



# Yogurt Sale Prediction

Customer Data Analytics – Migros

Group 6 – Case 1

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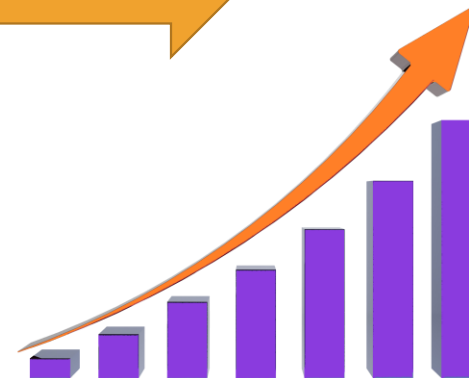
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Suter

# SEVEN-UP DATA FACTORY

# Agenda

- Research Question - Goal
- Data Basis & Data Enrichment
- Models for Sale Prediction
- Results
- Reflection & Dealing with Problems
- Implications

# USE CASE





## Research Question

How can we find the best model to predict sales data for multiple products on the example of yogurts?



# DATA

# WHICH FACTORS INFLUENCE SALES?

"Healthier lifestyle, more knowledge about food and higher income is associated with more yoghurt intake"

(D'Addezia et al., 2015; Possa et al, 2020; Robinson 2017)



**SALES**

**PRICES**



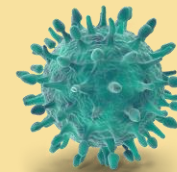
**Google  
TRENDS**

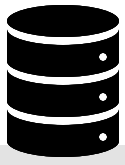


"During lockdown on average one kilogramm more yoghurt in Switzerland"


(BauernZeitung Donnerstag, 25. März 2021)

**EVENTS**





# DATA AS BASIC FOR OUR SOLUTION

Data Migros	Consolidate Data for Models	Enrich Data
Article		Yogurt prices
Year / month / week		Milk prices
Sales		Google search terms „gesund.essen“ & „joghurt“
Promotion 1-5		Lockdown dates
		Holidays



