

# **Pediatric Drug Development and Bayesian Method**

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# Disclaimer

**This talk reflects the views of the speaker  
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# Outline

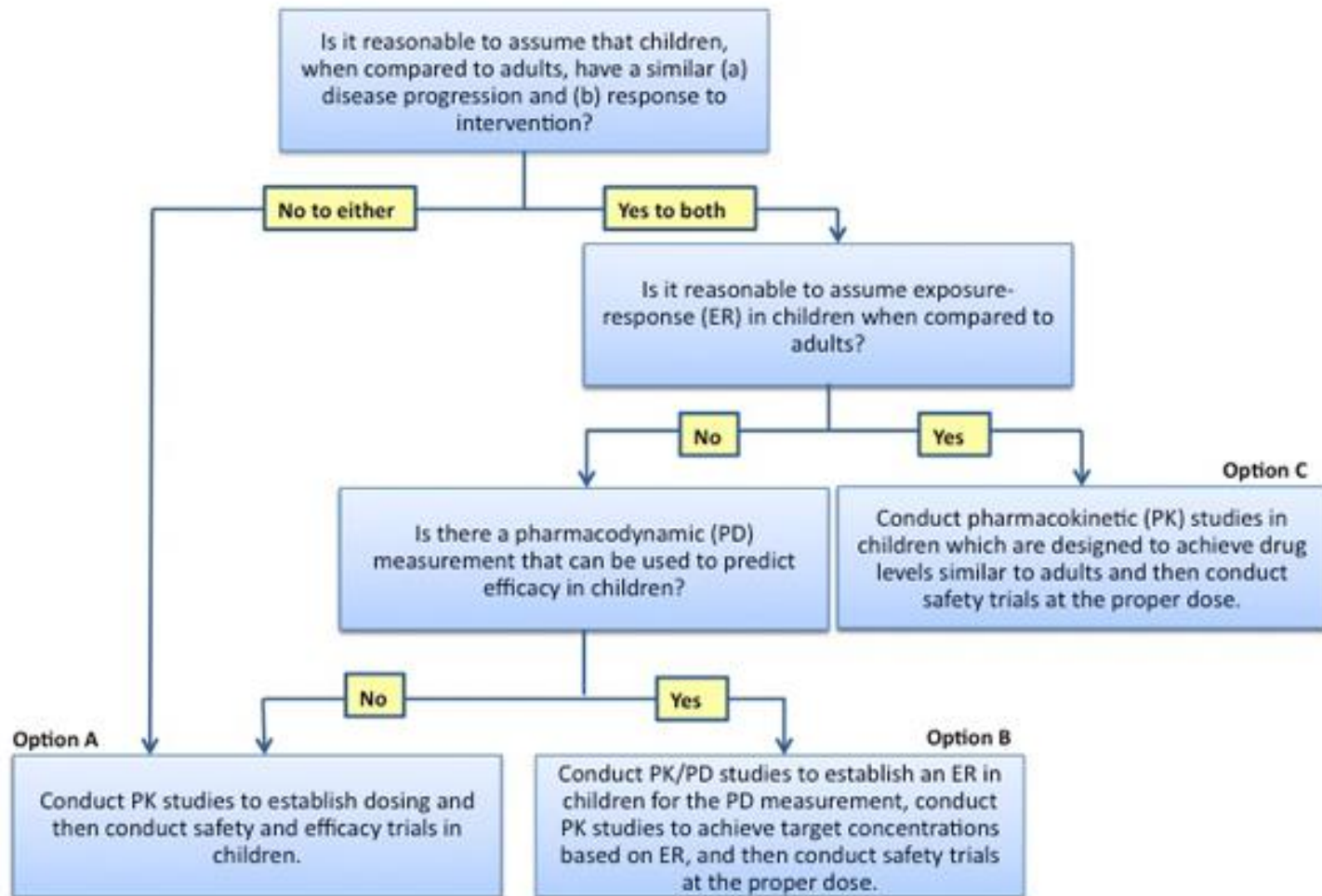
- Principles of pediatric drug development
- Challenges
- Applicability of Bayesian
- Case examples
- Conclusions

# Principles of Pediatric Drug Development

- Children are not miniature adults
  - Different physiologic, developmental, psychologic, and pharmacologic characteristics; Differ across pediatric spectrum
  - Different in how they metabolize and respond to drug: suboptimal therapy, unexpected response, AE, and toxicity
- Legal requirement of **SUBSTANTIAL EVIDENCE** to establish the effectiveness of drug
- For disease/condition that exists exclusively in children
  - Adequate and well-controlled trial(s)
- For disease/condition that exists in both adults and children
  - Adequate and well-controlled trial(s)
  - Extrapolation may be allowed under some circumstances

