

β -decay half-lives using the ANN model: Input for the r-process



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The R-Process Nucleosynthesis

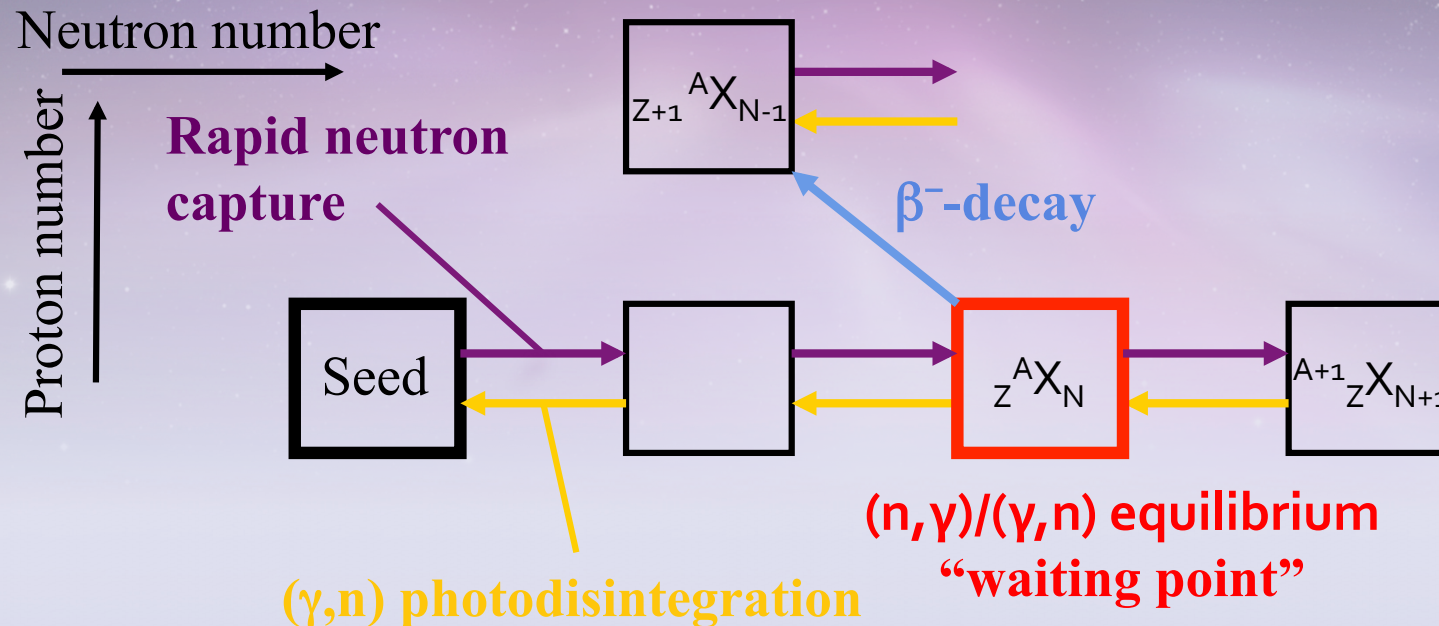
R → Rapid neutron-capture

$T \sim 1\text{-}2 \text{ GK}$, $n_n \sim 10^{20}\text{-}10^{26} \text{ /cm}^3$

Density: 300 g/cm^3 ($\sim 60\%$ neutrons !)

n-capture timescale (τ_n): $\sim 0.2 \text{ } \mu\text{s}$

$\tau_n \ll \tau_{\beta^-}$

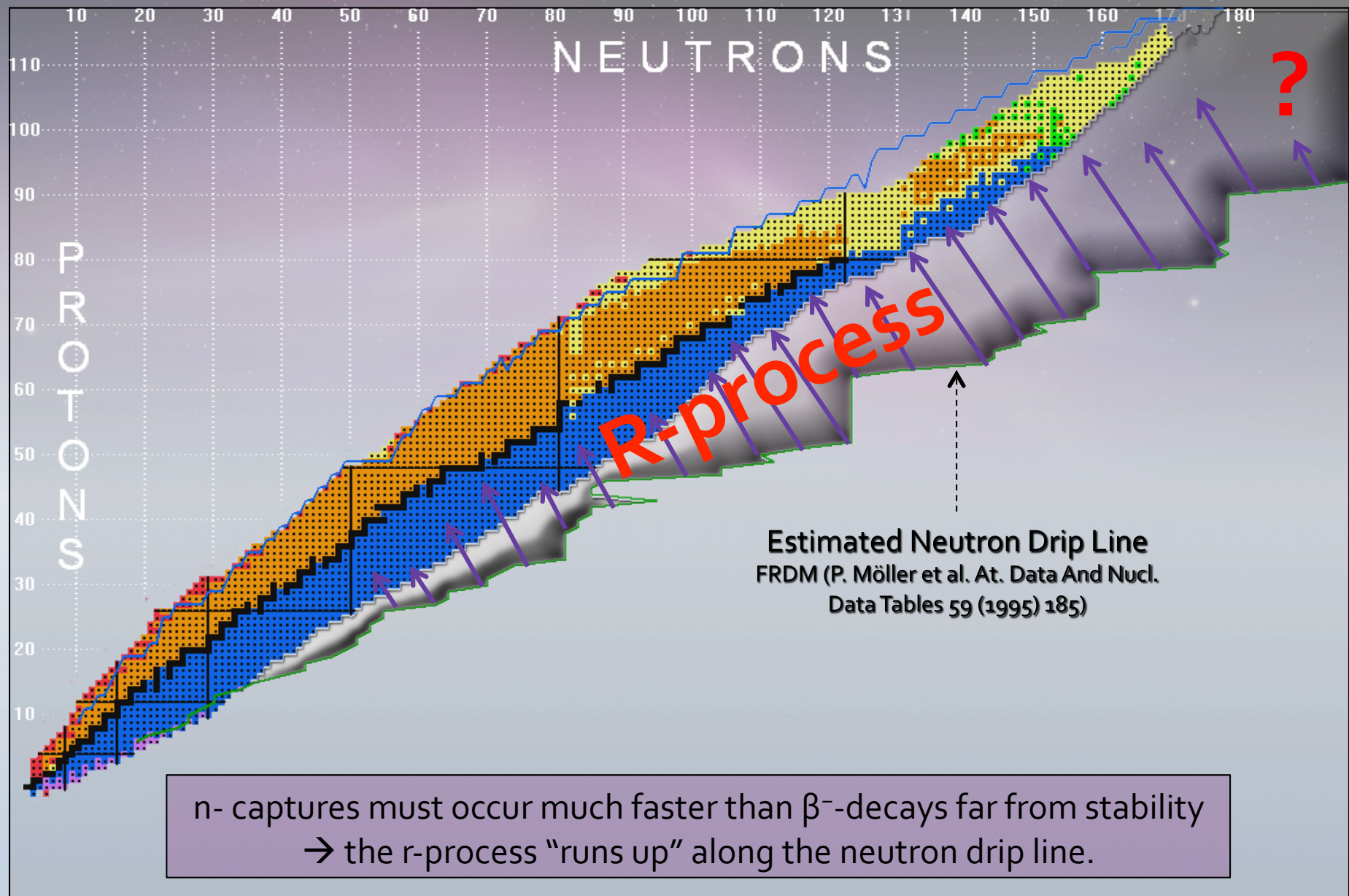


The **rapid n-capture process** (r-process) is today understood to be responsible for the synthesis of about **half of all** the nuclei abundances present in Solar System matter in the mass region from approximately Zinc through the Actinides (e.g., Thorium, Uranium, and Plutonium), as well as the bulk of the heavy elements in the mass region $A \geq 60$ in our Galactic and Cosmos matter.

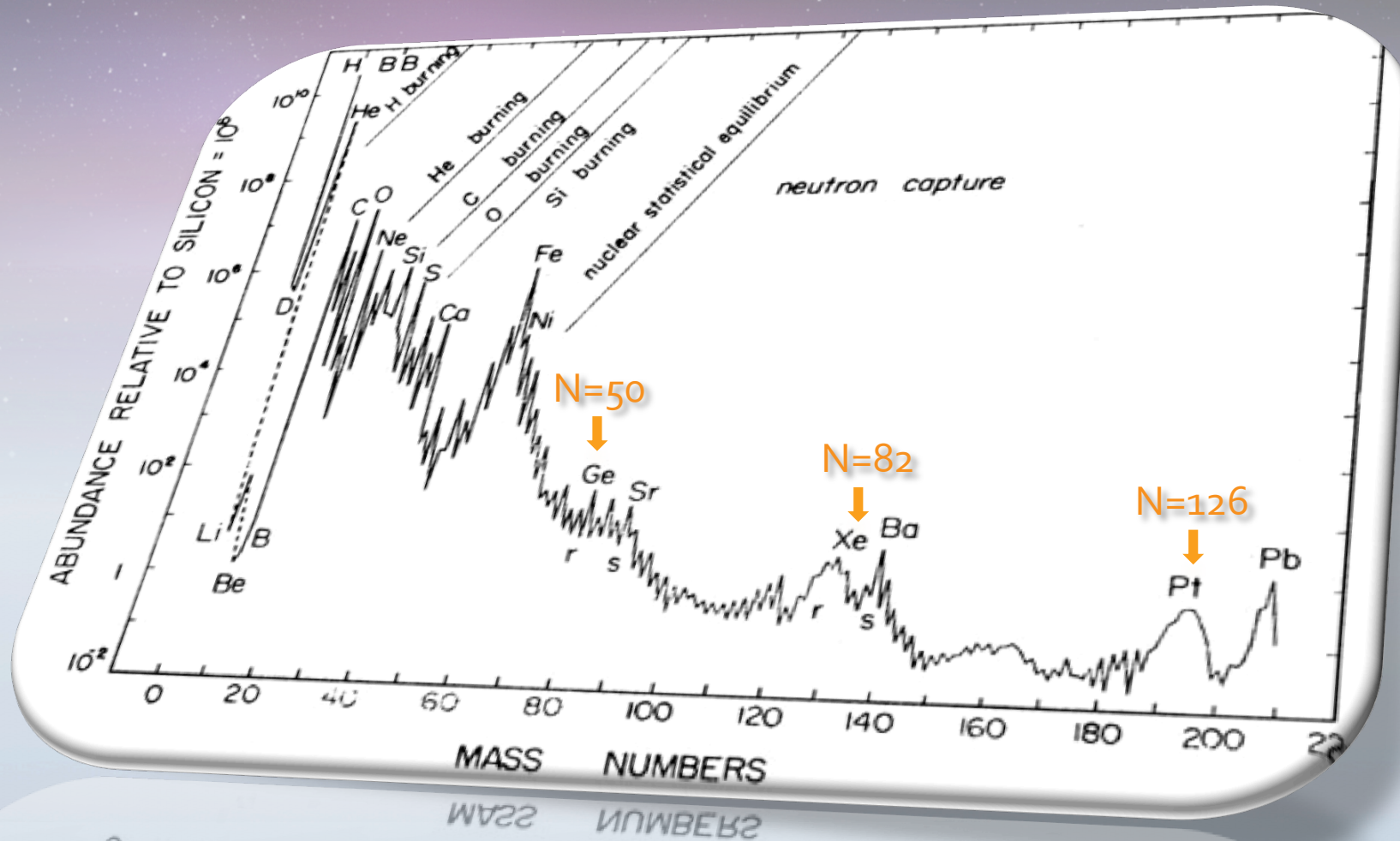
R-process still remains one of the most challenging open questions in modern nuclear astrophysics, since both the astrophysical scenarios where this process occurs and the needed nuclear physics input have not yet been unambiguously identified.

R-Process Path

chart of nuclides



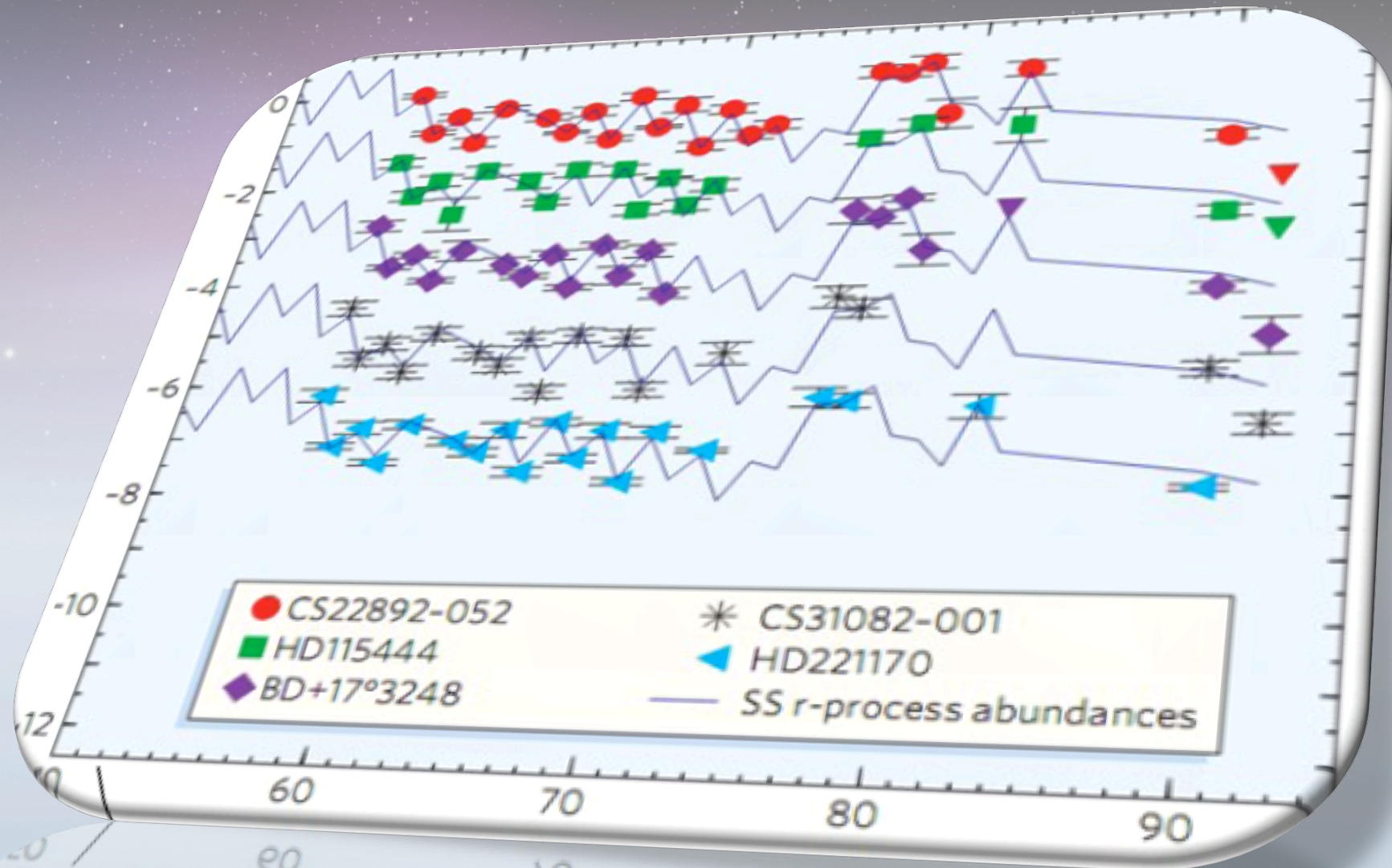
Relative Solar System Abundances



The solar elemental abundance distribution beyond Fe, shows peaks near $A = 80$, 130 , and 195 , corresponding to progenitors with closed neutron shells $N = 50$, 82 , and 126 .

Old Metal Poor Galactic Halo Stars

The r-process operating over the history of the Galaxy



→ R-process is unique in nature!

