

# From Power and Assurance to Bayesian Power: Application to probability of success

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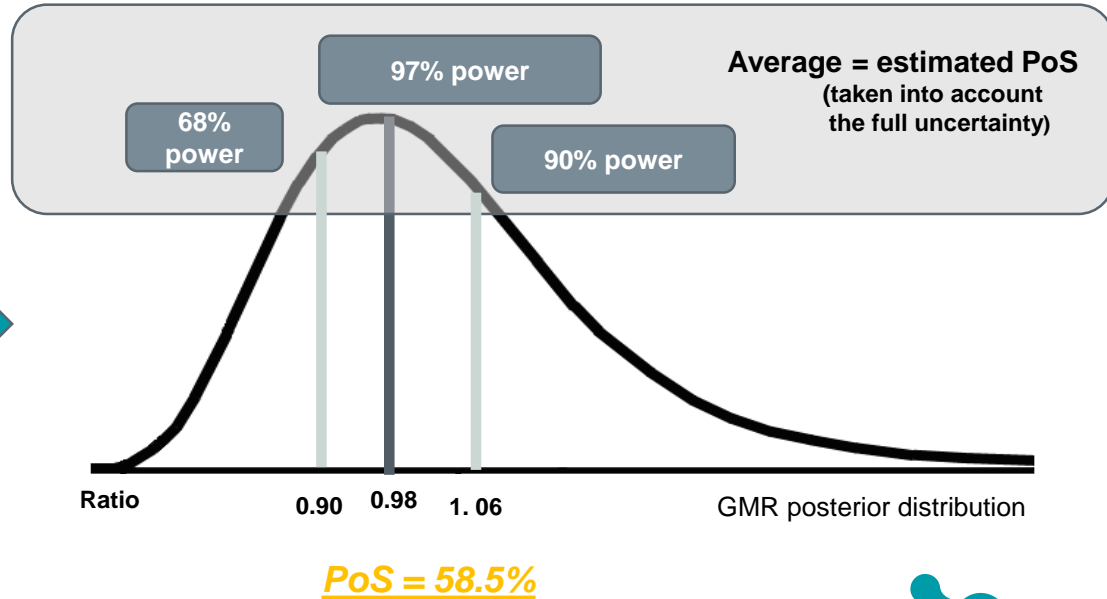
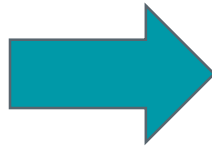
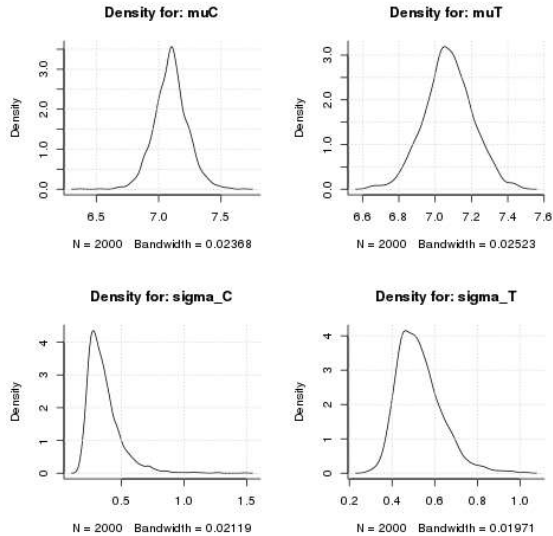
# Introduction/ Context

- ▶ **Bayesian network in France and Belgium (ARM of BAYES congress people!)**
  
- Study case presented:
  - In clinical early phase, estimation of the probability of success (PoS) of a second study taken into account the data of a first study:
    - Previous : Pilot study observed 24 patients (16 G1 and 8 G2))
    - Future: Similarity study with N=188 (94 patients per group G1 and G2)
    - Success rule criteria (frequentist) : 90% CI of the geometric mean ratio (GMR) is within [0.80; 1.25]
  
  - What is the PoS (probability of the success) for the future study based on the previous study data?



