

Application of Bayesian Extrapolation in Pediatric Drug Development Program

May Mo, Amy Xia | Amgen

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Outline

- **Pediatric Development Laws and Regulatory Guidelines**
- **Pediatric Extrapolation**
- **Statistical Extrapolation – Bayesian Methods**
- **Example 1 – a pediatric extrapolation proposal**
- **Example 2 – a statistical extrapolation analysis**
- **Summary**

US Pediatric Development Laws: PREA vs BPCA

- **Pediatric Research Equity Act (PREA)**

- Drugs and biologics
- Studies must be labeled
- Mandatory (required)
- No financial incentive
- For indication under review
- Orphan indication exempted

- **Best Pharmaceuticals for Children Act (BPCA)**

- Drugs and biologics
- Studies must be labeled
- Voluntary (written request)
- Financial incentive
- May expand to other indications
- May be requested in orphan indication

Recent Regulatory Guidelines

- **ICH. E11(R1)**: Addendum: Guideline on Clinical Investigation of Medicinal Products in the Pediatric Population. Step 2b; 12 October 2016.
- **FDA**. Guidance for Industry: Leveraging Existing Clinical Data for **Extrapolation** to Pediatric Uses of Medical Devices. June 2016.
- **FDA**. Guidance for Industry: Pediatric Study Plans: Content of and Process for Submitting Initial Pediatric Study Plans and Amended Initial Pediatric Study Plans. March 2016.
- **FDA**. Guidance for Industry: General Clinical Pharmacology Considerations for Pediatric Studies for Drugs and Biological Products. December 2014.
- **EMA**. Reflection Paper on **Extrapolation** of Efficacy and Safety in Pediatric Medicine Development, EMA/199678/2016.



Pediatric Study Plan Contents*

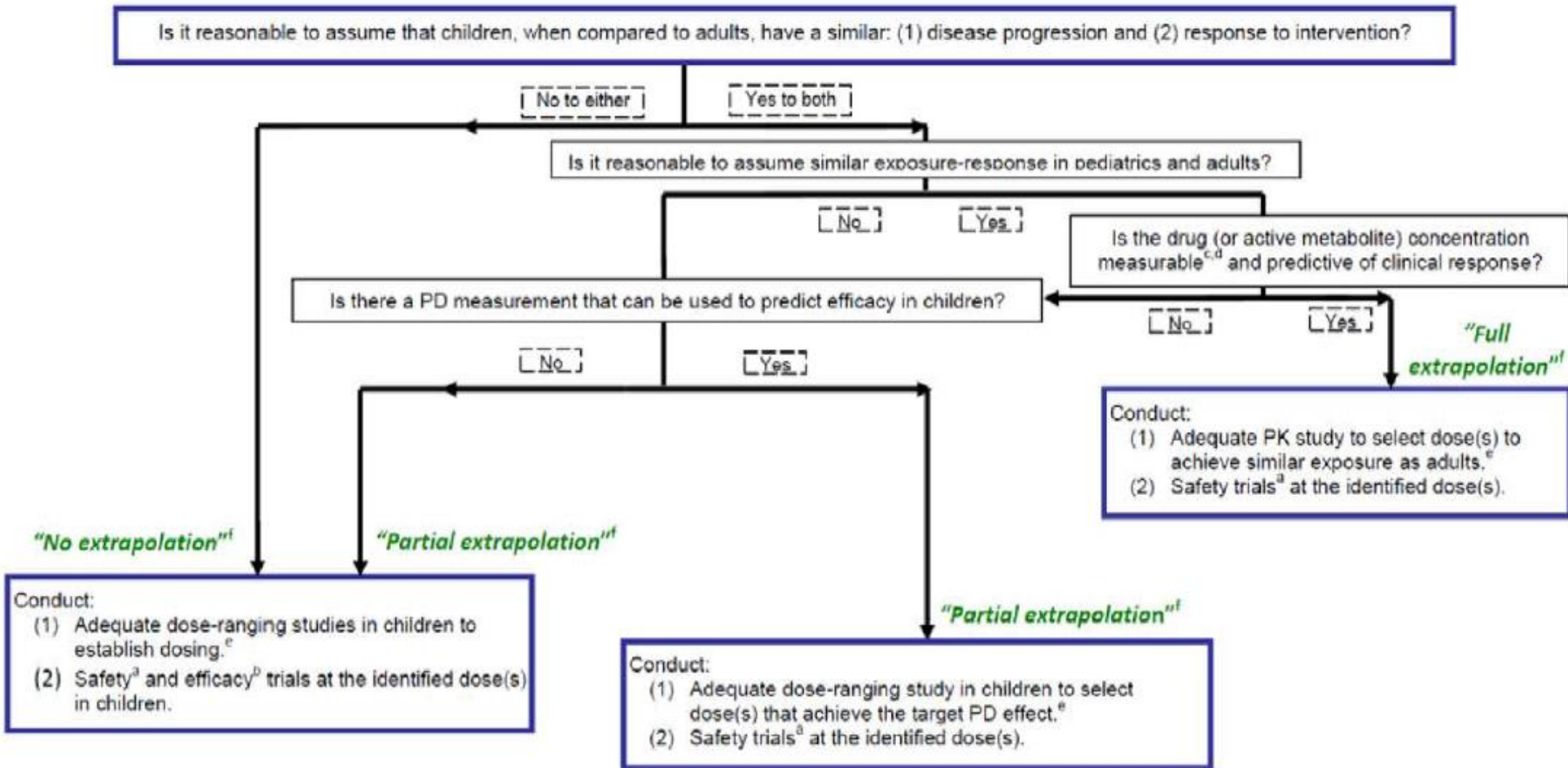
Required per Federal Food, Drug, and Cosmetic Act (FD&C Act), as amended by the Food and Drug Administration Safety and Innovation Act (FDASIA)

