Static Analysis for Data Science

MPRI 2-6: Abstract Interpretation, Application to Verification and Static Analysis



Caterina Urban

February 7th, 2022

Year 2021-2022

Data Science is Everywhere

data is cheap and ubiquitous



mobile devices



Sensors



data science is revolutionizing industries



retai

- · personalized recommendations
- targeted marketing



- finance
- predictive models
- customized product offerings



- pharmaceutical
- predictive models
- patient selection



energy

- · exploration and discovery
 - accident prevention



manufacturing

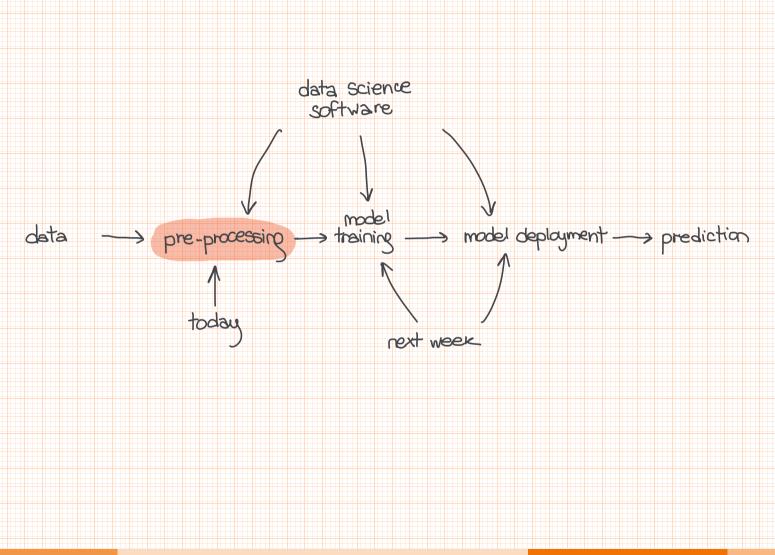
- · equipment failure predictions
- internet of things



- personalized treatments
- preventive care

Static Analysis for Data Science

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Ubiquitous Programming Errors

Julia

data science means programming



programming errors that do not cause failures can have serious consequences

Python





Anomalously Unused Data

The Reinhart-Rogoff Paper

American Economic Review: Papers & Proceedings 100 (May 2010): 573–578 http://www.aeaweb.org/articles.php?doi=10.1257/aer.100.2.573

Growth in a Time of Debt

By CARMEN M. REINHART AND KENNETH S. ROGOFF*

In this paper, we exploit a new multi-country historical dataset on public (government) debt to search for a systemic relationship between high public debt levels, growth and inflation.1 Our main result is that whereas the link between growth and debt seems relatively weak at "normal" debt levels, median growth rates for countries with public debt over roughly 90 percent of GDP are about one percent lower than otherwise; average (mean) growth rates are several percent lower. Surprisingly, the relationship between public debt and growth is remarkably similar across emerging markets and advanced economies. This is not the case for inflation. We find no systematic relationship between high debt levels and inflation for advanced economies as a group (albeit with individual country exceptions including the United States). By contrast, in emerging market countries, high public debt levels coincide with higher inflation.

Our topic would seem to be a timely one. Public debt has been soaring in the wake of the recent global financial maelstrom, especially in the epicenter countries. This should not be surprising, given the experience of earlier severe financial crises.² Outsized deficits and epic bank bailouts may be useful in fighting a downturn, but what is the long-run macroeconomic impact.

*Reinhart: Department of Economics, 4115 Tydings Hall, University of Maryland, College Park, MD 20742 (e-mail: creinhar@und.edu); Rogoff: Economics Department, 216 Littauer Center, Harvard University, Cambridge MA 02138-3001 (e-mail: Krogoff@harvard.edu). The authors would like to thank Olivier Jeanne and Vincent R. Reinhart for helpful comments.

¹ In this paper "public debt" refers to gross central government debt. "Domestic public debt" is government debt does not include debts carrying a government guarantee. Total gross external debt includes the external debts of all branches of government as well as private debt that is issued by domestic private entities under a foreign jurisdiction.

² Reinhart and Rogoff (2009a, b) demonstrate that the aftermath of a deep financial crisis typically involves a protracted period of macroeconomic adjustment, particularly in employment and housing prices. On average, public especially against the backdrop of graying populations and rising social insurance costs? Are sharply elevated public debts ultimately a manageable policy challenge?

Our approach here is decidedly empirical, taking advantage of a broad new historical dataset on public debt (in particular, central government debt) first presented in Carmen M. Reinhart and Kenneth S. Rogoff (2008, 2009b). Prior to this dataset, it was exceedingly difficult to get more than two or three decades of public debt data even for many rich countries, and virtually impossible for most emerging markets. Our results incorporate data on 44 countries spanning about 200 years. Taken together, the data incorporate over 3,700 annual observations covering a wide range of political systems, institutions, exchange rate and monetary arrangements, and historic circumstances.

We also employ more recent data on external debt, including debt owed both by governments and by private entities. For emerging markets, we find that there exists a significantly more severe threshold for total gross external debt (public and private)-which is almost exclusively denominated in a foreign currency-than for total public debt (the domestically issued component of which is largely denominated in home currency). When gross external debt reaches 60 percent of GDP, annual growth declines by about two percent; for levels of external debt in excess of 90 percent of GDP, growth rates are roughly cut in half. We are not in a position to calculate separate total external debt thresholds (as opposed to public debt thresholds) for advanced countries. The available time-series is too recent, beginning only in 2000. We do note, however, that external debt levels in advanced countries now average nearly 200 percent of GDP, with external debt levels being particularly high across Europe.

The focus of this paper is on the longer term macroeconomic implications of much higher public and external debt. The final section, how-

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The Reinhart-Rog

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0	В	C		J	K	L	М
Z		1.42.14	Real GDP growth				
3		-	Debt/GDP				
4	Country	Coverage	30 or less	30 to 60	60 to 90	90 or above 2	30 or less
26	and the second s		3.7	3.0	3.5	1.7	5.5
27	Minimum		1.6	0.3	1.3	-1.8	0.8
28	Maximum		5.4	4.9	10.2	3.6	13.3
29							
30	US	1946-2009	n.a.	3.4	3.3	-2.0	n.a.
31	UK	1946-2009	n.a.	2.4	2.5	2.4	n.a.
32	Sweden	1946-2009	3.6	2.9	2.7	п.а.	6.3
33	Spain	1946-2009	1.5	3.4	4.2	n.a.	9.9
34	Portugal	1952-2009	4.8	2.5	0.3	n.a.	7.9
35	New Zealand	1948-2009	2.5	2.9	3.9	-7.9	2.6
36	Netherlands	1956-2009	4.1	2.7	1.1	п.а.	6.4
37	Norway	1947-2009	3.4	5.1	n.a.	n.a.	5.4
38	Japan	1946-2009	7.0	4.0	1.0	0.7	7.0
39	Italy	1951-2009	5.4	2.1	1.8	1.0	5.6
40	Ireland	1948-2009	4.4	4.5	4.0	2.4	2.9
41	Greece	1970-2009	4.0	0.3	2.7	2.9	13.3
42	Germany	1946-2009	3.9	0.9	n.a.	n.a.	3.2
43	France	1949-2009	4.9	2.7	3.0	n.a.	5.2
44	Finland	1946-2009	3.8	2.4	5.5	n.a.	7.0
45	Denmark	1950-2009	3.5	1.7	2.4	n.a.	5.6
46	Canada	1951-2009	1.9	3.6	4.1	n.a.	2.2
47	Belgium	1947-2009	n.a.	4.2		2.6	n.a.
48	Austria	1948-2009	52	3.3	-3.8	п.а.	5.7
49	Australia	1951-2009	3.2	4.9	4.0	n.a.	5.9
50		/					
51			4.1	2.8	2.8	=AVERAGE(L30:L44)	

data excluded from the analysis

The Reinhart-Rog

FAQ: Reinhart, Rogoff, and the Excel Error That Changed History

By Peter Coy y April 18, 2013

The Excel Depression

By PAUL KRUGMAN Published: April 18, 2013 470 Comments

In this age of information, math errors can lead to disaster. NASA's Mars Orbiter crashed because engineers forgot to convert to metric measurements; JPMorgan Chase's "London Whale" venture went bad in part because modelers divided by a sum instead of an average. So, did an Excel coding error destroy the economies of the Western world?

Country

Minimum

Coverage

30 or less 30 to 60

1.6

60 to 90

0.3

3.5

1.3

f	FACEBOOK
y	TWITTER
Q +	GOOGLE+
	SAVE
\boxtimes	EMAIL
+	SHARE
₿	PRINT
ē	REPRINTS

M

5.5

0.8

13.3

n.a.

n.a.

6.3

9.9 7.9

90 or above 30 or less

1.8

3.6

-2.0

2.4

D.8

n.a.

n.a.



The story so far: At the beginning of 2010, two Harvard economists, Carmen Reinhart and Kenneth Rogoff, circulated a paper, "Growth

in a Time of Debt," that purported to identify a critical "threshold," a tipping point, for government indebtedness. Once debt exceeds 90 percent of gross domestic product, they claimed, economic growth drops off sharply.

Ms. Reinhart and Mr. Rogoff had credibility thanks to a widely admired earlier book on the history of financial

England Covid-19 Cases Error

figures have been contacted

Elisabeth Mahase

Details of nearly 16 000 cases of covid-19 were not

error in the process for updating the data.

reached by contact tracers.

transferred to England's NHS Test and Trace service

and were missed from official figures because of an

England's health and social care secretary, Matt

Hancock, told the House of Commons on Monday 5

October that after the error was discovered on Friday

2 October "6500 hours of extra contact tracing" had

morning only half (51%) of the people had been

In response, Labour's shadow health secretary,

bliesfully unawate they have been exposed to covid,

ding this deadly virus at a time when

US & WORLD \ TECH SCIENCE

Excel spreadsheet error blamed for UK's 16,000 missing coronavirus cases

The case went missing after the spreadsheet hit its filesize limit

By James Vincent | Oct 5, 2020, 9:41am EDT

Covid-19: Only half of 16 000 patients missed from England's official data and furthermore have issued guidance on

NEWS

validation and risk management for these products if they are to be used in such a safety critical manner." The error came as the Labour Party's leader, Keir Starmer, said that the prime minister had "lost control" of covid-19, with no clear strategy for beating it. Speaking to the Observer, Starmer set out his five point plan for covid-19, which starts with publishing the criteria for local restrictions, as the German government did. Secondly, he said public health messaging should be improved by adding a feature to the NHS covid-19 app so people can search their been carried out over the weekend. But as at Monday postcode and find out their local restrictions. Jonathan Ashworth, said, "Thousands of people are

Starmer has also said he would fix the contact tracing system by investing in NHS and university laboratories to expand testing and at the same time put local public health teams in charge of contact the their areas. Routine regular testing in high

Cite this as: BMJ 2020;371:m3891 http://dx.doi.org/10.1136/bmj.m3891 Published: 06 October 2020

