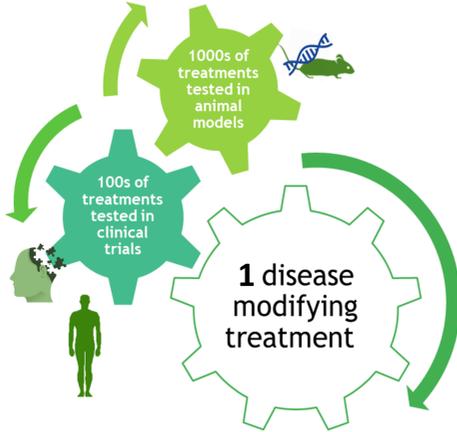


## 1. Objective

Meta-analysis of preclinical animal studies is vital in evaluating the internal and external validity. This informs the design and conduct of future preclinical studies and translation to a clinical setting. Investigating causes of heterogeneity between studies and choosing appropriate methods for estimation and modeling is crucial to perform a meta-analysis.



Assuming aggregate level data for a continuous primary outcome variable, we focus on two topics:

- Estimation of heterogeneity using three methods: method of moments (Dersimonian-Laird, DL), maximum likelihood (REML) and Bayesian approach (FB).
- Univariable versus multivariable meta regression for adjusting heterogeneity and exploring relative impact of multiple covariates on study level treatment effects.

## 2. Methodology

In random effects meta-analysis model, the treatment difference in studies are a random sample of independent observations.

$$\hat{\theta}_i = \theta + b_i + \epsilon_i \quad (1)$$

where  $b_i \sim N(0, \tau^2)$  and  $\epsilon_i \sim N(0, \xi_i^2)$  implying  $\hat{\theta}_i \sim N(\theta, \xi_i^2 + \tau^2)$ .

### Classical Estimation of Between Study Heterogeneity:

- REML estimator: A maximum likelihood-based method, where estimates of  $\tau^2$  and  $\theta$  are obtained through an iterative procedure.
- DL estimator: A method of moments-based estimate, where  $\tau^2$  can be derived by using the observed value of  $Q$  statistics and its expectation.

### Frequentist Approach to Meta Regression:

Heterogeneity between studies can be explored based on study design and characteristics by meta regression with  $\alpha_i = \beta_2 X_{2i}$  where  $X_{2i}$  is a covariate,  $b_i \sim N(0, \tau^2)$  and  $\epsilon_i \sim N(0, \xi_i^2)$ .

$$\hat{\theta}_i = \beta_1 + \alpha_i + b_i + \epsilon_i \quad (2)$$

### Bayesian Approach to Meta-Analysis and Heterogeneity Estimation:

$\hat{\theta}_i$  are the estimates for effect size  $N(\theta_i, \xi_i^2)$ , where prior for  $\theta_i$  is  $N(\theta, \tau^2)$  with hyperpriors  $\theta \sim N(\theta_0, \sigma_0^2)$ ,  $\tau^2 \sim \text{IG}(\alpha, \lambda)$ .

### Bayesian Meta Regression:

Prior for  $\theta_i$  is  $N(\mu_i, \tau^2)$ , where  $\mu_i = \beta_1 + \beta_2 x_{2i}$ .