1. OVERVIEW OF THE DISEASE IN THE PEDIATRIC POPULATION (1-3 pages)
2. OVERVIEW OF THE DRUG OR BIOLOGICAL PRODUCT (1-3 pages)
3. **OVERVIEW OF PLANNED EXTRAPOLATION OF EFFECTIVENESS TO SPECIFIC PEDIATRIC POPULATIONS (1-3 pages)**
4. PLANNED REQUEST FOR DRUG-SPECIFIC WAIVER(S) (1-3 pages)
5. PLAN TO REQUEST DEFERRAL OF PEDIATRIC STUDIES (1-3 pages)
6. TABULAR SUMMARY OF PLANNED NONCLINICAL AND CLINICAL STUDIES
7. AGE-APPROPRIATE FORMULATION DEVELOPMENT (1-3 pages)
8. NONCLINICAL STUDIES (1-3 pages)
9. CLINICAL DATA TO SUPPORT DESIGN AND/OR INITIATION OF STUDIES IN PEDIATRIC PATIENTS (1-5 pages)
10. PLANNED PEDIATRIC CLINICAL STUDIES
 - 10.1 Pediatric Pharmacokinetic Studies (1-10 pages)
 - 10.2 Clinical Effectiveness and Safety Studies (1-10 pages)
11. TIMELINE OF THE PEDIATRIC DEVELOPMENT PLAN (1 page)
12. AGREEMENTS FOR PEDIATRIC STUDIES WITH OTHER REGULATORY AUTHORITIES (1-3 pages)

* FDA. Guidance for Industry: Pediatric Study Plans: Content of and Process for Submitting Initial Pediatric Study Plans and Amended Initial Pediatric Study Plans. March 2016.



