

# Triforce

Junze

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# Overview

- Plots of Signal Background Accuracy
- **Outputs from GoogLeNet for  $\Pi_0$  vs. Gamma**
  - 94% test accuracy (Triforce+BatchNorm)
  - Using Batch-Normalization as pre-normalization process
  - Used to be 99% (Feature-Scaling: [0, 1])
- Add Feature-Scaling in `triforce.py` and `analyzer.py`
  - 95.95% training accuracy (Triforce)
  - 95.82% test accuracy (Triforce)
  - Not sure why different accuracy

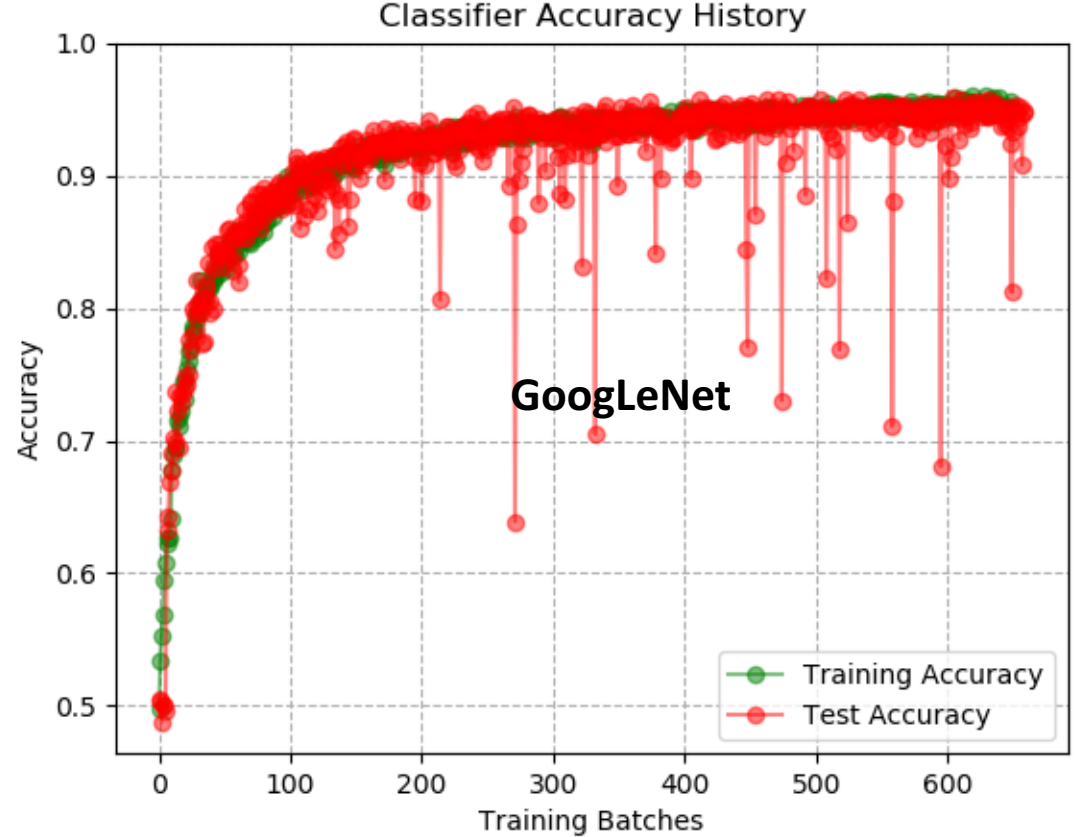
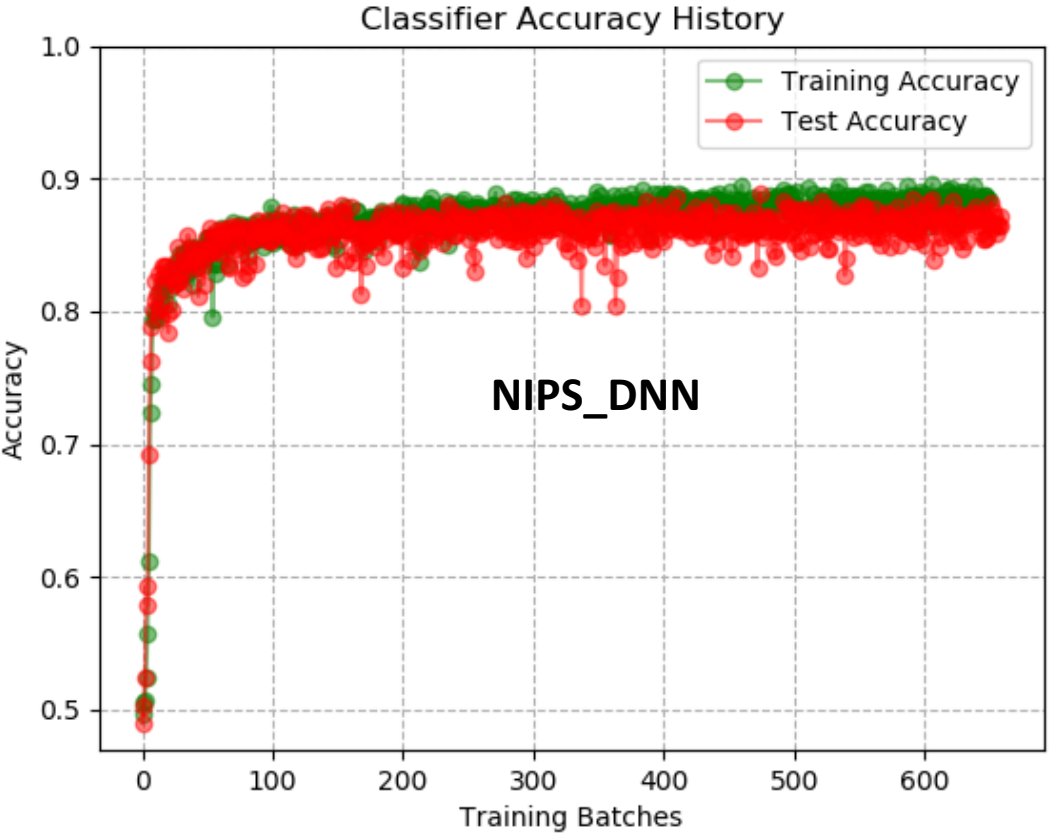
# Database

- Source: /data/LCD/V3/
- Particle: ECAL images of **Neutral Pion** and **Photon**
- Size: 20 x 10,000 = 200,000

