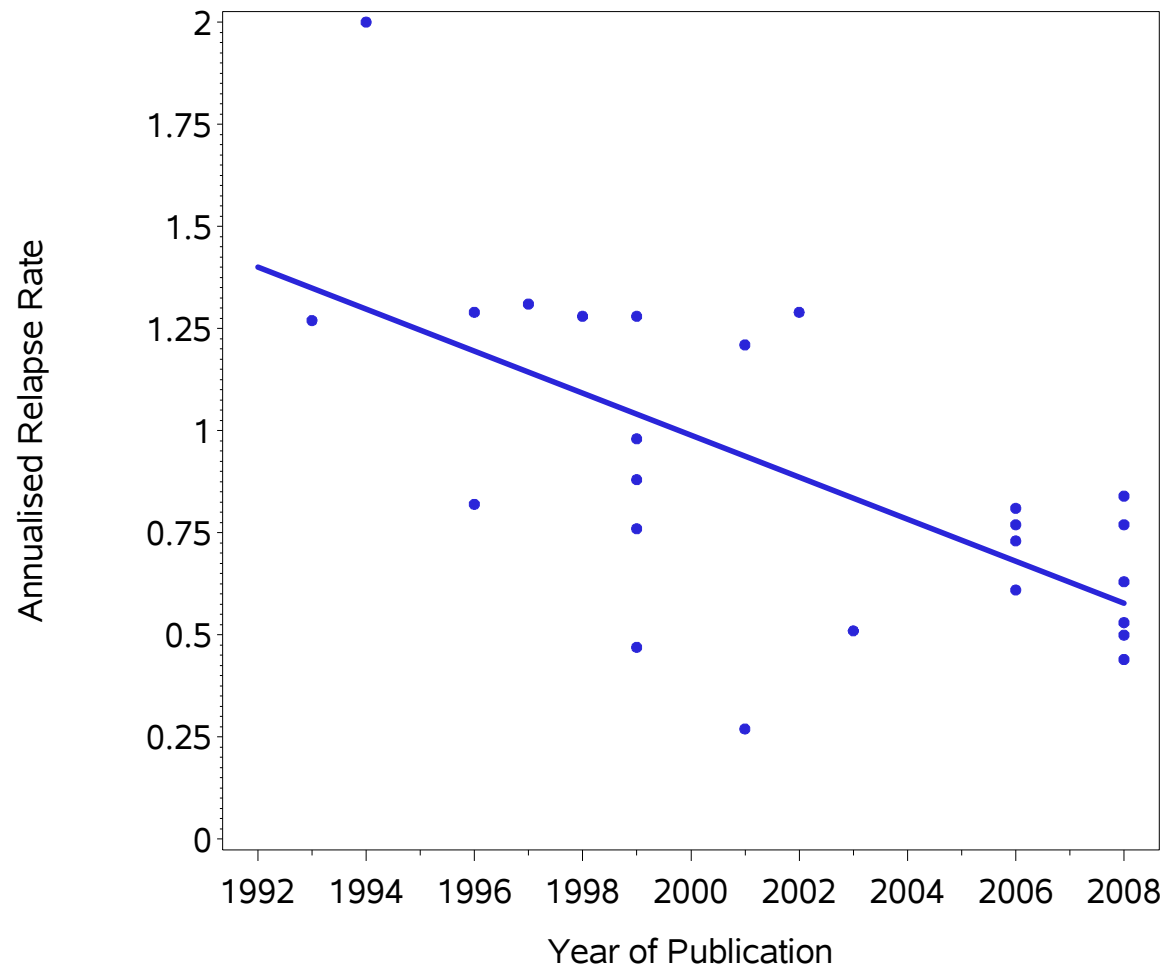

Blinded sample size reestimation with count data

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Motivating Example: Relapse Counts in Multiple Sclerosis



- **placebo ARR in RRMS** (Sormani et al, 2009)
- **trend over time and substantial variation**
- recently published placebo-controlled fingolimod trial: placebo ARR 0.7 expected in sample size calculation, observed 0.4

