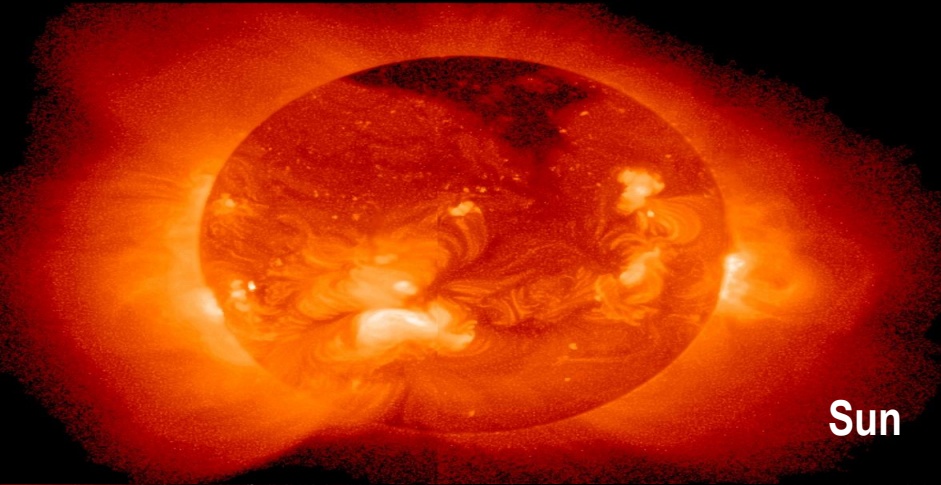
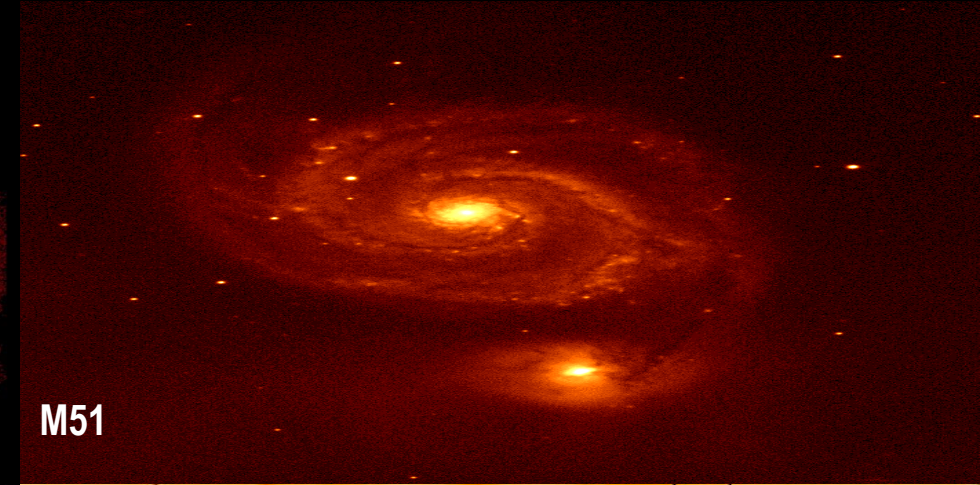


Unravelling the nonlinear physics of high-Z regions with neural networks



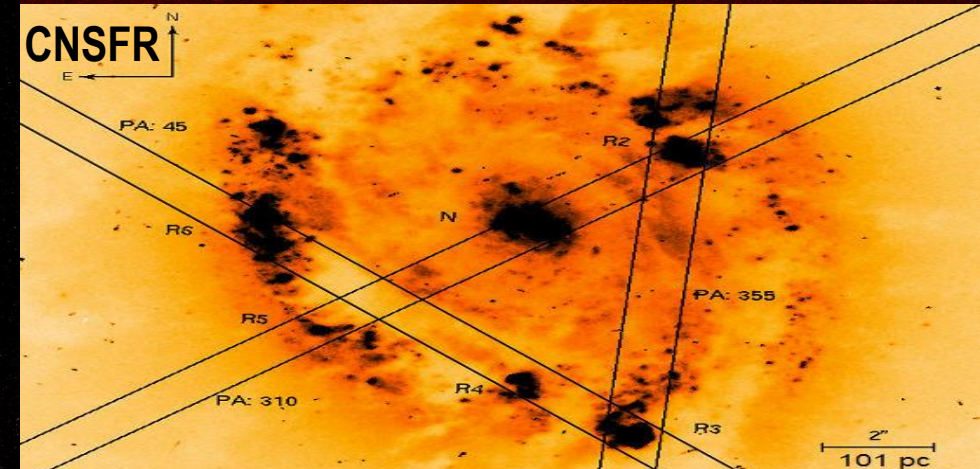
Sun



M51



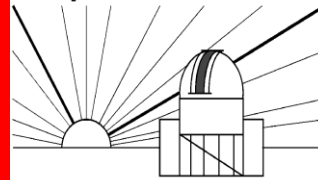
Rosette



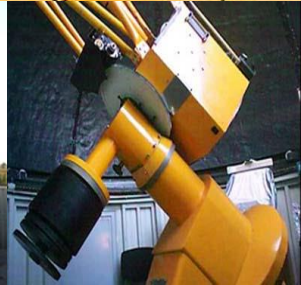
Mike Taylor, Pepe Vilchez
& Angeles Diaz

