

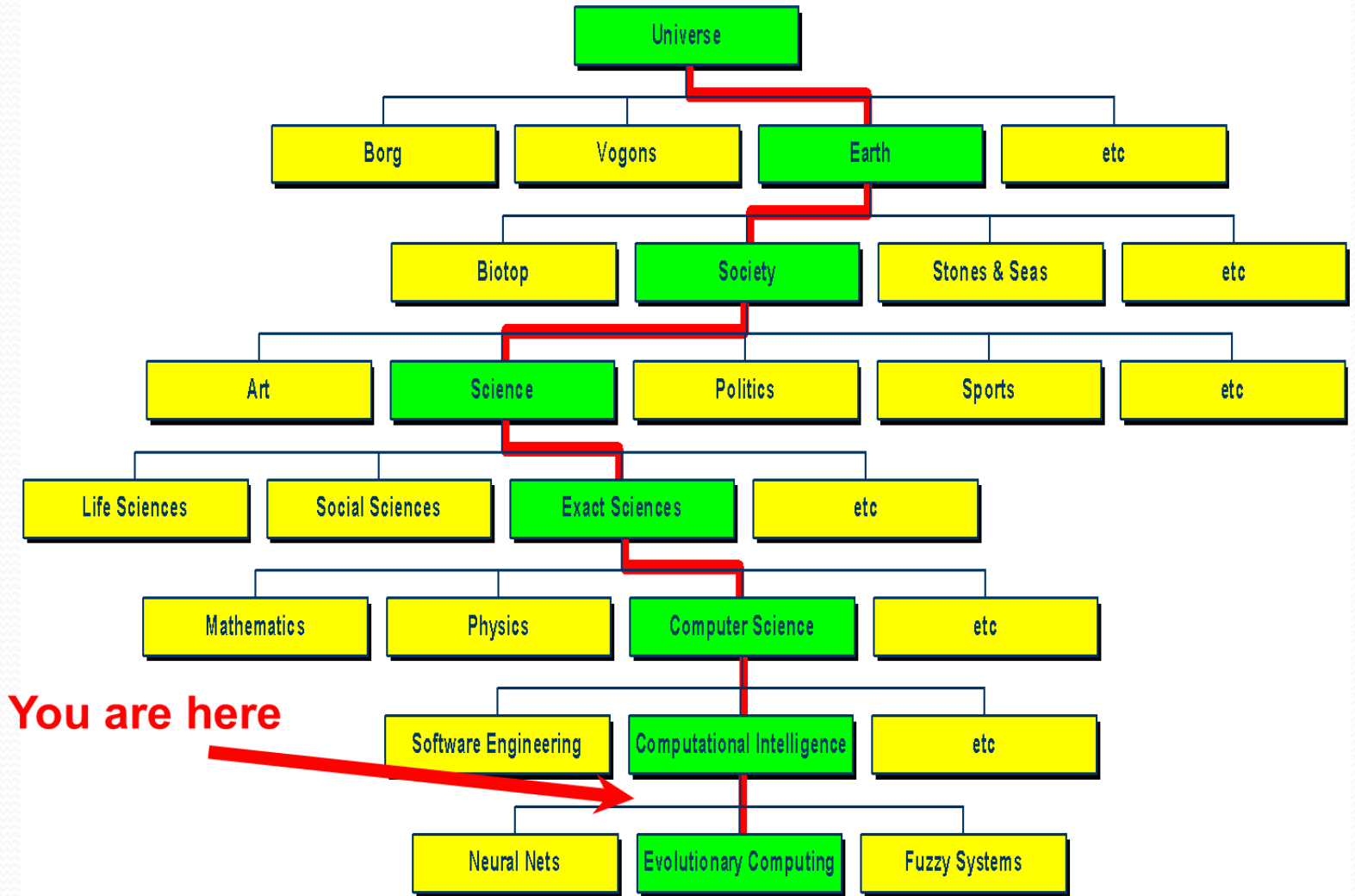
Introduction

Chapter 1

Contents

- Positioning of EC and the basic EC metaphor
- Historical perspective
- Biological inspiration:
 - Darwinian evolution theory (simplified!)
 - Genetics (simplified!)
- Motivation for EC
- What can EC do: examples of application areas
- Demo: evolutionary magic square solver

Positioning of EC



Positioning of EC

- EC is part of computer science
- EC is not part of life sciences/biology
- Biology delivered inspiration and terminology
- EC can be applied in biological research

The Main EC Metaphor

EVOLUTION

PROBLEM SOLVING

Environment



Problem

Individual



Candidate Solution

Fitness



Quality

Fitness → chances for survival and reproduction

Quality → chance for seeding new solutions CV

Fitness in nature: observed, 2ndary, EC: primary

Brief EC history

- 1948, Turing:
proposes “genetical or evolutionary search”
- 1962, Bremermann
optimization through evolution and recombination
- 1964, Rechenberg
introduces evolution strategies
- 1965, L. Fogel, Owens and Walsh
introduce evolutionary programming
- 1975, Holland
introduces genetic algorithms
- 1992, Koza
introduces genetic programming

