
ℓ_1 -SVD: An Algorithm for Designing Overcomplete Dictionaries for Sparse Representation

Type Journal Article
Author M. Aharon
Author M. Elad
Author A. Bruckstein
URL <http://ieeexplore.ieee.org/document/1710377/>
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Issue 11
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Publication IEEE Transactions on Signal Processing
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Journal Abbr IEEE Trans. Signal Process.
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Language en
Short Title ℓ_1 -SVD
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Attachments

- Aharon et al. - 2006 - ℓ_1 -SVD An Algorithm for Designing Overcomplet.pdf

A Study on Overfitting in Deep Reinforcement Learning

Type Journal Article
Author Chiyuan Zhang
Author Oriol Vinyals
Author Remi Munos
Author Samy Bengio
URL <http://arxiv.org/abs/1804.06893>
Publication arXiv:1804.06893 [cs, stat]
Date 2018-04-18
Extra 00032 arXiv: 1804.06893

Accessed 6/21/2019, 12:46:15 PM

Library Catalog arXiv.org

Abstract Recent years have witnessed significant progresses in deep Reinforcement Learning (RL). Empowered with large scale neural networks, carefully designed architectures, novel training algorithms and massively parallel computing devices, researchers are able to attack many challenging RL problems. However, in machine learning, more training power comes with a potential risk of more overfitting. As deep RL techniques are being applied to critical problems such as healthcare and finance, it is important to understand the generalization behaviors of the trained agents. In this paper, we conduct a systematic study of standard RL agents and find that they could overfit in various ways. Moreover, overfitting could happen "robustly": commonly used techniques in RL that add stochasticity do not necessarily prevent or detect overfitting. In particular, the same agents and learning algorithms could have drastically different test performance, even when all of them achieve optimal rewards during training. The observations call for more principled and careful evaluation protocols in RL. We conclude with a general discussion on overfitting in RL and a study of the generalization behaviors from the perspective of inductive bias.

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Tags:

Computer Science - Machine Learning, Statistics - Machine Learning

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- arXiv.org Snapshot
- Zhang et al_2018_A Study on Overfitting in Deep Reinforcement Learning.pdf

A tutorial on statistically sound pattern discovery

Type Journal Article

Author Wilhelmiina Hämmäläinen

Author Geoffrey I. Webb

URL <http://link.springer.com/10.1007/s10618-018-0590-x>

Volume 33

Issue 2

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Publication Data Mining and Knowledge Discovery

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Journal Abbr Data Min Knowl Disc

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Language en

Abstract Statistically sound pattern discovery harnesses the rigour of statistical hypothesis testing to overcome many of the issues that have hampered standard data mining approaches to pattern discovery. Most importantly, application of appropriate statistical tests allows precise control over the risk of false discoveries—patterns that are found in the sample data but do not hold in the wider population from which the sample was drawn. Statistical tests can also be applied to filter out patterns that are unlikely to be useful, removing uninformative variations of the key patterns in the data. This tutorial introduces the key statistical and data mining theory and techniques that underpin this fast developing field. We concentrate on two general classes of patterns: dependency rules that express statistical dependencies between condition and consequent parts and dependency sets that express mutual dependence between set elements. We clarify alternative interpretations of statistical dependence and introduce appropriate tests for evaluating statistical significance of patterns in different situations. We also introduce special techniques for controlling the likelihood of spurious discoveries when multitudes of patterns are evaluated. The paper is aimed at a wide variety of audiences. It provides the necessary statistical background and summary of the state-of-the-art for any data mining researcher or practitioner wishing to enter or understand statistically sound pattern discovery research or practice. It can serve as a general introduction to the field of statistically sound pattern discovery for any reader with a general background in data sciences.

Date Added 6/21/2019, 11:46:25 AM

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Attachments

- Hämmäläinen and Webb - 2019 - A tutorial on statistically sound pattern discover.pdf

An empirical analysis of dropout in piecewise linear networks

Type Journal Article

Author David Warde-Farley

Author Ian J. Goodfellow

Author Aaron Courville

Author Yoshua Bengio

URL <http://arxiv.org/abs/1312.6197>

Publication arXiv:1312.6197 [cs, stat]

Date 2013-12-20

Extra 00053 arXiv: 1312.6197

Accessed 6/21/2019, 11:56:11 AM

Library Catalog arXiv.org

Abstract The recently introduced dropout training criterion for neural networks has been the subject of much attention due to its simplicity and remarkable effectiveness as a regularizer, as well as its interpretation as a training procedure for an exponentially large ensemble of networks that share parameters. In this work we empirically investigate several questions related to the efficacy of dropout, specifically as it concerns networks employing the popular rectified linear activation function. We investigate the quality of the test time weight-scaling inference procedure by evaluating the geometric average exactly in small models, as well as compare the performance of the geometric mean to the arithmetic mean more commonly employed by ensemble techniques. We explore the effect of tied weights on the ensemble interpretation by training ensembles of masked networks without tied weights. Finally, we investigate an alternative criterion based on a biased estimator of the maximum likelihood ensemble gradient.

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Tags:

● No DOI found, Computer Science - Neural and Evolutionary Computing, Computer Science - Machine Learning, Statistics - Machine Learning

Notes:

Comment: Extensive updates; 8 pages plus acknowledgements/references

An Introduction to Variational Autoencoders

Type Journal Article

Author Diederik P. Kingma

Author Max Welling

URL <http://arxiv.org/abs/1906.02691>

Publication arXiv:1906.02691 [cs, stat]

Date 2019-06-06

Extra 00000 arXiv: 1906.02691

Accessed 6/21/2019, 6:44:44 PM

Library Catalog arXiv.org

Abstract Variational autoencoders provide a principled framework for learning deep latent-variable models and corresponding inference models. In this work, we provide an introduction to variational autoencoders and some important extensions.

Date Added 6/21/2019, 6:44:44 PM

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Tags:

Computer Science - Machine Learning, Statistics - Machine Learning

Attachments

- arXiv.org Snapshot
- Kingma_Welling_2019_An Introduction to Variational Autoencoders.pdf

Contents

- 1 Introduction
 - 1.1 Motivation
 - 1.2 Aim
 - 1.3 Probabilistic Models and Variational Inference
 - 1.4 Parameterizing Conditional Distributions with Neural Networks
 - 1.5 Directed Graphical Models and Neural Networks
 - 1.6 Learning in Fully Observed Models with Neural Nets
 - 1.7 Learning and Inference in Deep Latent Variable Models
 - 1.8 Intractabilities
- 2 Variational Autoencoders
 - 2.1 Encoder or Approximate Posterior
 - 2.2 Evidence Lower Bound (ELBO)
 - 2.3 Stochastic Gradient-Based Optimization of the ELBO
 - 2.4 Reparameterization Trick
 - 2.5 Factorized Gaussian posteriors
 - 2.6 Estimation of the Marginal Likelihood
 - 2.7 Marginal Likelihood and ELBO as KL Divergences
 - 2.8 Challenges
 - 2.9 Related prior and concurrent work
- 3 Beyond Gaussian Posteriors
 - 3.1 Requirements for Computational Tractability
 - 3.2 Improving the Flexibility of Inference Models
 - 3.3 Inverse Autoregressive Transformations
 - 3.4 Inverse Autoregressive Flow (IAF)
 - 3.5 Related work
- 4 Deeper Generative Models
 - 4.1 Inference and Learning with Multiple Latent Variables
 - 4.2 Alternative methods for increasing expressivity of generative models
 - 4.3 Autoregressive Models
 - 4.4 Invertible transformations with tractable Jacobian determinant
 - 4.5 Follow-Up Work
- 5 Conclusion
- 6 Appendix
 - 6.1 Notation and definitions
 - 6.2 Alternative methods for learning in DLVMs
 - 6.3 Stochastic Gradient Descent
- References

Attention Is All You Need

Type Journal Article
Author Ashish Vaswani
Author Noam Shazeer
Author Niki Parmar
Author Jakob Uszkoreit
Author Llion Jones
Author Aidan N. Gomez
Author Lukasz Kaiser
Author Illia Polosukhin
URL <http://arxiv.org/abs/1706.03762>
Publication arXiv:1706.03762 [cs]
Date 2017-06-12
Extra 02136 arXiv: 1706.03762
Accessed 6/21/2019, 5:04:35 PM
Library Catalog arXiv.org
Abstract The dominant sequence transduction models are based on complex recurrent or convolutional neural networks in an encoder-decoder configuration. The best performing models also connect the encoder and decoder through an attention mechanism. We propose a new simple network architecture, the Transformer, based solely on attention mechanisms, dispensing with recurrence and convolutions entirely. Experiments on two machine translation tasks show these models to be superior in quality while being more parallelizable and requiring significantly less time to train. Our model achieves 28.4 BLEU on the WMT 2014 English-to-German translation task, improving over the existing best results, including ensembles by over 2 BLEU. On the WMT 2014 English-to-French translation task, our model establishes a new single-model state-of-the-art BLEU score of 41.8 after training for 3.5 days on eight GPUs, a small fraction of the training costs of the best models from the literature. We show that the Transformer generalizes well to other tasks by applying it successfully to English constituency parsing both with large and limited training data.
Date Added 6/21/2019, 5:04:35 PM
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Tags:

Computer Science - Computation and Language, Computer Science - Machine Learning

Notes:

Comment: 15 pages, 5 figures

Attachments

- arXiv.org Snapshot
- Vaswani et al_2017_Attention Is All You Need.pdf

Contents

- 1 Introduction
- 2 Background
- 3 Model Architecture
 - 3.1 Encoder and Decoder Stacks
 - 3.2 Attention
 - 3.2.1 Scaled Dot-Product Attention
 - 3.2.2 Multi-Head Attention
 - 3.2.3 Applications of Attention in our Model
 - 3.3 Position-wise Feed-Forward Networks
 - 3.4 Embeddings and Softmax
 - 3.5 Positional Encoding
- 4 Why Self-Attention
- 5 Training
 - 5.1 Training Data and Batching
 - 5.2 Hardware and Schedule
 - 5.3 Optimizer
 - 5.4 Regularization
- 6 Results
 - 6.1 Machine Translation
 - 6.2 Model Variations
 - 6.3 English Constituency Parsing
- 7 Conclusion

Attentive Neural Processes

Type Journal Article
Author Hyunjik Kim
Author Andriy Mnih
Author Jonathan Schwarz
Author Marta Garnelo
Author Ali Eslami
Author Dan Rosenbaum
Author Oriol Vinyals
Author Yee Whye Teh
URL <http://arxiv.org/abs/1901.05761>

Publication arXiv:1901.05761 [cs, stat]

Date 2019-01-17

Extra 00007 arXiv: 1901.05761

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Library Catalog arXiv.org

Abstract Neural Processes (NPs) (Garnelo et al 2018a;b) approach regression by learning to map a context set of observed input-output pairs to a distribution over regression functions. Each function models the distribution of the output given an input, conditioned on the context. NPs have the benefit of fitting observed data efficiently with linear complexity in the number of context input-output pairs, and can learn a wide family of conditional distributions; they learn predictive distributions conditioned on context sets of arbitrary size. Nonetheless, we show that NPs suffer a fundamental drawback of underfitting, giving inaccurate predictions at the inputs of the observed data they condition on. We address this issue by incorporating attention into NPs, allowing each input location to attend to the relevant context points for the prediction. We show that this greatly improves the accuracy of predictions, results in noticeably faster training, and expands the range of functions that can be modelled.

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Tags:

Computer Science - Machine Learning, Statistics - Machine Learning

Attachments

- arXiv.org Snapshot
- Kim et al_2019_Attentive Neural Processes.pdf

Contents

- 1 Introduction
- 2 Background
 - 2.1 Neural Processes
 - 2.2 Attention
- 3 Attentive Neural Processes
- 4 Experimental Results
- 5 Related Work
- 6 Conclusion and Discussion
- A Architectural details for (A)NP
- B Experimental details of 1D function regression experiment
- C Additional figures for 1D regression on GP data
- D Experimental details of 2D Image regression experiment
- E Additional figures for 2D Image regression on MNIST and CelebA

Auto-Encoding Variational Bayes

- Type** Journal Article
- Author** Diederik P. Kingma
- Author** Max Welling
- URL** <http://arxiv.org/abs/1312.6114>
- Publication** arXiv:1312.6114 [cs, stat]
- Date** 2013-12-20
- Extra** 05075 arXiv: 1312.6114
- Accessed** 6/21/2019, 5:57:38 PM
- Library Catalog** arXiv.org
- Abstract** How can we perform efficient inference and learning in directed probabilistic models, in the presence of continuous latent variables with intractable posterior distributions, and large datasets? We introduce a stochastic variational inference and learning algorithm that scales to large datasets and, under some mild differentiability conditions, even works in the intractable case. Our contributions is two-fold. First, we show that a reparameterization of the variational lower bound yields a lower bound estimator that can be straightforwardly optimized using standard stochastic gradient methods. Second, we show that for i.i.d. datasets with continuous latent variables per datapoint, posterior inference can be made especially efficient by fitting an approximate inference model (also called a recognition model) to the intractable posterior using the proposed lower bound estimator. Theoretical advantages are reflected in experimental results.

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Tags:

Computer Science - Machine Learning, Statistics - Machine Learning

Attachments

- arXiv.org Snapshot
- Kingma_Welling_2013_Auto-Encoding Variational Bayes.pdf

Contents

- 1 Introduction
- 2 Method
 - 2.1 Problem scenario
 - 2.2 The variational bound
 - 2.3 The SGVB estimator and AEVB algorithm
 - 2.4 The reparameterization trick
- 3 Example: Variational Auto-Encoder
- 4 Related work
- 5 Experiments
- 6 Conclusion
- 7 Future work
- A Visualisations
- B Solution of - $DKL(q_{\boldsymbol{\mu}} \parallel p_{\boldsymbol{\mu}})$ 2005/06/28 ver: 1.3 subfig package(z) ||
 $p_{\boldsymbol{\mu}}$ 2005/06/28 ver: 1.3 subfig package(z)), Gaussian case
- C MLP's as probabilistic encoders and decoders
 - C.1 Bernoulli MLP as decoder
 - C.2 Gaussian MLP as encoder or decoder
- D Marginal likelihood estimator
- E Monte Carlo EM
- F Full VB
 - F.1 Example

Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift

Type Journal Article

Author Sergey Ioffe

Author Christian Szegedy
Pages 9
Library Catalog Zotero
Language en
Abstract Training Deep Neural Networks is complicated by the fact that the distribution of each layer's inputs changes during training, as the parameters of the previous layers change. This slows down the training by requiring lower learning rates and careful parameter initialization, and makes it notoriously hard to train models with saturating nonlinearities. We refer to this phenomenon as internal covariate shift, and address the problem by normalizing layer inputs. Our method draws its strength from making normalization a part of the model architecture and performing the normalization for each training mini-batch. Batch Normalization allows us to use much higher learning rates and be less careful about initialization, and in some cases eliminates the need for Dropout. Applied to a state-of-the-art image classification model, Batch Normalization achieves the same accuracy with 14 times fewer training steps, and beats the original model by a significant margin. Using an ensemble of batch-normalized networks, we improve upon the best published result on ImageNet classification: reaching 4.82% top-5 test error, exceeding the accuracy of human raters.

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Tags:

☹ No DOI found

Attachments

- Ioffe and Szegedy - Batch Normalization Accelerating Deep Network Tra.pdf

Convolutional Neural Networks for Sentence Classification

Type Journal Article
Author Yoon Kim
URL <http://arxiv.org/abs/1408.5882>
Publication arXiv:1408.5882 [cs]
Date 2014-08-25
Extra 04407 arXiv: 1408.5882
Accessed 6/21/2019, 11:30:28 AM
Library Catalog arXiv.org
Language en
Abstract We report on a series of experiments with convolutional neural networks (CNN) trained on top of pre-trained word vectors for sentence-level

classification tasks. We show that a simple CNN with little hyperparameter tuning and static vectors achieves excellent results on multiple benchmarks. Learning task-specific vectors through fine-tuning offers further gains in performance. We additionally propose a simple modification to the architecture to allow for the use of both task-specific and static vectors. The CNN models discussed herein improve upon the state of the art on 4 out of 7 tasks, which include sentiment analysis and question classification.

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Tags:

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- Kim - 2014 - Convolutional Neural Networks for Sentence Classif.pdf

Convolutional Sequence to Sequence Learning

Type Journal Article
Author Jonas Gehring
Author Michael Auli
Author David Grangier
Author Denis Yarats
Author Yann N. Dauphin
URL <http://arxiv.org/abs/1705.03122>
Publication arXiv:1705.03122 [cs]
Date 2017-05-08
Extra 00688 arXiv: 1705.03122
Accessed 6/21/2019, 5:04:24 PM
Library Catalog arXiv.org
Abstract The prevalent approach to sequence to sequence learning maps an input sequence to a variable length output sequence via recurrent neural networks. We introduce an architecture based entirely on convolutional neural networks. Compared to recurrent models, computations over all elements can be fully parallelized during training and optimization is easier since the number of non-linearities is fixed and independent of the input length. Our

use of gated linear units eases gradient propagation and we equip each decoder layer with a separate attention module. We outperform the accuracy of the deep LSTM setup of Wu et al. (2016) on both WMT'14 English-German and WMT'14 English-French translation at an order of magnitude faster speed, both on GPU and CPU.

