

## Big Data for building Inclusive Agroecosystems

Decoding the big-data ecosystems for efficiency, ecology, economy and impact

**Chandrashekhar Biradar** 

CGIARCSI Consortium for Scotlaf Information

cgiar.org

**Big Data in Agriculture** 10-14 December, 2018, Rabat, Morocco



icarda.org

International Center for Agricultural Research in the Dry Areas

A CGIAR Research Center CGIAR

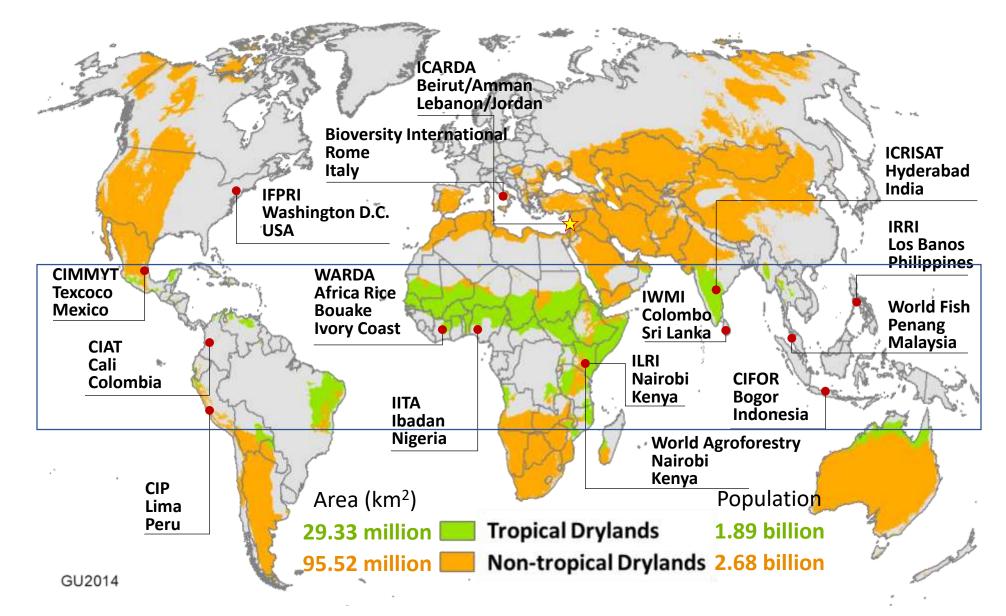
		Big D	ata in Ag	riculture (	Course - Rabat - 10 ·	- 14 December, 2018	}
Sunday 09 Dec 18	Time	Day1: Monday 10 Dec'18	Day2: Tuesda	ay 11 Dec'18	Day3: Wednesday 12 Dec'18	Day4: Thursday 13 Dec'18	Day5: Friday 14 Dec'18
	Theme>>	Big Data Big Picture	Data Collection	n and Curation	Geolocalization & Field Day	All about FAIR Data	Catalyst of Integration
	8:30-9:00	Opening Session and Introduction to Course (A Amri; C. Biradar)	Debriefing p (C. Bir	-	<b>Debriefing previous day</b> (F. Bonechi)	<b>Debriefing previous day</b> (K. El-Shama)	<b>Debriefing previous day</b> (B. Mueller)
	9:00-10:30	CGIAR Approach to Big Data (B. King)	<b>Data Repositor</b> (DV, Dspace, Bona	Schema - E.	<b>Big Data in Agriculture Analytics</b> (S. Faissal)	FAIR Data Principles and Applications (V. Graziano)	CapDev Tools for Better Integration (B. Mueller)
	10:30-11:00	Coffee break	Coffee	break	Coffee break	Coffee break	Coffee break
Arrival, Accomm	11:00-12:30	All About Big Data and ICTs in Agriculture (C. Biradar et al)	Enabling F (H. A		<b>Geotagging Research and</b> <b>Outreach activities</b> (K. El-Shama and C. Biradar)	Morocco Partners (INRA, ONCA, IAV, P6, +) Data and Presentations	Build your own Course in Moodle (B. Mueller)
odation,	12:30-14:00	Lunch break	Lunch	break	Lunch break	Lunch break	Lunch break
pre- planning, logistics	14:00-15:00	AGROVOC Use and Contribution (K. Kolshus, A. Turbati, I. Subiratis)	BCIP, GRS Dat Proce (Z. Ke	esses	Field Exercise and Demos in INRA/ICARDA premises (All)	Data Curation of Existing Datasets (F. Bonechi)	How Big Data Shaping CGIAR Ag Research (J. Koo)
check	15:00-16:00	Harmonizing CGIAR Data (M. Devare)	RALS Data, People and Process (M. Hilali)	WLEP Data, People and Process (M. Haddad)	Field Exercise and Demos in INRA/ICARDA premises (All)	International Nursery, Interaction, Data and Use (A. Niane and K. El-Shama)	Showcase Data on the Genebank (A. Tsivelikas)
	16:00-16.30	Coffee break	Coffee	break	Coffee break	Coffee break	Coffee break
	16:30-18:00	GDPR and IP Rights (A. Nour)	Motadata Spatial Data Curation		Big Data and Tools (GBDX, aWhere, etc)	Ontologies: Why should we use them? (E. Arnaud)	Wrap up, Certification and Closing Ceremony
		Intro Repositories - E.B.			(S. Ghosh, and C Biradar)		(B. Mueller, C. Biradar)
(core group discussion-	19:30-21:00						

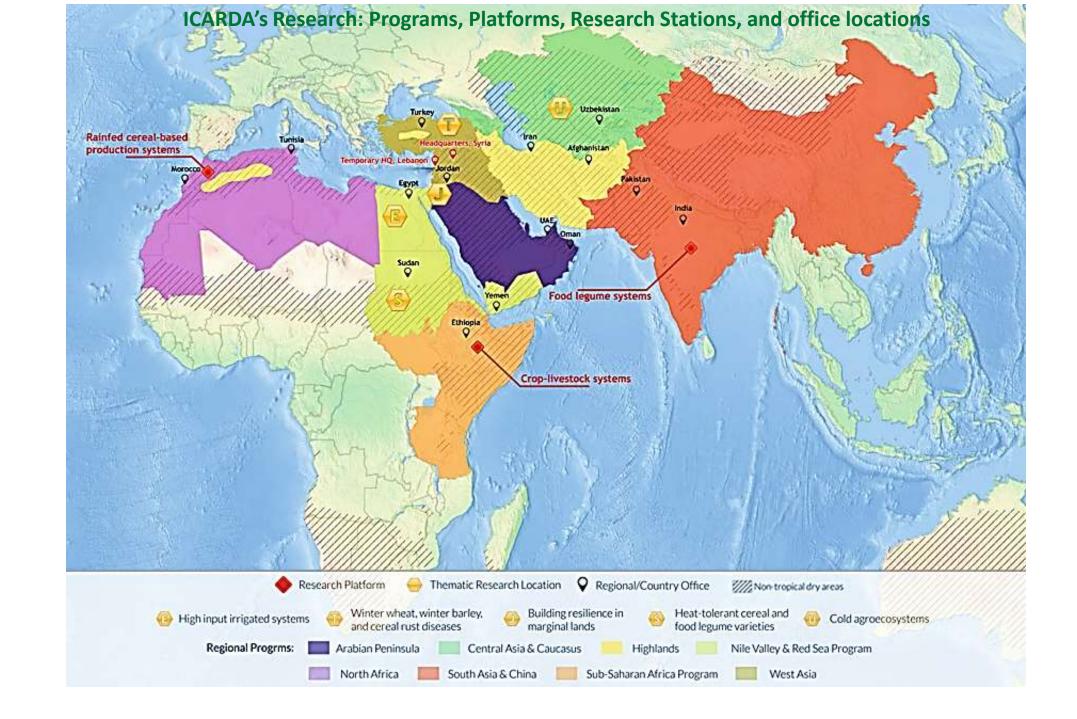
International Center for Agricultural Research in the Dry Areas

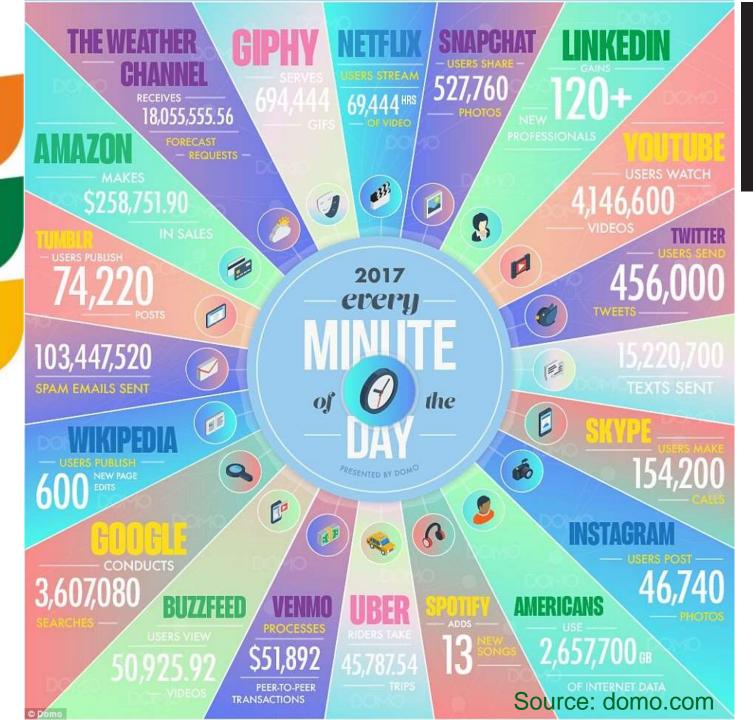
A CGIAR Research Center CGIAR

JUC

## **Global Drylands and CGIAR**





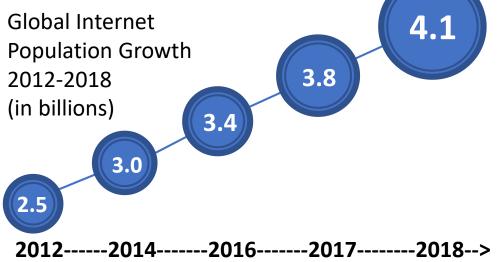


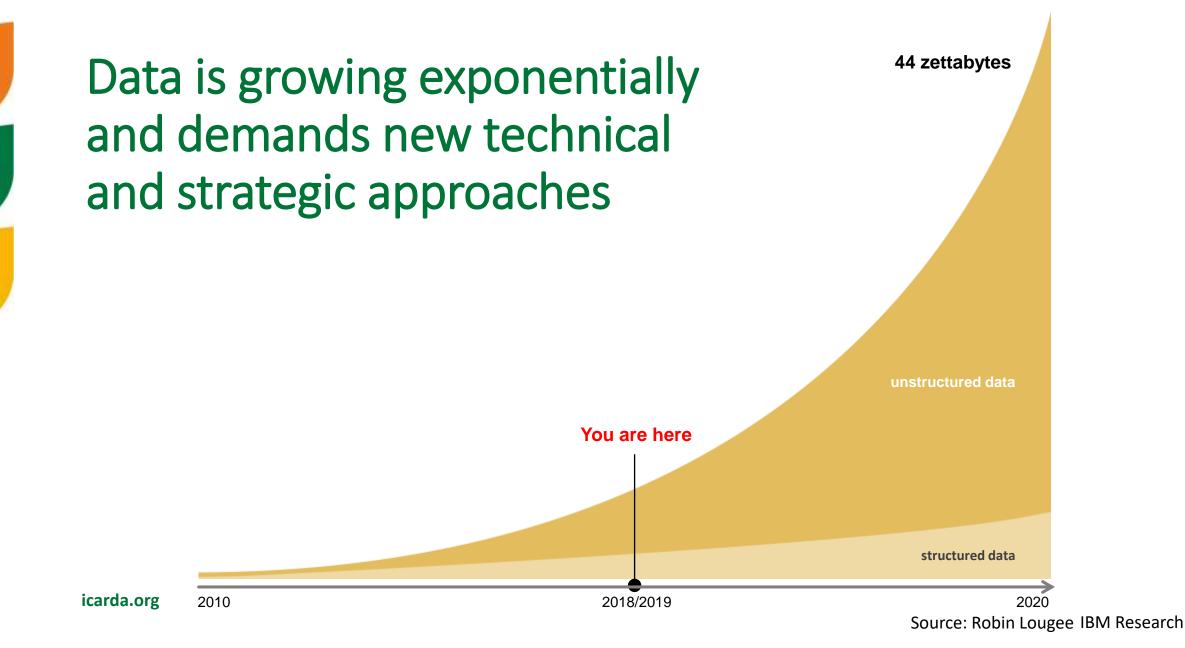
# **DATA NEVER SLEEPS 6.0**

### How much data is generated every minute?

There's no way around it: big data just keeps getting bigger. The numbers are staggering, but they're not slowing down. By 2020, it's estimated that for every person on earth, 1.7 MB of data will be created every second. In our 6th edition of Data Never Sleeps, we once again take a look at how much data is being created all around us every single minute of the day—and we have a feeling things are just getting started.

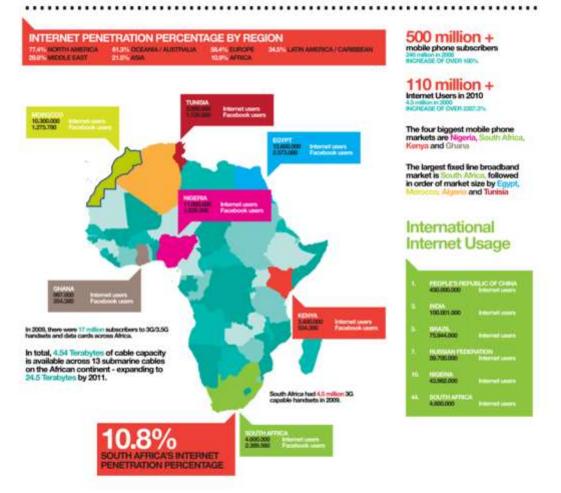
**The World's Internet users growing exponentially**. As of June **2018**, 55 % of the **world's** population has **internet** access. Its estimated nearly 4 billion people, or over half of the **world's** population online by the end of the year.





