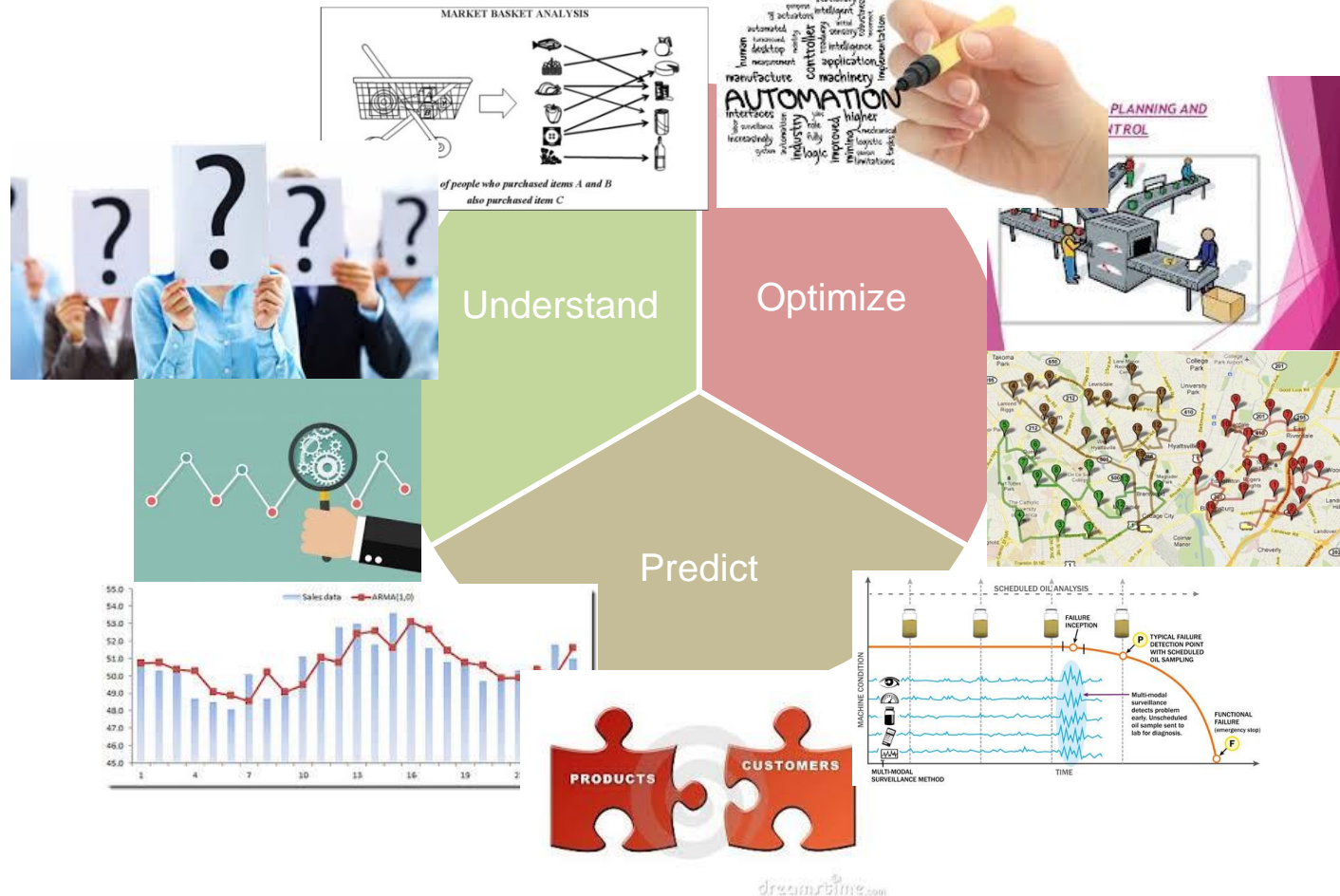


Data Science und Machine Learning in der Praxis

Üetliberg, September 12th, 2017
www.boak.ch



BUSINESS PROCESSES & DATA ANALYTICS



BUSINESS PROCESSES & DATA ANALYTICS

► Value Chain Model



BUSINESS PROCESSES & DATA ANALYTICS

› Value Chain Model



- › Forecasting of sales
- › Efficient fleet routing
- › ...



BUSINESS PROCESSES & DATA ANALYTICS

› Value Chain Model



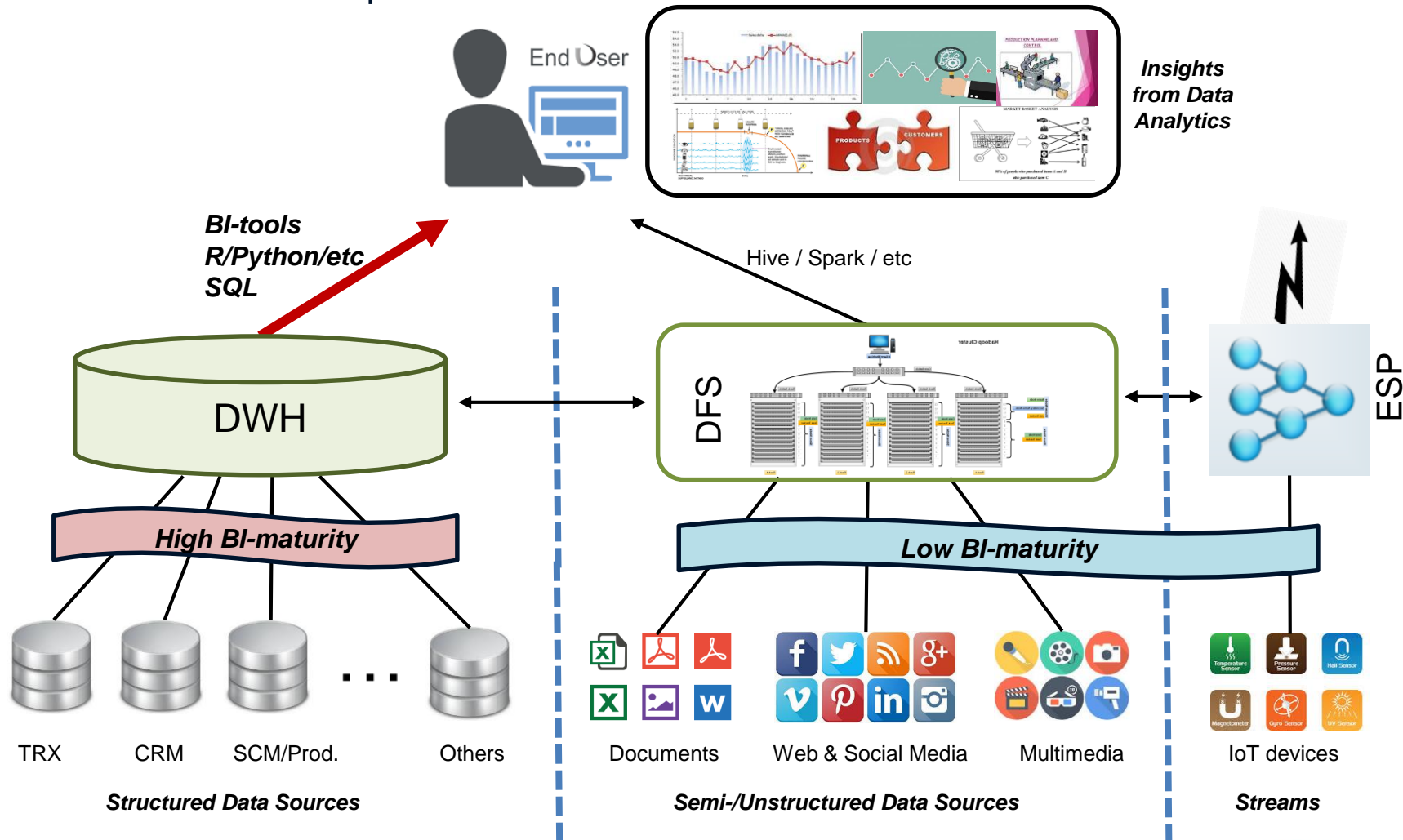
- › Customer understanding
- › 1:1 marketing
- › ...



