

High-Performance Traffic Classification on GPU

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Introduction & Background

Traffic classification

Network management

- Flow prioritization
- Traffic shaping
- Traffic policing

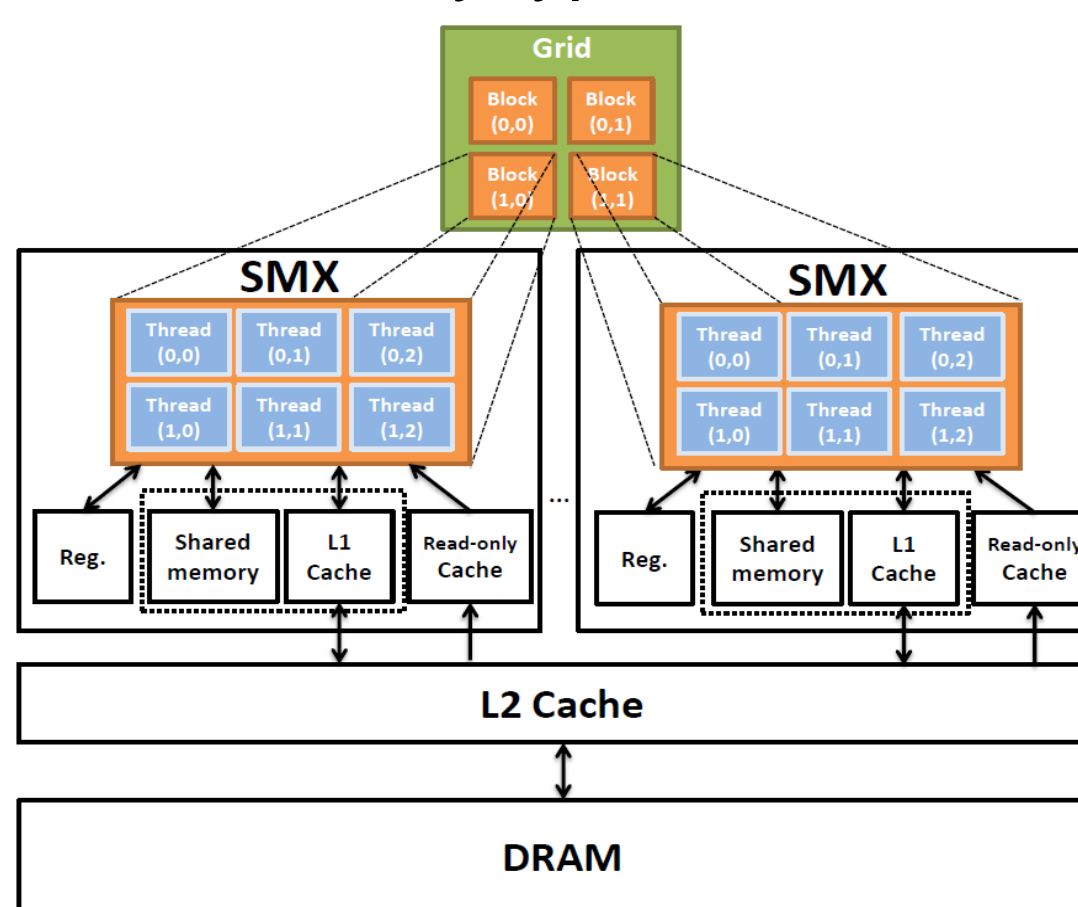
Approaches

- Transport layer port number-based
 - Port numbers can be dynamically allocated
- Deep packet inspection-based
 - Cannot handle encrypted traffic
- Heuristic-based
 - Low accuracy
- Statistics-based
 - Machine learning
 - High accuracy



CUDA programming model

- Host function + Kernel function
- SIMT execution model
- Various memory types on GPU



C4.5 Decision tree

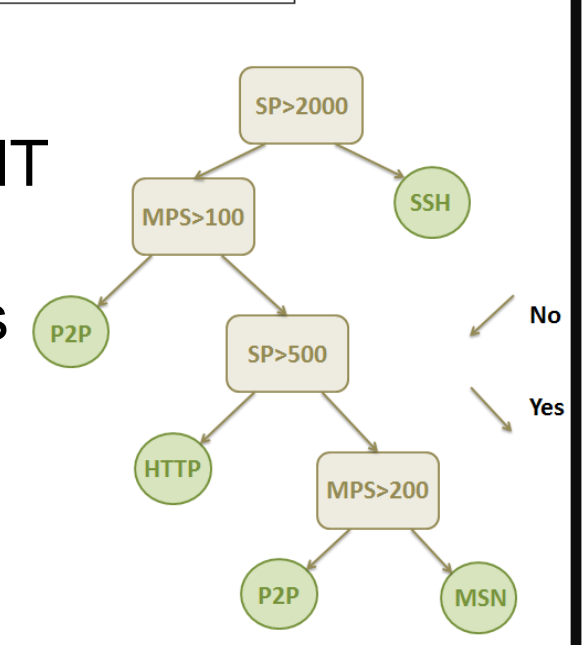
- Statistics-based approach
- Built Based on information entropy
- Examine features of packet header

