

# PROJECT SILVAFLUX

## Design Review 3

**Big Data Computing and Interface for  
Tropical Forest Regeneration**

By Team Clean Carbon

# Introduction

**Frontend team:**

Curtis McHone - Team Lead  
Justin Stouffer  
Jonathan Bloom

**Backend team:**

Richard McCue  
Shayne Sellner

**Team Sponsor:**

Allie (Alexander) Shenkin

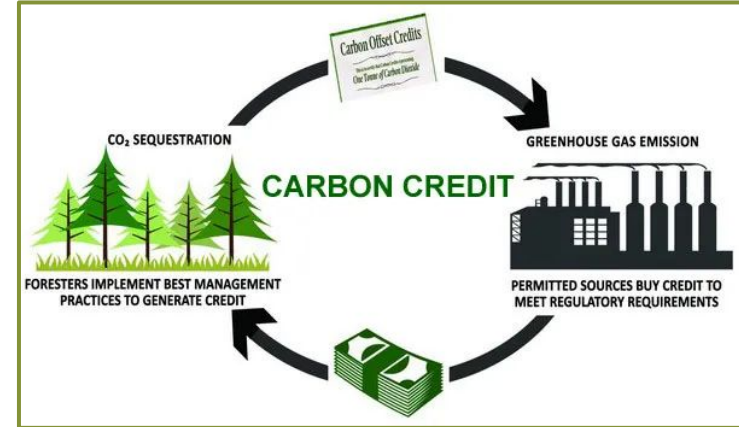
**Team Mentor:**

Vahid Nikoonejad Fard



# Introduction

- Our sponsor Allie (Alexander) Shenkin and his team have discovered a new climate cooling service that allows for 30% more carbon credits to be sold for a designated plot of land
- Carbon credits are purchasable credits that landowners and project developers can sell. These carbon credits directly correlate to the amount of carbon dioxide a certain plot of land takes in
- Developers and landowners sell these credits to help businesses or corporations offset their carbon footprint



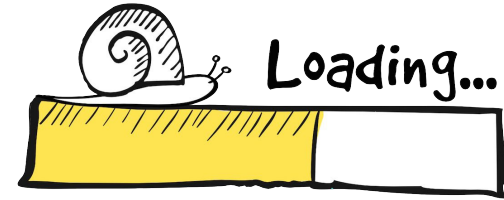
# Introduction

- Our team has been asked to create an application that is able to calculate the amount of carbon dioxide a certain area will uptake using this new discovery
- As a team we hope that our final application makes buying and selling carbon credits more profitable and accessible to the average person as well as revolutionize the way carbon credits are sold and predicted



# Problem Statement

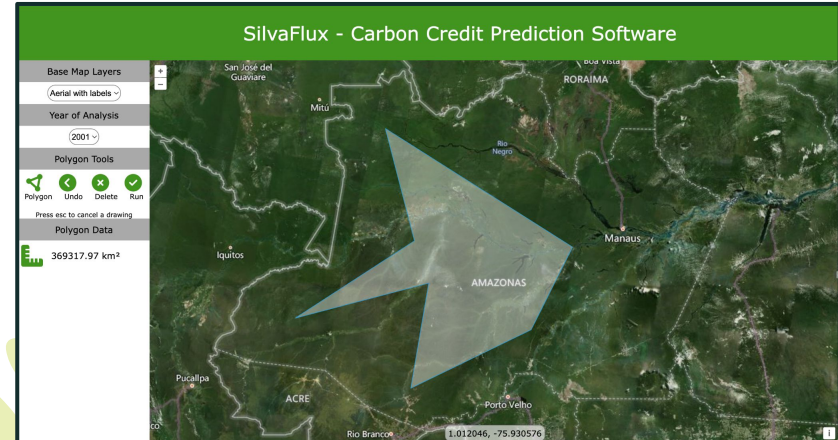
- There is not currently a way to accurately predict the amount of CO<sub>2</sub> uptake for a plot of land.
- Because of this land developers cannot predict how much money they are going to make or if the project is even going to be worth their time
- Allie's discovery makes investing in reforestation projects much more profitable, helping not only the investor but the planet, but what is the best way to implement this
- The current software prototype that is used to calculate the CO<sub>2</sub> prediction is simply too slow, inefficient, and not user friendly



