

Lightweight code verification for science

Dominic Orchard

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Matthew Danish



Andrew Rice



Ben Orchard

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A computational science agenda for programming language research

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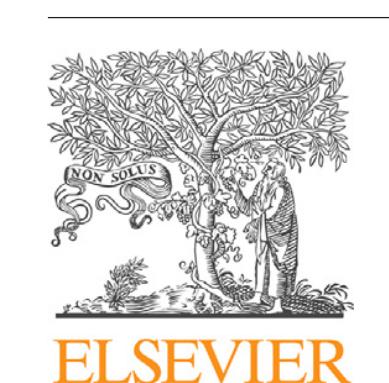
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Abstract

Scientific models are often expressed as large and complicated programs. These programs embody numerous assumptions made by the developer (*e.g.*, for differential equations, the discretization strategy and resolution). The complexity and pervasiveness of these assumptions means that often the only true description of the model is the software itself. This has led various researchers to call for scientists to publish their source code along with their papers. We argue that this is unlikely to be beneficial since it is almost impossible to separate implementation assumptions from the original scientific intent. Instead we advocate higher-level abstractions in programming languages, coupled with lightweight verification techniques such as specification and type systems. In this position paper, we suggest several novel techniques and outline an evolutionary approach to applying these to existing and future models. One-dimensional heat flow is used as an example throughout.

Keywords: computational science, modelling, programming, verification, reproducibility, abstractions, type systems, language design



Evolving Fortran types with inferred units-of-measure

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ABSTRACT

Dimensional analysis is a well known technique for checking the consistency of physical quantities, constituting a kind of type system. Various type systems, and its refinement to units-of-measure, have been proposed. In this paper, we implement a units-of-measure system for Fortran, provided as a plugin designed to aid adding units to existing code base: units may be polymorphic. Furthermore, we introduce a technique for reporting to the user a set of critical errors explicitly annotated with units to get the maximum amount of unit information from a small number of explicit declarations. This aids adoption of our type system to existing projects. There are many in computational science projects.

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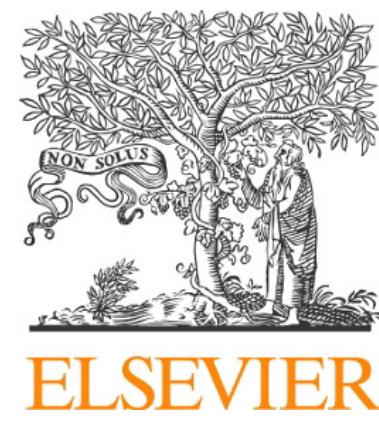
1. Introduction

Type systems are one of the most popular static techniques for recognizing and rejecting large classes of programming error. A common analogy for types is of physical quantities (*e.g.*, in [2]), where type checking excludes, for example, the non-sensical addition of non-comparable quantities such as adding 3 m to 2 J; they have different dimensions (length vs. energy) and different units (metres vs. joules). This analogy between types and dimensions/units goes deeper. The approach of *dimensional analysis* checks the consistency of formulae involving physical quantities, acting as a kind of type system (performed by hand, long before computers). Various automatic type-system-like approaches have been proposed for including dimensional analysis in programming languages (*e.g.*, [10] is a famous paper detailing one such approach,

circumstances. It therefore seems inevitable that units-of-measure will be used in computational science too.

The importance of units is often directly apparent when reading source code. We have seen source files careful to include units and dimensions of each variable and function argument, with programmers trying to use this information to quickly scroll up and down, repeatedly referring to the declaration of each parameter. Incorporating units-of-measure into the compiler would move the onus of responsibility for units-of-measure from the programmer to the compiler.

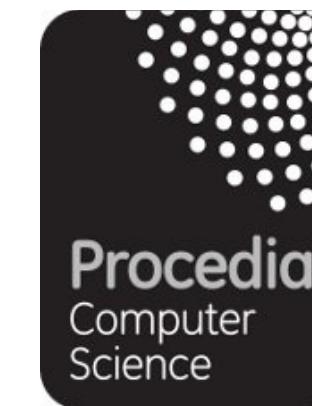
A recent ISO standards proposal (N1963) [1] proposes a units-of-measure system which follows the principle of dimensional analysis and its associated explicitness [7]. Every variable declaration must include a unit declaration and every composite unit (e.g., metres per second or seconds per meter) must itself be explicitly declared.



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A computational science agenda for programming language research

