L^2(\Omega)} \leq C |\Delta| (\| u - u_s \|_{L})
$$

for a positive constant $C$, where $|\Delta|$ is the size of the underlying tetrahedral $\Delta$.

**Proof** We use the same arguments as in [39] to establish a proof. \hfill $\blacksquare$
3 Our Proposed Algorithms and Their Convergence Analysis

3.1 A Spline Based Collocation Method for Monge-Ampère Equation

For convenience, let us explain our spline based collocation method for the 3D Monge-Ampère equation first. In the 3D setting, we will solve the following iterative equations as in [6]

\[ \Delta u_{k+1} = \sqrt[3]{(\Delta u_k)^3 + a(f - \det(D^2u_k))}, \quad k = 0, 1, \cdots, \]  

(18)

with an initial \( u_0 \) by solving

\[ \Delta u_0 = \sqrt[3]{27f} \]

together with the given boundary condition. Let us make two quick remarks. The initial \( u_0 \) so chosen is based on the following assumption. Writing \( \lambda_i, i = 1, 2, 3, \) to be the eigenvalues of \( \det(D^2u) \), the Monge-Ampère equation reads \( \lambda_1\lambda_2\lambda_3 = f \). If these eigenvalues are close to each other, e.g. all are equal to \( \lambda \), we have \( \lambda^3 = f \) and thus \( \lambda = \sqrt[3]{f} \). Since \( \Delta u = \lambda_1 + \lambda_2 + \lambda_3 \), we get \( \Delta u = 3\lambda = 3\sqrt[3]{f} = \sqrt[3]{27f} \). However, when these eigenvalues are quite different, such a choice of initial \( u_0 \) may not be a good one. We shall explain our approach later in this section.

Next about the parameter \( a \) in (18), we have tested different numbers for \( a \). Figure 1 shows that we can get more accurate results when using \( a = 27 \). We will use \( a = 27 \) in the rest of the paper.

For simplicity, we use the Dirichlet boundary condition \( u|_{\partial\Omega} = g \) to explain our numerical method.

3.2 Two Computational Algorithms

We shall present two computational algorithms for numerical solution of the Monge-Ampère equation. The first one is a standard approach which has been used by many researchers
based on finite different discretization and finite element discretization in the literature. We shall use multivariate spline functions to discretize the function space $H^2(\Omega)$ to demonstrate the efficiency and effectiveness of our multivariate spline approach as well as show that our numerical results are better than many methods in the literature. See our computational results in the next section.

**Algorithm 1: An Iterative Poisson Equation Algorithm**

Start with an initial solution $u_0$ by solving the following Poisson equation using our collocation method discussed in the previous section:

$$\Delta u_0 = \sqrt[3]{27f}$$

and $u_0 = g$ on the boundary $\partial \Omega$. We then iteratively solve the Poisson equation

$$\Delta u_{k+1} = \frac{3}{\sqrt[3]{\Delta u_k}}^3 + 27(f - \det(D^2 u_k)), \quad k = 0, 1, \ldots .$$

That is, we find $u_{k+1} \in S_\Delta(\triangle)$ satisfying the following equations approximately:

$$\begin{cases}
\Delta u_{k+1}(\xi_i) = \frac{3}{\sqrt[3]{\Delta u_k(\xi_i)}}^3 + 27(f(\xi_i) - \det(D^2 u_k(\xi_i))) \quad \xi_i \in \Omega \subset \mathbb{R}^3, \\
u_{k+1}(\xi_i) = g(\xi_i), \quad \xi_i \in \partial \Omega
\end{cases}$$

In other words, we numerically solve (16) with $f_k = \frac{3}{\sqrt[3]{\Delta u_k(\xi_i)}}^3 + 27(f(\xi_i) - \det(D^2 u_k(\xi_i)))$ by using the iterative algorithm in [39].

Terminate the iteration when $\|f - \det(D^2 u_{k+1})\|_{\infty} > \|f - \det(D^2 u_k)\|_{\infty}$.

Next we explain an averaged iterative algorithm. Assume $\Omega$ is bounded and the closure of $\Omega$ is of uniformly positive reach as explained in the previous section. For any $f \in L^2(\Omega)$, the solution of the Poisson equation with zero boundary condition is in $H^2(\Omega)$ by Theorem 6. Furthermore, the solution of the Poisson equation with boundary condition $g$ is in $H^2(\Omega)$ if $g \in H^{1/2}(\partial \Omega)$. Indeed, we consider a function $v \in H^2(\Omega)$ whose trace on $\partial \Omega$ is $g \in H^{1/2}(\partial \Omega)$. Define $w = u - v$ and have

$$\int_{\Omega} \nabla w \cdot \nabla \phi = \int_{\Omega} \nabla u \cdot \nabla \phi - \int_{\Omega} \nabla v \cdot \nabla \phi = \int_{\Omega} - f \cdot \phi - \int_{\Omega} \Delta v \cdot \phi = \int_{\Omega} (\Delta v - f) \cdot \phi.$$

for every $\phi \in H^1_0(\Omega)$. Then the solution $w$ satisfying the weak formulation of

$$\Delta w = f - \Delta v \text{ in } \Omega, \ w = 0 \text{ on } \partial \Omega$$

is in $H^2(\Omega)$ (cf. [31]). Therefore, $u = w + v$ is in $H^2(\Omega)$.

Let $T$ be an operator which maps $H^2(\Omega) \to H^2(\Omega)$ in the following sense: for any $v \in H^2(\Omega)$, let $u = T(v)$ be the solution of the Poisson equation:

$$\Delta u = \sqrt[3]{(\Delta v)^3 + 27(f - \det(D^2 v))} \text{ over } \Omega$$
and $u|_{\partial\Omega} = g$ with $g \in H^{1/2}(\partial\Omega)$. In other words, the operator $T$ on $H^2(\Omega)$ is defined by

$$T(u) = \Delta^{-1}\left[\frac{3}{2} (\Delta u)^3 + 27(f - \det(D^2u))\right].$$

It is easy to see that $T$ is a nonlinear operator $T$ maps $H^2(\Omega)$ to $H^2(\Omega)$. Also, we can see that the exact solution $u^*$ satisfying $\det(D^2u^*) = f$ is a fixed point of $T$.

Now we are ready to define an averaged iterative algorithm. In this way, we can find more accurate solutions than the one using Algorithm 1 only.

**Algorithm 2: The Averaged Iterative Algorithm**

Start with an initial $u_0$, where $\Delta u_0 = \sqrt[3]{27f}$ over $\Omega$ and $u_0 = g$ on $\partial\Omega$. We iteratively solve the Poisson equations

$$\Delta u_{k+1/2} = \sqrt[3]{(\Delta u_k)^3 + 27(f - \det(D^2u_k))},$$

(22)

together with the boundary condition $u_{k+1/2} = g$ on $\partial\Omega$ by using the minimization in (16) and then take

$$u_{k+1} = \frac{1}{2}u_{k+1/2} + \frac{1}{2}u_k.$$  

(23)

Stop the iteration if $\|f - \det(D^2u_{k+1})\|_{l_\infty} > \|f - \det(D^2u_k)\|_{l_\infty}$.