## 3. Motivating Dataset

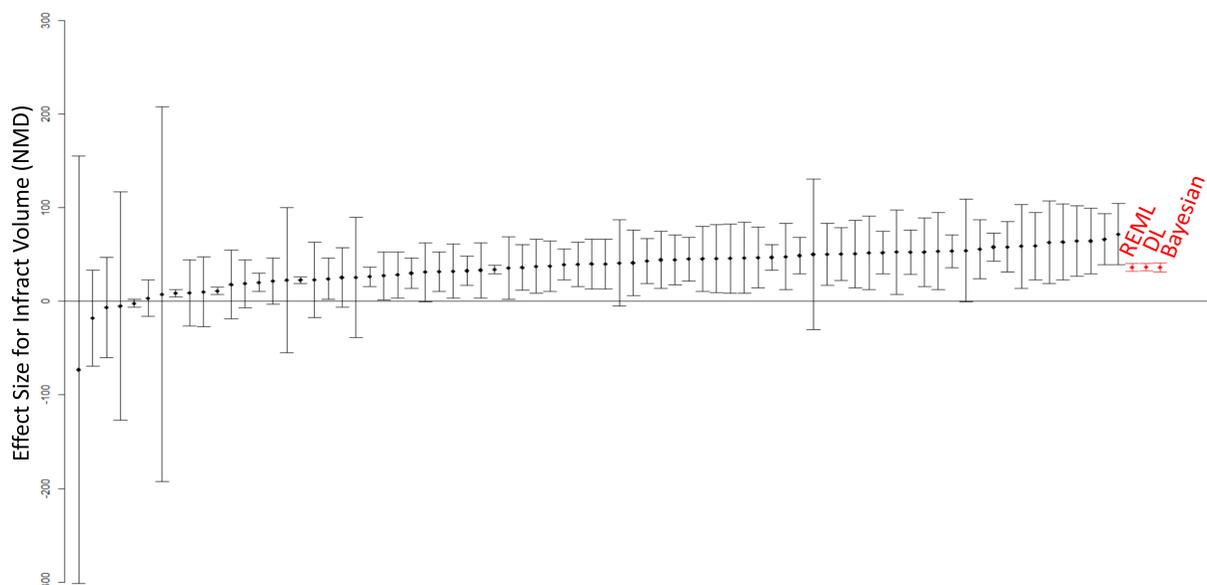


Figure 1: Effect of IL-1RA on infarct volume outcome with overall effect estimates obtained based on REML, DL and FB along with their 95% intervals.

- To exemplify the suggested methodology for meta-analysis, we reassess the study quality and design characteristics of systematic review data relating to the efficacy of Interleukin-1 Receptor Antagonist (IL-1RA) in animals exposed to focal cerebral ischaemia (modelling human with stroke).
- The difference in the effects of IL-1RA compared to a control group exposed to vehicle or to no treatment in animal studies of focal cerebral ischaemia is the outcome of interest. The primary endpoint was infarct volume.

## 4. Results

Table 1: (a) Summary of the selected univariate meta regression results using each study characteristic variables as a moderator to explain the extra heterogeneity present in infarct volume data, (b) Parameter estimates from multivariable meta regression model, where \* represents significant effect of a moderator. (FB noninformative priors:  $\beta \sim N(0, 10^4)$ ,  $\tau^2 \sim \text{IG}(0.001, 0.001)$  ran for 10000 iterations with 1000 burn-in period)

