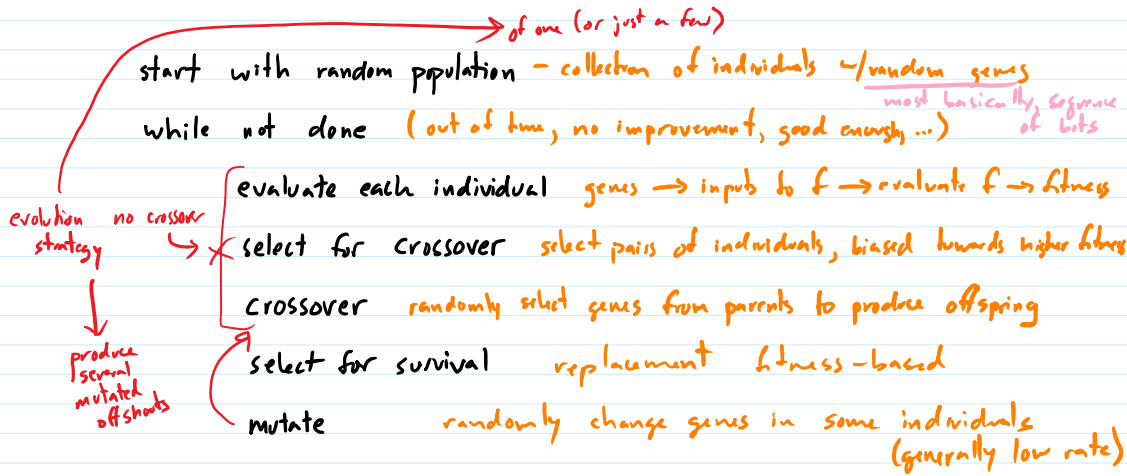


Genetic Algorithms

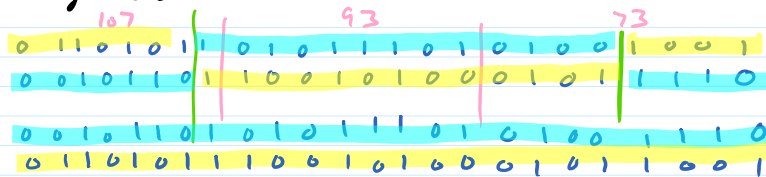
mimic natural selection
 individuals have genes
 genes determine phenotype
 phenotype contributes to fitness
 fitness contributes to propagation

nature-inspired Ant Colony Optimization
Particle Swarm Opt



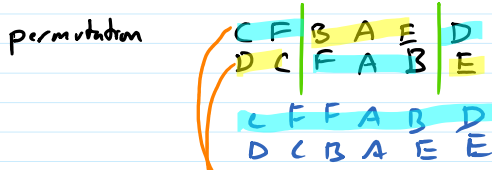
Representation: what is genetic code?

$f(x,y,z)$



Crossover: bitstrings ex: two-point crossover

0111010
 1101100



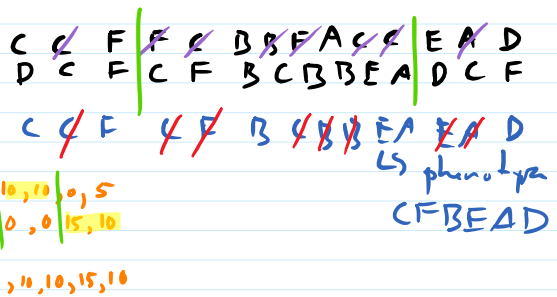
not a valid input!
 also not a permutation! infeasible

offspring not enough like parents

C	F	E	A	B	D
D	C	B	A	F	E
C	F	B	A	D	E
D	C	F	A	B	E
F	D	B	C	A	E

	A	B	C	D	E	F
C	50	70	100	20	10	80
D	37	30	50	75	26	40
F	50	70	50	75	26	80

solutions 1) assign low fitness to infeasible solns
 2) design genome to avoid infeasible offspring



