



# Detecting Malaria Using Convolutional Neural Networks

By Nathaniel Abraham Green





# Presentation Agenda



	1.	Background on Malaria	*
	2.	Stakeholder	·
	3.	Data	
	4.	Models	
<b>*</b>	5.	Conclusions	



# Quick Facts on Malaria

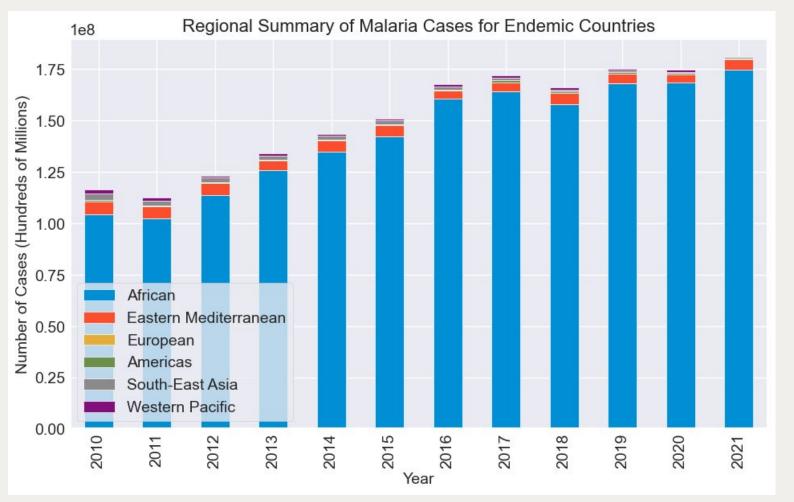
- Malaria is a parasite that can be transmitted to humans through mosquito saliva
- Once in the body, the malaria parasite lives in and destroy red blood cells
- This disease can cause flu like symptoms in most hosts and potentially death if untreated, especially in vulnerable populations
- The best method for malaria detection is to use microscopy to look at infected cells under a microscope.

### **Uninfected**



**Parasitized** 

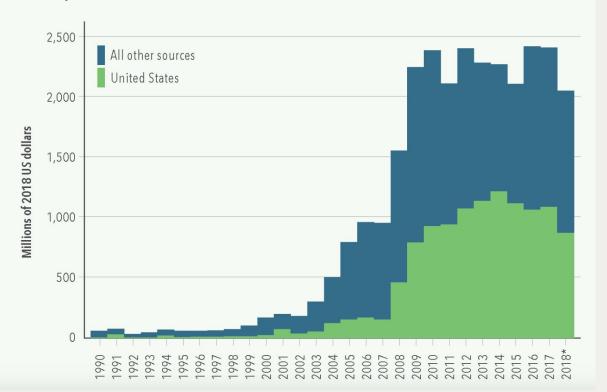




The number of worldwide malaria cases has increased in recent years with most cases occurring in Africa

### Donor funding for malaria was **flat** from 2010 to 2018.

### Development assistance for malaria, 1990-2018



In their 2022 report on Malaria, the World **Health Organization** said that a lack of consistent, sufficient funding was one of the greatest challenges in the effort to eradicate Malaria.





# Stakeholder: World Health Organization (WHO)



### The Problem

The WHO wants to reduce global malaria incidence by at least 90% by 2030, however inadequate global malaria funding has left the WHO's current malaria **monitoring** system understaffed and under resourced



### My Solution

Launched a CNN model through Streamlit that:

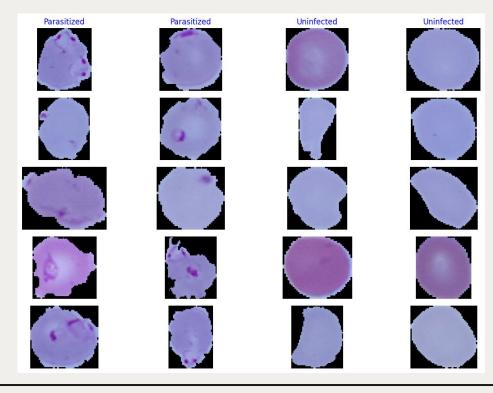
- Stream lines malaria detection efforts in these understaffed and under resourced areas
- Allows the WHO to more accurately and intentionally distribute malaria aid in areas impacted the most by malaria outbreaks

# My Dataset

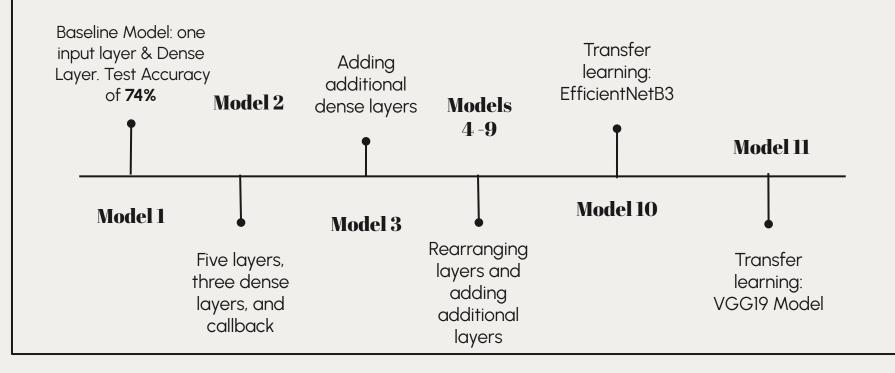
My dataset contains 27,558 blood cell samples photographed through a light microscope at 100X zoom from the National Library of Medicine

