

## Radial Undersampling that is Variable in $k_z$

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**Introduction:** 3D radial imaging, with projections in the x-y plane, and slice-encodings in z (i.e. “stack of stars”), can be accelerated with undersampling, with maintained spatial resolution and acceptable artifacts. To date, undersampling has been applied uniformly in  $kz$  (1-4) (Fig. 1B). This investigation compares a uniformly undersampled acquisition to a radial undersampling pattern which varies in  $kz$ -dimension, acquiring a greater number of projections for central  $kz$ -space (Fig. 1A).

**Theory:** For undersampled radial, only high spatial frequencies are undersampled. The critical  $k$ -space radius in the  $kx$ - $ky$  plane at which undersampling starts (the Nyquist limit, here called  $Nyq_{kr}$ ) is given by Eq. 1 for a given FOV and number of projections ( $N_p$ ) (5). We sought to develop a trajectory for which this critical radius is at a constant distance from the  $k$ -space origin, in 3D (Eq. 2). This requires that  $N_p$  vary with  $kz$  value, i.e. according to slice-encoding value (Fig. 1A). Figure 1 demonstrates that using a variable number of projections,  $Nyq_{kr}$  varies with  $kz$ , allowing smaller  $Nyq_{kr}$  for larger  $kz$ , as in Eq. 2, and shown as red contours in Fig. 1b. In Fig. 1b the  $kz$  value (green vertical arrow), and  $Nyq_{kr}(kz)$  (green horizontal arrow) determine the distance indicated by the blue arrows, indicating the absolute distance from the  $k$ -space origin at which undersampling begins for  $kz=0$ , and for non zero  $kz$ .

$$Nyq_{kr} = N_p \frac{1}{\pi \cdot FOV} \quad (1)$$

$$[Nyq_{kr}(kz)]^2 + [kz]^2 = [Nyq_{kr}(0)]^2 \quad (2)$$

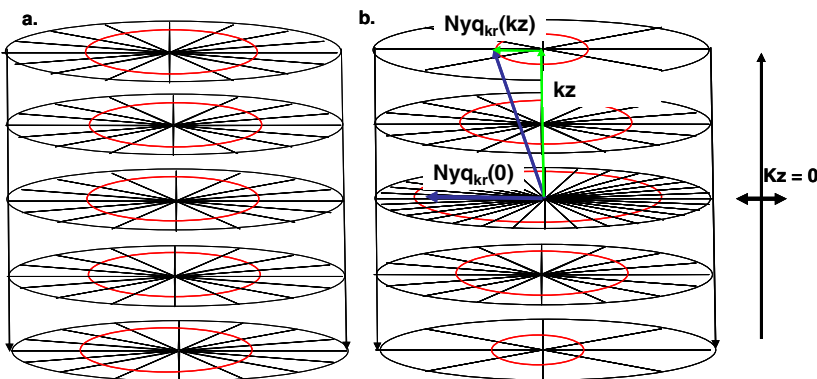


Figure 1: A) Uniform undersampling in  $kz$ . B) Variable undersampling in  $kz$ , by acquiring more projections for central  $kz$ -space, shown by the black sampling pattern. The red contours show the radius beyond which  $k$ -space is undersampled,  $Nyq_{kr}(kz)$ , which is dependent on  $N_p$  by Eq. 1. The blue arrows indicate the distance from the  $k$ -space origin to this critical radius, which depends on  $kz$  and  $Nyq_{kr}(kz)$ , as stated in Eq. 2.

**Methods:** All imaging was performed on a 1.5 T Philips Intera Achieva. Phantom data was acquired on the scanner using uniform ( $N_p=128$ ) and variable undersampling (128, 64, 32 for central, middle and edge  $kz$ -space data) with scan parameters: 3D radial SSFP, 256  $N_r \times 60 N_z$ , 1.2 x 1.2 x 2 mm. In healthy subjects, a 3D radial SSFP acquisition covered the whole left ventricle in short-axis. Scan parameters were: 160  $N_r$ , 320 cm FOV, 13  $N_z$  10 mm slices, TR/TE/ $\theta$ = 3.1ms/1.5ms/40°, 10 views per heart-beat. Fully sampled images ( $N_p=160$ ), uniformly undersampled images ( $N_p=38$ ), and variably undersampled images were acquired. The variably undersampled images were reconstructed offline and  $kz$ -space data was weighted according to the undersampling level. The simulations, performed in

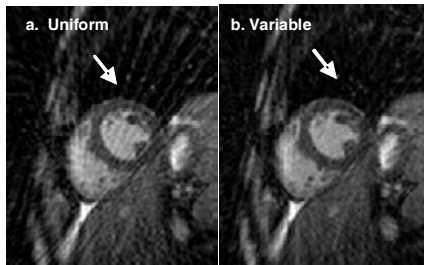


Figure 2: Simulation results for a) uniform ( $N_p=37$ ) and b) variable undersampling with equal total  $N_p$ . Note the reduced artifacts in b (arrows).

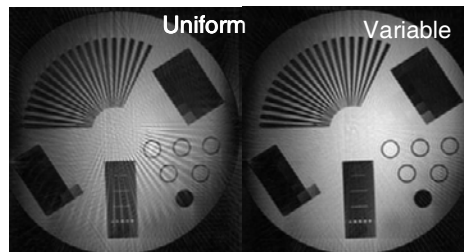


Figure 3: Phantom comparison of uniform and variable undersampling with equal scan time.

using this pattern. Figure 3 compares images of the phantom acquired in equal scan time, using variable and uniform undersampling, showing reduced artifacts.

**Conclusions:** This variable undersampling provides reduced artifacts compared to uniform undersampling for some applications. In the future, its application to contrast enhanced MR angiography will be investigated. **References:** 1) KK Vigen et al, MRM. 2000 Feb;43(2):170-6. 2) DC Peters et al, MRM. 2000 Jan;43(1):91-101. 3) AC Larson et al. MRM 2002 Oct;48(4):594-601. 4) E. Spuentrup, Radiology. 2004 May;231(2):581-6. 5) DC Peters et al. MRM. 2001 Apr;45(4):562-7.

Matlab, used the fully sampled data set as input, and compared variable undersampling ( $N_p$  varying from 20 to 48) to a uniform undersampling with identical number of total projections ( $N_p=37$ ).

**Results:** Figure 2 shows the results of simulations for a 13 slice 3D cardiac function acquisition. The optimal sampling pattern varied sinusoidally from 20 to 48 to 20  $N_p$  dependent on