

Learning-To-Rank Framework

How to create an army of models?

30.09.2024

### **HAYSTACK**

**Catarina Goncalves** Data Scientist, OLX Group Marcin Gumkowski ML Engineer, OLX Group



Image by pch.vector on Freepik



### About us



From **Olhão**, Portugal



BSc and MSc in Computer Science and Engineering

@ Instituto Superior Técnico (2017-2022)



Catarina Gonçalves

Junior Data Scientist
Lisbon, Portugal

DLX 2

### About us





BSc and MSc in Computer
Science

@ Poznań University of Technology (2009-2014)



**Marcin Gumkowski** 

Senior Machine Learning Engineer Poznań, Poland

### LTR framework

### **Meet the team**





**Ilan Dubois** 

Senior Machine Learning Engineer



Marcin Gumkowski

Senior Machine Learning Engineer



Catarina Gonçalves

Junior Data Scientist



Katarzyna Wagner-Wojciska

Senior Data Product Manager



**Cristian Martinez** 

Data Science Manager

### Agenda

- **■** OLX Group
- Learning to rank algorithm intro (OLX use case)
- OLX Group challenges
- **■** LTR Framework
- Status Quo
- Next Steps and Q&A



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#### We empower millions of people in making key life

decisions\*









Starts a family

01

# Learning to Rank Algorithm Introduction

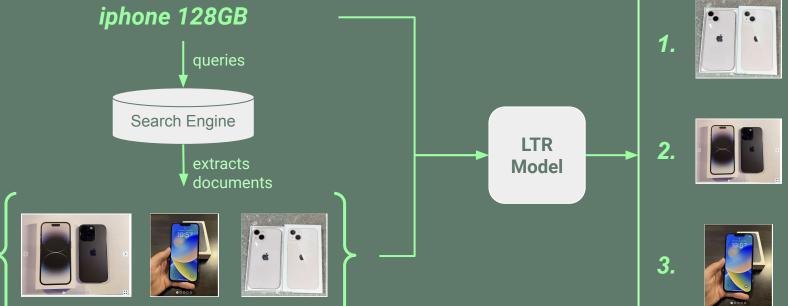




# Learning to Rank









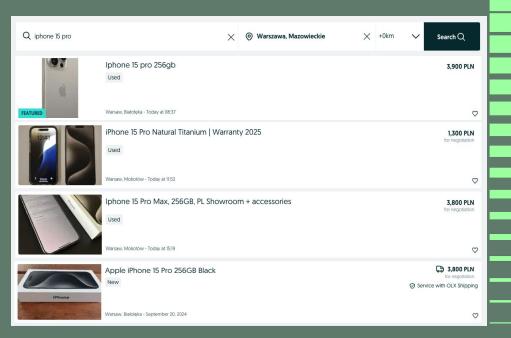
# We could ask our users what ads are relevant for each query



Is it relevant for you?