Pediatric Study Planning and Extrapolation Algorithm



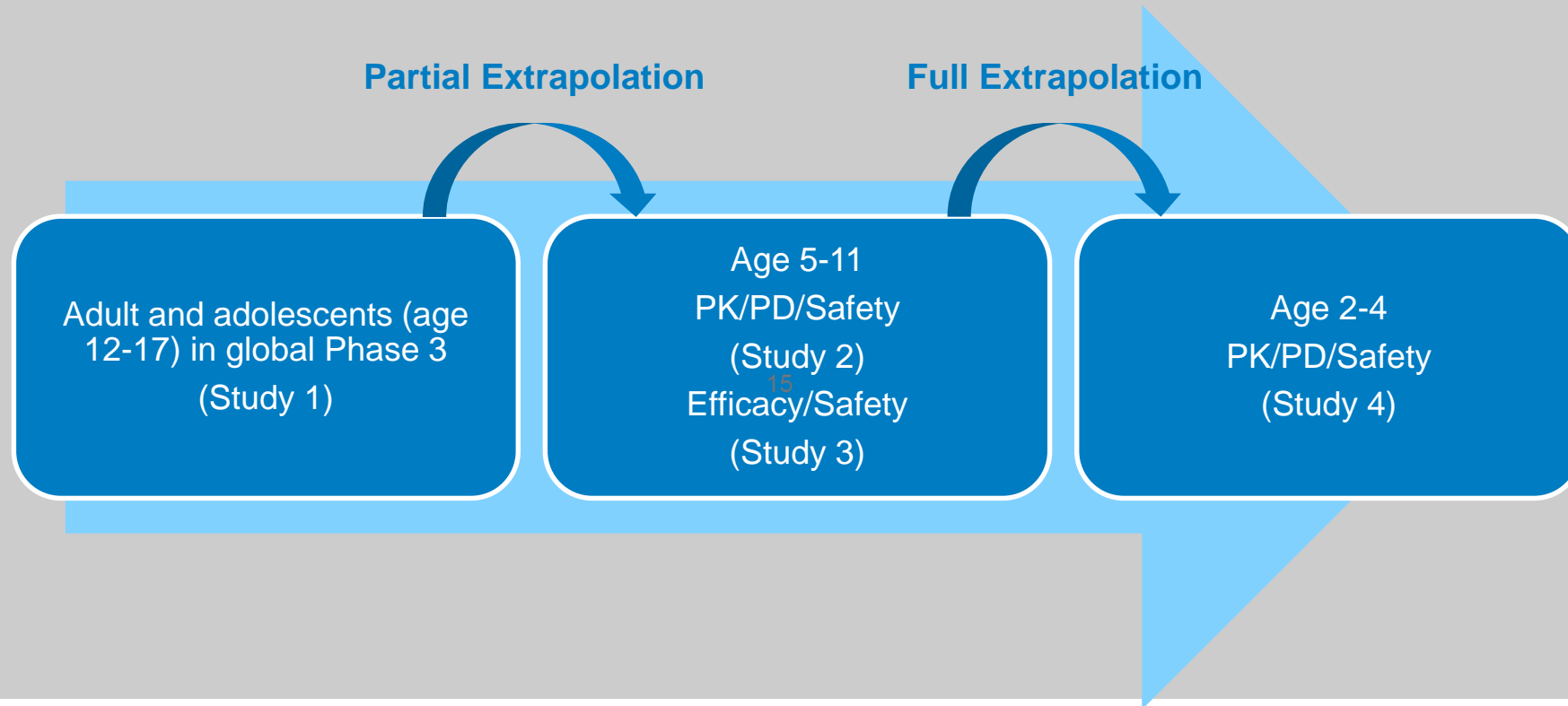
Extrapolation framework

- Extrapolation concept
 - Define source and target population
 - Form predictions and hypotheses
- Extrapolation plan
 - Prospective; including study planning and sample size
- Confirmation & extrapolation
 - Confirm consistency between prediction and observed target data
 - Iterative process of predicting and confirming
 - Conclusion based on confirmed concept
- Mitigating uncertainty and risk

