Final Report

A study into the feasibility of using machine learning techniques to process MRI scans of Cavalier king charles spaniels skulls and detect signs of chiari-like malformation.

Robert Clark

2019

# Abstract

Contents

[Abstract 1](#_Toc38650291)

[Chapter 1: Introduction 3](#_Toc38650292)

[1.1. Canine Chiari-Like Malformation 3](#_Toc38650293)

[1.2. Current Approaches to Diagnosis and Treatment of Canine Chiari-Like Malformation 3](#_Toc38650294)

[1.3. Magnetic Resonance Imagery (MRI) 4](#_Toc38650295)

[1.4. Aims and Objectives 5](#_Toc38650296)

[Chapter 2: Literature Review 6](#_Toc38650297)

[2.1. Prior Analysis Work on Canine Chiari-Like Malformation 6](#_Toc38650298)

[2.2. Machine Learning as a Diagnosis Aid for CLM 7](#_Toc38650299)

[2.3. Machine Learning as a Diagnosis Aid for Other Neurological Conditions 7](#_Toc38650300)

[2.4. Summary 7](#_Toc38650301)

[Chapter 3: Preliminary Investigation 9](#_Toc38650302)

[3.1. Data Description 9](#_Toc38650303)

[3.2. Data Processing 9](#_Toc38650304)

[3.2.1. ImageJ 9](#_Toc38650305)

[3.2.2. MatLab 9](#_Toc38650306)

[Chapter 4: Conclusions and Future Plans 10](#_Toc38650307)

[4.1. Conclusions 10](#_Toc38650308)

[4.2. Overview of Future Development 10](#_Toc38650309)

[4.3. Project Plan 10](#_Toc38650310)

[4.4. Final Report Plan 10](#_Toc38650311)

[Chapter 5: Summary 11](#_Toc38650312)

[Appendix 12](#_Toc38650313)

[References 13](#_Toc38650314)

[Table of Figures 17](#_Toc38650315)

# Chapter 1: Introduction

## Canine Chiari-Like Malformation

Syringomyelia (SM) is a term used by the medical community to refer to the formation of cavities or cysts known as “syrinxes” within the spinal cord, resulting in discomfort, paralysis and loss of sensation throughout the body [1]. The most common cause of Syringomyelia is Chiari-Like Malformation (CLM), which is typically characterised as an incongruity between the size of the brain and the size of the skull [2] and is believed to be present in up to 95% of the world wide population of Cavalier King Charles Spaniels (CKCS) [3].

The discrepancies between the shape of the skull and brain in patients suffering from Canine CLM cause an obstruction for cerebrospinal fluid and tissue compression within craniocervical junction where the skull meets the spinal cord [4]. It is believed the resulting irregular flow of fluid then results in a mismatch of timing between arterial blood flow and cerebrospinal fluid flow. The perivascular space (shown in Figure 2) widens during the lull of the cardiac cycle, resulting in the cerebrospinal fluid essentially “leaking” through while its own pressure is high. This could result in cerebrospinal fluid entering the central canal of the spinal cord, and the eventual formation of syrinxes as shown in Figure 1.

Figure - A cyst forming within a cervical spine, characteristic of Syringomyelia. [41]

