Introduction

Semantics and applications to verification

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Program of this first lecture

Introduction to the course:

- a study of some examples of software errors
 - what are the causes ? what kind of properties do we want to verify ?
- a panel of the main verification methods with a fundamental limitation: indecidability
 - many techniques allow to compute semantic properties
 - each comes with advantages and drawbacks
- an introduction to the theory of ordered sets (or, most likely, mostly a refresher...)
 - order relations are pervasive in semantics and verification
 - fixpoints of operators are also very common

Case studies Ariane 5, Flight 501 (1996)

Outline

Ariane 5 – Flight 501

Ariane 5:

- a satellite launcher
- replacement of Ariane 4, a lot more powerful
- first flight, June, 4th, 1996: failure!

Flight story:

- nominal take-off, normal flight for 36 seconds
- T + 36.7 s : angle of attack change, trajectory lost



• T + 39 s : disintegration of the launcher

Consequences:

- loss of satellites : more than \$ 370 000 000...
- launcher unusable for more than a year (delay !)

Full report available online:

http://esamultimedia.esa.int/docs/esa-x-1819eng.pdf

Trajectory control system design overview

Sensors: gyroscopes, inertial reference systems...

Calculators (hardware + software) :

- "Inertial Reference System" (SRI) : integrates data about the trajectory (read on sensors)
- "On Board Computer" (OBC) :

computes the engine actuations that are required to follow the pre-determined theoretical trajectory $% \left({{{\left[{{{c_{\rm{m}}}} \right]}_{\rm{max}}}} \right)$

Actuators: engines of the launcher follow orders from the OBC

Redundant systems (failure tolerant system):

- keep running even in the presence of one or several system failures
- traditional solution in embedded systems: duplication of systems aircraft flight system: 2 or 3 hydraulic circuits launcher like Ariane 5 : 2 SRI units (SRI 1 and SRI 2)
- there is also a control monitor

The root cause: an unhandled arithmetic error

Processor registers

Each register has a size of 16, 32, 64 bits:

- 64-bits floating point: values in range $[-3.6 \cdot 10^{308}, 3.6 \cdot 10^{308}]$
- 16-bits signed integers: values in range [-32768, 32767]
- upon copy of data: conversions are performed such as rounding
- when the values are too large:
 - interruption: run error handling code if any, otherwise crash
 - or unexpected behavior: modulo arithmetic or other

Ariane 5:

- the SRI hardware runs in interruption mode
- it has no error handling code for arithmetic interruptions
- an unhandled arithmetic conversion overflow crashes the SRI

From the root cause to the failure

A not so trivial sequence of events:

- a conversion from 64-bits float to 16-bits signed int is performed and causes an overflow
- 2 an interruption is raised
- Output to the lack of error handling code, the SRI crashes
- the crash causes an error return (negative integer value) value be sent to the OBC (On-Board Computer)
- the OBC interprets this illegal value as flight data
- this causes the computation of an absurd trajectory
- I hence the loss of control of the launcher

Let us discuss a few specific points

A crash due to an unaddressed software case

Several solutions would have prevented this mishappening:

O Deactivate interruptions on overflows:

- then, an overflow may happen, and produce wrong values in the SRI
- but, these wrong values will not cause the computation to stop! and most likely, the flight will not be impacted too much

② Fix the SRI code, so that no overflow can happen:

all conversions must be guarded against overflows:

```
double x = /* ... */;
short i = /* ... */;
if( -32768. <= x && x <= 32767. )
    i = (short) x;
else
    i = /* default value */;
```

this may be costly (many tests), but redundant tests can be removed

Image: Handle conversion errors (not trivial):

- the handling code should identify the problem and fix it at run-time
- the OBC should identify illegal input values

A crash due to a useless task

Piece of code that generated the error:

- part of a gyroscope re-calibration process
- very useful to quickly restart the launch process after a short delay
- can only be done before lift-off...
- ... but not after!

Re-calibration task shut down:

- normally planned **50** seconds after lift-off...
- no chance of a need for such a re-calibration after $T_0 + 3$ seconds
- the crash occurred at 36 seconds

A crash due to legacy software

Software history:

- already used in Ariane 4 (previous launcher, before Ariane 5)
- the software was tested and ran in real conditions many times yet never failed...
- but Ariane 4 was a much less powerful launcher

Software optimization:

- many conversions were initially protected by a safety guard
- but these tests were considered expensive

 (a test and a branching take processor cycles, interact with the
 pipeline...)
- thus, conversions were ultimately removed for the sake of performance

Yet, Ariane 5 violates the assumptions that were valid with Ariane 4

- higher values of horizontal bias were generated
- those were never seen in Ariane 4, hence the failure

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A crash not prevented by redundant systems

Principle of redundant systems: survive the failure of a component by the use of redundant systems

System redundancy in Ariane 5:

- one OBC unit
- two SRI units... yet running the same software

Obviously, physical redundancy does not address software issues

Other implementation of system redundancy (e.g., Airbus FBW):

- two independent set of controls
- three computing units per set of controls
- each computing unit comprises two computers with distinct softwares (design and implementation is also performed in distinct teams)

Ariane 501, a summary of the issues

A long series of design errors, all related to a lack of understanding of what the software does:

- O Non-guarded conversion raising an interruption due to overflow
- **2** Removal of pre-existing guards, too high confidence in the software
- Non revised assumptions on the inputs when moving from Ariane 4 to Ariane 5
- Redundant systems running the same software
- **O** Useless task **not shutdown** at the right time

Current status: such issues can be found by static analysis tools

Case studies Lufthansa Flight 2904, Warsaw (1993)

Outline

High-speed runway overshoot at landing

Landing at Warsaw airport, Lufthansa A320:

- bad weather conditions: rain, high side wind
- wet runway
- landing (300 km/h) followed by aqua-planing, and delayed braking
- runway overrun at 132 km/h
- impact against a hillside at about 100 km/h

Consequences:

- 2 fatalities, 56 injured (among 70 passengers + crew)
- aircraft completely destroyed (impact + fire)

Full report available online:

http://www.rvs.uni-bielefeld.de/publications/Incidents/ DOCS/ComAndRep/Warsaw/warsaw-report.html

Causes of the accident

O Root cause:

- bad weather conditions not well assessed by the crew
- side wind exceeding aircraft certification specification
- wrong action from the crew:
 - a "Go Around" (missed landing, acceleration $+\ {\rm climb})$ should have been done

② Contributing factor: delayed action of the brake system

time (seconds)	distance (meters)	events
	from runway threshold	
T ₀	770 m	main landing gear landed
$T_0 + 3 s$	1030 m	nose landing gear landed
		brake command activated
T ₀ + 12 s	1680 m	spoilers activated
$T_0 + 14 \mathrm{s}$	1800 m	thrust reversers activated
$T_0 + 31 \mathrm{s}$	2700 m	end of runway

Protection of aircraft brake systems

• Braking systems inhibition: Prevent in-flight activation !

