

# MKT 511 Marketing Analytics

## Tutorial 3 – Market Analytics



# Market Analytics

- **Part A) Market Basket Analysis**
  - Exercise 1
  - Exercise 2
- **Part B) Segmentation**
  - Exercise 3
  - Exercise 4
  - Exercise 5

# Part A) Market Basket Analysis

## Association Rules

- A **rule** is the association of one set of items conditional on another, e.g.,  $\{x\} \rightarrow \{y\}$
- Three measures to assess rules:
  - **support** $\{x, y\}$  is the proportion of all transactions that contain some set
  - **confidence** $(x \rightarrow y) = \frac{\text{support}\{x, y\}}{\text{support}\{x\}}$
  - **lift** $(x \rightarrow y) = \frac{\text{support}\{x, y\}}{\text{support}\{x\} \times \text{support}\{y\}}$

# Part A) Market Basket Analysis

## Exercise 1

- Transaction data for a Belgian supermarket
- Exploring the data and transforming it into a useful form **(a)**
- Finding association rules in the given data set **(b)**
- Selecting those rules that fulfill certain criteria **(c)**
- Simulating transaction profitability **(d)**

# Part A) Market Basket Analysis

## Exercise 2

- Non-transaction data of customer segments
- Converting the data into discrete form (a)
- Finding association rules in the given data set (b)
- Selecting those rules that fulfill certain criteria (c & d)

# Part B) Segmentation

## Exercise 3

- Music service subscription data
- Exploring the data and removing segment information (**a**)
- Applying various cluster algorithms to find groups in the data (**b & c**)

# Part B) Segmentation

## Exercise 3

- Plotting the results (d)

`labels` integer code, currently one of 0,1,2,3,4 and 5. If

`labels= 0,`  
no labels are placed in the plot;

`labels= 1,`  
points and ellipses can be identified in the plot (see [identify](#));

`labels= 2,`  
all points and ellipses are labelled in the plot;

`labels= 3,`  
only the points are labelled in the plot;

`labels= 4,`  
only the ellipses are labelled in the plot.

`labels= 5,`  
the ellipses are labelled in the plot, and points can be identified.



# Part B) Segmentation

## Exercise 4

- Music service subscription data
- Applying model-based clustering (a & b)

**Table 11.1** Interpretation of the Bayesian Information Criterion (BIC) when comparing two models. Lower BIC is better, and the difference in BIC indicates the strength of evidence. Adapted from Raftery [159], p. 139

BIC difference	Odds of model superiority	Strength of the evidence
0–2	50–75%	Weak
2–6	75–95%	Positive
6–10	95–99%	Strong
> 10	> 99%	Very strong

R for Marketing Research and Analytics (Chapman & Feit, 2019)



# Part B) Segmentation

## Exercise 4

- Recoding variables to binary factors (c)
- Fitting categorical latent class analysis (d)

<code>formula</code>	A formula expression of the form <code>response ~ predictors</code> . The details of model specification are given below.
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Model specification: Latent class models have more than one manifest variable, so the response variables are `cbind(dv1, dv2, dv3...)` where `dv#` refer to variable names in the data frame. For models with no covariates, the formula is `cbind(dv1, dv2, dv3) ~ 1`. For models with covariates, replace the `~1` with the desired function of predictors `iv1, iv2, iv3...` as, for example, `cbind(dv1, dv2, dv3) ~ iv1 + iv2 * iv3`.

```
music_LCA_3 <- polCA(cbind(age, sex, householdIncome, kidsAtHome, commuteCar,
                           musicEnthuse, subscribeToMusic) ~ 1,
                    data = musicdata_lca, nclass = 3)
music_LCA_4 <- polCA(cbind(age, sex, householdIncome, kidsAtHome, commuteCar,
                           musicEnthuse, subscribeToMusic) ~ 1,
                    data = musicdata_lca, nclass = 4)
```

# Part B) Segmentation

## Latent Class Analysis

- Can use **k-means clustering approach** to test for normality
  - Run cluster analysis with very high  $k$  (e.g., 100)
  - Eliminate those observations that are grouped into very small and unusual clusters (outliers)
  - Multivariable technique to find non-normal observations

# Part B) Segmentation

## Exercise 5

- Music service subscription data
- Splitting the data into training and test set (a)

### Description

`sample` takes a sample of the specified size from the elements of `x` using either with or without replacement.

### Usage

```
sample(x, size, replace = FALSE, prob = NULL)
```

### Arguments

<code>x</code>	either a vector of one or more elements from which to choose, or a positive integer. See 'Details.'
<code>n</code>	a positive number, the number of items to choose from. See 'Details.'
<code>size</code>	a non-negative integer giving the number of items to choose.

```
sample(x = nrow(musicdata_raw), size = nrow(musicdata_raw)*0.65)  
# works, since nrow( ) returns an integer  
sample(x = musicdata_raw, size = nrow(musicdata_raw)*0.65)  
# doesn't work, since musicdata_raw is not a vector of integer
```

# Part B) Segmentation

## Exercise 5

- Fitting different classification models (**b, c & e**)
- Determining important variables (**d**)

```
(music_rf_sub <- randomForest(subscribeToMusic ~ ., data = music_train,  
                               ntree = 5000, importance = TRUE))
```

- ( ) around an assignment to simultaneously run the code and print the results (Tutorial 1)
- ntree should be roughly 5 – 10 times the sample size

# Tutorial 3 – Market Analytics

## Questions?

Please feel free to ask all of them in the Q&A Forum on ILIAS!

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