

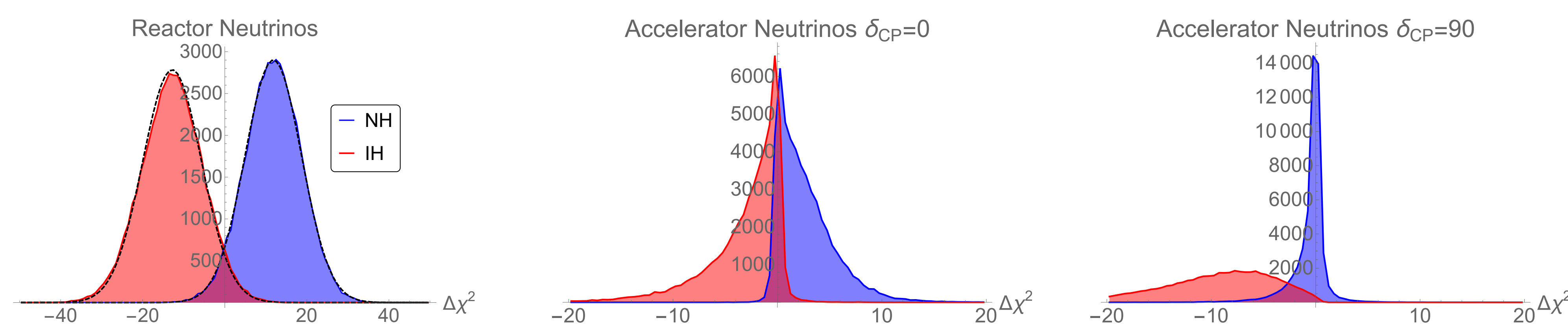
## STATISTICAL DISTRIBUTION

The two hierarchies are non-nested hypothesis, hence Wilks' theorem cannot be applied. This means that the  $\Delta\chi^2$ , defined as

$$\Delta\chi^2 = \chi_{IH}^2 - \chi_{NH}^2$$

does not follow a one-degree-of-freedom  $\chi^2$  distribution and the number of  $\sigma$ 's  $n \neq \sqrt{\Delta\chi^2}$ .

Under certain conditions it instead follows a Gaussian distribution [1, 2, 3] with  $\sigma = 2\sqrt{\overline{\Delta\chi^2}}$ , where  $\overline{\Delta\chi^2}$  is the expected  $\Delta\chi^2$ . These conditions seem to be satisfied in the case of reactor neutrino experiments, but not in the case of accelerator neutrino experiments.



**Figure 1:** Left Panel: Pdf of  $\Delta\chi^2$  in reactor neutrino experiments; the dashed curves are Gaussian fits. Central and right panels: Pdf of  $\Delta\chi^2$  for accelerator neutrinos, using different values of  $\delta_{CP}$  [4]

## ADDITIONAL PARAMETER

A possible solution for the non-nested problem is to introduce a new pull parameter, without physical meaning: this was proposed first in [5] for reactor neutrino experiments, writing

$$|\Delta m_{31}^2| = |\Delta m_{32}^2| + (2\eta - 1)|\Delta m_{21}^2|$$

$\eta = 1 \rightarrow \text{NH}$ ,  $\eta = 0 \rightarrow \text{IH}$ . A more general approach, that can be used also with accelerator neutrinos, was suggested in [6], considering the linear combination

$$\eta f(E) + (1 - \eta)g(E)$$

where  $f(E)$  and  $g(E)$  are the spectra for normal and inverted hierarchy, respectively. In this way the problem is reduced to parameter fitting;  $\Delta\chi^2$ , which can be defined for both hierarchies, now follows a  $\chi^2$  distribution

$$\Delta\chi_{NH}^2 = \chi_{\eta=1}^2 - \chi_{\min}^2 \quad \Delta\chi_{IH}^2 = \chi_{\eta=0}^2 - \chi_{\min}^2$$

## LAPLACE METHOD

In the Bayesian approach, if additional pull parameters  $\theta$  are present they must be **marginalized**, *i.e.*

