

# Article Enhanced μCT Imaging Protocol to Enable High-Resolution 3D Visualization of Microdamage in Rat Vertebrae

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Featured Application: The following work suggests enhanced  $\mu$ CT image acquisition parameters for high-resolution visualization of barium sulfate contrast in bone samples for use in investigations of microdamage in three dimensions.

Abstract: Contrast-enhanced  $\mu$ CT imaging has been used to provide non-destructive 3D images of microdamage, but at a lower quality than found in histology and 2D backscatter electron (BSE) imaging. This study aimed to quantify potential improvements in microdamage characterization by enhancing  $\mu$ CT scanning parameters. Eleven slides from 9 rat vertebrae (healthy = 3, osteolytic metastases = 3, mixed metastases = 3) previously stained for microdamage with  $BaSO_4$  and analyzed with BSE imaging ( $2\mu$ m voxel spacing) were used in this study.  $\mu$ CT imaging conducted under varying protocols (x-ray voltage, tube current, frame averaging) demonstrated enhanced scan parameters at 90 kVp, 44 µA, 0.5 mm aluminum filter, 8 times frame averaging, and 4.9 µm voxel spacing. Post-processing with Richardson-Lucy deconvolution further deblurred the µCT images. Labeled microdamage in the baseline, enhanced and deblurred µCT images were segmented and spatially quantified vs. BSE-labeled microdamage using a probability-based correlation metric at six inflation radii. Enhanced µCT scan parameters improved damage visualization and increased spatial correlation probability with BSE images. Deblurring improved the sharpness of stain boundaries but did not significantly improve spatial correlation probabilities in comparison to the enhanced scans. This enhanced µCT protocol facilitates 3D visualization of microdamage, an indicator of bone quality important to bone damage mechanics.

Keywords: bone; micro-computed tomography (µCT); microdamage; spatial correlation

# 1. Introduction

Microcracks and damage occur in bone during normal use, with damage playing a role in regulating bone turnover. Homeostasis in the healthy bone between osteoblastic and osteoclastic cell activity maintains bone integrity, replacing damaged bone tissue and adjusting the bone shape to mechanical loading. Diseases (cancer, osteoporosis) or treatment (bisphosphonates, radiation) are known to affect damage distribution by affecting bone turnover or bone material properties [1–4]. Skeletal microdamage has previously been studied with two-dimensional (2D) histological analyses, including light microscopy with basic fuchsin staining [5–7] or chelating fluorochromes [8]. Backscatter electron (BSE) microscopy with lead-uranyl acetate [9] or barium sulfate ( $BaSO_4$ ) stain [10–12] has also been used for high-resolution 2D imaging of skeletal microdamage, with the difference in atomic number between the labeling stain and the bony matrix providing excellent contrast for identifying damaged regions within bone tissue. While histologic analyses are the most common method of quantifying microdamage, both histology and BSE are destructive techniques limited to 2D analysis.



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Distinct from optical microscopes, scanning electron microscopes (SEMs) use electrons (rather than light) to produce images of specimens. BSEs are high-energy electrons that are reflected out of the specimen during SEM imaging that highlight the different elements contained in the specimen. Denser elements (those with a high atomic number) deflect incident electrons more strongly than lighter elements, thus appearing brighter in a BSE image acquired with an SEM [13]. This characteristic of BSE images allows for the study of mineralization in bone tissue, as regions high in hydroxyapatite (a mineral composed of calcium, phosphate, and hydroxide) appear bright. Visualizing high atomic number contrast agents is facilitated by BSE imaging as the bright contrast may be easily located and segmented from the images. SEM imaging acquires images under a vacuum. Thus, specimens must be dehydrated, or a cryo-fixation technique must be employed [13,14]. Bones' inorganic phase and relatively low water content compared to soft biological tissues allow for the dehydration method without the need for the more complicated cryo-fixation. The dehydration process, however, requires weeks of progressive dehydration to limit artifacts (microcracks) induced by drying [13]. As the surface topology of the samples must be flat, bone specimens require embedding in resin and subsequent surface polishing. This can further induce artifacts and requires additional time for resin polymerization [13]. To avoid static charge buildup on the sample surface, non-conductive samples such as bone must be rendered conductive. Strategies to achieve this include impregnation of the sample with heavy metals, applying a thin conductive coating (such as carbon), or use of ionic liquids [13].

