

Uncertainty quantification of neutron induced cross-section of Zr-90 with model defect treatment using Nuclear Data Evaluation Pipeline of Uppsala (NEPU)

The Curious Case of the Biased Zr-90 Fit

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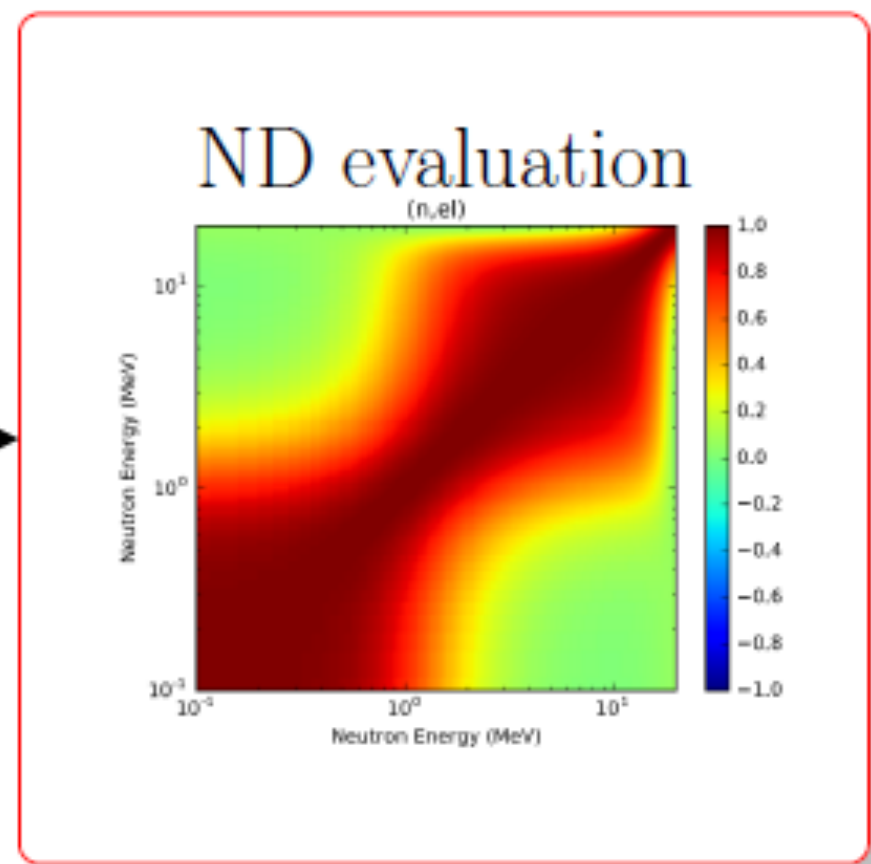
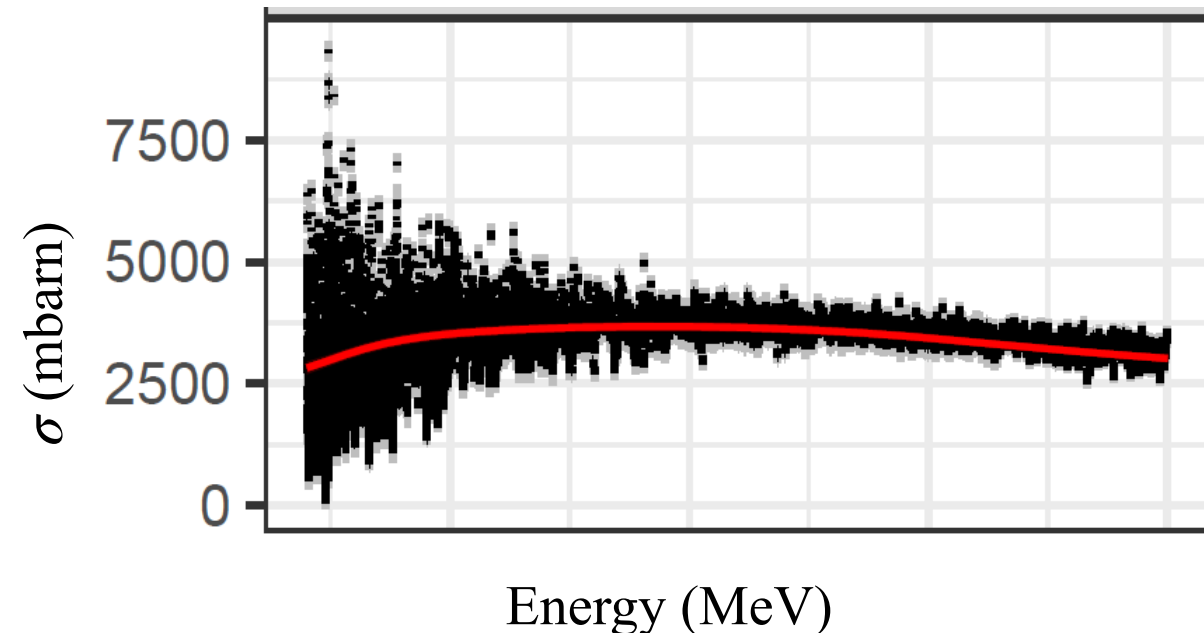
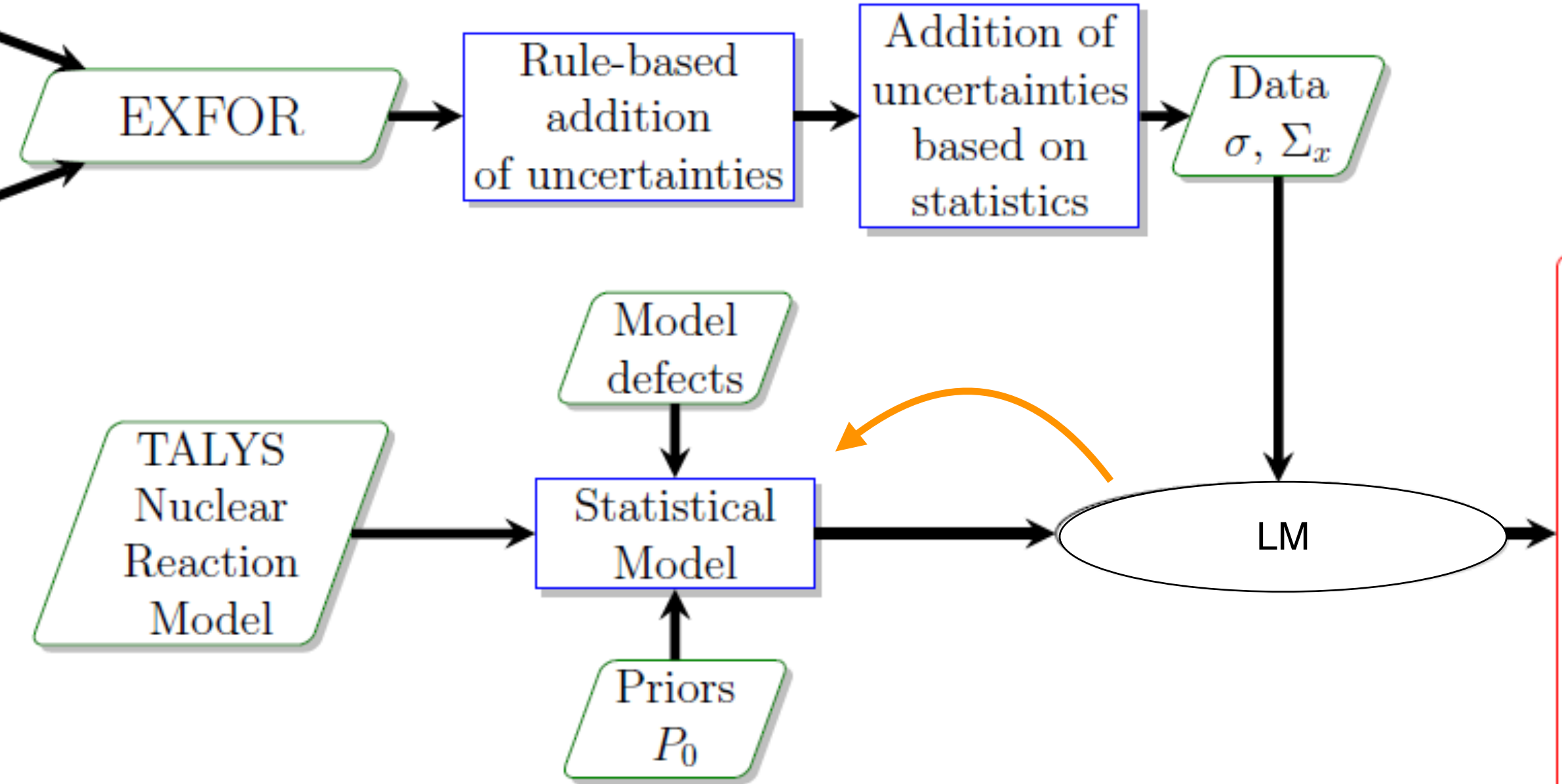
NEPU: Nuclear Data Evaluation Pipeline of Uppsala

A pipeline for nuclear data evaluation that implements (and further develops) methodology to **treat model defects** and **inconsistent experimental data** that has originated in research activities at UU.