Let us present some performance of these two algorithms to show that Algorithm 2 indeed very useful. Consider a testing function $u^3$ as in Sect. 4, the eigenvalues of the Hessian matrix $D^2u^3$ are 1, 5, 15. Although these three eigenvalues are not close to any real positive number, we use various positive numbers $p$ for the right-hand side of the Poisson equation $\Delta u_0 = p$ to solve $u_0$ as an initial solution and then apply Algorithm 1 and Algorithm 2. In Table 1, the results from both Algorithms are shown after the same number of iterations. We can see that Algorithm 2 produces more accurate solution than Algorithm 1 from various initial values except for $p$ which is close to $21 = \Delta u^3_{3ds1}$, i.e. $p \in [17.7, 26]$. Also, the $\ell_2$ and $h_1$ errors from Algorithm 2 are better than the errors from Algorithm 1 for testing functions $u^3_{3ds1}$ and $u^3_{3ds8}$ as shown in Fig. 2.

### 3.3 Convergence Analysis

According to [5], it is known that if $\det(D^2u^*) = f > 0$ and $u^*$ is convex, then there exist constants $m, M > 0$, independent of the mesh size $|\Delta|$ such that

$$0 < m \leq \lambda_3 \leq \lambda_2 \leq \lambda_1 \leq M,$$

where $\lambda_1, \lambda_2, \lambda_3$ are the eigenvalues of $(D^2u(x)), \forall x \in \Omega$. The following result is also known (cf. [6]). For clarity, we provide a proof below.

**Lemma 4** Suppose that the convex solution $u^* \in W^{2,\infty}$ satisfies $\det(D^2u^*) = f > 0$. There exists a $\delta > 0$ such that for any $u$ which is close enough to the exact solution $u^*$ in the sense that $|u - u^*|_{2,\infty} \leq \delta$, we have

$$\det(D^2u) \leq \frac{1}{27}(\Delta u)^3 < \frac{1}{a}(\Delta u)^3$$

for any $a < 27$.  

$$\textcircled{123}$$
Table 1 Errors of numerical solutions $u^{3ds1}$ for the Monge–Ampère equation over $[0, 1]^3$ with $D = 9, r = 1$ over the same tetrahedralization for various initial values $p$ by two algorithms

| $\Delta u_0 = p$ | Algorithm 1 $|e_s|_{l_2}$ | Algorithm 1 $|e_s|_{h_1}$ | Algorithm 2 $|e_s|_{l_2}$ | Algorithm 2 $|e_s|_{h_1}$ |
|----------------|-----------------|-----------------|-----------------|-----------------|
| 12.6           | 2.1291e−02      | 1.6315e−01      | 1.9230e−02      | 1.3670e−01      |
| 15.1           | 3.4135e−03      | 5.8902e−02      | 5.0004e−03      | 5.1503e−02      |
| 16.4           | 2.0124e−03      | 3.1978e−02      | 1.2081e−08      | 5.4356e−07      |
| 17.1           | 7.9130e−04      | 1.4517e−02      | 4.1401e−08      | 1.6448e−06      |
| 17.7           | 1.3980e−09      | 3.9929e−08      | 3.6103e−08      | 2.7446e−06      |
| 26.0           | 6.4074e−09      | 2.4003e−07      | 4.8324e−07      | 1.8097e−05      |
| 26.5           | 2.0899e−04      | 6.3176e−03      | 5.0149e−07      | 1.8781e−05      |
| 27.0           | 5.6423e−04      | 1.7172e−02      | 3.9592e−04      | 1.2049e−02      |
| 27.5           | 8.9037e−04      | 2.5381e−02      | 7.0701e−04      | 2.0696e−02      |

Fig. 2 Errors $\log(|e_s|_{l_2}), \log(|e_s|_{h_1})$ for $u^{3ds3}$ (Top) and $u^{3ds8}$ (Bottom)
Proof Recall that the eigenvalues of a symmetric matrix are continuous functions of its entries, as roots of the characteristic equation (cf. Ostrowski (1960) Appendix K [49]). Thus, for a given \( u^* \in W^{2,\infty}(\Omega) \), there exists \( \delta > 0 \) such that for \( u \in W^{2,\infty}(\Omega) \), \( |u - u^*|_{2,\infty} \leq \delta \) implies \( M \geq M_1(D^2u(x)) \geq M_2(D^2u(x)) \geq M_3(D^2u(x)) > 0 \). Now, we use the property that \( \det(D^2u) \) is the multiplication of all eigenvalues to have

\[
\det(D^2u) = \lambda_1 \lambda_2 \lambda_3 = \frac{1}{27} (3\lambda_1 \lambda_2 \lambda_3 + 6\lambda_1 \lambda_2 \lambda_3 + 18\lambda_1 \lambda_2 \lambda_3)
\]

\[
\leq \frac{1}{27} (\lambda_1^3 + \lambda_2^3 + \lambda_3^3 + 6\lambda_1 \lambda_2 \lambda_3 + 3\lambda_3 (\lambda_1^2 + \lambda_2^2) + 3\lambda_1 (\lambda_2^2 + \lambda_3^2) + 3\lambda_2 (\lambda_1^2 + \lambda_3^2))
\]

\[
= \frac{1}{27} (\lambda_1 + \lambda_2 + \lambda_3)^3 = \frac{1}{27} (\Delta u)^3.
\]

This completes all the proof. \( \square \)

We first consider the point-wise convergence of the sequence from Algorithm 1.

**Theorem 9** Fix a spline space \( S^p_D(\Delta) \) with \( \Delta \) being a tetrahedralization of the domain \( \Omega \). Let \( u_k \in S^p_D(\Delta), k \geq 1 \) be the sequence from Algorithm 1. Then, any average values of \( f - \det(D^2u_k), k \geq 1 \) are nonnegative in the following senses:

\[
\frac{1}{n+1} \sum_{k=0}^{n} (f(x) - \det(D^2u_k)(x)) \geq 0, \quad x \in \Omega \tag{24}
\]

for all \( n \geq 1 \). Furthermore, suppose that there exists a bound \( M > 0 \) such that \( |u_k(x)| \leq M \) over \( \Omega \) for all \( k \geq 0 \). Then

\[
\frac{1}{n+1} \sum_{k=0}^{n} (f(x) - \det(D^2u_k)(x)) \to 0 \tag{25}
\]

when \( n \to \infty \) for all \( x \in \Omega \).

We remark that the condition that \( |u_k(x)| \leq M \) above is a computational condition one can check during the iterative computation of Algorithm 1. Our numerical experiments show that for some testing functions \( u \), this condition does satisfy while for other testing functions, the condition does not satisfy. See Fig. 3 for these numerical phenomena.

