

## Lifecycle Triggers and Personalised Recommendations

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# sy INTRO



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RS Components



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since 1937

6000+ employees

32 locations

500k+ products

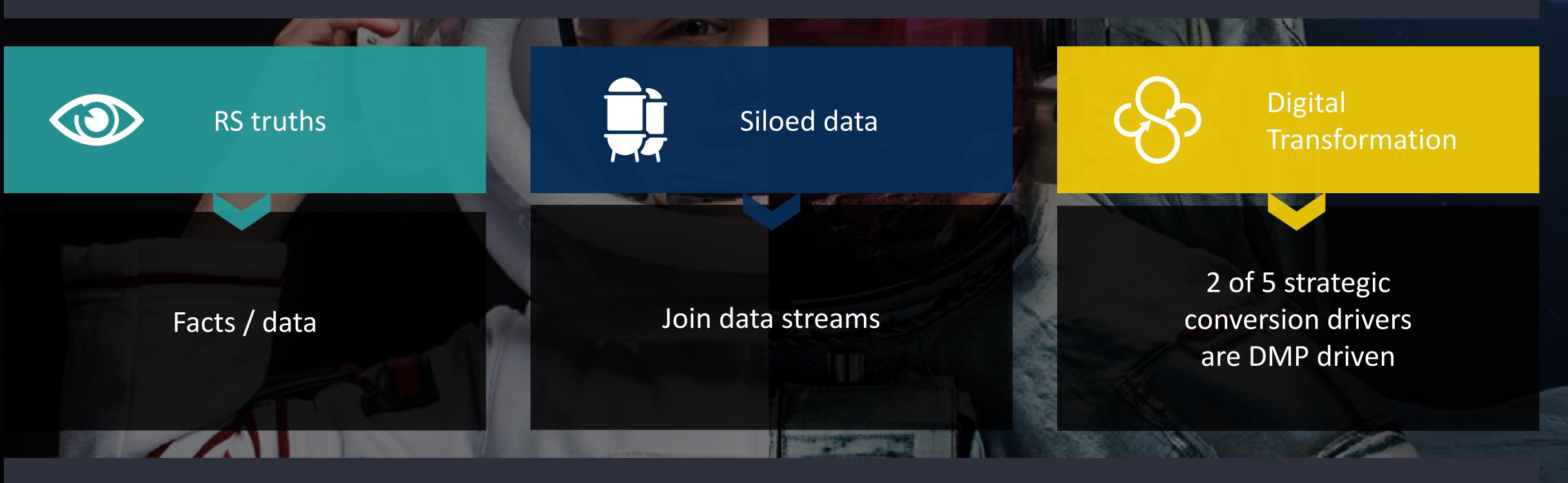
70% digital sales

100m visits

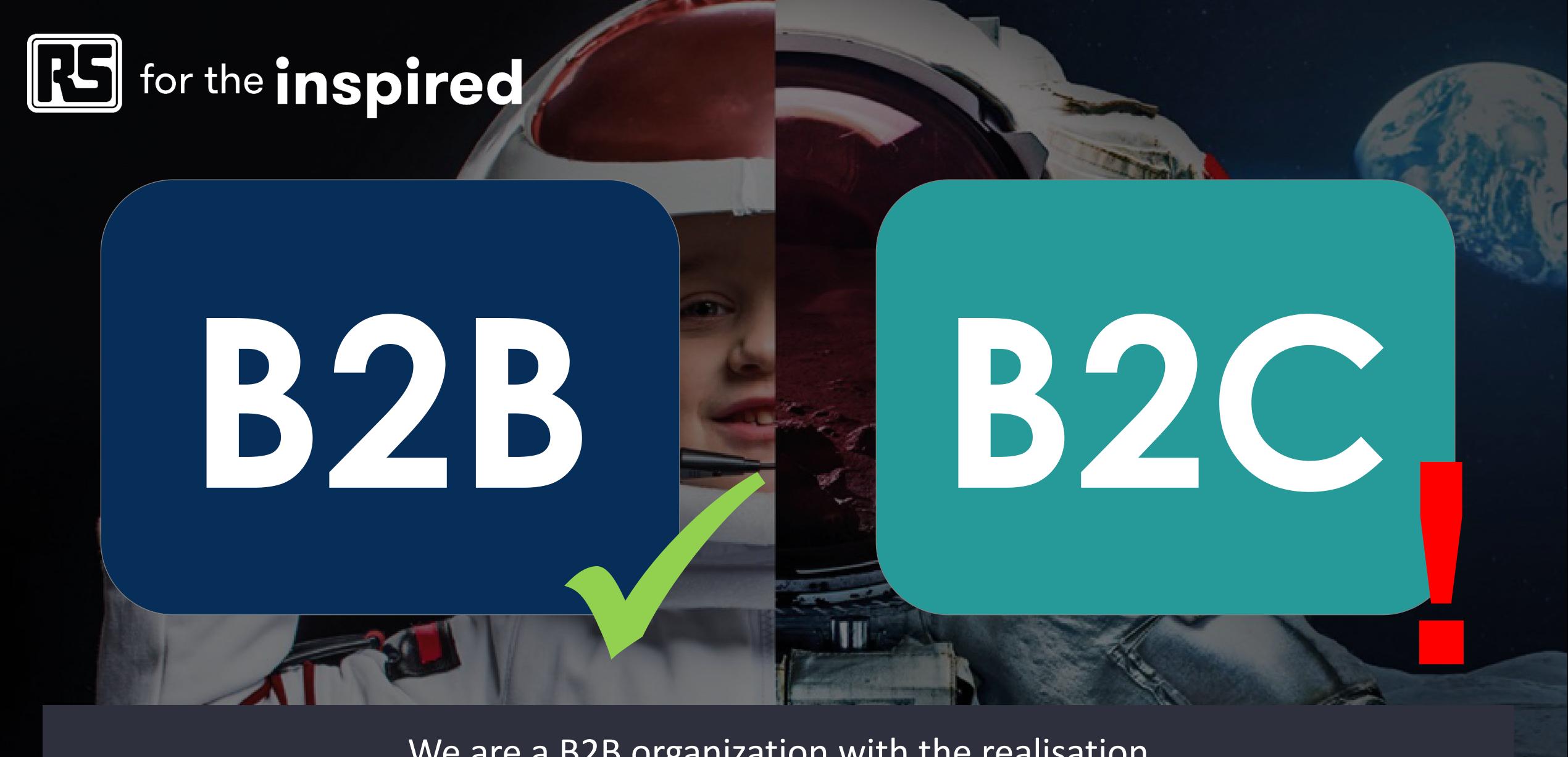
£1bn digital sales



