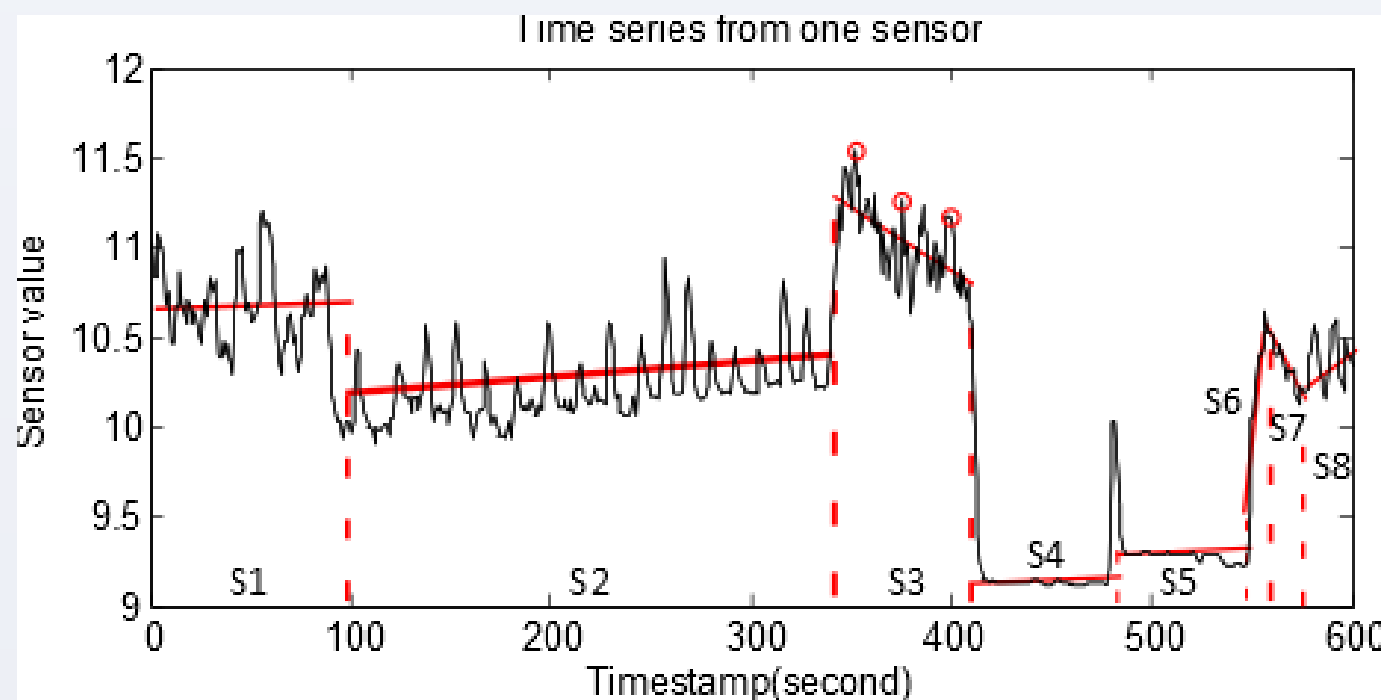


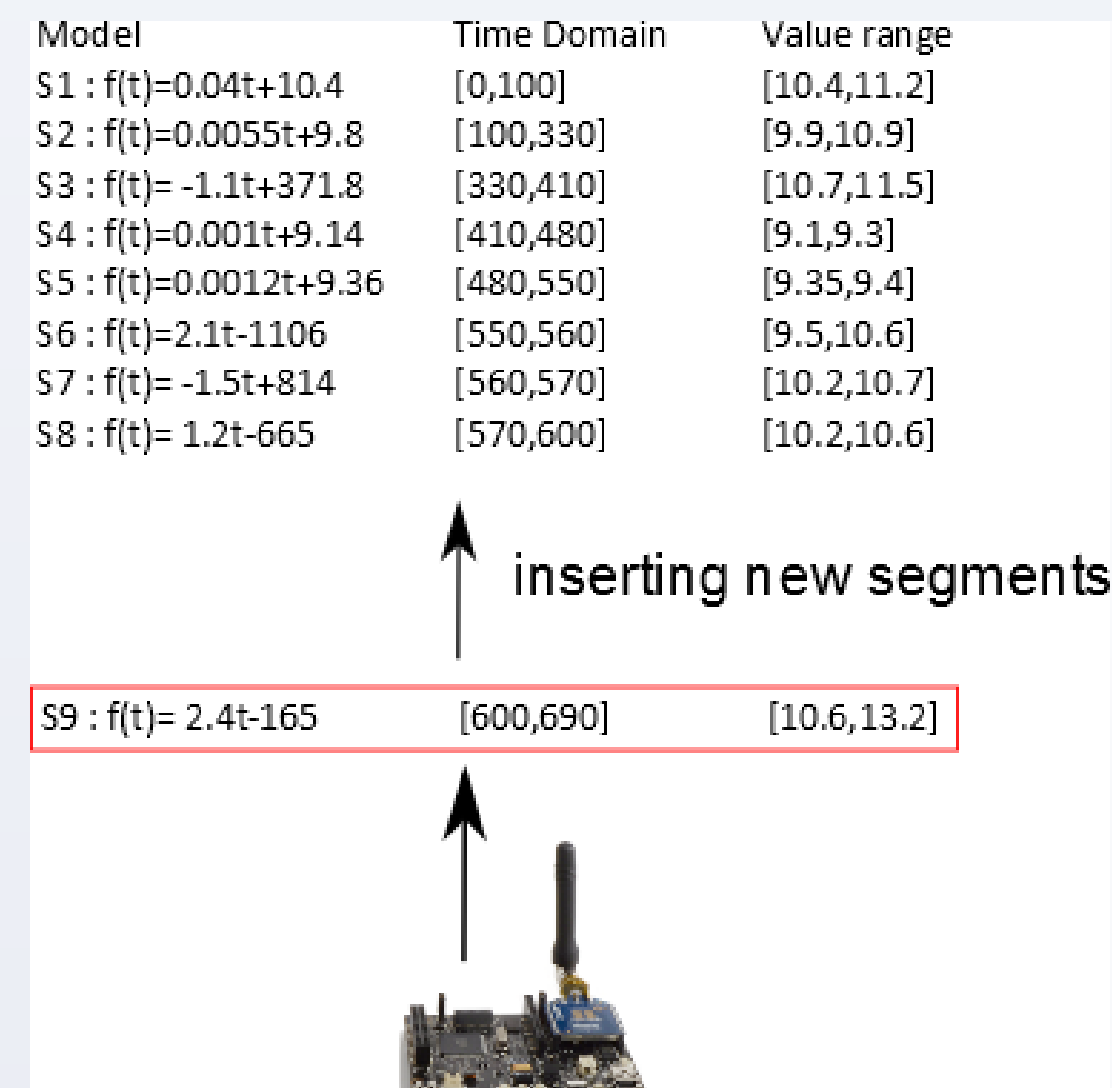
# Online indexing and distributed querying model-view sensor data in the cloud

## Model-view sensor data in the cloud

### Sensor time series segmentation



### Each segment approximated by one model



### Queries of interest:

Time(or value) range queries: SELECT sensor values (or time ranges)  
FROM sensor 1 WHEN  $t_1 \leq \text{time} \leq t_2$  (or  $v_1 \leq \text{value} \leq v_2$ )

Time(or value) point queries: SELECT sensor values (or time ranges)  
FROM sensor 1 WHEN  $\text{time} = t$  (or  $\text{value} = v$ )

Composite queries: SELECT sensor values (or time ranges)  
FROM sensor 1 WHEN  $t_1 \leq \text{time} \leq t_2$  AND  $v_1 \leq \text{value} \leq v_2$

## Challenges for managing model-view sensor data in the cloud

Store model-view sensor data in the distributed cloud store

### Challenges:

#### Storage scheme

- Model-view sensor data in key-value stores.

