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- 1 Original Full Length Article
- ² Development and optimization of a high-throughput micro-computed
- ³ tomography imaging method incorporating a novel analysis technique to
- ⁴ evaluate bone mineral density of arthritic joints in a rodent model of
- ⁵ collagen induced arthritis

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51 Introduction

Rheumatoid arthritis (RA) is a chronic and progressive autoimmune disease in which the immune system triggers multiple inflammatory responses against self-antigens, resulting in erosion of cartilage and bone in and around the peripheral joints [1]. The pathogenesis of RA includes a complex inflammatory response involving innate and adaptive immune cells, proinflammatory cytokines and autoantibodies that

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ABSTRACT

Rheumatoid arthritis (RA) is a chronic autoimmune disease resulting in joint inflammation, pain, and eventual 31 bone loss. Bone loss and remodeling caused by symmetric polyarthritis, the hallmark of RA, is readily detectable 32 by bone mineral density (BMD) measurement using micro-CT. Abnormalities in these measurements over time 33 reflect the underlying pathophysiology of the bone. To evaluate the efficacy of anti-rheumatic agents in animal 34 models of arthritis, we developed a high throughput knee and ankle joint imaging assay to measure BMD as a 35 translational biomarker. A bone sample holder was custom designed for micro-CT scanning, which significantly 36 increased assay throughput. Batch processing 3-dimensional image reconstruction, followed by automated 37 image cropping, significantly reduced image processing time. In addition, we developed a novel, automated 38 image analysis method to measure BMD and bone volume of knee and ankle joints. These improvements signif- 39 icantly increased the throughput of ex vivo bone sample analysis, reducing data turnaround from 5 days to 24 h 40 for a study with 200 rat hind limbs. Taken together, our data demonstrate that BMD, as quantified by micro-CT, is 41 a robust efficacy biomarker with a high degree of sensitivity. Our innovative approach toward evaluation of BMD 42 using optimized image acquisition and novel image processing techniques in preclinical models of RA enables 43 high throughput assessment of anti-rheumatic agents offering a powerful tool for drug discovery. $\overline{44}$ © 2013 The Authors. Published by Elsevier Inc. This is an open access article under the CC BY-NC-ND license

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infiltrate the synovia causing fluid accumulation, pain, and bone dam- 58 age in the affected joints [2,3]. 59

Animal models of RA have been used extensively to interrogate 60 the distinct mechanisms of disease pathology and identify potential 61 biological targets in pursuit of novel therapeutics. Collagen induced 62 arthritis (CIA) is a widely used rodent model of induced arthritis. The 63 rat CIA model exhibits multiple facets of human disease including pro- 64 found cartilage degradation, dependence on complement immunity, 65 periarticular inflammation, and bone resorption [4]. Following collagen 66 injection, rats display a severe polyarthritic phenotype consisting of 67 swollen extremities, cartilage degradation, and eventual loss of joint 68 function, which is, in some aspects, similar to RA [5,6]. The reproducibil- 69 ity, low variability, and rapid disease onset of this model make it a 70

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71valuable tool for the assessment of novel compounds [4,7–9]. Hind limb 72paw thickness is a routinely used surrogate marker of inflammatory pathology in this model [5,8,9]. In the rat CIA model, approximately 80% of 73 74 animals develop visually observable edema in the ankles and paws, while about 20% of animals develop visually observable edema in the 7576knee [8-10]. However, even when edema is not present, cellular infiltra-77 tion, cartilage degradation, and structural changes to bone occur at the 78knee and other affected joints [8,10]. While useful, paw thickness does 79not reveal the underlying structural changes to the bone caused by dis-80 ease since paw thickness only directly measures edema. Assessment of 81 bone morphology by imaging is a more sensitive and translational read-82 out; damage to its structural integrity, and in particular, periarticular demineralization, points to hyper-activation of osteoclast activity, 83 84 which is a critical component of RA [2].

Over the last two decades, imaging has made major advances in early 85 diagnosis and therapeutic monitoring of RA [11]. In the clinic, X-ray 86 radiography is the standard for imaging-assisted diagnosis of RA. Also 87 based on X-ray technology, computed tomography (CT) has demon-88 strated higher sensitivity for quantifying bone erosions in RA patients, 89 due to its 3-D visualization perspective [12]. It is ideal for bone measure-90 ments due to its high spatial resolution and the natural contrast be-91 tween bone and soft tissue. A reduction in bone mineral density 9293 (BMD), an important biomarker of disease assessed by CT, indicates 94 poor prognosis for patients with early stages of RA [13–15]. CT has also been demonstrated to be an excellent tool for assessment of bone dam-95age in preclinical rodent models of RA due to its high reproducibility and 96 inherently quantifiable properties [16,17]. Radiological examination has 9798 been applied successfully to monitor progression of RA and osteoarthritis in various rodent models to gain a deeper understanding of the path-99 ophysiology of disease [18-25]. Many metrics are used to track disease 100 progression, including bone volume changes, bone surface roughness, 101 102erosion scoring, 3-D tissue morphometry, and BMD measurements, as 103measured by X-ray, DEXA, or micro-CT [18-23,25]. Despite widespread use of both micro-CT and DEXA to gather BMD measurements, it has 104 been demonstrated, in at least one animal model, that the volumetric 105BMD measured by micro-CT is more sensitive than the 2-D BMD mea-106 surements acquired by DEXA [24]. 107