Dominic Orchard¹, Andrew Rice²

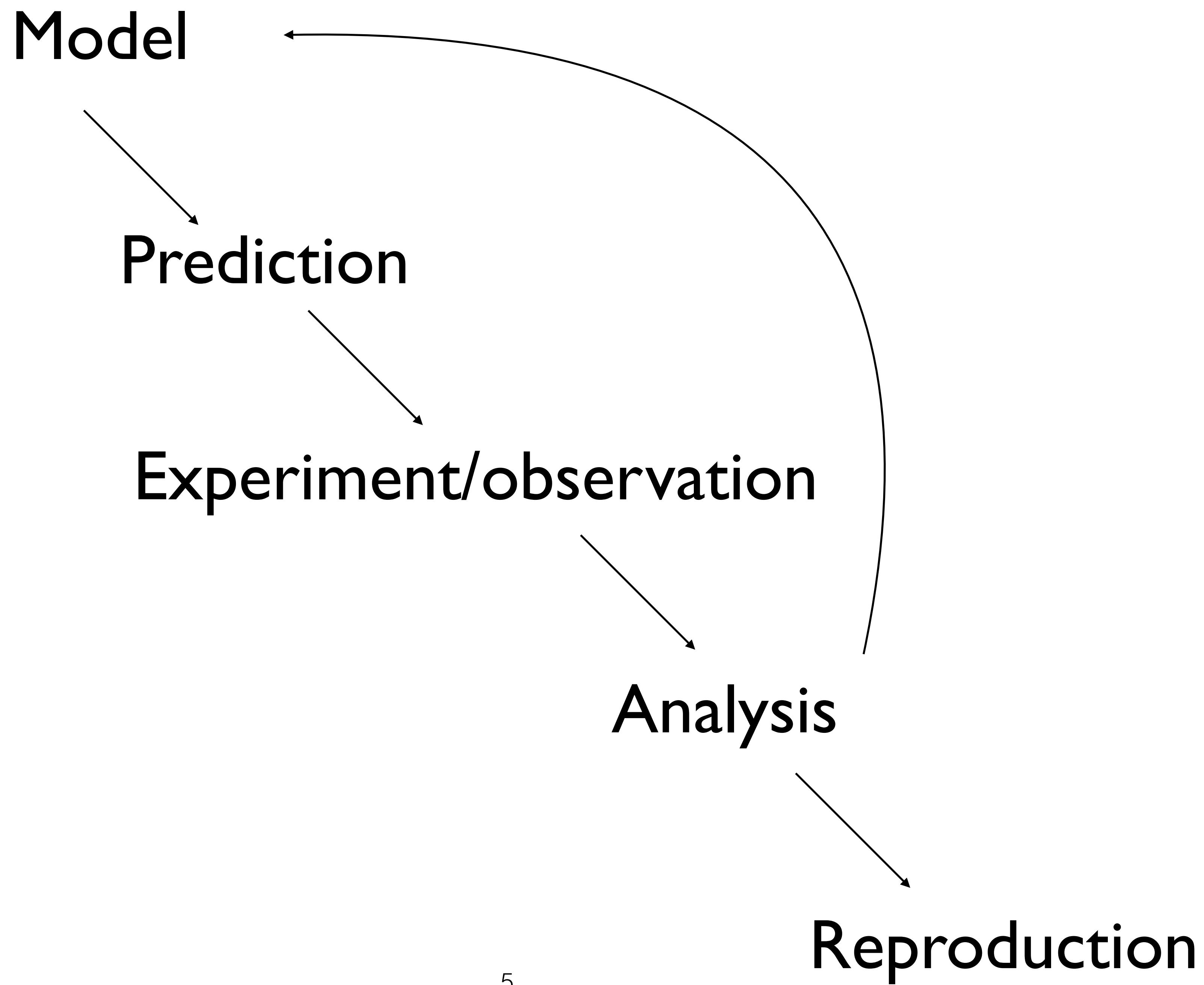
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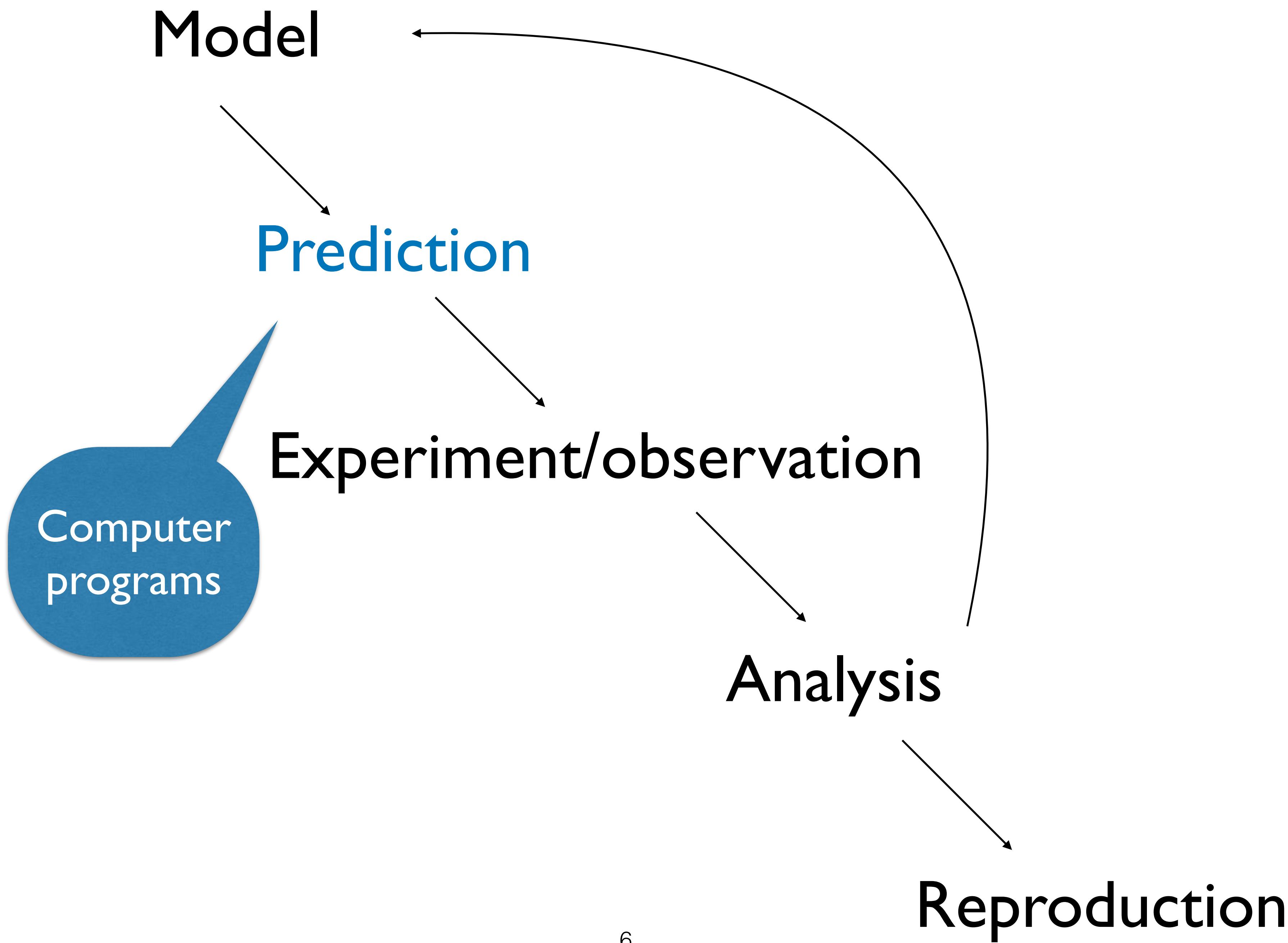
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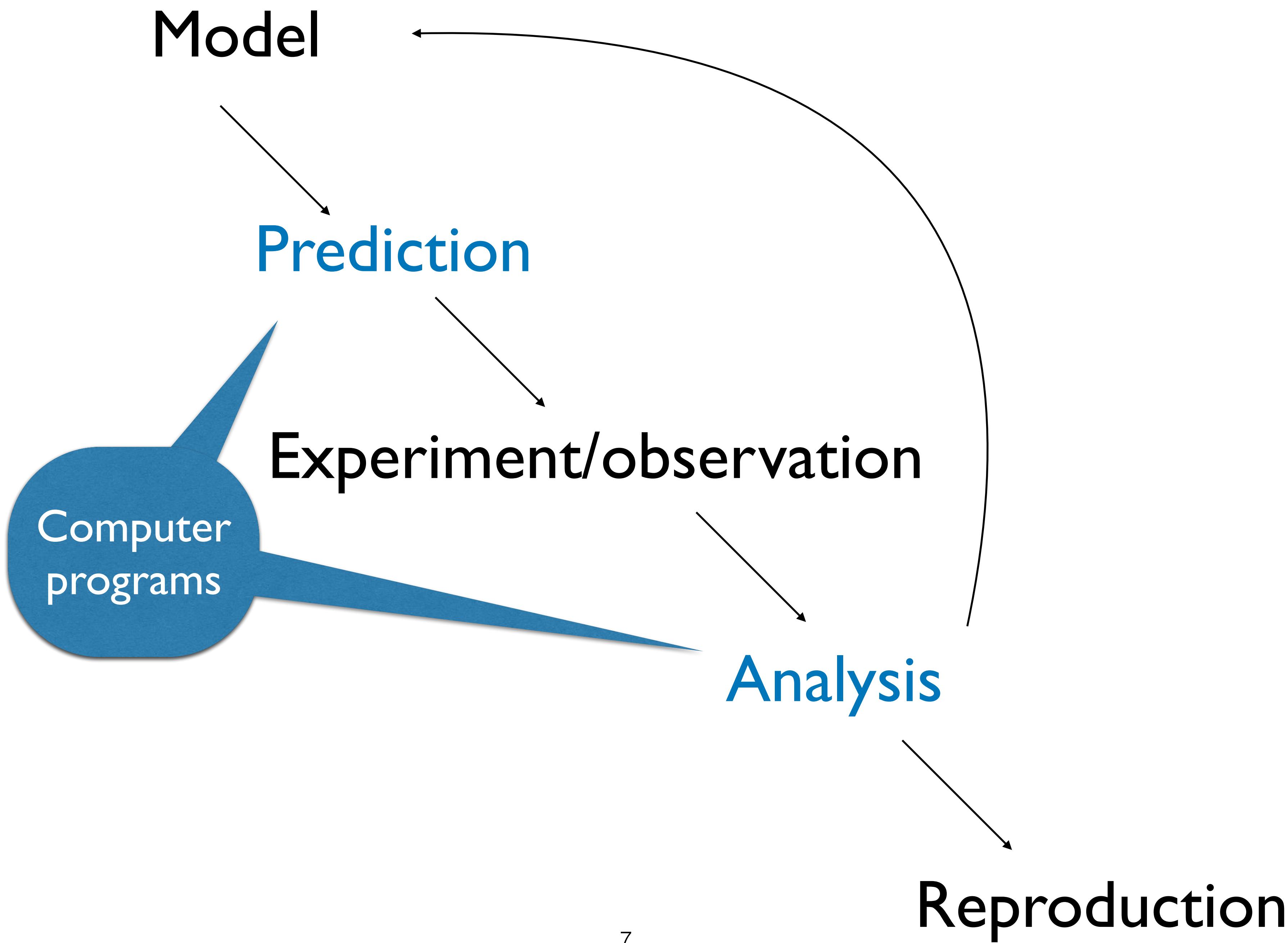
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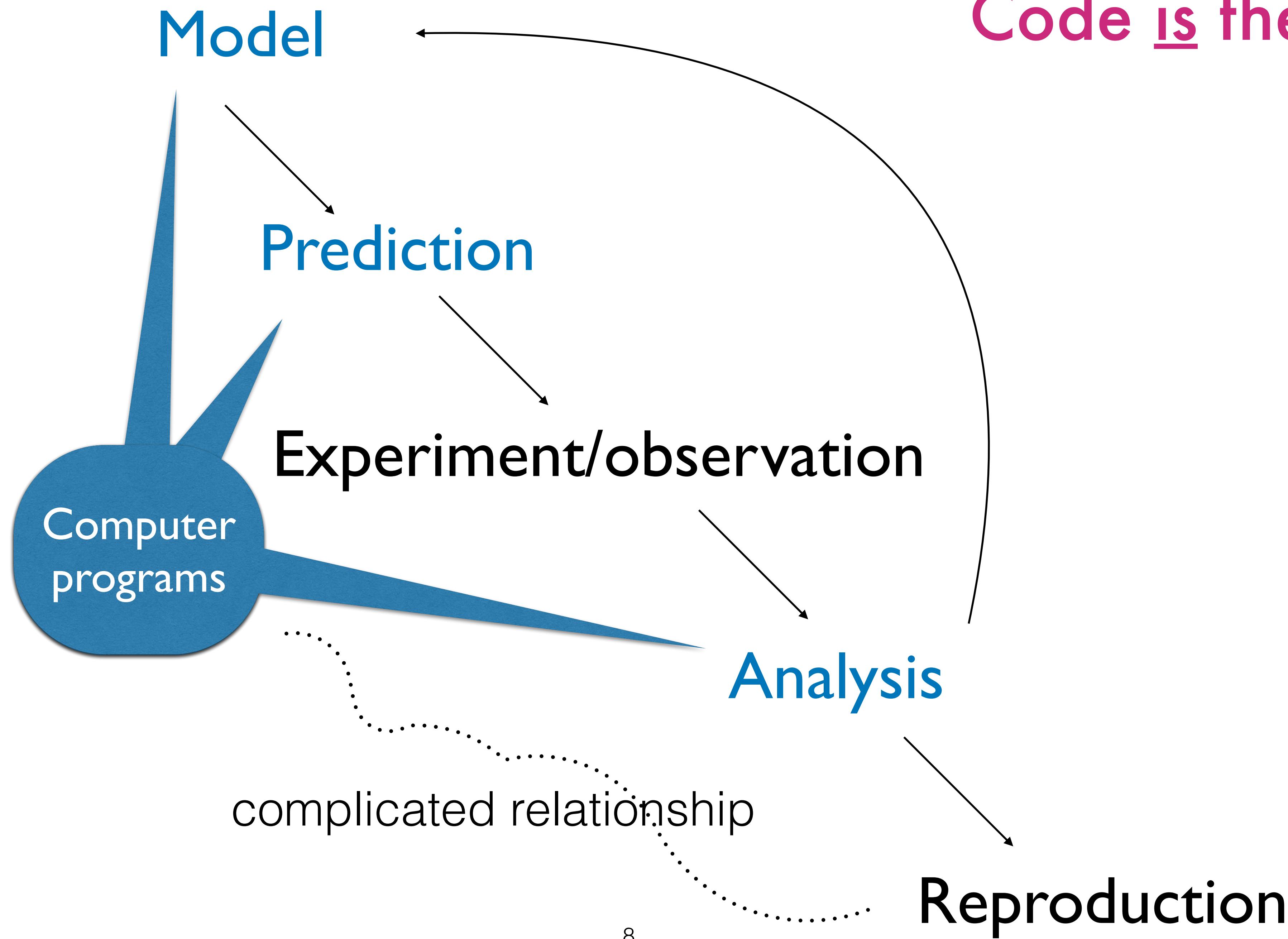
Keywords: computational science, modelling, programming, verification, reproducibility, abstractions, type systems, language design







