

# Melanoma cancer detection using image processing

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**Abstract:** Melanoma is the most lethal type of skin cancer in the world. Several attempts have been made to detect melanoma early using deep learning techniques based on dermoscopic images. For an accurate diagnosis of melanoma, it is important to distinguish complex lesion patterns. The typical lesion patterns, on the other hand, are not consistently present, resulting in sparse labelling issues in the results. We suggest a multi-tasking system in this paper

**Keywords:** opencv, feature extraction, image pre-processing, CNN algorithm etc.

## I. INTRODUCTION

With a total area of around 20 square feet, the skin is the body's largest organ. The skin protects us from pathogens and the elements, assists in body temperature regulation, and allows us to feel touch, heat, and cold. Skin has three layers. The epidermis, or outermost layer of skin, establishes our skin tone and serves as a waterproof shield. Tough connective tissue, hair follicles, and sweat glands are found under the epidermis in the dermis. The hypodermis (deeper subcutaneous tissue) is made up of fat and connective tissue. Melanocytes are cells that live in the epidermis' deepest layer, just above the dermis. Melanocytes are the cells that contain the pigment or colour of the skin. Melanoma is a cancerous tumour that develops when healthy melanocytes shift and grow out of control.

## II. OBJECTIVE

The objective of also this project use for melanoma cancer detection using image processing their ultimate objective is to improve DSS's overall decision support capability. The purpose of this paper is to use texture knowledge only for the classification of skin lesions.

## III. RELATED WORK OR LITERATURE SURVEY

[1] "A Study on Melanoma Skin Cancer Detection Techniques"

Author: Ms. Amulya P M

Both supervised and unsupervised classification are performed in the paper by Mhaskeet.al[11]. The supervised learning based classifiers are neural networks and support vector machines, while the unsupervised learning based classifier is the K-means clustering algorithm. The accuracy of the result is compared to the accuracy of these various classifiers. Using Support Vector Machine, they were able to achieve high accuracy. The accuracy of the K-means clustering algorithm is lower than the accuracy of the Neural Network and Support Vector Machine algorithms.

[2] A Method for Melanoma Skin Cancer Detection Using Dermoscopy Images

Author: Soniya Mane, Dr. Swati Shinde.

The dataset's original colour skin image is chosen in this process. An original colour skin image is chosen and transformed to a grayscale image. The skin picture includes several hairs, which will reduce classification accuracy. As a result, hair removal is necessary. The Gaussian filter is used to remove the hair.

[3] A Color-Based Approach for Melanoma Skin Cancer Detection

Author: Shalu, Aman Kamboj

A method for detecting melanoma skin cancer is developed in this paper. To begin, various preprocessing and segmentation techniques were used to enhance the image and extract the region of interest. KNN classifier, Decision

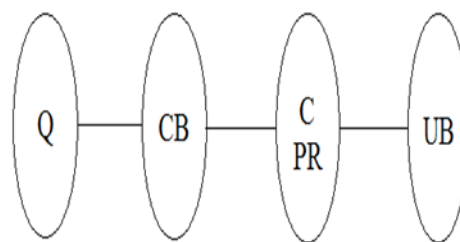
Tree, and Nave Bayes classifier In contrast, the Decision Tree classifier achieves an accuracy of 82.35 percent, which means that it outperforms the other classifiers. Since the evolved system's precision is higher than its sensitivity, it correctly recognises benign cases.

#### [4] LESION ATTRIBUTES SEGMENTATION FOR MELANOMA DETECTIONWITH MULTI-TASK U-NET

Author: Eric Z. Chen, Hongda Jiang

We only used the performance likelihood from the segmentation task for prediction in this paper. The classification task's performance probabili-ties may also be used. This may be a way to boost the efficiency of our proposed model, and it's worth exploring further. In the encoder portion of the U-Net, other pre-trained networks may be used. We also tried the DenseNet that had been pre-trained. However, the final result was worse than with VGG16 as the encoder due to the high GPU memory consumption and slow training speed.

