

# Article Title

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## Abstract

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## I. INTRODUCTION

Patellofemoral pain syndrome (PFPS) is a painful musculoskeletal condition that is presented as pain behind or around patella [1, 2]. PFPS affects 6-7 % of adolescents, of whom two thirds are highly physically active [3]. Additionally the prevalence is more than twice as high for females than males [4, 3]. PFPS may be present over longer periods of time where a high number of individuals experience a recurrent or chronic pain [5] and may also lead to osteoarthritis [4, 6].

Patellofemoral pain (PFP) is often described as diffuse knee pain, that can be hard to explain and localize [5]. Despite the fact that individuals feel pain in the knee, there is not any structural changes in the knee such as significant chondral damage or increased Q-angle. There is no definitive clinical test to diagnose PFPS and it is thereby often diagnosed based on exclusion criterias [4] to which PFPS is also described as an orthopaedic enigma, and is one of the most challenging pathologies to manage [7]. To assist diagnosis of PFPS, pain maps may be used as a helpful tool for the individuals to communicate their pain by drawing pain areas [8]. A study shows that through the use of

pain maps it is possible to find a correlation between the symptom duration and the size and morphology of pain area [9]. Another method to measure pain is by using visual analog scale (VAS), that scores pain between no-pain to the worst pain imaginable [10]. However it is a known problem that chronic pain is considered a multidimensional pain, because the perceived pain of an individual is influenced by biomedical, psychosocial and behavioral factors [? ].

Since PFP is associated with a lack of knowledge, and it has been shown that there is a correlation between pain maps and symptom duration as well as pain intensity, it is interesting to investigate if pain maps can be used to classify and thereby predict PFP related information.

A method that has not been found used in this context before is a deep learning. The deep learning method is chosen for this study because it is a state of the art method, that has shown greater performance in specific computation fields, compared to other machine learning methods [11]. Furthermore, the method is chosen because of its ability to find non-linear connections between input and output data [11], which is found relevant

for this study mainly based on the fact that PFP is subjective and may be affected by the multidimensionality of chronic pain.

The goals of this project is to explore how accurate a deep learning model can classify symptom duration and pain intensity associated to PFP pain maps using a limited dataset. Because the prevalence is more than twice as high for females than males, the gender is included as a feature in the deep learning model. Furthermore, morphology of the pain is considered to be relevant, based on the indication that morphology and size of pain area increase with prolonged symptom duration. To investigate the influence of morphology and location of the pain three types of pain map representations are created: a binary representation which reflect the morphology, a simplified representation of morphology based on knee regions that give information about the pain location, and a combined representation that contains the morphology divided into knee regions. The aim of this study is to explore classification performance of a deep learning model, using PFP pain maps and gender as input to classify either symptom duration or pain intensity.

*It is hypothesized that a deep learning model that uses pain maps and gender as input parameter has a higher performance when classifying according to symptom duration than pain intensity.*

The secondary aim is to investigate if multiple pain map representations, which reflect the morphology and location of the pain, affect the deep learning model classification performance.

*It is hypothesized that different data representations of pain maps, reflecting morphology and location of pain, affect the performance accuracy of a deep learning model when classifying according to symptom duration or pain intensity.*

## II. METHODS

### Data and manual data handling

Data used in this study were collected beforehand from an on-going FOXH trial which is

conducted in collaboration with Danish and Australian universities. The data consists of pain maps which were drawn by individuals with PFPs through the use of an application, Navigate Pain, in a clinical setting. The pain maps are both from individuals with uni- and bilateral PFP, an example of these are shown in figure 1.



**Figure 1:** Pain maps of the lower extremities from individuals with uni- and bilateral PFP. The red markings indicate the area of pain perceived by the individuals.

In addition to the pain maps appurtenant information was available, which contained information regarding the individuals. Before using the data in the deep learning models, a manual data handling was necessary. This incorporated matching the given pain maps and appertaining ID regarding the individuals, which resulted in 217 available pain maps. Furthermore, specific information like gender, symptom duration and pain intensity were collected from the appurtenant information. The number of pain maps with associated information, gender and symptom duration, was 205. Additionally, there were 197 pain maps with associated information, gender and pain intensity.