Goal: Automatization, Reproducible.

G. Schnabel, H. Sjöstrand, J. Hansson, D. Rochman, A. Koning, R. Capote Nucl. Data Sheets 173, 239 (2021)

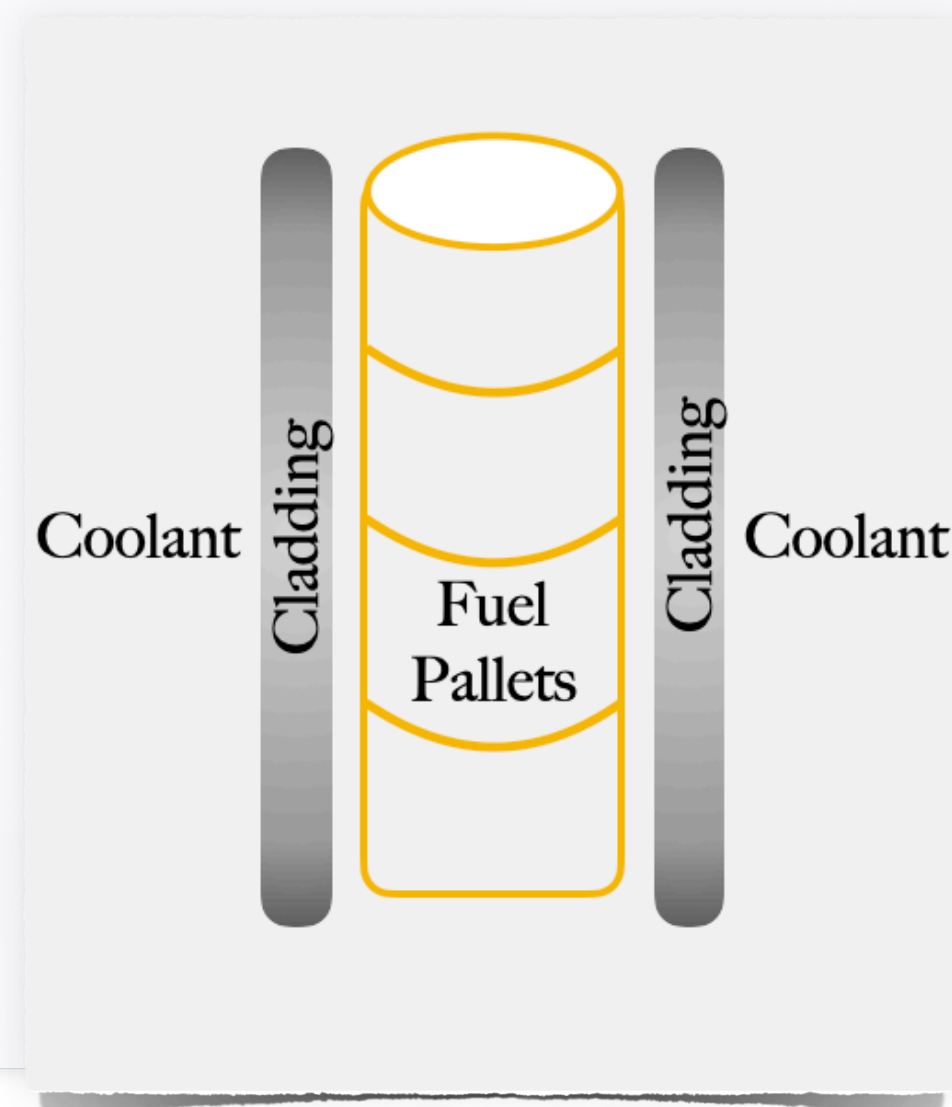
A.Gök, E. Andersson-Sundén, J. Hansson, and H. Sjöstrand, EPJ Web of Conf. 294, 04005 (2024)



Zr-90 on NEPU - Input setup

Why Zr-90?

- Assembly structure & cladding material
- Isotopic abundance of Zr-90 ~ 51%



For Zr-90 we have used TENDL-2019 input and TALYS-1.95 version.

We plan to use latest TENDL input with TALYS-2.2

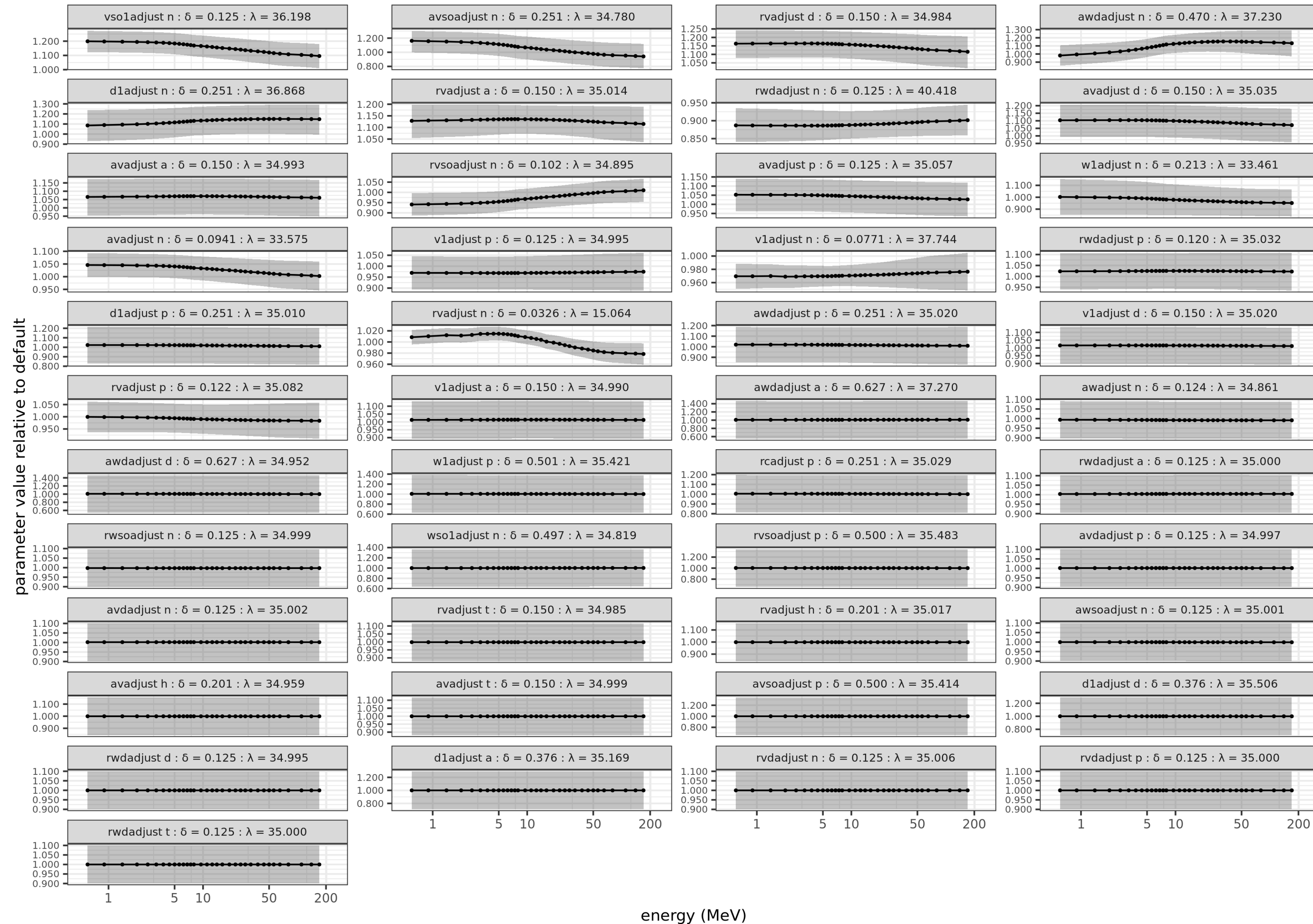
Total experimental data: 1727 across 9 reaction channels

