

Genetic Algorithms

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Wilkes University

CS 340 - Fall 2019

GREGOR MENDEL

(1822 - 1884) WAS AN AUGUSTINIAN MONK FROM BRÜNN, AUSTRIA. IN HIS SPARE TIME, MENDEL BRED PEA PLANTS IN THE MONASTERY GARDENS.



BUT MENDEL WAS
NOT JUST AN
AMATEUR GARDENER,
BUT A SCIENTIST
WHO STUDIED HIS PEA
PLANTS MOST
CAREFULLY —
HE CALLED THEM
HIS "CHILDREN."

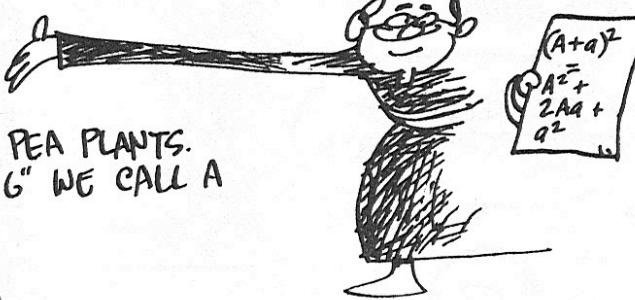


MENDEL'S INTERPRETATION:

IT'S MATHEMATICAL!

THERE IS
SOMETHING
IN POLLEN AND
EGG WHICH
DETERMINES
THE HEIGHT OF PEA PLANTS.
THIS "SOMETHING" WE CALL A

GENE.



EACH POLLEN GRAIN AND EGG HAS ONE HEIGHT GENE, SO
THE PLANT FORMED BY THEIR UNION HAS TWO.

THE GENE MAY BE ONE OF TWO DISTINCT TYPES, OR

ALLELLES.

ONE ALLELE, A , IS
FOR TALLNESS;
THE OTHER ONE, a ,
IS FOR SHORTNESS.

GENES MAKE
SHORTS?

CUT-OFFS....



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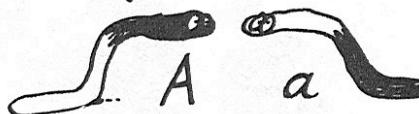
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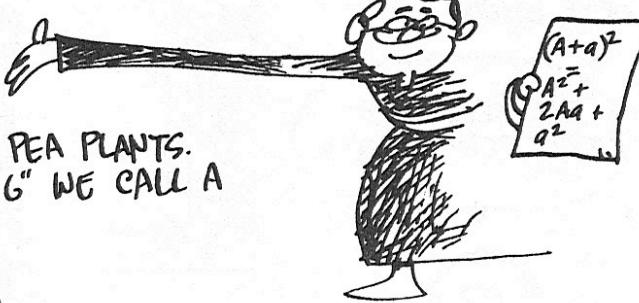


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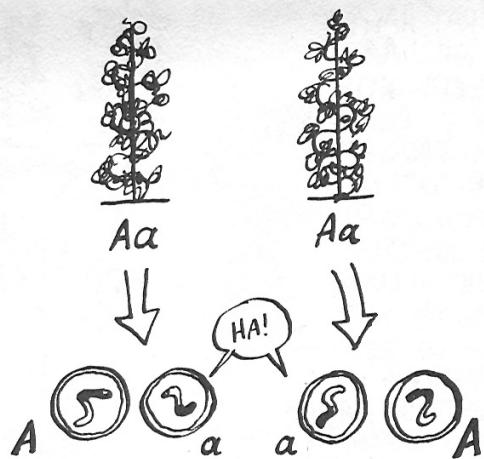
CUT-OFFS....



Individuals Chromosomes

Genes

WHEN THE HYBRID SELF-FERTILIZES, ITS ALLELES A AND α ARE SORTED OUT RANDOMLY AMONG THE POLLEN GRAINS AND EGGS. BOTH A AND α APPEAR, AND IN ROUGHLY EQUAL PROPORTIONS.



WHEN EGGS AND POLLEN UNITE, THERE ARE FOUR POSSIBILITIES:



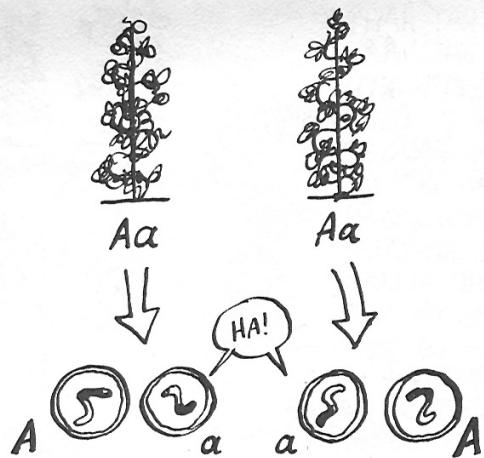
SHORT POLLEN,
SHORT EGG

TALL POLLEN.
SHORT EGG

SHORT POLLEN,
TALL EGG

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SHORT POLLEN,
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SHORT EGG

SHORT POLLEN,
TALL EGG

TALL POLLEN,
TALL EGG

Individuals
Chromosomes

Genes

Code:
IF aa THEN
be short
ELSE
be tall



First Steps

Pick a problem

- ◆ e.g. play blackjack



First Steps

Design a representation

oooooooooooooooooooo

0 = stay 1 = hit



First Steps

Design a fitness function

e.g. actually use
representation
to play Black Jack



Basic Genetic Algorithm

Start with initial population

→ Check for fitness

Breed fit individuals

Basic Genetic Algorithm

Start with initial population

→ Check for fitness

Breed fit individuals

Solution

Which individuals to breed?



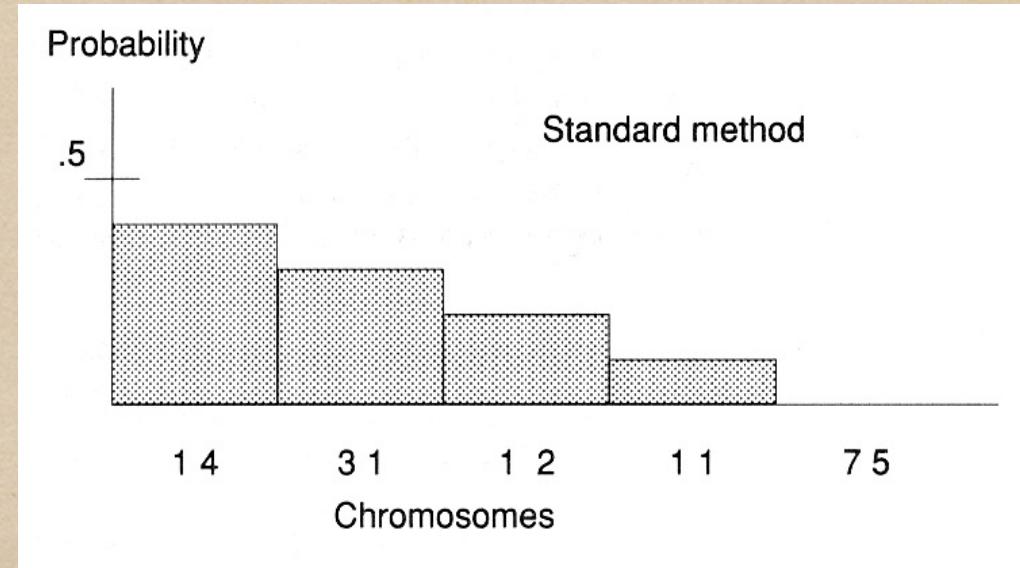
Which individuals to breed?

- ◆ Most fit
- ◆ Ranked by fitness
- ◆ Most Diverse



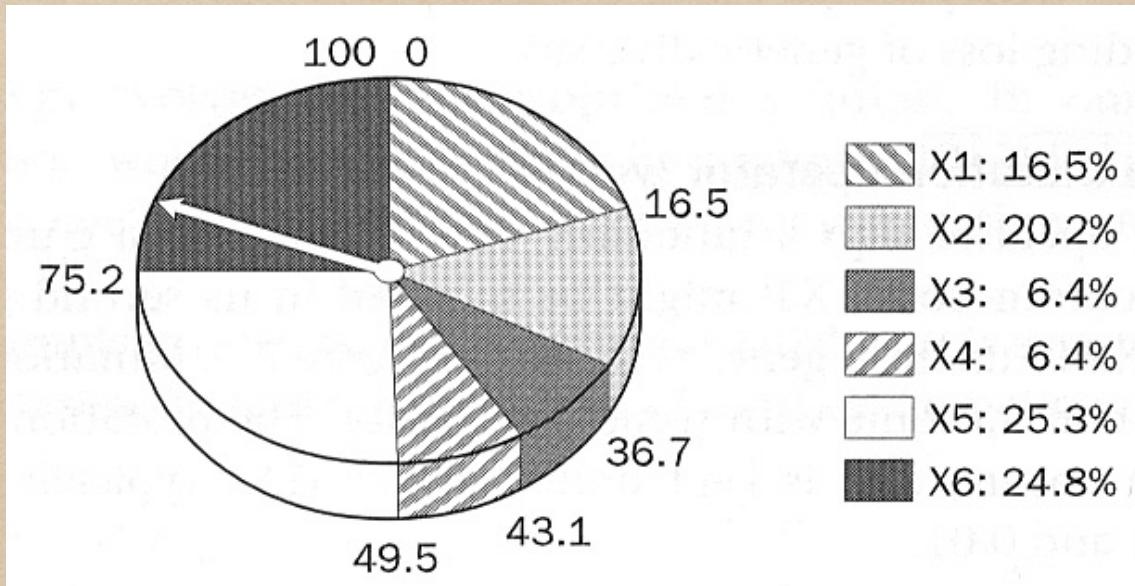
Standard Method

$$F_i \approx \frac{Q_i}{\sum_j Q_j}$$



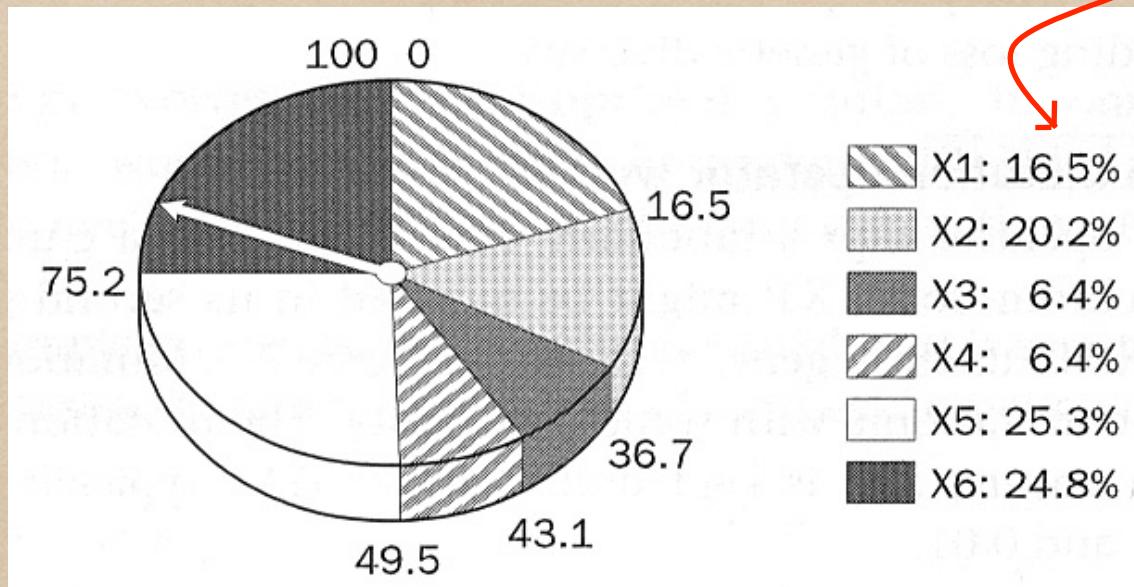
Fitness equated
with relative quality

Standard Method



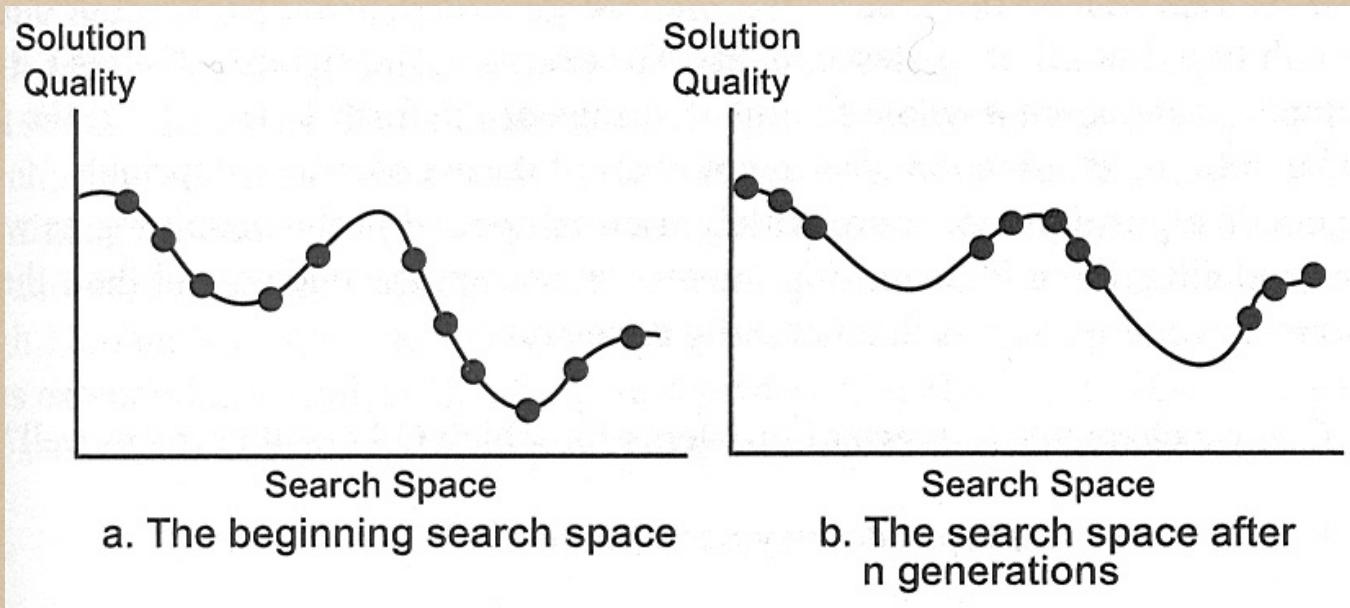
Randomly select individuals
with relative quality

