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## **The contribution of bone and cartilage to the near-infrared spectrum of osteochondral tissue**

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#### **Abstract**

Near-infrared (NIR) spectroscopy has been utilized to assess hyaline cartilage quality in human and animal osteochondral tissues. However, due to the lack of NIR signal from bone phosphate, and the relatively deep penetration depth of the radiation, the separate contributions of cartilage and bone to the spectral signatures have not been well defined. The objectives of the current study were 1) to improve the understanding of the contributions of bone and cartilage to NIR spectra acquired from osteochondral tissue, and 2) to assess the ability of this non-destructive method to predict cartilage thickness and modified Mankin grade of human tibial plateau articular cartilage. NIR spectra were acquired from samples of bovine bone and cartilage with varying thicknesses, and from twenty-two tibial plateaus harvested from patients undergoing knee replacement surgery. Spectra were recorded from regions of the tibial plateaus with varying degrees of degradation, and the cartilage thickness and modified Mankin grade of these regions were assessed histologically. Spectra from bone and cartilage samples of known thicknesses were investigated to identify spectral regions that were distinct for these two tissues. Univariate and multivariate linear regression methods were used to correlate modified Mankin grade and cartilage thickness with NIR spectral changes. The ratio of the NIR absorbances associated with water at 5270 and 7085 cm−1 were the best differentiator of cartilage and bone spectra. NIR prediction models for thickness and Mankin grade calculated using partial least squares regression were more accurate than univariate-based prediction models, with root mean square errors of cross validation of 0.42 mm (thickness) and 1.3 (modified Mankin grade), respectively. We conclude that NIR spectroscopy may be used to simultaneously assess articular cartilage thickness and modified Mankin grade, based in part on differences in spectral contributions from bone and cartilage.

#### **Keywords**

osteoarthritis; partial least squares regression; near-infrared spectroscopy; bone; cartilage

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#### **Introduction**

Clinical osteoarthritis (OA) has a large and increasing prevalence in the United States. Twenty one million people over 25 years old were estimated to have OA in 2001 (7% of the 2001 US population<sup>1</sup>), and 26.9 million people in 2008 (9% of the 2008 US population<sup>1</sup>).<sup>2</sup> Estimates of OA affliction are expected to continue to grow as age and obesity increase within the population.<sup>2–6</sup> Osteoarthritis may be debilitating and in later stages may only be effectively treated by replacing the damaged joint. In the 65–74 year old age category OA is the fifth most common reason for disability.<sup>7</sup> Reduction or reversal of OA progression in the initial stages of development would be the optimal treatment option.

Current methods for diagnosing OA include arthroscopy, radiography and magnetic resonance imaging  $(MRI)$ .<sup>8–11</sup> Arthroscopy, a minimally invasive technique, provides visual information for assessing OA development, including the color and texture of articular cartilage  $(AC)$ .<sup>12,13</sup> Radiography informs about joint space narrowing, joint space width, subchondral sclerosis, marginal osteophytes and malalignment.<sup>14–16</sup> Magnetic resonance imaging (MRI) provides similar information to radiography, although the three dimensional perspective allows for a better overview of the joint space.<sup>17, 18</sup> In addition, all tissues involved in OA (cartilage, bone, menisci) are observable using MRI. Advanced OA may be identified by any of these three methods; however, early stage OA is more difficult to diagnose with physical techniques.<sup>19, 20</sup>

The matrix composition of articular cartilage changes throughout the onset and development of OA. During the development of OA the collagen network of articular cartilage is disrupted, water content increases and proteoglycan content decreases, resulting in a loss of mechanical stress resistance.<sup>20, 21</sup> Histology, biochemistry and gravimetrics (water assessment) have been used to identify subtle compositional changes of articular cartilage during the onset of OA in animal models.<sup>22</sup> Histologically, Mankin grading,<sup>22</sup> and OARSI  $\gamma^{23, 24}$  are two grading systems that are widely used to assess severity of tissue degradation. While these techniques are informative and sometimes precise, they require harvesting and destruction of the tissues of interest, and thus are typically only utilized in pre-clinical studies. An ideal early diagnosis method would provide data on tissue status using a rapid and non-destructive, minimally (or non-) invasive analyses.