Bayesian Methods for Pediatric Extrapolation

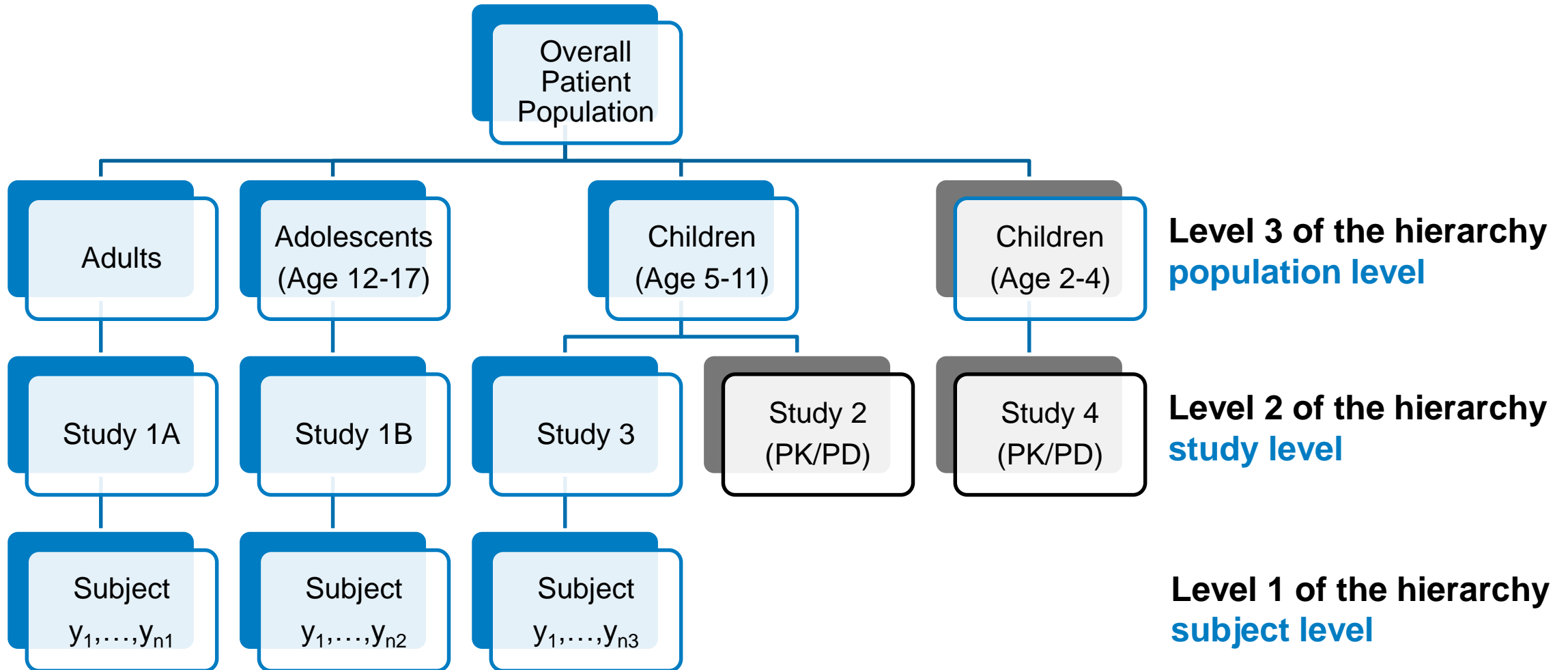
- Bayesian methods are a natural choice for quantitatively extrapolating information from source to target population in a **partial extrapolation** setting (ICH, 2014; US FDA, 2016).
- Bayesian methods to consider:
 - Use the posterior from source as the prior for target population
 - Power prior (Ibrahim and Chen, 2000)
 - Bayesian hierarchical modeling
- FDA’s “Guidance for Industry and Food and Drug Administration Staff: Leveraging Existing Clinical Data for Extrapolation to Pediatric Uses of Medical Devices” (US FDA, 2016) recommended **Bayesian Hierarchical Model** as an appropriate method for extrapolation.

