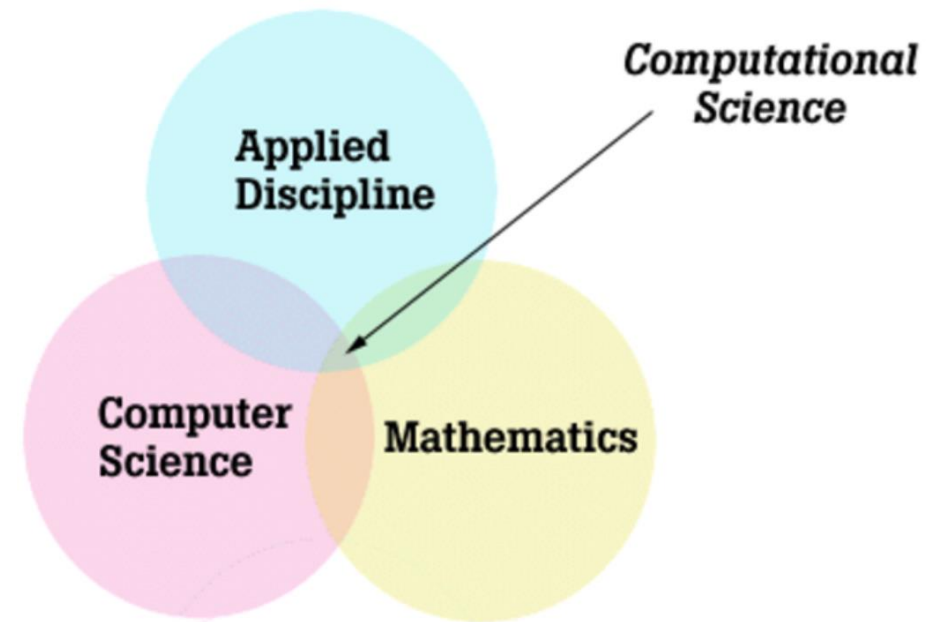
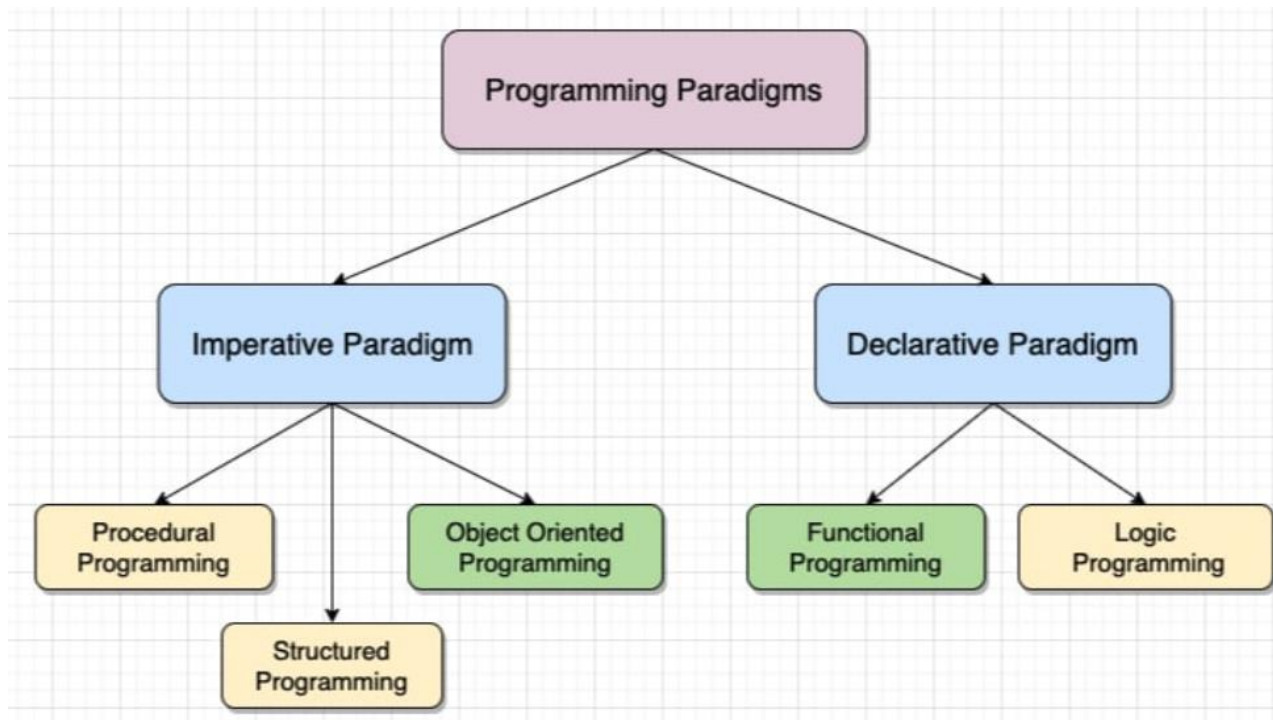


