

T. Vafeiadis*, A. Nizamis*, V. Pavlopoulos*, L. Giugliano†, V. Rousopoulou*, D.Ioannidis* and D. Tzovaras*

*Centre for Research and Technology Hellas-Information Technologies Institute (CERTH/ITI) † Links Foundation (LINKS)





International Workshop on IoT Applications and Industry 4.0



Introduction - Problem Definition

Smart waste management is a very important procedure in Industry 4.0

- Profitable asset for:
 - Waste producers
 - Waste management providers
- Connection with **IoT sensors** on industrial premises
- **Big data** availability
- Significant advances on data analytics

Enable the waste management companies:

- To smarten their domain
- To automate many of their solutions and processes





Introduction- Our Contribution

The main goal is to contribute to smart waste management optimization by providing:

- Sensor-based bins and dedicated waste management operations that will forcefully replace obsolete methods
- Transparency of waste spend and access to historical data towards the evaluation and improvement of waste management through analytics
- Variety of analytic services to end users related to waste management activities in order to enhance the decision-making and optimize planning





Overview



Optimization tool for waste management companies

State-of-the-art algorithms and methodologies for data analysis

Advanced data visualization



IoT devices connectivity and data analysis

Secure data exchange based on authentication mechanisms





Services

- ✓ Monitoring of bins fill level based on IoT sensors
- ✓ Analysis of the bins fill level trend
- Forecasting about the tonnage of wastes that is going
 to be transported by a waste management company
- Calculator for optimal pair of routes and tonnage should be transported
- ✓ Price forecasting for various waste types/materials
- ✓ Statistical analysis and visualization for better data exploration





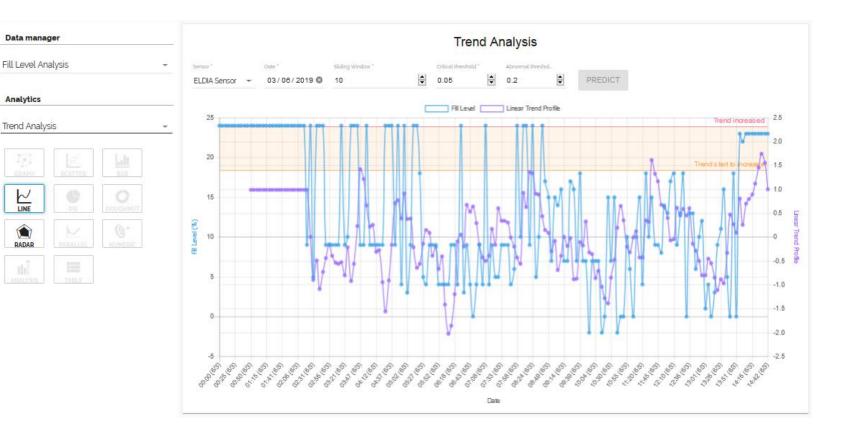
- Use of **Ultrasonic** and **IR** sensors for fill level measurement
- Use of LoRa network in order to cover low power needs and get data from sensors
- Measure the fill level of both indoor and outdoor industrial bins containing scrap metal and recycling materials
- Provide distant fill percentage monitoring for efficient logistics, between industry and waste collection companies
- Notification mechanisms (email) for fill level over 80%







- Real-time analysis of fill level sensor data
- Trend Analysis applied in order to create a profile for fill level trend
- Slope Statistic Profile method is applied on the time series of recordings (percentages) of a fill level sensor
- End user is able to select:
 - Sensor
 - Date
 - Sliding window for the analysis
 - Thresholds for the analysis
 - Type of visualization (line or radar)
- By using this analysis the waste management company is able to define which bin has the most aggressive trend in order to arrange a pick-up





