Analyzing Software usingDeep Learning

Reasoning about Types and Code Changes with Hierarchical Networks (Part 2)

Prof. Dr. Michael Pradel

Software Lab, University of Stuttgart Summer 2020

Overview

Hierarchical neural networks

Type prediction

Based on "TypeWriter: Neural Type Prediction with Search-based Validation" by Pradel et al., 2020

Representing code changes

Based on "CC2Vec: Distributed Representations of Code Changes" by Hoang et al., 2020

Types in Dynamic Progr. Langs.

- Dynamically typed languages:Extremely popular
- Lack of type annotations:
 - Type errors
 - Hard-to-understand APIs
 - Poor IDE support

```
def find_match(color):
    candidates = get_colors()
    for candidate in candidates:
        if color == candidate:
           return color
    return None

def get_colors():
    return ["red", "blue", "green"]
```

Gradual Typing

Annotate some code locations with types

E.g., parameter types and return types of some functions only

Gradual type checker

- Warn about inconsistencies
- Ignores missing information

Gradual Typing

Annotate some code locations with types

E.g., parameter types and return types of some functions only

Gradual type checker

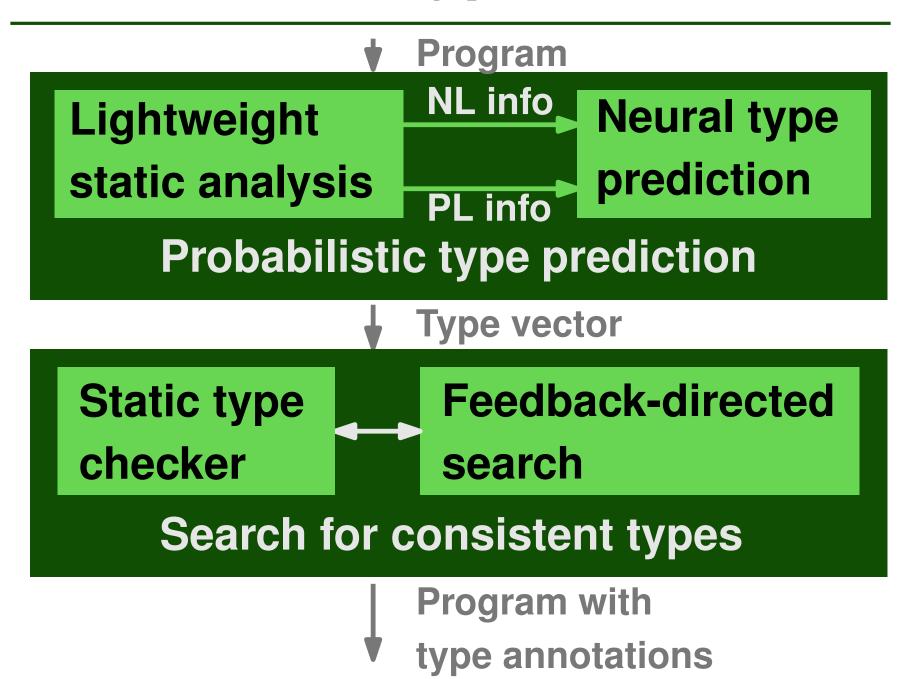
- Warn about inconsistencies
- Ignores missing information

But: Annotating types is painful

How to Add Type Annotations?

- Option 1: Static type inference
 - Guarantees type correctness, but very limited
- Option 2: Dynamic type inference
 - Depends on inputs and misses types
- Option 3: Probabilistic type prediction
 - Models learned from existing type annotations