**Proof** By (21) and Lemma 4, we get

\[
27\det(D^2u_{k+1})) \leq (\Delta u_{k+1})^3 = (\Delta u_k)^3 + 27(f - \det(D^2u_k))
\]

\[
= (\Delta u_{k-1})^3 + 27(f - \det(D^2u_{k-1})) + 27(f - \det(D^2u_{k-1}))
\]

\[
= (\Delta u_{k-1})^3 + 2 \cdot 27 f - 27 \det(D^2u_{k-1}) - 27 \det(D^2u_k)
\]

\[
= \cdots
\]

\[
= (\Delta u_0)^3 + 27(k + 1) f - 27 \sum_{j=0}^{k} \det(D^2u_j)
\]

\[
= 27 f + 27(k + 1) f - 27 \sum_{j=0}^{k} \det(D^2u_j).
\]
Hence, we have

\[ 0 \leq 27f + 27(k + 1)f - 27\sum_{j=0}^{k} \det(D^2u_j) - 27\det(D^2u_{k+1}) = 27\sum_{j=0}^{k+1} (f - \det(D^2u_j)). \]

which leads to (24). In addition, we also have

\[ (\Delta u_{k+1})^3 - (\Delta u_0)^3 = 27\sum_{j=0}^{k} (f - \det(D^2u_j)). \]

By the assumption of this theorem, \( u_{k+1} \) has a bound, i.e. \( \|u_{k+1}\|_{\infty, \Omega} \leq M \). Then we can use the Markov inequality to have

\[ \|\Delta u_{k+1}\|_{\infty, \Omega} \leq \frac{C}{|\Delta|^2} \|u_{k+1}\|_{\infty, \Omega} \leq \frac{CM}{|\Delta|^2} < \infty \]  \hspace{1cm} (26)

for a constant \( C > 0 \) independent of \( u_{k+1} \). It thus follows

\[ \frac{27\sum_{j=0}^{k} (f - \det(D^2u_j))}{k + 1} = \frac{(\Delta u_{k+1})^3 - (\Delta u_0)^3}{k + 1} \to 0. \]

Therefore, we finished a proof of Theorem 9. \( \square \)

Furthermore, we denote \( w(u, f) := \sqrt[3]{(\Delta u)^3 + 27(f - \det(D^2u))} \). We have

\[ \|\Delta u_{k+1} - \Delta u\|_{L^2(\Omega)} = \|\frac{(\Delta u_k)^3 + 27(f - \det(D^2u_k)) - (\Delta u)^3}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \|_{L^2(\Omega)} \]
\[ = \|\frac{(\Delta u_k)^3 - (\Delta u)^3 + 27(\det(D^2u) - \det(D^2u_k))}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \|_{L^2(\Omega)} \]

By simple calculations, we get

\[ (\Delta u_k)^3 - (\Delta u)^3 = (\Delta u_k - \Delta u)((\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2) \]
and by Lemmas 2.1, 2.2 and 2.3 in [5]

\[ \det(D^2u) - \det(D^2u_k) = \text{cof}((1 - t)D^2u_k + tD^2u) : (D^2u_k - D^2u) \]

for some \( t \in [0, 1] \). By simple calculation and Lemma 5, we have

\[
\begin{align*}
\| (\Delta u_k)^3 - (\Delta u)^3 \|_{L^2(\Omega)} &\leq \| (\Delta u_k - \Delta u)((\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2) \|_{L^2(\Omega)} \\
&\leq \| (\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2 \|_{L^2(\Omega)} \\
&\leq \| (\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2 \|_{L^2(\Omega)}.
\end{align*}
\]

Let \( M = (M_{ij}) \), \( N = (N_{ij}) \) be matrix fields and \( c \) be a real number. Then we get

\[
\begin{align*}
M : N &\leq \frac{1}{c^3} \sum_{i,j=1}^3 M_{ij} N_{ij} = \frac{1}{c^3} \sum_{i,j=1}^3 M_{ij} N_{ij} \\
&\leq \frac{M}{c} \| \infty \sum_{i,j=1}^3 |N_{ij}| \leq \frac{M}{c} \| \infty \left( \sum_{i,j=1}^3 N_{ij}^2 \right)^{1/2} \cdot 3, \quad (27)
\end{align*}
\]

where \( \| \frac{M}{c} \|_{\infty} = \max_{1 \leq i \leq 3} \sum_{j=1}^3 \frac{M_{ij}}{c} \). By (27) with, we have

\[
\begin{align*}
\| (w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2 \|_{L^2(\Omega)} &\leq 27 \| (\text{cof}((1 - t)D^2u_k + tD^2u) : (D^2u_k - D^2u)) \|_{L^2(\Omega)} \\
&= 27 \left[ \int \left( \left( \text{cof}((1 - t)D^2u_k + tD^2u) : (D^2u_k - D^2u) \right) \right)^2 \right]^{1/2} \\
&= 81 \| (1 - t)D^2u_k + tD^2u \|_{L^2(\Omega)} \\
&\leq 81 \| (1 - t)D^2u_k + tD^2u \|_{L^2(\Omega)}
\end{align*}
\]

for some \( t \in [0, 1] \). By these two equations, we can have

\[
\begin{align*}
\| \Delta u_{k+1} - \Delta u \|_{L^2(\Omega)} &\leq \frac{(\Delta u_k)^3 - (\Delta u)^3 + 27(\det(D^2u) - \det(D^2u_k))}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \|_{L^2(\Omega)} \\
&\leq \frac{(\Delta u_k - \Delta u)((\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2)}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \|_{L^2(\Omega)} \\
&+ \frac{27(\det(D^2u) - \det(D^2u_k))}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \|_{L^2(\Omega)} \\
&\leq \frac{(\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \|_{L^2(\Omega)} \\
&\leq \| (\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2 \|_{L^2(\Omega)} \| \Delta u_k \|_{L^2(\Omega)}
\end{align*}
\]
for some \( t \in [0, 1] \). Now, we need the following lemma from \([39]\) to prove one of the main convergence results in this paper.

