



RUB

DETECTION AND ASSESSMENT OF AGROFORESTRY SYSTEMS (AFS) AND MONOCULTURES IN NORTHERN BATKEN (KYRGYZSTAN)



Fig. 1: Agricultural land in the northern Batken region (Schierning 2024)

FLORIAN DREBES (MASTER'S THESIS)
LATINET ENVIRONMENTAL CONFERENCE 2025

Climate Change and Water Scarcity in Agriculture

- Rising temperatures, changes in precipitation, extreme weather events
- Increasing water scarcity → Threatening agriculture
- Affecting particularly (semi-)arid regions such as Kyrgyzstan (Abella & Ahn 2024)

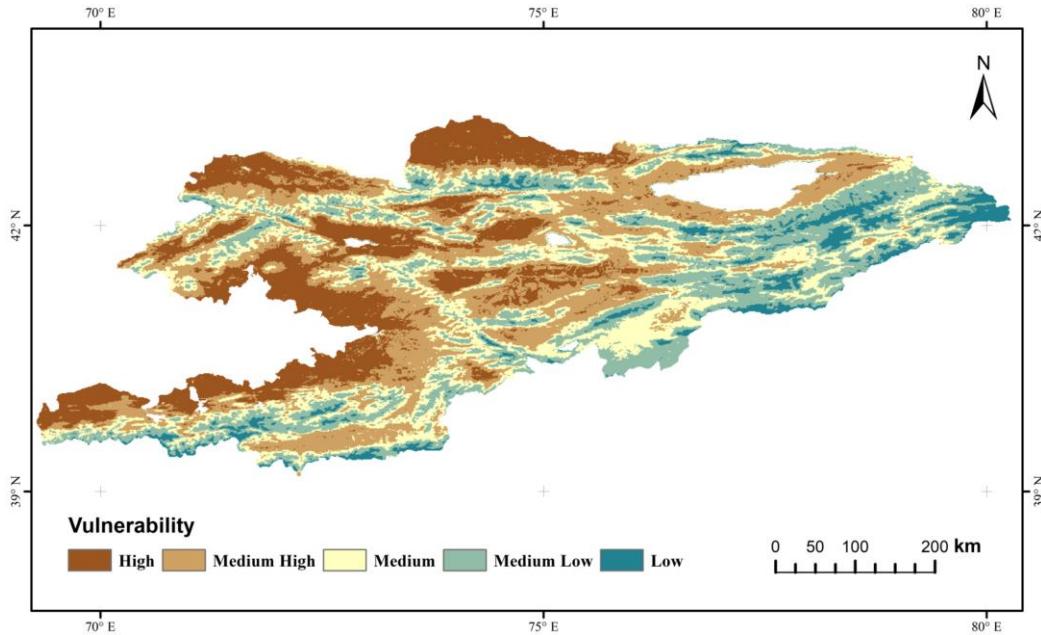
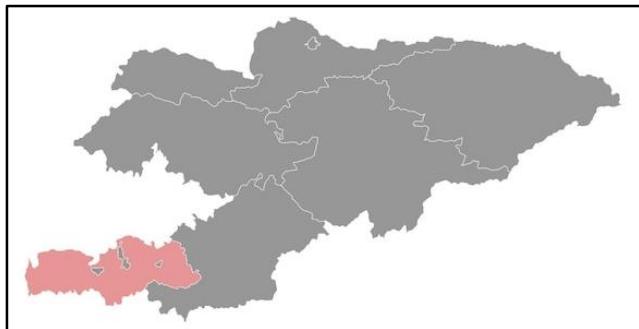


Fig. 2: Map of the drought exposure of Kyrgyzstan's agricultural systems (Liang et al. 2021)

Apricot Cultivation in Batken

Batken region

- Apricot cultivation = high importance
- Water scarcity and inefficient irrigation systems
- Dependence on the Isfara river and glacial meltwater (Djanibekov et al. 2016)

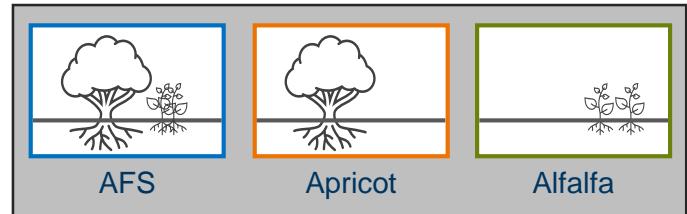


Alfalfa
ment, soil health,

Fig. 4: Map of Kyrgyzstan with Batken region highlighted



Fig. 3: Dead apricot tree in Batken (Schierning 2024)