Standard Method



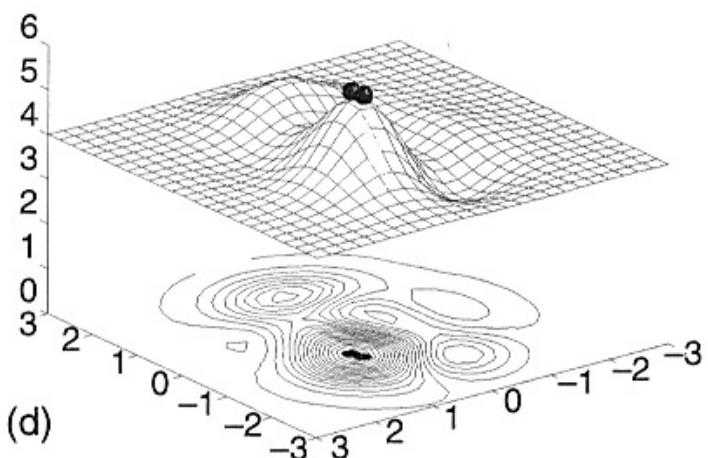
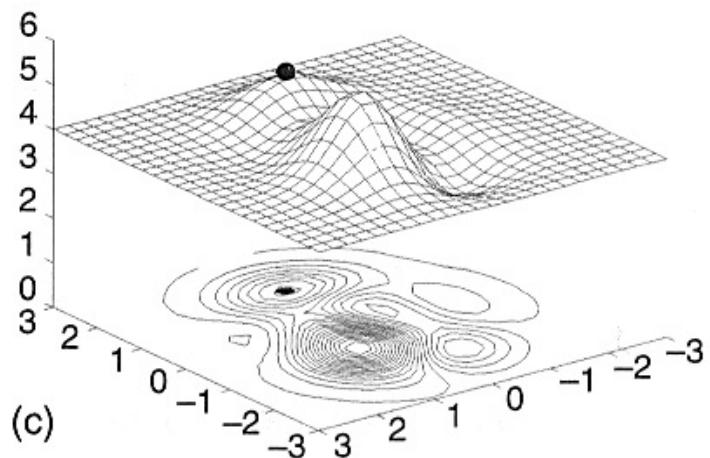
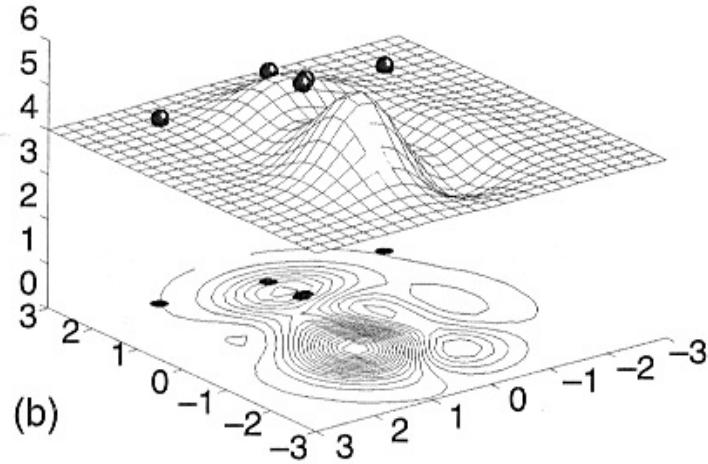
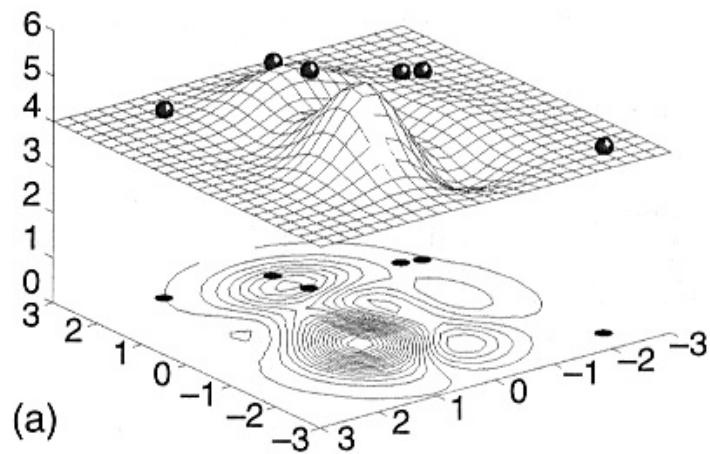
Randomly select individuals
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Like Parallel Hill Climbing



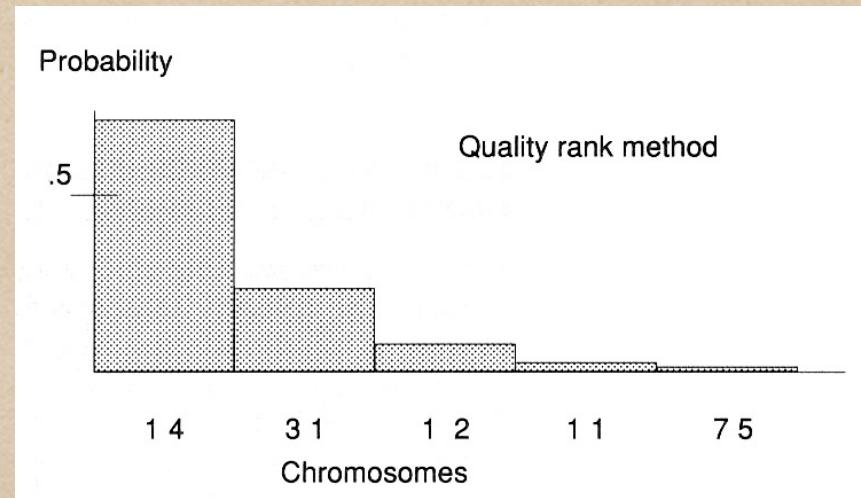
Fitness landscape

- ◆ Differs from heuristic state space search
 - no estimates of “effort” to move to new state
 - no strict ordering of states
- ◆ Has peaks, valleys, discontinuities
 - both real and introduced
 - e.g binary vs. gray coding



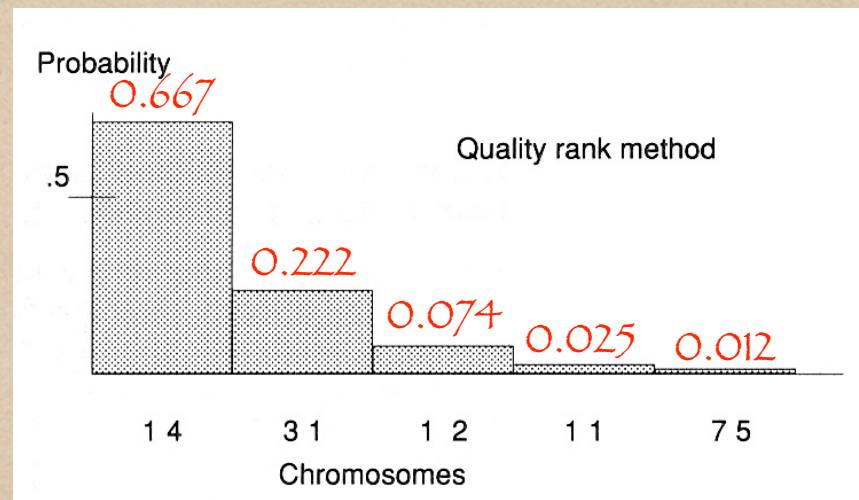
Quality Rank Method

Rank each
with fixed
Probability



Quality Rank Method

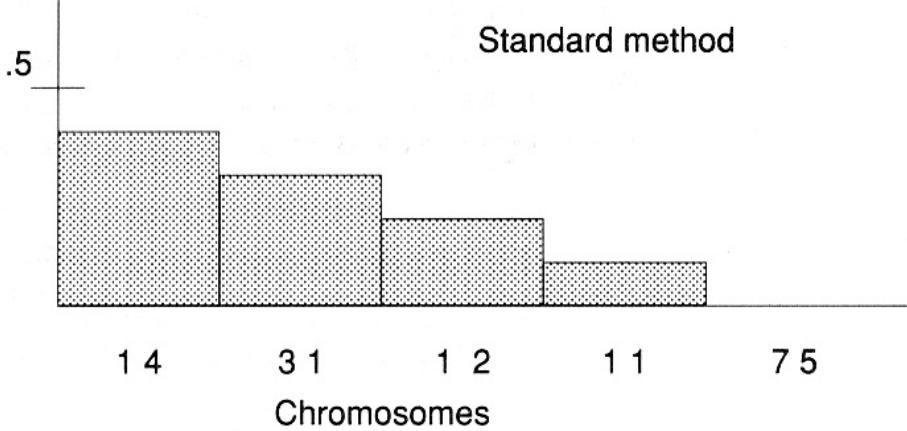
Rank each
with fixed
Probability



Here 0.667 and then 1/3 of the remaining

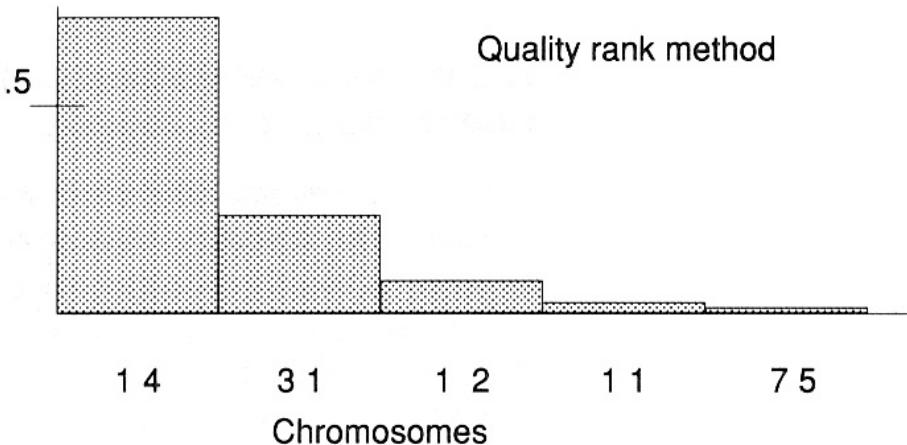
Probability

Standard method



Probability

Quality rank method





Example fitness function

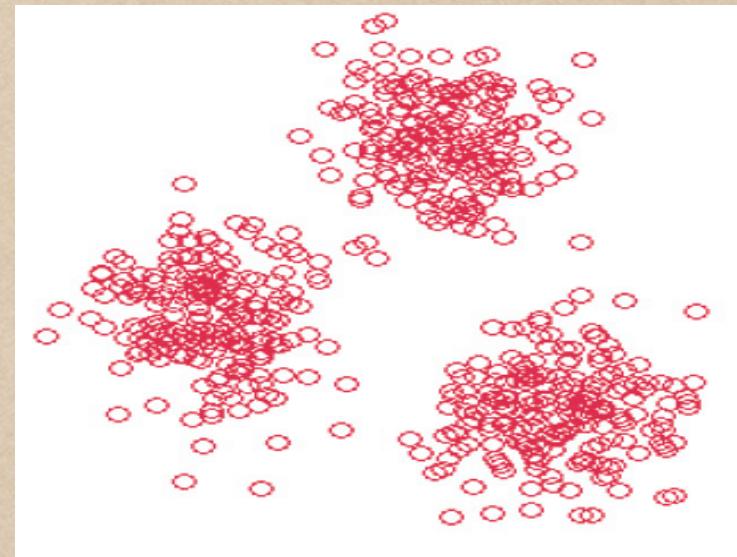
9	1	2	3	4	5	4	3	2	1
8	2	0	0	0	0	0	0	0	2
7	3	0	0	0	0	0	0	0	3
6	4	0	0	7	8	7	0	0	4
Sugar	5	5	0	0	8	9	8	0	0
4	4	0	0	7	8	7	0	0	4
3	3	0	0	0	0	0	0	0	3
2	2	0	0	0	0	0	0	0	2
1	1	2	3	4	5	4	3	2	1
	1	2	3	4	5	6	7	8	9
									Flour