### IV.MATHEMATICAL MODELING



Where,

Q = User entered input

CB = preprocess

C = apply classifier algorithm

PR = preprocess request evaluation

UB = response

#### Set Theory

1) Let S be as system which input image

$S = \{In, P, Op, \Phi\}$

2) Identify Input In as

$In = \{Q\}$

Where,

Q = User entered input (text)

Identify Process P as

$P = \{CB, C, PR\}$

Where,

CB = Preprocess

C = apply classifier algorithm

PR = Preprocess request evaluation

4) Identify Output Op as

$Op = \{UB\}$

Where,

UB = Predict outcome

$\Phi$  =Failures and Success conditions.

#### Failures:

1. Huge database can lead to more time consumption to get the information.
2. Hardware failure.
3. Software failure.

## Success:

1. Search the required information from available in Datasets.
2. User gets result very fast according to their needs.

## Space Complexity:

The space complexity depends on Presentation and visualization of discovered patterns. More the storage of data more is the space complexity.

## Time Complexity:

Check No. of patterns available in the datasets= n

If (n>1) then retrieving of information can be time consuming. So the time complexity of this algorithm is  $O(n^n)$ .

Above mathematical model is NP-Complete

## Algorithm:

### CNN:

A Convolutional Neural Network (ConvNet/CNN) is a Deep Learning algorithm which can take in an input image, assign importance (learnable weights and biases) to various aspects/objects in the image and be able to differentiate one from the other. The pre-processing required in a ConvNet is much lower as compared to other classification algorithms. While in primitive methods filters are hand-engineered, with enough training, ConvNets have the ability to learn these filters/characteristics.

### How CNN works:

- Convolution
- Relu layer
- Pooling
- Fully connected

The convolution of f and g, written as  $f * g$ , is defined as the integral of the product of the two functions after one is reversed and shifted:

$$s(t) = \int x(a)w(t-a)da \quad s(t) = (x * w)(t)$$

Convolution is commutative. Can be viewed as a weighted average operation at every moment (for this w need to be a valid probability density function).

Discrete Convolution (one-axis):

$$s[t] = (x * w)(t) = \sum_{a=-\infty}^{\infty} x[a]w[t-a]$$

Convolution and Cross-Correlation in Images

Convolution operator:  $G=H * F$

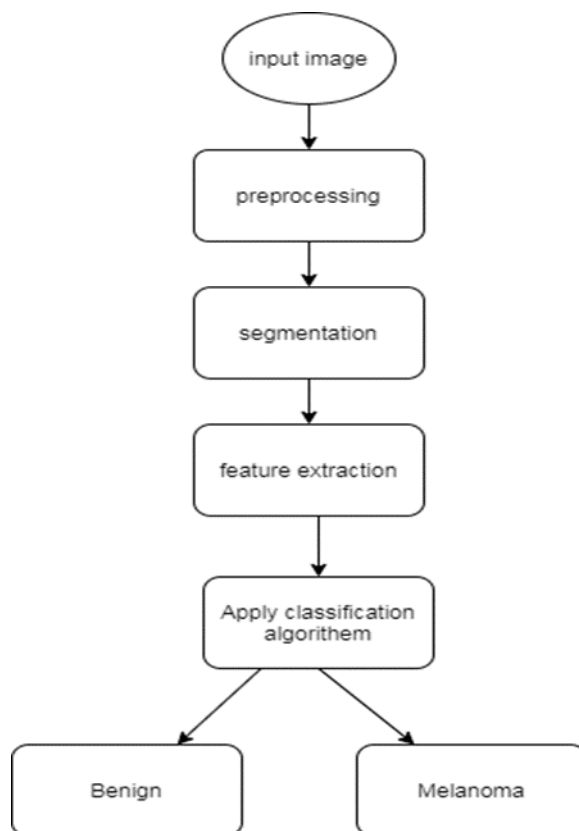
$$G[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k H[u, v] F[i - u, j - v]$$

## V. EXISTING SYSTEM AND DISADVANTAGES

In existing system there is no computerized system to identify the human query. Firstly, it is only suitable for the instance-level approaches that require an instance classifier. As we mentioned before, existing popular approaches of use with neural networks treat separated instances as inputs, then use a deep neural network to transform them into embedding space.

