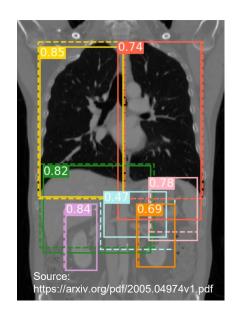
# **Technical Project Tutorial**

Data Science Online Summer School 04/08/2021

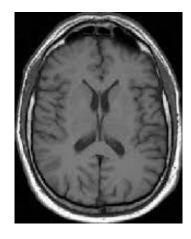
### Common Medical Image Analysis Tasks

#### Localisation



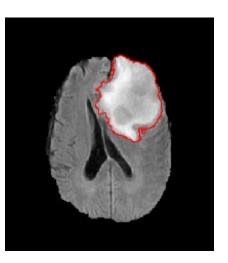
Output: bounding boxes

Classification



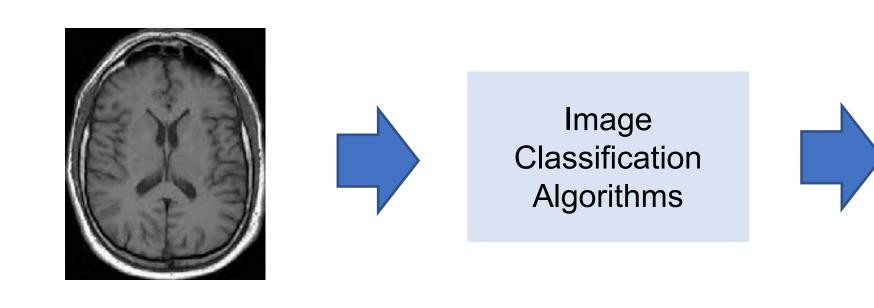
Output: tumour/no tumour

Semantic Segmentation



Output: segmentation map

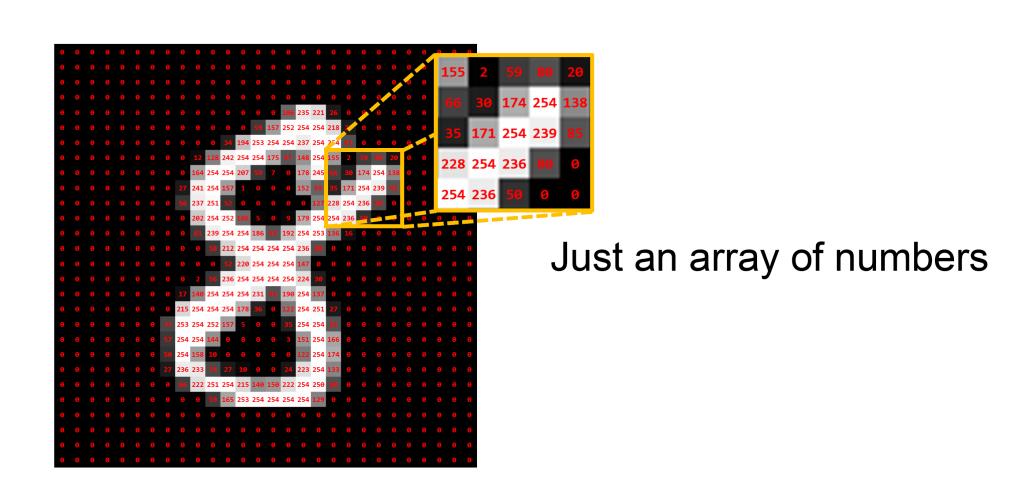
## Example: classification



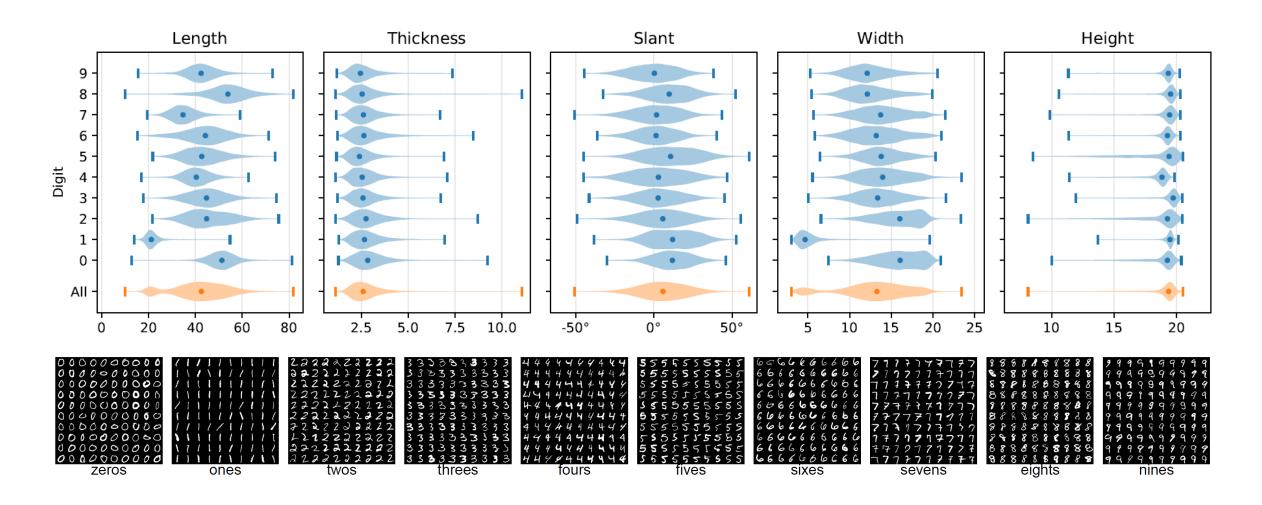
With/without

tumour

### Classification



### Classification



Source: Castro et al. 2018. https://arxiv.org/abs/1809.10780

### Convolution

Input Image

a <sub>11</sub>	a <sub>12</sub>	a <sub>13</sub>	***	a <sub>1</sub>
a <sub>21</sub>	a <sub>22</sub>	a <sub>23</sub>	1806	a <sub>21</sub>
a <sub>31</sub>	a <sub>32</sub>	a <sub>39</sub>	12.00	a <sub>31</sub>