- spoilers: increase in aerodynamic load (drag)
- thrust reversers: could destroy the plane if activated in-flight ! (ex : crash of a B 767-300 ER Lauda Air, 1991, 223 fatalities; thrust reversers in-flight activation, electronic circuit issue)

• Braking software specification:

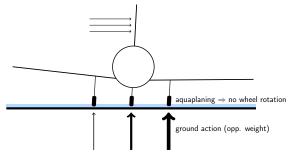
DO NOT activate spoilers and thrust reverse unless the following condition is met:

- thrust lever should be set to minimum by the flight crew
- **AND** either of the following conditions:
 - \star weight on the main gear should be at least 12 T
 - i.e., 6 T for each side
 - * OR wheels should be spinning, with a speed of at least 130 km/h

[Minimum Thrust] AND ([Weight] OR [Wheels spinning])

Understanding the braking delay

• Landing configuration:



• Braking systems: inhibited

- thrust command properly set to minimum
- no weight on the left landing gear due to the wind
- no speed on wheels due to aquaplanning

[Minimum Thrust] AND ([Weight] OR [Wheels spinning])

Flight 2904, a summary of the issues

Main factor is human (landing in weather conditions the airplane is not certified for), but the specification of the software is a contributing factor:

• Old condition that failed to be satisfied:

 $(P_{
m left} > 6T)$ AND $(P_{
m right} > 6T)$

• Fixed condition (used in the new version of the software):

$$(P_{\mathrm{left}} + P_{\mathrm{right}}) > 12T$$

- The fix can be understood only with knowledge of the environment
 - conditions which the airplane will be used in
 - behavior of the sensors

Outline

The anti-missile "Patriot" system

- Purpose: destroy foe missiles before they reach their target
- Use in wars:
 - first Gulf war (1991)

protection of towns and military facilities in Israël and Saudi Arabia (against "Scud" missiles launched by Irak)

- success rate:
 - * around 50 % of the "Scud" missiles are successfully destroyed
 - * almost all launched Patriot missiles destroy their target
 - * failures are due to failure to launch a Patriot missile
- Constraints on the system:
 - hit very quickly moving targets: "Scud" missiles fly at around 1700 m/s ; travel about 1000 km in 10 minutes
 - not to destroy a friendly target (it happened at least twice!)
 - very high cost: about \$1 000 000 per launch

System components

Detection / trajectory identification:

- detection using radar systems
- trajectory confirmation (to make sure a foe missile is tracked):
 - **1** trajectory identification using a sequence of points at various instants
 - **2** trajectory confirmation
 - computation of a predictive window (from position and speed vector)
 - + confirmation of the predicted trajectory
 - identification of the target (friend / foe)

Guidance system:

- interception trajectory computation
- launch of a Missile, and control until it hits its target high precision required (both missiles travel at more than 1500 m/s)

Very short process: about ten minutes

Dahran failure (1991)

Launch of a "Scud" missile

- Oetection by the radars of the Patriot system but failure to confirm the trajectory:
 - imprecision in the computation of the clock of the detection system
 - computation of a wrong confirmation window
 - the "Scud" cannot be found in the predicted window failure to confirm the trajectory
 - the detection computer concludes it is a false alert
- The "Scud" missile hits its target:
 28 fatalities and around 100 people injured

Fixed precision arithmetic

- Fixed precision numbers are of the form $\epsilon N 2^{-p}$ where:
 - p is fixed
 - $\epsilon \in \{-1, 1\}$ is the sign
 - $N \in [-2^n, 2^n 1]_{\mathbb{Z}}$ is an integer (n > p)
- In 32 bits fixed precision, with one sign bit, n = 31; thus we may let p = 20

• A few examples:

decimal value	sign	truncated value	fractional portion
2	0	0000000010	000000000000000000000000000000000000000
-5	1	0000000101	000000000000000000000000000000000000000
0.5	0	00000000000	100000000000000000000000000000000000000
-9.125	1	0000001001	0010000000000000000000

• Range of values that can be represented:

$$\pm 2^{12}(1-2^{-32})$$

Rounding errors in fixed precision computations

- Not all real numbers in the right range can be represented rounding is unavoidable may happen both for basic operations and for program constants...
- Example: fraction 1/10
 - ▶ 1/10 cannot be represented exactly in fixed precision arithmetic
 - let us decompose 1/10 as a sum of terms of the form $\frac{1}{2^i}$):

$$\begin{array}{rcl} \frac{1}{10} & = & \frac{1}{2} \cdot \frac{1}{5} \\ \frac{1}{5} & = & \frac{1}{8} + \frac{1}{16} + \frac{1}{16} \cdot \frac{1}{5} = \frac{1}{8} + \frac{1}{16} + \frac{1}{16} \cdot \left(\frac{1}{8} + \frac{1}{16} + \frac{1}{16} \cdot \frac{1}{5}\right) = \dots \end{array}$$

infinite binary representation: 0.00011001100110011001100...

• Floating precision numbers (more commonly used today) have the same limitation

The root cause: a clock drift

Trajectory confirmation algorithm (summary):

- hardware clock T_d ticks every tenth of a second
- time T_c is computed in seconds: $T_c = \frac{1}{10} \times T_d$
- in binary: $T_c = 0.00011001100110011001b \times_b T_d$!
- relative error is 10^{-6}
- after the computer has been running for 100 h :
 - the absolute error is 0.34 s
 - as a "Scud" travels at 1700 m/s : the predicted window is about 580 m from where it should be this explains the trajectory confirmation failure!

