A Semi-Lagrangian Scheme for the Game p-Laplacian via p-averaging

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Abstract

We present and analyze an approximation scheme for the two-dimensional game p-Laplacian in the framework of viscosity solutions. The approximation is based on a semi-Lagrangian scheme which exploits the idea of p-averages. We study the properties of the scheme and prove that it converges, in particular cases, to the viscosity solution of the game p-Laplacian. We encoded and ran a numerical implementation of the scheme for various values of p; the numerical tests show that the scheme is accurate.

Keywords: p-Laplacian, Tug of war game, Hamilton-Jacobi equations, semi-Lagrangian scheme, convergence, viscosity solutions.

1. Introduction

The game p-Laplace operator has been recently introduced in [24] to model a stochastic game called tug-of-war with noise. Part of the interest for this class of operators arises from the fact that it includes, as particular

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cases, the operator in the Aronsson equation, the infinity Laplacian [23], the motion by mean curvature operator [17], and for p = 2, a multiple of the ordinary Laplacian.

The equation associated to the game p-Laplacian has the same solutions as the variational p-Laplacian only in the homogeneous case, and has the advantage in the non-homogeneous case of being a combination of other p-Laplacians. In particular, we stress the fact that even for the non-homogeneous case the game ∞ -Laplacian is the limit as $p \to \infty$ of the game p-Laplacian.

Our work strongly relies on the general philosophy expressed in the paper of Peres and Sheffield [24], which indicates that to study p-harmonic functions one can look at discrete versions of the p-operators, obtained via interpolation between the limiting cases of the motion-by-curvature operator and the Aronsson equation. It is important to note that our work is in the framework of weak solutions in the viscosity sense (see [9] for an introduction and [7] for a guide to viscosity solutions for second order problems). However, the difference between their analytic case and our approximation scheme is in the fact that we need the value at a fixed point to depend only on a discrete number of values in space. In this respect, our construction starts from an approximation of p-averages in order to keep a strong link between the continuous and the discrete operator. Specifically, one of our goals is to prove that our scheme is consistent with the continuous operator proposed by Peres and Sheffield.

Semi-Lagrangian schemes for nonlinear Hamilton-Jacobi equations have been studied and analyzed by several authors. In particular, the starting point is the discrete version of the characteristic method which leads to a discrete Lax-Hopf formula for first order Hamilton-Jacobi equation [11]. More interestingly semi-Lagrangian schemes allow for large time steps still satisfying stability conditions. A comprehensive introduction to semi-Lagrangian methods for linear and first order Hamilton-Jacobi equations is contained in the book by Falcone and Ferretti [12]. For second order problems some results have been obtained for stationary and evolutive equations related to stochastic control problems in [5], whereas a presentation of the treatment of second order terms in SL schemes have been presented in [14]. Recent extensions to mean curvature driven flows have been analyzed in [6].

In the homogeneous case the game p-Laplacian coincides with the variational p-Laplacian for which several approximation methods have been proposed. Some of these schemes are based on finite elements and show convergence, also establishing a priori error estimates, see e.g. the paper by

Barrett and Liu [3]. However finite elements are not the most popular techniques for nonlinear degenerate equation. Finite difference approximation schemes for degenerate second order equations have been proposed and analyzed by Crandall and Lions in [8], and specifically for the infinity Laplacian by Oberman [20]. More recently, finite volumes schemes have been presented by Andreyanov, Boyer and Hubert in [1], for the variational p-Laplacian.

In the non-homogeneous case the game p-Laplacian interprets the non-homogeneity, f, as a multiple of a running payoff for one of the players in a two-player, zero-sum game while for the variational p-Laplacian the non-homogeneity is interpreted as a potential. The variational p-Laplacian is typically studied numerically with homogeneous Dirichlet boundary conditions. In this case one has Poincaré inequalities when the region and solutions are smooth, which help in the proofs of convergence and error estimates. The game p-Laplacian has fewer tools, relying essentially on monotonicity properties and on the notion of viscosity solutions.

The paper is organized as follows. In Section 2 we introduce the formal definition of the game p-Laplacian, and give a precise definition of viscosity solutions for our problem, as well as a brief description of the stochastic game tug-of-war with noise. We formulate our approximation scheme in Section 3, where we also give insights into its construction. The analysis of the properties of the scheme, and a proof of convergence for the homogeneous case when $p \geq 2$ are presented in Section 4. In Section 5 we discuss in detail the numerical implementation of the scheme on a rectangular grid, and we conclude with some numerical tests in Section 6. For the sake of completeness, we also add in Section 7 an appendix on some elementary properties of the p-average of finite sets of real numbers.

2. Game *p*-Laplacian

The p-Laplace operator, which we refer to as the variational p-Laplacian, for $1 \le p < \infty$, is defined by

$$\Delta_p u := \operatorname{div} \left(|\nabla u|^{p-2} \nabla u \right), \tag{1}$$

whereas for $p = \infty$, traditionally is given by

$$\Delta_{\infty} u := \sum_{i,j} \frac{\partial u}{\partial x_i} \frac{\partial u}{\partial x_j} \frac{\partial^2 u}{\partial x_i \partial x_j}.$$

The subject of our numerical study is the Dirichlet boundary value problem for the so-called game p-Laplacian introduced by Peres and Sheffield [24], which for 1 reads as follows:

$$\begin{cases}
 -\Delta_p^G u = f & \text{in } \Omega, \\
 u = F & \text{on } \partial\Omega,
\end{cases}$$
(2)

where

$$\Delta_p^G u := \frac{1}{p} |\nabla u|^{2-p} \operatorname{div} \left(|\nabla u|^{p-2} \nabla u \right). \tag{3}$$

We require f and F to be continuous in their domain of definition. Additionally, we assume f either identically equal to zero or never zero. In the sequel, we will then always consider the cases $f \equiv 0$ or f > 0 (without loss of generality). Here $\Omega \subset \mathbb{R}^n$ is a bounded smooth domain.

Note that for p=2, one has $\Delta_2^G=\frac{1}{2}\Delta_2$, that is one-half of the Laplacian, which is the infinitesimal generator for a Brownian motion.

If u is a smooth function, by expanding the derivatives, we obtain

$$\Delta_p^G u = \frac{1}{p} \Delta_2 u + \frac{p-2}{p} |\nabla u|^{-2} \sum_{i,j} \frac{\partial u}{\partial x_i} \frac{\partial u}{\partial x_j} \frac{\partial^2 u}{\partial x_i \partial x_j}, \tag{4}$$

therefore, by taking the limit for $p \to \infty$, one is naturally lead to the following definition of the game ∞ -Laplacian:

$$\Delta_{\infty}^{G} u := |\nabla u|^{-2} \sum_{i,j} \frac{\partial u}{\partial x_i} \frac{\partial u}{\partial x_j} \frac{\partial^2 u}{\partial x_i \partial x_j}.$$
 (5)

The game 1-Laplacian is defined in terms of the Laplacian and the game ∞ -Laplacian:

$$\Delta_1^G u := \Delta_2 u - \Delta_\infty^G u. \tag{6}$$

We would like to point out that with this notation, the expansion in (4) allows us to think of $\bar{\Delta}_p^G$ as the convex combination of the two limiting cases, that is

$$\Delta_p^G = \frac{1}{p} \Delta_1^G + \frac{1}{q} \Delta_\infty^G, \tag{7}$$

with q the conjugate exponent of p (i.e. $\frac{1}{p} + \frac{1}{q} = 1$). At the points where $\nabla u \neq 0$, the game 1-Laplacian and the game ∞ -Laplacian can be thought as the second derivative in the orthogonal direction of ∇u and in the direction of ∇u , respectively. That is,

$$\Delta_1^G u = |\nabla u^{\perp}|^{-2} < D^2 u \, \nabla u^{\perp}, \nabla u^{\perp} >, \tag{8}$$

and

$$\Delta_{\infty}^G u = |\nabla u|^{-2} < D^2 u \, \nabla u, \nabla u >, \tag{9}$$

where D^2u denotes the Hessian matrix.

The variational p-Laplacian can be obtained as the Euler-Lagrange equation of an energy functional, a fact that does not hold for the game p-Laplacian. Additionally, while the variational p-Laplacian is degenerate elliptic for $2 and singular for <math>1 \le p < 2$, the game p-Laplacian is singular for any $p \ne 2$, so suitable definitions of viscosity solutions are needed.

Juutinen and al. [16] have shown that for the variational p-Laplacian, when $1 , the notions of viscosity solution and weak solution are equivalent. The interested reader can find in the survey [7] a number of results on viscosity solutions for second order problems. Note that, in the homogeneous case, i.e. when <math>f \equiv 0$, the solutions of the two operators agree with each other.

Various definitions of viscosity solutions for the game p-Laplacian can be given and are found in the literature. The most suitable for our treatment is the one obtained by following the definition in the classical paper of Barles and Souganidis [2].

In what follows, we will restrict ourselves to the two-dimensional case, that is we will take $\Omega \subset \mathbb{R}^2$.

Definition 2.1. Consider a smooth domain $\Omega \subset \mathbb{R}^2$, and let 1 . If <math>f is a continuous function, we say that an upper semi-continuous function (respectively, lower semi-continuous) $u : \Omega \to \mathbb{R}$ is a viscosity subsolution (respectively, supersolution) of

$$-\Delta_p^G u(x) = f(x) \text{ in } \Omega,$$

if for any $\phi \in C^2$ such that $u - \phi$ has a local maximum (respectively, local minimum) at $x \in \Omega$, we have

(i)
$$-\Delta_p^G \phi(x) \le f(x)$$
 if $\nabla \phi(x) \ne 0$
(respectively, $-\Delta_p^G \phi(x) \ge f(x)$);

(ii)
$$-\frac{\lambda_1}{p} - \frac{\lambda_2}{q} \le f(x)$$
 if $\nabla \phi(x) = 0$ and $p \ge 2$
(respectively, $-\frac{\lambda_1}{q} - \frac{\lambda_2}{p} \ge f(x)$);

$$-\frac{\lambda_1}{q} - \frac{\lambda_2}{p} \le f(x) \quad \text{if } \nabla \phi(x) = 0 \text{ and } 1
$$(respectively, -\frac{\lambda_1}{p} - \frac{\lambda_2}{q} \ge f(x));$$

$$here \frac{1}{p} + \frac{1}{q} = 1, \text{ and } \lambda_1 \le \lambda_2 \text{ are the eigenvalues of } D^2 \phi(x).$$$$

Remark 2.1. Part (ii) of the definition of viscosity subsolution (supersolution) is implied by the condition:

(ii)'
$$-\Delta_2^G \phi(x) \le f(x)$$
 whenever $\nabla \phi(x) = 0$;
(respectively, $-\Delta_2^G \phi(x) \ge f(x)$).