10. Programming paradigms & computational science



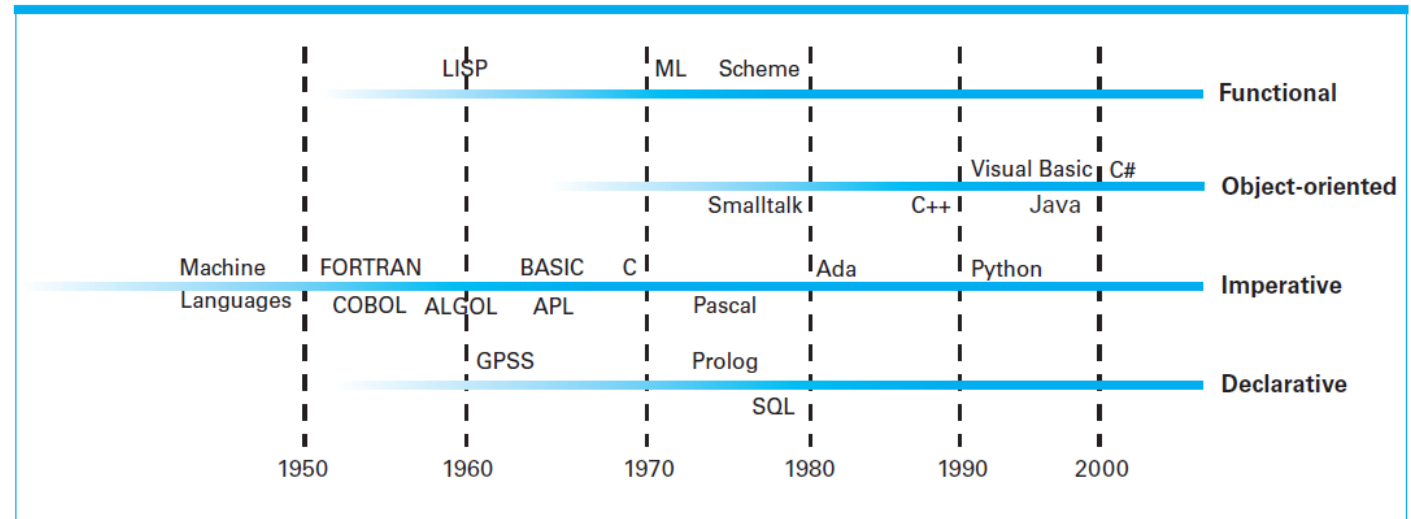
History of programming languages

- We've already discussed different programming languages, their evolution and translating from one language to another during the first half of this course
- Let's make a short revision:
 - Originally, computers were programmed using computer-specific (and essentially numeric) machine languages; these are commonly known as first-generation languages
 - Readability of code took a giant leap forward with the development of mnemonic assembly languages (which gave us the option to give variables and operations somewhat understandable names) – known as second-generation languages
 - The programs written in assembly language were still quite machine-dependent, because the primitives in them were equal to their machine language counterparts
 - Also, these primitives were very small as building blocks
 - Both these problems were solved by third generation languages, which were both machine independent as well as contained much larger primitives
 - Nowadays the generations span up to 5, but gen4 & gen5 languages are for specific purposes

Programming paradigms

- Generation-based approach is not the best for classification of languages
- Programming languages have developed along different paths as alternative approaches to the programming process – called *programming paradigms*
- These paradigms represent fundamentally different approaches on how to build solutions to presented problems
- Four main paradigms are
 - Imperative
 - Functional
 - Object-oriented
 - Declarative

Figure 6.2 The evolution of programming paradigms



Imperative paradigm

- Imperative paradigm (also known as the procedural paradigm; although some distinct procedural as a subset) is the traditional approach to programming
- Aim is to develop a sequence of imperative commands that (if followed) manipulates the data in such a way that the desired result is reached
 - Very engineer-ish “flowchart” method
- The structure of the program can be clarified by dividing it to subprograms called *procedures*
- Procedures can be executed sequentially (phases can be numbered) or concurrently
 - Concurrency requires additional control methods
- Examples: C, Ada, Pascal, FORTRAN, Basic, Cobol, ...