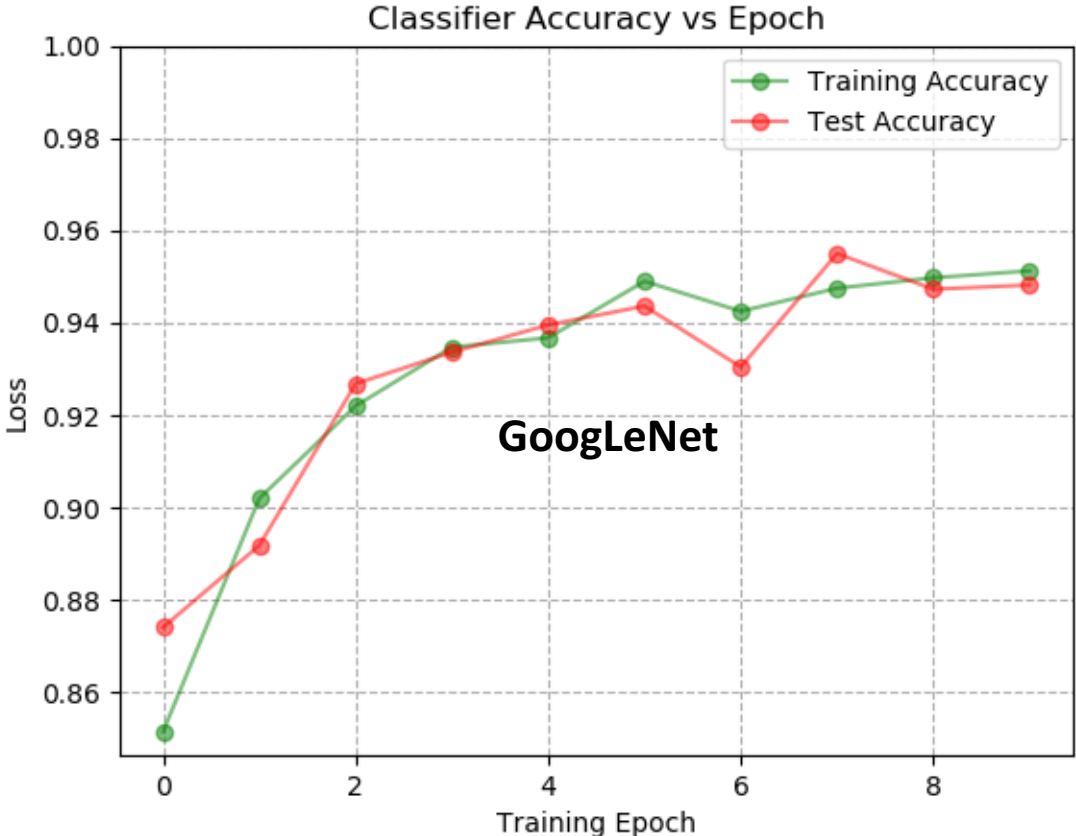
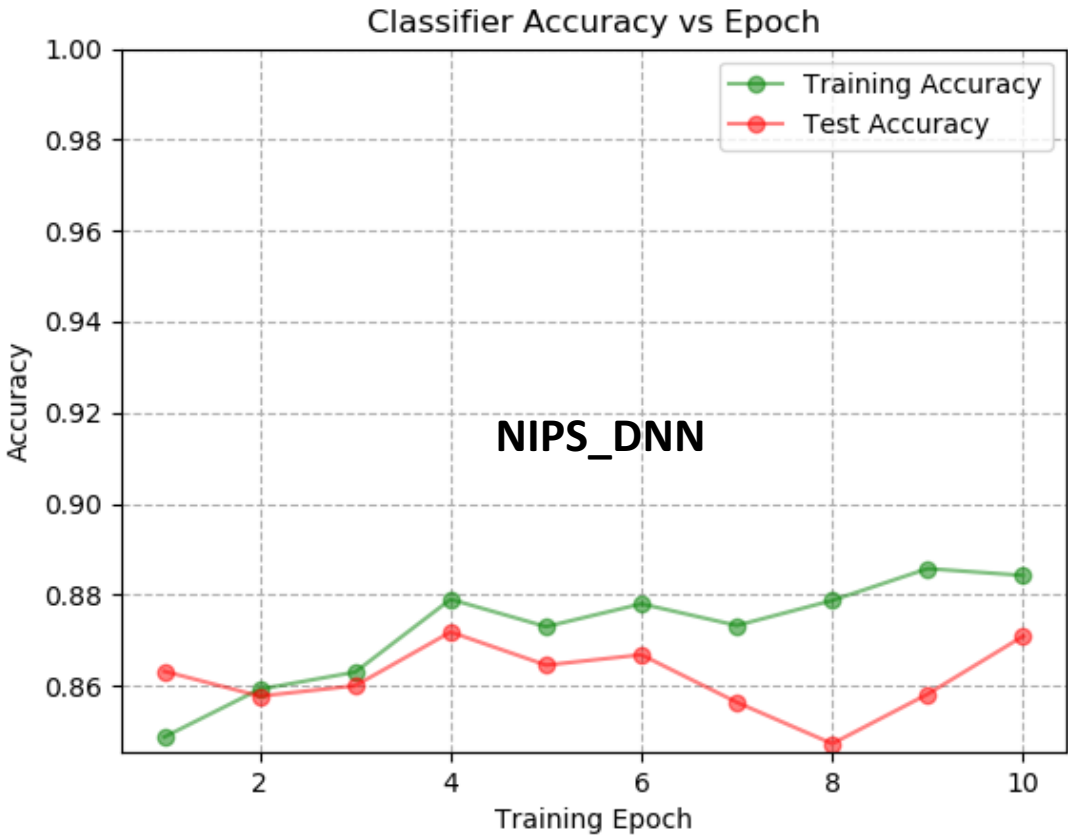
# Numerical Results

	<b>NIPS_DNN</b>	<b>Batch-Norm GoogLeNet</b>
Training Epochs	10	10
Accuracy	86.85%	93.99%
AUC	0.89	0.97
Signal Accuracy	87.02%	96.23%
Background Accuracy	87.23%	93.51%