TABLE I: Candidate Features

Source port number	Destination port number
Average packet size	Variance of packet size
Maximum packet size	Minimum packet size
Maximum inter-arrival time	Minimum inter-arrival time
# of Bytes in forward direction	# of Bytes in back direction

Highly unbalanced

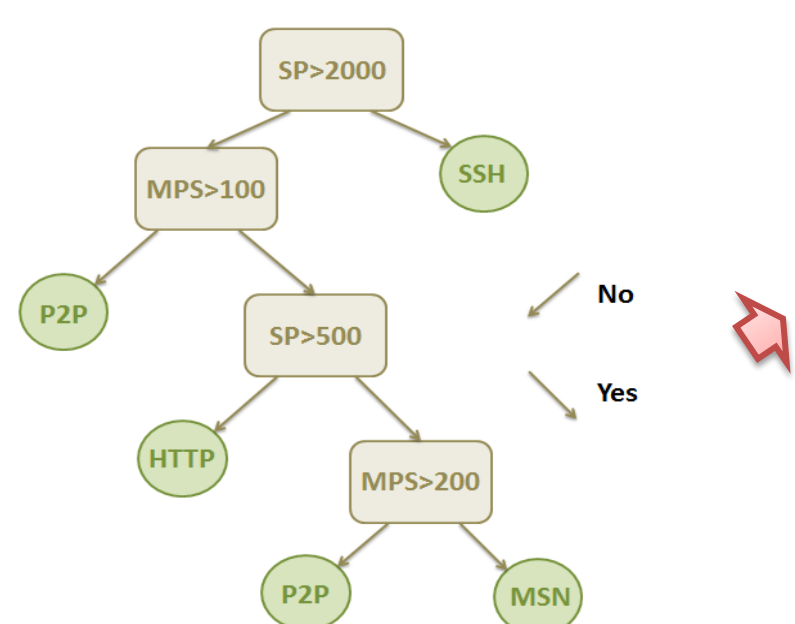
- Unappropriate for SIMT execution model
- Performance depends on traffic pattern



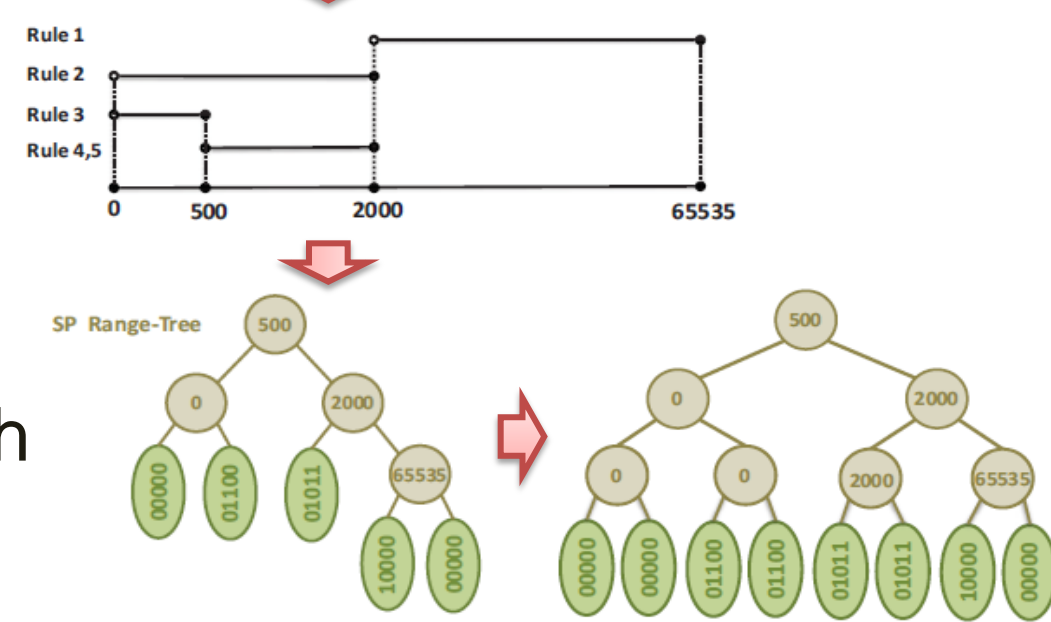
Algorithms

Convert C4.5 decision tree to rule set table

- Each leaf node → one rule



Rule ID	SP	MPS	App. Class
1	(2000, 65535]	(0, 65535]	SSH
2	(0, 2000]	(0, 100]	P2P
3	(0, 500]	(100, 65535]	HTTP
4	(500, 2000]	(0, 200]	P2P
5	(500, 2000]	(200, 65535]	MSN