Downsides to imperative paradigm

- Imperative (and procedural) algorithms have previously dominated programming, because they are very deterministic in nature and correspond to von Neumann-architecture
 - Troubleshooting is “easy”, because the problematic procedure can be identified
- On the other hand, if there are many procedures, meticulous attention needs to be put on the execution order; it is far too easy to design an algorithm that has large “visual complexity” and hence is prone for mistakes
 - For example, if there are several IF clauses: which “ELSE” is dedicated to which IF?
- Procedural algorithms are not intuitive in all cases
- Proof of correctness is a work-heavy task

Functional paradigm

- In functional paradigm, a program is seen as an entity that accepts inputs and produces outputs directly from the inputs by using smaller entities
 - These entities are called functions (hence the name)
- Program is constructed by connecting smaller functions in such a manner that the outputs of the previous function act as inputs to the next one
 - The "main function" = the solution to the problem
- Information is presented as common data list structure
 - Data as well as the program are presented in same fashion
- Based on Church's lambda calculus
 - Result of an algorithm is defined as a mathematical function (arguments = inputs)
- Examples: Lisp, Scheme, Erlang, Haskell, ...

Example: imperative vs. functional paradigm

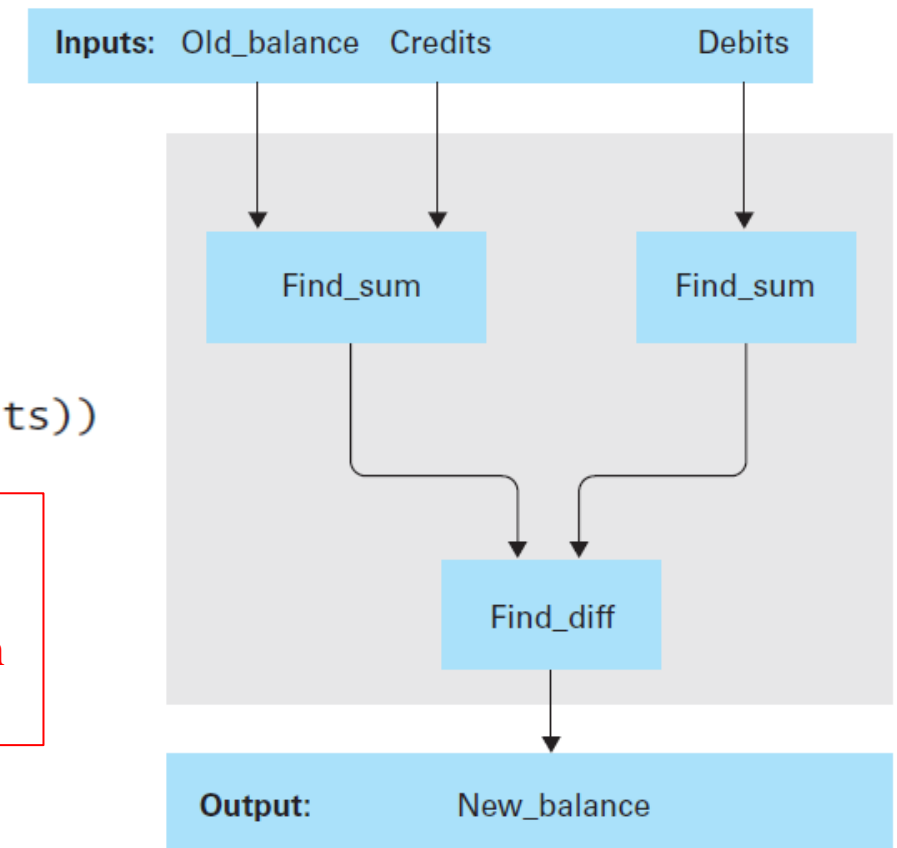
- Problem: find the balance of a checkbook
- Imperative:

```
Total_credits = sum of all Credits  
Temp_balance = Old_balance + Total_credits  
Total_debits = sum of all Debits  
Balance = Temp_balance - Total_debits
```

- Functional:

```
(Find_diff (Find_sum Old_balance Credits) (Find_sum Debits))
```

At first glance, these may seem quite identical.
But note: imperative version has several statements, and the mid-results of all such statements are stored! Functional version only has one, which results in better efficiency.



