UNIVERSITY OF TWENTE.

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Determining the Viscoelastic Response of Trabecular Bone

in collaboration with



Internship Report

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Abstract

The time independent mechanical properties of trabecular bone have been extensively studied including the relation between bone mineral density and Young's modulus. The time dependent properties (or) viscoelasticity of bone have not gained much attention. The purpose of this report is to determine the stress response and relate it to the bone mineral density and strain. The bone mineral density was calculated for human samples while the stress response and further correlations were made in bovine samples due to time constraints. Human distal femoral and proximal tibial samples were collected from 6 donors and the bone mineral density were determined. Simultaneously, stress relaxation experiments with a holding time of 30 minutes were performed on 16 cylindrical bovine femoral samples. The minimum duration data of the reference samples were fitted using the Schapery method and the Modified Superposition Method and validated with other samples. Among these two methods, Modified Superposition Method showed a better fit (error $\approx 46.43\%$) than Schapery Method (error $\approx 66.82\%$). The parameters of these methods were used to determine if there is a correlation between bone mineral density and strain together with stress response. While comparing the parameters of the Schapery and Modified Superposition Method equations with the bone mineral density, there was no correlation between the parameters of Modified Superposition Method and the non-linear parameters of the Schapery method meaning that the equation of Modified Superposition Method cannot be represented in terms of bone mineral density. On the other hand, the linear parameters of Schapery method related directly to the bone mineral density of the samples. Even if Modified Superposition Method essentially had a fairly better fit for the stress relaxation data, it has no parameters in its equation that relates to bone mineral density. Therefore, it was concluded that bone mineral density can be related to strain and stress with linear parameters of the Schapery model. The Schapery formula with bone mineral density, strain and stress had an error of 88.2%. Due to a high error rate, the correlation between bone mineral density and strain coupled with stress response using Schapery model was not ideal.

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1 Introduction

The trabecular bone is an open porous composite cellular solid material from an engineering perspective. The vertebral bodies of the axial skeleton and the ends of long bones of appendicular skeleton include trabecular bone. It is also present in the hollow long bones near the tip of the medullary chambers. The skeleton as a whole benefits from the intricate, porous spatial organization of the trabecular bone which helps to maximize strength 1. A large substrate is available for cellular interaction with bone mineral material provided by the high mineral surface area associated with the arrangement of the trabecular bone components. It creates a trabecular network of connecting rod- and plate-shaped structures that varies in density depending on the skeletal location. The proportions of the trabecular elements and their orientation in space, as well as trabecular architecture are all very diverse [2]. The mechanical properties of this cellular material depend on its heterogeneous microstructure, which varies with age, disease and anatomical site being considered [3]. The time-independent mechanical properties of the trabecular bone such as Young's modulus is generally related to its density and computer models of bone typically exploit these relationships. The density of the trabecular bone is usually determined with the help of Computed Tomography (CT) images [4, 5]. Studies have found that the trabecular bone has time-dependent mechanical properties [6].

Viscoelasticity is the time-dependent elastic behavior of materials. Viscoelastic behavior combines elastic and viscous behavior, with an immediate elastic strain followed by a viscous, time-dependent strain as a function of applied stress. Therefore, the response to an external stimulus is delayed, which results in a loss of energy inside the material [7]. There are certain experiments performed to determine the viscoelastic behavior of trabecular bones. Tensile tests [8, 9] test the force required to break the specimen and the extent to which the specimen elongates to that breaking point. Stress relaxation tests [5, 9, 10, 11] measures time-dependent varying force due to applied constant strain over time. Creep recovery tests [4, 12] measures the time-varying strain due to applied constant load over time. Dynamic Mechanical Analysis [13] measures the lag between sinusoidal stress and strain over a frequency range. Fatigue tests [10] determines the variation in Young's modulus at a constant strain rate and different cyclic loads. Various studies which determined viscoelastic behavior of the bones claim that the behavior linear [10, 12]. But in reality, the viscoelasticity is not linear and is dependent on the applied stress or strain level [11]. So, a material model cannot be developed with the assumption of linear viscoelasticity. At this point, the study on the non-linear behavior of trabecular bone comes into account.

A study by Manda.et.al (2016) attempted to explain the non-linear viscoelastic properties of the bovine bone sample using Multiple Load Creep Recovery tests [4] but crucially

fails to explain the effect of bone mineral density (BMD). Therefore, it is known that the effect of strain on stress response is non-linear. But the effect of BMD on stress response is still unclear. To determine this relationship, an existing model which relates strain to stress should be used to fit the stress relaxation data.

There are several models that were used to fit the stress relaxation data. The comparison of Quasi-Linear Viscoelasticity (QLV), Nonlinear Superposition and Schapery model on tendons and ligaments concluded that Schapery's nonlinear viscoelastic model correctly predicts recovery and reloading behavior while fitting a single relaxation curve and predicts the strain dependent relaxation behavior as well [14]. Many studies have used Schapery model to fit the stress relaxation data of polymers [15, 16, 17] and a few studies have used Modified Superposition Method to fit their stress relaxation data of ligaments and human dentin [18, 19, 20]. Provenzano et. al., decided to determine whether the Schapery theory or the modified superposition method could adequately model the strain-dependent stress-relaxation behavior of ligaments and found that both works the same [21]. So, it was decided to compare the Schapery model and Modified Superposition Method to fit the stress relaxation data. Both the Schapery method and Modified Superposition Method directly relates stress to applied stain level using Young's modulus of Tangent modulus. So, the variables of the models related to BMD were replaced with the relationship equations.

The aim of this study is to perform stress relaxation experiments on each sample for 24 hours, fit the results of a minimum duration to two existing models and determine if these models could be used to simulate nonlinear (or strain dependent) stress relaxation of bovine trabecular bone and if these can relate the bone mineral density to this stress either linearly or non-linearly.

The remainder of the report is organised as follows: First, the scenario that was presented before starting the internship was described in Section 2, followed by the problem statement in Section 3. Section 4 is about the planning for the entire assignment followed by methodology and results in Sections 5 and 6. The discussion and conclusion of the assignment is described in Sections 7 and 8 respectively.

2 Background

The Orthopedics Research Laboratory (ORL) at the Radboud University Medical Centre constructed an Finite Element (FE) model to reduce the chance of failure of a total knee implant. This model is able to predict the primary fixation of a prosthesis by calculating the relative displacement at the bone-implant interface, also known as micromotions. However, it was shown that this FE model underestimated the micromotions in comparison to experimental measurements [22]. It is believed to be due to the absence of

viscoelastic properties in the FE model. The solution is that a model must incorporate the viscoelastic behavior. So, the viscoelastic behavior of the trabecular bone from the femur and tibia need to be described numerically to make a model. Previous studies relate the viscoelastic behavior to strain level but not to BMD. The goal of this internship is to perform the material assignments on human bone sample to determine bone mineral density and characterize the viscoelastic behavior of bone specimens, through stress relaxation tests performed in the trabecular bone samples and to correlate the stress response to the bone mineral density and the initial strain levels. Due to time constraints, in this internship, the material assignments were carried out on human bone sample and stress relaxation tests were carried out on bovine femur samples.

The following section describes the problem statement presented throughout this internship report and the goals that were defined to address problem statement.

3 Problem Statement and goals

The primary objective of the study was to identify the correlation between the stress relaxation response and the initial strain level coupled with bone mineral density in bovine femoral trabecular bone.

To achieve this the following question need to be answered:

- 1. What is the BMD of each sample?
- 2. Which of the MSM or Schapery predict the viscoelastic behavior better?
- 3. How does BMD influence the viscoelastic behavior?
- 4. How does either MSM or Schapery incorporated with BMD predict the viscoelastic behavior?

The subsequent sections describe how the activities during the course of the internship were planned and the procedure in which the experiments were carried out in order to achieve the results towards the aforementioned problem statement.

4 Planning

The plan for the total duration of 14 weeks during the internship was scheduled as shown in Figure 1.

As stated before, due to time constraints, the entire assignment was carried out under two phases:

• Phase I (Human Samples): Sample collection, Material assignments

• Phase II (Bovine Samples): Stress Relaxation Experiments, Data Fitting



Figure 1: Internship Planning

5 Methodology

5.1 Sample Preparation

Human Femur and tibia were collected from 6 donors - 1 male and 5 female. The donors were classified into two groups based on age - 3 with mean age of 54 (53,57,53) and 3 with mean age of 80 (90,73,76). Cylindrical samples were harvested from the distal end of the femur and the proximal tibia as they form the knee joint. A 10 mm inner diameter cylindrical saw was used to acquire the trabecular bone core each of 30mm in length. A minimum of six samples were drilled from each tibia parallel to the shaft. In the femur, two samples at 80° from posterior condyles and six samples parallel to the femoral shaft. The femur and tibia after drilling as shown in Figure 2 were wrapped with saline cloth and stored at $-20^{\circ}C$.

After a few days, all the 12 drilled samples were scanned under a high resolution CT (Siemens Biograph mCT) at the Radboud UMC. These CT images (Figure 3) were used to determine bone mineral density as described in Section 5.2. The excess bones other than the sample were sawn off. The femur and tibia without the samples is shown is Figure 4. The samples with a size of around 21mm were selected for further procedures to get an aspect ratio of 1:2.

The bones were sliced from the selected samples to remove all the cortical bone on top or marrow on the bottom, wrapped with saline cloth and stored at $-20^{\circ}C$. The sample height, diameter and the height of the sliced portion from the top were measured using a digital caliper and noted.



(a) Femur in the drill



(b) Femur after drilling



(c) Tibia after drilling

Figure 2: Sample drilling

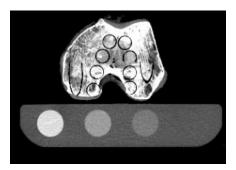


Figure 3: High Resolution CT image



(a) Femur after sawing



(b) Tibia after sawing

Figure 4: Sample sawing

The successive section explains the methods with which the further procedures were performed on the human sample and the bovine sample.

5.2 Material Assignments

The first step of the material assignment was to illustrate all the data (such as the sample height after sawing, the height of the sample cut from the top, diameter of each sample) obtained during the sample collection. A cylindrical Standard Triangle Language (.stl) file was created for each sample with the diameter same as the sample using Tinkercad. The STL file and the CT scans were loaded in a software called 3D Slicer (Slicer 4.11.20210226) and the cylinders were transformed to the locations that matches the exact projection of each sample. The height of the sample cut from the top was also adjusted while positioning each cylinder. Then the diameter of the cylinders were slightly adjusted so that it fits perfectly in the CT scan. These transformed cylinders were hardened to make sure it's position remains unchanged. The hardened cylinders as stl files were loaded in Hypermesh (HyperMesh Version: 2021.2.1 - HWDesktop) in which the 2D cylinders were filled with mesh and made 3D. The mesh created using Hypermesh is shown in Figure 5. The outputs of Hypermesh were saved as job files of the latest version. But our in-house software accepts a job file of version 2017.1. So, Marc Mentat (Marc Mentat 2021.4) is used to convert the job file of current version to job file of 2017.1 version and makes it suitable for use in Matlab. Finally, a custom written Matlab script (Matlab R2021b, Mathworks, MA, USA) with the job (.dat) file and the DICOM images (CT scans) of the sample as input was used to determine the BMD of the human trabecular bone of the femur and the tibia [23].

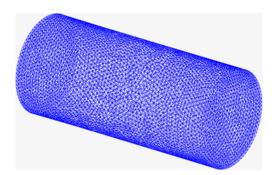


Figure 5: A sample of Hypermesh output

5.3 Stress Relaxation Tests

For this assignment, stress relaxation tests were done in the bovine femur samples. There were 16 bovine samples in total and on each of the samples 24 hours stress relaxation

experiments were performed. As the first step, the sample to be tested was cemented to the end caps using Autoplast Cold Curing Denture Base Material. The end caps were used to ensure that the bone does not get damaged during the application of stress. The sample after cementing is shown in Figure 6.