There are equal instances of parasitized (infected) and uninfected cells with diverse staining techniques

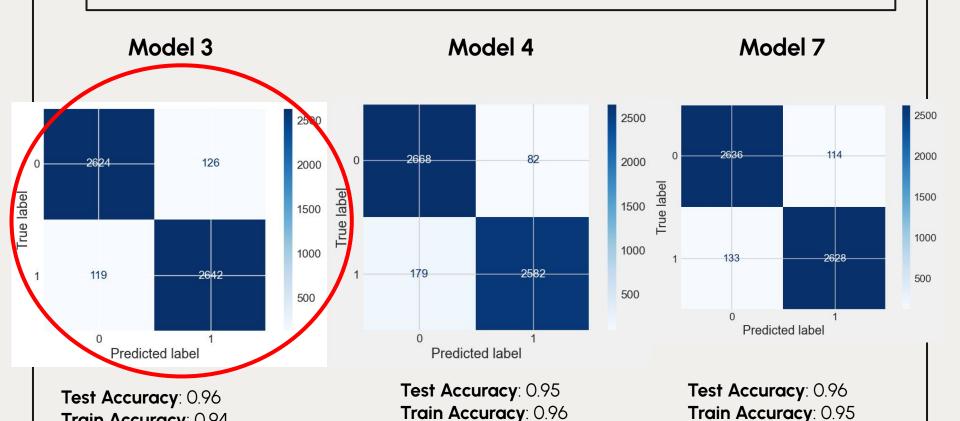
The parasitized cell samples show various stages of malaria infections



# **Model Timeline**

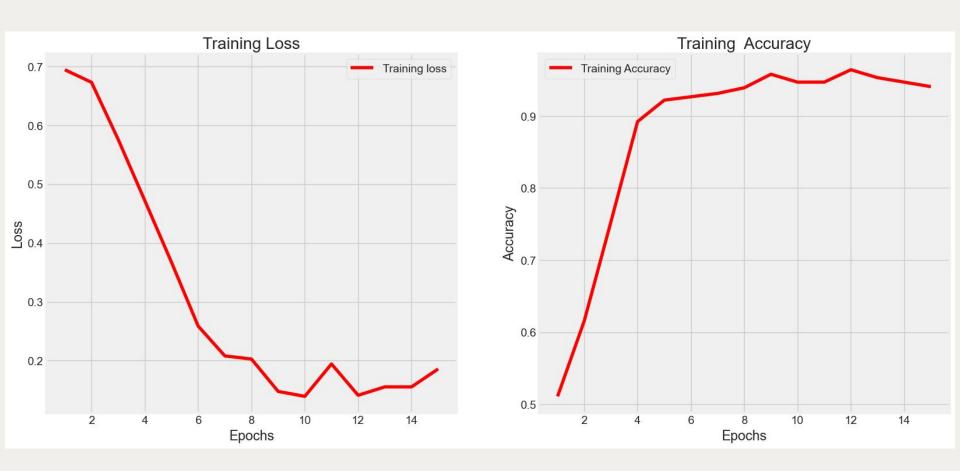


# **Model Performance**



Train Accuracy: 0.94

### **Model 3 Training Loss and Accuracy**



(None, 126, 126, 32)	
(None, 63, 63, 32)	0
(None, 61, 61, 64)	18496
(None, 30, 30, 64)	0
(None, 28, 28, 128)	73856
(None, 14, 14, 128)	0
(None, 12, 12, 256)	295168
(None, 6, 6, 256)	0
(None, 4, 4, 512)	1180160
(None, 2, 2, 512)	0
(None, 2048)	0
(None, 128)	262272
(None, 64)	8256
(None, 32)	2080
(None, 1)	33
	(None, 63, 63, 32)  (None, 61, 61, 64)  (None, 30, 30, 64)  (None, 28, 28, 128)  (None, 14, 14, 128)  (None, 12, 12, 256)  (None, 6, 6, 256)  (None, 4, 4, 512)  (None, 2, 2, 512)  (None, 2048)  (None, 128)  (None, 64)  (None, 32)

Model: "sequential\_2"

Non-trainable params: 0

# **Conclusions and Next Steps**

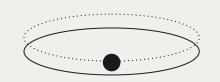
### Recommendations

- Given that my model is categorizing blood samples as being infected/uninfected with malaria with 96% accuracy, I would recommend the WHO use my Streamlit app.
- Next, I would recommend the WHO train field works, local doctors, and microscopists in high needs areas on how to use the Streamlit app.

### **Next Steps**

- Continue to run models with the goal of reducing false negatives (improve recall)
- Update streamlit app to intake patients information
- Connect Streamlit website to mongoDB in order to save results in a database to improve the WHO's malaria surveillance system

# Thanks!





## Do you have any questions?

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