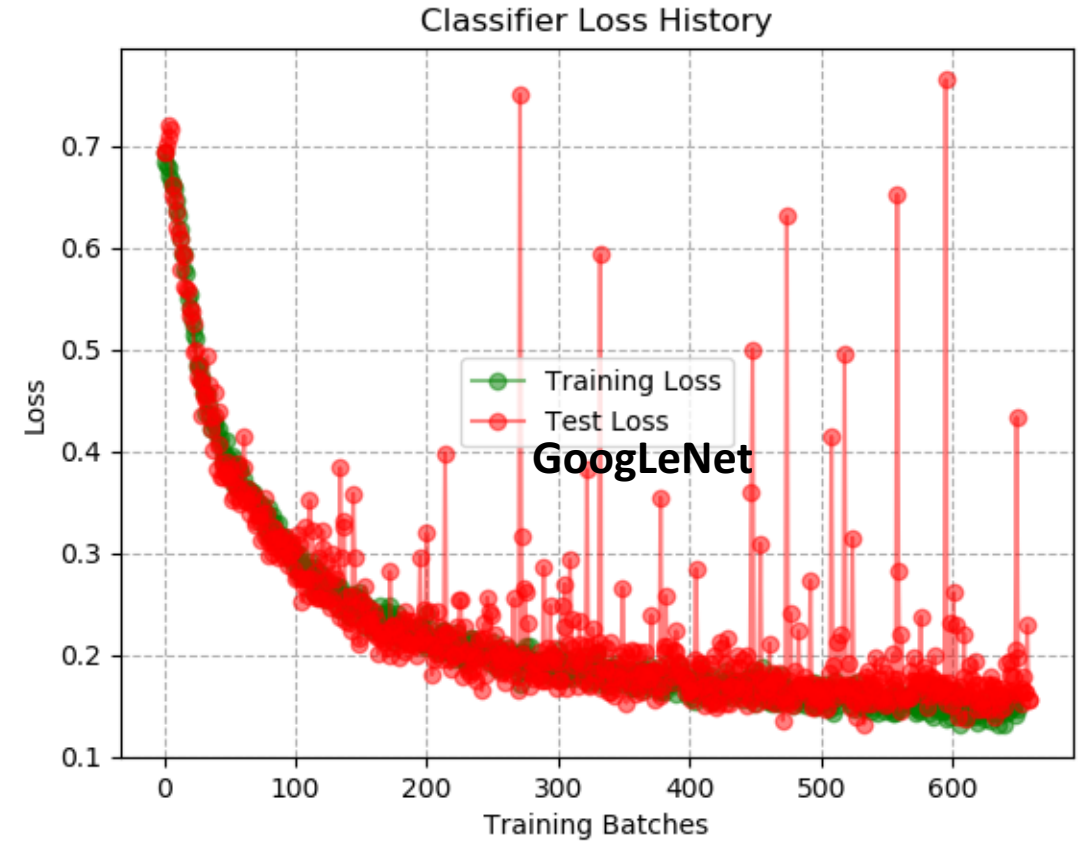
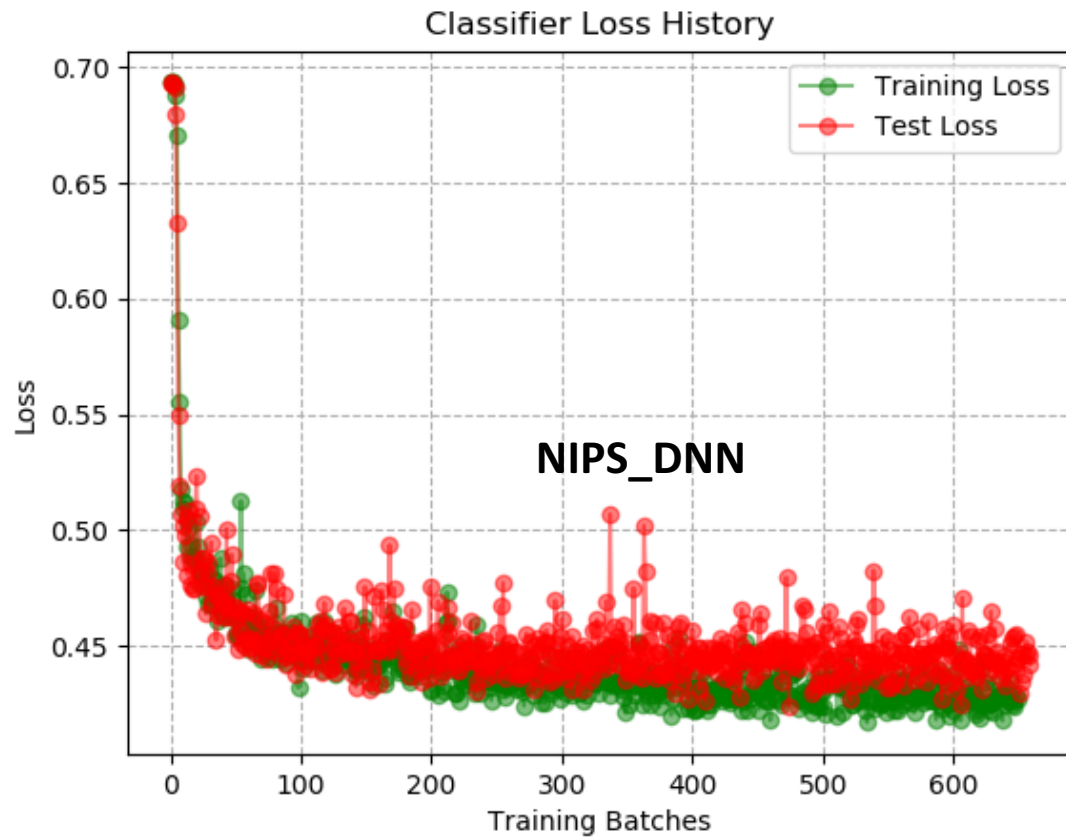
# Accuracy vs. Batches



