

## Summary

The Urban Co-Creation Data Lab (UCD Lab) project aimed to support decision-making at the municipality level to provide citizens with high quality services in the areas of micromobility, waste management, parking, pollution, and emergency. The project aimed developing a new generation of public services in the context of smart cities exploiting supercomputing facilities and public and private data to analyse complex combinations of large datasets in areas of public interest. The analytical model and service presented in this document was developed for the city of Lisbon regarding waste management and was made publicly available to any interested person or institution. The UCD Lab was co-financed by CEF Telecom, the EU instrument to facilitate cross-border interaction between public administrations, businesses and citizens, and the project beneficiaries were: Universidade Nova de Lisboa, Município de Lisboa, Agência para a Modernização Administrativa, I.P., NEC Portugal - Telecomunicações e Sistemas, S.A, and Barcelona Supercomputing Center - Centro Nacional de Supercomputación.

## Service description

This service allows to predict mixed waste production for 95 waste collection circuits for every week of the following month.

## Analytical model

### Input data and variables

In Table 1 are presented the datasets necessary to develop the analytical model for #2 Waste management use case.

*Table 1. Datasets necessary for the development of the analytical model for #2 Waste management use case.*

Dataset	Source	Open data
Undifferentiated waste collection loads	Urban Hygiene and Solid Residuals Department – Lisbon City Council	No
Holidays	NOVA Cidade – Urban Analytics Lab	Yes

The description of the input data necessary to run the analytical model that serve as the basis of the service is presented in Table 2.

*Table 2. Input data description of the waste management model necessary to run the service.*

Variable	Description	Type
Id_circuit	Unique id of the mixed waste collection circuit	INTEGER
Week_date	First day of the week (Monday)	DATE (yyyy-mm-dd HH:mm:ss)
waste_t	Quantity of mixed waste (t) collected in a specific week	FLOAT
holiday	Flag identifying holidays	INTEGER

## Model

The analytical model that supports the service is a Seasonal Auto-regressive Integrated Moving Average (SARIMAX) model (Box et al., 2008). For each mixed waste circuit, a SARIMAX model was trained with data from 01/04/2017 to 31/10/2019 and tested with data from 01/11/2019 to 31/10/2020, using the optimized parameters at waste collection circuit level.

## Output data

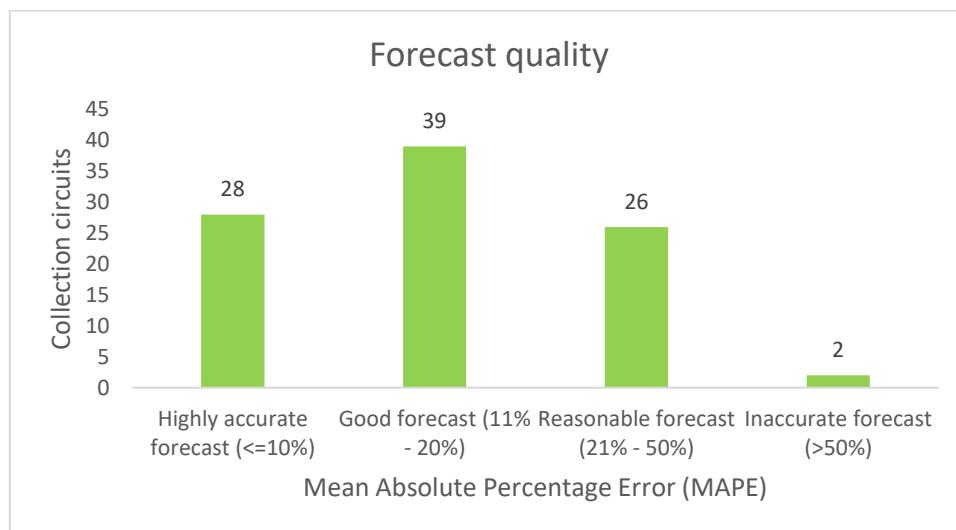
In Table 3 is presented the description of the output data provided by the analytical model.

*Table 3. Output data description of the waste management model.*

Column	Description	Type
Id_circuit	Unique id of the mixed waste collection circuit	INTEGER
Week_date	First day of the week (Monday) of the following month	DATE (yyyy-mm-dd HH:mm:ss)
predicted_waste_t	Quantity of predicted mixed waste (t) produced in a specific week	FLOAT

## Evaluation

The model's quality for each circuit was assessed through the computation of the root mean square error (RMSE), mean absolute error (MAE) and mean absolute percentage error (MAPE) (de Myttenaere et al., 2016). In Figure 1 is presented the percentage of circuits belonging to each forecast quality category derived from Lewis (1982), based on MAPE values. In Table 4 are presented the quality measures computed for each circuit.