michael@damir.iem.csic.es

Grupo de Astrofísica

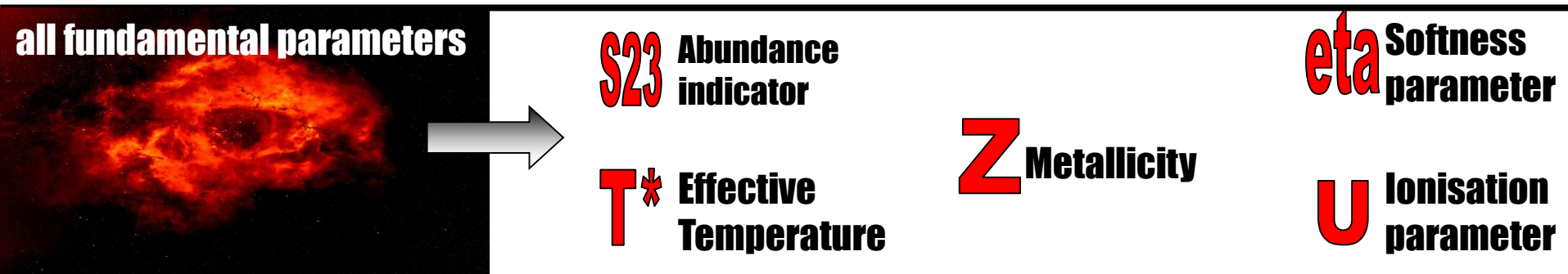
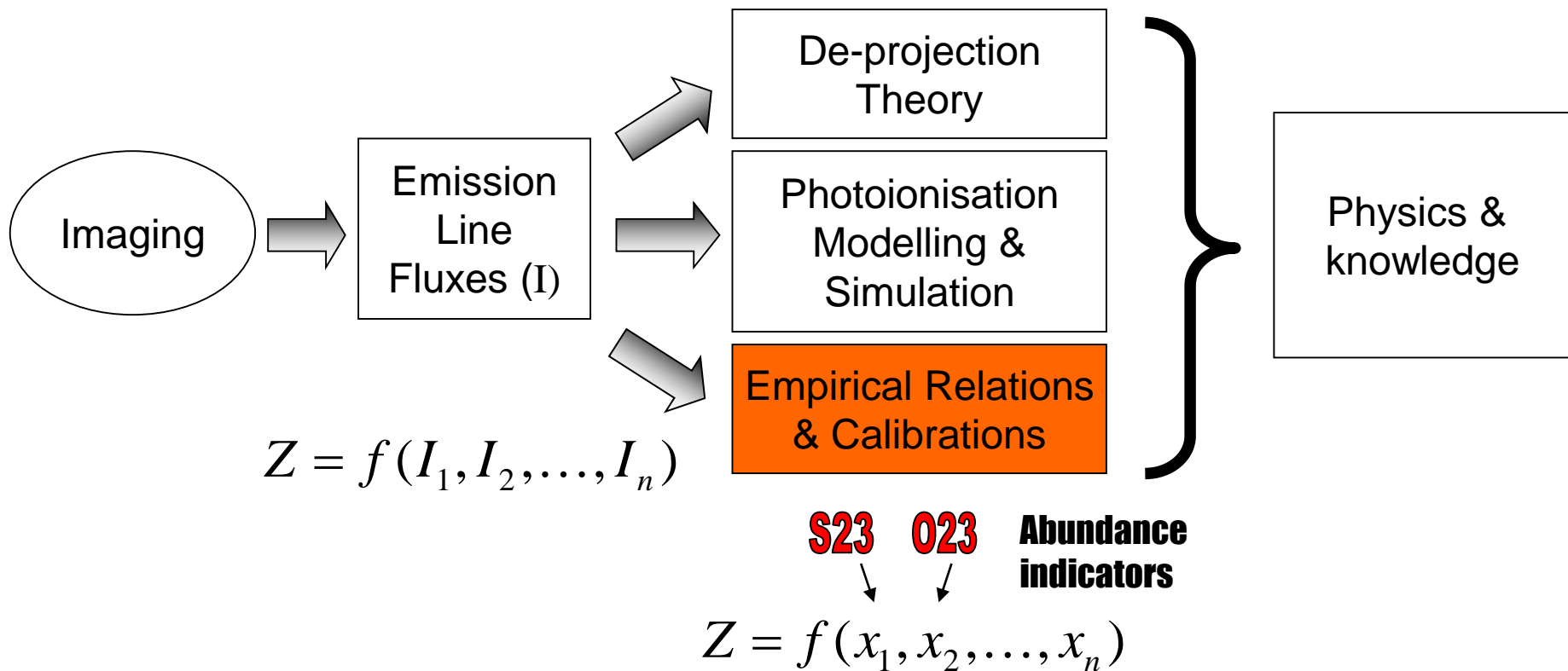


Universidad Autónoma de Madrid



Fundamental → Functional parameters

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



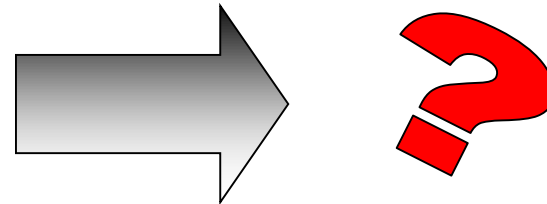
The **IDEAL** Indicator x_n ?

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

- Would be **MONOTONIC** with respect to metallicity **Z**
- Would be observationally **detectable** up to **high Z**
- Would have **low** average **dispersion** (least-square error)
- Would span a **W I D E** range of **Z** when empirically-calibrated using **REAL** galaxy data

& IF ALL THAT WENT PERFECTLY...