Figure 6: Bovine sample after cementing

Prior to pre-conditioning, the sample embedded in the endcaps was placed in a water basin filled with physiological saline, with a temperature of 37.0 ± 0.5 °C for 30 minutes. Then, each sample was preconditioned by applying 0.1% apparent strain for ten cycles [24] and was allowed to recover for 30 minutes again. The experiments were conducted using an MTS machine (MTS Systems Corporation, Eden Priairie, Minnesota, USA) with a measuring frequency of 10 Hz. The entire experimental setup is shown in Figure 7. It is known that trabecular exhibits nonlinearity and residual strain buildup below the yield strain of 0.8% [3, 25]. In order to quantify the viscoelastic response, static stresses of 0.2, 0.4, 0.6 and 0.8% were used. With a strain rate of 0.01 s⁻¹, the stresses were initially applied and then re-moved. Each strain level investigated four cylindrical trabecular bone samples, yielding a total of 16 specimens for the three studies. A holding period of 24 hours was applied to the static strain in order to measure the length of the stress relaxation response.

Digital imaging contrast (DIC) was used to measure the axial displacement applied on the bone samples. The dots in the end caps, shown in Figure 6, were marked to visualize if any displacement has occurred during the experiment. Images of the uniaxial compression test were continuously captured and deformations of the samples were calculated based on a custom-written Matlab script.

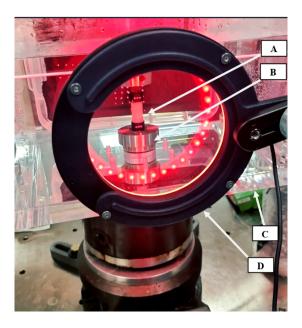


Figure 7: Sample inside MTS during experiment (A)Bovine femur sample; (B)MTS; (C)water bath with saline solution; (D)DIC

5.4 Data fitting

5.4.1 Force Versus Time

The fitting of stress relaxation data with the Schapery and MSM models was done using Matlab. The output text files of the 24 hours experiments from the MTS were taken as the input along with the images obtained using DIC. The DIC images were used to determine the applied strain (ϵ) on each sample. The data such as force, time and axial displacement were read from the MTS files. The data for a minimum duration of 30 minutes was chosen for further findings and illustrations. The force and time shown in Figure 8 were used to determine the time and the position of the force between the region of interest (i.e., between minimum force and maximum force).

Using the diameter of the sample which is determined during sample preparation and the strain applied on each sample which is determined using the DIC images, the stress (σ) and the stress relaxation modulus (E) for each sample were calculated using the equations 1 and 2 where Area is the area of the circle with the diameter of the sample.

$$\sigma = \frac{Force}{Area} \tag{1}$$

$$E = \frac{\sigma}{\epsilon} \tag{2}$$

The force position and time between the region of interest for the first 30 minutes of the experiment was determined. The respective $\sigma(t)$ and E(t) for each sample was

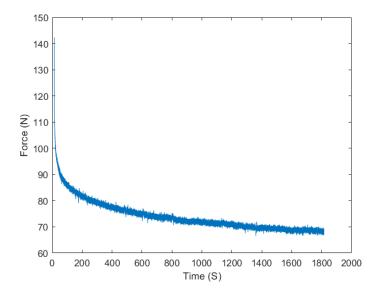


Figure 8: Force Versus Time graph for one bovine sample

calculated. The E(t) Vs time of this experimental data was plotted and fitted with power law to verify as shown in Figure 9. The power law is that a relative change in one quantity results in a proportional relative change in another and the equation of power law looks like $y = ax^b$.

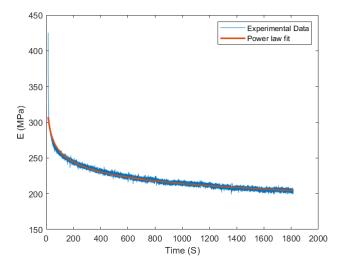


Figure 9: E Versus Time graph with fit for one bovine sample

5.4.2 Models for data fitting

As stated in the section 1, the stress relaxation data was fitted to the Schapery model and MSM model to determine which method has a better fit. Another reason for carrying out this comparison was to identify the correlation of BMD and stress response which

will be discussed later in section 7. A detailed explanation on these methods is given in Sections 5.4.3 and 5.4.4.

5.4.3 Schapery Method

Schapery's nonlinear viscoelastic theory can be derived using principles of irreversible thermodynamics [16, 26, 27]. When strain is treated as the independent state variable and the case of uniaxial loading is considered, Schapery's theory reduces to a single integral expression as given in Equation 3.

$$\sigma(\epsilon, t) = h_e(\epsilon) E_e \epsilon + h_1(\epsilon) C \int_0^t \Delta(E) (\rho(t) - \rho'(\tau)) \frac{dh_2(\epsilon) \epsilon}{d\tau} d\tau$$
 (3)

In this formula E_e is the equilibrium modulus which was taken as final value of the young's modulus (E) and $\Delta(E)$ is called the transient modulus. The reduced time ρ and ρ' is defined as

$$\rho = \int_0^t \frac{dt'}{a_e(\epsilon(t'))} d\tau \tag{4}$$

$$\rho = \int_0^\tau \frac{dt'}{a_e(\epsilon(t'))} d\tau \tag{5}$$

The terms h_e , h_1 , h_2 , and a_e are strain-dependent material properties. Theoretically, all these values were assumed to be 1 for a linear behaviour which would reduce the equation to a Boltzmann equation [21]. In reality, bovine trabecular bone is non-linear [4]. If the experiments were performed under isothermal conditions, the values of h1 and a_e are assumed to be 1. It is applicable for our data as we have performed our experiments under constant room temperature and the transient modulus is equated to power law as in Equation 6.

$$\Delta E(\rho) = C\rho^n \tag{6}$$

Substituting all these terms to Equation 3, the Schapery equation becomes,

$$\sigma(\epsilon, t) = h_e(\epsilon) E_e \epsilon + h_1(\epsilon) C \int_0^t (\rho - \rho')^n \frac{dh_2(\epsilon) \epsilon}{d\tau} d\tau$$
 (7)

Applying Heaviside step function, the final equation of Schapery model is

$$\sigma(\epsilon, t) = h_e(\epsilon) E_e \epsilon_0 + h_2(\epsilon) C \epsilon_0 t^n \tag{8}$$

In the above equation, h_e , h_2 , E_e can be found using curve fitting and ϵ_0 is the initial strain level from the experimental data. The curve fitting will be performed by fitting the above equation to the graph of E Vs time.

The general equations of h_2 and h_e can be written as a first order polynomials as Equation 9 and Equation 10 respectively.

$$h_2(\epsilon) = L * \epsilon + M \tag{9}$$

$$h_e(\epsilon) = O * \epsilon + P \tag{10}$$

The final equation of the Schapery fit with the functions can be written as Equation 11.

$$\sigma(\epsilon, t) = (O * \epsilon + P) * E_e * \epsilon + (L * \epsilon + M) * C * \epsilon * t^n$$
(11)

5.4.4 Modified Superposition Method (MSM)

The single integral formula of MSM is given by Equation 12

$$\sigma(\epsilon, t) = \int_0^t E(t - \tau, \epsilon(\tau)) \frac{d\epsilon(\tau)}{d\tau} d\tau \tag{12}$$

The form of the relaxation function will be chosen as a non-separable strain-dependent power law:

$$E(\epsilon, t) = A(\epsilon)t^{B(\epsilon)} \tag{13}$$

Applying Heaviside step function, the final equation of Schapery model is

$$\sigma(\epsilon, t) = A(\epsilon)\epsilon_0 t^{B(\epsilon)} \tag{14}$$

The general equations of $B(\epsilon)$ and $A(\epsilon)$ can be expressed as first order polynomials as Equations 15 and 16.

$$B(\epsilon) = D * \epsilon + F; \tag{15}$$

$$A(\epsilon) = H * \epsilon + J; \tag{16}$$

By substituting these two equations in Equation 14, the final equation of MSM becomes,

$$\sigma(\epsilon, t) = (H\epsilon + J) * \epsilon * t^{(D\epsilon + F)}$$
(17)

6 Parameter Estimation and Results

6.1 Schapery Method

In order to get a correlation between BMD and stress response, and a random distribution of BMDs over the strain levels, the samples were classified based on their BMD into 4 groups with 4 samples in each group in the order of increasing BMDs (i.e., lowest BMDs in group 1 and highest BMDs in group 4). All the 4 samples in each BMD group had different strain levels and different BMDs. Each group was assumed to have a common reference BMD (average of all 4 BMDs), say BMD_r . The Stress relaxation modulus (E) Versus Time graphs for each BMD group were plotted with the data obtained using the MTS output and DIC images. The Curve Fitter toolbox in Matlab was used to identify the parameters of the Schapery formula (Equation 8) as described below.

The lowest strains of each BMD group (0.2%, 0.19%, 0.18% and 0.16%) were taken as the initial strain levels (ϵ_0) for fitting. The bovine trabecular bone was assumed to behave linearly at these initial strain levels. In other words, the E vs time curve of these values were assumed to respond in a linear way. The curve of E vs time for the strain of 0.2% (initial strain of BMD group 1) was plotted in the curve fitter with the Equation 8 as fit. The values of h_e and h_2 were assumed to be 1 based on previous work by Provenzano et al., as given in the Section 5.4.3 and the values of E_e , C and n were determined from the Curve fitter. Fitting Equation 8 using the above values of E_e , C and n to the stress relaxation data of the other strain levels in BMD group 1, h_e and h_2 for other samples were calculated.

The h-strain curves of each BMD groups were fitted to the Equation 9 and Equation 10 with weights as shown in the Figure 10 to determine the variables of those equations. The weights were applied to the first values (values of initial strain level) to make it approximately 1.

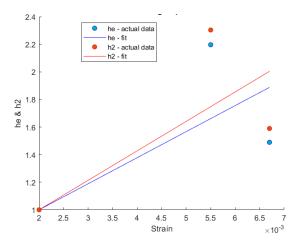


Figure 10: h_e and h_2 of BMD group 1 as a function of strain

By substituting the known values in Equation 11, the final equation of BMD group 1 can be obtained.

The same procedure was followed for the other BMD groups and the respective final equations were formed. The values of Ee, C and n of each BMD group are listed in table 1 and the figures 11, 12 and 13 show the plots of h_e & h_2 for BMD groups 2, 3 and 4 respectively. The values of the variables in equations 9 & 10 can be found in the table 2.

BMD Group	Ee	С	n
	(MPa)		
1	1215	-913.2	0.01439
2	958.4	-547.9	0.0304
3	1729	-384.5	0.08508
4	636.1	-149.3	0.1123

Table 1: Value of Ee, C & n of each BMD group

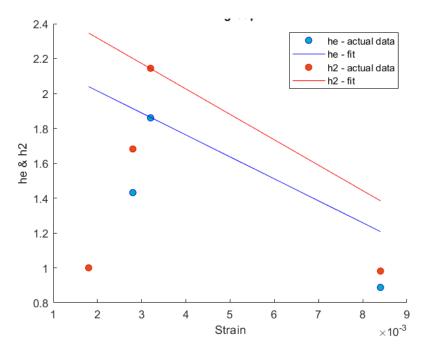


Figure 11: h_e and h_2 of BMD group 2 as a function of strain

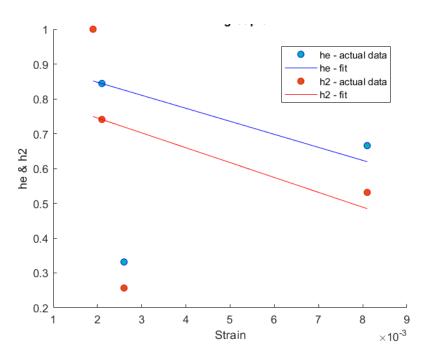


Figure 12: h_e and h_2 of BMD group 3 as a function of strain

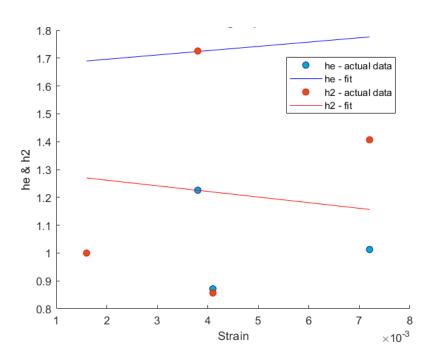


Figure 13: h_e and h_2 of BMD group 4 as a function of strain

BMD Group	L	M	О	Р
1	213.6	0.5730	188.9	0.6223
2	-145.8	2.609	-126	2.2661
3	-42.7	0.8303	-37.4	0.9224
4	-20.19	1.302	15.43	1.6650

Table 2: Value of variables of $h_2(\epsilon)$ and $h_e(\epsilon)$ of each BMD group

The Equation 11 with the values of Table 1 and Table 2 is plotted with experimental data in Figures 14,15, 16 and 17.