Computed tomography is a non-invasive imaging tool providing three-dimensional (3D) images of biological structures. Micro-computed tomography ( $\mu$ CT) allows for high-resolution images on a smaller scale, which has value for pre-clinical studies of biological tissues.  $\mu$ CT leverages the varying x-ray attenuations of biological tissues to generate images [15]. As the sample is rotated inside the  $\mu$ CT scanner (or the detector is rotated around the specimen), the intensity of x-rays transmitted through the tissues at different angles is measured by the detector [15,16]. The collection of 2D projections is then reconstructed post-acquisition to create the 3D image using a process called back-projection [15]. As the intensity of tissues in  $\mu$ CT images is determined by radiodensity (the relative inability of x-rays to pass through a material),  $\mu$ CT is best suited for distinguishing different types of tissues, such as bone versus muscle. The use of radiopaque contrast agents allows for otherwise undetectable structures (vasculature, bone microdamage) to be visualized with  $\mu$ CT imaging [16,17]. Images acquired with  $\mu$ CT can be acquired in as little as 20 min to 12 h depending on the sample size, desired voxel spacing, and image resolution. Biological samples may also remain hydrated and intact, as  $\mu$ CT scans are acquired at atmospheric pressure.

 $\mu$ CT image quality is determined by the image acquisition parameters and reconstruction algorithm employed by the scanner but may further be improved with post-image acquisition techniques.  $\mu$ CT images inherently contain spatial blurring, which can be quantified by a point spread function (PSF) [18–20]. The PSF defines image resolution and is the result of blurring induced by the finite size of the focal spot size, the x-ray detector aperture (the spatial resolution of the detector itself) [21], and the scattering of x-rays [22]. Spatial blurring may result in thin bone features (such as trabeculae) having a diffuse appearance, which overestimates their thickness and underestimates their intensity. For microstructural analyses of bone, this may negatively affect the quality of results. The localization and quantification of contrast agents may additionally be disrupted, as blurring may reduce their intensity. The application of deconvolution algorithms, which act to reduce blurring and noise, may correct the PSF to render images more closely to their true representations.

Non-destructive 3D assessment of damage is possible via  $\mu$ CT imaging with radioopaque contrast agents. Previous investigators have studied 3D skeletal microdamage distribution using  $\mu$ CT imaging of *BaSO*<sub>4</sub>-labeled bovine and rodent bone [12,23,24]. Microdamage accumulation found in  $\mu$ CT images of *BaSO*<sub>4</sub>-labeled human cortical bone cores has been strongly correlated with histologic crack density but with more variability [10]. BSE imaging of *BaSO*<sub>4</sub>-labeled rodent vertebrae [25] has been shown to yield superior visualization of microcracks and crack nucleation surfaces compared to high-resolution  $\mu$ CT imaging.

Trabecular bone contains a fine structure of mineralized tissue with individual strut thickness that can vary from 25–1000  $\mu$ m. Microdamage and cracks present within trabecular bone are on the order of microns to millimeters. As such,  $\mu$ CT-based microdamage assessments in trabecular bone are highly dependent on image acquisition parameters and the imaging system, given the overlap of resolution possible with  $\mu$ CT and the structures being imaged. This study aims to enhance  $\mu$ CT-based microdamage characterization in rat vertebrae labeled with *BaSO*<sub>4</sub> contrast by careful consideration of the x-ray physics (k-edge x-ray adsorption), physical specimen size, limitations of  $\mu$ CT systems (focal spot size interaction with tube current and voltage), and post-acquisition enhancement (deblurring) [18,26], and compare patterns of microdamage visualized with  $\mu$ CT to BSE imaging.