Build a perfect range-tree for each column (feature)

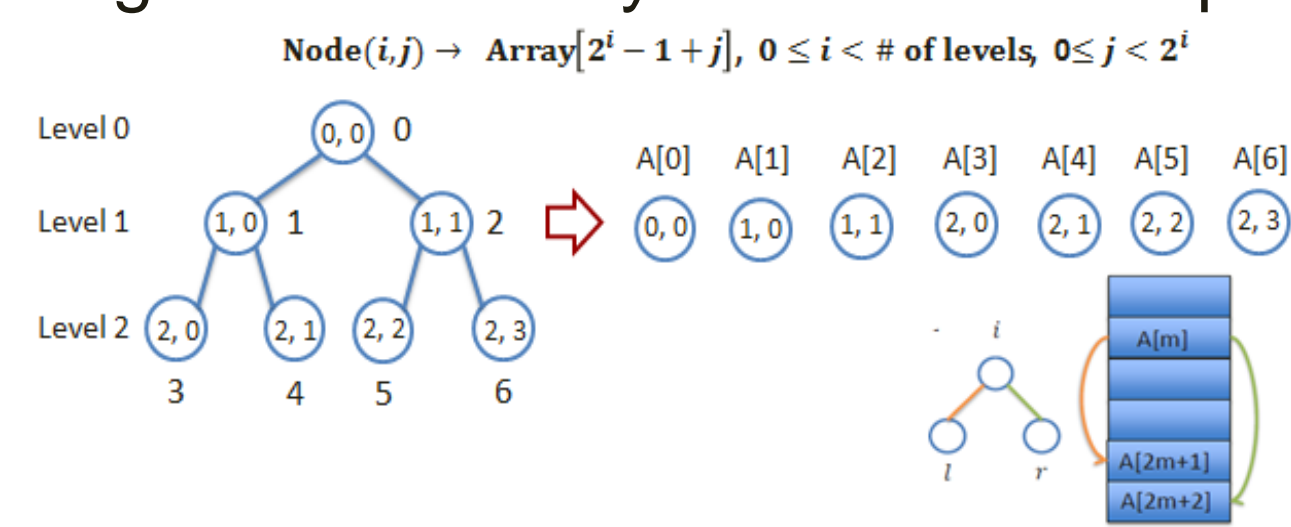
- All leaves have the same depth
- Each leaf contains a bit-vector
- One bit for one rule
- '1' in the i^{th} position: the i^{th} rule contains this range

Decomposition-based classification

- Search
 - Range-tree search for each field
 - Each field obtains a bit-vector
- Merge
 - Bitwise-AND all bit-vectors
 - Only one '1' in the result bit vector

Optimizations

- Use shared-memory to store range-trees
- Store range-trees in array fashion without pointers



