Predictive Implications of Space Weather on Quantum Network Functionality

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Introduction

Quantum communication networks at a glance:

- Utilize quantum properties for ultrasecure communications and sensing
- Depend on precision timing and signal integrity
- Relies on maintaining strict synchronization and control over signal polarization
- Sensitive to external disturbances

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Previous Work:

- Coexistence of classical & quantum channels over single fiber [1]
- Metro traffic vibrations impacting fiber links [2]
- Weather effects on time synchronization error [3][4]

Background [5]





Motivation

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8 APR 2024 Total Solar Eclipse

To what extent do space weather features contribute to explaining the variance in time synchronization errors?

Time Sync Experimental Setup





Dataset Sourcing

Ground-based observatory data (SWPC):

- > F10.7 Index (10.7 cm radio flux) solar radio emission measurements
- SESC sunspot numbers aggregated daily counts of visible sunspots
- ➤ Daily observations

Space-based observations (GOES-16):

- X-Ray flux solar X-ray emission measurements
- Magnetometer ambient magnetic field measurements
- Second-by-second observations

Temporal Aggregation:

- > SWPC data provided as daily observations; uses daily mean aggregate of Time Sync Error
- > GOES-16 hi-res data aggregated (mean) to match Time Sync Error 1 sec obs. rate



Methodology

Analytical Strategy:

Employ correlation and regression analyses

- Explore potential connection between space weather and time sync error
- Determine degree of time sync error variance explainability for each predictor

Test general linear models against observed data

- Apply data transforms to remediate violated model assumptions
- Account for curvilinear relationships
 - Include polynomial expansion of feature space with interaction terms

Model Evaluation:

►Quantitative:

➤Coefficient of Determination, RMSE

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➤Qualitative:

- ➤Residual analysis
- Examine suitability and limitations of regression model assumptions



Results (Correlation Analyses – Daily Aggregates)





Results (Correlation Analyses – Daily Aggregates)

Post-transform on target variable (normalized logit)

Correlation of Daily SWPC Features with Time Synchronization Error (Normalized Logit)











Results (Correlation Analyses – GOES-16 Data)

After selecting a subset of the combined data







Results (Regression Analyses – Baseline)





MSE	RMSE	R ²
16,485.845	128.397	-0.042

MSE	RMSE	R ²
251.522	15.860	0.000



Results (Regression Analyses – Daily Aggregates)





MSE	RMSE	R ²
8188.753	90.492	0.482

MSE	RMSE	R ²
12,216.808	110.530	0.228



Results (Regression Analyses – Daily Aggregates)

Post-transform on target variable (normalized logit)



MSE	RMSE	R ²
7.852	2.802	0.420

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MSE	RMSE	R ²
23.758	4.874	-0.756

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Results (Regression Analyses – GOES-16 Data)





MSE	RMSE	R ²
243.839	15.615	0.031

MSE	RMSE	R ²
243.003	15.589	0.034





Results (Regression Analyses – GOES-16 Data)

Using data subset

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MSE	RMSE	R ²
140.728	11.863	0.148



Residuals vs. Actual Values (Subset, GOES-16 Magnetometer and X-Ray Flux, Polynomial Features)

MSE	RMSE	R ²
139.217	11.800	0.158

Actual TimeTag

0

20

40

-40

-20

Discussion

Cons:

Limited Predictive Power

Space weather may not be as strong as local predictors

Model Assumption Violations

>Normality, homoscedasticity

➤Non-Ideal Sample Size

Limits statistical power of predictive models

➤Transform tradeoffs

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Result in higher correlation, with reduced explainability

Does not translate to better performance on original scale

Pros:

 Select features may explain some time sync error variance
F10.7 index, sunspot number
Could offer insights WRT daily aggregates

Informs future QNet modeling research

Summary

Key takeaways:

- Space weather may offer low to moderate explainability in variance of time synchronization error, but not for real-time applications
- >Other confounding factors may be greater drivers for error (i.e. local weather)
- Space weather is far more likely to impact hardware systems supporting quantum networks (e.g. repeaters, White Rabbit switches) [6][7]

Concluding remarks:

- Space weather is likely not a significant predictor of discrete time synchronization errors
- Additional data and modeling necessary to conclusively separate effects of local vs. space weather factors on QNet timing performance



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Thank you!

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