A fitness moat

	9	1	2	3	4	5	4	3	2	1
	8	2	0	0	0	0	0	0	0	2
	7	3	0	0	0	0	0	0	0	3
	6	4	0	0	7	8	7	0	0	4
Sugar	5	5	0	0	8	9	8	0	0	5
	4	4	0	0	7	8	7	0	0	4
	3	3	0	0	0	0	0	0	0	3
	2	2	0	0	0	0	0	0	0	2
	1	1	2	3	4	5	4	3	2	1
		1	2	3	4	5	6	7	8	9
										Flour

Rank Space Method

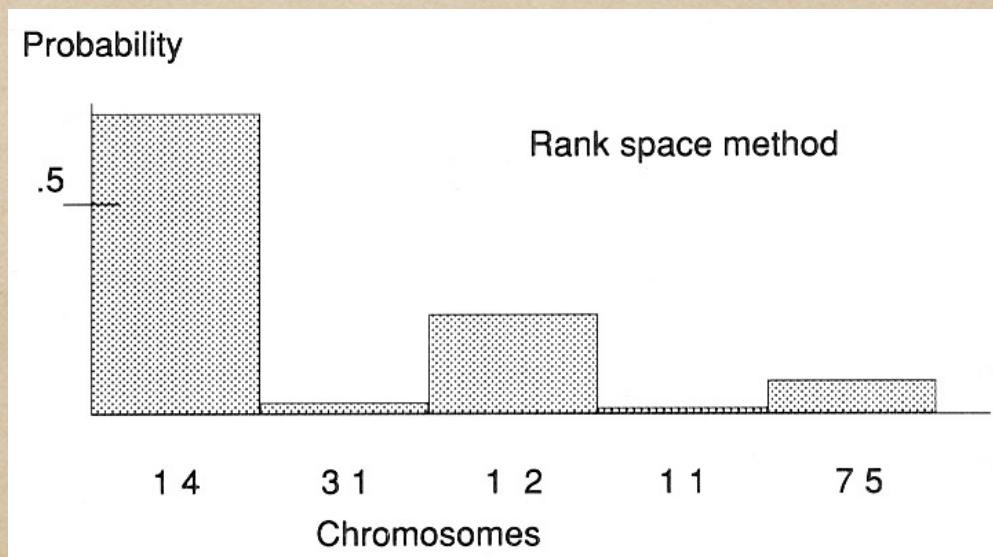
$$D_i = \sum_i \frac{1}{d_i^2}$$



Also compute a diversity rank

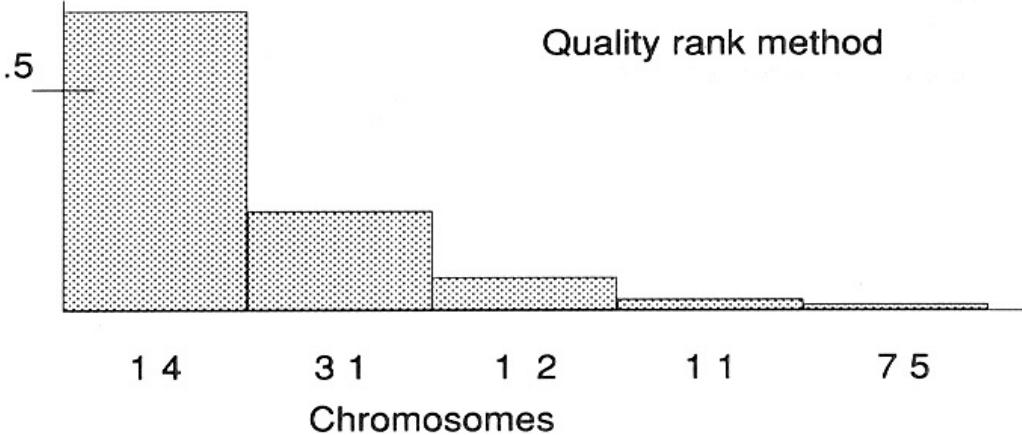
Rank Space Method

$$\frac{Q_i}{\sum_j Q_j} + \sum_i \frac{1}{d_i^2} \approx \text{Combined Fitness}$$



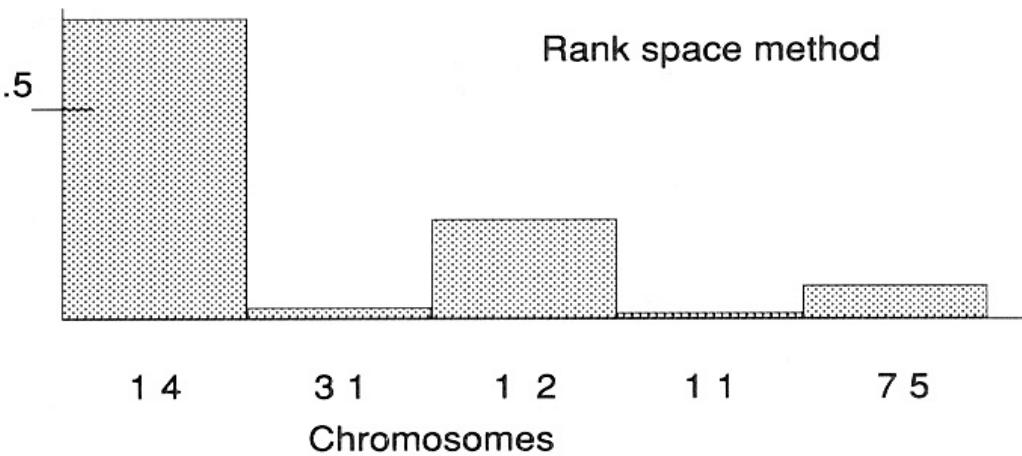
Probability

Quality rank method



Probability

Rank space method



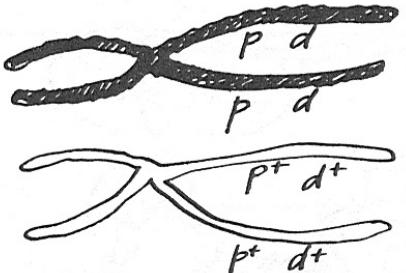
New Individuals?

During reproduction there are
two genetic operators:

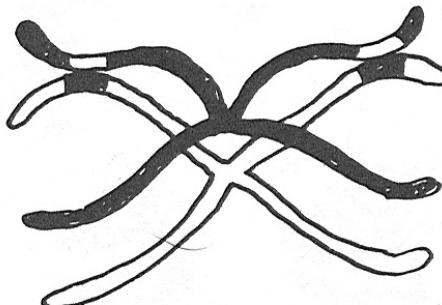
Crossover

Mutation

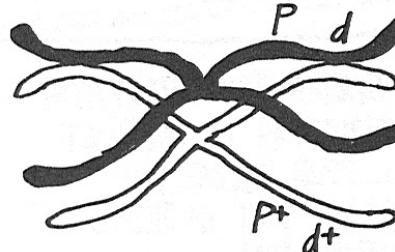
DURING MEIOSIS, HOMOLOGUES LINE UP WITH CORRESPONDING ALLELES OPPOSITE ONE ANOTHER.



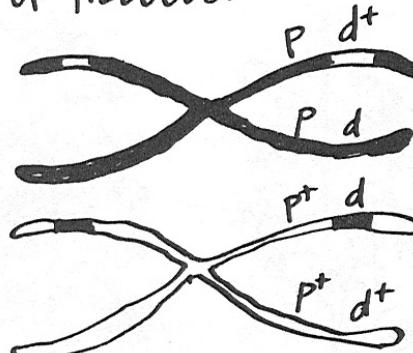
SOME SEGMENTS CROSS OVER:



AT CERTAIN POINTS, SEEMINGLY "CHOSEN" AT RANDOM, THE CHROMOSOMES TOUCH:



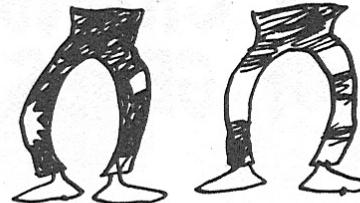
WHEN THEY SEPARATE, THEY HAVE NEW COMBINATIONS OF ALLELES.



THESE MUTATIONS - IT MEANS "CHANGES" IN LATIN - ARE FAIRLY RARE: THE CHANCE OF FINDING A MUTATION IN A GIVEN GENE IN AN INDIVIDUAL IS

1 IN 100,000

THOUGH SOME GENES ARE MORE PRONE TO CHANGE THAN OTHERS!



EVEN AT THIS RATE, THEY DO ADD UP! A HUMAN HAS SOME 200,000 GENES, SO WE CARRY AN AVERAGE OF TWO NEW MUTATIONS APIECE.

GREAT SHADES!

SORRY...
THOSE ARE
MY EYES...

BUT EVERY SO OFTEN
THE GENETIC "ERROR"
MAY BE OF POSITIVE
ADVANTAGE TO THE
LUCKY MUTANT!!

HMM! SO THE EGG
DID COME BEFORE
THE CHICKEN!





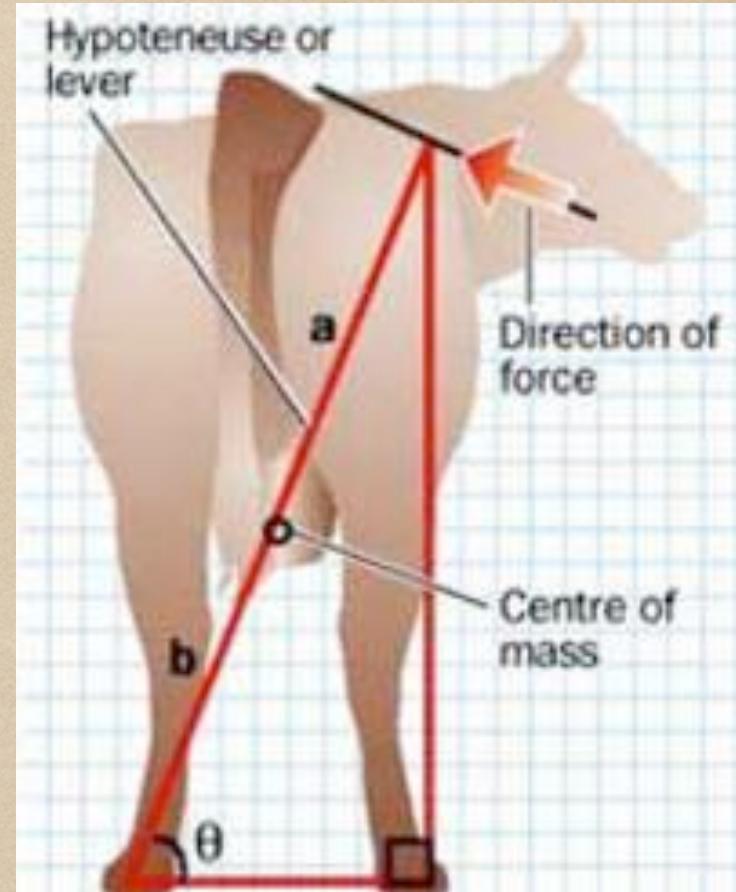
Cow Tipping

[<http://www.norcalblogs.com/bumpkins/>]



[AP Photo/Shane Jennings]

“Cows are heavy, man!” said Peter Park, a college student who tried to cow-tip on a trip to central Illinois. “There were four of us, but we had to run at the cow from 30 feet away and push as hard as we could to get it to even notice us. It woke up and looked at us like we were a bunch of flies, and walked away.”



