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The Use of Surrogate Endpoints in Regulatory and HTA Submissions: Methodological Considerations, Current Status and Best Practice Recommendations



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Surrogate endpoints in HTA decision making: methodological considerations

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Conflicts and disclosures

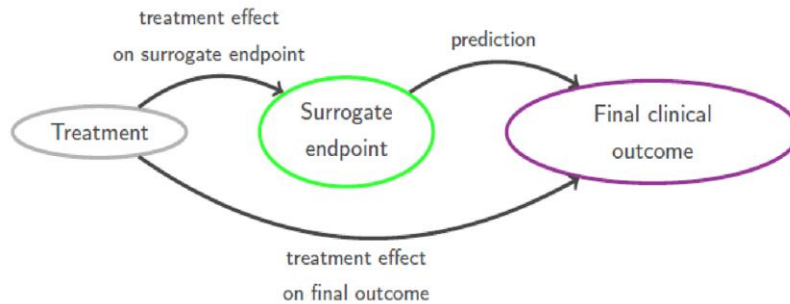
- I have served as a paid consultant, providing methodological advice, to NICE and Roche
- I have received research funding from European Federation of Pharmaceutical Industries & Associations (EFPIA) and Johnson & Johnson.
- I have received research funding from the [Medical Research Council \(MRC\)](#) , the National Institute for Health Research (NIHR) and EU Innovative Medicines Initiative (IMI).
- I have no financial or any other conflicts relating to specific products mentioned in my presentation

Surrogate endpoints: importance

- Modern trials of novel therapies bring a number of challenges for regulatory and HTA decision making, due to
 - trials of targeted therapies carried out in **small subsets** of the population
 - high success rates of new therapies **reduce number of events** (deaths) requiring long follow up time to measure effect on overall survival (OS)
 - This results in estimates of treatment effect on OS obtained with large **uncertainty** at early follow up.
- Regulatory agencies (EMA and FDA) carry out **conditional licensing** based on treatment effects measured on a short-term surrogate endpoint (PFS, TR).

Surrogate endpoints

Surrogate endpoints can be used to measure the effect of a new treatment **earlier** and with **higher precision** compared to OS.



Surrogate endpoint: A biomarker that is intended to substitute for a clinical (final) outcome. A surrogate end point is expected to **predict clinical benefit** (Biomarkers Definitions Working Group. *Clin Pharmacol Ther* 2001.)

Advantages of surrogate endpoints

- Surrogate endpoints are of interest in drug development process if they can be measured
 - less costly
 - less invasively
 - or require **shorter follow-up time**
 compared to a final clinical outcome.
- They are increasingly important in HTA decision making
 - at the early stages of drug development
 - when **conditional licensing** based on a biomarker
 - **evidence** on treatment effectiveness on a target outcome **limited**

Surrogate endpoints: examples

- A number of **putative** surrogate endpoints have been investigated, in particular in disease areas where the final clinical outcome requires long follow up time.
- Examples include:
 - **Progression free survival** (PFS) as surrogate to **overall survival** (OS) in advanced colorectal cancer
(Buyse et al. *Journal of Clinical Oncology* 2007, Ciani et al. *Journal of Clinical Epidemiology* 2015)
 - **Relapse rate** as surrogate for **disability progression** in relapsing remitting multiple sclerosis
(Sormani et al, *Multiple Sclerosis* 2010)



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Surrogate endpoint validation

- Validation at **three levels** (Taylor and Elston 2009, Ciani et al 2017)
 - level 3: evidence based on **biological plausibility** alone
 - level 2: evidence of a strong association between the surrogate and the final endpoint at the **individual patient level**
 - level 1: evidence showing that health technologies improving the surrogate also improve the final clinical outcome across many randomized controlled trials; **study level surrogacy**.
- For HTA decision-making, a modelling framework is required
 - to establish the **strength of the surrogate relationship** between the treatment effects on the surrogate and the final outcome
 - and to **predict** the likely treatment effect on the final outcome for the new health technology

Methods for surrogate endpoint validation

- Relying solely on **patient level association not sufficient** when evaluating surrogate endpoints, in particular when individual level association evaluated based on data from a single trial (*Fleming and DeMets, Annals of internal medicine 1996*).
- A **single trial validation** cannot guarantee that an association between effects confirmed based on individual data under **one treatment** will hold in other interventions.
- A **meta-analytic approach**, based on data from a number of trials to establish the association between the treatment effects on the candidate surrogate endpoint and on the final outcome is **more appropriate** for evaluation of surrogate endpoints.



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Methods for surrogate endpoint validation

- Putative surrogate endpoints are validated by estimating the **pattern of association** between the treatment effects on surrogate and final endpoints across trials.

**NICE DSU TECHNICAL SUPPORT DOCUMENT 20:
MULTIVARIATE META-ANALYSIS OF SUMMARY DATA
FOR COMBINING TREATMENT EFFECTS ON CORRELATED OUTCOMES AND
EVALUATING SURROGATE ENDPOINTS**

REPORT BY THE DECISION SUPPORT UNIT

22 October 2019

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Methods for surrogate endpoint validation

- Putative surrogate endpoints are validated by estimating the **pattern of association** between the treatment effects on surrogate and final endpoints across trials.
- **Bivariate meta-analysis methods**, that take account of the correlations between the average treatment effects on surrogate and final outcomes, are **suitable tools for modelling surrogate endpoints**
(Bujkiewicz et al, NICE DSU Technical Support Document 20; October 2019).
- **Individual patient data** hierarchical methods can be used to evaluate surrogate endpoint at both patients and study levels
(Buyse et al. Biostatistics 2000, Burzykowski et al. Journal of Royal Statistical Society A 2004).
- Bivariate methods for **summary data** can be used to model study-level surrogacy
(Daniels and Hughes, Statistics in Medicine 1997, Bujkiewicz et al. Stat Methods in Med Res 2018).

Novel methods for surrogate endpoint validation

- Bayesian meta-analytical methods to evaluate jointly **multiple surrogate endpoints**
(Bujkiewicz et al, Statistics in Medicine 2016)
- Bivariate **network meta-analysis** for surrogate endpoint evaluation.
(Bujkiewicz et al. Statistics in Medicine 2019)
 - allows modelling of surrogate relationships in greater detail
 - can help to disentangle information about surrogate relationships in data scenarios where such relationships vary across treatment contrasts
- Bayesian **hierarchical meta-analytic method** for modelling surrogate relationships that vary across **treatment classes** (Papanikos et al. Statistics in Medicine 2020)
 - allows for borrowing information about surrogate relationships between treatment classes whilst evaluating them within each treatment class

Discussion

- Use of putative surrogate endpoint – careful **evaluation** is needed.
- Rather than using pre-specified criteria, considering **a balance** between the **strength of the surrogate relationship** and the need for the decision to be made about the effectiveness of the new treatment may be important, in particular in the **high priority disease areas**.
- **Uncertainty** is key. The **strength of the surrogate relationship** will be reflected in the level of **uncertainty** around the predicted treatment effect on the final outcome.
- **Predicted** treatment effect (along with the uncertainty) may be used in HTA decision making framework.
- Novel methods may help improve predictions.



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