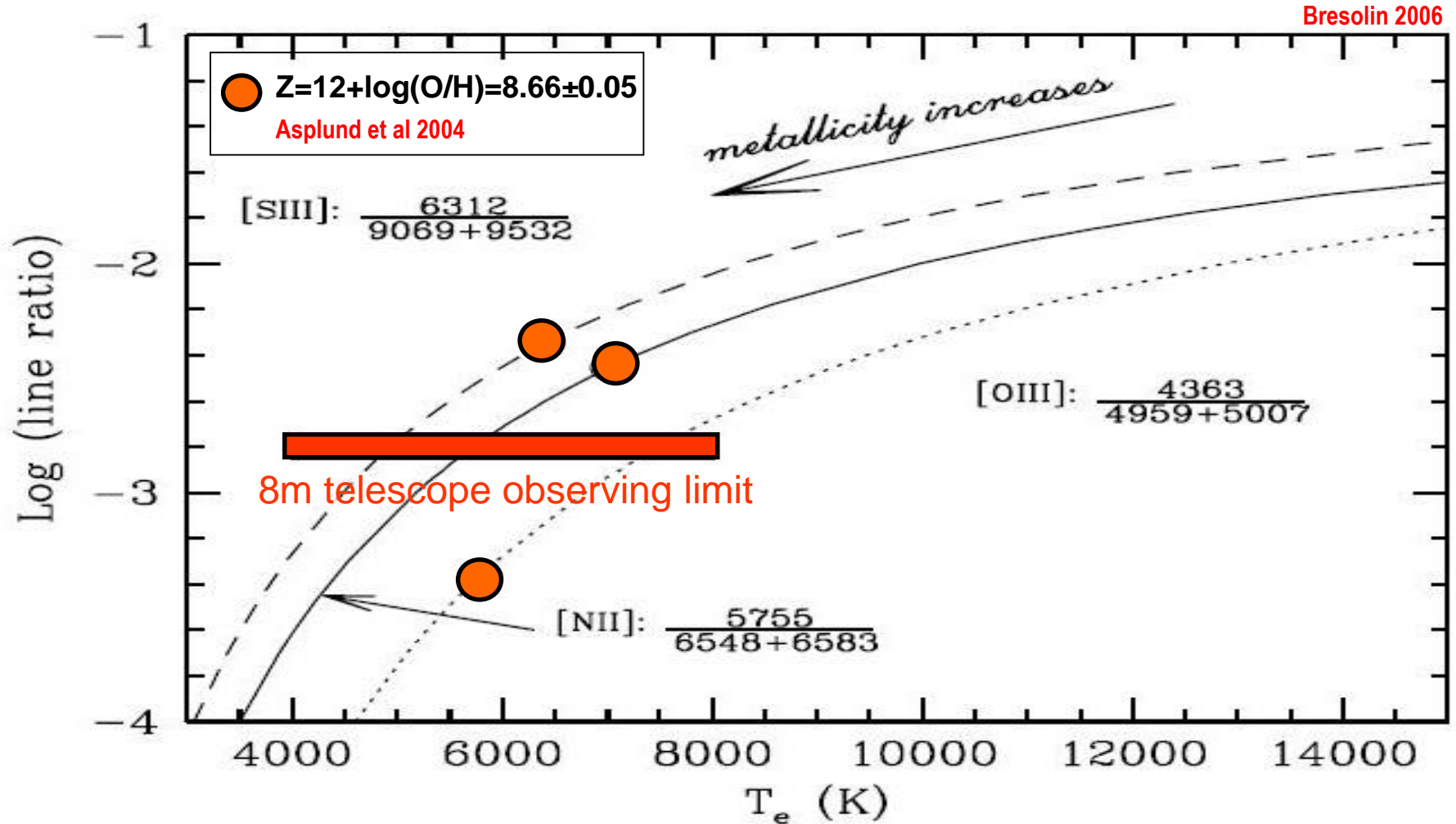
It would also be independent of chemical evolution and have an understandable behaviour with Z



Observational Limits at high Z

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

Commonly used auroral-to-nebular emission line ratios...



Indicators in HII-like regions

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

Galactic
HII-Regions

Extra-Galactic
HII-Regions

HII
Galaxies

CNSFRs?

Parameter

Validity

Dispersion

↑ O_{23}

$Z < 8.0$

0.12

≈ O_{23}

$8.0 \leq Z \leq 8.4$

1.30

↓ O_{23}

$Z > 8.4$

0.20

$N2$

up to Z_{\odot}

0.24

S_{23}

up to Z_{\odot}

0.16

S_{234}

up to Z_{\odot}

< 0.10

S_{23}/O_{23}

all Z

0.27

Perez-Montero 2002

AI: An opportunity

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



Scale-invariant And **Genetic Algorithm** Network (SAGAN)

