Static Analysis for Data Science MPRI 2-6: Abstract Interpretation, **Application to Verification and Static Analysis**

Caterina Urban

January 20th, 2024







Static Analysis for Data Science

predictions

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Data Scientists

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Data Scientist: The Sexiest Job of the 21st Century

Andrew McAfee and Erik Brynjolfsson

Andrew J Buboltz, silk screen on a page from a high school yearbook, 8.5" x 12", 2011 Tamar Cohen

When Jonathan Goldman arrived for work in June 2006 at <u>LinkedIn, the</u> <u>business networking site</u>, the place still felt like a start-up. The company had just under 8 million accounts, and the number was growing quickly as existing members invited their friends and colleagues to join. But users weren't seeking out connections with the people who were already on the site at the rate executives had expected. Something was apparently missing in the social experience. As one LinkedIn manager put it, "It was like arriving at a conference reception and realizing you don't know

Science is rate executives had expected to the Caterina Alphan missing in the social experience. As one Linkedin manager put it, "It was like arriving at a conference reception and realizing you don't know

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P. Subotić et al. - A Static Analysis Framework for Data Science Notebooks (ICSE 2022)

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P. Subotić et al. - A Static Analysis Framework for Data Science Notebooks (ICSE 2022)

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Anomalously Unused Data

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The Reinhart-Rogoff Paper

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

By Peter Coy Mapril 18, 2013

By PAUL KRUGMAN

Western world?

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

SAVE B EMAIL + SHARE B PRINT

The Excel Depression

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

Ma Doinhart and Mr Dogoff had andibility thanks to a

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.

REPRINTS

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SCIENCE VIS & WORLD VIECH

Excel spreadsheet error coronavirus cases

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

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print(passing) OUTPUT VARIABLES

the input variables english and science are unused

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ERROR: math SHOULD BE science

Data Usage Static Analysis [Urban18] **3-Step Recipe**

practical tools targeting specific programs

abstract semantics, abstract domains algorithmic approaches to decide program properties

concrete semantics mathematical models of the program behavior

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Dependency Fairness

$\mathcal{F}_i \stackrel{\text{def}}{=} \{ \llbracket M \rrbracket \mid \text{UNUSED}_i(\llbracket M \rrbracket) \}$

 \mathcal{F}_i is the set of all neural networks M (or, rather, their semantics [[M]]) that **do not use** the value of the sensitive input node $x_{0,i}$ for classification

Intuitively: inputs differing only on the value of the sensitive input node $x_{0,i}$ should lead to the same classification outcome

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$$\eta(t_0) = \eta(t'_0) \qquad \eta:$$

$$\pi(x_{0,j}) = t'_0(x_{0,j})) \qquad \eta:$$

$$\eta(x_{0,j}) = \begin{cases} \top & j = i \\ x_{0j} & \text{otherwise} \end{cases}$$

$$\mathscr{F}_i \Leftarrow \llbracket M \rrbracket \subseteq \llbracket M \rrbracket^{\natural} \subseteq \mathscr{F}_i$$

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Data (Non-)Usage

 $\mathcal{N}_{I} \stackrel{\text{def}}{=} \{ \llbracket P \rrbracket \mid \mathsf{UNUSED}_{I}(\llbracket P \rrbracket) \}$

 \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

UNUSED_J([[P]]) $\stackrel{\text{def}}{=} \forall t \in [[P]], V \in \mathscr{R}^{|J|} \colon t_0(J) \neq V \Rightarrow \exists t' \in [[P]] \colon$ $(\forall i : i \notin J \Rightarrow t_0(i) = t'_0(i))$ $\wedge t'_0(J) = V$ $\wedge t_{\omega} = t'_{\omega}$

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

Theorem

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

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practical tools

concrete semantics mathematical models of the program behavior

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Hierarchy of Semantics parallel semantics $\{[M]\}_{\sim}^{\mathbb{I}}$ $\{[M]\}^{\mathbb{I}}$ $\{[[M]]\}$ collecting semantics Caterina Urban 114 Lesson 12 Static Analysis for Machine Learning

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practical tools

abstract semantics, abstract domains algorithmic approaches to decide program properties

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Hierarchy of Semantics

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Data (Non-) Usage **Not a Subset-Closed Property**

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Data (Non-)Usage

 $\mathcal{N}_{I} \stackrel{\text{def}}{=} \{ [\![P]\!] \mid \text{UNUSED}_{I}([\![P]\!]) \}$

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

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Data (Non-) Usage Abstractions **Over-Approximation of the Used Input Data** \Rightarrow Under-Approximation of the Unused Input Data

