

Regulatory Issues in Survival Analysis

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Disclaimer



The views expressed in this presentation are those of the speaker, and are not necessarily those of MHRA.



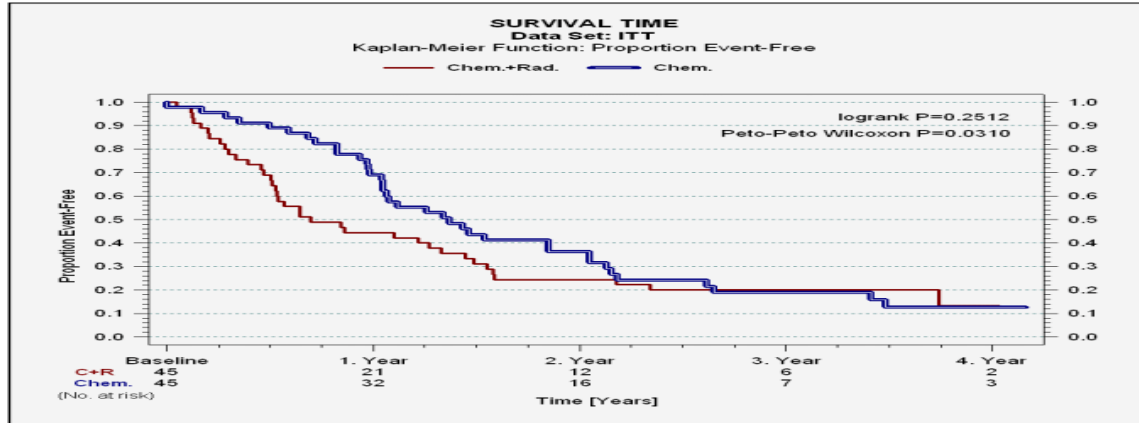
Main Issues



- Summary statistics
- Interim analyses
- Missing data
- Switching treatments



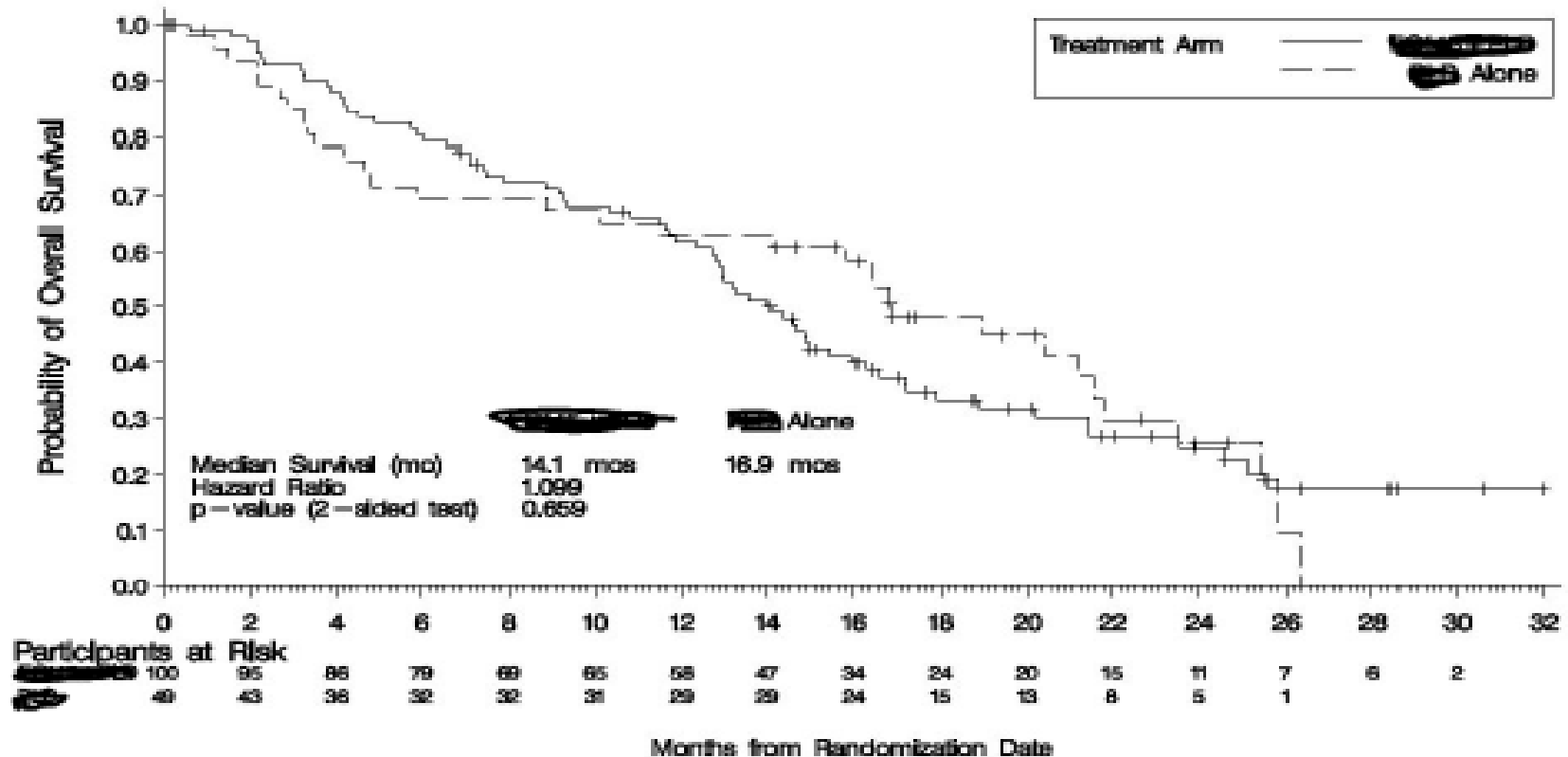
Summary Statistics



- Median time to event
- Percentage of subjects event-free at time t
- Hazard ratio



