

Deep Learning

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Contents

Website	vii
Acknowledgments	viii
Notation	xi
1 Introduction	1
1.1 Who Should Read This Book?	8
1.2 Historical Trends in Deep Learning	11
I Applied Math and Machine Learning Basics	29
2 Linear Algebra	31
2.1 Scalars, Vectors, Matrices and Tensors	31
2.2 Multiplying Matrices and Vectors	34
2.3 Identity and Inverse Matrices	36
2.4 Linear Dependence and Span	37
2.5 Norms	39
2.6 Special Kinds of Matrices and Vectors	40
2.7 Eigendecomposition	42
2.8 Singular Value Decomposition	44
2.9 The Moore-Penrose Pseudoinverse	45
2.10 The Trace Operator	46
2.11 The Determinant	47
2.12 Example: Principal Components Analysis	48
3 Probability and Information Theory	53
3.1 Why Probability?	54

3.2	Random Variables	56
3.3	Probability Distributions	56
3.4	Marginal Probability	58
3.5	Conditional Probability	59
3.6	The Chain Rule of Conditional Probabilities	59
3.7	Independence and Conditional Independence	60
3.8	Expectation, Variance and Covariance	60
3.9	Common Probability Distributions	62
3.10	Useful Properties of Common Functions	67
3.11	Bayes' Rule	70
3.12	Technical Details of Continuous Variables	71
3.13	Information Theory	72
3.14	Structured Probabilistic Models	75
4	Numerical Computation	80
4.1	Overflow and Underflow	80
4.2	Poor Conditioning	82
4.3	Gradient-Based Optimization	82
4.4	Constrained Optimization	93
4.5	Example: Linear Least Squares	96
5	Machine Learning Basics	98
5.1	Learning Algorithms	99
5.2	Capacity, Overfitting and Underfitting	110
5.3	Hyperparameters and Validation Sets	120
5.4	Estimators, Bias and Variance	122
5.5	Maximum Likelihood Estimation	131
5.6	Bayesian Statistics	135
5.7	Supervised Learning Algorithms	140
5.8	Unsupervised Learning Algorithms	146
5.9	Stochastic Gradient Descent	151
5.10	Building a Machine Learning Algorithm	153
5.11	Challenges Motivating Deep Learning	155
II	Deep Networks: Modern Practices	166
6	Deep Feedforward Networks	168
6.1	Example: Learning XOR	171
6.2	Gradient-Based Learning	177

6.3	Hidden Units	191
6.4	Architecture Design	197
6.5	Back-Propagation and Other Differentiation Algorithms	204
6.6	Historical Notes	225
7	Regularization for Deep Learning	229
7.1	Parameter Norm Penalties	231
7.2	Norm Penalties as Constrained Optimization	238
7.3	Regularization and Under-Constrained Problems	240
7.4	Dataset Augmentation	241
7.5	Noise Robustness	243
7.6	Semi-Supervised Learning	245
7.7	Multi-Task Learning	246
7.8	Early Stopping	247
7.9	Parameter Tying and Parameter Sharing	252
7.10	Sparse Representations	254
7.11	Bagging and Other Ensemble Methods	256
7.12	Dropout	258
7.13	Adversarial Training	268
7.14	Tangent Distance, Tangent Prop, and Manifold Tangent Classifier	270
8	Optimization for Training Deep Models	276
8.1	How Learning Differs from Pure Optimization	277
8.2	Challenges in Neural Network Optimization	284
8.3	Basic Algorithms	296
8.4	Parameter Initialization Strategies	302
8.5	Algorithms with Adaptive Learning Rates	308
8.6	Approximate Second-Order Methods	312
8.7	Optimization Strategies and Meta-Algorithms	320
9	Convolutional Networks	333
9.1	The Convolution Operation	334
9.2	Motivation	338
9.3	Pooling	342
9.4	Convolution and Pooling as an Infinitely Strong Prior	348
9.5	Variants of the Basic Convolution Function	350
9.6	Structured Outputs	361
9.7	Data Types	363
9.8	Efficient Convolution Algorithms	365
9.9	Random or Unsupervised Features	366

