



| GALY Resource Optimization Wizard

# GROW

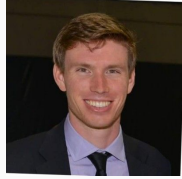
Team: Sustainobots



MIT ENERGY &  
CLIMATE HACK



# | Team Sustainobots!



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**SustainoBot**  
(Honorary Member)



## **| Sustainable Agriculture is key**

**29%**

GREENHOUSE GAS EMISSIONS  
COMES FROM AGRICULTURE

**70%**

FRESHWATER WITHDRAWAL IS  
FOR AGRICULTURAL USE

**37%**

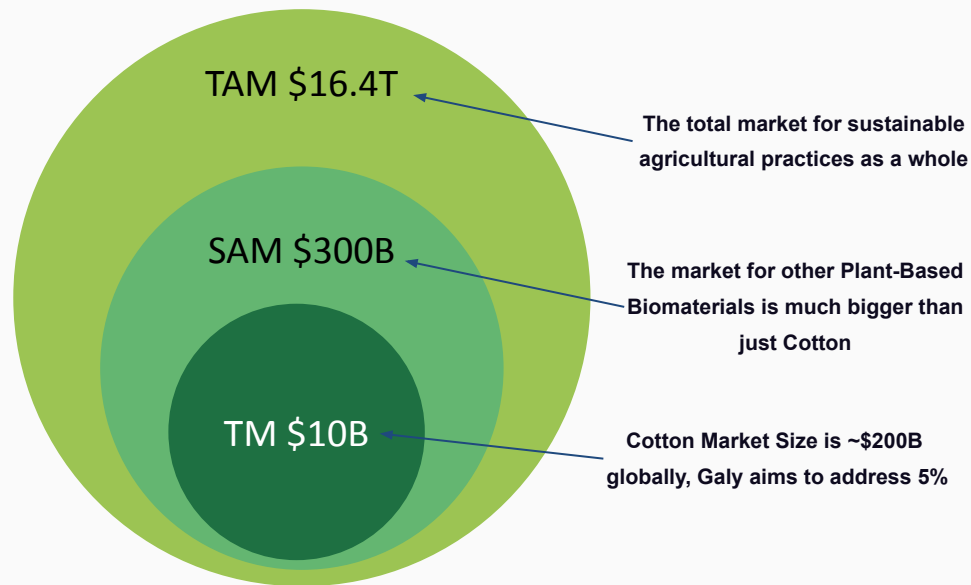
OF EARTH'S TERRESTRIAL  
SURFACE IS OCCUPIED  
BY AGRICULTURE

**33%**

OF THE WORLD'S SOIL IS  
DEGRADED DUE TO  
INTENSIVE AGRICULTURE



## | The Market (2030)





## | Problem Statement

### How can we identify the best crop options for GALY?

GALY is a climate tech company dedicated to developing novel cellular agriculture products. Recognizing the potential of its platform through its work with cotton, GALY has expanded its foundational technology to other crops in industries such as food and beverages, among many other possibilities.

