

Skin Cancer Detection

Cher Wee Zheng, Ng Wei, Ryan Yong



We Are



You Are





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Introduction & Problem Statement





Joseph Liu



Joseph Liu



Who is Joseph?

Joseph is a PR
construction foremen
and father of 2



Joseph Liu



Who is Joseph?

Joseph is a PR construction foreman and father of 2

What are his goals?

Joseph wants to pay off his 30 year hdb loan.



Joseph Liu



Who is Joseph?

Joseph is a PR construction foreman and father of 2

What are his goals?

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What are his concerns?

Joseph noticed a large black spot on his forearm

Joseph Liu



Who is Joseph?

Joseph is a PR construction foreman and father of 2

What are his goals?

Joseph wants to pay off his 30 year hdb loan.

What are his concerns?

Joseph noticed a large black spot on his forearm

What's the issue?

Joseph does not want to spend unnecessary money

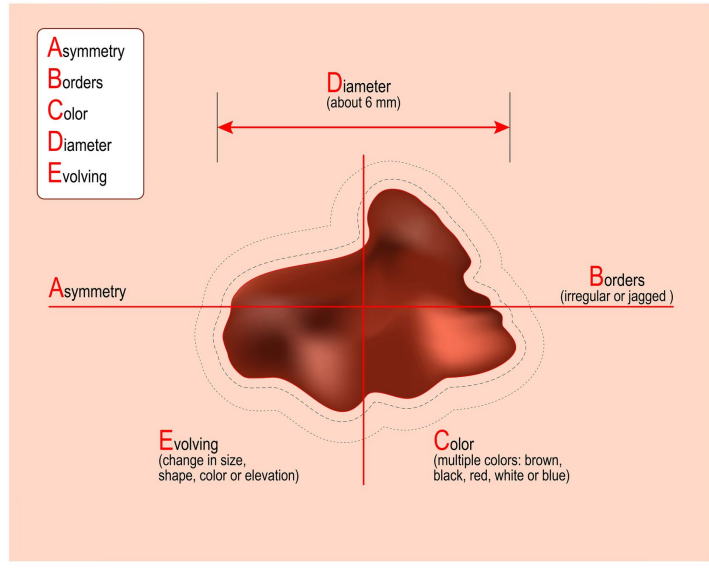


What should Joseph do?

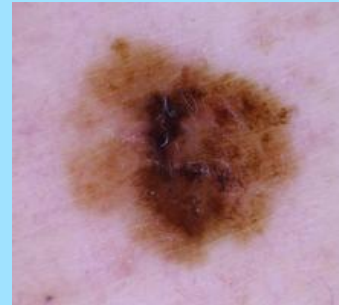
Skin Cancer

Source: CNA

ABCDE (the first signs of a melanoma)



Cancerous



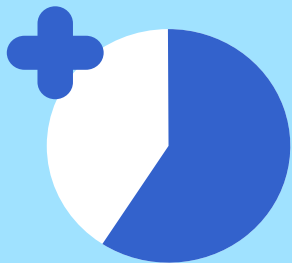
Cancerous




Non-cancerous


Types of Examination

Visual

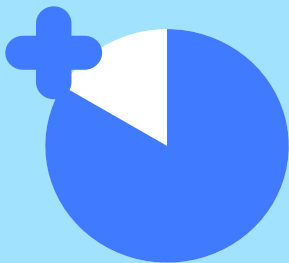


60%


 Instant


 \$35 - \$200

Dermoscopy

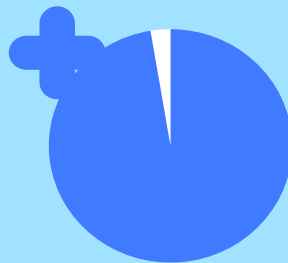


85%

 Instant


 From \$250

Biopsy



95%

 3 weeks

 \$500 - \$3000



Problem Statement

Can we develop a model to be able to **effectively** and **accurately** distinguish between a benign or malignant skin cancer spot visually, improving visual tests and **reducing the necessity** for more **invasive** biopsies?

Key Objectives



01

Maximise

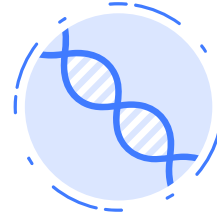
Accurate
diagnosis/treatment
at polyclinic level



02

Reduce

Unnecessary
specialist
referrals



03

Reduce

Costs



Modelling Process

Exploration

An overview of the dataset gathered and how the data is preprocessed

Modelling

Using our dataset, develop an effective and accurate model

Tuning

Improve our chosen model by altering its hyperparameters

Implementation

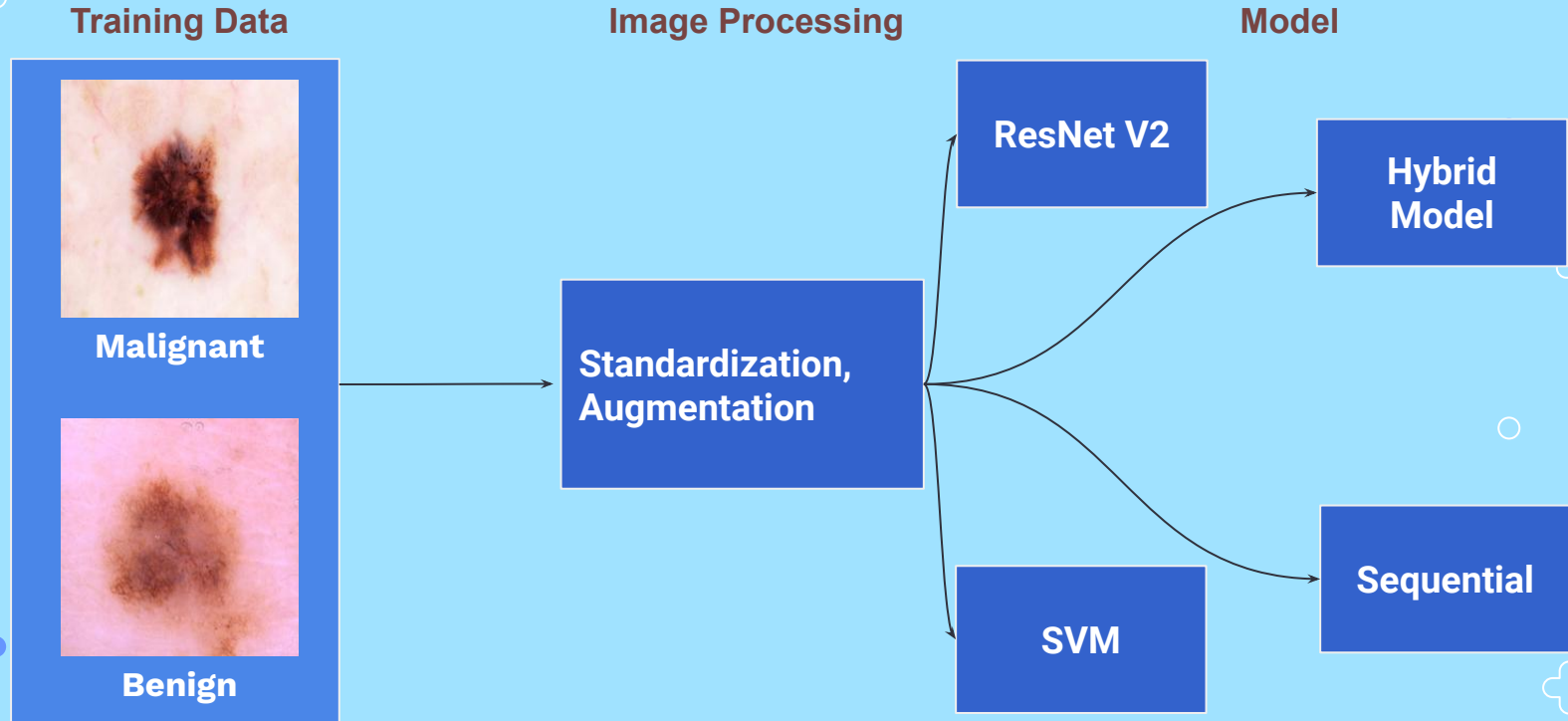
We work together with you to upload our app in your systems



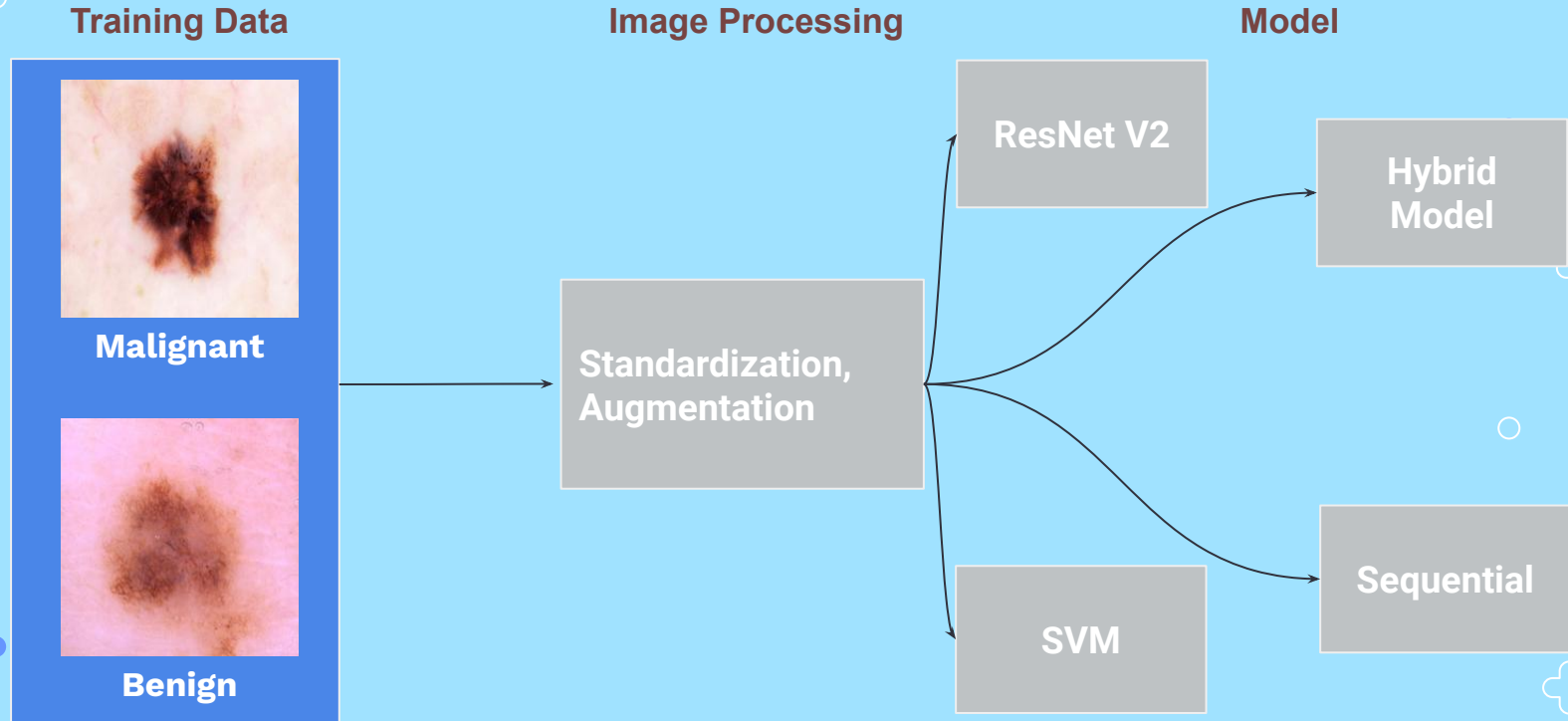
02

EDA & Preprocessing

Modelling Process

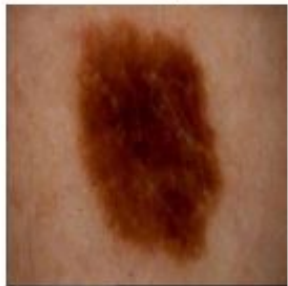


Modelling Process



Our Data are all images!

