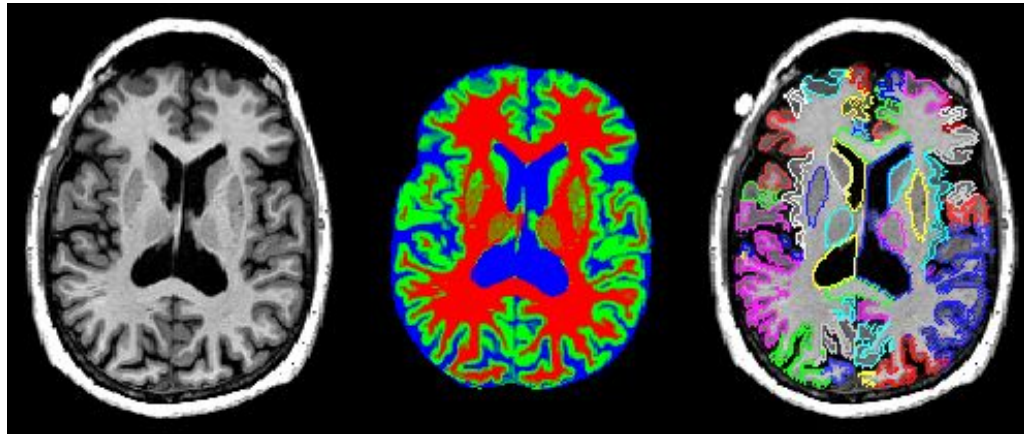


# Optimization-based Segmentation of Brain Using a Genetic Algorithm

Tim Ruesink, Grant Roberts, Lawrence Lechuga

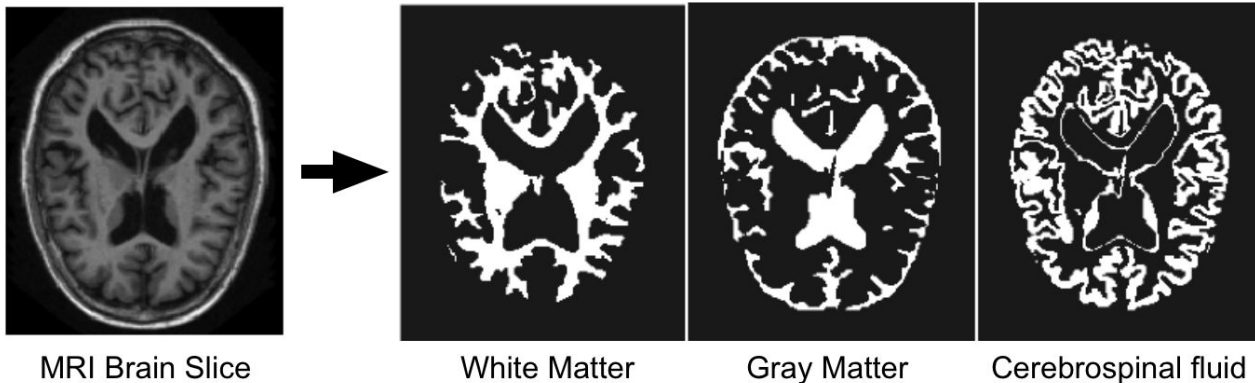


# Background

**Segmentation:** Divide an image into set of meaningful, homogenous, and non-overlapping regions, based on attributes in the image.

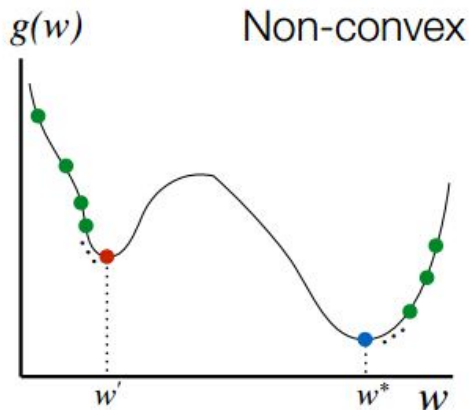
**Challenges:** Partial volume effect, artifacts (motion, etc), complex geometry of objects in question, and noise adds uncertainty to the process.