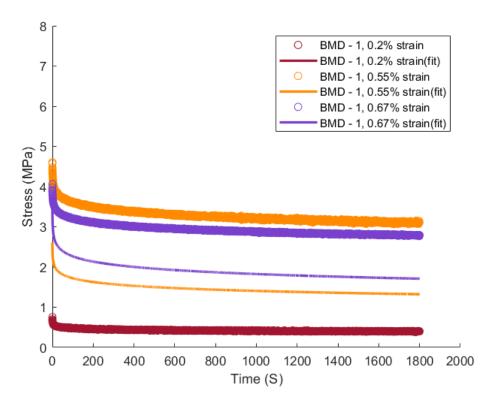


Figure 14: Experimental data with 0.2% Schapery model - BMD Group 1

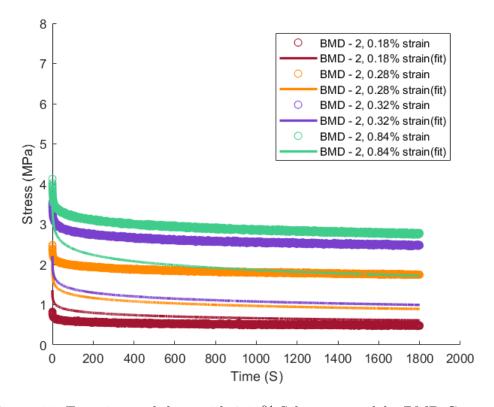


Figure 15: Experimental data with 0.18% Schapery model - BMD Group 2

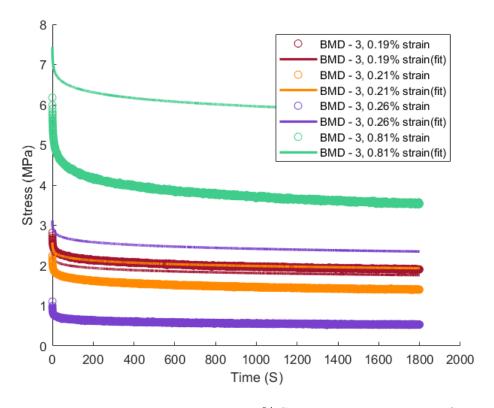


Figure 16: Experimental data with 0.19% Schapery model - BMD Group 3

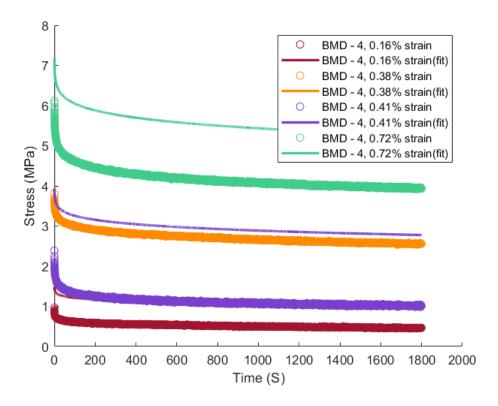


Figure 17: Experimental data with 0.16% Schapery model - BMD Group 4

The above procedure provided the equation of stress with respect to strain. Moreover, the aim of this study was to quantify the correlation between BMD and stress.

The variables of Equation 8 other than the strain and time were related to reference BMDs (BMD_r) to determine if they had any correlation. The graphs of C, n, h_e and h_2 versus BMD was shown in figures 18, 19 and 20.

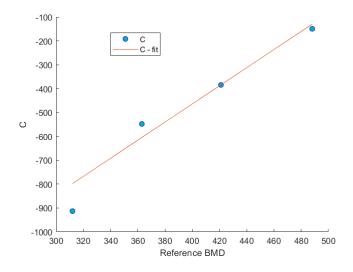


Figure 18: Variable C as a function of BMD

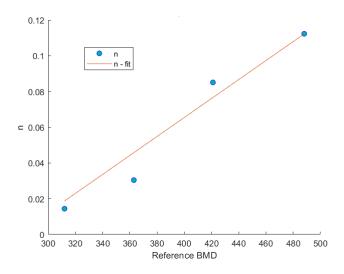


Figure 19: Variable n as a function of BMD

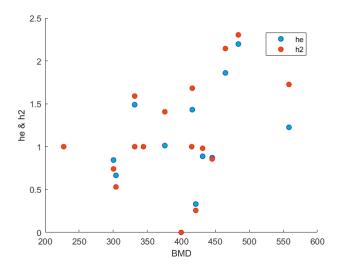


Figure 20: Overall h_e and h_2 as a function of BMD

The variables C and n show a correlation with BMD and their functions were given by equations 18 and 19.

$$f_c = 3.795 * BMD_r - 1982 (18)$$

$$f_n = 0.0005311 * BMD_r - 0.147 (19)$$

6.2 Modified Superposition Method (MSM)

For the same BMD groups, MSM model was implemented to determine the correlation. The initial strain levels and the reference BMDs were similar to the Shapery method. The Stress Versus Time graphs for each BMD group were plotted with the data obtained using the MTS output and DIC images. The logarithm of stress and time were calculated for further procedures in this method. The Curve Fitter toolbox in Matlab was used to identify the parameters of the Modified Superposition Method (Equation 14) as described below.

For each BMD group, the slope of the log-log stress-time graph of every sample was found and the slope-strain graph was fitted with Equation 15 (B(ϵ)). Then the initial relaxation modulus - strain graph was fitted Equation 16 (A(ϵ)).

The values of the variables H, J, D and F of Equation 17 are given in the table 3

BMD Group	D	F	Н	J
1	3.558	-0.070	67699	286.07
2	0.887	-0.052	-36024	882.73
3	-1.968	-0.055	-53228	1131.49
4	6.283	-0.098	37066	605.66

Table 3: Value of variables of D, F, H and J of each BMD group

The Equation 17 with the values of Table 3 is plotted with the experimental data as shown in figures 21, 22, 23 and 24.

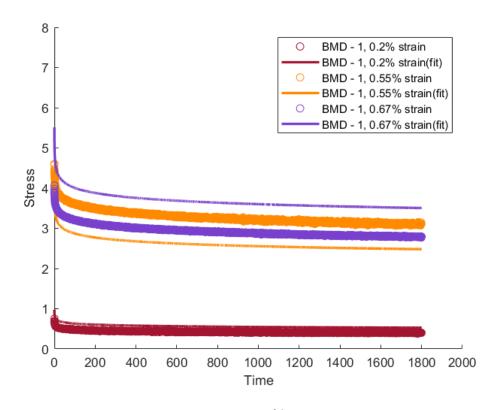


Figure 21: Experimental data with 0.2% MSM model - BMD Group 1

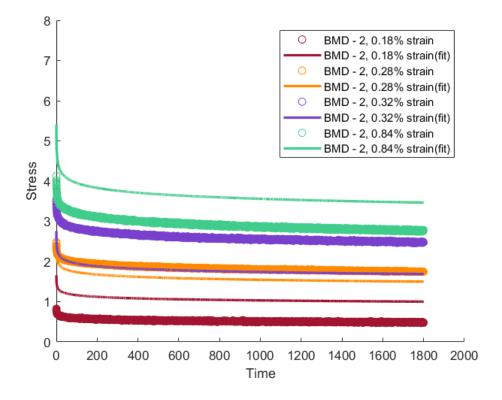


Figure 22: Experimental data with 0.18% MSM model - BMD Group 2

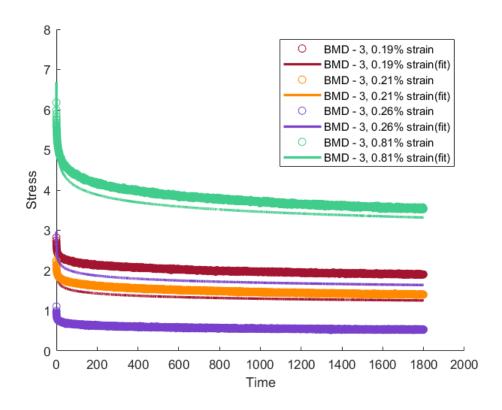


Figure 23: Experimental data with 0.19% MSM model - BMD Group 3

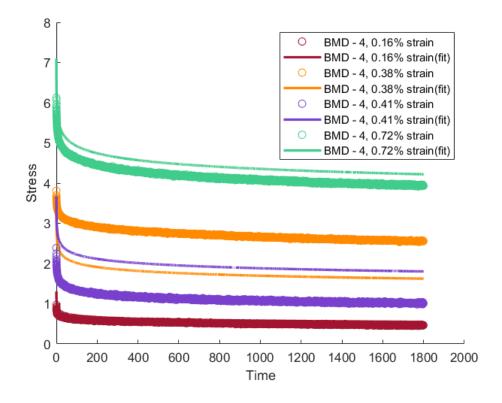


Figure 24: Experimental data with 0.16% MSM model - BMD Group 4

The above procedure provided the equation of stress with respect to strain. Moreover, the aim of this study was to quantify the correlation between BMD and stress.

The variables of Equation 17 other than the strain and time were related to reference BMDs to determine if they had any correlation. The functions of D, F, H and J versus BMD were found to be equations 20, 21, 22 and 23 and the graphs were plotted as shown in figures 25, 26, 27 and 28 to determine the correlation and the influence (linear/non-linear) of BMD on these variables.

$$f_D = 0.01165 * BMD_r - 2.422 \tag{20}$$

$$f_F = -0.0001626 * BMD_r - 0.004826 \tag{21}$$

$$f_H = -140 * BMD_r + 59330 (22)$$

$$f_J = 1.788 * BMD_r + 18.48 \tag{23}$$

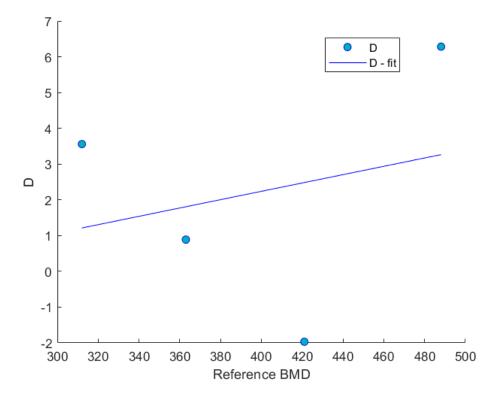


Figure 25: Variable D as a function of BMD

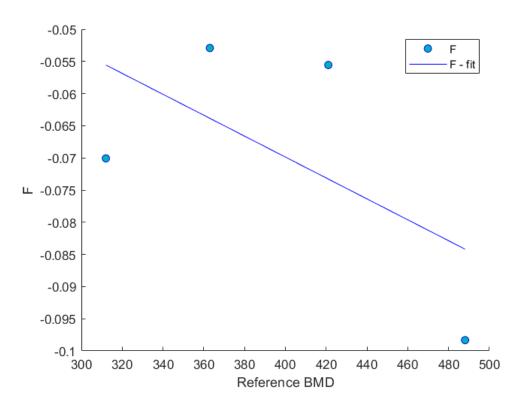


Figure 26: Variable F as a function of BMD

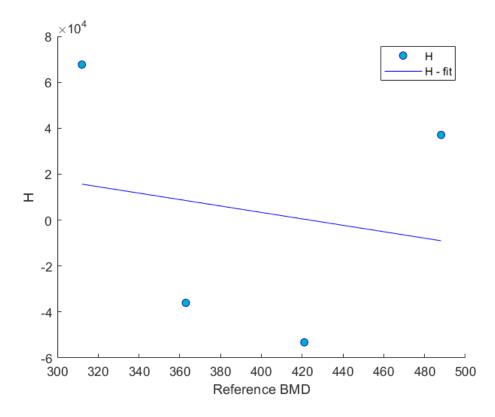


Figure 27: Variable H as a function of BMD

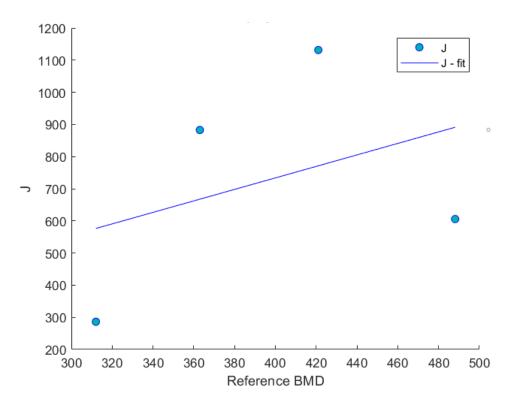


Figure 28: Variable J as a function of BMD

6.3 Comparison between Schapery & MSM models

While comparing the Schapery model and MSM model, MSM was easy to compute with less number of variables. Even if Schapery model related BMD, strain and stress, the mean of root mean square error (Coefficients of determinants or R^2) of Schapery method was 0.9451 while in MSM, it was 0.5807. In simple terms, there was about 66.82% variation in the fit of the final data point and the experimental data in Schapery method and it was 46.43% in MSM. This shows that MSM has a better fit with this stress relaxation data. The higher root mean square error and higher error percentage were assumed to be due to the lack of sufficient bovine samples. The linear parameters of Schapery method had a correlation with BMD while the non-linear parameters of the Schapery method and the parameters of MSM had no correlation with BMD. This might also be due to the lack of sufficient samples. The Schapery formula with bone mineral density, strain and stress had an error of 88.2%.