**Lemma 5** Suppose that \( \Omega \) is bounded and has uniformly positive reach \( r_\Omega > 0 \). Then there exist two positive constants \( A \) and \( B \) such that

\[
A \| u \|_{H^2(\Omega)} \leq \| \Delta u \|_{L^2(\Omega)} \leq B \| u \|_{H^2(\Omega)}, \quad \forall u \in H^2(\Omega) \cap H^1_0(\Omega).
\]  

By Lemma 5, we have

\[
\| \Delta u_{k+1} - \Delta u \|_{L^2(\Omega)} \leq \left\| \frac{(\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \right\|_{L^\infty(\Omega)} \| \Delta u_k \\
- \left(1-t\right)D^2 u_k + tD^2 u \right\|_{L^\infty(\Omega)} + 81 \left(1-t\right)D^2 u_k + tD^2 u \right\|_{L^\infty(\Omega)} \right\|_{L^\infty(\Omega)} + \frac{1}{A} \| \Delta u_k \\
- \| \Delta u \|_{L^2(\Omega)}
\]

and therefore

\[
\| \Delta u_{k+1} - \Delta u \|_{L^2(\Omega)} \leq \rho_k \| \Delta u_k - \Delta u \|_{L^2(\Omega)},
\]

where

\[
\rho_k := \left\| \frac{(\Delta u_k)^2 + \Delta u_k \cdot \Delta u + (\Delta u)^2}{(w(u_k, f))^2 + w(u_k, f)w(u, f) + (w(u, f))^2} \right\|_{L^\infty(\Omega)} + \frac{81}{A} \left(1-t\right)D^2 u_k + tD^2 u \right\|_{L^\infty(\Omega)} \right\|_{L^\infty(\Omega)} + \frac{1}{A} \| \Delta u_k \\
- \left(1-t\right)D^2 u_k + tD^2 u \right\|_{L^\infty(\Omega)} + 81 \left(1-t\right)D^2 u_k + tD^2 u \right\|_{L^\infty(\Omega)} \right\|_{L^\infty(\Omega)} + \frac{1}{A} \| \Delta u_k \\
- \| \Delta u \|_{L^2(\Omega)}
\]

We are now ready to conclude the following result

**Theorem 10** Suppose that \( \Omega \) is bounded and has uniformly positive reach. If \( \rho_k \leq \gamma < 1 \) for all \( k \geq 1 \), then the sequence \( \{ u_k \} \) from Algorithm 1 converges.

Note that our numerical experiments show that for some testing function \( u \), we have indeed \( \rho_k < 1 \) while there is other testing function \( u \) which gives \( \rho_k > 1 \). See Figs. 3 and 4. Also, it is hard to estimate \( \rho_k \) from the formula (29).

In Figs. 3 and 4, we plot \( \rho_k \) corresponding to the numerical solution for smooth solutions \( s_1, s_2, s_3, s_4 \) and non-smooth solutions \( ns_1, ns_2 \). They are defined as follows.

- \( s_1 \): polynomial function \((x^2 + 5y^2 + 15z^2)/2\);
- \( s_2 \): exponential function \( \exp((x^2 + y^2 + z^2)/2) \);
- \( s_3 \): radical function \(-\sqrt{6 - (x^2 + y^2 + z^2)} \);
- \( s_4 \): \((x^2 + y^2 + z^2)/2 - \sin(x) - \sin(y) - \sin(z) \);
- \( ns_1 \): \(-\sqrt{3 - (x^2 + y^2 + z^2)} \) where \( f \) is \( \infty \) at \((1, 1, 1) \);
- \( ns_2 \): \((x^2 + y^2 + z^2)^{3/4} \) where \( f \) is \( \infty \) at \((0, 0, 0) \).

The graphs in these figure above show that \( \rho_k < 1 \) and \( \| \Delta u_k \|_{\infty} \) are bounded for smooth testing solutions. However, \( \rho_k \) may be bigger than 1 and \( \| \Delta u_k \|_{\infty} \) may increase which maybe unbounded for nonsmooth testing functions.
When $\rho_k > 1$, the above analysis will not be useful to see convergence of the sequence $\{u_k\}$. The remaining case is $\rho_k \leq 1$. In this case, we need Algorithm 2. That is, we now study the convergence of our Algorithm 2. Letting $u^k$, $k \geq 1$ be the sequence from Algorithm 2, it is easy to see that

$$u_{k+1} - u^* = \frac{1}{2}(u_k - u^*) + \frac{1}{2}(T(u_k) - T(u^*))$$

(30)

for all $k \geq 1$ since $u^*$ is a fixed point of $T$. Since $\rho_k \leq 1$, we have $\|T(u_k) - T(u^*)\| \leq \|u_k - u^*\|$ and hence, $\|u_{k+1} - u^*\| \leq \|u_k - u^*\|$ for all $k \geq 1$. It follows that $u_k$, $k \geq 1$ are bounded in a $H^2(\Omega)$ norm. We now show the averaged iterative algorithm converges.

**Theorem 11** Suppose that $\Omega$ is a bounded domain which has a uniformly positive reach. Suppose that $g \in H^{1/2}(\partial \Omega)$. Suppose that $\rho_k \leq 1$. Then Averaged Iterative Algorithm 2 converges.

**Proof** Let $S = H^2(\Omega)$. By the assumptions, the operator $T$ defined above from $S$ to $S$ is a continuous and nonexpansive operator. We first recall the following equality: For any $x, y, z \in S$ and a real number $\lambda \in [0, 1]$, we have the following identity

$$\lambda \|x - z\|^2 + (1 - \lambda)\|y - z\|^2 - \lambda(1 - \lambda)\|x - y\|^2 = \|\lambda x + (1 - \lambda)y - z\|^2.$$

The proof is left to the interested reader. Let $\lambda = 1/2$ and $x = u^k$, $y = T(u^k)$, and $z = u^*$ which is a fixed point or the solution. Then we have

$$\|u_{k+1} - u^*\|^2 = \frac{1}{2}\|u_k - u^*\|^2 + \frac{1}{2}\|T(u_k) - u^*\|^2 - \frac{1}{4}\|u_k - T(u_k)\|^2$$

$$= \frac{1}{2}\|u_k - u^*\|^2 + \frac{1}{2}\|T(u_k) - T(u^*)\|^2 - \frac{1}{4}\|u_k - T(u_k)\|^2$$

$$\leq \|u_k - u^*\|^2 - \frac{1}{4}\|u_k - T(u_k)\|^2.$$

It follows that

$$\sum_{k=1}^{N} \frac{1}{4}\|u_k - T(u^k)\|^2 + \|u_{N+1} - u^*\|^2 \leq \|u_0 - u^*\|^2$$
for any integer $N > 1$. That is, $\|u_k - T(u_k)\| \to 0$ when $k \to \infty$.

We now claim that the sequence $u_k$, $k \geq 1$ converges. Note that due to the nonexpansiveness, $\|u_k\|, k \geq 1$ are bounded as explained above. Let $\hat{u}$ be the limit of a subsequence of $u^k$, $k \geq 1$. Then we have $\hat{u} = T(\hat{u})$ by the continuity of the operator $T$. So $\hat{u}$ is a fixed point of $T$. By the definition of $T$, we have

$$\Delta \hat{u} = \sqrt[3]{(\Delta \hat{u})^3 + 27(f - \det(D^2 \hat{u}))}$$

or $(\Delta \hat{u})^3 = (\Delta \hat{u})^3 + 27(f - \det(D^2 \hat{u}))$. It follows that $f = \det(D^2 \hat{u})$. Since the Monge-Ampère equation has a unique solution, we have $\hat{u} = u^*$. If there exists another $\bar{u}$ which is the limit of another subsequence of $u^k$, $k \geq 1$, we also have $\bar{u} = T(\bar{u})$. Then $\bar{u} = u^*$. Hence, the sequence $\{u_k, k \geq 1\}$ from Algorithm 2 converges. □

