



60 Years

IAEA

Atoms for Peace and Development

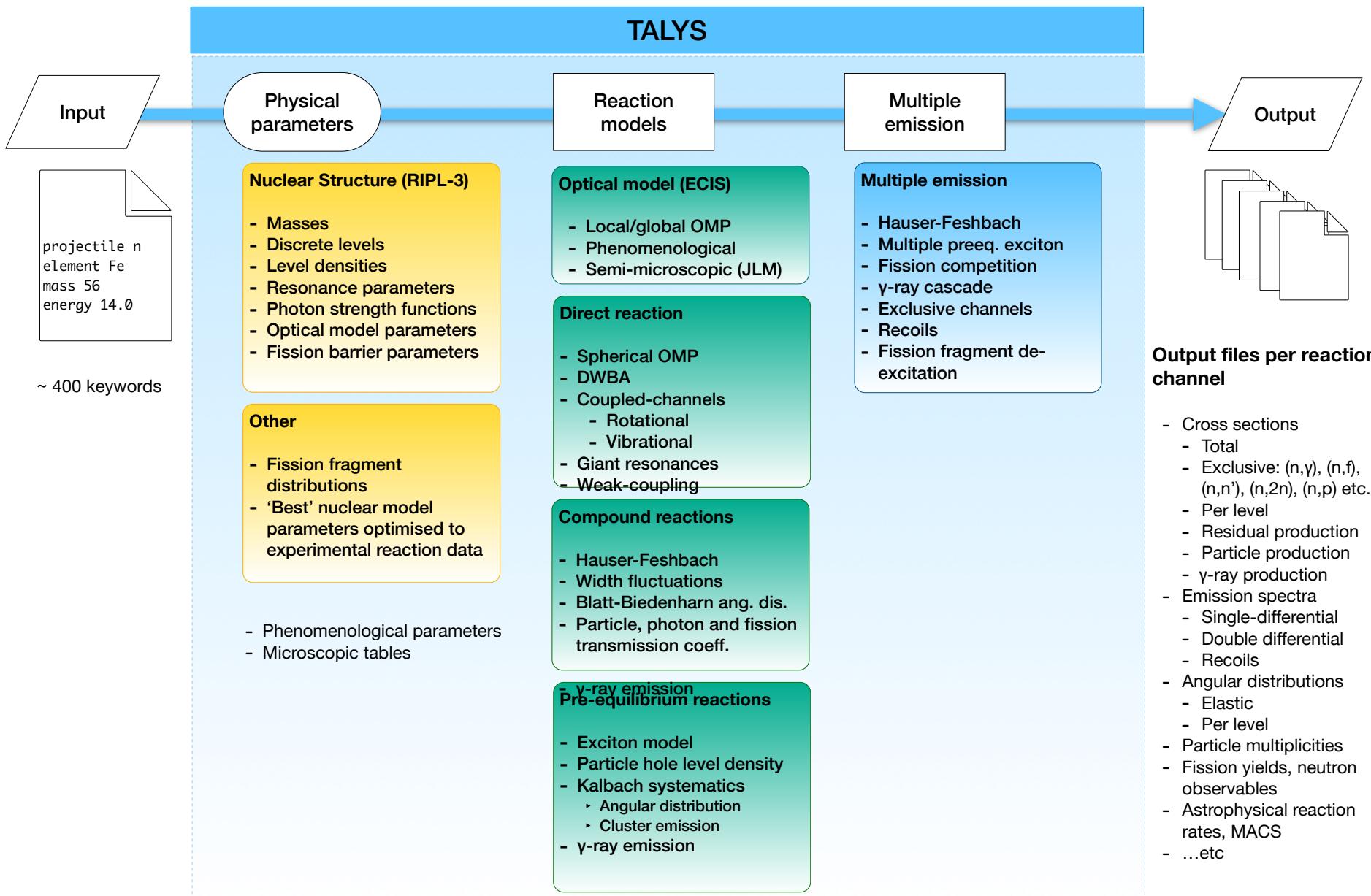
Nuclear model parameter optimisation with TALYS

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NDS library, November 3 2022

Introduction

- TALYS covers
 - All nuclides
 - All projectiles
 - 0-200 MeV
- 6000-7000 citations in all possible fields: neutron fission, medical isotope production, astrophysics, fusion, activation, etc.
- Objective: make the code globally better, regardless of the application
- Global good prediction: better physics, global optimisation to experimental data
- Global good description: nuclide-by-nuclide optimisation to experimental data