*Figure 1. Number of circuits in each class regarding forecast quality based on the scale developed by Lewis (1982).*

Table 4. Root Mean Squared Error (RMSE), Mean Absolute Error (MAE) and Mean Absolute Percentage Error (MAPE) obtained for each mixed waste collection circuit in the testing sample of the SARIMAX model.

Circuit ID	Circuit	order	seasonal order	RMSE (tons)	MAE (tons)	MAPE (%)
7749	l0208MK	(1, 1, 1)	(0, 0, 0, 52)	5.56	5.09	18
7750	l0502MFL	(0, 1, 1)	(0, 0, 0, 52)	3.55	2.35	12
7751	l0503	(2, 0, 0)	(0, 0, 0, 52)	11.65	11.25	41
7752	l0515MK	(0, 1, 1)	(0, 0, 0, 52)	1.87	1.36	6
7753	l0508MK	(0, 1, 1)	(1, 0, 0, 52)	5.40	5.04	31
7754	l0414MF	(1, 1, 1)	(1, 0, 0, 52)	5.83	5.24	9
7755	l0801	(1, 0, 1)	(0, 0, 0, 52)	5.44	4.47	15
7757	l0204A	(2, 1, 2)	(1, 0, 1, 52)	2.25	1.94	6
7764	l0501	(0, 1, 1)	(1, 0, 0, 52)	2.63	2.04	9
7765	l0810	(2, 0, 0)	(0, 0, 1, 52)	5.29	4.33	13
7768	l0804	(0, 0, 3)	(0, 0, 0, 52)	5.18	4.38	13
7769	l0812	(1, 1, 1)	(0, 0, 0, 52)	6.48	5.15	9
7770	l0806	(1, 0, 0)	(0, 0, 1, 52)	7.39	6.39	16
7772	l0402	(2, 1, 1)	(0, 0, 0, 52)	3.28	2.44	6
7773	l0207	(0, 1, 1)	(0, 0, 0, 52)	8.22	7.79	27
7778	l0709	(0, 1, 1)	(1, 0, 0, 52)	6.43	5.67	23
7780	l0809	(0, 0, 0)	(0, 0, 0, 52)	4.00	3.11	11
7781	l0509MK	(2, 0, 0)	(1, 0, 0, 52)	2.41	2.02	11
7782	l0611	(0, 1, 1)	(0, 0, 0, 52)	2.14	1.80	9
7787	l0206MK	(0, 1, 1)	(0, 0, 1, 52)	1.74	1.49	25
7788	l0711	(0, 1, 1)	(0, 0, 0, 52)	8.58	7.87	15
7795	l0803	(1, 0, 0)	(0, 0, 1, 52)	5.60	4.64	15
7796	l0412	(1, 1, 1)	(1, 0, 0, 52)	9.18	5.51	37
7798	l0817MF	(1, 0, 0)	(1, 0, 0, 52)	4.93	4.39	13
7799	l0816	(2, 0, 0)	(0, 0, 0, 52)	7.80	7.33	16
7801	l0802MF	(0, 1, 1)	(0, 0, 1, 52)	5.03	4.02	19
7802	l0807MK	(2, 0, 0)	(0, 0, 0, 52)	2.71	2.44	13
7803	l0808	(0, 0, 0)	(0, 0, 0, 52)	5.21	4.45	16
7804	l0805	(1, 0, 0)	(1, 0, 0, 52)	3.09	2.47	7
7805	l0608MK	(2, 1, 1)	(1, 0, 0, 52)	1.59	1.28	4
7806	l0401	(1, 0, 0)	(0, 0, 1, 52)	3.04	2.34	10
7809	l0306	(1, 0, 1)	(0, 0, 0, 52)	8.22	7.46	23
7810	l0210MK	(0, 0, 0)	(2, 0, 0, 52)	1.87	1.28	17
7811	l0211	(1, 1, 0)	(0, 0, 0, 52)	14.61	13.68	47
7812	l0215MK	(0, 1, 1)	(0, 0, 0, 52)	1.15	1.01	35
7814	l0604	(2, 1, 2)	(2, 0, 0, 52)	3.72	3.18	8
7821	l0212	(0, 1, 1)	(0, 0, 0, 52)	11.68	10.17	27
7822	l0403	(2, 0, 0)	(0, 0, 0, 52)	2.98	2.52	10
7823	l0404	(0, 1, 1)	(0, 0, 0, 52)	3.18	2.47	8
7824	l0307	(0, 1, 1)	(0, 0, 1, 52)	6.99	6.19	18
7825	l0406 INT	(1, 0, 1)	(0, 0, 0, 52)	3.90	3.11	12
7826	l0407MF	(1, 0, 1)	(0, 0, 1, 52)	20.15	19.57	79
7827	l0612	(2, 0, 0)	(0, 0, 1, 52)	2.90	2.38	11
7828	l0302MK	(1, 1, 1)	(1, 0, 0, 52)	1.29	1.07	5
7829	l0610INT	(0, 1, 1)	(0, 0, 0, 52)	2.74	2.15	6

