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PanelCheck – A Tool for Monitoring of Assessor and Panel Performance

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OUTLINE

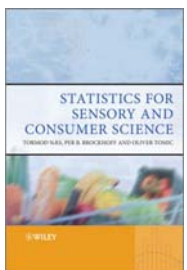
- What is PanelCheck and why use it?
- Data structure & data import
- Tucker-I and Manhattan plots
- One-way ANOVA for panelist performance
- Export of plots



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Literature

- Chapter 3 on sensory panel performance
- Tormod Næs, Per Brockhoff, Oliver Tomic
- Book info:
 - ISBN: 978-0-470-51821-2
 - Hardcover
 - 300 pages
 - June 2010



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Literature

- O. Tomic, A. Nilsen, M. Martens, T. Næs, **Visualization of sensory profiling data for performance monitoring**, LWT 40 (2005), 262-269
 - F plot, MSE plot, p*MSE plot (one-way ANOVA)
 - Correlation plot, Eggshell plot
- T. Deht, O. Tomic, J.P. Wold, T. Næs, **Some new tools for visualising multi-way sensory data**, Food Quality and Preference 19/1 (2008), 103-113
 - Tucker-I plots
 - Manhattan plots
- O. Tomic, G. Luciano, A. Nilsen, G. Hyldig, K. Lorensen, T. Næs, **Analysing sensory panel performance in a proficiency test using the PanelCheck software**, European Food Research and Technology 230 (2010), 497-511
 - Workflow
 - Comparison of performance from multiple panels
- O. Tomic, C. Forde, C. Delahunty, T. Næs, **Performance indices in descriptive sensory analysis - a complementary screening tool for assessor and panel performance**, Food Quality and Preference 28 (2013), 122-133
 - Performance indices



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What is PanelCheck and why use it?

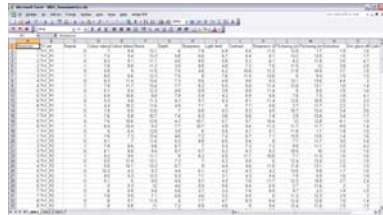


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What is PanelCheck and why use it?

«LARGE» AMOUNT OF NUMBERS (for non-statisticians)

- Example:
 - 10 assessors
 - 8 products
 - 2 replicates
 - 20 attributes
- → 160 rows of data
- → 3200 cells of data






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What is PanelCheck and why use it?

How to extract relevant information?

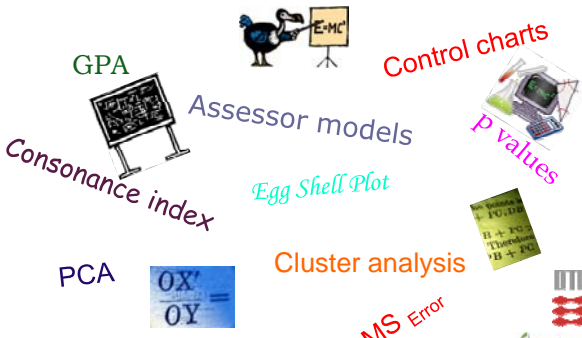


- Is the variation in your data based on:
 - Differences between the products?
 - Poor performance of the assessors?
- Is your sensory panel well enough trained?

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What is PanelCheck and why use it?

Which statistical method to use?








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What is PanelCheck and why use it?

Which statistical method to use?

- Plenty of different statistical methods available
- Various commercial data analysis software packages (costly)
- Data collection software contain some methods
- Various open source free software packages (might require programming skills)








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What is PanelCheck and why use it?

PanelCheck for practitioners and researchers

- An easy-to-use software tool for monitoring of sensory panel performance
- Taylor made for analysis descriptive analysis data
- Visual approach for performance analysis
- Does not require detailed knowledge in statistics
- Free open source software








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What is PanelCheck and why use it?

GENERAL PanelCheck CONCEPT

- There is **no** single plot / statistical method showing / yielding **ALL** the information in your data
- Use **different** plots (based on different statistical methods) to reveal **different** type of important information
- Use **joint information** from plots to get a **comprehensive overview** over performance (individuals, panel)








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What is PanelCheck and why use it?

WHAT CAN PanelCheck BE USED FOR?

- Checking performance of assessors and panel
- Checking performance of multiple panels in inter-collaborative tests
- Analysis of tested products






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What is PanelCheck and why use it?

Where to download and get information?

- www.panelcheck.com
- Download PanelCheck software
- Download posters, find info, etc
- LinkedIn: PanelCheck user group





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What is PanelCheck and why use it?

PanelCheck history and facts



- Collaboration between
 - Nofima (former Matforsk)
 - Technical University (DTU)
 - Norwegian industry
 - Danish industry
- First release in 2005
- About 20 000 downloads since
- Users from 94 countries

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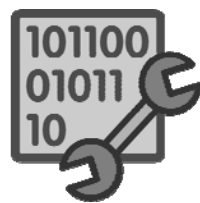

What is PanelCheck and why use it?

