





ADRIAN RAMLAL



ANIRUDDHA REDKAR



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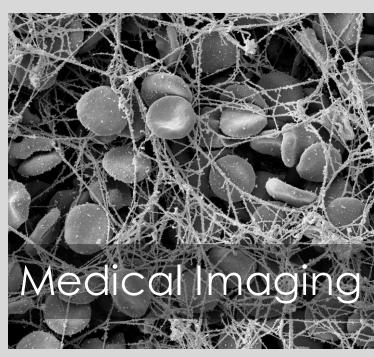
SIDDHARTH VIJAY

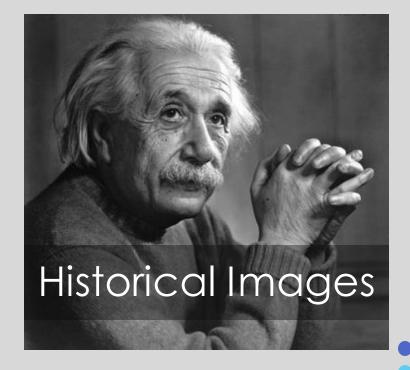
# THE PROBLEM

## We Are Limited By Grayscale Images





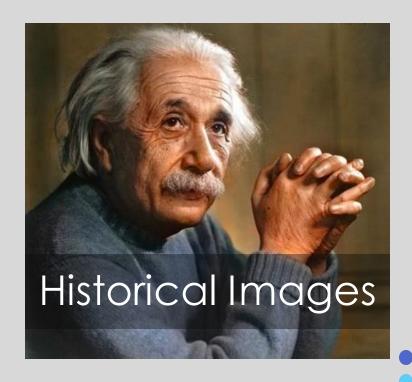




## We Are Limited By Grayscale Images

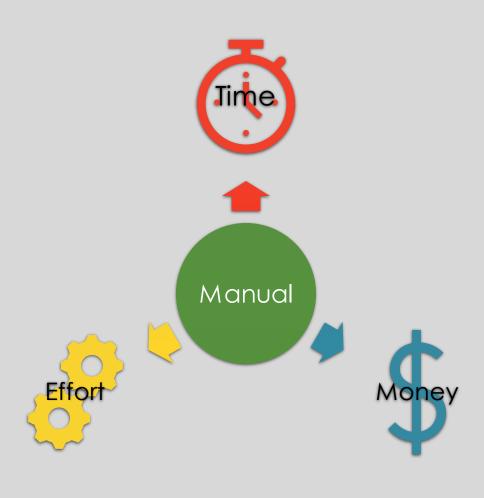


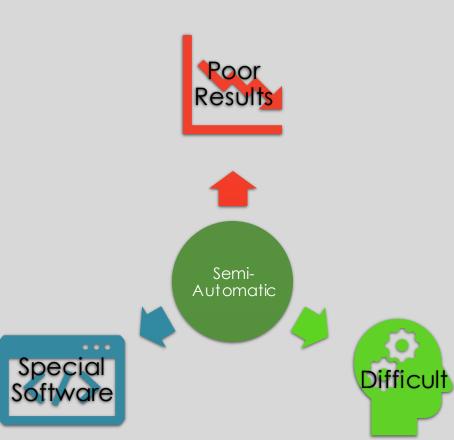




## Current Solutions are Unsatisfactory







## Goal





Train



Neural Network

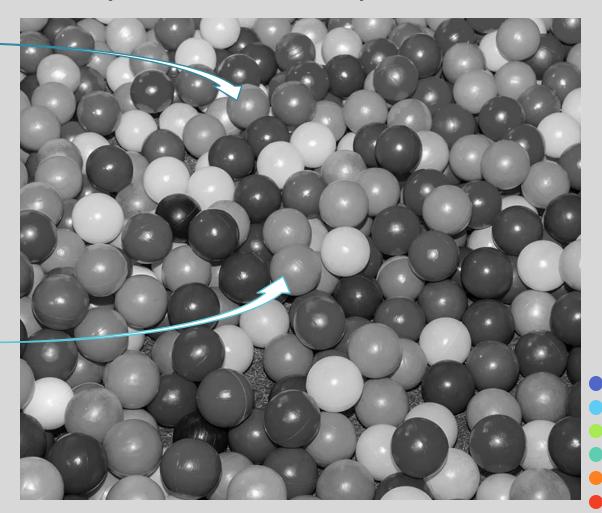


## Nuances That Increase Difficulty



Orange and Light Green have the Same Grayscale and Same Shape

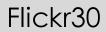




# DATA PROCESSING

## Data























#### Imagenet













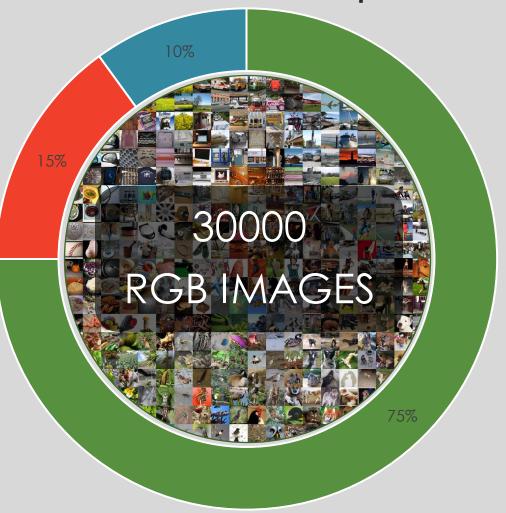






## Data Splits





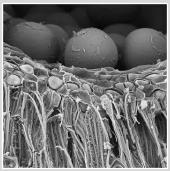
- Training
- Validation
- Testing