dreamstime.com

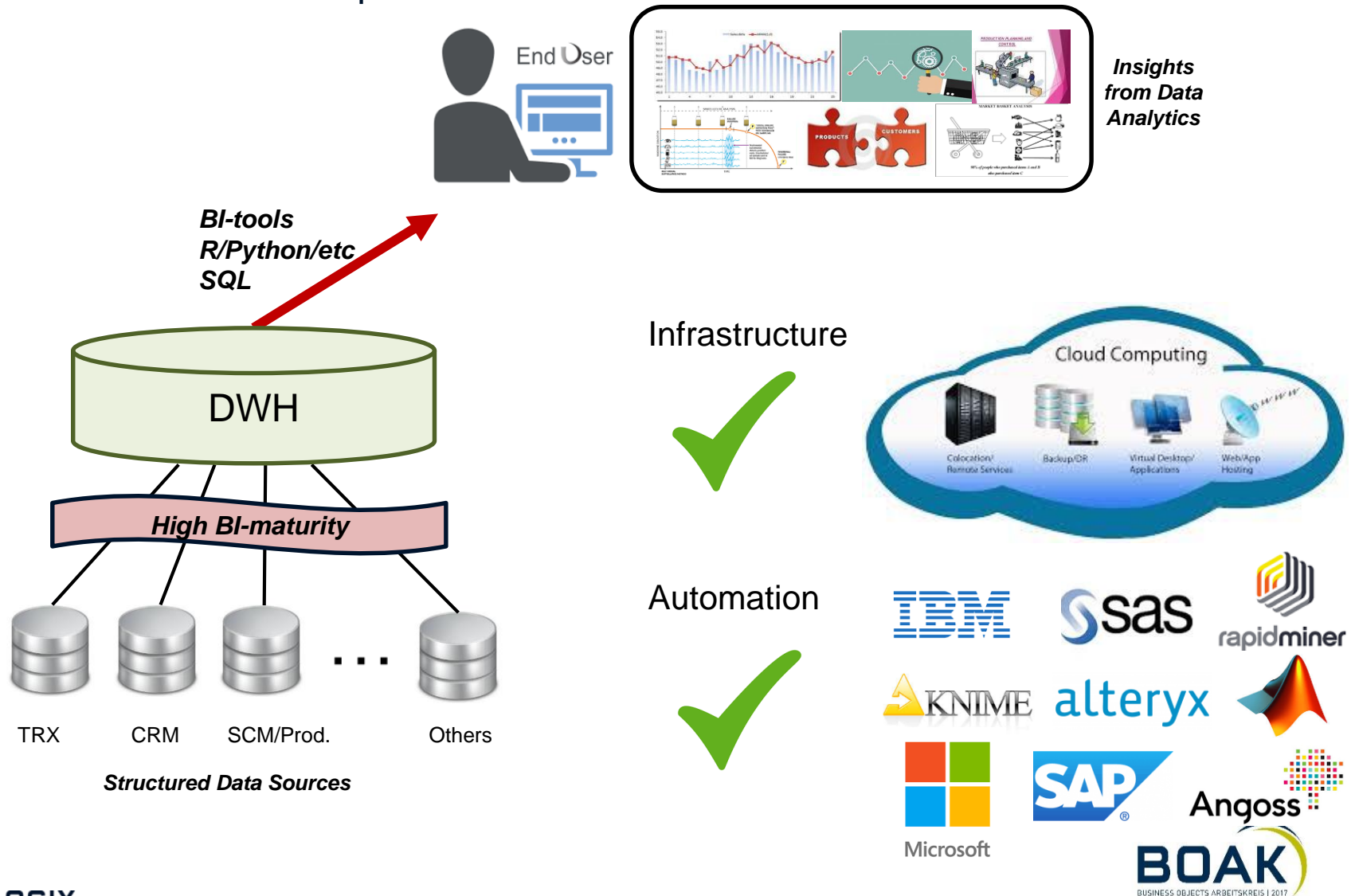
DATA SCIENCE & ARCHITECTURE

➤ Swiss mid-sized companies: current status



DATA SCIENCE & ARCHITECTURE

➤ Swiss mid-sized companies: current status



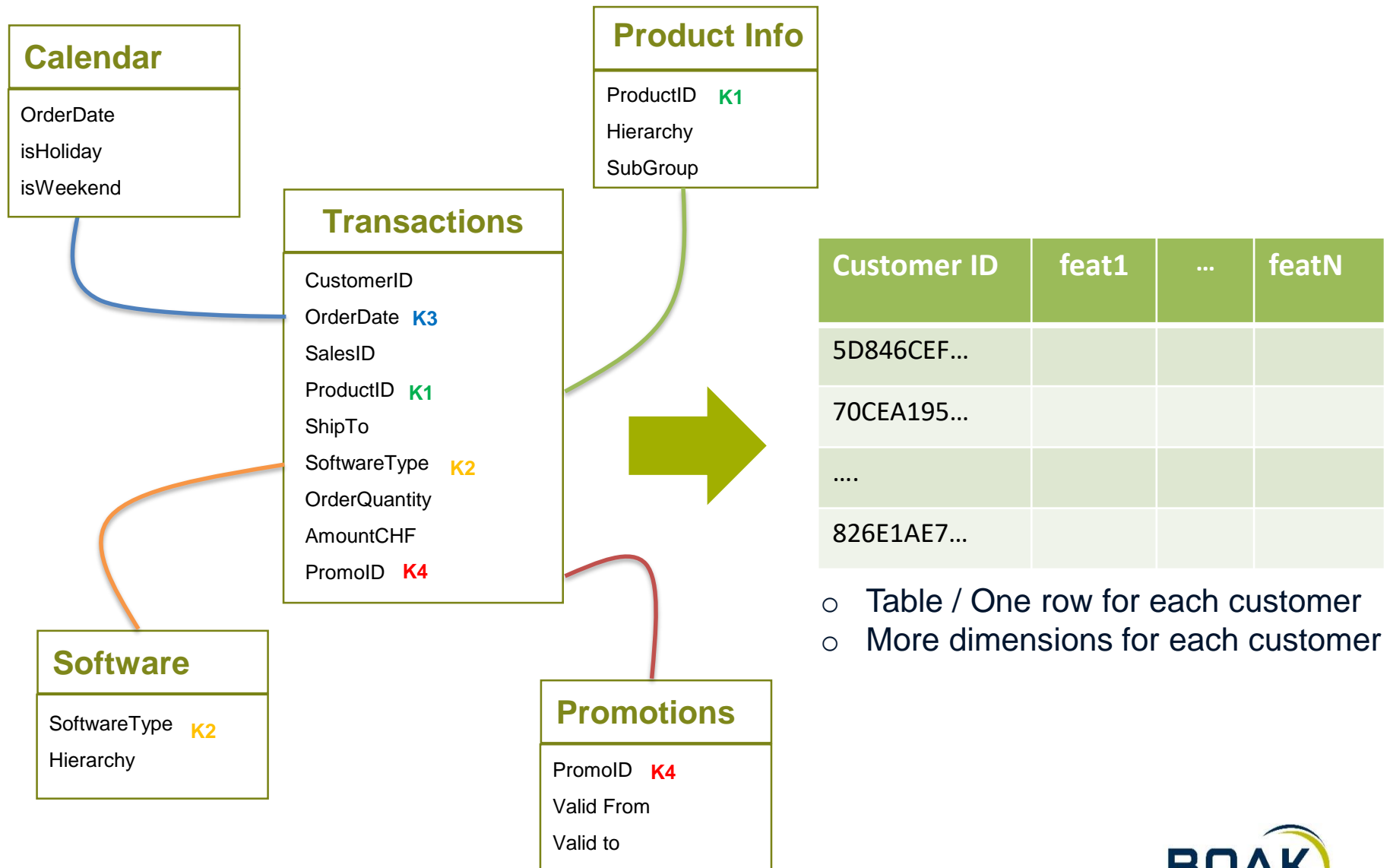
OUTLOOK

- Customer analytics
 - Better understand your customers
- Predictive analytics
 - Run predictions of quantities
- Geo-analytics
 - Leverage geodata
- Others
 - Data analytics projects @IT-Logix

CUSTOMER ANALYTICS

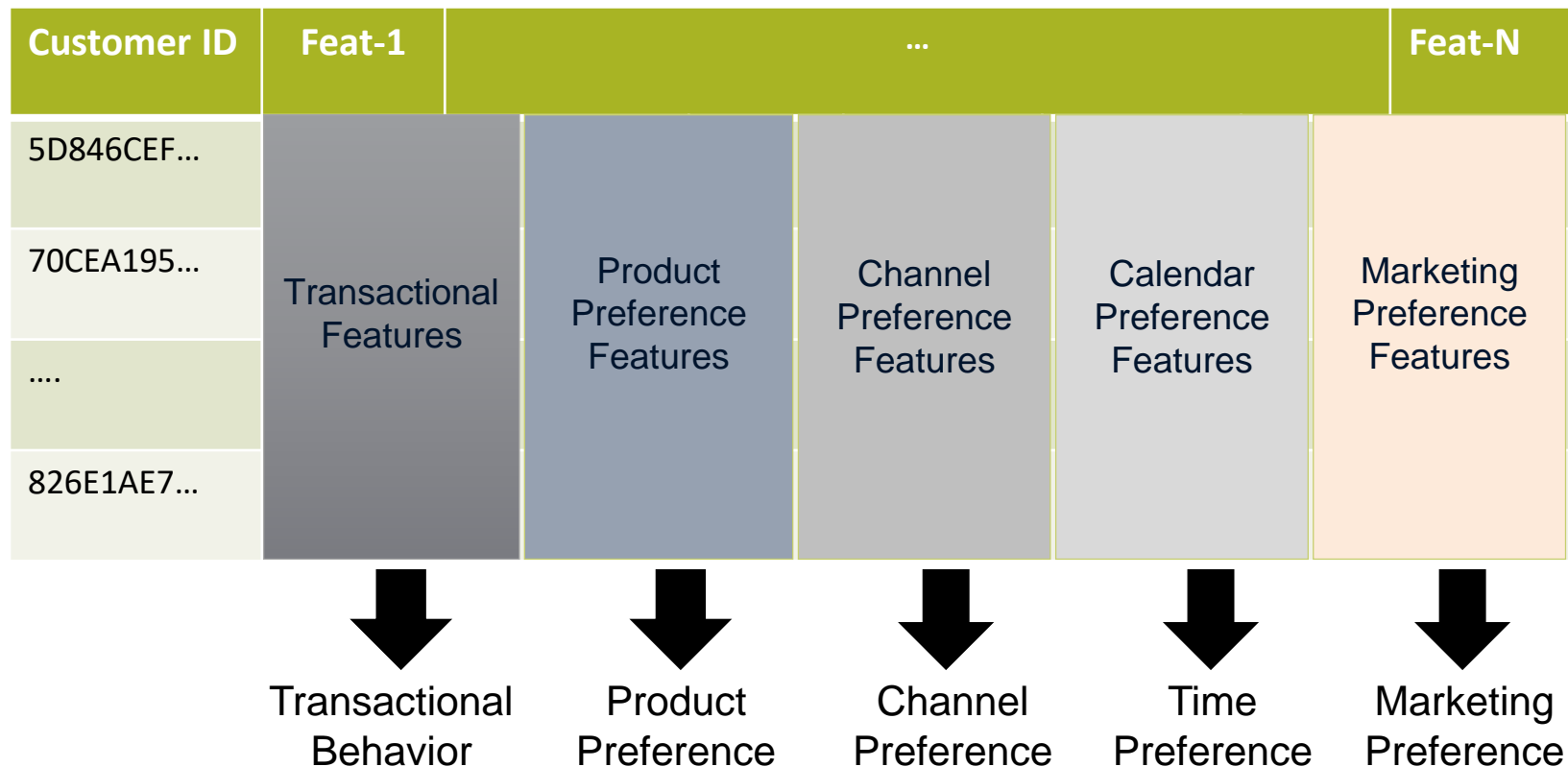


CUSTOMER ANALYTICS



CUSTOMER ANALYTICS

- Extraction of customer features
 - Table / One row for each customer
 - More dimensions for each customer