PanelCheck history and facts

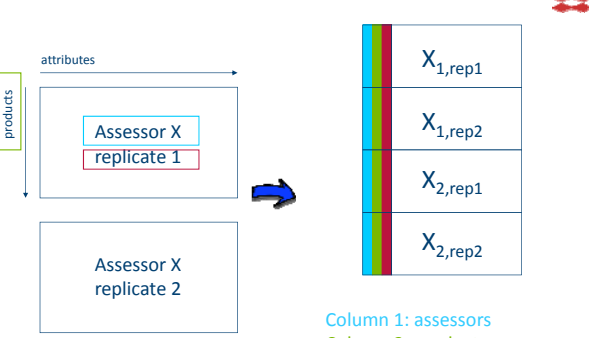
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Data structure & data import

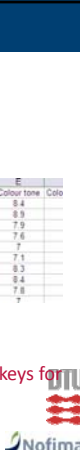



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Data structure & data import



Column 1: assessors
Column 2: products
Column 3: replicates



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Data structure & data import

How to organise the data

- Columns:
 - Column 1: assessor names
 - Column 2: product names
 - Column 3: replicate names
 - Column 4 ++ : attributes
- Rows:
 - Each row represents one tasting of one assessor
 - Order of rows is not important

Unique combination of keys for each row

	A	B	C	D	E
1	Assessor	Product	Replicate	Whiteness	Colour tone
2	1	Kaffe	1	2.9	8.4
3	1	Kaffe	2	2	8.9
4	1	Maarut	1	2.5	7.9
5	1	Maarut	2	3	7.6
6	1	Serlands	1	4.6	7
7	1	Serlands	2	4	7.1
8	2	Kaffe	1	3.9	8.3
9	2	Kaffe	2	3.8	8.4
10	2	Maarut	1	5	7.8
11	2	Maarut	2	6.7	7



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Data structure & data import

Balanced and unbalanced data

- ALL assessors need to have tested ALL products
- ALL assessor need an IDENTICAL number of replicates
- Unbalanced part will be stripped away (either assessor or product)

Order of products in NOT important

Whole row missing
→ unbalanced data

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Data structure & data import

Missing data

- Missing data in single cells are OK and will be imputed

Missing data

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Data structure & data import

File formats and ways of importing data

- Supported file formats:
 - .txt (delimited by: tab, comma, semicolon)
 - .csv
 - .xls
- Import through:
 - File → Import
 - File → Import recent
 - Drag & drop
 - Right-click on file and open with 'PanelCheck'

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Data structure & data import

EXCEL FILE (.xls)

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Data structure & data import

TEXT FILE (.txt)

(tab-delimited)

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COMMA SEPARATED VALUE TEXT (.csv)

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Data structure & data import

IMPORT DIALOG

Select columns for assessor, samples and replicates.

Assessor	TV test	Repeat	Alar saturation	Colour balance	Noise	Depth	Shurpen
1	TV1_P1	0.0	7.0	9.8	13.1	6.0	7.9
2	TV1_P1	1.0	7.5	9.4	13.3	5.8	6.6
3	TV1_P1	0.0	8.3	9.1	13.1	6.5	7.9
4	TV1_P1	1.0	7.8	8.8	13.2	5.9	8.8
5	TV1_P1	0.0	8.8	9.0	13.5	7.5	4.6
6	TV1_P1	1.0	8.5	9.6	13.3	7.5	8.0
7	TV1_P1	0.0	8.3	11.4	13.4	7.3	9.6
8	TV1_P1	1.0	7.9	11.7	13.4	7.7	8.2
9	TV1_P1	0.0	8.2	9.4	13.3	6.8	8.9
10	TV1_P1	1.0	6.9	10.8	8.8	7.9	5.4
11	TV1_P1	0.0	8.3	4.8	13.3	4.1	8.7
12	TV1_P1	1.0	4.4	10.2	13.6	4.9	7.1
13	TV1_P1	0.0	7.9	6.5	12.6	7.0	6.8
14	TV1_P1	1.0	7.6	8.8	10.7	7.4	6.3
15	TV1_P1	0.0	7.6	10.6	12.8	4.2	10.3
16	TV1_P1	1.0	8.4	10.8	9.3	7.7	10.7
17	TV1_P2	0.0	8.4	8.4	12.8	3.9	4.0
18	TV1_P2	1.0	7.8	7.3	13.4	4.6	7.2
19	TV1_P2	0.0	8.1	7.9	11.6	5.1	8.1

Data structure & data import

SUCCESSFUL DATA IMPORT

PanelCheck scans and finds highest and lowest value in data set.

PanelCheck suggests scale limits for some plots. you may change the limits.

Workflow

In which order should plots be analysed?