4 Numerical Results for 3D Monge-Ampère Equations

In this section, we present numerical results from various computational experiments. We will first test several smooth and nonsmooth solutions over convex domains, such as [0, 1]$^3$. Next, we show the numerical results over non-convex domains such as $C$, $L$, $S$-shaped domains. For all the experiments, we use 8 processors on a parallel computer, which has AMD Ryzen 7 4800H with Radeon Graphics 2.90 GHz. All the cases, the errors are computed based on $NI = 351 \times 351 \times 351$ equally spaced points $\{\eta_i\}_{i=1}^{NI}$ fell inside the domain of computation. The errors will be calculated according to the norms

$$\begin{align*}
|u|_{l_2} &= \sqrt{\frac{\sum_{i=1}^{NI} (u(i))^2}{NI}} \\
|u|_{h_1} &= \sqrt{\frac{\sum_{i=1}^{NI} (u(i))^2 + (u_x(i))^2 + (u_y(i))^2 + (u_z(i))^2}{NI}} \\
|u|_{\infty} &= \max |u(i)|,
\end{align*}$$

where $u(i) := u(\eta_i), u_x(i) := u_x(\eta_i), u_y(i) := u_y(\eta_i)$ and $u_z(i) := u_z(\eta_i)$ for given functions $u, u_x, u_y, u_z$. Tables in this section are the numerical results of $|e_s|_{l_2}$ and $|e_s|_{h_1}$, where $e_s := u - u_s$.

4.1 Smooth Testing Functions

Example 1 (Polynomial Examples) In [17], the researchers experimented the following two smooth exact solutions:

- $f^{3d1} = 75$ such that an exact solution is $u^{3d1} = \frac{1}{2}(x^2 + 5y^2 + 15z^2)$
- $f^{3d2} = 1000$ such that an exact solution is $u^{3d2} = \frac{1}{2}(x^2 + 10y^2 + 100z^2)$

Numerical results of their least squares/relaxation method (called LR method in this paper) are shown in Table 2. Together we present numerical results based on our spline collocation method by Algorithms 2. Table 2 shows our spline collocation method (called LL method) produces more accurate solutions than those presented in [17]. The eigenvalues of the Hessian matrix are 1, 5, 15 and therefore $\det(D^2u^{3d1}) = \lambda_1 \lambda_2 \lambda_3 = 75$ and $\Delta u^{3d1} = \lambda_1 + \lambda_2 + \lambda_3 = 21$. In Algorithm 2, we choose an initial value $\Delta u_0 = 14.55$ to approximate the exact solution $u^{3d1}$. This choice of initial value leads to converging iterations since 14.55 is close to $\Delta u^{3d1} = \sqrt[3]{27f} = \sqrt[3]{27 \cdot 75} = 12.65$. Similarly, we choose our initial value $u_0$ for $u^{3d2}$ satisfying $\Delta u_0 = 106.2$ which makes the iterations from Algorithm 2 converge. By choosing good initial value $u_0$ we achieve the numerical results shown in Table 2.
We also test other smooth solutions which were experimented in the literature, e.g., [4, 6, 13, 17], and etc.

**Example 2** (Smooth Exponential Functions) Consider a smooth exponential exact solution $u^{3ds_3} = e^{(x^2+y^2+z^2)^{3/2}}$ associated with $f^{3ds_3} = (1 + x^2 + y^2 + z^2)e^{(x^2+y^2+z^2)^{3/2}}$. We compare our methods with the least squares/relaxation method (LR method) in [17]. Table 3 shows comparison results including $l_2$, $h_1$ norm of these two methods for each mesh size $h$.

We can see that better convergence results using LL methods with $D = 5$, $r = 1$.

**Example 3** In [4] and [6], Awanou introduced the pseudo transient continuation, time marching methods and the spline element methods for Monge-Ampére equations. He presented several 2D and 3D numerical examples by his methods. For testing function $u^{3ds_4} = e^{(x^2+y^2+z^2)}$, it seems that the numerical results using the spline element method (SE method) in [6] is the best. We use our spline collocation method (LL method) and compare $L^2$, $H^1$, $H^2$ errors of our method and the SE method. Table 4 and 5 show that we can get better accuracy when $h = 1$ and $h = 1/2$ for each degree $D = 4, 5, 6$. 

| LR method | $u^{3ds_1}$ | $u^{3ds_2}$ |
|------------|-------------|-------------|
| $h$        | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ |
| 0.2        | 7.19e−02    | 1.58e−00    | 2.74e−02    | 5.16e−01    |
| 0.1        | 1.80e−02    | 7.91e−01    | 7.52e−03    | 2.81e−01    |
| 0.0625     | 7.06e−03    | 4.95e−01    | 3.06e−03    | 1.83e−01    |
| 0.04       | 2.89e−03    | 3.16e−01    | 1.26e−03    | 1.20e−01    |

| LL method | $u^{3ds_1}$ | $u^{3ds_2}$ |
|------------|-------------|-------------|
| $h$        | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ |
| 0.25       | 2.68e−07    | 1.01e−05    | 2.48e−04    | 4.63e−03    |

| LR method | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|------------|--------------|------|--------------|------|
| $h$        |              |      |              |      |
| 0.2        | 7.19e−02     | −    | 1.58e−00     | −    |
| 0.1        | 1.80e−02     | 1.99 | 7.91e−01     | 0.99 |
| 0.0625     | 7.06e−03     | 1.99 | 4.95e−01     | 1.00 |
| 0.04       | 2.89e−03     | 1.99 | 3.16e−01     | 0.99 |

| LL method | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|------------|--------------|------|--------------|------|
| $h$        |              |      |              |      |
| 1          | 1.17e−03     | −    | 1.24e−02     | −    |
| 0.5        | 4.36e−05     | 4.74 | 7.82e−04     | 3.99 |
| 0.25       | 1.42e−06     | 4.94 | 2.72e−05     | 4.84 |
| 0.125      | 1.10e−07     | 3.69 | 1.36e−06     | 4.32 |

| Table 2 | Errors of numerical solutions $u^{3ds_1}$, $u^{3ds_2}$ for Monge Ampère equation over $[0, 1]^3$ for LL methods with $D = 5$, $r = 1$ and LR method in [17] |
|---------|---------------------------------|
| LR method | $u^{3ds_1}$ | $u^{3ds_2}$ |
| $h$        | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ |
| 0.2        | 7.19e−02    | 1.58e−00    | 2.74e−02    | 5.16e−01    |
| 0.1        | 1.80e−02    | 7.91e−01    | 7.52e−03    | 2.81e−01    |
| 0.0625     | 7.06e−03    | 4.95e−01    | 3.06e−03    | 1.83e−01    |
| 0.04       | 2.89e−03    | 3.16e−01    | 1.26e−03    | 1.20e−01    |

