Every hit matters: White matter diffusivity changes in high school football athletes are correlated with repetitive head acceleration event exposure

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1. Introduction

Researchers have observed that retired American football athletes (who have extended histories of exposure to subconcussive impacts) may have a higher risk of developing neurodegenerative disorders such as chronic traumatic encephalopathy, Alzheimer’s disease, and Parkinson’s disease (Omalu et al., 2006; Broglio et al., 2009; McKee et al., 2009; Stern et al., 2011; Lehman et al., 2012). Previous neuroimaging work has demonstrated changes in brain function and chemistry are associated with the accumulation of exposure to head acceleration events (HAEs), even in the absence of a diagnosis of concussion. Exposure to these “subconcussive” HAEs has been observed to be associated with alterations in the brain’s response to task demands (Talavage et al., 2014; Robinson et al., 2015; Shenk et al., 2015), functional connectivity (Johnson et al., 2014; Abbas et al., 2015a,b), cerebrovascular reactivity (Svaldi et al., 2015, 2017, 2018), biochemical concentrations (Poole et al., 2014, 2015; Bari et al., 2018), and resting perfusion (Slobounov et al., 2017). Such alterations in function...
have been suggested as precursors to the symptoms normally resulting in the diagnosis of a concussion, with accumulation of HAEs put forth as a likely mechanism for symptom development (Bailes et al., 2013; Talavage et al., 2016).

While alterations in physiology like those reported above may arise from natural development or physical training (Lebel and Beaulieu, 2011; Lebel et al., 2012; Giorgio et al., 2010; Simmonds et al., 2014; Maddock et al., 2011), they may also arise from underlying structural damage to cells within the nervous system. Neural injury of this nature is typically assessed in MRI using diffusion-weighted imaging (DWI), with tensor-based analysis (diffusion tensor imaging, DTI) applied to focus on white matter integrity. In the healthy case, the DTI-measured diffusion of water molecules in white matter tracts is expected to be anisotropic—specifically, more directed along the length of an axon than outward through the cellular membrane and myelin sheath. Changes in the DTI-based measures of fractional anisotropy (FA) and the associated mean diffusivity (MD) are thus interpreted as markers or confirmation of changes to white matter health (e.g., Beaulieu, 2002; Arfanakis et al., 2002; Inglese et al., 2005; Bazarian et al., 2007).

For sport-related concussion and subconcussive trauma, measures of white matter health have been used to document changes associated with participation in a season of American football and soccer (Lipton et al., 2013; Bazarian et al., 2014; McAllister et al., 2014; Davenport et al., 2014, 2016; Chun et al., 2015; Bahrami et al., 2016; Gong et al., 2018; Myer et al., 2018). In these studies, alterations in white matter health have been quantified through comparison of pre- and post-participation measures, with alterations in mean FA (or associated measures) interpreted as evidence of the effect of the intervening (and only sometimes quantified) HAEs. However, while these studies have revealed a number of changes in white matter health resulting from said participation, the results have been highly variable. Some studies have reported reduced mean FA and/or elevated mean MD (Arfanakis et al., 2002; Inglese et al., 2005; Miles et al., 2008; Cabon et al., 2011; Lipton et al., 2013; Mustafi et al., 2018), while others have provided evidence for increased mean FA and/or decreased mean MD (Bazarian et al., 2007, 2012; Wilde et al., 2008; Mayer et al., 2010, 2012; Chu et al., 2010; Henry et al., 2011; Ling et al., 2012b; McAllister et al., 2012). Decreased FA and increased MD measurements are commonly considered to be indicative of impaired white matter structural integrity—e.g., myelin damage, including demyelination; or disruption of tissue structure, including axonal damage (Arfanakis et al., 2002; Kraus et al., 2007; Mac Donald et al., 2007a,b; Shitaka et al., 2011; Magnoni et al., 2015; Sundman et al., 2015; Pan et al., 2016; Kantarci et al., 2017). Conversely, increased FA and decreased MD are commonly attributed to either (a) axonal swelling arising from extra-neuronal injury, reducing space between fibers (Povlishock et al., 1983; Grady et al., 1993; Christman et al., 1994; Povlishock and Katz, 2005; Bazarian et al., 2012; Shultz et al., 2012; Johnson et al., 2013), or (b) cytotoxic edema and inflammation possibly resulting from ion homeostasis failure, membrane dysfunction, and microglial activation (Giza and Hovda, 2001; Hua Li et al., 2004; Giza and Prins, 2006; Marmarou et al., 2006; Wilde et al., 2008; Chu et al., 2010; Loane and Byrnes, 2010; Giza and Hovda, 2014; Giza et al., 2017).

Therefore, the critical question whether exposure to repeated HAEs produces what would be readily recognized as “injury” to the underlying brain structure remains open. Further the literature has not effectively addressed whether these reported changes—whether linked to injury or inflammation—are predominantly driven by natural growth, participation in intensive exercise, or are direct consequences of exposure to repeated subconcussive trauma. This retrospective study uses DTI acquired in a prospective study of high-school-aged American football athletes to identify the nature and extent of the changes in white matter health and structure associated with the accumulation of exposure to HAEs. While assessments at the whole-brain level can reasonably be expected to reflect severe injuries, such as those associated with vehicular accidents or falls (e.g., Lipton et al., 2008, 2013), the progression of damage to a symptom level likely requires a finer scale assessment. Confirmation that white matter alterations, likely to reflect one or both of inflammation or injury, are correlated with known mechanical exposures will provide key insight into the near-term risks of accumulation of repeated subconcussive events.

2. Methods

2.1. Participants

Previously-collected data from 181 high school-aged (i.e., ages 14–18) male athletes participating in American football (N = 162) or noncollision sports (N = 19) were evaluated for this study. Noncollision athletes indicated participation in track (N = 9), swimming (N = 6), cross-country (N = 7), or basketball (N = 2), with some participating in more than one sport. None of the included subjects reported having been diagnosed with a concussion within the 3 months prior to the period of study. Further, none of the athletes were diagnosed with a concussion by their team healthcare professionals during the period of study.