Dimensional Analysis

Taylor & Diaz 2006



10^3 ERROR REDUCTION

Taylor 2005

Net DNA Ecoding

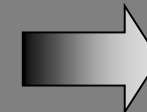
+ Genetic Operators

+ Pruning Algorithm

Architecture
Weights
Node
Functions
Mutation
Cross-over



EQUATIONS



Multi-Layer Perceptron

+ Back-Propagation

**UNIVERSAL FUNCTION
APPROXIMATOR**

Taylor 2005

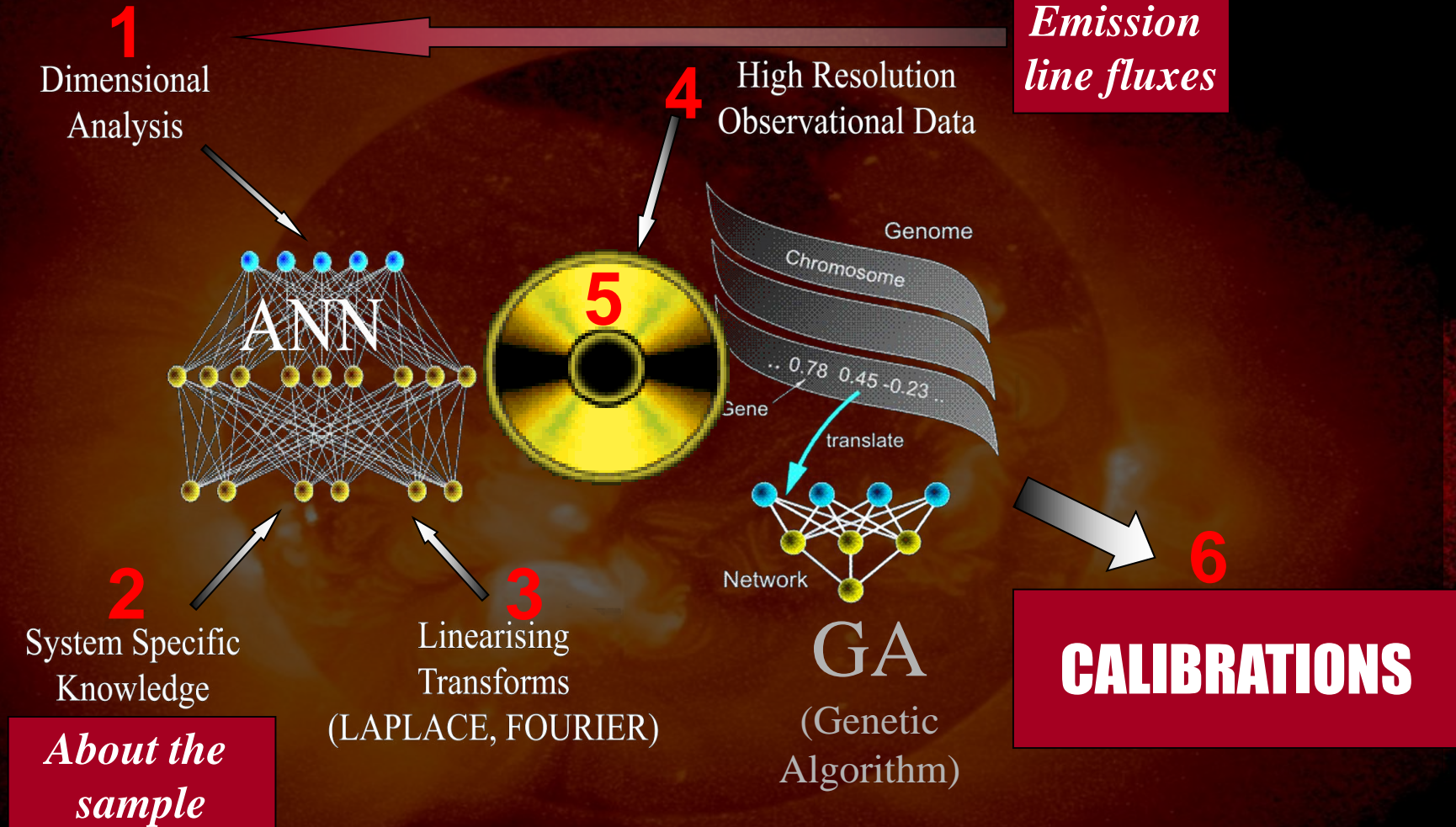
Unravelling the nonlinear physics of high-Z regions with neural networks

Mike Taylor, Pepe Vilchez
& Angeles Diaz

A new strategy

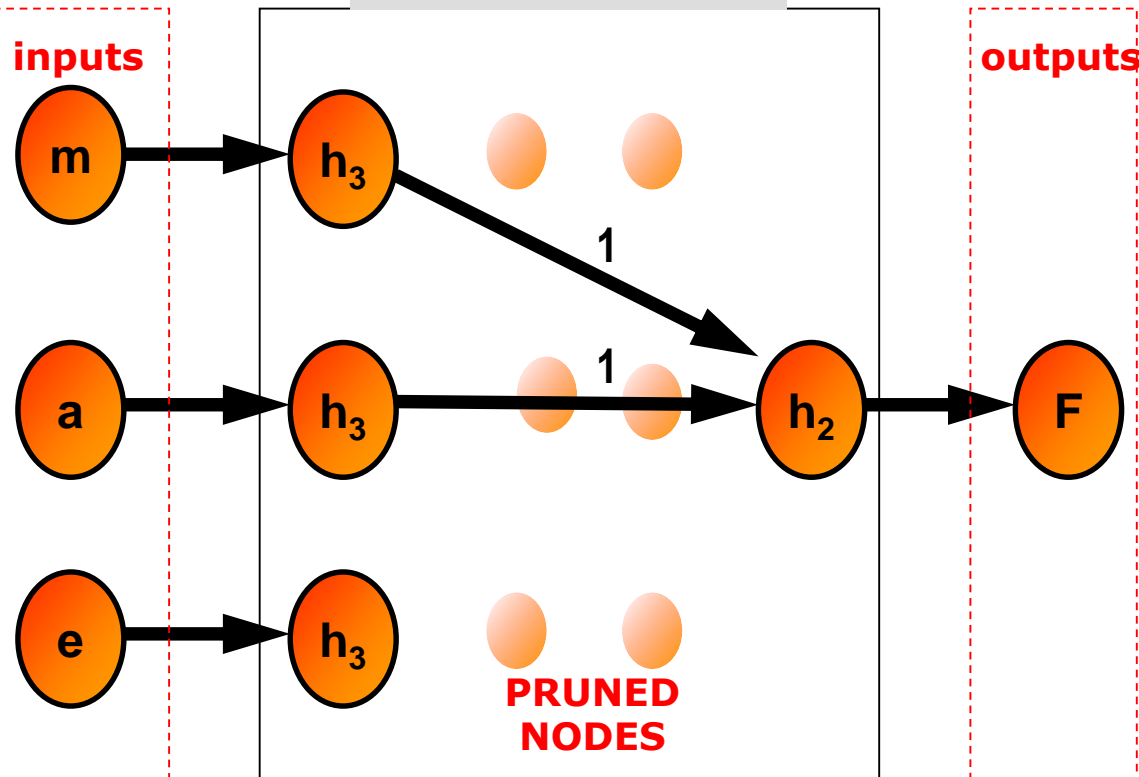
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