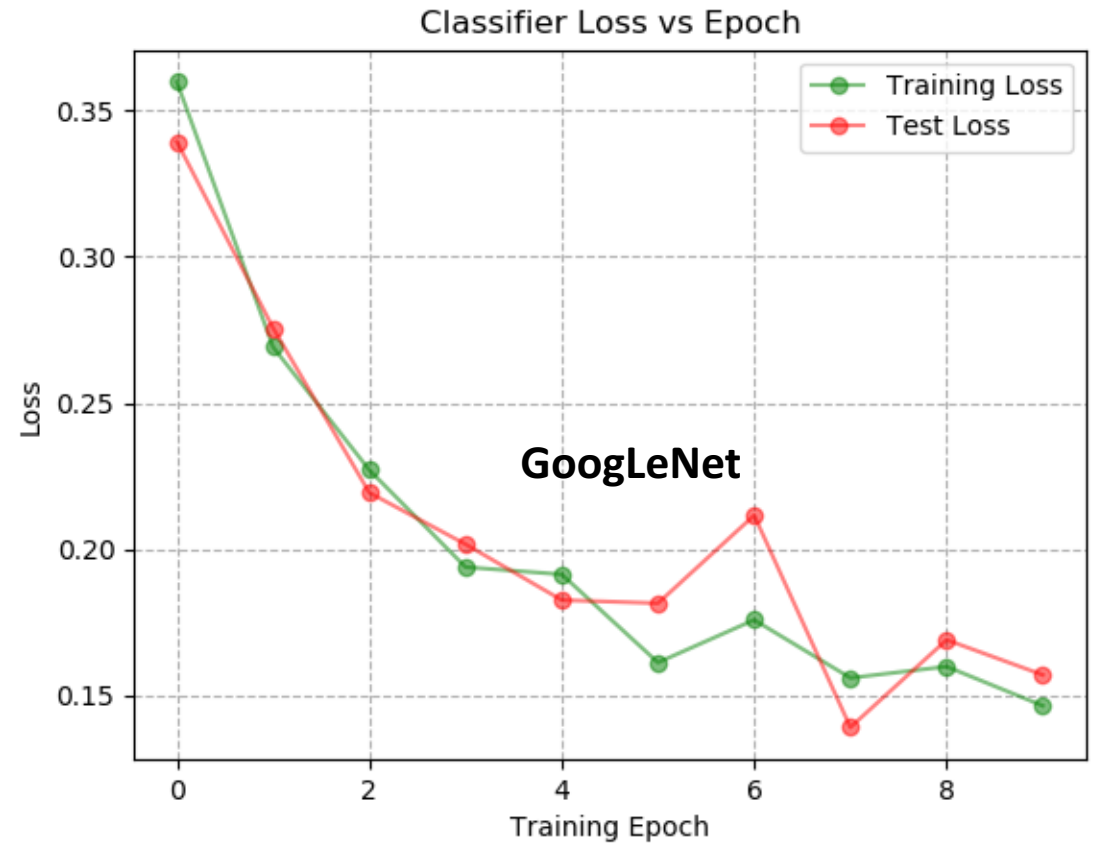
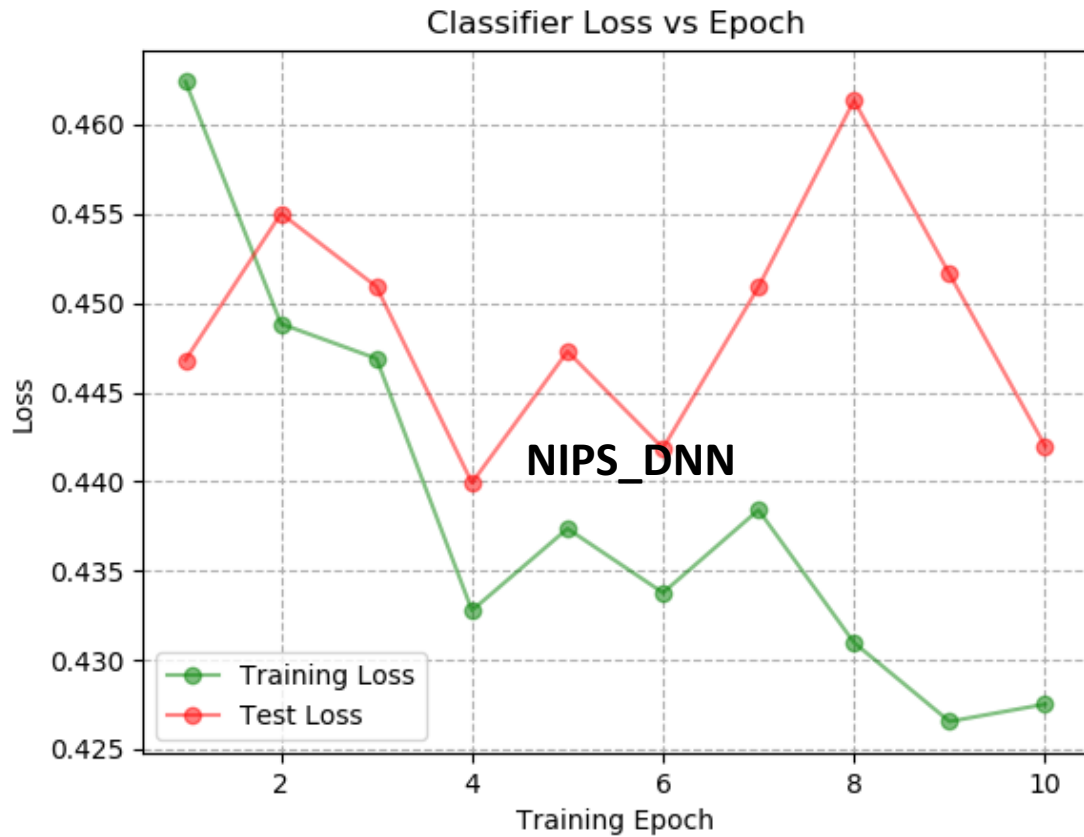
# Accuracy vs. Epoch



# Loss vs. Batches

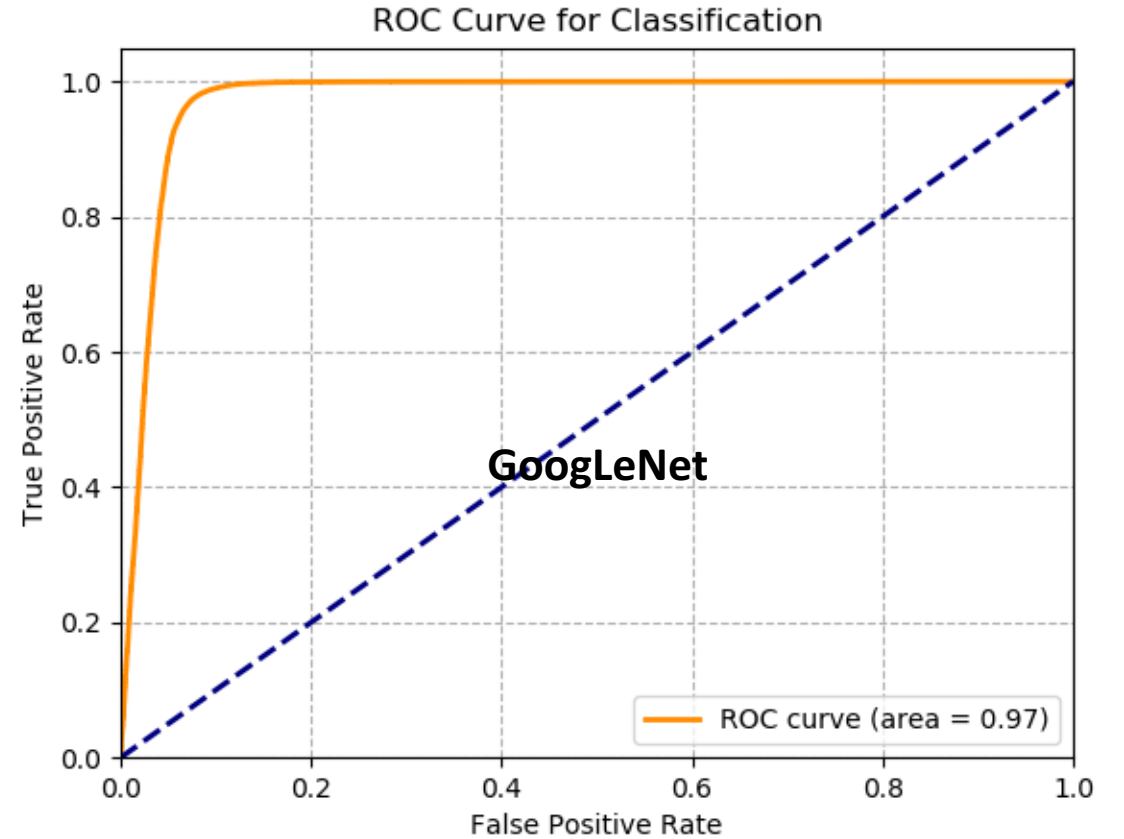
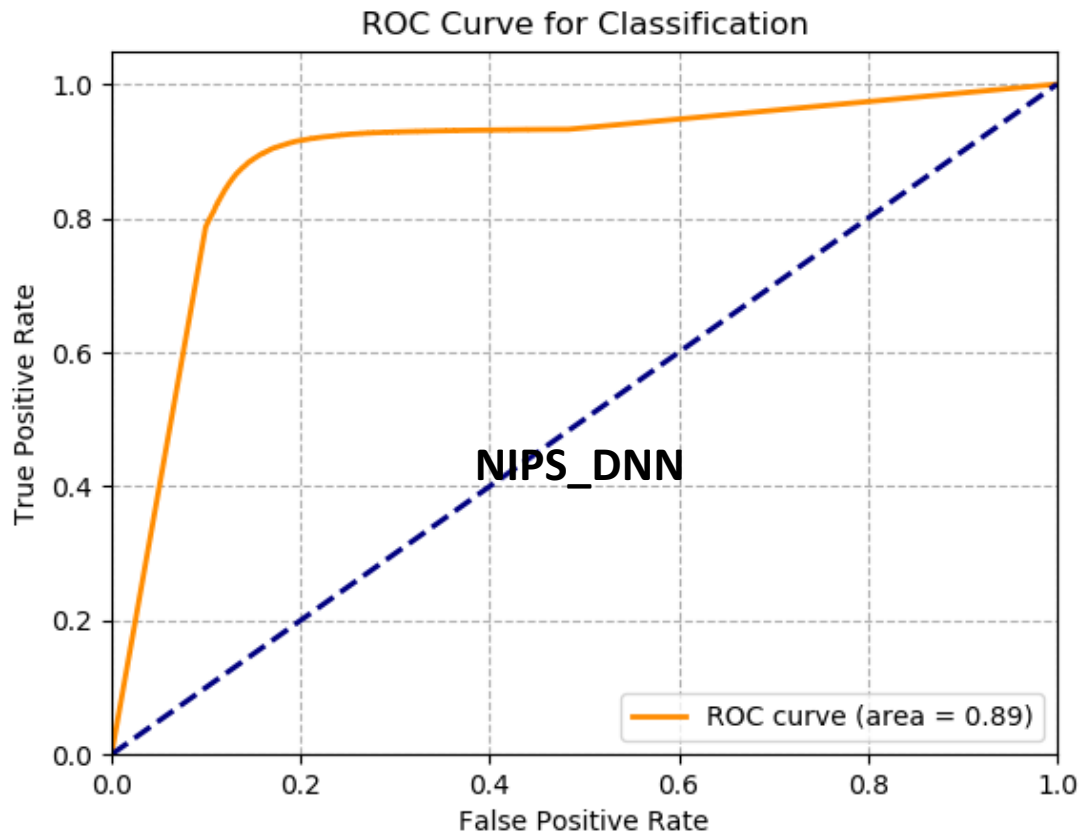


