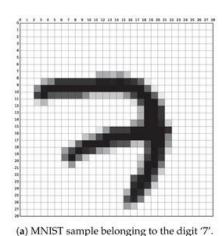
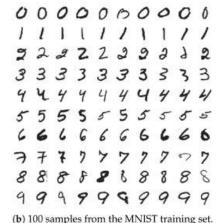
Application of the Lottery Ticket Hypothesis in NLP and Early Pruning



Intermission





Source: https://www.mdpi.com/applsci/applsci-09-03169/article_deploy/html/images/applsci-09-03169-g001-550.jpg



Source: https://www.bonaccorso.eu/wp-content/uploads/2016/07/28019400581_e1eb13ccc8_b.jpg

Prof. Dr.-Ing. Ralf Steinmetz KOM - Multimedia Communications Lab

Tim Unverzagt

Structure



Introduction

Motivation

Related Work

Task Definition

Progress

Remaining Work

Structure



Introduction

Motivation

Related Work

Context of the thesis

Task Definition

Progress

Remaining Work

Content of the thesis















Initial Thoughts

- Many good reasons to initialize & train neural networks with many parameters
- Empirical evidence that many networks can be reduced after training while maintaining performance
 - aka. "Pruning"
 - Desirable due to bias towards small models (Ockham's razor)















Initial Thoughts

- Good reasons to initialize & train neural networks with many parameters
- Empirical evidence that many networks can be reduced after training while maintaining performance
 - aka. "Pruning"
 - Desirable due to bias towards small models (Ockham's razor)

- Main Question:
 - "How important are the pruned weights during training?"





Lottery Ticket Hypothesis

Mot









Sheer number of subnetworks results in subnetworks with favorable initialization







- Mot
- Rel







- Sheer number of subnetworks results in subnetworks with favorable initialization
- Train a subnetwork with initial parameters
 - Similar performance ==> "lottery ticket"







- Mot
- Sheer number of subnetworks results in subnetworks with favorable initialization



Train a subnetwork with initial parameters



Similar performance ==> "lottery ticket"



Extraction of "lottery-ticket" after the full network is trained



Pruning weights based on magnitude finds a lottery ticket

Motivation









Task



Rem

Motivation

















Time & Memory

- Speedup during execution just as regular pruning
 - But remarkable compression rate: up to ~50x
- Decrease in memory usage during execution
- Possible speedup during development
 - There might be a way to identify lottery tickets early

Motivation

















Time & Memory

- Speedup during execution just as regular pruning
 - But remarkable compression rate: up to ~50x
- Decrease in memory usage during execution
- Possible speedup during development
 - There might be a way to identify lottery tickets early

Theory of Neural Networks

- "Lottery-tickets" contain weights necessary for training
- Identification of "lottery-tickets" might explain importance of weights

Related Work









Task









Fully Connected Neural Network

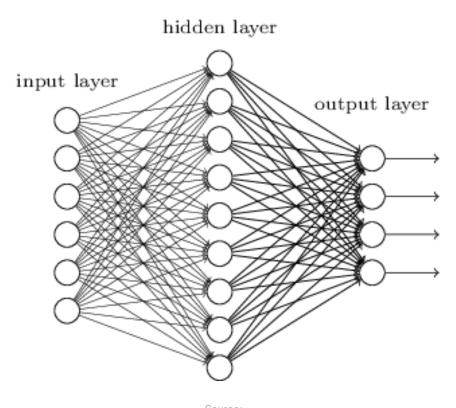












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Convolution in Neural Networks

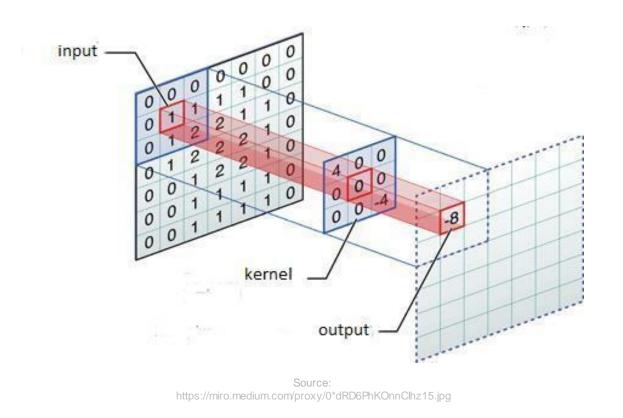
Mot

Rel

Task

Pro

Rem







Convolutional Neural Network Architecture (Lenet-5)