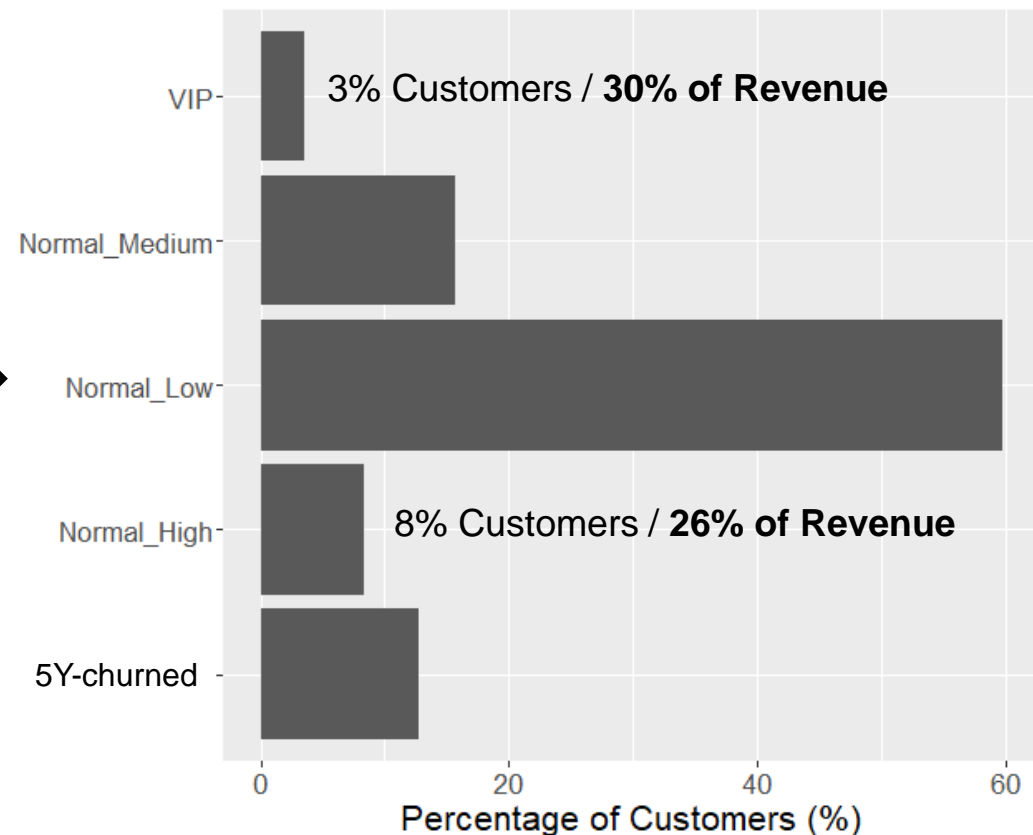
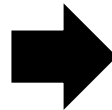
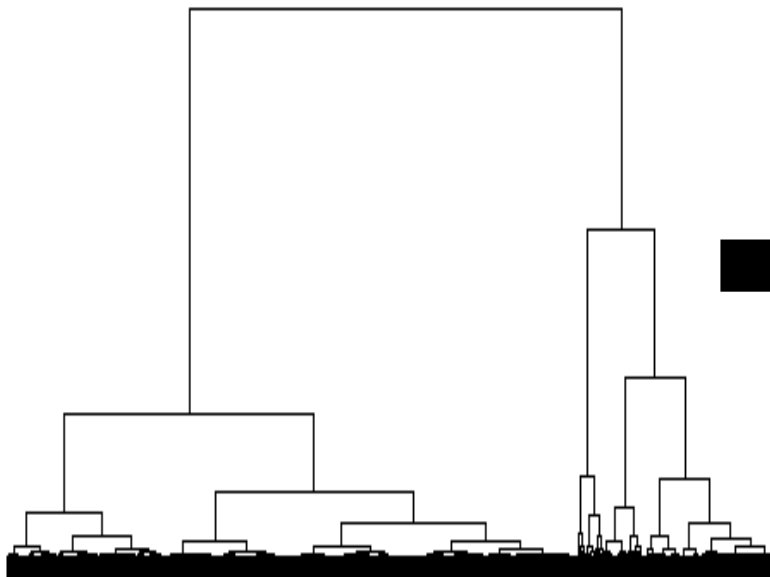
CUSTOMER FEATURES

Transactional
Features

► Amount Class

❑ Clustering based on the “Purchase.Amount.Score”

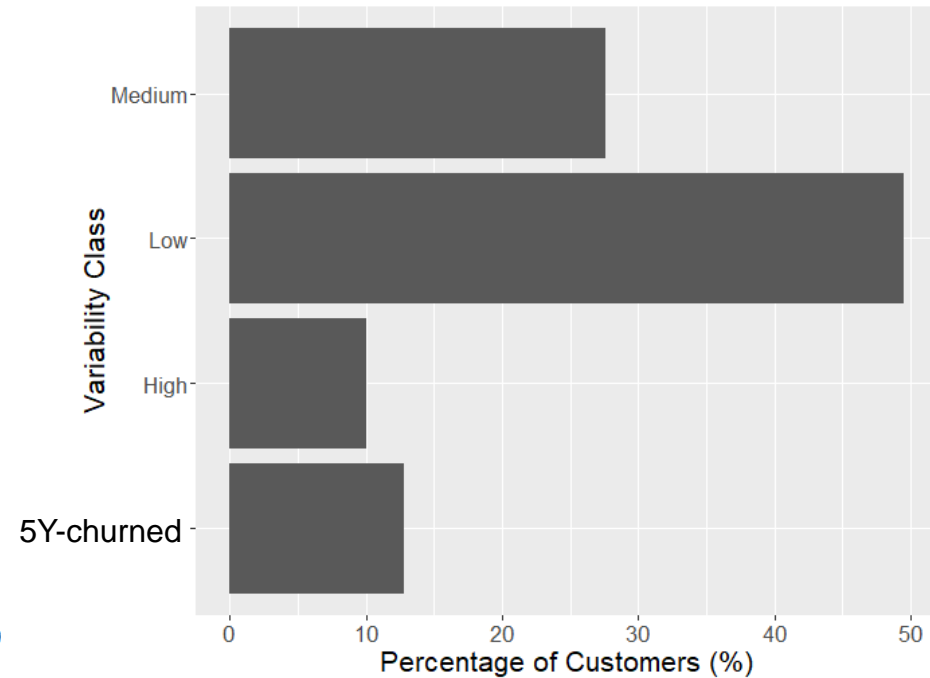
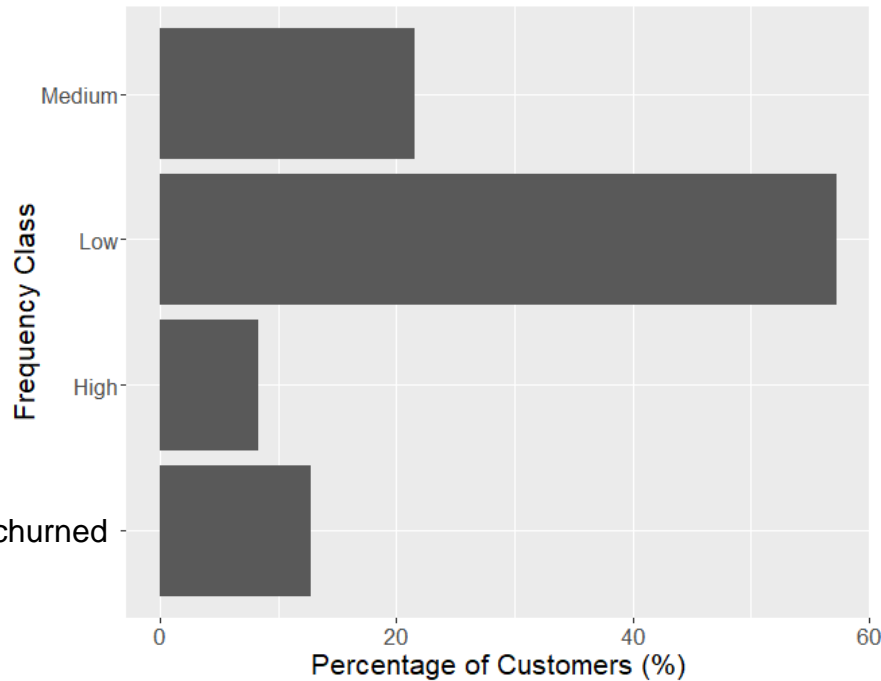
❑ $\text{Purchase.Amount.Score} = 4 \times [1Y]_{\text{CHF}} + 2 \times [2Y]_{\text{CHF}} + 1 \times [3Y-5Y]_{\text{CHF}}$



CUSTOMER FEATURES

Transactional
Features

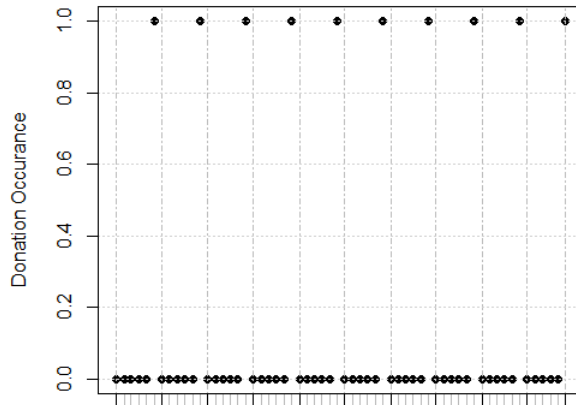
► Frequency & Variability Class



CUSTOMER FEATURES

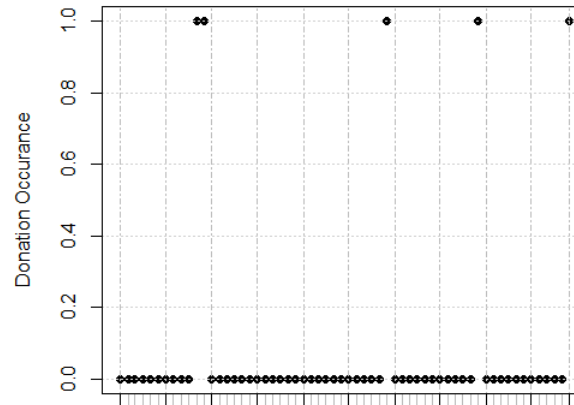
Transactional
Features

► Periodicity Class



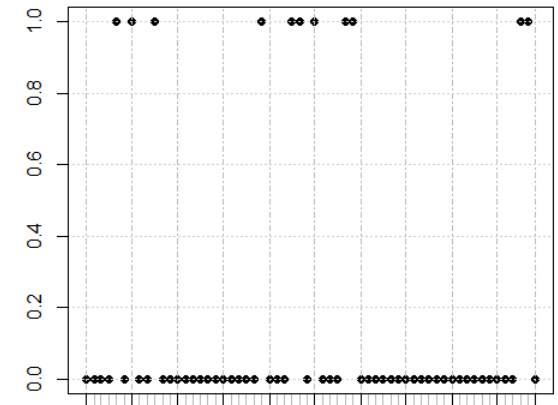
Periodicity Class	Customer Base (%)
-------------------	-------------------

