

SAX-PAC (Scalable And eXpressive PAcKet Classification)

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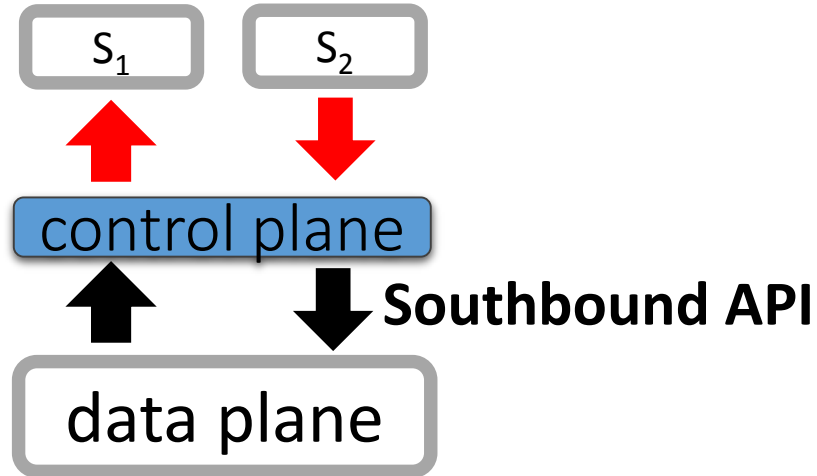
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Outline

- Current state of the art in packet classification
- Impact of structural properties on representation efficiency
- Classifiers as Boolean expressions
- Proposed solutions
- Evaluation
- Summary and future work

Representation of expressiveness on data plane



$$\begin{aligned} R_1 &= ([1, 3], [4, 31], [1, 28]) \\ R_2 &= ([4, 4], [2, 30], [4, 27]) \\ R_3 &= ([7, 9], [5, 21], [3, 18]) \end{aligned} \quad \begin{array}{c} \uparrow \\ \text{priority} \end{array}$$

SW-based vs. TCAM-based solutions

SW-based: $N = 4$ rules $K = 2$ fields ~~prefixes~~ ranges

$$\begin{aligned}
 R_1 &= (100*, 001*) \\
 R_2 &= (1010, 0001) \\
 R_3 &= (000*, ****) \\
 R_4 &= (001*, ****)
 \end{aligned}$$

Memory	Lookup time
$O(N)$	$O(\log^{k-1}N)$
$O(N^k)$	$O(\log N)$

TCAM-based: $N = 3$ rules $K = 3$ prefixes ranges

$$\begin{aligned}
 R_1 &= ([1, 3], [4, 31], [1, 28]) \\
 R_2 &= ([4, 4], [2, 30], [4, 27]) \\
 R_3 &= ([7, 9], [5, 21], [3, 18])
 \end{aligned}$$

Encoding	#TCAM entries
Binary	42+28+50=120
Gray	24+8+32=64

Order-independence

If the rules of a classifier do not “intersect”, their order is not important.

$$\begin{aligned} R_1 &= ([1, 3], [4, 31], [1, 28]) \\ R_2 &= ([4, 4], [2, 30], [4, 27]) \\ R_3 &= ([7, 9], [5, 21], [3, 18]) \end{aligned}$$

- Example: prefixes of the same length
- Implicit creation of order-dependence for service policies

	cisco1	cisco2	cisco3	fw	ipc	acl
Order-independent rules	120	249	329	39962	48294	49779
Total	148	269	364	45723	49840	49870
Order-independent %	81	93	90	87	97	99

Exploiting order-independence

- Adding new fields keep order-independence
- At most one rule is matched and it can be false-positive
- We can reduce space by skipping new fields