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Tags:

Computer Science - Computation and Language

Attachments

- arXiv.org Snapshot
- Gehring et al_2017_Convolutional Sequence to Sequence Learning.pdf

Deep learning

Type Journal Article

Author Yann LeCun

Author Yoshua Bengio

Author Geoffrey Hinton

URL <http://www.nature.com/articles/nature14539>

Volume 521

Issue 7553

Pages 436-444

Publication Nature

ISSN 0028-0836, 1476-4687

Date 5/2015

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Journal Abbr Nature

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Language en

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- LeCun et al. - 2015 - Deep learning.pdf

Design challenges and misconceptions in named entity recognition

Type Conference Paper
Author Lev Ratinov
Author Dan Roth
URL <http://portal.acm.org/citation.cfm?doid=1596374.1596399>
Place Boulder, Colorado
Publisher Association for Computational Linguistics
Pages 147
ISBN 978-1-932432-29-9
Date 2009
DOI 10/dkk8vf
Accessed 6/21/2019, 1:35:08 PM
Library Catalog DOI.org (Crossref)
Conference Name the Thirteenth Conference
Language en
Abstract We analyze some of the fundamental design challenges and misconceptions that underlie the development of an efficient and robust NER system. In particular, we address issues such as the representation of text chunks, the inference approach needed to combine local NER decisions, the sources of prior knowledge and how to use them within an NER system. In the process of comparing several solutions to these challenges we reach some surprising conclusions, as well as develop an NER system that achieves 90.8 F1 score on the CoNLL-2003 NER shared task, the best reported result for this dataset.
Proceedings Title Proceedings of the Thirteenth Conference on Computational Natural Language Learning - CoNLL '09
Date Added 6/21/2019, 1:35:08 PM
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Attachments

- Ratinov and Roth - 2009 - Design challenges and misconceptions in named enti.pdf

Dropout: A Simple Way to Prevent Neural Networks from Overfitting

Type Journal Article
Author Nitish Srivastava
Author Geoffrey Hinton
Author Alex Krizhevsky
Author Ilya Sutskever
Author Ruslan Salakhutdinov

Pages 30

Extra 12530

Library Catalog Zotero

Language en

Abstract Deep neural nets with a large number of parameters are very powerful machine learning systems. However, overfitting is a serious problem in such networks. Large networks are also slow to use, making it difficult to deal with overfitting by combining the predictions of many different large neural nets at test time. Dropout is a technique for addressing this problem. The key idea is to randomly drop units (along with their connections) from the neural network during training. This prevents units from co-adapting too much. During training, dropout samples from an exponential number of different “thinned” networks. At test time, it is easy to approximate the effect of averaging the predictions of all these thinned networks by simply using a single unthinned network that has smaller weights. This significantly reduces overfitting and gives major improvements over other regularization methods. We show that dropout improves the performance of neural networks on supervised learning tasks in vision, speech recognition, document classification and computational biology, obtaining state-of-the-art results on many benchmark data sets.

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Tags:

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- Srivastava et al. - Dropout A Simple Way to Prevent Neural Networks f.pdf

Efficient Object Localization Using Convolutional Networks

Type Journal Article

Author Jonathan Tompson

Author Ross Goroshin

Author Arjun Jain

Author Yann LeCun

Author Christopher Bregler

URL <http://arxiv.org/abs/1411.4280>

Publication arXiv:1411.4280 [cs]

Date 2014-11-16

Extra 00392 arXiv: 1411.4280

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Library Catalog arXiv.org**Language** en

Abstract Recent state-of-the-art performance on human-body pose estimation has been achieved with Deep Convolutional Networks (ConvNets). Traditional ConvNet architectures include pooling and sub-sampling layers which reduce computational requirements, introduce invariance and prevent over-training. These benefits of pooling come at the cost of reduced localization accuracy. We introduce a novel architecture which includes an efficient ‘position refinement’ model that is trained to estimate the joint offset location within a small region of the image. This refinement model is jointly trained in cascade with a state-of-the-art ConvNet model [21] to achieve improved accuracy in human joint location estimation. We show that the variance of our detector approaches the variance of human annotations on the FLIC [20] dataset and outperforms all existing approaches on the MPII-human-pose dataset [1].

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- Thompson et al. - 2014 - Efficient Object Localization Using Convolutional .pdf

Explaining Landscape Connectivity of Low-cost Solutions for Multilayer Nets

Type Journal Article
Author Rohith Kuditipudi
Author Xiang Wang
Author Holden Lee
Author Yi Zhang
Author Zhiyuan Li
Author Wei Hu
Author Sanjeev Arora
Author Rong Ge
URL <http://arxiv.org/abs/1906.06247>

Publication arXiv:1906.06247 [cs, stat]
Date 2019-06-14
Extra 00000 arXiv: 1906.06247
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Library Catalog arXiv.org
Abstract Mode connectivity is a surprising phenomenon in the loss landscape of deep nets. Optima---at least those discovered by gradient-based optimization---turn out to be connected by simple paths on which the loss function is almost constant. Often, these paths can be chosen to be piece-wise linear, with as few as two segments. We give mathematical explanations for this phenomenon, assuming generic properties (such as dropout stability and noise stability) of well-trained deep nets, which have previously been identified as part of understanding the generalization properties of deep nets. Our explanation holds for realistic multilayer nets, and experiments are presented to verify the theory.
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Tags:

Computer Science - Machine Learning, Statistics - Machine Learning

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- arXiv.org Snapshot
- Kuditipudi et al_2019_Explaining Landscape Connectivity of Low-cost Solutions for Multilayer Nets.pdf

Contents

- 1 Introduction
 - 1.1 Related work
- 2 Preliminaries
- 3 Connectivity of dropout-stable optima
- 4 Connectivity via noise stability
 - 4.1 Noise stability
 - 4.2 Noise stability implies dropout stability
- 5 Disconnected modes in two-layer nets
- 6 Experiments
- A Proofs for connectivity of dropout-stable optima
- B Proofs for connectivity via noise stability
- C Proofs for disconnected modes in two-layer nets
- D Experimental details and further results
 - D.1 Experimental details and hyperparameters
 - D.2 Verification of noise stability conditions
 - D.2.1 Layer cushion
 - D.2.2 Interlayer cushion
 - D.2.3 Activation contraction
 - D.2.4 Interlayer smoothness
- E Tools

Expressiveness of Rectifier Networks

Type	Journal Article
Author	Xingyuan Pan
Author	Vivek Srikumar
Pages	9
Extra	00022
Library Catalog	Zotero
Language	en
Abstract	Rectified Linear Units (ReLU) have been shown to ameliorate the vanishing gradient problem, allow for efficient backpropagation, and empirically promote sparsity in the learned parameters. They have led to state-of-the-art results in a variety of applications. However, unlike threshold and sigmoid networks, ReLU networks are less explored from the perspective of their

expressiveness. This paper studies the expressiveness of ReLU networks. We characterize the decision boundary of two-layer ReLU networks by constructing functionally equivalent threshold networks. We show that while the decision boundary of a two-layer ReLU network can be captured by a threshold network, the latter may require an exponentially larger number of hidden units. We also formulate sufficient conditions for a corresponding logarithmic reduction in the number of hidden units to represent a sign network as a ReLU network. Finally, we experimentally compare threshold networks and their much smaller ReLU counterparts with respect to their ability to learn from synthetically generated data.

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- Pan and Srikumar - Expressiveness of Rectifier Networks.pdf

FCNN: Fourier Convolutional Neural Networks

Type Book Section
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Editor Jaakko Hollmén
Editor Ljupčo Todorovski
Editor Celine Vens
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Author Bryan Williams
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ISBN 978-3-319-71248-2 978-3-319-71249-9
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Extra DOI: 10.1007/978-3-319-71249-9_47
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Language en

Abstract The Fourier domain is used in computer vision and machine learning as image analysis tasks in the Fourier domain are analogous to spatial domain methods but are achieved using different operations. Convolutional Neural Networks (CNNs) use machine learning to achieve state-of-the-art results with respect to many computer vision tasks. One of the main limiting aspects of CNNs is the computational cost of updating a large number of convolution parameters. Further, in the spatial domain, larger images take exponentially longer than smaller image to train on CNNs due to the operations involved in convolution methods. Consequently, CNNs are often not a viable solution for large image computer vision tasks. In this paper a Fourier Convolution Neural Network (FCNN) is proposed whereby training is conducted entirely within the Fourier domain. The advantage offered is that there is a significant speed up in training time without loss of effectiveness. Using the proposed approach larger images can therefore be processed within viable computation time. The FCNN is fully described and evaluated. The evaluation was conducted using the benchmark Cifar10 and MNIST datasets, and a bespoke fundus retina image dataset. The results demonstrate that convolution in the Fourier domain gives a significant speed up without adversely affecting accuracy. For simplicity the proposed FCNN concept is presented in the context of a basic CNN architecture, however, the FCNN concept has the potential to improve the speed of any neural network system involving convolution.

