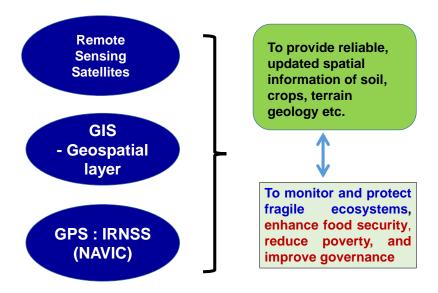
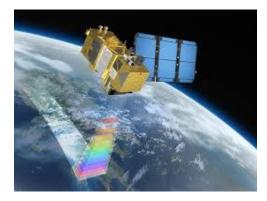


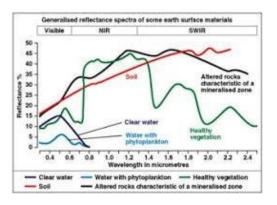
# **Geospatial Technology**



#### **Remote Sensing is Science and Art.**

It is science of obtaining spectral reflectance of object in several spectral bands and art to interpret these spectral reflectance / properties to derive information of the object.





#### Remote Sensing Programme has completed nearly 50 years.

First RS satellite : ERTS (Landsat MSS ) - NASA (USA) - July 1972

India - IRS LISS-1 & LISS II - ISRO- March 1982

Landsat 8 satellite: 07 spectral bands ,resolution is 30 m for bands 1 to 7. (Thermal infrared band 6 was collected at 120 m, but was resampled to 30 m. PAN (8<sup>th</sup> band) : 15 m (0.51-0.89 micro meter.

European Union's Sentinel-2A satellite 13 Multi spectral bands Download (Since Dec. 2016) : <u>EarthExplorer</u> ESA Sentinels Scientific Data

Sentinel-1 synthetic aperture radar (SAR) data processed by the European Space Agency



SLV-C40/Cartosat-2 Series , launched on Jan 12, 2018



**RISAT 1A & 1B** 

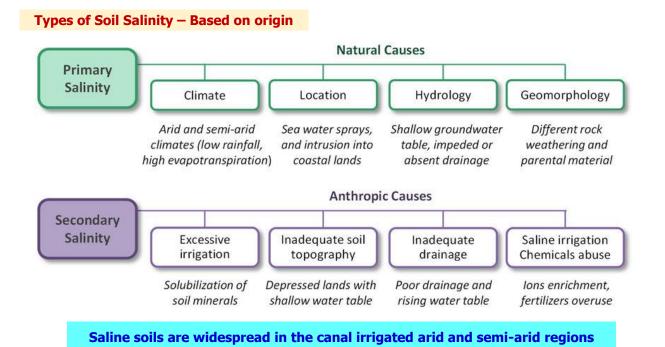
#### **Spatial Resolution** IIRS **RS Satellite** Spatial Resolution **Temporal Resolution** (Pixel Size – meter) (Days) Resourcesat 2 (2011) **Optical Satellite** Resourcesat 2A (2017) **LISS IV** 5.8 24 LISS III 23.5 24 5 AWiFS 56 PAN (1m) Cartosat 2A & 2B -**IKONAS (USA-2000)** PAN (0.8m); MS (4 m) 3-5 PAN (0.3m); MS (1.2m) World View- 3 (2014) 1-5 Planet Lab (USA 2014) Daily MS (5m); PAN (3m) Sentinal 2 A & B MS (10m) 5 Landsat 8 MSS MS (30 m); PAN (15m); 16 **TIRS (100m) Microwave Satellite** Sentinel 1 A & B C Band (5-25m) 12

C Band (5-25)

# Multispectral remote sensing data for mapping salt-affected soils

12

Image	Spatial resolution	Bands
Landsat Operational Land Imager (OLI)	30 m	Band 1 (Blue); Band 2 (Green) Band 3 (Red); Band 5 (NIR); Band 6 (SWIR1); Band 7 (SWIR2)
Sentinel 2A	10 m	Band 2(Blue); Band 3(Green); Band 4 (Red) Band 8(NIR)
	20 m	Band 11 (SWIR 1); Band 12 (SWIR 2)
MODIS (MOD09GA V6	500 m	Band 3(Blue); Band 4(Green); Band 1(Red) Band 2 (NIR); Band 6(SWIR 1); Band 7(SWIR 2



Characteristics of different soil salinity classes based on
saturation paste extracts

