

AqHive

Research for Impact

Hadi Jaafar

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Chair, Department of Agriculture

Faculty of Agricultural and Food Sciences

Remote Sensing Summer School – May 30- June 2, 2022



DEPARTMENT OF
AGRICULTURE



CORE OF REMOTE SENSING & GEOSPATIAL ANALYSIS AT FAFS

The Unit for the Remote Sensing
and Geospatial Lab and the Smart Irrigation Lab at the
American University of Beirut

01

ABOUT US

The unit

02

RESEARCH

Active Projects

03

OUTREACH

Workshops & Field Work



01

ABOUT US

Our philosophy, partners, and services

OUR PHILOSOPHY 01

The Middle East and other regions in the world are suffering from an expanding population, dwindling water resources, and an increase in the frequency of conflicts, thereby threatening food and global security. Monitoring water use and stress at the field scale is of great importance to the success of agriculture in such areas.

—Dr. Hadi Jaafar
Founder & Lab Lead, AgHive



OUR PILLARS 01

Remote Sensing

Smart Irrigation

Water Resources

Big Data

Software



DEPARTMENT OF
AGRICULTURE

OUR RESOURCES 01

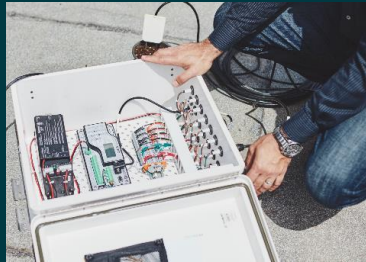
GRADUATE STUDENTS/ RESEARCHERS



HIGHLY MOTIVATED TEAM

OUR RESOURCES

2 Laboratories



2 Eddy
Covariance
towers;
2 Surface
Renewal
Stations
5 Infrared
Radiometers
3 Weather
Stations



Boundary Layer
Scintillometer

01 Experimental Field in Beqaa
(AREC)



OUR SERVICES

01



Agricultural Studies



Conflicts, Food & Water
Security Studies



GIS and Remote Sensing



Climate Change Impact Studies



Hydrology & Floodplain Studies



Irrigation System Design
Rehabilitation & Development



02

RESEARCH

Current projects and applications

WHAT SETS US APART?

Research for Social Good

Applications of GIS and remote sensing in smart irrigation, food security, and water resources management



Informed Action

We combine remotely sensed data, satellite imagery, and local knowledge



Motive

We aim to utilize and manage water resources for sustainable crop production



Responsiveness

We work directly with farmers and stakeholders and generate tools to help the user community combat pressing challenges

RECENT RESEARCH (Global Impact)

nature > scientific data > data descriptors > article


Data Descriptor | [Open Access](#) | Published: 12 August 2019

GCN250, new global gridded curve numbers for hydrologic modeling and design

Hadi H. Jaafar , Farah A. Ahmad & Naji El Beyrouthy

<https://jaafarhadi.users.earthengine.app/view/hydrologic-curve-number#GEE>

Earth Engine Apps ^{Experimental}

 Search places

GCN250m: Global Hydrologic Curve Number Explorer

This interface allows users to visualize the gridded hydrologic curve number dataset at ~250m resolution globally. Use the mouse-wheel to zoom in and out and click the map to query a wet, dry, or average antecedent runoff conditions. You can also zoom to a specific country of interest.

Version 1.0

Please cite: Jaafar, H. H., Ahmad, F. A., & El Beyrouthy, N. (2019). GCN250, new global gridded curve numbers for hydrologic modeling and design. *Scientific data*, 6(1), 1-9.

<https://www.nature.com/articles/s41597-019-0155-x>

Select layer:

Choose a runoff condition... 

Hint: Layer transparency can be changed at the top-right of the screen (layers box)

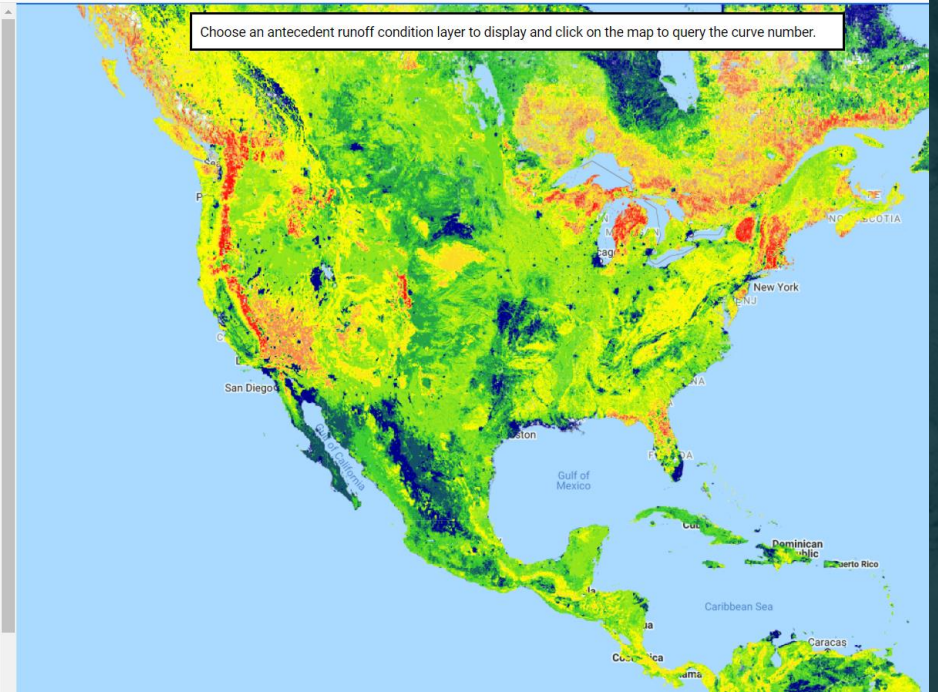
Zoom to Country:

Choose a country ... 

For more information:

Curve numbers for Hydrologic Runoff based on GLCC

Manuscript



RECENT RESEARCH (Global Im

remote sensing

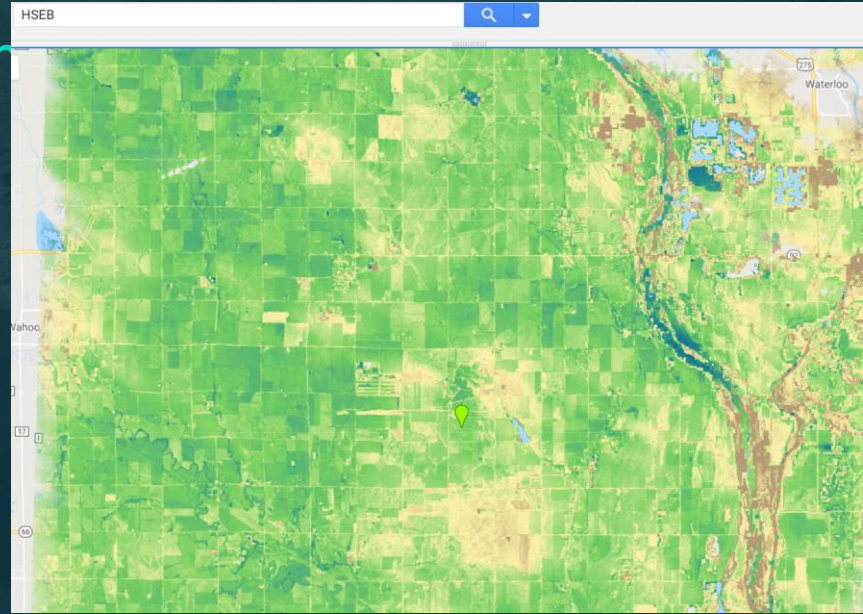


Article

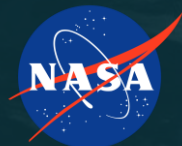
GYMEE: A global field-scale crop yield and ET mapper in Google Earth Engine based on Landsat, weather and soils data

Hadi Jaafar^{1,*}, Roya Mourad¹

¹ Department of Agriculture, Faculty of Agricultural and Food Sciences, American University of Beirut, Bliss St., Beirut, 2020-1100, Lebanon, hj01@aub.edu.lb, Corresponding Author
* Correspondence: hj01@aub.edu.lb; Tel.: +961-1-350-000



ACTIVE PROJECTS 02



MuSLI



Characterizing Field-Scale Water Use, Phenology and Productivity in Agricultural Landscapes using Multi-Sensor Data Fusion



Ministry of Foreign Affairs
Netherlands

ITSET



United Nations
Educational, Scientific and
Cultural Organization



Institute for
Water Education
in partnership with UNESCO

Integrating Time Series ET Mapping into Operational Irrigation Management Framework



TALANOA
WATER

Talanoa Water Dialogue for Transformational Adaptation to Water Scarcity Under Climate Change

