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CEDAR: Continuous Testing of Deep Learning Libraries





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Continuous testing for DL libraries is in high demand

OPyTorch

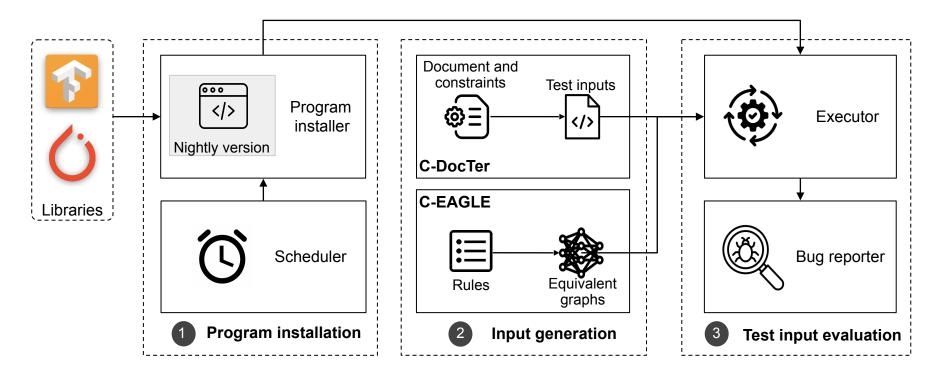
- Nightly build every few days
- 8 official versions in 2023
- 142,385 lines of code changes in a month

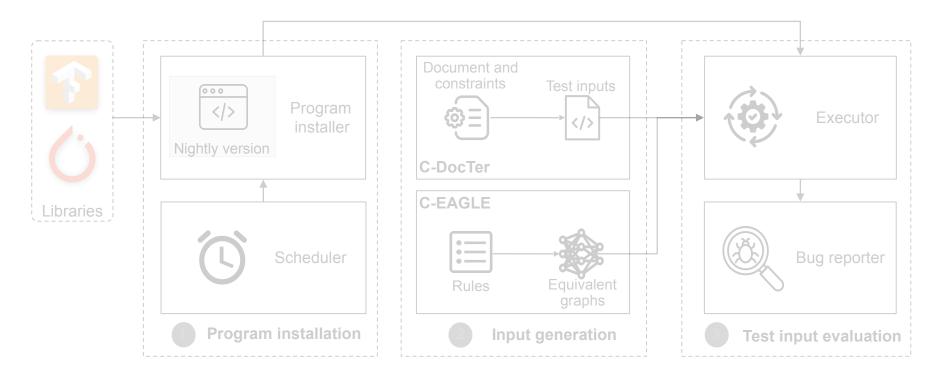


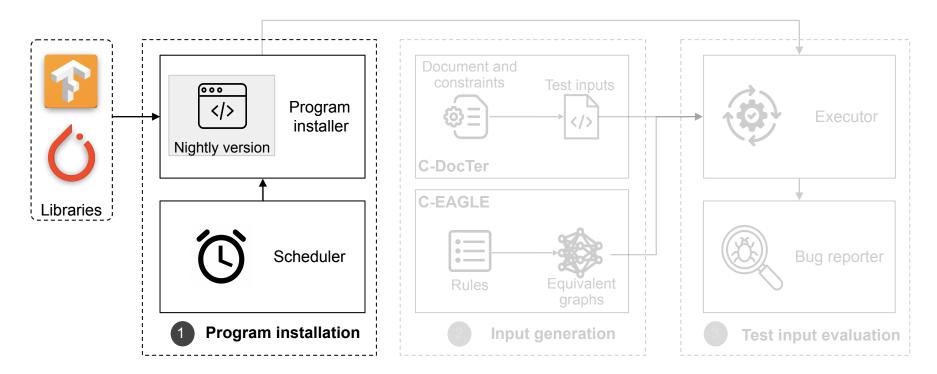
- **New bugs** are introduced along with the rapid changes.
- **Existing solutions** do not integrate cutting-edge DL testing tools including:
 - DocTer (ISSTA 22): documentation-guided fuzz testing framework for DL libraries.
 - EAGLE (ICSE 22): differential testing framework with equivalent graphs for DL APIs.
 - ...

