

# Predicting individual disease progression including parameter uncertainty in rare neurodegenerative diseases: the example of Autosomal-Recessive Spastic Ataxia Charlevoix Saguenay (ARSACS)

NIELS HENDRICKX<sup>①,1</sup>, FRANCE MENTRÉ<sup>1</sup>, REBECCA SCHÜLE<sup>3,4</sup>, CYNTHIA GAGNON<sup>5</sup>, ARCA STUDY GROUP, EVIDENCE-RND CONSORTIUM,

ANDREAS TRASCHÜTZ<sup>3,4</sup>, MATTHIS SYNOFZIK<sup>3,4</sup>, EMMANUELLE COMETS<sup>1,2</sup>

<sup>①</sup>niels.hendrickx@inserm.fr, <sup>1</sup>Université Paris Cité, IAME, Inserm, F-75018, Paris, France, <sup>2</sup>Univ Rennes, Inserm, EHESP, Irset - UMR\_S 1085, 35000, Rennes, France.,

<sup>3</sup>Division Translational Genomics of Neurodegenerative Diseases, Hertie Institute for Clinical Brain Research (HIH), University of Tübingen, Tübingen, Germany, <sup>4</sup>German Center for Neurodegenerative Diseases (DZNE), Tübingen, Germany,

<sup>5</sup>Centre de recherche du CHUS et du Centre de santé et des services sociaux du Saguenay-Lac-St-Jean, Faculté de médecine, Université de Sherbrooke, Québec, Canada

## Context

- Genetic cerebellar ataxias are progressive ultra-rare neurodegenerative diseases affecting the cerebellum, causing debilitating impairment of gait, balance, speech and fine motor skill
- Over a hundred ataxia diseases are autosomal-recessive cerebellar ataxias (ARCA), often starting in early childhood or early adulthood
- We use Autosomal Recessive Spastic Ataxia of Charlevoix-Saguenay (ARSACS), one of the more common ataxias, as a case study
- There is an urgent need to develop robust statistical methodology that allows to predict progression trajectories<sup>1</sup>
- We develop a disease progression model to describe the individual evolution of the Scale for the Assessment and Rating of Ataxia (SARA) score versus Time Since Onset of symptoms (TSO) and its uncertainty

## Modelling SARA score

### Data description

- ARCA Registry<sup>2</sup>: longitudinal study with 173 ARSACS patients (1-7 visits, median 2)
- Core dataset including age, age of onset (AoO) as reported by the patients and SARA score
- Missing AoO imputed to the median AoO in the population (2 years)
- Covariates at inclusion: BMI, INAS score (Inventory of Non Ataxia Signs), genotype of the mutation (missense or loss of function), sex and Age of Onset

