

**10th  
WORLD  
AVOCADO  
CONGRESS  
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2023**



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# Predicting irregular bearing of avocado crops using Sentinel 2 time series data

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# AARSC

Applied Agricultural  
Remote Sensing Centre



# Outline

- What is Irregular Bearing
- The research has been done until now
- Research questions
- Methodology
- Results
- Conclusion and Future work

# Background Information

- What is Irregular or Alternate Bearing???
- What causes Irregular Bearing??
  - Extreme weather events
  - Water deficit stress
  - Diseases and pests
  - Excessive fruit set
  - Excessive fruit retention



# Background Information

- Why understanding of Irregular bearing is important?
  - Improved crop management practices.
  - Accurate prediction of crops.
  - Better logistical planning of field operations.
  - Supports post-harvest decisions



# Research on Irregular Bearing

- Find the causes of irregular bearing
- Control Measures to reduce irregular bearing:
  - Develop new varieties/cultivars
  - Proper management to improve crop nutrition
  - Application of hormones
  - Pruning
  - Pollination

**However, no attempt has been made to understand irregular bearing using remote sensing technique and how we can predict this.**

# Research on Avocado by AARSC



- Rahman, M.M.; Robson, A.; Brinkhoff, J. Potential of Time-Series Sentinel 2 Data for Monitoring Avocado Crop Phenology. *Remote Sens.* **2022**, *14*, 5942. <https://doi.org/10.3390/rs14235942>
- Robson, A.; Rahman, M.M.; Muir, J. Using Worldview Satellite Imagery to Map Yield in Avocado (*Persea americana*): A Case Study in Bundaberg, Australia. *Remote Sens.* **2017**, *9*, 1223. <https://doi.org/10.3390/rs9121223>
- Satellite remote sensing based time series model for yield prediction of avocado crops. (Not yet published)



# Research Questions

- Is it possible to understand irregular bearing using satellite remote sensing data?
- How to predict 'On' or 'Off' crops using remote sensing time series data for Avocado?

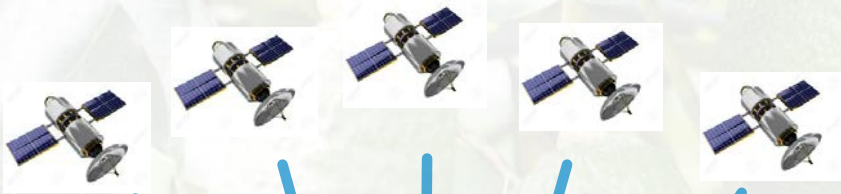


# Study Location

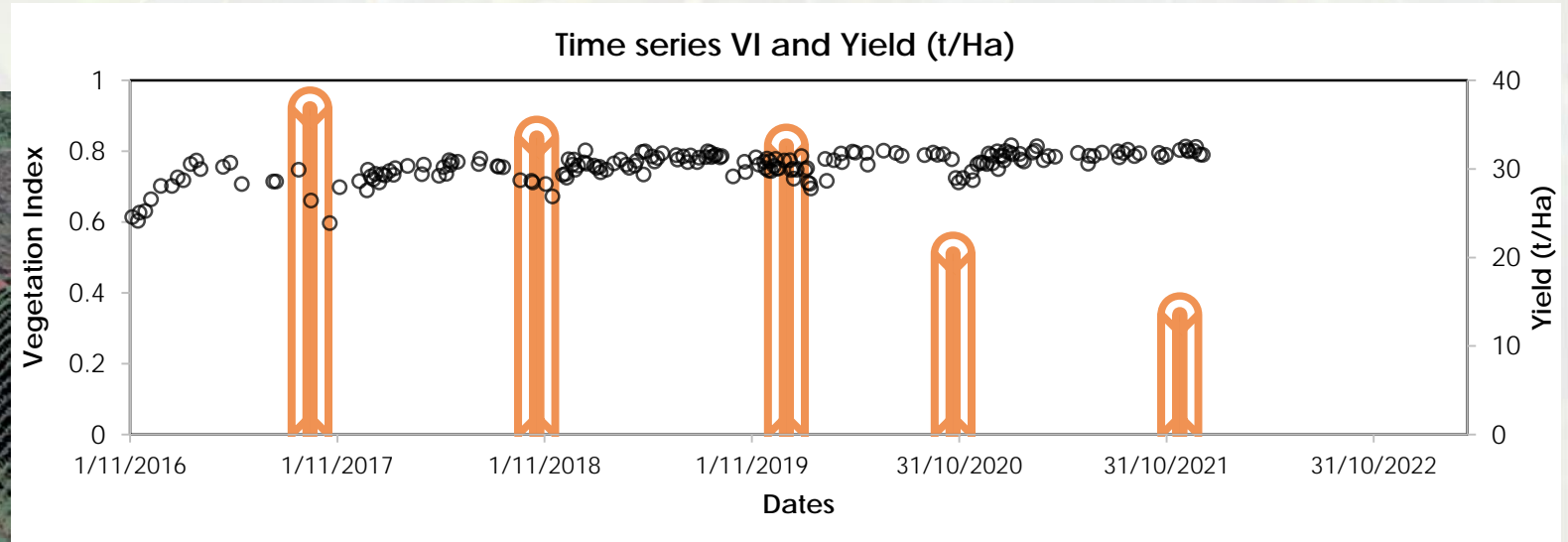
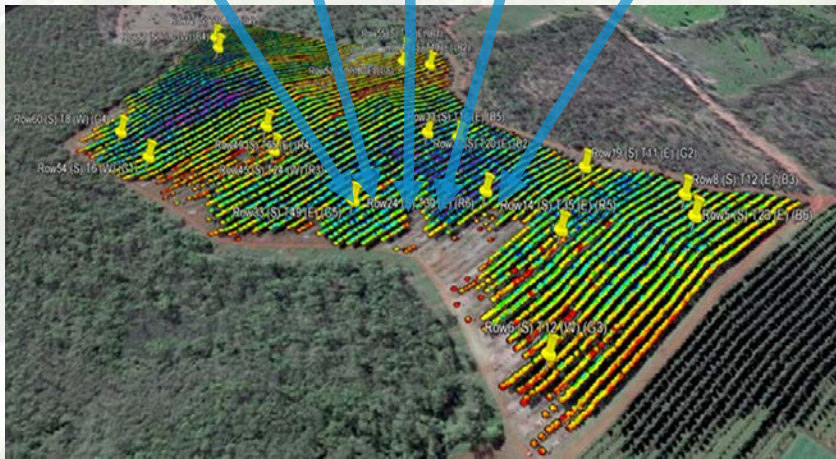
- Queensland – 3 farms
- South Australia – 1 farm
- Western Australia – 1 farm
  