Energy range: 1-150 MeV

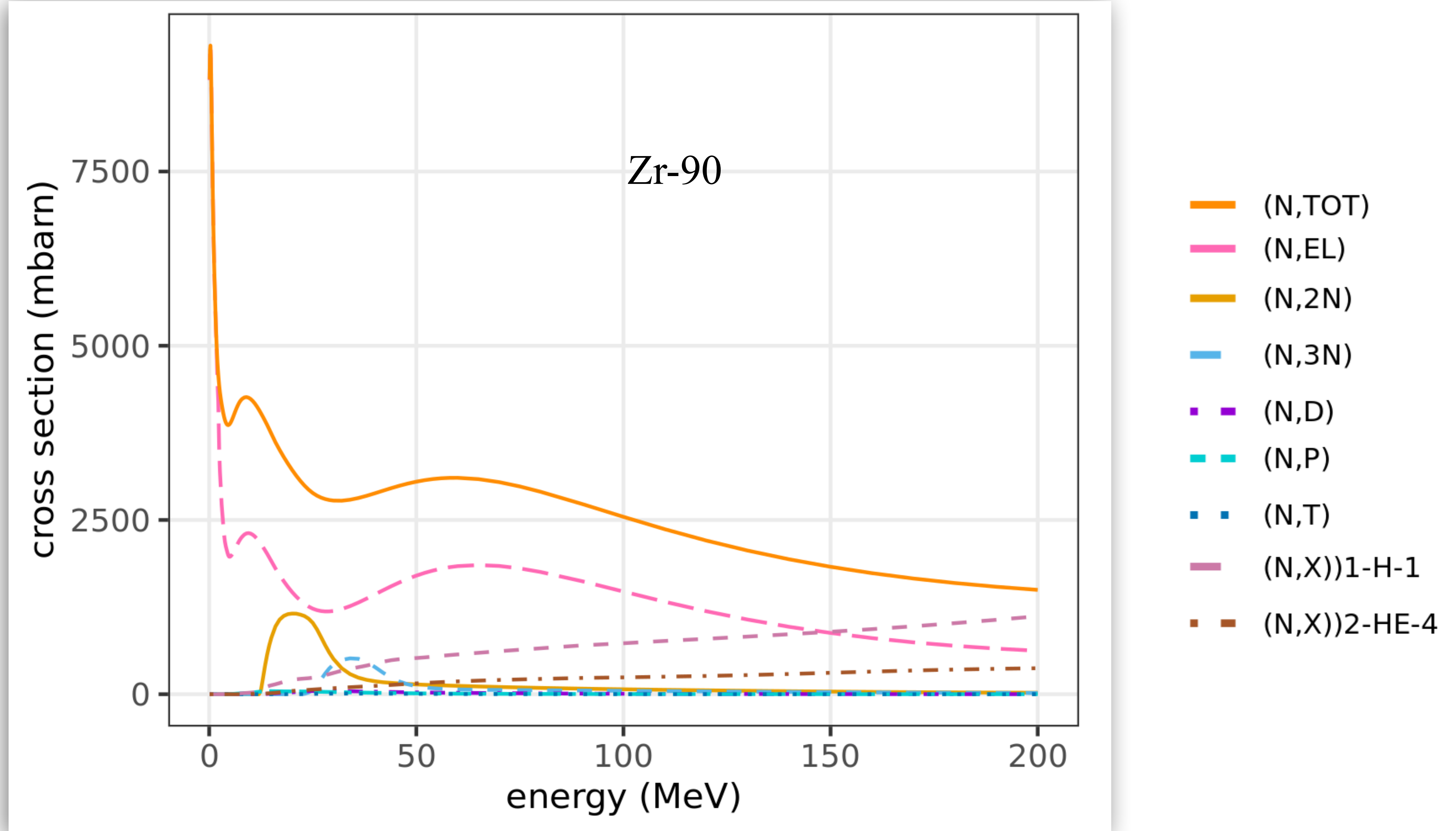
Free TALYS parameters: 3834, including energy dependence

Selected for LM fit: 1614 / 3834

GP treatment on energy dependent model parameters (preliminary)

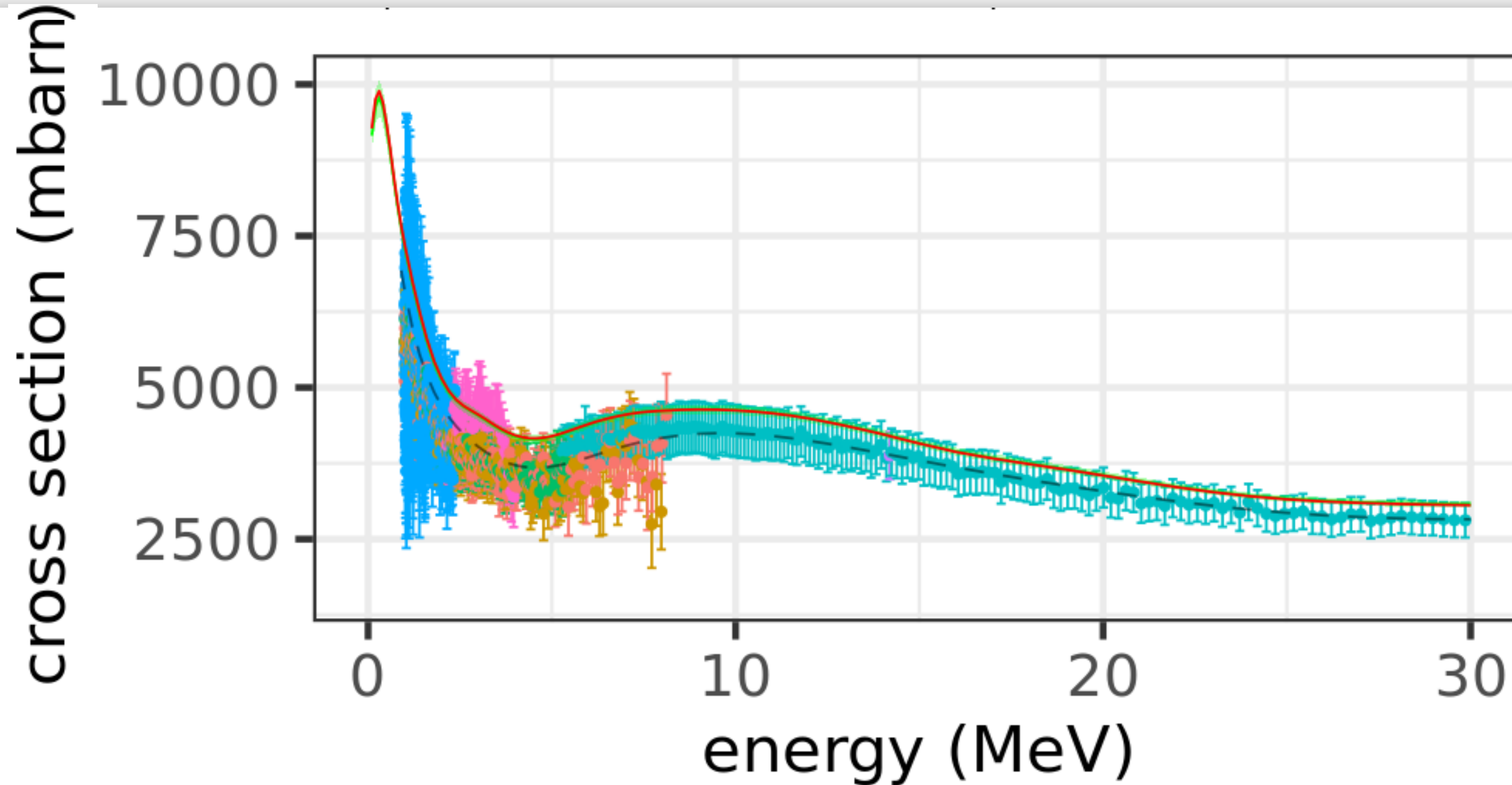


Posterior cross-section for neutron-induced channels (preliminary)