Salinity class	EC	ESP* *	рН	SAR	Dominant ions and their concentrations	Gypsum *	Carbonates of earth metals*
Non-saline, non-alkaline (normal)	< 4	< 15	<7<	Low		+, -	+, -
Saline	> 4	< 15	< 8.5	Medium	Cl <sup>-</sup> , SO <sub>4</sub> <sup>2-</sup> , HCO <sub>3</sub> <sup>-</sup> , - Low, CO <sub>3</sub> <sup>2-</sup> - absent, Na <sup>+</sup> > Ca <sup>2+</sup> + Mg <sup>2+</sup>	+	+
Alkaline	Mediu m	> 15	Usually > 8.5	high	Na <sup>+</sup> , may be present in the form of NaHCO <sub>3</sub> <sup>2-</sup> - NaCO <sub>3</sub>	Rare +	
Saline – Alklaine	> 4	> 15	Usually < 8.5		Like in case of saline, but with higher concentration of sodium		

USDA Handbook No. 60, 1954

Note: \* + present, - absent, \*\* the relative amount of the sodium ion expressed as a percentage (%) of the CEC or the sum of exchangeable bases

# Areal Extent

# Total degraded land (2010-11) is 120.40 Mha

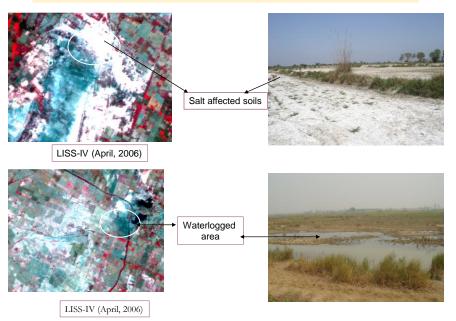
S. No.	Type of degradation	Area in Mha	% of Total Degraded lands
1.	Wind and Water Erosion	94.87	79
2.	Acidic soil	17.93	15
3.	Alkaline soil	3.70	3
4.	Saline soil	2.73	1.8
5.	Water logged	0.91	1
6.	Mining & Industrial waste	0.26	0.2
7.	Total	120.40	100.0

# Salt-affected Soils

- Variations in spectral reflectance of salt crusts of salt-affected soils affected by the physico-chemical properties of soil (e.g., soil moisture content, organic matter, texture, types of clay, color and surface roughness).
- The spectral reflectance of the salt features at the soil surface used as a direct indicator of soil salinity using remote sensing
- **Mapping soil salinity** based on only the soil surface **is limited**, despite decades of research (Allbed and Kumar, 2013; Metternicht and Zinck, 2013).
- Most notably, surface salts can only be readily detected in satellite data if the soils are sufficiently dry (salt content > 10-15%)

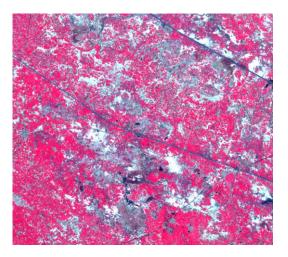
- Remote sensing is used for soil salinity mapping already for years (Metternicht and Zinck, 2009). Nevertheless, there are still <u>no universally acceptable methods</u> to derive soil salinity parameters from remote sensing data that can be used for different environments.
- Sub-surface salinity is not always associated with visible surface salts (surface reflectance is obstructed by overlying vegetation.
- Studies do not demonstrate the same high accuracy in different parts of the world, which means that scaling up to a global scale is problematic.

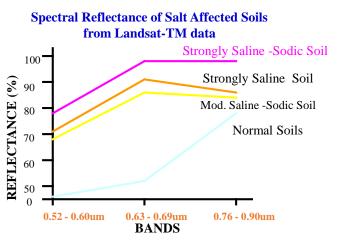
Approach for mapping salt-affected soils							
Input data requirements							
Data	Data type	Variables	Units				
Soil data		EC	dS/m				
		pH (H2O)	-				
	Georeferenced soil profile data (between 0-100 cm of soil depth)	ESP	%				
		Soluble ions*	Cmol/kg				
		TSS*	g/l				
Soil		Rainfall	mm				
forming factors	Climate (Mean annual)	Min Temperature	°C				
lactors	(nour unitual)	Max Temperature	°C				
	Land use/cover	Cover/use types	-				
	Soil map	Soil types	-				
	DEM	Elevation	m				
		Visible (RGB) reflectance	-				
	Remote sensing land surface reflectance	IR reflectance	-				
		SWIR reflectance	-				
	Geology	Lithology types	-				
Other data	Hydrogeology*	Groundwater level	m				
	Degradation*	Degradation types	-				
	Distance to the coastline	Distance	m				
	FAO 2020. Mapping of salt-affected soils: Technical specifications and country guidelines						

