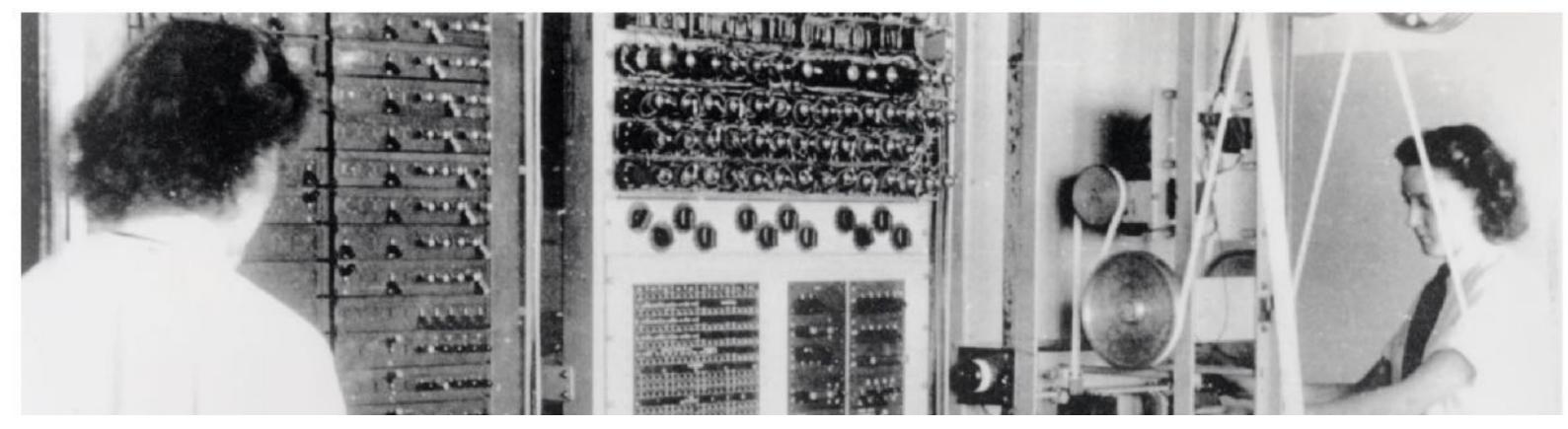
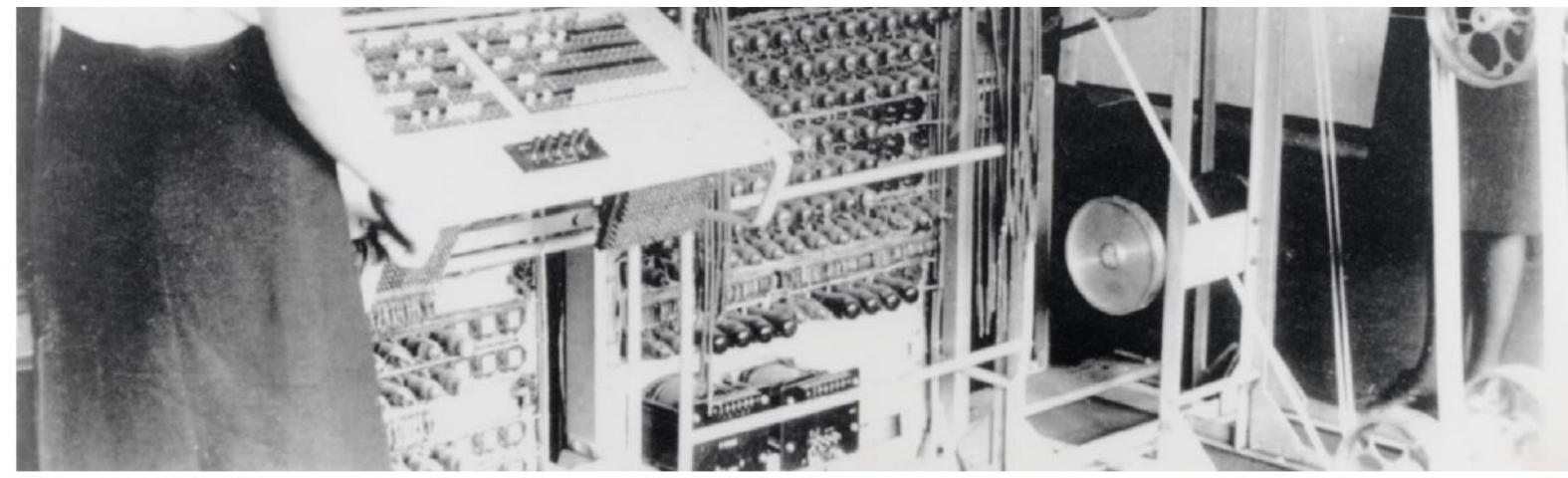


### PL/HCI Seminar (252R/279R) MW 12-1:15 PM in LISE 303 Pierce 209



### Making communicating with computers more accessible: easier, faster, and safer



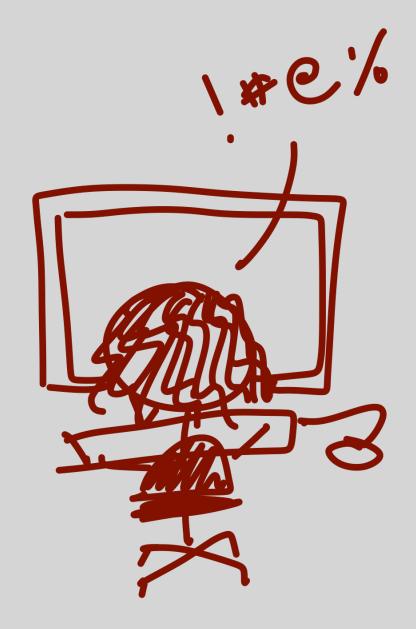
Harvard John A. Paulson School of Engineering and Applied Sciences

- This is a graduate course; undergrads are welcome.
  - can have taken 152 or 179 and be just fine, not necessarily both
- You (as a student presenter) will present and lead discussion for at least one paper
- You (as a non-presenting student) will post questions and a summary of the design arguments by Friday of the previous week
- Key learning outcomes:
  - (279r) to look at scientific publications, identify the core design arguments, write new design arguments, and evaluate them
  - (252r) understand, design and implement language abstractions for solving a task
- Group projects will be composed of both "HCI folks" and "PL folks"

## Welcome!

- In undergraduate courses, you consume knowledge and practice applying it.
- In graduate courses like this one, you attempt to generate new knowledge.

## Welcome!



## l'm an HCI person.



### I build novel interfaces and evaluate them in studies.





### Since ~2015, I have found PL technology useful for providing the magic behind the screen.



PL/HCI Seminar (252R/279R)

### **Evaluation is Hard.** But also, evaluation with respect to what?

Harvard John A. Paulson School of Engineering and Applied Sciences

### **1. Design Arguments**

- 2. Some HCI evaluation techniques
- 3. Means of communicating with computers

## Outline

## Design Arguments

### **Need Thesis**

Stakeholders + Domain Person P [in setting S] Core tension wants to achieve goal G but obstacles O<sub>1-N</sub> get in the way. **Evidence** 

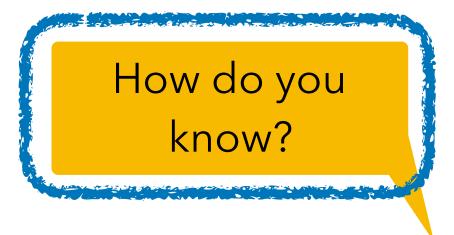
> Any solution also has to: satisfy constraints **X<sub>1-N</sub>**, minimize costs **Y<sub>1-N</sub>**, and avoid obstacles **Z<sub>1-N</sub>**.

Axioms As designers, we bring the following principles and constraints **A<sub>1-N</sub>**.

Approach Thesis Our approach, \_\_\_\_\_\_ has characteristics C<sub>1-N</sub>

that help stakeholders achieve their goal **G** while avoiding obstacles **O**<sub>1-N</sub>

### Evidence



Need



### **Evidence**

What characteristics have you borrowed from solutions that succeeded in analogous settings?