Understanding of R-Nucleosynthesis

r-abundances reproduction

Astrophysical Conditions

temperature (T_9)

matter density

neutron density (n_n)

...

radiation entropy (S_{rad})

r-process duration (τ_r)

Nuclear Physics Input

nuclear-masses (Q_β , S_n)

β -decay properties ($T_{1/2}$, P_n)

n-capture rates

fission properties

neutrino interactions

nuclear-structure-properties (J^π ...)

for 1000's of isotopes **FAR-OFF** β -stability

Astrophysical Site(s)

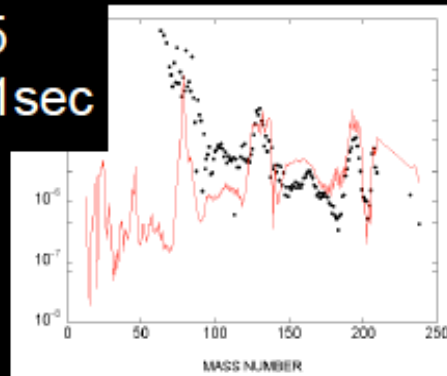
Dominant Candidates

for the "main" r-process

Neutrino-driven wind in SNe II

SN1987A

$S \sim 400$
 $Y_e \sim 0.45$
 $\tau_{\text{dyn}} \sim 0.1 \text{ sec}$

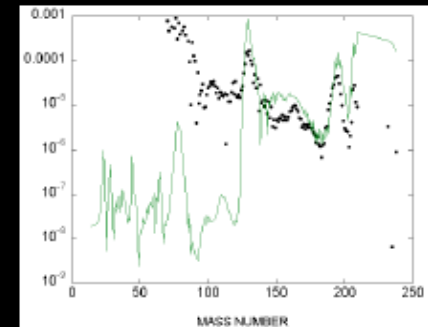


Neutron star mergers

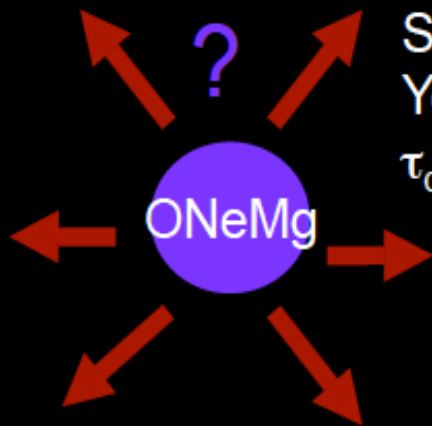
Simulation of NS mergers
(from Hayden planetarium)



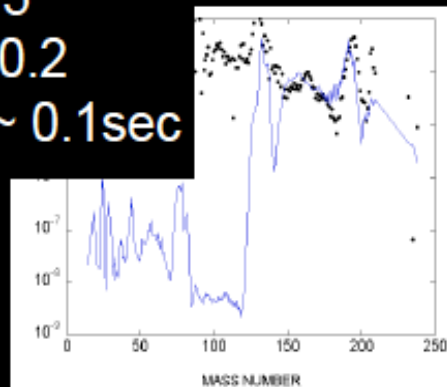
$S < 4$
 $Y_e < 0.2$



Prompt explosion of low mass supernovae



$S \sim 15$
 $Y_e \sim 0.2$
 $\tau_{\text{dyn}} \sim 0.1 \text{ sec}$



The most widely believed candidate site for the r-process are core-collapse supernovae (spectral Type Ib, Ic and II), which provide the necessary physical conditions for the r-process. However, the abundance of r-process nuclei requires that either only a small fraction of supernovae eject r-process nuclei to the interstellar medium, or that each supernova ejects only a very small amount of r-process material. A recently proposed alternative solution is that neutron star mergers (a binary star system of two neutron stars that collide) may also play a role in the production of r-process nuclei, but this has yet to be observationally confirmed.

If nuclear physics uncertainties are reduced, we could identify astrophysical site via observations.

Understanding of R-Nucleosynthesis

r-abundances reproduction

Astrophysical Conditions

temperature (T_9)

matter density

neutron density (n_n)

...

radiation entropy (S_{rad})

r-process duration (τ_r)

Nuclear Physics Input

nuclear-masses (Q_β , S_n)

β -decay properties ($T_{1/2}$, P_n)

n-capture rates

fission properties

neutrino interactions

nuclear-structure-properties (J^π ...)

for 1000's of isotopes **FAR-OFF** β -stability

Nuclear Physics Input

$T_{1/2}$ of very neutron-rich nuclei

Nuclear masses

- S_n -values $\Rightarrow$ r-process **paths** (i.e. $S_n \sim 3$ MeV)

β -decay properties

β^- Half-lives $\Rightarrow$ duration – speed – time scale
 $\Rightarrow$ r-process clock

$T_{1/2}$ (g.s., is.) $\Rightarrow$ r-process **progenitor** abundances ($N_{r,prog}$)

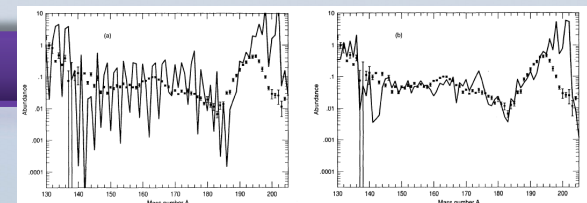
$T_{1/2}$ of waiting points $\Rightarrow$ pre freeze-out isobaric abundances and regulate the **speed** of the process towards heavier elements.

$T_{1/2}$ of heavy neutron-rich radionuclides $\Rightarrow$ **time scale** for the matter flow from the r-process seeds to even heavier nuclei.

Sum of $T_{1/2} \Rightarrow$ total r-process **duration** (τ_r)

β -delayed n-emission branching probabilities

$P_n \Rightarrow$ smoothing $N_{r,prog} \rightarrow N_{r,final} (N_{r,\odot})$



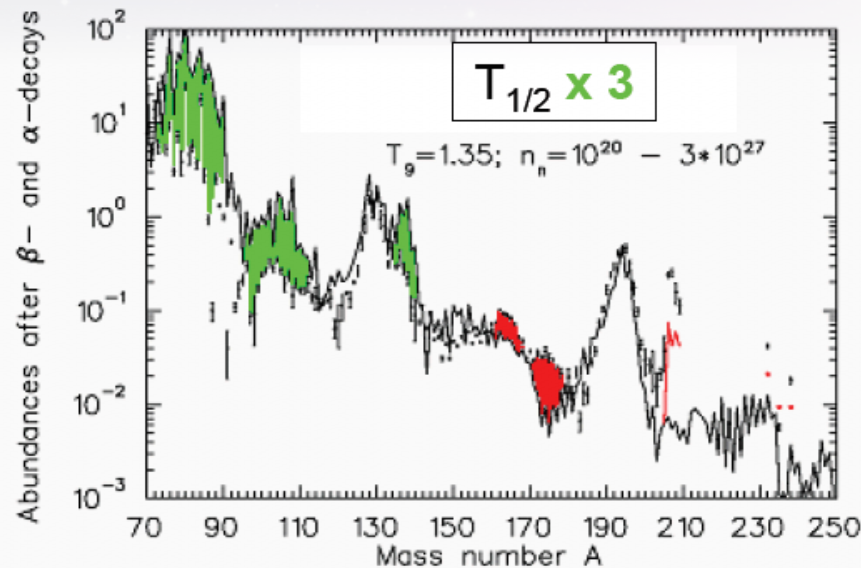
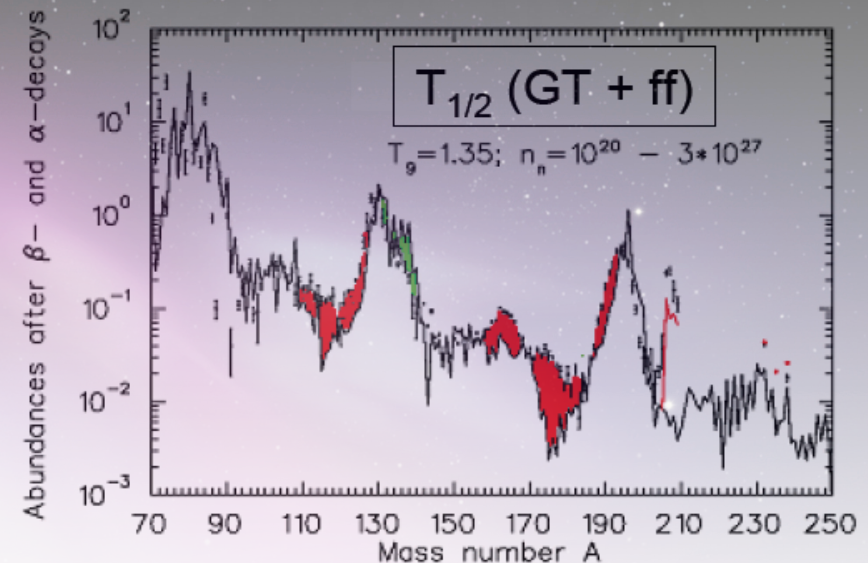
Effects of $T_{1/2}$ on matter flow

Mass model: **ETFSI-Q**

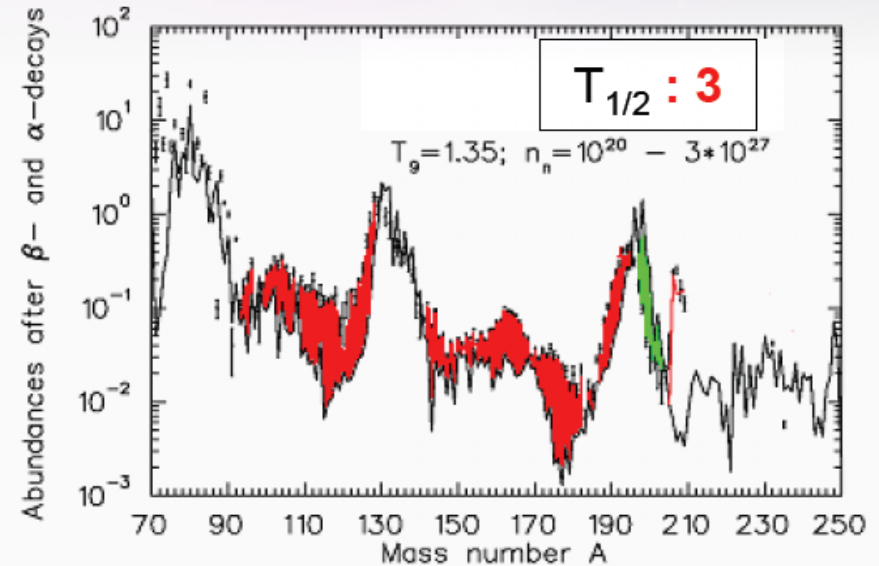
-all astro parameters kept constant

r-process model:

“waiting-point approximation”

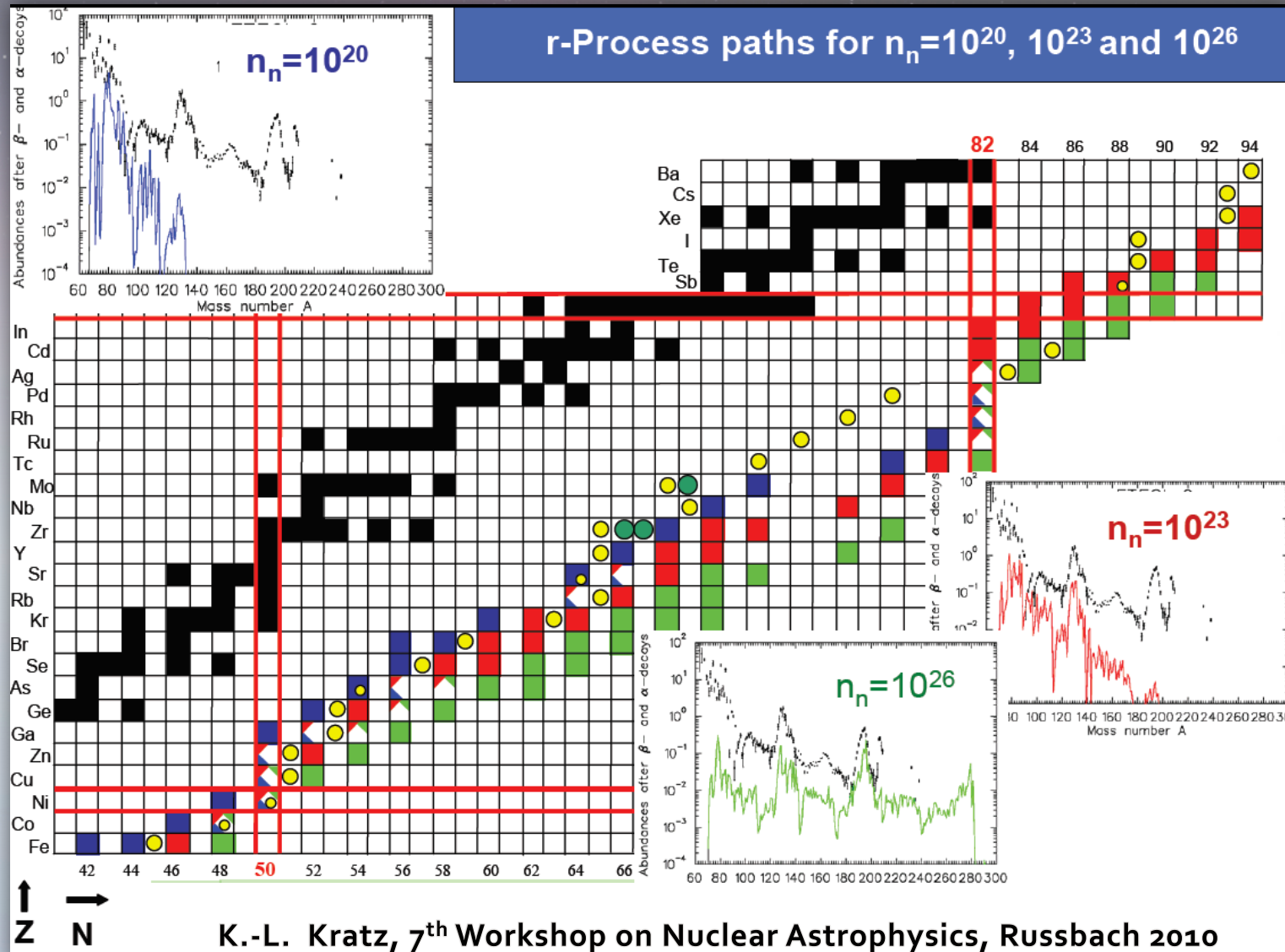


r-matter flow too **slow**



r-matter flow too **fast**

Today's Exp. Info - R-process paths



K.-L. Kratz, 7th Workshop on Nuclear Astrophysics, Russbach 2010

Today, altogether
60 r-process nuclei
are only known!

- Heaviest isotopes with measured $T_{1/2}$
- new (MSU, 2009)

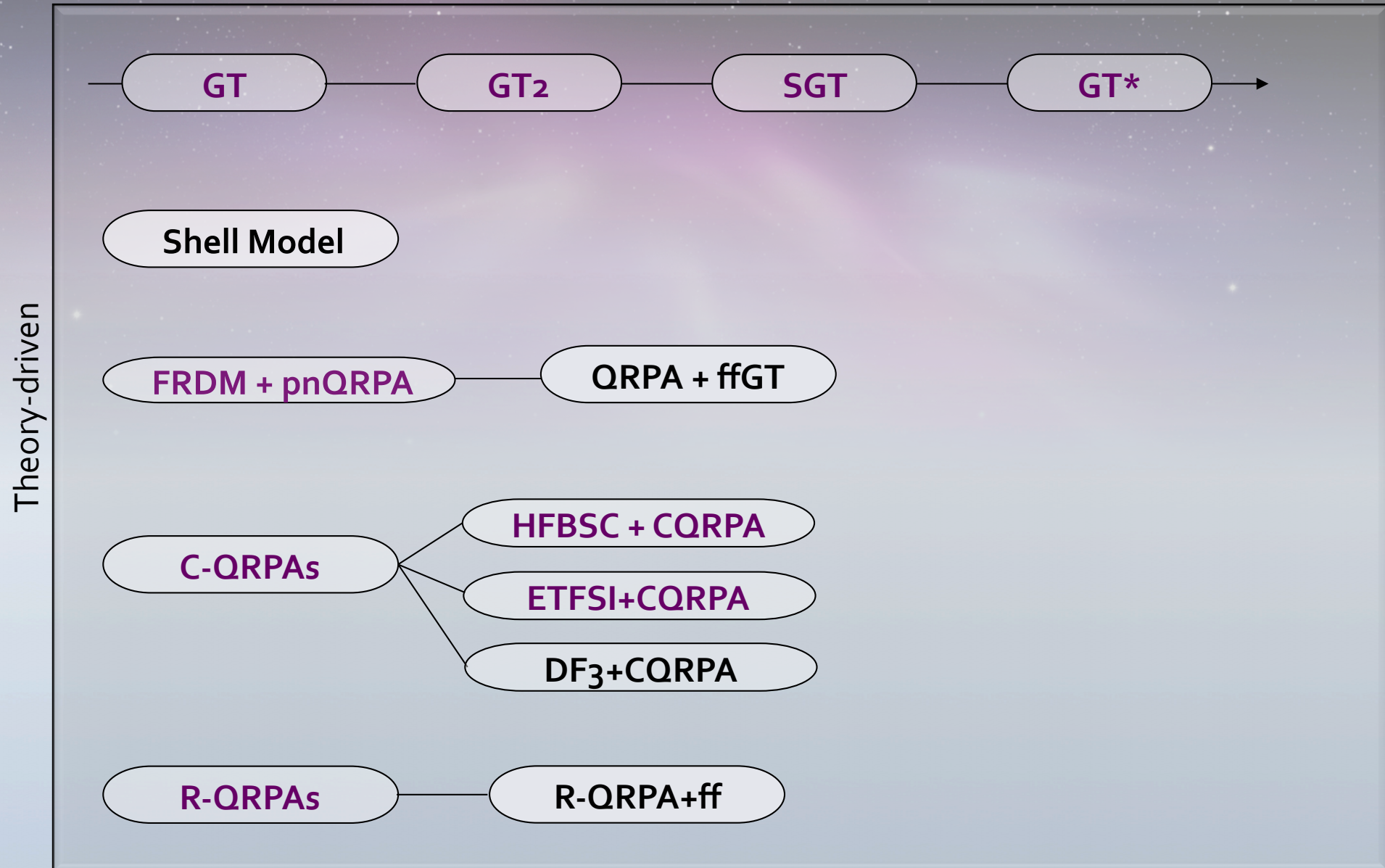
"Waiting point"
isotopes at:

- $n_n = 10^{20}$
- $n_n = 10^{23}$
- $n_n = 10^{26}$

freeze-out networks.

