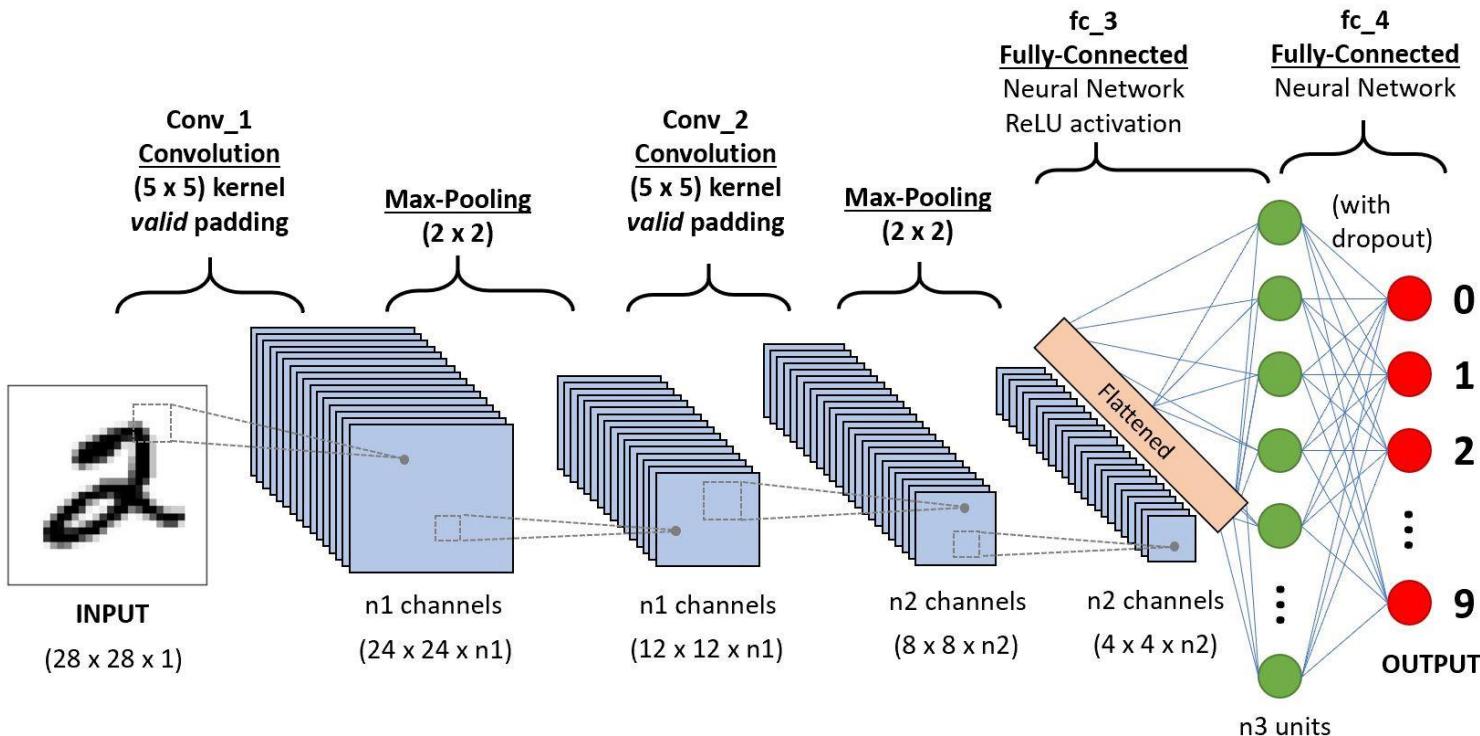


Comparing Pre-Trained Convolution Neural Networks on the CIFAR-10 Dataset

By: Arch Latham

Introduction

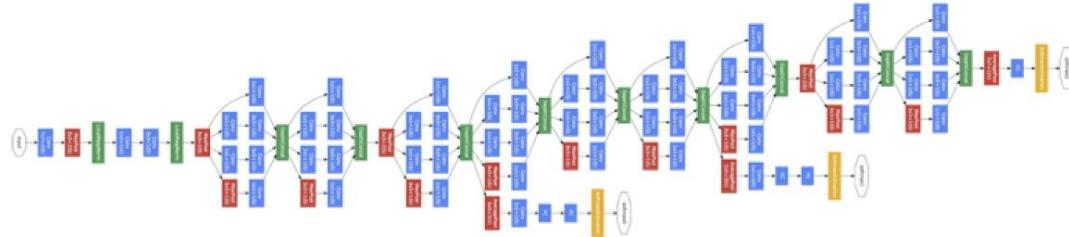
Convolution Neural Networks



Literature Review

GoogLeNet (2014)

