

A Hands-On Guide to Machine Learning, Deep Learning,  
AI, and Natural Language Processing

# MACHINE LEARNING IS FUN!



ADAM GEITGEY

SECOND EDITION

# Machine Learning Is Fun!

A Hands-On Guide to Machine Learning, AI, and Natural Language  
Processing

Adam Geitgey

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