Infrared (IR) spectroscopy, a technique based on the absorption of radiation that causes molecular vibrations, has been used to assess organic materials, and has proven to be useful in studies of OA.<sup>25–31</sup> Both mid-IR spectroscopy (frequency range 400–4000 cm<sup>-1</sup>) and near-IR (NIR) spectroscopy (frequency range 4000–13,000 cm−1) have been utilized to assess cartilage degradation.<sup>25–28, 30–32</sup> Mid-IR spectra are characterized by relatively sharp peaks that may be attributed to specific chemical functionality within the molecules of the sample, and the spectra for bone and cartilage are very well defined.<sup>25, 33</sup> In contrast, NIR spectra are characterized by broad and overlapping peaks that are mostly attributed to C–H, N–H and O–H chemical functionality, and have only recently had specific molecular species assigned to features in cartilage spectra.<sup>31, 34, 35</sup> Interestingly, there is no NIR signal from the phosphate molecules in bone, whereas in the mid-IR region, the phosphate absorbance is one of the largest. However, NIR spectroscopy has the distinct advantage of penetrating to a

greater sample depth (mm to cm) compared to mid-IR spectroscopy.36–38 Non-destructive mid-IR analyses are confined to the surface of samples, and a maximum depth on the order of ~2 μm is interrogated when using attenuated total reflectance (ATR) mid-IR spectroscopy.  $36$ 

Near-IR spectroscopy has previously proven to be useful for studying articular cartilage degradation,  $28-31$ ,  $35$ ,  $39-44$  including in a surgical environment.  $28-31$  Importantly, a significant correlation has been found between the International Cartilage Repair Society (ICRS) grading of cartilage and a ratio of peak areas in a NIR spectrum of cartilage.<sup>31, 35</sup> Furthermore, partial least squares regression (PLSR) has been used to correlate NIR spectra of bovine articular cartilage with thickness,40 and NIR spectra of rat articular cartilage with thickness,  $39$  Mankin grade,  $39$  subchondral bone volume<sup>41</sup> and mineral density.<sup>41</sup> The spectral basis for these correlations has not been well defined, particularly with respect to the contribution of subchondral bone. In the current study, we investigated NIR spectral contributions from bone and cartilage, and demonstrated that PLSR can be used to predict modified Mankin grade and thickness of human knee articular cartilage from NIR spectral data.

#### **Materials and methods**

#### **Tissues**

**Human tissues—**Twenty two tibial plateaus from male and female patients undergoing knee replacement surgery (ages 46–87) were harvested with institutional (Hospital for Special Surgery, New York, USA) IRB approval. Harvested tibial plateaus were removed from the operating room and taken immediately to an adjacent laboratory for infrared spectroscopic data collection. Tissues were kept moist during data collection, and synovial fluid was cleaned from the cartilage surface with 0.9% saline.

**Bovine Tissues—**Bovine nasal cartilage (BNC), articular cartilage, and bone (supplied by JBS, Souderton, PA, and by Research 87, Boylston, MA) were used for characterizing the depth of NIR penetration, and for characterizing the NIR spectrum of cartilage and bone. For the depth penetration studies, 14 sections of BNC were prepared, ranging in thickness from 0.37 to 5.45 mm. BNC is a hyaline cartilage that has a more homogeneous composition compared to articular cartilage, and thus was used for assessment of penetration to minimize compositional differences as a confounding factor. Each BNC section was mounted on a non-reflective surface for NIR data collection (three spectra for each section). The nonreflective surface was then lined with parafilm and a second set of NIR spectra were collected. Parafilm has a unique peak at  $5774 \text{ cm}^{-1}$  that was used to assess depth of penetration.

