



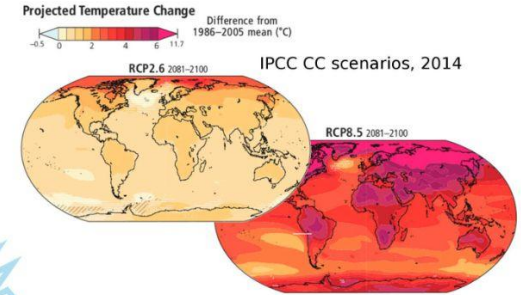
INTEGRATION OF MEASUREMENTS, REMOTE SENSING AND MODELS FOR SEMIARID SAVANNA-TYPE SYSTEMS MANAGEMENT.

WHAT ARE WE DOING?

PRECISION MANAGEMENT PLAN
(FARM AND BASIN)

CLIMATIC CHANGE

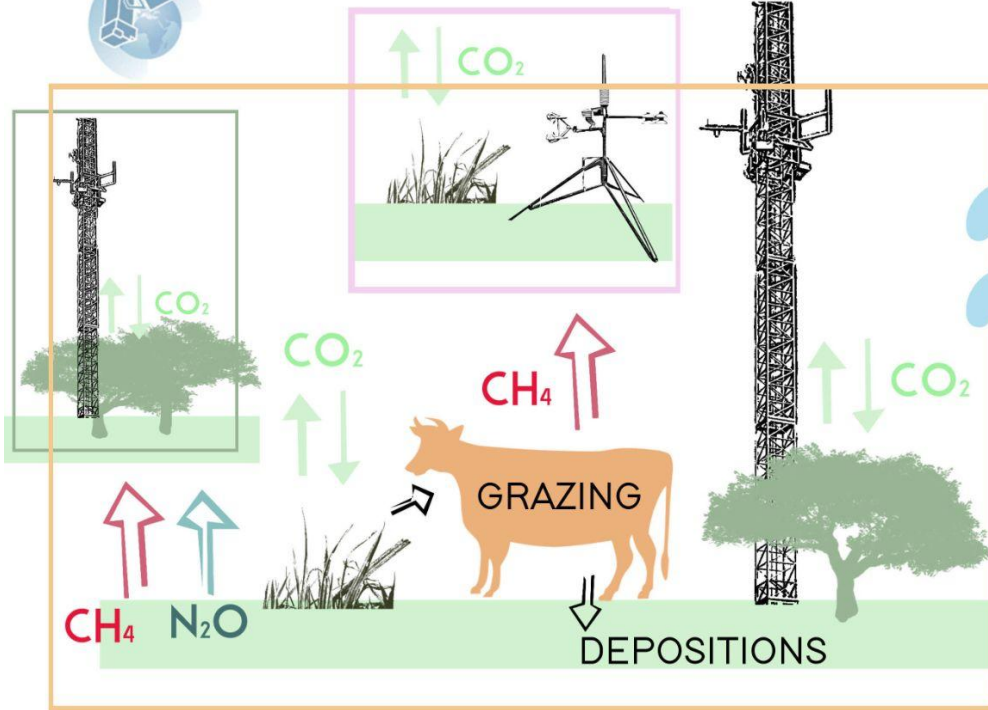
- ↑ AIR TEMPERATURE
- ↑ EXTREME EVENTS
- ↓ PRECIPITATION



$\text{CO}_2\text{eq} + \text{H}_2\text{O}$

MANAGEMENT PRACTICES

- REGENERATION
- GRASS COMMUNITY
- LIVESTOCK DENSITY
- LIVESTOCK MOBILITY
- FERTILIZATION
- INTENSIFICATION
- CLEARING
- ...



E.G. EXPERIMENTAL SITE STA CLOTILDE

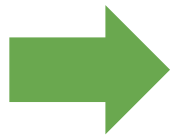
WHAT ARE WE DOING?

To provide recommendations for management that reinforce the ecosystem services that Mediterranean landscapes offer society, on local (e.g., landscape and resources management) and regional (e.g., management mitigation strategies regarding drought and carbon sequestration) scales, **integrating Earth Observation** and **low-cost sensors**

1 Improving the understanding of the processes governing biogeochemical flux exchanges between semiarid partially covered landscapes and the atmosphere.

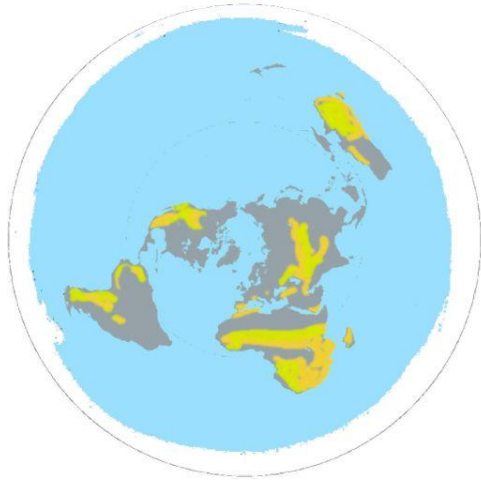
- Temperature heterogeneity
- Spectral Response
- Phenology...

2 Developing a robust modeling framework for water use and vegetation dynamics monitoring in semiarid partially covered ecosystems.



3 Development of scientific data series.

3. DATA IN SAVANNAS



1. SANTA CLOTILDE & LAS MORRILLAS DEHESAS IN CÓRDOBA, SPAIN.
2. SKUKUZA SAVANNA IN SOUTH AFRICA.
3. TONZI OAK SAVANNA IN CALIFORNIA, USA.



STA. CLOTILDE



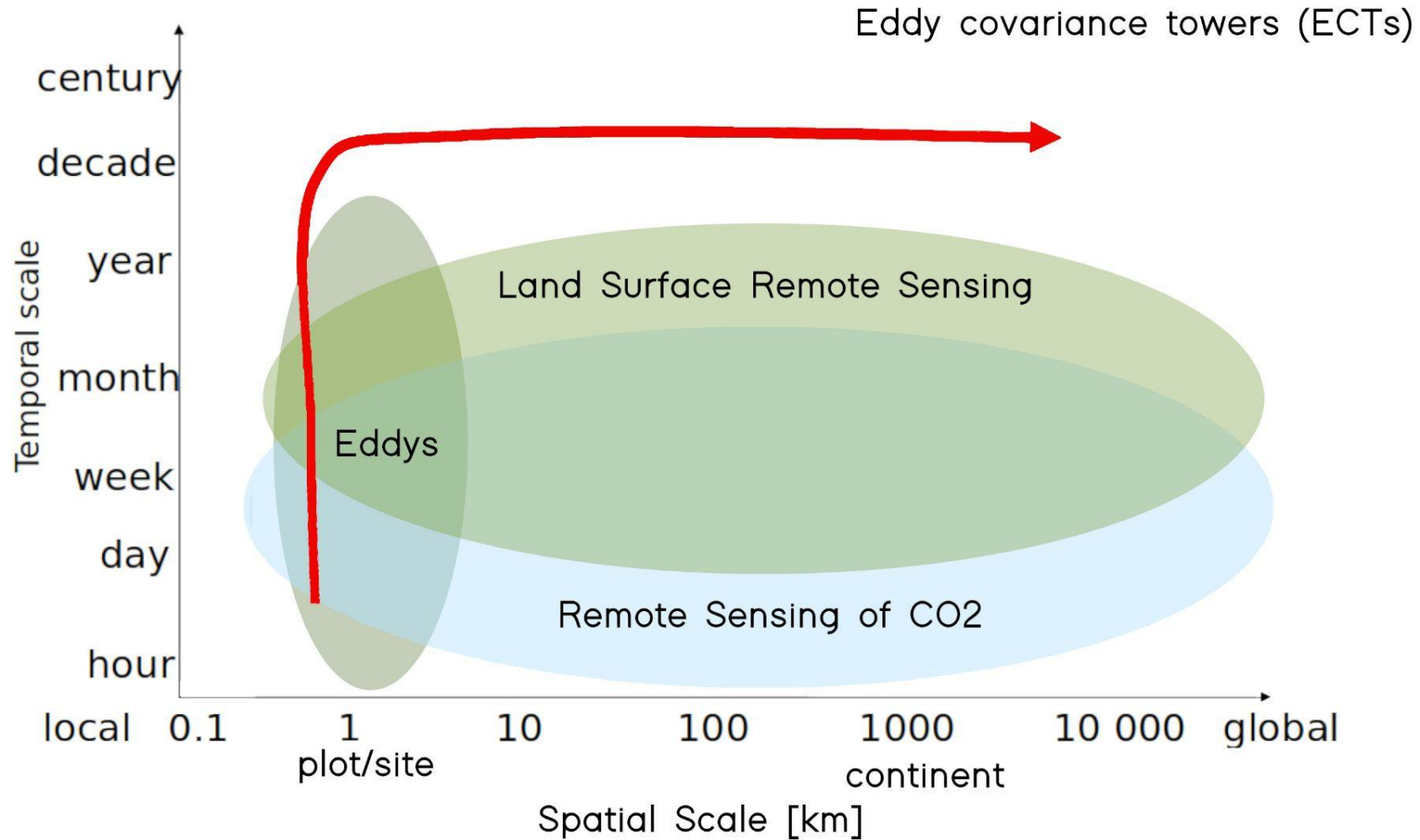
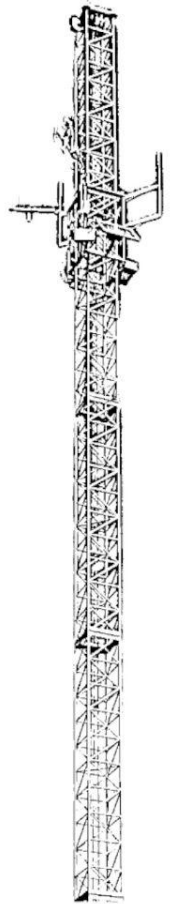
TONZI



SKUKUZA



3. DATA IN SAVANNAS



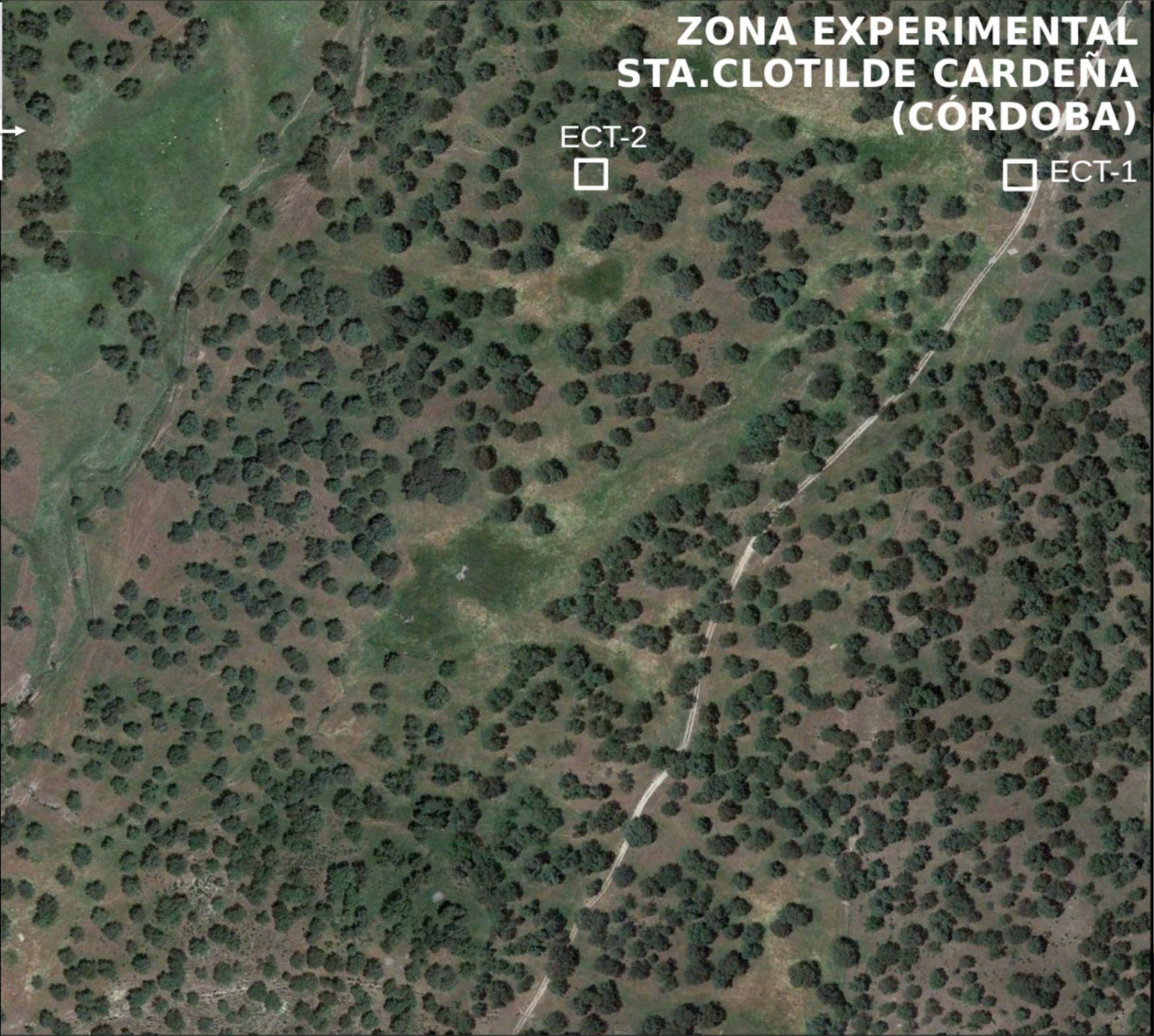
ADAPTED FROM MARKUS REICHSTEIN, MPI.