200				
			90	s

Mask

m <sub>11</sub>	m <sub>12</sub>	m <sub>13</sub>
m <sub>21</sub>	m <sub>22</sub>	m <sub>23</sub>
m <sub>31</sub>	m <sub>32</sub>	m <sub>33</sub>

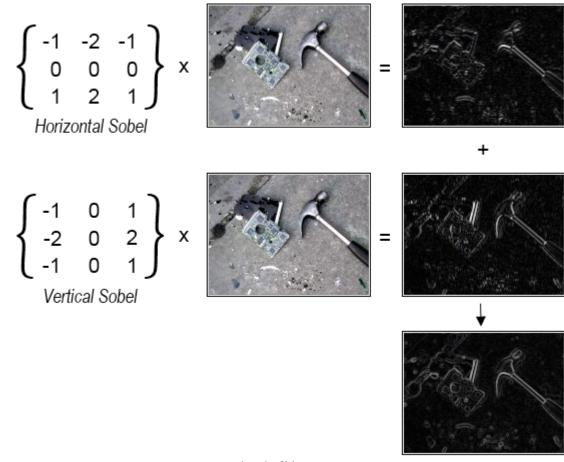
Output Image

b <sub>11</sub>	b <sub>12</sub>	b <sub>13</sub>	3.564	bıı
b <sub>21</sub>	b <sub>22</sub>	b <sub>23</sub>		b <sub>21</sub>
b <sub>31</sub>	b <sub>32</sub>	b <sub>33</sub>		b <sub>3i</sub>
	:	- 1		3

 $b_{22} = (a_{11} * m_{11}) + (a_{12} * m_{12}) + (a_{13} * m_{13}) + (a_{21} * m_{21}) + (a_{21} * m_{21}) + (a_{22} * m_{22}) + (a_{23} * m_{23}) + (a_{31} * m_{31}) + (a_{32} * m_{32}) + (a_{33} * m_{33}) + (a_{31} * m_{31}) + (a_{32} * m_{32}) + (a_{32} * m_{32}) + (a_{33} * m_{33}) + (a_{31} * m_{31}) + (a_{32} * m_{32}) + (a_{32} * m_{32}) + (a_{33} * m_{33}) + (a_{32} * m_{32}) + (a_{33} * m_{33}) + (a_{33} * m_{33$ 

