Soil moisture variability correlation with remotely sensed GLDAS Data using SWAT model output data for Upper Godavari River basin.

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## INTRODUCTION

#### Importance of Soil Moisture

- > Key to understanding the hydrological cycle.
- Monitoring of Crop-Water requirement.
- > Water resource management.
- To assess above and below ground water resources pertinent to Earth's climate system.
- To understand the implications of soil water storage on water and energy fluxes at land surfaces and atmosphere.
- > For estimating surface and groundwater availability and retrievals.



#### **Disadvantages of Conventional Methods**

#### Gravimetric Methods

- Destructive
- Requiring labour
- Time consuming

#### Neutron Probes

- High cost
- Licenced labour required
- Delay in receiving readings/reports

#### > Tensiometry

- Only allow indirect estimation of SMC
- Fragile
- Automated operation impractical

#### Gypsum blocks

- Requires labour input to take readings or download to computer
- Lifetime can be reduced in some soil types

## **USE OF REMOTE SENSING DATA**

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Challenges of direct soil moisture measurements

Expensive
 Necessity of using surface meteorological observations

- Not readily available over large areas
- Produce point type measurements
- Restricted to specific locations

Satellite remote sensing offers a means of measuring soil moisture

- Across a wide area
- Continuous

#### **Global Land Data Assimilation System (GLDAS)**



- Spatial Resolution -0.25°
- Collecting soil moisture since 2000
- Soil moisture is mapped monthly

## OBJECTIVE

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- Extracting soil moisture data from Global Land Data Assimilation System
- Calculating spatial variability of soil moisture of study area
- Calculating the soil moisture using SWAT-MODEL
- Establishing the spatial correlation between GLDAS data and SWAT-MODEL output data.

#### **STUDY AREA**

❑ Sub-Basin – Godavari Upper
 ❑ Area – 21443.23 km<sup>2</sup>



#### **METHODOLOGY**



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45.00 40.00 Soil Moisture (0-10cm) 35.00 30.00 25.00 20.00 15.00 10.00 5.00 0.00 2000 2002 2003 2004 2005 2007 2008 2009 2010 2011 2012 2013 2014 2015 2016 2001 2006 Post-Monsoon RABI Pre-Monsoon Monsoon Post-Monsoon Kharif

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# Development of SWAT-model for groundwater recharge estimation





### **RESULTS AND DISCUSSION**

	SWAT OUTPUT (SW)	
Post-Monsoon		
RABI	Pearson	Spearman
SM_0-10 cm	0.40	0.32
SM_10-40 cm	0.56	0.27
SM_40-100 cm	0.57	0.31
SM_100-200 cm	0.64	0.51
SM_RZ	0.55	0.28

	SWAT OUTPUT (SW)	
Monsoon	Pearson	Spearman
SM_0-10 cm	0.56	0.28
SM_10-40 cm	0.51	0.53
SM_40-100 cm	0.48	0.57
SM_100-200 cm	0.48	0.68
SM_RZ	0.50	0.56

#### SWAT OUTPUT (SW) **Pre-Monsoon** Pearson

SM_0-10 cm	0.36	0.27
SM_10-40 cm	0.62	0.65
SM_40-100 cm	0.34	0.35
SM_100-200 cm	0.34	0.35
SM_RZ	0.49	0.54

	SWAT OUTPUT (SW)	
Post-Monsoon		
RABI	Pearson	Spearman
SM_0-10 cm	0.66	0.44
SM_10-40 cm	0.74	0.55
SM_40-100 cm	0.72	0.47
SM_100-200 cm	0.75	0.46
SM_RZ	0.72	0.48

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Spearman

## CONCLUSION



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The spatial and temporal correlation of soil moisture between GLDAS-DATA and SWAT-MODEL DATA are harmonize in nature.

0-10cm	Pre-monsoon
100-200cm	<b>Post-monsoon RABI and KHARIF</b>

#### REFERENCES

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- Alvarez-Garreton, C., Ryu, D., Western, A., Crow, W.T., Robertson, D.E., 2014. The impacts of assimilating satellite soil moisture into a rainfall–runoff model in a semi-arid catchment. Journal of Hydrology, 519, 2763–2774.
- AlvarezGarreton, C., Ryu, D., Western, A., Su, C.H., Crow, W.T., Robertson, D.E., Leahy, C., 2015. Improving operational flood ensemble prediction by the assimilation of satellite soil moisture: comparison between lumped and semi-distributed schemes. Hydrology and Earth System Sciences Discussions, 11, 9, 10635–10681.
- Aubert, D., Loumagne, C., Oudin, L., 2003. Sequential assimilation of soil moisture and streamflow data in a conceptual rainfall-runoff model. Journal of Hydrology, 280, 1, 145–161.
- Brocca, L., Moramarco, T., Melone, F., Wagner, W., Hasenauer, S., Hahn, S., 2012. Assimilation of Surface-and Root- Zone ASCAT Soil Moisture Products Into Rainfall–Runoff Modeling.Geoscience and Remote Sensing, IEEE Transactions on 50(7): 2542–2555.
- Brocca, L., Moramarco, T., Dorigo, W., Wagner, W., 2013. Assimilation of satellite soil moisture data into rainfallrunoff modelling for several catchments worldwide. Geoscience and Remote Sensing Symposium (IGARSS), 2013 IEEE International, IEEE.
- Chen, F., Crow, W.T., Starks, P.J., Moriasi, D.N., 2011. Improving hydrologic predictions of a catchment model via assimilation of surface soil moisture. Advances in Water Resources, 34, 4, 526–536.
- Yongwei Liu., Wen Wang., Yiming Hu., 2017. Investigating the impact of surface soil moisture assimilation on state and parameter estimation in SWAT model based on the ensemble Kalman filter in upper Huai River basin.

## THANK YOU