**ZONA EXPERIMENTAL
STA. CLOTILDE CARDEÑA
(CÓRDOBA)**

ECT-2



ECT-1

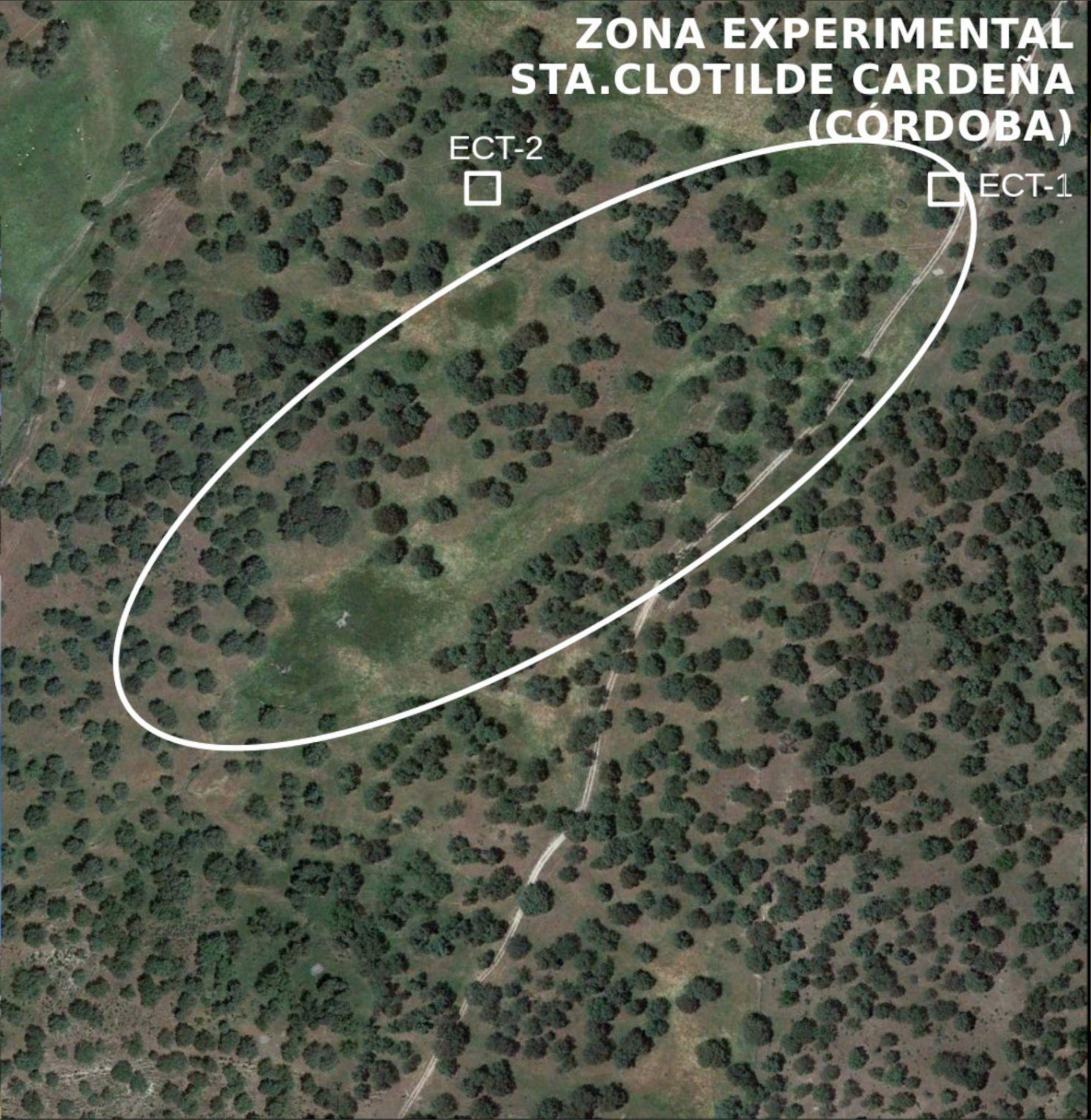




ZONA EXPERIMENTAL STA. CLOTILDE CARDEÑA (CÓRDOBA)

ECT-2
□

ECT-1
□



ZONA EXPERIMENTAL STA. CLOTILDE CARDEÑA (CÓRDOBA)



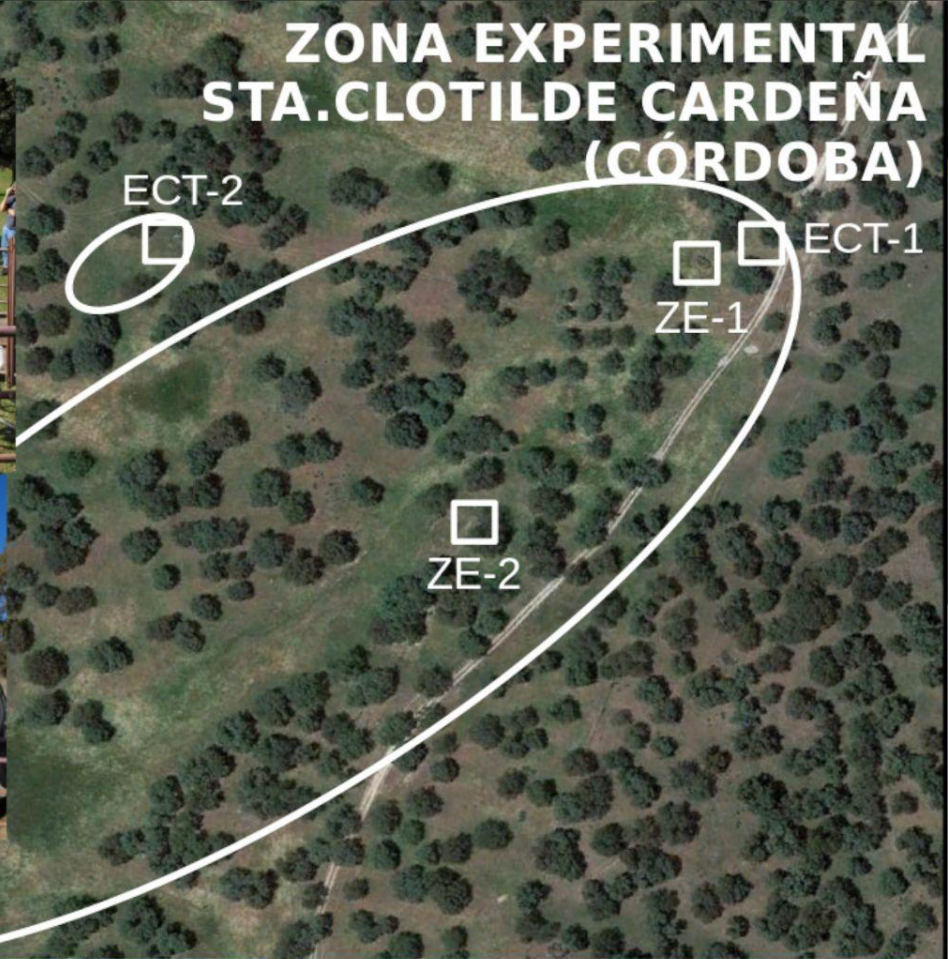
ECT-2



ECT-2



ECT-1



ECT-2

ZE-1

ECT-1

ZE-2



ZE-2



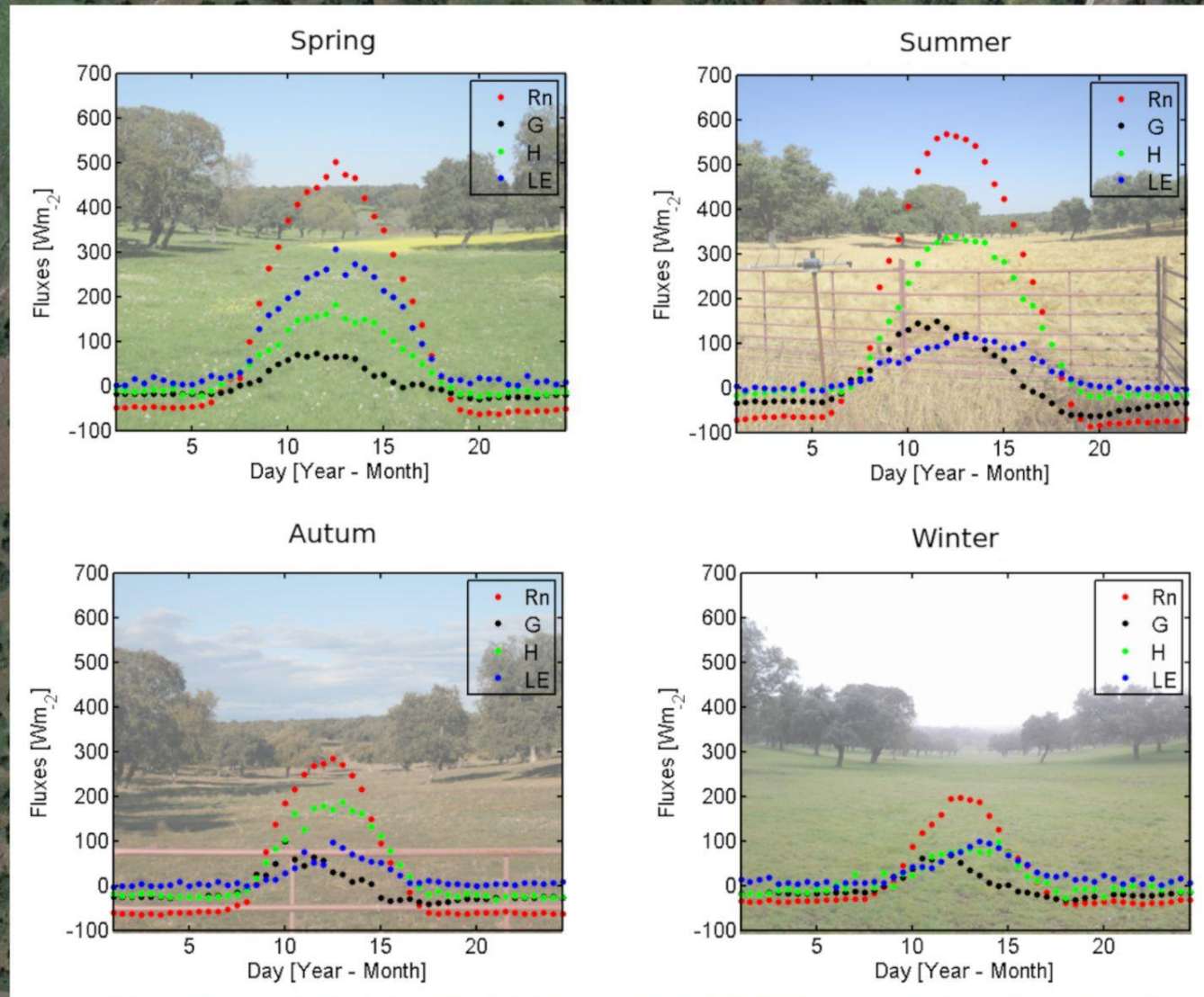
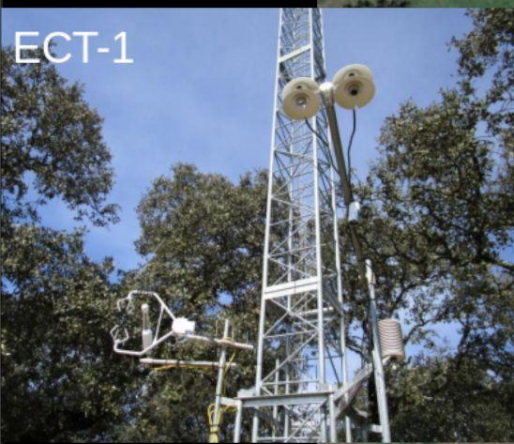
ZE-1

ZONA EXPERIMENTAL STA. CLOTILDE CARDEÑA (CÓRDOBA)

ECT-2



ECT-1



3. DATA IN SAVANNAS: SKUKUZA

African semiarid **rangelands** are a **mosaic of land uses**, where extensive **livestock** is the main economic activity.



From A. Andreu. Savanna Tiger Guide.

3. DATA IN SAVANNAS: SKUKUZA

Skukuza **area** is located in the Limpopo region, with great agricultural importance but subject to periodic droughts, and home of the **Kruger National Park**.



