

# Machine Learning Based Recognition of Crops Diseases By CNN

K.Naresh, G. Naga Satish, K.BhargavRam, Ch.Srinivasulu

**Abstract:** Machine learning plays a major role from past years in image detection, spam reorganization, normal speech command, product recommendation and medical diagnosis. Present machine learning algorithms used for the identification of crops, their diseases and also provide remedies to them. In the present paper, we discuss the Crops Diseases and remedies. Agro consultant application deals with crops and its diseases, nutritive deficiency indication and providing remedy using machine learning techniques for automated vision system used at agricultural field. The task of identifying plants is time-consuming even for botanists. Real-time identification of crop diseases, nutritive deficiency symptoms, and providing a remedy is an application that will aid users in spotting plant species through their leaves. The system will allow the user to search the database, browse the list of collected leaf samples, and take pictures of leaves and analyze them. Crop diseases are a major hazard to food security, but their swift naming remains not easy in many parts of the world due to the lack of the necessary communications. Using a public dataset of diseased and healthy plant leaves collected under controlled conditions, we train a deep convolutional neural network to identify crop species, their diseases and also provide remedies for diseased crops, weeds, and damaged pest.

**Index Terms:** Machine learning algorithm, Convolution Neural Network

## I. INTRODUCTION

What Is Learning? Vermin learning to avoid lethal Baits: Vermin normally stumble upon provisions items by its look and whiff and start eating in small amounts and later depending on food and physiological effect the feeding of provisions goes on. If the Vermin notices the illness of the provisions, the vermin will not touch that food. Similarly, the machine learning mechanism plays a vital role same as animal usage of experience for acquiring and expertise in detecting food safety. By mistake, if the knowledge with the food is negatively labeled, the prediction of the animal will also be negatively affected and encountered in the future. With the inspiration of the previous illustration of successful learning, we make obvious a typical machine learning algorithm.

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Likewise, we would want to program a machine that learns how to sort out spam e-mails. A credulous result might be comparable to the lifestyle of vermin that, how to keep away from lethal baits. The machine will remember the past e-mails that needed to be been named similarly as spam e-mails by the

individual user. When another email arrives, the machine will look for it in the past set about spam e-mails. Though it matches among them, it will be trashed. Otherwise it will make moving of the user's inbox organizer. At the same time, the first "learning by memorization" methodology may be useful, it fails to offer an imperative aspect known as learning systems – the capacity to mark un-noticed email messages. A fruitful apprentice ought to have the ability which will be the advancement from distinctive samples to more extensive generalization. This may be Likewise, as inductive thinking or inductive induction. In the attraction nervousness exhibited previously, after the vermin experience a sample of a certain sort about food; they apply their disposition at it once new, unseen illustrations from claiming nourishment of comparable emanation also taste. A situated about expressions whose presence for an email message is characteristic of spam. Then, at another email arrives, that machine could weigh if a standout among the suspicious expressions gives the idea on it, and foresee its mark appropriately. Such an arrangement might possibly have the ability effectively to foresee the name about unseen e-mails. Responsibilities further than Human Capabilities: an additional totally crew about errands that profit starting with machine Taking in systems are identified with the Investigation for extremely substantial and intricate information sets: galactic data, turning restorative chronicles under restorative knowledge, climate prediction, and dissection of ordering data, Web search engines, Also electronic trade. With an ever increasing amount accessible digitally traced data, it gets evident that there would riches about serious majority of the data covered clinched alongside information chronicles that would best approach excessively little and also as well perplexing to people with bode well. Taking in with recognizing serious examples over substantial also complex information sets may be a guaranteeing space for which the blending of projects that take for the Just about boundless memory limit and ever-increasing transforming velocity about PCs opens up new horizons. Regulated versus Unsupervised, since taking in includes an association between those learner and the environment, you quit offering on that one might separate taking in assignments as stated by those nature for that connection, to start with qualification will note the Contrast the middle of regulated and unsupervised Taking in. Likewise an illustrative example, think about that errand for Taking in will recognize spam email versus the undertaking aberrance identification. For the spam identification chore, we think about a setting to which the learner receives preparing e-mails to which the mark spam/not-spam may be Gave. On the support from



claiming such preparation the apprentice ought to further bolstering to evaluate a tenet to labeling a recently arriving email message. For contrast, for those assignment about aberrance detection, every last one of learner gets Concerning illustration preparing is an extensive form of email messages (with no labels) and the learner's errand is on identify "unusual" messages.

