

**PROTOTYPING** 

# **Crop Disease Detection**

End of internship presentation

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Prototyping Architect Intern AWS



# Agenda

Personal background

**Business objectives** 

Key features

Demo

**Technical Approach** 

**Next Steps** 

Mentors



# **Personal Background**

#### **Degree**

- Computer Networks and Telecommunications Engineering degree
- No prior ML knowledge
- No prior AWS Cloud knowledge

#### Internship

- Start date: 3<sup>rd</sup> April 2023
- End date: 18<sup>th</sup> August 2023

#### **School**

Location: Toulouse, France





## **Business Objectives**

• Crop monitoring is a labor-intensive task. Regular inspection for disease takes

10% to 30% of the overall time dedicated on a field.

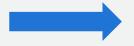
- Crop loss due to plant disease costs \$220 billion annually (global economy)
- This application targets Agronomists who operate on multiple fileds

aws

## **Business Objectives**

# Problem Statement

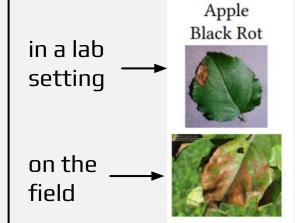
Why?



#### **Objectives and requirements**

Currently no viable ML solution is implemented for crop disease detection

Poor quality or lack of crop dataset



#### **Quality of data**

Leverage agronomists' knowledge of crops

#### **Usability**

Easy enough interface to be used by non-tech users

#### **Scalability**

Needs to encompass multiple crop types



# **Key features**

- Handle image upload
- Detect disease and crop type
- Allow data labeling (bounding boxes and class)
- Save image and corresponding annotations
- Enhance initial dataset with new images and annotations
- Monitor model performance and use enhanced dataset to re-train model

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# Demo



# **Technical Approach**





# **Preparation Phase**



# Identifying the problem and customer

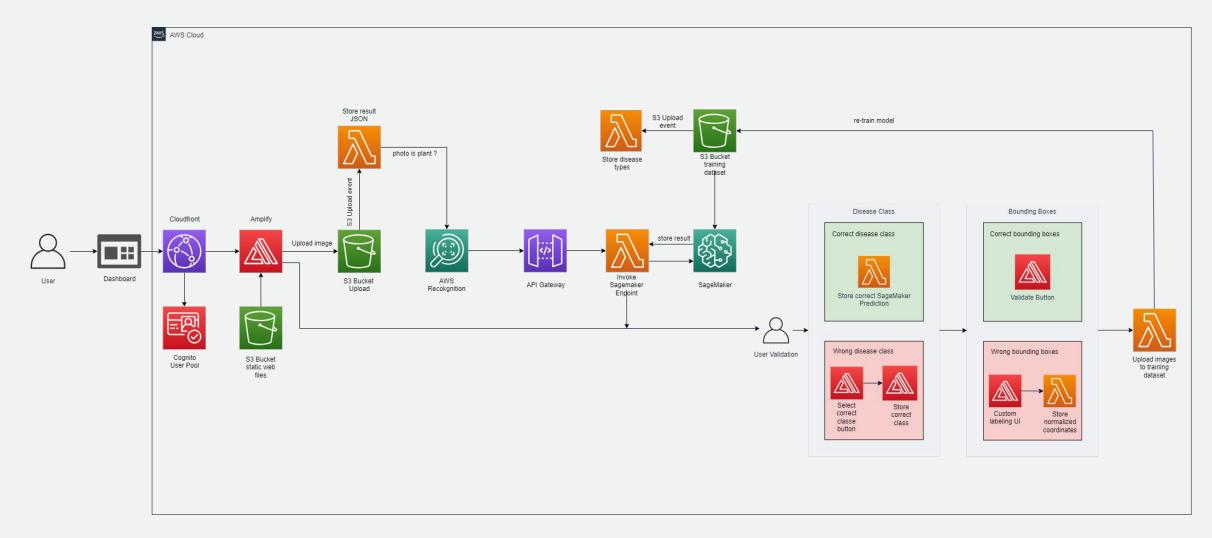
- TFC Input
- Choosing the model
- Defining Architecture