### WHY IS DATA SCIENCE IMPORTANT TO RS COMPONENTS?



DATA SCIENCE DOESN'T HAVE TO BE COMPLICATED



We are a B2B organization with the realisation that we need to act as a B2C business



## ST CHALLENGE STATEMENT

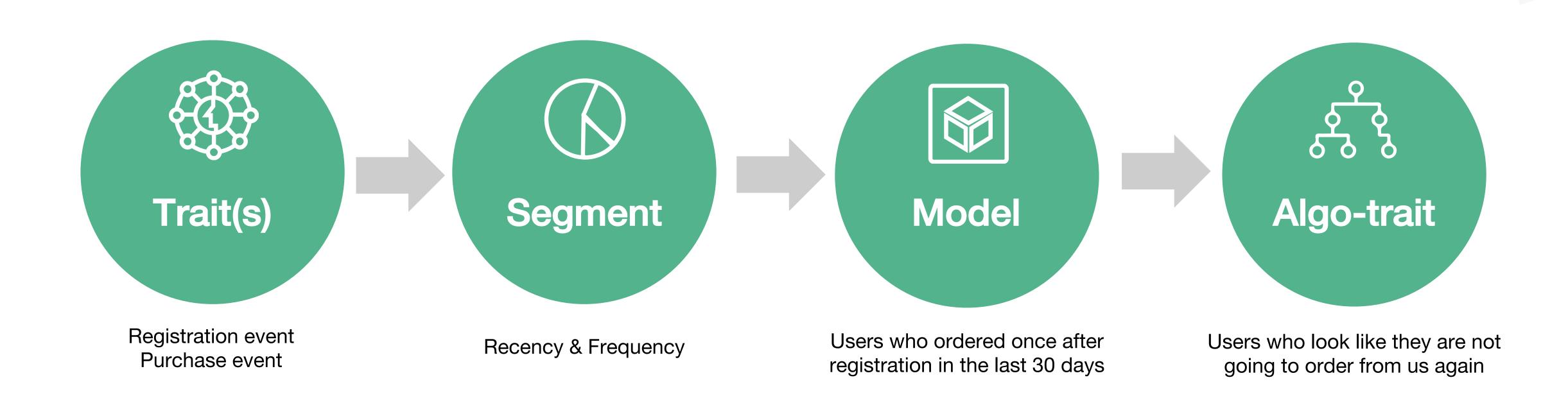
- 12,500 users register themselves and place their 1st order every month
- 70% don't place another order within 30 days
- How can we intelligently target them with better products to make them convert again?
- What is the next best action?
- Customer Lifecycle The Early Stage



