



# Simulx webinar: Optimizing Sample Size of a Phase III trial

We will start soon



# Introduction

## Reference asthma treatment

- Current standard-of-care
- Approved globally (also in China)

## Reference asthma treatment

- Current standard-of-care
- Approved globally (also in China)

## New asthma treatment

- FDA approved
- does not include Chinese patients

## Reference asthma treatment

- Current standard-of-care
- Approved globally (also in China)

## New asthma treatment

- FDA approved
- does not include Chinese patients



Plan a bridging study for approval in China

## Reference asthma treatment

- Current standard-of-care
- Approved globally (also in China)

## New asthma treatment

- FDA approved
- does not include Chinese patients



Plan a bridging study for approval in China



*What is the sample size* required for a trial in Chinese patients to show a difference in response between the two treatments?

## Data

- phase III, NEW treatment,  
NO Chinese patients

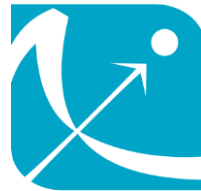
phase III, REF treatment,  
NO Chinese patients

phase II, REF treatment,  
ONLY Chinese patients

## Data



## Population modeling in Monolix



phase III, NEW treatment,  
NO Chinese patients

phase III, REF treatment,  
NO Chinese patients

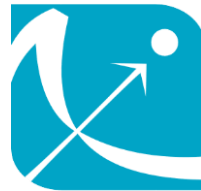
phase II, REF treatment,  
ONLY Chinese patients

- single **population model** covering the 3 datasets
- Investigate the **impact of the covariates** for Chinese and non-Chinese patients

## Data



## Population modeling in Monolix



## Clinical trial simulations in Simulx



phase III, NEW treatment,  
NO Chinese patients

phase III, REF treatment,  
NO Chinese patients

phase II, REF treatment,  
ONLY Chinese patients

- single **population model** covering the 3 datasets
- Investigate the **impact of the covariates** for Chinese and non-Chinese patients

- **Predict the response** to the new treatment in Chinese asthma patients
- Suggest a **minimal sample size** for China bridging study

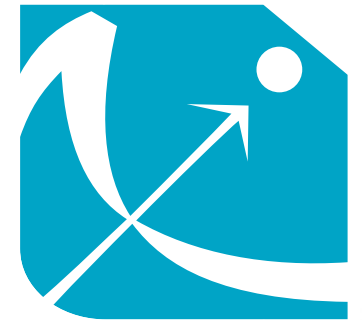


**Clinical trial simulation (CTS)** = abstraction of the clinical trial process based on a longitudinal model

It can be used to:

- investigate assumptions such as:
  - sample size
  - study duration
  - treatment schedules
  - population characteristics
  - ...
- quantify their influence on trial designs with:
  - outcomes
  - type I error
  - study power
  - ...
- evaluate the adequacy and efficiency of potential trial designs

# POPULATION MODELING WITH MONOLIX



## Measurements:

- FEV1: forced expiratory volume = maximum amount of air one can forcefully exhale in one second

## Studies:

- 2 studies from phase III (global) comparing the two treatments + placebo
- 2 studies from phase II (Chinese) for reference treatment only + placebo

## Covariates:

- age, sex, weight, height, race, treatment, baseline FEV1, disease severity

# Dataset: observations

Treatments:

Reference

Reference lower dose

Test

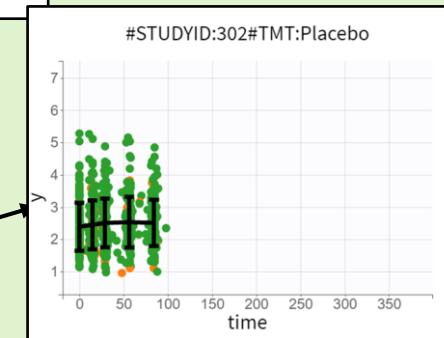
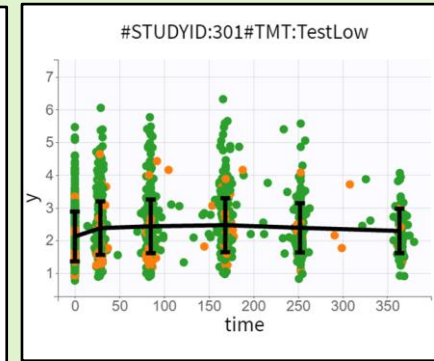
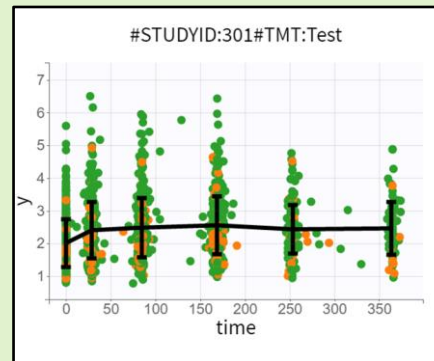
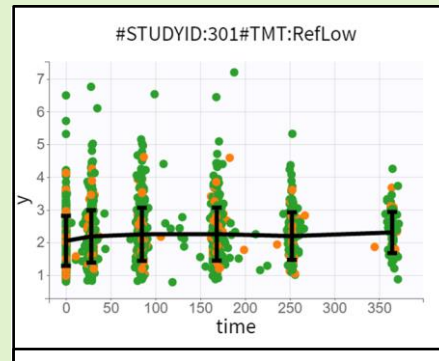
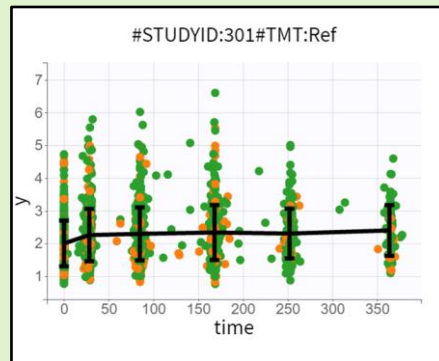
Test lower dose

Placebo

Phase III  
studies

Population:

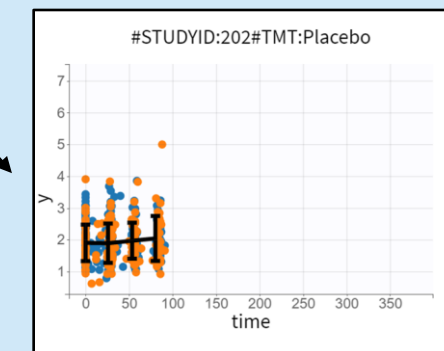
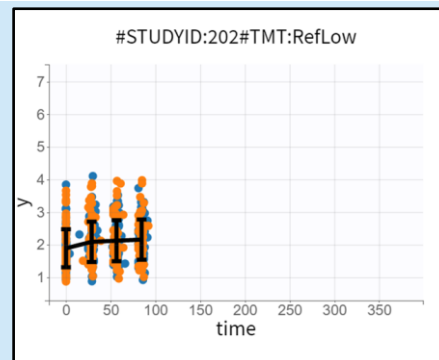
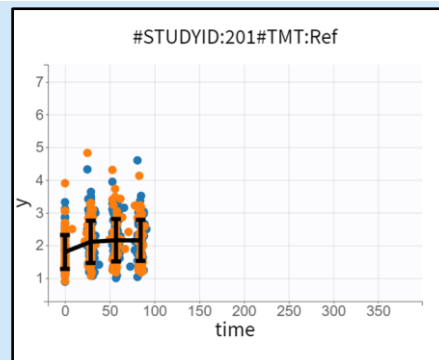
White  
Other



y = forced expiratory  
volume: maximum amount  
of air one can forcefully  
exhale in one second

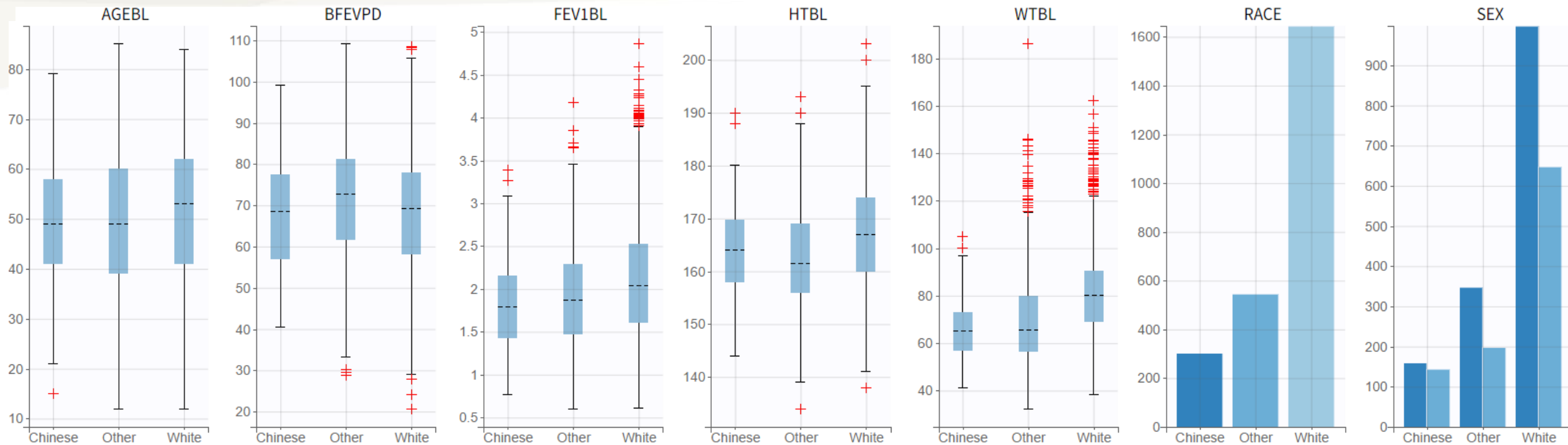
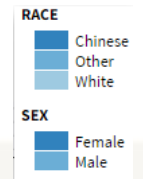
Phase II  
studies