# Solution Overview

## Front End

- Web interface
  - Django Web Framework
  - Utilizes the OpenLayers javascript library
  - Easy to use map interface
    - Bing Maps API base maps
    - Ability to Zoom, Pan, Scale
- User friendly and responsive
- Ability to draw a polygon
- Measure area
- Sends the polygon to the prediction script



# Solution Overview

## Back End

- High performance linux based server
- Python based prediction system
  - Static raster simulation
- Raster layer storage
- Send results to front end

## Database

- PostgreSQL database
- Stores user login information
- User access control
- Expandability for user query storage

# aws

The AWS logo consists of the lowercase letters 'aws' in a dark blue, sans-serif font. Below the letters is a thick, orange curved arrow that starts under the 'a' and points towards the 's'.

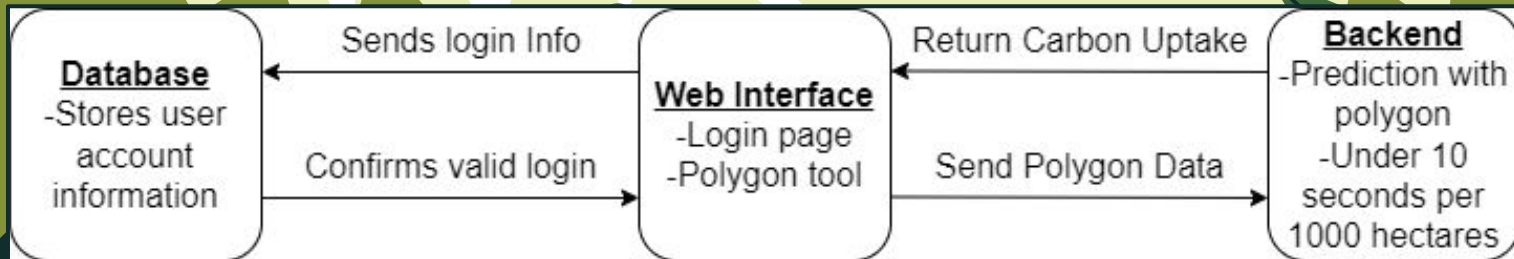
# Requirement/Architecture Overview

## Requirements Acquisition

- Weekly client meetings
- Initial project description and overview
- External research & sponsor recommendations

## Requirements

- Simple web interface with an interactable map
- Send polygon to backend
- Computationally efficient backend that computes carbon uptake
- Database with encryption to store user account information

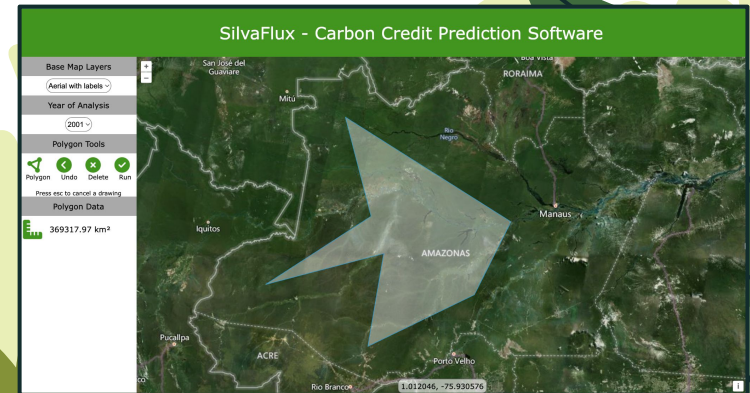




# Architecture/Implementation Overview

## Front End

- Django Framework
- OpenLayers map using bing maps baselayers
- Communicates with Django Postgre database to verify login information
- Polygon tool on map built into OpenLayers
- Calls backend script with polygon coordinates and year selected

The Django logo is displayed in white lowercase letters on a dark green rectangular background.

