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## Leveraging adult data in pediatric product development: The role of Bayesian statistics

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#### Outline

- Challenges in Pediatric Studies
- Extrapolation and Bayesian Model
- Prior Information Elicitation
- Bayesian Approaches
- Case Example
- Summaries



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#### **Challenges in Pediatric Studies**

- Smaller population size
- Less invasive measurement
- Unethical to include a placebo arm
- Shorter trial duration



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#### What Bayesian Can Do for YOU?

- Frustration:
  - Too many failed pediatric trials
- Purpose:
  - Less failed pediatric trials
  - Less inconclusive pediatric trials
  - Less pediatric trials



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#### **Statistical Significance**

### Significant

## Insignificant

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#### **Extrapolation (CDER)**



#### **Pediatric Study Planning & Extrapolation Algorithm**

#### Footnotes:

- a. For locally active drugs, includes plasma PK at the identified dose(s) as part of safety assessment.
- b. For partial extrapolation, one efficacy trial may be sufficient.
- c. For drugs that are systemically active, the relevant measure is systemic concentration.
- d. For drugs that are locally active (e.g., intra-luminal or mucosal site of action), the relevant measure is systemic concentration only if it can be reasonably assumed that systemic concentrations are a reflection of the concentrations at the relevant biospace (e.g., skin, intestinal mucosa, nasal passages, lung).
- e. When appropriate, use of modeling and simulation for dose selection (supplemented by pediatric clinical data when necessary) and/or trial simulation is recommended.
- f. For a discussion of no, partial and full extrapolation, see Dunne J, Rodriguez WJ, Murphy MD, et al. "Extrapolation of adult data and other data in pediatric drugdevelopment programs." Pediatrics. 2011 Nov;128(5):e1242-9.



#### **Extrapolation (CDRH)**



Guidance for Industry and FDA Staff: Leveraging Existing Clinical Data for Extrapolation to Pediatric Uses of Medical Devices (Draft) May 2015: Figure 1 10



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#### **Extrapolation and Bayesian Model**

- Extrapolation
  - Full
  - Partial
  - No
- Bayesian Model
  - Borrowing information from adult data (or other reliable data sources)
  - Minimize uncertainty incurred from using adult data



#### **Prior Information Elicitation**

- Adult Trial Data
  - Obvious choice?
  - Same disease with same treatment
  - Different population
- Similar Pediatric Trial Data
  - Similar population
  - Same disease with similar treatment
- PK/PD Data
  - Same population with same disease under same treatment
  - Different endpoint



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#### **Borrowing Information**

- Clinical input for reliable prior information
- Similarity
  - Population
    - Baseline characteristics and demographic information
  - Disease progression
    - Baseline disease characteristics
    - Placebo information
  - Treatment effect
    - Treatment group information
- Pre-specify criteria based on collected data



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#### **Bayesian Approaches**

- 1. Derive priors from the adult data
- 2. Bayesian Hierarchical Modeling
- CDRH 2015 guidance describes (2)
- Drug Information Association (DIA)/FDA Bayesian statistics working group has developed a concept paper describing (1) and (2) both as useful approaches for pediatric trials



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#### **Bayesian Approaches (cont.)**

- Bayesian Power Priors (Ibrahim & Chen, 2000)
  - Prior is a historical likelihood raised to a "power" to discount the information from the historical data
- Bayesian Commensurate Priors (Hobbs, et al., 2012)
  - Historical study data are on the same level as the current study data (no down-weighting)
  - Current study mean is centered at the historical study mean with precision that determines the commensurability of the studies



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#### **Case Example**

- Data: two adult clinical trials on a drug for a chronic disease
- Third trial: pediatric population
- Study treatment:
  - Adult: placebo, low dose & high dose
  - Pediatric: low dose
- Treatment is approved for both adult and pediatric patients based on these three trials



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#### **Bayesian Models**

- Borrow information on low dose only
- Adults: 121 and Pediatrics: 22
- Endpoint: Clinical response (yes vs. no)
- Model 1: Flat hierarchical model
- Model 2: Tier hierarchical model
- Model 3: Use adult posterior as prior for peds
- Model 4: Same as Model 3 w/ prior on k
- Model 5: Power prior



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#### **Results**



• Power prior is the most conservative model



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#### **Summaries**

- Pediatric studies pose unique challenges
- Explore innovative trial designs
- Informative prior data available
- Potential Bayesian models
- More efficient clinical trials



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#### References

 Extrapolation of efficacy and other data to support the development of new medicines for children: A systematic review of methods

Wadsworth I, et. al., Department of Mathematics and Statistics, Fylde College, Lancaster University, Lancaster, UK *Statistical Methods in Medical Research*; DOI: 10.1177/0962280216631359

 Stratification, Hypothesis Testing, and Clinical Trial Simulation in Pediatric Drug Development

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#### **Recent CDER Bayesian Work**

- Early stage studies (Phase 1, Phase 2)
  - Multi-stage dynamic treatment regime
  - Adaptive design
- Small sample studies
  - Rare diseases / Orphan drugs
  - Pediatric population
- Safety evaluation
  - Low adverse event rate
  - Continuous monitoring



#### **Recent CDER Bayesian Publication**

- Meta-Analysis: Meta-Analysis Using Dirichlet Process
  - S. Muthukumarana & R.C. Tiwari
  - Statistical Methods in Medical Research (July 2012)
- Non-Inferiority Study
  - Non-inferiority and networks: inferring efficacy from a web of data
    - J. Lin, M.A. Gamalo & R.C. Tiwari
    - Pharmaceutical Statistics, Dec 2015
  - Bayesian Approach to the Design and Analysis of Non-inferiority Trials for Antiinfective Products
    - M.A. Gamalo, R.C. Tiwari & L.M. LaVange
    - Pharmaceutical Statistics (Aug. 2013)
  - Bayesian Approach to Non-inferiority Trials for Normal Means
    - M.A. Gamalo, R. Wu & R.C. Tiwari
    - Statistical Methods in Medical Research (May 2012)



#### **Examples from Advisory Committees**

• Pediatric ODAC 2015:

http://www.fda.gov/AdvisoryCommittees/CommitteesMeetingMaterial s/Drugs/OncologicDrugsAdvisoryCommittee/ucm426351.htm

• Remicade UC 7/21/2011:

http://www.fda.gov/downloads/AdvisoryCommittees/CommitteesMee tingMaterials/Drugs/GastrointestinalDrugsAdvisoryCommittee/UCM2 66697.pdf

• Reslizumab Asthma 12/9/2015:

http://www.fda.gov/downloads/AdvisoryCommittees/CommitteesMee tingMaterials/Drugs/Pulmonary-AllergyDrugsAdvisoryCommittee/UCM477884.pdf



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# Thank you!

Questions?