Population:  
Chinese  
Other



# Dataset: covariates

## Comparison of covariate distributions in the different populations



## Structural model:

[LONGITUDINAL]

input = {**A**, **G**, Td, gamma}

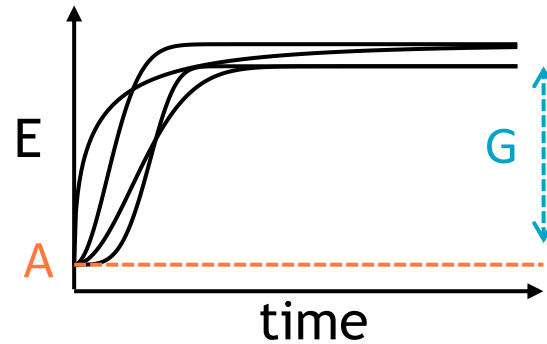
EQUATION:

$S = G * (1 - \exp(-(1/Td * t)^\gamma))$

$E = \max(1e-3, A + S)$

OUTPUT:

output = E



We use an exponential model to describe the placebo FEV1 trajectory over time.

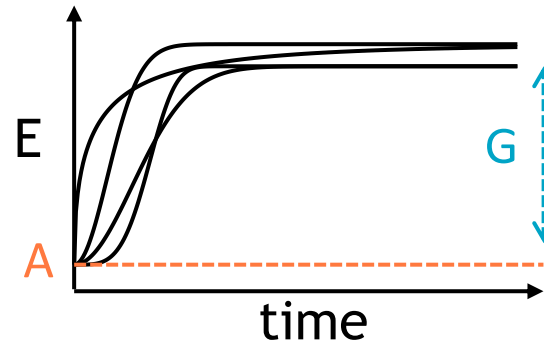
[LONGITUDINAL]

EQUATION:

$$E = \max(1e-3, A + S)$$

OUTPUT:

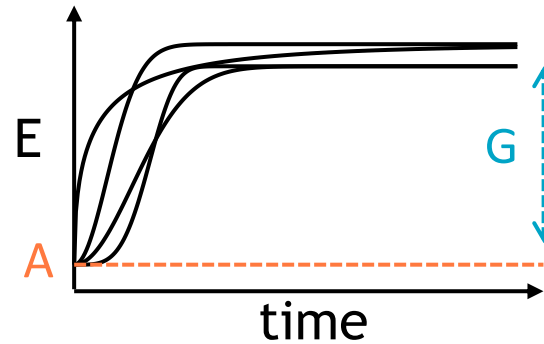
output = E



The final statistical model includes many covariate effects.

[illegible]

```
[LONGITUDINAL]
input = {A, G, Td, gamma}
EQUATION:
S = G * (1-exp(-(1/Td*t)^gamma))
E = max(1e-3, A + S)
OUTPUT:
output = E
```



The treatment has an effect on the gain parameter  $G$

[illegible]



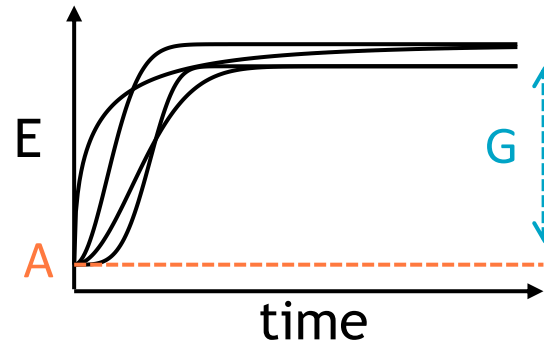
[LONGITUDINAL]

EQUATION:

$$E = \max(1e-3, A + S)$$

OUTPUT:

output = E



TMT	Typical G
Placebo	0.9
RefLow	1.34
Ref	1.59
TestLow	1.72
Test	2.30

# Statistical model

[illegible]

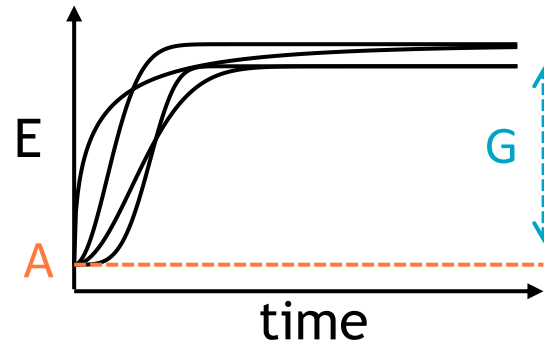
[LONGITUDINAL]

EQUATION:

$$E = \max(1e-3, A + S)$$

OUTPUT:

output = E



RACE and FEV1BL  
have an effect on the  
baseline parameter **A**

[illegible]

# PD model

## Structural model: exponential

[LONGITUDINAL]

input = {A, G, Td, gamma}

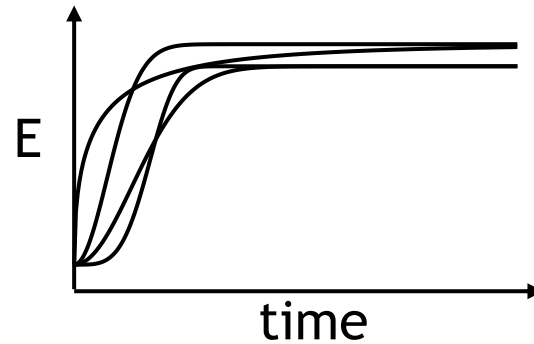
EQUATION:

$$S = G * (1 - \exp(-(1/T_d * t)^\gamma))$$

$$E = \max(1e-3, A + S)$$

OUTPUT:

output = E

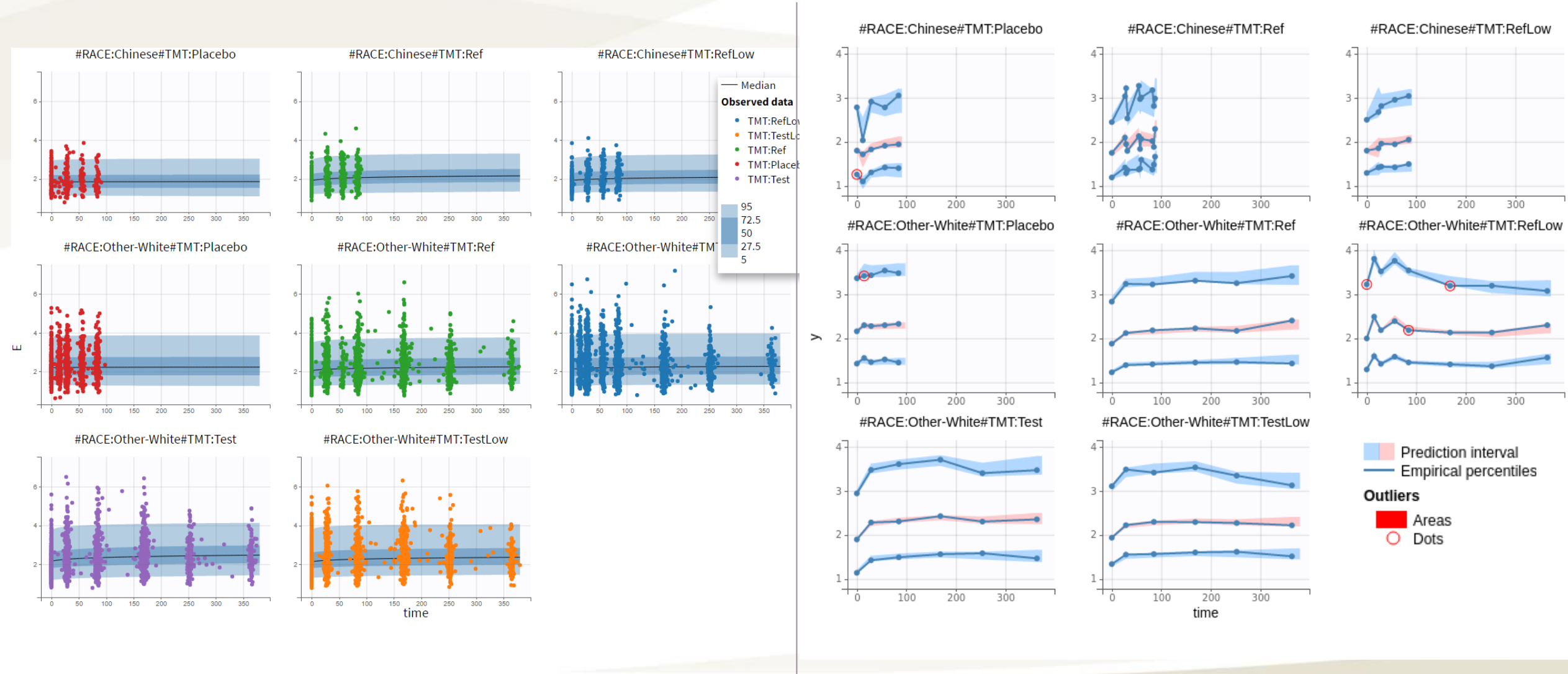


RACE	Typical A
White	1.32
Chinese	1.26
Other	1.27

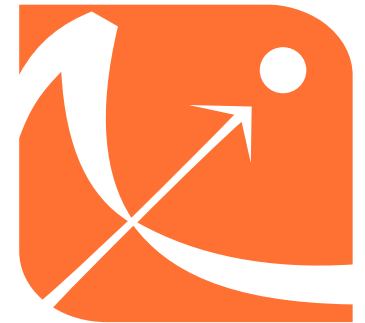
## Statistical model

[illegible]

# PD model estimated in Monolix



# CLINICAL TRIAL SIMULATION WITH SIMULX

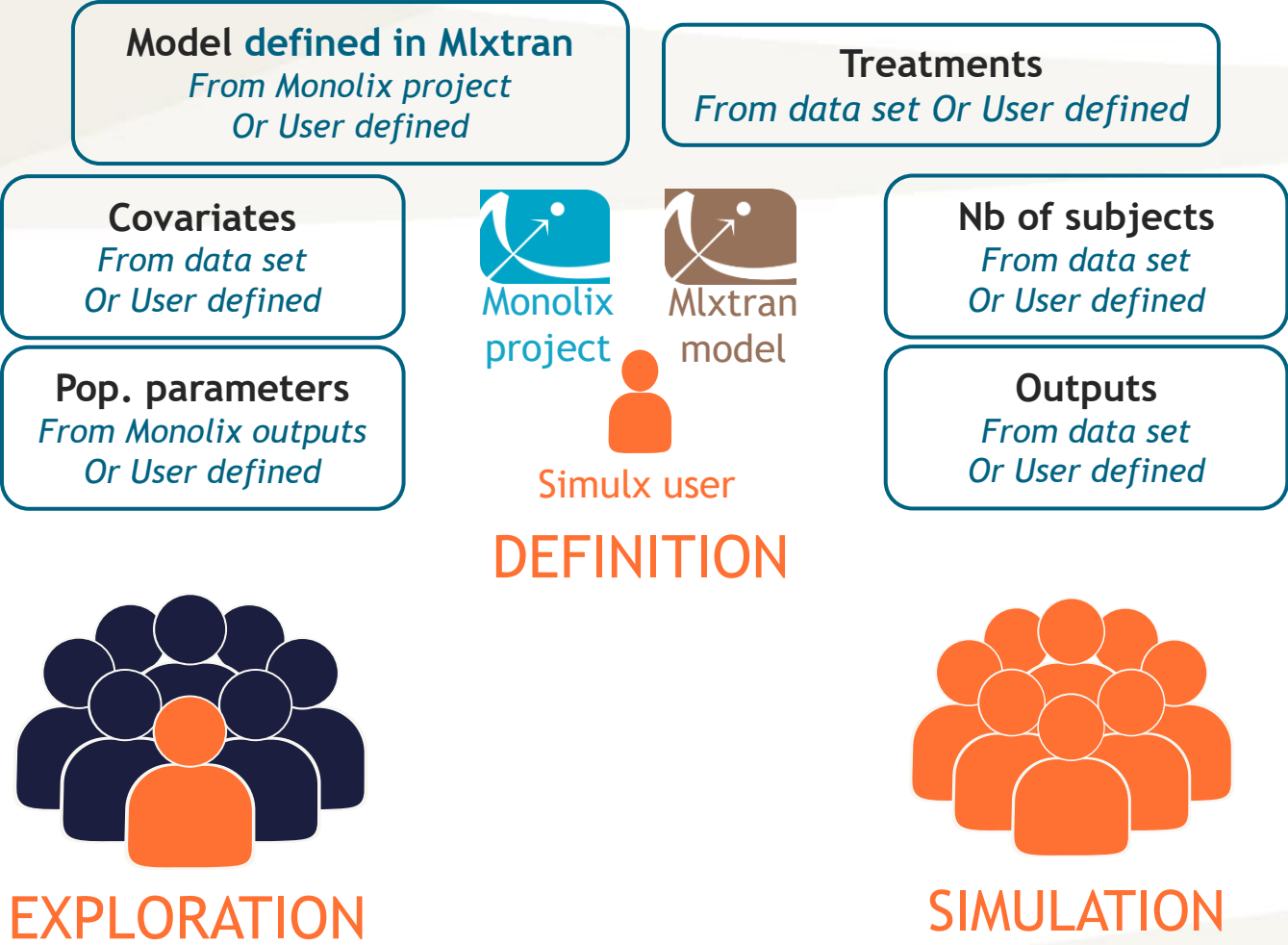


# What is Simulx?

- Advanced **simulator** of clinical trials and **decision-making** tool
- Easy-to-use and interactive **interface**
- **Interconnected** with MonolixSuite applications
- **Intuitive** workflow
- **Flexible** in building simulation scenarios
- Advanced computational capabilities with **C++ engine**
- Immediate **visual feedback**
- **Export** of plots and results



# Simulx: how does it work?



# Simulating a phase III clinical trial



## Simulation

Comparing reference and new treatment in a population of **75 Chinese patients**,  
**during 3 months or 6 months.**



# Simulating a phase III clinical trial



## Simulation

Comparing reference and new treatment in a population of **75 Chinese patients**,  
**during 3 months or 6 months.**



## Quantitative Outcome and Endpoint

What is the **mean FEV1 change from baseline at the end of the study?**

# Simulating a phase III clinical trial



## Simulation

Comparing reference and new treatment in a population of **75 Chinese patients**,  
**during 3 months or 6 months.**



## Quantitative Outcome and Endpoint

What is the **mean FEV1 change from baseline at the end of the study?**



## Trial Success

Is the mean FEV1 change from baseline **significantly higher with the new treatment?**

# Simulating a phase III clinical trial



## Simulation

Comparing reference and new treatment in a population of **75 Chinese patients**,  
**during 3 months or 6 months.**



## Quantitative Outcome and Endpoint

What is the **mean FEV1 change from baseline at the end of the study?**



## Trial Success

Is the mean FEV1 change from baseline **significantly higher with the new treatment?**



## Power of the study = probability of success

**Variability between replicate studies and uncertainty** of our model can affect the results

## LixoftConnectors

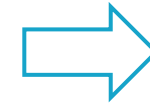
- API for R that allows to create/modify/run a Monolix, Simulx or Pkanalix project from R
- R package provided with the MonolixSuite installation



To automate a workflow

## Additional R packages

- R functions calling the connectors to go beyond what the interface allows
- Packages **RsMlx**, **RsSimulx**, and **RsPkx**



