# Statistical Considerations in Using Meta-analysis for Regulatory Decision Making for Medical Devices

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## Device Validation

- Different stages of medical product development
  - Exploratory stage (development)
  - Pivotal stage (validation)
  - Post-market stage
- Device validation
  - Pivotal clinical studies
    - Prospective study: subjects prospectively enrolled
    - Retrospective study: subject samples retrospectively obtained with a prospective plan
  - Systematic review with meta-analysis
    - Quantitatively combine and integrate comparable studies and trials through a systematical review.

# Objective Performance Criteria (OPC) and Performance Goals (PG)

- Design Considerations for Pivotal Clinical Investigations for Medical Devices
  - An OPC needs to be carefully constructed from a prior meta-analytic review of all relevant sources, and a subject-level meta-analysis is preferred.
- Use of Real-World Evidence to Support Regulatory Decision-Making for Medical Devices
  - From a sufficiently relevant and reliable observational data source, a PG can be constructed using appropriate statistical methods, such as a subject-level meta-analysis.

# Benefits and Challenges of using Meta-analysis

- Benefits in using Meta-analysis
  - Better precision of pooled estimate of the effect than from a single study
  - Allow an examination of the existence and the causes of heterogeneity
- Challenges in using Meta-analysis
  - Quality assessment
  - Selection bias, publication bias
  - Heterogeneity across studies
  - Aggregation bias (summary level data vs. individual patient data)

# VIDAS BRAHMS Procalcitonin (PCT) Assay

- To help clinicians better predict a patient's risk of mortality or becoming sicker due to sepsis.
- To use PCT as a biomarker to help making antibiotic management decisions (initiation/cessation) in patients with lower respiratory tract infections and sepsis.
- Panel on 11/10/2016; Cleared in Feb, 2017
- Systematic literature reviews and meta-analyses of published randomized control trials were conducted.

510k summary <u>https://www.accessdata.fda.gov/cdrh\_docs/reviews/K162827.pdf</u> Panel meeting material <u>https://www.fda.gov/AdvisoryCommittees/CommitteesMeetingMaterials/MedicalDevices/Med\_5</u> <u>icalDevicesAdvisoryCommittee/MicrobiologyDevicesPanel/ucm515517.htm</u>

## VIDAS BRAHMS Procalcitonin (PCT) Assay

### • Algorithm (Device):

LRTI<br/>AB initiationStrongly<br/>discouragedDiscouragedEncouragedStrongly<br/>encouragedAB initiationdiscouraged0.10-0.250.26-0.50>0.50AB cessation:PCT  $\leq$  0.25 ng/mL or decrease > 80%>0.50

- 2 groups: PCT-guided therapy vs. standard therapy
- Endpoints: AB initiation, AB duration, mortality, complications, length of hospital stay
- Hypothesis: Lower AB use in PCT guidance group
  + no significant increase in safety endpoints

# Assessment of Study Quality

- Conduct quality assessment before any quantitative analysis.
- The quality assessment of the literature review is crucial to meta-analysis because the validity and reliability of meta-analyses depend on the quality of data extracted from the studies.
  - Cochrane Risk of Bias Assessment tool (Higgins and Green, 2011)
  - Downs and Black instrument (Downs and Black, 1998)
  - Chalmers quality scale, etc.

# Assessment of Study Quality

- Treatment assignment mechanism (RCT, non-RCT or single arm)
- Masking (blinding of treatment assignment to physicians, patients, and evaluators of outcome)
- Prospective data or retrospective data
- Pre-specified protocol and sample size
- Cross-over, drop-out, missing data
- Generalizability of study results to current US medical practice, etc.

 $\rightarrow$ Quality score: selection, interpretation, weighting factors in the effect estimation.

### Bias Assessment for LRTI (PCT test)

Author, year	Random sequence generation (selection b ias)	Allocation concealment (selection bias)	Blinding of participants and personnel (performance bias)	Blinding of outcome assessment (detection bias)	Incomplete outcome data (attrition bias)	Selective reporting (reporting bias)
Branche, 2015	+	8		+	+	+
Briel, 2008	÷	+	+	?	+	-+
Burkhardt, 2010	÷	÷	+	+	<b>,</b>	÷+
Christ-Crain, 2004	+	?		?	+	+
Christ-Crain, 2006	3	+	3	8	198	+
Corti, 2016	*	+			+	ŧ
Kristoffersen, 2009	*	÷			+	+
Long, 2011	?			+	( <b>+</b> )	÷
Schuetz, 2009	+	÷	+	?	+	+
Stolz, 2007	?	?	*	+	*	+
Verduri, 2015	*	ŧ		2	+	?