Remarks:

- the issue was discovered by israeli users, who noticed the clock drift their solution: frequently restart the control computer... (daily)
- this was not done in Dahran... the system had been running for 4 days

Patriot missile failure, a summary of the issues

Precision issues in the fixed precision arithmetic:

- A scalar constant used in the code was invalid i.e., bound to be rounded to an approximate value, incurring a significant approximation the designers were unaware of
- There was no adequate study of the precision achieved by the system, although precision is clearly critical here !

Current status: such issues can be found by static analysis tools

Outline

Common issues causing software problems

The examples given so far **are not isolated cases** See for instance:

```
www.cs.tau.ac.il/~nachumd/horror.html
```

(not up-to-date)

Typical reasons:

- Improper specification or understanding of the environment, conditions of execution...
- Incorrect implementation of a specification
 - e.g., the code should be free of runtime errors
 - e.g., the software should produce a result that meets some property
- Incorrect understanding of the execution model
 - e.g., generation of too imprecise results

New challenges to ensure embedded systems do not fail

Complex software architecture: e.g. parallel softwares

- single processor multi-threaded, distributed (several computers)
- more and more common: multi-core architectures
- very hard to reason about
 - other kinds of issues: dead-locks, races...
 - very complex execution model: interleavings, memory models

Complex properties to ensure: e.g., security

- the system should resist even in the presence of an attacker (agent with malicious intentions)
- attackers may try to access sensitive data, to corrupt critical data...
- security properties are often even hard to express

Techniques to ensure software safety

Software development techniques:

- **software engineering**, with a focus on specification, and software quality (may be more or less formal...)
- programming rules for specific areas (e.g., DO 178 c in avionics)
- usually do not guarantee any strong property, but make softwares "cleaner"

Formal methods:

- should have sound mathematical foundations
- should allow to guarantee softwares meet some complex properties
- should be trustable (is a paper proof ok ???)
- increasingly used in real life applications, but still a lot of open problems

What is to be verified ?

What do the C programs below do ? What do these C programs do ?

P0.c

```
int x = 0
int f0( int y ){
  return x * y;
}
int f1( int y ){
  x = y;
  return 0;
}
void main( ){
  int z = f0( 10 ) +
   f1( 100 );
}
```

P1.c void main(){ int i; int t[100] = { 0, 1, 2, ..., 99 }; while(i < 100){ t[i]++; i++; } } }</pre>

P2.c

```
void main(){
  float f = 0.;
  for( int i = 0;
        i < 1000000;
        i++ )
      f = f + 0.1;
}</pre>
```

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Semantic subtleties...

P0.c

```
int x = 0
int f0( int y ){
  return x * y;
}
int f1( int y ){
  x = y;
  return 0;
}
void main( ){
  int z = f0( 10 ) + f1
      ( 100 );
}
```

Execution order:

- not specified in C
- specified in Java
- if left to right, z = 0
- if right to left, z = 1000

Semantic subtleties...

P1.c void main(){ int i: $int t[100] = \{ 0, 1, 2, \}$..., 99 }; while(i < 100){</pre> t[i]++; i++; } P2.c void main(){ float f = 0.;for(int i = 0; i < 1000000;i++) f = f + 0.1;}

Initialization:

- runtime error in Java
- read of a random value in C (the value that was stored before)

Floating point semantics:

- 0.1 is not representable exactly; what is it rounded to by the compiler ?
- rounding errors; what is the rounding mode at runtime ?

The two main parts of this course

Semantics

- allow to describe precisely the behavior of programs should account for execution order, initialization, scope...
- allow to express the properties to verify several important families of properties: safety, liveness, security...
- also important to transform and compile programs

② Verification

- aim at proving semantic properties of programs
- a very strong limitation: indecidability
- several approaches, that make various compromises around indecidability

Outline

The termination problem

Termination

Program P terminates on input X if and only if any execution of P, with input X eventually reaches a final state

- Final state: final point in the program (i.e., not error)
- We may want to ensure termination:
 - processing of a task, such as, e.g., printing a document
 - computation of a mathematical function
- We may want to ensure *non*-termination:
 - operating system
 - device drivers

The termination problem

Can we find a program Pt that takes as argument a program P and data X and that returns "TRUE" if P terminates on X and "FALSE" otherwise ?

The termination problem is not computable

- Proof by reductio ad absurdum, using a *diagonal argument* We assume there exists a program Pa such that:
 - Pa always terminates
 - Pa(P, X) = 1 if P terminates on input X
 - Pa(P, X) = 0 if P does not terminate on input X
- We consider the following program:

```
void P0( P ){
    if( Pa( P, P ) == 1 ){
      while( 1 ){
            // loop forever
        }
    } else {
        return; // do nothing
    }
}
```

• What is the return value of Pa(P0, P0) ? i.e., does P0 terminate on input P0 ?

The termination problem is not computable

• What is the return value of Pa(P0, P0) ?

We know Pa always terminates and returns either 0 or 1 (assumption). Therefore, we need to consider only two cases:

- if Pa(P0, P0) returns 1, then P0(P0) loops forever, thus Pa(P0, P0) should return 0, so we have reached a contradiction
- if Pa(P0,P0) returns 0, then P0(P0) terminates, thus Pa(P0,P0) should 1, so we have reached a contradiction
- In both cases, we reach a contradiction
- Therefore we conclude no such a Pa exists

The termination problem is not decidable

There exists no program Pt that always terminates and always recognizes whether a program P terminates on input X

Absence of runtime errors

- Can we find a program Pc that takes a program P and input X as arguments, always terminates and returns
 - ▶ 1 if and only P runs safely on input X, i.e., without a runtime error
 - 0 if P crashes on input X
- Answer: No, the same diagonal argument applies if Pc(P, X) decides whether P will run safely on X, consider

```
void P1( P ){
    if( Pc( P, P ) == 1 ){
        0 / 0; // deliberately crash
            (unsafe)
    } else {
        return; // do nothing
    }
}
```

Non-computability result

The absence of runtime errors is not computable

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Introduction

Rice theorem

- Semantic specification: set of correct program executions
- "Trivial" semantic specifications:
 - empty set
 - set of all possible executions
 - \Rightarrow intuitively, the non interesting verification problems...