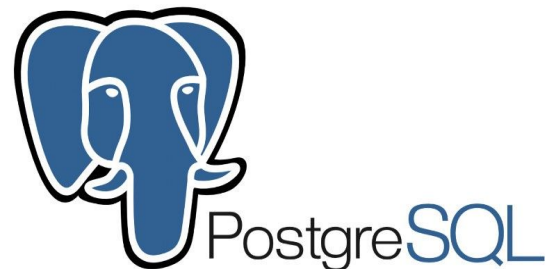
# Architecture/Implementation Overview

## Backend

- Precomputed rasters for years 2014-2021 for global prediction
- Python script to compute the carbon uptake on the polygons plot of land
- Return carbon uptake back to front end

## Database

- Stores user login information utilizing encryption
- (Stretch goal) Store user queries for usage based billing



# Prototype

**Silvaflux Login**

Username \_\_\_\_\_

Password \_\_\_\_\_

Submit


### SilvaFlux - Carbon Credit Prediction Software

Base Map Layers  
Aerial with labels

Year of Analysis  
2001

Polygon Tools  
Polygon Undo Delete Run  
Press esc to cancel a drawing

Polygon Data



40.428075, -116.900336

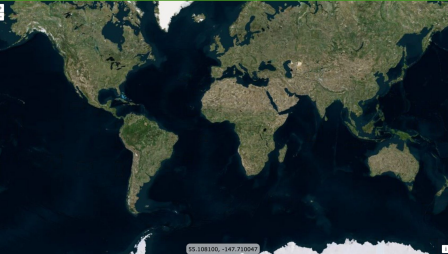
### SilvaFlux - Carbon Credit Prediction Software

Base Map Layers  
Aerial

Year of Analysis  
2001

Polygon Tools  
Polygon Undo Delete Run  
Press esc to cancel a drawing

Polygon Data



65.108105, -147.210543

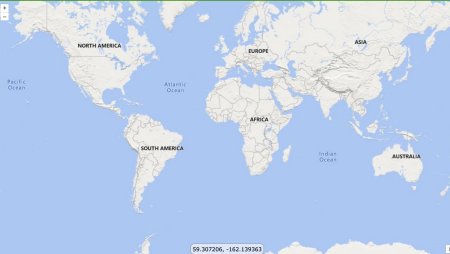
### SilvaFlux - Carbon Credit Prediction Software

Base Map Layers  
Road

Year of Analysis  
2001

Polygon Tools  
Polygon Undo Delete Run  
Press esc to cancel a drawing

Polygon Data



59.307206, -142.139363


### SilvaFlux - Carbon Credit Prediction Software

Base Map Layers  
Road dark

Year of Analysis  
2001

Polygon Tools  
Polygon Undo Delete Run  
Press esc to cancel a drawing

Polygon Data



64.307575, 176.910093

# Prototype

SilvaFlux - Carbon Credit Prediction Software

Base Map Layers  
Aerial with labels

Year of Analysis  
2001

Polygon Tools  
Polygon Undo Delete Run

Press esc to cancel a drawing

Polygon Data  
27742.62 km<sup>2</sup>

Barcelos

Reserva Extrativista Rio Urubaxi

Parque Nacional do Jaú

-1.515851, -64.928232



Annual tons CO<sub>2</sub> uptake: -0.323998

# Challenges

## Backend

### Raster storage:

- Our rasters are up to ~45GB in size, up to 21 rasters
- 45GB x 21 rasters = ~945GB total storage needed