Object-oriented paradigm

- In object-oriented paradigm, a software system is viewed as a collection of *objects* which can perform actions on themselves and/or request actions from other objects
 - For example, in a GUI, all icons are objects
- Methods linked to the object describe how the object responds to different events
 - Left mouse click, right mouse click, double click, ...
- Advantage: if several programs use the same object, the functions needed are already provided with the object
- Description of the properties of an object is called a class
 - Creating multiple objects with same properties is easy (class can be applied to new objects)
- Foundation of most OOP languages is imperative, though - methods are small imperative program units
- Examples: C++, Java, Python

Declarative paradigm

- In declarative paradigm, programmer is asked to describe the problem to be solved
- Declarative algorithms are non-deterministic, so the execution order of phases of the algorithm is not typically known
- The system applies a general-purpose problem-solving algorithm
 - Such algorithms unfortunately don't exist for very many problems
 - Commonly some deterministic search method is applied (depth-first search or similar)
- Declarative languages are therefore most suitable for some special applications
 - Hypothesis testing & predictions
 - Parallel computing
- Subfield: logic programming (find out whether claim x is true or not)
- Examples: Prolog, SQL

Concurrent vs. parallel computing

- The traditional programming method is to do everything in clearly ordered phases – so, in sequential fashion
- This limits us from making use of parallel computing
- Concurrent computing can be applied to sequential programs
 - Concurrent = multiple actions can be performed by rapidly switching the process in execution in order to give the user a feeling of parallelism
- True parallelism is based on parallel use of our system:
 - Multi-core processors / multiprocessor computers
 - Computer networks, clusters, decentralized computation
 - Supercomputers

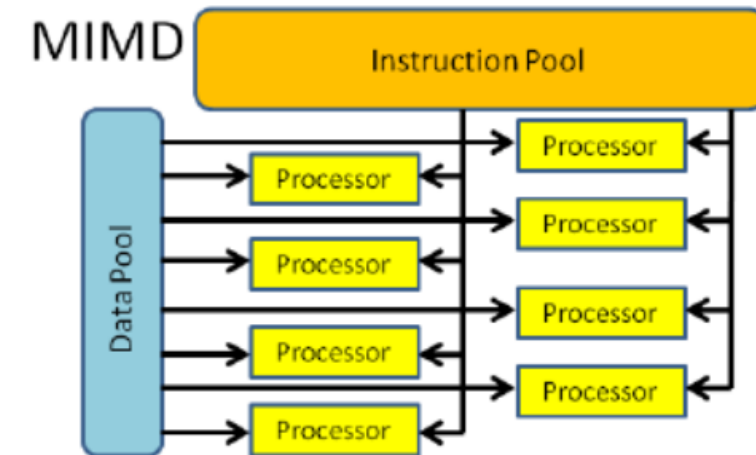
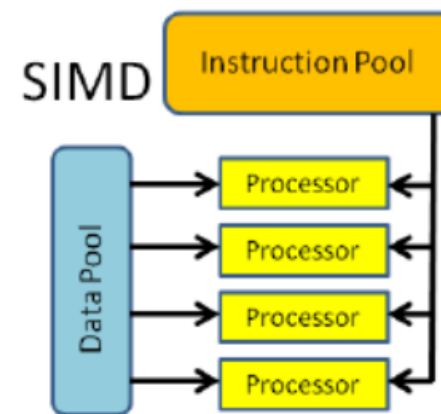
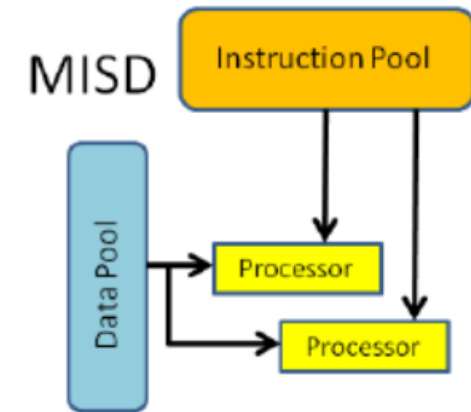
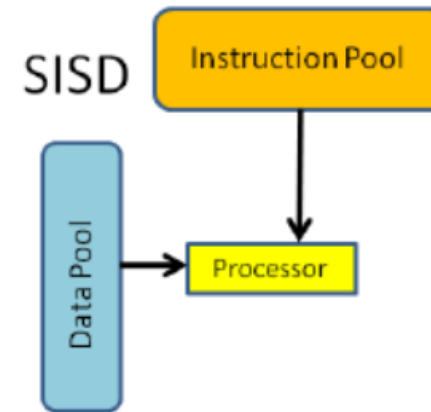
Parallel programming

- Not all problems are suitable for parallel computing
 - For example, problems where the previous phase produces the inputs of next phase
 - This is surprisingly common in practice
 - Parallel computing could be used in some phases, but not throughout the problem
- On the other hand, there are also problems for which parallelism feels natural
 - For example, face recognition: find a match from a database of 500 000 photos → if we have 10 computers available, the database can be divided so that each goes through 50 000 photos
- Designing a parallel algorithm may deviate a lot from a sequential algorithm
- There are programming languages which are especially suited for parallel programming (usually dependent on system or architecture)
- Parallelism in pseudolanguage: PARDO

```
FOR i := a,...,b PARDO  
    <body>  
END
```