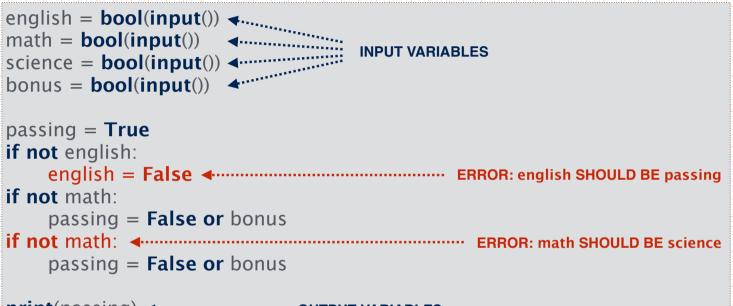
Check for updates

10.1136/bmj.m3891 on 6 October 2020. Downloa

BMJ: first published &

as

Example



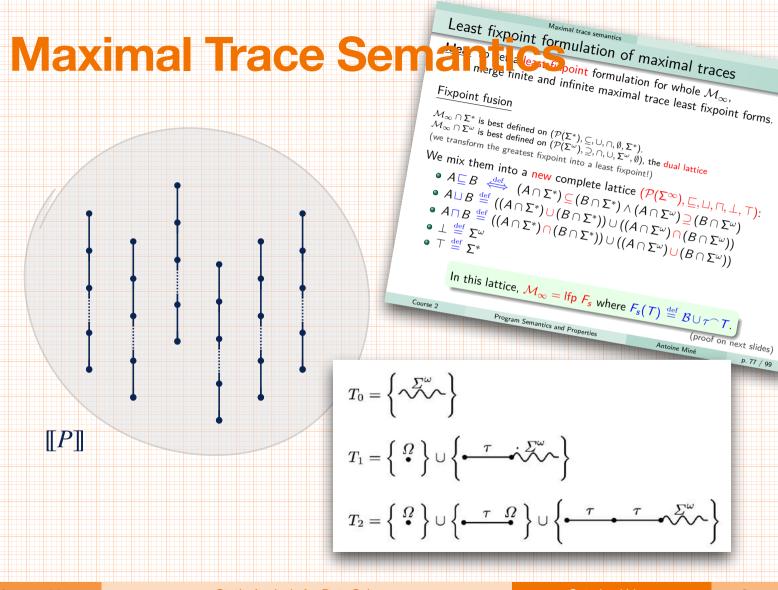
print(passing) <----- OUTPUT VARIABLES

the input variables english and science are unused

Lesson 14

Static Analysis for Data Science

Caterina Urban



Maximal Trace Semantics

 passing = True
 p

 if not english:
 english = False

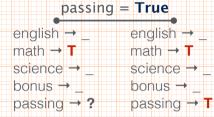
 english = False
 english =

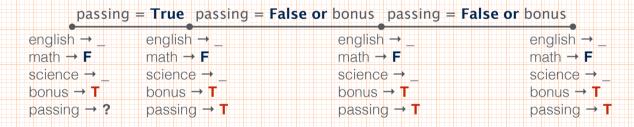
 if not math:
 math →

 passing = False or bonus
 science

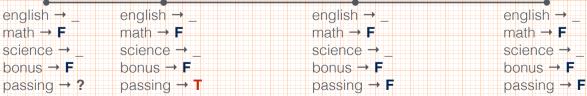
 if not math:
 bonus →

 passing = False or bonus
 passing









Lesson 14

_ set of all input variables of program P

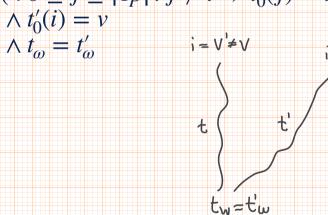
$\mathcal{N}_{J} \stackrel{\mathrm{def}}{=} \{ \llbracket P \rrbracket \in \mathscr{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} : \mathsf{UNUSED}_{i}(\llbracket P \rrbracket) \}$

 \mathcal{N}_J is the set of all programs P (or, rather, their semantics $[\![P]\!]$) that **do not use** the value of the input variables in $J \subseteq I_P$

$\mathcal{N}_{J} \stackrel{\mathsf{def}}{=} \{ \llbracket P \rrbracket \in \mathscr{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} \colon \mathsf{UNUSED}_{i}(\llbracket P \rrbracket) \}$

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 $\begin{array}{l} \text{UNUSED}_{i}(\llbracket M \rrbracket) \stackrel{\text{def}}{=} \forall t \in \llbracket P \rrbracket, v \in \mathscr{V}: t_{0}(i) \neq v \Rightarrow \exists t' \in \llbracket P \rrbracket: \\ (\forall 0 \leq j \leq |I_{P}|: j \neq i \Rightarrow t_{0}(j) = t_{0}'(j)) \end{array}$



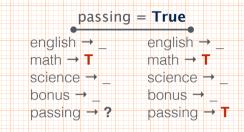
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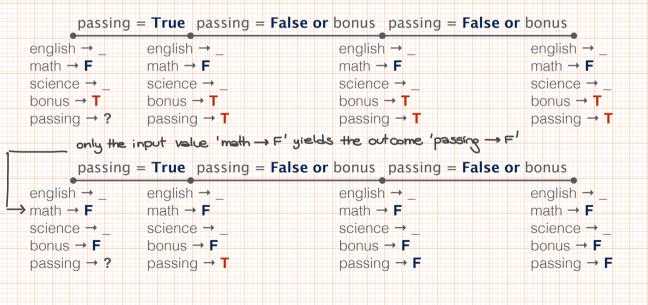
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Intuitively: any possible program outcome is possible from any value of the input variable i

passing = True if not english: english = False if not math: passing = False or bonus if not math: passing = False or bonus





Lesson 14

Static Analysis for Data Science

|P|

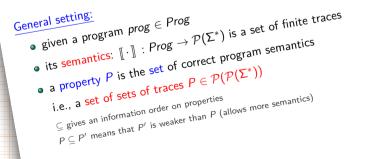
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 $\wedge t_{\omega} = t'_{\omega}$

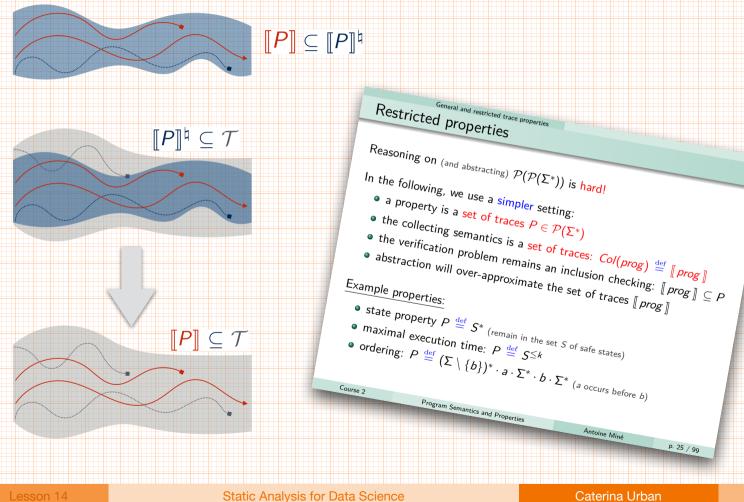
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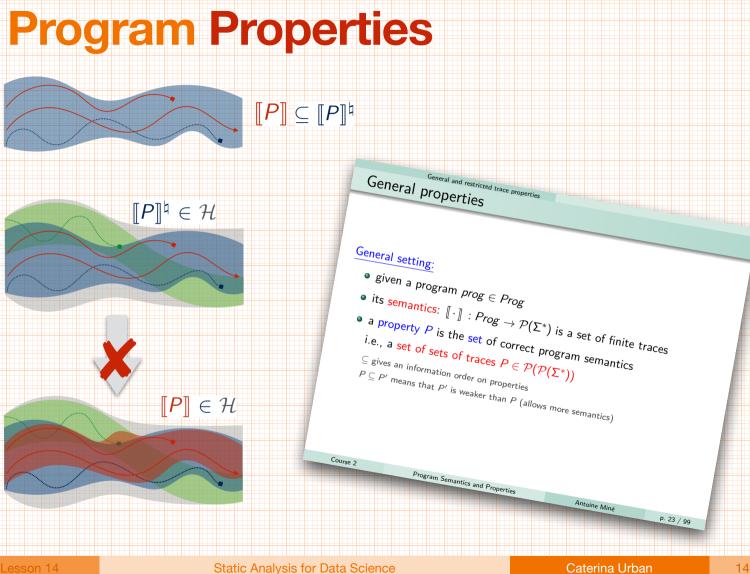
Intuitively: any possible program outcome is possible from any value of the input variable i General and restricted trace properties



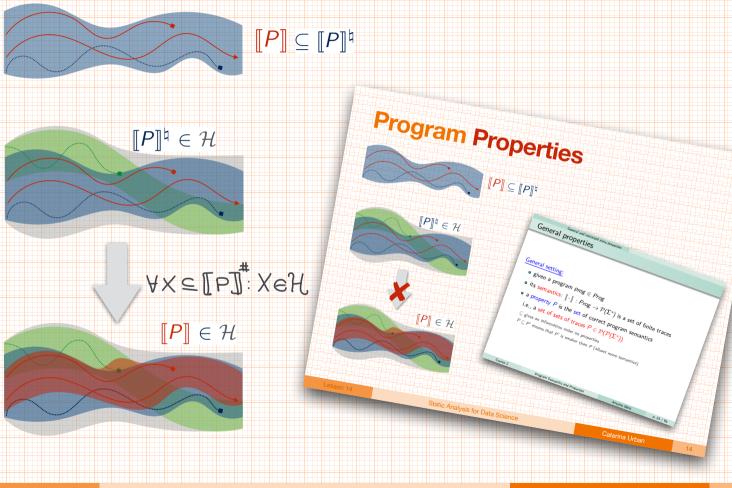
and Properties

Trace Properties



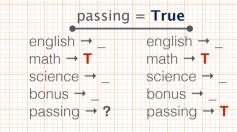


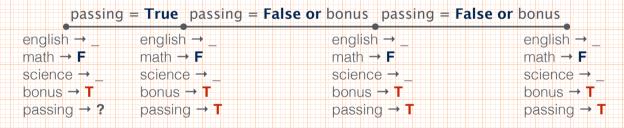
Subset-Closed Properties



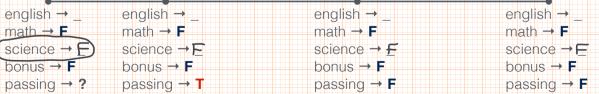
Lesson 14

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Lesson 14

Static Analysis for Data Science

NOTA

subset-closed

|P|

$\mathcal{N}_{J} \stackrel{\mathsf{def}}{=} \{ \llbracket P \rrbracket \in \mathscr{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} \colon \mathsf{UNUSED}_{i}(\llbracket P \rrbracket) \}$

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Theorem

$P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket\} \subseteq \mathcal{N}_J$

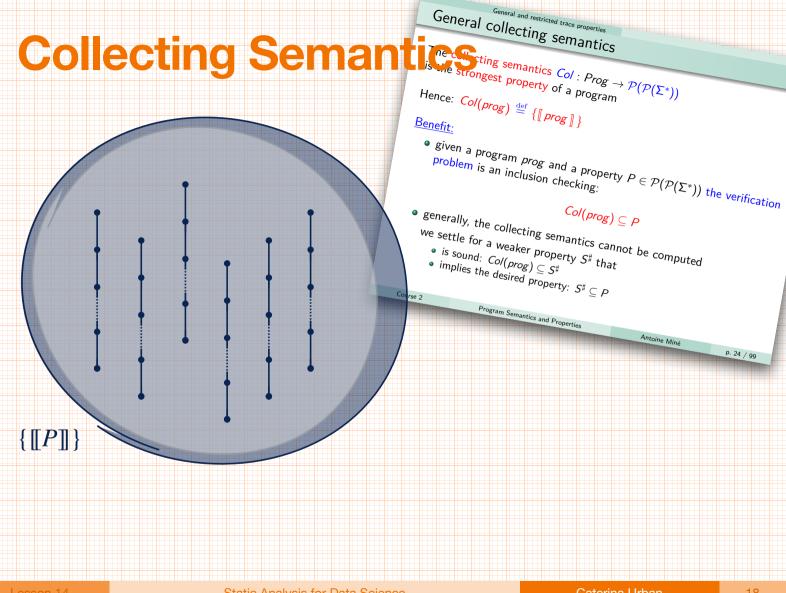
General and restricted trace properties

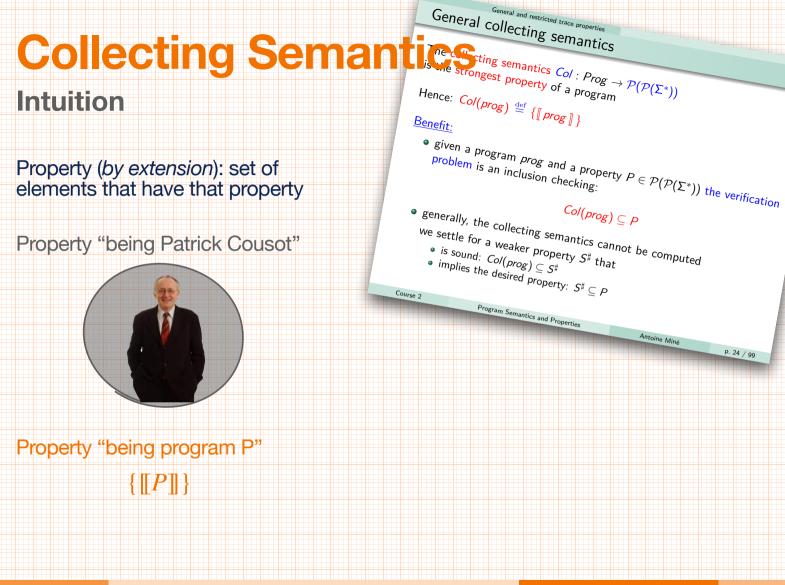
General setting: • given a program $prog \in Prog$ • its semantics: $[\cdot] : Prog \rightarrow \mathcal{P}(\Sigma^*)$ is a set of finite traces • a property \mathcal{P} is the set of correct program semantics • a property \mathcal{P} is the set of correct program semantics • i.e., a set of sets of traces $\mathcal{P} \in \mathcal{P}(\mathcal{P}(\Sigma^*))$ • i.e., a set of sets of traces $\mathcal{P} \in \mathcal{P}(\mathcal{P}(\Sigma^*))$ • gives an information order on properties • $\mathcal{P} \subseteq \mathcal{P}'$ means that \mathcal{P}' is weaker than \mathcal{P} (allows more semantics)

