

# Numerical Study of Novel Reconstruction Artifacts from Limited Sonar Tomographic Data



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## Introduction

- ▶ Two-dimensional sonar tomography uses indirect sound wave data to image the structural boundaries of an underwater object by inverting the circular **Radon transform** over a set of object density data.
- ▶ For a **limited data set**, e.g., in the case of object obstruction during imaging, visual artifacts arise from points on the boundary of the object due to the ill-posedness of the inverse integral operator, which makes it difficult to image the object.<sup>1</sup>
- ▶ Yet, there exist other physical settings in which morphological changes to the limited data set could incur noisy artifacts that have not yet been explained microlocally,
- ▶ We thus focus on finding, characterizing, and reducing novel reconstruction artifacts from limited 2-D sonar data; this will inform the field of tomography in a microlocal context and could have implications for modern inverse imaging technologies.

## Limited Data Set Generation

- ▶ Consider the ocean  $y > 0$  and a source of sound waves as a transceiver on the ocean surface at a point  $(x, 0)$ , where  $x$  is the transceiver center.
- ▶ The time distance to an object is the radius  $r$  of the circular wave that is reflected by the object. This admits a set  $C(x, r)$  of concentric circular waves transmitted at  $(x, 0)$  for all centers and radii  $x, r \in \mathbb{R}$ .
- ▶ We let the piecewise smooth object reflectivity function  $f(x, y)$  determine the points at which the waves are reflected, namely the boundaries of the object:

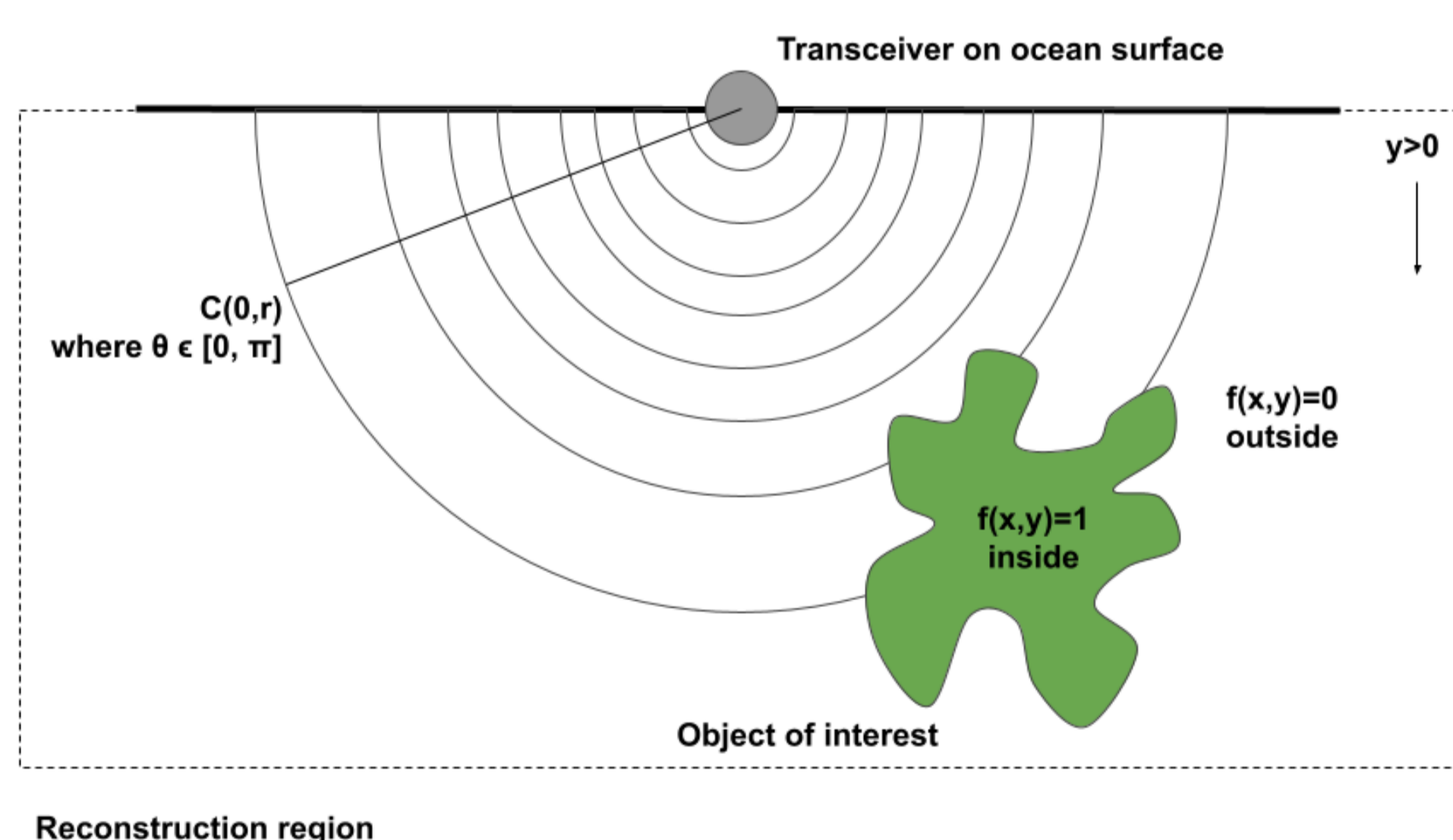


Figure 1: Graphical representation for transceivers with centers  $(x,0)$ .

- ▶ Thus, the wave intensity detected by the transceiver is approximated by the **Sonar transform**, defined parametrically<sup>2</sup>:

$$Rf(x, r) = \int_{C(x,r)} f(x, y) ds = \int_{\theta=0}^{\pi} f(x + r \cos \theta, r \sin \theta) d\theta.$$

- ▶ Only a compact subset of  $C(x, r)$  is considered such that  $(x, r) \in [-T, T] \times [0, R]$ , imposing a closed and bounded reconstruction region around the object. This mimics the canonical limited data case.

## Object Reconstruction

- ▶ Consider every point  $\bar{x} = (x_1, x_2)$  in the limited reconstruction region and define the **Sonar back-projection** as the inverse operator that recovers the object's properties  $f(x, y)$  for all such points:

$$R^*g(\bar{x}) = \int_{x=-T}^T g(x, \sqrt{(x_1 - x)^2 + x_2^2}) dx.$$

- ▶ We thus represent  $R^*Rf(\bar{x})$  as the distribution of values of  $f(x, y)$  over the entire reconstruction region, imaging the object with visual noise.
- ▶ It has been shown that reconstruction artifacts occur when circles  $C(x, r)$  that have  $(x, r) \in bd(Rf(x, r))$  are tangent to singular points on  $bd(f(x, y))$ ; the artifacts visually spread along such circles.<sup>1,3</sup>
- ▶ Similarly, a point on  $bd(f(x, y))$  will only be visible if there are any circles tangent to that point.

## Numerical Methods and Experiments

- ▶ To simulate the limited data set, we tabulate  $R^*Rf(\bar{x})$  into a data matrix and approximate integrals using trapezoidal rule.
- ▶ Since  $f$  is a piecewise smooth function, we apply a 3-point second order central difference  $D$  to accentuate boundaries and visual artifacts.