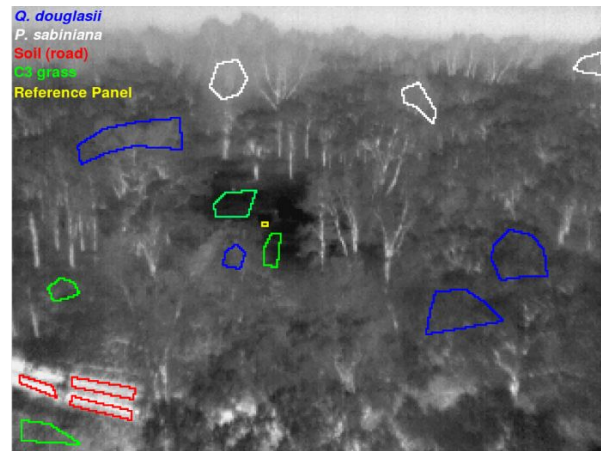
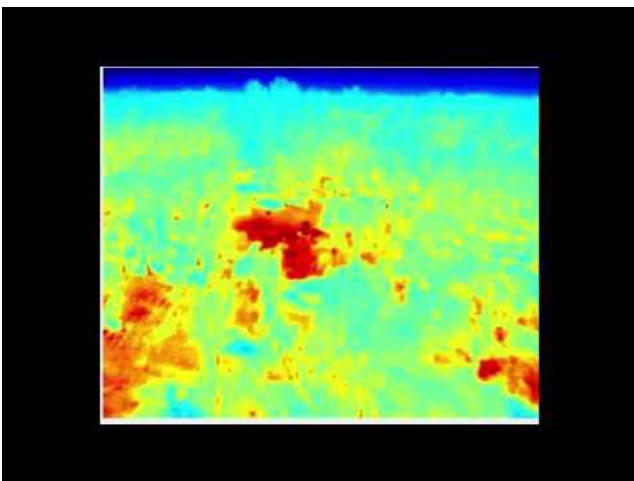
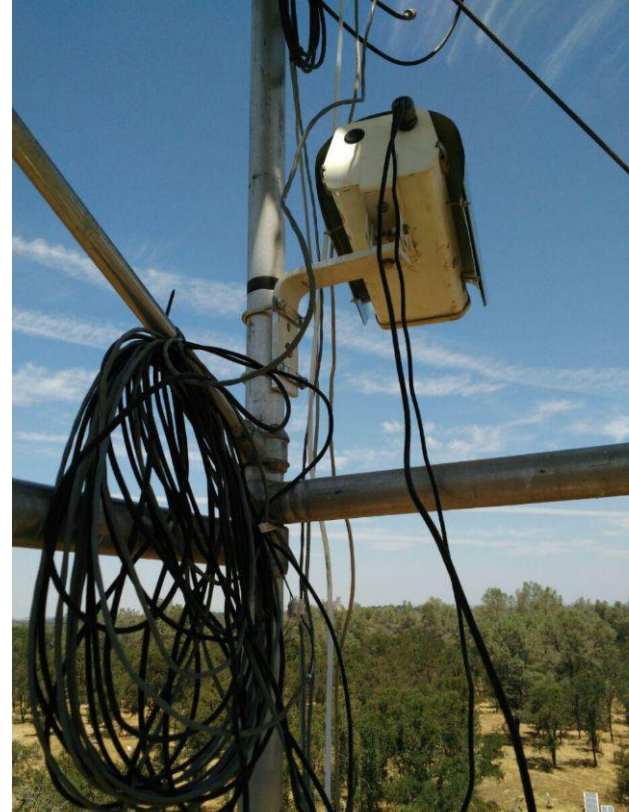
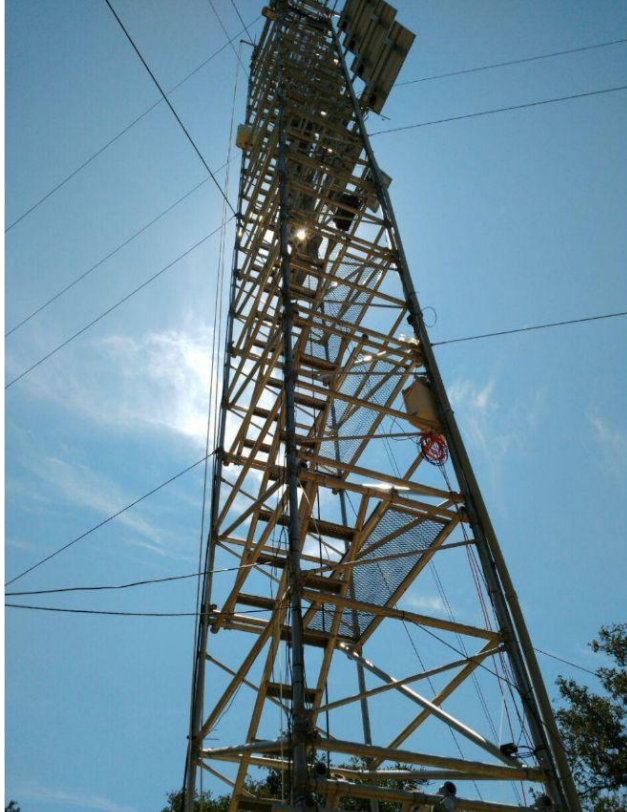
Bestok 25C 77F 11/02/2015 13:30 0000



*Drought in rangelands of South Africa
by Alexandra Dr. Sandhage-Hofmann*

3. DATA IN SAVANNAS: TONZI





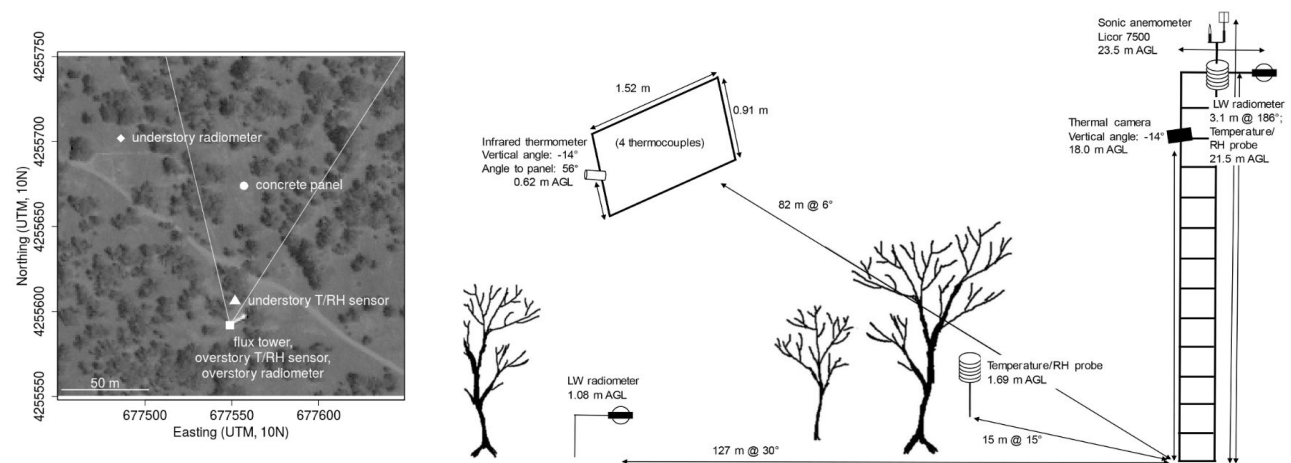
1. TEMPERATURE

Johnston et al., 2021 and 2022

Analyze canopy temperatures (fundamental controls on plant-level processes) of overstory vegetation, understory vegetation, and soil across space and time.

CHALLENGES:

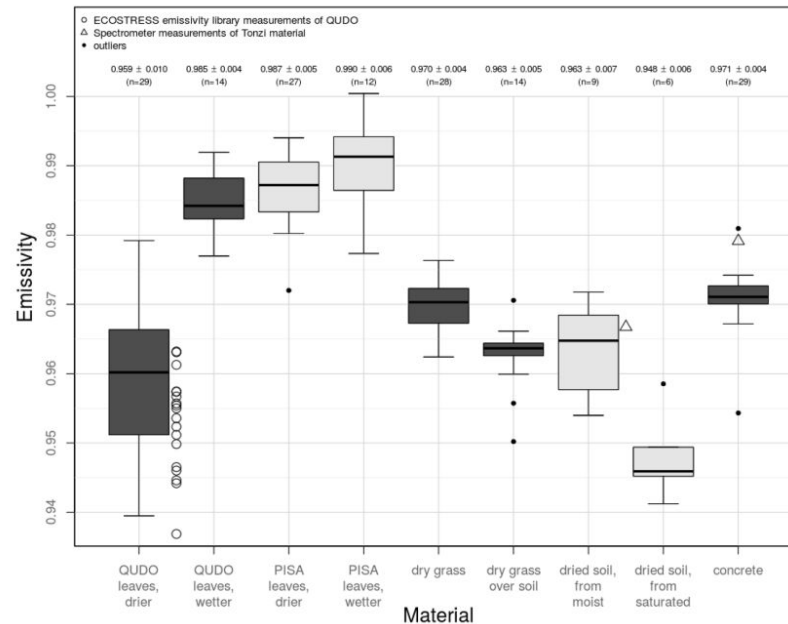
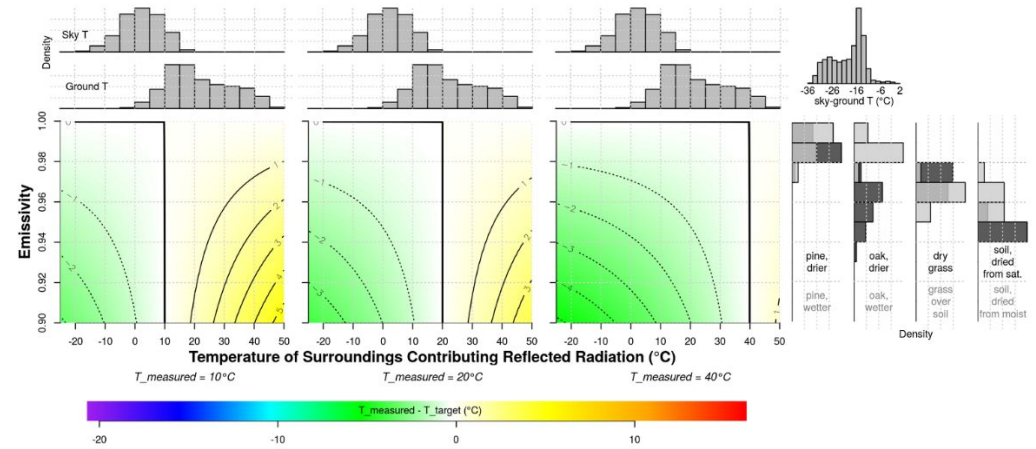
- 1 Emissivity (effectiveness in the energy emission as thermal radiation) and background radiation data required for image **calibration** are variable across the scene.
- 2 **Mixed pixels** are misleading as component temperatures diverge.
- 3 Targets of interest have very different pixel sizes associated with their different **distances** from the camera.



1. TEMPERATURE

Johnston et al., 2021 and 2022

Our ecosystem's relatively high emissivity values facilitate the calibration procedure. In contrast, **mixed pixels** (with the potential of divergent component temperatures) influence the calibrated target temperatures more.

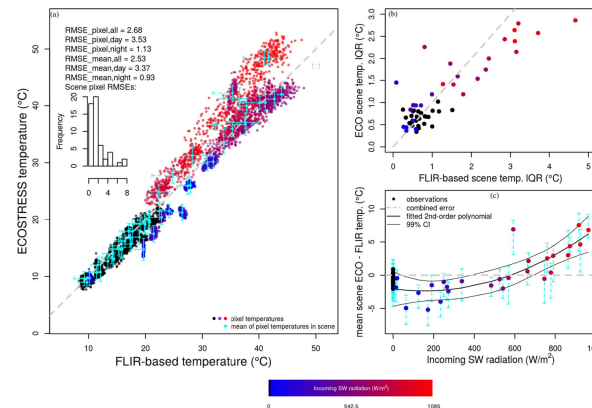
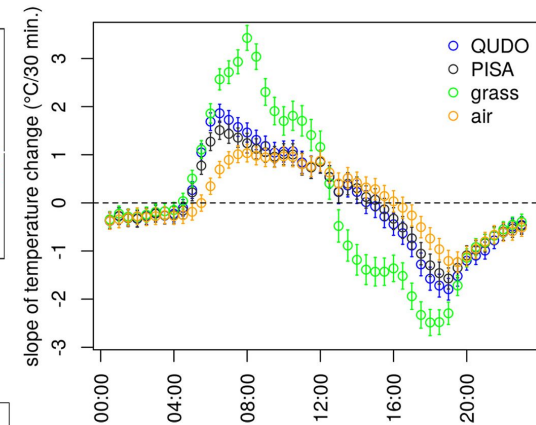
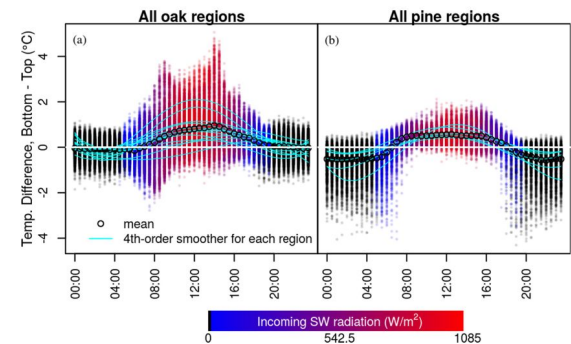
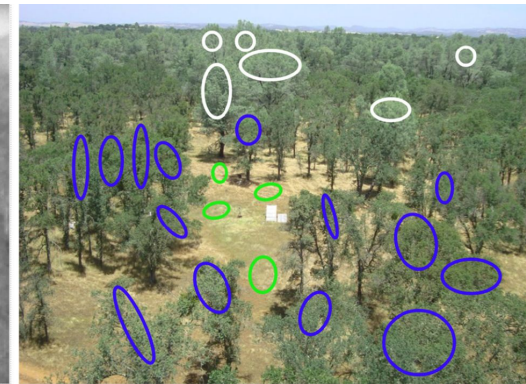
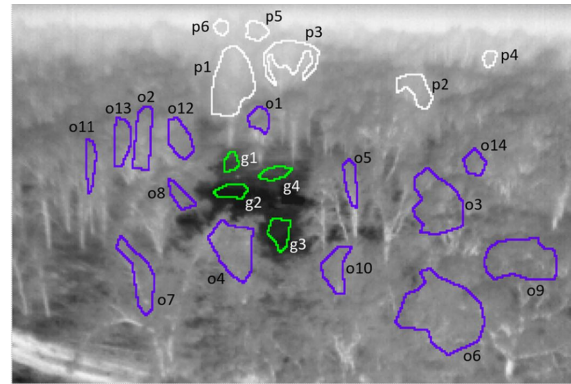


1. TEMPERATURE

Johnston et al., 2021 and 2022

HIGHLIGHTS

- Field-based thermal remote sensing can resolve vertical tree crown temperatures.
- In a woodland savanna, canopy tops are cooler than canopy bottoms at midday.**
- Satellite (ECOSTRESS) and field-based (thermal camera) measurements agree at night.
- During the day, ECOSTRESS and camera temperatures diverge considerably.
- ECOSTRESS/camera mismatch is more related to light than to crown thermal gradients.



