Operations on Signed Distance Functions*

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Abstract

We present a theoretical overview of signed distance functions and analyze how this representation changes when applying an offset transformation. First, we analyze the properties of signed distance and the sets they describe. Second, we introduce our main theorem regarding the distance to an offset set in \((X, \| \cdot \|)\) strictly normed Banach spaces. An offset set of \(D \subseteq X\) is the set of points equidistant to \(D\). We show when such a set can be represented by \(f(x) - c = 0\), where \(c \neq 0\) denotes the radius of the offset. Finally, we apply these results to gain a deeper insight into offsetting surfaces defined by signed distance functions.

Keywords: Signed Distance Functions, Sphere Tracing, Computer Graphics

1 Introduction

Surface representations for real-time graphics rely on linear approximations. With the advent of hardware accelerated tessellation units, parametric surfaces gained momentum in real-time computer graphics; however, implicit mappings are still considered infeasible for high-performance applications [2, 4, 6, 7, 9, 13]. Nevertheless, implicit functions simplify some otherwise challenging operations. For example, blending between different shapes does not necessitate the explicit representation of the target topologies when both objects are represented implicitly [3, 19]. Similarly, the result of set operations on these objects can be trivially computed [7, 12, 14, 15].

Our paper focuses on a particular class of implicit representations, signed distance functions (SDFs). Hart noted in [10] that SDFs could be rendered efficiently using a technique called sphere tracing [2, 9, 16]. This algorithm and the constant evolution of GPUs opened up the possibility of incorporating implicit representations into real-time applications, as exemplified by [1, 6, 18] more recently.

We discuss this class of functions and highlight their theoretical aspects that have practical consequences in rendering. In particular, we focus on offsetting SDF representations. Although both offsets and SDFs are simple concepts, their
combination does not always yield the expected simplicity when one tries to find a representation for the result, as highlighted in Section 6. Our paper begins with a set-theoretic overview in Section 2. We base our theorems upon these results.

Section 3 present a general algorithm for displaying surfaces defined by implicit functions, whereas Section 4 demonstrates the power SDFs provide in speeding up such tasks and their practical importance.

In Section 5, we propose a slightly different definition for signed distance function than seen in [10]. We show that the two definitions are equivalent.

We present our main result in Section 6. We show that it is possible to represent the radius $c \neq 0$ offset of $f(x) = 0$ by $f(x) - c = 0$; however, $f - c$ only produces a signed distance function on the subset of $\mathbb{R}^3$ for which $\frac{f(x)}{c} \geq 1$.

It has been observed that adding a constant value to a signed distance function produces a function that defines the offset set of the original surface [8, 10, 17]. In this paper, we analyze this operation mathematically and explain the reasons behind the effectiveness and limitations of the practical solutions.

2 Set-theoretic basics

This section reviews the definitions and results from the literature our paper relies on. Dyer et al. explain the topic in more detail in [5]. Let $(X, d)$ denote a metric space. We also use $d : X \times X \to [0, +\infty]$ to denote the distance to a set.

**Definition 1** (Distance to set). Let $A \subseteq X, p \in X$. Then

$$d(p, A) := \inf_{a \in A} d(p, a)$$
denotes the distance of \( p \) from the set \( A \). Let \( \inf \emptyset := +\infty \).

**Definition 2** (Neighborhood). Let us denote the \( r > 0 \) radius neighborhood of an element \( p \in X \) by

\[
S_r(p) := \{ x \in X : d(x, p) < r \}.
\]

\( A \subseteq X \) is open if \( \forall a \in A, \exists \epsilon > 0 : S_\epsilon(a) \subseteq A \). The set \( B \subseteq X \) is closed if \( X \setminus B \) is open. Note that \( \emptyset \) and \( X \) are both closed and open.

\( C \subseteq X \) is compact if every open covering of it can be reduced to be of finite cardinality. A compact set is closed and bounded, i.e. \( \exists R > 0 \) such that \( C \subseteq S_R(0) \). A bounded and closed set is compact if \( X \) is a finite dimensional metric space, for example \( X = \mathbb{R}^3 \).

**Lemma 1** (Existence of extremal element). Suppose \( A \subseteq X \) is closed and \( x \in X \) where \( (X, d) \) is a complete metric space. Then

\[
\exists a \in A : d(x, A) = d(x, a)
\]

The proof for Lemma 1 can be found in [11] on page 102 for \( \mathbb{R}^n \), the proof is analogous for this case [11, 5].

