Web Question Answering with Neurosymbolic Program Synthesis

Jocelyn Chen¹, Aaron Lamoreaux¹, Xinyu Wang², Greg Durrett¹, Osbert Bastani³, Isil Dillig¹

¹The University of Texas at Austin

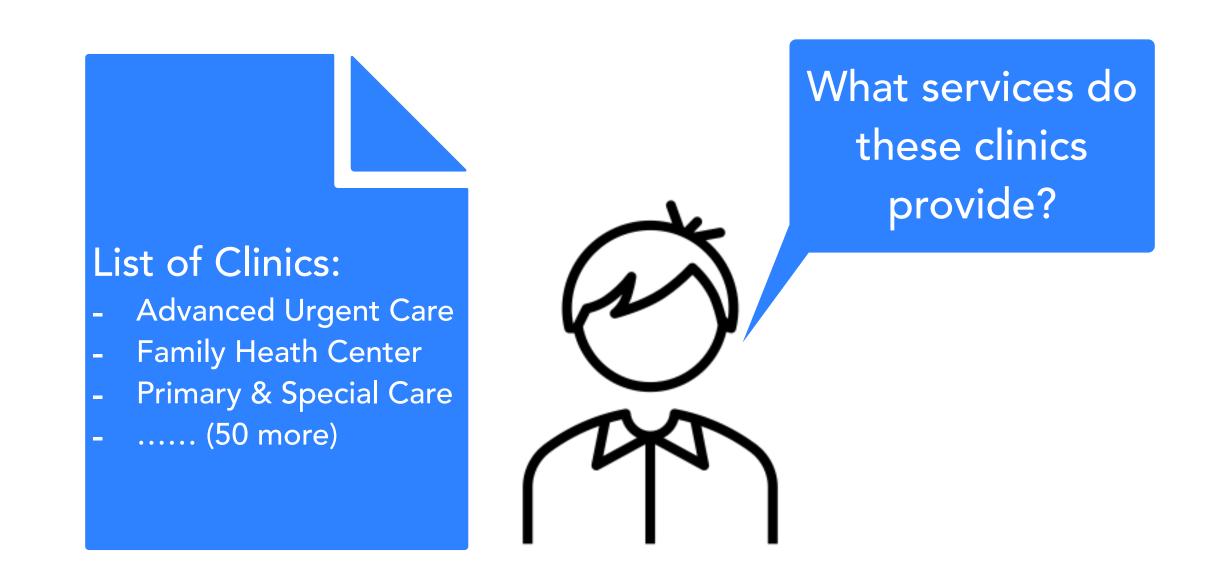
²University of Michigan

³University of Pennsylvania



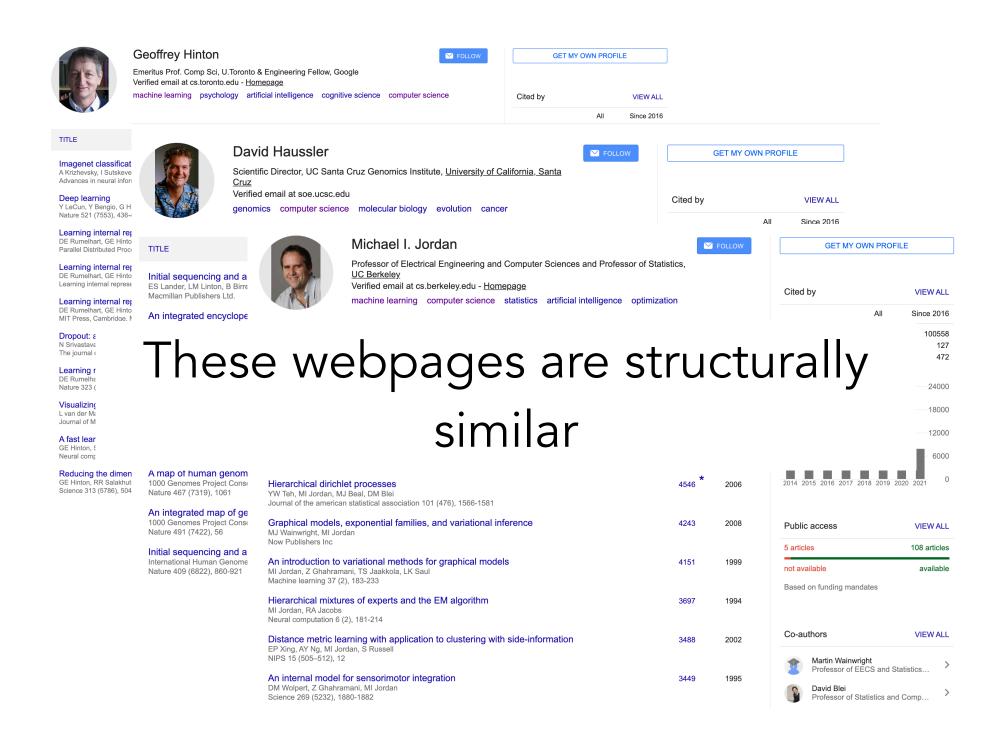
Motivation

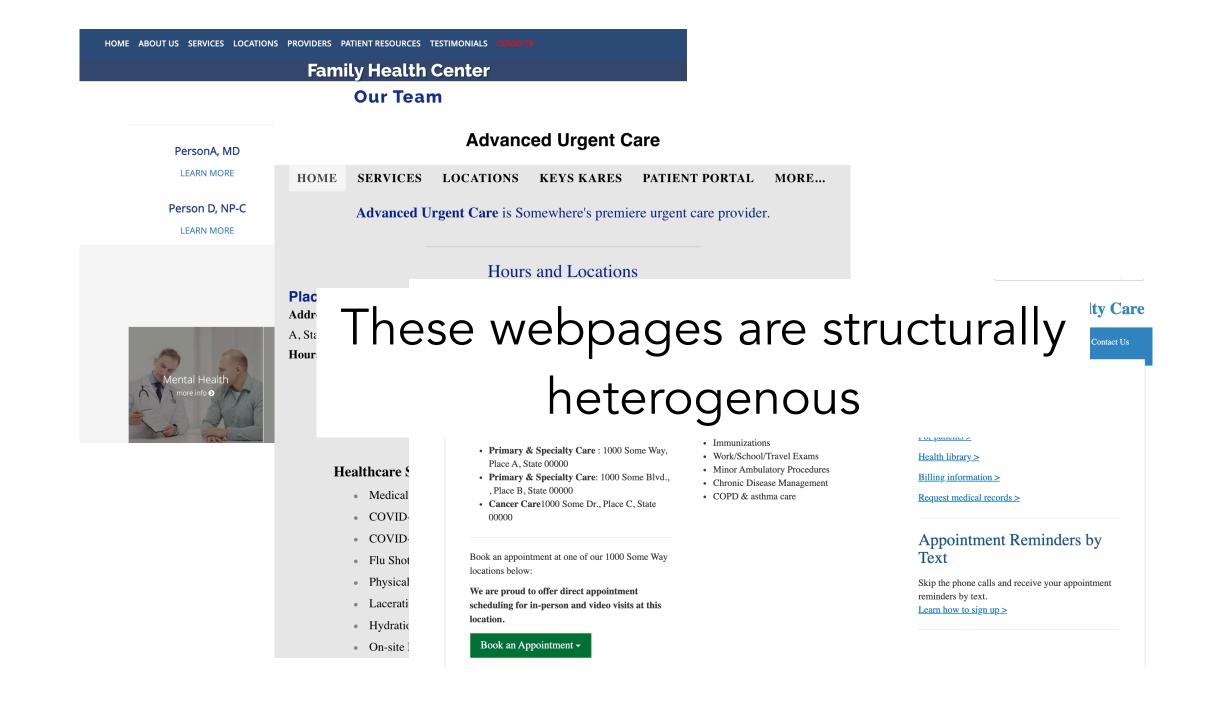
This work: automatically extract information from a large set of websites.



Possible Solution: Wrapper Induction/automatic scraping

Extract information from structurally similar webpages





Possible Solution: Question Answering

Passage Sentence

In meteorology, precipitation is any product of the condensation of atmospheric water vapor that falls under gravity

Question

What causes precipitation to fall?

Answer Candidate

Gravity

Answer questions from plain-text documents

Passage Sentence

Advanced Urgent Care. Home Service Locations ... 9AM - 3PM SAT. Service offered Healthcare Services Medical Exams Covid Testing...

Question

What services does clinic provides?

