

# An Algorithm for Generating t-wise Covering Arrays from Large Feature Models

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Franck Fleurey

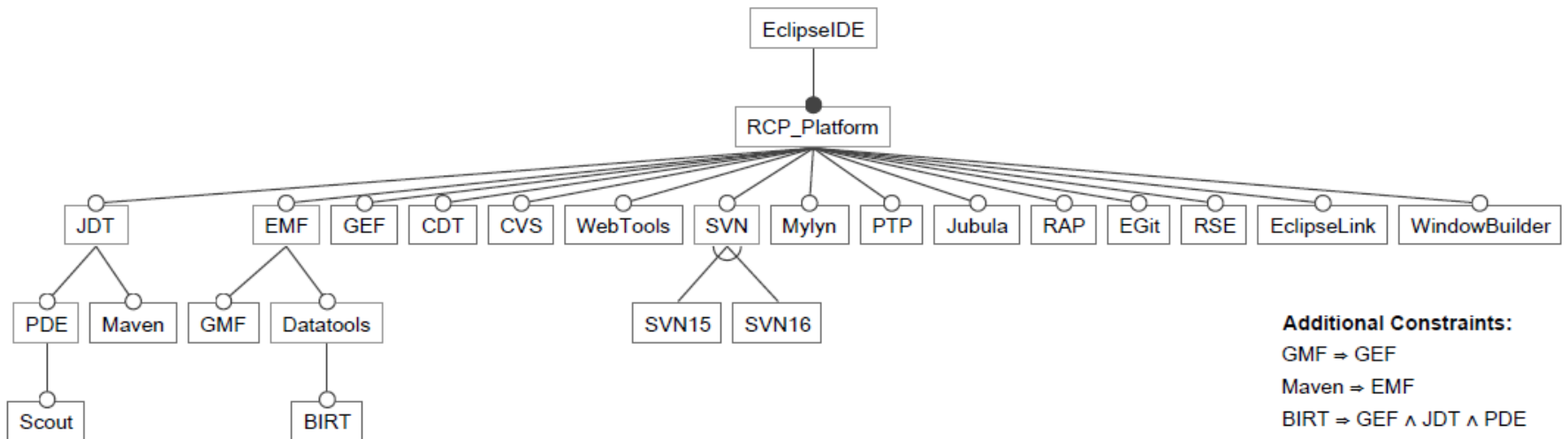


SPLC 2012 – Salvador, Brazil

# Example Product Line: The Eclipse IDEs

	 Java	 Java EE	 C/C++	 C/C++ Linux	 RCP/RAP	 Modeling	 Reporting	 Parallel	 Scout	 Testers	 Javascript	 Classic
RCP/Platform	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
CVS	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
EGit			✓	✓	✓	✓						
EMF	✓	✓				✓	✓					
GEF	✓	✓				✓	✓					
JDT	✓	✓			✓	✓			✓			✓
Mylyn	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	
Web Tools		✓					✓				✓	
Linux Tools			✓	✓				✓				
Java EE Tools		✓					✓					
XML Tools	✓	✓			✓		✓	✓				
RSE		✓	✓	✓			✓	✓				
EclipseLink		✓					✓			✓		
PDE		✓			✓	✓	✓		✓			✓
Datatools		✓					✓					
CDT			✓	✓				✓				
BIRT							✓					
GMF						✓						
PTP								✓				
MDT						✓						
Scout									✓			
Jubula										✓		
RAP					✓							
WindowBuilder	✓											
Maven	✓											