The r-process should be a dynamical process with continuously altered conditions and paths.

β^- -decay Half-lives Approaches



Statistical Modeling

Learning Machines

Data- driven

Artificial Neural Networks (ANNs)
Support Vector Machines (SVMs)

Ref: V. N. Vapnik, Statistical Learning Theory, Wiley-Interscience , 1998.

Statistical Modeling Using ANNs

data-driven, stand-alone statistical approaches

Fundamental Question

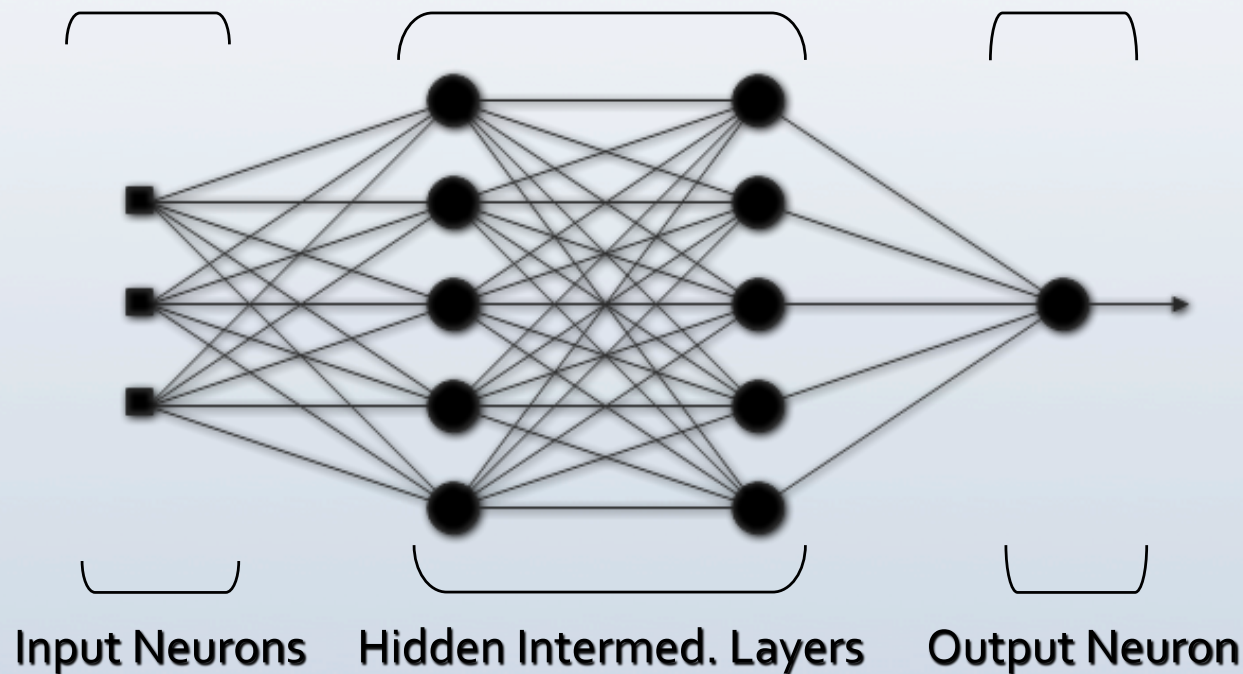
To what extent can mature learning machines (i.e. ANNs) decode the underlying β -decay half-lives systematics by **only** utilizing the hidden information over the **minimal input** of the **Z** and **N** nucleonic numbers?

Artificial Neural Networks

Learning Machines

classification – function approximation

Fully-connected Feed-forward Neural Network



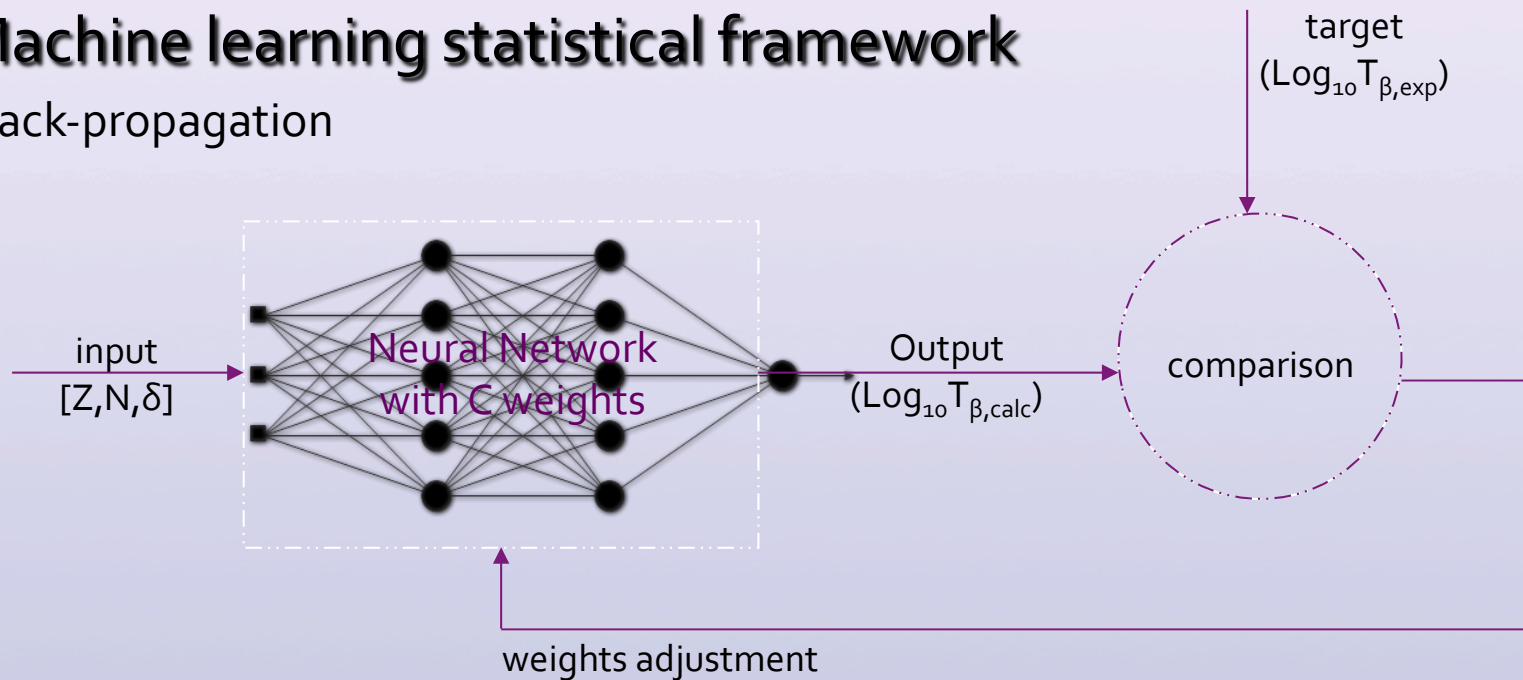
Learning Procedure

Objective: minimization of cost function

$$E_D = \sum_{i=1}^N \left(\log_{10} \left(\frac{T_i}{\beta_{\text{calc}}} \right) - \log_{10} \left(\frac{T_i}{\beta_{\text{exp}}} \right) \right)^2$$

Machine learning statistical framework

Back-propagation



Levenberg-Marquardt BP update rule

$$\vec{w}_{k+1} = \vec{w}_k - \left[\vec{J}_k^T \vec{J}_k + e \right]^{-1} \vec{J}_k^T \vec{e}_k$$

The ANN Model

General ANN Model Features

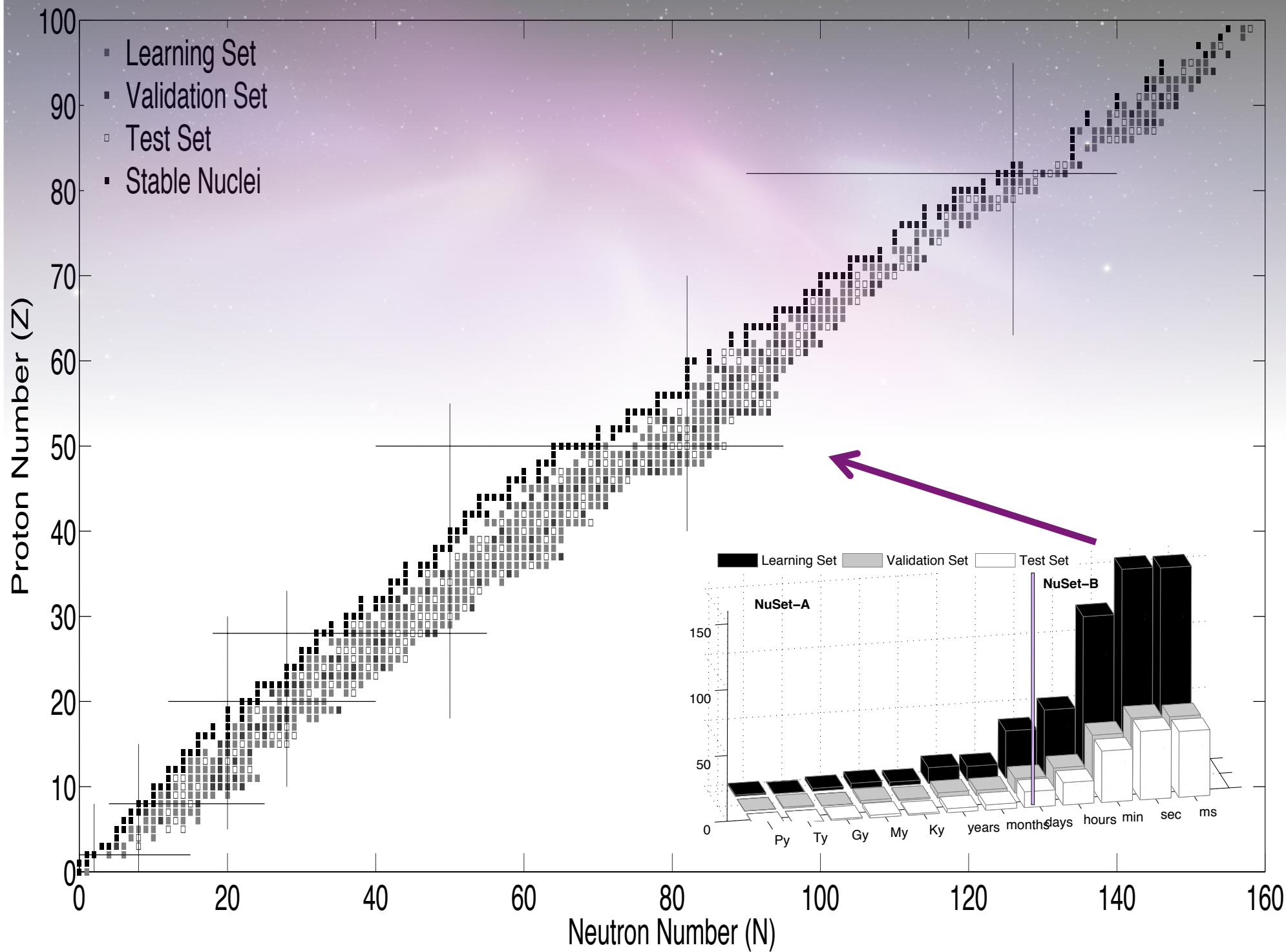
Network		Architecture		Activation Functions
Feed-forward Fully-connected		Size:	3-5-5-5-5-1	tanh-tanh-tanh-tanh-satlins
		Weights:	116	
Mode		Training Algorithm		Initialization Method
Batch		Bayesian Regularization		Nguyen-Widrow
Input Scaling [-1, 1]		Output Scaling [-1, 1]		Max Fail
Z	[0, 230]	Log ₁₀ T _{1/2} (ms)	[0.18 - 8.98]	300 epochs
N	[0, 230]			
Parity	[-1, 1]			

Database	Half-life Range	Decay Mode	NuSet-B	
NuBase2003 *	$0.15 \times 10^{-2} \text{ s} \rightarrow {}^{35}\text{Na}$ $2.43 \times 10^{23} \text{ s} \rightarrow {}^{113}\text{Cd}$	100% β^-	Cutoff	10^6 s

* A.H. Wapstra, G. Audi et al., Nucl Phys. A729 (2003) 337

To avoid overfitting effects		
Early Stopping		
Training Set	Validation Set	Test Set

NuSet-B. Uniform Nuclides Distribution over Half-lives			
All Nuclides	Training Set (60%)	Validation Set (20%)	Test Set (20%)
843	503	167	168



ANN Performance

Set	RMSE
Learning	0.53
Validation	0.60
Test	0.65
Overall	0.57

$$RMSE = \sqrt{\frac{1}{N} \sum_i \left(\log_{10} \frac{T_{\beta,calc}}{T_{\beta,exp}} - \log_{10} \frac{T_{\beta,exp}}{T_{\beta,calc}} \right)^2}$$

fact	T _{1/2} Exp (s)	Overall Mode			NBSC+pnQRPA		
		m(%)	<x> _K	σ _K	m(%)	<x> _K	σ _K
<10	<10 ⁶	90.5	2.46	1.72	76.7	3.00	-
	<60	96.5	2.21	1.52	87.2	2.81	-
	<1	97.6	2.10	1.39	95.7	2.64	-
<2	<10 ⁶	53.5	1.41	0.27	33.8	1.43	-
	<60	60.6	1.41	0.27	42.0	1.41	-
	<1	61.9	1.41	0.26	50.7	1.43	-

$$\bar{x}_K = \frac{1}{N} \sum_i x_i, \quad x_i = \begin{cases} T_{\beta,exp} / T_{\beta,calc} & \text{if } T_{\beta,exp} \geq T_{\beta,calc} \\ T_{\beta,calc} / T_{\beta,exp} & \text{if } T_{\beta,exp} < T_{\beta,calc} \end{cases} \quad \sigma_K = \left[\frac{1}{N} \sum_i (x_i - \bar{x}_K)^2 \right]^{1/2}$$

Results

Recently measured T_β for very neutron-rich nuclides.

