Next Frontier Technologies: How to Scale SDG Monitoring

Dean Angelides Corporate Director, Esri

# Vision

## GIS

Is Enabling a Sustainable World

## Our World Needs a Nervous System

An Intelligent and Responsive Platform

Creating More Understanding ....Collaboration and ....Action

. . Geography Is Essential

#### Your Work Is Already Creating Geospatial Infrastructure

Intelligent and Responsive . . .

ŏ

Integrating All Sources of Data . . .



Creating Digital Nervous Systems for Your Organizations and Countries

Connecting Everything

Applying The Science of Where . . .



#### Creating a Global Nervous System

A System for Understanding . . . And Collaborative Action

## **Geospatial Infrastructure** Supports Individuals . . . Distributed ... And Organizations of All Sizes . . . And Interconnected Communities Engaging Organizations Everyone **GIS at Scale** Departments Teams Individuals Sharing and Collaboration

## Integrating Open Science, AI and Machine Learning

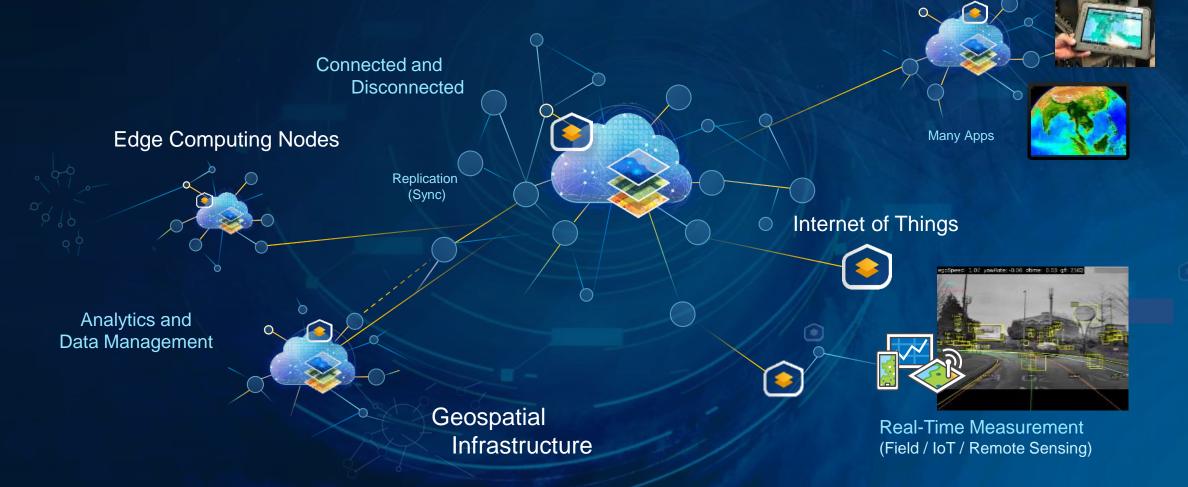
**Revolutionizing Spatial Analysis and Data Science** 



## Extending GIS to the Edge

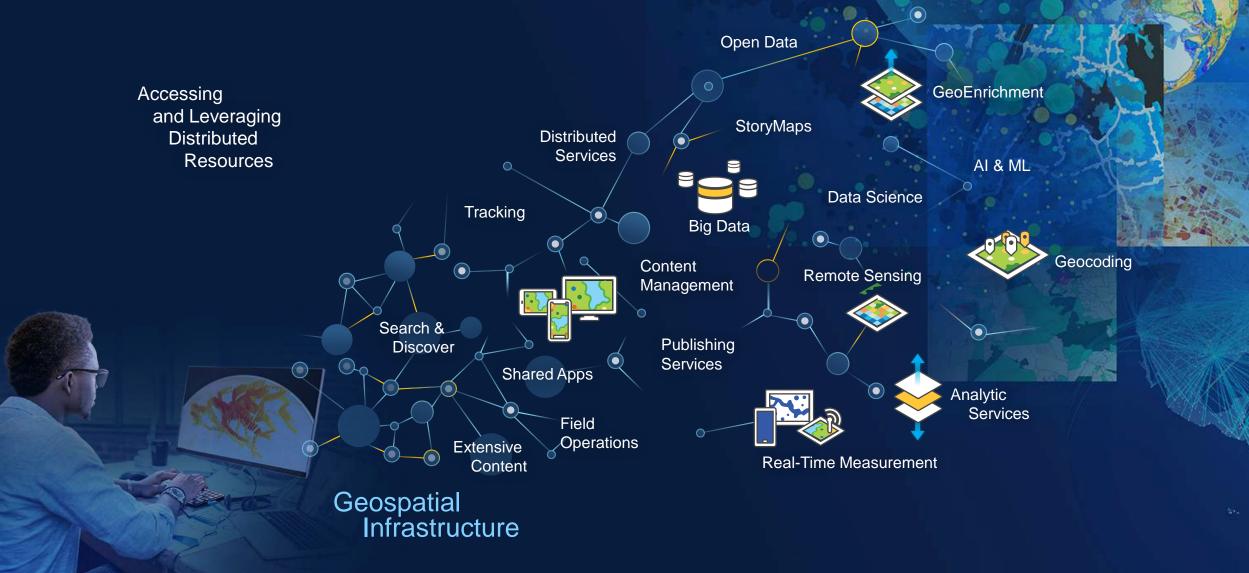
Including and Integrating . . . Challenged Network Environments

