Process issues and data abstraction

Lecture #5

CHSC 7362 A01 Systematic reviews and meta-analysis





Literature search

- Goal of search strategy:
 - sensitive enough to capture all relevant citations
 - **specific** enough to be feasible





Study selection

- Goal of Title/ Abstract screening phase (Level I Screening):
 - Exclude all irrelevant citations (e.g. obviously doesn't meet the PICOS criteria)
 - Include any citations that may be relevant
 - When in doubt or it is unclear then include for full-text screening
 - Note:
 - citations may present little details pertaining to the PICOS (e.g. details of randomization, age range, severity of disease, all outcomes in the study, etc.).
 - not have abstracts or key words (e.g. title only)
 - be published in a non-English language



Study selection

- Goal of Full-text screening phase (Level II Screening):
 - Exclude all citations of studies that don't meet all the PICOS elements (e.g. doesn't report outcomes of interest)
 - Exclusions must be classified (preferably using a hierarchy)
 - Example of reasons for exclusion:
 - **Population(s)** not of interest to this review
 - Intervention(s) not of interest to this review
 - Comparator(s) not of interest to this review
 - No **outcome(s)** related to this review reported
 - Setting(s) not of interest to this review
 - Study design(s) not of interest to this review
 - **Publication type(s)** not of interest to this review
 - Language of publication not English



	Literature Search Results (Database)	
	🕵 01 - Medline (Ovid)	(0)
	🕵 02 - Embase (Ovid)	(0)
	🕵 03 - Central (Wiley)	(0)
	□ Level I Screening	
	🕵 01 - Duplicates	(0)
	9 02 - Titles/ Abstracts - Deduped	(0)
	□ Level II Screening	
5	🕵 01 - Full-text screening - All	(0)
	02 - Awaiting assessment - English	(0)
5	03 - Awaiting assessment - Non-English	(0)
Ū	04 - Unavailable via library services	(0)
	■ Excluded Studies	
Š	🕵 01 - Excluded - All	(0)
	🕵 02 - Excluded - Population	(0)
5	🕵 03 - Excluded - Intervention/ Control	(0)
¥	🕵 04 - Excluded - Outcomes	(0)
1	🕵 05 - Excluded - Study design	(0)
Ś	🕵 06 - Excluded - Settings	(0)
_	🕵 07 - Excluded - Timing	(0)
	□ Included Studies	
	🕵 01 - Included - All	(0)
	🕵 02 - Included - Primary reports	(0)
	\$ 03 - Included - Secondary reports	(0)
	Background Papers	
	🕵 01 - Background citations	(0)
	9.02 - Systematic review citations	(0)





How important is the data extraction stage?

- How often do mistakes occur in data extraction in published reviews?
- Do mistakes decrease with reviewer experience?
- How can we decrease the incidence of mistakes?



Do mistakes get past peer- & editorial-review?

- Type of discrepancies
 - Abstract-text discrepancies
 - Within-the-full-text discrepancies
 - Text-figure discrepancies
 - Text-table discrepancies
 - Multiple discrepancies

Puljaka, J Clin Epidemiol, 2020



Mistakes & review experience

- Retrospective study re-extracted data from 34 Cochrane reviews
- Found at least one error in 59% of reviews
- Types of errors:
 - misinterpretation of reported data
 - incorrect calculations made when converting data in primary articles into data required for the review 59%
- All errors led
- Conclusions noulling and corrections

Jones, J Clin Epidemiol, 2005

y effect estimates



Mistakes & review experience

- Retrospective study compared results of 3 binary outcomes from published reviews compared to the authors own published review
- Found errors ranged from 8% to 42% (depending on outcome and review)
- Differences in pooled effect estimates were small (RR 0.01 to 0.05)

Carroll, BMC research notes, 2013



Mistakes & review experience

- Prospective, cross-sectional study on reviewer accuracy and efficiency of data extraction
- High, but similar, error rates across the various levels of reviewer experience (28% to 31%)
 - Errors of inaccuracy (14 to 18%)
 - Errors of omission (11 to 16%)
 - No significant differences in error rates or accuracy of meta-analysis results between groups
- However, time required for extraction tended to decrease with experience

