

# High-Performance Traffic Classification on GPU

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## Introduction & Background

### Traffic classification

- Network management

- Flow prioritization
- Traffic shaping
- Traffic policing

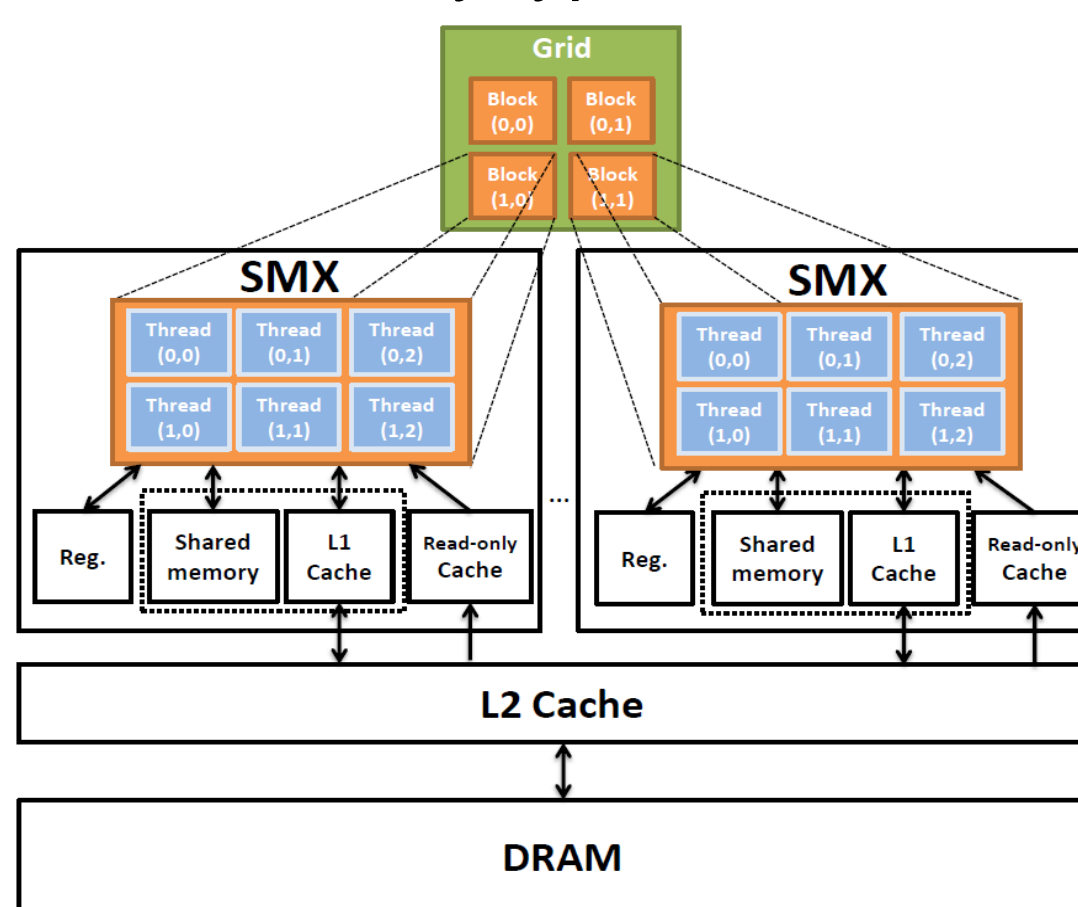
- Approaches

- 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

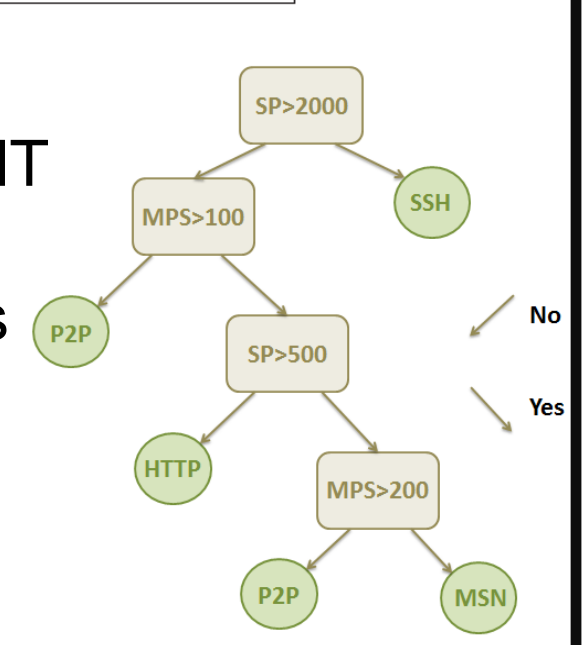
- Statistics-based approach
- Built Based on information entropy
- Examine features of packet header