The Anomaly

TALYS looks better : but the numbers say otherwise



n = 1535 data points

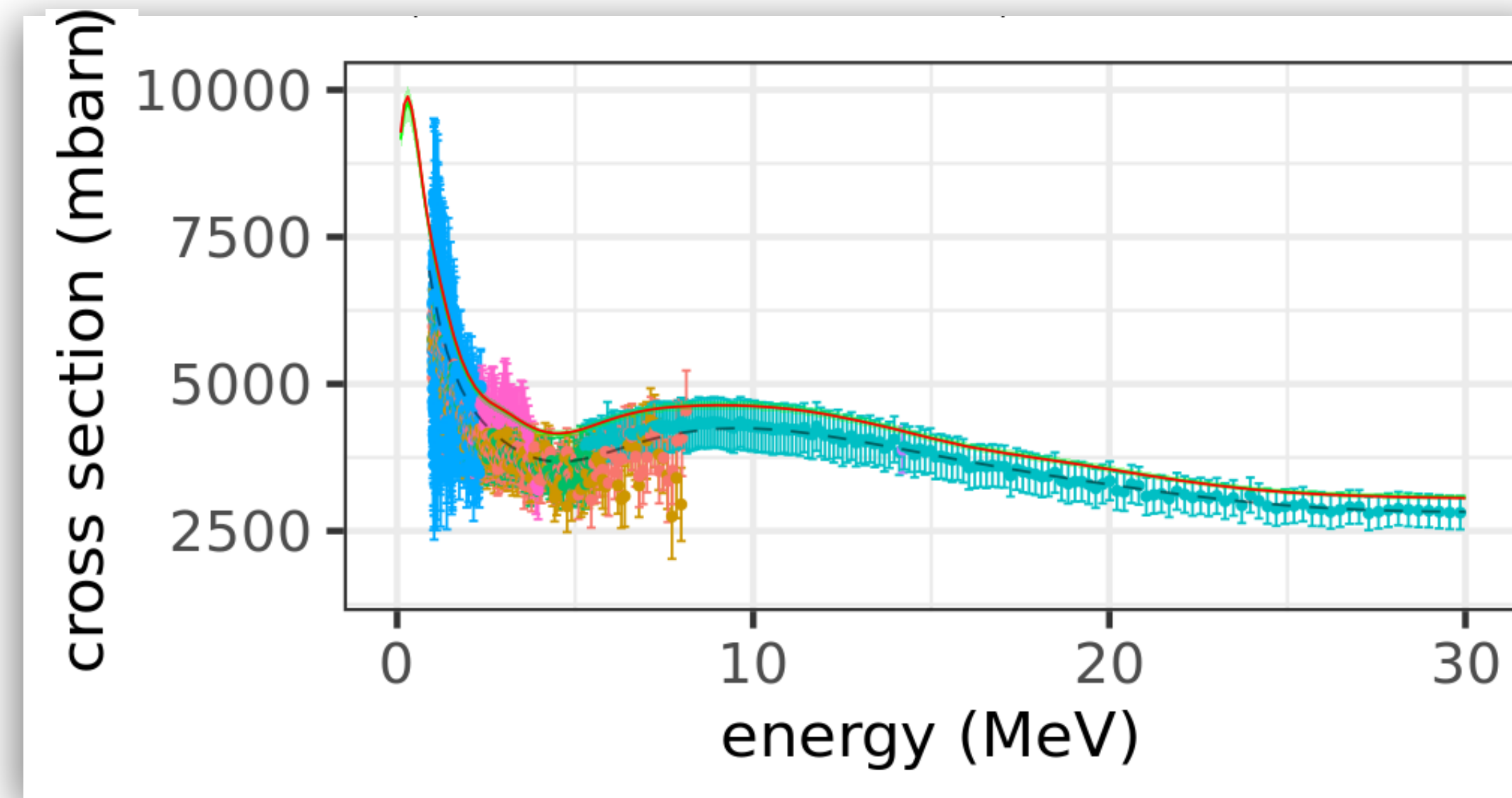
	χ^2	χ^2/n
TALYS default	944 119	615
Posterior fit	209 144	136

TALYS is visually better — yet $\chi^2/n = 615$

Rule-based addition of uncertainties in NEPU

NEPU on what EXFOR reports

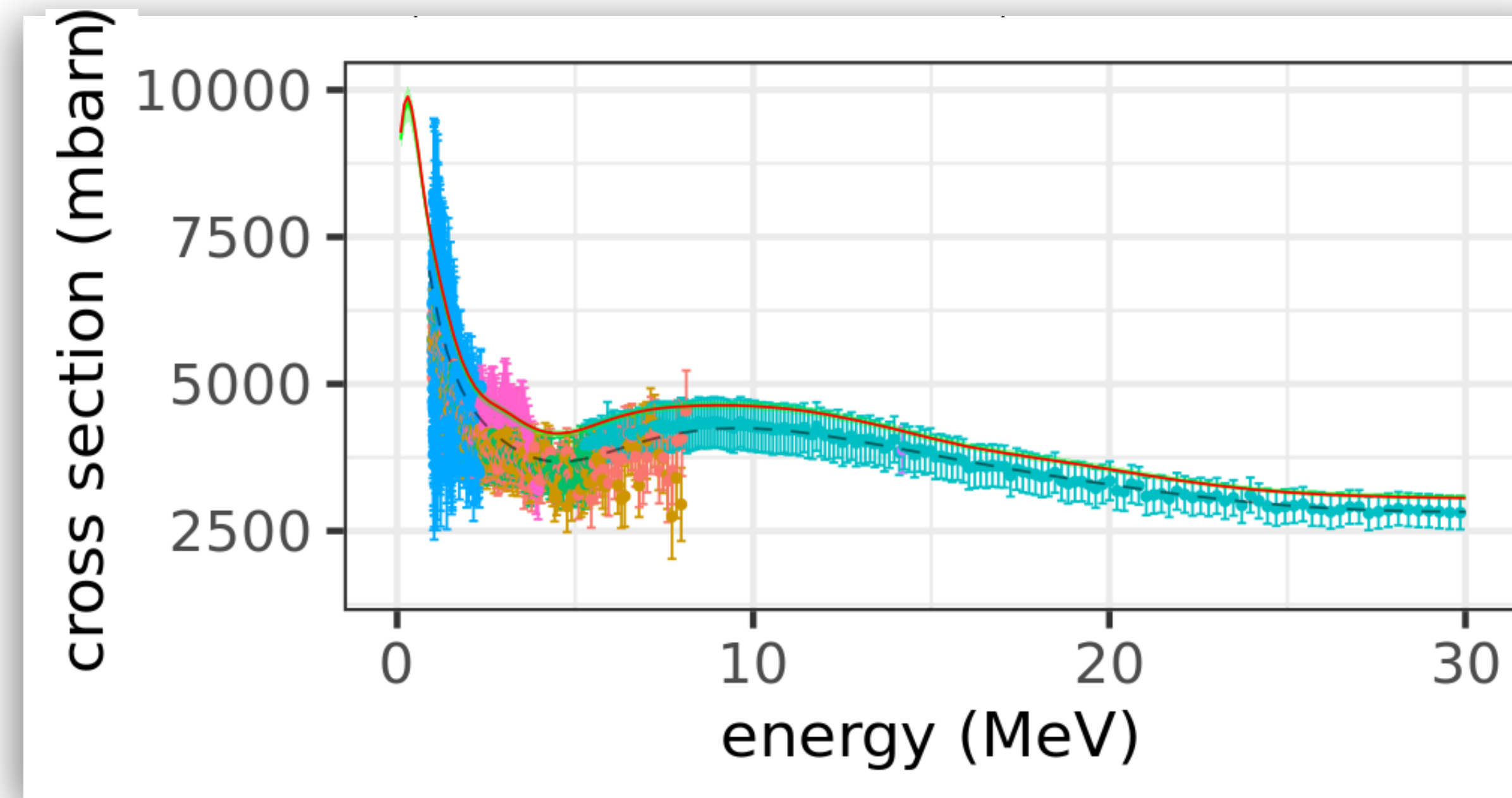
- 1. Statistical unc. reported**
 - use as random (diagonal)
 - use reported sys, or 10% penalty
- 2. Total unc. + systematic components reported**
 - random = $\sqrt{(\sigma^2_{\text{tot}} - \sigma^2_{\text{sys}})}$
 - use reported sys
- 3. Only total unc., no breakdown reported**
 - random = total unc. (as given)
 - add 10% systematic unc
- 4. No uncertainty reported**
 - random = 10%
 - sys = 10%



