

of downloading the TLE data from NORAD. Such as randomly creating the TLE and optimizing using the genetic algorithm (GA) methods, particle swarm optimization (PSO), etc.

ACKNOWLEDGMENT

The authors are grateful to the Malaysian Space Agency and special gratitude to the Universiti Teknologi Malaysia for realizing this study.

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