2.2. Participation schedule

2.2.1. Football athletes (FBA)

150 of the 162 football athletes participated in at least four MRI sessions were scheduled to encompass one competition season: one in the 2 months preceding onset of contact practices (Pre); one each within the first (In1) and second (In2) 6-week segments of the competition season, corresponding to an average of 6 (In1) and 12 (In2) weeks after Pre; and one 4–6 months after the end of the competition season (Post), at an average of 35 weeks after Pre (see Fig. 1, top). Note that the Pre session took place during summer conditioning, when FBA were engaged in regular workouts involving light contact (e.g., “shells”), but did not involve full-contact (e.g., “thud”) or tackling practices. The average intervals (± standard deviation) between the onset of contact activities and the corresponding follow-up sessions were (In1) 3.4 ± 2.0 weeks; (In2) 9.4 ± 2.1 weeks; and (Post) 32.1 ± 2.4 weeks, corresponding to 19.9 ± 2.9 weeks after the cessation of contact activities.

2.2.2. Noncollision-sport athletes (NCA)

The 19 noncollision-sport athletes were scanned twice (Test and Retest), at an interval of 5–18 weeks (average = 8 weeks), while actively engaged in training or competition, which are indistinguishable from one another in this population (see Fig. 1, bottom).

2.3. Head acceleration event monitoring

All football athletes were monitored for head acceleration events (HAEs) to assess relative mechanical loading across the population. Athletes were monitored throughout all team practices and games—for details, see Breedlove et al. (2012) and McCuen et al. (2015). Sensors used were either the HIT System (Simbex, LLC), a helmet-based telemetry system; or the xPatch (X2 Biosystems, Inc), a head-mounted sensor.

Both devices were set to record all events whose peak translational acceleration (PTA) exceeds 10 g, but analysis was conducted only on those events exceeding 20 g. Our previous work (Cummins et al., 2017) suggests that 20 g currently represents the lowest reasonable threshold for which the HIT System and xPatch are each reliable and consistent indicators of the presence of HAEs, as event counts exceeding this minimum threshold were found to be similar across both devices. It is critical to note that, based on laboratory testing, the (specific) magnitudes and locations provided for each HAE were not used in this work, as the errors associated with each individual measurement are substantial—sometimes exceeding 100% root-mean-square error—for...
these sensors (Jadischke et al., 2013; Cummiskey et al., 2017). However, given that both sensor systems have been found to be relatively unbiased on average (generally under 15% error; see Cummiskey et al., 2017), counts of events reported to exceed a given threshold may be used in a regression with increasing predictive power as the number of experienced events grows (i.e., the expected systematic undercounting will become more consistent across all subjects).

2.4. MRI data acquisition

All MRI sessions were performed at the Purdue MRI Facility, on a 3-T General Electric Signa HDxt (Waukesha, WI), using a 16-channel brain array (Nova Medical; Wilmington, MA). Head motion was minimized with restraining foam pads. DWI acquisitions used a two-dimensional single-shot spin-echo echo-planar imaging (EPI) sequence (repetition time [TR] = 12,000, echo time [TE] = 83.6ms, flip angle = 90°, field of view = 240 mm × 240 mm, in-plane resolution = 2.5mm × 2.5mm, slice thickness = 2.5 mm, slice gap = 0 mm, 46 contiguous axial slices, frequency readout = R/L) with 30 diffusion encoding directions at $b = 1000 \text{s/mm}^2$ and one volume acquired at $b = 0 \text{s/mm}^2$. Raw images were upsampled to a 256 × 256 matrix by the MRI system for an image voxel size of 0.938 mm × 0.938 mm × 2.5 mm.

2.5. Data processing and quality assessment

2.5.1. Pre-processing

DWI data were processed using FSL (Smith et al., 2004; Woolrich et al., 2009; Jenkinson et al., 2012). For each image, a brain mask was generated on the non-diffusion-weighted volume (i.e., $b = 0$) by segmenting brain from non-brain tissues (BET; Smith, 2002). Corrections were then applied for head movements and eddy current-induced distortions (Eddy; Andersson et al., 2016) while detecting slices with signal dropout and replacing them with Gaussian process predictions (−repol option in Eddy; Andersson and Sotiropoulos, 2016). Scalar diffusion tensor maps were then estimated by fitting the diffusion tensor model at each voxel (FDT; Behrens et al., 2003). Fractional anisotropy (FA) and mean diffusivity (MD) were subsequently calculated from the three primary eigenvalues.

2.5.2. Quality assurance

Prior to voxel-wise analysis, quality assessment was performed on the images output from the preceding process, to ensure that acquisitions were not significantly corrupted. The following criteria for exclusion were similar to, but more stringent than, Murugavel et al. (2014). First, a computational assessment was conducted. Head movements during imaging were estimated 1) between every consecutively-acquired volumes and 2) relative to the first volume, based on each volume’s registration parameters (Ling et al., 2012a). Subjects with at least one displacement relative to the first volume exceeding 5.0 mm were excluded from the study. For each scan, the average values of translations and rotations per unit time (i.e., time between two consecutive volumes within one DWI scan) were calculated across all 30 time points. Those subjects whose average movement exceeded three standard deviations (from the mean across all scans) in any translation/rotation along/around the x, y, or z-axis were also ruled out. Next a visual assessment was conducted, discarding any remaining data in which artifacts could be observed, or for which reconstruction had been improper.

After quality assessment, the resulting dataset comprised complete sets of data from 61 FBA (i.e., four valid imaging sessions and complete HAE data) and 15 NCA (i.e., both valid imaging sessions). See Table 1 for demographics of participants whose data passed screening and were included in analyses. The remaining 101 FBA were excluded for one or more of the following reasons: (a) they did not participate in all four

**Table 1**

| Demographics of participants with complete set of valid imaging data. | FBA ($N = 61$) | NCA ($N = 15$) |
|---|---|---|
| Age (years) | Mean ± StdDev | 16.6 ± 0.9 | 16.5 ± 1.2 |
| | [Min, Max] | [15, 18] | [14, 18] |
| Years of current sport (high school) | Mean ± StdDev | 2.2 ± 0.8 | 2.1 ± 1.0 |
| | [Min, Max] | [0, 3] | [0, 3] |
| Number of previously diagnosed concussions | Mean ± StdDev | 0.6 ± 1.0 | 0.5 ± 0.8 |
| | [Min, Max] | [0, 5] | [0, 2] |
| Racial and ethnic categories | | | |
| White | 43 | 13 |
| Black or African American | 14 | 0 |
| Hispanic or Latino | 1 | 0 |
| Asian | 1 | 2 |
| More than one | 2 | 0 |
imaging sessions, (b) they experienced an injury during the season that resulted in cessation of active participation, (c) their HAE data were incomplete (e.g., battery failure or sensor repeatedly fell off athlete), or (d) subject motion was excessive and the imaging data did not pass quality assurance. All 4 excluded NCA were on the basis of motion/failure to pass QA.