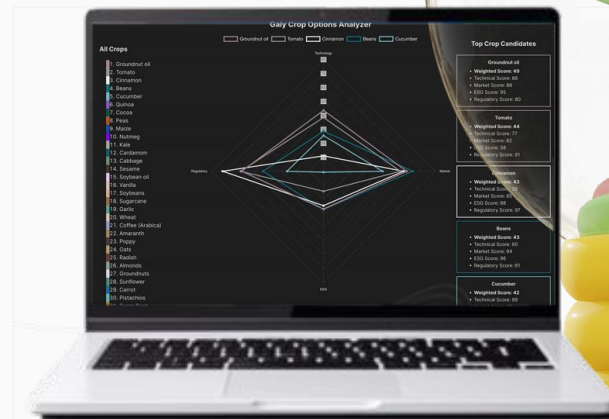


# | Solution

## Introducing GROW

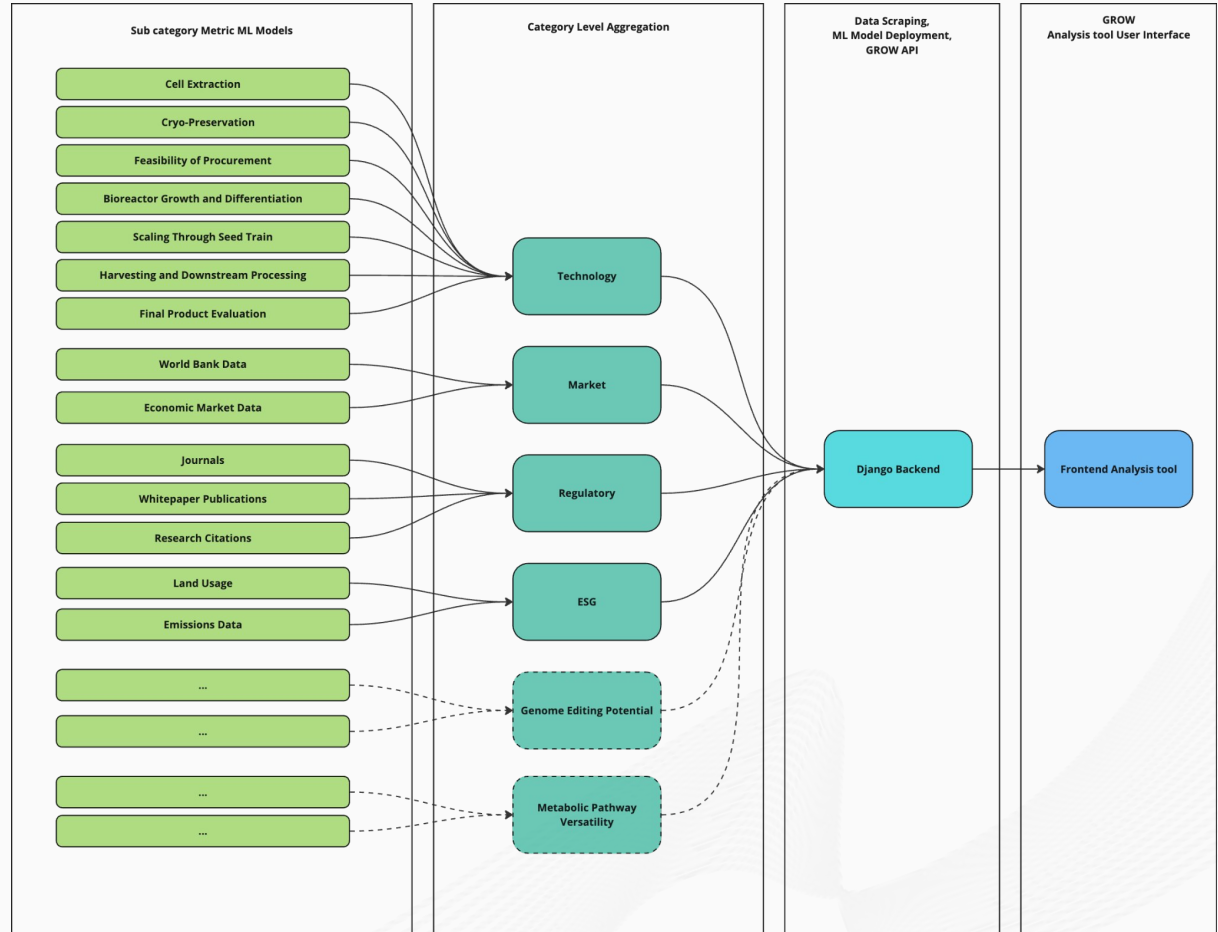
### GALY Resource Optimization Wizard

GROW leverages AI to identify crops for cellular agriculture and makes it easy for GALY researchers to find the optimal crop options