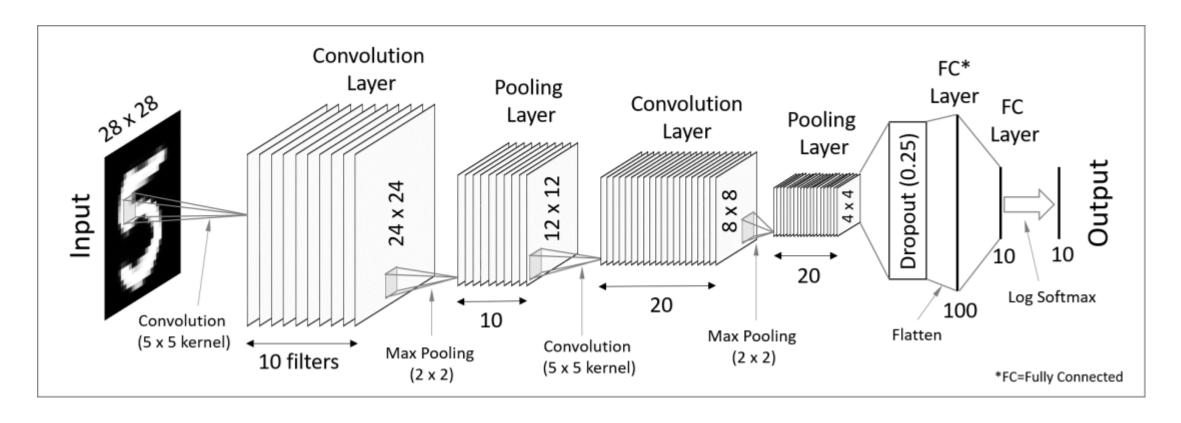
Convolution

### Convolution



Sobel filter

### Classification



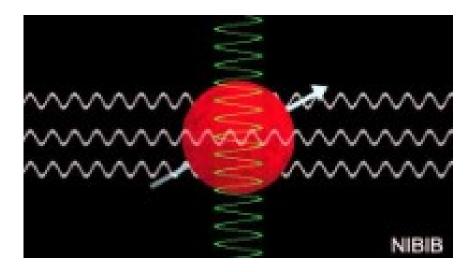
**CNN** model

# **Project Workflow**

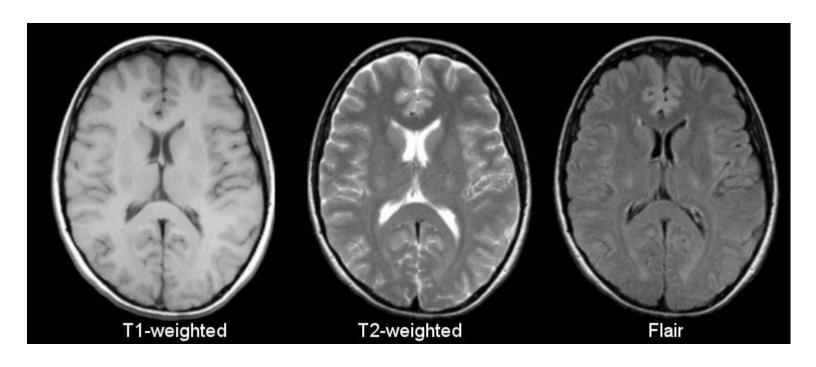
- Understand and preprocess your dataset
- Build the model
- Train the model
- Find the optimal hyperparameters
- Evaluation

### **Brain MRI**

- Non-invasive imaging technology
- Produce three dimensional detailed anatomical images
- Used for disease detection, diagnosis, and treatment monitoring



# Brain MR Images

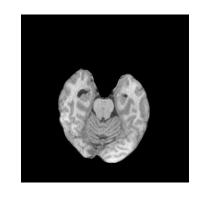


Different MRI sequences

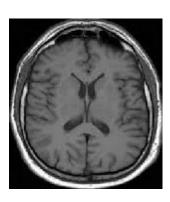
#### Classification Dataset

• 5391 images are given to you for training and testing the model.

• All the images have been preprocessed for you for better training result.



Brain image with tumour

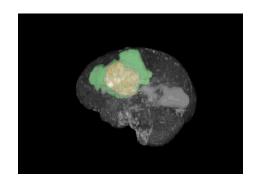


Brain image without tumour

## Segmentation Dataset

 251 MRI volumes and segmentation maps are given to you for training the model. Around 100 volumes will be given to you on 19<sup>th</sup> August for testing your model.

• All the MRI volumes have not been preprocessed.

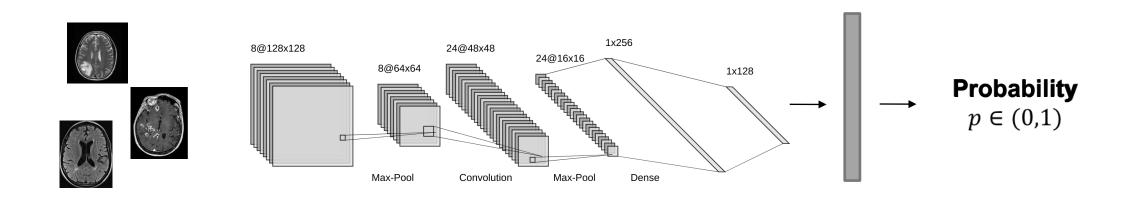


Brain tumour in 3D

# Data preprocessing

- Denoise
- Standardise images (resize, normalisation, etc)
- Data augmentation (rotation, sheer, scale, etc)