Workflow

- There is no "right" or "wrong" way of:
 - which plots should be considered
 - in which order they should be used
- Workflow only a recommendation based on our own experience

Workflow

PanelCheck menu: Help → Workflow


Plots based on Principal Component Analysis (PCA)

Results used in:

- » Tucker plot
- » Manhattan plot
- » Analysis of panel consensus

Plots based on Principal Component Analysis (PCA)

- For full details on Tucker-1 and Manhattan plots please read:



Available online at www.sciencedirect.com
ScienceDirect
 Food Quality and Preference 19 (2008) 103–113
www.elsevier.com/locate/foodqual


Food Quality and Preference

Some new tools for visualising multi-way sensory data

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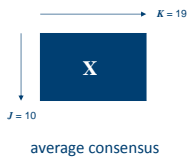



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PCA basics

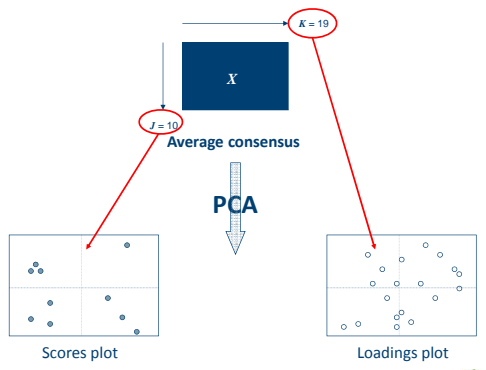

Example data

- 10 products
- 19 attributes
- (10 assessors, 2 reps)


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PCA basics





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PCA basics



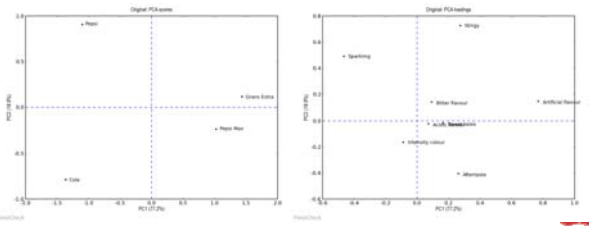

- Principle components (PC's) are orthogonal (uncorrelated)
- Each PC explains a certain amount of the total variance
- Scores provide information on the products
- Loadings provide information on how the attributes have contributed to the variation in the data



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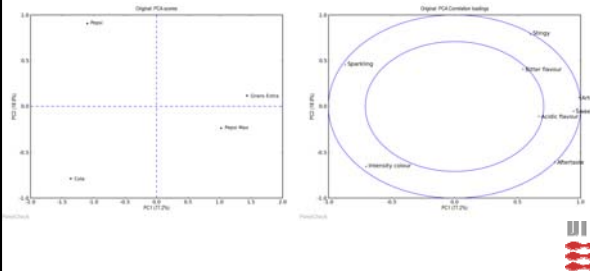

PCA basics

- Products that are close to each other are very similar and vice versa
- Attributes that are close to each other are highly correlated
- Superimpose scores and loadings plots to understand variation in data

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PCA basics

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PCA – Tucker 1

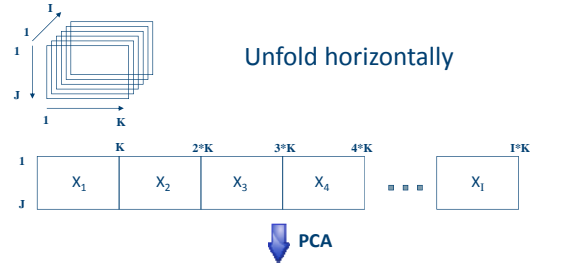
TUCKER-1 plot – DISPLAYING AGREEMENT BETWEEN ASSESSORS



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PCA – Tucker 1 – common scores

Unfold horizontally



Common scores plot showing J objects

Loadings or correlation loadings plot showing $I * K$ variables




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PCA – Tucker 1

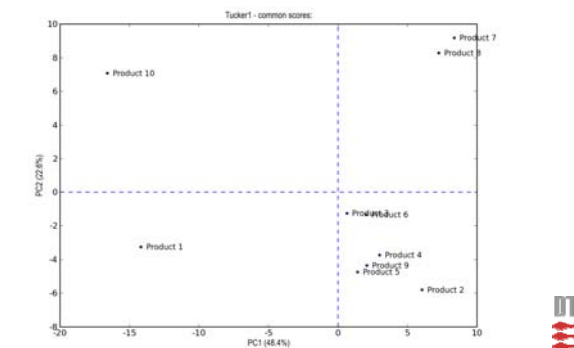

What is shown in the plots?

- **Common scores plot:** Shows how samples relate to each other
- **Correlation loadings plots:**
 - Shows $I * K$ dots in plot:
 - I : number of assessors
 - K : number of attributes
 - Each dot represents the attribute of one assessor
 - Focus on either assessors or attributes
 - Inner circle: 50% explained variance
 - Outer circle: 100% explained variance
 - Focus on assessor or attribute



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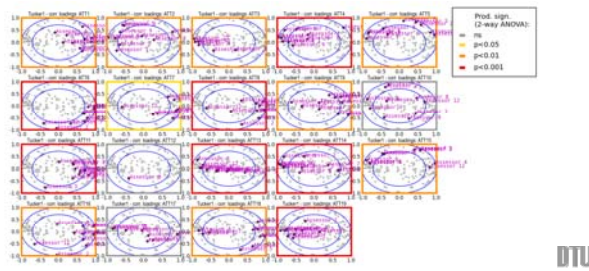

