

DETECTION AND CLASSIFICATION OF SKIN DISEASE USING DEEP LEARNING

G.Sasiakala¹, Bollineni AmruthaPriya², Gangavarapu LakshmiPriya³, Kothapalli Samyuktha⁴

¹Assistant Professor, Dept of Computer Science and Engineering, Vivekanandha College of Engineering for Women [Autonomous], Tamil Nadu, India.

^{2,3,4} Students, Dept of Computer Science and Engineering, Vivekanandha College of Engineering For Women [Autonomous], Tamil Nadu, India.

Abstract - The people health further than any other conditions. Skin conditions are substantially caused by fungal infection, bacteria, mislike, or contagions, etc. The spotlights advancement and Photonics grounded medical technology is used in opinion of the skin conditions snappily and directly. The medical accoutrements for similar opinion is limited and most precious. So, Deep literacy ways helps in discovery of skin complaint at an original stage. The point birth plays a crucial part in bracket of skin conditions. The operation of Deep Learning algorithms has reduced the need for mortal labor, similar as homemade point birth and data reconstruction for bracket purpose. A Dataset of 938 images has been taken for the Bracket of Skin conditions. They include Melanoma, Nevus, Seborrheic Keratosis. By using CNN algorithms, 70 delicacy is achieved in bracket of skin complaint. We've also tried with AlexNet, which gives 80 delicacy.

Keywords: Mobile User Authentication; Passwords; Biometrics; Handwriting; PIN; OTP; Touchscreen; Touch Interaction

1. INTRODUCTION

Skin is one of the largest and fastest growing tissue in the human body. Skin diseases are the common health problems in the world. The burden of skin disease is viewed as a multidimensional concept that comprehend psychological, social and financial importance of the skin disease on the patients and their families and also on society. It is the infections that occurring in people among all the ages. Skin is frequently damaged because it is very sensitive part of the body. There are 3000 and more unknown skin diseases. A cosmetically appearance spoiler disorder can have a significant impact, and can cause considerable pain and permanent injury. Most of the chronic skin conditions, such as atopic eczema, psoriasis, vitiligo and leg ulcers, are not immediately lethal, they are recognized as a considerable trouble on health status including physical, emotional and financial outcome. On the other hand, skin cancers, like malignant melanoma, are potentially lethal and their trouble is associated with the temporality that they carry. People of almost 73% are

affected with skin disorder do not seek medical advice. Chronic and several other incurable skin diseases, like psoriasis and eczema, are associated with significant sickness in the form of physical discomfort and impairment of patients life; whereas malignant diseases like malignant melanoma, carry substantial temporality. With the wide range of health status and quality-of-life measures, the effects of most skin diseases on patients lives can be measured efficiently. Along with some of the deep learning algorithms are used for detecting skin diseases in whole body. The convolutional neural network (CNN) is a category of deep learning neural networks. CNN represents a huge advance in image recognition. They are used to analyse the visual images and image classification. A convolutional neural network (CNN) is used to extract features from images. This eliminates the need of manual feature work extraction. The features from the set of images are not trained they are learned while the network trains on a set of images. It makes extreme accuracy for the deep learning models. documents in the training set involvement of the learned features. A particular amount dataset will be provided to detecting the skin diseases.

2. LITERATURE SURVEY

2.1. J. K. Mandal, S. C. Satapathy, M. K. Sanyal, P. P. Sarkar, and A. Mukhopadhyay

Data systems design and smart operations Proceedings of cover transcontinental conference India 2015, volume 1, "Adv.Intell.Syst.Comput.,vol. 339,pp. 301 - 310, 2015, doi10.1007/978-81-322-2250-7.

Due to the plethora of data available moment, text summarization has come truly essential to gain just the right amount of information from huge handbooks. We see long papers in news websites, blogs, guests' review websites, and so on. This review paper presents various approaches to induce summary of huge handbooks. Various papers have been studied for different styles that have been used so far for text summarization. Mainly, the styles described in this paper yield Abstractive (ABS) or Extractive (EXT) summaries of text documents. Query

based summarization ways are also mooted. The paper mainly discusses about the structured predicated and semantic predicated approaches for summarization of the text documents. Various datasets were used to test the summaries produced by these models, analogous as the CNN corpus, DUC2000, single and multiple text documents etc. We have studied these styles and also the tendencies, achievements, formerly work and future compass of them in text summarization as well as other fields. Moment's world is consolidated on computers and data. Data are our impalpable studies and imagination.

