ASSESSING
MALARIA
BLOOD
SAMPLES
USING
MACHINE
LEARNING

KIERNAN HARDING

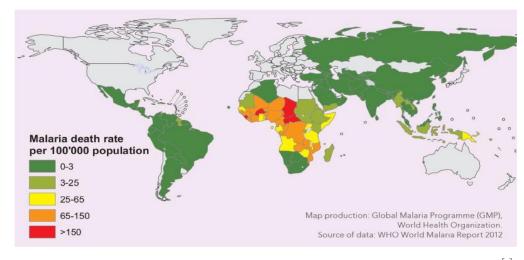
Mal.ARIA

IDENTIFICATION MADE SIMPLE



WHY?

- Malaria has a major impact on global health
- 2018: estimated 228 million cases worldwide & 405,000 deaths [2]
- Only 6 countries account for more than 50% of all cases [2]
- Nigeria accounts for 25% of all cases [2] and has 34% of the world average GDP (per Capita 2017) [8]
- Microscopists are required for testing and this can lead to many problems...







WHY USE AI?

- Yearly hundreds of millions of blood films examined by a trained microscopist
- This involves manual counting of parasites in red blood cells = timely
- Microscopists in poorer areas may have low quality training
 - Low-resource areas and poor-quality control setting
- Lead to incorrect diagnosis
 - False positives and <u>False negatives</u>
- Standardise using Al
- Faster testing
- Reduced workload
- More reliable (with an accurate model)





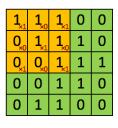
RESEARCH TOPICS

- Effects of malaria on world health and why the testing process would benefit from AI
- Using deep learning for image classifiers: specifically convolutional neural networks
- How to use TensorFlow and Keras to create a convolutional neural network (dogs and cats tutorial)
- How to implement a CNN model into a website (using python with an interactive website)





WHAT IS A CNN? (ML)



Image





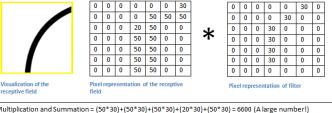




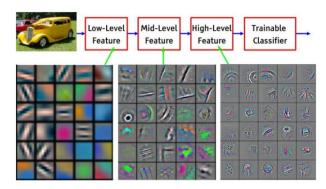
Class of deep learning neural networks

Convolutional Layer [10]

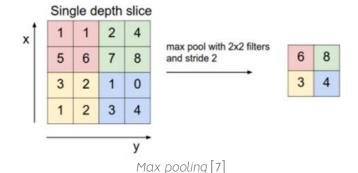
- Learns similarly to a child inspired by the human brain
- Trained through backpropagation:
 - Forward pass pass input data through network as normal as caches values
 - Backwards pass go back through network and alter loss (similar to telling a child they're incorrect, which they then learn from)
- Different types of layers:
 - Convolutional identifies features in images, such as straight lines and curves, aids learning process (becoming less abstract over time)
 - Activation (e.g. ReLU) aids the network to learn complex patterns in the data - increase non-linearity (ReLU: returns 0 if negative and value x if positive)
 - Pooling (e.g. max pooling) reduces sample size and thus speeds up processing
 - Fully connected essentially the output layer (provides probabilistic values)



Applying filters to an image [5]



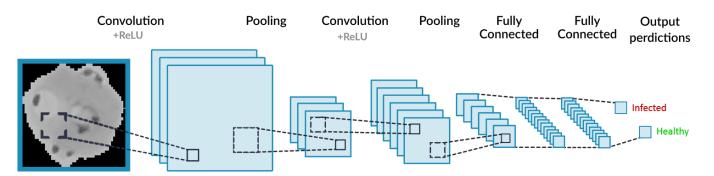
Convolved feature maps [9]





CREATING MY CNN MODEL

- Splitting the dataset into training and testing (80/20)
- Preparing the dataset for training resizing, greyscale...
- Created a convolutional neural network in python using TensorFlow and Keras
 - 3 main layers and out output layer (the fully connected layer)
 - These layers include: convolutional, activation (ReLU), pooling and fully connected
- Created a script and tested the models prediction accuracy: 71.4% (3,936/5,512) overfitting
 - Incredibly bias towards infected samples: parasitized = 91.1%, healthy = 51.7%
- Created a windows batch file which automatically installs the required libraries for running any project related code