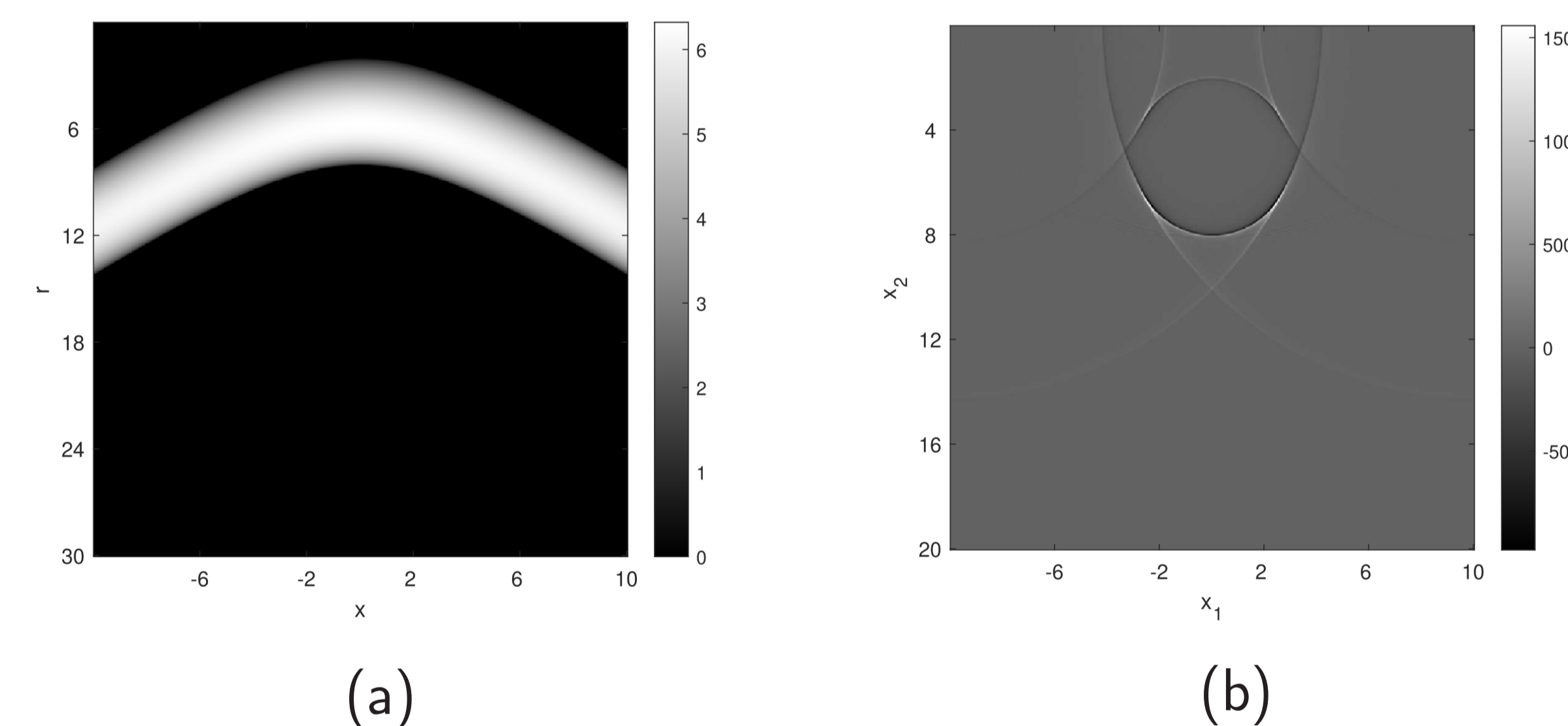


Figure 2: (a) Data set for a disk centered at  $(0,5)$  with radius 3 and (b) its subsequent filtered object reconstruction, generated using  $R^*DRf(\bar{x})$ .