LINE RATIOS



How it Works: e.g. Newton's Law

IDEAL SOLUTION



$$F = e^{\{\ln m + \ln a\}} = e^{\ln ma} = ma$$

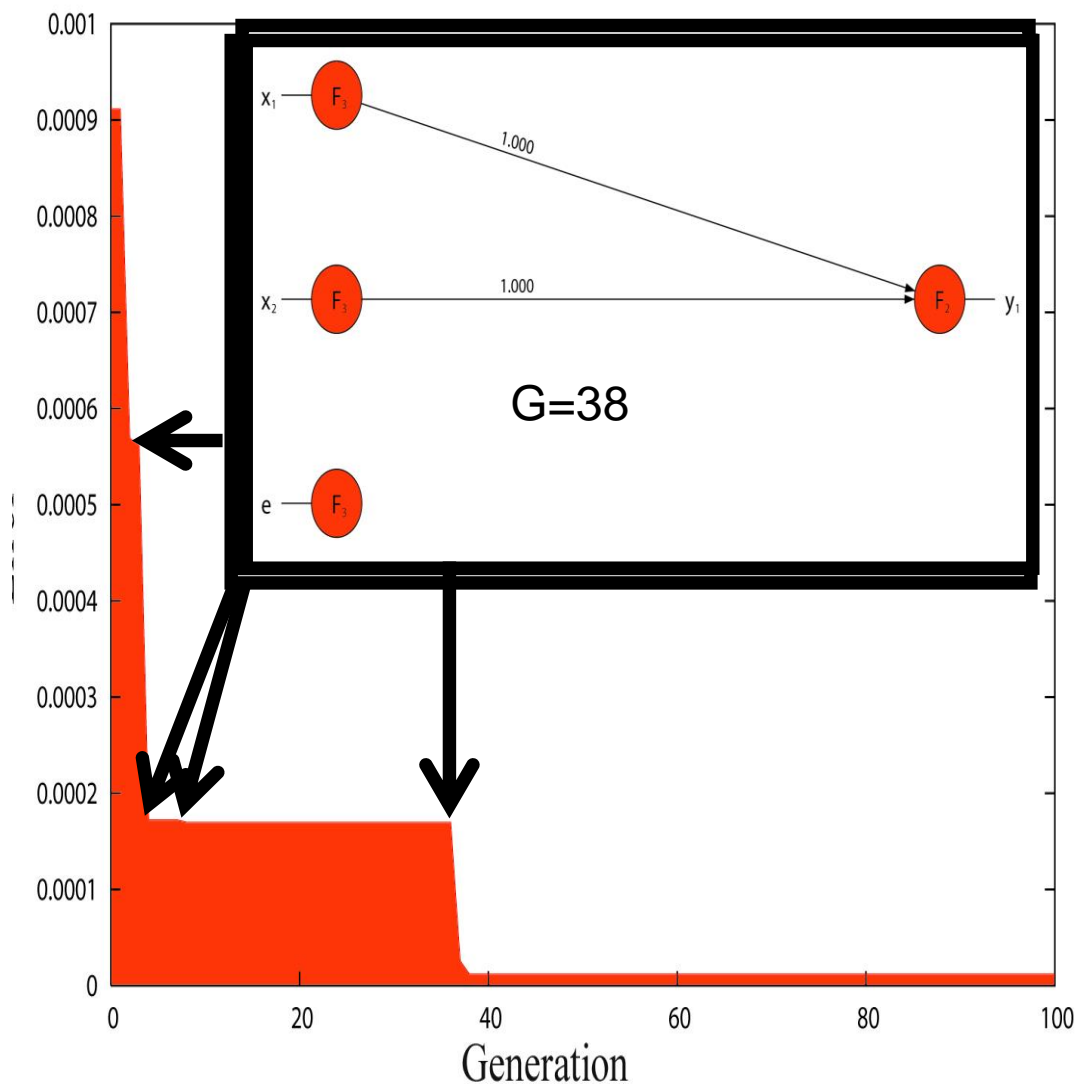
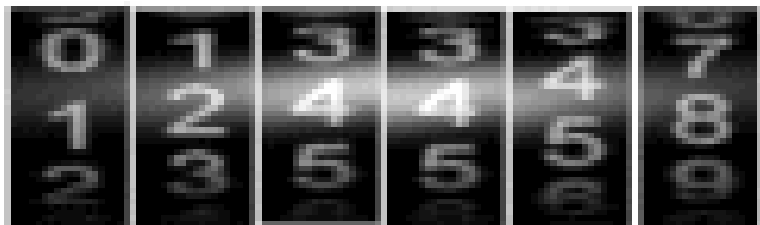
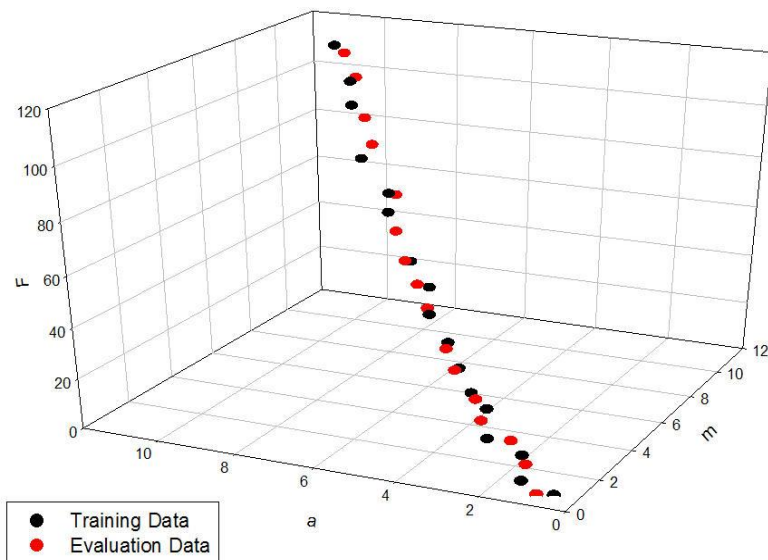
Neural Node Functions

$h(s)_1 = s$	Linear
$h(s)_2 = e^s$	Power Laws
$h(s)_3 = \ln s $	
$h(s)_4 = 1/(1 + e^{-s})$	
$h(s)_5 = (1 + e^{-s})/2(1 + e^{-s})$	Sigmoids
$h(s)_6 = \tanh s$	Trig. Functions
$h(s)_7 = \sin s$	
$h(s)_8 = \cos s$	
$h(s)_9 = e^{-s^2}$	Gaussians
$h(s)_{10} = 1$	Constants