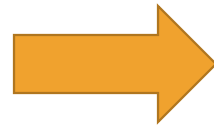


# PROCEDURE & ARCHITEKTURE

# PROBLEM

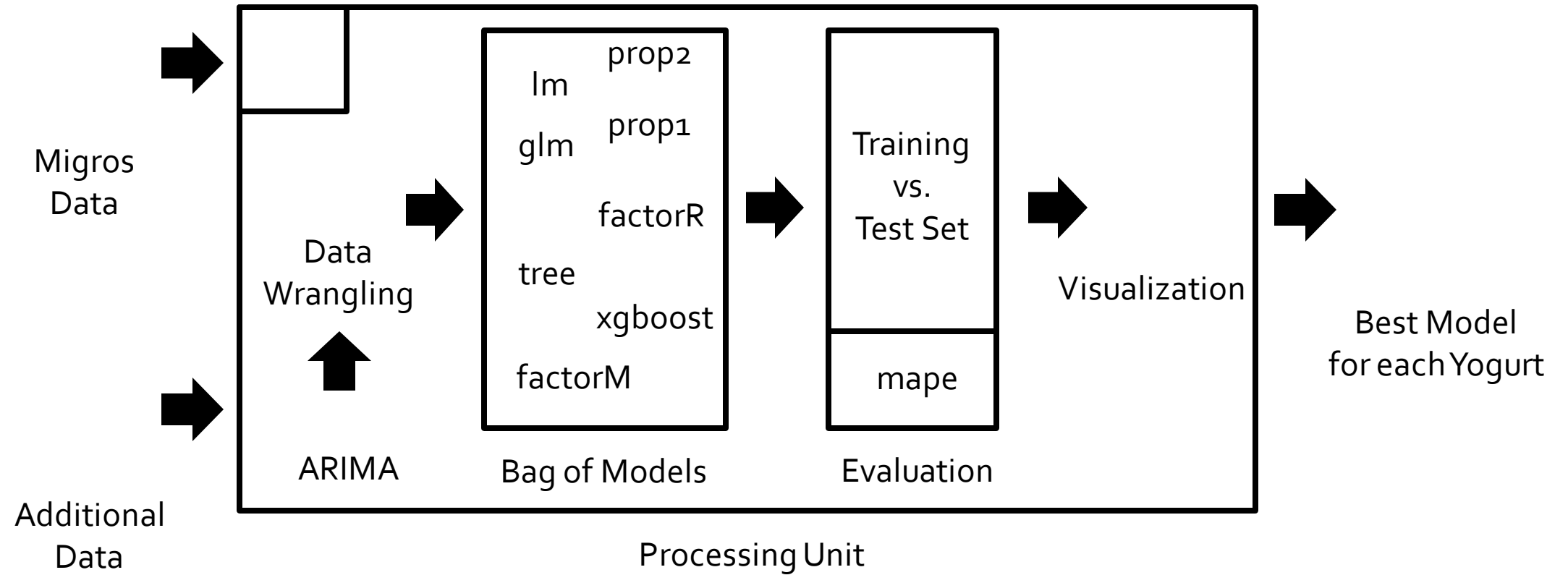


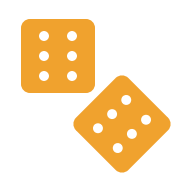
# SOLUTION





# PROCEDURE – ARCHITECTURE





# MODELS & RESULTS

# BEST MODEL FOR EACH YOGURT

lm	glm	tree	prop1	prop2	xgboost	factorM	factorR
0	0	1	1	2	4	10	1

prod	model	prod	model
BIO JOGHURT NATURE 180G	RegressionTree	M-CLAS JOG. HIMBEER 200G	ProphetScoresM
YOGOS GRECQUE NATURE 180G	Prophet1	M-CLAS JOG. SCHOKOLA 200G	ProphetScoresM
BIO FAIRT.JOG.MOKKA 180G	Prophet2	M-CLAS JOGHURT MOKKA 200G	ProphetScoresM
EXC JOGHURT TRUFFES 150G	Prophet2	M-CLAS JOG. APF/MANG 200G	ProphetScoresM
M-CLAS JOG. NATURE 200G	XGBOOST	M-CLAS JOG. ERDBEER 200G	ProphetScoresM
M-CLAS JOG. VANILLE 200G	XGBOOST	M-CLAS JOG. HASELNUS 200G	ProphetScoresM
BIFIDUS JOGH. NATURE 500G	XGBOOST	BIO JOGHURT NATURE 500G	ProphetScoresM
VALFLORA CREME FRAICHE NA	XGBOOST	BIFIDUS JOGH. NATURE 150G	ProphetScoresM
MB JOGHURT NATUR 500G	ProphetScoresM	AHA JOGHURT LAKTOSEF CLAS	ProphetScoresRest
M-CLAS JOG. HEIDELBE 200G	ProphetScoresM		



## ZOOM IN EXCELLENCE TRUFFES – MAPE IN TESTSET



Excellence  
Truffes 150g

	lm	glm	tree	prop1	prop2	xgboost	factorM	factorR