- $P \models \mathcal{N}_{J^{\natural} \subseteq J} \Leftarrow \{ \llbracket P \rrbracket \} \subseteq \llbracket P \rrbracket_A^{\natural} \subseteq \mathcal{N}_{J^{\natural} \subseteq J}$

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Secure Information Flow

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Secure Information Flow

- and the program is terminating

passing = **True** while not english:

$$P \models \mathcal{N}_J^+ \Leftarrow \{$$

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Strong Liveness

a variable is strongly live if

- it is used in an assignment to another strongly live variable
- it is used in a statement other than an assignment

$$\begin{split} e &::= v \\ S &::= \mathrm{sk} \\ \Theta_{\mathrm{X}} \llbracket \mathrm{skip} \rrbracket(S) \stackrel{\mathrm{def}}{=} S \\ \Theta_{\mathrm{X}} \llbracket x = e \rrbracket(S) \stackrel{\mathrm{def}}{=} \begin{cases} (S \setminus \{x\}) \cup \mathrm{VARS} \\ S \end{cases} \\ \Theta_{\mathrm{X}} \llbracket \mathrm{if} \ b \colon s_1 \ \mathrm{else} \colon s_2 \rrbracket(S) \stackrel{\mathrm{def}}{=} \mathrm{VARS}(b) \cup \Theta_{\mathrm{X}} \llbracket s_1 \rrbracket(S) \\ \Theta_{\mathrm{X}} \llbracket \mathrm{while} \ b \colon s \rrbracket(S) \stackrel{\mathrm{def}}{=} \mathrm{VARS}(b) \cup \Theta_{\mathrm{X}} \llbracket s_1 \rrbracket(S) \\ \Theta_{\mathrm{X}} \llbracket \mathrm{while} \ b \colon s \rrbracket(S) \stackrel{\mathrm{def}}{=} \Theta_{\mathrm{X}} \llbracket s_1 \rrbracket \circ \Theta_{\mathrm{X}} \llbracket s_2 \rrbracket(S) \\ \end{split}$$

$(P(X), \subseteq, U, \cap, \emptyset, X7 : abstract obmain$

Strong Liveness

$$\begin{split} & e ::= v \\ & \Theta_{\mathrm{X}}[\![\mathrm{skip}]\!](S) \stackrel{\mathrm{def}}{=} S \\ & \Theta_{\mathrm{X}}[\![x = e]\!](S) \stackrel{\mathrm{def}}{=} \begin{cases} (S \setminus \{x\}) \cup \mathrm{VARS} \\ S \end{cases} \\ & \Theta_{\mathrm{X}}[\![\mathrm{if}\ b:\ s_1\ \mathrm{else}:\ s_2]\!](S) \stackrel{\mathrm{def}}{=} \mathrm{VARS}(b) \cup \Theta_{\mathrm{X}}[\![s_1]\!](c) \\ & \Theta_{\mathrm{X}}[\![\mathrm{while}\ b:\ s]\!](S) \stackrel{\mathrm{def}}{=} \mathrm{VARS}(b) \cup \Theta_{\mathrm{X}}[\![s_1]\!](S) \\ & \Theta_{\mathrm{X}}[\![\mathrm{while}\ b:\ s]\!](S) \stackrel{\mathrm{def}}{=} \Theta_{\mathrm{X}}[\![s_1]\!] \circ \Theta_{\mathrm{X}}[\![s_2]\!](S) \end{split}$$

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- **U**: used in the current scope (or an inner scope)
- **S**: used in an outer scope
- N: not used

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x ---→ U y → S | y → U t ---→ N z ...→ N w ---→ O | w ---→ U • O: used in an outer scope and overridden in the current scope $\llbracket P \rrbracket_Q$ $\Theta_{\mathrm{Q}}\llbracket \texttt{skip}
rbracket(q) \stackrel{\scriptscriptstyle\mathrm{def}}{=} q$ $\Theta_{\mathrm{Q}}\llbracket x = e
rbracket(q) \stackrel{\mathrm{def}}{=} \mathrm{ASSIGN}\llbracket x = e
rbracket(q)$ $\Theta_{\mathbf{Q}}\llbracket \texttt{if } b: s_1 \texttt{ else: } s_2 \rrbracket(q) \stackrel{\text{def}}{=} \texttt{POP} \circ \texttt{FILTER}\llbracket b \rrbracket \circ \Theta_{\mathbf{Q}}\llbracket s_1 \rrbracket \circ \texttt{PUSH}(q)$ $\sqcup_{\mathbf{Q}} \operatorname{POP} \circ \operatorname{FILTER}[\![b]\!] \circ \Theta_{\mathbf{Q}}[\![s_2]\!] \circ \operatorname{PUSH}(q)$ $\Theta_{\mathbf{Q}}[[\texttt{while } b: s]](q) \stackrel{\text{\tiny def}}{=} \mathrm{lfp}_{t}^{\sqsubseteq_{\mathbf{Q}}} \ \Theta_{\mathbf{Q}}[[\texttt{if } b: s \texttt{ else: skip}]]$ $\Theta_{\mathrm{Q}}\llbracket s_1 \ s_2 \rrbracket(q) \stackrel{\text{\tiny def}}{=} \Theta_{\mathrm{Q}}\llbracket s_1 \rrbracket \circ \Theta_{\mathrm{Q}}\llbracket s_2 \rrbracket(q)$