2.2 J. K. Kandasamy and P. Koroth

An intertwined approach to spam bracket on Twitter using URL analysis, natural language processing and machine literacy ways, " 2014 IEEE Students' Conf. Electr. Electron. Comput. Sci. SCEECs 2014, pp. 1 – 5, 2014, doi10.1109/SCEECs.2014.6804508. With the increased operation of internet, emails and operation of social media, number of spams is also adding in these areas. Piecemeal from spam on social media, spammers are using spam accounts and targeting druggies on online social media. Whether a stoner access this social media through smartphone or web, he/ she is prone to the spammers on social media websites. This paper analyses some bracket ways that are presently being used in spam filtering in the environment of smartphones and social media. The contents of tweets are unique in nature due to lower content in them so some ways might be effective while some might not be due to the fact that there are some bowdlerizations used in tweets while Roman Urdu tweets are rudiments of English but language of Urdu. Roman Urdu tweets are collected from different metropolises of Pakistan. Some of the most generally used algorithms and ways for spam bracket are bandied and estimated on English and Roman Urdu tweets from Pakistan in this paper. With the increase in the operation of smartphones and internet everyone has access to the online social media websites like twitter, Facebook and numerous further. Considering the elaboration in technologies that are linked with internet, spammers have also changed their means and targeted other media for spamming.

3. EXISTING SYSTEM :

Manual diagnosis of skin disease by visiting dermatologists and consulting them is time- consuming. Most of the rural areas don't have this facility. Such people in rural areas have to travel to a nearby city for consulting and diagnosis. This requires a lot of human-effort and not to mention, that it costs a lot simply for consulting with the doctor. This also involves human contact which is an unnecessary evil during this pandemic crisis. Few diseases are contagious. In the existing system physical contact is unavoidable. The existing computer-aided diagnosis [2] involves identifying burns and injuries as skin diseases.

Existing methods use basic Matlab based image prediction methods. The accuracy of these methods are not as good as needed. Thus, there is a need to develop a computer-aided system that automatically diagnoses the skin disease problem and differentiates skin diseases with other skin issues.

3.1 Disadvantages:

1.Physical Contact: Manual diagnosis involves physical contact with the patient which is not very ideal during the pandemic. Diseases that are contagious can be easily spread.

2.Not Economic: Consulting a doctor before verifying the disease involves waste of money.

Finding dermatologists in remote areas is difficult. People in those areas have to travel to nearby cities for consultation. This also involves waste of money if the affected part is not a disease.

4.PROPOSED SYSTEM

Proposed system is a web application that acts as a preliminary step for the diagnosis of a disease where a person uploads the image of the affected area of the skin and then gets to know the type of the disease and few suggestions are given regarding the disease using this application. The proposed framework involves a deep learning based method [1] [8] to detect skin diseases. This system will utilize computational techniques to analyze, process, and relegate the image data predicated on various features of the images.

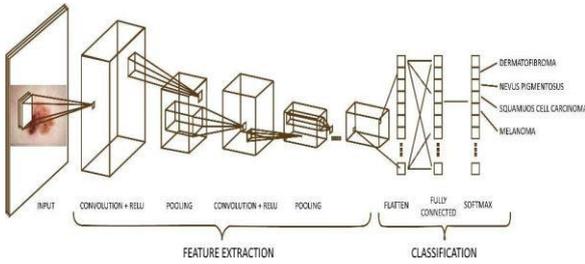
4.1Advantages

Applying these ml and dl model may predict the data with high accuracy in less time for treatment initiating AND Performance of model

5. SYSTEM DESIGN

User should draw each number of the word on the touch screen rather of codifying them as usual. the traditional authentication system are enhanced by incorporating dynamic handwritten biometric information. Our system involves two stages of authentication the drawn leg should be analogous to leg entered during enrollment process. Our alternate stage of authentication involves multiple options grounded on stoner preference where stoner can set multiple set of combinations. Stoner can set alternate stage word as stroke, time, screen brilliance or detector-grounded authentication system. The objectification of biometric information on traditional word- grounded systems can ameliorate the security through a alternate position of stoner authentication.

CNN ARCHITECTURE



6. MODULES

Detection and classification of skin diseases using deep learning consists of three modules

1. Disease Detection
2. Disease classification
3. Results of diagnosis

6.1 Disease Detection

First, an image of the skin's affected area is uploaded. With this uploaded image, it is then checked to see if it is actually a skin disease or not. If it is a disease then further classification will take place, if it is not a disease then a message will be displayed such as "not a skin disease".

6.2 Disease classification

Once the skin disease is detected, it is then further classified, this classification is done by using deep learning techniques. For the accuracy of classification,

INCEPTIONv3, ALEXNET Architecture, RESNET

Architectures of deep learning are used. Once the skin disease is detected, it is then further classified, this classification is done by using deep learning techniques. For the accuracy of classification, INCEPTIONv3, ALEXNET Architecture, RESNET Architectures of deep learning are used.