Rice theorem (1953)

Considering a Turing complete language, any non trivial semantic specification is not computable

- Intuition: there is no algorithm to decide non trivial specifications, starting with only the program code
- Therefore all interesting properties are not computable :
 - termination,
 - absence of runtime errors,
 - absence of arithmetic errors, etc...

Outline

Towards partial solutions

The initial verification problem is **not computable Solution**: solve a weaker problem

Several compromises can be made:

- simulation / testing: observe only finitely many finite executions infinite system, but only finite exploration (no proof beyond that)
- assisted theorem proving: we give up on automation (no proof inference algorithm in general)
- model checking: we consider only finite systems (with finitely many states)
- **bug-finding**: search for "patterns" indicating "likely errors" (may miss real program errors, and report non existing issues)
- static analysis with abstraction: attempt at automatic correctness proofs

(yet, may fail to verify some correct programs)

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Introduction

Safety verification method characteristics

Safety verification problem

- Semantics [P] of program P: set of behaviors of P (e.g., states)
- Property to verify S: set of admissible behaviors (e.g., safe states)

Goal: establish $\llbracket P \rrbracket \subseteq S$

- Automation: existence of an algorithm
- Scalability: should allow to handle large softwares
- Soundness: identify any wrong program
- Completeness: accept all correct programs
- Apply to program source code, i.e., not require a modelling phase

1. Testing by simulation

Principle

Run the program on finitely many finite inputs

- maximize coverage
- inspect erroneous traces to fix bugs

• Very widely used:

- unit testing: each function is tested separately
- integration testing: with all surrounding systems, hardware e.g., iron bird in avionics
- Automated
- Complete: will never raise a false alarm
- Unsound unless exhaustive: may miss program defects
- Costly: needs to be re-done when software gets updated

2. Machine assisted proof

Principle

Have a machine checked proof, that is partly human written

- tactics / solvers may help in the inference
- the hardest invariants have to be user-supplied

Applications

- software industry (rare): Line 14 in Paris Subway
- hardware: ACL 2
- academia: CompCert compiler, SEL4 verified micro-kernel
- ▶ also for math: four colour theorem, Feith-Thomson theorem
- Not fully automated

often turns out costly as complex proof arguments have to be found

• Sound and quasi-complete (in practice fine...)

3. Model-Checking

Principle

Consider finite systems only, using algorithms for

- exhaustive exploration,
- symmetry reduction...

• Applications:

- hardware verification
- driver protocols verification (Microsoft)
- Applies on a model: a model extraction phase is needed
 - for infinite systems, this is necessarily approximate
 - not always automated
- Automated, sound, complete with respect to the model

4. "Bug finding"

Principle

Identify "likely" issues, i.e., patterns known to often indicate an error

- use bounded symbolic execution, model exploration...
- rank "defect" reports using heuristics
- Intuition: model checking made unsound
- Example: Coverity
- Automated
- Not complete: may report false alarms
- Not sound: may accept false programs thus inadequate for safety-critical systems

5. Static analysis with abstraction (1/4)

Principle

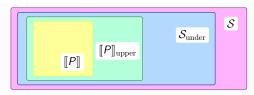
Use some approximation, but always in a conservative manner

- Under-approximation of the property to verify: $\mathcal{S}_{under} \subseteq \mathcal{S}$
- Over-approximation of the semantics: $\llbracket P \rrbracket \subseteq \llbracket P \rrbracket_{upper}$
- We let an automatic static analyzer attempt to prove that:

 $\llbracket P \rrbracket_{upper} \subseteq \mathcal{S}_{under}$

If it succeeds, $\llbracket P \rrbracket \subseteq \mathcal{S}$

 $\bullet~$ In practice, the static analyzer computes $[\![P]\!]_{\rm upper}, \mathcal{S}_{\rm under}$

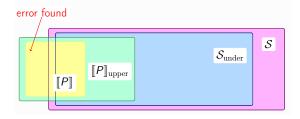


5. Static analysis with abstraction (2/4)

Soundness

The abstraction will catch any incorrect program

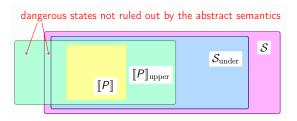
• If
$$\llbracket P \rrbracket \not\subseteq S$$
, then $\llbracket P \rrbracket_{upper} \not\subseteq S_{under}$
since $\begin{cases} S_{under} \subseteq S \\ \llbracket P \rrbracket \subseteq \llbracket P \rrbracket_{upper} \end{cases}$



5. Static analysis with abstraction (3/4)

Incompleteness

The abstraction may fail to certify some correct programs



Case of a false alarm:

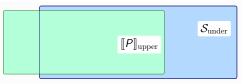
- program P is correct
- but the static analysis fails

5. Static analysis with abstraction (4/4)

Incompleteness

The abstraction may fail to certify some correct programs

• In the following case, the analysis cannot conclude anything



• One goal of the static analyzer designer is to avoid such cases

Static analysis using abstraction

- \bullet Automatic: $[\![P]\!]_{\rm upper}, \, \mathcal{S}_{\rm under}$ computed automatically
- Sound: reports any incorrect program
- Incomplete: may reject correct programs

A summary of common verification techniques

	Automatic	Sound	Complete	Source level	Scalable
Simulation	Yes	No ¹	Yes	Yes	sometimes ²
Assisted proving	No	Yes	Almost	Partially	sometimes ³
Model-checking	Yes	Yes	Partially ⁴	No	sometimes
Bug-finding	Yes	No	No	Yes	sometimes
Static analysis	Yes	Yes	No	Yes	sometimes

• Obviously, no approach checks all characteristics

• Scalability is a challenge for all

⁴only with respect to the finite models... but not with respect to infinite semantics

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¹unless full testing is doable

²full testing usually not possible except for small programs with finite state space ³quickly requires huge manpower