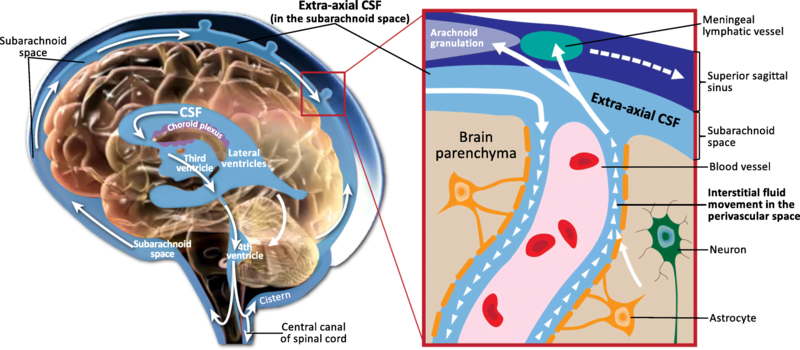


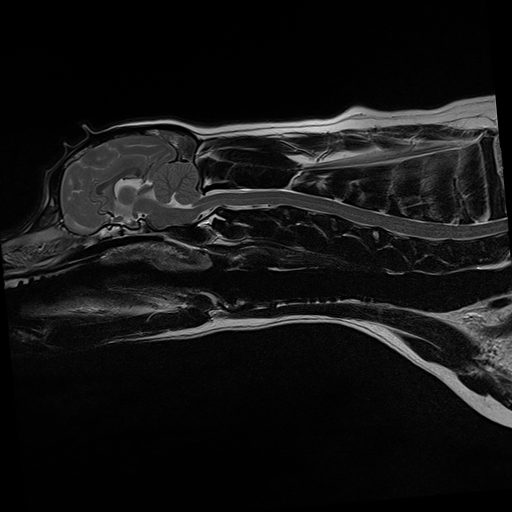
Figure - The flow of cerebrospinal fluid around the brain, with the perivascular space shown within the inset red box. [5] Licensed under Creative Commons Attribution 4.0 International.

In addition to Cavalier King Charles Spaniels, both Syringomyelia and Chiari-Like Malformation are known to present in humans. [6] Canine Chiari-Like Malformation is sufficiently analogous to it’s human counterpart for diagnostic and surgical techniques used to treat one to be effective on the other, resulting in extensive veterinary research being performed upon the Cavalier King Charles Spaniel breed as well as similarly affected dogs.

## Current Approaches to Diagnosis and Treatment of Canine Chiari-Like Malformation

Prior to the advent of Magnetic Resonance Imaging (MRI), only the behavioural signs of Chiari-like Malformation could be used in diagnosis. These would include disruption to motor skills, altered emotional state due to pain and excessive head rubbing and scratching and as such the condition was often misdiagnosed as epilepsy or an allergic reaction. [6]

After Chiari-like Malformation was observed in humans, the increased understanding of the condition allowed for the condition to be more easily diagnosed within animals. [6] Syringomyelia can be diagnosed via imaging of the spinal cord, where syrinxes will appear as anomalous regions along the central canal as shown in Figure 3.

A pair of black shoes

Description automatically generated

Figure - A central slice of an MRI scan depicting a healthy Cavalier King Charles Spaniel (left) and one affected by both CM and Syringomyelia (right), with the syrinx indicated by the red rectangle.

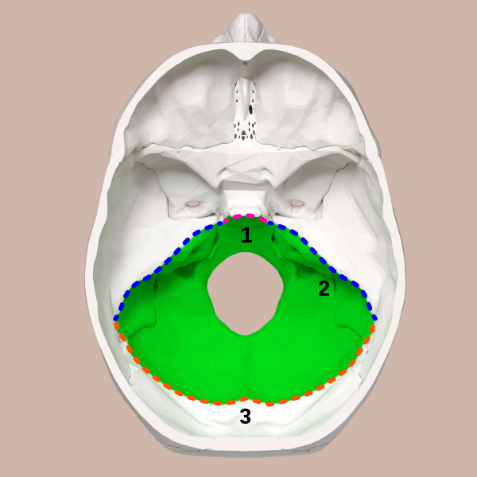
However, Chiari-like Malformation occurs independently of Syringomyelia and so syrinxes may not form until a considerable amount of time after the former condition develops. Diagnosis of Chiari-like Malformation alone is hence often done via the ruling out of other conditions and study of anatomical characteristics of the patient.

Figure - The posterior fossa of the human skull, shown in green. Licensed under the Creative Commons Attribution-Share Alike 2.1 Japan. [42]

Once Syringomyelia secondary to Chiari-like Malformation has been diagnosed, the treatment usually depends on the age of the patient. Younger canines typically have a higher recovery rate after surgery, and so removal of the syrinx is often recommended to prevent further development as the dog ages. Older dogs with fewer clinical symptoms are instead treated medically with opioids or antiepileptics to limit neuropathic pain. [7]

To limit production of further syrinxes and treat Chiari-like malformation directly, the most common treatment is Foramen Magnum Decompression (FMD). This involves enlarging the posterior fossa (Figure 4), the region of the skull which houses the brain stem and cerebellum, to alleviate pressure on the craniocervical junction. [8] This approach has an approximately 80% success rate, but has a 25% to 50% chance of relapse due to scar tissue formation at the decompression site.