This is a consequence of the fact that

$$-\frac{\lambda_1}{p} - \frac{\lambda_2}{q} \le -\Delta_2^G \phi(x) \le -\frac{\lambda_1}{q} - \frac{\lambda_2}{p}, \text{ if } p \ge 2,$$

and

$$-\frac{\lambda_1}{q} - \frac{\lambda_2}{p} \le -\Delta_2^G \phi(x) \le -\frac{\lambda_1}{p} - \frac{\lambda_2}{q}, \text{ if } 1$$

Uniqueness for viscosity solutions of nonlinear operators that are singular at isolated points, typically does not depend on the particular value one assigns to these points as long as this is chosen in a consistent manner (see for example Section 9 in [7]), additionally our numerical results show numerical convergence to solutions that verify (ii)'. Therefore, we will use the following definition for viscosity solution of the game p-Laplacian:

Definition 2.2. A function u is a viscosity solution of $-\Delta_p^G u = f$, for 1 if <math>u is a subsolution and a supersolution according to (i) of Definition 2.1 and (ii)' in Remark 2.1.

As we said in the introduction, the main tool of our approximation is the way we discretize p-averages. Let us recall the notion of p-average of a set of numbers.

Definition 2.3. Given a finite set of real numbers, $S = \{s_1, s_2,, s_m\}$, we denote by $A_p(S)$ the p-average of its elements, that is $A_p(S)$ is such that

$$\sum_{j=1}^{m} |s_j - A_p(S)|^p = \min_{c \in \mathbb{R}} \sum_{j=1}^{m} |s_j - c|^p \quad \text{if} \quad 1$$

$$A_{\infty}(S) = \frac{1}{2} \left[\max_{s_j \in S} s_j + \min_{s_j \in S} s_j \right], \tag{11}$$

and

$$A_1(S) = median of S. (12)$$

Since the median of an even number of points is not uniquely defined, in (12) we follow tradition and take it to be the average of the two middle points.

Note that by convexity $A_p(S)$ above is unique for 1 . For <math>p = 2, $A_2(S)$ is the arithmetic mean of the numbers in the set S:

$$A_2(S) = \frac{1}{m} \sum_{j=1}^m s_j$$
.

We end this section with a brief description of the two-player, zero-sum game called tug-of-war with noise. In this game, we fix a parameter $\epsilon > 0$, a $1 , a domain <math>\Omega$, a continuous running cost $h: \Omega \to \mathbb{R}$ with $h \geq 0$, and a continuous exit cost $F: \partial\Omega \to \mathbb{R}$, as well as an initial position $x = x_0 \in \Omega$. A token is placed at x_0 , and at each stage, k, a fair coin is flipped. The winner of the coin flip picks any direction vector, v_k , with $|v_k| \leq \epsilon$, to which a random noise vector z_k is added. The vector z_k is equally likely to be one of the two vectors orthogonal to v_k of length $\sqrt{\frac{1}{p-1}}|v_k|$. The token is moved to $x_k = x_{k-1} + v_k + z_k$. The play continues until the token is within a distance $\hat{\epsilon} = \left(1 + \sqrt{\frac{1}{p-1}}\right) \epsilon$ from $\partial \Omega$. In that case, the winner picks $x_k \in \partial \Omega$ with $|x_k - x_{k-1}| \leq \hat{\epsilon}$. The payoff to one of the players, say player I, from the other player, say player II, is $F(x_k) + \epsilon^2 \sum_{i=1}^{k-1} h(x_i)$. Under various conditions, Peres and Sheffield [24] show that when both players choose optimal strategies, as $\epsilon \to 0$ there is an expected value $u(x) = u(x_0)$ which solves the boundary value problem for the game p-Laplacian equation given in (2) with $f = \frac{2}{q}h$. Note that as $p \to \infty$ the noise vector disappears, while for p=2 it has the same length as the chosen direction vector, resulting in a two-dimensional random walk.

3. Construction of the approximation scheme

We arrive at our approximation scheme inspired by the work of Peres and Sheffield [24], where the game interpretation of the game p-Laplace operator is based on averaging over non-Markovian paths, by asking the question if the p-game operators have an averaging characteristic, and if this can be captured by some quantity. With this in mind, looking at the numerical approximations of Oberman [20, 21] of the 1-Laplacian and the ∞ -Laplacian, as well as at the standard central difference approximation of the 2-Laplacian, we notice that they all can be rewritten in terms of their corresponding p-average.

From these observations, we arrive at the conclusion that there should be an inherent averaging characteristic in all the p-operators, and that the notion of p-average is a possible candidate for the correct quantity describing it. We refer to the work [15], for an analytical result on averaging properties of the p-Laplacian, in terms of a continuous p-average.

To stress how the notion of p-average comes into play quite naturally, when dealing with approximation schemes for the game p-Laplacian, let's look at some known approximation schemes and rethink them in terms of p-averages. To avoid cumbersome notations we present the approximation scheme in \mathbb{R}^2 but the extension to the general multidimensional case follows exactly the same lines.

For p = 2, using standard central differences, at a point $\mathbf{x} = (x_1, x_2)$, for $h \in \mathbb{R}$ small, we have

$$\Delta_2^G u(x_1, x_2) \approx \frac{1}{2h^2} \left[u(x_1 + h, x_2) + u(x_1, x_2 + h) + u(x_1 - h, x_2) + u(x_1, x_2 - h) - 4u(x_1, x_2) \right],$$

which, reordering terms in a suitable way, gives

$$\Delta_2^G u(\mathbf{x}) \approx \frac{2}{h^2} \left[A_2(C_h(\mathbf{x}, u)) - u(\mathbf{x}) \right].$$

In general, we will denote by $C_h(\mathbf{x}, u)$ the set of values used to compute the approximation at a point \mathbf{x} , in this case $C_h(\mathbf{x}, u) = \{u(x_1 + h, x_2), u(x_1, x_2 + h), u(x_1 - h, x_2), u(x_1, x_2 - h)\}.$

For $p = \infty$, the scheme in [20] can be rewritten as

$$\Delta_{\infty}^G u(\mathbf{x}) \approx \frac{2}{h^2} \left[A_{\infty}(C_h(\mathbf{x}, u)) - u(\mathbf{x}) \right],$$

where now $C_h(\mathbf{x}, u)$ is a discrete set of values of u on the sphere of radius h centered at \mathbf{x} , and the distribution and number of points on the sphere influences the accuracy of the approximation in a fundamental way.

It is relevant to mention that also for p = 2 one could pick as $C_h(\mathbf{x}, u)$ a larger set of values of u on the sphere, but since the Laplacian is a linear operator this would not increase the accuracy.

A similar scheme that uses the 1-average can be constructed in the case p = 1, in view of the interpretation as second directional derivative given by (8), see [21] for the parabolic case.

The generalization to the game p-Laplacian of these interpretations using averages suggests the following approximation:

$$\Delta_p^G u(\mathbf{x}) \approx \frac{2}{h^2} \left[A_p(C_h(\mathbf{x}, u)) - u(\mathbf{x}) \right], \tag{13}$$

where again $C_h(\mathbf{x}, u)$ would be a suitable discrete set of values of u on the sphere of radius h centered at \mathbf{x} .

We are then lead to the following approximation scheme for the Dirichlet problem (2):

$$S(\rho, \mathbf{x}, u(\mathbf{x}), u) = 0 \text{ in } \overline{\Omega},$$
 (14)

where the discretization parameters are represented by $\rho := (h, \Delta \theta)$ (h the spatial step and $\Delta \theta > 0$ the angular resolution), and $S : [0, 1) \times (0, \pi/2] \times \overline{\Omega} \times \mathbb{R} \times L^{\infty}(\overline{\Omega}) \longrightarrow \mathbb{R}$ is defined as

$$S(\rho, \mathbf{x}, u(\mathbf{x}), u) = \begin{cases} \frac{2}{\alpha^2 h^2} \left[A_p(C_h^{\Delta \theta}(\mathbf{x}, u; \alpha)) - u(\mathbf{x}) \right] + f(\mathbf{x}), & \text{in } \Omega \\ u(\mathbf{x}) - F(\mathbf{x}) & \text{on } \partial \Omega. \end{cases}$$
(15)

Here, if $d_{\Omega} < \infty$ denotes the diameter of Ω , $\alpha = \alpha(\mathbf{x})$ is a dilation parameter such that $0 < \alpha(\mathbf{x}) \leq dist(\mathbf{x}, \partial\Omega) < d_{\Omega}$. Finally, $C_h^{\Delta\theta}(\mathbf{x}, u; \alpha)$ is now a suitably chosen discrete set of values of u taken on the sphere of center \mathbf{x} and radius $h \alpha$ associated to the angular resolution $\Delta\theta$.