BIO FAIRT.JOG.MOKKA 180G	12.16	11.32	11.59	8.32	5.42	7.08	11.06	10.37
MB JOGHURT NATUR 500G	11.84	15.01	7.79	8.76	8.21	6.52	5.95	9.56
M-CLAS JOG. HEIDELBE 200G	9.05	12.46	11.55	11.23	9.32	7.32	5.51	8.53
M-CLAS JOG. NATURE 200G	9.54	11.76	10.59	8.05	7.30	5.38	13.95	15.62
M-CLAS JOG. HIMBEER 200G	9.24	14.30	13.17	17.71	14.54	7.30	6.46	11.22
M-CLAS JOG. SCHOKOLA 200G	11.19	9.86	7.16	8.59	10.62	8.72	6.20	6.50
M-CLAS JOGHURT MOKKA 200G	11.00	10.23	7.27	8.43	9.23	10.82	5.89	7.54
M-CLAS JOG. APF/MANG 200G	10.62	11.88	7.86	9.90	9.00	8.38	6.52	7.18
M-CLAS JOG. ERDBEER 200G	9.06	13.76	5.64	10.92	8.29	8.86	3.55	7.38
M-CLAS JOG. VANILLE 200G	13.60	14.68	9.60	11.02	10.59	7.60	8.21	9.64
M-CLAS JOG. HASELNUS 200G	9.16	10.37	10.14	9.28	8.49	9.73	7.75	10.64
BIFIDUS JOGH. NATURE 500G	7.05	9.71	5.24	8.37	5.71	4.85	16.45	6.86
BIO JOGHURT NATURE 180G	9.64	9.82	6.51	11.36	11.34	7.82	6.63	14.42
EXC JOGHURT TRUFFES 150G	15.30	13.74	20.03	13.32	7.43	7.52	11.81	11.31
YOGOS GRECQUE NATURE 180G	7.30	11.70	13.14	5.16	6.06	6.72	20.42	9.07
BIO JOGHURT NATURE 500G	12.70	17.77	7.49	13.32	11.53	9.06	7.24	15.35
BIFIDUS JOGH. NATURE 150G	10.25	10.04	6.13	9.41	7.08	5.91	5.82	8.53
VALFLORA CREME FRATCHE NA	16.50	21.34	13.71	21.06	14.04	9.22	18.08	11.71