Newton's Law: $F=ma$ (EXACT)

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

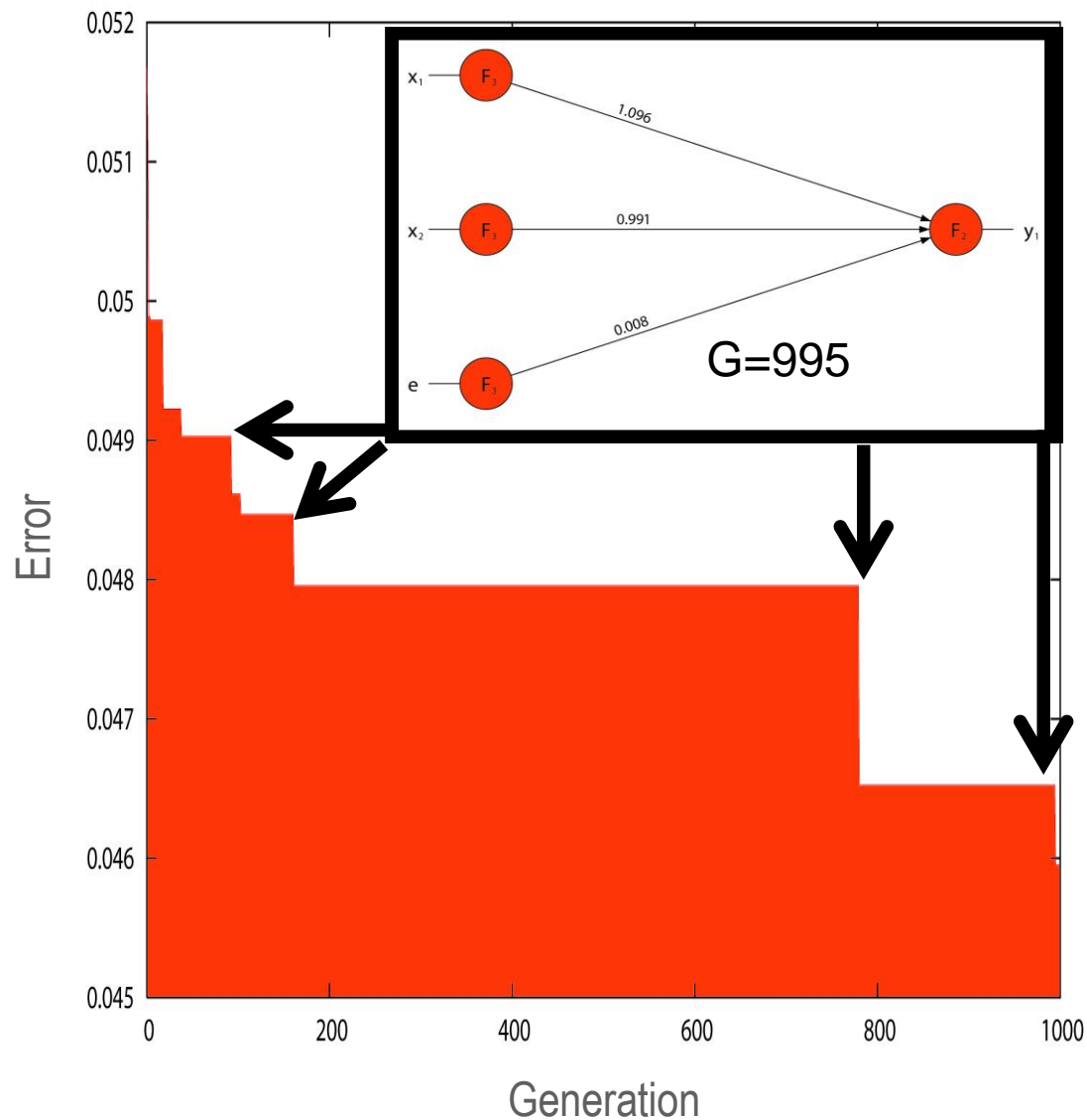
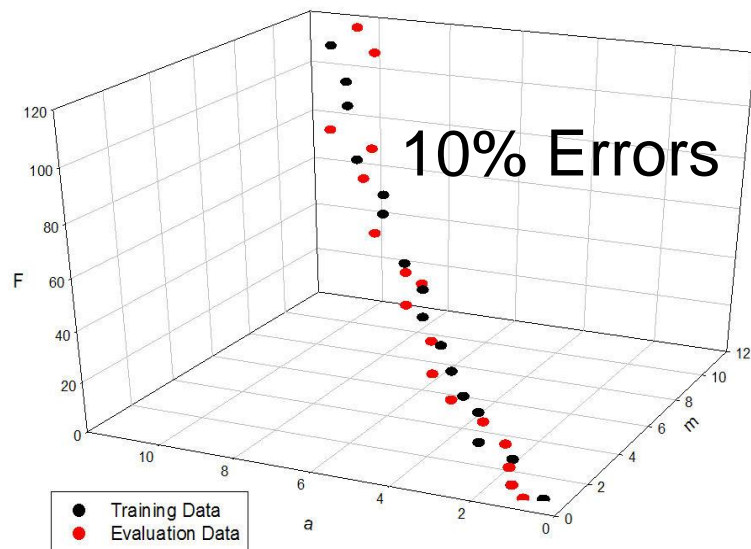
$F=ma$ (EXACT): Training & Evaluation Data



Newton's Law: $F=ma$ (10% Errors)