- Total 40 blocks from 5 farms
- From 2017 to 2021 data
- Varieties – Hass and Shepard



# Data Collection



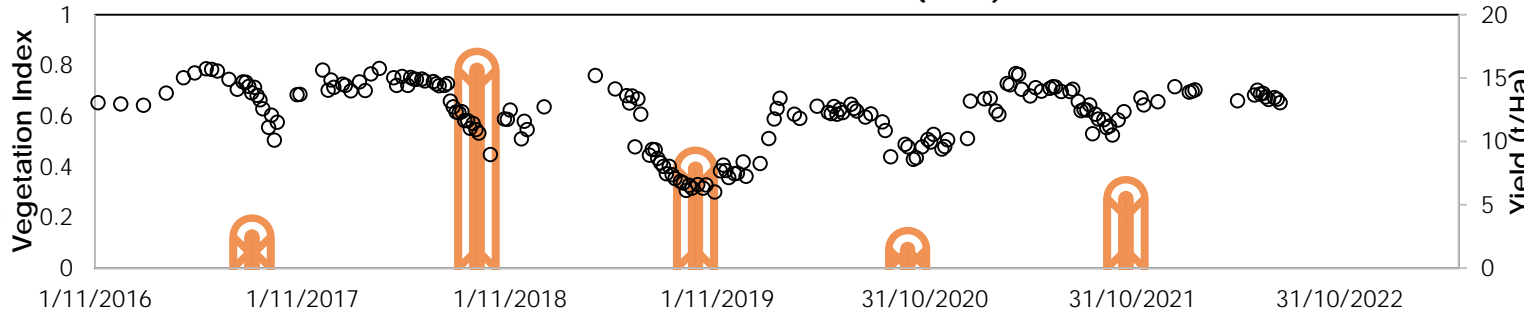
Vegetation Index (VI) = band ratio combination = crop reflectance



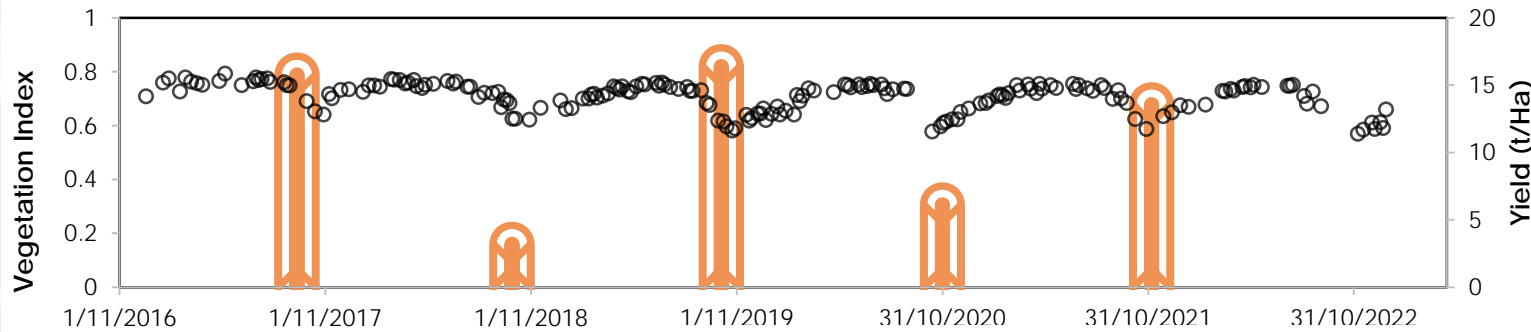
Sentinel 2 derived vegetation index (crop reflectance) and historical yield data from 2017 to 2021 for one block