## ST CUSTOMER WISH LIST

- Must be able to run the data from Adobe and Offline sources in the RS Virtual Private
   Cloud Data Environment
- Must be able to leverage open algorithmic approaches to build propensity scores across the known customer data set.
- Must be able to connect to the RS DMP to provide targeted lists to the RS Marketing Tech stack
- Must be measurable

Find the audience in Adobe Audience Manager



# ST THE SOLUTION

### Target audience



Consistent message across channels and site



#### Track success



Reduce the proportion of drop-offs after 1 order

#### WHAT IS THE MESSAGE?

# Next stage of customer lifecycle



Next best action model

Content options for campaigning for second order

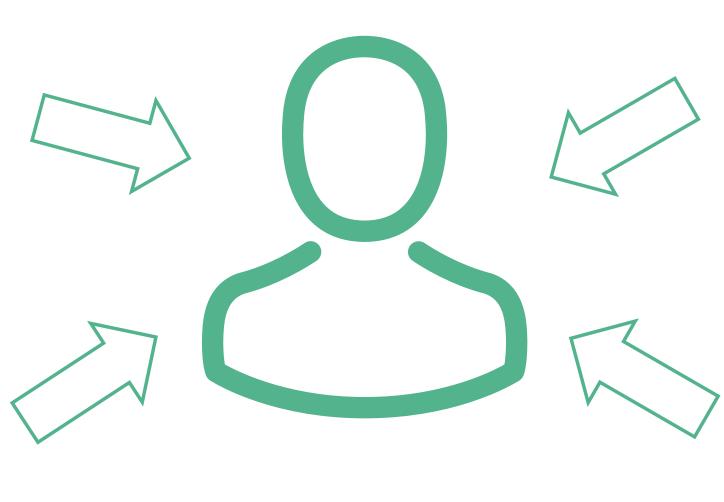
#### **GENERAL**

value proposition



### **PERSONALIZED**

product recommendation







"Utilise multiple data sources to truly understand RS customers and generate accurate product recommendations for our new customers"



