

FAKE CURRENCY DETECTION USING CNN

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Abstract - Right now, Fake Currency Acknowledgment System is intended to identify the fake paper cash to check whether it is phony or unique. The current fake issue due to demonetization impacts the financial framework and furthermore in other fields. Another methodology of Convolution Neural Network towards ID of phony money notes through their pictures is inspected right now is similarly better than past picture handling strategies. This technique is in light of Deep Learning, which has seen huge achievement in picture arrangement assignments lately. This method can support the two individuals and machine in recognizing a phony cash note continuously through a picture of the equivalent. The proposed framework can likewise be conveyed as an application in the cell phone which can push the general public to recognize the phony and unique cash notes. The Precision in the proposed framework can be expanded through the high number of unique and phony notes.

Key Words: CNN, Currency, Deep Learning.

1. INTRODUCTION

Acknowledgment of phony Indian cash is very significant in significant areas like financial these days. This framework is utilized to distinguish whether the money is phony or unique through the framework which is through convolution neural system, in profound learning. Profound learning exceeds expectations in the assignment of acknowledgment and order of pictures over a huge informational indexes, which is additionally principally utilized in object class acknowledgment. In the ongoing demonetization drive may be a stage towards destruction of debasement and dark cash, be that as it may, it neglects to address the issue of fake money. A profound neural system is a computational model that works likewise to the neurons in the human mind. Every neuron takes an information, per-frames a few tasks at that point passes its yield to the accompanying neuron that is to its shrouded layer.

2. RELATED WORKS

Concerns have been brought up as of late about the money acknowledgment framework because of increment in fake cash dissemination. Subsequently the target of any cash acknowledgment framework is to locate the phony money. A flourishing way to deal with distinguishing paper cash relies upon a progression of steps, first to catch picture, at that point change to grayscale, identification of edges, division, highlight extraction and picture correlation. Right

now, saw at various kinds of writing that portray various techniques for recognizing fake monetary standards. The archive likewise gives an diagram of strategies to distinguish fake Indian cash for extortion identification. Cash location will work productively by applying some compelling pre-handling and recuperation techniques . One of the best techniques for neutralizing falsifying can be to utilize effectively open what's more, powerful programming to distinguish fake money . A technique for deciding the cash continuously for grouping the cash is proposed. The money of picture obtaining depends on money revelation, when the size of different categories is acquired, and the highlights are removed, it gives the data to Kohonen's preparation arrange. This plan is acceptable with fast and acknowledgment exactness . In help vector machine based cash acknowledgment is introduced. It portions our cash into different areas with our own center, straight weighing through different types of fundamental preparing frameworks furthermore, is considered utilizing semi-characterized writing computer programs, are utilized to acquire perfect loads. Utilization of RGB and HSV shading space model in the picture preparing, helps in money discovery with new 500 and 2000 notes. It very well dependent on picture preparing. The cash picture is spoken to in the space of contrasts, which isa vector space developed by contrasting the picture and a progression of models. Each estimation quantifies the distinction between the picture in question and the model. To recognize two pictures, the nearby key purposes of each picture are recognized and portrayed. In light of money qualities, it is conceivable to successfully recognize the important key focuses between two pictures. Because of the set number of genuine fake monetary standards, SVM is utilized to identify fake monetary forms, so as it were bona fide monetary standards are required for verification and to train classifier. In late years falsifying utilizing neural system is acceptable at identification

3. Proposed Solution

We will fabricate a convolutional neural system as indicated by proposed calculation which will be prepared on the given phony and unique cash informational collection, and later have the option to foresee whether the given cash picture is phony or unique. Right now be illuminating a picture grouping issue, where our objective will be to tell which class the information picture has a place with. The manner in which we will accomplish it is by preparing a counterfeit neural system on picture informational collection of cash and make the NN (Neural

Network) to anticipate which class the picture has a place with, when it sees a picture having counterfeit note or unique note whenever. Convolutional neural systems (CNN's) are these days generally utilized in design acknowledgment and picture acknowledgment issues. They have numerous points of interest contrasted with other procedures. Normally, Convolution neural systems use around 10 particular layers of example acknowledgment. They take crude information, without the requirement for an underlying discrete pre-preparing or highlight extraction arrange: in a CNN, the highlight extraction and grouping happen normally inside a single structure. This is a significant favorable position when looked at to other picture handling procedures, while they need part of calculations only for pre-handling step.