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

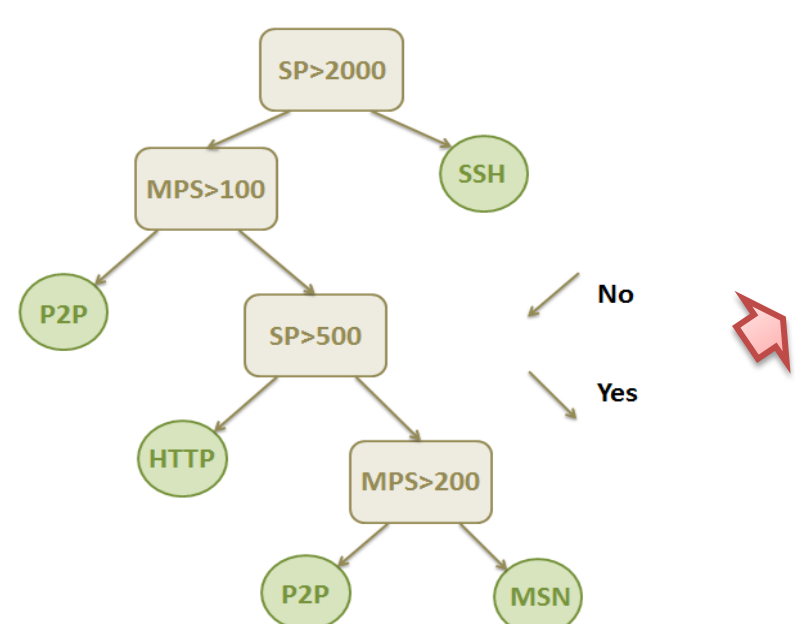
- Unappropriate for SIMT execution model
- Performance depends on traffic pattern



