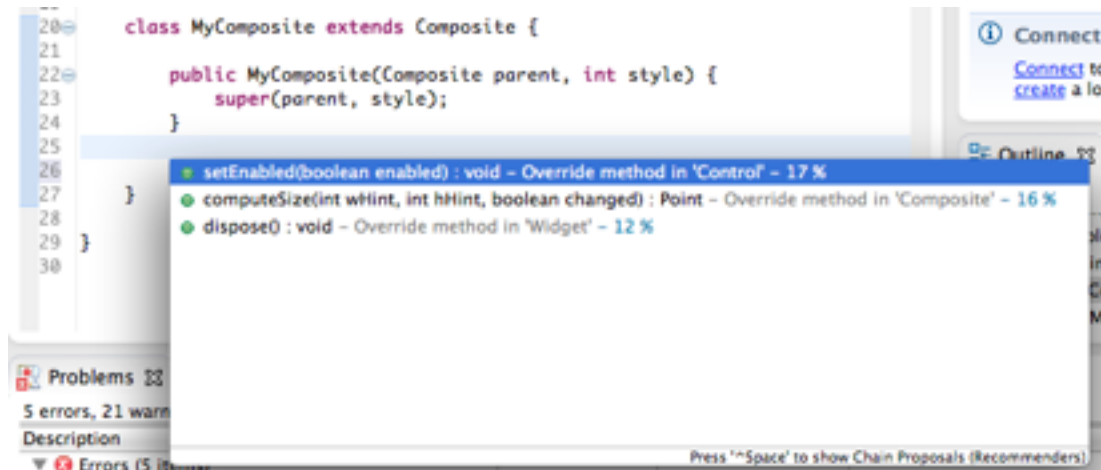


Statistical Models of Typed Syntax Trees

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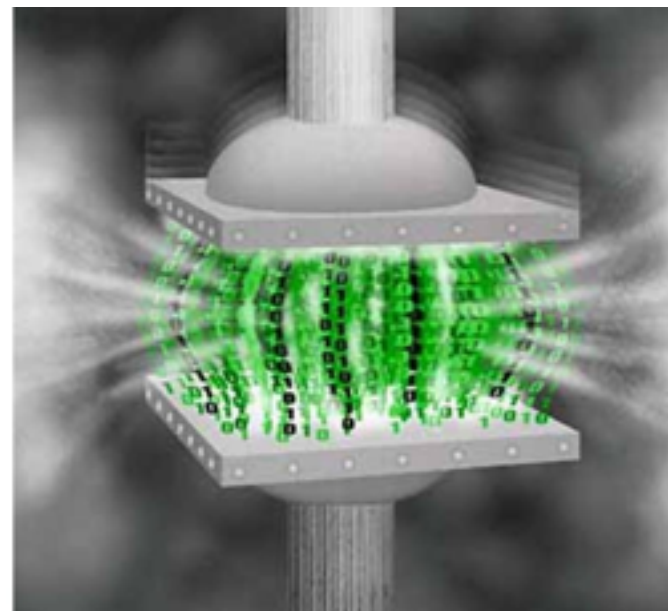
Many tools could benefit from an understanding of the statistics of *natural programs*.