lm	glm	tree	prop1	prop2	xgboost	factorM	factorR
15.30	13.74	20.03	13.32	7.43	7.52	11.81	11.31

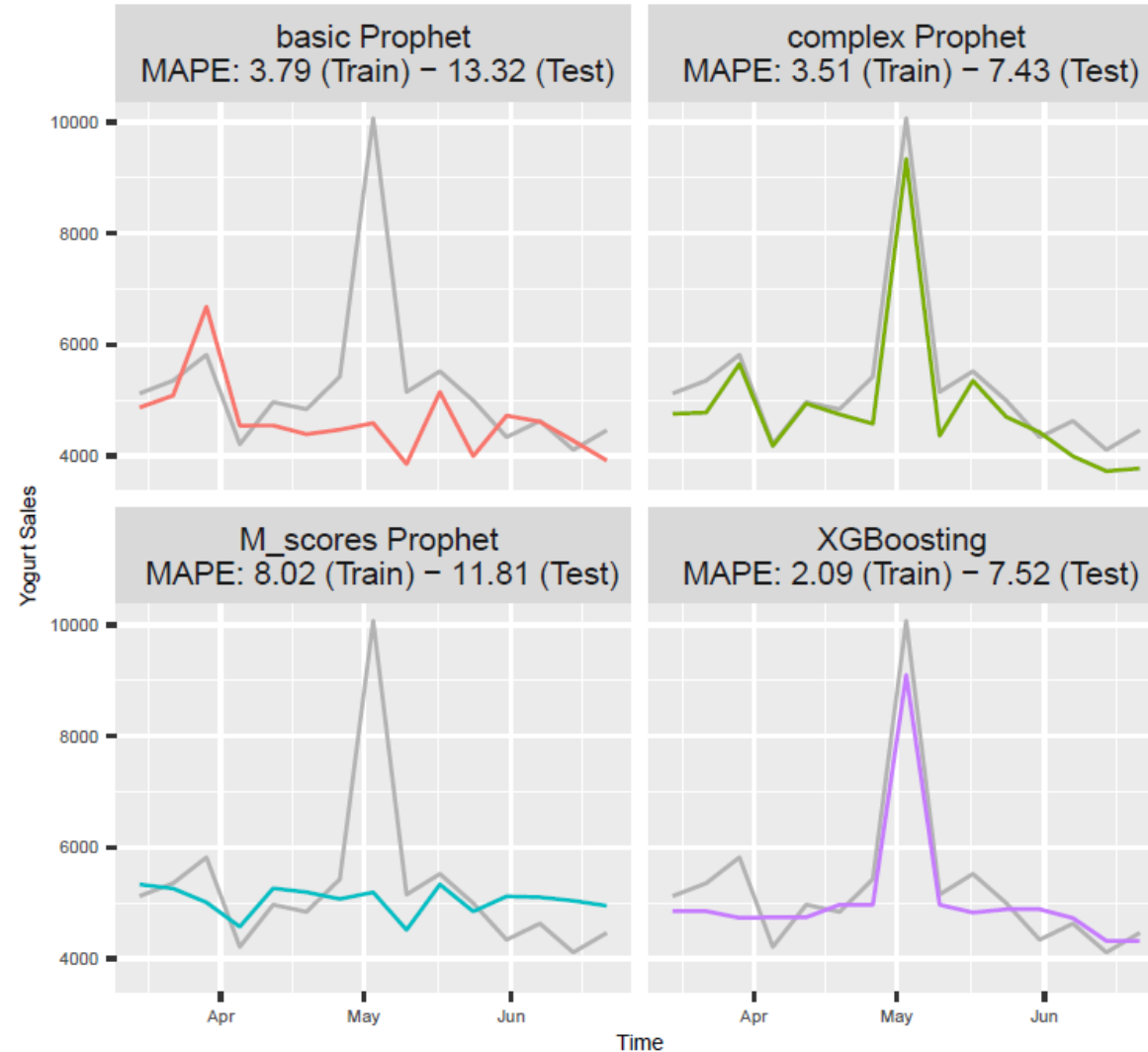
# ZOOM IN EXCELLENCE TRUFFES – MODEL COMPARISON