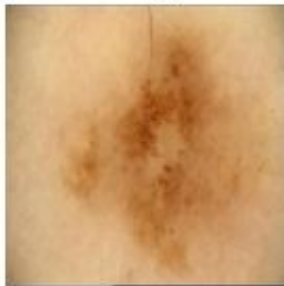
Benign



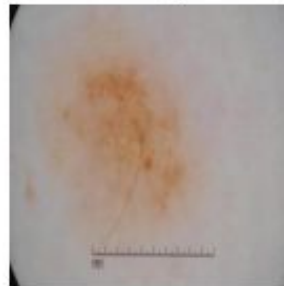
Benign



Benign



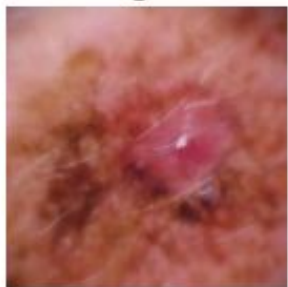
Benign



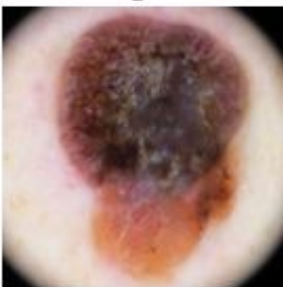
Benign



Malignant



Malignant



Malignant



Malignant



Malignant



Our Data



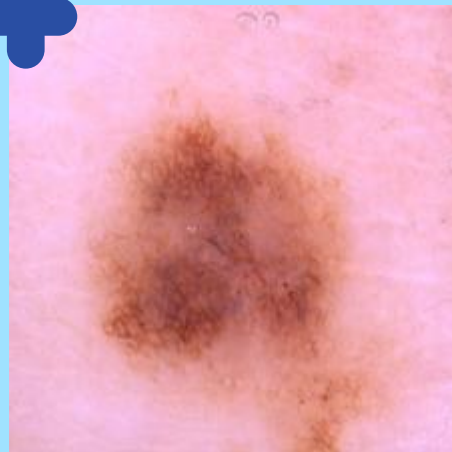
Malignant

5590

Training Data

1000

Testing Data



Benign

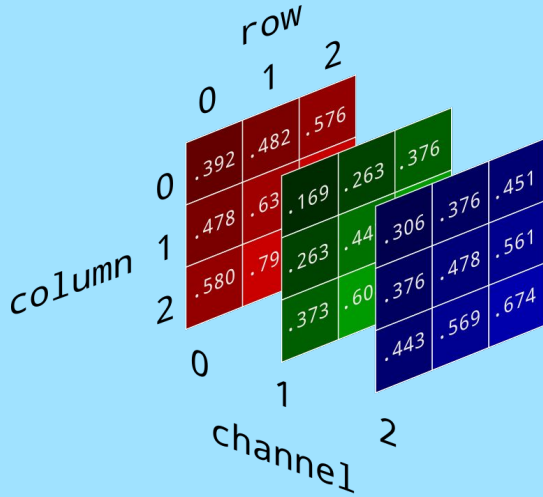
6289

Training Data

1000

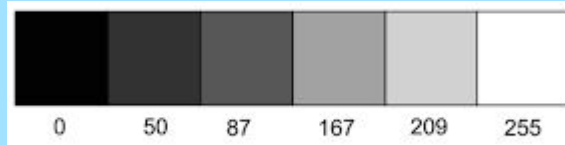
Testing Data

How does models see images?

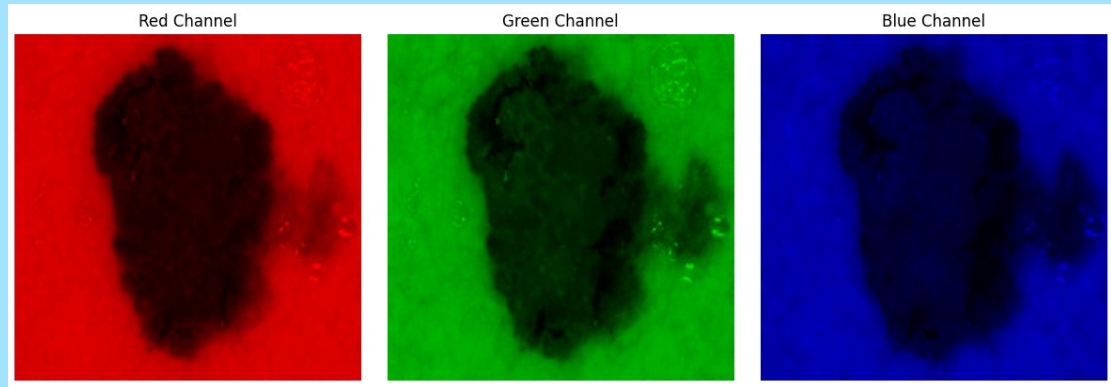


Images are **composed of pixels**
Each pixel contains **Red**, **Green** and **Blue** information.

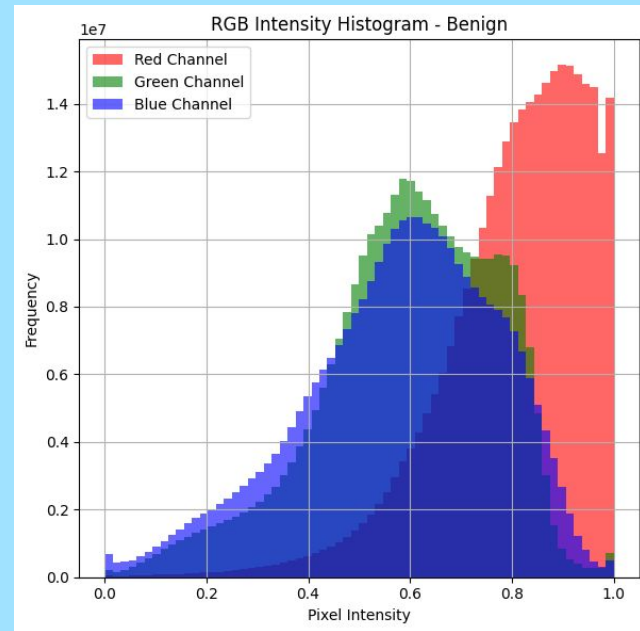
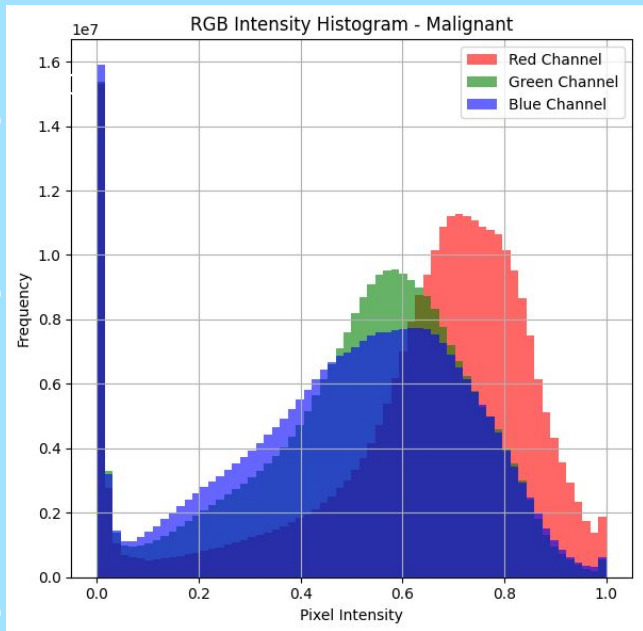
Normally **0-1** or **0-255** to representing each **color's intensity**.



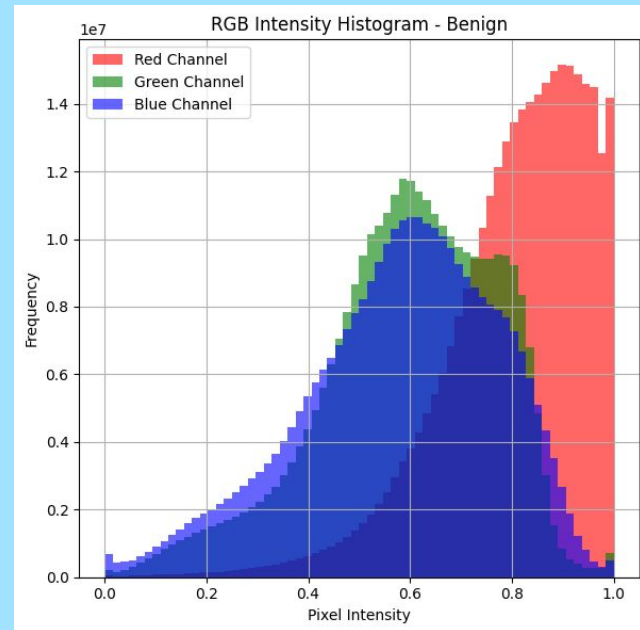
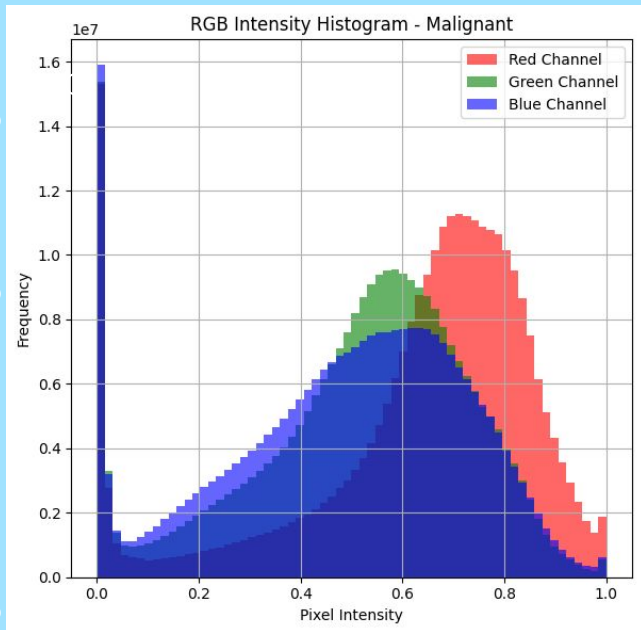
=