code completion
engines

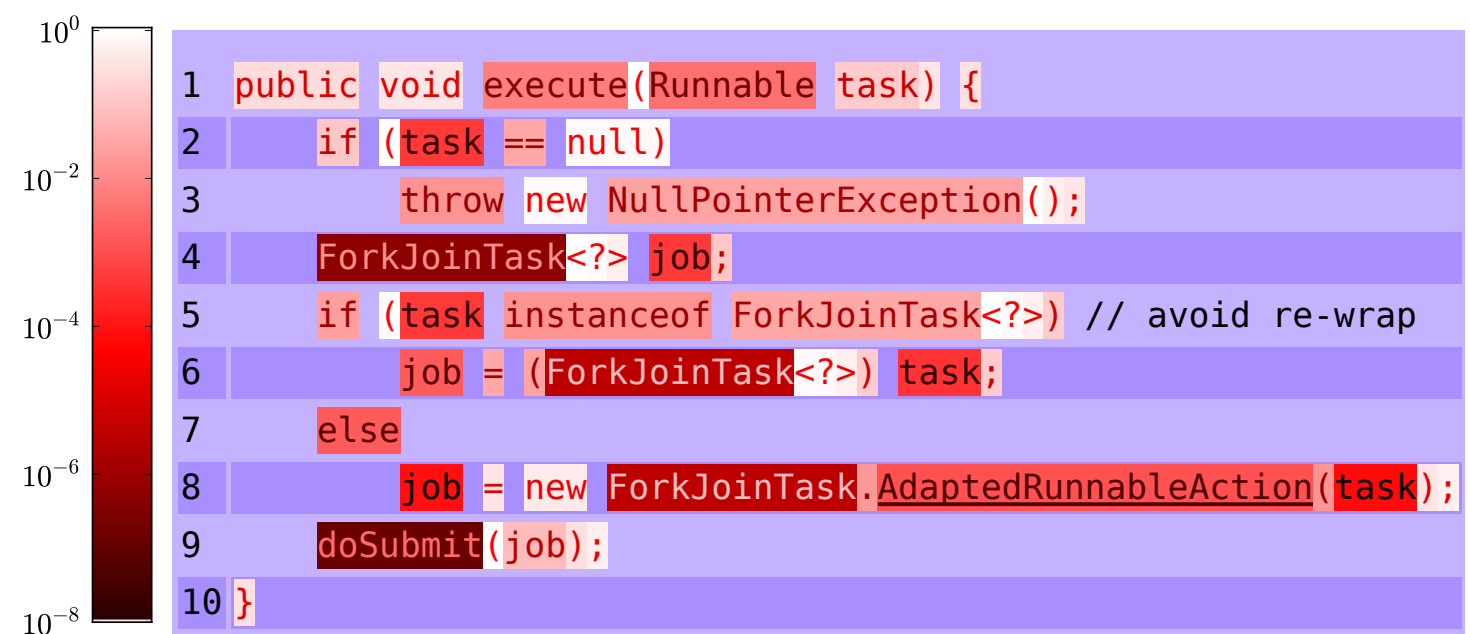


predictive programming
interfaces



code compression
engines

Many tools could benefit from an understanding of the statistics of *natural programs*.



code smell detectors

Running Example

```
enum Planet {  
    MERCURY,  
    VENUS,  
    ...,  
    NEPTUNE  
}  
  
class Astronomer {  
    void observe(Planet p) {...}  
}
```

```
class Sagan extends Astronomer {  
    void beforeBed() {  
        observe(favorite)  
    }  
  
    Planet favorite = Planet.NEPTUNE;  
}
```

To develop a probability distribution for code, we need to first choose a **representation of code**

Running Example

```
enum Planet {  
    MERCURY,  
    VENUS,  
    ...,  
    NEPTUNE  
}  
  
class Astronomer {  
    void observe(Planet p) {...}  
}
```

```
class Sagan extends Astronomer {  
    void beforeBed() {  
        observe(favorite)  
    }  
  
    Planet favorite = Planet.NEPTUNE;  
}
```

P(observe (favorite) | () { })

- *n*-grams (Hindle et al., ICSE 2012; Allamanis & Sutton, MSR 2013)
- topic modeling + part of speech analysis (Nguyen et al., FSE 2013)

Previous Work: Programs are Token Sequences

Running Example

```
enum Planet {  
    MERCURY,  
    VENUS,  
    ...,  
    NEPTUNE  
}  
  
class Astronomer {  
    void observe(Planet p) {...}  
}
```

```
class Sagan extends Astronomer {  
    void beforeBed() {  
        observe(favorite)  
    }  
  
    Planet favorite = Planet.NEPTUNE;  
}
```

P(call var this .observe(field var this .favorite)

| $\rho = \text{stmt},$
 role

Our Approach: Programs are **Syntax Trees**

Running Example

```
enum Planet {  
    MERCURY,  
    VENUS,  
    ...,  
    NEPTUNE  
}  
  
class Astronomer {  
    void observe(Planet p) {...}  
}
```

```
class Sagan extends Astronomer {  
    void beforeBed() {  
        observe(favorite)  
    }  
  
    Planet favorite = Planet.NEPTUNE;  
}
```

P(call var this .observe(field var this .favorite)

| $\rho = \text{stmt}, \tau = \text{void}, \Gamma$)

role

type

typing
context

Our Approach: Programs are *Typed Syntax Trees*

Running Example

```
enum Planet {
    MERCURY,
    VENUS,
    ...,
    NEPTUNE
}

class Astronomer {
    void observe(Planet p) {...}
}
```

```
class Sagan extends Astronomer {
    void beforeBed() {
        observe(favorite)
    }

    Planet favorite = Planet.NEPTUNE;
}
```

$= \mathbf{P}(\text{call } \boxed{\text{var this}}.\text{observe}(\text{field } \boxed{\text{var this}}.\text{favorite}))$

$| \phi = \text{call}, \rho = \text{stmt}, \tau = \text{void}, \Gamma) \mathbf{P}(\phi | \rho, \tau)$

syntactic
form

role

type

typing
context

Bayes' rule!