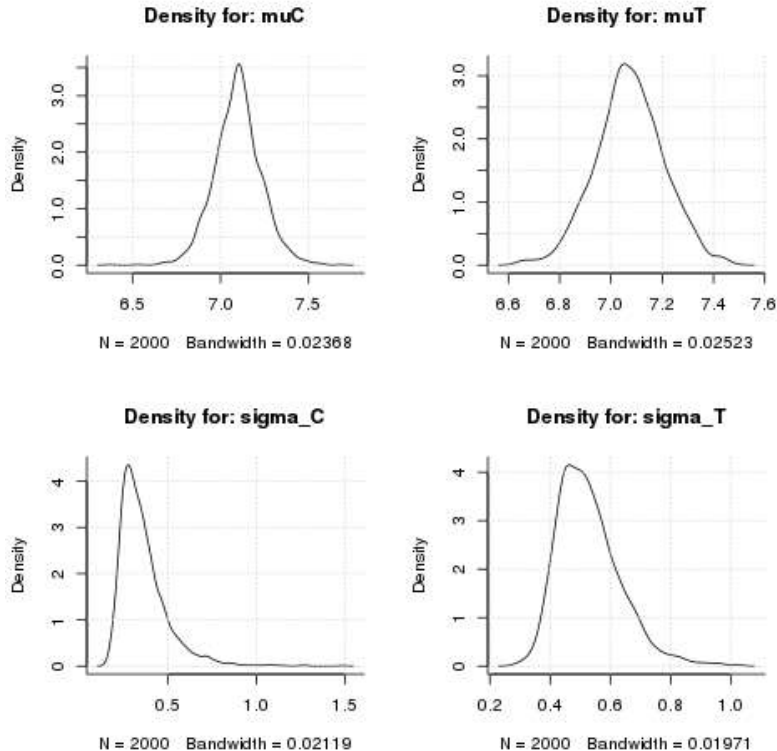
# Introduction/ Context Assurance

- ▶ Easy to obtain the posterior distribution using a classical Bayesian analysis with non informative prior on the first pilot study data



# Introduction/ Context

## Bayesian Power



- From each sample of the posterior densities:
  - Simulate a study
  - Assess whether it is a success yes or no
  - Repeat and take the mean overall all successes.

- $PoS = 50\%$

## Different way to estimate the POS

- ▶ Explanation of difference?
- ▶ How to do concretely?
- ▶ What are the recommendations?



## ● Power and Assurance: What is What ?

## (Frequentist) Power

- ▶ Let R denote the rejection of the null hypothesis, the power is , assuming parameter values of  $\theta = \theta^*$

$$\pi(\theta^*, n) := \Pr(R|\theta^*, n)$$

- ▶ It is a **conditional** probability. It is conditional on the parameters of the model, e.g. the “true effect size” in a frequentist test and the sample size.

## Assurance

- ▶ “Assurance is the **unconditional** probability that a trial will lead to a specific outcome”

$$\begin{aligned}\gamma(n) &:= \int \pi(\theta, n)f(\theta)d\theta \\ \gamma(n) &:= \Pr(R) = E_{\theta}[\pi(\theta)]\end{aligned}$$

It is thus also a function of n (and eventually other nuisance parameters)

The assurance is the expected power over all possible values of theta (-> **over its prior distribution...**)



## Assurance



- ▶ Assurance does not necessarily converge to one when  $n$  increases

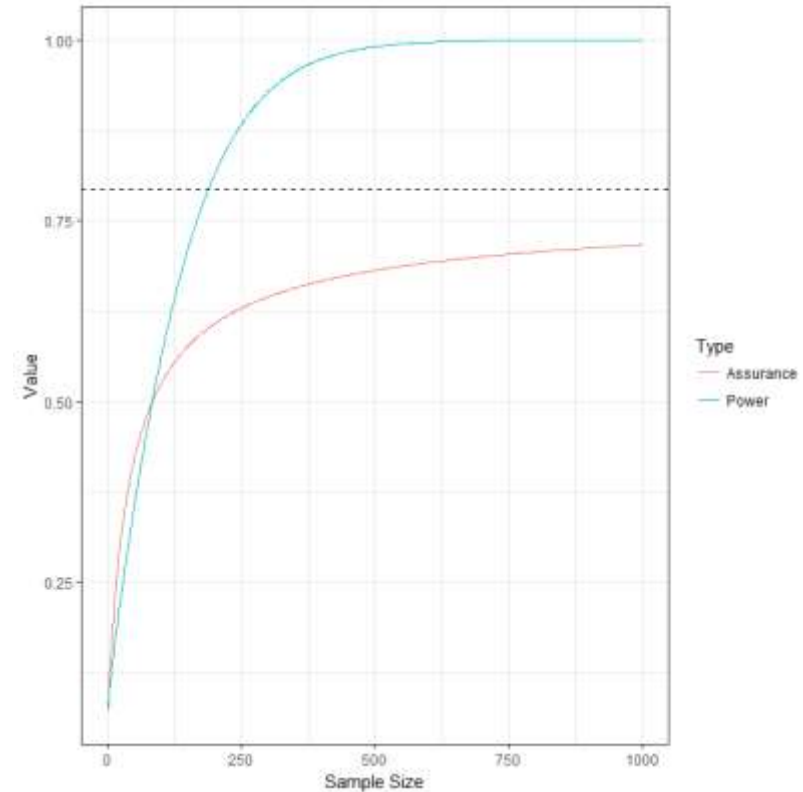
- ▶ Intuitively this is due to the fact that there can exist non-zero probability that we do not reject the null hypothesis.
- ▶ More precisely if the parameter space associated with “not  $H_1$ ” is a non-negligible set then assurance will not converge to one.





