

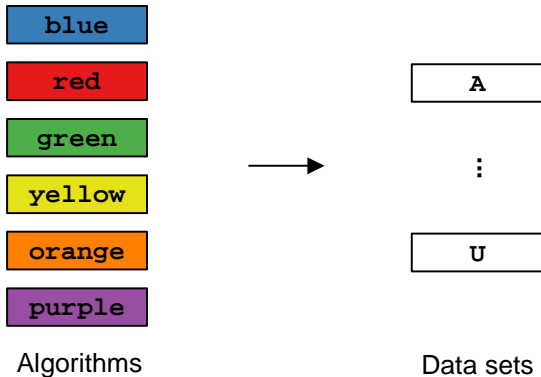
# Exploratory and Inferential Analysis of Benchmark Experiments

Manuel J. A. Eugster and Friedrich Leisch

Departement for Statistics  
Ludwig-Maximilians-Universität München

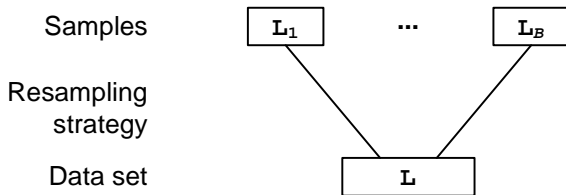
Statistical Computing 2008

# “Zehnkampf/Decathlon”



# Layers of abstraction

# Layer One: Setup



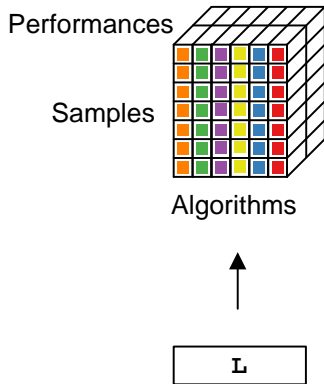
## Layer Two: Execution (1)

Performances

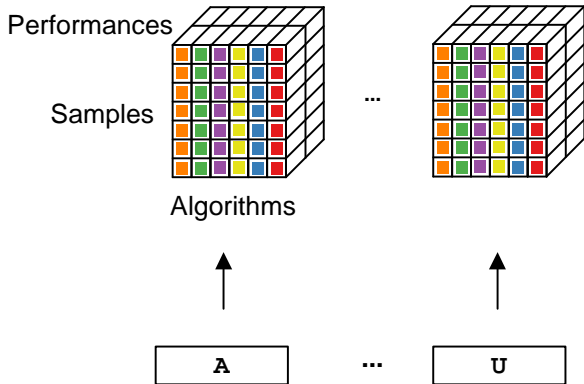
Samples

	⋮	
		-----
		0.8                      0.7
		8.3                      9.1
$P_2$		2.2                      1.9
		12.9                     12.3
		0.9                      1.1
		-----
		1.3                      1.3
		0.020                    0.011
		0.219                    0.350
$P_1$		0.372                    0.299
		0.014                    0.032
		0.386                    0.115
		-----
		0.299                    0.450
		-----
		$L_1$ ... $L_B$

## Layer Two: Execution (2)



## Layer Two: Execution (3)



## Layer Three: Analysis

**Exploratory:** get a better understanding of the benchmark experiment, “dig” for interesting information.

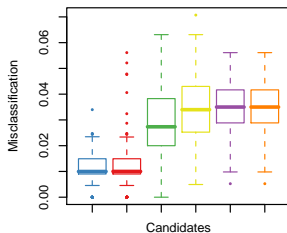
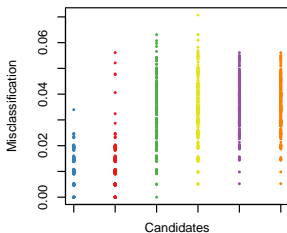
**Inferential:** test hypotheses of interest, infer a statistically correct order.



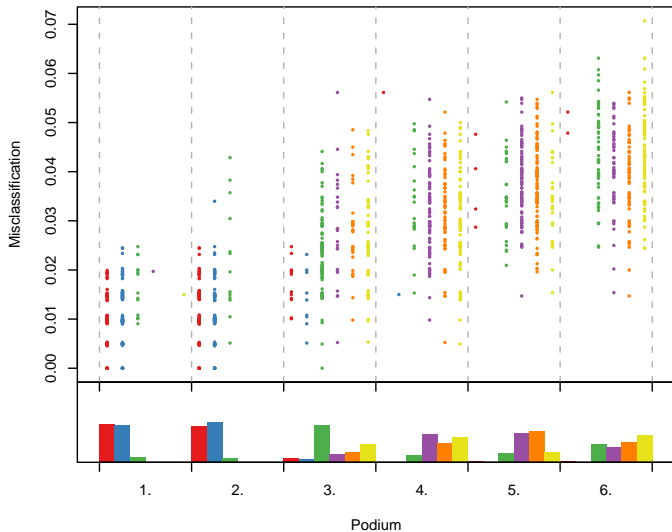
# **Analyses of benchmark experiments with one data set**