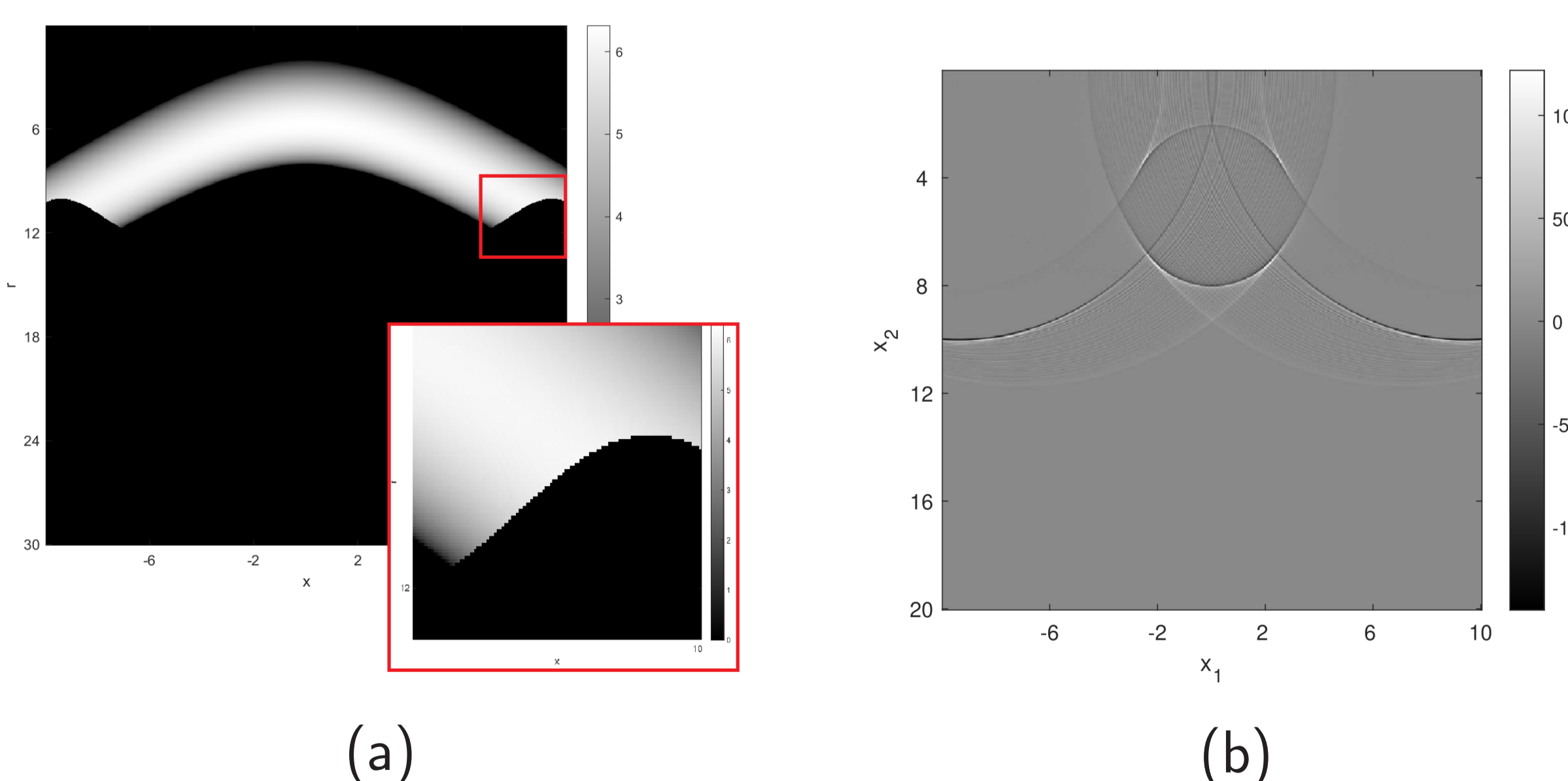


Figure 3: (a) Data set for a disk truncated by  $\cos(x)$  with non-smooth points on  $bd(Rf(x, r))$  and (b) its image with semi-concentric artifacts independent of the disk.

## Artifact Reduction

- ▶ A common artifact reduction method is multiplying the data with a smooth cutoff function  $\psi$  with compact support.<sup>3</sup> Yet, this conflates the data and does not localize to  $bd(Rf(x, r))$ .
- ▶ So, we convolve over a scanning window on the data matrix and smooth out jump discontinuities on  $bd(Rf(x, r))$  based on some tolerance.
- ▶ For some data set  $F$  and a function  $\phi$  symmetric about 0 such that  $\sum_{i=-n}^n \phi(i) = 1$ , the **n-point discrete convolution** is defined as:

$$C_\phi g(x, r) := (g * \phi)(x, r) = \sum_{j=-n}^n g(x, r - j) \phi(j).$$

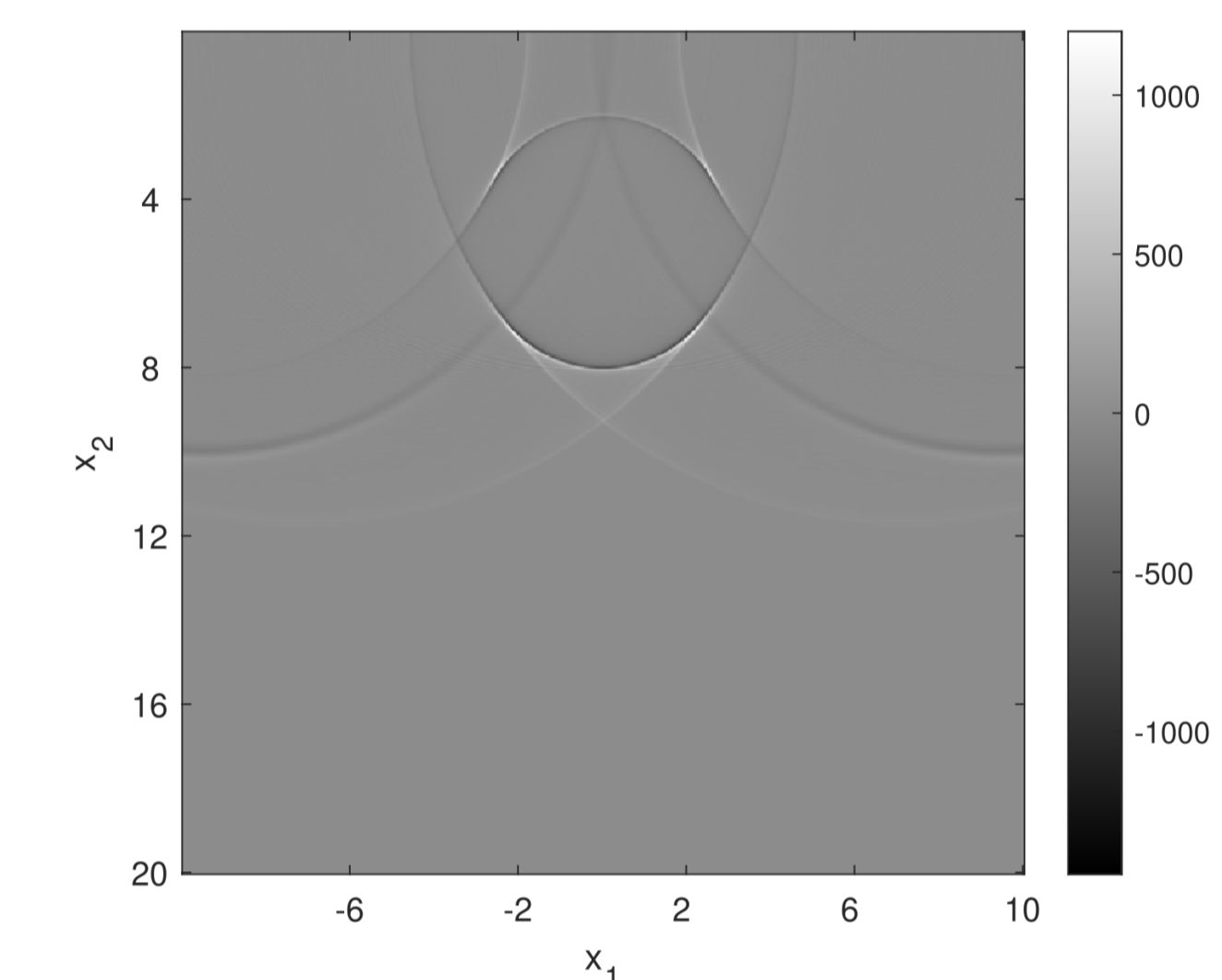


Figure 4: Disk smoothed by 7-point convolution  $R^*DC_\phi Rf(\bar{x})$  with dissipated artifacts.

## Conclusions and Future Directions

- ▶ We find novel reconstruction artifacts that occur from non-smooth points on  $bd(Rf(x, r))$  independent of  $bd(f(x, y))$ . We develop a convolution algorithm that minimizes these and any canonical artifacts.
- ▶ We aim to apply this research to microlocal theory, other limited data scenarios, and other general imaging problems.

## References

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