Book Title Machine Learning and Knowledge Discovery in Databases**Short Title** FCNN**Date Added** 6/21/2019, 11:33:35 AM**Modified** 6/21/2019, 11:33:35 AM**Attachments**

- Pratt et al. - 2017 - FCNN Fourier Convolutional Neural Networks.pdf

Forgetting Memories and Their Attractiveness
Type Journal Article**Author** Enzo Marinari**URL** https://www.mitpressjournals.org/doi/10.1162/neco_a_01162**Volume** 31**Issue** 3**Pages** 503-516**Publication** Neural Computation**ISSN** 0899-7667, 1530-888X**Date** 03/2019**Extra** 00001**Journal Abbr** Neural Computation

DOI 10/gf35pq
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Library Catalog DOI.org (Crossref)
Language en
Abstract We study numerically the memory that forgets, introduced in 1986 by Parisi by bounding the synaptic strength, with a mechanism that avoids confusion; allows remembering the pattern learned more recently; and has a physiologically very well-defined meaning. We analyze a number of features of this learning for a finite number of neurons and finite number of patterns. We discuss how the system behaves in the large but finite [Formula: see text] limit. We analyze the basin of attraction of the patterns that have been learned, and we show that it is exponentially small in the age of the pattern.
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Attachments

- Marinari - 2019 - Forgetting Memories and Their Attractiveness.pdf

Generative Adversarial Networks

Type Journal Article
Author Ian J. Goodfellow
Author Jean Pouget-Abadie
Author Mehdi Mirza
Author Bing Xu
Author David Warde-Farley
Author Sherjil Ozair
Author Aaron Courville
Author Yoshua Bengio
URL <https://arxiv.org/abs/1406.2661v1>
Date 2014/06/10
Extra 00485
Accessed 6/21/2019, 11:55:55 AM
Library Catalog arxiv.org
Language en
Abstract We propose a new framework for estimating generative models via an adversarial process, in which we simultaneously train two models: a generative model G that captures the data distribution, and a discriminative model D that estimates the probability that a sample came from the training data rather than G . The training procedure for G is to maximize the probability of D making a mistake. This framework corresponds to a minimax two-player game. In the space of arbitrary functions G and D , a

unique solution exists, with G recovering the training data distribution and D equal to 1/2 everywhere. In the case where G and D are defined by multilayer perceptrons, the entire system can be trained with backpropagation. There is no need for any Markov chains or unrolled approximate inference networks during either training or generation of samples. Experiments demonstrate the potential of the framework through qualitative and quantitative evaluation of the generated samples.

Date Added 6/21/2019, 11:55:55 AM

Modified 6/21/2019, 12:05:26 PM

Attachments

- Goodfellow et al_2014_Generative Adversarial Networks.pdf
- Snapshot

Glove: Global Vectors for Word Representation

Type Conference Paper

Author Jeffrey Pennington

Author Richard Socher

Author Christopher Manning

URL <http://aclweb.org/anthology/D14-1162>

Place Doha, Qatar

Publisher Association for Computational Linguistics

Pages 1532-1543

Date 2014

Extra 08011

DOI 10/gfshwg

Accessed 6/21/2019, 11:34:26 AM

Library Catalog DOI.org (Crossref)

Conference Name Proceedings of the 2014 Conference on Empirical Methods in Natural Language Processing (EMNLP)

Language en

Abstract Recent methods for learning vector space representations of words have succeeded in capturing fine-grained semantic and syntactic regularities using vector arithmetic, but the origin of these regularities has remained opaque. We analyze and make explicit the model properties needed for such regularities to emerge in word vectors. The result is a new global logbilinear regression model that combines the advantages of the two major model families in the literature: global matrix factorization and local context window methods. Our model efficiently leverages statistical information by training only on the nonzero elements in a word-word cooccurrence matrix, rather than on the entire sparse matrix or on individual context windows in a large corpus. The model produces a vector

space with meaningful substructure, as evidenced by its performance of 75% on a recent word analogy task. It also outperforms related models on similarity tasks and named entity recognition.

Proceedings Title Proceedings of the 2014 Conference on Empirical Methods in Natural Language Processing (EMNLP)

Short Title Glove

Date Added 6/21/2019, 11:34:26 AM

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Attachments

- Pennington et al. - 2014 - Glove Global Vectors for Word Representation.pdf

Google's Neural Machine Translation System: Bridging the Gap between Human and Machine Translation

Type Journal Article

Author Yonghui Wu

Author Mike Schuster

Author Zhifeng Chen

Author Quoc V. Le

Author Mohammad Norouzi

Author Wolfgang Macherey

Author Maxim Krikun

Author Yuan Cao

Author Qin Gao

Author Klaus Macherey

Author Jeff Klingner

Author Apurva Shah

Author Melvin Johnson

Author Xiaobing Liu

Author Łukasz Kaiser

Author Stephan Gouws

Author Yoshikiyo Kato

Author Taku Kudo

Author Hideto Kazawa

Author Keith Stevens

Author George Kurian

Author Nishant Patil

Author Wei Wang

Author Cliff Young

Author Jason Smith

Author Jason Riesa

Author Alex Rudnick

Author Oriol Vinyals

Author Greg Corrado

Author Macduff Hughes

Author Jeffrey Dean

URL <http://arxiv.org/abs/1609.08144>

Publication arXiv:1609.08144 [cs]

Date 2016-09-26

Extra 01464 arXiv: 1609.08144

Accessed 6/21/2019, 5:03:58 PM

Library Catalog arXiv.org

Abstract Neural Machine Translation (NMT) is an end-to-end learning approach for automated translation, with the potential to overcome many of the weaknesses of conventional phrase-based translation systems. Unfortunately, NMT systems are known to be computationally expensive both in training and in translation inference. Also, most NMT systems have difficulty with rare words. These issues have hindered NMT's use in practical deployments and services, where both accuracy and speed are essential. In this work, we present GNMT, Google's Neural Machine Translation system, which attempts to address many of these issues. Our model consists of a deep LSTM network with 8 encoder and 8 decoder layers using attention and residual connections. To improve parallelism and therefore decrease training time, our attention mechanism connects the bottom layer of the decoder to the top layer of the encoder. To accelerate the final translation speed, we employ low-precision arithmetic during inference computations. To improve handling of rare words, we divide words into a limited set of common sub-word units ("wordpieces") for both input and output. This method provides a good balance between the flexibility of "character"-delimited models and the efficiency of "word"-delimited models, naturally handles translation of rare words, and ultimately improves the overall accuracy of the system. Our beam search technique employs a length-normalization procedure and uses a coverage penalty, which encourages generation of an output sentence that is most likely to cover all the words in the source sentence. On the WMT'14 English-to-French and English-to-German benchmarks, GNMT achieves competitive results to state-of-the-art. Using a human side-by-side evaluation on a set of isolated simple sentences, it reduces translation errors by an average of 60% compared to Google's phrase-based production system.

Short Title Google's Neural Machine Translation System

Date Added 6/21/2019, 5:03:58 PM

Modified 6/21/2019, 5:04:17 PM

Tags:

Computer Science - Computation and Language, Computer Science - Machine Learning,
Computer Science - Artificial Intelligence

Attachments

- arXiv.org Snapshot
- Wu et al_2016_Google's Neural Machine Translation System.pdf

Contents

- 1 Introduction
- 2 Related Work
- 3 Model Architecture
 - 3.1 Residual Connections
 - 3.2 Bi-directional Encoder for First Layer
 - 3.3 Model Parallelism
- 4 Segmentation Approaches
 - 4.1 Wordpiece Model
 - 4.2 Mixed Word/Character Model
- 5 Training Criteria
- 6 Quantizable Model and Quantized Inference
- 7 Decoder
- 8 Experiments and Results
 - 8.1 Datasets
 - 8.2 Evaluation Metrics
 - 8.3 Training Procedure
 - 8.4 Evaluation after Maximum Likelihood Training
 - 8.5 Evaluation of RL-refined Models
 - 8.6 Model Ensemble and Human Evaluation
 - 8.7 Results on Production Data
- 9 Conclusion

Graphical Models, Exponential Families, and Variational Inference

Type Journal Article
Author Martin J. Wainwright
Author Michael I. Jordan
URL <http://www.nowpublishers.com/article/Details/MAL-001>
Volume 1
Issue 1–2
Pages 1-305

