

An Introduction to Data Science for Sensory and Consumer Scientists

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2020-12-29

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Chapter 1

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Welcome to the website for *Introduction to Data Science for Sensory and Consumer Scientists*. This book being written in the open and is currently under development.

Introduction

Chapter 2

Introduction

Here is a change.

Sensory and consumer science (SCS) is considered as a pillar of food science and technology and is useful to product development, quality control and market research. Most scientific and methodological advances in the field are applied to food. This book makes no exception as we chose a cookie formulation dataset as a main thread. However, SCS widely applies to many other consumer goods so are the content of this book and the principles set out below.

2.1 Core principles in Sensory and Consumer Science

2.1.1 Measuring and analyzing human responses

Sensory and consumer science aims at measuring and understanding consumers' sensory perceptions as well as the judgements, emotions and behaviors that may arise from these perceptions. SCS is thus primarily a science of measurement, although a very particular one that uses human beings and their senses as measuring instruments. In other words, sensory and consumer researchers measure and analyze human responses. To this end, SCS relies essentially on sensory evaluation which comprises a set of techniques that mostly derive from psychophysics and behavioral research. It uses psychological models to help separate signal from noise in collected data [ref O'Mahony, D.Ennis, others?]. Besides, sensory evaluation has developed its own methodological framework that includes most refined techniques for the accurate measurement of product sensory properties while minimizing the potentially biasing effects of brand identity and the influence of other external information on consumer perception [Lawless & Heymann, 2010]. A detailed description of sensory methods is beyond the scope of

this book and many textbooks on sensory evaluation methods are available to readers seeking more information. However, just to give a brief overview, it is worth remembering that sensory methods can be roughly divided into three categories, each of them bearing many variants: - Discrimination tests that aim at detecting subtle differences between two products. - Descriptive analysis (DA), also referred to as ‘sensory profiling’, aims at providing both qualitative and quantitative information about product sensory properties. - Hedonic tests. This category gathers affective tests that aim at measuring consumers’ liking for the tested products or their preferences among a product set. Each of these test categories generates its own type of data and related statistical questions in relation to the objectives of the study. Typically, data from difference tests consist in series of correct/failed binary answers depending on whether judges successfully picked the odd sample(s) among a set of three or more samples. These are used to determine whether the number of correct choices is above the level expected by chance. Conventional descriptive analysis data consist in intensity scores given by each panelist to evaluated samples on a series of sensory attributes, hence resulting in a product x attribute x panelist dataset (Figure 1). Note that depending on the DA method, quantifying means other than intensity ratings can be used (ranks, frequency, etc.). Most frequently, each panelist evaluates all the samples in the product set. However, the use of balanced incomplete design can also be found when the experimenters aim to limit the number of samples evaluated by each subject. Eventually, hedonic test datasets consist in hedonic scores (ratings for consumers’ degree of liking or preference ranks) given by each interviewed consumer to a series of products. As for DA, each consumer usually evaluates all the samples in the product set, but balanced incomplete designs are sometimes used too. In addition, some companies favor pure monadic evaluation of product (i.e. between-subject design or independent groups design) which obviously result in unrelated sample datasets. Sensory and consumer researchers also borrow methods from other fields, in particular from sociology and experimental psychology. Definitely a multidisciplinary area, SCS develops in many directions and reaches disciplines that range from genetics and physiology to social marketing, behavioral economics and computational neuroscience. So have diversified the types of data sensory and consumer scientists must deal with.

2.2 How should sensory and consumer scientists learn data science?

2.3 Caution: Don’t that everybody does

2.4 Example projects

Chapter 3

What is Data Science?

3.1 History

3.2 Workflow

3.2.1 Data preparation

3.2.2 Data analysis

3.2.3 Insight delivery

3.3 Benefits of data science

3.3.1 Reproducible research

3.3.2 Other benefits (machine learning?)

3.4 How to learn data science

3.5 How to use this book

3.6 Recommended data science tools

Chapter 4

Getting Started with R

4.1 R

4.2 RStudio

4.3 Git

4.4 GitHub

Data Scientific Workflow

Chapter 5

Example Project

5.1 Background

5.2 Other details

5.3 Conclusions?

Chapter 6

Data Preparation

6.1 Importation

6.2 Organization

6.3 Inspection

6.4 Manipulation

6.5 Cleaning

Chapter 7

Data Analysis

7.1 Transformation

7.2 Exploration

7.3 Modeling

Chapter 8

Data Visualization

8.1 Principles

8.2 Table Mechanics

8.3 Chart Mechanics

8.4 Examples

Chapter 9

Insight Delivery

9.1 Design principles

9.2 Scientific inquiry vs storytelling

9.3 Research reformulation

9.4 Interactive reporting

Reproducible Research

Chapter 10

Tools for Collaboration

10.1 Principles

10.2 Tools

10.2.1 GitHub

10.2.2 R scripts

10.2.3 RMarkdown

10.2.4 Shiny

10.3 Documentation

10.4 Version control

10.5 Online repositories for team collaboration

10.6 Building a code base

10.6.1 Internal functions

10.6.2 Packages

Chapter 11

Automated Reporting

11.1 Excel

11.2 Word

11.3 PowerPoint

11.3.1 Charts

11.3.2 Tables

11.3.3 Bullet Points

11.3.4 Images

11.4 HTML

Additional Topics

Chapter 12

Machine Learning

12.1 Concepts and general workflow (training/test)

12.2 Unsupervised learning

12.2.1 Cluster analysis

12.2.2 Factor analysis

12.2.3 Principle components analysis

12.2.4 t-SNE

12.3 Semisupervised learning

12.3.1 PLS regression

12.4 Supervised learning

12.4.1 Regression

12.4.2 K-nearest neighbors

12.4.3 Decision trees

12.4.4 Black boxes

12.4.4.1 Random forests

12.4.4.2 SVMs

12.4.4.3 Neural networks

12.5 Predictive modeling

Chapter 13

Text Analysis

13.1 Data import

13.1.1 Data sources

13.1.2 Tokenizing

13.1.3 Lemmatization, stemming, and stop word removal

13.2 Analysis

13.2.1 Frequency counts and summary statistics

13.2.2 Word clouds

13.2.3 Contrast plots

13.2.4 Sentiment analysis

13.2.5 Bigrams and word graphs

Chapter 14

Graph Databases

Conclusion

Chapter 15

Conclusion

Appendices