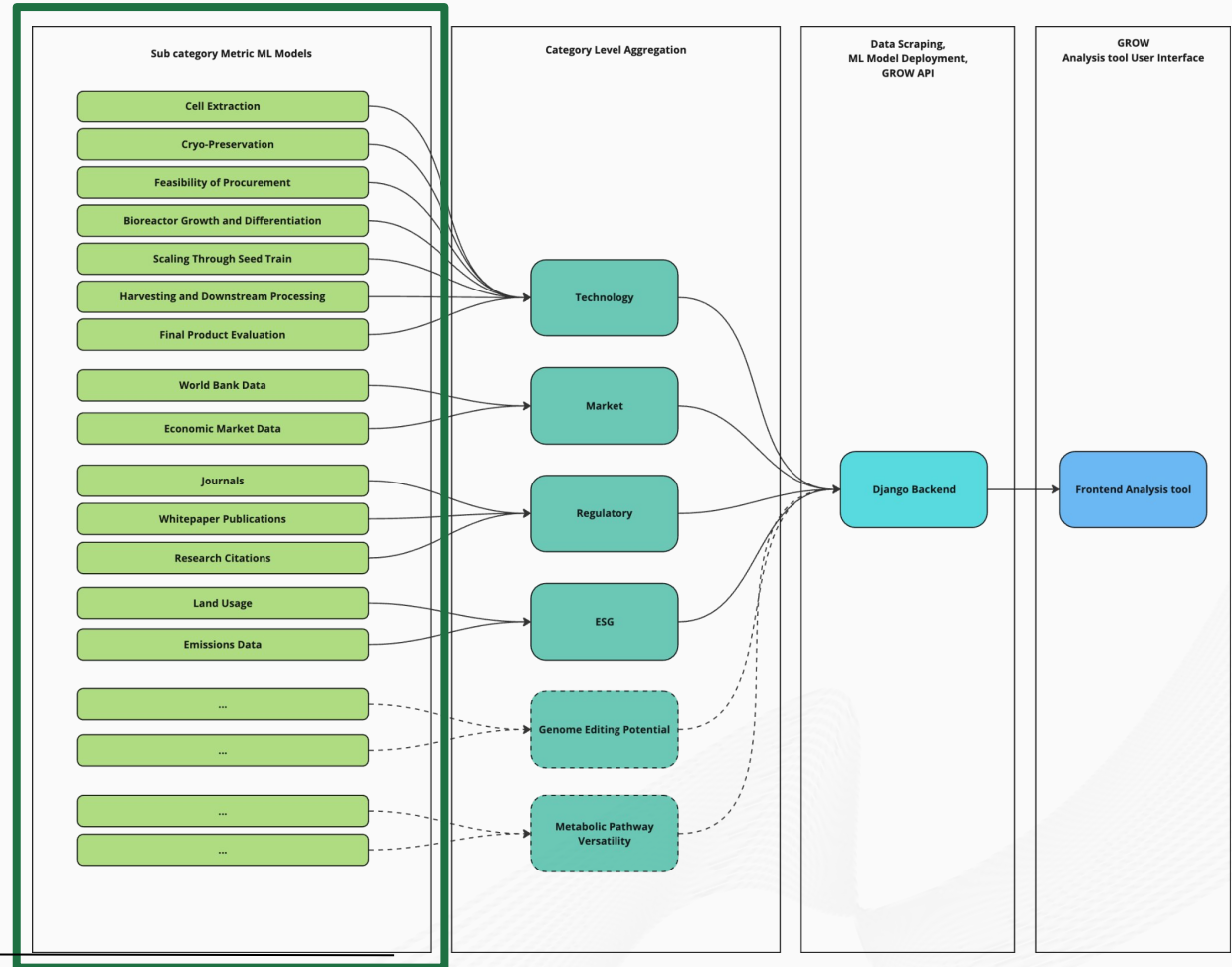
# System Overview

- Identified key metric categories and divided them up into sub categories
- ML Models built for each subcategory
- Category level score aggregated based on the data predictions
- Served through our Django Backend API
- Displayed to the researcher by our frontend



