Extrapolating sensitivity at infinite statistics

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Goal

Goal:

Show whether the Neutrino Mass Ordering sensitivity increases when using multiple categories in the PID variable $\,q\,$

 Method: introduce some binning scheme in PID variable q and calculate the sensitivity



Issue

- We see: Sensitivity increases with more `PID categories`
- Due to signal in PID variable q
- Due to statistics of response matrix
 - More categories \rightarrow less events per Response Matrix (track, shower, etc.)
 - Response matrix mapping gets worse with less events
 - Event distributions deviate more from a true detector response
 - Easier to distinguish two hypotheses H0, H1
 - Sensitivity to hypothesis H1 goes up
- Decouple these 2 effects





Response matrix property

When calculating the average chi2 for a RM [1]:

$$\begin{split} \langle \Delta \chi^2 \rangle (N) &= \langle \Delta \chi^2_{\infty} \rangle + \sum_{i}^{true \ reco} \frac{K_{ij}}{N_i} + O(N^{-\frac{3}{2}}) \\ K_{ij} &\approx N_j^A \Big(\frac{n_i^A}{N_j^A} - \frac{n_i^B}{N_j^B} \Big) \Phi_i(X_j) V_j e_i \\ \\ \hline \langle \Delta \chi^2 \rangle (N) &\approx \langle \Delta \chi^2_{\infty} \rangle + \frac{K}{N_{MC}} \end{split}$$

• See backup for details on K_{ij}

[1] Neutrino oscillations and Earth tomography with KM3NeT-ORCA, S. Bourret, 2018

Literature



Determining the Neutrino Mass Hierarchy with the Precision IceCube Next Generation Upgrade, *L. Schulte*, 2015

Measuring the neutrino mass hierarchy with the future KM3NeT/ORCA detector, *J. Hofestädt*, 2017

Neutrino oscillations and Earth tomography with KM3NeT-ORCA, *S. Bourret*, 2018

Fit to chi2 of sampled response matrix



Apply method to sensitivity curve

 $\langle \Delta \chi^2 \rangle$ at infinite statistics



Apply method to sensitivity curve

 $\langle \Delta \chi^2 \rangle$ at infinite statistics $\forall \Delta \chi^2$ IO at ∞ statistics 2 PID categories ∞ statistics 3 PID categories ∞ statistics 4 PID categories IO at ∞ statistics 5 PID categories $\langle \Delta \chi^2 \rangle$ IO at ∞ statistics 10 PID categories 0 40 42 44 46 48 50 θ_{23}

Backup

$$\langle \Delta \chi^2 \rangle(N) = \langle \Delta \chi^2_{\infty} \rangle + \sum_{i}^{true \ reco} \frac{K_{ij}}{N_i} + O(N^{-\frac{3}{2}})$$
$$K_{ij} \approx N_j^A \left(\frac{n_i^A}{N_j^A} - \frac{n_i^B}{N_j^B}\right) \Phi_i(X_j) V_j e_i$$

 N_j^A : average number of events in reco bin j under hypothesis A n_i^B : average number of events in true bin i under hypothesis B

 Φ_i : the pdf that maps true events from bin i to bin j

 X_j : bin center of reco bin j

 V_j : bin volume of reco bin j

 e_i : detector efficiency at true bin i







h chi2 dist 3

100

Chi2 dist at 3.98 percent MC in RM

100

10

52.79

4.869

13.96 / 16

h chi2 dist 6

Constant 9.649 ± 1.368

Entries

Mean

Std Dev

 χ^2 / ndf

Mean

Sigma

100

13.31

1.021

10.8/16

 13.36 ± 0.12

1.001 ± 0.109







10









14 15

12 13

Chi2 dist at 15.85 percent MC in RM