#### 2. Materials and Methods

## 2.1. Sample Generation

This protocol was approved by the animal care committee at the University Health Network prior to implementation. Healthy (n = 3) and metastatically involved (osteolytic (n = 3) mixed osteolytic/osteoblastic (n = 3)) athymic female rat vertebrae generated from previous work were used in this study [25]. The rodent model for the physiological development of vertebral metastases in immunocompromised rats was established previously [27–30]. Briefly, 5–6-week-old athymic female rats were randomly assigned to healthy, osteolytic metastatic, or mixed osteolytic/osteoblastic metastatic groups. Osteolytic or mixed metastases were generated with HeLa human cervical cancer cells or canine Ace-1 prostate cancer cells, respectively. The animals were anesthetized with nose-cone inhalation of a 2% isoflurane/air mixture, and  $\sim 1.5 \times 10^6$  cells (in 0.2 mL of Dulbecco's modified eagle medium/nutrient mixture F-12 media) were injected into the left heart ventricle. The animals were euthanized 21 days after cell injection via CO<sub>2</sub> asphyxiation, and the T12-L2 vertebral motion segments were separated, wrapped in saline-soaked gauze, and stored at -20 °C until use. These healthy and metastatically involved vertebrae provide a wide range with respect to the presence of microdamage accumulation. T13-L2 spinal motion segments were stained with 0.5 M barium chloride followed by 0.5 M sodium sulfate solution, each for 72 h, under vacuum [10]. Microdamage was induced in the L1 vertebra under axial compressive loading (50 N held for 3 h) with a loading rate of 35  $\mu$ m/s. The L1 vertebrae were separated and re-stained with BaSO<sub>4</sub> post-loading to label load-induced microdamage [25,31]. Staining was repeated in [17] to correlate load-induced microdamage with stresses and strains calculated with micro-finite element models. However, the method described here is focused solely on correlating BaSO<sub>4</sub> distribution across imaging modalities. The post-loaded specimens were used in the analyses to maximize the amount of damage present in the bone.

#### 2.2. Backscatter Electron Imaging and Baseline µCT Imaging

Previous work prepared the L1 vertebrae for BSE imaging by fixing the samples in 2% paraformaldehyde and dehydrating them in rising concentrations of ethanol immersed in an osteo-bed kit [25]. The sample blocks were polymerized, secured to a slide with epoxy, cut in the sagittal plane with a water-cooled precision diamond saw (Isomet<sup>®</sup> low-speed saw, Buehler, Lake Bluff, IL, USA), and polished with increasing grits of silicon carbide paper and diamond paste. Slides of the sagittal cross-sections of hard-embedded L1 vertebrae were carbon-coated, and BSE images were acquired at ×150 magnification with 2  $\mu$ m/pixel spacing (SS BSE detector, FEI, Hillsboro, OR, USA) using a Philips XL30 ESEM (FEI). Eleven BSE slides (healthy = 5, osteolytic = 3, mixed = 3) were  $\mu$ CT imaged ( $\mu$ CT-100, Scanco Medical, Brüttisellen, Switzerland) at baseline scan parameters (55 kVp, 200  $\mu$ A, 0.5 mm Al filter, 250 ms integration time, no frame averaging, 11.4  $\mu$ m voxel spacing) [25]. Image processing of the BSE and  $\mu$ CT images to quantify microdamage is described in Section 2.4.

#### 2.3. Enhanced µCT Imaging

These 11 slides prepared for BSE imaging were re-imaged with parameters selected to maximize contrast between BaSO<sub>4</sub> and bone, also considering voxel spacing, resolution, and signal-to-noise ratio (SNR). SNR is proportional to the square root of the products of current, integration time, and frame averaging. The contrast between *BaSO*<sub>4</sub> and bone is enhanced by maximizing the relative energy contribution of x-ray photons having energy just above the 37.4 keV k-edge of barium. The peak of the x-ray photon energy spectrum as a function of x-ray energy occurs at  $\sim 1/3$  of the tube voltage value. Selecting a tube voltage of 90 kVp, the maximum available, will therefore increase the fraction of x-ray photons with energy just below the k-edge of barium, maximizing contrast. By the same logic, the attenuation of lower energy photons also improves contrast and can be achieved by using an attenuating material with a sufficiently high atomic number such that photoelectric x-ray absorption becomes the dominant absorption process. Photoelectric absorption is more pronounced at lower photon energy, thereby removing the undesired lower energy photons. Thus, the 0.5 mm aluminum filter (1 of 3 options) used in the baseline scans sufficiently reduces the contribution of lower energy photons, reducing soft tissue contrast and beam hardening. This filter is suitable to be added to the x-ray beam for an enhanced set of image acquisition parameters.