6.3 Results of Diagnosis

After the diagnosis of the disease is done, the name of the disease and suggestions to cure the disease are mentioned.

```

/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:101: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:496: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:497: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:502: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:502: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:493: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:494: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:495: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:496: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))
/home/user/.local/lib/python3.8/site-packages/tensorflow/python/framework/dtypes.py:497: FutureWarning: Passing (type, 1) or 'type' as a synonym of type is deprecated; in a future version of numpy, it will be understood as (type, (1,)) / '(1,)type'.
  np._dtype = np.dtype(((quiet, 'no-quiet', 1)))

```

6.4 Modules Description:

•**Upload image:** Upload the test image to predict the disease. •**Analysis :** in analysis module first we resize the test image then we convert the test image into grey scale image finally we convert the grey scale image into numerical values like arrays.

•The conversion of image into grey scale is nothing but data pre-processing. We can do that with the help of opencv library.

•We can convert the image into numerical values by using numpy library.

•**Detect disease:** we already train the cnn with dataset. The training process have the following steps.

1.**Data collection:** we have to collect the dataset from Kaggle website.

2.**Data preprocessing:** we have to remove noise from images in this step.

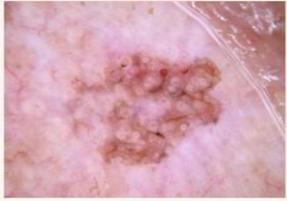
3.**Train & test Split:** we have to split the data into training and testing data.

4.**Model build:** we have to train the CNN algorithm with train data and test with test data to get the accuracy..

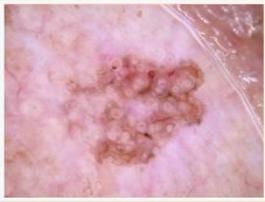
SCREEN SHOTS



Upload!



Click on this to see your Results!



Result: The predicted Disease is Actinic Keratosis - Must undergo Cryotherapy.

CONCLUSION

This work performed experiments using CNN structure for the skin image diagnosis of three common skin diseases and had constructed a dataset consisting mainly of skin disease images. The results demonstrate that CNNs have the ability to recognize and classify skin diseases. Further, our experiments also showed that a reasonable network structure could improve the performance of the model. The performance of the current network structure is used for classification in some diseases, but the overall performance is yet to be improved. As a result, if people want to actually use this technique to check their skin health in their daily life, specialized improvements should be done. In our opinion, with the increasing amount of image data of various skin diseases and the continuous improvement of the network structure, CNN-based skin disease diagnosis algorithms will continue to improve in performance. Apart from CNN and AlexNet, other architecture may also be implemented to improve the accuracy of classification.

REFERENCES

- 1) R.J. Hay, N. E. Johns, H.C. Williams, I. W. Bolliger, R. P. Dellavalle, and D. J. Margolis, "The global burden of skin disease in 2010: An analysis of the prevalence and impact of skin conditions," *J. Investigative Dermatology*, vol. 134, no. 6, pp. 1527-1534, 2014.
- 2) X. Huang, J. Zhang, J. Li, S. Zhao, Y. Xiao, Y. Huang, D. Jing, L. Chen, X. Zhang, J. Su, Y. Kuang, W. Zhu, M. Chen, X. Chen, and M. Shen, "Daily intake of soft drinks and moderate-to-severe acne vulgaris in Chinese Adolescents," *J. Pediatrics*, vol. 204, pp. 256-262, Jan. 2018.
- 3) Y. Deng, Q. Peng, S. Yang, D. Jian, B. Wang, Y. Huang, H. Xie, and J. Li, "The rosacea-specific quality-of-life instrument (RosQol): Revision and validation among Chinese patients," *PLoS ONE*, vol. 13, no. 2, Feb. 2018, Art. no. e0192487
- 4) C. Junchen, W. Zeng, W. Pan, C. Peng, J. Zhang, J. Su, W. Long, H. Zhao, X. Zuo, X. Xie, J. Wu, L. Nie, H.-Y. Zhao, H.-J. Wei, and X. Chen, "Symptoms of systemic lupus erythematosus are diagnosed in leptin transgenic pigs," *PLoS Biol.*, vol. 16, no. 8, Aug. 2018, Art. no. e2005354.
- 5) X. Xiaoyun, H. Chaofei, Z. Weiqi, C. Chen, L. Lixia, L. Queping, P. Cong, Z. Shuang, S. Juan, and C. Xiang, "Possible involvement of F1F0-ATP synthase and intracellular ATP in Keratinocyte differentiation in normal skin and skin lesions," *Sci. Rep.*, vol. 7, Feb. 2017, Art. no. 42672.