

# Automatic Measurement of Cartilage Thickness in the Human Knee

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## 1 Introduction

This paper seeks to discuss different methods to automate the measurement of cartilage thickness in the human knee using MRI scans. The methods are based on previous work by Wolfgang Wirth and Felix Eckstein. [WE08]

All computational methods make use of segmented MRI scans of the human knee, from the OAI dataset, and are written in Python. More precisely, there were two distinct sets of segmentations used: One set of manually segmented images, containing 507 samples, and one set of automatically segmented images, containing 24,783 samples. The set of automatic segmentations contained samples from the entire OAI dataset, i.e. over all time periods. Manual segmentations are stored as MHD and automatic segmentations as NIFTI files, respectively, which can be read and converted into Numpy arrays using the SimpleITK library. These arrays map each point in the three-dimensional space of the scan to an integer encoding, meaning for example points belonging to the femoral cartilage are assigned a value of 3. This makes isolating and extracting the cartilage volumes straightforward. Four different methods have been developed to determine mean cartilage thickness, plus other statistical measures. Thickness was measured for different subregions of the cartilage plate, which were determined as referenced in [WE08]. For a more detailed description of the procedure, refer to appendix A.

## 2 Mean cartilage thickness using meshes

This is a three-dimensional approach using meshes and normal vectors to determine the mean thickness of a cartilage volume. The main idea is building an upper and a lower mesh, and calculating the average distance between the two, for example by ray tracing along the normal vectors of the lower mesh against the upper mesh. Other methods, like a K-D-tree nearest neighbour search, are also possible. This works well for the tibial cartilage, because due to its physical shape, it is possible to simply group each point by  $x$  and  $y$  coordinates and add the point with the highest (lowest)  $z$  coordinate to the upper (lower) mesh. The result is two point clouds consisting of vectors  $(x, y, z)$ , as seen in figure 1 (red points make up the upper cloud, green the lower). These can then be converted into polygon meshes using the delaunay algorithm, which in turn allows for determining the normal vectors of the respective meshes. Mean cartilage thickness can be calculated by tracing along the normal vectors of the lower mesh against the upper mesh. For other methods not making use of the normals, like the previously mentioned nearest neighbour search, the delaunay conversion is optional.

The femoral cartilage makes things a bit more tricky. Due to its shape, building an upper and lower mesh is not trivial, because while for the tibial cartilage, there was only one respective vector for each coordinate pair  $(x, y)$ , there may now be multiple for the areas where the volume describes a curve, so just taking the minimum or maximum  $z$  coordinate no longer suffices. Using the