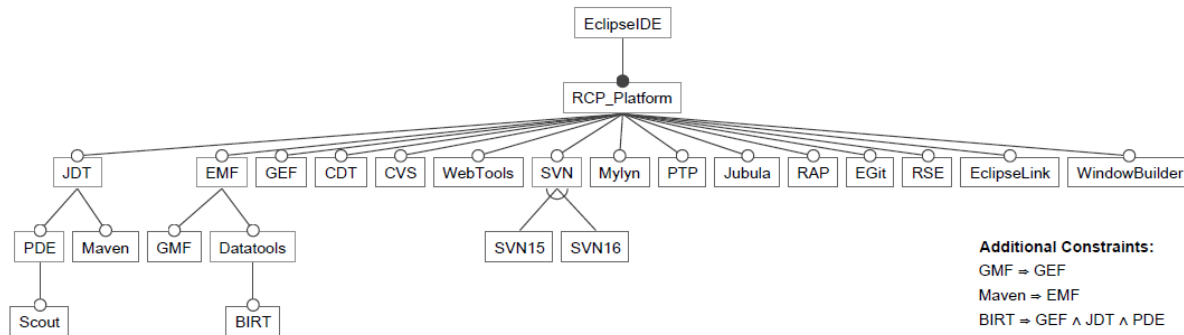
# Constraints Between Features



356,352 possible products

# Product Line Verification

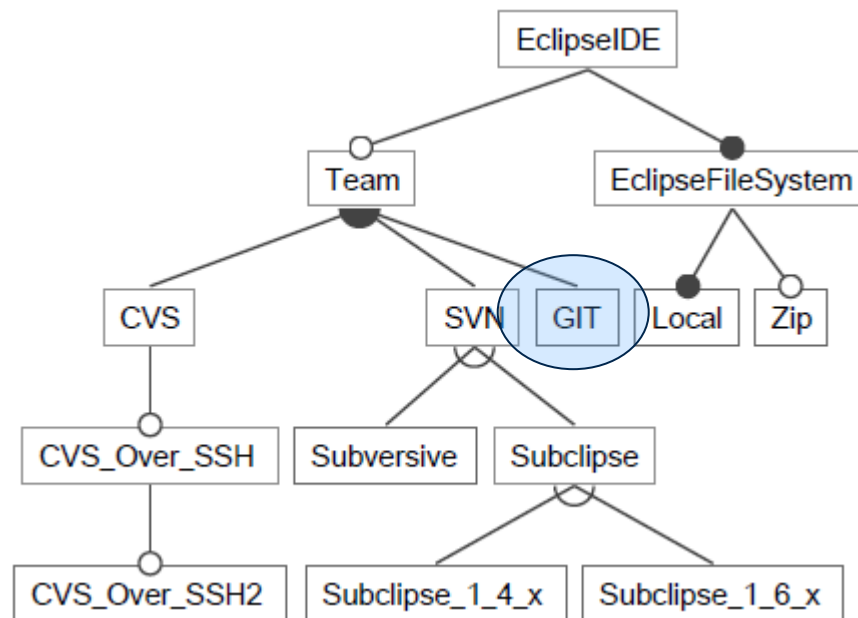
■ How do we gain confidence that any valid product works?



	Java	Java EE	C/C++	C/C++ Linux	RCP/RAP	Modeling	JEE BIRT Reporting	Parallel	Scout	Testers	Javascript	Classic
RCP/Platform	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
CVS	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
EGit			✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
EMF	✓	✓					✓					
GEF	✓	✓					✓					
JDT	✓	✓			✓	✓	✓		✓			✓
Mylyn	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
Web Tools	✓						✓				✓	
Linux Tools			✓	✓				✓				
Java EE Tools		✓					✓					
XML Tools	✓	✓			✓		✓	✓				
RSE	✓	✓	✓	✓			✓	✓				
EclipseLink	✓						✓			✓		
PDE					✓	✓	✓		✓			✓
Datatools		✓					✓					
CDT			✓	✓				✓				
BIRT							✓					
GMF						✓						
PTP								✓				
MDT						✓						
Scout									✓			
Jubula										✓		
RAP					✓							
WindowBuilder		✓										
Maven		✓										