4. Architecture of CNN

To diminish the quantity of parameters in such very profound systems, little 5x5 channels are utilized in all convolutional layers with the convolution walk set to 1. Toward the finish of the net-work are three completely associated layers. The systems utilize different 5x5 convolutional layers to speak to complex highlights.

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Layers =
10x1 Layer array with layers:
 1 " Image Input      431x620x3 images with 'zerocenter' normalization
 2 " Convolution      20 5x5 convolutions with stride [1 1] and padding [0 0]
 3 " ReLU             ReLU
 4 " Max Pooling      2x2 max pooling with stride [2 2] and padding [0 0]
 5 " Convolution      20 5x5 convolutions with stride [1 1] and padding [0 0]
 6 " ReLU             ReLU
 7 " Max Pooling      2x2 max pooling with stride [2 2] and padding [0 0]
 8 " Fully Connected  2 fully connected layer
 9 " Softmax          softmax
10 " Classification Output crossentropy
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5. Pre-processing

The fundamental point behind the information pre-handling is that to enhance the base worth which is the informational index produced. The primary bit of leeway of information pre-preparing is to show signs of improvement preparing set. The fundamental point behind the information pre-handling is that to increase the value of the base worth which is the informational collection created. The primary favorable position of information pre-handling is to show signs of improvement preparing set.

6. Training the CNN:

To prepare the neural system, it is in reality better to begin with an awful performing neural net work and raise the neural system with high exactness. As far as misfortune work, we need our misfortune capacity to be a lot of lower toward the finish of preparing. This shows our neural system has high learning rate and precision. The issue of preparing the system is equal to create the misfortune work with negligible mistake rate. It is significant and even proficient to limit the misfortune since, for reasons unknown, misfortune is a lot simpler capacity to optimize. Even if there are a great deal of calculations that actuation capacities and enhancement capacities, we pick ReLU(Rectifier direct unit) as our enactment work. In the event that we increment our picture informational collection through genuine examples can make the model all the more precisely prepared which may sum up our outcomes past 80% exactness which is a decent sign for expectation of results.

7. Conclusion

The discovery exactness is generally precise since the cash attributes highlights are found out through layer by layer. Here we have considered the entire cash picture, yet in future we will attempt to incorporate all the security highlights of money by utilizing reasonable basic structure and with appropriate preparing information. Further, clamor might be available in the caught picture which must be considered as a pre-handling step in money location process. The acknowledgment and phony money recognition can likewise be reached out by considering the examples of cash surface as highlights for improving the discovery exactness.

References

1. D-F Wang, and S-W Lian, "Automatic selling tickets machine on railway station and cash recognition technology", Computer System Applications, Vol. 7, pp. 12-14, 1999.
2. Z-X Ymg, Z-B Qian and J-G Li. "Currency recognition using mathematical morphology and neural networks", Journal of Shanghai Jiaotong University, Vol. 33, No. 9, pp. 1142-1145, 1999.
3. R Mirza, and V Nanda, "Design and implementation of Indian paper currency authentication system based on feature extraction by edge based segmentation using Sobel operator", International Journal of Engineering Research and Development, Vol. 3, No. 2, pp. 41-46, 2012.
4. J Guo, Y Zhao, and A Cai, "A reliable method for paper currency recognition based on LBP", IEEE International Conference on Network Infrastructure and Digital Content, pp. 359-363, 2010.

5. 5. PJ Grace, and A Sheema, "A survey on fake Indian paper currency identification system", International Journal of Advanced Research in Computer Science and Software Engineering, Vol. 6, No. 7, July 2016.
6. 6. ST Gouri, PK Akshay, M Sneha, and S Bharat, "Detection of fake Indian currency", International Journal of Advance Research, Ideas and Innovations in Technology, Vol. 4, No. 2, pp. 170-176, 2018.
7. 7. EH Zhang, B Jiang, JH Duan, and ZZ Bian, "Research on paper currency recognition by neural networks. In: International conference on machine learning and cybernetics, Vol. 4, pp. 2193-2197, 2003 8. CY Yeh, WP Su, and SJ Lee, "Employing multiple-kernel support vector machines for counterfeit banknote recognition", Applied Soft Computing (Elsevier), Vol. 11, No. 1, pp. 1439-1447, 2011.