Strong	0.5%
--------	------



Periodicity Class	Customer Base (%)
-------------------	-------------------

Weak	5%
------	----

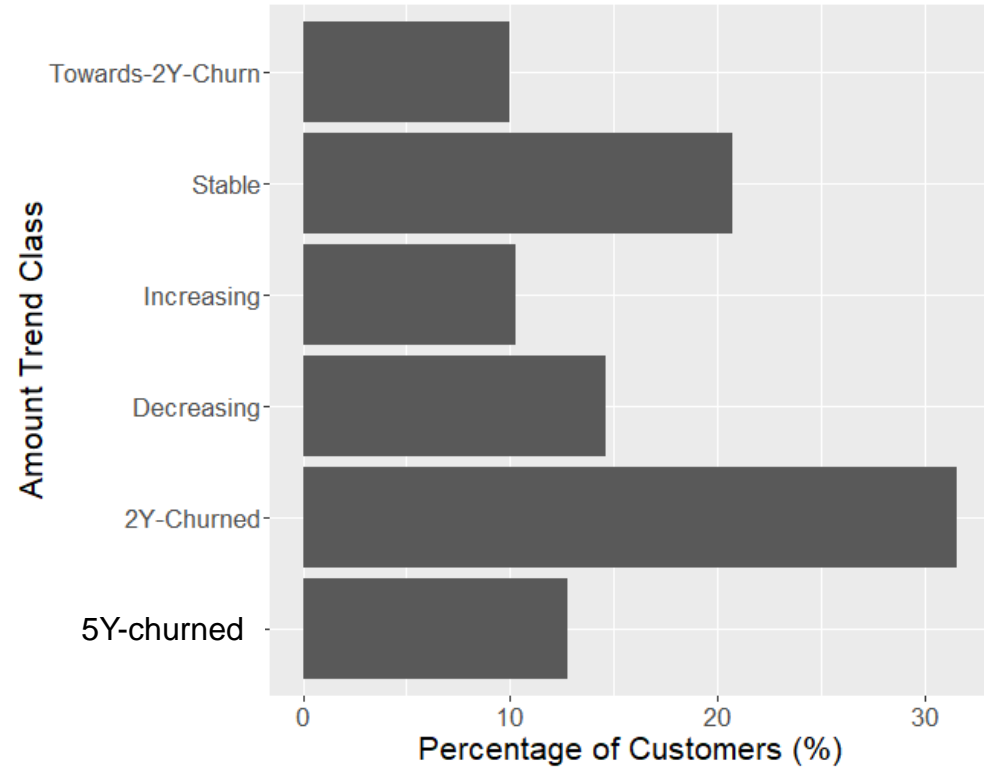
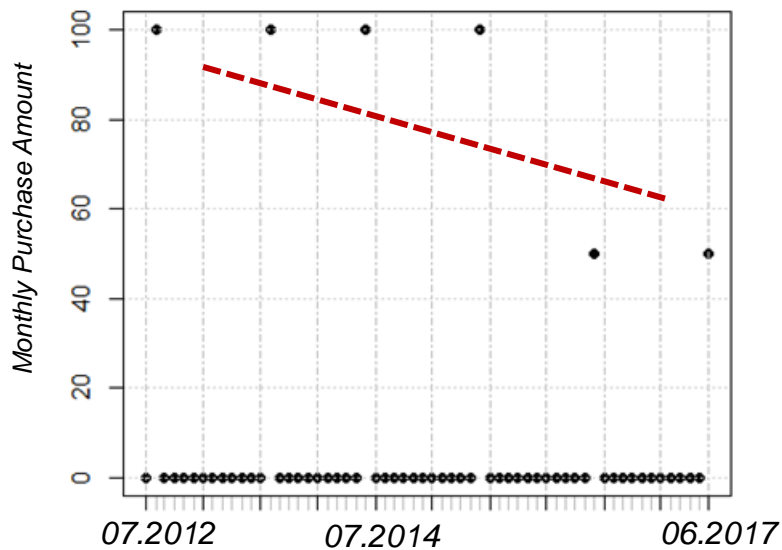


Periodicity Class	Customer Base (%)
-------------------	-------------------

Aperiodic	82%
-----------	-----

CUSTOMER FEATURES

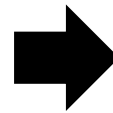
➤ Annual Amount Trend Class



Transactional
Features

CHANNEL PREFERENCE FEATURES

- Based on channel category



*Frequency of Mobile
Frequency of Download
Frequency of Online
Frequency of Diverse
Frequency of FALSE*

- Example:

Customer ID	Channel.Mobile	Channel.Online	Channel.*
07E58F094891FF5....	0.25	0.75	0



25% of times
via Mobile
Channel



75% of times
via Online
Channel

CUSTOMER ANALYTICS

➤ Customer Behavior Table

Customer ID	Feat-1	...			Feat-N
5D846CEF...	Transactional Features	Product Preference Features	Channel Preference Features	Calendar Preference Features	Marketing Preference Features
70CEA195...					
....					
826E1AE7...					

- ☐ Additional customer dimensions
- ☐ Contains DA in an “easy-to-use” form
- ☐ Basis for personalized marketing

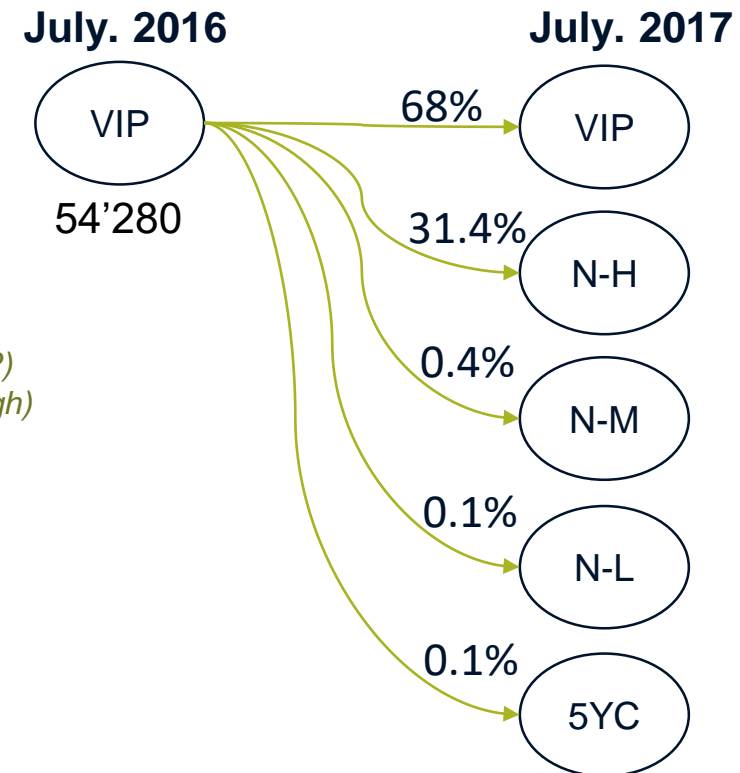
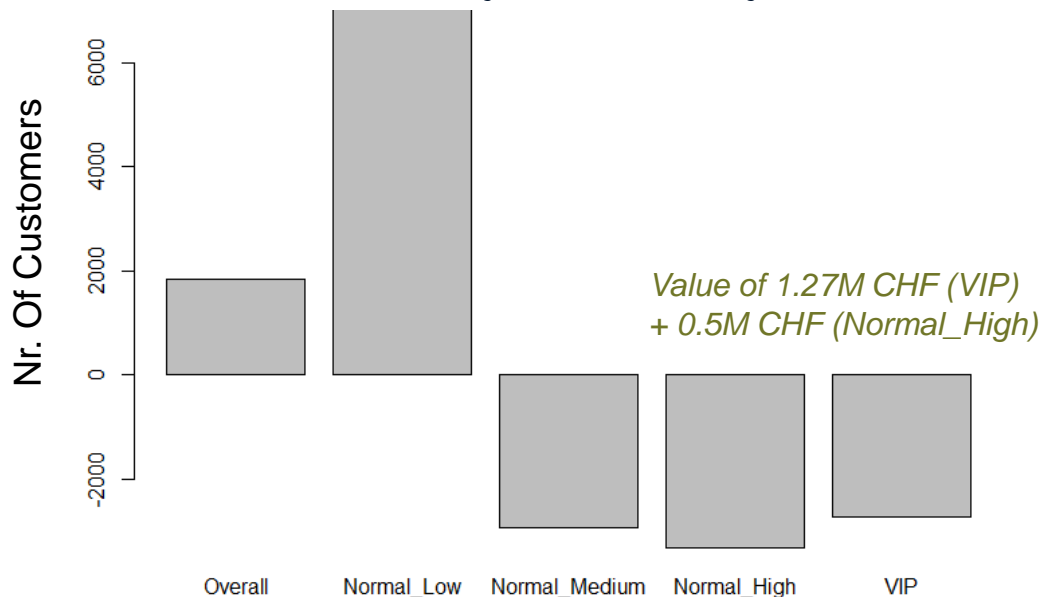
THE AFTER-"MATH"

➤ Immediate Value: Retention, Re-activation, Acquisition

Retention

- ❑ "VIP" & "Normal_High": **11%** of the customer base
- ❑ They make **~56%** of the total revenue

Difference between July. 2017 and July. 2016



THE AFTER-"MATH"

› A/B Testing

Empfänger:

VIP, Normal High

Testszenario wie folgt:

Testcases

A) 5'000 Adressen

B) 5'000 Adressen

C) 5'000 Adressen

Retention Cycle:

kein Rabatt

CHF 20.- Rabatt auf alle DPP's

20% Rabatt auf alle DPP's

Gültigkeit der Aktion: ca 4-6 Wochen. 14 Tage vor Ablauf der Gültigkeit wird nochmals ein Reminder an alle Nicht-Käufer versendet.

› Choice of customers based on:

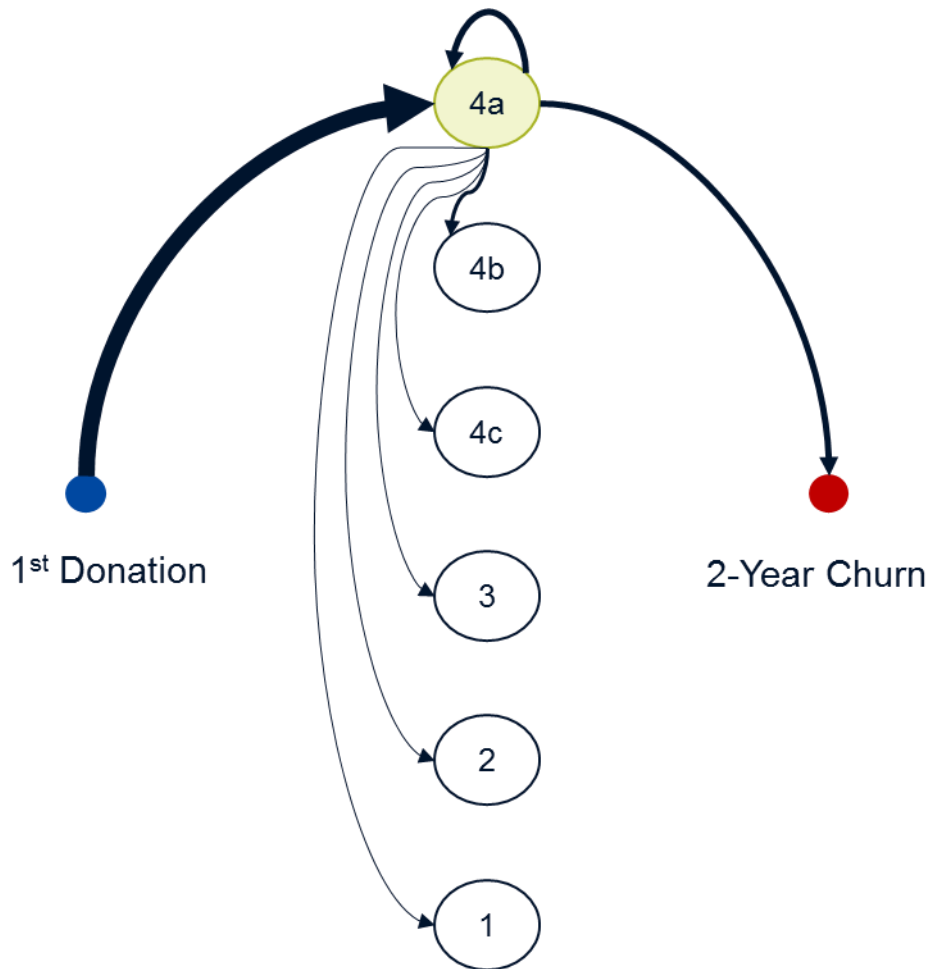
- › Amount.Trend = "Decreasing" or "Towards-Churn"