Code is the model



Example 1D heat equation

Abstract model

$$\frac{\partial \phi}{\partial t} = \alpha \frac{\partial^2 \phi}{\partial x^2}$$

Solution strategy

$$\phi_x^t = \phi_x^{t-1} + \frac{\alpha \Delta t}{\Delta x^2} (\phi_{x+1}^{t-1} + 2\phi_x^{t-1} + \phi_{x-1}^{t-1})$$

Example 1D heat equation

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Prediction calculation

```
1 tend = ... % end time
2 xmax = ... % length of material
3 dt = ... % time resolution
4 dx = ... % space resolution
5 alpha = ... % diffusion coefficient
6 nt = tend/dt % # of time steps
7 nx = xmax/dx % # of space steps
8 r = alpha*dt/dx^2 % constant in solution
9
10 real h(0,nx), % heat fun. (discretised
11     h_old(0, nx); % in space) at t and t-1
12
13 do t = 0, nt
14     h_old = h
15     do x = 1, nx - 1
16         h(i) = h_old(i) + r*(h_old(i-1))
17             - 2*h_old(i) + h_old(i+1)
18     end do
19 end do
```

Gap in explanation....

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APPLICATION PAPER  

A sensitivity analysis of a regression model of ocean temperature

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Keywords: Data science, interpretable ML, model sensitivity, oceanography, regression model

Abstract
There has been much recent interest in developing data-driven models for weather and climate predictions. However, there are open questions regarding their generalizability and robustness, highlighting a need to better understand how they make their predictions. In particular, it is important to understand whether data-driven models learn the underlying physics of the system against which they are trained, or simply identify statistical patterns without any clear link to the underlying physics. In this paper, we describe a sensitivity analysis of a regression-based model of ocean temperature, trained against simulations from a 3D ocean model, setup in a very simple configuration. We show that the regressor heavily bases its forecasts on, and is dependent on, variables known to be key to the physics such as currents and density. By contrast, the regressor does not make heavy use of inputs such as location, which have limited direct physical impact. The model requires nonlinear interactions between inputs in order to show any meaningful skill—in line with the highly nonlinear dynamics of the ocean. Further analysis interprets the ways certain variables are used by the regression model. We see that information about the vertical profile of the water column reduces errors in regions of convective activity, and information about the currents reduces errors in regions dominated by advective processes. Our results demonstrate that even a simple regression model is capable of learning much of the physics of the system being modeled. We expect that a similar sensitivity analysis could be usefully applied to more complex ocean configurations.

Impact Statement
Machine learning provides a promising tool for weather and climate forecasting. However, for data-driven forecast models to eventually be used in operational settings we need to not just be assured of their ability to perform well, but also to understand the ways in which these models are working, to build trust in these systems. We use a variety of model interpretation techniques to investigate how a simple regression model makes its predictions. We find that the model studied here, behaves in agreement with the known physics of the system. This work shows that data-driven models are capable of learning meaningful physics-based

```
1 module simulation_mod
2   use helpers_mod
3   implicit none
4
5   contains
6
7   subroutine compute_tentative_velocity(u, v, f, g, flag, del_t)
8     real u(0:imax-1, 0:jmax+1), v(0:imax+1, 0:jmax+1), f(0:imax+1, 0:jmax+1), &
9       g(0:imax-1, 0:jmax+1)
10    integer flag(0:imax+1, 0:jmax+1)
11    real, intent(in) :: del_t
12
13    integer i, j
14    real du2dx, duvdy, duvdx, dv2dy, laplu, laplv
15
16    do i = 1, imax-1
17      do j = 1, jmax
18        ! Only if both adjacent cells are fluid cells */
19        if (topological(iand(flag(i,j), C_F)) .and. &
20            topological(iand(flag(i+1,j), C_F))) then &
21
22          du2dx = ((u(i,j)+u(i+1,j))+u(i,j)+u(i+1,j))-&
23            gamma*abs(u(i,j)-u(i+1,j))*u(i,j)-u(i+1,j))-&
24            (u(i-1,j)+u(i,j))+u(i-1,j)+u(i,j))-&
25            gamma*abs(u(i-1,j)+u(i,j))*u(i-1,j)-u(i,j))/&
26            (4.0*delx)
27          duvdy = ((v(i,j)+v(i+1,j))+u(i,j)+u(i,j+1))+&
28            gamma*abs(v(i,j)-v(i+1,j))*u(i,j)-u(i,j+1))-&
29            (v(i,j-1)+v(i-1,j))+u(i,j-1)+u(i,j))-&
30            gamma*abs(v(i,j-1)+v(i+1,j-1))*u(i,j-1)-u(i,j))/&
31            (4.0*delx)
32          laplu = (u(i+1,j)-2.0*u(i,j)+u(i-1,j))/delx/delx+&
33            (u(i,j+1)-2.0*u(i,j)+u(i,j-1))/dely/dely
34
35          f(i,j) = u(i,j) + del_t*(laplu/Re-du2dx-duvdy)
36        else
37          f(i,j) = u(i,j)
38        end if
39      end do
40    end do
41
```