What differentiates your approach from previous solutions that failed?

How have stakeholders responded to/been able to use your approach?



### Visualizing API Usage Examples at Scale

Elena L. Glassman<sup>‡</sup>\*, Tianyi Zhang<sup>||</sup>\*, Björn Hartmann<sup>‡</sup>, Miryung Kim<sup>||</sup> <sup>‡</sup>\*UC Berkeley, Berkeley, CA, USA "\*UC Los Angeles, Los Angeles, CA, USA {eglassman, bjoern}@berkeley.edu, {tianyi.zhang, miryung}@cs.ucla.edu

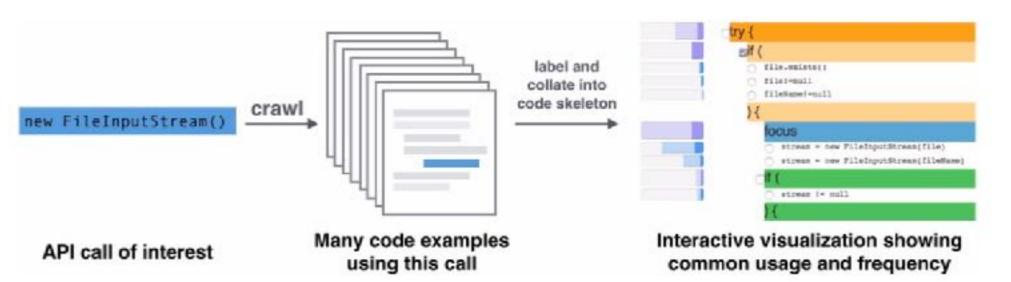


Figure 1. EXAMPLORE takes a focal API call that a programmer is interested in, locates uses of that API call in a large corpus of mined code examples, and then produces an interactive visualization that lets programmers explore common usage patterns of that API across the corpus.

### ABSTRACT

Using existing APIs properly is a key challenge in programming, given that libraries and APIs are increasing in number and complexity. Programmers often search for online code examples in Q&A forums and read tutorials and blog posts to learn how to use a given API. However, there are often a massive number of related code examples and it is difficult for a user to understand the commonalities and variances among them, while being able to drill down to concrete details. We introduce an interactive visualization for exploring a large collection of code examples mined from open-source repositories at scale. This visualization summarizes hundreds of code examples in one synthetic code skeleton with statistical distributions for canonicalized statements and structures enclosing an API call. We implemented this interactive visualization for a set of Java APIs and found that, in a lab study, it helped users (1) answer significantly more API usage questions correctly and comprehensively and (2) explore how other programmers have used an unfamiliar API.

### INTRODUCTION

Learning how to correctly and effectively use existing APIs is a common task - and a core challenge - in software development. It spans all expertise levels from novices to professional software engineers, and all project types from prototypes to production code. The landscape of publicly available APIs is massive and constantly changing, as new APIs are created in response to shifting programmer needs. Within companies, the same is true, perhaps even more so: joining a company can require learning a whole new set of proprietary APIs before a developer becomes an effective contributor to the company codebase. Developers often are stymied by various learning barriers, including overly specific or overly general explanations of API usage, lack of understanding about the interaction between multiple APIs, lack of alternative uses, and difficulty identifying program statements and structures related to an API [11, 19, 5].

One study found that the greatest obstacle to learning an API is "insufficient or inadequate examples." [19] Official documenta-

Figure 1. EXAMPLORE takes a focal API call that a programmer is interested in, locates uses of that API call in a and then produces an interactive visualization that lets programmers explore common usage patterns of that API

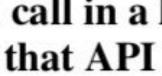
### ABSTRACT

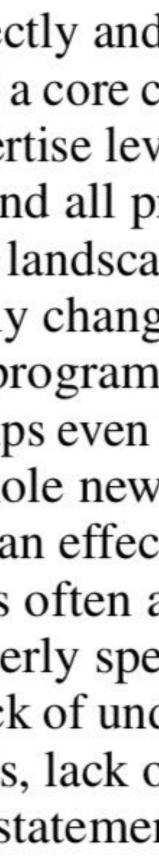
Using existing APIs properly is a key challenge in programming, given that libraries and APIs are increasing in number and complexity. Programmers often search for online code examples in Q&A forums and read tutorials and blog posts to learn how to use a given API. However, there are often a massive number of related code examples and it is difficult for a user to understand the commonalities and variances among them, while being able to drill down to concrete details. We introduce an interactive visualization for exploring a large collection of code examples mined from open-source repositories at scale. This visualization summarizes hundreds of code examples in one synthetic code skeleton with statistical distributions for canonicalized statements and structures enclosing an API call. We implemented this interactive visualization for a set of Java APIs and found that, in a lab study, it helped users (1) answer significantly more API usage questions correctly and comprehensively and (2) explore how other programmers have used an unfamiliar API.

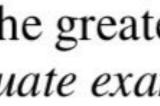
### INTRODUCTION

Learning how to correctly and a common task — and a core c ment. It spans all expertise lev software engineers, and all p production code. The landsca massive and constantly change response to shifting program the same is true, perhaps even require learning a whole new a developer becomes an effec codebase. Developers often a barriers, including overly spe tions of API usage, lack of und between multiple APIs, lack of identifying program statement API [11, 19, 5].

One study found that the great "insufficient or inadequate exa







- 1. Design Arguments
- 3. Means of communicating with computers

## Outline

### 2. Some HCI formative design and evaluation techniques

### Some HCI Formative Design Techniques

- Survey
- Interview
- Contextual inquiry
  - Observation in context
  - Requests for explanation
- Wizard of Oz
- Technology probe

## Some HCI Evaluation Techniques

- User study
  - Task design
  - Metrics
- Deployment
- Interview or survey of deployment participants
- Crowdsourcing, i.e., Amazon Mechanical Turk

- 1. Design Arguments
- 3. Means of communicating with computers

## Outline

# 2. Some HCI formative design and evaluation techniques

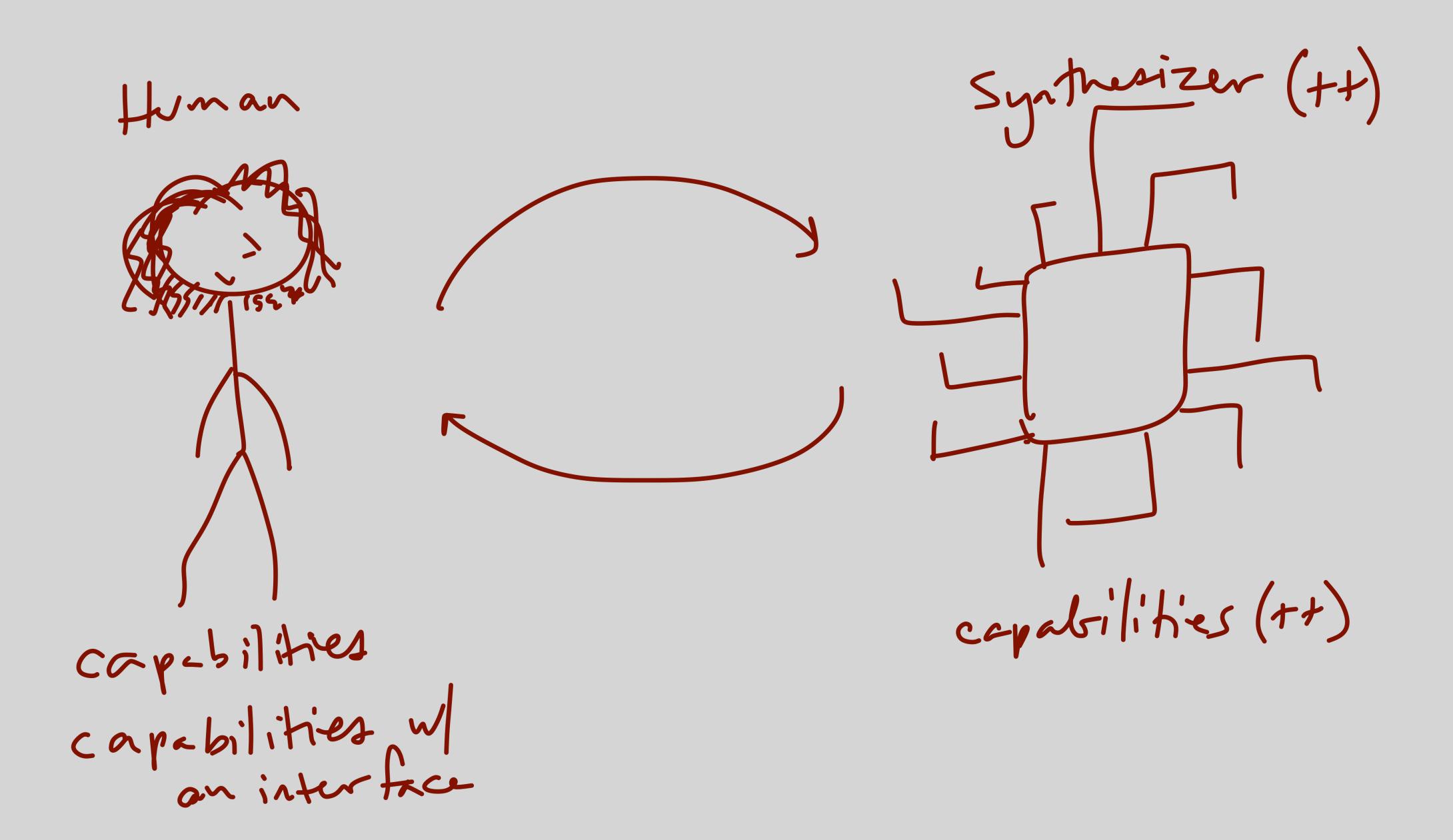
- Human intent
  - Examples
  - Statement(s) in a programming language
  - Natural language