# Regional Variation

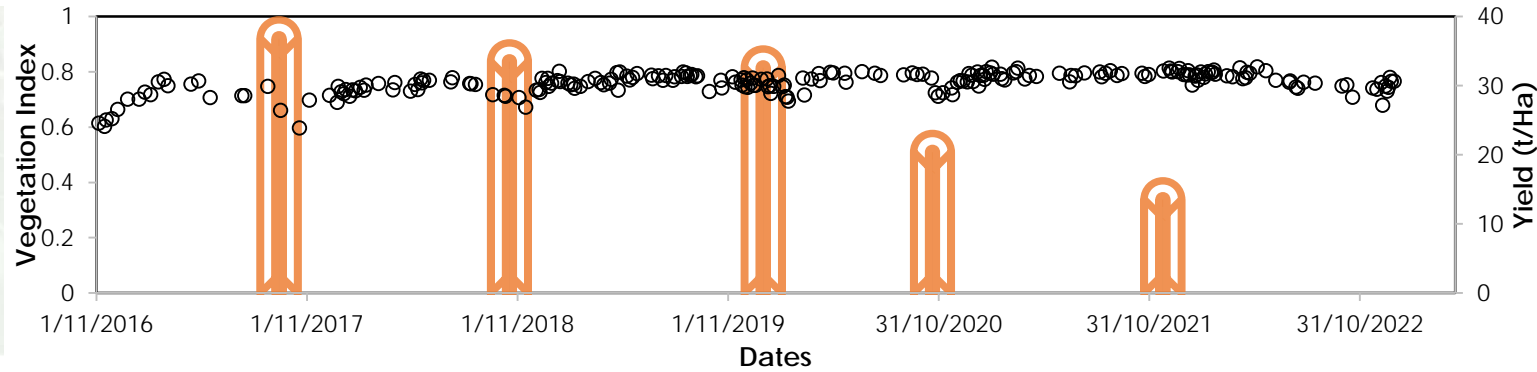
Time series VI and Yield (t/Ha)



Queensland



South Australia



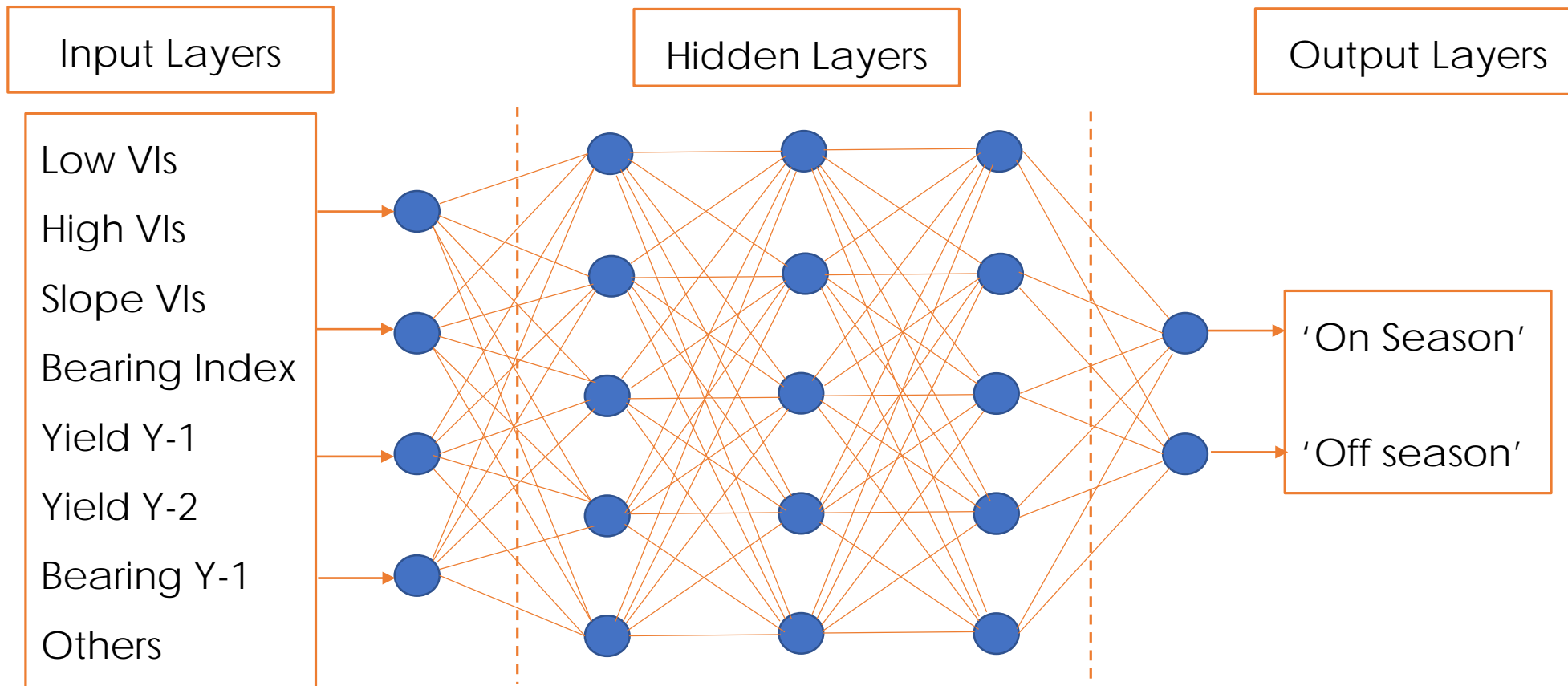
Western Australia



# Input Variables

- Lowest VIs values
- Highest VIs values
- Slope between two VIs values
- Bearing index =  $(\text{yield Y1} - \text{yield Y2}) / (\text{yield Y1} + \text{yield Y2})$
- Yield achieved in the previous season
- Yield achieved in the season 2 years prior
- Bearing previous year (Whether 'on' or 'off')
- Other weather or management variables (e.g. drought, floods, pruning etc.)