Edge Devices



Supporting GIS Workflows in All Environments

## Supporting and Integrating Advanced Technologies



## **3D** Visualization

#### New and Improved

- 3D Smart Mapping
- Mobile
- Point Clouds
- BIM Support
- Symbology



Data Visualization



Symbology



Effects



Power Lines



Extrusion

#### Augmented Reality / VR



Mobile

#### BIM as Scene Layers













Mobile Scene Packages

**BIM Integration** 



City Modeling





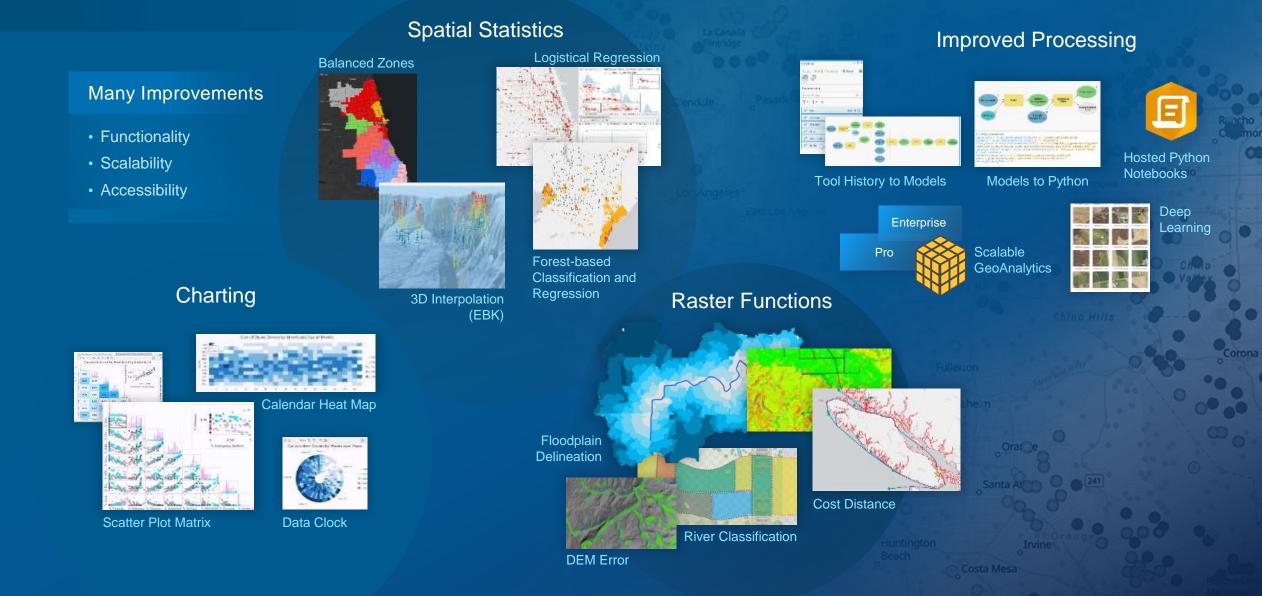
Underground



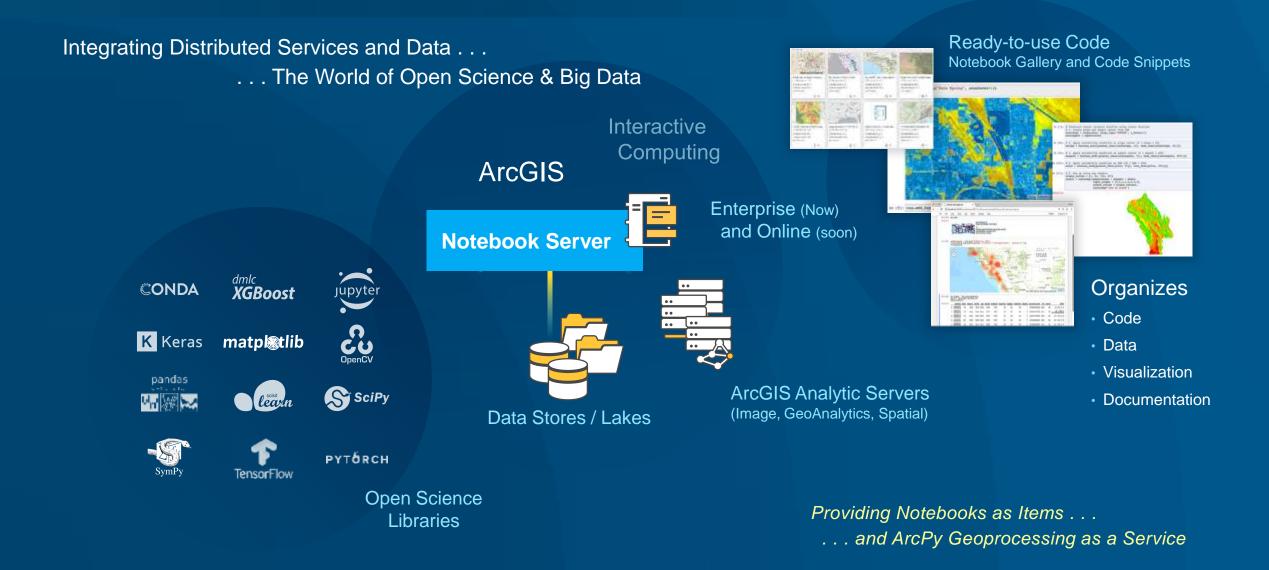
## Field Operations Location-Enabling All Aspects of Field Work



