

# Genetic Algorithms for Vision and Pattern Recognition

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# Objective

- To solve for optimization of computer vision problems using genetic algorithms

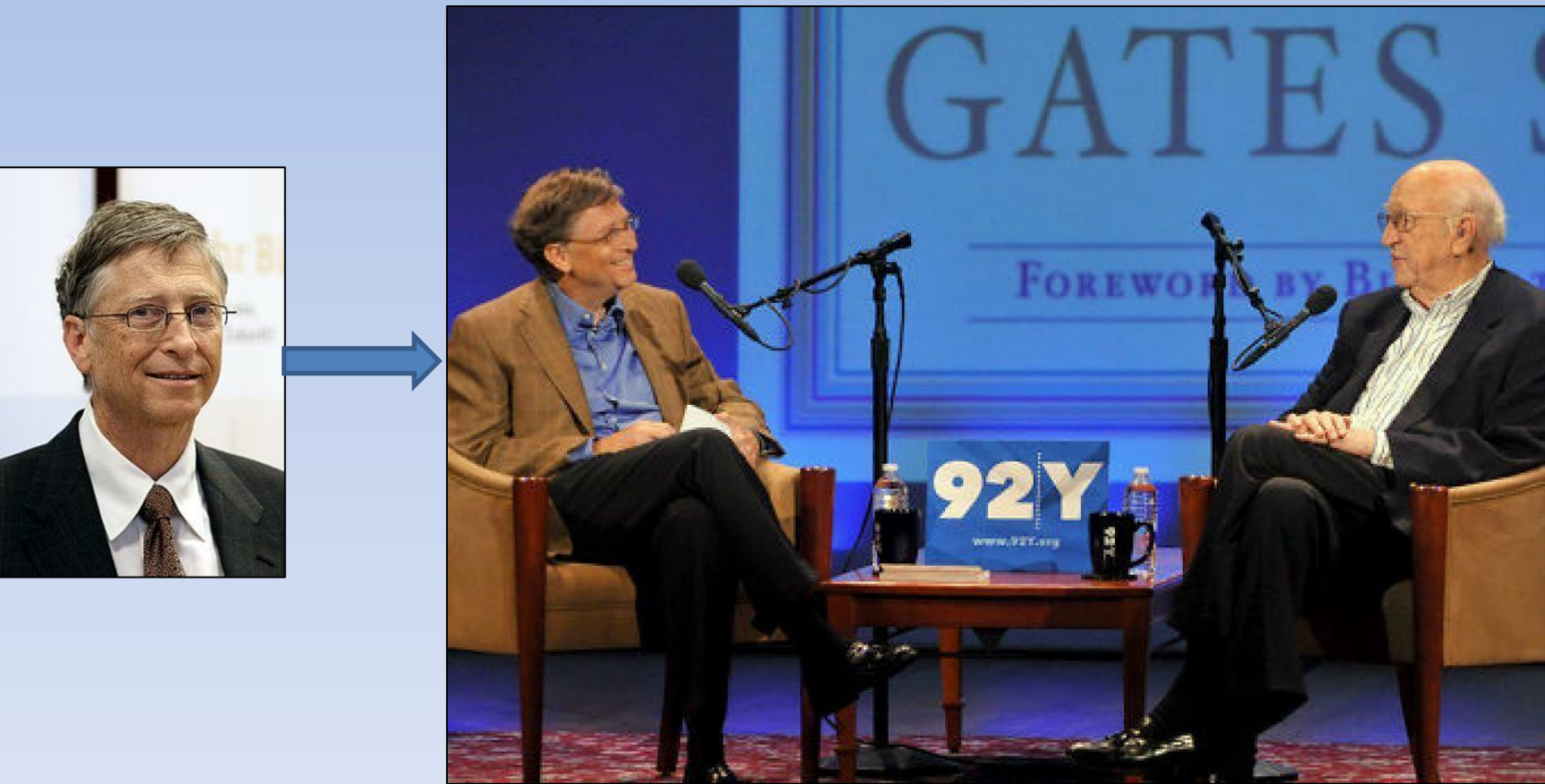
# Timeline

- Problem: Computer Vision
- Genetic Algorithms(GA)
- Solution by Genetic Algorithms
- Results and Discussions
- Questions?

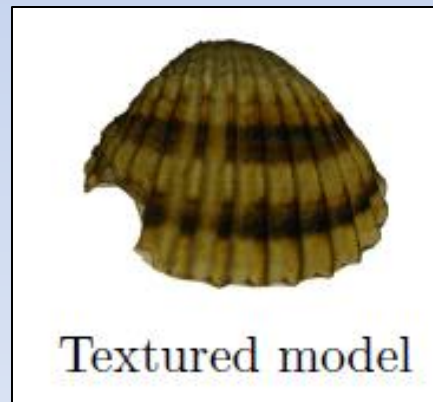
# Registration of Images

- Registration Problem is one of the fundamental problems in computer vision

# Registration of Images to Images



# Registration of Images to 3D Models

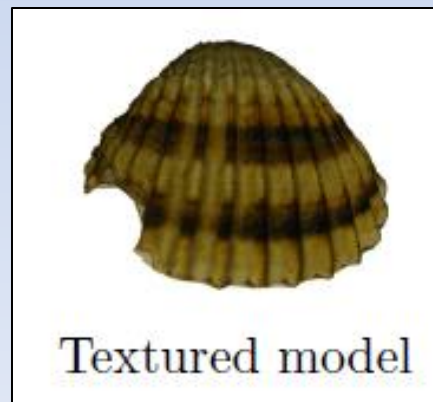


# Applications

- Augmented Reality
- Image Guided Surgery
- Rendering real objects in gaming environments

# Registration of Images to 3D Models[1]

- Goal: make textured 3D models



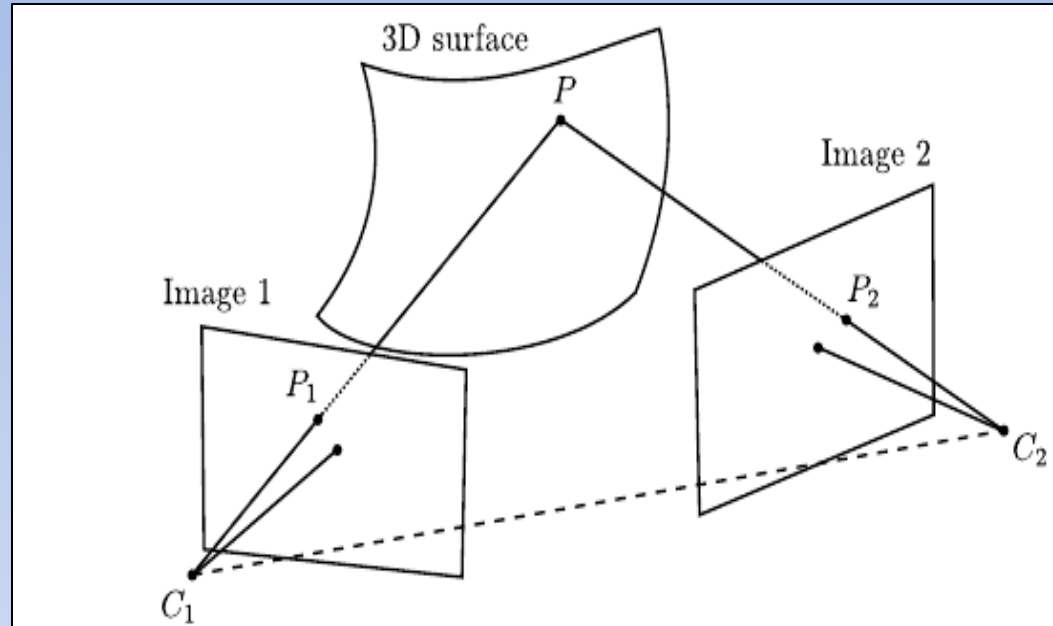


# How to do that?

- Sampling: Slice the 3D model
- Define cost function
- Solve for the optimization

# Define Cost Function

- 18 parameters
- $K(3 \times 2)$
- $R(3 \times 2)$
- $T(3 \times 2)$



**Figure 4.3:** Illustration of the basic principles of shape-from-stereo (stereo reconstruction). The reconstruction of a 3D point may be achieved using the epipolar geometry.