There is some freedom in how to choose the points in $C_h^{\Delta\theta}(\mathbf{x}, u; \alpha)$, but we follow [13] and in this work we take

$$C_h^{\Delta\theta}(\mathbf{x}, u; \alpha) = \{u(\mathbf{y}_i), i = 0, ..., 4m - 1\},$$
 (16)

where $\Delta \theta = \pi/(2m)$ and $\mathbf{y}_i = \mathbf{x} + h \alpha \mathbf{r}_i$ (with $\mathbf{r}_i = (\cos i \Delta \theta, \sin i \Delta \theta)$). Note that with this choice of m, if a direction \mathbf{r}_i is in the set of admissible directions so is its opposite, $-\mathbf{r}_i$, as well as its orthogonal and its reflections with respect to each of the axes. Also, note that with our choice of α we have that $\mathbf{y}_i \in \overline{\Omega}$ for every i, so our set $C_h^{\Delta \theta}(\mathbf{x}, u; \alpha)$ is well-defined.

4. Study of convergence

Let us analyze the convergence of our approximation scheme using the framework provided by the classical result of convergence for fully nonlinear second order elliptic equations of Barles and Souganidis presented in [2]. Assuming that a comparison principle is available for the exact solution, in this approach convergence to viscosity solutions is implied by the monotonicity, stability and consistency of the scheme.

We prove monotonicity for the general case, consistency for $p \geq 2$, and stability for the case f = 0. Therefore, we have formal convergence of the scheme only for the case $p \geq 2$ and f = 0. Nevertheless, the numerical experiments we run, and which we illustrate in Section 6, show convergence also in the general case.

We follow [2] and define $G_p: T^2 \times \mathbb{R}^2 \times \mathbb{R} \times \bar{\Omega} \to \mathbb{R}$, where T^2 is the set of 2×2 real symmetric matrices, by first introducing the function $H_p: T^2 \times \mathbb{R}^2 \times \Omega \to \mathbb{R}$:

$$H_p(M, \mathbf{r}, \mathbf{x}) = \begin{cases} -\frac{1}{p} < M \frac{\mathbf{r}^{\perp}}{|\mathbf{r}^{\perp}|}, \frac{\mathbf{r}^{\perp}}{|\mathbf{r}^{\perp}|} > -\frac{1}{q} < M \frac{\mathbf{r}}{|\mathbf{r}|}, \frac{\mathbf{r}}{|\mathbf{r}|} > -f(\mathbf{x}) & \text{if } \mathbf{r} \neq \mathbf{0} \\ -\frac{1}{2} \operatorname{tr}(M) - f(\mathbf{x}) & \text{if } \mathbf{r} = \mathbf{0}, \end{cases}$$

and then setting

$$G_p(M, \mathbf{r}, u, \mathbf{x}) = \begin{cases} H_p(M, \mathbf{r}, \mathbf{x}) & \text{for } \mathbf{x} \in \Omega, \\ u(\mathbf{x}) - F(\mathbf{x}) & \text{for } \mathbf{x} \in \partial \Omega. \end{cases}$$
(17)

I removed the variable u from the definition of H_p , since the dependence is not explicit. In the definition of G_p it is instead justified by the boundary value. Do you agree?

In this notation, the Dirichlet problem (2) is expressed as

$$G_p(D^2u, Du, u, \mathbf{x}) = 0 \quad \text{for } \mathbf{x} \in \overline{\Omega}.$$
 (18)

Remark 4.1. A viscosity solution of (18) is a function u that verifies Definition 2.2 in Ω and satisfy the boundary conditions

$$\max(H_p(D^2u, Du, \mathbf{x}), u - F) \ge 0 \text{ on } \partial\Omega,$$

$$\min(H_p(D^2u, Du, \mathbf{x}), u - F) \le 0 \text{ on } \partial\Omega.$$

The ellipticity of G_p is a trivial consequence of its definition:

Lemma 4.1. G_p is elliptic, namely for all $(\mathbf{r}, u, \mathbf{x}) \in \mathbb{R}^2 \times \mathbb{R} \times \bar{\Omega}$ we have

$$G_p(M, \mathbf{r}, u, \mathbf{x}) \le G_p(N, \mathbf{r}, u, \mathbf{x}),$$

for all $M, N \in T^2$ such that M - N is positive semidefinite.

Monotonicity of the scheme stems from the monotonicity property of the p-average, which we derive, for the convenience of the reader, in Lemma 7.2 of the Appendix.

Theorem 4.2. Let $u, v \in L^{\infty}(\overline{\Omega})$, if $u(\mathbf{x}) \geq \mathbf{v}(\mathbf{x})$ in $\overline{\Omega}$ then for all $p \geq 1$, $\rho \in [0, 1) \times (0, \pi/2]$, $\mathbf{x} \in \overline{\Omega}$, and $t \in \mathbb{R}$ it holds

$$S(\rho, \mathbf{x}, t, u) \le S(\rho, \mathbf{x}, t, v).$$

Proof. If $\mathbf{x} \in \partial \Omega$ then $S(\rho, \mathbf{x}, t, u) = t - F(\mathbf{x}) = S(\rho, \mathbf{x}, t, v)$ and the claim is clearly true. If $\mathbf{x} \in \Omega$ we have that

$$S(\rho, \mathbf{x}, t, u) = \frac{2}{\alpha^2 h^2} \left[A_p(C_h^{\Delta \theta}(\mathbf{x}, u; \alpha)) - t \right] + f(\mathbf{x})$$

$$\leq \frac{2}{\alpha^2 h^2} \left[A_p(C_h^{\Delta \theta}(\mathbf{x}, v; \alpha)) - t \right] + f(\mathbf{x}) = S(\rho, \mathbf{x}, t, v),$$

since $A_p(C_h^{\Delta\theta}(\mathbf{x}, u; \alpha)) \geq A_p(C_h^{\Delta\theta}(\mathbf{x}, v; \alpha))$, thanks to the assumption $u \geq v$ in $\overline{\Omega}$ and Lemma 7.2.

To prove consistency, we start by showing that in the case $p \geq 2$ our approximation has the correct behavior in the internal points of the domain.

Theorem 4.3. Let $p \geq 2$. For all $\mathbf{x} \in \Omega$ and $\phi \in C^{\infty}(\overline{\Omega})$, we have that

$$\lim_{(h,\Delta\theta)\to 0} \frac{2}{\alpha^2 h^2} \left[A_p(C_h^{\Delta\theta}(\mathbf{x}, \phi; \alpha)) - \phi(\mathbf{x}) \right] = \begin{cases} \Delta_p^G \phi(\mathbf{x}) & \text{if } \nabla\phi(\mathbf{x}) \neq 0 \\ \Delta_2^G \phi(\mathbf{x}) & \text{if } \nabla\phi(\mathbf{x}) = 0, \end{cases}$$
(19)

Proof. Assume $\nabla \phi(\mathbf{x}) \neq 0$, and denote by $\mathbf{e}_1 = (1,0)$; without loss of generality, we can assume $\mathbf{x} = (0,0)$ and $\nabla \phi(\mathbf{x}) = |\nabla \phi(\mathbf{0})| \mathbf{e}_1$. Equation (9) then gives

$$\Delta_{\infty}^{G}\phi(\mathbf{0}) = \partial_{11}\phi(\mathbf{0}), \tag{20}$$

while (8) yields

$$\Delta_1^G \phi(\mathbf{0}) = \partial_{22} \phi(\mathbf{0}). \tag{21}$$

The M := 4 m points in the set $C_h^{\Delta\theta}(\mathbf{x}, \phi; \alpha)$ are now on a sphere of center $\mathbf{0}$ and radius αh , with $0 < \alpha < d_{\Omega}$, see (16), that is they are given by

$$\xi_j = \alpha h \left(\cos \theta_j, \sin \theta_j\right) \text{ for } j = 1..M,$$
 (22)

where the θ_j are uniformly distributed angles verifying $|\theta_{j+1} - \theta_j| = \Delta \theta$. We use Taylor's expansion to obtain

$$\phi(\xi_j) = \phi(\mathbf{0}) + \nabla \phi(\mathbf{0}) \cdot \xi_j + \frac{1}{2} < D^2 \phi(\mathbf{0}) \xi_j, \xi_j > +o(\alpha^2 h^2),$$

and by Lemmas 7.3 and 7.1 it is enough to show that the *p*-average of $\frac{2}{\alpha^2 h^2} \left(\nabla \phi(\mathbf{0}) \cdot \xi_j + \frac{1}{2} < D^2 \phi(\mathbf{0}) \xi_j, \xi_j > \right) \text{ tends to } \Delta_p^G \phi(\mathbf{0}) \text{ as } h \text{ and } \Delta\theta \text{ tend to } 0.$

If $p = \infty$, consistency is proven as in Oberman [20]. If $p < \infty$, we employ (20), (21), (7) and the definition of ξ_j to rewrite our elements:

$$\nabla \phi(\mathbf{0}) \cdot \xi_{j} + \frac{1}{2} \langle D^{2}\phi(\mathbf{0})\xi_{j}, \xi_{j} \rangle$$

$$= \alpha h |\nabla \phi(\mathbf{0})| \cos \theta_{j} + \alpha^{2} h^{2} \partial_{12} \phi(\mathbf{0}) \cos \theta_{j} \sin \theta_{j}$$

$$+ \frac{1}{2} \alpha^{2} h^{2} \Delta_{\infty}^{G} \phi(\mathbf{0}) \cos^{2} \theta_{j} + \frac{1}{2} \alpha^{2} h^{2} \Delta_{1}^{G} \phi(\mathbf{0}) \sin \theta_{j}^{2}$$

$$= \alpha h |\nabla \phi(\mathbf{0})| \cos \theta_{j} + \alpha^{2} h^{2} \partial_{12} \phi(\mathbf{0}) \cos \theta_{j} \sin \theta_{j}$$

$$+ \frac{1}{2} \alpha^{2} h^{2} \left(\Delta_{\infty}^{G} \phi(\mathbf{0}) - \Delta_{1}^{G} \phi(\mathbf{0})\right) \cos^{2} \theta_{j} + \frac{1}{2} \alpha^{2} h^{2} \Delta_{1}^{G} \phi(\mathbf{0})$$

$$= \alpha h |\nabla \phi(\mathbf{0})| \cos \theta_{j} + \alpha^{2} h^{2} \partial_{12} \phi(\mathbf{0}) \cos \theta_{j} \sin \theta_{j}$$

$$+ \frac{1}{2} \alpha^{2} h^{2} \left(\Delta_{\infty}^{G} \phi(\mathbf{0}) - \Delta_{1}^{G} \phi(\mathbf{0})\right) \left(\cos^{2} \theta_{j} - \frac{1}{q}\right) + \frac{1}{2} \alpha^{2} h^{2} \Delta_{p}^{G} \phi(\mathbf{0}).$$