and Properties

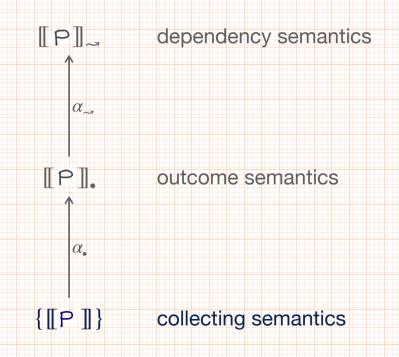
Antoine Miné

Lesson 14



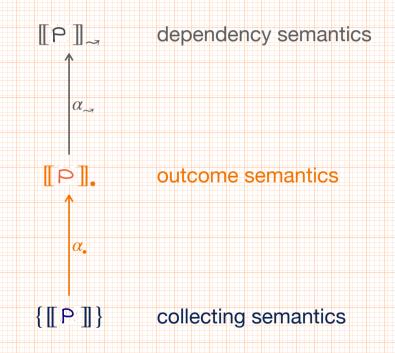


(Another) Hierarchy of Semantics



Lesson 14

(Another) Hierarchy of Semantics



Lesson 14

partitioning a set of traces that satisfies input data (non-)usage with respect to the program outcome yields sets of traces that also satisfy input data (non-)usage

Lesson 14

[[P]].

 $\mathbb{O} \stackrel{\text{def}}{=} \{ \begin{array}{c} \sum_{o_1=v_1,\ldots,o_k=v_k}^{+} \mid v_1,\ldots,v_k \in \mathcal{V} \} \cup \{ \Sigma^{\omega} \} \\ \text{outcomes} \end{array} \}$

Lemma

$P \models \mathcal{N}_{I} \Leftrightarrow \{ \llbracket P \rrbracket \cap O \mid O \in \mathbb{O} \} \subseteq \mathcal{N}_{I}$

input data (non-) usage can be decided independently for each possible outcome

$$\mathbb{O} \stackrel{\mathsf{def}}{=} \{\Sigma_{o_1=v_1,\ldots,o_k=v_k}^+ \mid v_1,\ldots,v_k \in \mathcal{V}\} \cup \{\Sigma^{\omega}\}$$

Lemma

 $P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket \cap O \mid O \in \mathbb{O}\} \subseteq \mathcal{N}_J$

 $\langle \mathscr{P}(\mathscr{P}(\Sigma^{+\infty})), \subseteq \rangle \qquad \langle \mathscr{P}(\mathscr{P}(\Sigma^{+\infty})), \subseteq, \rangle$

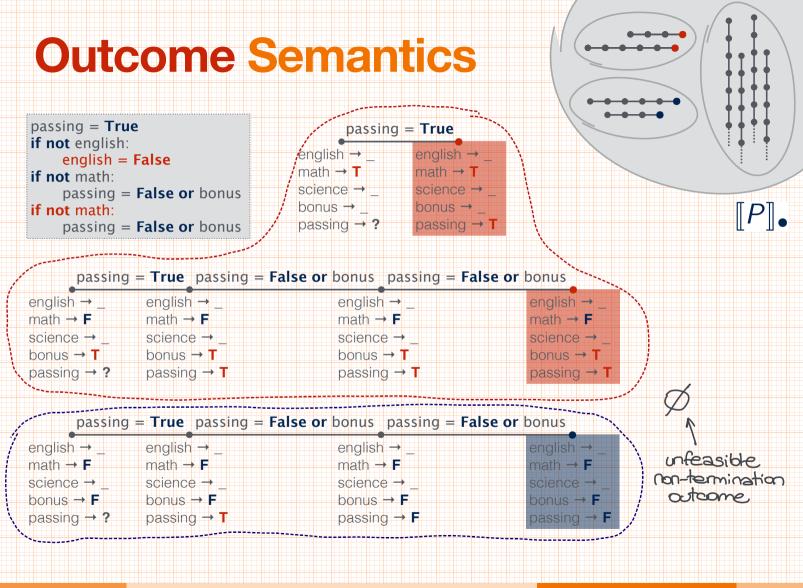
α.

 $\alpha_{\bullet}(S) \stackrel{\mathsf{def}}{=} \{ T \cap O \mid T \in S \land O \in \mathbb{O} \}$

outcome abstraction

outcomes

γ.



Lesson 14

$$\mathbb{O} \stackrel{\mathsf{def}}{=} \{\Sigma_{o_1=v_1,\ldots,o_k=v_k}^+ \mid v_1,\ldots,v_k \in \mathcal{V}\} \cup \{\Sigma^{\omega}\}$$

Lemma

 $P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket \cap O \mid O \in \mathbb{O}\} \subseteq \mathcal{N}_J$

 $\langle \mathscr{P}(\mathscr{P}(\Sigma^{+\infty})), \subseteq \rangle \qquad \langle \mathscr{P}(\mathscr{P}(\Sigma^{+\infty})), \subseteq, \rangle$

 α

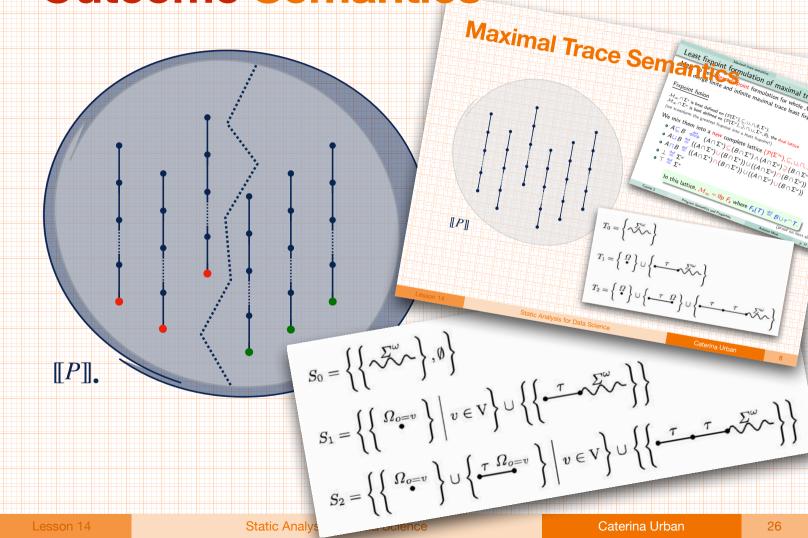
 $\alpha_{\bullet}(S) \stackrel{\mathsf{def}}{=} \{ T \cap O \mid T \in S \land O \in \mathbb{O} \}$

outcome abstraction

outcomes

 $\llbracket P \rrbracket_{\bullet} \stackrel{\mathsf{def}}{=} \alpha_{\bullet}(\{\llbracket P \rrbracket\}) = \{\llbracket P \rrbracket \cap O \mid O \in \mathbb{O}\}$

Lesson 14



$$S_{1} \subseteq S_{2} \stackrel{\text{def}}{=} \bigwedge_{v_{1},...,v_{k} \in V} S_{1}^{+}_{o_{1}=v_{1},...,o_{k}=v_{k}} \subseteq S_{2}^{+}_{o_{1}=v_{1},...,o_{k}=v_{k}} \wedge S_{1}^{\omega} \supseteq S_{2}^{\omega}$$

$$Theorem 1. The outcome semantics $\Lambda \in \mathcal{P}(\mathcal{P}(\mathcal{E}^{+\infty}))$ can be expressed as a least fixpoint in $(\mathcal{P}(\mathcal{P}(\mathcal{E}^{+\infty})), \Box, \sqcup, \sqcap, \{\mathcal{E}^{\omega}, \emptyset\}, \{\emptyset, \mathcal{E}^{+}\})$ as:

$$\Lambda_{\bullet} = \mathrm{lfp}^{\square} \Theta_{\bullet}$$

$$\Theta(S) \stackrel{\text{def}}{=} \{\Omega_{o_{1}=v_{1},...,o_{k}=v_{k}} \mid v_{1},...,v_{k} \in V\} \cup \{\tau ; T \mid T \in S\}$$

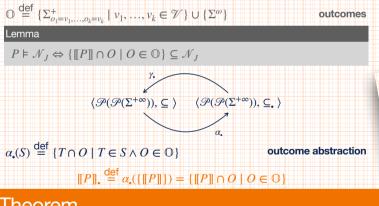
$$(proof by Kleenian Fixpoint Transfer [Urban18])$$

$$f_{0} = \{\{\mathcal{A}_{0} \stackrel{\omega}{=} \psi\}, \psi \in V\} \cup \{\mathcal{A}_{0} \stackrel{\omega}$$$$

$\mathcal{N}_{I} \stackrel{\mathsf{def}}{=} \{ \llbracket P \rrbracket \in \mathscr{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} \colon \mathsf{UNUSED}_{i}(\llbracket P \rrbracket) \}$

 \mathcal{N}_{I} is the set of all programs P (or, rather, their semantics [P]) that **do not use** the value of the input variables in $J \subseteq I_P$

Outcome Semantics



 $_{0}(i) \neq v \Rightarrow \exists t' \in \llbracket P \rrbracket$: $i \Rightarrow t_0(j) = t'_0(j)$

of the IVP

Theorem

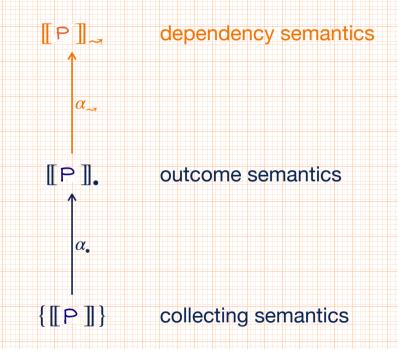
Input Data (Non-)Usage $\mathcal{N}_{J} \stackrel{\text{def}}{=} \{ [\![P]\!] \in \mathcal{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} : \text{UNUSED}([\![P]\!]) \}$ \mathcal{N}_{J} is the set of all programs P (or, rather, their semantics [[P]]) that do not use the value of the input variables in $J \subseteq I_P$ $\mathsf{VSED}_{i}(\llbracket M \rrbracket) \stackrel{\text{def}}{=} \forall t \in \llbracket P \rrbracket, v \in \mathcal{V} : t_{0}(i) \neq v \Rightarrow \exists t' \in \llbracket P \rrbracket: \\ (\forall 0 \leq j \leq |I_{P}| : j \neq i \Rightarrow t_{0}(j) = t'_{0}(j))$

 $P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket\} \subseteq \mathcal{N}_J$

Theorem

$P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket\} \subseteq \mathcal{N}_J \Leftrightarrow \alpha_{\bullet}(\{\llbracket P \rrbracket\}) \subseteq \mathcal{N}_J \Leftrightarrow \llbracket P \rrbracket_{\bullet} \subseteq \mathcal{N}_J$