To run advanced specific tasks

# Sample size optimization with LixoftConnectors

```
library(lixoftConnectors)
initializeLixoftConnectors(software="simulx")

loadProject(projectFile = "ChineseTrial.smlx")

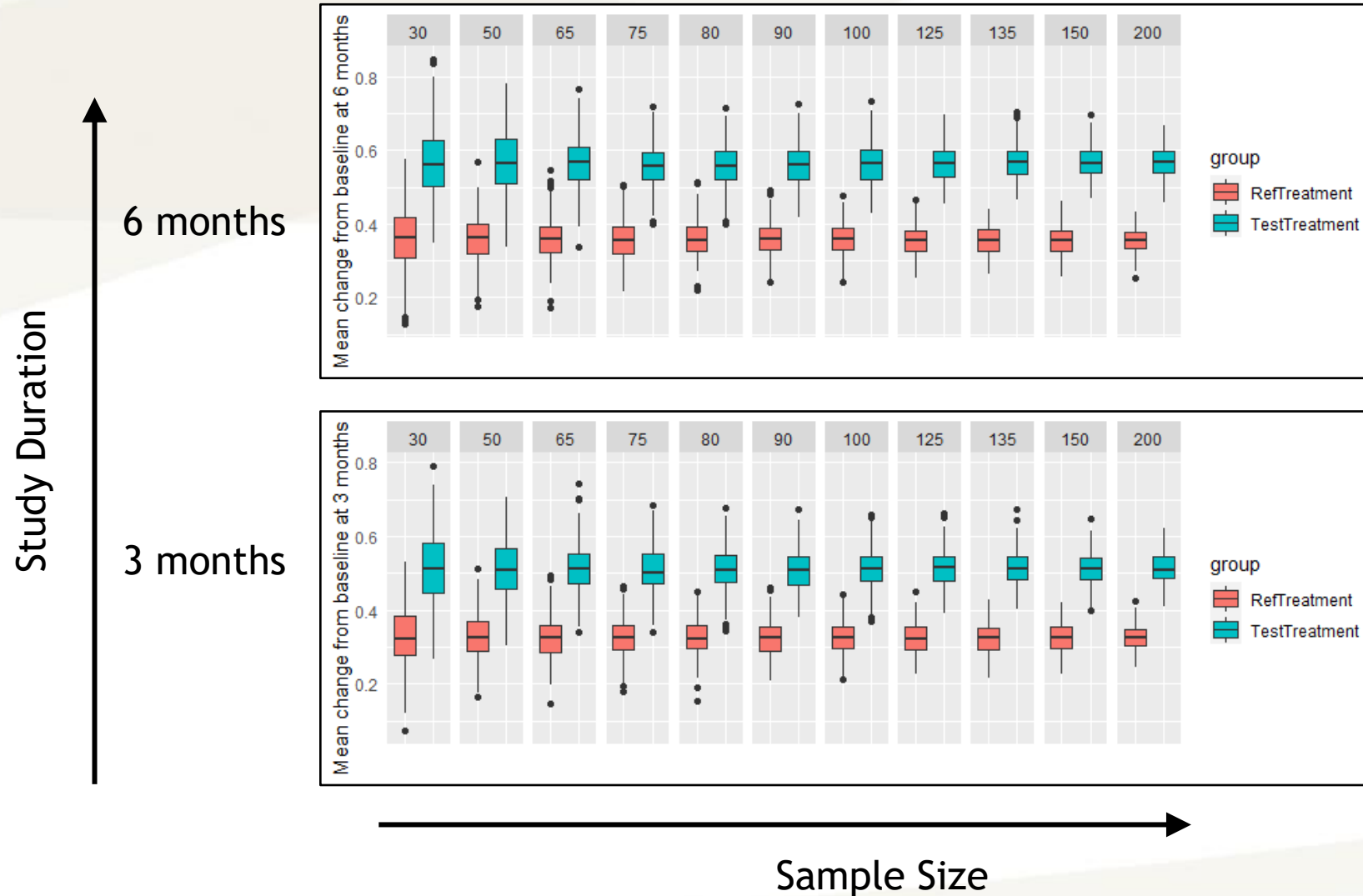
sample_sizes <- c(30, 50, 65, 75, 80, 90, 100, 125, 135, 150, 200)

for(N in sample_sizes){

  setGroupSize("test_treatment", N)
  setGroupSize("ref_treatment", N)
  runSimulation()

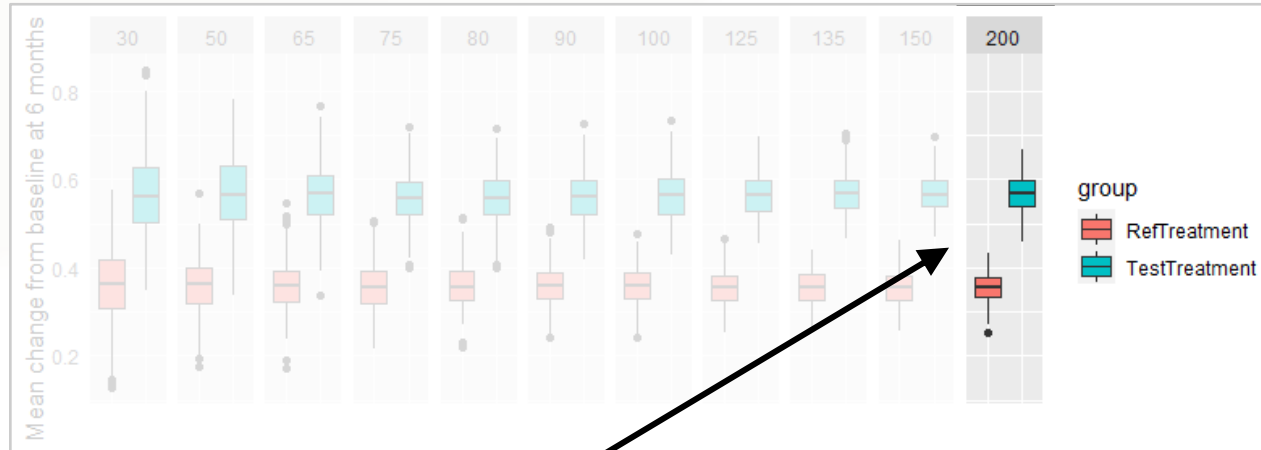
  sim <- getSimulationResults()$res$y
  ...
}
```

