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T. Fidler, M. Grasmair, H. Pottmann and O. Scherzer

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### Inverse Problems of Integral Invariants and Shape Signatures

T. Fidler<sup>1</sup>, M. Grasmair<sup>1</sup>, H. Pottmann<sup>2</sup>, O. Scherzer<sup>1,3</sup>

<sup>1</sup> Institute of Computer Science, University of Innsbruck, Technikerstraße 21a, A-6020 Innsbruck, Austria.

<sup>2</sup> Institute of Discrete Mathematics and Geometry, Vienna University of Technology,

Wiedner Hauptstraße 8-10/104, A-1040 Wien, Austria

<sup>3</sup> Johann Radon Institute for Computational and Applied Mathematics, Altenberger Str. 69, A-4040 Linz, Austria.

Thomas.Fidler@uibk.ac.at.
Markus.Grasmair@uibk.ac.at.
Pottmann@geometrie.tuwien.ac.at.
Otmar.Scherzer@uibk.ac.at.
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#### Abstract

Recently, integral invariants and according signatures have been identified to be useful for shape classification, which is an important research topic in computer vision, artificial intelligence and pattern recognition. The modelling of integral invariants and signatures for shape analysis and in particular the analysis have not attracted attention in the inverse problems community so far. This paper is to point out a novel research area in inverse problems. For that purpose we provide an "inverse problems point of view" of integral invariants and signatures and highlight some fundamental mathematical perspectives.

Keywords: signatures, integral invariants

#### 1 Introduction

For *shape matching* and *classification*, an object, given by its shape, is compared with representatives of classes within a data base. Similar problems are addressed in many areas of applied sciences such as *computer vision*, *artificial intelligence* and *pattern recognition*.

The problems are tackled by a-priori assigning each object class within a data base one or more typical representatives that capture the dominant features of the class. Then, the particular object under investigation is compared with the representative shapes using an appropriate notion of similarity. Common distance measures, such as the Hausdorff distance, are not appropriate, since they do not take into account the significance of dominant features.

We consider descriptors of shapes which emphasize on peaks, edges or ridges. These features can be expressed by differentials of the shape boundary, which are invariant under rigid motions. Differentials have been used successfully for shape matching and classification, but are difficult to handle numerically, since

they are unstable with respect to noise. Alternatively, *integral invariants* descriptors have been proposed by Manay et al. [5]. In comparison with differential invariants, a significant advantage of integral invariants is that the integration kernel can be adapted to capture either global or local features. Consequently, they can be used to distinguish object features on different scales.

Although integral invariants have proven to be successful in practical applications for object classification and shape matching, from a point of view of inverse problems, important issues have not been addressed so far. This paper is to point out some of the open questions in inverse problems theory, to highlight the relation to other inverse problems, as well as to introduce this novel area to the inverse problems community.

The first question – from an inverse problems point of view probably the most important one – is the theoretical possibility of reconstructing a shape from its integral invariant. A positive answer implies that an integral invariant uniquely determines a shape, and thus can be considered as a token. This question is even more involved when only partial data of integral invariants of an object are available; such a handicap can be observed if the object is partially occluded during recording or the data of integral invariants have been damaged. The second question concerns stability and the effect of noise on the shape matching and classification process. For this purpose appropriate distance measures of integral invariants have to be determined.

The outline of this paper is as follows. In Section 2 we introduce integral invariants and further two particular examples of integral invariants in detail. In Section 3 we introduce shape signatures derived from integral invariants, which are considered to be useful for object classification. In the last part (Section 4) of this paper we present mathematical formulations of inverse problems issues related to integral invariants and signatures. Some solutions and solution concepts are presented in Section 5.

## 2 Integral Invariants

We define integral invariants as follows:

Definition 2.1:

Let  $f: \mathbb{R}_{\geq 0} \times \mathbb{R}^n \to \mathbb{R}$ ,  $n \geq 2$ , satisfy the following conditions:

- For every  $r \geq 0$  the function  $\mathbf{x} \mapsto f(r, \mathbf{x})$  is locally summable.
- For every compact subset  $K \subset \mathbb{R}^n$  and  $r_0 \geq 0$  we have

$$\lim_{r \to r_0} \int_K |f(r, \mathbf{x}) - f(r_0, \mathbf{x})| d\mathcal{L}^n(\mathbf{x}) = 0.$$

• The function f is rotationally symmetric around  $e_1$ , that is,  $f(r, \mathbf{x}) = f(r, U\mathbf{x})$  whenever  $U \in O(n)$  is an orthogonal matrix satisfying  $U\mathbf{e}_1 = \mathbf{e}_1$ .