7 Discussion

Several studies have been performed to understand the time dependent mechanical properties of the trabecular bone [11, 24]. The studies that study the time independent properties of the trabecular bone relate the properties like Young's modulus to the density of the samples in order to develop a material model [1], but the relationship between time dependent viscoelastic properties with the density and strain has not been investigated to the finest of our information. A study related the viscoelastic properties of trabecular bone to the strain but failed to relate it to the density [4].

In this study, we conducted 24 hour stress relaxation experiments on 16 bovine trabecular femur samples and the behavior was quantified using non-linear viscoelastic theory based on the Schapery Method and Modified Superposition Method. Both the Schapery method and Modified Superposition Method directly relates stress to applied stain level using Young's modulus (time dependent properties) of a bone. But the goal of this assignment is to relate the time dependent and independent properties to strain level. The attempt to relate the variables of these methods to BMD in order to identify correlation resulted in the following discussions:

While comparing the Schapery model and MSM model, MSM was easy to compute with less number of variables. The mean of R^2 of Schapery method was 0.9451 while in MSM, it was 0.5807. The lower the root mean square error, the better the model is able to fit the data. This shows that MSM has a better fit with this stress relaxation data than the Schapery method.

In Schapery method, the graphs of the variables h_e and h_2 seems to have no correlation with BMD. This statement is supported by the overall graph of h_e vs BMD and h_2 vs BMD (Figure 20). The same holds true for the variables of Modified Superposition Method (Figure 25, 26, 27, 28). But there is a linear relationship between the variables C and n of the Schapery model with the BMD (Figure 18,19). It means with an increase in density, there is an increase in the value of C and n. This clearly defines that there is a positive correlation between the stress relaxation response and strain level coupled with BMD in relation to the linear variables. For non-linear variables, no correlation was observed.

After fitting the relationship equations of C and n to the final Schapery formula, the Schapery fit had an error of 88.2%. Due to a high error rate, the correlation between bone mineral density and strain coupled with stress response using Schapery model was not ideal.

There were a few challenges encountered during the course of the assignment. The lack of sufficient publications in determining the viscoelastic properties of trabecular bone made it difficult to conclude preconditioning and further procedures. The availability of less number of bovine samples intricated the process of data fitting. This reflected in

higher room mean square error in the results and it made the process of concluding the correlation between the non-linear parameters of the Schapery model and BMD difficult. But the minimum number of samples required to perform a correlation is still unkown. The concept of relating the BMD to the stress was quite complex as the BMD varied for every 16 sample. Initially, the applied strain levels were thought to be 0.2%, 0.4%, 0.6% and 0.8%. But, the DIC images confirmed that the strains were different for each sample. So, the concept of fitting the data of same strain level and different BMDs was not possible. This led to the assumption of the samples belonging to the same BMD group have a common (average) BMD. After grouping the BMDs, one sample in BMD group 1 varied much than the other samples in the same group. So the particular sample was omitted for fitting the curves of Schapery model and Modified Superposition Method.

Further studies related to this research might include the determination of the influence of non-linear parameters of the Schapery model on BMD again by including more number of samples. This might have a greater influence on the result that the correlation is possible with linear parameters in the model.

The problem statement and the goals of the assignment as in Chapter 3 were answered during the course of the internship. In bovine femoral trabecular bone, the stress relaxation response and the initial strain level coupled with bone mineral density had a correlation between each other in relation to the linear variables. The BMD of each sample was found using softwares Slicer, Hypermesh, Mentat and Matlab. A custom-written script was used in Matlab to find the density of each sample. After performing the stress relaxation experiments and data fitting, MSM model predicted the viscoelastic behavior better than Schapery method. The Schapery method incorporated with the density did not predict the viscoelastic behavior better while MSM method had no correlation with density making the method not possible to incorporate with density. The entire assignment was performed completely within the duration of the internship.

8 Conclusion

This research was designed to study the correlation between stress response and strain together with density in bovine trabecular bone using stress relaxation experiments. First, the density was found and then 24 hour stress relaxation experiments were performed. Due to time constraints, in this assignment, the density was found for human femoral and tibial samples which were harvested during the assignment and the experiment was performed on already harvested 16 bovine femur samples in which the material assignments were also performed previously.

The stress relaxation data of the reference samples of each BMD group were used to fit the two existing models: Schapery model and Modified Superposition Model. The

MSM produced better fit than Schapery model. MSM had less number of parameters to compute too. But, both of these methods related stress to strain and not to BMD. An attempt was taken to relate the parameters of these models to BMD. At this moment, MSM method did not show any correlation between BMD, strain and stress response due to the absence of linear parameters. The linear parameters of the Schapery model indicate the change of elastic modulus changes over time and had a better correlation with BMD. This brings the conclusion that BMD and strain level can be related to the stress response with a linear parameter in the model. The Schapery formula with bone mineral density, strain and stress had an error of 88.20%. Due to a high error rate, the correlation between bone mineral density and strain coupled with stress response using Schapery model was not ideal. With the data provided, this report helps one to understand the correlation between the viscoelastic properties and strain level coupled with density and develop a model with both time-dependent and independent properties of trabecular bone.

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Appendix A

Obtaining the ROI & parameters of ROI from Stress Relaxation Data

```
1 %% Run ReadDataMTS first
2 clc
4 [dataStruct] = ReadDataMTS_new();
6 %% obtain ROI for time Experiment
8 MaxForce = [];
9 MaxForceI = [];
10 EndTimeExperiment = [];
11 MinimumDifference = [];
12 MinForceI = [];
13 MinForce = [];
14 PercentageStressRelaxation = [];
15 ForcePosROI = [];
16 dataStruct_ROI = struct('TimeROI', [], 'ForcePosROI', [], 'StressSample'
     ,[],'StressRelMod',[],'StressRelMod_fit1',[]);
17 TimeROI = [];
for k = 1:size(dataStruct,2)
      if k==4 || k==9 || k==12 || k==13 || k==16
      [MaxForce(k), MaxForceI(k)] = min(dataStruct(k).kNForce);
      DurationExperiment = 86400; % aanpassen
      EndTimeExperiment(k) = dataStruct(k).Time(MaxForceI(k))+
     DurationExperiment;
      [MinimumDifference(k), MinForceI(k)] = min(abs(dataStruct(k).Time-
24
     EndTimeExperiment(k));
      MinForce(k) = dataStruct(k).Force(MinForceI(k));
      PercentageStressRelaxation(k) = 100 - (MinForce(k) / MaxForce(k) *
     100); %handige value
27
      dataStruct_ROI(k).TimeROI = dataStruct(k).Time(MaxForceI(k):
      dataStruct_ROI(k).ForcePosROI = dataStruct(k).kNForce(MaxForceI(k):
29
     MinForceI(k))*-1;
      NumberOfEndRemovers = 0;
while dataStruct_ROI(k).ForcePosROI(end) - dataStruct_ROI(k).
     ForcePosROI(end-1) > 5 || dataStruct_ROI(k).ForcePosROI(end) -
     dataStruct_ROI(k).ForcePosROI(end-1) < -5</pre>
      dataStruct_ROI(k).ForcePosROI(end) = [];
```

```
dataStruct_ROI(k).TimeROI(end) = [];
      NumberOfEndRemovers = NumberOfEndRemovers + 1;
  end
      else
38
      [MaxForce(k), MaxForceI(k)] = min(dataStruct(k).Force);
39
      DurationExperiment = 86400; % aanpassen
40
      EndTimeExperiment(k) = dataStruct(k).Time(MaxForceI(k))+
41
     DurationExperiment;
      [MinimumDifference(k), MinForceI(k)] = min(abs(dataStruct(k).Time-
42
     EndTimeExperiment(k));
      MinForce(k) = dataStruct(k).Force(MinForceI(k));
43
      PercentageStressRelaxation(k) = 100 - (MinForce(k) / MaxForce(k) *
44
     100); %handige value
45
      dataStruct_ROI(k).TimeROI = dataStruct(k).Time(MaxForceI(k):
     MinForceI(k));
      dataStruct_ROI(k).ForcePosROI = dataStruct(k).Force(MaxForceI(k):
     MinForceI(k))*-1;
      NumberOfEndRemovers = 0;
while dataStruct_ROI(k).ForcePosROI(end) - dataStruct_ROI(k).
     ForcePosROI(end-1) > 5 || dataStruct_ROI(k).ForcePosROI(end) -
     dataStruct_ROI(k).ForcePosROI(end-1) < -5</pre>
      dataStruct_ROI(k).ForcePosROI(end) = [];
      dataStruct_ROI(k).TimeROI(end) = [];
      NumberOfEndRemovers = NumberOfEndRemovers + 1;
55 end
     end
56
57 end
59 %% Assigning the applied strains
61 PercentageAppliedStrain = [];
62 AppliedStrain = [];
63 StressSample = [];
64 StressRelMod = [];
66 PercentageAppliedStrain = [0.26 0.18 0.20 0.55 0.20 0.19 0.32 0.67 0.84
      0.38 0.21 0.41 0.72 0.16 0.28 0.81];
12.85 12.8 12.85 12.85 12.85 ]; %aanpassen
68
69 for k=1:size(dataStruct,2)
      AppliedStrain(k) = PercentageAppliedStrain(k)/100;
      AreaSample(k) = (pi*(DiameterSample(k))^2) / 4;
```

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```
72 end
_{74} %% Obtain Stress and Relaxation Modulus for 24 hr data
76 for k=1:size(dataStruct,2)
      dataStruct_ROI(k).StressSample = dataStruct_ROI(k).ForcePosROI /
      AreaSample(k);
      dataStruct_ROI(k).StressRelMod = dataStruct_ROI(k).StressSample/
      AppliedStrain(k);
80 %% Stress Relaxation Modulus Fit
82 A = [];
83 B = [];
84 C = [];
86 [fitresult, gof] = CreatFit1_Powerlaw(dataStruct_ROI(k).TimeROI,
      dataStruct_ROI(k).StressRelMod); %nu nog met powerlaw, y=a*x^b+c
87 R2 = gof.rsquare;
88 Coefficients = coeffvalues(fitresult);
89 A(k) = Coefficients(1);
90 B(k) = Coefficients(2);
91 C(k) = Coefficients(3);
93 dataStruct_ROI(k).StressRelMod_fit1 = A(k)*dataStruct_ROI(k).TimeROI.^B
      (k)+C(k);
94 end
96 %% obtain ROI for 0.5 hour
98 dataStruct_ROI_half_hr = struct('TimeROI_half_hr',[], '
      ForcePosROI_half_hr',[],'StressSample_half_hr',[],'
      StressRelMod_half_hr',[],'StressRelMod_fit1_half_hr',[],'
      PercentageStressRelaxation_24hr_half_hr',[]);
99 EndTimeExperiment_half_hr = [];
100 MinimumDifference_half_hr = [];
101 MinForceI_half_hr = [];
102 MinForce_half_hr = [];
103 TimeROI_half_hr = [];
104 ForcePosROI_half_hr = [];
105 StressRelMod_half_hr = [];
106 StressSample_half_hr = [];
107 StressRelMod_fit1_half_hr = [];
108 PercentageStressRelaxation_24hr_half_hr = [];
DurationExperiment = 1800; % aanpassen
```