THE AFTER-"MATH"

➤ Flexible grouping of donors

ID	Features	Population (%)	Revenue (%)
4a	Amount_Group= Normal_low Nr. Of Reactions 5Y = 1	30%	5.5%
4b	Amount_Group= Normal_low Nr. Of Reactions 5Y > 1 Frequency = Low Periodicity = Aperiodic	33%	14%
4c	Amount_Group= Normal_Low Nr. Of Reactions 5Y > 1 Frequency = Medium, High Periodicity = weak, strong	22%	14%
3	Amount_Group = Normal_Medium	13%	28%
2	Amount_Group = Normal_High	2%	10%
1	Amount.Class = Special	1%	30%

➤ Monitoring of groups



ID	Features	Population	Revenue (%)
3p	Periodicity = Strong,Weak Frequency = Low	2.6%	5.4%

ID	Features	Population	Revenue (%)
3f	Frequency = Medium, High	7.2%	15 %

THE AFTER-"MATH"

➤ Optimize Marketing resources

Group	Donors (%) / Revenue (%)	Marketing Spending	PROPOSED ACTIONS
4a	30% / 5.5%	19%	<ul style="list-style-type: none"> • Less actions, but in proper timing • Mild Retention Process for new donors • Soft Retention Process for 2Y-churned donors
4b	33% / 14%	39%	<ul style="list-style-type: none"> • Less actions, but in proper timing • Mild Retention Process for new donors • Soft Retention Process for 2Y-churned donors
4c	22% / 14%	25%	<ul style="list-style-type: none"> • No changes • Full Retention Process
3	13% / 28%	14%	<ul style="list-style-type: none"> • Increase Marketing touchpoints • Full Retention Process • Fast Churn Monitoring
2	2% / 10%	2%	<ul style="list-style-type: none"> • Increase Marketing touchpoints • Full Retention Process • Fast Churn Monitoring / Personalized contact
1	1% / 30%	1%	<ul style="list-style-type: none"> • Increase Marketing touchpoints • Full Retention Process • Fast Churn Monitoring / Personalized contact

➤ Predictions and notifications



Prediction Status: "Delayed"



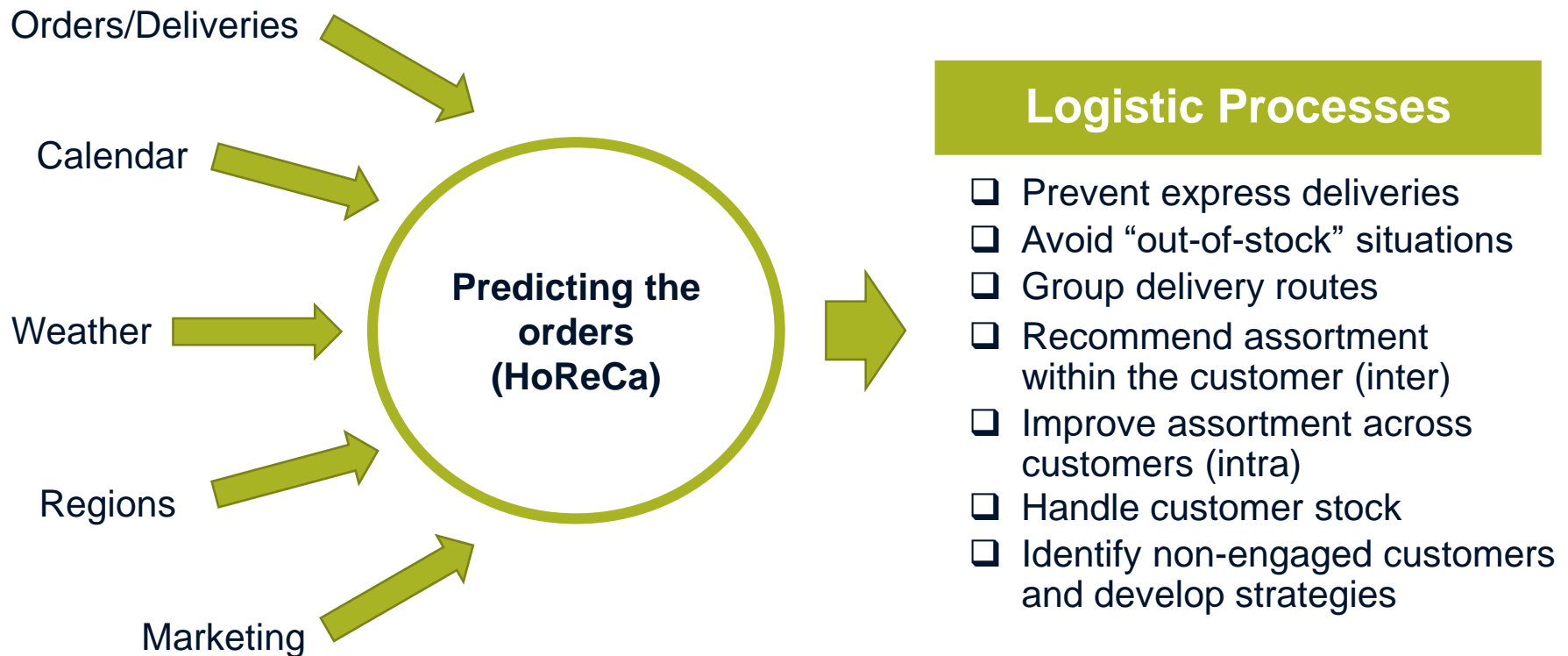
- Avoid churn
- Reduce forgotten donations

PREDICTIVE ANALYTICS



MOTIVATION: BUSINESS PERSPECTIVE

- Retailer of drinking products in **H**otels-**R**estaurants-**C**atering (HoReCa)
- Logistics Department



DESCRIPTION OF ANALYZED CUSTOMERS

► What type of analysis did we perform?

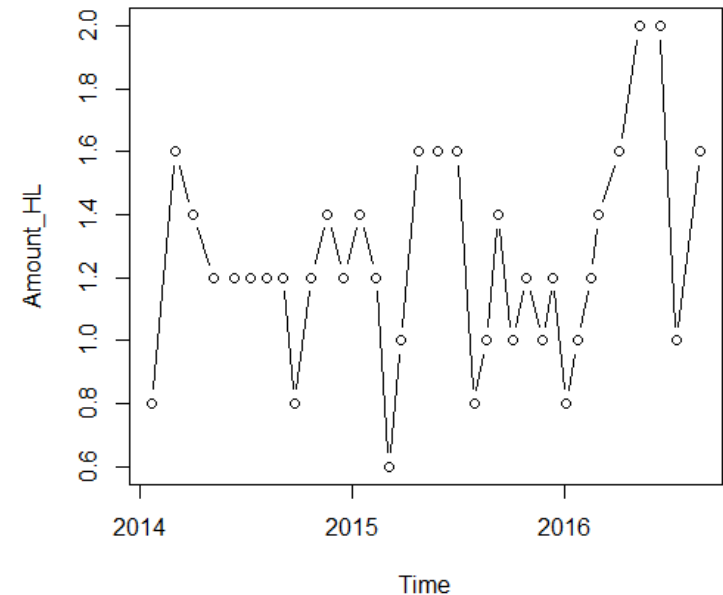
We analyzed
45'360 time-series

Example:

Customer ID =
77 294 025

- 43 deliveries in total
- 29 different products

Material	Nr. Of Deliveries
10041	35
10099	32
10152	29
10379	25
10476	24
10514	23
10601	19
10409	14
10153	12
10975	12
10606	10
10448	9
11542	8
11543	7
...	...



- Deliveries start from 2014-01-22
- Amount_HL in [0.6, 2] HL
- Time interval between consecutive orders in between 14 to 42 days

RESULTS

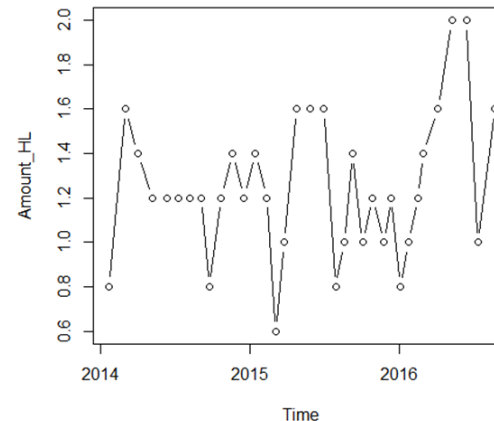
➤ Engaged and non-engaged customers

➤ Gr.1: 972 out of 1579 Customers (62%)

- Patterns exist for at least 1 product of these customers

- ✓ **88%** of the total deliveries

- ✓ **83%** of the total express-deliveries



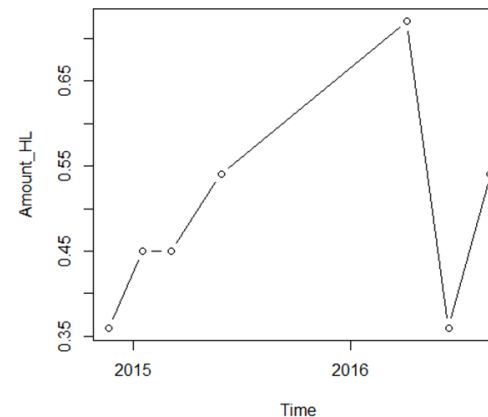
➤ Gr.2: 607 out of 1579 Customers (38%)

- Patterns do not exist (at least based on the history of deliveries)