In the following we refer to a function f satisfying above conditions as kernel function.

#### Definition 2.2:

Let f be a kernel function. For an open and bounded subset  $\Omega \subset \mathbb{R}^n$  with Lipschitz boundary  $\partial\Omega$  and  $\mathbf{x}_0 \in \mathbb{R}^n$  we define a mapping  $I[\Omega] : \partial\Omega \to \mathbb{R}$  by

$$I[\Omega](\mathbf{x}) := \int_{R_{\mathbf{x}-\mathbf{x}_0}(\Omega-\mathbf{x}_0)} f(\|\mathbf{x}-\mathbf{x}_0\|, \mathbf{y}) \, d\mathcal{L}^n(\mathbf{y}) \,,$$

where  $R_{\mathbf{y}}: \mathbb{R}^n \to \mathbb{R}^n$  is any rotation satisfying  $R_{\mathbf{y}}\mathbf{y} = \|\mathbf{y}\|\mathbf{e}_1$ .

In the definition above the point  $\mathbf{x}_0$  may either be fixed or depend on  $\Omega$ , as e.g. the choice  $\mathbf{x}_0 = \operatorname{cm}(\Omega)$  the center of mass of  $\Omega$ .

Note that the first item in the definition of the kernel function implies that  $I[\Omega](\mathbf{x})$  is finite for every  $\mathbf{x}$ , since we assume  $\Omega$  to be bounded. The second item implies continuity of  $I[\Omega]$ , that is, that  $I[\Omega](\mathbf{x_k}) \to I[\Omega](\mathbf{x})$  whenever  $\mathbf{x_k} \to \mathbf{x}$ . Finally, the last item implies that  $I[\Omega](\mathbf{x})$  is well defined in the sense that it does not depend on the choice of the rotation  $R_{\mathbf{x}-\mathbf{x_0}}$ .

#### **Examples:**

1. Let r > 0 and define  $f: \mathbb{R}_{>0} \times \mathbb{R}^n \to \mathbb{R}$  by

$$f(s, \mathbf{x}) := \chi_{B_r(s \cdot \mathbf{e_1})}(\mathbf{x}). \tag{1}$$

With this kernel function the corresponding integral invariant can be written as

$$I_{\text{Circle}}^r[\Omega](\mathbf{x}) := \mathcal{L}^n(\Omega \cap B_r(\mathbf{x})) = \chi_{\Omega} * \chi_{B_r(0)}(\mathbf{x}).$$
 (2)

The integral invariant  $I_{\text{Circle}}^r$  is called *circular area integral invariant* (see Figure 1).

2. Recall that a set  $\Omega$  is star-shaped with respect to  $\mathbf{x}_0$ , if for every  $\mathbf{x} \in \Omega$  the connecting line from  $\mathbf{x}_0$  to  $\mathbf{x}$  is contained in  $\Omega$ . Let  $\varepsilon > 0$  and  $\Omega$  be star-shaped with respect to  $\mathbf{x}_0$ . We consider  $f : \mathbb{R}_{\geq 0} \times \mathbb{R}^n \to \mathbb{R}$  defined by

$$f(s, \mathbf{x}) := \chi_{\left[-\frac{\varepsilon}{2}, \frac{\varepsilon}{2}\right]} \left(\arccos\left(\frac{\langle \mathbf{e}_1, \mathbf{x} \rangle}{\|\mathbf{x}\|}\right)\right),$$
 (3)

which is independent of s and equals the characteristic function of a cone with aperture  $\varepsilon$  and rotation axis  $\mathbf{e_1}$ . We call the corresponding integral invariant  $I_{\mathrm{Cone}}^{\varepsilon}$  cone area integral invariant (see Figure 1).