Overview of TypeWriter



Extracting NL and PL Info

NL information

- Names of functions and arguments
- Function-level comments

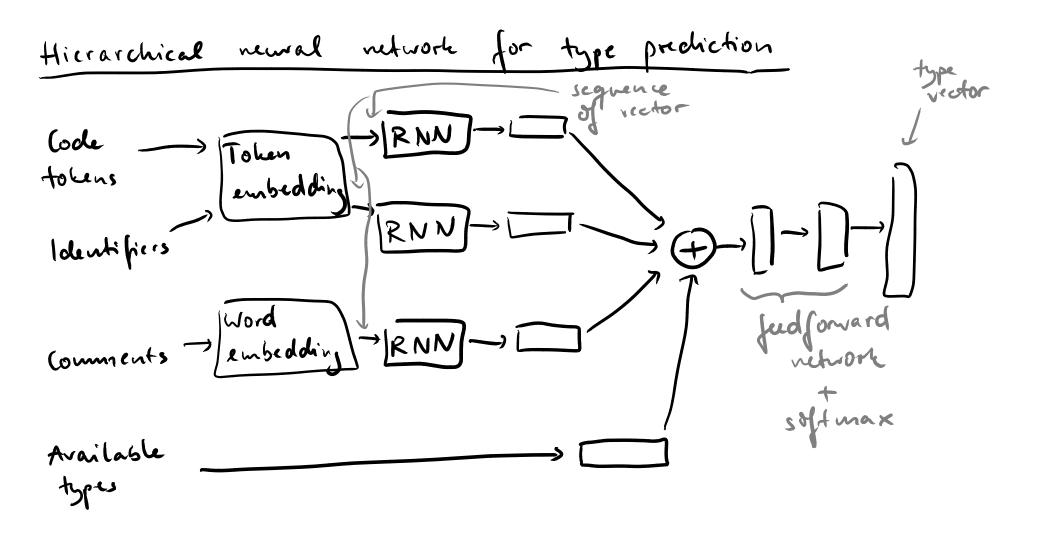
PL information

- Occurrences of the to-be-typed code element
- Types made available via imports

```
def find_match(color):
  ** ** **
  Args:
    color (str): color to match on and return
  ** ** **
  candidates = get_colors()
  for candidate in candidates:
    if color == candidate:
      return color
  return None
def get_colors():
  return ["red", "blue", "green"]
```

```
Names: find_match, color
def find_match(color):
  ** ** **
  Args:
    color (str): color to match on and return
  77 77 77
  candidates = get_colors()
                                        Function-level
  for candidate in candidates:
                                        comment
    if color == candidate:
      return color
  return None
                                  Names: get_colors
def get_colors():
  return ["red", "blue", "green"]
```

```
def find_match(color):
  ** ** **
 Args:
    color (str): color to match on and return
                                    Occurrences of
  77 77 77
  candidates = get_colors()
                                    parameters
  for candidates:
    if color == candidate:
      return color
  return None
                                        Return
def get_colors():
                                       statements
  return ["red", "blue", "green"]
```



Output: Type Vector

- Type prediction as a classification problem
- Output of the model: Type vector
 - One element for each of top-1000 types
 - During training:
 - All zero, except for the correct type
 - During prediction:
 - Interpreted as probability distribution over types

Training the Model

- Training data: Existing type annotations
 - Multi-million line code base
 - $_{\Box}$ Some types (pprox 20-50%) already annotated
- Learns to predict missing types from existing annotations

Example of Predictions

```
def find_match(color):
  ** ** **
  Args:
    color (str): color to match on and return
  ** ** **
  candidates = get_colors()
  for candidate in candidates:
    if color == candidate:
      return color
  return None
def get_colors():
  return ["red", "blue", "green"]
```

Example of Predictions

```
Predictions:
def find_match(color):
                                      int, str, bool
  ** ** **
  Args:
    color (str): color to match on and return
  ** ** **
  candidates = get_colors()
  for candidate in candidates:
                                      Predictions: str,
    if color == candidate:
                                      Optional[str],
      return color
  return None
                                      None
def get_colors():
  return ["red", "blue", "green"]
                                      Predictions:
                                      List[str],
                                      List[Any], str
```