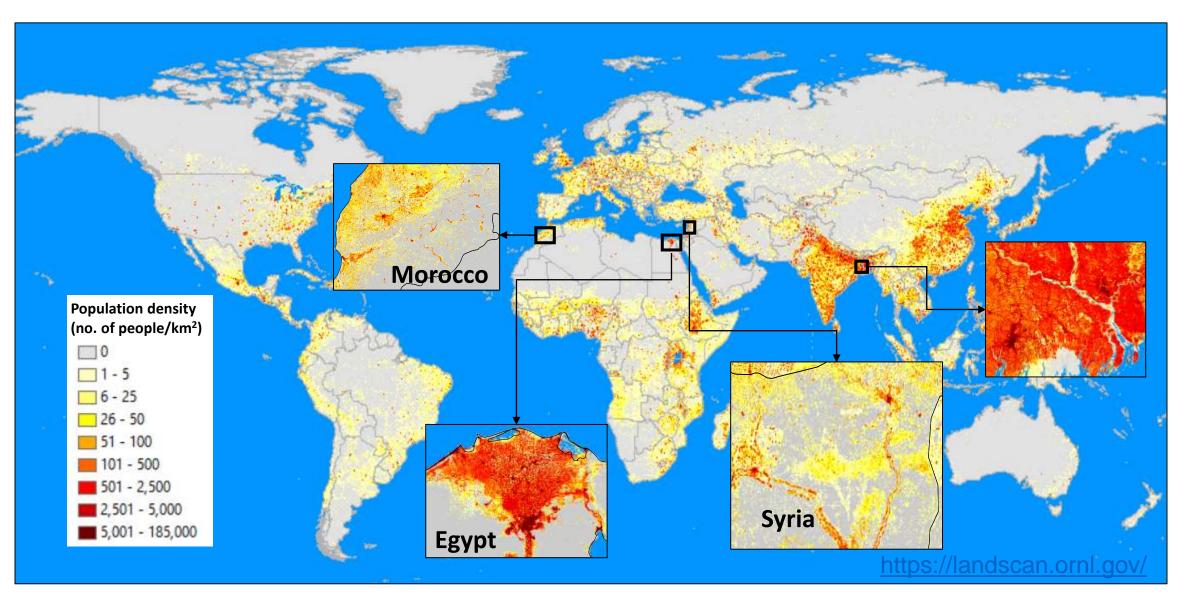
### The African Digital Frontier

1.0

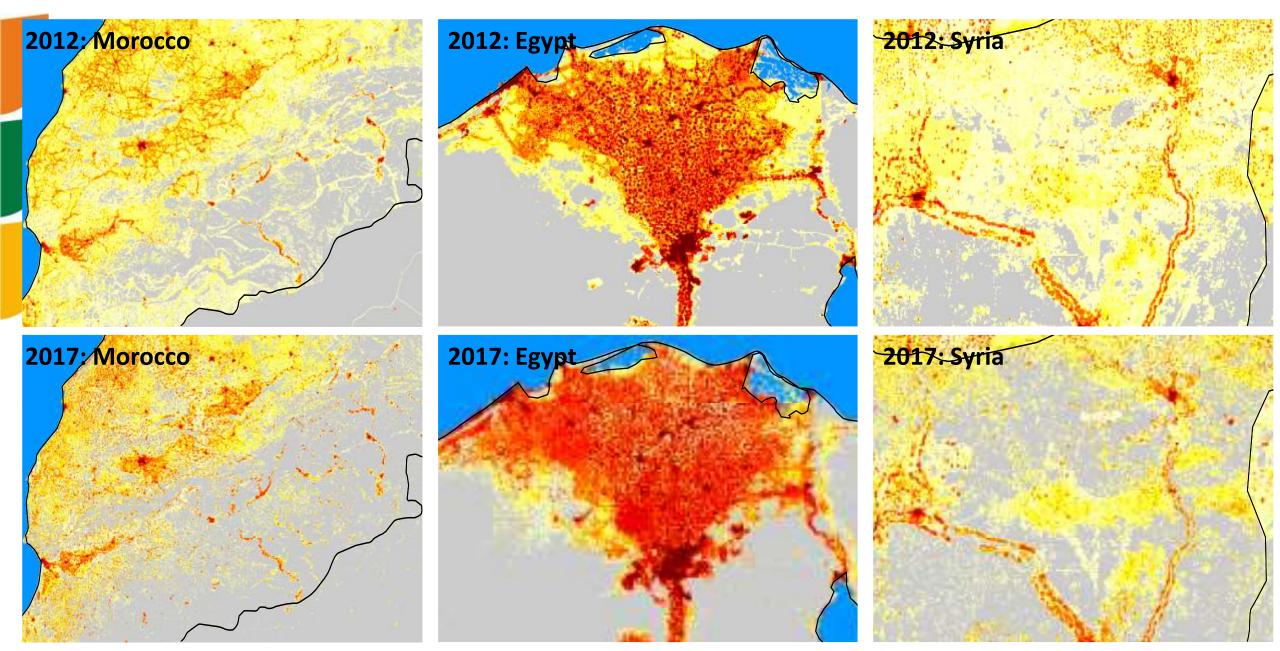


SOURCES: en.wikipedia.org www.internetworkdstats.com	MORE COOLNESS? #ODMACT @opihydma
www.nickburcher.com	Infoorabhic designed by
www.webtrendsng.com	Integraphic designed by Glvanisawesome

## Changing demography and diets



## Migration and aggregated expansion



## Integrated systems for resilience and risk reduction [mixed crops, livestock, fish and trees]

Legume as best alternative for improving diet related resource use efficacy

4,325





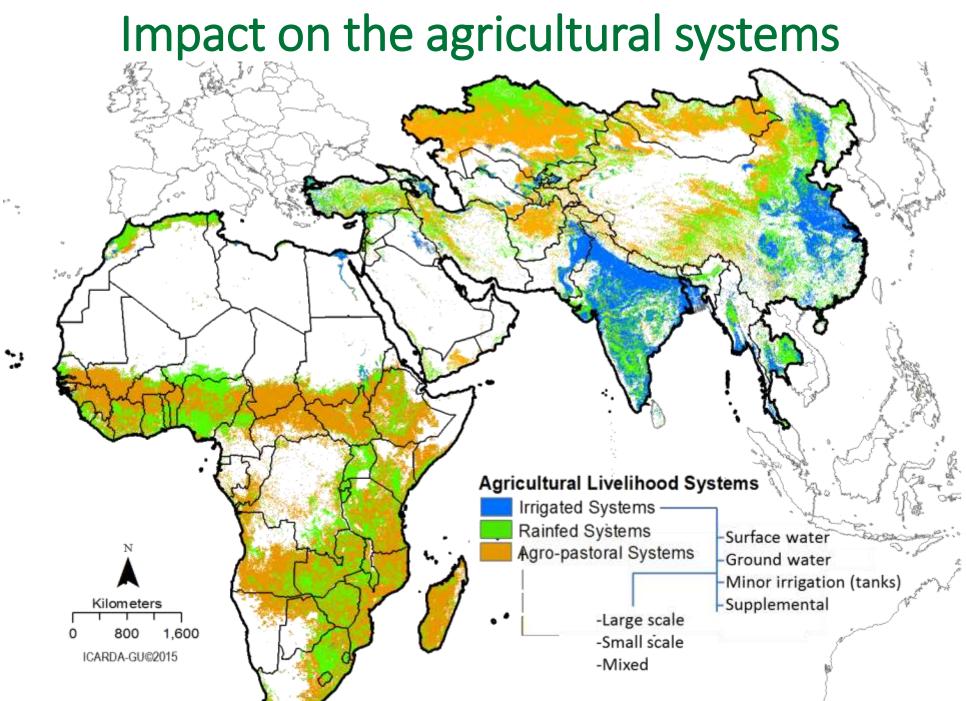




.250

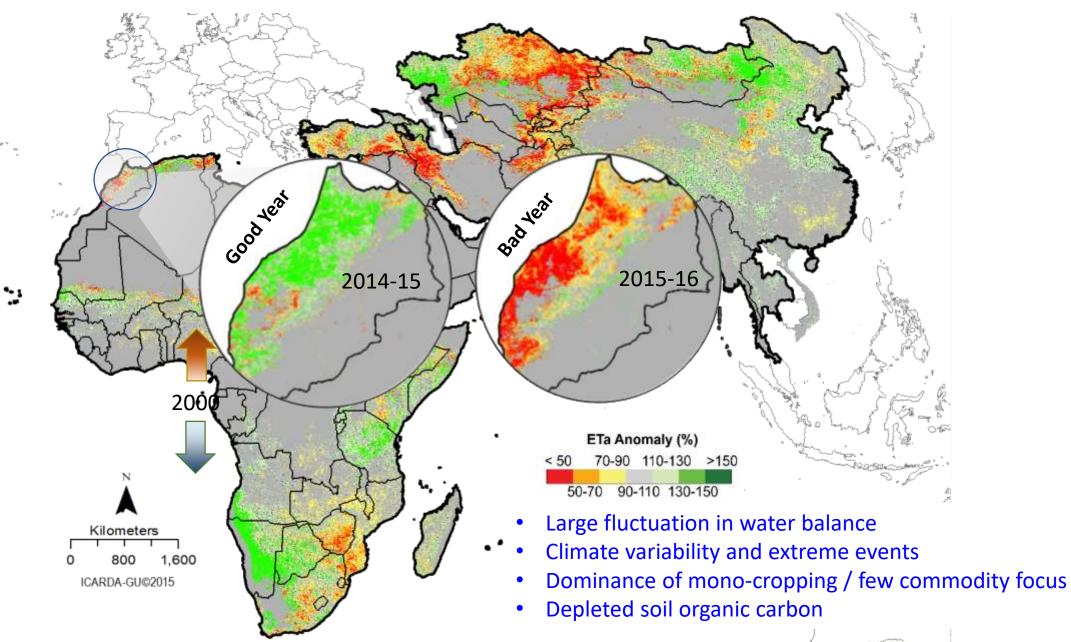
5,520



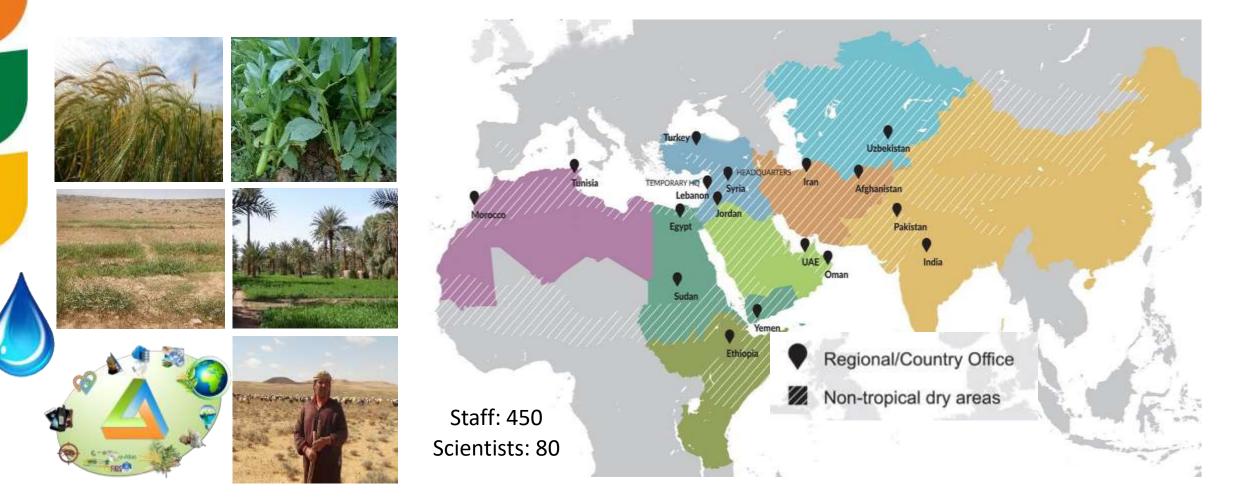


## Changing Water Balance

# Frequent deviation from long-term averages



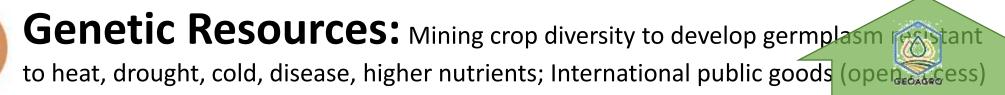
ICARDA is a **Decentralized R4D** International Institute on **Dryland Agriculture** combining **Component** Research and **Systems** Research



### www.icarda.org

icarda.org

## New 9: 5 SRPs + 4 CCTs





Adaption to Climate Change: Conventional and molecu CCTs breeding to develop climate-smart crops and livestock



Building resilience: Integrated crop-livestock farming systems to address economic, social, and environmental conditions



**Promoting value chains, policies:** Agriculture as a gincomegenerating business for many poor smallholder households



Enhancing water, land productivity: Rainfed, irrigated, and

agro-pastoral farming; Reversal of environmental degradation; Enhance intensification

SRPs-strategic research priorities + CCTs-cross cutting themes

### Resilient Agroecosystems Sustainable

Increased land, water and system productivity while safe guarding the environmental flows and ecosystem services