Prediction of reality

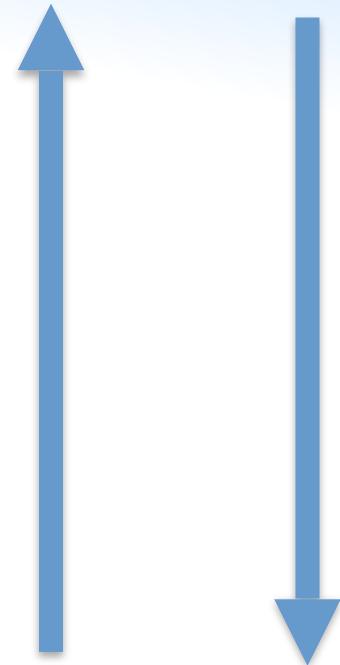
Microscopic global models

Phenomenological global models

Micro/pheno models with adjusted parameters

Micro/pheno models with unphysical parameters
or functions

Direct normalisation of model output to
experimental data

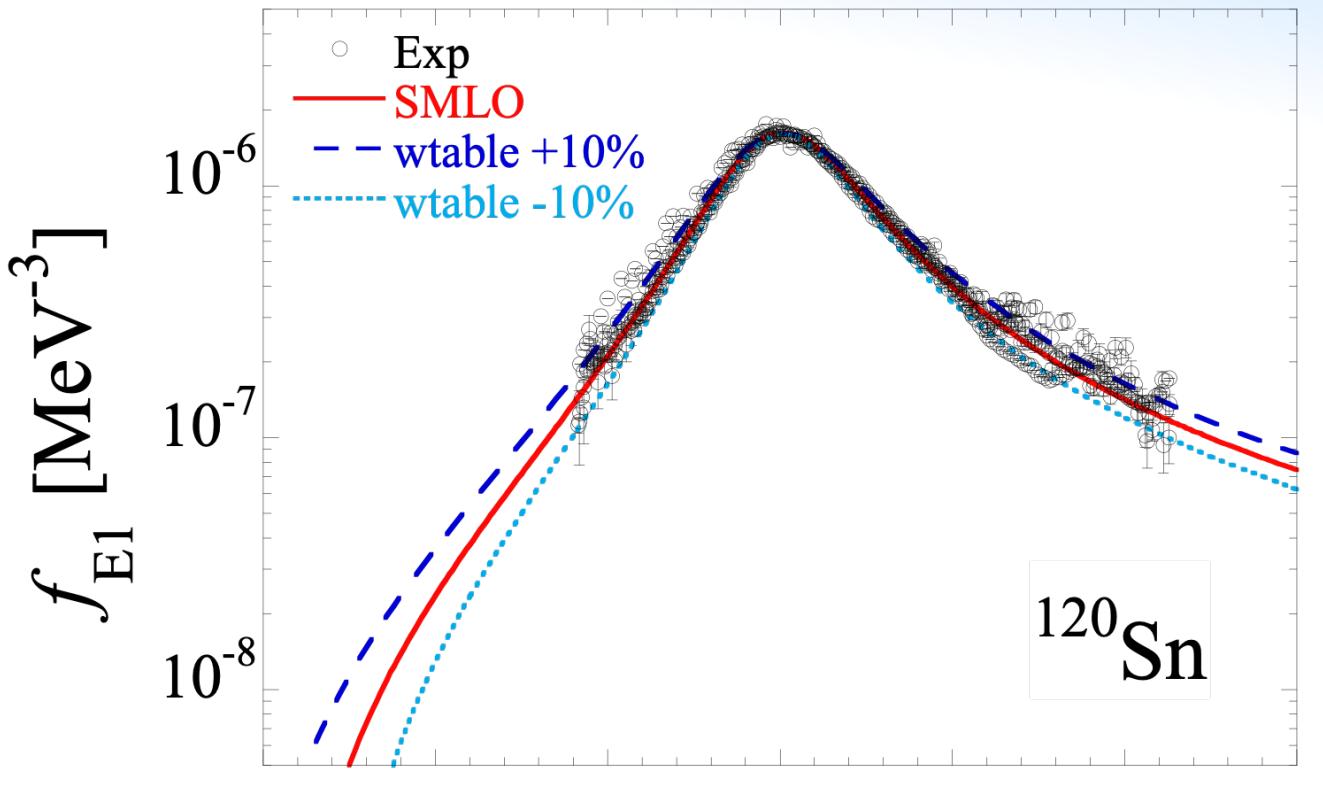


Description of reality
= exp. data

TALYS Application to (n, γ) cross sections

- Adjust width of the E1 SMLO photon strength function (TALYS: ‘wtable’) to match **the best nuclear data library** with
 - Best reproduction of MACS, around 30 keV
 - Best reproduction of experimental (n, γ) cross section from the top of the RR - 100 keV/1 MeV
 - Best nuclear data library **before** TENDL-2021:
 - JENDL-4.0: 97 target isotopes
 - JENDL-AD: 2 target isotopes
 - CENDL-3.2: 2 target isotopes
 - TENDL-2019: 106 target isotopes
 - JEFF-3.3: 8 target isotopes
 - ENDF/B-VIII.0: 61 target isotopes
- **Autotalys** automatically optimizes ‘wtable’ to match **above libraries in a restricted energy range**, e.g. autotalys -element Eu -mass 151 -Lttarget 000 -Liso 0 -proj n -bins 40 -search -energyfile /Users/koning/samples/psf/smlo/ctm/energies -best -noautosearch -noparauto -talysfile /Users/koning/samples/psf/smlo/ctm/talys.add -tasmanfile /Users/koning/samples/psf/smlo/ctm/tasman.add

Adjusted width parameter does not affect original photon strength function very much

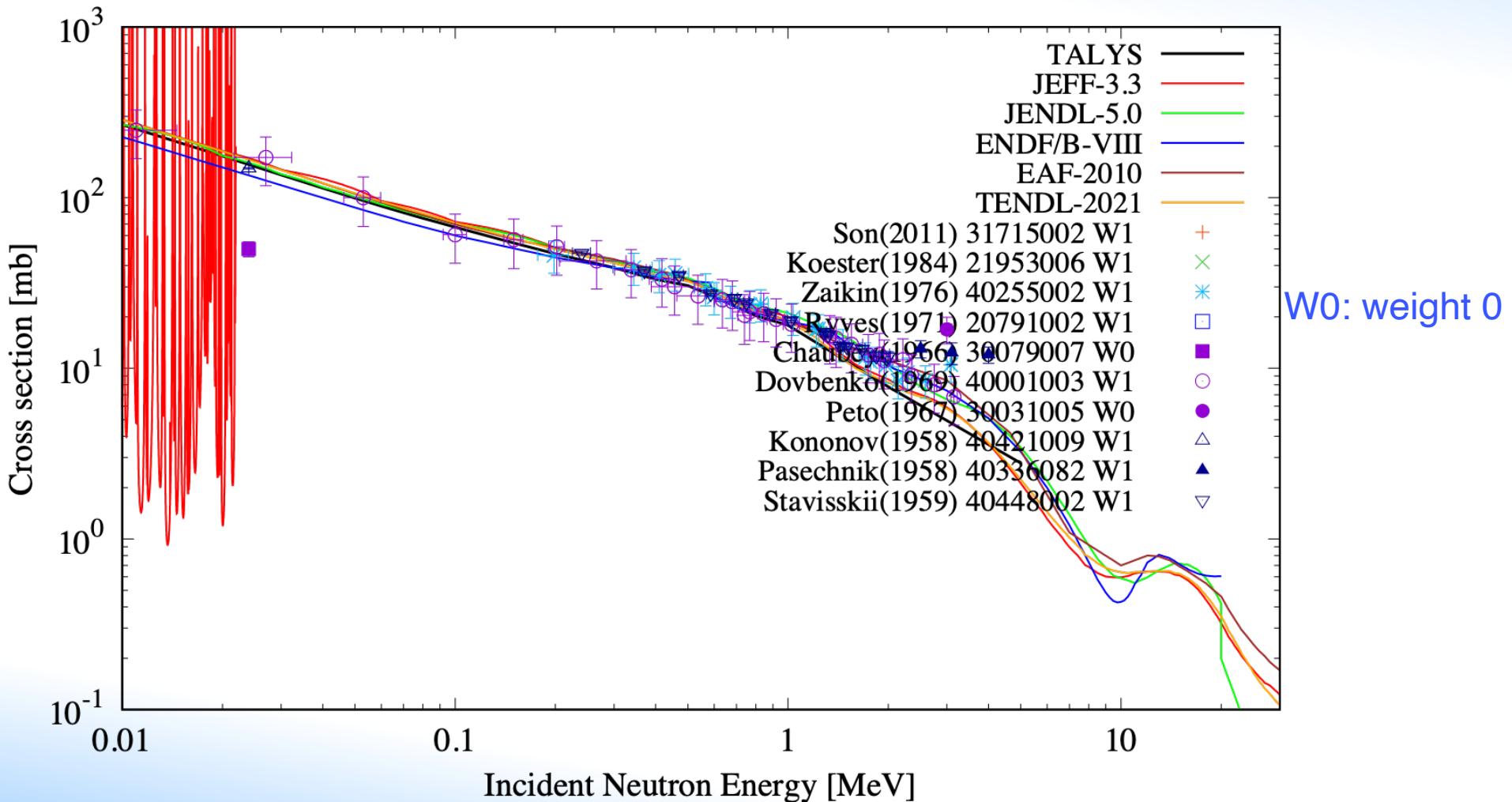


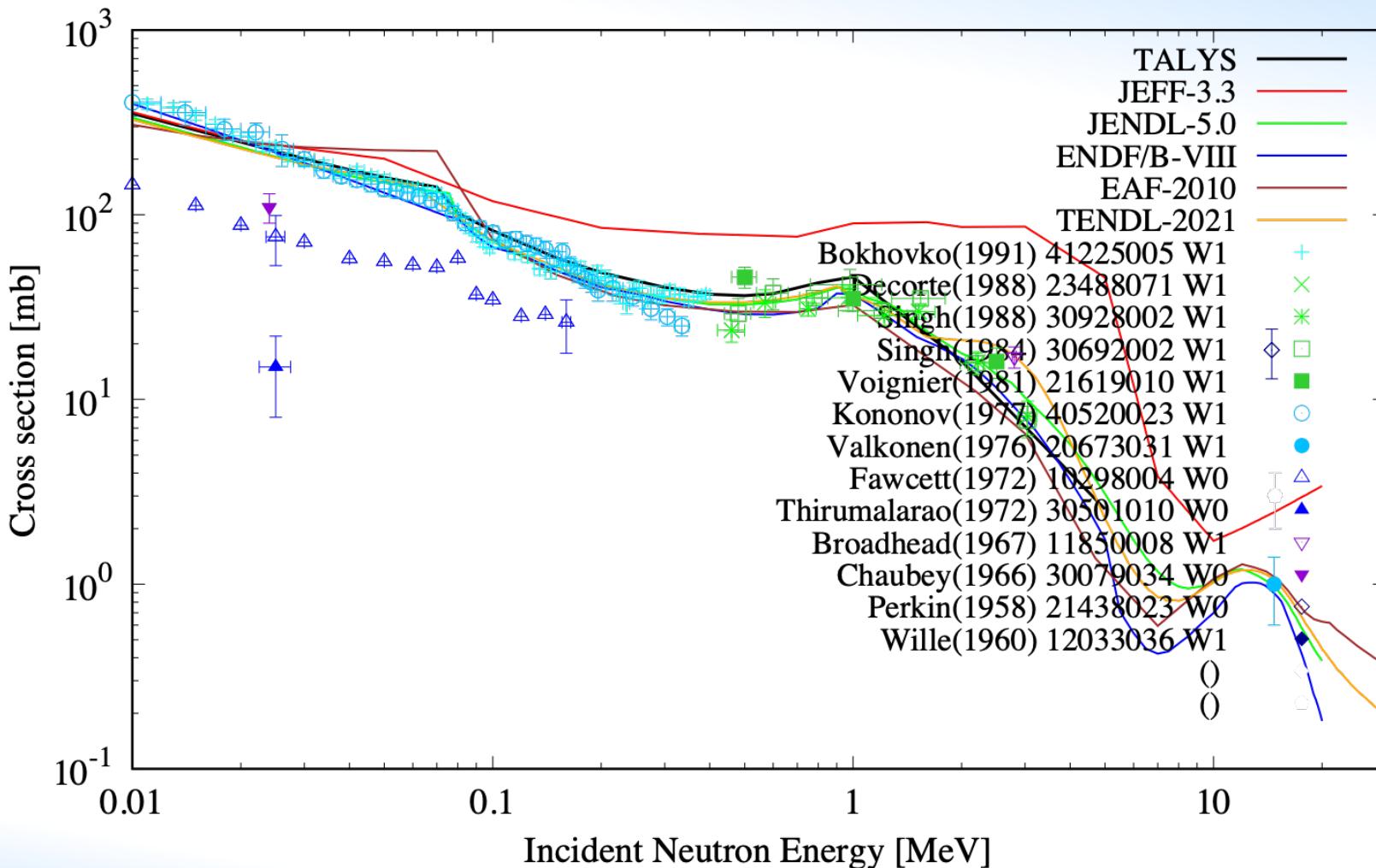
$S_n + E_{in}$: range
important for
(n, γ)