Answer Candidate

Healthcare Services

Websites are not plain-text documents

Our Solution: WebQA

Targets at structurally heterogeneous websites with no global schemas

Handles diverse schemas while able to reason about the content

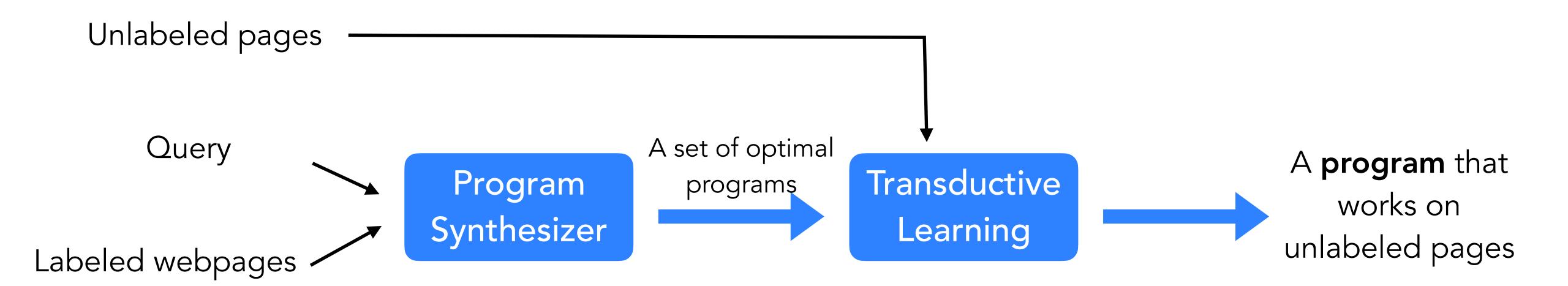
Key idea



Better reasoning about the content

Handle the tree structure of the webpage

WebQA Workflow

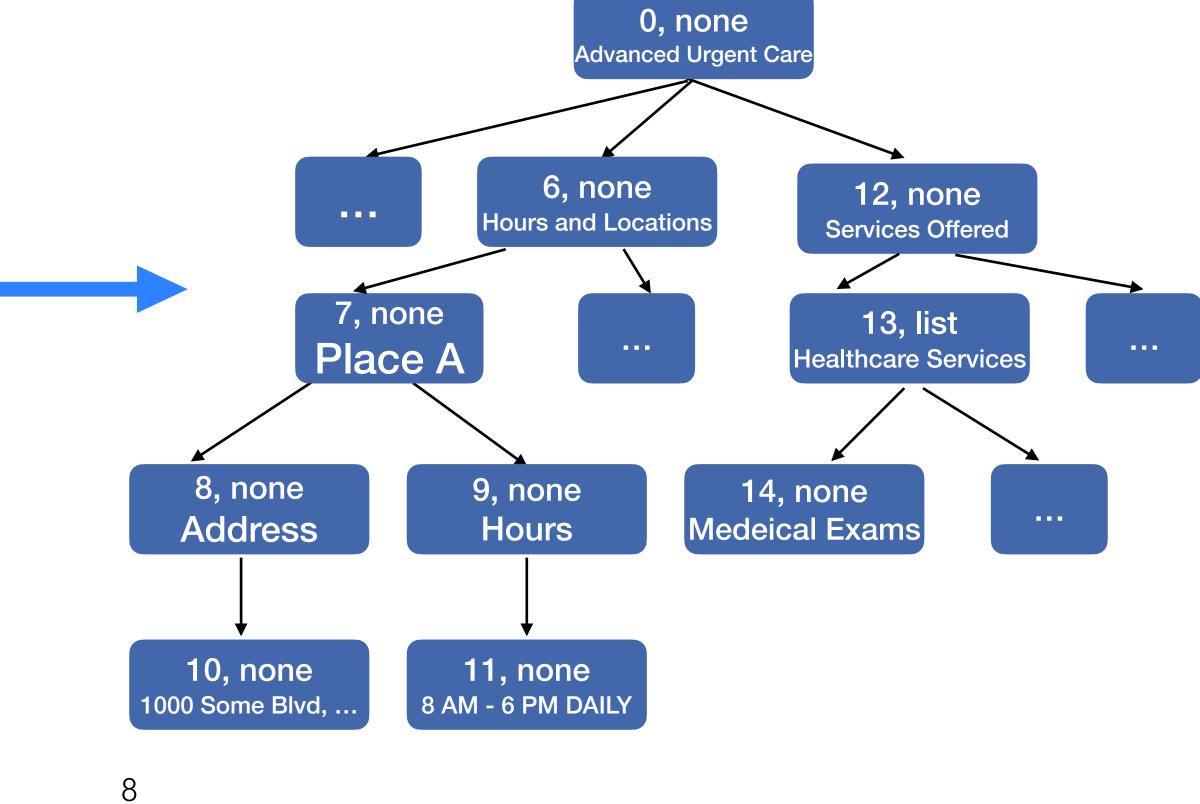


Webpage

A tree that captures the relationship between text elements on the rendered webpage

Advanced Urgent Care





WebQA Program Workflow

Input:

Q (question), K (keyword), W (webpage)

Top-level Program:

$$\lambda Q, K, W. \{\psi_1 \rightarrow \lambda x. e_1, \dots, \psi_n \rightarrow \lambda x. e_n\}$$

Guard

Identify schema and locate relevant section

Extractor

Extract information in the located section under this schema

Constructs in Guards and Extractors

GetChildren(v, p)

GetDescendents(v, p)

Substring(t, p)

Split(t, c)

containsKeyword(t, K)

hasAnswer(t, Q)

hasEntity(t, PERSON)

Tree Navigation

Given a node v, get its children/descendent that satisfy a predicate p

String Processing

Given a text t, get its

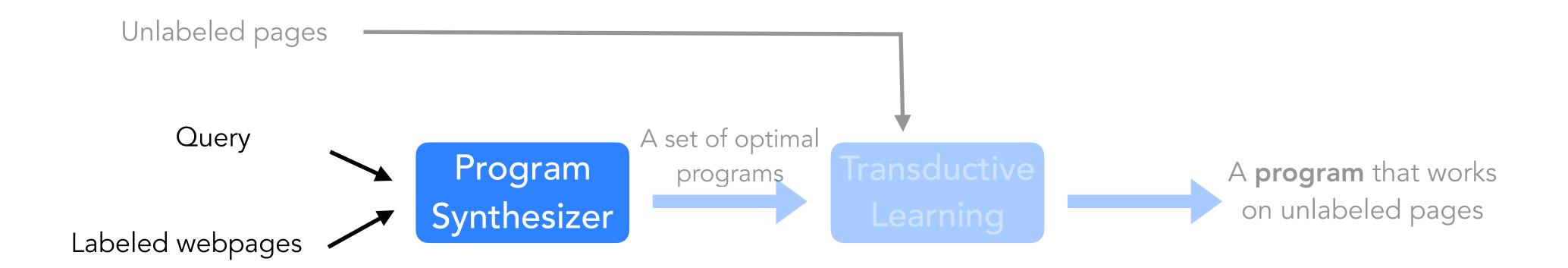
- substrings that satisfy some predicate p
- split substrings using delimiter c