Example 1: Stepwise extrapolation proposal in a PSP to satisfy PREA requirement



Different extrapolation strategies may be adopted

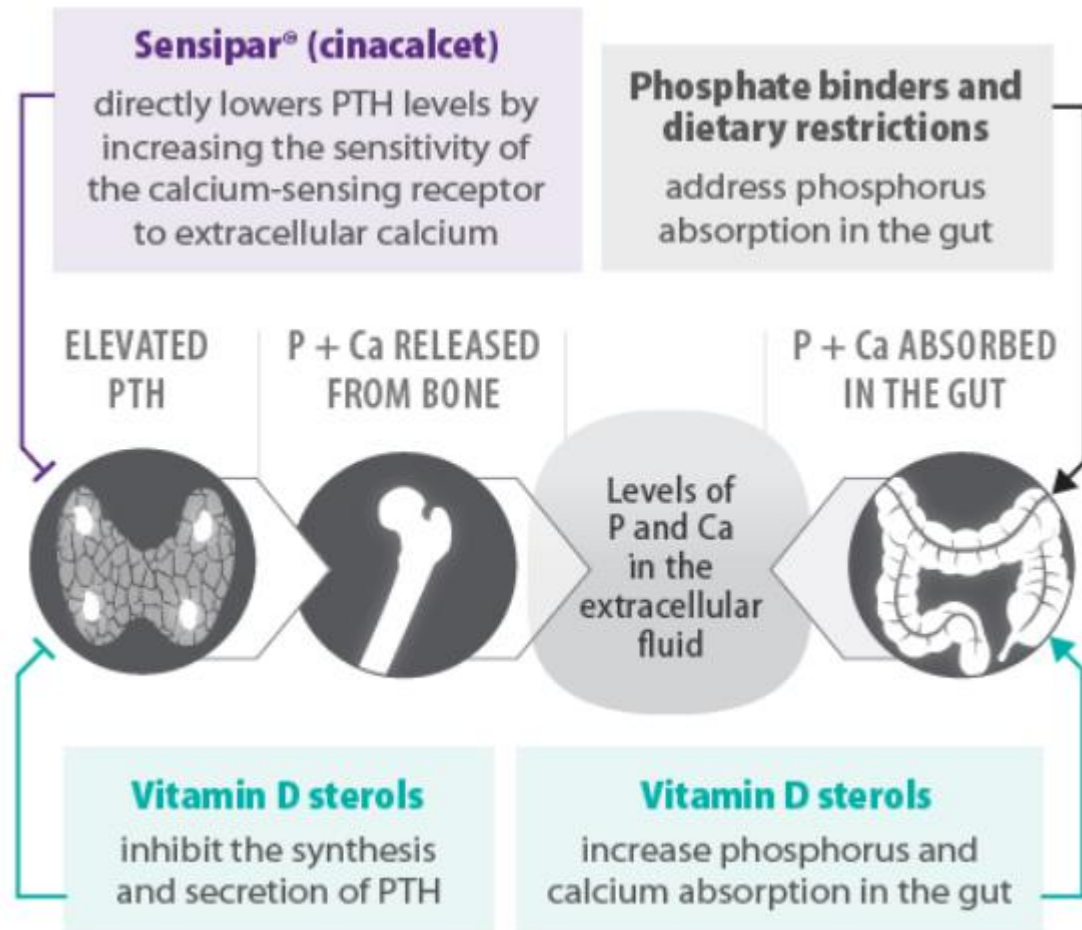
3-level Bayesian Hierarchical Model



Extrapolation Reduced Proposed Sample Size

- PSP sample size justification for Study 3 (children 5-11 years)
 - Source data can be extrapolated
 - Adult, adolescent data from Study 1
 - Placebo data from other pediatric (age 5-11) program
 - Extrapolation can save up to 50% of subjects in trial 3 assuming extrapolated information cannot exceed that are observed in the target population

Example 2: Cinacalcet Bayesian Extrapolation Analysis Under BPCA Written Request



- Cinacalcet is indicated for the treatment of secondary hyperparathyroidism (HPT) in adult patients with chronic kidney disease receiving dialysis
- Cinacalcet is a calcimimetic agent which acts as a modulator of the calcium-sensing receptor (CaR) and regulates PTH secretion
- **Pediatric, orphan indication**
 - < 1000 pediatric patients on dialysis will develop secondary HPT. Of these, approximately 300 patients are estimated to be 0 to 5 years of age

Rationale for the Pediatric Extrapolation Strategy

- The adult and pediatric populations are similar in the following aspects according to ICH E11 (ICH, 2000):
 - Pediatric population in which cinacalcet has been studied are similar to those of the approved population in adults (i.e., patients with secondary HPT treated with dialysis).
 - The pathophysiology and course of the disease process (secondary HPT) is similar in adult and pediatric populations with CKD receiving dialysis.
 - The outcome of therapy is likely to be comparable.