ISDC

INTERNATIONAL
SECURITY AND
DEVELOPMENT
CENTER

Seeds for
Recovery



Department
for International
Development

The Long-Term Impacts of a Complex Agricultural Intervention on Welfare, Behavior and Stability in Syria

TIDES

A Force for Social Good



SPIRM

Google.org

Smart and Precision Irrigation with Remote Sensing & Machine Learning

WORKSHOPS

03



FIELD WORK

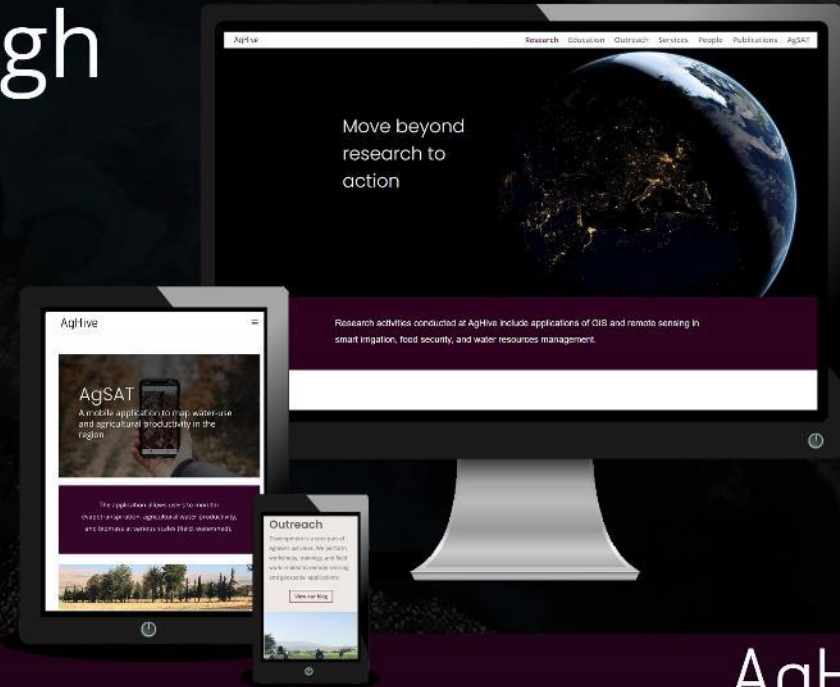
03

Micasense multispectral and thermal sensor



Access the full list of **projects** and **publications** on our website

See the world through AqHive



sites.aub.edu.lb/aqhive/

AqHive

HSEB: A global 30-m ET model using harmonized Landsat and Sentinel-2, MODIS and VIIRS



Remote Sensing of Environment

Volume 274, 1 June 2022, 112995



A global 30-m ET model (HSEB) using
harmonized Landsat and Sentinel-2, MODIS and
VIIRS: Comparison to ECOSTRESS ET and LST

Hadi Jaafar ^{a, *}, Roya Mourad ^a, Mitch Schull ^{b, c}

Outline

- Introduction

 - Why measure ET?

 - How to measure/estimate ET?

- Methodology

 - HSEB Model Description

 - Input Data

 - Thermal Sharpening

 - HSEB ET validation

- Results

 - HSEB ET performance

 - HSEB ET benchmarked with ECOSTRESS ET

 - Assessment of sharpened LSTs

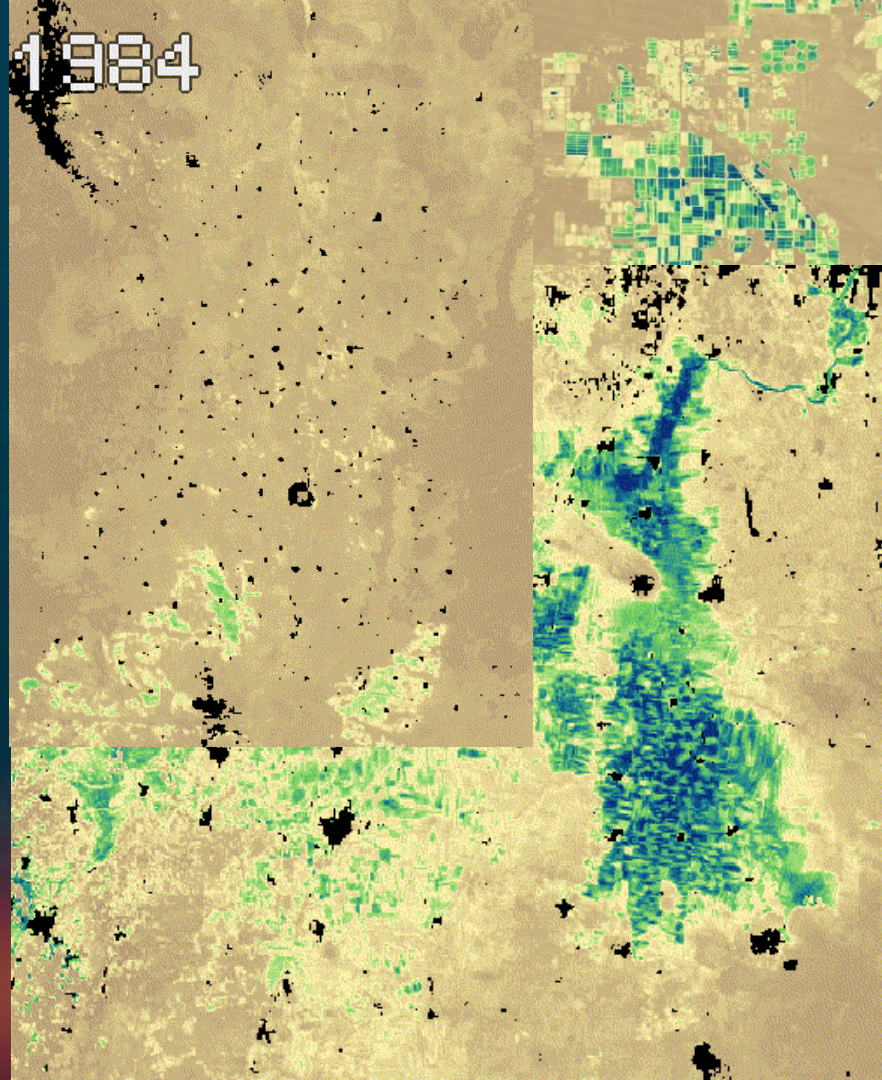
 - Impact of LST on HSEB ET results

- Applications and Conclusions

Why Measure Evapotranspiration (ET) in Agricultural Applications?

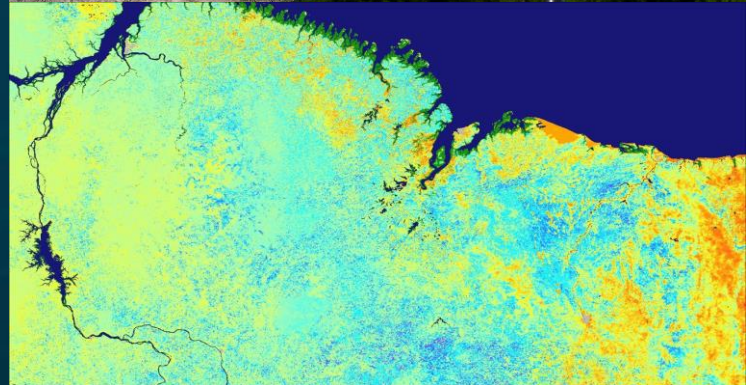
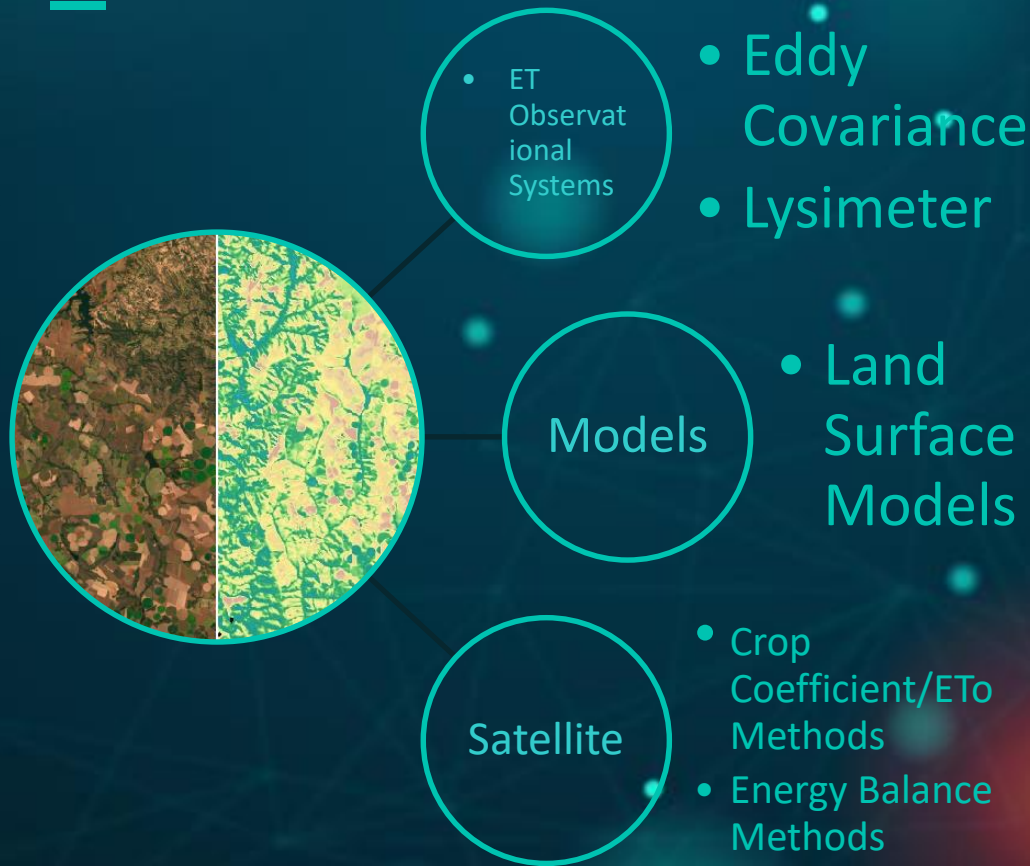
- Irrigation management
- Monitoring drought and crop stress
- Yield prediction
- Water use accounting (crop per drop)
- Water rights regulations and planning