- Example: Between all consecutive orders

Min Day Difference = 49

Max Day Difference = 313



RESULTS

➤ Engaged customers (Gr.1): Main Products - Concept



● ● Main Products: at least one of them is ordered almost always

● ● ● ● Secondary Products: ordered infrequently

RESULTS

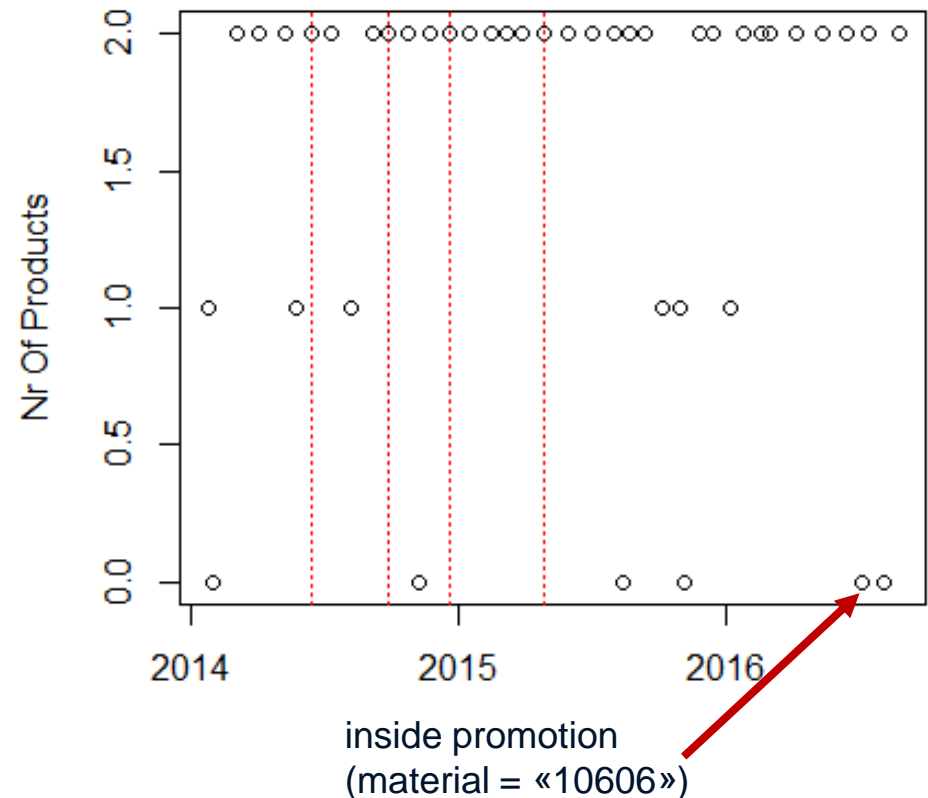
› Engaged customers (Gr.1): Main Products - Example

Example: Customer ID = 77 294 025

«main» products

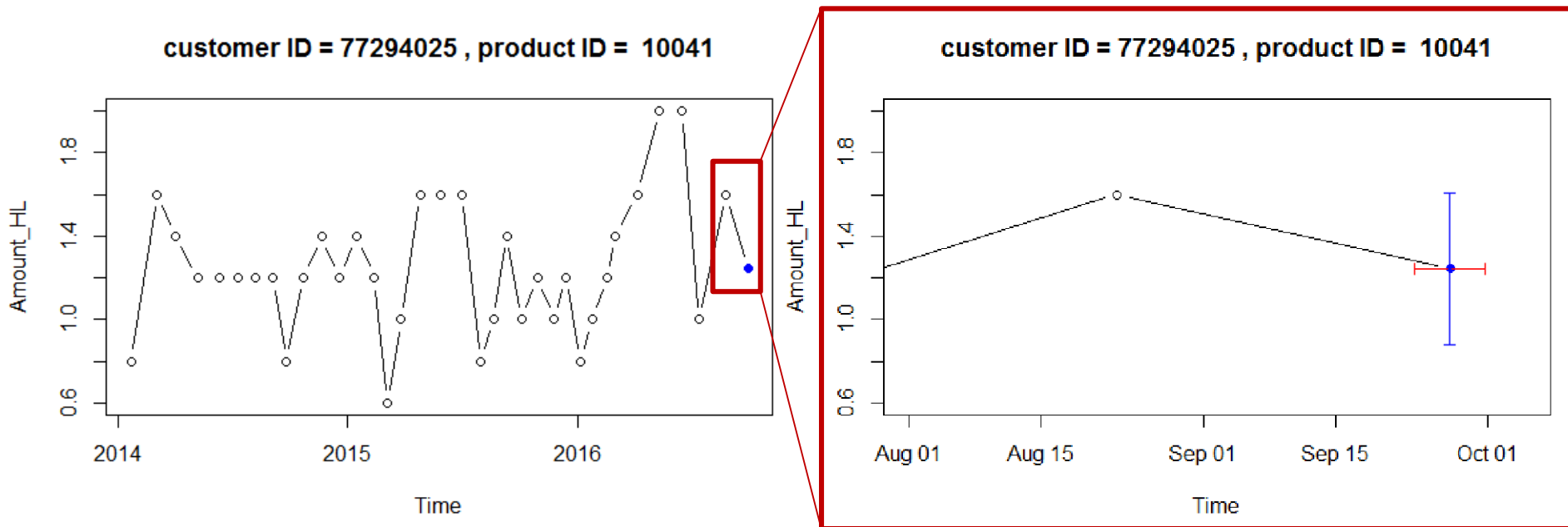
Material	Nr. Of Orders
10041	35
10099	32

- ✓ 37 out of 43 times (86%) at least one of the 2 «main» products was delivered
- ✓ 297 out of 307 product-deliveries (97%) happened in these 37 deliveries
- ✓ 4 out of 4 express-deliveries (100%) happened in these 37 deliveries (red dashed lines)



RESULTS

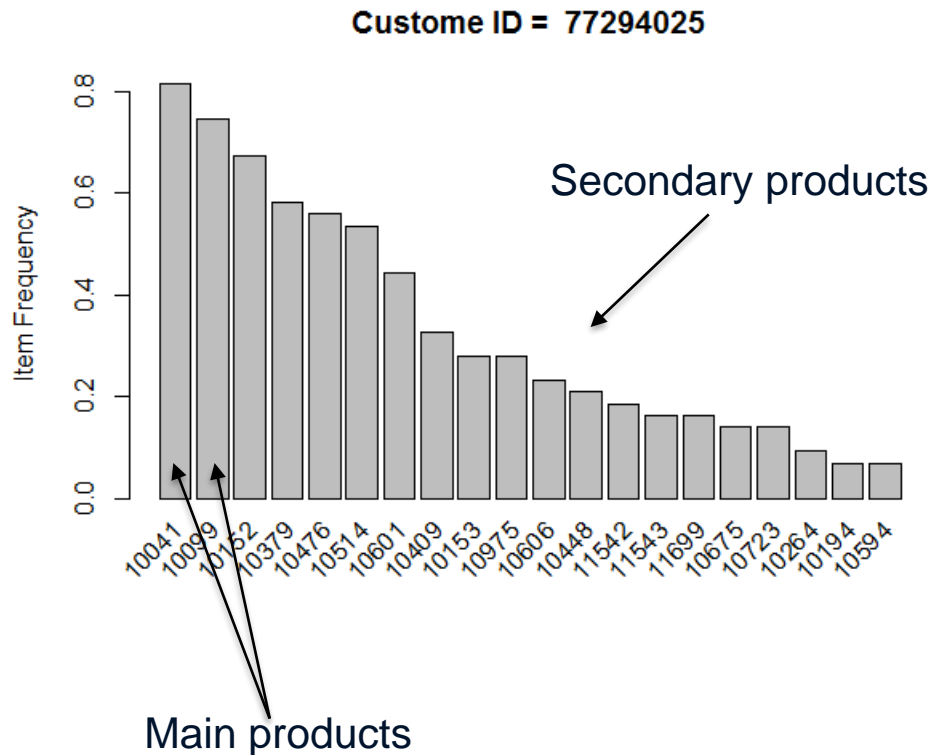
Engaged customers (Gr.1): Main Products - Predictions



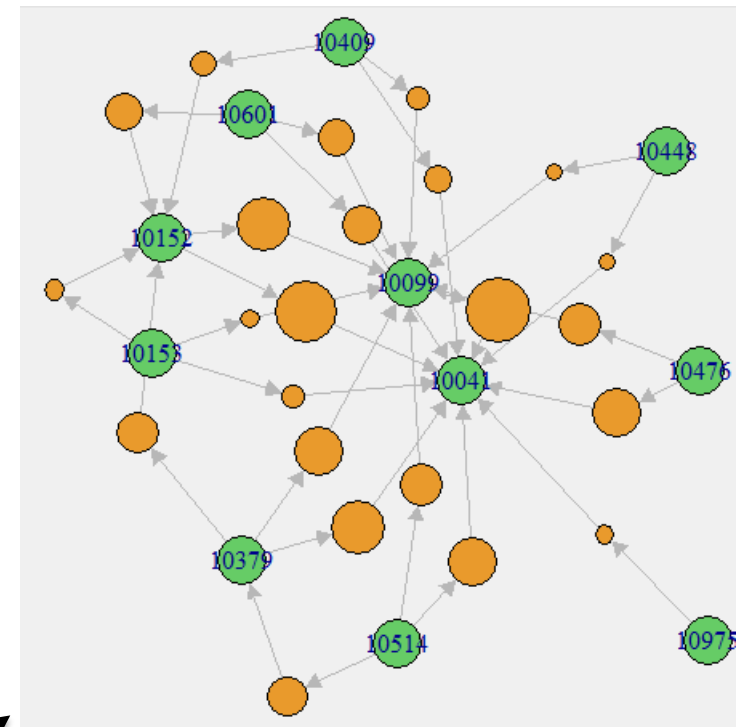
- For every prediction 2 confidence intervals are computed:
 1. Time of delivery (red line)
 2. Amount of product delivered (blue line)

RESULTS

Engaged customers (Gr.1): Secondary products



Association Graph of products



Can be used for product recommendations

RESULTS

‣ Engaged customers (Gr.1): Overall Strategy & Benefits

- ✓ For 62% of the customers (972 out of 1579) we can make reliable predictions
- ✓ 88% of the all deliveries , and 83% of all the express deliveries are made to the Gr.1 customers
- ✓ For the Gr.1 customers, deliveries can be predicted based on the «main» product(s)
 - Majority of total deliveries can be predicted
 - Majority of express deliveries can be avoided
 - Change the order-behavior via recommendations for known «secondary» products (reduce single orders of «secondary» products)

GEO-ANALYTICS



GEO-DATA IN SWITZERLAND

➤ BFS as a source of information

Regional Level

**State
(Kanton)**

**District
(Bezirk)**

**Community
(Gemeinde)**

**Hectare
(100m x 100m)**



Features / Attributes

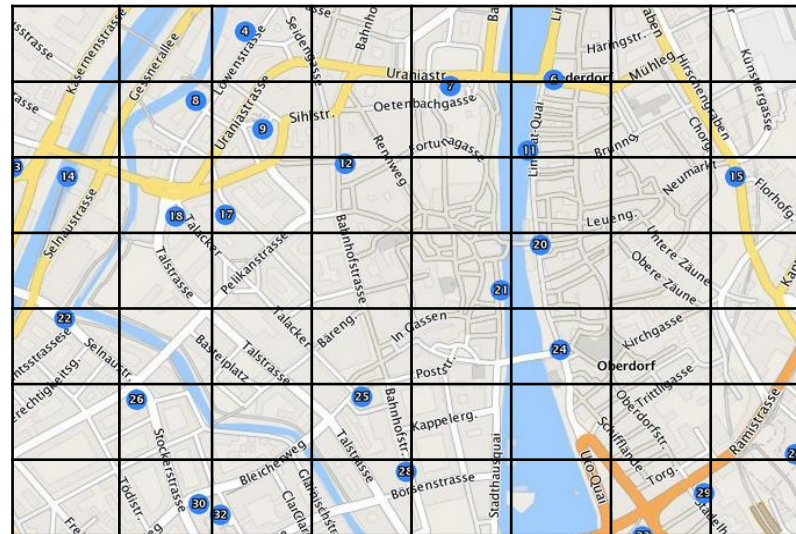
- Age groups
- Housing Information
- Country of origin
- Marital status
- Language
- Work sectors
- Family and kids
- Religion
- Political opinions
- Commuting
- Urbanization
- Educational level
- Income
- etc

