






PROTOCOLS, METHODS, AND RESOURCES

Impact of pulse sequence, analysis method, and signal to noise ratio on the accuracy of intervertebral disc T_2 measurement

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Abstract

Noninvasive assessments of intervertebral disc health and degeneration are critical for addressing disc degeneration and low back pain. Magnetic resonance imaging (MRI) is exceptionally sensitive to tissue with high water content, and measurement of the MR transverse relaxation time, T_2 , has been applied as a quantitative, continuous, and objective measure of disc degeneration that is linked to the water and matrix composition of the disc. However, T_2 measurement is susceptible to inaccuracies due to Rician noise, T_1 contamination, and stimulated echo effects. These error generators can all be controlled for with proper data collection and fitting methods. The objective of this study was to identify sequence parameters to appropriately acquire MR data and to establish curve fitting methods to accurately calculate disc T_2 in the presence of noise by correcting for Rician noise. To do so, we compared T_2 calculated from the typical monoexponential (MONO) fits and noise corrected exponential (NCEXP) fits. We examined how the selected sequence parameters altered the calculated T_2 in silico and in vivo. Typical MONO fits were frequently poor due to Rician noise, and NCEXP fits were more likely to provide accurate T_2 calculations. NCEXP is particularly less biased and less uncertain at low SNR. This study showed that the NCEXP using sequences with data from 20 echoes out to echo times of ~ 300 ms is the best method for calculating T_2 of discs. By acquiring signal data out to longer echo times and accounting for Rician noise, the curve fitting is more robust in calculating T_2 despite the noise in the data. This is particularly important when considering degenerate discs or AF tissue because the SNR of these regions is lower.

KEYWORDS

degenerative disc disease, intervertebral disc, MRI, Rician noise, T_2

1 | INTRODUCTION

Noninvasive assessment of intervertebral disc degeneration is critical for addressing low back pain, for evaluating treatment efficacy in

patients, and for evaluating preclinical animal models of disc disorders. Magnetic resonance imaging (MRI) is exceptionally sensitive to tissue with high water content. For this reason, MRI is widely used for the disc with grading schemes based on structure and signal intensity

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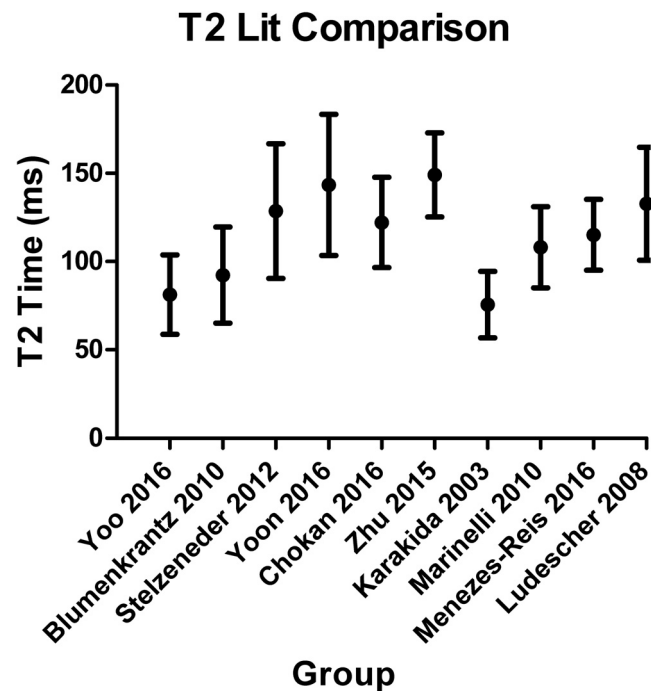


FIGURE 1 Reported T_2 for healthy discs from volunteers without back pain. Despite sampling a similar population, reported mean T_2 vary by up to a factor of two across studies, ranging from a mean of 76 to 149 ms. Dots and bars represent mean \pm SD

TABLE 1 Sequence parameters used to acquire data for T_2 calculations shown in Figure 1

Citation	TR (ms)	First TE (ms)	Last TE (ms)	# echoes	Average T_2 (ms)
Yoo, 2016	1836	6	38.1	4	81.30
Blumenkrantz, 2010	Not reported	9.6	77.2	7	92.30
Stelzeneder, 2012	1200	13.8	82.8	6	128.60
Yoon, 2016	120	9.9	89.1	9	143.47
Chokan, 2016	2000	13	103	8	122.10
Zhu, 2015	2000	13.9	111	8	149.10
Karakida, 2003	2000	30	120	4	75.56
Marinelli, 2010	2000	9	144	16	108.00
Menezes-Reis, 2016	3000	20	160	8	115.10
Ludescher, 2008	3000	9	288	32	132.80
Recommended from present study	≥ 3000		≥ 300		

Note: There is a large range of sequence parameters used, which lead to wide variations in reported T_2 .

from T_2 -weighted (T_2w) images. The contrast provided by T_2w MRI is particularly well-suited for structural evaluation because it provides contrast between a bright nucleus pulposus (NP), a dark annulus fibrosus (AF), and a dark vertebral body at the inferior and superior boundaries of the disc. In the Pfirrmann grading scheme and others, structural features are evaluated and graded based on NP brightness and uniformity, NP-AF distinction, and disc height.¹ Unfortunately, such evaluation is highly subjective and is nonquantitative in terms of degree of pathology. Further, the phenotypes of human disc degeneration are a continuum that are too complex to be categorized by the five Pfirrmann grades. For these reasons, measurement of the MR

transverse relaxation time, T_2 , has been proposed and applied as a quantitative, continuous, and objective measure of disc degeneration that is linked to the water and matrix composition of the disc. T_2 is defined as the time constant of the decay, or relaxation, of the transverse signal and is calculated from a series of T_2w images.^{2,3} T_2 is longer in healthy discs, which have higher water content and water mobility, and decreases as the disc degrades and loses proteoglycan and water content.

T_2 measurement is becoming more widely used as a measure of disc degeneration,⁴⁻¹⁵ however, there are limitations in its application that can make comparisons across studies problematic and, in the

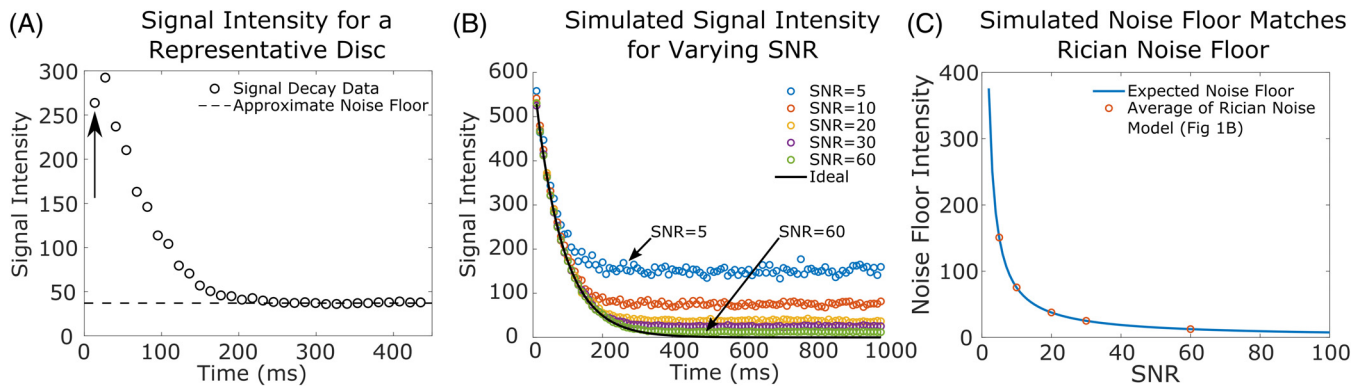


FIGURE 2 (A), Signal intensity as a function of echo time for a representative disc, where each data point represents a signal intensity measured at a TE . It can be observed that the signal decays to a nonzero noise floor (dashed line). Note that the first echo (arrow) has lower signal because of the stimulated echo effect affecting all subsequent echoes. (B), The effect of the amount of SNR on the noise floor for simulated Rician noise at varying SNR levels. Without noise (solid line) the signal intensity decays to 0, but as the SNR decreases, the noise floor level increases. (C), The expected noise curve vs SNR. The noise floor for each SNR level in (B) is shown and matches the expected curve almost exactly