### Multiple data sources

#### **Adobe Analytics**

Daily, automated upload

#### Adobe Campaign logs

Daily, automated upload

#### Offline transactional data

Weekly, automated upload





Adobe Audience Manager

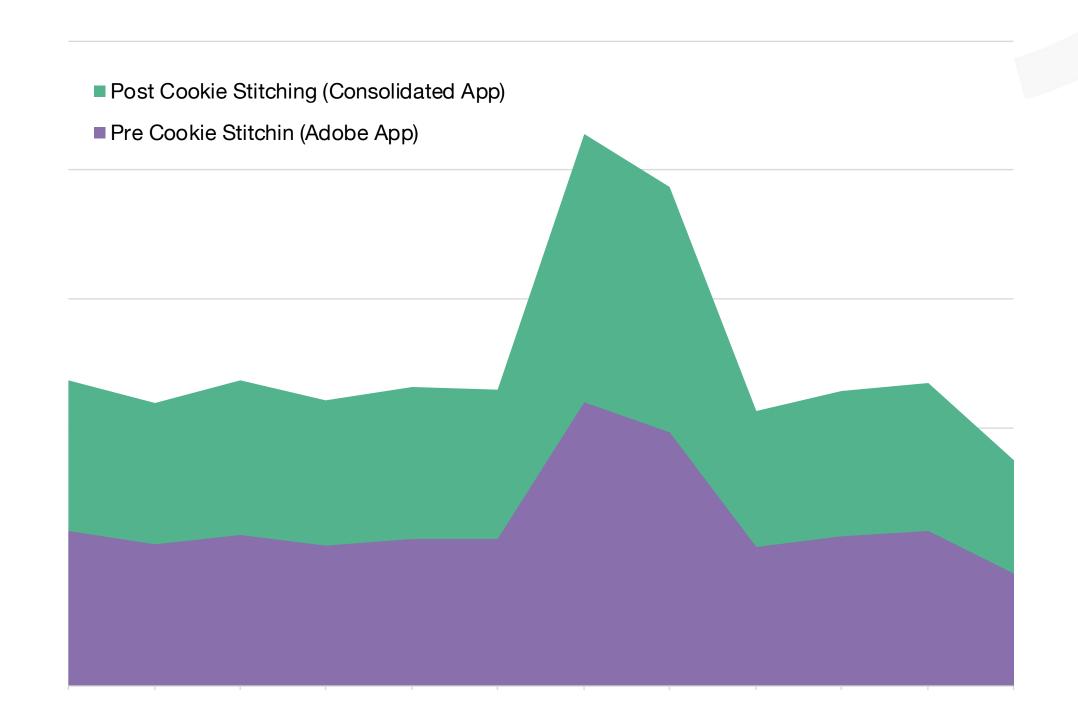
Adobe Campaign

Adobe Target

Paid Advertising

### Cookie stitching results

- Increased the amount of known customer data by 115%
- Process uses cookie to customer ID matches retrospectively & proactively
- Optimises data for modelling, segmentation & analysis



# ST THE SOLUTION

#### PLANNED WORK

#### **All Products Recommendation Model**

 Generating up to 50 Product Family Recommendations for each website visitor

#### **RS Pro Recommendation Model**

 A second model will work as per the above but be restricted to products in the RS Pro product range

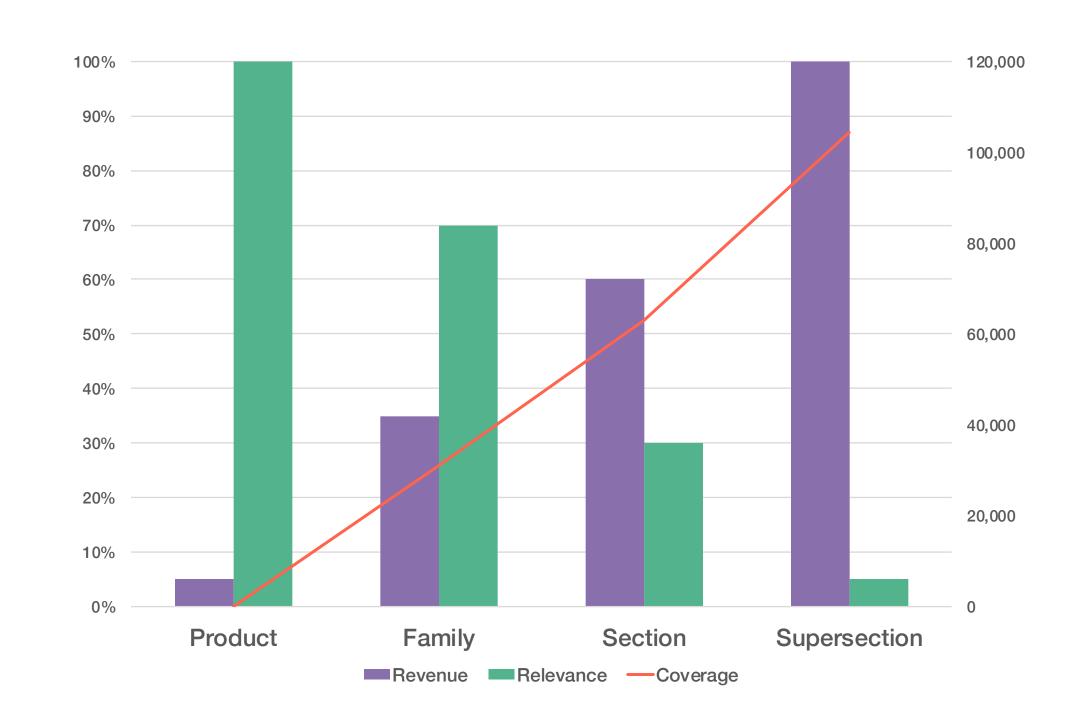


# **SY** EXPERIMENTATION

#### EXPERIMENTATION

#### **Initial Data Analysis**

- 5 Products out of 1 Product family vs. 5 Product families
- Generate recommendation at what level?



## **STEAMENTATION PLAN**

- Recommend 5 families out of 50 families per customer
- Models based on Apr and May 2018 data
- Evaluation metric precision@k (with k =5)
  - Recommend 5 families to a customer and if any of those 5 families were included in next 2 purchases, consider the recommendation to be correct
- Pick a baseline to compare effectiveness of model
- Baseline of recommending top 5 families to every customer
  - Precision of 0.13