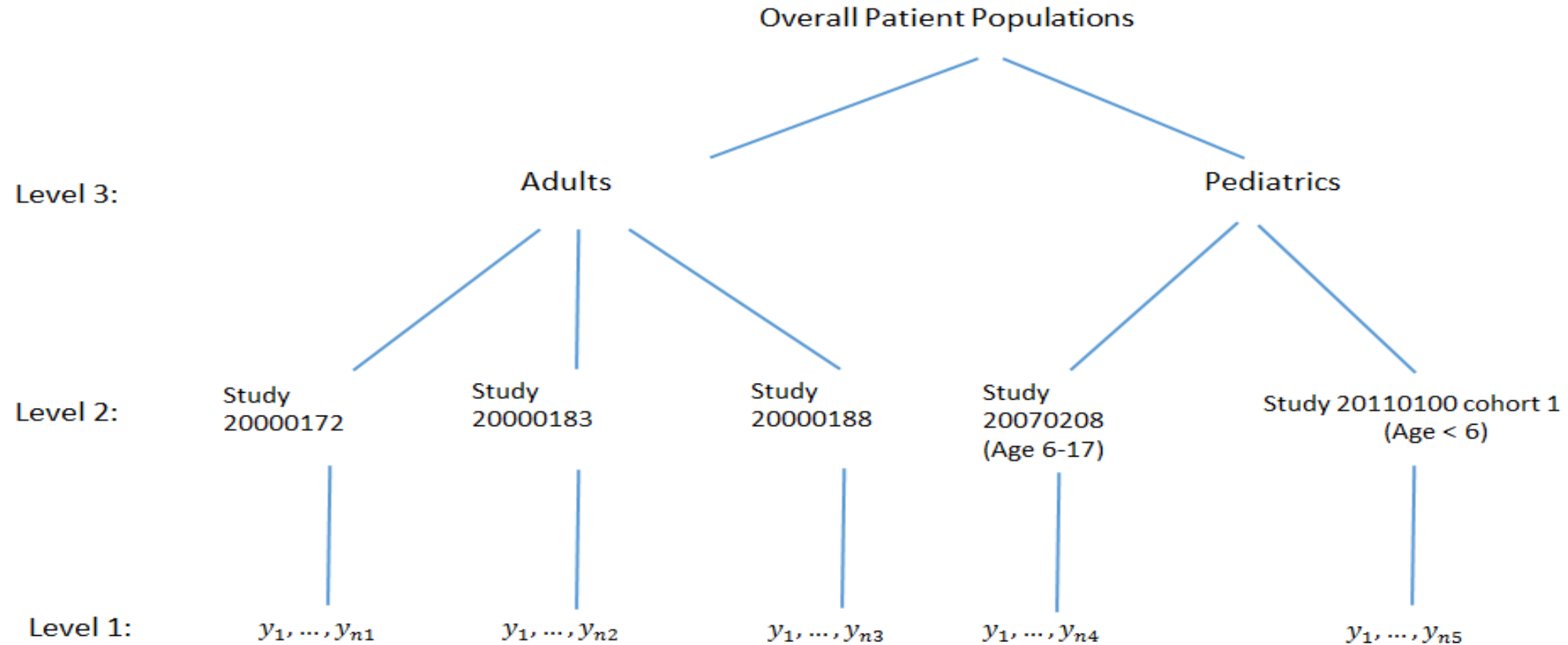
Pediatric Extrapolation Strategy by Age Category

- Partial extrapolation for age 6 – 17
 - Sample sizes of pediatric studies were too small to have adequate power
 - Adult data were borrowed to estimate treatment effect
- Full extrapolation for age younger than 6
 - Younger age study focused on safety rather than efficacy
 - No control arm in younger age study, and relative treatment effect in younger age cannot be directly estimated

Bayesian Statistical Models for Extrapolation

- **Bayesian hierarchical model**
 - Assumes exchangeability at population level, study level and subject level
 - Objective amount of borrowing depending on the consistency of evidence across information sources: more borrowing if results are more similar, less borrowing if results are less similar
 - **Recommended in FDA guidance (2016); primary method in FDA filing**
- **Power prior**
 - Discount adult data by the power prior parameter
 - Pre-specified power prior parameter: subjective amount of borrowing
 - Relation to hierarchical models: power prior parameter connects to the variance parameters in hierarchical models