## VI. ADVANCED SYSTEM AND ADVANTAGES

The proposed system would help replicate the customer service experience with one difference that the customer would be interacting with a virtual trial room instead of a real person and yet get the queries attended and resolved.



**Figure: Advance System Architecture**

Step 1: User select the clothes

Step 2: Then apply data pre-processing

Step 3: Then after extract the feature of clothes using cascaded classifier

Step 4: Apply CNN algorithm and train the model

Step 5: Then after load the model and given input model predict the virtual clothes.

**Algorithm****CNN:**

- A Convolutional Neural Network (ConvNet/CNN) is a Deep Learning algorithm which can take in an input image, assign importance (learnable weights and biases) to various aspects/objects in the image and be able to differentiate one from the other.
- CNNs are used for image classification and recognition because of its high accuracy. ... The CNN follows a hierarchical model which works on building a network, like a funnel, and finally gives out a fully-connected layer where all the neurons are connected to each other and the output is processed
- One of the main parts of Neural Networks is Convolutional neural networks (CNN). ... They are made up of neurons with learnable weights and biases. Each specific neuron receives numerous inputs and then takes a weighted sum over them, where it passes it through an activation function and responds back with an output.
- There are three types of layers that make up the CNN which are the convolutional layers, pooling layers, and fully-connected (FC) layers. When these layers are stacked, a CNN architecture will be formed

**Technique:****Image preprocessing**

- Image pre-processing is the name for operations on images at the lowest level of abstraction whose aim is an improvement of the image data that suppress undesired distortions or enhances some image features important for further processing. It does not increase image information content.
- Local illumination can be enhanced using gradient filters, local histogram equalization, and rank filters. Blur and focus enhancements. Many well-known filtering methods for sharpening and blurring may be employed at the pre-processing.

**Advantages:**

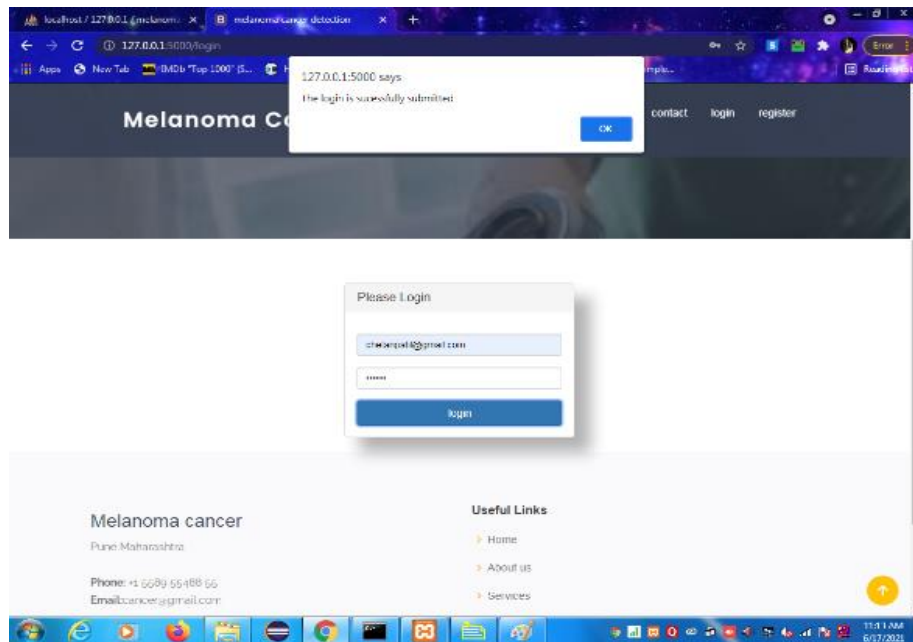
- 1) Secure and efficient system.

