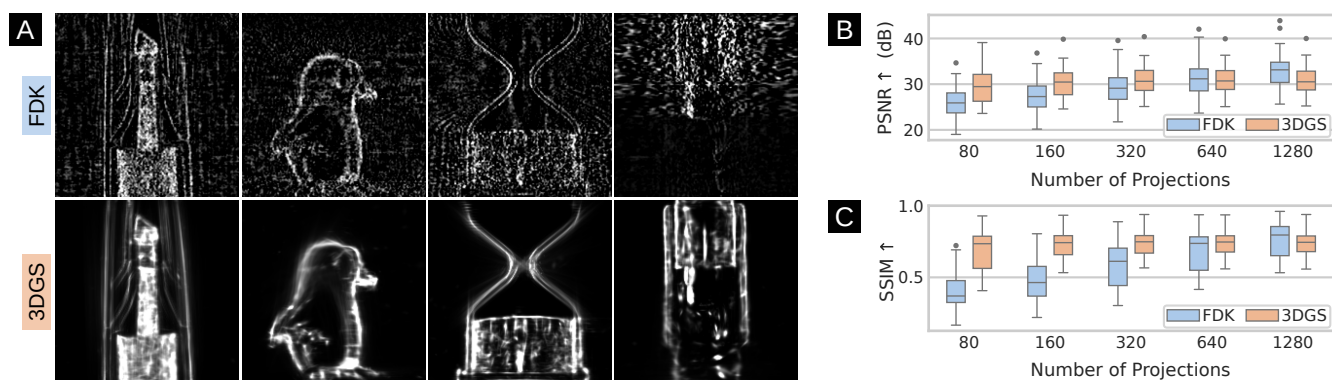


# ROBUST SPARSE-VIEW DARK-FIELD CT WITH 3D GAUSSIAN SPLATTING

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**Fig. 1.** Lateral slices from 80-view reconstructions (A) demonstrate the robustness of 3DGS to sparse inputs, while FDK suffers from high-frequency noise and streak artifacts. 3DGS consistently recovers smooth, continuous structures and maintains geometric integrity across 17 different samples evaluated with PSNR (B) and SSIM (C) on isotropic reconstruction volumes.

## 1. INTRODUCTION

Streak artifacts constitute a major challenge in X-ray dark-field computed tomography (DFCT). Convolutional neural networks have shown promising results for streak reduction, but they rely on dedicated ground truth data with limited availability [1]. In contrast to post-processing approaches, this work aims to prevent streak formation at the reconstruction stage using 3D Gaussian splatting (3DGS) [2].

## 2. METHODS

3DGS was implemented using  $R^2$ -Gaussian, as described in [2], and sparsely acquired dark-field projections reconstructed in comparison to Feldkamp–Davis–Kress (FDK). Reconstruction quality was assessed based on peak signal-to-noise ratio (PSNR) and structural similarity index measure (SSIM) in reference to full-view FDK (2324 projections).

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## 3. RESULTS AND CONCLUSION

Visual and quantitative evaluation highlight the superior performance of 3DGS compared to FDK, as illustrated in Fig. 1. Due to imperfections in the full-view FDK reference, metrics can deviate from the visual impression. SSIM provides a more reliable assessment owing to its larger noise resilience compared to PSNR. Our findings underscore the potential of 3DGS to enable streak-free DFCT reconstruction.

## 4. REFERENCES

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