For the sake of simplicity of presentation, in the following we restrict our attention to integral invariants of two dimensional domains  $\Omega \subset \mathbb{R}^2$ , where the boundary  $\partial\Omega$  is parameterized by a homeomorphism  $\gamma:\mathbb{S}^1\to\partial\Omega$ . Then it is reasonable to regard an integral invariant not as function on  $\partial\Omega$  but rather on  $\mathbb{S}^1$ . Hence, we define

$$I[\gamma](\varphi) := I[\Omega](\gamma(\varphi)). \tag{4}$$

In this case it has been shown in [4] that for a two times differentiable curve  $\gamma$  the circular area integral invariant  $I_{\text{Circle}}^r[\gamma]$  and the curvature  $\kappa$  of  $\gamma$  are related as follows:

$$I_{\text{Circle}}^{r}[\gamma](\varphi) = \frac{\pi}{2}r^{2} - \frac{\kappa_{\gamma}(\varphi)}{3}r^{3} + O(r^{4}).$$
 (5)

In general, if  $\gamma$  is not twice differentiable, we have

$$I_{\text{Circle}}^{r}[\gamma](\varphi) = \frac{\delta}{2}r^{2} - \frac{\kappa_{\gamma}^{-}(\varphi) + \kappa_{\gamma}^{+}(\varphi)}{6}r^{3} + O(r^{4}), \tag{6}$$

where  $\delta$  denotes the aperture of the circular sector centered at the non-smooth point of the curve, and  $\kappa_{\gamma}^{-}$ ,  $\kappa_{\gamma}^{+}$  denote the curvature to the right and left, respectively (see [6]).

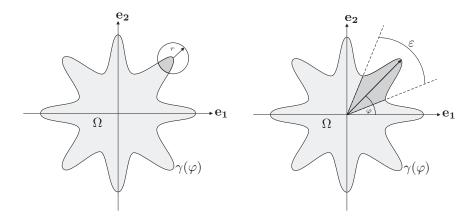


Figure 1: (2D Visualization of different integral invariants) Left: Circular area integral invariant. For each point located on the curve  $\gamma$  the area of intersection of  $\Omega$  with a ball of radius r is measured. Right: Cone area integral invariant. For each value of the angle  $\varphi$  the area of intersection of a cone with aperture  $\varepsilon$  and the domain  $\Omega$  is measured.

In the definitions of the circular and cone area integral invariants, the parameters r and  $\varepsilon$ , respectively, control the sensitivity of each invariant with respect to local variations in  $\gamma$ . Large parameters guarantee that the invariants are less sensitive to local variations. Moreover, from Table 1 it becomes evident that  $I_{\text{Cone}}^{\varepsilon}$  is less sensitive to local variations of the curve than  $I_{\text{Circle}}^{r}$ , which in turn is much less sensitive than the curvature as an example of a differential invariant. The higher stability of  $I_{\text{Cone}}^{\varepsilon}$  results from the fact that the barycenter is less affected by small changes of the curve than the center of the ball of integration, both being the reference point for the corresponding integral invariant. Thus, from a point of view of stability the cone area integral invariant is preferable; the use of  $I_{\text{Cone}}^{\varepsilon}$ , however, is restricted to star-shaped domains. Furthermore, in the case of incomplete data it is impossible to determine the barycenter of the domain, and subsequently the cone area integral invariant turns out to be useless.

## 3 Shape Signatures

Integral invariants are invariant with respect to rotations and translations but suffer from the dependence of the parameterization of the curve  $\gamma$ . Without a specification of the parameterization the integral invariant is not suitable for