Reconstruction voxel spacing requirements are governed by the size of details of interest but must also consider computation limitations like processing time, memory, and storage, limiting the number of voxels. The image resolution should also be commensurate with voxel spacing, as storing a blurry image with a small pixel spacing is wasteful. The specification of smaller reconstructed voxel spacings limits the maximum sample holder diameter because the  $\mu$ CT uses geometric magnification, trading off the field of view to increase resolution at the detector. In this application, a voxel spacing of 4.9  $\mu$ m (1 of 7 choices) was chosen with the sample holder of 14 mm inner diameter, the maximum value having an x-ray projection that can be encompassed by the detector.

A higher x-ray tube current creates more x-ray photons per unit time, thereby increasing the SNR achievable per unit time of imaging. This requires a larger x-ray focal spot so as not to damage the x-ray anode due to excessive energy deposition density, increasing x-ray projection blurring at the detector. The tube current was thus limited to a value of 44  $\mu$ A (1 of 5 choices available), so the associated (dependent) focal spot size does not cause noticeable blurring.

Finally, the process of analyzing images and comparing modalities require µCT images with a minimum SNR, which depends on the square root of the number of detected xray photons for a given x-ray spectrum. This number, for a given x-ray tube current, is proportional to the amount of time that the x-ray detector collects radiation and is also known as the integration time [32,33]. The Scanco  $\mu$ CT-100 additionally allows several images to be acquired at each projection angle and combined, a process labeled as frame averaging. Thus, the number of detected x-ray photons used to image a sample with a 2000 ms integration time and no averaging is the same as imaging with a 250 ms integration time and 8 times frame averaging. The former approach will, however, lead to a faster scan time. Frame averaging is useful when SNR requirements cannot be achieved by increasing the integration time alone. In these experiments, we chose to only vary frame averaging for descriptive simplicity. The SNR was evaluated by examining images with the above-specified parameters; frame averaging was increased up to 8 times to improve the SNR to offset limitations encountered with the use of low current. Note: frame averaging greatly increases the scan time, which may not always be practical depending on the type and number of samples to be analyzed.

The x-ray beam is sufficiently penetrating that no spectral adjustments (kVp, filtration) are required to compensate for limited attenuation differences of the samples of interest due to variations in extent and morphology. Compensations for said changes in attenuation can be achieved by varying the total x-ray exposure time and/or tube current alone. Larger

x-ray focal spots associated with larger tube currents might be tolerable if the larger sample is also imaged with a larger voxel spacing, so blurring is not noticeable.

Semi-quantitative evaluation criteria focused on maximum contrast between the  $BaSO_4$  and bone without saturation, sharpness of trabecular bone boundaries, presence of artifacts (beam hardening, scatter, noise), and graininess of the images to tune the  $\mu$ CT image acquisition parameters.

#### 2.4. Image Processing and Deblurring Algorithm

 $\mu$ CT images were cropped to only include five slices from the top face of the BSE slides to facilitate comparison and reduce computational expense [34]. 3D Slicer software (3D Slicer 4.8.1) was used to transform, resample, and register the five  $\mu$ CT slices and individual BSE images of each sample. A single  $\mu$ CT plane was extracted from registration with the BSE image. Registration error was measured using fiducials placed on paired BSE and  $\mu$ CT images. Five fiducials were placed on all image pairs, and the distance between fiducials was averaged for each sample.

An established deblurring algorithm (based on Richardson-Lucy deconvolution) was applied to the enhanced  $\mu$ CT images [18,26]. The PSF was modeled as a Gaussian and using a custom module (https://bitbucket.org/OrthopaedicBiomechanicsLab/deblurring, accessed on 20 October 2020), the in- and out-of-plane blurring were estimated by placing profiles (implemented as rulers) across thin regions of cortical bone. Profile locations were chosen as thin cortical bone structures with non-bone regions on both sides (air or marrow) such that there were three distinct layers, or greyscale values, across the profile. In-plane blurring was estimated by profiles placed in the sagittal plane, and out-of-plane blurring was estimated by profiles placed in the coronal and axial planes. Theoretical PSF should be equal in identically acquired images (same  $\mu$ CT scanner using the same parameters). Thus, an average was calculated based on the estimated in-plane blurring in each image  $(0.0067 \pm 0.0007 \text{ mm})$ . Due to their thinness (~500 µm), the slides did not have enough visible structures in the coronal and axial directions to estimate out-of-plane blurring. A whole fresh frozen vertebra (age-matched, identical HeLa cell injection protocol) was  $\mu$ CT imaged with the enhanced scan parameters. Out-of-plane blurring was estimated from this image as 0.0068 mm. All enhanced  $\mu$ CT images were deblurred using the average in-plane and single out-of-plane blurring parameters as inputs to the Richardson-Lucy deconvolution algorithm.

