

# Back-projection filtration image reconstruction approach for reducing high-density object artifacts in digital breast tomosynthesis

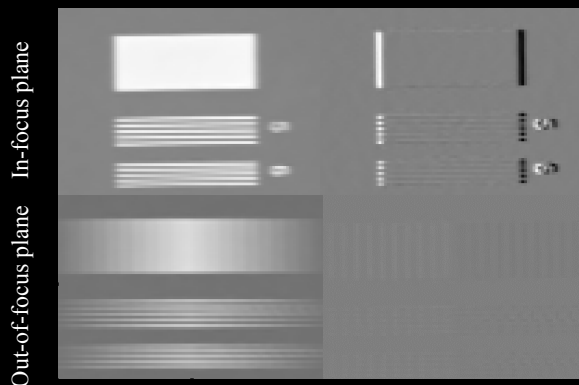
Presenting author: Hyeongseok Kim, Date: 07/31/2018, Time: 08:20 AM, Session title: Breast Imaging, Location: Room 202

## Innovation

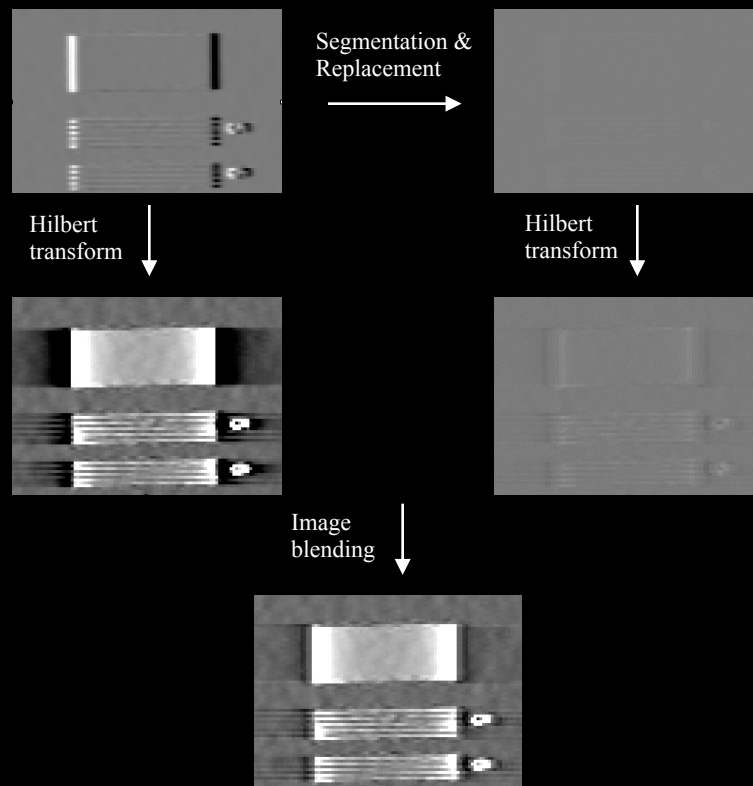
## Key Results

### Ripple artifacts reduction

BP image      DBP image

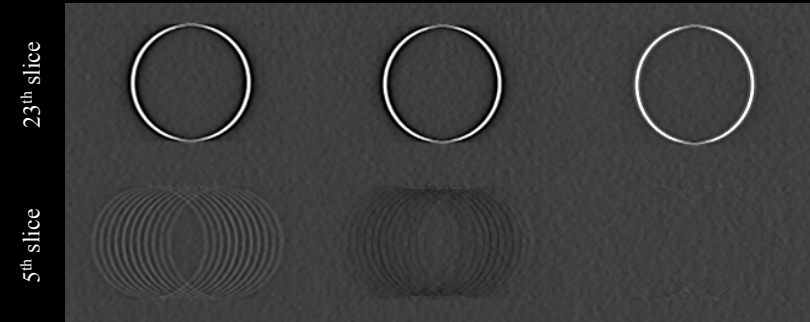


### Undershoot artifacts reduction

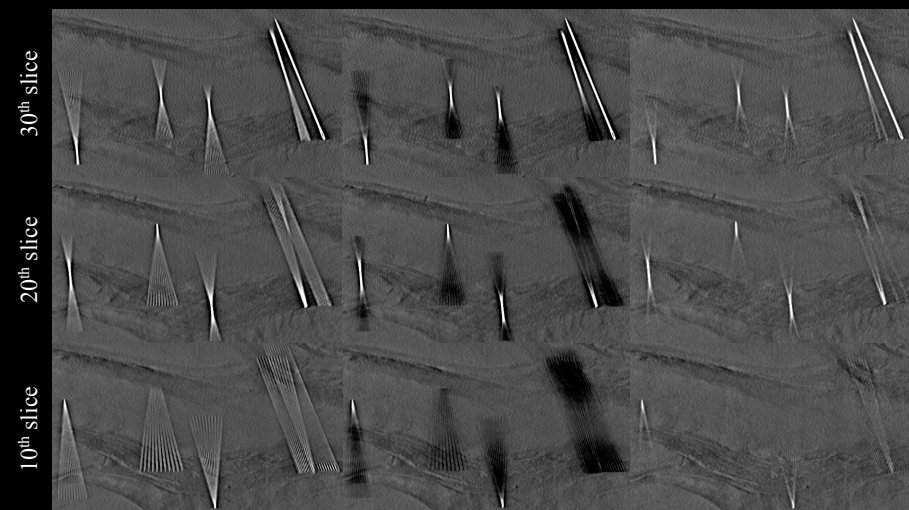


\* DBP images before segmentation (left top) and after segmentation and replacement (right top). Hilbert transformed images in the middle row. Final image in the bottom-most. We blended the two images in the middle row. to generate final image. In blending procedure, larger values were taken between the two images.

Conventional FBP      wFBP      Proposed method



Conventional FBP      wFBP      Proposed method



\* Conventional FBP is the filtered back-projection algorithm that is modified for DBT reconstruction, and wFBP is the FBP algorithm involving weighted back-projection for ripple artifacts reduction.

$$\mu_j = \frac{\sum_n (I_{differentiated}(i_n, j))}{N}$$

$$\sigma_j^2 = \frac{\sum_n (I_{differentiated}(i_n, j) - \mu_j)^2}{N}$$

$$w(i_n, j) = \frac{1}{\sigma_j \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{I_{differentiated}(i_n, j) - \mu_j}{\sigma_j} \right)^2}$$

$$V_w(j) = \frac{\sum_n w(i_n, j) * (I_{differentiated}(i_n, j))}{\sum_n w(i_n, j)}$$

where  $I_{differentiated}(i_n, j)$  is the back-projected value from  $i$ -th pixel in  $n$ -th differentiated projection to  $j$ -th voxel in image,  $\mu_j$  is mean,  $\sigma_j^2$  is standard deviation,  $w(i_n, j)$  is weights, and  $V_w(j)$  is weighted sum of back-projected value of each reconstructed voxel.