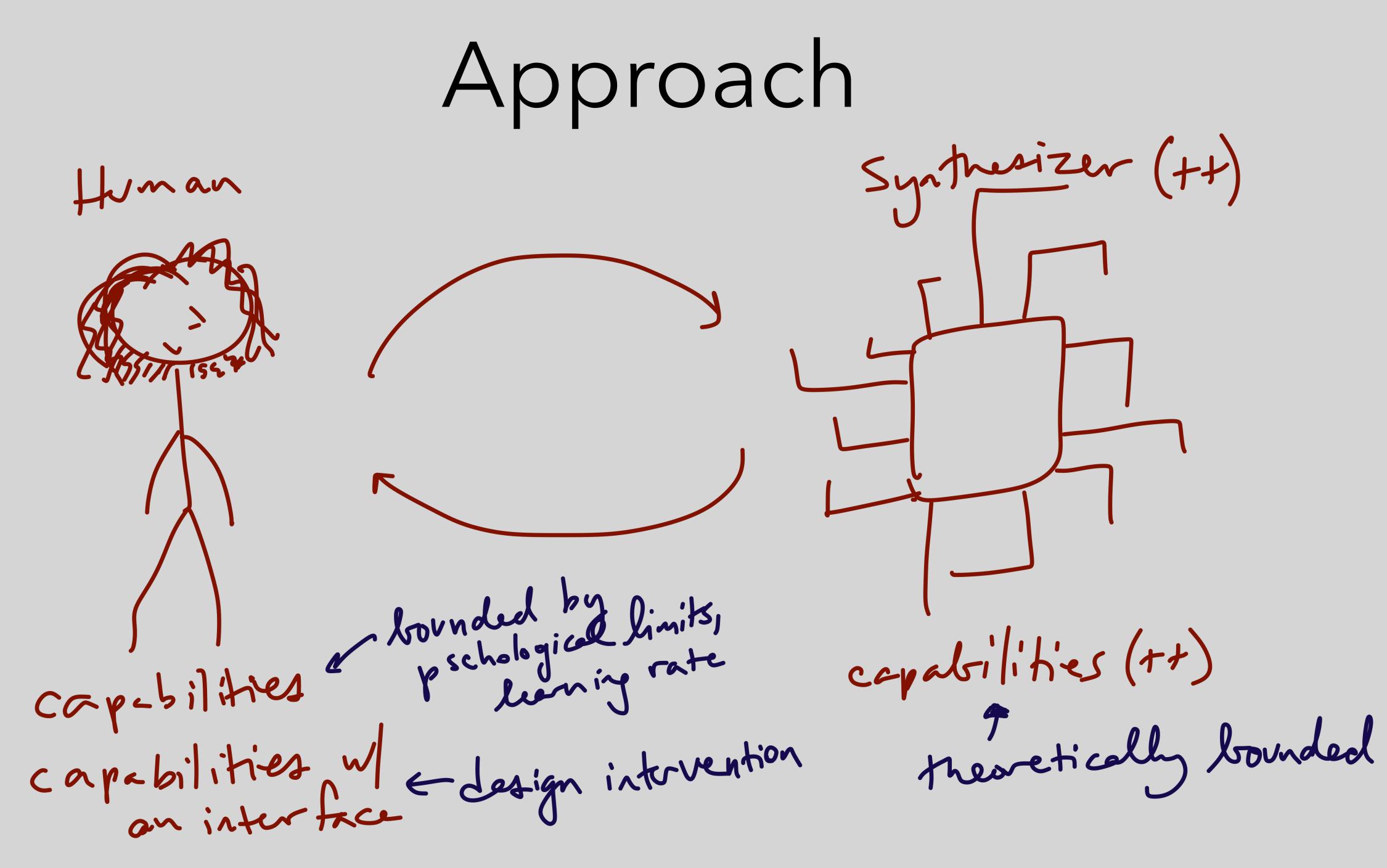
• ....

- Computer's interpretation
  - Program
  - Behavior
    - Action in response to a human request

## Communicating with Computers

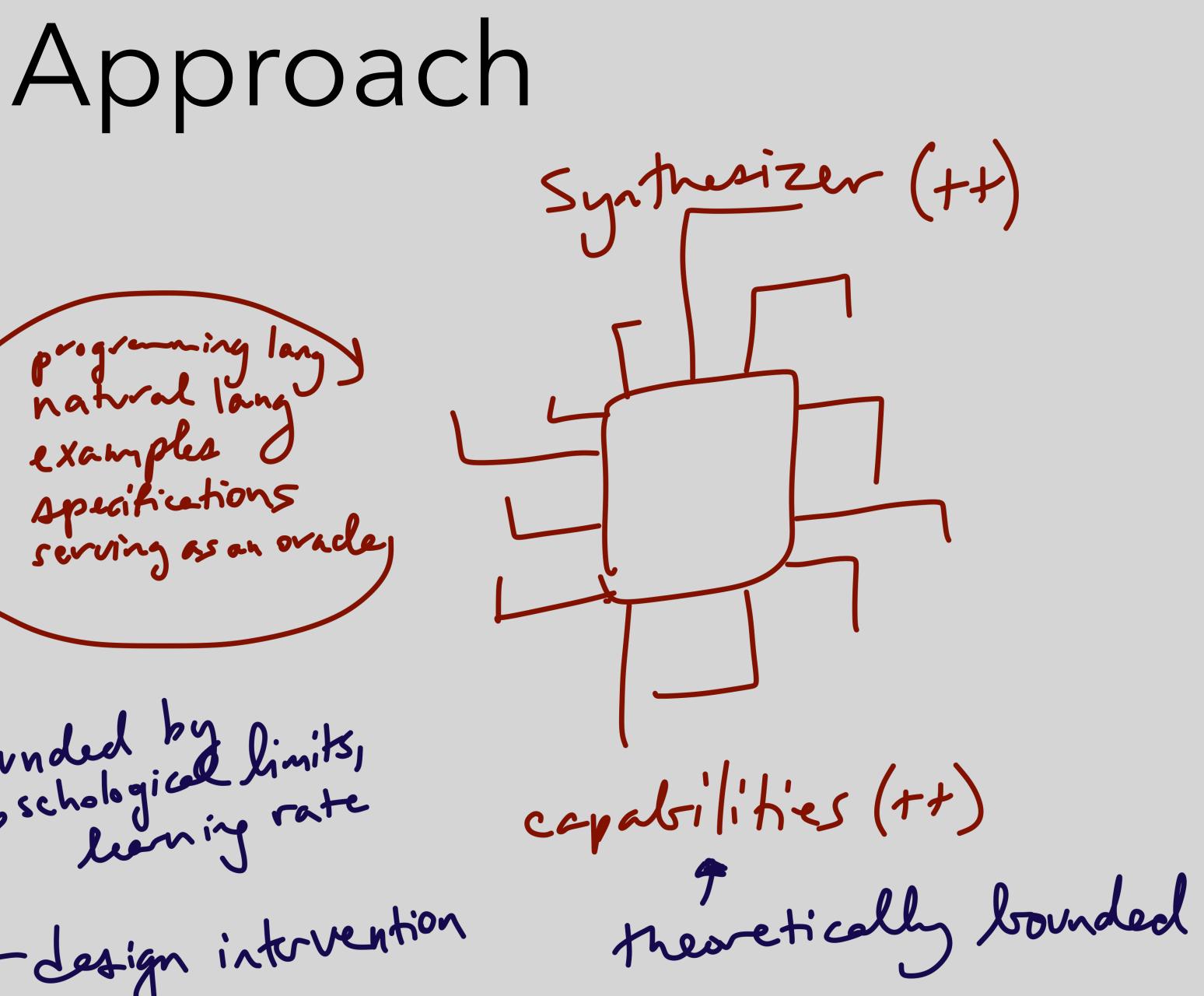
Results of running understood program on additional data



