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A NOVEL ALGORITHM FOR GATED EXPERT NEURAL NETWORKS

BY

Rasha Saleh Aiyad

A Thesis Submitted to the
Faculty of Engineering at Cairo University
in Partial Fulfillment of the
Requirements for the Degree of
MASTER OF SCIENCE

in

ELECTRONICS & COMMUNICATIONS ENGINEERING (COMPUTER ENGINEERING)

Under the Supervision of

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FACULTY OF ENGINEERING, CAIRO UNIVERSITY
GIZA, EGYPT
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