/* iComment: Bugs or Bad Comments? */

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iComment

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• Program comments express assumptions.

```
linux/drivers/scsi/in2000.c:
   /* Caller must hold instance lock!*/
   static int reset_hardware(...) {...}
```

• Millions lines of comments exist in software.

| Software | Linux | Mozilla |
|--|-------|---------|
| Lines of code (excluding copyright notices and blank lines) | 5.0M | 3.3M |
| Lines of Comment (excluding copyright notices and blank lines) | I.0M | 0.51M |

- Comments are not fully utilized yet.
 - Ignored by compilers and bug detection tools.





Code vs. Comments

| Code | Comment | Implication |
|----------------------|----------------------|--|
| Precise | Imprecise | Comments are harder to analyze. |
| Can be tested | Can NOT be tested | Comments may become less reliable as software evolves. |
| Harder to understand | Easier to understand | Likely to read comments. Wrong comments mislead programmers. |

- Many assumptions are difficult to infer from source code alone.
 - Inferring from source code alone may fail
 - for cases that no (or only a few) places of the code follow the assumption.
- Use comment-code redundancy to detect comment-code mismatches.





Possibility (1): Bugs

- Mismatches indicate:
 - Possibility (1): Bugs
 - Due to time-constraints or other reasons.
 - Old code is not updated according to a new assumption.

A bug automatically detected by iComment:





ILLINOIS

Possibility (2): Bad Comments

- Possibility (2): Bad comments can cause new bugs
 - Comments are not updated accordingly.

A bad comment automatically detected by iComment:



 Our paper contains bad comment examples that already caused new bugs.





• Goal: Detect comment-code inconsistencies.

- Challenges of understanding comments written in natural language
 - Various ways to paraphrase natural language
 - /* We need to acquire the write IRQ lock before calling ep_unlink(). */
 - /* Lock must be acquired on entry to this function. */
 - /* Caller must hold instance lock! */
- Use Natural Language Processing (NLP) techniques?

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NLP alone is not enough.

• NLP only analyzes sentence structures.



- NLP is far from "understanding" natural language text.
- Many comments are not even grammatically correct.
- Almost impossible to automatically analyze any arbitrary comments.





Idea & Contributions

- Took the first step to automatically analyze comments written in natural language to check for mismatches
 - Combine Natural Language Processing (NLP), Machine Learning, Statistics, and Program Analysis
- Automatically extracted 1832 rules and detected 60 new bugs and bad comments (19 confirmed by developers)
 - 2 topics, lock-related and call-related.
 - Latest versions of 4 large software projects, Linux, Mozilla, Apache and Wine.





Outline

- Motivation, Challenges & Contributions
- Our Approach
 - Analyze comments written in natural language
 - Detect comment-code inconsistencies
- Methodology & Results
- Related work
- Conclusions





What to Analyze?

• What information is useful to extract?

• What information can be checked against code?





What is useful to extract?

- Two types of comments (examples from Linux):
 - Explain code segment: /* Set the clock rate */
 - Express assumptions/rules: /* Caller must hold instance lock! */
- We focus on rule-containing comments.
 - Likely to be inconsistent with code.
 - Likely to mislead programmers to introduce bugs.





- Not everything in comments can be checked.
- Checking can only be done topic by topic.
 - Race detectors race bugs
 - Purify, Valgrind, etc memory bugs
- So our comment analysis is topic by topic.
 - A general framework allowing users to choose the topic, such as lock and call-from.





Rule Template Examples

| ID | Rule Template Examples | |
|----|---|----------------|
| I | <lock l=""> must be held before entering <function f="">.</function></lock> | ר |
| I | <lock l=""> must NOT be held before entering <function f="">.</function></lock> | lock related |
| 2 | <lock l=""> must be held in <function f="">.</function></lock> | lock related |
| 2 | <lock l=""> must NOT be held in <function f="">.</function></lock> | J |
| 3 | <function a=""> must be called from <function b=""></function></function> |) call related |
| 3 | <function a=""> must NOT be called from <function b=""></function></function> | J |
| | | |

- L, F, A and B are rule parameters.
- See our paper for many other templates supported.
- Many other templates can be added.