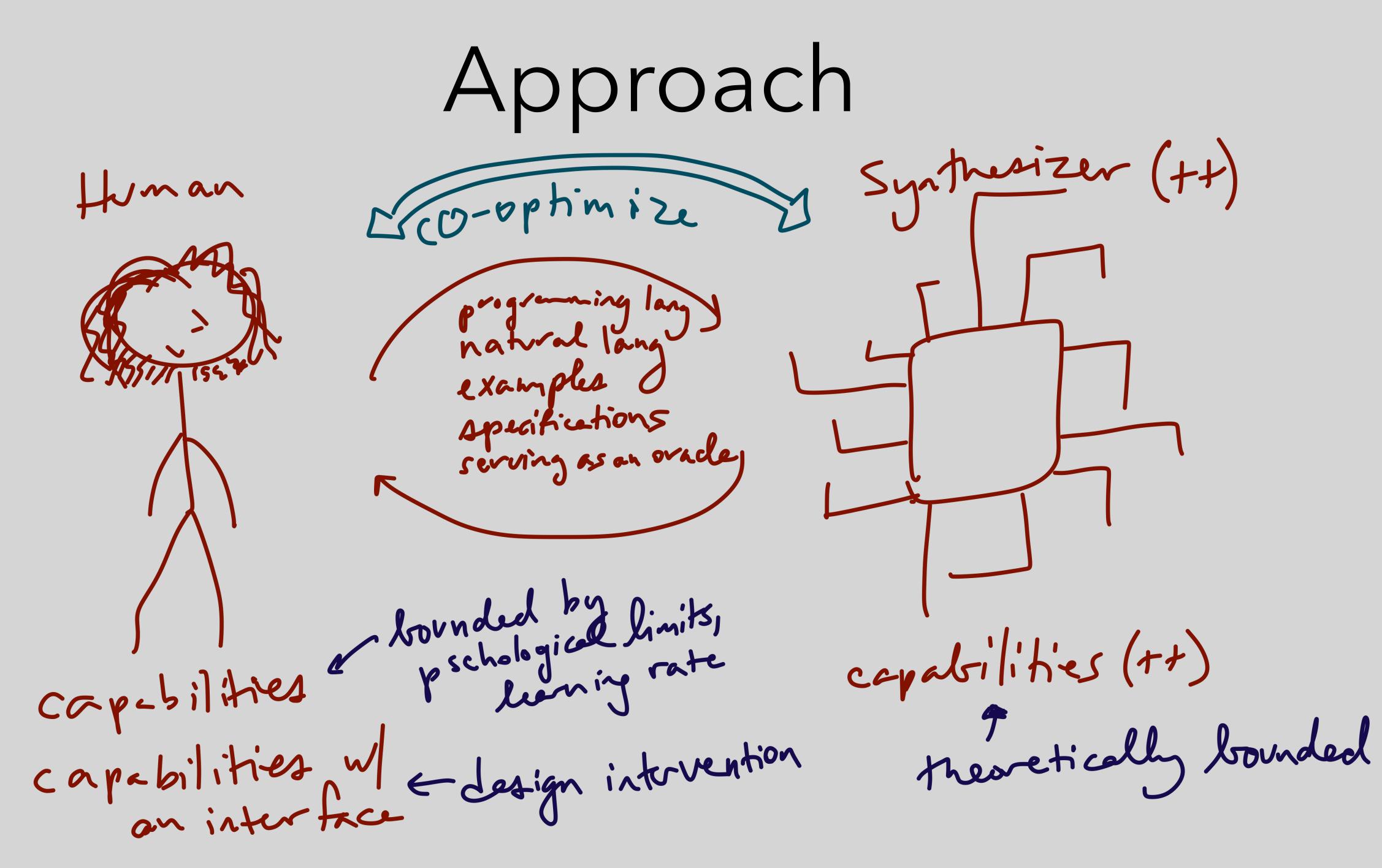




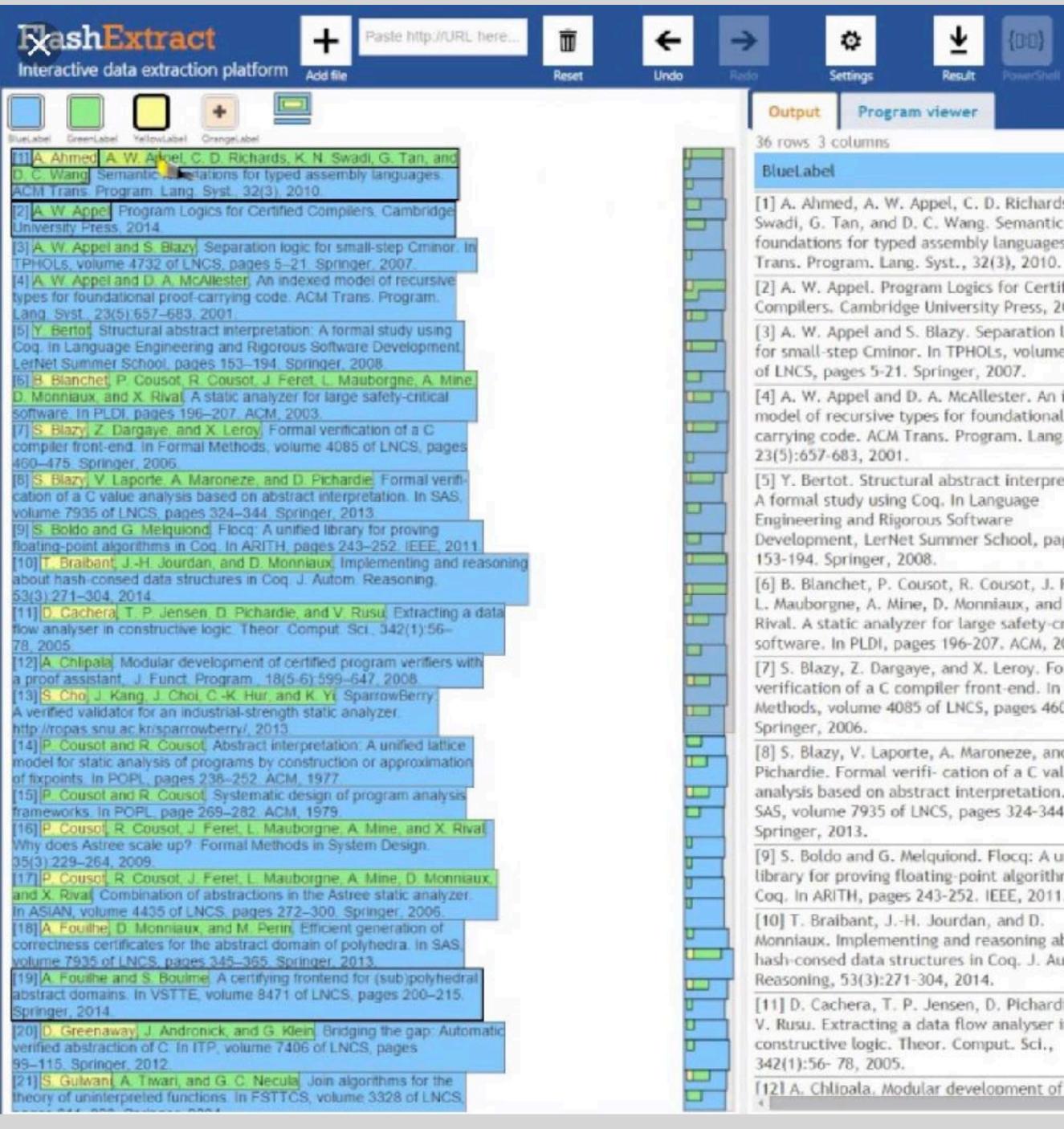
Lunan examples specifications serving as an oracle bounded by linits, pschological linits, leaving rate Capebilities capabilities n'e design intervention on interface