Bovine articular cartilage and bone samples were harvested from a 2–3 week old bovine stifle joint using a 5 mm diameter biopsy punch. The bone and cartilage of the punched sample were separated and NIR spectra were recorded from each tissue. Five samples that were approximately 5 mm thick total but contained different amounts of bone and cartilage were generated for assessment of the underlying bone contribution to the cartilage spectra; the ratio of bone to cartilage thickness ranged from 0.2 to 2.7.

#### **Near-infrared spectroscopy**

Diffuse reflectance NIR spectroscopy was performed using a Remspec NIR probe (Charlton, MA) coupled to a matrix-F spectrometer (Bruker Optics, Billerica, MA). Data were collected with OPUS version 5.5 (Bruker Optik GmbH, Ettlingen, Germany). The probe was placed in contact with the cartilage or bone surface. Spectra were the sum of 32 co-added scans across the spectral range of 12,000–4000 cm<sup>-1</sup> with a spectral resolution of 16 cm<sup>-1</sup>. Background spectra of air were collected with a mirror before each series of measurements.

Several regions were sampled from each tibial plateau (medial and lateral) and three spectra were collected from each region. A total of 35 sampling sites were identified for a total of 105 spectra. Triplicate spectra were also collected from each BNC section (with and without parafilm beneath) and from the bovine articular cartilage and bone samples. Bovine nasal and articular cartilage, and bone samples were placed on a non-reflective background when spectra were collected.

#### **Histology**

The 35 tibial plateau regions that were evaluated spectroscopically were also sectioned for histology. The cartilage and subchondral bone were removed from each region (that had been spectroscopically sampled) with a 5 mm diameter biopsy punch, and then immersed in a mixture of 80% ethanol and 1% cetylpyridinium chloride. The samples were subsequently decalcified in a 10% EDTA/Tris buffer solution, embedded in paraffin and sectioned to a thickness of 7 μm onto glass slides. These sections were stained with safarin-O, hematoxylin and eosin, and graded according to a modified Mankin scale.45 The Mankin grading system is a standard for the assessment of cartilage and incorporates grading based on structure fissuring, cell cloning, loss of proteoglycans, and tidemark integrity. In the modified scale, a grade of 0 represents normal cartilage and a grade of 13 represents severely degenerated cartilage (Table I).46, 47 Two investigators independently evaluated samples that had been blind-coded and randomized. The average of these two grades was used as the final modified Mankin grade. The Cohen's Kappa index was calculated using the 'wkappa' function (psych package version  $1.2.4$ )<sup>48</sup> in R version  $2.15.2^{49}$  to assess the level of agreement between the two investigators. The thickness of cartilage was measured from the histology images using Image J version  $1.45s$ ;<sup>50</sup> the lengthes of ten lines drawn from the cartilage surface to the bone spanning the width  $(-5 \text{ mm})$  of the section were averaged.

#### **Data analysis**

Spectral preprocessing, principal component analysis (PCA) and partial least squares regression (PLSR) were performed with Unscrambler X (CAMO Inc, Norway), and peak integrations were performed in OPUS version 5.5. PCA was used to identify spectral outliers in the tibial plateau dataset; two outlying spectra were identified in the principal component one (PC1) versus PC2 scores plot. Visual inspection of the raw data indicated these spectra had a low signal to noise level and were subsequently removed from the dataset. Spectra from each tibial plateau region were truncated to 11,000–4000 cm<sup>-1</sup>, averaged and baseline offset corrected, prior to the calculation of absorbance areas in the following spectral regions: 8820–8060, 8695–8197, 7460–6780, 7280–6040, 5716–5496 cm−1. Relationships between the integrated band intensities or ratios thereof (8820–8060/5716–5496, 7460–