Outline

Order relations

Very useful in semantics and verification:

- logical ordering, expresses implication of logical facts
- computational ordering, useful to establish well-foundedness of fixpoint definitions, and for termination

Definition: partially ordered set (poset)

Let a set S and a binary relation $\sqsubseteq \subseteq S \times S$ over S. Then, \sqsubseteq is an order relation (and (S, \sqsubseteq) is called a **poset**) if and only if it is

- reflexive: $\forall x \in S, x \sqsubseteq x$
- transitive: $\forall x, y, z \in S, x \sqsubseteq y \land y \sqsubseteq z \implies x \sqsubseteq z$
- antisymmetric: $\forall x, y \in S, x \sqsubseteq y \land y \sqsubseteq x \implies x = y$

• notation:
$$x \sqsubset y ::= (x \sqsubseteq y \land x \neq y)$$

Graphical representation

We often use Hasse diagrams to represent posets:

Extensive definition:

•
$$S = \{x_0, x_1, x_2, x_3, x_4\}$$

• \square defined by:

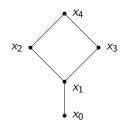


- By reflexivity, we have, e.g., $x_1 \sqsubseteq x_1$
- By transitivity, we have, e.g., $x_1 \sqsubseteq x_4$

Order relations are very useful in semantics...

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Example: semantics of automata

In the following, we **illustrate order relations** and their **usefulness in semantics** using **word automata**.

We consider the classical notion of finite word automata and let

- L be a finite set of letters
- Q be a finite set of states
- ${\it q}_{
 m i}, {\it q}_{
 m f} \in {\it Q}$ denote the initial state and final state
- $\rightarrow \subseteq Q \times L \times Q$ be a transition relation

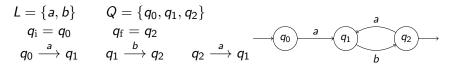
Semantics of an automaton

The set of words recognized by $\mathcal{A}=(\mathit{Q},\mathit{q_{\rm i}},\mathit{q_{\rm f}},\rightarrow)$ is defined by:

$$\mathcal{L}[\mathcal{A}] = \{a_0 a_1 \dots a_n \mid \exists q_0 \dots q_{n-1} \in Q, \ q_i \xrightarrow{a_0} q_0 \xrightarrow{a_1} q_1 \dots q_{n-1} \xrightarrow{a_n} q_f\}$$

Example: automata and semantic properties

A simple automaton:



A few semantic properties:

• \mathcal{P}_0 : no recognized word contains two consecutive b

 $\mathcal{L}[\mathcal{A}] \subseteq L^* \setminus L^* bbL^*$

• \mathcal{P}_1 : all recognized words contain at least one occurrence of *a*

$$\mathcal{L}[\mathcal{A}] \subseteq L^* a L^*$$

• we could also consider under-approximation properties (of the form $\mathcal{P}_2 \subseteq \mathcal{L}[\mathcal{A}]$), but do not in this lecture

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Total ordering

Definition: total order relation

Order relation \sqsubseteq over ${\mathcal S}$ is a total order if and only if

 $\forall x, y \in \mathcal{S}, \ x \sqsubseteq y \lor y \sqsubseteq x$

Examples:

• real numbers:

 (\mathbb{R},\leq) is a total ordering

• powerset:

if set S has at least two distinct elements x, y then its powerset $(\mathcal{P}(S), \subseteq)$ is **not** a total order indeed $\{x\}, \{y\}$ cannot be compared

Most of the order relations we will use are *not* be total indeed: very often, powerset or similar

Minimum and maximum elements

Definition: extremal elements

Let $(\mathcal{S}, \sqsubseteq)$ be a poset and $\mathcal{S}' \subseteq \mathcal{S}$. Then x is

- minimum element of S' if and only if $x \in S' \land \forall y \in S', x \sqsubseteq y$
- maximum element of S' if and only if $x \in S' \land \forall y \in S', y \sqsubseteq x$
- maximum and minimum elements may not exist example: {{x}, {y}} in the powerset, where x ≠ y
- infimum \perp ("bottom"): minimum element of S
- supremum \top ("top"): maximum element of $\mathcal S$

Exercise:

what are the logical interpretations of infimum / supremum elements ?

Upper bounds and least upper bound

Definition: bounds

Given poset $(\mathcal{S}, \sqsubseteq)$ and $\mathcal{S}' \subseteq \mathcal{S}$, then $x \in \mathcal{S}$ is

• an upper bound of \mathcal{S}' if

$$\forall y \in \mathcal{S}', \ y \sqsubseteq x$$

• the least upper bound (lub) of S' (noted $\sqcup S'$) if $\forall y \in S', \ y \sqsubseteq x \land \forall z \in S, (\forall y \in S', \ y \sqsubseteq z) \implies x \sqsubseteq z$

 if it exists, the least upper bound is unique: if x, y are least upper bounds of S, then x ⊑ y and y ⊑ x, thus x = y by antisymmetry

• notation:
$$x \sqcup y ::= \sqcup \{x, y\}$$

• upper bounds and least upper bounds may not exist

• dual notions: lower bound, greatest lower bound (glb, noted $\sqcap S'$) Exercise: logical interpretations ?

Duality principle

So far all definitions admit a symmetric counterpart

• dual relation: given an order relation \sqsubseteq , \mathcal{R} defined by

$$x\mathcal{R}y \iff y \sqsubseteq x$$

is also an order relation

This is the duality principle:

minimum element infimum lower bound greatest lower bound least upper bound

... more to follow

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Complete lattice

Definition: complete lattice

A complete lattice is a tuple $(S, \sqsubseteq, \bot, \top, \sqcup, \sqcap)$ where:

- $(\mathcal{S}, \sqsubseteq)$ is a poset
- \perp is the infimum of ${\cal S}$
- ullet o o is the supremum of ${\mathcal S}$
- \bullet any subset \mathcal{S}' of \mathcal{S} has a lub $\sqcup\,\mathcal{S}'$ and a glb $\sqcap\,\mathcal{S}'$

Properties:

- $\bot = \sqcup \emptyset = \sqcap \mathcal{S}$
- $\top = \sqcap \emptyset = \sqcup \mathcal{S}$