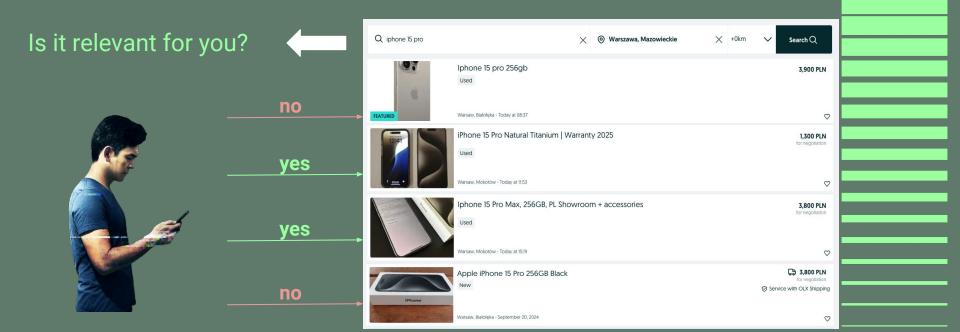






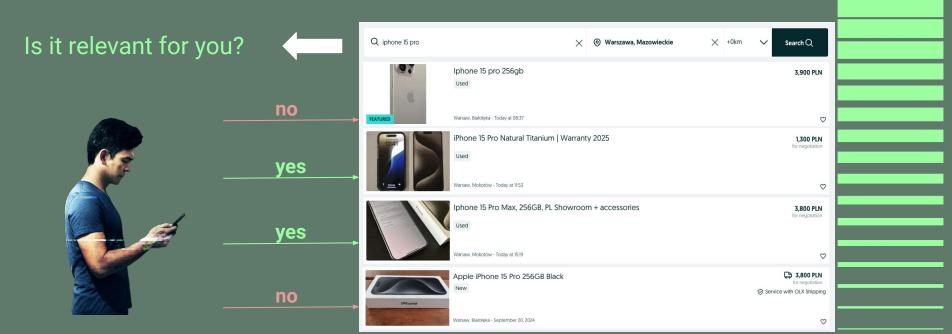
# We could ask our users what ads are relevant for each query





# We could ask our users what ads are relevant for each query





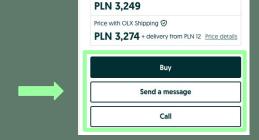
# What are the characteristics that make an ad relevant?