Non-proportional hazards



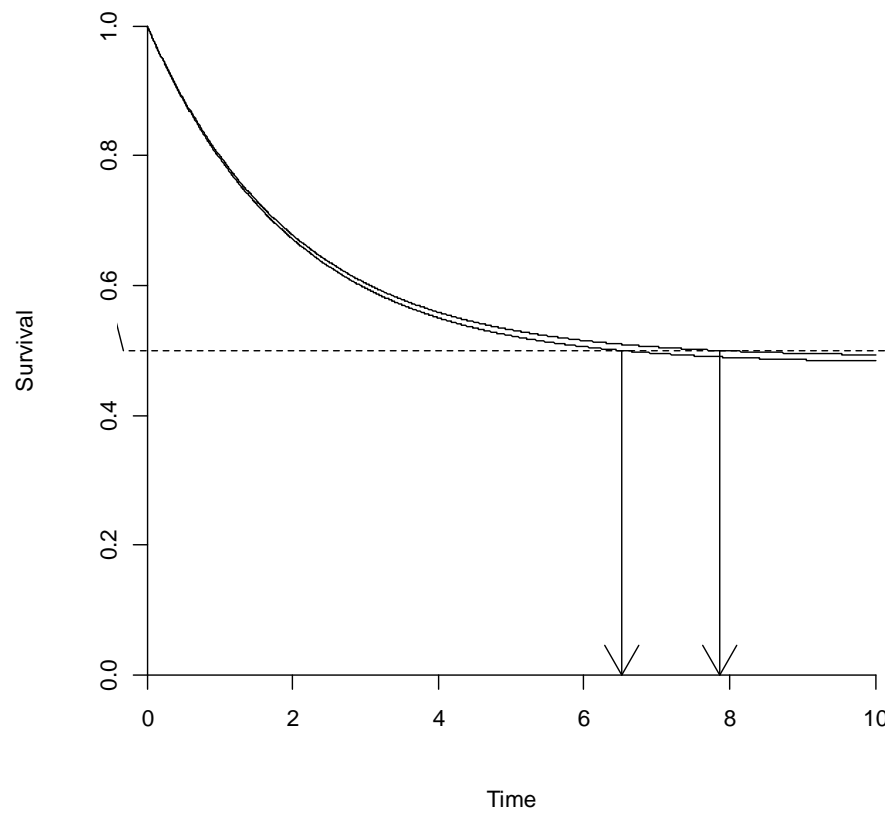
Proportional hazards



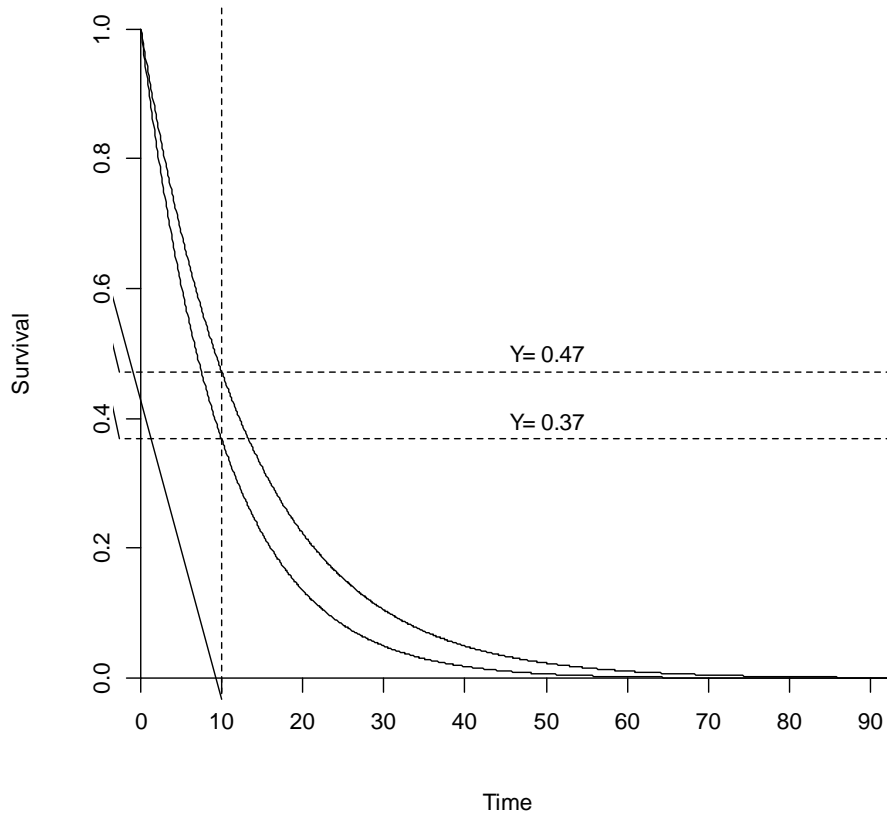
- Hazard ratio useful statistic – but not good enough on its own
- Can be misleading on its own – and can't be compared with other hazard rates without context. For example – hazard ratio of 0.5 – looks impressive???
- Halving risk from 0.9 to 0.45 more impressive than from 0.002 to 0.001 – the latter will hardly save any events...
- So more information is needed...