[http://seedmagazine.com/content/article/udder_impossibility/]

[<http://www.englishwithrae.com/>]

Fitness landscape

- ◆ Differs from heuristic state space search
 - no estimates of “effort” to move to new state
 - no strict ordering of states
- ◆ Has peaks, valleys, discontinuities
 - both real and introduced
 - e.g **binary** vs. **gray coding**

Binary vs. Gray Coding

0000 ----- 0000

0001 ----- 0001

0010 ----- 0011

0011 ----- 0010

0100 ----- 0110

0101 ----- 0111

0110 ----- 0101

0111 ----- 0100

1000 ----- 1100

Binary vs. Gray Coding

0000 ----- 0000

0001 ----- 0001

0010 ----- 0011

0011 ----- 0010

0100 ----- 0110

0101 ----- 0111

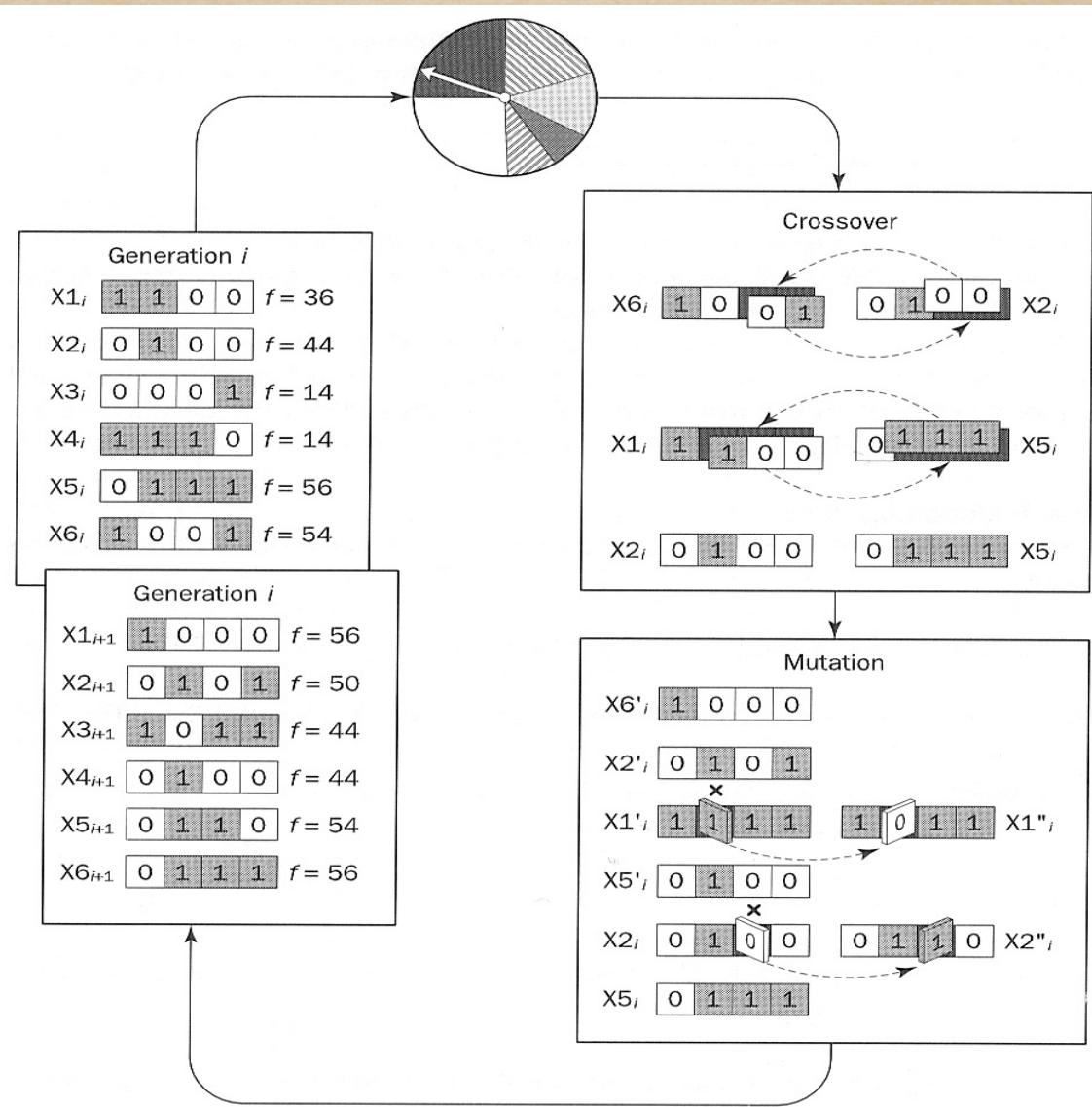
0110 ----- 0101

0111 ----- 0100

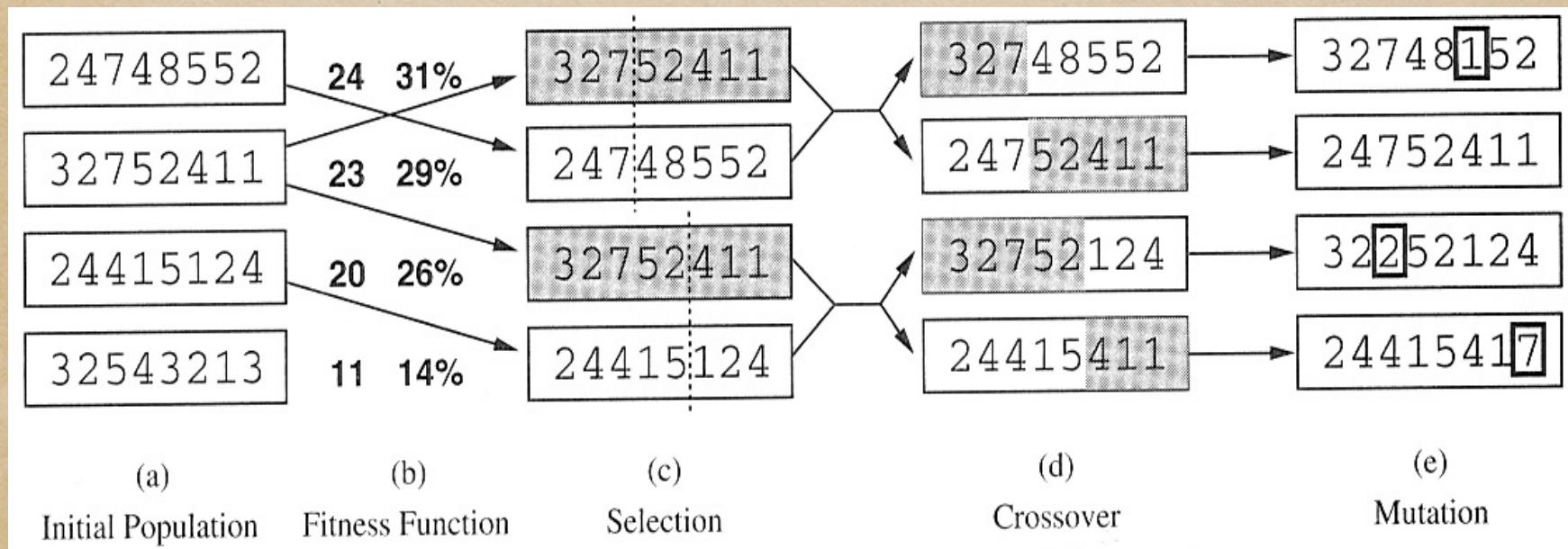
1000 ----- 1100

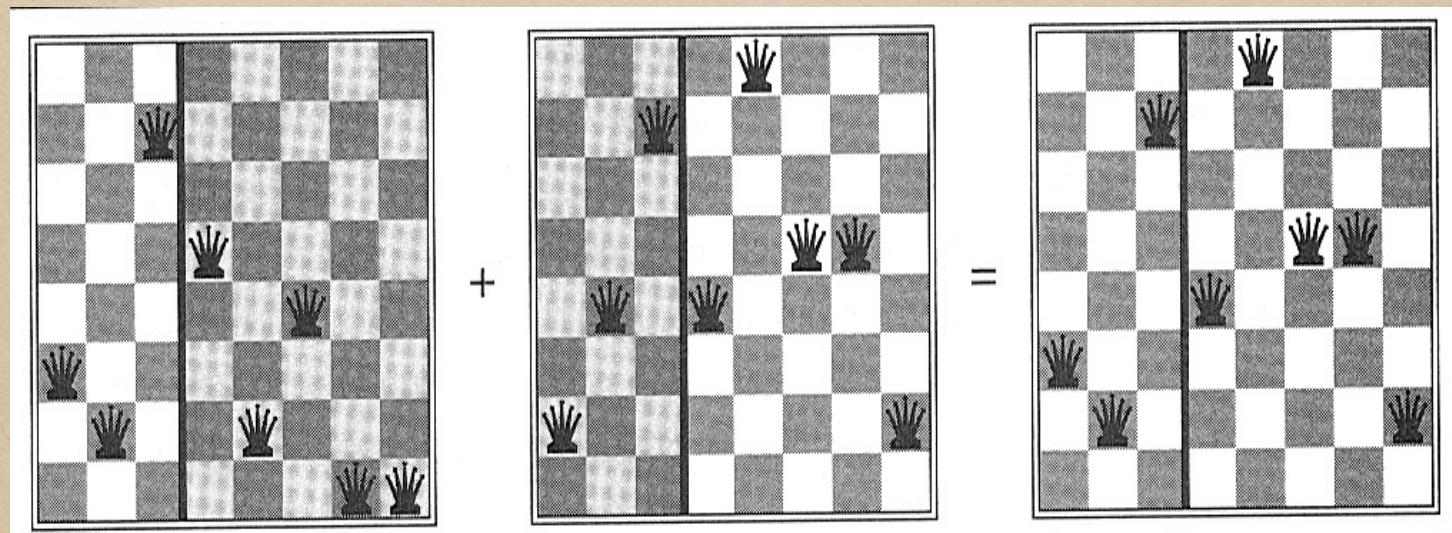
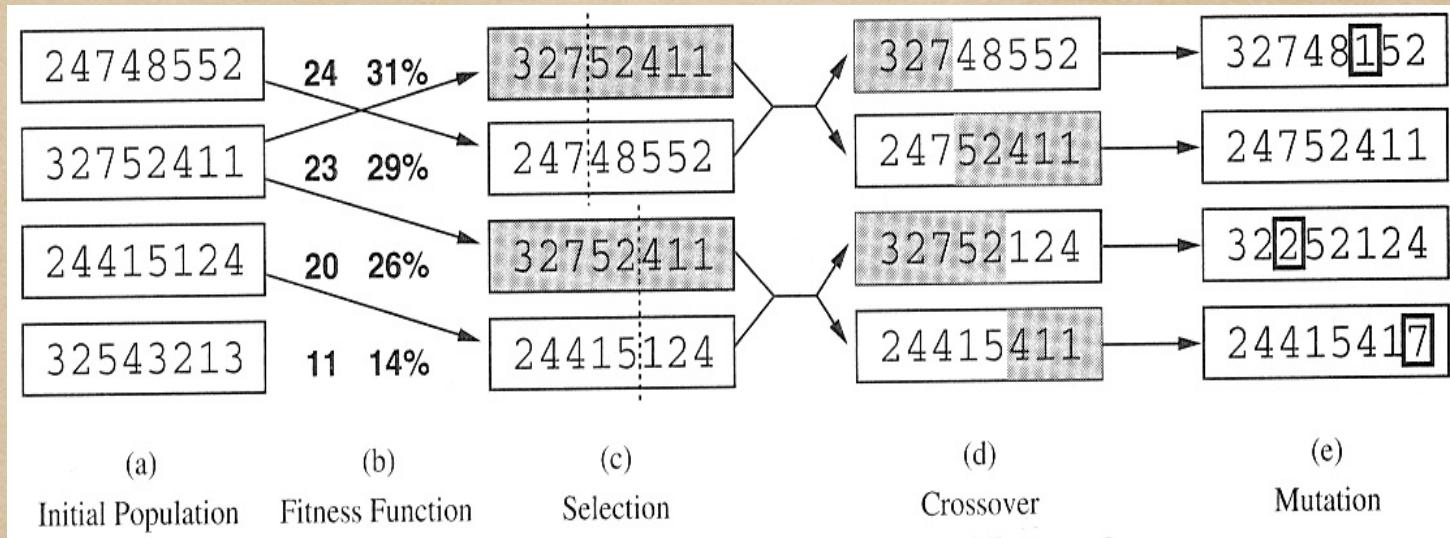
Hamming Distance

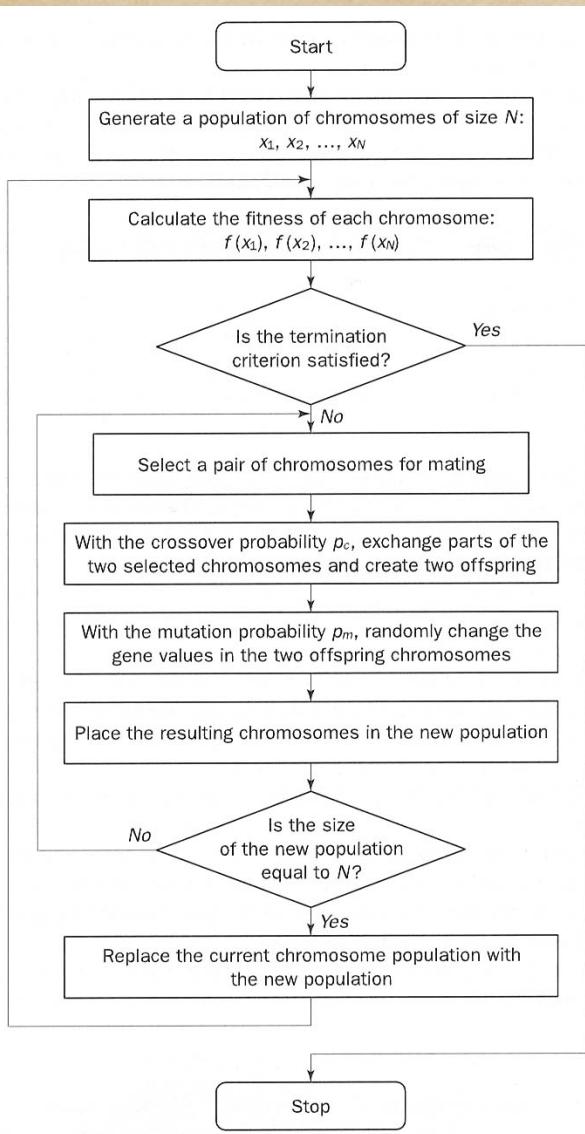
4 versus 1!