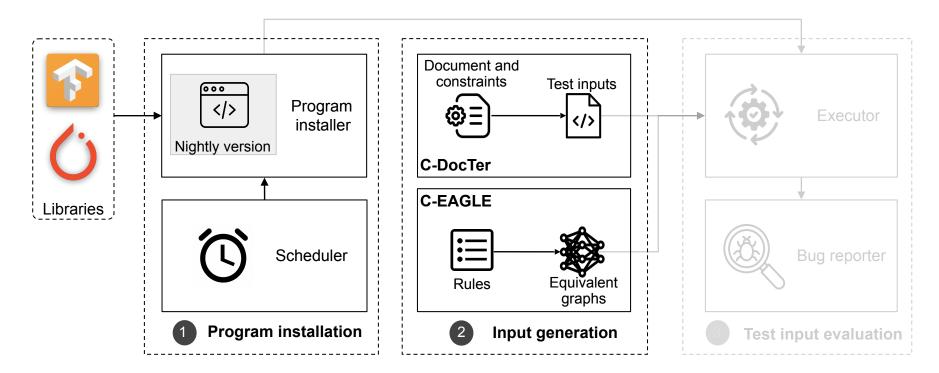
CEDAR: a continuous testing framework

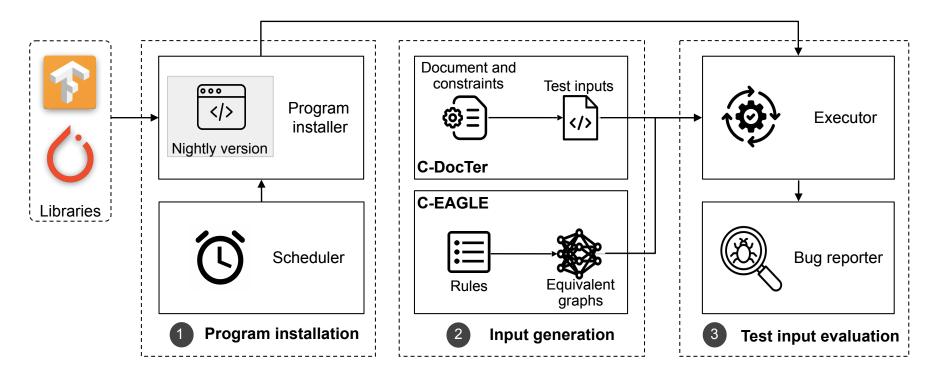
- Integrates two state-of-the-art DL testing approaches (DocTer and EAGLE).
- Effective: detecting 83 bugs affecting 140 PyTorch and TensorFlow APIs, including 23 previously unknown bugs.
- **Efficient**: with tool-specific optimization strategies to reduce the time and space overhead.
- Shortens the bug detection latency by **338.6 days** on average.