The *BaSO*<sub>4</sub> contrast was segmented from the  $\mu$ CT (baseline, enhanced (Figure 1a), and deblurred) and BSE images (3D Slicer segmentation editor). First, the whole bone was segmented with automated thresholding using the Otsu method [35–37]. The segmentation was shrunk by 50  $\mu$ m from the outer edge (Figure 1b, light green) of the cortical shell to remove non-specific contrast not caused by microdamage [10,11]. *BaSO*<sub>4</sub> was segmented inside the shrunken label field (Figure 1c, light blue) using thresholding at a greyscale intensity of 253 for the BSE images and ~2500 mgHA/cm<sup>3</sup> for the  $\mu$ CT images. The BSE images were 8-bit images with no intensity calibration for *BaSO*<sub>4</sub>. As the difference in density between barium and calcium is high, and the greyscale intensity bins are limited to the 8-bit range for the BSE images, the pixel intensities corresponding to *BaSO*<sub>4</sub> were concentrated around the upper intensities of the image. As such, pixel intensities of 253 or above were sufficient in segmenting *BaSO*<sub>4</sub> from the BSE images.



**Figure 1.** Stain segmentation procedure on an enhanced  $\mu$ CT image. (**a**) unlabeled enhanced  $\mu$ CT image with background removed; (**b**) bone label field (blue) less than 50  $\mu$ m of the outer edge of the cortical shell (light green); (**c**) labeled *BaSO*<sub>4</sub> (light blue) used for damage volume fraction calculations and showing excluded *BaSO*<sub>4</sub> contrast (green).

#### 2.5. Spatial Correlation

The spatial correlation between labeled microdamage in paired BSE and  $\mu$ CT images (i.e., from the same sample) was determined using a probability-based method [38,39]. All calculations and comparisons were performed within image pairs. For each sample, there were four imaging modalities: baseline  $\mu$ CT parameters, enhanced  $\mu$ CT parameters, enhanced µCT parameters with deblurring, and BSE imaging. The probability of observing labeled microdamage in a  $\mu$ CT image within some radius of labeled microdamage in a BSE image was determined for all labeled voxels. Additionally, the probability of observing labeled microdamage in a BSE image within some radius of labeled microdamage in a  $\mu$ CT image was determined. Probabilities were calculated at radii of 0, 10, 20, 30, 40, 50, and  $60 \mu m$  from labeled pixels, chosen based on the trabecular thickness and the  $\mu CT$  spatial resolution. To demonstrate, Figure 2 shows the label fields used for the probability of spatial correlation calculations at a 30 µm radius between a BSE and enhanced µCT image. To model varying radii considered as correlated,  $BaSO_4$  labeled pixels in one image (i.e., the BSE image in Figure 2a, the enhanced  $\mu$ CT image in Figure 2b) was "inflated" using convolution with circular kernels with radii of  $10-60 \ \mu m$ . The probability of observing spatially close pixels was calculated as the relative number of intersecting BaSO<sub>4</sub>-labeled pixels in paired images. As the inflation radius increases, more pixels in the inflated image are labeled as  $BaSO_4$ , thus increasing the likelihood that a  $BaSO_4$ -labeled pixel in the noninflated image will intersect. The probability is calculated twice for each pair to ensure mutual correlation. Figure 2a shows the label fields used to calculate the probability that  $BaSO_4$ -labeled pixels in an enhanced  $\mu$ CT image fall within 30  $\mu$ m of  $BaSO_4$ -labeled pixels in a BSE image. Figure 2b shows the label fields used to calculate the probability for the opposite direction, namely that BaSO<sub>4</sub>-labeled pixels in a BSE image fall within 30 μm of  $BaSO_4$ -labeled pixels in an enhanced  $\mu$ CT image.



