

# Considerations for the Dose-Response Analyses of Inorganic Arsenic Health Outcomes

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## Disclaimer

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## Presentation Outline

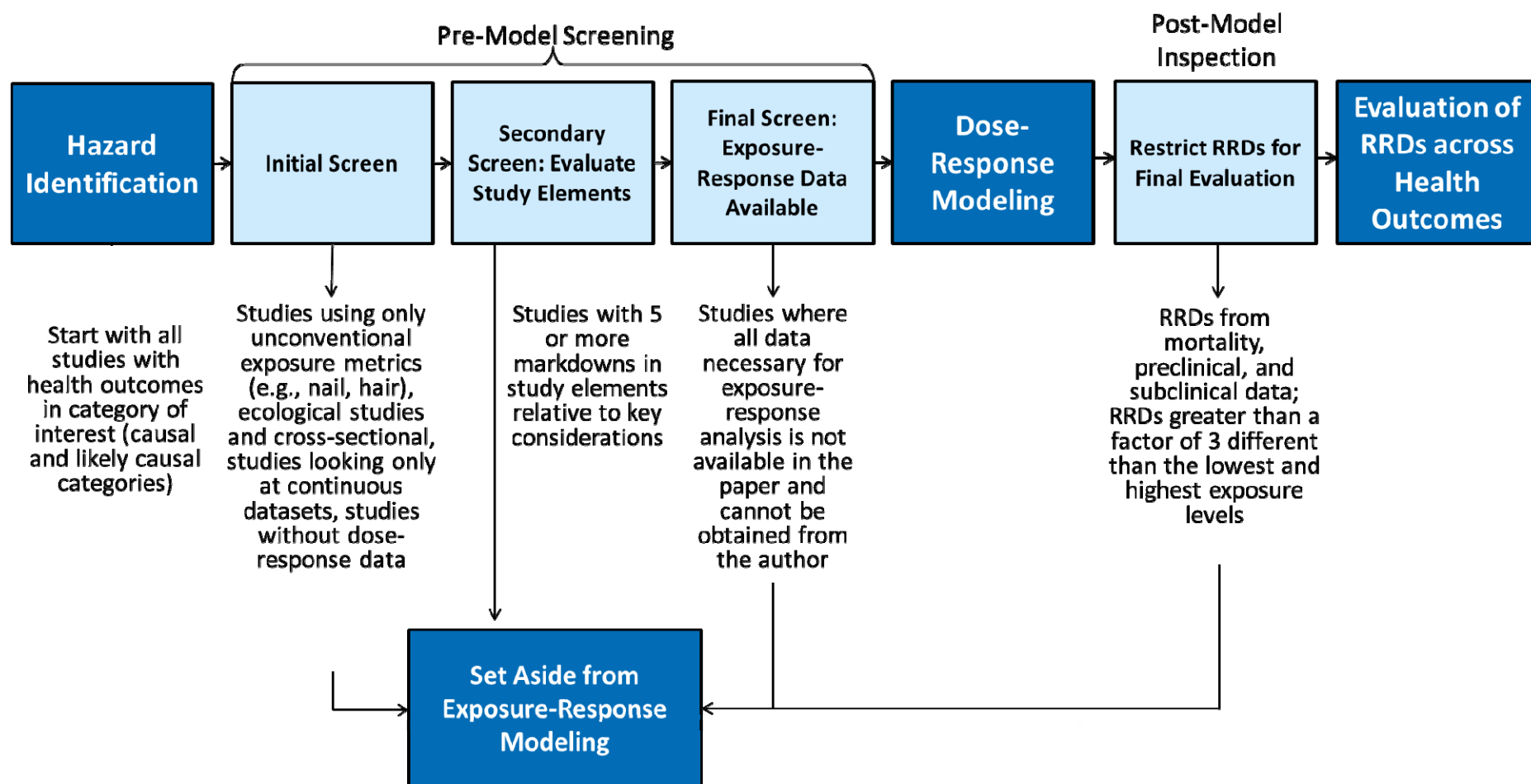
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- Overview of Potential Tiered Dose-response Approach
  - Margin of Exposure (MOE) analysis
  - Model averaging to assess within study model uncertainty
  - Bayesian meta-regression multiple study analysis
  - Extrapolation to target, U.S. population using lifetable approach
- Potential Science Issues
- Acknowledgments