Abstract model

Solution strategy

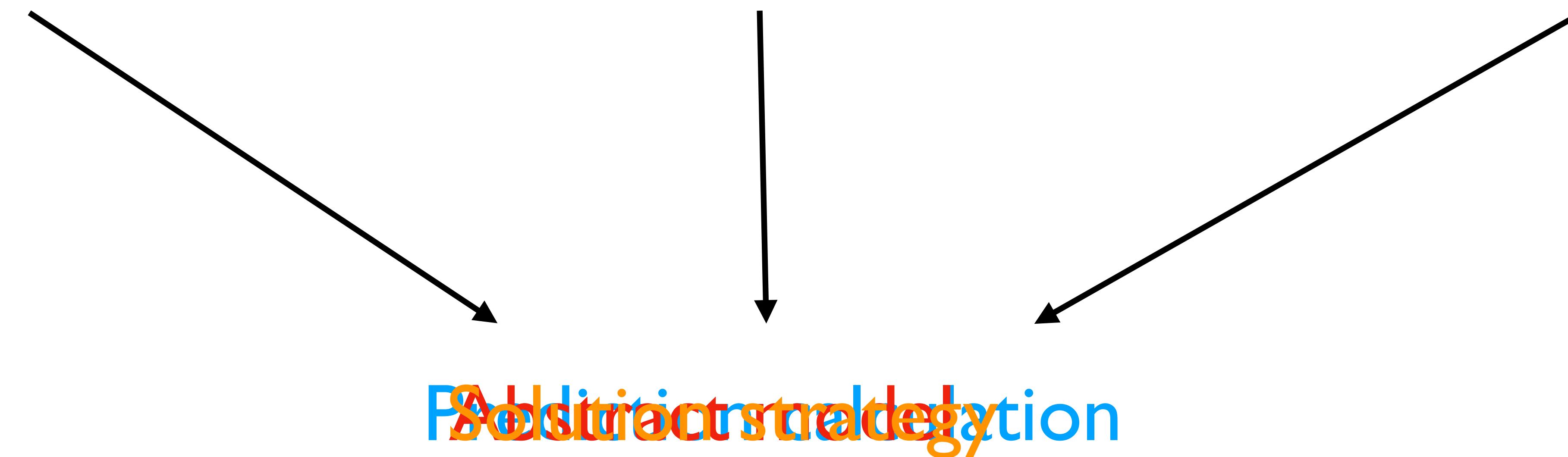
Prediction calculation

Conflation of concerns

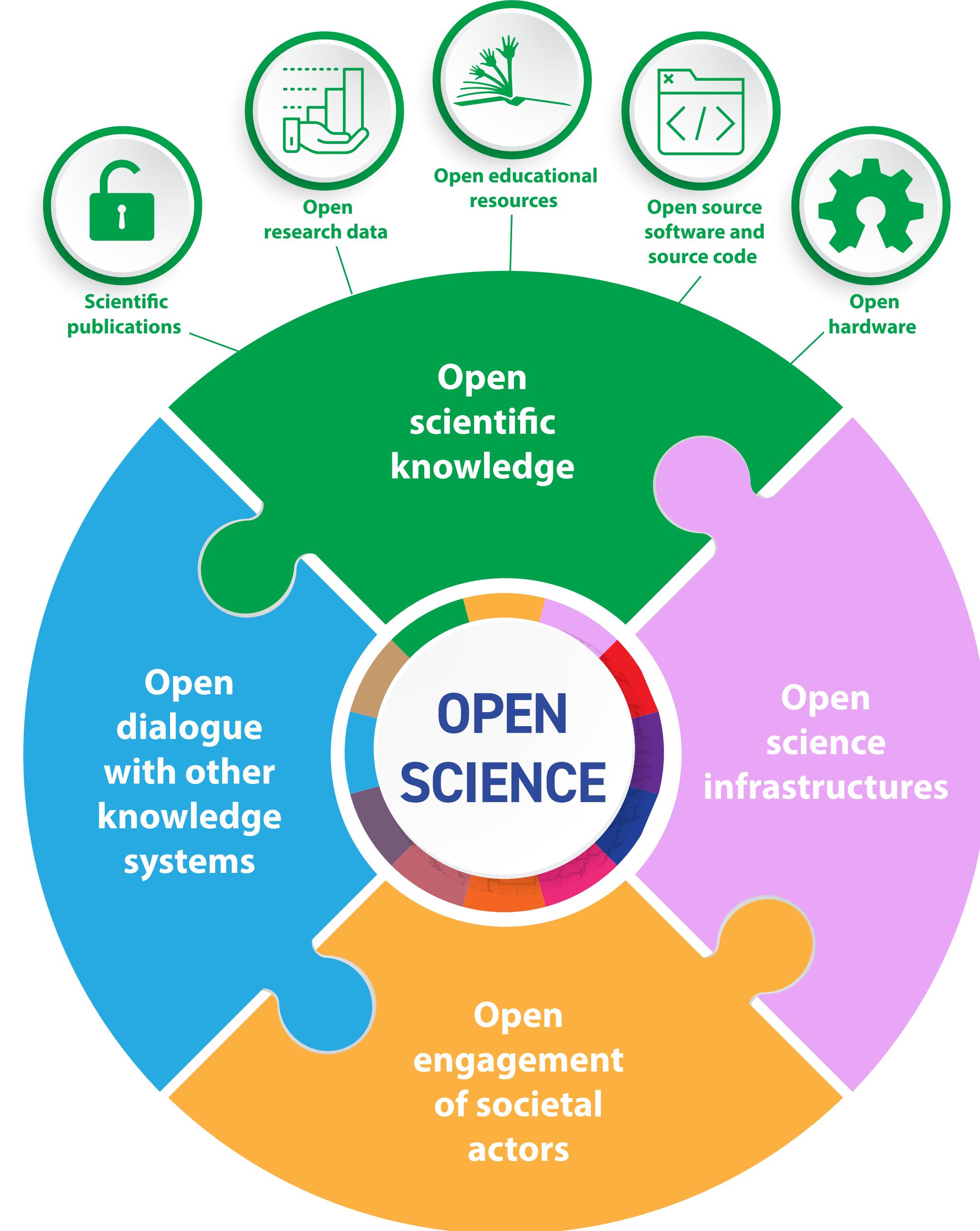
Abstract model

Solution strategy

Prediction calculation



Code conflates & hides many aspects of the model



Open scientific

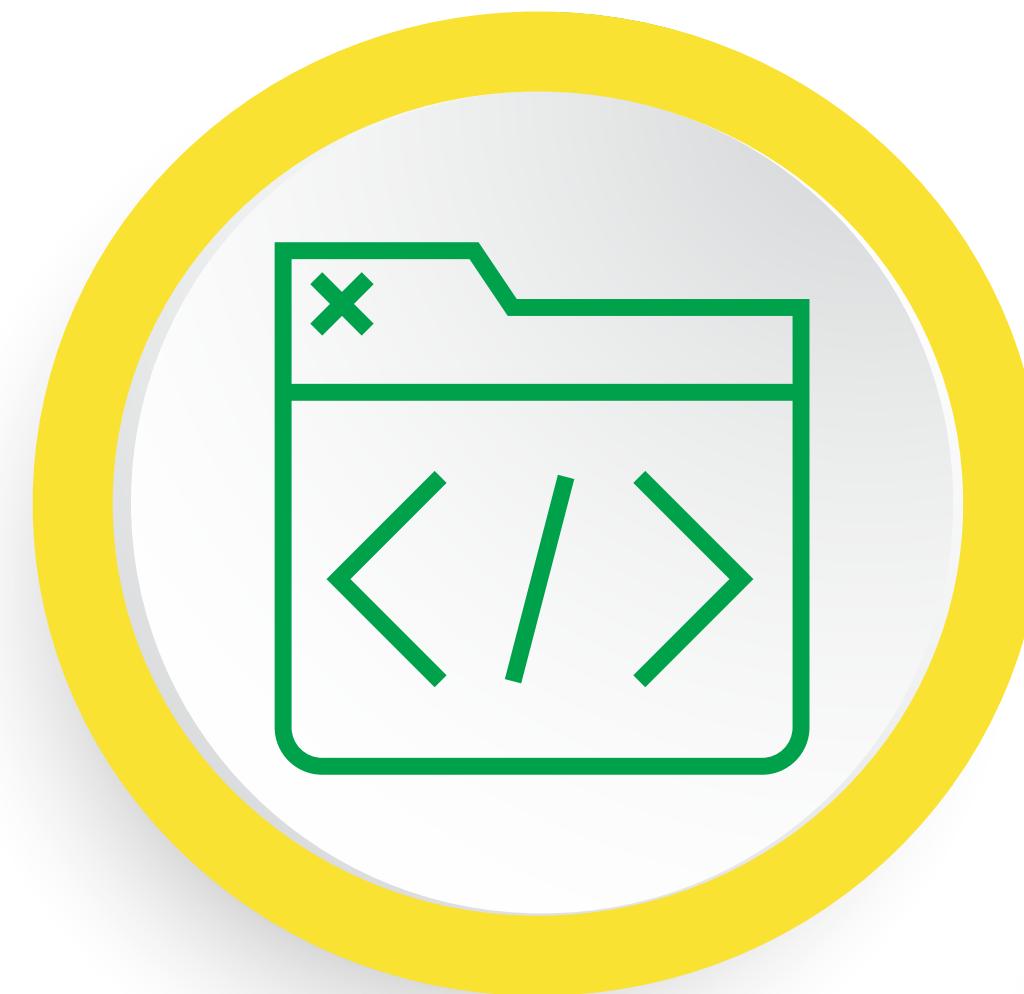
Open
research data



Open educational
resources



Open source
software and
source code



Open
hardware



But.. sharing code includes sharing **bugs**



- + assumptions
- + incidental decisions
- + approximations

Open problem: separating and relating concerns



Abstract model

Solution strategy

Prediction calculation

Partial solutions

- ▶ Extra technical documentation
- ▶ Clear systems design
- ▶ High modularity

Could there be better support via a **programming language tailored to science?**

**BETTER
LANGUAGES**

**BETTER
SOFTWARE**

**BETTER
RESEARCH**

The four Rs of programming language design... (Orchard, 2011)

Reading
wRiting
Reasoning
Running

The *Four Rs* of Programming Language Design

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Categories and Subject Descriptors D.1.0 [Software]: Programming Techniques—General; I.0 [Computing Methodologies]: GENERAL

General Terms Design, Languages

Keywords Programming language design, The Four Rs, Domain-specific languages

"I can learn the poor things *reading*, *writing*, and '*rithmetic*, and counting as far as the rule of three, which is just as much as the likes of them require;" *Lawrie Todd: Or the Settlers in the Woods*, Galt (1832) [4].

Many will be familiar with the old adage that at the core of any child's education should be the *three Rs*: *reading*, *writing*, and '*rithmetic'. The phrase, which appeared first in print in 1825 [12] has been appropriated and parodied at length ("*read*, *reason*, *recite*", "*reduce*, *reuse*, *recycle*", etc.). Each permutation has the same purpose: to express succinctly the core tenets of an approach or philosophy.*

The *four Rs of programming language design* is another such parody of this old phrase, providing a rubric, or framework, for the design and evaluation of effective programming languages and language features.

Since the very first programming language back in the 1940s [14] *thousands* of programming languages have been developed, representing a broad spectrum of paradigms, perspectives, and philosophies. And yet, there is no single language which is "*all things to all men*" (and women!).

The *four Rs* were born out of trying to answer a number of questions about the nature of programming languages and programming language design: what makes a programming language effective or ineffective? What should be the core aims of a language designer? How should programming languages and features be compared? Why is there no single "perfect" language? The *four Rs* go some way towards answering these questions.

Before I reveal the *four Rs*, let's first consider some more foundational questions:

Why programming languages? The development of programming languages has greatly aided software engineering. As hard-

languages have developed to manage this complexity more effectively, aiding us in expressing ideas and solving increasingly complex problems.

Programming languages provide *abstraction*, by both hiding details and allowing components to be reused, allowing programmers to more effectively manage complexity in software and hardware. While it is in principle possible for any program to be written in machine code, it's hard to imagine some of the larger computer programs we interact with daily being developed in such a way. By building layers of abstraction with languages, increasingly complex systems can be constructed.