1. SPECTRAL RESPONSE

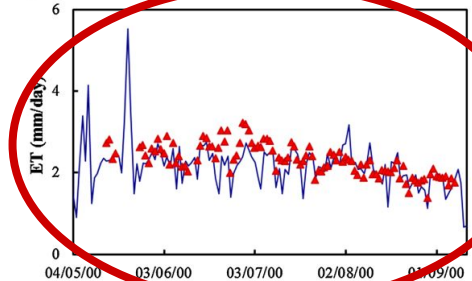
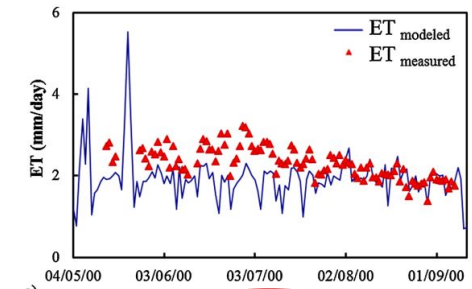
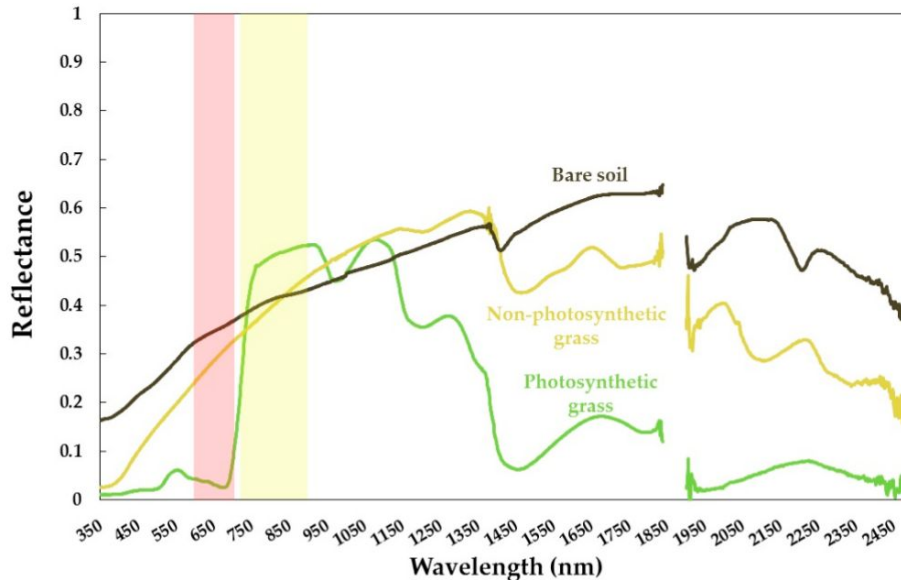
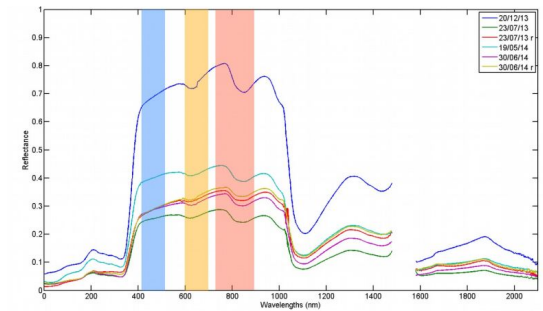
Carpintero 2020a,b, Andreu et al., 2018a



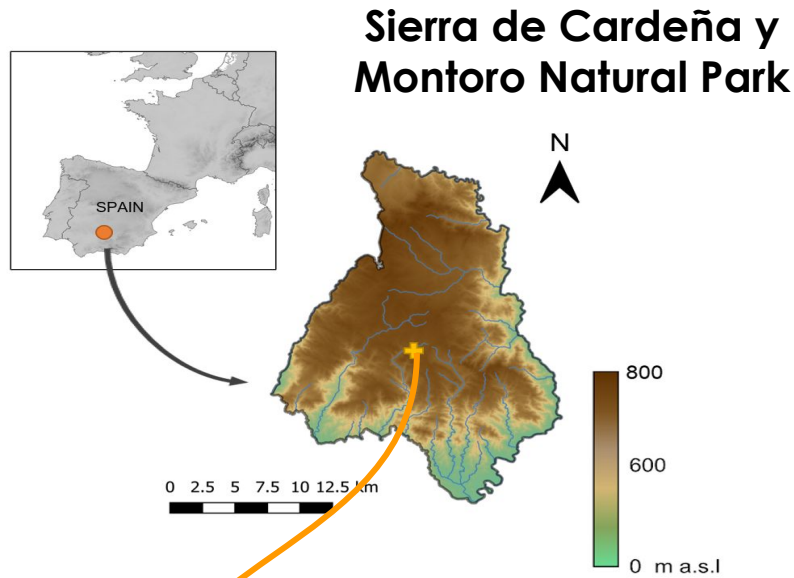
Differential spectral response of oak trees influenced the estimation ET.

Assumption of non-variability of the spectral properties of the holm oak throughout the year for modeling.

Consider the influence of the dead grass in the ecosystem.



WHEN INCLUDING SPECTRAL CHARACTERISTICS



Santa Clotilde experimental site:

- Terrestrial cameras
- Meteorological station



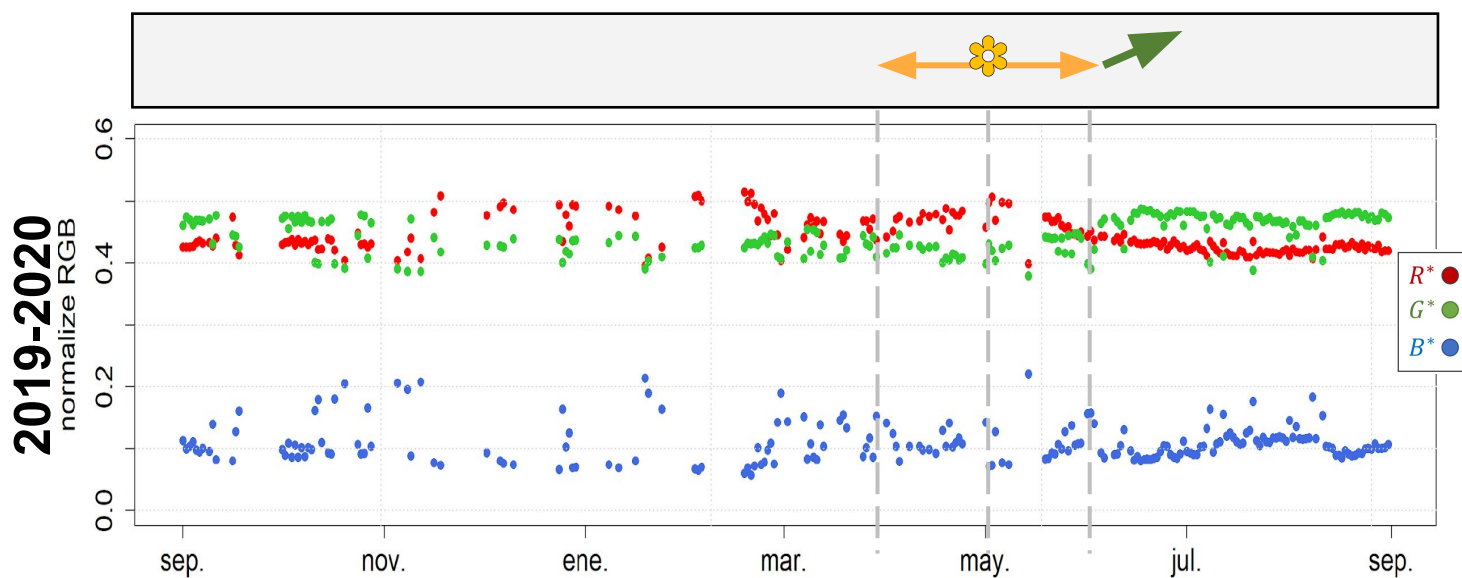
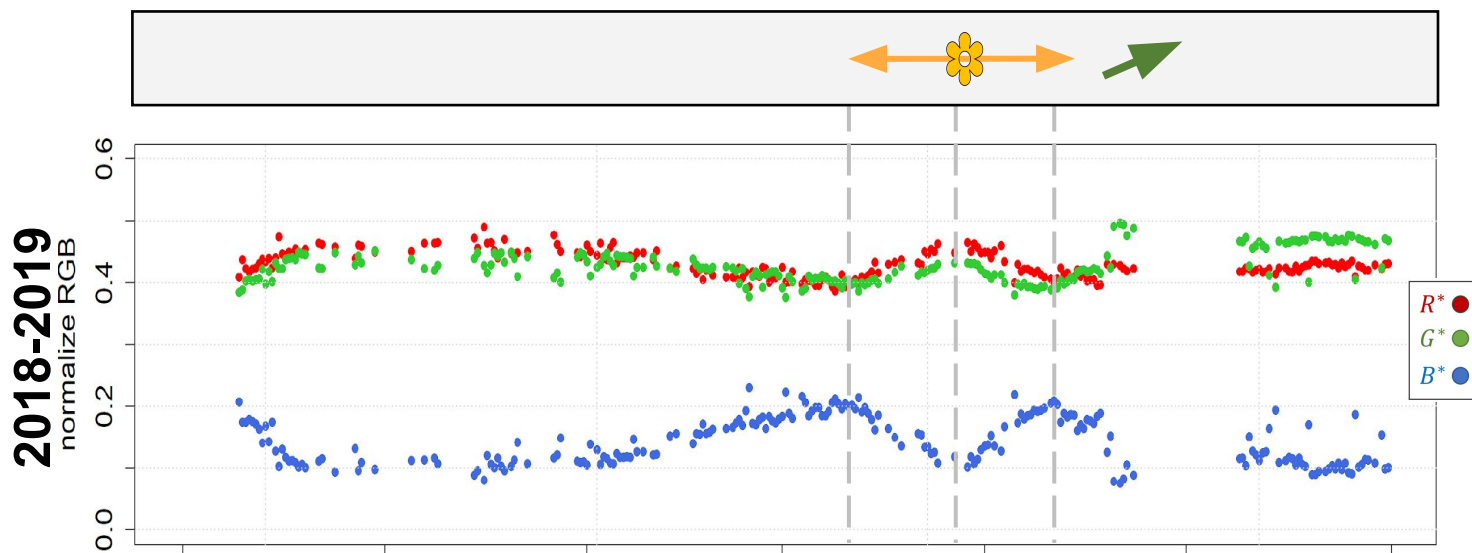
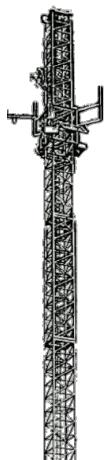
DETAIL SCALE