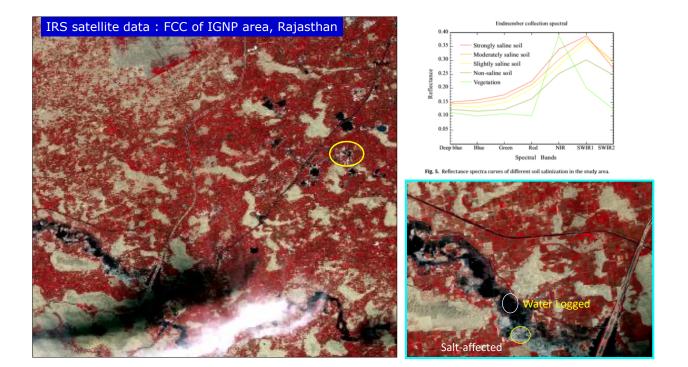


# RS of Salt-affected soils and waterlogged areas

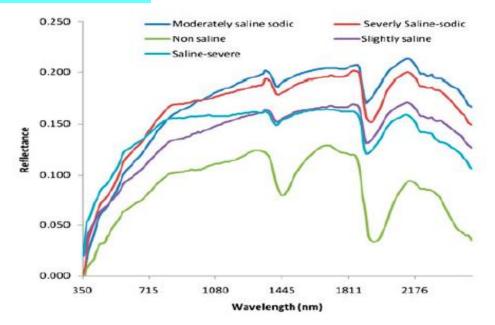
# Salt affected and water logged land

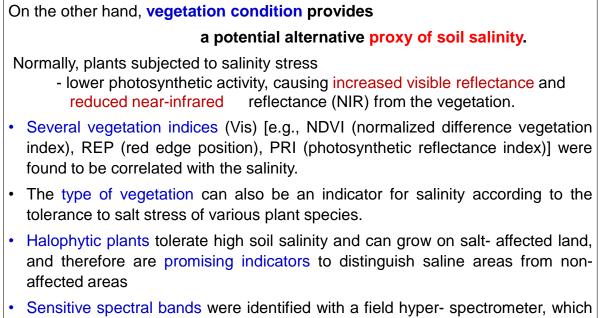




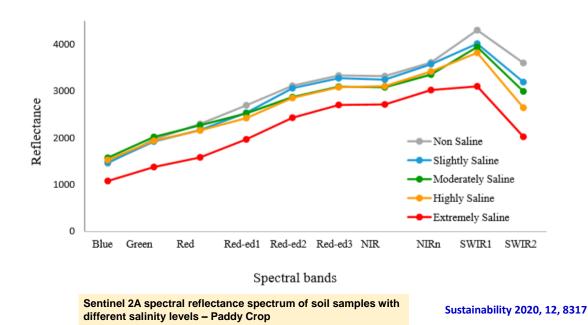


# **Spectral Characteristics**

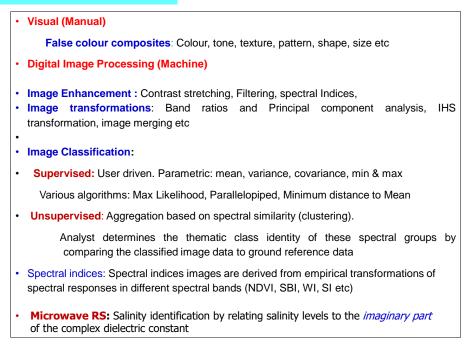




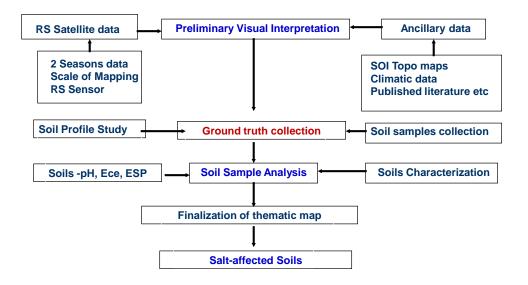
could enhance monitoring salinization across various vegetation types.



#### RS data analysis and interpretation methods



#### Methodology For Land Degradation Studies Visual Interpretation



#### **Digital Image Processing**

#### **Image enhancement**

- Contrast stretching
- Spatial filtering
- Spectral indices
- Principal components analysis

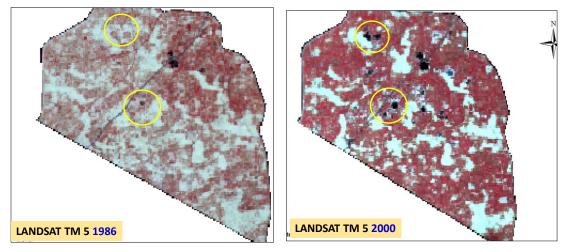
## **Digital Image classification**

- Supervised classification are : Maximum Likelihood Classifier (MLC), Minimum Distance to Mean Classifier (MDMC) and Parallelopiped Classifier (PC).
- In unsupervised classification approach, the image data are first classified by aggregating them into the natural spectral groupings or clusters present in the scene using clustering algorithms.

