

# EVOLVE

Applications

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www.evolve-h2020.eu

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# **Seven EVOLVE pilots**

### **Automotive Industry**

### **Space Hellas**

 Maritime Surveillance using observation data, historic metadata and classification models

### **Thales Alenia**

 Change detection on satellite images

### Cybeletech

Optimising agro production yield using numerical models and massive historic data

### **Automotive Industry**

AVL

Predictive Maintenance
Application

### Koola

 Data driven vehicle engineering processes support Transport

### **Tiemme/MEMEX**

• Public Transport planning and operations

### BMW

 Demand Mobility Services

# **Applications Stack Profiles**

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Pilot Application	Technology Stack	
Maritime Surveillance	Spark, Tensorflow, GPUs, external visualisation and Weblyzard	
Change detection on satellite		Goals
imaging	Kafka, Tensorflow, GPUs, DASK	
Optimising agri-production yield	Tensorflow, GPUs, MPI, Memoscale	- Performance
Improvement of bus		
transportation service	Kafka, Spark, external visualisation and Weblyzard	- Flexibility to
Advanced mobility services		adapt
optimization	Kafka, Tensorflow, GPUs	
Predicitive Vehicle		
Maintenance	Tensorflow, GPUs, R	
Data-assisted automotive		
service development	Spark, R, plotpy and Weblyzard	

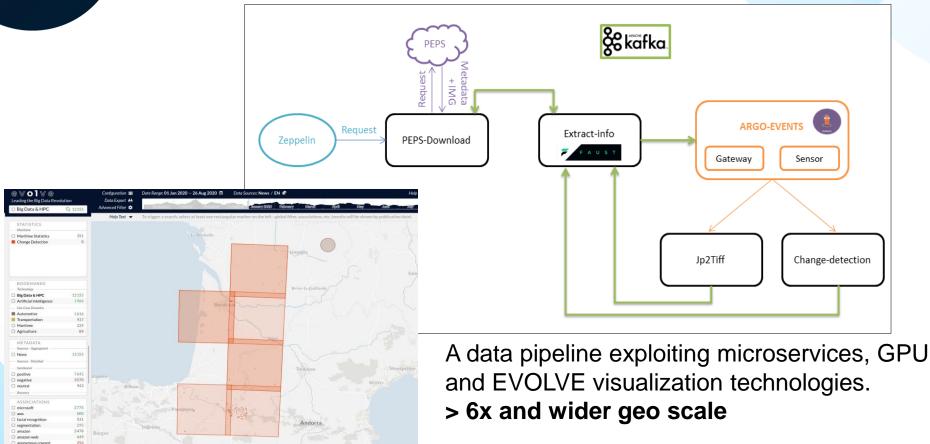


# **Change Detection**

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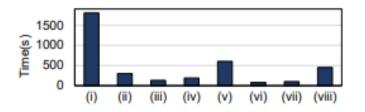
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# **Change Detection**



#### 8 different versions of this application:

i) CPU-only,

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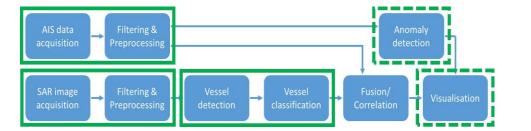
- ii) iCPU-only using TensorFlow(TF),
- iii) iGPU and TF,
- iv) DASK,
- v) DASK and GPU,
- vi) KAFKA and DASK in a small dataset,
- vii) KAFKA and GPU in small dataset,
- viii) KAFKA and DASK in a big dataset.

KPI Description	EVOLVE Impact
	(speedup)
Average time to generate one change detection map from a couple of Sentinel-2 images with the tile generation step running on CPU and DASK is used for parallel computing	x10 faster
Average time to generate one change detection map from a couple Sentinel-2 images with tile generation running on GPU	x15 faster
Average time to generate one change detection map from 200 Sentinel-2 images	x120 faster for the CPU+Dask version

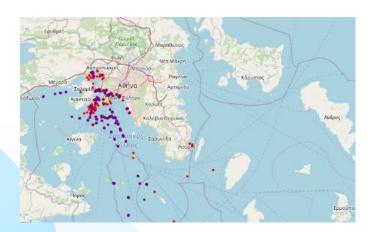
When ?	Processing time	Number of results
Before Evolve	> 20min	1 CD map
Dask	3min15s	1 CD map
Dask + Kafka	6min30s	5 CD maps



## **Maritime Surveillance**



IME, microservices and stream processing, GPU, Zeppelin visualization brought substantial performance benefits



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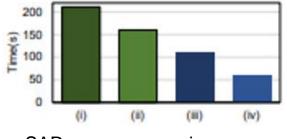
Data Volume		
Data processed per batch		
Vessel Detection	1 SAR Image ~ 1Gb	
Anomaly Detection	8 records for testing, 4096 for training	
Time for batch processing		
Vessel Detection	~60 secs/image	
Anomaly Detection	~0.015 secs/batch with GPU	
Batches Required		
Vessel Detection	depends on user criteria	
Anomaly Detection	895 batches/epoch, 500 epochs	
Comparison with previous situations		
Vessel Detection	~210 secs (previous) VS ~60 secs (now)	
Anomaly Detection	>8 hours (previous) VS <2 hours (now) – training phase	