7830	l0411MF	(1, 0, 0)	(0, 0, 0, 52)	5.71	5,02	12
7831	l0811MF	(1, 0, 1)	(0, 0, 0, 52)	14.52	13,45	23
7832	l0304MK	(0, 1, 1)	(0, 0, 0, 52)	6,36	5,62	23
7836	l0301MK	(0, 1, 1)	(1, 0, 0, 52)	1,67	1,39	5
7837	l0601	(2, 1, 1)	(0, 0, 0, 52)	3.74	3,17	8
7838	l0303	(1, 1, 2)	(2, 0, 0, 52)	2,03	1,46	5
7839	l0511MK	(0, 1, 1)	(0, 0, 0, 52)	2,58	2,12	14
7840	l0703	(1, 1, 1)	(0, 0, 0, 52)	7,65	7,03	23
7845	l0510	(1, 0, 0)	(1, 0, 0, 52)	6,93	6,38	22
7847	l0603	(0, 1, 1)	(0, 0, 0, 52)	5,99	5,41	19
7849	l0605	(1, 1, 1)	(2, 0, 0, 52)	4,34	4,15	16
7850	l0606MF	(0, 1, 1)	(0, 0, 0, 52)	3,06	2,63	12
7851	l0413	(2, 0, 0)	(0, 0, 0, 52)	2,72	2,23	5
7853	l0712	(1, 0, 1)	(0, 0, 0, 52)	4,33	3,62	10
7855	l0602	(1, 1, 2)	(1, 0, 0, 52)	4,16	3,80	14
7865	l0704	(0, 1, 1)	(0, 0, 0, 52)	7,27	6,69	23
7874	l0702	(0, 0, 0)	(1, 0, 0, 52)	6,04	5,01	13
7875	l0513	(0, 1, 1)	(2, 0, 0, 52)	7,87	7,21	24
7877	l0710	(3, 0, 1)	(0, 0, 0, 52)	5,06	3,49	7
7880	l0512MFL	(1, 0, 0)	(0, 0, 0, 52)	6,54	5,97	19
7882	l0713INT	(1, 0, 2)	(0, 0, 0, 52)	2,59	2,25	12
7885	H0801	(1, 0, 0)	(1, 0, 0, 52)	4,90	4,19	10
7887	l0708	(0, 1, 2)	(0, 0, 2, 52)	4,60	3,50	9
7891	l0705	(0, 0, 0)	(0, 0, 0, 52)	6,87	6,53	21
7892	l0813	(1, 0, 0)	(1, 0, 0, 52)	6,73	5,84	20
7893	l0716	(0, 0, 0)	(2, 0, 0, 52)	6,34	5,96	21
7894	l0607	(1, 0, 0)	(1, 0, 0, 52)	7,72	7,35	28
7895	l0505	(3, 1, 1)	(0, 0, 1, 52)	6,35	5,75	29
7897	l0609MFL	(0, 1, 2)	(1, 0, 0, 52)	2,50	2,10	9
7901	l0707	(0, 0, 4)	(0, 0, 0, 52)	3,76	3,10	11
7902	l0714	(1, 0, 1)	(0, 0, 0, 52)	5,18	4,13	14
7904	l0504MF	(2, 0, 0)	(0, 0, 0, 52)	8,31	7,72	36
7906	l0613	(1, 0, 0)	(1, 0, 0, 52)	3,05	2,28	4
7963	M9901	(0, 1, 1)	(1, 0, 0, 52)	5,66	5,19	16
8112	l0105	(0, 1, 1)	(0, 0, 0, 52)	5,54	4,78	13
8114	l0202MK	(0, 1, 1)	(0, 0, 1, 52)	1,88	1,20	32
8115	l0201MK	(0, 0, 0)	(0, 0, 1, 52)	3,49	3,01	52
8116	l0214MK	(1, 0, 0)	(0, 0, 0, 52)	2,68	2,00	34
8117	l0209	(0, 1, 1)	(0, 0, 0, 52)	5,95	5,29	15
8118	l0107MK	(2, 1, 1)	(0, 0, 0, 52)	1,73	1,41	6
8120	l0106	(0, 1, 1)	(0, 0, 0, 52)	6,58	5,92	16
8122	l0104	(0, 1, 1)	(0, 0, 2, 52)	2,46	2,18	10
8123	l0103	(3, 0, 1)	(0, 0, 0, 52)	4,71	4,13	16
8124	l0102	(0, 0, 0)	(0, 0, 0, 52)	8,17	7,02	28
8128	l0205	(2, 1, 1)	(0, 0, 0, 52)	4,35	3,95	15
8136	l0203MK	(0, 1, 1)	(0, 0, 1, 52)	8,02	7,02	29
8336	l0701	(0, 1, 1)	(0, 0, 2, 52)	4,95	4,65	16
16369	l0715	(4, 0, 1)	(0, 0, 0, 52)	2,13	1,53	4
16372	l0108MFL	(1, 1, 2)	(0, 0, 2, 52)	6,17	5,75	29
16520	IESA01MF	(1, 1, 1)	(0, 0, 0, 52)	10,85	8,68	15