Example







procedure GeneticAlgorithm

{ $t = 0$;

 initialize $P(t)$;

 evaluate chromosomes in $P(t)$;

 while (not satisfied) do

 { $t += 1$;

 select $P(t)$ from $P(t-1)$;

 recombine chromosomes in $P(t)$;

 evaluate chromosomes in $P(t)$;

 }

}

[Negnevitsky 2002, Grefenstette 1990]

the GENESIS system

Genes can be bit sequence or multiple floats

Can specify or randomly initialize population

Selection by standard or ranked method

You write the evaluation function in C

```
double eval(str, length, vect, genes)
char str[];      /* bits as string */
int length;      /* length of str */
double vect[];   /* float genes */
int genes;       /* length of vect */
{ /* your function here */ }
```



Using Genesis for BlackJack

BJ eval function

- ◆ Use bit vector strategy
11111111111110000000
- ◆ Play BJ against dealer
- ◆ Report losing %age



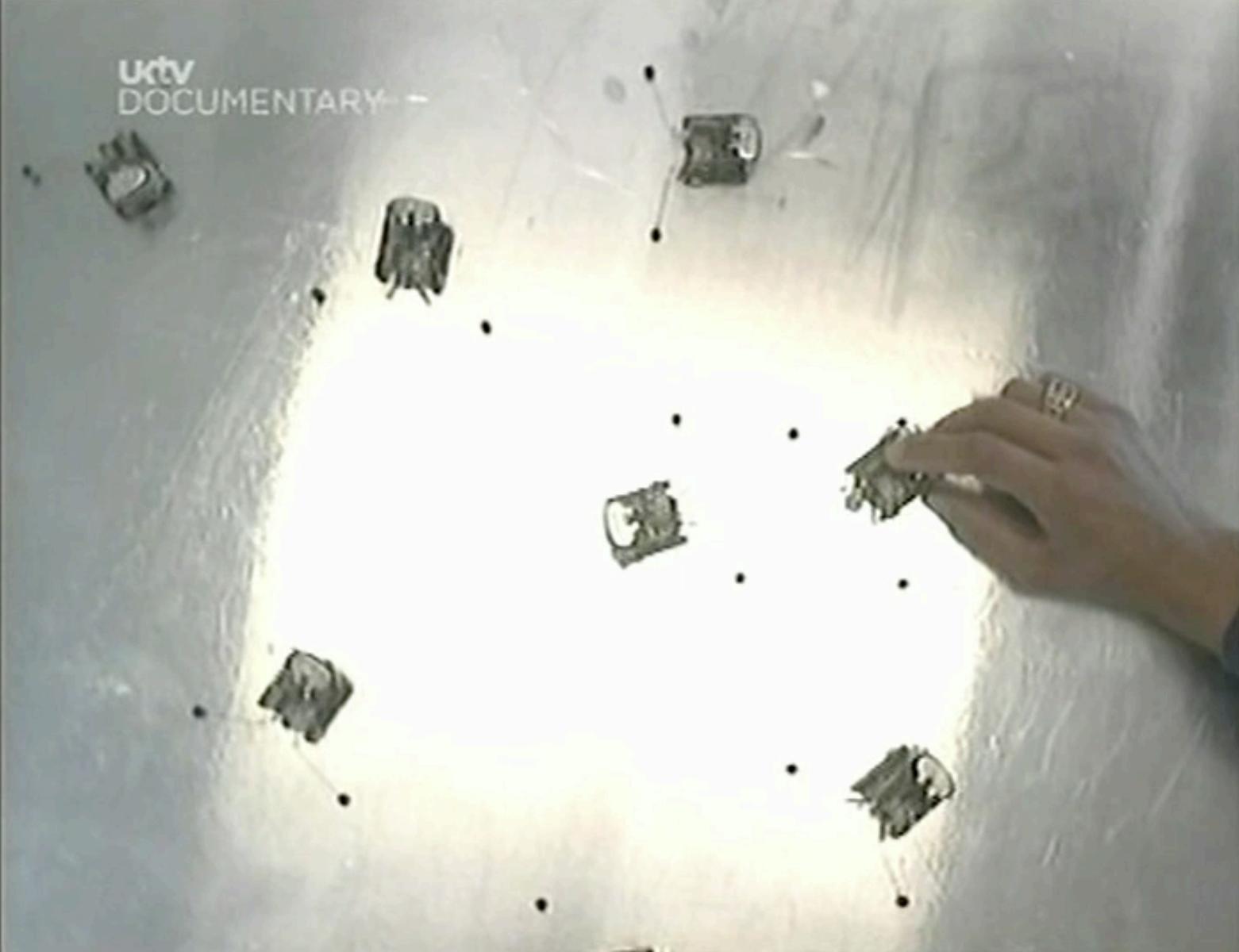
Genesis Setup

```
File suffix []: bj
Floating point representation [y]: n
Experiments [1]:
Total Trials [1000]: 5000
Population Size [50]:
Structure Length [30]: 22
Crossover Rate [0.6]:
Mutation Rate [0.001]: 0.01
Generation Gap [1.0]:
Scaling Window [5]:
Report Interval [100]:
Structures Saved [10]:
Max Gens w/o Eval [2]:
Dump Interval [0]: 5
Dumps Saved [0]: 5
Options [cel]:
Random Seed [123456789]:
Rank Min [0.75]:
```