Publication Foundations and Trends® in Machine Learning
ISSN 1935-8237, 1935-8245
Date 2007
Extra 03181
Journal Abbr FNT in Machine Learning
DOI 10/bpnwrm
Accessed 6/21/2019, 6:23:56 PM
Library Catalog DOI.org (Crossref)
Language en
Date Added 6/21/2019, 6:23:56 PM
Modified 6/21/2019, 6:23:58 PM

Attachments

- Wainwright and Jordan - 2007 - Graphical Models, Exponential Families, and Variat.pdf

Introduction to Information Retrieval

Type Journal Article
Author Christopher Manning
Author Prabhakar Raghavan
Author Hinrich Schuetze
Pages 581
Date 2009
Extra 16180
Library Catalog Zotero
Language en
Date Added 6/21/2019, 2:01:47 PM
Modified 6/21/2019, 2:01:50 PM

Tags:

- No DOI found

Attachments

- Manning et al. - 2009 - Introduction to Information Retrieval.pdf

lecun-89.pdf

Type Attachment

Date Added 6/21/2019, 11:45:20 AM

Modified 6/21/2019, 11:45:20 AM

Long Short-Term Memory

Type Journal Article

Author Sepp Hochreiter

Author Schmidhuber, Jurgen

Volume 9

Pages 1735-1780

Publication Neural Computation

Extra 19087

DOI 10/bxd65w

Date Added 6/21/2019, 12:51:09 PM

Modified 6/21/2019, 12:53:23 PM

Attachments

- Hochreiter and Schmidhuber, Jurgen - Long Short-Term Memory.pdf

Maxout Networks

Type Journal Article

Author Ian J. Goodfellow

Author David Warde-Farley

Author Mehdi Mirza

Author Aaron Courville

Author Yoshua Bengio

URL <http://arxiv.org/abs/1302.4389>

Publication arXiv:1302.4389 [cs, stat]

Date 2013-02-18

Extra 01610 arXiv: 1302.4389

Accessed 6/21/2019, 11:53:02 AM

Library Catalog arXiv.org

Abstract We consider the problem of designing models to leverage a recently introduced approximate model averaging technique called dropout. We define a simple new model called maxout (so named because its output is the max of a set of inputs, and because it is a natural companion to dropout) designed to both facilitate optimization by dropout and improve the accuracy of dropout's fast approximate model averaging technique. We empirically verify that the model successfully accomplishes both of these

tasks. We use maxout and dropout to demonstrate state of the art classification performance on four benchmark datasets: MNIST, CIFAR-10, CIFAR-100, and SVHN.

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Tags:

Computer Science - Machine Learning, Statistics - Machine Learning

Notes:

Comment: This is the version of the paper that appears in ICML 2013

Attachments

- Goodfellow et al_2013_Maxout Networks.pdf

Neocognitron: A hierarchical neural network capable of visual pattern recognition

Type Journal Article
Author Kuniyiko Fukushima
URL <https://linkinghub.elsevier.com/retrieve/pii/0893608088900147>
Volume 1
Issue 2
Pages 119-130
Publication Neural Networks
ISSN 08936080
Date 1/1988
Extra 00924
Journal Abbr Neural Networks
DOI 10/bw8fwf
Accessed 6/21/2019, 11:33:40 AM
Library Catalog DOI.org (Crossref)
Language en
Abstract A neural network model for visual pattern recognition, called the "neocognitron," was previously proposed by the author. In this paper, we discuss the mechanism of the model in detail. In order to demonstrate the ability of the neocognitron, we also discuss a pattern-recognition system which works with the mechanism of the neocognitron.
Short Title Neocognitron

Date Added 6/21/2019, 11:33:40 AM**Modified** 6/21/2019, 11:33:42 AM**Attachments**

- Fukushima - 1988 - Neocognitron A hierarchical neural network capabl.pdf

Neocognitron: A self-organizing neural network model for a mechanism of pattern recognition unaffected by shift in position

Type Journal Article

Author Kunihiro Fukushima

URL <http://link.springer.com/10.1007/BF00344251>

Volume 36

Issue 4

Pages 193-202

Publication Biological Cybernetics

ISSN 0340-1200, 1432-0770

Date 4/1980

Extra 03322

Journal Abbr Biol. Cybernetics

DOI 10/fhrjiv

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Library Catalog DOI.org (Crossref)

Language en

Abstract A neural network model for a mechanism of visual pattern recognition is proposed in this paper. The network is self-organized by "learning without a teacher", and acquires an ability to recognize stimulus patterns based on the geometrical similarity (Gestalt) of their shapes without affected by their positions. This network is given a nickname "neocognitron". After completion of self-organization, the network has a structure similar to the hierarchy model of the visual nervous system proposed by Hubel and Wiesel. The network consists of an input layer (photoreceptor array) followed by a cascade connection of a number of modular structures, each of which is composed of two layers of cells connected in a cascade. The first layer of each module consists of "S-cells", which show characteristics similar to simple cells or lower order hypercomplex cells, and the second layer consists of "C-cells" similar to complex cells or higher order hypercomplex cells. The afferent synapses to each S-cell have plasticity and are modifiable. The network has an ability of unsupervised learning: We do not need any "teacher" during the process of selforganization, and it is only needed to present a set of stimulus patterns repeatedly to the input layer of the network. The network has been simulated on a digital computer. After repetitive presentation of a set of stimulus patterns, each stimulus pattern

has become to elicit an output only from one of the C-cells of the last layer, and conversely, this C-cell has become selectively responsive only to that stimulus pattern. That is, none of the C-cells of the last layer responds to more than one stimulus pattern. The response of the C-cells of the last layer is not affected by the pattern's position at all. Neither is it affected by a small change in shape nor in size of the stimulus pattern.

Short Title Neocognitron

Date Added 6/21/2019, 11:33:47 AM

Modified 6/21/2019, 11:33:48 AM

Attachments

- Fukushima - 1980 - Neocognitron A self-organizing neural network mod.pdf

Neural Machine Translation by Jointly Learning to Align and Translate

Type Journal Article

Author Dzmitry Bahdanau

Author Kyunghyun Cho

Author Yoshua Bengio

URL <http://arxiv.org/abs/1409.0473>

Publication arXiv:1409.0473 [cs, stat]

Date 2014-09-01

Extra 07290 arXiv: 1409.0473

Accessed 6/21/2019, 5:03:31 PM

Library Catalog arXiv.org

Abstract Neural machine translation is a recently proposed approach to machine translation. Unlike the traditional statistical machine translation, the neural machine translation aims at building a single neural network that can be jointly tuned to maximize the translation performance. The models proposed recently for neural machine translation often belong to a family of encoder-decoders and consists of an encoder that encodes a source sentence into a fixed-length vector from which a decoder generates a translation. In this paper, we conjecture that the use of a fixed-length vector is a bottleneck in improving the performance of this basic encoder-decoder architecture, and propose to extend this by allowing a model to automatically (soft-)search for parts of a source sentence that are relevant to predicting a target word, without having to form these parts as a hard segment explicitly. With this new approach, we achieve a translation performance comparable to the existing state-of-the-art phrase-based system on the task of English-to-French translation. Furthermore, qualitative analysis reveals that the (soft-)alignments found by the model agree well with our intuition.

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Tags:

Computer Science - Computation and Language, Computer Science - Neural and Evolutionary Computing, Computer Science - Machine Learning, Statistics - Machine Learning

Notes:

Comment: Accepted at ICLR 2015 as oral presentation

Attachments

- arXiv.org Snapshot
- Bahdanau et al_2014_Neural Machine Translation by Jointly Learning to Align and Translate.pdf

Contents

- 1 Introduction
- 2 Background: Neural Machine Translation
 - 2.1 RNN Encoder–Decoder
- 3 Learning to Align and Translate
 - 3.1 Decoder: General Description
 - 3.2 Encoder: Bidirectional RNN for Annotating Sequences
- 4 Experiment Settings
 - 4.1 Dataset
 - 4.2 Models
- 5 Results
 - 5.1 Quantitative Results
 - 5.2 Qualitative Analysis
 - 5.2.1 Alignment
 - 5.2.2 Long Sentences
- 6 Related Work
 - 6.1 Learning to Align
 - 6.2 Neural Networks for Machine Translation
- 7 Conclusion
- A Model Architecture
 - A.1 Architectural Choices
 - A.1.1 Recurrent Neural Network
 - A.1.2 Alignment Model
 - A.2 Detailed Description of the Model
 - A.2.1 Encoder
 - A.2.2 Decoder
 - A.2.3 Model Size
- B Training Procedure
 - B.1 Parameter Initialization
 - B.2 Training
- C Translations of Long Sentences