# Loss vs. Epoch

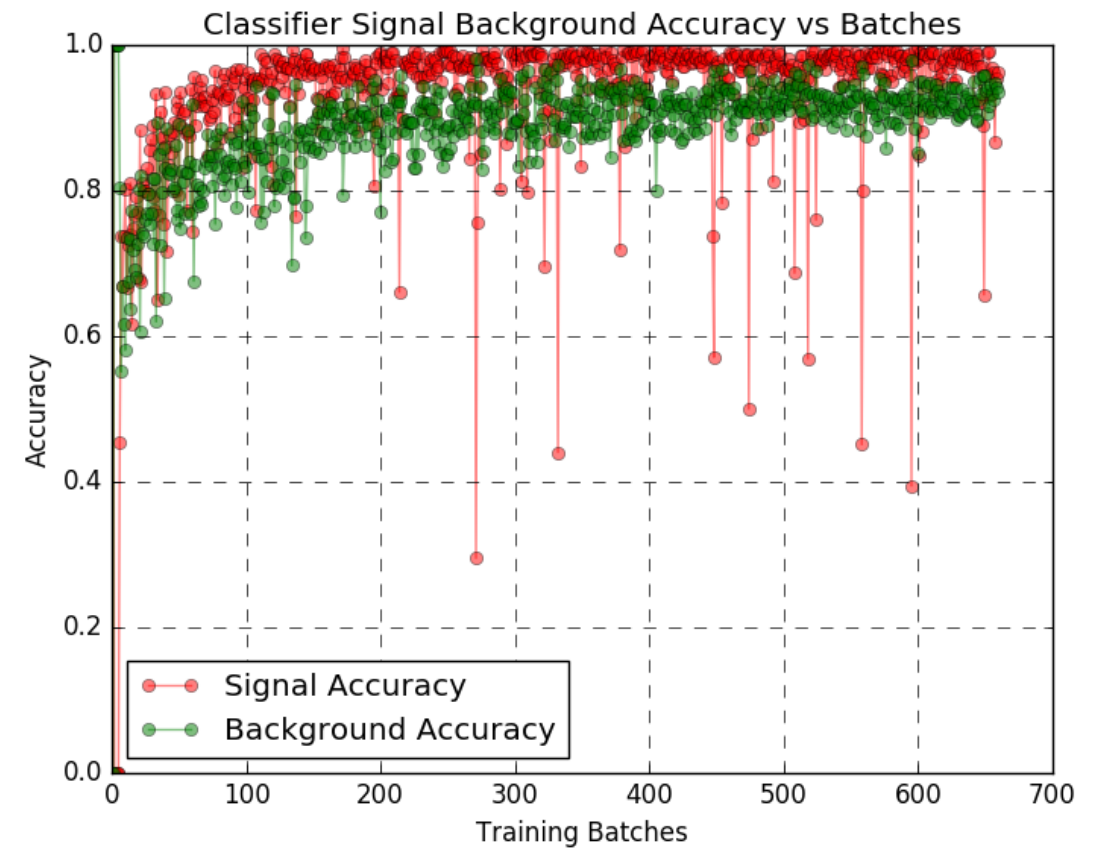
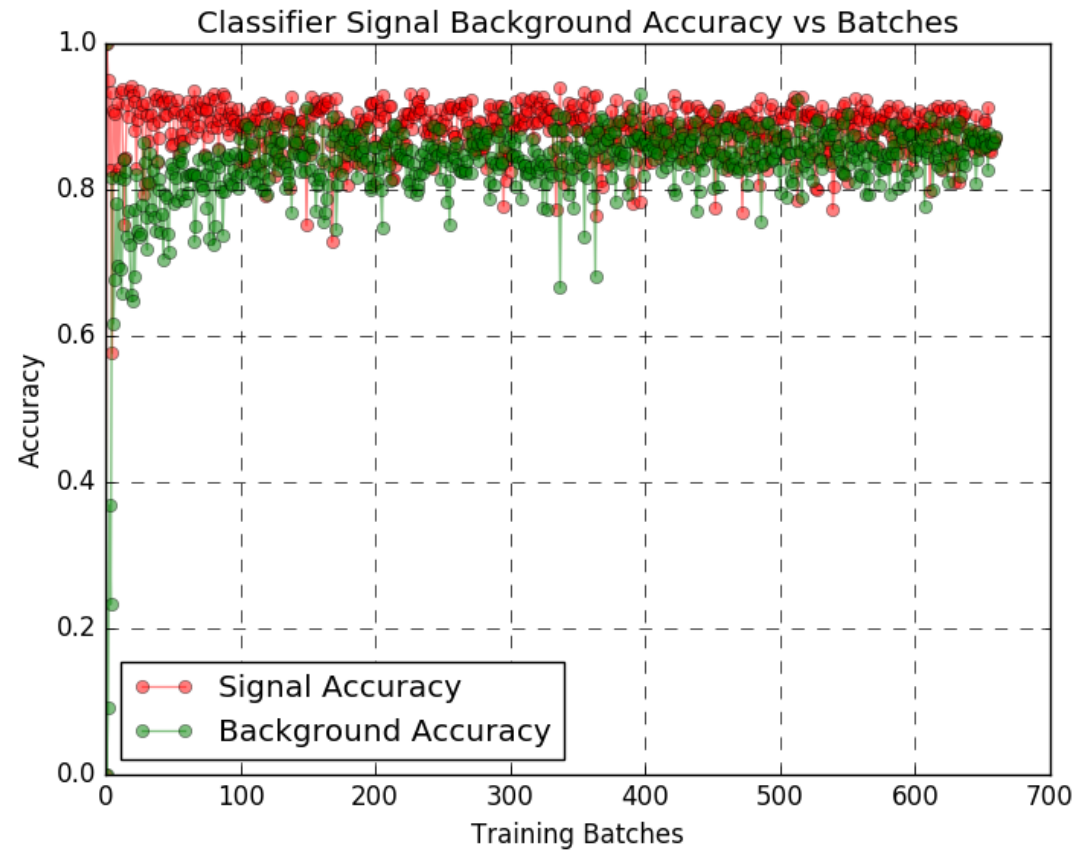