Research Question

Lack of knowledge about the distribution and potential benefits of AFS in Batken

*Can AFS and monocultures
(apricots and alfalfa)
in the northern Batken region
be accurately distinguished and
classified using remote sensing data?*

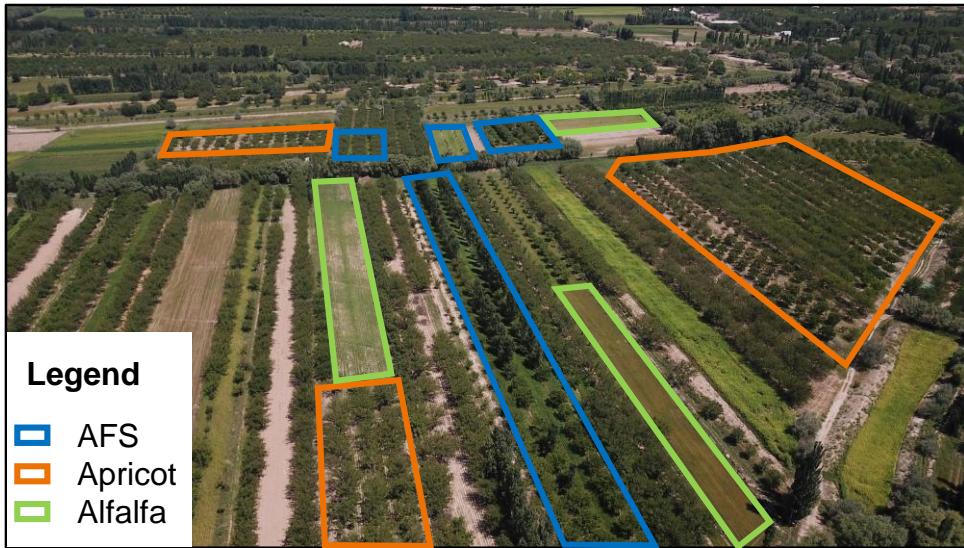


Fig. 5: Drone image with examples of the three land use categories (Schierning 2024)

Study Area

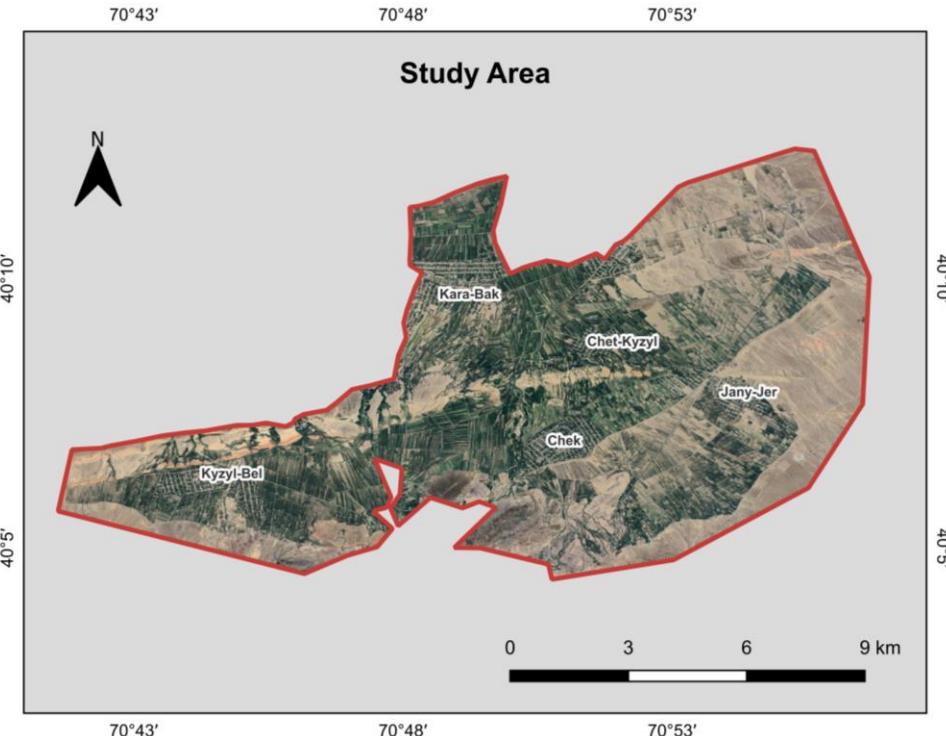
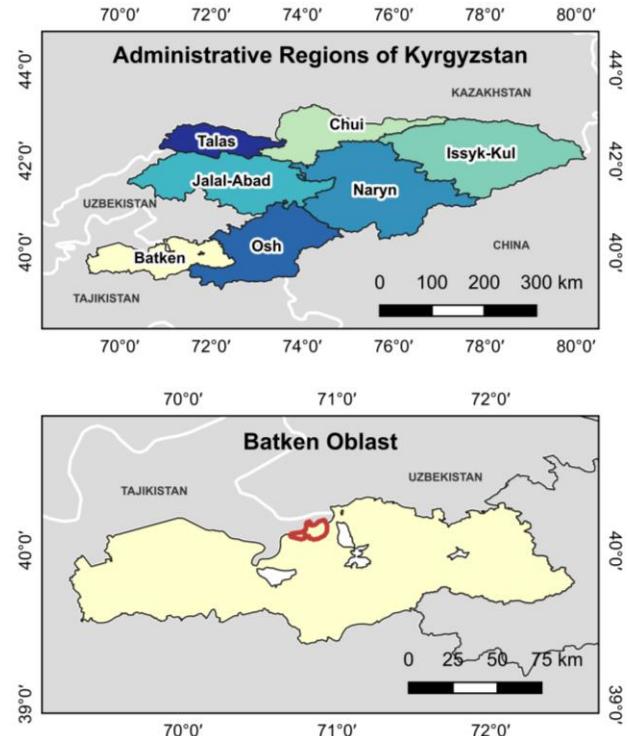