# Machine Learning Model Development

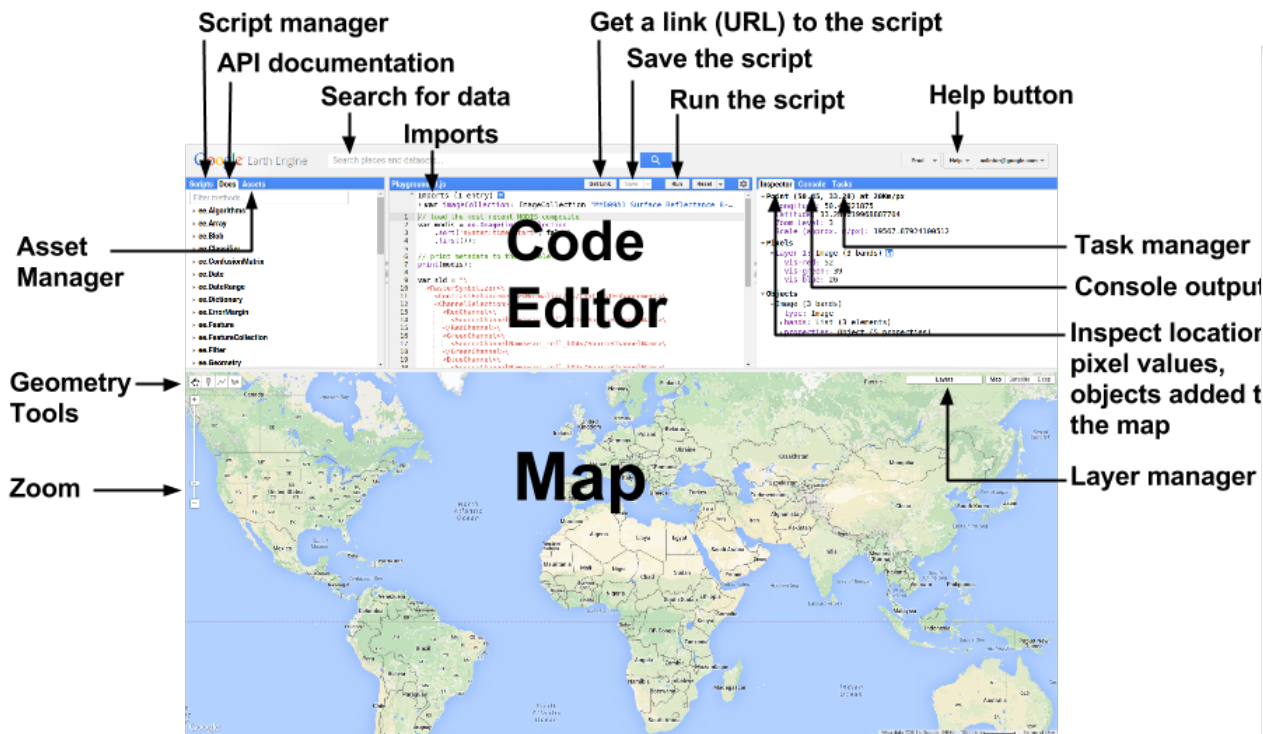


# Machine Learning Models

## Machine Learning Models used for classification of 'On' or 'Off' season:

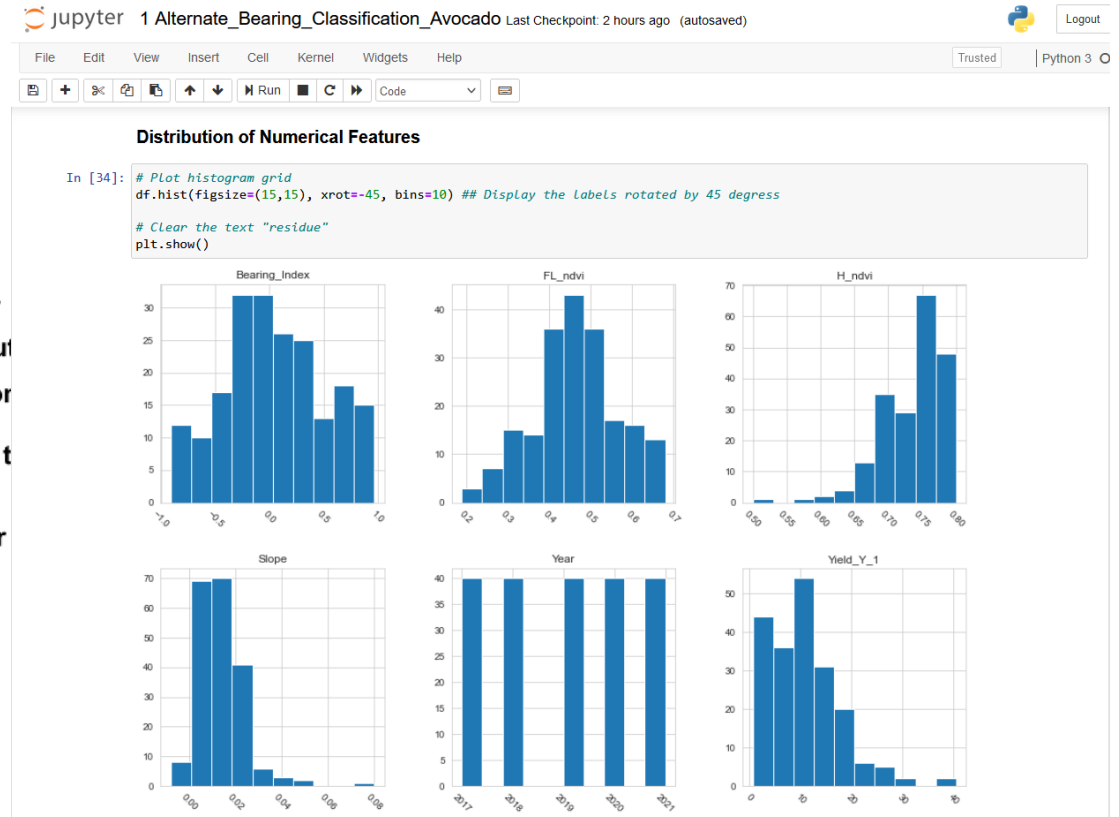
- Logistic Regression
- Random Forest Classification
- XGBoost Classification
- KNN Classification
- Decision Trees
- Support Vector Classification
- Gradient Boosting

# Model Development Platforms



**Script manager**  
**API documentation**  
**Search for data**  
**Imports**  
**Get a link (URL) to the script**  
**Save the script**  
**Run the script**  
**Help button**  
**Asset Manager**  
**Code