Challenges

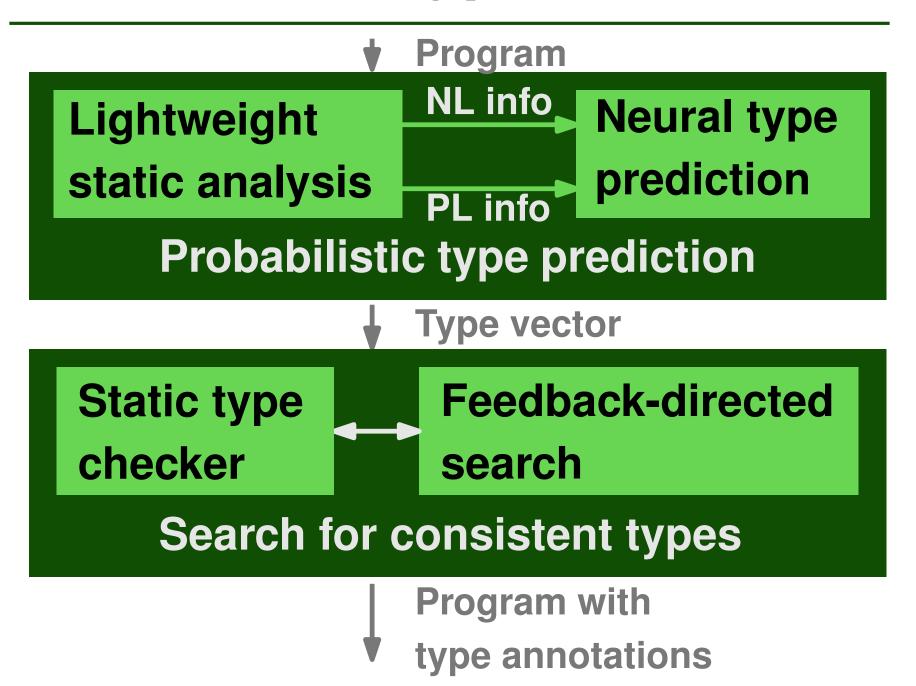
Imprecision

- Some predictions are wrong
- Developers must decide which suggestions to follow

Combinatorial explosion

- For each missing type: One or more suggestions
- Exploring all combinations:
 - Practically impossible

Overview of TypeWriter



Searching for Consistent Types

Top-k predictions for each missing type

- Filter predictions using gradual type checker
- E.g., pyre and mypy for Python, flow for JavaScript

Combinatorial search problem

 $\ \square$ For type slots S and k predictions per slot: $(k+1)^{|S|}$ possible type assignments

Searching for Consistent Types

Top-k predictions for each missing type

- Filter predictions using gradual type checker
- □ E.g., pyre and mypy for Python, flow for **JavaScript**

Combinatorial search problem

- $\ \square$ For type slots S and k predictions per slot:
 - $(k+1)^{|S|}$ possible type assignments

Feedback Function

 Goal: Minimize missing types without introducing type errors

Feedback score (lower is better):

$$v \cdot n_{missing} + w \cdot n_{errors}$$

Feedback Function

 Goal: Minimize missing types without introducing type errors

Feedback score (lower is better):

$$v \cdot n_{missing} + w \cdot n_{errors}$$



Default: v = 1, w = 2,

i.e., higher weight for errors

Search Strategies

Optimistic vs. pessimistic

Add top-most predicted type everywhere and then remove types

Add one type at a time

Greedy vs. non-greedy

If score decreases, keep the type

Backtrack to avoid local minima

```
def find_match(color):
  ** ** **
  Args:
    color (str): color to match on and return
  ** ** **
  candidates = get_colors()
  for candidate in candidates:
    if color == candidate:
      return color
  return None
def get_colors():
  return ["red", "blue", "green"]
```

```
Predictions:
def find_match(color):
                                      int, str, bool
  ** ** **
  Args:
    color (str): color to match on and return
  ** ** **
  candidates = get_colors()
  for candidate in candidates:
                                      Predictions: str.
    if color == candidate:
                                      Optional[str],
      return color
  return None
                                      None
def get_colors():
  return ["red", "blue", "green"]
                                      Predictions:
                                      List[str],
                                      List[Any], str
```

Results

Neural model

- Precision: 58-73% in top-1, 50-92% in top-5
- □ Recall: 50-58% in top-1, 69-72% in top-5

Model and search together

 Best strategy adds 72% of type-correct types and completely annotates 44% of files

In use at Facebook

 Thousands of suggested types accepted by developers with minimal changes