ANL488 Final Project References

**Comparison of SARIMA and LSTM models**

**in forecasting solar irradiance**

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Presented to Singapore University of Social

Sciences in partial fulfillment of the requirements

for the Degree of Bachelor of Science

in Business Analytics

2022

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