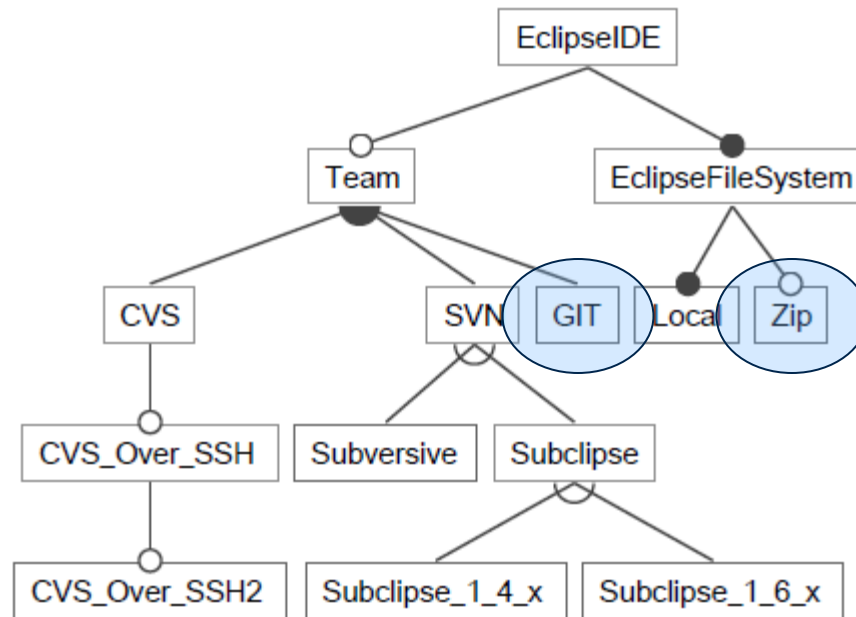
# Faulty Features

- Unit tests may find faults inside a single feature.
  - n test suites required for a product line with n features.
- What about faulty cooperation between features?
  - What if they *interact* incorrectly?



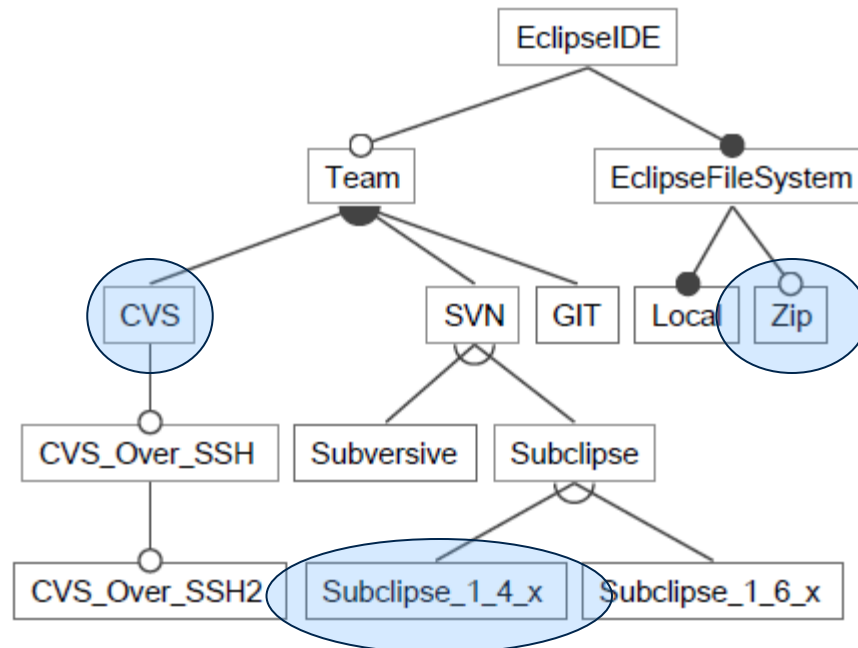
# Interaction Faults

- 2-wise interaction fault
  - reproducible by including 2 specific features
  - the others do not matter



# Interaction Faults

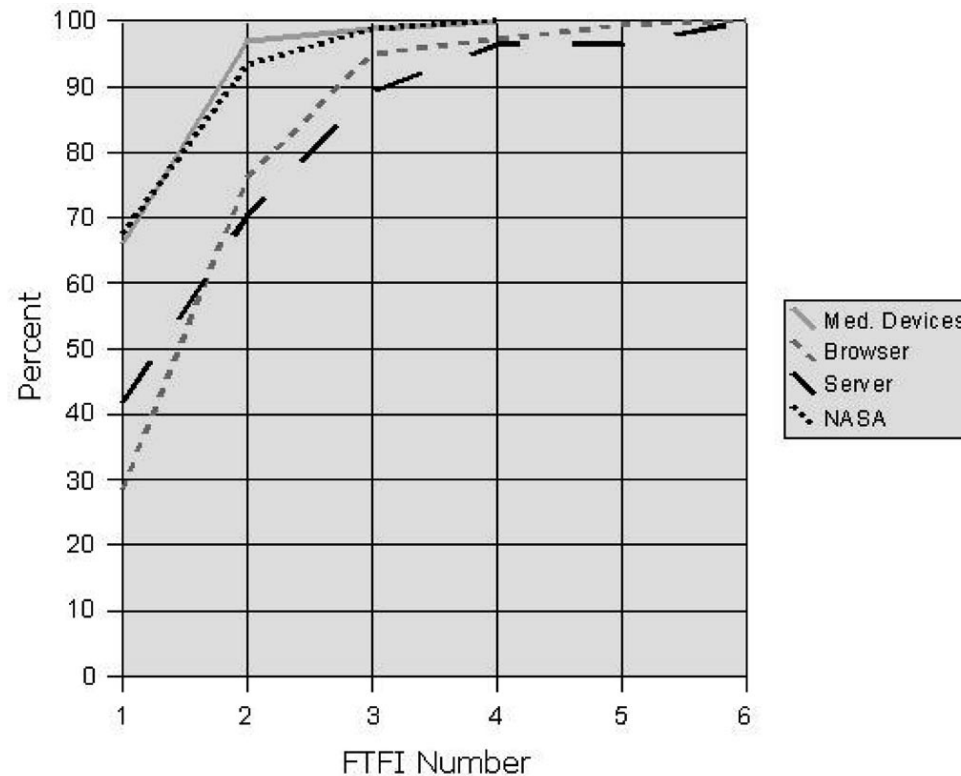
- 3-wise interaction fault
  - reproducible by including 3 specific features
  - the others do not matter



