

### Water Use With Cloud Computing: **Prospective for Food and Water Security in Smallholder Irrigation Schemes**

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#### Introduction

Agriculture is experiencing a global water shortage.

Food security can be improved by understanding how irrigated lands use water.

Thriving evapotranspiration models incorporating remote sensing data -> new unprecedented spatial and temporal resolutions.

#### **Objective and Study Area**

#### **Spatiotemporal Trends in Crop Water Use**



- To investigate long-term irrigated crop water consumption in two smallholder irrigation schemes in a west African semi-arid area (Burkina Faso)
- Accomplished by concentrating on the dry season (October to May) from 2000 to 2020, i.e., croplands are irrigated areas.
- Surroundings areas with significant land cover changes in the last 20 years (Figure 1)

#### Irrigation system: Lake Bam -> unmanaged; Bagré reservoir -> managed

#### Methods

Entire study done in the cloud-computing environment Google Earth Engine (Gorelick et al., 2017).

#### (1) Image classification

croplands are distinguished from other land uses categories using random forest image classification.

#### (2) Evapotranspiration mapping

The indicator "water use"  $\rightarrow$  estimated amount of water consumed per irrigated hectare (m<sup>3</sup>ha<sup>-1</sup>).

derived from an automated tool implemented in the Google Earth Engine platform named geeSEBAL (Laipelt et al., 2021).

Figure 2: Actual evapotranspiration per year (mm year<sup>-1</sup>) during the dry season (October to May), averaged over 5-year intervals from 2000 to 2020. 30 m is the resolution.



• based on the Surface Energy Balance Algorithm for Land (SEBAL) and uses actual evapotranspiration from a surface energy balance model (Bastiaanssen et al., 1998).

#### Actual evapotranspiration (mm) x Irrigated surface (ha) = Water use ( $m^{3}ha^{-1}$ )

Daily information from 233 Landsat images (Landsat 5, 7 and 8- Bagré: 115) and Bam: 118) was used to convert monthly to seasonal actual evapotranspiration using ERA5 Land as meteorological inputs.

# Results **Image Classification** Lake Bam Bagré reservoir\_ Bam Burkina Faso Lake

# Figure 4: Change in crop water use (%) relative to the previous period over a 5-year interval from 2000 to 2020. 30 m is the resolution.

#### Table 1: Summary statistics in crop water use

Area	Rate of change (m <sup>3</sup> year <sup>-1</sup> )	Relative change (%)	Irrigated lands expansion	
			2000 (km²)	2020 (%)
Bam	0.14*	-6	9.8	+1.4
Bagré	0.45	+55	4.3	-24
* a statistically significant				

creasing

not statistically significant

#### References

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Figure 1: Land use and land cover classification in 2000 and 2020 using random forest image classification with a spatial resolution of 30 m.

#### Conclusion

#### Google Earth Engine

2006-2010

2001-2005

Change in crop water use (%)

• Significant potential for long-term assessment of crop water use

2016-2020

Lake Bam

2011-2015

Bagré reservoir

- More timely and cost-effective information updates for water management in areas where data is scarce and informal.
- Managed Bagré reservoir: increasing crop water use and shrinking water demand.
- Unmanaged Lake Bam: decreasing crop water use and growing water demand.
- Future with the increasingly limited water resources: investigate crop water productivity
- improve food security and
- better water management strategies