

EXPONENTIAL RADON TRANSFORM INVERSION BASED ON HARMONIC ANALYSIS OF THE EUCLIDEAN MOTION GROUP

Can Evren Yarman, Birsen Yazıcı

Rensselaer Polytechnic Institute,
Electrical, Computer and System Engineering
Troy, NY

ABSTRACT

This paper presents a new method for the exponential Radon transform inversion based on harmonic analysis of the Euclidean motion group $M(2)$. The exponential Radon transform is modified to be formulated as a convolution over $M(2)$. The convolution representation leads to a block diagonalization of the modified exponential Radon transform in the Euclidean motion group Fourier domain, which provides a deconvolution type inversion for the exponential Radon transform. Numerical examples are presented to show the viability of the proposed method.

1. INTRODUCTION

For a uniform attenuation coefficient $\mu \in \mathbb{C}$, the exponential Radon transform of a compactly supported real valued function f over \mathbb{R}^2 is defined as

$$\mathcal{T}_\mu f(\boldsymbol{\theta}, t) = \int_{\mathbb{R}^2} f(\mathbf{x}) \delta(\mathbf{x} \cdot \boldsymbol{\theta} - t) e^{\mu \mathbf{x} \cdot \boldsymbol{\theta}^\perp} d\mathbf{x}, \quad (1)$$

where $t \in \mathbb{R}$, $\boldsymbol{\theta} = (\cos \theta, \sin \theta)^T$ is a unit vector on S^1 with $\theta \in [0, 2\pi)$ and $\boldsymbol{\theta}^\perp = (-\sin \theta, \cos \theta)^T$.

The exponential Radon transform constitutes a mathematical model for imaging modalities such as x-ray tomography ($\mu = 0$), single photon emission tomography (SPECT) ($\mu \in \mathbb{R}$) [7], and optical polarization tomography of stress tensor field ($\mu \in i\mathbb{R}$) [8].

A number of different approaches have been proposed for the exponential Radon transform inversion. These can be classified into three categories: Fourier, filtered back projection and circular harmonic decomposition [1, 9, 3, 2, 4, 6] type inversion methods. A unified framework and a detailed review of the inversion methods can be found in [6].

This paper presents an alternative approach and provides a new inversion method for the exponential Radon transform based on the harmonic analysis of the Euclidean motion group, denoted by $M(2)$. This method of inversion leads to new algorithms for the inversion of the exponential Radon transform. The reconstructed images for μ real and μ imaginary are presented to show the viability of the proposed method.

2. MODIFIED EXPONENTIAL RADON TRANSFORM AS A CONVOLUTION OVER THE GROUP $M(2)$

The rigid motions of \mathbb{R}^2 form a group called the Euclidean motion group denoted by $M(2)$. The elements of the group are the 3×3

dimensional matrices of the form

$$(R_\theta, \mathbf{r}) = \begin{bmatrix} R_\theta & \mathbf{r} \\ \mathbf{0}^T & 1 \end{bmatrix}, \quad (2)$$

where $R_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \in SO(2)$ is the rotation component and $\mathbf{r} = (r_1, r_2)^T \in \mathbb{R}^2$ is the translation component. The group operation is the usual matrix multiplications and inverse of an element is obtained by matrix inversion as $(R_\theta, \mathbf{r})^{-1} = (R_\theta^{-1}, -R_\theta^{-1}\mathbf{r})$.

Let $g = (R_\theta, \mathbf{r})$ and $h = (R_\phi, \mathbf{x})$. Convolution over $M(N)$ is defined as

$$(f_1 *_{M(2)} f_2)(g) = \int_{M(2)} f_1(h) f_2(h^{-1}g) d(h) \quad (3)$$

where $\int_{M(2)} d(h) = \int_{SO(2)} \int_{\mathbb{R}^2} d\mathbf{x} d(\phi)$. Then, multiplying the exponential Radon transform of a real valued function f with $e^{\mu r_2}$, $r_2 \in \mathbb{R}$, the resulting integral can be expressed as a convolution operation over $M(2)$ as follows:

$$\begin{aligned} \mathcal{T}'_\mu f(g) &= e^{\mu r_2} \mathcal{T}_\mu f(\boldsymbol{\vartheta}, -r_1) \\ &= \int_{\mathbb{R}^2} f(\mathbf{x}) \delta(\mathbf{x} \cdot \boldsymbol{\vartheta} + r_1) e^{\mu \mathbf{x} \cdot \boldsymbol{\vartheta}^\perp + \mu r_2} d\mathbf{x} \\ &= (\Lambda_\mu *_{M(2)} f^*)(g), \end{aligned} \quad (4)$$

where $f^*(h) = \overline{f(h^{-1})}$ and $\Lambda_\mu(h) = \delta(\mathbf{x} \cdot \mathbf{e}_1) e^{\mu \mathbf{x} \cdot \mathbf{e}_2}$. Λ_μ will be called the convolution filter.