PLOT SCALE

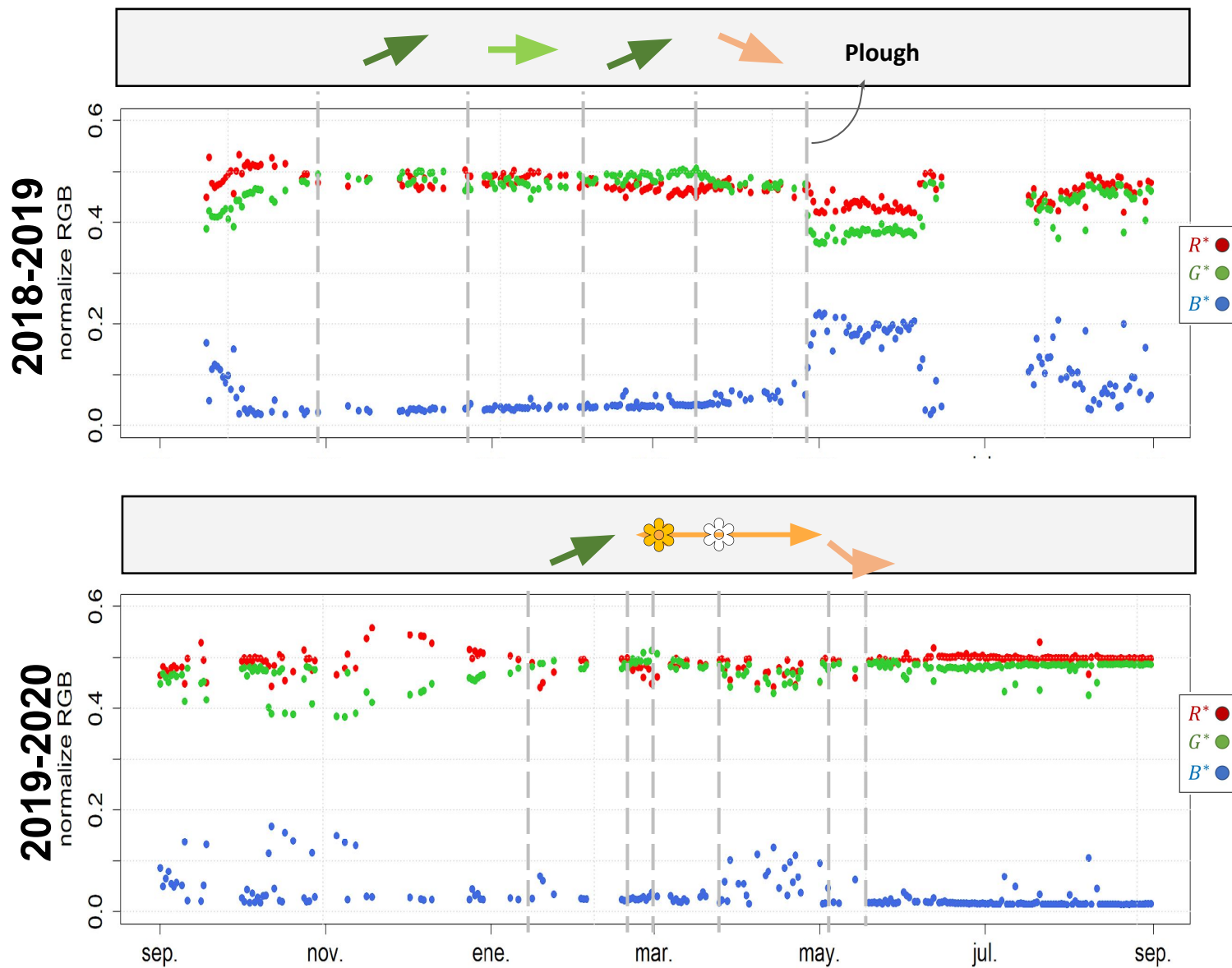
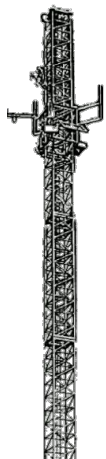
1. PHENOLOGY: HOLM OAK

Pimentel et al., 2021 and 2022



1. PHENOLOGY: PASTURE

Pimentel et al., 2021 and 2022





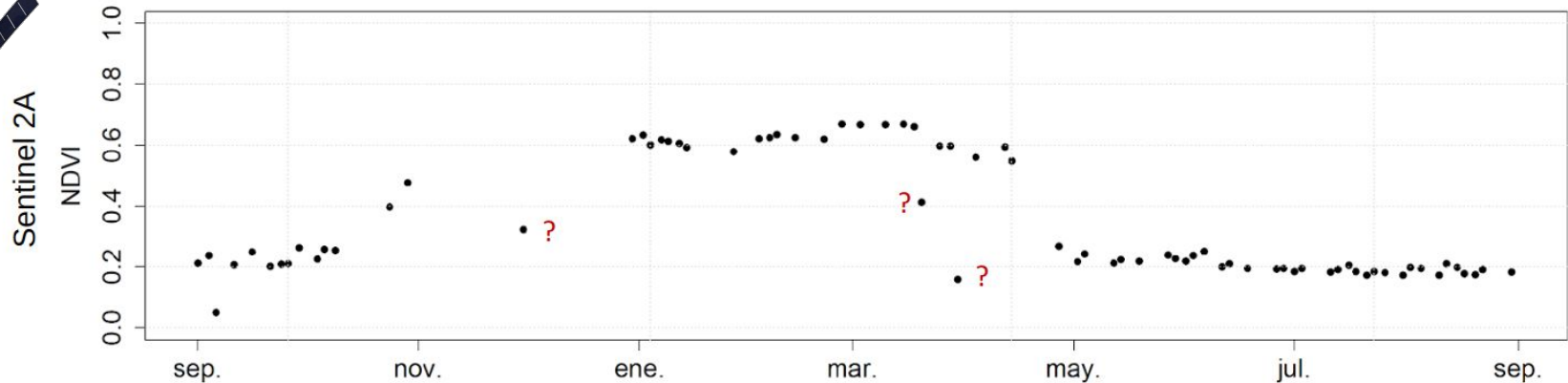
EXPLORING
ECOSYSTEMS'
HEALTH
FROM SPACE

1. PHENOLOGY: OAK & PASTURE

Pimentel et al., 2021 and 2022



2018-2019



Pasture



Holm Oak Tree



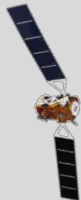
- In general NDVI dynamics are linked to the pasture evolution
- However, autumn' NDVI increase (nov) not directly linked to any change in the pasture but rather a modification in holm tree greenness due to precipitation during this month.
- Holm oak flowering seems not impact NDVI

2. WATER & BIOMASS

WATER

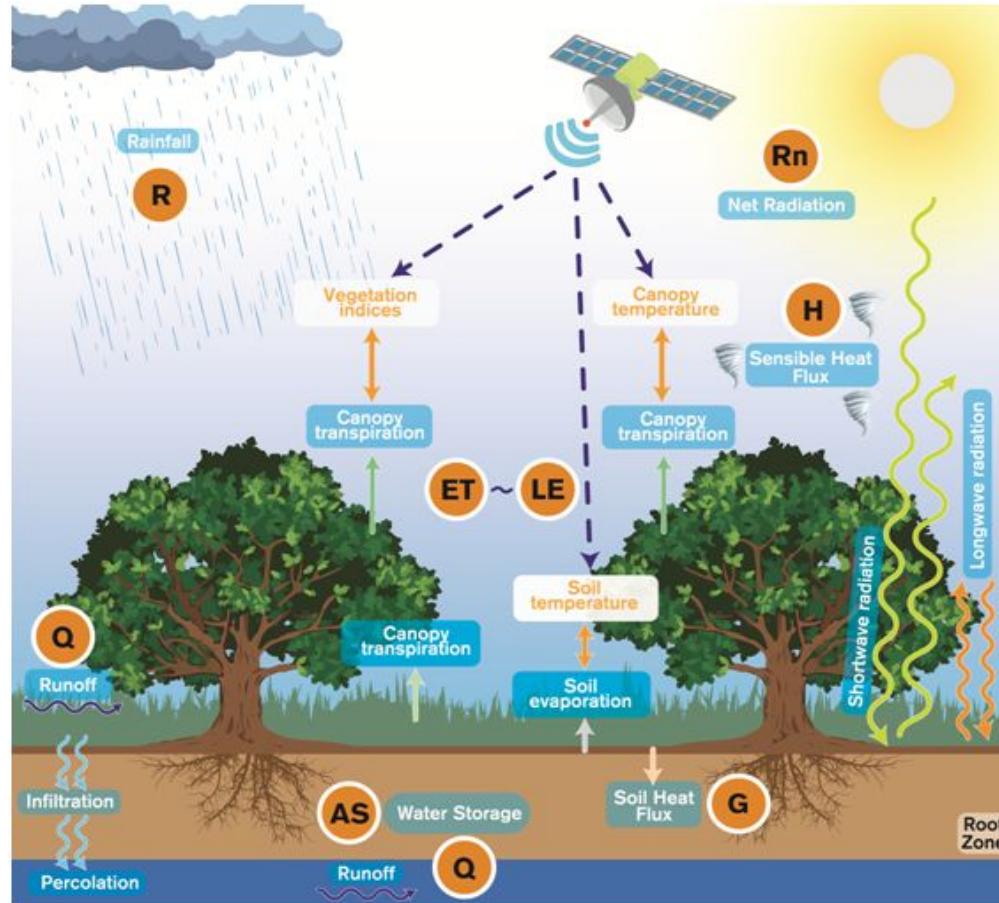
BIOMASS

Base	Surface EB	Soil WB	Light use efficiency
Model	Two-Source EB (TSEB) ²³	WiMed ²⁶	LUE model ²⁷
ED data	Surface thermal data (TIR)	Spectral reflectance (VIS/NIR)	Spectral reflectance (VIS/NIR)
Other	Meteorological data, vegetation characteristics (VIS/NIR)	Meteorological data, soil/vegetation characteristics, precipitation, etc.	Light use efficiency
Why?	-Best accounts for partial canopy cover ^{24, 25} -Strong physical base ⁵	-Distributed hydrological model for Mediterranean watersheds -Strong physical base	-Evaluated and validated in mostly of the world ¹¹ -Strong relation between VI & fraction of radiation absorbed by green canopies



2. WATER: COMPONENTS OF WATER AND ENERGY BUDGETS

Kc-FA056 approach



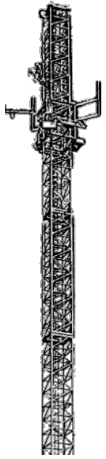
Water Balance
 $R = ET + Q + AS$

Energy Balance
 $Rn = LE + H + G$

TSEB model

2. WATER: ENERGY BALANCE

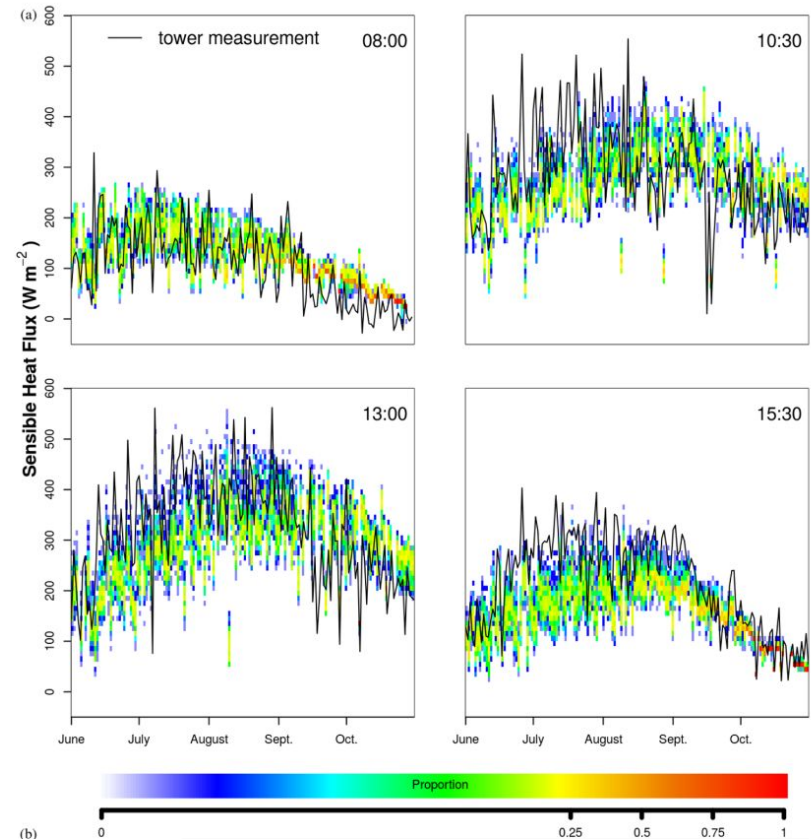
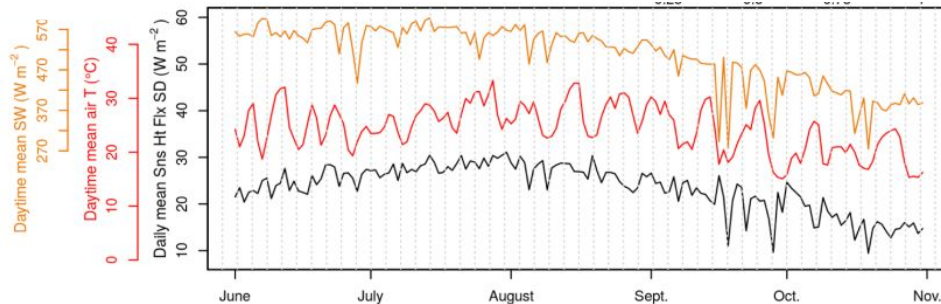
Johnston et al, 2021, Andreu et al., 2018a, Andreu et al., 2015.



Explore thermal model mechanisms and their physical base.

How the parameterization of EB models influences estimates of transpiration and evaporation?

How thermal images can drive EB models using disaggregate temperatures → uncertainty.