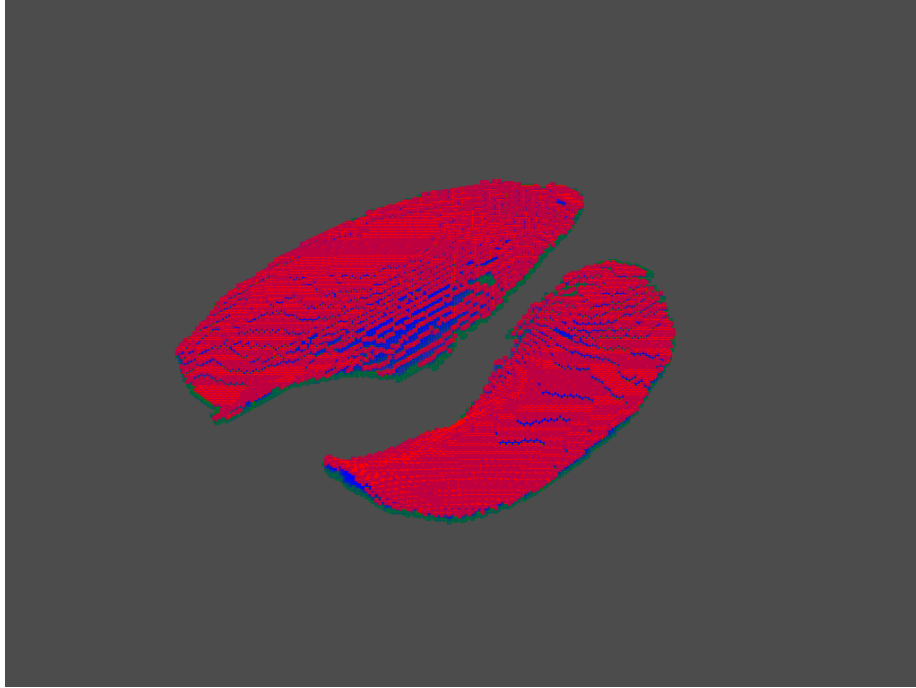


Figure 1: Point clouds of the tibial cartilage

previous approach results in point clouds where some areas are left bare, as illustrated in figure 2. One solution, used in this approach, is splitting the volume into parts, and rotating the curved sections, such that it is once again possible to choose points according to the  $z$  coordinate. For a detailed description of how this splitting is achieved, refer to appendix B. This allows for building an upper and lower point cloud and corresponding delaunay mesh for each part, calculating the mean distance in the same manner as before (aka ray tracing), and finally combining the results.

### 3 Mean cartilage thickness using ray tracing from a central point

This is a three-dimensional approach using ray tracing along normal vectors to determine the mean thickness of a cartilage volume. This is another proposed solution to the previously discussed issue with the shape of the femoral cartilage volume. Instead of using meshes, this approach takes a central point underneath or above the cartilage and utilizes ray tracing from that point against the volume to discover intersection points. The central point is determined by calculating the halfway points between minimum and maximum  $x$ ,  $y$  and  $z$  coordinates of

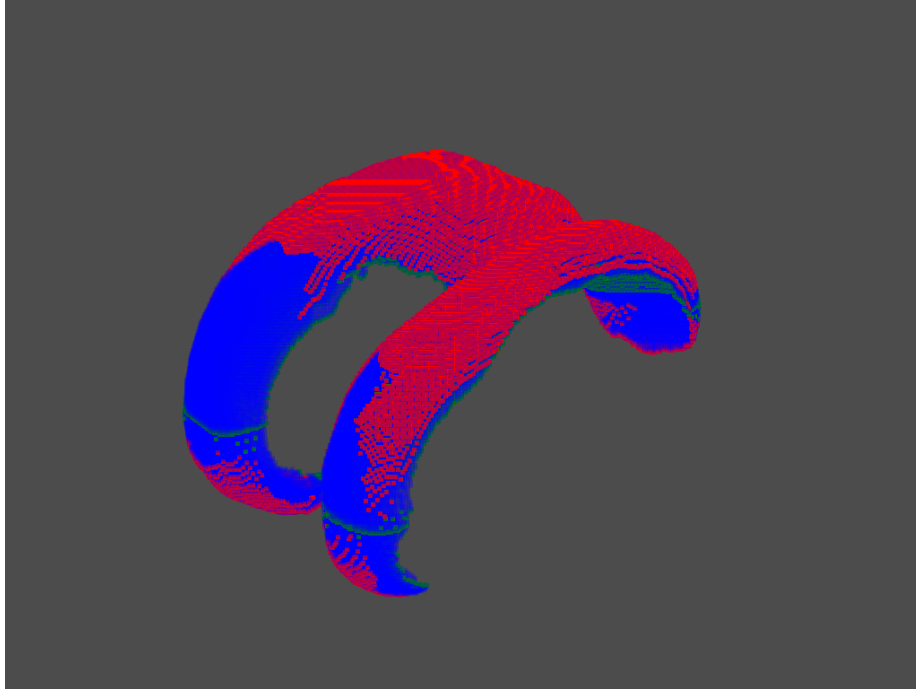


Figure 2: Point clouds of the femoral cartilage

the cartilage volume, respectively.