Slowly declining curves

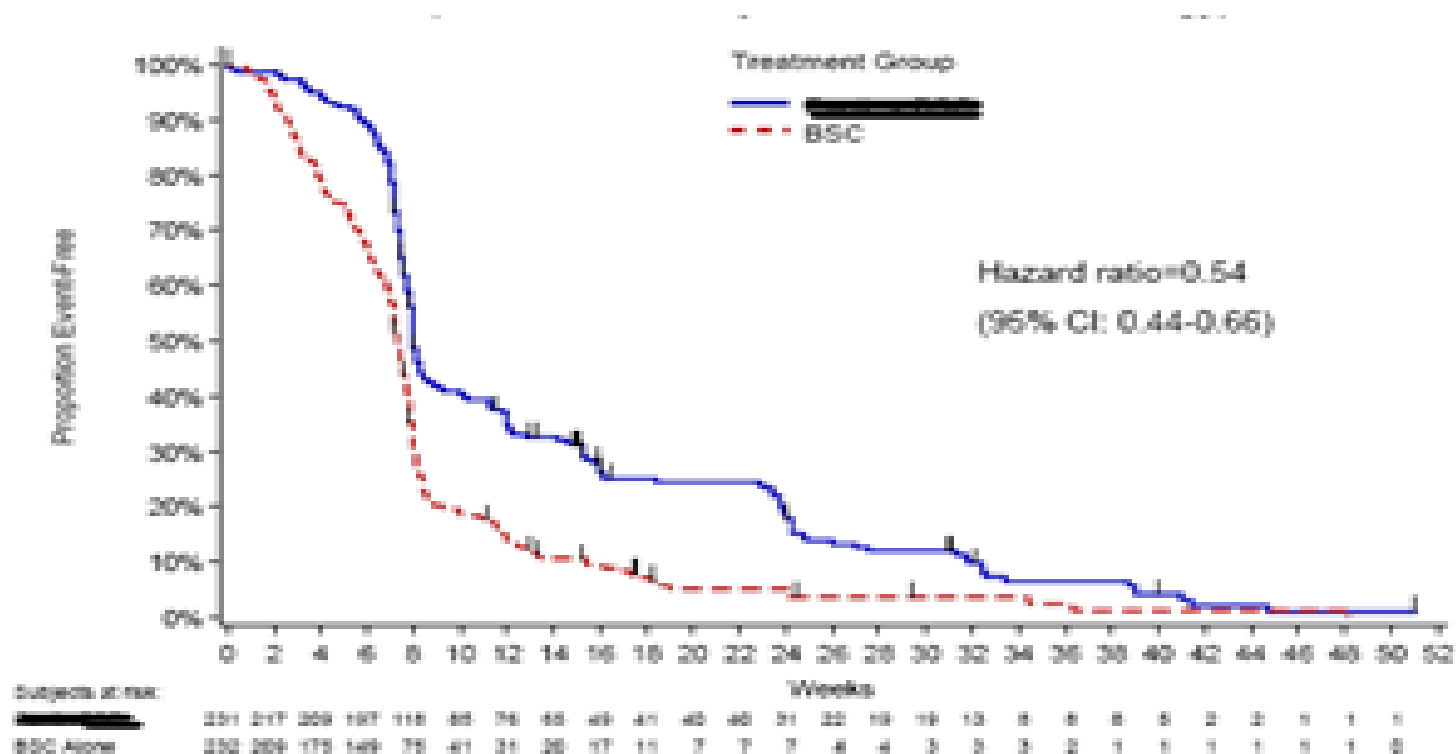


Quickly declining curves



Example

Kaplan-Meier Estimate of Progression-free Survival by Randomized Treatment (All Enrolled Analysis Set, Central Radiology)



Cardiovascular studies



- Typically primary endpoint is a composite of death and hospitalisation analysed using time-to-first event approach.
- However patients often have multiple hospitalisations as well as terminal events
- Each recurrent HF hospitalisation substantially increases the risk of death
- New approaches to analysis include joint frailty models and the win ratio.



Interim Analyses



- More of a concern in oncology than heart disease – much smaller studies
- Usually only 1 pivotal study
- Often open label
- If study stops – very limited data set for B/R assessment
- Results need to be as robust and convincing as possible



Interim Analyses



When planning - need to consider:

- No use stopping for PFS if OS is needed
- If any subgroups of interest will have enough representation
- If more follow up/patients are needed anyway – e.g. safety endpoints



Interim Analyses



- If study stops for success at interim – limited data for benefit-risk assessment
- Results at this point should be representative of whole population
- Fewer patients with long follow up - better than many patients with short follow up
- E.g. If 1000 patients are randomised and 250 events are required for the final analysis (25% maturity), then interim when 125 events have been observed on the first 500 patients randomised



Missing Data

- Often overlooked in time to event trials
- But censored patients considered to have similar survival to those continuing
- Important to know if censoring 'informative' or 'non-informative'
- Imbalance in informative censoring gives cause for concern



Non-informative censoring



- Censoring caused by the study design:
- Study finished and event not reached
- Planned 2-year follow-up only – event not reached
- Study planned for certain number of events
- Imbalance in non-informative censoring is not a concern – just reflects different efficacy of treatment (so usually good if more censoring in treatment group).



Informative censoring



- The fact a patient is censored gives information about the chances of having an event. Censoring not caused by the study design:
- Took rescue medication (maybe should count as event?)
- Trial withdrawal
- Moved away
- Patient lost to follow-up??



Informative censoring



- It is not possible to judge if censoring is informative – e.g. could decision to move away be based on progress of condition?
- Safest policy to assume all informative unless caused directly by trial design.
- Imbalance in informative censoring is cause for concern – could lead to bias



Sensitivity analyses

- Sometimes some patients are censored, but it might also be a reasonable approach to say they may have had an event
- e.g. rescue medication use, adverse event causes discontinuation
- Sensitivity analysis can provide reassurance that the censoring had no real influence on the results