Running Example

```
enum Planet {
    MERCURY,
    VENUS,
    ...,
    NEPTUNE
}

class Astronomer {
    void observe(Planet p) {...}
}
```

```
class Sagan extends Astronomer {
    void beforeBed() {
        observe(favorite)
    }

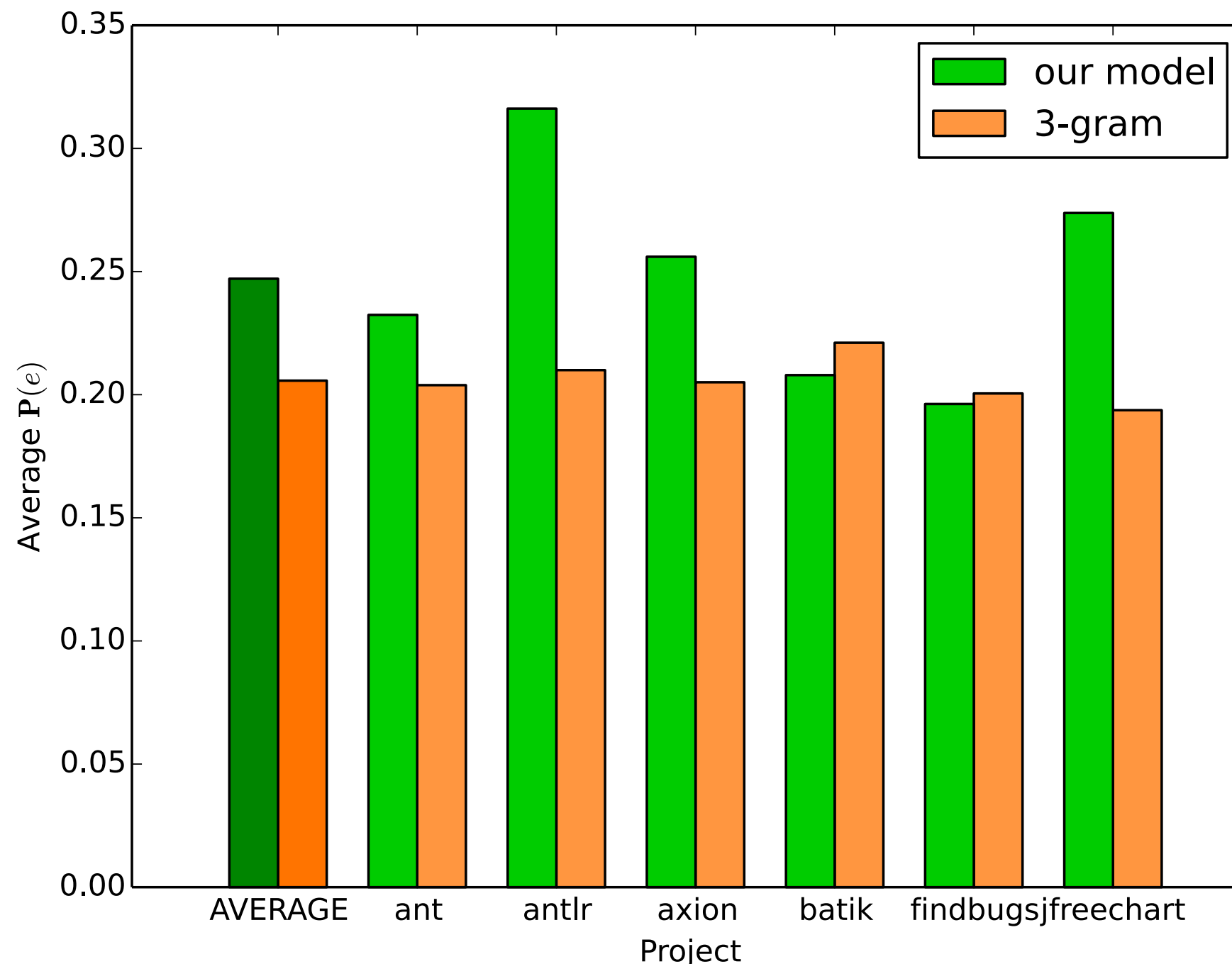
    Planet favorite = Planet.NEPTUNE;
}
```

$$\begin{aligned}
 &= \mathbf{P}(\text{call } \boxed{\text{var this}}.\text{observe}(\text{field } \boxed{\text{var this}}.\text{favorite}) \\
 &\quad | \phi = \text{call}, \rho = \text{stmt}, \tau = \text{void}, \Gamma) \mathbf{P}(\phi | \rho, \tau) \\
 &= \mathbf{P}(\text{Sagan.observe} | \rho = \text{stmt}, \tau = \text{void}, \Gamma) \\
 &\quad \mathbf{P}(\boxed{\text{var this}} | \rho = \text{targ}, \tau = \text{Sagan}, \Gamma) \\
 &\quad \mathbf{P}(\text{field } \boxed{\text{var this}}.\text{favorite} | \rho = \text{arg}, \tau = \text{Planet}, \Gamma) \\
 &\quad \mathbf{P}(\phi | \rho, \tau)
 \end{aligned}$$

We are starting to **implement** this model for Java using the **Eclipse JDT** for parsing and keeping track of Γ

<http://www.github.com/cyrus-/syzygy>

To test our implementation, we perform **10-fold cross-validation** of our model on a corpus of several large open source projects and **compare it to the 3-gram model** used in Hindle et. al, 2012.



(collaboration with **Salil Joshi** and **Flavio Cruz**)

We are taking a first-principles approach to
source code prediction
that combines the foundational techniques of both
statistics and **semantics**.

P(call `var this` .observe(field `var this` .favorite)
| $\rho = \text{stmt}, \tau = \text{void}, \Gamma$)
 role type typing
 context