

# Notation

This section provides a concise reference describing the notation used throughout this book. If you are unfamiliar with any of the corresponding mathematical concepts, we describe most of these ideas in chapters 2–4.

## Numbers and Arrays

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

## CONTENTS

## Sets and Graphs

$A$	A set
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$\mathbb{R}$	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 $b$
$A \setminus B$	Set subtraction, i.e., the set containing the elements of $A$ that are not in $B$
$G$	A graph
$\text{Pa}_G(x_i)$	The parents of $x_i$ in $G$

#### Indexing

$a_i$	Element $i$ of vector $a$ , with indexing starting at 1
$a_{-i}$	All elements of vector $a$ 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 $A$
$A_{:,:,i}$	2-D slice of a 3-D tensor
$a_i$	Element $i$ of the random vector $a$

#### Linear Algebra Operations

$A^\top$	Transpose of matrix $A$
$A^+$	Moore-Penrose pseudoinverse of $A$
$A \oslash B$	Element-wise (Hadamard) product of $A$ and $B$
$\det(A)$	Determinant of $A$

## CONTENTS

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#### Calculus

$\frac{dy}{dx}$	Derivative of $y$ with respect to $x$
$\frac{\partial y}{\partial x}$	Partial derivative of $y$ with respect to $x$
$\nabla_x y$	Gradient of $y$ with respect to $x$
$\nabla_x Y$	Matrix derivatives of $y$ with respect to $X$
$\nabla_x v$	Tensor containing derivatives of $v$ with respect to $x$

$\nabla_x f$	Gradient of $f$ at input point $x$
$\frac{\partial f}{\partial x}$	Jacobian matrix $J \in \mathbb{R}^{m \times n}$ of $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$
$\nabla_x^2 f(x)$ or $H(f)(x)$	The Hessian matrix of $f$ at input point $x$
$\int f(x)dx$	Definite integral over the entire domain of $x$
$\int_S f(x)dx$	Definite integral with respect to $x$ over the set $S$

### Probability and Information Theory

$a \perp b$	The random variables $a$ and $b$ are independent
$a \perp b \mid c$	They are conditionally independent given $c$
$P(a)$	A probability distribution over a discrete variable
$p(a)$	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 $P$
$E_{x \sim P} [f(x)]$ or $Ef(x)$	Expectation of $f(x)$ with respect to $P(x)$
$\text{Var}(f(x))$	Variance of $f(x)$ under $P(x)$
$\text{Cov}(f(x), g(x))$	Covariance of $f(x)$ and $g(x)$ under $P(x)$
$H(x)$	Shannon entropy of the random variable $x$
$D_{KL}(P \parallel Q)$	Kullback-Leibler divergence of $P$ and $Q$
$N(x; \mu, \Sigma)$	Gaussian distribution over $x$ with mean $\mu$ and covariance $\Sigma$

## CONTENTS

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### Functions

$f : A \rightarrow B$	The function $f$ with domain $A$ and range $B$
$f \circ g$	Composition of the functions $f$ and $g$
$f(x; \theta)$	A function of $x$ parametrized by $\theta$ . (Sometimes we write $f(x)$ and omit the argument $\theta$ to lighten notation)
$\log x$	Natural logarithm of $x$
$\sigma(x)$	Logistic sigmoid, $\frac{1}{1 + \exp(-x)}$
$\zeta(x)$	Softplus, $\log(1 + \exp(x))$
$\ x\ _p$	$L^p$ norm of $x$
$\ x\ $	$L^2$ norm of $x$

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

$1_{\text{condition}}$  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:  $f(x)$ ,  $f(X)$ , or  $f(X)$ . This denotes the application of  $f$  to the array element-wise. For example, if  $C = \sigma(X)$ , then  $C_{i,j,k} = \sigma(X_{i,j,k})$  for all valid values of  $i, j$  and  $k$ .

### Datasets and Distributions

$p_{\text{data}}$  The data generating distribution

$\hat{p}_{\text{data}}$  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  $y^{(i)}$  The target associated with  $x^{(i)}$  for supervised learning

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