Bioinformatics and Medical Applications

Big Data Using Deep Learning Algorithms

Edited by A. Suresh S. Vimal Y. Harold Robinson Dhinesh Kumar Ramaswami and R. Udendhran





This edition first published 2022 by John Wiley & Sons, Inc., 111 River Street, Hoboken, NJ 07030, USA and Scrivener Publishing LLC, 100 Cummings Center, Suite 541J, Beverly, MA 01915, USA © 2022 Scrivener Publishing LLC

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Library of Congress Cataloging-in-Publication Data

ISBN 978-1-119-79183-6

Cover image: Pixabay.Com Cover design by Russell Richardson

Set in size of 11pt and Minion Pro by Manila Typesetting Company, Makati, Philippines

Printed in the USA

10 9 8 7 6 5 4 3 2 1

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