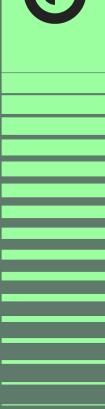


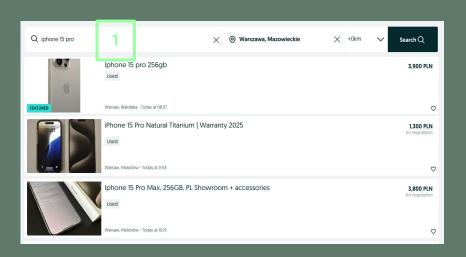
**User Signals** - clickstream events



Just an item

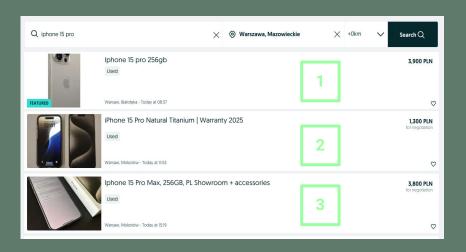






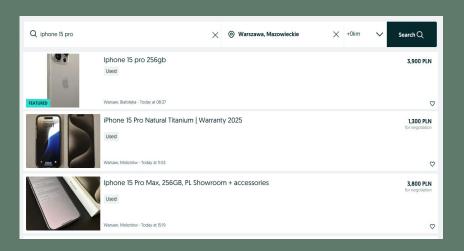






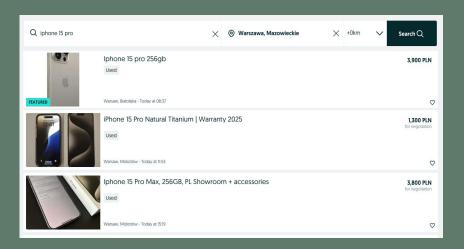
query_id	doc_id
1	1
1	2
1	3





query_id	doc_id	features
1	1	
1	2	
1	3	

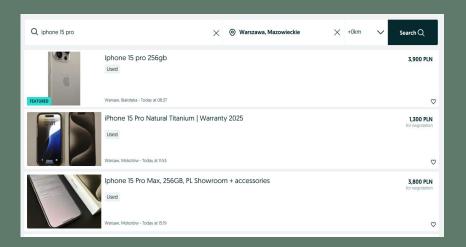






e.g. ad price

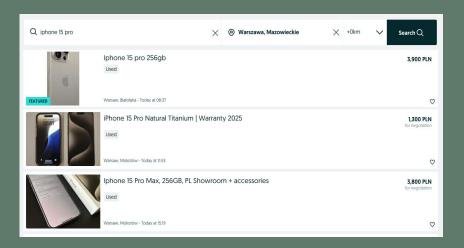






e.g. ad freshness

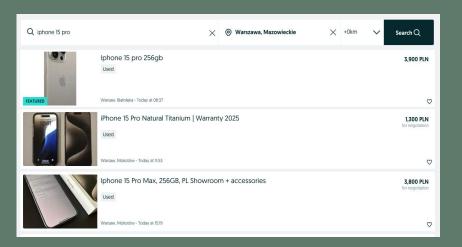




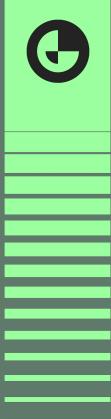


e.g. Click Through Rate



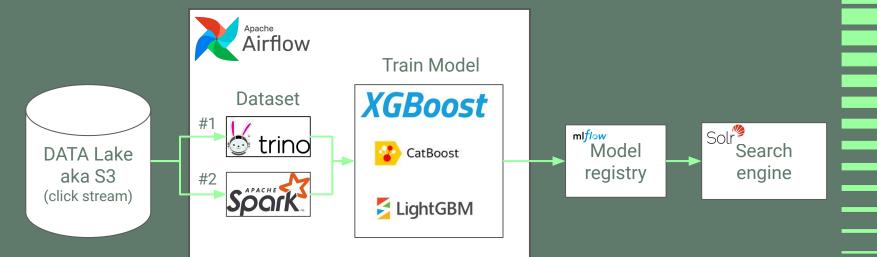


query_id	doc_id	_id features			relevance		
1	1	f0	f0 f1 f2		0		
1	2	fO	f1	f2	1		
1	3	f0	f1	f2	0		





# Usual LTR pipeline



## 02

# Challenges in OLX Group



Core

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**Real Estate** 

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**Motors** 

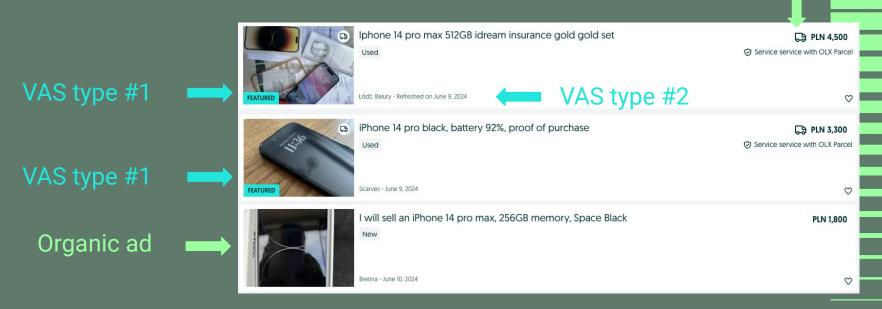


#### #1 challenge



Many Targets to optimize...