Brief EC history 2

- 1985: first international conference (ICGA)
- 1990: first international conference in Europe (PPSN)
- 1993: first scientific EC journal (MIT Press)
- 1997: launch of European EC Research Network EvoNet

EC in the early 21st Century

- 3 major EC conferences, about 10 small related ones
- 4 scientific core EC journals
- 1000+ EC-related papers published last year(estimate)
- uncountable (meaning: many) applications
- uncountable (meaning: ?) consultancy and R&D firms
- part of many university curricula

Darwinian Evolution 1: Survival of the fittest

- All environments have finite resources
(i.e., can only support a limited number of individuals)
- Life forms have basic instinct/ lifecycles geared towards reproduction
- Therefore some kind of selection is inevitable
- Those individuals that compete for the resources most effectively have increased chance of reproduction
- Note: fitness in natural evolution is a derived, secondary measure, i.e., we (humans) assign a high fitness to individuals with many offspring

Darwinian Evolution 2:

Diversity drives change

- Phenotypic traits:
 - Behaviour / physical differences that affect response to environment
 - Partly determined by inheritance, partly by factors during development
 - Unique to each individual, partly as a result of random changes
- If phenotypic traits:
 - Lead to higher chances of reproduction
 - Can be inheritedthen they will tend to increase in subsequent generations,
- leading to new combinations of traits ...

Darwinian Evolution: Summary

- Population consists of diverse set of individuals
- Combinations of traits that are better adapted tend to increase representation in population

Individuals are “units of selection”

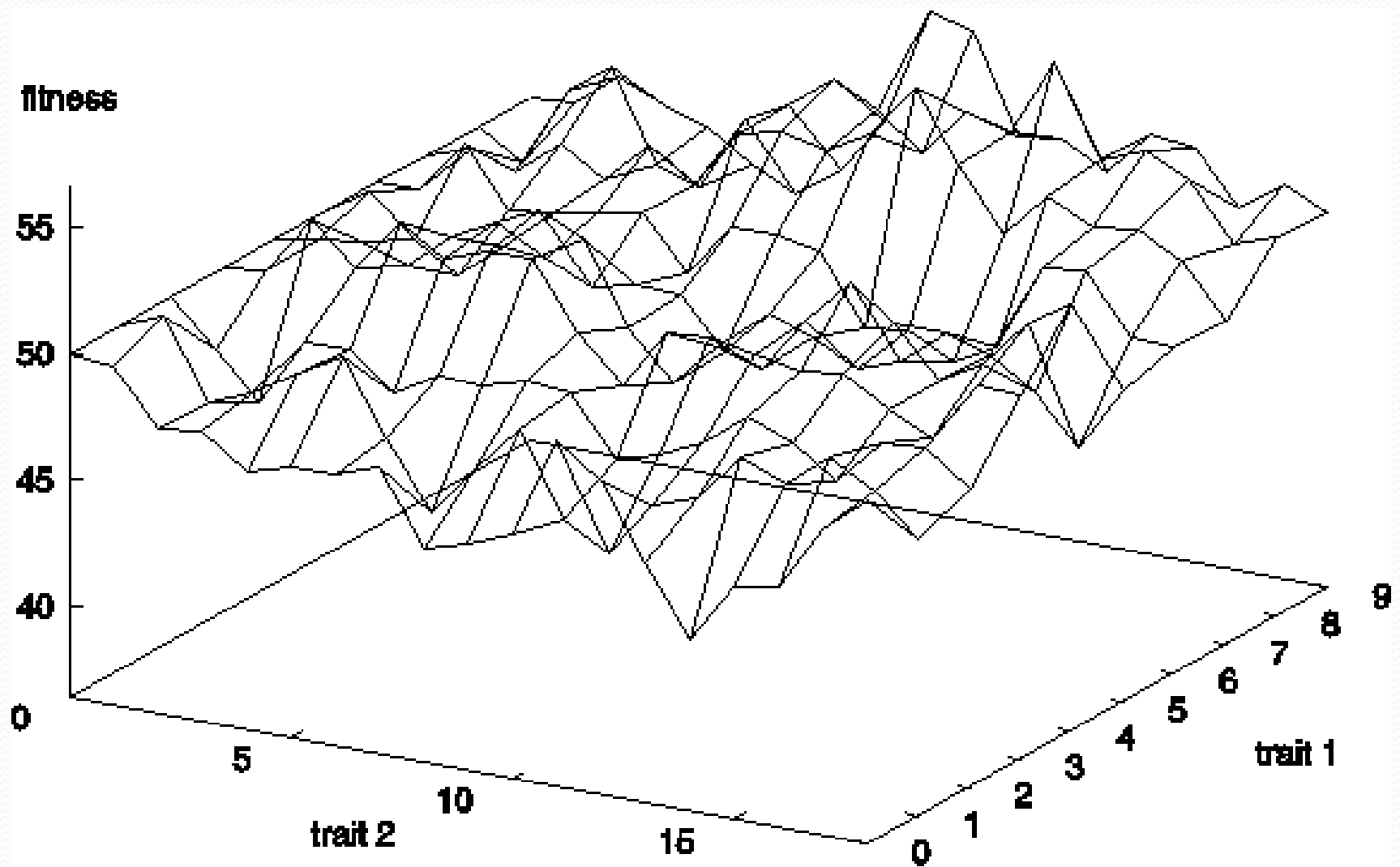
- Variations occur through random changes yielding constant source of diversity, coupled with selection means that:

Population is the “unit of evolution”

- Note the absence of “guiding force”

Adaptive landscape metaphor (Wright, 1932)

- Can envisage population with n traits as existing in a $n+1$ -dimensional space (landscape) with height corresponding to fitness
- Each different individual (phenotype) represents a single point on the landscape
- Population is therefore a “cloud” of points, moving on the landscape over time as it evolves - adaptation



Adaptive landscape metaphor (cont'd)

- Selection “pushes” population up the landscape
- Genetic drift:
 - random variations in feature distribution
 - (+ or -) arising from sampling error
 - can cause the population “melt down” hills, thus crossing valleys and leaving local optima

Natural Genetics

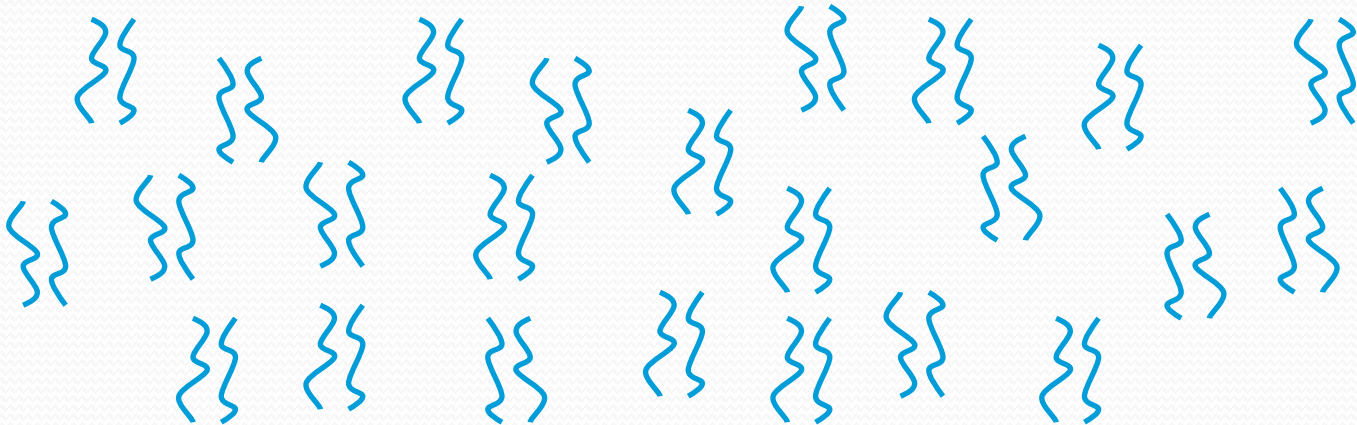
- The information required to build a living organism is coded in the DNA of that organism
- Genotype (DNA inside) determines phenotype
- Genes → phenotypic traits is a complex mapping
 - One gene may affect many traits (pleiotropy)
 - Many genes may affect one trait (polygeny)
- Small changes in the genotype lead to small changes in the organism (e.g., height, hair colour)

Genes and the Genome

- Genes are encoded in strands of DNA called chromosomes
- In most cells, there are two copies of each chromosome (diploidy)
- The complete genetic material in an individual's genotype is called the Genome
- Within a species, most of the genetic material is the same

Example: Homo Sapiens

- Human DNA is organised into chromosomes
- Human body cells contains 23 pairs of chromosomes which together define the physical attributes of the individual:

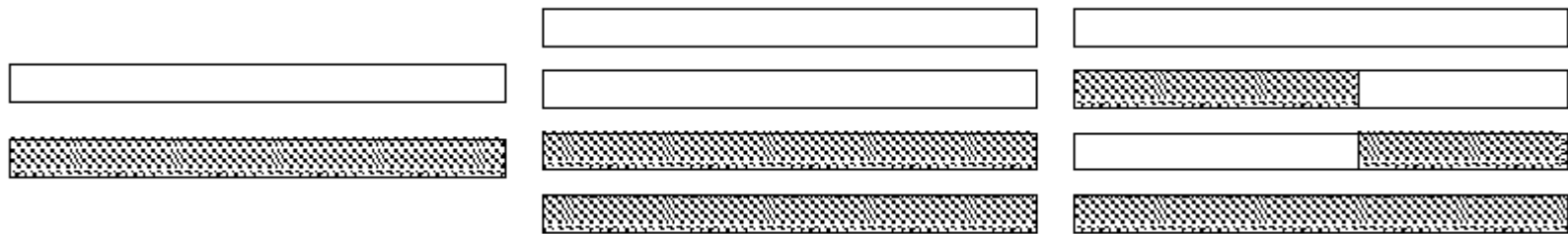


Reproductive Cells

- Gametes (sperm and egg cells) contain 23 individual chromosomes rather than 23 pairs
- Cells with only one copy of each chromosome are called haploid
- Gametes are formed by a special form of cell splitting called meiosis
- During meiosis the pairs of chromosome undergo an operation called *crossing-over*

