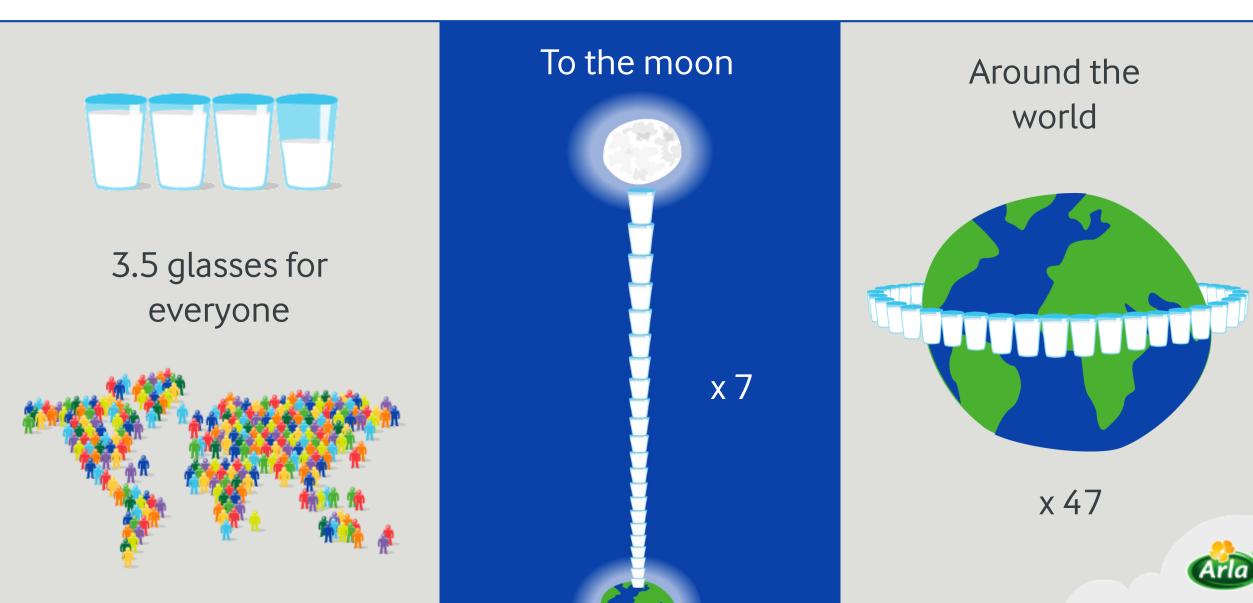


# About Arla





# We are the **4th largest dairy** company in the world based on milk intake



We provide dairy that nourishes body and soul throughout life













# We take responsibility

- throughout the entire value chain





# **Building internal BI capacity at Arla**

Why is it important to build internal capacity?

Finding the balance between external and internal capacity is key

#### External capacity

- Flexibility in the workforce
- Easier to find specialist skills
- Reduced risk
- Very fast to start up a new project

#### Internal capacity

- The knowledge remains with Arla
- More cost efficient
- More ownership over the solutions we build
- Stronger connection with our values
- Better understanding of our mission, vision and industry



# **Building internal BI capacity at Arla**

What kind of capacity is being built?

### Finding the balance between business and IT skills is key

#### Skills

- SAP
- SAP BW
- Power BI
- R
- Python
- .net
- SQL

#### Roles

- Subject Matter Expert (Business)
- Scrum Master (Business)
- Product Owner (Business)
- Data scientist (Business / IT)
- Front end Developer (IT)
- Back end developer (IT)
- Data engineer (IT)
- Architect (IT)



# **Building internal BI capacity at Arla**

# How is Arla building and attracting the internal capacity?

#### Scale

- The Arla valuechain is particularly wide offering a wide range of opportunities and complexity in BI
- Provides opportunities for the individual employee

#### Agility

- Arla is Agile resulting in fast turn around time for new solutions and features
- Focus on delivering value