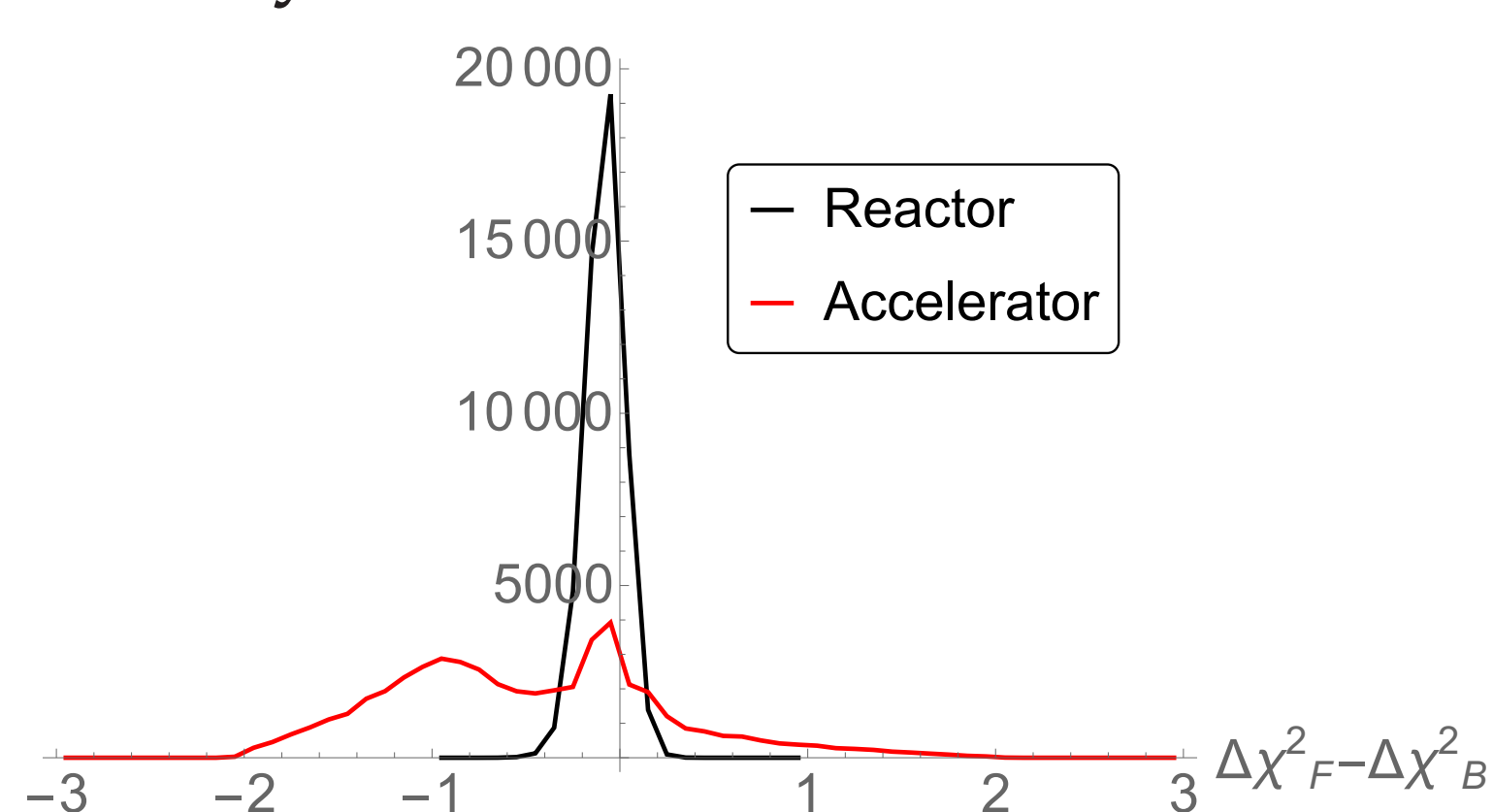
$$\Pr(\mathbf{D}|\text{MH}) = \int \Pr(\mathbf{D}|\text{MH}, \theta)\pi(\theta)d^n\theta$$

while in the frequentist approach they are **minimized**, using the best fit value for  $\theta$ .