Flynn's taxonomy

- Computer architectures can be divided in four categories in relation to their ways to handle instruction & data streams:
 - SISD (single instruction, single data stream; all old single-core processors)
 - MISD (multiple instruction, single data; used only for fault detection)
 - SIMD (single instruction, multiple data; basically all modern GPUs)
 - MIMD (multiple instruction, multiple data; all modern multi-core processors)
- This division is known as *Flynn's taxonomy*



Communication and speedup

- Especially in MIMD implementations, good communication between processors/cores is the key to performance improvements
 - Cores can share mid-results with each other etc.
- There are two ways to communicate and share information:
 - Shared memory (all cores can access the information)
 - Message passing (if memory is distributed; cores send “private messages” to each other)
- A benchmark quantity for measuring quality of parallelism is speedup
 - If the problem is divided to n cores/CPU's, how multiple will the performance rise?
 - Best-case: linear speedup = n (theoretical maximum)
- In special cases, a superlinear speedup is possible
 - For example, if breaking the problem to smaller parts results in all parts fitting in the cache (greatly reduced memory fetch times)

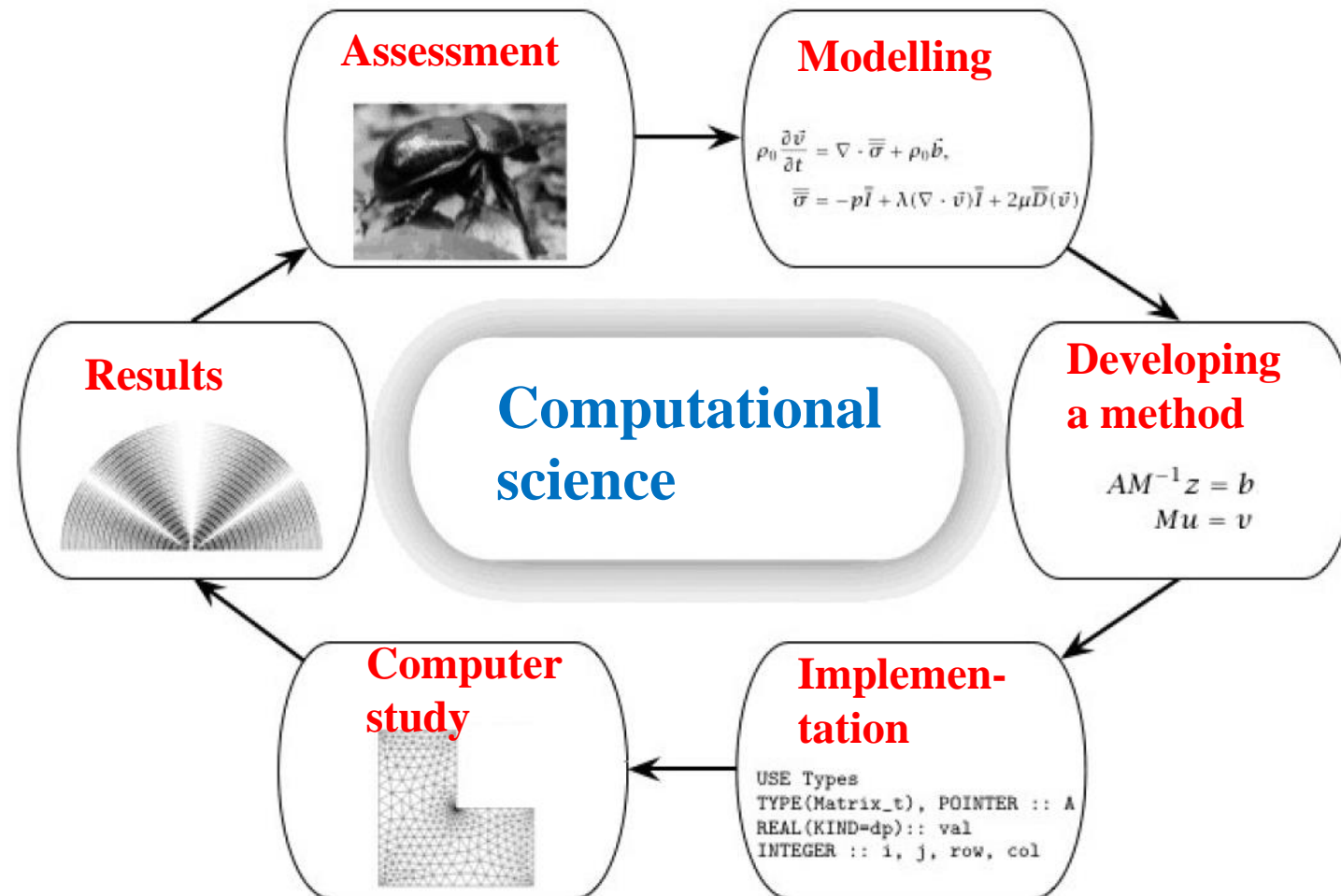
Computer science and mathematics

- Previously we've looked at things from the angle of a programmer
- We might remember from the first lecture that the first computer scientists were actually mathematicians; computer science and mathematics have a special bond
- There are problems in mathematics which are far too complex to solve in an analytical fashion
- The field that aims to solve these problems by taking advantage of modern computing capabilities is called *computational science*
- Computational science primarily deals with the most mathematical side:
 - Mathematical models
 - Algorithms
 - Optimization

Definition of computational science

- Computational science is a field mainly concentrated in three topics:
 - 1) Algorithms and programs, modelling & simulation knowhow
