Object Detection using CNN

Group 6



Road Map

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Objective

Our idea for this project is to build a model that is capable to **detect certain objects** in the images and highlight them with selective color.

Applications:

- •This technique is highly used in **healthcare**.
- **Digital artists** take few hours to color the image but now with deep learning it is possible to color image in few seconds.

Data

We have sourced data from Caltech256 image data set.

- Caltech's image collection procedure was by rating the images good, bad, not applicable.
- 92,652 images were collected from google and pic search, out of which **32.1%(30,608)** were good and kept.
- There are 256 object categories, and at least **80 images** per category.
- To name few categories: Sports equipment, plants, insects, animals, etc.

http://www.vision.caltech.edu/Image Datasets/Caltech256/

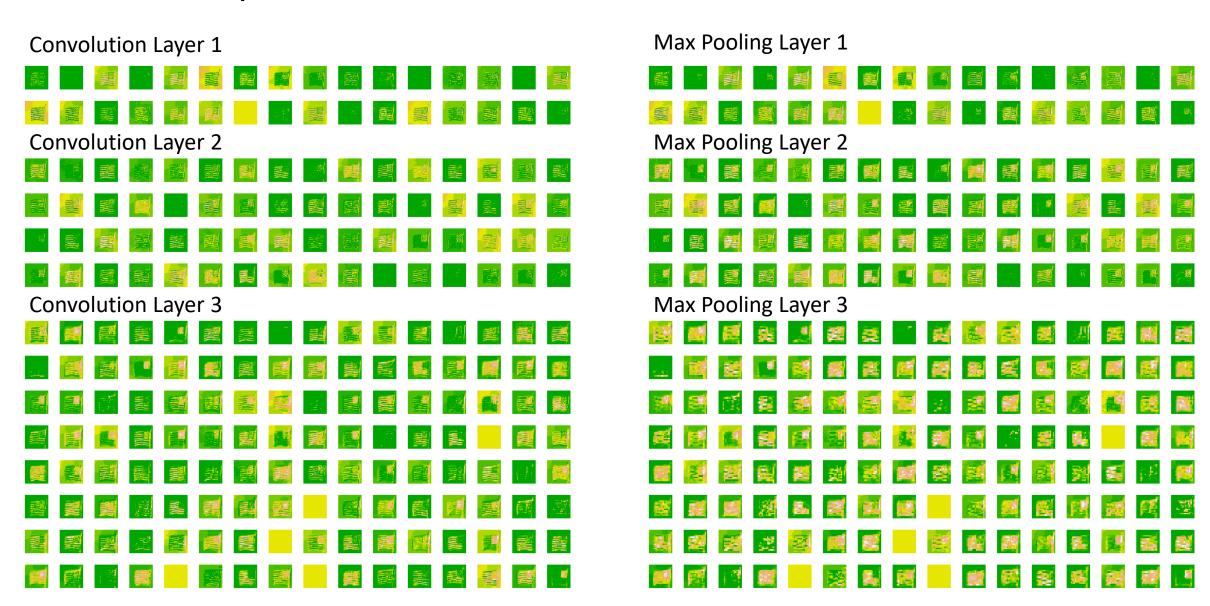
Model Strategy

- Recently Deep Learning has seen many advancements, especially objective detection in images.
- Convolution Neural Network (CNN) are complex feed forward networks.
- CNNs are used for image classification and recognition because of its high accuracy.
- Our model was built for image classification. So, CNN was the best fit.

Technique

Stage 1 Stage 5 Stage 2 Stage 3 Stage 4 Stage 6 Stage 7 Stage 8 Stage 9 Stage 10 Input Layer Convnet Layer 1 Max Pooling Layer 2 Convet Layer 2 Max Pooling Layer 2 Convet Layer 3 Max Pooling Layer 3 Convet Layer 4 Max Pooling Layer 4 Dense Layer Here we are giving input We are introducing the This layer consists of Kernel Size: 3x3 Window Size: 2x2 Kernel Size: 3x3 Window Size: 2x2 Kernel Size: 3x3 Window Size: 2x2 We will convert image shape as 150, 150 input shape to convolution extracting windows from Filters: 64 Strides: 2 Filters: 128 Strides: 2 Filters: 128 Strides: 2 Convolution layer to ,3 i.e height, width, layer with kernel size 3x3. the inpute feature maps Activation: Relu Activation: Relu Activation: Relu Dense Layers. To achieve channel. Since, its in RGB This layer learns local and outputting maximum this we have flatten the patterns of image in small value of each channel. Here, we increase filters layer and used dropout color space. Difference from previous to make network more regularization to prevent 2D window of 3x3 layer is that it down complex and extract more over fitting patterns samples the feature maps feature maps. Since its a classification Filters: 32 by 2x2 task and has two levels, Activation : Relu we used "sigmoid" as Filters: 32 activation function. Activation: Relu

Technique



Model Refinement

- In order to achieve **82.5**% accuracy of the model, we have tuned several hyper-tuning parameters.
 - Every convolution layer we have increased model complexity by incrementing number of filters.
 - We flatten the pooled feature map into a **1D tensor**. By introducing dropout regularization we tried to prevent model from over fitting.
 - In dense layers, we have used **512 nodes** for a hidden layer.

Conclusion

- We successfully built a model to classify the images and corresponding RGB values of these images were stored in array format.
- When we give a B/W image to the model, it will be able to classify the image.
- Due to lack of computing power, we couldn't proceed to our secondary objective, where this model is capable to colorize the given B/W image.

Further Improvements

- Presuming we have enough computing power;
 - We could train this model further with more image data and extract RGB color values for every object in an image.
 - Using this trained models, we can colorize a black and white image or an old distorted images.
- References:
 - https://towardsdatascience.com/colorizing-images-with-a-convolutional-neural-network-3692d71956e2
 - https://towardsdatascience.com/colorizing-old-b-w-photos-and-videos-with-the-help-of-ai-76ba086f15ec