| Table 3 | Errors of numerical solutions $u^{3ds_3}$ for Monge Ampère equation over $[0, 1]^3$ for LL methods with $D = 5$, $r = 1$ and LR method in [17] |
|---------|---------------------------------|
| LR method | $u^{3ds_1}$ | $u^{3ds_2}$ |
| $h$        | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ | $|e_s|_{l_2}$ | $|e_s|_{h_1}$ |
| 0.2        | 7.19e−02    | 1.58e−00    | 2.74e−02    | 5.16e−01    |
| 0.1        | 1.80e−02    | 7.91e−01    | 7.52e−03    | 2.81e−01    |
| 0.0625     | 7.06e−03    | 4.95e−01    | 3.06e−03    | 1.83e−01    |
| 0.04       | 2.89e−03    | 3.16e−01    | 1.26e−03    | 1.20e−01    |

| LL method | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|------------|--------------|------|--------------|------|
| $h$        |              |      |              |      |
| 0.25       | 2.68e−07    | 1.01e−05 | 2.48e−04    | 4.63e−03 |

| LR method | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|------------|--------------|------|--------------|------|
| $h$        |              |      |              |      |
| 0.2        | 7.19e−02     | −    | 1.58e−00     | −    |
| 0.1        | 1.80e−02     | 1.99 | 7.91e−01     | 0.99 |
| 0.0625     | 7.06e−03     | 1.99 | 4.95e−01     | 1.00 |
| 0.04       | 2.89e−03     | 1.99 | 3.16e−01     | 0.99 |

| LL method | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|------------|--------------|------|--------------|------|
| $h$        |              |      |              |      |
| 1          | 1.17e−03     | −    | 1.24e−02     | −    |
| 0.5        | 4.36e−05     | 4.74 | 7.82e−04     | 3.99 |
| 0.25       | 1.42e−06     | 4.94 | 2.72e−05     | 4.84 |
| 0.125      | 1.10e−07     | 3.69 | 1.36e−06     | 4.32 |
Fig. 5 Convergence rates of $l_2$, $h_1$ errors for solutions $u^{3ds5}$ (Left) and $u^{3ds5}$ (Right) with respect to $|\Delta|$ based on the LL method.

Table 4 Errors of numerical solutions $u^{3ds4}$ for Monge Ampère equation over $[0, 1]^3$ for LL methods with $D = 3, 4, 5, 6, r = 1, h = 1$ and SE method in [6]

| $D$ | SE method $L^2$ norm | $H^1$ norm | $H^2$ norm | LL method $L^2$ norm | $H^1$ norm | $H^2$ norm |
|-----|-----------------------|-------------|-------------|-----------------------|-------------|-------------|
| 3   | 1.2338e−02            | 7.6984e−02  | 4.4411e−01  | 1.6916e−02           | 1.0879e−01  | 3.8250e−01  |
| 4   | 1.6289e−03            | 1.4719e−02  | 1.3983e−01  | 6.4696e−04           | 6.1874e−03  | 3.7146e−02  |
| 5   | 1.5333e−03            | 8.7312e−03  | 6.0412e−02  | 1.7440e−04           | 2.2203e−03  | 1.7392e−02  |
| 6   | 1.2324e−04            | 9.7171e−04  | 1.0584e−02  | 4.6740e−05           | 6.2257e−04  | 3.5432e−03  |

Table 5 Errors of numerical solutions $u^{3ds4}$ for Monge Ampère equation over $[0, 1]^3$ for LL methods with $D = 3, 4, 5, 6, r = 1, h = 1/2$ and SE method in [6]

| $D$ | SE method $L^2$ norm | $H^1$ norm | $H^2$ norm | LL method $L^2$ norm | $H^1$ norm | $H^2$ norm |
|-----|-----------------------|-------------|-------------|-----------------------|-------------|-------------|
| 3   | 3.1739e−03            | 2.3005e−02  | 2.4496e−01  | 2.4294e−03           | 1.5806e−02  | 1.0139e−01  |
| 4   | 3.2786e−04            | 3.5626e−03  | 5.2079e−02  | 9.5591e−05           | 1.1644e−03  | 9.8077e−03  |
| 5   | 2.4027e−05            | 3.9210e−04  | 8.8868e−03  | 5.8750e−06           | 1.2214e−04  | 1.4292e−03  |
| 6   | 1.3821e−06            | 2.2369e−05  | 6.0918e−04  | 6.0635e−07           | 1.4198e−05  | 1.6487e−04  |

4.2 Non-smooth Testing Functions

Example 4 In [17], the researchers considered the following problem which do not have exact solution with the $H^2(\Omega)$ regularity or may have no solution at all. For $R \geq \sqrt{3}$, let $u = -\sqrt{R^2 - (x^2 + y^2 + z^2)}$ be a testing function. When $R > \sqrt{3}$, this function belongs to $C^\infty(\Omega)$, while $u \in C^0(\Omega) \cap W^{1,s}(\Omega)$, with $1 \leq s < 2$, if $R = \sqrt{3}$. More precisely, let us consider the following two solutions

$$u^{3ds5} = -\sqrt{6 - (x^2 + y^2 + z^2)} \text{ with } f^{3d5} = 6(6 - (x^2 + y^2 + z^2))^{-\frac{5}{7}}$$

and

$$u^{3ds6} = -\sqrt{3 - (x^2 + y^2 + z^2)} \text{ with } f^{3d6} = 3(3 - (x^2 + y^2 + z^2))^{-\frac{5}{2}}.$$
The numerical results from the least squares/relaxation method in [17] (called LR method) are shown in Table 6. In Fig. 5, we can see that the rate of convergences of $u^{3ds5}$ are about $O(h^{4.82})$. In addition, we show our spline collocation method (called LL method) in the same table for comparison.

It is clear to see that when the solution $u^{3ds5}$ is smooth, both methods work nicely and our collocation method is much accurate.

Next let us consider the non-smooth solution $u^{3ds6}$ in Table 7. Table 7 shows numerical results such as $l_2$, $h_1$ errors of these two methods for various mesh sizes. Our method can get more accurate solution with $D = 5$, $r = 1$ with the large mesh size $|\Delta|$. However, it is clear that an adaptive method is needed to improve the approximation since the maximal error, $e_s = u - u_s$, is worst near the point $(1, 1, 1)$.