Delivery ad

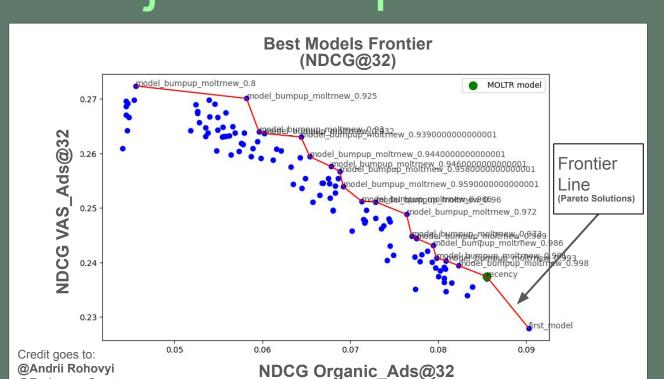


\*VAS - value added-service

### #1 challenge

@Enderson Santos

# Multi objective-optimization

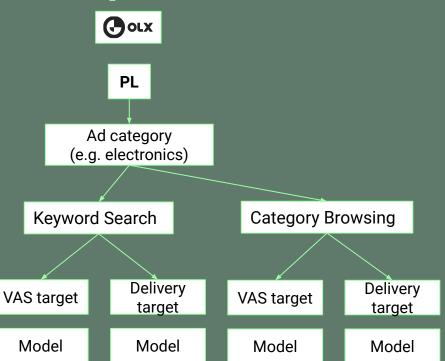




#### #2 challenge

OLX

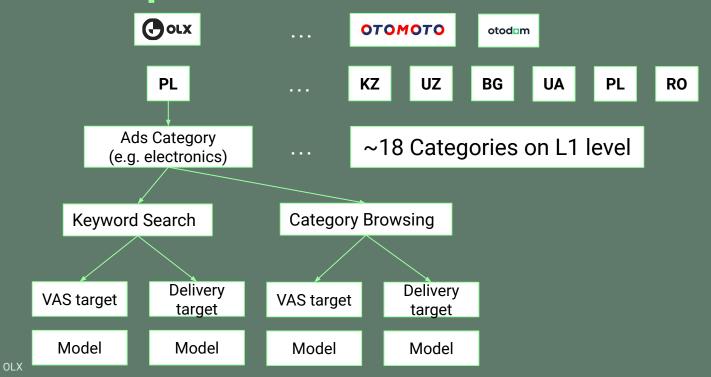
# Multiple models





#### #2 challenge

# Multiple models



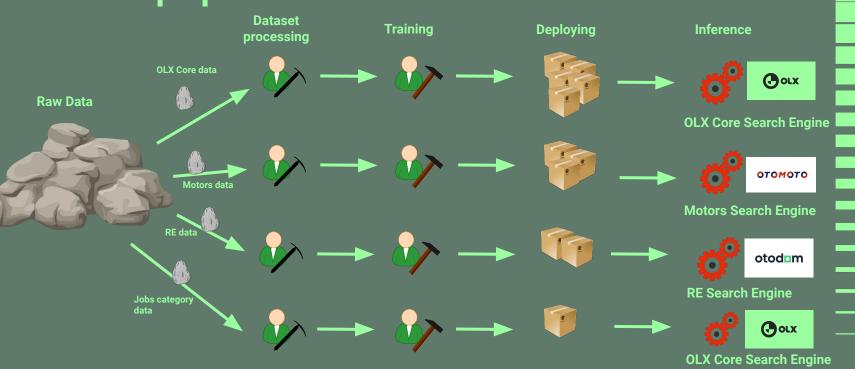


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### #3 challenge

OLX

# The pipelines hell



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	OLX Core	Motors	Real Estate	Jobs in OLX
countries	7 (3 P&S, 4 non P&S)	3	3	7
repos to maintain	3	1	1	1
optimization targets	Delivery, VASes, Successful Events	VASes, Successful Events	VASes, Successful Events	VASes, Successful Events
models to maintain	11+	3+	3+	7+
duplicated o∟€ode	YES	YES	NOT YET	NOT YET