$$R_1 = ([1, 3], [4, 31])$$

$$R_2 = ([4, 4], [2, 30])$$

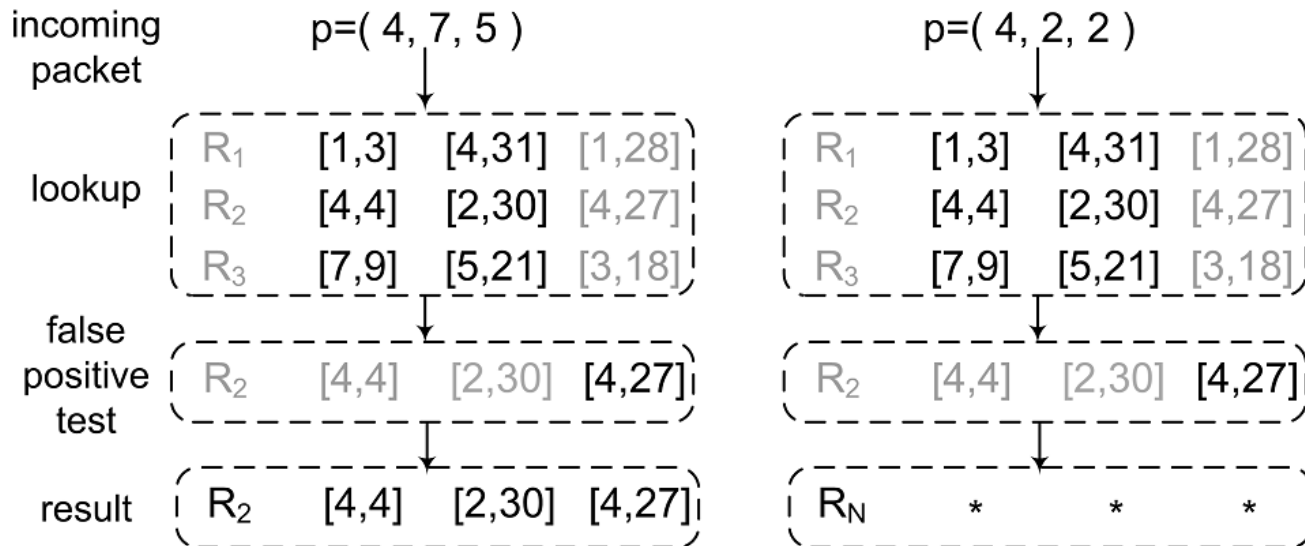
$$R_3 = ([7, 9], [5, 21])$$

$$R_1^{+1} = ([1, 3], [4, 31], [1, 28])$$

$$R_2^{+1} = ([4, 4], [2, 30], [4, 27])$$

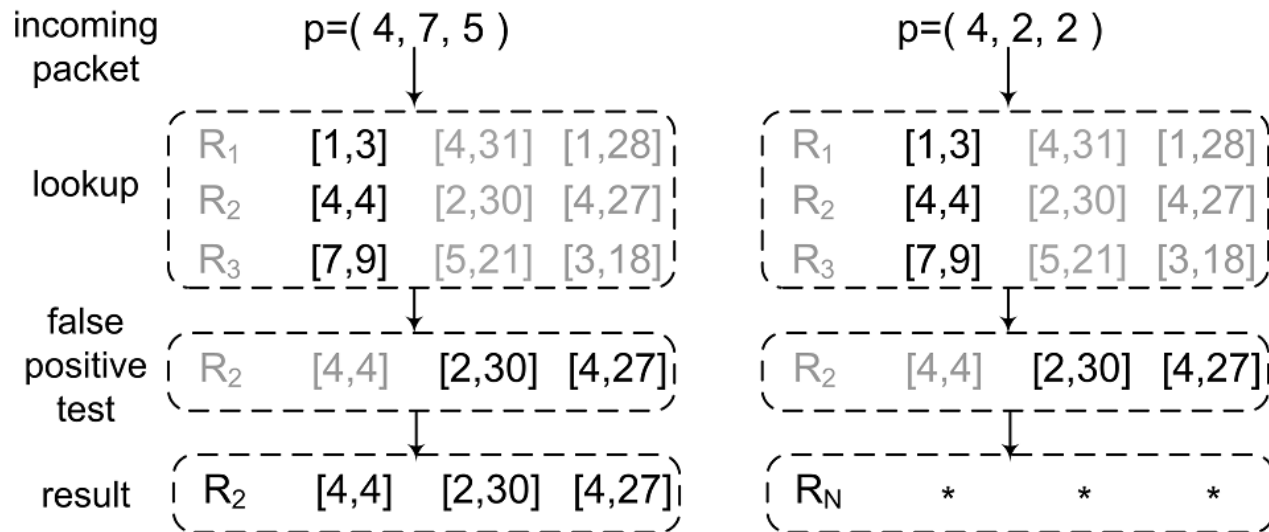
$$R_3^{+1} = ([7, 9], [5, 21], [3, 18])$$

#Fields	Bin Encoding	Gray Encoding
2	6+7+10=23	6+4+8=18
3	42+28+50=120	24+8+32=64



Fields subset minimization (FSM)

Problem 1: Find a maximal subset M of fields of an order-independent classifier K s.t. K^{-M} is order-independent