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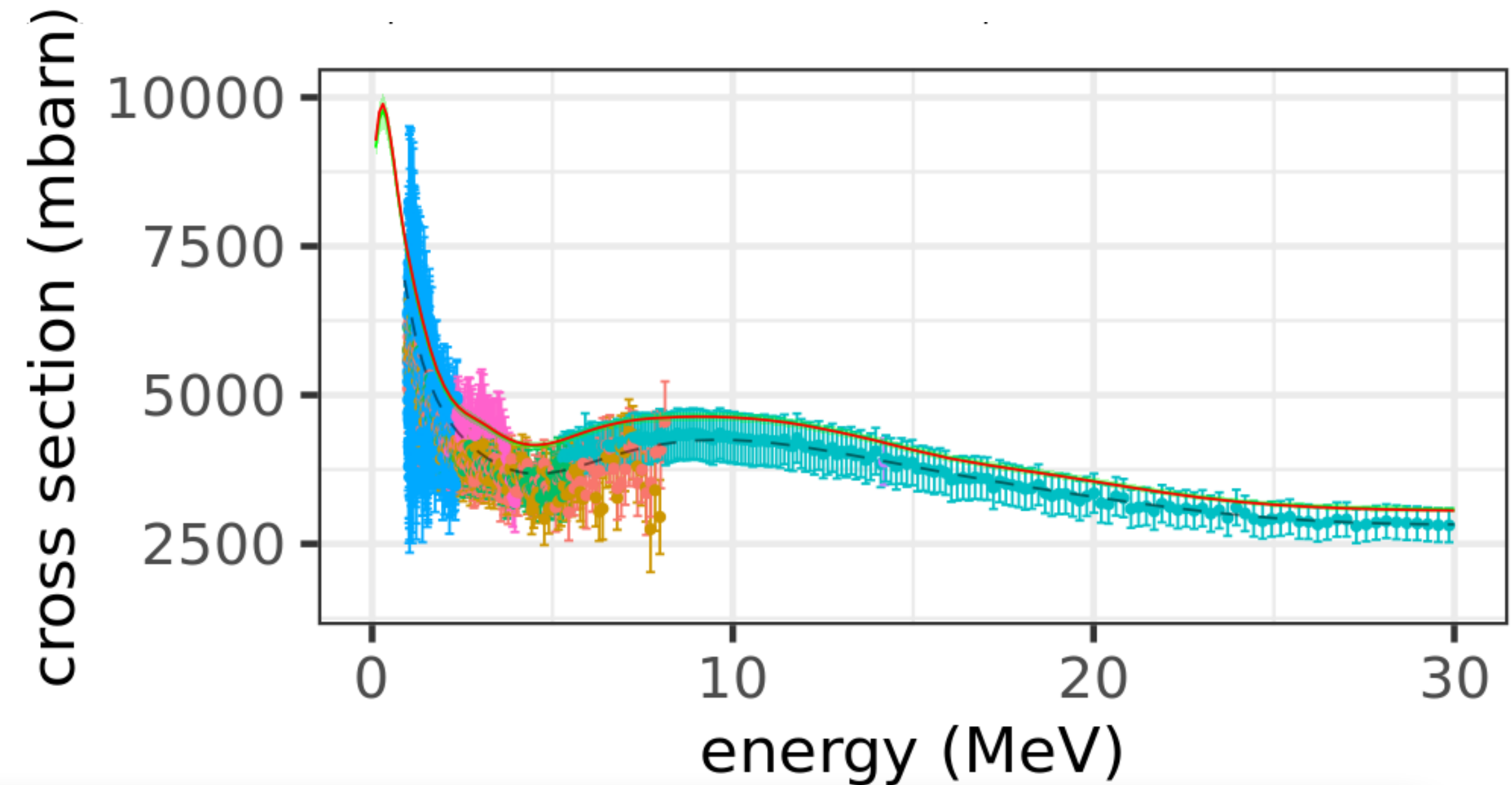


The Covariance Model

What does GLS actually minimise?

$$\chi^2 = (y - f)^T C^{-1} (y - f)$$

where $C_{ij} = \sigma_{i,\text{rand}}^2 \delta_{ij} + \sigma_{i,\text{sys}} \sigma_{j,\text{sys}}$



Random uncertainty

- Diagonal: $C_{ij} = \sigma_i^2 \delta_{ij}$
- Each data point residual penalised independently
- GLS is sensitive to individual point offsets
- A curve displaced upward pays a cost at every point

Systematic (fully correlated)

- Off-diagonal: $C_{ij} = \sigma_{i,\text{sys}} \sigma_{j,\text{sys}}$
- all points move together mode
- A coherent level shift costs almost nothing in χ^2
- GLS minimizes shape mismatch - not the overall offset

When C is dominated by large correlated systematics, GLS optimises shape. A systematic level bias is invisible to χ^2 .

Reconstructing the uncertainty

Before

Uncertainties almost entirely assigned as fully correlated systematic (to penalise inconsistent experiments) → level offset → high posterior bias

After

Partial reallocation to random component → individual point residuals now penalized