**VII. CONCLUSION**

This paper demonstrates an effective tool for detecting skin cancer. This technology uses computer-assisted diagnosis to diagnose skin cancer. Manually detecting skin cancer is a time-consuming and boring process. Biopsy is a traditional technique for detecting skin cancer. This procedure involves scraping a portion of a suspected lesion and sending it to a lab for examination. As a result, this procedure is invasive, painful, and time-consuming. As a result, computer-assisted diagnosis is needed to address the aforementioned issues. The skin image is pre-processed in this method, and then segmentation is performed.

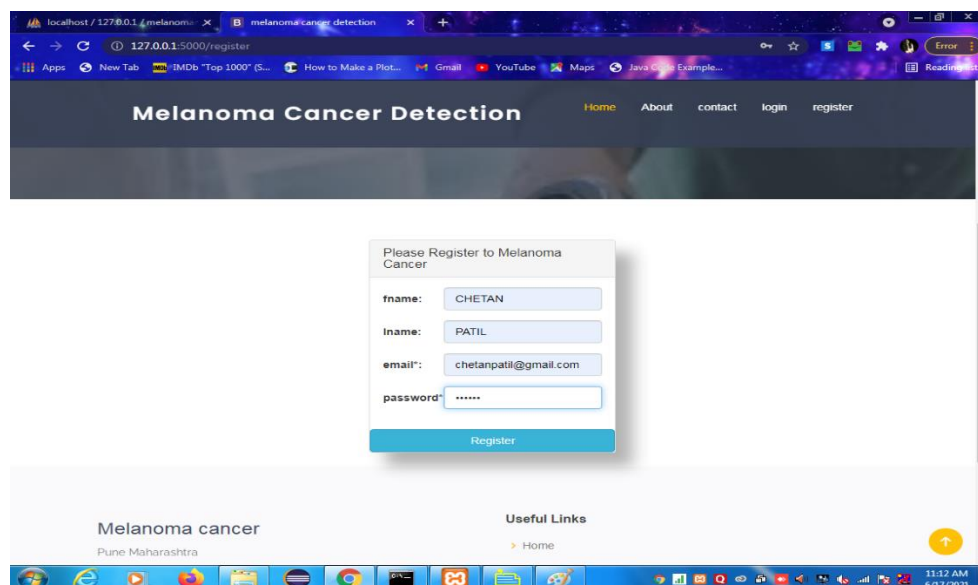
## VIII. SCREENSHOT OUTPUT

### 1. Login:



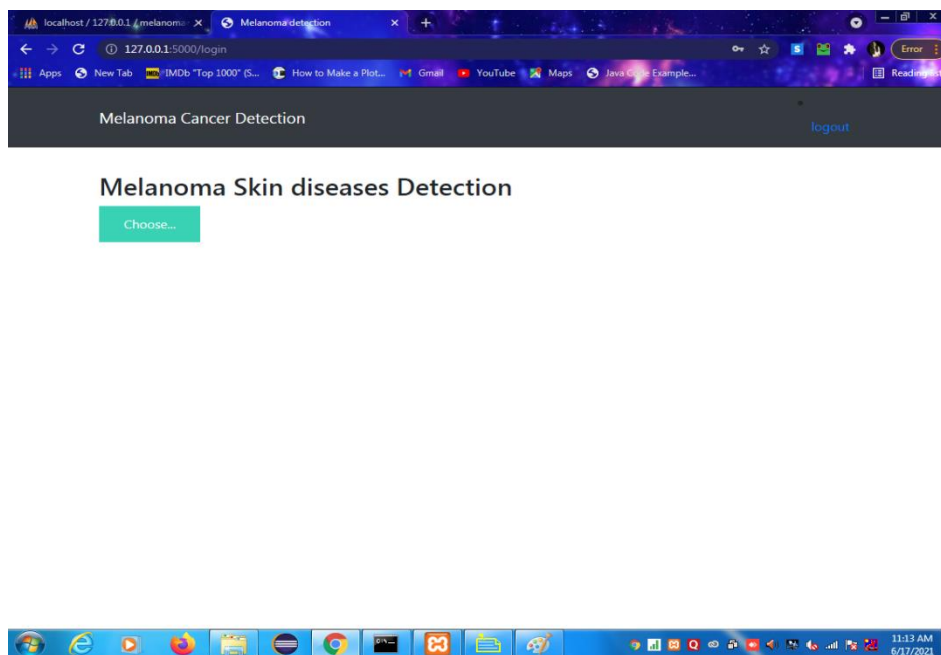
**Fig.** This is the Login and the first page. The user must have to login before any certain operation in the application. The user have to give their login credentials as Username and Password.