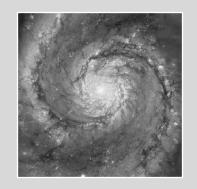
## Practical Testing Images



Medical Astronomical Historical



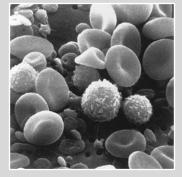


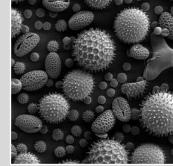


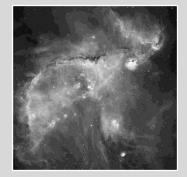


















## Data Cleaning

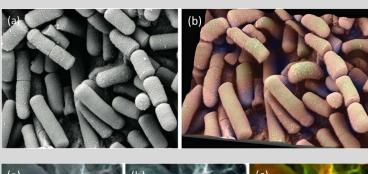


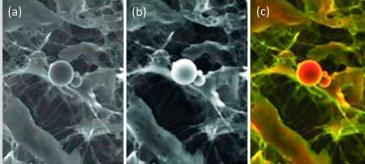


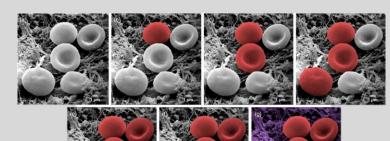
#### Data Pipeline Ground Truth Labels UV channels 224 x 224 x 2 LUV Image 224 x 224 x 3 L(greyscale) Generated channel UV channels Neural 224 x 224 x 2 224 x 224 x 1 Network Generated LUV Image RGB Image 224 x 224 x 3 224 x 224 x 3

# Practical Images May Not Be Real Color

Medical







Manually Colorized Astronomical







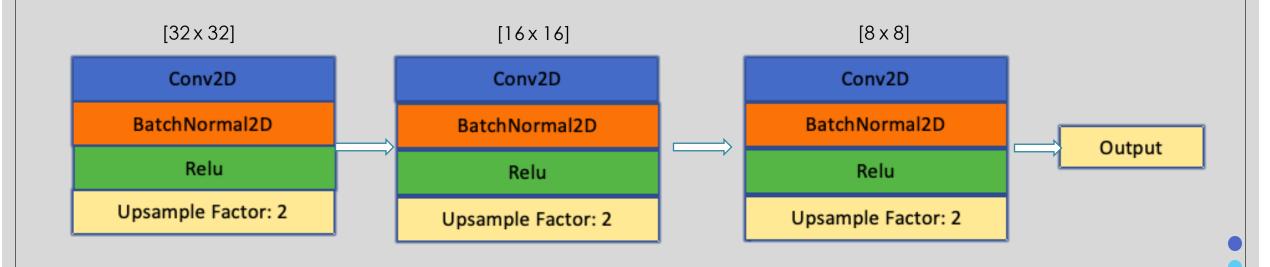
# BASELINE MODEL

REGRESSION BASED MODEL

### Starter Baseline Model



A regression-based CNN Model altered to only take in greyscale inputs.