#### Service

A report with predictive capabilities was developed in a dashboard with the information about the weekly prediction of mixed waste (Figure 2). The report contains visuals that allow to compare the observed and predicted mixed waste produced, for the studied period, for specific cultural/sports events and holidays. The waste production prediction can be also assessed by parish or for a specific or all mixed waste collection circuits.

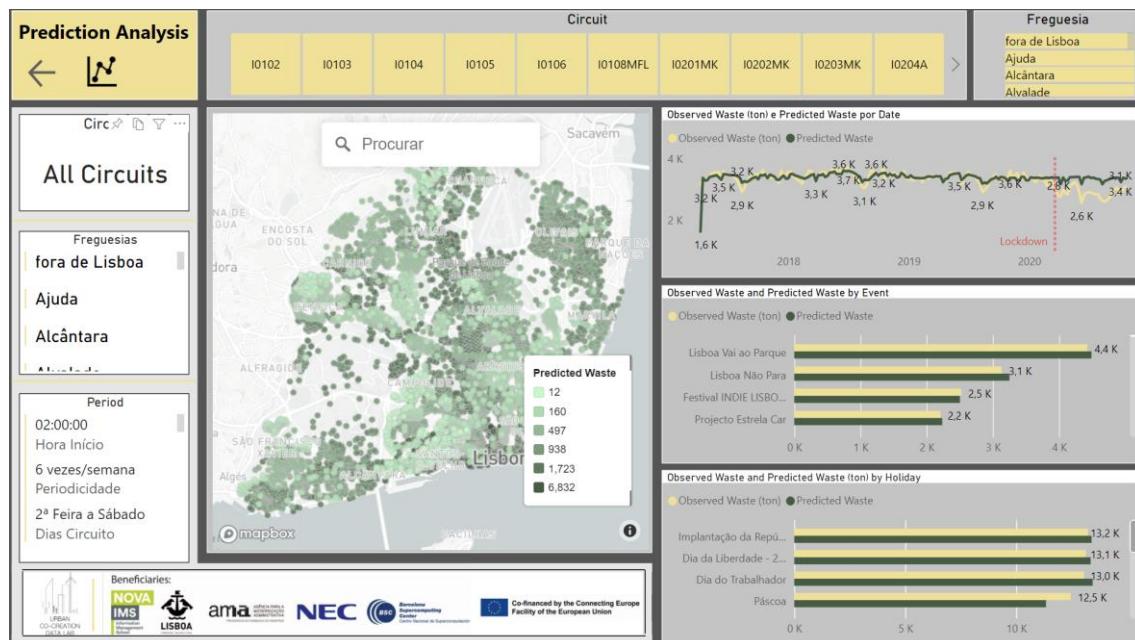


Figure 2. Report on the prediction of mixed waste by waste collection circuit.

#### References

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