#### Insight and foresight

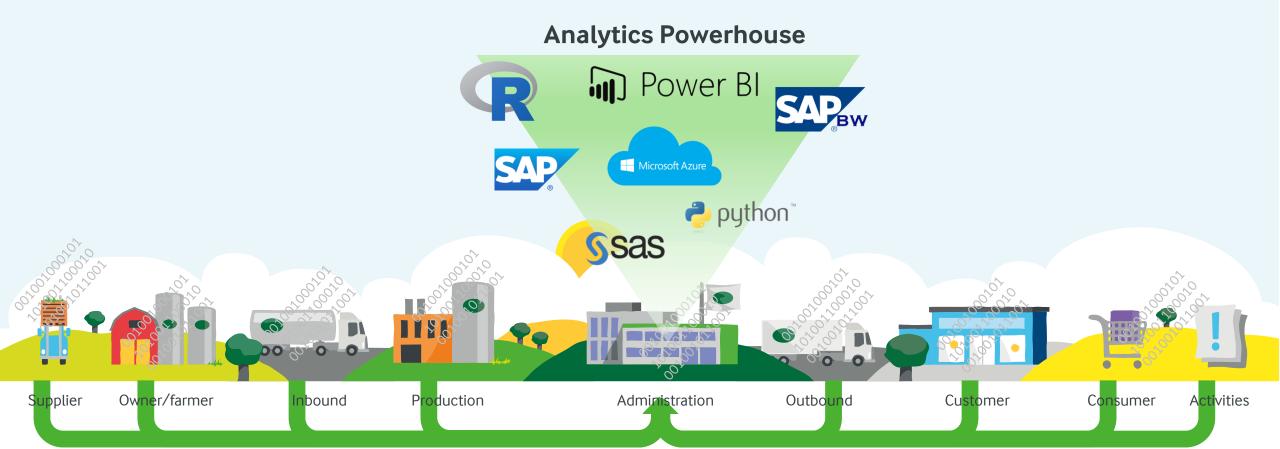
- Arla is more than traditional BI, it is also data science and analytics using state of the art tools and methods
- BI becomes part of running the business

#### Analytics Powerhouse

- Centre of competency for BI and Analytics at Arla
- Resources in both Denmark and Poland



# The **Analytics Powerhouse** use data from across the valuechain to provide Arla with insight and foresight

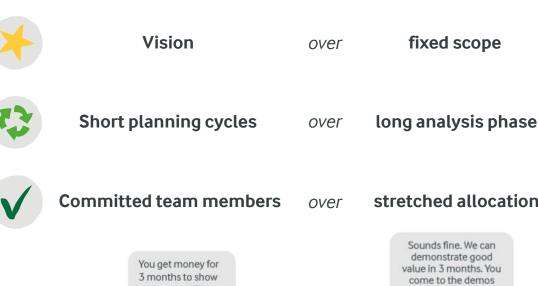




# Arla is agile!

## Time to build new technological capacity is significantly reduced

- Arla IT Solutions has adopted Agile with BI and Analytics being the first 18 months ago
- Delivery teams are cross functional with a mix of people from IT and business and a mix of internal and external resources
- Time to first User Go Live significantly reduced