# Empirics Show:

## ■ Kuhn et al. 2004:

- Most bugs can be attributed to the interaction of a few features.

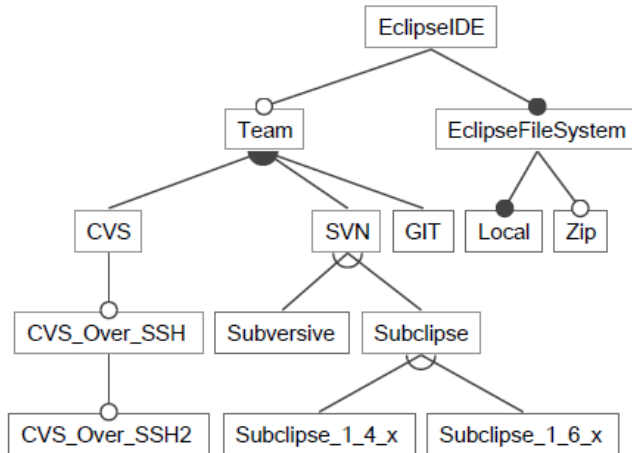




# Covering Arrays

## Mathematical property:

- Only a few products needed to cover all simple interactions




Feature\Product	1	2	3	4	5	6	7	8	9	10	11	12
EclipseIDE	X	X	X	X	X	X	X	X	X	X	X	X
Team	X	X	-	X	X	X	X	X	X	X	-	X
CVS	X	X	-	X	X	-	X	-	X	-	-	-
CVS_Over_SSH	X	-	-	X	X	-	X	-	X	-	-	-
CVS_Over_SSH2	-	-	-	X	X	-	X	-	X	-	-	-
SVN	X	X	-	X	X	X	-	X	X	X	-	-
Subversive	X	-	X	-	X	-	-	X	-	-	-	-
Subclipse	-	X	-	-	X	X	-	-	X	X	-	-
Subclipse_1_4_x	-	X	-	-	X	-	-	-	-	X	-	-
Subclipse_1_6_x	-	-	-	-	-	X	-	-	X	-	-	-
GIT	X	-	-	-	X	X	-	X	-	-	-	X
EclipseFileSystem	X	X	X	X	X	X	X	X	X	X	X	X
Local	X	X	X	X	X	X	X	X	X	X	X	X
Zip	-	X	X	X	-	-	-	X	X	-	-	-

## Other examples (pair-wise testing):

- For the "e-shop product line" with 287 features: 21 products
- For the Linux kernel with almost 7,000 features: 480 products



# Combinatorial Interaction Testing (CIT)

1. Generate a covering array  scalability issue
  - Can be reused until the feature model is changed
2. Build each product
3. Apply a single system testing technique to each product

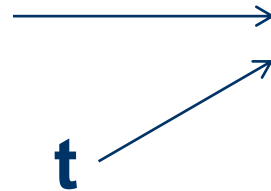
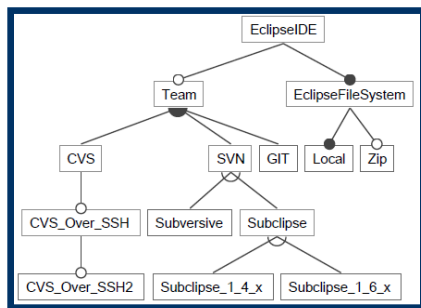
- Note: CIT was originally intended for single system testing
  - Covering arrays over input instead of interactions.

