

# Bayesian Statistics in CMC

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**Transform data into knowledge** by  
applying quantitative expertise to...

make the best  
possible decision



using optimised  
designs



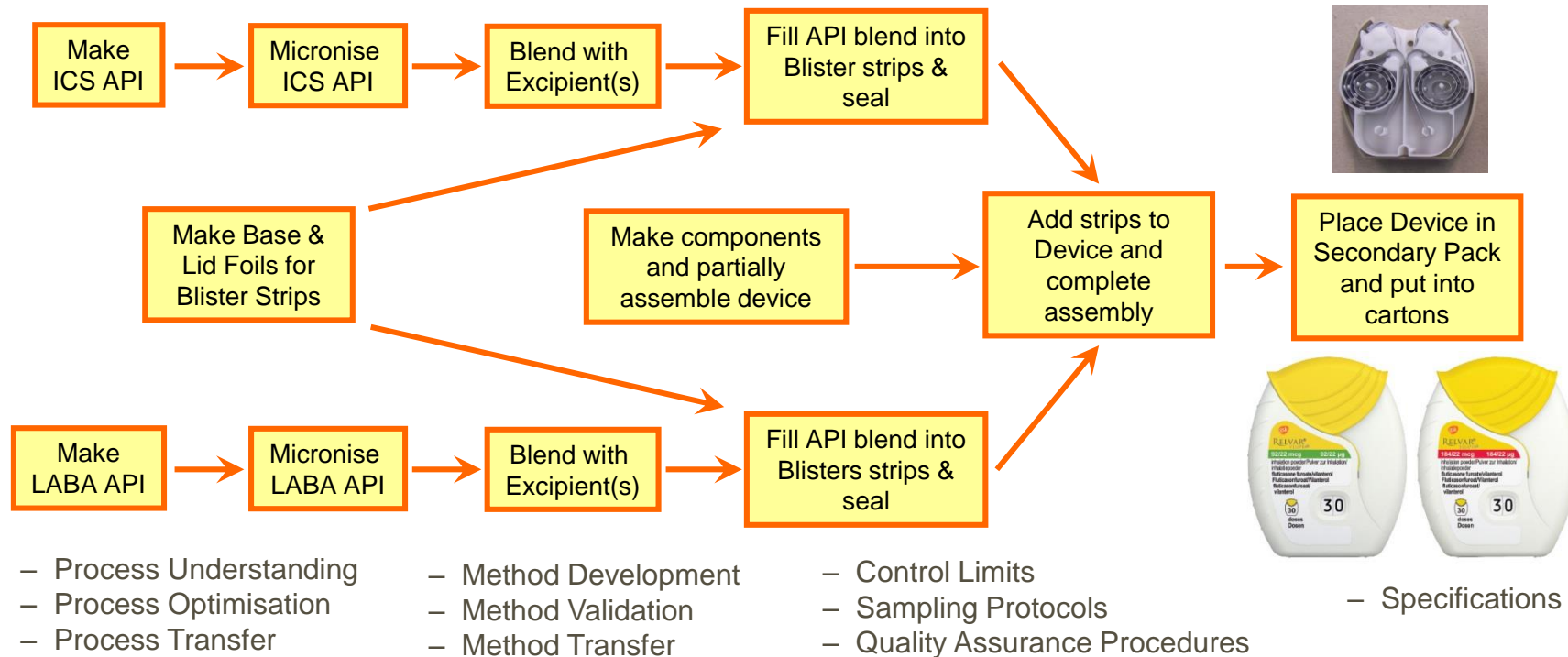
in the most  
efficient manner



# What is Chemistry Manufacturing and Controls (CMC)?



## Example of manufacture of RELVAR ELLIPTA<sup>1</sup>



## 26 Statisticians globally

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- Compounds across multiple therapeutic areas e.g. Respiratory, Infectious Diseases, Oncology
- Both small molecules and biopharmaceuticals
- Statistical Activities:
  - Consultancy, enablement and training for R&D scientists and engineers
  - Experimental Design
    - Factorial/Fractional Factorial, Response Surface, Screening, bespoke...
  - Data Modelling
    - Linear and Linear Mixed models (repeated measures and random effects)



# Bayesian Analytical Biosimilarity

Supporting heading



## With acknowledgements to:



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EFSPI Working Group on Statistical methodology for comparative assessment of quality attributes in drug development:

Christophe Agut (Sanofi)  
Armin Boehrer (Boehringer-Ingelheim)  
Bruno Boulanger (Pharmalex)  
Piet Hoogkamer (Abbott)  
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Martina Kron (Abbvie)  
Beate Krueger (Boehringer-Ingelheim)  
Jens Lamerz (Roche)  
Timothy Mutsvari (Pharmalex)  
Christian Seifert (Boehringer-Ingelheim)

