

Uncertainty and parameter optimisation in high-energy models

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Work done by **J. Hirtz** (Post-doc)

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Plan

- Introduction
- Method
- Results
- Conclusions

Introduction

- The main project → Error assessment and model improvement + associated uncertainty
 - Tool → bayesian statistics(inference)
 - Information → experimental data
- Done within CHANDA
 - Building a method to assess error
- Done within SANDA
 - Building a method to improve the model (INCL) via parameter optimisation
- Next steps (project funded by ANR for 4 years, 2024 → 2027)
 - Combining both
 - Applying to specific cases

Introduction

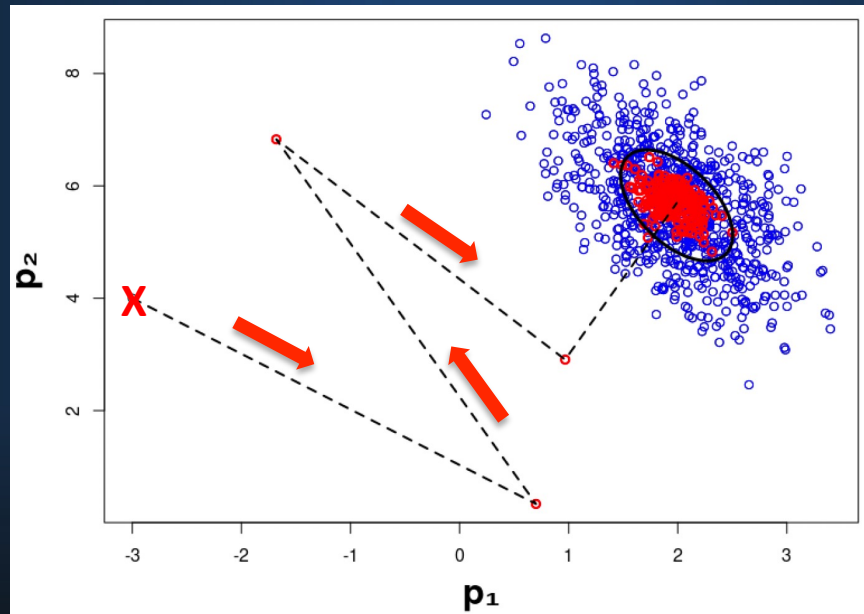
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Method

Parameter optimisation

Two steps

- GLS (Generalized Least Square method) → **Optimized parameters**
- Gibbs-sampling-like method → **Uncertainties of the values**



More details...

Method

Parameter optimisation

GLS

GLS gives the new values of the parameters and related covariance matrices

$$\vec{p}_{op} = \vec{p}_{ref} + \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} (\vec{\sigma}_{exp} - \mathcal{M}(\vec{p}_{ref}))$$

$$\Sigma_{op} = \Sigma_p - \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} J_p \Sigma_p$$

Method

Parameter optimisation

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reference

link between
parameter and
model

weight of the
information

information

$$\Sigma_{op} = \Sigma_p - \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} J_p \Sigma_p$$

Method

Parameter optimisation

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But

1- Right for linear model \rightarrow Approximation: $\mathcal{M}(\vec{p}_{ref}) \rightarrow \mathcal{M}(\vec{p}_i) + J_{p_i} (\vec{p}_{ref} - \vec{p}_i)$

Jacobian

2- So, Covariance matrix not the right one to get uncertainties

~~$$\Sigma_{op} = \Sigma_p - \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} J_p \Sigma_p$$~~

Method

Parameter optimisation

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Steps

(INCL is not linear, so p_{op} from iterations)

$$\vec{p}_{i+1} = \vec{p}_{ref} + \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} (\vec{\sigma}_{exp} - T_i(\vec{p}_{ref}))$$

with

$$T_i(\vec{p}_{ref}) = \mathcal{M}(\vec{p}_i) + J_{p_i}(\vec{p}_{ref} - \vec{p}_i)$$

Method

Parameter optimisation

Gibbs-sampling-like

Similar process as GLS, But...

$$T_i(\vec{p}_{ref}) = \mathcal{M}(\vec{p}_i) + \mathcal{J}_{p_i}(\vec{p}_{ref} - \vec{p}_i)$$

No more from

$$\vec{p}_{i+1} = \vec{p}_{ref} + \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} (\vec{\sigma}_{exp} - T_i(\vec{p}_{ref}))$$