### Software application: Navigate Pain

Navigate Pain is a software application that

is used to visualise the location, shape and spatial distribution of pain from individuals to healthcare personnel. The application permits individuals to draw their pain into a body outline with different colors and line thickness. Navigate Pain android was developed at Aalborg University and a commercial web application is available at Aglance Solutions (Denmark).[12]

### Data representations

It is presumed that different representations of the pain maps affect the performance accuracy of a deep learning models, which is why different data representations are created. A study by Boudreau et al. found a correlation between a prolonged symptom duration and the size of the pain area. It was shown that the pain area increased for individuals that have a symptom duration for longer than five years compared to those with a symptom duration below. Likewise, pain intensity had a correlation with the size of pain area for individuals. Furthermore, the morphology of the pain developed from a U-shape to an O-shape for individuals with a symptom duration above five years.[9]

Based on this study the morphology of PFP is considered to be relevant to investigate, and therefore the morphology constitutes a data representation.

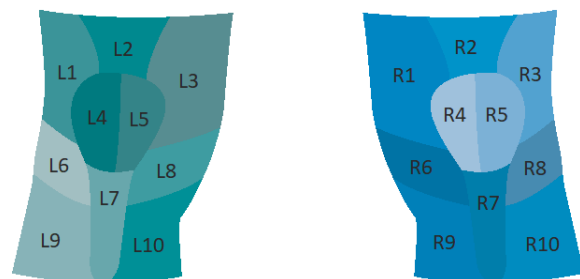
The PFP is often described as diffuse pain and therefore difficult to describe and localise [5]. To accommodate this it is chosen to divide the pain into different knee regions, which may indicate whether a specific region of the knee influence the PFP. This is converted to a simplified data representation that indicate active pain regions. A combination of the morphology and location of the pain constitutes third data representation. Furthermore, gender is an interesting parameter to use as an input, because the prevalence is more than twice as high for females than males. Thereto, perceived pain is subjective and depends on the individual's character and personality. The distribution of gender is investigated by creating a histogram, which showed that the prevalence is higher for females than males, given that females consti-

tute 156 of the 206 individuals.

It is chosen to classify the three data representation in proportion to symptom duration, based on the study by Boudreau et al. which indicated that symptom duration seems to affect the size and morphology of the pain area. Furthermore the study showed that there was a correlation between pain intensity and size of pain area, which is why it is chosen to classify pain maps according to pain intensity. The three data representations are referred to as morphology-, regions- and combined-representation.

### Knee regions

Patients with PFPs often describe the knee pain as a diffuse pain, and when looking at pain drawing samples from multiple patients it is also evident that there is a high variability in the distribution of pain patterns across different areas of the knee. To distinguish between different pain areas, the knee can be divided into various regions as seen in figure 2, where the division of the left and right anterior knees are illustrated.



**Figure 2:** The regions of the left and right knees, where each knee is split into ten regions.

The divisions is inspired by Photographic Knee Pain Map (PKPM) which is designed to categories location of knee pain, diagnostic and research purposes. PKPM represent both knees that makes it possible to identify unilateral and bilateral pain.[13]

The regions are based on the anatomic structures according to the areas where individuals often indicate pain. There is ten regions, where region 1 and 3 represent the superior lateral and superior medial areas for patella.

Region 2 refers to quadriceps tendon. The patella is divided into lateral and medial regions, which are region 4 and 5. Region 6 and 8 are lateral and medial joint line areas. Patella tendon is region 7 and the two last regions, 9 and 10, are tibia lateral and medial.[13]

In relation to the regions-representation, it is necessary to find a threshold that decides when a knee region contains enough pain pixels to be considered active. A threshold is required to increase the confidence of an active pain region by avoiding minimal contributions e.g. small pain areas in the associated regions. Simultaneously the threshold may not be too large so that potential pain regions will not be incorporated. The threshold to indicate active pain regions is decided based on an analysis, where threshold values of 0, 5, 10 and 15 percent are tested. The analysis of the threshold is tested on five random pain maps to get a general impression of the data. Based on analysis of the five pain maps is 5 percent chosen as the threshold, that defines when a pain region is active.

#### Pre-processing of data

The data is pre-processed in MatLab to prepare it to the three different deep learning models. Each model has an appurtenant data representation which are prepared in three different ways. In general are the pain maps resized, since it was collected at different resolutions (screen sizes) and cropped to sort out unnecessary data like the areas inferior and superior to the knee. Each data representation is reflected in a matrix consisting of the pain maps, gender and the output, symptom duration and pain intensity. The morphology-representation is not further processed before using it in the models. The regions-representation is a simplified representation of morphology of the pain and reflects only the active pain regions. Thereto is a threshold used to define when a region is active. The third data representation is a combination of morphology- and regions-representations, which means that the matrix reflects the morphology of the pain in each

region.

### III. RESULTS

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### IV. DISCUSSION

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### V. CONCLUSION

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