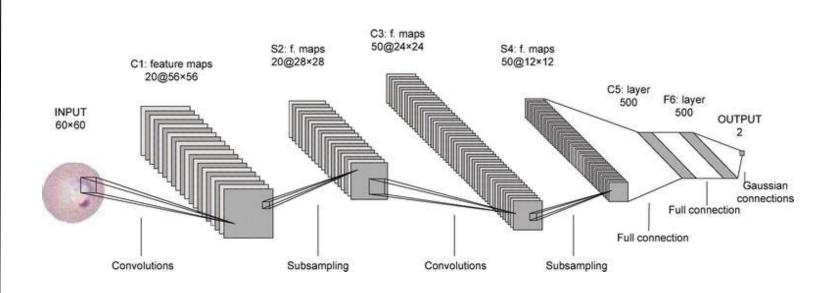












Source: https://api.intechopen.com/media/chapter/58989/media/F4.png





Language Models

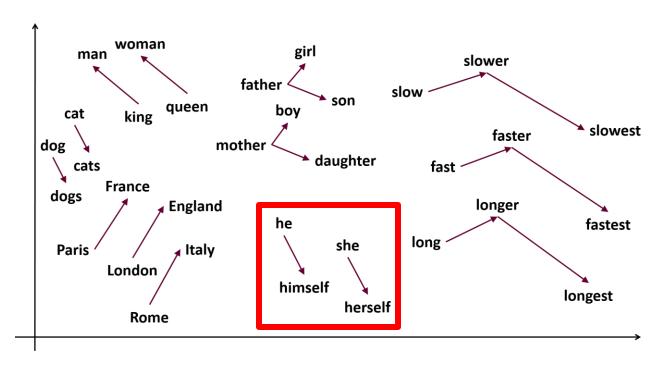
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Source: https://samyzaf.com/ML/nlp/w ord2vec2.png

Related Work – CNN in NLP















"Convolutional Neural Networks for Sentence Classification"

- 2014
- Task:
 - Varying Classifications
- Datasets:
 - Movie reviews
 - SST-1, SST-2
 - Subjectivity dataset
 - TREC question dataset
 - Customer reviews
 - **MPQA**

Related Work – CNN in NLP





"Convolutional Neural Networks for Sentence Classification"

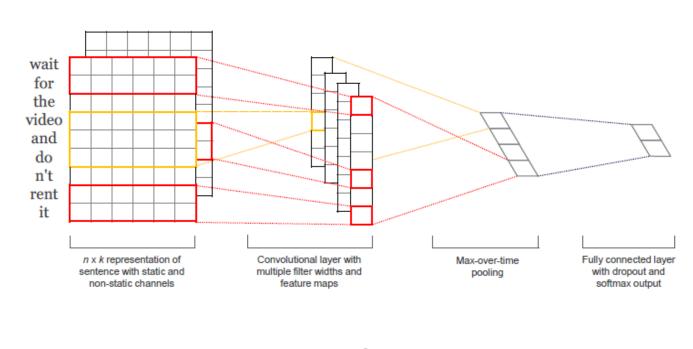
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Task

Pro

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Source: "Convolutional Neural Networks for Sentence Classification" Figure 1

Related Work - Pruning





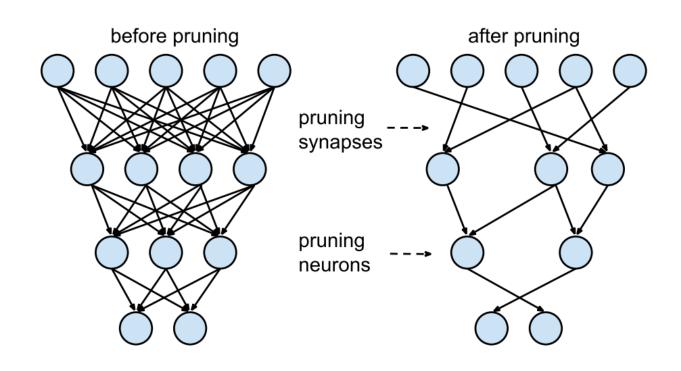












Source:

https://www.mdpi.com/applsci/applsci-09-03169/article_deploy/html/images/applsci-09-03169-g001-550.jpg