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## **Architecture**

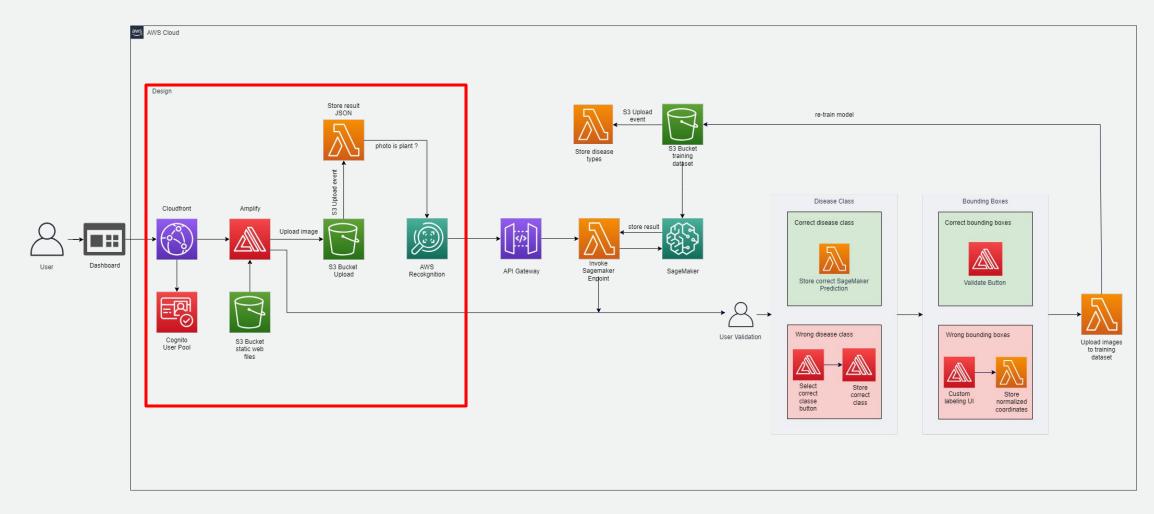




# Design Phase



# **Design Phase**





# **Design Phase**

#### **Amplify / React**

- Front end development
- Back end integration with AWS Services



#### **Cognito**

Authentication and authorization



#### Rekognition

Checking that image is a leaf



#### **S3**

- Storing result
- Storing images



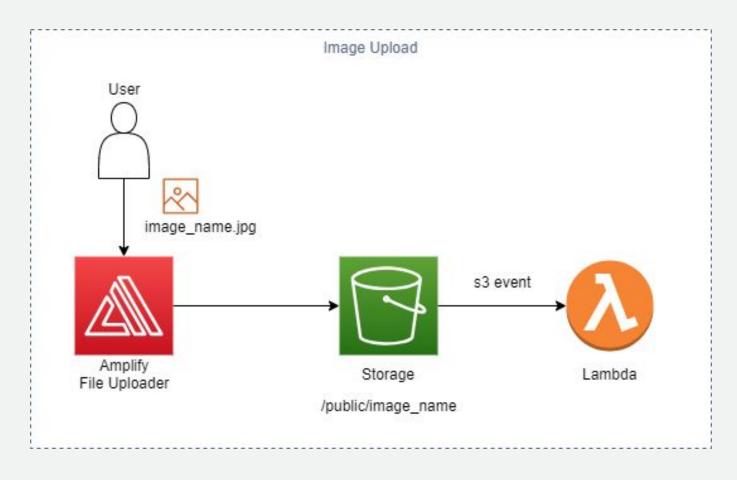