Fig. 6: Maps of Kyrgyzstan, Batken Oblast and the study area

Segmentation using Segment Anything Model (SAM)



- No available vector data of agricultural areas
- SAM as an efficient solution for automatic segmentation by Meta AI
- Segments new images and recognizes unseen objects without additional training data
(Wu & Osco 2023)

Fig. 7: Example of results using the model (Meta 2023)

Segmentation using Segment Anything Model (SAM)

Input Google Satellite



Segmentation Result

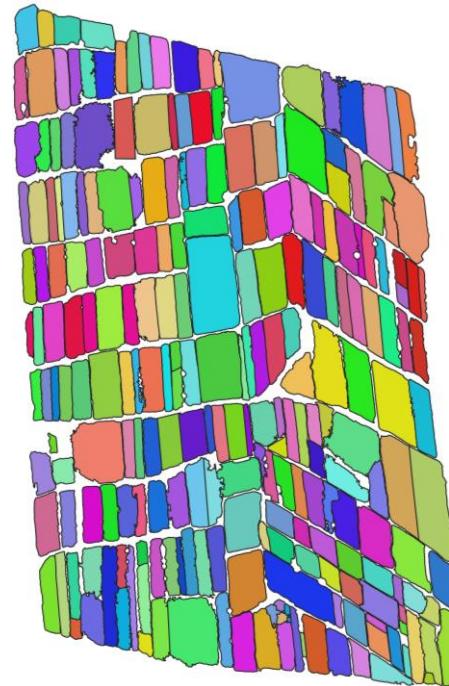


Fig. 8: Segmentation result using SAM for an example area

Segmentation Results

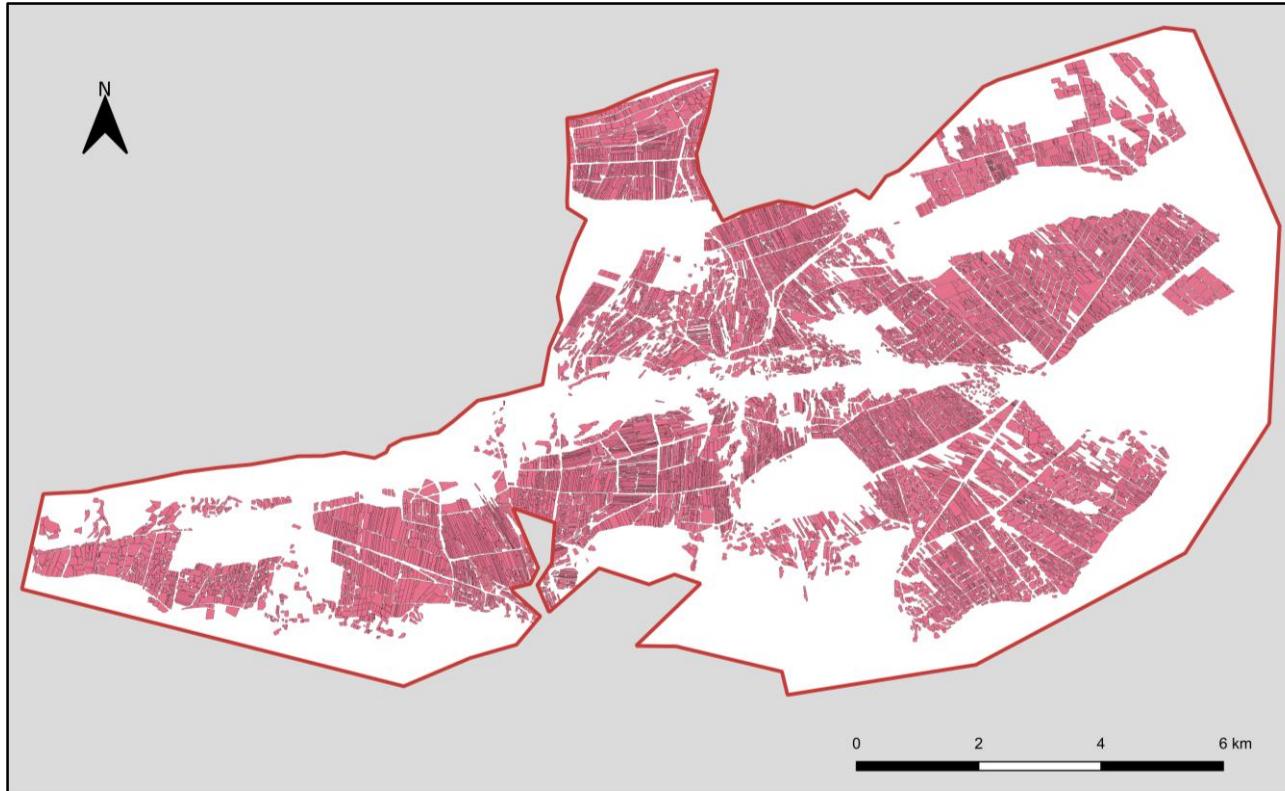


Fig. 9: Segmentation of all potential agricultural land in the study area

Excursus: Remote Sensing

- Satellites or aerial sensors gather data about the Earth's surface without physical contact
- Different materials reflect light uniquely
- Used to monitor vegetation, water, land use and environmental changes (USGS 2024)

