

Genetic Algorithms

mimic natural selection nature-inspired
 ← $\max f(x,y,z)$ ← individuals have genes
 genes determine phenotype (physical characteristics)
 phenotype contributes to fitness
 fitness contributes to propagation

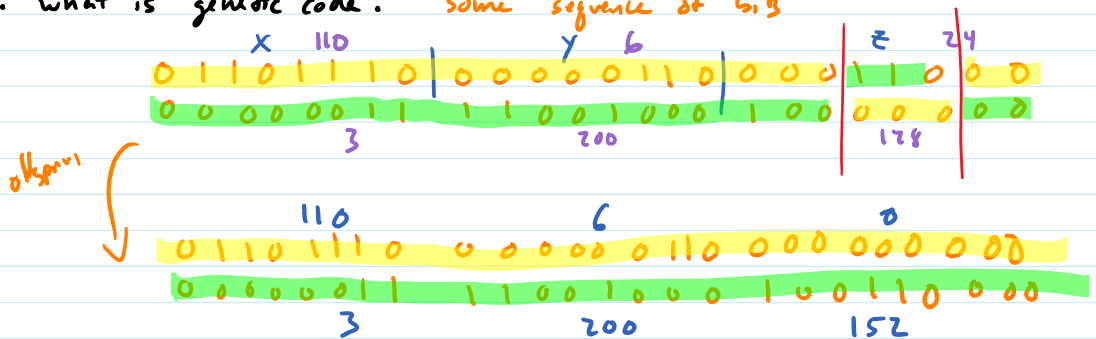
— ant colony opt
 — particle swarm opt

start with random population — random collection of individuals
 while not done (out of time, no recent improvement) (random genes)

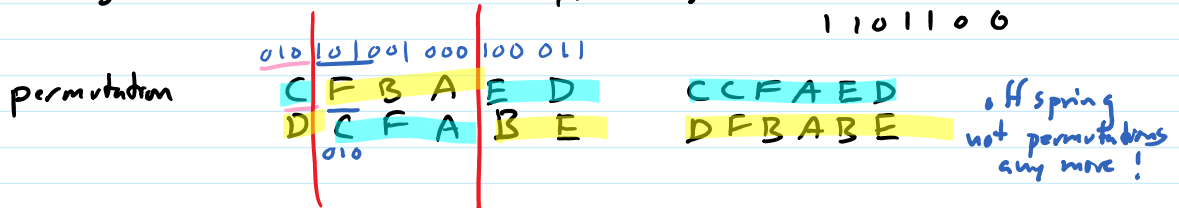
generation [

- evaluate individuals genes → inputs to f → evaluate f → fitness
- select for crossover randomly select genes from each parent for offspring
- crossover
- select for survival randomly change genes for some individuals
- mutate] exploration

Representation: what is genetic code? some sequence of bits



Crossover: bitstrings - choose start/end randomly, exchange



ABCDEF 24101
 012345 32300

24300 CFEABD
 32101 DCBAFE

Genetic Programming:

```

if seeds[0] > 5
{
  many++
}
  
```

```

if seeds[i] == 0 and seeds[i - 1] == 1
{
  return i-1
}
  
```

```
if seeds[i] == 0
{
    many++
}
```

O. David et al

Chess heuristic : for each player, count

9 for queen
5 for rook
3 for bishop
3 for knight
1 for pawn

+4 breakout pawn

+1 king protection

+1 rook mobility

⋮

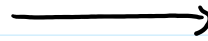
Problems for GA :

coevolution : fitness based on results of competing with other individuals

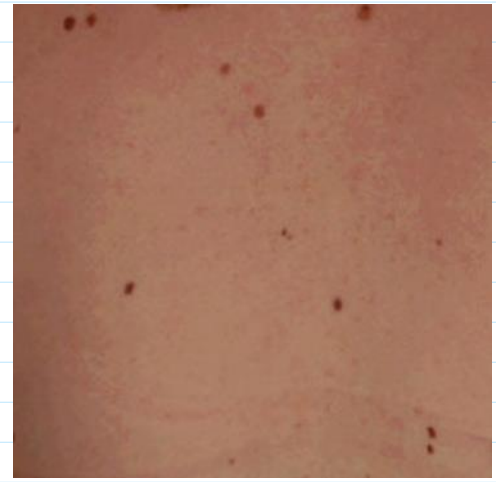
mentor : compare output to experts choices

GAs in Noisy Environments

Medical Image Registration



scale
translation
rotation
color shift



Optimization: Find parameters to

Fitness function: sum of difference of all pixels

of sum of difference of randomly selected pixels
(now a noisy fitness function)

Evolving a die with a high average

Parameter: num of sides

Fitness: mean value when rolling

Die 1	6 sided: 5 1 4
2	20-sided: 15 18 20
3	4 sided: 3 2 2
4	100-sided: 11 15 16
5	10-sided: 3 6 3

Evaluation Scheduling: Given budget T of fitness estimations, allocate across individuals

generations
phases

to optimize evolution

Game Playtesting : Input
Fitness