Dive into Deep Learning

xAI-Proj-B: Bachelor Project Explainable Machine Learning

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Abstract

//TODO: write abstract? Was muss da rein?

1 Introduction

//TODO

- 2 Methods
- 2.1 Datasets

//TODO

2.1.1 MNIST

//TODO

2.1.2 MedMNIST

//TODO

- 2.2 Model Architectures
- 2.2.1 SimpleCNN

//TODO

2.2.2 ResNet

//TODO

Resnet18 //TODO

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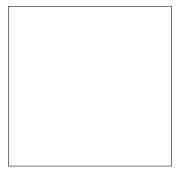


Figure 1: Sample figure caption.

Table 1: Sample table title

	Part	
Name	Description	Size (μm)
Dendrite Axon Soma	Input terminal Output terminal Cell body	~ 100 ~ 10 up to 10^6

Resnet50 //TODO

ResnetXX //TODO

2.2.3 Xception

//TODO

3 Discussion

//TODO

4 Conclusion

//TODO

4.1 Footnotes

Footnotes should be used sparingly. If you do require a footnote, indicate footnotes with a number⁴ in the text. Place the footnotes at the bottom of the page on which they appear. Precede the footnote with a horizontal rule of 2 inches.

Note that footnotes are properly typeset after punctuation marks.⁵

4.2 Figures

4.3 Tables

5 Notation

This section provides a concise reference describing notation as used in the book by Goodfellow et al. (2016). If you are unfamiliar with any of the corresponding mathematical concepts, Goodfellow et al.

⁴Sample of the first footnote.

⁵As in this example.

Numbers and Arrays

- a A scalar (integer or real)
- a A vector
- **A** A matrix
- A A tensor
- I_n Identity matrix with n rows and n columns
- I Identity matrix with dimensionality implied by context
- $e^{(i)}$ Standard basis vector $[0, \dots, 0, 1, 0, \dots, 0]$ with a 1 at position i
- $\operatorname{diag}(a)$ A square, diagonal matrix with diagonal entries given by a
 - a A scalar random variable
 - a A vector-valued random variable
 - A A matrix-valued random variable

Sets and Graphs

- \mathbb{A} A set
- \mathbb{R} The set of real numbers
- $\{0,1\}$ The set containing 0 and 1
- $\{0, 1, \dots, n\}$ The set of all integers between 0 and n
 - [a, b] The real interval including a and b
 - (a, b] The real interval excluding a but including b
 - $\mathbb{A} \setminus \mathbb{B}$ Set subtraction, i.e., the set containing the elements of \mathbb{A}
 - that are not in \mathbb{B}
 - \mathcal{G} A graph
 - $Pa_{\mathcal{G}}(\mathbf{x}_i)$ The parents of \mathbf{x}_i in \mathcal{G}

Indexing

- a_i Element i of vector a, with indexing starting at 1
- a_{-i} All elements of vector \boldsymbol{a} except for element i
- $A_{i,j}$ Element i, j of matrix A
- $A_{i::}$ Row i of matrix A
- $A_{:,i}$ Column i of matrix A
- $A_{i,j,k}$ Element (i,j,k) of a 3-D tensor **A**
- $\mathbf{A}_{:,:,i}$ 2-D slice of a 3-D tensor
- a_i Element i of the random vector \mathbf{a}

Linear Algebra Operations

- ∂y Partial derivative of y with respect to x $\overline{\partial x}$ $\nabla_{\boldsymbol{x}} y$ Gradient of y with respect to x $\nabla_{\mathbf{X}} y$ Matrix derivatives of y with respect to X $\nabla_{\mathbf{X}} y$ Tensor containing derivatives of y with respect to X ∂f Jacobian matrix $\boldsymbol{J} \in \mathbb{R}^{m \times n}$ of $f: \mathbb{R}^n \to \mathbb{R}^m$ The Hessian matrix of f at input point \boldsymbol{x} $\int_{\mathbb{S}} f(\boldsymbol{x}) d\boldsymbol{x}$ $\int_{\mathbb{S}} f(\boldsymbol{x}) d\boldsymbol{x}$ Definite integral over the entire domain of xDefinite integral with respect to \boldsymbol{x} over the set $\mathbb S$

Probability and Information Theory

	Probability and information Theory		
$a\bot b$	The random variables a and b are independent		
$a \bot b \mid c$	They are conditionally independent given c		
$P(\mathbf{a})$	A probability distribution over a discrete variable		
$p(\mathbf{a})$	A probability distribution over a continuous variable, or over a variable whose type has not been specified		
$a \sim P$	Random variable a has distribution P		
$\mathbb{E}_{\mathbf{x} \sim P}[f(x)]$ or $\mathbb{E}f(x)$	Expectation of $f(x)$ with respect to $P(\mathbf{x})$		
Var(f(x))	Variance of $f(x)$ under $P(x)$		
Cov(f(x), g(x))	Covariance of $f(x)$ and $g(x)$ under $P(x)$		
$H(\mathbf{x})$	Shannon entropy of the random variable x		
$D_{\mathrm{KL}}(P\ Q)$	Kullback-Leibler divergence of P and Q		
$\mathcal{N}(oldsymbol{x};oldsymbol{\mu},oldsymbol{\Sigma})$	Gaussian distribution over x with mean μ and covariance Σ		

Functions

 $f:\mathbb{A}\to\mathbb{B}$ The function f with domain \mathbb{A} and range \mathbb{B}

 $f \circ g$ Composition of the functions f and g

 $f(x; \theta)$ A function of x parametrized by θ . (Sometimes we write

f(x) and omit the argument θ to lighten notation)

 $\log x$ Natural logarithm of x

 $\sigma(x)$ Logistic sigmoid, $\frac{1}{1 + \exp(-x)}$

 $\zeta(x)$ Softplus, $\log(1 + \exp(x))$

 $||\boldsymbol{x}||_p$ L^p norm of \boldsymbol{x}

||x|| L^2 norm of x

 x^+ Positive part of x, i.e., max(0, x)

 $\mathbf{1}_{\mathrm{condition}}$ is 1 if the condition is true, 0 otherwise

Sometimes we use a function f whose argument is a scalar but apply it to a vector, matrix, or tensor: $f(\boldsymbol{x}), f(\boldsymbol{X})$, or $f(\boldsymbol{X})$. This denotes the application of f to the array element-wise. For example, if $C = \sigma(\boldsymbol{X})$, then $C_{i,j,k} = \sigma(X_{i,j,k})$ for all valid values of i,j and k.

Datasets and Distributions

 p_{data} The data generating distribution

 \hat{p}_{data} The empirical distribution defined by the training set

X A set of training examples

 $x^{(i)}$ The *i*-th example (input) from a dataset

 $y^{(i)}$ or $oldsymbol{y}^{(i)}$ The target associated with $oldsymbol{x}^{(i)}$ for supervised learning

X The $m \times n$ matrix with input example $x^{(i)}$ in row $X_{i,:}$

References

I. Goodfellow, Y. Bengio, A. Courville, and Y. Bengio. Deep learning, volume 1. MIT Press, 2016.

Declaration of Authorship

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A Appendix

Optionally include extra information (complete proofs, additional experiments and plots) in the appendix. This section will often be part of the supplemental material.