Spectral Indices	Formula
Salinity index (SI)	SI= (B4×B2) <sup>0.5</sup>
Salinity index (SI1)	SI1= (B4×B3) <sup>0.5</sup>
Salinity index (SI2)	$SI2=[(B5)^2+(B4)^2\times(B3)^2]^{0.5}$
Salinity index (SI3)	SI3= [(B4) <sup>2</sup> +(B3) <sup>2</sup> ] <sup>0.5</sup>
Salinity index I	B2/B4
Salinity index II	(B2-B4)/(B2+B4)
Salinity index III	B3×B4/B2
Normalized Difference Salinity Index (NDSI)	NDSI = (red - NIR)/(red + NIR)
Normalized Difference Vegetation Index (NDVI)	NDVI = (NIR - red)/(red + NIR)
Enhanced vegetation index (EVI)	g×(B5-B4)/(B5+C1×B4-C2×B2+L)
Soil Adjusted Vegetation Index (SAVI)	$SAVI = (1 + L) \times NIR - red/L + NIR + red$
Vegetation Soil Salinity Index (VSSI)	VSSI = 2 × green - 5× (red + NIR)
Tasseled cap transformation for Landsat TM	BI=B1*0.3561+ B2*0.3972 + B3* 0.3904 + B4*0.6966 +
Brightness Index (BI)	B5*0.2286 + B7*0.1596
Greenness Index (GI)	GI= - B1*0.3344 – B2*0.3544 – B3*0.4556 + B4*0.6966 -
	B5*0.0242 – B7*0.2630
Wetness Index (WI)	WI= B1*0.2626 + B2 *0.2141 +B3* 0.0926 + B4 *
	0.0656 – B5* 0.7629 – B7*0.5388

#### MONITORING OF LAND DEGRADATION USING TEMPORAL SATELLITE DATA

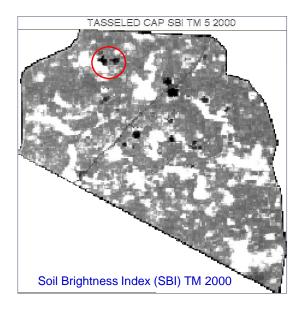
IN IGNP COMMAND AREA ,HANUMANGARH, RAJASTHAN

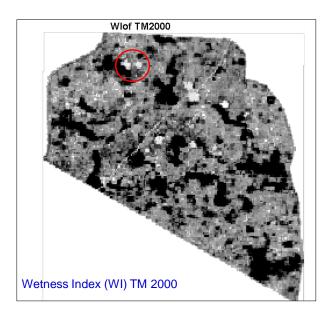


#### **Geographical location :**

74<sup>0</sup> 22'12''E-29<sup>0</sup> 27'36''N 74<sup>0</sup>34'12''E-29<sup>0</sup>21'36''N.

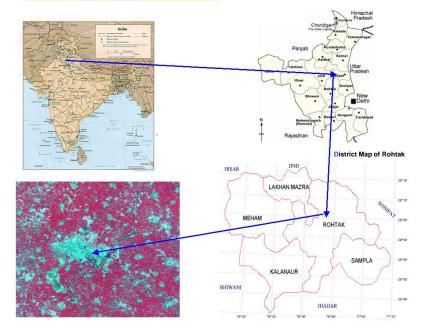
Hanumangarh District of Rajasthan



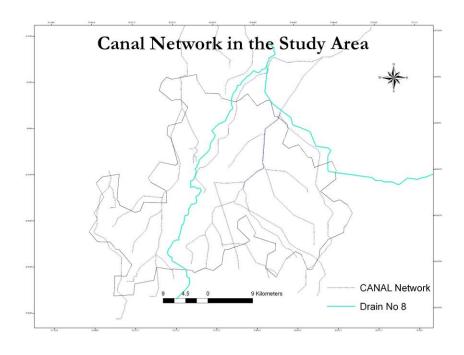


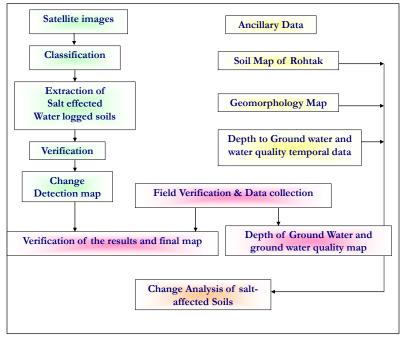
### Change Analysis using Temporal Landsat data

- Post classification comparison method
  - Supervised classification of various types of land cover levels including salt-affected soils between two image dates
- Spectral enhancement methods
  - PCA combination PCs to generate FCCs
  - Multi-date PCA
- Tasseled Cap transformations
  - Soil brightness Image (SBI),
  - Wetness Image (WI)
  - Greenness image (GI)
- Hybrid FCCs : combination of PCA and Tasseled cap transformation images