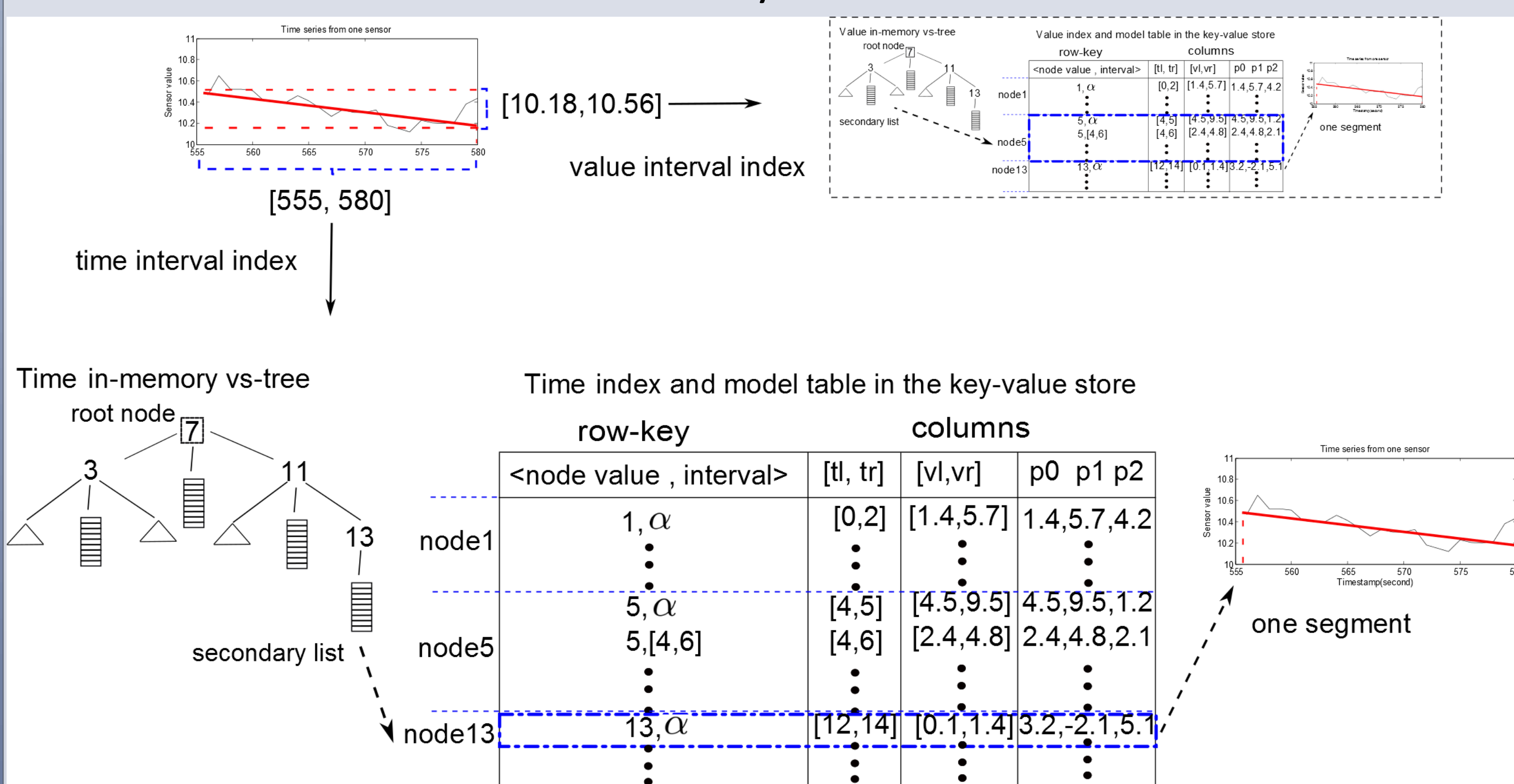
#### Model index

- Update the segments of model-view sensor data on the fly
- Support various queries over model-view sensor data via distributed computing

### Key-Value model index

### Index structure:

- Index time and value dimension respectively.
- two-tire structure of time (or value) index:  
in-memory virtual search tree (vs-tree)  
index and model table in the key-value store



### Registration node of one segment:

- It is the highest one that overlaps the interval to index on vs-tree.
- It is the node of which secondary structure one segment is materialized into.