Feature\Product	1	2	3	4	5	6	7	8	9	10	11	12
EclipseIDE	X	X	X	X	X	X	X	X	X	X	X	X
Team	X	X	-	X	X	X	X	X	X	X	-	X
CVS	X	X	-	X	X	-	X	-	X	-	-	-
CVS_Over_SSH	X	-	-	X	X	-	X	-	X	-	-	-
CVS_Over_SSH2	-	-	-	X	X	-	X	-	X	-	-	-
SVN	X	X	-	X	X	X	-	X	X	X	-	-
Subversive	X	-	-	X	-	-	-	X	-	-	-	-
Subclipse	-	X	-	-	X	X	-	-	X	X	-	-
Subclipse_1.4_x	-	X	-	-	X	-	-	-	-	X	-	-
Subclipse_1.6_x	-	-	-	-	-	X	-	-	X	-	-	-
GIT	X	-	-	-	X	X	-	X	-	-	-	X
EclipseFileSystem	X	X	X	X	X	X	X	X	X	X	X	X
Local	X	X	X	X	X	X	X	X	X	X	X	X
Zip	-	X	X	X	-	-	-	X	X	-	-	-

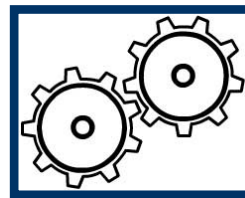
# Background

- Our MODELS 2011 paper concludes:
  - Covering array generation is tractable in practice.
    - Difficult to satisfy FMs imply no products to sell, which is absurd.
  - An efficient algorithm was not provided.
    - 2-wise testing limit: about 500 features
    - 3-wise testing limit: about 200 features
- An efficient algorithm is contributed in this paper.
  - 2-wise testing
    - Now works for the Linux Kernel feature model (6888 features)
  - 3-wise testing
    - Now works for the eCos feature model (1244 features)
    - (An optimized C/C++ implementation + some good hardware should work even for the Linux Kernel feature model.)

# Overview of the Algorithm



ICPL



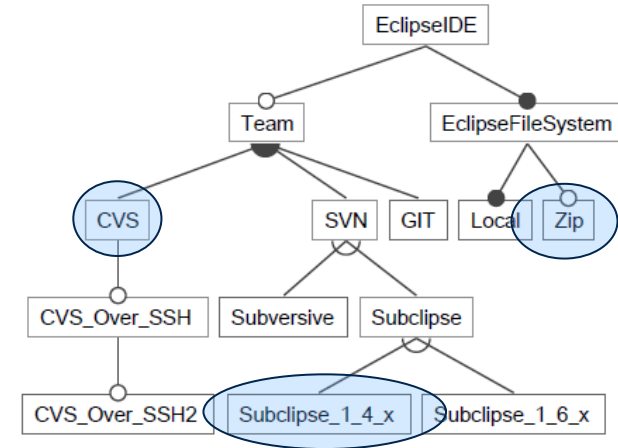
Feature\Product	1	2	3	4	5	6	7	8	9	10	11	12
EclipseIDE	X	X	X	X	X	X	X	X	X	X	X	X
Team	X	X	-	X	X	X	X	X	X	X	-	X
CVS	X	X	-	X	X	-	X	-	X	-	-	-
CVS_Over_SSH	X	-	X	X	-	X	-	X	-	-	-	-
CVS_Over_SSH2	-	-	-	X	X	-	X	-	X	-	-	-
SVN	X	X	-	X	X	X	-	X	X	X	-	-
Subversive	X	-	X	-	-	X	-	-	-	-	-	-
Subclipse	-	X	-	-	X	X	-	X	X	-	-	-
Subclipse_1.4.x	-	X	-	-	X	-	-	-	X	-	-	-
Subclipse_1.6.x	-	-	-	-	X	-	-	X	-	-	-	-
GIT	X	-	-	X	X	-	X	-	-	-	-	X
EclipseFileSystem	X	X	X	X	X	X	X	X	X	X	X	X
Local	X	X	X	X	X	X	X	X	X	X	X	X
Zip	-	X	X	X	-	-	X	X	-	-	-	-