In this way, due to Lemma 7.1, we will prove our conclusion if we show that when h and $\Delta\theta$ tend to zero, the p-average of

$$\alpha h \cos \theta_j + \alpha^2 h^2 \frac{\partial_{12} \phi(\mathbf{0})}{|\nabla \phi(\mathbf{0})|} \cos \theta_j \sin \theta_j + \alpha^2 h^2 \frac{\Delta_{\infty}^G \phi(\mathbf{0}) - \Delta_1^G \phi(\mathbf{0})}{2 |\nabla \phi(\mathbf{0})|} \cos^2 \theta_j ,$$

times
$$\frac{2}{\alpha^2 h^2}$$
, tends to $\frac{\Delta_{\infty}^G \phi(\mathbf{0}) - \Delta_1^G \phi(\mathbf{0})}{|\nabla \phi(\mathbf{0})|} \frac{1}{q}$.

By definition of p-average, we then need to compute the argmin of the function Z(t):

$$Z(t) = \sum_{j=1..M} \left| \alpha h \cos \theta_j + \alpha^2 h^2 \frac{\partial_{12} \phi(\mathbf{0})}{|\nabla \phi(\mathbf{0})|} \cos \theta_j \sin \theta_j \right.$$
$$+ \left. \alpha^2 h^2 \frac{\Delta_{\infty}^G \phi(\mathbf{0}) - \Delta_1^G \phi(\mathbf{0})}{2 |\nabla \phi(\mathbf{0})|} \cos^2 \theta_j - t \right|^p,$$

but $argmin Z = \alpha^2 h^2 argmin z$, with

$$z(t) = \Delta \theta \sum_{j=1..M} |\cos \theta_j + \alpha h \frac{\partial_{12} \phi(\mathbf{0})}{|\nabla \phi(\mathbf{0})|} \cos \theta_j \sin \theta_j + \alpha h \frac{\Delta_{\infty}^G \phi(\mathbf{0}) - \Delta_1^G \phi(\mathbf{0})}{2 |\nabla \phi(\mathbf{0})|} \cos^2 \theta_j - \alpha h t|^p.$$
 (23)

We set $a = \partial_{12} \phi(\mathbf{0})/|\nabla \phi(\mathbf{0})|$ and $b = (\Delta_{\infty}^G \phi(\mathbf{0}) - \Delta_1^G \phi(\mathbf{0}))/(2 |\nabla \phi(\mathbf{0}))|$, and recall equation (45) from the Appendix, which holds only if $p \geq 2$, to derive:

$$z'(t) = -p \,\Delta\theta \,\alpha \,h \,\sum_{j=1}^{M} [(\cos\theta_j + \alpha \,h \,a \cos\theta_j \,\sin\theta_j + \alpha \,h \,b \,\cos^2\theta_j - \alpha \,h \,t)$$
$$\cdot |\cos\theta_j + \alpha \,h \,a \cos\theta_j \,\sin\theta_j + \alpha \,h \,b \,\cos^2\theta_j - \alpha \,h \,t|^{p-2}] ;$$

we next use (46), and the fundamental theorem of calculus to see that for any d and e:

$$-p |d + e|^{p-2} (d + e) = -p |d|^{p-2} d - p(p-1) \int_{d}^{d+e} |s|^{p-2} ds,$$

hence

$$z'(t) = \alpha h \Delta \theta \sum_{j=1..M} [-p | \cos \theta_j|^{p-2} \cos \theta_j$$
$$-p (p-1) \int_{\cos \theta_j}^{\cos \theta_j + \alpha h} a \cos \theta_j \sin \theta_j + \alpha h b \cos^2 \theta_j - \alpha h t \\ |s|^{p-2} ds].$$

And the change of variable $s = \cos \theta_j + \alpha h u$ gives

$$z'(t) = \alpha h \Delta \theta \sum_{j=1..M} [-p |\cos \theta_j|^{p-2} \cos \theta_j$$
 (24)

$$-\alpha h p (p-1) \int_0^a \cos \theta_j \sin \theta_j + b \cos^2 \theta_j - t |\cos \theta_j + \alpha h u|^{p-2} du].$$

We remarked previously that for our original choice of m, if a direction \mathbf{r}_j is in the set of admissible directions so is its opposite, its orthogonal, as well as its reflections in each of the axes. After rotating the coordinate system, so that $\nabla \phi(\mathbf{x}) = |\nabla \phi(\mathbf{0})| \mathbf{e}_1$, we can not make this claim any longer, as for example we lose the reflections with respect to the axes. Nevertheless, given θ_j , we still have $\theta_j + \pi$, so that the sum of the first term in (24) is zero, and we obtain

$$z'(t) = -\alpha h p (p-1) \int_0^a \cos \theta_j \sin \theta_j + b \cos^2 \theta_j - t |\cos \theta_j + \alpha h u|^{p-2} du. \quad (25)$$

The argmin of z, call it t_0 , is bounded by a constant independent of αh . This can be seen by noticing that $\alpha h t_0$ is the p-average of the values $\{\cos \theta_j + \alpha h a \cos \theta_j \sin \theta_j + \alpha h b \cos^2 \theta_j\}$; but for our choice of angles the p-average of the values $\{\cos \theta_j\}$ is zero (values are symmetric about zero), hence Lemma 7.3 implies $|\alpha h t_0| < \alpha h |a| + \alpha h |b|$, that is $t_0 < |a| + |b|$.

We would like to show that the t_0 is equal to $\frac{b}{q}$ up to an order of $O(\Delta\theta) + O(\epsilon)$. To prove our claim, we set $C(a,b,q) = 2 \max \left\{ |a| + |b|, \frac{|b|}{q} \right\}$, so that both t_0 and $\frac{b}{q}$ belong to the interval |t| < C(a,b,q), and the upper limit of integration in (25) verifies $|a\cos\theta_j\sin\theta_j + b\cos^2\theta_j - t| < |a| + |b| + C(a,b,q)$ for |t| < C(a,b,q).

On the other hand, by uniform continuity if |u| < |a| + |b| + C(a, b, q), for any $\epsilon > 0$ there is a $\delta_{\epsilon} := \delta_{\epsilon}(a, b, q)$ such that for $0 < \alpha h < \delta_{\epsilon}$ it holds $|\cos \theta_j + \alpha h u|^{p-2} = |\cos \theta_j|^{p-2} + O(\epsilon)$.

Therefore, as long as |t| < C(a, b, q), for a fixed $\epsilon > 0$, there is a δ_{ϵ} such that if $0 < \alpha h < \delta_{\epsilon}$, we have

$$z'(t) = -\alpha^{2}h^{2} \Delta\theta \ p(p-1) \sum_{j=1..M} \int_{0}^{a \cos \theta_{j} \sin \theta_{j} + b \cos^{2} \theta_{j} - t} [|\cos \theta_{j}|^{p-2} + O(\epsilon)] ds$$

$$= -\alpha^{2}h^{2} \Delta\theta \ p(p-1) \sum_{j=1..M} (a \cos \theta_{j} \sin \theta_{j}) |\cos \theta_{j}|^{p-2}$$

$$-\alpha^{2}h^{2} \Delta\theta \ p(p-1) \sum_{j=1..M} (b \cos^{2} \theta_{j} - t) |\cos \theta_{j}|^{p-2} + \alpha^{2}h^{2}O(\epsilon) ;$$

here we used the fact that $\Delta\theta \sum_{j=1..M} = 2\pi$.

Given an angle θ_j , as mentioned before, we can not assume we still have the angle $\theta_j + \frac{\pi}{2}$ as well; nevertheless, we have among our angles an approximation of it up to order $\Delta\theta$. In other words, given a certain direction, in our pool of directions we also have its reflection up to an error of order $O(\Delta\theta)$, so that

$$z'(t) = \alpha^{2} h^{2} O(\Delta \theta) + \alpha^{2} h^{2} O(\epsilon) - \alpha^{2} h^{2} \Delta \theta \ p(p-1) \sum_{j=1...M} (b \cos^{2} \theta_{j} - t) |\cos \theta_{j}|^{p-2}.$$
 (26)

To proceed in our proof, we recall the elementary equality

$$\frac{\int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} (\cos \theta)^p d\theta}{\int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} (\cos \theta)^{p-2} d\theta} = \frac{1}{q} \quad \text{for any } 1$$

which implies

$$z'\left(\frac{b}{q}\right) = \alpha^2 h^2 O(\Delta\theta) + \alpha^2 h^2 O(\epsilon), \tag{27}$$

since from it we deduce

$$\Delta \theta \sum_{j=1..M} \left(b \cos^2 \theta_j - \frac{b}{q} \right) |\cos \theta_j|^{p-2} = O(\Delta \theta).$$

We next notice that for any $c_0 > 0$ with $|b/q \pm c_0(O(\Delta\theta) + O(\epsilon))| < C(a, b, q)$, we can use (26) and (27) to obtain:

$$z'\left(\frac{b}{q} \pm c_0(O(\Delta\theta) + O(\epsilon))\right)$$

$$= z'\left(\frac{b}{q}\right) \pm c_0(O(\Delta\theta) + O(\epsilon)) \alpha^2 h^2 \Delta\theta \ p(p-1) \sum_{j=1..M} |\cos\theta_j|^{p-2}$$

$$= \alpha^2 h^2 \left(O(\Delta\theta) + O(\epsilon)\right) \left(1 \pm c_0 \Delta\theta \ p(p-1) \sum_{j=1..M} |\cos\theta_j|^{p-2}\right). \tag{28}$$