PCA – Tucker 1 – common scores

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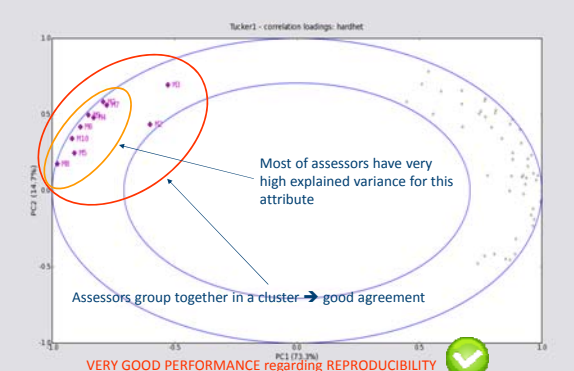
PCA – Tucker 1

TUCKER1 – CORRELATION LOADINGS: OVERVIEW PLOT

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PCA – Tucker 1 – correlation loadings



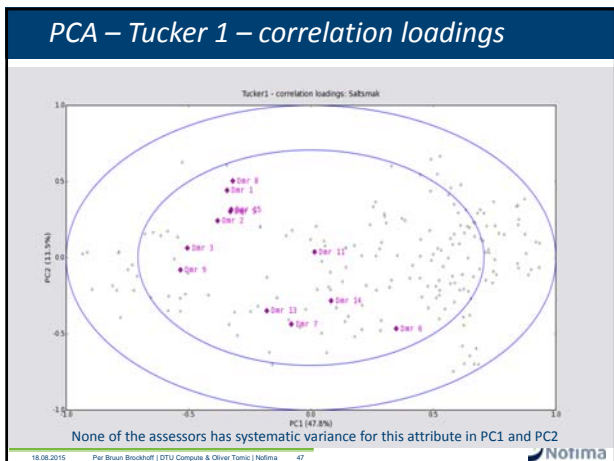
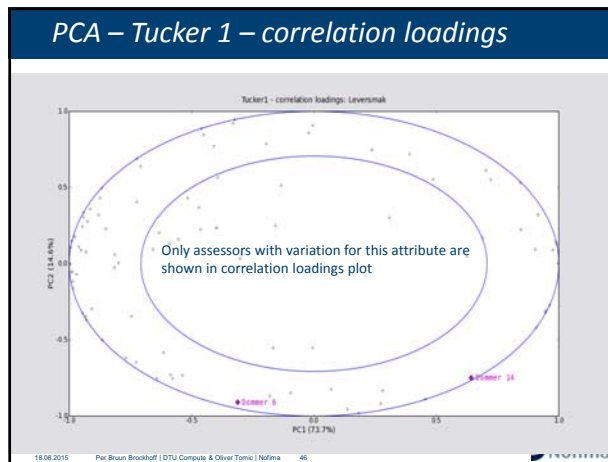
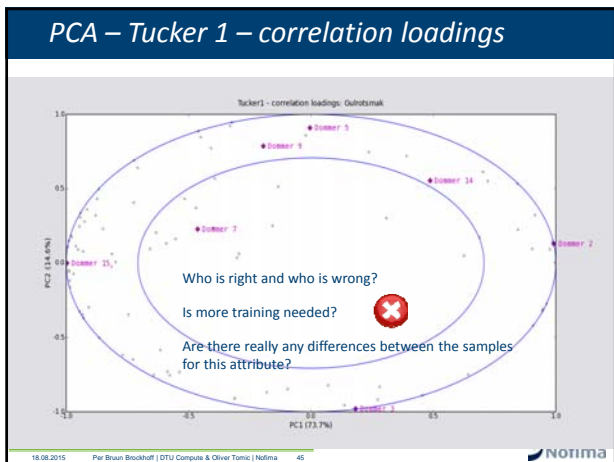
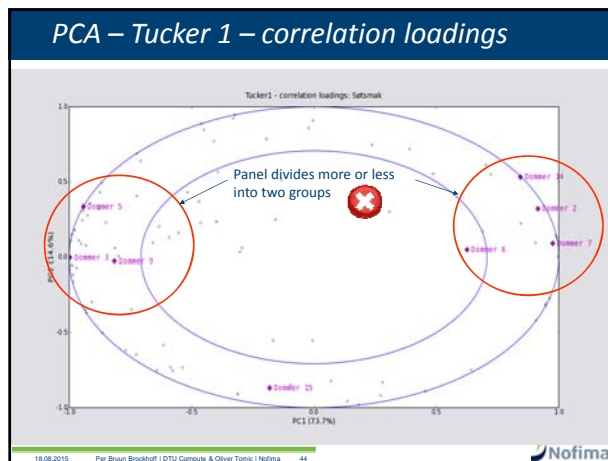
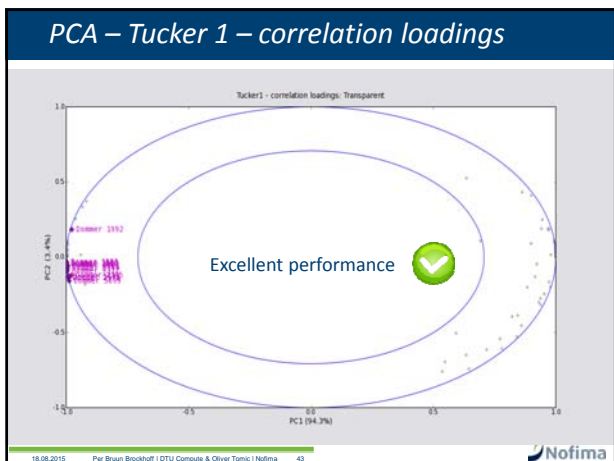
Most of assessors have very high explained variance for this attribute

