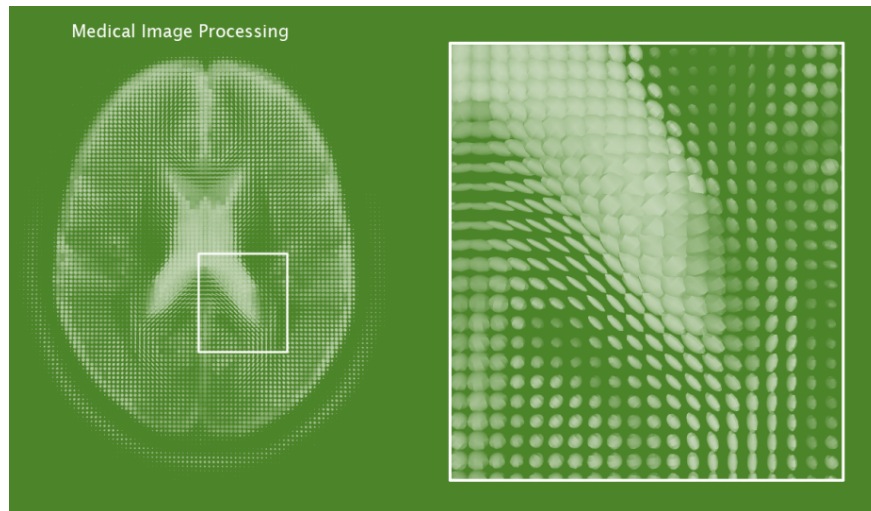




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## Medical Image Analysis with Deep Learning —IV



[Nvidia GTC conference 2017](#) was an excellent source for all the effort on work on health care in Deep learning. Deep learning experts such as [Ian GoodFellow](#), [Jeremy Howard](#) and others shared their perspective on Deep learning. Top medical schools ([Mount Sinai](#), [NYU](#), [Massachusetts General Hospital](#), etc.) and [Kaggle](#)—lung cancer BOWL winners explained their modeling strategies. Coming back to our series, in the last article we talked about basic deep-learning on text and image data. In this article we will focus on the medical images and their formats.

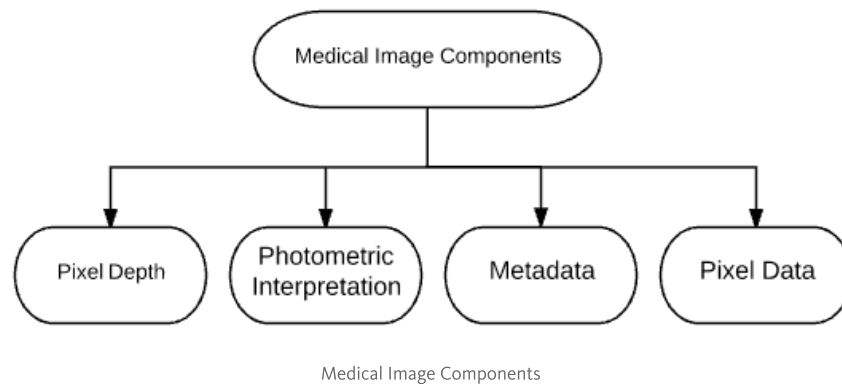
This article is structured into 3 parts—Medical Images and their components, Medical Image formats and their format conversions. The goal is to develop knowledge to help us with our ultimate goal—medical image analysis with deep learning.

### Medical Images & Components

A very good resource for this discussion is [the paper published by Michele Larobina & Loredana Murino](#) from, Institute of bio structures and bioimaging (IBB), Italy. This is part of The National Research Council (CNR). It is the largest public research facility in Italy. Another good source of reference is the paper [“Working with the DICOM and NIfTI Data Standards in R”](#).

**What is a medical Image?** : A medical image is an representation of the internal structure or function of an anatomic region. It is in the form of an array of picture elements called pixels (2 dimensional) or voxels (3 dimensional) . It is a discrete representation resulting from a sampling or reconstruction process that maps numerical values to positions of the space. The number of pixels used to describe the field-of-view of a certain acquisition modality is an expression of the detail with which the anatomy or function can be depicted. What the numerical value of the pixel expresses depends on the imaging modality, the acquisition protocol, the reconstruction, and eventually, the post-processing. (Source: [Link](#))

### Medical Image Components



Medical Images have 4 key constituents—Pixel Depth, Photometric Interpretation, Metadata and Pixel data. These constituents are responsible for the size and resolution of the image.