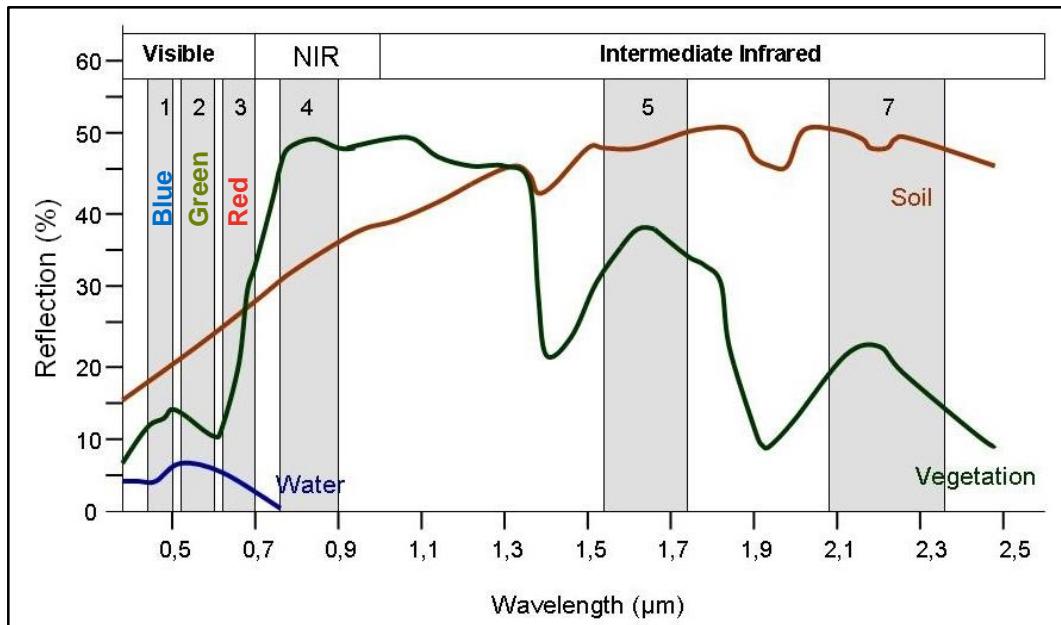


Fig. 10: Spectral signatures of soil, vegetation and water (Siegmund & Menz 2005)

Excursus: Remote Sensing Indices

- Spectral indices combine different bands to highlight specific surface features
- NDVI (Normalized Difference Vegetation Index) measures vegetation health by comparing red and near-infrared
- High values = dense, healthy vegetation
- Low values = bare soil or stressed plants

(Shurlaeva et al. 2021)

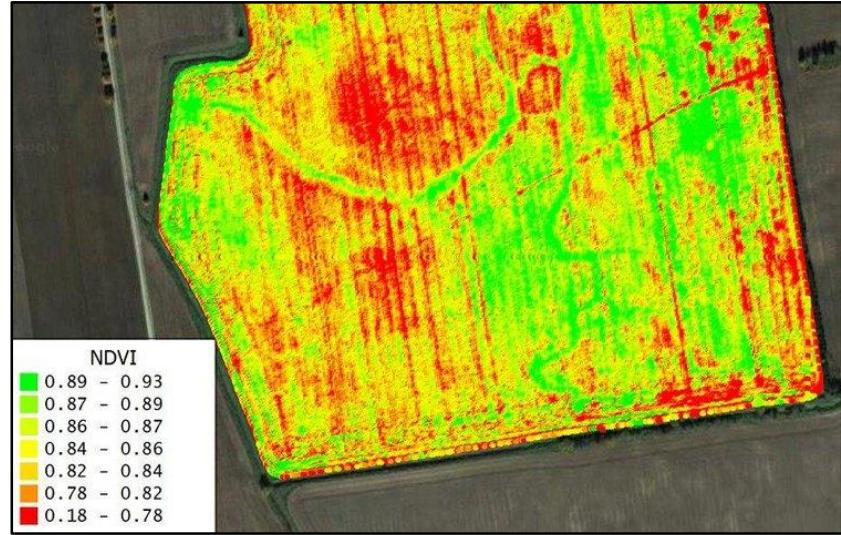