# Three Fundamental Requirements

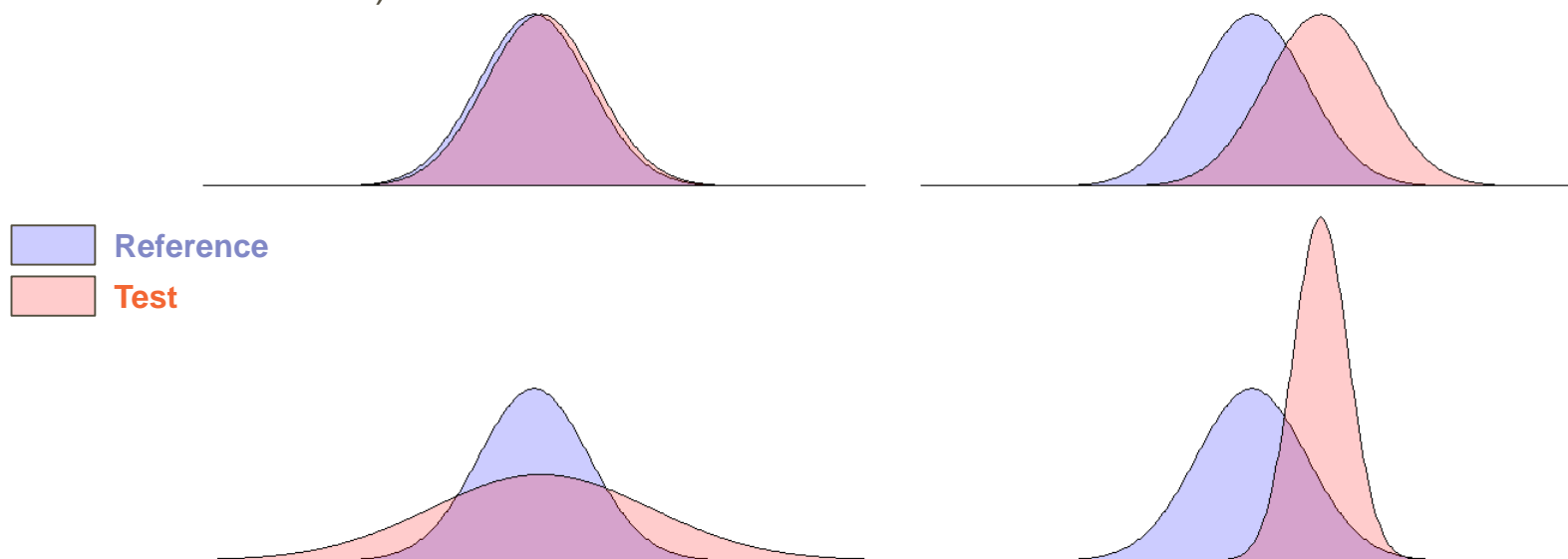


1. Define what we mean by equivalence/comparability
2. Provide a well-defined decision procedure
3. Demonstrate the operating characteristics of the procedure
  - What is the probability of deciding in favour of equivalence/comparability?
  - What is the patient risk?
    - (Test product is deemed equivalent/comparable and a patient receives a bad lot from the Test product)
  - What is the producer risk?
    - (Test product is deemed not to be equivalent/comparable when it is)

# What do we mean by equivalent/comparable?



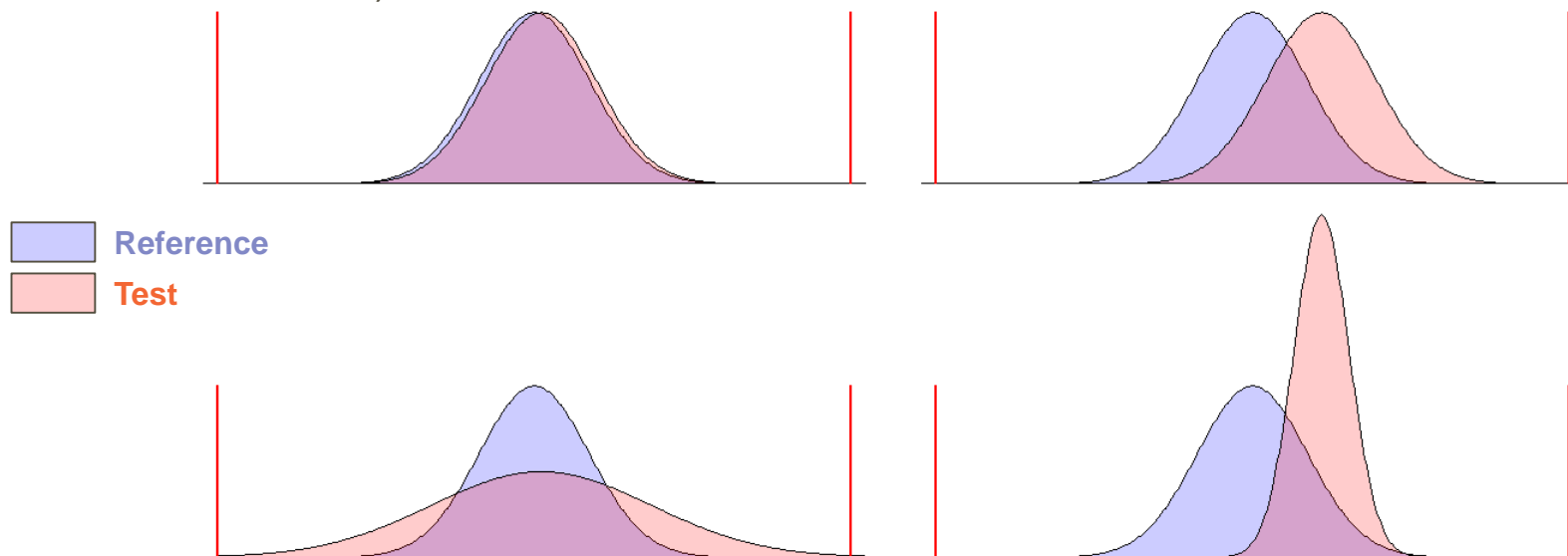
- We wish to demonstrate that a **proposed new process produces** lots of **Test** product that are (analytically) “equivalent/comparable” to those of a **Reference** product (both now and in the future).