```
111
for k=1:size(dataStruct,2)
      if k==4 || k==9 || k==13 || k==16 || k==12
      EndTimeExperiment_half_hr(k) = dataStruct(k).Time(MaxForceI(k))+
114
      DurationExperiment;
       [MinimumDifference_half_hr(k), MinForceI_half_hr(k)]=min(abs(
      dataStruct(k).Time-EndTimeExperiment_half_hr(k)));
      MinForce_half_hr(k) = dataStruct(k).kNForce(MinForceI_half_hr(k));
      PercentageStressRelaxation_half_hr(k) = 100 - (MinForce_half_hr(k)
117
      / MaxForce(k) * 100); %handige value
118
      dataStruct_ROI_half_hr(k).TimeROI_half_hr = dataStruct(k).Time(
119
     MaxForceI(k):MinForceI_half_hr(k));
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr = dataStruct(k).
120
      kNForce(MaxForceI(k):MinForceI_half_hr(k))*-1;
      NumberOfEndRemovers_1hr = 0;
123
      while dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end) -
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end-1) > 5 ||
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end) -
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end-1) < -5
             dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end) = [];
             dataStruct_ROI_half_hr(k).TimeROI_half_hr(end) = [];
126
             NumberOfEndRemovers_1hr = NumberOfEndRemovers_1hr + 1;
127
      end
129
      else
      EndTimeExperiment_half_hr(k) = dataStruct(k).Time(MaxForceI(k))+
131
      DurationExperiment;
       [MinimumDifference_half_hr(k), MinForceI_half_hr(k)]=min(abs(
132
      dataStruct(k).Time-EndTimeExperiment_half_hr(k)));
      MinForce_half_hr(k) = dataStruct(k).Force(MinForceI_half_hr(k));
133
      PercentageStressRelaxation_half_hr(k) = 100 - (MinForce_half_hr(k)
134
      / MaxForce(k) * 100); %handige value
      dataStruct_ROI_half_hr(k).TimeROI_half_hr = dataStruct(k).Time(
     MaxForceI(k):MinForceI_half_hr(k));
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr = dataStruct(k).Force
137
      (MaxForceI(k):MinForceI_half_hr(k))*-1;
138
      NumberOfEndRemovers_1hr = 0;
139
140
      while dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end) -
141
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end-1) > 5 ||
      dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end) -
```

```
dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end-1) < -5
             dataStruct_ROI_half_hr(k).ForcePosROI_half_hr(end) = [];
             dataStruct_ROI_half_hr(k).TimeROI_half_hr(end) = [];
143
             NumberOfEndRemovers_1hr = NumberOfEndRemovers_1hr + 1;
144
      end
145
      end
146
147 end
149 %% obtain Stress & Relaxation modulus for 0.5 hour
151 for k=1:size(dataStruct,2)
dataStruct_ROI_half_hr(k).StressSample_half_hr = dataStruct_ROI_half_hr
      (k).ForcePosROI_half_hr ./ AreaSample(k);
dataStruct_ROI_half_hr(k).StressRelMod_half_hr = dataStruct_ROI_half_hr
      (k).StressSample_half_hr / AppliedStrain(k);
156 A_half_hr = [];
157 B_half_hr = [];
158 C_half_hr = [];
160 %% Stress Relaxation Modulus Fit for 0.5 hour data
161
162 [fitresult_half_hr, gof_half_hr] = CreatFit1_Powerlaw(
      dataStruct_ROI_half_hr(k).TimeROI_half_hr, dataStruct_ROI_half_hr(k)
      .StressRelMod_half_hr); %nu nog met powerlaw, y=a*x^b+c
R2_half_hr = gof_half_hr.rsquare;
164 Coefficients_half_hr = coeffvalues(fitresult_half_hr);
165 A_half_hr(k) = Coefficients_half_hr(1);
166 B_half_hr(k) = Coefficients_half_hr(2);
167 C_half_hr(k) = Coefficients_half_hr(3);
168
169
170 dataStruct_ROI_half_hr(k).StressRelMod_fit1_half_hr = A_half_hr(k)*
      dataStruct_ROI(k).TimeROI.^B_half_hr+C_half_hr;
dataStruct_ROI_half_hr(k).PercentageStressRelaxation_24hr_half_hr = 100
       - ((-dataStruct_ROI_half_hr(k).StressRelMod_fit1_half_hr(end)*
      AppliedStrain(k) * AreaSample(k)) / MaxForce(k) * 100);
173 %% Calculate relevant values between fits
174
175 % Error Determination
176
y0 = dataStruct_ROI(k).StressRelMod_fit1;
y1 = dataStruct_ROI_half_hr(k).StressRelMod_fit1_half_hr;
```

```
dy = y0 - y1; % error
abs_dy = abs(y0 - y1); % absolute error
relerr = abs(y0 - y1)./y0; % relative error
mean_err_SR = mean(abs(y0 - y1));
                                     % mean absolute error
MSE_SR = mean((y0 - y1).^2);
                                    % Mean square error
186 RMSE_SR = sqrt(mean((y0 - y1).^2)); % Root mean square error
mean_err_Force = mean_err_SR*AppliedStrain(k)*AreaSample(k);
MSE_SR_Force = MSE_SR*AppliedStrain(k)*AreaSample(k);
190 RMSE_SR_Force = RMSE_SR*AppliedStrain(k)*AreaSample(k);
191 end
193 %% Plot for Force versus Time - for ref
194
195 figure()
plot(dataStruct_ROI_half_hr(1).TimeROI_half_hr,dataStruct_ROI_half_hr
     (1).ForcePosROI_half_hr)
197 title('Force Versus Time')
198 xlabel('Time (S)')
199 ylabel('Force (N)')
201 %% Plot for E versus Time - for ref
203 figure()
204 plot(dataStruct_ROI(1).TimeROI, dataStruct_ROI(1).StressRelMod)
206 plot(dataStruct_ROI(1).TimeROI, dataStruct_ROI(1).StressRelMod_fit1,'
     LineWidth',2)
plot(dataStruct_ROI(1).TimeROI, dataStruct_ROI_half_hr(1).
     StressRelMod_fit1_half_hr,'LineWidth',2)
208 hold off
200 legend('Experimental Data', 'Fit 24 hours', 'Fit 0.5 hour')
210 xlabel('Time (S)')
211 ylabel('E (MPa)')
```

Appendix B

Schapery Method