(Another) Hierarchy of Semantics



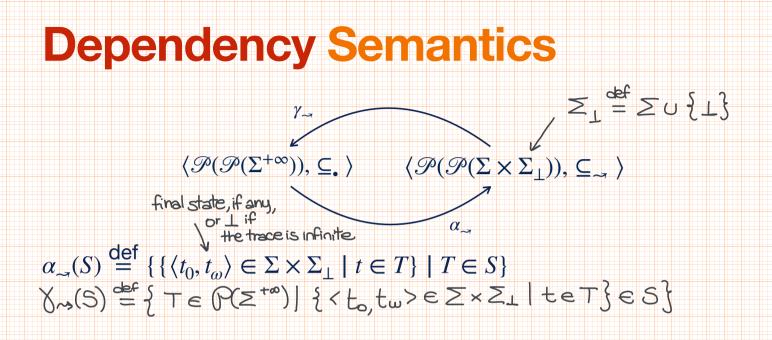
Lesson 14

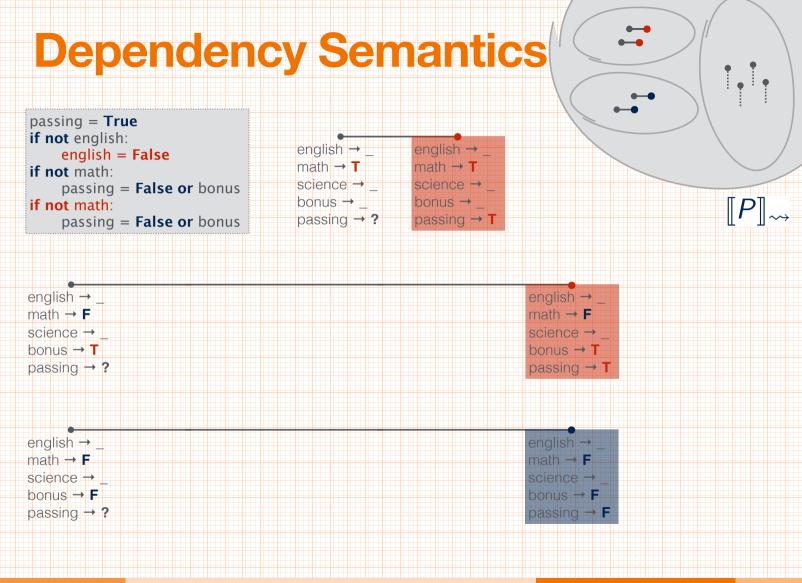
Dependency Semantics

to reason about input data (non-)usage we do not need to consider all intermediate computations between the initial and final states of a trace (if any)

Lesson 14

[[P]]~





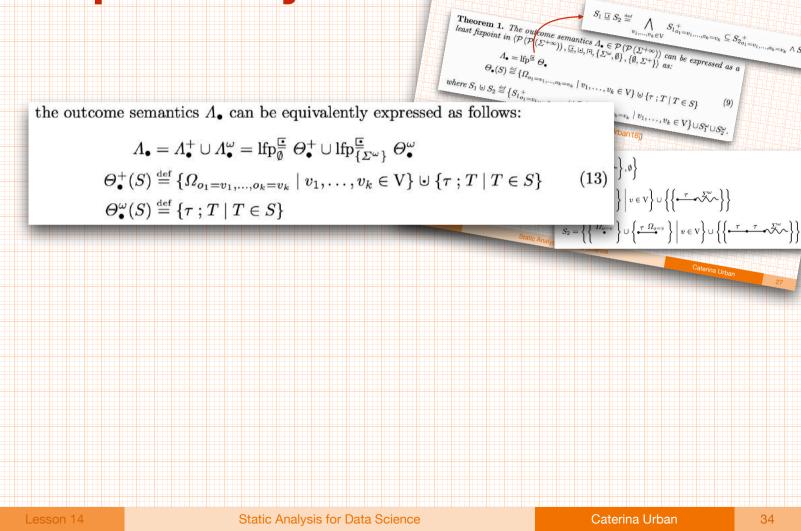
$\langle \mathscr{P}(\mathscr{P}(\Sigma^{+\infty})), \subseteq, \rangle \qquad \langle \mathscr{P}(\mathscr{P}(\Sigma \times \Sigma_{\perp})), \subseteq, \rangle$

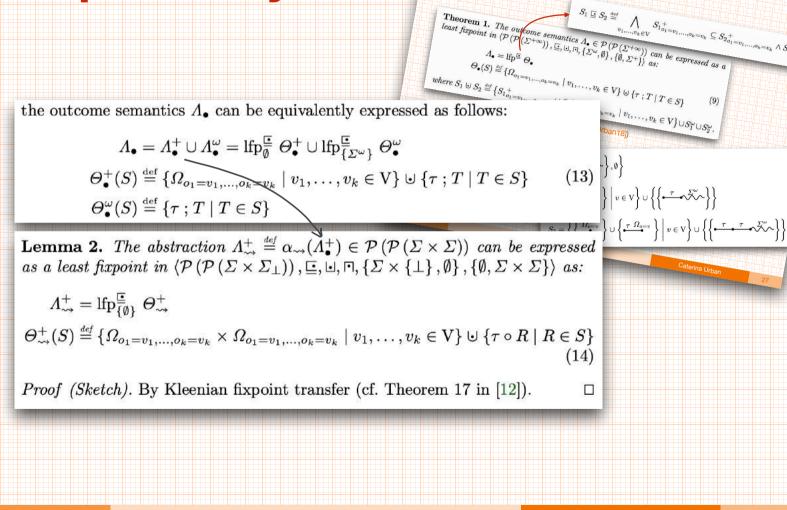
 α

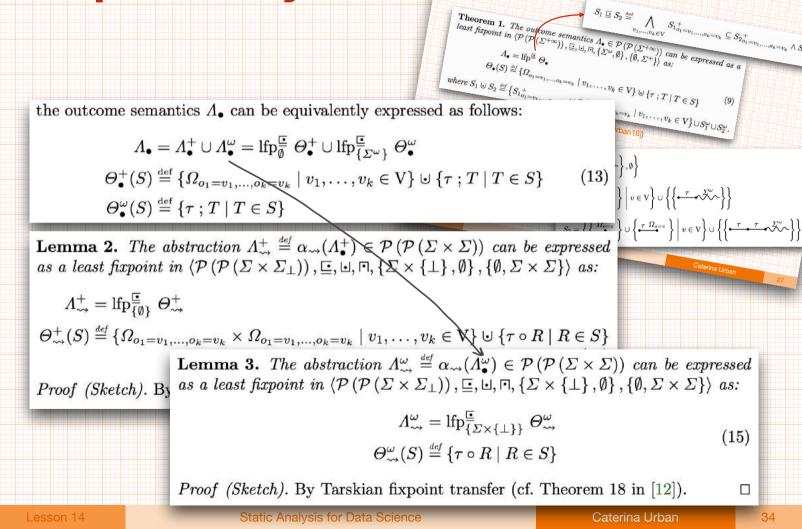
$$\begin{split} \alpha_{\prec}(S) &\stackrel{\mathsf{def}}{=} \{ \{ \langle t_0, t_{\omega} \rangle \in \Sigma \times \Sigma_{\perp} \mid t \in T \} \mid T \in S \} \\ \gamma_{\prec}(S) &\stackrel{\mathsf{def}}{=} \{ T \in \mathscr{P}(\Sigma^{+\infty}) \mid \{ \langle t_0, t_{\omega} \rangle \in \Sigma \times \Sigma_{\perp} \mid t \in T \} \in S \} \end{split}$$

 $\llbracket P \rrbracket_{\prec} \stackrel{\mathsf{def}}{=} \alpha_{\prec}(\llbracket P \rrbracket_{\bullet}) = \{ \{ \langle t_0, t_\omega \rangle \in \Sigma \times \Sigma \mid t \in \llbracket P \rrbracket \cap O \} \mid O \in \mathbb{O} \}$

Lesson 14







Lemma 2. The abstraction $\Lambda^+_{\rightarrow \rightarrow} \stackrel{\text{def}}{=} \alpha_{\rightarrow \rightarrow}(\Lambda^+_{\bullet}) \in \mathcal{P}(\mathcal{P}(\Sigma \times \Sigma))$ can be expressed as a least fixpoint in $\langle \mathcal{P}(\mathcal{P}(\Sigma \times \Sigma_{\perp})), \sqsubseteq, \lor, \neg, \{\Sigma \times \{\bot\}, \emptyset\}, \{\emptyset, \Sigma \times \Sigma\} \rangle$ as:

$$\Theta_{\rightarrow}^+(S) \stackrel{\text{\tiny def}}{=} \{\Omega_{o_1=v_1,\dots,o_k=v_k} \mid v_1,\dots,v_k \in \mathbf{V}\} \cup \{\tau \circ R \mid R \in S\}$$

Proof (Sketch). By as a least fixpoint in $\langle \mathcal{P}(\mathcal{P}(\Sigma \times \Sigma_{\perp})), \subseteq, \sqcup, \sqcap, \{\Sigma \times \{\bot\}, \emptyset\}, \{\emptyset, \Sigma \times \Sigma\} \rangle$ as:

$$\Lambda^{\omega}_{\rightarrow } = \operatorname{lfp}_{\{\Sigma \times \{\bot\}\}}^{\sqsubseteq} \Theta^{\omega}_{\rightarrow}
\Theta^{\omega}_{\rightarrow }(S) \stackrel{\scriptscriptstyle def}{=} \{\tau \circ R \mid R \in S\}$$
(15)

Proof (Sketch). By Tarskian fixpoint transfer (cf. Theorem 18 in [12]).

Theorem 3. The dependency semantics $\Lambda_{\rightarrow} \in \mathcal{P}\left(\mathcal{P}\left(\Sigma \times \Sigma_{\perp}\right)\right)$ can be expressed as a least fixpoint in $\langle \mathcal{P}\left(\mathcal{P}\left(\Sigma \times \Sigma_{\perp}\right)\right), \boxdot, \lor, \neg, \{\Sigma \times \{\bot\}, \emptyset\}, \{\emptyset, \Sigma \times \Sigma\}\rangle$ as: $\Lambda_{\rightarrow} = \Lambda_{\rightarrow}^{+} \cup \Lambda_{\rightarrow}^{\omega} = \operatorname{lfp}_{\{\Sigma \times \{\bot\}, \emptyset\}}^{\boxdot} \Theta_{\rightarrow}$ $\Theta_{\rightarrow}(S) \stackrel{\text{def}}{=} \{\Omega_{o_{1}=v_{1},...,o_{k}=v_{k}} \times \Omega_{o_{1}=v_{1},...,o_{k}=v_{k}} \mid v_{1},\ldots,v_{k} \in \mathcal{V}\} \cup \{\tau \circ R \mid R \in S\}$ (16)

Proof (Sketch). The proof follows immediately from Lemma 2 and Lemma 3. \Box

 $\Lambda^+_{\infty} = \operatorname{lfp}_{(A)}^{\textcircled{\bullet}} \Theta^+_{\infty}$

Input Data (Non-)Usage

$\mathcal{N}_{I} \stackrel{\mathsf{def}}{=} \{ \llbracket P \rrbracket \in \mathscr{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} \colon \mathsf{UNUSED}_{i}(\llbracket P \rrbracket) \}$

 \mathcal{N}_{I} is the set of all programs P (or, rather, their semantics [P]) that **do not use** the value of the input variables in $J \subseteq I_P$

Dependency Semantics

 $\langle \mathscr{P}(\mathscr{P}(\Sigma^{+\infty})), \subseteq \rangle \qquad \langle \mathscr{P}(\mathscr{P}(\Sigma \times \Sigma_{\perp})), \subseteq \rangle$

 $\alpha_{\sim}(S) \stackrel{\text{def}}{=} \{ \{ \langle t_0, t_{\omega} \rangle \in \Sigma \times \Sigma_1 \mid t \in T \} \mid T \in S \}$ $\gamma_{\prec}(S) \stackrel{\mathsf{def}}{=} \{ T \in \mathscr{P}(\Sigma^{+\infty}) \mid \{ \langle t_0, t_{\omega} \rangle \in \Sigma \times \Sigma_1 \mid t \in T \} \in S \}$

 $\llbracket P \rrbracket_{\sim} \stackrel{\text{def}}{=} \alpha_{\sim}(\llbracket P \rrbracket_{\bullet}) = \{\{\langle t_0, t_{\omega} \rangle \in \Sigma \times \Sigma \mid t \in \llbracket P \rrbracket \cap O\} \mid O \in \mathbb{O}\}$