We shall use the Fourier transform over the group $M(2)$ to express the modified exponential Radon transform, $\mathcal{T}'_\mu f$, as a multiplication in the $M(2)$ -Fourier domain.

3. $M(2)$ -FOURIER TRANSFORM

Let $f \in L^2(M(2))$. The $M(2)$ -Fourier transform of f is defined as

$$\mathcal{F}_{M(2)}(f)_{mn}(\lambda) = \widehat{f}_{mn}(\lambda) = \int_{M(2)} f(g) u_{mn}^{(\lambda)}(g^{-1}) d(g), \quad (5)$$

for $\lambda \geq 0$, and the corresponding inverse $M(2)$ -Fourier transform is given by

$$\mathcal{F}_{M(2)}^{-1}(\widehat{f}_{mn})(g) = f(g) = \int_0^\infty \sum_{m,n} \widehat{f}_{mn}(\lambda) u_{nm}^{(\lambda)}(g) \lambda d\lambda, \quad (6)$$

where $u_{mn}^{(\lambda)}(g)$ are the matrix elements of the irreducible unitary representations of $M(2)$ given by [10]

$$u_{mn}^{(\lambda)}(g) = \frac{1}{2\pi} \int_0^{2\pi} e^{-im\omega} e^{-i\lambda(r_1 \cos \omega + r_2 \sin \omega)} e^{in(\omega - \theta)} d\omega. \quad (7)$$

Let $f, f_1, f_2 \in L^2(M(2))$. Then, $M(2)$ -Fourier transform satisfies the following properties:

1. Adjoint property:

$$\widehat{f^*}_{mn}(\lambda) = \overline{\widehat{f}_{nm}(\lambda)}, \quad (8)$$

where $f^*(g) = \overline{f(g^{-1})}$.

2. Convolution property:

$$\mathcal{F}(f_1 * f_2)_{mn}(\lambda) = \sum_q \widehat{f}_{2mq}(\lambda) \widehat{f}_{1qn}(\lambda). \quad (9)$$

3. If f is an $SO(2)$ invariant function over $M(2)$, i.e. $f(g) = f(\mathbf{r}) \in L^2(\mathbb{R}^2)$, then

$$\widehat{f}_{mn}(\lambda) = \delta_m \widetilde{f}_n(-\lambda) \quad (10)$$

where δ_m is the Kronecker delta function.

4. INVERSION OF EXPONENTIAL RADON TRANSFORM USING $M(2)$ -FOURIER TRANSFORM

Treating the projections $\mathcal{T}'f_\mu$ and the filter Λ_μ as distributions over $M(2)$, the modified exponential Radon transform can be expressed as a multiplication in the $M(2)$ -Fourier transform domain. Using the convolution property of the $M(2)$ -Fourier transform, (4) becomes:

$$\widehat{\mathcal{T}'f}_{mn}(\lambda) = \sum_q \overline{\widehat{f}_{qm}(\lambda)} \widehat{\Lambda}_{\mu qn}(\lambda) = \overline{\widehat{f}_{0m}(\lambda)} \widehat{\Lambda}_{\mu 0n}(\lambda), \quad (11)$$

for $m, n \in \mathbb{Z}$. Equation 11 provides a block diagonal representation of the modified exponential Radon transform in the $M(2)$ -Fourier domain, where each block is parameterized by $\lambda \geq 0$. It also provides an immediate inversion formula for the exponential Radon transform:

$$f = \mathcal{F}_{M(2)}^{-1} \left(\sum_k \left[\widehat{\mathcal{T}'f}_{nk} \left[\widehat{\Lambda}_\mu \right]_{km}^{-1} \right]^\dagger \right). \quad (12)$$

Since $\widehat{\Lambda}_\mu$ is rank one, as long as $\widehat{\Lambda}_{\mu 0n}(\lambda) \neq 0$, by (11), the $M(2)$ -Fourier coefficients of f is given by

$$\widehat{f}_{0m}(\lambda) = \left(\frac{\widehat{\mathcal{T}'f}_{mn}(\lambda)}{\widehat{\Lambda}_{\mu 0n}(\lambda)} \right). \quad (13)$$

Thus, the inversion formula for the exponential Radon transform becomes:

$$\begin{aligned} f(\mathbf{x}) &= \mathcal{F}_{M(2)}^{-1} \left(\delta_k \widetilde{f}_{-m}(-\lambda) \right) = \mathcal{F}^{-1} \left(\delta_k \left(\frac{\widehat{\mathcal{T}'f}_{mn}(\lambda)}{\widehat{\Lambda}_{\mu 0n}(\lambda)} \right) \right) \\ &= \int_0^\infty \sum_m \left(\frac{\widehat{\mathcal{T}'f}_{mn}(\lambda)}{\widehat{\Lambda}_{\mu 0n}(\lambda)} \right) u_{m0}^{(\lambda)}(h) \lambda^{N-1} d\lambda. \end{aligned} \quad (14)$$