Excellence  
Truffes 150g

variable

- pred\_Prophet1
- pred\_Prophet2
- pred\_ProphetScoresM
- pred\_XGBOOST

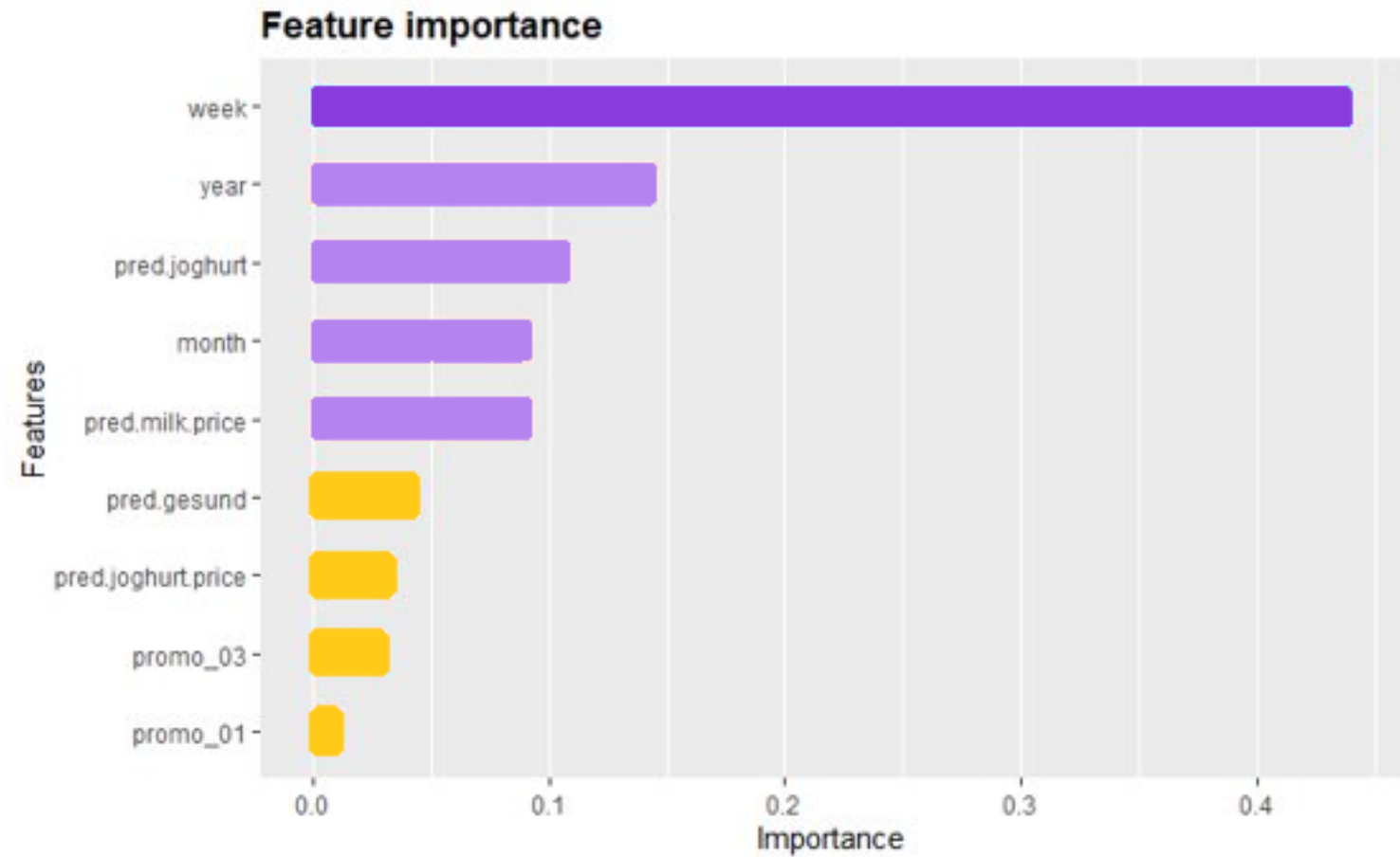




# ZOOM IN EXCELLENCE TRUFFES – XGBOOSTING



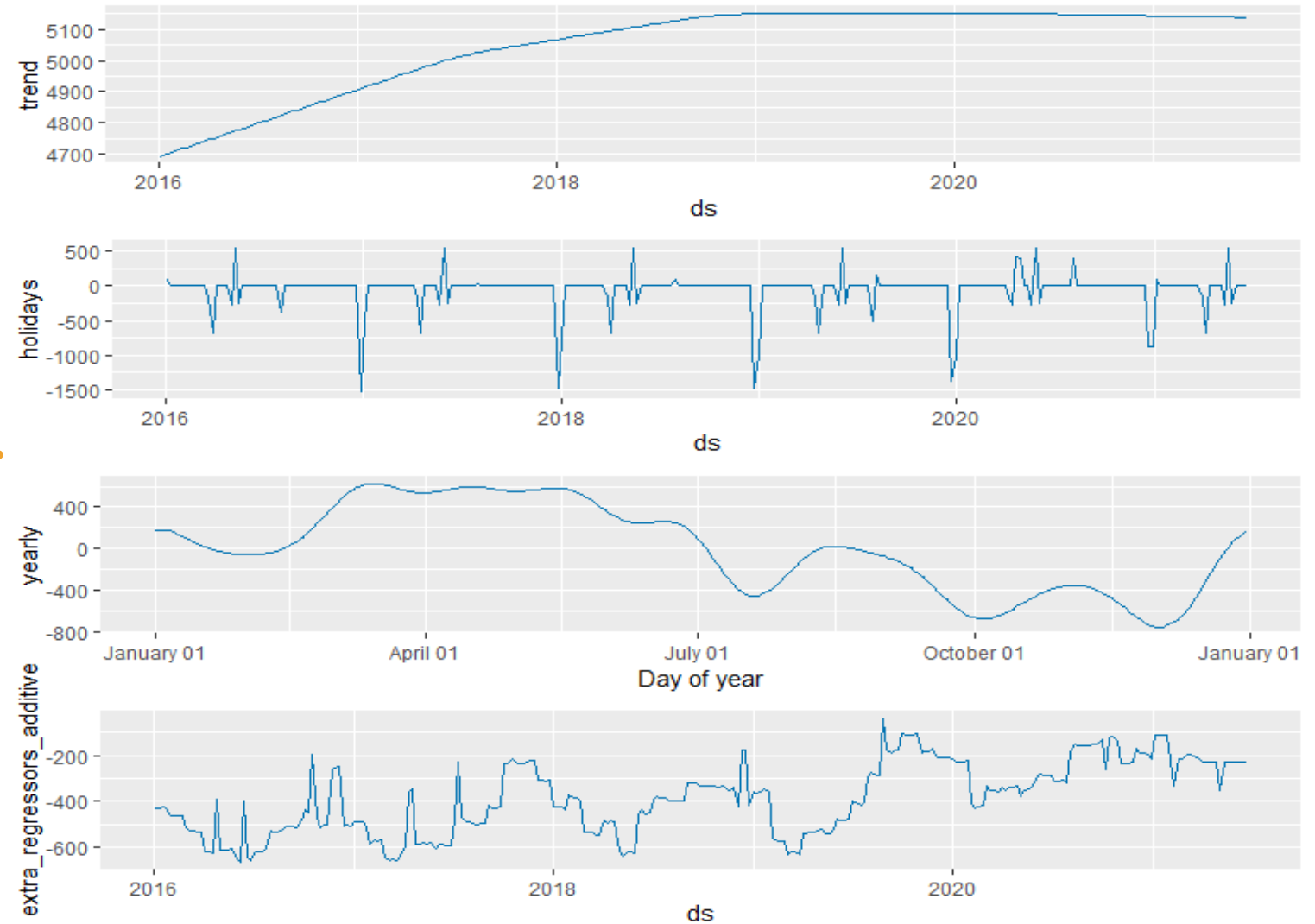
Excellence  
Truffles 150g



# ZOOM IN EXCELLENCE TRUFFES- PROPHET<sub>2</sub>



Excellence  
Truffles 150g



## Additional Features

- Automatically predict future values with best model (based on MAPE of Testset)
- Other models or multiple models > specify in script
- PDF with code and all important plots



# REFLECTION & IMPLICATIONS

# REFLECTION – DATA



## GTrends API

- API retrieves time interval dependent on input
- Limited amount of API calls per day -> Future: get data in a separate micro service

## Yogurt & Milk Prices

- just monthly
- no yogurt prices from Migros

Other data like Cumulus or data per store

# REFLECTION – PROCEDURE & MODELS



## **Regressors must be predicted**

- Google Trends, Milk & Yogurt Prices
- Bad predictions can strongly worsen the result

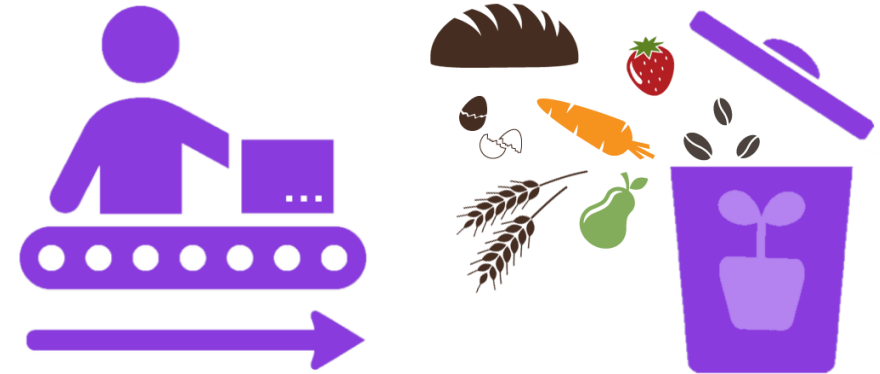
## **Model Evaluation**

- Cross validation only for one time interval
- Better: cross validation on rolling basis