# Common exploratory tools

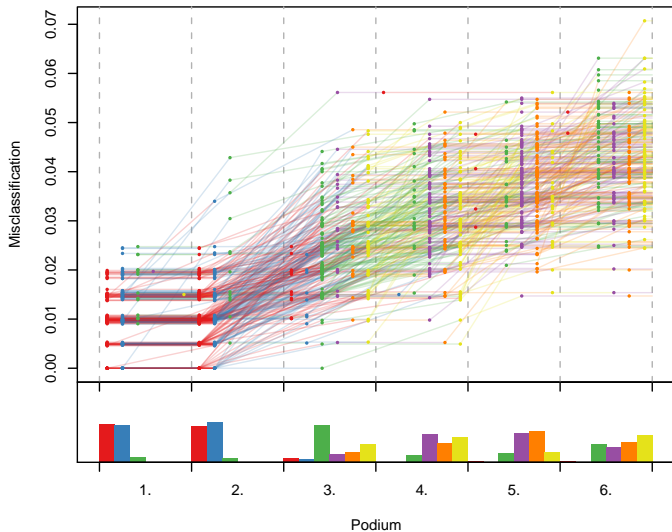
$\phi =$	Mean	SD	Median	Max
blue	0.0110	0.0059	0.0100	0.0340
red	0.0116	0.0080	0.0100	0.0561
green	0.0293	0.0123	0.0273	0.0631
yellow	0.0344	0.0118	0.0340	0.0707
purple	0.0352	0.0094	0.0350	0.0561
orange	0.0353	0.0094	0.0350	0.0561



# Benchmark experiment plot



# “Full” Benchmark experiment plot



# Inferential analysis

## Random block design:

$$\begin{aligned} \rho_{ij} &= \kappa_0 + \kappa_j + b_i + \epsilon_{ij}, \\ i &= 1, \dots, B, j = 1, \dots, (K - 1), \end{aligned}$$

with different assumptions on  $\kappa_j$ ,  $b_i$  and  $\epsilon_{ij}$ .

## Test problem:

$$\begin{aligned} H_0 &: \kappa_1 = \dots = \kappa_{K-1} = 0, \\ H_A &: \exists j : \kappa_j \neq 0, \end{aligned}$$

using parametric and non-parametric methods.

# Linear mixed effects model

## Assumptions:

$\kappa_j$  fixed effect,  $b_i$  random effect,

$$b_i \sim N(0, \sigma_b^2), \epsilon_{ij} \sim N(0, \sigma^2).$$

## Test problem:

Pairwise comparisons with Tukey contrasts.

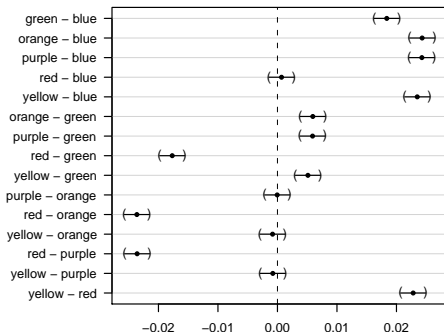
# Pairwise comparisons based on LME

## General Linear Hypotheses

Multiple Comparisons of Means:  
Tukey Contrasts

Linear Hypotheses:

	Estimate
green - blue == 0	1.837e-02
orange - blue == 0	2.431e-02
purple - blue == 0	2.427e-02
red - blue == 0	6.863e-04
yellow - blue == 0	2.349e-02
orange - green == 0	5.941e-03
purple - green == 0	5.899e-03
red - green == 0	-1.769e-02
yellow - green == 0	5.121e-03
purple - orange == 0	-4.188e-05
red - orange == 0	-2.363e-02
yellow - orange == 0	-8.202e-04
red - purple == 0	-2.359e-02
yellow - purple == 0	-7.783e-04
yellow - red == 0	2.281e-02



# Order relation and topological sort

In case of a significant difference between two algorithms we define a strict total order  $<$ , otherwise the algorithms are  $\approx$ -related.

## Pairwise orders:

red  $\approx$  blue, purple  $\approx$  orange, blue  $<$  green, ...