Example:

the powerset $(\mathcal{P}(\mathcal{S}),\subseteq,\emptyset,\mathcal{S},\cup,\cap)$ of set \mathcal{S} is a complete lattice

Lattice

The existence of lubs and glbs for all subsets is often a very strong property, that may not be met:

Definition: lattice

- A lattice is a tuple $(S, \sqsubseteq, \bot, \top, \sqcup, \sqcap)$ where:
 - $(\mathcal{S},\sqsubseteq)$ is a poset
 - \perp is the infimum of ${\mathcal S}$
 - op is the supremum of ${\mathcal S}$
 - any pair $\{x, y\}$ of S has a lub $x \sqcup y$ and a glb $x \sqcap y$

• let $Q = \{q \in \mathbb{Q} \mid 0 \le q \le 1\}$; then (Q, \le) is a lattice but not a complete lattice indeed, $\{q \in Q \mid q \le \frac{\sqrt{2}}{2}\}$ has no lub in Q

• property: a finite lattice is also a complete lattice

Chains

Definition: increasing chain

Let $(\mathcal{S}, \sqsubseteq)$ be a poset and $\mathcal{C} \subseteq \mathcal{S}$.

It is an increasing chain if and only if

- it has an infimum (thus it is not empty)
- poset $(\mathcal{C}, \sqsubseteq)$ is total (i.e., any two elements can be compared)

Example, in the powerset $(\mathcal{P}(\mathbb{N}), \subseteq)$:

$$\mathcal{C} = \{c_i \mid i \in \mathbb{N}\}$$
 where $c_i = \{2^0, 2^2, \dots, 2^i\}$

Definition: increasing chain condition

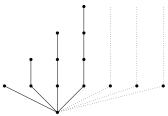
The poset (S, \sqsubseteq) satisfies the increasing chain condition if and only if any increasing chain $C \subseteq S$ is finite.

Complete partial orders

Definition: complete partial order

A complete partial order (cpo) is a poset (S, \sqsubseteq) such that any increasing chain C of S has a least upper bound. A pointed cpo is a cpo with an infimum \bot .

- clearly, any complete lattice is a cpo
- the opposite is not true:



Outline

Towards a constructive definition of the automata semantics

We now look for a **constructive version of the automaton semantics** as hinted by the following observations

Observation 1: $\mathcal{L}[\mathcal{A}] = \llbracket \mathcal{A} \rrbracket(q_f)$ where

$$\begin{bmatrix} \mathcal{A} \end{bmatrix} : \begin{array}{ccc} Q & \longrightarrow & \mathcal{P}(L^*) \\ q & \longmapsto & \{ w \in L^* \mid \exists n, \ w = a_0 a_1 \dots a_n \\ & & \exists q_0 \dots q_{n-1} \in Q, \ q_i \xrightarrow{a_0} q_0 \xrightarrow{a_1} \dots q_{n-1} \xrightarrow{a_n} q \} \end{array}$$

Observation 2: $[\![\mathcal{A}]\!] = \bigcup_{n \in \mathbb{N}} [\![\mathcal{A}]\!]_n$ where

$$\begin{split} \llbracket \mathcal{A} \rrbracket_n : & Q & \longrightarrow & \mathcal{P}(L^*) \\ & q & \longmapsto & \{ a_0 a_1 \dots a_{n-1} \mid \\ & & \exists q_0 \dots q_{n-2} \in Q, \ q_i \xrightarrow{a_0} q_0 \xrightarrow{a_1} \dots q_{n-1} \xrightarrow{a_{n-1}} q \} \end{split}$$

Observation 3: $[\![A]\!]_{n+1}$ can be computed directly from $[\![A]\!]_n$

$$\llbracket \mathcal{A} \rrbracket_{n+1}(q) = \bigcup_{q' \in Q} \{ wa \mid w \in \llbracket \mathcal{A} \rrbracket_n(q') \land q' \stackrel{a}{\longrightarrow} q \}$$

Towards a constructive definition of the automata semantics

Alternate approach:

• Let $[A]_n$ denote recognized words of length at most *n*:

$$\llbracket \mathcal{A} \rrbracket(q) = \{ w \in \llbracket \mathcal{A} \rrbracket(q) \mid \mathsf{length}(w) \leq n \}$$

- **2** Compute $\llbracket \mathcal{A} \rrbracket_{n+1}$ from $\llbracket \mathcal{A} \rrbracket_n$
- Obefine the semantics of the automaton as the union of the iterates of this sequence:

$$\llbracket \mathcal{A} \rrbracket = \bigcup_{n \in \mathbb{N}} \llbracket \mathcal{A} \rrbracket_n$$

In the following, we study such a way of defining semantics, based on general mathematical tools, that we will use throughout the course

Operators over a poset

Definition: operators and orderings

Let $(\mathcal{S}, \sqsubseteq)$ be a poset and $\phi : \mathcal{S} \to \mathcal{S}$ be an operator over \mathcal{S} . Then, ϕ is:

- monotone if and only if $\forall x, y \in S, x \sqsubseteq y \Longrightarrow \phi(x) \sqsubseteq \phi(y)$
- continuous if and only if, for any chain S' ⊆ S then: {
 if ⊔ S' exists, so does ⊔{φ(x) | x ∈ S'}
 and φ(⊔ S') = ⊔{φ(x) | x ∈ S'}

 ⊔-preserving if and only if: ∀S' ⊂ S, {
 if ⊔ S' exists, then ⊔{φ(x) | x ∈ S'} exists

, and
$$\phi(\sqcup S') = \sqcup \{\phi(x) \mid x \in S'\}$$

Notes:

- "monotone" in English means "croissante" in French ; "décroissante" translates into "anti-monotone" and "monotone" into "isotone"
- the dual of "monotone" is "monotone"

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Operators over a poset

A few interesting properties:

Continuity implies monotonicity

If ϕ is continuous, then it is also monotone

We assume ϕ is continuous, and $x, y \in S$ are such that $x \sqsubseteq y$: Then $\{x, y\}$ is a chain with lub y, thus $\phi(x) \sqcup \phi(y)$ exists and is equal to $\phi(\sqcup\{x, y\}) = \phi(y)$. Therefore $\phi(x) \sqsubseteq \phi(y)$.