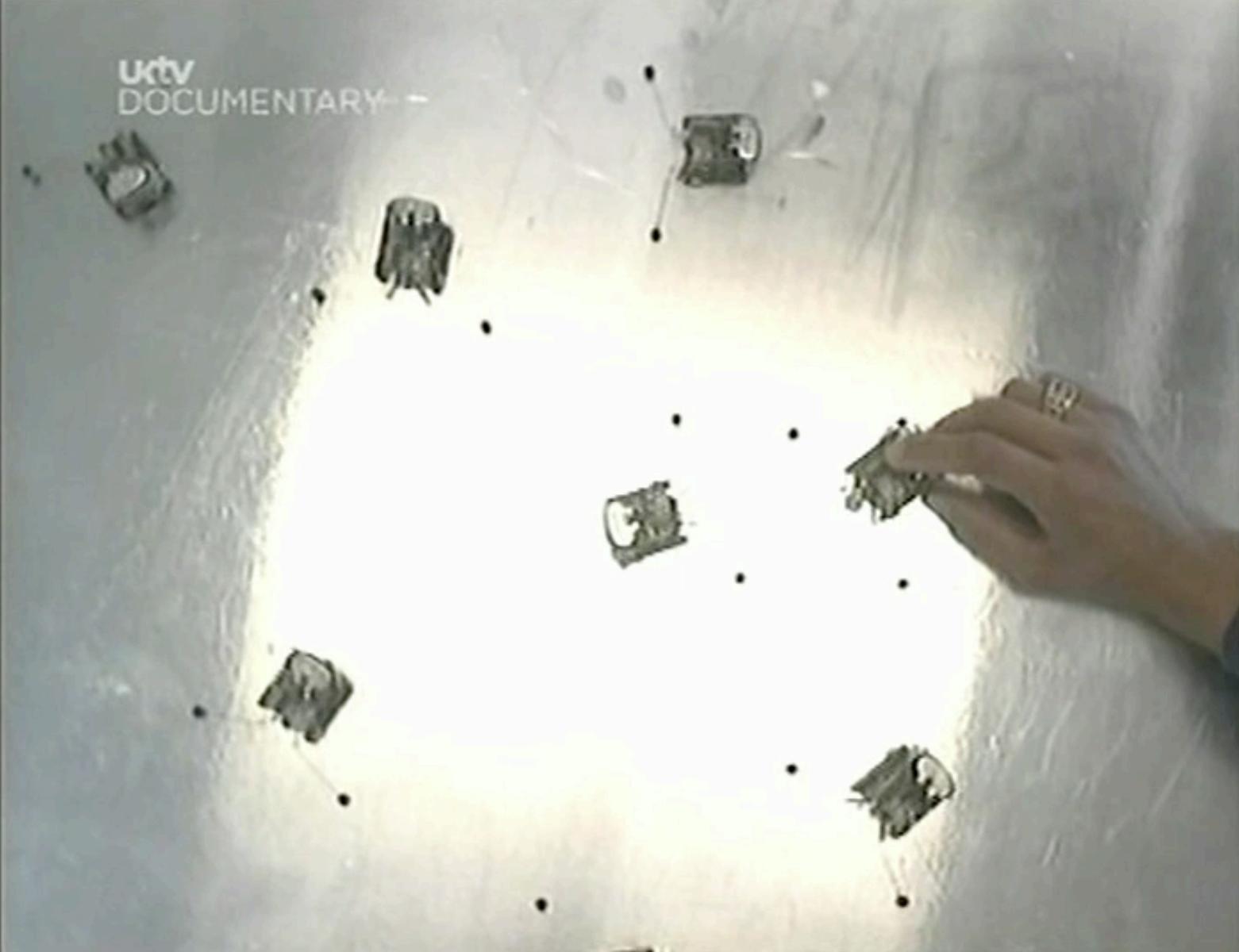
Individuals After 150 Generations

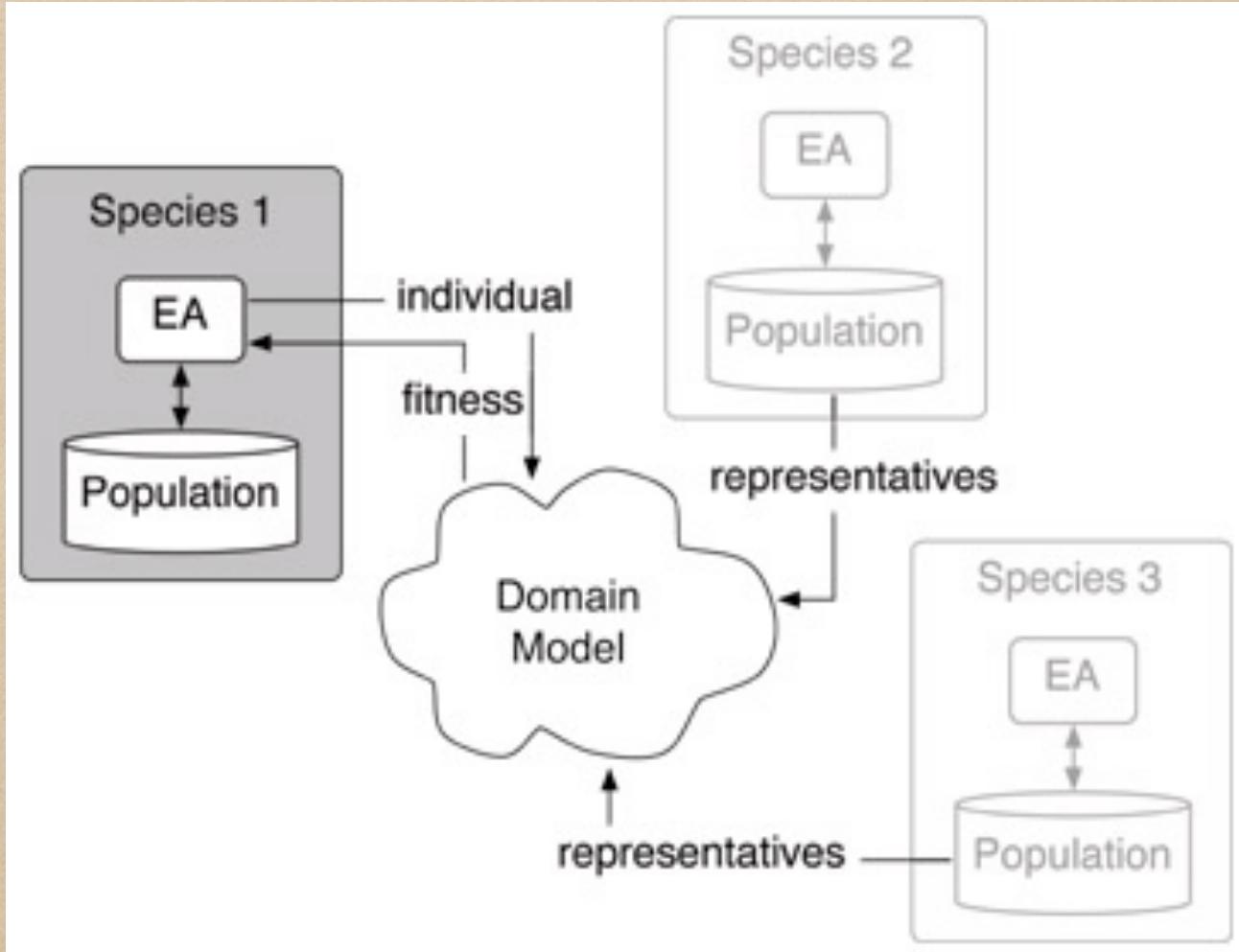
0000100011110111000000	4.5000e-01	111	2551
0010110011110111000000	4.6000e-01	133	2692
1000110111011100000000	4.6000e-01	74	1831
0000110011110111000000	4.3000e-01	83	2018
0000110011101111000000	4.4000e-01	3	122
1000000101110010100000	4.4000e-01	18	500
1000000101110010000000	4.6000e-01	43	1114
1000000101111110000000	4.7000e-01	33	863
1000110011101111000000	4.5000e-01	51	1309
1000110110100110000000	4.6000e-01	92	2219

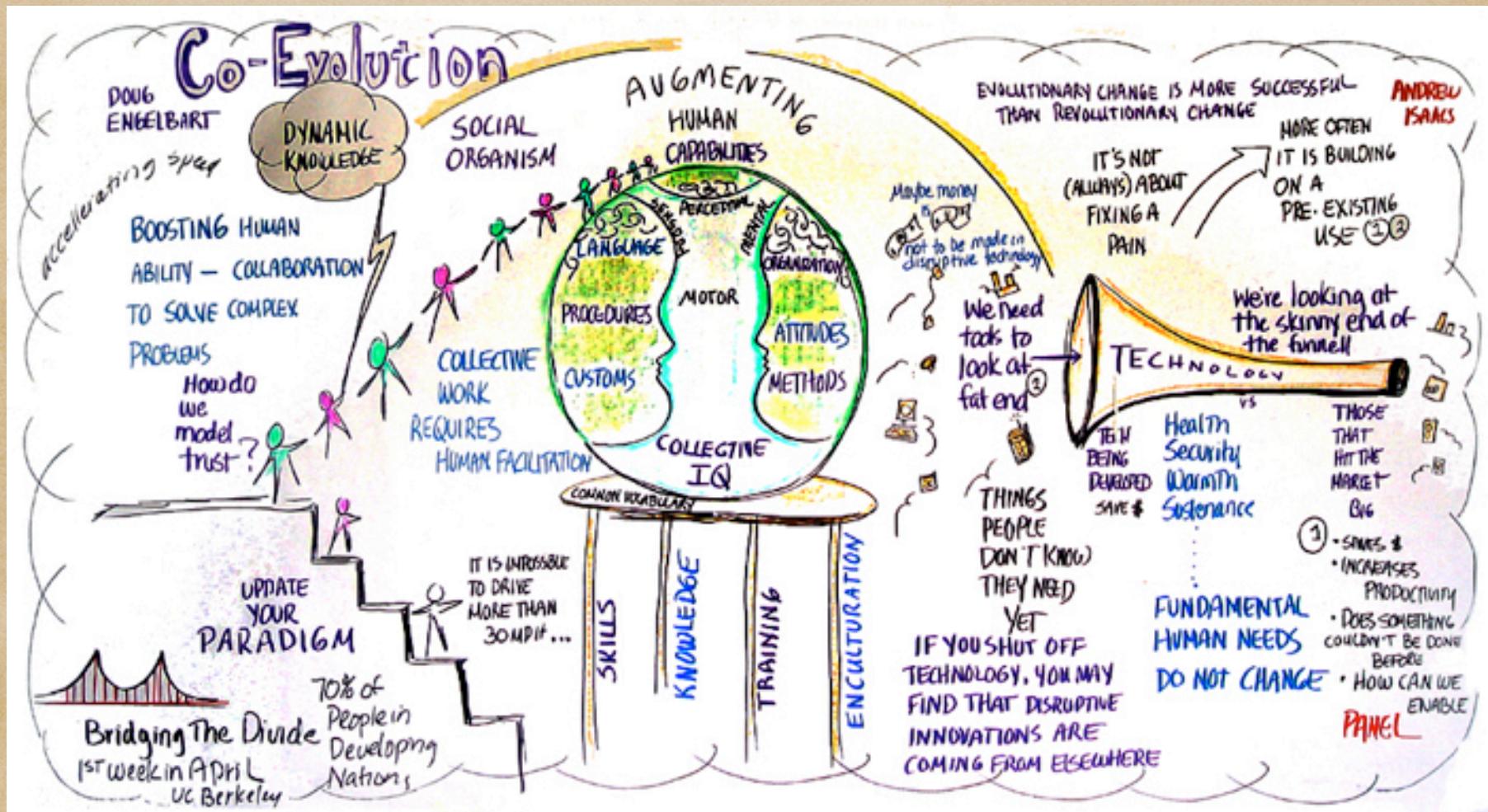
UKTV
DOCUMENTARY -



UKTV
DOCUMENTARY -







Co-Evolution Business & Technology Innovation

Small Step.
Major Change

MARKET-
PLACE
VOLATILITY

INVISIBLE
INTANGIBLES

TRANSFORM
ORGANIZATIONAL
FORMS

Thursday
morning
9.25.03

93% OF
LEADERS
WORRIED ABOUT
ADAPTATION
TO CHANGE

INFORMATION
ARBITRAGE

ORGANIZATIONAL
INTELLIGENCE

* LOOK AT CAUSAL
FACTORS
WHAT REALLY MOTIVATES?

JEFF KELLY

NEW TOOLS
AND
NEW APPROACH

TO HUMAN
RELATIONSHIPS:

BUILDING
RELATIONSHIPS

FOR
MUTUAL
VALUE

EXCHANGE

RELATIONSHIP
MANAGEMENT

CONTENT

COMMUNITY

LEARNING

CONTEXT

FLUID

FACE
TO
FACE

CONVERSATION

ON DEMAND

BUSINESS

Zero Degrees of
Separation

On Demand
Business

The Customer

is the

Business —

Peter Drucker

ORGANIZATIONAL
INTELLIGENCE

* LOOK AT CAUSAL
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WHAT REALLY MOTIVATES?

JEFF KELLY

CONTROLLED
MUTUALITY

Rob Flaherty

AWAY FROM
BUSINESS AS VALUE ALLOCATION

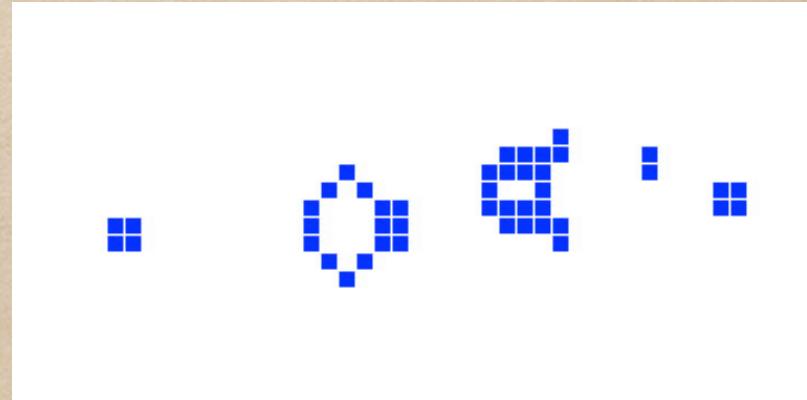
LAYERS
OF
PROBLEMS

CAN ON-DEMAND
HELP IMPROVE
EDUCATION?

LANGUAGE IS CHANGING
AS WE ARE CREATING
A NEW DIRECTION

* MAKING THE SOFT HARD —
MUTUAL VALUE EXCHANGE —
LET'S PUT SOME METRICS BEHIND IT

Which is alive?



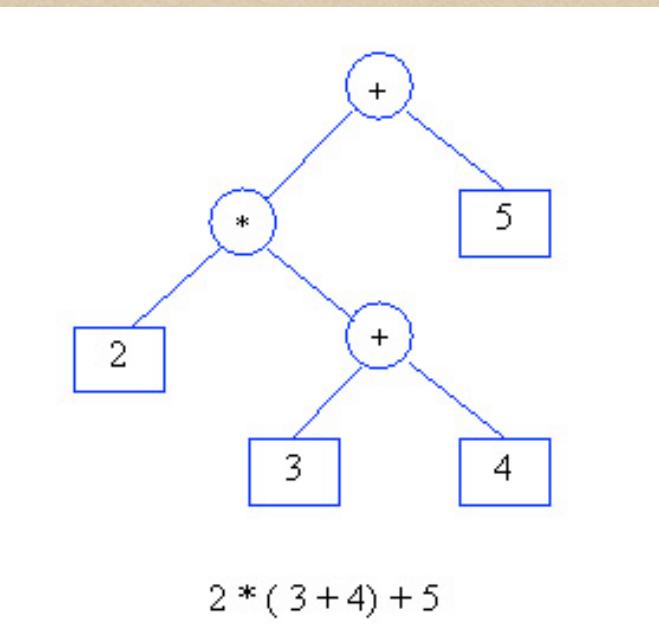
Suppose we wanted to learn something about right triangles.

Side a	Side b	Hypotenuse c	Side a	Side b	Hypotenuse c
3	5	5.830952	12	10	15.620499
8	14	16.124515	21	6	21.840330
18	2	18.110770	7	4	8.062258
32	11	33.837849	16	24	28.844410
4	3	5.000000	2	9	9.219545

That is, learn a function that takes a and b and predicts the hypotenuse.

Genetic Programming

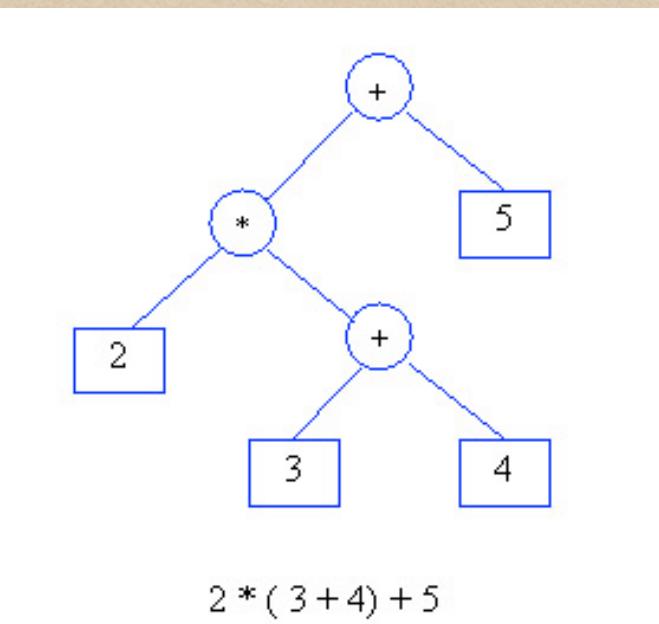
Representing Equations



Simple expression tree
(could use variables a & b)

Abstract Syntax Tree
(parse tree / compilers)

Representing Equations



Simple expression tree
(could use variables a & b)

Abstract Syntax Tree
(parse tree / compilers)