2.5.3. Image registration

Datasets that passed quality assessment were input to a subset of the Tract-Based Spatial Statistics (TBSS) pipeline (Smith et al., 2004, 2006; Andersson et al., 2007a,b) in FSL, to obtain and assess predominant fiber tracts within white matter of the brain. All FA images were non-linearly registered to the 1 mm isotropic FMRIB58-FA standard-space image, yielding a transformation for each subject. Consistent with prior studies (e.g., Kraus et al., 2007; Oni et al., 2010; Gajawelli et al., 2013; Myer et al., 2016b; Slobounov et al., 2017; Kuzminski et al., 2018), a mean FA image was calculated from the registered images from all subjects, and thresholded at 0.2 to create a mean white matter (WM) skeleton, hereafter referred to as ROI WM.

Subsequently, the aligned FA image of the i-th subject obtained at the j-th session was projected onto ROI WM to form an FA skeleton specific to each session and individual (FA WM,i,j). MD skeletons (MD WM,i,j) were similarly created by applying the subject's transformation to the raw MD images followed by projection to ROI WM.

2.6. Statistical analysis

Standard methodologies as described in Glantz (2012) were applied throughout for statistical testing purposes.

2.6.1. Statistical tools

Statistical analyses were performed using STATA 14.0 (StataCorp LP; College Station, TX), MATLAB R2016 (MathWorks), and FMRIB Software Library (FSL) 5.0.

2.6.2. Repeated measures testing

In cases of analyses across sessions and groups, the Shapiro-Wilk test for normality and Bartlett test of sphericity were conducted to ensure the validity of using a one-way repeated measures analysis of variance (ANOVA). If the normality assumption was violated, the Friedman non-parametric test was used in its place. If the sphericity assumption was violated, the Huynh-Feldt correction was applied to the resulting statistics.

2.6.3. Two-sample hypothesis testing

For comparison across two sets of data, the Shapiro-Wilk test for normality was conducted to ensure the validity of using a t-test. If this normality assumption was violated, the non-parametric Wilcoxon rank-sum test (also called the Mann-Whitney U test) was used.

2.7. Image analysis

Several analyses were conducted to detect and evaluate the dependence of changes in white matter diffusion on exposure to the repetitive subconcussive HAEs (associated with a single competition season of American football). First, group-level longitudinal changes were sought by analyzing the magnitude and spatial extent (effectively a volume, comprising a collection of voxels) of changes in FA and MD. Second, the spatial extents of substantial changes in FA or MD (“change masks”) were identified at the individual subject level—note that white matter locations at which FA or MD decreased and increased were identified separately, given these alterations have different pathophysiologic implications. Third, correlations between the spatial extents of the aforementioned change masks and the accumulated HAE exposure were analyzed.

2.7.1. Group-level longitudinal changes

2.7.1.1. Longitudinal changes in mean FA/MD

Consistent with prior literature (e.g., Meier et al., 2015; Henry et al., 2011; McAllister et al., 2014; Myer et al., 2016a), one-way repeated measures ANOVAs were conducted to determine if either the NCA or FBA groups exhibited longitudinal changes across sessions in mean FA (ΔFA WM) or mean MD (ΔMD WM), as averaged over the entire white matter skeleton (i.e., ROI WM)—i.e., FA WM,i,j and MD WM,i,j, where for i ∈ FBA: j ∈ (Pre, In1, In2, Post); and for i ∈ NCA: j ∈ (Test, Retest). A further one-way ANOVA was performed to assess whether race had a significant effect on mean FA or mean MD.

Given the hypothesis that FBA who are no longer exposed to repeated HAEs might exhibit natural recovery toward baseline values of FA and MD (e.g., after the competition season ends), a second set of repeated measures ANOVAs was conducted on the mean FA and mean MD for this population to evaluate changes, relative to Pre, only for those follow-up sessions acquired during the active accrual of HAEs (i.e., In1, In2).

When significant effects of session were observed, pairwise comparison post hoc tests were conducted on the (subject-level) values of mean FA or mean MD in the corresponding population to identify those sessions that significantly differed in mean from baseline (i.e., estimated marginal mean).

2.7.2. Subject-level longitudinal changes

Individual subject changes during and after exposure to HAEs were obtained through subject-specific quantification of extent of volumetric masks comprising voxels in which FA or MD values were significantly-altered over time.

2.7.2.1. Individual subject mask generation.

Given that we might expect changes in FA (or MD) over time in adolescent athletes (Giorgio et al., 2010; Lebel and Beaulieu, 2011; Lebel et al., 2012; Simmonds et al., 2014), it is desirable to further refine our analysis to focus only on those voxels in which we observe across-session changes unlikely to be associated with normal white matter development. Masks of change in FA (ΔFA) or MD (ΔMD) were generated using the population statistics from the NCA as a reference, assuming that changes markedly outside the range observed in the NCA population are potentially a consequence of exposure to repeated HAEs.

To this end, “change masks” that reflected any substantial alteration in one of FA or MD at a follow-up session, relative to baseline, were constructed for all subjects (both NCA and FBA) as follows. First, for each voxel in ROI WM a 95% confidence interval (CI) was constructed from the NCA pool based on FA (or MD) changes observed at Retest relative to Test. Second, a change mask was created for each athlete (FBA and NCA) at each follow-up session (FBA: In1, In2, and Post; NCA: Retest), through identification of those voxels for which the subject’s change in FA (or MD), relative to Pre, fell outside the corresponding voxel-specific NCA-defined 95% CI. Note that voxels for which the observed change in FA (or MD) fell outside the 99.9% CI are likely to be erroneous, and were excluded when creating subject-specific masks. (See Fig. 2.) The resulting (unsigned) change masks (roi ΔFA WM,i,j, roi ΔMD WM,i,j) represent the spatial extents of substantial change for the corresponding measures for subject i at follow-up session j.