Editor**  
**Task manager**  
**Console output**  
**Inspect location pixel values, objects added to the map**  
**Layer manager**  
**Geometry Tools**  
**Zoom**  
**Map**

Google Earth Engine Code Editor



**Jupyter 1 Alternate\_Bearing\_Classification\_Avocado** Last Checkpoint: 2 hours ago (autosaved)

```

In [34]: # Plot histogram grid
df.hist(figsize=(15,15), xrot=-45, bins=10) ## Display the Labels rotated by 45 degrees

# Clear the text "residue"
plt.show()
    
```

**Distribution of Numerical Features**

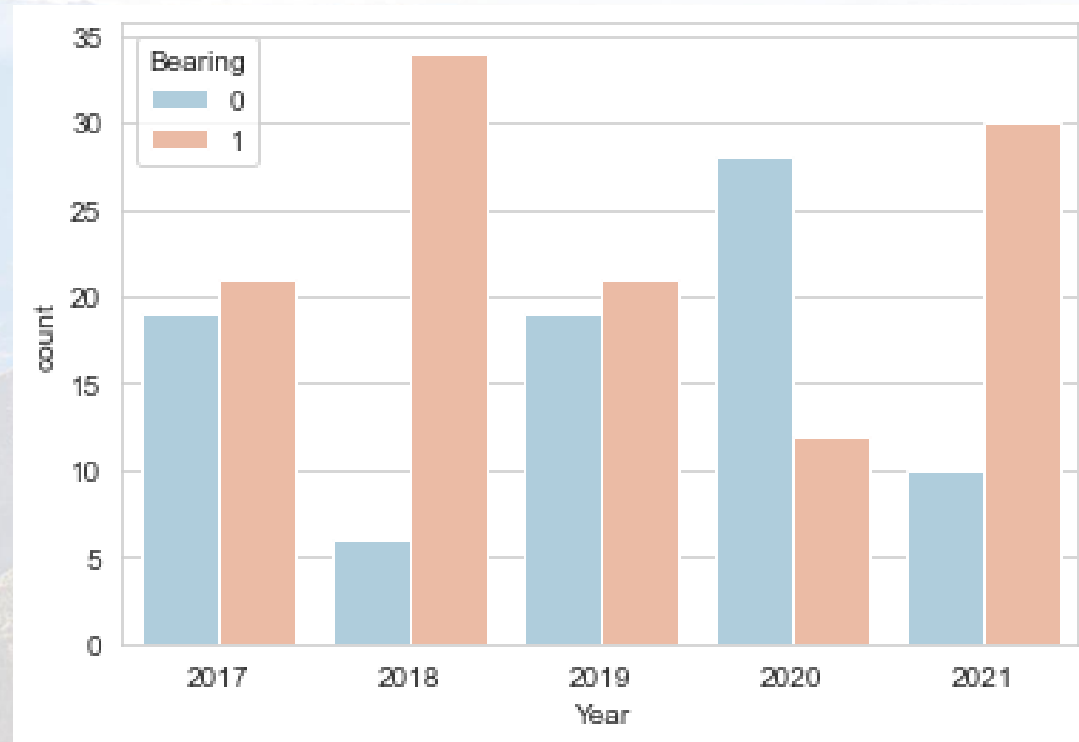
- Bearing\_Index
- FL\_ndvi
- H\_ndvi
- Slope
- Year
- Yield\_Y\_1