worst case, provide inaccurate T_2 values. First, a wide variety of MR sequences and calculation methods have been implemented, leading to a large range of reported average T_2 from 75 to 150 ms for similar populations of young, healthy, nondegenerative discs (Table 1, Figure 1). This is likely due in large part to the variability in the sequence parameters (TR , TE , and number of echoes) which are sometimes outside the recommended range based on disc material parameters (T_1 and T_2), as described in the next section. In addition, greater accuracy is often obtained by excluding the first echo in data fits,¹⁶ as seen in some cartilage studies.¹⁷ The first echo is excluded because it is the only data point for which the phenomenon of stimulated echoes does not occur, and therefore it follows a different decay than the subsequent data points^{16,18} (Figure 2A). It is assumed that most published work has excluded this first echo, but it is not always explicitly stated and researchers new to the field may not be aware of this limitation, which is ultimately based on the unavoidable inhomogeneity of RF pulse amplitude throughout the sample. Finally, MR imaging is susceptible to noise that can corrupt fits of the T_2 decay curve. Notably, Rician noise results in an altered signal decay curve that decays to a nonzero value called the “noise floor” (Figure 2A), which causes error in T_2 calculation because the fit equation assumes a monoexponential signal decay to zero. The impact of this noise is that the calculated T_2 , which is a material parameter, can be inaccurate. This effect is influenced by the signal to noise ratio (SNR) and whether the number of echoes acquired is sufficient so that the noise floor is approached during signal decay. This effect can be addressed through careful design of data acquisition and modeling of noise characteristics as will be shown in this study.

A noise-corrected exponential (NCEXP) has successfully accounted for Rician Noise in cartilage and has been used to calculate accurate T_2 ^{19,20}; however, it has not been applied to the disc. It is critical to address the effect of the Rician noise in the disc because the disc signal decays to low SNR, in consideration of the Rician distribution of noise important. Moreover, the disc loses hydration and signal

intensity with degeneration, which will decrease the initial SNR and exacerbate the role of Rician noise. Therefore, the objective of this study is to identify appropriate sequence parameters to acquire MR data and to establish curve fitting methods to accurately calculate disc T_2 in the presence of noise by correcting for Rician noise. To do so, we compared T_2 calculated from the typical monoexponential (MONO) fits and noise corrected exponential (NCEXP) fits and examined how the selected sequence parameters altered the calculated T_2 . We also evaluated how T_2 is affected by performing fits of intensity data averaged over a region-of-interest (ROI) to suppress noise vs performing fits at each individual voxel. Based on these results, we recommend methods to select sequence parameters and to calculate T_2 for the disc to address the effect of Rician noise and the low signal intensity at long echo times, particularly for the degenerating disc.

2 | THEORY

Determining T_2 requires a longer and more complex imaging sequence than acquisition of the single T_2w image that is used for grading. However, the advantage of this sequence is that it gives quantitative material property information that is less susceptible than single T_2w signal intensity to environmental factors, scanner strength, magnetic field inhomogeneities, and subject traits such as body weight. The signal intensity in each voxel in a conventional MR image is often modeled as

$$\text{Signal intensity} \propto [H] \cdot \left(1 - e^{-TR/T_1}\right) \cdot e^{-TE/T_2} \quad (1)$$

where the repetition time, TR , is the time between individual spin excitation pulses, and the echo time, TE , is the time of echo occurrence after the initial excitation pulse. TR and TE are user-defined input parameters, while the T_1 and T_2 relaxation times are material

properties, with T_1 being the spin-lattice, or longitudinal, relaxation time; H is the number of protons in a voxel. Although Equation (1) omits the dependence of signal intensity on a number of nonmodeled effects (e.g., pulse errors, diffusion, chemical exchange, non-monoexponential relaxation behavior), it correctly describes the dominant dependences on TR and TE in the noise-free case. Clinical MR images are formed from the absolute magnitude of the signal (i.e., signal intensity being the square root of the sum of the squares of the signal acquired in real and imaginary channels); this is a crucial concept for the understanding of Rician noise as described below. See References 21, 22 for an overview of MRI fundamentals and definitions.

The TR selected for the sequence is the dominant determinant of the total scan time. In T_2 w MRI, TR is often selected to minimize T_1 weighting in the acquired signal (i.e., with TR satisfying $e^{-TR/T_1} \rightarrow 0$), so that residual image intensity weighting is primarily dependent on T_2 , or more precisely, on the ratio TE/T_2 . Ideally, TR would be $\geq T_1 * 5$ in order to achieve $<1\%$ T_1 weighting, however since disc T_1 is approximately 1200 ms,²³ this would lead to very lengthy scans. Thus, the selection of TR is a compromise between scan duration and the desired limit $e^{-TR/T_1} \rightarrow 0$.

The echo time in Equation (1), TE , defines the delay between an initial excitation pulse and signal acquisition at the peak of the subsequent echo, with a refocusing pulse applied between these at a time $TE/2$ after spin excitation. To determine T_2 , signal intensity is measured across a series of echo times, which are conventionally multiples of a minimum TE . This means that samples are obtained at times $n * TE$ following excitation, with n ranging from one to some large value defining the number of echoes acquired. From Equation (1), it is clear that signal amplitude will decrease as an exponential function of TE , with time constant T_2 . Note that this assumes a monoexponential signal equation, which is commonly used when curve fitting signal decay data to determine T_2 in disc. The minimum TE used should be small enough to permit signal acquisition from rapidly relaxing tissue, and the number of echoes should be large enough to permit near complete signal decay during the echo train. We chose 85% signal decay as a criterion, as further decay would generally result in signal at the noise floor. Since disc T_2 is approximately 150 ms in healthy NP, the maximum TE should be at least 300 ms, to capture 85% of the signal decay. Because TR is much longer than TE , multi-echo MRI sequences can collect many TE s without any cost to overall imaging time.