Furthermore, we denote the interior of the set \( A \subseteq X \) as

\[
\text{int } A := \{ a \in A | \exists \epsilon > 0 : S_\epsilon(a) \subseteq A \}
\]

The closure of \( A \subseteq X \) is

\[
\overline{A} := \{ a \in X | \forall \epsilon > 0 : S_\epsilon(a) \cap A \neq \emptyset \}
\]

The boundary of \( A \) is denoted by \( \partial A := \overline{A} \setminus \text{int } A \). For any set \( A \subseteq X \) it follows from the definitions that \( \text{int } A \) is open, \( \overline{A} \) and \( \partial A \) are closed sets.

### 3 Raymarching

From now on, let us consider surfaces defined by an \( f : \mathbb{R}^3 \rightarrow \mathbb{R} \) implicit function, such that the surface is the \( \{ f \equiv 0 \} := \{ x \in \mathbb{R}^3 | f(x) = 0 \} \) level-set. For example, the characteristic function \( 1 - X_D = X_{\mathbb{R}^3 \setminus D} : \mathbb{R}^3 \rightarrow \{0, 1\} \) is an implicit function of any \( D \subseteq \mathbb{R}^3 \) set.

A ray is a half line originating from a particular point, for example, the camera. Let us represent rays by their origin \( p \in \mathbb{R}^3 \) and unit length direction vector \( v \in \mathbb{R}^3, \|v\|_2 = 1 \). Then a ray is written as

\[
s(t) := s_{p,v}(t) := p + t \cdot v \in \mathbb{R}^3 \quad (t \geq 0).
\]

Therefore, the ray-surface intersection problem can be expressed as a root finding problem. We need to find the smallest positive root of the

\[
f \circ s : [0, +\infty) \rightarrow \mathbb{R}
\]
Algorithm 1 Raymarching a continuous implicit surface

Input: Ray defined by \( \mathbf{p} \) and \( \mathbf{v} \in \mathbb{R}^3 \), where \( \| \mathbf{v} \|_2 = 1 \)

Input: Continuous implicit function \( f : \mathbb{R}^3 \to \mathbb{R} \)

Input: \( \Delta t > 0 \) step size

Output: \( t \in [0, +\infty) \) distance traveled along the ray

1: \( t := 0; \quad f_0 := f(s(0)); \quad f_1 := f(s(\Delta t)) \)
2: for \( t < t_{\text{max}} \) and \( f_0 \cdot f_1 > 0 \) do
3: \( t := t + \Delta t; \quad \) Raymarch cycle – the bottleneck
4: \( f_0 := f_1; \)
5: \( f_1 := f(s(t)); \)
6: end for
7: \( t := \text{RefineSolution}(f \circ s, [t - \Delta t, t]); \quad \) For example, using secant method
8: return \( t \)

composite function. Usually, one can infer that \( f \) is continuous in which case raymarching that is shown in Algorithm 1 can be used to find an approximate solution. The method takes \( \Delta t \) sized steps along the ray looking for two consecutive values of different signs.

Despite being a popular algorithm for implicit surface rendering, raymarching is expensive, and it may even skip over solutions, causing visible artifacts. To provide a better ray tracing algorithm, \( f \) needs to be restricted even further which is explained in the next section.

4 Sphere Tracing

Throughout this section, we adapt the definitions from Hart [10]. Let us consider the Banach-space \((\mathbb{R}^3, \|\cdot\|_2)\) where we denote the induced metric as \( d(\mathbf{x}, \mathbf{y}) := \| \mathbf{y} - \mathbf{x} \|_2 \) \( (\mathbf{x}, \mathbf{y} \in \mathbb{R}^3) \).

Definition 3 (Distance function). \( f : \mathbb{R}^3 \to [0, +\infty) \) is a distance function if
\[
f(\mathbf{p}) = d(\mathbf{p}, \{f \equiv 0\}) \quad (\forall \mathbf{p} \in \mathbb{R}^3).
\]

Example. The distance function of the unit sphere is
\[
f_{\text{sphere}}(\mathbf{p}) = d(\mathbf{p}, S_1(0)) = \max(\| \mathbf{p} \|_2 - 1, 0) \quad (\mathbf{p} \in \mathbb{R}^3).
\]

Definition 4 (Unbounding sphere). The unbounding sphere for the distance function \( f : \mathbb{R}^3 \to [0, +\infty) \) at \( \mathbf{p} \in \mathbb{R}^3 \) is the open neighbourhood \( S_f(\mathbf{p})(\mathbf{p}) \).

It follows from Definition 3 that there are no surface points closer to \( \mathbf{p} \) than \( f(\mathbf{p}) \), i.e. \( S_f(\mathbf{p})(\mathbf{p}) \cap \{ f \equiv 0 \} = \emptyset \).
Figure 2: The sphere tracing algorithm takes distance sized steps, thereby it does not overstep a solution, yet it converges quickly. Each step defines an unbounding sphere that is disjoint from the surface.