⊔-preserving implies monotonicity

If ϕ preserves \sqcup , then it is also monotone

Same argument.

Fixpoints

Definition: fixpoints

- Let $(\mathcal{S}, \sqsubseteq)$ be a poset and $\phi : \mathcal{S} \to \mathcal{S}$ be an operator over \mathcal{S} .
 - a fixpoint of ϕ is an element x such that $\phi(x) = x$
 - a pre-fixpoint of ϕ is an element x such that $x \sqsubseteq \phi(x)$
 - a post-fixpoint of ϕ is an element x such that $\phi(x) \sqsubseteq x$
 - the least fixpoint lfp ϕ of ϕ (if it exists, it is unique) is the smallest fixpoint of ϕ
 - the greatest fixpoint gfp ϕ of ϕ (if it exists, it is unique) is the greatest fixpoint of ϕ

Note: the existence of a least fixpoint, a greatest fixpoint or even a fixpoint is *not guaranteed*; we will see several theorems that establish their existence under specific assumptions...

Tarski's Theorem

Theorem

Let $(S, \sqsubseteq, \bot, \top, \sqcup, \sqcap)$ be a complete lattice and $\phi : S \to S$ be a monotone operator over S. Then:

- ϕ has a least fixpoint **Ifp** ϕ and **Ifp** $\phi = \sqcap \{x \in S \mid \phi(x) \sqsubseteq x\}$.
- **2** ϕ has a greatest fixpoint **gfp** ϕ and **gfp** $\phi = \sqcup \{x \in S \mid x \sqsubseteq \phi(x)\}$.
- **③** the set of fixpoints of ϕ is a complete lattice.

Proof of point 1:

We let
$$X = \{x \in \mathcal{S} \mid \phi(x) \sqsubseteq x\}$$
 and $x_0 = \sqcap X$.

Let $y \in X$:

- $x_0 \sqsubseteq y$ by definition of the glb;
- thus, since ϕ is monotone, $\phi(x_0) \sqsubseteq \phi(y)$;
- thus, $\phi(x_0) \sqsubseteq y$ since $\phi(y) \sqsubseteq y$, by definition of X.

Therefore $\phi(x_0) \sqsubseteq x_0$, since $x_0 = \sqcap X$ and $\phi(x_0)$ is a lower bound.

Tarski's Theorem

We proved that $\phi(x_0) \sqsubseteq x_0$. We derive from this that:

- $\phi(\phi(x_0)) \sqsubseteq \phi(x_0)$ since ϕ is monotone;
- $\phi(x_0)$ is a post-fixpoint of ϕ , thus $\phi(x_0) \in X$;
- $x_0 \sqsubseteq \phi(x_0)$ by definition of the greatest lower bound

We have established both inclusions so $\phi(x_0) = x_0$.

If x_1 is another fixpoint, then $x_1 \in X$, so $x_0 \sqsubseteq x_1$.

Proof of point 2: similar, by duality.

Proof of point 3:

- if X is a set of fixpoints of φ, we need to consider φ over
 {y ∈ S | y ⊑_S ⊓ X} to establish the existence of a glb of X in the
 poset of fixpoints
- the existence of least upper bounds in the poset of fixpoints follows by duality

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Tarski's theorem: example (1)

A function over the powerset:

We consider a set \mathcal{E} , and a subset $\mathcal{A} \subseteq \mathcal{E}$ We let:

$$egin{array}{cccc} f: & \mathcal{P}(\mathcal{E}) & \longrightarrow & \mathcal{P}(\mathcal{E}) \ & X & \longmapsto & X \cup \mathcal{A} \end{array}$$

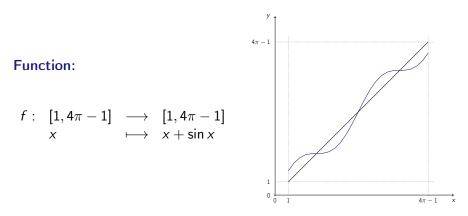
Exercise:

• apply Tarski's theorem, characterize the least and greatest fixpoints

Orderings, lattices, fixpoints

Operators over a poset and fixpoints

Tarski's theorem: example (2)



Exercise:

• apply Tarski's theorem, and derive the fixpoints of the function

Automata example, fixpoint definition

Lattice:

- $\mathcal{S} = Q \rightarrow \mathcal{P}(L^*)$
- \bullet the ordering is the pointwise extension \sqsubseteq of \supseteq

Operator:

Proof steps to complete:

- the existence of Ifp ϕ follows from Tarski's theorem
- the equality Ifp $\phi = [\![A]\!]$ can be established by induction and double inclusion... but there is a simpler way

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Introduction

Kleene's Theorem

Tarski's theorem guarantees existence of an lfp, but is not constructive.

Theorem

Let (S, \sqsubseteq, \bot) be a pointed cpo and $\phi : S \to S$ be a continuous operator over S. Then ϕ has a least fixpoint, and

If
$$\phi = \bigsqcup_{n \in \mathbb{N}} \phi^n(\bot)$$

First, we prove the existence of the lub:

Since ϕ is continuous, it is also monotone. We can prove by induction over n that $\{\phi^n(\bot) \mid n \in \mathbb{N}\}$ is a chain:

•
$$\phi^0(\bot) = \bot \sqsubseteq \phi(\bot)$$
 by definition of the infimum;

• if
$$\phi^n(\bot) \sqsubseteq \phi^{n+1}(\bot)$$
, then

$$\phi^{n+1}(\bot) = \phi(\phi^n(\bot)) \sqsubseteq \phi(\phi^{n+1}(\bot)) = \phi^{n+2}(\bot)$$

By definition of the cpo structure, the lub exists. We let x_0 denote it.

