

**Performance Improvement in Non-invasive Brain-Computer
Interface Using Combinatorial Search Space Optimization and
Machine Learning Techniques**

*A thesis submitted in fulfilment of the requirements for the degree
of*

Doctor of Philosophy

By

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Under the Supervision

of

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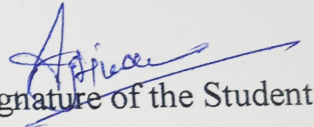
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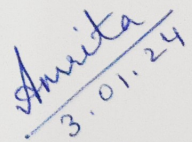
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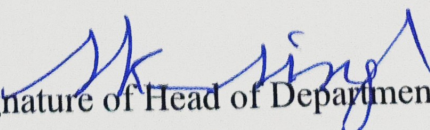
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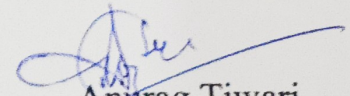
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