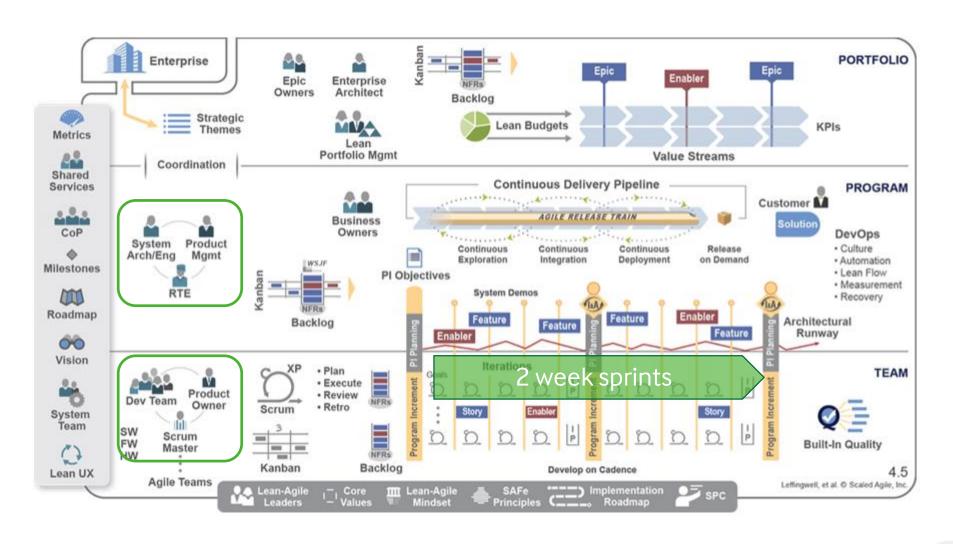




# Arla is agile!

# Analytics Powerhouse is a delivery train







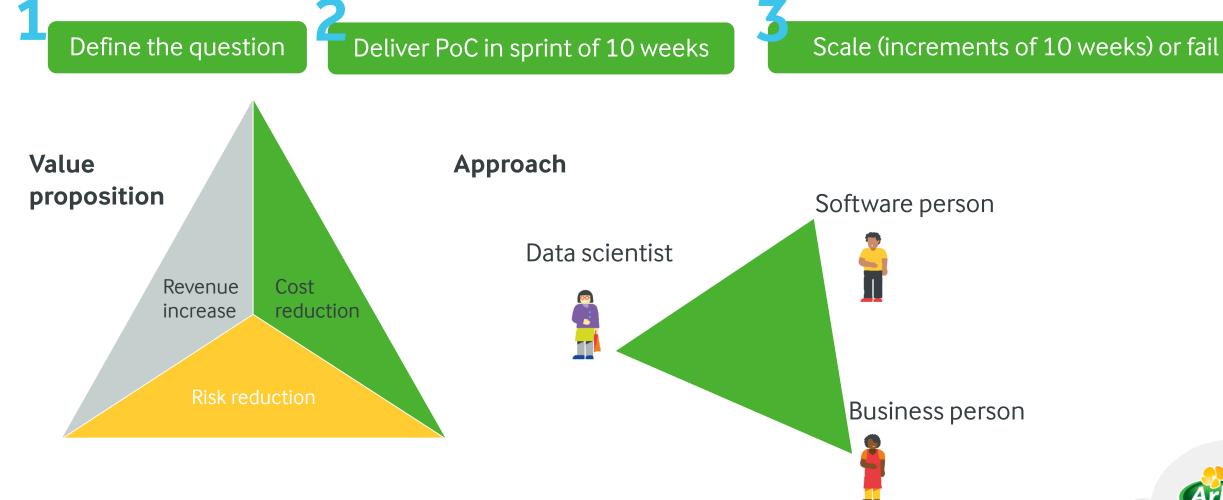
# Building internal BI capacity at Arla Specific BI examples delivered by the Analytics Powerhouse

- 1. Optimising trade promotion spend by analysing impact on sales
- 2. Integrate & automate data for Stock Keeping Unit optimization & complexity reduction
- 3. Improving milk intake forecasting accuracy using machine learning learning
- 4. Consumer segmentation using analytics on online behaviour



# Let your data innovate!

- And work with the Analytics Powerhouse to make it happen





# Something about data

#### The four V's

- **Volume:** "data at rest", i.e. the amount of data.
- Variety: "data in many forms", i.e. different types of data (e.g. structured, semi-structured and unstructured, e.g. text, web or multimedia data), data sources (e.g. internal, external, public) and data resolutions.
- **Velocity:** "data in motion", i.e. the speed by which data are generated and need to be handled.
- **Veracity:** "data in doubt", i.e. the varying levels of noise and processing errors.

Big data makes new types of analytics possible

Big data forces us to

 change the way we
 collect, store, manage,
 analyze and visualize
 data.



#### Optimising trade promotion spend by analysing impact on sales

#### What was done?



#### **Situation**

 The promotion planners were missing insight into ROI on the trade promotion. This was preventing them from learning about the impacts of promotions.

#### What was done

 Using PoS data to automate the calculation of the campaign impacts, by comparing numbers on regular weeks to the ones where promotions were planned.

# Technology Big Data Dimensions Volume Velocity Analytics method Descriptive analytics Veracity Veracity

#### **Data sources**

- Promotion planning data
- AC Nielsen
  - Contains SKU sales data on a weekly level.
- Customer codebook
  - Is used when promotions are on sub-chain level, it is used for calculating approximately sub chain sales from the Nielsen data.