1984

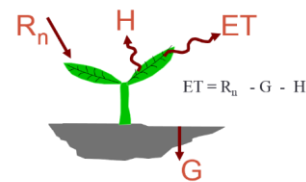


INTRODUCTION

How to measure/estimate ET?

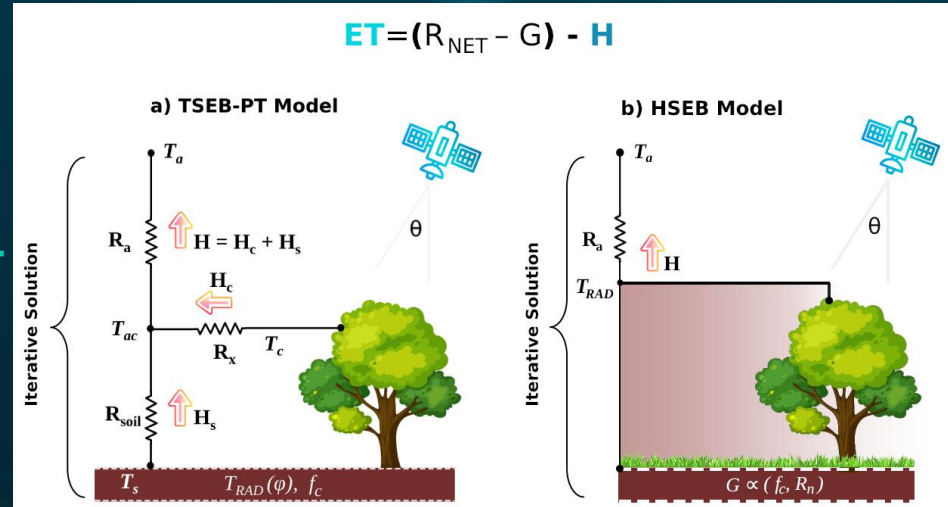


$$ET_o \times \begin{matrix} K_s \times K_c \text{ adjusted} \\ \text{water \& environmental stress} \end{matrix} = \begin{matrix} ET_c \text{ adj} \\ \uparrow \end{matrix}$$



Hybrid single-source energy balance (HSEB)

- Single-source energy balance models, where vegetation & soil are analyzed in a combined energy budget
- ET is calculated as a “residual” of the energy balance.
- Thermal imagery is key + solar irradiance



HSEB Model Description

—

$$R_n - G = H + LE$$

- Components of the radiation balance are used to determine the net radiation (R_n) – $SW \downarrow$, α , ϵ , T_s , $LW \downarrow$

$$R_n = (1 - \alpha) SW \downarrow + \epsilon LW \downarrow - LW \uparrow$$

- The ground heat flux (G) is parameterized as a function of fractional cover – LAI/NDVI relationships

HSEB Model Description

- HSEB calculates H using similarity theory

$$L = \frac{-u_*^3}{k \left(\frac{g}{\rho c_p T_a} \right) H + 0.608 c_p E T_a}$$

$$u_* = \frac{\bar{u} k}{\ln \left(\frac{z_u - d_0}{z_{OM}} \right) - \psi_M \left(\frac{z_u - d_0}{L} \right) + \psi_M \left(\frac{z_{OM}}{L} \right)}$$

$$H = \frac{(T_s - T_a) k u_* \rho c_p}{\left[\ln \left(\frac{z - d_0}{z_{0h}} \right) - \psi_h \left(\frac{z - d_0}{L} \right) + \psi_h \left(\frac{z_{0h}}{L} \right) \right]}$$

Wind, air temperature, humidity
(aerodynamic resistance ,
)

Various sub-modules for calculating needed components...

Input Data

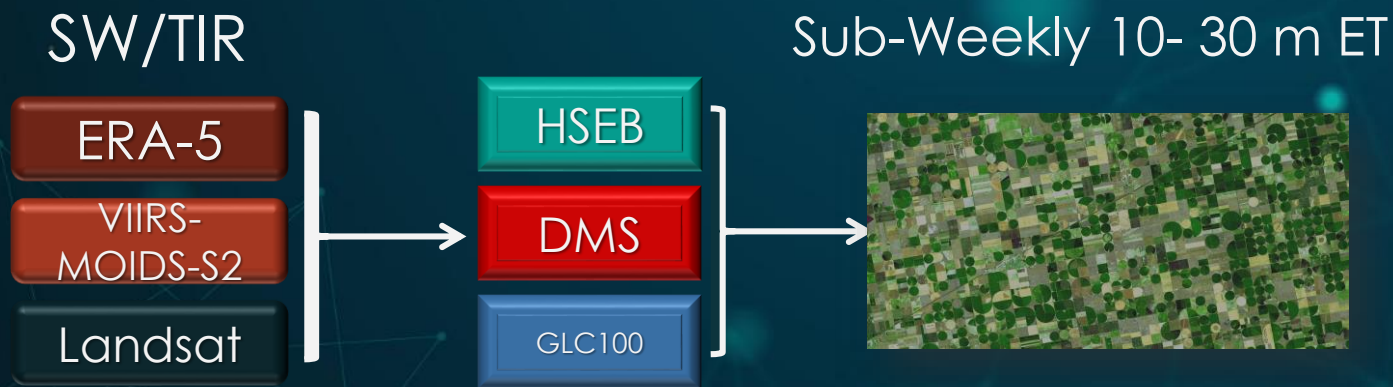
- HSEB model leverage globally available state-of-the-art thermal, surface reflectance, reanalysis, & land cover datasets in Google Earth Engine (GEE).



Google Earth Engine

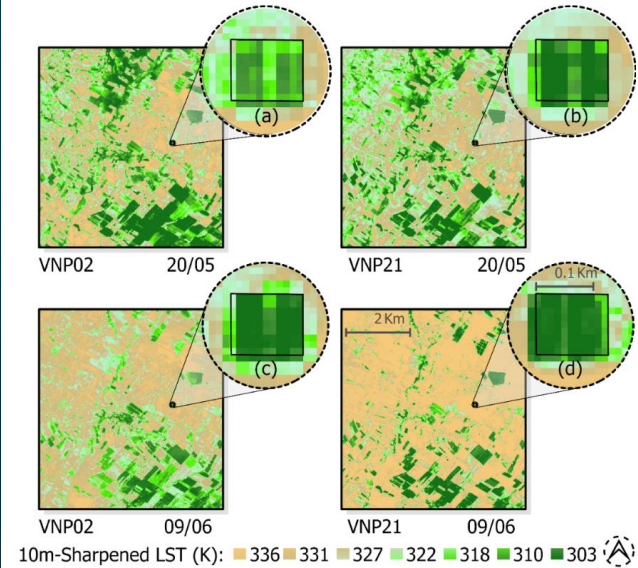
Data layers	Source	Resolution
Land Surface Temp	Landsat-7 & 8 / VIIRS-VNP02/VIIRS-VNP21/ MODIS-Terra (MOD11A1.006)	60-100 m/ 375 m/ 750 m/1000m
Surface reflectance	Landsat-4,5,7 & 8/9/Sentinel-2	30,20, 10 m
Meteorological forcing	ECMWF	0.2°/0.1°
Elevation	SRTM	30 m
Landcover	CGLS-LC100	100 m

