# **Smart Agriculture Crop Layout Optimization**

#### **Overview**

This project leverages Genetic Algorithms (GA) to optimize crop layouts while considering key farming factors like sunlight, water availability, economic value, and crop compatibility. It helps in maximizing yield and profit by optimizing land usage.

#### **Features**

- Optimized Crop Layout Generation using AI
- Random Layout vs. Optimized Layout Comparison
- Sunlight & Water Heatmaps for better decision-making
- Real-time Fitness & Profit Analysis
- Interactive Crop Selection Feature
- Download Optimized Layout as CSV

## **Technology Stack**

- Python: Data Processing & AI Algorithms
- Streamlit: Interactive Web UI
- Matplotlib & Seaborn: Visualizations
- Pandas & NumPy: Data Processing
- Genetic Algorithms (GA): Crop Layout Optimization

## **Installation & Setup**

1. Clone the Repository:

```
git clone https://github.com/yourusername/smart-agriculture-optimization.git cd smart-agriculture-optimization
```

2. Install Dependencies:

```
pip install -r requirements.txt
```

3. Run the Streamlit App:

```
streamlit run main.py
```

#### code

```
import pandas as pd
import numpy as np
import random
import matplotlib.pyplot as plt
import streamlit as st
import seaborn as sns
import time
# Load dataset
file_path = 'synthetic_agriculture_field_large.csv'
dataset = pd.read_csv(file_path)
# Add price per unit for cost & profit analysis
dataset["Price_per_unit"] = np.random.randint(50, 200, size=len(dataset)) # Random price
for each crop
# Define category mappings
category_mapping = {
  'Sunlight': {'high': 2, 'medium': 1, 'low': 0},
  'Water': {'high': 2, 'medium': 1, 'low': 0},
  'Soil_type': {'loamy': 2, 'clay': 1, 'sandy': 0},
  'Water_efficiency': {'high': 2, 'medium': 1, 'low': 0},
  'Temperature_range': {'hot': 2, 'moderate': 1, 'cool': 0},
  'Pest_resistance': {'high': 2, 'medium': 1, 'low': 0},
```

```
'Crop_rotation': {'yes': 1, 'no': 0},
  'Harvest_frequency': {'biannual': 2, 'annual': 1}
}
for column, mapping in category_mapping.items():
  dataset[column] = dataset[column].map(mapping)
rows = st.number_input("Enter number of rows:", min_value=5, max_value=50, value=10,
step=1)
cols = st.number_input("Enter number of columns:", min_value=5, max_value=50, value=10,
step=1)
population_size = st.slider('Population Size', 50, 500, 100)
generations = st.slider('Generations', 10, 500, 200)
mutation_rate = st.slider('Mutation Rate', 0.01, 0.2, 0.02)
crossover_rate = st.slider('Crossover Rate', 0.5, 1.0, 0.9)
compatibility_scores = {('Beans', 'Corn'): 1, ('Lettuce', 'Wheat'): -1}
sunlight_zones = np.random.choice([0, 1, 2], size=(rows, cols))
water_zones = np.random.choice([0, 1, 2], size=(rows, cols))
def fitness(layout):
  score = 0
  economic_value = 0
  total_profit = 0
```

```
grid = np.array(layout).reshape(rows, cols)
 for i in range(rows):
    for j in range(cols):
      crop_index = grid[i, j]
      crop_info = dataset.iloc[crop_index]
      if crop_info['Sunlight'] == sunlight_zones[i, j]: score += 2
      if crop_info['Water'] == water_zones[i, j]: score += 2
      economic_value += crop_info['Economic_value']
      total_profit += crop_info['Economic_value'] * crop_info['Price_per_unit']
      if j < cols - 1:
        neighbor_crop_index = grid[i, j + 1]
        neighbor_crop_name = dataset.iloc[neighbor_crop_index]['Crop']
        crop_name = dataset.iloc[crop_index]['Crop']
        score += compatibility_scores.get((crop_name, neighbor_crop_name), 0)
 return score, economic_value, total_profit
# Generate a Completely Random Layout
def generate_random_layout():
 return np.array(random.choices(range(len(dataset)), k=rows * cols)).reshape(rows, cols)
# Genetic Algorithm Functions
```

```
def generate_population(pop_size):
 return [random.choices(range(len(dataset)), k=rows * cols) for _ in range(pop_size)]
def tournament_selection(population, tournament_size=3):
 selected = random.sample(population, tournament_size)
 return sorted(selected, key=lambda x: fitness(x)[0], reverse=True)[:2]
def crossover(parent1, parent2):
 point = random.randint(1, rows * cols - 1)
 return parent1[:point] + parent2[point:]
def mutate(child):
 for i in range(len(child)):
   if random.random() < mutation_rate:</pre>
      child[i] = random.choice(range(len(dataset)))
 return child
def genetic_algorithm():
 population = generate_population(population_size)
 best_fitness = []
 best_value = []