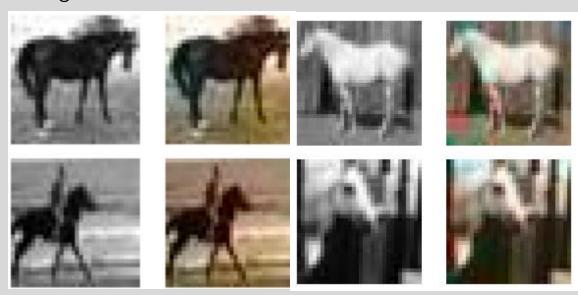
#### Results

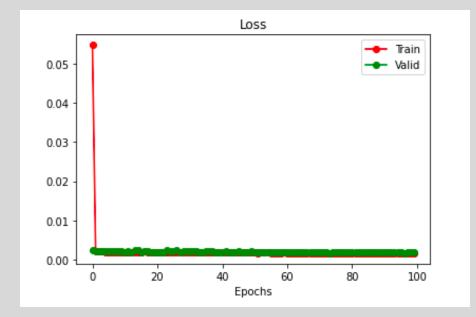


• Decent results! A good starter But definitely lots of scope for improvement.

• Avg Loss for training: 0.0017

• Avg Loss for validation: 0.0035





# INITIAL MODEL

CLASSIFICATION BASED MODEL



#### Scoping & Defining our Problem

- We will scope the task of image colourization into a pixel-wise classification task
  - We label each pixel with one of 24 colours
- •The 24 colours are selected using <u>k-means clustering</u> over colours, and selecting cluster centers
  - Given a grayscale image, predict each pixel as a color amongst the 24 colors



#### **Dataset & Constraints**



• CIFAR-10, which has images of small dimensions 32\*32 pixels

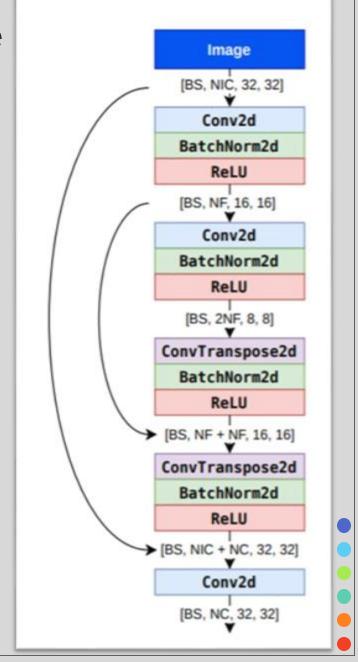
We just focus on the "horse" category of the dataset

• Furthermore, the **error** is calculated by defining distances over RGB space.



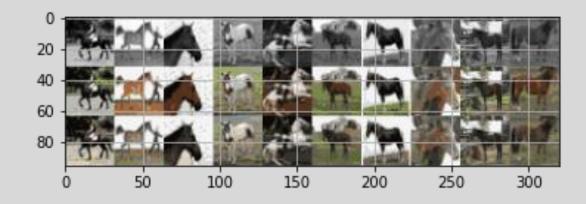
#### Model Architecture

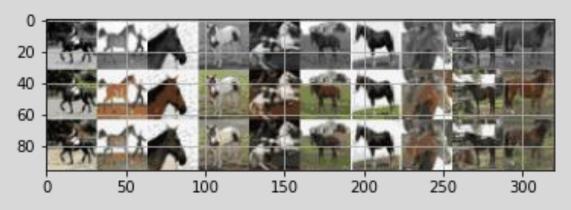
- •The architecture we use is U-NET
- This architecture has following additions over a more basic CNN architecture:
  - Strided & Transposed Convolutions instead of upsampling functions
  - Skip connections