Neural Components

Given a text t, check if it

- contains answers to questions Q
- contains words similar to keywords K
- contains entity such as person

WebQA Workflow



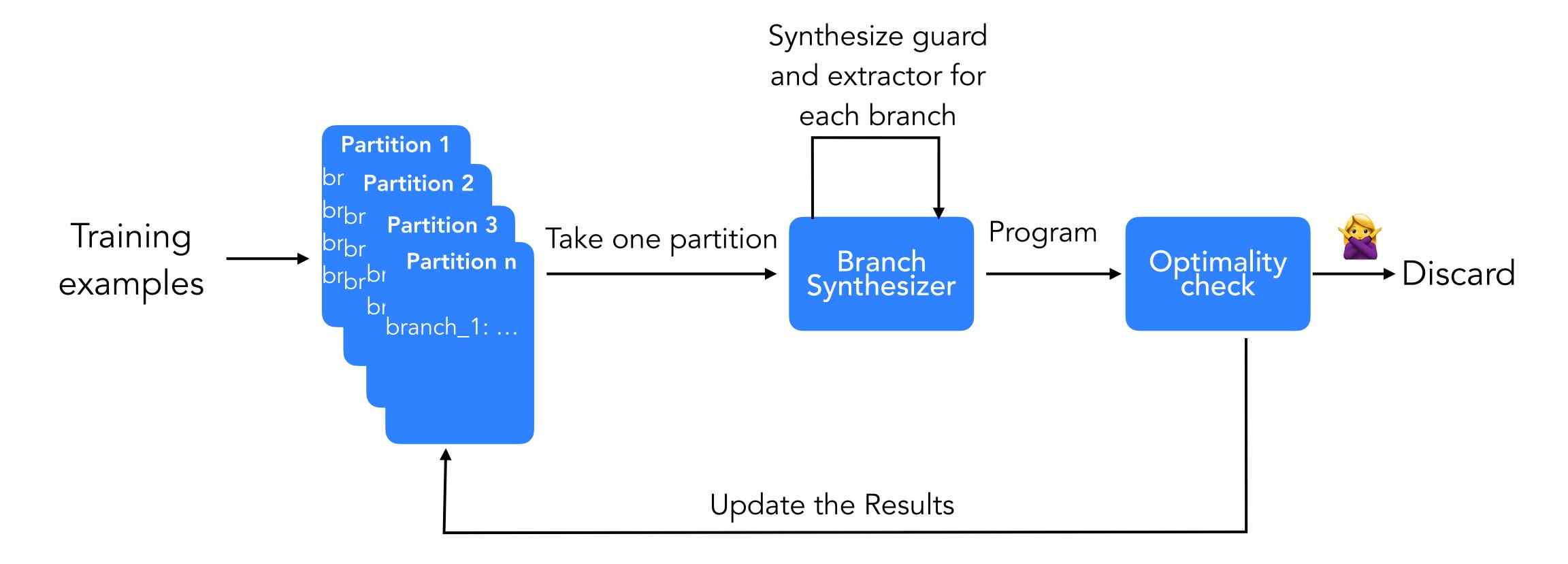
Generate **a set of programs** that **extract correct information** from labeled webpage

Difficult to get exact match

Generate a set of programs achieve the highest F1 score on the labeled webpage

$$F_1 = \frac{2 \cdot \text{Precision} \cdot \text{Recall}}{\text{Precision} + \text{Recall}}$$