Rician noise causes an alteration in the shape of the signal decay along with causing the signal to decay to a nonzero value. This Rician noise can be modeled to calculate more accurate T_2 values. Rician noise occurs in the MR signal magnitude due to the Gaussian noise characteristics of the real and imaginary signals acquired to compute the magnitude image. The Rician probability distribution is given by

$$p_M(M) = \frac{M}{\sigma^2} e^{-\frac{M^2 + S^2}{2\sigma^2}} I_0\left(\frac{S \cdot M}{\sigma^2}\right) \quad (2)$$

where M is the measured signal intensity, S is the signal intensity without noise, σ is the SD of the Gaussian noise in the real and imaginary components, and I_0 is the modified zeroth order Bessel function of

the first kind.²⁴ A key characteristic of the Rician distribution is that at high SNR it approximates a Gaussian distribution. However, when the SNR is low, the noise associated with the signal is no longer Gaussian because taking the magnitude of real and imaginary components is a nonlinear function. At the extreme of SNR = 0 (no tissue signal, only noise signal) the measured MR signal takes on a nonzero mean value (Rayleigh distribution) that we call the noise floor (dashed line in Figure 2A).^{20,24} This phenomenon was previously studied with simulation and articular cartilage samples using a noise corrected exponential (NCEXP) that fit the signal while incorporating the expected value of the Rician noise, thereby more accurately determining T_2 .¹⁹ The expected noise corrected signal intensity is given by

$$\text{Signal intensity}(S, \sigma) = \sqrt{\frac{\pi \sigma^2}{2}} e^{-\alpha} ((1 + 2\alpha) I_0(\alpha) + 2\alpha I_1(\alpha)) \quad (3)$$

where $\alpha = (S/2\sigma)^2$, and I_1 is the modified first order Bessel function of the first kind.

3 | METHODS AND RESULTS

3.1 | Simulation of Rician noise and dependence on SNR

We first confirmed that simulated Rician noise recapitulated the noise floor observed in the disc (Figure 2A) and that we could model the signal intensity of the noise floor as a function of the SNR.

Using MATLAB (MathWorks), an ideal monoexponential decay was generated with an initial signal of 600 (S_0) and T_2 of 80 ms.

$$S(TE) = S_0 * e^{-TE/T_2} \quad (4)$$

Rician noise was simulated by adding Gaussian noise to both the real and imaginary components of the ideal monoexponential decay. The resulting signal intensity, calculated as the magnitude of the real and imaginary components, was averaged for 100 simulated voxels to mimic the size of an ROI in an in vivo measurement. The σ of the Gaussian noise was given by S_0/SNR , and this simulation was repeated for SNR ranging from 5 to 60. The resulting signal was plotted over time (every 10 ms out to 1000 ms, that is, $TE = 10$ ms with 100 echoes) and compared qualitatively to the observed signal intensity decay and noise floor from an in vivo disc (Figure 2A).

The noise floor of the simulated signal was calculated as the average signal of the last 50 points for each simulated SNR (Figure 2B). These values were then plotted as a function of SNR for the 5 SNR values used (Figure 2C, circles). The expected relationship between noise floor and SNR (mean of Rayleigh distribution) is also plotted (Figure 2C, curve), based on the following equation

$$\text{Expected noise floor} = \sigma * \sqrt{\pi/2} = \frac{S_0}{\text{SNR}} * \sqrt{\pi/2} \quad (5)$$

where the initial signal, S_0 , was equal to 600.

Our simulation of Rician noise as a function of SNR confirms the noise floor observed in the disc (Figure 2A,B) and the expected noise floor as derived from MR physics (Figure 2C). The signal intensity with simulated Rician noise (Gaussian noise applied to the real and imaginary components) demonstrated a nonzero noise floor that depends on the SNR (Figure 2B) and is consistent with in vivo disc imaging data (Figure 2A). The signal intensity of the noise floor is greater with a low SNR of 5 and is negligible with a high SNR of 60 (Figure 2B). Moreover, when the noise floors of the simulated signal intensity values are compared to the curve of the expected noise floor, a perfect overlap is observed that demonstrates the accuracy of our Rician noise simulation (Figure 2C).

3.2 | Simulation of T_2 sensitivity to fitting model and number of echoes

Signal intensity data with simulated noise was used to demonstrate the sensitivity of the calculated T_2 to (a) the degree of noise (low vs high SNR), (b) the model used to calculate T_2 (monoexponential or noise corrected exponential), and (c) the effect of the number of echoes in the dataset (simulating the number of TE chosen for the MR sequence during data acquisition). Data were simulated using physiological T_2 for an unhealthy disc ($T_2 = 80$ ms) and $TE = 13.6$ ms with 20 echoes out to 272 ms to reflect typical MR data acquisition. Low and high SNR data were both simulated (SNR = 5 and 30) using the definition $SNR = S_0/\sigma$. For each SNR, 100 voxels of signal intensity data were

simulated, to represent an ROI, and these signal intensities were averaged and plotted against time to be used for curve fitting to calculate T_2 .

T_2 was then calculated using two different curve fits: typical mono-exponential (MONO) and noise corrected exponential (NCEXP).¹⁹ As described previously, MONO represents the decay to zero which is typically used in the field, and NCEXP represents the corrected signal model that includes Rician noise and the resulting nonzero noise floor. The MONO fit was calculated using MATLAB's built in *fit* function under default options using a monoexponential curve (Equation 4). Two parameters, S_0 and T_2 , were allowed to vary and were unbounded in the fit. The parameters producing the lowest residual error between data and curve fit were used. The NCEXP fit was calculated using nonlinear least squares methods using MATLAB's *fmincon* function with a multi-start approach using 100 randomized initial guesses. The NCEXP curve fit has three parameters: T_2 , S_0 , and σ . Their lower and upper bounds were set to: [0 300 ms], [100 1000], and [20 80], respectively. The fit with the lowest residual error was taken as the best fit, and its parameters reported.

The curve fit procedure was repeated multiple times with different numbers of echoes in the dataset to simulate differing number of TE s chosen during data acquisition. Curve fitting started with just the first four echoes (acquisition to 54.4 ms) and was repeated, increasing the number of echoes by one each time, up to 20 echoes (acquisition to 272 ms). Note that the first echo ($TE = 13.6$ ms) was ignored in the fitting to mimic how real data should be handled. The calculated T_2 and goodness of fit (mean squared error, MSE) were plotted vs the