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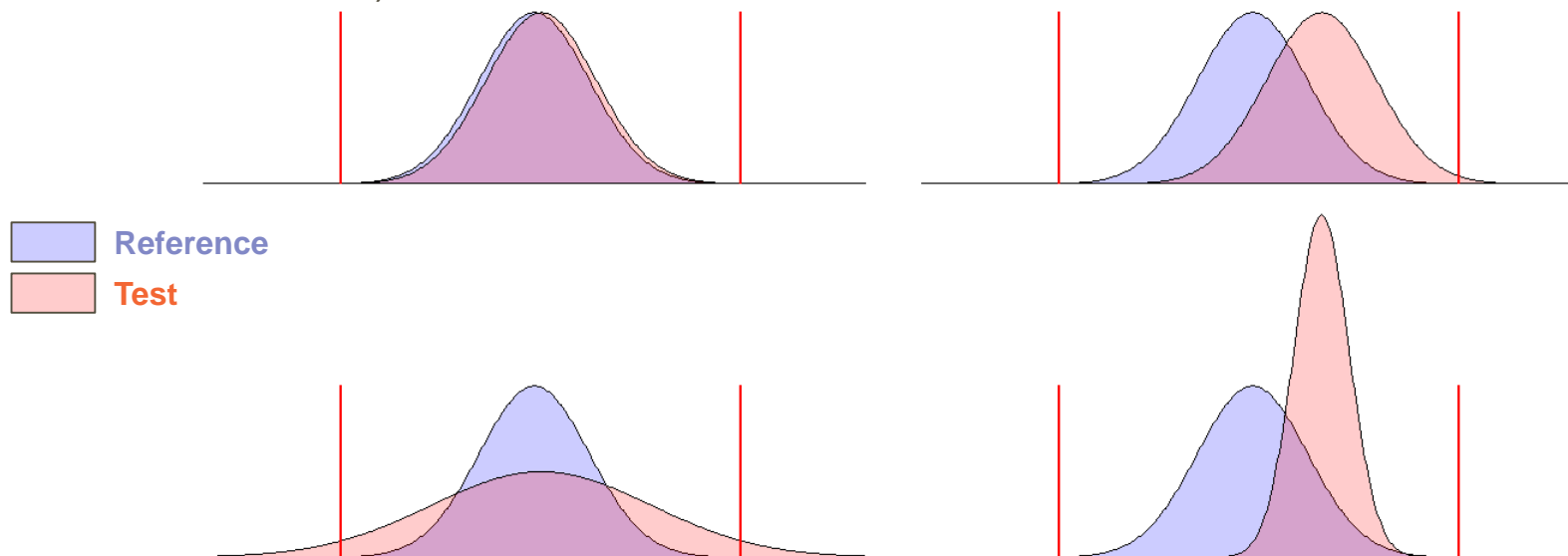
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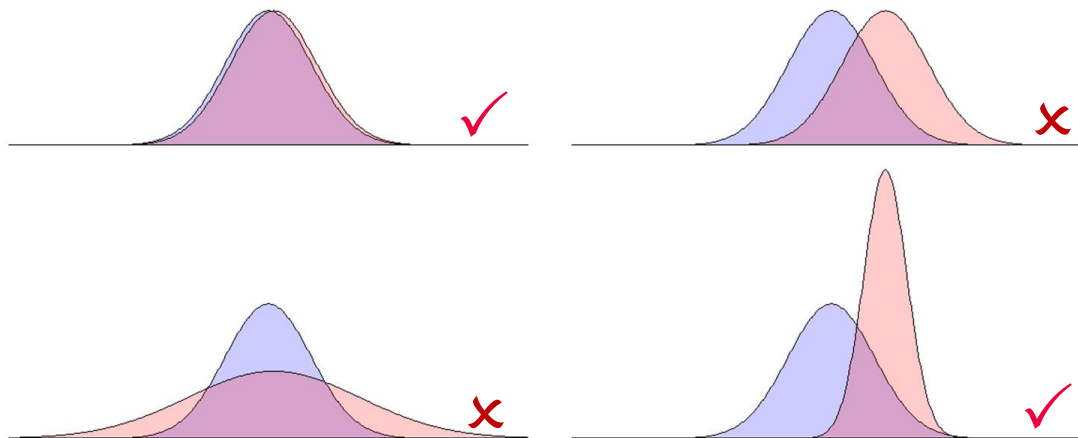


# A Definition of (Analytical) Biosimilarity

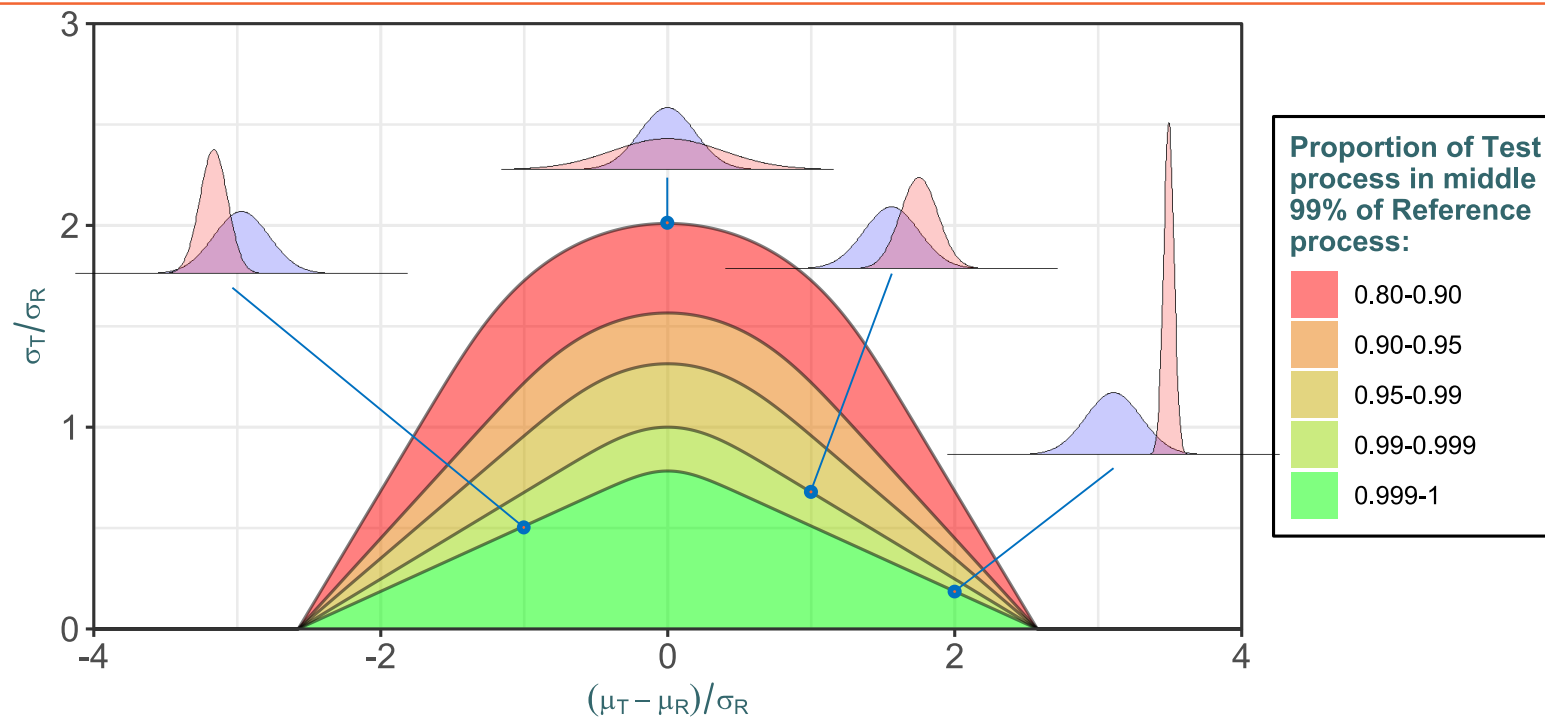


- The test product is analytically comparable (for a given attribute) to the **Reference** product if at least  $P_T\%$  of all lots produced by the **Test** product process lie within the middle  $P_R\%$  of the lots produced by the Reference product process.