 - 2) Development of hardware, software, data communications technology and information processing components needed for solving challenging tasks
 - 3) Computing infrastructure that supports the problem-solving & development of related sciences
- Possibilities:
 - Create more realistic models for solving complex and wide study problems (climate change, garbage problem of seas, fusion energy, ...)
 - Decrease the need for expensive experiments & prototypes
 - Study the correlations and causations between phenomena
 - Gain better understanding of things in the “big picture”
 - Improve risk control and tolerance to uncertainty

Process cycle of computational science



Modelling with data

- What part of the data is essential considering the question?
- How correct is the data we get from the sensors?
 - Calibration errors, measurement errors
- What internal relations does the data have?
 - Which data is “raw” and measured, which has been calculated from raw data?
- How can the data be saved during the measurement, are there delays?
 - For example, digital image correlation (DIC) systems require a quick hard drive
- Accessibility and transferability of data
 - Cloud storage or something else? Transferring hundreds of GBs of data is hard – especially if security is an issue

Types of mathematical models

Model type

Examined
phenomenon

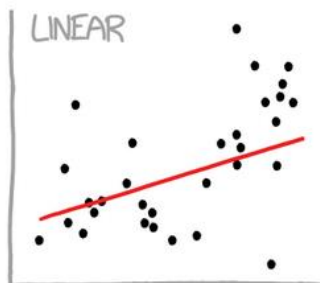


Linear \leftrightarrow Nonlinear
Deterministic \leftrightarrow Stochastic
Static \leftrightarrow Dynamic
Homogenous \leftrightarrow Heterogenous
Black box \leftrightarrow White box

—————→ Verification
—————→ Optimization
—————→ Adjustment
—————→ Decision-
making

What model fits the data?

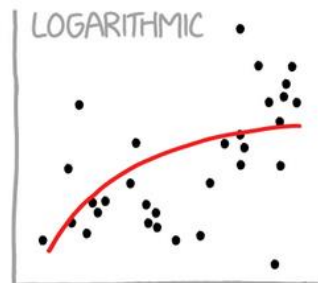
CURVE-FITTING METHODS AND THE MESSAGES THEY SEND



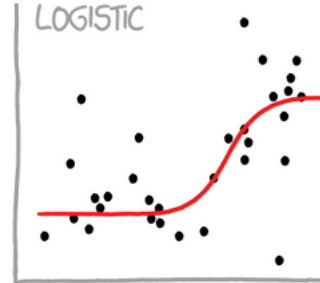
"HEY, I DID A REGRESSION."



"I WANTED A CURVED LINE, SO I MADE ONE WITH MATH."



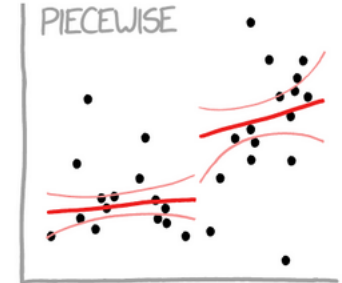
"LOOK, IT'S TAPERING OFF!"



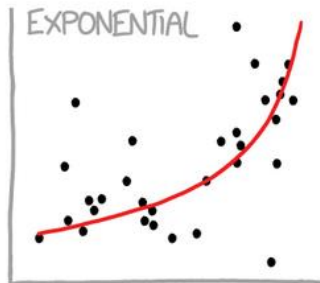
"I NEED TO CONNECT THESE TWO LINES, BUT MY FIRST IDEA DIDN'T HAVE ENOUGH MATH."