# Cost Function

$$C(P_1, P_2) = \frac{1}{|P|} \sum_{X \in P} (I_1(P_1 \mathbf{X}) - I_2(P_2 \mathbf{X}))^2$$

- $P_1, P_2$  are the projection matrices
- $P$  is the number of points in 3D model
- $I_i(P\mathbf{X})$  is the color of point  $X$  on the image

# Why genetic algorithms?

- Unpredictable shape of cost function
- Local minima failure by using:
  - Newton method
  - Levenberg-Marquardt(LM)
  - BGFS variable metric method

# Genetic Algorithms[2][3]

Review

# Theory Of Evolution

- Organisms evolve to fit into the environment
- Only the best individuals are kept by nature

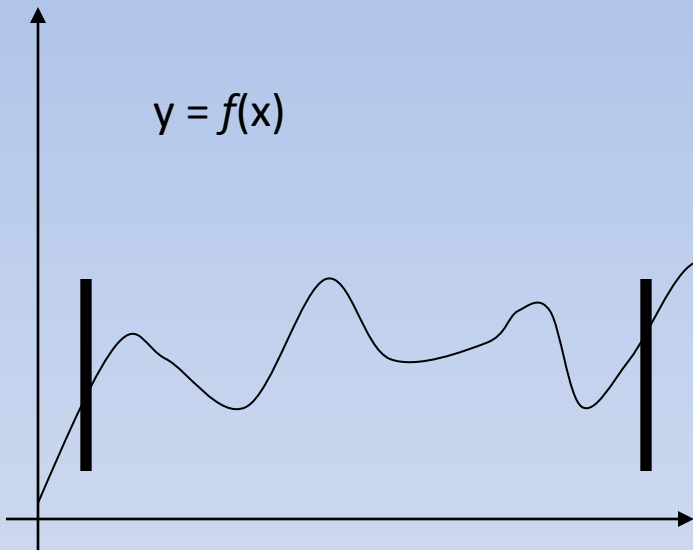
Oh seriously...

- From a set of random solutions only the best ones are picked

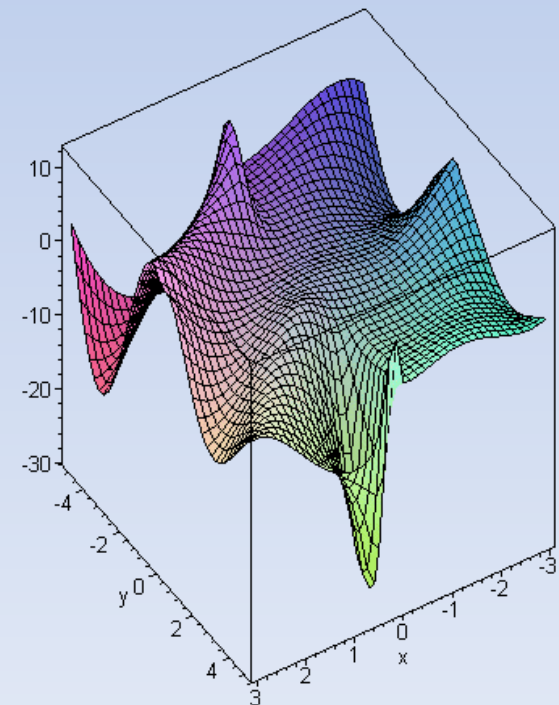
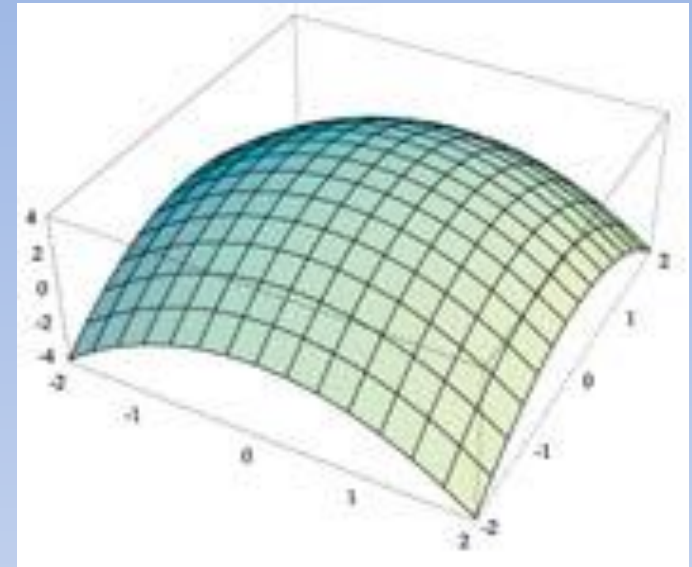
# Purpose of Genetic Algorithms

“Genetic Algorithms are good at taking large, potentially huge search spaces and navigating them, looking for optimal combinations of things, solutions you might not otherwise find in a lifetime.”

