Approximate reconstruction from circular mean data via classical summability

W. R. Madych

Univ. of Connecticut, Storrs, CT USA

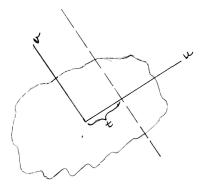
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Report of work done in collaboration with Marcus Ansorg, Frank Filbir, and Ruben Seyfried at the Institute of Biomathematics and Biometry, Helmholtz Center Munich, Germany. Notation:

$$\begin{split} &x = (x_1, x_2) \in \mathbb{R}^2, \text{ also use } y, \ \xi \text{ etc.} \\ &\text{Scalar (inner, dot) product: } \langle x, y \rangle = x_1 y_1 + x_2 y_2. \\ &u = u(\theta) = (\cos \theta, \sin \theta), \\ &v = u^{\perp} = u(\theta + \pi/2). \end{split}$$

Radon transform:

$$Rf(u,t) = \int_{-\infty}^{\infty} f(tu + sv) ds = f_u(t).$$

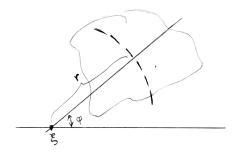


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Applications are well known.

Circular mean transform:

$$\mathcal{M}f(\xi,r)=\int_{0}^{2\pi}f(\xi+ru(\phi))d\phi$$



Models the data acquisition scheme in photoacoustic, also known as thermoacoustic, tomography that is currently being tested for possible clinical applications. Typical assumptions are that f has support in the unit disk and $\xi = u(\theta)$, $0 \le \theta < 2\pi$.

We are interested in reconstruction of f in terms of the data $\mathcal{M}f(\xi, r), \xi \in \Xi$, where Ξ is some appropriate collection of detectors exterior to the unit disk. An important special case is $\Xi = S^1$ the unit circle, the boundary of the unit disk with data $\mathcal{M}f(u(\theta), r), 0 \le \theta < 2\pi$.

In the case $\Xi = S^1$ exact inversion formulas are known but, as in the case of the classical Radon transform, need to be regularized for numerical work.

Inversion formulas for f in terms of the data $\mathcal{M}f(u(\theta), r)$, $0 \le \theta < 2\pi$, were first published in FHR = D. Finch, M. Haltmeier, and Rakesh, Inversion of spherical means and the wave equation in even dimensions, *SIAM J. Appl. Math.* 68, no. 2, (2007), 392-412.

An alternate derivation and further generalizations can be found in Y. A. Antipov, R, Estrada, and B. Rubin, Inversion formulas for spherical means in constant curvature spaces, (2011) preprint.

A different inversion formula can be found in L. A. Kunyansky, Explicit inversion formulae for the spherical mean Radon transform, *Inv. Prob.* 23, (2007), 373-383.

Our method is a adaptation of a variant of a classical summability procedure used for Radon transform data outlined in W. R. Madych, Summability and approximate reconstruction from Radon Transform data, *Contemporary Mathematics*, Vol 113 (1990), 189-219.

Recall the notion of a *ridge function*:

 $H(x) = h(\langle x, u \rangle)$

If f(x) has compact support

$$H * f(x) = \int_{\mathbb{R}^2} h(y) f(x - y) dy$$

= $\int_{-\infty}^{\infty} h(t) Rf(u, \langle x, u \rangle - t) dt$
= $h * f_u(\langle x, u \rangle).$

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If K(x) is a sum of ridge functions, i. e.

$$K(x) = \int_0^{2\pi} h(\langle x, u(\theta) \rangle) \frac{d\theta}{2\pi}$$

then

(1)

$$K * f(x) = \int_{0}^{2\pi} \left\{ \int_{-\infty}^{\infty} h(t) Rf(u(\theta), \langle x, u(\theta) \rangle - t) dt \right\} \frac{d\theta}{2\pi}$$

$$= \int_{0}^{2\pi} h * f_{u(\theta)}(\langle x, u(\theta) \rangle) \frac{d\theta}{2\pi}.$$

If K is a an approximation of the identity then (1) gives rise to a reconstruction algorithm for f in terms of its Radon transform data.

Remark: There is a formula for h in terms of K that can be explicitly evaluated in certain cases.