$(+ (* 2 (+ 3 4)) 5)$

LISP S-expressions

$(/ (- (sqrt (+ (* a a) (- a b)))) a) (* a b)),$

which is equivalent to

$$\frac{\sqrt{a^2 + (a - b)} - a}{ab},$$

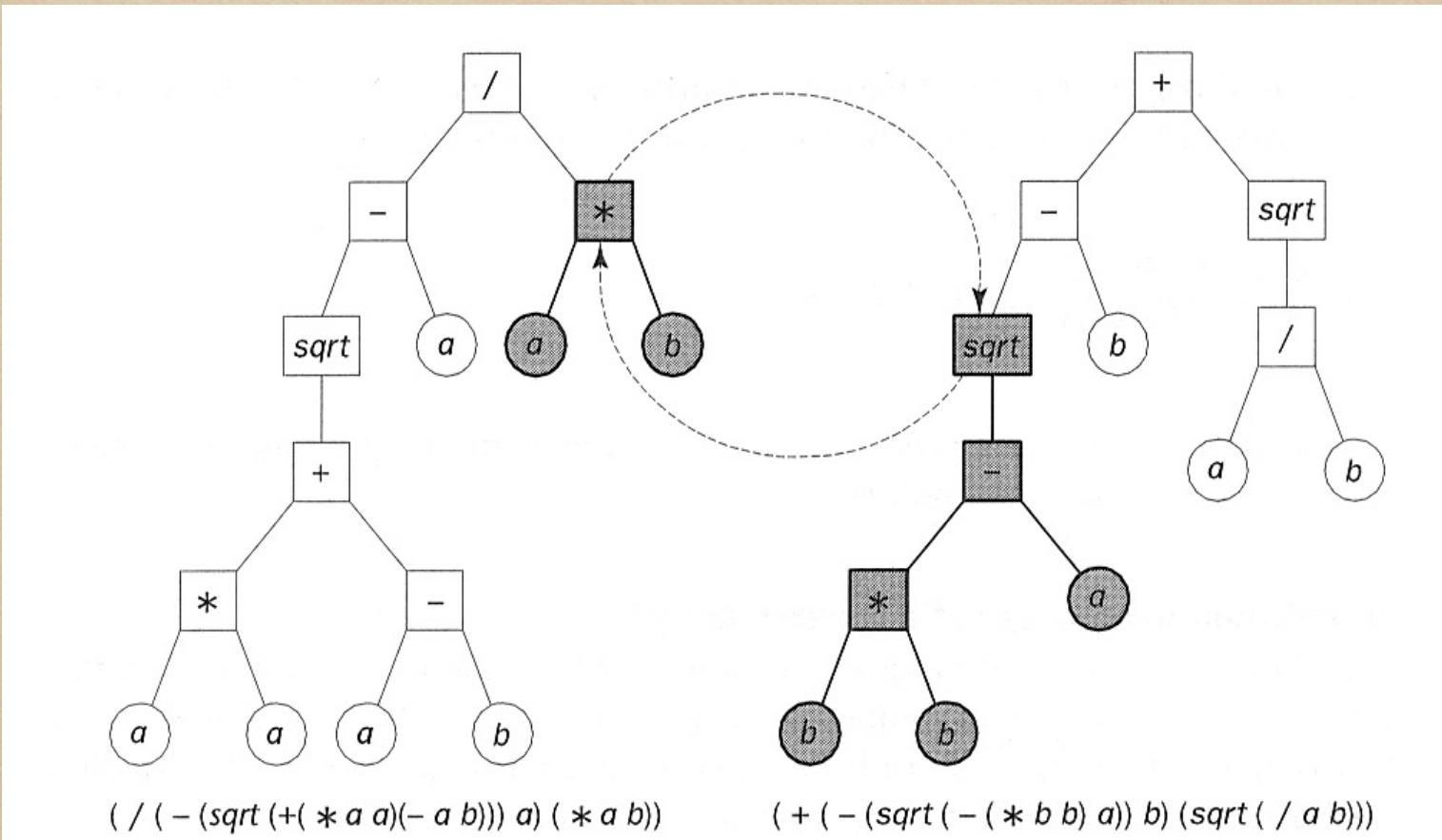
and

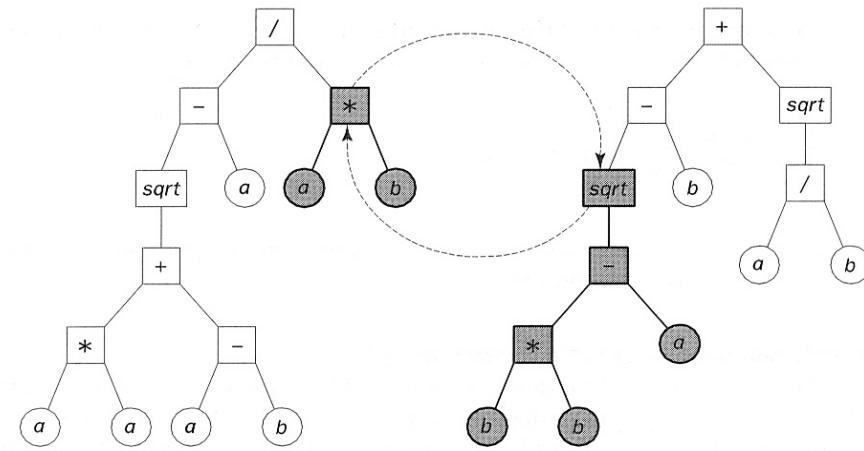
$(+ (- (sqrt (- (* b b) a))) b) (sqrt (/ a b))),$

which is equivalent to

$$\left(\sqrt{b^2 - a} - b \right) + \sqrt{\frac{a}{b}}.$$

Crossover

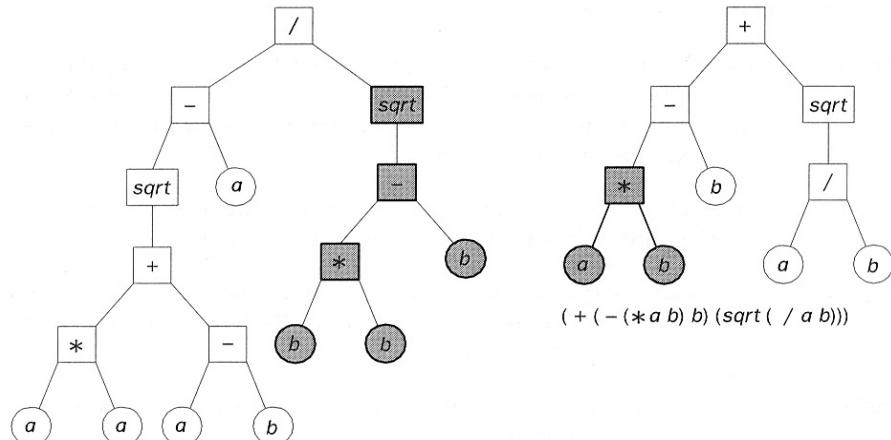




$(/ (- (sqrt (+(* a a)(- a b))) a) (* a b))$

$(+ (- (sqrt (- (* b b) a)) b) (sqrt (/ a b)))$

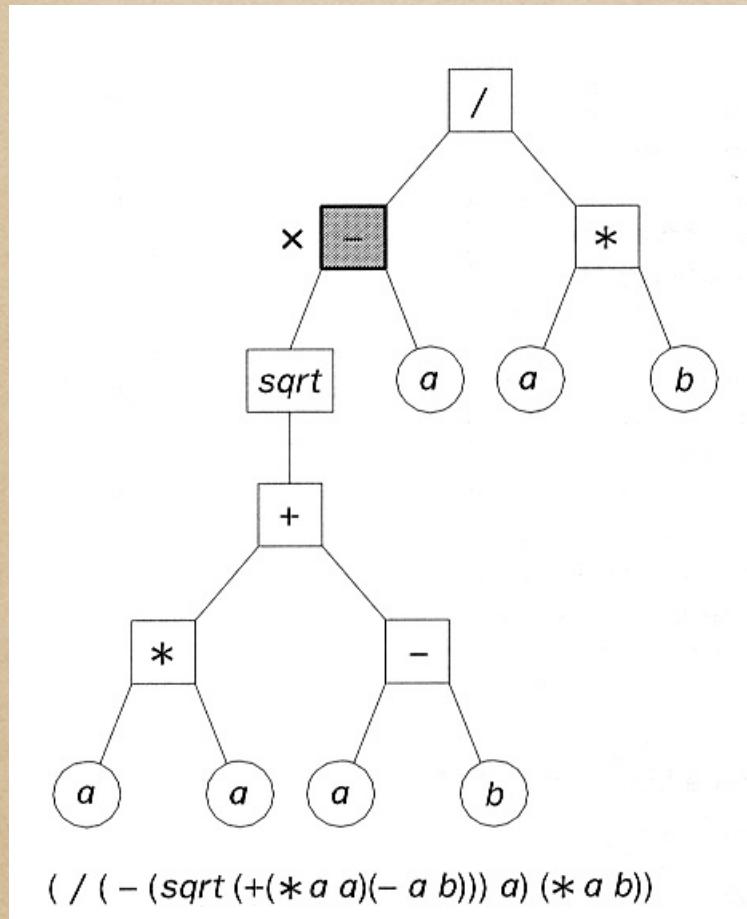
(a)



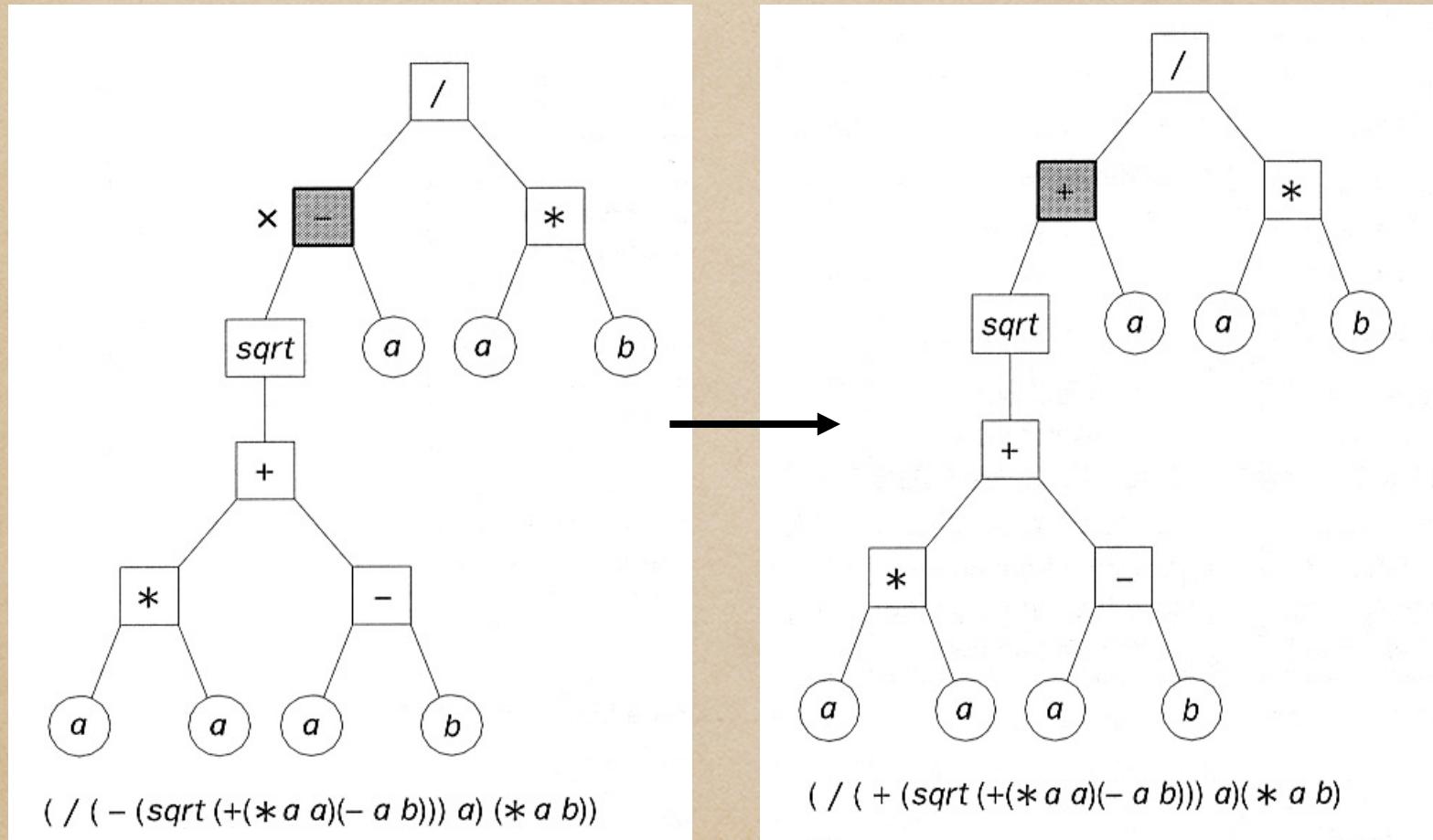
$(/ (- (sqrt (+(* a a)(- a b))) a) (sqrt (- (* b b) a)))$

(b)