2. WATER: ENERGY BALANCE

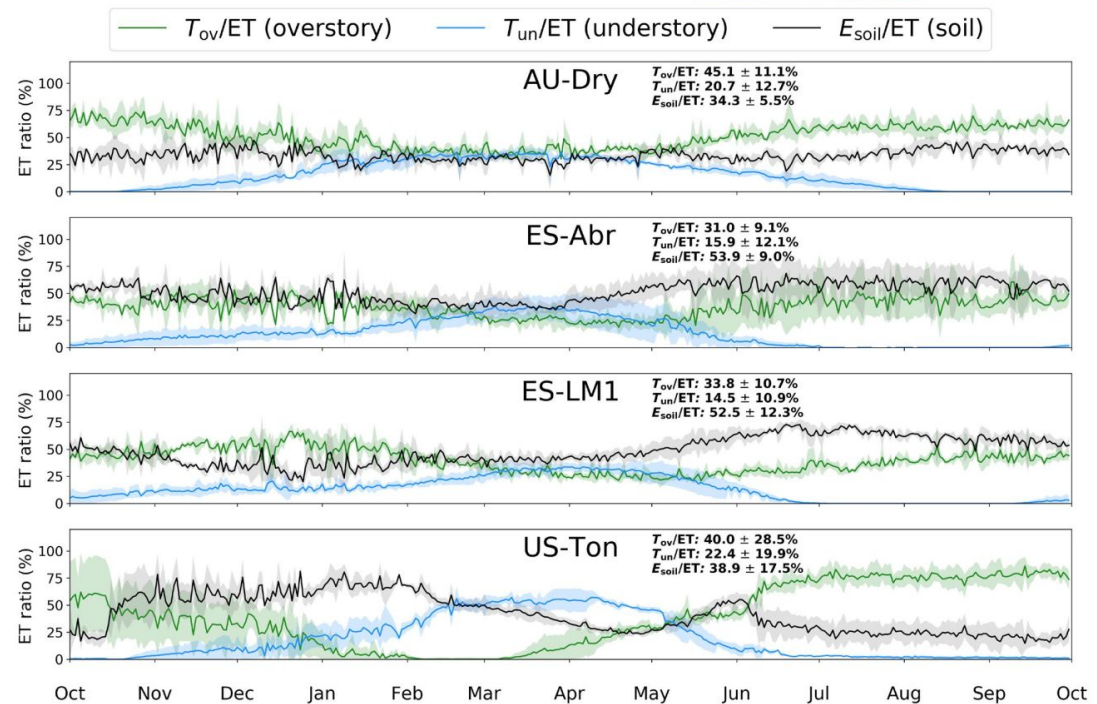
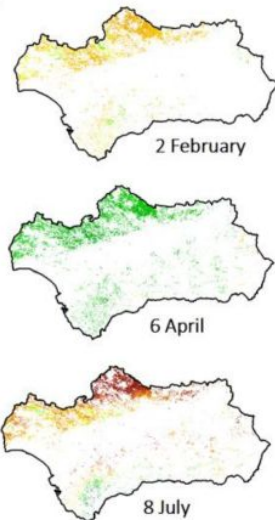
Burchard-Levine 2022, Andreu et al., 2018b.



How the architecture of the vegetation affects the model parameterization?

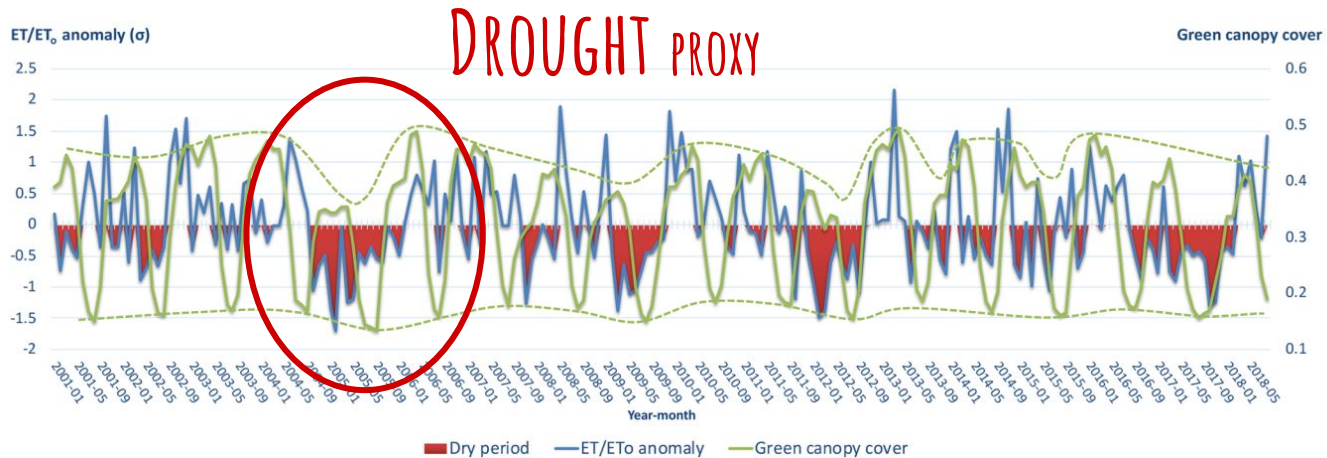
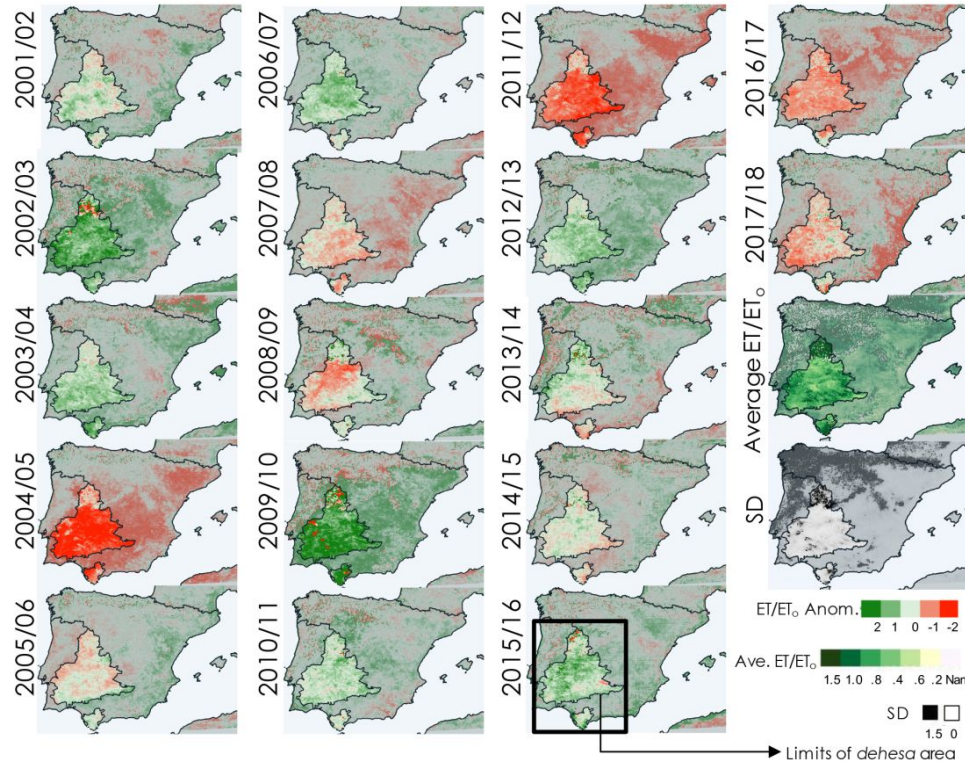
- Monitoring water use in Andalusian dehesa.
- Developing a three-source model (tree, grass, soil) applied and validated in semiarid savannas worldwide.

Latent Heat Flux [Wm^{-2}]



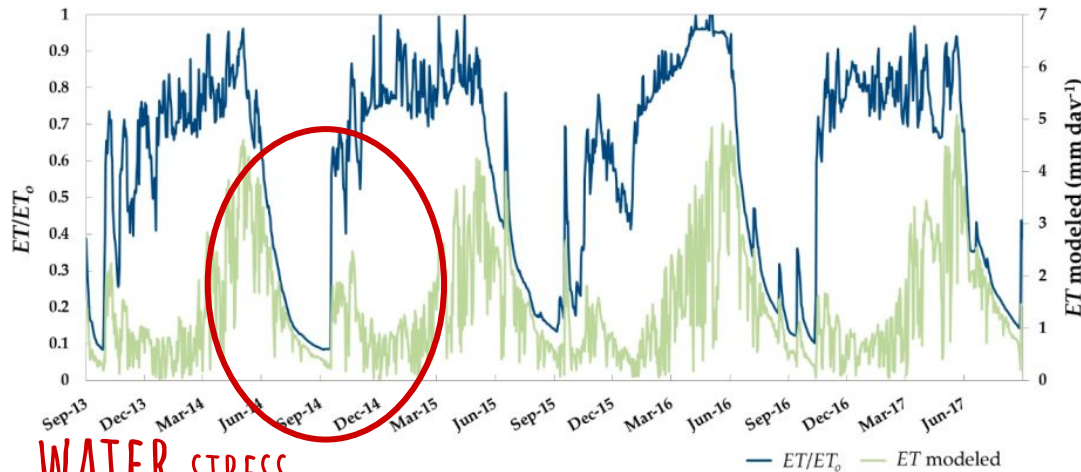
2. WATER: ENERGY BALANCE

Gonzalez-Dugo et al., 2020



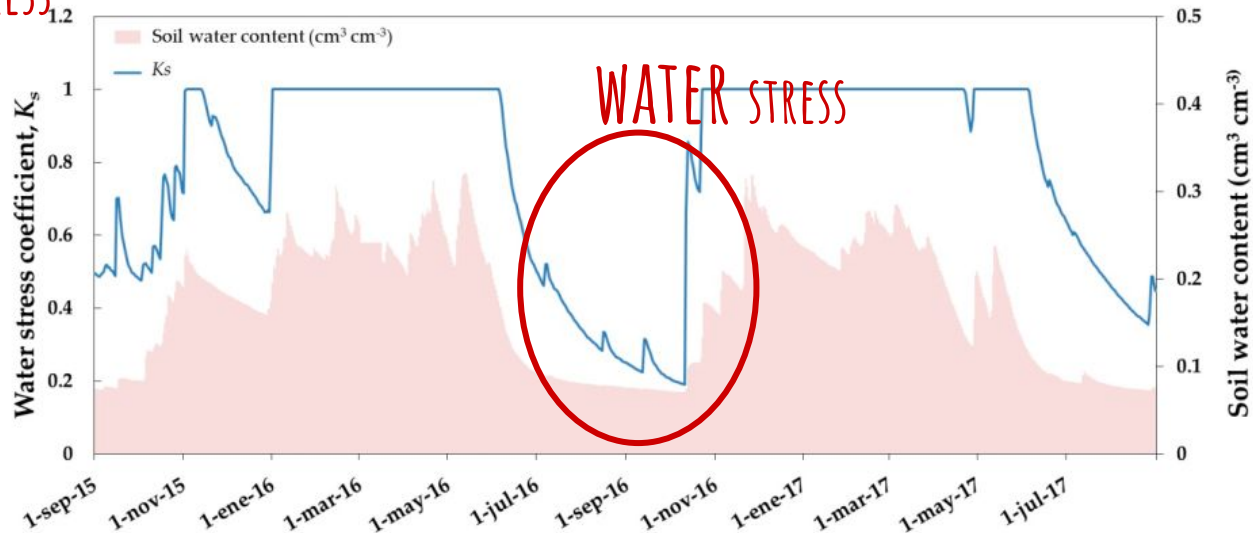
2. WATER: WATER BALANCE

Carpintero et al., 2020



WATER STRESS

The ET/ET₀ ratio helped to identify periods of water stress, confirmed for the grassland by measured soil water content.