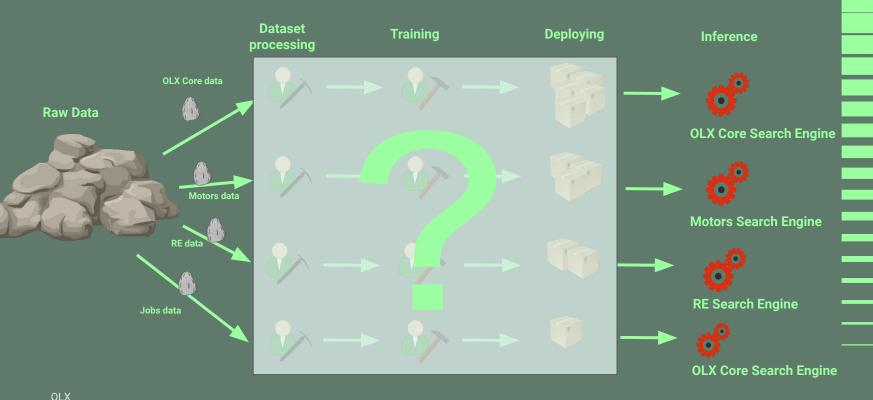
# Ctrl+C, Ctrl+V







## Can we do better?



### LTR Framework motivation



stop duplicating efforts for rankings



stop growing **technical debt** 🧊



add synergy to the LTR projects 🕰





reduce costs across teams 💸



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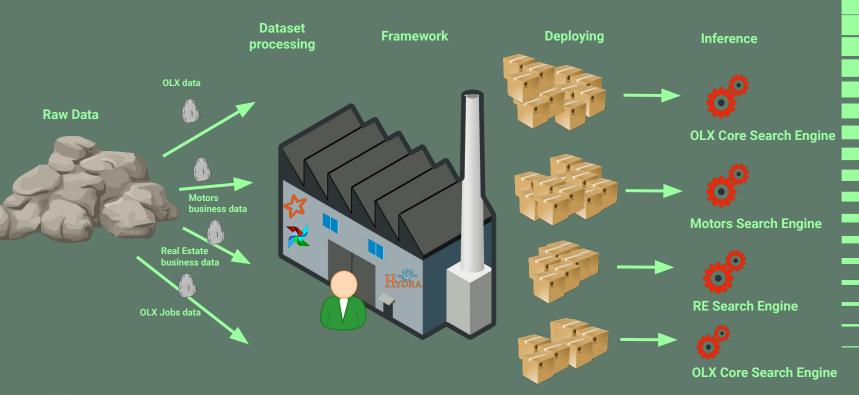
## LTR Framework







### LTR Framework





### Dataset unification

Query ID	Doc ID	Rank	f_*	r_*	e_*	Partition
Unique identifier of the user query	of the ad displayed to the user	Position of the ad on the result page	Features relative to the document or the user	Relevance signals (# clicks, # replies,)	<b>Extra</b> features not previously classified	country, platform, date, 
Required	Required	Required	At least 1	At least 1	Optional	At least 1

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q_id	doc_id	rank	f_price	f_ctr	r_replied	r_click	e_source	country	year	month	day
42	19	0	10	0.98	true	true	promoted	pl	2024	09	30
42	57	1	89	0.95	false	false	promoted	pl	2024	09	30
42	102	2	43	0.95	false	true	organic	pl	2024	09	30

## **Configuration** Setup



3 units 👁 🖦 🕮

1+ model

**3+** countries

1+ target

**3** environments

**GOAL** 

**Increase Flexibility** 

**Enhance Experimentation** 

**Better Organization** 

OLX :

## Configuration Setup

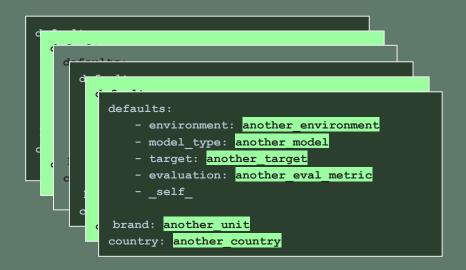
**Scaling with Hydra** 



## Hierarchical Configuration

- Modularity
- Scalability
- Reusability







## Configuration Setup

**Scaling with Hydra** 



## Runtime CLI Overrides

Overriding, appending, removing

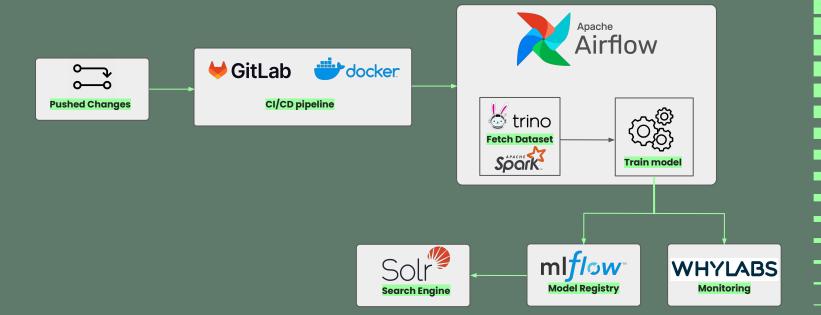
- Flexibility
- Customization
- Dynamic Adjustments











## Monitoring

LTR Framework's setup





and clean data









WHYLABS mlflow

> Save run and log metrics

Log dataset

Train Model

DI X



# Monitoring whylabs



**Established nomenclature** 

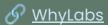
LTR\_FWK\_{unit}\_{country}

### **Key Features**

- Data Quality Observation
- **Performance** Monitoring
- Customizable Dashboards
- Detect and root cause ML issues
- Supports Ranking Projects