## Spatial Analysis and Data Science



## Hosted Python Notebooks For Integration, Modeling and Automation

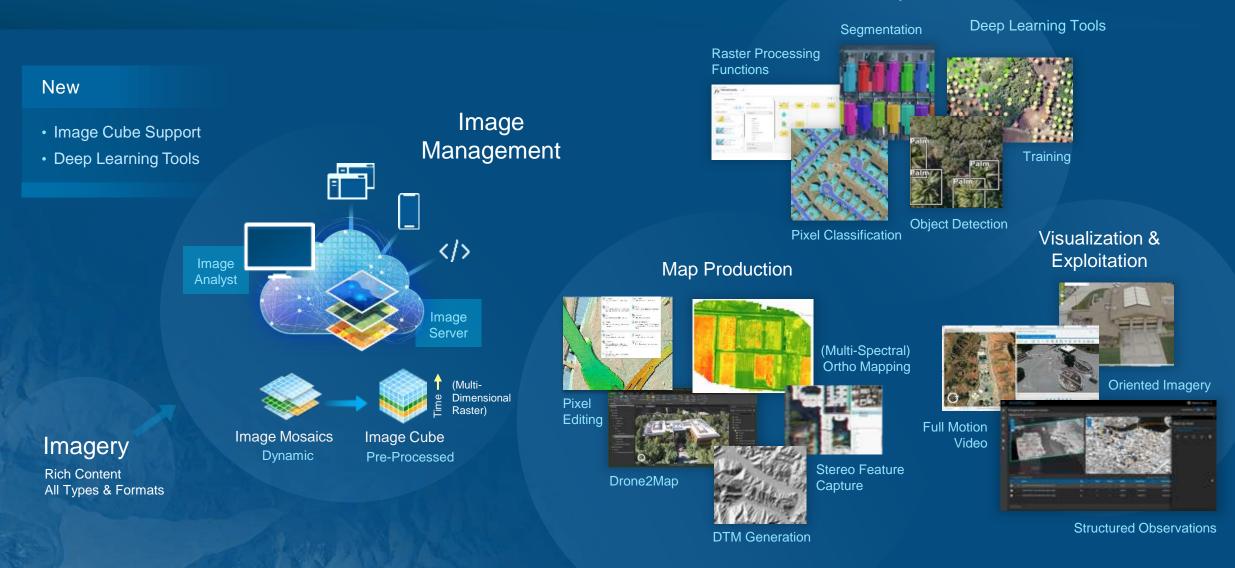


## AI, ML and Deep Learning Integrating Open Science



### **Magery** A Comprehensive System for Imagery and Remote Sensing





## Real-Time Analytics Integrating Sensor Networks and IoT

Supporting High-Velocity Data Streams Tracking, Monitoring and Alerting

#### Improved

- Performance
- Scalability
- Resiliency
- Cloud Connectors
- Actuation

Sensors

Vehicles



Real-Time Environment Data





Enterprise Now . . . . . . SaaS Coming

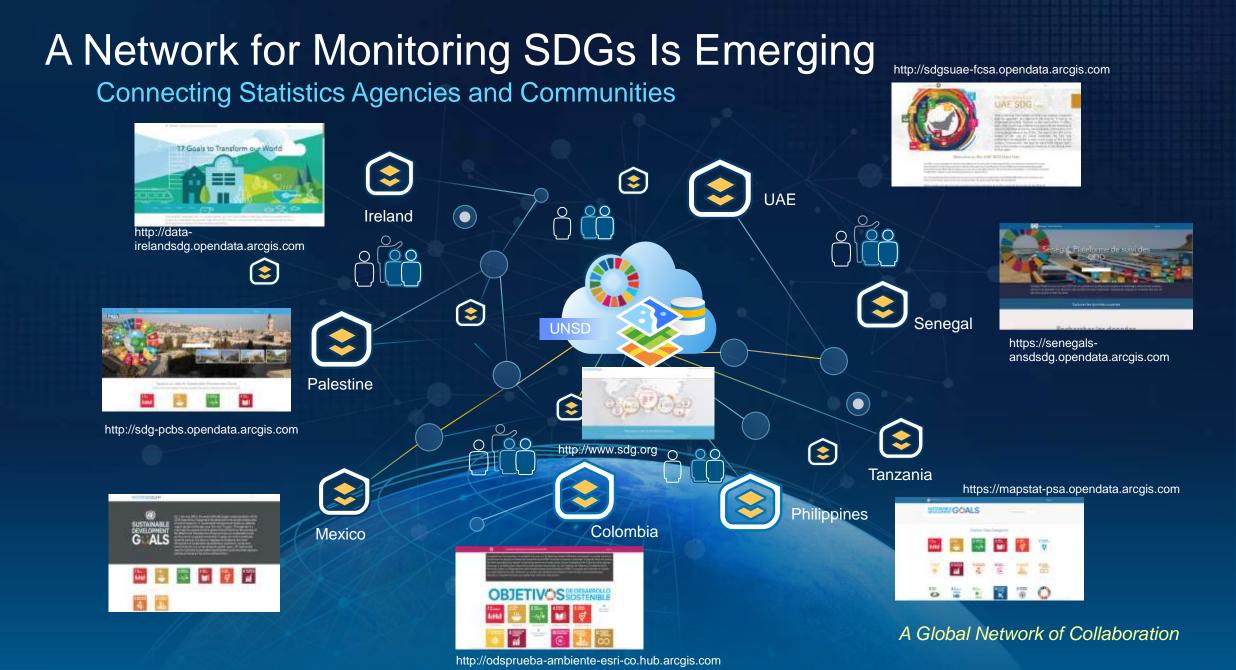
Collapsing the Time from Measurement to Decision Making

## Engaging and Interconnecting Communities

Bringing Together People, Organizations and Stakeholders



. . Collaborating Around Common Interests and Initiatives



## Geospatial Infrastructure provides SDG monitoring at scale

Enabling federated data management, and sharing best practices and advanced analytics

#### **UNSD Federated Information System UN Big Data Global Working Group** for the SDGs Palestine Philippines Mobile Phone Data Scanner Analytic Services Data Palestine Central Bureau **Philippine Statistics** </> </> </> </> </> </> </> </> </> of Statistics Authority Mexico **Big Data** Integration GWG Partner UN SDG Python **Open Science Tools** Hub Network Notebooks Instituto Nacional de Estadística y Geografía Ireland (INEGI) Social Media Big Data Data Spatial Analysis & UAE, Geoprocessing Central Statistics Office & Partner **Ordnance Survey Ireland** Publish ... Share Create **Federal Competitiveness &** A Global Network of Collaboration **Statistics Authority**