### Table 6

Errors of numerical approximation of the solution $u^{3ds5}$ for Monge Ampère equation over $[0, 1]^3$ by the and LR method and by the LL method with $D = 5$, $r = 1$

| $|\Delta|$ | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|---|---|---|---|---|
| 0.2 | 4.96e−03 | – | 8.60e−02 | – |
| 0.1 | 1.28e−03 | 1.95 | 4.41e−02 | 0.96 |
| 0.0625 | 5.09e−04 | 1.96 | 2.78e−02 | 0.97 |
| 0.04 | 2.10e−04 | 1.97 | 1.79e−02 | 0.98 |

| $|\Delta|$ | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|---|---|---|---|---|
| 1 | 3.75e−05 | – | 3.88e−04 | – |
| 0.5 | 1.10e−06 | 5.09 | 1.73e−05 | 4.49 |
| 0.25 | 5.05e−08 | 4.45 | 6.73e−07 | 4.69 |
| 0.125 | 3.52e−09 | 3.84 | 3.06e−08 | 4.46 |

### Table 7

Errors of numerical approximation of the solution $u^{3ds6}$ for Monge Ampère equation over $[0, 1]^3$ by the and LR method and by the LL method with $D = 5$, $r = 1$

| $|\Delta|$ | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|---|---|---|---|---|
| 0.2 | 1.15e−02 | – | 6.60e−01 | – |
| 0.1 | 3.06e−03 | 1.91 | 6.31e−01 | – |
| 0.0625 | 1.24e−03 | 1.92 | 6.25e−01 | – |
| 0.04 | 5.17e−04 | 1.96 | 6.22e−01 | – |

| $|\Delta|$ | $|e_s|_{l_2}$ | rate | $|e_s|_{h_1}$ | rate |
|---|---|---|---|---|
| 1 | 8.07e−02 | – | 7.02e−01 | – |
| 0.5 | 7.06e−03 | 3.52 | 1.63e−01 | 2.10 |
| 0.25 | 4.78e−04 | 3.88 | 2.21e−02 | 2.89 |
| 0.125 | 3.85e−04 | 0.31 | 2.54e−02 | −0.20 |
| 0.0625 | 3.57e−04 | 0.11 | 1.98e−02 | 0.35 |
Fig. 6  Several 3D domains (Top: Cube, Letter L, Letter C, Bottom: Letter S, Subset of the unit ball)

Table 8  CPU time results and numbers of vertices and tetrahedrons over domains in Fig. 6 when $D = 9$, $r = 1$

| Domain   | No. of Vertices | No. of Tetrahedrons | Total CPU(s) |
|----------|-----------------|---------------------|--------------|
| Cube     | 125             | 384                 | 56.0         |
| Letter L | 105             | 288                 | 42.7         |
| Letter C | 190             | 431                 | 129.2        |
| Letter S | 115             | 171                 | 45.9         |

4.3 Numerical Results over Nonconvex Domains

In this section, we test various solutions for each domain in Fig. 6. We display CPU times versus the number of vertices and triangles in Table 8 for each domain in Fig. 6 when $D = 9$, $r = 1$.

Example 5  We use our method to numerically solve three smooth testing functions $u^{3ds3}$, $u^{3ds4}$, $u^{3ds5}$ over 5 solids which are not strictly convex or not convex. They even do not have an uniformly positive reach. Table 9 shows our method performs very well.

Example 6  In [17], the researchers considered the problem over the unit ball $\Omega = \{(x, y, z)|x^2 + y^2 + z^2 < 1\}$ and a convex solution

$$u^{3ds7} = -\frac{1}{2\sqrt{3}}(1 - x^2 - y^2 - z^2)$$

of the Monge-Ampére-Dirichlet problem with $f = \frac{1}{3\sqrt{3}}$. They experimented their numerical solutions (called LR method) over the unit ball as well as the 3/4 ball as shown in Fig. 6.
Table 9 Errors of numerical solutions \( u^{3ds3} - u^{3ds5} \) for Monge Ampère equations over several domains in Fig. 6 for LL methods with \( D = 9, r = 1 \)

| Solution | Cube | Letter L | Letter C | Letter S |
|----------|------|----------|----------|----------|
| \( u^{3ds3} \) | \(|e_s|_{l_2}\) | \(|e_s|_{h_1}\) | \(|e_s|_{l_2}\) | \(|e_s|_{h_1}\) | \(|e_s|_{l_2}\) | \(|e_s|_{h_1}\) |
| 1.76e−09 | 1.64e−07 | 6.63e−10 | 2.58e−08 | 1.48e−08 | 8.67e−07 | 3.39e−11 | 1.75e−09 |
| \( u^{3ds4} \) | 2.82e−11 | 1.91e−10 | 3.90e−11 | 1.04e−09 | 2.31e−08 | 3.56e−07 | 5.84e−11 | 2.70e−09 |
| \( u^{3ds5} \) | 5.05e−08 | 3.61e−07 | 2.47e−08 | 9.56e−07 | 3.87e−08 | 3.32e−06 | 6.03e−10 | 5.03e−08 |

Table 10 Errors of numerical approximation of solution \( u^{3ds7} \) for Monge Ampère equation over the unit ball for the LR method and the LL method with \( D = 5, r = 1 \)

| LR method | | | |
|-----------|-----|-----|-----|
| \(|\Delta|\) | \(|e_s|_{l_2}\) | rate | \(|e_s|_{h_1}\) | rate |
| 2.98e−01 | 3.26e−02 | – | 2.60e−01 | – |
| 1.61e−01 | 1.11e−02 | 1.74 | 1.28e−01 | 1.14 |
| 8.32e−02 | 3.22e−03 | 1.88 | 6.16e−02 | 1.11 |
| 4.34e−02 | 9.89e−04 | 1.80 | 2.86e−02 | 1.17 |

| LL method | | | |
|-----------|-----|-----|-----|
| \(|\Delta|\) | \(|e_s|_{l_2}\) | rate | \(|e_s|_{h_1}\) | rate |
| 1 | 3.71e−13 | – | 3.15e−12 | – |
| 0.5 | 2.97e−14 | 3.64 | 1.39e−13 | 4.51 |

Table 11 CPU time and errors of our spline solution to \( u^{3ds7} \) for Monge Ampère equation over the domain in Fig. 6 with \( D = 5, r = 1 \), the number of vertices=585, the number of tetrahedrons=2304

| LL method | CPU time | \(|e_s|_{l_2}\) | \(|e_s|_{h_1}\) |
|-----------|----------|-----|-----|
| 174.70 | 1.9005e−08 | 2.4941e−06 |

In Table 10, we first include the numerical results from [17] and then compare the \( L^2(\Omega), H^1(\Omega) \) norms of the computed approximation error \( u^{3ds7} - u_s \) by our spline collocation method. In addition, we tested the solution \( u^{3ds7} \) over the subset of the unit ball as shown in Fig. 6. The numerical results we obtained are displayed in Table 11.