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# Backend - Score estimation for Regulatory attribute

```
[1]: import random
import os
import openai
from openai import OpenAI
import networkx as nx
import json

# Set up your API key
OPENAI_API_KEY = "sk-proj-5GG0RpBmF8fA6zqJLIHqWTIYP61yDALegw2MBw0l8zyAFQNG9MAyDYipe7b0YkuuyxirZVL0DT3BlbkfJ5bk08soxp9K21phY-bCe2GTQVksSYkeFU4"

# Creating randomized batches with intentional duplicates (for variability in rankings)
def create_batches(crops, batch_size=5, duplicate_probability=0.3):
    random.shuffle(crops) # Shuffle the crop list randomly
    batches = [crops[i:i + batch_size] for i in range(0, len(crops), batch_size)]

    # Introduce some duplicates in the batches for variability
    for i in range(len(batches)):
        if random.random() < duplicate_probability: # 30% chance to duplicate a crop
            crop_to_duplicate = random.choice(crops)
            batch_to_duplicate = random.choice(batches)
            batch_to_duplicate.append(crop_to_duplicate)

    return batches

# Function to call OpenAI API
def get_gpt_response(**kwargs):
    openai_client = OpenAI(api_key=OPENAI_API_KEY)
    # Ensure the correct structure of messages
    for message in kwargs.get("messages", []):
        for k in ["function_call", "tool_calls"]:
            if k in message and message[k] is None:
                message.pop(k)

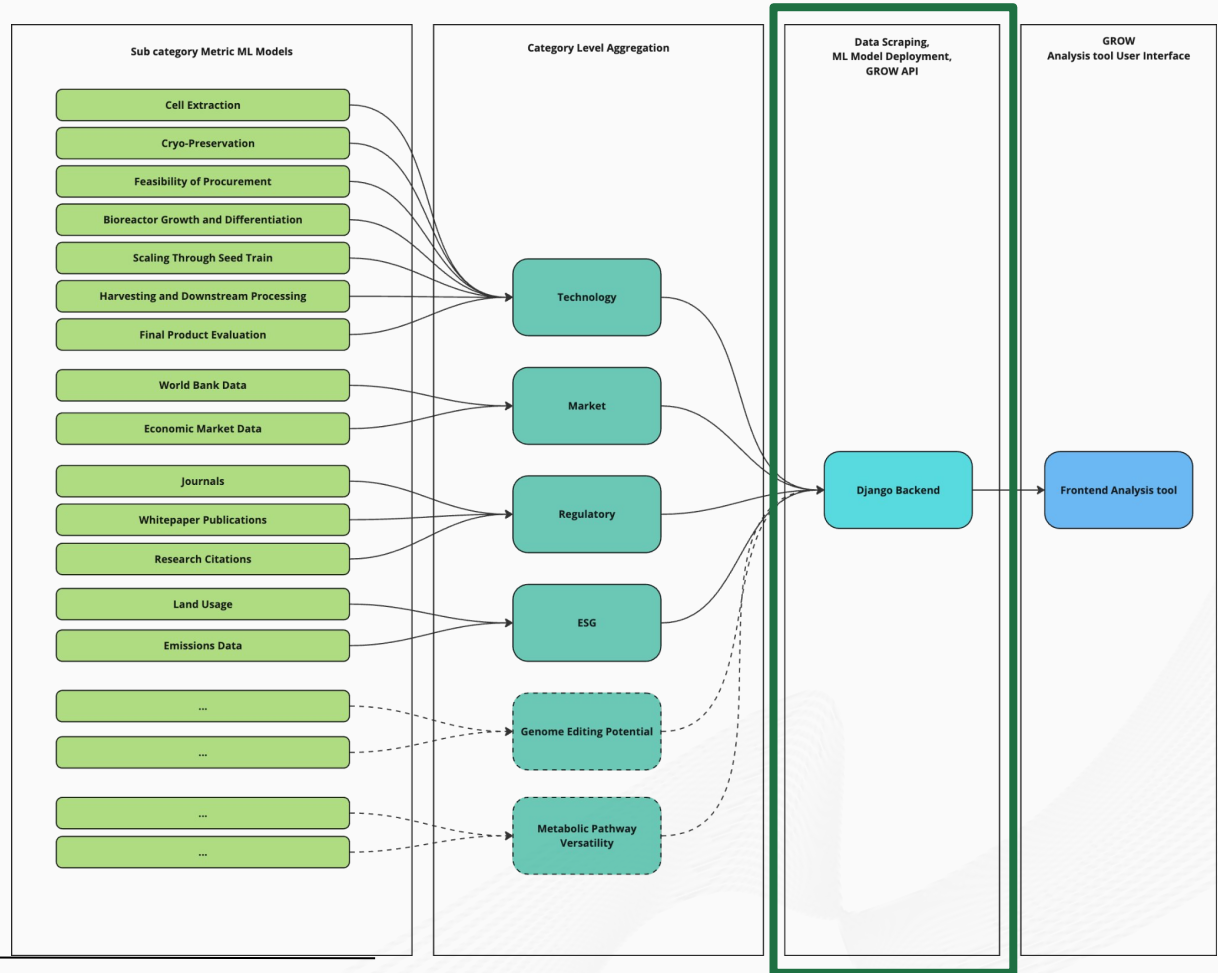
    # Call OpenAI's API
    response = openai_client.chat.completions.create(**kwargs)
    return response
```