# Extrapolation vs. No Extrapolation

Figure 1: FDA Pediatric Study Decision Tree



# Challenges

- **Low enrollment**
  - Limited disease population, fear of the risk, concern over randomization to placebo, drug off-label use
- **High dropout**
  - Too many study visits, number of invasive procedures, feel drug is ineffective, concern about risk/benefit profile
- **Ethical considerations**
  - Participants are expected to benefit from clinical trials, benefit and risk balance, minimizing risk by minimizing the number of participants at design stage
- **Delayed study initiation**
  - Formulation can take time, waiting for adult approval, safety considerations

**Operational challenges + Design challenges  
+ Analysis challenges**

# Innovative Thinking and Strategies

**Span over the entire drug development cycle –  
Opportunity for statisticians to play an essential role**

- Global pediatric trial – MRCT (ICH E17)
- Age-staggered enrollment and initiation
- Enrolling adolescents into adult trials
- Master protocols – umbrella, platform, basket
- Real World Data – Historical (external) Controls
- Bayesian methods

# Applicability of Bayesian Methods in Pediatric Trials

**Similarities between adult and pediatric patients on the course of disease and effect of drug; Similar study design, conduct, and endpoints; Adult/older age group data are available and hard to ignore**

- Leveraging adult/older age group data

**Ethical considerations: Limit the number of exposure; Stop the trial early for efficacy or futility**

- Borrowing external data as control / augment the concurrent control when reasonable
- Update knowledge or decision-making when information accumulates, e.g. Bayesian sequential monitoring

# Case Example 1

- Chronic Hepatitis C – MAVYRET
  - Approved for adults in 2017
  - Course of disease and the effect of the drugs are sufficiently similar in adults and pediatric patients, but not identical
  - Extrapolation of efficacy from PK data to support approval, SVR12 provides supportive evidence of efficacy
- Pediatric clinical trial
  - Open-label, single-arm, adolescent subjects 12 to <18 years (n=47).
  - PK data were comparable to adults. SVR12 rate was 100% (47/47)
  - Approved in May, 2019. First treatment for all genotypes of HCV in pediatric patients

# A Related Hypothetical Scenario

- What if SVR12 is 95%, 90%, 85%...?
- Adult SVR12 – genotype 1-6, no cirrhosis, treatment-naïve, treatment experienced (peginterferon, ribavirin, sofosbuvir)

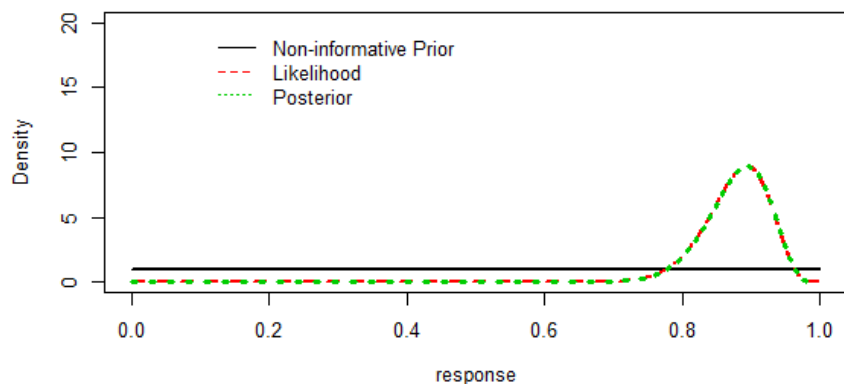
<b>Adult</b>	<b>N=663</b>	<b>GT1-6: 93% - 100%</b> <b>95% CI for GT1-6 ranges 82%, 100%</b>
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- Bayesian analysis – interested in posterior prob. of SVR12 > xx% (e.g. 82%, 30%)
  - Selection of prior
    - $\text{Prior} = (1-\alpha) \times f(D) + \alpha \times g(D)$
    - $f(D)$ : skeptical prior/non-informative prior
    - $g(D)$ : adult posterior
    - $\alpha = P(\text{applicability of adult results})$