```
1 % Run the MTSdatafitter_durationofstressrelaxation first
2 %% Log transformation & plot
4 BMD = [421.2866 331.5744 415.2768 483.8211 399.8866 227.2884 464.8261
     331.5744 431.4026 558.237 300.4444 445.4421 375.8612 344.386
     416.2005 304.0382];
5 \text{ BMD1} = [300.4444 \ 304.0382 \ 331.5744];
6 BMD2 = [331.7045 344.386 375.8612 399.8866];
7 BMD3 = [415.2768 416.2005 421.2866 431.4026];
8 \text{ BMD4} = [445.4421 \ 464.8261 \ 483.8211 \ 558.237];
10 BMD_r(1) = mean(BMD1);
BMD_r(2) = mean(BMD2);
12 BMD_r(3) = mean(BMD3);
BMD_r(4) = mean(BMD4);
15 epsilon = [0.0026 0.0018 0.0020 0.0055 0.0020 0.0019 0.0032 0.0067
     0.0084 0.0038 0.0021 0.0041 0.0072 0.0016 0.0028 0.0081];
16 epsilon_r = [0.002 0.0018 0.0019 0.0016]; % reference strain level
_{17} epsilon1 = [0.002 \ 0.0055 \ 0.0067];
18 epsilon2 = [0.0032 0.0018 0.0084 0.0028];
19 epsilon3 = [0.0021 0.0026 0.0019 0.0081];
20 epsilon4 = [0.0038 0.0041 0.0072 0.0016];
22 dataStruct_final = struct('t',[],'E',[],'Stress',[],'log_t',[],'log_E'
     ,[],'log_stress',[],'E_schapery',[]);
23
24 for k = 1:size(dataStruct,2)
      dataStruct_final(k).t = dataStruct_ROI_half_hr(k).TimeROI_half_hr -
      dataStruct_ROI_half_hr(k).TimeROI_half_hr(1);
      dataStruct_final(k).t(1) = 0.1;
      dataStruct_final(k).Stress = dataStruct_ROI_half_hr(k).
     StressSample_half_hr;
      dataStruct_final(k).E = dataStruct_final(k).Stress./(epsilon(k));
28
      dataStruct_final(k).log_t = log10(dataStruct_final(k).t);
29
      dataStruct_final(k).log_E = log10(dataStruct_final(k).E);
      dataStruct_final(k).log_stress = log10(dataStruct_final(k).Stress);
32 end
34 %% Fit reference strain - BMD Group 1
36 E_1 = dataStruct_final(3).E;
```

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```
37 Time_1 = dataStruct_final(3).t;
_{38} he(3) = 1; h2(3) = 1;
39 \text{ epsilon1} = [0.002 \ 0.0055 \ 0.0067];
_{41} G(1) = -913.2;
42 \text{ Ee}(1) = 1215;
n(1) = 0.01439;
46 %% Fit for other strain levels - BMD Group 1
48 \text{ for } k = 1:16
     if k == 8 || k == 4
49
               [fitresult_s1, gof_s1] = createFit_s1(dataStruct_final(k).t
50
      , dataStruct_final(k).E);
               Coeff_s1 = coeffvalues(fitresult_s1);
51
               he(k) = Coeff_s1(1);
               h2(k) = Coeff_s1(2);
      end
55 end
he_1 = [he(3) he(4) he(8)];
h2_1 = [h2(3) \ h2(4) \ h2(8)];
60 %% Fit reference strain - BMD Group 2
62 E_2 = dataStruct_final(2).E;
63 Time_2 = dataStruct_final(2).t;
64 \text{ he}(2) = 1; h2(2) = 1;
66 G(2) = -547.9;
67 \text{ Ee}(2) = 958.4;
n(2) = 0.0304;
_{70} %% Fit for other strain levels - BMD Group 2
_{72} for k = 1:16
    if k == 7 || k == 9 || k == 15
                [fitresult_s2, gof_s2] = createFit_s2(dataStruct_final(k).t
      , dataStruct_final(k).E);
               Coeff_s2 = coeffvalues(fitresult_s2);
75
               he(k) = Coeff_s2(1);
76
               h2(k) = Coeff_s2(2);
      end
79 end
```

```
he_2 = [he(7) he(2) he(9) he(15)];
h2_2 = [h2(7) \ h2(2) \ h2(9) \ h2(15)];
84 %% Fit reference strain - BMD Group 3
86 E_3 = dataStruct_final(6).E;
87 Time_3 = dataStruct_final(6).t;
88 he(6) = 1; h2(6) = 1;
90 G(3) = -384.5;
91 \text{ Ee}(3) = 1729;
92 n(3) = 0.08508;
94 %% Fit for other strain levels - BMD Group 3
96 \text{ for } k = 1:16
      if k == 11 || k == 1 || k == 16
                [fitresult_s3, gof_s3] = createFit_s3(dataStruct_final(k).t
      , dataStruct_final(k).E);
               Coeff_s3 = coeffvalues(fitresult_s3);
               he(k) = Coeff_s3(1);
100
               h2(k) = Coeff_s3(2);
101
      end
102
103 end
105 he_3 = [he(11) he(1) he(6) he(16)];
h2_3 = [h2(11) \ h2(1) \ h2(6) \ h2(16)];
108 %% Fit reference strain - BMD Group 4
110 E_4 = dataStruct_final(14).E;
Time_4 = dataStruct_final(14).t;
he(14) = 1; h2(14) = 1;
_{115} G(4) = -149.3;
116 \text{ Ee}(4) = 636.1;
n(4) = 0.1123;
119 %% Fit for other strain levels - BMD Group 4
120
121 for k = 1:16
      if k == 10 || k == 12 || k == 13
               [fitresult_s4, gof_s4] = createFit_s4(dataStruct_final(k).t
      , dataStruct_final(k).E);
               Coeff_s4 = coeffvalues(fitresult_s4);
```

```
he(k) = Coeff_s4(1);
                                       h2(k) = Coeff_s4(2);
                  end
127
128 end
130 \text{ he}_4 = [\text{he}(10) \text{ he}(12) \text{ he}(13) \text{ he}(14)];
h2_4 = [h2(10) \ h2(12) \ h2(13) \ h2(14)];
133 %% Weights for he & h2 fit
w_1 = [1000 \ 1 \ 1];
w_2 = [1000 \ 1 \ 1];
w_3 = [1000 \ 1 \ 1];
w_4 = [1000 \ 1 \ 1 \ 1];
_{140} %% Schapery fit & functions of he and h2
L = [213.6 - 145.8 - 42.7 - 20.19];
M = [0.5730 \ 2.609 \ 0.8303 \ 1.302];
144\ 0 = [188.9\ -126\ -37.4\ 15.43];
      P = [0.6223 \ 2.2661 \ 0.9224 \ 1.665];
146
_{147} for k = 1:16
                 if k == 8 || k == 4 || k ==3
148
                            f_h2(k) = L(1)*epsilon(k)+M(1);
149
                            f_he(k) = O(1) * epsilon(k) + P(1);
                            dataStruct_final(k).E_schapery = epsilon(k)*(f_he(k)*Ee(1) +
               f_h2(k)*G(1)*(dataStruct_final(k).t.^(n(1))));
                  elseif k == 7 || k == 2 || k == 9 || k ==15
152
                            f_h2(k) = L(2) * epsilon(k) + M(2);
153
                            f_he(k) = O(2) * epsilon(k) + P(2);
154
                            \label{lem:dataStruct_final(k).E_schapery = epsilon(k)*(f_he(k)*Ee(2) + f_he(k)*Ee(2)) + f_he(k)*Ee(2) + f_h
155
               f_h2(k)*G(2)*(dataStruct_final(k).t.^(n(2))));
                  elseif k == 11 || k == 1 || k == 6 || k == 16
                            f_h2(k) = L(3) * epsilon(k) + M(3);
                            f_he(k) = O(3) * epsilon(k) + P(3);
                             dataStruct_final(k).E_schapery = epsilon(k)*(f_he(k)*Ee(3) +
               f_h2(k)*G(3)*(dataStruct_final(k).t.^(n(3))));
                  else
160
                            f_h2(k) = L(4) * epsilon(k) + M(4);
161
                            f_he(k) = O(4) * epsilon(k) + P(4);
162
                            dataStruct_final(k).E_schapery = epsilon(k)*(f_he(k)*Ee(4)+
163
               f_h2(k)*G(4)*(dataStruct_final(k).t.^(n(4))));
                  end
164
165 end
```

```
f_{he_1} = [f_{he_3} f_{he_4} f_{he_5}];
f_h2_1 = [f_h2_3] f_h2_4 f_h2_8];
f_{he} = [f_{he}(7) f_{he}(2) f_{he}(9) f_{he}(15)];
f_h^2 = [f_h^2 = [f_h^2 = f_h^2 = f_
f_{he} = f_{he} = [f_{he} = (11) f_{he} = (1) f_{he} = (16) f_{he} = (16)];
f_h^2 = [f_h^2(11) f_h^2(1) f_h^2(6) f_h^2(16)];
f_he_4 = [f_he(10) f_he(12) f_he(13) f_he(14)];
f_h^2 = [f_h^2(10) f_h^2(12) f_h^2(13) f_h^2(14)];
175
        %% Plot experimental data with Schapery Model - BMD group 1
177
178
179 figure()
180 hold on
181 scatter(dataStruct_final(3).t,dataStruct_final(3).Stress)
182 hold on
plot(dataStruct_final(3).t,dataStruct_final(3).E_schapery,'LineWidth'
               ,2)
184 hold on
185 scatter(dataStruct_final(4).t,dataStruct_final(4).Stress)
plot(dataStruct_final(4).t,dataStruct_final(4).E_schapery,'LineWidth'
                ,2)
188 hold on
189 scatter(dataStruct_final(8).t,dataStruct_final(8).Stress)
190 hold on
plot(dataStruct_final(8).t,dataStruct_final(8).E_schapery,'LineWidth'
192 hold off
193 ylim([0 8])
194 xlim([0 2000])
        legend('BMD - 1, 0.2% strain', 'BMD - 1, 0.2% strain(fit)', 'BMD - 1,
              0.55% strain','BMD - 1, 0.55% strain(fit)','BMD - 1, 0.67% strain','
              BMD - 1, 0.67% strain(fit)')
         ylabel('Stress (MPa)');
         xlabel('Time (S)');
199 newcolors =
                                           [0.635 0.078 0.184
                                             0.635 0.078 0.184
200
                                             1.00 0.54 0.00
201
                                              1.00 0.54 0.00
202
                                              0.47 0.25 0.80
203
                                              0.47 0.25 0.80
204
                                              0.25 0.80 0.54
205
                                              0.25 0.80 0.54];
207 colororder (newcolors)
```

```
208
210 %% Plot experimental data with Schapery Model - BMD group 2
211
212 figure()
213 scatter(dataStruct_final(2).t,dataStruct_final(2).Stress)
214 hold on
plot(dataStruct_final(2).t,dataStruct_final(2).E_schapery,'LineWidth'
      ,2)
216 hold on
217 scatter(dataStruct_final(15).t,dataStruct_final(15).Stress)
218 hold on
219 plot(dataStruct_final(15).t,dataStruct_final(15).E_schapery,'LineWidth'
      , 2)
220 hold on
scatter(dataStruct_final(7).t,dataStruct_final(7).Stress)
222 hold on
plot(dataStruct_final(7).t,dataStruct_final(7).E_schapery,'LineWidth'
      , 2)
224 hold on
225 scatter(dataStruct_final(9).t,dataStruct_final(9).Stress)
plot(dataStruct_final(9).t,dataStruct_final(9).E_schapery,'LineWidth'
      ,2)
228 hold off
229 ylim([0 8])
230 xlim([0 2000])
  legend('BMD - 2, 0.18% strain','BMD - 2, 0.18% strain(fit)','BMD - 2,
      0.28% strain','BMD - 2, 0.28% strain(fit)','BMD - 2, 0.32% strain','
      BMD - 2, 0.32% strain(fit)', 'BMD - 2, 0.84% strain', 'BMD - 2, 0.84%
     strain(fit)')
   ylabel('Stress (MPa)');
232
   xlabel('Time (S)');
233
234
    newcolors = [0.635 \ 0.078 \ 0.184]
235
                  0.635 0.078 0.184
                  1.00 0.54 0.00
                  1.00 0.54 0.00
                  0.47 0.25 0.80
239
                  0.47 0.25 0.80
240
                  0.25 0.80 0.54
241
                  0.25 0.80 0.54];
243 colororder (newcolors)
_{245} %% Plot experimental data with Schapery Model - BMD group 3
```

```
247 figure()
248 scatter(dataStruct_final(6).t,dataStruct_final(6).Stress)
plot(dataStruct_final(6).t,dataStruct_final(6).E_schapery,'LineWidth'
      , 2)
251 hold on
252 scatter(dataStruct_final(11).t,dataStruct_final(11).Stress)
253 hold on
254 plot(dataStruct_final(11).t,dataStruct_final(11).E_schapery,'LineWidth'
      ,2)
255 hold on
256 scatter(dataStruct_final(1).t,dataStruct_final(1).Stress)
257 hold on
258 plot(dataStruct_final(1).t,dataStruct_final(1).E_schapery,'LineWidth'
      ,2)
259 hold on
260 scatter(dataStruct_final(16).t,dataStruct_final(16).Stress)
261 hold on
262 plot(dataStruct_final(16).t,dataStruct_final(16).E_schapery,'LineWidth'
      ,2)
263 hold off
264 ylim([0 8])
265 xlim([0 2000])
legend('BMD - 3, 0.19% strain', 'BMD - 3, 0.19% strain(fit)', 'BMD - 3,
      0.21% strain','BMD - 3, 0.21% strain(fit)','BMD - 3, 0.26% strain','
      BMD - 3, 0.26% strain(fit)', 'BMD - 3, 0.81% strain', 'BMD - 3, 0.81%
      strain(fit)')
   ylabel('Stress (MPa)');
   xlabel('Time (S)');
268
269
      newcolors = [0.635 \ 0.078 \ 0.184]
270
                  0.635 0.078 0.184
271
                  1.00 0.54 0.00
272
                  1.00 0.54 0.00
273
                  0.47 0.25 0.80
                  0.47 0.25 0.80
                  0.25 0.80 0.54
                  0.25 0.80 0.54];
278 colororder (newcolors)
279
   %% Plot experimental data with Schapery Model - BMD group 4
280
282 figure()
283 scatter(dataStruct_final(14).t,dataStruct_final(14).Stress)
284 hold on
```

```
plot(dataStruct_final(14).t,dataStruct_final(14).E_schapery,'LineWidth'
      , 2)
286 hold on
287 scatter(dataStruct_final(10).t,dataStruct_final(10).Stress)
288 hold on
289 plot(dataStruct_final(10).t,dataStruct_final(10).E_schapery,'LineWidth'
      ,2)
290 hold on
291 scatter(dataStruct_final(12).t,dataStruct_final(12).Stress)
292 hold on
plot(dataStruct_final(12).t,dataStruct_final(12).E_schapery,'LineWidth'
      , 2)
294 hold on
295 scatter(dataStruct_final(13).t,dataStruct_final(13).Stress)
296 hold on
plot(dataStruct_final(13).t,dataStruct_final(13).E_schapery,'LineWidth'
      ,2)
298 hold off
299 ylim([0 8])
300 xlim([0 2000])
   legend('BMD - 4, 0.16% strain', 'BMD - 4, 0.16% strain(fit)', 'BMD - 4,
      0.38% strain','BMD - 4, 0.38% strain(fit)','BMD - 4, 0.41% strain','
      BMD - 4, 0.41% strain(fit)', 'BMD - 4, 0.72% strain', 'BMD - 4, 0.72%
      strain(fit)')
   ylabel('Stress (MPa)');
302
   xlabel('Time (S)');
303
304
      newcolors = [0.635 \ 0.078 \ 0.184]
                  0.635 0.078 0.184
306
                  1.00 0.54 0.00
307
                  1.00 0.54 0.00
308
                  0.47 0.25 0.80
309
                  0.47 0.25 0.80
310
                  0.25 0.80 0.54
311
                  0.25 0.80 0.54];
313 colororder (newcolors)
315 %% Calculate relevant values between fits
317 % Error Determination
318
struct('y00',[],'y11',[],'dy1',[],'abs_dy1',[],'relerr1',[],'
      pererr1',[]);
320 \text{ for } k = 1:16
322 error(k).y00 = dataStruct_final(k).Stress;
```

```
error(k).y11 = dataStruct_final(k).E_schapery;
error(k).dy1 = error(k).y00 - error(k).y11 ; % error
326 error(k).relerr1 = error(k).abs_dy1./ y00 ; % relative error
327 error(k).pererr1 = error(k).relerr1*100 ;
                                             % percentage error
mean_err_SR1(k) = mean(abs(error(k).y00 - error(k).y11));
                                                               % mean
     absolute error
MSE_SR1(k) = mean((error(k).y00 - error(k).y11).^2);
                                                              % Mean
     square error
RMSE_SR1(k) = sqrt(mean((error(k).y00 - error(k).y11).^2)); % Root
     mean square error
331
mean_RMSE1 = mean(RMSE_SR1); % Mean Root Mean Square error
mean_pererr1(k) = mean(error(k).pererr1(end));
mean_perc = mean(mean_pererr1);
336 %% Correlation between he, h2 & strain
338 % plot he & h2 for BMD group 1
340 figure()
341 hold on
scatter(epsilon1, he_1,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7
     0.71):
343 hold on
344 plot(epsilon1, f_he_1,'b');
345 hold on
346 scatter(epsilon1, h2_1,'MarkerEdgeColor','r','MarkerFaceColor',[0.8500
     0.3250 0.0980]);
347 hold on
348 plot(epsilon1, f_h2_1,'r');
349 hold off
350 title('BMD group 1')
legend('he - actual data','he - fit','h2 - actual data','h2 - fit')
352 ylabel('he & h2');
   xlabel('Strain');
355 %% plot he & h2 for BMD group 2
357 figure()
358 hold on
scatter(epsilon2, he_2,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7
     0.7]);
360 hold on
general state | plot(epsilon2, f_he_2,'b');
362 hold on
```

```
scatter(epsilon2, h2_2,'MarkerEdgeColor','r','MarkerFaceColor',[0.8500
      0.3250 0.0980]);
364 hold on
365 plot (epsilon2, f_h2_2, 'r');
366 hold off
367 title('BMD group 2')
   legend('he - actual data','he - fit','h2 - actual data','h2 - fit')
   ylabel('he & h2');
   xlabel('Strain');
370
   %% plot he & h2 for BMD group 3
372
373
374 figure()
375 hold on
scatter(epsilon3, he_3,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7
      0.7]);
377 hold on
plot(epsilon3, f_he_3,'b');
379 hold on
scatter(epsilon3, h2_3,'MarkerEdgeColor','r','MarkerFaceColor',[0.8500
      0.3250 0.0980]);
381 hold on
382 plot(epsilon3, f_h2_3,'r');
383 hold off
384 title('BMD group 3')
   legend('he - actual data','he - fit','h2 - actual data','h2 - fit')
   ylabel('he & h2');
   xlabel('Strain');
388
   %% plot he & h2 for BMD group 4
389
390
391 figure()
392 hold on
scatter(epsilon4, he_4,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7
      0.7]);
394 hold on
395 plot(epsilon4, f_he_4,'b');
396 hold on
scatter(epsilon4, h2_4,'MarkerEdgeColor','r','MarkerFaceColor',[0.8500
      0.3250 0.0980]);
398 hold on
399 plot(epsilon4, f_h2_4,'r');
400 hold off
401 title('BMD group 4')
legend('he - actual data', 'he - fit', 'h2 - actual data', 'h2 - fit')
403 ylabel('he & h2');
```

```
xlabel('Strain');
406 %% Overall he & h2 to predict correlation
407
408 figure()
409 hold on
       scatter(epsilon,he,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7
      0.7]);
411 hold on
       scatter(epsilon,h2,'MarkerEdgeColor','r','MarkerFaceColor',[0.8500
      0.3250 0.0980]);
413 hold off
414 legend('he','h2')
415 ylabel('he & h2');
416 xlabel('Strain')
418 %% Correlation between G, n & BMD
420 for k=1:4
f_g(k) = 3.795*BMD_r(k) - 1982;
f_n(k) = 0.0005311*BMD_r(k) - 0.147;
423 end
424
  %% plot G & n vs BMD
425
427 figure()
428 hold on
429 scatter(BMD_r, G,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7 0.7]);
430 hold on
431 plot(BMD_r, f_g);
432 hold off
title('C vs BMD')
1434 legend('C','C - fit')
   ylabel('C');
435
   xlabel('Reference BMD');
438 figure()
439 hold on
440 scatter(BMD_r, n, 'MarkerEdgeColor', 'b', 'MarkerFaceColor', [0 0.7 0.7]);
441 hold on
442 plot(BMD_r, f_n);
443 hold off
title('n vs BMD')
1445 legend('n','n - fit')
446 ylabel('n');
xlabel('Reference BMD');
```

```
448
449 %% Correlation between he,h2 & BMD
450
451 % Overall he & h2 to predict correlation
452
453 figure()
454 hold on
455 scatter(BMD,he,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7 0.7])
456 hold on
457 scatter(BMD,h2,'MarkerEdgeColor','r','MarkerFaceColor',[0.8500 0.3250 0.0980]);
458 hold off
459 legend('he','h2')
460 ylabel('he & h2');
461 xlabel('BMD')
```