 total_profit = []
 for generation in range(generations):
   new_population = []
   while len(new_population) < population_size:
```

```
parent1, parent2 = tournament_selection(population)
      child = crossover(parent1, parent2) if random.random() < crossover_rate else</pre>
parent1
      child = mutate(child)
      new_population.append(child)
    population = new_population
   best_individual = max(population, key=lambda x: fitness(x)[0])
   best_fit, best_econ_value, best_profit = fitness(best_individual)
   best_fitness.append(best_fit)
   best_value.append(best_econ_value)
   total_profit.append(best_profit)
 return np.array(best_individual).reshape(rows, cols), best_fitness, total_profit
if st.button("Generate Optimized Layout"):
  optimized_layout, fitness_progress, profit_progress = genetic_algorithm()
 # Display Heatmaps BEFORE the layout
 st.write(" **Sunlight Heatmap**")
 plt.figure(figsize=(8, 6))
 sns.heatmap(sunlight_zones, cmap="YlOrBr", annot=True, fmt=".0f")
 st.pyplot(plt)
 st.write("♦ **Water Availability Heatmap**")
 plt.figure(figsize=(8, 6))
```

```
sns.heatmap(water_zones, cmap="Blues", annot=True, fmt=".0f")
 st.pyplot(plt)
 # Display Optimized Layout
 st.subheader(" **Optimized Crop Layout**")
 fig, ax = plt.subplots(figsize=(10, 10))
 crop_colors = {'Wheat': 'gold', 'Lettuce': 'green', 'Corn': 'yellow', 'Beans': 'brown'}
 ax.set_xlim(0, cols)
 ax.set_ylim(0, rows)
 ax.set_xticks([])
 ax.set_yticks([])
 ax.set_aspect('equal', 'box')
 for i in range(rows):
   for j in range(cols):
      crop_index = optimized_layout[i, j]
      crop_name = dataset.iloc[crop_index]['Crop']
      ax.add_patch(plt.Rectangle((j, i), 1, 1, color=crop_colors[crop_name],
edgecolor='black'))
      ax.text(j + 0.5, i + 0.5, crop_name, ha='center', va='center', fontsize=8, color='black')
 st.pyplot(fig)
 # Compare with Random Layout
 st.subheader(" **Comparison: Random vs Optimized Layout**")
```

```
random_layout = generate_random_layout()
  fig, axes = plt.subplots(1, 2, figsize=(16, 8))
  def plot_layout(ax, layout, title):
    ax.set_xlim(0, cols)
    ax.set_ylim(0, rows)
    ax.set_xticks([])
    ax.set_yticks([])
    ax.set_aspect('equal', 'box')
    ax.set_title(title)
    for i in range(rows):
      for j in range(cols):
        crop_index = layout[i, j]
        crop_name = dataset.iloc[crop_index]['Crop']
        ax.add_patch(plt.Rectangle((j, i), 1, 1, color=crop_colors[crop_name],
edgecolor='black'))
        ax.text(j + 0.5, i + 0.5, crop_name, ha='center', va='center', fontsize=8, color='black')
  plot_layout(axes[0], random_layout, " Random Layout (Before Optimization)")
  plot_layout(axes[1], optimized_layout, " Optimized Layout (After AI Processing)")
  st.pyplot(fig)
  # [11] Calculate fitness, economic value, and profit for the random layout
  random_fitness, random_economic_value, random_total_profit = fitness(random_layout)
```

```
# [11] Calculate fitness, economic value, and profit for the optimized layout
  optimized_fitness, optimized_economic_value, optimized_total_profit =
fitness(optimized_layout)
 # Display insights before vs. after optimization
 st.subheader(" **Insights: Before vs. After Optimization**")
 col1, col2 = st.columns(2)
 with col1:
   st.write("### **Before Optimization (Random Layout)**")
   st.write(f"♦ **Fitness Score:** {random_fitness}")
   st.write(f" **Total Economic Value:** {random_economic_value}")
    st.write(f" **Total Profit:** ${random_total_profit}")
 with col2:
   st.write("###  **After Optimization (AI-Optimized Layout)**")
    st.write(f"♦ **Fitness Score:** {optimized_fitness} (**↑ {((optimized_fitness -
random_fitness) / random_fitness) * 100:.2f}%**)")
   st.write(f" **Total Economic Value: ** {optimized_economic_value} (**1
{((optimized_economic_value - random_economic_value) / random_economic_value) *
100:.2f}%**)")
   st.write(f" **Total Profit:** ${optimized_total_profit} (**1 {((optimized_total_profit -
random_total_profit) / random_total_profit) * 100:.2f\%**)")
```

```
# Save Optimized Layout for Download
optimized_layout_df = pd.DataFrame(optimized_layout)
optimized_layout_df.to_csv("optimized_layout.csv", index=False)
st.download_button("Download Optimized Layout", "optimized_layout.csv")
```

## License

This project is open-source under the MIT License.

# **Developer & Contact Info**

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GitHub Repository: https://github.com/subanaveen/Insyde.Io