- more <u>crop</u> per <u>drop</u> -water focus

-system focus

- in a **inch of land** and a **bunch of crop** 

Knowledge based prioritization (space & time) for better strategy for investment, intervention, implementation and impact

Ecological intensification Target specific interventions Bridging yield/data gaps Resource use efficiency Nutrition and resilience Halt land/water degradation Technology scaling

- food and nutritional security

- resilience and risk reduction
- agro-ecosystem sustainability
- adaption and mitigation
- citizen science and collective actions
- trade, social security and stability

- Big Data and IC Technologies can support resilient Agri-food systems under Climate Change, Demography, Variability and Uncertainty
- These data driven smart farming systems have huge potential in the dry areas where resource use efficiency is much below its actual potential
- But they can only deliver if applied to Inclusive Farming Systems

**Goal**: to harness the capabilities of **Big Data** to accelerate and enhance the impact of international agricultural research, and *solve development problems faster, better and at greater scale* 



### ORGANIZE

Support and improve data generation, access, and management in CGIAR





### CONVENE

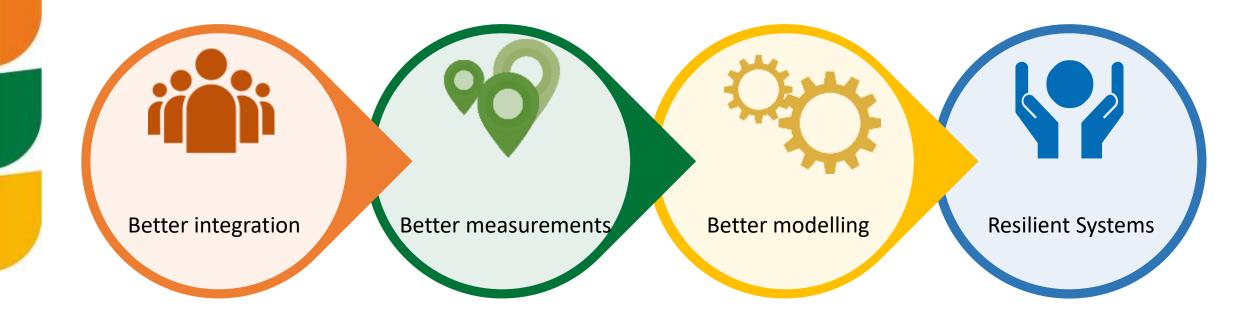
Collaborate and convene around big data and agricultural development

### INSPIRE

Lead by example and inspire how big data can deliver development outcomes

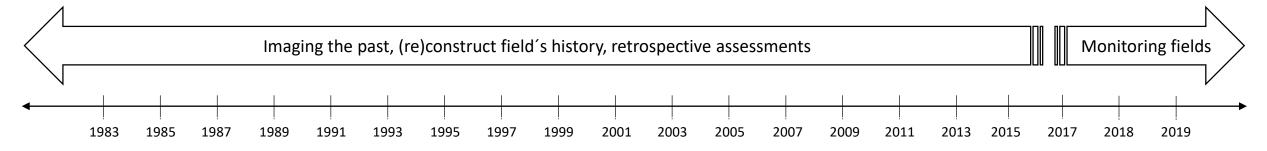
### bigdata.cgiar.org

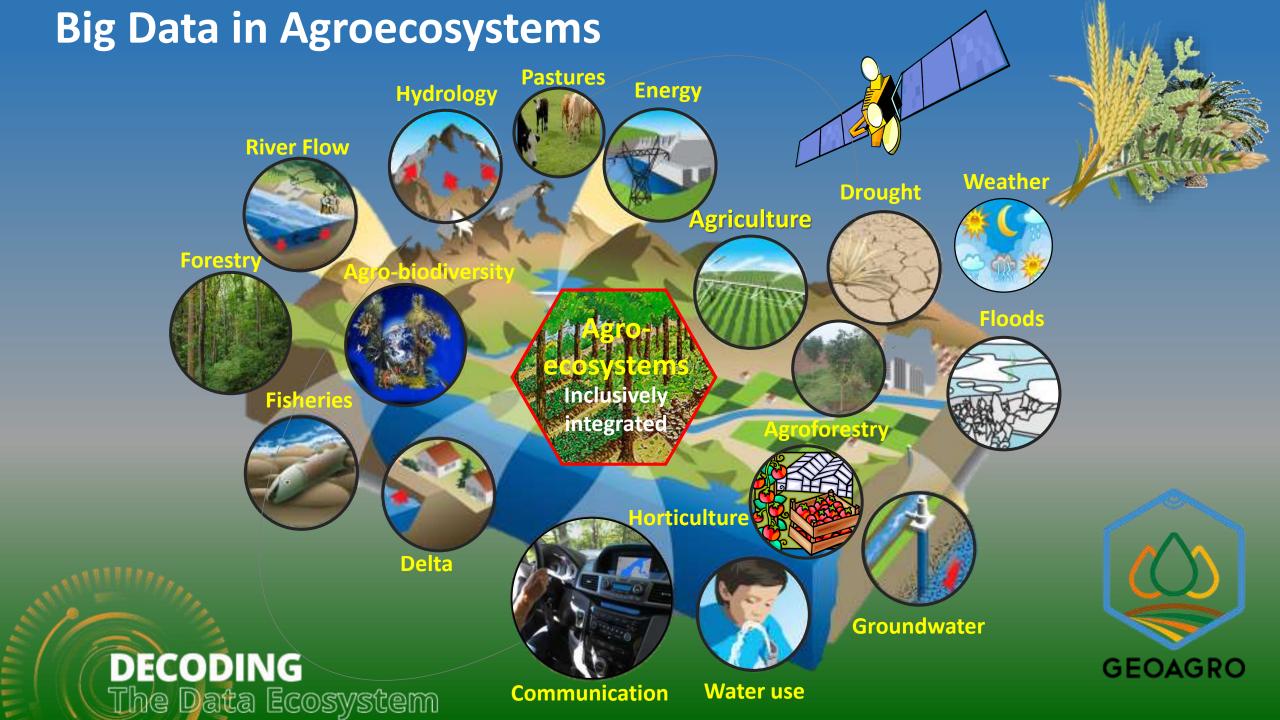
## **Big Data in Building Resilience**



Crop fields for functional productivity are key to exponential efficiency in world largest and oldest industry

### "Agroecosystems"







### Earth Observation Systems for Agro-Ecosystem Research

#### ACTIVE SATELLITE SENSORS AND CHARACTERSTICS



#### Very High Resolution (Up to - 1 m)

Satellite Sensors		Country (Inco		
Satellite Sensors	Spatial (m)*	Temporal (days)	Spectral (Bands)	Swath (km)
GEOEYE-1	1.65 (0.41)	1	B, G, R, IR, P	15.2
IKONOS	3.2 (0.82)	14	B, G, R, IR, P	11.3
PLEIADES-1A	2 (0.5)		B, G, R, IR, P	
PLEIADES-1B	3 (0.5)		B, G, R, IR, P	20
Quick Bird	2.4 (0.6)		B, G, R, IR, P	
WorldView-1	(0.4)	1.2		17.6
WorldView-2	1.8 (0.4)		P, C, B, G, Y, R, RE, IR (2)	16.4
CARTOSAT-2				9.6
CARTOSAT-2a				9.6
CARTOSAT-2B			Р	9.6
SKYSAT-1	2 (0.9)	<1 (hourly)	B, G, R, IR, P	
KOMPSAT-3	2.8 (0.7)	14	B, G, R, IR, P	16.8
KOMPSAT-2			B, G, R, IR, P	
OrbView-3	4 (1)	3	B, G, R, IR, P	14

#### High Resolution (1 to 5 m)

Satellite Sensors=				
	Spatial (m)*	Temporal (days)	Spectral (Bands)	

#### \*=Resolution in parenthesis is panchromatic +=Bands: B-Blue, G-Green, R-Red, IR-Infra Red, C-Coastal blue, Y-Yellow, SW-Shortwave Infrared, M-Mid infrared, P-Panchromatic, H-Horizonal, V-vertcial

#### Medium resolution (5 - 30 m)

Satellite	Multispectral resolution (m)	B, s	Swath width (
ASTER (15m)			
VNIR (Visible Near Infrared)		VIR (4)	60
SWIR (Shortwave Infrared)		SW (6)	
TIR (Thermal Infrared)	60	TIR (5)	60
CBERS - 2			
WFI	260	R, IR	890
CCD		B, G, R, IR	
IRMSS	(2.7)	P-	
LANDSAT 5TM -7ETM	30 (14.8)	B, G, R, IR, SW1, TIR, SW2, P	
Nigeriasat-X	22	G, R, IR	
Resourcesat-2/Liss-III		R, G, IR, SW	
Deimos-1		G, R, IR	600
UK-DMC-2/SLIM6	22	G, R, IR	638
BILSAT-1	26 (12)	R, B, G, IR, P	640
Nigeriasat-1		G, R, IR	640
ALSAT-1		G, R, IR	640
UK-DMC/EC (DMC)	32	G, R, IR	600
EO-1/ALI-MS	30	B (2), G, R, IR (3), SW (2), P	
EO-1/ Hyperion	30	220 bands	7,7
ASTER (15m)	15, 30, 90	G, R, IR (2) SW(6), TIR (4)	60
LANDSAT 7ETM+	30m (14.5)	B, G, R, IR, SW (2), TIR, P	
SPOT-4	20 (10)	G, R, IR, SW, P	60
SPOT-3	20 (10)	G, R, IR+P	60
JERS-1	24 (18)	G, R, IR, IR	75
SPOT-2	20 (10)	G, R, IR	
SPOT-1	20 (10)	G, R, IR	60
Landsat 5/MSS	80	G, R, IR, IR	185
Landsat 5/TM	30, 120	B, G, R, IR, SW, SW, TIR	185
RESURS-01-1	45	G, R, IR	600

#### Low or Medium resolution

Satellite	Multispectral resolution (m)	) B, s	Swath width (km)
Landsat 8	30 (14.8)	P, C, B, G, R, IR, SW (3)	185
	40000		
SPOT5/VEGETATION 2		B, R, IR, SW (4)	
SPOT4/VEGETATION 1		B, R, IR, SW (4)	
Orbview-2/ SeaWiFS		B(2), G (3), IR (8)	
RESURS-01-1/ MSU-S		G, R, IR (3)	
ResourceSat/AWiFS		R, G, IR, SW	740
Landsat 2/ RBV		G, R, IR	
Landsat 1/ RBV	80	G, R, IR	183

Satellite	Bands	Band (Polarity)	Swath width (km)
Sentinel-1			
COSMO-SKYMED 4	1, 5, 15, 30, 100	X-B (HH, VV, HV, VH)	10, 40, 30, 100 200
COSMO_SKYMED 2	1, 5, 15, 30, 100	X-B (HH, VV, HV, VH)	10, 40, 30, 100 200
RADARSAT 2	3, 8, 12, 18, 25, 30, 40, 50 100		
COSMO-SKYMED 1	1, 5, 15, 30, 100	X-B (HH, VV, HV, VH)	10, 40, 30, 100 200
ALOS (PALSAR)	10, 20, 30, 100	L-B (HH, VV, HH, HV, VH)	
ENVISAT (ASAR)		C-B (VV)	5 - 406
RADARSAT 1 (SAR)	8,25, 30, 35, 50, 100	C-B (HH)	
ERS 2 (AMI)		C-B (VV)	
ERS 1 (AMI)		C-B (VV)	