### 2. Register:



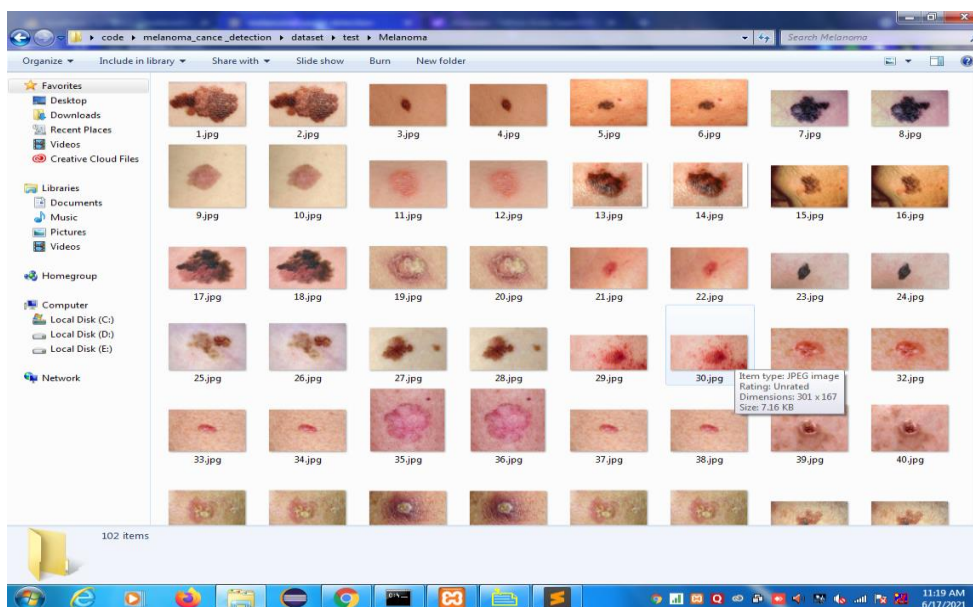
**Fig.** The above picture is register page. If the user is not registered then user have the option to create an new user account to for certain operation. User must have to give its Full Name, In Name i.e. Last Name their Email and create new Password and then simply click on Register Button.

### 3. Dashboard:



**Fig.** After Login/Register user will see the next page i.e. Dashboard. The Dashboard is the main page which enables the user to choose the images from the data set to detect whether its Melanoma or Non-Melanoma.

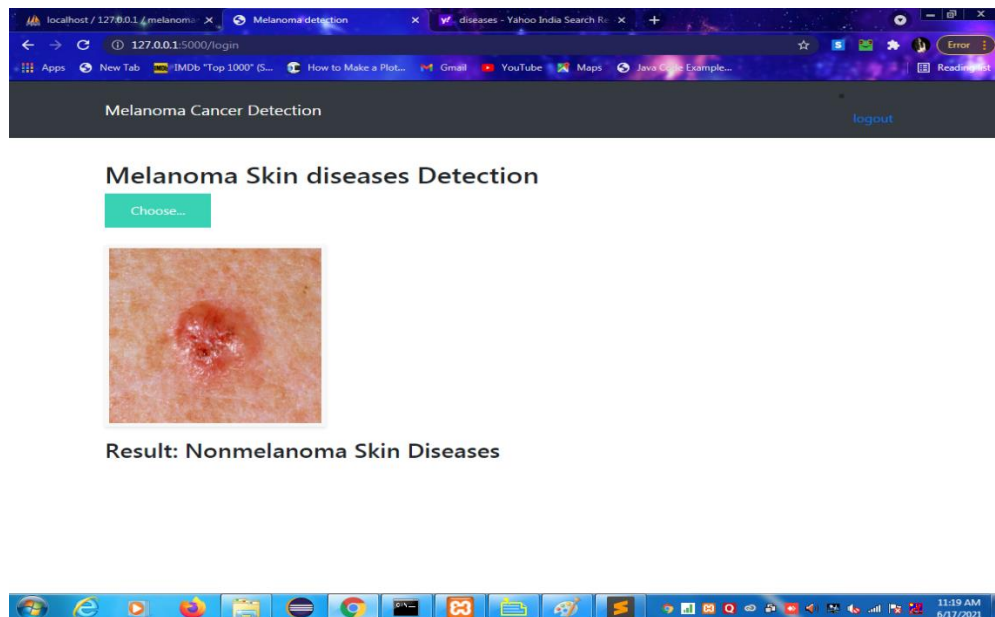
### 4. Dataset:



**Fig.** This is the Dataset of the skin in which user can select the images and detect whether the selected image is Melanoma or Non-Melanoma.

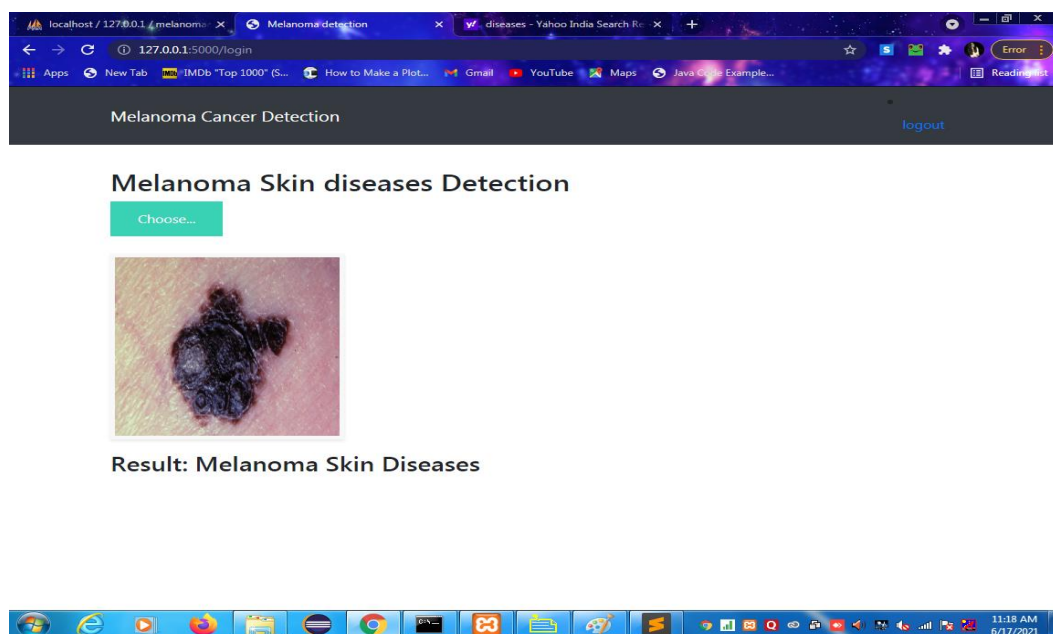


## 5. Output 1:



**Fig.** The above selected image is Non melanoma Skin. The above image is the normal skin infection and not a Melanoma Cancer.