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x ---→ U y ---→ S | y ---→ U t ---→ N z ...→ N w ---→ O | w ---→ U • O: used in an outer scope and overridden in the current scope $\Theta_{\mathrm{Q}}\llbracket \texttt{skip}
rbracket(q) \stackrel{\scriptscriptstyle\mathrm{def}}{=} q$ $\Theta_{\mathrm{Q}}\llbracket x = e
rbracket(q) \stackrel{\mathrm{def}}{=} \mathrm{ASSIGN}\llbracket x = e
rbracket(q)$ $\Theta_{\mathbf{Q}}\llbracket \texttt{if } b \colon s_1 \texttt{ else} \colon s_2 \rrbracket(q) \stackrel{\text{def}}{=} \texttt{POP} \circ \texttt{FILTER}\llbracket b \rrbracket \circ \Theta_{\mathbf{Q}}\llbracket s_1 \rrbracket \circ \texttt{PUSH}(q)$ $\sqcup_{\mathbf{Q}} \operatorname{POP} \circ \operatorname{FILTER}[\![b]\!] \circ \Theta_{\mathbf{Q}}[\![s_2]\!] \circ \operatorname{PUSH}(q)$ $\Theta_{\mathbf{Q}}[[\text{while } b: s]](q) \stackrel{\text{def}}{=} \operatorname{lfp}_{t}^{\sqsubseteq_{\mathbf{Q}}} \Theta_{\mathbf{Q}}[[\text{if } b: s \text{ else: skip}]]$ $\Theta_{\mathrm{Q}}\llbracket s_1 \ s_2 \rrbracket(q) \stackrel{\text{\tiny def}}{=} \Theta_{\mathrm{Q}}\llbracket s_1 \rrbracket \circ \Theta_{\mathrm{Q}}\llbracket s_2 \rrbracket(q)$ (any other variable maps to N)

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Lesson 13

$$x \rightarrow U$$

$$y \rightarrow S | y \rightarrow U$$

$$t \rightarrow N$$

$$z \rightarrow N$$

$$w \rightarrow 0 | w \rightarrow U$$

$$W \rightarrow$$

Lesson 13



Lesson 13





Lesson 13





Lesson 13





Lesson 13





Lesson 13

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x ---→ U y ---→ S | y ---→ U t ---→ N z ...→ N w ---→ O | w ---→ U • O: used in an outer scope and overridden in the current scope $\Theta_{\mathrm{Q}}\llbracket \mathtt{skip}
rbracket(q) \stackrel{\mathrm{def}}{=} q$ $\Theta_{\mathrm{Q}}[x = e](q) \stackrel{\mathrm{def}}{=} \mathrm{ASSIGN}[x = e](q)$ $\Theta_{\mathbf{Q}}\llbracket \texttt{if } b \colon s_1 \texttt{ else} \colon s_2 \rrbracket(q) \stackrel{\text{def}}{=} \texttt{POP} \circ \texttt{FILTER}\llbracket b \rrbracket \circ \Theta_{\mathbf{Q}}\llbracket s_1 \rrbracket \circ \texttt{PUSH}(q)$ $\sqcup_{\mathbf{Q}} \operatorname{POP} \circ \operatorname{FILTER}[\![b]\!] \circ \Theta_{\mathbf{Q}}[\![s_2]\!] \circ \operatorname{PUSH}(q)$ $\Theta_{\mathbf{Q}}[[\text{while } b: s]](q) \stackrel{\text{def}}{=} \operatorname{lfp}_{t}^{\sqsubseteq_{\mathbf{Q}}} \Theta_{\mathbf{Q}}[[\text{if } b: s \text{ else: skip}]]$ $\Theta_{\mathcal{Q}}\llbracket s_1 \ s_2 \rrbracket(q) \stackrel{\text{def}}{=} \Theta_{\mathcal{Q}}\llbracket s_1 \rrbracket \circ \Theta_{\mathcal{Q}}\llbracket s_2 \rrbracket(q)$ math, bonus, passing ---> S | math, bonus, passing ---> U math, bonus, passing ----> U math \rightarrow S, bonus \rightarrow U, passing \rightarrow O | ... math, bonus, passing ---> S | math, bonus, passing ---> U math, bonus, passing ----> U passing --- U