In the absence  
of  
specification  
limits



# A Definition of (Analytical) Biosimilarity



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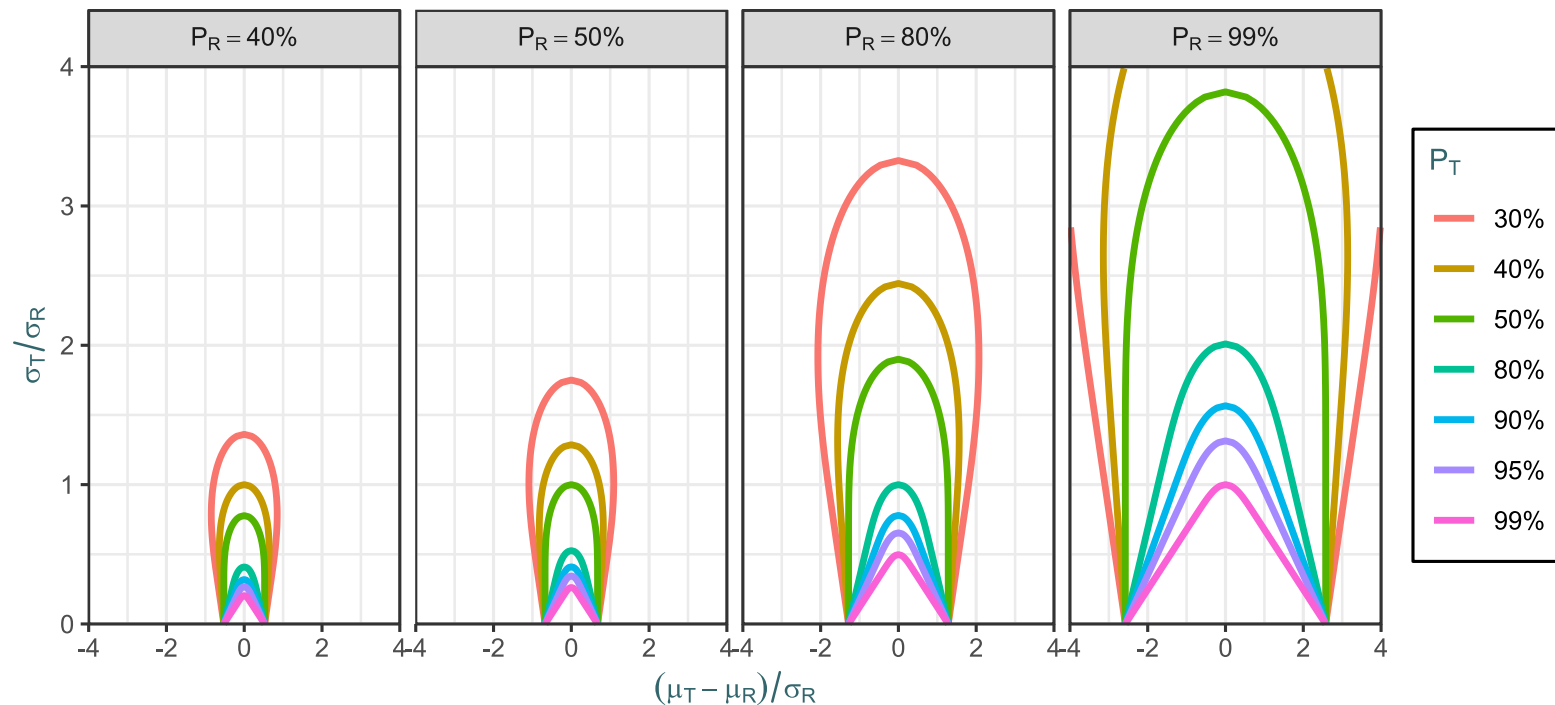
- For normally distributed processes  $N(\mu_T, \sigma_T^2)$ ,  $N(\mu_R, \sigma_R^2)$  comparability region only depends on difference in means and standard deviation relative to the Reference standard deviation:

$$\mu_{\Delta} = \frac{\mu_T - \mu_R}{\sigma_R}; \rho = \frac{\sigma_T}{\sigma_R}$$

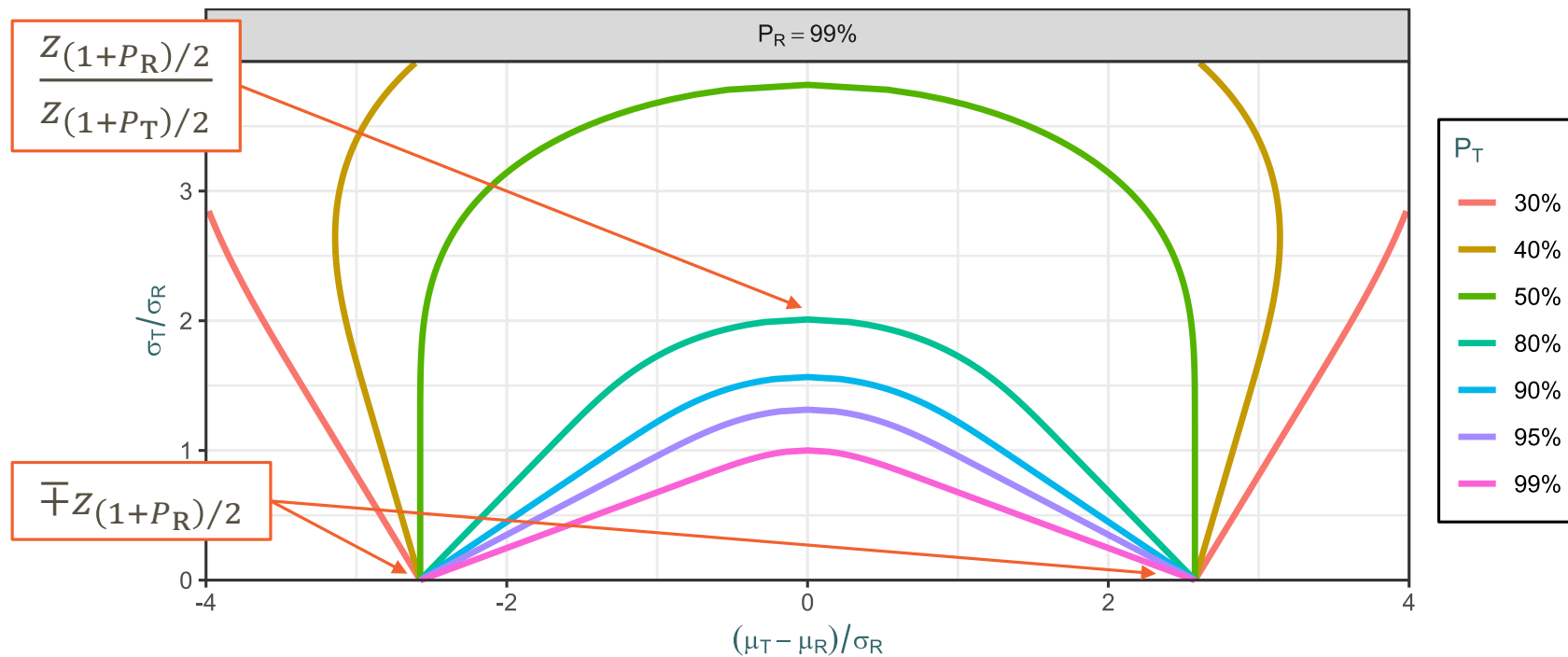
$$\left\{ (\mu_{\Delta}, \rho) : \Phi \left[ \frac{z(1+P_R)/2 - \mu_{\Delta}}{\rho} \right] - \Phi \left[ \frac{-z(1+P_R)/2 - \mu_{\Delta}}{\rho} \right] \geq P_T \right\}$$