$$\langle \Gamma_\gamma \rangle = \frac{D_0}{2\pi} \sum_{X,L,J,\pi} \int_0^{S_n + E_n} T_{XL}(\varepsilon_\gamma) \times \rho(S_n + E_n - \varepsilon_\gamma, J, \pi) d\varepsilon_\gamma$$

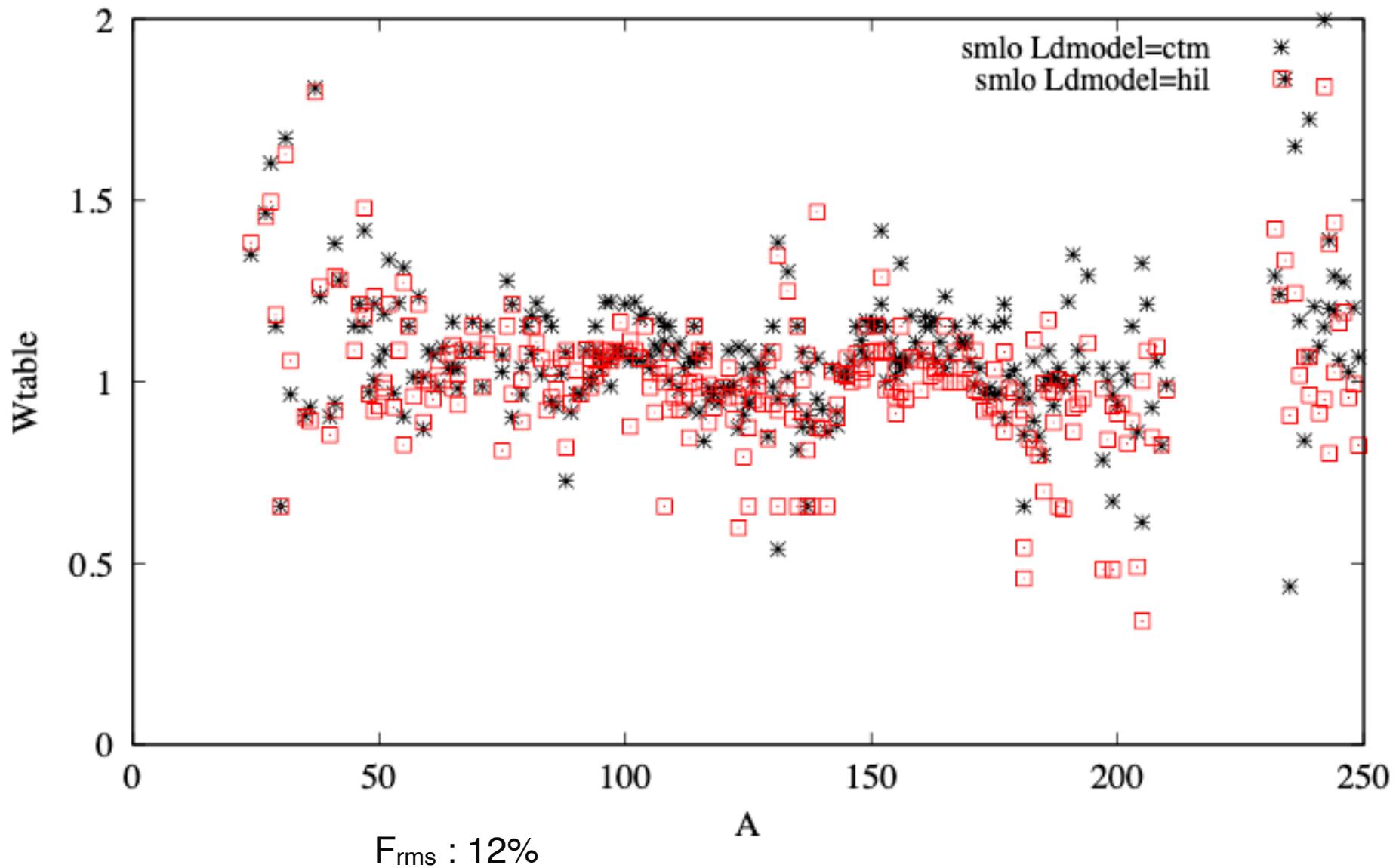
Optimization to included exp. data

$^{69}\text{Ga}(\text{n},\gamma)^{70}\text{Ga}$ GOF = 1.037



$^{160}\text{Gd}(\text{n},\gamma)^{161}\text{Gd}$ GOF= 1.094


Wtable for (n,g) with exp. MACS, fitted to best library



TALYS optimisation to EXFOR

- Two birds with one stone:
 - **Exforcism** (flag evil data in EXFOR)
 - Add to current database of experimental ‘outliers’ by Alhassan, Gaughan and Dzysiuk
 - ‘Graphical’ outliers
 - Full JSON files available
 - Optimization of TALYS parameters to all cross sections
 - Use of XC5 format of EXFOR (V. Zerkin)
 - Dimension reduction
 - Same parameter set for all nuclides