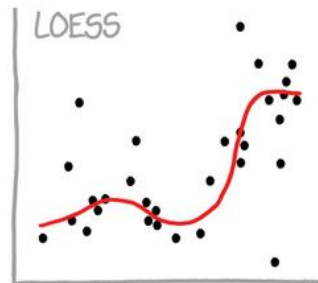
"LISTEN, SCIENCE IS HARD. BUT I'M A SERIOUS PERSON DOING MY BEST."



"I HAVE A THEORY, AND THIS IS THE ONLY DATA I COULD FIND."



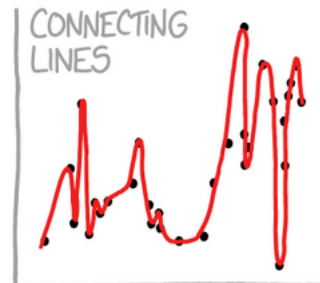
"LOOK, IT'S GROWING UNCONTROLLABLY!"



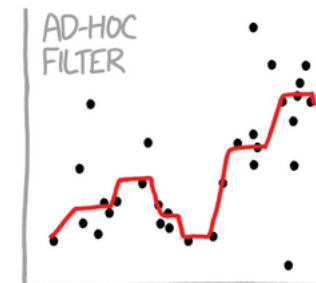
"I'M SOPHISTICATED, NOT LIKE THOSE BUMBLING POLYNOMIAL PEOPLE."



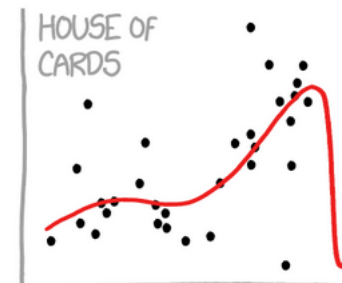
"I'M MAKING A SCATTER PLOT BUT I DON'T WANT TO."



"I CLICKED 'SMOOTH LINES' IN EXCEL."



"I HAD AN IDEA FOR HOW TO CLEAN UP THE DATA. WHAT DO YOU THINK?"