Crossing-over during meiosis

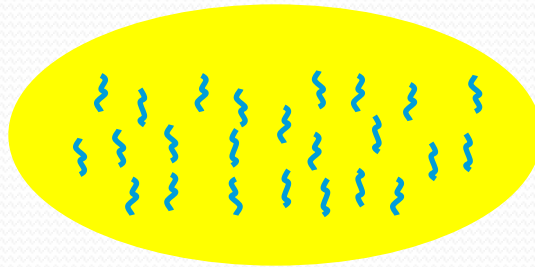
- Chromosome pairs align and duplicate
- Inner pairs link at a *centromere* and swap parts of themselves



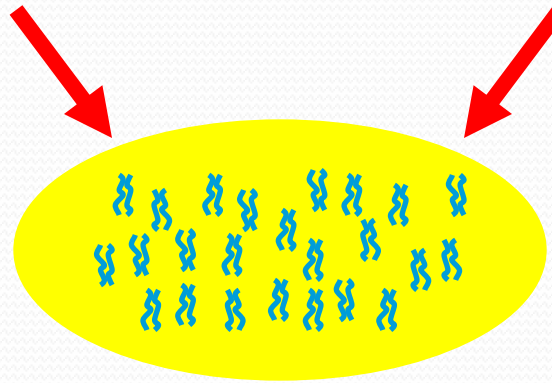
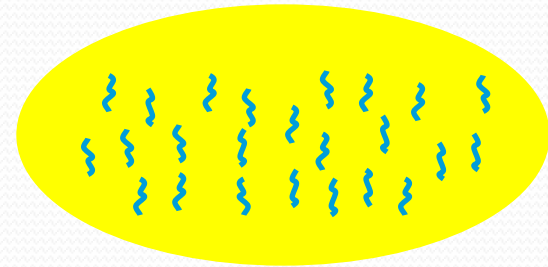
- Outcome is one copy of maternal/paternal chromosome plus two entirely new combinations
- After crossing-over one of each pair goes into each gamete

Fertilisation

Sperm cell from Father



Egg cell from Mother



New person cell (zygote)

Introduction

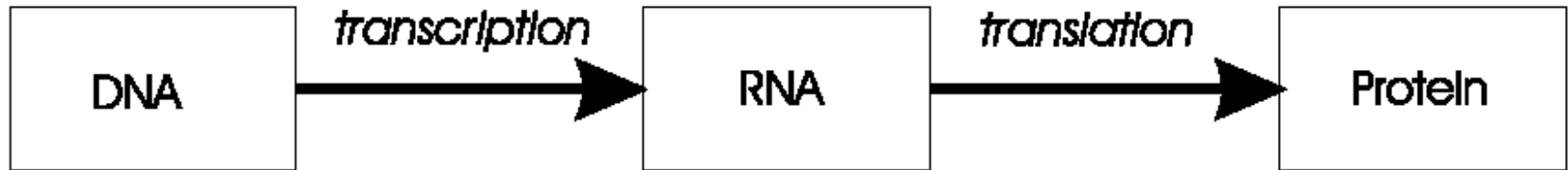
After fertilisation

- New zygote rapidly divides etc creating many cells all with the same genetic contents
- Although all cells contain the same genes, depending on, for example where they are in the organism, they will behave differently
- This process of differential behaviour during development is called ontogenesis
- All of this uses, and is controlled by, the same mechanism for decoding the genes in DNA

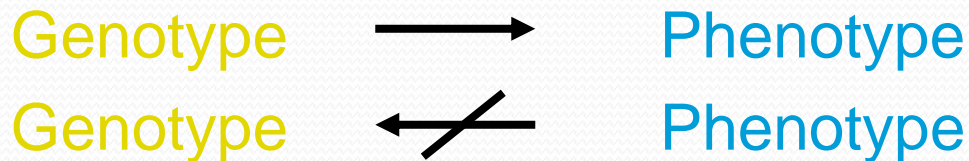
Genetic code

- All proteins in life on earth are composed of sequences built from 20 different amino acids
- DNA is built from four nucleotides in a double helix spiral: purines A,G; pyrimidines T,C
- Triplets of these form *codons*, each of which codes for a specific amino acid
- Much redundancy:
 - purines complement pyrimidines
 - the DNA contains much rubbish
 - $4^3=64$ codons code for 20 amino acids
 - genetic code = the mapping from codons to amino acids
- **For all natural life on earth, the genetic code is the same !**

Transcription, translation



A central claim in molecular genetics: only one way flow



Lamarckism (saying that acquired features can be inherited) is thus wrong!

