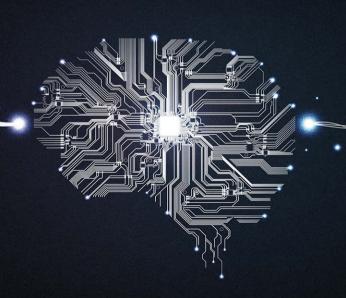


MACHINE LEARNING WITH PYTHON FOR EVERYONE





MARK E. FENNER

Machine Learning with Python for Everyone

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