## **Any model for each yogurt**

- Not always the same method worked best
- LM/GML models not optimized for the individual yogurt

# IMPLICATIONS



**added value:**  
automatically calculate all  
models for each yogurt and  
predicts with the best one

adapted production ->  
reduction of  
production costs & food waste

# OUTLOOK



Evaluation of Forecast

OPTIMIZE ?



Cost Benefit Analysis





QUESTIONS

## Sources

- <https://www.bauernzeitung.ch/artikel/landleben/pro-kopf-wurde-2020-rund-ein-kilo-mehr-joghurt-verspeist-353195>
- D'Addezio, L., Mistura, L., Sette, S., & Turrini, A. (2015). Sociodemographic and lifestyle characteristics of yogurt consumers in Italy: Results from the INRAN-SCAI 2005-06 survey. *Mediterranean Journal of Nutrition and Metabolism*, 8(2), 119-129.
- Possa, G., de Castro, M. A., Marchioni, D. M. L., Fisberg, R. M., & Fisberg, M. (2015). Probability and amounts of yogurt intake are differently affected by sociodemographic, economic, and lifestyle factors in adults and the elderly—results from a population-based study. *Nutrition Research*, 35(8), 700-706.
- Robinson, R. A. (2017). Examination of the Demand for Yogurt by Brand, 2009-2011 (Doctoral dissertation).