Python 3 in Jupyter Notebook



# Alternate Bearing in 5 years

Yearly 'On' or 'Off' season data from 40 blocks over 5 years

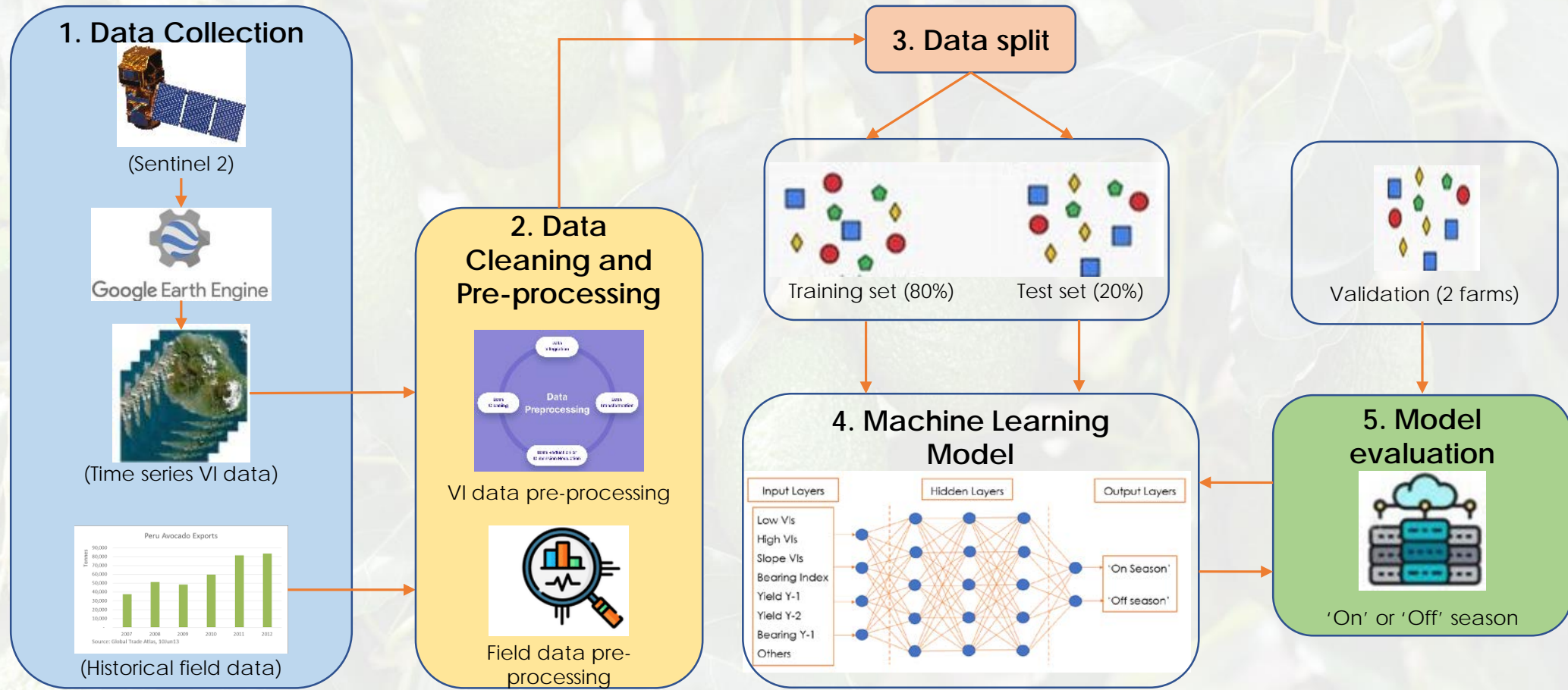


'0' = 'Off' crops = 82  
'1' = 'On' crops = 118

Yearly 'On' and 'Off' season in 40 blocks  
(Total data 200 from 5 years)