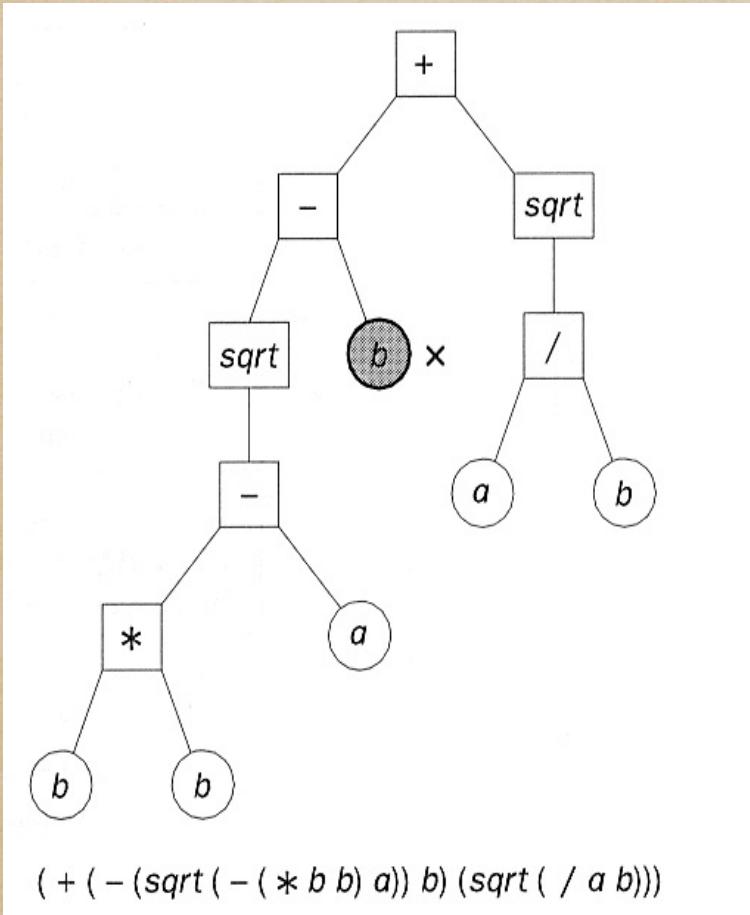
Mutation - operators



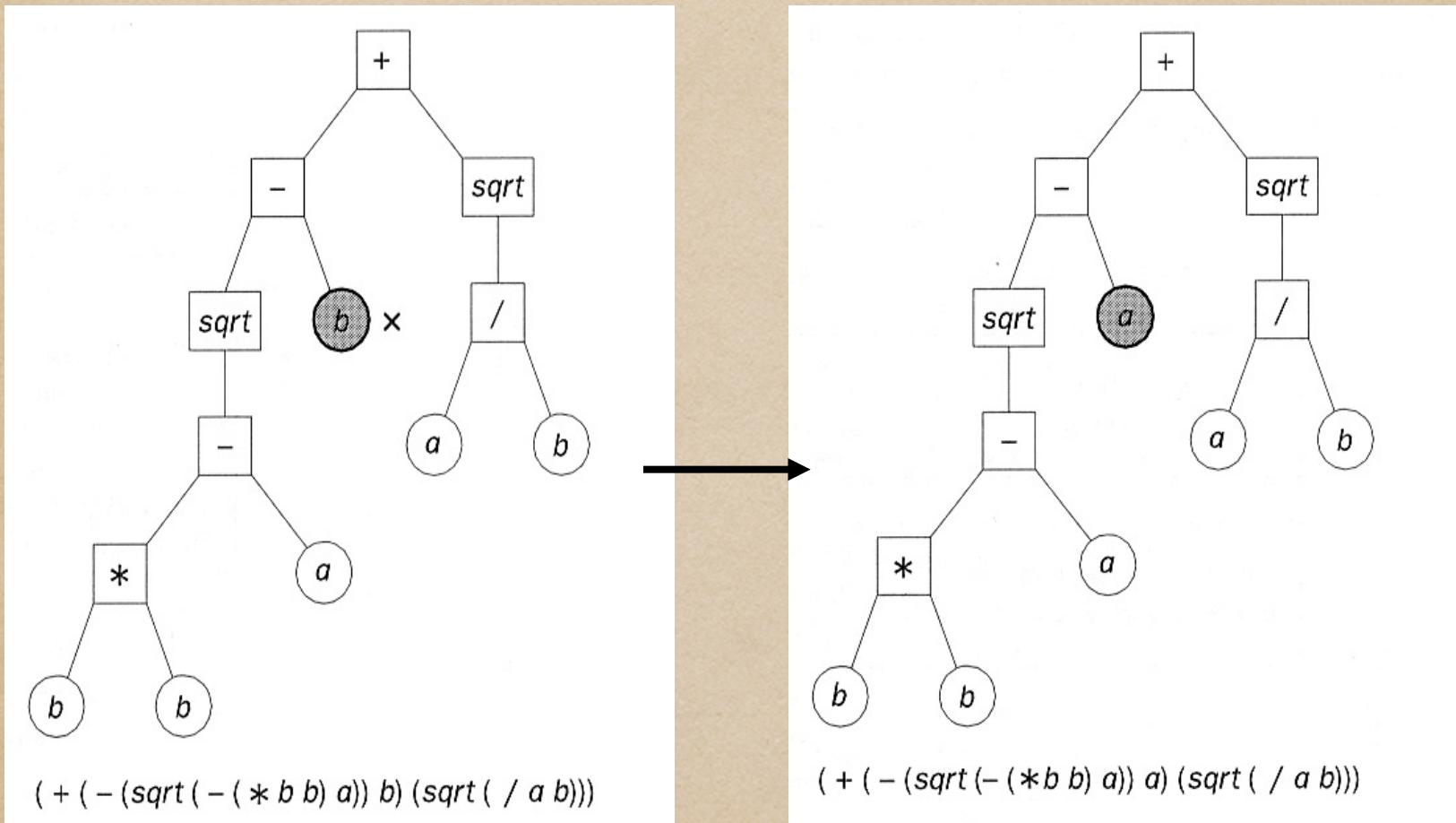
Mutation - operators



Mutation - operands



Mutation - operands

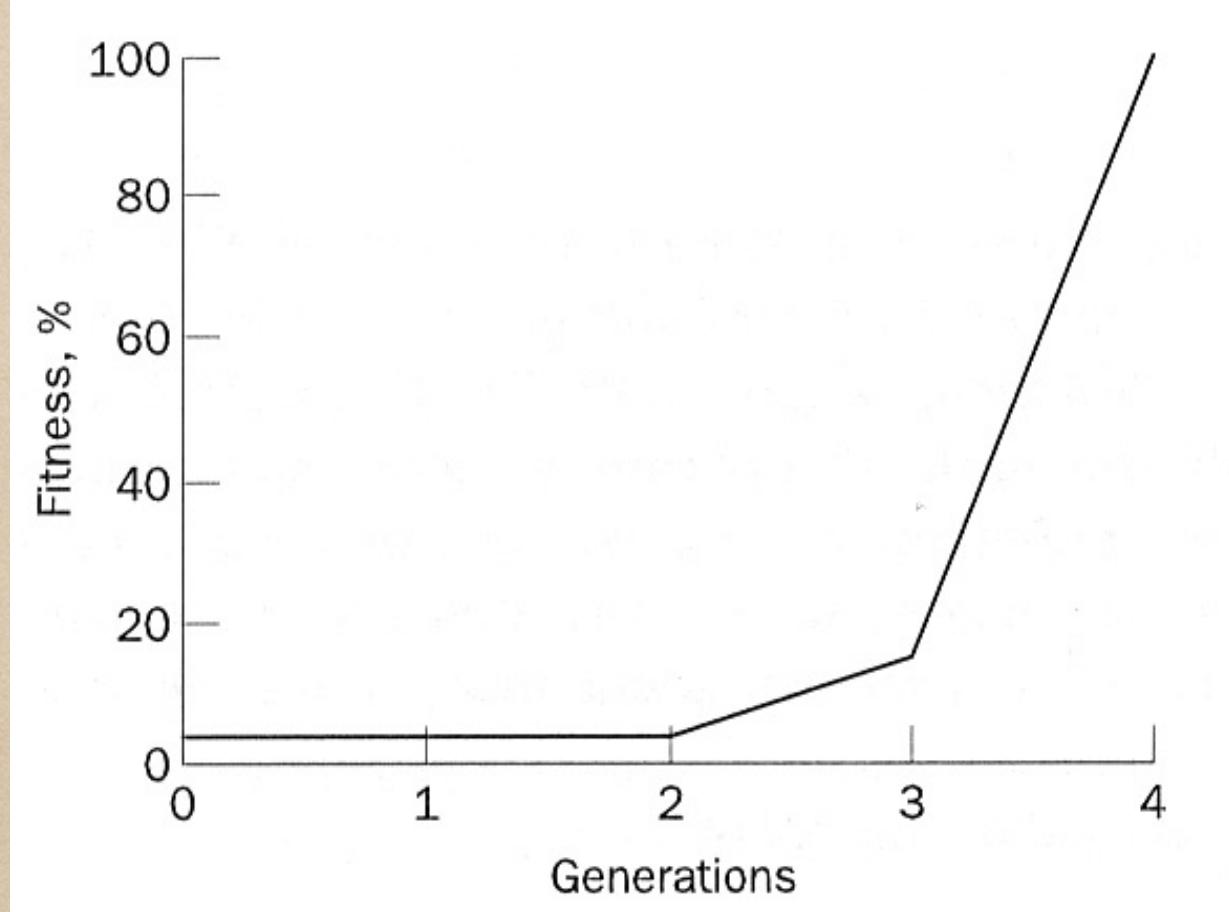


Fitness function

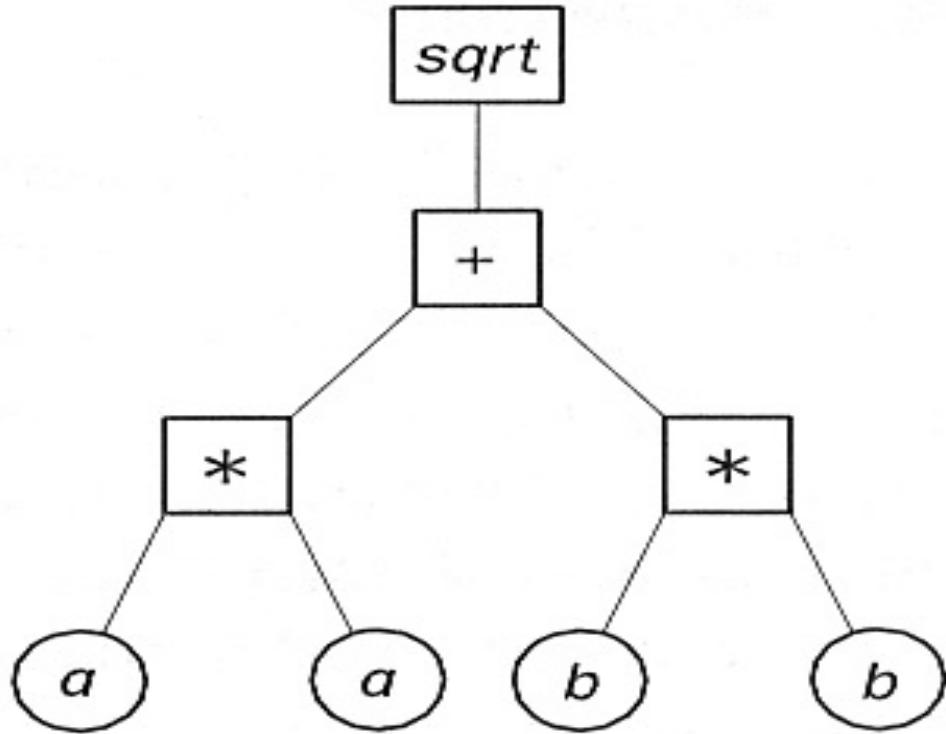
Side a	Side b	Hypotenuse c	Side a	Side b	Hypotenuse c
3	5	5.830952	12	10	15.620499
8	14	16.124515	21	6	21.840330
18	2	18.110770	7	4	8.062258
32	11	33.837849	16	24	28.844410
4	3	5.000000	2	9	9.219545

Test on sample instances,
compute similarity to desired value.

Results



Results



Best of generation

(Online) Readings

- ◆ Why GA work
- ◆ Case Studies

THERE ARE NO
"STUPID" QUESTIONS
ONLY
AWKWARD SILENCES.

— your course
illustrator

Study Questions

Draw a flowchart implementing the main steps
in a genetic algorithm.

Explain crossover and mutation on both bit
strings and Lisp S-expressions.

What is a schema? How does Holland's Schema
Theorem explain why genetic algorithms work?



Do genetic algorithms cover
the complete search space?