## Magnetic Resonance Imagery (MRI)

The dataset used for this project is composed of MRI scans of Cavalier King Charles Spaniel skulls, as shown in Figure 3. To obtain these images, the subjects were athetized to reduce movement and placed within a strong oscillating magnetic field. This aligns the positively charged protons within the water molecules of the subject. Targeted bursts of radio waves were then fired at the subjects in A picture containing device

Description automatically generatedorder to “knock” the water molecules out of alignment. As they realign, they emit radio signals of their own which are detected by the receiving coil with different tissues aligning at different speeds producing varying signals. The three gradient coils within the MRI machine allow the oscillating magnetic field to be moved within the uniform field generated by the primary coil, allowing for data to be obtained for a number of different perspectives which can be combined to obtain an overall image.

Figure - A cross section of an MRI scanner with labelled components. [43] Licensed under the Creative Commons Attribution-ShareAlike 3.0.

The “frequency content” of the signal emitted as the hydrogen nuclei realign can then be analysed through use of Fourier Transforms, and the different tissues associated with the frequencies identified alongside their spatial location. In order to image the brain for this project, T2-weighted imaging was used in order to produce a higher contrast between the soft tissues of the brain and the hard tissue of the bone which will appear as light and dark regions respectively. This involved using comparatively long intervals between repeated radio bursts in order to allow magnetization to decay. [9]

## Aims and Objectives

As explained in Section 1.2, early diagnosis of chiari-like malformation can allow for surgical intervention while the Cavalier King Charles Spaniel is still young enough to have a high rate of recovery and before Syringomyelia can develop and require addition surgery to remove syrinxes. Currently, this is difficult due to the ambiguity as to what directly qualifies as chiari-like malformation and the process of diagnosis often requiring specialist knowledge which may not be available to all patients.

If accurate diagnosis of the condition through machine learning is possible, access to an MRI machine would be the only obstacle to diagnosis of the condition. This would allow for the condition to be identified earlier, before it can develop into Syringomyelia or intrusive surgery becomes too damaging for the patient and so result in a significant quality of life improvement for dogs presenting with the disease. Evidence also suggests that chiari-like malformation is a hereditary condition [10], meaning that dogs which suffer from the condition but which do not present many clinical conditions could be identified before they produce less fortunate descendants and hence provide better breeding values for healthy dogs.

The similarity of canine chiari-like malformation to its human equivalent could also potentially mean that any machine learning based approach to diagnosis could be used on humans as well, hence reducing the time needed to diagnose the condition and expedite pain alleviating surgery.

# Chapter 2: Literature Review

## 2.1. Prior Analysis Work on Canine Chiari-Like Malformation

A picture containing text, map

Description automatically generatedThough the definition of chiari-like malformation as a product of an underdeveloped skull compressing an overdeveloped brain is well understood, the exact dimensions needed to produce clinical symptoms are still debated. Cerebellar compression as a result of overlapping of the atlas and occipital bones (Figure 6) so that the atlas bone is partly within the cranium has a stronger incidence rate of CLM than other causes [11], meaning that the nature of the occipital bone may be one of the deciding factors of the condition. However, atlantooccipital overlapping also resulted in more severe incidences of cerebellar compression and so the correlation may instead be to the that rather than any particular cause.

Figure - The Atlantooccipital joint and atlantoaxial ligament, located at the base of the skull. [44]

Alternative studies have instead linked CLM developing into Syringomyelia to lesions on the atlantoaxial ligament known as “bands” [12] or “medullary kinking” of the craniocervical junction [13] where the spinal cord meets the brain stem at a non-continuous angle as illustrated in Figure 7.