We employed a combination of randomization and GPT-based ranking to analyze a list of crops based on regulatory complexity.

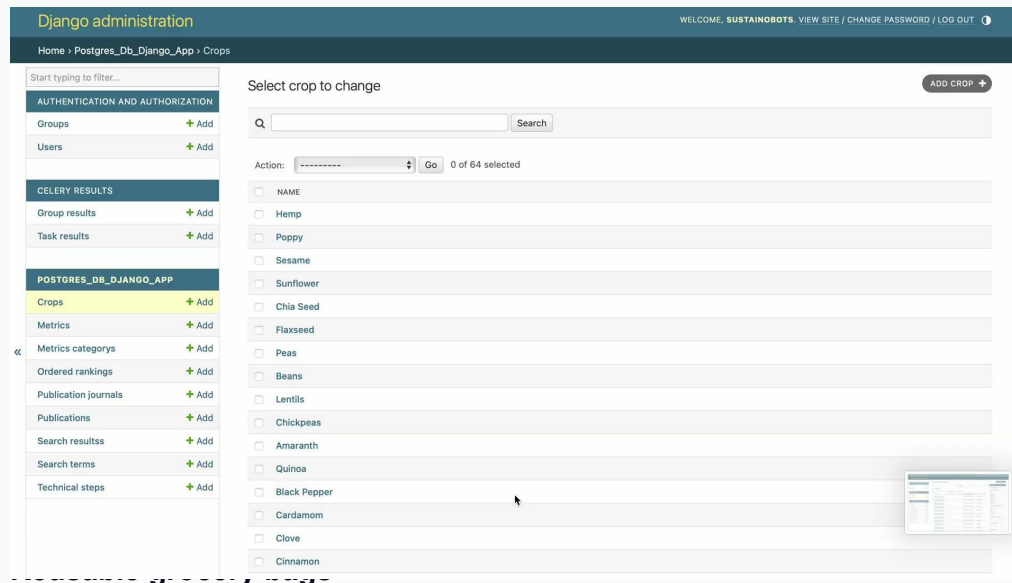
We created randomized batches of crops, ranking them through GPT-4 based API calls, and further refining the rankings using a PageRank algorithm with added random noise to ensure unbiased final scores.