When several pull parameters are present, marginalization requires multi-dimensional integrals. However, if the  $\chi^2$  is strongly peaked around the minimum, and  $\text{Det}(\mathbf{C})$  is the same for the two hierarchies ( $\mathbf{C}$ =covariance matrix), using the Laplace method we can prove that

$$\Delta\chi_B^2 = \Delta\chi_F^2$$

$\Delta\chi_B^2, \Delta\chi_F^2$  are the  $\Delta\chi^2$  obtained marginalizing or minimizing over the pull parameters, respectively.



**Figure 2:** Difference between marginalization and minimization in accelerator and reactor neutrino experiments (simplified models with only one parameter)

## CONCLUSIONS

- Some results regarding the pdf of  $\Delta\chi^2$ , but not applicable to all cases
- Different approaches available: no "right" or "wrong" choices, but important to report the convention used
- Bayesian method gives only **one quantity**, *i.e.*  $\Pr(\text{MH}|\mathbf{D})$ , while frequentist gives the compatibility of **both** hierarchies with the data
- **Different and complementary** information: more accurate analysis reporting both

## FREQUENTIST APPROACH

**Hypothesis Test** [3] We test the two hierarchies separately: before the experiment, we define two threshold  $T_{c,NH}$  and  $T_{c,IH}$ ; if  $\Delta\chi^2 > T_{c,IH}$  the inverted hierarchy is rejected, if  $\Delta\chi^2 < T_{c,NH}$  the normal hierarchy is rejected :

- The confidence level (CL) that can be achieved depends only on  $T_{c,NH}, T_{c,IH}$ , not on the actual result of the experiment (that determines only if such a CL is achieved or not)  $\Rightarrow$  convenient to quantify the precision of future experiments
- It is possible to accept or reject both hierarchies at the same time
- To calculate CL we need to know the pdf of  $\Delta\chi^2$ ; problems when it depends strongly on pull parameters

Possible definition of sensitivity (for the Gaussian, symmetric case, generalization is trivial):

- **Median Sensitivity:**  $T_{c,NH(IH)} = -(+)\overline{\Delta\chi^2}$ ,  $n = \sqrt{\overline{\Delta\chi^2}}$ , power=0.5
- **Crossing Sensitivity:**  $T_{c,NH} = T_{c,IH} = 0$ ,  $n = \sqrt{\overline{\Delta\chi^2}/2}$ , power=CL

**p-value** Probability of getting a "more extreme" result than the one obtained from the experiment

$$\text{p-value} = \Pr(\Delta\chi^2 > (<)\Delta\chi_{obs}^2 | \text{IH}(\text{NH}))$$

- Also relies on the knowledge of  $\Delta\chi^2$ 's pdf
- **Two** quantities must be considered: **both** the p-values for NH and IH. Ex: NH excluded at  $5\sigma$ 's, IH excluded at  $1\sigma$ 's or NH excluded at  $5\sigma$ 's, IH excluded at  $5\sigma$ 's
- p-value(NH)  $\ll$  1 does not necessarily implies IH