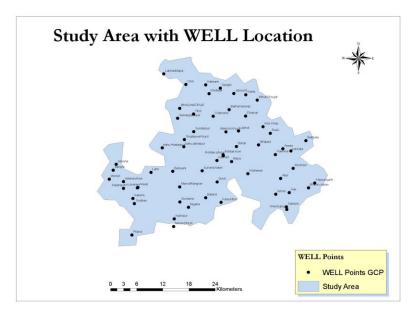


#### A case study : Rohtak (Haryana)

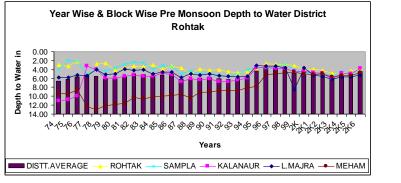


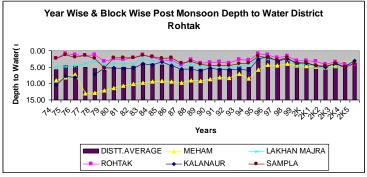


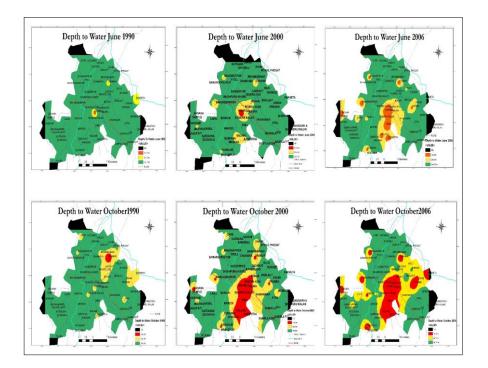
Flowchart showing Methodology:

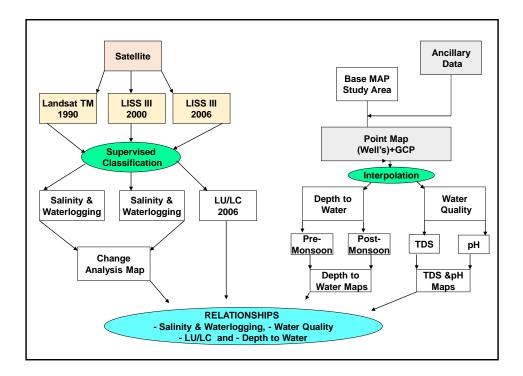


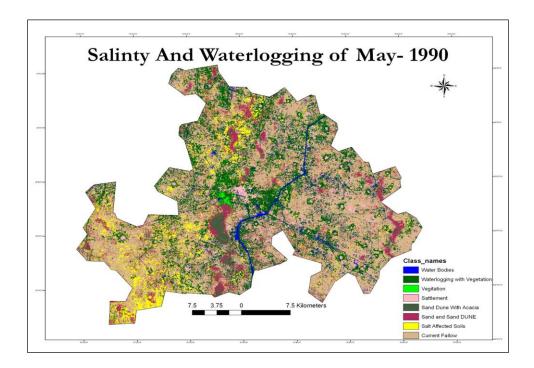
Point coverage showing the exact location of each observation well. (Source: Ground Water cell, Agriculture Department Rohtak Haryana.)