## Implementation Supports

- Simple XML Feature Models (SXFM)
- GUI DSL
- DIMACS
- CVL (Proposed OMG standard)

## CSV-file

# Groundwork



## ■ Data Structures

- $a = (\text{feature, included})$  – an assignment
- $e = \{a_1, a_2, \dots, a_t\}$  – a t-set – a set of t assignments
- $T_t$  – the set of all t-sets
- $I_t$  – the set of all invalid t-sets
- $U_t$  – the set of all valid t-sets (the "universe")
- $C$  – configuration – a set of assignments, one for each feature
- $CA_t = \{C_1, C_2, \dots, C_x\}$  – a Covering Array of strength t

## ■ Equations

- $|T_t| = 2^t \binom{f}{t}$ , i.e. 95 million pair-wise interactions for the Linux kernel
  - Empirically,  $|I_t| \ll |U_t|$
- $CA_{t-1} \subseteq CA_t$ , thus, generating  $CA_{t-1}$  before  $CA_t$  is an option.

- Recursive generation of covering arrays of lesser strength

- Greedy Loop

- Fit as many as possible
- Remove those covered
- ... and repeat until all interactions are covered

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Algorithm 2  $ICPL(FM, t) : (C_t, I_t)$

---

```
1: if  $t = 1$  then
2:    $(C_t, I_t) \leftarrow genCompleteI_1(FM)$ 
3:    $T_t \leftarrow FM.genT(1) \setminus I_t$ 
4:    $invalidRemoved \leftarrow true$ 
5: else
6:    $(C_t, I_{t-1}) \leftarrow ICPL(FM, t - 1)$ 
7:    $(T_t, I_t) \leftarrow generateTsets(FM.getT(t), I_{t-1}, C_t)$ 
8:    $invalidRemoved \leftarrow false$ 
9: end if
10: while  $T_t \neq \emptyset$  do
11:    $C \leftarrow genConfiguration(FM, T_t)$ 
12:    $C_t \leftarrow C_t \cup \{C\}$ 
13:    $CO \leftarrow getCovered(FM, C, T_t)$ 
14:    $T_t \leftarrow T_t \setminus CO$ 
15:   if  $(\neg invalidRemoved) \wedge (\lfloor \log_{10} |CO| \rfloor \leq \lfloor \log_{10} |FM| \rfloor)$ 
16:     then
17:        $(T_t, TempI) \leftarrow genInvalid(FM, T_t)$ 
18:        $I_t \leftarrow I_t \cup TempI$ 
19:        $invalidRemoved \leftarrow true$ 
20:     end if
21: end while
22: return  $(C_t, I_t)$ 
```

---

- Not parallel

**Algorithm 6** *genConfiguration*( $FM, T$ ) :  $C$

```
1:  $(C, CF, PF) \leftarrow (\emptyset, \emptyset, \emptyset)$ 
2: for each t-set  $e$  in  $T$  do
3:   for each assignment  $a$  in  $e$  do
4:      $PF \leftarrow PF \cup \{a.f\}$ 
5:   end for
6: end for
7: Loop:
8: for each t-set  $e$  in  $T$  do
9:    $F \leftarrow \emptyset$ 
10:  for each assignment  $a$  in  $e$  do
11:     $F \leftarrow F \cup \{a.f\}$ 
12:  end for
13:  if  $F \subseteq CF$  then
14:    continue
15:  end if
16:  if  $FM.is\_satisfiable(C \cup e)$  then
17:     $C \leftarrow C \cup e$ 
18:     $CF \leftarrow CF \cup F$ 
19:  else
20:    for each assignment  $a$  in  $e$  do
21:      if  $((e \setminus \{a\}) \subseteq C) \wedge (a \notin C)$  then
22:         $b \leftarrow a$  with assignment inverted
23:         $C \leftarrow C \cup \{b\}$ 
24:         $CF \leftarrow CF \cup \{b.f\}$ 
25:      end if
26:    end for
27:  end if
28:  if  $|CF| = |PF|$  then
29:    break Loop
30:  end if
31: end for
32: return  $FM.satisfy(C)$ 
```

■ Pick an interaction

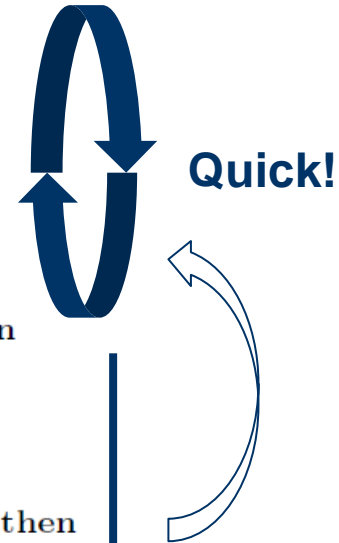
- Skip if covered

■ Does it fit the product?