Customer

Value area Org & value chain



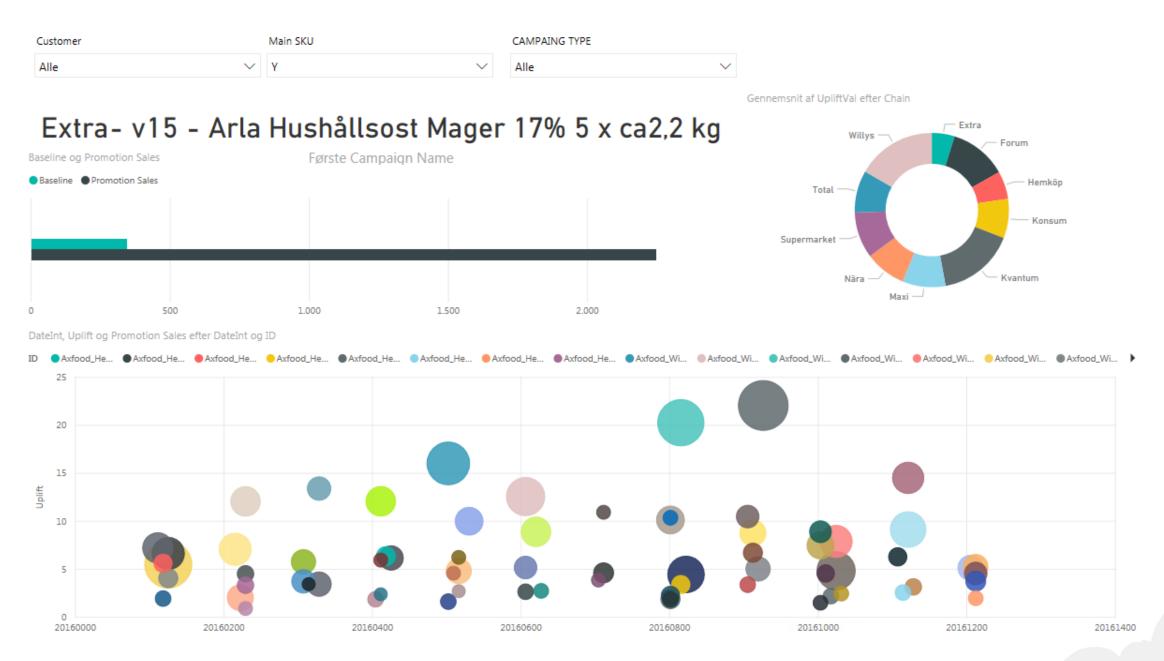
Revenue growth



Commercial Europe



#### Optimising trade promotion spend by analysing impact on sales





#### Integrate & automate data for SKU optimization & complexity reduction

#### What was done



#### **Situation**

- SKU Management Process is time consuming more than 6 months.
- Decisions on SKU refinements are based on a data dump copied into more than 50 different Excel sheets (1 per dairy)
- Data is outdated by the time it is ready for decision making and often does not include all parameters relevant for decision making

#### What was done

- A user dashboard in PowerBI based on 3 sites was created.
- The portal shows the daily updated status of SKUs, from multiple perspectives, with complexity as well as commercial analysis and automated reports.
- This was done to enable better monitoring and optimization of product portfolio.
- The data foundation for future commercial attractiveness and complexity analysis and understanding are built into HANA.

#### Technology





**Analytics method**Descriptive analytics

#### **Big Data Dimensions**

Volume

Velocity

Variety

Veracity

#### **Data sources**

- Production Volumes
- Sales data
- Bill of Material and costing data
- Global Master Data catalogue
- Financial Forecast data
- Custom data sets