5. RECONSTRUCTION ALGORITHMS

Two algorithms are proposed for exponential Radon transform inversion. The algorithms differ mainly in the way the $M(2)$ -Fourier coefficients, $\widehat{f}_{0m}(\lambda)$, of the function are computed (step 3 of the reconstruction). In the first algorithm, (14) is used to compute $\widehat{f}_{0m}(\lambda)$, whereas in the second algorithm, (12) is used. Let $f(\mathbf{x}) = 0$ for $|\mathbf{x}| > a$ and hence $\mathcal{T}_\mu f(\theta, r_1) = 0$ for $|r_1| > a$. Proposed exponential Radon transform inversion can be implemented in four steps:

Step 1. Extend $\mathcal{T}_\mu f(\theta, -r_1)$ to $\mathcal{T}'_\mu f(g)$ by multiplying with $e^{\mu r_2}$ for $r_2 \in [-a, a]$, where $0 < a < \infty$.

Step 2. Compute $\widehat{\mathcal{T}'_\mu f}_{mn}(\lambda)$, the $M(2)$ -Fourier transform of $\mathcal{T}'_\mu f$ for $m, n = 0, \pm 1, \dots, \pm S - 1$, and $\lambda_k = \frac{k\lambda_0}{S}$, $k = 0, \dots, S - 1$ for some $\lambda_0 > 0$.

Step 3. Compute $\widetilde{f}_{-m}(-\lambda)$ by either of the following ways:

Algorithm 1. For each λ , let $[\widehat{\Lambda}_0(\lambda)]$ and $[\widehat{\mathcal{T}'_\mu f}_m(\lambda)]$ denote the row vectors with their elements given by $\widehat{\Lambda}_{0n}(\lambda)$ and $\widehat{\mathcal{T}'_\mu f}_{mn}(\lambda)$, respectively. For each m , compute $\widetilde{f}_{-m}(-\lambda)$ by

$$\widetilde{f}_{-m}(-\lambda) = \frac{[\widehat{\mathcal{T}'_\mu f}_m(\lambda)][\widehat{\Lambda}_0(\lambda)]^T}{[\widehat{\Lambda}_0(\lambda)][\widehat{\Lambda}_0(\lambda)]^T + \sigma}, \quad (15)$$

where σ is a positive constant close to zero.

Algorithm 2. For each λ , let $[\widehat{\Lambda}(\lambda)]$, $[\widehat{\mathcal{T}'_\mu f}(\lambda)]$ and $[\widehat{f}(\lambda)]$ denote the matrices with their corresponding elements given by $\delta_m \widehat{\Lambda}_{0n}(\lambda)$, $\widehat{\mathcal{T}'_\mu f}_{mn}(\lambda)$ and $\delta_m \widetilde{f}_{-n}(-\lambda)$, where m and n denote the row and column numbers, respectively. Then,

$$[\widehat{f}(\lambda)] = [\widehat{\Lambda}(\lambda)] \left([\widehat{\Lambda}(\lambda)]^T [\widehat{\Lambda}(\lambda)] + \sigma I \right)^{-1} [\widehat{\mathcal{T}'_\mu f}(\lambda)]^T, \quad (16)$$

where σ is a positive constant close to zero. Note that a generalization of *Algorithm 2* for $\mu = 0$ was presented in our earlier work [11].

Step 4. Using Equation 10, form $\widehat{f}_{mn}(\lambda)$ and take the inverse $M(2)$ -Fourier transform to obtain f .

In step 3, $\widehat{\Lambda}_{\mu 0n}$ can be very close to zero, making the inverse filtering numerically unstable. Therefore, the inverse filter is replaced with its regularized linear least square version to stabilize the inversion.

A fast implementation of the $M(2)$ -Fourier transform is implemented as described in [5, 11].

6. NUMERICAL SIMULATIONS

Numerical simulations are performed on a two-dimensional modified Shepp-Logan phantom image of $13.1 \times 13.1 \text{cm}^2$, discretized by 129×129 pixels. The $M(2)$ -Fourier transform was numerically implemented as described above. All numerical implementations were performed using MATLAB. Figure 1 and 2 present the reconstructed images using the proposed algorithms for $\mu = 0.154 \text{cm}^{-1}$ and $\mu = i0.154 \text{cm}^{-1}$, respectively. In all reconstructions, σ is set to 10^{-10} . The case for $\mu = 0$ was studied previously in [11, 12]. These simulations demonstrate the viability of the proposed method and algorithms.