We can also find a Δ_0 , such that for any $\Delta \theta < \Delta_0$, there is a positive constant c_1 independent of $\Delta \theta$ and ϵ , for which

$$c_1 < \Delta\theta \ p(p-1) \sum_{j=1..M} |\cos\theta_j|^{p-2}.$$

Hence, since for any $\Delta\theta$ and ϵ small enough to have $|O(\Delta\theta) + O(\epsilon)| < c_1 \left[C(a,b,q) - \frac{|b|}{q} \right]$, we can pick a c_0 for which $\frac{1}{c_1} < c_0 < \frac{2}{c_1}$ and $|b/q \pm c_0(O(\Delta\theta) + O(\epsilon))| < C(a,b,q)$, from (28) we obtain

$$z'\left(\frac{b}{q}-c_0\left|O(\Delta\theta)+O(\epsilon)\right|\right)<0, \quad z'\left(\frac{b}{q}+c_0\left|O(\Delta\theta)+O(\epsilon)\right|\right)>0.$$

Recalling that $t_0 = argmin$ of z is its only critical point, and z'(t) > 0 for every $t > t_0$, while z'(t) < 0 for $t < t_0$, see Remark 7.1, we conclude that

$$\left| \operatorname{argmin} z(t) - \frac{b}{q} \right| \le \frac{2}{c_1} \left| O(\Delta \theta) + O(\epsilon) \right|, \tag{29}$$

for every $\Delta\theta$ and ϵ small enough.

Therefore, we have that for ϵ and $\Delta\theta$ small enough, there is a δ_{ϵ} such that if $0 < \alpha h < \delta_{\epsilon}$ then

$$\left| \frac{2}{\alpha^2 h^2} \operatorname{argmin} Z(t) - \frac{\Delta_{\infty}^G \phi(\mathbf{0}) - \Delta_1^G \phi(\mathbf{0})}{|\nabla \phi(\mathbf{0})|} \frac{1}{q} \right| \le |O(\Delta \theta) + O(\epsilon)|,$$

and the theorem follows for the case $\nabla \phi(\mathbf{x}) \neq 0$.

Assume $\nabla \phi(\mathbf{x}) = 0$, then by Definition 2.2 we need to show that the p-average of $\frac{2}{\alpha^2 h^2} \left(\frac{1}{2} < D^2 \phi(\mathbf{0}) \xi_j, \xi_j > \right)$ tends to $\Delta_2^G \phi(\mathbf{0})$ as h and $\Delta \theta$ tend to 0. By our choice of angles, $\xi_j = \alpha h \left(\cos j \Delta \theta, \sin j \Delta \theta \right)$, so that

$$\frac{2}{\alpha^2 h^2} \left(\frac{1}{2} < D^2 \phi(\mathbf{0}) \xi_j, \xi_j > \right) = \partial_{11} \phi(\mathbf{0}) \cos^2(j\Delta\theta)
+ \partial_{22} \phi(\mathbf{0}) \sin^2(j\Delta\theta) + 2 \partial_{12} \phi(\mathbf{0}) \cos(j\Delta\theta) \sin(j\Delta\theta)
= \frac{1}{2} \Delta_2 \phi(\mathbf{0}) + \frac{1}{2} (\partial_{11} \phi(\mathbf{0}) - \partial_{22} \phi(\mathbf{0})) \cos(2j\Delta\theta)
+ 2 \partial_{12} \phi(\mathbf{0}) \cos(j\Delta\theta) \sin(j\Delta\theta),$$

but by our assumptions if θ is in our selection of angles, so is $\theta + \frac{\pi}{2}$, thus the p-average of

$$\frac{1}{2}(\partial_{11}\phi(\mathbf{0}) - \partial_{22}\phi(\mathbf{0}))\cos(2j\Delta\theta) + 2\,\partial_{12}\phi(\mathbf{0})\cos(j\Delta\theta)\,\sin(j\Delta\theta)$$

is zero, being a set of symmetric data with respect to 0. Therefore, using (48),

$$A_p \left[\frac{2}{\alpha^2 h^2} \left(\frac{1}{2} < D^2 \phi(\mathbf{0}) \xi_j, \xi_j > \right) \right] = \frac{1}{2} \Delta_2 \phi(\mathbf{0}) = \Delta_2^G \phi(\mathbf{0})$$

and the theorem is proven.

We are now ready to show consistency of our approximation scheme. Before doing so we remark that although the definition of consistency we use is slightly different from the one given in [2], their convergent result applies also for this formulation.

Theorem 4.4. Let $p \geq 2$. Our approximation scheme is consistent, that is for all $\mathbf{x} \in \overline{\Omega}$ and $\phi \in C^{\infty}(\overline{\Omega})$, we have that

$$\limsup_{\substack{\rho \to 0 \\ \mathbf{y} \to \mathbf{x} \\ \xi \to 0}} S(\rho, \mathbf{y}, \phi(\mathbf{y}) + \xi, \phi + \xi) \le \limsup_{\substack{\mathbf{y} \to \mathbf{x} \\ \mathbf{y} \in \overline{\Omega}}} G_p(D^2 \phi(\mathbf{y}), D\phi(\mathbf{y}), \phi(\mathbf{y}), \mathbf{y})$$
(30)

(where, as before, $\rho = (h, \Delta \theta)$) and

$$\liminf_{\substack{\rho \to 0 \\ \mathbf{y} \to \mathbf{x} \\ \xi \to 0}} S(\rho, \mathbf{y}, \phi(\mathbf{y}) + \xi, \phi + \xi) \le \liminf_{\substack{\mathbf{y} \to \mathbf{x} \\ \mathbf{y} \in \overline{\Omega}}} G_p(D^2 \phi(\mathbf{y}), D\phi(\mathbf{y}), \phi(\mathbf{y}), \mathbf{y}).$$
(31)

Proof. If $\mathbf{x} \in \Omega$ both statements are a consequence of our previous theorem, i.e. Theorem 4.3. On the other hand, if $\mathbf{x} \in \partial \Omega$, we have that

$$\limsup_{\substack{\mathbf{y} \to \mathbf{x} \\ \mathbf{y} \in \overline{\Omega}}} G_p(D^2 \phi(\mathbf{y}), D\phi(\mathbf{y}), \phi(\mathbf{y}), \mathbf{y}) \\
= \max \left(H_p(D^2 \phi(\mathbf{x}), D\phi(\mathbf{x}), \mathbf{x}), \phi(\mathbf{x}) - F(\mathbf{x}) \right),$$

while

$$\lim_{\substack{\mathbf{y} \to \mathbf{x} \\ \mathbf{y} \in \overline{\Omega}}} G_p(D^2 \phi(\mathbf{y}), D\phi(\mathbf{y}), \phi(\mathbf{y}), \mathbf{y}) \\
= \min \left(H_p(D^2 \phi(\mathbf{x}), D\phi(\mathbf{x}), \mathbf{x}), \phi(\mathbf{x}) - F(\mathbf{x}) \right),$$

and the theorem follows by the definition of S.

However, as far as an explicit scheme in time is applied to

$$u_t - G_p(D^2u(\mathbf{y}), Du(\mathbf{y}), u(\mathbf{y}), \mathbf{y}) = 0, \tag{32}$$

the consistency of the scheme S with respect to the stationary nonlinear operator G_p implies the consistency for the evolutive operator as in [2]. In fact, by applying the Euler approximation in time we have, for a given initial datum u^0 , the explicit time marching scheme

$$u^{n+1} = u^n + \Delta t \ S(\rho, \mathbf{y}, u^n(\mathbf{y}), u^n), \tag{33}$$

which implies, taking $\Delta t = |\rho|$,

$$\frac{u^{n+1} - u^n}{|\rho|} = S(\rho, \mathbf{y}, u^n(\mathbf{y}), u^n).$$

Passing to the limit for ρ which tends to 0, we get consistency in the usual sense.

Theorem 4.5. Let f = 0. For all $h > 0, \Delta \theta > 0$, there exists a solution $u_{\rho} \in L^{\infty}(\overline{\Omega})$ of (14) such that $||u_{\rho}||_{L^{\infty}(\overline{\Omega})} \leq ||F||_{L^{\infty}(\partial\Omega)}$.

Proof. We consider the operator $E_{\rho}: L^{\infty}(\overline{\Omega}) \to L^{\infty}(\overline{\Omega})$ defined as

$$E_{\rho}(u)(\mathbf{x}) = \begin{cases} A_{p}(C_{h}^{\Delta\theta}(\mathbf{x}, u; \alpha)) & \text{if } \mathbf{x} \in \Omega \\ F(\mathbf{x}) & \text{if } \mathbf{x} \in \partial\Omega, \end{cases}$$

and notice that thanks to Lemma 7.1 in the Appendix, we have

$$||E_{\rho}u||_{L^{\infty}(\overline{\Omega})} \le \max(||u||_{L^{\infty}(\overline{\Omega})}, ||F||_{L^{\infty}(\partial\Omega)}). \tag{34}$$

Additionally, E_{ρ} is a nonexpansive operator in the L^{∞} norm, that is

$$||E_{\rho}(u) - E_{\rho}(v)||_{L^{\infty}(\overline{\Omega})} \le ||u - v||_{L^{\infty}(\overline{\Omega})}$$
(35)

since if $\mathbf{x} \in \Omega$, by Lemma 7.3 we know that

$$|E_{\rho}(u)(\mathbf{x}) - E_{\rho}(v)(\mathbf{x})| = |A_{p}(C_{h}^{\Delta\theta}(\mathbf{x}, u; \alpha)) - A_{p}(C_{h}^{\Delta\theta}(\mathbf{x}, v; \alpha))| \le ||u - v||_{L^{\infty}(\overline{\Omega})},$$

while

$$|E_{\rho}(u)(\mathbf{x}) - E_{\rho}(v)(\mathbf{x})| = 0$$
, if $\mathbf{x} \in \partial \Omega$.