Poisson Counts, Rate Ratio and Wald Test

- **data:** D_T and D_C Poisson counts with ...
 - event rates λ_T and λ_C
 - follow-up times y_T and y_C
- **hypotheses:** $H_0 : \theta = 1$ versus $H_1 : \theta < 1$ with $\theta = \lambda_T/\lambda_C$
- **hypothesis test:** reject H_0 , iff $Z \geq z_{1-\alpha}$ whereby

$$Z = \frac{\log(\hat{\lambda}_T/\hat{\lambda}_C)}{\sqrt{1/D_T + 1/D_C}} \quad \text{with} \quad \hat{\lambda}_i = D_i/y_i$$

Sample Size Calculation for Poisson Counts

- **desired power** $1 - \beta$ and **relevant effect size** θ^*
- assuming same follow-up for all patients and 1:1 randomisation
- **overall event rate** $\bar{\lambda} = (\lambda_C + \lambda_T)/2$
- approximate **sample size** per group (see e.g. Ng & Tang, 2005)

$$n = \frac{1}{\bar{\lambda}} \frac{(1 + \theta^*)^2}{2 \theta^*} \frac{(z_\alpha + z_\beta)^2}{(\log \theta^*)^2}$$

Internal Pilot Study Design (Wittes & Brittain, 1990)

- **initial sample size estimation** $n_0 = n(\alpha, 1 - \beta, \Delta^*, \hat{\sigma}_0^2)$
 - significance level α , desired power $1 - \beta$, clinically relevant effect Δ^*
 - initial estimate $\hat{\sigma}_0^2$ of the nuisance parameter σ^2 (from other studies)
- **sample size review:**
 - after recruitment of $n_1 = \pi n_0$ patients (e.g., $\pi = 1/2$)
 - estimation of nuisance parameter $\rightarrow \hat{\sigma}^2$
 - sample size *re-estimation* $\hat{N} = n(\alpha, 1 - \beta, \Delta^*, \hat{\sigma}^2)$
 - * “restricted”: $n_2 = \max(n_0, \hat{N}) - n_1$
 - * “unrestricted”: $n_2 = \max(n_1, \hat{N}) - n_1$ (Birkett & Day, 1994)
- **final analysis**
 - estimation of treatment effect and hypothesis test
 - with all $n_1 + n_2$ patients

Sample Size Re-estimation and International Guidelines

- **ICH Guideline E9 (1998)**

The steps taken to preserve blindness and consequences, if any, for the type I error [...] should be explained.

- **CHMP Reflection Paper on Adaptive Designs (2007)**

Whenever possible, methods for blinded sample size re-assessment [...] that properly control the type I error should be used.

- **draft FDA guidance on adaptive designs (2010), Sec. V.B**

Sample size adjustment using blinded methods to maintain desired study power should generally be considered for most studies.

Blinded Sample Size Recalculation for Poisson Counts

- overall event rate $\bar{\lambda}$ estimated in blinded fashion from data of IPS

$$\hat{\lambda}_{1.} = \frac{D_{1T} + D_{1C}}{y_{1T} + y_{1C}} = \frac{D_{1.}}{y_{1.}}$$

similar to Gould's approach for binomial data, see Gould (1992), Friede & Kieser (2004)

- assuming constant event rates within treatment groups over the course of the study, all data available can be used
- plugging in the observed overall event rate $\hat{\lambda}_{1.}$ for $\bar{\lambda}$ in sample size formula leads to a new sample size estimate

Extension to Overdispersed Poisson Counts

- **overdispersion:**

- so far $Var(D_i) = \lambda_i y_i$

- now $Var(D_i) = \sigma^2 \lambda_i y_i$ with overdispersion parameter $\sigma^2 \geq 1$

- quasi-likelihood approach (McCullagh & Nelder, 1989)

- **test statistic** allowing for overdispersion: $Z^{(O)} = Z/\tilde{\sigma}$

- **sample size:** $n^{(O)} = \sigma^2 n$

Extension to Overdispersed Poisson Counts (cont.)