Fig. 11: An example of NDVI map visualization (Shurlaeva et al. 2021)

Model Training with Time Series of Spectral Indices

Data basis for model training

- 25 Sentinel-2 images (2024)
- 66 AFS, 88 apricot, 59 alfalfa plots

Spectral indices

- Normalized Difference Vegetation Index – **NDVI**
- Enhanced Vegetation Index – **EVI**
- Normalized Difference Water Index – **NDWI**
- Bare Soil Index – **BSI**

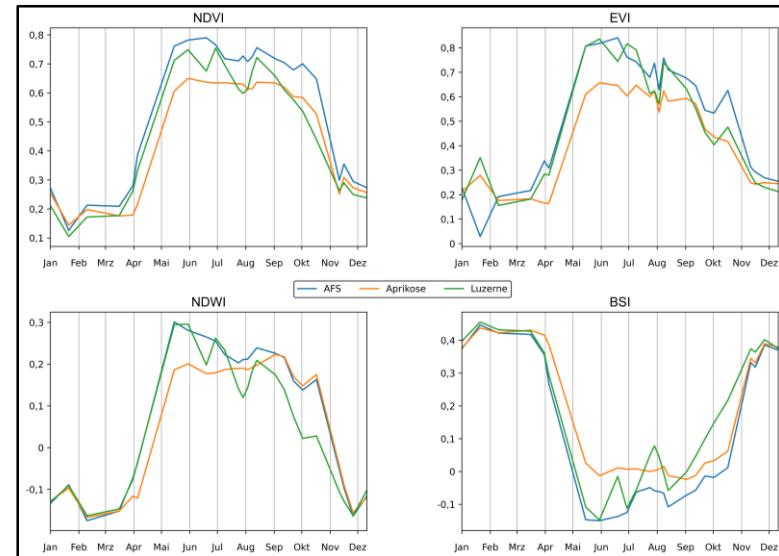


Fig. 12: Time series of average indices per category (2024)