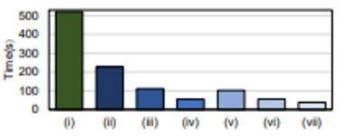


# **Maritime Surveillance**

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SAR scene processing



Anomaly Detection based on AIS data

KPI Description	EVOLVE
	impact
Vessel detection time	71% reduction
AIS anomaly detection training time	89% reduction
Detection Accuracy	5% increase
Operational spatio-temporal window increase	10x



#### <<sup>™</sup> **Predictive Maintenance** DATA Train Training Data Batches Model or API Score Evaluate Model Model

Learning Algorithm **KPI** Description **EVOLVE impact** Preparation, Training and Serving Data 2x – 3x speedup (without GPUs) models time More expected via GPUs (in progress M36)

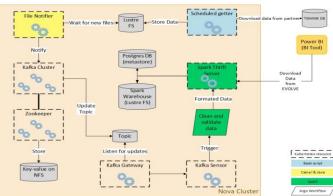
New Data

Different set of Machine Learning libraries are assessed with different and extensive datasets. Models deployed on EVOLVE are showing significant speedups (from 2x to 3x) in four different algorithms (logistic regression, linear discriminant analysis, classification and regressions trees, support vector machines)

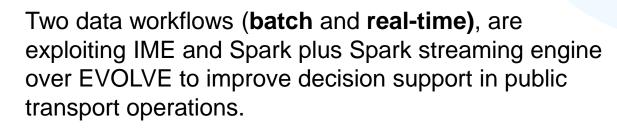
More to expect with tensorflow and GPU



# **Public Transport**



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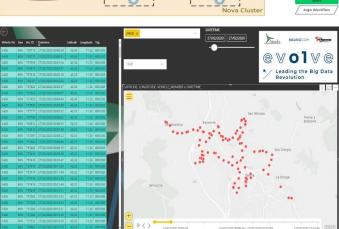


Finally exporting data to external Business Intelligence tools and WebLyzard visualisation

Past analysis made on 1-month data  $\rightarrow$  with EVOLVE can be made on yearly data

Past analysis only on historical data  $\rightarrow$  with EVOLVE real-time ability was easily deployed





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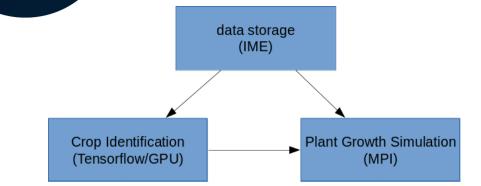
# **Public Transport**

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KPI Description	Before Evolve	After Evolve
Time period covered by the analysis (for the offline workflow context), which can be considered by the management authority as possible to analyse	nununu aara involvina	1 year and related all the 4 areas served by Tiemme fleet
Listing of trips operated in a period of interest along with the extracted relevant bus events (for the offline context) Indicative KPI : The time required to extract 86370 records (corresponds to 2 days service) from SQL	20 min and 35 seconds	>20x speedup Less than one minute
Daily query: summary of transits time analysis carried out within a period of interest: 30-60-90 days of service (for the offline context)	For 30 days data: 2 min 15 seconds For 90 days data: 29 minutes and 49 seconds	For 30 days data 11,21 seconds For 60 days data 13 seconds For 90 days data 16 secods From 10x to 120x speedup
Necessary time to visualize the real time status of 2 or 3 selected lines/routes (for "Real Time" context) and related information, on map	Not supported	Under 6 seconds to analyse 2 or 3 lines.
Arcs number which can be processed and visualized simultaneously with related updated information (for "Real Time" context) on the map	Possibility to analyze only a few arcs in an acceptable interval	Possibility to process 21K arcs in under 20 seconds

# **Agri production optimisation**

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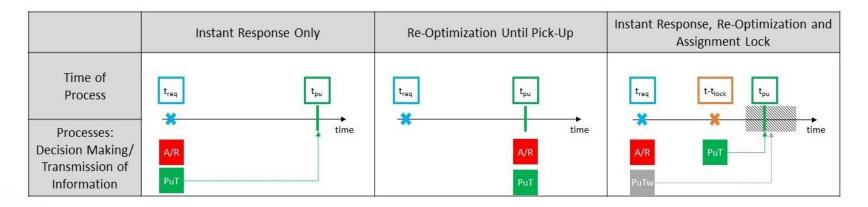
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Integration of MPI within EVOLVE

Use of IME, GPUs and stream analytics engine by tensorflow on 20 Gb datasets. Memoscale compression stage at data transfer phase.

KPI Description	EVOLVE impact
Crop Detection Rate - The success rate of crop detection is improved with larger learning datasets	90% → 97%
Learning dataset size	x4 Geographcal area coverage enlarged
Plant Growth Simulation – Computation time reduction	:4
Number of simulations	x2
Overall process execution time	:5, More robust, stable and easy to deploy

Exploiting EVOLVE HPC aims to accelerate three different on-demand mobility services (allocating cars to requests)



Evident speedups based on New York demand datasets.

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Three simulations are end-to-end possible on EVOLVE, assessed over different distributed parallel schemes of Branch-and-Bound algorithm using MPI.





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Remarkable Performance Improvement: The use cases are using technology components, such as parallelization frameworks, stream analytic engines, fast storage, and hardware accelerators that automate, optimize, and boost the performance of applications.

### Business-level achievements

- Increase of quality and quantity business indicators
- Development efficiency  $\rightarrow$  lower efforts and costs, quicker time-to-market:



# The Consortium

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# **The Use Case Providers**

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