#### Value area

## Org & value chain







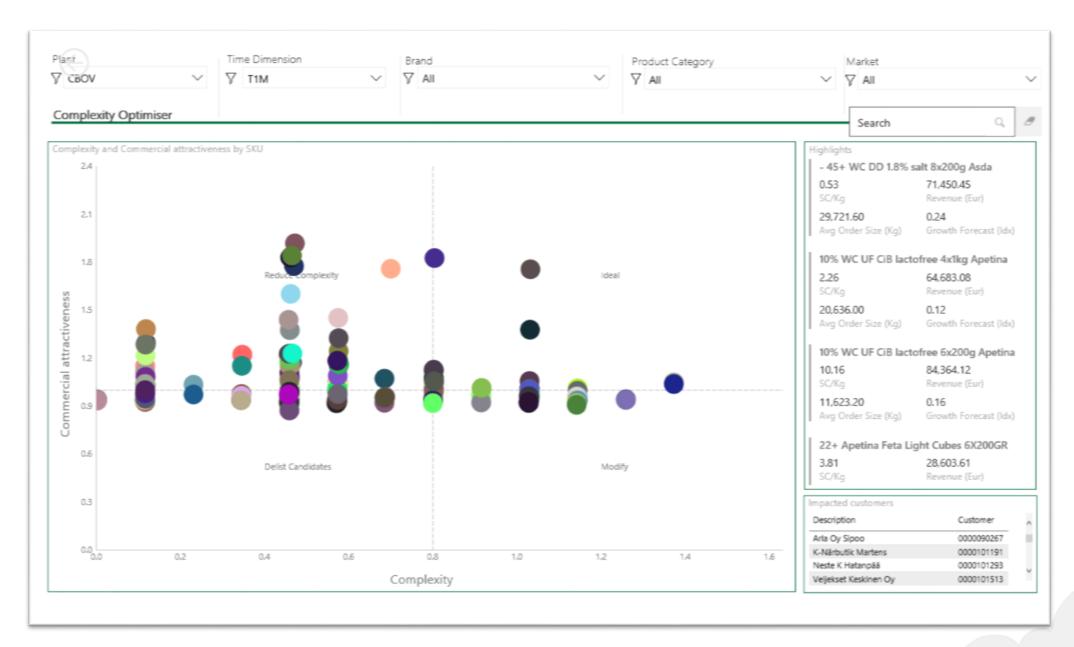




Supply Chain



#### Integrate & automate data for SKU optimization & complexity reduction





#### Improving milk intake forecasting accuracy using machine learning

#### What was done?



#### **Situation**

- Currently, there is no unified approach for constructing the milk intake forecasts across markets.
- The data foundation for the current approach is very limited and the use of Excel makes is very time consuming.

#### What was done

- A forecasting methodology with underlying clear drivers and assumptions was developed and visualized as an interactive dashboard.
- Machine learning techniques made it possible to build on a much richer data foundation, thus improving accuracy.

#### **Technology**





( Al/Machine learning

**Analytics method** 

Penalized regression

#### **Big Data Dimensions**

Volume

Velocity

Variety

Veracity

#### **Data sources**

- Daily milk intake data from all farmers
- National herd data from DK:
  - No. of cows
  - Delivery percent
- Arla on-account milk price
- Feed price (SEGES)