FIELD-SCALE DAILY EVAPOTRANSPIRATION



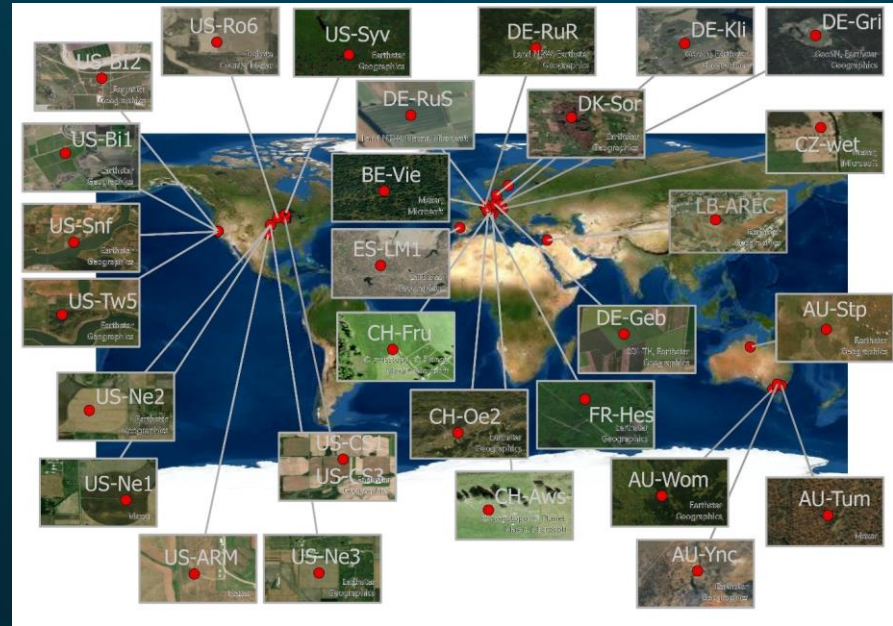
Thermal Sharpening

- HSEB runs TIR data sharpened with harmonized SR data from Landsat, S-2, & DEM.
- Random forest sharpener or TsHarp algorithm.
- Over fields in Lebanon, sharpened LST data were compared to LST observations



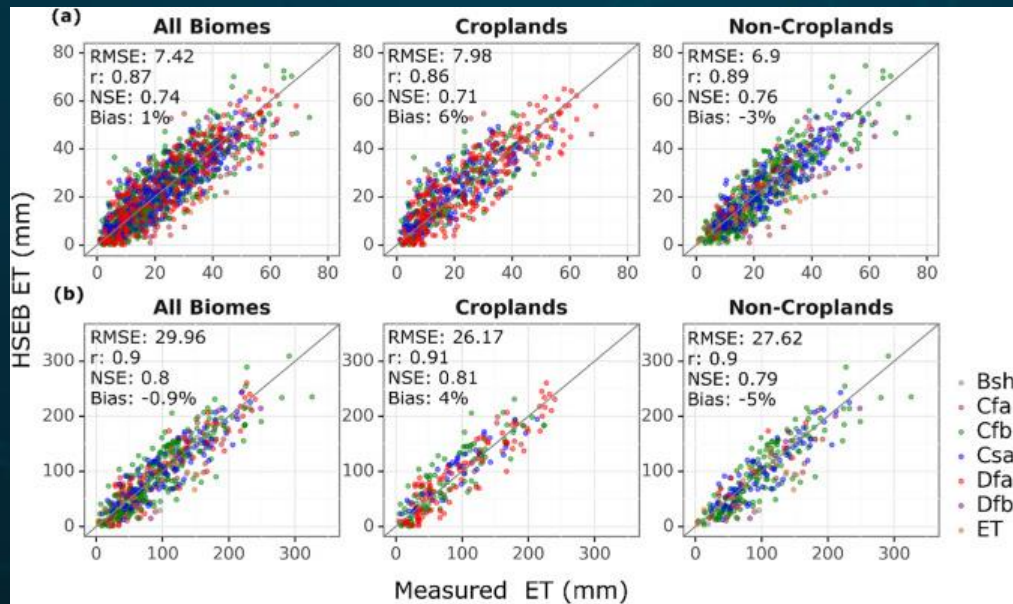
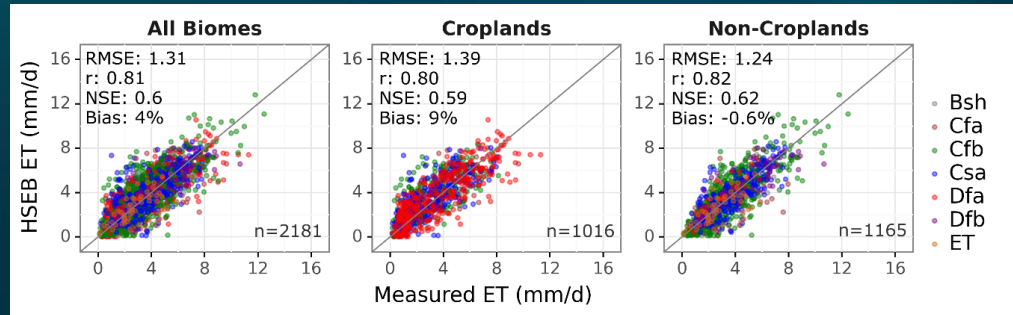
HSEB comparison with ground observations

- HSEB is validated over 2018–2020 with 2181 observations from 29 flux towers
- HSEB was constructed over target investigated sites (US, EU, and Australia)



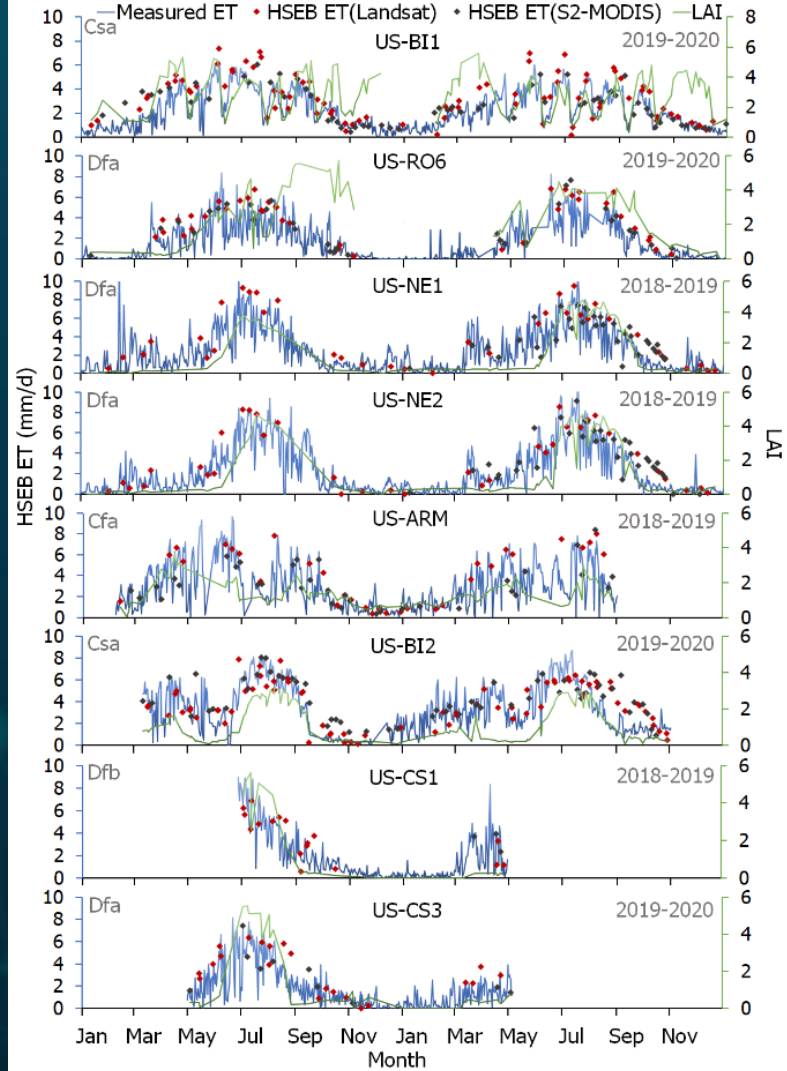
HSEB ET performance

- HSEB performed well in all considered biome types and climatic conditions ($r = 0.81, 0.87, \text{ and } 0.8$, NSE of 0.6, 0.74, 0.8, and a bias of 4%, 1%, and -0.9% at the daily, weekly, and monthly scales, respectively).
- RMSE averaged at 1.31 mm/day