 $_{0}(i) \neq v \Rightarrow \exists t' \in \llbracket P \rrbracket$: $i \Rightarrow t_0(j) = t'_0(j)$

Input Data (Non-)Usage $\mathcal{N}_{J} \stackrel{\text{def}}{=} \{ [\![P]\!] \in \mathcal{P}(\Sigma^{+\infty}) \mid \forall i \in J \subseteq I_{P} : \text{UNUSED}_{i}([\![P]\!]) \}$ \mathcal{N}_{J} is the set of all programs P (or, rather, their semantics [[P]]) that do not use the value of the input variables in $J \subseteq I_P$ $\begin{array}{l} \underset{i \Rightarrow t_0(j) = t_0(j)}{\overset{0(i) \neq v \Rightarrow \exists t' \in \llbracket P \rrbracket:} \\ i \Rightarrow t_0(j) = t_0(j) \end{array}$ come Semantics outcomes

Theorem

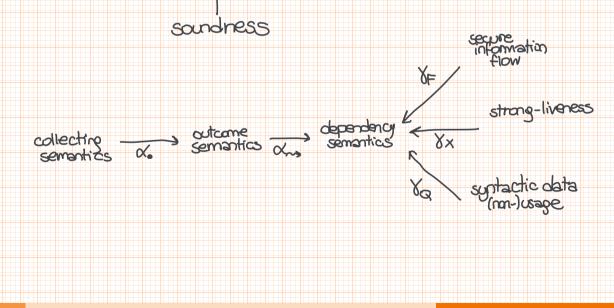
$P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket\} \subseteq \mathcal{N}_J \Leftrightarrow \llbracket P \rrbracket_{\bullet} \subseteq \mathcal{N}_J \Leftrightarrow \gamma_{\prec}(\llbracket P \rrbracket_{\sim}) \subseteq \mathcal{N}_J$ $P \models \mathcal{N}_J \Leftrightarrow \{\llbracket P \rrbracket\} \subseteq \mathcal{N}_J \Leftrightarrow \alpha_*(\{\llbracket P \rrbracket\}) \subseteq \mathcal{N}_J \Leftrightarrow \llbracket P \rrbracket_* \subseteq \mathcal{N}_J$

Input Data (Non-)Usage Abstractions

Over-Approximation of the Used Input Data

⇒ Under-Approximation of the Unused Input Data





I ---→ X

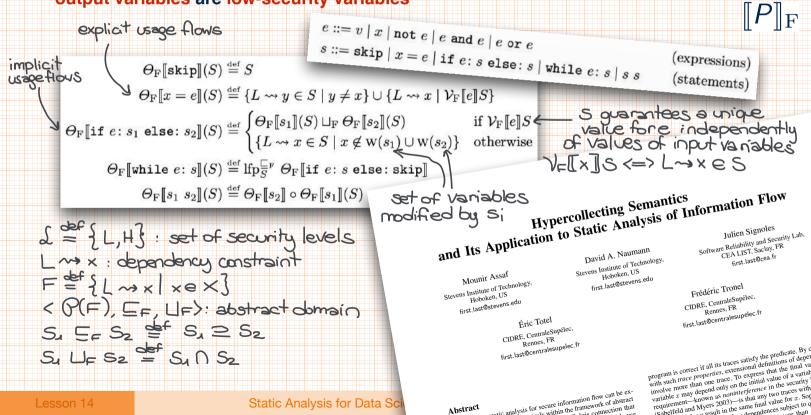
H ... + t

L ... > Z

H ... > W

---+ V

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains all input variables:
 input variables are high-security variables
 output variables are low-security variables



I ...→ X

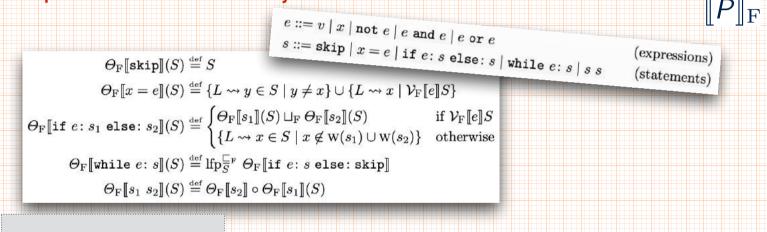
L ...→ Y

H → t

H ... → w

L ... > Z

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains all input variables:
input variables are high-security variables
output variables are low-security variables



passing = True
if not english:
 english = False
if not math:
 passing = False or bonus
if not math:
 passing = False or bonus

L …→ v

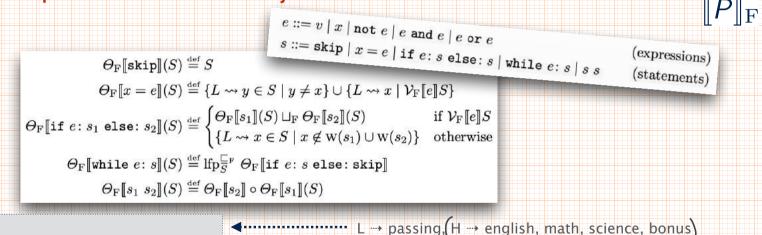
I ...→ X

H ---→ t

L ··· > Z

H ... > W

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains all input variables:
input variables are high-security variables
output variables are low-security variables



passing = True
if not english:
 english = False
if not math:
 passing = False or bonus
if not math:
 passing = False or bonus

I ...→ X

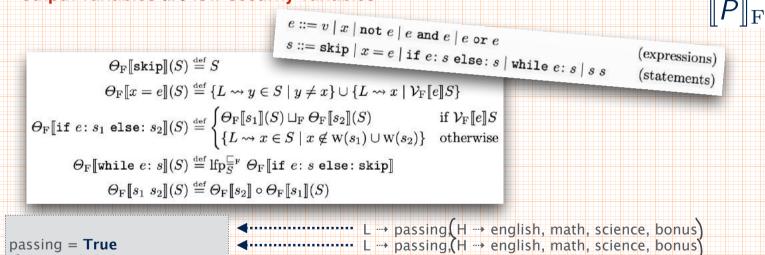
H ---→ t

L ⊶ y

L ···→ z

H ... > W

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains all input variables:
input variables are high-security variables
output variables are low-security variables



passing = True if not english: english = False if not math: passing = False or bonus if not math: passing = False or bonus

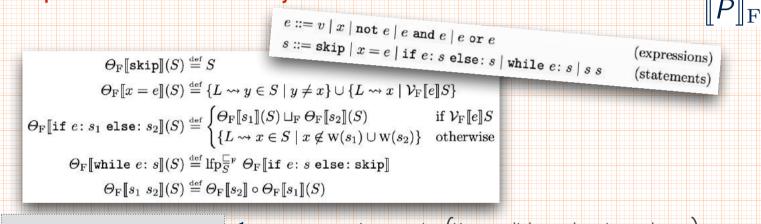
I ...→ X

L ...→ y

H ... > W

L ···→ z

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains *all* input variables:
input variables are high-security variables
output variables are low-security variables



passing = True
if not english:
 english = False
if not math:
 passing = False or bonus
if not math:
 passing = False or bonus

Lesson 14

I ...→ X

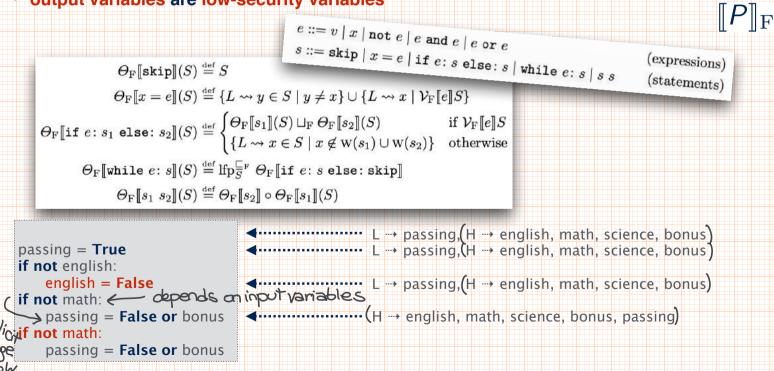
L …→ y

H ... → W

H ...→ t

L ... > Z

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains all input variables:
input variables are high-security variables
output variables are low-security variables



Lesson 14

L ...→ V

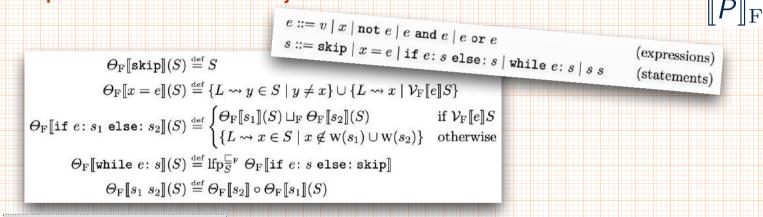
| ...→ x

H ...→ t

H ... > W

L ... > Z

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains all input variables:
input variables are high-security variables
output variables are low-security variables



Lesson 14

I ...→ X

H ...→ t

L ...→ A

H ... > W

(expressions) (statements)

L ...→ Z

 $\|P\|_{\mathrm{F}}$

possibilistic non-interference coincides with input data (non-)usage when the set J of unused input variables contains *all* input variables:

- input variables are high-security variables
- output variables are low-security variables

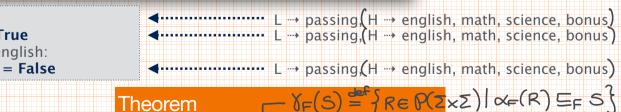
and the program is terminating

$$\begin{array}{l} \Theta_{\mathrm{F}}[\![\mathrm{skip}]\!](S) \stackrel{\mathrm{def}}{=} S \\ \Theta_{\mathrm{F}}[\![\mathrm{skip}]\!](S) \stackrel{\mathrm{def}}{=} S \\ \Theta_{\mathrm{F}}[\![x = e]\!](S) \stackrel{\mathrm{def}}{=} \{L \rightsquigarrow y \in S \mid y \neq x\} \cup \{L \rightsquigarrow x \mid \mathcal{V}_{\mathrm{F}}[\![e]\!]S\} \\ \Theta_{\mathrm{F}}[\![\mathrm{if}\ e \colon s_{1}\ \mathrm{else}\colon s_{2}]\!](S) \stackrel{\mathrm{def}}{=} \begin{cases} \Theta_{\mathrm{F}}[\![s_{1}]\!](S) \sqcup_{\mathrm{F}} \Theta_{\mathrm{F}}[\![s_{2}]\!](S) & \mathrm{if}\ \mathcal{V}_{\mathrm{F}}[\![e]\!]S \\ \{L \rightsquigarrow x \in S \mid x \notin \mathrm{w}(s_{1}) \cup \mathrm{w}(s_{2})\} & \mathrm{otherwise} \end{cases} \end{array}$$