# Impact of choice of $P_T$ and $P_R$ on Comparability Region



# Comparability regions for $P_R = 99\%$



- In principle, calculating the posterior probability of Biosimilarity is easy:

- 1) Define an indicator function for Biosimilarity in terms of the unknown parameters:

$$B(\mu_{\Delta}, \rho) = \begin{cases} 1 & \Phi\left(\frac{z_{(1+P_R)/2} - \mu_{\Delta}}{\rho}\right) - \Phi\left(\frac{-z_{(1+P_R)/2} - \mu_{\Delta}}{\rho}\right) \geq P_T \\ 0 & \text{otherwise} \end{cases}$$

- 2) “Integrate” this over the unknown parameters weighted by the joint posterior density:

$$P(\text{Biosimilarity}|\text{Data}) = \int_{\sigma_R, \sigma_T, \mu_T, \mu_R} B\left(\frac{\mu_T - \mu_R}{\sigma_R}, \frac{\sigma_T}{\sigma_R}\right) P(\mu_T, \mu_R, \sigma_T, \sigma_R|\text{Data})$$

# Posterior Probability of Biosimilarity



(Independent Samples from Test and Reference Processes)

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Using Jeffreys' Prior:  $P(\mu_T, \sigma_T, \mu_R, \sigma_R) = \sigma_T^{-1} \times \sigma_R^{-1}$

Posterior:

$$P(\mu_T, \mu_R, \sigma_T, \sigma_R | \text{Data}) = P(\mu_T, \sigma_T | \bar{X}_T, s_T, N_T) \times P(\mu_R, \sigma_R | \bar{X}_R, s_R, N_R)$$

$$(\mu_T | \sigma_T) \sim N(\bar{X}_T, \sigma_T^2 / N_T); \quad (\sigma_T^2) \sim \Gamma^{-1}((N_T - 1)/2, (N_T - 1)s_T^2/2)$$

$$(\mu_R | \sigma_R) \sim N(\bar{X}_R, \sigma_R^2 / N_R); \quad (\sigma_R^2) \sim \Gamma^{-1}((N_R - 1)/2, (N_R - 1)s_R^2/2)$$

- With a bit of effort it's possible to show that

$$P(\text{Biosimilarity}|\text{Data}) = \int_{\sigma_R, \sigma_T, \mu_T, \mu_R} B\left(\frac{\mu_T - \mu_R}{\sigma_R}, \frac{\sigma_T}{\sigma_R}\right) P(\mu_T, \mu_R, \sigma_T, \sigma_R|\text{Data})$$

can be re-written as a double integral over two  $\Gamma^{-1}$  random variates

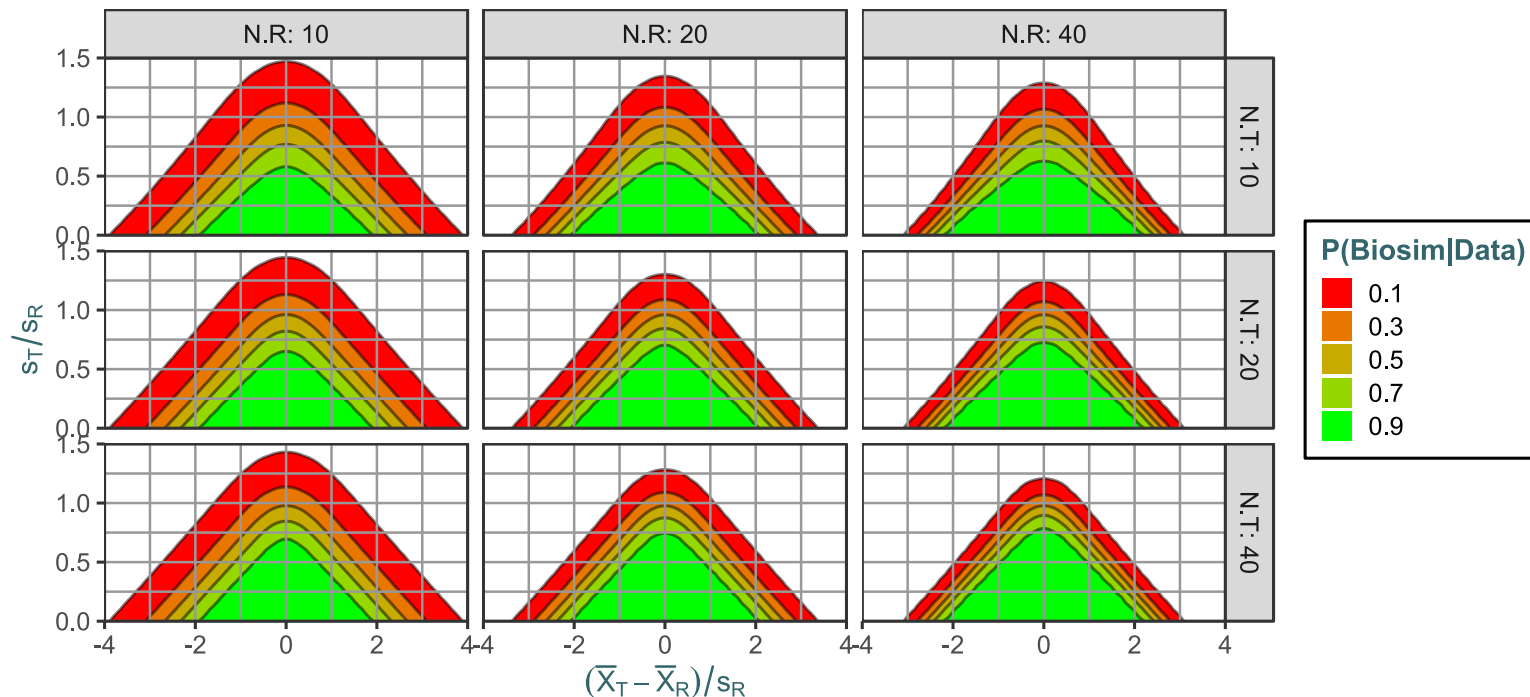
- Value only depends on:

$$(\bar{X}_T - \bar{X}_R)/s_R; \quad s_T/s_R; \quad N_T; \quad N_R.$$

# Posterior Probability of Biosimilarity



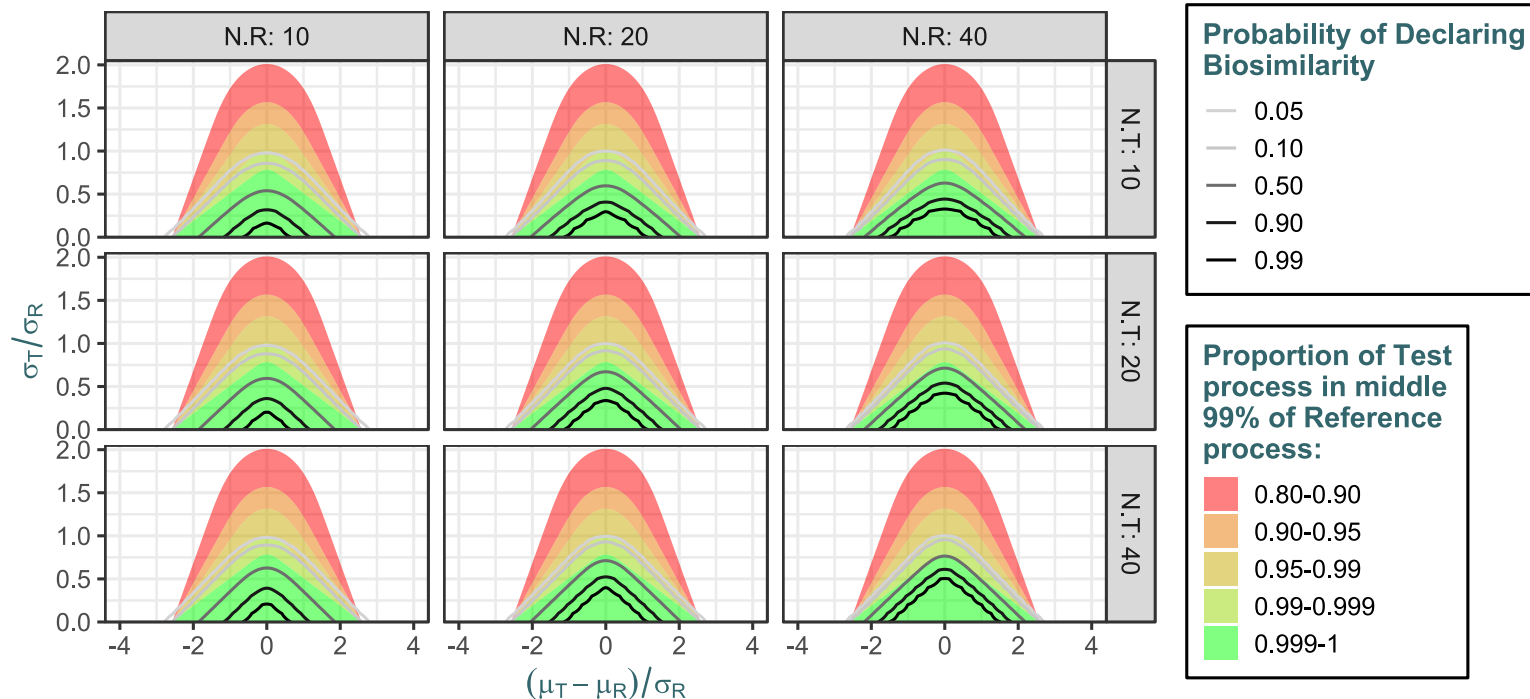
Biosimilarity based on  $P_T = P_R = 99\%$



# Operating Characteristic based on $P(\text{Biosim}|\text{Data}) > 0.9$



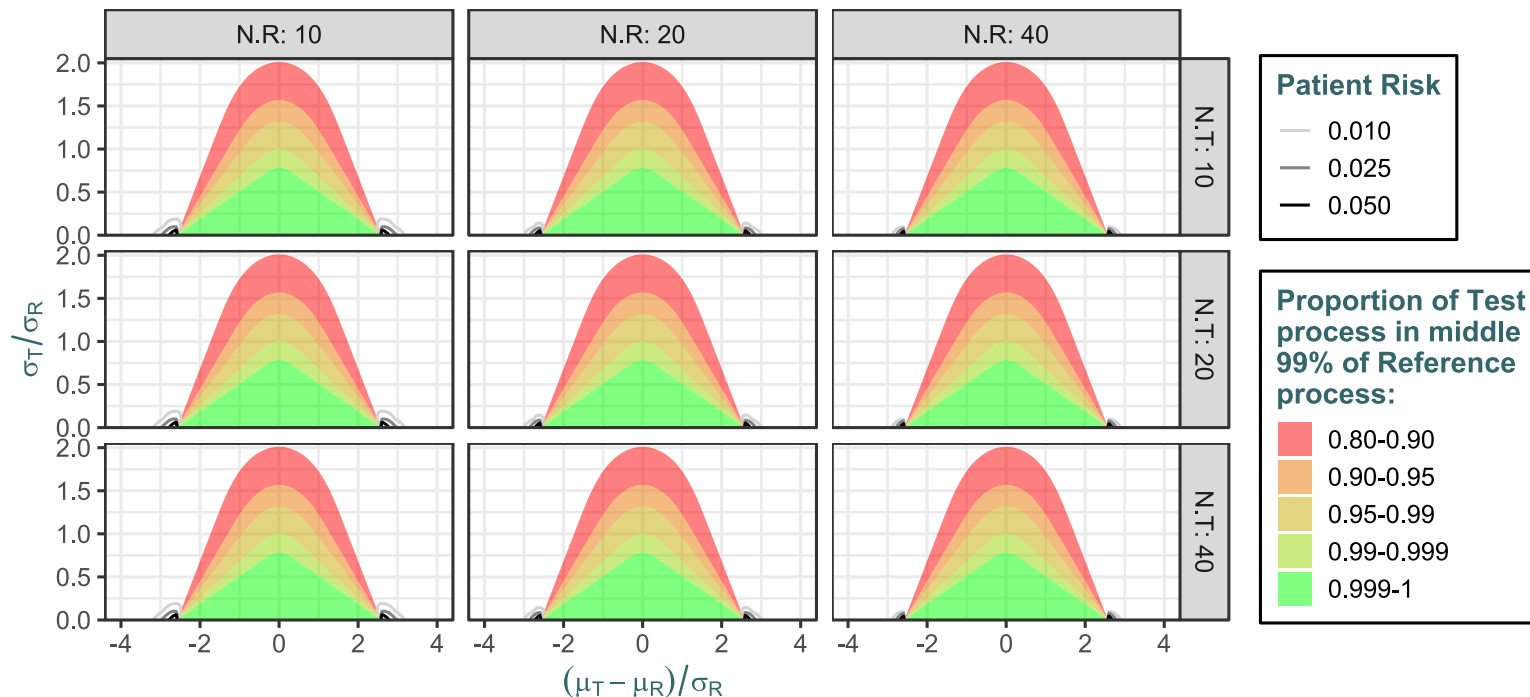
Biosimilarity based on  $P_T = P_R = 99\%$



