### Using Machine Learning to Improve Security Analysis of Source Code

Mark Sherman Sept 23, 2020 Cylab Partners Conference

Software Engineering Institute Carnegie Mellon University Pittsburgh, PA 15213

Carnegie Mellon University Software Engineering Institute [DISTRIBUTION STATEMENT A] Approved for public release and unlimited distribution.

Copyright 2020 Carnegie Mellon University.

This material is based upon work funded and supported by the Department of Defense under Contract No. FA8702-15-D-0002 with Carnegie Mellon University for the operation of the Software Engineering Institute, a federally funded research and development center.

The view, opinions, and/or findings contained in this material are those of the author(s) and should not be construed as an official Government position, policy, or decision, unless designated by other documentation.

NO WARRANTY. THIS CARNEGIE MELLON UNIVERSITY AND SOFTWARE ENGINEERING INSTITUTE MATERIAL IS FURNISHED ON AN "AS-IS" BASIS. CARNEGIE MELLON UNIVERSITY MAKES NO WARRANTIES OF ANY KIND, EITHER EXPRESSED OR IMPLIED, AS TO ANY MATTER INCLUDING, BUT NOT LIMITED TO, WARRANTY OF FITNESS FOR PURPOSE OR MERCHANTABILITY, EXCLUSIVITY, OR RESULTS OBTAINED FROM USE OF THE MATERIAL. CARNEGIE MELLON UNIVERSITY DOES NOT MAKE ANY WARRANTY OF ANY KIND WITH RESPECT TO FREEDOM FROM PATENT, TRADEMARK, OR COPYRIGHT INFRINGEMENT.

[DISTRIBUTION STATEMENT A] This material has been approved for public release and unlimited distribution. Please see Copyright notice for non-US Government use and distribution.

This material may be reproduced in its entirety, without modification, and freely distributed in written or electronic form without requesting formal permission. Permission is required for any other use. Requests for permission should be directed to the Software Engineering Institute at permission@sei.cmu.edu.

Carnegie Mellon® is registered in the U.S. Patent and Trademark Office by Carnegie Mellon University.

DM20-0770

## Research Projects at SEI/ CERT

Source Code as Natural Language Combining Source Code Analyzers Automated Source Code Repair

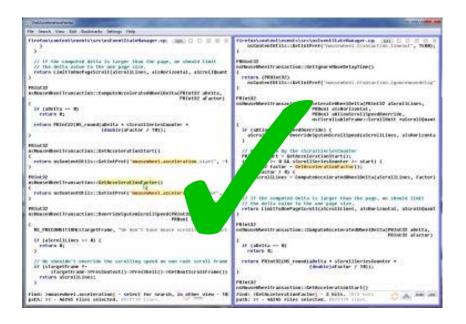
### **Research Teams**

<u>PIs</u> Lori Flynn Will Klieber Carson Sestili

Team members Jennifer Burns Matt Churilla Zachary Kurtz Jiyeon Lee Derek Leung Guillermo Marce-Santurio Ruben Martins (SCS) Mike McCall

Team members (cont) Ebonie McNeil **Richard Qin** Will Snavely Ryan Steele Robert Stoddard David Svoboda Nathan VanHoudnos Joseph Yankel David Zubrow

## Finding Code Defects – ML that Considers Source Code as Natural Language



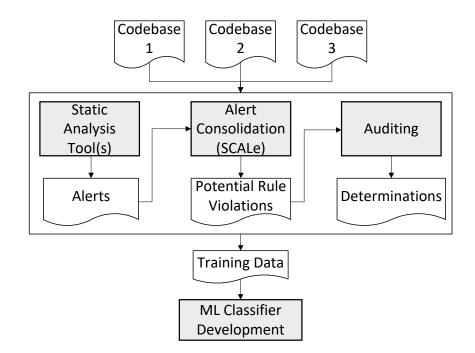
Analyze Source Code for Insecure Coding

- Supplements Compiler-style Checking
- Treats Programs Like Natural Language

Sources: Carson D. Sestili, William S. Snavely, Nathan M. VanHoudnos, Towards security defect prediction with AI, Sep 12, 2018, <u>read the article online here</u>.