A sphere is constructed around the central point, and each of its normal vectors gets extended until it hits the cartilage point cloud, or a maximum number of iterations is reached. If a point is hit, it is saved and the vector again gets extended until it doesn't hit any points anymore, i.e. it is extended past beyond the cartilage. The distance between the first and last point hit is calculated and added to the result set. This way, it is possible to determine the average thickness of the cartilage. One issue with this approach is that it is computationally expensive, as intersection problems tend to be; in essence, there is a trade-off between accuracy and runtime: The more vectors are used for the ray tracing, the more accurate the result is going to be, but each added vector makes the computation more expensive. The resolution of the sphere was set to  $30 \times 30$ , resulting in  $30 + 29 \cdot 28 = 842$  normal vectors. No optimization analysis has been performed as of yet.

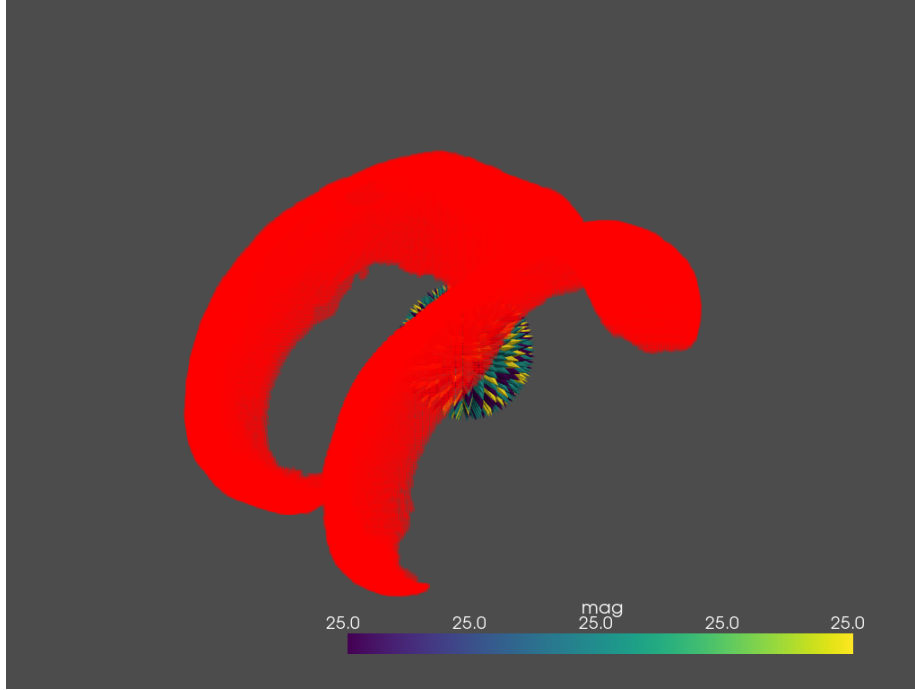


Figure 3: Ray tracing against the femoral cartilage using a central sphere

## 4 Mean cartilage thickness using two-dimensional function fitting

### 4.1 Determine thickness via function normals

This is a two-dimensional approach using a least-squares fit of single cartilage layers to determine the mean thickness of a cartilage volume. As the MRI scans consist of a number of slice exposures, it is possible to do the thickness calculation layer by layer. For each layer, a polynomial function gets fitted over the data points, i.e. through the middle of the volume, and a variable number of normals is calculated along the fitted function. For each normal, the inner- and outermost intersection points with the cartilage are determined, and the distance between these two points is added to the result set, as illustrated in figure 4.

### 4.2 Determine thickness via integration/function values

This is another two-dimensional variant, where two functions are fitted over the data points, i.e. not through the middle of the volume but rather along the outlines of the volume, as illustrated in figure 5. The distance between the