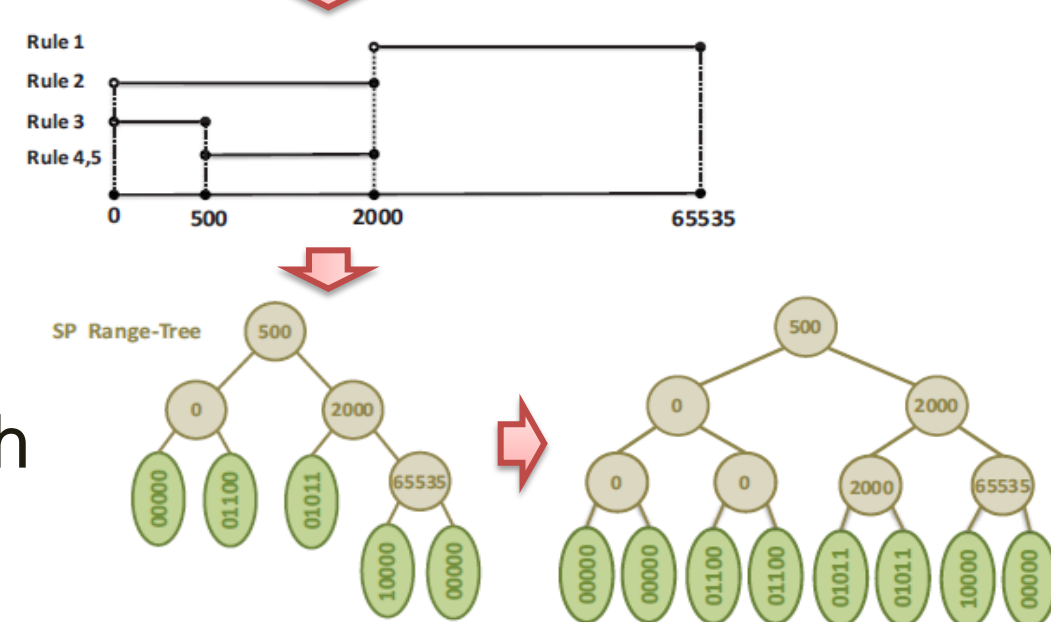
## Algorithms

### 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

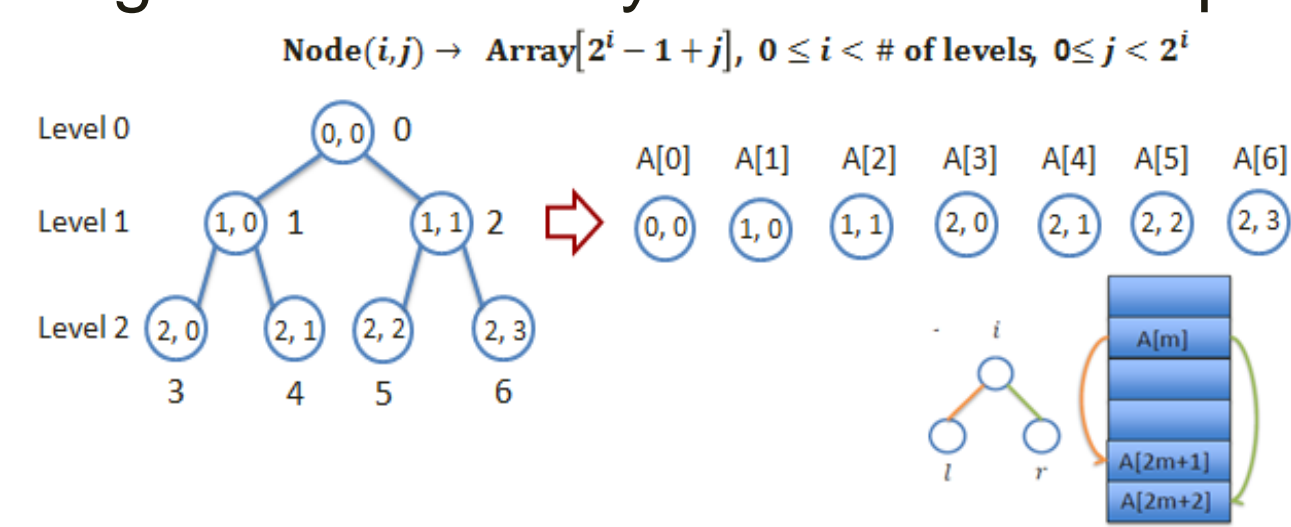
- One bit for one rule
- '1' in the  $i^{th}$  position: the  $i^{th}$  rule contains this range