Histogram - Malignant vs Benign

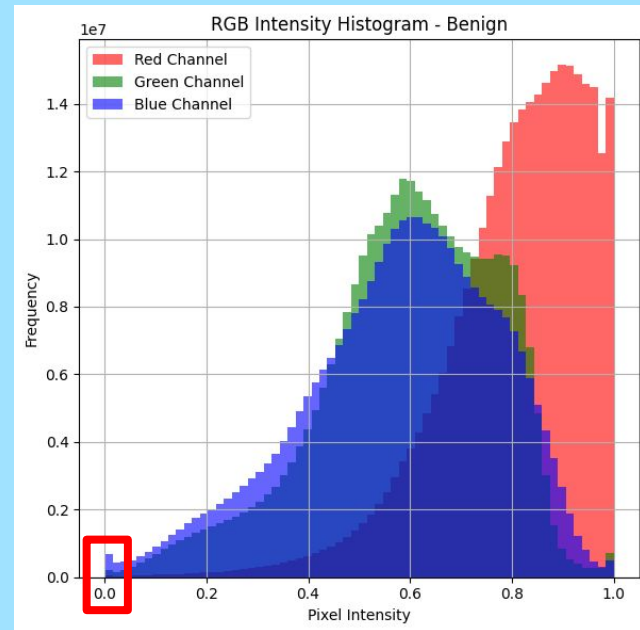
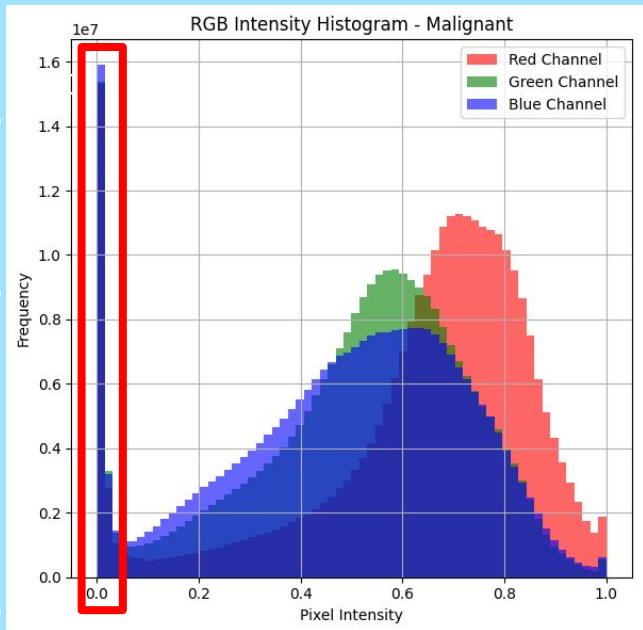


Histogram - Malignant vs Benign



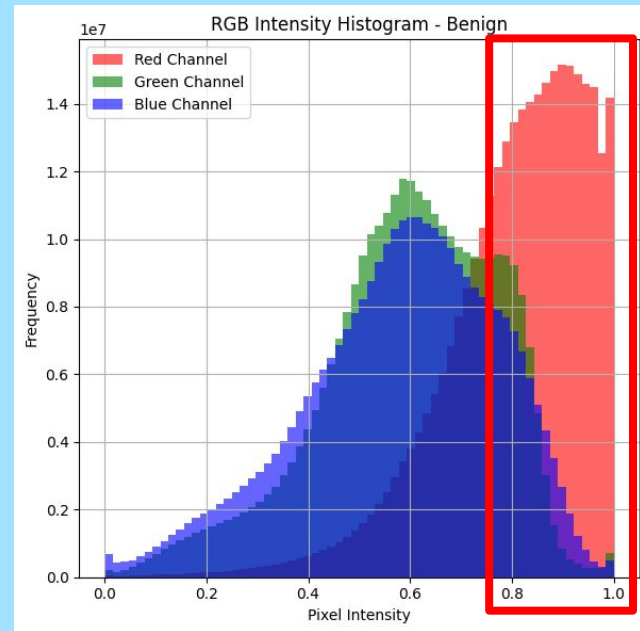
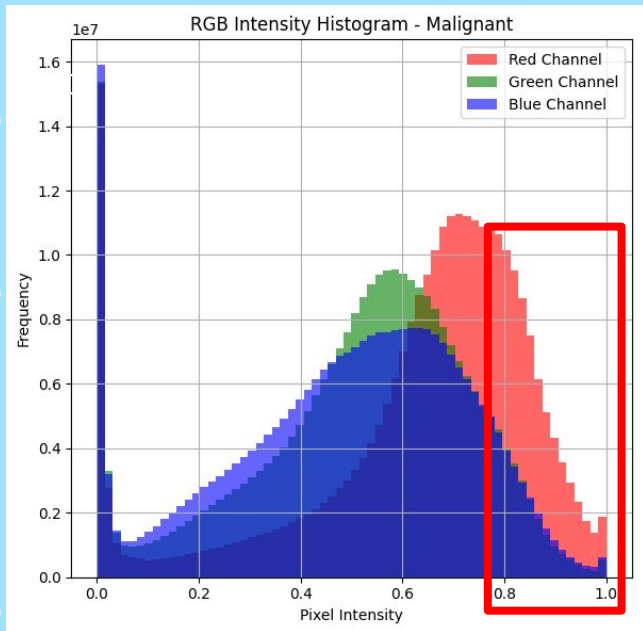
Finding 1:
Based on the average of RGB value, benign seem like brighter than malignant

Histogram - Malignant vs Benign



Finding 2:
Malignant image has many black pixels (R,G,B = 0) compare to Benign

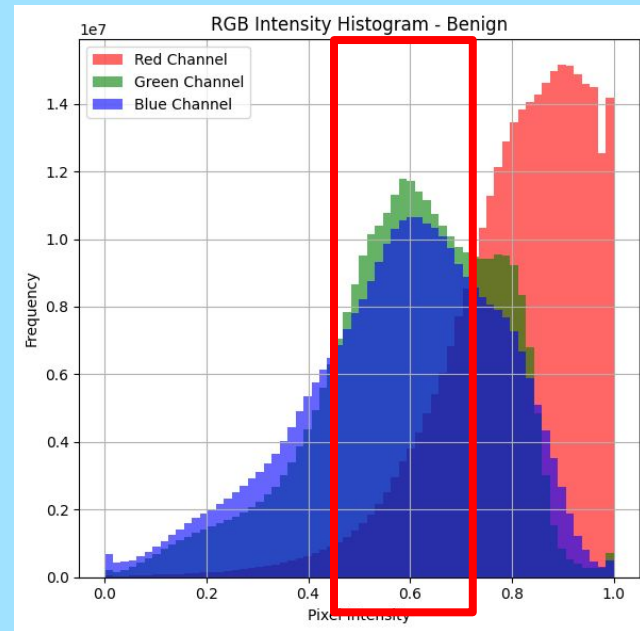
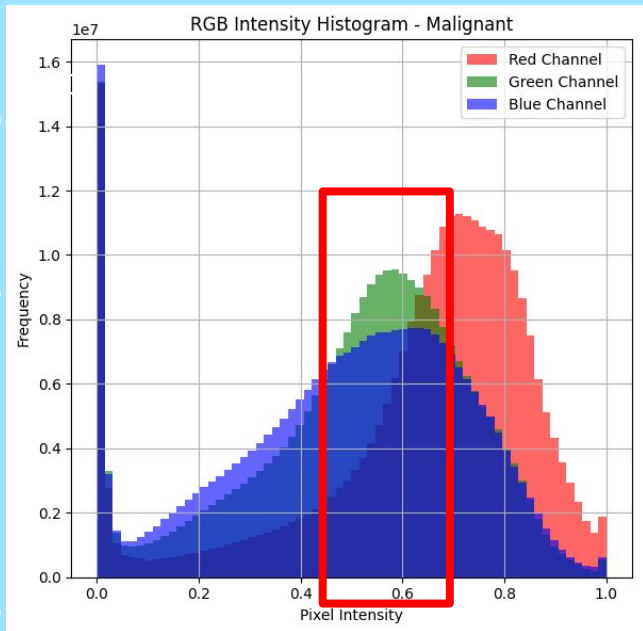
Histogram - Malignant vs Benign



Finding 3:

Benign image is more “red” than Malignant, most red pixel > 0.75

Histogram - Malignant vs Benign

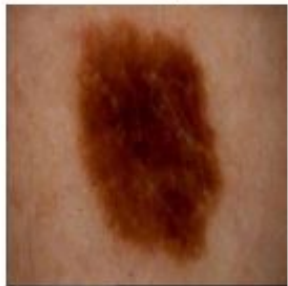


Finding 4:

Beside the black pixel from malignant, both Green and Blue mostly distributed in between 0.45 - 0.7 intensity

What we expected to see.

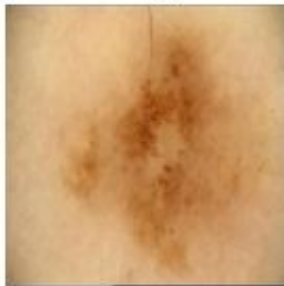
Benign



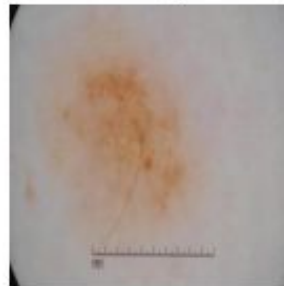
Benign



Benign



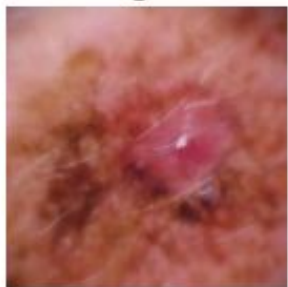
Benign



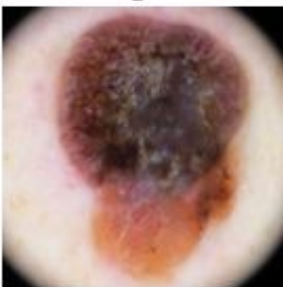
Benign



Malignant



Malignant



Malignant



Malignant



Malignant



The Outliers!