- yes
- no

■ Make sure all features are assigned

■ Not parallel



## Algorithm 3

- Is the assignment valid? →

Algorithm 3 *getInvalidAssignments*( $FM, T_1$ ) :  $I_1$

```
1:  $I_1 \leftarrow \emptyset$ 
2: for each t-set  $e$  in  $T_1$  do
3:   if  $\neg FM.is\_satisfiable(e)$  then
4:      $I_1 \leftarrow I_1 \cup \{e\}$ 
5:   end if
6: end for
7: return  $I_1$ 
```

## Algorithm 7

- For all uncovered interactions →
- Is the interaction covered? →

Algorithm 7 *genCovered*( $FM, C, T$ ) :  $CO$

```
1:  $CO \leftarrow \emptyset$ 
2: for each t-set  $e$  in  $T$  do
3:   if  $e \subseteq C$  then
4:      $CO \leftarrow CO \cup \{e\}$ 
5:   end if
6: end for
7: return  $CO$ 
```

## Algorithm 8

- Pick an interaction →
- Check if it is invalid →

Algorithm 8 *genInvalid*( $FM, T$ ) : ( $V, I$ )

```
1: ( $I, V$ )  $\leftarrow$  ( $\emptyset, \emptyset$ )
2: while  $T \neq \emptyset$  do
3:    $e \leftarrow$  any t-set in  $T$ 
4:   if  $\neg FM.is\_satisfiable(e)$  then
5:      $I \leftarrow I \cup \{e\}$ 
6:      $T \leftarrow T \setminus \{e\}$ 
7:   else
8:      $V \leftarrow V \cup \{e\}$ 
9:      $T \leftarrow T \setminus \{e\}$ 
```

- These are data-parallel sub-algorithms



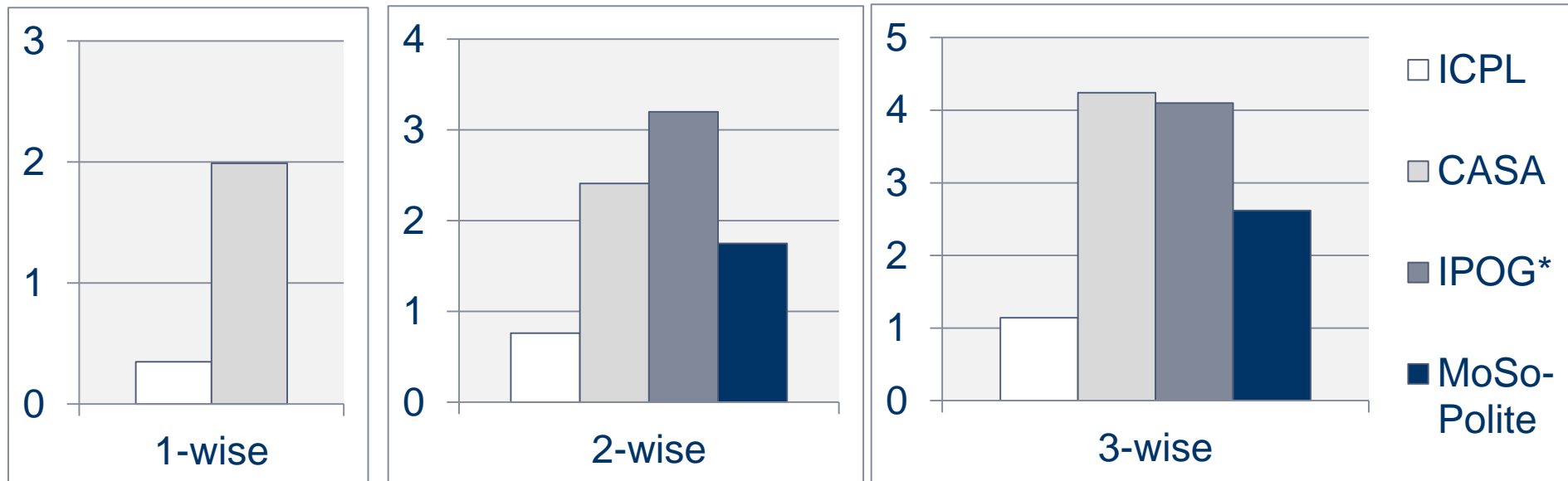
# Compared to Other Tools

- ICPL – our new algorithm
- CASA – Simulated annealing algorithm by Garvin et al.
- MoSo-Polite – algorithm by Oster et al.
- IPOG – algorithm by Lei et al.
  