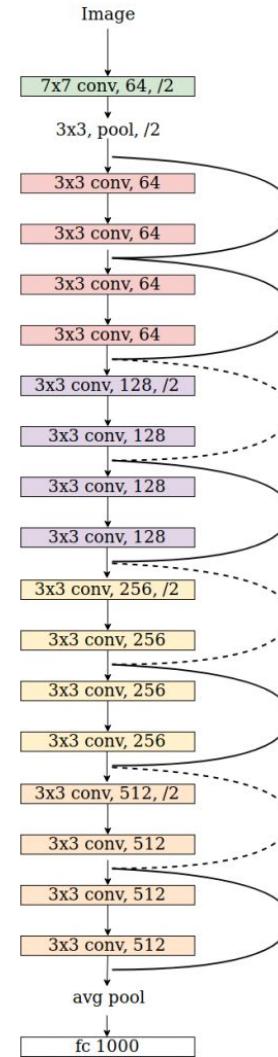
- Developed by Google
- Won ImageNet Large Scale Image Recognition Contest (ILSVRC)
- Based off of LeNet by Yann LeCun (1998)
- Inception v1, v2, v3, v4
- 22 Layers



Convolution
Pooling
Softmax
Other

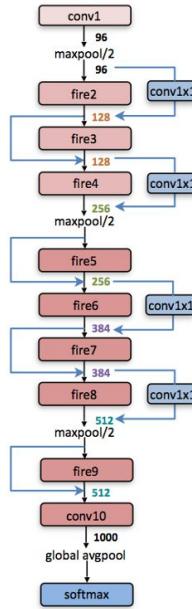
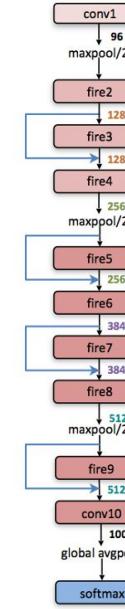
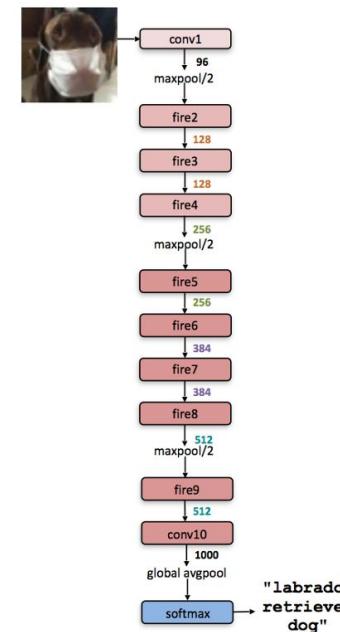
ResNet (2015)

- Developed by Microsoft and Facebook
- Won the ILSVRC
- Several Variations
 - 18
 - 34
 - 50
 - 101
 - 152



SqueezeNet (2016)

- Developed by DeepScale, UC Berkeley, Stanford
- Based off of AlexNet (2012)
 - Won the ILSVRC
- Smaller Package
 - 50x Fewer Parameters
 - 0.5 MB



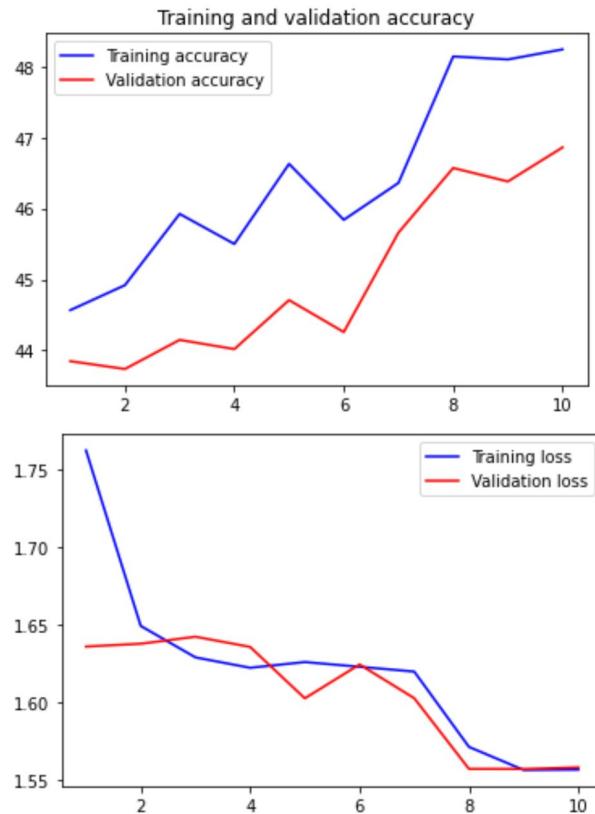
Method

- Use CIFAR-10 Dataset
- Modified Open Source Code
 - Pytorch
 - CUDA
- Run each CNN for 10 Epochs
- Record
 - Training Accuracy
 - Training Loss
 - Test Accuracy
 - Test Loss