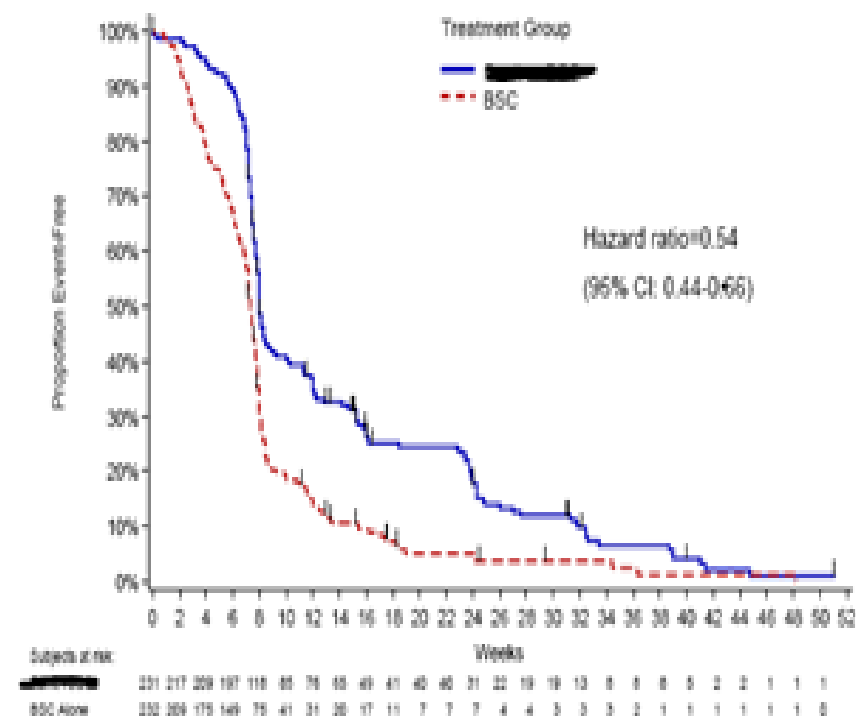
Interval censoring

- This can arise in trials where events can only be seen at visits (e.g. PFS). So we know the event occurred – but we don't know exactly when, only a range.
- Doesn't cause important bias as long as the visit schedule is the same for both groups and is strictly followed.
- If there are lots of unscheduled visits bias can be caused (especially in an open trial).
- A sensitivity analysis where all events assigned to next scheduled visit can correct this.

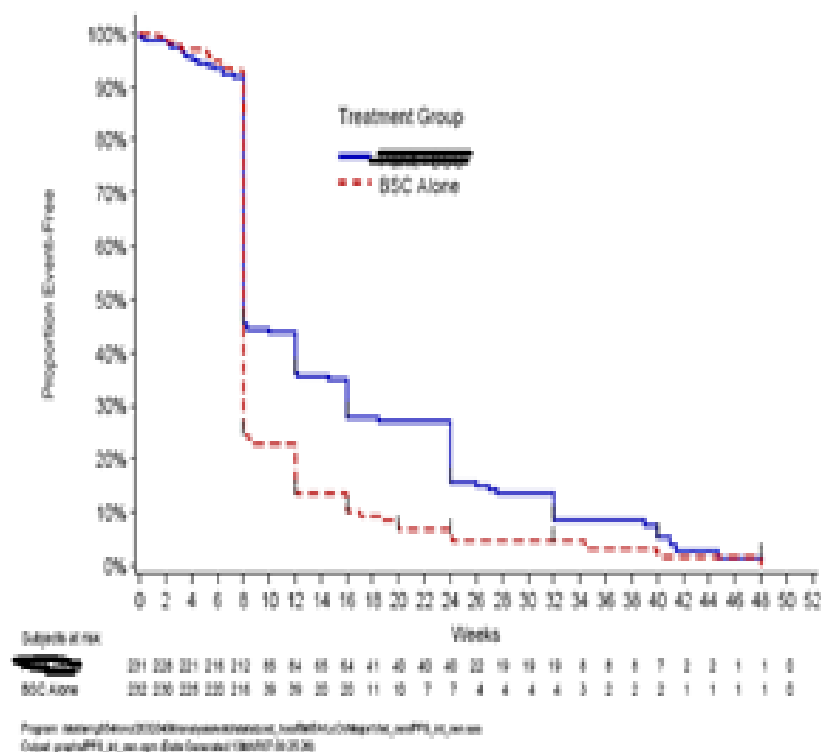


Example

Kaplan-Meier Estimate of Progression-free Survival by Randomized Treatment (All Enrolled Analysis Set, Central Radiology)



Kaplan-Meier Estimates of Progression-free Survival Imputing Radiologic Progression at the Closest Scheduled Assessment Time Point



Switching Treatments



- Mainly an oncology problem
- Often both PFS and OS are needed to demonstrate a positive B/R – however patients often treated with different treatments at PD:
 - Patients on control swap to investigational
 - Patients on both treatments go to next line therapy as per local guidelines
- No longer comparing just new versus control



Switching Treatments

- Ideally both OS and PFS show benefit of new drug regardless of switching
- But often the benefit in PFS is no longer seen in OS – switching usually blamed
- Modelling approaches to try and correct for this are of limited use - rely on unrealistic assumptions
- Improvement in PFS would be expected to create improvement in OS unless subsequent treatment very unbalanced (which needs explaining)



Switching Treatments

- Occasionally the opposite happens: More treatment effect on OS than PFS
- Could be the result of a very efficacious next line treatment
- Patients with shorter PFS swap earlier – making a bad treatment look good



Conclusions



- Important to consider shape of curve – summary statistics
- Interim analyses: limited but representative dataset
- Consider possible bias by informative censoring
- Inconsistencies between PFS and OS need to be investigated. Not enough to blame switching.