Nucleus	Experiment	T_β (ms)	
		ANN Model	$pnQRPA+ffGT$
^{36}Mg	3.9 ± 1.3	14.5	15.9
^{37}Al	10.7 ± 1.3	15	9.9
^{38}Al	7.6 ± 0.6	7.7	4.7
^{39}Al	7.6 ± 1.6	7.2	4.3
^{39}Si	47.5 ± 2.0	27.7	101.5
^{40}Si	33.0 ± 1.0	48.5	30.6
^{42}Si	12.5 ± 3.5	15.5	43.4
^{44}P	18.5 ± 2.5	12.7	17.2
^{46}S	50 ± 8	39.5	30.8
^{47}Cl	101 ± 6	77.9	51.5
^{48}Ar	475 ± 40	447.6	181.9
^{49}Ar	170 ± 50	130.2	54.9
^{64}V	19 ± 8	23.4	7.6
^{73}Co	41 ± 4	103.6	30.7
^{115}Tc	73^{+32}_{-22}	84.2	70.7
^{118}Ru	123^{+48}_{-35}	69.1	211.8
^{120}Rh	136^{+14}_{-13}	196.2	82.7
^{121}Rh	151^{+67}_{-58}	90.7	62.3
^{122}Pd	175 ± 16	227.2	951.2
^{124}Pd	38^{+38}_{-19}	124.2	288.7
^{163}Eu	7.8 ± 5 (s)	7.8 (s)	17.2 (s)
^{164}Eu	4.2 ± 2 (s)	3.3 (s)	8.6 (s)
^{165}Eu	2.3 ± 2 (s)	3.1 (s)	5.7 (s)
^{199}Ir	6^{+5}_{-4} (s)	73 (s)	370.6 (s)
σ_{rms}		0.31	0.53

Relevant to R-process Results

Some r-process waiting-point nuclei at $N = 50$ and $N = 82$ regions.

Nuc.	Exp. Data	ANN Model	$T_{\beta-} (ms)$		
			$pnQRPA+ffGT$	$pn-RQRPA+ff$	DF3+CQRPA
$^{78}\text{Ni}_{50}$	110^{+100}_{-60}a	57	224	150	108
$^{79}\text{Cu}_{50}$	188 ± 25	115	157	-	257
$^{80}\text{Zn}_{50}$	545 ± 16	371	1.26(s)	970	839
$^{128}\text{Pd}_{82}$	-	81	74	63	43
$^{129}\text{Ag}_{82}$	44 ± 7	77	32	-	56
$^{130}\text{Cd}_{82}$	162 ± 7	158	502	299	147
$^{131}\text{In}_{82}$	280 ± 30	307	139	-	201
$^{132}\text{Sn}_{82}$	$39.7 \pm 0.5(s)$	30(s)	23.8(s)	472.5(s)	29.8(s)

^a Recent first time experimentally deduced half-life value at the CCF of the NSCL at MSU by Hosmer et al.

Recently measured* T_{β} of eight heavier abuse to ff neutron-rich isotopes close to the neutron shell $N = 126$ around $A = 195$.

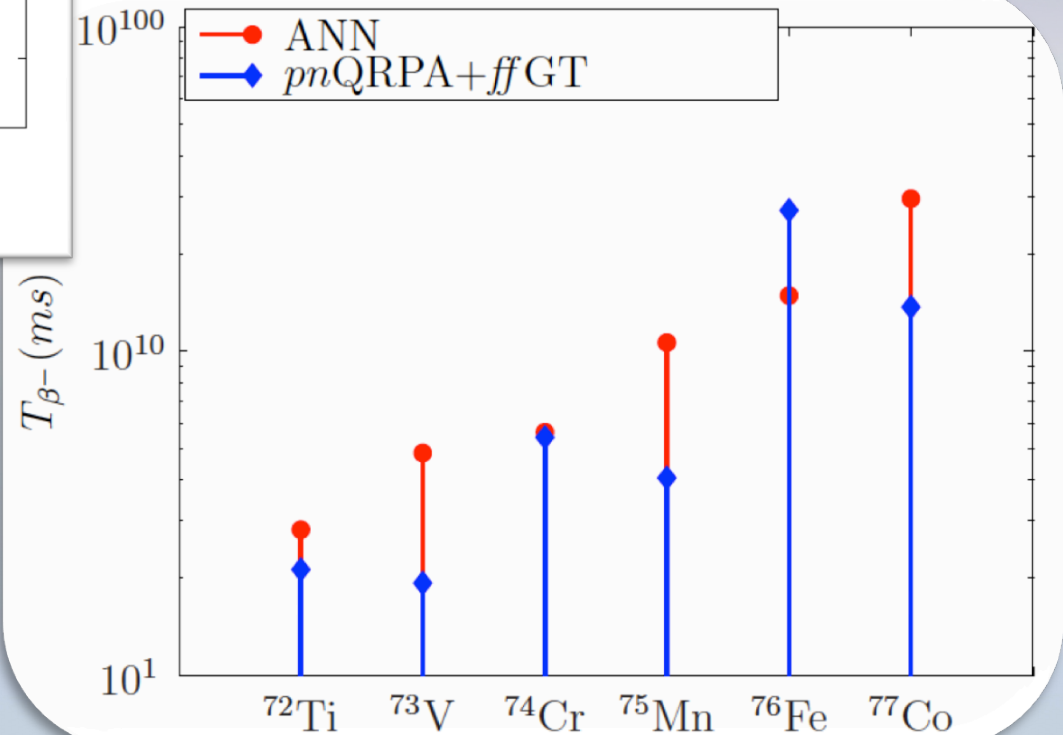
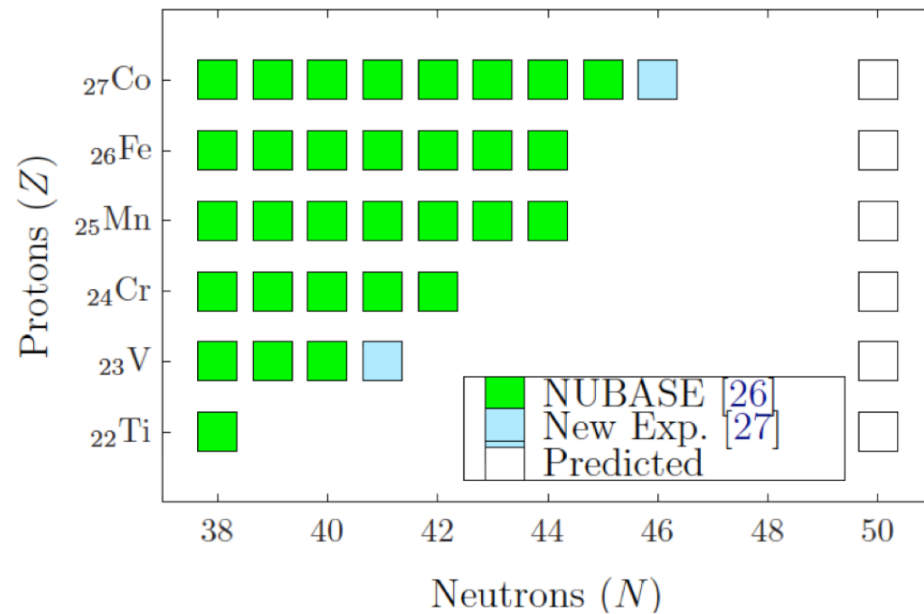
Nucleus	Exp. Data	ANN Model	$pnQRPA+ffGT$	DF3+CQRPA	Gr. Th.
^{194}Re	$1^{+0.5}_{-0.5}$	20.8	70.8	2.1	16.1
^{195}Re	6^{+1}_{-1}	23.9	3.3	8.5	10.3
^{196}Re	3^{+1}_{-2}	8.8	3.6	1.4	5.1
^{199}Os	5^{+4}_{-2}	13.6	106.8	6.6	17.2
^{200}Os	6^{+4}_{-3}	21.7	187.1	6.9	16
^{198}Ir	8^{+2}_{-2}	57.6	377.1	19.1	-
^{199}Ir	6^{+5}_{-4}	73	370.6	46.7	96.6
^{202}Ir	11^{+3}_{-3}	8.6	68.4	9.8	8.5
σ_{rms}		0.77	1.33	0.39	-

*T. Kurtukian-Nieto et al.

N=50 Closed Shell

$A = 70 - 80$ mass region

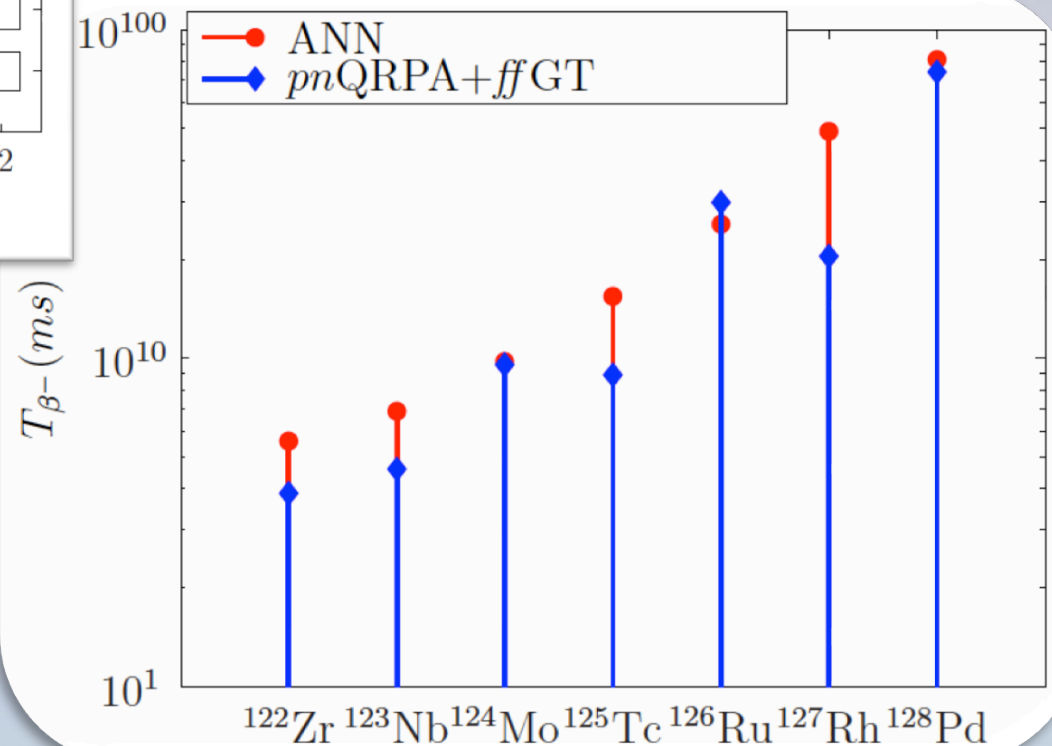
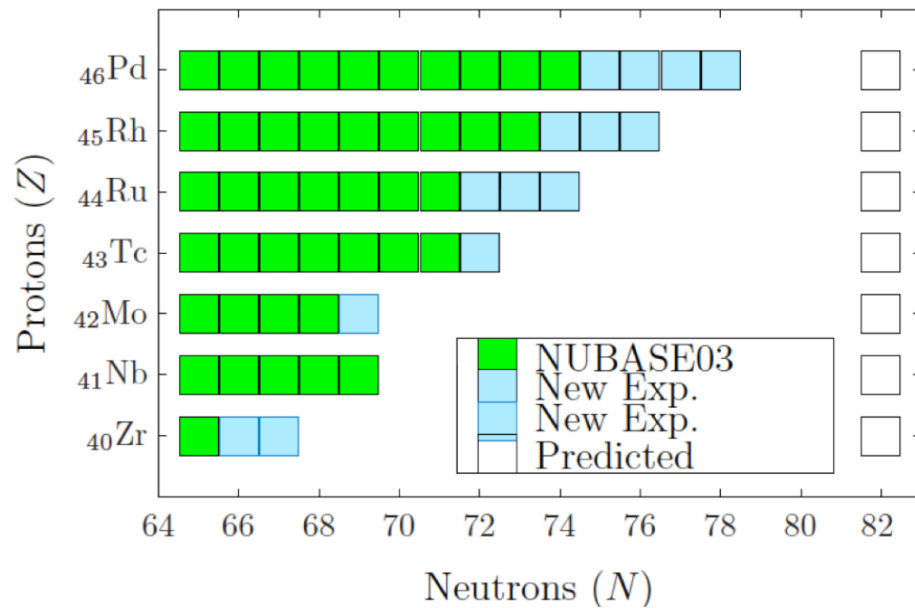
Chart of nuclides up to $N = 50$