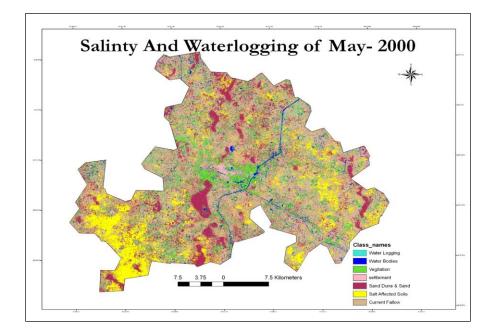


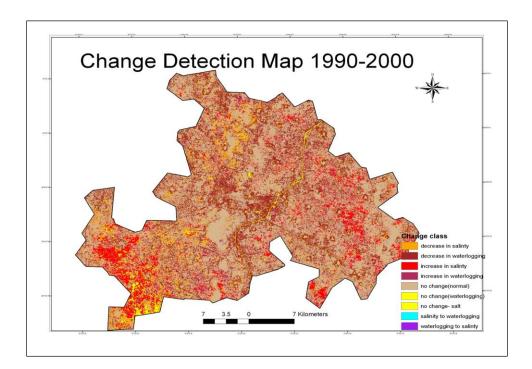


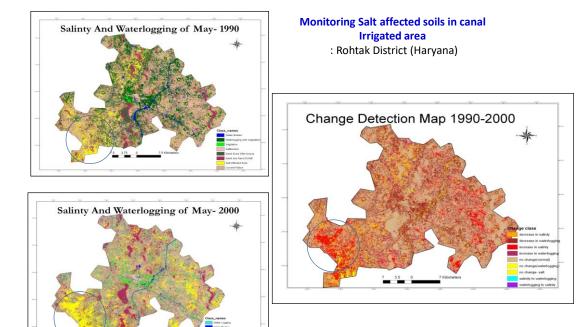












- Hyperspectral remote sensing (HRS) is emerging as a promising tool for these application owing to its capability to measure the reflectance of earth surface features
- Advantage Hyper spectral data can improving the accuracy in quantitative estimation of soil chemical compounds

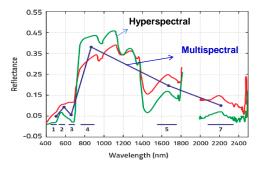
• Due to continuous spectral bands HRS contains the exact information over the full length of the spectrum, where as in MRS information lies only in the discrete patches of spectrum thus lacks information content.

• Specific chemical compounds has specific absorption feature at specific wavelength, hence narrow band width helps to pick those specific absorption feature unlike the broad bands in MRS

Thus Hyper spectral data can improving the accuracy in quantitative estimation of soil chemical compounds

## Hyperspectral Vs Multispectral Remote Sensing

- Powerful and versatile tool for characterization, mapping and monitoring of salt-affected soils as it records continuous sampling and the high spectral resolution (<5 nm).</li>
- Multispectral broadband sensors (>50 nm) may lose important spectral information, while narrow bands can discriminate critical spectral differentials in detail.



Hyperspectral provide spectral signature that is unique of one object

Hyperspectral sensing technology should be preferred when it comes to sense chemical and physical properties of an object.

# Remote Sensing of Salt-affected Soils

- Bare soil surface with salt incrustation
- · Vegetation as proxy for monitoring salinity
- Saline soils contain liquid water molecules in their crystal lattice, the absorption range of liquid water is useful to estimate soil salinity.
- With increasing salinity, the depth and width of the absorption increase especially to the long wavelength side, and spectral feature shows "shoulder" around 1970-1980 nm.
- BenDor et al. (2009) showed that the position of peak depth around 1400 and 1900 nm shifted to long wavelength with increasing soil salinity.

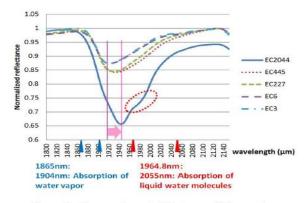


Figure 2. Change of spectral features with increasing soil salinity.

# Hyperspectral vegetation indices as a proxy

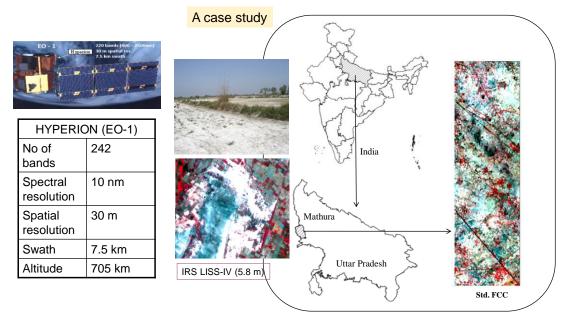
- Vegetation stress responses to chlorophyll reduction and cell structure damage
- Increased visible reflectance (VIS) and the reduced near-infrared reflectance (NIR)

## Study shows:

- Wavelengths at 395–410, 483–507, 632–697, 731–762, 812–868, 884–909, and 918–930nm were determined to be the most sensitive bands (VNIR).
- By combining the most sensitive bands in a SAVI form, four soil adjusted salinity indices (SASIs) for all plant species (Halophytes and Non- Halophytes).
- Findings indicate the potential to monitor salinity with the hyperspectra of saltsensitive and halophyte plants.

- IOP Conf. Series: Materials Science and Engineering **274** (2017) - Zhang et al. (2011). Ecological Indicators 11 : 1552–1562

# Spectral Characterization and Mapping of Salt-Affected Soils Using Hyperspectral Satellite



#### Objectives

- 1. To study soil laboratory spectra to characterize soil salinity using portable Field Spectrometer and to compare with the Hyperspectral satellite derived spectra
- 2. Characterization and mapping of salt-affected lands using Hyperspectral satellite data with the following sub-objectives:

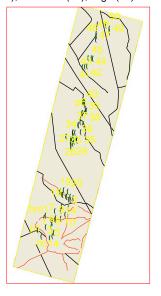
(a) Sorting of suitable spectral bands for characterizing salt - affected soils

(b) Study relationships between hyperspectral indices and soil salinity and mapping salinity severity

(c) Spectral unmixing analysis for mapping various kinds of salt-affected lands.