### Sì

### MODEL 1 EXPERIMENTATION

#### FP GROWTH

- Algorithm to identify frequent item-sets
- Originally developed to do market basket analysis
- Useful for identifying patterns like
  - "Customers who bought Processor & Microcontroller Development Kits are likely to buy Plug In Power Supply"
  - "Customers who bought Ceramic Multilayer Capacitors are likely to buy Surface Mount Fixed Resistors"
- Generated a precision of 0.17 overall (1.3X lift)
  - This model generated 2 types of cases and above precision is by combining below cases
    - Customers for whom there was a recommendation (precision 0.4)
    - Customers for whom there was no recommendation (all considered as misclassified)

# MODEL 2 EXPERIMENTATION

#### ENSEMBLE - FP GROWTH + REPURCHASE PROBABILITY

- RS's customers tend to repurchase same family with a high rate (16% of existing customers and 25% of new customers)
  - This is not a normal occurrence in 'recommendation problems'
  - A standard model will not be enough for our problem
- An ensemble model was generated based on
  - Repurchase probability based on families
  - FP-Growth
- Generated a precision of 0.42 overall (3X lift)
- Even after eliminating families that are not purchased often, this model recommends a total of 450 families out of roughly 3000 families
- Reducing this down to only 50 families will reduce the lift drastically

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### MODEL 3 EXPERIMENTATION

#### MULTI CLASS CLASSIFICATION MODEL

- Considering that we can only recommend a static set of 50 families and there is a high rate of repurchase, a classification model is a good fit
- A Multi-Class classification model was generated that
  - Predicts how likely a customer is to purchase any of the 50 product families
  - Identify top 5 products a customer is interested in
- This model generated a precision of 0.5 overall
- Considerations:
  - This model recommends 5 families for all customers (existing and new)
  - Model does not wait until customer makes very first purchase to recommend next product
    - About 10% of new customers explore products, register but do not purchase on that day and purchase products later
    - Having recommendations in place for those customers can be beneficial

# FUTURE POSSIBILITIES

### Good fit for full scale Amazon style recommendation

- Large (individual) Product space
- Stable customer base

### Different purchasing patterns observed – in-depth analysis needed to confirm

- Loyal customers who repeatedly buy similar products regularly
- Not so regular customers
- Customers at risk
- Further benefits towards Personalization, Promotions (RS Pro brand products)

### Customer acquisition Journey Analytics



Activation



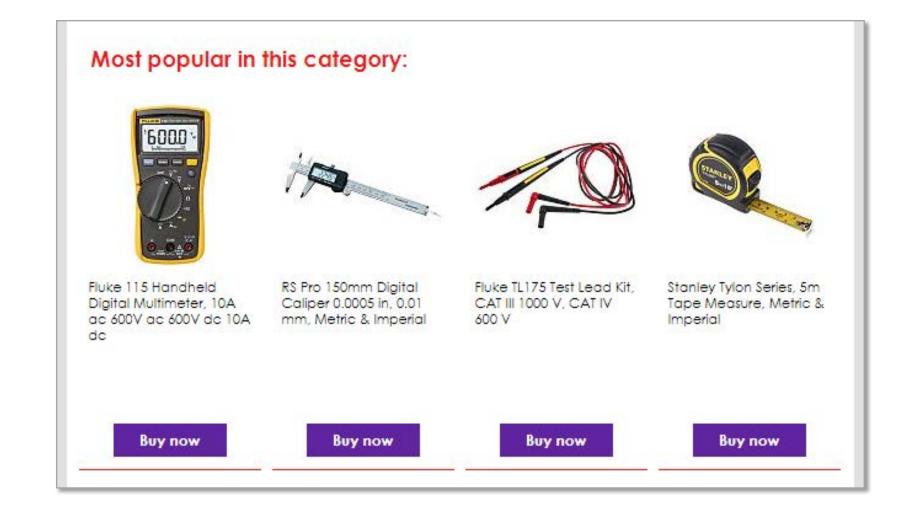


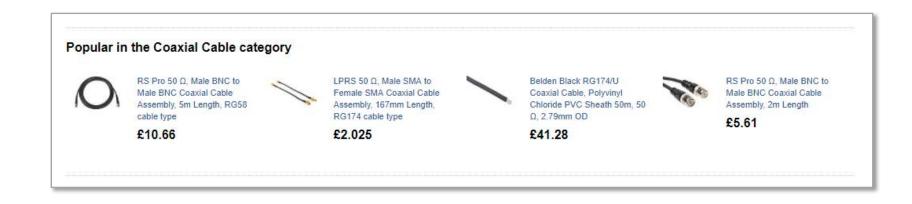
#### Adobe Audience Manager

Adobe Campaign

**Adobe Target** 

Paid Advertising





SI Q&A