"Learning both Weights and Connections for Efficient Neural Netw orks" Figure.3

Related Work – Pruning





"Learning both Weights and Connections for efficient Neural Networks"



2015



- Task:
 - Image Classification (ImageNet)



Architectures:



LeNet (300-100-FC, 5-CNN)



AlexNet



- VGG-16
- Compression:
 - 9x to 13x

Related Work – Pruning















"ThiNet: A Filter Level Pruning Method for Deep Neural Network Compression"

- **2017**
- Task:
 - Image Classification (ImageNet)
- Architectures:
 - VGG-16
 - ResNet-50
- Compression:
 - Up to ~17x

Related Work – Pruning





Mot











"The Lottery Ticket Hypothesis: Finding Sparse, Trainable Neural Networks"

- 2019
- Task:
 - Image Classification (MNIST)
- Architectures:
 - Lenet-FCN (300-100-FCN)
 - Simple CNN (Conv-2, Conv-4, Conv-6)
 - VGG-19
 - ResNet-18
- Compression: ~20x to ~50x

Related Work – Early Pruning





Mot









"Really should we pruning after model be totally trained? Pruning based on a small amount of training"

- **2019**
- Task:
 - Image Classification (MNIST, CIFAR-10)
- Architectures:
 - Unspecified CNN
 - VGG-19
- Compression --- Training Speed-Up:
 - ~10x --- 10x















"Rethinking the Value of Network Pruning"

- **2018**
- Observation:
 - Randomizing weights does not worsen a pruned network
- Weights are not essential to the quality of pruned network
- Pruning at its core is about finding suitable network architectures















"Network Architecture Search: A Survey"

2019

- Categorization of NAS-Algorithms:
 - Search Space:
 - Space of possible architectures
 - Search Strategy:
 - Space of possible architectures
 - Performance Estimation Strategy:
 - Space of possible architectures

















"Deconstructing Lottery Tickets: Zeros, Signs, and the Supermask"

- **2019**
- Alteration of the Search Strategy (based on Magnitude):
 - large final (original strategy)
 - small final
 - large initial
 - small initial
 - large init & large final
 - small initial & small final
 - magnitude increase
 - movement
 - random (baseline strategy)

















"Deconstructing Lottery Tickets: Zeros, Signs, and the Supermask"

- **2**019
- Alteration of the Search Strategy (based on Magnitude):
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Task





"Deconstructing Lottery Tickets: Zeros, Signs, and the Supermask"

- **2019**
- Alteration of the Search Strategy (based on Magnitude):
 - large final (original strategy)
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 - random (baseline strategy)

Large final & same sign

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Introduction

Motivation

Related Work

Context of the thesis

Task Definition

Progress

Remaining Work

Content of the thesis

Task I















Reproduction

- No source-code available
 - ⇒ Produce own source-code
- Verify source-code by running experiments from the paper
 - Lenet-FCN
 - CNN-4
 - VGG-18

Task II





Transfer to NLP



Rel



Task





- Original context for the paper
 - Task: Image Classification
 - Dataset: "MNIST"
 - Model: Varying FCN and CNN
- Find comparable context in NLP
 - Task: Topic Classification
 - Dataset: "Reuters-21578"
 - Model: TBD
- Check if the Lottery-Ticket-Hypothesis holds

