Quantifying exchange using optimized small-tip fast recovery (STFR) sequences

Naveen Murthy, Jeffrey A. Fessler, Nicole Sieberlich

University of Michigan

MRI Workshop on Acquisition and Reconstruction

September 9, 2021



 Exchange captures inter-compartmental interactions (exchange of water + electrolytes)

Figure: 2-D exchange map from MRF-X¹

²Harkins, Kevin D., et al. "Effect of intercompartmental water exchange on the apparent myelin water fraction in multiexponential T2 measurements of rat spinal cord." MRM 67.3 (2012): 793-800.

¹Hamilton, Jesse, et al. "MR fingerprinting with chemical exchange (MRF-X) for in vivo multi-compartment relaxation and exchange rate mapping." ISMRM 2016.



- Exchange captures inter-compartmental interactions (exchange of water + electrolytes)
- Exchange can be a useful clinical biomarker²

Figure: 2-D exchange map from MRF-X¹

¹Hamilton, Jesse, et al. "MR fingerprinting with chemical exchange (MRF-X) for in vivo multi-compartment relaxation and exchange rate mapping." ISMRM 2016.

²Harkins, Kevin D., et al. "Effect of intercompartmental water exchange on the apparent myelin water fraction in multiexponential T2 measurements of rat spinal cord." MRM 67.3 (2012): 793-800.

Quantifying exchange using STFR



Figure: 2-D exchange map from MRF-X¹

 Exchange captures inter-compartmental interactions (exchange of water + electrolytes)

- Exchange can be a useful clinical biomarker²
- Hard to estimate exchange in vivo currently

²Harkins, Kevin D., et al. "Effect of intercompartmental water exchange on the apparent myelin water fraction in multiexponential T2 measurements of rat spinal cord." MRM 67.3 (2012): 793-800.

¹Hamilton, Jesse, et al. "MR fingerprinting with chemical exchange (MRF-X) for in vivo multi-compartment relaxation and exchange rate mapping." ISMRM 2016.



 $\Delta \omega_{\mathrm{f}}$: Myelin-specific frequency shift

Two pool exchanging model



Objectives

- Obtain a scan design that is able to estimate exchange with high precision
- Quantify exchange rates using an appropriate estimator

 $[\]Delta \omega_{\mathrm{f}}$: Myelin-specific frequency shift

Two pool exchanging model



Setup

$$\bullet f_{\rm f} + f_{\rm s} = 1$$

• Well-mixed:
$$f_f/f_s = \tau_{f \rightarrow s}/\tau_{s \rightarrow f}$$

▶ 8 unknowns: $(M_0, f_f, T_{1f}, T_{1s}, T_{2f}, T_{2s}, \tau_{f \rightarrow s}, \Delta \omega_f)$

 $\Delta \omega_{\mathrm{f}}$: Myelin-specific frequency shift

Scan design

- Cramer-Rao Bound (CRB) based optimization (weighted wrt exchange)
- Pulse sequence: Small-tip Fast Recovery (STFR) sequence³

³Sun, Hao, et al. "Steady-state functional MRI using spoiled small-tip fast recovery imaging." *Magnetic resonance in medicine* 73.2 (2015): 536-543.

High-level overview

Scan design

- Cramer-Rao Bound (CRB) based optimization (weighted wrt exchange)
- Pulse sequence: Small-tip Fast Recovery (STFR) sequence³

Estimation

- Parameter Estimation via Regression with Kernels (PERK)⁴
- Fast dictionary-free parameter estimation method
- Lifts measurements to a higher-dimensional space followed by kernel regression

³Sun, Hao, et al. "Steady-state functional MRI using spoiled small-tip fast recovery imaging." *Magnetic resonance in medicine* 73.2 (2015): 536-543.

⁴Nataraj, Gopal, et al. "Dictionary-free MRI PERK: Parameter estimation via regression with kernels." IEEE transactions on medical imaging 37.9 (2018): 2103-2114.

Small-tip Fast Recovery (STFR)



 Goal: Design a set of D STFR scans.

Small-tip Fast Recovery (STFR)



- Goal: Design a set of D STFR scans.
- Scan parameters: $(T_{\text{free}}, T_{\text{g}}, \alpha, \beta, \phi, T_{\text{E}})_d,$

where d = 1, 2, ... D.

Small-tip Fast Recovery (STFR)



- Goal: Design a set of D STFR scans.
- Scan parameters: $(T_{\text{free}}, T_{\text{g}}, \alpha, \beta, \phi, T_{\text{E}})_d$,

where
$$d = 1, 2, ... D$$
.

Hypothesis

Optimized STFR sequences are sensitive to $\Delta \omega_{\rm f},$ which might lead to more precise exchange estimates.

 $\Delta \omega_{\mathrm{f}}$: Myelin-specific frequency shift

Setup:

D = 20 scans

• max FA =
$$40^{\circ}$$
, range of T_{free} = 8-20ms

Scan design

Setup:

- D = 20 scans
- max FA = 40° , range of T_{free} = 8-20ms
- B0 and B1+ maps are computed through separate SPGR and Bloch-Siegert scans
- Total scan time: ~21mins for 10 slices

Optimized STFR design



 Coeff. of variation is 20% for exchange estimation in WM (vs. 93% for initial design).

Simulations



(a) Exchange map - ground truth (in ms) ⁵

Parameter	White Matter	Gray Matter
Mo	0.77	0.86
ff	0.15	0.03
7 _{1,f} (ms)	400	500
T _{1,s} (ms)	832	1331
T _{2,f} (ms)	20	20
T _{2,s} (ms)	80	80
$\Delta \omega_{\rm f}$ (Hz)	15	5
$\Delta \omega$ (Hz)	Varies	Varies
κ	Varies	Varies

(b) Nominal values for WM and GM at 3T $^{\rm 6}$

⁶Collins, D. Louis, et al. "Design and construction of a realistic digital brain phantom." *IEEE TMI* 1998.

-250 -225 -200 -175 -150 -125 -100 -75

⁷Whitaker, Steven T., et al. "Myelin water fraction estimation using small-tip fast recovery MRI." MRM 2020.

M University of Michigan

Quantifying exchange using STFR

Results









Ground truth

Initial rand design (RMSE in WM: 38ms)

Opt STFR design (RMSE in WM: 23ms)

Results



(RMSE in WM: 38ms)

(RMSE in WM: 23ms)

Opt STFR design has an RMSE of 15% in WM (vs. 25% for initial design).

Results



- Opt STFR design has an RMSE of 15% in WM (vs. 25% for initial design).
- Performs as worse as a mean estimator in GM (caveat: very little myelin content in GM)

Invivo data (in progress)



Acknowledgements for data: Melissa Haskell

Conclusions:

- Obtained an optimized STFR design that has ~20% coefficient of variation for estimating exchange in WM.
- Optimized design performs better in WM than a random design (by ~10%); does as worse as a mean estimator in GM.

Conclusions

Conclusions:

- Obtained an optimized STFR design that has ~20% coefficient of variation for estimating exchange in WM.
- Optimized design performs better in WM than a random design (by ~10%); does as worse as a mean estimator in GM.

Next steps:

- Investigate the performance of the estimator in the presence of off-resonance.
- Investigate how the opt design does for invivo data; compare with other state-of-the-art methods.