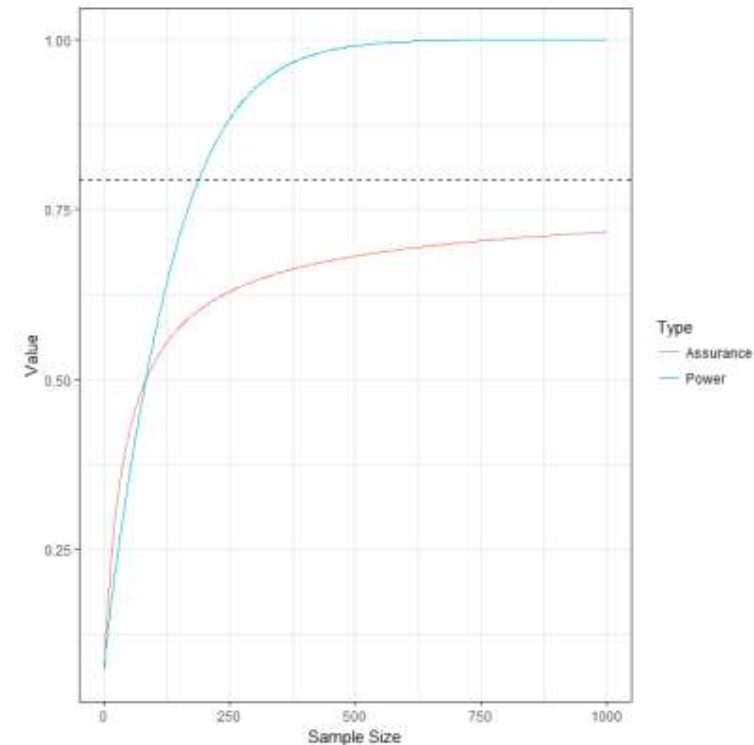
## An Example: A simple T- test

The assurance converges to 0.793, that is the prior probability that the new drug is indeed superior. (assumed in this example)



## Link to Probabilities of Success

- ▶ This can also be expressed in terms of probability of success quite easily
- ▶ Imagine that in the previous example rejecting the null hypothesis is considered as a success. This implies:
  - Frequentist Power will always allow to have a perfect probability of success if the “true effect” is in the rejection region
  - Assurance will never give a higher probability of success than that which is assumed in the prior belief
  - **!/\** At this stage assurance is still a frequentist concept ! **!/\**