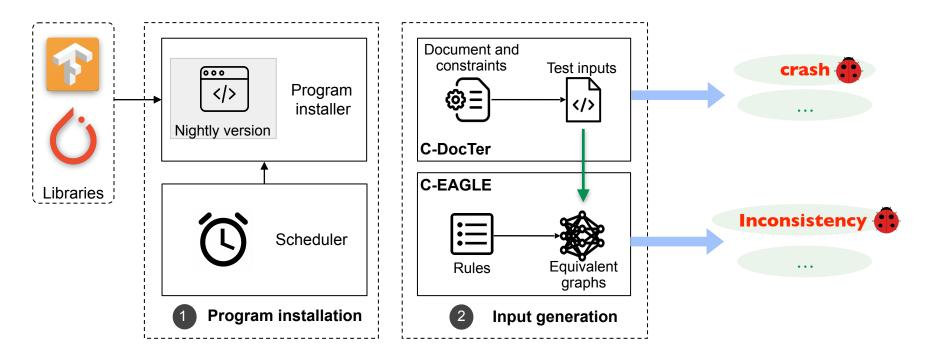




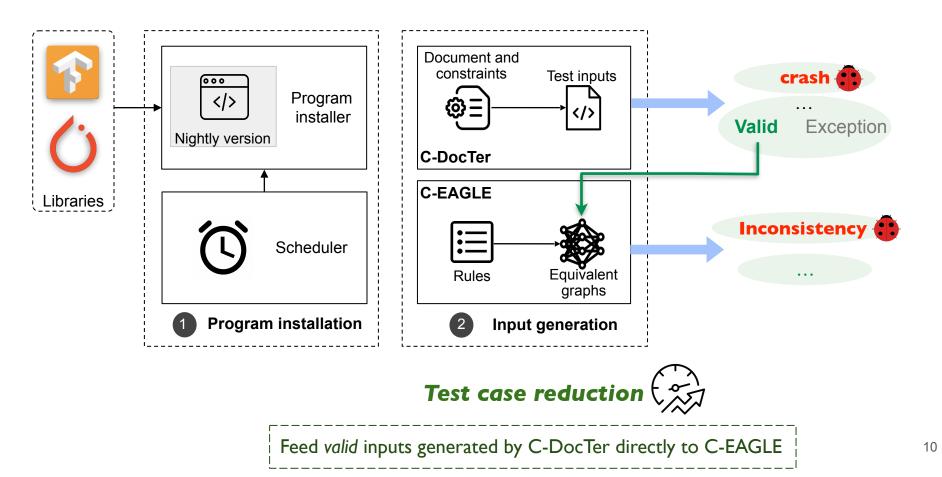


optimization strategies

CEDAR: tool-specific optimization to accelerate input generation



CEDAR: tool-specific optimization to accelerate input generation



Experimental Setup

2 libraries

versions of each library evaluated

5 9 and **925** APIs from PyTorch and TensorFlow

1,000 test inputs for each API

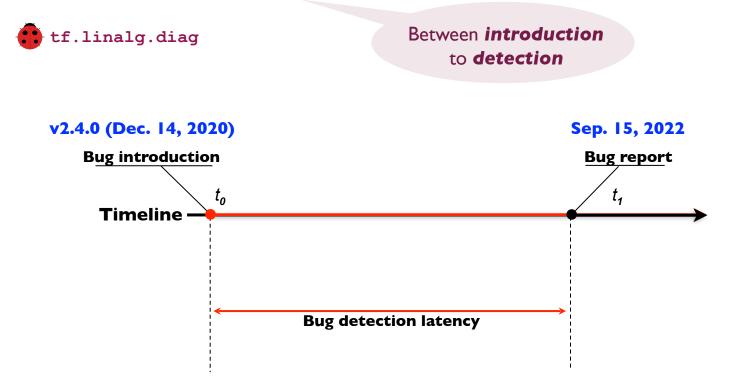
24 parallel processes

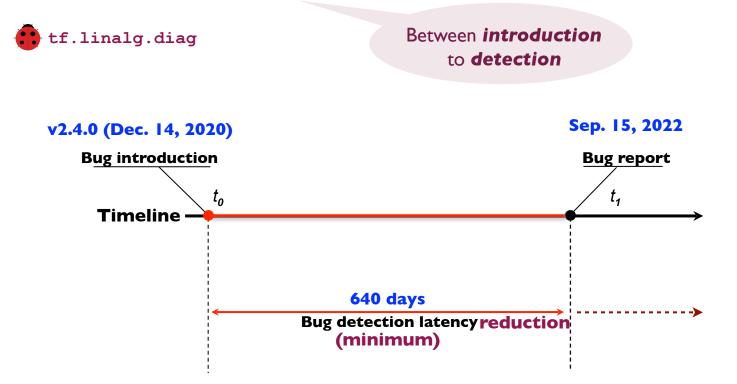
CEDAR's Bug detection results over continuous testing

83 bugs affecting 40 APIs,
including 2 high-priority bugs (24 APIs),
23 of the 83 bugs are new,
21 of 23 are confirmed or fixed.

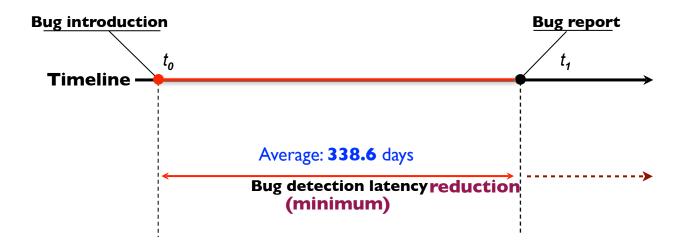
| | Verified | New | All | API |
|------------|----------|-----|-----|-----|
| PyTorch | 6 | 6 | 8 | 35 |
| TensorFlow | 15 | 17 | 75 | 105 |
| Total | 21 | 23 | 83 | 140 |