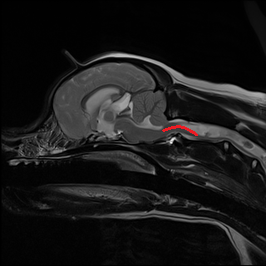
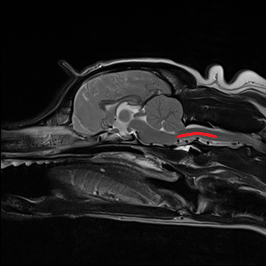


Figure - A healthy CKCS (left) with a continuous craniocervical junction and a CLM/SM affected CKCS (right) with a noticeably elevated caudal medulla oblongata. Both junctions are highlighted with a red line.

These investigations are hampered by the sheer prevalence of chiari-like malformation, with control groups being hard to establish when the condition simply could not have presented itself yet.

## 2.2. Machine Learning as a Diagnosis Aid for CLM

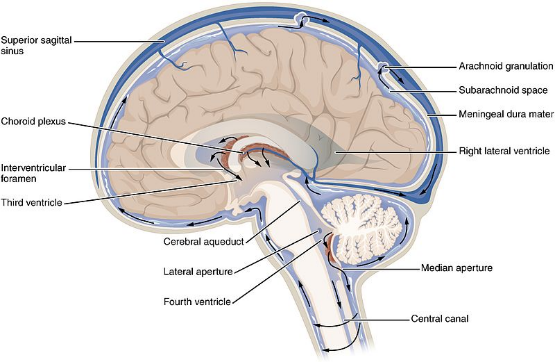
Some investigation into machine learnings use diagnosing chiari-like malformation has already been performed. The floor of the third ventricle (shown and a region in the sphenoid bone as potential biomarkers for CLM by locating morphological differences between MRI scans of affected and healthy dogs. [14] By quantifying these morphological differences and using them as features within a Support Vector Machine (SVM), researchers were able to produce a binary classification system with an Area Under the Curve (AUC) of 77.77. This is sufficient to lend credence to these biomarkers being related to CLM and eventual pain but lacks the reliability necessary for consistent use within medicine.

Figure - A labelled cross section of a brain with ventricles exposed. [45] Licensed under the Creative Commons Attribution 4.0 International.

Additional supervised learning approaches have also been tried on MRI scans of human chiari malformation patients, producing mixed results. Distance from the Foranem Magnum (FM) to the peak of the fourth ventricle, distance from the FM to the brain stem and the angle of the brain stem were identified as key characteristics through use of SVMs and produced sensitivity and specificity rates of above 90% when testing for type 1 Chiari Malformation [15]. This condition is analogous to CLM in canines, but structural differences between human and canine brains such as the enlarged cerebellum may decrease the importance of these specific measurements in the diagnosis of CLM.

## 2.3. Machine Learning as a Diagnosis Aid for Other Neurological Conditions

In addition to research into CLM, machine learning has also proved to be invaluable in the study of other neurological conditions. By obtaining MRI data from Alzheimer’s patients and providing it to a Convolutional Neural Network (CNN), researchers were able to create a binary classifier capable of diagnosing the condition with an average accuracy 96.8588% [16]. The dataset used here consisted of a “stack” of images obtained from each patient, with non-brain tissue removed from each layer and the eight outermost layers discarded and was able to obtain such a high accuracy rate through use of the LeNet CNN.

SVMs have also been used in the study of Parkinson’s Disease, with a dataset of MRI scans of Parkinson’s afflicted brains reduced via Principle Component Analysis (PCA) to identify regions within the 3D space which expressed significant variation and then treated these as features. This resulted in an overall accuracy of 92.7% when classifying Parkinson’s afflicted and healthy brains [17], a significant improvement over the 39.53% obtained by a similar study performed without dimensionality reduction via PCA [18]. This could potentially be attributed to differences in the dataset, with the former study having an approximately 27% larger dataset which may have allowed for more generalisations but should still be considered a testament to the potential use of PCA within this investigation.

## 2.4. Summary

Though research into this area has already been performed to some success, they have typically been performed using larger datasets and required a greater deal of pre-processing before classification. Regions of interest within the brain have been identified through both machine learning and clinical studies, but there does not appear to one definitive agreed cause and so research performed here will have to pursue multiple avenues.

