## GENITIC ALGORITHM FOR OPTIMIZATION

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import cv2
import numpy as np
import random
import matplotlib.pyplot as plt
# Genetic Algorithm Parameters
POP SIZE = 20
                     # Population size
N_GENERATIONS = 30 # Number of generations
MUTATION RATE = 0.2 # Probability of mutation
IMG PATH = r"C:\Users\prami\OneDrive\Documents\Family.jpg"
def fitness(image, threshold):
  """Calculate fitness based on between-class variance (Otsu's method)."""
  _, binary = cv2.threshold(image, threshold, 255, cv2.THRESH_BINARY)
  hist = cv2.calcHist([image], [0], None, [256], [0,256]).ravel()
  total pixels = image.size
  w0 = np.sum(hist[:threshold]) / total pixels
  w1 = np.sum(hist[threshold:]) / total pixels
  m0 = np.sum([i*hist[i] for i in range(threshold)]) / (np.sum(hist[:threshold])+1e-6)
  m1 = np.sum([i*hist[i] for i in range(threshold, 256)]) / (np.sum(hist[threshold:])+1e-6)
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variance = w0 * w1 * (m0 - m1) ** 2
  return variance
def initialize_population():
  return [random.randint(0, 255) for in range(POP SIZE)]
def selection(population, scores):
  """Roulette wheel selection."""
  scores = np.array(scores)
  probs = scores / (scores.sum() + 1e-6)
  idx = np.random.choice(range(POP SIZE), size=2, p=probs)
  return population[idx[0]], population[idx[1]]
def crossover(p1, p2):
  """Single point crossover."""
  point = random.randint(0, 7)
  mask = (1 << point) - 1
  child1 = (p1 \& mask) | (p2 \& \sim mask)
  child2 = (p2 \& mask) | (p1 \& \sim mask)
  return child1, child2
def mutation(threshold):
  """Random bit flip mutation."""
  if random.random() < MUTATION RATE:
    bit = 1 \ll random.randint(0, 7)
    threshold ^= bit
  return max(0, min(255, threshold))
```

```
# Main GA Optimization
def genetic_thresholding(image):
  population = initialize population()
  for generation in range(N GENERATIONS):
    scores = [fitness(image, t) for t in population]
    new population = []
    for in range(POP SIZE // 2):
      p1, p2 = selection(population, scores)
      c1, c2 = crossover(p1, p2)
      new population.append(mutation(c1))
      new population.append(mutation(c2))
    population = new_population
    best score = max(scores)
    best threshold = population[np.argmax(scores)]
         print(f'Gen {generation+1}: Best Threshold = {best threshold}, Fitness =
{best score:.2f}")
  return best threshold
if name == " main ":
  # Load image in grayscale
  img = cv2.imread(IMG PATH, cv2.IMREAD GRAYSCALE)
  best t = genetic thresholding(img)
```

```
# Apply best threshold
  _, final_img = cv2.threshold(img, best_t, 255, cv2.THRESH_BINARY)
  # Show result
  plt.subplot(1,2,1), plt.imshow(img, cmap="gray"), plt.title("Original")
  plt.subplot(1,2,2), plt.imshow(final img, cmap="gray"), plt.title(f'GA Threshold {best t}")
  plt.show()
output:
Gen 1: Best Threshold = 73, Fitness = 2550.92
Gen 2: Best Threshold = 6, Fitness = 2550.92
Gen 3: Best Threshold = 146, Fitness = 2543.74
Gen 4: Best Threshold = 203, Fitness = 2527.72
Gen 5: Best Threshold = 134, Fitness = 2543.74
Gen 6: Best Threshold = 70, Fitness = 2543.74
Gen 7: Best Threshold = 114, Fitness = 2549.20
Gen 8: Best Threshold = 66, Fitness = 2552.17
Gen 9: Best Threshold = 82, Fitness = 2552.17
Gen 10: Best Threshold = 82, Fitness = 2543.74
Gen 11: Best Threshold = 144, Fitness = 2543.74
Gen 12: Best Threshold = 178, Fitness = 2543.74
Gen 13: Best Threshold = 178, Fitness = 2408.85
Gen 14: Best Threshold = 210, Fitness = 2408.85
Gen 15: Best Threshold = 147, Fitness = 2537.01
Gen 16: Best Threshold = 146, Fitness = 2537.01
Gen 17: Best Threshold = 128, Fitness = 2537.01
Gen 18: Best Threshold = 148, Fitness = 2537.01
Gen 19: Best Threshold = 144, Fitness = 2537.01
```

- Gen 20: Best Threshold = 184, Fitness = 2537.01
- Gen 21: Best Threshold = 136, Fitness = 2537.01
- Gen 22: Best Threshold = 144, Fitness = 2537.01
- Gen 23: Best Threshold = 183, Fitness = 2537.01
- Gen 24: Best Threshold = 140, Fitness = 2537.01
- Gen 25: Best Threshold = 182, Fitness = 2537.01
- Gen 26: Best Threshold = 170, Fitness = 2537.01
- Gen 27: Best Threshold = 138, Fitness = 2532.83
- Gen 28: Best Threshold = 136, Fitness = 2527.72
- Gen 29: Best Threshold = 136, Fitness = 2488.61
- Gen 30: Best Threshold = 154, Fitness = 2527.72