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 $\Theta_{\mathrm{Q}}\llbracket \texttt{skip}
rbracket(q) \stackrel{ ext{def}}{=} q$



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Data Usage Static Analysis [Urban18] **3-Step Recipe**

practical tools targeting specific programs



strongly-live variable analysis

concrete semantics mathematical models of the program behavior

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secure information flow

Hierarchy of Semantics

syntactic non-usage























Data Leakage

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Conference on Fairness, Accountability, and Transparency

Proceedings of Machine Learning Research 81:1–15, 2018

A case study of algorithm-assisted decision making in child maltreatment hotline screening decisions

Alexandra Chouldechova Carnegie Mellon University Pittsburgh, PA, 15213, USA Emily Putnam-Hornstein Suzanne Dworak-Peck School of Social Work University of Southern California Los Angeles, CA, 90089, USA Diana Benavides-Prado

Oleksandr Fialko Rhema Vaithianathan Centre for Social Data Analytics Auckland University of Technology Auckland, New Zealand Editors: Sorelle A. Friedler and Christo Wils

Abstract Every year there are more than 3.6 lion referrals made to child protection a

cies across the US. The practice of scr ing calls is left to each jurisdiction t low local practices and policies, poter leading to large variation in the which referrals are treated across the try. Whilst increasing access to link ministrative data is available, it is for welfare workers to make system of historical information about all dren and adults on a single refe Risk prediction models that use collected administrative data can workers to better identify case. likely to result in adverse outcor ever, the use of predictive anal. area of child welfare is content is a possibility that some cor such as those in poverty or t lar racial and ethnic groupsadvantaged by the reliance of administrative data. On the

these analytics tools can augment or replace human judgments, which themselves are biased and imperfect. In this paper we describe our work on developing, validatuditing, and deploying a risk EHORNSTE@USC.EDU

Family separation in Allegheny county

In 2016, Allegheny county in Pennsylvania adopted the Allegheny Family Screening Tool (AFST) to predict which children are at risk of maltreatment. AFST is used to decide which families should be investigated by social workers. In these investigations, social workers can forcibly remove children from their families and place them in foster care, even if there are no allegations of abuse-only poverty-based neglect.

Two years later, it was discovered that AFST suffered from data leakage, leading to exaggerated claims about its performance. In addition, the tool was systematically biased against Black families. When questioned, the creators trotted out the familiar defense that the final decision is always made by a human decision-maker.

to make systematic use of historical informatio about all the children and adults on a single refer-

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ACHOULD@CMU.EDU
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https://www.aisnakeoil.com/p/the-bait-and-switch-behind-ai-risk

Withianathan.



A STAT INVESTIGATION

Epic's sepsis algorithm is going off the rails in the real world. The use of these variables may explain why



By Casey Ross y Sept. 27, 2021



Epic's sepsis prediction debacle

Epic is a large healthcare software company. It stores health data for over 300 million patients. In 2017, Epic released a sepsis prediction model. Over the next few years, it was deployed in hundreds of hospitals across the U.S. However, a 2021 study from researchers at the University of Michigan found that Epic's model vastly underperformed compared to the developer's claims.

The tool's inputs included information about whether a patient was given antibiotics. But if a patient is given antibiotics, they have already been diagnosed with sepsis—making the tool's prediction useless. These cases were still counted as successes when the developer evaluated the tool, leading to exaggerated claims about how well it performed. This is an example of data leakage, a common error in building AI tools.

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https://www.aisnakeoil.com/p/the-bait-and-switch-behind-ai-risk









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Data Leakage Static Analysis [Drobnjaković24] **3-Step Recipe**

practical tools targeting specific programs

abstract semantics, abstract domains algorithmic approaches to decide program properties

concrete semantics mathematical models of the program behavior

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(Absence of) Data Leakage **Independence of Training and Testing Data**