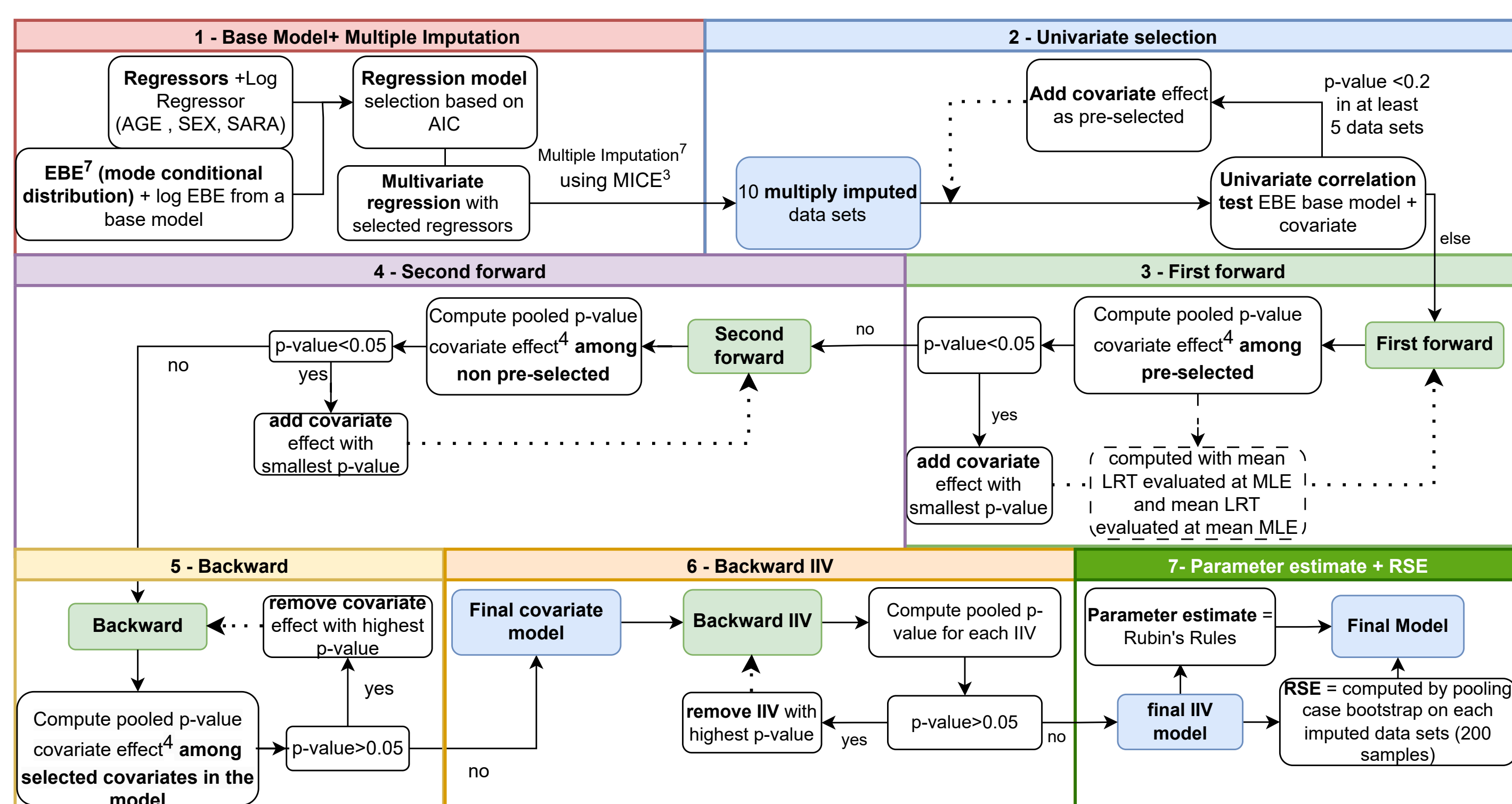
### Non linear mixed effect model

- Structural model: 4 parameter logistic function
- $f(TSO) = S_0 + (S_0 - S_m) \times \frac{1}{1 + e^{\alpha(TSO - T_{50})}}$
- constant error model, lognormal distribution for parameters, diagonal covariance matrix, covariate effect: linear on log parameter, (and log covariate for continuous)

Continuous Covariate	median	25-75%	nb missing
BMI ( $kg.m^{-2}$ )	24.4	21.4-29	36
INAS	7	5-8	67
AoO (yr)	2	1-6	22
Categorical covariate	Amount	percentage	nb missing
genotype of mutation	115 loss of function	82%	32
AoO 0-7	131	87%	-
AoO 8-14	7	4.5%	-
AoO 15-40	13	8.5%	-
Sex	92 females	52%	0

### Building covariate model via multiple imputation

- Many covariates have missing values = a specific approach is used<sup>4</sup>



### Implementation

- Parameter estimation performed with the SAEM algorithm (saemix<sup>5</sup> in R 4.2.0<sup>6</sup>).

### Results

Model	Without Covariates	With covariates		
Parameter	Value	RSE(%)	Value	RSE(%)
$S_0$	6.9	11	8.1	9
$\beta_{male, S_0}$	-	-	-0.28	50
$S_m$	35.8	2	34.3	2
$\beta_{male, S_m}$	-	-	0.08	38
$\alpha (yr^{-1})$	0.11	13	0.12	12
$T_{50} (yr)$	36.8	3	38.1	3
$\beta_{AOO > 15, T_{50}}$	-	-	-0.43	20
$\omega_{S_0}$	0.27	37	0.19	42
$\omega_{\alpha}$	0.59	19	0.58	22
$\omega_{T_{50}}$	0.21	11	0.16	11
$\sigma$	1.92	6	1.86	6

