PAFL: Enhancing Fault Localizers by Leveraging Project-Specific Fault Patterns

Artifact Overview

1 Introduction

This artifact aims to reproduce the results of PAFL in our paper "PAFL: Enhancing Fault Localizers by Leveraging Project-Specific Fault Patterns" submitted to OOPSLA 2025. By following this manual, you can replicate the results of PAFL as shown in Tables 2, 3, 4, 6, and 7 of the paper. Additionally, we provide instructions on how to reuse this artifact for further applications.

2 Hardware Dependencies

By default, this artifact is configured to reproduce the results of the *cpp_peglib* project, requiring 2GB of disk space. This default setup is chosen due to the substantial storage requirements for reproducing all results. To reproduce the results for all projects, 200GB of disk space is necessary. The complete datasets can be accessed via the following links: https://doi.org/10.5281/zenodo.14583938 and https://figshare.com/s/8d82745d78ade3bbab5d. To enable the artifact to reproduce results for all projects, extract the files and place them in the "data/source" and "data/coverage" directories.

3 Getting Started Guide

3.1 Setup

We provide a Docker image that contains all the dependencies. The following command builds the Docker image (this takes about 5 minutes):

\$./install.sh

The command will start the Docker container.

3.2 Verifying Installation (Basic Testing)

To verify the installation, please run the following command:

\$ train_and_run ochiai cpp_peglib

The necessary project sources and code coverage files for training and evaluation are stored in the directories, "data/source" and "data/coverage". The above command will train and run baseline FL (OCHIAI) and PAFL for each version of the *cpp_peglib* project. If the artifact is successfully installed, you will see the following results:

```
#Used threads: 1
Elapsed time (ochiai) : 0.1 secs
Elapsed time (ochiai + PAFL): 3.6 secs
Average overhead: 0.4 secs
```

The above results illustrate that the training and evaluation took 0.4 seconds on average for each version of the *cpp_peglib* project. The elapsed time reported in the above results varies depending on the experiment environment. In our work, all the experiments on PAFL are conducted on a machine with Intel Xeon Silver 4214 with 12 cores.

4 Step by Step Instructions

The following instructions will reproduce the results in Table 2, 3, 4, 6, and 7.

4.1 Reproducing Table 2

Following the instructions reproduce Table 2 of our paper.

Training and running PAFL models. To reproduce the results in Table 2 (PAFL for the SBFL baselines), start by training PAFL using the following command:

```
$ train_and_run <baseline>   -t <numthreads>
```

where baseline can be one of the following SBFL baselines:

```
ochiai, dstar, barinel
```

and project can be one of the following projects¹.

```
cpp_peglib, cppcheck, exiv2, libchewing, libxml2, proj,
  openssl, yaml_cpp, thefuck, fastapi, spacy, youtube-dl
```

In the above commane, numthreads determines the number of threads (default = 1) for training and evaluating the models. For example, if you type \$ train_and_run ochiai cpp_peglib -t 8, the command will train the model and produce the output of Ochiai and PAFL for each version of the cpp_peglib project using eight threads. The trained model will be stored in the "/opt/PAFL/profile" directory. If the training and evaluation are successfully finished, the command will produce the following results:

The above results illustrate that the training and running use eight threads and took 0.1 seconds on average for each version of the cpp_peglib project.

Reproducing the results in Table 2. After the training, type the following command to reproduce the results in Table 2:

```
evaluate <baseline> <project>
```

Then, the command will reproduce FR of each version, and MFR, MAR, Top-1, Top-5, and Top-10 for the project in Table 2. For example, if you type \$ evaluate ochiai cpp_peglib, the command will produce the following results:

¹ As noted in Section 2, the datasets projects (other than *cpp_peglib*) must first be downloaded and placed in the data/source and data/coverage directories before running the above command.

1	oc]	hiai	I	P	'AFL		
AR :	247	/1775 3/1775.	5	240	.9/	1775	
	==-	======					
otal							
Γotal ======	===: 	===== ochi	==== .ai	====	:=== 	 PAI	====== FL
Γotal ====== MFR	===: :						
		151.0/	 1426	 6.6	. <u>.</u>	138.8	 /1426.6
MFR	I	151.0/ 208.6/	/1426 /1426	 6.6	 	138.8	 /1426.6
MFR MAR	 	151.0/ 208.6/	 /1426 /1426 1	6.6 6.6	 	138.8	 /1426.6

The above results illustrate that FR and AR of PAFL for version $cpp_peglib\#10$ are 2 and 240.9, respectively. The total results (i.e., Total) reproduced the cell corresponds to row " cpp_peglib " and column "Ochiai" and "PAFL" in Table 2. For example, MFR, MAR, and Top-10 of PAFL for the cpp_peglib project are 138.8, 195.4, and 3, respectively. Other cells in Table 2 can be reproduced by changing the project and baseline in the command.

4.2 Reproducing Table 3

Since the baselines in Table 3 are deep learning-based techniques with inherent stochasticity, the following instructions may yield results that differ slightly from those reported in the paper. For reproducing the results of PAFL for the DLFL baselines in Table 3, run the following command:

\$ evalDLFL <baseline> -t <numthreads>

baseline can be one of the following DLFL baselines:

CNN, RNN, MLP

project can be one of the following projects for CNN and RNN:

 ${\tt cpp_peglib,\ libchewing,\ thefuck,\ fastapi,\ spacy}$

project can be one of the following projects for MLP:

For example, if you type \$ evalDLFL CNN cpp_peglib -t 8, the command will produce the following results:

-=====	==	etrics for 5 iter		====
	I	CNN I	PAFL	
MFR	 	328.6/1426.6	300.3/142	26.6
MAR		399.3/1426.6	385.7/142	26.6
Top-1	1	0.0		0.0
Top-5		0.0		0.0
Top-10	Ι	0.0		0.0

The above execution attempted to reproduce the results corresponding to the row "cpppeglib" and columns "CNN" and "PAFL" in Table 3, but the results slightly differ from those in the paper due to the stochastic nature of the DL-based approaches.