3-level Bayesian Hierarchical Model – Primary Method for US submission



Bayesian Hierarchical Model Requires Exchangeability

- The model assumes exchangeability at each hierarchical level
 - subjects within a study are exchangeable (at Level 1)
 - studies within a patient population are exchangeable (at Level 2)
 - patient populations are exchangeable (at Level 3)
- Exchangeable means there is nothing known a priori that would imply one (subject, study, or population) would be better or worse in the outcome of interest than another (US FDA, 2016).

3-level Bayesian Hierarchical Model

- **Notations**

- treatment group c
 - $c = 0$ for placebo
 - $c = 1$ for cinacalcet
- patient population p
 - $p = 0$ for pediatric
 - $p = 1$ for adult
- study k
 - when $p = 0$: $k = 1$ for Study 20070208; $k = 2$ for Study 20110100 cohort 1
 - when $p = 1$: $k = 1$ for adult Study 20000172; $k = 2$ for adult Study 20000183; $k = 3$ for adult Study 20000188
- n_{pkc} be the number of subjects for population p , study k and treatment group c
- Y_{pkc} be the number of responders (number of subjects achieving $\geq 30\%$ reduction from baseline in mean iPTH)
- p_{pkc} be the proportion of responders (response rate)

3-level Bayesian Hierarchical Model

- **Level 1 of the hierarchy – subject level**

- Assuming binomial distribution for Y_{pkc} , that is

$$Y_{pkc} | p_{pkc} \sim \text{Binomial}(n_{pkc}, p_{pkc})$$

- **Level 2 of the hierarchy – study level**

- Study level random effects are introduced to account for the between-study variabilities:

$$\text{logit}(p_{pkc}) = \beta_{pc} + \gamma_{pkc}$$

where

$$\text{logit}(x) = \log\left(\frac{x}{1-x}\right),$$

β_{pc} is the population effect,

γ_{pkc} is the study level random effect,

$\gamma_{pkc} \sim N(0, \sigma_{pc}^2)$, where σ_{pc}^2 accounts for the between-study variation.

3-level Bayesian Hierarchical Model

- **Level 3 of the hierarchy – population level**

- Pediatric and adult population parameters are assumed to be random samples from an overall patient population, that is:

$$\beta_{pc} = \beta_c + \mu_{pc}$$

- β_c is the treatment effect such that $\text{expit}(\beta_c)$ is the overall response rate for treatment group c across all age groups and $\mu_{pc} \sim N(0, \sigma_c^2)$ which captures the between-population variability