## II. OBJECTIVES

The chore of identifying plants is protracted even for botanists. Real-time identification of crops and its diseases nutritive deficiency symptoms and providing a remedy is an application that will aid users in spotting plant species through their leaves. The system will allow the user to investigate the database, browse the list of collected leaf samples, and take pictures of leaves and analyse them. Crop diseases are a foremost hazard to food security, but their speedy identification vestiges difficult in many parts of the world due to the deficient of the necessary communications.

## III. IPROPOSED MODEL

CNN architectures vary with the type of the problem at hand. The proposed model is expressed by three convolutional layers, each followed by a max pooling layer. The final layer is fully connected MLP. ReLu activation function is applied to the output of every convolutional layer and fully connected layer. The first convolutional layer filters the input image with 32 kernels of size 3x3. After max pooling is applied, the output is given as an input for the second convolutional layer with 64 kernels of size 4x4. The last convolutional layer has 128 kernels of size 1x1, followed by a fully connected layer of 512 neurons. The output of this layer is given to softmax function, which produces a probability distribution of the four output class. The model is trained using adaptive moment estimation (Adam) with batch size of 100 for 1000 epochs.

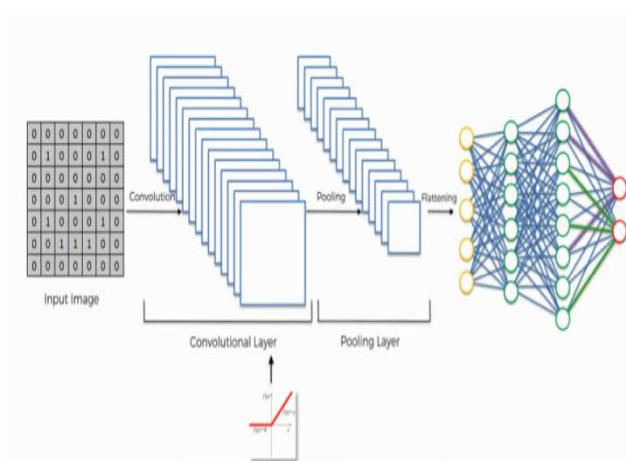


Fig 1: Convolution Neural Network

There are four CNN algorithm steps,

**Convolution:** The phrase convolution refers to the algebraic combination of two functions to produce a third function. It merges two sets of information. In the case of a CNN, the convolution is carrying out on the input data with the use of a filter or kernel to then produce an attribute map. Here are the three rudiments that enter into the convolution operation:

- Input image

- Feature detector
- Feature map

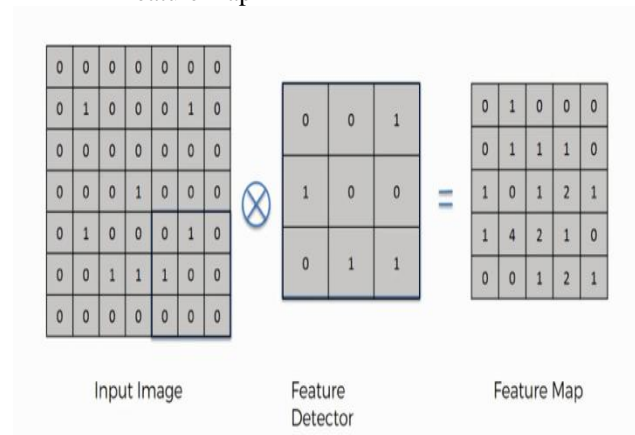


Fig 2 Convolution in CNN

**Max pooling:** Max pooling is a sample-based prejudice process. The intent is to down-sample an input illustration (image, hidden-layer output matrix, etc.), plummeting its dimensionality and permitting for suppositions to be made about attributes contained in the sub-regions binned.