Optimal Synthesis Workflow



Yields programs with highest F1 score on labeled training examples

Highlight of our Synthesis Technique

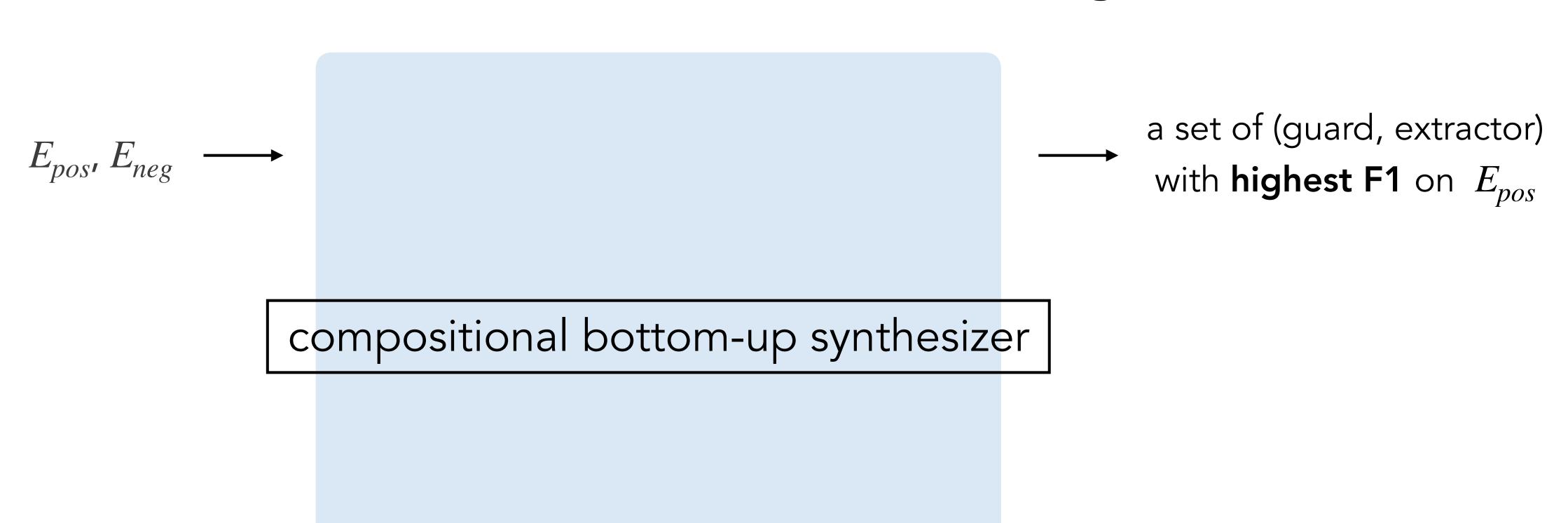
Decomposition

separate extractor synthesis and guard synthesis

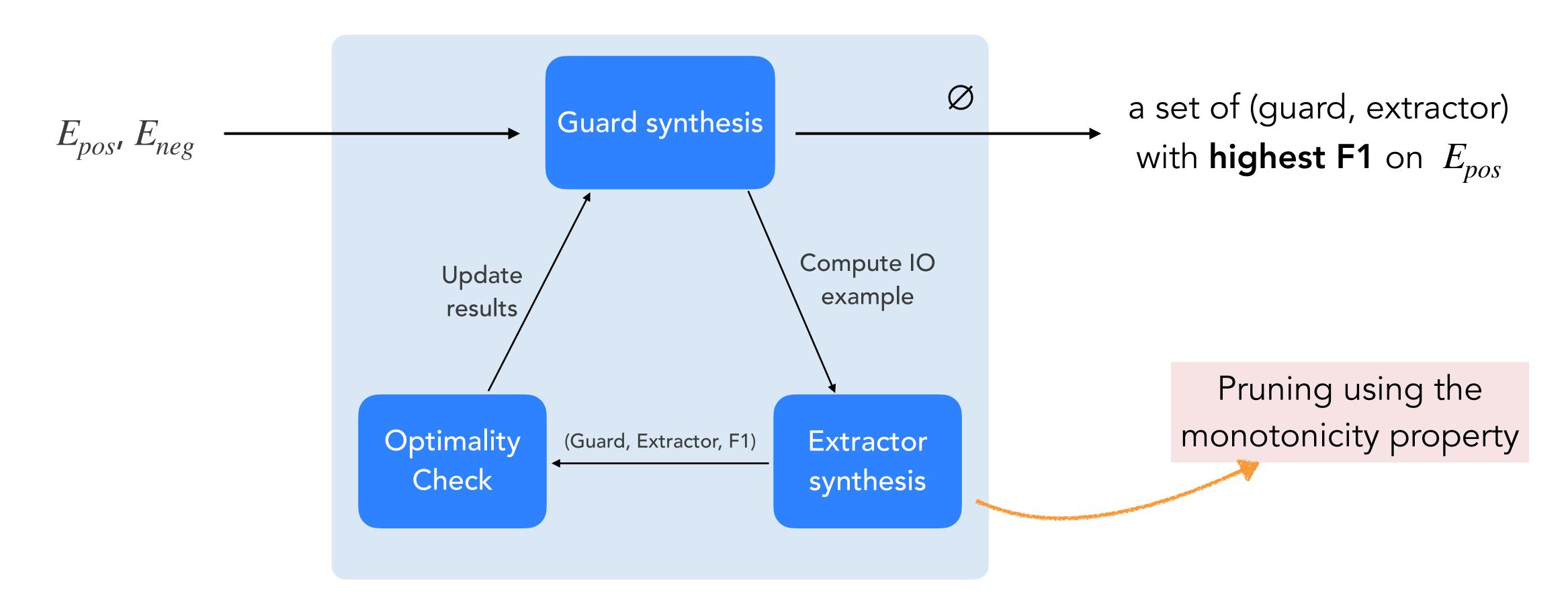
Pruning

using a monotonicity property of the DSL with respect to recall

Interface of Branch Synthesis



Branch Synthesis Workflow



Decomposition reduces the number of (guard, extractor) pairs!