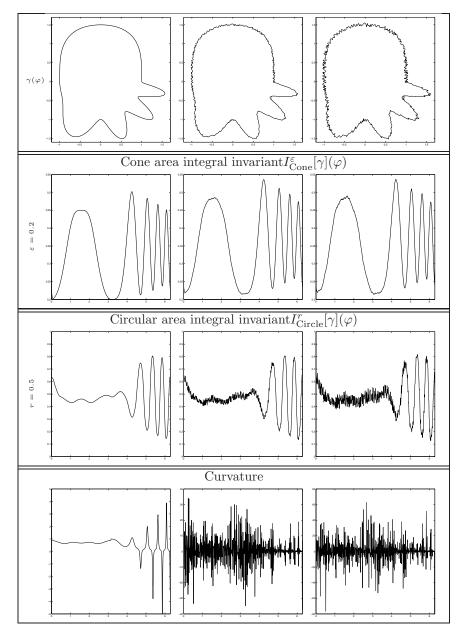


Table 1: (Circular versus cone area integral invariant) First row: Three test curves. The original curve (first image) is perturbed with  $\pm 1.25\%$  (second image) and  $\pm 2.5\%$  uniformly distributed noise (third image). Second row: Cone area integral invariant  $I_{\text{Cone}}^{\varepsilon}[\gamma]$  with  $\varepsilon=0.2$ . Third row: Circular area integral invariant  $I_{\text{Circle}}^{r}[\gamma]$  with r=0.5. Fourth row: Curvature of the test curves.

shape classification. In particular, a different choice of the starting point of the parameterization results in a shift of the integral invariant  $I[\gamma]$ , which is undesirable for shape classification and matching. To get rid of the effects of reparameterization, Manay et al. [5] defined signatures based on integral invariants, which are curves, where  $I[\gamma]$  is plotted against its derivative  $I[\gamma]$ . Thus, they followed the standard approach for defining differential signatures (cf. [2]) where the curvature  $\kappa$  is plotted against  $\kappa'$ . In both definitions the derivatives of integral and differential invariants, respectively, are required, which have to be provided numerically. Since the stable numerical computation of derivatives is quite demanding we introduce a slightly different concept of signatures the idea of which can be found in [6].

Definition 3.1:

A signature of a curve  $\gamma$  is the set

$$\{(I_1[\gamma](\varphi), I_2[\gamma](\varphi)) : \varphi \in \mathbb{S}^1\}, \tag{7}$$

where  $I_1[\gamma]$ ,  $I_2[\gamma]$  denote two arbitrary integral invariants.

The signature does not depend on the parameterization of the curve  $\gamma$ . As an advantage to previous approaches no numerical derivatives are required. Therefore, the approach is expected to be more robust with respect to curve perturbations than the original definition given by Manay.

In principle it is possible to combine any two integral invariants to a signature. In practice, however, it seems more reasonable to use two of the same class. Using the invariants introduced in Section 2 this leads to the signatures

$$\left\{ \left( I_{\text{Circle}}^{r_1}[\gamma](\varphi), I_{\text{Circle}}^{r_2}[\gamma](\varphi) \right) : \varphi \in \mathbb{S}^1 \right\}, \quad r_1, r_2 > 0, \ r_1 \neq r_2, \\
\left\{ \left( I_{\text{Cone}}^{\varepsilon_1}[\gamma](\varphi), I_{\text{Cone}}^{\varepsilon_2}[\gamma](\varphi) \right) : \varphi \in \mathbb{S}^1 \right\}, \quad \varepsilon_1, \varepsilon_2 > 0, \ \varepsilon_1 \neq \varepsilon_2,$$
(8)

respectively. The signatures form closed curves that may contain self-intersections or segments that are passed through several times (cf. Table 2).

Although the signature formed by the circular integral invariant seems to be heavily affected by noise, its most prominent features vary only slightly. Moreover, these spikes in the signature correspond to the features in the object. This effect can also be seen in Table 3, which shows the influence of occlusions on the signature.

## 4 An Inverse Problems Point of View on Invariants and Shape Signatures

From an inverse problems point of view, questions to be addressed in classification applications concern the identifiability of a shape or some properties of the integral invariant or signature, respectively.

The integral invariant as defined in (4) can be regarded as a mapping  $I: C^0(\mathbb{S}^1) \to C^0(\mathbb{S}^1)$ , where  $C^0(\mathbb{S}^1)$  denotes the set of all continuous functions on  $\mathbb{S}^1$  taking values in  $\mathbb{R}^2$ . Injectivity of this mapping can only be achieved up to some restrictions, which are highlighted in the following.