HSEB ET performance

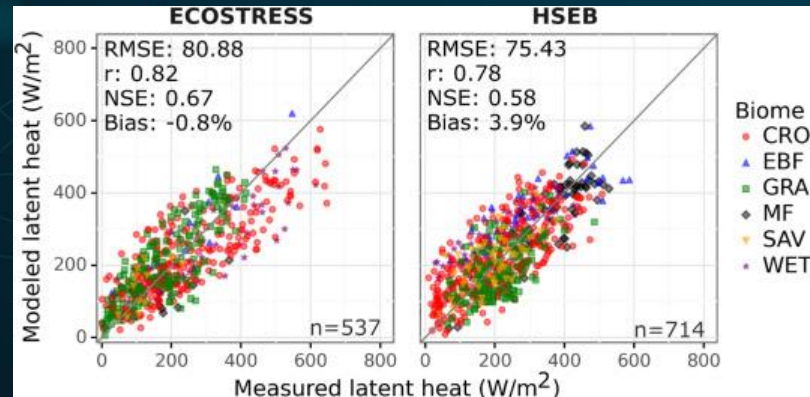
- The time series over US cropland sites, demonstrate good agreement between HSEB with the observed ET over the growing cycle.
- ET at these cropland sites generally correlates well with LAI



HSEB ET Benchmarked with ECOSTRESS ET



- HSEB produces acceptable results, although ECOSTRESS had a slightly lower bias (3.8% for HSEB vs. -0.8% for ECOSTRESS)
- HSEB performed better over croplands

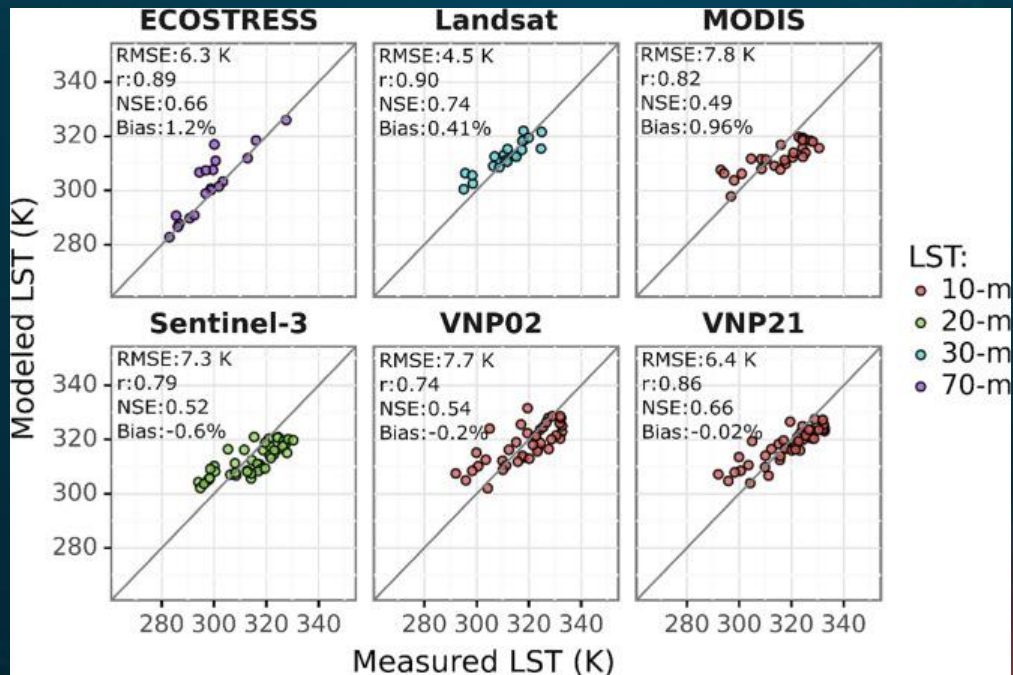


RESULTS

Assessment of Sharpened LSTs

RESULTS

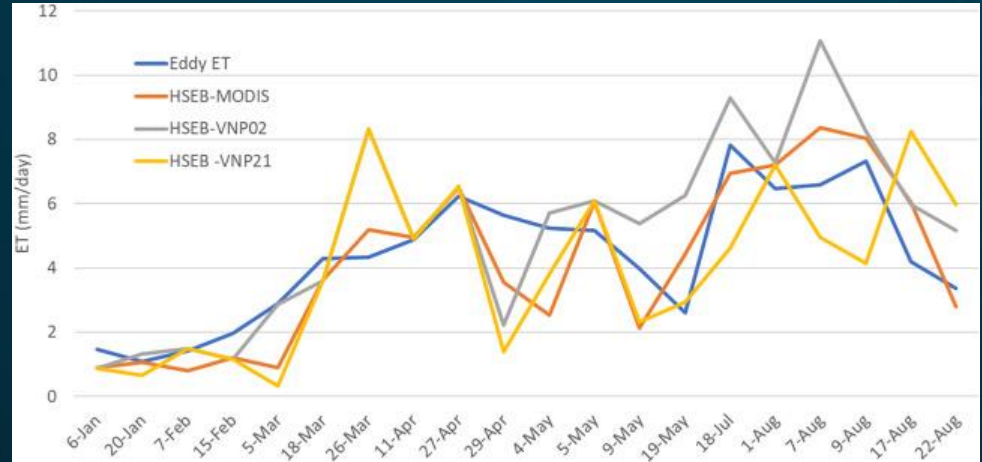
- Both ECOSTRESS and Landsat showed better performance at different LST ranges and time of day when compared to LST observation collected over a small potato field in Lebanon.



Case Study: potato field at AREC study site in Lebanon (2020 growing season)

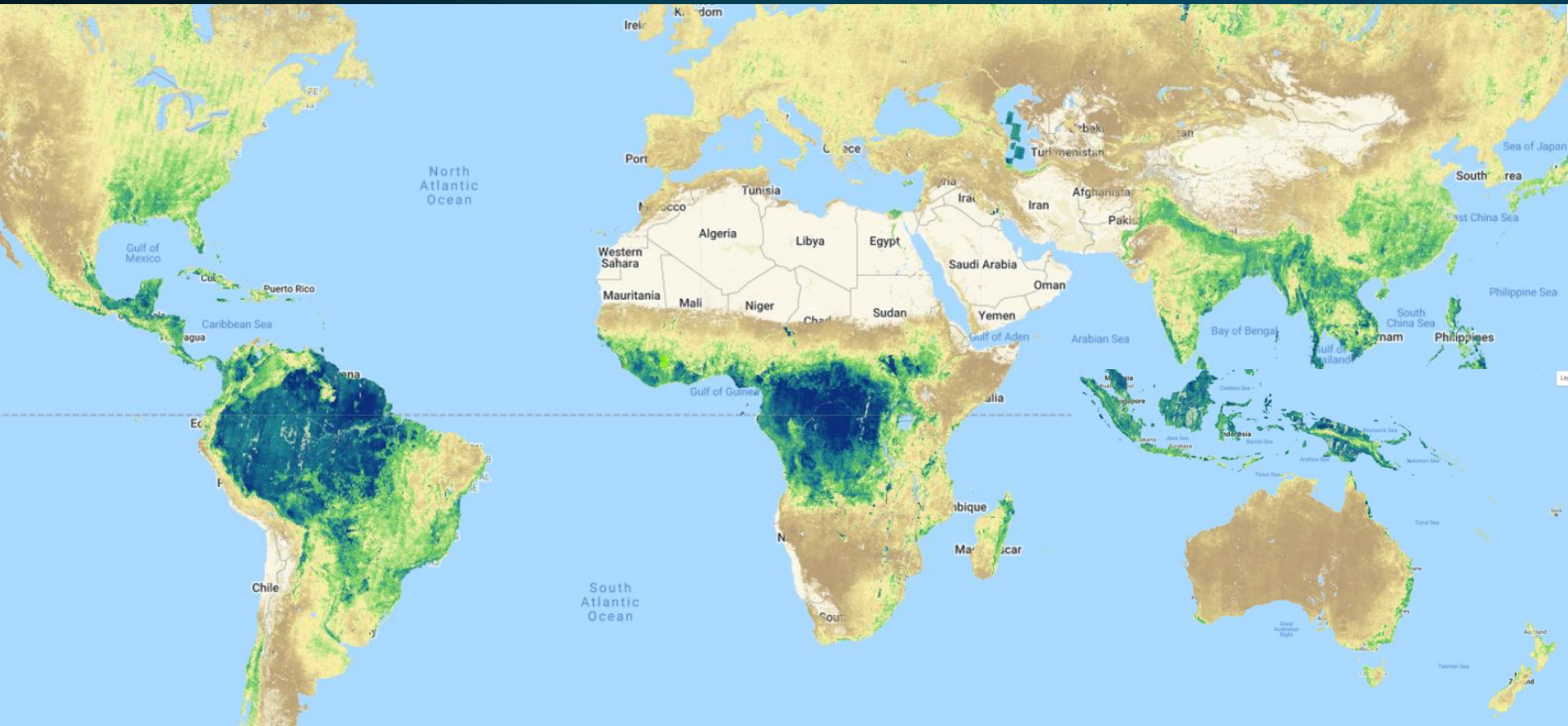
Impact of the LST product on HSEB ET results

- The analysis of impact of LST product used in HSEB on ET results at US-ARM site showed that HSEB with MODIS LST outperforms LST from VNP02 (an overestimate) and VNP21 (an underestimate).