- RSE computed with case bootstrap (200 samples)
- 8 covariates selected as pre-selected, 3 final covariates selected during first forward (none have missing values)

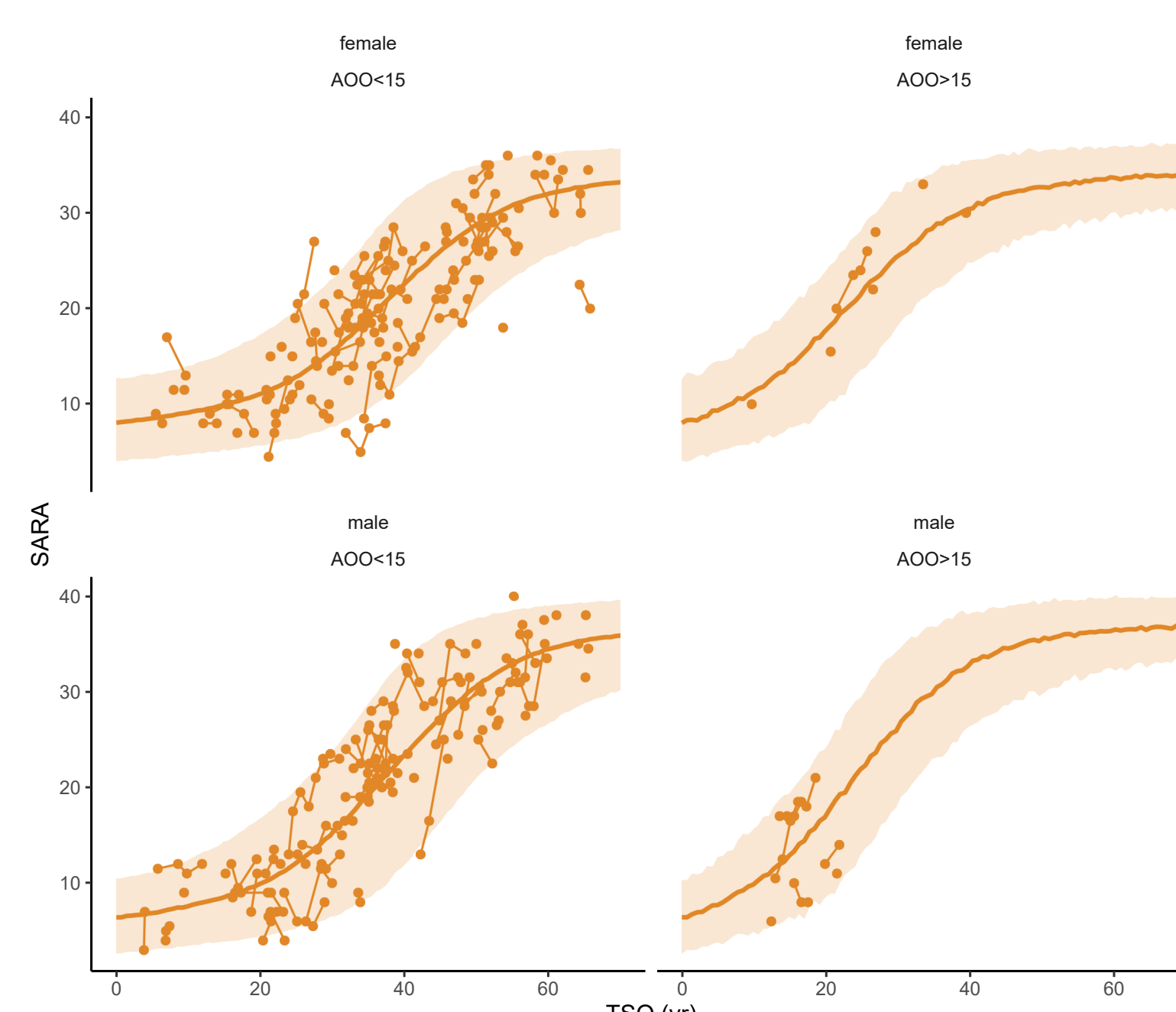
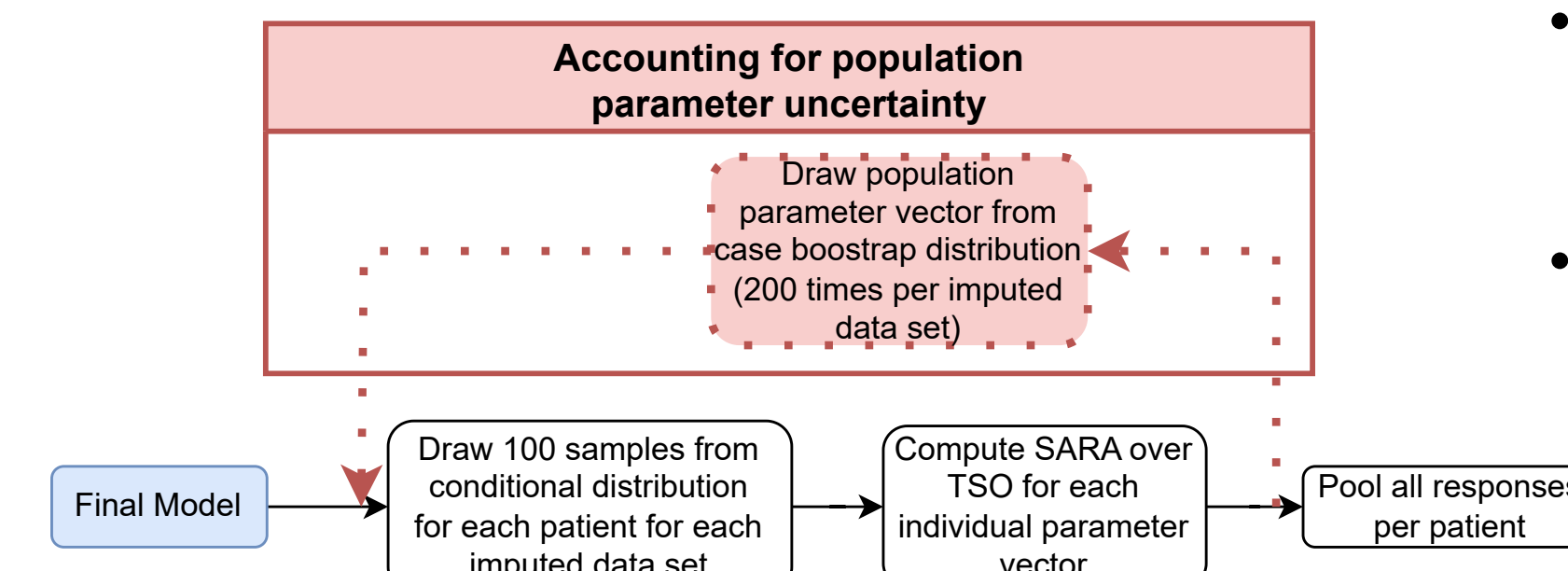