Appendix C

Modified Superposition Method

```
1 % Run the MTSdatafitter_durationofstressrelaxation first
3 %% Log transformation & plot
5 BMD = [421.2866 331.5744 415.2768 483.8211 399.8866 227.2884 464.8261
     331.5744 431.4026 558.237 300.4444 445.4421 375.8612 344.386
     416.2005 304.0382];
_{6} BMD (5) = [];
7 \text{ BMD1} = [300.4444 \ 304.0382 \ 331.5744];
8 BMD2 = [331.7045 344.386 375.8612 399.8866];
9 \text{ BMD3} = [415.2768 \ 416.2005 \ 421.2866 \ 431.4026];
10 \text{ BMD4} = [445.4421 \ 464.8261 \ 483.8211 \ 558.237];
BMD_r(1) = mean(BMD1);
BMD_r(2) = mean(BMD2);
14 BMD_r(3) = mean(BMD3);
15 BMD_r(4) = mean(BMD4);
17 epsilon = [0.0026 0.0018 0.0020 0.0055 0.0020 0.0019 0.0032 0.0067
     0.0084 0.0038 0.0021 0.0041 0.0072 0.0016 0.0028 0.0081];
18 epsilon_r = [0.002 0.0018 0.0019 0.0016]; % reference strain level
19 epsilon1 = [0.002 \ 0.0055 \ 0.0067];
20 epsilon2 = [0.0032 0.0018 0.0084 0.0028];
21 epsilon3 = [0.0021 0.0026 0.0019 0.0081];
22 epsilon4 = [0.0038 0.0041 0.0072 0.0016];
24 dataStruct_final = struct('t',[],'E',[],'Stress',[],'log_t',[],'log_E'
      ,[],'log_stress',[],'E_MSM',[]);
25
26 for k = 1:size(dataStruct,2)
      dataStruct_final(k).t = dataStruct_ROI_half_hr(k).TimeROI_half_hr -
      dataStruct_ROI_half_hr(k).TimeROI_half_hr(1);
      dataStruct_final(k).t(1) = 0.1;
      dataStruct_final(k).Stress = dataStruct_ROI_half_hr(k).
     StressSample_half_hr;
      dataStruct_final(k).E = dataStruct_final(k).Stress./(epsilon(k));
      dataStruct_final(k).log_t = log10(dataStruct_final(k).t);
      dataStruct_final(k).log_E = log10(dataStruct_final(k).E);
      dataStruct_final(k).log_stress = log10(dataStruct_final(k).Stress);
34 end
36 %% Slope of Stress Vs time graph
```

```
38 coefficients_slope = struct('Coefficients_slope',[]);
39 for k=1:16
40 coefficients_slope(k).Coefficients_slope = polyfit(dataStruct_final(k).
     log_t, dataStruct_final(k).log_stress, 1);
slope(k) = coefficients_slope(k).Coefficients_slope(1);
42 end
43
44 %% BMD - group 1
46 figure
47 % hold on
           scatter(dataStruct_final(5).log_t, dataStruct_final(5).
     log_stress)
49 hold on
         scatter(dataStruct_final(3).log_t, dataStruct_final(3).
     log_stress)
51 hold on
         scatter(dataStruct_final(8).log_t, dataStruct_final(8).
     log_stress)
53 hold on
         scatter(dataStruct_final(4).log_t, dataStruct_final(4).
     log_stress)
55 hold off
     legend('BMD - 1, 0.2% strain','BMD - 1, 0.67% strain','BMD - 1,
     0.55% strain')
     ylabel('log stress');
      xlabel('log time');
60 %% BMD group 1 - slope
62 slope1 = [slope(3) slope(4) slope(8)];
64 %% Polynomial Fit
66 [fitresult_1, gof_1] = createFit_poly(epsilon1, slope1);
68 Coefficients_poly1 = coeffvalues(fitresult_1);
69 D(1) = Coefficients_poly1(1);
F(1) = Coefficients_poly1(2);
72 Stress_fit1 = []; % B(epsilon)
73
   Stress_fit1(3) = D(1)*(epsilon1(1)) + F(1);
   Stress_fit1(4) = D(1)*(epsilon1(2)) + F(1);
   Stress_fit1(8) = D(1)*(epsilon1(3)) + F(1);
```

```
78 %% Initial Stress Relaxation Modulus
80 E1(1) = dataStruct_final(3).E(1);
81 E1(2) = dataStruct_final(4).E(1);
82 E1(3) = dataStruct_final(8).E(1);
84 %% Linear Fit
86 [fitresult1, gof1] = createFit_linear(epsilon1, E1);
87 Coefficients_linear1 = coeffvalues(fitresult1);
88 H(1) = Coefficients_linear1(1);
89 J(1) = Coefficients_linear1(2);
91 E1_fit = []; % A(epsilon)
93 E1_{fit}(3) = H(1)*(epsilon1(1)) + J(1);
94 E1_{fit}(4) = H(1)*(epsilon1(2)) + J(1);
95 E1_fit(8) = H(1)*(epsilon1(3)) + J(1);
97 %% BMD - group 2
99 figure
100 hold on
          scatter(dataStruct_final(7).log_t, dataStruct_final(7).
      log_stress)
102 hold on
         scatter(dataStruct_final(2).log_t, dataStruct_final(2).
     log_stress)
         scatter(dataStruct_final(9).log_t, dataStruct_final(9).
     log_stress)
106 hold on
         scatter(dataStruct_final(15).log_t, dataStruct_final(15).
     log_stress)
108 hold off
      legend('BMD - 2, 0.32% strain','BMD - 2, 0.18% strain','BMD - 2,
      0.84% strain', 'BMD - 2, 0.28% strain')
      ylabel('log stress');
110
      xlabel('log time');
      %% BMD group 2 - slope
slope2 = [slope(7) slope(2) slope(9) slope(15)];
117 %% Polynomial Fit
```

```
118 [fitresult_2, gof_2] = createFit_poly(epsilon2, slope2);
120 Coefficients_poly2 = coeffvalues(fitresult_2);
121 D(2) = Coefficients_poly2(1);
F(2) = Coefficients_poly2(2);
124 \text{ Stress\_fit1}(7) = D(2)*(epsilon2(1)) + F(2);
125 Stress_fit1(2) = D(2)*(epsilon2(2)) + F(2);
126 Stress_fit1(9) = D(2)*(epsilon2(3)) + F(2);
127 \text{ Stress\_fit1}(15) = D(2)*(epsilon2(4)) + F(2);
128 %% Initial Stress Relaxation Modulus
130 E2(1) = dataStruct_final(7).E(1);
131 E2(2) = dataStruct_final(2).E(1);
132 E2(3) = dataStruct_final(9).E(1);
133 E2(4) = dataStruct_final(15).E(1);
135 %% Linear Fit
136 [fitresult2, gof2] = createFit_linear(epsilon2, E2);
137 Coefficients_linear2 = coeffvalues(fitresult2);
138 H(2) = Coefficients_linear2(1);
139 J(2) = Coefficients_linear2(2);
140
E1_{fit}(7) = H(2)*(epsilon2(1)) + J(2);
142 E1_fit(2) = H(2)*(epsilon2(2)) + J(2);
143 E1_fit(9) = H(2)*(epsilon2(3)) + J(2);
144 E1_fit(15) = H(2)*(epsilon2(4)) + J(2);
146 %% BMD - group 3
148 figure
149 hold on
          scatter(dataStruct_final(11).log_t, dataStruct_final(11).
     log_stress)
151 hold on
          scatter(dataStruct_final(1).log_t, dataStruct_final(1).
      log_stress)
          scatter(dataStruct_final(6).log_t, dataStruct_final(6).
     log_stress)
155 hold on
          scatter(dataStruct_final(16).log_t, dataStruct_final(16).
     log_stress)
157 hold off
      legend('BMD - 3, 0.21% strain','BMD - 3, 0.26% strain','BMD - 3,
      0.19% strain', 'BMD - 3, 0.81% strain')
```

```
ylabel('log stress');
      xlabel('log time');
161
162 %% BMD group 3 - slope
slope3 = [slope(11) slope(1) slope(6) slope(16)];
165
166 %% Polynomial Fit
[fitresult_3, gof_3] = createFit_poly(epsilon3, slope3);
169 Coefficients_poly3 = coeffvalues(fitresult_3);
170 D(3) = Coefficients_poly3(1);
F(3) = Coefficients_poly3(2);
173 Stress_fit1(11) = D(3)*(epsilon3(1)) + F(3);
174 \text{ Stress\_fit1}(1) = D(3)*(epsilon3(2)) + F(3);
Stress_fit1(6) = D(3)*(epsilon3(3)) + F(3);
176 Stress_fit1(16) = D(3)*(epsilon3(4)) + F(3);
178 %% Initial Stress Relaxation Modulus
180 E3(1) = dataStruct_final(11).E(1);
181 E3(2) = dataStruct_final(1).E(1);
182 E3(3) = dataStruct_final(6).E(1);
183 E3(4) = dataStruct_final(16).E(1);
184
185 %% Linear Fit
186 [fitresult3, gof3] = createFit_linear(epsilon3, E3);
187 Coefficients_linear3 = coeffvalues(fitresult3);
188 H(3) = Coefficients_linear3(1);
189 J(3) = Coefficients_linear3(2);
190
191 E1_{fit}(11) = H(3)*(epsilon3(1)) + J(3);
192 E1_fit(1) = H(3)*(epsilon3(2)) + J(3);
193 E1_{fit}(6) = H(3)*(epsilon3(3)) + J(3);
194 E1_fit(16) = H(3)*(epsilon3(4)) + J(3);
196 %% BMD - group 4
198 figure
199 hold on
          scatter(dataStruct_final(10).log_t, dataStruct_final(10).
     log_stress)
201 hold on
          scatter(dataStruct_final(12).log_t, dataStruct_final(12).
      log_stress)
```

```
203 hold on
          scatter(dataStruct_final(13).log_t, dataStruct_final(13).
      log_stress)
205 hold on
          scatter(dataStruct_final(14).log_t, dataStruct_final(14).
206
      log_stress)
207 hold off
      legend('BMD - 4, 0.38% strain','BMD - 4, 0.41% strain','BMD - 4,
208
      0.72% strain', 'BMD - 4, 0.16% strain')
      ylabel('log stress');
      xlabel('log time');
210
211
212 %% BMD group 4 - slope
214
slope4 = [slope(10) slope(12) slope(13) slope(14)];
217 %% Polynomial Fit
218 [fitresult_4, gof_4] = createFit_poly(epsilon4, slope4);
220 Coefficients_poly4 = coeffvalues(fitresult_4);
221 D(4) = Coefficients_poly4(1);
F(4) = Coefficients_poly4(2);
224 \text{ Stress\_fit1}(10) = D(4)*(epsilon4(1)) + F(4);
225 Stress_fit1(12) = D(4)*(epsilon4(2)) + F(4);
226 Stress_fit1(13) = D(4)*(epsilon4(3)) + F(4);
227 Stress_fit1(14) = D(4)*(epsilon4(4)) + F(4);
229 %% Initial Stress Relaxation Modulus
231 E4(1) = dataStruct_final(10).E(1);
232 E4(2) = dataStruct_final(12).E(1);