Assessors group together in a cluster → good agreement

VERY GOOD PERFORMANCE regarding REPRODUCIBILITY ✓



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PCA – Tucker 1 – correlation loadings

SUMMARY TUCKER-1 PLOTS

- Features:
 - Provides information about reproducibility across panel
 - Provides information about systematic variation for each assessor – attribute combination
- Weakness:
 - No information on use of scale

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PCA – Tucker 1 – one attribute

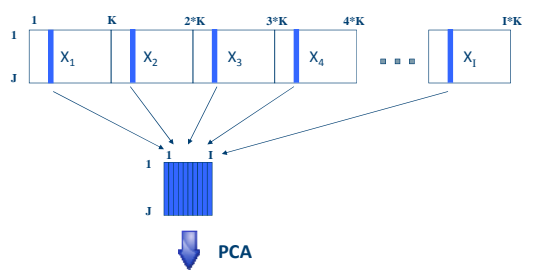
Tucker 1 with a little twist

- Look at only ONE attribute at the time
- → detach from influence of other attributes
- → easier to evaluate agreement across assessors

- When agreement data should be unidimensional
 - → explained variance should be as close as possible to 100% in PC 1
 - → all assessors should cluster together on one side of PC 1


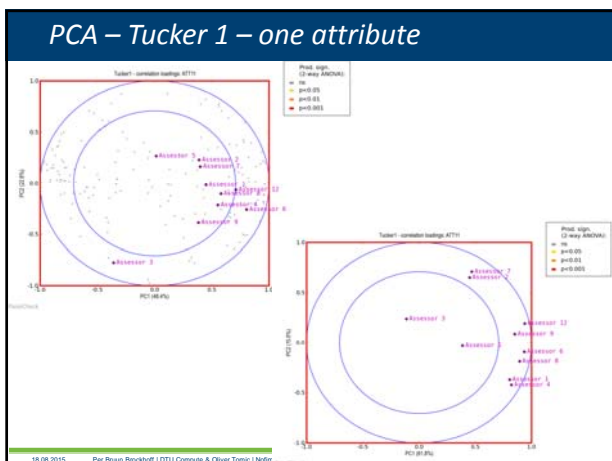


PCA – Tucker 1 – one attribute




Common scores plot showing J objects

Correlation loadings plot showing I variables

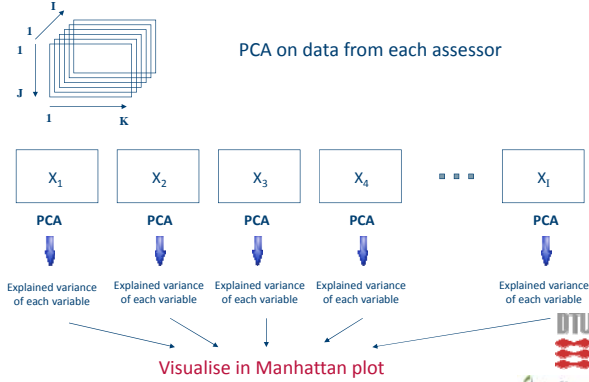
PCA – Manhattan plot

Manhattan plot – A SCREENING TOOL FOR CHECKING GENERAL DATA STRUCTURE


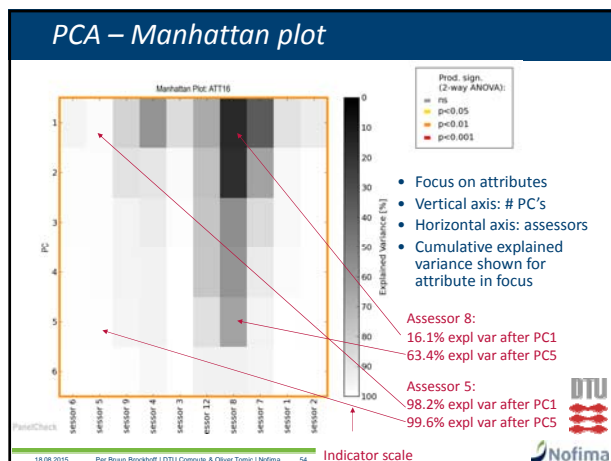


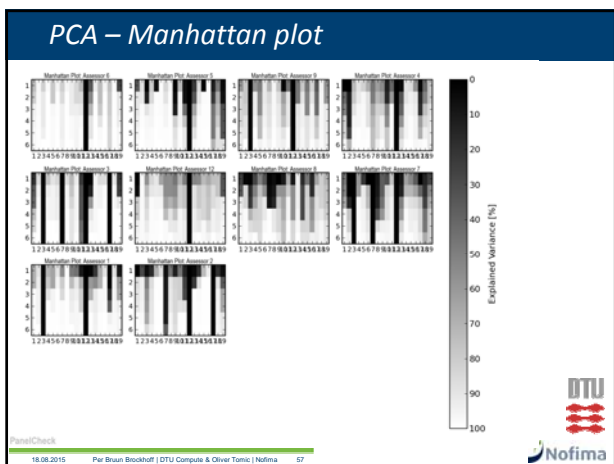
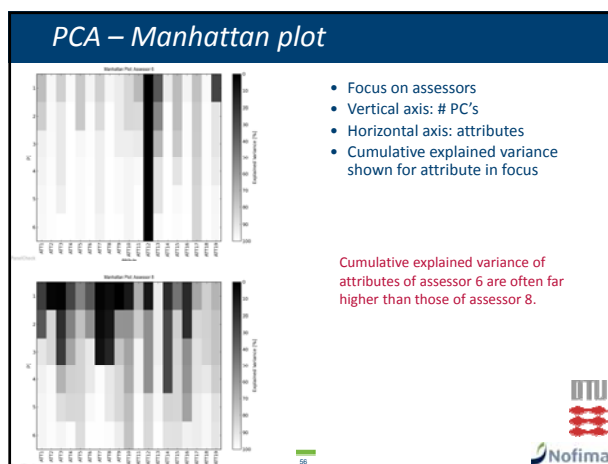
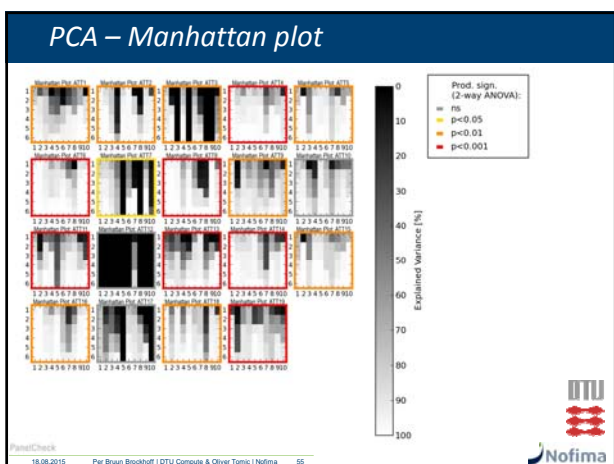
PCA – Manhattan plot

PCA on data from each assessor



Visualise in Manhattan plot