Figure 1: Spaghetti plot with simulations, median + 90% Simulation Interval, stratified by covariates

## Method



## Results

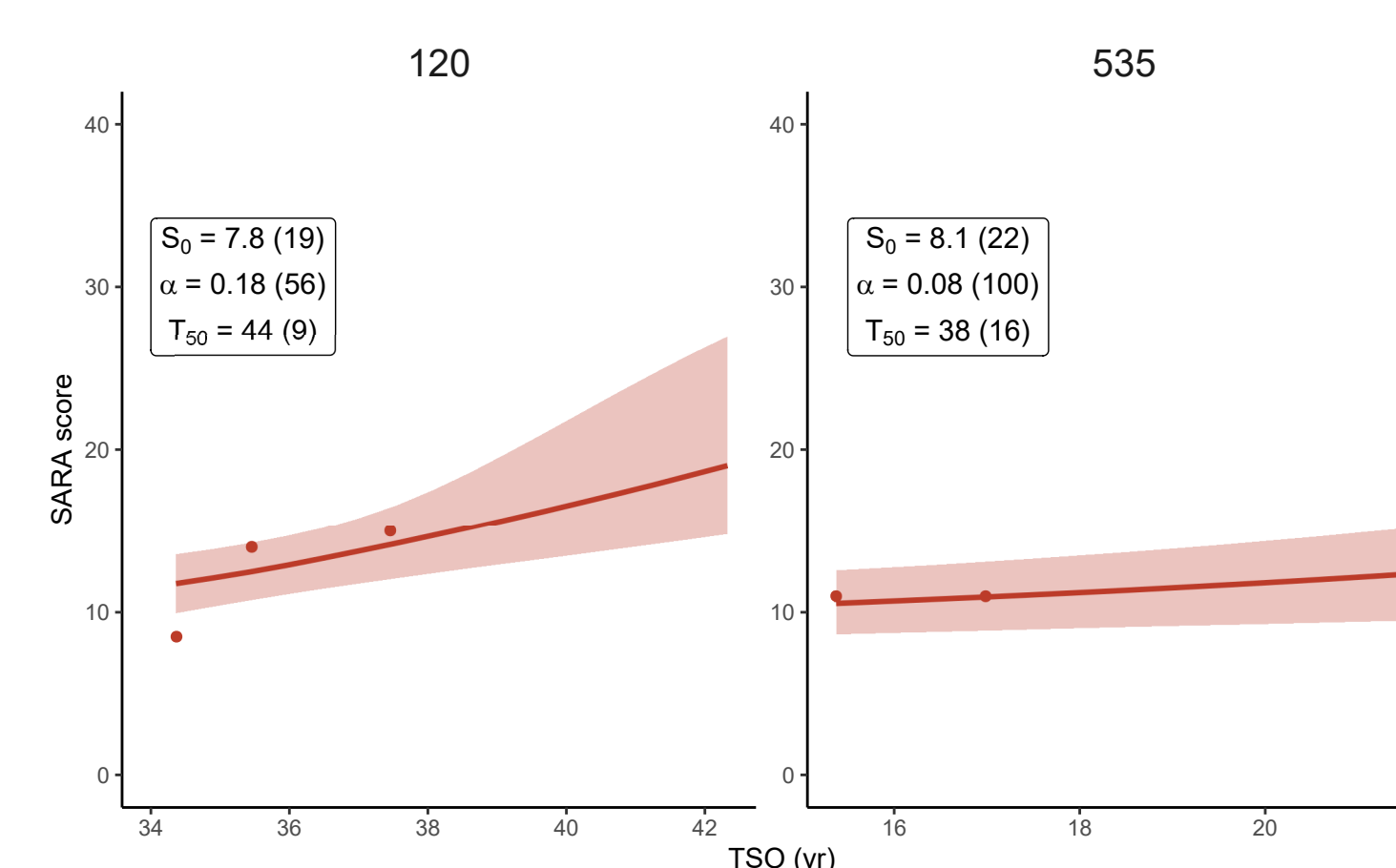


Figure 2: Example of individual predictions with population parameter uncertainty (median + 90% PI) and individual parameter estimates (estimated as the median + RSE (%) of conditional distributions)

