Balancing Soundness and Efficiency for Practical Testing of Configurable Systems



Configurable Systems





Testing Configurable Systems System Configurations **Tests Monolithic Tests** Combinatorial explosion! System ...

. . .













Sampling (one-enabled)

SPLat

S-SPLat (one-enabled)





Notepad • 17 configuration variables Sampling (one-enabled) Only 3 are reached by toolbar() Test ... class Notepad { **17 configurations** \rightarrow void toolbar() { if (TOOLBAR) { 1... S-SPLat (one-enabled) (WORDCOUNT) **SPLat** if 11 (MENUBAR) if 11 . . . } 11 . . . 6 configurations 14 }

Notepad • 17 configuration variables Sampling (one-enabled) Only 3 are reached by toolbar() Test ... class Notepad { **17 configurations** woid toolbar() { if (TOOLBAR) { 1 . . . S-SPLat (one-enabled) (WORDCOUNT) **SPLat** if 11 (MENUBAR) if 11 . . . 2 configurations } 11 . . . 6 configurations 15 one-enabled }

S-SPLat

Input

Output









Tests executed with reachable and satisfiable configurations
C1, T1 C2, T1 C1, T2 C5, T2 C4, T3

S-SPLat

Output Input For all tests ... $Y_{es} \rightarrow Run the test T_i$ Instrumented Configurable System Find reachable variables Tests executed with reachable and satisfiable configurations C1, T1 Look for next reachable configuration Č2, T1 \ 🗅 \ C1, T2 Otherwise C5, T2 Tests C4, T3 🗅 ... 🛆 Sampling Heuristic **Check:** ß - Sampling heuristic - Feature model Feature Model (Optional) 17

EVALUATION

Research Questions

RQ1 → Which heuristics maximize efficiency (#samples)?

RQ2 \rightarrow Which heuristics maximize efficacy (#failures)?

RQ3 \rightarrow Which heuristics (basic or combination) maximize efficiency and efficacy?



Evaluation SPLs

Evaluation SPLs

Evaluation Techniques

SPLs

8 subjects

Techniques:

- 1. SPLat
- 2. SPLat + med
- 3. SPLat + oe
- 4. SPLat + od
- 5. SPLat + pw
- 6. SPLat + ran

[ICSE'16,ASE'14]



One-enabled (oe)









RQ1: Which heuristics maximize efficiency (#samples)?



RQ2: Which heuristics maximize efficacy (#failures)?



RQ3: Which heuristics maximize efficiency (#samples) and efficacy (#failures)?

Combinations of heuristics

Evaluation SPLs

•oe x od x med x pw

- **c1** = oe+od
- **c2** = oe+med
- **c3** = oe+pw ...
- •c11 = oe+od+med+pw

RQ3: Which heuristics maximize efficiency (#samples) and efficacy (#failures)?



SPLat+Most-enabled-disabled optimized #samples at the expense

optimized #samples at the expense of #failures

Evaluation SPLs

SPLat+c11 (oe + od + med + pw) optimized #failures at the expense of #samples

SPLat did not scale for some subjects

The sampling heuristics **reduced the number of samples** explored by SPLat yet retaining their ability to **reveal failures**.

Evaluation



Evaluation

Evaluation

Evaluation Techniques





RQ2: Which heuristics maximize efficacy (#failures)?



Evaluation SPLs



RQ3: Which heuristics maximize efficiency (#samples) and efficacy (#failures)?



It is preferable to pick the best performing heuristics in the leftmost group \rightarrow **the best choices**!

Bugs found







Avoid heuristics with a large number of requirements

S-SPLat found a good balance between bugs and samples The sampling heuristics helped to reduced the number of samples explored by SPLat without loss the ability to find bugs

S-SPLat could deal with scalability It revealed bugs in potentially large configuration spaces



BACKUP SLIDES





#samples





Discussion

- c2 found all crashes with a relatively low number of configurations
- •c7 performed better, it detected most failures and crashes through a relatively small number of configurations
- Combine different simple heuristics instead of using one that entails a larger number of test requirements
- **S-SPLat** is promising to reveal errors in potentially large configuration spaces

Handling Constraints



Complex models

- 54% of the selected configurations are invalid
- 43% of failures are false positives



The use of validation is not necessary

Crashes was only revealed in valid configurations

The techniques performed consistently with and without feature constraints



Additional Evaluations

S-SPLat x Regular Sampling

Regular Sampling detected the same bugs as S-SPLat with more configurations. Random Sampling with more rates: 10% and 30%

New results are proportional to the change in the sampling rates of random.

Threats to Validity and Limitations

- The selection of subjects
 - We used subjects from a variety of sources, including a large configurable system with hundreds of options
- Eventual implementation errors
 - We thoroughly checked our implementation and our experimental results
 - Our datasets and implementations are publicly available: https://sabrinadfs.github.io/s-splat/
- SPLat currently only supports systems with dynamically bound feature variables])
 - It remains to investigate how SPLat and S-SPLat would perform on systems with #ifdef variability