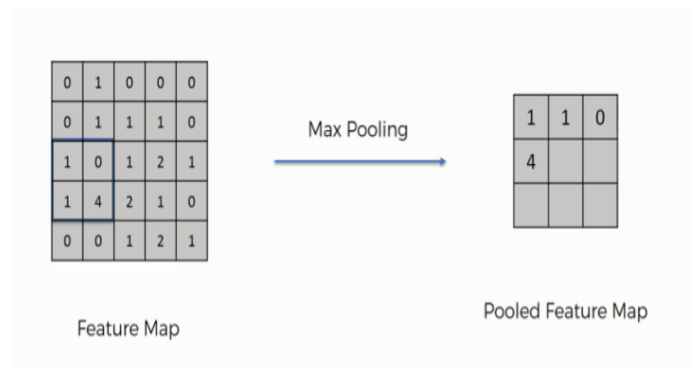
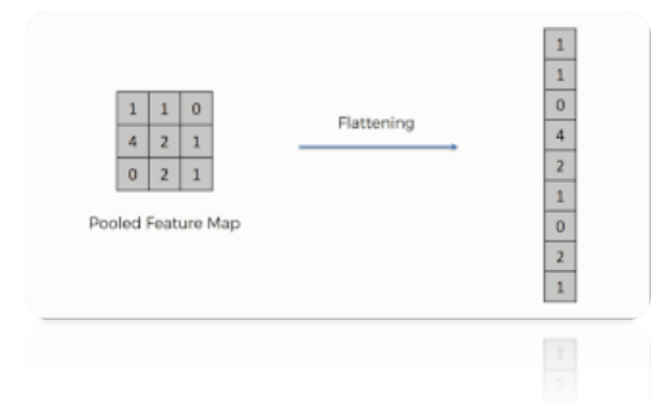


Fig.3 Max Pooling in CNN

**Flattening:** Flattening is the progression of adapting all the ensuing two-dimensional arrays into a single long constant linear vector.



Fi g 4 Flattening in CNN

**Full Connection:** At the end of a CNN, the output of the last Pooling Layer actives as an input to the so-called fully Connected Layer. There can be one or more of these layers ("fully connected" means that every node in the first layer is allied to every node in the second layer).

- Input layer



- Fully-connected layer
- Output layer

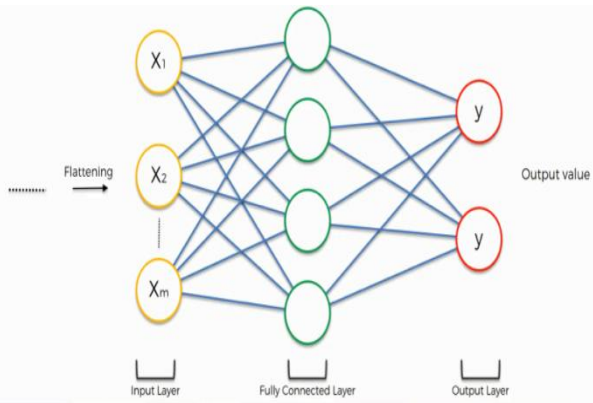


Fig 5. Full Connection in CNN

IV. EXECUTION AND OUTPUTS

Test Case ID	Test Scenario	Expected Result	Actual Result	Pass/Fail
TC01	Check whether application is working after copying the url	Home page should be opened	As Expected	Pass
TC02	Check whether Train option is working on navbar	Train page should be opened	As Expected	Pass
TC03	Check whether Predict option is working on navbar	Predict page should be opened	As Expected	Pass
TC04	Check whether process option is working in predict page	Should get the result below	As Expected	Pass
TC05	Check whether image is getting browsed from the images	Image should get browsed	As Expected	Pass
TC06	Check whether image is uploaded from the images	Image should get uploaded	As Expected	Pass
TC07	Check whether getting the correct result	Should get correct result	As Expected	Pass
TC08	Check whether getting correct leaf name, disease and remedy	Should get correct leaf name, disease and remedy	As Expected	Pass
TC09	Check whether clear option is working	Image should be cleared	As Expected	Pass
TC10	Check whether change option is working	Image should be changed	As Expected	Pass

```

C:\Windows\System32\cmd.exe
Microsoft Windows [Version 10.0.17134.523]
(c) 2018 Microsoft Corporation. All rights reserved.

C:\Users\SAI\Desktop\ImageClassifier>python retrain.py --image_dir=tf_files/plant_images --bottlenecks=tf_files/bottlenecks --output_graph=tf_files/retrained_graph.pb
    
```