**Common Methods:** Thresholding, k-means clustering, atlas, and manual methods



# Motivation

Find an optimization method that performs well in a non-convex solution space.



Specifically, we want to compare a gradient-based approach to a genetic algorithm approach using the Pott's model as our formulation.

# Objective Function: Potts Model

- Formulation:  $\text{minimize } \gamma \|\nabla \mathbf{x}\|_0 + \|\mathbf{x} - \mathbf{y}\|_2^2$
- Jump Penalty: Forces Piecewise constant solutions  $\gamma \|\nabla \mathbf{x}\|_0$
- Data term: Couples minimizing candidate  $\mathbf{x}$  to data  $\mathbf{y}$   $\|\mathbf{x} - \mathbf{y}\|_2^2$

Non-Convex

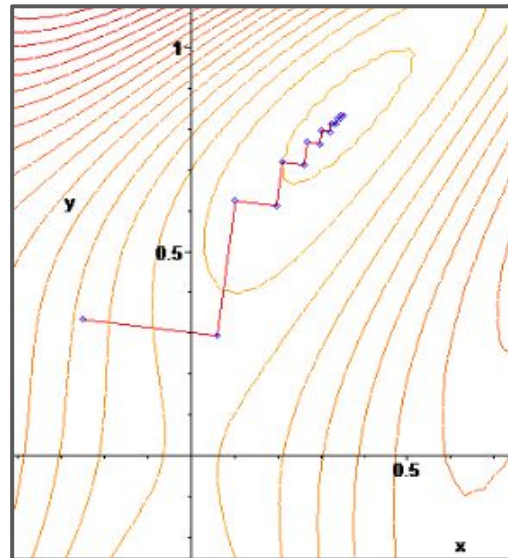
# Gradient-Based Optimization Segmentation

**How do Gradient-Based Methods work?** Take a step proportional to the gradient of the cost function.

**What are common gradient-based methods?** Steepest descent, conjugate gradients, Newton's method, Quasi-Newton

**Advantages:** Generality and simplicity, can handle many types of cost functions

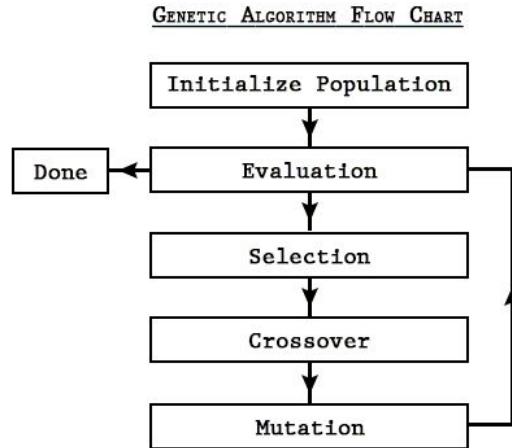
**Drawbacks:** Converge slowly, sensitive to local minima



# Genetic Algorithm Segmentation

**What are genetic algorithms?** Used to find global minimum solution for non-linear or non-convex optimization problems

**How do Genetic algorithms work?** Conceptualized using evolutionary biology



**Selection:** Keep best performing Solutions (Parents)

**Crossover:** Create new solutions (Children) from the Successful Parents

**Mutation:** Create new solutions (Children) by taking successful parent and mutating certain variable to take on random values.

FIGURE 2

# Analysis and Comparison

1. Shepp-Logan Phantom segmentation using:
  - a. Gradient-Based Algorithm
  - b. Genetic Algorithm
2. Computational Time
3. Performance/Correctness
  - a. Qualitative Results
  - b. DICE Coefficient

$$DC = \frac{2|\hat{A} \cap A^*|}{|\hat{A} \cup A^*|}$$



# Objectives

1. Understanding and modeling the segmentation problem using the Potts Model (Definition of cost function)
2. Learn how to implement a Gradient-Based Algorithm for segmentation optimization
3. Learn how to implement a Genetic Algorithm for segmentation optimization
4. Use both algorithm for a realistic medical imaging application
5. Compare the performance of each algorithm



Questions