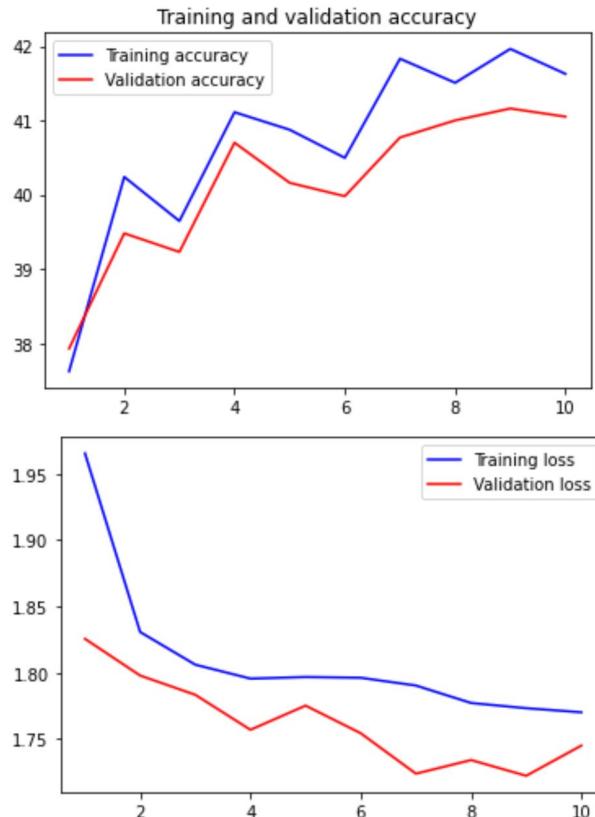
```
def validate(model, loss_fn, optimizer):  
  
    model.eval()  
  
    predictions = []  
  
    with torch.no_grad():  
        validation_batch_losses = []  
  
        for batch, labels in valid_loader:  
            batch = batch.to(cuda)  
            labels = labels.to(cuda)  
  
            labels_pred = model(batch)  
            loss = loss_fn(labels_pred, labels)  
  
            validation_batch_losses.append(float(loss))  
  
    mean_loss = statistics.mean(validation_batch_losses)  
  
    return mean_loss
```

Results

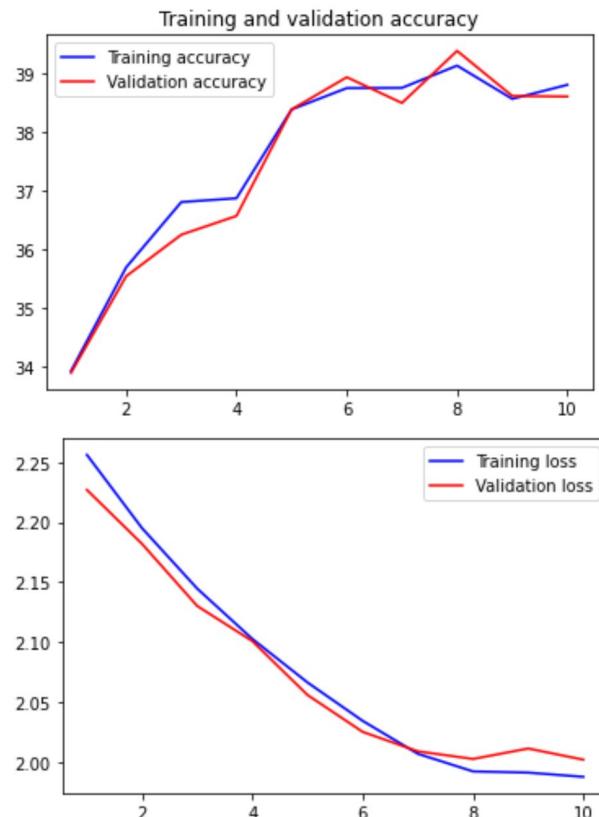
GoogLeNet



ResNet18



SqueezeNet



Conclusion

References

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Images

<https://www.google.com/url?sa=i&url=https%3A%2F%2Ftowardsdatascience.com%2Fa-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way-3bd2b1164a53&psig=AOvVaw22CBzrKQRo7xwUcuajTaHc&ust=1588390397317000&source=images&cd=vfe&ved=0CAMQjB1qFwoTCPCTv8DdkekCFQAAAAAdAAAAABAD>

https://www.google.com/url?sa=i&url=https%3A%2F%2Fmedium.com%2Fanalytics-vidhya%2Fcnn-architectures-lenet-alexnet-vgg-googlenet-resnet-and-more-666091488df5&psig=AOvVaw0Yw0EcjOWNg9OY-ZcQy0kt&ust=1588390421097000&source=images&cd=vfe&ved=0CAMQjB1qFwoTCOC_-8vdkekCFQAAAAAdAAAAABAH

<https://www.google.com/url?sa=i&url=https%3A%2F%2Fwww.learnopencv.com%2Ffully-convolutional-image-classification-on-arbitrary-sized-image%2F&psig=AOvVaw2BTufsG5DJcP-vOVnM4w9F&ust=1588390441033000&source=images&cd=vfe&ved=0CAMQjB1qFwoTCMCx8trdkekCFQAAAAAdAAAAABAD>

<https://www.google.com/url?sa=i&url=https%3A%2F%2Ftowardsdatascience.com%2Freview-squeezenet-image-classification-e7414825581a&psig=AOvVaw2Fwlq1Den310LoIJLRtxyD&ust=1588390471242000&source=images&cd=vfe&ved=0CAMQjB1qFwoTCLjoi-TdkekCFQAAAAAdAAAAABAD>

Questions?