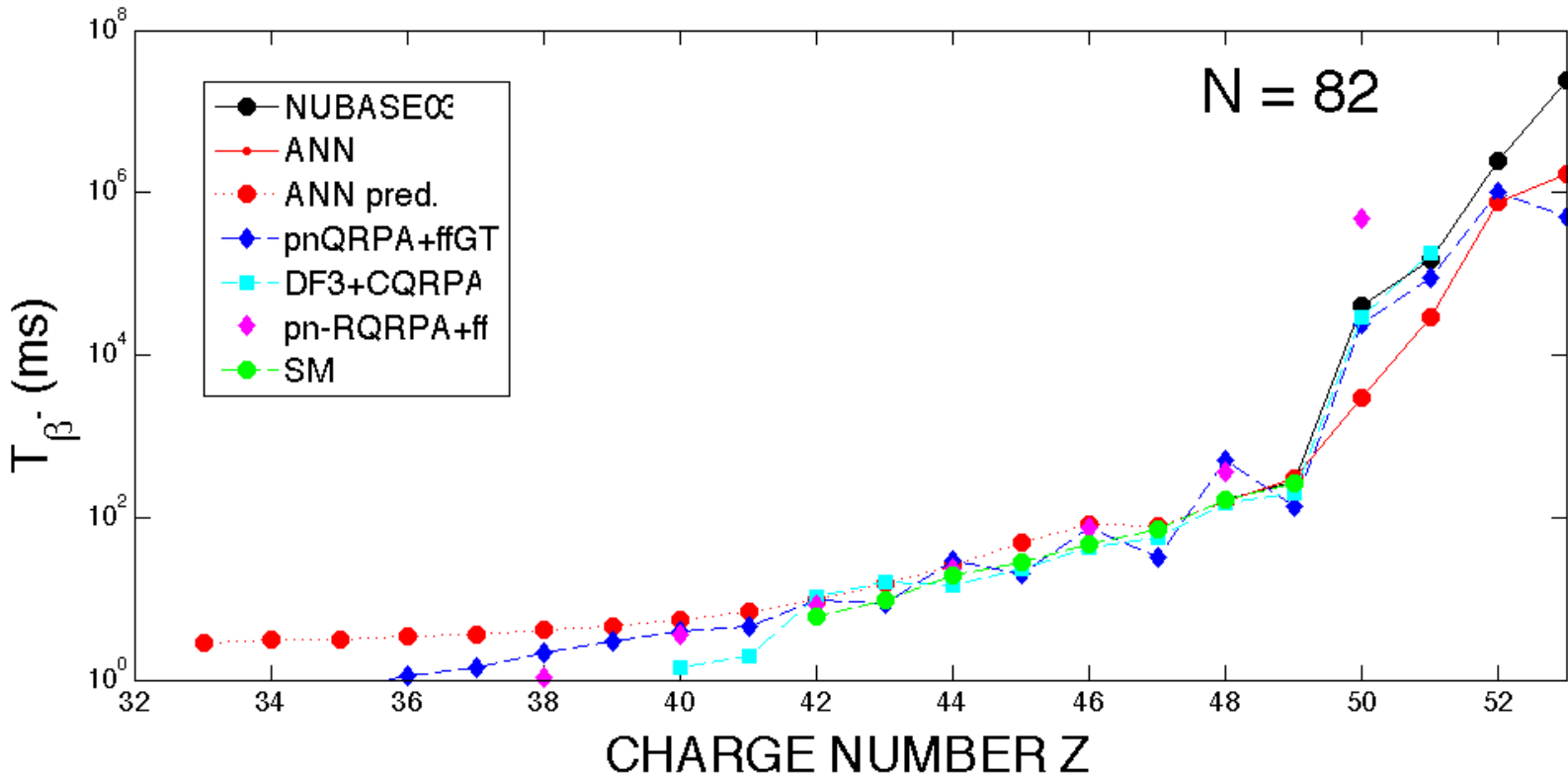
N=82 Closed Shell

$A = 120 - 130$ mass region

Chart of nuclides up to $N = 82$



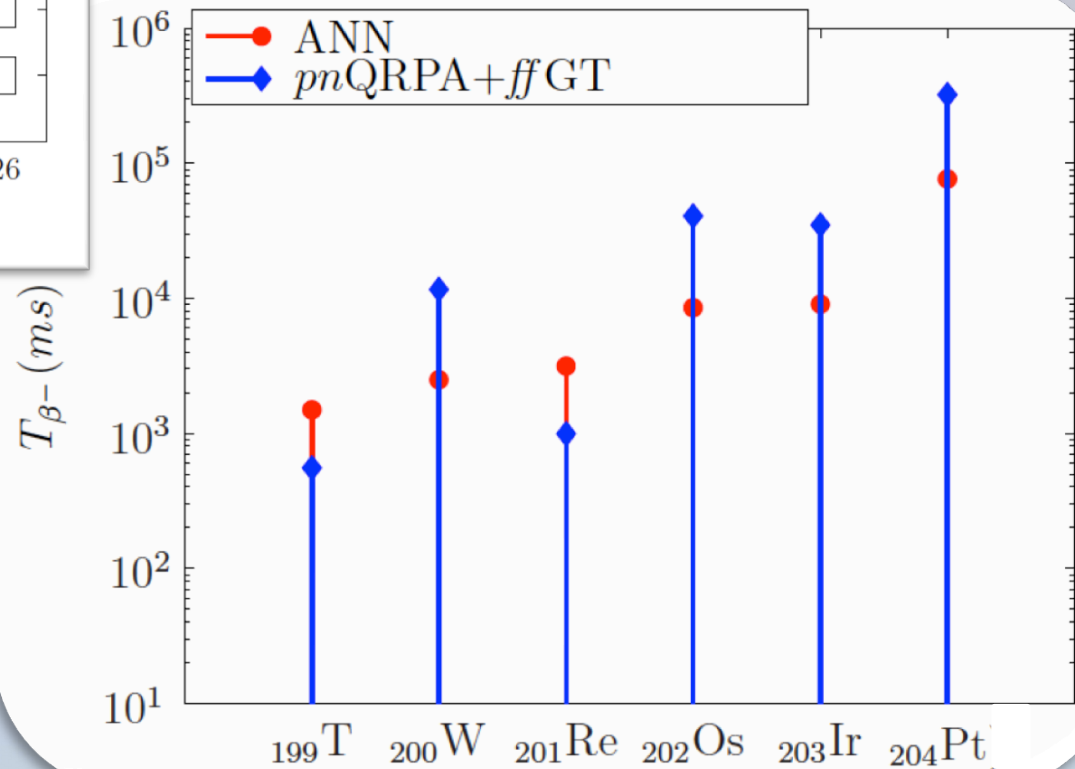
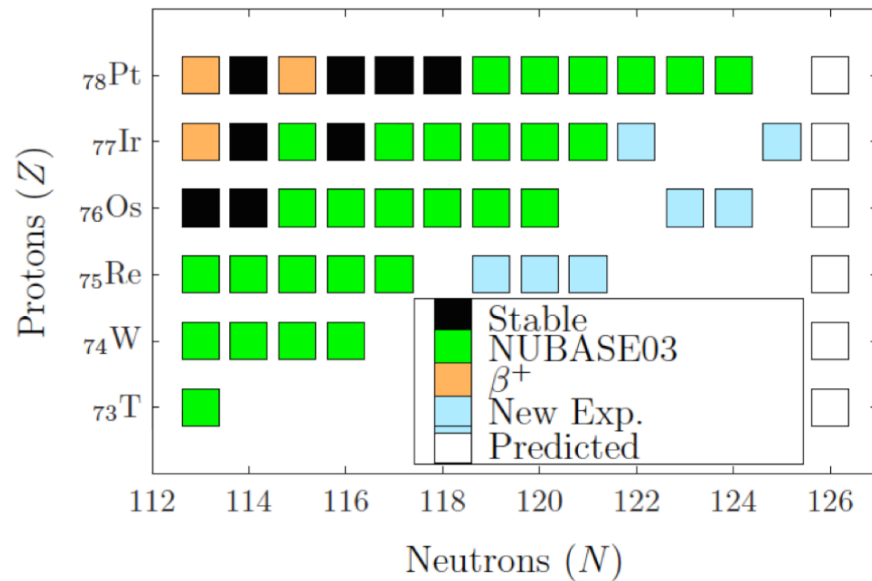
INTELLECTUAL PROPERTY