(a)



(b)

**Figure 2.** Label fields used to calculate spatial correlation probability of  $BaSO_4$ -labeled pixels between BSE and enhanced µCT images at an inflation radius of 30 µm. (**a**) the probability of  $BaSO_4$ -labeled pixels in an enhanced µCT image (blue) being within 30 µm of  $BaSO_4$ -labeled pixels in an inflated BSE image (white). The probability of spatial correlation is calculated as the ratio of red to blue pixels. (**b**) the probability of  $BaSO_4$ -labeled pixels in a BSE image (white) being within 30 µm of  $BaSO_4$ -labeled pixels in an inflated enhanced µCT image (blue). Red areas represent overlapping  $BaSO_4$ -labeled pixels in both images. The probability of spatial correlation is calculated as the ratio of red to white pixels. Damage area fraction (DAF) was defined as the ratio of intersecting bone and  $BaSO_4$  pixels to bone pixels, described in Equation (1):

$$DAF = \frac{\#(p_{BaSO_4} \cap p_{bone})}{\#p_{bone}} \tag{1}$$

where  $p_{BaSO_4}$  is the set of all  $BaSO_4$ -labeled pixels and  $p_{bone}$  is the set of all bone-labeled pixels. DAF was calculated for every modality after each convolution for inflation.

#### 2.6. Statistical Analysis

Measures of DAF exhibited non-normal residual distributions; therefore, a Friedman non-parametric test was used with sample number as a random effect. Post-hoc comparisons were performed using the Nemenyi test (RStudio). Mean microdamage correlation probability in both correlation directions (BSE to  $\mu$ CT and  $\mu$ CT to BSE) was compared between image pair groups with a two-way ANOVA with sample number as a random effect. Tukey pairwise post-hoc comparisons were performed between each image pair at each radius. The level of significance for all tests was set at 0.05.

## 3. Results

## 3.1. Enhanced µCT Imaging and Deblurring

Varying currents (all other parameters fixed, 70 kVp, no frame averaging) demonstrated that lower currents improved the visualization of trabecular edges. Increasing frame averaging using a fixed low current improved the SNR. Increased voltage improved the contrast between  $BaSO_4$  and bone while maintaining low current and high frame averaging. Tuned  $\mu$ CT parameters (90 kVp, 44  $\mu$ A, 0.5 mm Al filter, 200 ms integration time, 8 frame averaging, 4.9 µm voxel spacing) enhanced visualization of the damaged regions compared to the baseline  $\mu$ CT scan parameters. Table 1 shows a summary of the baseline and enhanced  $\mu$ CT acquisition parameters. The current was kept low enough for the tube to operate with a small focal spot size to yield sharp images. Due to the limitations with respect to the current, frame averaging was increased to improve damage visualization. Each scan took 2.1 h collecting images for a volume of  $\sim$ 500 µm in height. The enhanced  $\mu$ CT images show microdamage that is obscured by reduced resolution and low contrast in the baseline  $\mu$ CT images (Figure 3). The deblurring algorithm visually sharpened the regions of the  $BaSO_4$  contrast agent seen in the enhanced  $\mu$ CT images and identified smaller areas of *BaSO*<sup>4</sup> not picked up in the enhanced image. The mean registration error between paired BSE and  $\mu$ CT images (baseline, enhanced, deblurred) was 0.01  $\pm$  0.008 mm.

**Table 1.** Summary of  $\mu$ CT acquisition parameters for baseline and enhanced protocols for visualization of *BaSO*<sub>4</sub>-labeled microdamage in rat vertebral bone.

Acquisition Protocol	Voltage (kVp)	Tube Current (µA)	Filter	Integration Time (ms)	Frame Averaging	Voxel Spacing (µm)
Baseline	55	200	0.5 mm Al	250	N/A	11.4
Enhanced	90	44	0.5 mm Al	200	8	4.9



**Figure 3.** BaSO<sub>4</sub> visualization of microdamage in trabecular bone of a healthy rat L1 vertebra with four imaging modalities. (a) baseline  $\mu$ CT; (b) enhanced  $\mu$ CT; (c) enhanced  $\mu$ CT with a deblurring algorithm applied post-acquisition; (d) BSE imaging. The red arrow identified matching *BaSO*<sub>4</sub> stained structures in all images.