Neural Transfer Learning for Natural Language Processing

Type Journal Article
Author Sebastian Ruder
Pages 329
Extra 00002
Library Catalog Zotero
Language en
Date Added 6/21/2019, 11:34:13 AM
Modified 6/21/2019, 11:34:18 AM

Tags:

⊖ No DOI found

Attachments

- Ruder - Neural Transfer Learning for Natural Language Proc.pdf

Pattern recognition and machine learning

Type Book
Author Christopher M. Bishop
Series Information science and statistics
Place New York
Publisher Springer
ISBN 978-0-387-31073-2
Date 2006
Call Number Q327 .B52 2006
Extra 36259
Library Catalog Library of Congress ISBN
Language en
of Pages 738
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Modified 6/21/2019, 2:08:43 PM

Tags:

Machine learning, Pattern perception

Attachments

- Bishop - 2006 - Pattern recognition and machine learning.pdf

Random Search for Hyper-Parameter Optimization

Type Journal Article
Author James Bergstra
Author Yoshua Bengio
Pages 25
Extra 02538
Library Catalog Zotero
Language en

Abstract Grid search and manual search are the most widely used strategies for hyper-parameter optimization. This paper shows empirically and theoretically that randomly chosen trials are more efficient for hyper-parameter optimization than trials on a grid. Empirical evidence comes from a comparison with a large previous study that used grid search and manual search to configure neural networks and deep belief networks. Compared with neural networks configured by a pure grid search, we find that random search over the same domain is able to find models that are as good or better within a small fraction of the computation time. Granting random search the same computational budget, random search finds better models by effectively searching a larger, less promising configuration space. Compared with deep belief networks configured by a thoughtful combination of manual search and grid search, purely random search over the same 32-dimensional configuration space found statistically equal performance on four of seven data sets, and superior performance on one of seven. A Gaussian process analysis of the function from hyper-parameters to validation set performance reveals that for most data sets only a few of the hyper-parameters really matter, but that different hyper-parameters are important on different data sets. This phenomenon makes grid search a poor choice for configuring algorithms for new data sets. Our analysis casts some light on why recent “High Throughput” methods achieve surprising success—they appear to search through a large number of hyper-parameters because most hyper-parameters do not matter much. We anticipate that growing interest in large hierarchical models will place an increasing burden on techniques for hyper-parameter optimization; this work shows that random search is a natural baseline against which to judge progress in the development of adaptive (sequential) hyper-parameter optimization algorithms.

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Tags:

● No DOI found

Attachments

- Bergstra and Bengio - Random Search for Hyper-Parameter Optimization.pdf

Real-Time Single Image and Video Super-Resolution Using an Efficient Sub-Pixel Convolutional Neural Network

Type Journal Article
Author Wenzhe Shi
Author Jose Caballero
Author Ferenc Huszár
Author Johannes Totz
Author Andrew P. Aitken
Author Rob Bishop
Author Daniel Rueckert
Author Zehan Wang
URL <http://arxiv.org/abs/1609.05158>
Publication arXiv:1609.05158 [cs, stat]
Date 2016-09-16
Extra 00751 arXiv: 1609.05158
Accessed 6/21/2019, 4:49:34 PM
Library Catalog arXiv.org
Abstract Recently, several models based on deep neural networks have achieved great success in terms of both reconstruction accuracy and computational performance for single image super-resolution. In these methods, the low resolution (LR) input image is upsampled to the high resolution (HR) space using a single filter, commonly bicubic interpolation, before reconstruction. This means that the super-resolution (SR) operation is performed in HR space. We demonstrate that this is sub-optimal and adds computational complexity. In this paper, we present the first convolutional neural network (CNN) capable of real-time SR of 1080p videos on a single K2 GPU. To achieve this, we propose a novel CNN architecture where the feature maps are extracted in the LR space. In addition, we introduce an efficient sub-pixel convolution layer which learns an array of upscaling filters to upscale the final LR feature maps into the HR output. By doing so, we effectively replace the handcrafted bicubic filter in the SR pipeline with more complex upscaling filters specifically trained for each feature map, whilst also reducing the computational complexity of the overall SR operation. We evaluate the proposed approach using images and videos from publicly available datasets and show that it performs significantly better (+0.15dB on Images and +0.39dB on Videos) and is an order of magnitude faster than previous CNN-based methods.
Date Added 6/21/2019, 4:49:34 PM
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Tags:

Computer Science - Computer Vision and Pattern Recognition, Statistics - Machine Learning

Notes:

Comment: CVPR 2016 paper with updated affiliations and supplemental material, fixed typo in equation 4

Attachments

- arXiv.org Snapshot
- Shi et al_2016_Real-Time Single Image and Video Super-Resolution Using an Efficient Sub-Pixel.pdf

Realistic Evaluation of Deep Semi-Supervised Learning Algorithms

Type Journal Article

Author Avital Oliver

Author Augustus Odena

Author Colin Raffel

Author Ekin D. Cubuk

Author Ian J. Goodfellow

URL <http://arxiv.org/abs/1804.09170>

Publication arXiv:1804.09170 [cs, stat]

Date 2018-04-24

Extra 00063 arXiv: 1804.09170

Accessed 6/21/2019, 11:55:34 AM

Library Catalog arXiv.org

Abstract Semi-supervised learning (SSL) provides a powerful framework for leveraging unlabeled data when labels are limited or expensive to obtain. SSL algorithms based on deep neural networks have recently proven successful on standard benchmark tasks. However, we argue that these benchmarks fail to address many issues that these algorithms would face in real-world applications. After creating a unified reimplementation of various widely-used SSL techniques, we test them in a suite of experiments designed to address these issues. We find that the performance of simple baselines which do not use unlabeled data is often underreported, that SSL methods differ in sensitivity to the amount of labeled and unlabeled data, and that performance can degrade substantially when the unlabeled dataset contains out-of-class examples. To help guide SSL research towards real-world applicability, we make our unified reimplementation and evaluation platform publicly available.

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Tags:

⊖ No DOI found, Computer Science - Machine Learning, Statistics - Machine Learning

Attachments

- Oliver et al_2018_Realistic Evaluation of Deep Semi-Supervised Learning Algorithms.pdf

Recurrent Convolutional Neural Networks for Scene Labeling

Type Journal Article
Author Pedro O Pinheiro
Author Ronan Collobert
Pages 9
Extra 00563
Library Catalog Zotero
Language en
Abstract The goal of the scene labeling task is to assign a class label to each pixel in an image. To ensure a good visual coherence and a high class accuracy, it is essential for a model to capture long range (pixel) label dependencies in images. In a feed-forward architecture, this can be achieved simply by considering a sufficiently large input context patch, around each pixel to be labeled. We propose an approach that consists of a recurrent convolutional neural network which allows us to consider a large input context while limiting the capacity of the model. Contrary to most standard approaches, our method does not rely on any segmentation technique nor any task-specific features. The system is trained in an end-to-end manner over raw pixels, and models complex spatial dependencies with low inference cost. As the context size increases with the built-in recurrence, the system identifies and corrects its own errors. Our approach yields state-of-the-art performance on both the Stanford Background Dataset and the SIFT Flow Dataset, while remaining very fast at test time.
Date Added 6/21/2019, 11:45:24 AM
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Tags:

⊖ No DOI found

Attachments

- Pinheiro and Collobert - Recurrent Convolutional Neural Networks for Scene .pdf

Show and Tell: A Neural Image Caption Generator

Type Journal Article
Author Oriol Vinyals
Author Alexander Toshev
Author Samy Bengio
Author Dumitru Erhan
URL <http://arxiv.org/abs/1411.4555>
Publication arXiv:1411.4555 [cs]
Date 2014-11-17
Extra 02562 arXiv: 1411.4555
Accessed 6/21/2019, 12:44:10 PM
Library Catalog arXiv.org
Abstract Automatically describing the content of an image is a fundamental problem in artificial intelligence that connects computer vision and natural language processing. In this paper, we present a generative model based on a deep recurrent architecture that combines recent advances in computer vision and machine translation and that can be used to generate natural sentences describing an image. The model is trained to maximize the likelihood of the target description sentence given the training image. Experiments on several datasets show the accuracy of the model and the fluency of the language it learns solely from image descriptions. Our model is often quite accurate, which we verify both qualitatively and quantitatively. For instance, while the current state-of-the-art BLEU-1 score (the higher the better) on the Pascal dataset is 25, our approach yields 59, to be compared to human performance around 69. We also show BLEU-1 score improvements on Flickr30k, from 56 to 66, and on SBU, from 19 to 28. Lastly, on the newly released COCO dataset, we achieve a BLEU-4 of 27.7, which is the current state-of-the-art.
Short Title Show and Tell
Date Added 6/21/2019, 12:44:10 PM
Modified 6/21/2019, 12:44:24 PM

Tags:

Computer Science - Computer Vision and Pattern Recognition

Attachments

- arXiv.org Snapshot
- Vinyals et al_2014_Show and Tell.pdf

Contents

- 1 . Introduction
- 2 . Related Work
- 3 . Model
 - 3.1 . LSTM-based Sentence Generator
- 4 . Experiments
 - 4.1 . Evaluation Metrics
 - 4.2 . Datasets
 - 4.3 . Results
 - 4.3.1 Training Details
 - 4.3.2 Generation Results
 - 4.3.3 Transfer Learning, Data Size and Label Quality
 - 4.3.4 Generation Diversity Discussion
 - 4.3.5 Ranking Results
 - 4.3.6 Human Evaluation
 - 4.3.7 Analysis of Embeddings
- 5 . Conclusion

Show and Tell: Lessons learned from the 2015 MSCOCO Image Captioning Challenge

Type Journal Article
Author Oriol Vinyals
Author Alexander Toshev
Author Samy Bengio
Author Dumitru Erhan
URL <http://arxiv.org/abs/1609.06647>
Volume 39
Issue 4
Pages 652-663
Publication IEEE Transactions on Pattern Analysis and Machine Intelligence
ISSN 0162-8828, 2160-9292
Date 2017-4-1
Extra 00281 arXiv: 1609.06647
Journal Abbr IEEE Trans. Pattern Anal. Mach. Intell.
DOI 10/f9xnjx

Accessed 6/21/2019, 11:30:39 AM

Library Catalog arXiv.org

Language en

Abstract Automatically describing the content of an image is a fundamental problem in artificial intelligence that connects computer vision and natural language processing. In this paper, we present a generative model based on a deep recurrent architecture that combines recent advances in computer vision and machine translation and that can be used to generate natural sentences describing an image. The model is trained to maximize the likelihood of the target description sentence given the training image. Experiments on several datasets show the accuracy of the model and the fluency of the language it learns solely from image descriptions. Our model is often quite accurate, which we verify both qualitatively and quantitatively. Finally, given the recent surge of interest in this task, a competition was organized in 2015 using the newly released COCO dataset. We describe and analyze the various improvements we applied to our own baseline and show the resulting performance in the competition, which we won ex-aequo with a team from Microsoft Research, and provide an open source implementation in TensorFlow.

Short Title Show and Tell

Date Added 6/21/2019, 11:30:39 AM

Modified 6/21/2019, 11:30:43 AM

Tags:

Computer Science - Computer Vision and Pattern Recognition

Notes:

Comment: arXiv admin note: substantial text overlap with arXiv:1411.4555

Attachments

- Vinyals et al. - 2017 - Show and Tell Lessons learned from the 2015 MSCOC.pdf

State-Space Representations of Deep Neural Networks

Type Journal Article

Author Michael Hauser

Author Sean Gunn

Author Samer Saab

Author Asok Ray

URL https://www.mitpressjournals.org/doi/10.1162/neco_a_01165

Volume 31

Issue 3
Pages 538-554
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ISSN 0899-7667, 1530-888X
Date 03/2019
Extra 00001
Journal Abbr Neural Computation
DOI 10/gf35pr
Accessed 6/21/2019, 12:55:58 PM
Library Catalog DOI.org (Crossref)
Language en
Abstract This letter deals with neural networks as dynamical systems governed by finite difference equations. It shows that the introduction of [Formula: see text]-many skip connections into network architectures, such as residual networks and additive dense networks, defines [Formula: see text]th order dynamical equations on the layer-wise transformations. Closed-form solutions for the state-space representations of general [Formula: see text]th order additive dense networks, where the concatenation operation is replaced by addition, as well as [Formula: see text]th order smooth networks, are found. The developed provision endows deep neural networks with an algebraic structure. Furthermore, it is shown that imposing [Formula: see text]th order smoothness on network architectures with [Formula: see text]-many nodes per layer increases the state-space dimension by a multiple of [Formula: see text], and so the effective embedding dimension of the data manifold by the neural network is [Formula: see text]-many dimensions. It follows that network architectures of these types reduce the number of parameters needed to maintain the same embedding dimension by a factor of [Formula: see text] when compared to an equivalent first-order, residual network. Numerical simulations and experiments on CIFAR10, SVHN, and MNIST have been conducted to help understand the developed theory and efficacy of the proposed concepts.
Date Added 6/21/2019, 12:55:58 PM
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Attachments

- Hauser et al. - 2019 - State-Space Representations of Deep Neural Network.pdf

Statistical learning with sparsity: the lasso and generalizations

Type Book
Author Trevor Hastie
Author Robert Tibshirani
Author Martin Wainwright

Series Monographs on statistics and applied probability
Place Boca Raton
Publisher CRC Press, Taylor & Francis Group
ISBN 978-1-4987-1216-3
Date 2015
Call Number QA275 .H38 2015
Extra 00982
Series Number 143
Library Catalog Library of Congress ISBN
Short Title Statistical learning with sparsity
of Pages 351
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Tags:

Least squares, Linear models (Statistics), Mathematical statistics, Proof theory

Notes:

"A Chapman & Hall book."

Attachments

- Hastie et al. - 2015 - Statistical learning with sparsity the lasso and .pdf

T-CGAN: Conditional Generative Adversarial Network for Data Augmentation in Noisy Time Series with Irregular Sampling

Type Journal Article
Author Giorgia Ramponi
Author Pavlos Protopapas
Author Marco Brambilla
Author Ryan Janssen
URL <http://arxiv.org/abs/1811.08295>
Publication arXiv:1811.08295 [cs, stat]
Date 2018-11-20
Extra 00001 arXiv: 1811.08295
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Library Catalog arXiv.org
Language en

Abstract In this paper we propose a data augmentation method for time series with irregular sampling, Time-Conditional Generative Adversarial Network (T-CGAN). Our approach is based on Conditional Generative Adversarial Networks (CGAN), where the generative step is implemented by a deconvolutional NN and the discriminative step by a convolutional NN. Both the generator and the discriminator are conditioned on the sampling timestamps, to learn the hidden relationship between data and timestamps, and consequently to generate new time series.

Short Title T-CGAN

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Tags:

☹ No DOI found, Computer Science - Machine Learning, Statistics - Machine Learning

Attachments

- Ramponi et al. - 2018 - T-CGAN Conditional Generative Adversarial Network.pdf

The elements of statistical learning: data mining, inference, and prediction

Type Book
Author Trevor Hastie
Author Robert Tibshirani
Author J. H. Friedman
Series Springer series in statistics
Edition 2nd ed
Place New York, NY
Publisher Springer
ISBN 978-0-387-84857-0 978-0-387-84858-7
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Call Number Q325.5 .H39 2009
Extra 00355
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of Pages 745
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Tags:

Machine learning, Bioinformatics, Computational intelligence, Data mining, Forecasting, Inference, Methodology, Statistics

Attachments

- Hastie et al. - 2009 - The elements of statistical learning data mining,.pdf

The great time series classification bake off: a review and experimental evaluation of recent algorithmic advances

Type Journal Article
Author Anthony Bagnall
Author Jason Lines
Author Aaron Bostrom
Author James Large
Author Eamonn Keogh
URL <http://link.springer.com/10.1007/s10618-016-0483-9>
Volume 31
Issue 3
Pages 606-660
Publication Data Mining and Knowledge Discovery
ISSN 1384-5810, 1573-756X
Date 5/2017
Journal Abbr Data Min Knowl Disc
DOI 10/f9586n
Accessed 6/21/2019, 11:44:25 AM
Library Catalog DOI.org (Crossref)
Language en
Abstract In the last 5 years there have been a large number of new time series classification algorithms proposed in the literature. These algorithms have been evaluated on subsets of the 47 data sets in the University of California, Riverside time series classification archive. The archive has recently been expanded to 85 data sets, over half of which have been donated by researchers at the University of East Anglia. Aspects of previous evaluations have made comparisons between algorithms difficult. For example, several different programming languages have been used, experiments involved a single train/test split and some used normalised data whilst others did not. The relaunch of the archive provides a timely opportunity to thoroughly evaluate algorithms on a larger number of datasets. We have implemented 18 recently proposed algorithms in a common Java framework and compared them against two standard benchmark classifiers (and each other) by performing 100 resampling experiments on each of the 85 datasets. We use these results to test several hypotheses relating to whether the algorithms

are significantly more accurate than the benchmarks and each other. Our results indicate that only nine of these algorithms are significantly more accurate than both benchmarks and that one classifier, the collective of transformation ensembles, is significantly more accurate than all of the others. All of our experiments and results are reproducible: we release all of our code, results and experimental details and we hope these experiments form the basis for more robust testing of new algorithms in the future.