## **D-R Methods – Margin of Exposure (MOE) Screening Level Analysis**

- **Screening Level Margin of Exposure (MOE) Analysis:**
  - **Purpose – verify NRC health outcome tiers, prioritize studies and endpoints for more complex analyses and inform cost-benefit analyses for a broad set of endpoints**
  - **Approach:**
    - Develop Relative Risk Dose (RRD; like BMD) for causal & likely causal health outcomes
    - Identify exposure-response datasets from adequate studies
    - Model unaltered exposures and adjust responses for selected datasets
    - Focus on single study, single best fitting standard model

# MOE Analysis – Identification of Adequate Dose-response Datasets



## “Mark-Down” Elements for Pre-model Study Screening

“Mark-Down” Element	Explanation
<b>Endpoint Selection</b>	Incidence preferred to mortality; odds ratio or relative risk preferred to SMR
<b>Number of subjects and cases reported</b>	Numbers of cases & controls/subjects highly preferred for dose-response; reporting only summary measures (OR, RR, SMR) is serious shortcoming
<b>Exposure ascertainment</b>	Prefer individual measurements, then small-groups, then large groups
<b>Exposure uncertainty</b>	Mean or median with variance preferred; ranges of exposures less desirable
<b>Exposure/Dose metric</b>	Cumulative intake or exposure preferred over point-in-time measurements. Urinary markers (adequately characterized) is considered a reliable indicator of intake; other biomarkers (hair, nails, blood) less desirable
<b>Exposure timing, duration</b>	Evidence supporting long-term exposure patterns highly desirable
<b>Measurement of, adjustment for, covariates</b>	Estimates control for at least smoking, gender, age, as well as other covariates (nutrition, genetic polymorphisms, etc.) as appropriate
<b>Number of exposure grps</b>	Referent plus two or more exposure groups preferred; having only on exposed group is not a categorical disqualification for meta-regression
<b>Representativeness of referent group</b>	Referent group characteristics (case-control) should be well-documented; referent group should be similar to exposed, with regard to key covariates
<b>Numbers of cases, controls, subjects</b>	Having sufficient numbers of cases, controls, subjects to support reliable statistical analyses is highly desirable; most serious issue is small number of referents in case-control studies
<b>Sensitive populations</b>	Where sensitive populations are identified, it is desirable to have separate exposure/dose-response evaluations for them

## D-R Methods – Multiple Study Bayesian Meta-Regression

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- **Bayesian Meta-Regression Analysis of Select NRC (2013) “Tier 1” health outcomes:**
  - **Purpose - Combine multiple studies to benefit from more (e.g., low dose) data**
  - **Approach:**
    - Multiple studies, single model, Bayesian meta-regression, and lifetable analysis
    - Model select NRC (2013) “Tier 1” health outcomes (e.g., bladder cancer, lung cancer, diseases of the circulatory system)

# EPA PBPK Model

## EPA PBPK model used for evaluation of Tier 1 health outcomes (El-Masri and Kenyon, 2008)

- validate with data from 11,500 people in Taiwan and Nevada
- Allow the incorporation of studies that reported urinary concentrations of iAs

## Dose-Response variabilities and uncertainties evaluated

- Study and model choice
- Exposure variability
- Population characteristics: e.g. age, BMI, smoking, sex, background incidence of disease, nutrition, arsenic metabolism and genetic polymorphisms

