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## QCE24 Submission 949

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### Submission 949

Title	Multi-Task Quantum Annealing for Rapid Multi-Class Classification
Paper:	 (Jul 21, 17:31 GMT)
Track	Poster Submissions -- deadline July 22
Author keywords	Quantum annealing Multi-tasking Quantum Annealing (MTQA) Support Vector Machines (SVMs) Machine Learning Optimization Quantum Computing Applications
EasyChair keyphrases	tokyo university of agriculture and technology (180), tasking quantum annealing (126), agriculture and technology tokyo (120), machine learning (110), multi tasking quantum (95), train accuracy test accuracy train accuracy test (84), accuracy test accuracy train accuracy test accuracy (84), rapid multi class classification (80), parallel on a quantum annealer (69), quantum annealing for rapid multi (69), enhance multi class support vector (69), multi class support vector machines (69), support vector machines (63), multi class classification (63), quadratic unconstrained binary optimization (60), test accuracy train accuracy test accuracy train (56), accuracy train accuracy test accuracy train accuracy (56), dataset blob subsets of handwritten digits (51), quantum algorithm optimization and parameter tuning (51), d wave advantage (47), quantum annealing qa (47), parallel quantum annealing (47), is class label (47), cycles to a single cycle (46), sa and standard qa methods (46), cycles to classify multi class (46), multiple cycles to classify multi (46), qa methods require multiple cycles (46), iris dataset with different feature (46), application of multi tasking quantum (46), emergent potential of quantum computing (46), empirical selection of model parameters (46), optimization smo simulated annealing sa (46), rapid and accurate computational solutions (46), classical sequential minimal optimization smo (46), promising approaches to optimization problems (46)
Abstract	Quantum computing offers new paradigms in machine learning, notably through Quantum Annealing (QA), which offers promising approaches to optimization problems. This study introduces an application of Multi-tasking Quantum Annealing (MTQA) to enhance multi-class Support Vector Machines (SVMs). Traditional QA methods require multiple cycles to classify multi-class datasets, significantly increasing computational demands. In contrast, MTQA embeds multiple SVM classifiers in parallel on a quantum annealer, thereby reducing the number of required annealing cycles to a single cycle. Utilizing the D-Wave Advantage 6.4 system, we evaluated our method across three benchmark datasets: a synthetic dataset (Blob), subsets of handwritten digits,

	and the Iris dataset with different feature focuses. Our findings demonstrate that MTQA achieves accuracy equivalent to that of classical Sequential Minimal Optimization (SMO), Simulated Annealing (SA), and standard QA methods while significantly improving computational efficiency. Despite the current limitations in quantum technology and the empirical selection of model parameters, these results highlight the emergent potential of quantum computing in machine learning. Further advancements in quantum algorithm optimization and parameter tuning are anticipated to enhance the efficacy of QA and MTQA, solidifying their practical application roles.
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