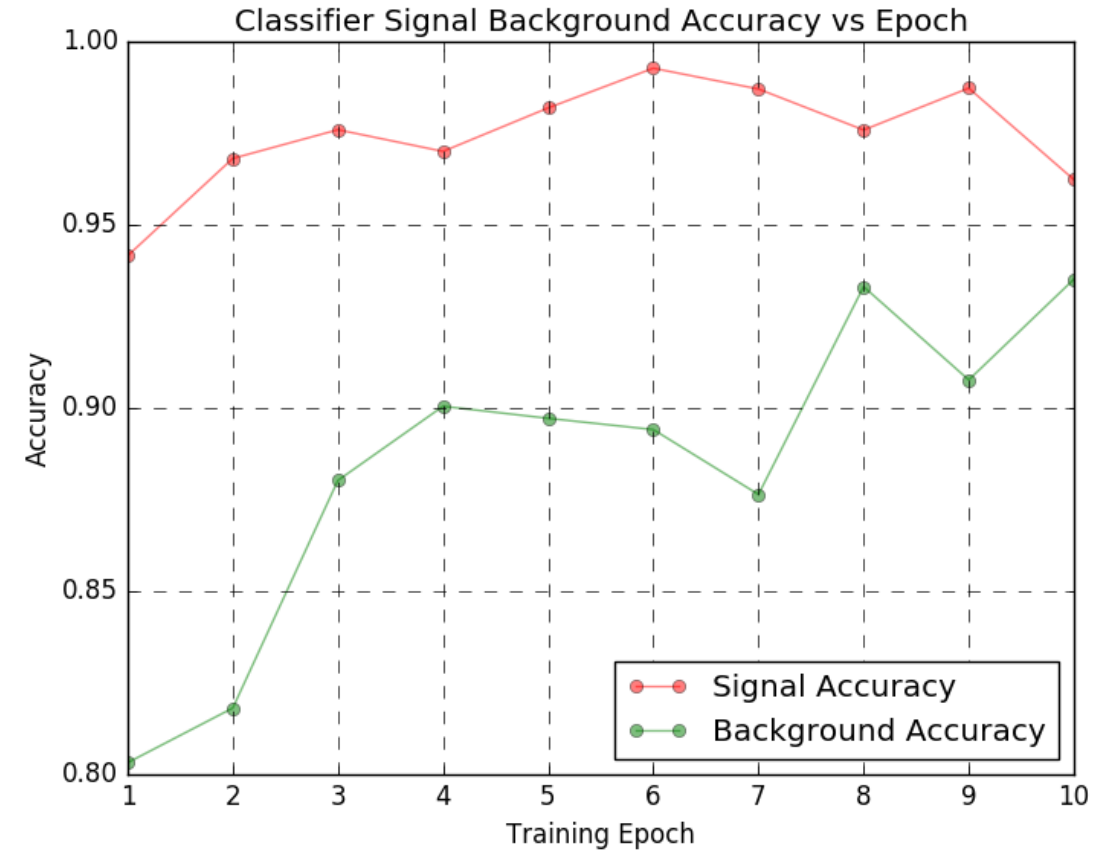
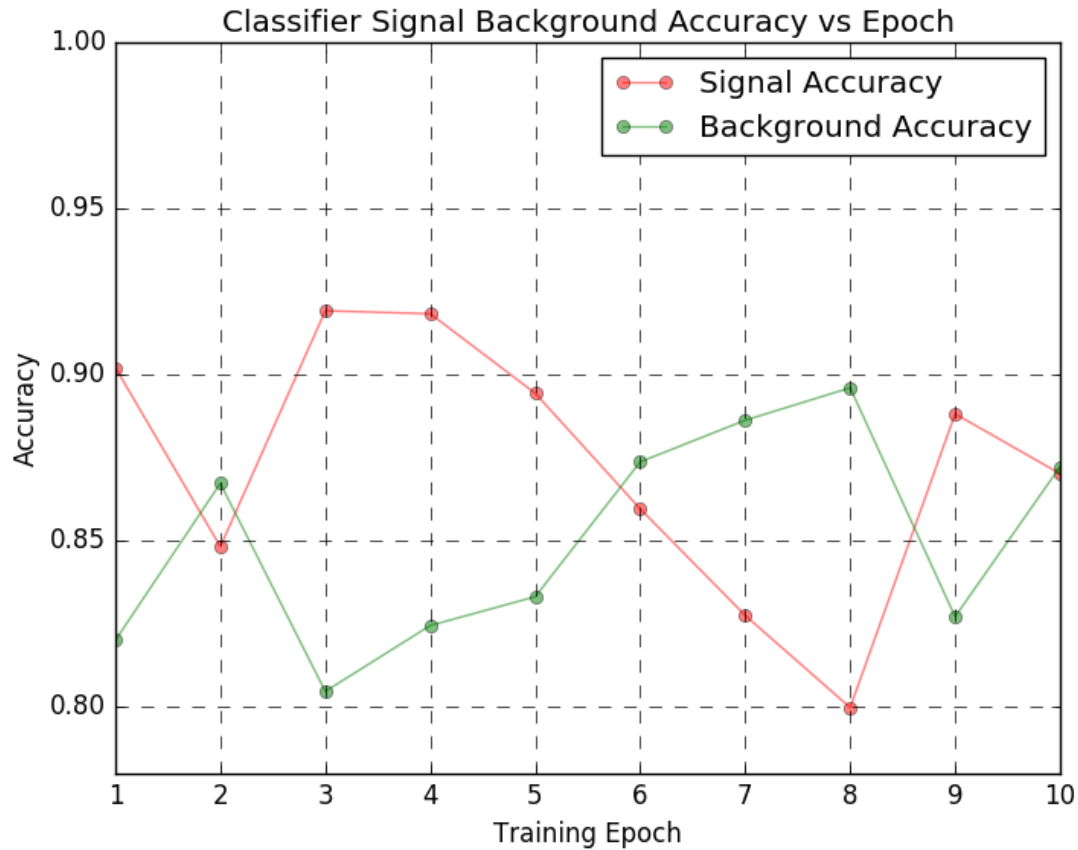
# ROC Curve