**Pixel Depth or Bit Depth or Color Depth** is the number of bits used to encode the information of each pixel. For example, an 8-bit raster can have 256 unique values that range from 0 to 255.

| Bit depth             | Range of values that each cell can contain |
|-----------------------|--|
| 1 bit                 | 0 to 1                                     |
| 2 bit                 | 0 to 3                                     |
| 4 bit                 | 0 to 15                                    |
| Unsigned 8 bit        | 0 to 255                                   |
| Signed 8 bit          | -128 to 127                                |
| Unsigned 16 bit       | 0 to 65535                                 |
| Signed 16 bit         | -32768 to 32767                            |
| Unsigned 32 bit       | 0 to 4294967295                            |
| Signed 32 bit         | -2147483648 to 2147483647                  |
| Floating-point 32 bit | -3.402823466e+38 to 3.402823466e+38        |
| Unsigned 64 bit       | 0 to 18446744073709551616                  |

Source: <http://desktop.arcgis.com/en/arcmap/10.3/manage-data/raster-and-images/bit-depth-capacity-for-raster-dataset-cells.htm>

**Photometric Interpretation** specifies how the pixel data should be interpreted for the correct image display as a monochrome or color image. To specify if color information is or is not stored in the image pixel values, we introduce the concept of samples per pixel, also known as number of channels. Monochrome images have one sample per pixel and no color information stored in the image. A scale of shades of gray from black to white is used to display the images. The number of shades of gray depends clearly from the number of bits used to store the sample that, in this case, coincide with the pixel depth. Clinical radiological images, like x-ray computed tomography (CT) and magnetic resonance (MR) images have a gray scale photometric interpretation. Nuclear medicine images, like positron emission tomography (PET) and single photon emission tomography (SPECT), are typically displayed with a color map or color palette. [Source: [link](#)]

**Metadata** is the information that describe the image. It can seem strange, but in any file format, there is always information associated with the image beyond the pixel data. This information called metadata is typically stored at the beginning of the file as a header and contains at least the image matrix dimensions, the spatial resolution, the pixel depth, and the photometric interpretation.

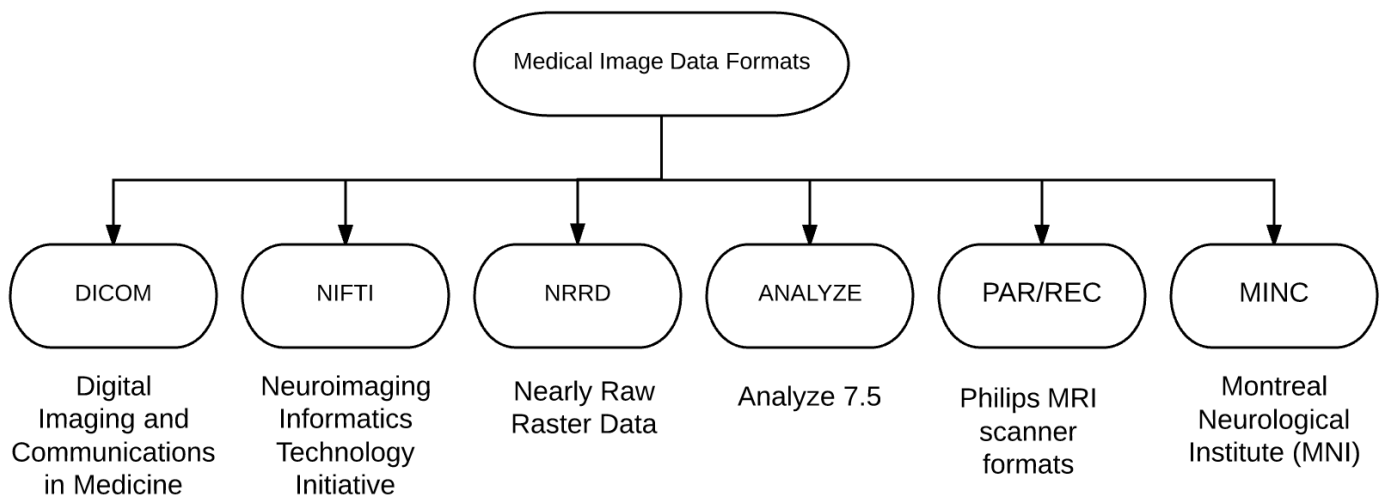
**Pixel Data**—This is the section where the numerical values of the pixels are stored. According to the data type, pixel data are stored as

integers or floating-point numbers using the minimum number of bytes required to represent the values