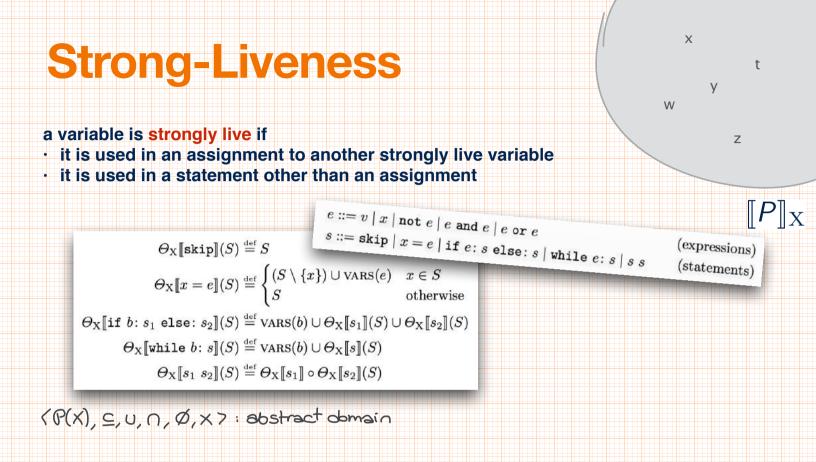
P ··· al

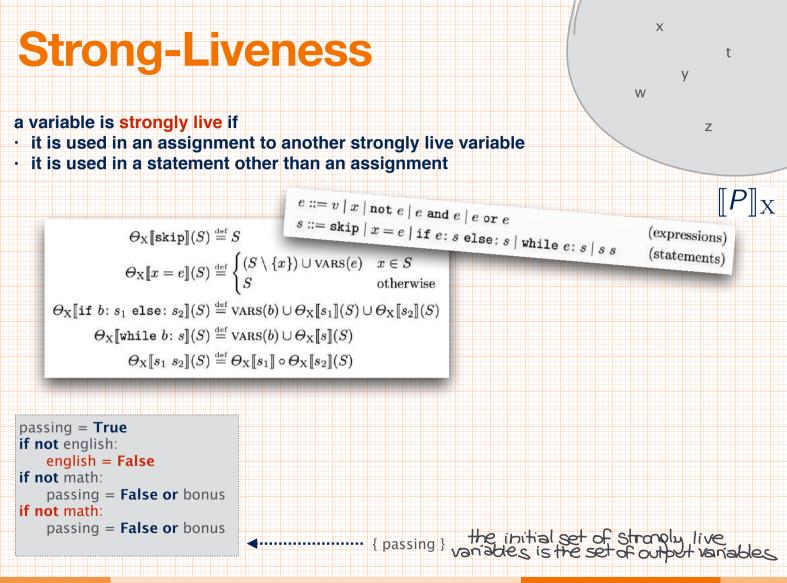
 $\Theta_{\mathrm{F}}\llbracket \texttt{while } e \colon s \rrbracket(S) \stackrel{\text{def}}{=} \mathrm{lfp}_{S}^{\sqsubseteq_{\mathrm{F}}} \Theta_{\mathrm{F}}\llbracket \texttt{if } e \colon s \texttt{ else} \colon \texttt{skip} \rrbracket$ $\Theta_{\mathrm{F}}\llbracket s_{1} \ s_{2} \rrbracket(S) \stackrel{\text{def}}{=} \Theta_{\mathrm{F}}\llbracket s_{2} \rrbracket \circ \Theta_{\mathrm{F}}\llbracket s_{1} \rrbracket(S)$

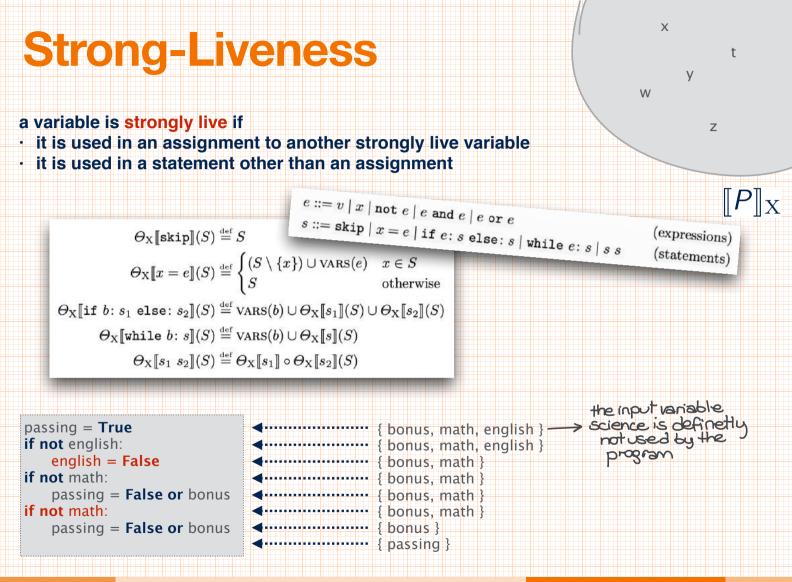
passing = **True** while not english: english = **False**

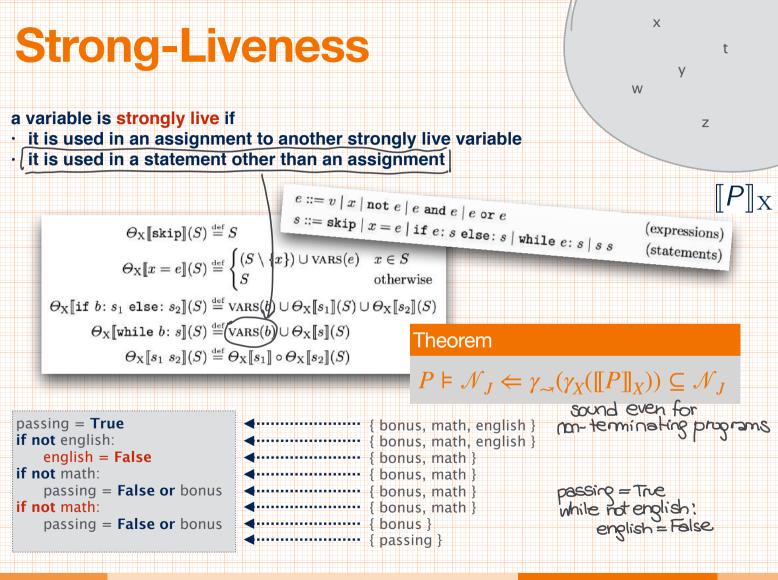


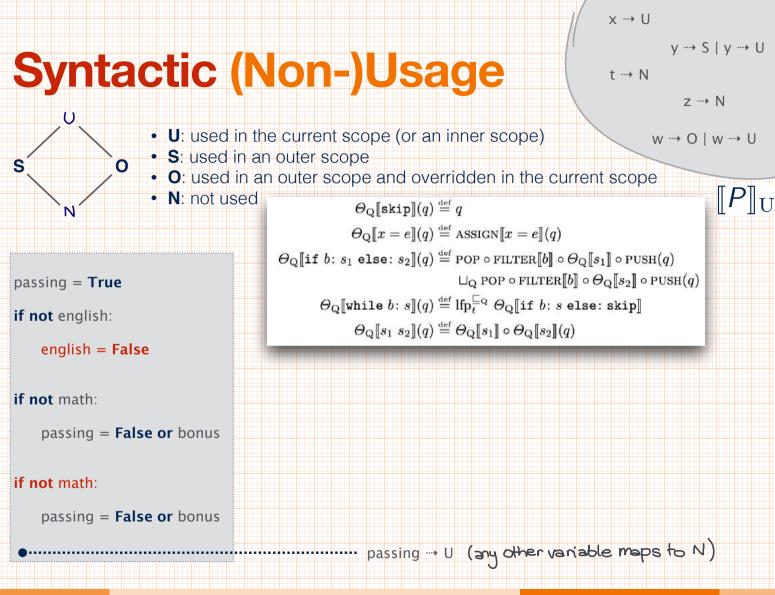
$$P \models \mathcal{N}_{I_P}^* \Leftarrow \gamma_{\prec}(\gamma_F(\llbracket P \rrbracket_F)) \subseteq \mathcal{N}_{I_P}^*$$