THE GLOBAL GOALS For Sustainable Development

<b>Target</b> Contribute to progress on the Target, not necessarily the Indicator										Goal	Indicator Direct measure or indirect support to the Indicator					
							1.4	1.5	1	No poverty	1.4.2					
						2.3	2.4	2.c	2	Zero hunger	2.4.1					
					3.3	3.4	3.9	3.d	3	Good health and well-being	3.9.1					
									4	Quality education						
								5.a	5	Gender equality	5.a.1					
		6.1	6.3	6.4	6.5	6.6	6.a	6.b	6	Clean water and sanitation	6.3.1	6.3.2	6.4.2	6.5,1	6.6.1	
					7.2	7.3	7.a	7.b		Affordable and clean energy	7.1.1					
								8.4	8	Decent work and economic growth						
					9.1	9.4	9.5	9.a	9	Industry, innovation and infrastructure	9.1.1	9.4.1				
						10.6	10.7	10.a	10	Reduced inequalities						
	11.1	11.3	11.4	11.5	11.6	11.7	11.b	11.c	11	Sustainable cities and communities	11.1.1	11.2.1	11.3.1	11.6.2	11.7.1	
				12.2	12.4	12.8	12.a	12.b	12	Responsible consumption and production	12.a.1					
					13.1	13.2	13.3	13.b	13	Climate action	13.1.1					
		14.1	14.2	14.3	14.4	14.6	14.7	14.a	14	Life below water	14.3.1	14.4.1	14.5.1			
	15.1	15.2	15.3	15.4	15.5	15.7	15.8	15.9	15	Life on land	15.1.1	15.2.1	15.3.1	15.4.1	15.4.2	
								16.8	16	Peace, justice and strong institutions						
17.2	17.3	17.6	17.7	17.8	17.9	17.16	17.17	17.18	17	Partnerships for the goals	17.6.1	17.18.1				

EARTH OBSERVATION AND GEOSPATIAL INFORMATION LINKAGES TO SDG GOALS, TARGETS AND INDICATORS

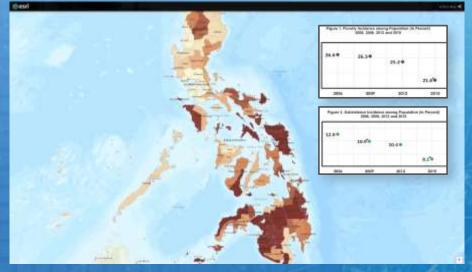


#### Population Below Poverty Line



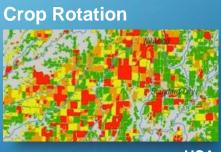
Ireland

#### **Poverty Incidence**



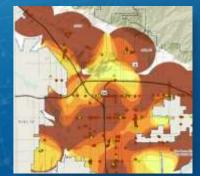
Philippines





USA

#### **Healthy Food Access**



California

#### Food Supply



**UN-Yemen** 

#### Malnutrition



World

Precision Agriculture



**New Zealand** 



### Machine Learning using Drone Data

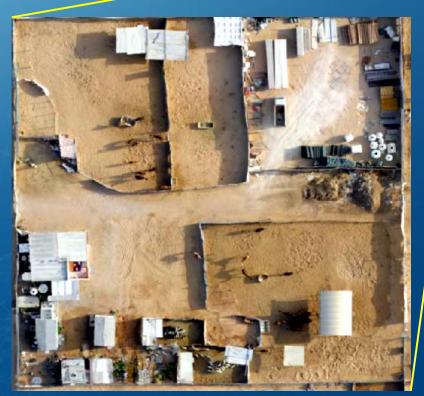
#### Captured images for two study areas

- Animal Farms
- Crop Farms

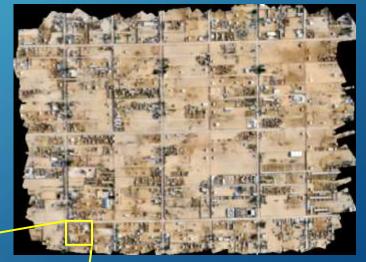
#### **Use Esri Artificial Intelligence tools**