But from

$$\vec{p}_{i+1} = \mathcal{N}(\vec{p}_i, \Sigma_i) \quad \leftarrow \text{--- (break the linear approximation)}$$

and with

$$\Sigma_{i+1} = \Sigma_p - \Sigma_p J_p^T (\Sigma_e + J_p \Sigma_p J_p^T)^{-1} J_p \Sigma_p$$

Results

Two examples - two cases

- Model needs improvement  Subthreshold production of K^+
- Model already not bad  Double differential neutron cross section

Results

Subthreshold production of K⁺

Description

- K⁺ production from p+A (E~1GeV)
- A few parameters involved 😊
- Very rare event ($\sigma \sim \text{nb}$) + experimental data for each set (target, E) 😞
→ long calculations
- Parameters:
 - $\sigma(\text{NN} \rightarrow \text{K}^+ + \text{X}) \rightarrow \mathbf{a_{NN}} \times \sigma(\text{NN} \rightarrow \text{K}^+ + \text{X})$
 - $\sigma(\pi\text{N} \rightarrow \text{K}^+ + \text{X}) \rightarrow \mathbf{a_{\pi N}} \times \sigma(\pi\text{N} \rightarrow \text{K}^+ + \text{X})$
 - $\sigma(\Delta\text{N} \rightarrow \text{K}^+ + \text{X}) \rightarrow \mathbf{a_{\Delta N}} \times \sigma(\Delta\text{N} \rightarrow \text{K}^+ + \text{X})$
 - Fermi momentum (270 MeV/c) → **Fm**
- Time: 7 days with 20 cores

Results

Subthreshold production of K+

Results

Name	old	new
a_{NN}	1.0 +/- 0.1	1.5
$a_{\pi N}$	1.0 +/- 0.1	0.26
$a_{\Delta N}$	1.0 +/- 0.1	0.43
F_m	270. +/- 5.	232.

! Exercice! !

New values, acceptable?
consequences on other quantities?

A complete study must go beyond!

χ^2 was bad

- . delicate mechanisms;
- . only 3% for exp. uncertainties...

And now...

- . not « wonderful »,
- . but falls rapidly!

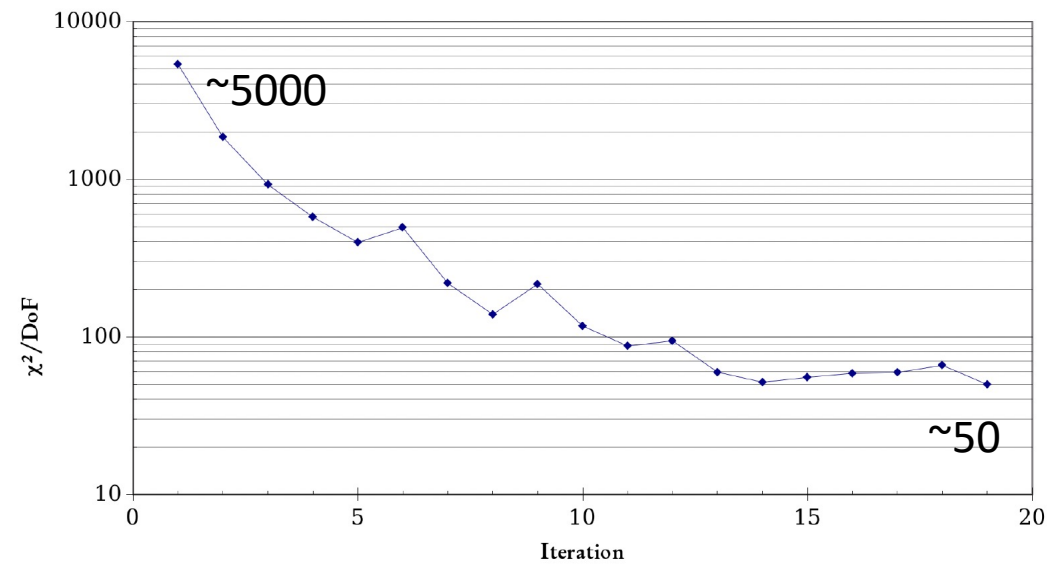


Fig. 2 Figure of merit showing the evolution of the χ^2/DoF after each iteration. Iteration 1 corresponds to the initial version of INCL.