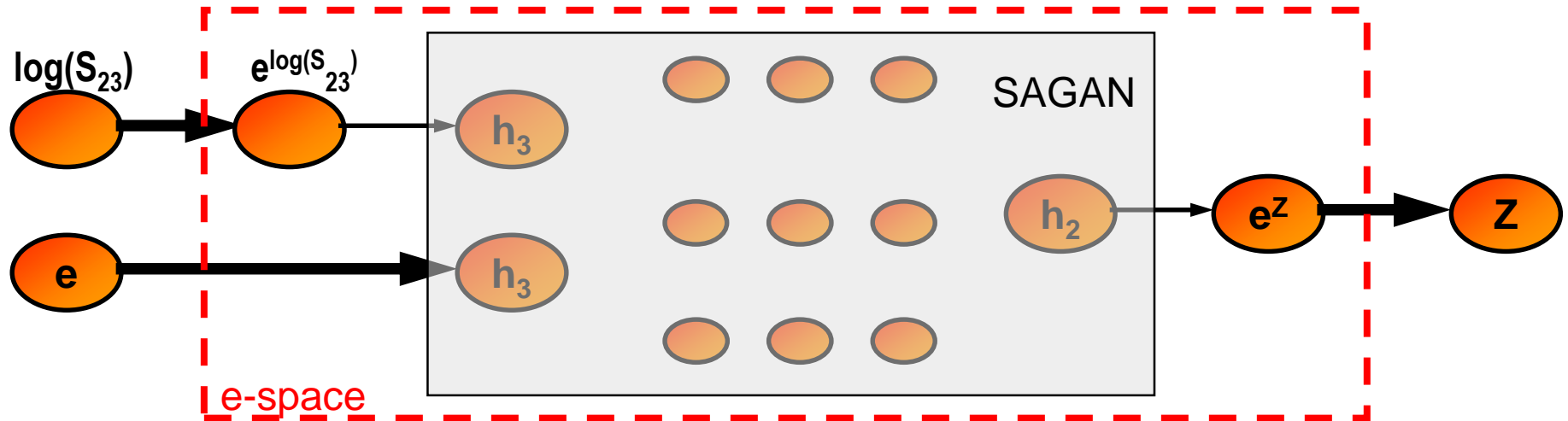
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