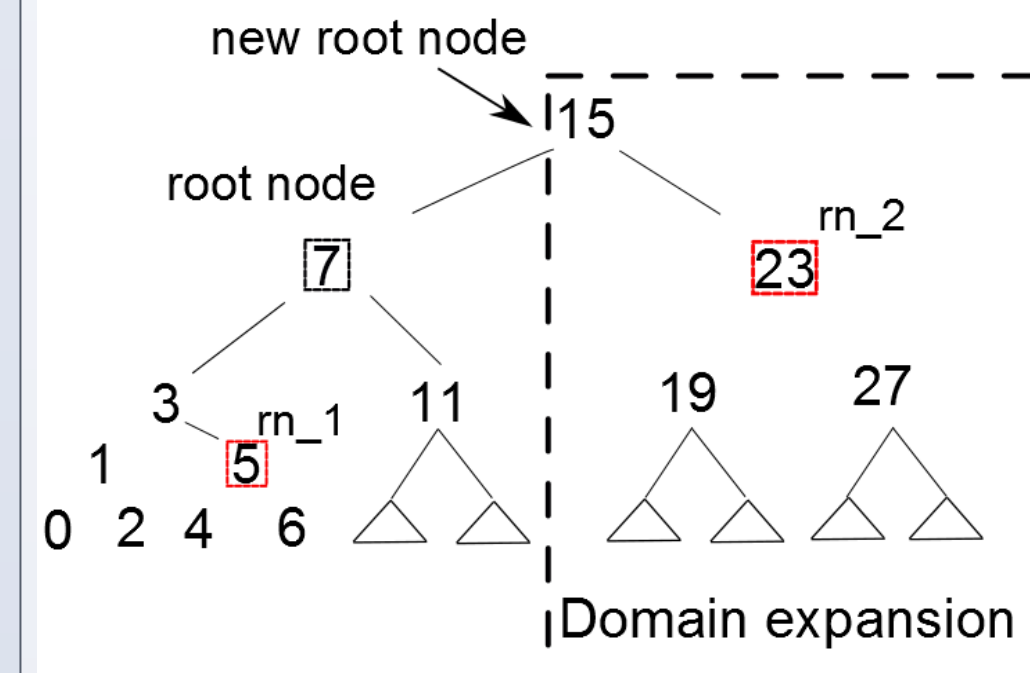
### Index update:

- in-memory search on vs-tree.
- segment materialization in the key-value store.

### Registration node searching on vs-tree

E.g.: insert time intervals of two new models

model 1: [4, 6]  
model 2: [20, 25]



### Segment materialization

#### row-key construction

model 1: 5,[4,6]