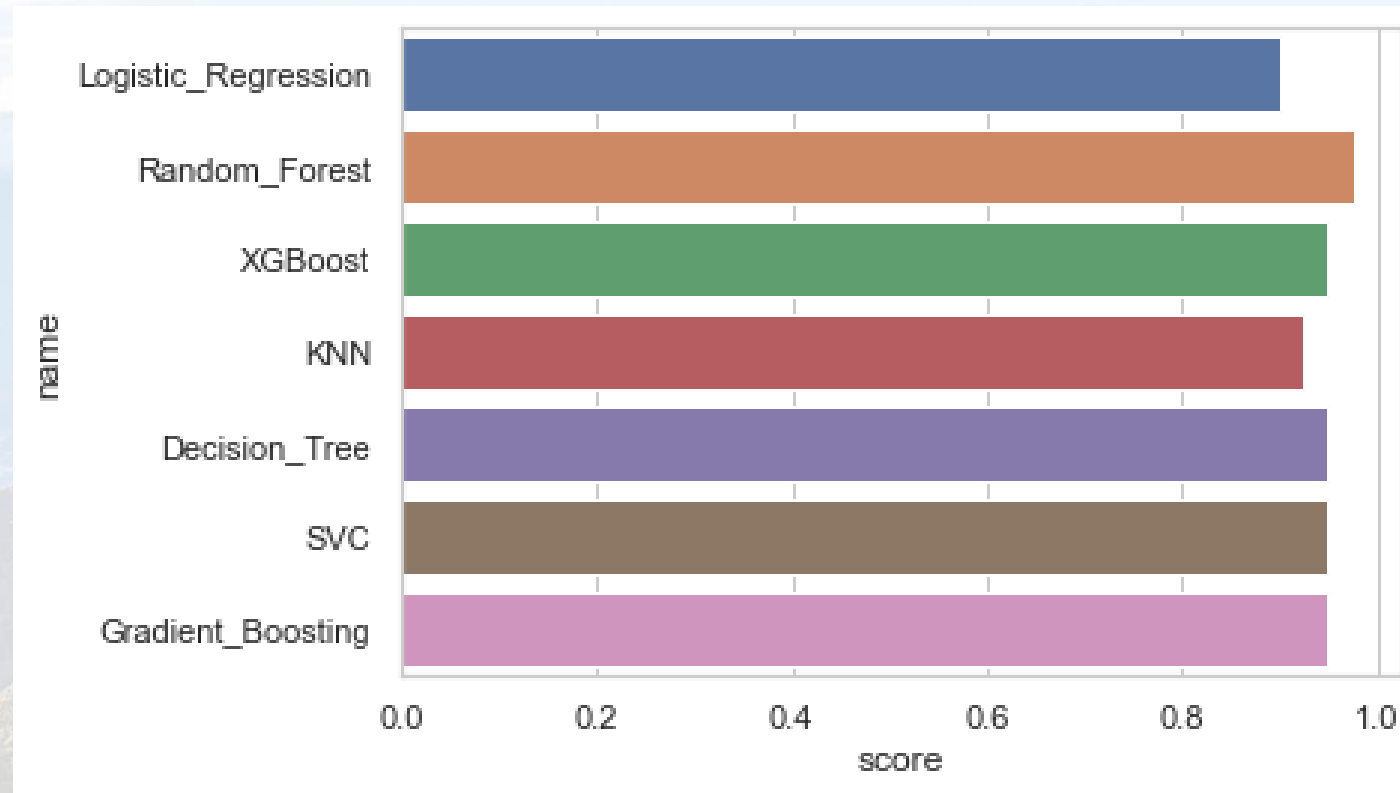


# Model development flowchart



Model development and validation flowchart

# Model Performance

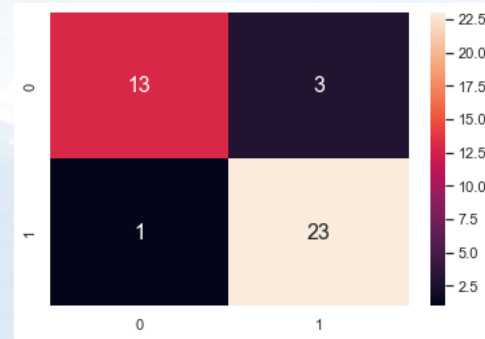


Model performance over test set (40 blocks)

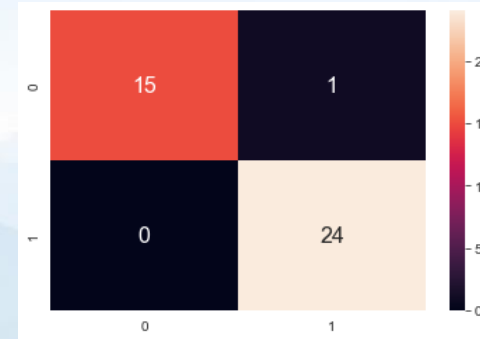
# Confusion Matrices

## Confusion Matrix

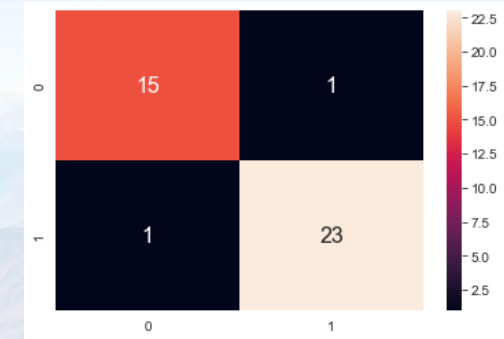
	Actually Negative (0)	Actually Positive (1)
Predicted Negative (0)	True Negatives (TNs)	False Negatives (FNs)
Predicted Positive (1)	False Positives (FPs)	True Positives (TPs)



