

Uncertainty quantification of network availability for networks of optical ground stations Iñigo del Portillo, Marc Sanchez Net, Bruce G. Cameron, Edward F. Crawley Massachusetts Institute of Technology



INTRODUCTION

space optical communications is envisioned as the next Free milestone in space communications, due to the higher data-rates achievable (an increase of 10 to 100 times compared to current RF technology), and its lower size, mass, and power. The main drawback of this technology is the decrease in network availability due to link outages caused by cloud coverage.

For **correlated ground stations**, an approximation Monte Carlo method has been proposed¹. We quantify the uncertainty associated to the randomness inherent to this method as a function of cloud probability, correlation index between ground stations, and number of samples.



RESULTS

Revisiting our motivational example, all the monthly average link outage probability points fall in the 99% CI predicted by our model. Also the yearly average LOP.



In the last few years, several studies have been conducted to determine the optimal location of the optical ground stations, both for networks that serve GEO satellites and LEO satellites. However, no analyses have been conducted to quantify the uncertainty of the results when a) the inputs of the models come from different datasets, and b) the network availability is computed using different **methods**. See the motivational example below:



Fig. 1 Average monthly link outage probability (years 2004-2008)

UNCERTAINTY ANLYSIS

Approximations and assumptions incorporated in the link outage probability (LOP) estimation methods.



Fig. 3 Uncertainty vs. cloud probability, correlation index, and number of samples for the approximation MCS method.

UNCERTAINTY IN DATASETS

A pairwise comparison of cloud probabilities monthly averages from different

datasets reveals mean standard deviations of 0.15 and biases of 0.02.



Fig. 4 Pairwise comparison of cloud probabilities among datasets.

PROPOSED METHOD

Fig. 5 Results of motivational example

Comparing our model to the results obtained when using other datasets:

• Low bias (except with MODEIS-CE and POLDER)

• Relatively low variance.



Dataset	Yearly	
	$E[e_y]$ [%]	$\sigma_y \ [\%]$
AIRS-LMD	-2.5	50.3
ISCCP	23.7	20.7
MODIS-CE	-126.5	110.3
MODIS-ST	9.5	30.9
PATMOSX	22.6	21.0
POLDER	-285.6	186.8
ISCCP-hf	-65.8	189.3
EUMETSAT-hf	24.9	61.6

Fig 6 Relative errors of yearly average link outage probability between the values of our model and those obtained using other datasets

- Variations in repeated estimations of the LOP under identical conditions.
- Inexact values of cloud probabilities obtained from external sources and used in the estimation algorithms

UNCERTAINTY IN METHODS

Uncorrelated ground stations in which LOP is defined as having less than M ground stations with clear skies:

 $LOP = \sum_{m=0}^{M-1} {N \choose m} (1-p)^m p^{N-m}$





We provide a dataset that includes average cloud probability and the associated monthly standard uncertainty. This uncertainty takes into account both a) inter-annual variations, and b) different cloud

probability estimations obtained using different datasets.



Fig. 4 Global model for cloud probability and associated uncertainty

CONCLUSIONS

- We present a method to quantify uncertainty in network availability estimations.
- We quantified uncertainty due to different methods (uncorrelated OGSs and correlated OGSs) and different datasets.
- We produced a global dataset of monthly cloud probabilities and associated uncertainty.
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Fig. 2 Link outage probability for multiple station availability (M = 1, 2, 3) for different single-site cloud probabilities (p) and relative uncertainty (u = 10%) vs. different number of stations in the network.

We provide a 5 step method to compute the LOP

and the associated uncertainty, accounting for

inter-annual variations and differences in clouds

datasets.

1. Reference, *Name of reference* 2. Reference, *Name of reference* 3. Reference, *Name of reference*

4. Reference, Name of reference 5. Reference, Name of reference 6. Reference, Name of reference

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