Typical examples:

(2)
$$K(x) = \frac{1}{\pi} \begin{cases} 1 & \text{if } |x| \le 1\\ 0 & \text{otherwise} \end{cases}$$

(3)
$$K(x) = \frac{1}{\pi} \begin{cases} 1 & \text{if } |t| \le 1\\ 1 - |t|/(t^2 - 1)^{1/2} & \text{otherwise,} \end{cases}$$

(3)
$$K(x) = \frac{3}{\pi} \begin{cases} 1 - |x| & \text{if } |x| \le 1\\ 0 & \text{otherwise} \end{cases}$$

$$h(t) = \frac{3}{\pi} \begin{cases} 1 - \frac{\pi}{2}|t| & \text{if } |t| \le 1\\ 1 - t \arcsin(1/t) & \text{otherwise.} \end{cases}$$

(4)
$$K(x) = \frac{1}{2\pi} \frac{1}{(1+|x|^2)^{3/2}}$$
 $h(t) = \frac{1}{2\pi} \frac{1-t^2}{(1+t^2)^2},$

Note that in all the above cases the family of functions parametrized by ϵ

$$K_{\epsilon}(x) = rac{1}{\epsilon^2} Kigg(rac{x}{\epsilon}igg)$$

are well known approximations of the identity as $\epsilon \rightarrow 0$. The corresponding functions $h_{\epsilon}(t)$ of course, are given by

$$h_{\epsilon}(t) = rac{1}{\epsilon^2} h\!\left(rac{t}{\epsilon}
ight)$$

We use the same philosophy to reconstruct f from its circular mean transform data.

G is radial with center ξ :

$$G(x) = g(|x - \xi|)$$

If f has compact support

$$\int_{\mathbb{R}^n} G(x)f(x)dx = \int_0^\infty \int_0^{2\pi} g(r)f(\xi + ru(\theta))rd\theta dr$$
$$= \int_0^\infty g(r)\mathcal{M}f(\xi, r)rdr.$$

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If K(x, y) is a sum of radial functions in the variable y, i. e.

$$K(x,y) = \int_{\Xi} k(x,\xi,|y-\xi|) d\mu(\xi)$$

then

(5)
$$\int_{\mathbb{R}^2} \mathcal{K}(x,y) f(y) dy = \int_{\Xi} \left\{ \int_0^\infty k(x,\xi,r) \mathcal{M}f(\xi,r) r dr \right\} d\mu(\xi).$$

If f is sufficiently regular and has support in a region Ω and K(x, y) is a good approximation of the identity in y at each $x \in \Omega$ then then identity (5) represents an approximate reconstruction of f in terms of the data $\mathcal{M}f(\xi, r), \xi \in \Xi$ and r > 0.

Such a kernel K(x, y) can be conveniently viewed as a family of functions in the y variable parameterized by x, each member of which is a sum of radial functions with centers in Ξ .

As alluded to earlier, we will study the case $\Xi = S^1 = \{\xi = u(\theta) : 0 \le \theta < 2\pi\}$ with $d\mu(\xi) = \frac{d\theta}{2\pi}$ and fsupported in $B = \{x : |x| < 1\}$.

Remark: I don't know how to solve

$$G(y) = \int_{S^1} g(|y-u|) d\mu(u)$$

for g and μ in terms of G.

This leaves us with the problem of how to construct such K(x, y) and $k(x, \xi, |y - \xi|)$ pairs?

Note that

$$\lim_{r\to\infty}\left\{|x-ru|-|y-ru|\right\}=\langle y-x,u\rangle.$$

This suggests that, roughly speaking, if the detector $\xi = ru$ is relatively far from x and y then $|x - \xi| - |y - \xi|$ looks like $\langle y - x, u \rangle$.

We know that

$$\epsilon^{-2}K((y-x)/\epsilon) = \int_0^{2\pi} \frac{1}{\epsilon^2} h\Big(\frac{\langle y-x, u(\theta) \rangle}{\epsilon}\Big) \frac{d\theta}{2\pi}$$

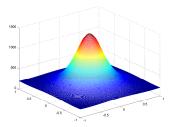
is a good approximation of the identity at x with an appropriate choice of h. i. e. one of the examples of K, h pairs.

Hence, it is not unreasonable to expect that

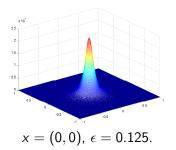
$$\mathcal{K}_1(x,y;\epsilon) = \int_0^{2\pi} \frac{1}{\epsilon^2} h\Big(\frac{|x-u(\theta)|-|y-u(\theta)|}{\epsilon}\Big) \frac{d\theta}{2\pi}$$

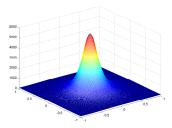
where *h* comes from the ridge function representation of a kernel *K*, i. e. one of the examples (3) or (4), looks like a summability kernel or approximate identity at *x*. At least for *x* and *y* close to the origin.

Plots of $K_1(x, y; \epsilon)$ for fixed x and ϵ as a function of y. Here $h(t) = \frac{1}{2\pi} \frac{1-t^2}{(1+t^2)^2}$ as in example (4).