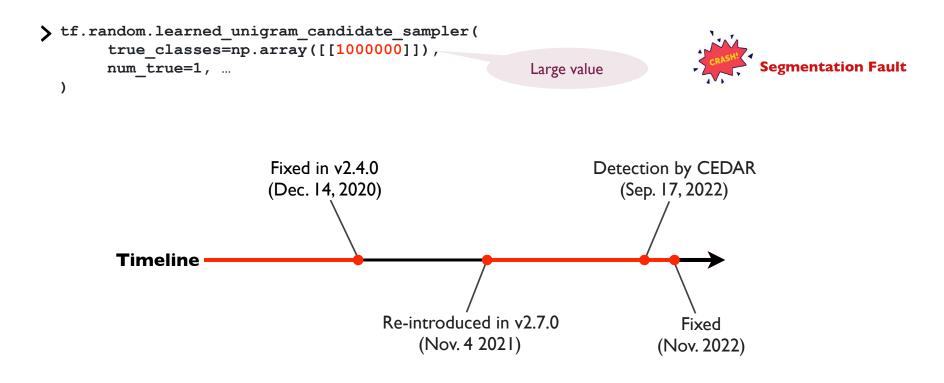


Between *introduction* to *detection*



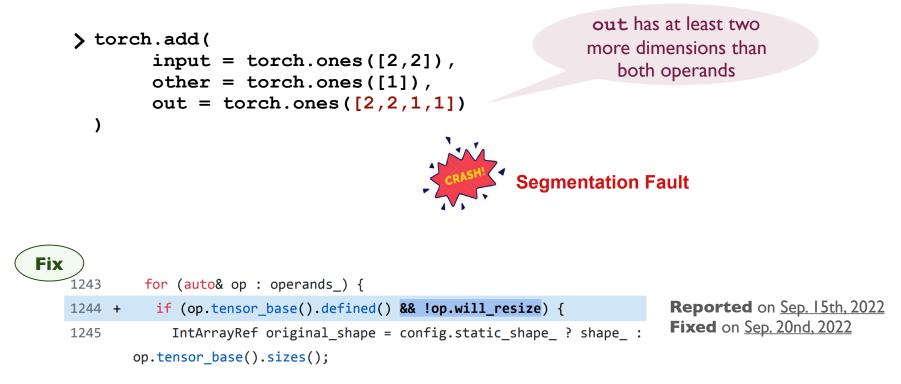
CEDAR leads to bugs being detected at least on average 338.6 days earlier

CEDAR detected **regression bug** through continuous testing



CEDAR shows its effectiveness in regression and continuous testing

A new high-priority bug affecting 23 PyTorch APIs



```
A new inconsistency bug
                                                   Zero values
original
   01 = tf.signal.stft(..., frame length=0, ...)
>
optimized
    @tf.function
                                                          Reported on May. 9th, 2022
                                                          Confirmed on May. 11th, 2022
    def fun wrapper(x):
         return tf.signal.stft(*x)
    O2 = fun wrapper(..., frame length=0, ...)
                                                                       Large
                                                                     inconsistencies
inconsistency
    np.max(o1 - o2) # (1.2623837153272947e+180+2.19373012209e-312j)
>
```

Effectiveness of the optimization strategies

Time efficiency

- shorten the execution time from 130:36 to 8:29
- reduce the **time overhead** by a factor of **3**

Space efficiency

- remove **3**, **I40**, **929** redundant files in total
- release **I 59.2 GB** space
- reduce the **space overhead** by a factor of **9.7**

Conclusion and Discussion

- We propose CEDAR, a continuous testing framework for DL libraries that efficiently integrates two state-of-the-art DL testing approaches to test DL libraries for detecting bugs effectively.
- CEDAR detected **83 bugs** affecting 140 PyTorch and TensorFlow APIs, including **23 previously unknown** bugs with **21** confirmed or fixed.
- The optimization strategies reduce the time and space overhead by a factor of 1.3 and 9.7.
- CEDAR's continuous application through 20 versions has effectively shortened the bug detection latency by almost a year (338.6 days).