(d). SVM technique in mapping severity of soil salinity

• Soil sampling plan : Three transects : No. of samples: 64 : Normal (14), Slight (14),Moderate (15), High (21)



EC 1:2 (ds/m) and pH						
Bands	Wavelength (nm)	R <sup>2</sup> EC ( ds /m)	R <sup>2</sup> pH			
Band 9	436.99	0.6977	0.5576			
Band 10	447.17	0.6839	0.5279			
Band 11	457.34	0.672	0.5203			
Band 13	477.69	0.6779	0.5349			
Band 14	487.87	0.6813	0.5204			
Band 16	508.22	0.6744	0.539			
Band 19	538.74	0.639	0.5321			
Band 20	548.92	0.6359	0.5321			
Band 22	569.27	0.6151	0.5208			
Band 26	609.97	0.5973	0.5043			
Band 28	630.32	0.5976	0.5058			
Band 29	640.50	0.5917	0.5083			
Band 30	650.67	0.5887	0.5053			
Band 40	752.43	0.5761	0.5125			
Band 46	813.48	0.5717	0.5226			

#### Sensitive Bands based on the Correlation of Mean Reflectance Value and EC 1:2 (ds(m) and pH

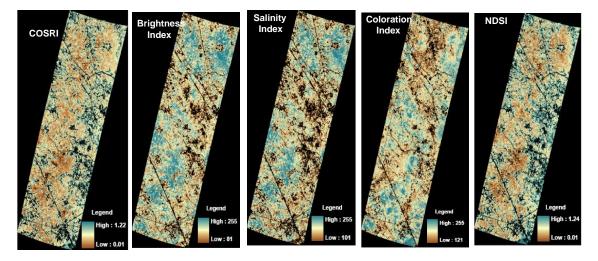
## Formulae of spectral indices based on sensitive spectral bands

S. No	Spectral Indices	Formulae			
1	NDVI	(Band 40(752.43 nm) – Band 30(650.67 nm) / Band 40(752.43 nm) + Band 30(650.67 nm))			
2	COSRI	(Band 9(436.99 nm) + Band 20(548.93nm) / Band 28(630.32 nm) + Band 46(813.48 nm))*NDVI			
3	NDSI	(Band 29(640.50 nm) – Band 46(813.48 nm) / Band 29(640.50 nm) + Band 46(813.48 nm))			
4	BI	$\checkmark$ ((Band 9 <sup>2</sup> (436.99 nm) + Band 20 <sup>2</sup> (548.93nm) + Band 28 <sup>2</sup> (630.32 nm))/3)			
5	CI	(Band 29 (640.50) – Band 22 (569.27) / Band 29 (640.50) + Band 22 (569.27))			
6	SI	√ Band 9(436.99 nm) * Band 28(630.32 nm)			

Spectral Indices	EC	рН	ECe	ESP	SAR
Salinity index	0.814	0.519	0.777	0.804	0.801
Brightness Index	0.773	0.517	0.732	0.796	0.772
COSRI	0.591	0.436	0.524	0.596	0.552
NDSI	0.522	0.228	0.534	0.271	0.389
Coloration Index	0.433	0.312	0.312	0.444	0.451

# Correlation coefficient (r<sup>2</sup>) between spectral indices and soil salinity parameters

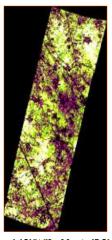
# Spectral Indices



Spectral Indices	No.of randomly selected sites	EC	SAR	ESP
Salinity index	10	7.48	18.14	7.85
Brightness index	10	7.7	33.36	9.60
COSRI	10	11.7	48.21	13.79
NDSI	10	10.4	62.02	27.67
Coloration index	10	21.2	68.67	18.15

#### RMSE : Observed and Predicted EC, SAR and ESP

Hyperspectral remote sensing data derived spectral indices in characterizing salt-affected soils: a case study of Indo-Gangetic plains of India. *J. Environmental Earth Sciences*, DOI:10.1007/s12665-014-3613-y.



y = 1.1518( SI \_ Map )-47.516 SAR Map using Salinity Index

#### Legend

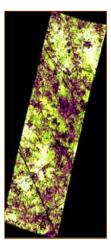
 Normal Soil (6.16 - 12.36)

 Slightly Salt Affected Soil (16.05 - 19.36)

 Mod. Salt Affected Soil (38.06 - 113.99)

 Highly Salt Affected Soil (117.58 - 270.30)

Environ Earth Sci (2015) 73:3299-3308



y = 0.1189 ( SI \_ Map)+4.3019

#### ESP Map using Salinity Index

 Normal Soil (5.48 - 13.16)