Mutation

- Occasionally some of the genetic material changes very slightly during this process (replication error)
- This means that the child might have genetic material information not inherited from either parent
- This can be
 - catastrophic: offspring is not viable (most likely)
 - neutral: new feature does not influence fitness
 - advantageous: strong new feature occurs
- Redundancy in the genetic code forms a good way of error checking

Motivations for EC 1

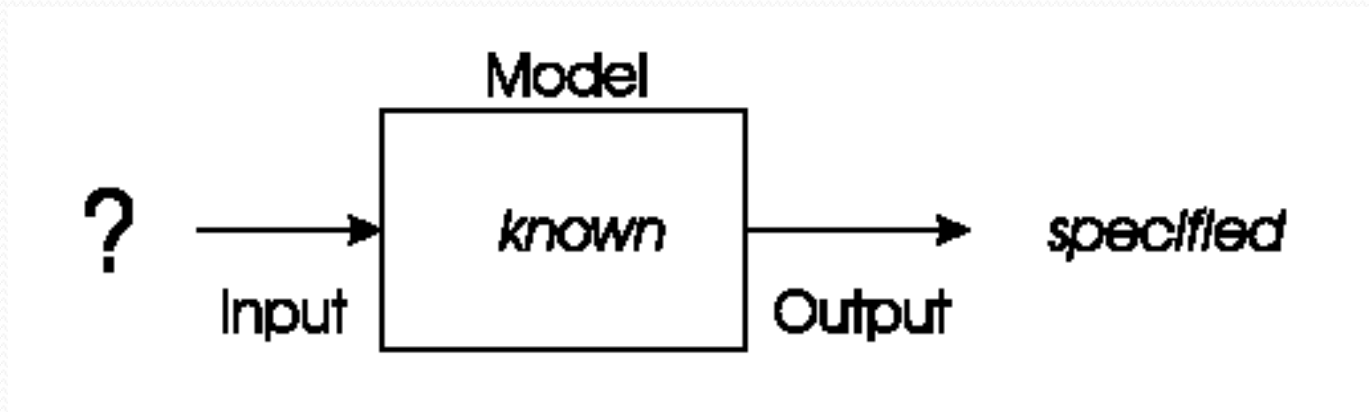
- Nature has always served as a source of inspiration for engineers and scientists
- The best problem solver known in nature is:
 - **the (human) brain** that created “the wheel, New York, wars and so on” (after Douglas Adams’ Hitch-Hikers Guide)
 - **the evolution mechanism** that created the human brain (after Darwin’s Origin of Species)
- Answer 1 → neurocomputing
- Answer 2 → evolutionary computing

Motivations for EC 2

- Developing, analyzing, applying **problem solving** methods a.k.a. algorithms **is a central theme** in mathematics and computer science
- **Time** for thorough problem analysis **decreases**
- **Complexity** of problems to be solved **increases**
- Consequence: **ROBUST PROBLEM SOLVING** technology needed

Problem type 1 : Optimisation

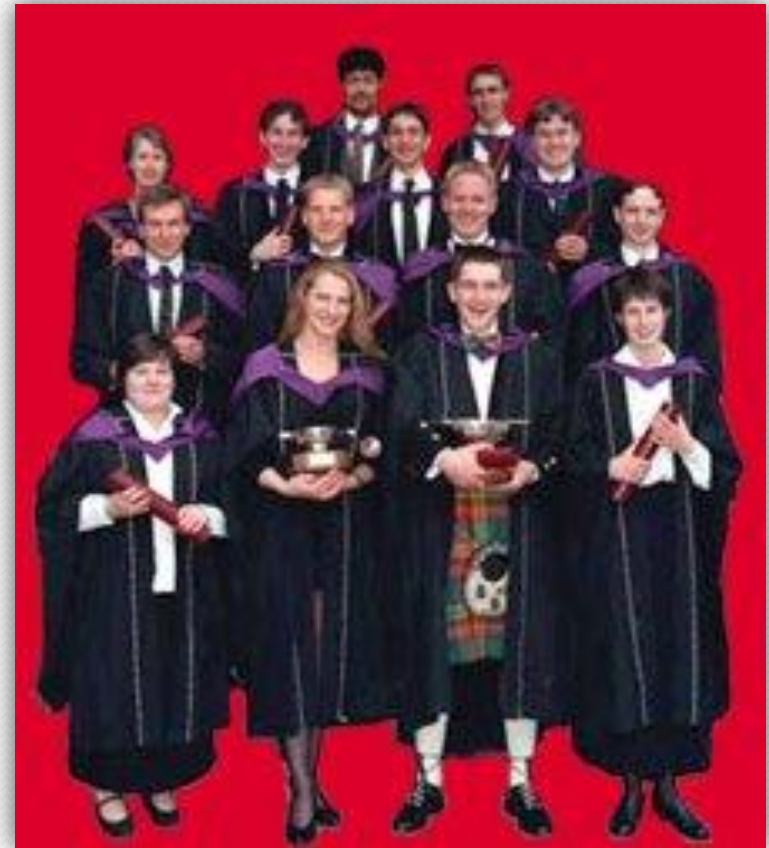
- We have a model of our system and seek inputs that give us a specified goal



- e.g.
 - time tables for university, call center, or hospital
 - design specifications, etc etc