- various estimators for σ^2 proposed, e.g. method of moments estimator

$$\tilde{\sigma}^2 = \left(\sum_{i=T,C} \sum_{j=1}^{n_i} \frac{(D_{ij} - \hat{\lambda}_i)^2}{\hat{\lambda}_i} \right) / (n_T + n_C - 2)$$

- for the purpose of a blinded review the dispersion parameter σ^2 estimated in blinded fashion by

$$\tilde{\sigma}_{1.}^2 = \left(\sum_{i,j} \frac{(D_{1ij} - \hat{\lambda}_{1.y1.})^2}{\hat{\lambda}_{1.y1.}} \right) / (n_T + n_C - 1)$$

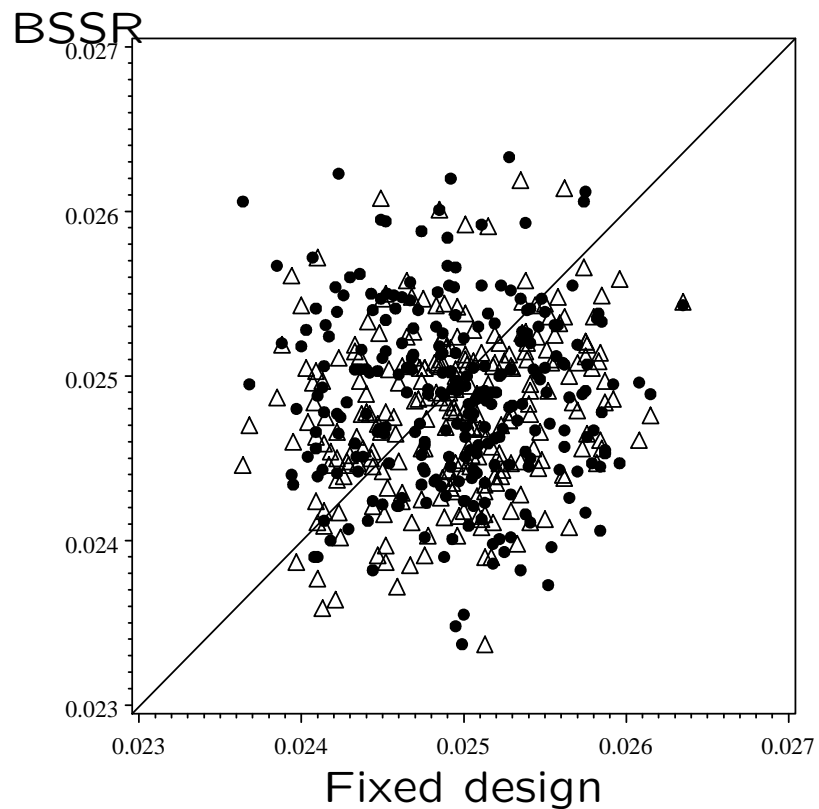
Simulation Study

- **motivated by trials in relapsing MS**
 - phase III: relapse counts (no or small overdispersion)
 - phase II: MRI lesion counts (large overdispersion)
- **number of simulated trials per scenario:** 100,000 for power (type I error rate) and 10,000 for sample size distribution
- **distributional assumptions:** Poisson and Negative Binomial (as an example for overdispersed counts)