If we consider $B(0,||F||_{L^{\infty}(\partial\Omega)})$, to be the sphere centered at the zero function and of radius $||F||_{L^{\infty}(\partial\Omega)}$, equations (34) and (35) imply that E_{ρ} is a nonexpansive operator mapping the closed sphere $\overline{B(0,||F||_{L^{\infty}(\partial\Omega)})} \subset L^{\infty}(\overline{\Omega})$ into itself. Therefore, by a classical fixed point theorem result (see Corollary 1 and Remark thereafter in [25]) we conclude that E_{ρ} has a fixed point in $\overline{B(0,||F||_{L^{\infty}(\partial\Omega)})}$, and the theorem follows.

Theorem 4.6. Assume $f \equiv 0$ and $p \geq 2$, then the solution u_{ρ} of the approximation scheme (14) converges as $\rho = (h, \Delta \theta) \rightarrow 0$ to the viscosity solution u of (18).

Proof. For f = 0, we know that (18) has a unique bounded viscosity solution u, with $||u||_{L^{\infty}(\overline{\Omega})} \leq ||F||_{L^{\infty}(\partial\Omega)}$ and that a comparison principle holds (see [18, 16] for details). Therefore, thanks to Theorems 4.2, 4.4 and 4.5, we can apply Theorem 2.1 in [2].

5. Numerical implementation

Let us now introduce in Ω a structured grid according to a space discretization parameter h, with 0 < h < 1, and denote by \mathbf{x}_j its nodes. An important step to implement our approximation scheme for a fixed angular resolution $\Delta\theta$ is the way we reconstruct the values $C_h^{\Delta\theta}(\mathbf{x}, u; \alpha)$ in (13) starting from the known values of u at the nodes \mathbf{x}_j . This is done via interpolation, but in order to obtain a convergence result we must restrict to monotone interpolation techniques. This means that denoted by u a function defined in a domain D and by $I[u](\cdot)$ its local interpolation based on the values at the nodes we can only use an interpolation operator such that

$$m = \min_{x \in D} u(x) \le I[u](x) \le \max_{x \in D} u(x) = M.$$
 (36)

Note that linear and bilinear interpolation in \mathbb{R}^2 are monotone interpolation. Another important property of bilinear interpolation is that it is translation invariant, i.e. given a constant δ we have

$$I[u+\delta](x) = I[u] + \delta. \tag{37}$$

This property will guarantee that the resulting scheme is also invariant with respect to the addition of constants.

We implement our approximation scheme on a structured grid using the classical time marching approximation (33): at the j-th node, $\mathbf{x_j}$, of the space grid we consider the iterate

$$u^{n+1}(\mathbf{x}_{j}) = u^{n}(\mathbf{x}_{j}) + \Delta t f(\mathbf{x}_{j})$$

$$+ \frac{2 \Delta t}{\alpha_{j}^{2} h^{2}} \left[A_{p}(\hat{C}_{h}^{\Delta \theta}(\mathbf{x}_{j}, u^{n}; \alpha_{j})) - u^{n}(\mathbf{x}_{j}) \right] \text{ if } \mathbf{x}_{j} \in \Omega, \qquad (38)$$

$$u^{n+1}(\mathbf{x}_{j}) = F(\mathbf{x}_{j}) \text{ if } \mathbf{x}_{j} \in \partial \Omega;$$

if for an integer m we assume $\Delta \theta = \pi/(2m)$, now

$$\hat{C}_h^{\Delta\theta}(\mathbf{x}_j, u; \alpha_j) = \{ w(\mathbf{x}_{(j,i)}, u), \ i = 0, ..., 4m - 1 \},$$
(39)

where $\mathbf{x}_{(j,i)} = \mathbf{x}_j + h \, \alpha_j \, \mathbf{r}_i$ (with $\mathbf{r}_i = (\cos i \Delta \theta, \sin i \Delta \theta)$), and $w(\mathbf{x}_{(j,i)}, u)$ is the value of the function u at $\mathbf{x}_{(j,i)}$ obtained by using bilinear interpolation on the four closest grid-points (in fact this is the main difference with respect to $C_h^{\Delta \theta}$). Here, α_j is a parameter that may vary at each grid point \mathbf{x}_j , and which verifies $0 < \alpha_* < \alpha_j < dist(\mathbf{x}_j, \partial \Omega) < d_{\Omega}$, for some constant α_* and for d_{Ω} , the diameter of Ω . In other words, α_j is a function of j, which is uniformly bounded independently of h from above and below, and such that $\mathbf{x}_{(j,i)}$ belongs to the computational domain whenever 0 < h < 1.

More precisely, if $\mathbf{x} = (x, y)$ is a point contained in a grid cell with vertices (x_k, y_l) (the lower left corner), $(x_{k+1}, y_l), (x_k, y_{l+1})$ and (x_{k+1}, y_{l+1}) , and we denote by $u_{k,l} = u(x_k, y_l)$, etc., the known values of a function u on them, the bilinear interpolation computes in (x, y) the second order polynomial

$$I[u](x,y) = axy + bx + cy + d,$$

where the coefficients a, b, c and d can be determined solving the linear system 4×4 which corresponds to the four height conditions at the four vertices of the cell. Those values can also be written as linear combinations of the values of u at the vertices of the cell, i.e.

$$I[u](x,y) = \lambda_{k,l}u_{k,l} + \lambda_{k+1,l}u_{k+1,l} + \lambda_{k,l+1}u_{k,l+1} + \lambda_{k+1,l+1}u_{k+1,l+1}$$

where the coefficients are given by

$$\lambda_{k,l} = (x_{k+1} - x)(y_{l+1} - y)/V$$

$$\lambda_{k+1,l} = (x - x_k)(y_{l+1} - y)/V$$

$$\lambda_{k,l+1} = (x_{k+1} - x)(y - y_l)/V$$
$$\lambda_{k+1,l+1} = (x - x_k)(y - y_l)/V$$

and $V := (x_{k+1} - x_k)(y_{l+1} - y_l)$ is the area of the cell, i.e. $V = h^2$ for our uniform grid (see for example [13]).

The *p*-average in (38) is then computed through the 4m values of $I[u^n]$ on a set of equally distributed points on the sphere of centre \mathbf{x}_j and radius $h\alpha_j$.

Then the approximate solution of the Dirichlet problem (2) for the game p-Laplacian is computed by using the scheme above and running it until the stopping rule

$$E_n = \max_j |u_j^{n+1} - u_j^n| \le \varepsilon \tag{40}$$

is satisfied, for a given tolerance ε . Another option is to use the simpler scheme obtained by setting $\frac{2\Delta t}{\alpha_i^2 h^2} = 1$ in (38), that is

$$u^{n+1}(\mathbf{x}_j) = A_p(\hat{C}_h^{\Delta\theta}(\mathbf{x}_j, u^n; \alpha_j)) + \Delta t \, f(\mathbf{x}_j) \quad \text{if } \mathbf{x}_j \in \Omega, \qquad (41)$$

$$u^{n+1}(\mathbf{x}_j) = F(\mathbf{x}_j) \quad \text{if } \mathbf{x}_j \in \partial \Omega.$$

until convergence (i.e. until (40) is satisfied).

We next show that for a suitable initial configuration the iteration generated by (38) converges if f = 0. Although we can not claim that this proves convergence to the solution of our approximation scheme, the numerical tests that we present in Section 6 show convergence to the correct viscosity solution in all cases were the exact solution is known, also for $f \neq 0$.

Let N be the total number of grid points, $\{\mathbf{x}_j\}_{j=1..N}$, and \mathbf{u}^n be the iterate at step n, that is the vector in \mathbb{R}^N of components $u_j^n = u^n(\mathbf{x}_j)$; we write our numerical iteration as

$$\mathbf{u}^{n+1} = T_{\rho}(\mathbf{u}^n),\tag{42}$$

where for $\mathbf{u} \in \mathbb{R}^N$, we have

$$(T_{\rho}(\mathbf{u}))_{j} = u_{j} + \Delta t \, f(\mathbf{x}_{j}) + \frac{2 \, \Delta t}{\alpha_{j}^{2} \, h^{2}} \left[A_{p}(\hat{C}_{h}^{\Delta \theta}(\mathbf{x}_{j}, \mathbf{u}; \alpha_{j})) - u_{j} \right], \, \mathbf{x}_{j} \in \Omega$$

$$(T_{\rho}(\mathbf{u}))_{j} = u(\mathbf{x}_{j}), \qquad \mathbf{x}_{j} \in \partial \Omega.$$

$$(43)$$

It is not difficult to show that for a given grid, the fact that the boundary nodes have a fixed value over each iteration prevents the numerical solution from blowing up even in the presence of the source term f. On the other hand, for $f \neq 0$ the bound depends on the grid size and f jointly, in a manner for which we don't have an independent bound. Instead, if $f \equiv 0$ it is very easy to derive that the initial values will provide a bound for any subsequent iteration as the following result shows.

Theorem 5.1. Let $\frac{2 \Delta t}{\alpha_*^2 h^2} \le 1$. Assume $f \equiv 0$, then for $n \ge 1$ it holds

$$\sup_{j=1..N} |u_j^n| \leq \sup_{j=1..N} |u_j^{n-1}|,$$

where \mathbf{u}^n is defined by (42).