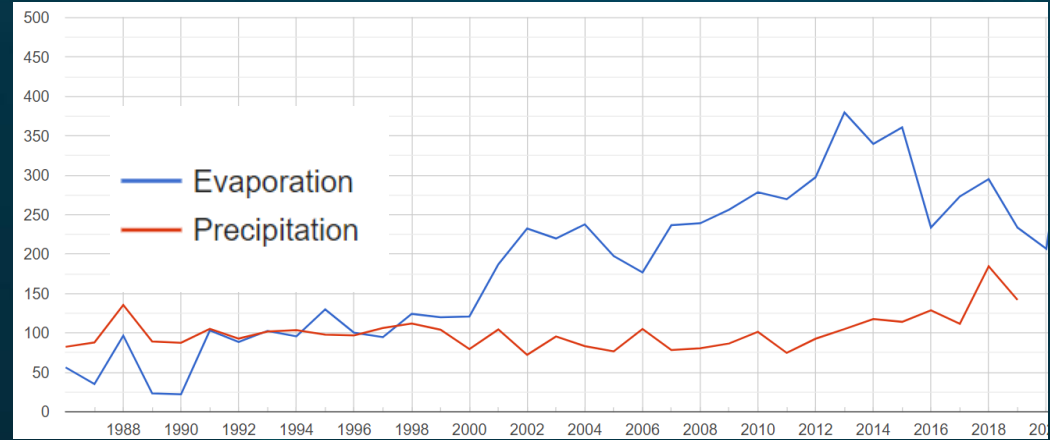


Case Study: HSEB ET time comparison with corrected Eddy ET at the US-ARM site for 2020

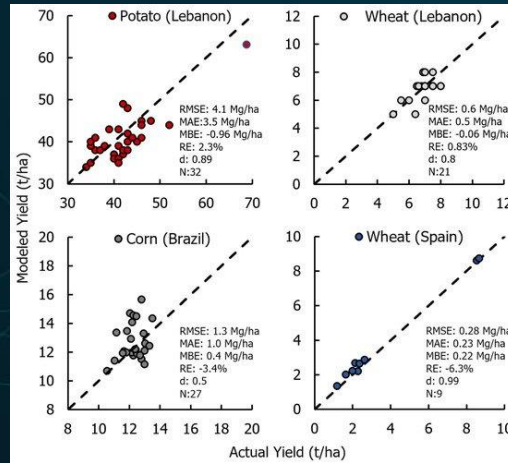
HSEB: Global 30m Landsat-based ET



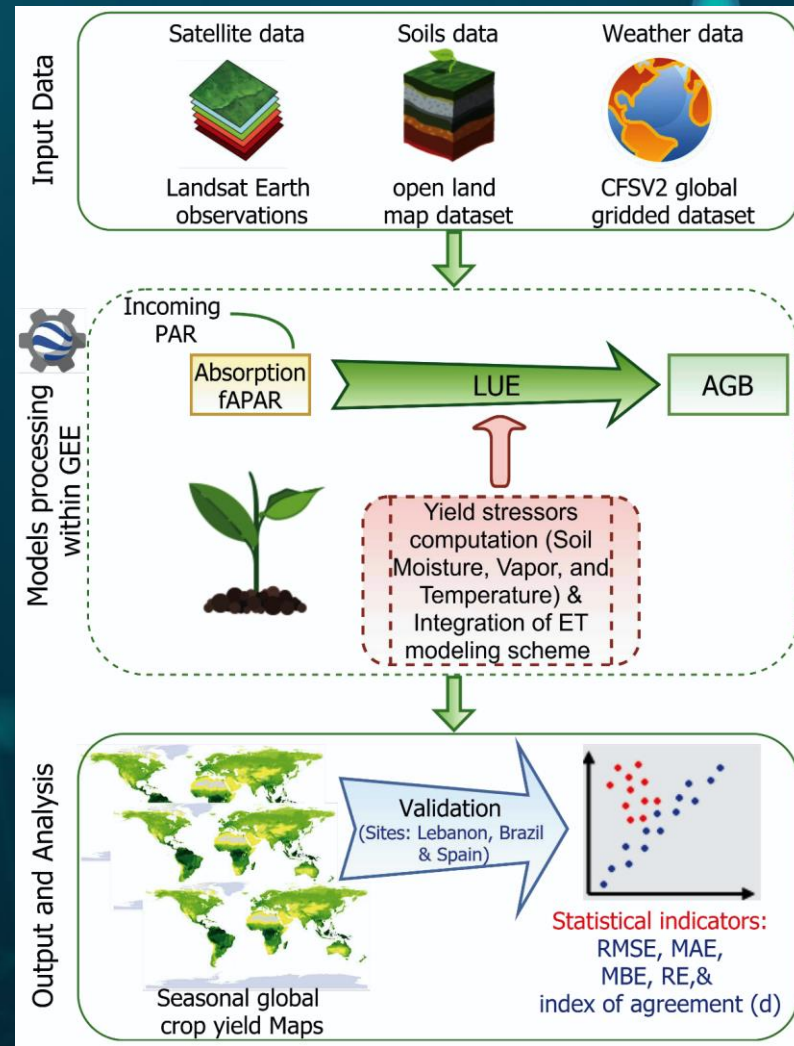
1986 @HadiHJaafar



GYMEE- global water use, biomass, and productivity, at the sub-field level using multisource satellites and GEE



Jaafar, Hadi, and Roya Mourad. "GYMEE: A Global Field-Scale Crop Yield and ET Mapper in Google Earth Engine Based on Landsat, Weather, and Soil Data." *Remote Sensing* 13.4 (2021): 773.



Conclusions

- A robust field-scale global evaporation mapper is presented
- HSEB generates a 2–3-day ET product globally at the 30-m and 10–20-m resolution
- HSEB provides a good representation of daily evapotranspiration and compares well with ECOSTRESS
- HSEB provides reasonably accurate results at the weekly and monthly time-scales
- HSEB has a high capacity to estimate field-scale ET over long periods accurately with high temporal frequency