Fig 5 Train the Dataset

```

C:\Windows\System32\cmd.exe --img_dir=FileSet\image --bottlenecks=tf_files/bottlenecks --output_graph=FileSet\retrain.pb
INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06550778-8092-4308-8382-0326099F2D52 Com_0_TG_FL_9246_396_Inception_v3.txt
INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06046448-6c3d-4263-824d-8c04b0408044 Com_0_TG_FL_8825_396_Inception_v3.txt
INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06849179-798c-4319-8737-46b791409127 Com_0_TG_FL_8879_396_Inception_v3.txt
INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06652741-8456-4d7f-868c-45c20e206383 Com_0_TG_FL_9923_396_Inception_v3.txt
INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06055051-1307-4d34-7546-476709F51204 Com_0_TG_FL_8078_396_Inception_v3.txt
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INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06240873-8079-4d33-838c-450433340369 Com_0_TG_FL_9889_396_Inception_v3.txt
INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06639352-7687-4646-8417-4eb346649499 Com_0_TG_FL_7985_396_Inception_v3.txt
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INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06120988-604c-4e07-806c-304449974742 Com_0_TG_FL_9732_396_Inception_v3.txt
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INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\06615159-739f-4471-4a42-4e4464618147 Com_0_TG_FL_9933_396_Inception_v3.txt
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INFO:tensorflow:Creating bottleneck at /tmp/hot11meck\Ymazo_Target_Sort\0670e486-8021-4484-9793-80e451512277 Com_0_TG_FL_8135_396_Inception_v3.txt
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```

```

C:\Windows\System32\cmd.exe
Microsoft Windows [Version 10.0.17134.523]
(c) 2018 Microsoft Corporation. All rights reserved.

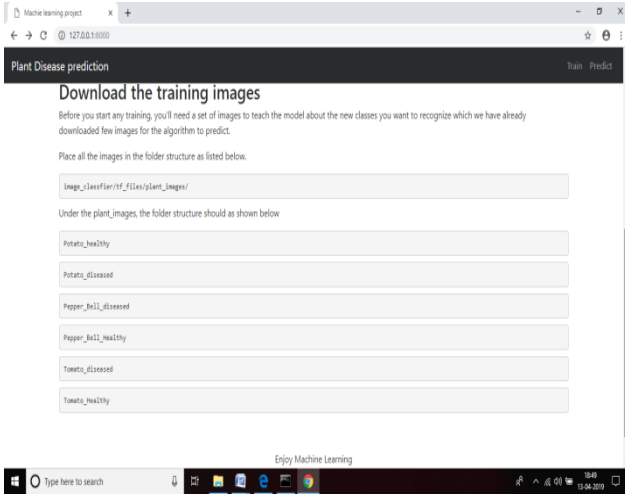
C:\Users\SAI\Desktop\ImageClassifier>python manage.py runserver
    