- Multi-spectral image analysis
- Auto-detect features

## Focus on-site inspections to farms that have regulatory issues







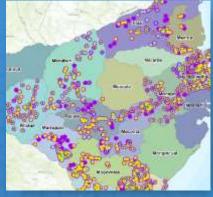
**Animal Farms** 





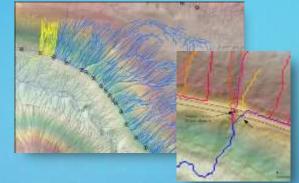


Water and Sanitation Projects



Mozambique

#### Drainage Network Modeling



Sewer

Sanitation Cleanout

Locations

California

Water Monitoring

Los Angeles



Switzerland

#### Water Quality Monitoring



China

#### Work Order Management

Washington



**New Jersey** 

#### **Pipeline Alignment**



Montana



#### **Rooftop Solar Potential**



**Massachusetts** 

Solar and Wind Energy



Philippines

**Solar Potential** 

Renewable Energy Connection Network

**Southern California** 

#### Wind Farm Design



Bavaria, Germany

Wind Resources

Singapore



England

#### **Renewable Energy Monitoring**



China



## Electricity Consumption per Capita:

Uganda (2016): 71 kWh/Capita
Germany (2014): 7,035 kWh/Capita
EU (2014): 5,909 kWh/Capita
World (2014): 3,128 kWh/Capita



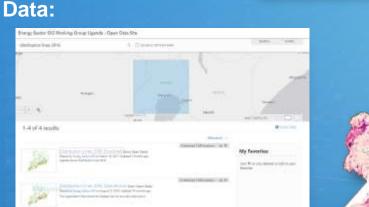




#### Goal:



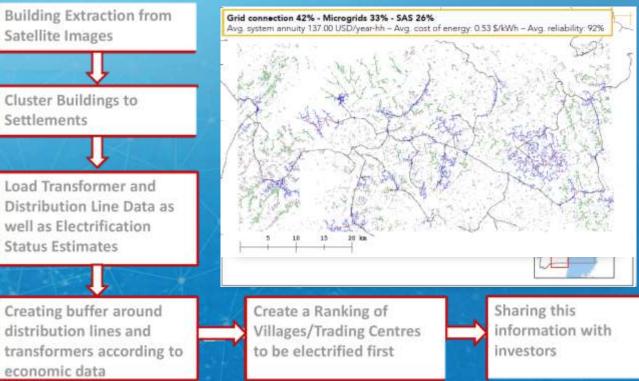
Developing a GIS based decision support Model to decide whether it is more economical to electrify a village using Solar Home Systems, Mini-Grid or On-Grid Solutions



Energy Sector GIS Working Group Uganda Open Data Site

# 

Facebook Population Data https://ciesin.columbia.edu/data/hrsl/

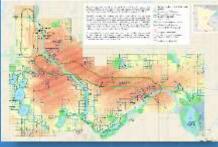


#### Electrification Planning in Uganda using Satellite Data

Solar Containers for rural communities



#### **Urban Heat Islands**



Minneapolis

#### **Urban Planning**



Abu Dhabi, UAE

#### **Vertical Intensification**



Zoning







California

Land Use



**Miami-Dade** 



Switzerland



Greece





Honolulu



#### Recycling Communications





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## Walking and Transit Model

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Wisconsin

#### **Rail Status Monitoring**



USA



Germany

#### **Public Transit**



Washington



#### **Postal Delivery**



Los Angeles

Charlotte

## **13** CLIMATE ACTION



#### **Glacial Melt**

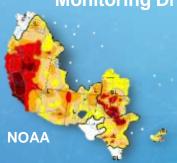


Monitoring Drought





South Carolina





San Francisco



Biomass Assessment

Africa

Groundwater

Change



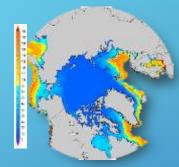
Turkey

Forest Carbon Reserves



South America

#### Calculating First Ice Freeze



Forest Restoration



Wallowa-Whitman NF

Californi



#### LIFE BELOW WATER 14



Marine and **Terrestrial Habitat** 



Abu Dhabi, UAE

Ecologically Significant Areas



**NOAA**—Monterey Bay

**Ocean Modeling** 

Coral Communities



Martinique







#### **Biodiversity**



Philippines

NOAA

#### **Marine Sanctuary**



California

#### **Reef Health**



**Cook Islands Protection Planning** 



Australia

Marine

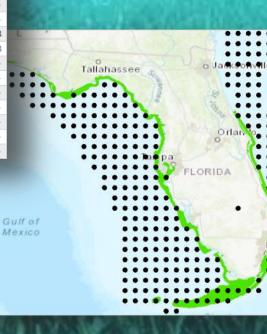




Predicting Environmental Phenomena Where Seagrasses Grows, Empirical Bayesian Kriging (EBK), Random Forest classifier