What is programming? In essence, programming is a communication process between one or more programmers and one or more computer systems. Programming languages are the medium of this communication.

Programming is not only a communication process, it is also a *translation process*. Each participant in the programming process has an internal language, both programmers and machines. In the case of a machine, the internal language comprises the instructions of the underlying hardware. In the case of a programmer, the internal language is far more nebulous, perhaps comprising natural and formal languages, along with other incorporeal, abstract thoughts.

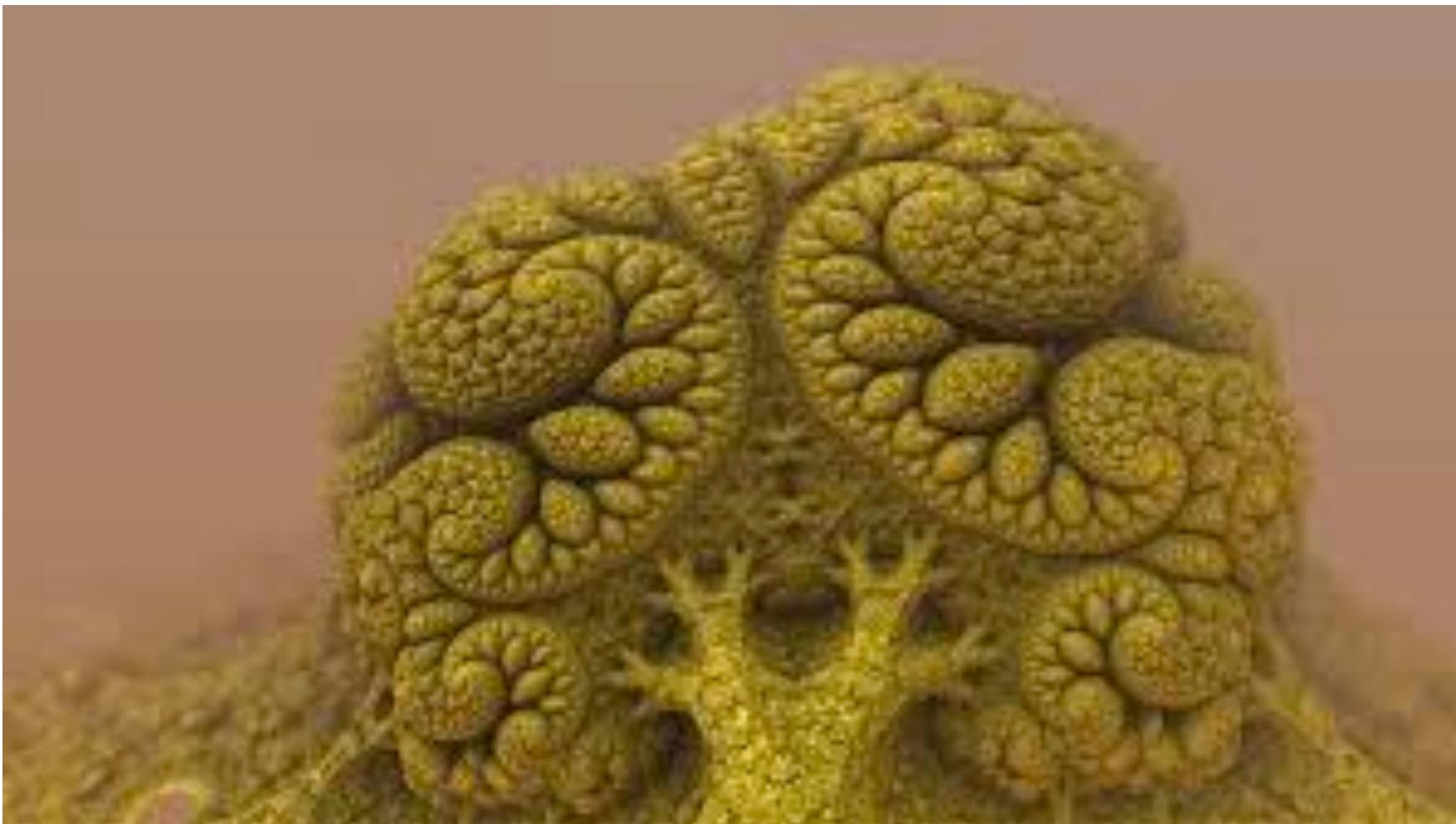
In any case, a programming language acts as the intermediate language of translation between the participants. *Programming* is the translation from a programmer's internal language to a programming language, and *execution* is the translation from the programming language to the machine's internal language. McCracken, in 1957, captured some of this sentiment, saying "Programming [...] is basically a process of translating from the language convenient to human beings to the language convenient to the computer" where the convenient language for humans was "mathematics or English statements of decisions to be made" [8]. Here we consider the "language convenient to human beings" to be programming languages, bridging the gap between our ideas and the underlying, low-level instructions of a computer system.

Sometimes, programming is more *exploration* than communication. In which case, a programmer explores and learns about a problem by translating their internal thoughts into a program and the re-internalising the result to gain further insight. Again the process is a *translational*.

It is from this view of programming, as a translation, communication, and exploration, that the *four Rs* are created.

The Two Complexities

Inherent



Accidental



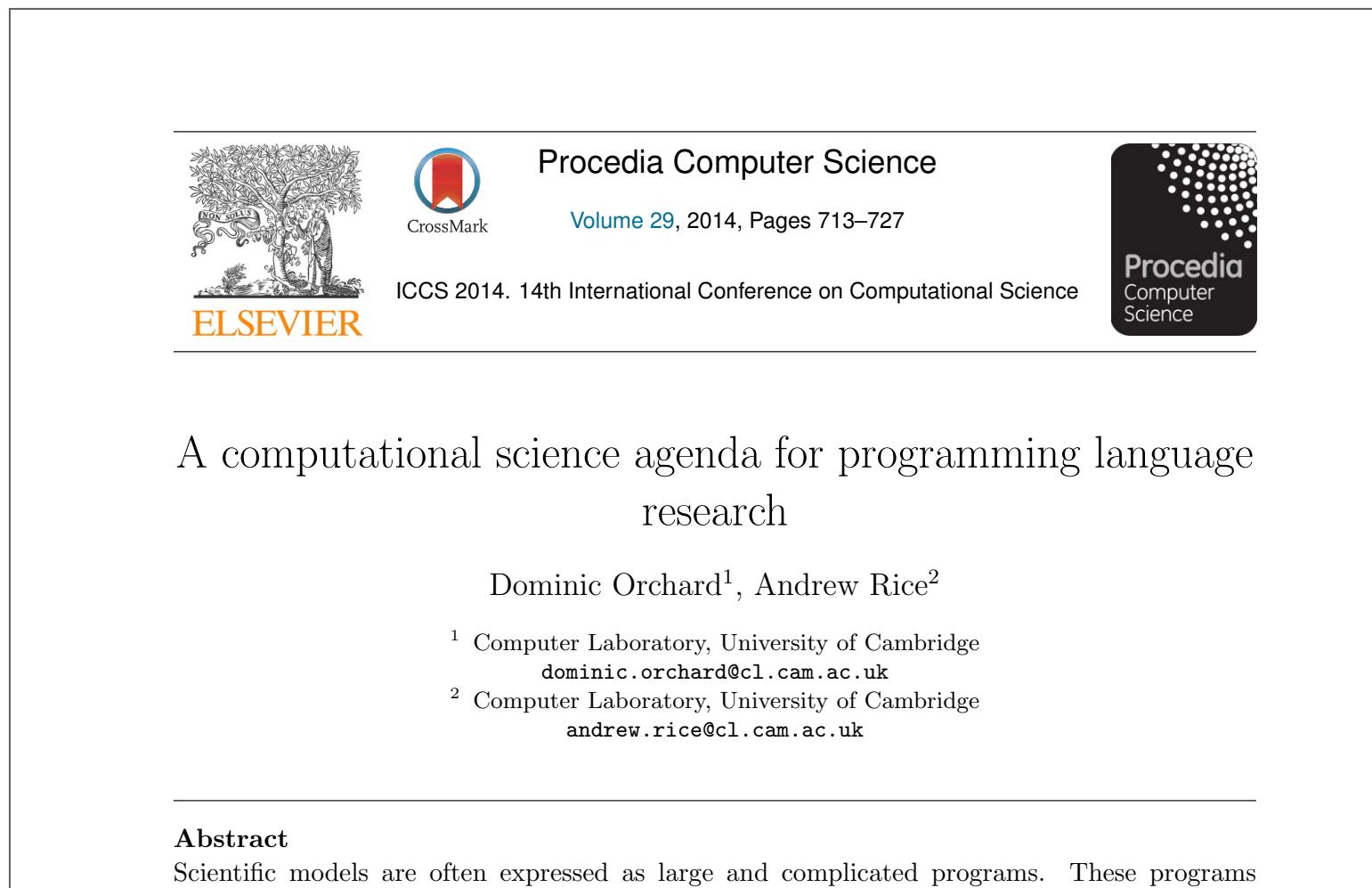
Inadequately supported

Too easy to introduce

Both hinder scientific progress, only one is necessary

Roadmap

1. Computer science engagement with scientists
2. New systems for abstraction and specification
3. Evolutionary approach for languages



Validation

Did we implement the right equations?

vs

Verification

Did we implement the equations right?