Low risk

high risk

# Generalizability using Non-US Studies

Meta-	Disease	Selected RCT	Sample size			
Analysis	type	Studies	РСТ	Cntrl	US SILES	
Study- Level	LRTI 11 RCTs		2040	2050	1 (year 2015) PCT: n=151 Cntrl: n=149	
	Sepsis	10 RCTs	1735	1754		
Patient- Level	LRTI	LRTI 13 RCTs		1606		
	Sepsis	5 RCTs	287	311	1 in Stolz 2009	

# Selection Bias

- Publication bias: studies with insignificant results or poor outcomes are typically not published.
- Approaches to minimize the selection bias
  - Two reviewers perform the literature search and data extraction independently.
  - Redact the study outcomes from abstract, text, etc.
  - Mask author names, affiliations, journal name, etc.
  - Pre-define the inclusion and exclusion criteria
    - E.g. Randomized control trial

# Funnel Plot

- A descriptive approach for evaluating if selection bias is present (Sterne and Harbord, 2004).
- X-axis: treatment effect Y-axis: precision of effect size estimate
- Statistical test (Egger et al. 1997; Harbord 2005; Begg & Mazumdar, 1994).



# Funnel Plots (PCT test)



# Heterogeneity Across Studies

- Heterogeneity is inevitable in a meta-analysis (Higgins 2003).
- Clinical heterogeneity
  - Study populations (enrollment criteria), endpoints, length of follow-up, treatment arm, control arm, available data, device used in studies, etc.
- Statistical heterogeneity
  - exists when the true effects being evaluated differ between studies.
- Cochran's  $\chi^2$  or Q (Higgins and Thompson 2002; 2003)

# Forest Plot of OR: Antibiotic Initiation, LRTI



#### Figure 7: Antibiotic initiation (fixed effects model)

### Different Devices for PCT Measurement

- LRTI (study level)
  - 2 out of 11 studies used <u>VIDAS BRAHMS PCT</u>
  - 9 out of 11 studies used BRAHMS PCT sensitive Kryptor
- Sepsis (study level)
  - 1 out of 10 studies used <u>VIDAS BRAHMS PCT</u>
  - 2 out of 10 studies used VIDAS BRAHMS PCT as one of multiple assays
  - 5 out of 10 studies used BRAHMS PCT sensitive Kryptor
  - 2 out of 10 studies used BRAHMS PCT LIA

# Different Cutoffs in Guidance Algorithms

### Algorithm (Device):

LRTIStronglyDiscouragedEncouragedStronglyAB initiationdiscouraged--encouraged<0.10</td>0.10-0.250.26-0.50>0.50

LRTI AB cessation: PCT  $\leq$  0.25 ng/mL or decrease > 80%

Sepsis AB cessation: PCT  $\leq$  0.5 ng/mL or decrease > 80%

# Different Follow-up Times and Rates

- Follow-up time is different across studies: ranges from 5 days, 1 month to 6 months.
- Follow-up rate varied across studies:
  - LRTI: range was 83% to 99% with 1 study unreported
  - Sepsis: range was 67% to 99% with 4 studies unreported

# Summary Level analysis (Aggregation Bias)

- Meta-regression using summary level data (aggregate data) can be subject to aggregation bias (ecological fallacy, Berlin et al., 2002).
- The phenomenon that a relationship across studies does not reflect the relationships within studies (Harbord & Higgins, 2008; Higgins, Thompson, Deeks, & Altman, 2002)

# Aggregation Bias



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# Patient Level Analysis

- Individual patient-level data (IPD)
  - Whether patient characteristics are related to treatment/outcome
  - Controlling for the covariate effects (confounding risk factors, baseline characteristics)
- IPD is considered as a gold standard approach
- But NOT a solution

## Verification of Meta-analysis

- Compare IPD analysis to the summary-level analysis if possible (Fortin et al, 1995; Olkin and Sampson, 1998)
- Predict the results for the Nth study from a metaanalysis of the first N – 1 studies (Simon, 1999; Pennello and Thompson, *J Biopharmaceutical Statistics*, 2008)

# Summary

- An opportunity to combine and integrate comparable studies of the device identified through systematic review.
- Many challenges to be overcome for a metaanalysis result to be interpretable and generalizable.

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