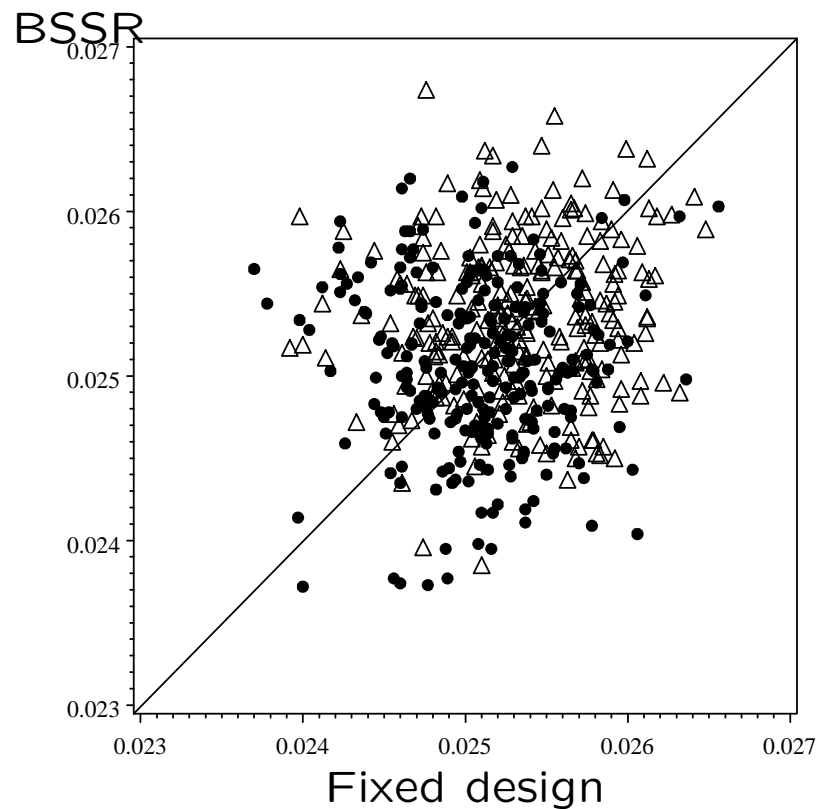
Simulation Study Setup

	Relapses	MRI lesions
Proportion π in IPS	0.2, 0.5	0.4
Overall event rate $\bar{\lambda}$	0.5, 0.51, ..., 1	5, 6, ..., 15
Dispersion parameter σ^2	1, 1.1, ..., 2	20, 25, ..., 40
<hr/> <u>Under the null hypothesis $H_0 : \theta = 0$</u>		
Required sample size N	100, 200, ..., 500	50, 75, ..., 150
<hr/> <u>Under the alternative $H_1 : \theta = \theta_a$</u>		
Assumed overall event rate $\bar{\lambda}_a$	0.75	10
Assumed dispersion σ_a^2	1	30
Assumed rate ratio θ_a	0.6, 0.75	0.5
Target power $1 - \beta$	0.90	0.80

Relapses: Type I Error Rate

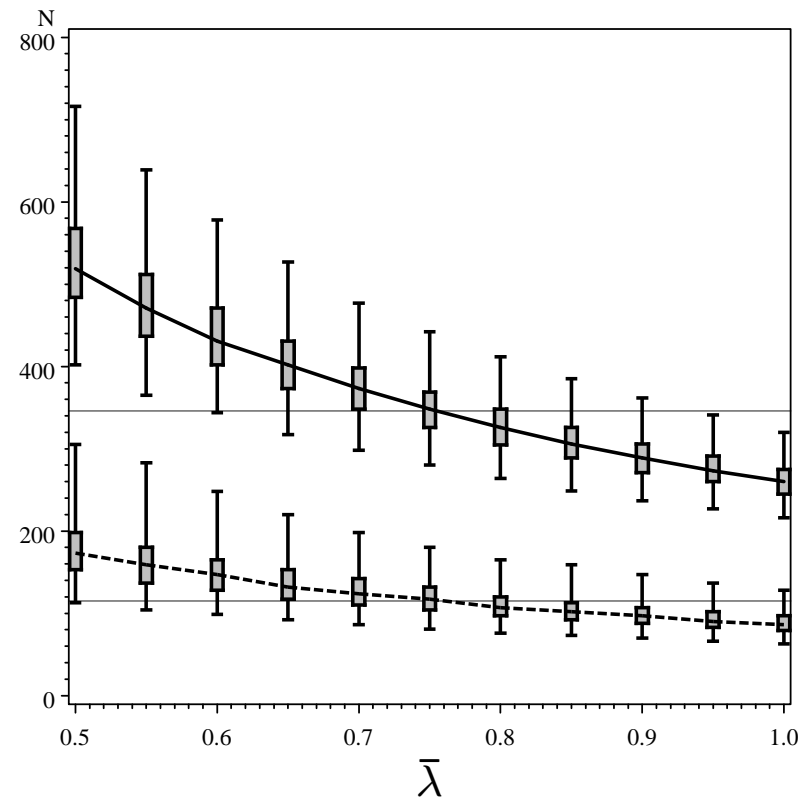
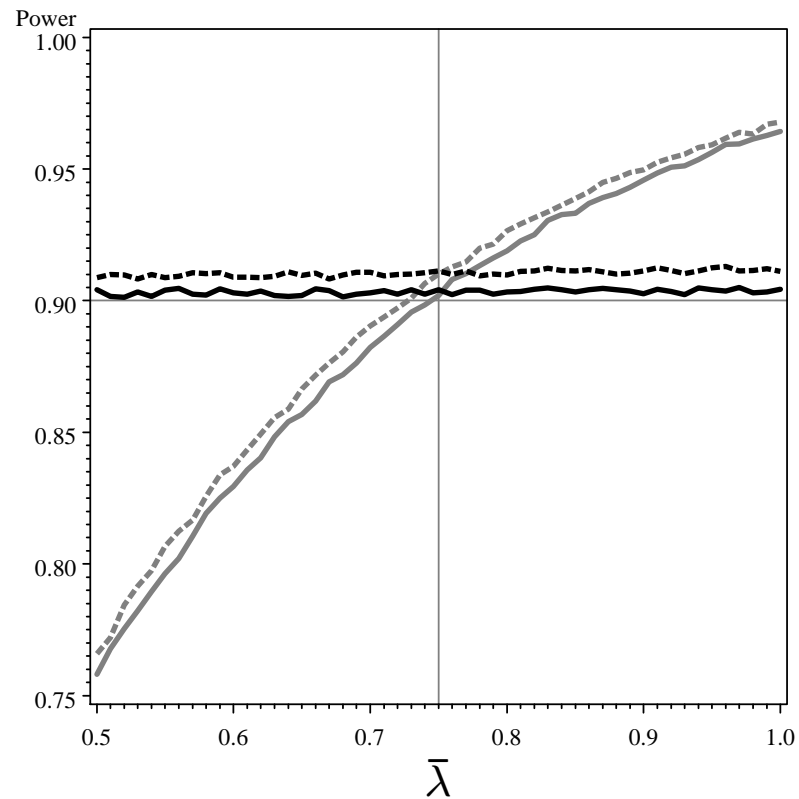