Random Unc: Maximum of reported or 1% of data

Systematic Unc: Maximum of reported or 5% of data

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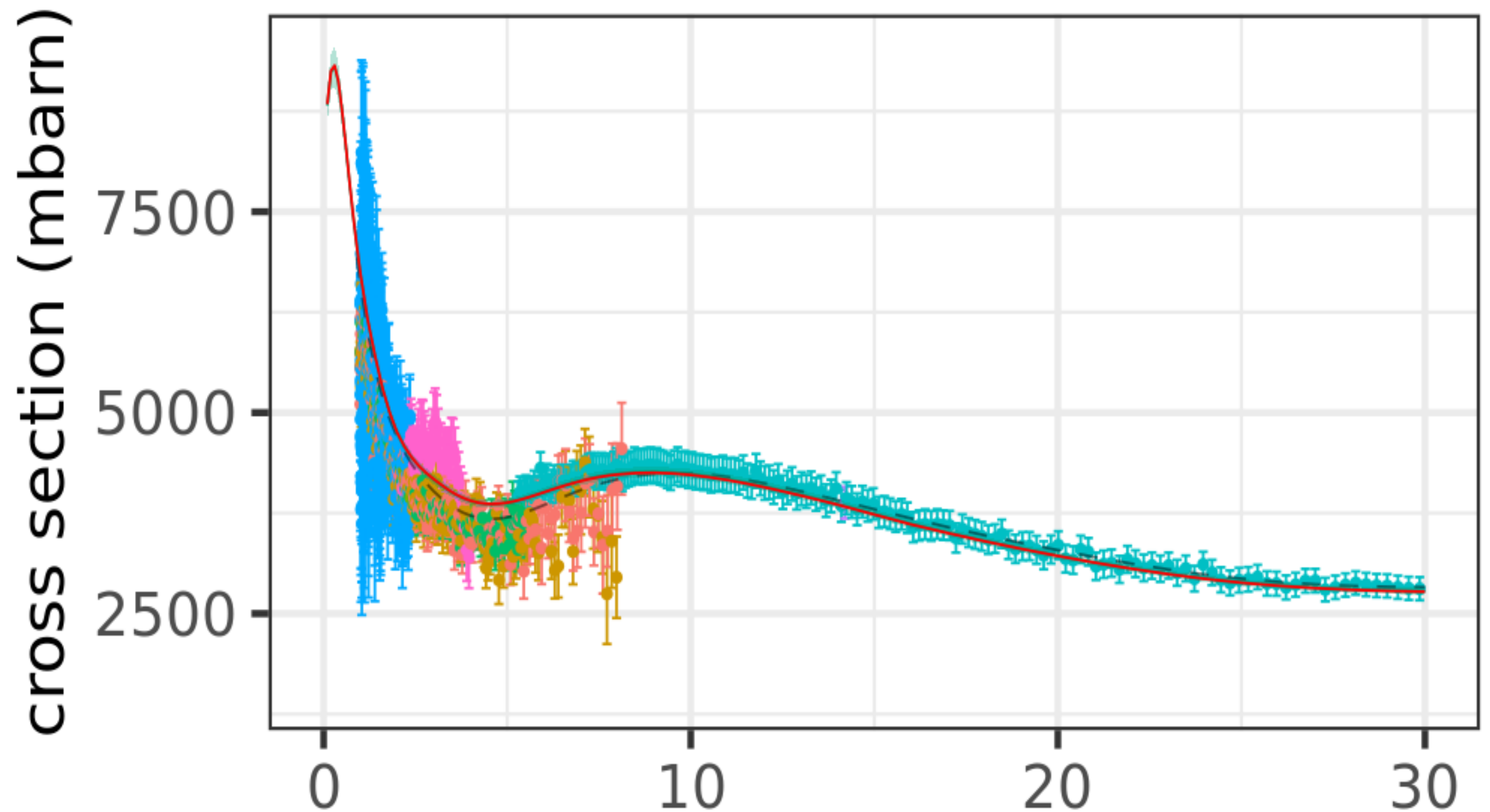
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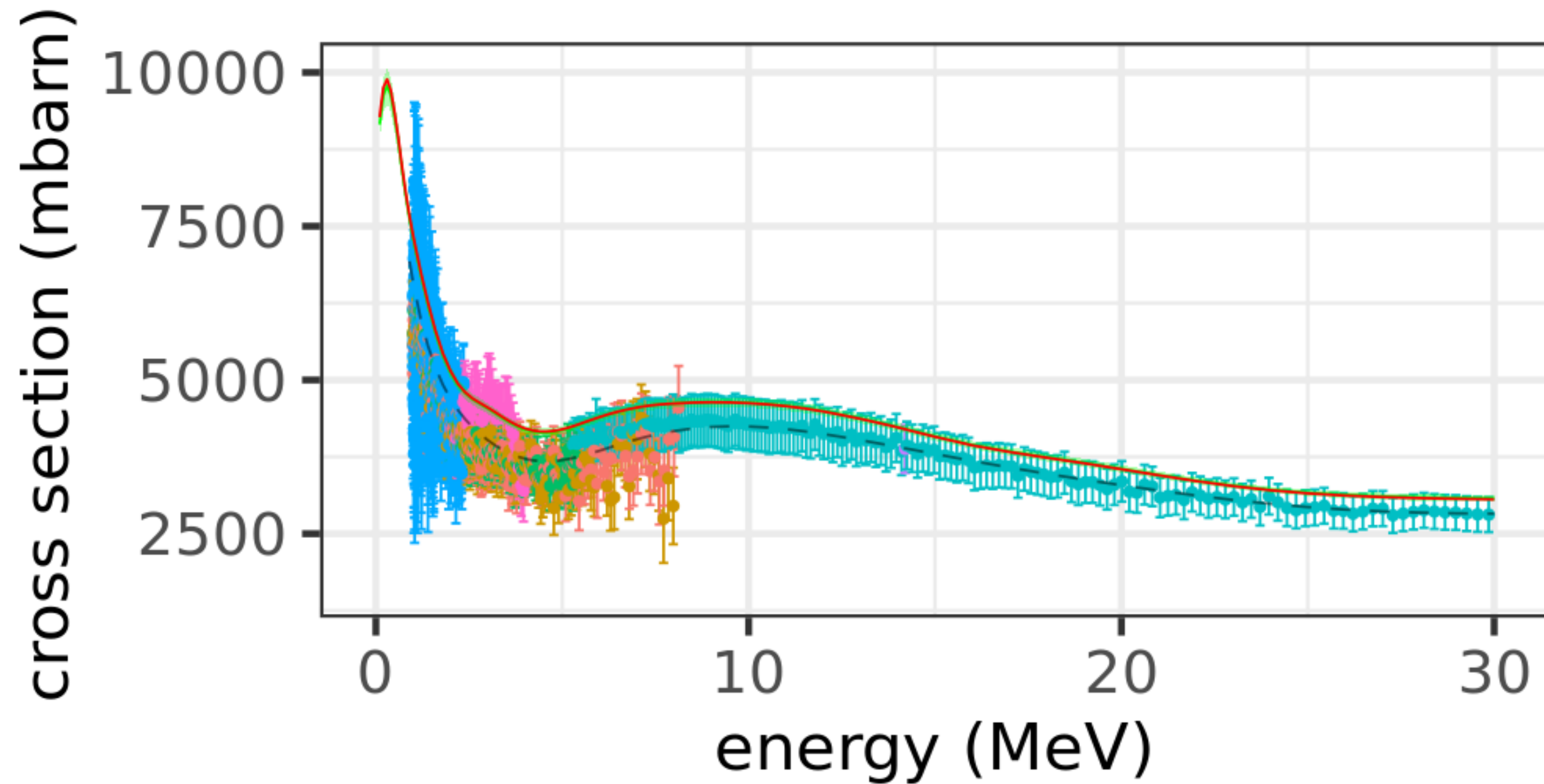
Random Unc: Maximum of reported or 1% of data

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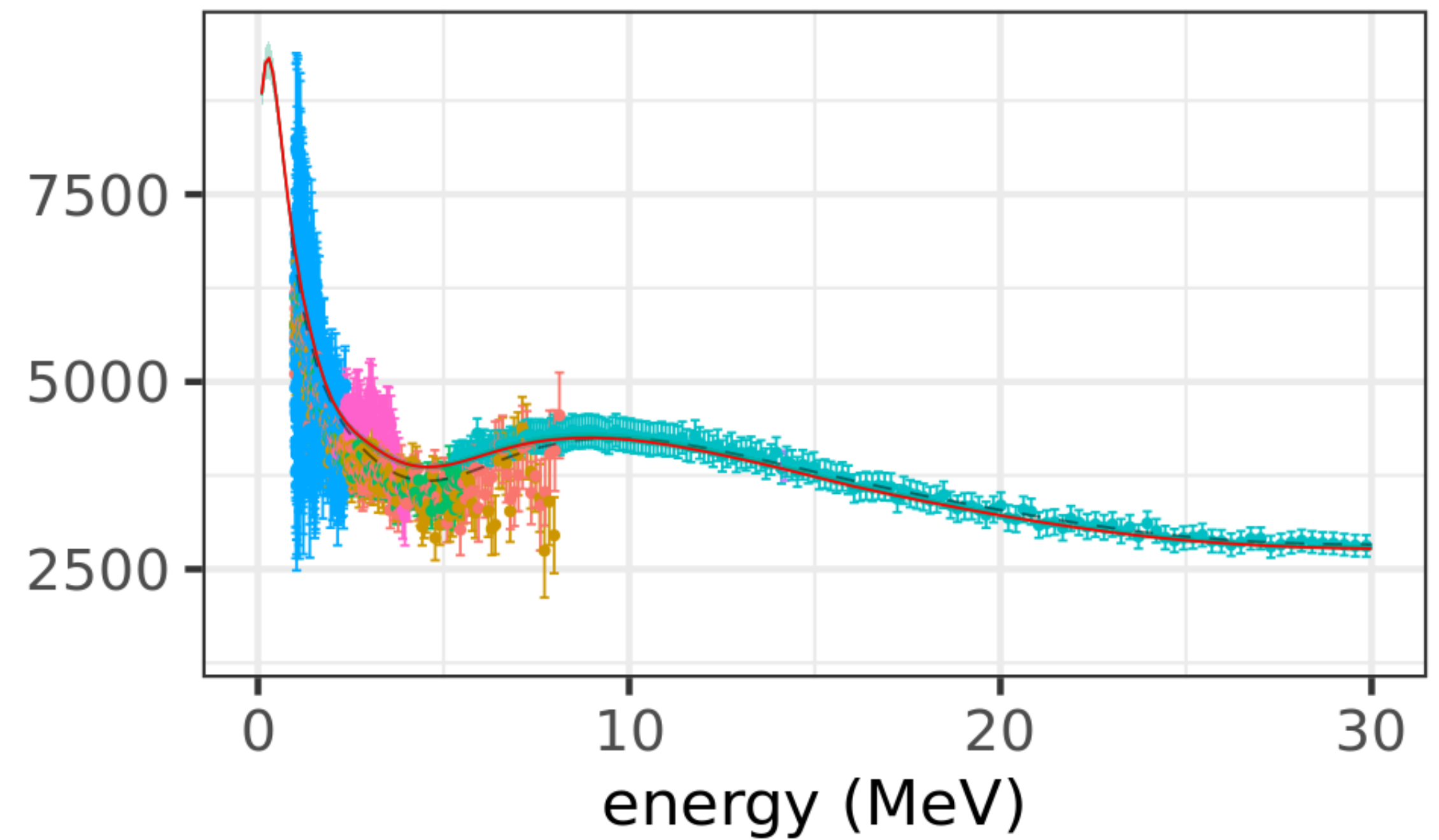
→ **Better fit**