Proof. It is enough to look at the internal nodes, since the approximation at the boundary nodes has fixed values. For j fixed, the set $\hat{C}_h^{\Delta\theta}(\mathbf{x}_j, \mathbf{u}; \alpha_j)$ consists of values $w(\mathbf{x}_{(j,i)}, u)$ computed by using bilinear interpolation of u^n . Hence, each $w(\mathbf{x}_{(j,i)}, u)$ is controlled by the values u_k^n , and Lemma 7.1 in the Appendix implies that

$$|u_j^n| = \left| (T_\rho(\mathbf{u}^{n-1}))_j \right|$$

$$= \left| \left(1 - \frac{2\Delta t}{\alpha_j^2 h^2} \right) u_j^{n-1} + \frac{2\Delta t}{\alpha_j^2 h^2} A_p(\hat{C}_h^{\Delta \theta}(\mathbf{x}_j, \mathbf{u}^{n-1}; \alpha_j)) \right|$$

$$\leq \left(1 - \frac{2\Delta t}{\alpha_j^2 h^2} \right) |u_j^{n-1}| + \frac{2\Delta t}{\alpha_j^2 h^2} \sup_{j \dots N} |u_j^{n-1}|,$$

and stability follows.

The next theorem shows that if the initial iteration is appropriately chosen, then the iterates generated by our schemes are point-wise increasing; this fact together with the previous stability result implies that they pointwise converge if f = 0.

Theorem 5.2. Let $\frac{2\Delta t}{\alpha_*^2 h^2} \le 1$. For $n \ge 1$, there exists an initial iteration, \mathbf{u}^0 , for which the iterations generated by the scheme (42) verify

$$u_j^n \ge u_j^{n-1}$$
 for any $j = 1..N$, and $n \ge 1$.

Proof. We choose as initial iteration:

$$u_j^0 = \begin{cases} \min_{\partial \Omega} F & \text{if } \mathbf{x}_j \in \Omega, \\ F(\mathbf{x}_j) & \text{if } \mathbf{x}_j \in \partial \Omega, \end{cases}$$

and since $u_j^0 \ge \min_{\partial \Omega} F$ for any j we have

$$A_p(\hat{C}_h^{\Delta\theta}(\mathbf{x}_j, \mathbf{u}^0; \alpha_j)) \ge \min_{\partial\Omega} F$$
 if $\mathbf{x}_j \in \Omega$.

Therefore,

$$u_j^1 = \left(1 - \frac{2\Delta t}{\alpha_j^2 h^2}\right) u_j^0 + \frac{2\Delta t}{\alpha_j^2 h^2} A_p(\hat{C}_h^{\Delta \theta}(\mathbf{x}_j, \mathbf{u}^0; \alpha_j)) + \Delta t f(x_j)$$

$$\geq \min_{\partial \Omega} F + \Delta t f(x_j) \geq u_j^0, \quad \text{if } \mathbf{x}_j \in \Omega,$$

because by assumption $f \geq 0$.

From the above we conclude $u_i^1 \geq u_i^0$ for every j, as well as

$$A_p(\hat{C}_h^{\Delta\theta}(\mathbf{x}_j, \mathbf{u^1}; \alpha_j)) \ge A_p(\hat{C}_h^{\Delta\theta}(\mathbf{x}_j, \mathbf{u^0}; \alpha_j)), \text{ if } \mathbf{x}_j \in \Omega$$

thanks to Lemma 7.2. Inserting this last inequality in the definition of u_j^2 leads to

$$u_j^2 = \left(1 - \frac{2\Delta t}{\alpha_j^2 h^2}\right) u_j^1 + \frac{2\Delta t}{\alpha_j^2 h^2} A_p(\hat{C}_h^{\Delta \theta}(\mathbf{x}_j, \mathbf{u}^1; \alpha_j)) + \Delta t f(x_j)$$

$$\geq \left(1 - \frac{2\Delta t}{\alpha_j^2 h^2}\right) u_j^0 + \frac{2\Delta t}{\alpha_j^2 h^2} A_p(\hat{C}_h^{\Delta \theta}(\mathbf{x}_j, \mathbf{u}^0; \alpha_j)) + \Delta t f(x_j)$$

$$= u_j^1, \quad \text{if } \mathbf{x}_j \in \Omega,$$

that is $u_j^2 \ge u_j^1$ for every j. We then obtain the desired conclusion by induction.

Remark 5.1. In the case $f \equiv 0$, if we pick as initial iteration

$$u_j^0 = \begin{cases} \max_{\partial \Omega} F & \text{if } \mathbf{x}_j \in \Omega, \\ F(\mathbf{x}_j) & \text{if } \mathbf{x}_j \in \partial \Omega, \end{cases}$$

and follow the steps of the proof of Theorem 5.2, we obtain a non-increasing sequence, which dominates element by element the one of Theorem 5.2. Then, if the two sequences were to converge to the same limit, we could prove that as long as we pick as initial iteration a \mathbf{u}^0 such that u_j^0 is between the minimum and maximum values of F, we would have a convergent sequence, which converges to this common limit. In particular, if we could show that our numerical implementation converges to the viscosity solution of problem (2), since for $f \equiv 0$ we know that this problem has a unique viscosity solution, we would have convergence to the viscosity solution for any initial iteration with $\min_{\partial\Omega} F \leq u_j^0 \leq \max_{\partial\Omega} F$.

6. Numerical tests

We present in this section some experiments obtained with our numerical implementation coded in MATLAB, and executed on a MacBook Pro desktop machine with a 2.2 GHz Intel Core 2 Duo processor. As described in Section 5, we have used structured uniform grids, and the values of the approximated solution at the points in $\hat{C}_h^{\Delta\theta}$ of (39) have been computed via bilinear interpolation using the four closest grid points. If one of the points lies on a line joining two grid points, the bilinear interpolation reduces to the linear interpolation between them. The subroutine used to compute the p-average at each node is a rather standard minimization algorithm for convex functions: we believe that an optimization of this part of the procedure can improve the speed of calculations.

Depending on the examples we have picked different values for the parameter α_j at different grid points. To better illustrate our choices, we introduce the following simple definition, where d_j denotes the distance of $\mathbf{x_j}$ from $\partial\Omega$ along the grid lines, and therefore $\frac{d_j}{h}$ is an integer greater or equal to 1 (see Figure 1).

Definition 6.1. An iteration generated by (38) is called n-level circles iteration if for every j the parameter α_j is chosen so that $\alpha_j = \beta \min(n, \frac{d_j}{h})$, for a given $0 < \beta \le 1$.

Test 1. For comparison, first of all we run the schemes of the previous section on the examples for the ∞ -Laplacian presented in [20], using the same specifications there provided (in this case convergence is assumed to be reached when $E_n \leq 2h10^{-2}$, where E_n is the quantity defined in (40)).

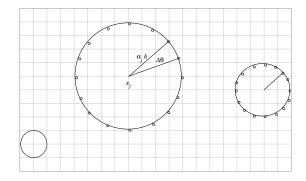


Figure 1: The 4-level circles stencil, with $\beta = 1$

We have applied scheme (38) to the case $\Omega = (-1,1) \times (-1,1), p = \infty$, f(x,y) = 0, $F(x,y) = |x|^{4/3} - |y|^{4/3}$, and the results are summarized in Table 1 below. The explicit solution of this problem is the well-known Aronsson function $u(x,y) = |x|^{4/3} - |y|^{4/3}$. We denote by N^2 (so that h = 2/N) the total number of nodes in the square grid, and by n the number of iterations (reported between parentheses), while the error is given in the maximum norm and a 2-level circles iteration is used, for $\beta = 0.99$. The initial value \mathbf{u}^0 was assumed to be a perturbation of the exact solution (of order $\pm 20\%$). The choice of grid points and number of directions makes possible to compare our results with the ones for the 17-point and the 25-point stencils presented in Table 2 of [20, pg. 1227]. Note that the errors decrease on the rows and on the columns in a regular way and that also the number of iterations decreases if we increase the number of directions and N simultaneously.

Dir.	N = 41	N = 81	N = 161	N = 241	N = 401
4	0.1105 (250)	0.0765 (448)	0.0373 (584)	0.0225 (589)	0.0122 (621)
8	0.0274 (80)	0.0182 (161)	0.0084 (214)	0.0069 (190)	0.0048 (188)
16	0.0084 (54)	0.0070 (75)	0.0043 (105)	0.0033 (108)	0.0023 (112)
24	0.0088 (57)	0.0081 (73)	0.0050 (91)	0.0035 (103)	0.0024 (107)

Table 1: L^{∞} - errors and iterations for Test 1 on Aronsson function

In Figure 2 one can see the numerical solutions and its contour plots obtained, when again $\Omega = (-1,1) \times (-1,1)$, $p = \infty$ and f(x,y) = 0, for three different choices of the boundary data. These computations are included

for comparison with Figure 2 in [20, pg. 1228]. In the first two cases, we consider $F(x,y) = |x|^2 |y|^2$ and $F(x,y) = x^3 - 3xy^2$. We use 24 controls, 2-level circles iterations, $\beta = 0.99$, and a grid with 401^2 nodes. The initial iteration for this grid is generated by a multi-resolution type approach. More specifically, we start from a 21 by 21 coarse grid with initial values set to zero in the interior nodes; after a few iterations we interpolate the numerical solution on a finer grid and repeat the procedure up to the desired resolution. This start up procedure requires some time, but significantly improves the rate of convergence. The approximations in Figure took n=15 and n=47 iterations, respectively, to converge. In the third example, F is the characteristic function of the point (1,0). We proceed as described above, but we use 24 controls on a 4-level circles iteration, again for $\beta = 0.99$. In this test, the approximation took n=2428 iterations to converge with the required accuracy.

Test 2. We next consider the problem for the game p-Laplacian in a case where we were able to compute the exact solution. Starting with $\Omega=B(0,1)$, f=1 and F=0, and working in radial coordinates we derived the solution for any $p\geq 2$, that is $v(x,y)=\frac{1-x^2-y^2}{2}$. This is an analytic function in the whole plane and has a unique extrema at the origin. By looking at v on the unit square $Q=(-1,1)\times(-1,1)$, we see that this function verifies, for any $p\geq 2$, $-\Delta_p^G v=1$ in Q. To test our code, we then implemented it on the problem: $-\Delta_p^G u=1$ in Q, with the choice of $F=\frac{1-x^2-y^2}{2}$ on the boundary. We summarize the numerical results obtained with scheme (41) for the cases p=5 and $p=\infty$ in Tables 2 and 3 below showing the L^∞ -errors and the number of iterations of the algorithm until convergence ($E_n\leq 10^{-5}$, for this test) for different combinations of levels and directions. The initial iteration was set to min F=-1/2 at the interior nodes.