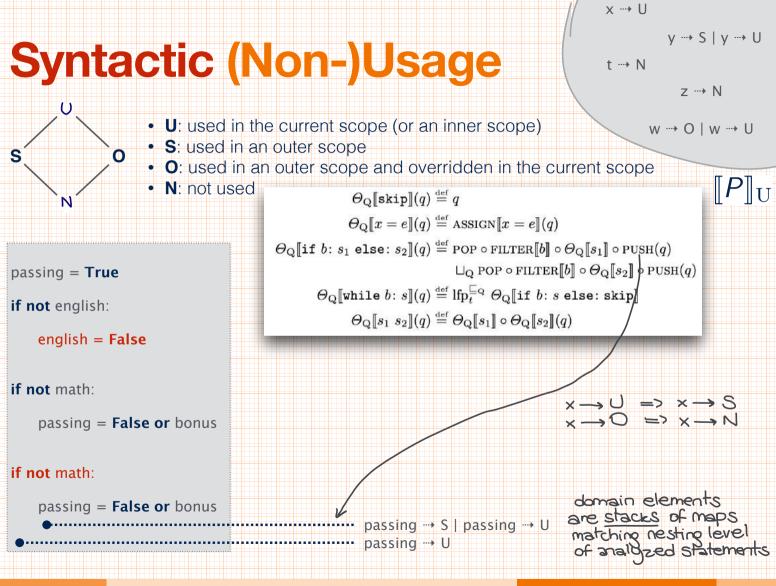


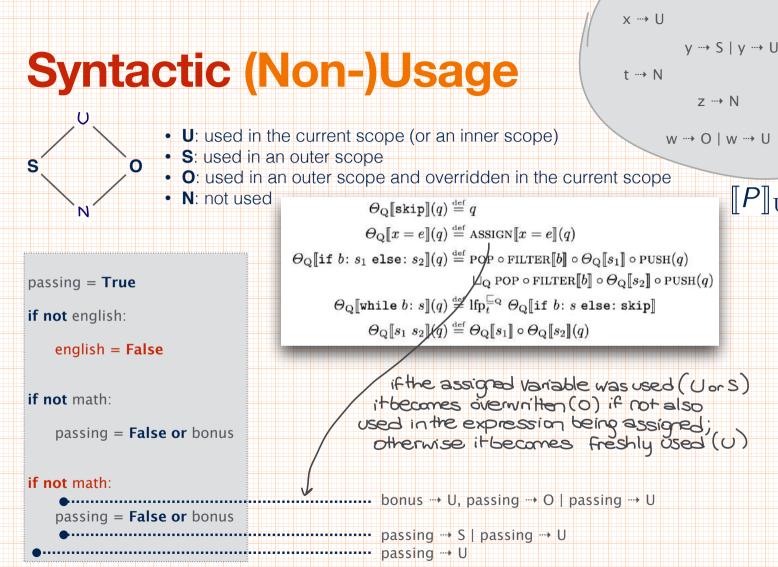




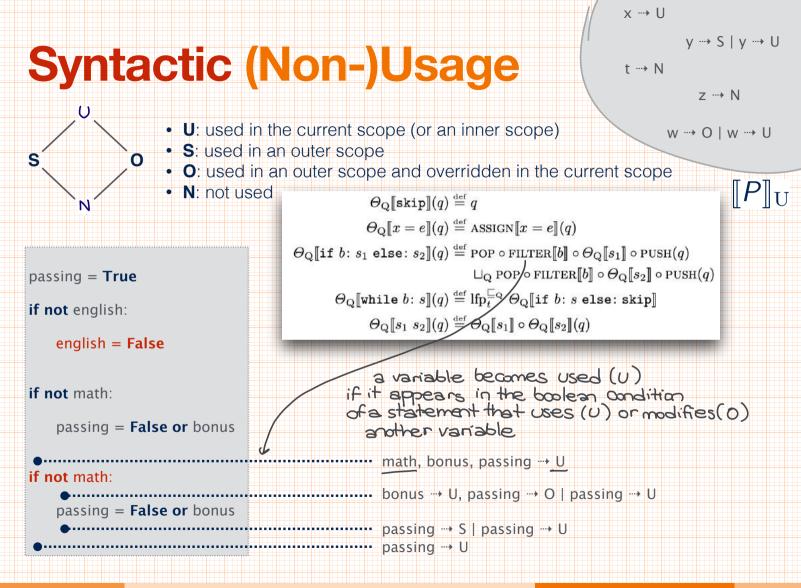


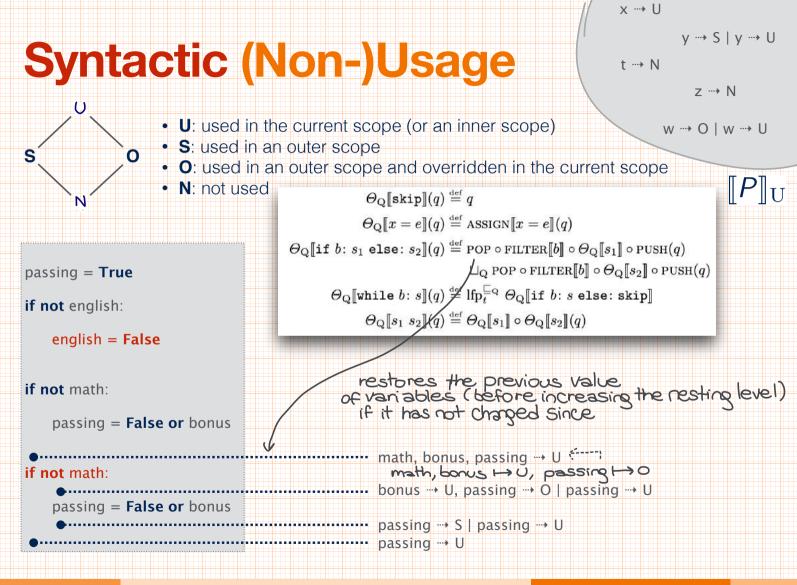


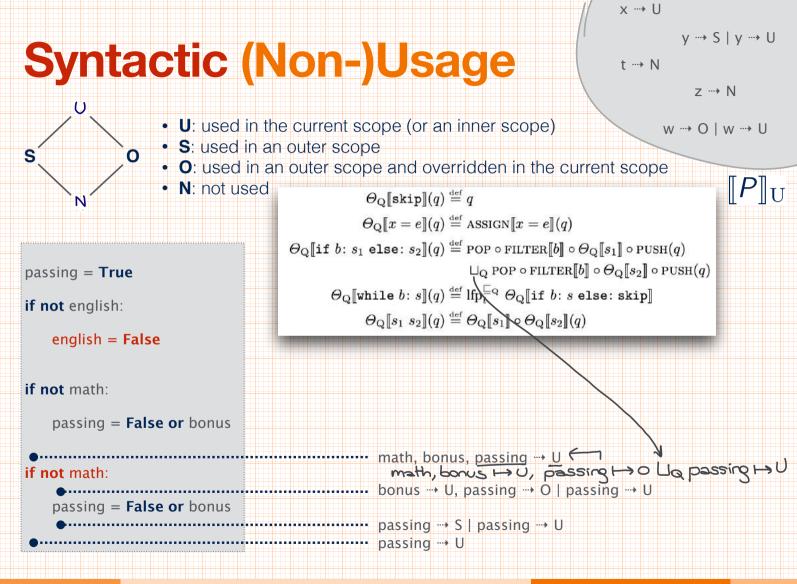


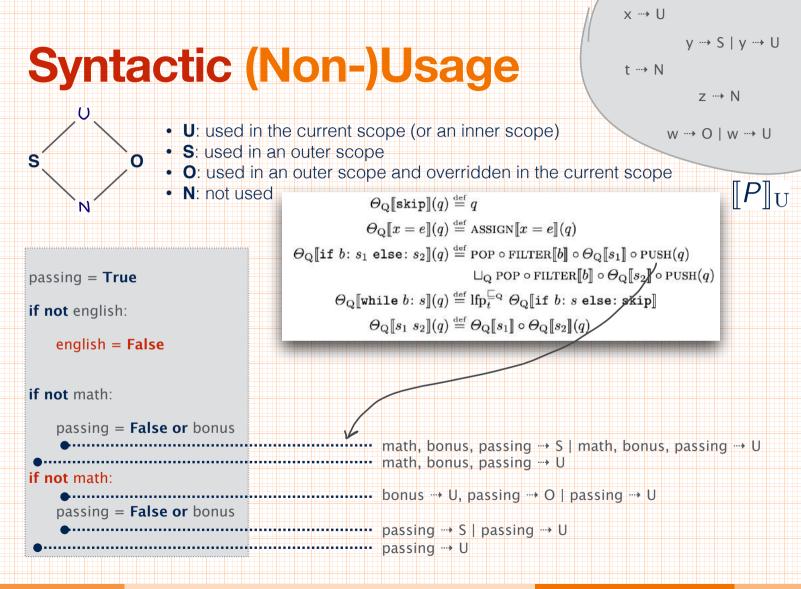


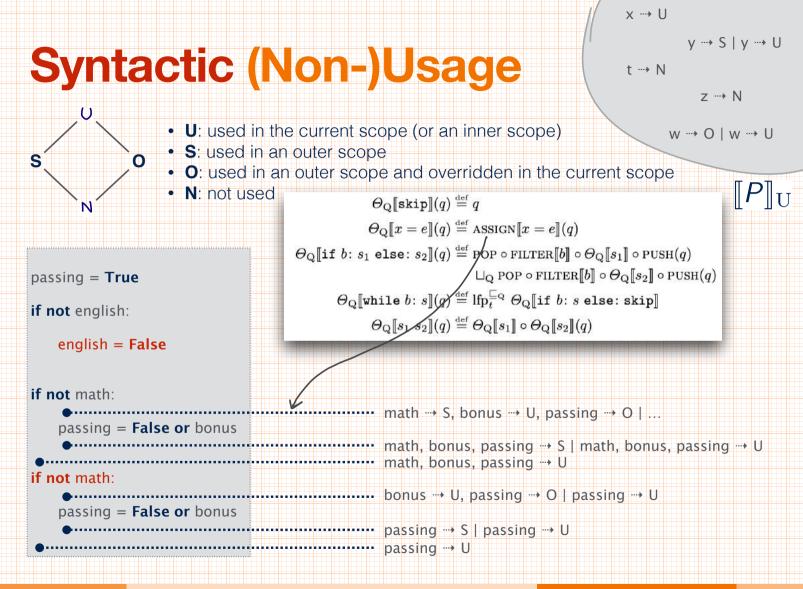
 $\llbracket P \rrbracket_{\mathrm{U}}$

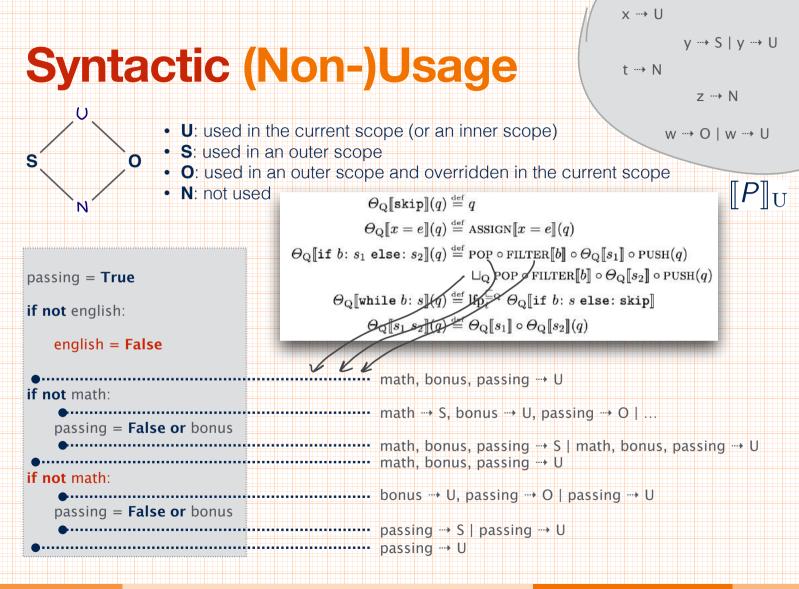


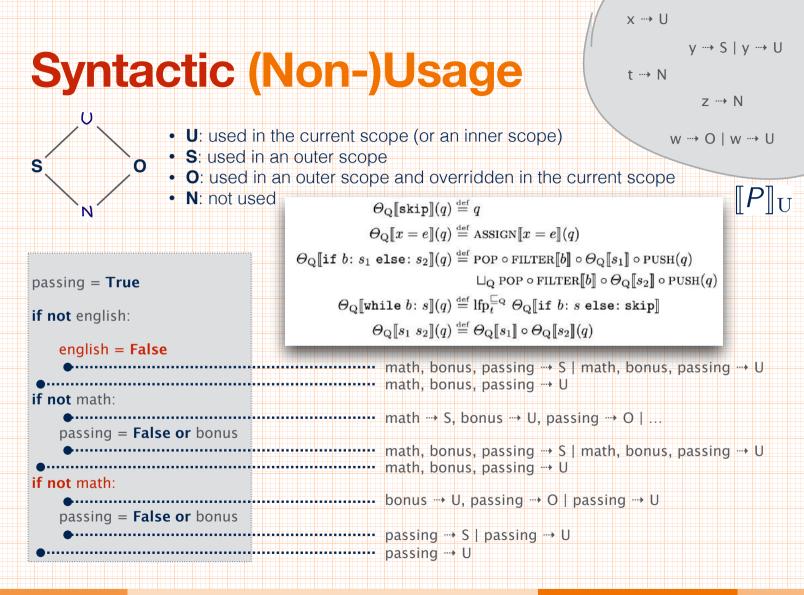


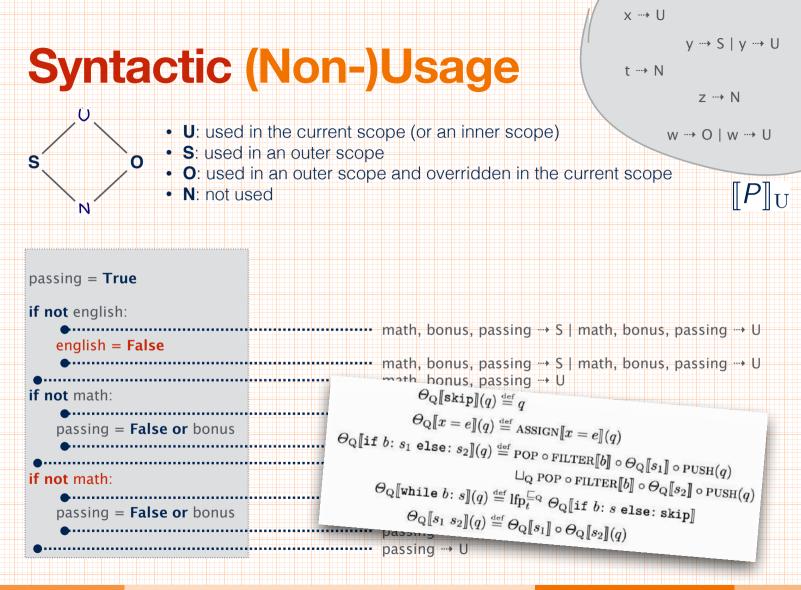












Syntactic (Non-)Usage

y ---→ S | y ---→ U

z …→ N

W ---> O | W ---> U

 $\|P\|_{U}$

x --- + U

t ... → N

U: used in the current scope (or an inner scope)
S: used in an outer scope

• O: used in an outer scope and overridden in the current scope

• N: not used

passing = True if not enalish: math, bonus, passing \rightarrow S | math, bonus, passing \rightarrow U english = False math honus, passing --- U if not math: $\Theta_{\mathbf{Q}}[\![\texttt{skip}]\!](q) \stackrel{\text{def}}{=} q$ $\Theta_{\mathrm{Q}}\llbracket x = e
rbracket(q) \stackrel{\scriptscriptstyle\mathrm{def}}{=} \mathrm{ASSIGN}\llbracket x = e
rbracket(q)$ passing = False or bonus $\Theta_{\mathbf{Q}}\llbracket \texttt{if } b \colon s_1 \texttt{ else} \colon s_2 \rrbracket(q) \stackrel{\text{\tiny def}}{=} \texttt{POP} \circ \texttt{FILTER}\llbracket b \rrbracket \circ \Theta_{\mathbf{Q}}\llbracket s_1 \rrbracket \circ \texttt{PUSH}(q)$ $\sqcup_{\mathbf{Q}} \operatorname{POP} \circ \operatorname{FILTER}\llbracket b \rrbracket \circ \Theta_{\mathbf{Q}}\llbracket s_2 \rrbracket \circ \operatorname{PUSH}(q)$ if not math: $\Theta_Q[\![\texttt{while } b: s]\!](q) \stackrel{\text{\tiny def}}{=} \mathrm{lfp}_t^{\sqsubseteq_Q} \; \Theta_Q[\![\texttt{if } b: s \; \texttt{else: skip}]\!]$ $\Theta_{\mathrm{Q}}\llbracket s_1 \ s_2
rbracket(q) \stackrel{\scriptscriptstyle\mathrm{def}}{=} \Theta_{\mathrm{Q}}\llbracket s_1
rbracket \circ \Theta_{\mathrm{Q}}\llbracket s_2
rbracket(q)$ passing = False or bonus

()

S

x --- > U Syntactic (Non-)Usage ()• U: used in the current scope (or an inner scope) • S: used in an outer scope S • O: used in an outer scope and overridden in the current scope N: not used N > the input variables english and science are definetly not used by the program math, bonus ---> U, passing ---> O

math, bonus, passing ---> U if not enalish:

········ math, bonus, passing → S | math, bonus, passing → U english = False

math, bonus, passing
$$\rightarrow$$
 S | math, bonus, passing \rightarrow U
math honus, passing \rightarrow U
 $\Theta_Q[[skip]](q) \stackrel{\text{def}}{=} q$
 $\Theta_Q[[x = e]](q) \stackrel{\text{def}}{=} ASSIGN[[x = e]](q)$
 $\Theta_Q[[if b: s_1 else: s_2]](q) \stackrel{\text{def}}{=} POP \circ FILTER[[b]] \circ \Theta_Q[[s_1]] \circ PUSH(q)$
 $\sqcup_Q POP \circ FILTER[[b]] \circ \Theta_Q[[s_2]] \circ PUSH(q)$
 $\Theta_Q[[while b: s]](q) \stackrel{\text{def}}{=} Ifp_t^{\sqsubseteq_Q} \Theta_Q[[if b: s else: skip]]$
 $\Theta_Q[[s_1 s_2]](q) \stackrel{\text{def}}{=} \Theta_Q[[s_1]] \circ \Theta_Q[[s_2]](q)$

passing = **True**

if not math:

if not math:

passing = False or bonus

passing = False or bonus

Static Analysis for Data Science

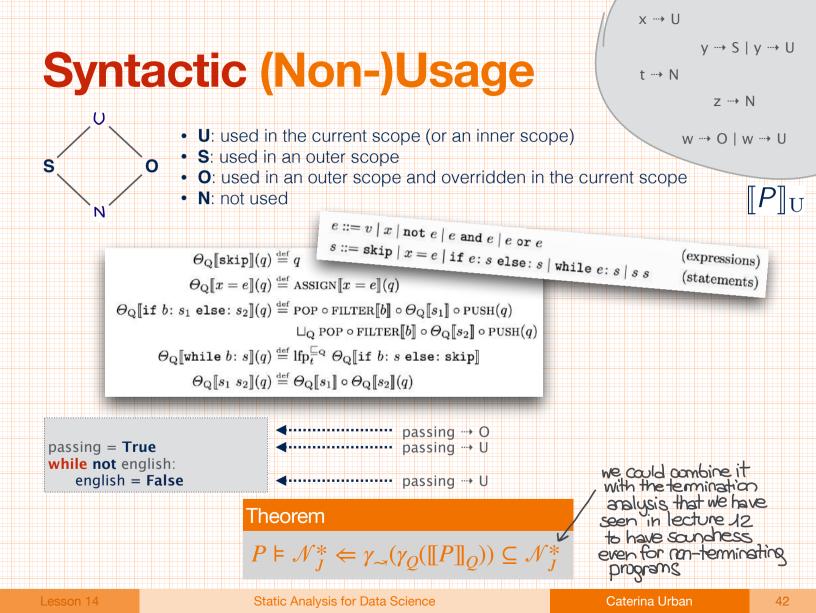
v ---→ S | v ---→ U

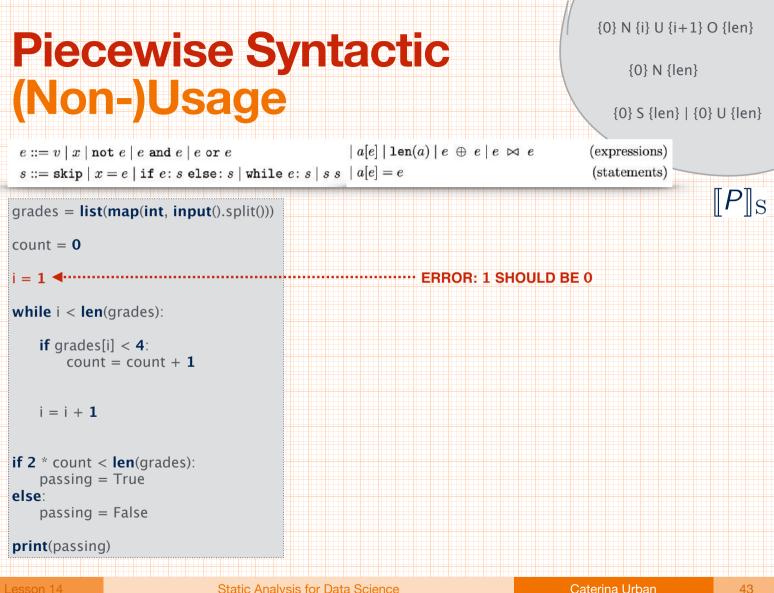
7 ··· → N

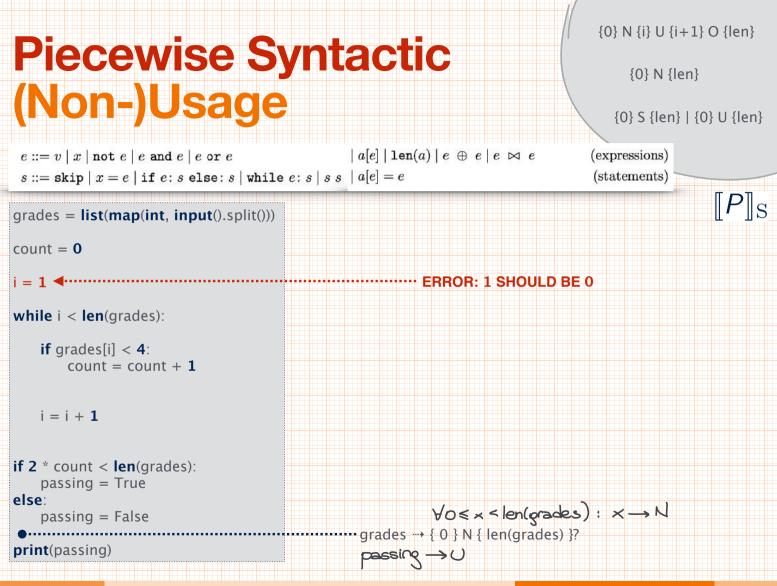
t ... → N

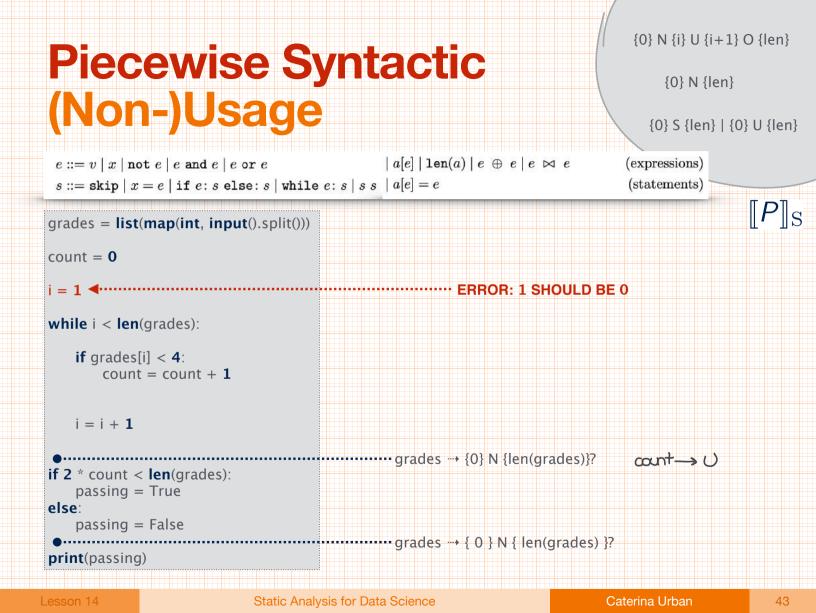
W ---> O | W ---> U

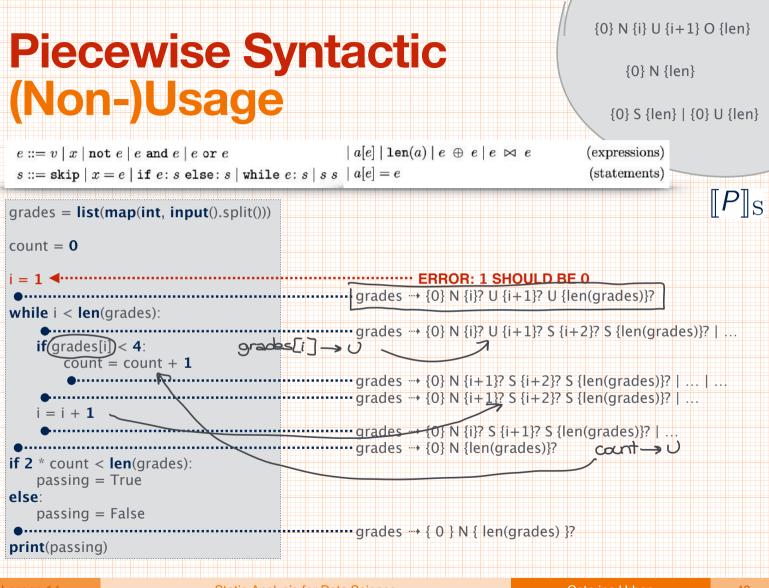
 $\llbracket P
rbracket_{\mathrm{U}}$

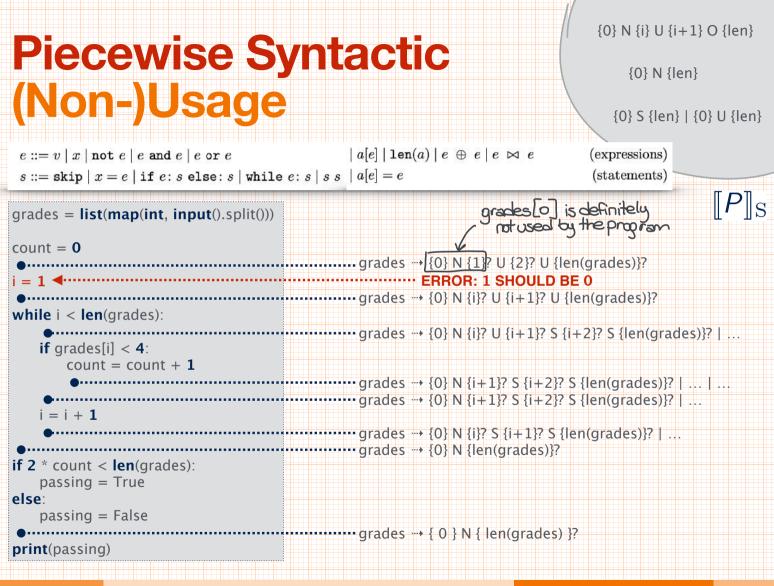












Implicit Assumptions on Data

What Programs Want Automatic Inference of Input Data Specifications

Caterina Urban^{1,2}

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² DIENS, École Normale Supérieure, CNRS, PSL University, Paris, France caterina.urban@inria.fr

Abstract. Nowadays, as machine-learned software quickly permeates our society, we are becoming increasingly vulnerable to programming errors in the data pre-processing or training software, as well as errors in the data itself. In this paper, we propose a static shape analysis framework for input data of data-processing programs. Our analysis automatically infers necessary conditions on the structure and values of the data read by a data-processing program. Our framework builds on a family of underlying abstract domains, extended to indirectly reason about the input data rather than simply reasoning about the program variables. The choice of these abstract domain is a parameter of the analysis. We describe various instances built from existing abstract domains. The proposed approach is implemented in an open-source static analyzer for PYTHON programs. We demonstrate its potential on a number of representative examples.

1 Introduction

Due to the availability of vast amounts of data and corresponding tremendous advances in machine learning, computer software is nowadays an ever increasing presence in every aspect our society. As we rely more and more on machinelearned software, we become increasingly vulnerable to programming errors but (in contrast to traditional software) also errors in the data used for training.

In general, before software training, the data goes through long pre-processing pipelines³. Errors can be missed, or even introduced, at any stage of these pipelines. This is even more true when data pre-processing stages are disregarded as single-use glue code and, for this reason, are poorly tested, let alone statically analyzed or verified. Moreover, this kind of code is often written in a rush and is highly dependent on the data (e.g., the use of magic constants is not uncommon) All this together, greatly increases the likelihood for errors to be noticed extremely late in the pipeline (which entails a more or less important waste of time), or more dangerously, to remain completely unnoticed.

³ https://www.nytimes.com/2014/08/18/technology/for-big-data-scientistshurdle-to-insights-is-janitor-work.html

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