Machine learning approaches to diagnosis for other neurological conditions, such as transfer learning and SVMs have yielded significant results. Reproducing these results within this investigation however may not be feasible; these studies have mostly been results driven and have been performed on conditions with clearly understood origins. As such, identification has been easier than it will be for this hypothesis driven study.

# Chapter 3: Methodology

## 3.1. Data Description

The dataset used for this project originated from an agreement between the University of Surrey and Fitzpatrick Referrals, and consisted of sagittal T2-weighted MRI scans of 19 anaesthetised CKCS. Of those 19, 14 had been previously diagnosed as CLM afflicted by a trained veterinarian, leaving a control group of 5.

Of the 14 affected by CLM, 11 also presented with syringomyelia and this will be reflected within the MRI scans.

|  |  |  |
| --- | --- | --- |
| Group | Affected | Control |
| Sex (M:F) | 5:9 | 5:0 |
| Age Range (Years) | 0.92 - 7 | 2 - 8 |
| Age Median (Years) | 4 | 7 |
| Weight Range (kg) | 6.4 – 12.5 | 9.2 - 16.15 |
| Weight Median (kg) | 8.6 | 11.95 |

Table - Summary of patient data within dataset

## 3.2. Data Processing

### 3.2.1. ImageJ

### 3.2.2. MatLab

Chapter 4: Conclusions and Future Plans

## Conclusions

## Overview of Future Development

## Project Plan

## Final Report Plan

Due to similarities in structure between this report and the final report due at the conclusion of this project, topics discussed in Chapters 1, 2 and 3 of this report may be reiterated there. In addition to discussing the preliminary analysis, an overview will also be given of the work performed with the larger dataset as explained in Section 4.1.

The results obtained over the course of the project will then be examined and the effectiveness of a CNN as method of identifying the different types of canine elbow dysplasia evaluated. From this, the potential for future researchers to further investigate the role of joint incongruity in elbow dysplasia using machine learning can then be assessed and recommendations made.

It is expected the following chapter headings will be used:

* Title page
* Abstract
* Chapter 1: Introduction
* Chapter 2: Literate Review
* Chapter 3: Methodology
* Chapter 4: Results and Discussion
* Chapter 5: Conclusion
* Chapter 6: Potential for Future Development
* References and Table of Figures
* Appendices

# Chapter 5: Summary

This report explains intentions to use machine learning techniques to evaluate the role of joint incongruency in canine elbow dysplasia, and to identify specific connections between the shape of the joint and the disease.

An overview of canine dysplasia is then given, including causes, diagnosis and treatment, followed by a brief explanation of how the CT scans used within the dataset are obtained. A literature review then explains the successes of previous attempts to use machine learning techniques within medicine, focusing primarily on orthopaedic medicine. Support is also given for the hypothesis that elbow dysplasia is related to joint incongruity, with the research gap which we will attempt to fill identified.

From a limited dataset of CT scans of Labrador joints, binary masks depicting the hard-exterior cortical tissue of the joint have been produced which can now be used to create a three-dimensional model depicting the joint. Aims of using these binary masks and three-dimensional models to produce deformation patterns describing the difference between healthy and diseased joints are then given, with plans to train a CNN to recognise key areas responsible for elbow dysplasia also given. A detailed project plan in the form of a Gantt chart is provided.

Preliminary research performed in preparation for the arrival of the larger dataset is nearing completion, with more in depth work soon ready to begin.