Pruning

Monotonicity property of the DSL:

Let e, e' be two extractors, if e' is a sub-expression of e, then $Recall(e) \le Recall(e')$

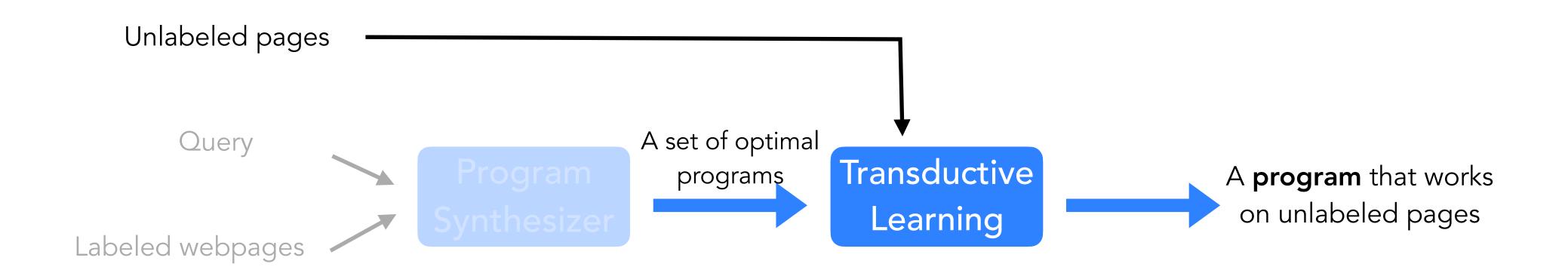
 F_1 Upper bound for any program uses e as sub-expression on example E:

$$UB(e, E) = \frac{2 \cdot 1 \cdot Recall(e, E)}{1 + Recall(e, E)}$$

precision = 1 (best case)

We prune out any expression e if $UB(e, E) < current_best_F1$

WebQA Workflow



- Working with websites of *diverse* schemas: **not all** optimal programs on the label webpages **work well** on unlabeled webpages
- Performance of randomly chosen program varies a lot

Given a **large** set of **optimal programs**, use the **unlabeled data** to select **a program** that **works well to the unlabeled data** with **low variance**

Program selection via transductive learning

Transductive learning: use unlabeled data to help obtain predictions

Key idea: Ensemble of optimal programs

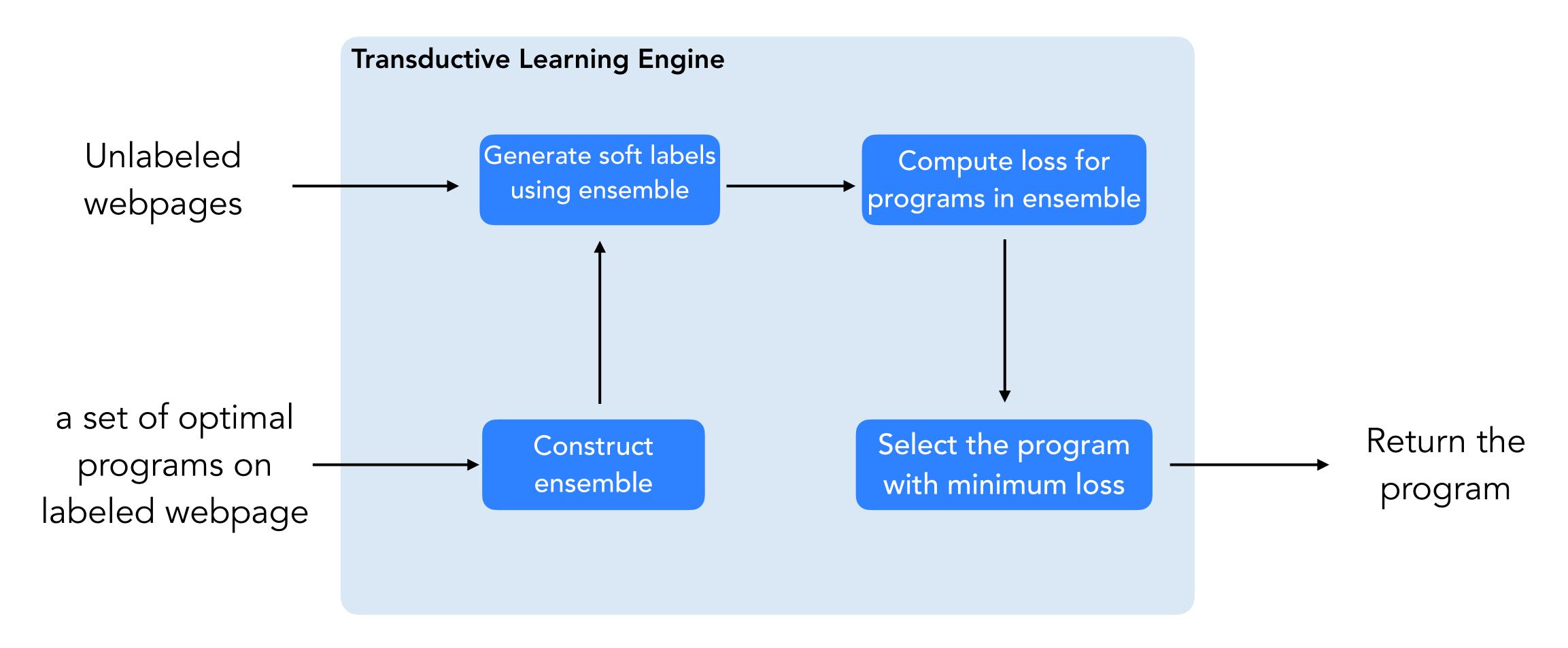
- Aggregate predication over samples of the optimal programs
- Generalize better than a single program with lower variance