Kleene's Theorem

Secondly, we prove that it is a fixpoint of ϕ : Since ϕ is continuous, $\{\phi^{n+1}(\bot) \mid n \in \mathbb{N}\}$ has a lub, and

$$\begin{array}{lll} \phi(x_0) &=& \phi(\sqcup\{\phi^n(\bot) \mid n \in \mathbb{N}\}) \\ &=& \sqcup\{\phi^{n+1}(\bot) \mid n \in \mathbb{N}\} & \text{by continuity of } \phi \\ &=& \bot \sqcup(\sqcup\{\phi^{n+1}(\bot) \mid n \in \mathbb{N}\}) & \text{by definition of } \bot \\ &=& x_0 & \text{by simple rewrite} \end{array}$$

Last, we show that it is the **least** fixpoint:

Let x_1 denote another fixpoint of ϕ . We show by induction over *n* that $\phi^n(\bot) \sqsubseteq x_1$:

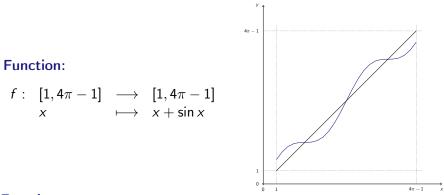
•
$$\phi^0(\bot) = \bot \sqsubseteq x_1$$
 by definition of \bot ;

• if $\phi^n(\perp) \sqsubseteq x_1$, then $\phi^{n+1}(\perp) \sqsubseteq \phi(x_1) = x_1$ by monotony, and since x_1 is a fixpoint.

By definition of the lub, $x_0 \sqsubseteq x_1$

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Kleene's theorem: example



Exercise:

• apply Kleene's theorem and sketch the iterations

Automata: constructive semantics

We can now state a **constructive definition** of the automaton semantics. Operator ϕ is defined by

$$\phi(f) = \lambda(q \in Q) \cdot \begin{cases} f(q) \cup \phi_0(f)(q_i) \cup \{\epsilon\} & \text{if } q = q_i \\ f(q) \cup \phi_0(f)(q) & \text{otherwise} \end{cases}$$

Proof steps:

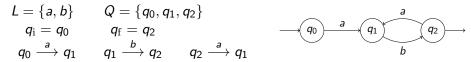
- ϕ is continuous
- thus, Kleene's theorem applies so Ifp ϕ exists and Ifp $\phi = \bigcup_{n \in \mathbb{N}} \phi^n(\bot) \dots$ \dots this actually saves the double inclusion proof to establish that $\llbracket \mathcal{A} \rrbracket = \mathsf{lfp} \phi$

Furthermore, $\llbracket \mathcal{A} \rrbracket = \bigcup_{n \in \mathbb{N}} \phi^n(\bot)$.

This fixpoint definition will be very useful to infer or verify semantic properties.

Automata: constructive semantics iterates

A simple automaton:



Iterates of function ϕ from \perp :

Iterate	0	1	2	3	4	5
q_0	Ø	$\{\epsilon\}$	$\{\epsilon\}$	$\{\epsilon\}$	$\{\epsilon\}$	$\{\epsilon\}$
q_1	Ø	Ø	{a}	{a}	$\{a, aba\}$	$\{a, aba\}$
<i>q</i> ₂	Ø	Ø	Ø	$\{ab\}$	{ <i>ab</i> }	$\{ab, abab\}$

Duality principle

We can extend the duality notion to fixpoints:

monotonemonotoneanti-monotoneanti-monotonepost-fixpointpre-fixpointleast fixpointgreatest fixpointincreasing chaindecreasing chain

Furthermore both Tarski's theorem and Kleene's theorem have a dual version (Tarski's theorem is its own dual).

On the topic of inductive reasoning...

Formalizing inductive definitions: Definition based on inference rules: $x \in \mathcal{X}$ Same property based on a least-fixpoint:

 $\frac{x \in \mathcal{X}}{x_0 \in \mathcal{X}} \qquad \frac{x \in \mathcal{X}}{f(x) \in \mathcal{X}} \qquad \qquad \mathsf{lfp}(Y \longmapsto \{x_0\} \cup Y \cup \{f(x) \mid x \in Y\})$

Proving the inclusion of a fixpoint in a given set:

- Let $\phi : S \longrightarrow S$ be a continuous operator
- Let $\mathcal{I} \in \mathcal{S}$ such that:

$$\forall x \in \mathcal{S}, \ x \sqsubseteq \mathcal{I} \Longrightarrow \phi(x) \sqsubseteq \mathcal{I}$$

- We obviously have $\bot \sqsubseteq \mathcal{I}$
- We can prove that ${\sf lfp}\,\phi\sqsubseteq \mathcal{I}$

Exercise: language of a grammar

Language of a grammar as a least-fixpoint

Assumptions:

- Alphabet \mathcal{A} , finite set of nodes \mathcal{N}
- Finite set of rules $\mathcal{R} \subseteq \mathcal{N} \times (\mathcal{A} \uplus \mathcal{N})^*$
- Starting node $S \in \mathcal{N}$

Questions:

- Define the set of words recognized by the grammar with inductive rules
- Do the same using a least-fixpoint

Hints:

- start with a function that maps each node into the set of words recognized by this node
- compute such a function by induction

Conclusion

Outline

Main points to remember

Foundations:

- program semantics: express program behaviors
- target semantic property: express proof goal
- conservative approximation usually required due to undecidability

Order relations:

- counterpart for logical implication (among other)
- will be pervasive in this course

Fixpoints and induction:

- encode general iteration
- will also be pervasive in this course

In the next lectures...

- Families of semantics, for a general model of programs
- Families of semantic properties of programs
- Verification techniques:
 - abstract interpretation based static analysis
 - machine assisted theorem proving
 - model checking

Next week: transition systems and operational semantics

Conclusion

Practical information about the course

The course will be taught by:

- Marc Chevalier (DIENS, TDs)
- Sylvain Conchon (LRI, Paris-Orsay, Model-Checking / SMT)
- Jérôme Feret (DIENS, Semantics, Abstract interpretation)
- Xavier Rival (DIENS, Semantics, Abstract interpretation, Coq)

Practical organization:

• 1h30 Cours + 2h00 TD or TP depending on week

Evaluation:

$$n=\frac{p+e}{2}$$

- project p: several projects will be proposed in a few weeks
- exam e: 31st of May, 2019

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