**Radar Satellites** 







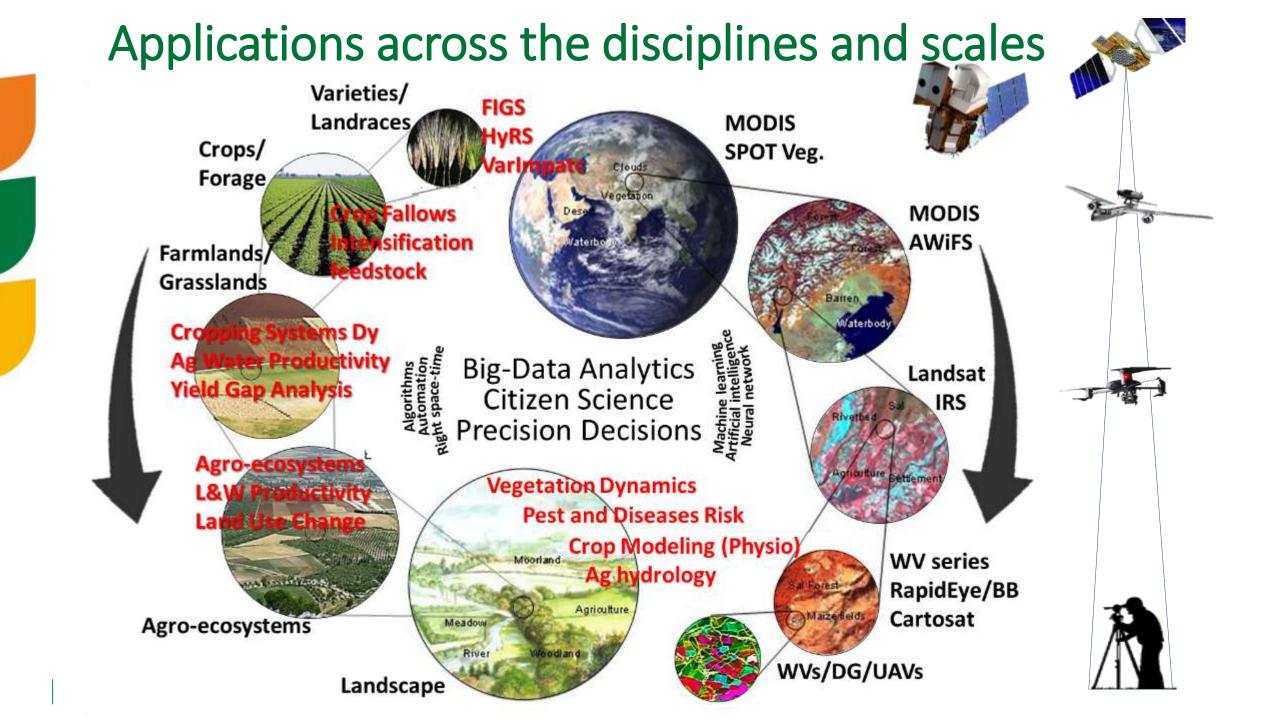


for Better Livelihoods in Dry Areas

### **EOS Matrix at Farmscape to Landscape**

### Biospectral – Biophysical

-	Platforms	Ground/	'in-situ	Airb	orne			Spaceborne		
The	Mode	Hyperspectral	Multispectral	Optical	Lidar		Optical		Lidar	SAR
	Sensor	ASD FieldSpec	M× Camera	APs/UAVs	Lidar	WorldView-2	Landsat	MODIS	ICESat*	PALSAR
RS data characteristics	Spectral	350-2500nm	4 bands	3-4 bands	1264nm	8 bands	7 bands	7/36 bands*	1264 & 532nm	Lband
RS data racterist	Spatial resolution	0.1-1.5m	0.1-0.2m	1-m	20 - 80cm	0.46m Pan;	15m Pan;	250m, 500m,	70m	10m, 20m,
RS						1.84m MS	30m MS	1000m MS		100m
cha	Swath	1-4m	2-10m		1-2km	16.4km	185km	2330km		35-250km
	Revisit			3-year		1.1 days	16 days	1 day	91 days	46 days
cal	Plant biomass	×	×		×	×	×	×		×
Biophysical	Plant height				×				×	×
do	LAI, fPAR, LST	×	×			×	×	×		
B	NDVI, EVI, LSWI	×	×	×		×	×	×		
	Erosion, Salinity	×	×	×	×	×	×	×		
cal	Soil moisture	×	×			×	×			×
emi	Chlolophyll	×	×	×		×	х	×	f Area Index 💋	
Biochemical	Nitrogen	×	×	×		×	×	Lea	NDVI	
Bi	Phosphorous	×	×			×			Les	
	Plant water	×	×					LeafPi	gments orophyll	
duc	GPP	×	×	×		×			lorophyll	
Produc tion	NPP	×				×	×	E	VI	
	land cover/use	×	×	×		×	×			×
LULC	phenology	×	×				х	×		×
	Irrigation	×	×	×		×	×	×		×
in	DEM		×	×	×	×			×	×
Terrain	Derivatives									
Ĕ			×	×	×				×	×
	Tier 1 AOIs	×	×	×	×	×	×	×	×	×
Scale	Tier 2 action sites	×	×	×			×	×	×	×
Sc	Tier 3 AEZs	×	×	×				×	×	×
	Tier 4 Target			×				×		×



## Satellite constellations for Agricultural Applications

Sensor	Wavelengths	Spatial Resolution	Revisit frequency	Temporal coverage
Landsat	Optical (6-9) + thermal	30,60m	16 day	1984-present
RapidEye	Optical (5)	5 m	5 day	2009-present
Skysat	Optical (4)	1m	~weekly	2013-present
WV-2/3	Optical (8)	0.3, 1m	5 day	2015-present
Planet Scope	Optical (4)	3 m	~daily	2014-present
Sentinel-1	C-band radar	10, 20m	6 day	2014-present
Sentinel-2	Optical (13)	10,20,60m	5 day	2015-present
Sentinel-3	Optical (13)	10,20,60m	3 day	2015-present