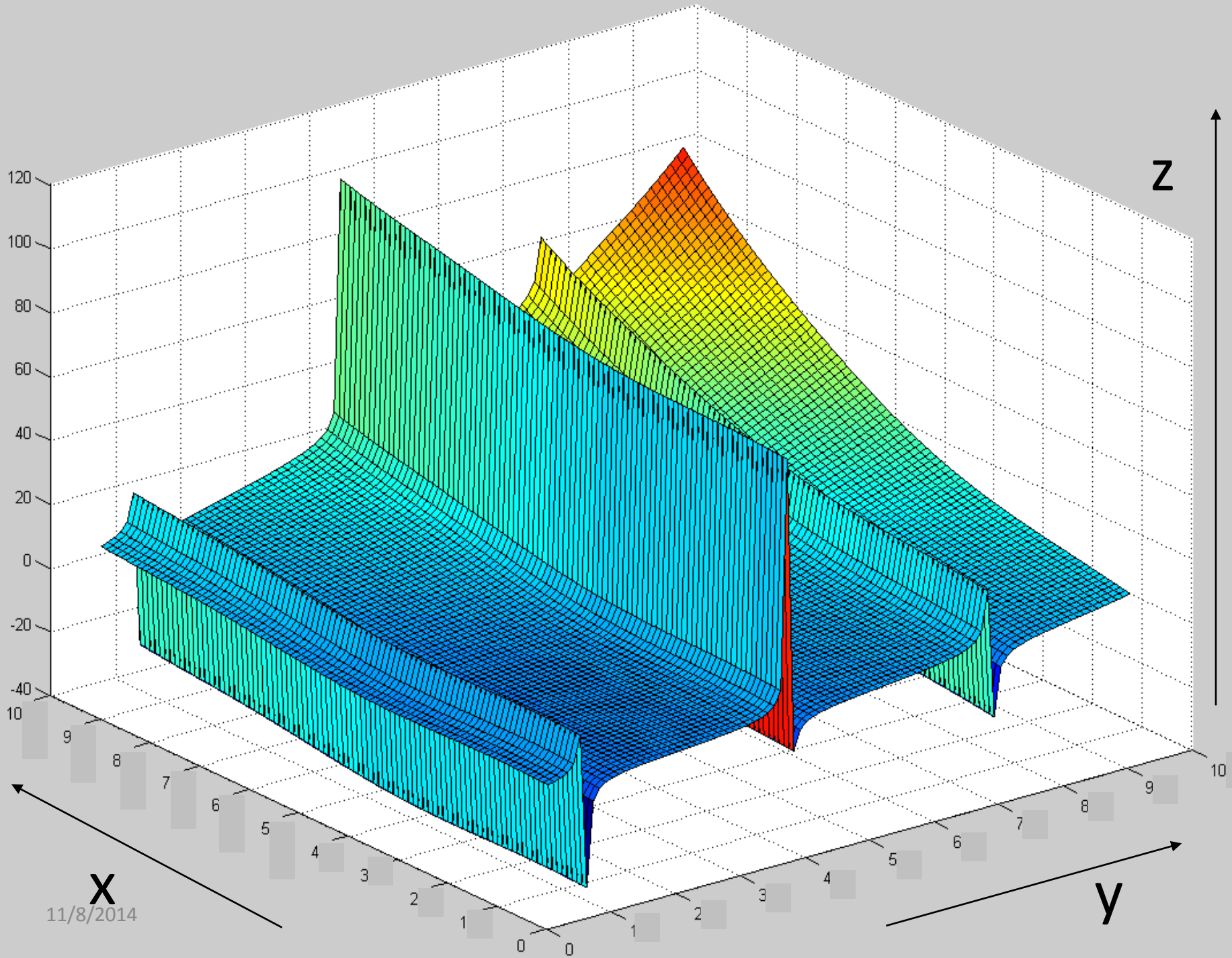
# Example Problem I



Finding the maximum (minimum) of some function (within a defined range).







**X**  
11/8/2014

**y**

**z**

# Problem?

- Numbers from 1..100
- Find the set of numbers that give a sum of 313
- Any ideas??

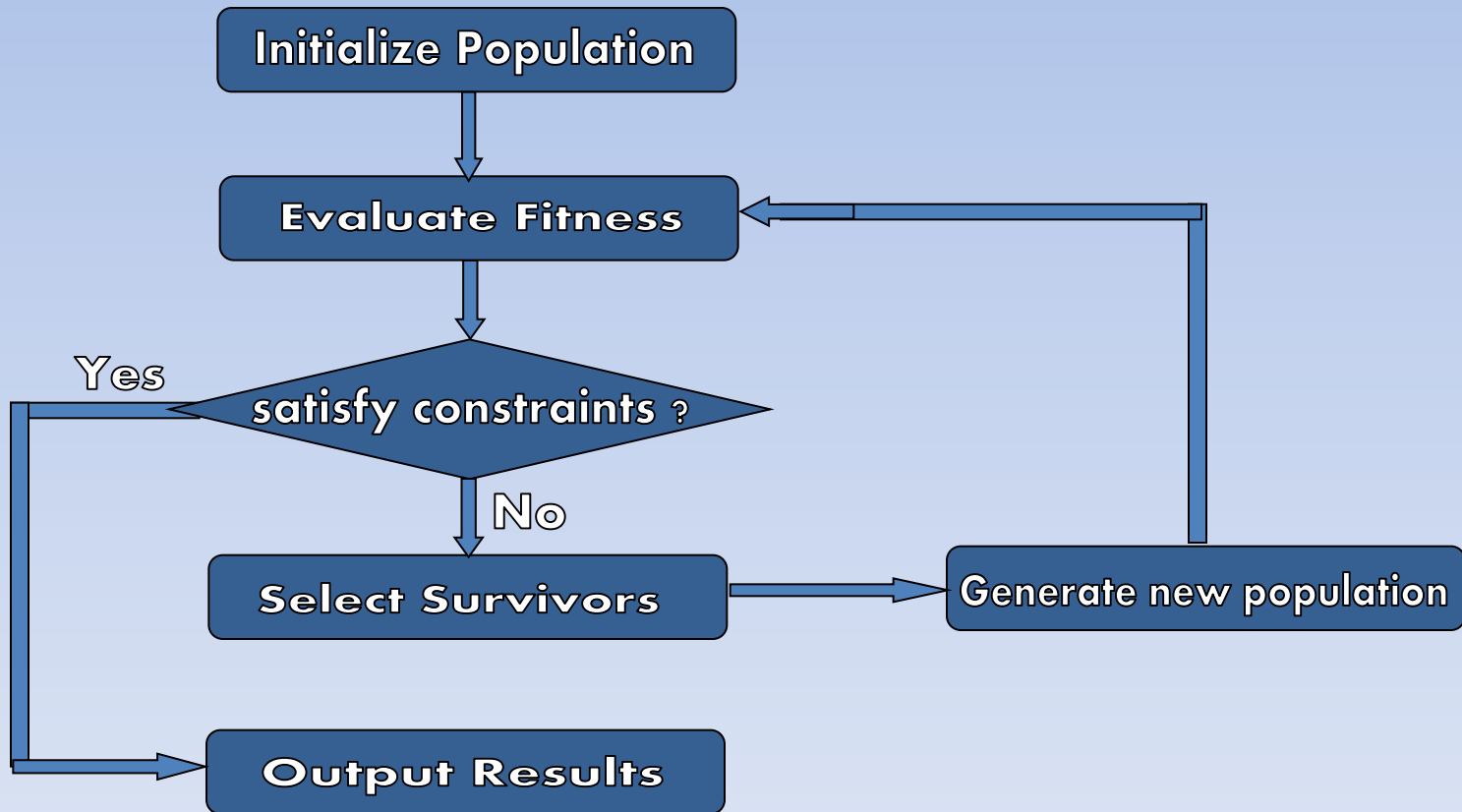
# Genetics - Promo

- A **gene** is hereditary unit of inheritance
- Multiple **genes** are stringed together to form **chromosomes**
- A gene, if expressed in an organism is called a **trait**
- Offsprings **inherit** traits from their parents
- A gene may get **mutated** during mating process

# How is the process done?

- Genetic algorithm (GA) introduces
  - the principle of evolution genetics into search among possible solutions to given problem
- To simulate the process in natural systems
- How: by the creation within a machine a population of individuals

# Genetic Algorithms: Process



# Parameters of GA

- Fitness Function
- Mechanism of selection
- Crossover
- Mutation

# Fitness Function

- Evaluates how good an individual is
- Computes this for each individual in a population
  
- Fitness function is application dependent
- Examples: Mean squared error, Classification rate

# Mechanism of Selection

- Parent/Survivor Selection
  - Roulette Wheel Selection
  - Tournament Selection
  - Rank Selection
  - Elitist Selection