# Let's Make things Bayesian

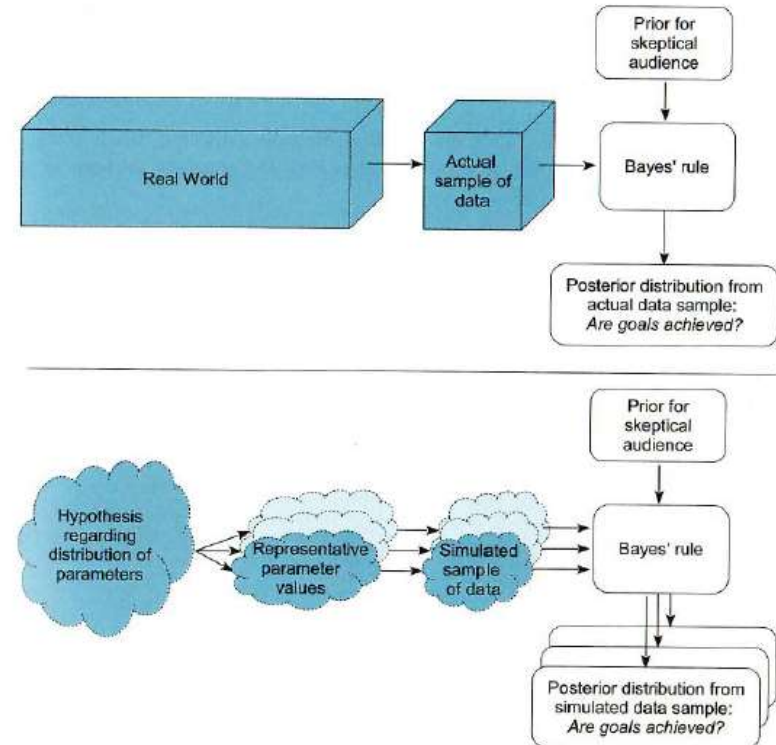
## ● Going Further to Bayesian Methods

- ▶ Assurance is a relatively old concept
  - It predates the golden age of Bayesianism that we now live in
- ▶ Power calculation in the Bayesian methodology (Chapter 13: Goals, Power, and sample size p359-398 in Doing Bayesian Data Analysis A tutorial with R,JAGS and Stan, John K.Kruschke, Edition 2, 2015 ,Elsevier)
  - Differentiates power in the NHST world (Null Hypothesis Significance Testing) which is the rejection of the null from Bayesian power which can accommodate other goals and sampling plans and hypotheses