## Data fusion has huge potential!

## New era of analytics

Tabulating

Systems

Era

icarda.org

Programmable Systems Era

Conscious Systems







26

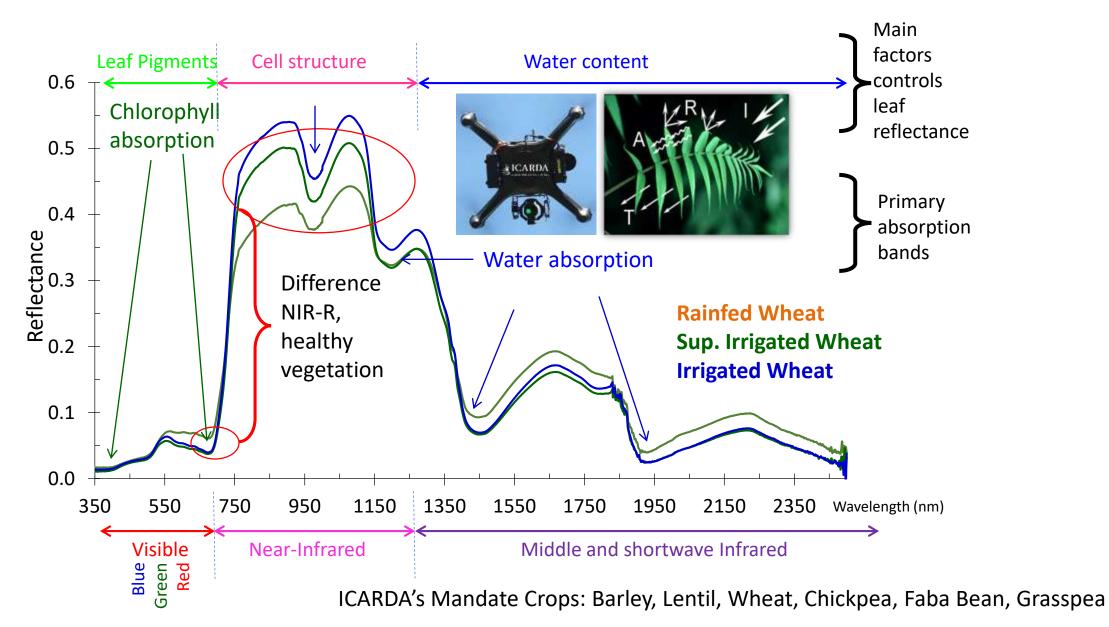
Cognitive

Systems

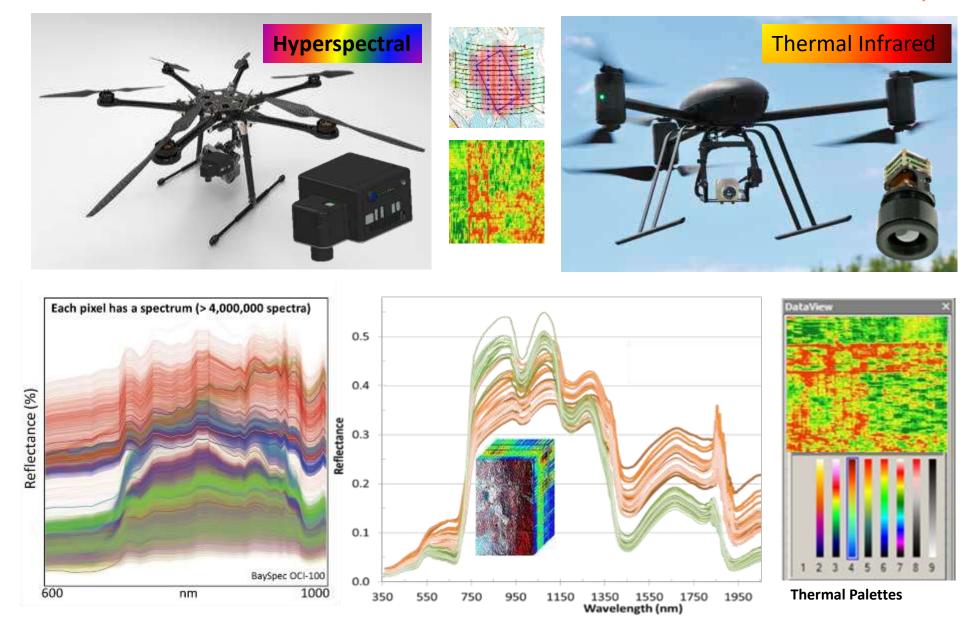
Era

# Reflectance-Based Characterization of Bio-Physical Traits

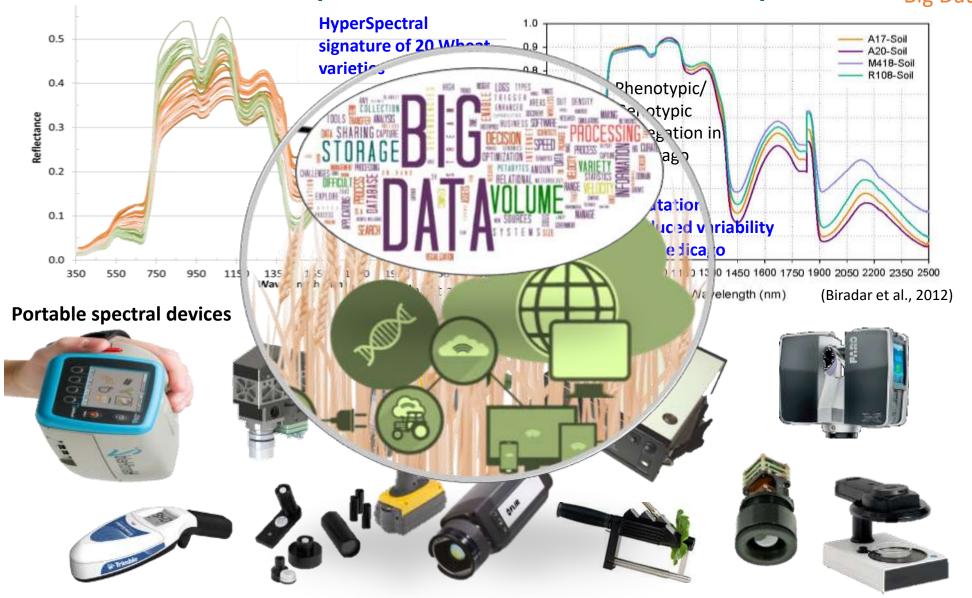
@ cell, leaf, canopy and landscape

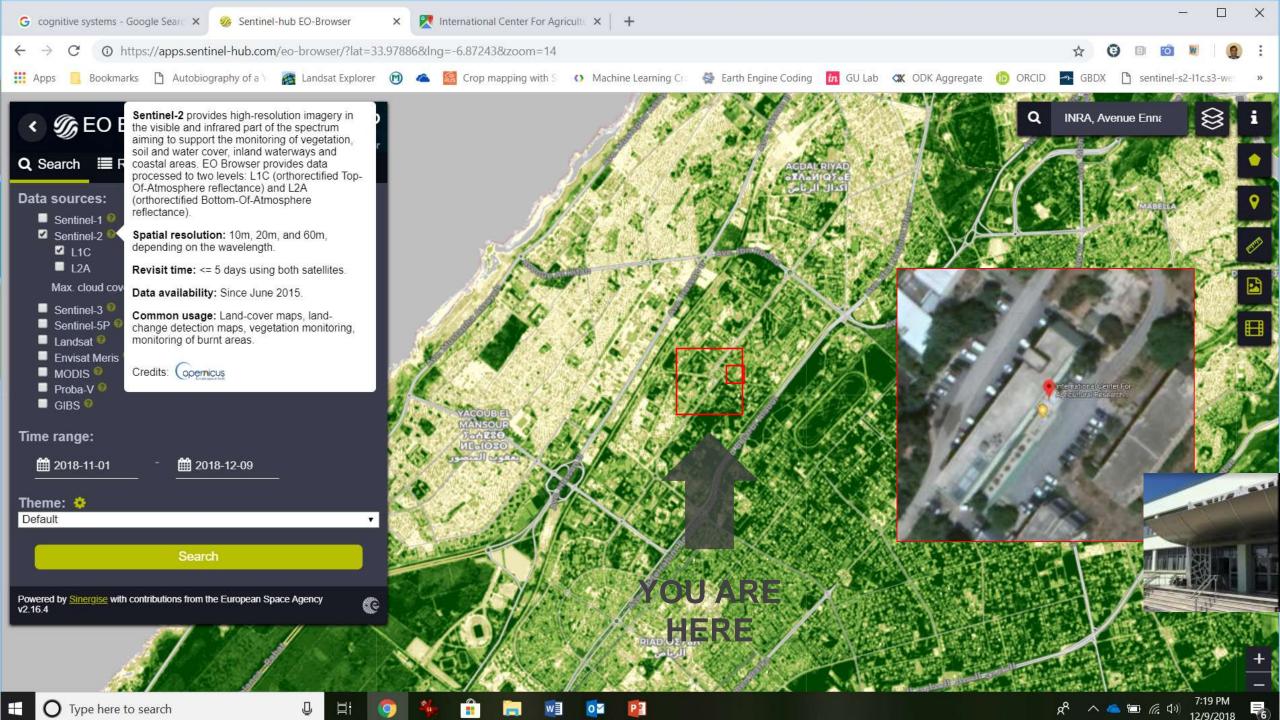


# Use of Portable Observation Systems in Research



## Advanced Sensors and Tools: Hyperspectral, Multispectral, Thermal, Ultraspatial





### Farming system dynamics at given time and scale

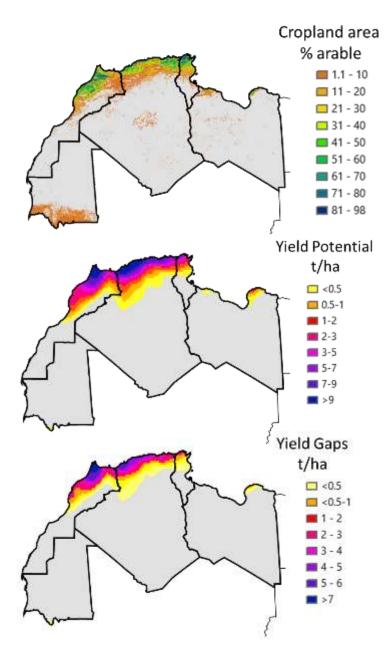
Jamdigri-mo	oved	Pakurseni l	LDTW	Hariharpur	•	Kalisara LD	w	Kundra - IV	PDW	Gosain Bun	dh SFMIS-mo
	amdigri	Pa	kurseni LDTW	1	Hariharpur	Kalis	ara LDTW	Kun	dra - IV PDW	Gosair	n Bundh SFMIS
FID	11	FID	115	FID	40	FID	24	FID	294	FID	71
BATCH	1	BATCH	2	BATCH	1	BATCH	1	BATCH	3	BATCH	2
SchemeName	e Jamdigri		e Pakurseni LDTW	SchemeName	Hariharpur	SchemeNam	ne Kalisara LDTW	SchemeNam	e Kundra - IV PDW	SchemeNam	Gosain Bundh
District	BANKURA	District	PASCHIM MIDNAPORE	District	PASCHIM MIDNAPORE	District	BIRBHUM	District	BIRBHUM	District	PURULIA
Block	JOYPUR	Block	NARAYANGARH	Block	SABANG	Block	MAYURESWAR I	Block	RAJNAGAR	Block	KASHIPUR
Scheme_Typ	MDTW	Scheme_Typ	TW	Scheme_Typ	Mini(E) RLI	Scheme_Typ	LDTW	Scheme_Typ	PDW	Scheme_Typ	
Village_Mo	Jamdigri	Village_Mo	Pakurseni	Village_Mo	Hariharpur	Village_Mo	Kalisara	Village_Mo	Kundra	Village_Mo	Uluberia
Lat	23.07006	Lat	22.19834	Lat	22.138147	Lat	24.05688	Lat	23.965694	Lat	23.477367
Long	87.47454	Long	87.44147	Long	87.630084	Long	87.84444	Long	87.356806	Long	86.790317
PhysicalPr	100	PhysicalPr	100	PhysicalPr	100	PhysicalPr	100	PhysicalPr	100	PhysicalPr	100
HODate	November 6, 2015	HODate	July 18, 2016	HODate	March 23, 2015	HODate	June 29, 2016	HODate	November 14, 2017	HODate	September 10, 2015
	N2VI — Find	1.00	NDVI — fitted	1.00	- NDVI - fitted	1.00			- NDVI fited	8	NIM fited
2016 2	Jan Jul Jan Jul 2017 2018 2018 select by system (me_shert)	0.75 0.50 0.25 0.00 <u>Jul</u> 2016 Image	Jan Jul Jan Jul 2017 2017 2018 2018 (labeled by system time_start)				0.75 0.50 0.25 0.27 0.00 0.25 0.00 2017 2018 2018 by system time_start)	3 J.M. Jan 2016 2017	Jul         Jan         Jul         01           2017         2018         2018         01           system.time_start)         01         01         01	4 2 0 	Aul Jan Jul 2017 2018 2018 system time_start)

### **Highly intensive systems**

Less intensive systems



## Mapping Yield Gaps



Grain yield at national average (FAO, 2015)

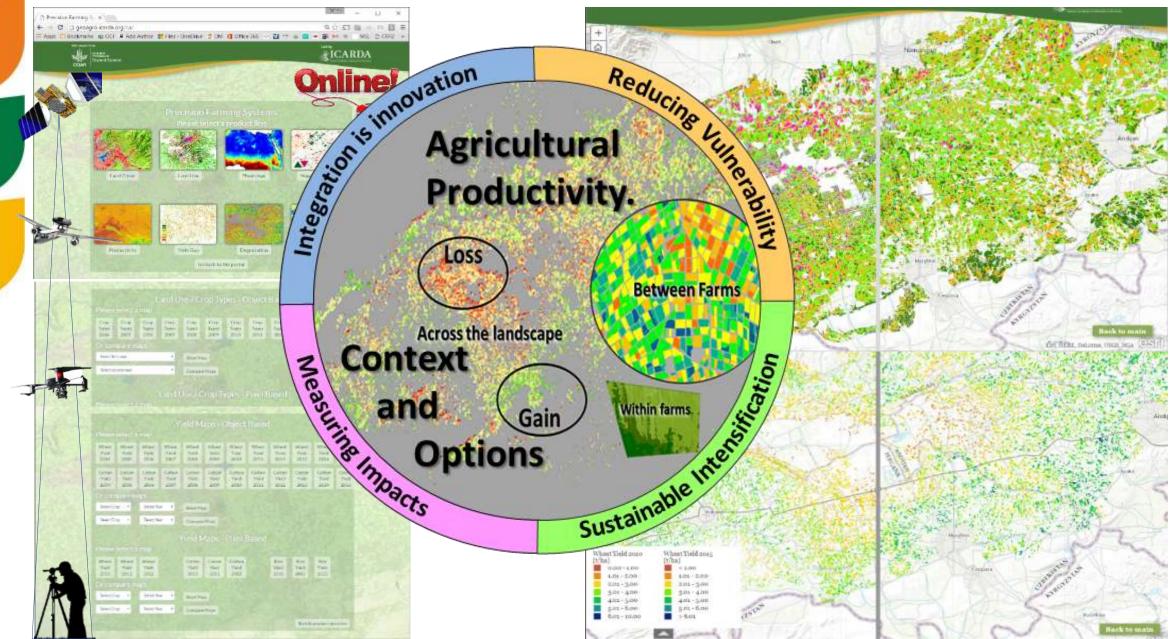
Country	2014	15yrs ave	15yr trend
Morocco	1713	1495	~~~~
Algeria	1475	1404	~~~
Tunisia	2149	1820	M
Libya	1250	871	~

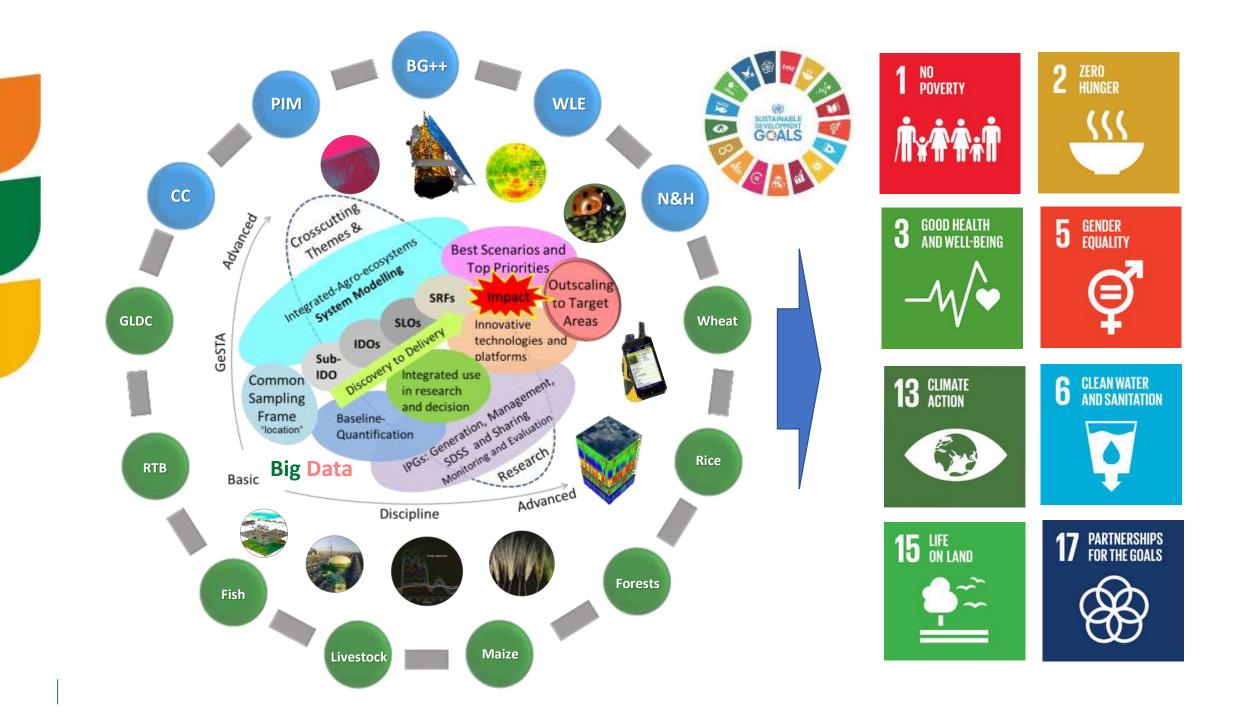
#### Actual grain yield measured at farm level

Farms	Longitude	Latitude	(kg/ha)	
Farm1	6.722111	33.620782	2,125	
Farm2	6.716233	33.612049	5,014	Range b/v
Farm3	6.704697	33.578011	2,039	< 1 to 5
Farm4	6.683507	33.564448	1,579	
Farm5	6.695998	33.552394	3,049	l t/h
Farm6	6.717472	33.569803	2,703	
Farm7	6.729286	33.573701	2,452	
Farm8	6.690977	33.573454	4,057	Median
Farm9	6.676971	33.599369	2,525	
Farm10	6.666595	33.604833	2,195	2 t/h.
Farm11	6.707888	33.595405	478	
Farm12	6.703041	33.586931	1,233	
Farm13	6.695853	33.572357	1,526	
Farm14	6.702122	33.562839	1,412	
Farm15	6.68507	33.585016	4,626	

### Quantification of Farming Systems @ multiple-scales

Smart Farming Platform







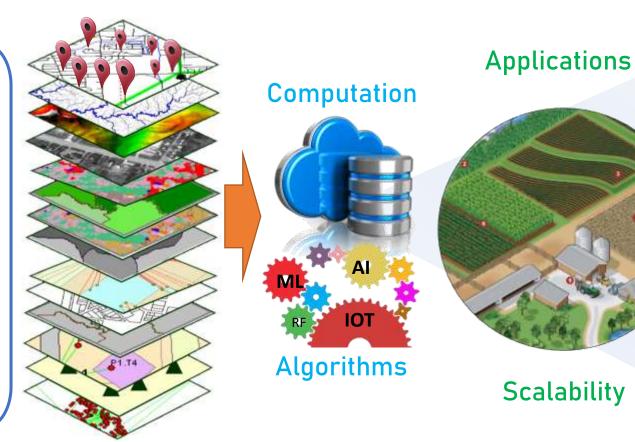
1

# Resilient Agroecosystems



### The Big Data

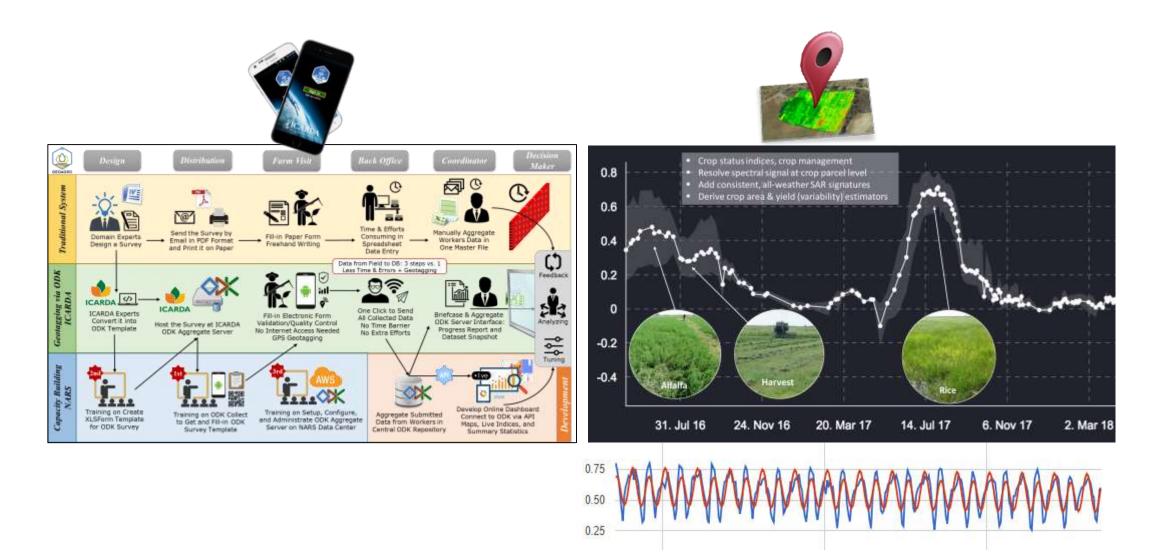
Geo-Tagging research & outfreach data Satellite data Crop data Climate data Soil data Water data Topography Demography Ecological data