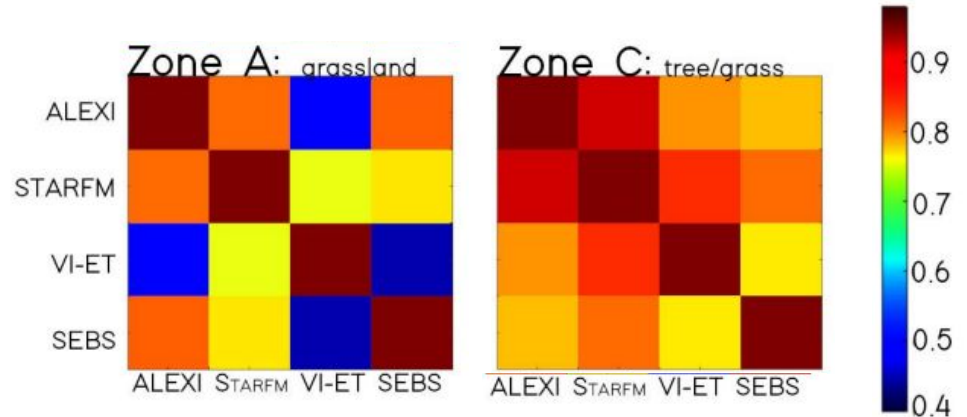
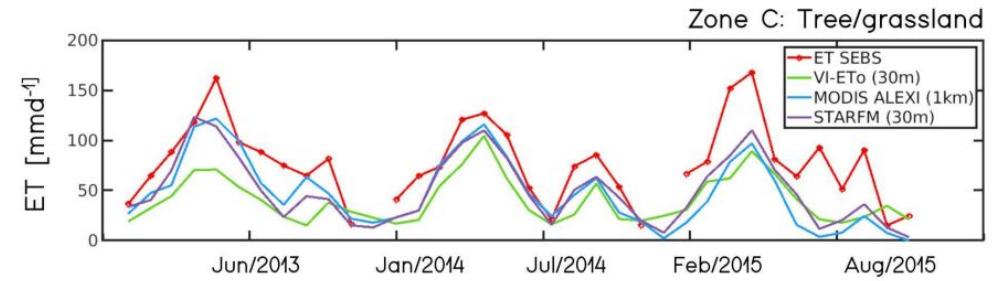
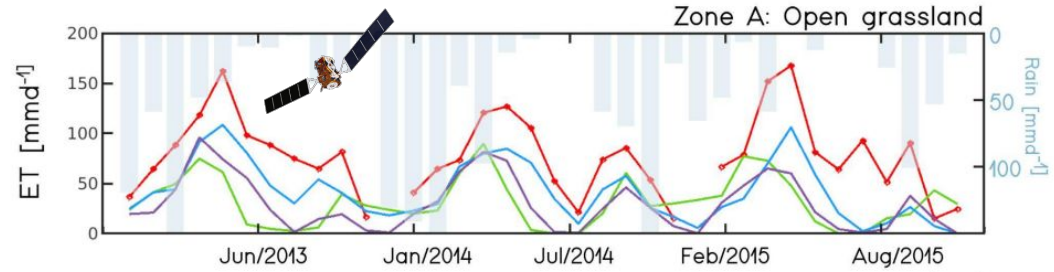
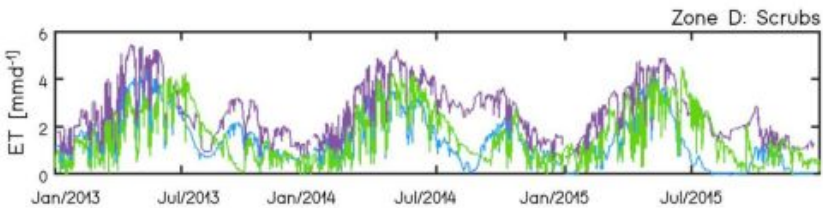
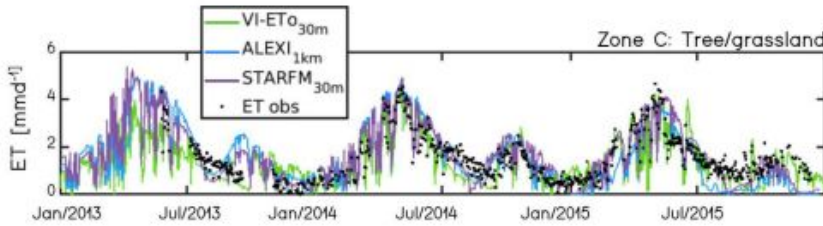
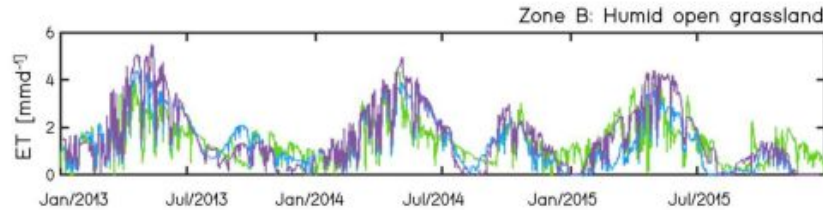
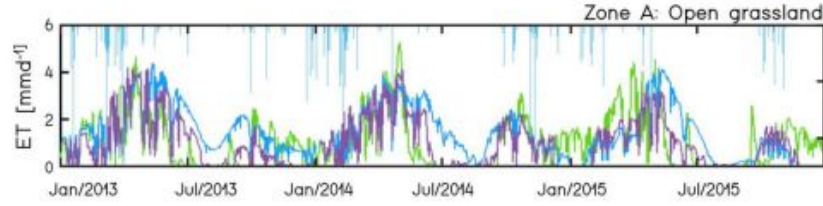
WATER STRESS

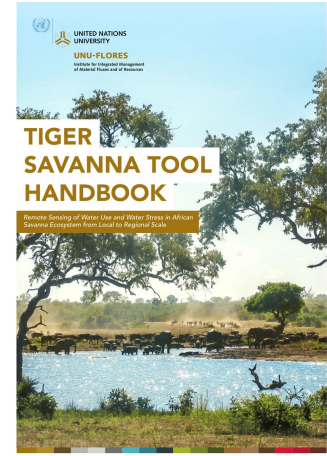


2. WATER > DIFFERENT MODELS AND RESOLUTIONS



Andreu et al., 2023 and 2022



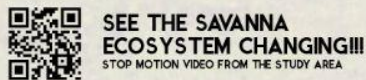
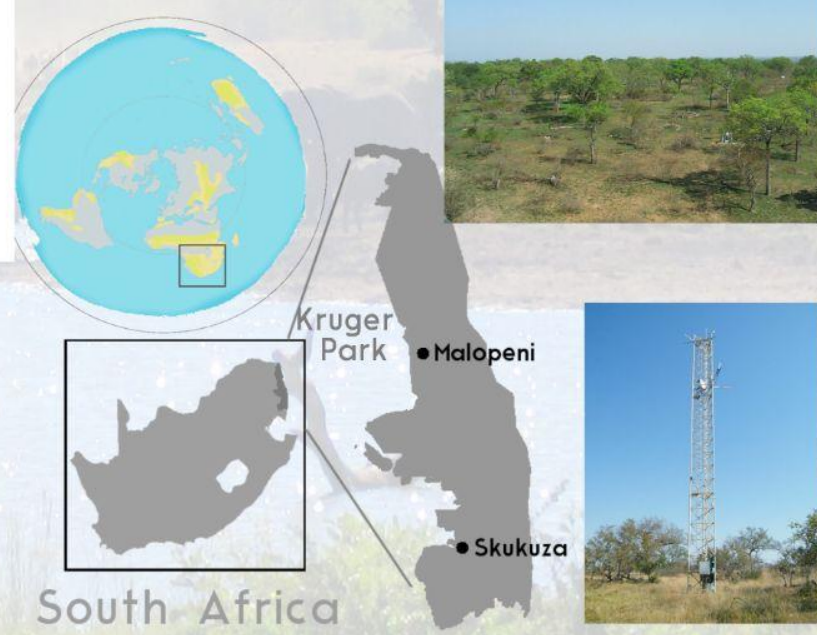


Andreu et al., 2019

MONITORING OF SOUTH AFRICAN SAVANNA'S WATER USE & STRESS USING EARTH OBSERVATION

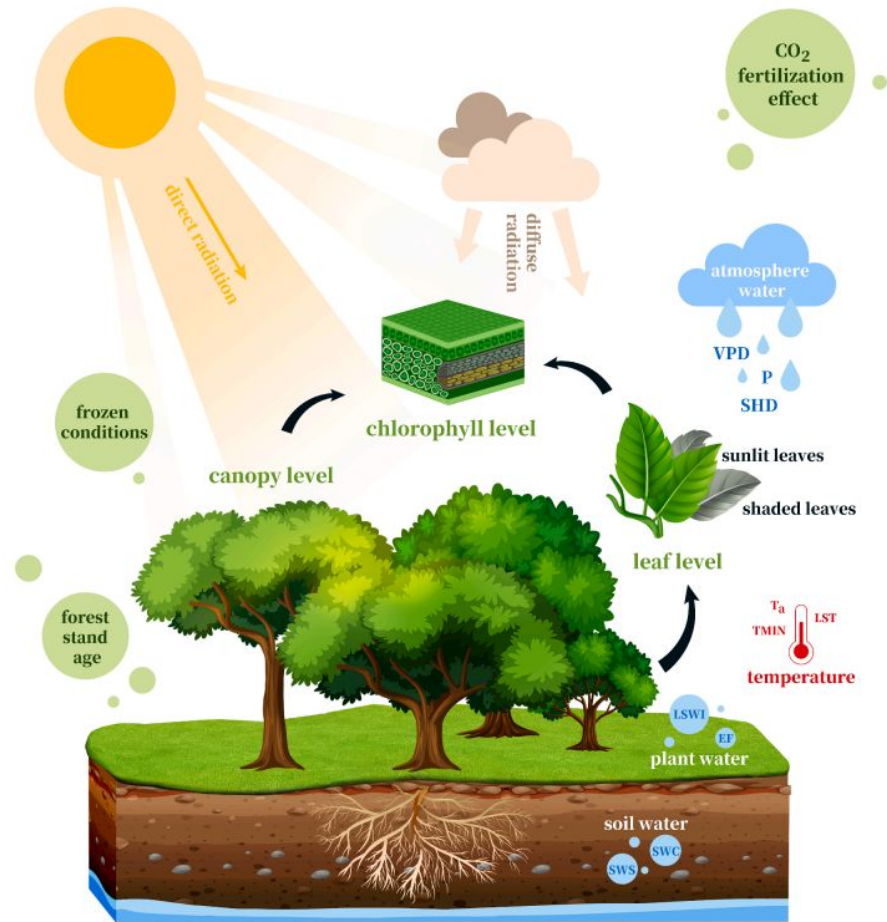
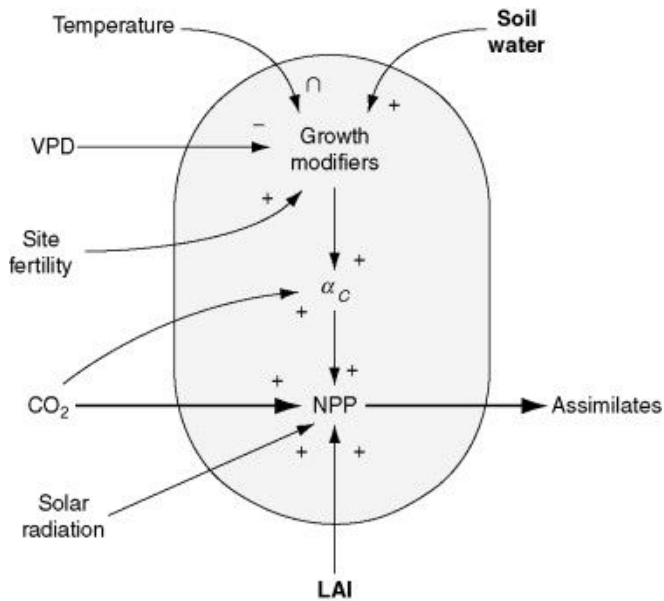
TIGER PROJECT

Savannas are among Africa's most productive multifunctional landscape - supporting wildlife, livestock, crops and livelihoods - but experiencing frequent droughts, aggravated by climate change and other human-induced changes. To maintain ecosystem productivity without reaching the tipping points, while ensuring food security, we should rely on an integrated management and monitoring of resources. The aim of this project is to **map African savannas water use (evapotranspiration-ET) and water stress using Earth Observation data**, to support decision-making at different scales (from local to regional), using as a pilot experience South Africa. The modeling framework was tested during 2010 -2012 with AATSR (thermal data) & SPOT 4/5 (visible and NIR data) satellites and will be next applied with Sentinel 2 & 3 from 2015 to present.



2. VEGETATION PRODUCTION...

$$PNN = \int FPAR PAR LUE DT$$



J. Landsberg and P. Sands, 2011.
<https://doi.org/10.1016/B978-0-12-374460-9.00009-3>

Y. Pei et al. 2022.
<https://doi.org/10.1016/j.agrformet.2022.108905>



UNIVERSIDAD
DE
CÓRDOBA



EXCELENCIA
MARÍA
DE MAEZTU
2020-2023



UNIVERSITY of the
WESTERN CAPE



UNIVERSITY OF THE
WITWATERSRAND,
JOHANNESBURG



UNIVERSITEIT VAN PRETORIA
UNIVERSITY OF PRETORIA
YUNIBESITHI YA PRETORIA



CONSEJO SUPERIOR DE INVESTIGACIONES CIENTÍFICAS



Instituto de
Investigación y
Formación Agraria
y Pesquera



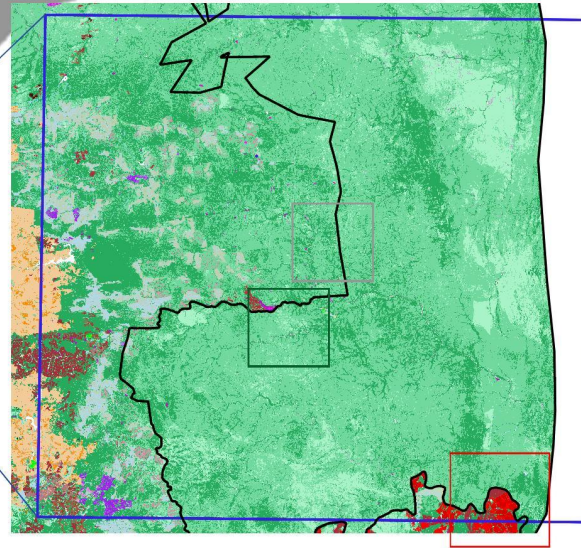
TACTIC: Drought impACT on the vegeTation of South African semlarid mosaiC landscapes. Implications on grass-crop-lands primary production.