# Appendix



# References

|  |  |
| --- | --- |
| [1] | E. L. Hilton and L. J. Henderson, “Neurosurgical considerations in posttraumatic syringomyelia,” *AORN Journal,* vol. 77, no. 1, pp. 135-150, 2003. |
| [2] | “Use of Morphometric Mapping to Characterise Symptomatic Chiari-Like Malformation, Secondary Syringomyelia and Associated Brachycephaly in the Cavalier King Charles Spaniel,” *PLoS One,* vol. 12, no. 1, 2017. |
| [3] | C. A. Loughin, “Chiari-like Malformation,” *Veterinary Clinics of North America: Small Animal Practice,* vol. 46, no. 2, pp. 231-242, 2016. |
| [4] | C. Rusbridge, “Chiari–like malformation and syringomyelia,” *European Journal of Companion Animal Practice,* vol. 23, no. 3, p. 70, 2013. |
| [5] | M. D. Shen, “Wikimedia Commons,” 13 December 2018. [Online]. Available: https://commons.wikimedia.org/wiki/File:CSF\_circulation.png. [Accessed 20 April 2020]. |
| [6] | C. Rusbridge, F. Stringer and S. P. Knowler, “Clinical Application of Diagnostic Imaging of Chiari-Like Malformation and Syringomyelia,” *Front Vet Sci,* vol. 5, no. 280, 2018. |
| [7] | A. C. Hechler and S. A. Moore, “Understanding and Treating Chiari-like Malformation and Syringomyelia in Dogs,” *Topics in Companion Animal Medicine,* vol. 33, no. 1, pp. 1-11, 2018. |
| [8] | P. Salunke, M. Karthigeyan and P. Malik, “Foramen magnum decompression without bone removal: C1–C2 posterior fixation for Chiari with congenital atlantoaxial dislocation/basilar invagination,” Department of Neurosurgery, Postgraduate Institute of Medical Education & Research (PGIMER), Chandigarh, 2019. |
| [9] | D. W. McRobbie, E. A. Moore, M. J. Graves and M. R. Prince, MRI from Picture to Proton, Cambridge: Cambridge University Press, 2006, pp. 33-34. |
| [10] | P. Lemay, S. P. Knowler, S. Bouasker, Y. Nédélec, S. Platt, C. Freeman, G. Child, L. B. Barreiro, R. G. A., C. Rusbridge and Z. Kibar, “Quantitative Trait Loci (QTL) Study Identifies Novel Genomic Regions Associated to Chiari-Like Malformation in Griffon Bruxellois Dogs,” *PLoS One,* vol. 9, no. 4, 2014. |
| [11] | J. J. Sackman, D. J. Marino, C. A. Loughin, C. W. Dewey, L. J. Marino, M. L. Lesser and M. B. Akerman, “Morphometric features of the craniocervical junction region in dogs with suspected Chiari-like malformation determined by combined use of magnetic resonance imaging and computed tomography,” *American Journal of Veterinary Research,* vol. 73, no. 1, pp. 105-111, 2012. |
| [12] | S. Cerda‐Gonzalez, N. J. Olby and E. H. Griffith, “Dorsal Compressive Atlantoaxial Bands and the Craniocervical Junction Syndrome: Association with Clinical Signs and Syringomyelia in Mature Cavalier King Charles Spaniels,” *Journal of Veterinary Internal Medicine,* vol. 29, no. 3, pp. 887-892, 2015. |
| [13] | S. Cerda‐Gonzalez, N. J. Olby and E. H. Griffith, “Medullary Position at the Craniocervical Junction in Mature Cavalier King Charles Spaniels: Relationship with Neurologic Signs and Syringomyelia,” *Journal of Veterinary Internal Medicine,* vol. 29, no. 3, p. 882–886, 2015. |
| [14] | S. Micheal, S. P. Knowler, C. Rusbridge and K. Wells, “Using machine learning to understand neuromorphological change and image‐based biomarker identification in Cavalier King Charles Spaniels with Chiari‐like malformation‐associated pain and syringomyelia,” *Journal of Veterinary Internal Medicine,* vol. 33, no. 6, pp. 2665-2674, 2019. |
| [15] | A. Urbizu, B. A. Martin and D. Moncho, “Machine learning applied to neuroimaging for diagnosis of adult classic Chiari malformation: role of the basion as a key morphometric indicator,” *Journal of Neurosurgery,* vol. 129, no. 3, pp. 567-851, 2018. |
| [16] | S. Sarraf and G. Tofighi, “Deep learning-based pipeline to recognize Alzheimer's disease using fMRI data,” in *2016 Future Technologies Conference (FTC)*, San Francisco, 2016. |
| [17] | C. Salvatore, A. Cerasa, I. Castiglioni, F. Gallivanone, A. Augimeri, M. M. López, G. Arabia, M. Morelli, M. C. Gilardi and A. Quattrone, “Machine learning on brain MRI data for differential diagnosis of Parkinson's disease and Progressive Supranuclear Palsy,” *Journal of Neuroscience Methods,* vol. 222, no. 1, pp. 230-237, 2014. |
| [18] | N. K. Focke, G. Helms, S. Scheewe, P. M. Pantel, C. G. Bachmann, J. Ebentheuer, P. Dechent, A. Mohr, W. Paulus and C. Trenkwalder, “Individual voxel‐based subtype prediction can differentiate progressive supranuclear palsy from idiopathic parkinson syndrome and healthy controls,” *Human Brain Mapping,* vol. 32, no. 11, pp. 1905-1915, 2011. |
| [19] | M. J. Pead and S. Guthrie, “Elbow dysplasia in dogs - a new scheme,” British Veterinary Association (BVA), London, 2011. |