#### **CloudFront**

Deployment





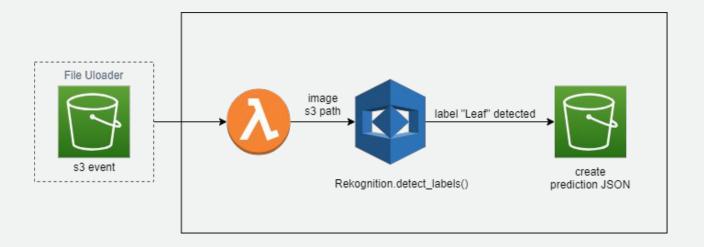
# **Image Upload**



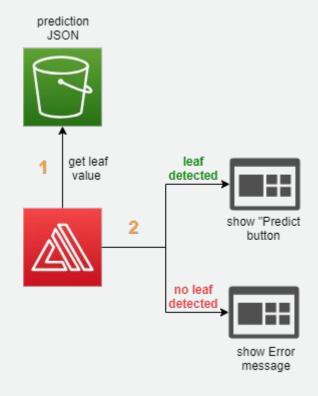


# **Image Analysis**

#### Step 1



#### Step 2



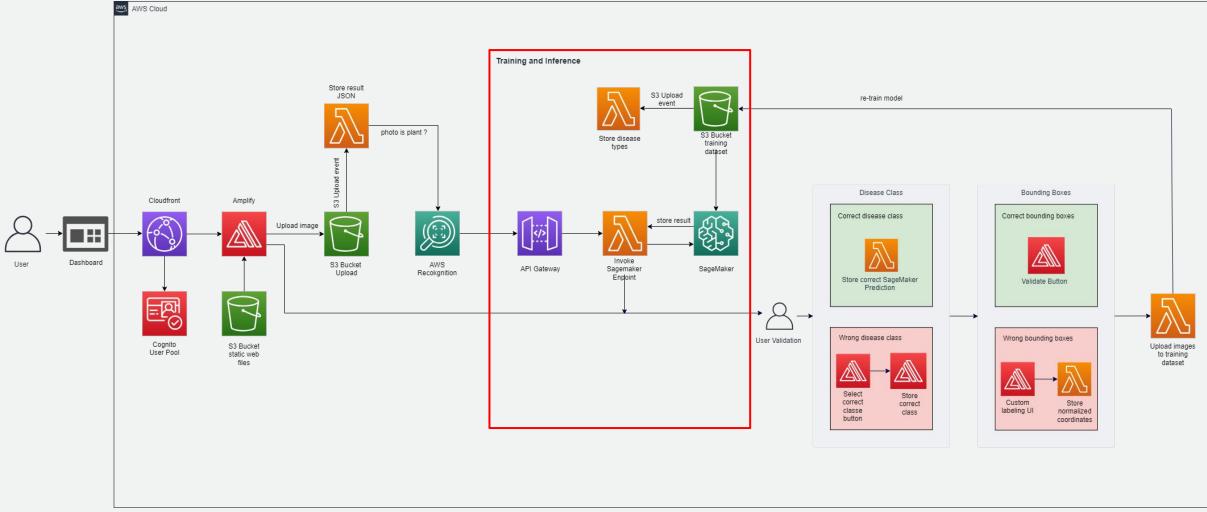


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# **ML Phase**



## **ML Phase**

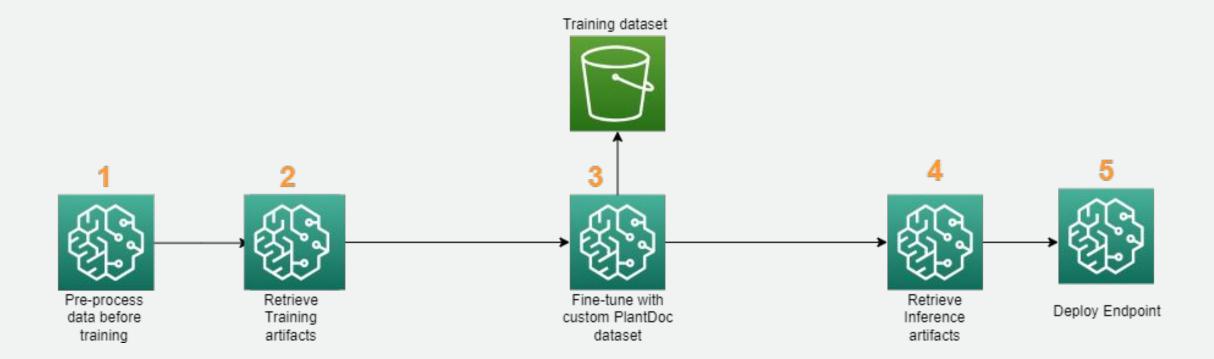


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# **ML** pipeline





# **Data Analysis**

#### **Pre-trained Model**

Tensorflow Resnet50