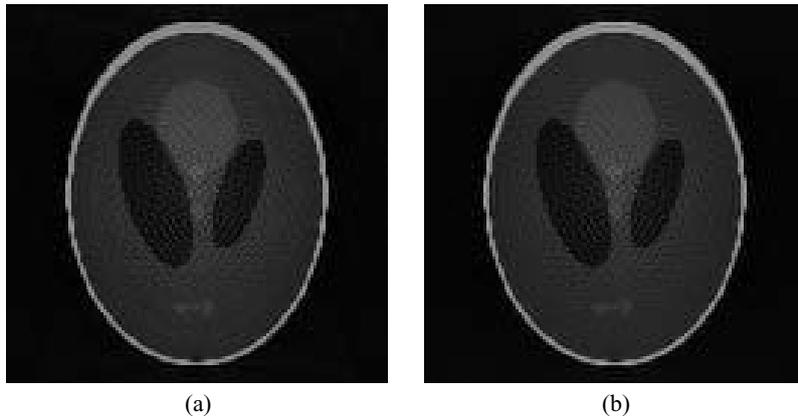


Fig. 1. Reconstruction of the modified Shepp-Logan phantom using the proposed algorithms for $\mu = 0.154\text{cm}^{-1}$. (a) *Algorithm 1*, $\sigma = 10^{-10}$. (b) *Algorithm 2*, $\sigma = 10^{-10}$.

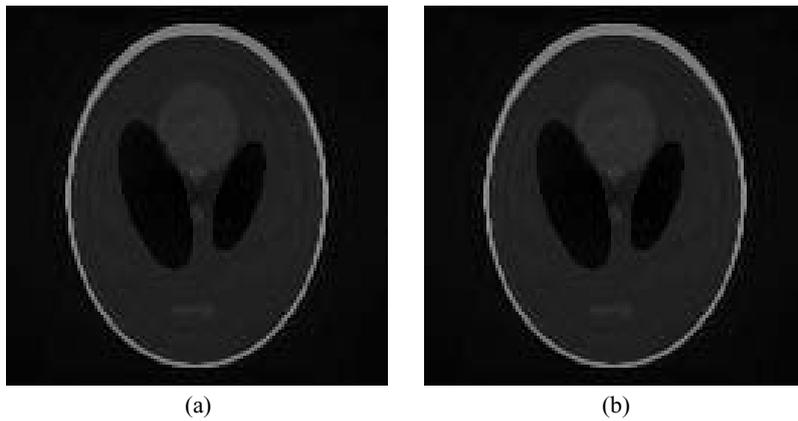


Fig. 2. Reconstruction of the modified Shepp-Logan phantom using the proposed algorithms for $\mu = i0.154\text{cm}^{-1}$. (a) *Algorithm 1*, $\sigma = 10^{-10}$. (b) *Algorithm 2*, $\sigma = 10^{-10}$.

7. CONCLUSION

In this paper, a block diagonal form of the modified exponential Radon transform in the $M(2)$ -Fourier domain is derived and a new method of inversion for the exponential Radon transform using the $M(2)$ -Fourier transform is introduced. Numerically stable algorithms are introduced. Viability of proposed method and algorithms is demonstrated by numerical simulations.

8. REFERENCES

- [1] Bellini S, Piancentini M, Cafforio C, Rocca F 1979 *IEEE Transactions on Acoustics, Speech, and Signal Processing ASSP-27* 213–218
- [2] Hawkins G W, Leichner P K, Yang, N-C 1988 *IEEE Transactions on Medical Imaging* 7 135–148
- [3] Inouye T, Kose K, Hasegawa A 1989 *Phys. Med. Biol.* **34** 299–304
- [4] Kuchment P, Shneiberg I 1994 *Applicable Analysis* **53** 221–231
- [5] Kyatkin A B, Chirikjian G S 2000 *Applied Computational Harmonic Analysis* **9** 220–241
- [6] Metz C E, Pan X 1995 *IEEE Transactions on Medical Imaging* **14** 643–658
- [7] Natterer F 1986 *The Mathematics of Computerized Tomography* (New York: Wiley-Teubner)
- [8] Puro, A 2001 *Inverse Problems* **17** 179–1888
- [9] Tretiak O, Metz C 1980 *SIAM J. Appl. Math.* **39** 341–354
- [10] Vilenkin N J 1988 *Special functions and the theory of representations* (Providence: American Mathematical Society)
- [11] C.E. Yarman, B. Yazıcı, “Radon transform inversion via Wiener filtering over the Euclidean motion group,” *Proceedings of IEEE international Conference on Image Processing, vol 2*, Barcelona, Spain, pp. 811–814, 2003.
- [12] C.E. Yarman, B. Yazıcı, 2004 “Radon transform inversion based on harmonic analysis of the motion group,” (preprint)