# Sample size optimization with LixoftConnectors

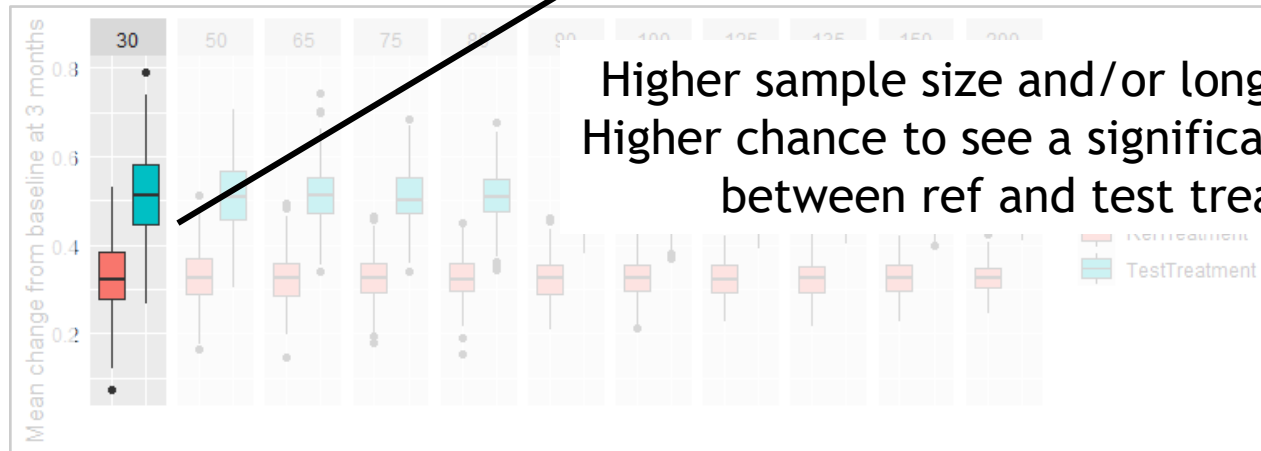


# Sample size optimization with LixoftConnectors

6 months



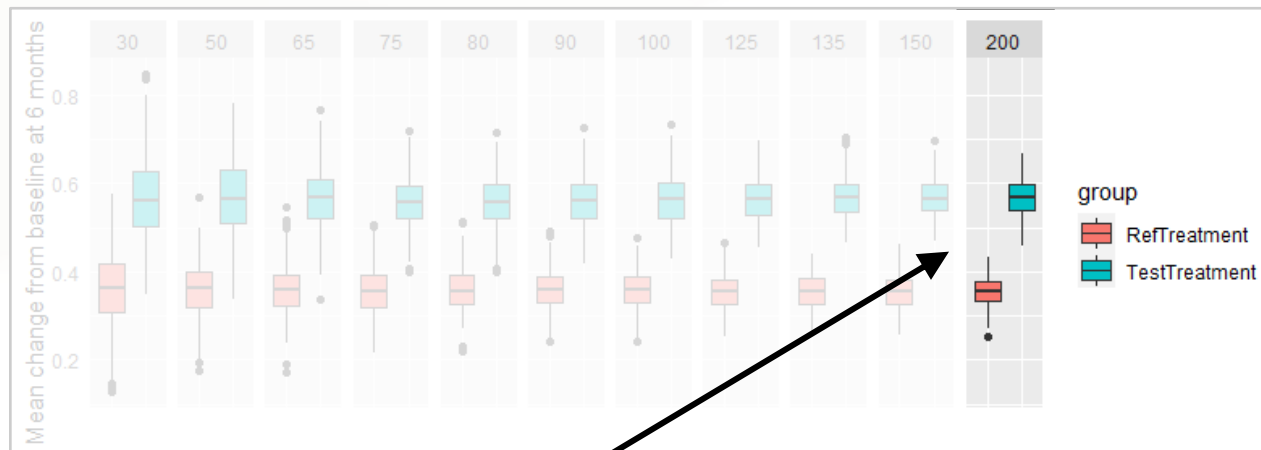
3 months



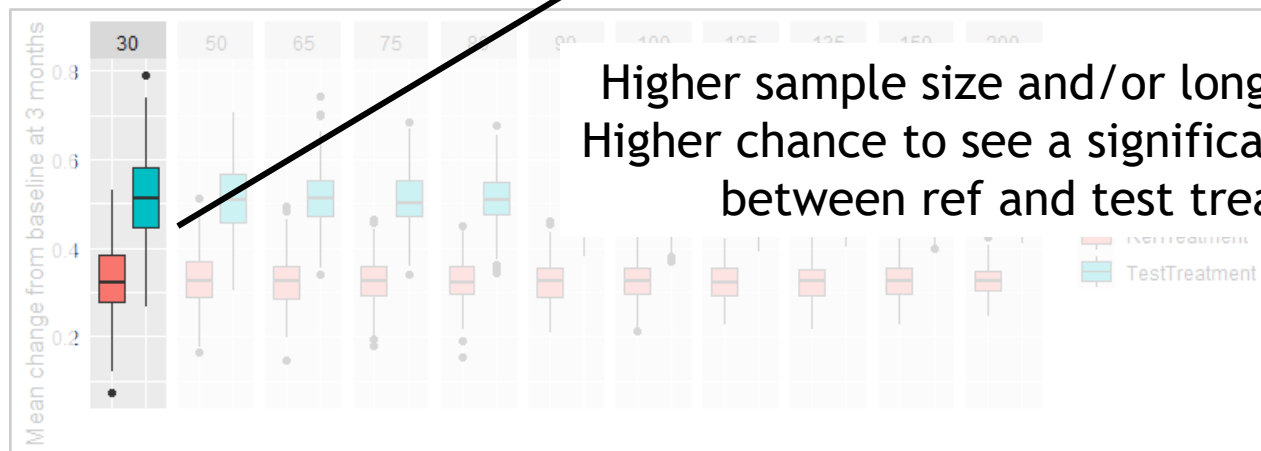
Higher sample size and/or longer duration:  
Higher chance to see a significant difference  
between ref and test treatment

# Sample size optimization with LixoftConnectors

6 months



3 months



Higher sample size and/or longer duration:  
Higher chance to see a significant difference  
between ref and test treatment

Which minimum sample size is required to reach 85% probability of success?

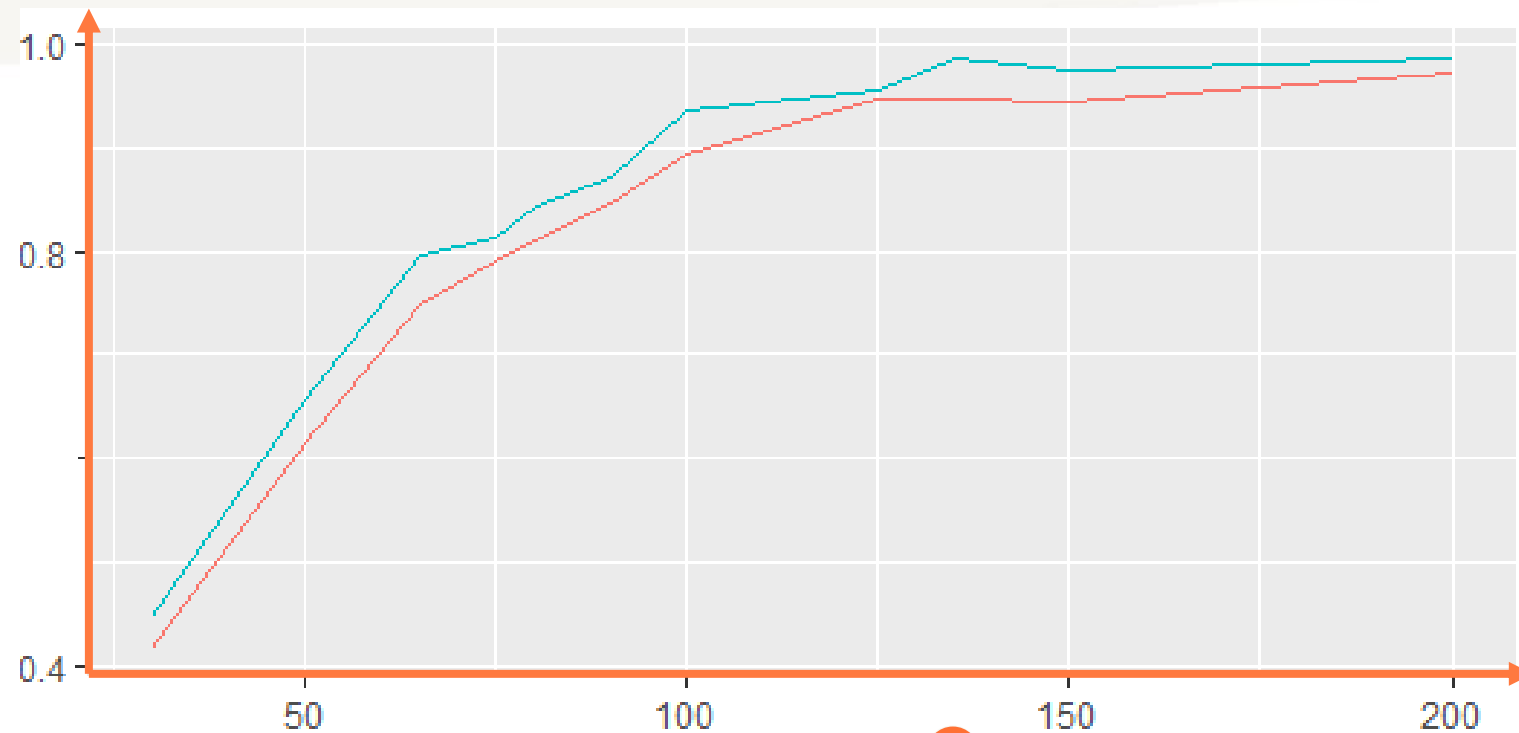


# Sample size optimization with LixoftConnectors

Endpoint: change from baseline

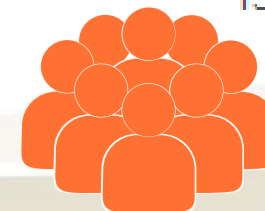
— 3 months

— 6 months

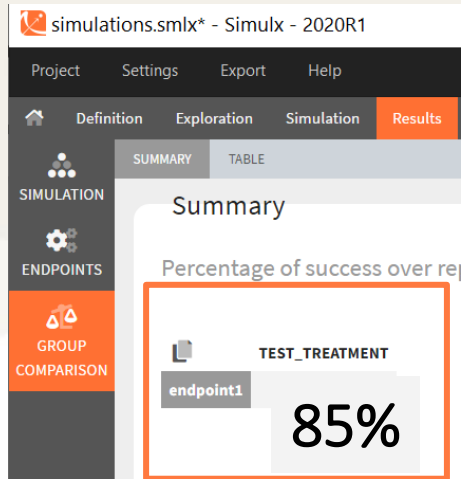


Power of Study:  
probability of  
success

Sample Size



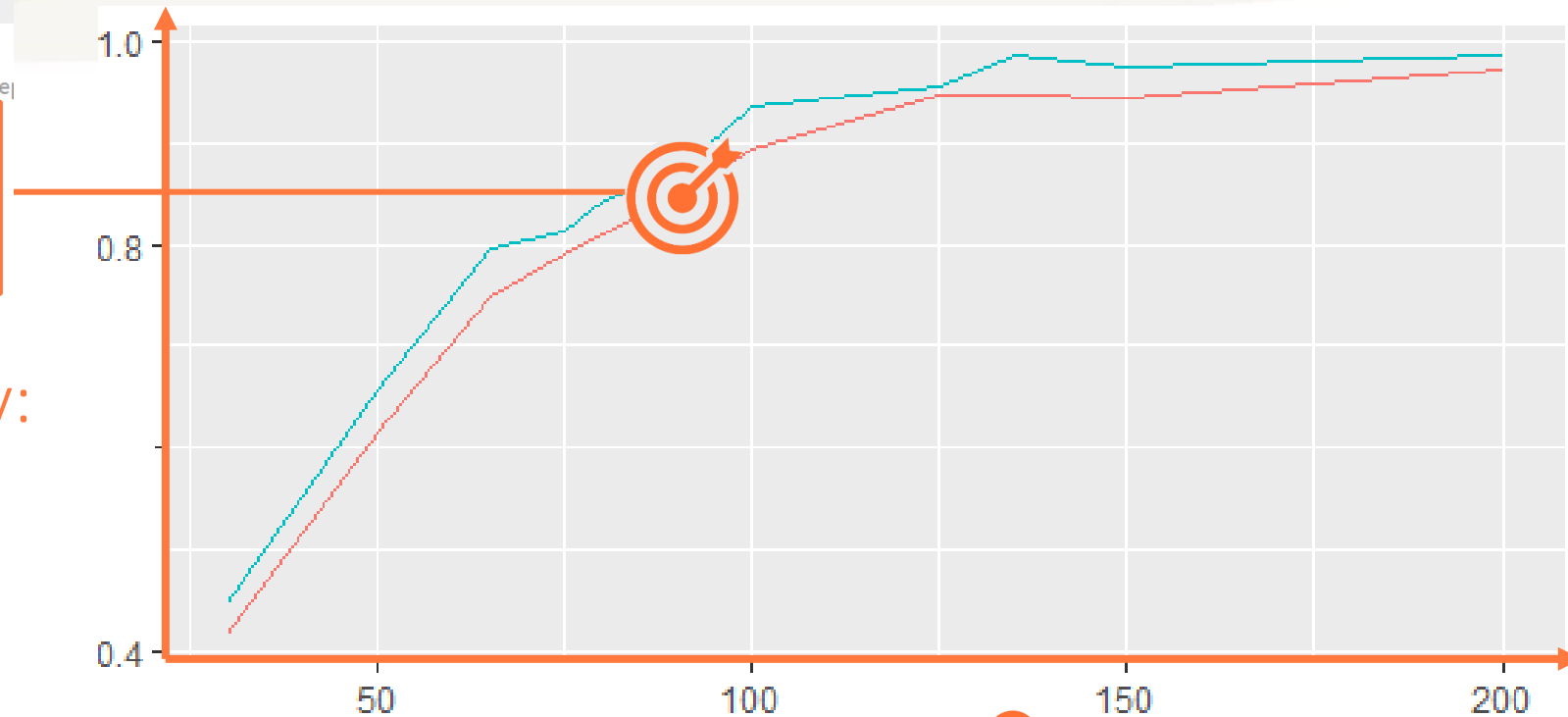
# Sample size optimization with LixoftConnectors