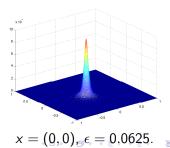


 $x = (0, 0), \epsilon = 0.5.$

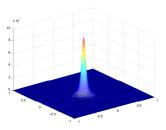




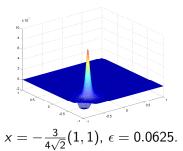
 $x = (0, 0), \epsilon = 0.25.$

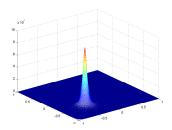


More plots of $K_1(x, y; \epsilon)$ for fixed x and ϵ as a function of y with the same h(t).

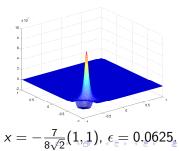


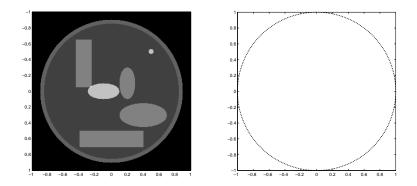
$$x = (0, 0), \epsilon = 0.0625.$$





 $x = -\frac{1}{2\sqrt{2}}(1,1)$, $\epsilon = 0.0625$.





Phantom and detectors.

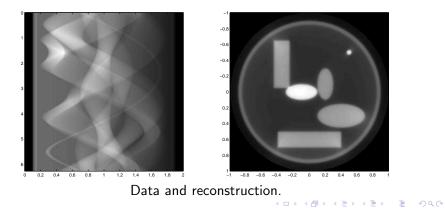
We use a discretization of

$$\int_{|y|<1} K_1(x,y;\epsilon)f(y)dy$$

= $\int_0^{2\pi} \left\{ \int_0^2 h_\epsilon(|x-u(\theta)|-r)\mathcal{M}f(u(\theta),r)rdr \right\} \frac{d\theta}{2\pi}.$

Reconstruction

$$\tilde{f}(x) = \frac{C}{MN} \sum_{j=1}^{N} \left\{ \sum_{i=1}^{M} h_{\epsilon}(|x - u_{\theta_{j}}| - r_{i})\mathcal{M}f(u_{\theta_{j}}, r_{i})r_{i} \right\}$$
$$h_{\epsilon}(t) = \frac{\epsilon^{2} - t^{2}}{(\epsilon^{2} + t^{2})^{2}}, \quad \epsilon = 0.01, \quad M = 299, \quad N = 300$$



These and similar numerical experiments suggest that $K_1(x, y; \epsilon)$ is a summability kernel and a good approximation of the identity at x for |x| < 1 as a function of y, |y| < 1, for sufficiently small ϵ .

Further numerical experiments suggest that the set of detectors Ξ need not be restricted to circles. For example

$$\mathcal{K}(x,y;\epsilon) = C \sum_{\xi_i \in \Xi} h_\epsilon(|x-\xi_i|-|y+\xi_i|)$$

will still be a good approximation of the identity at x for |x| < 1 as a function of y, |y| < 1, for appropriate ϵ as long as, roughly speaking, the set of detectors Ξ is a sufficiently dense set surrounding the unit disk. F. Filbir, R. Hielscher, and W.R. Madych, Reconstruction from circular and spherical mean data, *Applied and Computational Harmonic Analysis* 29, (2010), 111-120. Is it true that

$$\lim_{\epsilon \to 0} \int_{|y|<1} K_1(x,y;\epsilon) f(y) dy = f(x)$$

whenever f is bounded, vanishes outside the unit disk, and is continuous at x?

To hopefully simplify the matter try working with

$$\mathcal{K}_{2}(x,y;\epsilon) = \int_{0}^{2\pi} \frac{1}{\epsilon^{2}} h\Big(\frac{|x-u(\theta)|^{2} - |y-u(\theta)|^{2}}{2\epsilon}\Big) \frac{d\theta}{2\pi}$$

which is also a sum of radial functions in the variable y, seems pretty much like $K_1(x, y; \epsilon)$, but the argument $|x - u(\theta)|^2 - |y - u(\theta)|^2$ is algebraically easier to work with.

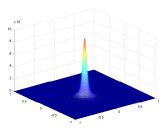
Note that $K_2(x, y; \epsilon)$ can be re-expressed as

$$K_2(x,y;\epsilon) = \int_0^{2\pi} \frac{1}{\epsilon^2} h\left(\left\langle \frac{x-y}{\epsilon}, u(\theta) - \frac{x+y}{2} \right\rangle\right) \frac{d\theta}{2\pi}$$

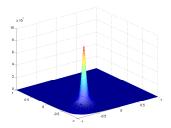
or

$$K_2(x,y;\epsilon) = \int_0^{2\pi} \frac{1}{\epsilon^2} h\Big(\Big\langle \frac{x-y}{\epsilon}, u(\theta) \Big\rangle + \frac{|y|^2 - |x|^2}{2} \Big) \frac{d\theta}{2\pi}.$$

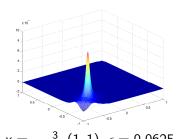
Plots of $K_2(x, y; \epsilon)$ for fixed x and ϵ as a function of y with h(t) as in (4).