8400 JSON outlier/inlier files, one per EXFOR subentry

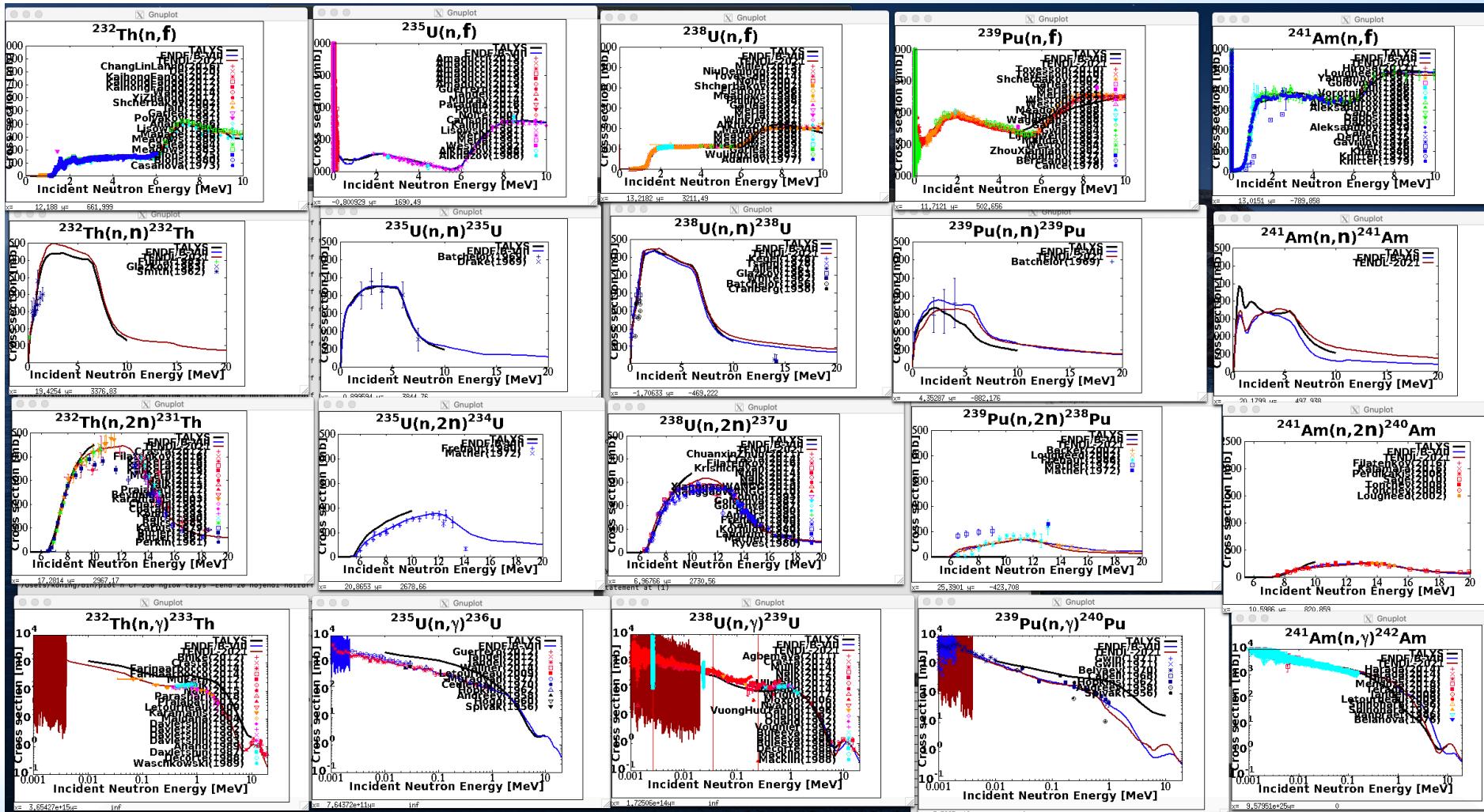
```
"Subentry"      : "A0001004",
"Author"        : "Skakun",
"Year"          : 1975,
"Projectile"    : "p",
"Target Z"      : 48,
"Target A"      : 111,
"Target state": "0",
"X4 Reaction" : "48-CD-111(P,N)49-IN-111,,SIG",
"Evaluations"  :
[
  {
    "Evaluator"    : "Arjan Koning",
    "Date"         : "2022-06-05",
    "Weight"       : 0,
    "Comment"      : [
      " Excluded from evaluation: graphical outlier"
    ]
  },
  {
    "Evaluator"    : "Erwin Alhassan",
    "Date"         : "2019-11-08",
    "Weight"       : 0,
    "Comment"      : [
      " Erwin Alhassan (PSI, 2018) 0",
      " (1 -> accept and 0 -> reject)",
      " Reasons for inclusion/exclusion",
      " 1) Experimental data set not consistent with other experiments such as Takacs (2005) between about 10 - 15
MeV (The cross sections are systematically lower)"
    ]
  },
  {
    "Evaluator"    : "Natalie Gaughan",
    "Date"         : "2019-03-15",
    "Weight"       : 1,
    "Comment"      : [
      " IAEA-TECDOC-1211 - Data selected"
    ]
  }
]
```

TALYS parameters for optimization

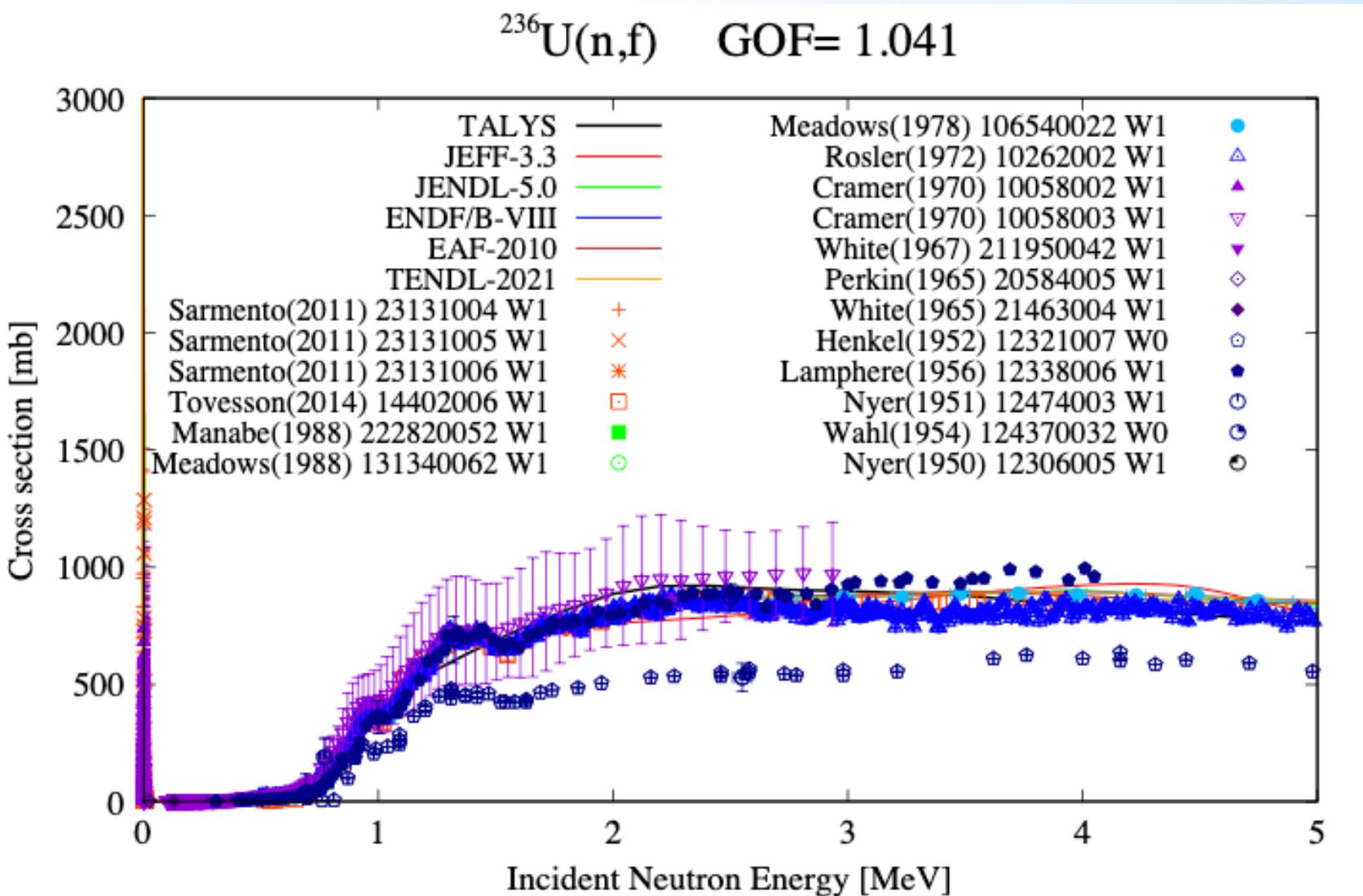
Reaction	Nuclides	Parameter	Parameter	Parameter	Parameter	Parameter	Parameter
(n, γ)	278	wtable					
(n,f)	34	vfiscor	betafiscor	ctable(1)	ptable(1)	ctable(2)	ptable(2)
(n,n'), (n,2n), (n,p)	210	rv(p)	g _{ph} (0)	g _{ph} (n)	ctable(n)	ctable(p)	
(N,α)	157	rv(α)	Cstrip(α)	g _{ph} (0)	ctable(α)		
(p,n)	142	rv(p)	rwd(p)	rv(n)	g _{ph} (0)	g _{ph} (n)	ctable(n)
(γ,n)	77	wtable	ftable	etable			
(α,n)	93	rv(α)	rwd(α)	rv(n)	g _{ph} (0)	ctable(α)	
(d,n)	40	rv(p)	rwd(p)	rv(n)	g _{ph} (0)	g _{ph} (n)	ctable(n)

TASMAN code (AK): Nelder-Mead optimisation.
 Number of TALYS trials: N(parameters) × 20