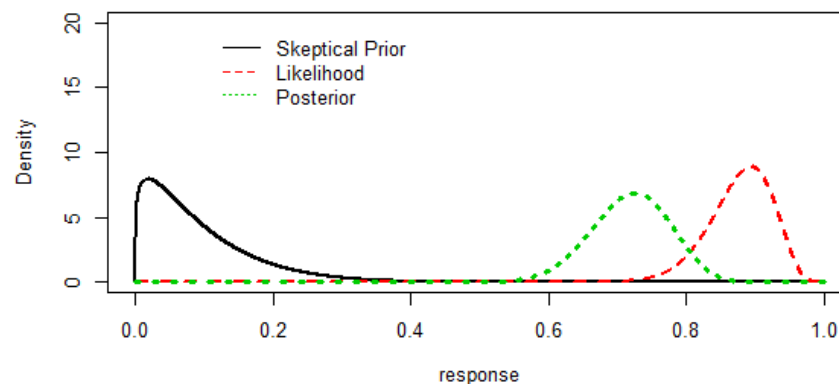
# A Related Hypothetical Scenario

## Posterior Distribution

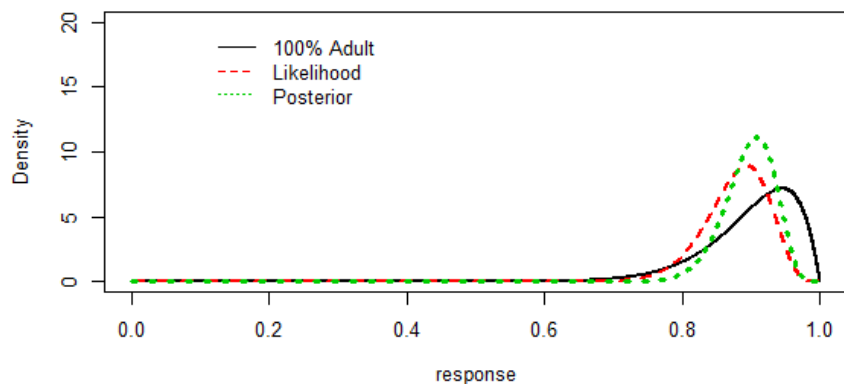
**Non-informative Prior**



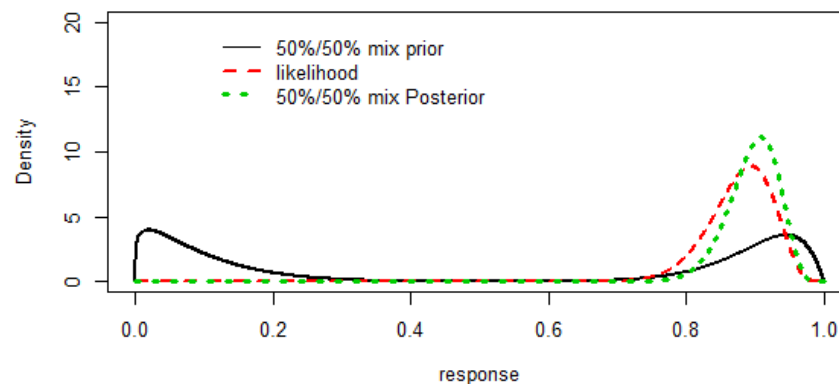
**Skeptical Prior**



**100% Adult**



**50%/50% mix prior**



# A Related Hypothetical Scenario

## Posterior Probability of Efficacy

Prior	If observed pediatric SVR 12 = 89%		
	Bayesian Estimate (95% Cred. Int.)	Post. Prob. SVR12 > 82%	Post. Prob. SVR12 > 30%
Non-informative	88.3 (77.3, 95.3)	88.5%	>99.9%
Skeptical	72.0 (59.9, 82.3)	2.9%	>99.9%
100% adult	89.9 (81.3, 95.6)	96.5%	>99.9%
50%/50% mixture	89.9 (81.2, 95.6)	96.5%	>99.9%

## Case Example 2

### Benlysta (belimumab)

Approved for adult patients with active, seropositive lupus erythematosus (SLE) in 2011.

- A pediatric post-marketing study was required under PREA:
  - A randomized, double-blind, placebo-controlled trial targeting to enroll 100 pediatric subjects 5 to 17 years of age with active systemic SLE
- Study was not fully powered by design. Efficacy was planned to be descriptive; no formal statistical hypothesis testing
- Trial was started in 2012. Due to difficulties in enrolling patients, the overall target enrollment was reduced to 70 in 2016. The study was completed in 2018
- There was no approved treatment for pediatric SLE patients

[https://www.accessdata.fda.gov/drugsatfda\\_docs/label/2019/125370s064,761043s007lbl.pdf](https://www.accessdata.fda.gov/drugsatfda_docs/label/2019/125370s064,761043s007lbl.pdf)