Return the ensemble

- Ensemble is less interpretable
- Ensemble is expensive to compute

Select **one program** whose outputs are **most similar** to the ensemble

Program selection via transductive learning

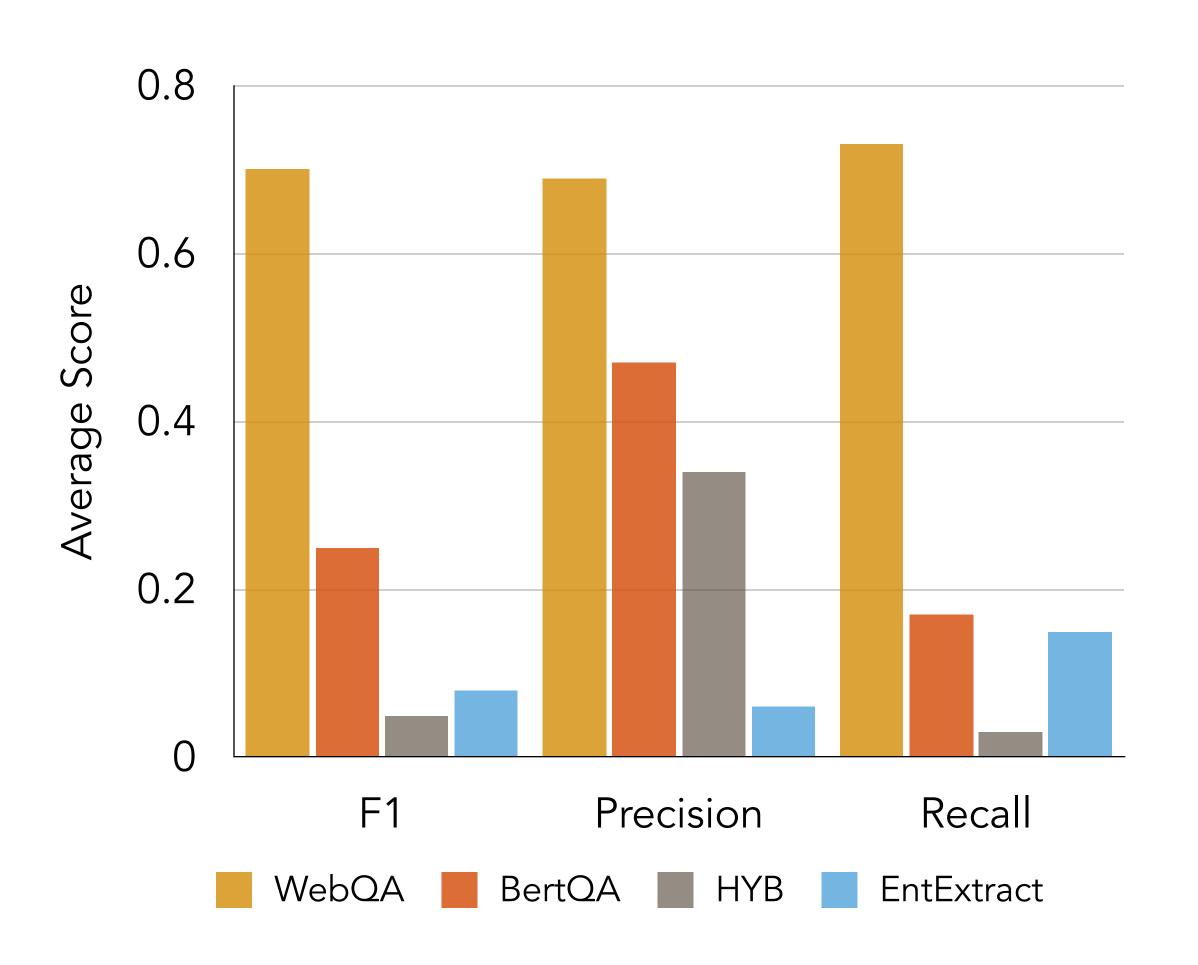


Evaluation: dataset

- We collect webpages:
 - 4 domains (Faculty profile, Conference, Class, Clinic)
 - 40 webpages for each domain
 - ~6 tasks per domain
 - 5 for training and remaining for testing