# Patient Risk<sup>1</sup>



Biosimilarity based on  $P_T = P_R = 99\%$  and  $P(\text{Biosim}|\text{Data}) > 0.90$

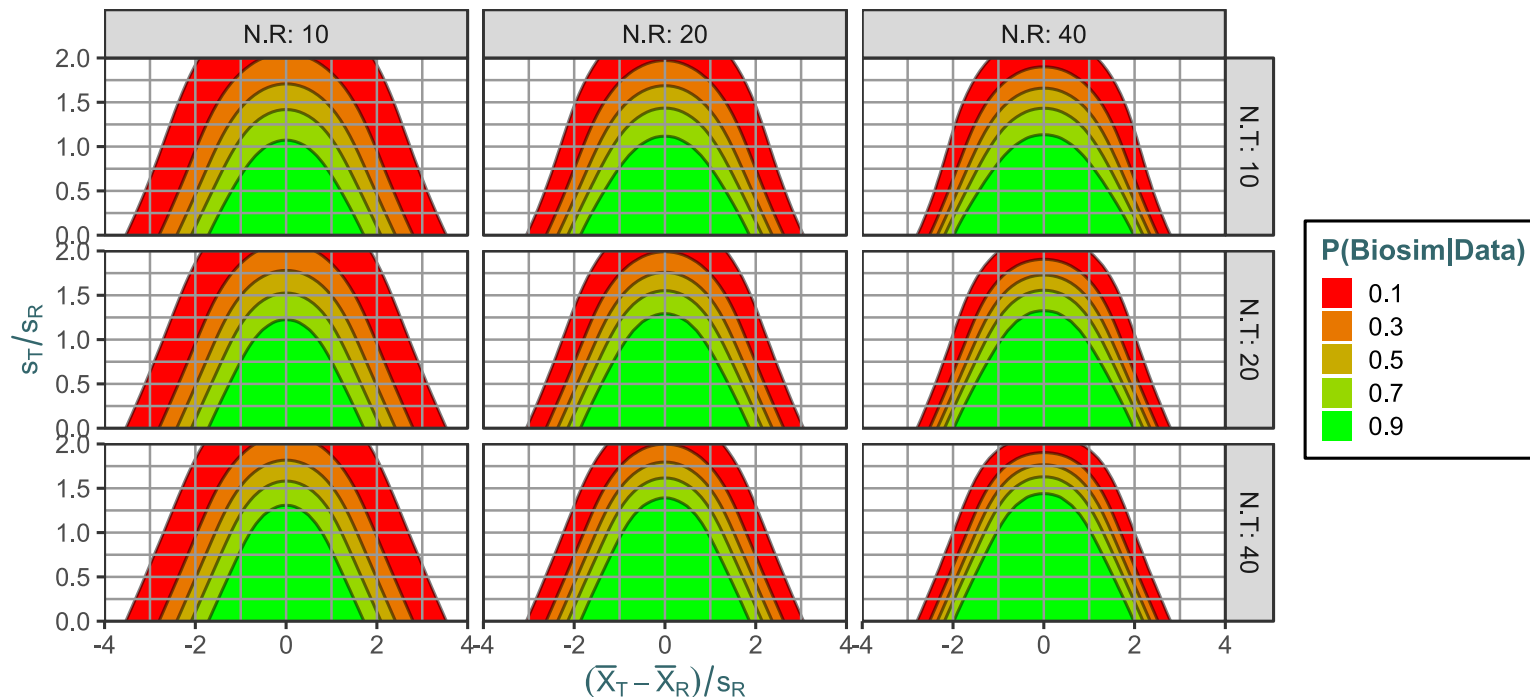




# Posterior Probability of Biosimilarity



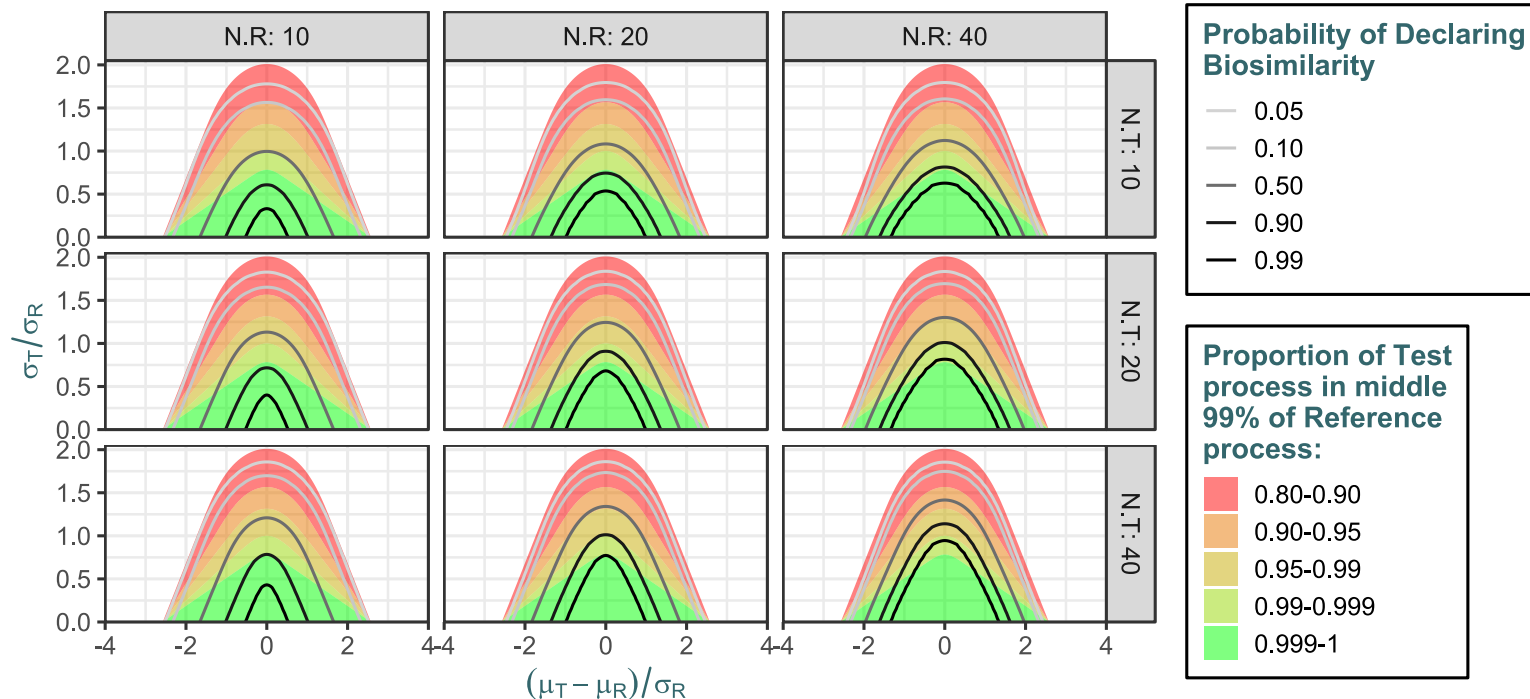
Biosimilarity based on  $P_T = 80\%$   $P_R = 98\%$



# Operating Characteristic based on $P(\text{Biosim}|\text{Data}) > 0.9$



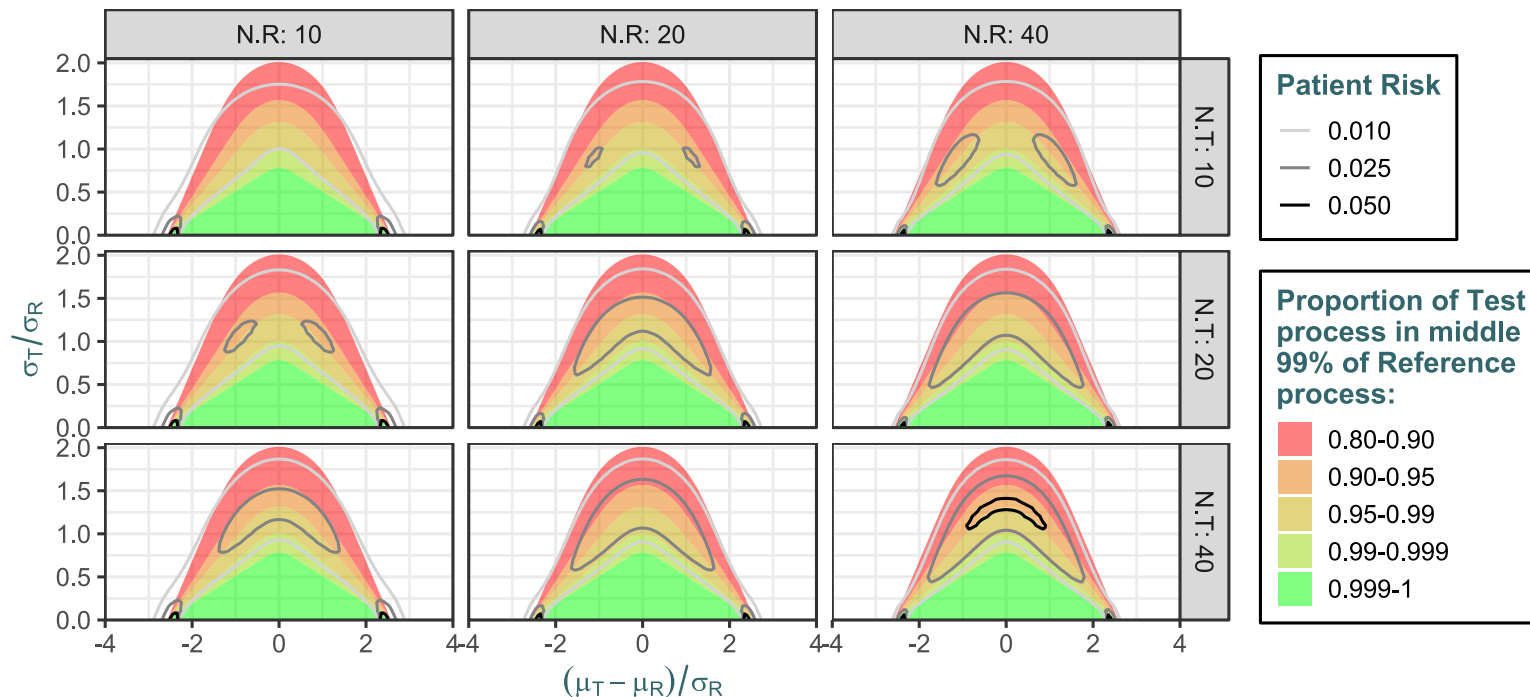
Biosimilarity based on  $P_T = 80\%$   $P_R = 98\%$



# Patient Risk<sup>1</sup>



Biosimilarity based on  $P_T = 80\%$   $P_R = 98\%$  and  $P(\text{Biosim}|\text{Data}) > 0.90$



# Patient Risk<sup>1</sup>



Biosimilarity based on  $P_T = 80\%$   $P_R = 98\%$  and  $P(\text{Biosim}|\text{Data}) > 0.90$