```

Fig 6: Input

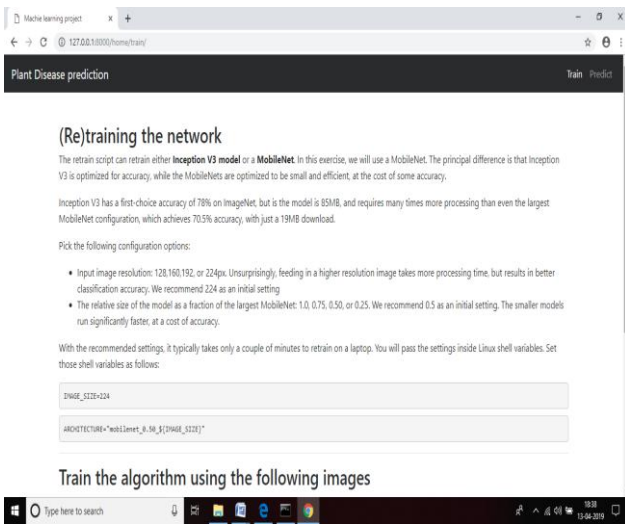
OUTPUT:



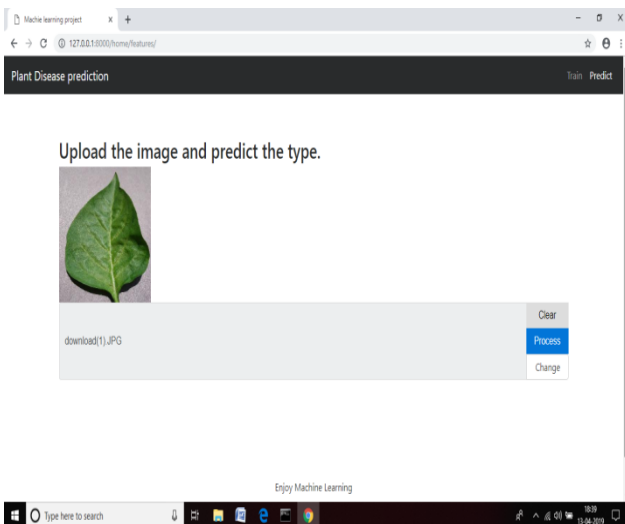




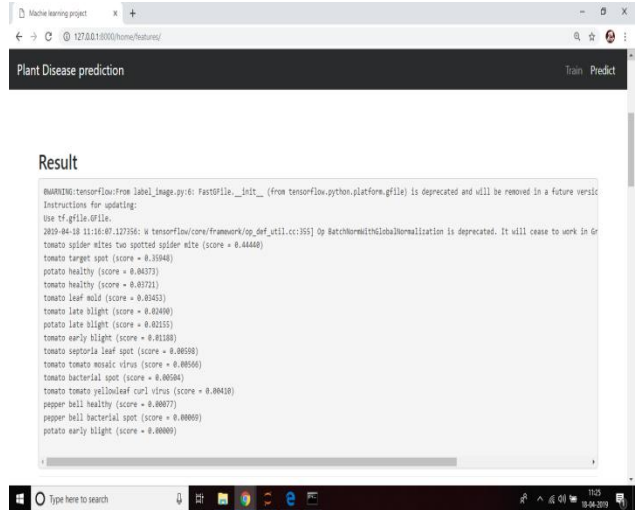
**Fig 7: Test case showing description about the plant disease prediction**



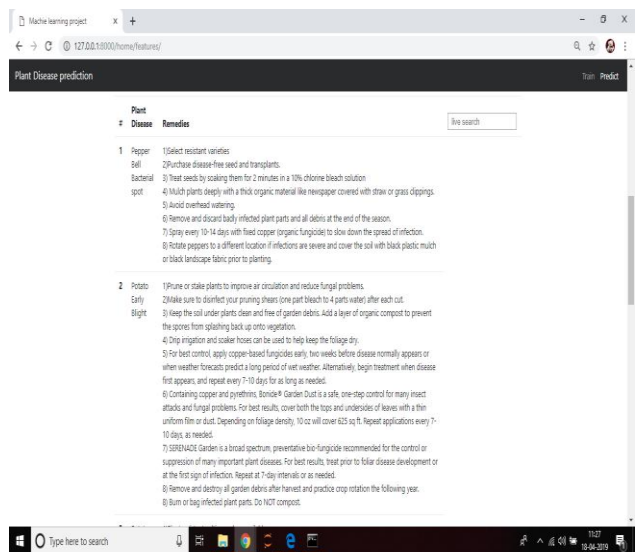
**Fig 8: Test case showing a description of the training**



**Fig 9: Test case uploads the image**



**Fig 10: Test case showing plant name, disease and remedy**



**Fig 11: Test case showing plant name, disease and remedy**

## V. CONCLUSION

The application helps farmers to identify diseased plant. Crop diseases are a major hazard to food security, but their swift discovery remains thorny in many parts of the world due to the not have of the obligatory communications. A data set containing 12,673 leaf images of four classes including healthy leaves are downloaded and trained a deep convolutional neural network to make out diseased plant. We can increase the accurateness and therapy dynamically in prospect.

## REFERENCES

1. T. Weier, C. Stocking, M. Barbour, and T. Rost, Botany “An Introduction to Plant Biology. John Wiley and Sons” 1982.
2. D. Woodland, Contemporary Plant Systematics” Andrews University Press, 2009.
3. “Plant species numbers.” <http://www.bgci.org/ourwork/1521/>. Accessed: 2015-02-25.
4. C. University, U. of Maryland, and S. Institution, “Leafsnap.” <https://itunes.apple.com/us/app/leafsnap/id430649829?mt=8>, 2014. Accessed: 2014-10-15.
5. <https://plantvillage.psu.edu>



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