Extracting Target Comments

- Statistics & NLP
 - Topic keyword filtering automatic
 - Correlated word filtering automatic

 $cosine(A,B) = \frac{P(A,B)}{\sqrt{P(A)P(B)}}$

See our paper for details.

| Linux | hold | acquire | call | unlock | protect |
|---------|------|---------|--------|---------|---------|
| Mozilla | hold | acquire | unlock | protect | call |

Take lock as the topic:

```
#A: /* return -EBUSY if a lock is held. */
#B: /* Lock must be held on entry to this function. */
#C: /* Caller must acquire instance lock! */
#D: /* Mutex locked flags */
...
```





Classifying Comments

- Machine Learning & NLP
- Automatically classify comments to different templates (give each comment a unique label)
 - Core technique: Use learning classifier <u>automatically</u> <u>built</u> from a small set of manually labeled comments









General-purpose Training

- The training is optional for the users
 - Done by us before releasing iComment (only once per topic).
 - Feasible because:
 - Programmers share wording and phrasing (confirmed by our correlated word results)
 - Cross-software training results show decision trees trained on one software can classify comments from other software with high accuracy (~89%)
- Took only about 2 hours to manually classify comments of 2 topics for Linux, Mozilla, Apache and Wine





/* Caller must hold

instance lock! */

Generating Rules

- NLP & Program Analysis
- What are the parameters?
 - The function name is right after the comment.
 - The lock name is the object of the verb.
- Is the rule positive or negative?
 - Positive if the verb is not modified by a negation word.





Rule Checker

- Use static analysis for checking
 - Flow-sensitive, and context sensitive
 - Simple point-to analysis
- Mismatch report ranking
 - Support
 - Violation





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| Software | LOC | LOM | Language | Description |
|----------|------|-------|----------|----------------------------------|
| Linux | 5.0M | I.0M | С | OS |
| Mozilla | 3.3M | .51M | C&C++ | Browser Suite |
| Wine | I.5M | .22M | С | Program to Run WinApp on Unix |
| Apache | .27M | .057M | С | Web Server |

- Latest versions of 4 large software projects
- 2 topics: lock-related and call-related
- 18% of comments are used for training on average.
 - Our training sensitivity analysis provides guidance on how much training data to use (find detailed results in our paper).





Overall Results



| Software | Mismatches | Bugs | BadCom | FP | Rules |
|----------|------------|---------|--------|----|-------|
| Linux | 51 (14) | 30 (11) | 21 (3) | 32 | 1209 |
| Mozilla | 6 (5) | 2 (1) | 4 (4) | 3 | 410 |
| Wine | 2 | I | I | 3 | 149 |
| Apache | _(| 0 | | 0 | 64 |
| Total | 60(19) | 33 (12) | 27(7) | 38 | 1832 |

- Automatically detected 60 new bugs and bad comments
 - 19 new bugs and bad comments already confirmed by the corresponding developers.
- Major causes of false positives
 - Mostly caused by inaccuracy from checking
 - Incorrectly generated rules







- Accuracy = the percentage of correctly labeled comments
- Software-specific training accuracy (lock-related)

| Linux | Mozilla | Wine | Apache |
|-------|---------|-------|--------|
| 90.8% | 91.3% | 96.4% | 100% |

Other measures, such as Kappa and Macro-F score, show similar results. Accuracies for call-related comments are similar.

• Cross-software training accuracy (lock-related)

| Training SW | Mozilla | Wine | Apache |
|---------------|---------|-------|--------|
| Linux | 81.5% | 78.6% | 83.3% |
| Linux+Mozilla | / | 89.3% | 88.9% |

 Training can be done by us before releasing iComment to analyze users' software.





- Extracting rules from source code and execution behaviors [SOSPOI & OSDI06 Engler et. al., Daikon, ...]:
 - Our approach complements these techniques.
- Annotation Language [Microsoft SAL, Java annotations, Splint, SafeDrive, Sparse, ...]:
 - Not as expressive: usability
 - Not widely adopted vs. millions lines of comments already exist.
- Automatic document generation from comments
 [C# XML comments, JavaDoc, Doxygen, RDoc, ...]:
 - Do NOT analyze the natural language part
 - Share similar challenges of analyzing unstructured comments.





Conclusions

- Comment-code inconsistencies hurt software quality and reliability.
- First work to automatically analyze comments written in natural language for mismatch detection
 - iComment automatically extracted 1832 rules on 2 topics and detected 60 new bugs and bad comments (19 confirmed by developers)
- More work in this direction!
 - Analyze other system documents in natural language

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Thank you! Questions?

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