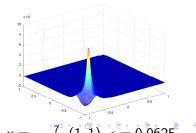


$$x = (0, 0), \epsilon = 0.0625.$$

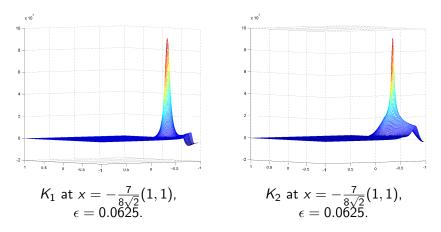


 $x = -\frac{1}{2\sqrt{2}}(1,1), \ \epsilon = 0.0625.$





Comparison of the plots of $K_1(x, y; \epsilon)$ and $K_2(x, y; \epsilon)$ for fixed x and ϵ as a functions of y.



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Theorem: If h is the function in example (4), that is

$$h(t) = rac{1}{2\pi} rac{1-t^2}{(1+t^2)^2},$$

then

$$\lim_{\epsilon \to 0} \int_{|y|<1} \kappa_2(x,y;\epsilon) f(y) dy = c(x) f(x)$$

where

$$c(x) = \frac{\pi}{1-|x|^2}$$

whenever f is bounded, vanishes outside the unit disk, and is continuous at x.

This is a corollary of the following:

Lemma: If h is the function in example (4) then

$$\left|\frac{1}{2\pi}\int_{0}^{2\pi}h(\langle z,u(heta)-x
angle)d heta
ight|\leqrac{C}{1+|z|^{3}}$$

for all $z \in \mathbb{R}^2$ where C is a constant that depends only on x when |x| < 1.

Proof of Lemma: Use residues to to evaluate the integral and get

$$\int_0^{2\pi} h(\langle z, u(\theta) - x \rangle) d\theta = c \operatorname{Re} \frac{\langle z, x \rangle + i}{\left((\langle z, x \rangle + i)^2 - |z|^2 \right)^{3/2}}.$$

Follow this by several pages of algebraic manipulations together with applications of appropriate inequalities to get the desired result.

Note 1: The analytic nature of h(t) and the change to $|x - \xi|^2 - |y - \xi|^2$ allowed us to do this.

Note 2: Want argument which depends only on the integrability of

$$K(x) = \int_0^{2\pi} h(\langle x, u(heta)
angle) d heta$$

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over R^2 .

Corollary 1 of Theorem: If x is in the unit disk B, and f is in $C^{2}(B)$ then

$$(6) \quad \frac{f(x)}{(1-|x|^2)} = \frac{-1}{\pi^2} \int_0^{2\pi} \left\{ \frac{1}{2} \int_0^{\sqrt{2}a} \left\{ M(r) + M\left(\sqrt{2a^2 - r^2}\right) - 2M(a) \right\} \frac{r}{\left(r^2 - a^2\right)^2} dr + \int_{\sqrt{2}a}^{\infty} M(r) \frac{r}{\left(r^2 - a^2\right)^2} dr - \frac{M(a)}{a^2} \right\} d\theta$$

where $M(r) = \mathcal{M}f(\xi, r)$ and $a = |x - u(\theta)|$.

Remark: f in $C^2(B)$ is overkill. I suspect that f in $L^p(B)$ for $p \ge p_0$ is sufficient.

Corollary 2 of Theorem: If Ω is the unit disk, x is in Ω , and f is in $C^2(\Omega)$ then f(x) =

$$\frac{1-|x|^2}{2\pi}\int_0^{2\pi}\left\{\int_0^2\log\left(\left|r^2-|x-u(\theta)|^2\right|\right)\frac{d}{dr}\left(\frac{1}{2r}\frac{d\mathcal{M}f(u(\theta),r)}{dr}\right)dr\right\}d\theta$$

Remark 1: Analogues of the Theorem and Corollaries are valid when the unit disk is replaced by a more general elliptical region and the set of detectors Ξ , currently a circle, is replaced by a more general ellipse. In this case $d\mu(\xi)$ is not the usual arclength.

Remark 2: The inversion formula of the Corollary should be compared with the inversion formula in FHR.

$$f(x) = \frac{1}{2\pi} \int_0^{2\pi} \int_0^2 \log(\left|r^2 - |x - u(\theta)|^2\right|) \frac{d}{dr} \left(r \frac{d\mathcal{M}f(u(\theta), r)}{dr}\right) dr d\theta.$$

Analogous results are valid for spherical mean transforms in higher dimensions n.

In fact, in the case n = 3 a stronger version of the Theorem is valid. The function h need not be analytic. It suffices that

$$K(x) = \frac{1}{4\pi} \int_{S^2} h(\langle x, u \rangle) d\sigma(u)$$

be integrable over \mathbb{R}^3 .