# MONITORING OF CROP GROWTH AND SOIL MOISTURE RETRIEVAL USING MULTI-SENSOR SATELLITE DATA



Thesis submitted in partial fulfillment for the Award of Degree of Aphilosophy

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# LIST OF ABBREVIATIONS

yntax	EIST OF ABBREVIATIONS
VIR	Description
AR	Near- infrared
TIR	Synthetic Aperture Radar
R	Thermal infrared
	Infrared
iDAR	Light Detection 1.5
WIR	Light Detection and Ranging
/WIR	Short wave infrared
WIR	Mid wave infrared
	Long wave infrared
TIR	Far infrared
LISS	Linear Imaging Self-Scanner
CSI	Classification Success Index
CC	
OLI	Jaccard's Coefficient of community
RISAT-1	Operational Land Imager
VWC	Radar Imaging Satellite-1
PH	Vegetation water content
	Plant height
LAI	Leaf area index
SVM	Support vector machine
ANN	Artificial neural network
RF	Random forest
ML	Maximum likelihood
DT	Decision tree
ГМ	Thematic Mapper
MODIS	Moderate Resolution Imaging
AVUDB	Spectroradiometer
AVHRR	Advanced Very High Resolution Radiometer
LWAI	Leaf water area index
RTM	Radiative transfer models
WCM	Water cloud model
RFR	Random forest regression