4.4 Comparison with Numerical Method in [13]

In this section, we compare our LL method with the Cascadic method in [13]. The researchers presented several examples in [13] over the irregular domains in Fig. 7 by using the following test functions

\[
\begin{align*}
 u^{3ds3} &= e^{\frac{x^2+y^2+z^2}{2}}, \\
 u^{3ds6} &= -\sqrt{3-(x^2+y^2+z^2)}, \\
 u^{3ds8} &= \frac{x^2+y^2+z^2}{2} - \sin(x) - \sin(y) - \sin(z),
\end{align*}
\]
Fig. 7 Several 3D domains (Cube, Letter L, Torus)

Table 12 The CPU time, DOFs, errors $|e_1|_{L^2}$, $|e_1|_{H^1}$ using LL method with $D = 6$, $r = 1$ and $|e_1|_{L^2}$ using the Cascadic method in [13] over the cube $[0, 1]^3$

| solution | LL method | Cascadic method |
|----------|------------|-----------------|
|          | CPU time   | DOFs | $|e_1|_{L^2}$  | $|e_1|_{H^1}$  | $|e_1|_{L^2}$  |
| $u^{3ds3}$ | 18.712     | 32256 | 4.4010e−08  | 1.6144e−06  | 9.8659e−04  |
| $u^{3ds6}$ | 6.9873     | 32256 | 4.0285e−04  | 2.6783e−02  | 1.7831e−04  |
| $u^{3ds8}$ | 14.852     | 32256 | 1.8152e−11  | 6.8884e−10  | 3.5044e−04  |
| $u^{3ds9}$ | 12.826     | 32256 | 1.3242e−04  | 1.1264e−03  | 2.2255e−04  |

Table 13 The CPU time, DOFs, errors $|e_1|_{L^2}$, $|e_1|_{H^1}$ using LL method with $D = 6$, $r = 1$ and $|e_1|_{L^2}$ using the Cascadic method in [13] over L-shaped domain

| solution | LL method | Cascadic method |
|----------|------------|-----------------|
|          | CPU time   | DOFs | $|e_1|_{L^2}$  | $|e_1|_{H^1}$  | $|e_1|_{L^2}$  |
| $u^{3ds3}$ | 12.826     | 24192 | 1.5727e−07  | 7.0173e−07  | 4.8655e−03  |
| $u^{3ds6}$ | 4.5050     | 24192 | 6.5992e−04  | 7.6198e−04  | 1.6238e−04  |
| $u^{3ds8}$ | 12.935     | 24192 | 1.4826e−11  | 6.9814e−10  | 1.2240e−04  |
| $u^{3ds9}$ | 4.8951     | 24192 | 2.3076e−04  | 2.0929e−03  | 4.2183e−04  |

We use our method (LL method) to compute numerical solutions based on the same testing functions over the same testing domains. Our numerical results are shown in Tables 12, 13 and 14.

4.5 Free Movement of Transportation

Finally, we consider the free movement problem. In this case, the Monge-Ampère Eq. (4) becomes

$$\det(D^2 u(x)) = 1, \quad x \text{ in } \Omega \subseteq \mathbb{R}^3 \quad (31)$$

$$\nabla u(x) = \partial W, \quad x \text{ on } \partial \Omega. \quad (32)$$
Table 14 The CPU time, DOFs, errors $|\varepsilon_h|_2$, $|\varepsilon_h|_{L^1}$ using LL method with $D = 4$, $r = 1$ and $|\varepsilon_h|_2$ using the Cascadic method in [13] over Torus

| solution     | LL method CPU time | DOFs  | $|\varepsilon_h|_2$ | $|\varepsilon_h|_{L^1}$ | Cascadic method $|\varepsilon_h|_2$ |
|--------------|-------------------|-------|--------------------|-----------------------|-------------------------------|
| $u_{3d3}^3$  | 141.01            | 80990 | 6.0125e−07         | 1.3250e−05            | 3.1914e−04                   |
| $u_{3d6}^3$  | 90.420            | 80990 | 5.6340e−04         | 1.1438e−02            | 1.9850e−04                   |
| $u_{3d8}^3$  | 119.42            | 80990 | 4.9540e−07         | 9.5117e−06            | 2.1182e−04                   |
| $u_{3d9}^3$  | 125.64            | 80990 | 2.3272e−07         | 1.3522e−05            | 1.7504e−04                   |

To completely determine $u$, we need to specify an oblique boundary condition. When both $\Omega$ and $W$ are star-shaped domains, e.g., convex domains, we can match the centers of $\Omega$ and $W$ by shifts and use a ray $R$ from the center to intercept the boundary of $\Omega$ and the boundary of $W$. This can build up a map from $\partial \Omega$ to the boundary of $W$. Using the outward normal of $\partial \Omega$ and the outward normal of $W$, we solve the Neumann boundary condition of the Monge-Ampére Eq. (4) becomes

$$\det(D^2 u(x)) = 1, \quad x \in \Omega \subset \mathbb{R}^3 \quad (33)$$

$$n_{\partial \Omega} \nabla u(x) = n_W \partial W, \quad x \text{ on } \partial \Omega. \quad (34)$$

Now we can apply our computational approach to find a solution of $u$ and form a transportation map from $\Omega$ to $W$. So that we can show which points in $\Omega$ is mapped to which points in $W$ by using an image or stack of images.

**Example 7** For simplicity, we first consider the 2D setting over a rectangular domain $\Omega$ and $W$ is a circular domain as shown in Fig. 8. On the left-hand side, we have an image density function and on the right-hand side, the density is transported to the circular domain. The center of the rectangular domain $\Omega = [-1, 1]^2$ is the origin $(0, 0)$ and the same for the circular domain $W = \{(x, y), x^2 + y^2 \leq 1\}$. In Fig. 9 and 10, we use the point $(-0.4, 0)$ and $(0, 0.4)$ as the center of the circular domain, respectively. The image on the right-hand side of Fig. 9 is clearly deformed and the similar for the image (right) of Fig. 10. The cost for transportation in these two cases

$$\frac{1}{2} \int_{\Omega} \|x - \nabla u(x)\|^2 f(x) dx \quad (35)$$

is larger than the cost for the transportation in Fig. 8.

**Example 8** We now show the transportation from the points in the cube $\Omega = [-1, 1]^3$ to the unit ball $W = \{(x, y, z) : x^2 + y^2 + z^2 \leq 1\}$. Again we use the density $f(x)$ which is a stack of the same image over $\Omega$ to show how a point in $\Omega$ is transported to the point in $W$.

We can see that our computation is reliable as the points in $\Omega$ are completely transported into $W$. In fact the map $\nabla u$ is a bijection due to the convexity of the Brenier potential $u$ as shown in Fig. 12.

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Fig. 8 A density over a rectangular domain $\Omega = [-1, 1]^2$ (on the left) and a transported image over the circular domain (on the right).

Fig. 9 A density over a rectangular domain $\Omega = [-1, 1]^2$ (on the left) and a transported image over the circular domain (on the right). Note the point $(-0.4, 0)$ in the circular domain was chosen as the center.

Fig. 10 A density over a rectangular domain $\Omega = [-1, 1]^2$ (on the left) and a transported image over the circular domain (on the right). Note the point $(0, 0.4)$ in the circular domain was chosen as the center.
Fig. 11 A density over a rectangular domain $\Omega = [-1, 1]^3$ (on the left) and a transported image over the spherical domain $W$ (on the right).

Fig. 12 An iso-surface plot of the Brenier potential over a rectangular domain $\Omega = [-1, 1]^3$.

Data Availability The datasets generated during and/or analyzed during the current study are available from the corresponding author upon request.

Declarations

Conflict of interest The authors declare that they have no competing interests.

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