0000111100001110 0011001100110010

A SINGLE ROW CHANGE IN THE INPUT DATA AFFECTS **BOTH TRAIN AND TEST DATA**

00001111000001110 0011001100110010 0101010101010100 000000011111110

INPUT DATA MIN-MAX NORMALIZATION **TRAIN DATA TEST DATA**

INPUT DATA

TRAIN DATA TEST DATA

MIN-MAX NORMALIZATION 3339999333399999 3399339933993399 3939393939393939393939 3333333999999999

3333999933339999 33993399339933993399

A SINGLE ROW CHANGE IN THE INPUT DATA AFFECTS ONLY TRAIN OR ONLY TEST DATA

> 0000110000001100 0011000001100000 010101010000000000 00000001010101010

> > **Caterina Urban**







(Absence of) Data Leakage **Independence of Training and Testing Data**

$\mathcal{J} \stackrel{\mathsf{def}}{=} \{ \llbracket P \rrbracket \mid \mathsf{INDEPENDENT}(\llbracket P \rrbracket) \}$

input data frame variables INDEPENDENT($\llbracket P \rrbracket$) $\stackrel{\text{def}}{=} \forall \sigma \in \llbracket P \rrbracket$, $i \in I_P$, $r \in R_i$: UNCHANGED($\sigma, i, r, U_P^{\text{test}}$) \lor UNCHANGED($\sigma, i, r, U_P^{\text{train}}$) $\text{UNCHANGED}(\sigma, i, r, U) \stackrel{\text{def}}{=} \forall \bar{v} \in \mathbb{V}^{C_i} : \sigma(i)[r] \neq \bar{v} \Rightarrow (\exists \sigma' \in \Upsilon[P]] : \sigma'(i)[r] = \bar{v} \land \eta(\sigma) = \eta(\sigma') \land \sigma(U) = \sigma'(U))$ $\eta(\sigma) \stackrel{\text{def}}{=} \lambda j \colon \lambda r' \colon \begin{cases} \sigma(j)[r'] & j \in I_P \setminus \{i\} \lor r' \in R_i \colon r' \neq r \\ \top & \text{otherwise} \end{cases}$ $\sigma(X) = \sigma'(X) \stackrel{\text{\tiny def}}{=} \forall x \in X \colon \sigma(x) = \sigma'(x)$



Data Leakage Static Analysis [Drobnjaković24] **3-Step Recipe**

practical tools

concrete semantics mathematical models of the program behavior

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Data Leakage Semantics

INPUT DATA

MIN-MAX NORMALIZATION

> **TRAIN DATA TEST DATA**

3333999933339999 3399339933993399 393939393939393939 333333399999999

0000111100001110 0011001100110010 010101010101010100 000000011111110

0000111100001110 0011001100110010 0101010101010100 000000011111110

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practical tools

abstract semantics, abstract domains algorithmic approaches to decide program properties

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Data Frame Sources Abstract Domain

- $X \qquad \longmapsto \qquad \left\langle \left\{ \text{source1}_{[1,10]}^{\{\text{id}\}}, \text{source2}_{[1,10]}^{\{\text{id}\}} \right\}, \text{FALSE} \right\rangle$
- $\mathsf{Z} \qquad \longmapsto \qquad \Big\langle \Big\{ \mathsf{source3}^{\{\mathsf{id},\mathsf{zip}\}}_{[0,\infty]} \Big\}, \mathsf{FALSE} \Big\rangle$
- $W \mapsto \left\langle \left\{ \text{source2}_{[500,1000]}^{\{\text{id}\}} \right\}, \text{TRUE} \right\rangle$

DATA SOURCE

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DATA FRAME VARIABLES









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concrete semantics mathematical models of the program behavior

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Experimental Evaluation 7378 Executions in 2111 Notebooks from Kaggle

Implementation	True P	Ealaa Daaitiyaa	
Implementation	Taint Data Leakage	Overlap Data Leakage	raise Positives
NBLyzer + Original Data Leakage Analysis	10	0	2
NBLyzer + Our Data Leakage Analysis	10	15	2
	IN 5 NOTEBOOKS	IN 11 NOTEBOOKS	CONFIRMED BY 4 DATA SCIENTISTS AT MICROSOFT

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practical tools targeting specific programs

abstract semantics, abstract domains algorithmic approaches to decide program properties