Benign Outlier (RGB < 0.35)

4627.jpg



5211.jpg



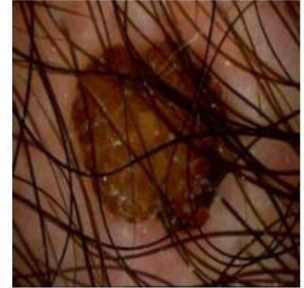
5160.jpg



3573.jpg

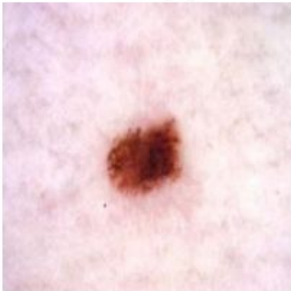


2089.jpg



Malignant Outlier (Red > 0.75, Green and Blue > 0.45)

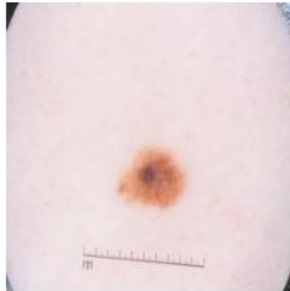
2474.jpg



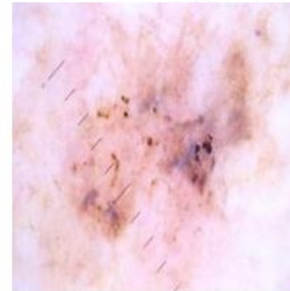
5119.jpg



1028.jpg



5291.jpg



52.jpg



Modelling Process

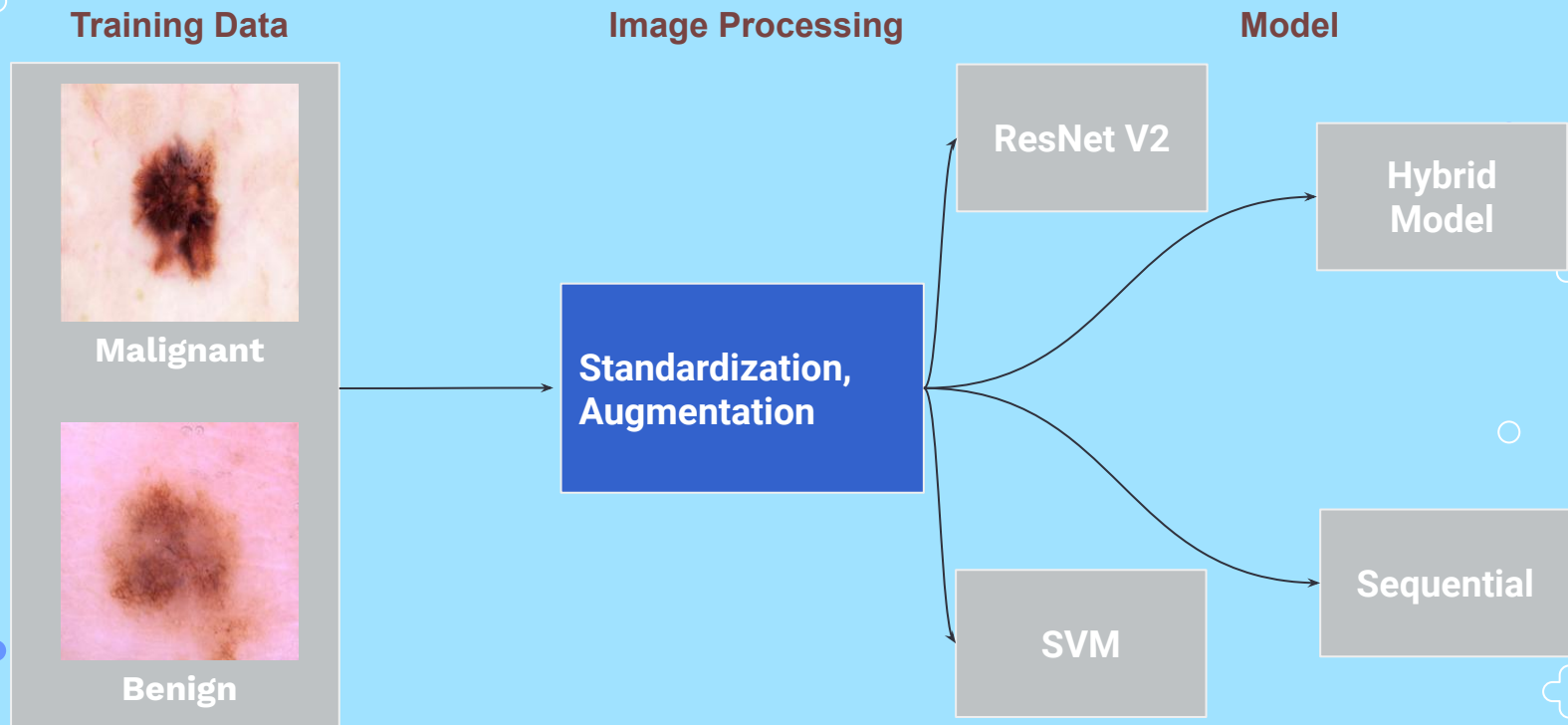
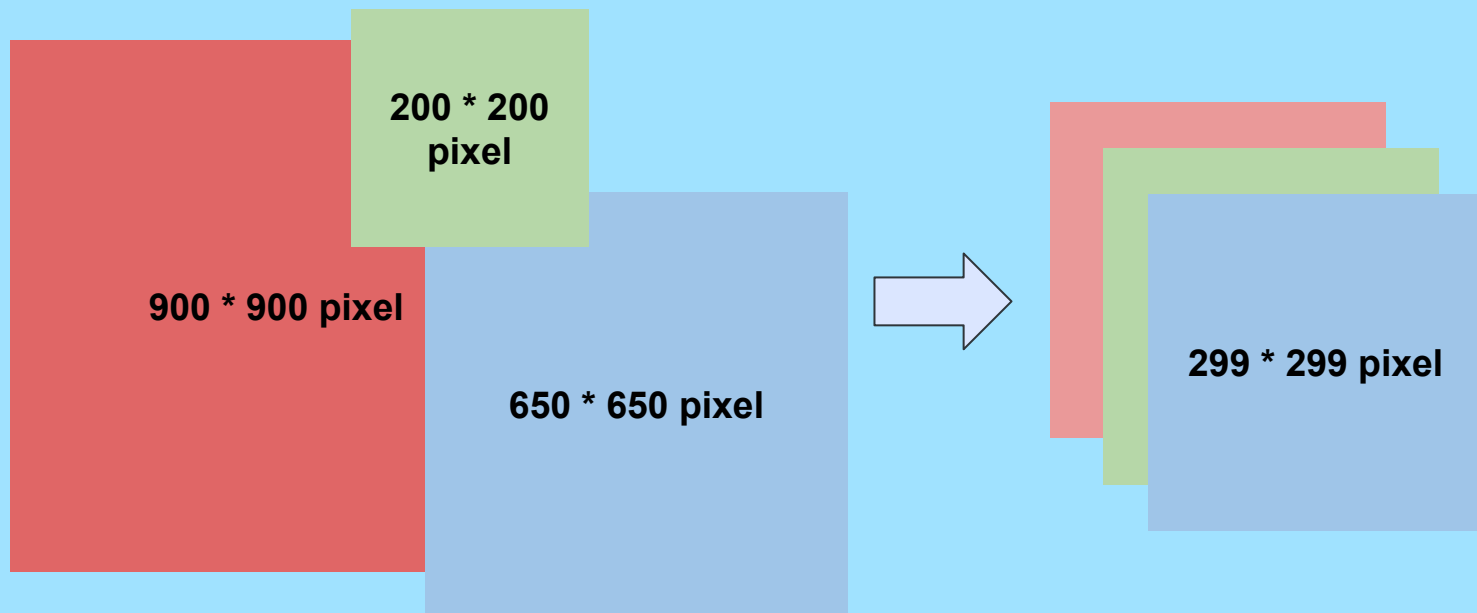


Image Preprocessing

- **Standardization - Resize image to 299*299 pixel**



In order to obtain discriminative internal features that can lead to excellent performance, we adopt an image preprocessing method to **reduce variance of images** and **enhance inherent characteristics of images**.

Image Preprocessing

- Standardization - Scale the RGB from 0-255 to 0-1

Original Pixel Information

0	16	32
64	4	8
128	255	2

Normalizing the Pixel Info

$0/255$	$16/255$	$32/255$
$64/255$	$4/255$	$8/255$
$128/255$	$255/255$	$2/255$

After Normalized Pixel Info

0.00	0.06	0.13
0.25	0.02	0.03
0.50	1.00	0.01

Instead of using StandardScaler, we can just simply divide the original intensity value with 255 to normalize the value.