Challenge

Telling these two apart when results are not as expected

Software bugs undermine reproduction

Testing

“Smoke” testing

Unit testing

Integration testing

Important, but incomplete
(does not rule out all bugs)

vs.

Verification

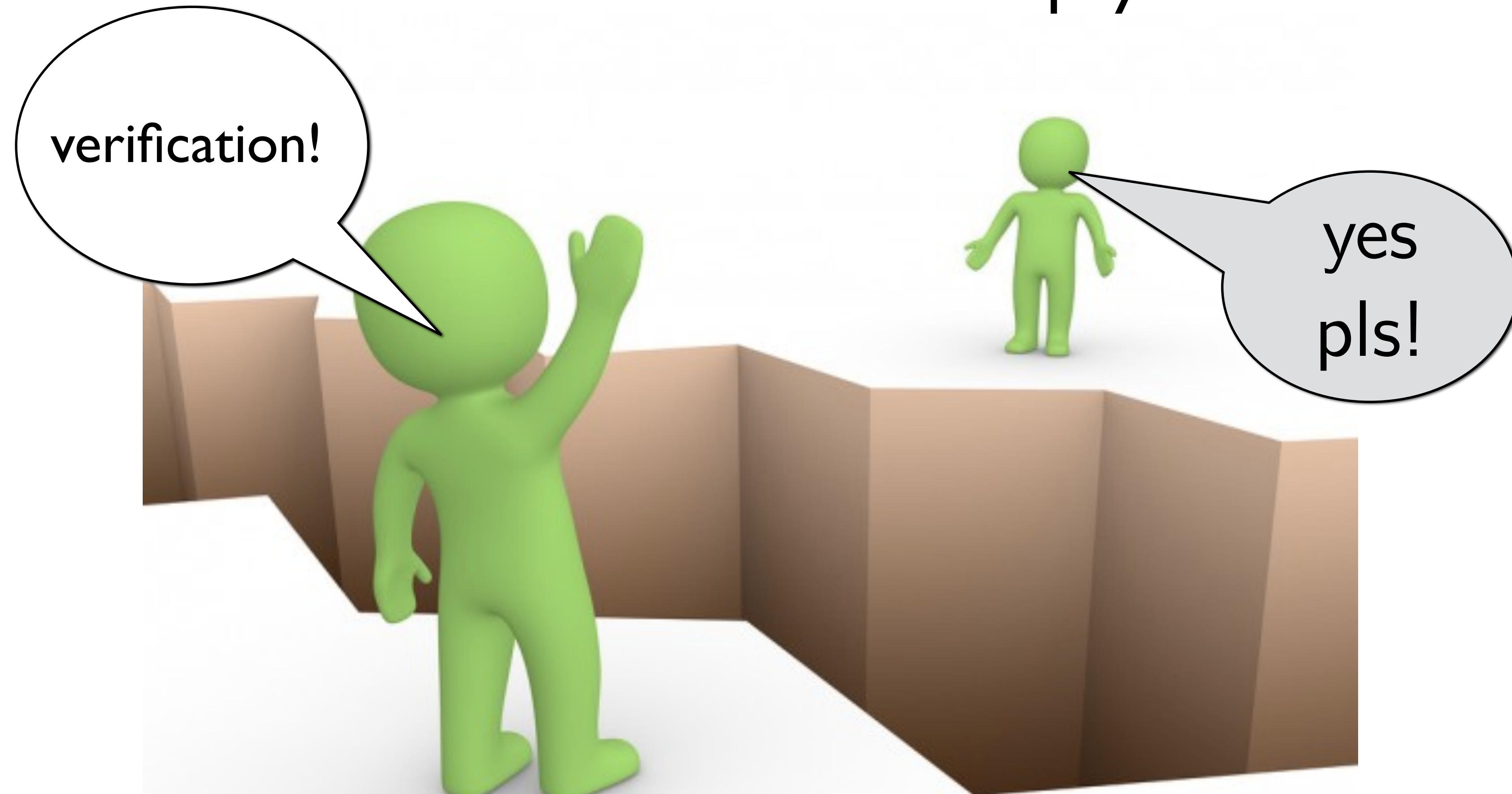
Types

Static analysis

Specify-and-check

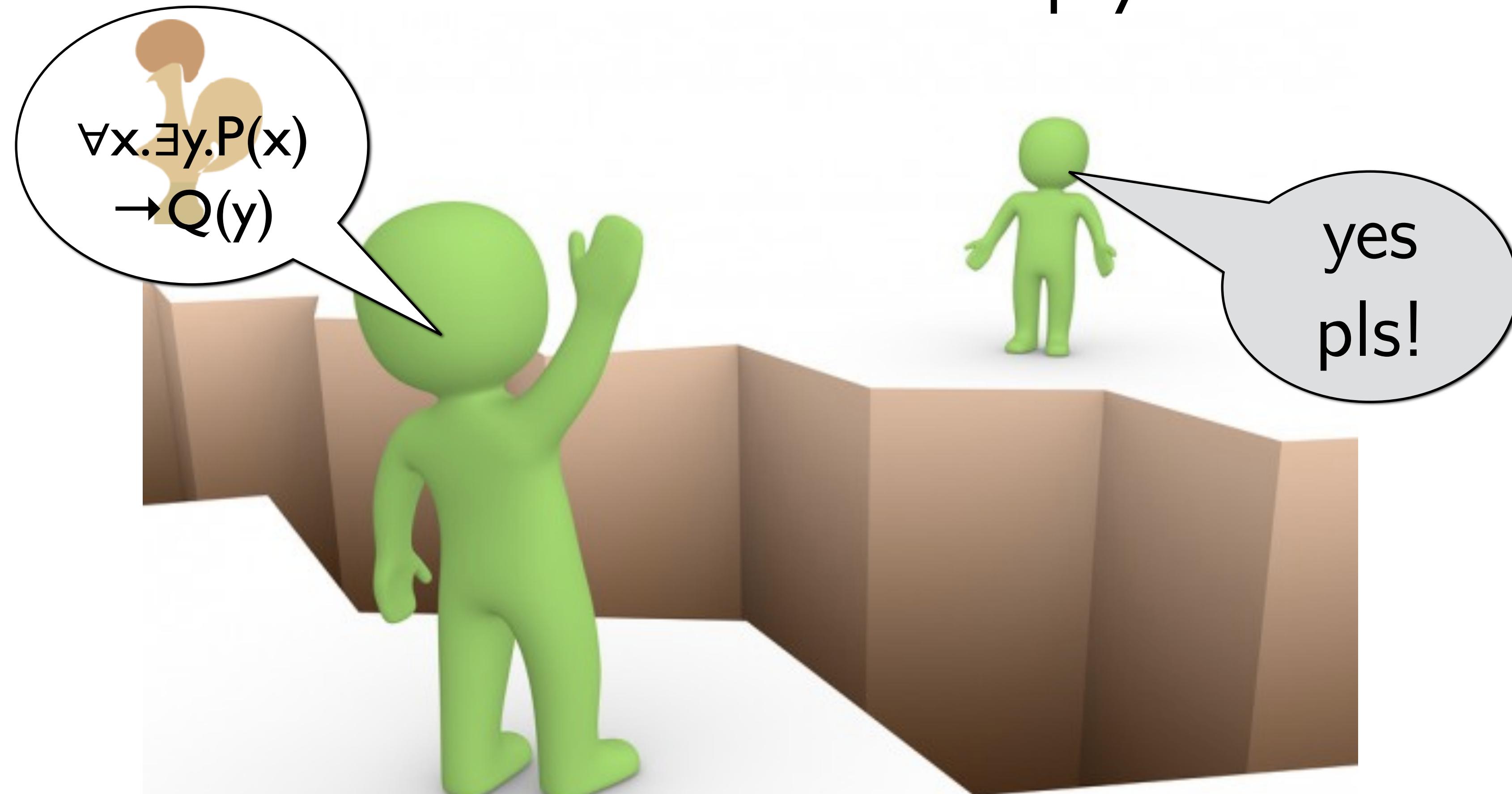
Largely unexplored
in climate science

natural & physical sciences



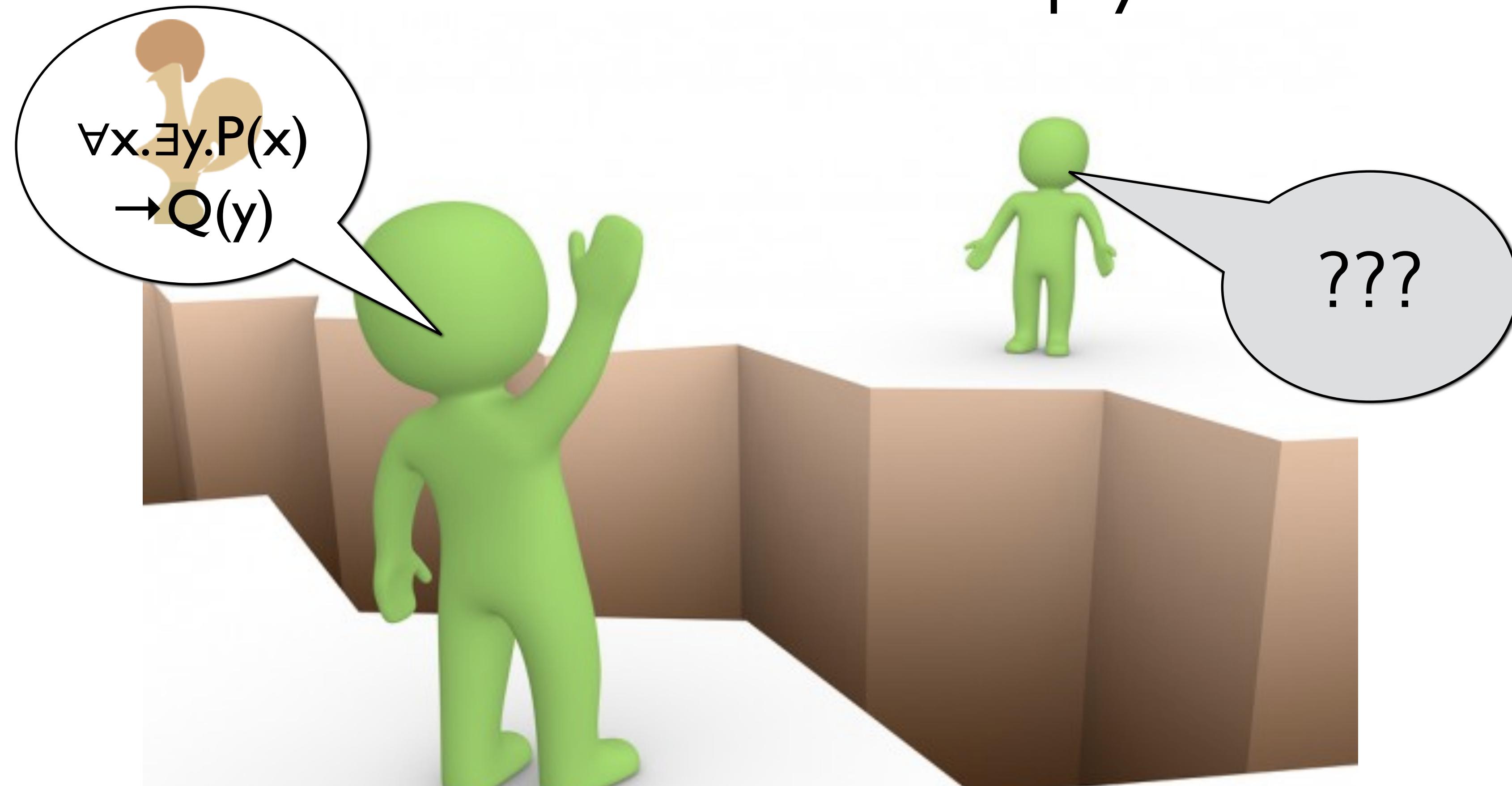
computer science

natural & physical sciences



computer science

natural & physical sciences



computer science

Let's bridge the chasm!

Approaches to verification

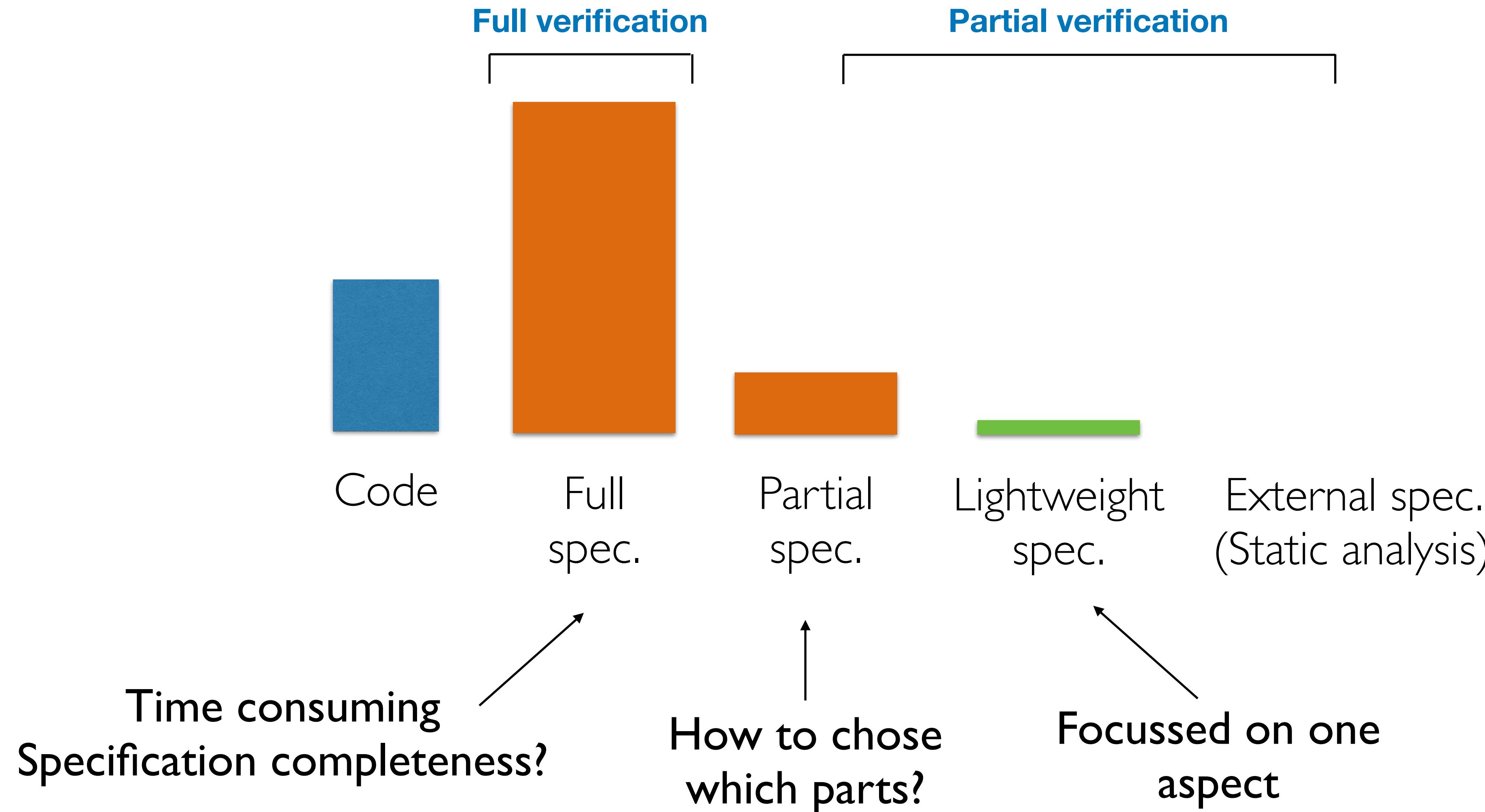




photo from Andrew Kennedy's website
<http://research.microsoft.com/en-us/um/people/akenn/units/>



Evolving Fortran types with inferred units-of-measure

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1. Introduction

Type systems are one of the most popular static techniques for recognizing and rejecting large classes of programming error. A common analogy for types is of physical quantities (e.g., in [2]), where type checking excludes, for example, the non-sensical addition of non-comparable quantities such as adding 3 m to 2 J; they have different *dimensions* (length vs. energy) and different *units* (metres vs. joules). This analogy between types and dimensions/units goes deeper. The approach of *dimensional analysis* checks the consistency of formulae involving physical quantities, acting as a kind of type system (performed by hand, long before computers). Various automatic type-system-like approaches have been proposed for including dimensional analysis in programming languages (e.g. [10] is a famous paper detailing one such approach, which also cites much of the relevant history of other systems).

Failing to ensure that the dimensions (or units) of values are correctly matched can be disastrous. An extreme example of this is the uncaught unit mismatch which led to the destruction of the

circumstances. It therefore seems inevitable that these errors are likely in computational science too.

The importance of units is often directly acknowledged in source code. We have seen source files carefully commented with the units and dimensions of each variable and parameter. We have also watched programmers trying to use this information: a process of scrolling up and down, repeatedly referring to the unit specification of each parameter. Incorporating units into the type system would move the onus of responsibility from the programmer to the compiler.