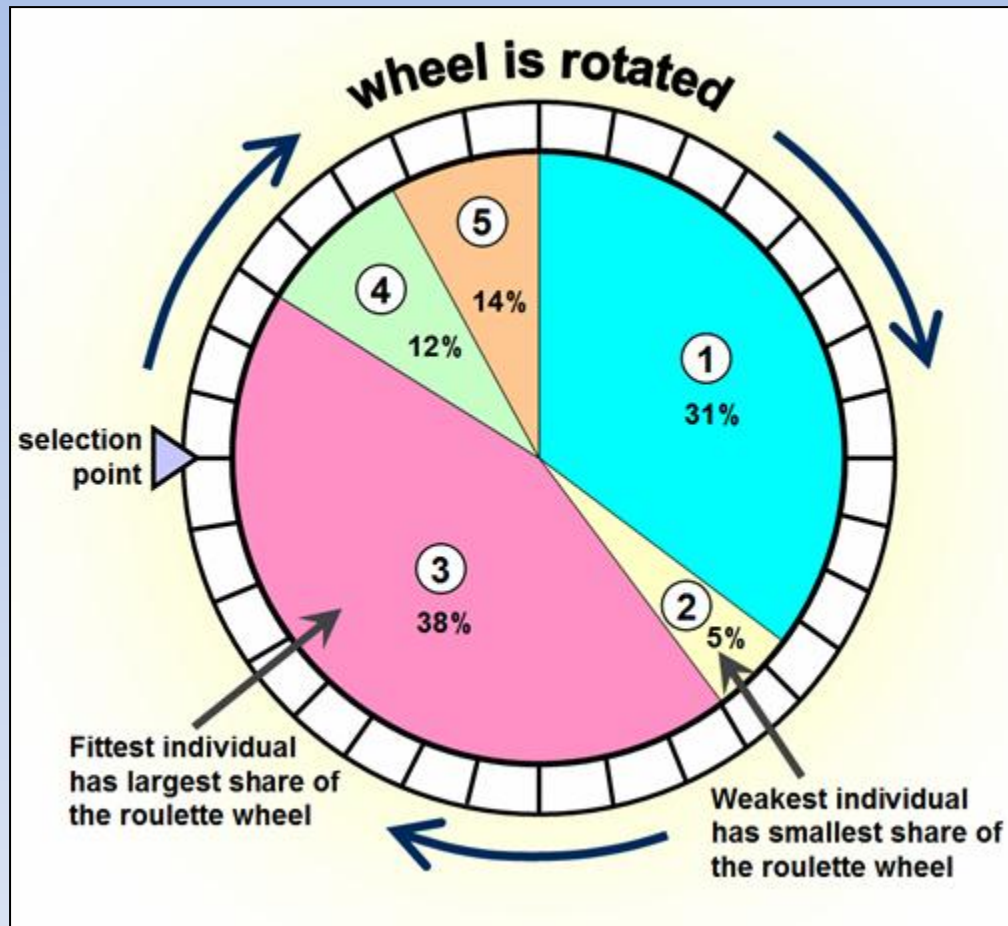
# Roulette Wheel Selection

- **Main idea:** better individuals get higher chance
- Individuals are assigned a probability of being selected based on their fitness.

$$p_i = f_i / \sum f_j$$

- $p_i$  probability that individual  $i$  will be selected
- $f_i$  is the fitness of individual  $i$
- $\sum f_j$  represents the sum of all the fitness(s) of the individuals with the population

# Roulette Wheel: Mechanism



# Tournament Selection

- **Binary tournament**
  - Two individuals are randomly chosen; the fitter of the two is selected as a parent
- **Larger tournaments**
  - $n$  individuals are randomly chosen; the fittest one is selected as a parent

# Other Methods

- Rank Selection

- Each individual in the population is assigned a numerical rank based on fitness, and selection is based on this ranking.

- Elitism

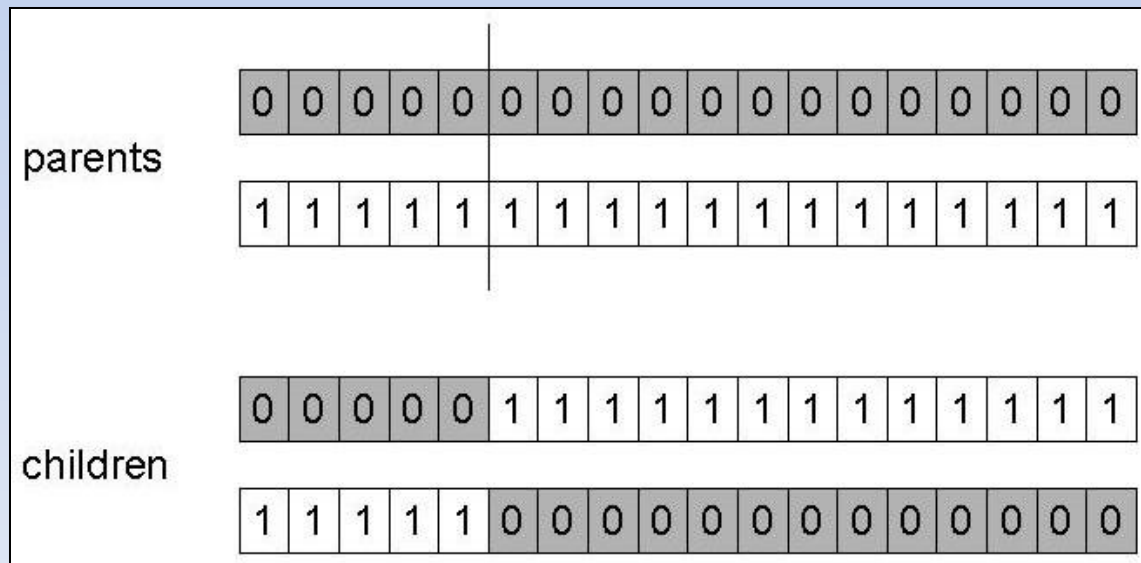
- Reserve  $k$  slots in the next generation for the highest scoring/fittest chromosomes of the current generation

# Crossover

- Generating offspring from two selected parents
  - Single point crossover
  - Two point crossover (Multi point crossover)
  - Uniform crossover

# One Point Crossover

- Choose a random point on the two parents
- Split parents at this crossover point
- Create children by exchanging tails