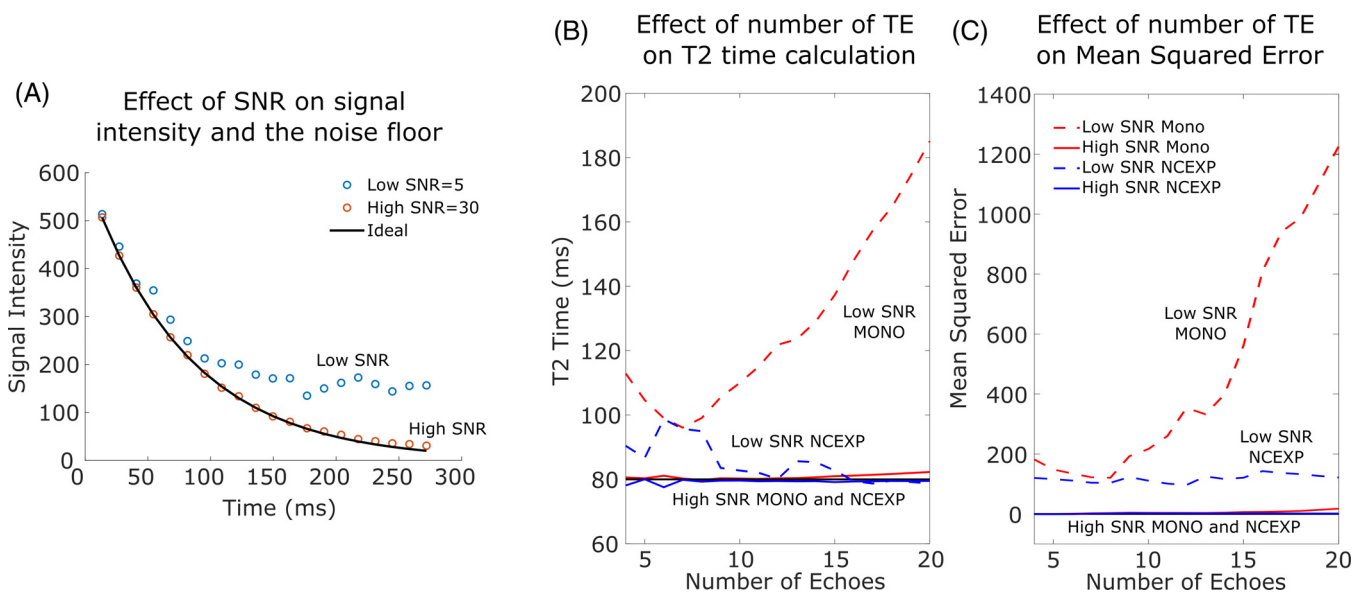


FIGURE 3 Simulation of the effect of SNR on the noise floor and the effect of the fit model on the calculated T_2 for a high SNR of 30 and a low SNR of 5. A, The high SNR disc follows the ideal decay closely out to 272 ms, but the low SNR disc hits a nonzero noise floor. For each curve fitting method, monoexponential (MONO) or noise corrected exponential (NCEXP), the calculated, B, T_2 and C, MSE are shown as a function of number of echoes in the fit. For a MONO fit, the high SNR disc maintains a near perfect 80 ms T_2 and MSE below 20 across all echo train lengths, but the low SNR disc has inaccurate T_2 values and worse fit at all number of echoes. For the NCEXP, T_2 and MSE are closer to ideal than the monoexponential for the low SNR disc, but slightly inaccurate at low number of echoes. The high SNR disc is well fit with the NCEXP giving an accurate T_2 across all number of echoes

number of echoes used for both model types (MONO and NCEXP) and for both low and high SNR (5 and 30) to illustrate the impact of number of echoes, correction for noise, and SNR on the accuracy of calculated T_2 and fit quality.

The signal intensity data simulated with and without noise for a material with 80 ms T_2 (Figure 3A) demonstrates that a low SNR alters signal decay and generates a nonzero noise floor, while a larger SNR of 30 is close to the ideal signal decay. Indeed, for a large SNR of 30, both curve fitting methods produce a calculated T_2 near the expected value of 80 ms and a goodness of fit (MSE) near the ideal value of 0 (Figure 3B,C). Specifically, MONO T_2 is within 3 ms for all echo train lengths (max MSE = 18), and NCEXP T_2 is within 2 ms for all echo train lengths (max MSE = 3). Therefore, high SNR does not require noise correction or long echo trains; however, this high SNR is not typically achievable in practice. With a more realistic SNR of 5, the effect of the model and number of echoes is large. For the low SNR of 5, when fewer than 6 echoes are used in the fit, the MONO fit overestimates T_2 due to the impact of noise on the low number of data points in the fit. Importantly, for low SNR, the MONO fit never matches the expected T_2 of 80 ms regardless of the number of echoes used because of the altered decay curve and noise floor (Figure 3B, red dashed). The MONO fit is closest to the expected 80 ms at 7 echoes (approximately 95 ms), but thereafter increases with increasing number of echoes, reaching a maximum calculated T_2 of 186 ms at 20 echoes or a $>2\times$ overestimation of the expected T_2 . Accordingly, MONO has the highest MSE at all echo train lengths and has a maximum of 1226 with 20 echoes in the echo train (Figure 3C, red dashed). In contrast, NCEXP is within 6 ms of the expected T_2 for echo trains with 8 or more echoes (Figure 3B, blue dashed) and has a much lower MSE than MONO at longer echo train lengths, with a maximum MSE of 143 with 16 echoes in the echo train (Figure 3C, blue dashed). Overall, the MONO fit has large errors that depend on the number of echoes, while NCEXP matches expected values and the number of echoes has less impact.

3.3 | Sensitivity of T_2 to fitting model and number of echoes for in vivo human disc data

To evaluate the sensitivity of T_2 to the fitting model and number of acquired echoes, in vivo spine imaging was performed, and disc signal intensity was fit with two models and two different number of echoes. Lumbar spines from healthy volunteers with no history of back pain were scanned under IRB approved protocols after providing informed written consent ($n = 8$, 24–31 years old). All lumbar discs from each spine were imaged and included in the analysis for a total sample size of 40 discs ($n = 35$ Pfirrmann grade I–II, $n = 5$ grade III–IV). To minimize variation, each scan was acquired at 8 AM (after a full night's sleep) with minimal activity prior to scanning and with the subject laying supine at the MRI facility for at least 45 minutes prior to scanning. A single sagittal slice CPMG sequence was used to collect T_2 data on a 3 T Siemens Magnetom Prisma scanner.³ Sequence parameters included: FoV = 165×220 mm, TR = 3000 ms, TE = 13.6,

27.2, ..., 272 ms (20 total echoes), voxel size $0.57 \times 0.57 \times 5.00$ mm. TR was selected to minimize the contribution of T_1 in the measured signal (see Section 2²³) and 20 echoes were acquired to obtain several data points for evaluating model fits. Scan time was 14:29 minutes.

Image analysis was performed with an in-house code by first creating a circular ROI in the middle of the NP and calculating the average signal intensity in the ROI for each echo, analogous to simulations above. The ROI included 80 to 120 voxels, depending on disc size. Curve fits were then performed on these average intensities to find T_2 . The first echo was excluded for curve fitting.¹⁶

Signal intensity data were fit using two different models (mono-exponential and noise corrected exponential) and two different number of echoes (6 and 20), resulting in the following four groups for comparison: MONO6, MONO20, NCEXP20, and NCEXP6 defined based on fit method and number of echoes collected. The cases with 6 collected echoes are similar to the shorter sequences used in several published studies reporting disc T_2 , and the case with 20 collected echoes represents a larger number of data points in the fit that captures nearly the entire decay and likely extends into the noise floor. It is important to note that only one MRI sequence was acquired per subject to obtain the data. The same data is being analyzed for each fit case, with a different curve fit and number of echoes being considered. Unlike in simulation, where SNR could be specified, the SNR of in vivo scans cannot be easily controlled but is the result of several factors including the subject, the disc characteristics, the MR system, and the selected sequence.

Results were tested for normality using a Jarque-Bera test, and failed to reject the null hypothesis that the data was normally distributed.²⁵ To test for the effect of curve fit on the calculated T_2 , a 2-way ANOVA was run followed by post-hoc matched pair t-tests between all pairings with a significance level set at $\alpha = 0.05$.

Representative data for a Pfirrmann grade IV disc is plotted vs time with fits for MONO and NCEXP curves with 6 or 20 echoes and their respective T_2 (Figure 4A). The dashed vertical line represents the last echo in the fit for the MONO6 and NCEXP6 curve fits (the first echo is not shown). As with the simulated data, the MONO20 curve fit overestimates T_2 (124 ms) and the fit is not very strong (MSE = 521). NCEXP20 and MONO6 give very similar T_2 (66 and 69 ms, respectively). The goodness of fit for MONO6 is great when considering the first six echoes (MSE = 27), but when compared to the full data set the fit is much worse (MSE = 1679) than the fit for NCEXP20 (MSE = 20) meaning it does not accurately describe the full data. This is to be expected as the MONO6 fit only uses the first six echoes and assumes that the signal will decay monoexponentially to zero. The NCEXP6 T_2 is slightly lower than NCEXP20 and MONO6 (58 ms), and while the NCEXP6 fit is good for 6 echoes (MSE = 17), it does not fit the whole data set (MSE = 208) as well as NCEXP20.