Performance Evaluation

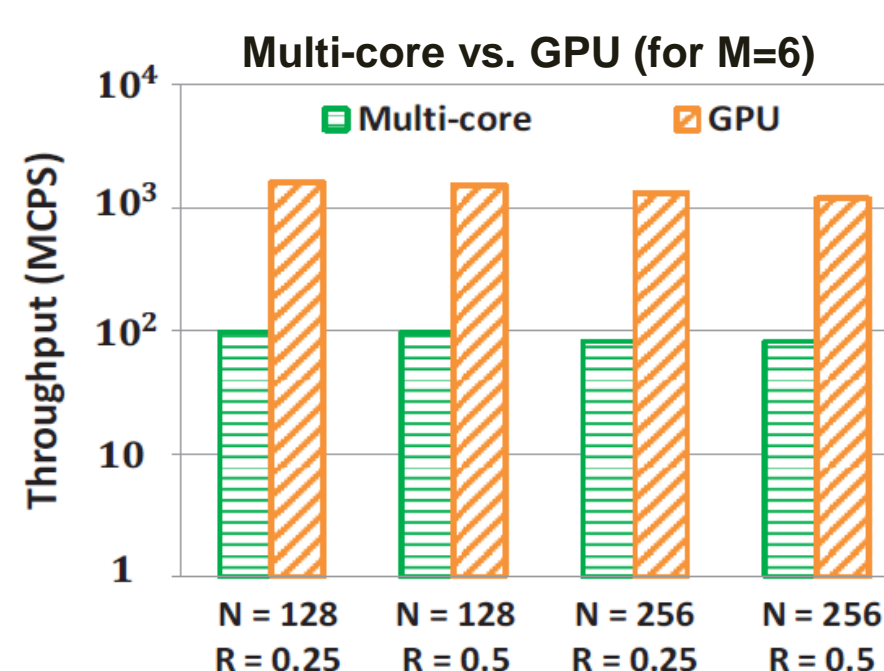
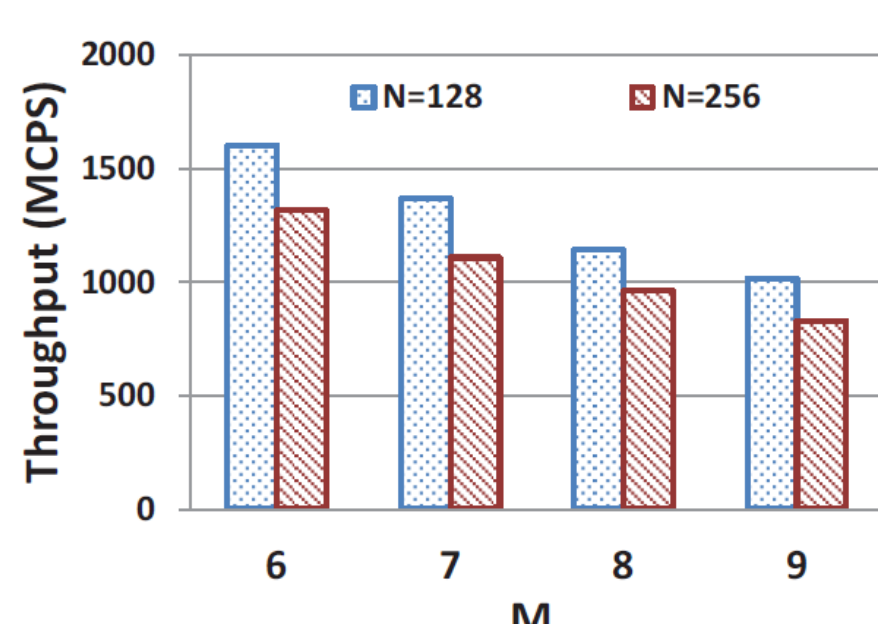
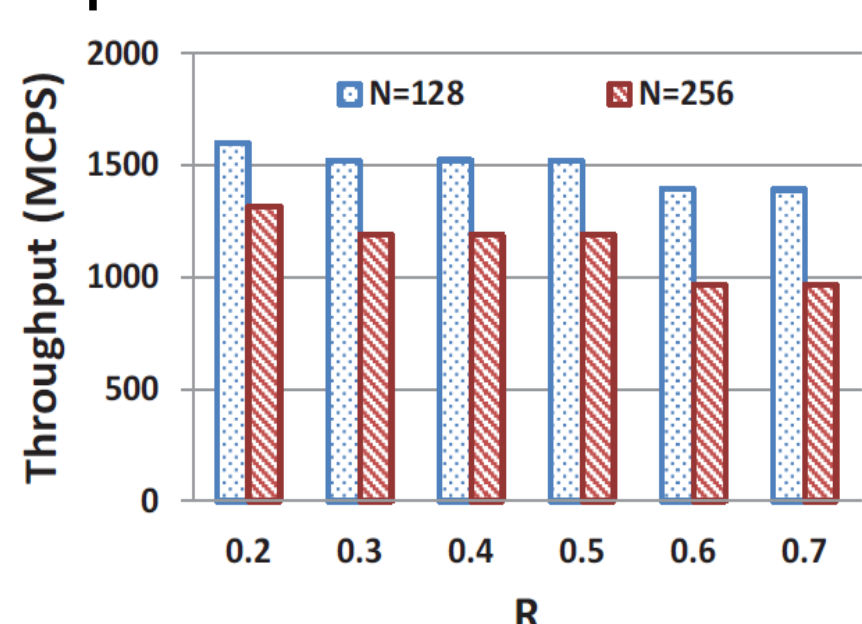
Platform

- Dual 8-core Intel E5-2665; 2.4 GHz
- NVIDIA K20 Kepler
 - 2496 CUDA cores
 - 705.5 MHz

Notations

- N : the number of rules
- M : the number of features
- R : the percentage of the unique values in each feature

Experimental Results



Conclusion

- Convert unbalanced C4.5 decision tree into perfect binary range-trees
 - Decomposition-based classification
 - 1.88x higher throughput compared with C4.5 decision tree
- Store range-trees using compact arrays without explicit pointers in shared memory
- Achieve a high throughput of 1.6 billion classifications per second (GCPS)
 - 128 leaf nodes; 6 features
 - 16x improvement compared with state-of-the-art multi-core implementation