Optimisation example 1: university timetabling

- Enormously big search space
- Timetables must be *good*
- “Good” is defined by a number of competing criteria
- Timetables must be feasible
- Vast majority of search space is infeasible





Evaluations: 717

Last change at: 431

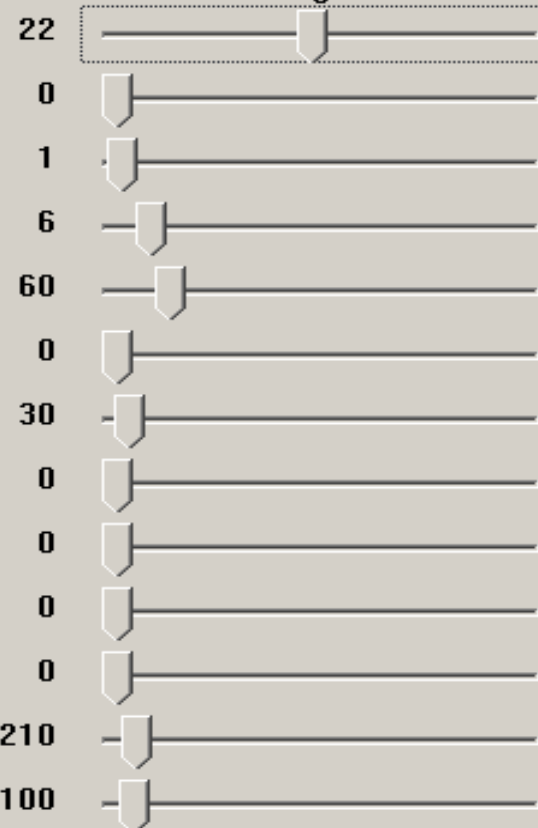
Evaluations per minute: 14952

Displaying: Best

Run No: 0

Placing Events is a Special Priority

Targets



Unplaced Events: 1

Changes: 0

Five O'Clock Classes: 13

Wed Afternoon Classes: 13

Gaps in Student Day: 7046

Lone Classes: 17708

Long Intensive: 0

Overloaded Lecturers: 26

No Teaching Free Day: 52

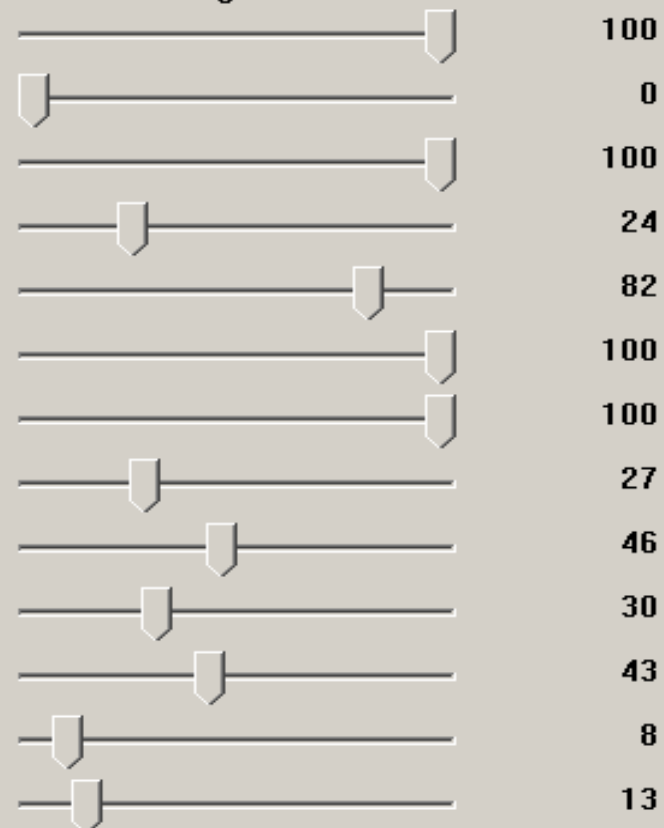
Instant Site Changes: 0

Site Changes: 0

Location Changes: 49738

Room Changes: 11869