| [20] | A & C Black, Black's Veterinary Dictionary, London: A & C Black, 2014. |
| [21] | J. Temwichitr, A. P. A. Leegwater and A. H. A. Hazewinkel, “Fragmented coronoid process in the dog: A heritable disease,” *The Veterinary Journal,* vol. 185, no. 2, pp. 123-129, 2010. |
| [22] | J. Michelsen, “Canine elbow dysplasia: Aetiopathogenesis and current treatment recommendations,” *The Veterinary Journal,* vol. 196, no. 1, pp. 12-19, 2013. |
| [23] | T. Gemmill and D. Clements, “Fragmented coronoid process in the dog: is there a role for incongruency?,” *Journal of Small Animal Practice,* no. 48, pp. 361-368, 2007. |
| [24] | J. D. Demko and R. D. D. McLaughlin, “Developmental Orthopedic Disease,” *Veterinary Clinics of North America: Small Animal Practice,* vol. 35, no. 5, pp. 1111-1135, 2005. |
| [25] | N. M. C. C. Fitzpatrick and R. M. V. Yeadon, “Working Algorithm for Treatment Decision Making for Developmental Disease of the Medial Compartment of the Elbow in Dogs,” *Veterinary Surgery,* vol. 38, no. 2, pp. 285-300, 2009. |
| [26] | E. Chandler, Ettinger Feldman Textbook of Veterinary Internal Medicine 4th Edn, London: W. B. Saunders, 1996. |
| [27] | H. Eljack, H. Werner and P. Bottcher, “Sensitivity and Specificity of 3D Models of the Radioulnar Joint Cup in Combination With a Sphere Fitted to the Ulnar Trochlear North for Estimation of Radioulnar Incongruence In Vitro,” *Veterinary Surgery,* vol. 42, pp. 365-370, 2013. |
| [28] | S. Alves-Pimenta, M. M. Ginja, A. M. Fernandes, A. J. Ferreira, P. Melo-Pinto and B. Colaco, “Computed Tomography and Radiographic Assessment of Congruity Between the Ulnar Trochlear Notch and Humeral Trochle in Large Breed Dogs,” *Veterinary and Comparative Orthopaedics and Traumatology,* vol. 1, no. 30, pp. 8-14, 2017. |
| [29] | J. M. M. Pérez and J. Pascau, Image processing with ImageJ : discover the incredible possibilities of ImageJ, from basic image procesing to macro and plugin development, Birmingham: Packt Publishing, 2013. |
| [30] | T. J. Gemmill, D. J. Mellor, D. N. Clements, S. P. Clarke, M. Farrell, D. Bennett and S. Carmichael, “Evaluation of elbow incongruencyusing reconstructed CT in dogssuffering fragmented coronoid process,” *Journal of Small Animal Practice,* vol. 46, pp. 327-333, 2005. |
| [31] | P. Proks, L. Stehlík, K. Irova, M. Dvořák, R. Srnec and A. Nečas, “Relationship between Radioulnar Incongruity of Elbow Joints and the Type ofFragmented Processus Coronoideus Medialis,” *Acta Veterinaria Brno,* vol. 79, pp. 307-312, 2010. |
| [32] | U. Gille, “Wikimedia Commons,” 13 June 2006. [Online]. Available: https://commons.wikimedia.org/wiki/File:Elbow-Dysplasia-Manifestations.svg. [Accessed 3 December 2019]. |
| [33] | S. M., “CT Scan,” in *Encyclopedia of Cancer*, Springer, Berlin, Heidelberg, 2011, p. 78. |
| [34] | T. Toyaski, “Wikicommons,” 2 September 2012. [Online]. Available: https://commons.wikimedia.org/wiki/File:CT\_PRINCI\_PB.jpg. [Accessed 3 December 2019]. |
| [35] | K. Jun and S. Yoon, “Alignment Solution for CT Image Reconstruction using Fixed Point and Virtual Rotation Axis,” *Sci Rep,* vol. 7, no. 41218, 2017. |
| [36] | I. Kononeko, “Machine learning for medical diagnosis: history, state of the art and perspective,” *Artificial Intelligence in Medicine,* vol. 23, no. 1, pp. 89-109, 2001. |
| [37] | C. Dezateux, “Developmental dysplasia of the hip,” *The Lancet,* vol. 369, no. 9572, pp. 1541-1552, 2007. |
| [38] | Y. Xue, R. Zhang, Y. Deng, K. Chen and T. Jiang, “A preliminary examination of the diagnostic value of deep learning in hip osteoarthritis,” *PLoS One,* 2017. |
| [39] | D. Maturana and S. Scherer, “VoxNet: A 3D Convolutional Neural Network for real-time object recognition,” in *2015 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*, Hamburg, IEEE, 2015, pp. 922-928. |
| [40] | J. Broder, Diagnostic Imaging for the Emergency Physician, Philadelphia: Saunders, 2011. |
| [41] | E. Linsey, “Wikimedia Commons,” 19 February 2007. [Online]. Available: https://commons.wikimedia.org/wiki/File:Syringomyelia.jpg. [Accessed 20 April 2020]. |
| [42] | BodyParts3D, “Wikimedia Commons,” 21 July 2019. [Online]. Available: https://commons.wikimedia.org/wiki/File:Posterior\_cranial\_fossa\_boundaries.svg. [Accessed 20 April 2020]. |
| [43] | Wikipedia, “Wikipedia,” 10 November 2017. [Online]. Available: https://en.wikipedia.org/wiki/File:Mri\_scanner\_schematic\_labelled.svg. [Accessed 21 April 2020]. |
| [44] | H. V. Carter and H. Gray, Anatomy of the Human Body, Philadelphia: Lea and Febiger, 1918. |
| [45] | OpenStax, “Wikimedia Commons,” 18 May 2016. [Online]. Available: https://commons.wikimedia.org/wiki/File:1317\_CFS\_Circulation.jpg. [Accessed 22 April 2020]. |