Algorithm 2 Sphere tracing a surface defined by a distance function

\textbf{Input}: Ray defined by } p \text{ and } v \in \mathbb{R}^3, \text{ where } \|v\|_2 = 1

\textbf{Input}: Distance function } f : \mathbb{R}^3 \rightarrow \mathbb{R}

\textbf{Output}: } t \in [0, +\infty) \text{ distance traveled along the ray}

1: } t := 0; \quad i := 0;
2: for } i < i_{\text{max}} \text{ and } f(p + t \cdot v) > \epsilon \text{ do}
3: } t := t + f(p + t \cdot v);
4: } i := i + 1;
5: end for

This property shows that sphere tracing shown in Algorithm 2 can be used to find the first ray-surface intersection robustly. The algorithm iteratively takes distance-sized steps along the ray; thus no ray-surface intersection is skipped while large empty spaces are traversed quickly.

As a consequence of the above, as we approach the surface along the ray, the distance to the surface cannot change more than what we have travelled. We generalize this using the Lemma 2 and Corollary 1 below.

**Lemma 2.** Let the set } A \subseteq \mathbb{R}^n \text{ be a closed set and } x, y \in \mathbb{R}^n. \text{ Then}

\[ |d(x, A) - d(y, A)| \leq d(x, y) . \]

**Proof.** Since } A \text{ is a closed set, there exist } x', y' \in A \text{ such that } d(x, x') = d(x, A) \text{ and } d(y, y') = d(y, A) \text{ according to Lemma 1. Using the definition of the distance, we provide a lower bound to } d(x, y') \text{ and } d(y, x') \text{ respectively. The upper bound}
is given by the triangle inequality in the \( xyy' \) and \( yxx' \) triangles, respectively:

\[
d(x, x') \leq d(x, y') \leq d(x, y) + d(y, y'),
\]

(1)

\[
d(y, y') \leq d(y, x') \leq d(x, y) + d(x, x').
\]

(2)

Using (1) for the upper bound and (2) for the lower bound of \( d(x, x') \) we have:

\[
d(y, y') - d(x, y) \leq d(x, x') \leq d(y, y') + d(x, y).
\]

This proves Lemma 2.

**Definition 5** (Lipschitz constant). *Let the function \( f : \mathbb{R}^3 \to \mathbb{R} \) be arbitrary, we define the set of Lipschitz constants as

\[
\text{Lip}_f := \{ L > 0 : \forall x, y \in \mathbb{R}^3 : |f(x) - f(y)| \leq L \cdot d(x, y) \}.
\]

(3)

The function \( f \) is Lipschitz continuous if \( \text{Lip}_f \neq \emptyset \).

![Figure 3: A visualization for the proof of Lemma 2 and Proposition 1.](image)

**Corollary 1.** Every signed distance function is Lipschitz continuous and their smallest Lipschitz constant is 1. Formally:

\[
\forall f : \mathbb{R}^3 \to \mathbb{R} \text{ SDF} : \inf \text{Lip}_f = \min \text{Lip}_f = 1.
\]

Proof. First, the Lemma 2 above implies that \( \text{Lip}_f \geq 1 \) element-wise with \( D := A \).

Second \( 1 \in \text{Lip}_f \), because if \( y := x' \), then \( y = x' = y' \in A \) in the proof, then inequalities turn to equities in Equation 1. □
5 Signed Distance Functions

**Definition 6 (SDF).** If \( f : \mathbb{R}^3 \to \mathbb{R} \) is continuous and \( |f| \) is a distance function, then \( f \) is a signed distance function.

Signed distance functions (SDFs) can represent an entire volume by classifying the points of \( \mathbb{R}^3 \) belonging to its "interior" (\( \{ f < 0 \} \)), "exterior" (\( \{ f > 0 \} \)), or to the surface (\( \{ f \equiv 0 \} \)). For example, \( \mathbb{R}^3 \ni p \to \| p \|_2 - 1 \in [-1, +\infty) \) is a signed distance function of the unit sphere.