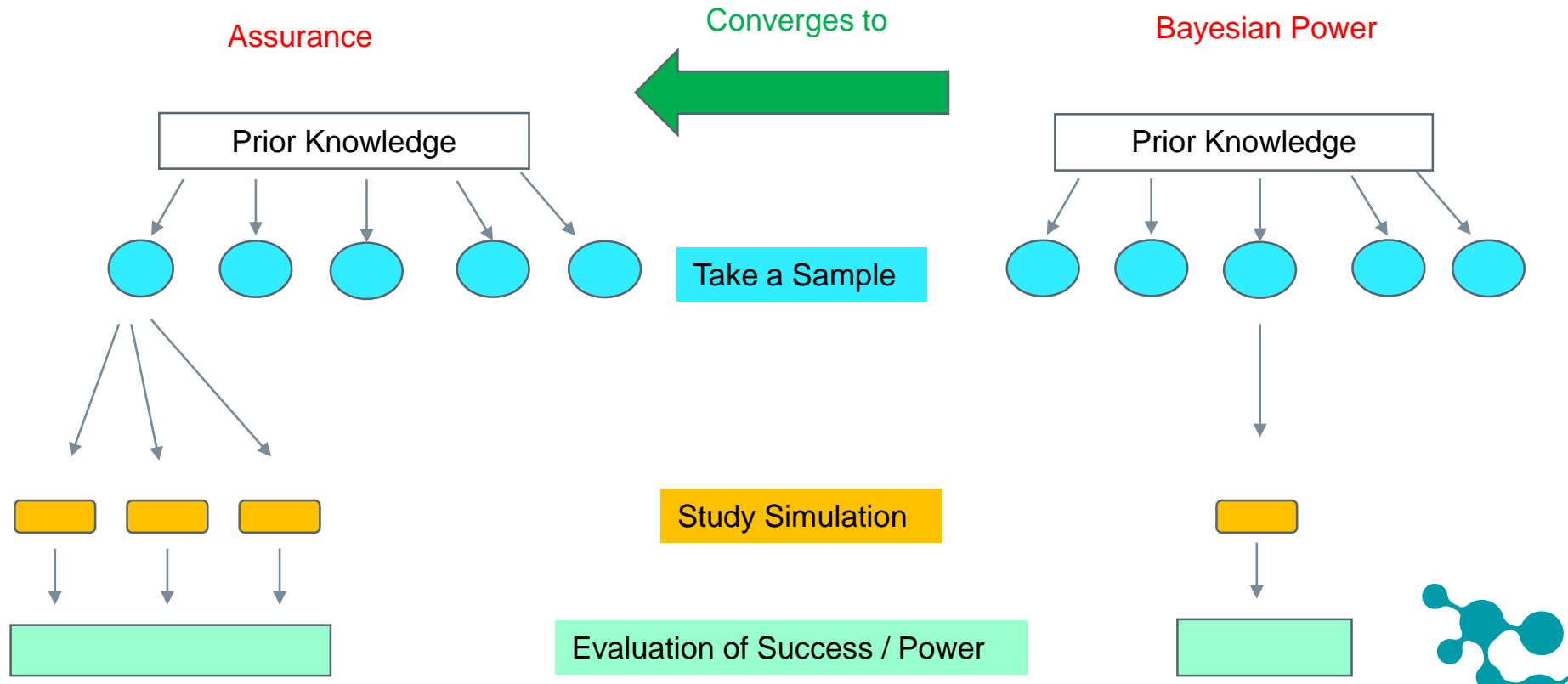


## Going Further to Bayesian Methods

- ▶ From the hypothetical distribution of parameter values, generate representative values.
- ▶ Generate data from these values
- ▶ Compute the posterior on this sample using appropriate Bayesian analysis
- ▶ Tally , from the posterior if the goal is achieved
- ▶ Repeat
- ▶ **THIS IS NEARLY ASSURANCE**
  - It is assurance where power at each value of the prior is approximated by a unique sample (1-0)



# Bayesian Power vs Assurance



## ● Back To the Real Case Study

# Different Methods to Compute PoS

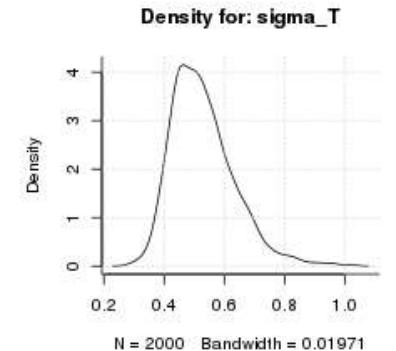
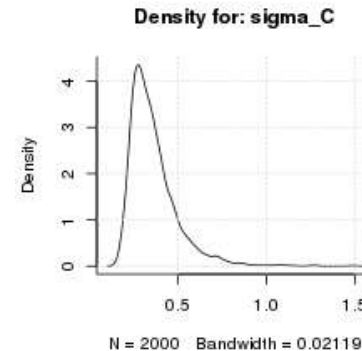
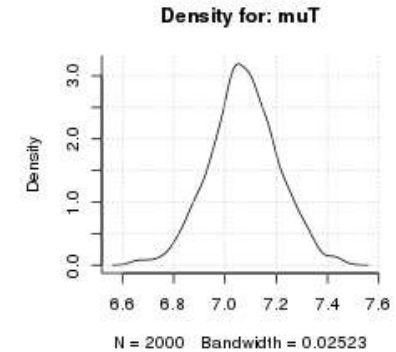
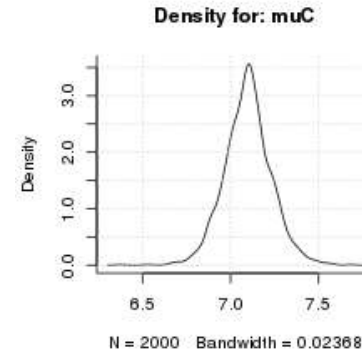
- ▶ Assurance
  - Directly from posterior distribution:
    - Calculation of power for each difference and precision of the posterior distribution
    - PoS=Mean of power
- ▶ Bayesian Power (as described by Kruschke)
  - From Posterior Distribution:
    - Calculation of the success or failure for difference and precision of the posterior distribution
    - PoS = Mean number of successes
  - It is an assurance methodology where the power is computed on a unique sample