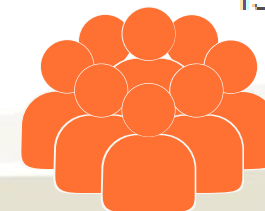
Endpoint: change from baseline

— 3 months

— 6 months

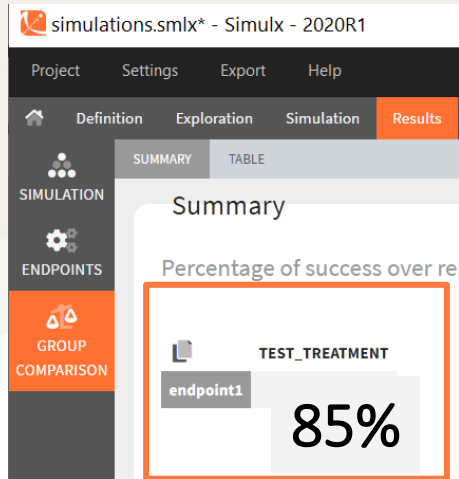


Power of Study:  
probability of  
success

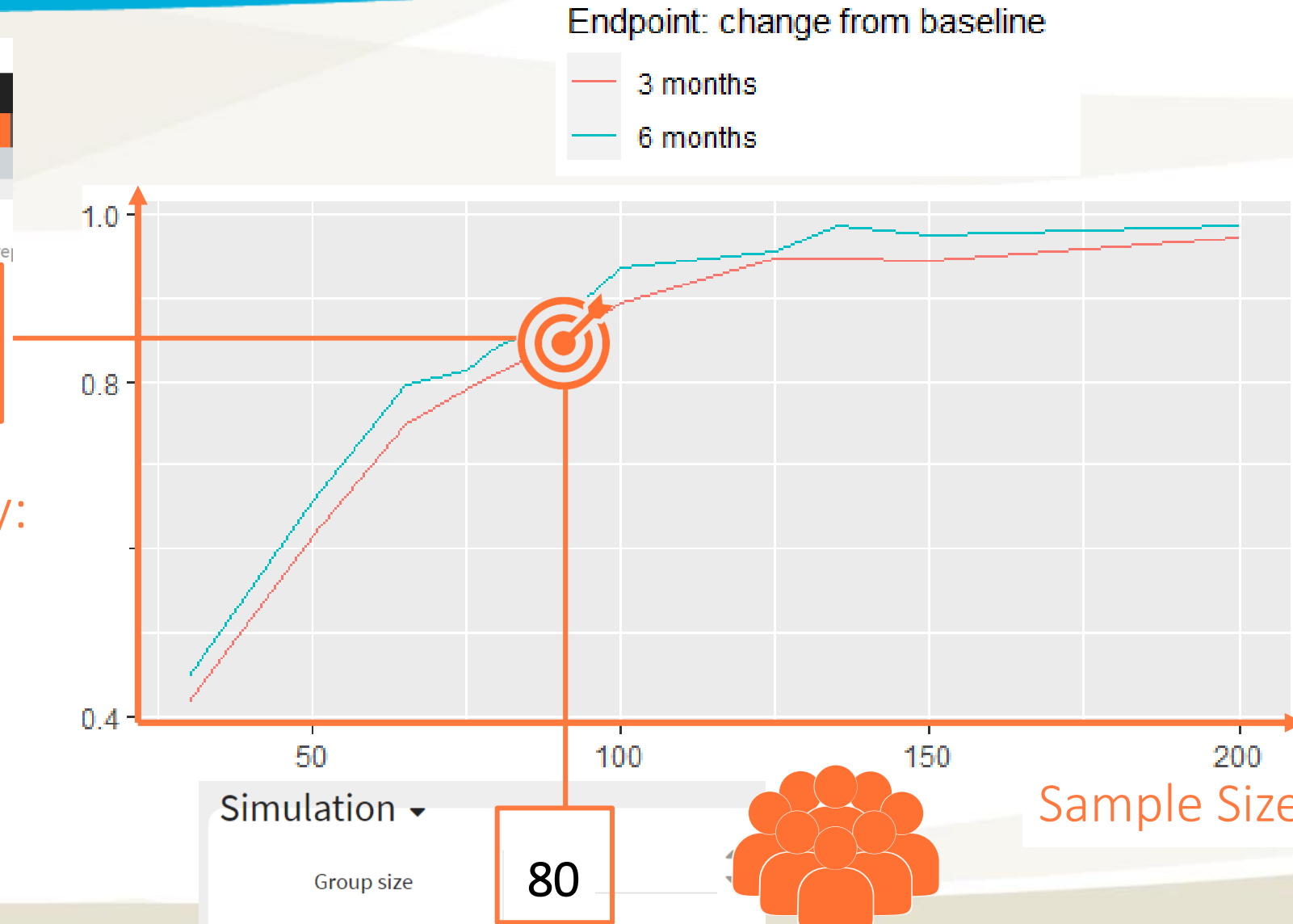


Sample Size

# Sample size optimization with LixoftConnectors



Power of Study:  
probability of  
success



# More complex simulations

## Assumptions for this clinical trial simulation:

1. Covariates can be described with simple independent distributions
2. Population parameter estimates are reliable

## Assumptions for this clinical trial simulation:

1. Covariates can be described with simple independent distributions
2. Population parameter estimates are reliable

## Using external tables for more complex simulations:

1. Sampling covariates from an existing table of covariates
2. Sampling population parameters to consider uncertainty

# Sampling from a table of covariates

Simulation ▾

☐ Single simulation  
☒ Replicates

300

☐ Same individuals among groups

Shared

Shared ids  
All | None

RefTreatment



Sampling method

↓ Keep Order

↑ Sample with replacement

↑ Sample without replacement

Group size

75

Parameters

mlx\_Pop

Treatments

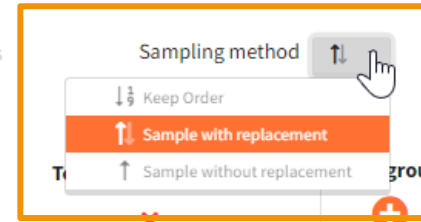
Outputs

FEV1\_3months

FEV1\_6months

Covariates

RefTrt\_Chinese\_table



Different sampling methods to choose between:

- Using all covariate values
- Covariate bootstrapping

RefTrt\_Chinese\_table [read-only]

ID	AGEBL	WTBL	HTBL	BFEVPD	FEV1BL	SEX	RACE	TMT
1	36.84	63.66	172.73	45.52	2.44	Female	Chinese	Ref
2	42.27	54.58	160.06	88.93	1.57	Female	Chinese	Ref
3	43.12	78.27	153.92	51.19	1.19	Female	Chinese	Ref
4	53.9	79.97	159.19	69.82	2.12	Female	Chinese	Ref
5	63.31	71.5	160.88	65.92	0.84	Female	Chinese	Ref
6	37.72	63.45	179.78	64.47	2.27	Female	Chinese	Ref
7	59.7	89.09	160.91	39.59	1.98	Female	Chinese	Ref
8	51.36	60.59	163.13	78.78	2.45	Female	Chinese	Ref
9	37.02	59.38	175.72	90.89	1.81	Male	Chinese	Ref