Given the potentially different pathophysiologic causes of increases and decreases in FA and MD, separate (signed) change masks were generated for each direction of substantial alteration. An “increase change mask” was generated for each FBA and NCA subject, at each follow-up session, by identifying the subset of FA (or MD) change mask voxels exhibiting changes from Pre exceeding the upper 95% CI bound (roi ΔFA WM,i,j, roi ΔMD WM,i,j). Similarly, a “decrease change mask” was generated for each FBA and NCA subject, at each follow-up session, by identifying the subset of FA (or MD) change mask voxels exhibiting changes from Pre falling below the lower 95% CI bound (roi ΔFA WM,i,j, roi ΔMD WM,i,j).
2.7.2.2. Comparison of change mask spatial extents between FBA and NCA. For each of the six change masks (unsigned, increase, decrease; FA and MD) created for each subject above, the spatial extent of the change mask, as a percentage of the total voxel count in ROI\(_{WM}\), was computed and compared across the FBA and NCA pools. These spatial extents are indicated as \(\|\) operating on a given change mask (i.e., \(\|\text{roi}_{\Delta MD_{j,i}}\|\) indicates the percentage extent, relative to the white matter skeleton, of the signed change mask associated with substantial decreases in MD).

2.7.2.3. Correlation of longitudinal changes in FBA masks with HAE exposure. For all FBA subject change masks in ROI\(_{WM}\), Pearson (linear relationship) and Spearman (monotonic relationship) correlation analyses were conducted across each FBA follow-up session (In1, In2, Post), comparing the spatial extent of the change mask with the number of HAEs experienced to-date in practices and games that exceeded a given PTA threshold. PTA thresholds examined were 20g, 30g, 40g, 50g, 60g, and 70g. The upper-bound threshold of 70g was selected as it represents approximately the 90th percentile of all recorded HAEs (Bari et al., 2018), and was expected to preserve a sufficient number of events per athlete such that the count remained accurate across subjects (cf. Cumminsley et al., 2017). Thresholds at which the correlation met a corrected significance level of \(p_{\text{Bonferroni}}<0.05\) were noted.

At the PTA threshold exhibiting the most significant correlation, a linear regression analysis was conducted to characterize the relationship between the change mask spatial extents at each FBA follow-up session and the number of HAEs experienced to-date, exceeding said threshold. For this regression, the confidence intervals for the true regression lines (i.e., confidence bands) were determined. Regressions for which the confidence bands did not contain a slope of zero were interpreted to be suggestive of a contribution to white matter alterations from exposure to repeated (sub-concussive) HAEs exceeding the identified threshold.

2.7.2.4. Group-level altered WM mask. Based on the outcome of the regression analysis for the entire white matter skeleton, we desired to assess the degree to which exposure to repeated HAEs was driving the observation of statistically-significant changes in white matter measures. Therefore, the ensemble of voxels exhibiting a statistically-significant change, relative to baseline (FBA: Pre; NCA: Test) at any follow-up session (FBA: In1, In2, or Post; NCA: Retest) was identified through use of a pair-wise (by session) permutation \(t\)-test (50,000 permutations for each of Pre vs. In1, Pre vs. In2, Pre vs. Post; NCA: Retest). All results of the permutation \(t\)-test underwent threshold-free cluster enhancement (Smith and Nichols, 2009) and multiple-comparison correction for family-wise error (FWE), and voxels exhibiting \(p_{\text{FWE}}<0.05\) were identified at each follow-up session. The resulting voxels were combined across tests to produce a mask of altered white matter, ROI\(_{ALT}\). In other words, ROI\(_{ALT}\) represents the subset of voxels within ROI\(_{WM}\) that exhibited any significant change (increase or decrease) relative to baseline, at any follow-up session (see Fig. 3).

For the purpose of understanding the spatial distribution of the voxels found to exhibit statistically-significant changes in any follow-up session, the WM tracts containing voxels within ROI\(_{ALT}\) were identified through matching with the digital white matter atlas from Johns Hopkins University (JHU ICBM-DTI-81; Mori et al., 2005).

2.7.2.5. Subject-level longitudinal changes in altered WM mask. Using the same methods as described above, six change masks (unsigned: \(\text{roi}_{\Delta FA_{ALT,j,i}}\), \(\text{roi}_{\Delta MD_{ALT,j,i}}\); signed: \(\text{roi}_{\Delta FA_{ALT,j,i}}\), \(\text{roi}_{\Delta MD_{ALT,j,i}}\)) were generated from the voxels comprising ROI\(_{ALT}\). Note that this process is identical to the intersection of ROI\(_{ALT}\) with each of the subject change masks.

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**Fig. 2.** Illustration of the process by which FA and MD change masks (unsigned and signed) were generated at each imaging session for each member of the FBA and NCA populations, in both the white matter skeleton (ROI\(_{WM}\)) and in the altered white matter mask (ROI\(_{ALT}\)). See text (Individual subject mask generation) for details.
generated earlier (e.g., roi_ΔMDALT = roi_ΔMDWM ∩ ROIALT). The resulting change masks were evaluated using the same methodologies as above (for ROIWM), assessing (1) longitudinal changes in mean FA and mean MD within ROIALT, (2) group differences, between FBA and NCA, in FA and MD change mask spatial extents within ROIALT, and (3) the longitudinal relationship between FBA change mask spatial extents within ROIALT and HAE exposure.

3. Results

3.1. HAE exposures

Fig. 4 summarizes the distribution (median, 1st and 3rd quartile) of counts of HAEs experienced by the corpus of 61 FBA in practices and games, from the beginning of the scholastic season to the time of the corresponding follow-up sessions (In1, In2, Post). A repeated measures ANOVA revealed no effect of race (categories: White, Black/African American, Hispanic/Latino, Asian, More than one)—the only demographic factor that differed across the NCA and FBA groups—on to-date HAEs at each follow-up session.