*Image Size = Header size(includes metadata) + Rows\* Columns \* Pixel Depth \* (Number of Frames)*

## Medical Image Formats

There are 6 predominant formats for radiology images—DICOM (Digital Imaging and Communications in Medicine) , NIFTI (Neuroimaging Informatics Technology Initiative) , PAR/REC (Philips MRI scanner formats), ANALYZE (Mayo Medical Imaging), NRRD (Nearly Raw Raster Data) and MINC.



Medical Image formats as of May 2017

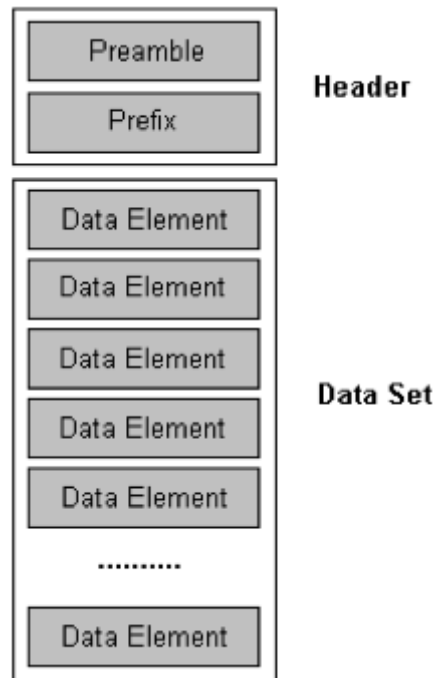
Of these 5, DICOM and NIFTI are the more popular ones.

## DICOM Format Basics

DICOM stands for Digital Imaging and Communications in Medicine. DICOM is standard created by the National Electrical Manufacturers Association (NEMA). It defines a standard for handling, storing, printing and transmitting information in medical imaging. These are the format of files that you can expect right off a scanner or hospital PACS (picture archiving and communication system).

It includes a file format and a network communications protocol that uses TCP/IP to communicate between entities that are capable of receiving image and patient data in DICOM format.

A DICOM file consists of a header and the image data in the same file (\*.dcm). The size of the header depends on how much header information is provided. The header contains information such the Patient Id, Patient Name, Modality and other information. It defines also how many frames are contained and in which resolutions. This is used by image viewers to display the image. For a single acquisition there will be a lot of DICOM files.



Source: <https://www.leadtools.com/sdk/medical/dicom-spec1>

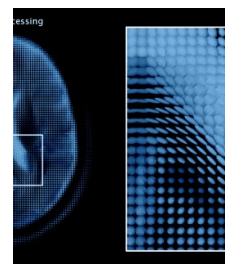
A python library to read dicom files is pydicom. Refer to the code sample in the part 1 of this article.

## Medical Image Analysis with Deep Learning

—|

Analyzing images and videos, and using them in various applications such as self driven cars, drones etc. with...

medium.com



A R-package for reading dicom data is “oro.dicom”.

```
2 #Example to load Dicom Image in R
3 #install.packages("oro.dicom")
4 library(oro.dicom)
5 #Folder where the dicom images are. I am using uncompressed Dicom Images
6 folder = "/Users/toposh/Downloads/dicom_Images/uncompressed/"
7 #Loading dicom images using Read Dicom
8 dicomImages <- readDICOM(folder, verbose = TRUE, recursive = TRUE)
9
```

5:73 (Top Level) R Script

Console Deploy

~/workspace/shinyprojects/shiny-examples/ ↗

```
> #Example to load Dicom Image in R
> #install.packages("oro.dicom")
> library(oro.dicom)
> #Folder where the dicom images are. I am using uncompressed Dicom Images
> folder = "/Users/toposh/Downloads/dicom_Images/uncompressed/"
> #Loading dicom images using Read Dicom
> dicomImages <- readDICOM(folder, verbose = TRUE, recursive = TRUE)
176 files to be processed by readDICOM()
```

100%

Using oro.dicom package to read an Uncompressed DICOM File

## NIFTI Format Basics

I remember Nifti was originally created for Neuroimaging. The Nifti format was envisioned by the Neuroimaging Informatics Technology Initiative (NIFTI) as a replacement for the ANALYZE 7.5 Format. It has its origin in the field of neuro-imaging but can be used in other fields as well. A major feature is that the format contains two affine coordinate definitions which relates each voxel index (i,j,k) to a spatial location (x,y,z).