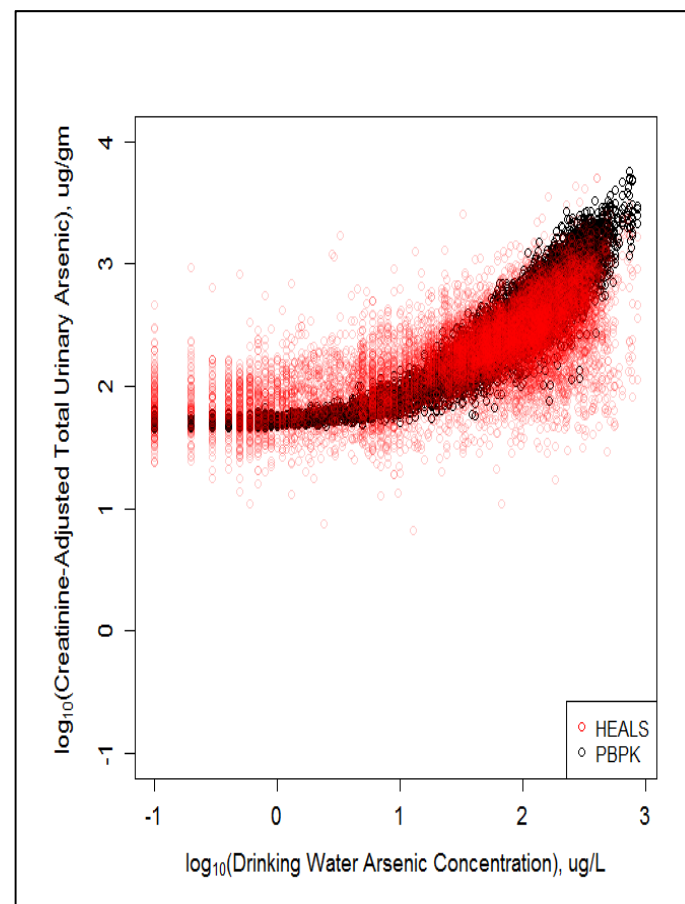