#### **Dataset**

Plantdoc dataset

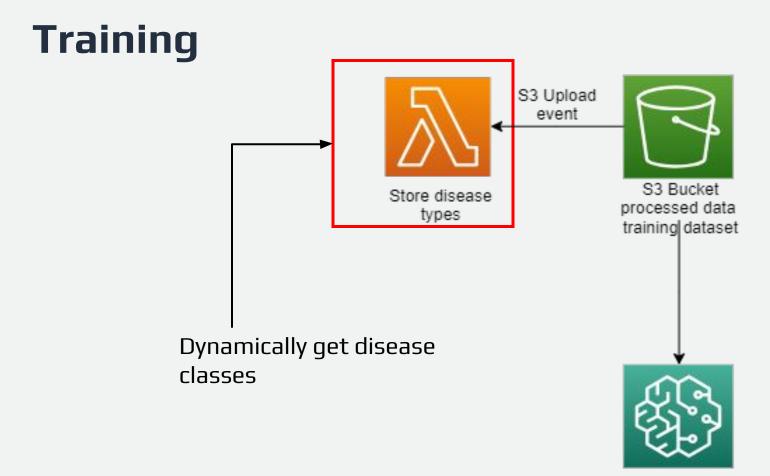
#### **Pre-processing**

- Image size
- Color channels



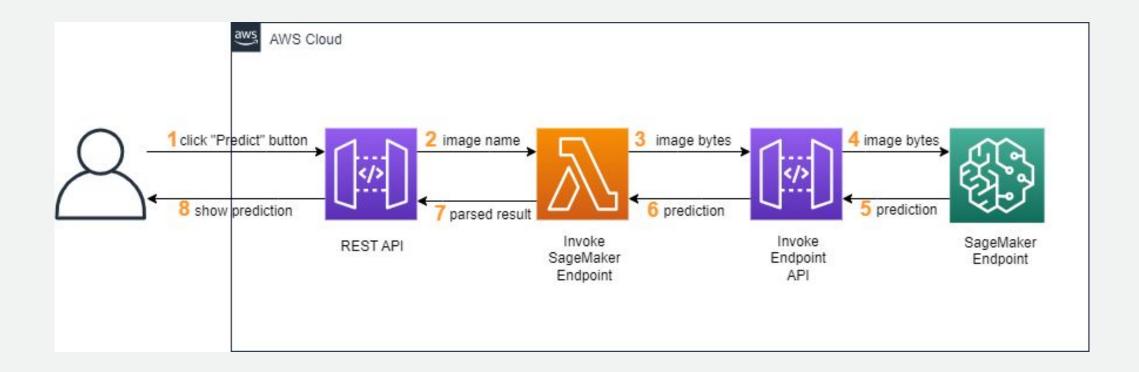








## Inference

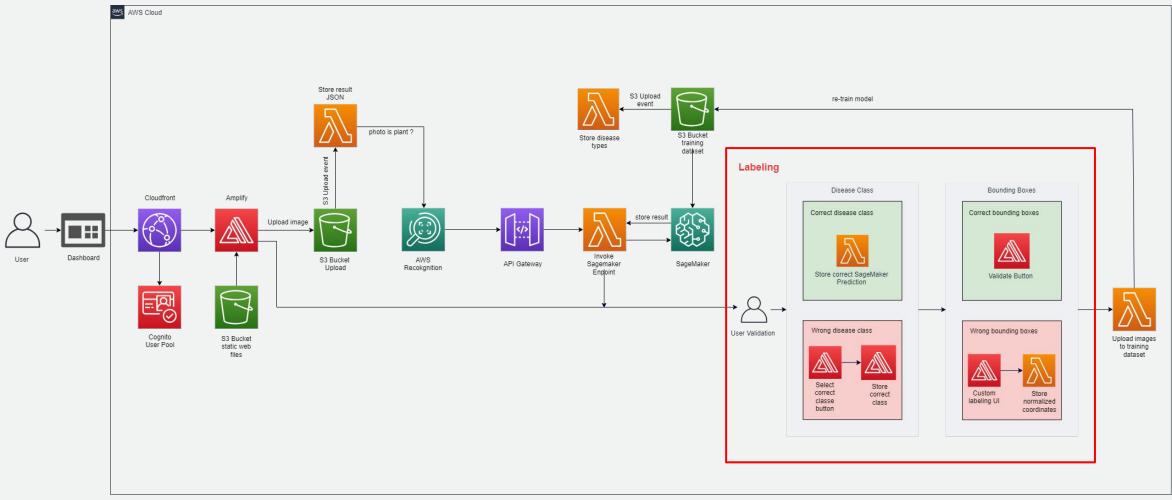




# **Data Labeling Phase**



# **Data Labeling Phase**





# **Ground Truth vs Custom Labeling UI**

#### **Ground Truth**

- Impossible to integrate with existing user pool
- Limited in defining disease classes