#### Ⅲ EMU\_Global\_90m ×

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⊿	OBJECTID	SHAPE	pointid	temp	salinity	appO2ut	dissO2	nitrate	percO2sat	
	11	Point Z	24	-1.433144	34.18222	<null></null>	<null></null>	<null></null>	<null></null>	
	13	Point Z	26	-1.439945	34.17537	<null></null>	<null></null>	<null></null>	<null></null>	
	118	Point Z	307	-1.387401	34.32391	<null></null>	<null></null>	<null></null>	<null></null>	
	753	Point Z	1739	-1.600642	34.03786	1.110779	7.211782	22.96304	86.85204	
	754	Point Z	1740	-1.56238	34.02853	0.997919	7.310482	21.98382	88.18443	
	871	Point Z	2184	-1.619098	33.9525	<null></null>	<null></null>	<null></null>	<null></null>	
	872	Point Z	2185	-1.678768	33.97821	<null></null>	<null></null>	<null></null>	<null></null>	
	882	Point Z	2211	-1.616092	33.94105	<null></null>	<null></null>	<null></null>	<null></null>	
	884	Point Z	2219	-1.697907	33.92028	<null></null>	<null></null>	<null></null>	<null></null>	
	885	Point Z	2220	-1.68471	33.92426	<null></null>	<null></null>	<null></null>	<null></null>	
	886	Point Z	2221	-1.69101	33.94196	<null></null>	<null></null>	<null></null>	<null></null>	
	887	Point Z	2222	-1.69061	33.93676	<null></null>	<null></null>	<null></null>	<null></null>	





#### **Empirical Bayesian Kriging**

from sklearn.ensemble import RandomForestClassifier import numpy as NUM import arcpy as ARCPY import arcpy.da as DA import pandas as PD import seaborn as SEA import matplotlib.pyplot as PLOT import arcgisscripting as ARC import SSUtilities as UTILS import os as OS



**Habitat Corridors** 

Atlanta

#### Wildlife Conservation



South Africa

Watershed



#### **Wilderness Tour**

Wildlife Imagery

**Invasive Species** 



**Steens Mountain Wilderness, Oregon** 

#### **Ecosystem Sensitivity**



**New Jersey** 

**Green Infrastructure** 

Bolivia

#### **Habitat Monitoring**



California



Using Deep Learning to Assess Palm Tree Health





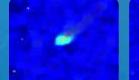


Leaf Spots and Leaf Blights of Palm

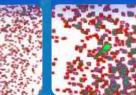
**Bud Rot of Palm** 



Graphiola Leaf Spot (False Smut) of Palm







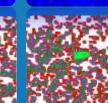


Image Classification to help Infer presence of contamination



Inferring presence of fungal & bacterial diseases using image classification enabling an immediate response to identify containment zones & to contain contaminations

#### • Benefits:

- Supervised Classification for autonomous systems
- Real-Time Detection & Accelerated Response







**Fire Station** Location/Allocation



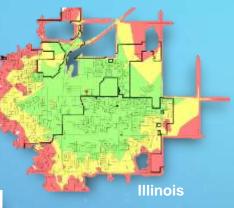
Texas

#### **EMS** Resources



**Tel Aviv, Israel** 

#### **Fire Response Times**



**Marathon Viewshed** 



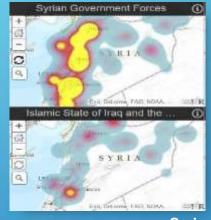
London, England

**Protection** 

# **Acts of Terrorism** Infrastructure

DHS

#### Violence Hot Spots



Syria

#### **Officer Involved Shootings**



Texas

Spatiotemporal Crime Patterns

Louisiana

16 PEACE AND JUSTICE STRONGINSTITUTIONS

### **City in Motion** Geography-Wide



**CRM** Demographics, Visitor Lines



Analytics Home/Work Locations



Monitoring

Signaling Network Movement, Roaming



Data Packet Inspection DPI Web Activity 1 Billion Records Daily



















