

## RAIN/NO RAIN CLASSIFICATION OVER TROPICAL **REGIONS USING TRMM TMI**

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## ABSTRACT

INTRODUCTION

>For almost over a decade, passive sensing of upwelling microwave

radiation has been recognized as a promising source for the

Microwave instruments aboard Tropical Rainfall Measuring Mission

(TRMM) namely, Microwave Imager (TMI) and Precipitation Radar (PR) offer excellent opportunity to study atmospheric phenomenon over

Overland rain retrieval algorithms using passive radiometers, hinge mainly on the scattering signature from high microwave frequencies. Hence, problematic compared to ocean rainfall algorithms that

utilize both emission and scattering signatures from observations at

Studies dealing with physically based retrieval over land, are few and have not indicated better performance relative to purely statistical

>A deterministic method known as rain /no rain classification (RNC)

applied before actual rain retrieval is adopted to reduce computational

DATA

>Orbital data (version 7) from TRMM Microwave Imager (TMI)

and Precipitation Radar (PR) namely 1B11, 2A25 and 2A23 were

used for study. (From trmm.gsfc.nasa.gov made available since

The effective field of view (EFOV) of the different data channels

along the down track (DT) and cross track (CT) directions are shown

EFOV (CT x DT)

91Kmx632Km

9.1 Km x 30.1 Km

91Km x 226 Km

9.1 Km x 16.0 Km

4.6 Km x 7.2 Km

RESULTS

cipitation Rada (PR)

13.80 GHz (5.1 Km x 5.1 Km

burden as well as for quality checks (Seto et al. 2005).

TRMM Microwa Imager (TMI)

estimation of precipitation (Ferraro et al. 1996)

multiple frequencies (You et al. 2011).

algorithms

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in Figures 1 & 2

Channels

10.65 GHz (H & V)

19.35 GHz (H &V)

37.00 GHz (H & V)

85.50 GHz (H &V)

21 30 GHz 00

This study focalizes on the comparative assessment of existing "rain" or "no rain" classification (RNC) methodologies over land surface. Using the data products from Tropical Rainfall Measuring Mission's Precipitation Radar (PR) and Microwave Imager (TMI), this work reveals the shortcomings of existing overland RNC methods hinged on scattering signatures from 85GHz high frequency channel with special reference to the Indian subcontinent. Overland RNC of microwave radiometer brightness temperatures (Tb) offers a myriad of complications as land surface presents itself as a radiometrically warm and highly variable background. Hence, sensitivity analysis of Tb captured at different microwave frequencies to near surface rain (NSR) rate is of supreme importance. Variability of Tb to NSR suprementation of a second sec channel for the computation of estimated 85GHz Tb is avoidable, due to its prominent correlation with NSR. Also relevant, were the contributions from 19 GHz channels (both horizontal and vertical). A novel attempt has been made to assess the performance of some powerful statistical descriptors to increase the accuracy of RNC. The comparative results are presented extensively in the form of contingency tables and kappa values. The descriptor giving best results from the analysis was used to formulate a regression relation with the rain rate (RR). Furthermore, this work also conducts a detailed examination on the use of Empirical Orthogonal Functions (EOF), to improve rain retrieval accuracy. Analysis reveals that, for TMI, the first two EOF's were deemed sufficient to fully explain the variability offered by the 9 channels. A regression based relationship was established between RR and EOF. A comparative analysis was conducted between the regression relation of RR with EOF and RR with "best RNC descriptor". Results reveal that the use of efficient methodologies for overland RNC does improve the classification accuracy (upto less than 10%). Even though the performance was not too high, it was proven to improve the retrieval accuracy.











Elising the best results from RNC non linear regression relations were developed between the highly correlated channel combination and NSR. The relation is shown in Figure 5.

>For TMI data, first two principal components were deemed sufficient to explain the variability of the data. Hence, polynomial regression based relation, linking first two principal components with NSR is shown in Figure 6



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