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- Choose a random point on the two parents
- Split parents at this crossover point
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```
Parent 1:   XX | XXXXX  
Parent 2:   YY | YYYYY  
Offspring 1: XXYYYYY  
Offspring 2: YYXXXXX
```

# Uniform Crossover

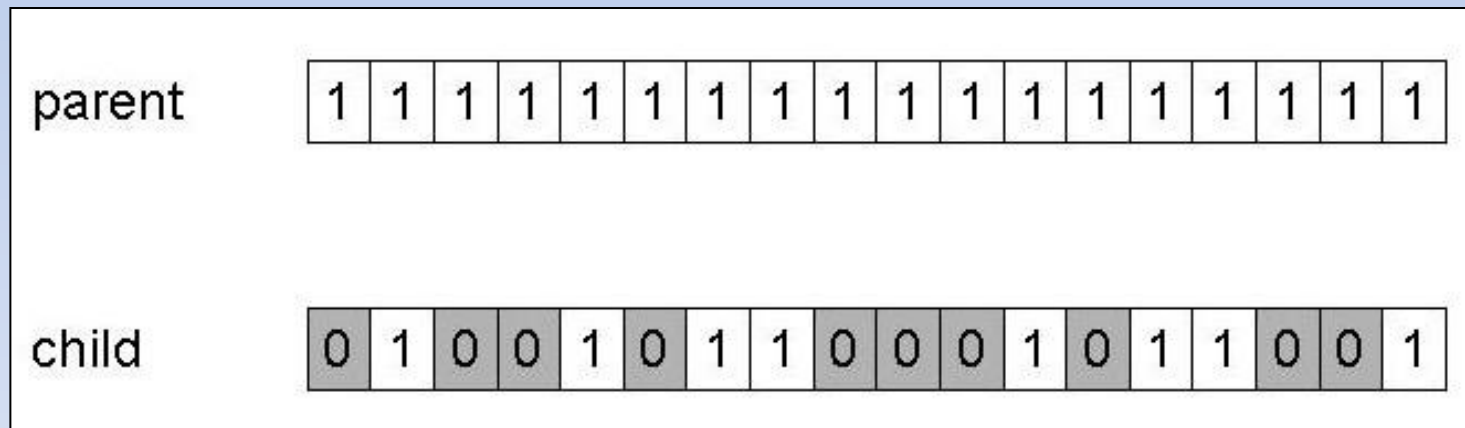
- A random mask is generated
- The mask determines which bits are copied from one parent and which from the other parent
- Bit density in mask determines how much material is taken from the other parent

<b>Mask:</b>	0110011000	(Randomly generated)
<b>Parents:</b>	<u>1</u> 01 <u>00</u> 0 <u>1</u> 110	<u>0</u> 0 <u>1</u> <u>1</u> 0 <u>1</u> 0 <u>0</u> <u>1</u> 0