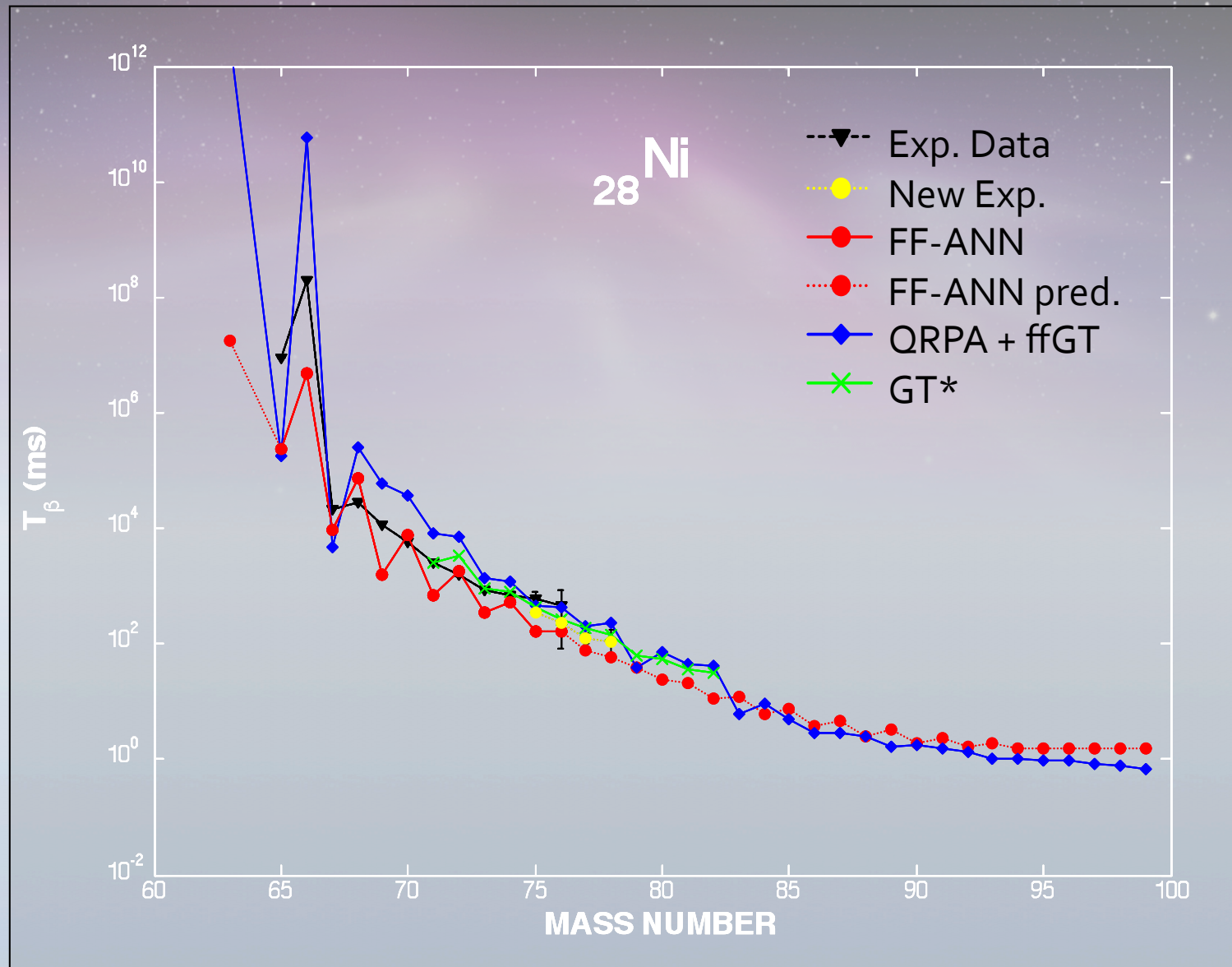
N=126 Closed Shell

$A = 195 - 205$ mass region

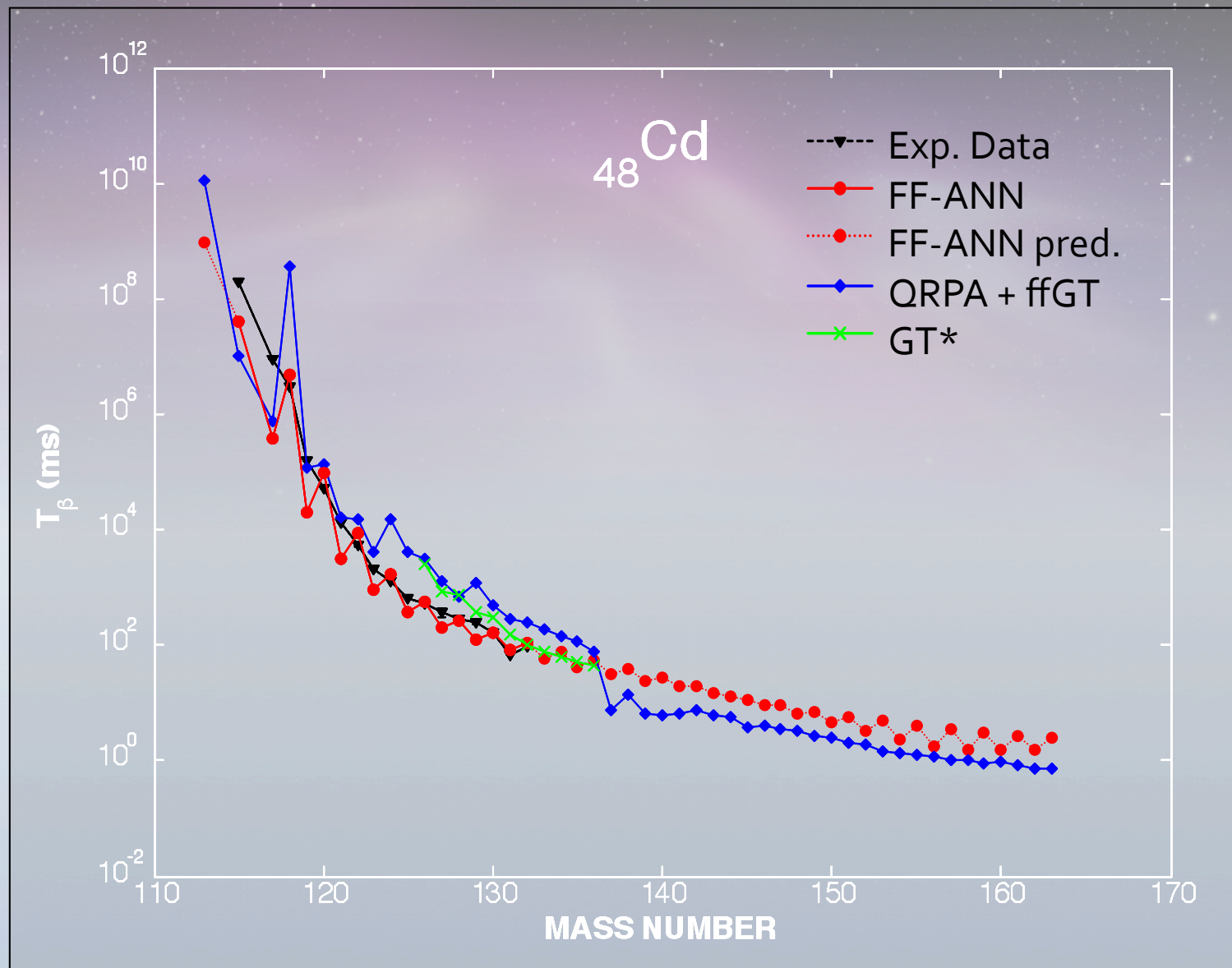
Chart of nuclides up to $N = 126$



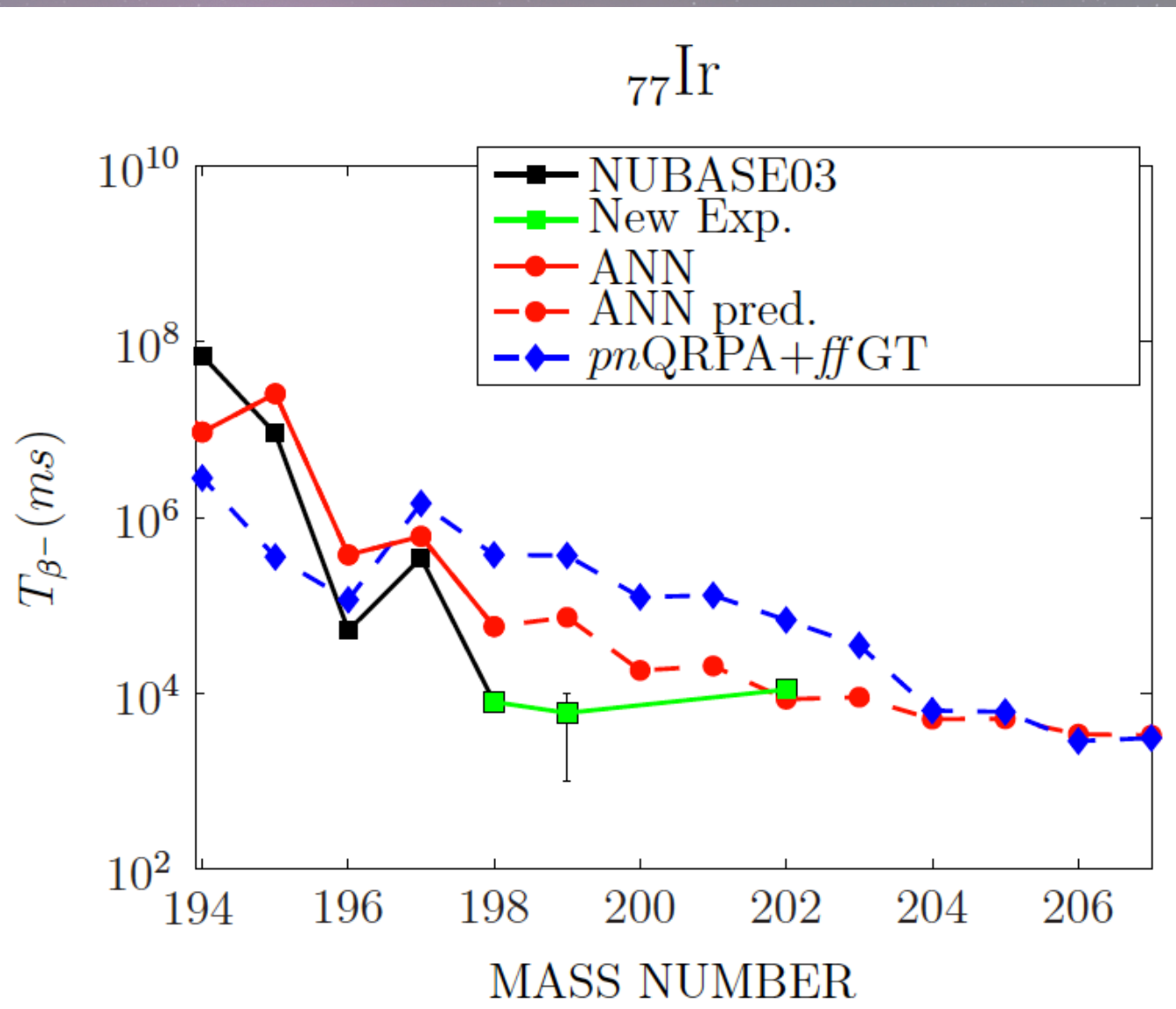
Isotopic Chain of $_{28}\text{Ni}$



Isotopic Chain of $_{48}\text{Cd}$

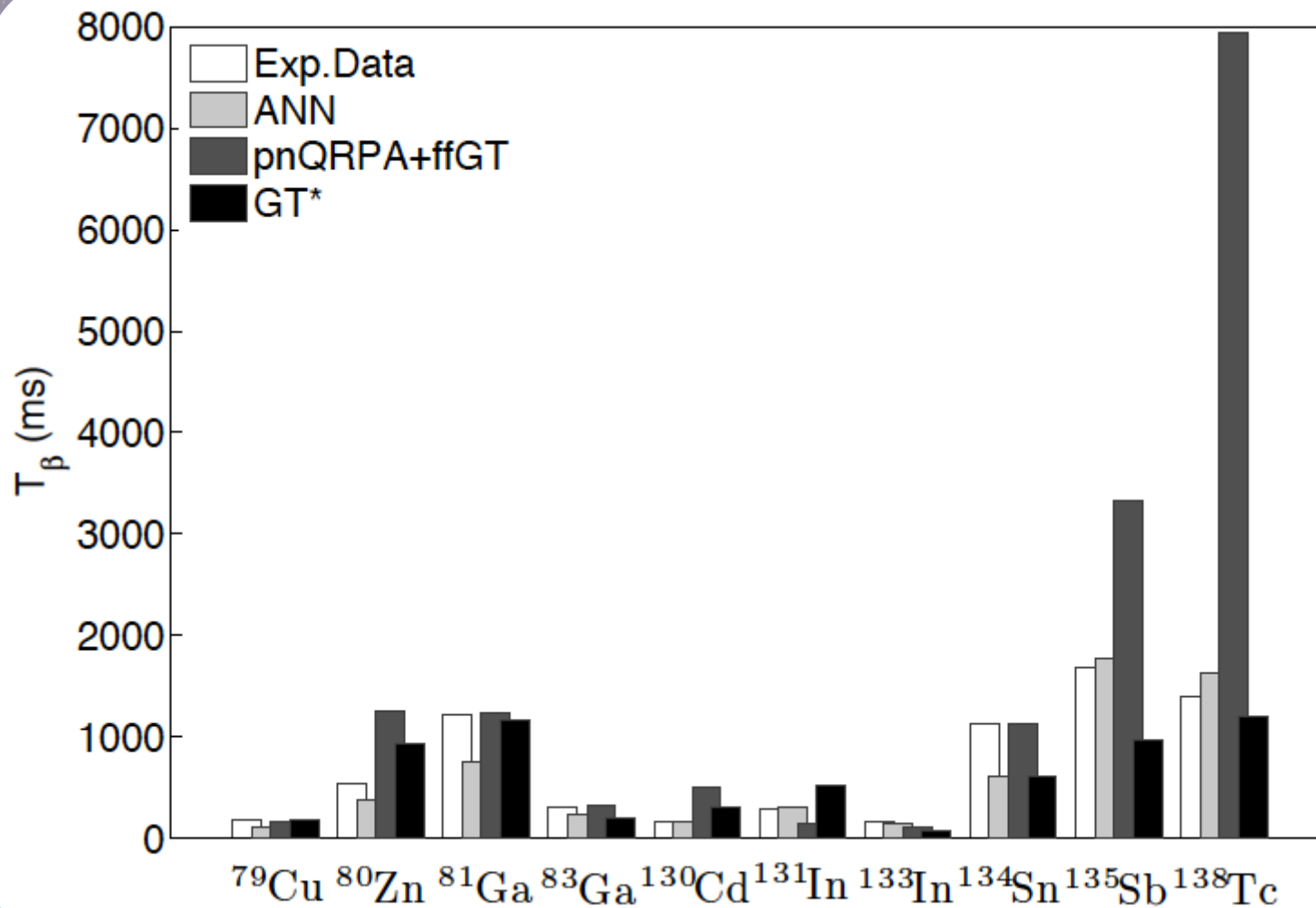


Isotopic Chain of ^{77}Ir



Nuclides on or near a typical r-process path

with S_n up to 3 MeV

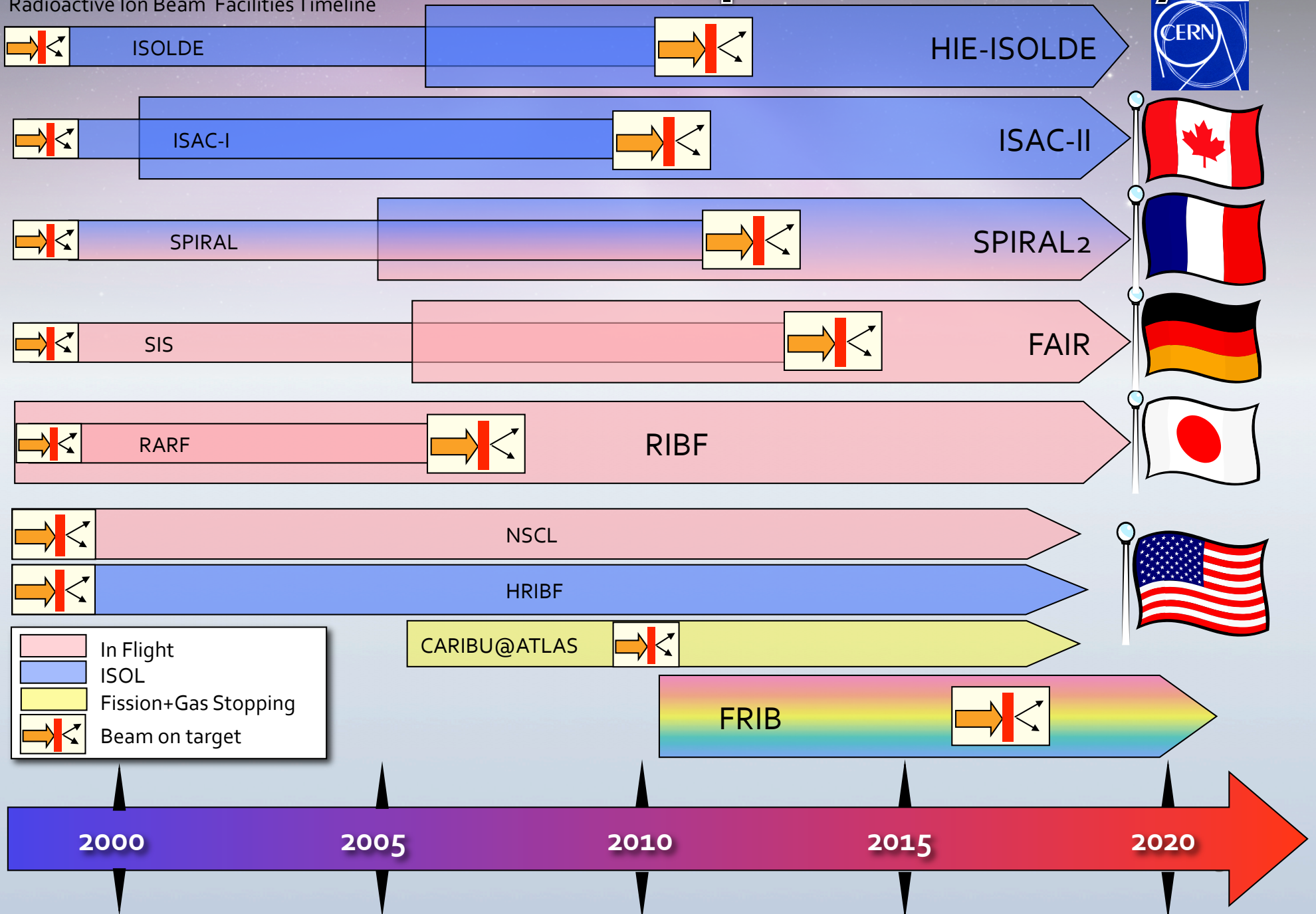


^{79}Cu ^{80}Zn ^{81}Ga ^{83}Ga ^{130}Cd ^{131}In ^{133}In ^{134}Sn ^{135}Sb ^{138}Tc



New data will help enormously

Radioactive Ion Beam Facilities Timeline



Conclusions & Prospects

- **Conclusions**

Theory-thin statistical global models of nuclear properties developed by learning machines should provide a valuable, robust additional tool to **complement** the nuclear systematics studies **beyond the standard nuclear landscape** planned with the present and future very neutron- rich, rare-isotope experimental facilities.

- **Prospects**

We plan further studies of nuclear properties relevant to r-process: masses, neutron capture cross-sections with the already developed by us ANN and SVM techniques.

We also plan further studies of nuclear properties using Artificial Intelligence's more mature learning strategies, such as **committee of machines** (CoM) - a collection of different feed-forward Artificial Neural Networks (ANNs) instead of a single ANN, with a view to refine current results.

Questions



Thank You

β -decay half-lives using the ANN model:
Input for the r-process*

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[†] Department of Physics, Section of Nuclear & Particle Physics, University of Athens, 15771 Athens, Greece

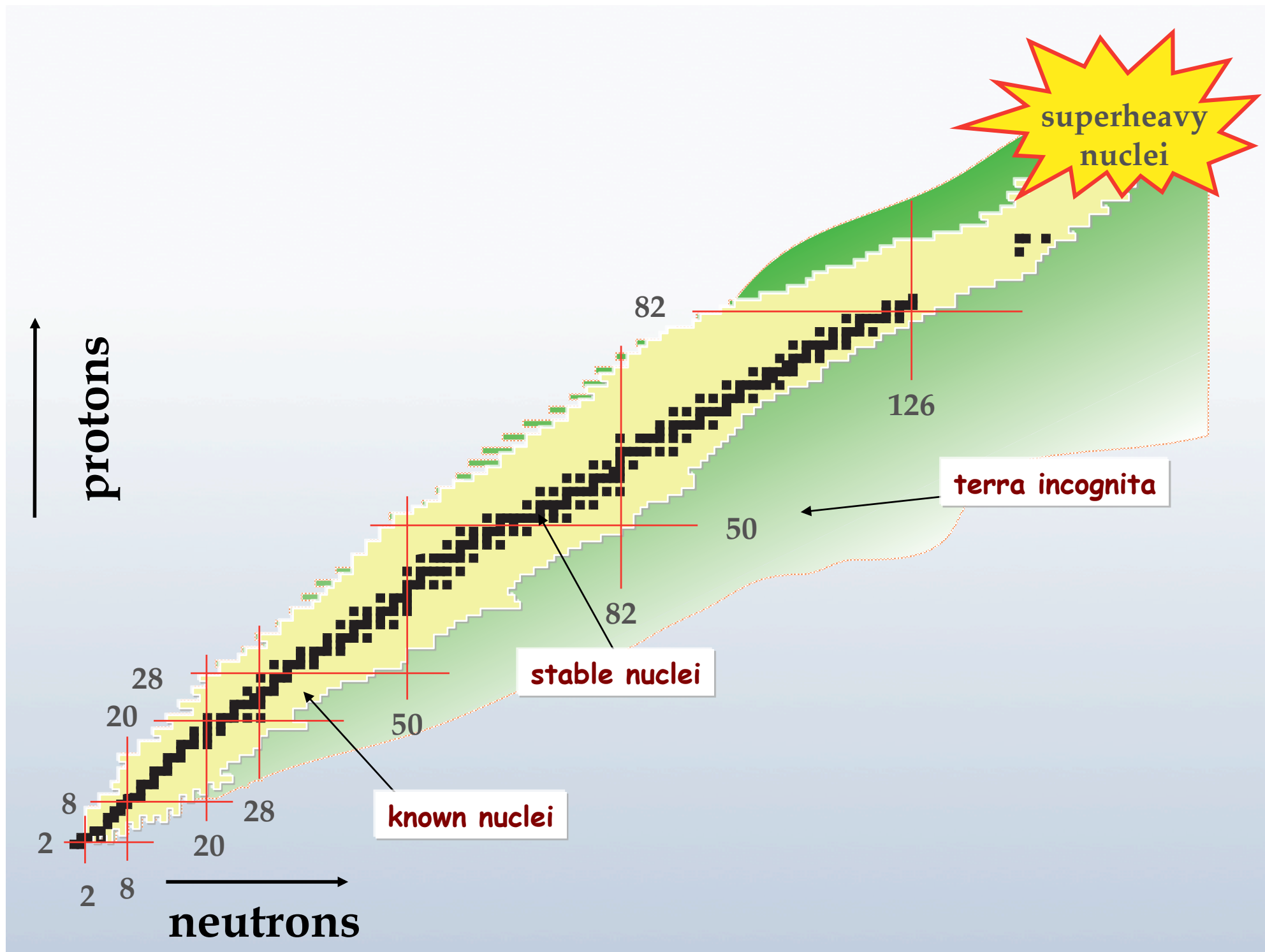
[‡] McDonnell Center for the Space Sciences and Department of Physics, Washington University, St. Louis, Missouri 63130, USA

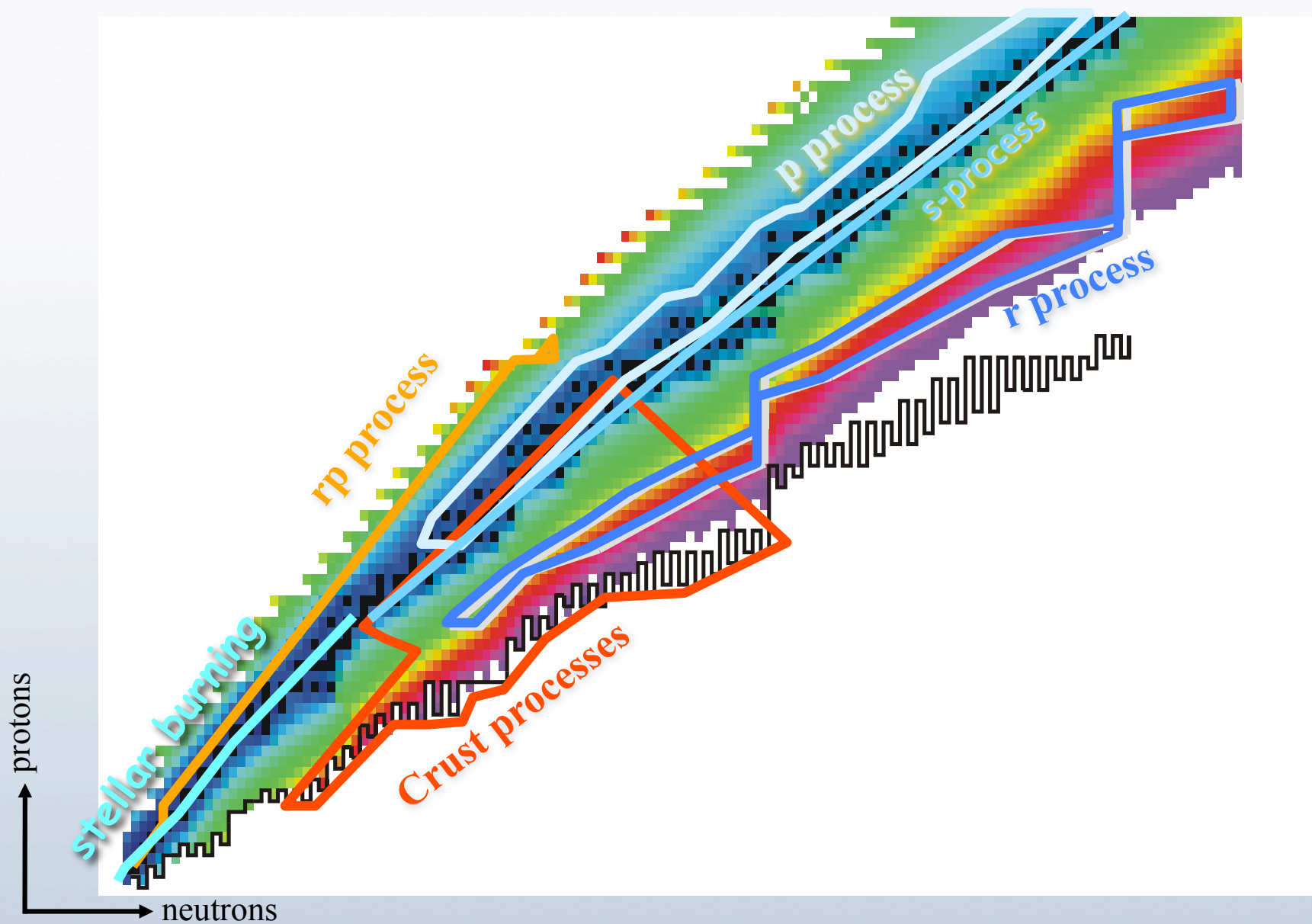
[§] School of Physics & Astronomy, Schuster Building, The University of Manchester, Manchester, M13 9PL, United Kingdom