model 2: 23,α

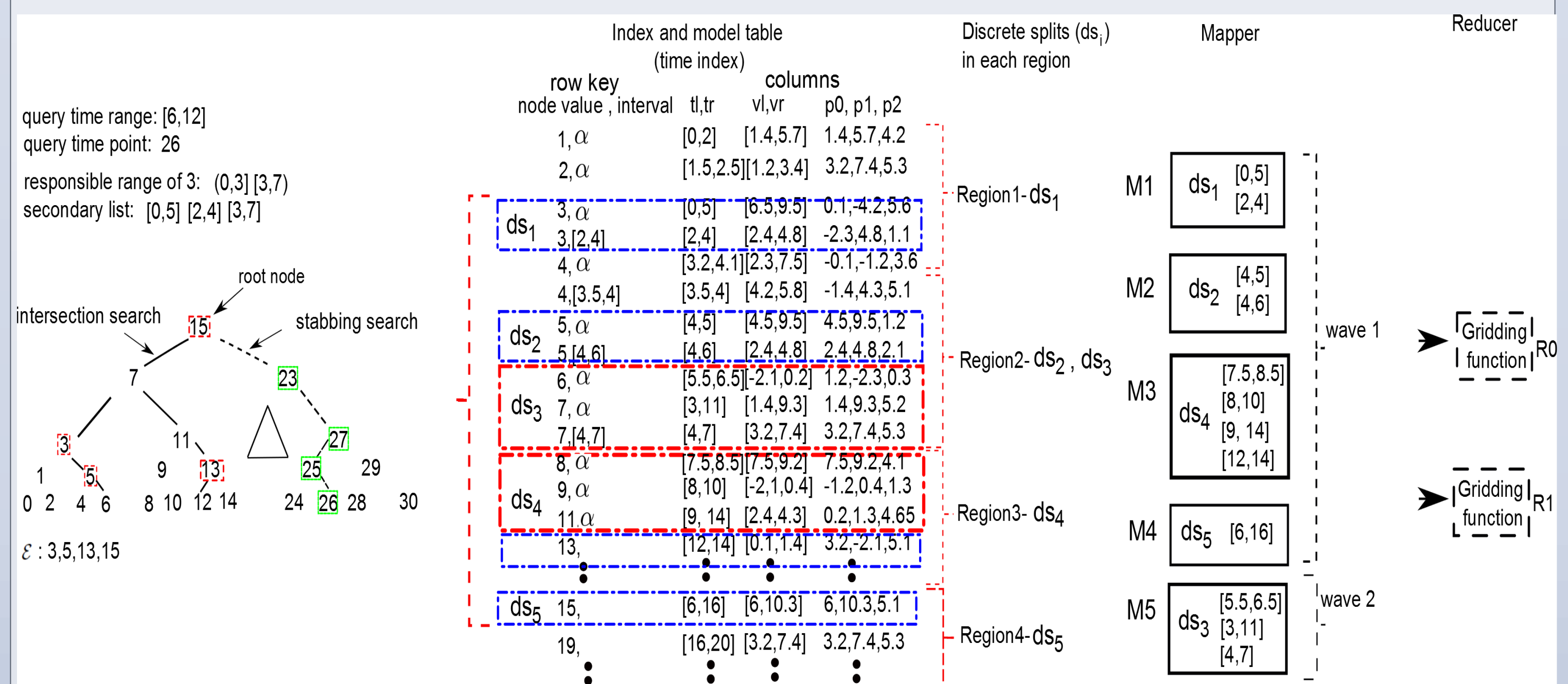
#### index and model table

row key		columns		
node	value, interval	tl,tr	vl,vr	p0, p1, p2
	1, $\alpha$	[0,2]	[1.4,5.7]	1.4,5.7,4.2
	5, $\alpha$	[4,5]	[4.5,9.5]	4.5,9.5,1.2
→	5,[4,6]	[4,6]	[2.4,4.8]	2.4,4.8,2.1
	7, $\alpha$	[3,11]	[1.4,9.3]	1.4,9.3,5.2
	7,[4, 10]	[4, 10]	[3.2,7.4]	3.2,7.4,5.3
	11, $\alpha$	[9, 14]	[2.4,4.3]	0.2,1.3,4.65
	15, $\alpha$	[6,16]	[6,10.3]	6,10.3,5.1
→	23, $\alpha$	[20,25]	[7.5,9.2]	7.5,9.2,4.1
	⋮	⋮	⋮	
	⋮	⋮	⋮	

## Query processing

### Point and Range queries:

- search on vs-tree to find relevant nodes.
- modify the data access method of MapReduce to enable to process the discrete parts of the index and model table.



Composite queries: adaptive index selection decides which index-model table to be processed by MapReduce.

Based on the splits  $n_s$  given the maximum mappers slots  $M$  waves is defined as

$$W = \frac{n_s}{M}$$

data transferring cost is defined as

$$D_t = \sum_{i=1}^{n_r} g([W] * m_i - |S_i|) \quad g(x) = \begin{cases} 0, & x \geq 0 \\ |x|, & \text{otherwise} \end{cases}$$

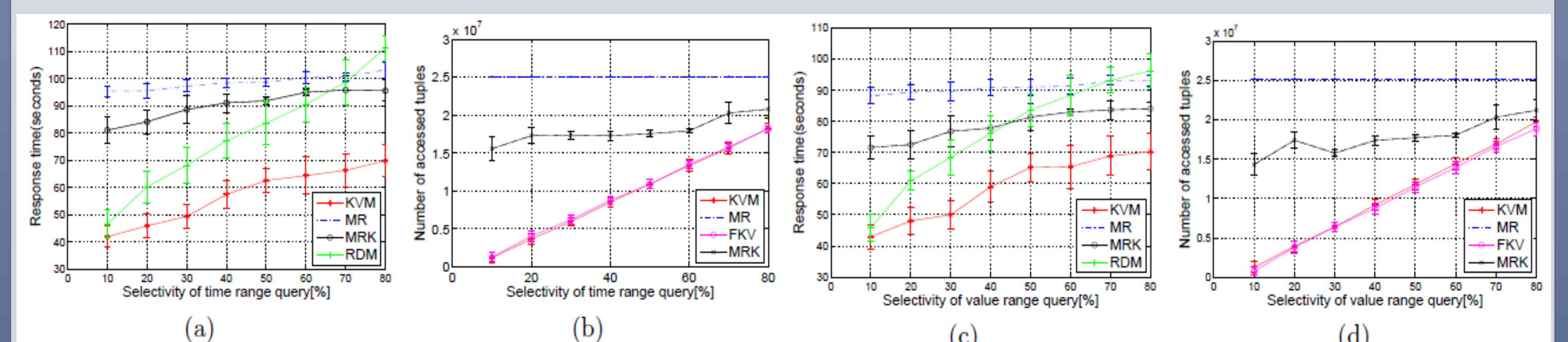
the number of discrete splits  $|S_i|$  in each region  
 $n_r$  is the number of regions of the index-model table.  
mapper slots  $m_i$  of each node

the combined cost of one composite query for time (or value) index is defined as

$$C = \alpha * W + (1 - \alpha)D_t, (0 \leq \alpha \leq 1)$$

## Experimental evaluation

### Comparison of range query performance:



## Acknowledgements

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