Task III













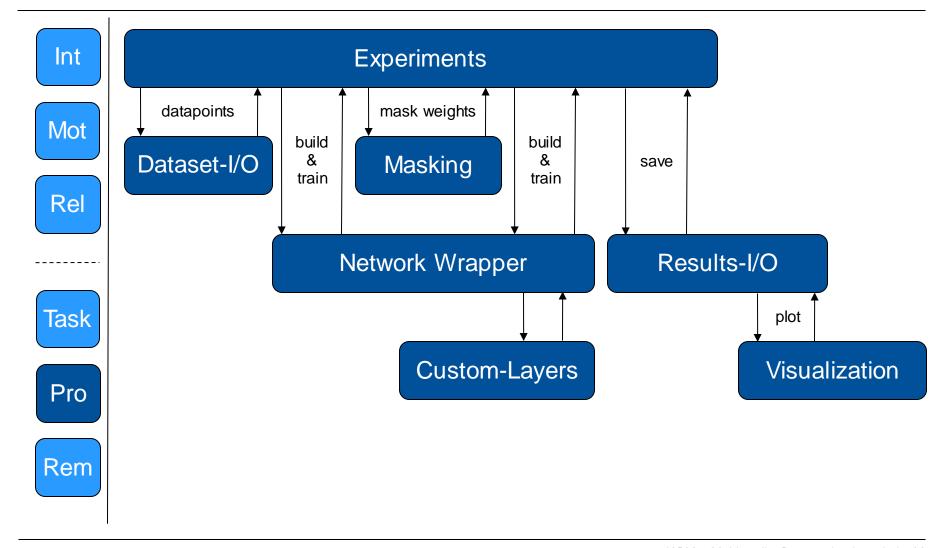


Early Retrieval of Lottery Tickets

- Original method
 - Keep all weights with large final weights
 - Reset weights to original initial value
 - Retrain network
 - Repeat (Optional)
- Adaptation
 - "Select" weights earlier ~ develop early stopping criteria
 - Keep weights based on other metrics (Optional)