Horton, J Clin Epidemiol, 2010



What can we do to decrease mistakes

- Randomized trial compared the frequency of errors from single vs. double data extraction
- Single data extraction resulted in more errors

(RD = 22%, P = 0.02)

- No substantial difference between methods in effect estimates for most outcomes
- Average time spent for single data extraction was less

(RD = 36%, P = 0.003)

Buscemi, J Clin Epidemiol, 2006



Example statement:

- 'The data extraction form will be pilot tested on a sample of
 - three trial publications. Data from trial reports will be
- extracted independently by two reviewers with disagreements
 - resolved through consensus, or by a third reviewer'



Components of Data Form

- Plan... Plan... Plan... Careful think and plan
- How much information to collect:

overly vs. insufficiently detailed

• General rule of thumb:

'if you extract it then it has to have a place in the final review... so where is that place?'

Logical to entry into RevMan (e.g., copy/ paste)



Plan... Plan... Plan...

• Decide what you want to do before you start...

before you start doing anything

- Sketch it out on paper or electronically
- Pilot test the form (e.g., 3 5 studies)
- Document changes between the protocol and conducting the review (e.g., new outcomes)



Generic Data Extraction Items

- I. Coder Information
 - Study ID

II. Publication

- First author, pub year
- Funding source
- III. General Study Characteristics
 - Inclusion/exclusion criteria
 - Setting
- IV. Baseline Participant Characteristics
 - Number of participants randomized/ analyzed

- V. Intervention
 - drug name, dose
- VI. Outcomes
 - Primary outcome(s)
 - Secondary outcome(s)
 - Adverse outcome(s)
- VII. Risk of Bias
 - Sequence generation
 - Allocation concealment
 - Blinding
 - Etc.

VIII. Notes



A. PICO-based

- **B. Determine volume** of included studies and expected data available
- C. Determine data types required for comparisons
- **D. Determine data required** for subgroup analyses
- E. Determine data required for risk of bias assessment



- A. PICO-based:
 - 1. Information on Participants (e.g., age, weight, height)
 - 2. Information on Interventions (e.g., dosage, intervals)
 - 3. Information on Comparisons (e.g., placebo, intervention)
 - 4. Information on Outcomes (e.g., primary, secondary, AE)



- B. Determine volume of included studies and expected data available
 - Scenario #1:
 - 2 included trials
 - 3 outcomes per trial

• Scenario #2:

- 100 included trials
- 2 primary outcomes per trial
- 5 secondary (surrogate) outcomes per trial
- All adverse events described in all trials



- C. Determine data types required for comparisons
 - Most common:
 - Dichotomous (e.g. n/ N)
 - Continuous (e.g. mean ± SD)
 - Special situations
 - Categorical
 - Ordinal
 - Counts and rates
 - Time to event (e.g. Survival analysis)
 - Computed effect sizes (e.g. MD, 95% CI)



Participants

Patient Baseline Demographics:

	Intervention 1	Intervention 2
Age (yr)		
$Mean \pm SD$	±	±
Body weight (Kg) Mean ± SD	±	±
Height (cm)		
Mean ± SD	±	±
$BMI (Kg/m^2)$		
$Mean \pm SD$	±	±
Sex		
n		



Interventions/ comparators

	Intervention 1	Intervention 2	Intervention 3	Intervention 4
Classification	Starch	Crystalloid		
Туре	6% HES (130/0.4)	0.9% NaCl		
-	maximum dose	maximum dose		
Rate of	50ml/kg body	50ml/kg body		
administration	wt/ day then	wt/ day then		
	0.9% saline	0.9% saline		
Administration trigger	Hypovolemia	Hypovolemia		
Additional interventions				
Number randomized	3500	3500		



Outcomes

Primary outcome measures:

	Intervention (1)	Intervention (2)	Intervention (3)	Intervention (4)
Mortality	n = 597	n = 566	n =	n =
RIFLE categories				
Risk (R)	n=1788	n = 1912	n =	n =
Injury (I)	n = 1130	n = 1253	n =	n =
Failure (F)	n = 336	n = 301	n =	n =
Loss (L)	n =	n =	n =	n =
End-stage (E)	n =	n =	n =	n =
Creatinine (post-treatment)				
Mean ± SD	$112.11 \pm$	102.11 ±	±	±
	(n = 3260)	(n = 3283)	(n =)	(n =)







- Make sure you know what outcome you are measuring
- Mortality...

are we extracting how many died or how many are still alive???