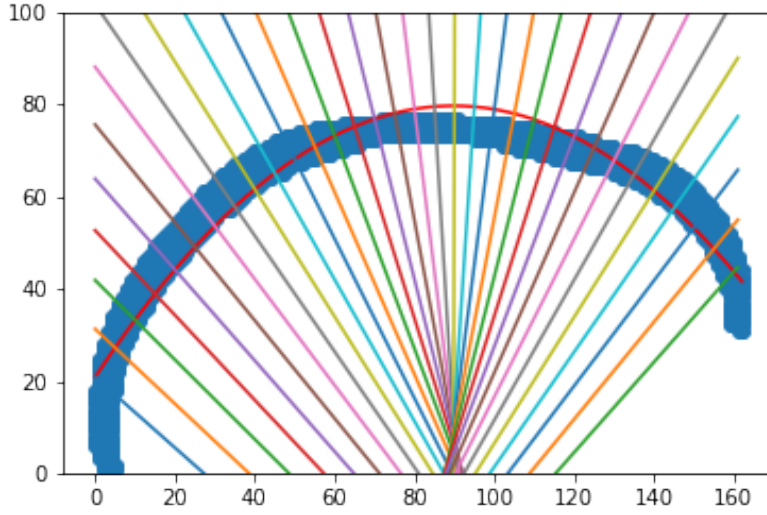


Figure 4: Normals along a two-dimensional least squares fit

functions, i.e. the thickness of the cartilage, can then easily be calculated, for example by integrating both functions and taking the difference ( $\int f(x)dx - \int g(x)dx$ ), or calculating the difference between the function values for every  $x$  value ( $[\sum_{x_i=0}^{max(x)} f(x_i) - g(x_i)] \cdot \frac{1}{max(x)}$ ).

One issue with both of these approaches is that the function fitting may not work well for certain shapes, especially for layers where the number of data points is very sparse. The degree for the polynomial fit was chosen as four. This is because in practice, no layer is going to have more than two inflection points (layers of the femoral cartilage can generally be approximated by a second-order polynomial, while the layers of the tibial cartilage volume follow a third-order form.) and choosing a higher degree than needed is undesirable (overfitting, Runge's phenomenon, etc). While in theory, a degree of two and three, respectively, would be sufficient, it has been set to four to accommodate for edge cases where a lower-order fit might yield poor or no results, e.g. for the layers on the edges of the volume, and the higher flexibility that comes with a higher degree is beneficial.

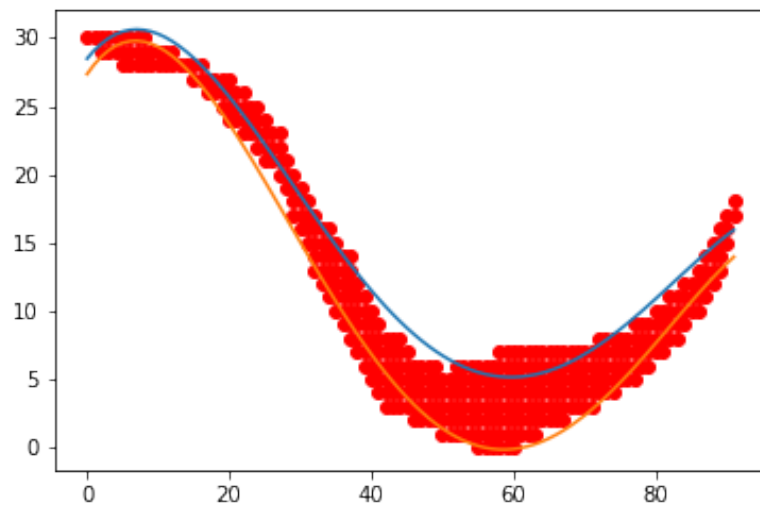


Figure 5: Two functions fit over the outlines of a tibial cartilage slice exposure

## A Determination of the cartilage subregions

As proposed by Wirth & Eckstein, the two cartilage plates of the medial / lateral femoral condyle are divided into three subregions each (central, internal and external), while the two cartilage plates of the medial / lateral tibia are divided into five subregions each (central, internal, external, anterior, posterior). For nomenclature, refer to the table of acronyms below.

### A.1 Determination of the femoral subregions

For each plate of the femur, each subregion should encompass approximately 33% of the plate volume. This was achieved in practice by splitting the volume into three parts along the x-axis; two splitting points  $x_1, x_2$  are chosen arbitrarily, such that  $x_{min} < x_1 < x_2 < x_{max}$ ,  $x_1 = \frac{x_{max} - x_{min}}{3}$ ,  $x_2 = 2 \cdot x_1$ , where  $x_{max}, x_{min}$  are the maximum / minimum x values, respectively.  $x_1$  and  $x_2$  are then iteratively moved along the x-axis until the constraint is satisfied, i.e. 33% of all data points lie left of  $x_1$ , 33% lie between  $x_1$  and  $x_2$  and 33% lie right of  $x_2$ .