### **Benefits**

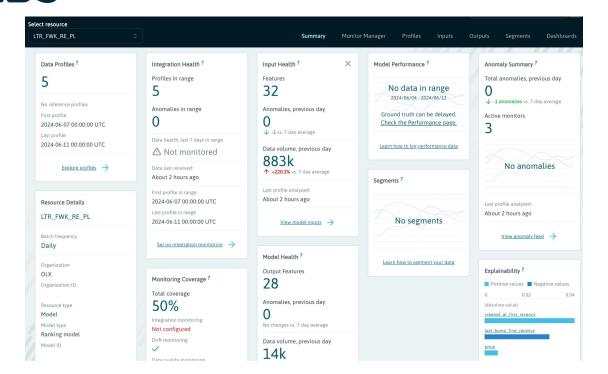
- Enhanced Project Organization
- Improve **Model Reliability**
- Reduce Downtime



# 0

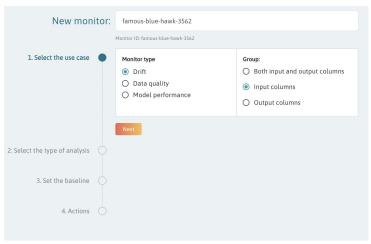
# Monitoring whylabs

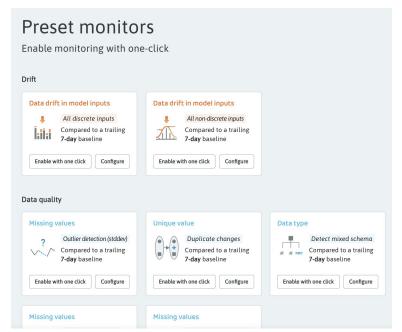
**Overview** 



# Monitoring whylabs

### **Monitors**



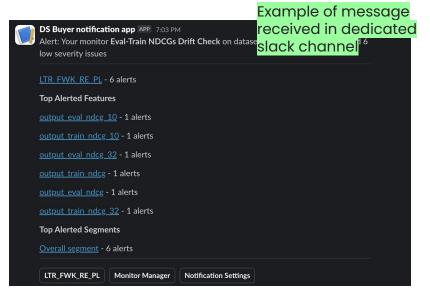


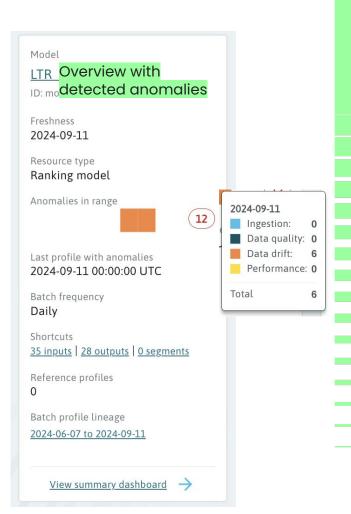
LX 44



# Monitoring **WHYLABS**

### **Monitors Alerts**



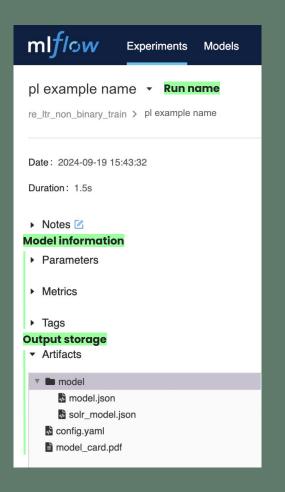


# Monitoring mlflow

### **Key Features**

- Experiment Tracking
- Reproducibility
- Up-to-date Model Storage
- Customizable









## Monitoring

## Model Cards mlflow

#### Ranking insights

#### **Avg position Before After**

Bumped ads 14.6 14.0 Clicks 14.6 13.9 Replies 14.1 13.0

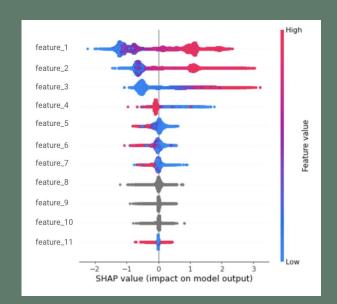
We present here the offline NDCG metrics.