### Runtime:

- Takes very long to read the raster into memory

## Frontend

### Local CSS:

- CSS for OpenLayers is not working locally

### Django:

- Could not start up the web server



# Resolutions

## Backend

### Storage Solutions:

- Increase our AWS storage size
- Support less rasters

### Runtime Solutions:

- Reading only sections of the raster into memory



## Frontend

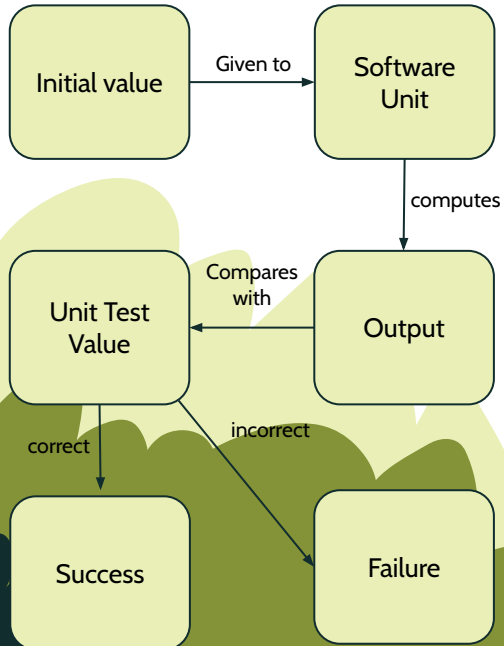
### CSS and Django Solutions:

- More research
  - On static CSS files and linking
  - On Django file architecture and server startup

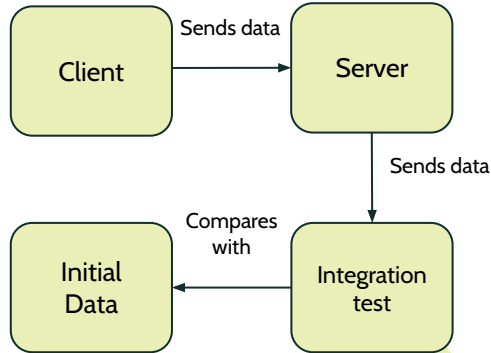


# Testing Plan

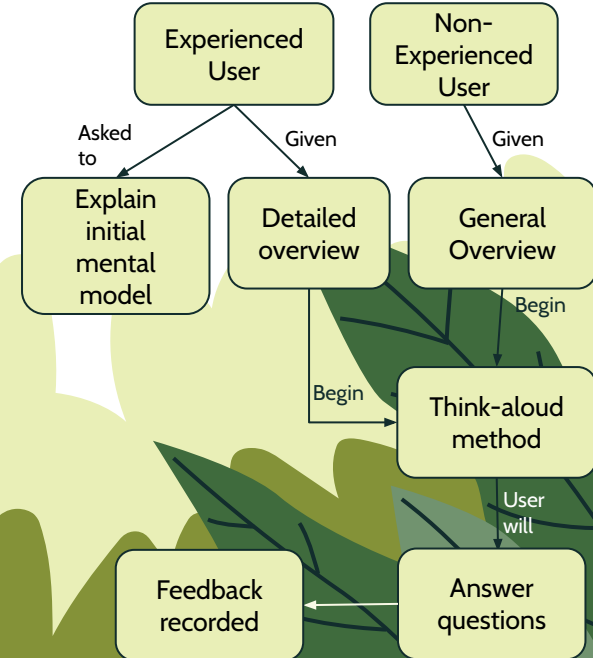
## Unit testing



## Integration testing



## Usability testing



# Testing Plan

## Unit testing

Ensures our software and code is accurate and reliable

Tested Units:

- Correct coordinates
- Map projection
- CO2 uptake amount
- Year of prediction selection

Once Unit testing is complete, our team will know exactly which functions/isolated pieces of software work correctly

## Integration testing

Four different tests:

- Query results are correctly sent over to postgres from Django
- Logout function restricts access to home page
- Prediction page vulnerabilities
- Frontend coordinates correctly sent and match to backend coordinates

Once Integration testing is complete, our team will know exactly which software systems are correctly communicating with one another

## Usability testing

Users will provide feedback on:

- Initial Impressions
- Ease of use, navigation around interface
- Expected functionality of features
- Interpretation of results, useful or redundant data

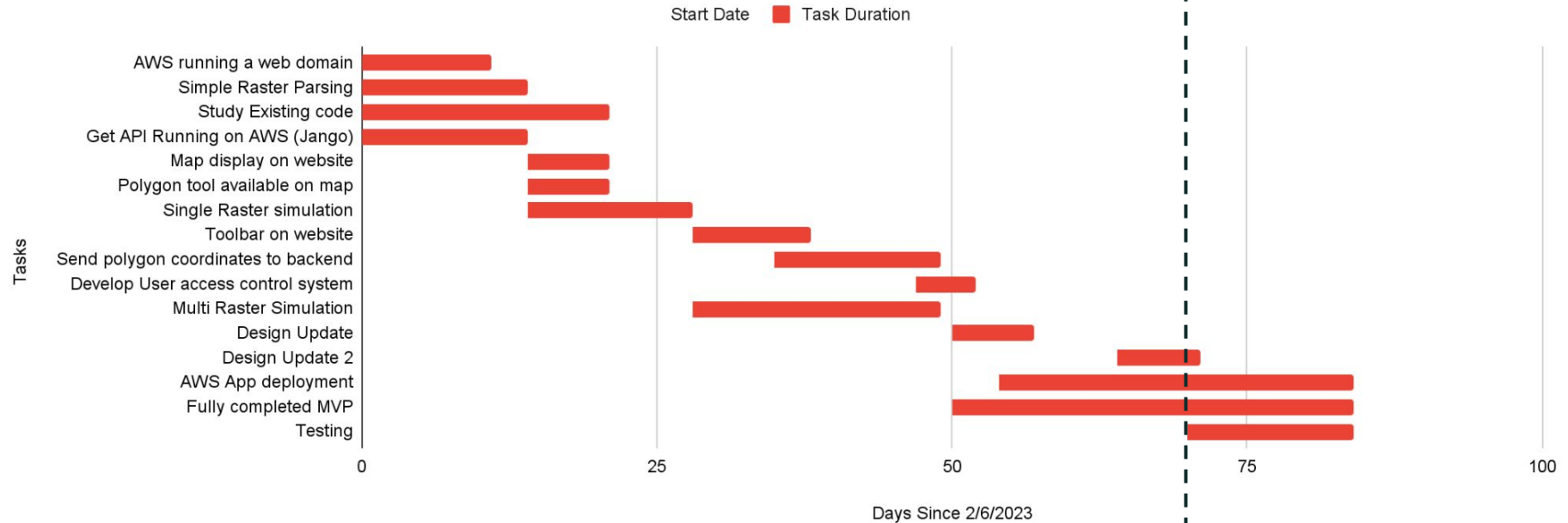
Once Usability testing is complete, our team will have a better understanding of what users are thinking when using our system, and the different functionalities that they find important



# Schedule

Now

## Implementation for Project Silvaflux 2023



# Conclusion

Our current implemented solution includes:

- A responsive and convenient web interface front-end with authenticated user access control system (Django)
- Fast and secure data transmission between our web interface and backend (Django)
- Secure database that holds all of the user access information (PostgreSQL)
- Correct data sent back to the user with additional useful information

## Functional Requirements:

Simple web interface with a interactable map, with fast communication with backend



Computationally efficient backend that will run the prediction on a specified plot of land



Database with encryption to store user account information



# Conclusion

## System Performance

- Instantaneous Login/Logout connection with database
- Under 10 sec runtime for a polygon of 1000 hectares



## Challenges/Resolutions:

- Computational performance was accelerated to meet runtime requirements
- AWS storage was upgraded to hold a larger amount of rasters
- Django and Web front end are up and running with working CSS

## Architecture Implementation

- Prediction computation located on the backend, implemented in Django framework
- High performance AWS backend server
- PostgreSQL database used to store authenticated user information



OpenLayers