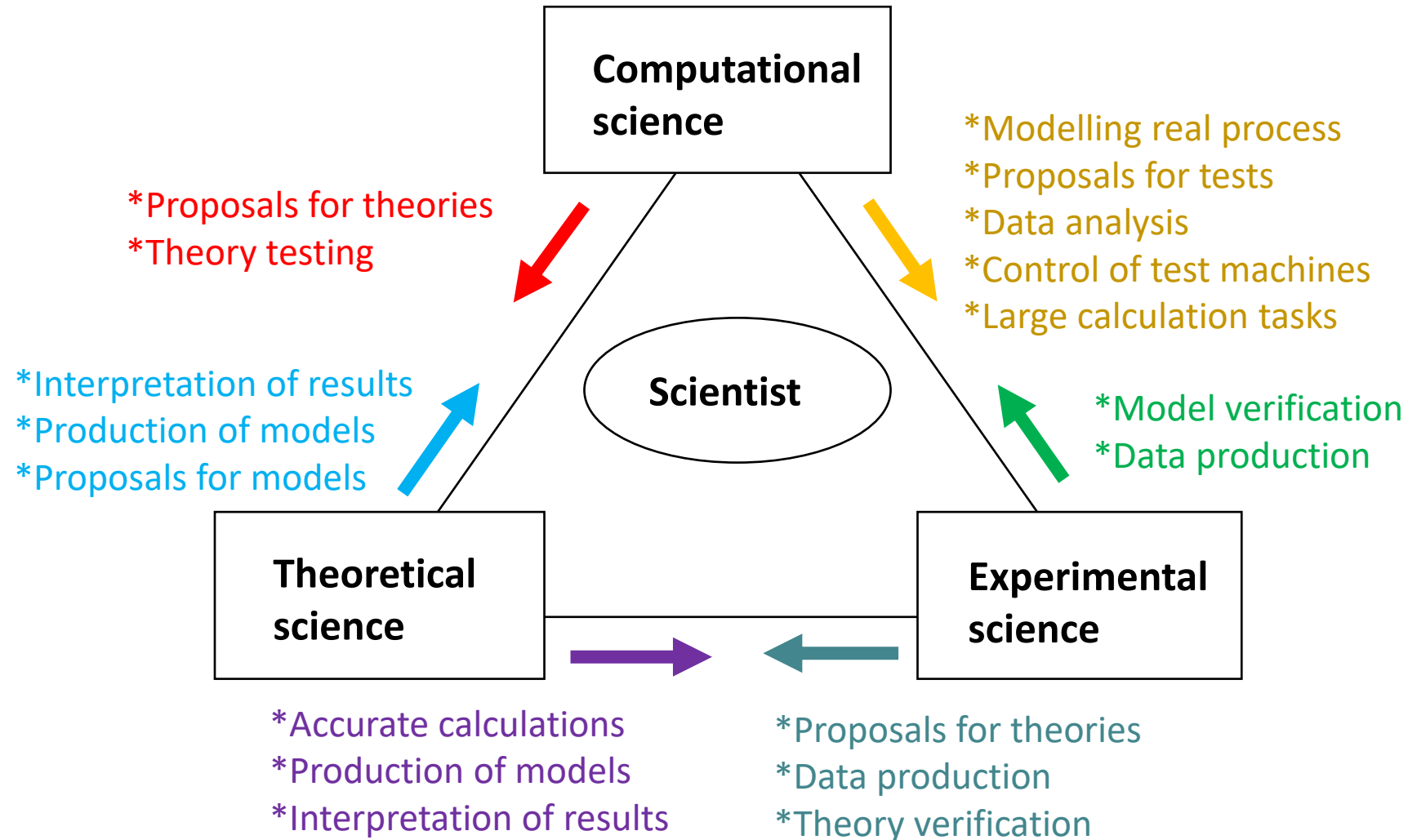


"AS YOU CAN SEE, THIS MODEL SMOOTHLY FITS THE— WAIT NO NO DON'T EXTEND IT AAAAAA!!!"

Optimization

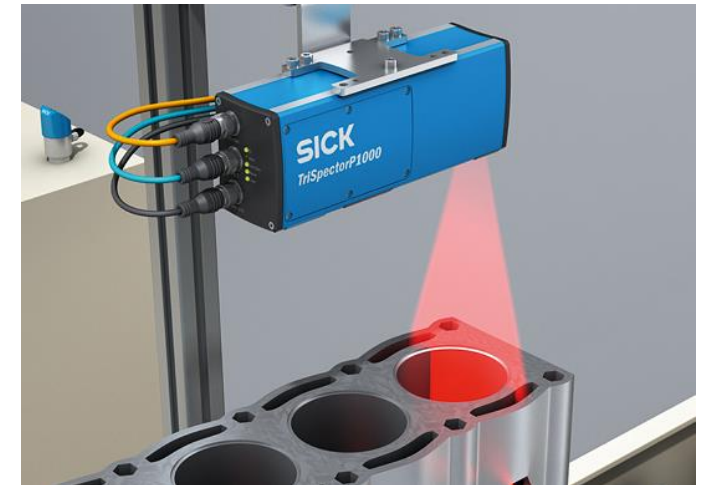
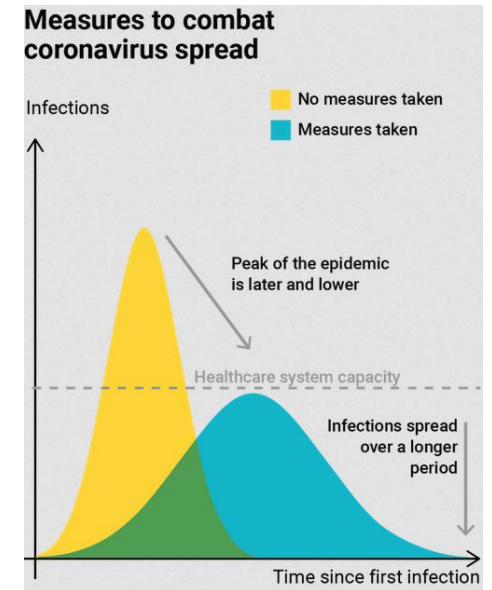
- After a model has been created, it can be *optimized*
- Computational science has given rise to many advanced optimization techniques
- One good example are genetic algorithms, which mimic evolution:
 - 1) Initiation: create a random population of solutions
 - 2) Selection: select the best solutions using a fitness function and delete the worst ones. If a good enough solution has been found (or time has run out), stop and go to 5.
 - 3) Combination: perform crossovers and mutations to the remaining population in order to create a new population – the next evolution version
 - 4) Repetition: go back to stage 2 with the new population
 - 5) Post-processing & visualization of results, termination
- Especially good algorithms when problem is supposed to have multiple local minima where derivative-based methods may get stuck

Role of computational science



Application areas of computational science

- Modelling fusion reactor behavior
 - Global epidemic models (ebola, covid-19)
 - Streaming services (recommendation algorithms, user demand)
 - Measurement & instrumentation technology
 - Remote mapping of natural resources (laser scanning of forests etc.)
 - Process diagnostics in factories
 - Raw material analysis (optimal sawing of a log)
 - Dynamic traffic steering
 - Machine vision & pattern recognition applications
- ...and many, many more!



Thank you for listening!