Poisson data with reviews including
20% (●) and 50% (△) of the patients



Reviews with 50% of the patients;
 $\sigma^2 = 1$ (●) and $\sigma^2 = 2$ (△)

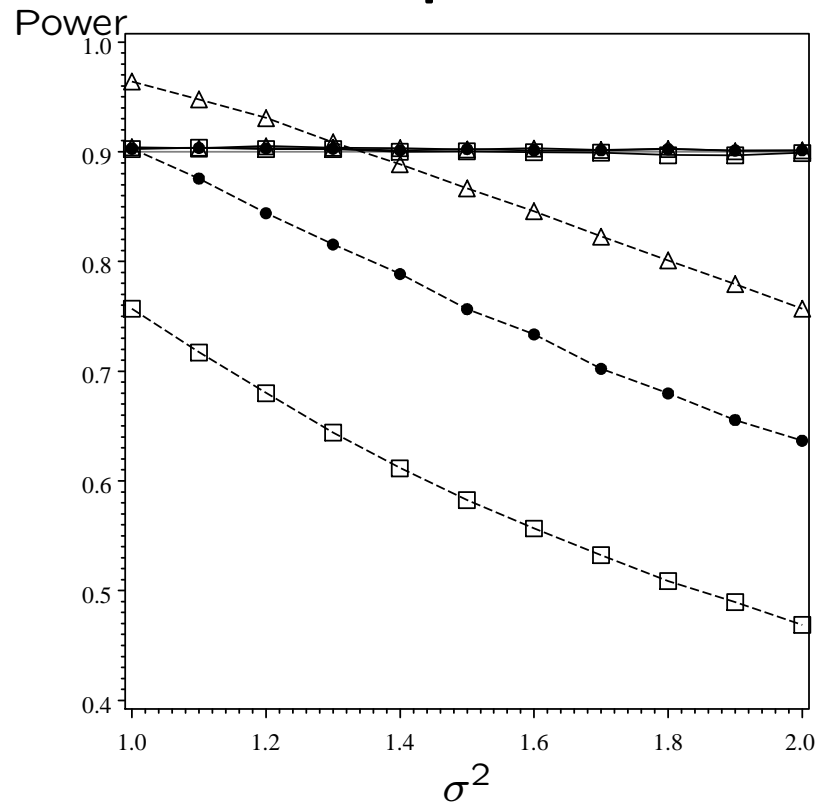
Relapses: Misspecification of the Overall Event Rate