<https://www.fda.gov/media/127912/download>

# Case Example 2 - continued

## Primary Efficacy Result

**Primary endpoint:** response rate at Week 52 as assessed by SRI (SLE Responder Index - composite efficacy measure, used in the approved adult trials)

	Pediatric Study		Adult Study 1		Adult Study 2	
	Placebo N=40	Belimumab N=53	Placebo N=275	Belimumab N=273	Placebo N=287	Belimumab N=290
Response, n(%)	17 (43.6)	28 (52.8)	93 (34)	118 (43)	125 (44)	167 (58)
Observed difference	9.2%		9%		14%	
Odds ratio (95% CI)	1.5 (0.6, 3.5)		1.5 (1.1, 2.1)		1.8 (1.3, 2.6)	

**Secondary endpoint:** proportion of subjects meeting PRINTO/ACR Juvenile SLE Response Evaluation criteria for improvement at Week 54, and the 5 components of this endpoint – all results favor the belimumab group

**Post-hoc exploratory analyses of biomarkers:** all results favor the belimumab group

## Case Example 2 - continued

### A Post-hoc Bayesian Analysis Was Requested by FDA

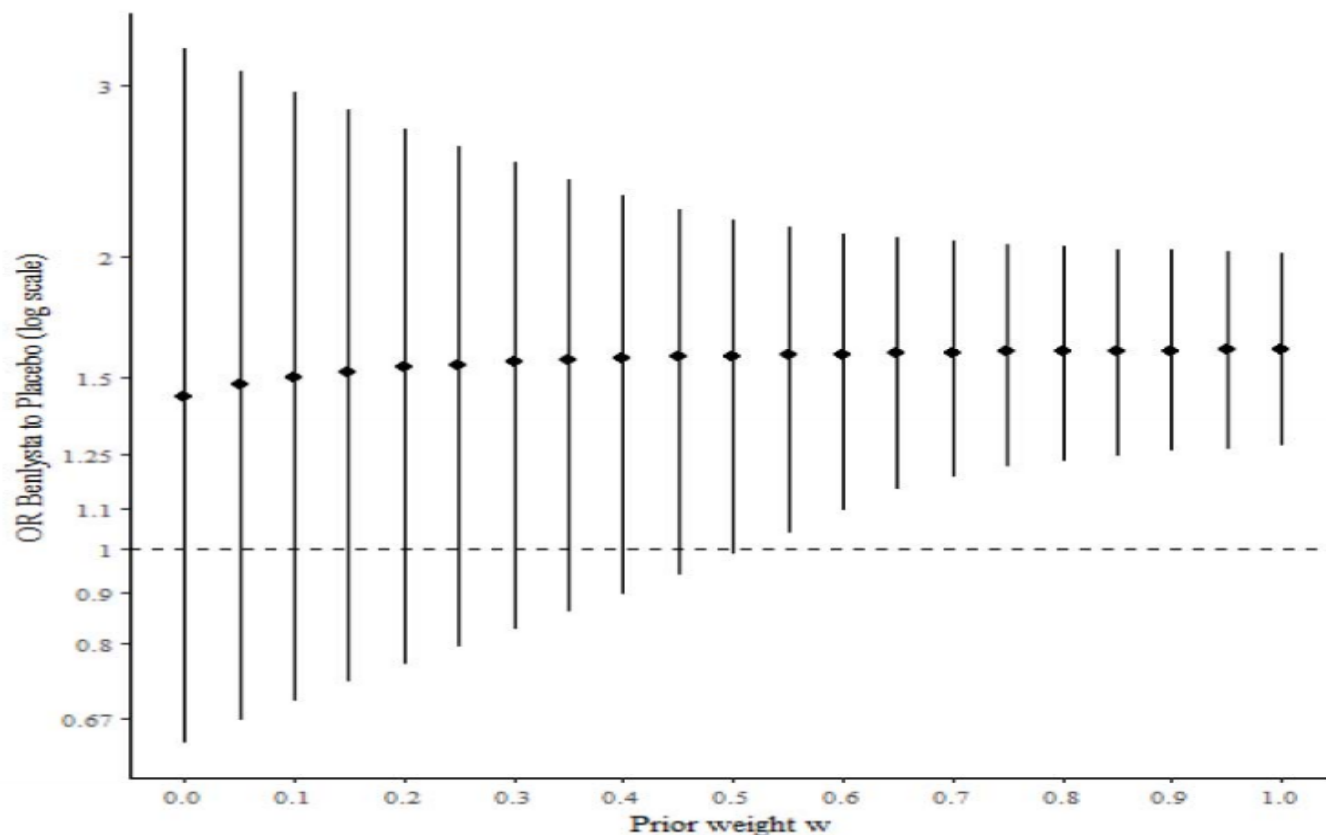
Comparisons between adults and pediatric patients/studies