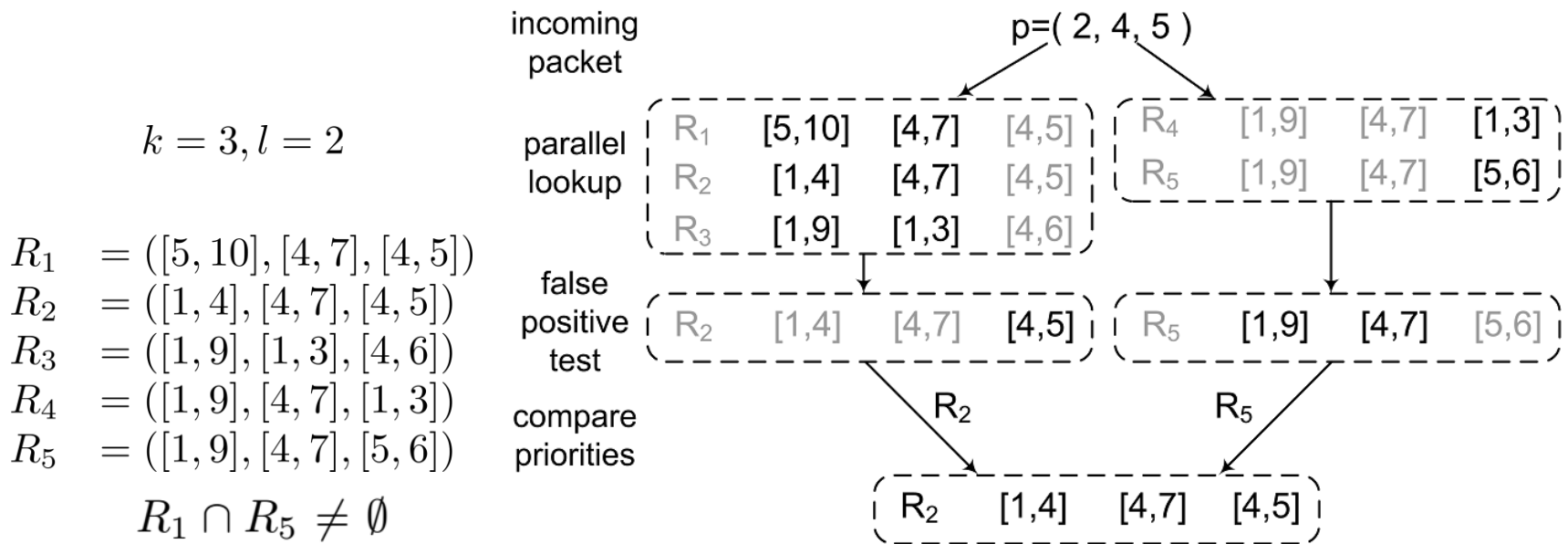


Fields	Binary Encoding	Gray Encoding
{1,2,3}	42+28+50=120	24+8+32=64
{1,2}	6+7+10=23	6+4+8=18
{1}	2+1+2=5	2+1+2=5

Multi-group Representation

A classifier K is irreducible or order-dependent?

Problem 2: Given a classifier K on k fields and $0 < l < k$. Find an assignment of rules to a minimal number of disjoint groups s.t. every group is order-independent on l fields.



Multi-group representation of rules subset

Number of groups $>$ supported level of parallelism β ?

Problem 3: Given a classifier K on k fields, $0 < l < k$, and $\beta > 0$. Find an assignment of a maximal subset of rules to β disjoint groups s.t. every group is order-independent on l fields.

$$k = 3, l = 2$$

$$\begin{aligned} R_1 &= ([5, 9], [4, 4], [4, 4]) \\ R_2 &= ([2, 4], [5, 7], [5, 5]) \\ R_3 &= ([2, 3], [1, 4], [4, 6]) \\ R_4 &= ([1, 5], [1, 7], [1, 3]) \\ R_5 &= ([1, 9], [1, 7], [1, 6]) \end{aligned}$$

$$\{R_1, R_2, R_3, R_4\}$$

Order-independent part (I)

$$\left\{ \begin{array}{l} R_1 \ [5,9] \ [4,4] \ [4,4] \\ R_2 \ [2,4] \ [5,7] \ [5,5] \\ R_3 \ [2,3] \ [1,4] \ [4,6] \end{array} \right\} \left\{ R_4 \ [1,5] \ [1,7] \ [1,3] \right\}$$

The rest (D)

$$R_5 \ [1,9] \ [1,7] \ [1,6]$$

Order-independent part (I)

$$\left\{ \begin{array}{l} R_1 \ [5,9] \ [4,4] \ [4,4] \\ R_2 \ [2,4] \ [5,7] \ [5,5] \\ R_4 \ [1,5] \ [1,7] \ [1,3] \end{array} \right\}$$

The rest (D)

$$\left\{ \begin{array}{l} R_3 \ [2,3] \ [1,4] \ [4,6] \\ R_5 \ [1,9] \ [1,7] \ [1,6] \end{array} \right\}$$