GEO-DATA IN SWITZERLAND

➤ BFS as a source of information

Regional Level

**Hectare
(100m x 100m)**



Features / Attributes

- Age groups
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GEO-DATA IN SWITZERLAND

➤ BFS as a source of information

Regional Level

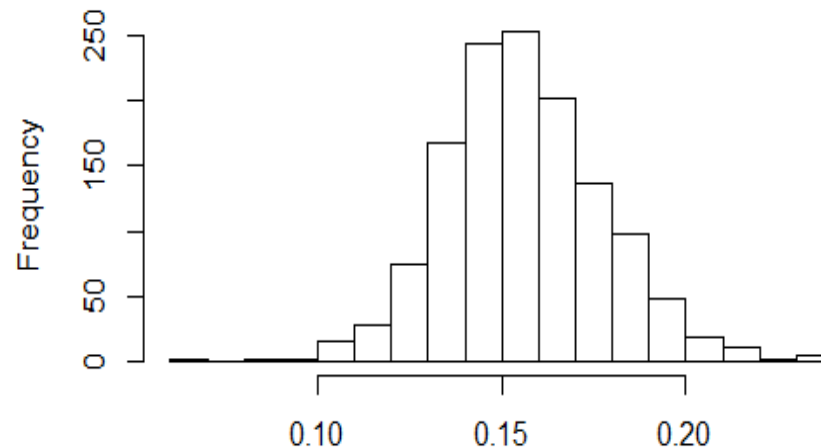
State
(Kanton)

District
(Bezirk)

Community
(Gemeinde)

Hectare
(100m x 100m)

*Quota of persons of age 0-14
across political communities in CH*



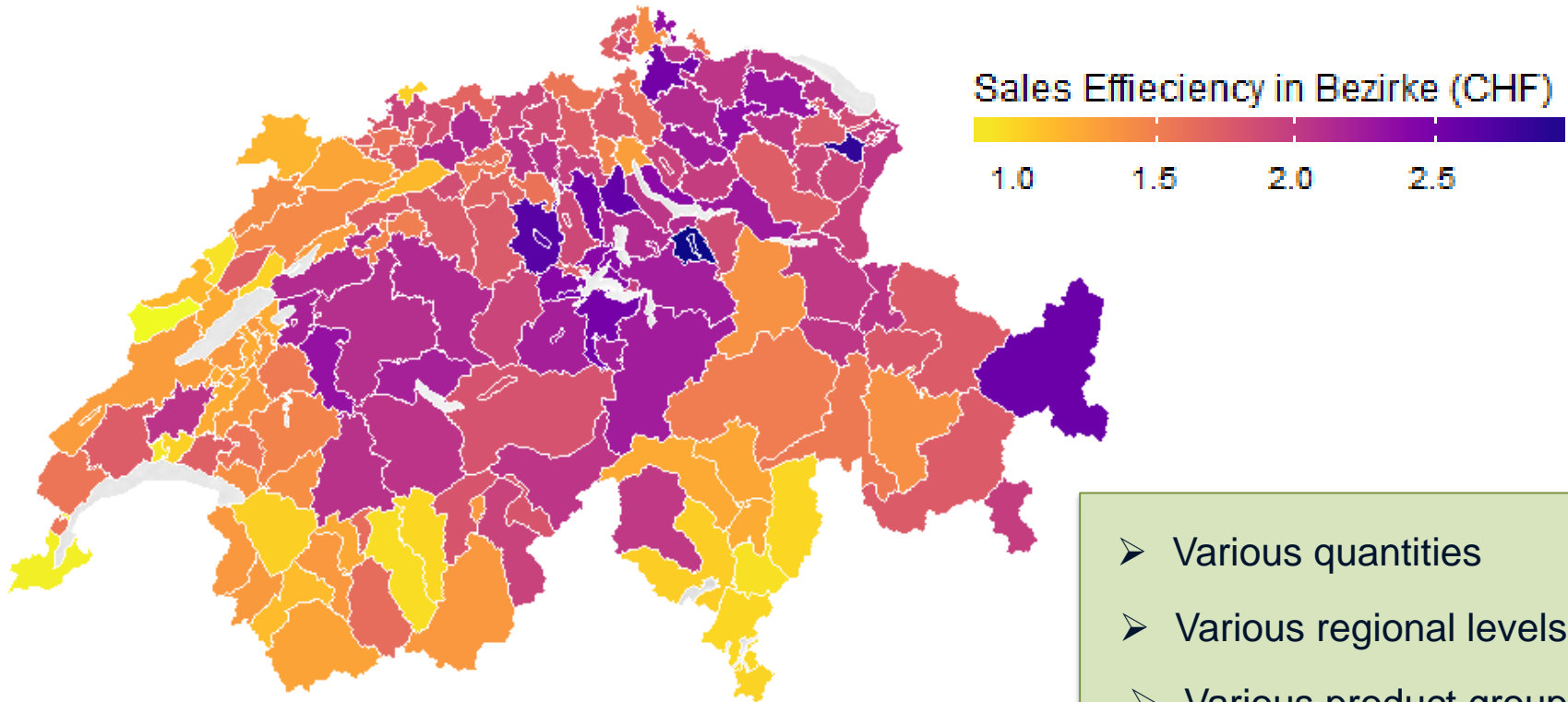
Diversity is the key to informativeness!

Features / Attributes

- **Age groups**
 - 0-4
 - 5-9
 - 10-14
 - 15-19
 - 20-24
 - ...
 - 89 and more
- Housing Information
- Country of origin
- Marital status
- Language
- Work sectors
- etc

GEO-ANALYTICS: RESULTS

➤ Quantification sales efficiency at the regional level

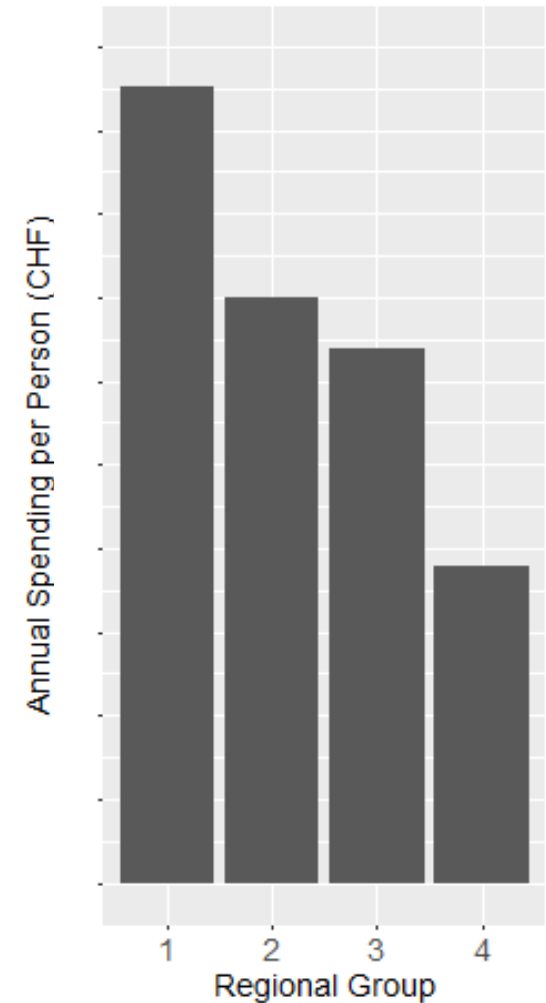
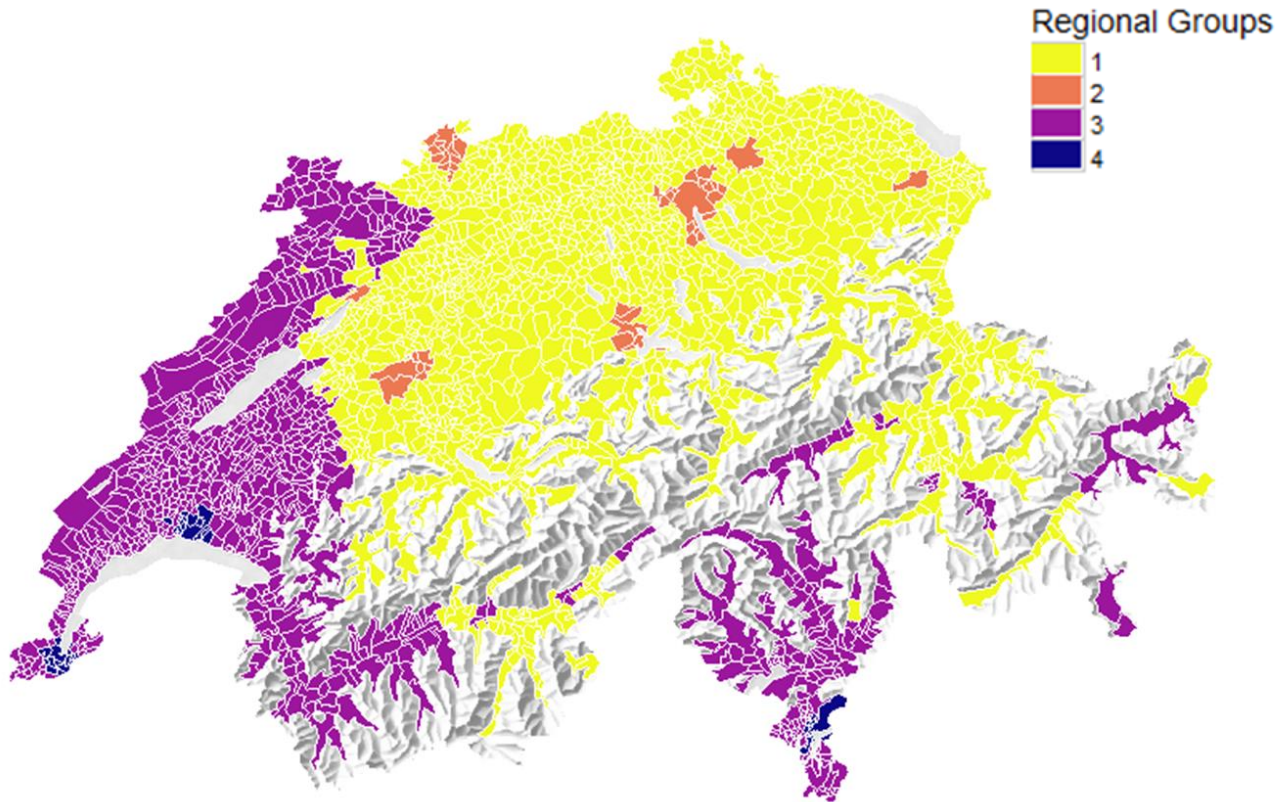