4.3 Reproducing Table 6

The instructions below reproduce the results in Table 6.

Training and running PAFL models. To fully reproduce the results (specifically the rows labeled "together") in Table 6, start by training PAFL with AENEAS using the following command:

```
$ train_and_run <baseline>   -t <numthreads>
```

To reproduce Table 6, baseline can be one of the following AENEAS combined with SBFL:

```
aeneas-ochiai, aeneas-dstar, aeneas-barinel
```

project can be one of the following projects:

```
cpp_peglib, cppcheck, libchewing, libxm12,
yaml_cpp, thefuck, fastapi, spacy, youtube-dl
```

If the training and running are successful, the command (e.g., \$ train_and_run aeneas-ochiai cpp_peglib -t 8) will produce the following results:

Reproducing the results in Table 6. To reproduce the results, run the following command:

```
evaluate <baseline> <project>
```

Then, the command will reproduce *MFR*, *MAR*, *Top-1*, *Top-5*, and *Top-10* for the project in Table 6. For example, if you type \$ evaluate aeneas-ochiai cpp_peglib, the command will produce the following results:

Total				
		aeneas-ochiai		PAFL
MFR		125.0/1426.6	 	118.9/1426.6
MAR	1	165.3/1426.6		159.2/1426.6
Top-1	1	0		0
Top-5	1	2		2
Top-10	Ι	2		2

Note that the above results only show the performance of AENEAS and PAFL only for the cpp_peglib project. To reproduce the cells in Table 6, where each cell presents the total performance across all the project, you need to integrate the results of all the other projects into a single result. To reproduce the cells in Table 6, where each cell represents the overall performance across all projects, you need to integrate the results from all the projects into a single aggregated result.

4.4 Reproducing Table 4

The instructions below reproduce the results in Table 4 (robustness of PAFL). Before proceeding, ensure that you have fully reproduced the results in Tables 2 and 3 by following the instructions in Section 4.1 and Section 4.2.

Reproducing the results in Table 4. To reproduce Table 4, run the following command:

```
robustness <baseline> <project>
```

Then, the command will reproduce the robustness of PAFL for the project. For example, if you type \$ robustness ochiai cpp_peglib, the command will produce the following results:

```
Baseline FL: ochiai
Target project: cpp_peglib
------
Acheiving equal or better FR in 10 out of 10 (100%) versions
```

The above results illustrate that 100% of versions of cpp_peglib are improved by PAFL or equal compared to OCHIAI.

4.5 Reproducing Table 7

Following the instructions below reproduce the results in Table 7.

Training with various configurations. To fully reproduce the results in Table 7, train PAFL using seven different configurations with the following command:

In the above command, config can be one of the following seven configurations:

```
1111, 0000, 2111, 1211, 1121, 1112, 2222
```

Again, project can be one of the following projects:

```
cpp_peglib, cppcheck, exiv2, libchewing, libxml2, proj,
  openssl, yaml_cpp, thefuck, fastapi, spacy, youtube-dl
```

Reproducing the results in Table 7. After the training, the following command will reproduce the results in Table 7:

```
evaluate ochiai <project> -u <config>
```

Then, the command will reproduce *MFR*, *MAR*, and *Top-10* for the project in Table 7. For example, if you type \$ evaluate ochiai cpp_peglib -u 2222, the command will produce the following results:

```
. . .
Total
ochiai
                  | PAFL (2222)
        151.0/1426.6 | 138.8/1426.6
MFR.
        208.6/1426.6 | 195.5/1426.6
Top-1
               1
                   -
                            2
Top-5
               3
                            3
                   -1
Top-10 |
               3
                            3
```

The above results illustrate that MFR, MAR, Top-1, Top-5, and Top-10 of the PAFL model with configuration (2,2,2,2) for the cpp_peglib project are 138.8, 195.5, 2, 3, and 3, respectively.

5 Reusability Guide

The artifact can be reused for improving other (baseline) fault localizers through the following steps:

5.1 Storing Suspiciousness Scores from Baseline Fault Localizers

To apply PAFL to a new fault localizer, you need to store the suspiciousness scores computed by the new fault localizer as follows:

In the above, method is the name of the baseline fault localizer. project is the target project name. version.json stores suspicious score for each line of a version in the project as follows:

In the above, file_name is the name of the source file. line_number is the line number (i.e., statement) in the source file. new_score is the suspiciousness score of the line. For example, if you

want to apply PAFL to a new FL method 'example_FL' in the cpp_peglib project, place the files as follows:

In the above, 1.json includes the suspicious scores of the cpp_peglib project for the first version computed by the 'example_FL' method. <pre

The above file shows that the suspiciousness score of the 2155th line in the file 'peglib.h' is 0.4, the 2156th line is 0.3, and the 2122th line is 0.2.

5.2 Applying PAFL

To train and run PAFL with the new baseline fault localizer, run the following command:

```
$ train_and_run <method>   -u <config> -t <numthreads>
```

To evaluate the performance of PAFL with the new scores, run the following command:

```
$ evaluate <method>   u <config>
```

For example, the following commands will produce the results of PAFL with the baseline FL method, 'example_FL', for the cpp_peglib project:

```
$ train_and_run example_FL cpp_peglib
$ evaluate example_FL cpp_peglib
```