Note that distance functions are a subset of SDFs, but they cannot differentiate between interior and surface points. For signed distance functions, we give the following equivalent definition:

**Proposition 1 (SDF equivalence).** The function \( f : \mathbb{R}^3 \to \mathbb{R} \) is a signed distance function if, and only if there exists a \( \emptyset \neq D \subseteq \mathbb{R}^3 \) set for which

\[
  f(p) = \begin{cases} 
    d(p, \partial D) & \text{if } p \notin D \\
    -d(p, \partial D) & \text{if } p \in D 
  \end{cases} \tag{4}
\]

Proof. First, let us assume that \( f \) is defined according to equation (4). In this case, it follows that \( |f| \) is a distance function of the \( \partial D = \{ f \equiv 0 \} \) set. Using Lemma 2 with \( A := \partial D \), with \( x, y \in \{ f \geq 0 \} \subseteq \mathbb{R}^3 \) we know that

\[
  |f(x) - f(y)| = |d(x, \partial D) - d(y, \partial D)| \leq d(x, y),
\]

and therefore, \( f \) is uniformly continuous function on the set \( \{ f \geq 0 \} \). One can analogously show that \( f \) is continuous on the set \( \{ f \leq 0 \} \).

Assuming that \( |f| \) is a distance function where \( f : \mathbb{R}^3 \to \mathbb{R} \) is a continuous function, we have to show that the \( D := \{ f \leq 0 \} \) set satisfy equation (4). It indeed does, because \( \partial D = \{ f \equiv 0 \} \), and \( |f(p)| = d(p, \{ f \equiv 0 \}) \), and if, for example, \( f(p) > 0 \), then \( f(p) = d(p, \partial D) \) and \( p \notin D \). \qed

Hart [10] defined signed distance functions that are distance functions in absolute value. Definition 1 is similar to that of Hart, but the represented object \( D \) appears in it. Moreover, the sign is not allowed to jump on the same side of the surface, so there is a distinct "inside" and "outside" region associated with the surface. However, this intuitive definition lacks the simplicity of the original, hence the need for Definition 6.

6 Offset theorem

Let us investigate the geometric operation of offsetting on SDF representations.

**Definition 7 (Offset surface).** The offset surface at signed distance \( c \in \mathbb{R} \) of the surface defined by the SDF \( f : \mathbb{R}^3 \to \mathbb{R} \) is the \( \{ f \equiv c \} \) (level-)set.
Intuitively, offsets are obtained by inflating or deflating an initial volume by some fixed radius \( c \in \mathbb{R} \). Contrary to the naive assumption, however, offsets cannot be represented by \( f(x) - c = 0 \) in general, see the counterexample on Figure 5. Nevertheless, there’s a subset of \( \mathbb{R}^3 \) where the SDF of the offset can be written this way, as shown in Theorem 1.

First, we define strict convexity. Strictly convex Banach spaces include \( \mathbb{R}^n \), \( \mathbb{C}^n \), and \( L^p \) spaces with \( p \)-norms, if \( 1 < p < +\infty \).

**Definition 8** (Strictly convex normal space). The \((X, \|\cdot\|)\) normal space is strictly convex, if for all \( x, y, z \in X \), the following holds:

\[
d(x, z) + d(z, y) = d(x, y) \iff \exists \lambda \in [0, 1] : z = (1 - \lambda) \cdot x + \lambda \cdot y,
\]

where \( d(x, y) \) denotes the induced metric, i.e. \( d(x, y) := \|y - x\| \) (\( x, y \in X \)).

Second, the definition of the open offset set follows, which is a generalization of neighborhood in Definition 2.

**Definition 9** (Offset set). For any \( D \subseteq X \) in the metric space \((X, d)\), one can define an open offset set from \( D \) with \( r \geq 0 \) range, as

\[
S_r(D) := \{ x \in X : d(x, D) < r \}.
\]

Finally, we present the main contribution of this paper in the following

**Theorem 1** (Offset theorem). Let \((X, \|\cdot\|)\) be a strictly convex Banach space and \( D \subseteq X \) closed. Then for any \( c \geq 0 \),

\[
\forall p \in X \setminus S_c(D) : d(p, D) - c = d(p, S_c(D)).
\] (5)

Figure 4: A visualization of the proof for the offset theorem.
Proof. Since the set containing the single element \( \{p\} \subset \mathbb{R}^3 \) is compact and \( D \) is closed, the extremal points exist between the two sets according to Lemma 1:

\[
\exists p_0 \in D : d(p, D) = d(p, p_0).
\]

The \( e(t) := (1 - t) \cdot p_0 + t \cdot p \in X \), \((t \in [0, 1])\) is the parametric form of the \( \overline{p_0p} \) line segment. First we show that

\[
\forall x \in \overline{p_0p} : d(x, p_0) = d(x, D).
\]