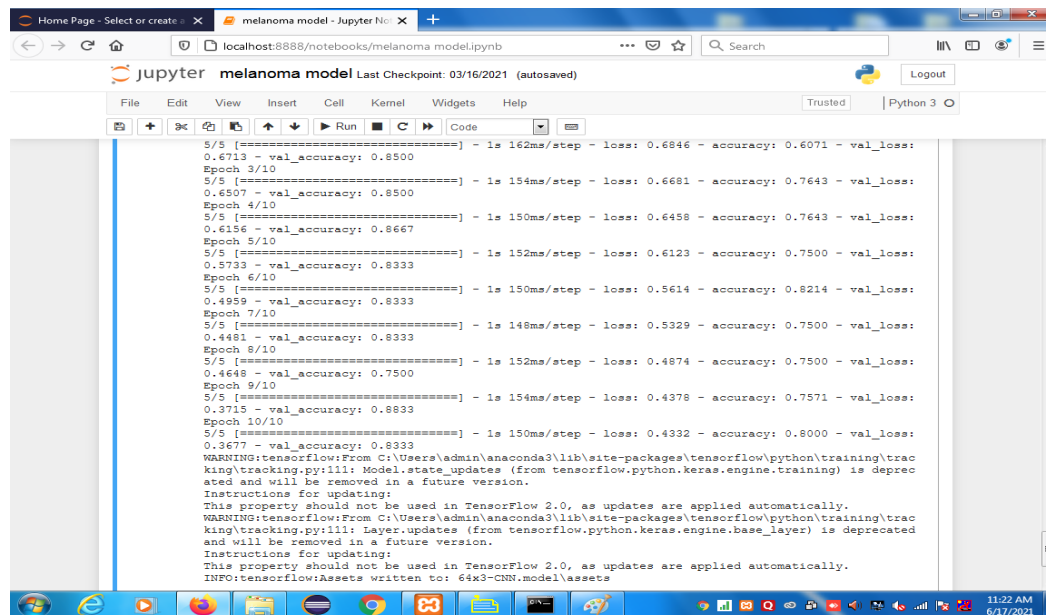
## 6. Output 2



**Fig.** In the above Output image, the selected image from dataset is detected as Melanoma Cancer on the Skin and the patient need an serious medication.



## 7. Accuracy of the result:



```

5/5 [=====] - 1s 162ms/step - loss: 0.6846 - accuracy: 0.6071 - val_loss:
0.6713 - val_accuracy: 0.8500
Epoch 3/10
5/5 [=====] - 1s 154ms/step - loss: 0.6681 - accuracy: 0.7643 - val_loss:
0.6507 - val_accuracy: 0.8500
Epoch 4/10
5/5 [=====] - 1s 150ms/step - loss: 0.6458 - accuracy: 0.7643 - val_loss:
0.6156 - val_accuracy: 0.8667
Epoch 5/10
5/5 [=====] - 1s 152ms/step - loss: 0.6123 - accuracy: 0.7500 - val_loss:
0.5733 - val_accuracy: 0.8333
Epoch 6/10
5/5 [=====] - 1s 150ms/step - loss: 0.5614 - accuracy: 0.8214 - val_loss:
0.4959 - val_accuracy: 0.8333
Epoch 7/10
5/5 [=====] - 1s 148ms/step - loss: 0.5329 - accuracy: 0.7500 - val_loss:
0.4481 - val_accuracy: 0.8333
Epoch 8/10
5/5 [=====] - 1s 152ms/step - loss: 0.4874 - accuracy: 0.7500 - val_loss:
0.4648 - val_accuracy: 0.7500
Epoch 9/10
5/5 [=====] - 1s 154ms/step - loss: 0.4378 - accuracy: 0.7571 - val_loss:
0.3715 - val_accuracy: 0.8833
Epoch 10/10
5/5 [=====] - 1s 150ms/step - loss: 0.4332 - accuracy: 0.8000 - val_loss:
0.3677 - val_accuracy: 0.8333
WARNING:tensorflow:From C:\Users\admin\anaconda3\lib\site-packages\tensorflow\python\taining\trac
king\ttracking.py:111: Model.state_updates (from tensorflow.python.keras.engine.training) is deprec
ated and will be removed in a future version.
Instructions for updating:
This property should not be used in TensorFlow 2.0, as updates are applied automatically.
WARNING:tensorflow:From C:\Users\admin\anaconda3\lib\site-packages\tensorflow\python\taining\trac
king\ttracking.py:111: Layer.updates (from tensorflow.python.keras.engine.base_layer) is deprecate
d and will be removed in a future version.
Instructions for updating:
This property should not be used in TensorFlow 2.0, as updates are applied automatically.
INFO:tensorflow:Assets written to: 64x3-CNN.model\assets
  
```

**Fig.** The image is the result of the Accuracy which shows how the process is going on in the backend.

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