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# Backend - Django

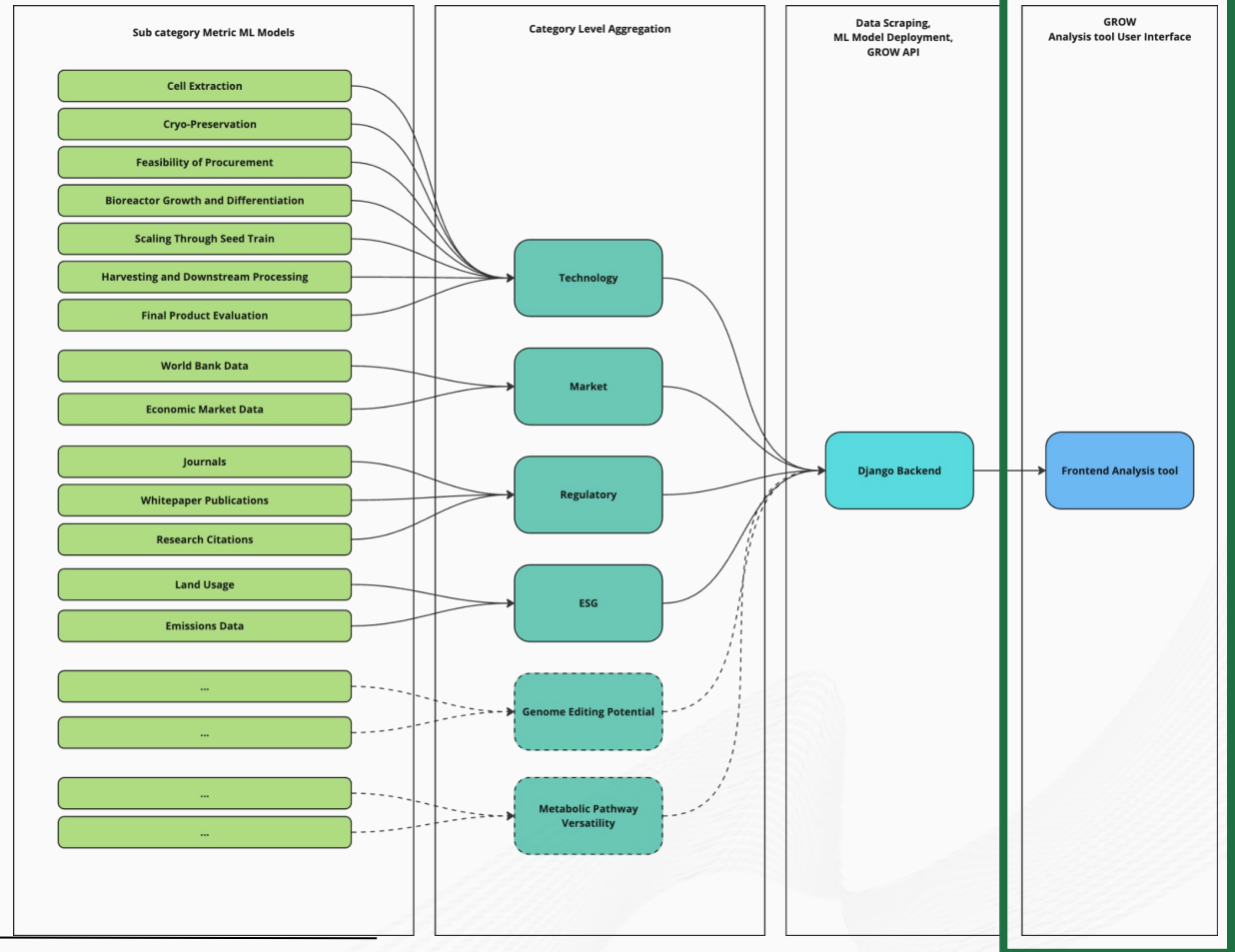


We used Django to create a backend interface to store our researched short list of crops, desired metrics, submetrics and categories, and outputs of web scraping across a list of publications.

