# Assessment of the Heterogeneity of the Treatment Effect among Subgroups by Detecting Qualitative Interactions

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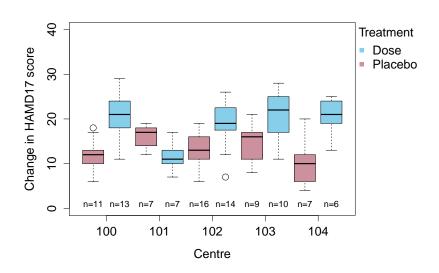
# Subgroup analysis of clinical trials

## Data are separated into various specific subgroups of subjects:

- Gender (male, female)
- Race (Asian, Black, White)
- Initial health status (blood pressure, severity of disease)
- ▶ Age (<50 years, ≥50 years)</p>
- Centre (multi-centre trial)

# Motivating Example - multi-centre clinical trial

Example based on "Analysis of Clinical Trials using SAS: A Practical Guide" (Dmitrienko et al., 2005)



# Detection of heterogeneity of treatment effects in subgroups

ICH guidance "Statistical principles for clinical trials" (1998)

"Marked heterogeneity may be identified by graphical display of the results of individual centres or by analytical methods, such as a significance test of the treatment-by-centre interaction"

#### Several terms

- treatment-by-centre interaction
- treatment-by-clinic interaction
- treatment-by-stratum interaction
- heterogeneity of treatment effects

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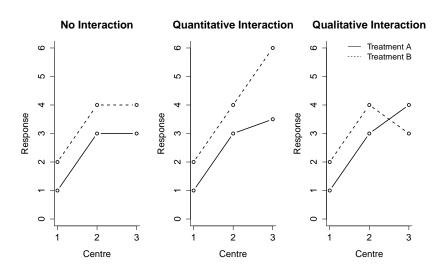
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# Types of interactions



## The model

#### **ANOVA** model

$$Y_{ijk} = \mu + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \epsilon_{ijk}$$
, where  $\epsilon_{ijk} \sim N(0, \sigma^2)$ 

#### Cell means model

$$Y_{ijk} = \mu_{ij} + \epsilon_{ijk}$$

- ▶ main effect A:  $\alpha_i = \mu_{i.} \mu_{..}$
- ▶ main effect B:  $\beta_i = \mu_{.i} \mu_{.i}$
- interaction effect:

$$(\alpha\beta)_{ij} = (\mu_{ij} - \mu_{..}) - (\mu_{i.} - \mu_{..}) - (\mu_{.j} - \mu_{..}) = \mu_{ij} - \mu_{i.} - \mu_{.j} + \mu_{..}$$

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product type interaction contrast as a direct (Kronecker) product of the two one-way contrasts ( $C_{AB} = C_B \otimes C_A$ ).

Example for a balanced design with I=2 and J=5: Define:  $\mu=(\mu_{Dose(1)},\mu_{Placebo(1)},\mu_{Dose(2)},\mu_{Placebo(2)},\dots,\mu_{Dose(5)},\mu_{Placebo(5)})$ 

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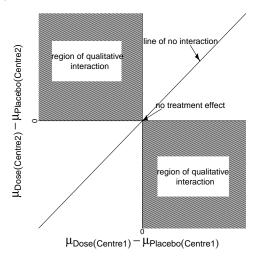
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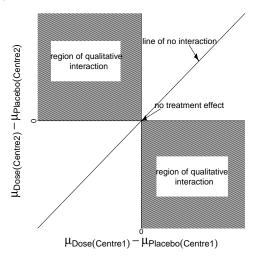
# Parameter space for treatment effects





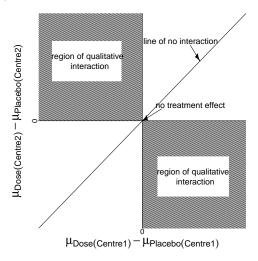
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- $\frac{\mu_{Dose(Centre1)}}{\mu_{Dose(Centre2)} \mu_{Placebo(Centre2)}} < 0 \Rightarrow \text{qualitative interaction}$

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- $\frac{\mu_{Dose(Centre1)} \mu_{Placebo(Centre1)}}{\mu_{Dose(Centre2)} \mu_{Placebo(Centre2)}} \ge 0 \Rightarrow \text{quantitative interaction}$
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# Simultaneous inference to test for qualitative interaction

Interest is in simultaneous estimation of I = 1, ..., L ratios of treatment effects

$$\gamma_I = rac{\mathbf{n}_{\mathbf{I}}' oldsymbol{\mu}}{\mathbf{d}_{\mathbf{I}}' oldsymbol{\mu}}$$

 $\boldsymbol{\mu} = (\mu_{Dose(1)}, \mu_{Placebo(1)}, \mu_{Dose(2)}, \mu_{Placebo(2)}, \dots, \mu_{Dose(J)}, \mu_{Placebo(J)})$   $\mathbf{n}_{1}'$  and  $\mathbf{d}_{1}'$  user defined contrast vectors.

The goal is to simultaneous test the *L* hypotheses:

$$H_{0_I}: \gamma_I \geq 0$$
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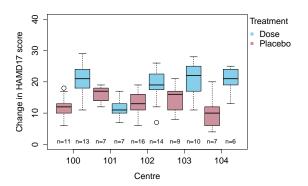
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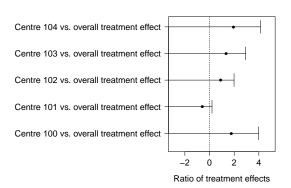
# Example - Results



	Df	Sum Sq	Mean Sq	F value	Pr(>F)
DRUG	1	888.04	888.04	40.07	<.0001
CENTER	4	87.14	21.78	0.98	0.4209
DRUG:CENTER	4	507.45	126.86	5.72	0.0004
Residuals	90	1994.38	22.16		

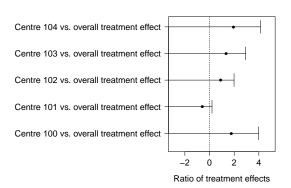
# Evaluated example using ratios of differences

Hypothesis	Estimate	adj. p-value
Centre 100 treatment effect/Overall treatment effect < 0	1.77	1
Centre 101 treatment effect/Overall treatment effect < 0	-0.58	0.187
Centre 102 treatment effect/Overall treatment effect < 0	0.90	1
Centre 103 treatment effect/Overall treatment effect < 0	1.35	1
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#### References



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