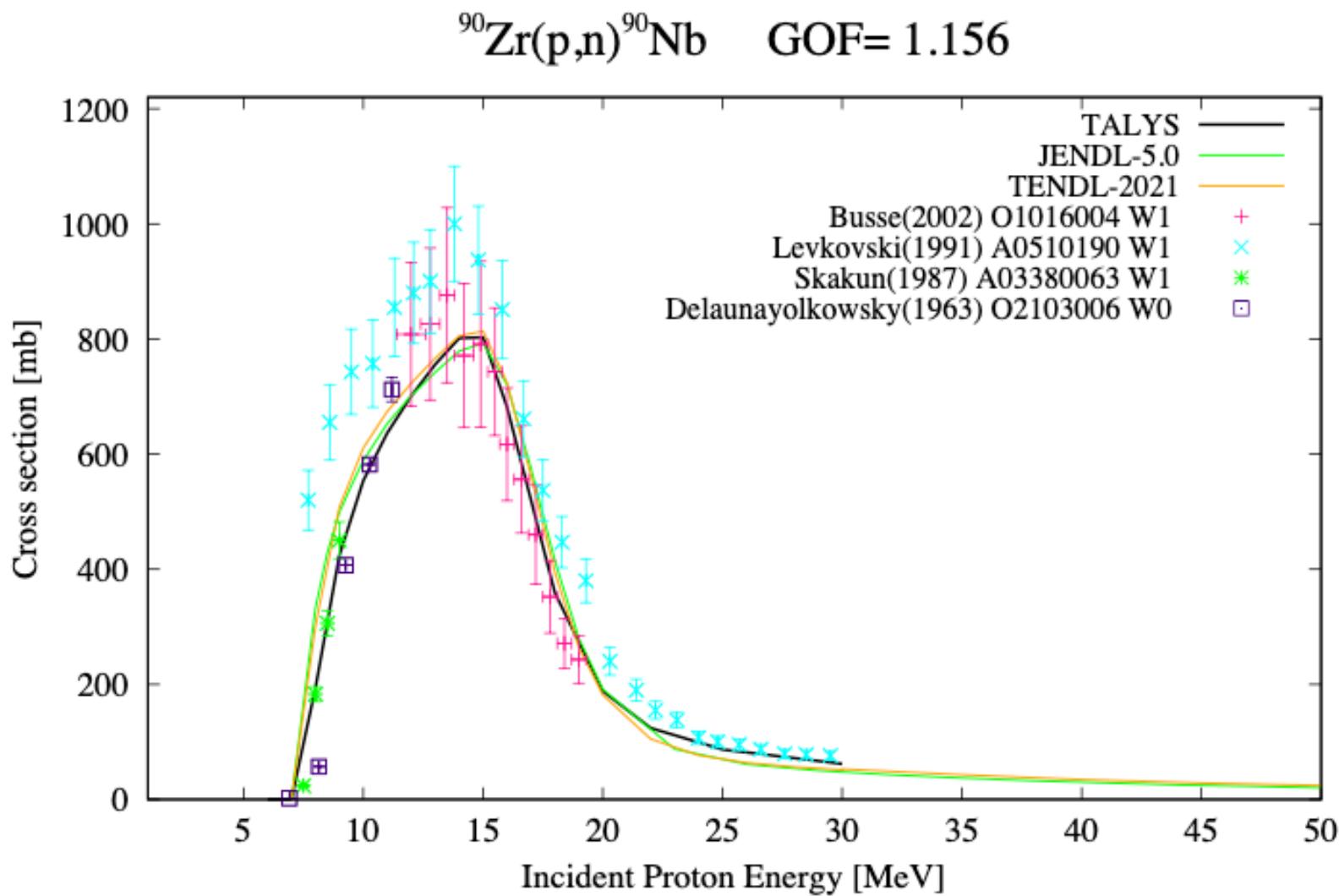
Fitting all actinides at the same time



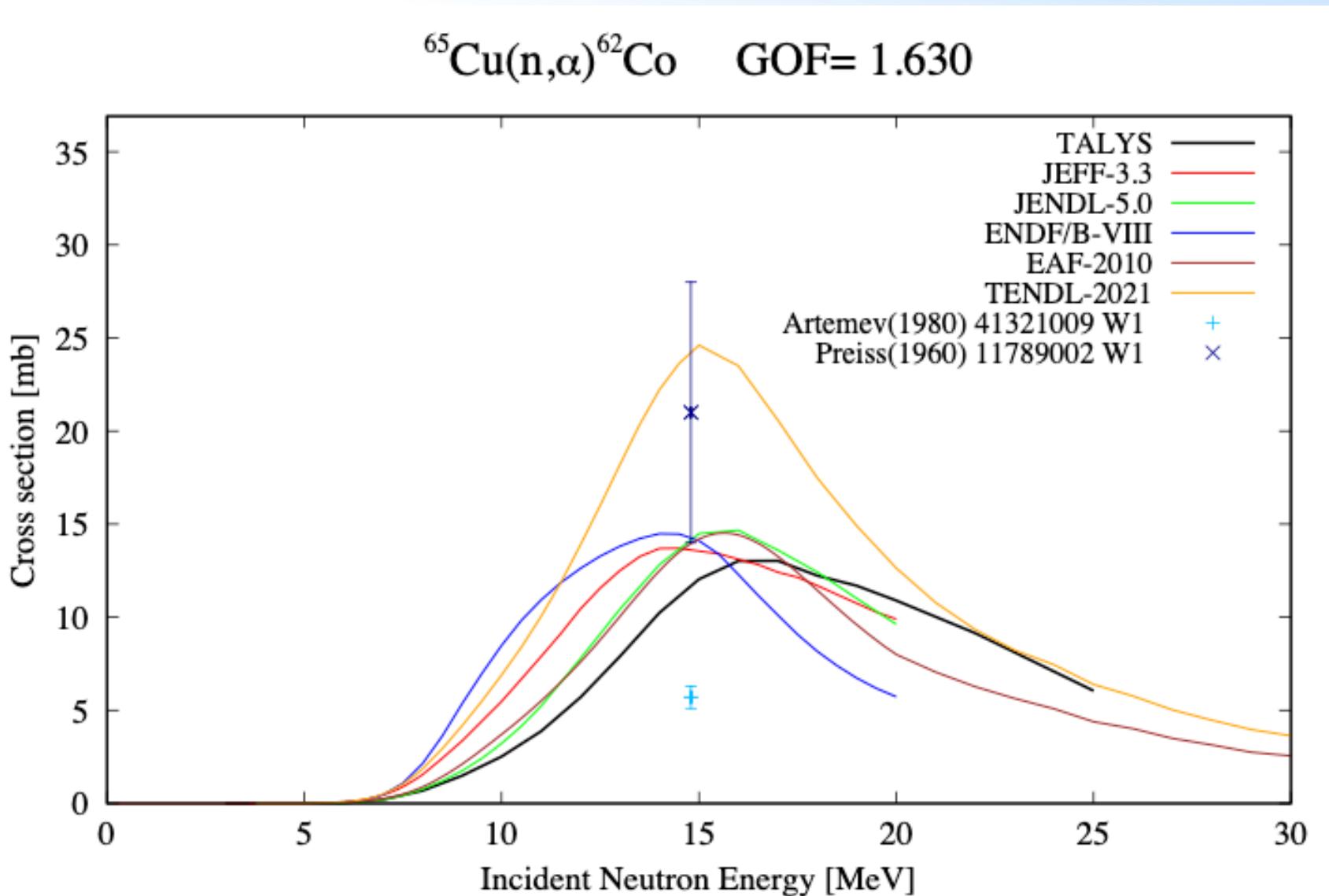
Fission: WKB (Sin, Capote) + HFB level densities (Hilaire, Goriely)