233 E4(3) = dataStruct_final(13).E(1);
E4(4) = dataStruct_final(14).E(1);
236 %% Linear Fit
237 [fitresult4, gof4] = createFit_linear(epsilon4, E4);
238 Coefficients_linear4 = coeffvalues(fitresult4);
H(4) = Coefficients_linear4(1);
240 J(4) = Coefficients_linear4(2);
242 E1_fit(10) = H(4)*(epsilon4(1)) + J(4);
E1_{fit}(12) = H(4)*(epsilon4(2)) + J(4);
E1_{fit}(13) = H(4)*(epsilon4(3)) + J(4);
E1_{fit}(14) = H(4)*(epsilon4(4)) + J(4);
```

```
247 %% Modified Superposition Method
_{249} for k = 1:16
       dataStruct_final(k).E_MSM = E1_fit(k)*epsilon(k)*((dataStruct_final
250
      (k).t).^Stress_fit1(k));
251 end
252
       %% BMD group 1 plot
253
255 figure()
256 scatter(dataStruct_final(3).t,dataStruct_final(3).Stress)
257 hold on
plot(dataStruct_final(3).t,dataStruct_final(3).E_MSM,'LineWidth',2)
259 hold on
260 scatter(dataStruct_final(4).t,dataStruct_final(4).Stress)
261 hold on
262 plot(dataStruct_final(4).t,dataStruct_final(4).E_MSM,'LineWidth',2)
263 hold on
264 scatter(dataStruct_final(8).t,dataStruct_final(8).Stress)
plot(dataStruct_final(8).t,dataStruct_final(8).E_MSM,'LineWidth',2)
267 hold off
268 ylim([0 8])
269 xlim([0 2000])
   legend('BMD - 1, 0.2% strain', 'BMD - 1, 0.2% strain(fit)', 'BMD - 1,
      0.55% strain','BMD - 1, 0.55% strain(fit)','BMD - 1, 0.67% strain','
      BMD - 1, 0.67% strain(fit)')
   ylabel('Stress');
   xlabel('Time');
272
273
      newcolors = [0.635 \ 0.078 \ 0.184]
274
                  0.635 0.078 0.184
275
                   1.00 0.54 0.00
276
                   1.00 0.54 0.00
                   0.47 0.25 0.80
                   0.47 0.25 0.80
                   0.25 0.80 0.54
280
                   0.25 0.80 0.54];
282 colororder (newcolors)
283
   %% BMD group 2 plot
284
285
286 figure()
scatter(dataStruct_final(2).t,dataStruct_final(2).Stress)
288 hold on
```

```
plot(dataStruct_final(2).t,dataStruct_final(2).E_MSM,'LineWidth',2)
291 scatter(dataStruct_final(15).t,dataStruct_final(15).Stress)
292 hold on
plot(dataStruct_final(15).t,dataStruct_final(15).E_MSM,'LineWidth',2)
294 hold on
295 scatter(dataStruct_final(7).t,dataStruct_final(7).Stress)
296 hold on
plot(dataStruct_final(7).t,dataStruct_final(7).E_MSM,'LineWidth',2)
298 hold on
299 scatter(dataStruct_final(9).t,dataStruct_final(9).Stress)
300 hold on
got plot(dataStruct_final(9).t,dataStruct_final(9).E_MSM,'LineWidth',2)
302 hold off
303 ylim([0 8])
304 xlim([0 2000])
   legend('BMD - 2, 0.18% strain','BMD - 2, 0.18% strain(fit)','BMD - 2,
      0.28% strain','BMD - 2, 0.28% strain(fit)','BMD - 2, 0.32% strain','
      BMD - 2, 0.32% strain(fit)','BMD - 2, 0.84% strain','BMD - 2, 0.84%
      strain(fit)')
   ylabel('Stress');
306
   xlabel('Time');
307
308
     newcolors = [0.635 \ 0.078 \ 0.184]
309
                  0.635 0.078 0.184
310
                  1.00 0.54 0.00
                  1.00 0.54 0.00
                  0.47 0.25 0.80
                  0.47 0.25 0.80
314
                  0.25 0.80 0.54
315
                  0.25 \ 0.80 \ 0.54;
317 colororder (newcolors)
318
   %% BMD group 3 plot
319
   figure()
   scatter(dataStruct_final(6).t,dataStruct_final(6).Stress)
plot(dataStruct_final(6).t,dataStruct_final(6).E_MSM,'LineWidth',2)
325 hold on
scatter(dataStruct_final(11).t,dataStruct_final(11).Stress)
327 hold on
328 plot(dataStruct_final(11).t,dataStruct_final(11).E_MSM,'LineWidth',2)
329 hold on
scatter(dataStruct_final(1).t,dataStruct_final(1).Stress)
331 hold on
```

```
332 plot(dataStruct_final(1).t,dataStruct_final(1).E_MSM,'LineWidth',2)
334 scatter(dataStruct_final(16).t,dataStruct_final(16).Stress)
335 hold on
336 plot(dataStruct_final(16).t,dataStruct_final(16).E_MSM,'LineWidth',2)
337 hold off
338 ylim([0 8])
339 xlim([0 2000])
   legend('BMD - 3, 0.19% strain','BMD - 3, 0.19% strain(fit)','BMD - 3,
      0.21% strain', 'BMD - 3, 0.21% strain(fit)', 'BMD - 3, 0.26% strain', '
      BMD - 3, 0.26% strain(fit)','BMD - 3, 0.81% strain','BMD - 3, 0.81%
      strain(fit)')
   ylabel('Stress');
341
   xlabel('Time');
342
343
     newcolors = [0.635 \ 0.078 \ 0.184]
344
                  0.635 0.078 0.184
                  1.00 0.54 0.00
                  1.00 0.54 0.00
                  0.47 0.25 0.80
348
                  0.47 0.25 0.80
349
                  0.25 0.80 0.54
350
                  0.25 0.80 0.54];
351
352 colororder (newcolors)
353
   %% BMD group 4 plot
354
355
   figure()
357 scatter(dataStruct_final(14).t,dataStruct_final(14).Stress)
358 hold on
s59 plot(dataStruct_final(14).t,dataStruct_final(14).E_MSM,'LineWidth',2)
360 hold on
361 scatter(dataStruct_final(10).t,dataStruct_final(10).Stress)
362 hold on
363 plot(dataStruct_final(10).t,dataStruct_final(10).E_MSM,'LineWidth',2)
364 hold on
scatter(dataStruct_final(12).t,dataStruct_final(12).Stress)
plot(dataStruct_final(12).t,dataStruct_final(12).E_MSM,'LineWidth',2)
368 hold on
scatter(dataStruct_final(13).t,dataStruct_final(13).Stress)
370 hold on
371 plot(dataStruct_final(13).t,dataStruct_final(13).E_MSM,'LineWidth',2)
372 hold off
373 ylim([0 8])
374 xlim([0 2000])
```

```
legend('BMD - 4, 0.16% strain', 'BMD - 4, 0.16% strain(fit)', 'BMD - 4,
      0.38% strain','BMD - 4, 0.38% strain(fit)','BMD - 4, 0.41% strain','
      BMD - 4, 0.41% strain(fit)','BMD - 4, 0.72% strain','BMD - 4, 0.72%
      strain(fit)')
   ylabel('Stress');
   xlabel('Time');
377
378
      newcolors = [0.635 \ 0.078 \ 0.184]
379
                  0.635 0.078 0.184
                  1.00 0.54 0.00
                  1.00 0.54 0.00
382
                  0.47 0.25 0.80
383
                  0.47 0.25 0.80
384
                  0.25 0.80 0.54
385
                  0.25 0.80 0.54];
386
  colororder (newcolors)
387
388
   %% Calculate relevant values between fits
391 % Error Determination
393 error = struct('y00',[],'y11',[],'dy1',[],'abs_dy1',[],'relerr1',[],'
      pererr1',[]);
394 \text{ for } k = 1:16
395
396 error(k).y00 = dataStruct_final(k).Stress;
397 error(k).y11 = dataStruct_final(k).E_MSM;
error(k).dy1 = error(k).y00 - error(k).y11 ; % error
ass error(k).abs_dy1 = abs(error(k).dy1);  % absolute error
400 error(k).relerr1 = error(k).abs_dy1./ y00 ; % relative error
401 error(k).pererr1 = error(k).relerr1*100; % percentage error
mean_err_SR1(k) = mean(abs(error(k).y00 - error(k).y11));
                                                                    % mean
      absolute error
MSE_SR1(k) = mean((error(k).y00 - error(k).y11).^2);
                                                                   % Mean
      square error
404 RMSE_SR1(k) = sqrt(mean((error(k).y00 - error(k).y11).^2)); % Root
      mean square error
406 mean_RMSE1 = mean(RMSE_SR1); % Mean Root Mean Square error
407 mean_pererr1(k) = mean(error(k).pererr1(end));
408 mean_perc = mean(mean_pererr1);
409 end
410
   %% Correlation between H, J, D, F and BMD
_{413} for k = 1:4
```

```
f_H(k) = -140*BMD_r(k) + 59330;
414
      f_J(k) = 1.788*BMD_r(k) + 18.48;
      f_D(k) = 0.01165*BMD_r(k) - 2.422;
      f_F(k) = -0.0001626*BMD_r(k) - 0.004826;
417
418 end
419
420 %% plot H, J, D & F vs BMD
422 figure()
423 hold on
424 scatter(BMD_r, H,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7 0.7]);
425 hold on
426 plot(BMD_r, f_H,'b');
427 hold off
title('H vs BMD')
   legend('H','H - fit')
430 ylabel('H');
   xlabel('Reference BMD');
433 figure()
434 hold on
435 scatter(BMD_r, J,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7 0.7]);
436 hold on
437 plot(BMD_r, f_J,'b');
438 hold off
title('J vs BMD')
440 legend('J','J - fit')
   ylabel('J');
   xlabel('Reference BMD');
444 figure()
445 hold on
446 scatter(BMD_r, D,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7 0.7]);
447 hold on
448 plot(BMD_r, f_D,'b');
449 hold off
title('D vs BMD')
legend('D','D - fit')
   ylabel('D');
   xlabel('Reference BMD');
453
454
figure()
456 hold on
457 scatter(BMD_r, F,'MarkerEdgeColor','b','MarkerFaceColor',[0 0.7 0.7]);
458 hold on
459 plot(BMD_r, f_F,'b');
```

```
460 hold off
461 title('F vs BMD')
462 legend('F','F - fit')
463 ylabel('F');
464 xlabel('Reference BMD');
```