Results

Subthreshold production of K⁺

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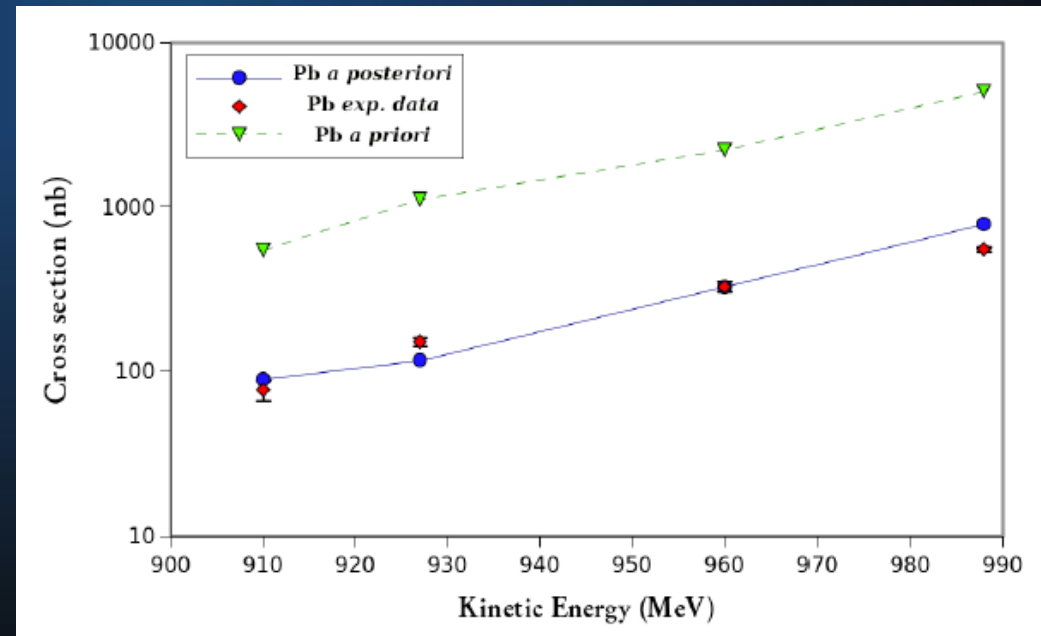
χ^2 was bad

- . delicate mechanisms;
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And now...

- . not « wonderful »,
- . but falls rapidly!

And great improvement for Pb target



Results

Double differential neutron cross section

Description

- Neutron DDXS from p+A $A < \text{Aluminum}; E_p < 200 \text{ MeV}$
- *Toxic* data (error bar: 0.45%!!!) → exp. data must be studied before 😞
- Are we able to do better, for an already not too bad DDXS? 🤔
- Parameters:
 - $\sigma(N\Delta \rightarrow NN) \propto 3 \sigma(NN \rightarrow N\Delta)$ → $\sigma(N\Delta \rightarrow NN) \propto \mathbf{DB} \times \sigma(NN \rightarrow N\Delta)$
 - $t_{\text{stop}} = 29.8 \times A^{0.16} \text{ fm/c}$ → $t_{\text{stop}} = \mathbf{a} \times A^{\mathbf{b}} \text{ fm/c}$
 - Fermi momentum (270 MeV/c) → **Fm**
- Time: 60 hours with 20 cores

Results

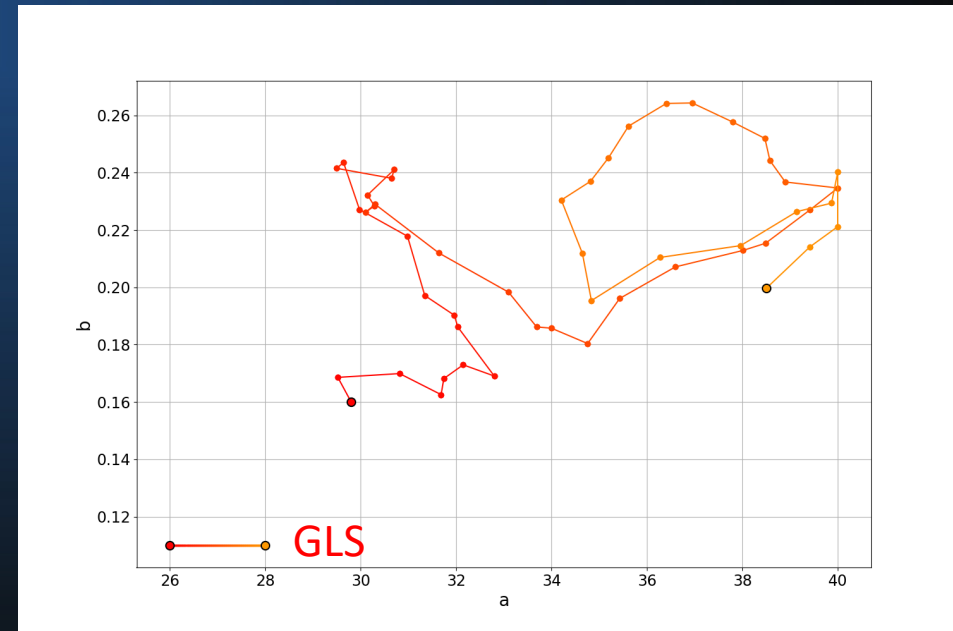
Double differential neutron cross section