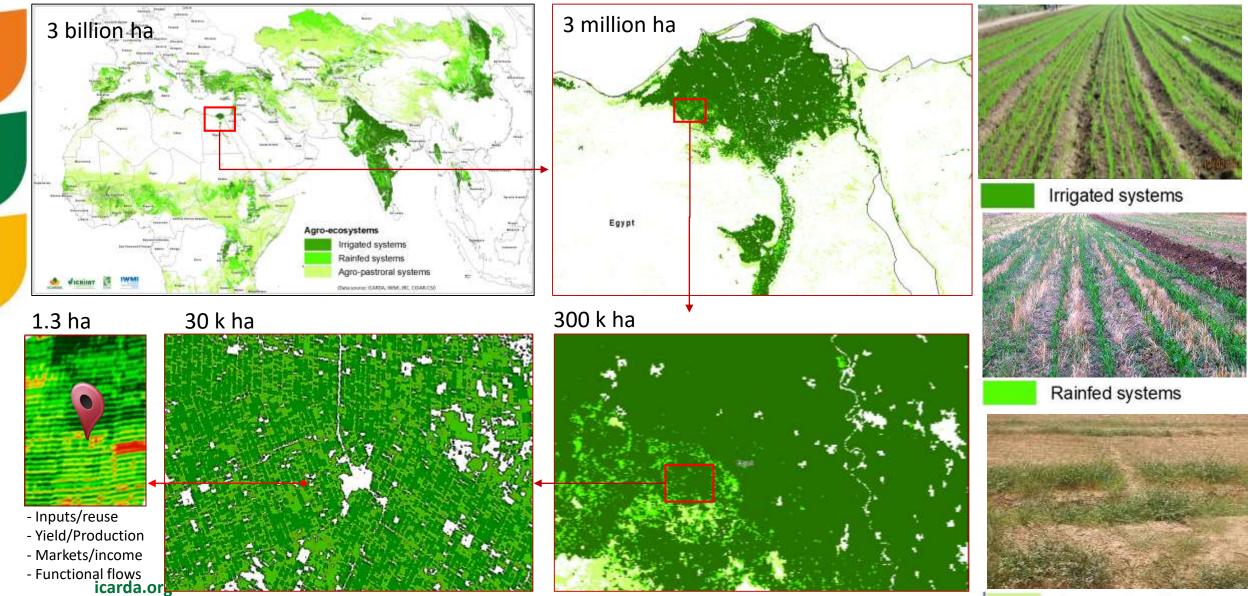
Mapping Monitoring Targeting Estimating Forecasting Warning Lending Insurance Value chains

### **Biggest drivers**

## Geotagging and AgroTagging



#### Agroecosystems: scaling innovations and measuring impacts



Agro-pastroral systems

#### Easily detectable (RS) evidences of potential yield gaps at farm level





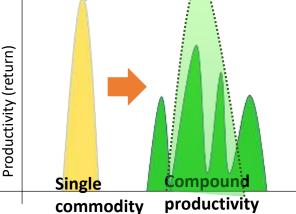


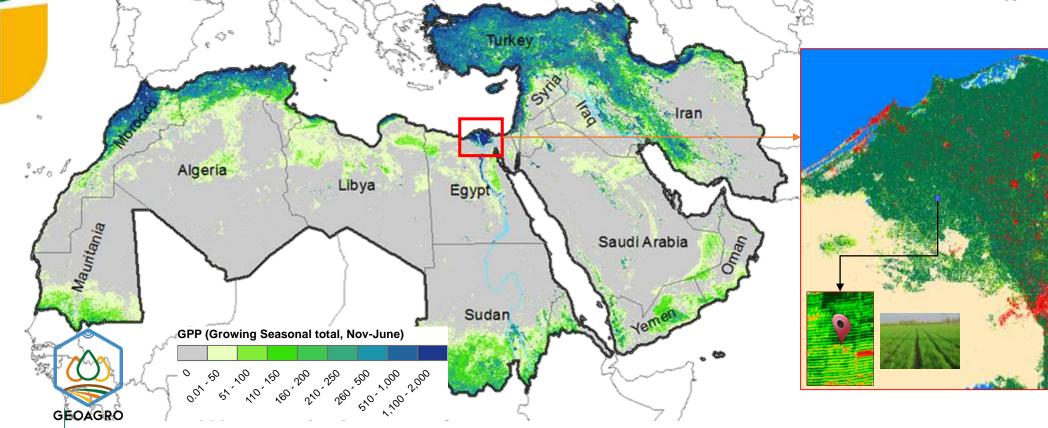


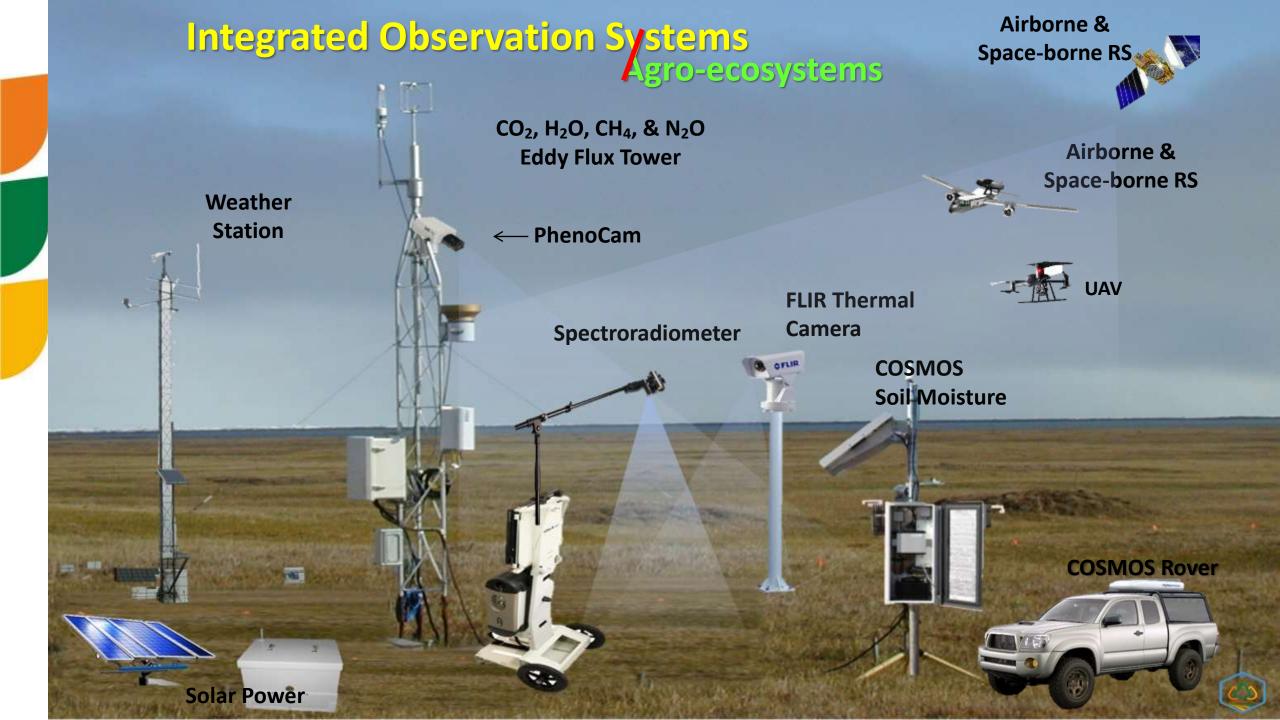


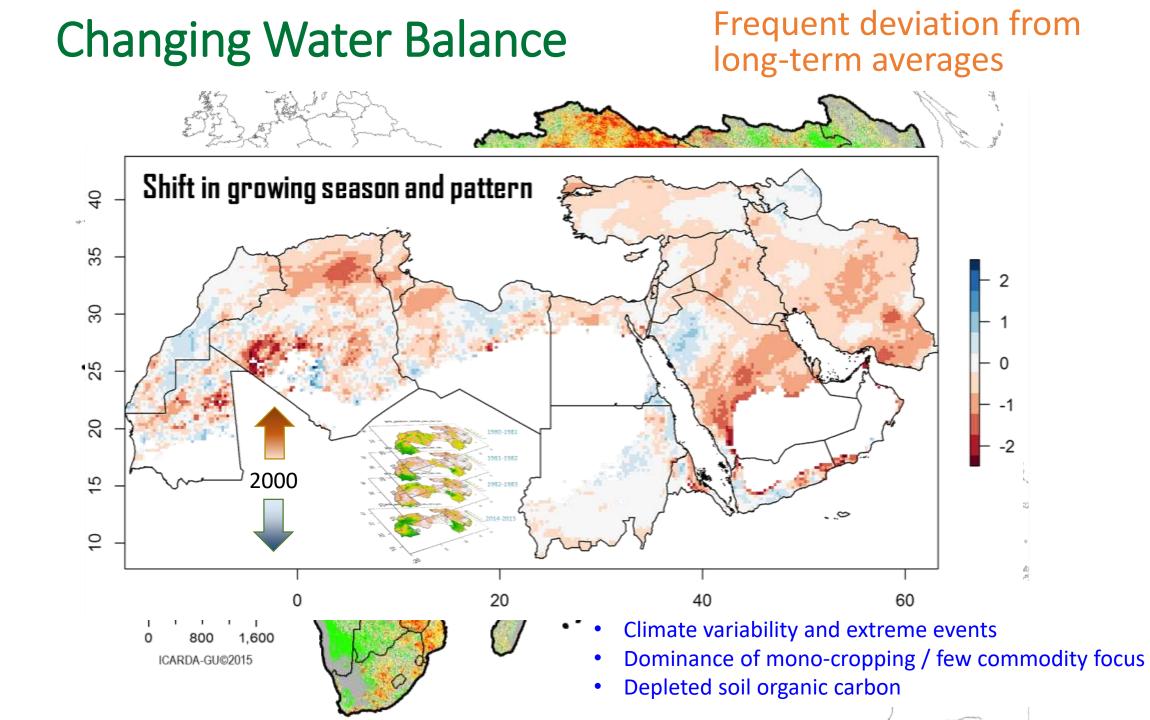
# Shift in paradigm for sustainable agri-food systems

Disaggregating yield gaps at farm/pixels scales to target appropriate interventions Shift in the paradigm from commodity to functional productivity









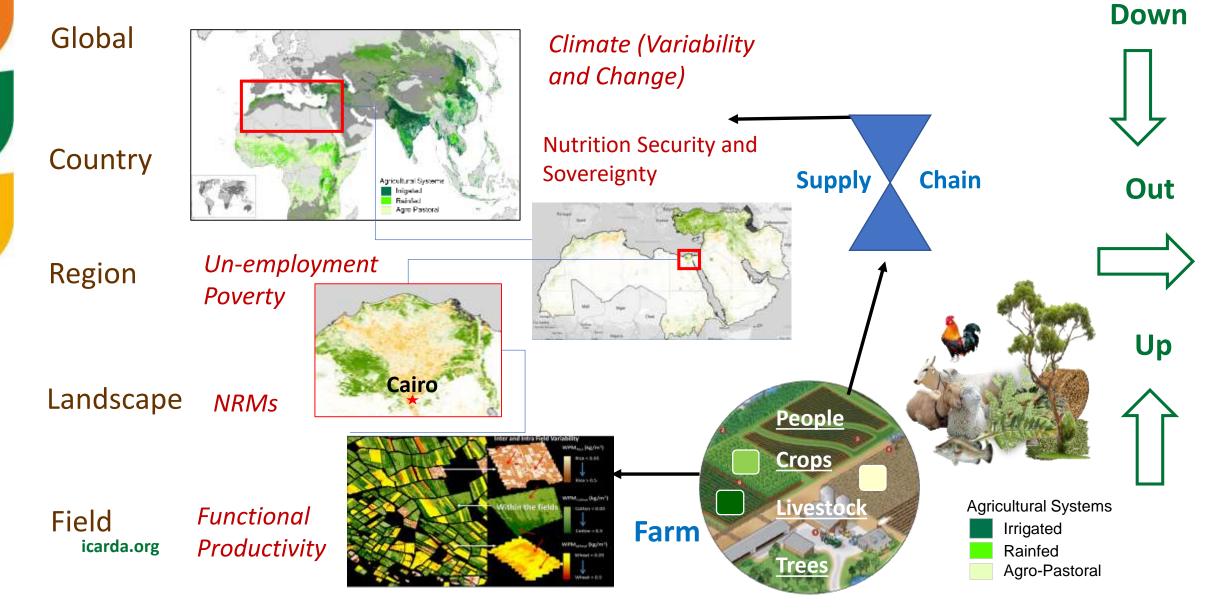
#### Copying with water demand **Blockchain for ecological intensification** Build resilient Incentivizing best agricultural practices 1. Crop growth agroecosystems through blockchain based 2. Yield & Rotation WATER/CARBON CREDITS associated with improved agroecosystem health by **Financial inclusion** 3. Water productivity adoption of the water saving agri-food systems (integrated agroecosystems) **Better livelihoods** Incentives 4. 3 days revisit Sustainable intensification 30m 1 **Right crop at right place and time** 10m 1.0m **Resilient cropping systems** Pixel/Farm/Parcel 0.3m better integration of crops, livestock, fish, trees & people A single entity for each & every developmental entry point **Optimizing agricultural water use by integrated approach** compound intensification in cereal based systems

Open source

Agreements

<Biggest drivers

### Integrated Agroecosystems combining Component Research & Systems Research A multi-scale and multi-criteria R4D



# Machine Learning and AI in GeoAgro analytics





Thousands of research and outreach data points in each season across the agro-ecosystems

Open source near-real time earth observation data at field, farm to landscape scales

Enormous power of cloud computing, open access, algorithms, analytics to process data on time

Smart phone enabled apps and cloud web-GIS for decision making at point, farm and administrative units





### **Technological Innovations**

(integrated systems approach for reducing risks)