# Table of Figures

[Figure 1 - A cyst forming within a cervical spine, characteristic of Syringomyelia. [33] 3](file:///D:\Coursework\Final-Year-Project-2\Reports\Final%20Report.docx#_Toc38330689)

[Figure 2 - The flow of cerebrospinal fluid around the brain, with the perivascular space shown within the inset red box. [5] Licensed under Creative Commons Attribution 4.0 International. 3](#_Toc38330690)

[Figure 3 - A central slice of an MRI scan depicting a healthy Cavalier King Charles Spaniel (left) and one affected by both CM and Syringomyelia (right), with the syrinx indicated by the red rectangle. 4](file:///D:\Coursework\Final-Year-Project-2\Reports\Final%20Report.docx#_Toc38330691)

[Figure 4 - The posterior fossa of the human skull, shown in green. Licensed under the Creative Commons Attribution-Share Alike 2.1 Japan. [34] 4](file:///D:\Coursework\Final-Year-Project-2\Reports\Final%20Report.docx#_Toc38330692)

[Figure 5 - A cross section of an MRI scanner with labelled components. [35] Licensed under the Creative Commons Attribution-ShareAlike 3.0. 5](file:///D:\Coursework\Final-Year-Project-2\Reports\Final%20Report.docx#_Toc38330693)

[Figure 6 - A "healthy" hip joint (top) and one presenting symptoms of hip osteoarthritis with arrows at regions indicating this (bottom) [16]. 6](file:///D:\Coursework\Final-Year-Project-2\Reports\Final%20Report.docx#_Toc38330694)