### Build the classification model



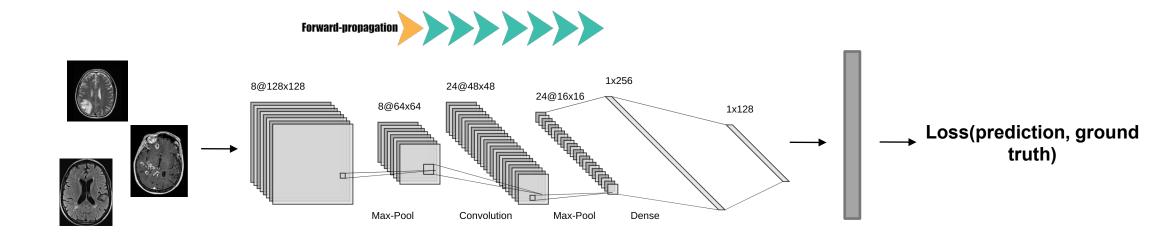
Preprocessed dataset

CNN-based backbone model for feature extraction (VGG16, AlexNet, ResNet, etc)

Classifier

Output

#### Train the classification model



Preprocessed dataset

CNN-based backbone model for feature extraction (VGG16, AlexNet, ResNet, etc)

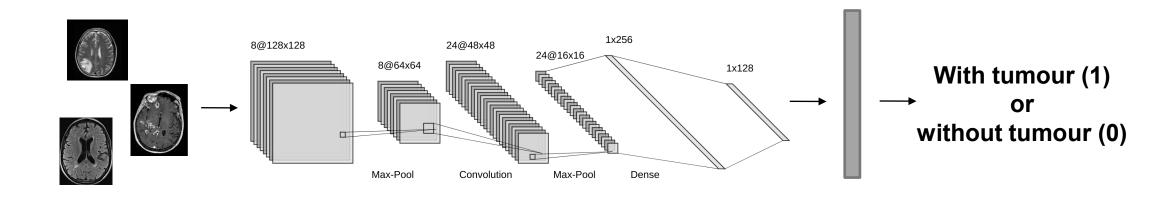
Classifier

Training loss

### Improve the accuracy of the prediction

- Choose the right augmentation methods to tackle the overfitting problem;
- Find the best backbone model;
- Choose to finetune the model or train it from the scratch;
- Set the number of epochs to train the model and early stopping criteria;
- Tune other hyperparameters to achieve the best performance.

#### Test the classification model



Test data

CNN-based backbone model for feature extraction (VGG16, AlexNet, ResNet, etc)

Classifier

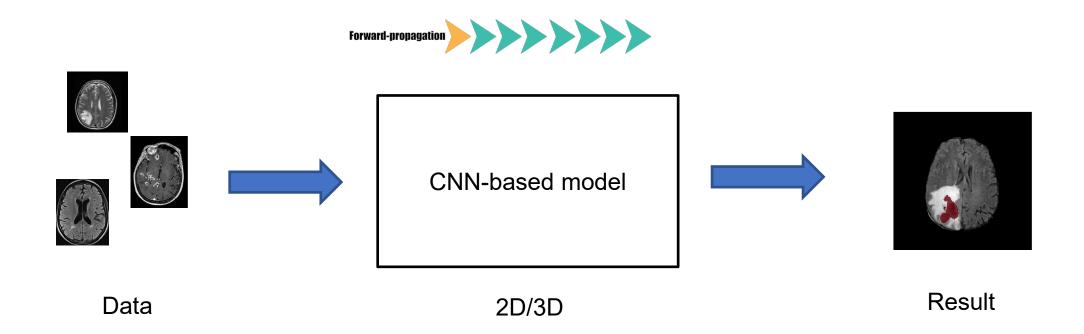
Result

#### Result evaluation

Accuracy is used to evaluate the performance of your classification model, which is defined as

$$Accuracy = \frac{Number\ of\ correctly\ classified\ images}{Number\ of\ total\ images} \times 100\%$$

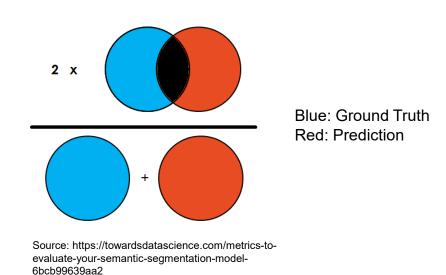
# Segmentation model



#### Result evaluation

Dice score is used to evaluate the performance of your segmentation model, which is defined as

Dice score = (2 \* Area of Overlap)/(total pixels combined)



# Agony of choice

- Model architecture: depth, width, scales, residuals,...
- Loss function: (weighted) cross-entropy, IoU, Dice,...
- Sampling strategy: equally per class, fore/background, uniform,...
- Optimization: optimizer, learning rate, momentum, regularization,...
- Data normalization/standardisation: z-score, bias field correction, histogram matching,...

#### **Setting Up Python Development Environment**

Anaconda Installation (2020.02)

https://repo.anaconda.com/archive/

For using Nvidia GPU

https://github.com/antoniosehk/keras-tensorflow-windows-installation

Packages:

Tensorflow

Scikit-learn

SimpleITK