A python library to read nifti files is nibabel. A R-package for reading nifti data is "oro.nifti".

## Differences between DICOM and NIFTI

The main difference between DICOM and Nifti is that the raw image data in Nifti is saved as a 3d image, where in DICOM you have 2d image slices. This makes NIFTI more preferable for some machine learning applications over DICOM, because it is modeled as a 3d image. Handling a single NIFTI file instead of several hundreds of DICOM is easier. Nifti stores 2 files per 3d image as opposed to dozens in DICOM.

## NRRD Format Basics

The flexible Nrrd format includes a single header file and image file(s) that can be separate or combined. A Nrrd header accurately represents N-dimensional raster information for scientific visualization and medical image processing. National Alliance for Medical Image Computing (NA-MIC) has developed a way of using the Nrrd format to represent Diffusion Weighted Images (DWI) volumes and Diffusion Tensor Images (DTI). Nrrd DWI and Nrrd DTI data can be read into 3D Slicer, to visually confirm that the orientation of the tensors is consistent with expected neuroanatomy[link]

The general format of a NRRD file (with attached header) is:

```
NRRD000X
<field>: <desc>
<field>: <desc>
# <comment>
...
<field>: <desc>
<key>:=<value>
<key>:=<value>
<key>:=<value>
# <comment>

<data><data><data><data><data><data>...
```

Source : <http://teem.sourceforge.net/nrrd/format.html#general.1>

## MINC Format Basics

MINC stands for Medical Imaging NetCDF Toolkit. The MINC file format development was started in 1992 at Montreal Neurological Institute (MNI). Currently there is active work going on at McGill's Brain Imaging center (BCI). The first version of Minc format (Minc1) was based on the standard Network Common Data Format (NetCDF).

Minc2 switched from NetCDF to Hierarchical Data Format version 5 (HDF5). HDF5 supports an unlimited variety of datatypes, and is designed for flexible and efficient I/O and for high volume and complex data. These added features helped Minc2 to work with large and complex data-sets.

Some comparisons obtained from research papers in headers of these formats.

| Format  | Header                                | Data Types   |
|---------|---------------------------------------|--|
| DICOM   | Variable length binary format         | Signed and unsigned integer, (8-, 16-bit; 32-bit only allowed for radiotherapy dose)                       |
| NIFTI   | Fixed-length: 352 byte binary format  | Signed and unsigned integer (from 8- to 64-bit), float(from 32- to 128-bit), complex (from 64- to 256-bit) |
| NRRD    | Extensible with Attached and detached |  |
| ANALYZE | Fixed-length: 348 byte binary format  | Unsigned integer (8-bit), signed integer (16-, 32-bit),float (32-, 64-bit), complex (64-bit)               |
| MINC    | Extensible binary format              | Signed and unsigned integer (from 8- to 32-bit),float (32-, 64-bit), complex (32-, 64-bit)                 |

Source: Medical Image Formats, Springer Publication 2014.

# Format Conversions

## DICOM to NIFTI

A popular tool for converting DICOM to NIFTI is [dcm2nii](#). A python library to read and write nifti files is [nibabel](#). If one would like to convert DICOM to Nifti, there are tools for automatic conversion (e.g. [dcm2nii](#)). Python 2 library “[dcmstack](#)” allows series of DICOM images to be stacked into multi-dimensional arrays. These arrays can be written out as Nifti files with an optional header extension (the *DcmMeta* extension) containing a summary of all the meta data from the source DICOM files. A newer library [dcm2nifti](#) is available for Python 3. I would also recommend the reader to check out the [nipy](#) project.

## DICOM to MINC

MINC team at [BIC](#) has developed a tool to convert from DICOM to MINC images. This program is written in C and the github repo is [here](#).