Changing the Irrigation World



THE HEART OF EFFICIENT WATER MANAGEMENT

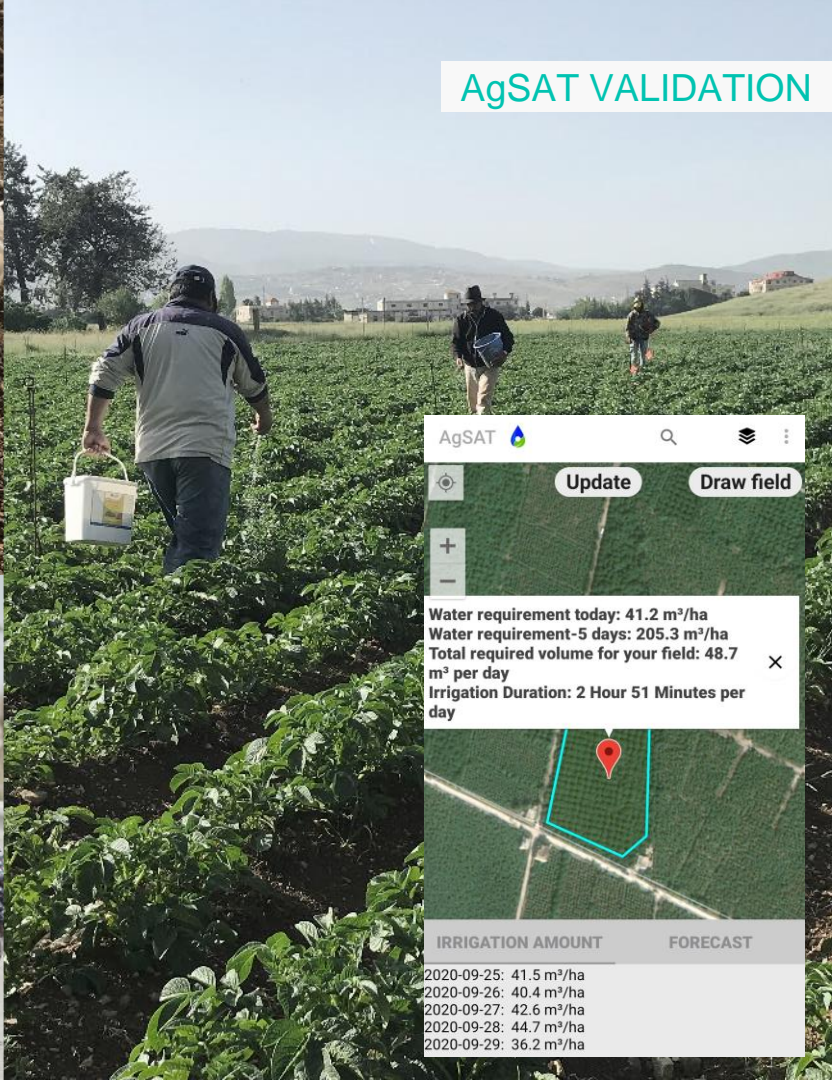
TO COMBAT WATER
SCARCITY







Now available on Google Play






AgSAT




AgSAT VALIDATION

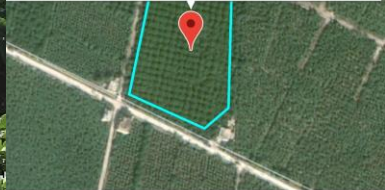
AgSAT    

 **Update** **Draw field**

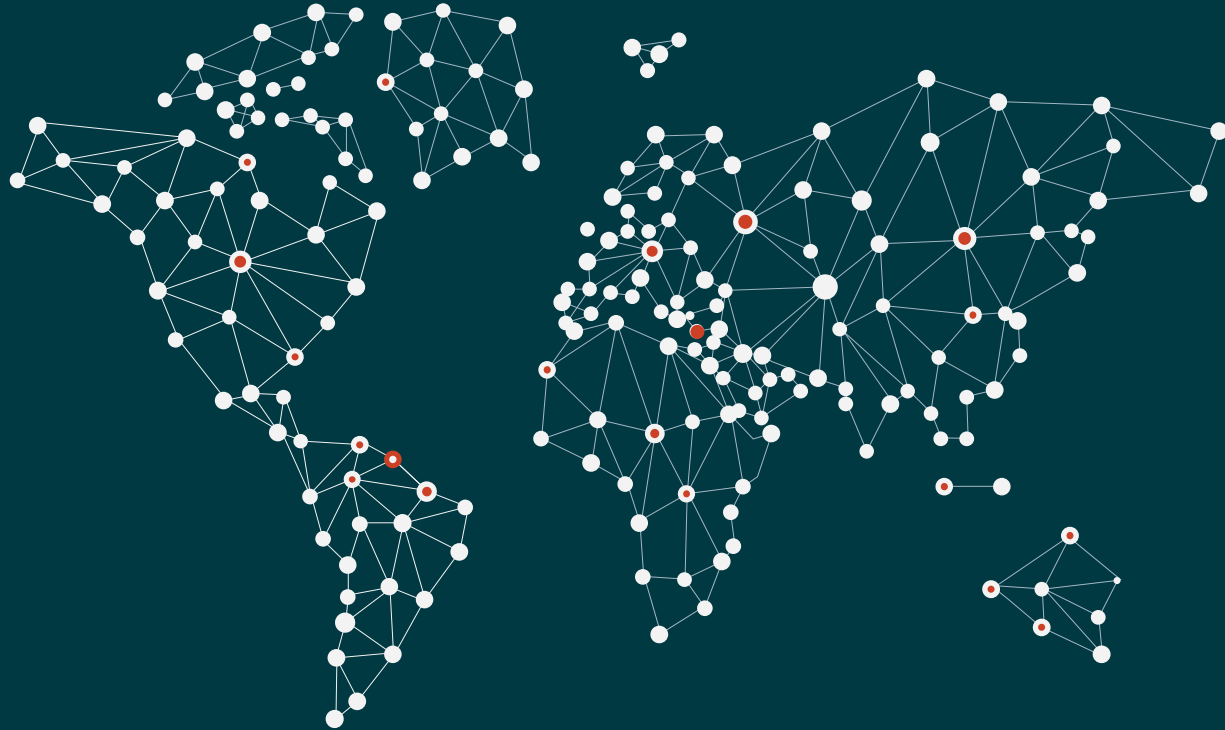
Water requirement today: 41.2 m³/ha
Water requirement-5 days: 205.3 m³/ha
Total required volume for your field: 48.7 m³ per day 

Irrigation Duration: 2 Hour 51 Minutes per day



IRRIGATION AMOUNT	FORECAST
2020-09-25: 41.5 m ³ /ha	
2020-09-26: 40.4 m ³ /ha	
2020-09-27: 42.6 m ³ /ha	
2020-09-28: 44.7 m ³ /ha	
2020-09-29: 36.2 m ³ /ha	

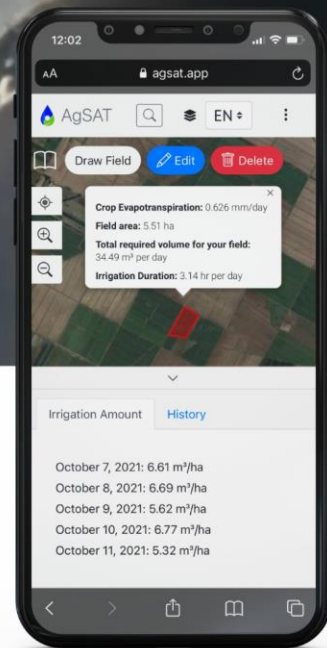
COUNTRIES WHERE AgSAT IS BEING USED





AgSAT

FOR A SUSTAINABLE WORLD



**IRRIGATE
MORE PRECISELY**

Anywhere, Anytime

✉ hj01@aub.edu.lb  agsat.app

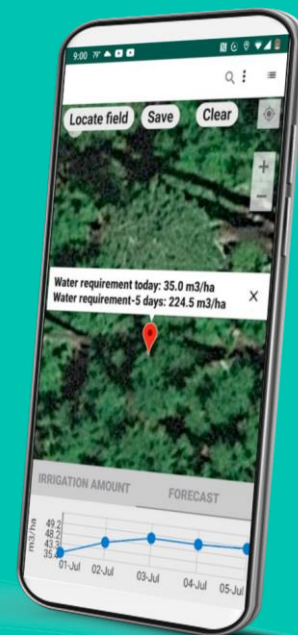
Increase
CROP YIELD
Save
THE ENVIRONMENT

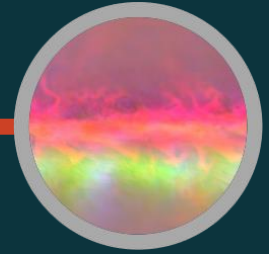
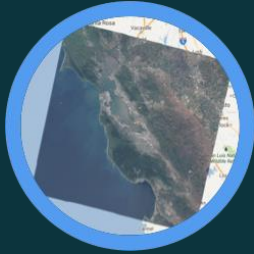
powered by 

A CASE STUDY

Mobile & web application for water management— AgSAT

- AgSAT—is a mobile app developed at AUB with GEE as a backend
- Calculates daily water requirements using remote sensing and weather data
- to all types of users, from small-holder famers to irrigation districts and regional water planners in the Arab countries
- Farmers input information specific to their farms to receive irrigation water information (volume & run time) based on crop vegetation status, weather conditions, and irrigation system type