 Slight Salt Affected Soil (14.36 - 15.73)

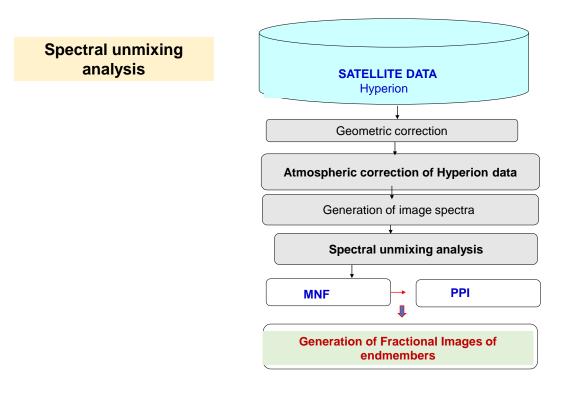
 Mod. Salt Affected Soil (16.27 - 23.95)

 Highly Salt Affected Soil (24.33 - 36.21)

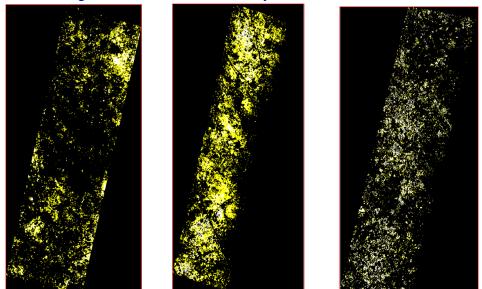


y = 0.0438(SI\_Map)+ 4.347 EC Map using Salinity Index





# Fractional Images of endmembers as salinity levels



Highly Salt Affected Soil

Moderately Salt Affected Soil

l Slightly Salt Affected Soil J. Indian Soc. Remote Sens., 40 (1): 129-136.

# **Future Global Hyperspectral Satellite**

Satellite	Country	Resolution (m)	Spectral bands
HySIS on IMS-2	India (Nov. 2018)	550	64
HySIS on IMS-1	India (May, 2008)	550	64
AHySIS-2	India (2020-24)	30	55 (0.4 to 0.95µm)
EnMAP	Germany (2017)	30	232
EnMAP	Germany (2020)	30	232 (420-2450 nm)
ALOS-3	Japan (2019)	30	VNIR - 57 bands,SWIR- 128 bands

#### India : Future Geostationary Imaging SATellite (GISAT) - 2019-20

Spectral bands	# channels	Resolution (m)	Spectral range	
MX-VNIR	4	50	0.45-0.86	5 minute (0.5 -1.5 Km)
HyS-VNIR	>60	500	0.375- 1.0	and 30 minutes (50 m)
Hys-SWIR	>150	500	0.9-2.5	
MX-LWIR	6	1500	7.0-13.5	

**Thermal infrared imagery** was used to distinguish between different levels of soil salinity on agricultural lands. The principle behind this approach is that **canopy temperature of the plants** grown in affected areas will be higher than of plants growing in non-affected areas.

The approach was tested on regional and local scales and showed its robustness in different climatic conditions and on areas covered with different crops. Therefore, it **seems promising** for use on a global scale.

But needs to establish because of the different climatic zones and extreme temperature differences between regions, and use without normalization will just lead to characterization of climate, rather than soil salinity.

- Ivushkin et al. (2019). Global mapping of soil salinity change. Remote Sensing of Environment 231 (2019) 111260
- Ivushkin, K., Bartholomeus, H., Bregt, A.K., Pulatov, A., 2017. Satellite thermography for soil salinity assessment of cropped areas in Uzbekistan. Land Degrad. Dev. 28, 870–877. https://doi.org/10.1002/ldr.2670.





#### **References:**

- Metternicht, G., Zinck, J.A., 2009. Remote Sensing of Soil Salinization Impact on Land Management. CRC Press, Boca Raton, FL.
- FAO, 2018. Salt-affected soils. http://www.fao.org/soils-portal/soil-management/ management-of-some-problem-soils/salt-affected-soils/more-. information-on-saltaffected-soils/en/, Accessed date: 10 April 2018.
- Allbed, A., Kumar, L., 2013. Soil salinity mapping and monitoring in arid and semi-arid regions using remote sensing technology: a review. Advances in Remote Sensing 02, 373-385. https://doi.org/10.4236/ars.2013.24040.
- Allbed, A., Kumar, L., Sinha, P., 2014b. Mapping and modelling spatial variation in soil salinity in the Al Hassa oasis based on remote sensing . indicators and regression techniques. Remote Sens. 6, 1137–1157. https://doi.org/10.3390/rs6021137.