Automated analysis can dramatically increase throughput, so special 108 attention has been given to performance tuning the analysis procedure 109by developing new automation methods. For example, Barck et al. de-110 scribe an automated method to measure BMD of the paw joints in 111 112 mice and Huber et al. describe automated BMD measurements of femur in human [23,26]. However, other biomarkers, such as changes 113 in bone architecture, are more heavily emphasized in the literature on 114 automated bone analysis. There are examples of automation techniques 115 that can separate trabecular bone from cortical bone or segment trabec-116 117 ular bone using high resolution images [27,28]. An in vivo method using registration techniques to detect differences in bone lesion volume in a 118 rat model using magnetic resonance has also been described [29]. 119Segmenting specific bone components, such as trabecular or cortical 120bone, and then measuring shape and attributes such as 2-D area and 1211223-D volume, or bone erosion are the main analysis endpoints in these 123publications. Of note, the automation techniques in these studies were not able to avoid manually separating their regions of interest (ROIs) 124from their samples; instead, ROIs were either prepared prior to imaging 125or were manually segmented from a larger area after imaging and be-126127fore automated image analysis was used to extract measurements.

Herein, we describe a high-throughput method to evaluate BMD and 128track volume changes of peripheral joints for pharmacological assess-129 ment of drug candidates to support discovery of novel therapeutics. 130Our objective was to apply BMD measurements of the knee and ankle 131 joints in the rat CIA model as a rapid and efficient biomarker for drug 132screening and optimization. The methods we detail had to replicate 133 the sensitivity and accuracy of slower acquisition processes, while 134 speeding up work-flow. The automated analysis method also needed 135136 to correlate with manual analysis and show high reproducibility from study to study. To achieve this goal, we developed a streamlined process 137 to image eight bone samples simultaneously and perform batch image 138 reconstruction and automated image cropping. In addition, we demon- 139 strate a novel automated method of combining image processing 140 techniques, such as intensity thresholding and skeletonization, with 141 mathematical techniques in curve fitting and curvature calculations, to 142 find and place a bounding box around the ROIs in CT images quickly 143 and consistently. The algorithm can process individual images or entire 144 data sets and provides various metrics of interest including volume and 145 mean intensity of Hounsfield units of the bone ROI within the bounding 146 box. This manuscript further expounds new methods of data acquisition 147 and analysis that utilize the predictive potential of BMD assessment as it 148 relates to RA outcome and therapeutic treatment in the rat CIA model. 149

Materials and methods

Development

Micro-CT: acquisition, image reconstruction, and cropping

Conical tubes containing fixed rat hind limbs stored in 70% ethanol 153 solution were loaded into the micro-CT holder. Samples were posi- 154 tioned symmetrically on the perimeter of the holder for equal X-ray 155 beam exposure, with a phantom located in the middle of the holder 156 (Fig. 1B). CT scans were acquired as described in Haines et al. (2009) 157 with the exception that the three dimensions of the image data set 158 (X, Y, and Z) were adjusted to $1000 \times 900 \times 1300$ slices at 100 μm 159 cubic voxel dimensions and scaled to Hounsfield units (HU) [30]. The 160 reconstructed data were then cropped into images containing a single 161 bone sample, using MATLAB (MathWorks, Inc., Natick, MA, USA) soft- 162 ware to automate this process. 163

Automated image analysis and manual correction

The CT images were downsampled and thresholded to 300 HU. We 165 chose this range to be consistent between studies and because it 166 showed the best dynamic range while still excluding soft tissue. Impor-167 tantly, the bone samples on which this threshold is applied undergo 168 rapid formalin fixation and are then stored in ethanol filled conical 169 tubes. This threshold is designed to take the tubes, scanning holder, 170 and ethanol into account. After thresholding, a morphological closing 171 operation was applied to the bone mask to fill the holes, and this resul- 172 tant mask was skeletonized to obtain the main shape of the hind limb 173 bone. A 6th degree polynomial curve was fitted using a Chebyshev poly- 174 nomial, as detailed previously [31]. The curvature of the skeleton was 175 calculated and the peaks of maximum curvature were found, using a 176 previously published equation [32]. Using the location of maximum cur- 177 vature, segmentation of the knees and ankles was performed by placing 178 predefined boxes of sizes [X, Y, Z] around each ROI. The two knee and 179 ankle joint regions of interest at X, Y, and Z were set to $15 \times 15 \times 180$ 15 mm and $13 \times 13 \times 13$ mm, respectively. These bounding boxes 181 were placed by aligning the center point of the turning point of the 182 joint (Figs. 2A, B, C). After box placement, the bone inside is quantified 183 using thresholding to obtain volume and HU data for our desired final 184 output: BMD of knee and ankle regions. By thresholding the box, only 185 the bone is selected as our ROI; air and other materials are excluded 186 from our measurements. We evaluate BMD as bone mineral content 187 (mg)/bone region (ROI in cm³) and track volume of the ROI as well. 188 After automated analysis was complete, cross-sectional views of the 189 boxes were exported as 2-D image files for visual inspection to ensure 190 that the algorithm worked properly (Fig. 2D) and a master Excel sheet 191 was generated containing the volumes and average HUs for the ROIs 192 (Fig. 2E). Bounding box placement was visually reviewed and samples 193 were manually re-analyzed if the boxes were misplaced or if samples 194 were not analyzed correctly using Amira® 5.4.2 image analysis software 195 (Mercury Computer Systems, Inc., Chelmsford, MA). Detailed methods 196 are provided in the Supplementary methods section. 197

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