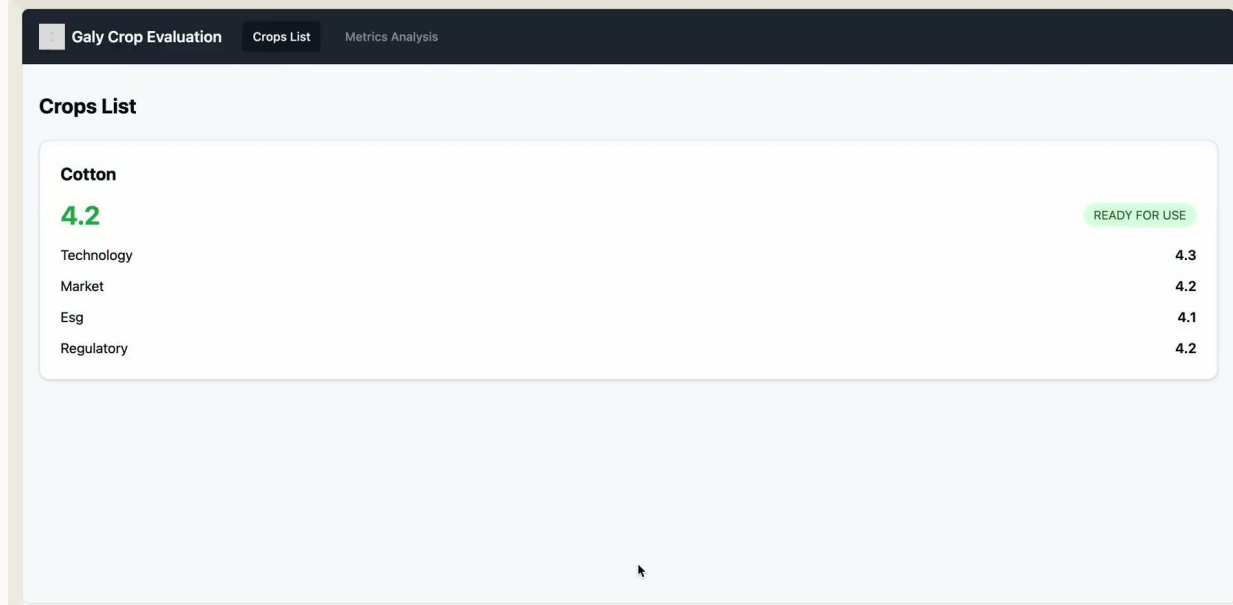
We use keywords pertinent to the crop, metrics and submetrics to extract research papers from these publications.

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# Frontend - Mock Prototype

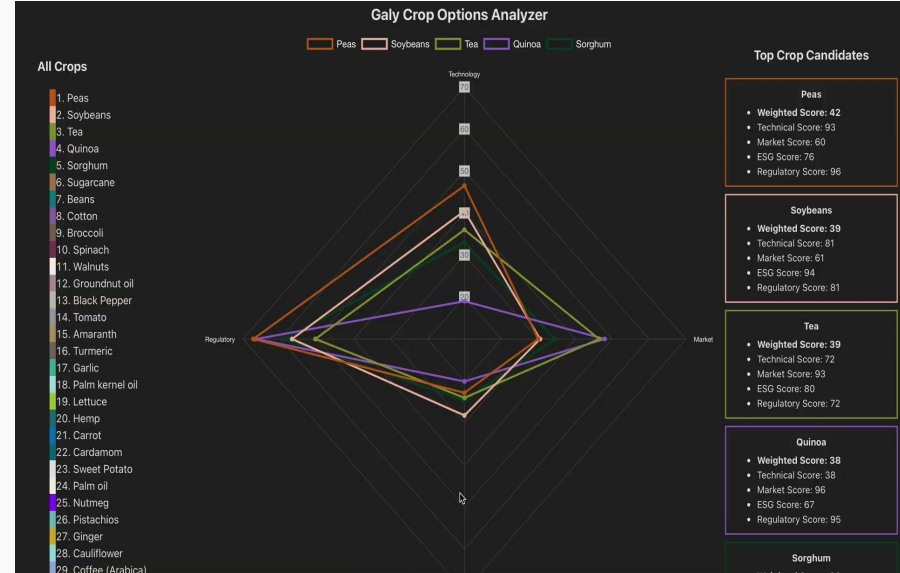
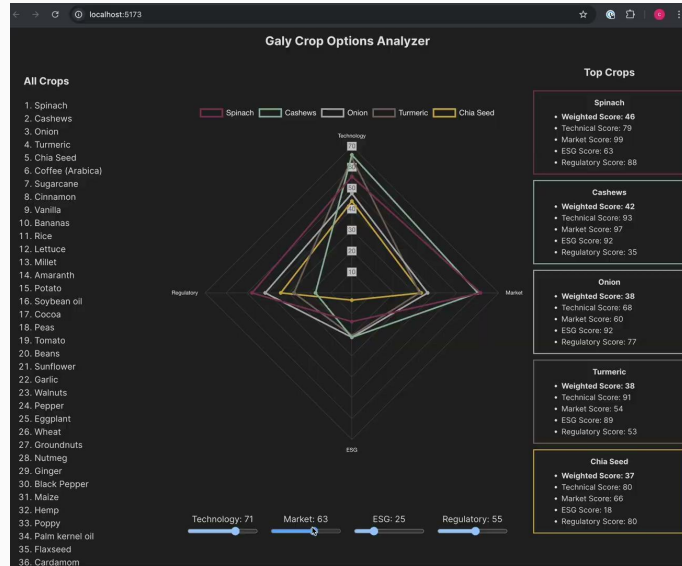