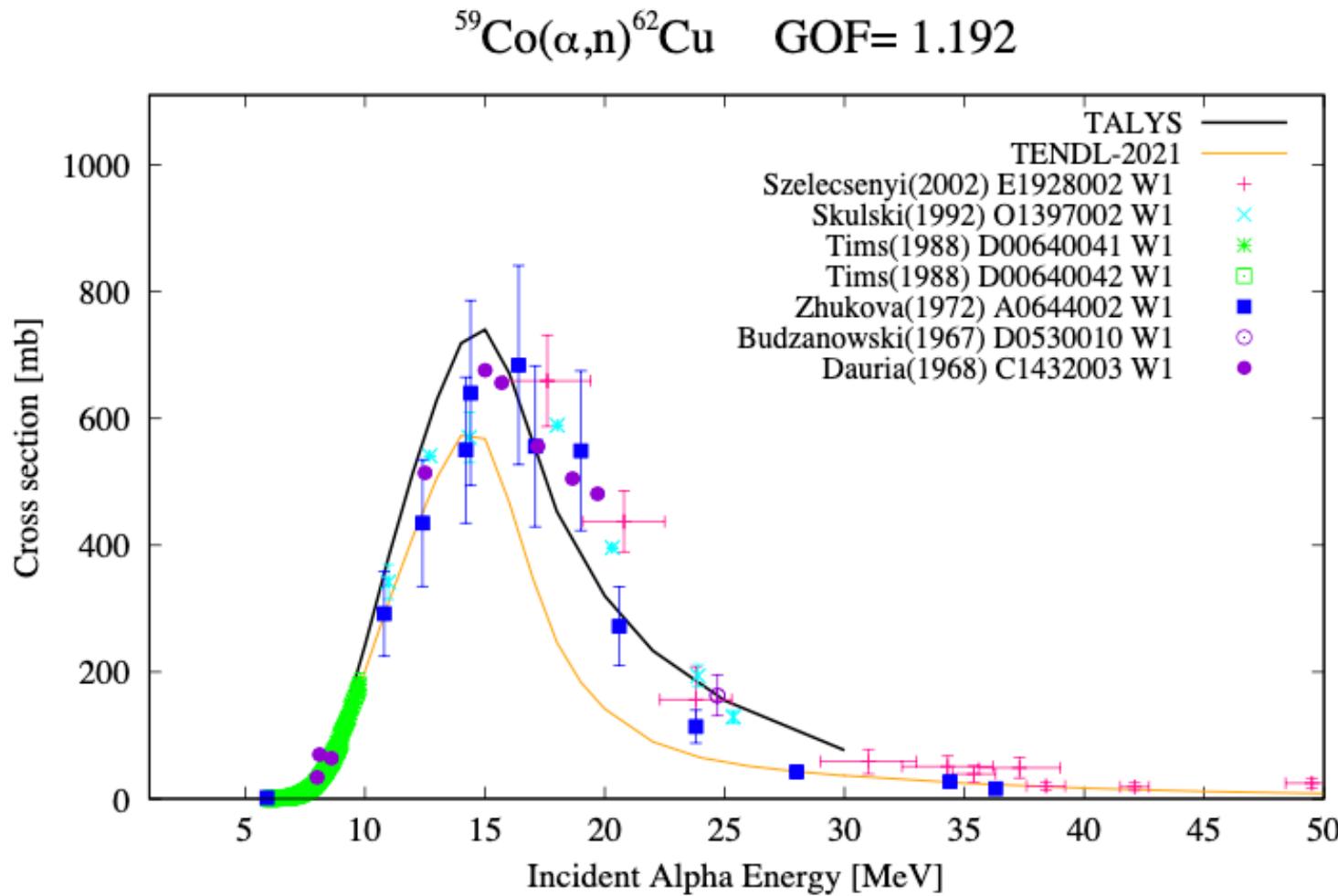
(p,n): several nuclides with JENDL-5 evaluation



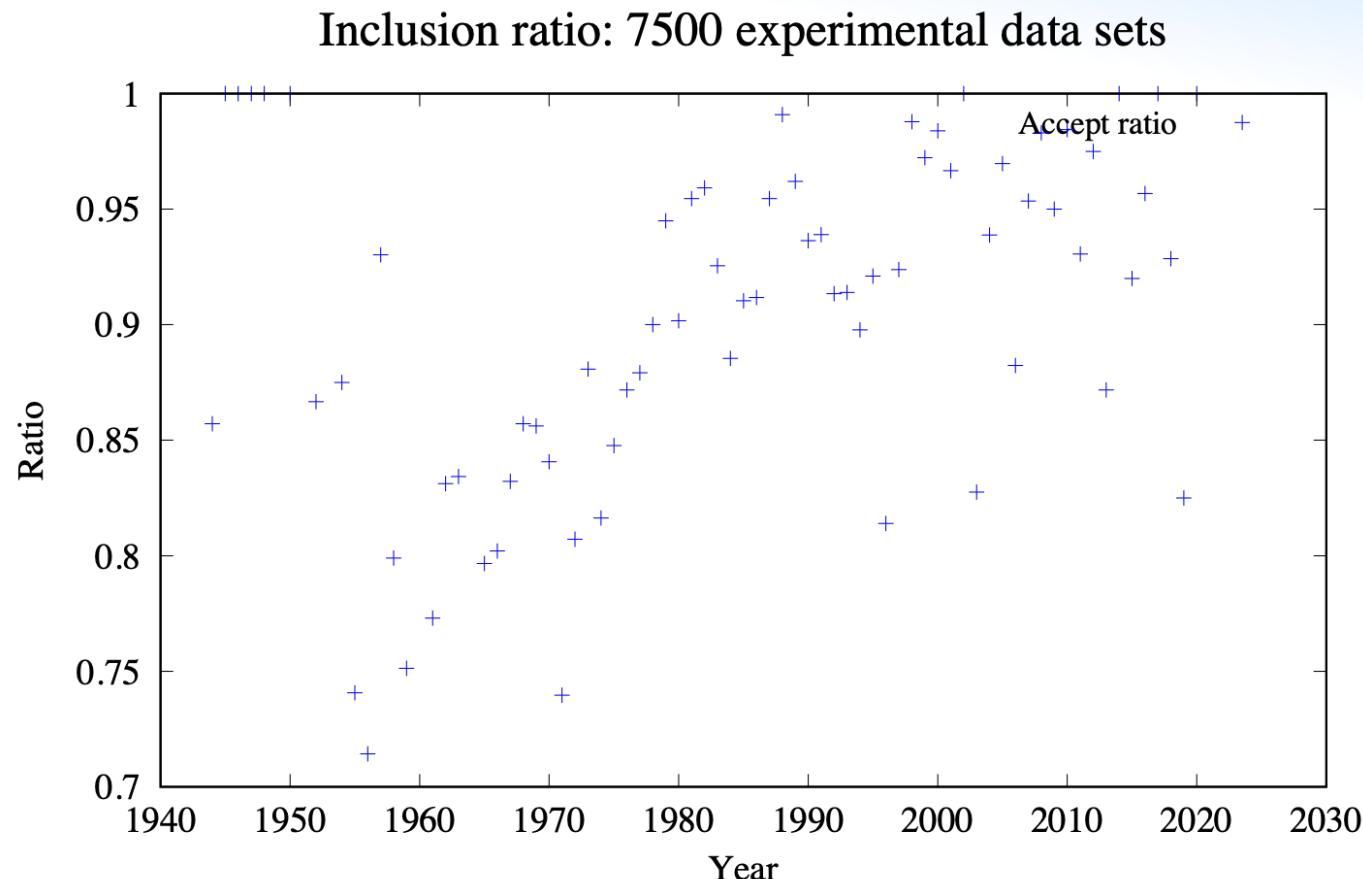
Segal's law: A man with a watch knows what time it is.
A man with two watches is never sure.



(α ,n): TENDL rules! (OK OK, because there is nothing else for A > 20)