Nodes (levels)	16 directions	24 directions
21 (2)	0.0634 (163)	0.0617 (180)
21 (4)	0.0241 (50)	0.0192 (107)
41 (4)	0.0201 (213)	0.0191 (163)

Table 2: Test 2, L^{∞} -errors and iterations for $p=5, f=1, F=(1-x^2-y^2)/2, \beta=0.9$.

Test 3. (*The tug-of-war game*) We finally run the code on the rectangle $(-2,2)\times(-1,1)$, for different values of p with f=1 and F=0. In particular,

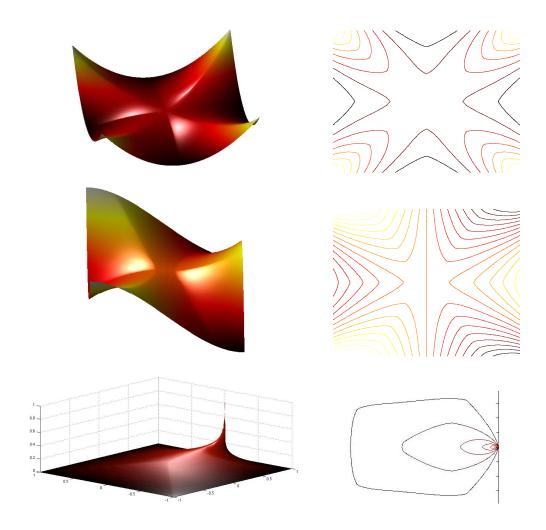


Figure 2: Test 1, numerical solutions and corresponding contour plots when $\Omega=(-1,1)^2$, $p=\infty,\ f=0$, with boundary data $F(x,y)=|x|^2\,|y|^2,\ F(x,y)=x^3-3xy^2,$ and F the characteristic function of point (1,0).

Nodes (levels)	16 directions	24 directions
21 (2)	0.0590(249)	0.0563(248)
21 (4)	0.0211 (80)	0.0185 (77)
41 (4)	0.0192 (272)	0.0156 (272)

Table 3: Test 2, L^{∞} -errors and iterations for $p = \infty$, f = 1, $F = (1 - x^2 - y^2)/2$, $\beta = 0.9$.

the case $p=\infty$ and f=1 corresponds to a running cost in the tug-of-war game. The exact solution is not known, and to our knowledge these are the first numerical results for a solution without radial symmetry. More precisely, the explicit solution is known only in a part of the domain, where it can be computed based on ideas from [23]. That is, for any -1 < x < 1, the exact solution is given by $u(x,y) = \frac{1-y^2}{2}$. We recover such solution in this part of the domain with our numerics. We set the initial iteration \mathbf{u}^0 to $\min F = 0$ in the interior nodes, and run the scheme (41) with 4-level circles iterations, with $\beta = 0.8$, until the difference of the values at (0,0) of two consecutive approximations is less than 10^{-6} (note that u(0,0) = 0.5 is the maximum value of the exact solution). Our tests have shown that the scheme is able to converge for any choice of interior values for \mathbf{u}^0 , with of course only a different number of iterations. Here we summarize in Table 4 and in Figure 3 below the results of the simulation with 16 controls (due to the axis-oriented solution, there is no real advantage in this case to use more directions).

We have also included Table 5 in order to compare the multi-level circle approaches: as expected the table shows some acceleration of convergence with larger multi-level circles, since the information from the boundary can reach the interior of the domain quicker for larger circles.

grid	error at $(0,0)$	iterations	CPU time
161×81	0.0276	1112	236
241×121	0.0155	2205	957
321×161	0.0094	3578	2681

Table 4: Test 3, $p = \infty$, $\Omega = (-2, 2) \times (-1, 1)$, f = 1, F = 0, $\beta = 0.8$, 16 directions.

levels	error at $(0,0)$	iterations	CPU time
4	0.0276	1112	236
2	0.0260	3330	621
1	0.0917	9206	1881

Table 5: Test 3, $p = \infty$, $\Omega = (-2, 2) \times (-1, 1)$, f = 1, F = 0, $\beta = 0.8$, 16 directions, 161×81 nodes.

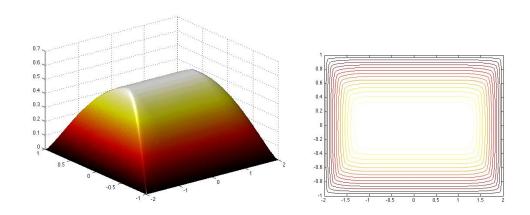


Figure 3: Test 3, surface and contour plots of the approximate solution in the rectangle $(-2,2)\times(-1,1)$ for $p=\infty,\,f=1,\,F=0,\,201\times101$ grid points and 16 directions.

7. Appendix: some p-average properties

In this appendix, for the convenience of the reader we provide some elementary properties satisfied by the *p*-average of finite sets of real numbers.

For a fixed set $S = \{s_1, ..., s_m\} \subset \mathbb{R}, s \in \mathbb{R} \text{ and } p > 1$, we define the function

$$f(s,S) = \sum_{j=1}^{m} |s_j - s|^p,$$
(44)

whose derivative with respect to s is easily computed as

$$\frac{\partial f}{\partial s}(s,S) = -p \sum_{j=1}^{m} |s_j - s|^{p-1} \operatorname{sgn}(s_j - s) = p \sum_{s_j \neq s} |s - s_j|^{p-2} (s - s_j).$$
 (45)

For $p \geq 2$, one can also compute the second derivative of f(s, S) with respect

to s:

$$\frac{\partial^2 f}{\partial s^2}(s,S) = p(p-1) \sum_{j=1}^m |s_j - s|^{p-2}.$$
 (46)

Remark 7.1. The function f has exactly one extremum and is convex in s, and the p-average $A_p(S)$ is the only value for which $\frac{\partial f}{\partial s}(A_p(S),S)=0$, hence $\frac{\partial f}{\partial s}(s,S)<0$ for every $s< A_p(S)$, and $\frac{\partial f}{\partial s}(s,S)>0$ for $s>A_p(S)$. Additionally, $A_p(S)$ solves implicitly the following equation:

$$A_p(S) = \frac{\sum_{s_j \neq A_p(S)} |s_j - A_p(S)|^{p-2} s_j}{\sum_{s_j \neq A_p(S)} |s_j - A_p(S)|^{p-2}} .$$
(47)

Lemma 7.1. Let $S = \{s_1, s_2, ..., s_m\}$ be a finite set of real numbers, and for $k \in \mathbb{R}$ let $S + k = \{s_1 + k, s_2 + k, ..., s_m + k\}$. The following assertions hold true for $1 \le p \le \infty$:

$$A_p(S+k) = A_p(S) + k, (48)$$

$$\min_{j=1..m} s_j \le A_p(S) \le \max_{j=1..m} s_j \ . \tag{49}$$

Proof. The cases p=1, $p=\infty$ are trivial. For p>1, the first assertion follows by the uniqueness of $A_p(S)$, while the second follows by Remark 7.1, and the fact that if $s_* = \min_{j=1..m} s_j$ and $s^* = \max_{j=1..m} s_j$ then by (45) one has

$$\frac{\partial f}{\partial s}(s_*, S) \le 0 \text{ and } \frac{\partial f}{\partial s}(s^*, S) \ge 0.$$

Lemma 7.2. Let $S = \{s_1, s_2, ..., s_m\}$ and $T = \{t_1, t_2, ..., t_m\}$ be two finite sets of real numbers having the same number m of elements, and let $1 \le p \le \infty$ be fixed. If it holds that $t_j \le s_j$, for every j = 1, ..., m, then we have $A_p(T) \le A_p(S)$.

Proof. According to the definition of p-average given in (10) the lemma clearly holds for the cases p = 1, and $p = \infty$.

Assume $1 . For <math>r \in \mathcal{R}$ define the function $M_p(r) = |r|^{p-2}r$ if $r \neq 0$ and $M_p(0) = 0$, note that $M_p(r)$ is a continuous increasing function.

Given $t := A_p(T)$, by Remark 7.1 we know that $\frac{\partial f}{\partial s}(t,T) = 0$, thus by equation (45) we obtain $\sum_{j=1}^m M_p(t-t_j) = 0$. But, since $t_j \leq s_j$ it holds $M_p(t-s_j) \leq M_p(t-t_j)$, therefore $\sum_{j=1}^m M_p(t-s_j) \leq \sum_{j=1}^m M_p(t-t_j) = 0$, which gives $\frac{\partial f}{\partial s}(t,S) \leq 0$. By Remark 7.1, we conclude $A_p(S) \geq t := A_p(T)$.

Lemma 7.3. Let S and T be two finite sets of real numbers having the same number of elements, and let $1 \le p \le \infty$ be fixed. Assume that $S = \{s_1, s_2, ..., s_m\}$ and $T = \{t_1, t_2, ..., t_m\}$ verify $t_j = s_j + \delta_j$, for every j = 1, ..., m, where $|\delta_j| < \delta$ for some $\delta > 0$, then one has

$$A_p(S) - \delta \le A_p(T) \le A_p(S) + \delta.$$

Proof. Since $s_j - \delta \leq t_j \leq s_j + \delta$, Lemma 7.2 implies $A_p(S - \delta) \leq A_p(T) \leq A_p(S + \delta)$. But, from equation (48) we know $A_p(S - \delta) = A_p(S) - \delta$, and $A_p(S + \delta) = A_p(S) + \delta$, so the lemma follows.

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