UI - WEB APPLICATION

- Use of web development to increase the applications accessibility in poorer areas
- Microscope, camera, ideally an internet connection
- No specific requirements (e.g. android application = limits reach)
- Uses the Flask library to connect the python script to a website
- Simple as possible
- Takes an image input of a blood sample
- Classifies and outputs whether the sample is infected

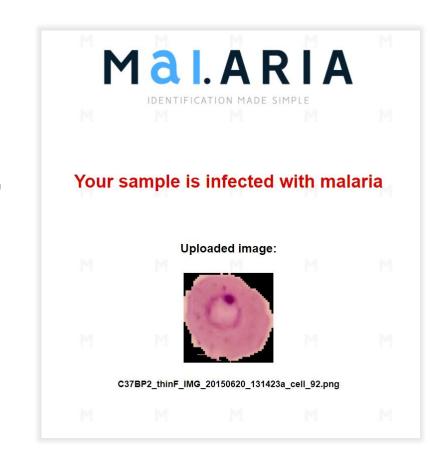




LIVE DEMO

- A web application that includes a back-end machine learning classifying model
- Examples of infected blood cells being classified
- Examples of healthy blood cells being classified







MAIN ISSUES ENCOUNTERED

- Installation of the required local environment e.g. libraries such as TensorFlow
- Issues with array types between libraries Occurred whilst creating a working convolutional neural network using tutorials (using the Kaggle dogs and cats dataset)
- Trying to save, use and classify the input image whilst designing the web application





MAIN INTERIM GOALS – HOW DID IT GO?

What was adjusted and achieved?

Specific MVP milestones	Estimated start/end dates
Create and train a convolutional neural network using python that identifies whether a red blood cell is infected.	19/10/20 - 16/11/20
Test the CNN and adjust accordingly to increase accuracy and remove unfavorable features such as overfitting.	16/11/20 - 23/11/20
Connect the algorithm to a local host web page. At this point it should take an image input and return whether the sample is infected.	23/11/20 – MVP Presentation



TIMELINE OF FUTURE OBJECTIVES

Improve my convolutional neural network model through research and structured testing (e.g. add colour)

8/2/21

into my website to improve usability across all devices

22/2/21

Improve the visual styling of the user interface

8/3/21

Slack for any issues encountered before the submissions

17/3/21

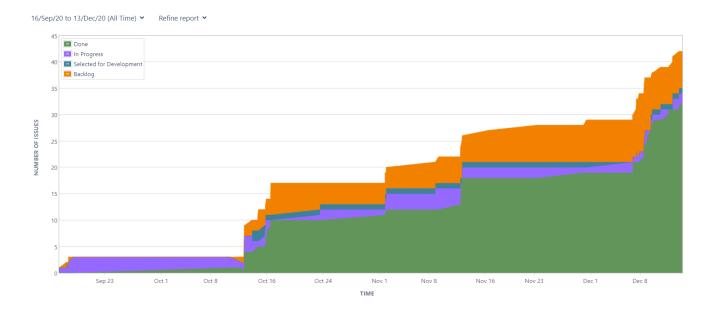
Now 17/3/21 – Open Day Submissions



USE OF JIRA

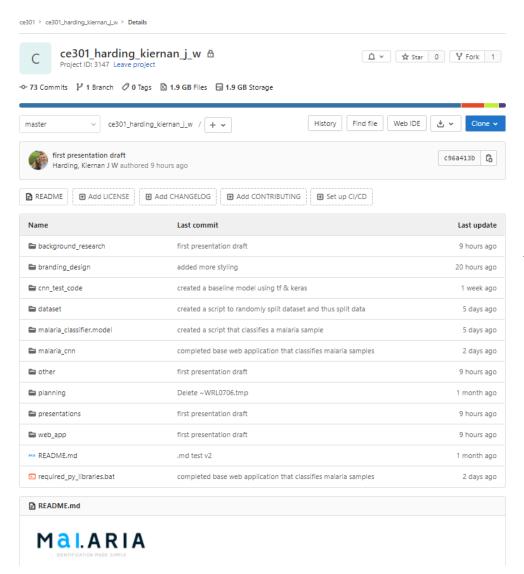
- Consistent use of Jira
- Use of tasks, sub-tasks comments and releases to document updates and plan future tasks

Cumulative flow diagram since the start of September





- Consistent use of Git
- Regular commits to Git for:
 - Significant changes
 - Keep important work backedup!
 - Small required alterations: deletions
- Short, meaningful commit messages





Snapshot of Git 14/12/20

THANK YOU, I HOPE YOU ENJOYED.

ANY QUESTIONS?



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