PCA – Manhattan plot

SUMMARY MANHATTAN PLOTS

- Features:
 - Provides quick information data structure for each assessor
 - Provides insight “deep” into data (beyond PC2)
- Weakness:
 - No information about repeatability
 - No information on use of scale
 - No information on how products are distributed

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Plots based on one-way ANOVA

Results used in:

- » *F plot*
- » *MSE plot*
- » *p*MSE plot*

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One-way ANOVA

- Applied on data from only ONE assessor at a time
- Requires a minimum 2 replicates
- Main effect: products (fixed)


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One-way ANOVA

- Cumbersome to do this manually for each assessor and each attribute
- Example:
 - 10 assessors
 - 30 attributes
 - → 10 * 30 = 300 analyses in Minitab
- PanelCheck does this automatically for the user
- Results are used in various plots


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One-way ANOVA – p*MSE plot

p * MSE plot – SIMULTANEOUSLY CHECK ASSESSOR'S REPEATABILITY AND PRODUCT DISCRIMINATION ABILITY

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One-way ANOVA – p*MSE plot

What is shown in the plots?

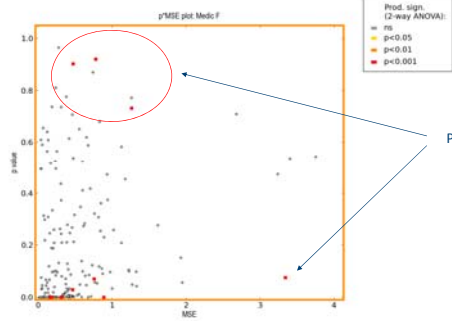
- Shows $I * K$ dots in plot:
 - I : number of assessors
 - K : number of attributes
- Each dot represents the attribute of one assessor
- Focus on either assessors or attributes