9.10	The Neuroscientific Basis for Convolutional Networks	367
9.11	Convolutional Networks and the History of Deep Learning	374
10	Sequence Modeling: Recurrent and Recursive Nets	376
10.1	Unfolding Computational Graphs	378
10.2	Recurrent Neural Networks	381
10.3	Bidirectional RNNs	397
10.4	Encoder-Decoder Sequence-to-Sequence Architectures	399
10.5	Deep Recurrent Networks	401
10.6	Recursive Neural Networks	403
10.7	The Challenge of Long-Term Dependencies	404
10.8	Echo State Networks	407
10.9	Leaky Units and Other Strategies for Multiple Time Scales	409
10.10	The Long Short-Term Memory and Other Gated RNNs	411
10.11	Optimization for Long-Term Dependencies	416
10.12	Explicit Memory	419
11	Practical methodology	424
11.1	Performance Metrics	425
11.2	Default Baseline Models	428
11.3	Determining Whether to Gather More Data	429
11.4	Selecting Hyperparameters	430
11.5	Debugging Strategies	439
11.6	Example: Multi-Digit Number Recognition	443
12	Applications	446
12.1	Large Scale Deep Learning	446
12.2	Computer Vision	455
12.3	Speech Recognition	461
12.4	Natural Language Processing	464
12.5	Other Applications	480
III	Deep Learning Research	489
13	Linear Factor Models	492
13.1	Probabilistic PCA and Factor Analysis	493
13.2	Independent Component Analysis (ICA)	494
13.3	Slow Feature Analysis	496
13.4	Sparse Coding	499

13.5	Manifold Interpretation of PCA	502
14	Autoencoders	505
14.1	Undercomplete Autoencoders	506
14.2	Regularized Autoencoders	507
14.3	Representational Power, Layer Size and Depth	511
14.4	Stochastic Encoders and Decoders	512
14.5	Denoising Autoencoders	513
14.6	Learning Manifolds with Autoencoders	518
14.7	Contractive Autoencoders	524
14.8	Predictive Sparse Decomposition	526
14.9	Applications of Autoencoders	527
15	Representation Learning	529
15.1	Greedy Layer-Wise Unsupervised Pretraining	531
15.2	Transfer Learning and Domain Adaptation	539
15.3	Semi-Supervised Disentangling of Causal Factors	544
15.4	Distributed Representation	549
15.5	Exponential Gains from Depth	556
15.6	Providing Clues to Discover Underlying Causes	557
16	Structured Probabilistic Models for Deep Learning	561
16.1	The Challenge of Unstructured Modeling	562
16.2	Using Graphs to Describe Model Structure	566
16.3	Sampling from Graphical Models	583
16.4	Advantages of Structured Modeling	585
16.5	Learning about Dependencies	585
16.6	Inference and Approximate Inference	586
16.7	The Deep Learning Approach to Structured Probabilistic Models	588
17	Monte Carlo Methods	593
17.1	Sampling and Monte Carlo Methods	593
17.2	Markov Chain Monte Carlo Methods	599
17.3	Gibbs Sampling	603
17.4	The Challenge of Mixing between Separated Modes	603
18	Confronting the Partition Function	609
18.1	The Log-Likelihood Gradient of Undirected Models	610
18.2	Stochastic Maximum Likelihood and Contrastive Divergence . . .	612
18.3	Pseudolikelihood	620

18.4	Score Matching and Ratio Matching	622
18.5	Denoising Score Matching	625
18.6	Noise-Contrastive Estimation	625
18.7	Estimating the Partition Function	628
19	Approximate inference	636
19.1	Inference as Optimization	638
19.2	Expectation Maximization	639
19.3	MAP Inference and Sparse Coding	640
19.4	Variational Inference and Learning	643
19.5	Learned Approximate Inference	656
20	Deep Generative Models	659
20.1	Boltzmann Machines	659
20.2	Restricted Boltzmann Machines	661
20.3	Deep Belief Networks	665
20.4	Deep Boltzmann Machines	668
20.5	Boltzmann Machines for Real-Valued Data	682
20.6	Convolutional Boltzmann Machines	690
20.7	Boltzmann Machines for Structured or Sequential Outputs	692
20.8	Other Boltzmann Machines	694
20.9	Back-Propagation through Random Operations	695
20.10	Directed Generative Nets	699
20.11	Auto-Regressive Networks	713
20.12	Drawing Samples from Autoencoders	718
20.13	Generative Stochastic Networks	722
20.14	Other Generation Schemes	724
20.15	Evaluating Generative Models	725
20.16	Conclusion	727
	Bibliography	729
	Index	787

Website

www.deeplearningbook.org

This book is accompanied by the above website. The website provides a variety of supplementary material, including exercises, lecture slides, corrections of mistakes, and other resources that should be useful to both readers and instructors.

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