Model accuracy of 80 % with Std Dev of 10.4 %

Model Results

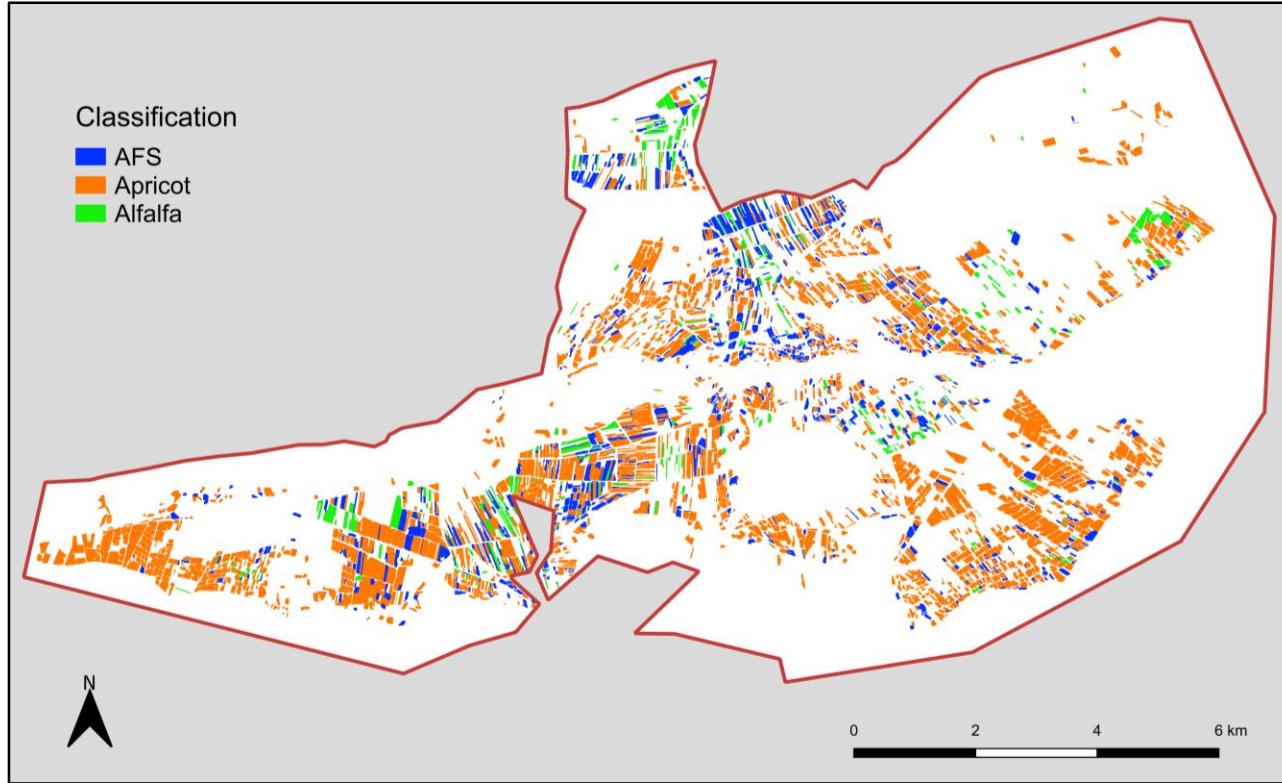


Fig. 13: Identified AFS, apricot orchards and alfalfa fields in the study area

Model Results

Tab. 1: Classification results for agricultural land

Category	Identified Plots	Ø Size per Plot	Total Size	Share
AFS	1234	0.43 ha	527.5 ha	9.3 %
Apricot	3417	0.58 ha	1984.4 ha	34.8 %
Alfalfa	547	0.43 ha	233.5 ha	4.1 %
Other Agricultural Land	6554	0.45 ha	2954.5 ha	51.8 %
Total	11752	0.49 ha	5699.9 ha	100 %

Conclusions

Overall model performance was highly satisfactory, with clear differentiation between AFS and monocultures. The results demonstrate the potential for using index-based classifications in the Batken region.