URL: <http://www.pythaim.phys.uoa.gr>

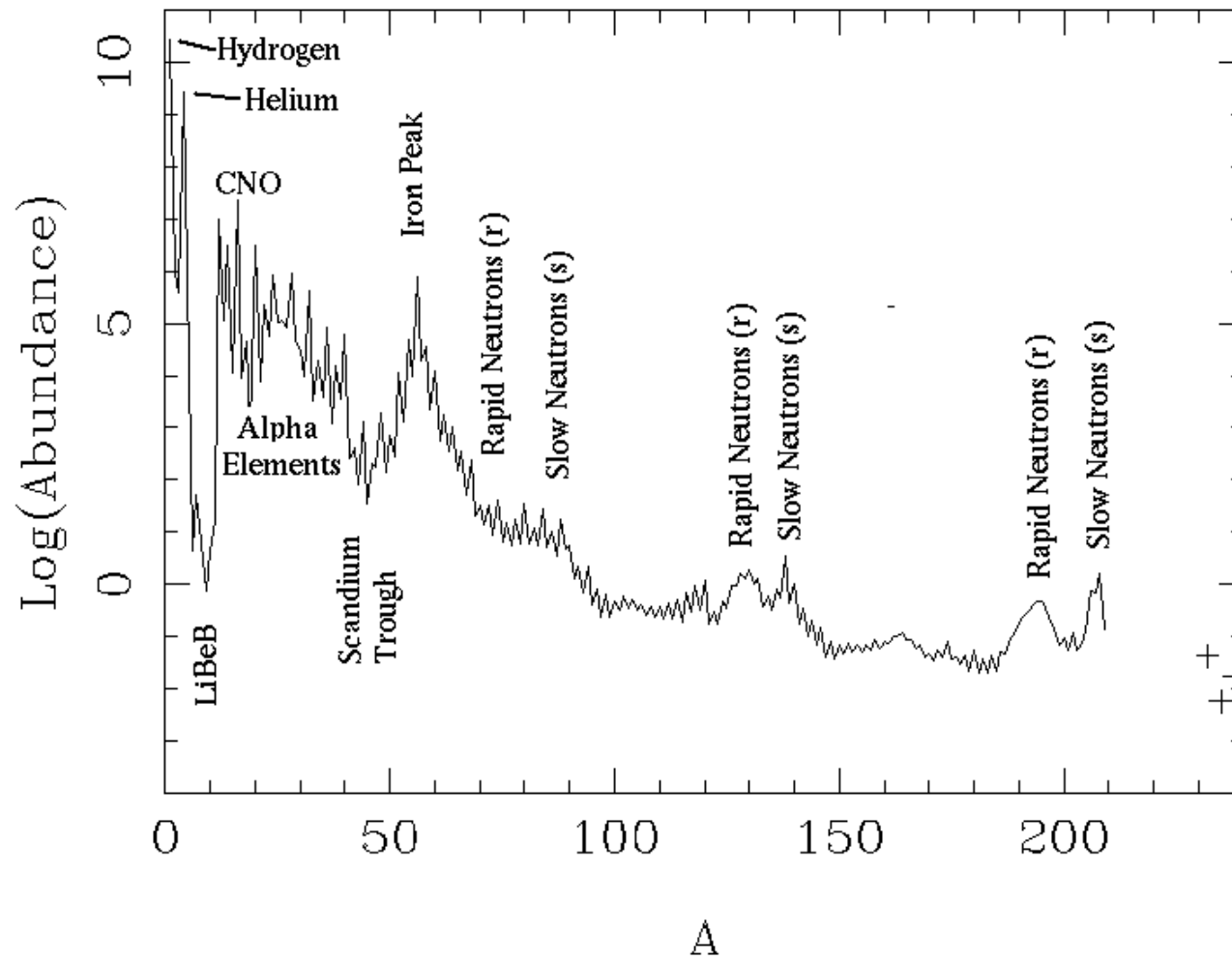
E-mail: pythaim@phys.uoa.gr

Backup Slides

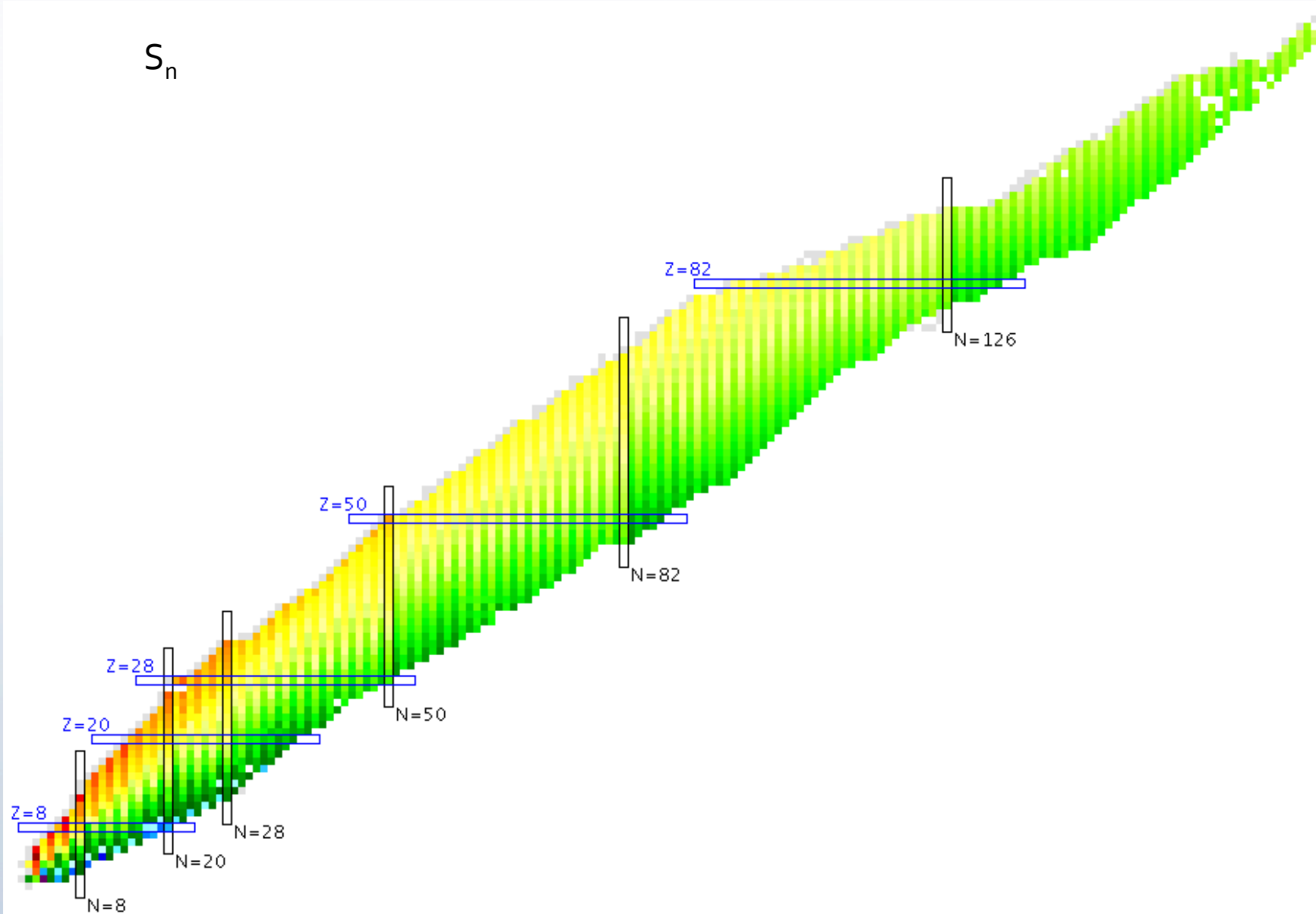


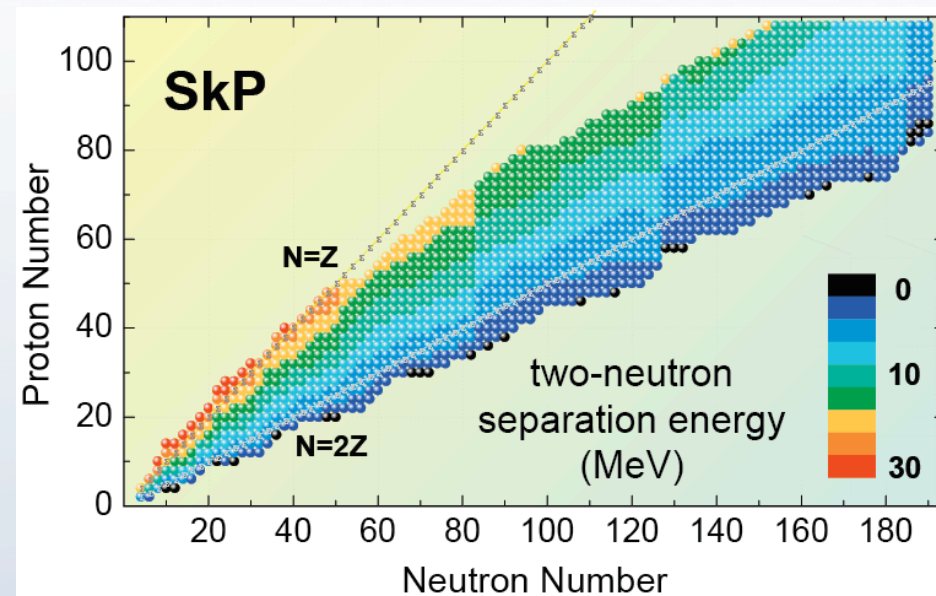


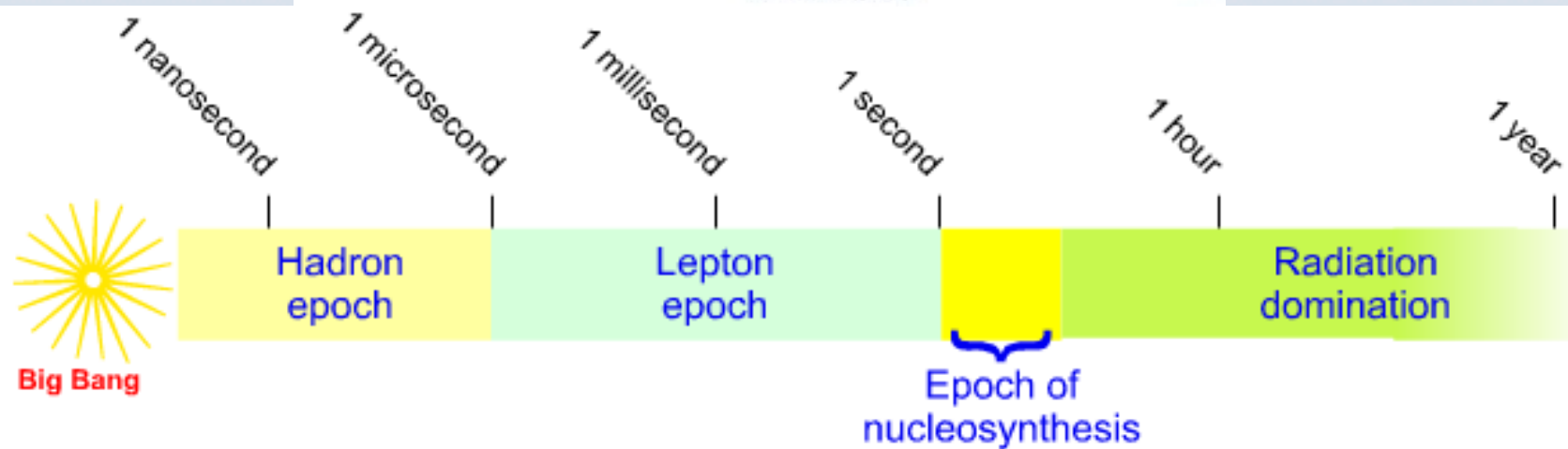
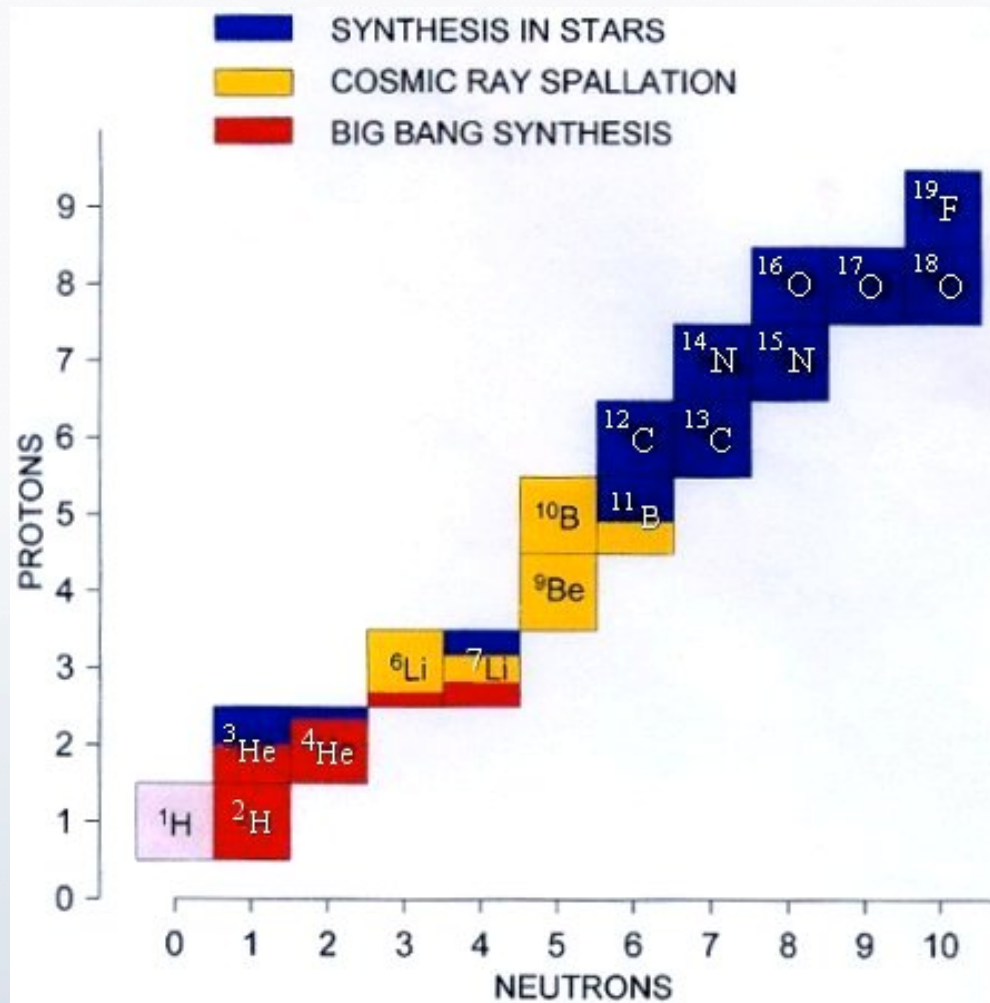
Standard Abundance Distribution (SAD) vs. A