### A.2 Determination of the tibial subregions

For each plate, an ellipse around the center of the plate, aka the central region, should encompass approximately 20% of the plate volume, and four triangles surrounding the ellipse should be of variable size. This was achieved in practice by first determining the center of gravity of the plate through K-Means clustering and constructing an ellipse around this point; a radius  $r$  is chosen arbitrarily, and is lengthened iteratively until the constraint is satisfied, i.e. 20% of all data points lie within the ellipse. The four triangles surrounding the ellipse are determined by calculating the corners  $a, b, c, d$  of the plate, and data points are assigned to subregions according to their position relative to the vectors  $\vec{ac}$  and  $\vec{db}$ , which is calculated via cross product.

```
1      Procedure to determine femoral subregions
2      Given:
3      x_axis := range of x-axis
4      plate := data points (x, y, z) making up a cartilage
           ↪ plate
5
6      Procedure:
7      x_min, x_max := min(x_axis), max(x_axis)
8      x_range := x_max - x_min
9      x1 := x_range / 3
10     x2 := 2 * x1
11     first_third := empty set
12     second_third := empty set
13
```



```

14 while len(first_third) / len(plate) is not .33 do
15     first_third := {d ∈ plate | d.x < x1}
16     if len(first_third) > .33
17         x1 := x1 - 1
18     else
19         x1 := x1 + 1
20
21 while len(second_plate) / len(plate) is not .33 do
22     second_plate := {d ∈ plate | x1 < d.x < x2}
23     if len(second_plate) > .33
24         x2 := x2 - 1
25     else
26         x2 := x2 + 1

```

1 Procedure to assign a femoral point to a subregion  
2 Given:  
3 plate := data points (x, y, z) making up a cartilage  
4     ↪ plate  
5 x1, x2 := split points  
6  
7 for point in plate do  
8 if point.x < x1  
9 point.region = external/internal #  
10 ↪ depending on whether point lies in  
11 ↪ left or right plate  
12 if x1 < point.x < x2  
13 point.region = central  
14 else  
15 point.region = external/internal #  
16 ↪ depending on whether point lies in  
17 ↪ left or right plate

1 Procedure to determine tibial subregions  
2 Given:  
3 plate := data points (x, y, z) making up a cartilage  
4     ↪ plate  
5  
6 Procedure:  
7 r := 20  
8 c := KMeans(plate)  
9 points\_in\_ellipse := empty set  
10  
11 while len(points\_in\_ellipse) / len(plate) is not .2 do  
12 points\_in\_ellipse := {d ∈ plate | dist(d, c) < r}  
13 if len(points\_in\_ellipse) > .2  
14 r = r / 2

```

14         else
15             r = r + .5
16
17         x_min := {min(d.x) | d ∈ plate}
18         x_max := {max(d.x) | d ∈ plate}
19         y_min := {min(d.y) | d ∈ plate}
20         y_max := {max(d.y) | d ∈ plate}
21
22         a := (x_min, y_min)
23         b := (x_max, y_min)
24         c := (x_max, y_max)
25         d := (x_min, y_max)

```

1 Procedure to assign a tibial point to a subregion  
2 Given:  
3 plate := data points (x, y, z) making up a cartilage  
4     ↪ plate  
5 a, b, c, d := plate corners  
6 points\_in\_ellipse := set of points lying within the  
7     ↪ central ellipse

```

6
7 Procedure:
8 for point in plate do
9     if point is in points_in_ellipse
10         point.region := central
11
12         ac := c - a
13         db := b - d
14         pc := c - point
15         pb := b - point
16
17         if ac × pc > 0
18             if db × pc > 0
19                 point.region := internal
20             else
21                 point.region := posterior
22         else
23             if db × pc > 0
24                 point.region := anterior
25             else
26                 point.region := external