#### **Custom Labeling UI**

Custom UI using Typescript and AWS services

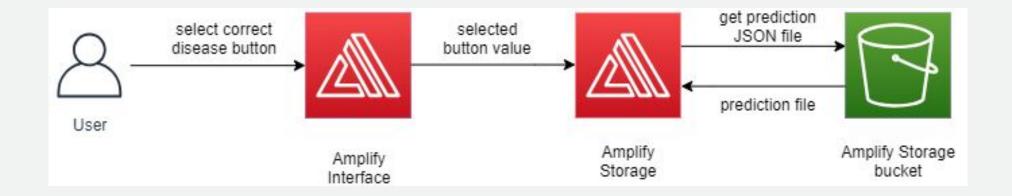




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Leverage AWS Amplify and S3 integration

## **Correct disease class**

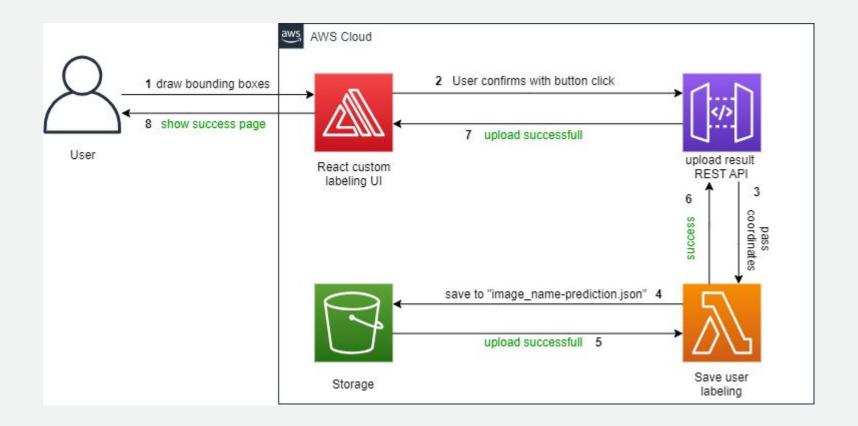




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# **Correct bounding boxes**





# Model Monitoring and retraining



# Steps to monitor and retrain model

#### **Accuracy Goal**

- Need to define a goal accuracy
- What parameters to implement? New photo numbers per category? Other parameter?
- Periodic retraining?

#### Oversampling issue

Need to deal with inequality of sample size per disease

#### Sagemaker Pipeline

- Set up a Sagemaker pipeline
- Move the annotated images from the upload bucket to the dataset bucket

#### Alarm to track error percentage

- Track the percentage of user rectifications versus the number of inference
- Set up an alarm to trigger re-training for over 30% of rectifications



# **Next Steps**

	Suggestion	Advancement
Smart Crops	Possible add-on to the smart crops demonstrator	Process in progress
Open source	Open source the project on AWS	Process in progress
TFC Demo	Handover the project to TFC	Process in progress
Handover	Handover to a team member (Paul Devillers)	Process planned



# **Next Steps:** road to optimization

#### WHAT TO IMPROVE

- Add Infrastructure As Code (Terraform)
- Add other crop types to asses model performance

#### **FEEDBACK FROM TFC**

- Keep track of image metadata
- Localize images with metada on a map
- Image classification for crop types before detecting the disease



# Fun Experiment : Synthetic Data

Text to image : « Apple rust disease leaf »



Image to image:





#### **Mentors**



**Paul Devillers** 

Overall mentorship in different phases



**Bishesh Adhikari** 

Data Analysis and Preparation assitance



**Mohamed Ali Jamaoui** 

Data Analysis and ML pipeline assistance



Ion Kleopas

Labeling Phase assitance



**Florian Clanet** 

ML inference and training assistance



**Stephen Hibbert** 

Labeling Phase assistance



## **Additional resources**



https://gitlab.aws.dev/rkhemiri/crops-disease-detection
Gitlab repository (Internal only)

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# Thank you!

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