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One-way ANOVA – p*MSE plot

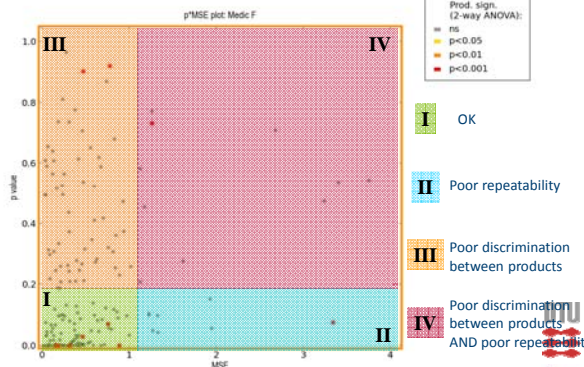
FOCUS ON ATTRIBUTE




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One-way ANOVA – p*MSE plot

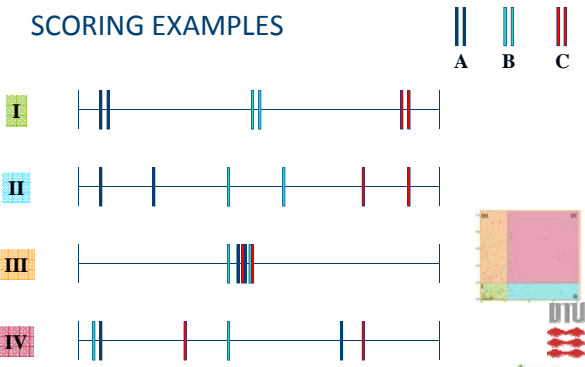


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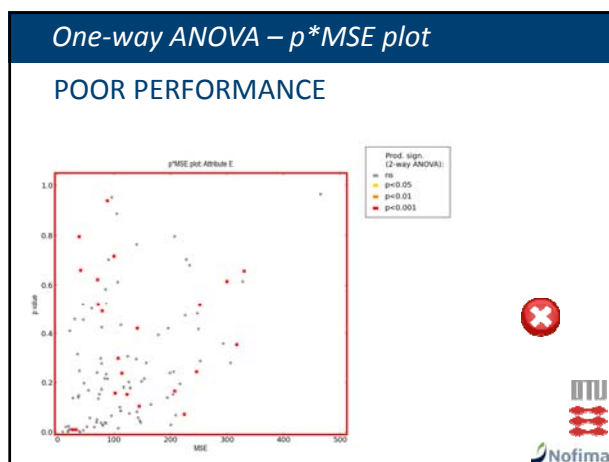
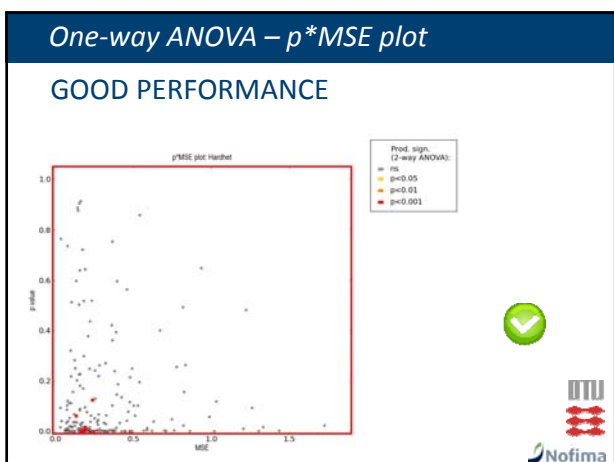
One-way ANOVA – p*MSE plot

SCORING EXAMPLES



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One-way ANOVA – p *MSE plot

p *MSE PLOT - SUMMARY

- Features:
 - Provides information about assessor's repeatability AND ability to discriminate between samples
 - Shows all attributes and assessors in one plot
- Weakness:
 - No information about sample ranking (scale upside down?)

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One-way ANOVA – F plot

F plot – CHECK ASSESSOR'S PRODUCT DISCRIMINATION ABILITY

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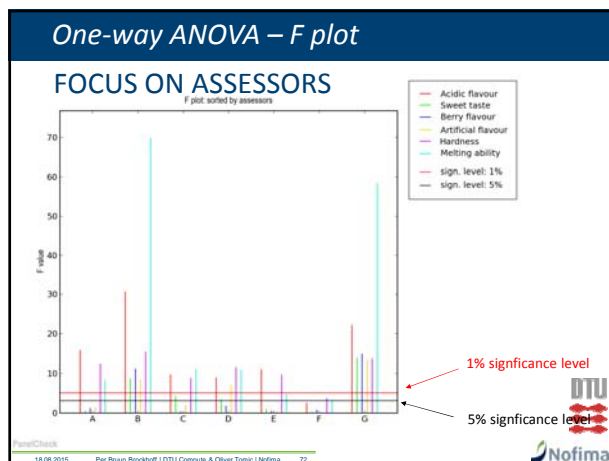
One-way ANOVA – F plot

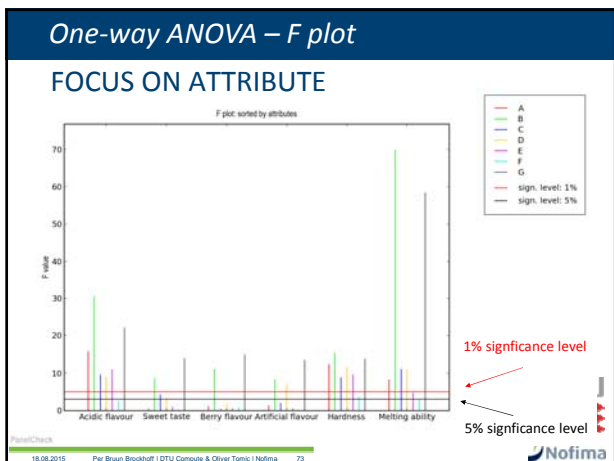
What is shown in the plots?