- Experiment Machine
  - Could execute 6 threads in parallel
  - 32 GiB RAM

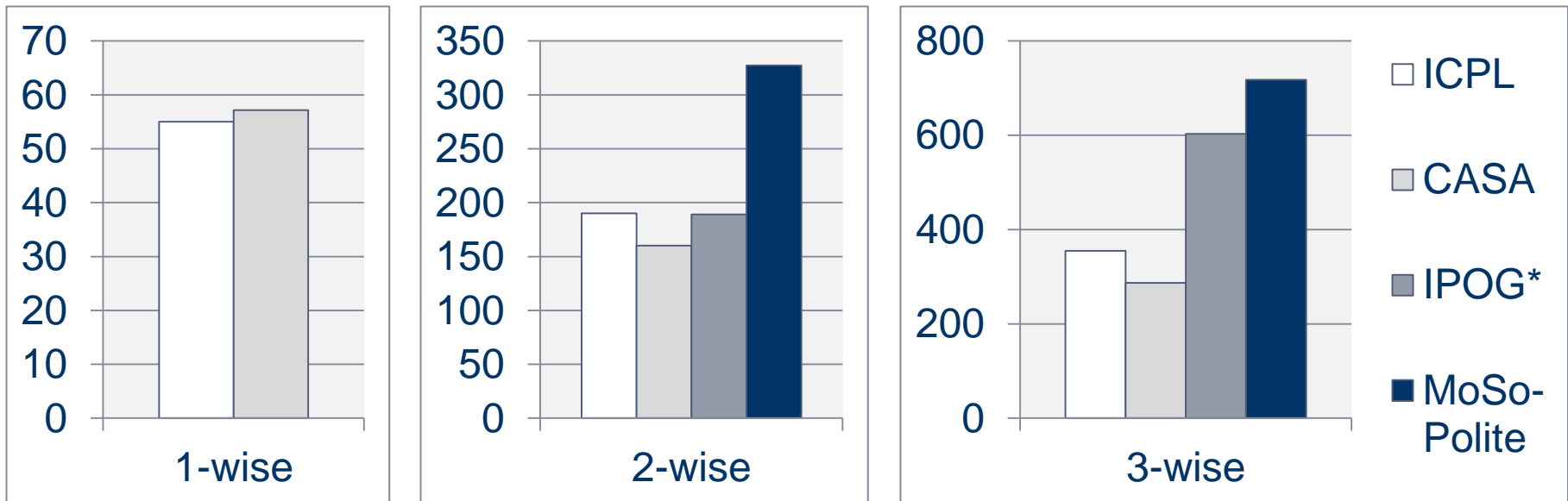
# Time Taken to Generate

## Statistic estimates

- The  $c$  in  $O(f^c)$  where  $f$  is the number of features



# Size of Covering Arrays



# Large Feature Models

Feature Model\keys	Features	Constraints (CNF clauses)	SAT time (ms)	1-wise size	1-wise time (s)	2-wise size	2-wise time (s)	3-wise size	3-wise time (s)
2.6.28.6-icse11.dimacs	6,888	187,193	125	25	89	480	33,702	n/a	n/a
freebsd-icse11.dimacs	1,396	17,352	18	9	10	77	240	*78	*2,540
ecos-icse11.dimacs	1,244	2,768	12	6	2	63	185	*64	*973
Eshop-fm.xml	287	22	5	3	0.16	21	5	108	457

■ ■ ■

# Summary

## ■ Our contribution

- A scalable algorithm for t-wise (1-3) covering array generation
- An empirical evaluation and comparison

## ■ Implementation available

- The implementation is available as open source (EPL)
- Experiments are reproducible

## ■ All the data is available

- All 28,500 measurements available for the paper's resource website + charts and summaries