6780/8695–8197, 7280–6040/8820–8060, 7280–6040/5716–5496) and modified Mankin grade or thickness, were evaluated using Spearman correlation with a  $P < 0.05$  level of significance. The integrated bands were attributed to CH stretching second overtones associated with matrix components (8820–8060 and 8695–8197 cm<sup>-1</sup>), OH stretching first overtones primarily associated with water (7460–6780 and 7280–6040 cm−1) and CH stretching first overtones (5716–5496 cm<sup>-1</sup>).<sup>51</sup> The two spectral regions, 8695–8197 and 7280–6040 cm<sup>-1</sup> were integrated for direct comparison to the study of Spahn et al.<sup>35</sup> The height of five peaks (8640, 8266, 6418, 6025, 5824 cm<sup>-1</sup>) were evaluated in spectra that were preprocessed from 11,000–4000 cm−1 using standard normal variate (SNV) and Savitzky-Golay second derivatives (third order polynomial, 339 cm−1 window: 2dSG3,339). These peaks were attributed to CH stretching second overtones (8640 and 8266 cm<sup>-1</sup>), NH stretching first overtones (6418 cm<sup>-1</sup>) and CH stretching first overtones (6025 and 5824  $\text{cm}^{-1}$ ).<sup>51</sup> In addition, the minimum peak height in the spectral ranges 8660–8600, 7100– 7040 and 5290–5240 cm−1 and the wavenumber of this minimum peak height were investigated. Ratios of these minimum peaks heights were calculated: 5290–5240/7100– 7040, 5290–5240/8660–8600 and 7100–7040/8660–8600 cm−1. The relationships between these peak heights and ratios with thickness or modified Mankin grade were tested using Spearman correlation with a 0.05 level of significance.

PLSR prediction models for modified Mankin grade and cartilage thickness were calculated using the 9000–4400 cm<sup>-1</sup> spectral region. These spectral data were preprocessed using the standard normal variate transform followed by calculating the Savitzky-Golay (SG) second derivatives (third order polynomial) using one of three window widths: 170, 255 and 339 cm−1. Large window widths were required for second derivative preprocessing to adequately smooth the NIR data. Mean centering was applied prior to all PLSR calculations.

Modified Mankin grading and cartilage thickness were analysed with PLSR, a multivariate linear regression method based on the calculation of latent variables which represent the maximum covariance between the X (spectral) and Y (modified Mankin grade or cartilage thickness) matrices.<sup>52</sup> PLSR prediction models were validated using an eight segment cross validation procedure; samples were randomly assigned to each segment. Outlying samples were identified using Hotelling  $T^2$  residuals; these spectra were further investigated. Prediction models were evaluated on the basis of the number of factors required, coefficient of correlation of calibration and cross validation and root mean square error of calibration and cross validation (RMSEC and RMSECV respectively; equation 1).

$$
RMSE\!=\!\sqrt{\frac{\sum_{i=1}^{i=N}(Y_{i,Predicted}\!-\!Y_{i,Measured})^2}{N}}\quad \text{Equation 1}
$$

where N is the number of samples.

#### **Results**

#### **Sample characteristics**

The averaged modified Mankin grades of the dataset ranged from 2 to 11, with a median of 6.5. The difference between the two investigators' grades ranged from 0 to 3, the mean difference was 1. The inter-rater agreement was moderate as the linearly weighted kappa value was 0.6.53 Cartilage thickness of the samples ranged from 0.95 to 3.68 mm with an average of 2.35 mm (standard deviation 0.65 mm). Within a sample the smallest relative standard deviation (RSD) of cartilage thickness was 0.82% and the largest was 30.33%, the average sample RSD was 9.34%.

#### **NIR penetration depth into cartilage**

The unique parafilm peak at  $5774 \text{ cm}^{-1}$  was readily apparent in the second derivative of the NIR spectrum (Fig. 1). The intensity of this peak decreased with increasing thickness of the cartilage layer overlaying the parafilm section, and was not observed when samples were greater than 5 mm thick. Thus, the penetration depth of the NIR radiation was estimated to be under 5 mm in this spectral region.

