Image Super-Resolution using a Multi Layer CNN

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Overview

- Our project aims to take compressed or small resolution images and expand them to a larger size, and enhance or restore full size quality.
- A possible application would be to send images at a lower resolution and allow the recipient to be able to view a full quality image.
- Another application is that images taken with a lower quality camera could be improved
 - Security footage (License Plate Readers)
 - Older phones
 - Images with too high of an exposure

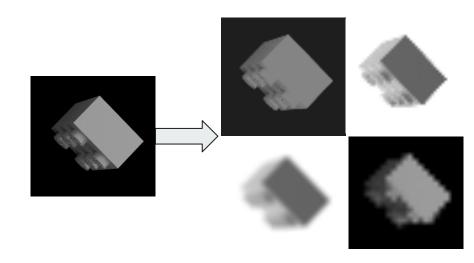
Our Dataset

- We found a dataset of 40,000 unique images of LEGO bricks rendered in blender
 - 50 Unique LEGO bricks, each rotated in 800 different orientations
- We chose this dataset because it contains a lot of simple shapes with very sharp edges and curves
- When compressed the image will lose the sharp edges, which we aimed to teach the model how to recreate

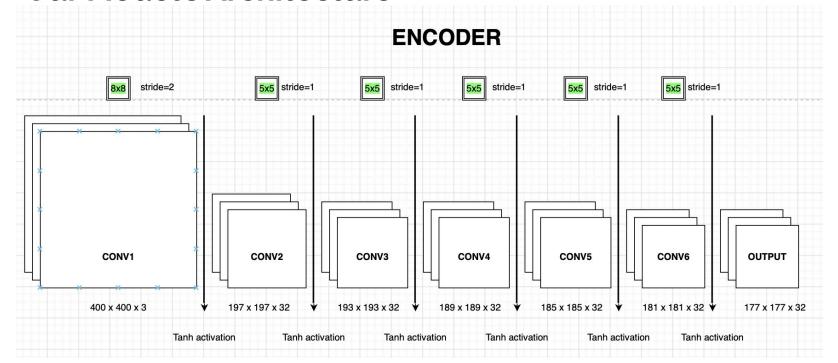


Image Preprocessing

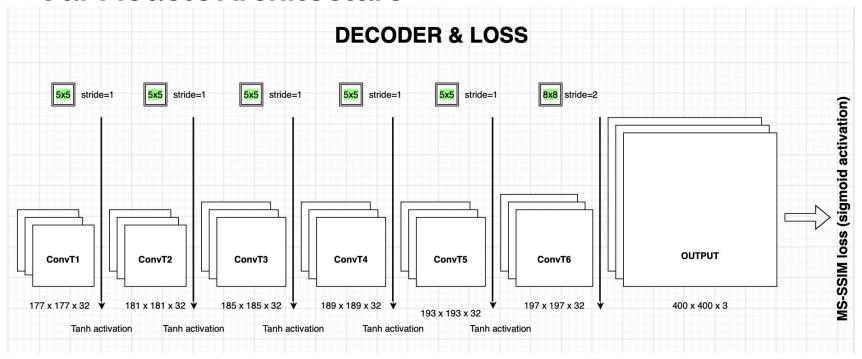
- To create the input for our model, we applied 1 of 4 different transformations to the image
 - Apply a gaussian blur
 - Scale image down 12x then back up, then do one of the following:
 - Lighten the image to remove contrast
 - Invert colors
 - Do nothing



Our Model's Architecture



Our Model's Architecture

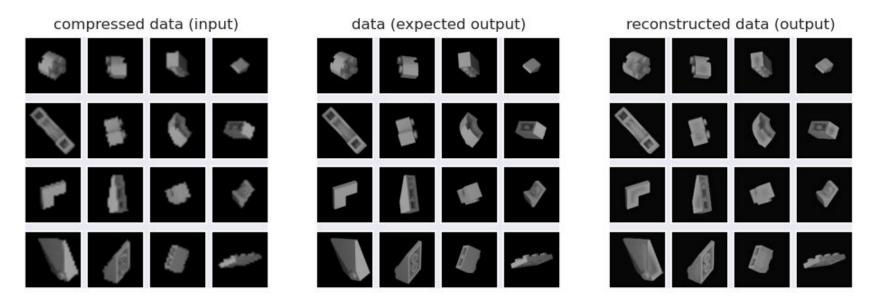


Loss Function

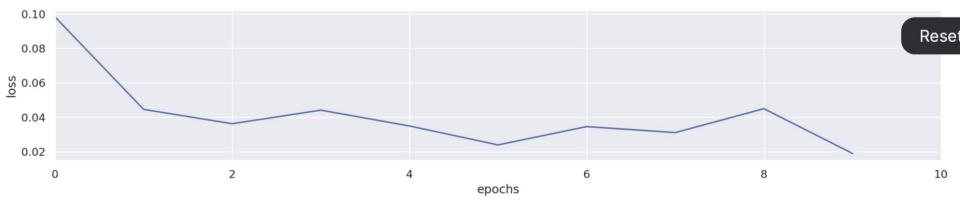
- We tried multiple different loss functions, and found Multi Scale-SSIM performed the best for our model
 - MSE was able to create a gray blob in the shape of the piece, but was unable to create any detail within the piece
- SSIM is the comparison of of luminance, contrast, and structure in a limited window for both images
- Using Multi Scale-SSIM, it performs the SSIM calculations on multiple differently sized windows.

$$ext{SSIM}(x,y) = rac{(2\mu_x \mu_y + c_1)(2\sigma_{xy} + c_2)}{(\mu_x^2 + \mu_y^2 + c_1)(\sigma_x^2 + \sigma_y^2 + c_2)}$$

Results



Results (cont.)



Running the Model on Non-LEGO Images