concrete semantics mathematical models of the program behavior

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Unexpected Data

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Unexpected Data



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Example

Insert Cell Insert Cell ▲ ▲ Prt Pandas as pd. read_csv('(head() Name Q1 Q2 Q Alice A A B Bob F B	<pre>kpoint: a few seconds ago Kernel Help Code d Grades.csv', index A</pre>	(autosaved) ↓ □ _col=0)			Trusted	Logout Python 3 O	
Insert Cell	Kernel Help C Code d Grades.csv', index A	(dutosaved) .col=0)			Trusted	Python 3 O	
Insert Cell	Kernel Help ■ C Code d Grades.csv', index 3 A	<pre></pre>			Trusted	Python 3 O	
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<pre>pd.read_csv('(head() Name Q1 Q2 Q A Alice A A B Bob F B</pre>	d Grades.csv', index_ 03 A	_col=0)					
pd.read_csv('(nead() Name Q1 Q2 Q A Alice A A B Bob F B	Grades.csv', index_ 03 A	_col <mark>=0</mark>)					
Name Q1 Q2 Q A Alice A A B Bob F B	A						
Alice A A Bob F B	A						
Alice A A Bob F B	A						
Bob F B							
	В						
Carol F A	с						
B David D F	С						
de2gpa = { 'A': iloc[:, df.colur	4.0, 'B': 3.0, 'C mns.str.startswith	': 2.0, 'D': 1.0, 'F': ('Q')] = df.iloc[:, df	<pre>0.0 } .columns.str.startswith</pre>	n(<mark>'Q'</mark>)].applymap(g	rade2gpa.ge	et)	
df['Mean'] = df.iloc[:, df.columns.str.startswith('Q')].mean(axis=1)							
es = pd.read_csv('Emails.csv', index_col=0)							
= df.join(es)							
<pre>= un[["Email", head()</pre>	"Mean"]]						
Email Mea	n						
)	_						
1 alice@uni.eu 4.0	0						
bob@uni.eu 2.0	0						
david@uni.eu 1	0						
	Bob F B Carol F A David D F Pavid P Pavid Pavid P Pavid P Pavid P Pavid Pavid P Pavid Pavid P Pavid Pavid Pavid P Pavid Pavid	<pre>Bob F B B Carol F A C David D F C Below and Below</pre>	<pre>Bob F B B Carol F A C David D F C le2gpa = { 'A': 4.0, 'B': 3.0, 'C': 2.0, 'D': 1.0, 'F': .loc[:, df.columns.str.startswith('Q')] = df.iloc[:, df Mean'] = df.iloc[:, df.columns.str.startswith('Q')].me pd.read_csv('Emails.csv', index_col=0) df.join(es) = un[["Email", "Mean"]] head() Email Mean alice@uni.eu 4.0 bob@uni.eu 2.0 carol@uni.eu 2.0 david@uni.eu 1.0</pre>	<pre>Bob F B B Carol F A C David D F C loc[:, df.columns.str.startswith('Q')] = df.iloc[:, df.columns.str.startswith Mean'] = df.iloc[:, df.columns.str.startswith('Q')].mean(axis=1) pd.read_csv('Emails.csv', index_col=0) df.join(es) = un[["Email", "Mean"]] head() Email Mean dailee@uni.eu 4.0 bob@uni.eu 2.0 david@uni.eu 1.0</pre>	<pre>Bob F B B Carol F A C David D F C E2gpa = { 'A': 4.0, 'B': 3.0, 'C': 2.0, 'D': 1.0, 'F': 0.0 } loc[:, df.columns.str.startswith('Q')] = df.iloc[:, df.columns.str.startswith('Q')].applymap(g Mean'] = df.iloc[:, df.columns.str.startswith('Q')].mean(axis=1) pd.read_csv('Emails.csv', index_col=0) df.join(es) = un[["Email", "Mean"]] head() Email Mean Email Mean Email Mean D A alice@uni.eu 4.0 A Alice@uni.eu 4</pre>	<pre>Bob F B B Garol F A C David D F C le2gpa = { 'A': 4.0, 'B': 3.0, 'C': 2.0, 'D': 1.0, 'F': 0.0 } Loc[:, df.columns.str.startswith('Q')] = df.iloc[:, df.columns.str.startswith('Q')].applymap(grade2gpa.gr Mean'] = df.iloc[:, df.columns.str.startswith('Q')].mean(axis=1) pd.read_csv('Emails.csv', index_col=0) fd.join(es) function func</pre>	