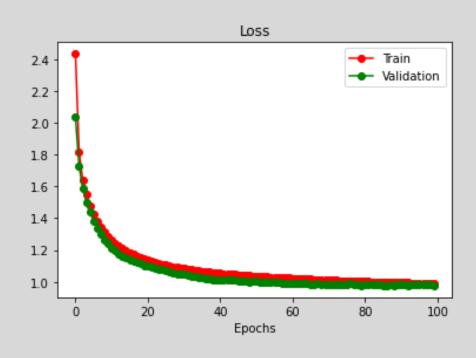


#### Results & Discussion









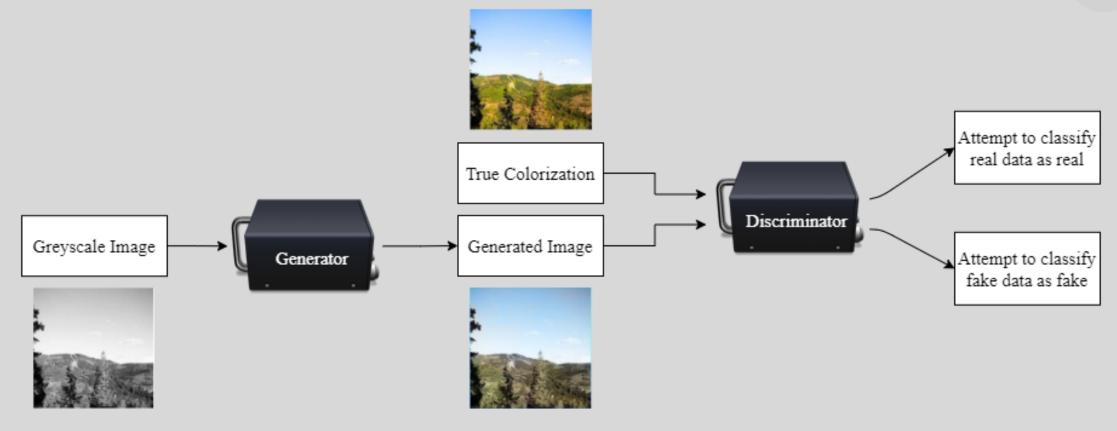
Epoch [100/100], Val Loss: 0.9769, Val Acc: 59.8%, Time(s): 98.27

# FINAL MODEL

# GENERATIVE ADVERSARIAL NETWORK

### GAN Architecture Overview



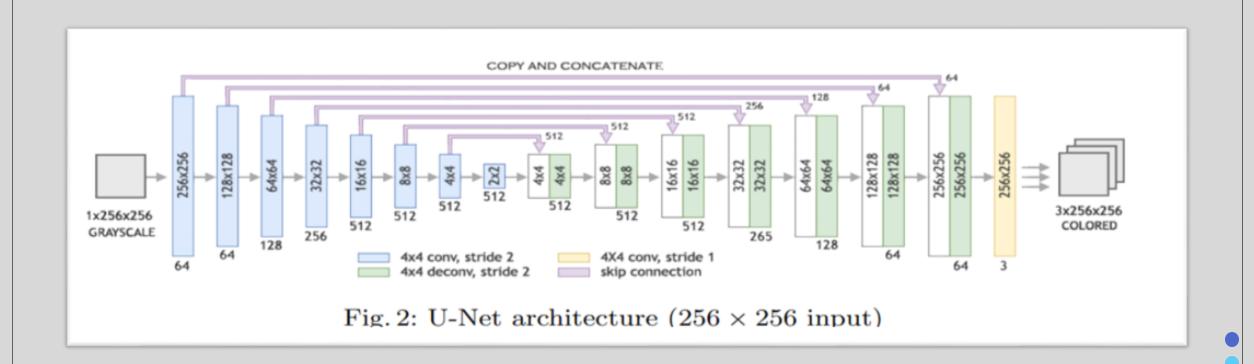


Generator tries to "trick" the Discriminator as they compete and train in parallel