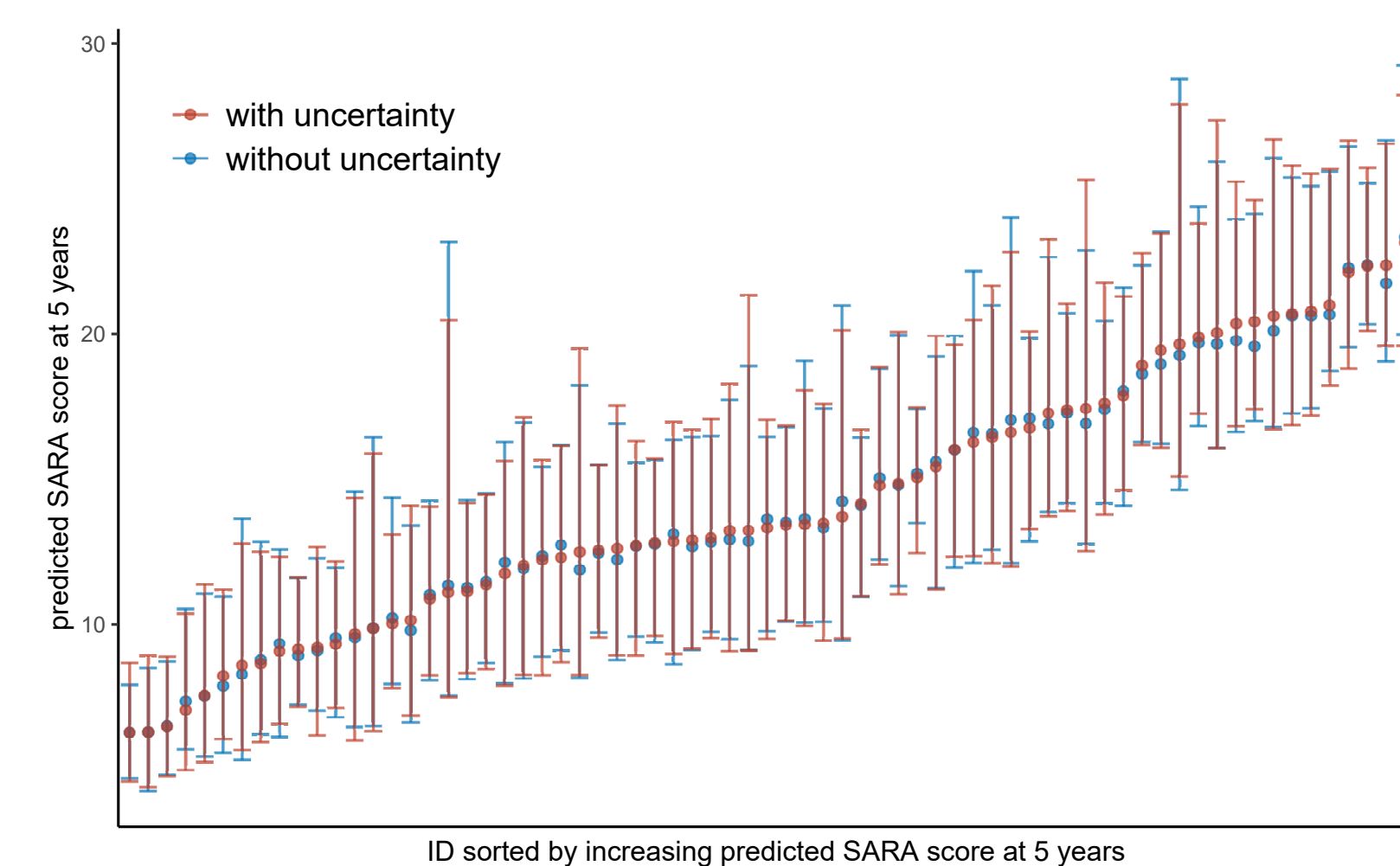


Figure 3: Distribution of the predicted SARA score and its 90% PI at 5 years, with and without population parameter uncertainty

## Individual predictions

- Select all patients with last SARA < 20 (70)
- Compute predicted SARA and width 90% Prediction Interval (PI) at 5 years after last visit
- Compare values with and without population parameter uncertainty
- Parameter standard deviation depends on the number of observation (ex: patient 120) but the size of the PI is more influenced by the fact that TSO is close to the inflexion point ( $T_{50} = 44$ )
- Median (90% PI) increase in SARA score at 5 years of +2.9 points (0.33-5.1)