Before vs. After

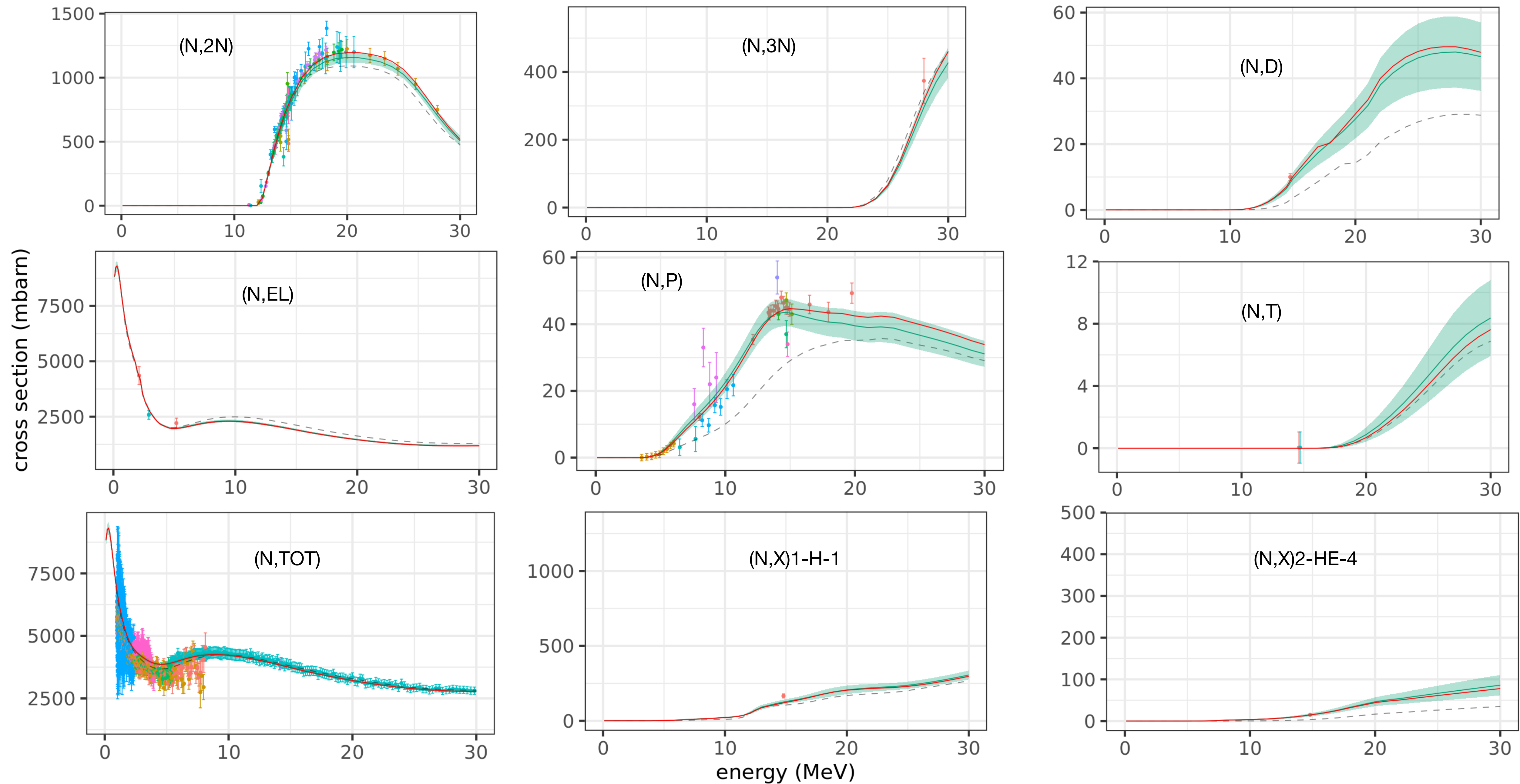


Large χ^2/n | large correlated unc.



$\chi^2/n \approx 1.01$

Uncertainty quantification of $\sigma_{(N,*)}$ for Zr-90 (preliminary result)



Peelle's Pertinent Puzzle

Relative uncertainties should be converted to absolute before entering the covariance matrix.

The PPP bias (without correction)

$$\sigma_{\text{abs},i} = p_i \times y_i$$

If relative uncertainty p_i is quoted on the measured value y_i , and y_i scatters above the true value, σ is inflated — the uncertainty correlates with the fluctuation.
GLS gives them less weight → the result is biased downward

The standard fix

$$\sigma_{\text{abs},i} = p_i \times f(E_i)$$

Replace measured y with a smooth reference f (e.g. TALYS default).
Breaks the correlation.

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When PPP renormalization goes wrong

Example

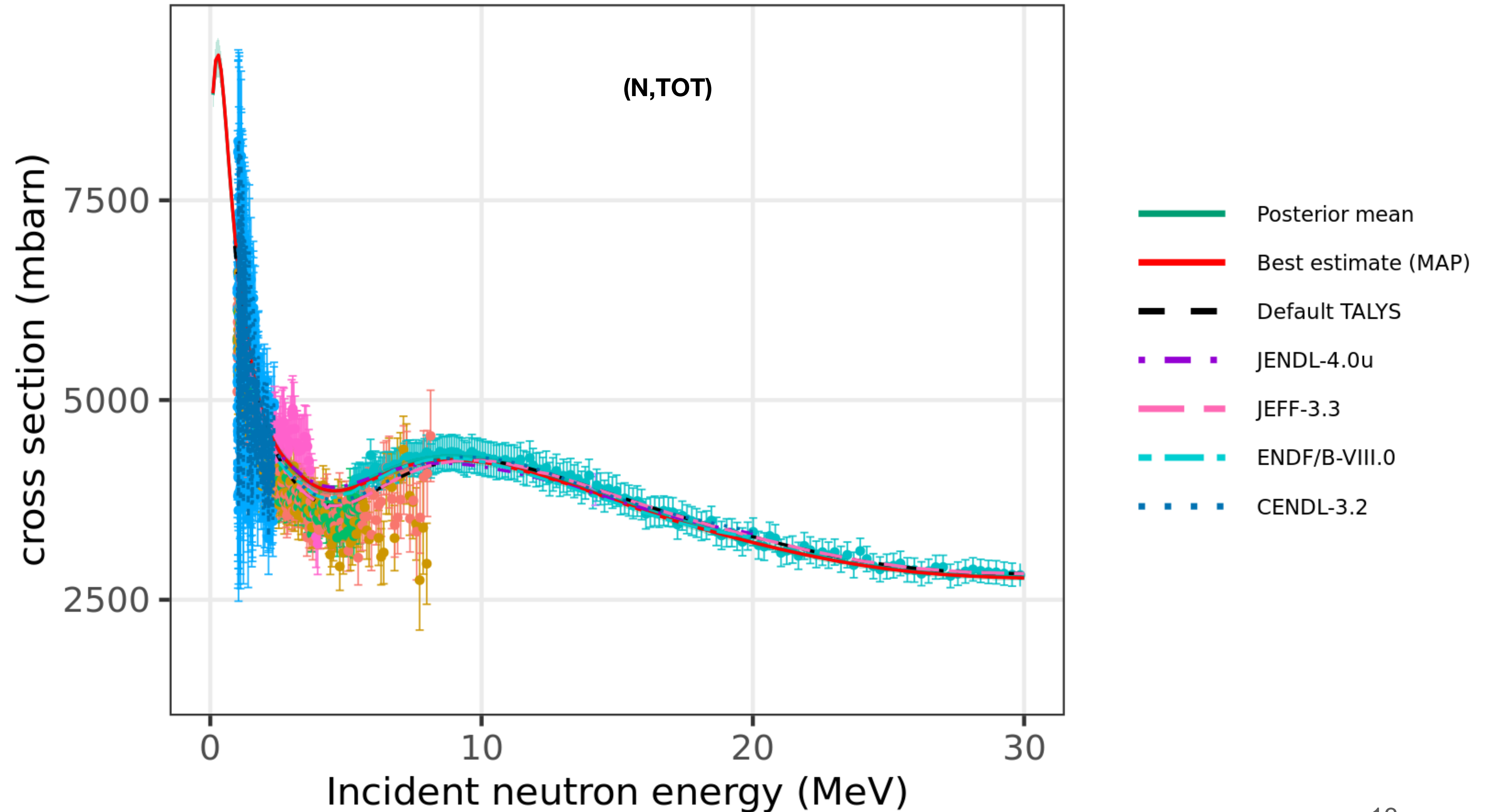
Experiment: $y = 5000$ mb, $p = 10\% \rightarrow \sigma = 500$ mb
TALYS reference: $f = 1000$ mb
→ after PPP fix: $\sigma = 100$ mb
Uncertainty reduced by factor 5.