### 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

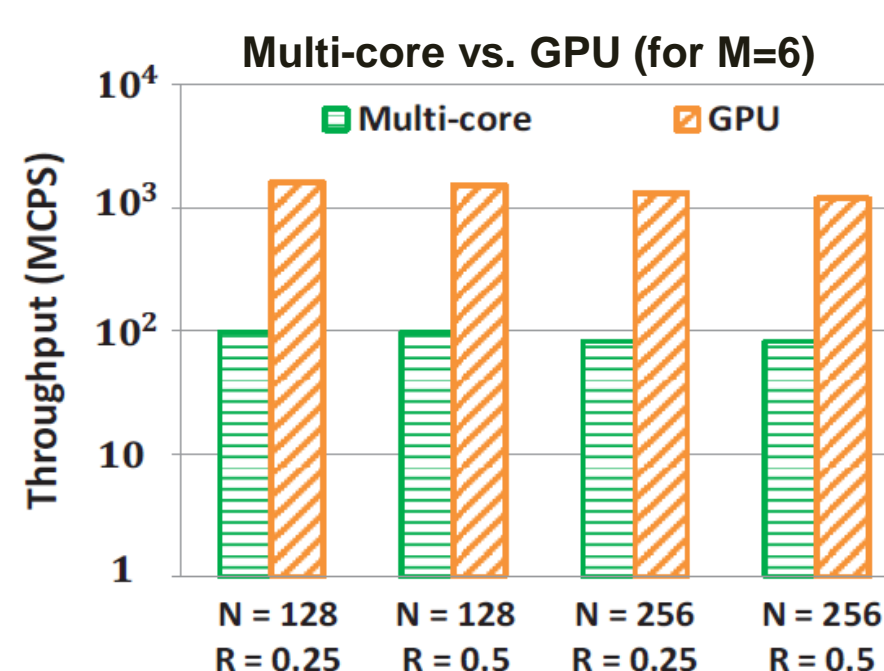
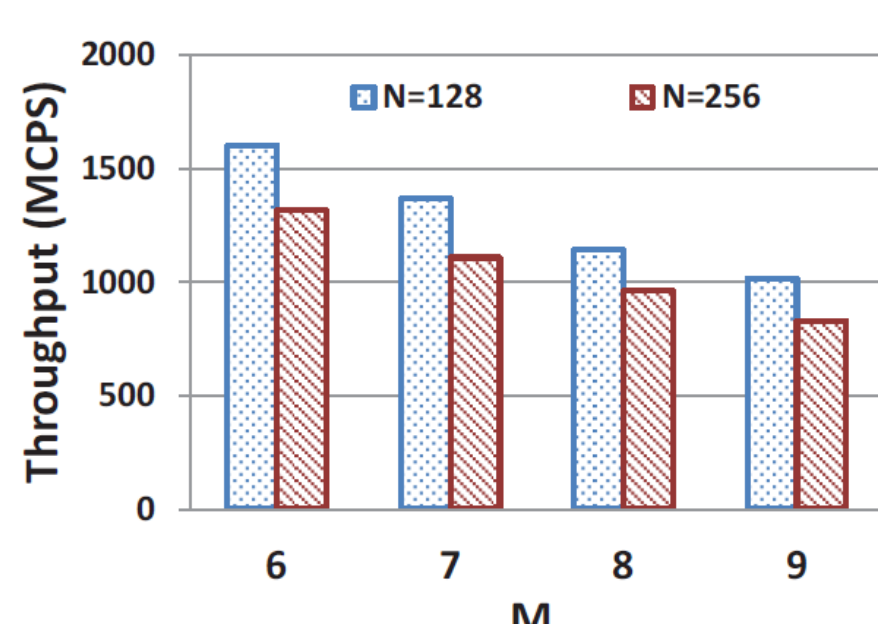
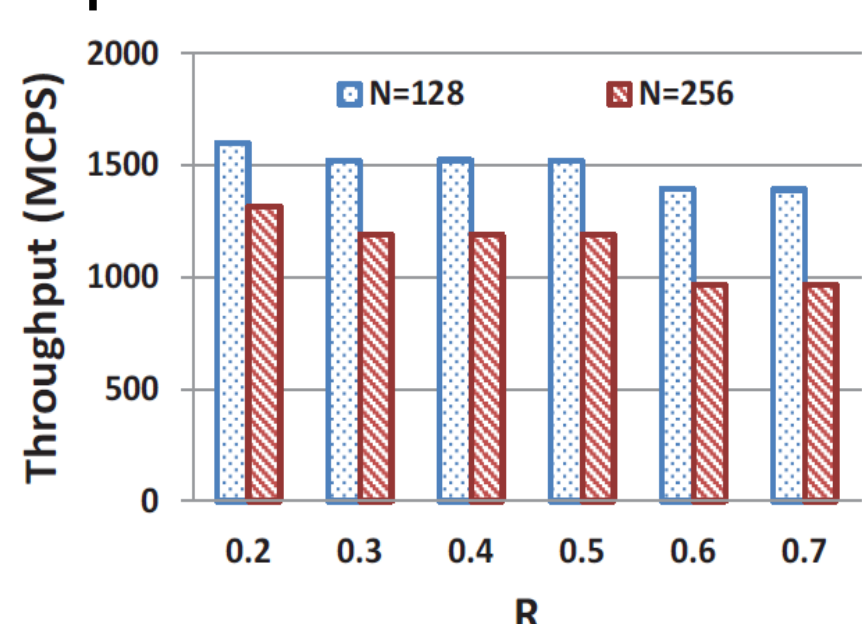
### Platform

- 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