3.2. Group-level longitudinal changes

3.2.1. Longitudinal changes in mean FA/MD

Population distributions over the white matter skeleton (ROIWM) of mean FA (FAWM) and mean MD (MDWM) are presented as a function of group and session in Fig. 5. Note that at baseline, there was no significant difference in either mean FA or mean MD between FBA and NCA. Further, as with HAEs, above, a repeated measures ANOVA revealed no effect of race or age on mean FA or mean MD across sessions.

As seen in Table 2, there was no effect on session for NCA or FBA when all available (baseline plus follow-ups) were evaluated by repeated measures ANOVA. However, the repeated measures ANOVA conducted on FBA using only those follow-up sessions acquired during period of exposure to HAEs (i.e., Pre, In1, and In2) revealed a statistically-significant effect of session for mean FA. Post hoc pairwise analysis of the three FBA sessions revealed significant changes in mean FA relative to Pre, at both In1 and In2 (Table 3). Note that in Table 2 one session (at In1) of one FBA subject was excluded as an outlier from the repeated measures ANOVA for mean MD, because the subject's
In WM, the value was more than 250% of the interquartile range below the 25th percentile (see Fig. 5, bottom).

Scatter plots of changes, relative to baseline, in mean FA and mean MD for all FBA and NCA subjects at each follow-up session (i.e., $\Delta{\text{FA}}_{\text{WM}}$, $\Delta{\text{MD}}_{\text{WM}}$) are shown in Fig. 6. It may be readily observed that FBA consistently exhibit broader distributions of changes (both increases and decreases) than do NCA.

### 3.3. Subject-level longitudinal changes

#### 3.3.1. Comparison of change mask spatial extents between FBA and NCA

For the entirety of the white matter skeleton ($\text{ROI}_{\text{WM}}$) the spatial extents of substantial changes in both FA and MD—i.e., $\|\text{roi}_i \Delta{\text{FA}}_{\text{WM},i}\|$ and $\|\text{roi}_i \Delta{\text{MD}}_{\text{WM},i}\|$—were significantly larger ($p < 0.0001$; t-test or Wilcoxon rank-sum test) at each follow-up session for FBA relative to NCA (Fig. 7).

Consistent with the differences observed for the unsigned change masks, all signed change masks associated with $\text{ROI}_{\text{WM}}$ were also found to be significantly larger in spatial extent ($p < 0.001$; t-test or Wilcoxon rank-sum test) at each follow-up session for FBA relative to NCA (Fig. 8).

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**Table 2**

Repeated measures ANOVA outcomes evaluating mean FA and mean MD in $\text{ROI}_{\text{WM}}$ across indicated sessions.

| Group | FA/MD | Sessions | $F$-statistic | Corrected-$p$ |
|-------|-------|----------|---------------|---------------|
| NCA   | FA$_{\text{WM}}$ | All sessions | $F(1, 14) = 0.08$ | 0.785 |
|       | MD$_{\text{WM}}$ | All sessions | $F(1, 14) = 0.23$ | 0.642 |
| FBA   | FA$_{\text{WM}}$ | All sessions | $F(3, 180) = 2.44$ | 0.140 |
|       | MD$_{\text{WM}}$ | All sessions | $F(3, 179) = 2.59$ | 0.114 |

Where the sphericity assumption was violated, $p$-values were corrected with Huynh–Feldt’s method. Final $p$-values were corrected using a Bonferroni correction, given FBA data underwent two tests. * indicates a result significant at the (corrected) $p < 0.05$ level.

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**Table 3**

Post hoc pairwise t-test comparisons of mean FA measurements in $\text{ROI}_{\text{WM}}$ of FBA, acquired in sessions associated with accumulation of HAEs, for which a significant effect of session was observed (see Table 2).

| Pairwise comparisons | t-Statistic | Corrected-$p$ |
|----------------------|-------------|---------------|
| In1 vs. Pre          | 2.64        | 0.013         |
| In2 vs. Pre          | 2.13        | 0.035         |
| In2 vs. In1          | $-0.51$     | 0.611         |

Duncan’s method was used to correct for multiple comparisons. (Note that a positive sign for $t$ indicates that the measurement in second listed session is greater than the measurement in the first.) * indicates a result significant at the (corrected) $p < 0.05$ level.
3.3.2. Correlation of longitudinal changes in FBA masks with HAE exposure

To-date HAE accumulations were only found to be significantly correlated with the spatial extent of the change mask reflecting substantial decreases in FA (i.e., $\|\text{roi}_\text{FAWM}\|$), and only for PTA thresholds of 20 g and 30 g (see Table 4).

Fig. 7. At all follow-up sessions, football athletes (FBA) exhibited significantly ($p < 0.001$) greater spatial extents of substantial changes in FA and MD, in ROI$_{WM}$, than did noncollision athletes (NCA). Box-and-whisker plots are presented at each follow-up session for (A) $\|\text{roi}_\Delta\text{FA}_{WM,j,i}\|$; (B) $\|\text{roi}_\Delta\text{MD}_{WM,j,i}\|$.

Regressions against the HAE count at 20 g, the threshold associated with the most significant correlation with signed change mask spatial extents, are shown for all signed change masks in Fig. 9. Consistent with Table 4, the regression fit for $\|\text{roi}_\Delta\text{FA}_{WM,j,i}\|$ against the to-date accumulation of HAEs exceeding 20 g was found to be statistically significant.

Fig. 8. At all follow-up sessions, football athletes (FBA) exhibited significantly ($p < 0.001$) greater spatial extents of substantial increase/decrease in FA and MD, within ROI$_{WM}$, than did noncollision athletes (NCA). Box-and-whisker plots are presented at each follow-up session for (A) $\|\text{roi}_\Delta\text{FA}_{WM,j,i}\|$; (B) $\|\text{roi}_\Delta\text{MD}_{WM,j,i}\|$; (C) $\|\text{roi}_\Delta\text{FA}_{WM,j,i}\|$; (D) $\|\text{roi}_\Delta\text{MD}_{WM,j,i}\|$.
3.3.3. Group-level altered WM mask

The mask comprising white matter skeleton voxels exhibiting significant alterations (pFWE<0.05; permutation t-test with 50,000 iterations) relative to FBA and/or NCA baseline, ROIALT, is depicted via three-view projection in Fig. 10. This group-level region represents 3.96% of the white matter skeleton, comprising 4398 of the 110,939 (interpolated) voxels in ROIWM. Note that no voxels were found to be significantly changed in FA from Test to Retest in NCA. The resulting altered white matter mask, ROIALT, intersects with 14 of the WM tracts defined in the JHU ICBM-DTI-81 atlas (Mori et al., 2005), as indicated in Table 5.