## 3.2. Damage Area Fraction

DAF in all image types increased with inflation radius (Figure 4). Uninflated (radius = 0  $\mu$ m), DAF in the baseline  $\mu$ CT scans is significantly higher than the enhanced and BSE images (p = 0.026, p < 0.001, respectively). However, for every non-zero inflation radius, the BSE and deblurred  $\mu$ CT images have significantly higher DAFs than the baseline and enhanced images.

Convolving (or inflating) the  $BaSO_4$ -labeled pixels blur the damage, which increases the area identified as  $BaSO_4$ . Higher stain areas are measured on the BSE images when there is inflation as smaller areas of  $BaSO_4$ -labeled pixels are connected, forming larger regions. In  $\mu$ CT images, more of the  $BaSO_4$ -labeled pixels are clumped in regions representing larger areas of damage; pixels inflated at the center of a clump do not contribute to additional stain.

While DAF is equal across the enhanced, deblurred, and BSE images without inflation, the deblurred and BSE images likely identify smaller areas of stain and estimate smaller areas for concentrated *BaSO*<sub>4</sub> clumps than the enhanced images.



**Figure 4.** Damage area fractions ( $BaSO_4$  area/bone area) of  $BaSO_4$  label fields in  $\mu$ CT and BSE images before (radius = 0  $\mu$ m) and after convolution inflating the label fields at six radii. The data represented are from eleven sagittal BSE slides from rat L1 vertebrae (healthy = 5, osteolytic = 3, mixed = 3). \* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001, \*\*\*\* p < 0.001, Nemenyi pair-wise comparison.

## 3.3. Damage Spatial Correlation

Considering both directions of probability calculations, the enhanced and deblurred  $\mu$ CT imaging showed a greater spatial correlation of damage with BSE than the baseline  $\mu$ CT imaging (BSE-labeled damage occurring near  $\mu$ CT labeled damage and vice versa). As expected, spatial correlation probability between image pairs (BSE and baseline  $\mu$ CT, BSE and enhanced  $\mu$ CT, BSE and deblurred  $\mu$ CT) increased with larger inflation radii (Figure 5). Post-hoc pairwise comparison showed the probability of finding *BaSO*<sub>4</sub>-labeled pixels in enhanced  $\mu$ CT images near *BaSO*<sub>4</sub>-labeled pixels in BSE images was significantly greater than in the baseline  $\mu$ CT images for inflation radii of 10–40  $\mu$ m (p = 0.0002–0.045, Figure 5a). When BSE images were inflated (Figure 5a), the deblurring algorithm showed no significant differences compared to enhanced  $\mu$ CT at any inflation radii (p = 0.32–0.99). The probability of spatial correlation approached 1 when the BSE images were inflated (Figure 5a), as the DAF of BSE images also approached 1 (due to the inflation of many small, distributed pixels throughout the specimens), increasing the likelihood of finding a  $\mu$ CT pixel in this area.

The probability of finding a  $BaSO_4$ -labeled pixel in a BSE image spatially close to one in a  $\mu$ CT image was significantly different for each  $\mu$ CT modality at all inflation radii. Post-hoc pairwise comparison showed the deblurring algorithm to have the highest probability of spatial correlation compared to baseline and enhanced images at all inflation radii (p < 0.0001-0.002, Figure 5b). For both directions, there were no significant differences across  $\mu$ CT modalities when no convolution was applied (p = 0.082-0.99, radius = 0  $\mu$ m).