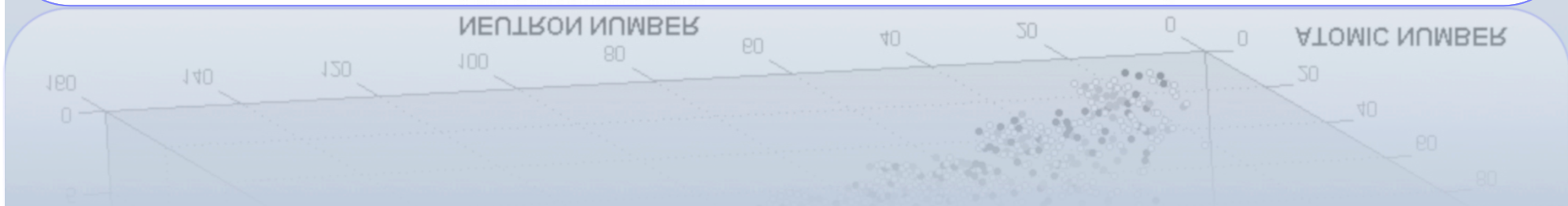
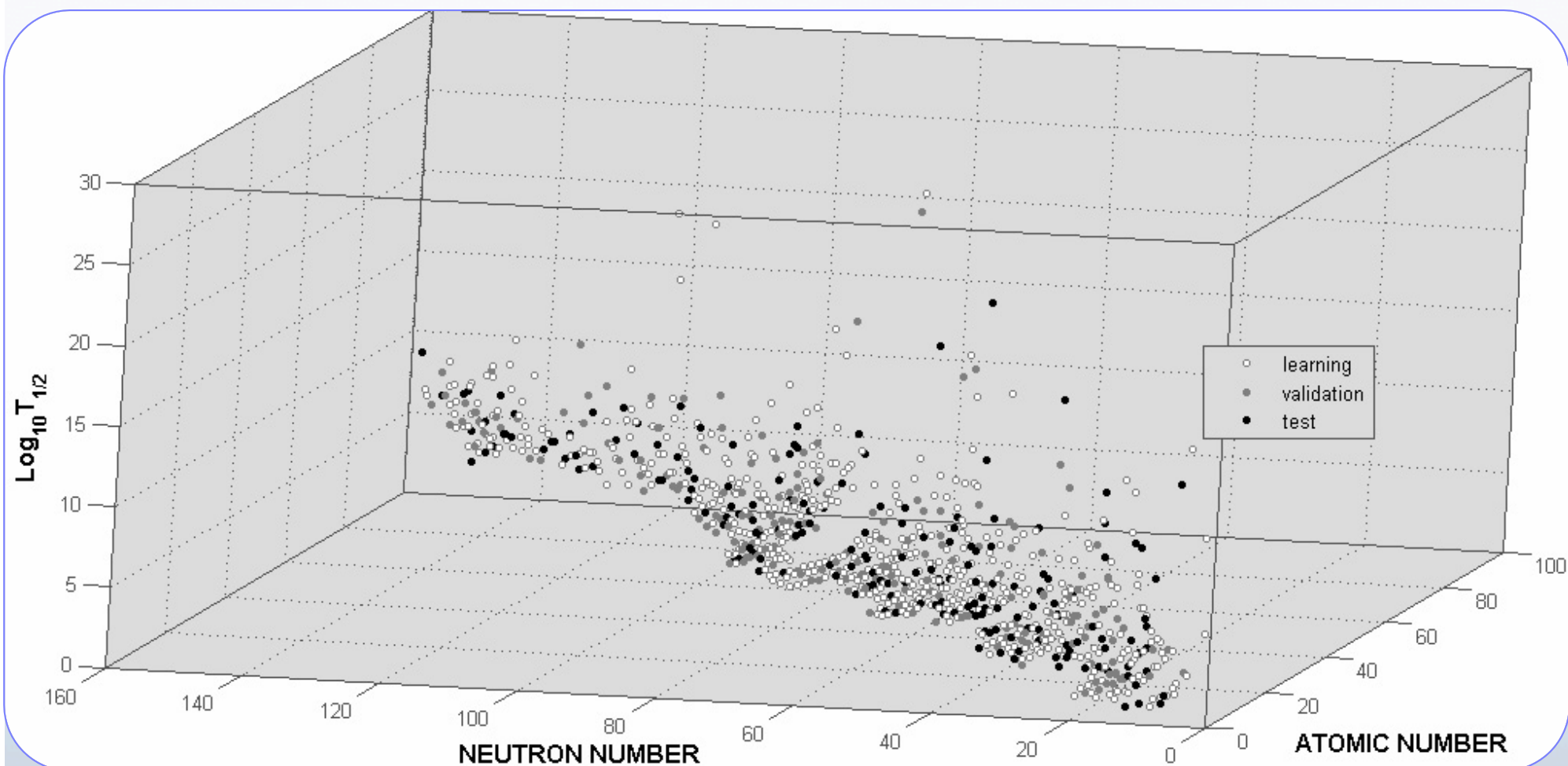
S_n



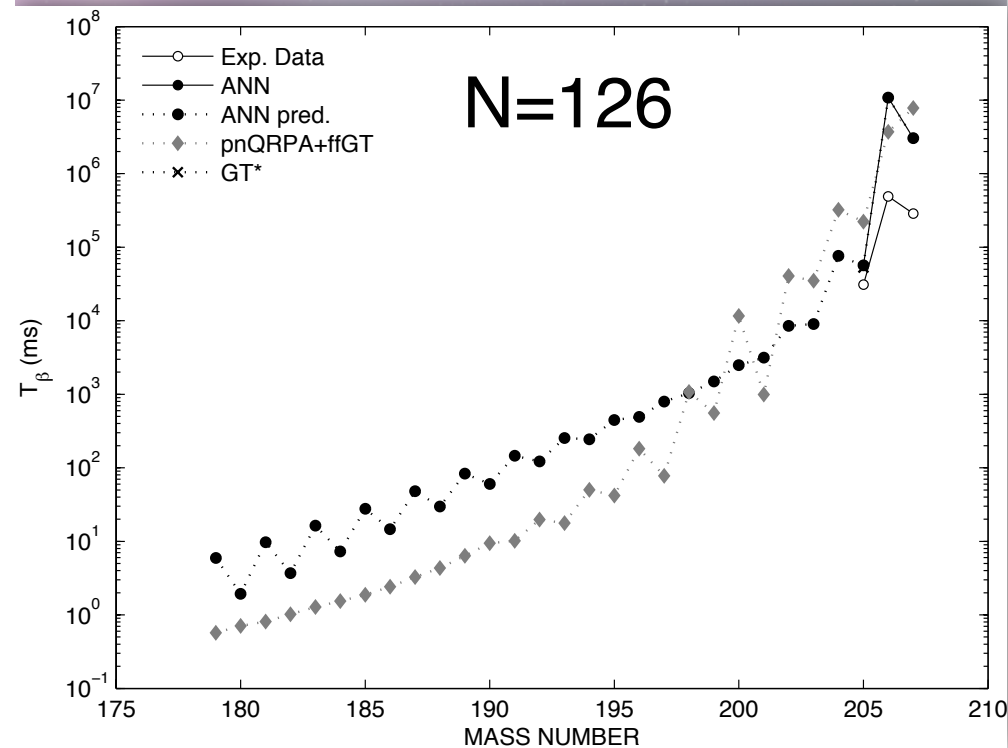
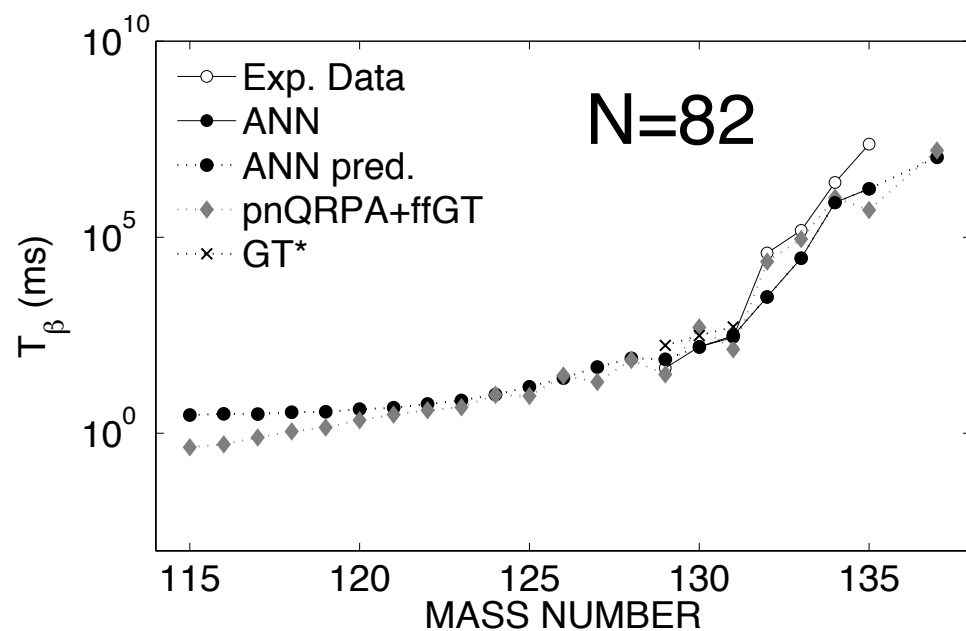
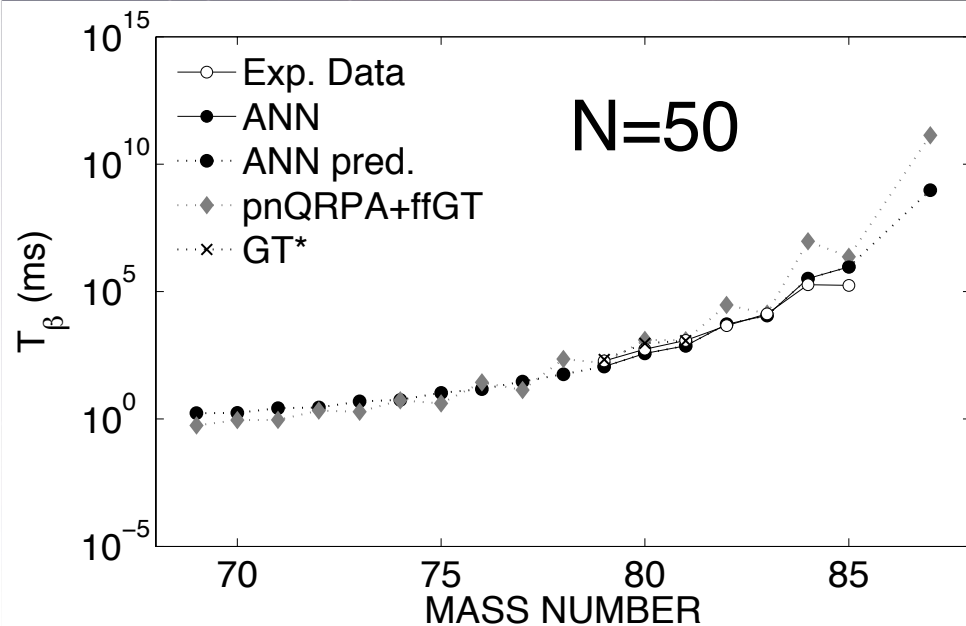




NuBase03

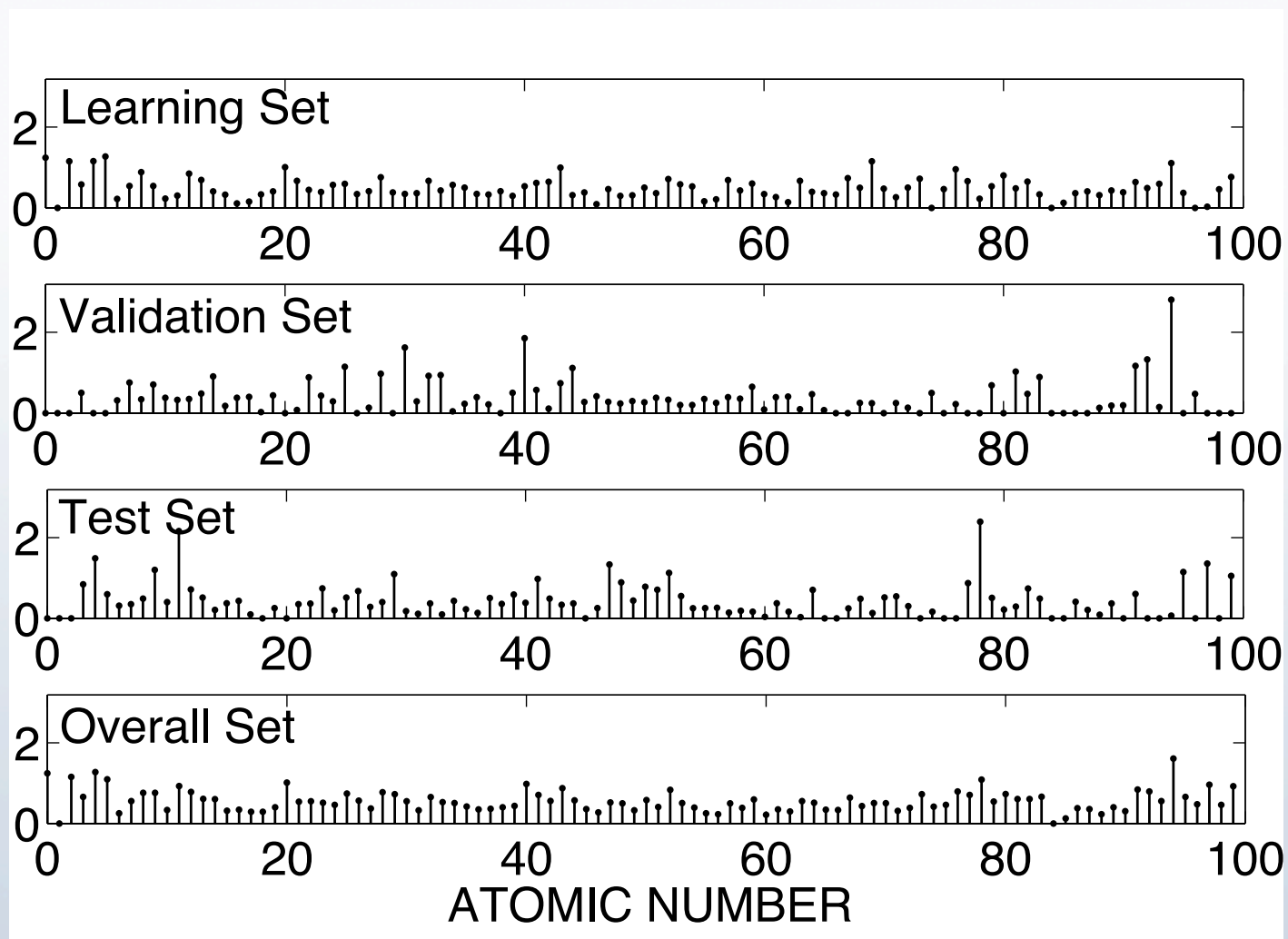


Isotonic Chains



Nucleus	NUBASE03	ANN Model	$T_{\beta-} (s)$		
			$pnQRPA+ffGT$	$pn-RQRPA$	$pn-RQRPA+ffGT$
^{70}Ni	6.00 (0.30)	7.58	36.72	35.28	4.89
^{72}Ni	1.57 (0.05)	1.83	7.03	9.11	1.42
^{74}Ni	0.680 (0.120)	0.530	1.20	3.57	0.563
^{76}Ni	0.470 (0.390), $0.238^{+0.015}_{-0.018}$	0.164	0.433	2.13	0.279
^{78}Ni	$0.110^{+0.100}_{-0.600}$	0.057	0.224	1.71	0.150
^{122}Cd	5.24 (0.03)	8.74	15.17	0.549	2.40
^{124}Cd	1.25 (0.02)	1.71	15.20	0.416	1.21
^{126}Cd	0.515 (0.017)	0.560	3.23	0.329	0.673
^{128}Cd	0.280 (0.040)	0.265	0.695	0.297	0.429
^{130}Cd	0.162 (0.007)	0.158	0.502	0.299	0.299
^{132}Cd	0.097 (0.010)	0.107	0.243	0.184	0.172
^{198}Ir	8^{+2}_{-2}	57.6	377.1	19.1	-
^{199}Ir	6^{+5}_{-4}	73	370.6	46.7	96.6
^{202}Ir	11^{+3}_{-3}	8.6	68.4	9.8	8.5

RMSE



Klapdor's Metrics*

* H. Homma, M. Bender et al., Phys. Rev. C 54:6 (1996) 2972

fact	T _{1/2} Exp (s)	Overall Mode			Prediction Mode			NBSC+pnQRPA		
		m(%)	<x> _K	σ _K	m(%)	<x> _K	σ _K	m(%)	<x> _K	σ _K
<10	<10 ⁶	90.5	2.46	1.72	90.5	2.69	1.85	76.7	3.00	-
	<60	96.5	2.21	1.52	96.1	2.48	1.64	87.2	2.81	-
	<1	97.6	2.10	1.39	98	2.24	1.30	95.7	2.64	-
<5	<10 ⁶	82.8	1.99	0.95	79.2	2.10	0.97	-	-	-
	<60	90.2	1.88	0.84	87.3	2.05	0.91	-	-	-
	<1	93.7	1.88	0.8	94	2.04	0.89	-	-	-
<2	<10 ⁶	53.5	1.41	0.27	49.4	1.48	0.28	33.8	1.43	-
	<60	60.6	1.41	0.27	53.9	1.48	0.27	42.0	1.41	-
	<1	61.9	1.41	0.26	60	1.50	0.27	50.7	1.43	-

$\bar{x}_K$ must tends towards 1 while σ_K must tends towards 0.

$$\bar{x}_K = \frac{1}{N} \sum_i x_i, \quad x_i = \begin{cases} T_{\beta, \text{exp}} / T_{\beta, \text{calc}} & \text{if } T_{\beta, \text{exp}} \geq T_{\beta, \text{calc}} \\ T_{\beta, \text{calc}} / T_{\beta, \text{exp}} & \text{if } T_{\beta, \text{exp}} < T_{\beta, \text{calc}} \end{cases} \quad \sigma_K = \left[\frac{1}{N} \sum_i (x_i - \bar{x}_K)^2 \right]^{1/2}$$