#### **Features of cartilage NIR spectra**

NIR spectra of cartilage were characterized by a broad absorption band in the range 5450– 4300 cm<sup>-1</sup>, a broad and intense peak around 7000 cm<sup>-1</sup>, a broad and low intensity band around 8450 cm<sup>-1</sup>, and minor peaks between 4300–4000 cm<sup>-1</sup>, and between 6000–5450  $cm^{-1}$  (Fig. 2a). The band centered around 5200 cm<sup>-1</sup> has been attributed to a combination of 'free' water and bound water, while the absorption peak in the  $7000 \text{ cm}^{-1}$  spectral region has been attributed to free water.<sup>34</sup> The NIR spectra of bone had absorbances in the same spectral regions as cartilage, however, the peaks were less broad than in cartilage. Peaks at 5270 and 7085 cm−1 were particularly intense in second derivative-transformed spectra of cartilage, with lower intensity bands observed at 4675, 5624, 6653, and 8640 cm−1 (Fig. 2b). Second-derivative transformed spectra of bone had low intensity peaks at 5624 and 8645, and dominant peaks at 5220 and 7050  $\text{cm}^{-1}$ .

The ratio of these two bands informs about the ratio of bound and free water in a sample, and the 5270:7085 cm−1 ratio was higher in bone when compared with cartilage (Fig. 2c). For samples with both cartilage and bone, the ratio of 5270 to 7085 cm−1 increased as the ratio of bone thickness to cartilage thickness increased (samples in the studied series had the same total thickness). Several bands of changing intensity were also identified in the second derivative spectra of degraded cartilage samples (Fig. 3), including 8266, 6025 and 5824 cm−1, which have been attributed to C-H stretching second and first overtones (cartilage sections, spectra and  $2<sup>nd</sup>$  derivative spectra: Fig 3a, b, c, respectively).<sup>51</sup> Within bone spectra the bands around 5624 and 8645 cm−1 arise from bone collagen, and the 5220 and 7085 cm−1 are attributed to water.

Although there was a negative trend between modified Mankin grade and cartilage thickness,  $(R = -0.24)$ , the correlation did not reach significance  $(P = 0.20)$ . However, both cartilage thickness and modified Mankin grade were significantly correlated with several

peak areas and second derivative parameters. R values ranged from −0.57 to 0.66 for thickness and −0.56 to 0.61 for modified Mankin grade with respect to peak area parameters and −0.75 to 0.70 for thickness and −0.60 to 0.36 for modified Mankin grade with respect to second derivative peak height parameters (Table II).

The PLSR model for the prediction of cartilage thickness, using spectra pretreated with SNV and second derivative transformation (third order,  $170 \text{ cm}^{-1}$  window), used three factors and had a calibration  $R^2 = 0.75$ , RMSEC = 0.32 mm, validation  $R^2 = 0.61$  and RMSECV = 0.42 mm. The model for the prediction of modified Mankin grade, using spectra pretreated with SNV and second derivative transformation (third order, 339 cm<sup>-1</sup> window), used five factors and had a calibration  $R^2 = 0.84$ , RMSEC = 0.9, validation  $R^2 = 0.64$  and RMSECV = 1.3. A large difference between RMSEC and RMSECV is indicative of an over-fit model, and a comparison of these parameters indicate that the models were not over-fit; the RMSECV was 70 and 80% of the RMSEC for modified Mankin grade and thickness respectively. The loading weights of the three factors of the cartilage thickness model were dominated by the 5300 and 7100 cm−1 spectral regions (Fig. 4). In the modified Mankin grade model, in addition to the water absorbances, other lower wavenumber features attributable to matrix absorbances were also strong contributors to the loading weights.

### **Discussion**

Near-IR light penetrates up to 5 mm into a cartilage sample in the 5774 cm<sup>-1</sup> spectral region. Though the penetration depth will vary with respect to frequency (wavelength) with higher frequencies having a greater penetration depth it is likely the near infrared wavelengths used here interact with all four zones of articular cartilage from the superficial to the calcified cartilage zone, as well as the underlying subchondral bone. The spectra in our study are therefore informative about the chemical and physical composition of cartilage beyond the direct interface between the probe tip and the sample surface; cartilage in our study ranged in thickness from 0.95 to 3.68 mm. Furthermore, the articular cartilage in the medial condyle of a healthy human knee is 2.92 mm thick on average.<sup>54</sup> Hence, the diffuse reflectance spectra collected from each tibial plateau sampled the entire thickness of cartilage and interrogated the underlying subchondral bone. Although the penetration depth of NIR spectroscopy may offer a substantial advantage over the limited penetration of attenuated total internal reflection mid-IR spectroscopy, the NIR signal cannot directly separate the bone and cartilage contributions, and represents an average of the tissues being sampled.