In resonance regions and in outliers

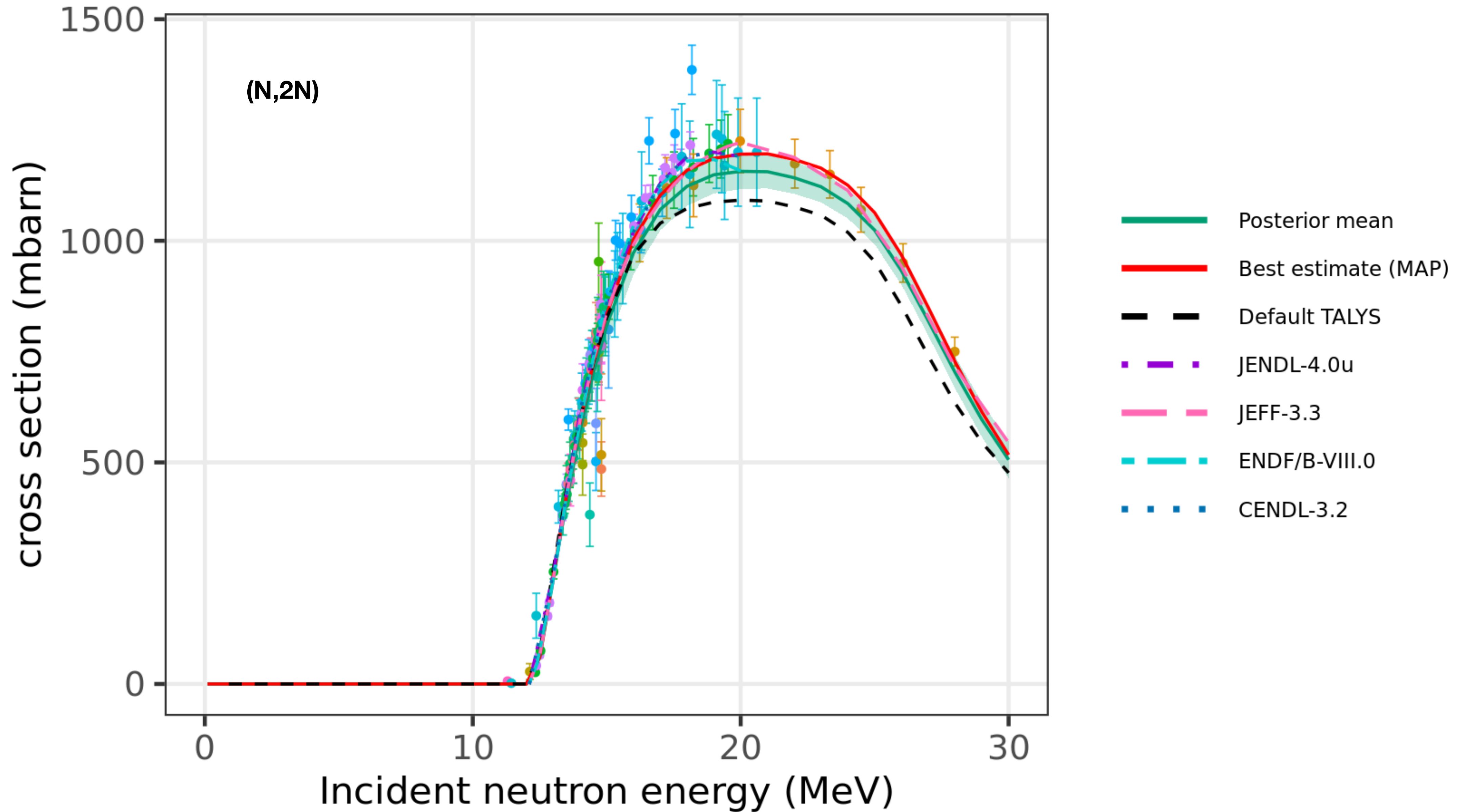
- Outlier uncertainty is actually a representation of disbelief.
- The smooth reference f lacks the resonance structure that the data sees. $y \gg f$, the uncertainty gets underestimated.

PPP renormalization is valid only when the model is a good estimate of the true value. In resonance regions, and for outliers, it may create a problem.

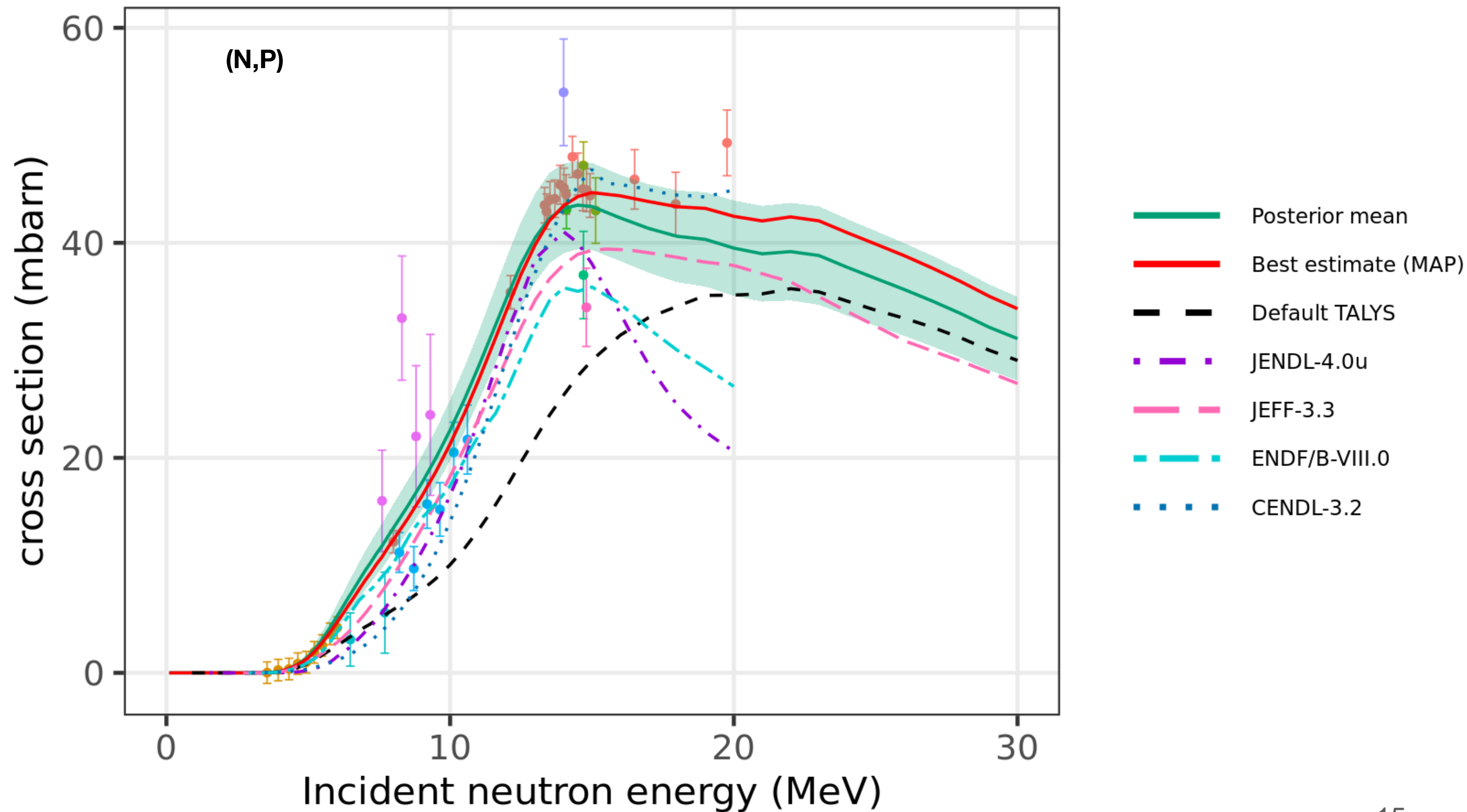
Uncertainty quantification of $\sigma_{(N,TOT)}$ for Zr-90



Uncertainty quantification of $\sigma_{(N,2N)}$ for Zr-90



Uncertainty quantification of $\sigma_{(N,P)}$ for Zr-90



Summary

- Development of Nuclear data Evaluation pipeline of Uppsala (NEPU)
 - ✓ Based around the TALYS
 - ✓ Treatment of inconsistent experiment
 - ✓ Model defects treatment with GP
 - ✓ Generalizability and automatization

We present preliminary investigation of uncertainty in Zr-90.

- **The χ^2 paradox**
When the covariance matrix is dominated by fully correlated systematics, GLS optimizes shape over magnitude. A visually biased fit can be statistically correct given the covariance model assumed.
- **PPP renormalization can introduce new problems when the model does not describe the data.**



<https://github.com/UU-nuclear>

THANK YOU.

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