- May experience difficulties with clearly identifying numbers:
 - poor reporting
 - provided in graph
 - percentage (%)
 - per protocol, not ITT



- Per protocol analysis:
 - Number analyzed based on patient compliance and lack of protocol violations.
 - Leads to biased results not particularly reflective of the 'truth'
- Intention-to-treat analysis:
 - Number analyzed based on initial treatment assignment and not on the treatment eventually received
 - Everyone randomized is analyzed, even with missing data
 - ITT analysis avoids misleading artifacts and biases (e.g., non-random attrition)



- Modified intention-to-treat analysis:
 - Same as intention-to-treat analysis but definition varies by study description
 - Commonly include all randomized participants that received at least one dose of intervention and had at least one post-treatment measure of efficacy or safety.



Continuous outcomes

 Outcomes that can take any value in a specific range – numerical or ordered categories

(e.g., weight, height, length of hospital stay)

- Is the scale validated (e.g., age, BP, VAS)?
- Is there a measure of variance (e.g., SD, SE, CI, P)?
- May experience difficulties with clearly identifying numbers:
 - poor reporting
 - provided in graph
 - per protocol, not ITT
 - measure of variance is not SD (convert first)







Categorical (nominal) outcomes

- Participants are classified into two or more categories with no intrinsic ordering to the categories (e.g., male, female)
- If clinically relevant, the data can be combined to form one group...
- Regardless of clinical relevance, data can be extracted separately to allow for subgroup analyses and in-between group comparisons



Ordinal outcomes

- Participants are classified into categories with a natural order (e.g., disease severity)
 - Short: small number of categories

(disease severity: mild, moderate, severe)

• Long: larger number of categories

(e.g., risk assessment: low, low-moderate, moderate,

moderate-high, high, very high)

• How to analyze:

Short \rightarrow often meta-analyzed as **binary data**

 $Long \rightarrow$ often meta-analyzed as continuous data

All cases ightarrow if in doubt ightarrow consult a statistician



Count of events

- Events that can happen more than once to the same individual (e.g., MI, stroke, headache)
- Example: 100 reported cases of MI were reported during a study of 100 individuals... with 20 people each having 5 MI during the follow-up period
- If we assumed it was dichotomous data, then 100/100 (100%) of population had MI... while the truth is only 20/100 (20%) of population suffered from MI (unit-of-analysis error)
- Analyzing counts of events... statistician





Time-to-Event Data

- Analysis of whether the event occurred and when
- 'Survival data' in stats (e.g., mortality, recurrence)
- Can sometimes be analyzed as dichotomous
- Hazard ratio analysis most appropriate



Time-to-Event Data (Kaplan Meier Curve)



Special situations

- Handling of missing data from included studies
 - Imputing data
 - Data from figures
 - Data from other sources
- How to reduce errors and bias associated with data extraction
 - Clear written guidelines (SOP) modified as changes occur
 - Data verification (double, independent data extraction vs. checking extracted data)





Figure 3. Visual analog scale values for anxiety. True intervention differs from sham intervention significantly (p < 0.001).

Handling of missing data

- Imputation of data
 - From the same study (e.g. P-value \rightarrow SD)
 - From other studies in the same review

(e.g. SD from another included study)

- From other sources

 (e.g. Probability of event from other sources)
- Data from figures
- Data from other sources (e.g. Unpublished data presented in another review)







Introduction to RevMan 5



Structure in RevMan

- Comparison
 - > Outcome

Sub-category (subgroup analyses, subdivision of outcome)

Study (data for each study entered in a standardized format, specific to each outcome

May not have subcategories...

then all studies fall directly under the outcome



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