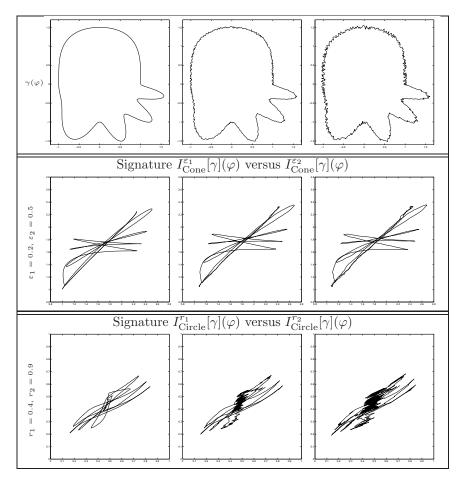


Table 2: (Signatures) First row: Three test curves. The original curve (first image) is perturbed with  $\pm 1.25\%$  (second image) and  $\pm 2.5\%$  uniformly distributed noise (third image). Second row: Corresponding signatures using the cone area integral invariant  $I_{\text{Cone}}^{\varepsilon}[\gamma]$  with  $\varepsilon_1=0.2$  and  $\varepsilon_2=0.5$ . Third row: Corresponding signatures using the circular area integral invariant  $I_{\text{Circle}}^{r}[\gamma]$  with  $r_1=0.4$  and  $r_2=0.9$ .

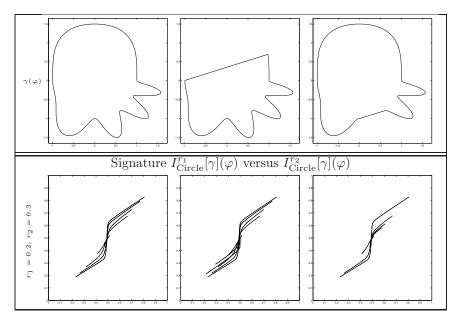


Table 3: First row: Original curve and curves with occlusions. Second row: Corresponding signatures using the circular area integral invariant  $I_{\text{Circle}}^r[\gamma]$  with  $r_1 = 0.2$  and  $r_2 = 0.3$ .

- 1. Since I is invariant with respect to translations and rotations of the curve, it is evident that I is only injective up to rigid motions.
- 2. The choice of the parameters defining the integral invariants affects the possibility of differing between shapes. For instance, if  $\varepsilon = 2\pi$ , the function  $I_{\text{Cone}}^{\varepsilon}$  is constant with function value the total area of  $\Omega$ . Therefore, in this situation the functional can only be used to discriminate between objects of different size. Similarly, if  $r \geq \operatorname{diam} \Omega$ , then  $I_{\text{Circle}}^{r}$  equals the constant curve  $\varphi \mapsto \mathcal{L}^{2}(\Omega)$ .
- 3. In applications, a difficulty for shape classification by integral invariants is the dependence on the choice of the parameterization of the curve. In fact, assume that  $J: \mathbb{S}^1 \to \mathbb{R}$  is topologically equivalent to the invariant  $I[\gamma]$ , that is, there exists a homeomorphism  $\eta$  of  $\mathbb{S}^1$  such that  $J \circ \eta = I[\gamma]$ . Then  $I[\gamma] = I[\gamma \circ \eta^{-1}] \circ \eta$ . Therefore,  $J = I[\gamma \circ \eta^{-1}]$ , is the integral invariant of the reparameterized curve  $\gamma \circ \eta^{-1}$ . Consequently, we may only obtain injectivity results if the parameterization of the boundary of  $\Omega$  is specified, e.g. by prescribing  $\gamma(0)$  and  $\dot{\gamma}(0)$ . A natural choice for parameterizing an arbitrary domain  $\Omega$  is the arclength parameterization and for a star-shaped domain an angular parameterization with respect to the barycenter.

Taking into account the considerations above, we formulate some inverse problems related to shape classification with integral invariants.

PROBLEM 4.1:

Denote by  $C^0_{\mathrm{arc}}(\mathbb{S}^1)$  the set of all arclength parameterized closed curves  $\gamma$  in  $\mathbb{R}^2$ 

satisfying  $\gamma(0) = \mathbf{0}$  and  $\dot{\gamma}(0) = \mathbf{e_1}$ , and by

$$C_{\text{angle}}^{0}(\mathbb{S}^{1}) := \left\{ \gamma \in C^{0}(\mathbb{S}^{1}) : \gamma(\varphi) = \|\gamma(\varphi)\|(\cos\varphi, \sin\varphi), \ \operatorname{cm}(\Omega) = \mathbf{0} \right\}$$
(9)

the set of all angular parameterized closed curves such that the center of mass of the enclosed area  $\Omega$  is **0**.