Short Title The great time series classification bake off

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Attachments

- Bagnall et al. - 2017 - The great time series classification bake off a r.pdf

U-Net: Convolutional Networks for Biomedical Image Segmentation

Type Journal Article

Author Olaf Ronneberger

Author Philipp Fischer

Author Thomas Brox

URL <http://arxiv.org/abs/1505.04597>

Publication arXiv:1505.04597 [cs]

Date 2015-05-18

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Library Catalog arXiv.org

Abstract There is large consent that successful training of deep networks requires many thousand annotated training samples. In this paper, we present a network and training strategy that relies on the strong use of data augmentation to use the available annotated samples more efficiently. The architecture consists of a contracting path to capture context and a symmetric expanding path that enables precise localization. We show that such a network can be trained end-to-end from very few images and outperforms the prior best method (a sliding-window convolutional network) on the ISBI challenge for segmentation of neuronal structures in electron microscopic stacks. Using the same network trained on transmitted light microscopy images (phase contrast and DIC) we won the ISBI cell tracking challenge 2015 in these categories by a large margin. Moreover, the network is fast. Segmentation of a 512x512 image takes less than a second on a recent GPU. The full implementation (based on Caffe) and the trained networks are available at <http://lmb.informatik.uni-freiburg.de/people/ronneber/u-net>.

Short Title U-Net

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Tags:

Computer Science - Computer Vision and Pattern Recognition

Notes:

Comment: conditionally accepted at MICCAI 2015

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- arXiv.org Snapshot
- Ronneberger et al_2015_U-Net.pdf

Contents

U-Net: Convolutional Networks for Biomedical Image Segmentation

Very Deep Convolutional Networks for Large-Scale Image Recognition

Type Journal Article
Author Karen Simonyan
Author Andrew Zisserman
URL <http://arxiv.org/abs/1409.1556>
Publication arXiv:1409.1556 [cs]
Date 2014-09-04
Extra 23902 arXiv: 1409.1556
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Library Catalog arXiv.org
Language en
Abstract In this work we investigate the effect of the convolutional network depth on its accuracy in the large-scale image recognition setting. Our main contribution is a thorough evaluation of networks of increasing depth using an architecture with very small (3×3) convolution filters, which shows that a significant improvement on the prior-art configurations can be achieved by pushing the depth to 16–19 weight layers. These findings were the basis of our ImageNet Challenge 2014 submission, where our team secured the first and the second places in the localisation and classification tracks respectively. We also show that our representations generalise well to other datasets, where they achieve state-of-the-art results. We have made our two best-performing ConvNet models publicly available to facilitate further research on the use of deep visual representations in computer vision.
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Tags:

● No DOI found, Computer Science - Computer Vision and Pattern Recognition

Attachments

- Simonyan and Zisserman - 2014 - Very Deep Convolutional Networks for Large-Scale I.pdf

Visualizing and Understanding Convolutional Networks

Type Book Section
Editor David Fleet
Editor Tomas Pajdla
Editor Bernt Schiele
Editor Tinne Tuytelaars
Author Matthew D. Zeiler
Author Rob Fergus
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Date 2014
Extra 00000 DOI: 10.1007/978-3-319-10590-1_53
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Library Catalog DOI.org (Crossref)
Language en
Abstract Large Convolutional Network models have recently demonstrated impressive classification performance on the ImageNet benchmark Krizhevsky et al. [18]. However there is no clear understanding of why they perform so well, or how they might be improved. In this paper we explore both issues. We introduce a novel visualization technique that gives insight into the function of intermediate feature layers and the operation of the classifier. Used in a diagnostic role, these visualizations allow us to find model architectures that outperform Krizhevsky et al. on the ImageNet classification benchmark. We also perform an ablation study to discover the performance contribution from different model layers. We show our ImageNet model generalizes well to other datasets: when the softmax classifier is retrained, it convincingly beats the current state-of-the-art results on Caltech-101 and Caltech-256 datasets.

Book Title Computer Vision – ECCV 2014
Date Added 6/21/2019, 11:34:07 AM
Modified 6/21/2019, 11:34:09 AM

Attachments

- Zeiler and Fergus - 2014 - Visualizing and Understanding Convolutional Networ.pdf

Visualizing the Loss Landscape of Neural Nets

Type Journal Article
Author Hao Li
Author Zheng Xu
Author Gavin Taylor
Author Christoph Studer
Author Tom Goldstein
URL <http://arxiv.org/abs/1712.09913>
Publication arXiv:1712.09913 [cs, stat]
Date 2017-12-28
Extra 00129 arXiv: 1712.09913
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Library Catalog arXiv.org
Abstract Neural network training relies on our ability to find "good" minimizers of highly non-convex loss functions. It is well-known that certain network architecture designs (e.g., skip connections) produce loss functions that train easier, and well-chosen training parameters (batch size, learning rate, optimizer) produce minimizers that generalize better. However, the reasons for these differences, and their effects on the underlying loss landscape, are not well understood. In this paper, we explore the structure of neural loss functions, and the effect of loss landscapes on generalization, using a range of visualization methods. First, we introduce a simple "filter normalization" method that helps us visualize loss function curvature and make meaningful side-by-side comparisons between loss functions. Then, using a variety of visualizations, we explore how network architecture affects the loss landscape, and how training parameters affect the shape of minimizers.
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Tags:

Computer Science - Computer Vision and Pattern Recognition, Computer Science - Machine Learning, Statistics - Machine Learning

Notes:

Comment: NIPS 2018 (extended version, 10.5 pages), code is available at <https://github.com/tomgoldstein/loss-landscape>

Attachments

- arXiv.org Snapshot
- Li et al_2017_Visualizing the Loss Landscape of Neural Nets.pdf

Contents

- 1 Introduction
 - 1.1 Contributions
- 2 Theoretical Background
- 3 The Basics of Loss Function Visualization
- 4 Proposed Visualization: Filter-Wise Normalization
- 5 The Sharp vs Flat Dilemma
- 6 What Makes Neural Networks Trainable? Insights on the (Non)Convexity Structure of Loss Surfaces
- 7 Visualizing Optimization Paths
 - 7.1 Why Random Directions Fail: Low-Dimensional Optimization Trajectories
 - 7.2 Effective Trajectory Plotting using PCA Directions
- 8 Conclusion
- A Comparison of Loss Surfaces
 - A.1 The Change of Weights Norm during Training
 - A.2 Comparision of Normalization Methods
 - A.3 Small-Batch vs Large-Batch for ResNet-56
 - A.4 Repeatability of the Loss Surface Visualization
 - A.5 Implementation Details
 - A.6 Training Curves for VGG-9 and ResNets

What is the Role of Recurrent Neural Networks (RNNs) in an Image Caption Generator?

Type Journal Article
Author Marc Tanti
Author Albert Gatt
Author Kenneth P. Camilleri
URL <http://arxiv.org/abs/1708.02043>
Publication arXiv:1708.02043 [cs]
Date 2017-08-07

Extra 00005 arXiv: 1708.02043
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Library Catalog arXiv.org
Abstract In neural image captioning systems, a recurrent neural network (RNN) is typically viewed as the primary 'generation' component. This view suggests that the image features should be 'injected' into the RNN. This is in fact the dominant view in the literature. Alternatively, the RNN can instead be viewed as only encoding the previously generated words. This view suggests that the RNN should only be used to encode linguistic features and that only the final representation should be 'merged' with the image features at a later stage. This paper compares these two architectures. We find that, in general, late merging outperforms injection, suggesting that RNNs are better viewed as encoders, rather than generators.
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- arXiv.org Snapshot
- Tanti et al_2017_What is the Role of Recurrent Neural Networks (RNNs) in an Image Caption.pdf

Where to put the Image in an Image Caption Generator

Type Journal Article
Author Marc Tanti
Author Albert Gatt
Author Kenneth P. Camilleri
URL <http://arxiv.org/abs/1703.09137>
Volume 24
Issue 3
Pages 467-489
Publication Natural Language Engineering

ISSN 1351-3249, 1469-8110
Date 05/2018
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Journal Abbr Nat. Lang. Eng.
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Abstract When a recurrent neural network language model is used for caption generation, the image information can be fed to the neural network either by directly incorporating it in the RNN -- conditioning the language model by 'injecting' image features -- or in a layer following the RNN -- conditioning the language model by 'merging' image features. While both options are attested in the literature, there is as yet no systematic comparison between the two. In this paper we empirically show that it is not especially detrimental to performance whether one architecture is used or another. The merge architecture does have practical advantages, as conditioning by merging allows the RNN's hidden state vector to shrink in size by up to four times. Our results suggest that the visual and linguistic modalities for caption generation need not be jointly encoded by the RNN as that yields large, memory-intensive models with few tangible advantages in performance; rather, the multimodal integration should be delayed to a subsequent stage.
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