Lesson 13

Static Analysis for Data Science



Missing Values

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B + % 4	▲ ↓ ▶ Run ■ C ▶ Code	₽ + ≈ 4	▲ ▲ ► Run	Code			
In [1]:	<pre>import pandas as pd</pre>	In [1]:	import pandas as	pd			
In [2]:	<pre>df = pd.read_csv('Grades.csv', index_col= df.head()</pre>	In [2]:	df <mark>=</mark> pd.read_csv(df.head()	'Grades.csv', index_co	.=0)		
Out[2]:	Name Q1 Q2 Q3	Out[2]:	Name Q1 Q2	2 Q3			
	ID		ID				
	2394 Alice A A A		2394 Alice A A	A			
	4583 Bob F B B		4583 Bob F B	В			
	3956 Carol F A C		3956 Carol NaN A	С			
	9578 David D F C		9578 David D F	C			
In [3]:	<pre>grade2gpa = { 'A': 4.0, 'B': 3.0, 'C': 2. df.iloc[:, df.columns.str.startswith('Q')</pre>	In [3]:	grade2gpa = { 'A' df.iloc[:, df.col	: 4.0, 'B': 3.0, 'C': 2 .umns.str.startswith('Q'	<pre>2.0, 'D': 1.0, 'F': 0.0 })] = df.iloc[:, df.columns.st</pre>	tr.startswith(<mark>'Q'</mark>)].applym	nap(grade2gpa.get)
In [4]:	<pre>df['Mean'] = df.iloc[:, df.columns.str.st</pre>	In [4]:	df['Mean'] = df.i	loc[:, df.columns.str.s	<pre>startswith('Q')].mean(axis=1)</pre>		
In [5]:	<pre>es = pd.read_csv('Emails.csv', index_col=</pre>	In [5]:	es = pd.read_csv(<pre>'Emails.csv', index_col</pre>	.=0)		
In [6]:	un = df.join(es)	In [6]:	un = df.join(es)				
In [7]:	<pre>res = un[["Email", "Mean"]] res.head()</pre>	In [7]:	<pre>res = un[["Email" res.head()</pre>	, "Mean"]]			
Out[7]:	Email Mean	Out[7]:	Email M	ean			
	ID		ID				
	2394 alice@uni.eu 4.0		2394 alice@uni.eu	4.0			
	4583 bob@uni.eu 2.0		4583 bob@uni.eu	2.0			
	3956 carol@uni.eu 2.0	l I	3956 carol@uni.eu	3.0			
	9578 david@uni.eu 1.0		9578 david@uni.eu	1.0			
Lesson 13		Static Ana	vsis for Data	Science		Caterina	a Urban



Extra Values

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In [1]:	<pre>import pandas as pd</pre>					
In [2]:	<pre>df = pd.read_csv('Grades.csv', index_cole df.head()</pre>	In [1]:	<pre>import pandas as pd</pre>			
Out[2]:	Name Q1 Q2 Q3	In [2]:	<pre>df = pd.read_csv('Grades.csv', index_ df.head()</pre>	col=0)		
	ID	Out[2]:	ID Name Q1 Q2 Q3			
	2394 Alice A A A		2394 Alice A A A A			
	4583 Bob F B B		4583 Bob F B B NaN			
	3956 Carol F A C		3956 Carol F A C NaN			
In [3]:	grade2gpa = { 'A': 4.0, 'B': 3.0, 'C': 2	To [2].	9578 David D F C NaN $arade2ana = \begin{cases} A + A + A + B $	• 2 0 'D'• 1 0 'E'• 0 0 }		
	<pre>df.iloc[:, df.columns.str.startswith('Q'</pre>)	df.iloc[:, df.columns.str.startswith('Q')] = df.iloc[:, df.columns.	<pre>str.startswith('Q')].applymap</pre>	p(grade2gpa.get)
In [4]:	<pre>df['Mean'] = df.iloc[:, df.columns.str.s</pre>	t In [4]:	<pre>df['Mean'] = df.iloc[:, df.columns.st</pre>	r.startswith(' <mark>0</mark> ')].mean(axis=1	.)	
In [5]:	<pre>es = pd.read_csv('Emails.csv', index_col</pre>	In [5]:	<pre>es = pd.read_csv('Emails.csv', index_</pre>	col <mark>=0</mark>)		
In [6]:	un = df.join(es)	In [6]:	un = df.join(es)			
In [7]:	<pre>res = un[["Email", "Mean"]] res.head()</pre>	In [7]:	<pre>res = un[["Email", "Mean"]] res.head()</pre>			
Out[7]:	Email Mean	Out[7]:	Email Moon			
	ID		2394 alice@uni.eu 4.0			
2:	2394 alice@uni.eu 4.0		4583 bob@uni.eu 3.0			
	4583 bob@uni.eu 2.0		3956 carol@uni.eu 3.0			
	3956 carol@uni.eu 2.0	•	9578 david@uni.eu 1.0			
	9578 david@uni.eu 1.0					

