COLORIZATION OF GRAYSCALE IMAGES

Team Name: SkyNet

Area: Computer Generated Art

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1 Problem Statement

Given a grayscale image as input, find the plausible color version of the photograph.

The problem to be solved is basically a fully automatic approach to colorization without any human interference.

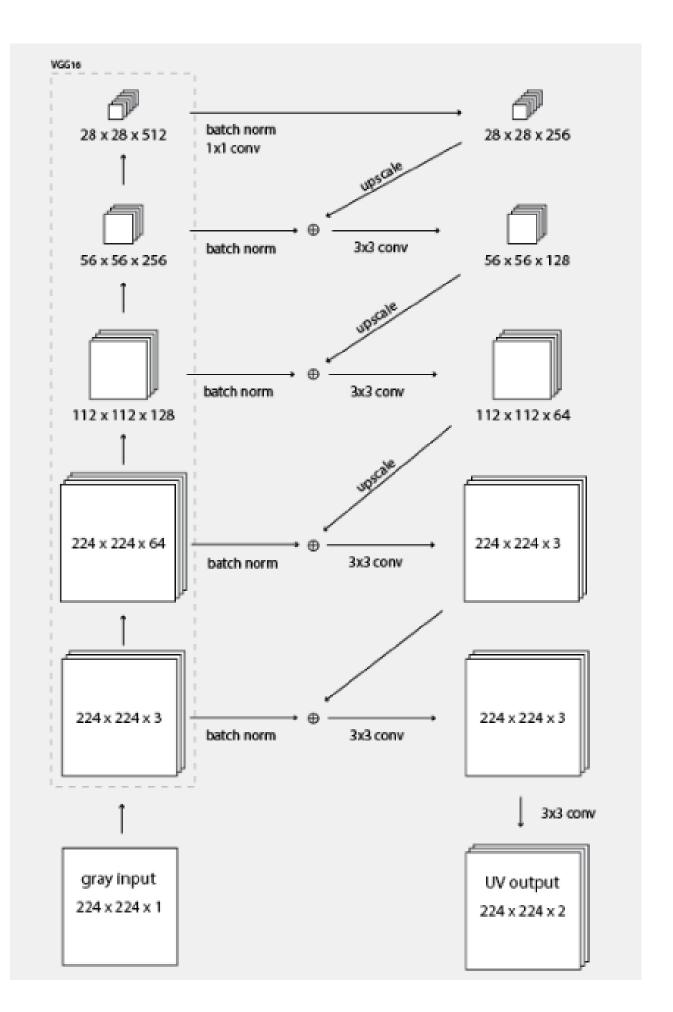
2 Architecture

The VGG-16 convolutional network is a standard architecture used for object detection in images using CNN. However, in our PS, we need to not only detect objects in the grayscale image, but also to associate U and V values with it for colorization. For this, we remove the final class labels of the VGG-16 output (truncated VGG-16). Instead, we have used additional architecture, wherein we have first passed the grayscale image through VGG-16. Then, using the highest layer, we have performed batch normalisation to infer some color, which is upscaled and merged with the batch normalised color information from the next highest layer. Proceeding like this, we have worked our way to the bottom of VGG-16 architecture and obtained the UV output. From there, it's only a matter of linear transform to convert orthonormal basis YUV to RGB color output.

3 Detailed Explanation

In CNN classification models), more info can be extracted than just the final classification. Intermediate layers in classification models can provide useful color information.

Concept of a hypercolumn in CNN is used. A hypercolumn for a pixel in the input image is a vector of all the activations above that pixel. The way this has been implemented was by forwarding an image through the VGG network and then extracting a few layers (specifically the tensors before each of the first 4 max-pooling operations), upscaling them to the original image size, and concatinating them all together.



The resulting hypercolumn tensor has tons of information about what's in that image. Using this information I should be able to color the image.

The residual encoder model (shown on page 2) has been used. Its almost an autoencoder but from black and white to color, with residual connections. The model forwards a grayscale image through VGG16 and then using the highest layer infers some color information. Then it upscales the color guess and adds in information from the next highest layer, and so on working down to the bottom of the VGG16 until there is $224 \times 224 \times 3$ tensor. We used residual connections to add in information as it works its way down the VGG16.

4 Change in output images with iterations

Intermediate outputs were obtained while training the model.

The input to the model is the left-side grayscale image. The output is the middle image. The right image is the true color.

1. Iteration = 1000



2. Iteration = 6000



3. Iteration = 22000



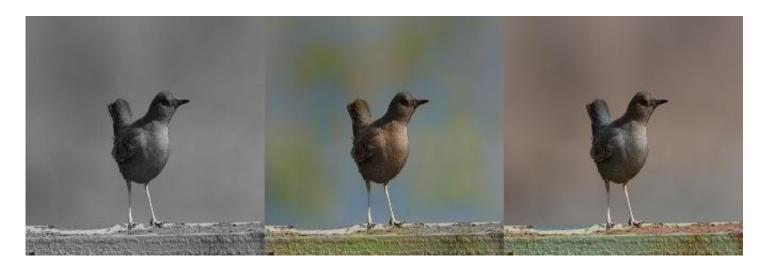
4. Iteration = 38000

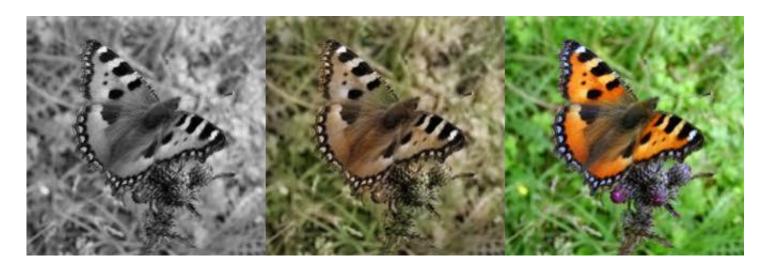


Ws we can see the model is able to identify features more sharply and add more colors to the image (it learns more number of colors) as it goes through successive iterations (as number of iterations increase).

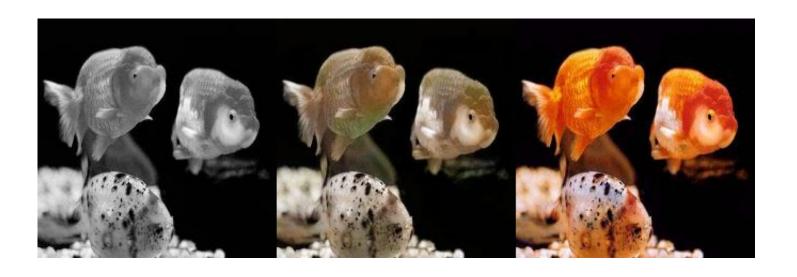
5 Results

The input to the model is the left-side grayscale image. The output is the middle image. The right image is the true color which the model never gets to see























6 Relevant Links

Team: https://github.com/orgs/cs60050/teams/skynet

Repository: https://github.com/cs60050/SkyNet

Website: https://cs60050.github.io/SkyNet/website/