When comparing the 4 different calculation methods across all 40 discs (Figure 4B), the same pattern emerges as with the representative disc (Figure 4A). Two-way ANOVA showed significant differences by both factors: echo and curve fit ($p < .001$). MONO6 and NCEXP20 give similar T_2 , 131.12 ± 27.86 ms and 133.66 ± 31.25 ms, respectively, and are not different from one another, despite

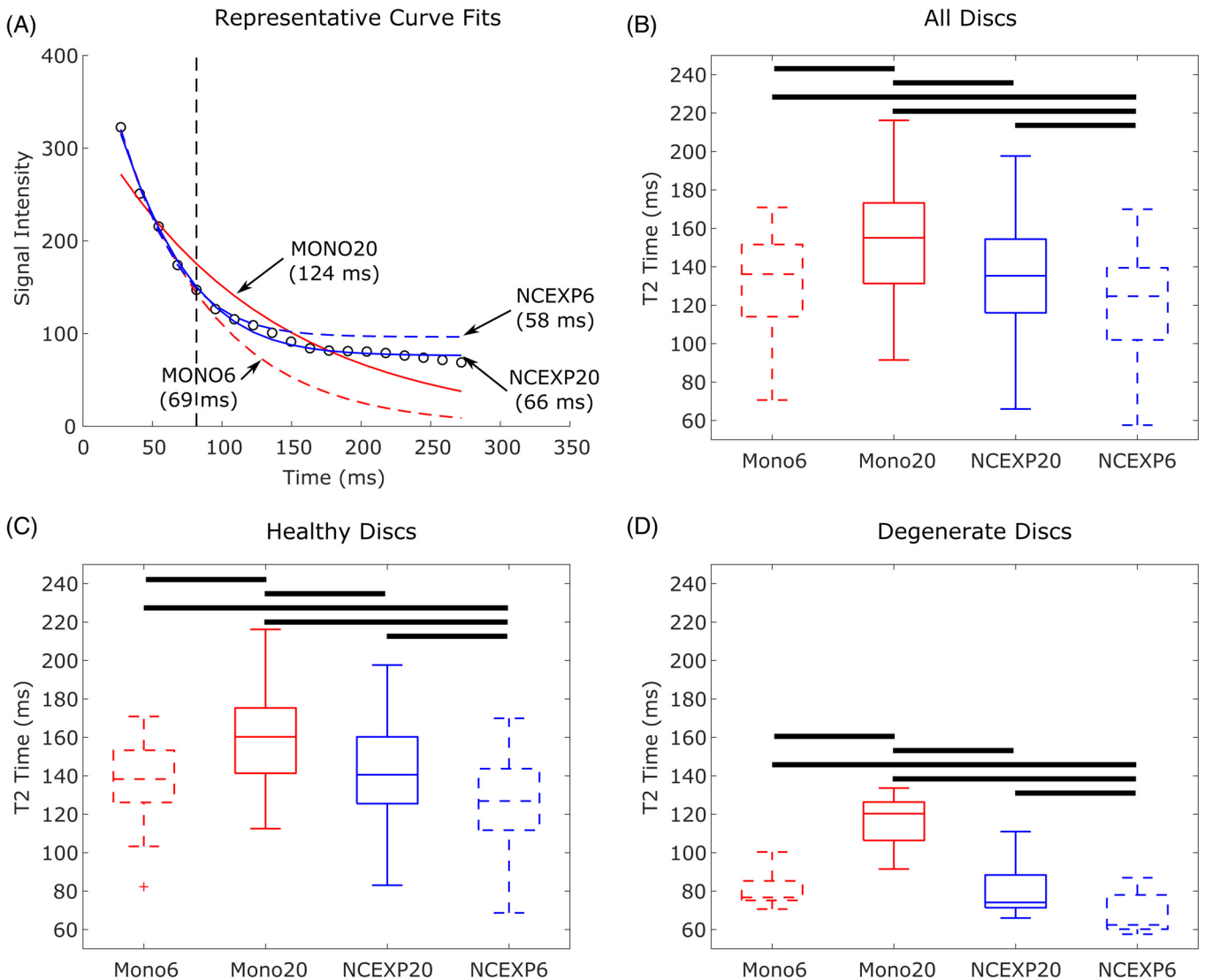


FIGURE 4 A, Representative dataset for a degenerated disc showing all four curve fitting methods. Vertical dashed line represents the last echo used for fitting for MONO6 and NCEXP6 fits (first echo is ignored and not shown) and the rest of the curve is extrapolated. NCEXP fits look better than MONO fits. NCEXP20 and MONO6 give very similar T_2 , but MONO6 is a bad fit. B-D, calculated T_2 for, B, all discs ($n = 40$), C, healthy discs ($n = 35$), and D, degenerate discs ($n = 5$). MONO20 is higher than other measures and likely overestimated because of the noise floor. NCEXP6 is lower than others and likely underestimated because of insufficient data points to predict the noise floor. For all cases MONO6 and NCEXP20 are not different from each other even though the fits for MONO6 are not as strong as those for NCEXP20. Boxes represent median, 25th, and 75th percentiles. Whiskers extend to most extreme data point that is not considered an outlier (+ symbol)

NCEXP20 fitting better than MONO6 ($p > .05$). MONO20 gives T_2 that are significantly higher than all other measures (154.4 ± 29.28 ms, $p < .001$) and likely overestimates T_2 due to sampling of the noise floor. NCEXP6 gives values that are significantly lower than all other measures (119.01 ± 28.85 ms, $p < .001$) and likely underestimates T_2 due to incorporation of an insufficient number of data points (Figure 4B). This finding is the same when the discs are separated into nondegenerate ($n = 35$, Figure 4C) and degenerate ($n = 5$, Figure 4D) discs. The error introduced by using the MONO20 fit in healthy discs is approximately 20 ms compared to NCEXP20, while in the degenerate discs the difference is much larger at approximately 40 ms. It should be noted that NCEXP20 predicted an average

SNR of 10.07 (min 4.65, max 29.26) across all 40 discs, as calculated from the fitting parameters using $SNR = 50\%$.

3.4 | Accuracy of MONO6 and NCEXP20

To determine which fitting method was most accurate, simulations were performed to find the bias and uncertainty from MONO6 and NCEXP20 curve fitting methods. The MONO20 and NCEXP6 fitting methods were omitted from the bias and uncertainty analysis as they were found to be inferior to MONO6 and NCEXP20 fitting methods in the preceding analysis (see Section 3.3). Signal data was simulated

to find the bias and uncertainty of the MONO6 and NCEXP20 curve fits over the SNR range from 5 to 40 (steps by 5) and T_2 range from 50 to 200 ms (steps by 10 ms), for 128 combinations of SNR and T_2 . Each combination of SNR and T_2 was simulated 100 times (representing 100 voxels), and the MONO and NCEXP fits performed to find T_2 for each simulated voxel. The percent error of each calculated T_2 was then found as $\frac{|\text{Calculated } T_2 - \text{Expected } T_2|}{\text{Expected } T_2} \times 100$, and the percent errors were averaged across simulations for every combination of SNR and T_2 . The bias was defined as the average percent error of the 100 voxels for each combo of SNR and T_2 . The uncertainty was defined as the SD of the percent error of the 100 voxels. Heatmaps were generated from bias and uncertainty data using MATLAB's *contourf* function. Zero bias or uncertainty was mapped to white, while 50% error was mapped to red, meaning that areas of white indicated an accurate T_2 calculation while areas of red indicated poor accuracy.