Outlook

- Analysis of water availability and drought stress across the different land use categories
- Validation of the classification results through field work
- Derivation of practical recommendations for agricultural practice and possible future applications of remote sensing techniques in similar regions

An aerial photograph of a vast, green agricultural landscape. The foreground is filled with numerous rectangular fields, likely orchards or vineyards, separated by dirt roads. In the middle ground, a dense forest of tall, thin trees stretches across the horizon. The background features a range of majestic, snow-capped mountains under a clear blue sky.

THANKS!

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Any Slides for Backup

Possible Advantages of Agroforestry Systems

- Improvement of water management through:
 - Increased water infiltration → Root systems penetrate soils deeper → Water absorption
 - Less water loss through less surface runoff
 - Trees reduce soil temperature through shading effect → reduces evaporation
- Promotion of biodiversity through different microhabitats
- Improved soil health through organic material and nitrogen-fixing through trees
- Reduction of income risk through diversification of income

Apricot Cultivation in Batken



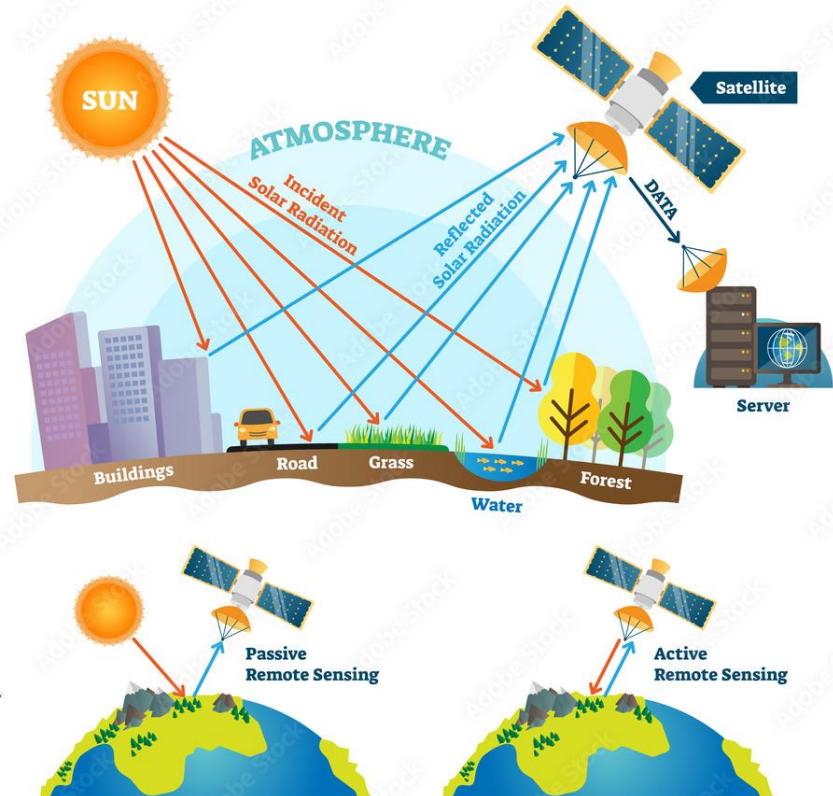
Fig. XX: Apricot monoculture (Schierning 2024)



Fig. XX: Apricot AFS (Oetker 2023)

Remote Sensing

REMOTE SENSING



Sentinel-2 Band Setting

Band	Spatial Resolution	Central Wavelength	Description
B2	10 m	490 nm	Blue
B3	10 m	560 nm	Green
B4	10 m	665 nm	Red
B8	10 m	842 nm	NIR
B11	20 m	1610 nm	SWIR1

Spectral Indices utilized in this Study

Index	Equation	Description	Reference
Normalized Difference Vegetation Index (NDVI)	$\frac{NIR - Red}{NIR + Red}$	Measures vegetation health and density	Rouse et al. 1974
Enhanced Vegetation Index (EVI)	$2,5 * \frac{NIR - Red}{NIR + 6 * Red - 7,5 * Blue + 1}$	Adaptation of the NDVI for dense vegetation by reducing atmospheric disturbances and ground reflection	Huete et al. 2002
Normalized Difference Water Index (NDWI)	$\frac{NIR - SWIR1}{NIR + SWIR1}$	Records the liquid water content in the vegetation	Gao 1996
Bare Soil Index (BSI)	$\frac{SWIR1 + Red - NIR + Blue}{SWIR1 + Red + NIR + Blue}$	Estimates the proportion of uncovered ground	Chen et al. 2004