- Disease and patient response to the treatment are likely to be similar
- Same efficacy endpoints
- Study design and conduct were highly similar

Borrow information from adult studies

- Bayesian mixture prior - Determine the range of weight where the posterior probability of efficacy exceeds the thresholds of 97.5%
- $\text{Prior} = (1 - \alpha) \times f(D) + \alpha \times g(D)$ 
  - $f(D)$ : skeptical prior -  $N(0, m \times s_P^2)$
  - $g(D)$ : adult posterior -  $N(y_A, s_A^2)$
  - $y_A$ : log odds ratio from adult studies
  - $s_A^2$ : variance of adult log odds ratio
  - $s_P^2$ : variance of pediatric log odds ratio
  - $m$ : was chosen so that the effective sample size of this component worth one pediatric subject per arm
  - $\alpha$ : weight

## Case Example 2 - continued



- A prior weight of 0.55 or larger led to the posterior probabilities > 97.5%
- $\geq 0.55$  weight on the relevance of the adult information is reasonable
- Approved in pediatric population ( $\geq 5$  years) in 2019

## Case Example 3

### PREVAIL II and PALM – A Story of ZMapp

- PREVAIL II – A randomized, controlled trial with a “barely Bayesian” design
  - Prior: Independent uniform distribution on  $[0, 1]$  for arms A and B
  - Posterior probability: Observe data  $\rightarrow$  update priors to posterior distributions  $\rightarrow$  compute posterior probability of  $p_A < p_B$
  - Primary analysis: Posterior probability of Day 28 mortality; Pre-specified threshold for success was 97.5%
  - Aimed to enroll 200 patients (1:1), but ended up with 72 patients because of the waning Ebola epidemic.
  - Result: Mortality rates were 22% vs 37% for ZMapp+SOC vs SOC; Posterior probability that ZMapp was better was 91.2%

# Case Example 3 - continued

## PREVAIL II and PALM – A Story of ZMapp

- PALM – A randomized, controlled platform trial
  - Four treatment arms: ZMapp, remdesivir, Mab114, REGN-EB3 (1:1:1:1)
  - Targeted sample size was 725 in total
  - Primary endpoint: Day 28 mortality
  - Comparisons were restricted to patients enrolled concurrently
  - Results: ZMapp arm was discontinued early due to high mortality

Population	ZMapp	Remdesivir	Difference, Remdesivir vs. ZMapp	MAb114	Difference, MAb114 vs. ZMapp	REGN-EB3	ZMapp Subgroup	Difference, REGN-EB3 vs. ZMapp Subgroup
	<i>no. of deaths/ total no. (%)</i>	<i>no. of deaths/ total no. (%)</i>	<i>percentage points (95% CI)</i>	<i>no. of deaths/ total no. (%)</i>	<i>percentage points (95% CI)</i>	<i>no. of deaths/ total no. (%)</i>	<i>no. of deaths/ total no. (%)</i>	<i>percentage points (95% CI)</i>
Overall	84/169 (49.7)	93/175 (53.1)	3.4 (–7.2 to 14.0)	61/174 (35.1)	–14.6 (–25.2 to –1.7)*	52/155 (33.5)	79/154 (51.3)	–17.8 (–28.9 to –2.9)*
Patients with high viral load†	60/71 (84.5)	64/75 (85.3)	0.8 (–15.3 to 17.2)	51/73 (69.9)	–14.6 (–33.0 to –0.5)	42/66 (63.6)	56/65 (86.2)	–22.5 (–41.8 to –5.1)
Patients with low viral load†	24/98 (24.5)	29/100 (29.0)	4.5 (–9.1 to 19.1)	10/101 (9.9)	–14.6 (–32.4 to –2.6)	10/89 (11.2)	23/89 (25.8)	–14.6 (–32.6 to –2.3)

# Conclusion

- Challenges exist in pediatric drug development
- Demonstration of substantial evidence is required
- Bayesian methods can be useful in pediatric trials
- Prior beliefs affect the estimates of efficacy
- Type I error rate may be inflated depending on the choice of prior
- Pre-specification is key
- A promising result may not be a real effect for both pediatric and adult studies
- Early interaction with the FDA is encouraged

# Acknowledgement

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**Thank You!**