Lesson 13


Different Format

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	2	■ + % 4	▲ ▲ ► Run ■	Code			
In [1]	: import pandas as pd	In [1]:	<pre>import pandas as pd</pre>				
In [2]	<pre>df = pd.read_csv('Grades.csv', index_col= df.head()</pre>	In [2]:	<pre>df = pd.read_csv('Gra df.head()</pre>	des.csv', index_col=0)			
Out[2]	Name Q1 Q2 Q3	Out[2]:	Name Q1 Q2 Q3				
	ID		ID				
	2394 Alice A A A		2394 Alice A A A				
	4583 Bob F B B		4583 Bob F B+ B				
	3956 Carol F A C		3956 Carol F A C				
	9578 David D F C		9578 David D F C				
In [3]]: grade2gpa = { 'A': 4.0, 'B': 3.0, 'C': 2. df.iloc[:, df.columns.str.startswith('Q')	In [3]:	<pre>grade2gpa = { 'A': 4. df.iloc[:, df.columns</pre>	0, 'B': 3.0, 'C': 2.0, '[.str.startswith('Q')] = 0	D': 1.0, 'F': 0.0 } df.iloc[:, df.columns.	.str.startswith(' <mark>0</mark> ')].applymap	(grade2gpa.get)
In [4]]: df['Mean'] = df.iloc[:, df.columns.str.st	In [4]:	<pre>df['Mean'] = df.iloc[</pre>	:, df.columns.str.starts	with(' <mark>0</mark> ')].mean(axis=1	1)	
In [5]	<pre>es = pd.read_csv('Emails.csv', index_col=</pre>						
In [6]]: un = df.join(es)	In [6]:	un <mark>=</mark> df.join(es)				
In [7]	<pre>res = un[["Email", "Mean"]] res.head()</pre>	In [7]:	<pre>']: res = un[["Email", "Mean"]] res.head()</pre>				
Out[7]	Email Mean	Out[7]:	Email Mean				
	ID		ID				
	2394 alice@uni.eu 4.0		2394 alice@uni.eu 4.0				
	4583 bob@uni.eu 2.0		4583 bob@uni.eu 1.5				
	3956 carol@uni.eu 2.0		3956 carol@uni.eu 2.0				
	9578 david@uni.eu 1.0		9578 david@uni.eu 1.0				
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Data Expectations Static Analysis

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In [1]:	import pandas as po	1			
In [2]:	df = pd.read csv('(rades.csv', index co	1=0)		
	df.head()	rudesresr , index_co	(-0)		
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	Name Q1 Q2 Q	3			
	2394 Alice A A	4			
	4583 Bob F B	3			
	3956 Carol F A	5			
	9578 David D F	3			
In [3]:	grade2gpa = { 'A':	4.0. 'B': 3.0. 'C':	2.0. 'D': 1.0. 'F': 0.0 }		
	df.iloc[:, df.colum	ns.str.startswith('Q	')] = df.iloc[:, df.columns.	str.startswith(' <mark>0</mark> ')].applymap	o(grade2gpa.get)
In [4]:	df['Mean'] = df.ild	c[:, df.columns.str.	<pre>startswith('Q')].mean(axis=1</pre>)	
T- (5).	and and and any fit				
TU [2]:	es = pu.read_csv('	marts.csv , index_co	(
In [6]:	un = df.join(es)				
In [7]:	<pre>res = un[["Email", res.head()</pre>	"Mean"]]			
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	2394 alice@uni.eu 4.0)			
	4583 bob@uni.eu 2.0	1			
	3956 carol@uni.eu 2.0)			





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1st Challenge: Multi-Dimensional Data Structures



Lesson 13

Static Analysis for Data Science







Data Expectations Static Analysis 3-Step Recipe

practical tools targeting specific programs

abstract semantics, abstract domains algorithmic approaches to decide program properties

concrete semantics mathematical models of the program behavior

Lesson 13

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Concrete Semantics 2nd Challenge: Indirect Reasoning



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Abstract Semantics 3rd Challenge: Complex Library Calls

import pandas as pd df = pd.read_csv("HousePrices.csv")

ex = df[df.SalePrice >= 1000000]

ex['Profit'] = ex['SalePrice'] - ex['BuyPrice']

```
:
dL = pd.read_csv("L.csv")
dP = dL.pivot(index=c, columns=y, values=1)
dR = pd.read_csv("R.csv")
dG = dP.loc[:, 0:35].groupby(dR[r])
```

Static Analysis for Data Science





Implementation Wish List



Static Analysis for Data Science



(Un)expected + (Un)used Data



2



Static Analysis for Data Science

Lesson 13

$df -> \{ : W \}$

 $df -> \{ : W \}$ import pandas as pd $df -> \{ : W \}$ df: pd.DataFrame = read_csv(pandas, "...") df -> {"id": N, "t": U, _: N} drop(df, ["id"]) df -> {"t": U, _: N} head(df["t"]) $df -> \{ : N \}$

 $df -> \{ : N \}$







Static Analysis for Data Science

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Lesson 13



Bibliography

[Urban18] Caterina Urban and Peter Müller. An Abstract Interpretation Framework for Data Usage. In ESOP, pages 683-710, 2018. data (non-) usage

[Drobnjaković24] Filip Drobnjaković, Pavle Subotić, and Caterina Urban. An Abstract Interpretation-Based Data Leakage Static Analysis. In TASE, pages 109-126, 2024. data leakage static analysis

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