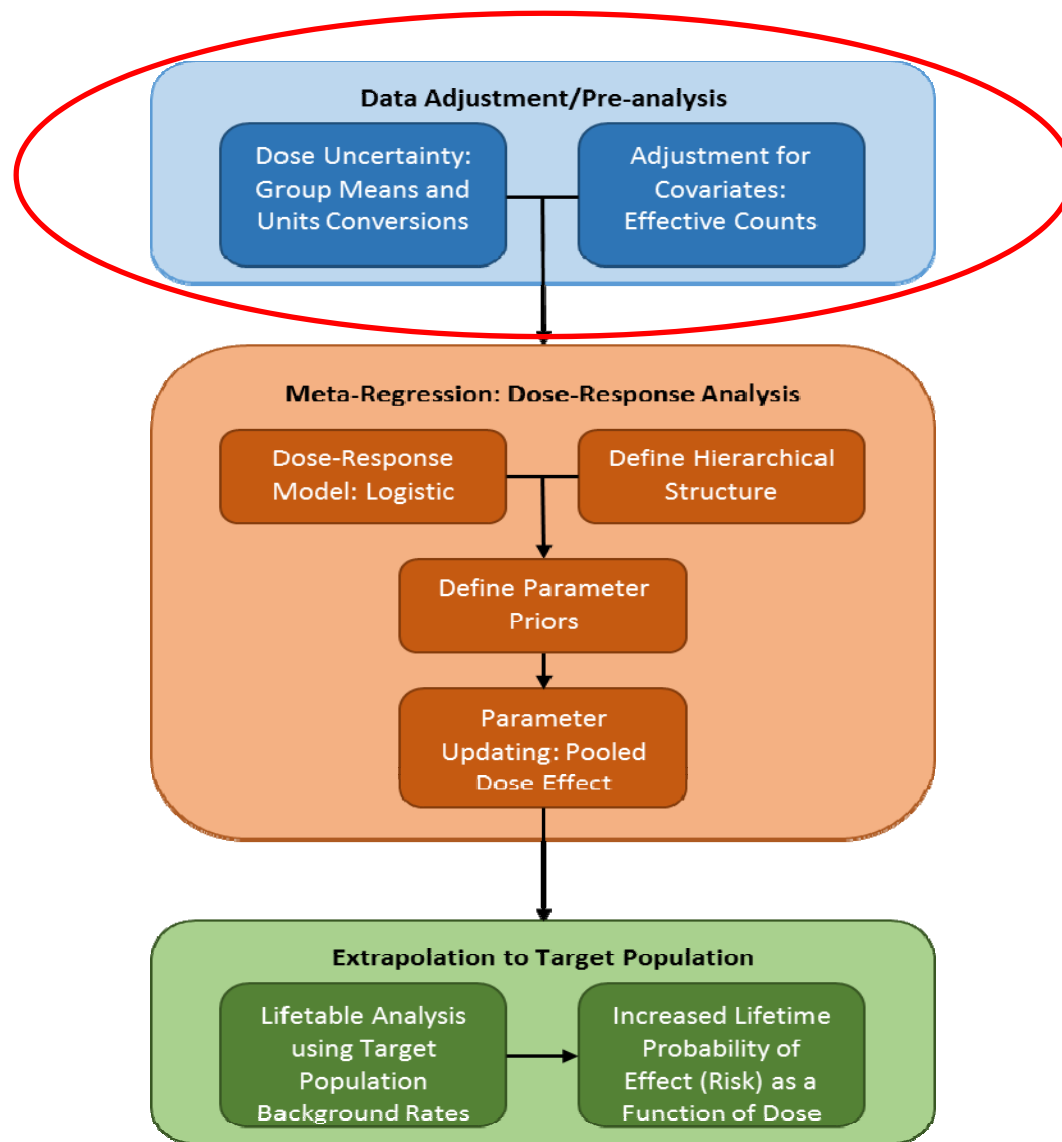