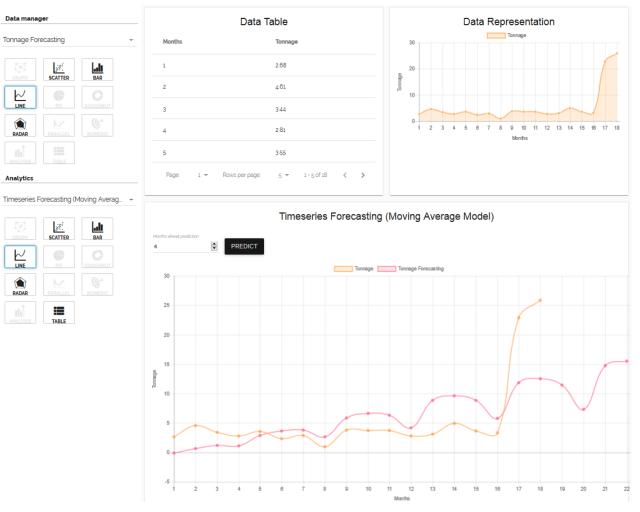
Fill Level Analysis

Trend Analysis



Tonnage Forecasting Time series forecasting

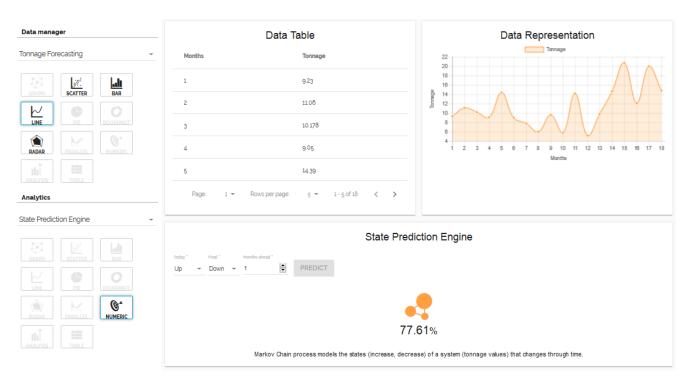
- Forecasting about the tonnage of wastes that is going to be transported by a waste management company
- Time series forecasting using moving average model
- Visualization of predictions
- End user is able to select:
 - Number of months for ahead prediction
 - **Type** of visualization (line, scatter, table, bar and radar)





State Prediction Engine Markov Chain Models

- Predict the probability of future increase or decrease of the transferred tonnage based on current state
- Markov chain models used to determine the probability of moving from one state to another
- End user is able to select:
 - Months ahead period for the prediction
 - **Current** and **future state** of transferred tonnage (Up/Down)







- **Deep Learning** algorithm to provide **prediction** for the **price** per ton at which specific commercial partners is likely to accept to buy/sell waste material
- Prediction model based on historical price data – Recurrent Neural Network (RNN) is used
- Initial network for regression is composed of four hidden Long Short Term Memory (LSTM) layers with 64, 32, 24 and 8 neurons respectively
- Visualization of prediction values and the coefficient of determination expressed as accuracy rate in the same diagram
- End user is able to select:
 - Material for price prediction
 - **Type** of visualization (line, scatter, table, bar and radar)





Price Forecasting Deep Learning



Optimal Routes Calculator

Genetic Algorithm

- Optimal Routes Calculator based on monthly data ٠ about routes and transferred weight of wastes per route
- Calculates 10 best solutions/combinations for the pair routes/weights that can be transferred for a material
- End user is able to: ٠
 - Load monthly data per material
 - Select type of visualization (table, scatter, bar, radar and line)

Data manager	Data Table		Data Representation		
Optimal Routes -	Date Routes	Weight	70 Novies Weight 1400		
GRAPH SCATTER BAR	2015-Jan 30	676.09			
	2015-Feb 45	92142	20 400		
	2015-Mar 62	1189.25	10 10 10 10 10 10 10 10 10 10		
RADAR DARALLEL NUMERIC	2015-Apr 58	1167.88	and		
	2015-May 48	103788			
Analytics	Page: 1 - Rows per page:	5 - 1-5 of 48 < >			
Optimal Routes Calculator 👻					
	Optimal Routes Calculator Top 10 Simulated Solutions Material: Brown paper				
	#	Routes	Weights (tons)		
RADAR PARALLEL NUMERIC	1	30	761		
	2	36	922		
	3	30	760		
	4	20	507		
	5	15	383		
	6	43	1091		
	7	23	588		
	8	33	842		
	9	25	618		
	10	33	837		





Statistical Analysis

- Statistical analysis of the end user's data
 - Price averages, tonnage averages etc.
- Visualization of statistical analysis output

Data manager		Data Table		Data Representation
Statistical Analysis	- Date	Price	Tonnage	Price Tonnage 120
	02/29/2016	300	28.52	80 gr
	08/31/2016	305	99.71	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
	12/31/2016	290	56.82	230 6 10 5 10 10 5 10 5 10 10 10 10 10 10 10 10 10 10 10 10 10
	07/31/2017	290	106.91	్ స్
	08/31/2017	280	13.27	
Analytics	Page: 1 👻	Rows per page: 5	- 1-5 of 13 < >	
Statistical Analysis	•			
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LINE PIE DOUGHNUT	300			0
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	280 B			e0
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	260			Avg: 3778 40
	250			
	240			20
	230 Feb 2016 Apr 2016 Ju	an 2016 Aug 2016 Oct 201	6 Dec 2016 Feb 2017 Apr 2017	





Conclusions

Proposed solution

- IoT data analytics platform for waste management optimization
- To-the-point Data analytics solutions
 - Waste bins fill level monitoring/analysis
 - Forecasting of transported tonnage
 - Price forecasting of waste materials
 - Optimal transportation KPIS (routes/tonnages)

Gains to waste management companies

- Supervised control of waste level
- Access to historical data
- Planning and optimization potential
- Financial and environmental benefits





Thank you

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