We made a quick mockup of our tool using an AI tool Claude to iterate on what the interface could look like for the analysis tool.

We then built out the tool as a webapp using React.



# Frontend - GALY Crop Options Analyzer



# Features of the Front End



## User Friendly

The user can adjust the importance of each metric in determining the overall score



## Research Backed

Analysis based on research availability on the distinguished science journals online



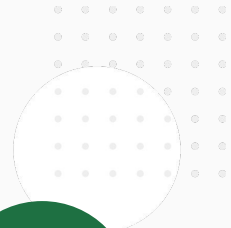
## Use of Keywords

Web scraping depends on keywords that best describe the metrics and sub-metrics



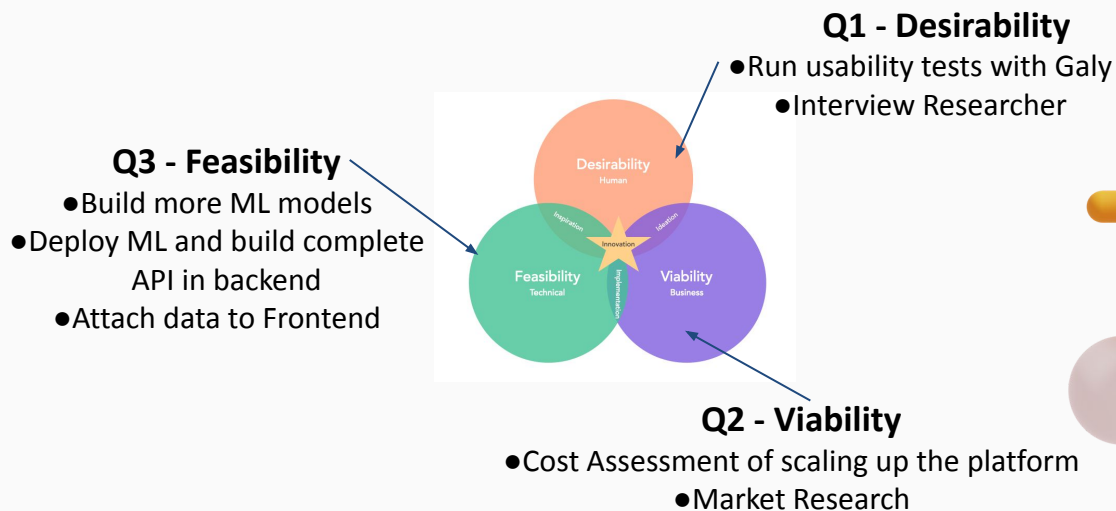
## Ranking

The user can see the top five ranked crops based on our analysis and their assigned weightage to each metric



## Next Steps

### The future looks bright for GROW



| Team: Sustainobots

# Thank You

for doing good for the planet.



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