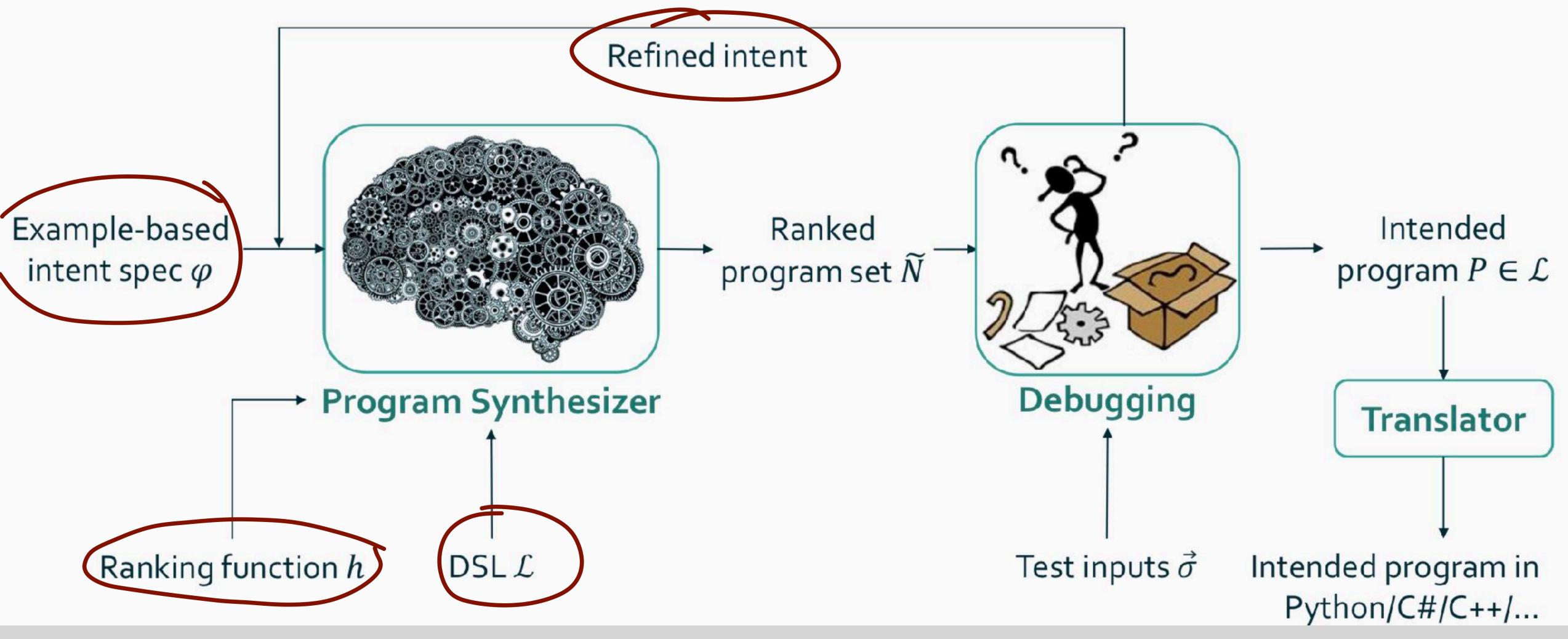




	GreenLabel	YellowLabel
ppel, C. D. Richards, K. N. C. Wang. Semantic assembly languages. ACM Syst., 32(3), 2010.	A. Ahmed, A. W. Appel, C. D. Richards, K. N. Swadi, G. Tan, and D. C. Wang	A. Ahmed
am Logics for Certified University Press, 2014.	A. W. Appel	<null></null>
Blazy. Separation logic In TPHOLs, volume 4732 pringer, 2007.	A. W. Appel and S. Blazy	<null></null>
A. McAllester. An indexed bes for foundational proof- ans. Program. Lang. Syst.,	A. W. Appel and D. A. McAllester	<null></null>
al abstract interpretation: oq. In Language ous Software Summer School, pages 08.	Y. Bertot	<null></null>
usot, R. Cousot, J. Feret, p. D. Monniaux, and X. r for large safety-critical es 196-207. ACM, 2003.	B. Blanchet, P. Cousot, R. Cousot, J. Feret, L. Mauborgne, A. Mine, D. Monniaux, and X. Rival	B. Blanchet
e, and X. Leroy. Formal piler front-end. In Formal of LNCS, pages 460-475.	S. Blazy, Z. Dargaye, and X. Leroy	S. Blazy
e, A. Maroneze, and D. fi- cation of a C value ract interpretation. In NCS, pages 324-344.	S. Blazy, V. Laporte, A. Maroneze, and D. Pichardie	S. Blazy
lquiond. Flocq: A unified ating-point algorithms in 243-252. IEEE, 2011.	5. Boldo and G. Melquiond	<null></null>
Jourdan, and D. ng and reasoning about ctures in Coq. J. Autom. 304, 2014.	T. Braibant, JH. Jourdan, and D. Monniaux	T. Braibant
Jensen, D. Pichardie, and data flow analyser in eor. Comput. Sci.,	D. Cachera, T. P. Jensen, D. Pichardie, and V. Rusu	D. Cachera
lar development of		

Log off				
	Î			
	+			
3				

## PROSE Architecture



**Credit: Alex Polozov** 

"FlashFill in Excel is designed to cater to users that care not about the program but about its behavior on the small number of input rows in the spreadsheet.

Such users can simply eye-ball the outputs of the synthesized program and provide another example if they are incorrect.

However, this becomes much more cumbersome (or impossible) with a larger spreadsheet."

	Α	В	С
1	First Name	Last Name	Full Name
2	Jay	Shasthri	Jay Shasthri
3	Pratap	Pillai	Pratap Pillai
4	Madhu	Srivastava	Madhu Srivastava
5	Victoria	Marsh	Victoria Marsh
6	David	Pizarro	David Pizarro
_			

support.office.com

- Lu et al. "Interactive Program Synthesis" (2017)



"We have observed that inspecting the synthesized program directly also does not establish enough confidence in it even if the user knows programming.

Two main reasons for this are (i) program readability, and (ii) the users' uncertainty in the desired intent due to hypothetical unseen corner cases in the data."