- Let  $0 < \varepsilon < 2\pi$ . Is the function  $I_{\text{Cone}}^{\varepsilon}$  restricted to  $C_{\text{angle}}^{0}(\mathbb{S}^{1})$  injective?
- Let r>0. Denote by  $C_r^0(\mathbb{S}^1)$  the set of all curves  $\gamma$  in  $\mathbb{R}^2$  satisfying

$$\max\{\|\gamma(\varphi) - \gamma(\tau)\| : \varphi, \tau \in \mathbb{S}^1\} > r. \tag{10}$$

Is the function  $I_{\text{Circle}}^r$  restricted to  $C_{\text{arc}}^0(\mathbb{S}^1) \cap C_r^0(\mathbb{S}^1)$  injective?

- Is it possible to reconstruct a curve from its integral invariant numerically in a stable way?
- Is it possible to identify features of an object from its invariant? If it is possible, how can this additional information be used to tackle the problem of incomplete data? If applicable, can an algorithm be derived that partially recovers the object of interest?

Discussing the injectivity of the mapping which assigns to each Jordan curve its signature is more complicated. Signatures have been introduced in Definition 3.1 as mere point sets to derive a shape characterization which is independent of the parameterization. In most applications, though, at least some information on the parameterization is still present, e.g. the direction in which this point set is passed through and the number of times one specific point in the signature is attained. This can be modelled by assuming that we only know the mapping  $\varphi \mapsto \left(I_1[\gamma](\varphi), I_2[\gamma](\varphi)\right)$  up to a reparameterization. Therefore, we define the quotient space

$$C_T^0(\mathbb{S}^1) := C^0(\mathbb{S}^1)/\sim,$$
 (11)

where two mappings  $\beta_1$ ,  $\beta_2$  are equivalent, if there exists a homeomorphism  $\eta$  of  $\mathbb{S}^1$  such that  $\beta_1 = \beta_2 \circ \eta$ . For given integral invariants  $I_1$  and  $I_2$ , and a curve  $\gamma$  we can define

$$\Sigma([\gamma]) := [(I_1[\gamma], I_2[\gamma])], \tag{12}$$

where  $[\gamma]$  and  $[(I_1[\gamma], I_2[\gamma])]$  denote the equivalence classes in  $C_T^0(\mathbb{S}^1)$  of the curves  $\gamma$  and  $(I_1[\gamma], I_2[\gamma]) \in C^0(\mathbb{S}^1)$ .

#### PROBLEM 4.2:

Let  $I_1$ ,  $I_2$  be integral invariants.

- For which choices of  $I_1$ ,  $I_2$  and subsets  $X \subset C^0_T(\mathbb{S}^1)$  of admissible curves is the mapping  $\Sigma: X \to C^0_T(\mathbb{S}^1)$  injective?
- Does there exist a numerical algorithm for reconstructing a curve (up to reparameterization) from its signature?
- Similar to the case of integral invariants: Is it possible to identify features in the signature which are correlated to features of the equivalence class of a curve? If there exists a correlation, how can this be taken into account for identifying curves from incomplete signatures?

As a prerequisite step for integral invariants calculation, the shape has to be extracted from data which contains the object of interest. By this extraction perturbations and noise are introduced in the shape. To measure the influence of data errors in the shape representation of integral invariants and signatures, appropriate distance measures for objects and the integral invariants have to be introduced.

#### PROBLEM 4.3:

Let  $\gamma_1$ ,  $\gamma_2$  be Jordan curves.

- What is a reasonable distance measure between two curves, for instance a noisy, perturbed curve and a representative of a class in a data base?
- Does there exist a continuous dependence of the invariants and signatures from the curves?

Similar questions have to be answered in the case of integral invariants for higher dimensional objects  $\Omega \subset \mathbb{R}^n$ , n > 2.