# Signal Background Accuracy vs. Batch



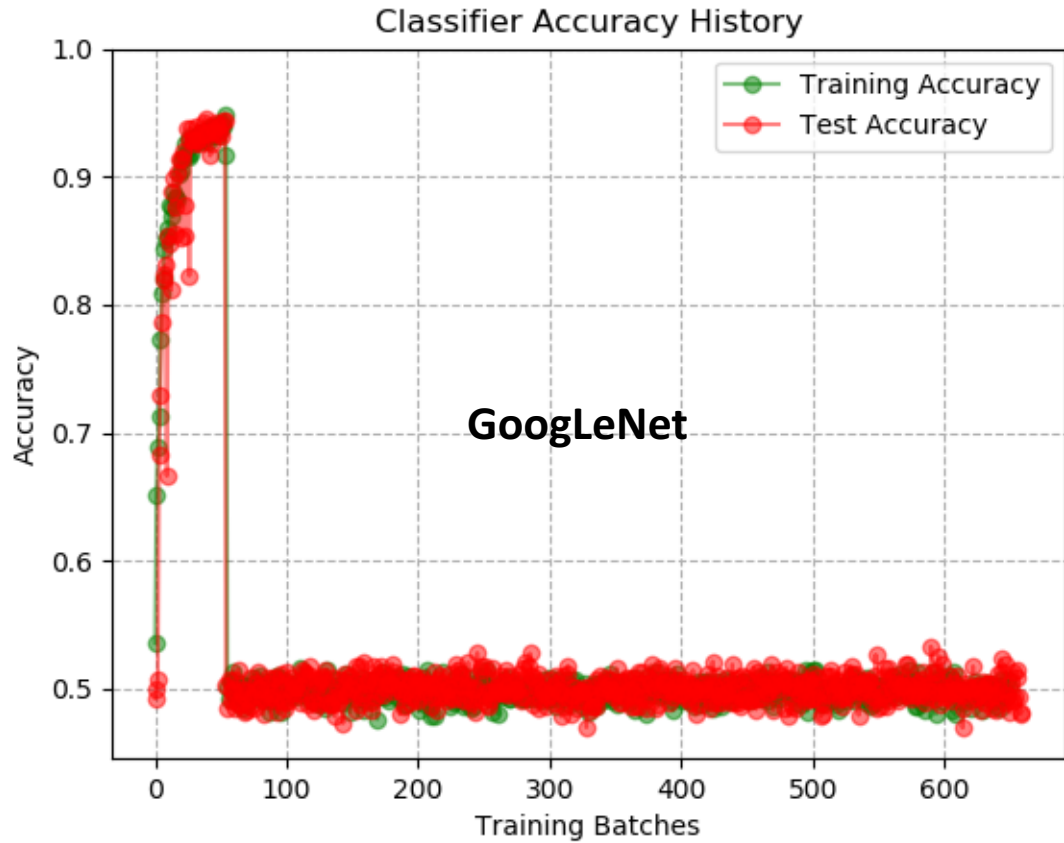
# Signal Background Accuracy vs. Epoch



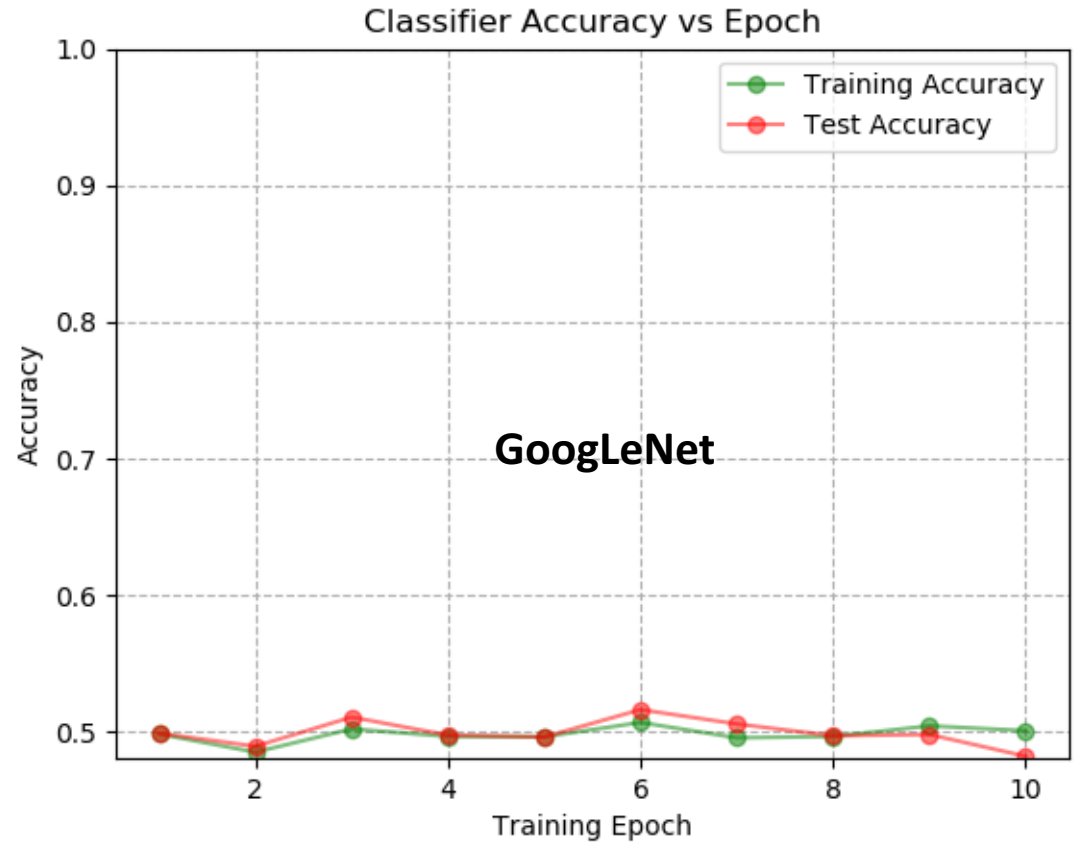
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  - 94% test accuracy (Triforce+BatchNorm)
  - Using Batch-Normalization as pre-normalization process
  - Used to be 99% (Feature-Scaling: [0, 1])
- **Add Feature-Scaling in triforce.py and analyzer.py**
  - 94.90% best training accuracy (Triforce)
  - 94.55% best test accuracy (Triforce)
  - Not sure why different accuracy