- Shows $I * K$ vertical lines in plot:
 - I : number of assessors
 - K : number of attributes
- Each line represents one attribute of one assessor
- Focus on either assessors or attributes

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One-way ANOVA – F plot

F plot - SUMMARY

- Features:
 - Shows all assessors and all attributes in one plot
 - Easy to compare assessors/attributes with each other
 - Info about whether an assessor can discriminate between products or not
- Disadvantages:
 - No information about use of scale
 - No information about quality of replicates
 - No information about product ranking

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One-way ANOVA – MSE plot

MSE plot – CHECK ASSESSOR’S REPEATABILITY

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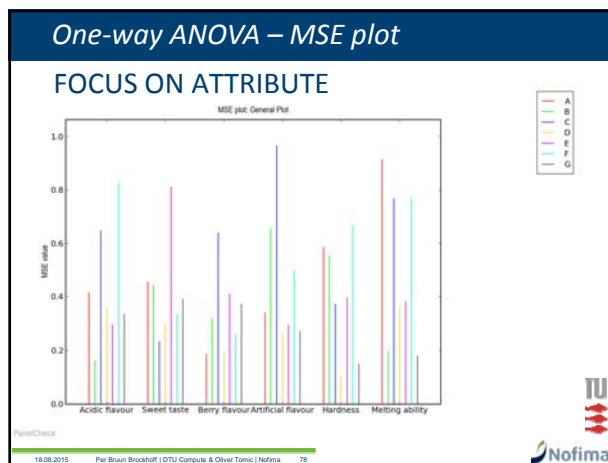
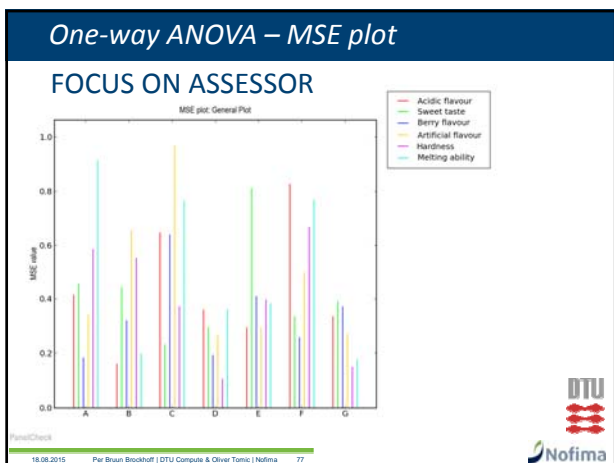
One-way ANOVA – MSE plot

What is shown in the plots?

- Shows $I * K$ vertical lines in plot:
 - I : number of assessors
 - K : number of attributes
- Each line represents one attribute of one assessor
- Focus on either assessors or attributes

DTU


Nofima





One-way ANOVA – MSE plot

MSE plot - SUMMARY

- Features:
 - Shows all assessors and all attributes in one plot
 - Easy to compare assessors/attributes with each other
 - Info about whether an assessor can discriminate between products or not
- Disadvantages:
 - No information about use of scale
 - No information about quality of replicates
 - No information about product ranking




Export of plots and results

Export of plots and results

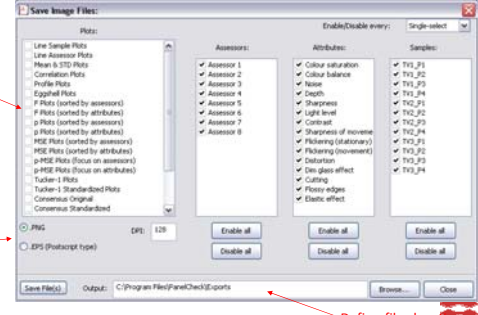
WAYS TO EXPORT PLOTS

- Copy (ctrl-c) and paste (ctrl-v)
- Save images to files:
 - All selected images are being saved to .png-files
 - .png-files should be handled most programs (like .jpg-files)
- Save images to PowerPoint:
 - All selected files are exported to a new PowerPoint-file
 - All selected images are being saved to .png-files



Export of plots and results


SAVE IMAGES TO FILES



Choose which plots to export

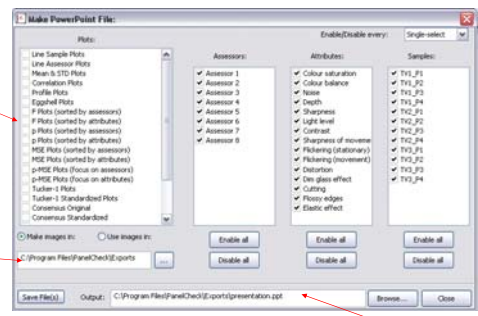
Choose file format

Define files location



Export of plots and results

SAVE IMAGES TO POWER POINT



Choose which plots to export

Define location of image files

Define location of PowerPoint presentation