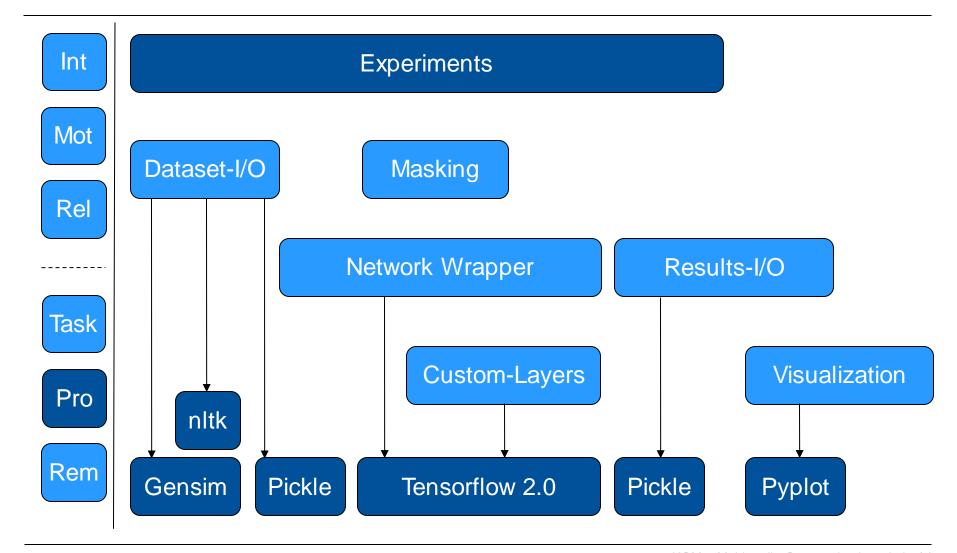
Progress – Python-project





Progress – Backend





Progress – Experiments





Mot





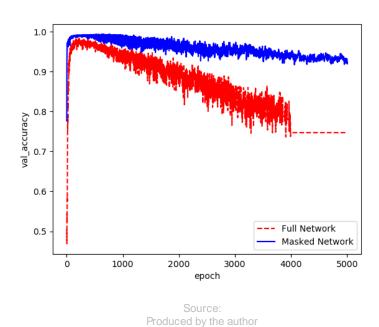


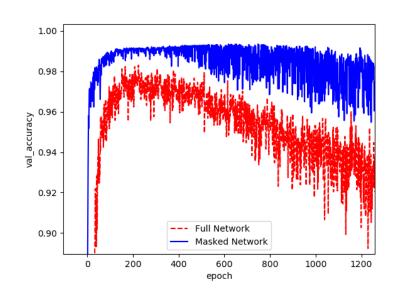




Lenet-FCN-MNIST

- Validation-Accuracy
 - 20% pruned weights





Produced by the author

Progress – Experiments





Lenet-FCN-MNIST

Training-Accuracy

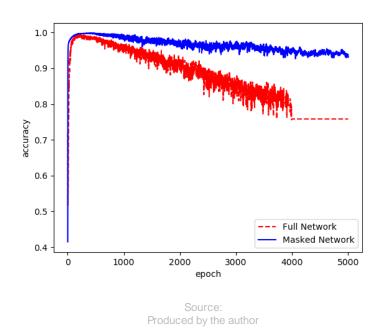
Mot

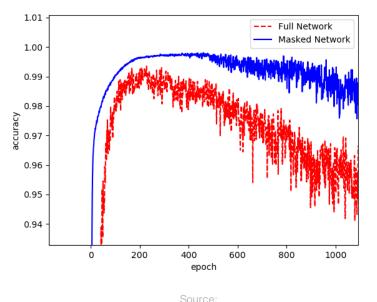
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Task



Rem





Progress – Background





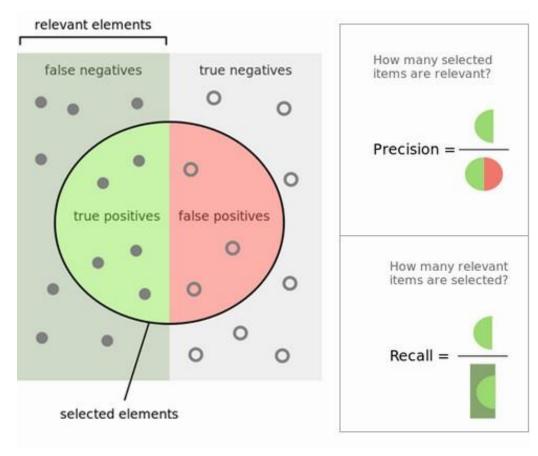












Source: https://www.kdnuggets.com/images/precision-recall-relevant-selected.jpg

Progress – Experiments





Lenet-FCN-MNIST

Validation-Recall

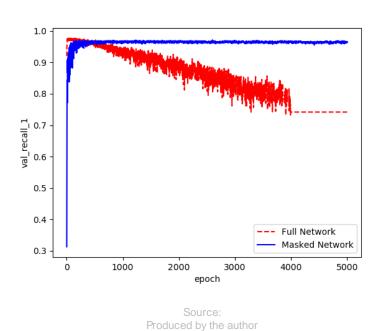
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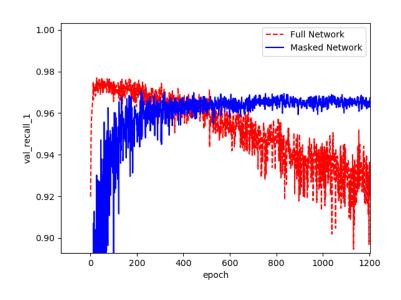
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Task

Pro

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Progress – Experiments



Int

Lenet-FCN-MNIST

Validation-Precision

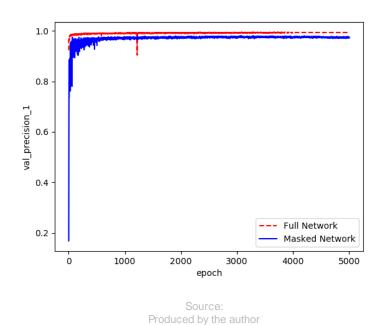
Mot

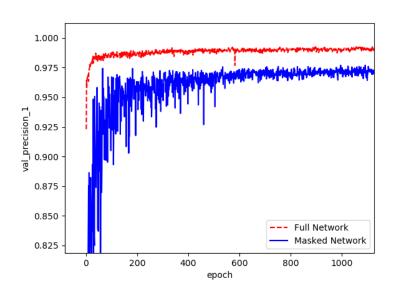
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Task



Rem





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Remaining Work

















Remaining parts of the framework

- Custom Convolutional Layer
- Support for iterative Pruning

More experiments

- MNIST / CNN-4
- MNIST / VGG-18
- Reuters / TBD
- MNIST / Lenet-FCN / Early Pruning

Thank you for your attention! Questions?