A recent ISO standards proposal (N1969) for Fortran introduces a units-of-measure system which follows Fortran's tradition of explicitness [7]. Every variable declaration must have an explicit unit declaration and every composite unit (e.g., metres times seconds) must itself be explicitly declared. This imposes the extra burden of annotating variables directly on the programmer. As an example, we studied two medium-sized models (roughly 10,000 lines of code each) and found roughly a 1:10 ratio between variable declarations and lines of code. Thus, adding explicit units of



Units-of-Measure Correctness in Fortran Programs

Mistral Contrastin, Andrew Rice, and Matthew Danish | University of Cambridge
Dominic Orchard | Imperial College London

Much of mathematics' use in science revolves around measurements of physical quantities, both abstractly and concretely. Such measurements are naturally classified by their *dimension*, that is, whether the measurement is of distance, energy, time, and so on. Dimensionality is further refined by a measurement's units-of-measure (or units, for short), such as

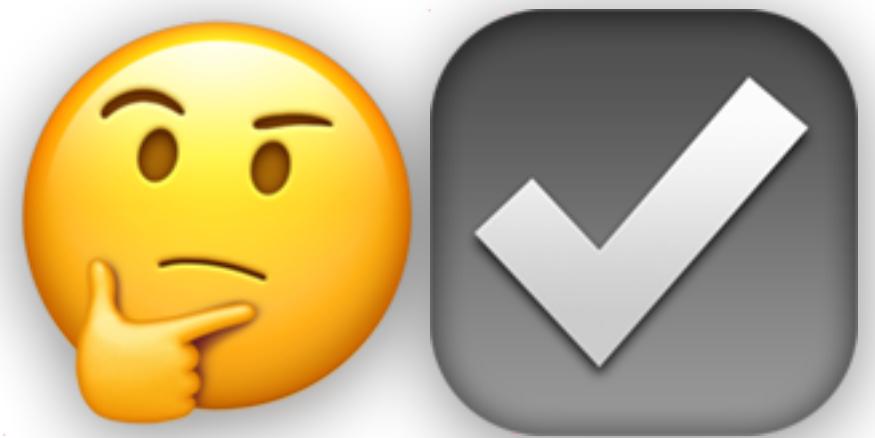
lightweight, nonbinding specification and analysis tools that can help find bugs in programs before they strike. We show that, in general, these kinds of program analysis tools will become more widely used by scientists to save time and reduce risk during the development process, as well as increase confidence in results of numerical models.

Ensuring the consistent use of units is an important part of this process.

CamFort

<https://github.com/camfort/camfort/>

Verification

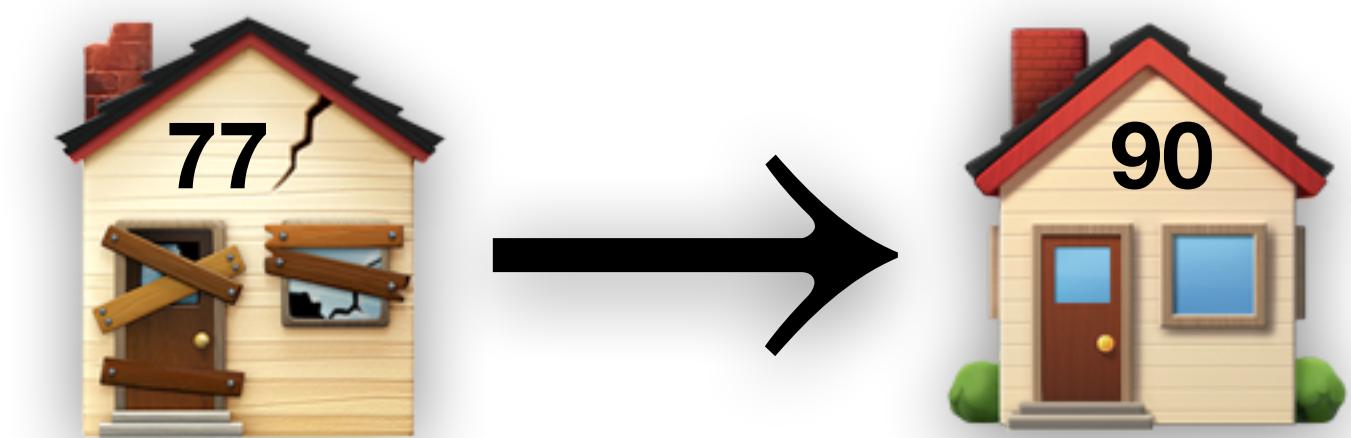


Analysis



Engineering and
Physical Sciences
Research Council

Refactoring



Bloomberg

Units-of-measure verification

```
1 program energy
2   real :: mass = 3.00, gravity = 9.91, height = 4.20
3   real :: potential_energy
4
5   potential_energy = mass * gravity * height
6 end program energy
```

Suggest

```
$ camfort units-suggest energy1.f90
```

```
Suggesting variables to annotate with unit specifications in 'energy1.f90'
```

```
...
```

```
energy1.f90: 3 variable declarations suggested to be given a
specification:
```

```
  energy1.f90 (2:43)      height
  energy1.f90 (2:14)      mass
  energy1.f90 (3:14)      potential_energy
```

Units-of-measure verification

```
1 program energy
2   != unit kg :: mass
3   != unit m  :: height
4   real :: mass = 3.00, gravity = 9.91, height = 4.20
5   != unit kg m**2/s**2 :: potential_energy
6   real :: potential_energy
7
8   potential_energy = mass * gravity * height
9 end program energy
```

Check

```
$ camfort units-check energy1.f90
```

```
energy1.f90: Consistent. 4 variables checked.
```

Units-of-measure verification

```
1 program energy
2   != unit kg :: mass
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4   real :: mass = 3.00, gravity = 9.91, height = 4.20
5   != unit kg m**2/s**2 :: potential_energy
6   real :: potential_energy
7
8   potential_energy = mass * gravity * height
9 end program energy
```

Synthesise

```
$ camfort units-synth energy1.f90 energy1.f90
```

```
Synthesising units for energy1.f90
```

Units-of-measure verification

```
1 program energy
2   != unit kg :: mass
3   != unit m  :: height
4   != unit m/s**2  :: gravity
5   real :: mass = 3.00, gravity = 9.91, height = 4.20
6   != unit kg m**2/s**2 :: potential_energy
7   real :: potential_energy
8
9   potential_energy = mass * gravity * height
10 end program energy
```

Synthesise

```
$ camfort units-synth energy1.f90 energy1.f90
```

```
Synthesising units for energy1.f90
```

Check

Does it do what I think it does?

Infer

What does it do?

Synthesise

Capture what it does for documentation & future-proofing

Suggest

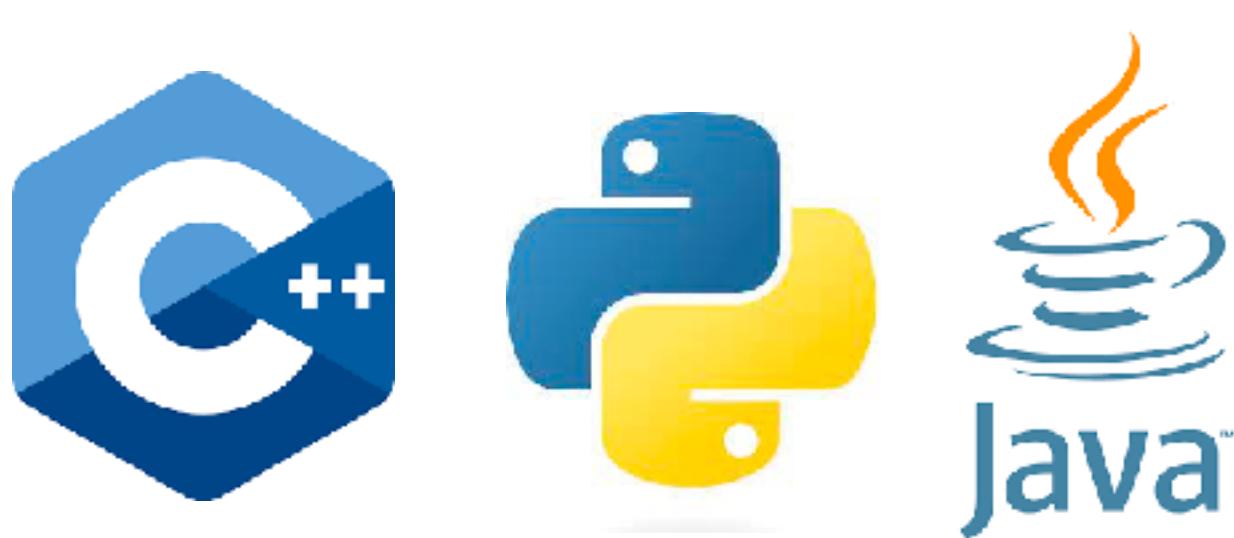
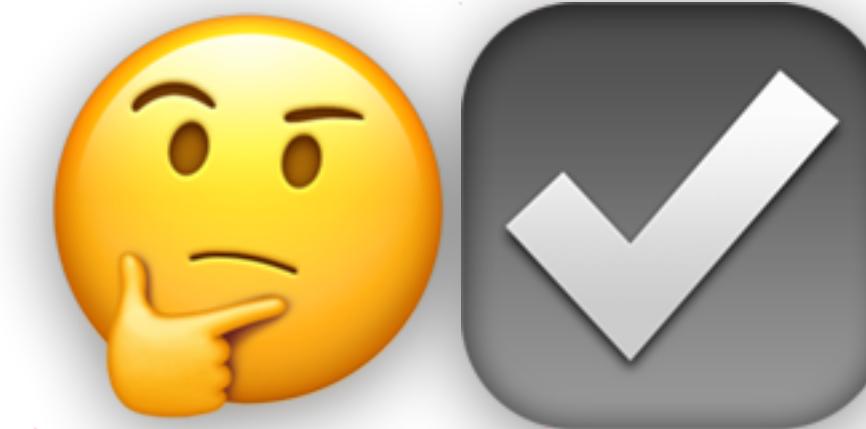
Where should I add a specification to get the most information?

Analysis



CamFort

Verification



See general tools
e.g.



Take home messages

- Are we done with language developments? **No!** But changes slow
- Verification for:
 - ▶ Increasing trust
 - ▶ Speeding up development
 - ▶ Enabling reuse
- I am keen to explore more ideas in this space



@dorchard

<https://dorchard.github.io>

<https://camfort.github.io>