Histopathologic cancer detection using Convolutional Neural Networks

By:

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- Kaggle
 - Primarily classifications competitions on various types of data using Machine Learning
 - Around 1200 participants in my competition
- Histopathology examination of changes human tissue

The data

- Classify wheter small microscopic images of human tissue have tumour present or not
- Input images of the size (96, 96) in RBG
- Hard to tell the difference between positive and negative class





The data

- 220025 training and 57458 test images.
- Distribution amonst training images
 - 130908 negative examples
 - 89117 positive examples





Data preprocessing

- Normalize data before feeding it to the neural net
 - Subract mean and divide by the standard deviation for each of the training example
- Image augmentation as one experiment more on that later

Evaluation metric

- Predetermined evaluation metric on Kaggle
- ROC AUC score
 - Plots TPR (true positive rate) versus FPR (false positive rate)
 - Score is based on the area under the curve (0.5 for a purely random classifier and 1 for a perfect one)



Implementation and CNNs

- Trained on a Google Cloud virtual machine
 - Nvidia TESLA K80 GPU
- Implemented using Python and high-level library Keras and the use of Neural Networks
 - Convolutional layers
 - Fully connected layers



Architectures – user defined



Architectures – pre trained

- Tried two pretrained convolutional neural networks
 - VGG16 23 layers
 - InceptionResNetV2 572 layers
- Both pre trained on ImageNet data.



Results

Model	Score
Baseline	0.9358
VGG16	0.9347
Image augmentation	0.9279
InceptionResNetV2	0.9485
Final	0.9530

Suggestions for improvement

- Ensemble learning
- More extensive image augmentation
- Build a classifier that classes more false negatives than false positives.