Let us prove this by contradiction: let \( x_0 \in D \) such that \( d(x, x_0) < d(x, p_0) \). Using the definition of distance to the set, the triangle inequality in \( xx_0p_0 \), the indirect assumption, and the strict concavity, in order, we have the following:

\[
d(p, D) \leq d(p, x_0) \leq d(p, x) + d(x, x_0)
\]

\[
< d(p, x) + d(x, p_0) = d(p, p_0) = d(p, D)
\]

Which is a contradiction, so all \( x \in \overline{p_0p} \), the \( p_0 \) is a closest point in \( D \). When \( x = e(t) \), one can deduce that the distance from \( D \) along \( e \) is linear:

\[
d(e(t), D) = d(e(t), p_0) = t \cdot d(p, p_0), \quad (t \in [0, 1]).
\]

Because \( 0 \leq c \leq d(p, D) \), \( p_c := \left( \frac{c}{d(p, p_0)} \right) \in \overline{p_0p} \). Then

\[
\{p_c\} = \partial S_c(D) \cap \overline{p_0p}
\]

because the offset surface \( \partial S_c(D) = \{x \in X : d(x, D) = c\} \) contains \( p_c \) since \( d(p_c, D) = c \); moreover, \([0, 1] \ni t \rightarrow d(e(t), D) \) function is strictly increasing, so the intersection is unique. This implies half of the proposed equality (5), because

\[
d(p, S_c(D)) = d(p, \partial S_c(D)) \leq d(p, p_c) = d(p, p_0) - d(p_0, p_c) = d(p, D) - c.
\]

For the other direction, let us assume indirectly that \( d(p, S_c(D)) < d(p, p_c) \), so there exist an \( y_c \in S_c(D) \) such that \( d(p, y_c) < d(p, p_c) \) as it is shown on Figure 4b. Since \( D \) is a closed set, \( y_c \) also has a closest point in \( D \) that we denote \( y_0 \in D \).

Using the definition for the distance, the triangle inequality in \( yy_0p \), the indirect assumption, and that

\[
d(y_c, D) = d(y_c, y_0) \leq c,
\]

we arrive at a contradiction:

\[
d(p, D) \leq d(p, y_0) \leq d(p, y_c) + d(y_c, y_0)
\]

\[
< d(p, p_c) + c = d(p, p_0) = d(p, D).
\]

\[
\square
\]

Remark. i). Because equation (6) is generally false for \( t \not\in [0, 1], p \) must not be inside \( S_c(D) \).
ii). Note that the proof does not require that the closest point $p_0$ to be unique, any one of them will suffice.

iii). Consider the signed distance function form of this theorem, Corollary 2. Because of equation (6), if $f$ is differentiable at point $x \in pp_0 \subseteq \mathbb{R}^3$, then $
abla f(x) = \frac{p - p_0}{\|p - p_0\|_2}$.

We can now state the theorem on offsetting SDFs:

**Corollary 2 (Offset of an SDF).** If $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ is an SDF, then for any $0 \neq c \in \mathbb{R}$ offset, the function $f - c$ is an SDF on the set $\{\frac{f}{c} \geq 1\}$.

**Remark.**

i). $\left\{\frac{f}{c} \geq 1\right\} = \begin{cases} \{f \leq c\} & \text{if } c < 0 \\ \{f \geq c\} & \text{if } c > 0 \end{cases}$.

ii). The theorem is untrue for other points, as a counterexample is demonstrates this on Figure 5. Let $c < 0$, and $p$ be a point on a highly convex point on the surface as seen on the figure, so $\{f \equiv 0\} \ni p \not\ni \{\frac{f}{c} \geq 0\}$. Then, let $p_0$ be a closest point to $p$ on the original surface $\{f \equiv 0\}$, and $p_c$ be the closest point on offset surface $\{f \equiv c\}$. Clearly $p = p_0$, but because of the said convexity, $|c| < d(p_0, p_c) = d(p, \{f \equiv c\})$; and therefore, $d(p, \{f \equiv 0\}) - c = -c \neq d(p, \{f \equiv c\})$.

### 7 Conclusion

This paper presented a theoretical overview of surfaces defined by signed distance functions. We formulated equivalent definitions to emphasize the geometric properties of this implicit representation.

We defined an abstract offset set of an arbitrary set in Banach spaces. Our main theoretical contribution is a theorem stating a distance equivalence for points outside of the offset set.
Most importantly, Theorem 1 exposes a way to compute a signed distance function of an offset surface defined by an SDF by merely subtracting the offset radius from the function. However, this formulation is limited to the exterior of the offset volume, and the error can be arbitrarily large as we demonstrated on Figure 5.

The simple subtraction formula for offsetting a signed distance function was often used in practice, but it was only validated empirically. Our paper gave this missing guarantee and explained when this formula does not work.

References