CSV




# Parameter uncertainty with replicates

Simulation ▾


☐ Single simulation

☒ Replicates 100

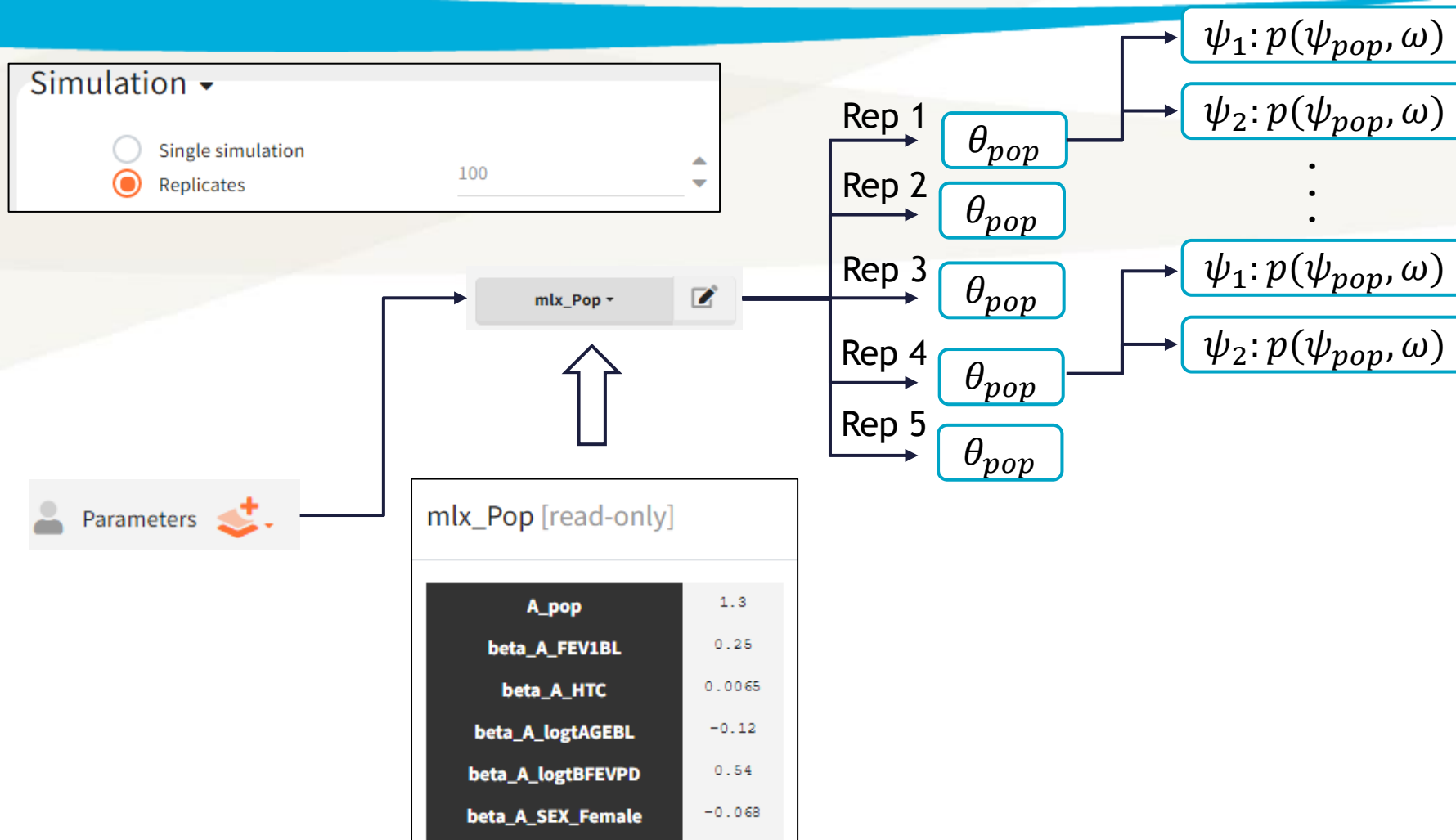
Parameters 

mlx\_Pop [read-only]

<b>A_pop</b>	1.3
<b>beta_A_FEV1BL</b>	0.25
<b>beta_A_HTC</b>	0.0065
<b>beta_A_logtAGEBL</b>	-0.12
<b>beta_A_logtBFEVPD</b>	0.54
<b>beta_A_SEX_Female</b>	-0.068