Max order-independent set on 3 fields

Classifiers as Boolean expressions

$$R_1 = (100*, 001*)$$

$$R_2 = (1010, 0001)$$

$$R_3 = (000*, ***)$$

$$R_4 = (001*, ****)$$

$$(x_1 \wedge \bar{x}_2 \wedge \bar{x}_3 \wedge \bar{x}_5 \wedge \bar{x}_6 \wedge x_7) \vee$$

$$(x_1 \wedge \bar{x}_2 \wedge x_3 \wedge \bar{x}_4 \wedge \bar{x}_5 \wedge \bar{x}_6 \wedge \bar{x}_7 \wedge x_8) \vee$$

$$(\bar{x}_1 \wedge \bar{x}_2 \wedge \bar{x}_3) \vee$$

$$(\bar{x}_1 \wedge \bar{x}_2 \wedge x_3)$$

MinDNF: For a given Boolean function f , find a minimal size of DNF for f

MinDNF:

$$R_1 = (100*, 001*)$$

$$R_2 = (1010, 0001)$$

$$R_{34} = (00**, ***)$$

**FSM with per
field resolution:**

$$R_1^{-1} = (100*)$$

$$R_2^{-1} = (1010)$$

$$R_3^{-1} = (000*)$$

$$R_4^{-1} = (001*)$$

**FSM with per
bit resolution:**

$$R_1^{-6} = (10)$$

$$R_2^{-6} = (11)$$

$$R_3^{-6} = (00)$$

$$R_4^{-6} = (01)$$

Exact algorithm

FSM is NPC (reduction from SetCover)

Algorithm FSMBINSEARCH(k, min, max)

```
1:  $m = \lfloor \frac{min+max}{2} \rfloor$ 
2: if  $min = max$  then return  $m$ 
3: for  $M \subseteq \{1, \dots, k\}, |M| = m$  do
4:   if ISORDERINDEPENDENT( $M$ ) then
5:      $min = m$ 
6:     return FSMBINSEARCH( $k, min, max$ )
7:  $max = m - 1$ 
8: return FSMBINSEARCH( $k, min, max$ )
```

Theorem: $FSMBinSearch(k, 0, k - 1)$ runs in time $O(k2^{k-1}N^2)$

Approximate algorithms

$$\begin{array}{rcl} & S_1 & S_2 & S_3 \\ R_1 & = & ([1, 3], [4, 31], [1, 28]) \\ R_2 & = & ([4, 4], [2, 30], [4, 27]) \\ R_3 & = & ([7, 9], [5, 21], [28, 31]) \end{array}$$

Heuristics: algorithms for *SetCover* and *MaxSetCoverage*

$$U = \{(i, j) \mid i < j, i, j \in [1, N]\} = \{(1, 2), (1, 3), (2, 3)\}$$

Define k sets S_1, \dots, S_k (one per field) to cover U

$S_i, 1 \leq i \leq k$, contains all pairs of rules that do not intersect in this field

$$S_1 = \{(1, 2), (1, 3), (2, 3)\} \quad S_2 = \emptyset \quad S_3 = \{(2, 3)\}$$

Theorem: *FSM* is reducible to *SetCover* in $O(kN^2)$ time with approximation factor $2 \ln(N) + 1$

Evaluation

6 classifiers with real parameters (see paper for more examples)

#	Total	Max OI part 5 fields			Multi-group representation									
		OI size	FSM	{0,1}	1-field groups					2-field groups				
rules	OI size	FSM	{0,1}	all	95%	99%	≤ 2	≤ 5	all	95%	99%	≤ 2	≤ 5	
1	584	538	0,1,3,4	406	15	8	13	2	8	10	4	7	3	4
2	269	249	0,1,4	246	4	2	3	1	1	2	2	2	0	0
3	364	329	0,1,3,4	324	7	3	5	2	4	4	2	3	1	14
4	49870	49779	0,1,4	49768	16	1	1	6	9	12	1	1	5	7
5	47276	44178	0,1,3,4	43819	67	5	13	19	32	39	2	5	10	20
6	48885	48826	0,1,2,4	48755	20	3	3	15	15	12	1	1	5	9

Summary

- Semantically equivalent: more fields imply less efficient representation
- Representation with false-positive: more fields – more efficient representation
- Structural properties can significantly improve time-space trade-off
- No restrictions on representation of every group (we define only additional abstraction layer)

Ongoing and future work

- Consider special cases as a FIB representation

G. Retvari et al., **Compressing IP Forwarding Tables: Towards Entropy Bounds and Beyond** SIGCOMM 13

- Identify additional structural properties
- Composition of structural properties
- Application to neighboring fields: data bases, program optimization, etc.

Thank you