## BAYESIAN APPROACH [2, 4]

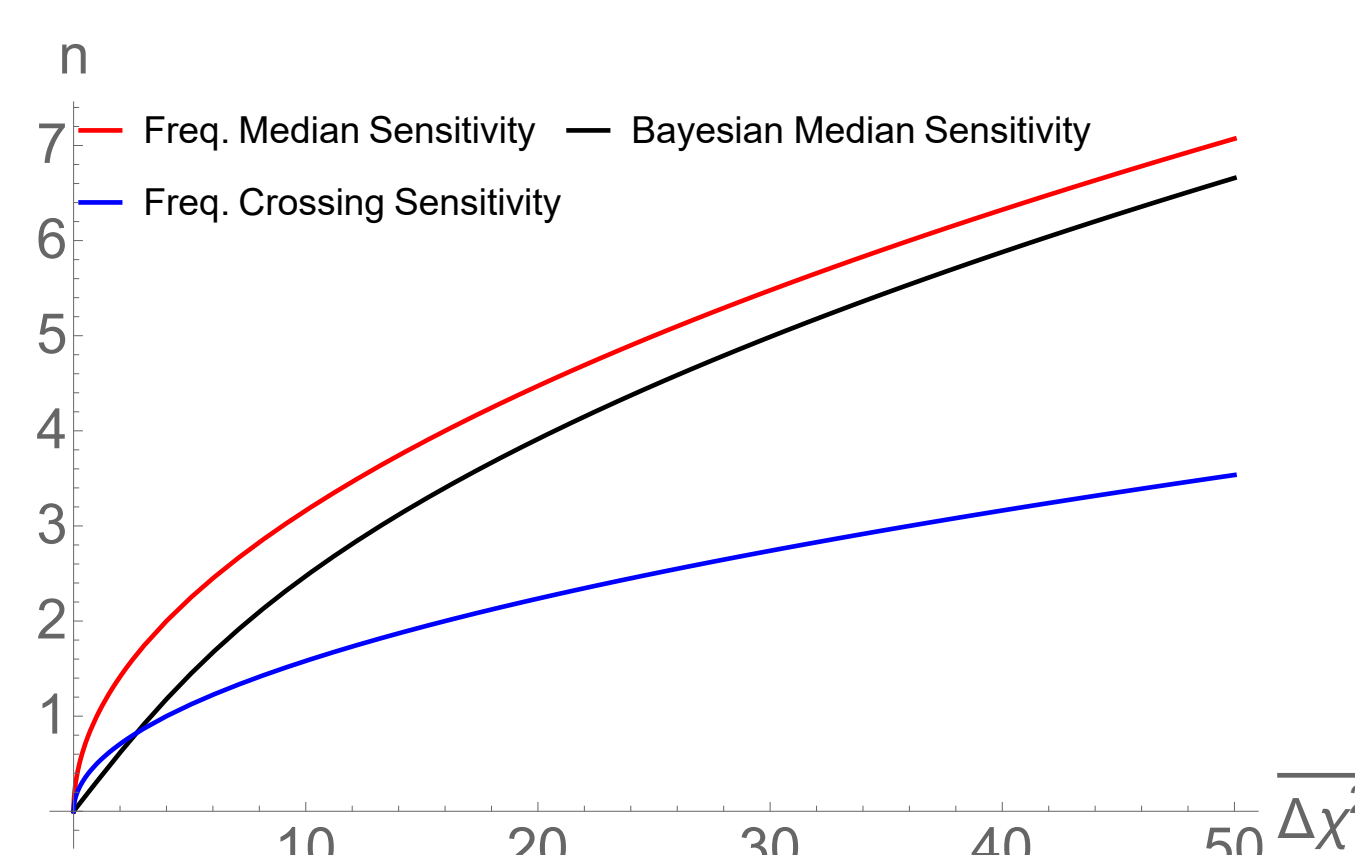
- **Frequentist Approach**  $\Rightarrow \Pr(\mathbf{D}|\text{MH})$ : probability of the data  $\mathbf{D}$  given MH. Suppose we find p-value(NH)=0.0001: alone not enough to say NH is unlikely
- **Bayesian Approach**  $\Rightarrow \Pr(\text{MH}|\mathbf{D})$ : probability that the hierarchy is normal (or inverted) given the data

Using Bayes theorem

$$\Pr(\text{NH}|\mathbf{D}) = \frac{\pi(\text{NH})}{\pi(\text{NH}) + \pi(\text{IH})K^{-1}}$$

$$\text{Bayes Factor: } K = \frac{\Pr(\mathbf{D}|\text{NH})}{\Pr(\mathbf{D}|\text{IH})} = e^{-\Delta\chi^2/2}$$

- Provide a single quantity  $\rightarrow$  better quantify likelihood of NH vs. IH
- Depends upon the choice of priors! However very natural choice for MH:  $\pi(\text{NH}) = \pi(\text{IH}) = 0.5$
- Bayesian and frequentist give **different (and complementary) information!**
- More complete analysis by using both approaches



## REFERENCES

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