Figure 2. El-Masri and Kenyon (2008) PBPK model calibration against measured iAs total urinary concentrations and drinking water concentrations.



## D-R Methods - Meta-Regression Analysis Flow Chart



## Meta-Regression - Dose Uncertainty Pre-analysis

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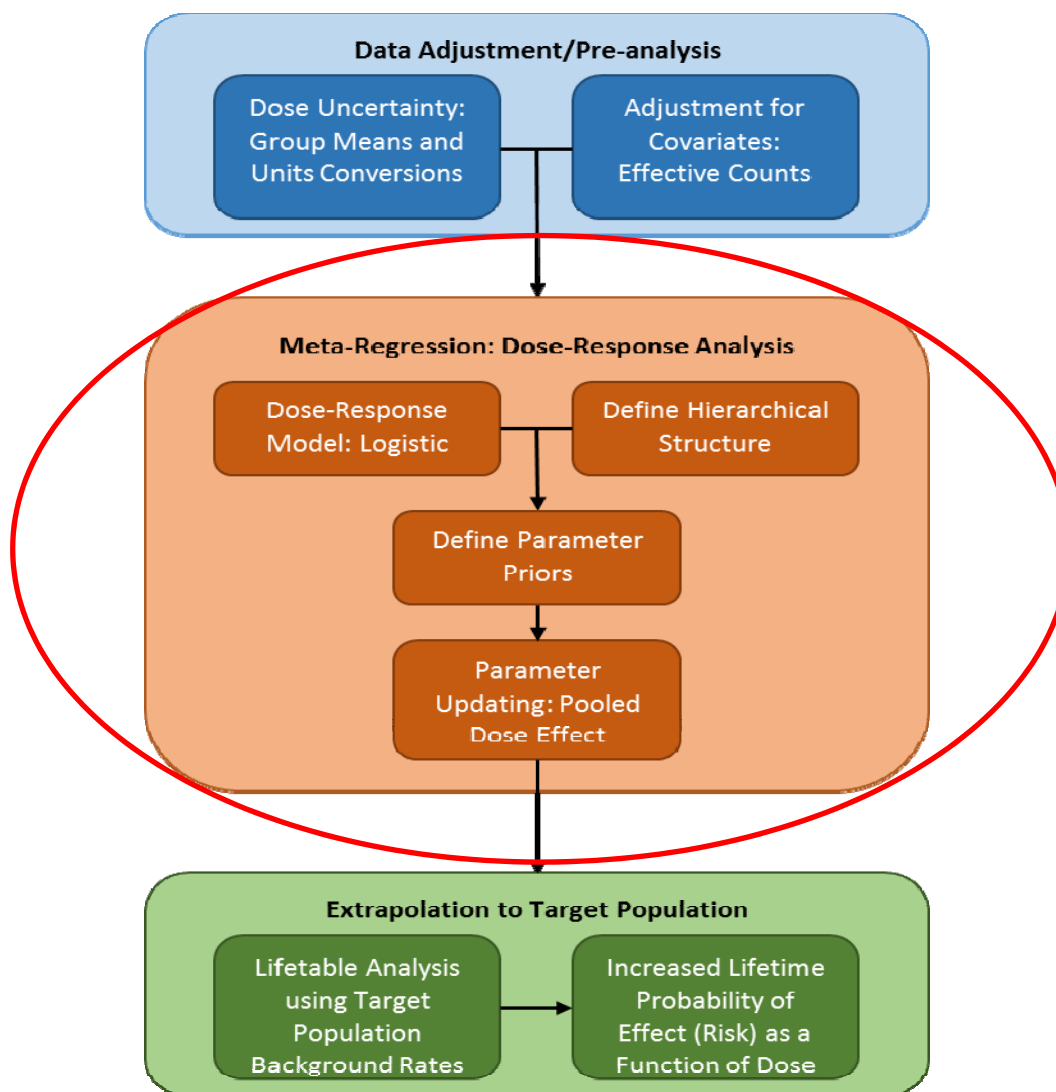
- **Most published reports provide ranges of exposures. Proposed approach:**
  1. **Compute means** (in units given in the published papers) – assume lognormal distribution; estimate mean for maximized profile likelihood (best estimate); then find MLE while minimizing (low estimate) and maximizing (high estimate) value of the high dose group (most uncertain and most influential dose group)
  2. **Convert to daily intake estimates.** If intake data not provided in study convert exposures to intake using population-specific estimates of key factors such as average duration of well exposure (yrs), average age at diagnosis (yrs), low water exposure ( $\mu\text{g/L}$ ), water consumption rate ( $\text{ml/kg-d}$ ) and dietary intake ( $\mu\text{g/kg-d}$ )
  3. **Estimate uncertainty about the mean daily intake estimates** via a Monte Carlo analysis that takes into account confidence intervals for key factors
    - a. Each iteration consists of drawing values from all the “key factor” distributions necessary for conversion to dose, then averaging according to the number of individuals in the exposure group
    - b. This will be repeated 1,000 times to derive a distribution and median, 2.5<sup>th</sup> and 97.5<sup>th</sup> percentiles obtained to characterize the best, low-end and high-end dose values

