Moving Vehicle Identification using Artificial Neural Networks

by

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A thesis submitted in fulfillment of the requirements for the degree of Master of Science (Mechatronic Engineering)

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