# 2 3 1 4

#### Biodiversity & Crop Improvement Program

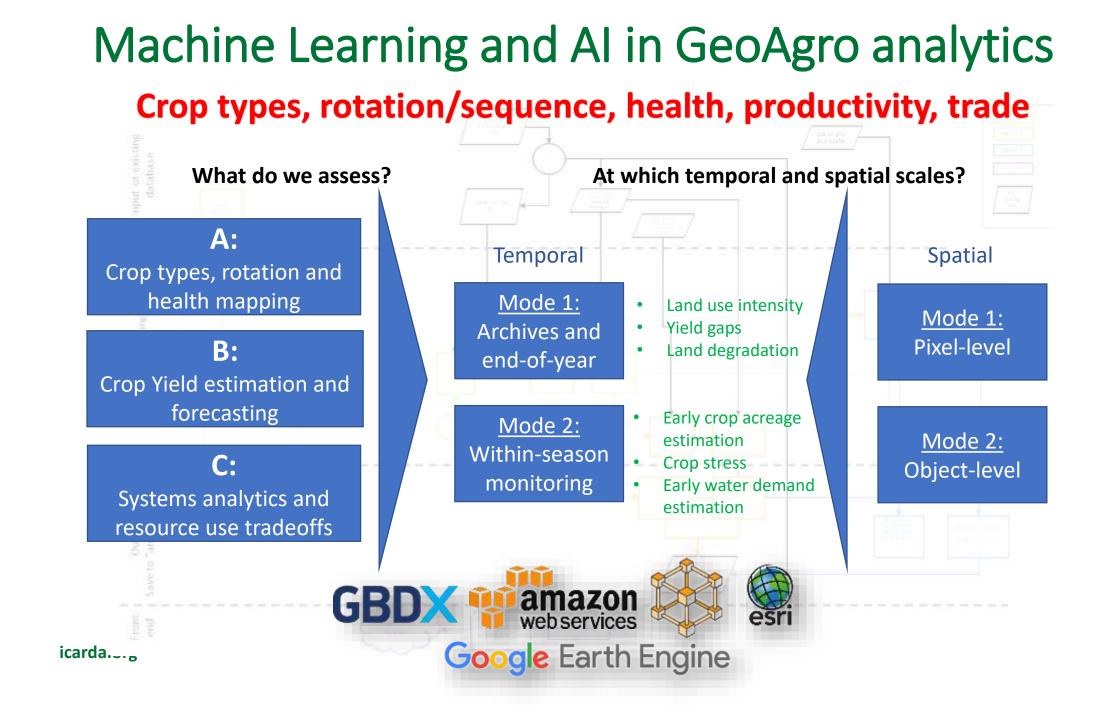
>> research plots, > farm
trials / demonstrations, >
international nurseries >
germplasms, > NARS
partners feedbacks, > etc.

Resilient AgriculturalWater, Land Management &Cross Cutting ThemesLivelihood Systems ProgramEcosystems ProgramBig Data and ICTs

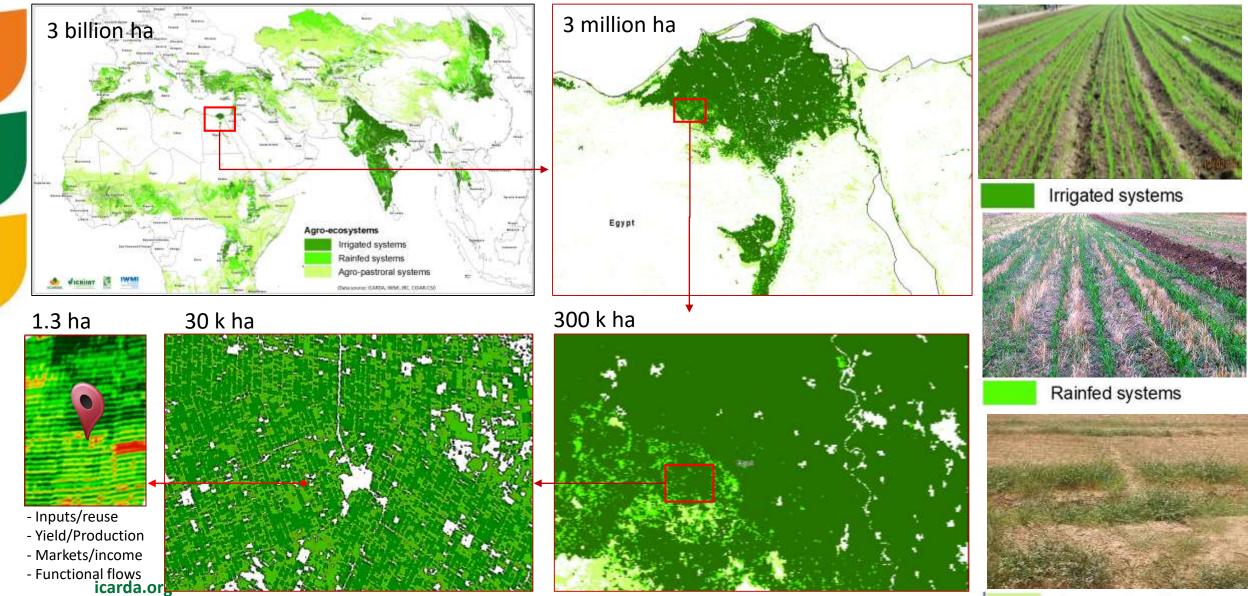
>> research plots, >
agronomy, >CA/Zero tillage>
livestock, > rangelands, >
household surveys, > value
chains, > etc.

>> field data, > raised
beds> Field ETs, > AWPs,
> soils, > hydrology, > land
degradation,> erosion>,
hydrology, > etc.

>> big-data, > open access
resources, > cloud
computing, > gender data, >
> scaling > capacity dev., >
modelling, > etc.



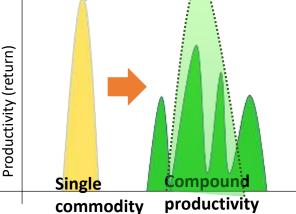
#### Agroecosystems: scaling innovations and measuring impacts

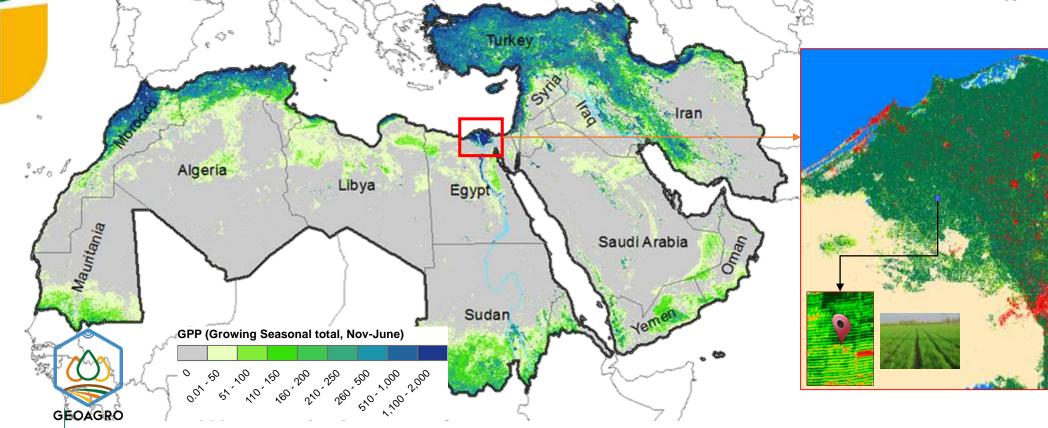


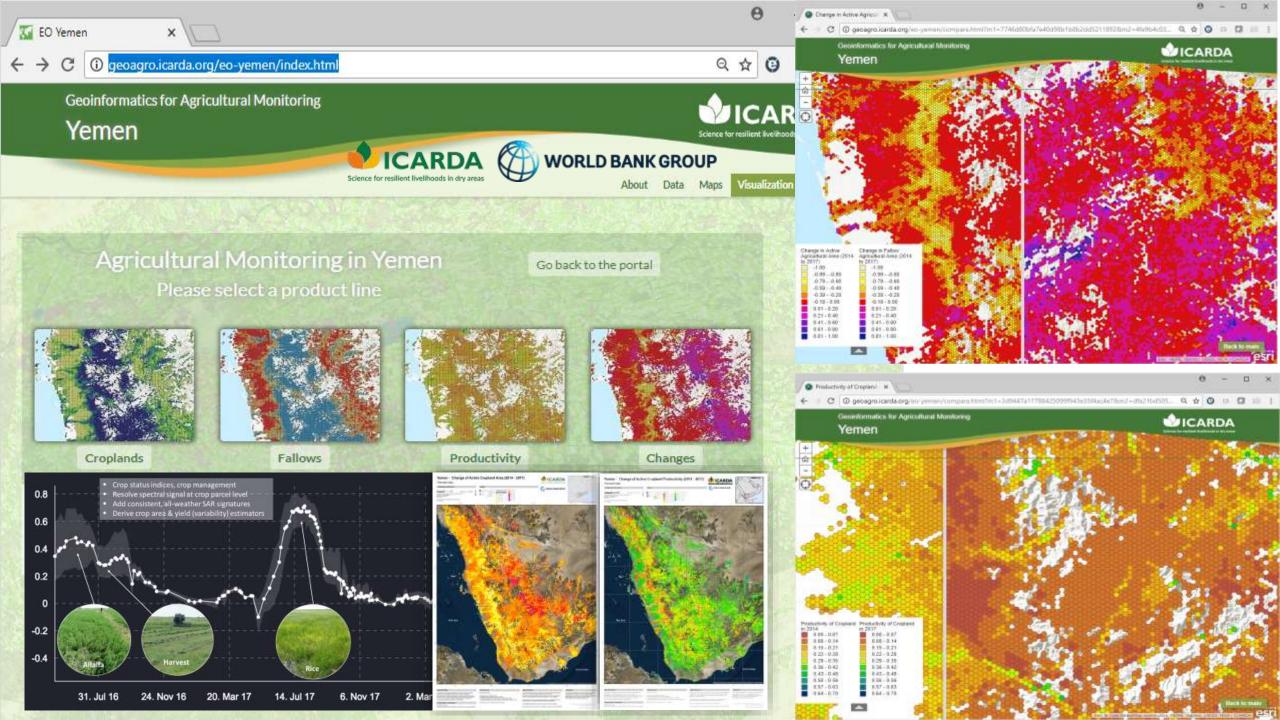
Agro-pastroral systems

# Shift in paradigm for sustainable agri-food systems

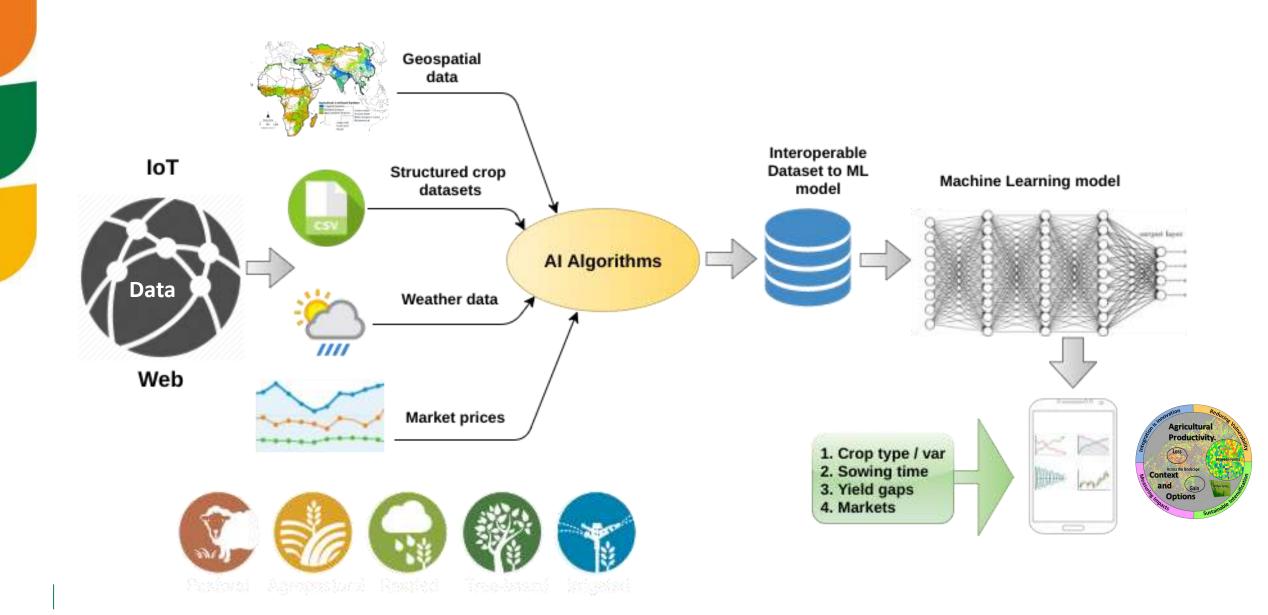
Disaggregating yield gaps at farm/pixels scales to target appropriate interventions Shift in the paradigm from commodity to functional productivity