*Annual Revenue Efficiency (= Annual Revenue / Population)
of a product-group for all Bezirke in 2016*

- Various quantities
- Various regional levels
- Various product groups

GEO-ANALYTICS: RESULTS

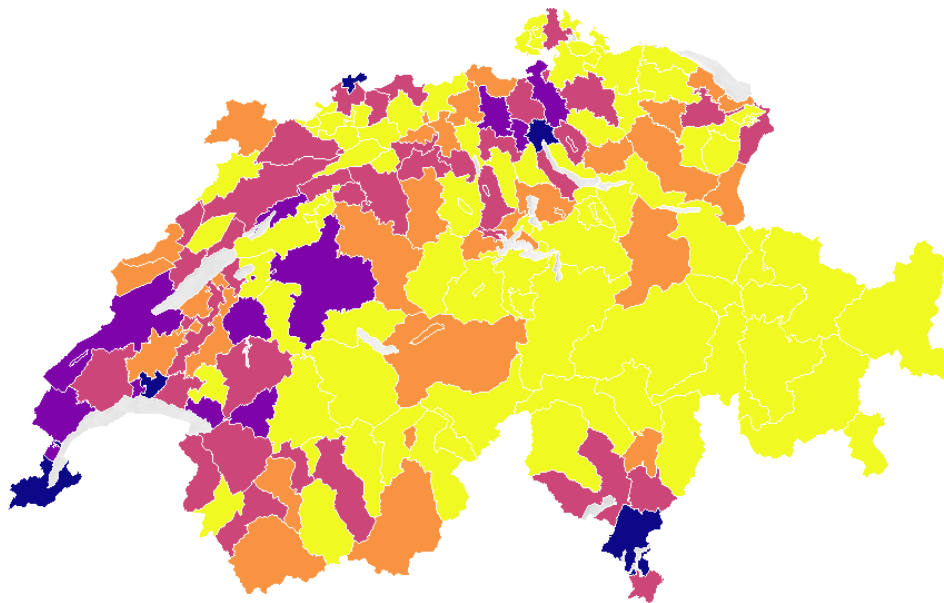
➤ Regional grouping



GEO-ANALYTICS: RESULTS

➤ Annual revenue potential

➤ Which regions should be targeted first?



Revenue Potential (CHF)



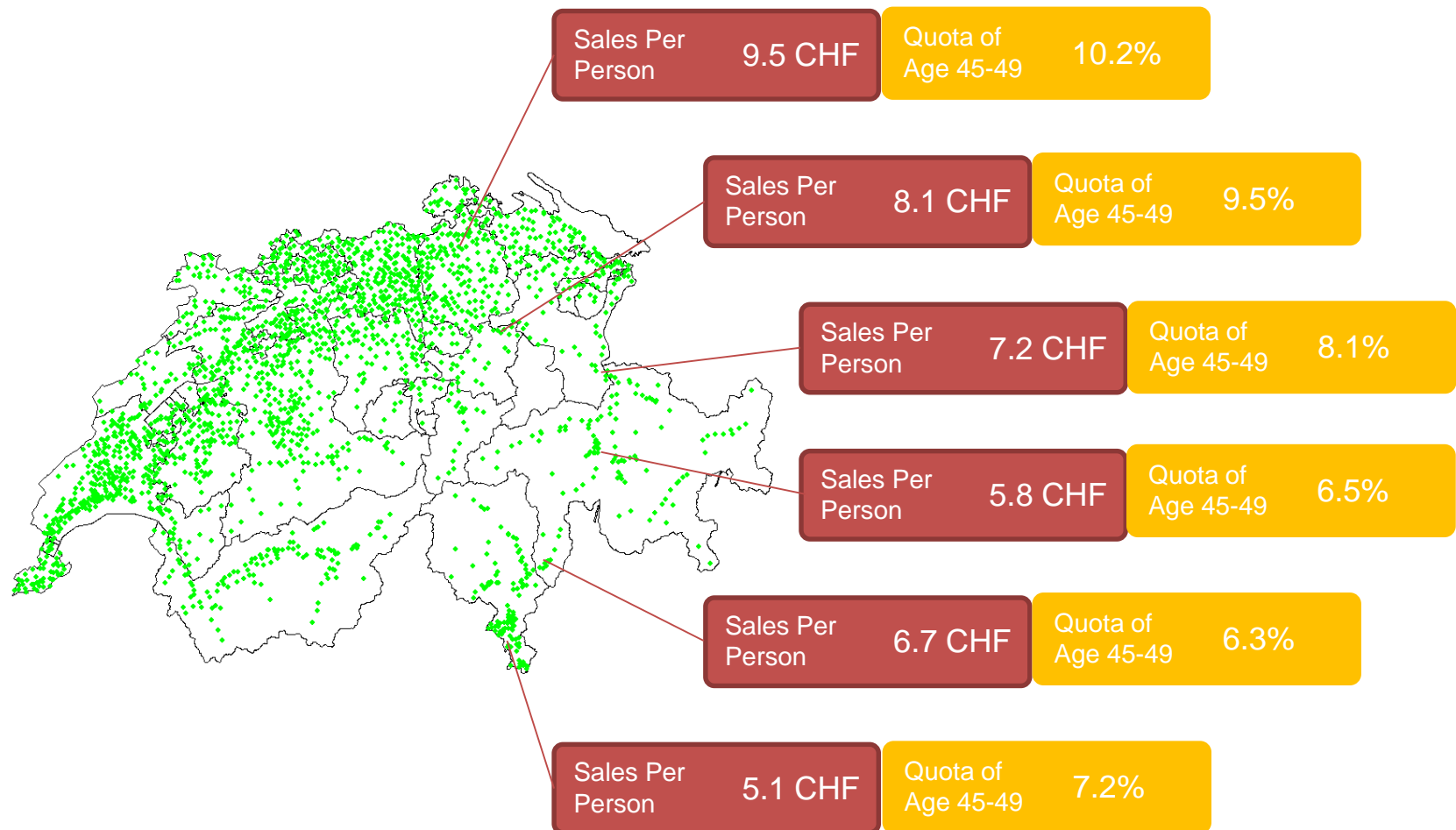
Annual revenue potential of a product-group at the Bezirk-level

KANTON_NAME	Revenue Potential (CHF)	Cumulative Percentage (%)
Vaud	5.69*X	14.2
Genève	5.58*X	28.1
Zürich	5.39*X	41.6
Ticino	3.47*X	50.3
...

Annual revenue potential of a product-group at the Kanton level

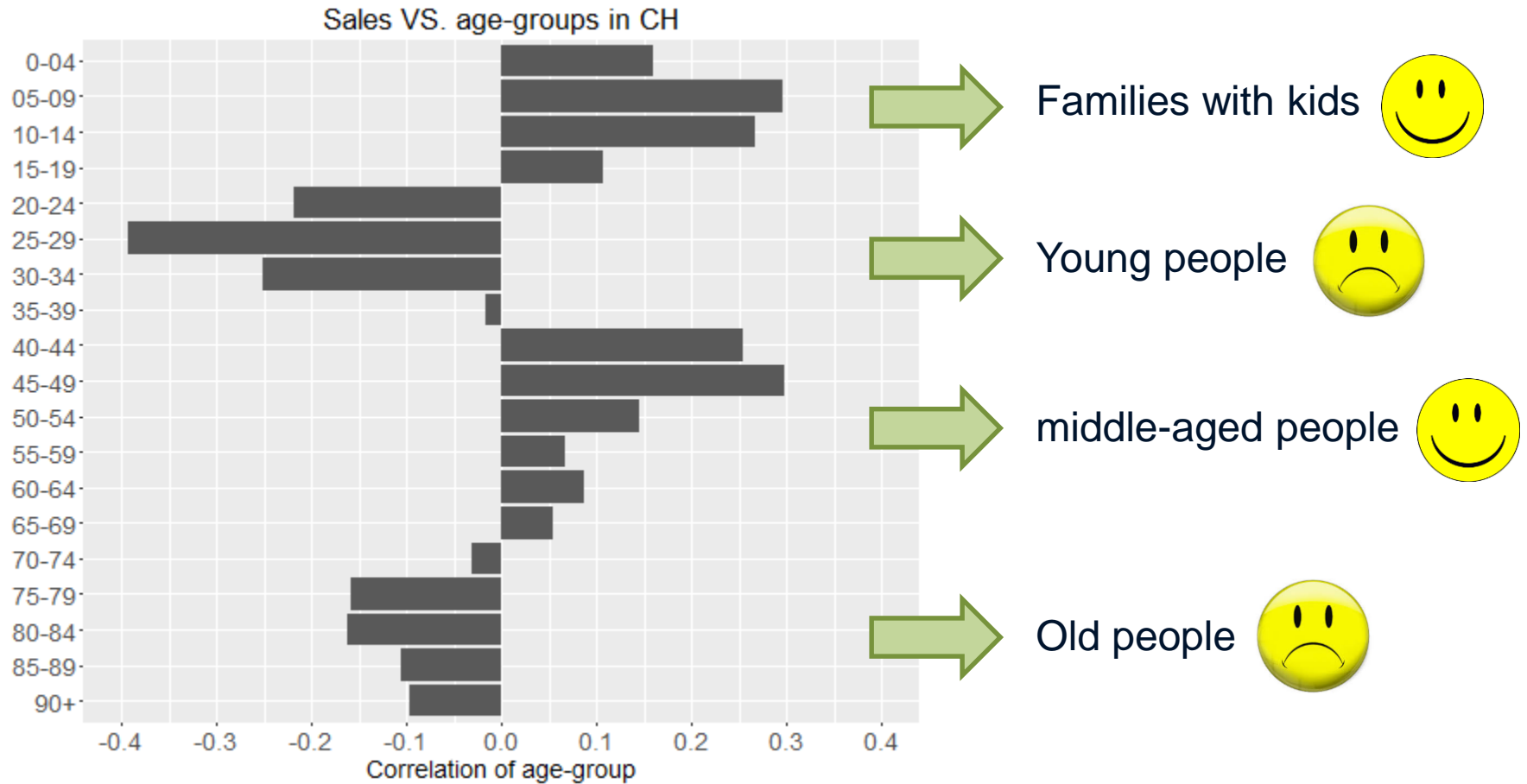
GEO-ANALYTICS: RESULTS

► Analysis for age-groups: exploit regional diversity



GEO-ANALYTICS: RESULTS

► Sales Efficiency for age-groups



OTHERS



OTHERS

- Requirements Engineering with Data Science
- Fraud detection (mobilezone)
- Predicting defaulting customers (card center of major swiss bank)
- Optimization of campaigns (Young Swiss)
- Infer customer segments behind products (B2B client)

Wir freuen uns auf angeregte Gespräche mit Ihnen

➤ Dr. Sotiris Dimopoulos
Senior Data Science Consultant

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 <http://blog.it-logix.ch>