```

## B Splitting of femoral cartilage volume for meshes

As described above, constructing meshes for the femoral cartilage is not possible without some pre-processing. The convex parts of the volume on either side need to be split off and rotated by  $90^\circ$  so that minimum and maximum points can reliably be extracted without missing a significant amount. The sections are isolated as follows:

The cartilage volume is a point cloud made up of vectors  $(x, y, z)$ . Group the points by  $(x, y)$  and for each of these combinations, calculate the range of  $z$ . Then calculate the mean or median of  $z$  and take all combinations  $(x, y)$  for which the range of  $z$  is less than the previously calculated mean/median, i.e. those points where the minimum and maximum  $z$  values are not too far apart. Note that the convex parts of the cartilage are characterized by a large ranges for  $z$ . Having now isolated the central part of the volume, it is a trivial task to extract the convex parts: We can calculate the minimum and maximum  $x$  value of the central vectors, and everything left of the minimum gets assigned to the left convex part, and everything right of the maximum gets assigned to the right convex part.

With the sections isolated, the problematic convex parts can then be rotated by  $90^\circ$ . Now, again for every combination  $(x, y)$ , the vector  $(x, y, \max(z))$  can be added to the upper mesh, and the vector  $(x, y, \min(z))$  to the lower mesh ( $\min(z)/\max(z)$  may be replaced by distance to a central point, depending on implementation).

```
1      Procedure to split a femoral cartilage volume into three
      ↪ parts
2      Given:
3      volume := vectors (x, y, z) making up the volume
4
5      Procedure:
6      z_range := volume.group_by((x, y)).z.max() -
      ↪ volume.group_by((x, y)).z.min()
7      z_med := median(z_range)
8      z_index := {d ∈ volume | d.z < z_med}
9      lower_bound := {min(d.x) | d ∈ z_index}
10     upper_bound := {max(d.x) | d ∈ z_index}
11     left_part := {d ∈ volume | d.x < lower_bound}
12     middle_part := {d ∈ volume | lower_bound < d.x <
      ↪ upper_bound}
13     right_part := {d ∈ volume | d.x > upper_bound}
14     left_part := rotate(90)
15     right_part := rotate(90)
```

## C Acronyms

- ecLF** Mean thickness of the external cartilage subregion of the central part of the lateral femoral condyle
- ccLF** Mean thickness of the central cartilage subregion of the central part of the lateral femoral condyle
- icLF** Mean thickness of the internal cartilage subregion of the central part of the lateral femoral condyle
- icMF** Mean thickness of the internal cartilage subregion of the central part of the medial femoral condyle
- ccMF** Mean thickness of the central cartilage subregion of the central part of the medial femoral condyle
- ecMF** Mean thickness of the external cartilage subregion of the central part of the medial femoral condyle
- cLT** Mean thickness of the central cartilage subregion of the lateral tibia
- aLT** Mean thickness of the anterior cartilage subregion of the lateral tibia
- eLT** Mean thickness of the external cartilage subregion of the lateral tibia
- pLT** Mean thickness of the posterior cartilage subregion of the lateral tibia
- iLT** Mean thickness of the internal cartilage subregion of the lateral tibia
- cMT** Mean thickness of the central cartilage subregion of the medial tibia
- aMT** Mean thickness of the anterior cartilage subregion of the medial tibia
- eMT** Mean thickness of the external cartilage subregion of the medial tibia
- pMT** Mean thickness of the posterior cartilage subregion of the medial tibia
- iMT** Mean thickness of the internal cartilage subregion of the medial tibia
- x.aSD** Standard deviation of the thickness of the respective (x) cartilage subregion
- x.aMav** Mean value of the maximum 1% measurements of cartilage thickness of the respective (x) cartilage subregion
- x.aMiv** Mean value of the minimum 1% measurements of cartilage thickness of the respective (x) cartilage subregion

## References

- [WE08] Wolfgang Wirth and Felix Eckstein. A technique for regional analysis of femorotibial cartilage thickness based on quantitative magnetic resonance imaging. *IEEE transactions on medical imaging*, 27(6):737–744, 2008.