#### Value area

#### Org & value chain





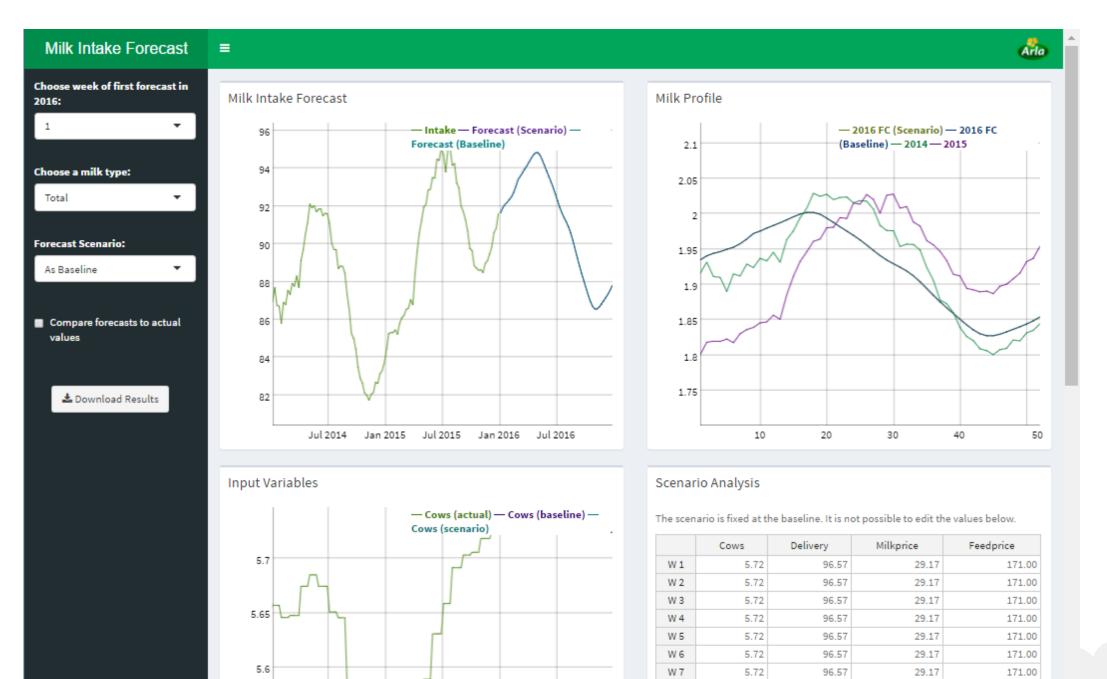


Millk. Members & Trading





#### Improving milk intake forecasting accuracy using machine learning





#### Consumer segmentation using analytics on online behaviour

#### What was done?



#### **Situation**

Arla owns webpage data that reveals consumer behaviour and measures digital engagement scores. This can be used for precision marketing.

#### What was done

- An investigation of the feasibility of using Arla's web data for customer segmentation was conducted.
- A preliminary cluster analysis of recipes using the web data was conducted in order to identify clusters of recipes that consumers are interested in.

#### **Technology**



Big data & analytics

**Analytics method** Clustering

#### **Big Data Dimensions**

Volume

Velocity

Variety

Veracity

#### Data sources

- Traffic data from Arla websites.
- Usage data from "Karolines Køkken" app.
- Recipe data from "Karolines Køkken".

Value area

Org & value chain



aains



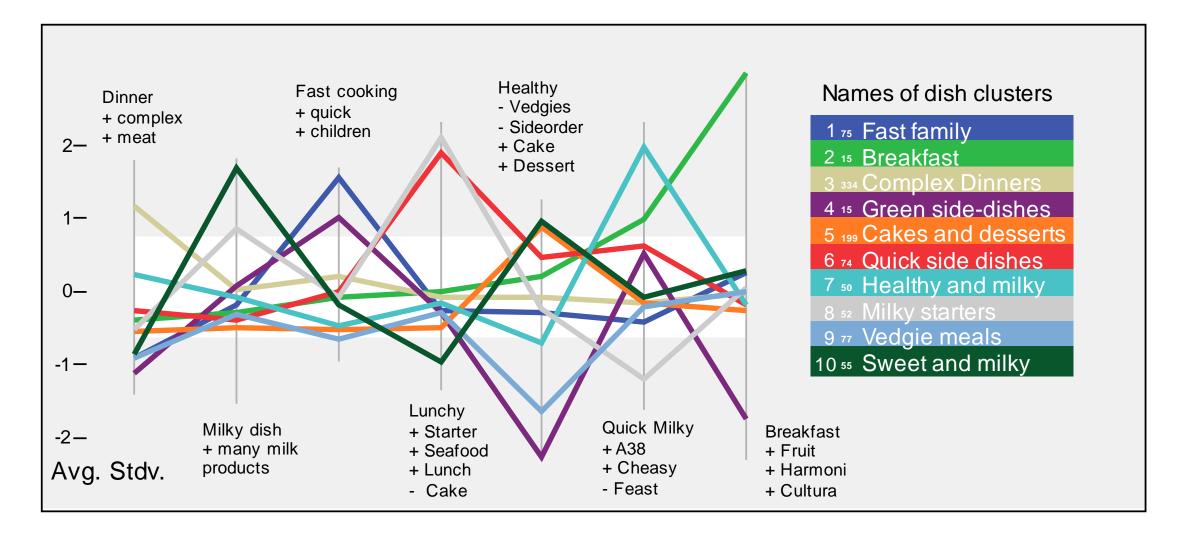
growth







#### Consumer segmentation using analytics on online behaviour





# Thank you!