# Big-data, Machine Learning and AI algorithms



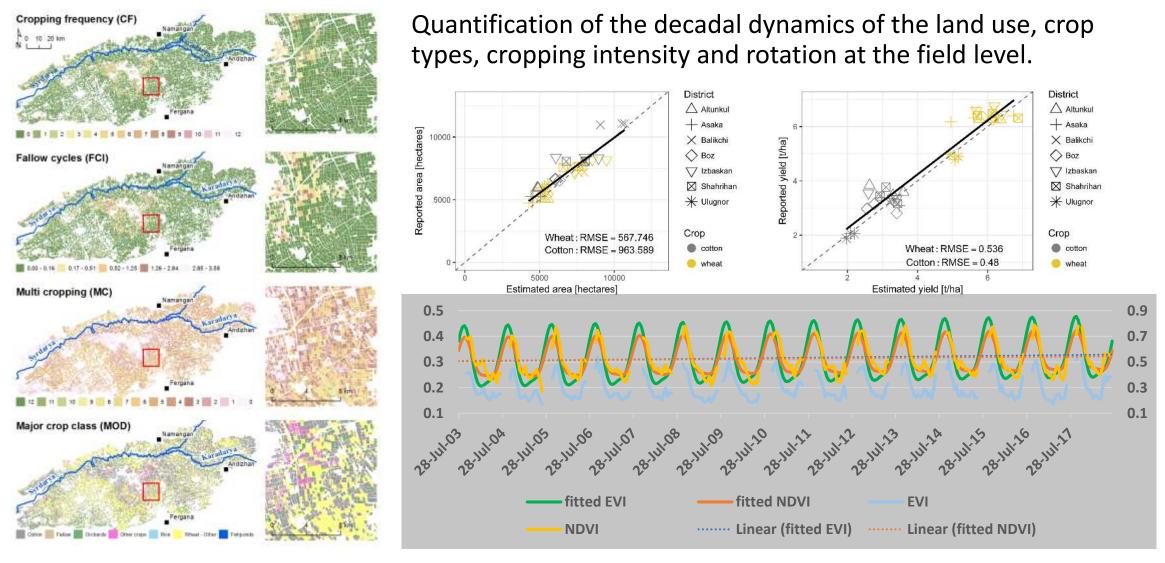
#### Big-data, Machine Learning and AI algorithms AI-ML-BigData @ genetics, chemistry, weather, agronomies, trade... Big A **Meta Analytics** Data ML People Crops deep learning machine learning Livesto predictive analytics translation natural language classification & clustering processing (NLP) information extraction Demand driven contexts Inclusive Sustainable options options Agroecosystems location

expert systems

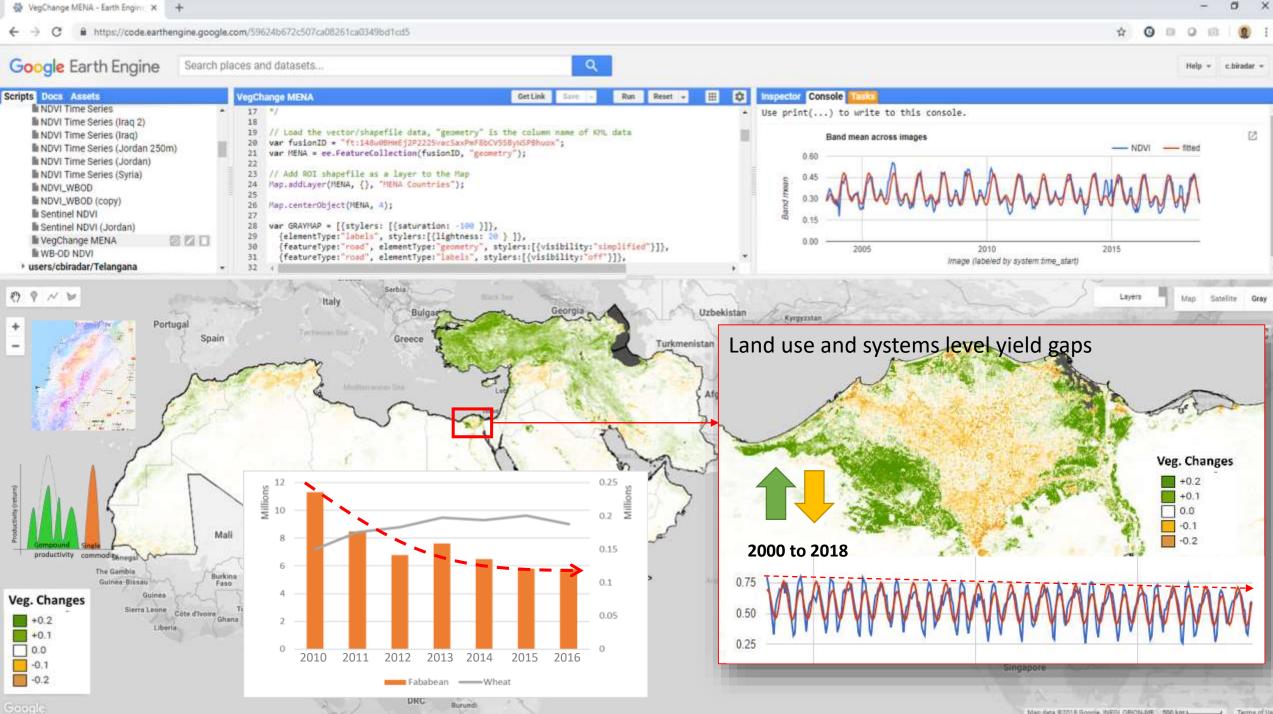
planning, scheduling & optimization robotics image recognition vision machine vision

typology

# Multi-year mapping of the crops, intensity and rotation

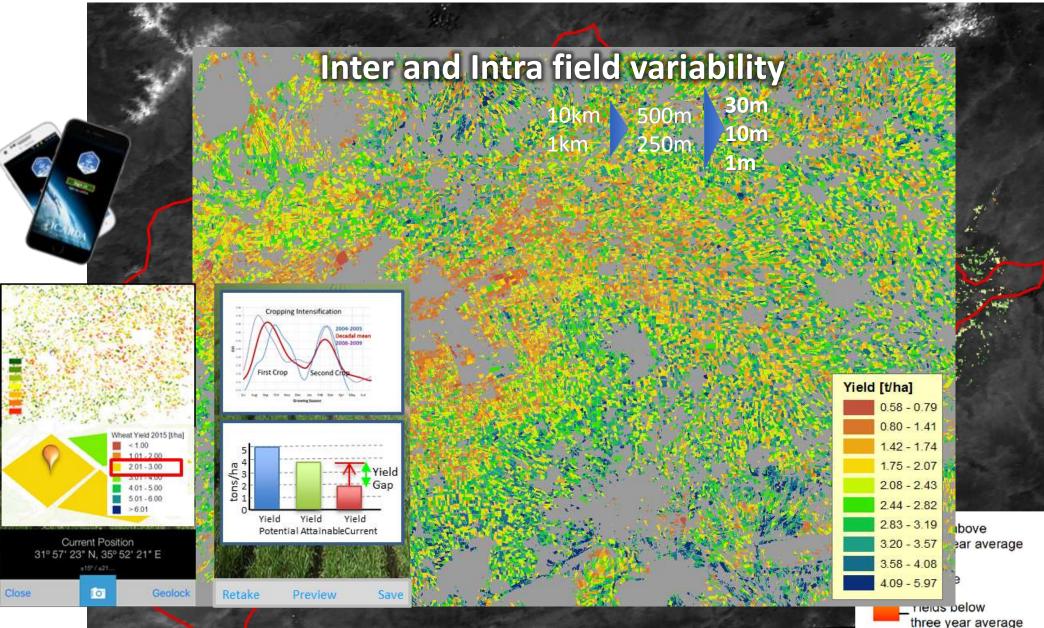


icarda.org

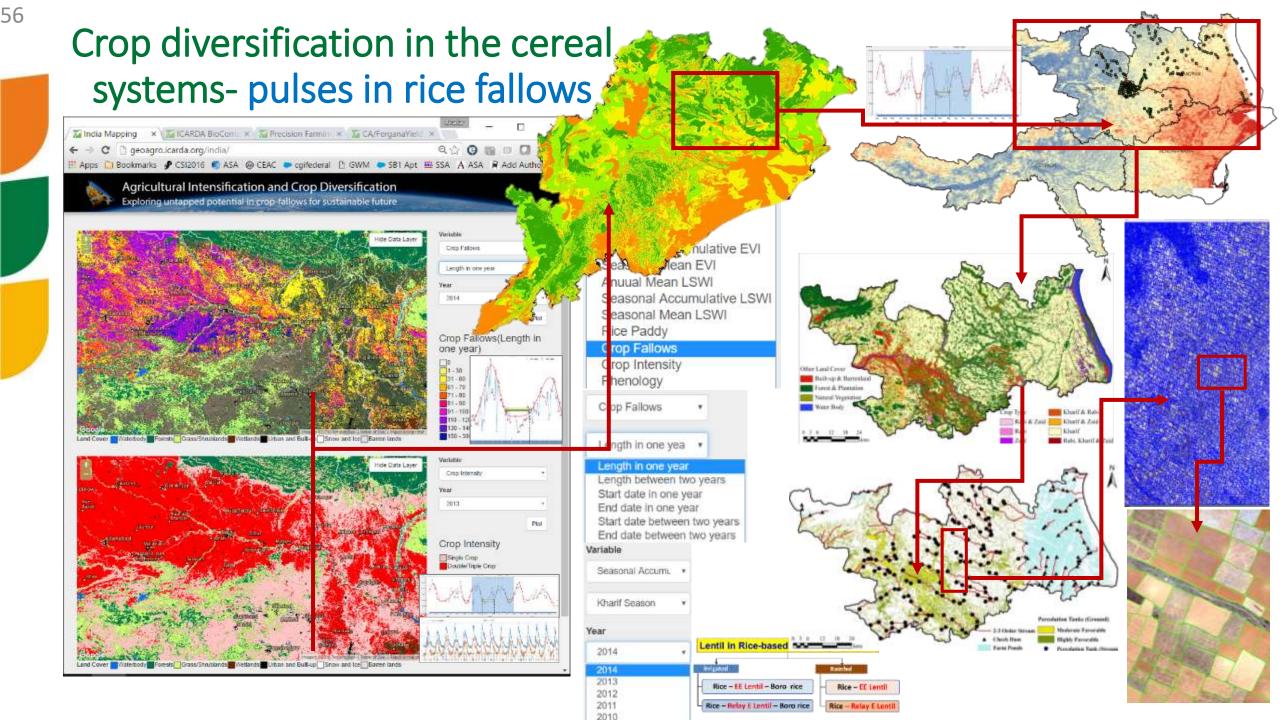


Map data #2018 Google, INEGI, ORION-ME11:000 kmt.L.

#### Productivity dynamics, pattern and yield gaps



System efficiency at farm/field/pixel level



### **Smart Farming Systems Platform**

NN

Cadastral, Object

& Pixel based

Biophysical and socio-ecological

**Machine Learning** 

Crop types, crop intensity, rotation, fallows, crop stress,

AET-I8, soil moisture-

SMAP Citizen-Science

> Cellphone feedback

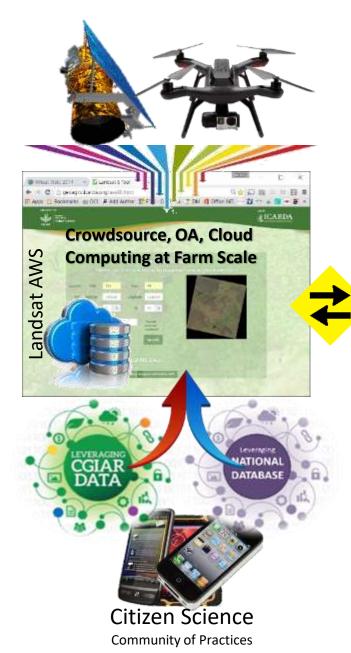
Direct Access and Markets/Business

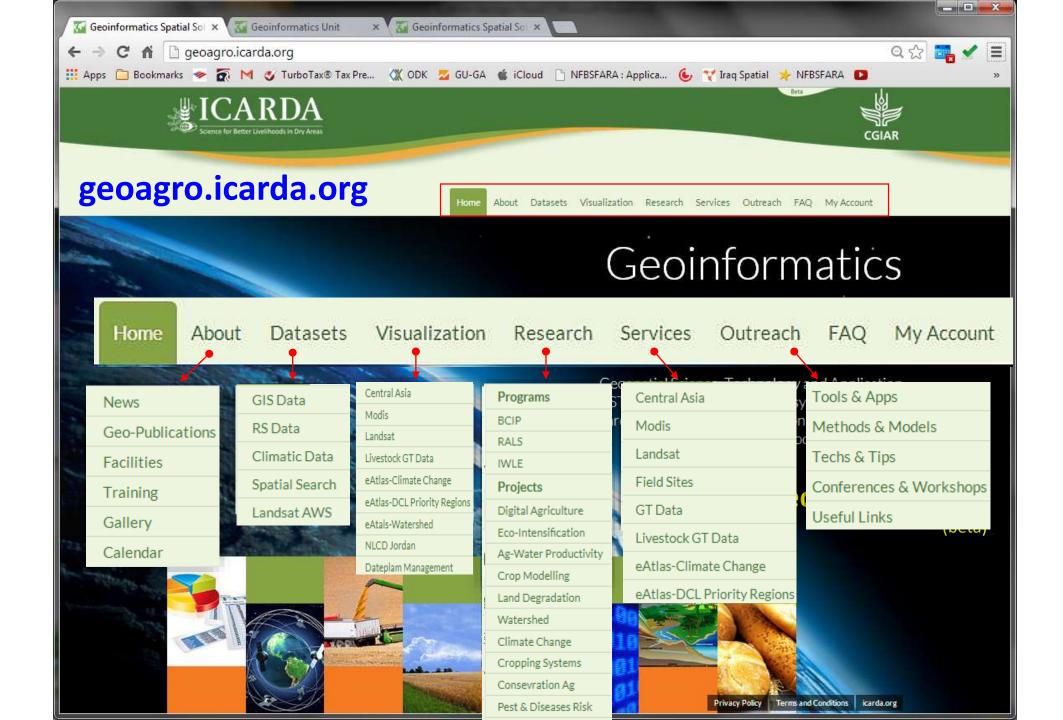
**Precision-Decision** 

# Digital Augmentation for accelerating sustainable intensification











#### **Production follows functions**

Building functional feedback system through integration of crops, trees and animals



-IPCC Confronting Climate Change:



# **Thank You**

#### c.biradar@cgiar.org

Chandrashekhar Biradar, PhD Head-Geoinformatics Unit Principal Scientist (Agro-Ecosystems)