3.3.4. Subject-level longitudinal changes in altered WM mask

Repeated measures ANOVA and post hoc pairwise analysis of sessions confirmed that the altered white matter mask (ROIALT) was associated with statistically-significant effects of session in both mean FA and mean MD for FBA, but not for NCA (see Tables 6 and 7). Comparisons in FBA with In1 yielded the strongest observed effects, while the smallest session-wise changes were observed for Post. As observed for ROIWM, the spatial extents of substantial changes in both FA and MD within the altered white matter mask—||roi_ΔFAALT,j,i|| and ||roi_ΔMDALT,j,i||—were found to be significantly larger (p < 0.0001; t-test or Wilcoxon rank-sum test) at each

Table 4

| ROI extent | [||roi_ΔFAWM||] | [||roi_ΔFAWM||] | [||roi_ΔMDWM||] | [||roi_ΔMDWM||] |
|------------|----------------|----------------|----------------|----------------|
| Included HAEs | Pearson | Spearman | Pearson | Spearman | Pearson | Spearman | Pearson | Spearman |
| PLA > 20 g | 0.096 | 0.087 | 0.231 | 0.171 | 0.170 | 0.158 | 0.089 | 0.003 |
| PLA > 30 g | 0.100 | 0.069 | 0.201 | 0.154 | 0.130 | 0.138 | 0.111 | −0.015 |
| PLA > 40 g | 0.125 | 0.074 | 0.164 | 0.142 | 0.096 | 0.130 | 0.143 | −0.013 |
| PLA > 50 g | 0.122 | 0.058 | 0.122 | 0.109 | 0.057 | 0.091 | 0.127 | −0.021 |
| PLA > 60 g | 0.130 | 0.082 | 0.085 | 0.064 | 0.026 | 0.040 | 0.121 | 0.012 |
| PLA > 70 g | 0.131 | 0.103 | 0.102 | 0.113 | 0.033 | 0.076 | 0.113 | 0.018 |

* Correlation significant at the pBonferroni < 0.05 level.

Fig. 9. Linear predictions and associated confidence bands are superimposed on 183 FBA samples (61 subjects at each of three follow-up sessions) for (A) [||roi_ΔFAWM||]; (B) [||roi_ΔMDWM||]; (C) [||roi_ΔFAWM||]; (D) [||roi_ΔMDWM||]. A significant linear regression (Bonferroni-corrected; see Table 4) was found in FBA for the spatial extent of the ROIWM signed change mask associated with a decrease in FA (||roi_ΔFAWM||), as a function of the cumulative count of HAEs exceeding 20 g. This white matter change (decreased FA) is typically associated with neural injury (e.g., Wiesmann et al., 1999, Arfanakis et al., 2002, Kraus et al., 2007, Mac Donald et al., 2007a, 2007b; Mac Donald et al., 2007a, 2007b; Shitaka et al., 2011, Magnoni et al., 2015, Sundman et al., 2015, Pan et al., 2016, Kantarci et al., 2017). The symbol r represents a Pearson’s correlation coefficient.
follow-up session for FBA relative to NCA (Fig. 11).

Similarly, all signed change masks associated with ROI_{ALT} were found to be significantly larger in spatial extent ($p < 0.001$; t-test or Wilcoxon rank-sum test) at each follow-up session for FBA, relative to baseline (FBA: Pre; NCA: Test). White matter tracts in which ROI_{ALT} voxels were observed are listed in Table 5.

Table 5
White matter tracts (JHU ICBM-DTI-81 atlas) in which voxels from ROI_{ALT}, shown in Fig. 10, were located. The percentage of voxels comprising ROI_{ALT} that were located in each tract is indicated. (The total may not sum to 100% due to rounding.)

| White matter tract                          | Proportion of ROI_{ALT} | JHU WM atlas labels                        |
|---------------------------------------------|--------------------------|--------------------------------------------|
| Corpus callosum                             | 27.1%                    | Body of corpus callosum                    |
| Splenium of corpus callosum                 |                          | Genu of corpus callosum                    |
| Superior longitudinal fasciculus            | 26.0%                    | Superior longitudinal fasciculus R         |
| Limb of internal capsule                    | 17.8%                    | Anterior limb of internal capsule L        |
|                                             |                          | Posterior limb of internal capsule L       |
| Corona radiata                              | 17.7%                    | Anterior corona radiata L                 |
|                                             |                          | Superior corona radiata L                 |
| Superior corona radiata R                   |                          | Superior corona radiata L                 |
| Superior corona radiata R                   |                          | Superior corona radiata R                 |
| Cingulum (cingulate gyrus)                  | 8.6%                     | Cingulum (cingulate gyrus) L               |
| Superior fronto-occipital fasciculus        | 2.8%                     | Superior fronto-occipital fasciculus L     |
| Posterior thalamic radiation                | 0.1%                     | Posterior thalamic radiation R             |

L = Left hemisphere, R = Right hemisphere.

Table 6
Repeated measures ANOVA outcomes evaluating mean FA and mean MD in ROI_{ALT} across indicated sessions.

| Group | FA/MD | Sessions | F-statistic | Corrected-p |
|-------|-------|----------|-------------|-------------|
| NCA   | FA_{ALT} | All sessions | F(1, 14) = 0.21 | 0.651       |
|       | MD_{ALT} | All sessions | F(1, 14) = 0.48 | 0.504       |
| FBA   | FA_{ALT} | All sessions | F(3, 180) = 20.82 | $<10^{-4}$ |
|       | MD_{ALT} | All sessions | F(3, 180) = 7.91 | $<10^{-4}$ |

Where the sphericity assumption was violated, $p$-values were corrected with Huynh–Feldt’s method.