Clinical osteoarthritis disrupts the extracellular matrix of articular cartilage, degrading the cartilage surface and reducing the cartilage thickness. Despite this known relationship, we did not recover a strong, significant correlation between cartilage thickness and modified Mankin grade. Instead, we observed substantial variation in the thickness of tibial plateau cartilage at different modified Mankin grades. Thus, cartilage thickness and modified Mankin grade may not correlate in patients with osteoarthritis. This is consistent with a previous study that showed substantial variation in the thickness of normal tibial plateau cartilage (average minimum = 0.35,  $\sigma$  = 0.27 mm; average maximum = 6.25 mm,  $\sigma$  = 0.27 mm;  $n = 12$ ,<sup>54</sup> and highlights the patient to patient variation.

Nevertheless, correlations were sought between NIR spectral parameters and thickness or modified Mankin grade. Here, the spectral parameters of interest were peak areas calculated from baseline offset corrected spectra, and peak heights obtained from second derivative spectra. Modified Mankin grade was most strongly correlated ( $R = 0.61$ ,  $P < 0.05$ ) with a peak area ratio representing the ratio of water to organic matrix in the sample (7280– 6040/8820–8060 cm−1). This positive correlation tracks the increase in water content relative to proteoglycan and collagen content that occurs with OA. The correlation coefficient between modified Mankin grade and the peak area ratio of 7460–6780 to 8695– 8197 cm<sup>-1</sup> (R = 0.60) was similar to a previous study based on ovine knee articular cartilage  $(R = 0.69)$ <sup>35</sup> Second derivative preprocessing was effective at resolving peaks within spectra of bone and cartilage. The strongest correlation between modified Mankin grade and second derivative spectral parameters involved the 7201–7000 cm<sup>-1</sup> region, i.e. the O-H stretching second overtone region that is largely associated with water. More specifically, the O-H stretching second overtone region increased in intensity as modified Mankin grade increased ( $R = -0.60$ ; peaks have negative intensity in second derivative spectra).

Correlations were also sought between second derivative spectral parameters and cartilage thickness. The strongest correlations for thickness involved the C-H stretching second overtone region, 8755–8500 cm<sup>-1</sup>. The peak intensity in the 8755–8500 cm<sup>-1</sup> region was more prominent in thicker cartilage samples, which have a proportionally higher organic component. Thickness correlations were found to be useful for assigning organic peaks in NIR spectra. With increased cartilage thickness the presence of a shoulder at 5824 cm<sup>-1</sup> was reduced, associating this particular C-H stretching first overtone with bone. Furthermore, the 8640 and 8266 cm−1 peaks have elsewhere been attributed to C-H stretching second overtones,<sup>51</sup> and the opposing signs of the correlation coefficients indicate that the peaks are due to cartilage and bone respectively.

Prediction models for modified Mankin grade and thickness were calculated using the 4400– 9000 cm<sup>-1</sup> spectral region and PLSR. The increased complexity of determining modified Mankin grade, compared to the relatively simple process of measuring cartilage thickness, likely accounts for the greater number of factors used in the modified Mankin grade prediction model. The 5200 and 7100 cm<sup>-1</sup> bands dominated the loading weights of all three factors of the thickness PLS prediction model; therefore, the spectral differences between bone and cartilage, based on differences in relative amount of bound and free water, formed the basis of the thickness calibration. However, modified Mankin grade predictions were based on more than just the spectral differences between bone and cartilage. Loading weights of the modified Mankin grade prediction model included bands associated with C-H and N-H functionality, and are likely due to changes in proteoglycan and collagen content and environment.