<b>Offspring:</b>	0011001010	1010010110



# Mutation

- Alter each gene independently with a probability  $p_m$
- $p_m$  is called the mutation rate



# Summary – Reproduction cycle

- Select parents for producing the next generation
- For each consecutive pair apply crossover with probability  $p_c$  , otherwise copy parents
- For each offspring apply mutation (bit-flip with probability  $p_m$ )
- Replace the population with the resulting population of offsprings

# How to solve cost?

$$C(P_1, P_2) = \frac{1}{|P|} \sum_{X \in P} (I_1(P_1 X) - I_2(P_2 X))^2$$

- $P_1, P_2$  are the projection matrices
- $P$  is the number of points in 3D model
- $I(PX)$  is the color of point  $X$  on the image

# Initial Settings for GA[1]

- Initial estimation of projection matrices by manual registration
- This acts as initial population of the genetic algorithm

# Pipeline for Images to 3D Models

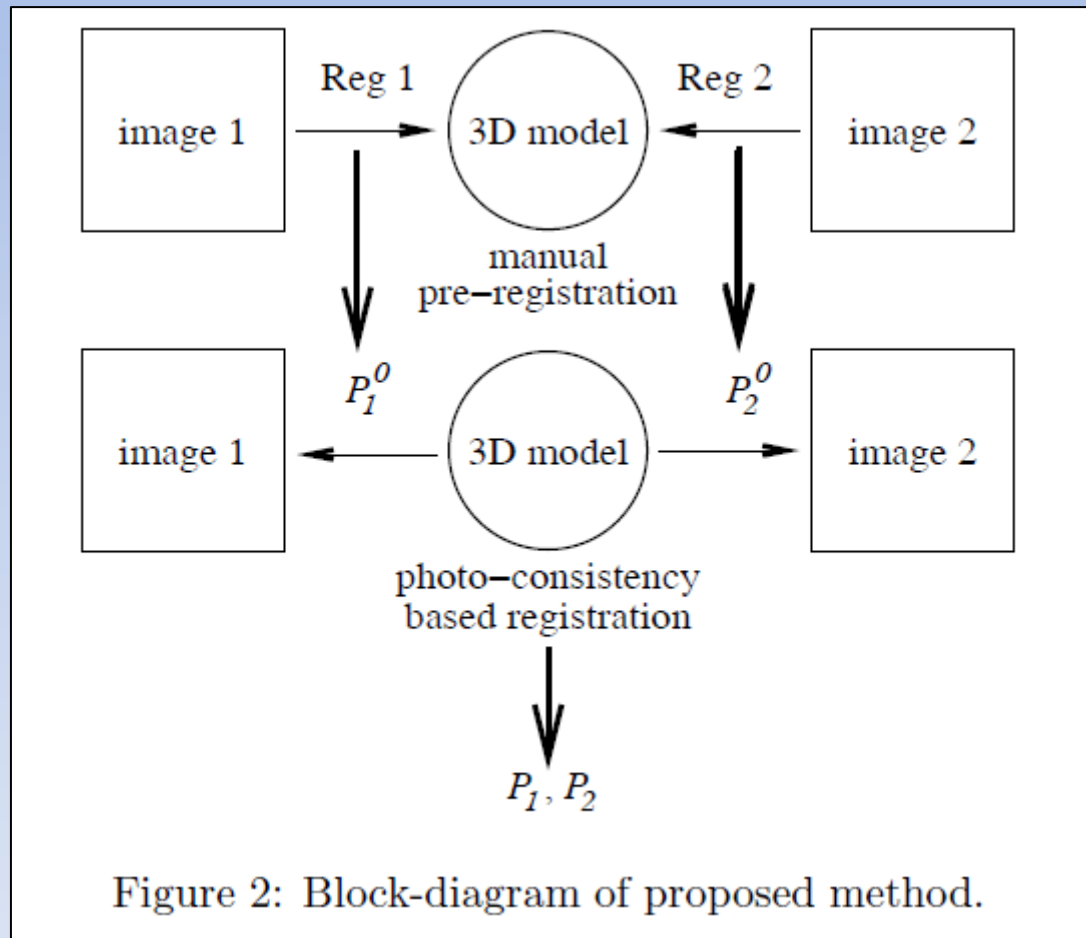


Figure 2: Block-diagram of proposed method.

# Results



Images



3D model



Textured model



Images



3D model



Textured model

# Results



Images

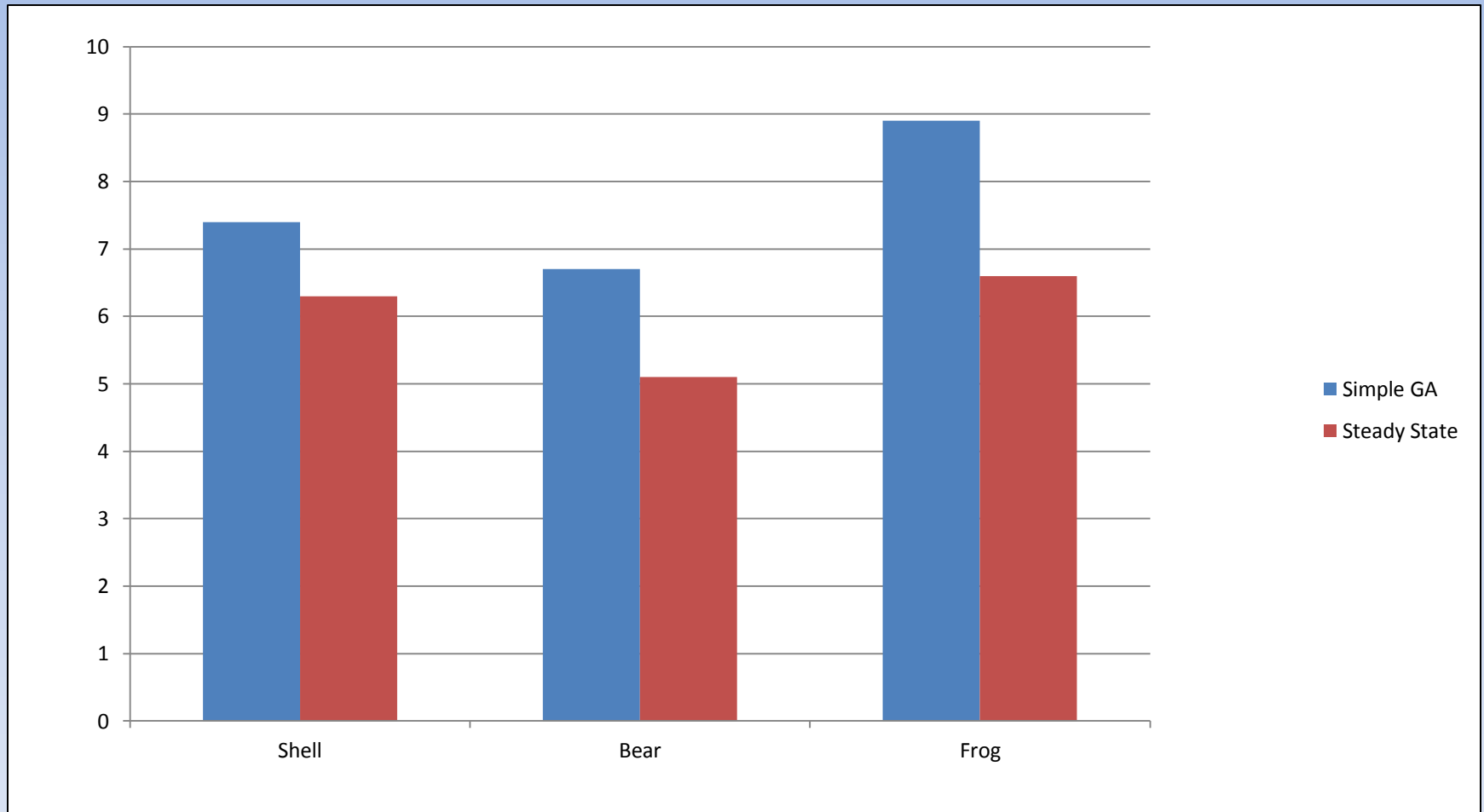


3D model



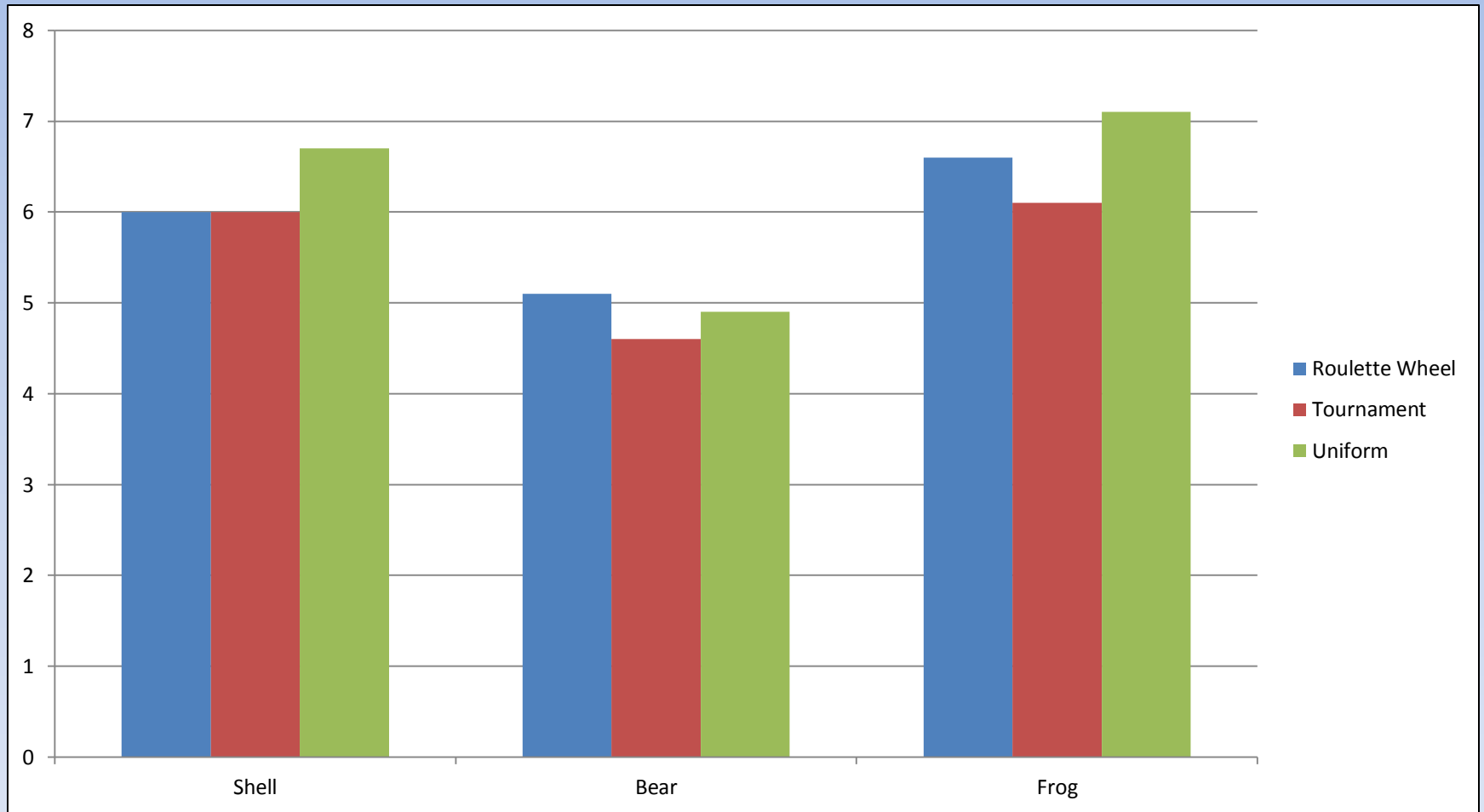
Textured model

# Discussions: Preservation

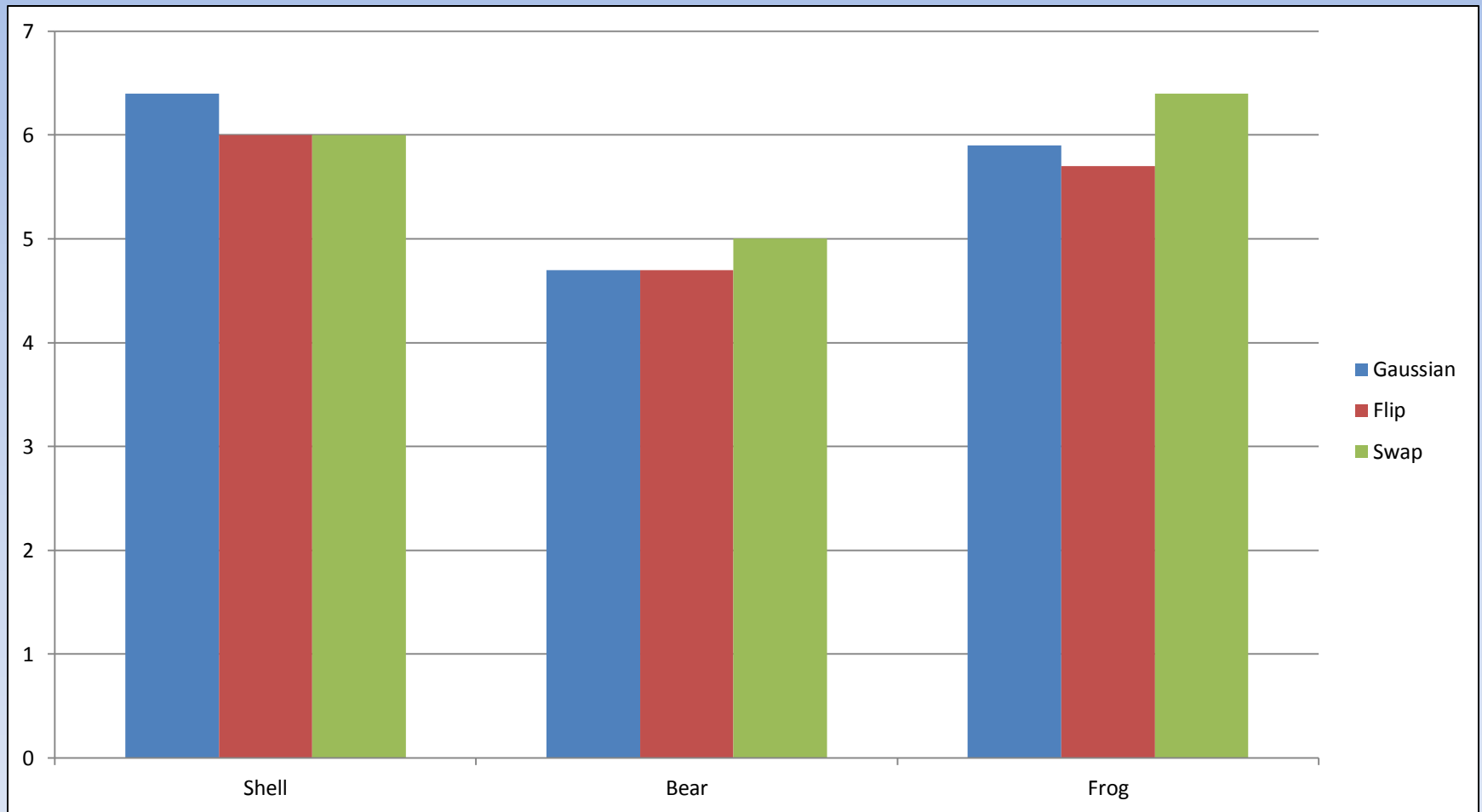




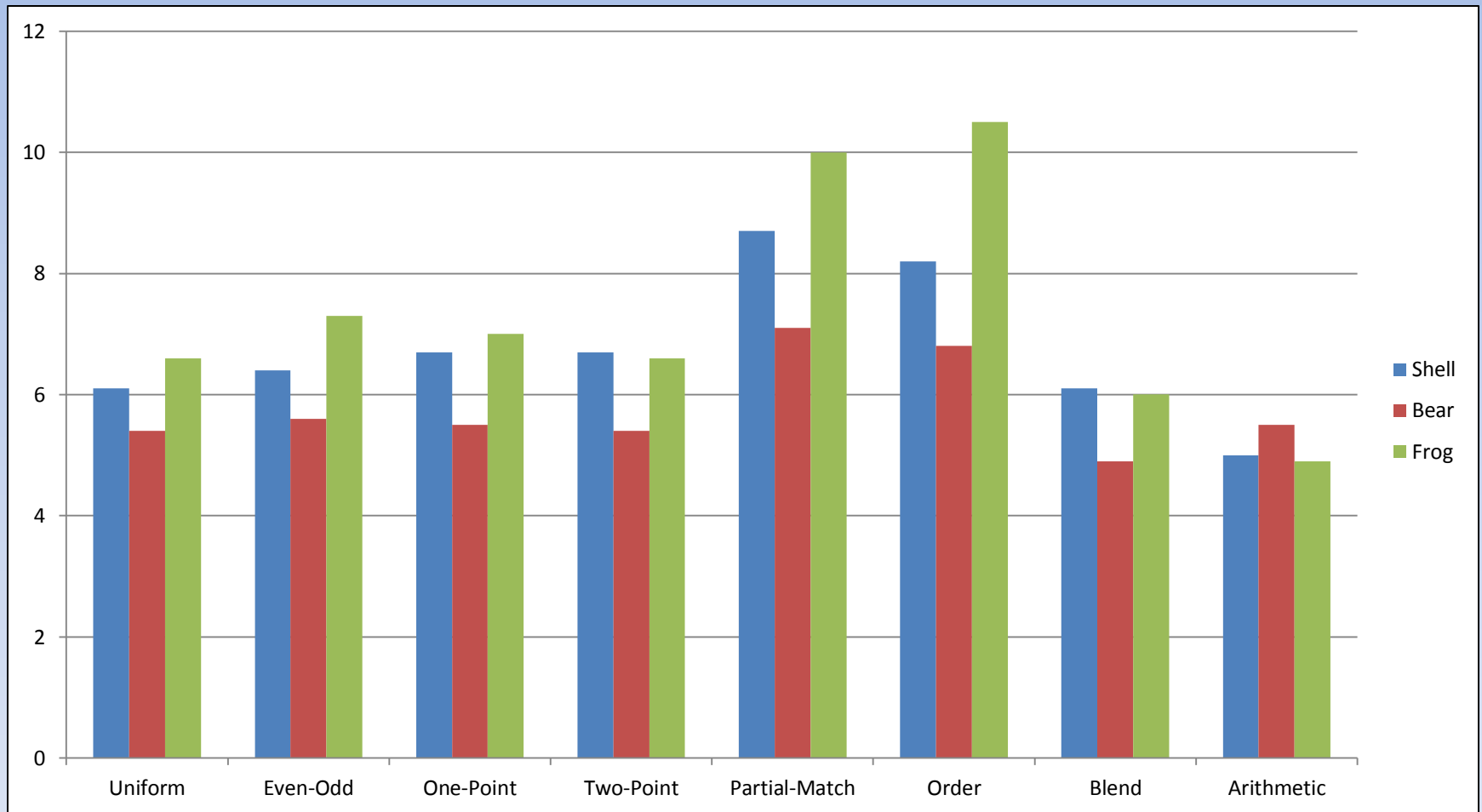
# Discussions: Selections Criteria



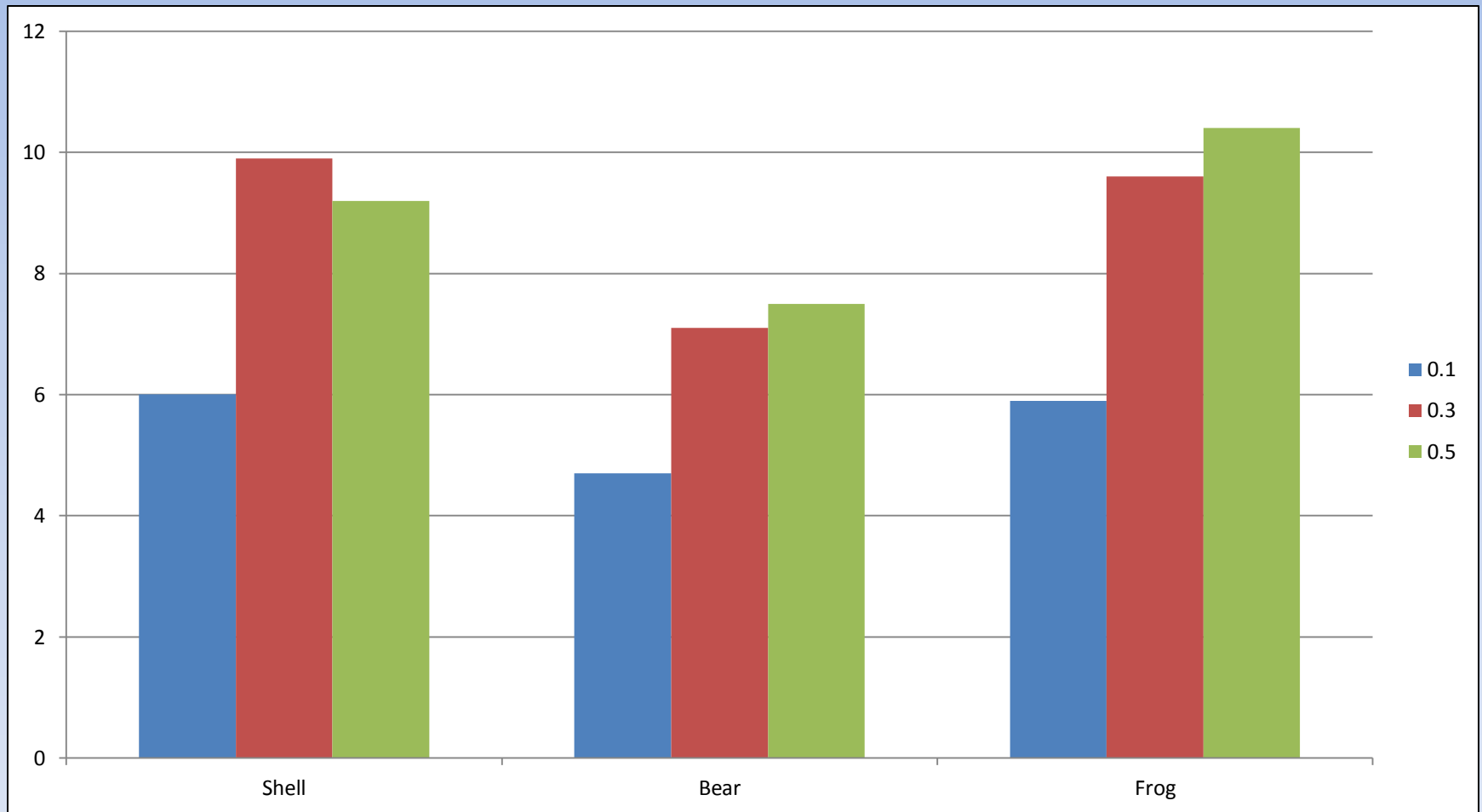