* Indicates a result significant at the (corrected) $p < 0.05$ level.

Similarly, all signed change masks associated with ROI_{ALT} were found to be significantly larger in spatial extent ($p < 0.001$; t-test or Wilcoxon rank-sum test) at each follow-up session for FBA, relative to NCA (Fig. 11).

Change mask spatial extents within ROI_{ALT} exhibited greater linkage to HAE exposure than for ROI_{WM}. As seen in Table 8, correlations significant at the $p_{\text{Bonferroni}} < 0.05$ level were observed between accrual of HAEs (at thresholds of 20 g, 30 g, and 40 g) and the spatial extent of the signed change masks in the altered white matter.

Regressions against the HAE count at 20 g, the threshold associated with the most significant correlation with signed change mask spatial extents, are shown for all signed change masks in Fig. 13. Consistent
Post hoc pairwise t-test comparisons of session-wise mean FA and mean MD measurements in ROIs of FBA as acquired over all sessions, given a significant effect of session observed in Table 6.

| Group  | FA/MD | Pairwise comparisons | t-statistic | Corrected-p |
|--------|-------|----------------------|-------------|-------------|
| FBA    | FA_ALT | In1 vs. Pre          | 7.84        | <10⁻⁴       |
|        |        | In2 vs. Pre          | 4.10        | <10⁻⁴       |
|        |        | Post vs. Pre         | 3.19        | 0.002       |
|        |        | In2 vs. In1          | −3.75       | <10⁻³       |
|        |        | Post vs. In1         | −4.65       | <10⁻⁴       |
|        |        | Post vs. In2         | −0.90       | 0.367       |
|        | MD_ALT | In1 vs. Pre          | −4.83       | <10⁻⁴       |
|        |        | In2 vs. Pre          | −2.69       | 0.014       |
|        |        | Post vs. Pre         | −1.97       | 0.056       |
|        |        | In2 vs. In1          | 2.23        | 0.027       |
|        |        | Post vs. In1         | 2.86        | 0.007       |
|        |        | Post vs. In2         | 0.63        | 0.532       |

Duncan’s method was used to correct for multiple comparisons. (Note that a positive sign for t indicates that the measurement in first listed session is greater than the measurement in the second.)

* Indicates a result significant at the (corrected) p < 0.05 level.

with Table 8, the regression fits against the to-date accumulation of HAEs exceeding 20 g for \( ||\text{ROI}_{\Delta FA_{ALT}}|| \) and \( ||\text{ROI}_{\Delta MD_{ALT}}|| \) were found to be statistically significant.

4. Discussion

Retrospective examination of DWI data, collected longitudinally over single seasons of participation by male high school athletes, revealed that athletes who experience repetitive HAEs exhibit greater changes in white matter diffusivity than athletes who do not experience such HAEs, and that these white matter changes were significantly correlated with the longitudinal accumulation of exposure to HAEs. A novel regional analysis of signed changes in diffusion measures provided enhanced granularity of pathophysiology detection beyond that achieved through traditional whole-brain or regional averages. It facilitated identification of spatial extents of white matter exhibiting significant alteration in FA and/or MD (ROI ALT) at the group level were predominantly (53% of the volume) located in the corpus callosum (Arfanakis et al., 2002; Inglese et al., 2005; Bazarian et al., 2007, 2014; Kraus et al., 2007; Lipton et al., 2008; Wilde et al., 2008; Mayer et al., 2010, 2012; Cubon et al., 2011; Koerte et al., 2012; Ling et al., 2012b; McAllister et al., 2012, 2014; Gajawelli et al., 2013; Gong et al., 2018; Mustafi et al., 2018) and superior longitudinal fasciculus (Kraus et al., 2007; Cubon et al., 2011; Koerte et al., 2012; Mayer et al., 2012; Bazarian et al., 2014; Bahrami et al., 2016; Manning et al., 2017; Myer et al., 2018; Yuan et al., 2018).

From a biomechanical perspective, blows randomly incident on the head will exhibit a primary intersection of induced strain in the vicinity of the center of the brain (Corsellis et al., 1973; Gurdijian and Gurdijian, 1976; Ji et al., 2014). Therefore, centrally-located tracts are a priori expected to be at greater risk of accumulation of strain (and associated pathophysiology) from exposure to repeated HAEs arising from blows to the head or body.

Longitudinal changes in diffusion measures are unlikely to have arisen

#### Table 7

Post hoc pairwise t-test comparisons of session-wise mean FA and mean MD measurements in ROIs of FBA as acquired over all sessions, given a significant effect of session observed in Table 6.

| Group  | FA/MD | Pairwise comparisons | t-statistic | Corrected-p |
|--------|-------|----------------------|-------------|-------------|
| FBA    | FA_ALT | In1 vs. Pre          | 7.84        | <10⁻⁴       |
|        |        | In2 vs. Pre          | 4.10        | <10⁻⁴       |
|        |        | Post vs. Pre         | 3.19        | 0.002       |
|        |        | In2 vs. In1          | −3.75       | <10⁻³       |
|        |        | Post vs. In1         | −4.65       | <10⁻⁴       |
|        |        | Post vs. In2         | −0.90       | 0.367       |
|        | MD_ALT | In1 vs. Pre          | −4.83       | <10⁻⁴       |
|        |        | In2 vs. Pre          | −2.69       | 0.014       |
|        |        | Post vs. Pre         | −1.97       | 0.056       |
|        |        | In2 vs. In1          | 2.23        | 0.027       |
|        |        | Post vs. In1         | 2.86        | 0.007       |
|        |        | Post vs. In2         | 0.63        | 0.532       |

Duncan’s method was used to correct for multiple comparisons. (Note that a positive sign for t indicates that the measurement in first listed session is greater than the measurement in the second.)

* Indicates a result significant at the (corrected) p < 0.05 level.