Although true T_2 values are not available in vivo so error cannot be determined, resultant T_2 maps were calculated for each fitting method for the entire disc to compare the heterogeneity of the methods. To do this, the entire disc region was defined in MATLAB

using the polygon tool, then signal data across echoes for every voxel inside the ROI was curve fit individually with the MONO6 and NCEXP20 methods. From the resulting T_2 values, maps were made such that each voxel was assigned a color based on a jet colormap with a range of T_2 values from 0 to 250 ms (black/dark blue indicated low T_2 and red indicated high T_2). These maps were overlaid on the black and white T_2 w MR images to visualize T_2 and its heterogeneity from MONO6 and NCEXP20 fit methods.

MONO6 has higher bias and uncertainty than NCEXP20, particularly at low SNR (Figure 5). The two methods are similar when SNR is greater than 40, but NCEXP20 is still superior. Considering typical disc values, SNR of 10 and T_2 of 130 ms, MONO6 has a bias of 125% while NCEXP20 has a bias of 13.4%. NCEXP20 has a maximum bias of 24.8% at T_2 of 50 ms. Further at SNR of 10, MONO6 bias is >50% error for all but three T_2 values and >100% error for most T_2 . Even at SNR of 20, MONO6 exhibits a bias >20% error in half of the T_2 , while NCEXP20 has a maximum of 10.1% error. Uncertainty follows the same pattern where MONO6 has much higher maximum uncertainty at low SNR (Figure 5). Overall, NCEXP20 is a more precise and accurate method.

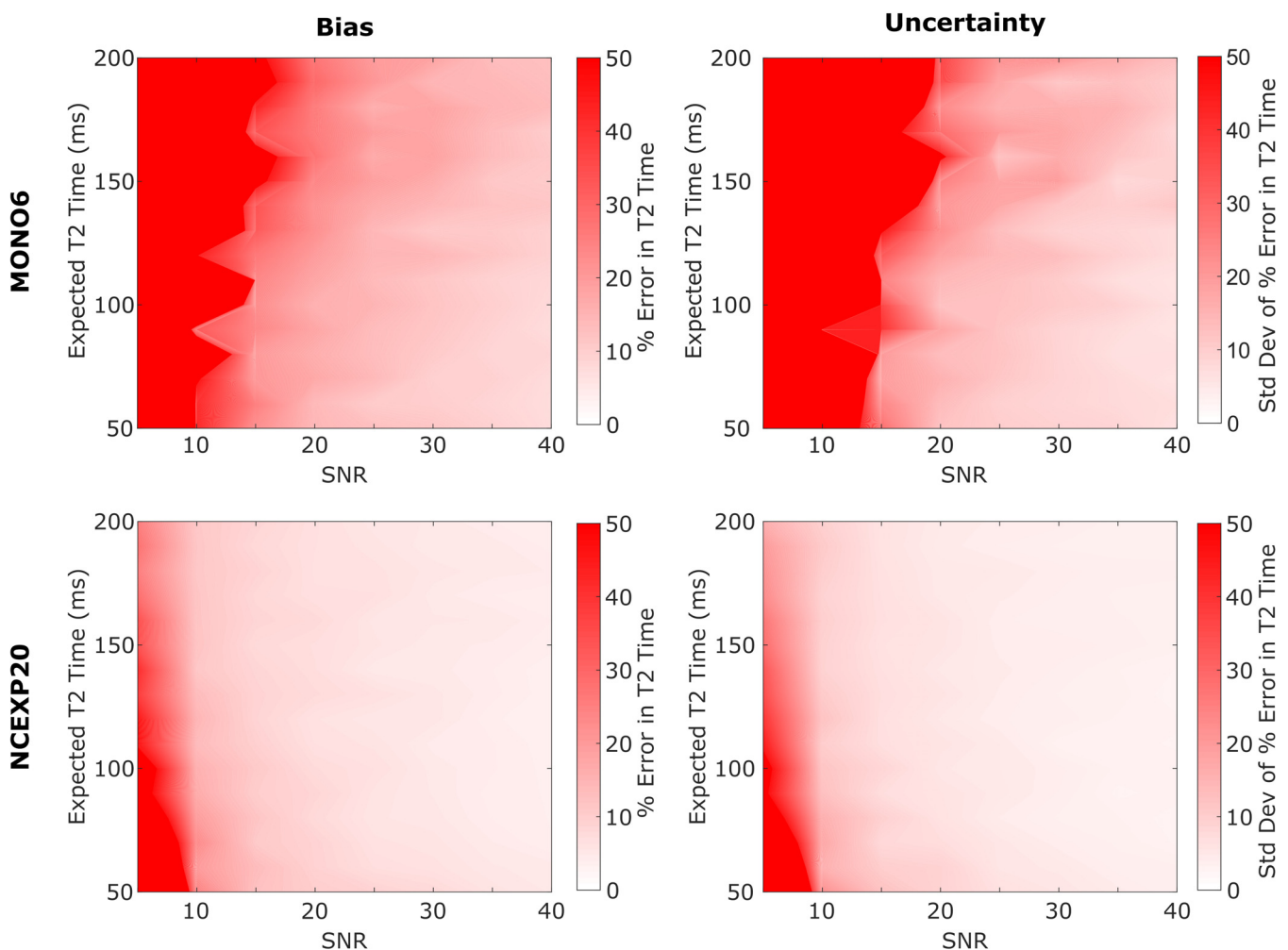
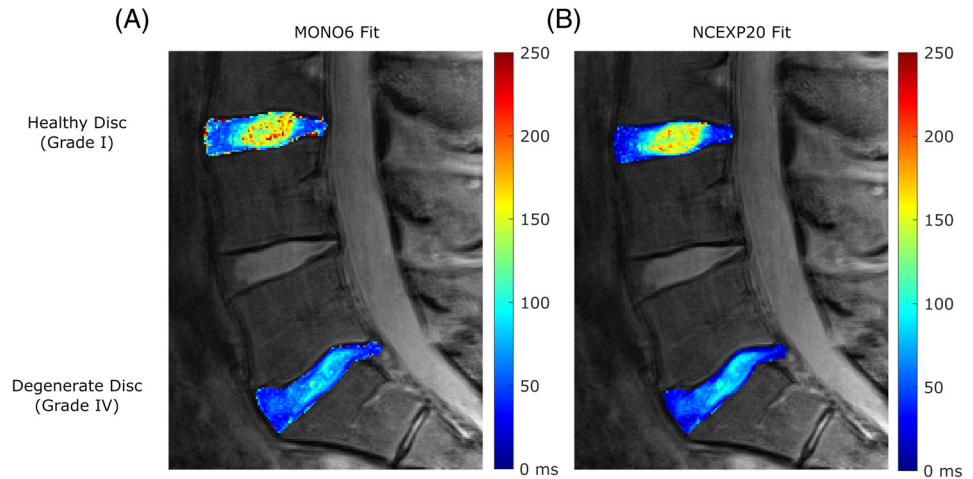