2. VEGETATION PRODUCTION... IN THE FACE OF DROUGHT

South Africa

Land Use map for the Sentinel 2 36JUT Tile

Kruger Park



LEGEND

- SAVANNA
- GRASSLAND
- SCRUBS
- IRRIGATED CROPS
- IRRIGATED ORCHARDS
- WOODLAND
- RAINFED CROPS
- WOODLAND
- URBAN
- EC (middle down box)



HUNTING RESERVE AREA - RGB S2

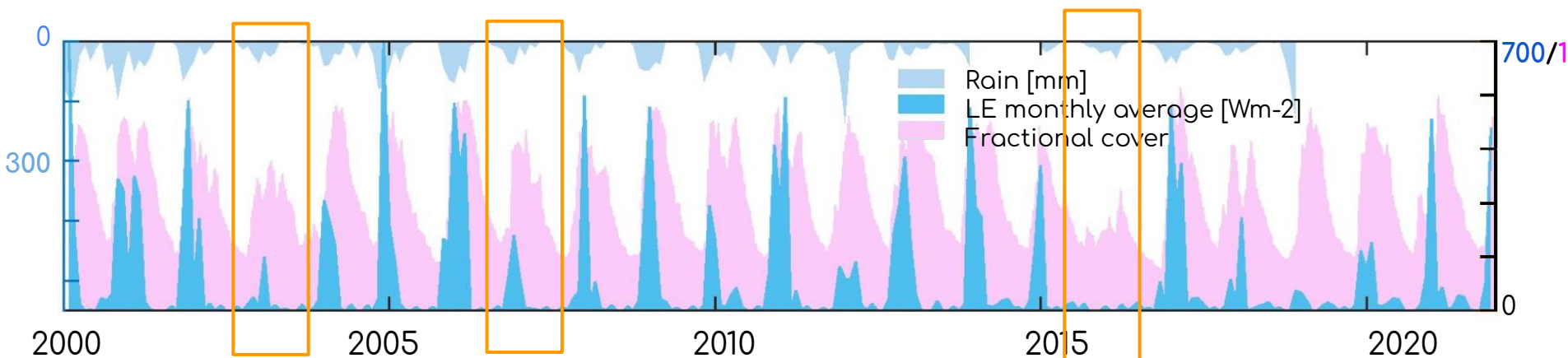
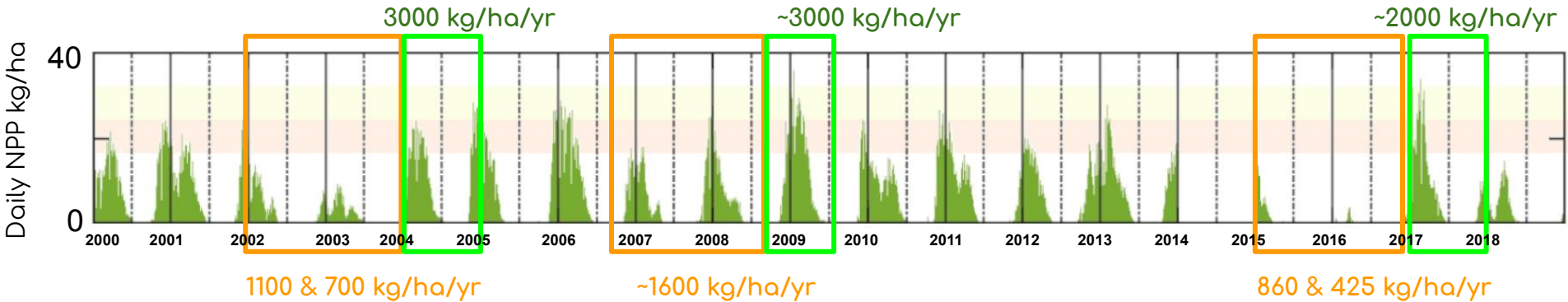
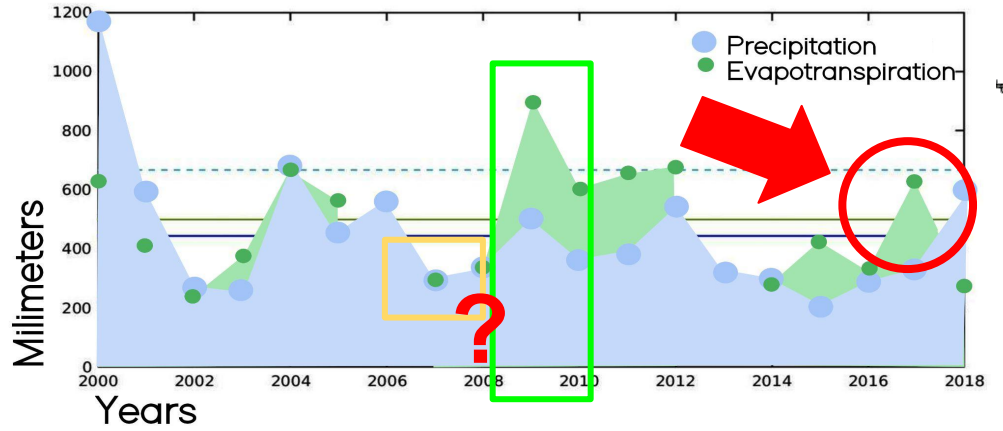
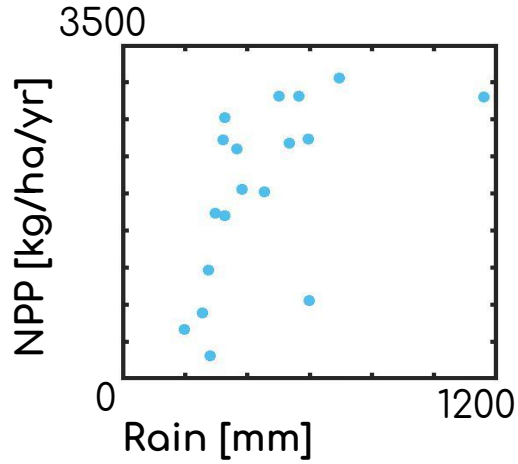


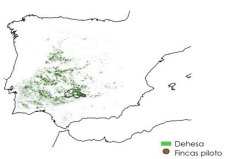
SKUKUZA AREA - RGB S2



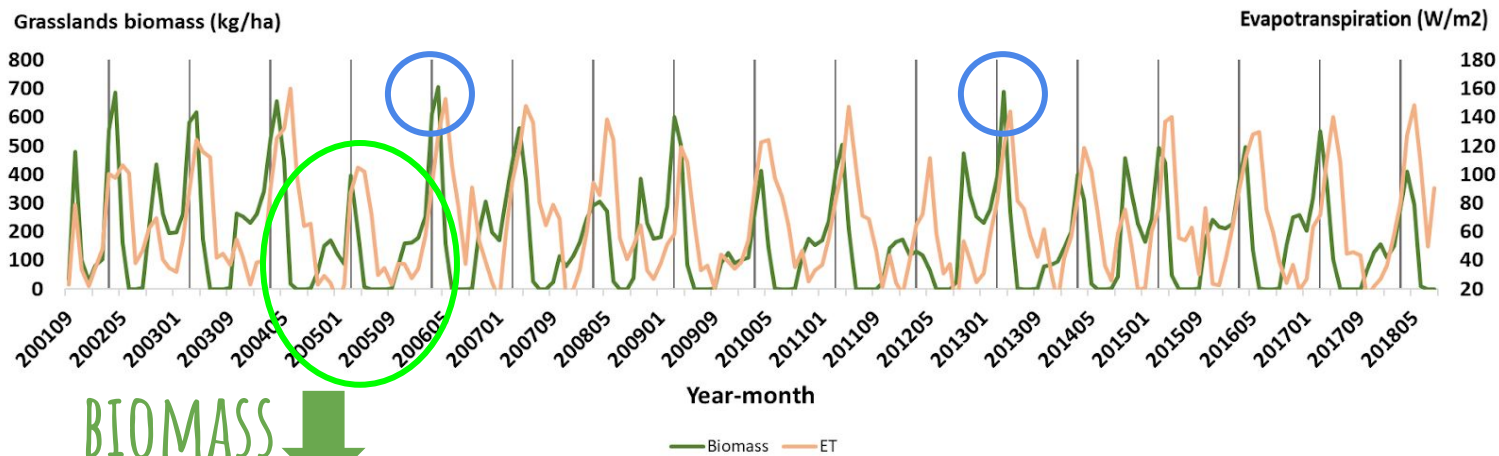
CROP AREA - RGB S2

Skukuza savanna

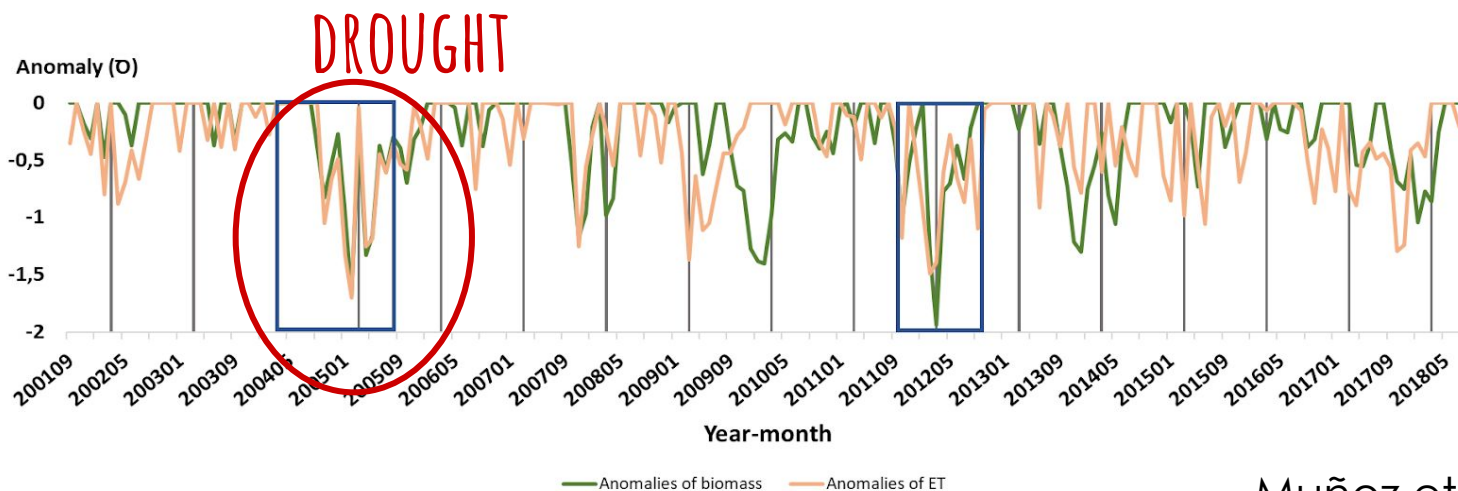




2. VEGETATION PRODUCTION...IN THE FACE OF DROUGHT

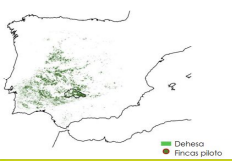


BIOMASS ↓



Muñoz et al., 2022



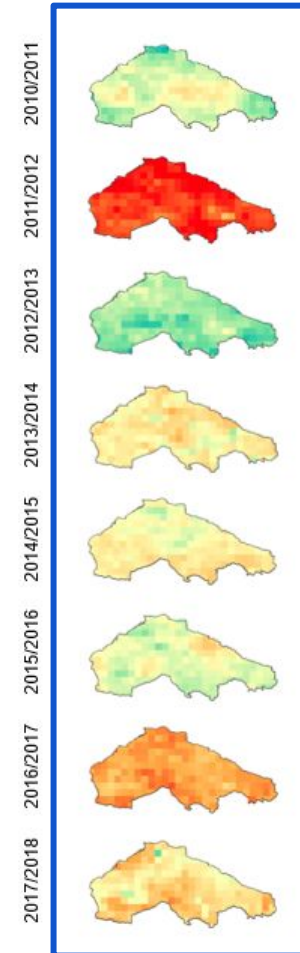
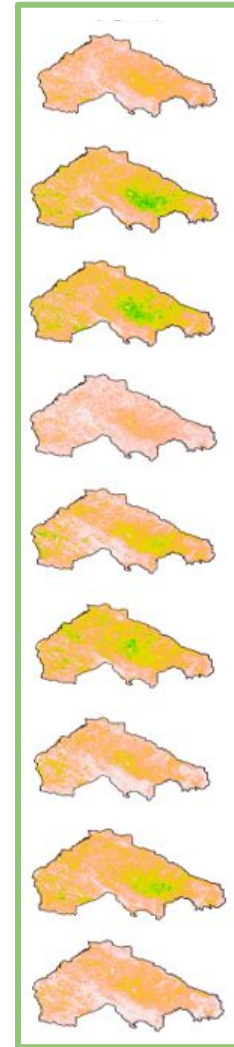
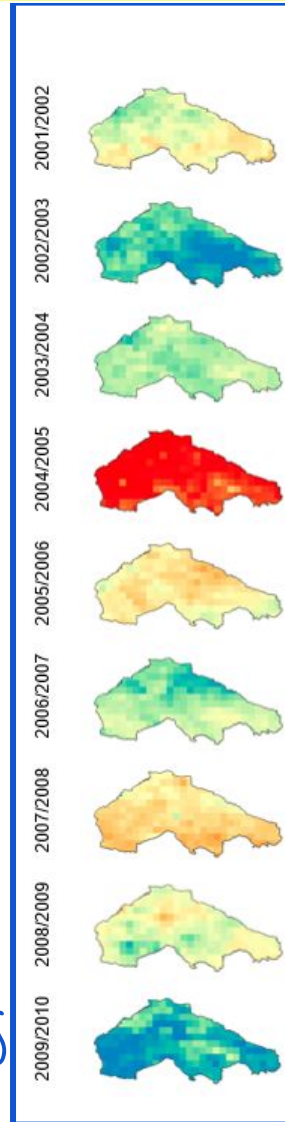


2. VEGETATION PRODUCTION...IN THE FACE OF DROUGHT

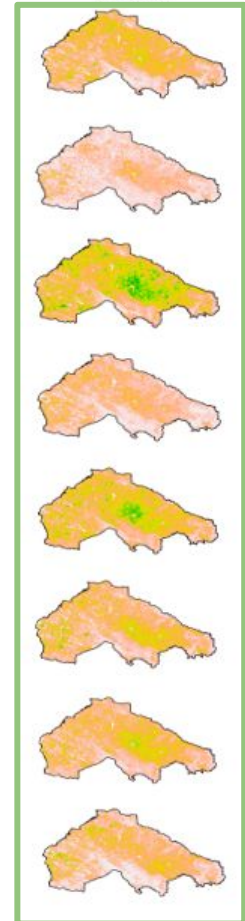


The Unseen Attack of Drought
by Alwyn Biju

ET ANOMALIES



BIOMASS



Muñoz et al., 2022

Thanks a lot!!