Using Machine Learning to Improve Security Analysis of Source Code © 2020 Carnegie Mellon University

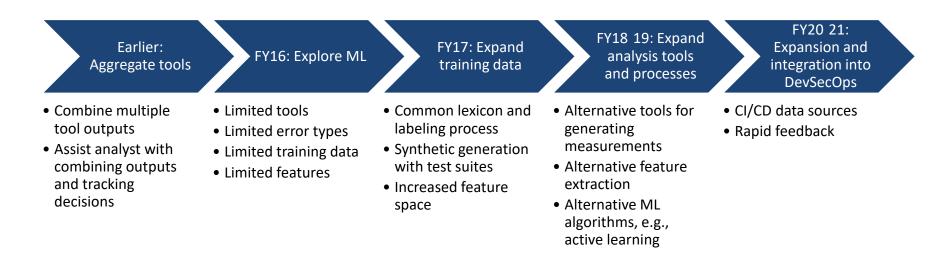
## Finding Code Defects Using Machine Learning



Read the L. Flynn publications at SEI Digital Library here.

Using Machine Learning to Improve Security Analysis of Source Code © 2020 Carnegie Mellon University

# Introduction and Integration of Machine Learning



## Automated Code Repair (ACR) Tool as a Black Box

Input: Buildable codebase

**Output:** Repaired source code that is still human-readable and maintainable.

We currently support C code. Support for C++ can likely be added without too much difficulty.

### **ACR Tool**



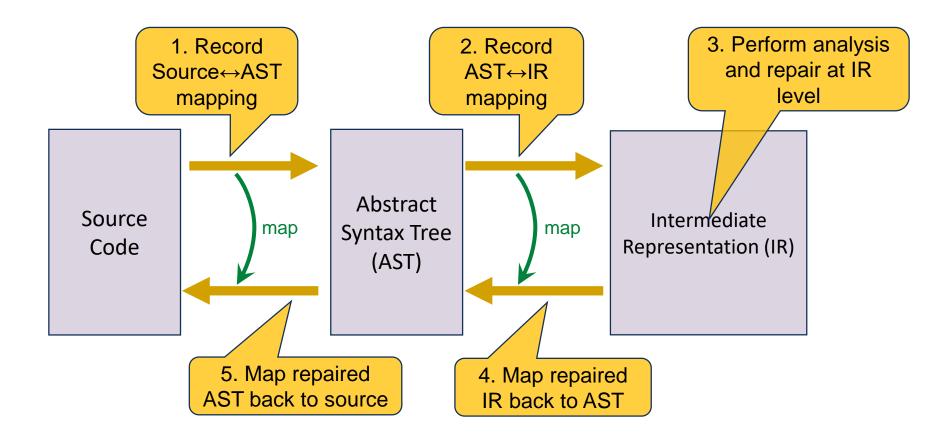
Read Will Klieber, Automated Code Repair to Ensure Memory Safety, Feb 24, 2020, here.

Using Machine Learning to Improve Security Analysis of Source Code © 2020 Carnegie Mellon University

## Why repair of source code instead of as a compiler pass?

Repair of source code	Repair as a compiler pass
Easily audited (if desired).	Must trust the tool.
Repairs can easily be tweaked to improve performance, if necessary.	Difficult to remediate performance issues caused by repair.
Changes to source code are frequent and easily handled.	Changes to the build process may be more difficult and error-prone.
Okay to do slow, heavy-weight static analysis; produces a persistent artifact.	Slowing down every test build is not okay.

### Source Code Repair Pipeline



Using Machine Learning to Improve Security Analysis of Source Code © 2020 Carnegie Mellon University

## Challenges

- In translating repairs from AST to source, the C preprocessor is main difficulty.
  - Repairs to macro uses
  - Repairs to #included code
  - Conditional-compilation directives (#ifdef, #endif, etc.) inside expressions
- Considerations of whitespace
- The C preprocessor can conditionally include or exclude pieces of code depending on the configuration chosen at compile time.
  - We repair configurations separately and then merge the results such that the final repaired code is correct under all desired configurations.

## Ways to Engage with Us



- Download software and tools
- Explore <u>research and capabilities</u>
- Participate in <u>education</u> offerings
- Attend an event
- Search the <u>digital library</u>
- Read the <u>SEI Year in Review</u>
- <u>Collaborate</u> with the SEI on a new project

#### Software Engineering Institute

Carnegie Mellon University 4500 Fifth Avenue Pittsburgh, PA 15213-3890 412-268-5800 - Phone 888-201-4479 - Toll-Free 412-268-5758 - Fax info@sei.cmu.edu - Email www.sei.cmu.edu - Web