Dataset figures are the performance that actually happened. Evaluation values represent the theorical performance on new data.

NDCG@10 NDCG@32

Dataset 0.277 0.396 Evaluation 0.317 [14.6%] 0.424 [7.0%]

#### Target: Non-tech people

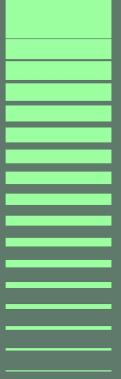


example

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## **Status Quo**





## Releases

Motors **Planned** 

OLX Core 30% Progress



All 7 countries

Real Estate Integrated







Portugal





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**MLFlow Integration** 



**WhyLabs Monitoring** 







Poland Portugal







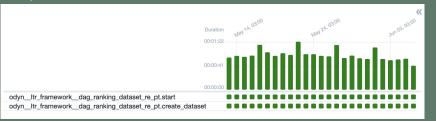
#### **Automated Pipelines**

Daily automated pipelines for efficient dataset creation and model training, ensuring continuous tracking and up-to-dateness of models.

#### DAG for each Country



#### Example



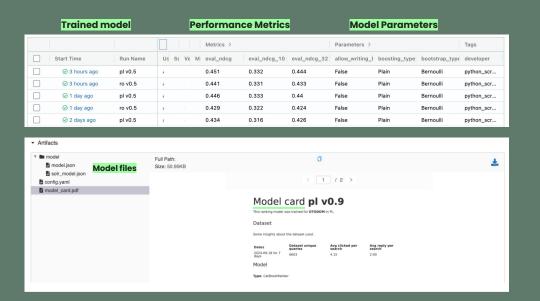




# mlflow

## MLflow Integration & Automation

After training, the resulting trained models, features, parameters, and performance metrics, all complemented by detailed model cards, are stored in the desired MLflow experiment.





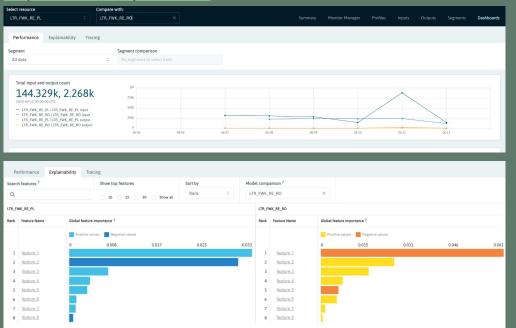
## Real Estate Release

## WHYLABS

#### **WhyLabs Monitoring**

For each country, there is a WhyLabs resource to track training data and model outputs, utilizing both WhyLabs' and custom metrics for comprehensive performance evaluation and data quality observation.

#### Comparison with other WhyLabs resources



# LTR framework speeding up the development

	OLD WAY		FRAMEWORK	
Nr of weeks needed to conduct each step	adding new feature	scaling features to other country	adding new feature	scaling features to other country
Calculate new feature				
Add new feature to the dataset				
Train new models and select the best one				
Run AB test				
Analyse results & evaluate acceptance criteria				
TOTAL # weeks				

Adding new features ~25% faster

Training a model ~50-75% faster



# LTR framework speeding up the development



Adding **new features** to dataset

**75%** 

Adding new features (ENTIRE PROCESS)

**25%** faster

**Model Training** 

75% faster

**OLD WAY VS LTR FRAMEWORK** 

Percentual impact on time needed to conclude each step

## LTR framework **Motivation**

# **Overview**



MOTIVATION 🚀 Status Stop duplicating efforts Stop growing technical debt Add **synergy** to the LTR projects

Enable rapid experimentation

Reduce costs across teams



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## Next Steps





## What's Next?



### **Migrations**

Migrate the remaining units into the Framework.







### Feature Engineering

Take advantage of cross-project synergies to identify features that proved value in other contexts.



## Scaling Up

Accommodate growth, ensure performance, and meet the increasing demands of the system.





**Q&A**Any questions?



