

# PROSPECTIVE SNR OPTIMIZATION IN $k$ - $T$ -BASED SENSITIVITY-ENCODED DYNAMIC IMAGING USING A FAST GEOMETRIC ALGORITHM

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**Introduction** MR data acquisition and image reconstruction are typically formulated using Fourier transform (FT) theory, i.e., the classical  $k$ -space relationship. However, in dynamic MR imaging (e.g., 2D Cartesian sampling) it takes a few milliseconds from collecting one  $k$ -space phase-encode (PE) line to the next, i.e., sampling in the PE direction is time-sequential and not instantaneous [1]. Hence, a more accurate model for MR data acquisition is the so called “ $k$ - $t$ -space perspective” ( $k$ =spatial frequency,  $t$ =time) [1,2]. Most well known  $k$ - $t$ -based schemes use  $k$ - $t$  sampling patterns that are restricted to a lattice (generalized sheared grid) [1-3]. Based on sampling theory,  $k$ - $t$  sampling on a lattice will result in aliasing of the object’s spectrum in the reciprocal domain, referred to as dual  $k$ - $t$  domain [4-6] or the  $x$ - $f$ -space ( $x$ =PE direction,  $f$ =temporal frequency) [3]. The replication pattern is on the corresponding point-spread function (PSF) which is the FT of the sampling lattice [2,4]. The acceleration in  $k$ - $t$ -based techniques results from multifold undersampling that the  $k$ - $t$  lattice provides relative to the Nyquist rate. In UNFOLD [3] and PARADIGM [4], the  $k$ - $t$  lattice is designed so that there is no overlap of the support region in  $x$ - $f$ -space. On the contrary, in UNFOLD-based parallel imaging (e.g., TSENSE),  $k$ - $t$  SENSE, and PARADISE [5-6], there is overlap in  $x$ - $f$ -space which can potentially be undone using sensitivity encoding (SE) and prior information (signal model). The prior information is in form of statistics for  $k$ - $t$  SENSE, cross-shaped  $x$ - $f$  support model (Fig 1a) for UNFOLD [3], and multi-banded patient-adaptive  $x$ - $f$  support model (Fig 1b) for PARADISE [4-6]. Optimizing the reconstruction SNR in this context has been previously introduced and its significance has been demonstrated *in-silico* for PARADISE [5]. A similar study has been conducted for  $k$ - $t$  SENSE [7]. We focus on *nongated* cardiac MR techniques that use  $x$ - $f$  support models and SE, namely, PARADISE [5-6] or UNFOLD-based parallel imaging. We propose a novel algorithm for designing the SNR-optimal  $k$ - $t$  sampling pattern and study its performance *in-vivo*. The proposed algorithm is geometric, i.e., it is based on geometry of overlap patterns in  $x$ - $f$  space and does not require any knowledge of coil sensitivity profiles.

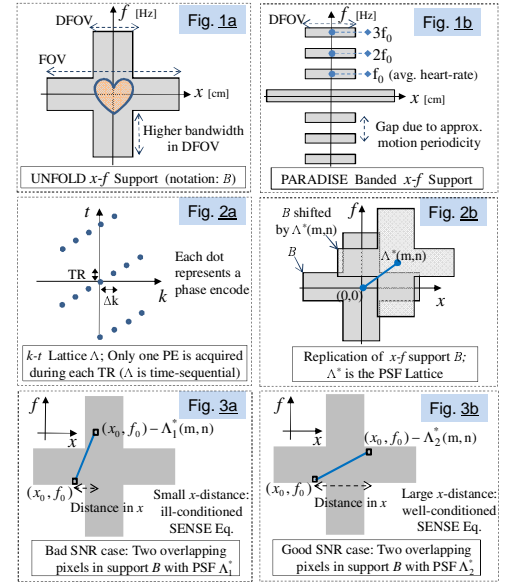
**Theory** Fig 1a depicts a  $x$ - $f$  support model  $B$  similar to the one used in UNFOLD [3] characterized by a dynamic FOV (DFOV) and a temporal bandwidth that is varying in  $x$ . Fig 1b shows a more refined  $x$ - $f$  support used in PARADISE [6], referred to as the banded support model [4], which formulates the approximate heart motion periodicity (hence the gap between bands) and also models heart-rate (HR) variability [6]. Given a fixed spatial resolution, a  $k$ - $t$  lattice  $\Lambda$  shown in Fig 2a is parameterized by  $(TR, \Delta k)$  and can be expressed as sum of delta functions located on a sheared grid [2,5]. The PSF for  $\Lambda$ , denoted by  $\Lambda^*$  is given by its 2D FT (inverse FT in  $k$  and forward in  $t$ ). Fig 2b depicts two representative replicas of the  $x$ - $f$  support (Fig 1a), one centered at the origin and another on a grid point  $\Lambda^*(m,n)$  of the PSF lattice  $\Lambda^*$ . As a result of this replication, pixels within  $B$  overlap with each other and hence need to be un-mixed. The reconstruction task is to recover all pixels within support  $B$  by un-mixing the overlapped pixels in  $x$ - $f$ -space. This overlap pattern is depicted in Figs 3a and 3b for two PSF lattices. In parallel  $k$ - $t$ -based techniques that only use  $x$ - $f$  support information (unlike  $k$ - $t$  SENSE), this procedure relies on the SE along  $x$  (no encoding along  $f$ ) provided by multiple receiver channels. If two overlapping pixels have close  $x$ -coordinates, as shown in Fig 3a, then they are experiencing almost similar SE. Hence, the underlying SENSE equations in  $x$ - $f$ -space would be closely dependent which will result in an ill-conditioned matrix equation for un-mixing of the overlapped pixels. Therefore, compared to Fig 3a, the overlap pattern in Fig 3b is expected to have a better conditioning (lower spatio-temporal g-factor [5,7]) and result in a better SNR. Motivated by this observation, for a pixel  $(x_0, f_0)$  in  $B$ , we denote by  $p(x_0, f_0)$  the minimum  $x$ -coordinate distance between the set of pixels that overlap onto  $(x_0, f_0)$ . Let  $\rho(TR, \Delta k, B)$  denote the minimum of  $p(x, f)$  among all pixels within  $B$  resulting from sampling on  $\Lambda$  parameterized by  $(TR, \Delta k)$  (Fig 2a). We postulate that by designing  $\Lambda$  so that  $\rho(TR, \Delta k, B)$  is maximized, the resulting overlap patterns will correspond to well-conditioned (low g-factor) SENSE matrix equations for  $x$ - $f$ -space reconstruction. Eq. (1) provides a functional form for the optimization problem. Fig 4 summarizes the computational steps for evaluating the cost function for a set of feasible  $(TR, \Delta k)$  choices.