Image Preprocessing

- **Augmentation - Generate new transformed images**

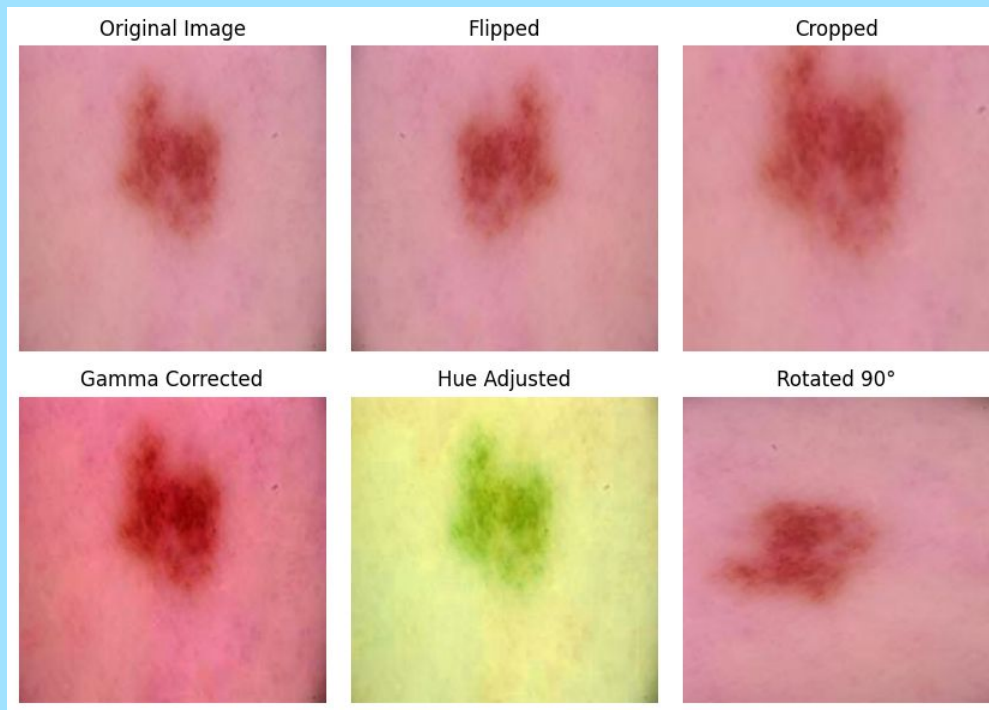


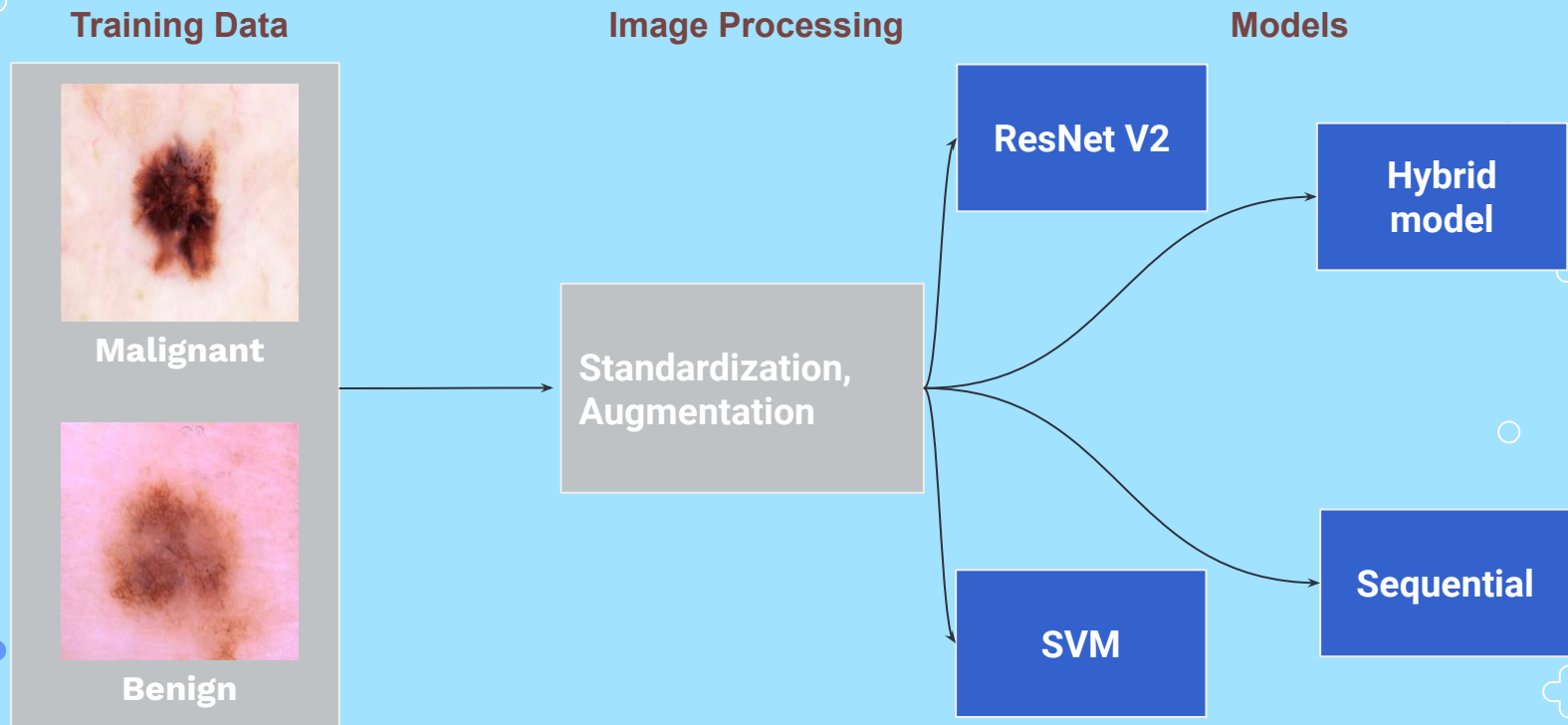
Image augmentation supplements the creation of image variations that can **help models improve its prediction accuracy** and **enhance the generalization ability of models.**