## NIFTI or ANALYZE to MINC

MINC team at [BIC](#) has developed another tool to convert from NIFTI or Analyze images to MINC images. This program is called [nii2mnc](#). A list of conversion tools by BIC including [nii2mnc](#) is [here](#).

# Conclusions

As we see there are several formats of storing imaging and utilizing them for Deep learning. Our goal is to use the best format that will enable us to get all features we need for our Convolutional Neural Net (CNN) to predict accurately.

In the next article we will discuss how to segment the lungs from a CT scan using one of the formats.

## References

BIC - TheMcConnell Brain Imaging Centre: Home F

McConnell Brain Imaging Centre[www.bic.mni.mcgill.ca](http://www.bic.mni.mcgill.ca)



## DICOM - Wikipedia

Digital Imaging and Communications in Medicine (DICOM) is a standard for handling, storing, printing, and transmitting...

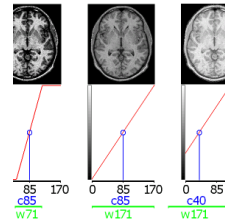
[en.wikipedia.org](http://en.wikipedia.org)



## DICOM introduction and free software

Introduction to the DICOM standard. Includes open source Windows freeware DICOM viewer.

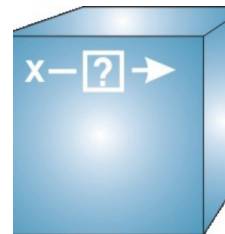
[www.cabiatl.com](http://www.cabiatl.com)



## NIfTI-1 Data Format - Neuroimaging Informatics Technology Initiative

Source Code, Documentation, Example Files and further information

[nifti.nimh.nih.gov](http://nifti.nimh.nih.gov)



## The NIFTI file format

(This article is about the nifti-1 file format. For an overview of how the nifti-2 differs from the nifti-1, see this...

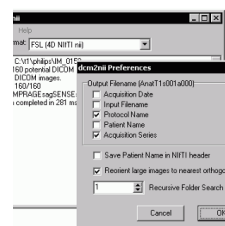
[brainder.org](http://brainder.org)



## dcm2nii DICOM to NifTI conversion

DCM2NII attempts to convert images from the proprietary scanner format to the NifTI format used by FSL, SPM5, MRICron...

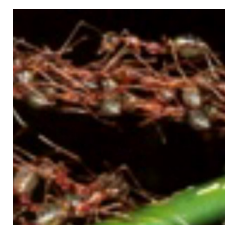
[people.cas.sc.edu](http://people.cas.sc.edu)



## Teem: nrrd: Definition of NRRD File Format

The magic, field specifications, key/value pairs, and comments comprise the NRRD header. After the header, there is a...

[teem.sourceforge.net](http://teem.sourceforge.net)



## NetCDF: Introduction and Overview



NetCDF is a set of software libraries and self-describing, machine-independent data formats that support the creation...

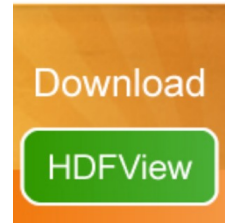
[www.unidata.ucar.edu](http://www.unidata.ucar.edu)



## HDF Group - HDF5

The HDF Group is a not-for-profit corporation with the mission of sustaining the HDF technologies and supporting HDF...

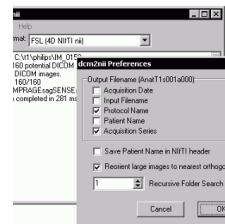
[support.hdfgroup.org](mailto:support.hdfgroup.org)



## dcm2nii DICOM to NIfTI conversion

DCM2NII attempts to convert images from the proprietary scanner format to the NIfTI format used by FSL, SPM5, MRIcron...

[people.cas.sc.edu](mailto:people.cas.sc.edu)



[http://www.medicalimagingandgraphics.com/article/o895-6111\(89\)90285-1/pdf](http://www.medicalimagingandgraphics.com/article/o895-6111(89)90285-1/pdf)

[https://www.researchgate.net/publication/239808108\\_Working\\_with\\_the\\_DICOM\\_and\\_NIfTI\\_data\\_standards\\_in\\_R](https://www.researchgate.net/publication/239808108_Working_with_the_DICOM_and_NIfTI_data_standards_in_R)

[https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3948928/pdf/10278\\_2013\\_Article\\_9657.pdf](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3948928/pdf/10278_2013_Article_9657.pdf)