Domain	Sample tasks	
Faculty	Extract current PhD studentsExtract papers published in 2012	
Conference	Extract program committee membersExtract paper submission deadlines	
Class	Extract the time of the lecturesExtract name of the TAs	
Clinic	Extract the list of doctors or providersExtract the provided services	

Evaluation: Benefit of Neurosymbolic Approach



• Baselines:

- BERTQA: Question answering system takes a webpage and a question as input and outputs the answers [Zero shot]
- HYB: **PBE system** that takes a set of webpages as example inputs and synthesize **a Xpath program**
- EntExtract: **Entity extraction tool** for webpages using a natural language query as input [Zero shot]

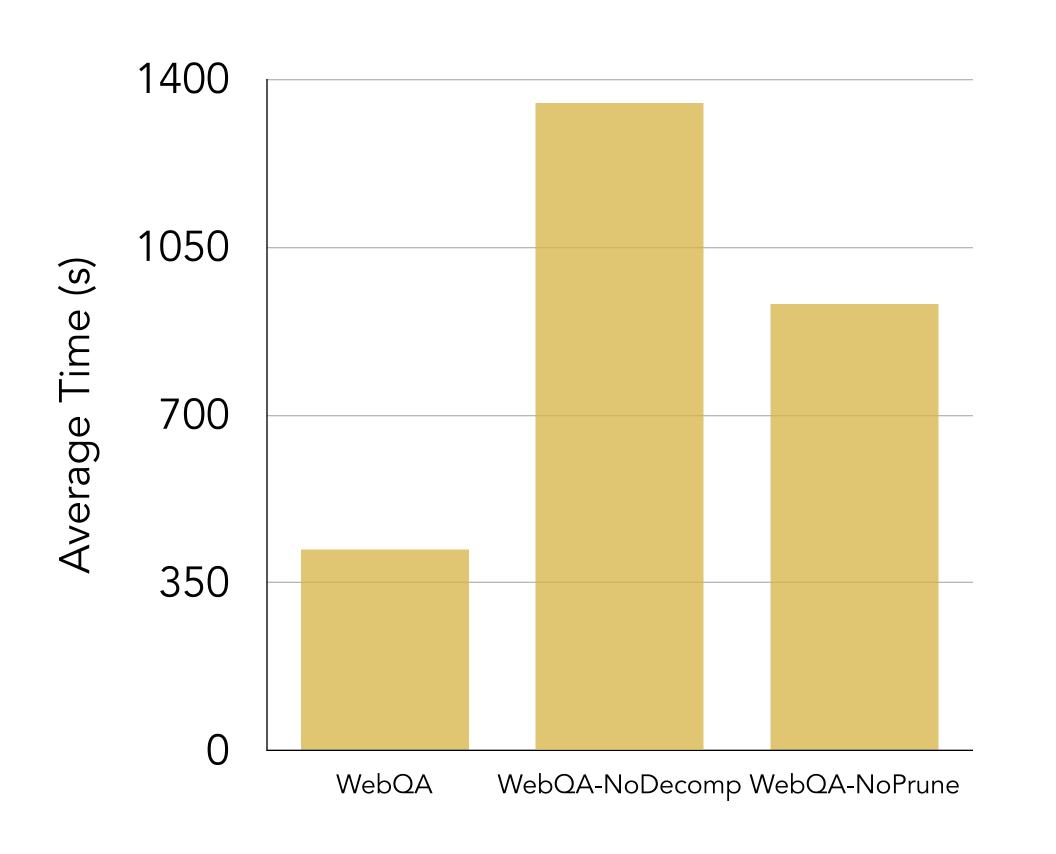
WebQA outperforms prior works for our task

Evaluation: ablation study

Three key ideas:

- Decomposition between guard and extractor synthesis
- Pruning using the monotonicity property of the DSL
- Reduce variance of the output program using transductive learning

Evaluation: Effectiveness of Decomposition and Pruning



• Baselines:

- WebQA-NoDecomp: Synthesizes guards and extractors jointly
- WebQA-NoPrune: Does not compute upper bound of partial programs for pruning

WebQA achieves 3.6x speedup compare to WebQA-NoPrune and 2.4x speedup compare to WebQA-NoDecomp.

Evaluation: Effectiveness of the Transductive Learning

Technique	% Improvement in F1	Reduction in Variance
Random	6.0%	1550X
Shortest	6.3%	1570X

- Two Baselines :
 - Random: randomly chooses one of the optimal programs
 - Shortest: chooses randomly one program with the smallest AST size among the optimal programs
- We measured the % of improvement and reduction in variance that WebQA achieved over the baseline

WebQA achieves more stable performance in obtaining high quality synthesized programs

Conclusion

- Extract information from websites with globally different schemas using neurosymbolic program synthesis
- Performs better than techniques for wrapper induction and question answering for such structurally heterogeneous websites

Try WebQA at https://github.com/utopia-group/WebQA