Online trajectory analysis with scalable event recognition

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Online trajectory analysis with scalable event recognition

- Monitoring of moving objects:
 - Detect critical situations to avoid accidents and ensure regulations.
 - Detect them as fast as possible to provide margins for action.
- Solution:
 - Use Wayeb, a Complex Event Processing tool, to identify such situations.
 - Implement and compare parallel techniques on top of Wayeb to improve scalability.

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vehicle id	78986	78986	78986	78986	78986	
speed	85	93	99	104	111	
timestamp	1	2	3	4	5	

- Events in a stream come in the form of tuples with both numerical and categorical values
- Through the use of patterns on such events we identify complex events.
- e.g. A vehicle exceeding a speed limit of 100 km/h for two consecutive events.

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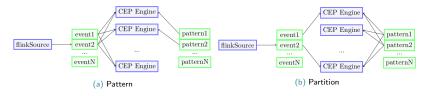
Automata-based Event Recognition

start
$$\rightarrow \bigcirc^{\frown} speed > 100 \bigcirc 1 \xrightarrow{speed > 100} 2$$

- Wayeb transforms these patterns to symbolic automata as its computational model.
- Transitions use boolean predicates that operate on the attributes of the events.

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Event Recognition with Flink 1/2



(a) Pattern-based parallelization:

- Patterns distributed in round robin format to engines
- Events are broadcast to all engines
- (b) Partition-based parallelization:
 - Each engine receives the full set of patterns.
 - Each event goes to one engine.
 - A partitioner decides in which engine each event is sent to.

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Event Recognition with Flink 2/2



(c) Partition Based parallelization with parallel sources

- Flink gives us the option to use parallel input streams.
 - Used to emulate streams of greater event rate.
- (c) Special case of Partition-Based parallelization when one-to-one relation between sources and CEP engines exists.
 - Events are forwarded to the corresponding engine.
 - Each engine receives the full set of patterns.

- Fleet Management: 270M events. Covers a period of 5 months, from June 30,2018 11:00:00 PM to November 30, 2018 11:59:59 PM
- Maritime Monitoring: 18M events from 5K vessels sailing in the Atlantic Ocean around the port of Brest, France, between October 1st 2015 and 31st March 2016 (6 months).

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Experts have provided us with patterns:

- Route:
 - Basic element of vehicle management.
 - Defined as the time period between periods of the vechicle being parked.
- Identify erroneous patterns in the data that imply:
 - hardware malfunctions,
 - bad connection during the device installation
 - lack of satellites tracking the vehicle
- Extra: Refuel opportunity and dangerous driving

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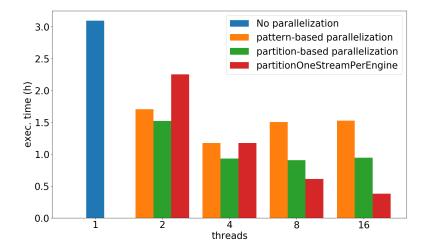
Maritime Patterns

- Patterns deemed of special significance in previous works by domain experts:
 - High speed near coast
 - Anchored
 - Drifting
 - Search and rescue
 - Trawling
 - Loitering.
- Vessels approaching a port:
 - Use the same pattern with different port eachtime
 - Emulate loads of higher magnitude.

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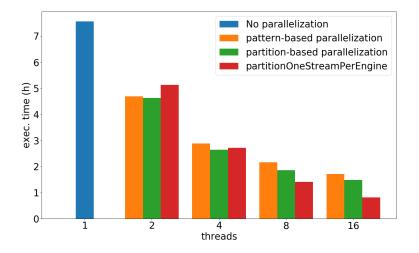
Fleet Management Execution Times, 16 patterns



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Fleet Management Execution Times, 48 patterns



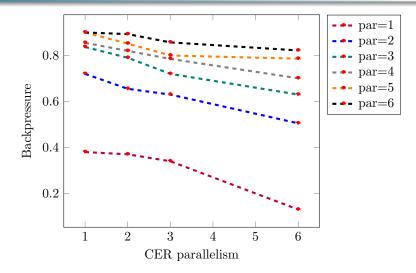
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- BackPressure: Flink metric that notifies you if your operators can consume the data from the stream source as fast as they arrive.
 - In our case, the operator is the distributed version of Wayeb.
- Measured as a percentage:
 - 0% pressure means the operators consume the events as fast as they arrive.
 - I.e., their throughput is at least equal to the incoming event rate.

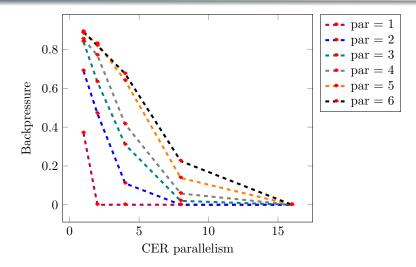
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Pattern parallelization - (6 Patterns)



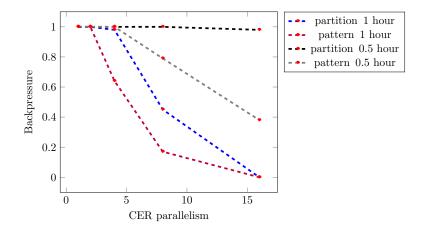
par: parallelism of the source. Black line (6 threads) is 490K events/second. Pink line (1 thread) is 110K events/second

Partition parallelization - (6 Patterns)



par: parallelism of the source. Black line (6 threads) is 490K events/second. Pink line (1 thread) is 110K events/second

BackPressure - Maritime (200 Patterns)



 Simulated the whole stream to run in half and in one hour due to larger load.

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Summary

Conclusion:

- effectively process streams of event rate at least 490K events/second
- superiority of partition-based over pattern-based parallelization, when patterns are few.
- pattern-based parallelization is viable for many patterns.

Future Work:

- Combine various distribution techniques
- Construct more patterns for the domains presented.
- Compare our automata-based method against other approaches.

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