	А	В	С
1	First Name	Last Name	Full Name
2	Jay	Shasthri	Jay Shasthri
3	Pratap	Pillai	Pratap Pillai
4	Madhu	Srivastava	Madhu Srivastava
5	Victoria	Marsh	Victoria Marsh
6	David	Pizarro	David Pizarro
_			

support.office.com

- Lu et al. "Interactive Program Synthesis" (2017)



## Discussion Preview

### What is particularly hard about evaluating methods for communicating with computers?



PL/HCI Seminar (252R/279R)

### Evaluation Why is it hard: some puzzlers from programming languages

Harvard John A. Paulson **School of Engineering** and Applied Sciences

## puzzler 1 easy vs safe

## Array<Cat>

# <: Array<Animal>

7

## puzzler 2 semantics

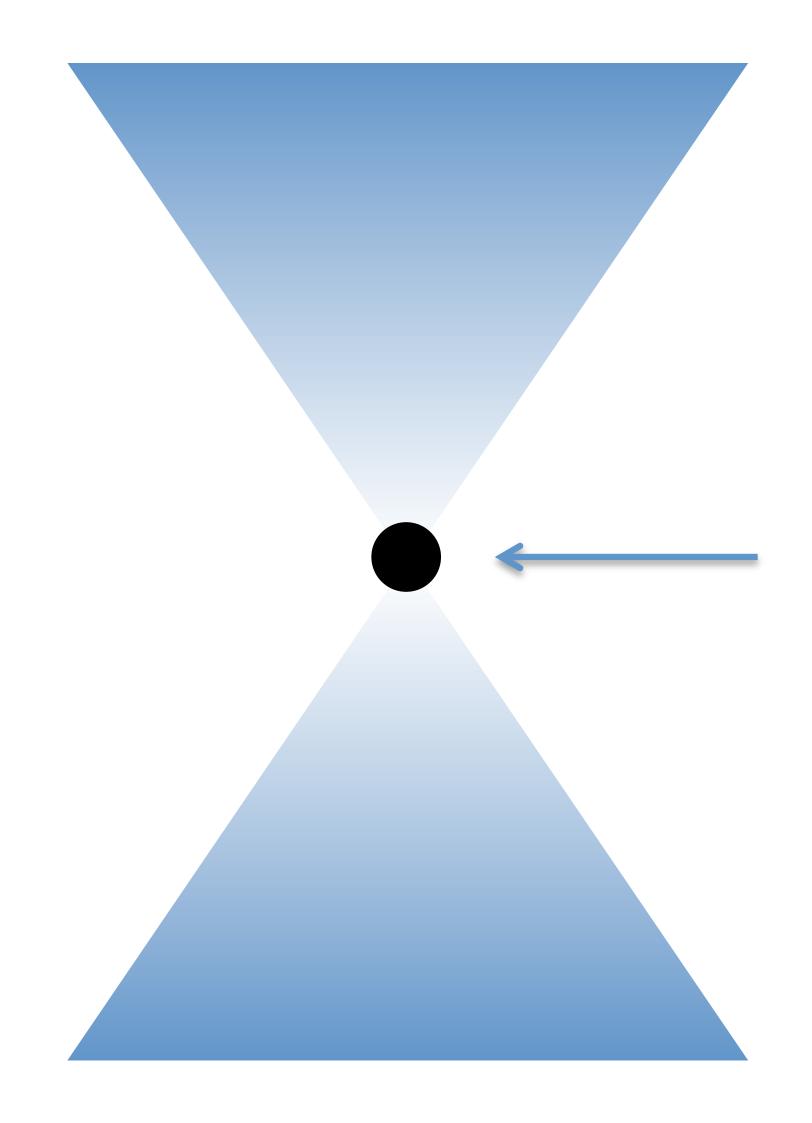
• let x = 1 in let x = 3 in f 0

• Is the result 1, 2 or 3?

### let f = let x = 2 in $(y \rightarrow y+x)$

## puzzler 3 cognitive overhead

### Programmer

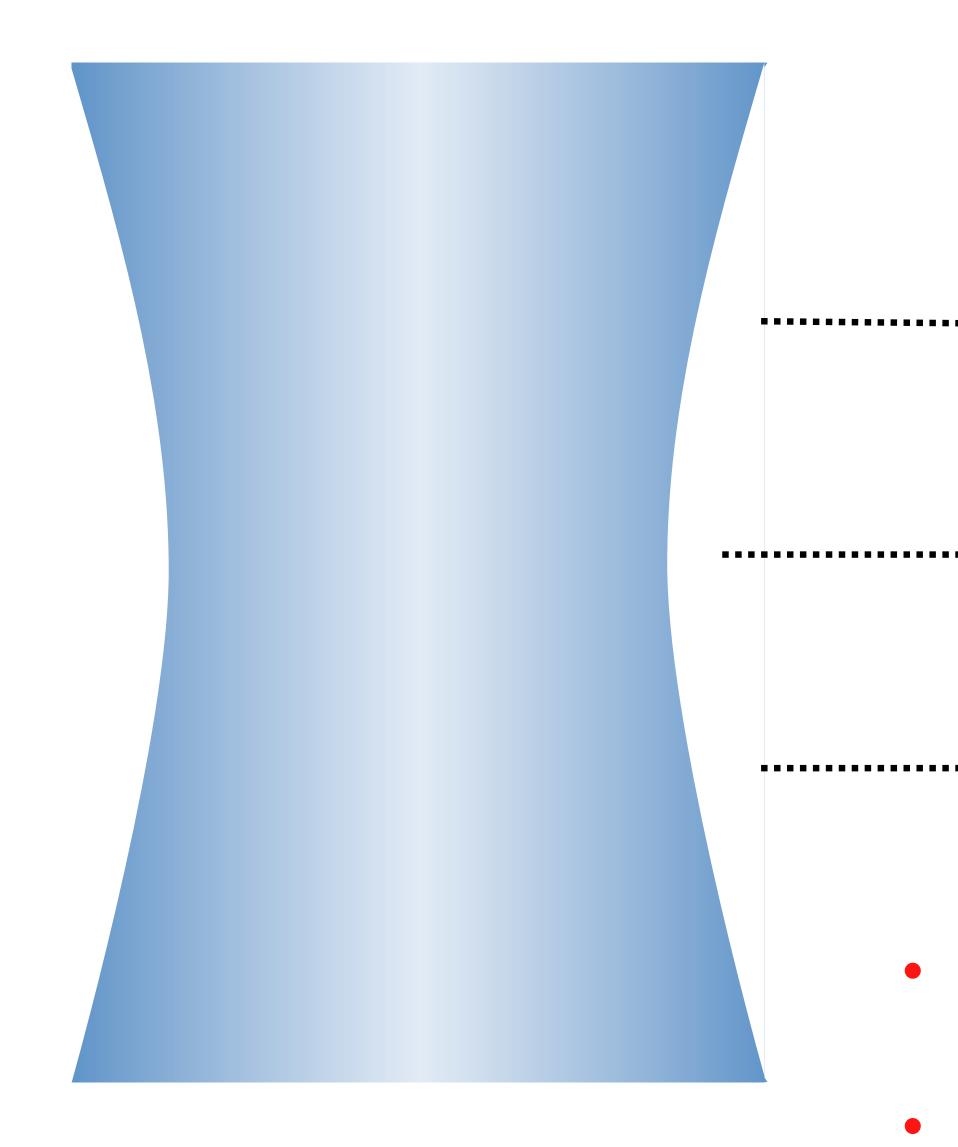


Hardware

### general purpose compiler

(illustration: Markus Püschel)

### Programmer



### Hardware

..... Matrix, Graph, ...

Array, Struct, Loop, ...

SIMD, GPU, cluster, ...

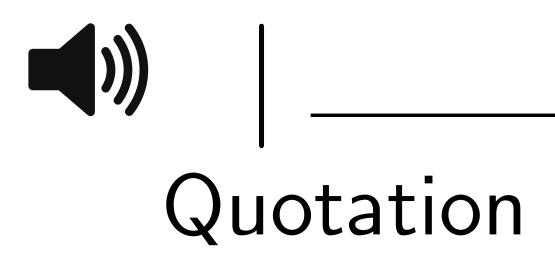
horizontal and vertical
extensibility
generic optimizations
at each level (cse, dce, ...)

## Staging?

- multi-level language n | x | e  $\mathbb{Q}^{b}$  e |  $\lambda^{b}$  x e | ...
- •MetaML / MetaOCaml n | x | e e |  $\lambda x \cdot e$  |  $\langle e \rangle$  |  $\langle e \rangle$  | run e
- driven by types: T vs Rep[T]

# Lightweight Modular Staging (LMS) in Scala

"People confuse the familiar for the simple. For new features, people insist on LOUD explicit syntax. For established features, people want terse notation."