Logistic Regression



Random Forest



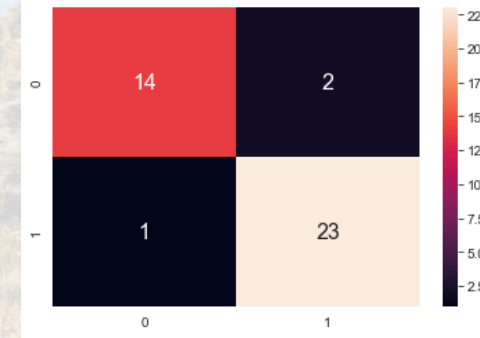
XGBoost



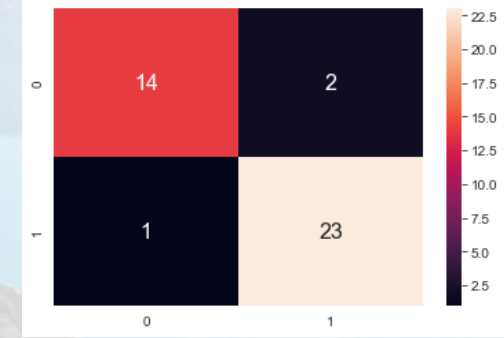
K Nearest Neighbour



Decision Tree



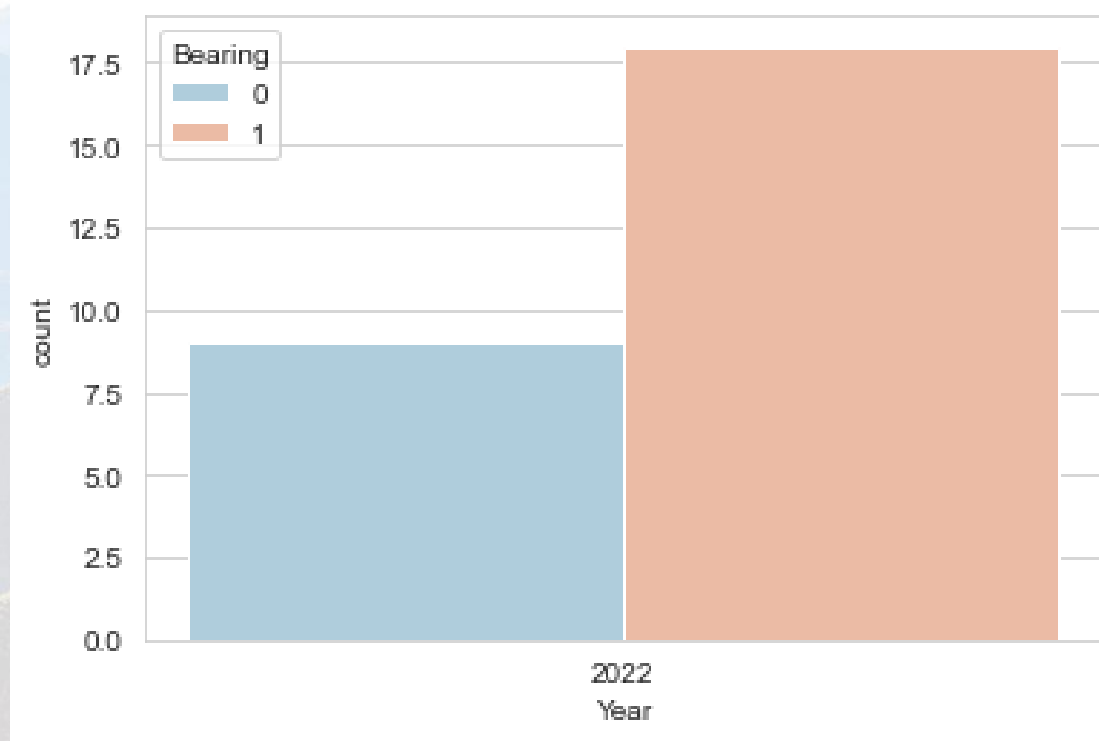
Support Vector



Gradient Boosting

# Alternate Bearing in 2022

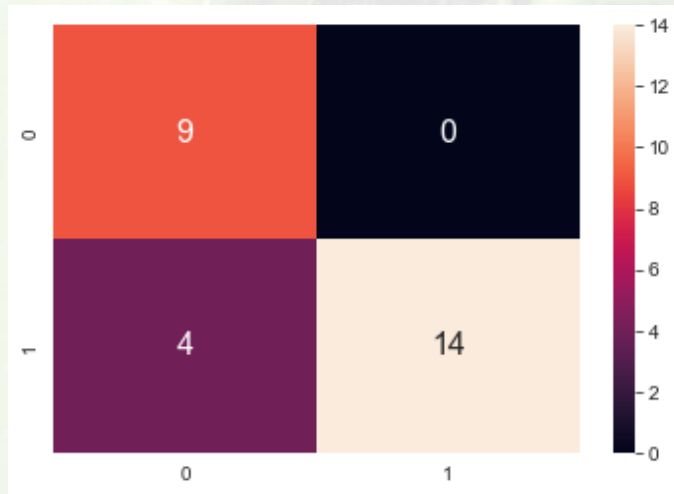
'On' or 'Off' season data from 27 blocks on 2022 year



'0' = 'Off' crops = 9  
'1' = 'On' crops = 18

'On' and 'Off' season crop in 2022 in 27 blocks

# Random Forest on 2022



## Confusion Matrix

	Actually Negative (0)	Actually Positive (1)
Predicted Negative (0)	True Negatives (TNs)	False Negatives (FNs)
Predicted Positive (1)	False Positives (FPs)	True Positives (TPs)

Accuracy = 0.85

# Conclusions

- Satellite remote sensing has the potential to predict 'On' or 'Off' crops.
- The 'Random Forest' provides better accuracy.
- The developed model can provide information at the early stage of season (4 months prior to harvest).
- This is exciting at this stage, since it supports growers to improve management decisions.

# Future Works

- Acquire more data from growers to improve model accuracy and generalize the model
- Apply at different regions (internationally) with various management practices
- Apply the model as an input to improve yield prediction of Avocado crops

# Acknowledgements



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Australian Government  
Department of Agriculture  
and Water Resources



Hort  
Innovation





# Questions



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