Our PLSR results compared favorably with other spectroscopic studies. For example, an earlier prediction of modified Mankin grade used mid-IR spectroscopy to study human tibial plateau articular cartilage (modified Mankin grades 2–10) and produced a prediction model with a root mean squared error of prediction (RMSEP) of 1.4.<sup>45</sup> Furthermore, a study of articular cartilage from rats coupled NIR spectroscopy and PLSR to predict modified Mankin grade (approximate range 5–13.5) and produced an RMSECV of 0.9 using leave-

one-out cross validation.39 The thickness of bovine patellar cartilage was also studied with NIR spectroscopy: a PLSR prediction using the 8850–7200 cm<sup>-1</sup> spectral region produced an RMSEP of 0.07 mm from a cartilage thickness range of 1.20–2.42 mm.40 The influence of thickness on cartilage spectra has also been investigated in the visible and short-wave NIR spectral region.<sup>55-58</sup> Intensities and RGB coordinates from visible images usefully predicted cartilage thickness below 1.5 mm (RMSEP 0.28 mm), although prediction efficacy was reduced for samples between 1.5 mm and less than 2.5 mm thick.<sup>55</sup>

In conclusion, near-IR spectroscopy may be used to simultaneously assess the thickness and modified Mankin grade of human tibial plateau articular cartilage. The NIR spectroscopic measurements presented here were performed *ex vivo*, however, this method could be adapted for *in vivo* analysis of arthroscopically accessible cartilage surfaces.

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#### **Fig. 1.**

Second derivative (third order polynomial, 170 cm−1 window) near infrared spectra of parafilm and cartilage sections of varying thickness, with (black spectra) and without (grey spectra) an underlying layer of parafilm. Evidence of parafilm was observed when the overlying layer of cartilage was less than 5 mm. The parafilm spectrum has been divided by a factor of 6 so that the  $5774 \text{ cm}^{-1}$  peak is on the same scale.



#### **Fig. 2.**

(a) Baseline offset corrected and (b) second spectra derivative (third order polynomial, 355 cm−1 window) near infrared spectra of bone (grey) and cartilage (black). The NIR spectra were collected from bone and cartilage samples that were 3.56 and 3.52 mm thick respectively. (c) The peak height ratio of 5270 (free plus bound water) to 7085 cm−1 (free water) as determined from the second derivative spectra of samples with varying amounts of bone and cartilage.



#### **Fig. 3.**

(a) Safranin-O sections, (b) baseline offset corrected and (c) second derivative (third order polynomial 355 cm−1 window) near infrared spectra of samples with modified Mankin grades of 2.5 (black solid line), 6.5 (grey dashed line) and 11 (grey solid line; 2.30, 3.12 and 1.89 mm thick respectively). The spectra more closely resemble the spectrum of bone as Mankin grade increases.



#### **Fig. 4.**

(a) First three factor loading weights of the cartilage thickness prediction model based on sample averaged 4400–9000  $\text{cm}^{-1}$  spectra that were pretreated with standard normal variate and second derivative (third order polynomial,  $170 \text{ cm}^{-1}$  window) methods. (b) First five factor loading weights of the modified Mankin grade prediction model based on sample averaged 4400–9000 cm−1 spectra that were pretreated with standard normal variate and second derivative (third order polynomial, 355 cm<sup>-1</sup> window) methods.

#### **Table I**

#### Modified Mankin scoring system



Table II<br>Spearman correlation coefficients for cartilage thickness and Mankin grade with peak area and height parameters. Significant correlation coefficients (P < 0.05) are in bold. Spearman correlation coefficients for cartilage thickness and Mankin grade with peak area and height parameters. Significant correlation coefficients (P < 0.05) are in bold.





*c*Values in bold are statistically significant.

 $\ensuremath{^{\mathrm{c}}\mathbf{V}}$  alues in bold are statistically significant.

*d*These parameters were calculated from second derivative (Savitzky-Golay 3rd order polynomial, 339 cm−1 window).

 $d$ These parameters were calculated from second derivative (Savitzky-Golay 3rd order polynomial, 339 cm<sup>-1</sup> window).