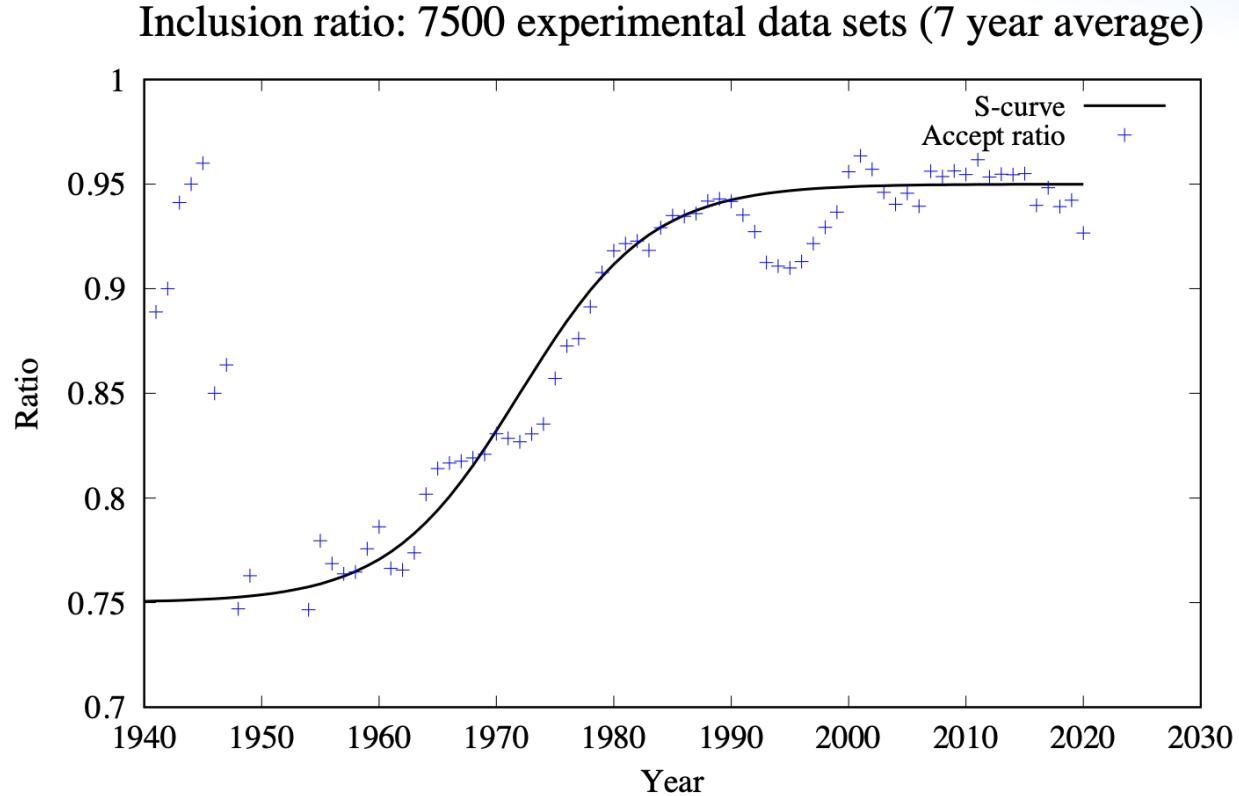


Outlier assignment



Summed over all (n,g), (n,f), (n,n'), (n,2n), (n,p), (n,a), (p,n), (g,n), (a,n), (d,n) reactions we could mine from EXFOR. 6500 accepts, 1000 rejects

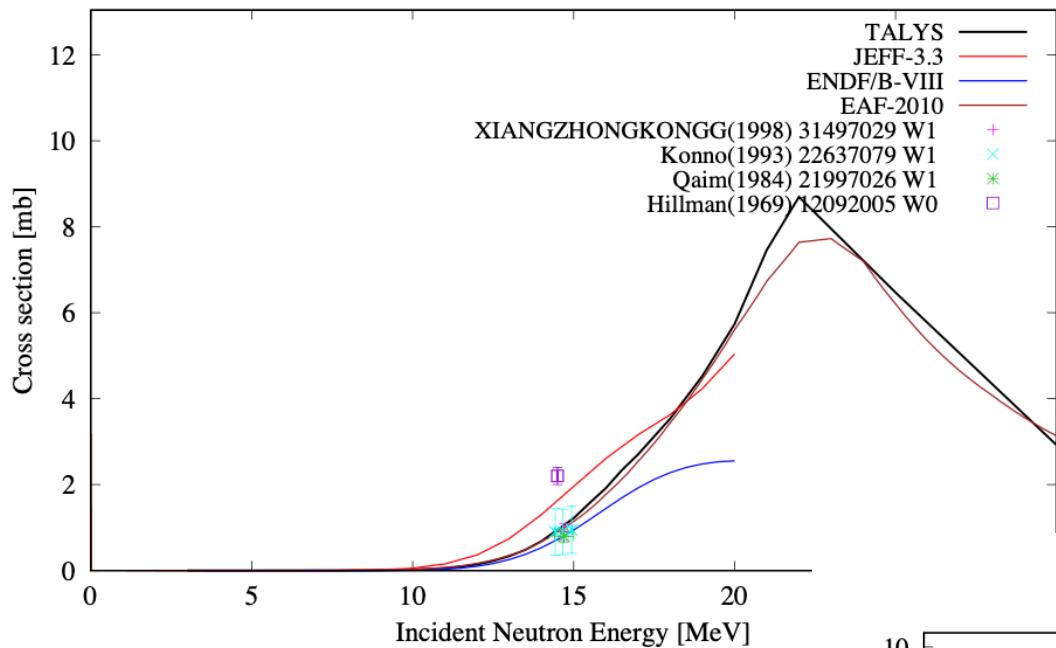
Outlier assignment: a learning curve?



Other analyses possible:

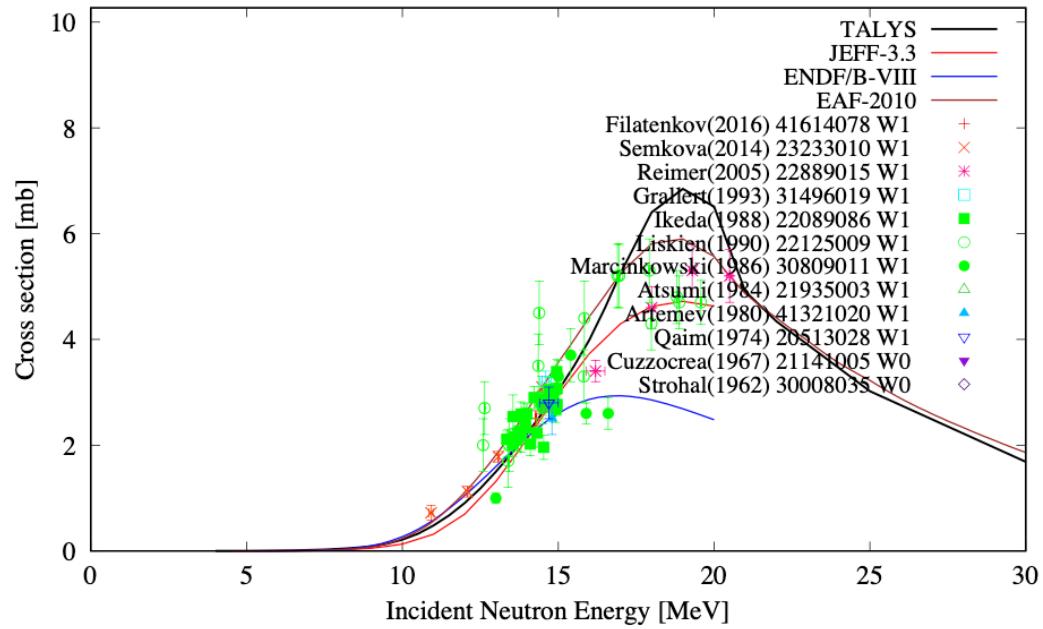
- per reaction channel
- per author, co-author, lab, etc
- per incident energy (e.g. 14 MeV)
- re-insert this as prior in the next Bayesian update

$^{180}\text{Hf}(\text{n},\alpha)^{177}\text{Yb}$ GOF = 1.014

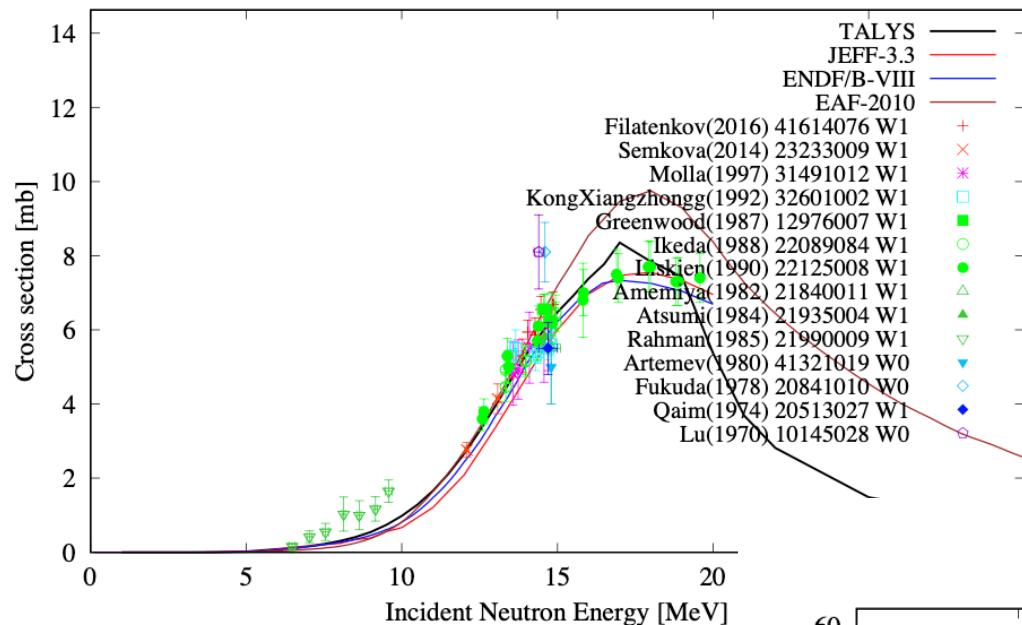


rv(α) Cstrip(α) g_{ph}(0) ctable(α)