# Discussions: Mutations



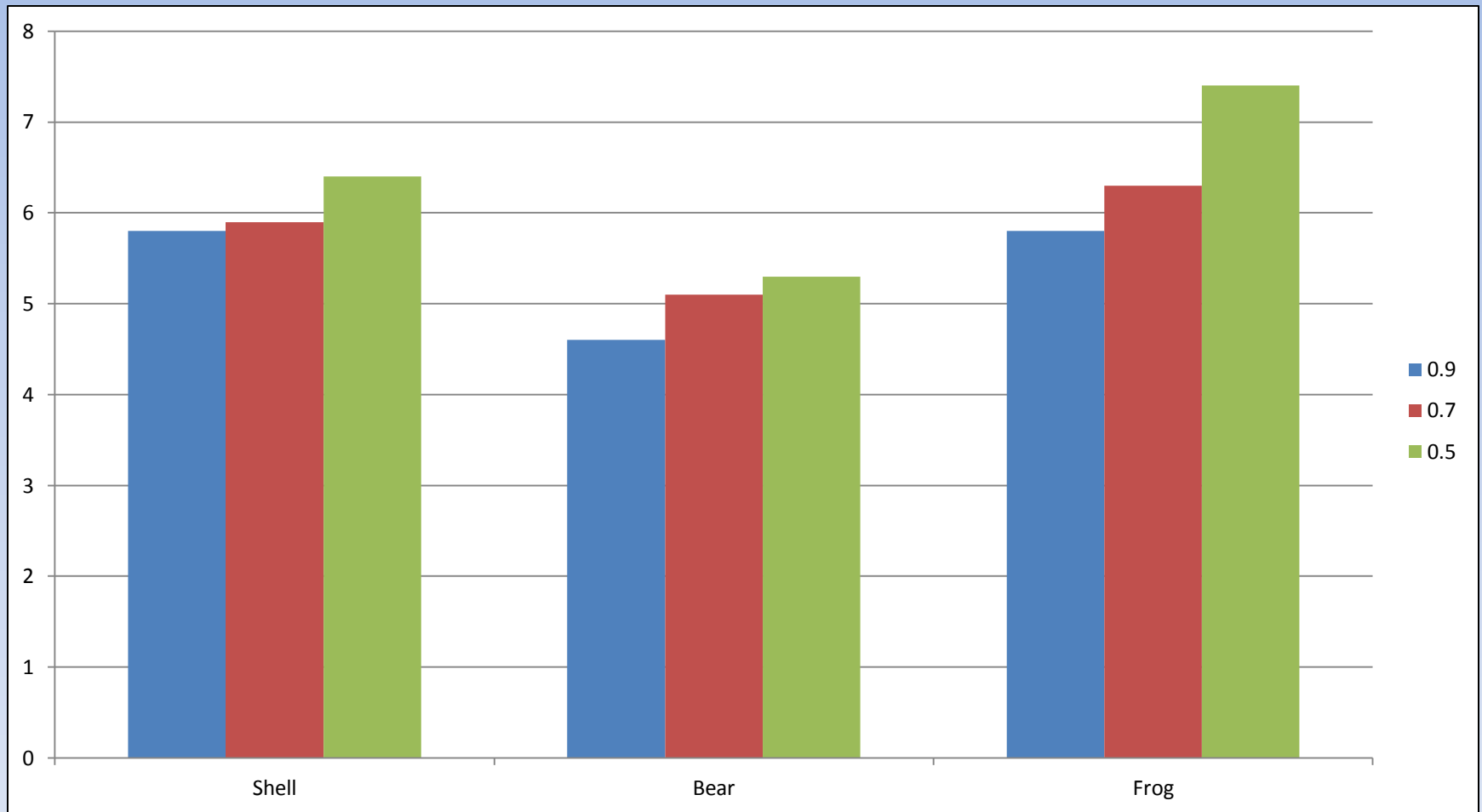
# Discussions: Crossover



# Discussions: P(Mutation)



# Discussions: P(CrossOver)



# Summary

- Genetic Algorithms useful for solving optimization problems
- Four elements:
  - Fitness Function
  - Mechanism of Selection
  - Crossover Mechanism and Frequency
  - Mutation Mechanism and Frequency
- Important! Problem representation for GA

# Bibliography

1. Janko, Z., Chetverikov, D., Ekart, A., *Using Genetic Algorithms in Computer Vision: Registering Images to 3D Surface Model*, Acta Cybernetica, 2007.
2. Eiben, A., Smith, J., *Introduction to Evolutionary Computing*, Springer, 2008.
3. <http://www.cs.cmu.edu/afs/cs/project/theo-20/www/mlbook/ch9.pdf>

# Questions