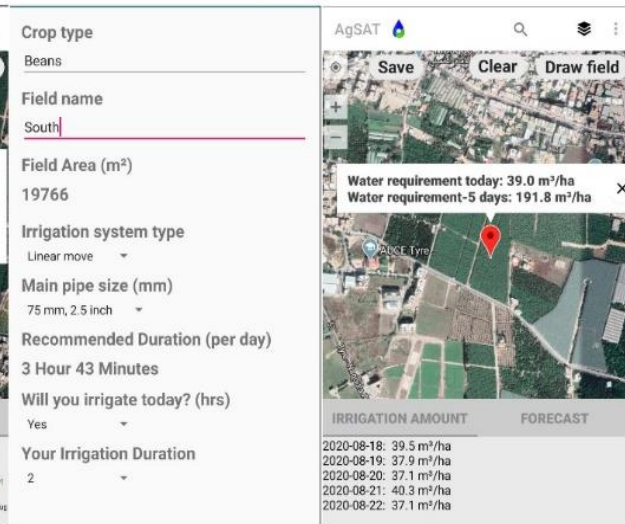
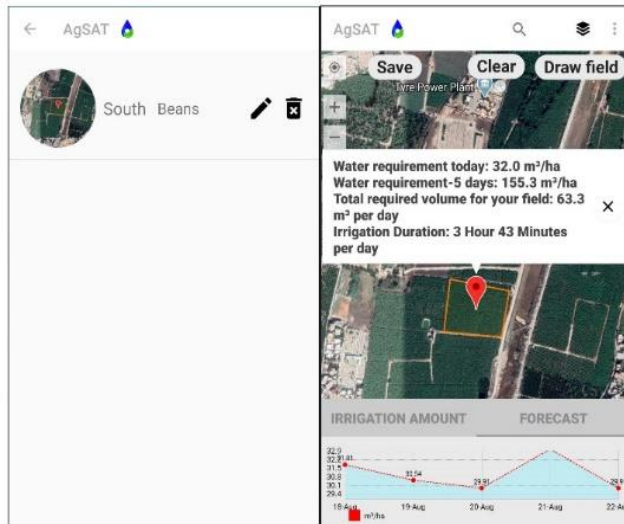
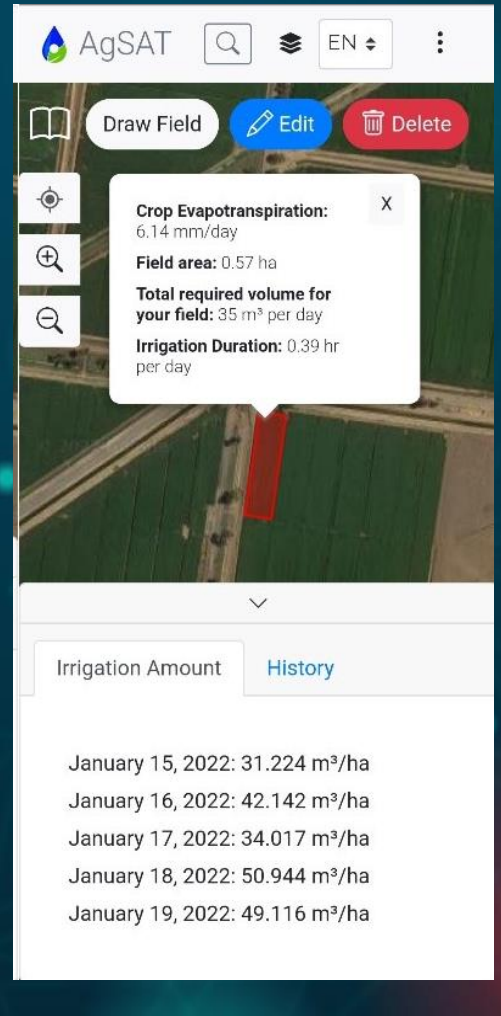
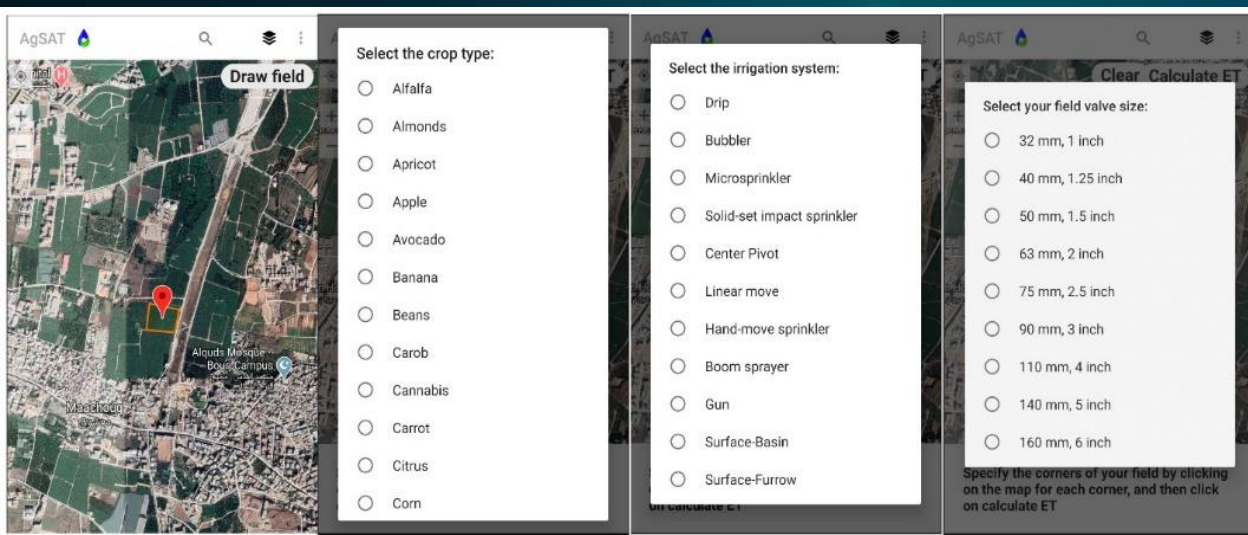


Farmer's own data





Reflectance-based
Growth Coefficients


ETref



Volumes per ha, Irrigation Run times, 5-Days ETc






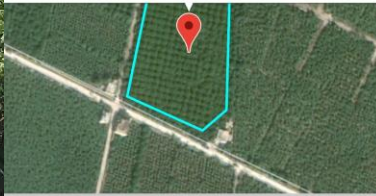
AgSAT    

 **Update** **Draw field**

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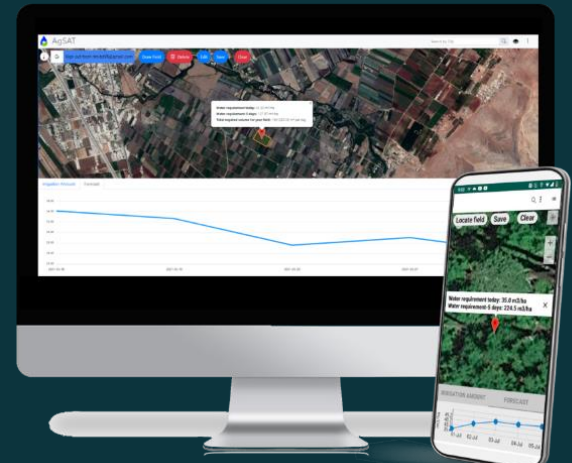
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WHERE CAN AgSAT BE USED

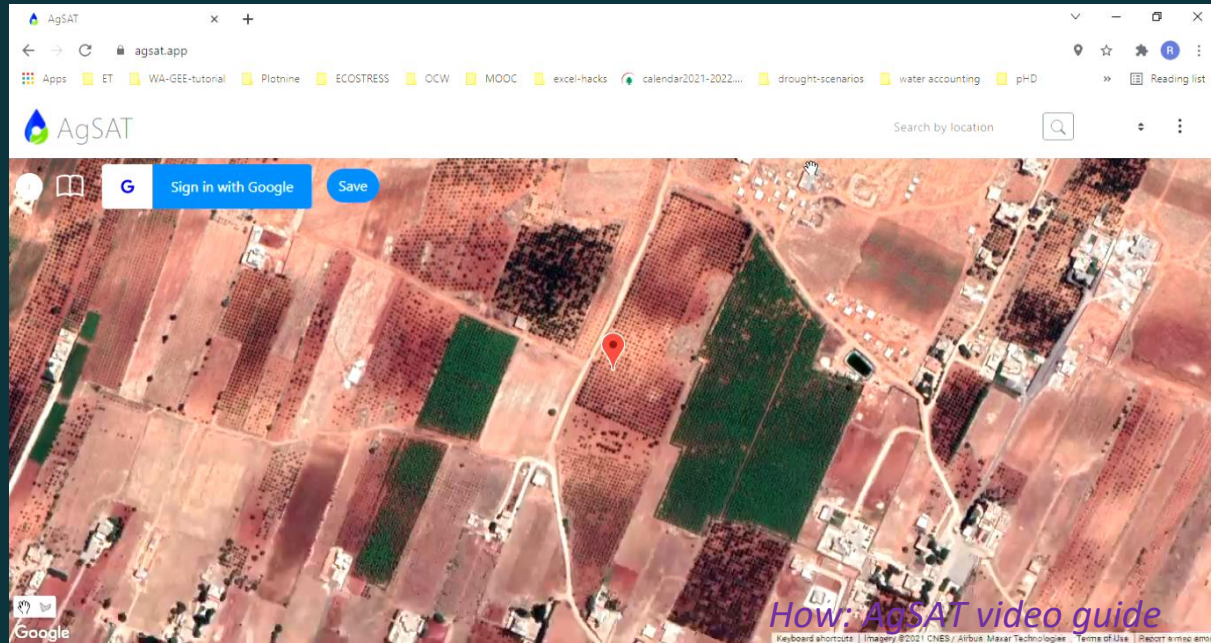
- Anywhere in all countries where irrigation is practiced
- where cellular network coverage is available
- where farmers have smart phones
- at the district level or water authority level
- Currently the app is being used in Sudan, Egypt, UEA, Saudi Arabia, Lebanon, as well as other countries

Trends in precipitation in the Maghreb region



HOW & WHO CAN UTILIZE AgSAT

- **Ministries and Municipalities**
- **Farmers**
- **Irrigation engineers**
- **Water authorities**
- **Large and small farming enterprises**
- **NGOs**
- **Ag companies**
- **Leading organizations (ESCWA, WB...)**



Case study- III

TALANOA-WATER

Integrated management of water resources in the Mediterranean



THANK YOU!

hj01@aub.edu.lb

<https://sites.aub.edu.lb/aghive/>