### GAN Architecture - Overview



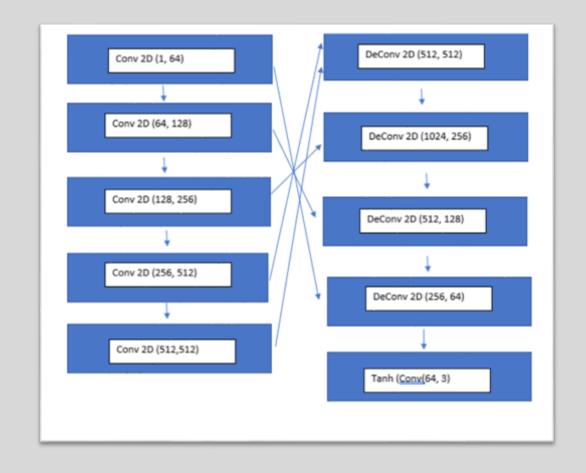
Convolutional Layers/Deconvolutional layers connected via skip net connections



### GAN Architecture - Generator



- 5 Convolutional Blocks
  - Conv2d
  - BatchNorm2d
  - ReLU
- 5 Deconvolutional Blocks
  - ConvTranspose2D
  - BatchNorm2D
  - ReLU
- Tanh activation Function

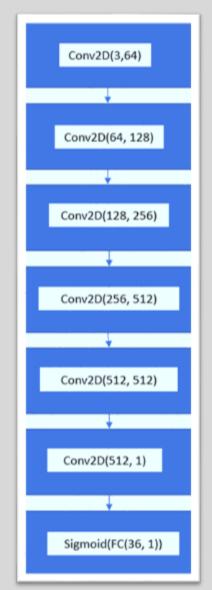


### GAN Architecture - Discriminator



- 6 Convolutional Blocks
  - Conv2d
  - BatchNorm2d
  - ReLU
- 1 Fully Connected Layer

Sigmoid Activation



## **GAN Results**













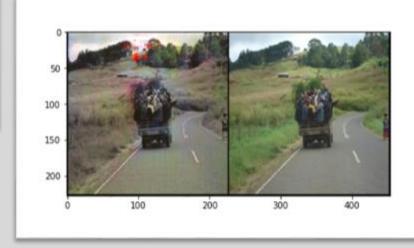








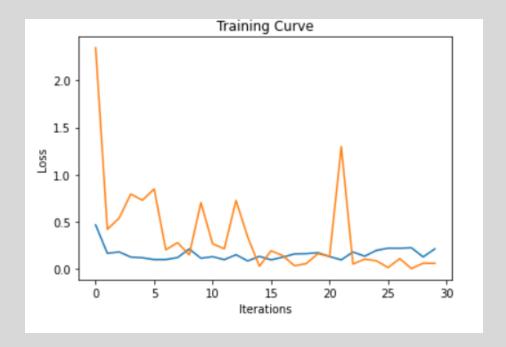




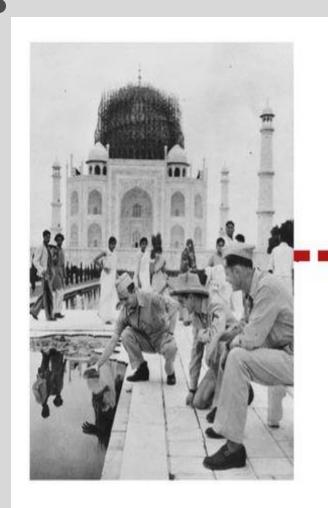
#### GAN Discussion



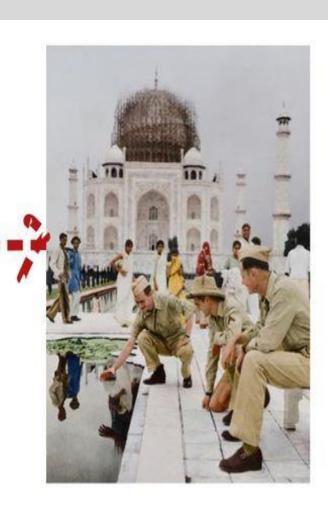
- GAN has abnormal loss graphs
  - Competition between models induces settling at value instead of typical trends
- Qualitative results are best for determining success rate of model
  - Difficult to quantify
- Colors not as vibrant and some incorrect coloring found
  - Could be due to shorter training period/training size



# Using AI to do this!



Generator



# DEMONSTRATION