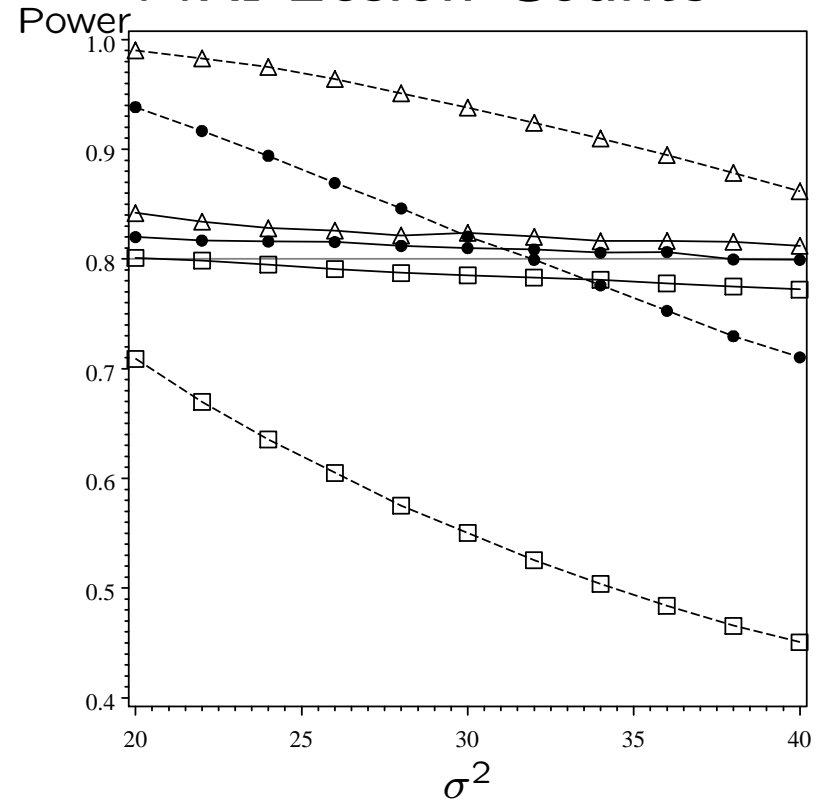
Poisson counts (no overdispersion); blinded reviews including 20% of patients in the internal pilot; rate ratios $\theta = 0.6$ (dashed) and $\theta = 0.75$ (solid); fixed design (grey) for comparison.

Misspecification of Overall Event Rate and Overdispersion

Relapses



MRI Lesion Counts



Likelihood Approach: Negative Binomial Distribution

- **likelihood-based inference** assuming a particular mixture distribution, e.g. negative binomial (Aban et al, 2009)
- Cook et al (2009) proposed a blinded procedure based on an **EM algorithm**
 - computationally more demanding than our approach outlined above
 - on the other hand, our approach does not utilize all information under sampling from negative binomial distributions
- **new approach**: assuming negative binomial distributions ML estimates of the overall event rate and of the shape parameter are derived in blinded review ignoring treatment groups

Non-inferiority Trials

- other indications: for instance exacerbation counts in asthma and COPD (Keene et al 2007, 2008)
- in asthma / COPD standard treatment exist and placebo therefore (at least long-term) unethical
- active controlled non-inferiority trials

Non-inferiority: Type I Error Rates and Power

Simulation study motivated by trials in COPD with average study sizes ranging from 700 to 1300 patients

Scenario		Type I	Power
ϕ	λ	error rate	
0.4	1	0.0253	0.7985
	1.5	0.0255	0.7984
	2	0.0269	0.8054
0.5	1	0.0227	0.8073
	1.5	0.0241	0.7984
	2	0.0243	0.8010
0.6	1	0.0235	0.8015
	1.5	0.0259	0.7947
	2	0.0259	0.7991

Conclusions

- type I error rate: similar to fixed design tests
- power robust against misspecifications of both overall event rate and overdispersion parameter
- blinded reviews fulfill regulatory requirements

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