03

Model

Modelling Process





ResNetV2



Hybrid Model

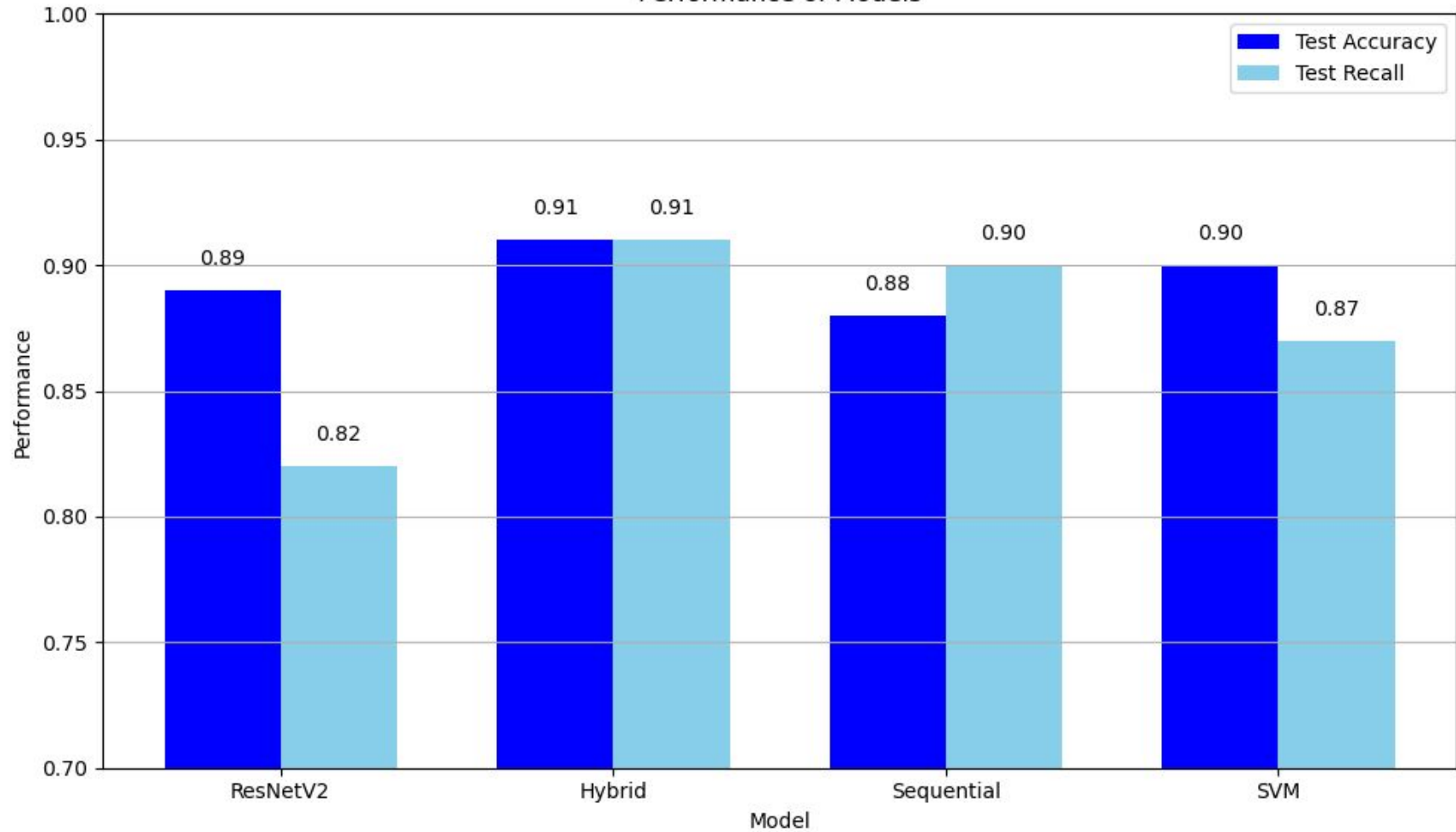


Sequential

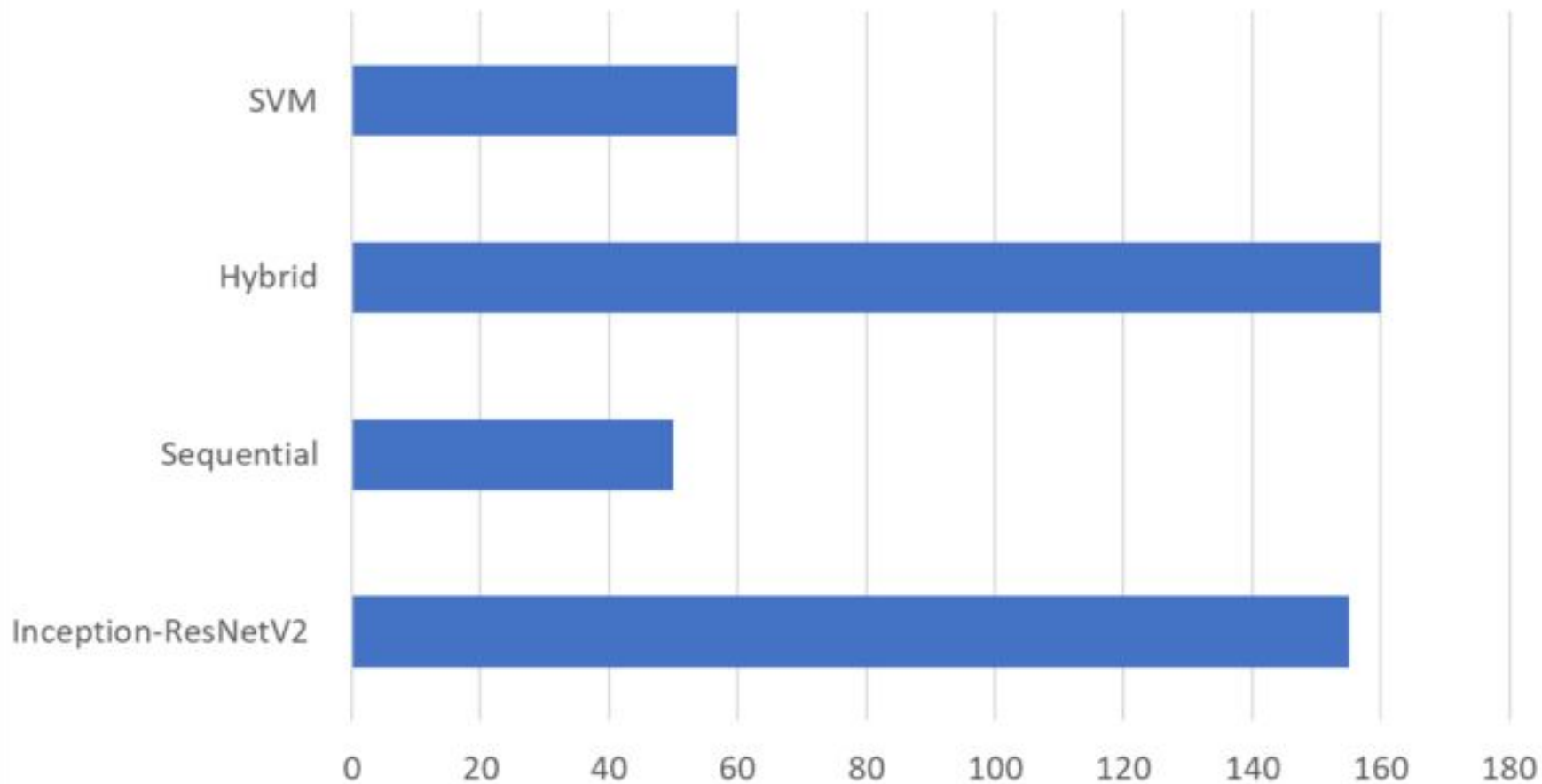


SVM

Performance of Models



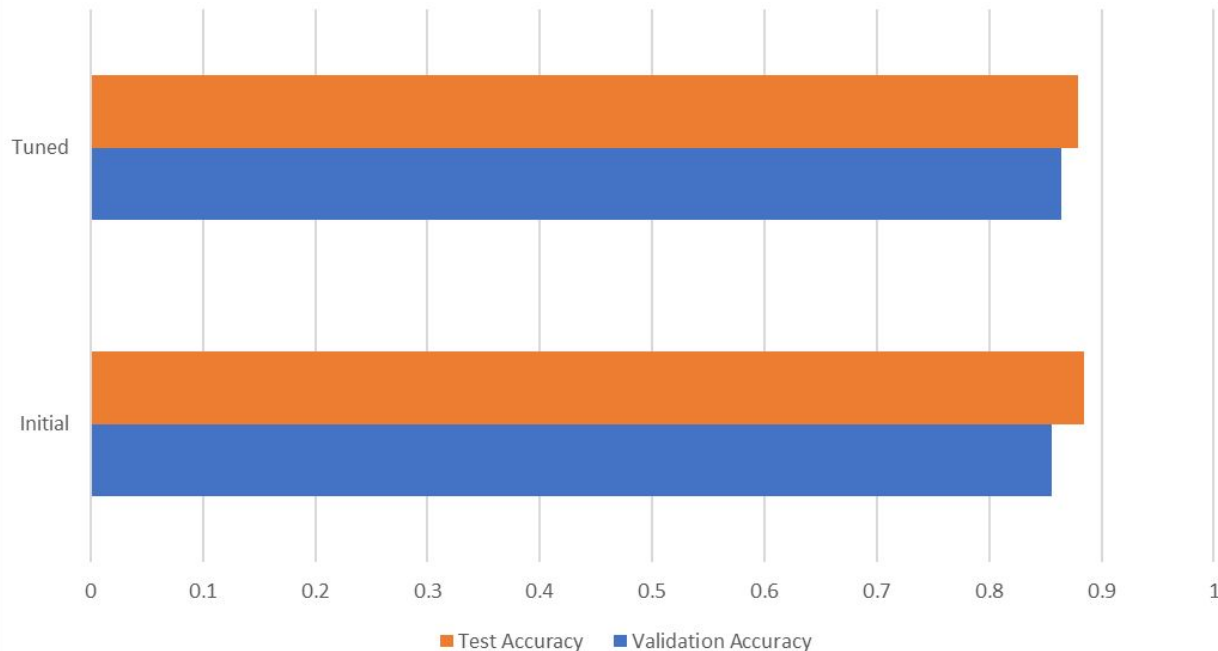
Model vs Process Time (Min)





Model Tuning

Initial vs Tuned (Sequential)



Learning rate:
0.0001, 0.001, 0.01

Number of epochs:
5, 10, 20

RandomSearchCV:
6 iterations



04

App Demo



Hi Doctor King, I noticed these spots on my arm which weren't there before. Should I be worried?

Please upload to the app pictures of your spots, I'll have a look.

<image1.jpeg>
<image2.jpeg>
<image3.jpeg>



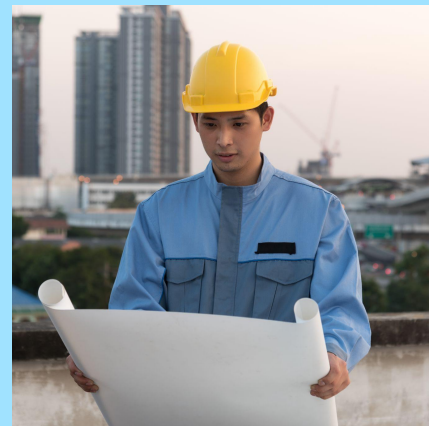
App Demo



Hi Joseph, I am going to refer you to a specialist for further tests.

Oh I see, I didn't want to visit a specialist since I wasn't sure if it was serious, but since you recommended, I guess I will do it now. Thank you!

You're welcome Joseph!
Glad I was able to help.





05

Conclusion

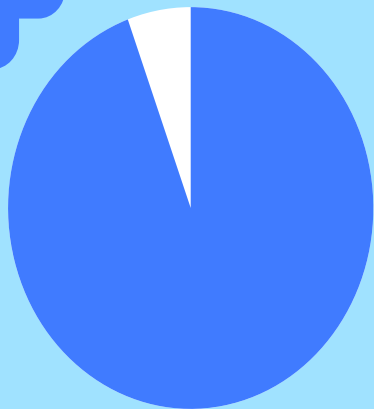
Cost Benefit Analysis

	Visual	Image Detection	Biopsy
Accuracy	60%	89%	95%
Cost	\$100-\$200	\$20-\$30	\$500 - \$3000
Relative Survival Rate*	~95%	~98.7%	~99.5%

SEER stage	5-year relative survival rate
Localized	>99%
Regional	74%
Distant	35%
All SEER stages combined	94%

Conclusion

Skin Cancer Detector



89%



Instant results



Minimal costs



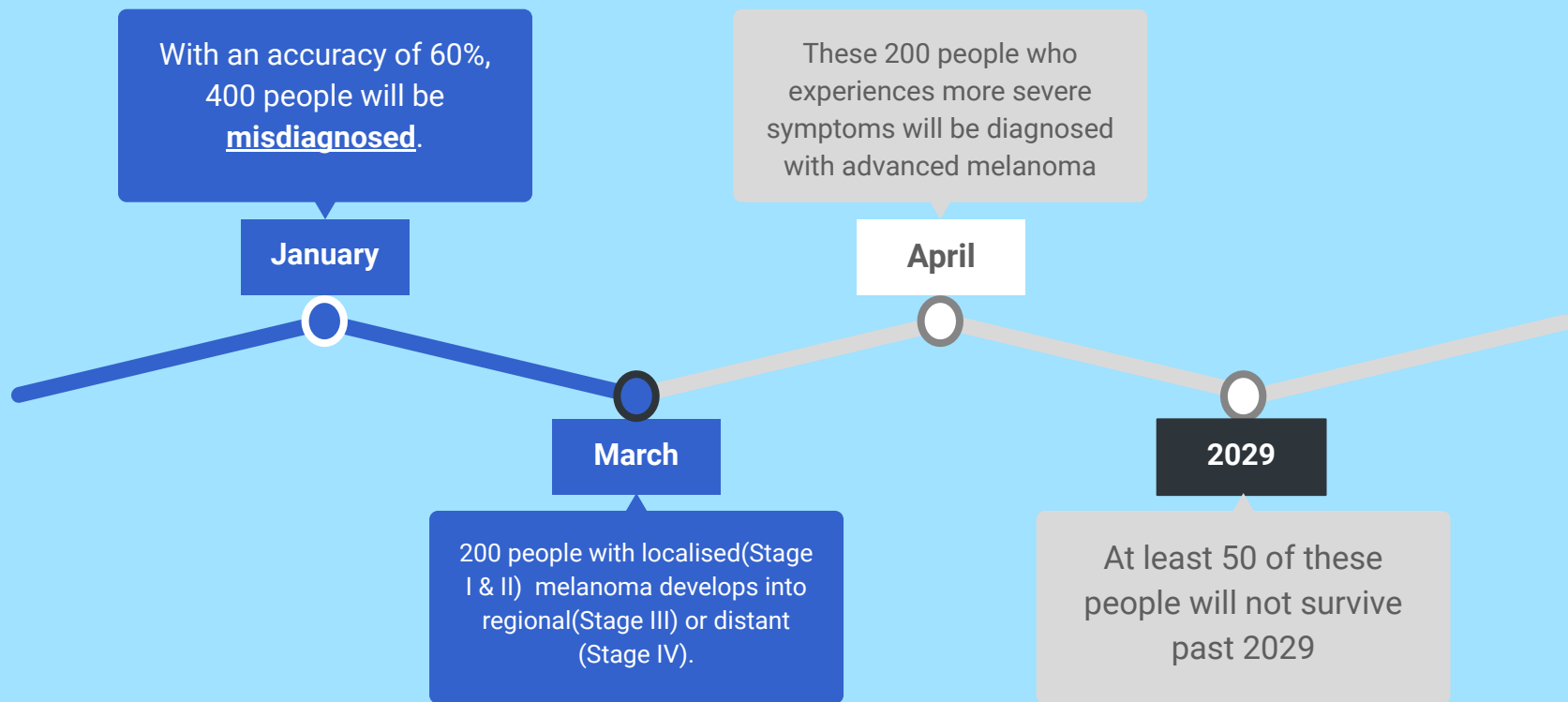
Online accessibility



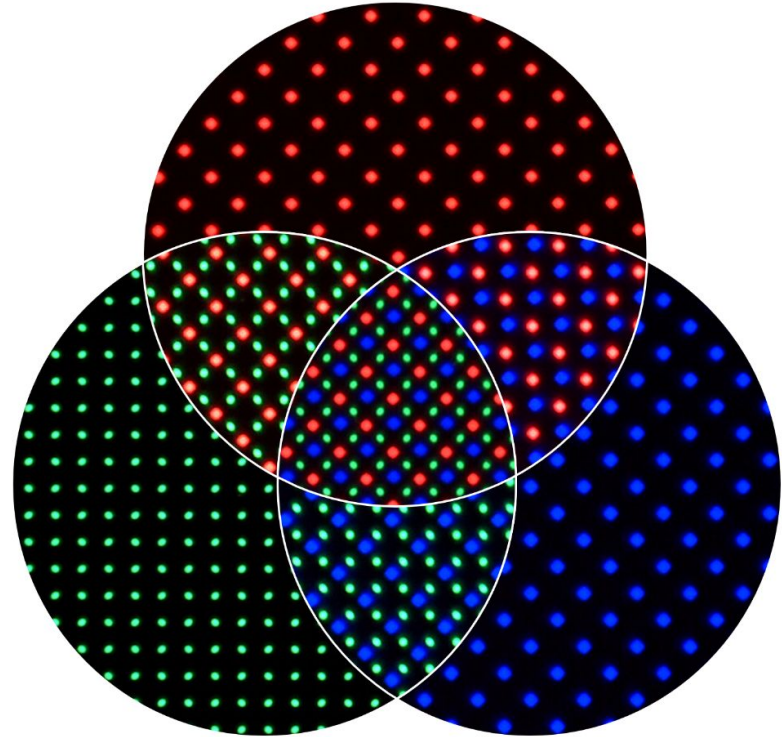
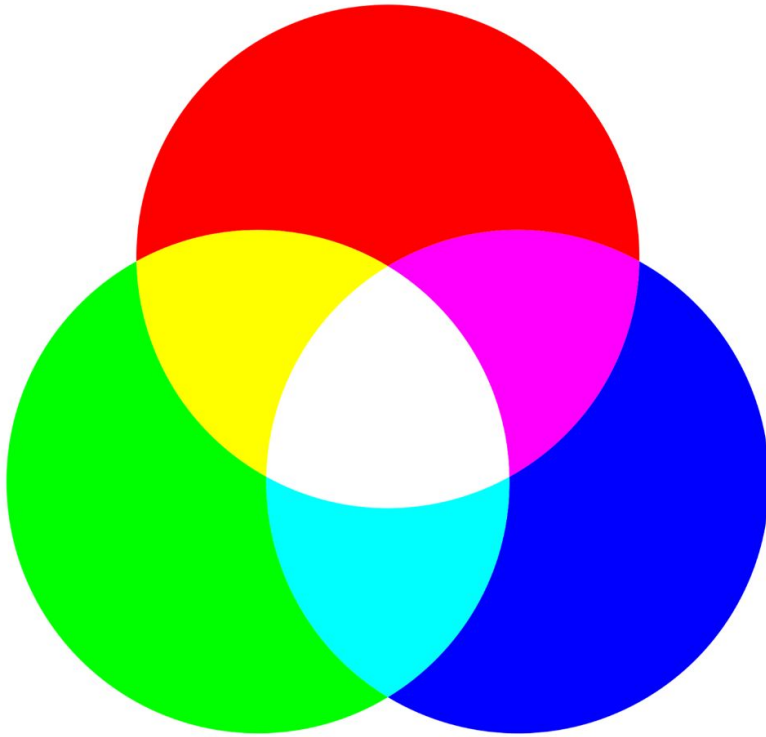
Thank You!!