mlx\_Pop ▾ 

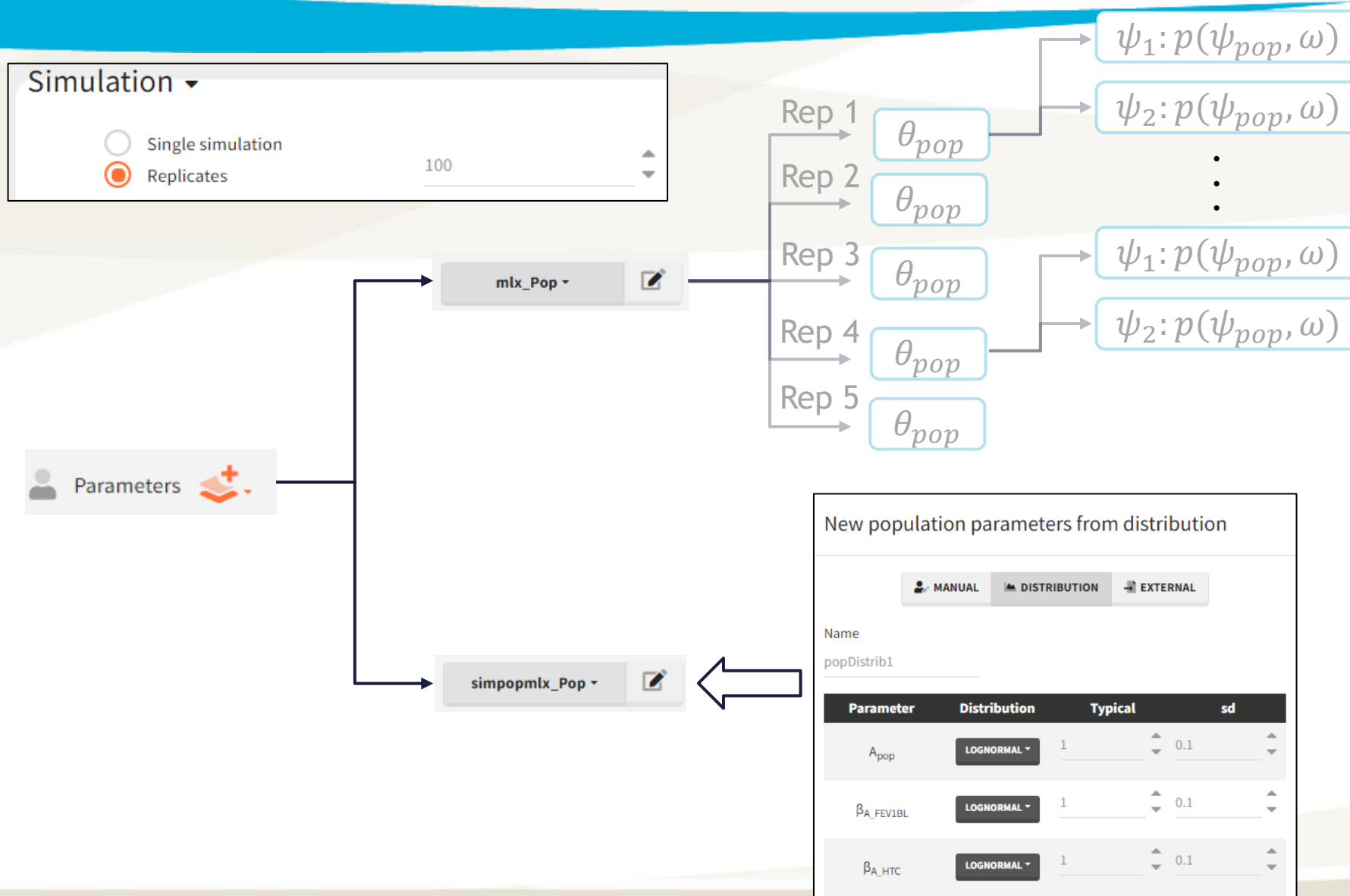
# Parameter uncertainty with replicates



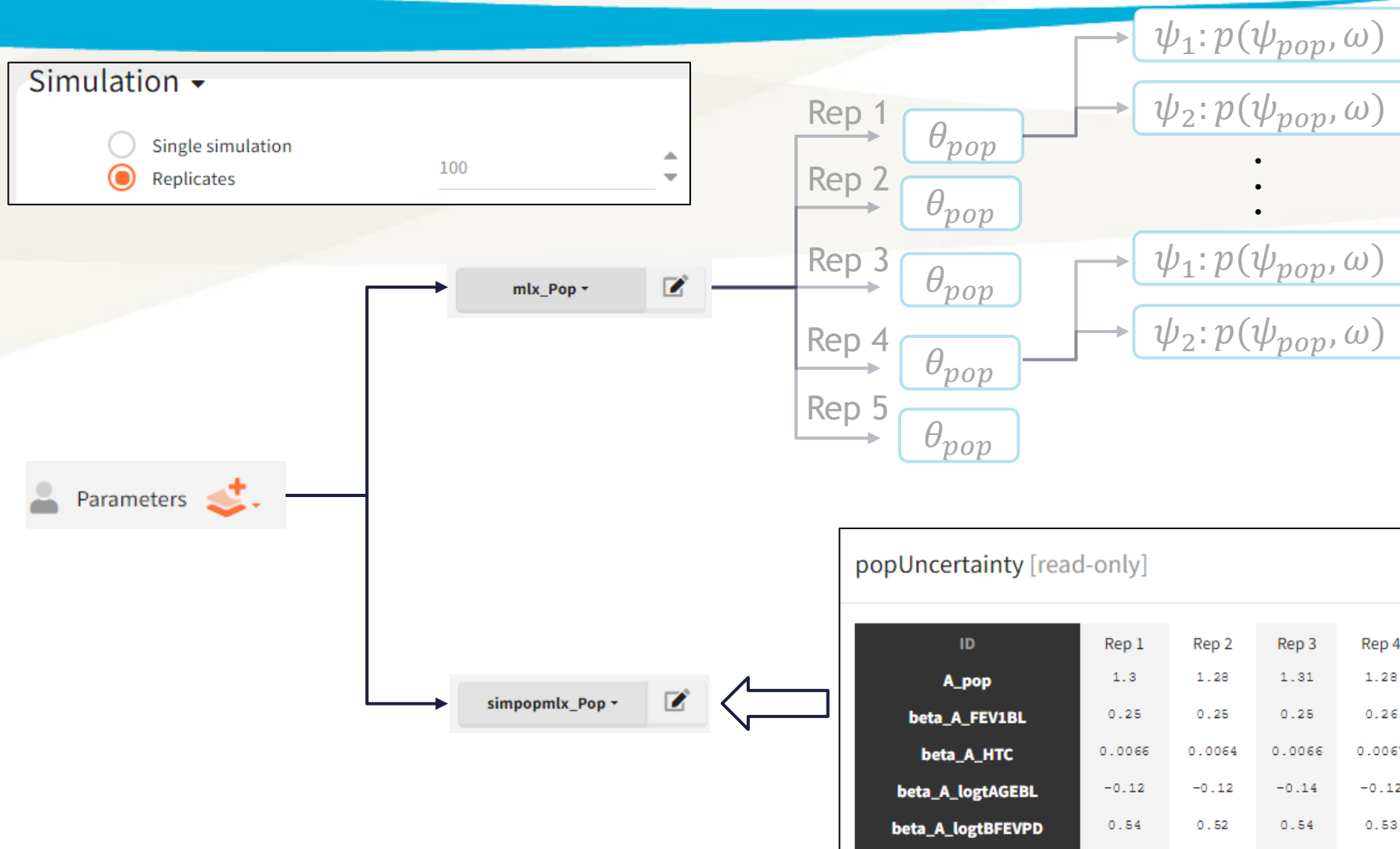
Individual parameters in **all replicates** are simulated using **THE SAME** population parameters



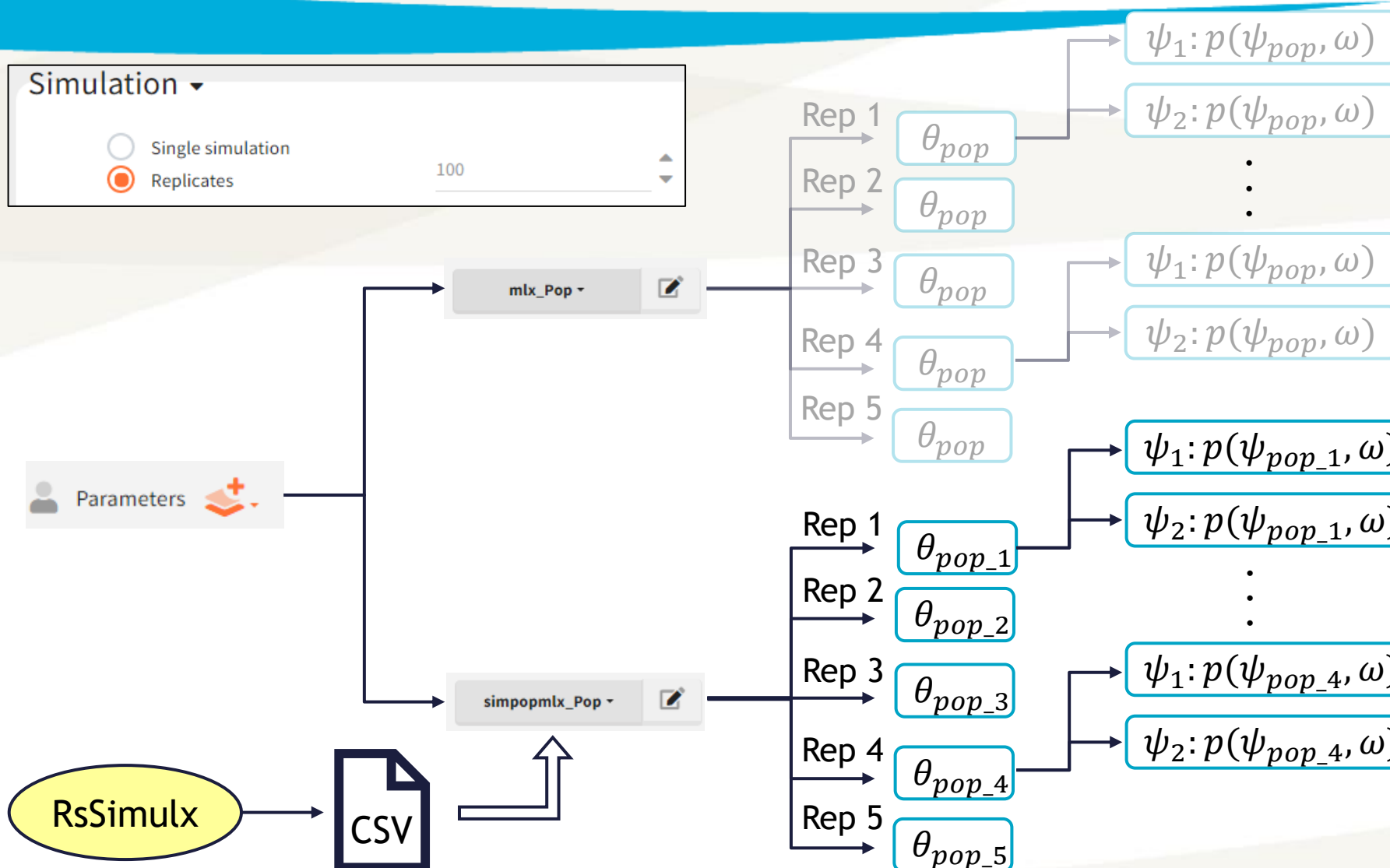
# Parameter uncertainty with replicates



# Parameter uncertainty with replicates



# Parameter uncertainty with replicates



Individual parameters in **all replicates** are simulated using **DIFFERENT** population parameters

# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



Optimize trial  
design with  
LixoftConnectors

## Result:

A sample size of **90 individuals** per treatment arm **is sufficient** to obtain a successful clinical trial showing the greater effect of the new drug in a **3-months study**, with an **85% probability**.

# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



Simulation  
of clinical trial arms (parallel, cross-  
over) with different treatments

# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



Simulation



Quantitative  
outcome  
post-processing results



# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



Simulation



Quantitative  
outcome



Trial success  
comparison of different groups  
using post-processed target



# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



Simulation



Quantitative  
outcome



Trial success



Power of study

Accounting for inter-individual variability  
and uncertainty of parameter estimates

# Summary

Already  
available  
data



Population  
modeling in  
Monolix



Clinical trial  
simulations  
in Simulx



Simulation



Quantitative  
outcome



Trial success



Power of study



Optimize trial  
design with  
LixoftConnectors

# Questions & Answers

Please type your questions in the Q&A section  
of the Zoom interface

The recordings and the slides will be available on our webpage by the end of the  
week



# Endpoints & statistical tests in Simulx

Outcome	Endpoint	Metric	Statistical test	
			Same indiv = True	Same indiv = False
Continuous	Geometric mean	Ratio of means	Paired t-test on log-transformed	Unpaired t-test on log-transformed
	Arithmetic mean	Difference of means	Paired t-test	Unpaired t-test
	Median	Difference of medians	Wilcoxon signed rank test	Wilcoxon rank sum test
Binary true/false	Percent true	Odds ratio	McNemar's exact test	Fisher's exact test
Time-to-event	Median survival	Difference in median survival	Logrank test with variance correction	Logrank test