## Meta-Regression – “Effective Counts” Pre-analysis

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- **The proposed Bayesian approach is based on likelihoods of observing a particular number of cases** (e.g., the number of observed cases in a cohort study) where the expected number given is described by a Poisson distribution
- **“Effective counts”** are adjusted counts of cases and controls that reflect only the effect of arsenic
- **For all groups**, adjustments are made so as to mimic data that might have been collected had the covariate levels remained the same as in the referent group.

## iAs D-R Methods Overview- Meta-Regression Analysis Flow Chart



## Meta-Regression – Dose-response Modeling

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- **The Bayesian hierarchical under consideration** involves simultaneous modeling of individual studies to yield study-specific and pooled logistic model “ $\beta$ ” parameter estimates that describe the relationship between iAs exposure and a health outcome.
- **Use of hierarchical (mixed effects) modeling** recognizes that there are differences among studies that lead to variation in study-specific estimates.
- **Prior information about specific model parameter values** are incorporated.
- **Variation** is expressed in terms of Bayesian posterior distributions rather than single, constant-across-studies point-estimates.
- **The degree of heterogeneity across study-specific estimates** is estimated to indicate whether more complex hierarchical methods are necessary to account for study-specific differences in effects, or whether a fixed-effects model is more appropriate.

## Meta-Regression – Bayesian Priors

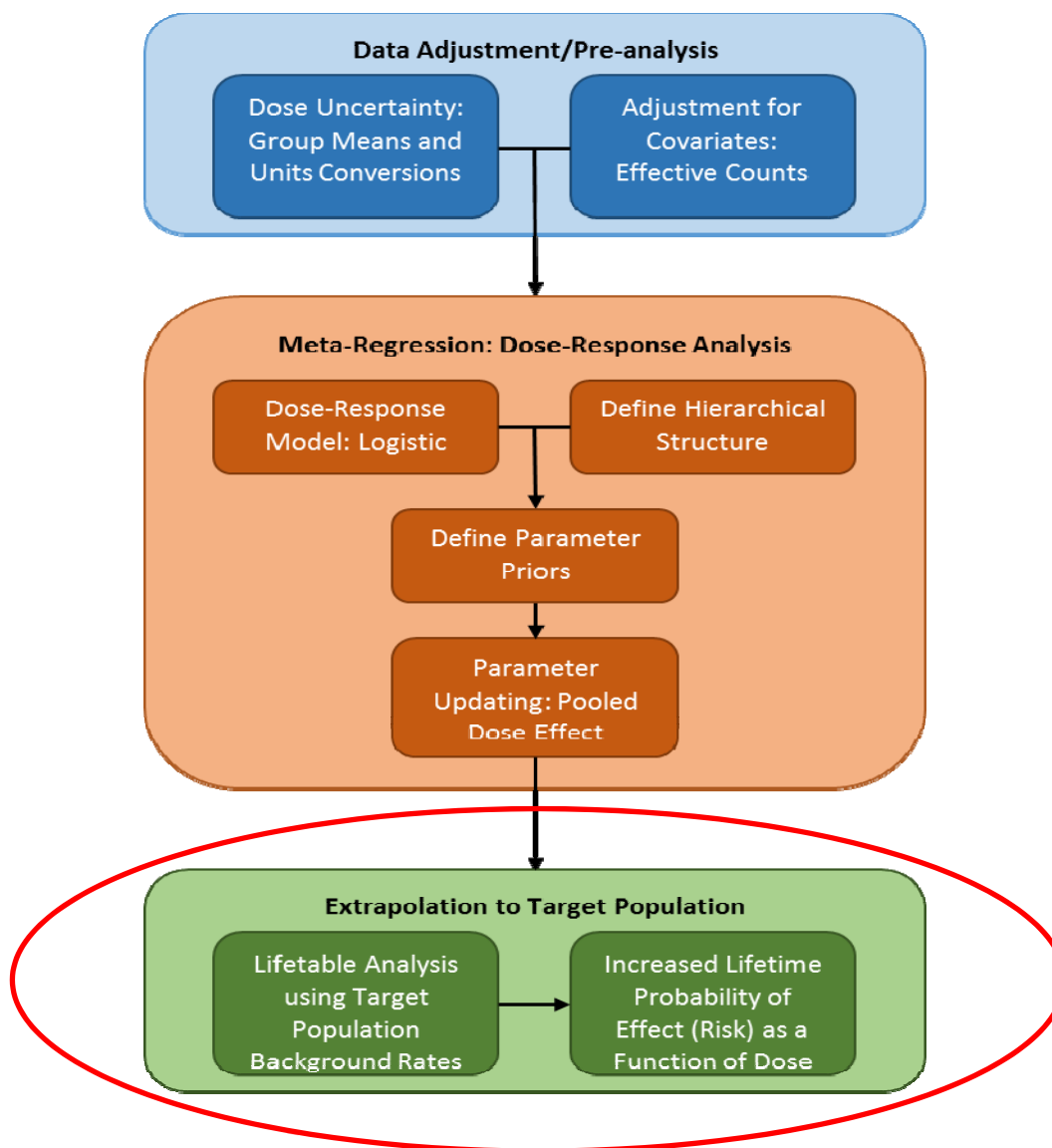
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- **Study-specific  $\beta$ s** – normal distribution assumed; allows positive and negative  $\beta$  values.
- **Pooled  $\beta$ s** - Gamma distribution assumed; allows only positive  $\beta$  values; reflects NRC (2013) determination that arsenic is causal for bladder cancer.
- **The  $\alpha$  and  $b$  parameters of the standard Gamma distribution (below)** can be set to reflect expected upper and lower risk boundaries at a given iAs dose.

$$f(x) = \alpha e^{-\alpha x} (\alpha x)^{b-1} / \Gamma(b) \quad x \geq 0$$

- **Gamma distributions** give greatest weight to  $\beta_{\text{mean}}$  values closest to zero.