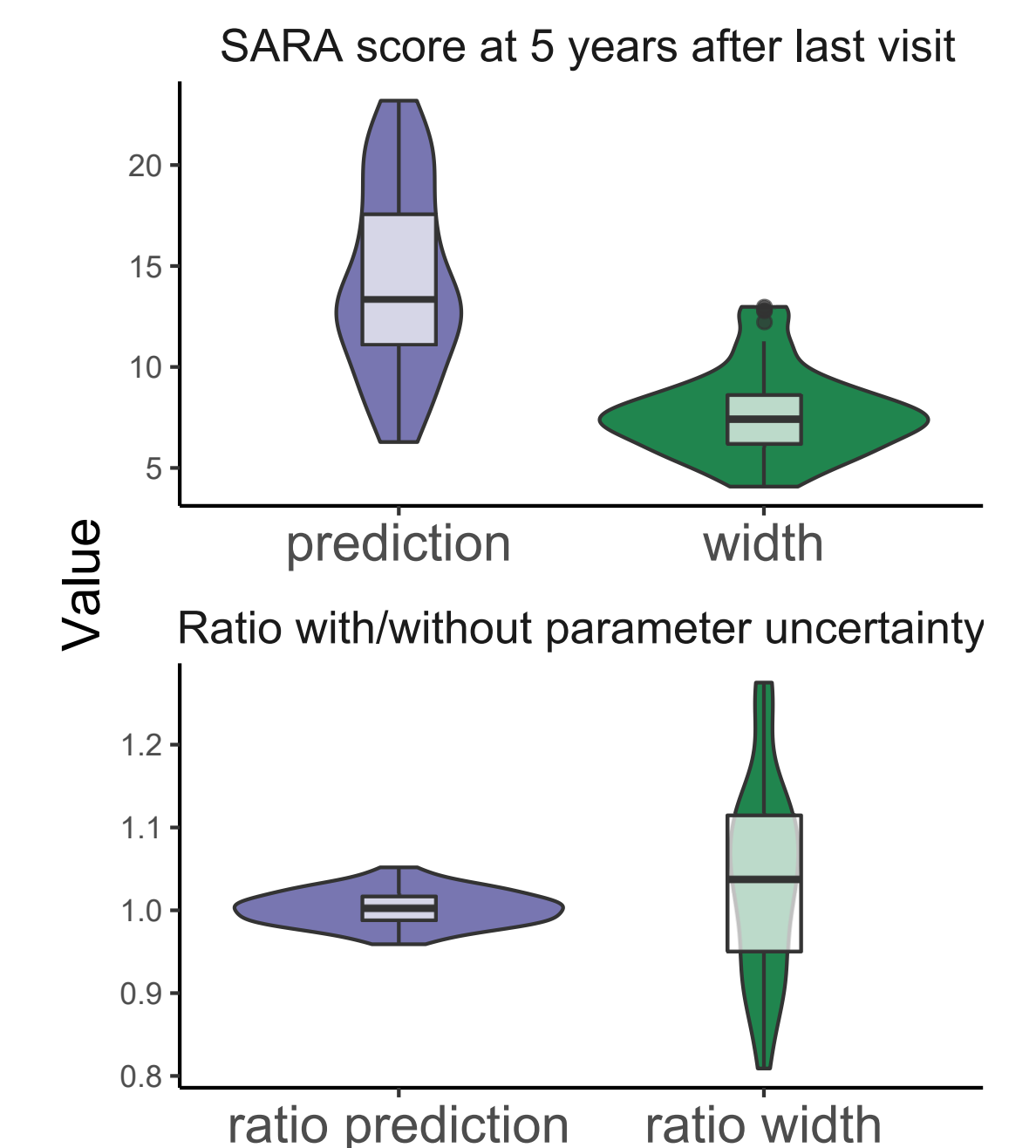


Figure 4: Violin plot of the predicted SARA score and its width at 5 years, and its ratio with/without uncertainty

- Median (90% PI) SARA score/width at 5 years of 13/7.4 (7.3-22)/(4.7-12)
- Population parameter uncertainty: little impact, likely due to the small RSE in the model

## Conclusion

- Disease progression in ARSACS could be described using a logistic growth function. The score at onset of disease ( $S_0$ )/maximum SARA score ( $S_m$ ) was estimated at 8.1/34.3 for females, 6.1/37.2 for males. Model predicts faster progression for late onset patients
- We used a method to select a covariate model with Multiple Imputation, accounting for missing values in covariates. In the final model, none of the selected covariates had missing values
- Parameter uncertainty did not change the predicted individual SARA score and width of PI
- Individual predictions while accounting for parameter uncertainty will be implemented in smaller populations in the registry (and most likely with higher RSE due to the small amount of patients)

<sup>1</sup> Synofzik et al. *J Nucl. Med.* 2022; <sup>2</sup> Träschütz et al. *Front Neurol* 2021; <sup>3</sup> van Buuren et al. *J Stat Softw* 2011; <sup>4</sup> Meng et al. *Biometrika* 1992; <sup>5</sup> Comets et al. *J Stat Softw* 2017; <sup>6</sup> R core team et al. *R Foundation for Statistical Computing* 2022. <sup>7</sup> Johansson et al. *AAPS J* 2013.

## Acknowledgements

This work was supported by the European Joint Programme on Rare Diseases (EJPRD) WP20 Innovation Statistics project "EVIDENCE-RND" and the EJPRD PROSPAX consortium (DFG, German Research Foundation, No 441409627), both under the EJP RD Grant Agreement (n°825575). It was further supported, in part, by the Clinician Scientist program "PRECISE.net" funded by the Else Kröner-Fresenius-Stiftung.