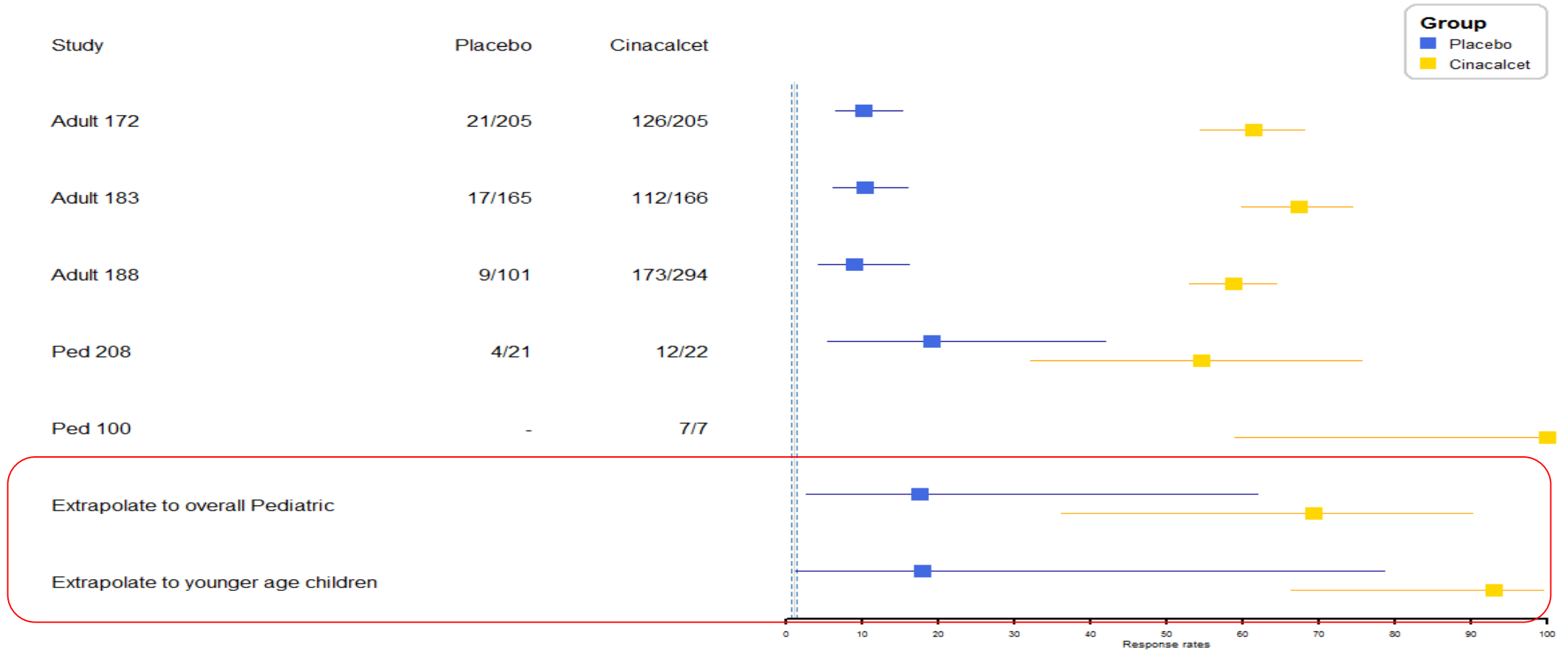
Effective Sample Size

- The information borrowed from the source population can be quantified using effective sample size (ESS) calculated based on variance reduction
- ESS was selected using a grid search method following the FDA Guidance (US FDA, 2010) that the ESS could not exceed the observed sample size in the target population.
- The restriction was implemented to prevent the adult data from being too informative and dominate the limited pediatric data.

3-level Bayesian Hierarchical Model

	Placebo	Cinacalcet	Difference (Δ)	Posterior Exceedance Probability			
	Median (95% CrI)	Median (95% CrI)	Median (95% CrI)	$\Delta > 0\%$	$\Delta > 10\%$	$\Delta > 20\%$	$\Delta > 30\%$
<i>No extrapolation, fit 208 and 100 data only</i>							
Overall Pediatric (28 days to < 18 years) response rates (%)	21.5 (2.1, 86.6)	80.8 (27.4, 98.2)	53.2 (-19.9, 90)	92.4	87.9	81.9	74.2
<i>With extrapolation, fit adult and pediatric data together</i>							
Pediatric patients 28 days to < 18 years response rates (%) ^a	17.5 (2.6, 62)	69.3 (36.1, 90.2)	48.6 (-2.3, 79.7)	97.0	93.9	88.4	79.2
ESS /n_ped	21.0 / 21 ^b	25.1 / 29 ^c	-	-	-	-	-
Pediatric patients 28 days to < 6 years response rates (%) ^d	18.0 (1.2, 78.7)	93.1 (66.4, 99.5)	71.8 (10.7, 95.1)	99.0	97.6	95.5	92.5
ESS/n_100	_*	-0.6/7 ^e	-	-	-	-	-

Observed Data and Posterior Estimates



Summary

- **Partial or full extrapolation are supported by HA to reduce the need to run large pediatric trials given similarities in disease progression and response to intervention**
 - Justifications for extrapolation need to be carefully examined
 - Collaborative work with PK/PD and Clinical teams
- **Bayesian extrapolation has broad applications to help with sample size limitations and missing control arms in pediatric setting**
 - Extrapolation from one population to another population
 - Extrapolation from historical data to current studies
- **The usefulness of sensitivity analysis using different statistical methods**
 - Power parameter in the power prior method
 - Commensurate prior method