FIGURE 5 Heatmaps of bias (left) and uncertainty (right) in calculated T_2 for MONO6 (top) and NCEXP20 (bottom) using simulated Rician noise. NCEXP20 has lower bias and uncertainty compared to MONO6, making it a more robust calculation method for finding accurate T_2

FIGURE 6 T_2 maps of a healthy L34 disc and degenerate L51 disc with A, MONO6 and B, NCEXP20 fits. NCEXP20 maps are more homogenous with less variation or outlier pixels. MONO6 is more susceptible to error due to noise in signal decay



T_2 maps of full discs allowed visualization of variance of T_2 across the disc (Figure 6). Fits obtained with NCEXP20 display less random variation than those obtained using MONO6. Fitting each voxel with MONO6 shows areas of overestimated T_2 , likely outliers, as can be seen by dark red voxels near the edges of the disc and in the NP. These were more frequent in MONO6 fits. MONO6 is much more susceptible to voxels becoming very overestimated or biased, particularly in areas of low signal at the disc edge, but also in the NP. The overall T_2 averaged for all voxels in the disc are impacted by fitting methods. In the healthy disc, where there are larger differences, MONO6 calculates a T_2 of 3547 vs 84 ms for NCEXP20. In the degenerate disc, MONO6 calculates a T_2 of 53 vs 52 ms for NCEXP20. When using the ROI of the NP, the calculated T_2 are more similar. In the healthy disc, MONO6 calculates a T_2 of 155 vs 147 ms with NCEXP20, and in the degenerate disc both curve fits calculate a T_2 of 80 ms. This is likely because the NP has higher SNR compared to the AF and disc edge regions, so there is less bias and fewer outliers. The higher bias in MONO6 at low SNR likely drives the differences between MONO6 and NCEXP20 and results in inaccurate overestimation of T_2 .

3.5 | Comparison of average intensity of ROI vs voxel calculation methods

We also determined if there is a difference between calculated T_2 when first averaging the signal in the ROI and performing a single curve fit to obtain disc T_2 vs performing the curve fit to the signal in each voxel individually and then averaging the T_2 of each within an ROI to obtain disc T_2 . ROI results using NCEXP20 curve fitting were taken from section 3.3 and voxel-wise maps were generated as described in section 3.4. The same circular ROI used in the ROI method in the NP was applied to the disc map from the previous section, and the average T_2 of all voxels inside the ROI was calculated. The resulting distributions of T_2 from all 40 discs were compared with a paired *t* test, and a histogram of the differences between ROI calculation methods was generated as ROI T_2 —voxel-wise T_2 .

The two methods give very similar results with an average absolute difference of 3.54 ms and average percent difference of 2.77% ($p > .05$, Figure 7A). Of the 40 discs investigated, only two discs exhibited a difference between methods of more than 10 ms (11.19 and 30.23, both Grade I, Figure 7B). The average absolute differences are within the range of standard deviations of reported T_2 . Although voxel methods require much more computation time and single voxels may create outliers (Figure 6), both methods should be assumed to be accurate when SNR is reasonable. However, care may need to be taken when the ROI has a small number of voxels or contains very low SNR regions that may create outliers that may skew the data. Of note, the fit for the voxel method required 60 to 300 seconds depending on ROI size, while a single fit for the ROI method takes less than 5 seconds.

4 | DISCUSSION

This study showed that using the noise corrected exponential (NCEXP) to fit data from 20 echoes out to ~ 300 ms is the best method for calculating T_2 of discs because it is the least likely to be biased by low SNR. By acquiring data out to longer echo times and accounting for Rician noise, the curve fitting is more robust in calculating T_2 despite the noise in the data. This is particularly important when considering degenerate discs or AF tissue because the SNR of these regions will be lower. Additionally, there is little difference between the calculated T_2 from either averaged intensity fitting or voxel-wise T_2 calculation, so either method is viable.

NCEXP20 was more accurate at fitting simulated and in vivo data than other fitting methods and had smaller bias and uncertainty compared to MONO6 across all T_2 and SNR combinations. Collecting data out to longer echo times and taking the Rician noise into account during curve fitting resulted in better fitting of signal decay data and more accurate calculation of T_2 . In simulated data, at high SNR, either MONO or NCEXP fitting worked reasonably well, as the signal decays to zero and not into the noise floor. But with low SNR,

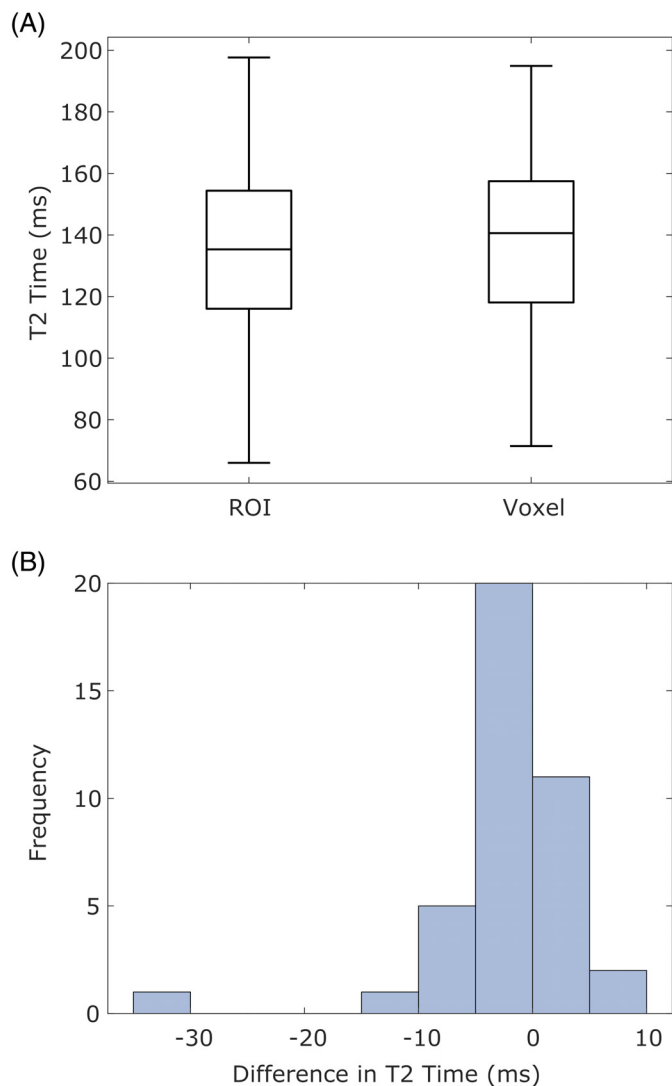


FIGURE 7 A, Comparison of NCEXP20 for average ROI intensity (left) and Voxel (right) calculation methods for a circular ROI in the NP. Distributions are very similar between methods. B, histogram of the differences between the two ROI methods. Differences are centered around 0 with one outlier that has much higher voxel T_2 . Only two discs have T_2 differences greater than 10 ms. Boxes represent median, 25th, and 75th percentiles. Whiskers extend to most extreme data point that is not considered an outlier (+ symbol)

the signal decay is altered by noise and does not decay to zero, so MONO fitting overestimates the T_2 . As a result, the difference in fit quality is most noticeable in degenerate discs that have lost water content in the NP and therefore lost signal in T_2w images and have a lowered SNR. Attempting to fit only the first several echoes (representing some studies that do not have sufficient number of echoes in their protocols) with NCEXP was also inaccurate because the data give little information about the eventual noise floor that the NCEXP curve is trying to fit. Thus, using a short TE with an acquisition time out to $2\times$ the expected T_2 of the tissue (producing a high number of TE and $\sim 85\%$ signal decay) and a NCEXP curve fit is optimal for disc T_2 measurement.