-Bjarne Stroustrup

Type-Based



PE

- def square(x: Int): Int = x\*x
- def power(b: Int, n: Int): Int = if (n == 0) 1 else if (n % 2 == 0) square(power(b, n/2)) else b \* power(b, n-1)
- // power(2, 7) == 128

## Power in Scala



def square(x: Rep[Int]): Rep[Int] = x\*x

def power(b: Rep[Int], n: Int): Rep[Int] = if (n == 0) 1else if (n % 2 == 0) square(power(b, n/2)) else b \* power(b, n-1)

def snippet(b: Rep[Int]) = power(b, 7)

## Staged Power in Scala/LMS

### Generated Power n=7

class Snippet extends ((Int)=>(Int)) {
 def apply(x0:Int): Int = {
 val x1 = x0 \* x0
 val x2 = x0 \* x1
 val x3 = x2 \* x2
 val x4 = x0 \* x3
 x4
 }
}

- let square x = x \* x
- let rec power n x =
  - if n = 0 then 1

  - else x \* (power (n-1) x)
- (\* val power : int -> int -> int = <fun> \*)

### Power in OCaml

# else if n mod 2 = 0 then square (power (n/2) x)

- let square x = x \* x
- let rec spower n x =
  - if n = 0 then .<1>.

  - else .<.x \* .~(spower (n-1) x).
- (\* val spower : int -> int code -> int code = <fun> \*)

### Staged Power in MetaOCaml

else if n mod 2 = 0 then .<square .~(spower  $(n/2) \times$ )>.

### Generated Code

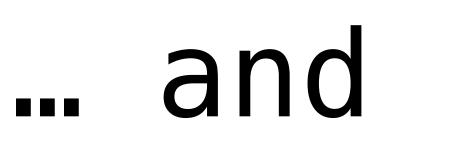
- let spower7\_code = .<fun x -> .~(spower 7 .<x>.)>.;;(\*
- val spower7\_code : (int -> int) code = .<</pre>

fun x\_1 ->

x\_1 \* ((\* CSP square \*) (x\_1 \* ((\* CSP square \*) (x\_1 \* 1)))>.

\*)

### Program generically...





### ... and run specialized!

# Program generically... (save human time)

# ... and run specialized! (save computer time)



A matrix vector product, where the matrix is known (static) but the vector is unknown (dynamic),

e.g., for a Hidden Markov Model (HMM) where a single transition matrix is multiplied by many different observation vectors.

-Shonan Challenge for Generative Programming (PEPM'13)

def matrix\_vector\_prod(a: Array[Array[Int]],

val n = a.length val v1 = new Array[Int](n) for (i <- (0 until n)) {</pre> for (j <- (0 until n)) {</pre> v1(i) = v1(i) + a(i)(j) \* v(j)} } v1

### Unstaged

### v: Array[Int]) = $\{$

def matrix\_vector\_prod(a0: Array[Array[Int]],

val n = a0.length val a = staticData(a0)val v1 = NewArray[Int](n) for (i <- (0 until n):Range) {</pre> val sparse = a0(i).count(\_ != 0) < 3 for (j <- unrollIf(sparse, 0 until n)) {</pre> v1(i) = v1(i) + a(i).apply(j) \* v(j)

# Shonan Challenge

- v: Rep[Array[Int]]) = {

def matrix\_vector\_prod(a0: Array[Array[Int]],

val n = a0.length val a = staticData(a0)val v1 = NewArray[Int](n) for (i <- (0 until n):Range) {</pre> val sparse = a0(i).count(\_ != 0) < 3 for (j <- unrollIf(sparse, 0 until n)) {</pre> v1(i) = v1(i) + a(i).apply(j) \* v(j)

# Shonan Challenge

- v: Rep[Array[Int]]) = {

### Example Matrix

- val a0 =
  - A(A(1, 1, 1, 1, 1), // dense
    - A(0, 0, 0, 0, 0), // null
    - A(0, 0, 1, 0, 0), // sparse
    - A(0, 0, 0, 0, 0),
    - A(0, 0, 1, 0, 1))

### Generated Code

class Snippet(px6:Array[Int]) extends ((Array[Int])=>(Array[Int])) { val x6 = px6 // static data: Array(1,1,1,1,1) var x4 : Int = 0 val x13 = while (x4 < 5) { val x5 = x2(0); val x7 = x6(x4); val x8 = x0(x4) val x9 = x7 \* x8; val x10 = x5 + x9;val x11 = x2(0) = x10 x4 = x4 + 1val x14 = x2(1); val x17 = x2(1) = x14 val x22 = x2(2); val x24 = x2(2) = x22 val x19 = x0(2); val x25 = x22 + x19 val x26 = x2(2) = x25; val x27 = x2(3); val x29 = x2(3) = x27 val x30 = x2(4); val x32 = x2(4) = x30 val x33 = x30 + x19; val x34 = x2(4) = x33 val x21 = x0(4); val x35 = x33 + x21; val x36 = x2(4) = x35 x2 }}

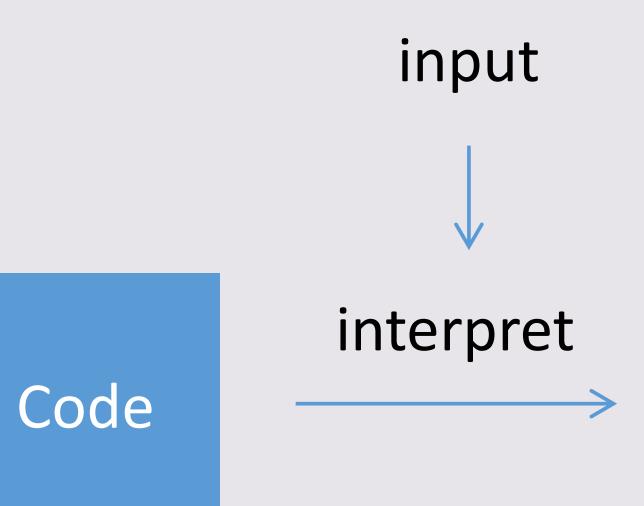
def apply(x0:Array[Int]): Array[Int] = { val x2 = new Array[Int](5)