## Topological sort:

blue  $\approx$  red  $<$  green  $<$  orange  $\approx$  purple  $\approx$  yellow



# Overall order

## Performance measures $P_i$ :

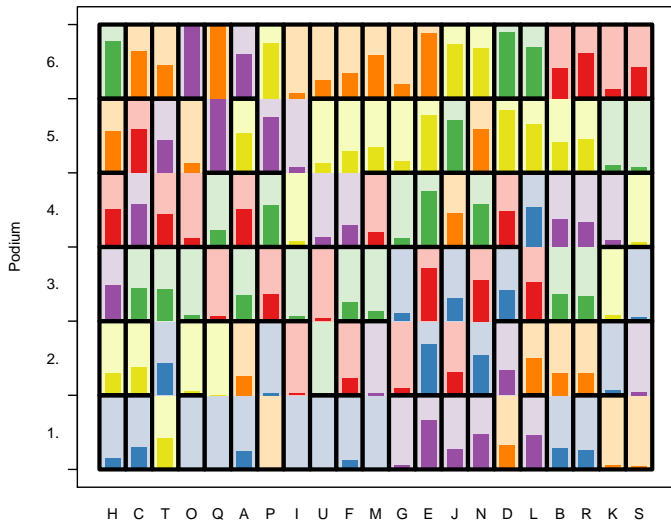
Mcl : blue  $\approx$  red < green < orange  $\approx$  purple  $\approx$  yellow

Time : red < purple < orange < yellow < green < blue

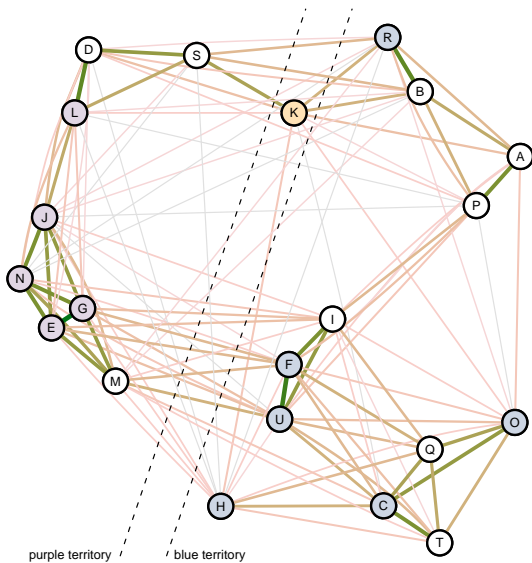
**Overall order:** Hierarchical order\*, Consensus ranking\*

# **Analyses of benchmark experiments with more than one data set**

# Benchmark survey plot



# Benchmark survey graph



## Further formal analyses

**Consensus:** overall order based on the set of order relations.\*

**Inference:** model the design with two experimental factors, their interactions and blocking factors at two levels.\*

**Overall:** sum up order relations based on different data sets and different performance measures.\*

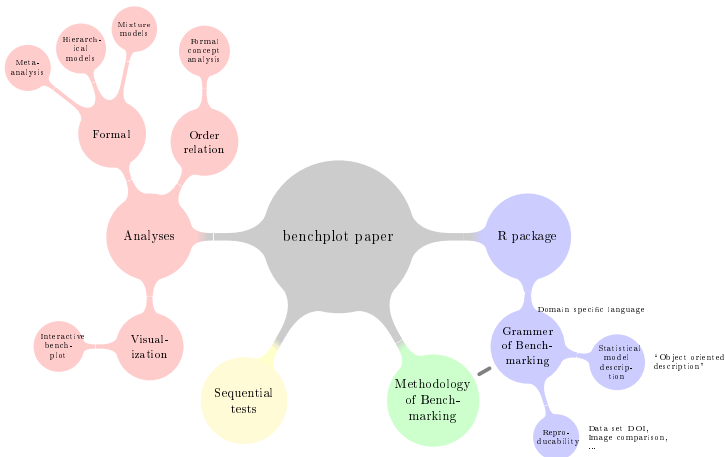
## Statistically correct order

Algorithms {blue, green, orange, red, purple, yellow}, data sets  $\{A, \dots, U\}$ , performance measures  $P_1 = \text{misclassification}$ ,  $P_2 = \text{computation time}$ :

blue < red  $\approx$  orange  $\approx$  green < yellow < purple

# Perspective

# Goals and future work





# References

*Bench Plot and Mixed Effects Models: First steps toward a comprehensive benchmark analysis toolbox.*

Manuel J. A. Eugster and Friedrich Leisch. Technical Report 26, LMU Munich. Accepted for the Compstat 2008-Proceedings in Computational Statistics.

(\*) *Exploratory and Inferential Analysis of Benchmark Experiments.*

Manuel J. A. Eugster, Torsten Hothorn and Friedrich Leisch. Technical Report 30, LMU Munich.

<http://www.statistik.lmu.de/~eugster>