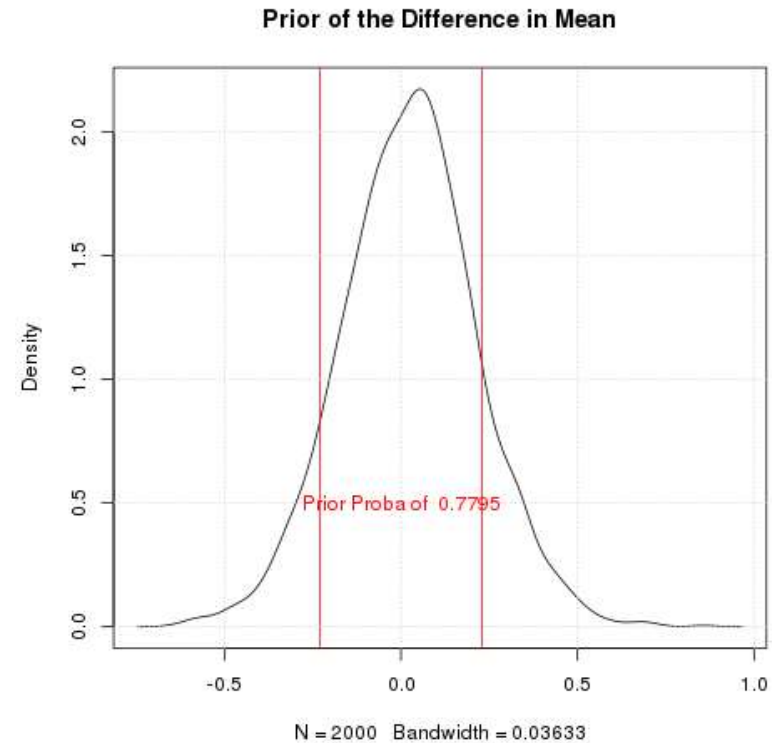
## Posterior distributions from previous studies estimation

- ▶ From historical data we can obtain the following distributions.
- ▶ These are our prior probabilities for constructing our beliefs of future studies
- ▶ The aim is to assess if the 90% credible interval of the difference between the two means is contained within the interval [-0.23;0.23]



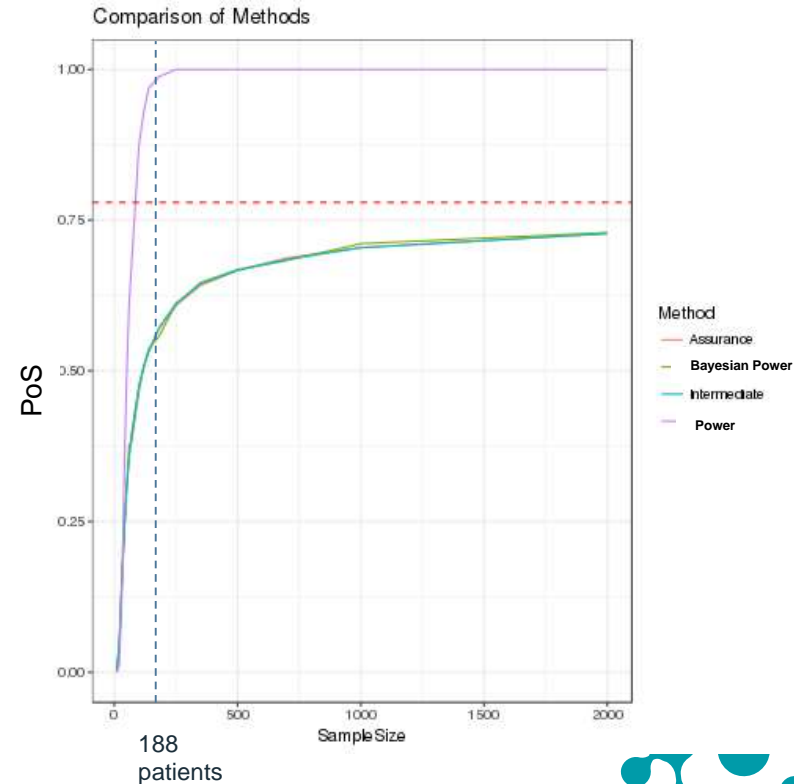
## Comparison of Methods: A practical case

- ▶ The difference in mean has a prior probability of .7795 of being within the defined equivalence margins



## Comparison of Methods: A practical case

- ▶ We see that the profiles for the power built on the predictive methodology converges to 1 very quickly.
- ▶ For very small sample sizes the three other methods actually yield a higher “power”
- ▶ The three other methods converge to  $\pm 0.78$ , that is the prior probability that the difference between the two groups is actually within the equivalence margins defined
- ▶ Dashed line represent the case with a total sample size of 188 patients



## Conclusion

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- ▶ Power in the frequentist view does not account for prior beliefs
- ▶ Power converges to 1 if the condition is in the acceptance region
- ▶ Bayesian Power and Assurance (asymptotically equivalent) do not converge to one because under the prior beliefs, there could exist a chance that we will never be able to show what we desire
- ▶ This reflects better the true world probabilities