Image Pair 👼 Baseline µCT and BSE 🧔 Enhanced µCT and BSE 🔄 Deblurred µCT and BSE

**Figure 5.** Probability of spatial correlation of  $BaSO_4$ -labeled pixels in paired µCT (baseline, enhanced, and deblurred) and BSE images at six inflation radii. (**a**) the probability that  $BaSO_4$ -labeled pixels in a µCT image (baseline, enhanced, or deblurred) fall within some radius of  $BaSO_4$ -labeled pixels in a BSE image; (**b**) the probability that  $BaSO_4$ -labeled pixels in a BSE image; (**b**) the probability that  $BaSO_4$ -labeled pixels in a BSE image; (**b**) the probability that  $BaSO_4$ -labeled pixels in a BSE image; (**b**) the probability that  $BaSO_4$ -labeled pixels in a BSE image; (**b**) the probability that  $BaSO_4$ -labeled pixels in a BSE image; (**b**) the probability that  $BaSO_4$ -labeled pixels in a BSE image fall within some radius of  $BaSO_4$ -labeled pixels in a µCT image (baseline, enhanced, or deblurred). The data presented are from eleven sagittal BSE slides from rat L1 vertebrae (healthy = 5, osteolytic = 3, mixed = 3). \* p < 0.05, \*\* p < 0.01, \*\*\* p < 0.001, \*\*\*\* p < 0.001, Tukey pair-wise comparison.

## 4. Discussion

Enhancements to  $\mu$ CT scanning parameters and deblurring post-acquisition both qualitatively and quantitatively improved visualization of damage deposition. Enhanced parameter voxel spacing of 4.9 µm represents higher resolution imaging than historically used for 3D microdamage analysis with *BaSO*<sub>4</sub> labeling [10,25]. Visualization of microdamage with this µCT protocol is dependent on the efficacy of the *BaSO*<sub>4</sub> staining (ability to precipitate into all microdamage sites). This is an important consideration, as contrary to BSE imaging, the enhanced µCT parameters are not able to distinguish microdamage that is not labeled by the *BaSO*<sub>4</sub> contrast. *BaSO*<sub>4</sub> labels all damage types (microfractures, diffuse damage, linear microcracks), limiting the specificity of damage characterization by intensity alone; the enhanced techniques presented in this investigation potentially allow the examination of morphology to determine the damage type. Effective assessment of damage distribution in 3D allows for the study of how damage interacts with remodeling, prediction of failure, and treatment responses.

This investigation studied trabeculae with thicknesses of 75–95  $\mu$ m and spacing of 90–126  $\mu$ m [25], which may explain the lack of difference between the enhanced techniques and baseline  $\mu$ CT imaging at larger inflation radii (Figure 5a). Our tuning of  $\mu$ CT acquisition parameters was limited by pre-set voltage and current combinations available on the scanner, preventing any determination if higher voltage could further increase the contrast between bone and *BaSO*<sub>4</sub>. Spatial correlation accuracy was impacted by registration errors between BSE and  $\mu$ CT image pairs, limiting the strength of findings at the smaller inflation

radii. Frame averaging and small voxel spacing in the enhanced scan parameters resulted in long scan times and large file sizes for whole rat vertebrae (~11 h, ~8 GB for ~1500 slices), which is not suitable for samples without fixation. The deblurring algorithm requires extensive computational resources that may be unavailable for total volume deblurring using the enhanced parameters.

Eight times frame averaging is a major contributor to increased scan acquisition time in this protocol. However, the reduced noise in these images facilitates the segmentation of small  $BaSO_4$ -labeled regions. Increasing integration time may have further reduced noise, however, at the expense of greater scan acquisition times when combined with eight times averaging. Thus, this study did not investigate the magnitude of noise reduction when increasing integration time and frame averaging concurrently. While the acquisition time is long, the process for obtaining BSE images is extensive, requiring weeks for sample dehydration and hard embedding (5 weeks total for the BSE slides used in this study). The enhanced  $\mu$ CT protocol allows results to be obtained by the next day, is non-destructive to the sample, and provides 3D spatial location and distribution of damage. Scan acquisition time and file size prevents the enhanced  $\mu$ CT parameters outlined in this protocol from being used for the microdamage analysis of whole bones from large mammals or humans. However, bone cores (trabecular and cortical) are often used to study microdamage distribution in larger bone samples [10,35,40,41], which would be feasible for this protocol.

#### 5. Conclusions

Enhancing  $\mu$ CT protocols provides high-resolution visualization of  $BaSO_4$ -labeled microdamage that spatially locates  $BaSO_4$  near BSE-identified microdamage. 3D visualization of microdamage allows global observation of accumulation and distribution, which could influence further crack propagation and potential fracture locations. The use of a deblurring algorithm may be beneficial but may not be practical for high-resolution  $\mu$ CT scans because of the significant computational expense. Further research regarding factors affecting microdamage accumulation and its influence on bone material properties, including fracture behavior, can be facilitated with this protocol.

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