**Methods** MR imaging with informed consent was performed under the NHLBI IRB using a 1.5T Siemens Avanto scanner with a 32-element cardiac array. Initially, a gated segmented SSFP cine scan was acquired ( $|FOV|=420$ mm square, matrix= $256 \times 224$ , 30 phases,  $TR=3.5$ ms, temporal resolution= $28$ ms, GRAPPA rate 4). A customized SSFP pulse sequence was developed to allow for operator-defined  $TR$  and ordering of phase-encodes -- hence capable of acquiring a general  $k$ - $t$  lattice (Fig 2a). MR data for the non-gated PARADISE scheme [5] with a  $320 \times 256$  image matrix ( $1.3 \times 1.6$ mm resolution) was collected during a single breathhold (scan time= $16.4$ s). The  $x$ - $f$  support model was chosen to be the banded model [5] as in Fig 1b with 11 bands (DC centered) and the following parameters: (1) DFOV location was estimated from localizer scans ( $|DFOV|=0.32*|FOV|$ ) (2) Average HR during the gated scan was used as an estimate for  $f_0$  ( $=1.1$ Hz) (3) The thickness of the bands (Fig 1b) was set to  $0.33*f_0$  to account for HR mis-estimation and variability during the scan. The proposed  $k$ - $t$  lattice design algorithm (Fig 4) was run to find the SNR-optimal lattice by maximizing  $\rho$  (result:  $\rho^*=0.24*|FOV|$ ). For comparison, a suboptimal lattice was computed so that its  $\rho$  would satisfy  $\rho \approx \rho^*/3$  (result:  $\rho=0.09*|FOV|$ ). The search space for  $(TR, \Delta k)$  was limited according to the acquisition and SSFP pulse sequence specifications (resolution, min  $TR$ , etc). Computation time was less than 30s and each  $k$ - $t$  lattice was immediately preceded by a conventional coil-profile calibration acquisition (128 Nyquist-spaced phase-encodes) to provide the final  $k$ - $t$  sampling schedule.

**Discussion & Conclusion** Fig 5 shows the end-diastolic (ED) frame for the gated cine. Reconstructed ED frames for PARADISE acquired (non-gated) according to the SNR-optimized and suboptimal  $k$ - $t$  lattices are shown in Figs 6a and 6b, respectively. The optimized lattice acquisition results in a much better reconstruction SNR (Fig 6b) because its underlying SENSE equations (in  $x$ - $f$  space [5]) are better conditioned (have lower g-factor). In conclusion, the SNR difference seen in Fig 6 demonstrates the effectiveness and significance of the proposed  $k$ - $t$  lattice design technique. In contrast to previous  $k$ - $t$  acquisition design techniques [5-7], all computations in the proposed algorithm are geometric-based and independent of coil sensitivities. This feature enables fast computation of the SNR-optimal  $k$ - $t$  sampling schedule for  $k$ - $t$ -based parallel imaging prior to running the MR scan.

**References** [1] Liang, Lauterbur, IEEE Med. Imaging 13:677-86, 1994 [2] Willis, Bresler, IEEE Info. Theory 43:190-207, 1997 [3] Tsao, MRM 47:202-07, 2002 [4] Aggarwal *et al*, Inverse Problems (24) 2008 [5] Sharif, Bresler, Proc IEEE ISBI’07, 1020-23 [6] Sharif *et al*, ISMRM’07(15), p151 [7] Malik *et al*, ISMRM’08(16), p11

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$$(TR^*, \Delta k^*) = \arg \max_{(TR, \Delta k) \text{ feasible}} \rho(TR, \Delta k, B) \quad \text{Eq. (1)}$$

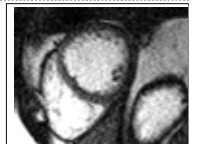
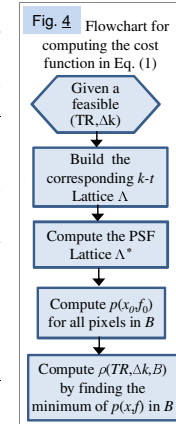


Fig. 5 Gated Cine ED Frame; 256x224 matrix;

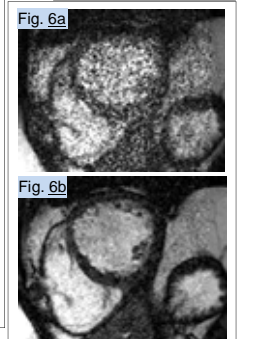


Fig. 6 PARADISE Reconstruction; Non-gated; 320x256 matrix; (a) Suboptimal; (b) Optimized