Method	Variable	Estimate (SE/MC Error)	95% CI	p-value	$\tau^2$	$I^2$	$R^2$
REML	No covariate	35.5(2.35)	(30.8, 40.2)	<.0001	213.9	82.4%	-
	Time to Outcome Measure(Intercept)	37.7(2.75)	(32.3, 43.2)	<.0001	206.4	80.8%	3.50%
	Time to Outcome Measure	-0.03(0.02)	(-0.06, 0.01)	0.12			
	Route of Drug Del.(Intercept)	31.5(3.27)	(24.9, 38.0)	<.0001	120.1	69.6%	43.9%
	Route of Drug Del.(ICerebVentricular)	20.3(5.54)	(9.25, 31.4)	0.001			
	Route of Drug Del. (Other)	-14.6(5.67)	(-26.0, -3.31)	0.01			
	Route of Drug Del.(Ivenous)	5.79(4.99)	(-4.19, 15.8)	0.25			
	Dose(Intercept)	38.0(2.91)	(32.2, 43.8)	<.0001	221.9	81.7%	0%
DL	No covariate	35.7(2.35)	(31.0, 40.4)	<.0001	235.2	83.7%	-
	Time to Outcome Measure(Intercept)	38.0(2.75)	(32.5, 43.5)	<.0001	253.3	83.8%	0%
	Time to Outcome Measure	-0.03(0.02)	(-0.06, 0.01)	0.13			
	Route of Drug Del.(Intercept)	32.4(3.31)	(25.8, 39.1)	<.0001	177.8	77.2%	24.4%
	Route of Drug Del.(ICerebVentricular)	19.3(5.49)	(8.28, 30.3)	0.001			
	Route of Drug Del. (Other)	-15.5(5.99)	(-27.4, -3.50)	0.01			
	Route of Drug Del.(Ivenous)	4.48(5.02)	(-5.57, 14.5)	0.38			
	Dose(Intercept)	38.4(2.91)	(32.5, 44.2)	<.0001	263.5	84.2%	0%
Bayesian	No covariate	35.5(0.09)	(30.3, 40.7)	-	224.2	-	-
	Time to Outcome Measure(Intercept)	37.6(0.05)	(31.7, 37.6)	-	215.2	-	-
	Time to Outcome Measure	-0.03(0.0004)	(-0.07, 0.01)	-			
	Route of Drug Del.(Intercept)	31.5(0.10)	(24.4, 39.3)	-	127.7	-	-
	Route of Drug Del.(ICerebVentricular)	20.2(0.15)	(7.78, 32.3)	-			
	Route of Drug Del. (Other)	-14.6(0.12)	(-27.7, -1.99)	-			
	Route of Drug Del.(Ivenous)	5.85(0.13)	(-5.50, 16.9)	-			
	Dose(Intercept)	37.9(0.06)	(31.7, 44.7)	-	232.9	-	-
REML	$\beta_1$ (Intercept)	42.9(5.53)*	(31.8, 54.0)				
	$\beta_2$ (Blinded Induction of Ischemia:True)	-11.5(7.03)	(-25.5, 2.60)				
	$\beta_3$ (Route of Delivery: Intracerebroventricular)	10.1(6.80)	(-3.55, 23.7)				
	$\beta_4$ (Route of Delivery: Intravenous)	-2.85(6.10)	(-15.1, 9.37)				
	$\beta_5$ (Route of Delivery: Other)	-24.4(6.66)*	(-37.8, -11.1)				
	$\beta_6$ (Time to Outcome Measure)	-0.03(0.02)	(-0.07, 0.01)				
	$\beta_7$ (Dose)	0.01(0.02)	(-0.03, 0.05)				
	DL	$\beta_1$ (Intercept)	43.6(5.47)*	(32.6, 54.5)			
$\beta_2$ (Blinded Induction of Ischemia:True)		-10.9(6.92)	(-24.7, 2.97)				
$\beta_3$ (Route of Delivery: Intracerebroventricular)		9.28(6.69)	(-4.12, 22.7)				
$\beta_4$ (Route of Delivery: Intravenous)		-3.71(6.04)	(-15.8, 8.39)				
$\beta_5$ (Route of Delivery: Other)		-24.7(6.84)*	(-38.4, -11.0)				
$\beta_6$ (Time to Outcome Measure)		-0.03(0.02)	(-0.07, 0.01)				
$\beta_7$ (Dose)		0.01(0.02)	(-0.03, 0.04)				
Bayesian		$\beta_1$ (Intercept)	39.0(0.22)	(27.4, 50.2)			
	$\beta_2$ (Blinded Induction of Ischemia:True)	-13.7(0.34)	(-31.1, 3.83)				
	$\beta_3$ (Route of Delivery: Intracerebroventricular)	12.6(0.23)	(-1.67, 27.2)				
	$\beta_4$ (Route of Delivery: Intravenous)	0.05(0.22)	(-13.1, 12.9)				
	$\beta_5$ (Route of Delivery: Other)	-20.9(0.21)	(-35.3, -6.25)				
	$\beta_6$ (Time to Outcome Measure)	-0.003(0.001)	(-0.05, 0.05)				
	$\beta_7$ (Dose)	0.01(0.001)	(-0.03, 0.06)				

Three methods yielded similar estimates for overall treatment effect, but between study variabilities vary.

Proportion of heterogeneity explained by a variable is larger for REML and FB than for DL.

Substantial drop in heterogeneity using multiple meta regression (Table 1 (b)):

- $\tau_{REML}^2=109.3, I_{REML}^2=68.3\%, R_{REML}^2=50.2\%$
- $\tau_{DL}^2=168.5, I_{DL}^2=76.9\%, R_{DL}^2=28.8\%$
- $\tau_{FB}^2=123.6$

## 5. Conclusions

- DL gives higher  $\tau^2$  than REML and FB; it is negatively biased when there is high heterogeneity due to false assumption that within study variabilities are known.
- REML reduces bias by excluding the summary effect parameter in  $\tau^2$  estimation but due to its iterative process, convergence is not guaranteed.
- Bayesian approach accounts for uncertainty in estimation and does not assume that within study variances are known akin to DL and REML.
- Multivariable meta-regression explains more heterogeneity between studies and leads to more generalisable results than simple random effects meta-analysis or univariate meta-regression.

## 6. Acknowledgements

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