Calculation of Relative Survival Rate



How does machine see images?



Original Image - Benign



RGB Intensity Histogram



RGB Colour Mixer - Used by Screens

rgb(200,180,180)

Red - Currently set to 200



Green - Currently set to 180



Blue - Currently set to 180



☒ Decimal ☐ Hexadecimal

rgb(255,180,180)

Red - Currently set to 255



Green - Currently set to 180

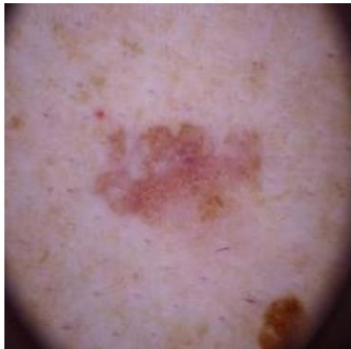


Blue - Currently set to 180

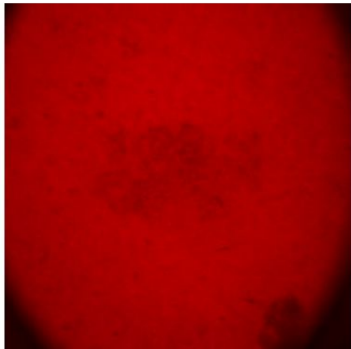




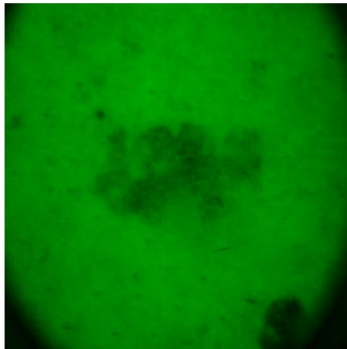
Original Image



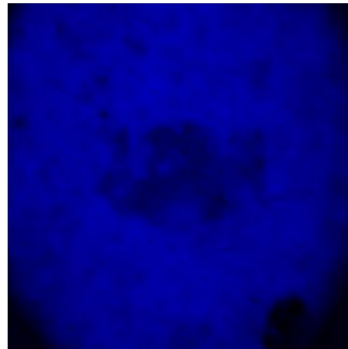
Red Channel



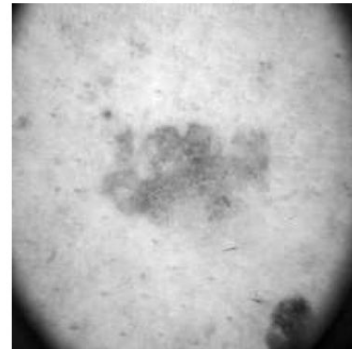
Green Channel



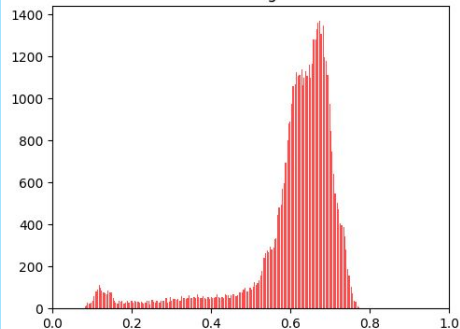
Blue Channel



