# Program Synthesis in the Industrial World: Inductive, Incremental, Interactive

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#### PROgram Synthesis using Examples





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We are hiring! Interns or full-time 😳

#### R&D team, MSR → industrial Microsoft

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Simmons



#### Outline

- Programming by Examples (PBE) & PROSE: Quick Background
- Mass-Market Deployment
  - ↔ Goals
    - ↔ Challenges↔ Solutions
- Discussion

# PBE & PROSE

A 3-slide Background

#### Motivation

# 99% of spreadsheet users do not know programming



# Data scientists spend 80% time wrangling raw data



#### **PROSE** Timeline





# Mass-Market Deployment

Goals & Challenges





## **Engineering practices**

- Production-quality library code
  - Prototyping still exists, but it's not the final form
- Unit tests & TDD
- Integration tests: real-life scenarios
  - Close to 8K for all DSLs in total
  - Most are mined from *public* sources (e.g. help forums)
  - In preparation: benchmark suite release for the community



#### Performance-minded engineering

- Parallelization of learning matters
  - *E.g.:* multi-user log file processing in Azure Log Analytics
- Performance of program execution matters
  - *E.g.:* "Big Data" on an end-user's machine
  - Smallest  $\neq$  fastest!
  - (1) Synthesize many correct programs, then (2) optimize for the fast ones

Robustness-based ranking

Performance-based ranking

#### Development

Should I process the string "25-06-11" with regexes? Treat it as a numeric computation? A date?

- DSL design:  $\approx 10$  months  $\rightarrow \approx 2$  weeks
  - This is not a bottleneck!\*
- Ranking: bulk of the effort
  - Designing a score for an operator *F* is 2-3× longer than designing *F* (incl. synthesis!)
  - *E.g.:* rock-paper-scissors among string processing operators

\* Once you learn the skill...



[1] A. Ahmed, A. W. Appel, C. D. Richards, K. N. Swadi, G. Tan, and D. C. Wang. Semantic foundations for typed assembly languages. ACM Trans. Program. Lang. Syst., 32(3), 2010.
[2] A. W. Appel. Program Logics for Certified Compilers. Cambridge University Press, 2014.

From:

...and up to  $10^{20}$  more candidates

all lines ending with "Number • Dot"

"Space • Number • Dot"

starting with "Word • Space • CamelCase"

Extract:

the first "Number" before a "Dot" the last "Number" before a "Dot" the last "Number" before a "Dot • LineBreak" the last "Number" text between the last "Space" and the last "Dot" the first "Comma • Space" and the last "Dot • LineBreak"



• FlashFill was not accepted to Excel until it solved the most common scenarios from 1 example

| Adam Smith     | Adam |
|----------------|------|
| Alice Williams | Alic |

• Some users still don't know you can give 2!



**Option 1:** machine-learned robustness-based ranking

- Idioms/patterns from test data can influence search & ranking
- *E.g.:* bucketing

| 100 | 76-100 |
|-----|--------|
| 51  | 51-75  |
| 86  |        |

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#### **Option 2:** interactive clarification









#### Hypothesizer

Given a program set  $\tilde{N}$ , find program constraints ("hypotheses")  $\varphi$ that best disambiguate among programs in  $\tilde{N}$ , and present them to the user as multiple-choice questions.

- Reduces the cognitive load on the user
- Reduces the number of iterations by choosing the most effective disambiguating questions

Increases the user's confidence in the system ("proactive = smart")

#### Example



#### Example



#### Example



#### Example – alternative



## Picking the right question

"Distinguishability" = effectiveness for disambiguation

- 1. An *input* is distinguishing if many top-ranked candidate programs disagree on the intended output on it.
  - Any response will partition the program set well
- 2. A *question* is distinguishing if the alternative candidate programs corresponding to all potential responses have high ranks.
  - Any response will lead to a good alternative program

Preliminary results: good questions yield just 4-6 iterations until convergence



## **Big Data**



## Big Data + Program Synthesis





#### Problem definition

Given a program set  $\widetilde{N}_i \subset \mathcal{L}$  that satisfies the currently accumulated spec  $\varphi_i$ , and a new constraint  $\psi_{i+1}$ , learn a subset  $\widetilde{N}_{i+1} \subset \widetilde{N}_i$  of programs that satisfy the new spec  $\varphi_{i+1} = \varphi_i \wedge \psi_{i+1}$ 

- $\mathcal{L}$  is an industrial DSL (e.g., FlashFill)
- $\left|\widetilde{N}_{i}\right| \approx 10^{20}$
- Time limit:  $\approx 1 \sec \theta$

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#### VSAs and CFGs are two

isomorphic representations for a language

# $\operatorname{Filter}(\widetilde{N},\psi) \equiv \operatorname{Learn}(\mathcal{L}(\widetilde{N}),\psi)$

#### Incremental Inductive Synthesis

- 1. Implicitly represent  $\widetilde{N}_i$  (already a VSA!) as an isomorphic CFG  $\mathcal{L}(\widetilde{N}_i)$ .
- 2. Analyze the *descriptive power* of  $\psi_{i+1}$ :
  - **Definitive** (*e.g.*, examples, set membership, subsequence constraints): Apply regular top-down deductive synthesis on  $\mathcal{L}(\widetilde{N}_i)$
  - Locally refining (e.g., datatypes, input relevance): Re-run backpropagation only on relevant parts of  $\mathcal{L}(\widetilde{N}_i)$
  - Globally refining (e.g., negative examples): Filter  $\widetilde{N}_i$  at the top level



#### Preliminary results



- Big improvement when VSA fragmentation is limited
- Not the final results; work in progress has orders-of-magnitude improvements

#### Lessons & Conclusions

- Robust engineering is the key to PBE deployment
- Ranking >> learning (in industrial PBE)
- Interaction models should be first-class citizens in synthesis frameworks
  - Great theoretical results: e.g., OGIS [Jha & Seshia 2015]
  - Also need: HCI evaluations, comparison of query types, worst-case TD optimization
- Proactive ambiguity analysis of current candidate programs
- Incrementality: treat the previous program set as the new search space
  - Requires *full* program set computation (possibly in the background)

Thank you!

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