Genetic Programming: genetic algorithms for programs

```

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

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

fitness function
 input: heuristic function
 output: performance when

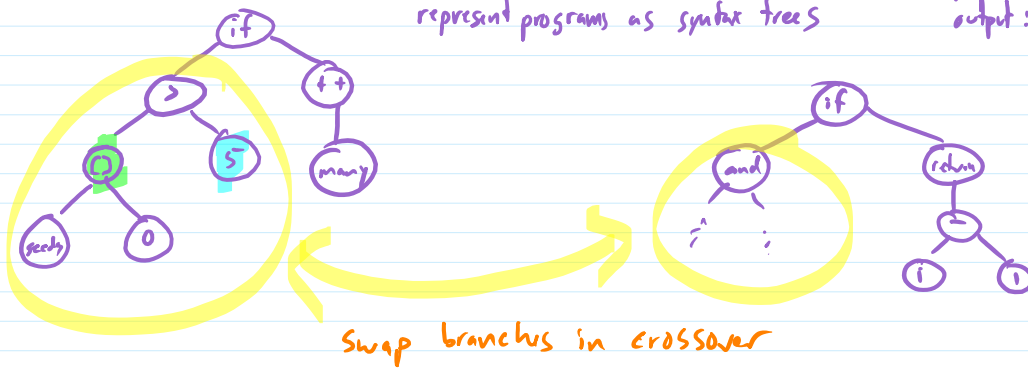
represent programs as syntax trees

```
many++  
}
```

```
return i-1  
}
```

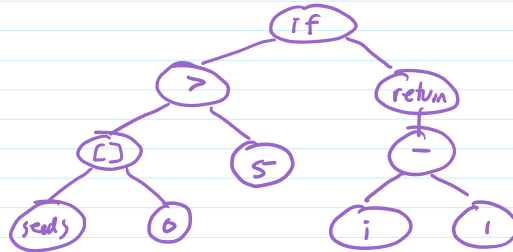
represent programs as syntax trees

! fitness function
input: heuristic function
output: performance when
playing w/ that heuristic



Swap branches in crossover

```
if seeds[0] > 5  
{  
return i - 1;  
}
```



G. David et al

Chess Heuristic: for each player, count

- 9 for queen
- 5 for rook
- 3 for bishop
- 3 for knight
- 1 for pawn
- +1 for rook mobility
- +1 for king protection
- +3 for breakaway pawn
- ⋮

Find parameters to maximize Elo → numeric measure of strength of player
adjust rating after each game
larger adjustments for more surprising result

Problems for GA: what is fitness?

performance against existing agents
or population of agents
how well moves chosen match experts (or other good agents)

coevolution: play individuals against each other

fitness function is #wins in round-robin tournament

mentor: how often individual's choice matches mentor's

GAs in Noisy Environments

→ noise in fitness fun

$f(\text{heuristic constants}) = \text{expected turns to complete game}$
in Rule of 28

difficult to compute
so estimate

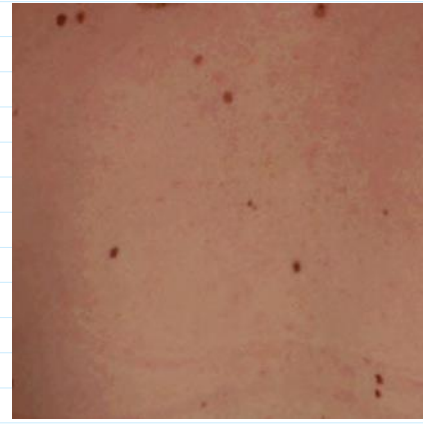
play 10000 times

if 0 for one game = 1

0 for estimate over 10000 games

$$= \frac{1}{\sqrt{10000}} = \frac{1}{100}$$

Medical Image Registration



Optimization: Find parameters to maximize # matching pixels
↳ rotation angles, zoom, skew, translation, ...

Fitness function:

in general, GAs can deal with noisy fitness funs

- don't need more samples per individual
- randomness leads to exploration