### Turning Interpreters into Compilers

### Turning Interpreters into Compilers





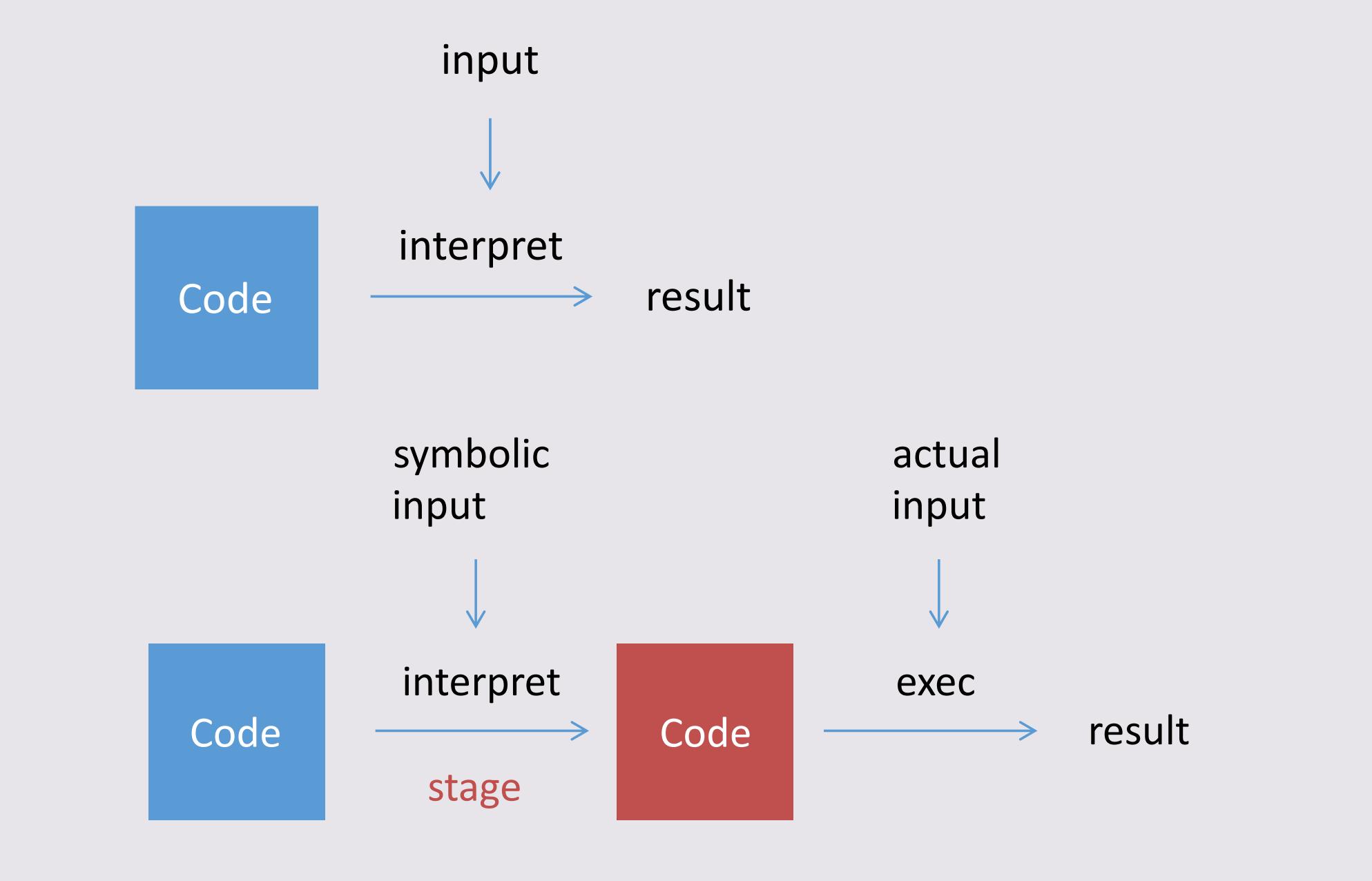




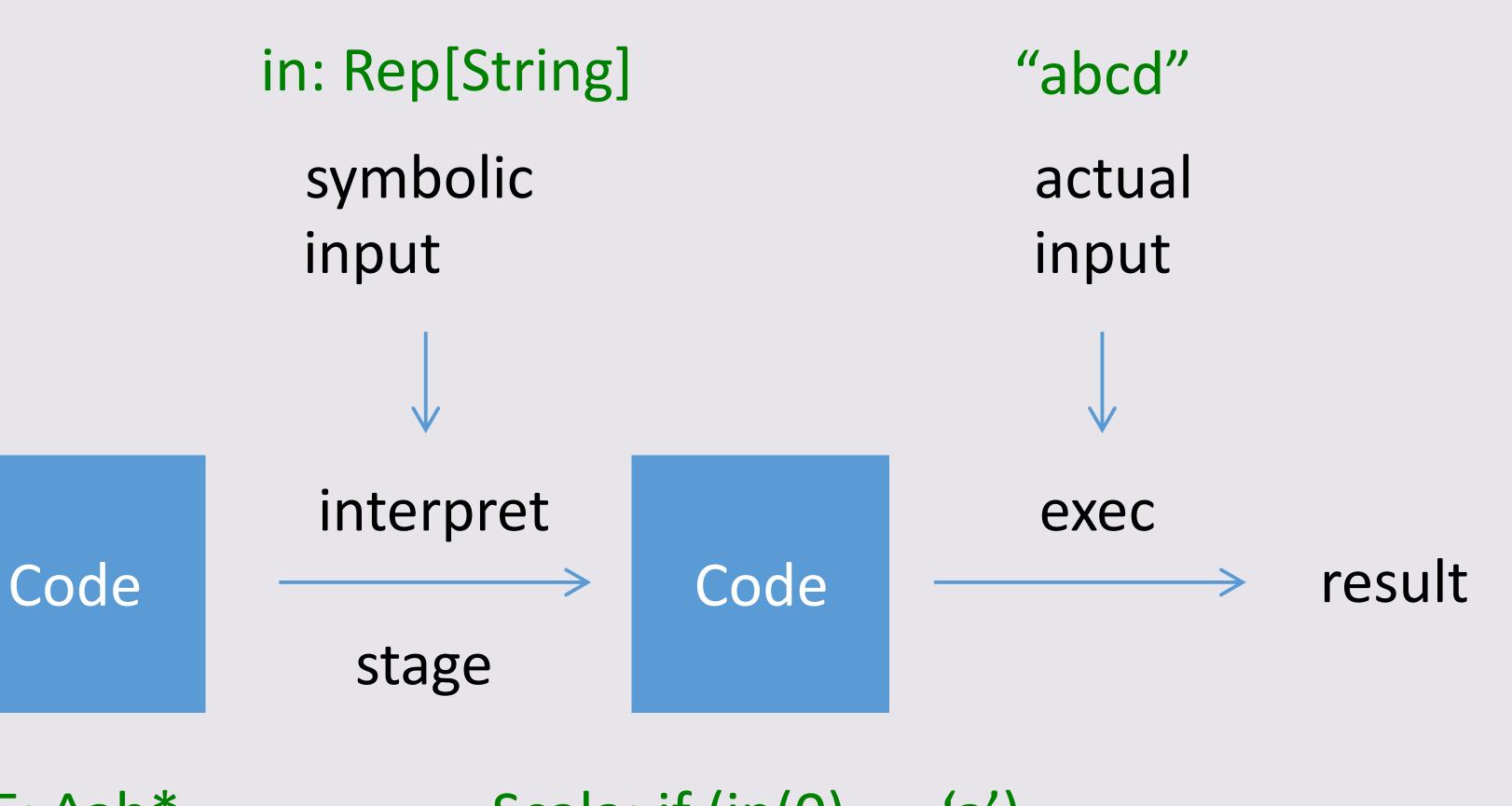
 $\rightarrow$ 

### result





### Turning interpreters into compilers



RE: ^ab\*

Scala: if (in(0) == 'a') ...

### Usability Issues

- cognitive overhead of multi-level language
- notational overhead?
- multi-stage errors

puzzler 4
synthesis

## Programming by Example

- f(1)=2, f(2)=3, ...
- g(1)=2, g(2)=4, ...
- h(1)=3, h(2)=5, ...

### SMT solver underspecified

- (assert (and (= (f 1) 2) (= (f 2) 3))
- Model: f(x) = if x=2 then 3 else if x=2 then 3 else 2
- Try it: <u>https://rise4fun.com/Z3/bpMYx</u>

- (assert (forall ((x Int) (=> (or (and (= x 1) (= (and (= x 2) (=(= y (+ (\* a x) b))
- Solution: a=1, b=1, so f(x)=x+1.
- Try it: <u>https://rise4fun.com/Z3/HcJY</u>

### SMT solver overspecified

(declare-const a Int) (declare-const b Int) (define-fun f ((x Int)) Int (+ (\* a x) b))(assert (and (= (f 1) 2) (= (f 2) 3)))(check-sat) (get-model)

• Try it: <u>https://rise4fun.com/Z3/jhr8</u>

# SMT solver full program

## Usability Issues

- Need to be precise induces cognitive overhead.
- Precision is brittle.
- Many ways to encode.

### Trade-offs, e.g. between expressivity and decidability.

# Why evaluation is hard

- Many conflicting dimensions: faster, safer, easier
- Does a new toolkit enable new ways
  - of thinking?
  - of programming?
  - of creating?

### Discussion

### What is particularly hard about evaluating methods for communicating with computers?

- Systems HCI requiring heavy-duty PL
  - Humans modifying DSLs for PBD (programming by demonstration)
  - Examplore with interactively defined templates
- Generic human-centered PL
  - Pick language feature, design it in a human-friendly way
  - Pick a language, describe how—and to what extent-its features are being used in the wild

## Group Projects

- Usable + X (PL technique)
  - Usable Generative Programming
  - Usable Probabilistic Programming
  - Usable Type System / Verification
  - Usable Synthesis
    - inductive bias alignment between human and machine
    - ranking function improvements
    - DSL improvements
    - expressing constraints on intermediate states, i.e., equivalence values or types