$^{100}\text{Mo}(\text{n},\alpha)^{97}\text{Zr}$ GOF = 1.136

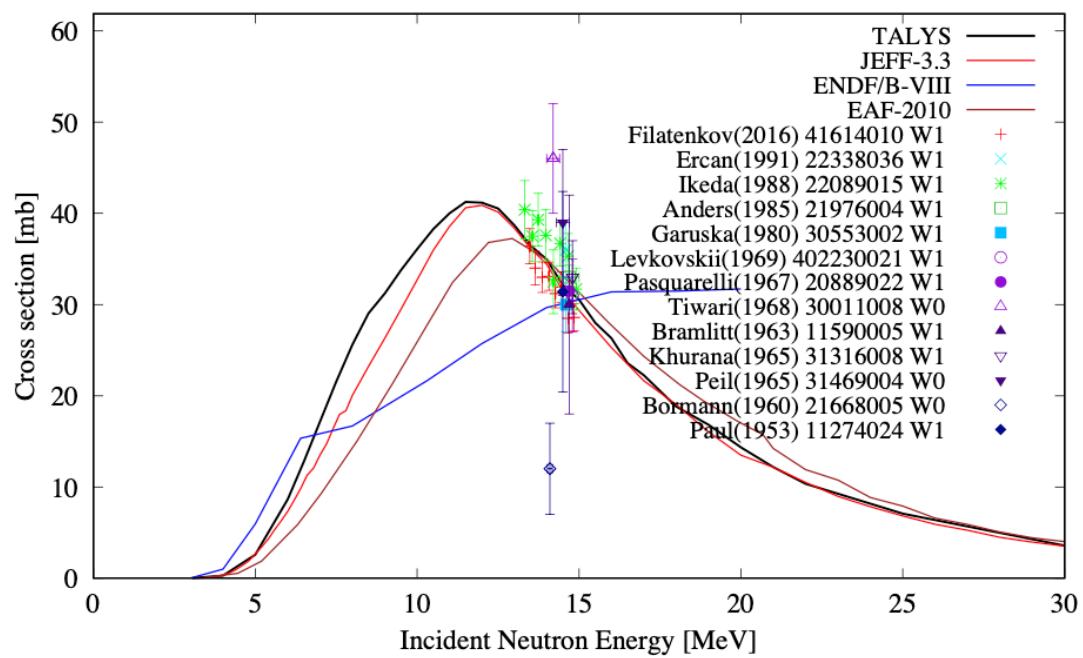


$^{98}\text{Mo}(\text{n},\alpha)^{95}\text{Zr}$ GOF= 1.058



rv(α) Cstrip(α) g_{ph}(0) ctable(α)

$^{41}\text{K}(\text{n},\alpha)^{38}\text{Cl}$ GOF= 1.031



Can ML help here for unmeasured nuclides?

```
#Fitted parameters for (n,a)
#alphaomp    6
#ldmodel    5
#coenhance n
#nuclides   163
# Z   A      rv-a      cstrip     gadjust-0    ctable-a
  4   9      1.00000    1.00000    1.00000    0.00000
  5  10      0.21085    2.21082    1.64359    1.66936
  5  11      0.11639    4.20759    1.67975    2.86544
  6  12      0.48959    0.02787    1.21481   -0.44030
  6  13      1.14567    1.21152    0.28736    1.64602
  7  14      0.92861    0.31371    1.20199   -0.44030
  8  16      1.11293    8.86975    0.84572   -0.92541
  8  17      0.96495    2.43752    1.01593   -1.75192
  8  18      0.22696    7.14760    0.90733    3.71203
  9  19      0.89539    0.81826    1.10131    0.89627
 10  20      0.91654    0.22750    1.38561    0.48796
 10  21      0.95587    1.78427    1.20656   -0.34957
 11  23      0.96817    0.53236    1.12941    0.17828
 12  24      0.15974    2.96245    1.98147   -4.79352
 12  25      0.29923    2.89807    1.63549   -7.02134
 12  26      1.01241    1.33169    1.04872    0.34515
 13  26      1.01873    0.76462    1.02671   -0.22364
 13  27      1.02951    0.57233    1.14429    0.33124
 14  28      0.96387    1.50342    1.25148   -2.66615
 14  29      1.00194    1.35584    0.96969    0.90763
 14  30      0.97181    2.11276    0.89030   -0.29783
 15  31      1.04834    0.63263    1.01261   -0.08361
 16  32      1.02006    0.36600    1.01089   -1.45398
 16  33      0.27283    1.43855    2.22639    0.41495
 16  34      0.97253    5.25031    1.13227   -0.07678
 17  35      0.81829    0.89638    1.17646   -1.32655
 17  36      0.99394    0.87904    1.08989    0.15246
 17  37      0.93518    0.78168    0.83646    1.06129
 18  36      0.45342    0.90863    0.94192    0.69087
 18  40      0.85430    3.04631    1.36767   -0.46891
 19  39      1.04509    1.20812    0.94896   -1.08482
 19  41      1.01116    0.57316    1.04658   -0.24002
 20  40      0.98171    0.22697    1.08368   -0.08800
```

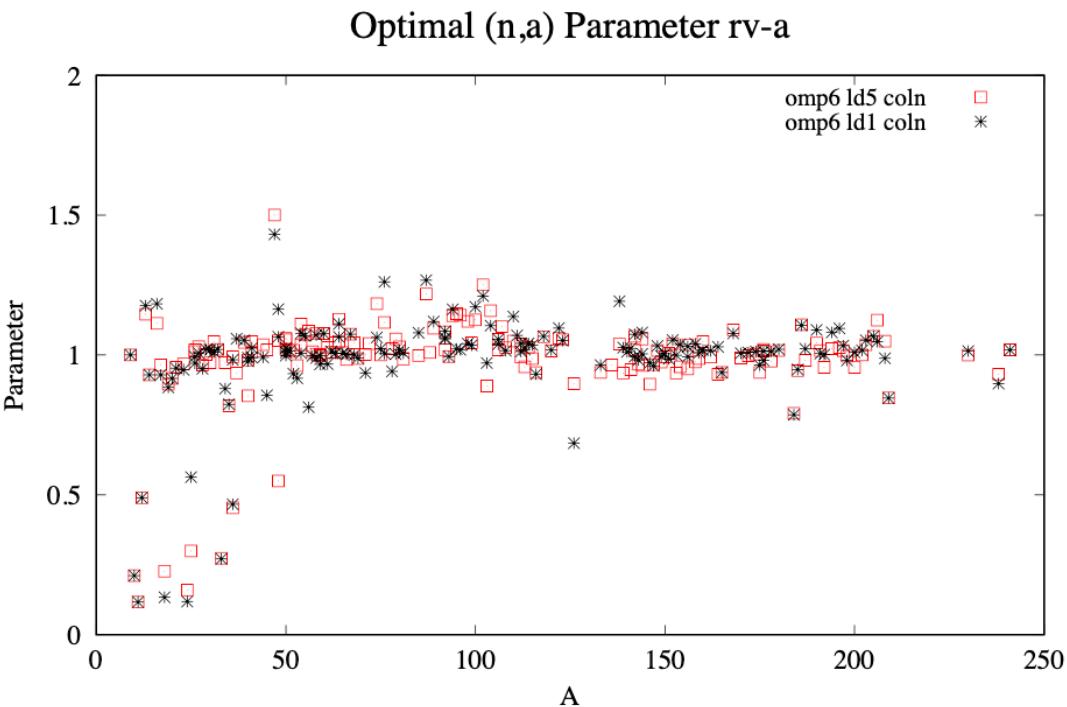
Can we discover trends that a human can not see?

We know that for $A < 20$ the physics of TALYS is not good enough

Are 160 nuclides enough to select a 20% training set?

Can an ML algorithm be predictive for nuclides at or close to the valley of stability?

Optical model parameter rv(a) for 2 different level density models



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Summary

- Trying to put human knowledge and opinions in consistent metadata
- Part of the objective to make TALYS more predictive and more descriptive (zeroing in on the truth)
- Accessible experimental database essential
- “Standard” optimisation with Nelder-Mead done for each reaction and target nuclide: scripts running scripts running search code running TALYS
- Per reaction channel: scattered parameters for up to 5 dimensions, for typically 200 nuclides, looking for trend and parameter guesses for unmeasured nuclides closed to the measured ones



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Thank you!