Weights

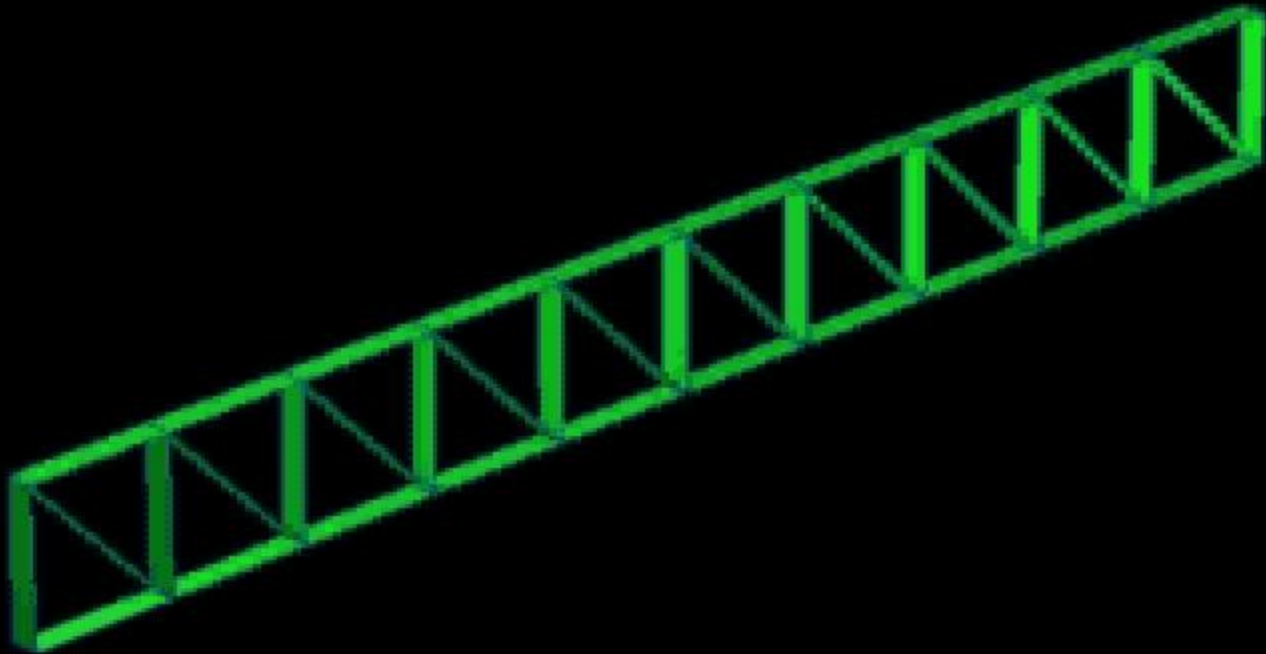


Progress: 55%

Optimisation example 2: satellite structure

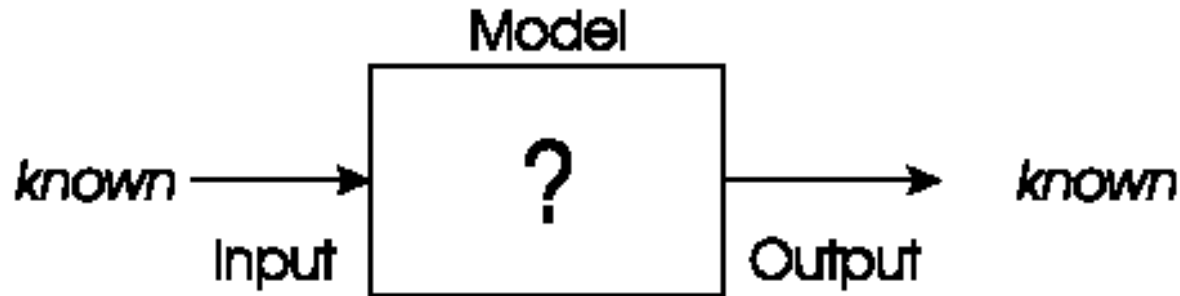
- Optimised satellite designs for NASA to maximize vibration isolation
- Evolving: design structures
- Fitness: vibration resistance
- **Evolutionary “creativity”**





Problem types 2: Modelling

- We have corresponding sets of inputs & outputs and seek model that delivers correct output for every known input



- Evolutionary machine learning

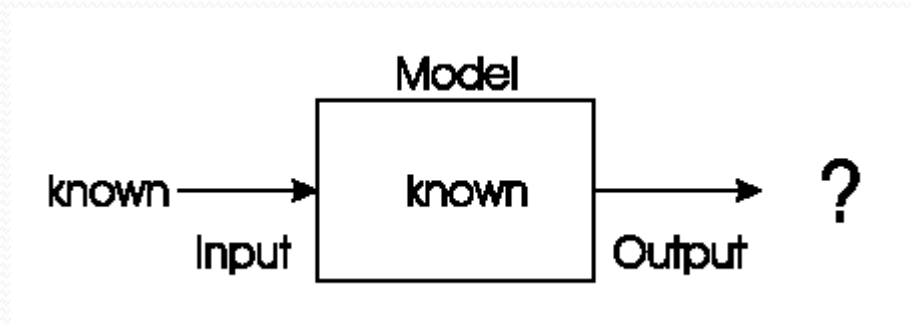
Modelling example: load applicant creditibility

- British bank evolved creditability model to predict loan paying behavior of new applicants
- Evolving: prediction models
- Fitness: model accuracy on historical data