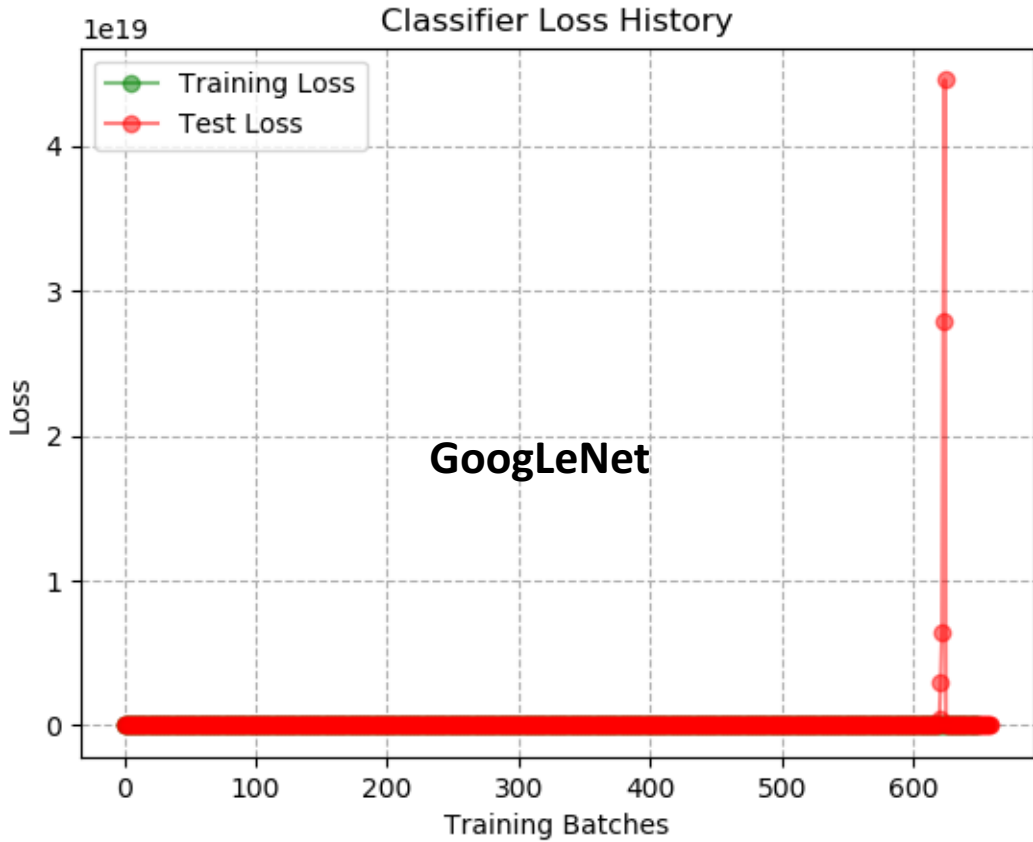
# Accuracy vs. Batches



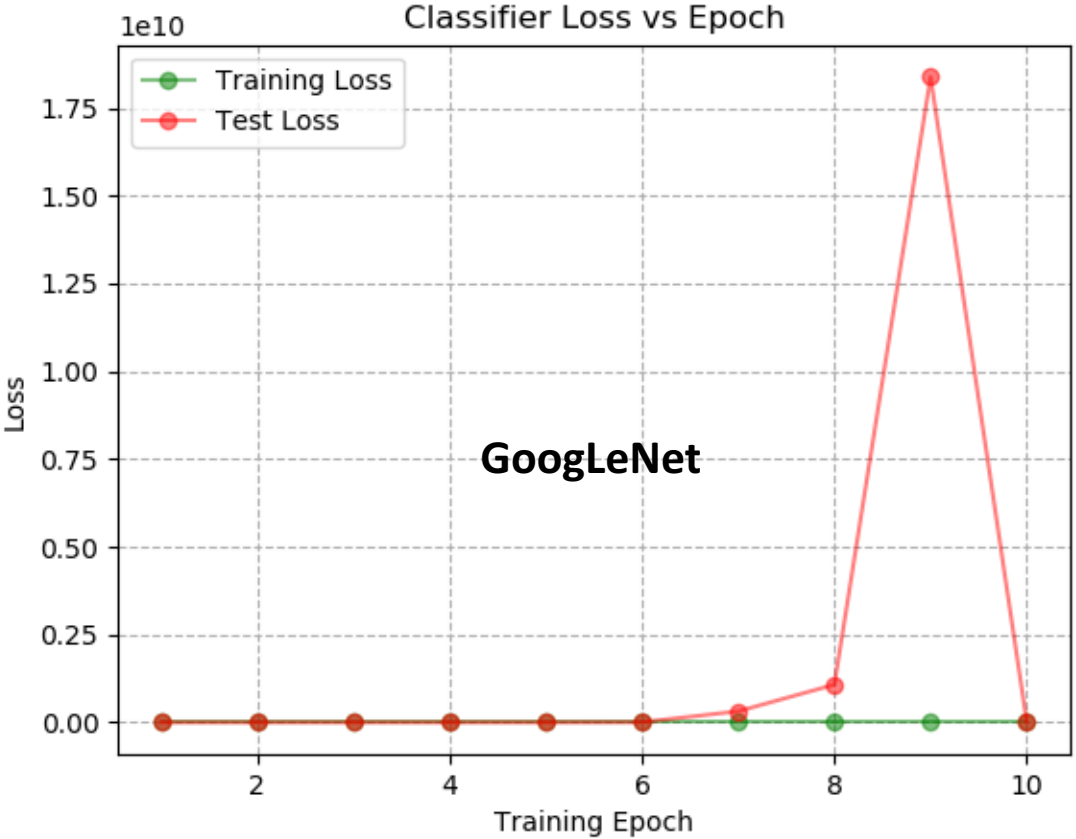
# Accuracy vs. Epoch



# Loss vs. Batches



# Loss vs. Epoch





# Future Work

- Generating results of NIPS\_DNN with different size of window
  - Based on Random-Angle new samples
- Looking into feature-scaling GoogLeNet code