#### Fig. 11.

At all follow-up sessions, football athletes (FBA) exhibited significantly (p < 0.001) greater spatial extents of substantial changes in FA and MD, in ROI ALT, than did noncollision athletes (NCA). Box-and-whisker plots are presented at each follow-up session for (A) \( ||\text{ROI}_{\Delta FA}\_alt|| \); (B) \( ||\text{ROI}_{\Delta MD}\_alt|| \).
from either normal white matter development or differences in the level of physical activity across the groups, across assessment times. Given that these spatial extents were defined based on the test-retest variability of the NCA population over a time-window (average = 8 weeks) comparable with the inter-session period (average = 6 weeks) for the FBA, it is unlikely that the substantive changes observed in FBA from Pre to In1 may be explained by normal development. Further, given that both groups are actively involved in conditioning activities at the time of their baseline sessions, these changes in FBA are less likely to be the consequence of altered levels of physical activity.

### 4.2. Implied HAE-induced damage accumulation

The signed change masks, focusing on the spatial extent of the brain exhibiting a particular direction of deviation in white matter diffusion, may help us explain why previous studies of FA and/or MD have obtained variable outcomes. Our findings suggest that we are observing a continuous white matter injury and repair process. This potential white matter tract injury (tissue damage) may begin with inflammation or axonal swelling (associated with increased FA) and progresses to axonal injury (associated with decreased FA).
Conversely, when examining only the mean effect on FA or MD, athletes early in the process may balance athletes in the later stages, and the mean observed alteration in diffusion measures would be expected to be close to zero. See Table 3 and Figs. 5 and 6, in which the average change in mean FA over the entire white matter skeleton (i.e., ROIWM) would be expected to be slightly positive at $In1$ and $In2$, but to approach zero at $Post$. The opposite trend is observed for mean MD. Thus, within an athlete, different magnitudes or signs of change in mean FA or mean MD could be observed, depending on the time since initiation of the injury and repair process. Given the rate at which athletes accumulate HAEs depends on a wide range of factors (e.g., position, playing time, style of play, competition level), a population assessment at the end of a season captures the averaged effects of this potential white matter injury process.

The overall hypothesis of an active injury and repair process is supported by our observation of multiple, sparsely-distributed locations in the white matter that exhibit either increases or decreases in FA and MD. When examined on a regional basis, these distributions generally led to changes in FA and MD that fluctuated between increases and decreases across sessions. Thus, some studies (e.g., McAllister et al., 2012; Lipton et al., 2013; Bahrami et al., 2016) may have captured post-season measures at a point when regions with increased FA may have balanced those regions exhibiting decreased FA (e.g., see $Post$ sessions for FA-related mask spatial extents in Fig. 8). Therefore, this lack of mean change in diffusion measures may not imply the absence of underlying disturbances in axonal architecture or tissue injury.

4.3. Limitations

In a study of athletes involved in collision-based sports, it must also be acknowledged that participants may be exposed to HAEs outside the known times of practices and games, possibly through non-sanctioned play or involvement in other collision-based activities (e.g., club sports). Coupled with such additional potential exposure, complete knowledge of the use of anti-inflammatory drugs (e.g., aspirin, ibuprofen, naproxen) is unknown, and such medications could affect measurements of diffusion, particularly if the associated control of inflammation is variable across measurement sessions. Although the study incorporates an age-appropriate and gender-matched non-collision sport control population that adjusts for confounds such as environmental factors (Tian et al., 1998; Dechent et al., 1999; Babb et al., 2004) and exercise (Maddock et al., 2011), it would be desirable to achieve a greater balance of racial and ethnic categories across the subject populations. Further, individuals in this age bracket are experiencing many biological changes, including rapid growth of the brain (Giorgio et al., 2010; Lebel and Beaulieu, 2011; Lebel et al., 2012; Simmonds et al., 2014), for which collection of a longer-term longitudinal dataset for the NCA population would facilitate more powerful parallel comparisons.

From a technical perspective, the exact relationship between diffusion MRI markers of white matter and underlying tissue damage is still a matter of debate. Myelin and cellular membranes each play a role in restricting water diffusion in the nervous system (Barkovich, 2000;
Lancaster et al., 2003; Sotak, 2004), with cellular membranes bounding water between pools of different mobility. Thus, there may not exist a strict one-to-one relationship between a given structural alteration and a particular MR measure. For example, FA may increase as a result of restricted axial diffusivity, facilitated parallel diffusivity, or some combination of the two. In addition, the use of traditional, single b-value diffusion weighted imaging is partially limiting (e.g., to correct susceptibility distortions).

We note that the location of significantly altered white matter is unexpectedly structured, running from the left frontal lobe through the corpus callosum to the right parietal lobe. In theory, one might expect such a region to be more uniformly distributed if the athletes are experiencing an unbiased distribution of impacts to the head. Given that the HAE monitoring devices used in this study have not been shown to provide meaningful source location information, we cannot effectively assess whether the incoming blows were uniformly distributed or biased toward one side.

Complex white matter structures (e.g., crossing fibers) will potentially invalidate the assumption of a primary orientation of fibers, which is normally assumed to be represented by the diffusion tensor’s main eigenvector (Mori and Tournier, 2013). A longitudinal approach that applies a more powerful technique such as diffusion kurtosis imaging (e.g., Davenport et al., 2016) could provide additional insight.

5. Conclusion

This study documents that changes in diffusivity of white matter in high school-aged athletes participating in American football are correlated with accumulation of HAEs throughout the course of a season of participation. This correlation provides evidence of a potential mechanically-initiated white matter injury and repair process. Such a process is consistent with evidence from previous work involving both neuroimaging and HAE monitoring (e.g., Breedlove et al., 2012; Lipton et al., 2013; Bazarai et al., 2014; Davenport et al., 2014, 2016; Bahrami et al., 2016; Svaldi et al., 2018; Bari et al., 2018) suggesting deviations of measures of brain structure and function are correlated with aggregate HAE exposure. While there may be a threshold beyond which physiologic function is acutely altered (Svaldi et al., 2018; Bari et al., 2018), this study demonstrates that the potential damage accumulation is more gradual and represents the cumulative effect of possibly all HAE exposures. Future effort should be directed at reducing the number and magnitude of events experienced by collision-sport athletes, whether through enhanced protective equipment, improved technique instruction, or modification of rules.

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