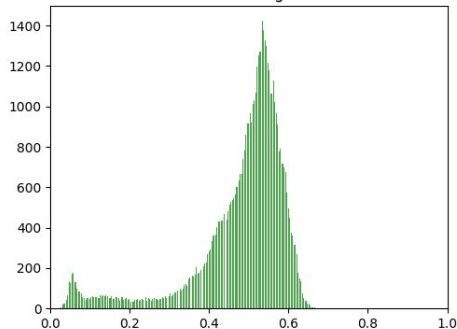
Grayscale Image



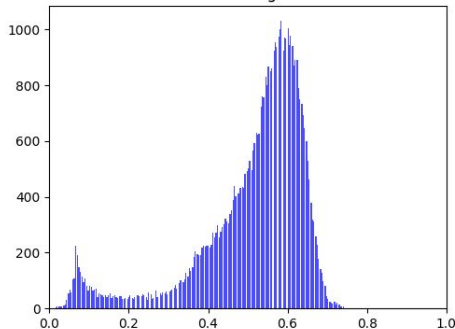
Red Histogram



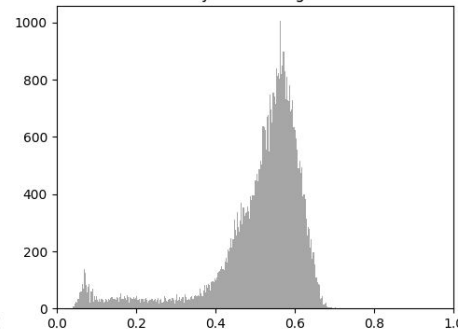
Green Histogram



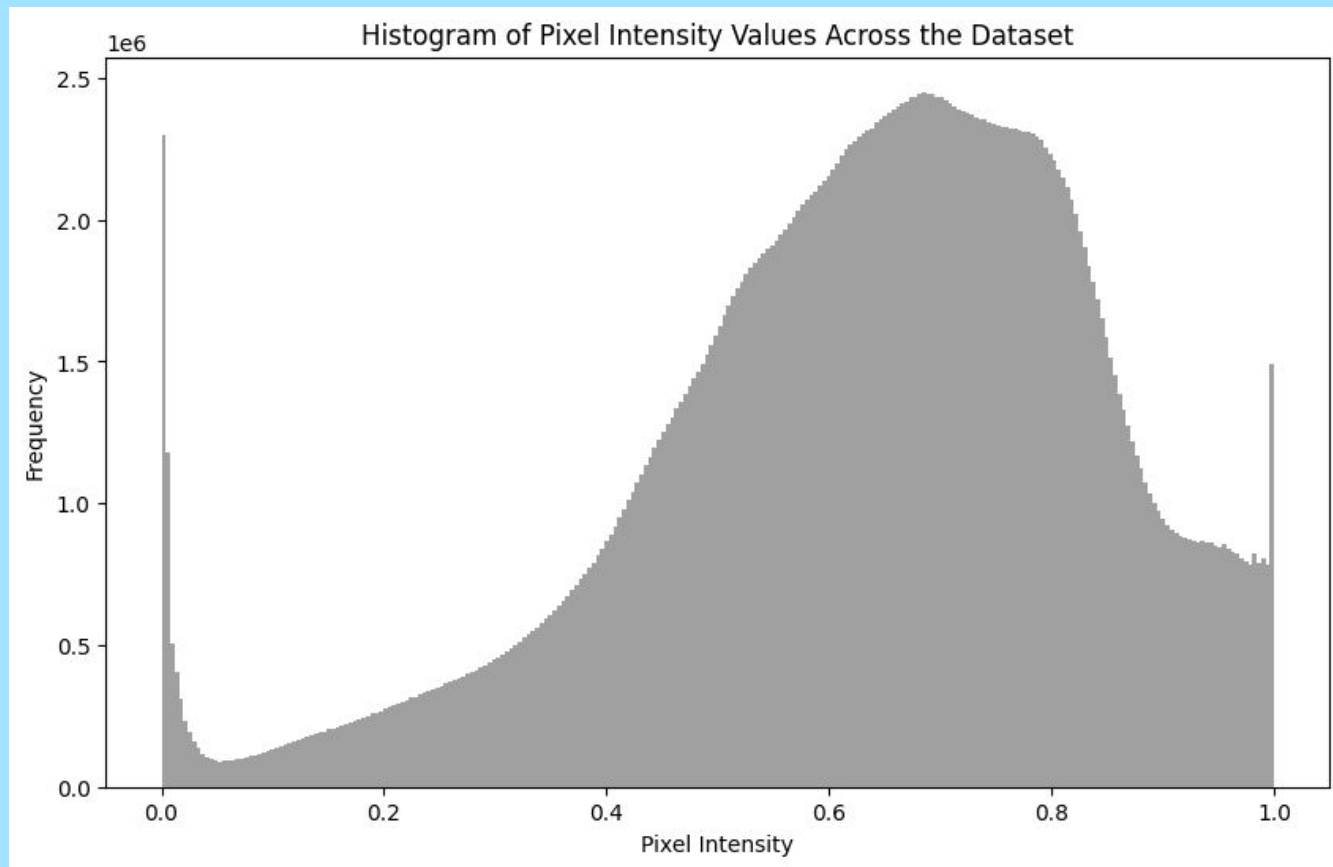
Blue Histogram



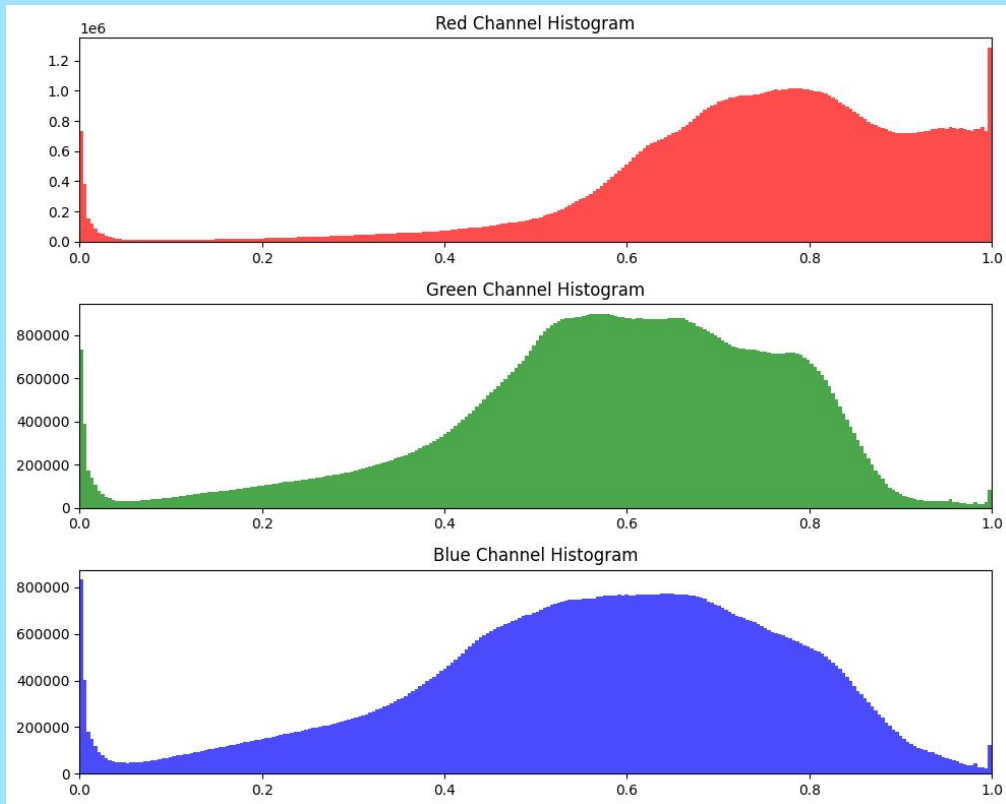
Grayscale Histogram



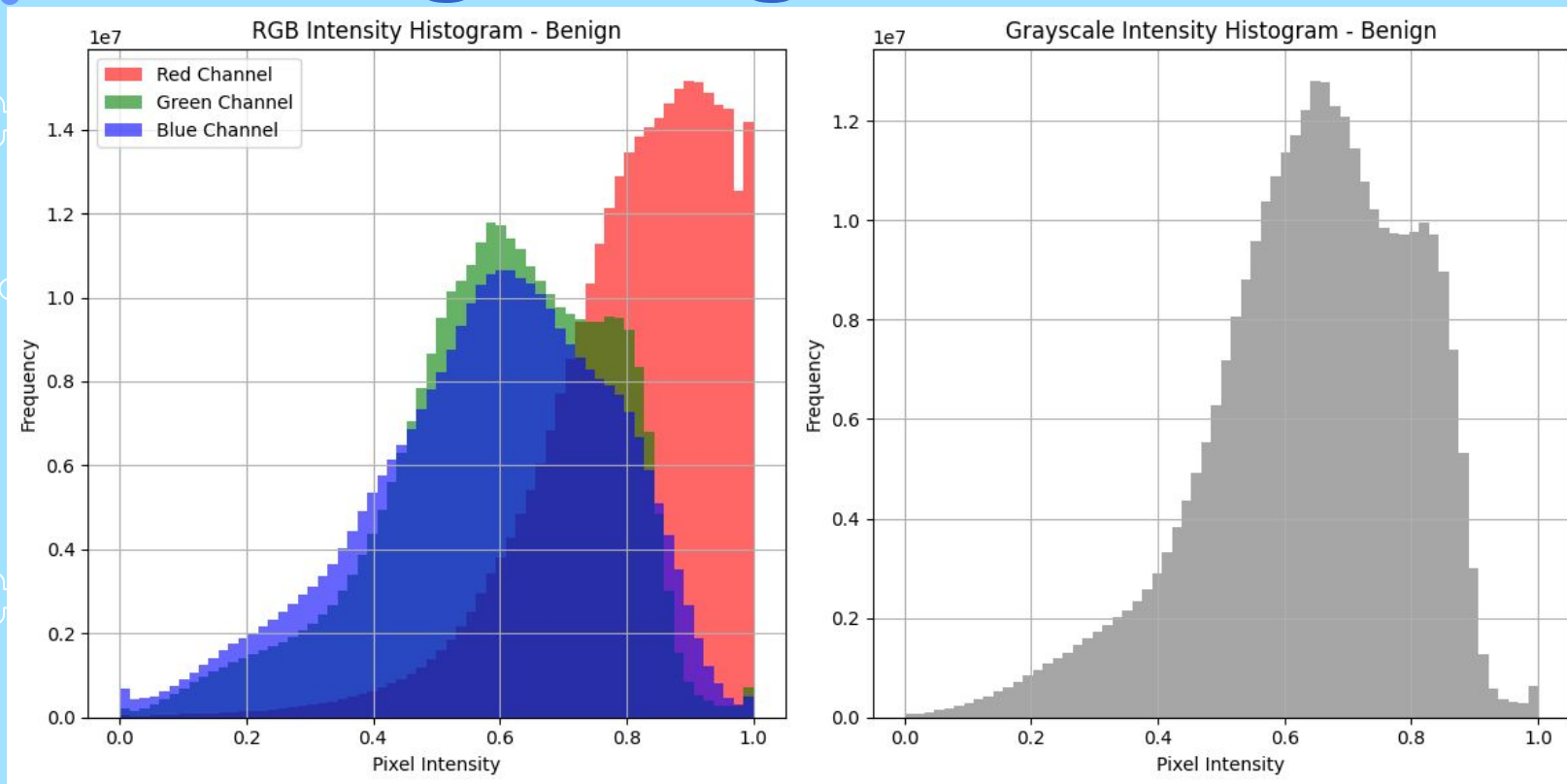
Histogram - Malignant & Benign



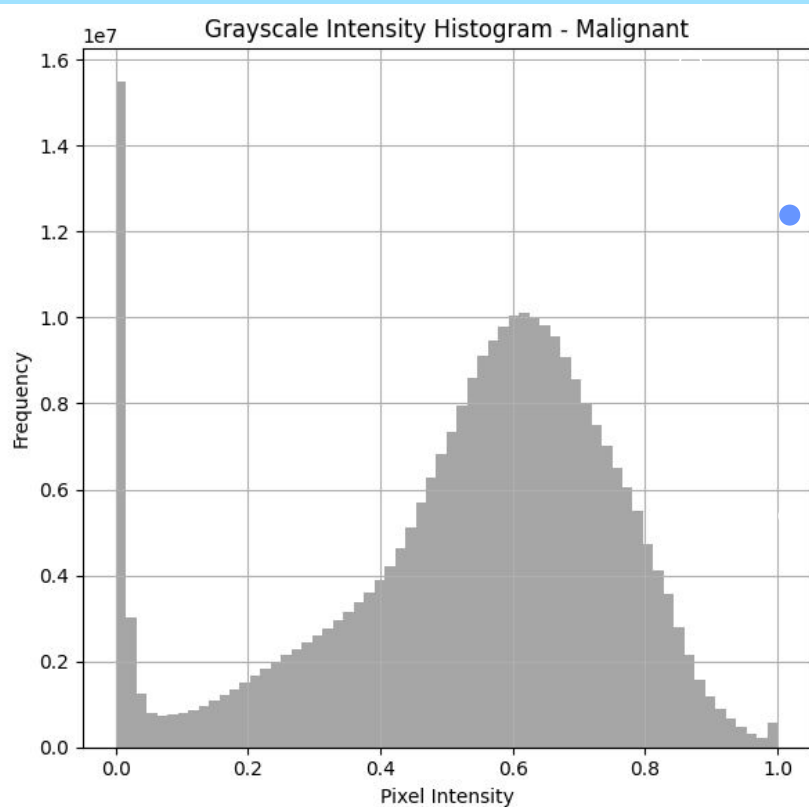
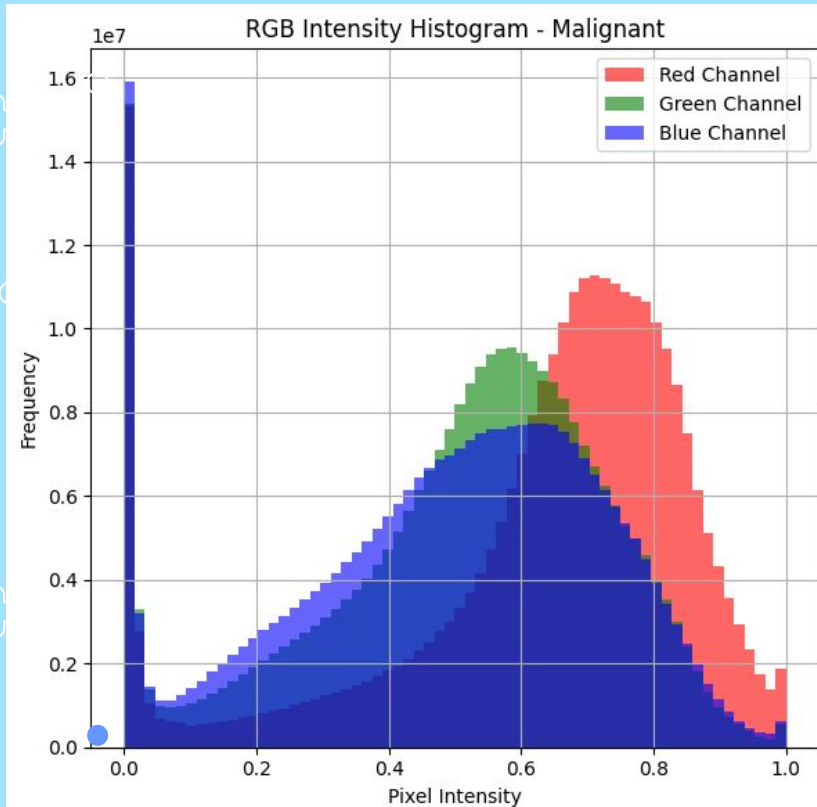
Histogram - Malignant & Benign



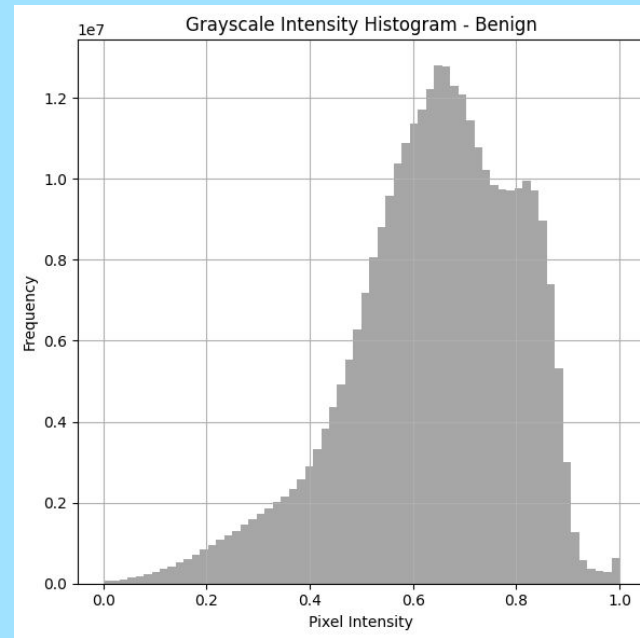
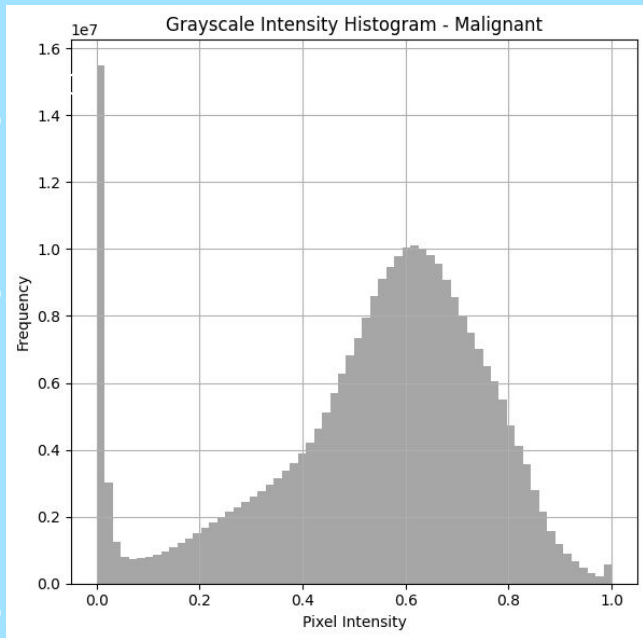
Benign Image - Statistic



Malignant Image - Statistic



Histogram - Malignant vs Benign



“Real men don’t quit
smoking,
Real men fight cancer.”

— WZ