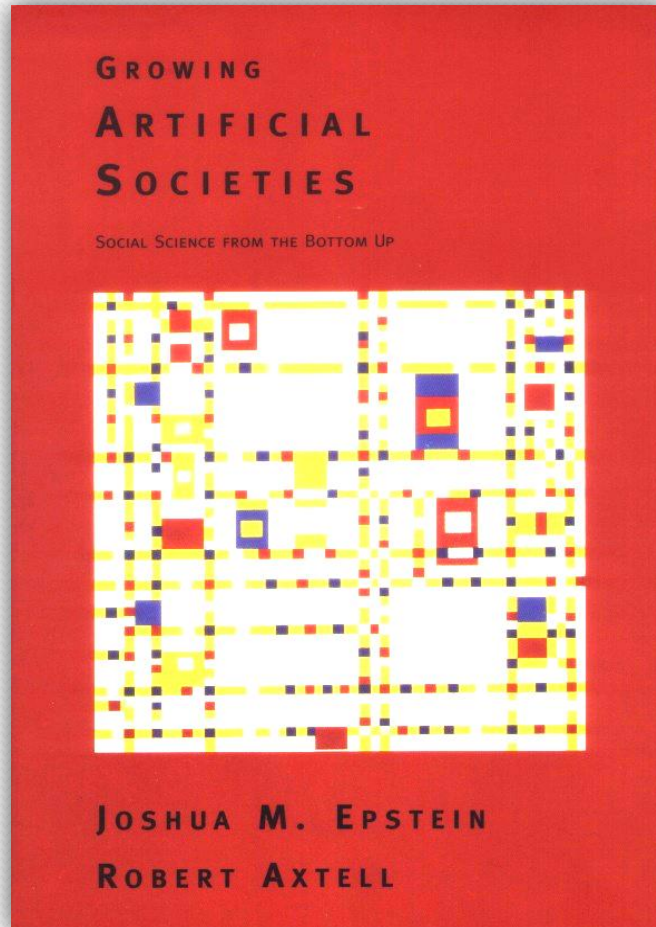
Problem type 3: Simulation

- We have a given model and wish to know the outputs that arise under different input conditions



- Often used to answer “what-if” questions in evolving dynamic environments
- e.g. Evolutionary economics, Artificial Life

Simulation example: evolving artificial societies



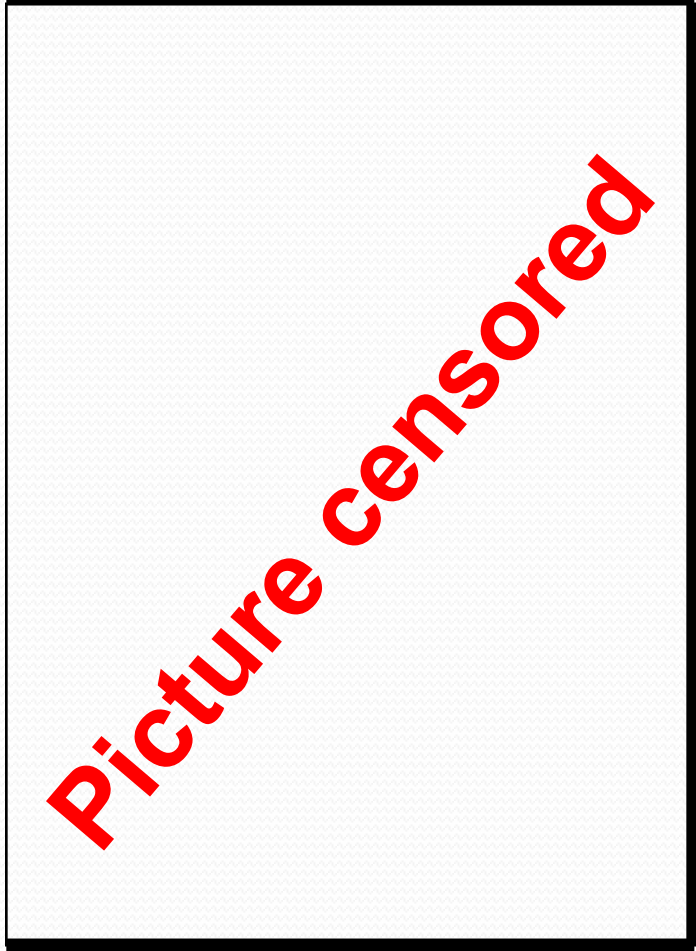
Simulating trade, economic competition, etc. to calibrate models

Use models to optimise strategies and policies

Evolutionary economy

Survival of the fittest is universal (big/small fish)

Simulation example 2: biological interpretations



Incest prevention keeps evolution from rapid degeneration
(we knew this)

Multi-parent reproduction, makes evolution more efficient
(this does not exist on Earth in carbon)

2nd sample of Life

Demonstration: magic square

- Given a 10x10 grid with a small 3x3 square in it
- Problem: arrange the numbers 1-100 on the grid such that
 - all horizontal, vertical, diagonal sums are equal (505)
 - a small 3x3 square forms a solution for 1-9

Demonstration: magic square

Evolutionary approach to solving this puzzle:

- Creating random begin arrangement
- Making N mutants of given arrangement
- Keeping the mutant (child) with the least error
- Stopping when error is zero

Demonstration: Magic Square

- Software by M. Herdy, TU Berlin
- Interesting parameters:
 - Step1: small mutation, slow & hits the optimum
 - Step10: large mutation, fast & misses (“jumps over” optimum)
 - Mstep: mutation step size modified on-line, fast & hits optimum
- Start: double-click on icon below
- Exit: click on TUBerlin logo (top-right)



Application