## 5 Relation to Literature on Inverse Problems and Partial Solutions

It is shown in [3] that  $I_{\text{Cone}}^{\varepsilon}$  is injective on  $C_{\text{angle}}^{0}(\mathbb{S}^{1})$  if and only if  $\varepsilon/\pi$  is irrational. A similar characterization of the injectivity of  $I_{\text{Cone}}^{\varepsilon}$  is also given for  $\Omega \subset \mathbb{R}^{n}$  with n>2. Therefore, we expect that the signatures corresponding to cone integral invariants  $I_{\text{Cone}}^{\varepsilon_{1}}$ ,  $I_{\text{Cone}}^{\varepsilon_{2}}$  are unique, if  $\varepsilon_{1} \neq \varepsilon_{2}$  and  $\varepsilon_{i}/\pi$  irrational for i=1,2. However, a rigorous proof is missing so far.

From the results in [4] (see also (5)) it follows that the circular invariant  $I^r_{\text{Circle}}$  approximates the curvature of  $\gamma$  as  $r \to 0$ . Since the curvature uniquely determines an arclength parameterized curve up to rigid motions, it follows that  $I^r_{\text{Circle}}$  is injective in the limit as  $r \to 0$ . In [3] we have used the Landweber iteration method for reconstructing the original curve  $\gamma \in C^0_{\text{arc}}(\mathbb{S}^1) \cap C^0_r(\mathbb{S}^1)$  from its circular area integral invariant  $I^r_{\text{Circle}}$  for r > 0. Numerical experiments have shown that injectivity of that function for positive fixed radii can hold, but the proof is still missing.

There exist various results concerning the question whether a function  $f \in L^1_{loc}(\mathbb{R}^n)$  is uniquely determined by integrals over certain subsets. One of the results is that the integrals

$$\int_{B_{r_i}(\mathbf{x})} f(\mathbf{y}) d\mathcal{L}^n(\mathbf{y}), \quad \mathbf{x} \in \mathbb{R}^n, i = 1, 2,$$

determine f, if and only if  $r_1/r_2$  is not the ratio of two zeros of certain Bessel functions (see [7, Chpt. 2, Thm. 1.6]).

Moreover, there exists a relation between the circular integral invariant and the spherical Radon transform  $\mathcal{R}[f]: \mathbb{R}^2 \times \mathbb{R}_{\geq 0} \to \mathbb{R}$ , which is defined by

$$\mathcal{R}[f](\mathbf{x},r) := r^{-1} \int_{\mathbb{S}^1} f(\mathbf{x} + r\boldsymbol{\sigma}) d\mathcal{L}^1(\boldsymbol{\sigma})$$
 (13)

for a function  $f \in L^1_{loc}(\mathbb{R}^2)$ . Indeed, we have

$$I_{\text{Circle}}^{r}[\gamma](\varphi) = \int_{0}^{r} \mathcal{R}[\chi_{\Omega}](\gamma(\varphi), s) d\mathcal{L}^{1}(s). \tag{14}$$

For the spherical Radon transform there exist injectivity results, which show that for certain subsets  $S \subset \mathbb{R}^2$  it is possible to reconstruct a function f with compact support from its spherical Radon transform restricted to  $S \times \mathbb{R}_{\geq 0}$  (cf. [1]). In other words, it is enough to know  $\mathcal{R}[f]$  for a relatively small subset S of  $\mathbb{R}^2$  but all radii r, in order to be able to identify f. In [8], identifiability of f from data  $\{\mathcal{R}[f](\mathbf{x},r): \mathbf{x} \in \mathbb{R}^2, r_1 < r < r_2\}$  for appropriate  $r_1, r_2 \in \mathbb{R}_{\geq 0}$  has been shown.

Considering (14) the circular area integral invariant  $I_{\text{Circle}}^r$  represents the average of the spherical Radon transform over an interval [0, r] evaluated at  $\mathbf{x} \in \partial \Omega$ . Thus, a lot less information is given than required for the two injectivity results mentioned above. However, we additionally know, that the function to be reconstructed is a characteristic function. This information is not taken into account in the standard references on injectivity of the spherical Radon transform [1].

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