Results

Name	old	new
DB	3.0 +/- 0.1	4.406 +/- 0.131
a	29.8 +/- 0.5	37.13 +/- 0.59
b	0.16 +/- 0.05	0.226 +/- 0.005
Fm	270. +/- 3.	266.4 +/- 0.97

χ^2/DoF 7.805 \rightarrow 7.34
(6% better)

GLS gives the values
... and uncertainties, but not really usable (linearity)



**Uncertainty and parameter optimisation
in high-energy models**

Results

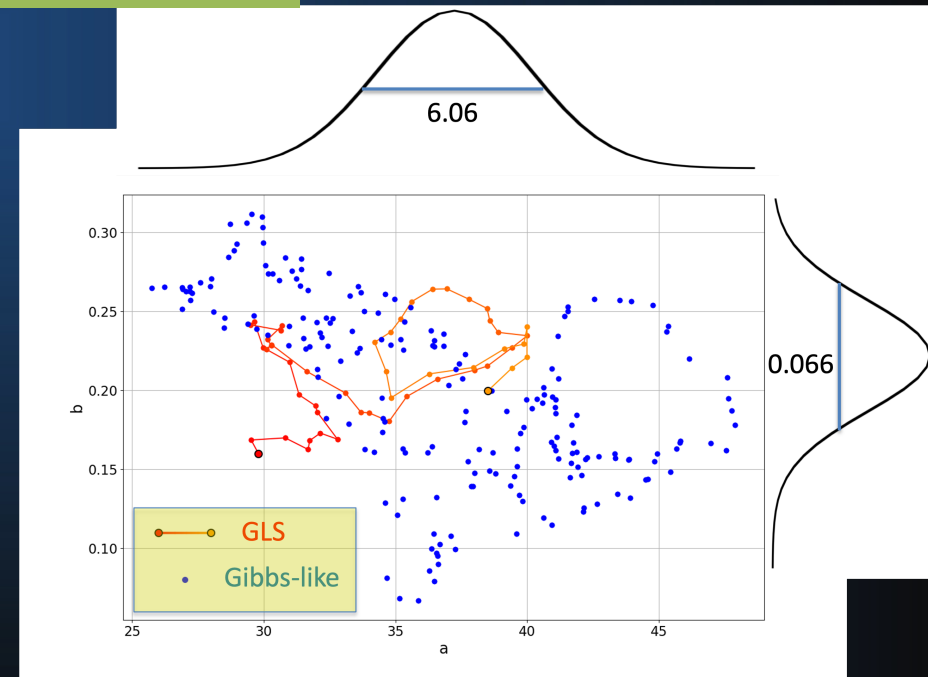
Double differential neutron cross section

Results

Name	old	new
DB	3.0 +/- 0.1	4.406 +/- 0.131 +/- 0.986
a	29.8 +/- 0.5	37.13 +/- 0.59 +/- 6.06
b	0.16 +/- 0.05	0.226 +/- 0.005 +/- 0.066
Fm	270. +/- 3.	266.4 +/- 0.97 +/- 8.82

GLS gives the values
... and uncertainties, but not really usable (linearity)

Gibbs-like, distributions of the possible values
Fit by a gaussian → uncertainties



Conclusions

- The method to optimize parameters of Monte Carlo nuclear reaction models like INCL is built (with associated uncertainty).
- The report has been delivered.
- An article has been accepted in EPJA and will be published soon.

*Thanks for your
attention!*

And thanks to

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in collaboration with
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