The T_2 values calculated were consistent with previous literature for both healthy and degenerate discs. The healthiest of discs feature NP T_2 in the 150 to 200 ms range, while degenerate grade IV discs were closer to 80 ms. Most literature reports T_2 for healthy discs near 150 ms, while very degenerate discs can be as low as 50 ms.^{11,13,14} Specifically, our data matches closest with literature that has sequence parameters with echoes out to 288 ms and TR of 3000 ms.¹³ Based on MR physics and this observation, we recommend sequence parameters of $TR = 3000$ ms ($3\times$ the expected T_1 of the tissue) and TE out to at least 300 ms ($2\times$ the expected T_2 of the tissue). A short TE should be utilized to maximize the number of echoes that can be curve fit and to capture quick decaying signal. Some published data report healthy disc T_2 in the 75 to 100 ms range. This is surprisingly low for healthy discs and contrary to our data. These discrepancies could be explained by possible combinations of the low TR times leading to signal contamination with T_1 signal, the inclusion of the first echo in the fit, or a low number of echoes in the fit; however, the source of the discrepancy cannot be determined without examining the studies' raw data and curve fits. Old data can be re-analyzed by its owners to determine if systematic T_2 calculation errors occurred, but in the absence of open data we cannot determine if reported T_2 are accurate or proper calculation methods were used. Thus, values reported in the literature should be examined and cautiously used when the methods for calculation are not clear or when methods between papers are not similar.

T_2 can be used as a marker for disc health because of its relation to water content and matrix integrity. As the disc begins to degenerate, the NP loses proteoglycan content and water content. T_2 measurement quantifies the biochemical state because T_2 decreases as water content and water mobility decrease, and thus T_2 serves as a marker for disc degeneration. Further, changes in quantitative T_2 are more robust than simple changes in T_2 weighted signal intensity. It should be noted that it is usually accepted that T_2 does not change very much, if at all, with magnetic field strengths of 3.0T or lower,²⁶ though the signal intensity of a disc can vary depending on the scanner, magnetic field strength, coils used, sequence used, temperature of the subject, and many other factors. T_2 is more robust to these factors because it is a signal decay time constant. Last, unlike Pfirrmann grading, T_2 is quantitative, continuous, and objective, all of which are important for a measurement scheme that is intended to be used across studies. Pfirrmann grading has been shown to correlate with T_2 across grades, but T_2 avoids the problem of subjective binning of discs into five grades.^{6,11} Even more importantly, T_2 is easy to calculate, with sequences readily available on clinical scanners, so T_2 measurements can be easily added to existing imaging protocols as a diagnostic tool or for evaluation of treatments.

Our application of NCEXP in disc follows application of this approach in articular cartilage, which generally has a higher water content and higher SNR. The NCEXP fitting with long echo trains is more robust to noise and more accurate in finding T_2 in phantoms and articular cartilage.^{19,20} We applied these methods to the disc in order to improve T_2 calculation in the spine. This approach can be applied to

other quantitative MRI methods that are based on fitting mono-exponential signal decay (e.g., $T_{1\rho}$) and to other fibrous tissues with low SNR (e.g., meniscus or tendon). $T_{1\rho}$ also follows a mono-exponential signal decay over time and is also susceptible to Rician noise and the presence of a noise floor.²⁷ Our pilot studies with agarose phantoms show that NCEXP curve fitting can be applied to $T_{1\rho}$ data for better $T_{1\rho}$ time calculation at low SNR. Meniscus has very low MR signal because the tightly packed collagen matrix leads to low water mobility and low water content (compared to disc or cartilage), leading to low SNR in the tissue. NCEXP fitting may be an appropriate method to get accurate T_2 or $T_{1\rho}$ for this tissue.

Many previous reports calculated T_2 on a voxel basis, but T_2 is often reported for a specific ROI, for example, the NP in the disc, and calculation of an ROI by averaging the signal intensity and only fitting the averaged data once is computationally faster and suppresses effects of noise. On the other hand, voxel-wise maps are useful for observing inhomogeneities in a region but require a curve fit for every voxel of interest. There appears to be no prior comparison of these two methods in the literature. Depending on the goal of the research, a voxel-wise method can be used to look at the heterogeneity of a region or for looking at the T_2 of the disc across its width in either the anterior-posterior or lateral directions. There should be clear lower T_2 regions at the edges of the disc while the NP region will have higher T_2 . The differences between the AF and NP may be smaller in less healthy discs. For simply calculating the T_2 of a whole region, the average intensity of the ROI method measures T_2 as robustly as traditional voxel-wise measures and takes less computation time.

In conclusion, NCEXP curve fitting of long echo trains with short TE should be adopted as the primary method for calculating T_2 in the disc. Researchers should use TR times that are at least 3000 ms so that nearly full recovery of magnetization is achieved and signal is not contaminated by T_1 weighting. The first echo should be ignored in multi-echo sequences because of the stimulated echo effect. Either the average ROI intensity method or voxel-wise method can be used to calculate T_2 depending on the goal of the research, but the average intensity ROI method will be computationally quicker. Moving forward, sequence parameters and calculation methods need to be clearly defined in reports of T_2 in disc and other tissues. If similar methods are adopted by the field, results can be compared more usefully, promoting faster scientific discovery.

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CONFLICT OF INTEREST

We have no conflicts of interest to disclose.

AUTHOR CONTRIBUTIONS

Kyle D. Meadows, Curtis L. Johnson, and John M. Peloquin contributed to experiments and analysis of the study, Kyle D. Meadows, Curtis L. Johnson, Richard G. Spencer, Edward J. Vresilovic, and Dawn M. Elliott contributed to study design, to results interpretation, and to writing the manuscript. All authors have read and approved submission.

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APPENDIX A

NCEXP function:

```
NCEXP = @(T2, So, sigma) sqrt(pi*sigma^2/2) *
exp(-((So*exp(-TE/T2)/(2*sigma)).^2)).*...
((1+2*((So*exp(-TE/T2)/(2*sigma)).^2)).*...
besseli(0, ((So*exp(-TE/T2)/(2*sigma)).^2))+...
2*((So*exp(-TE/T2)/(2*sigma)).^2)).*...
besseli(1, ((So*exp(-TE/T2)/(2*sigma)).^2))));
```

Where TE is a vector of echo times at which signal data was collected.

Cost function:

```
fun = @(dummy) sum((Data-NCEXP(dummy(1), dummy(2),
dummy(3))).^2);
```

Where “dummy” is a 3 × 1 vector of T₂, So, sigma, and “Data” is a vector of signal intensity decay over TE.

Curve fit:

```
[par_fit, fval] = fmincon(fun, par0, [], [], [], [], lb,
ub, [], []);
```

Where “par_fit” is a 3 × 1 output vector of fit parameters T₂, So, sigma, and “fval” is the value of the objective function for the fit using “par_fit.” “par0” are initial guesses for fit parameters, and “lb” and “ub” are lower and upper bounds of fit parameters, respectively.