# iAs D-R Methods Overview- Meta-Regression Analysis Flow Chart



## Meta-Regression – Extrapolation to Target Population

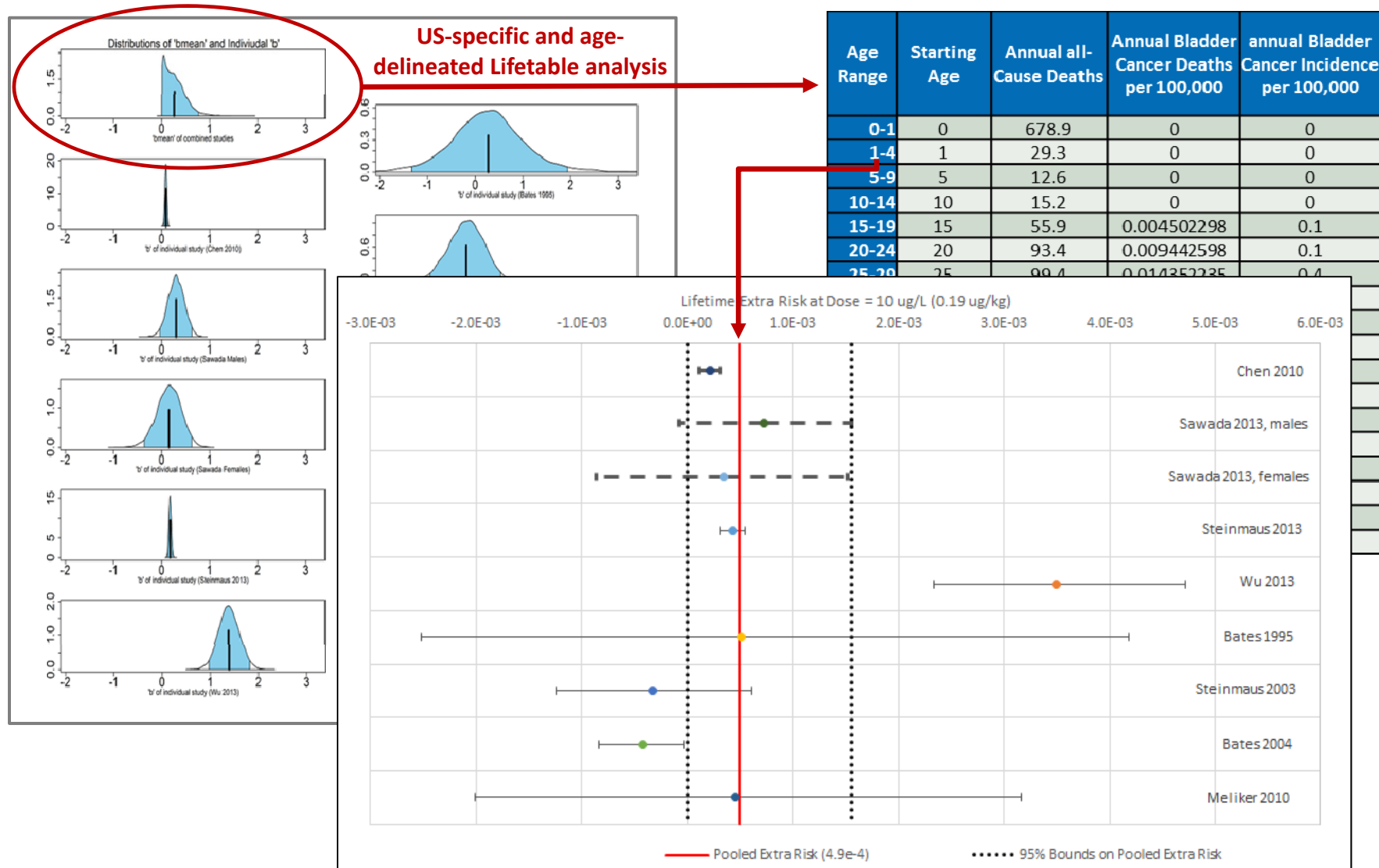
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- **Life-table type analyses for extrapolating extra risk to target populations** typically involve summing the probability of cancer at age intervals adjusted by the probability of dying from other causes at each age interval
- **Posterior distribution of a pooled “slope” (i.e., logistic  $\beta$  parameter)** can be used to characterize uncertainty about risk estimates at any level of daily intake.
- **Background rates reported for U.S. health outcomes can be expressed as zero extra risk from iAs at a mean U.S. background dose**
- **A benefit of this approach** is that one can consider impact of both positive changes (additional sources of iAs above background) and negative changes (reduction in background associated with, for example, clean-up of background sources) on lifetime cancer risks.



# Bayesian Meta-Regression (example; not an actual assessment)

## Combining 10 Studies, Logistic Modeling, and Lifetable Analysis



## Meta-regression – Additional Sensitivity Analyses

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### **Additional sensitivity analyses could include:**

- Expansion of dose-uncertainty Monte Carlo pre-analysis (slide 10),
- Choice of datasets (“leave-one-out” type analysis)
- Assumption of zero inhalation background exposure
- Consideration of alternative gamma prior distribution values

## Potential Science Issues

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- “Risk at a dose” (with confidence intervals) tables and equations versus traditional cancer CSF and noncancer RfD?
- Prior assumptions for meta-regression
  - Normal distribution for individual study slope estimates
  - gamma distribution of pooled slope estimate
  - $\alpha$  and  $b$  parameters of the standard Gamma distribution
- Model choice - possible future incorporation of model averaging with multiple fractional forms of the logistic model (raises issues for implementation of lifetable analysis)
- Nationwide background estimate and CDC cancer statistics - more targeted lifetable analyses could be done given population-specific background exposure estimate and age-interval cancer mortality and incidence data

## Acknowledgments for Development of Dose-Response Methods

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