$F=ma$ (Errors): Training & Evaluation Data

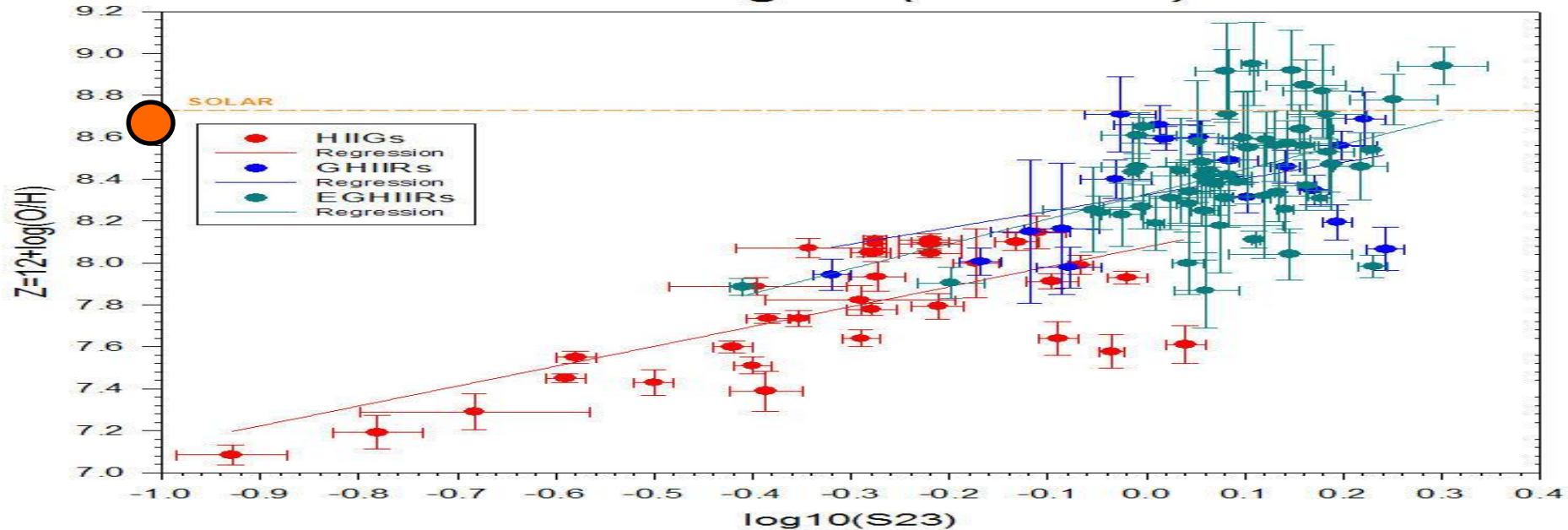


HII-like regions: HIIGs, GHIIRs & EGHIRs

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

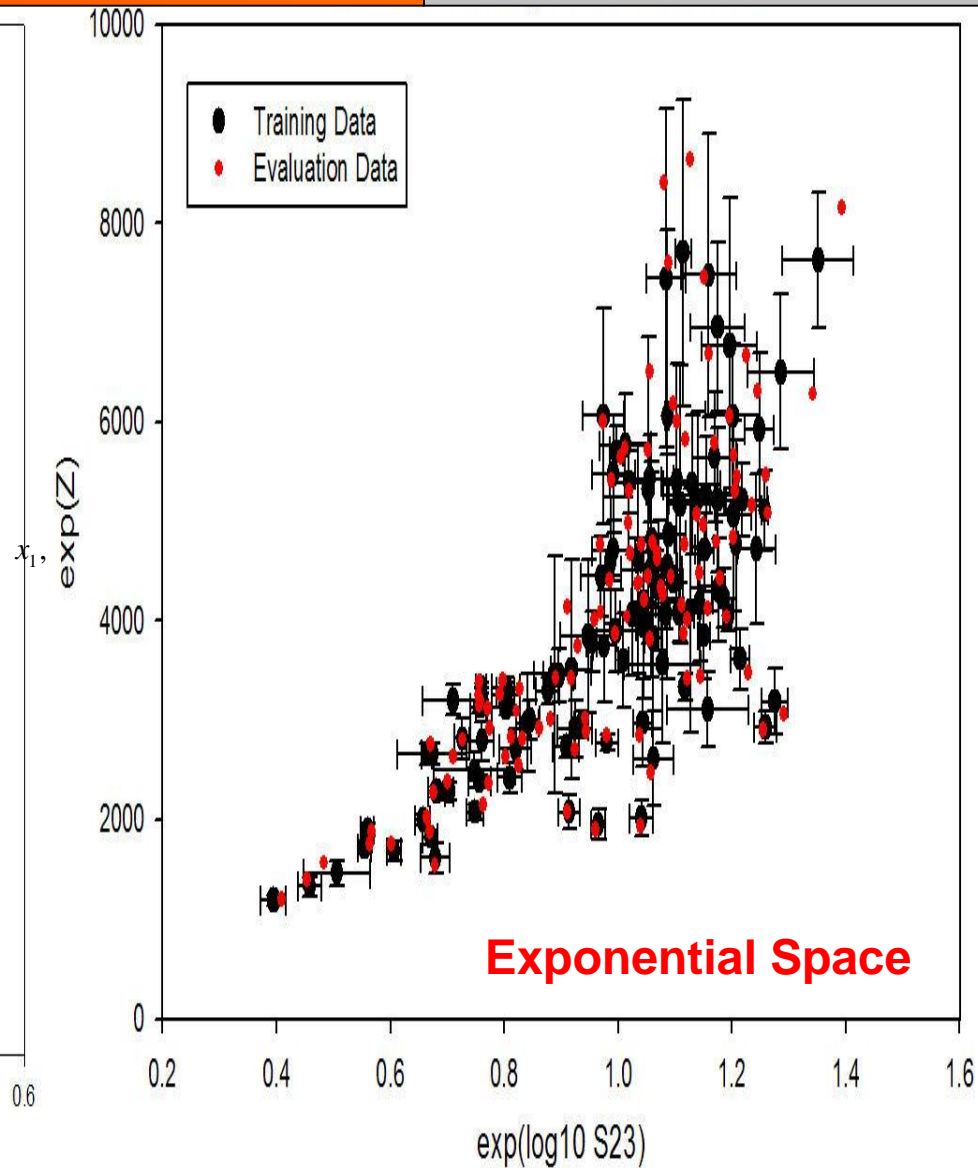
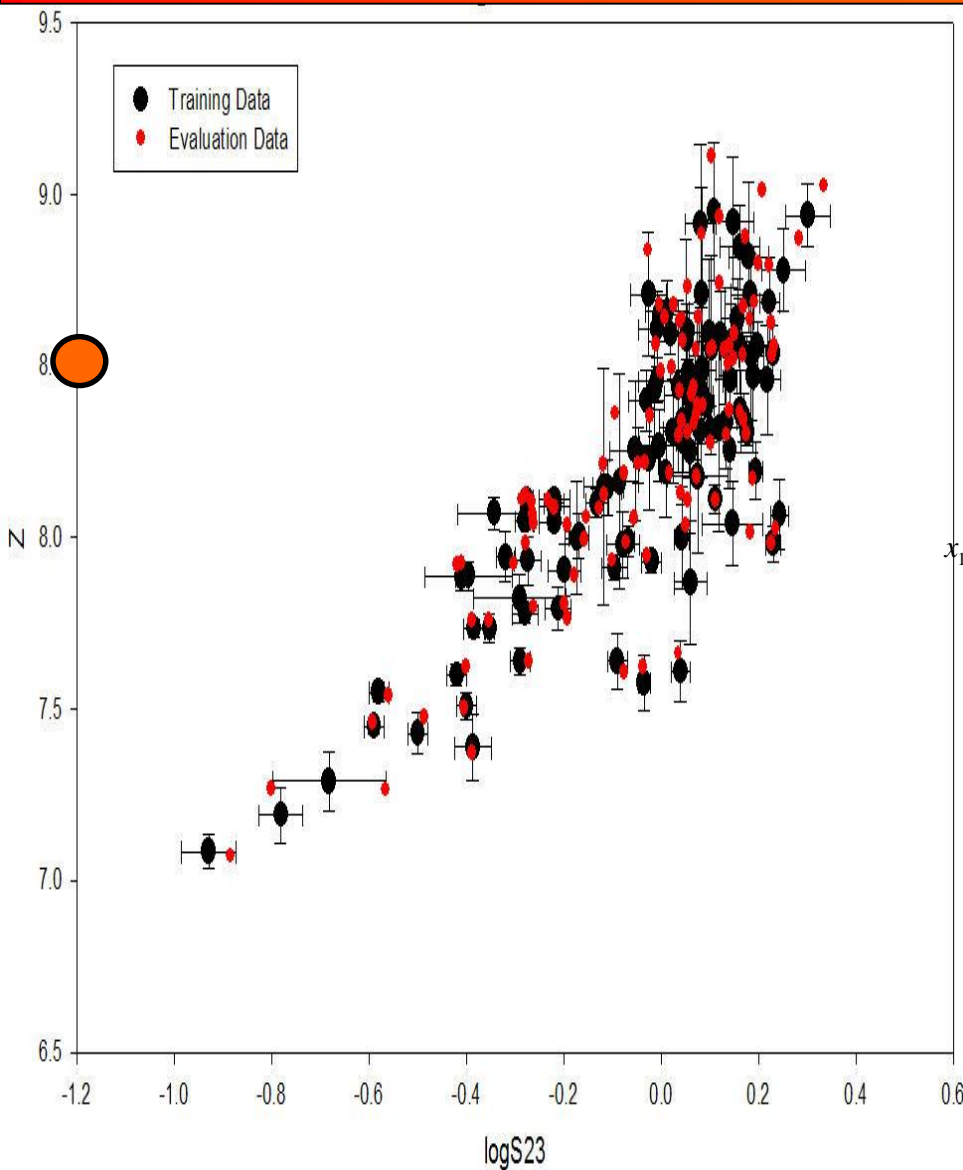


HII-like Regions (109 values)



Training & Evaluation Data

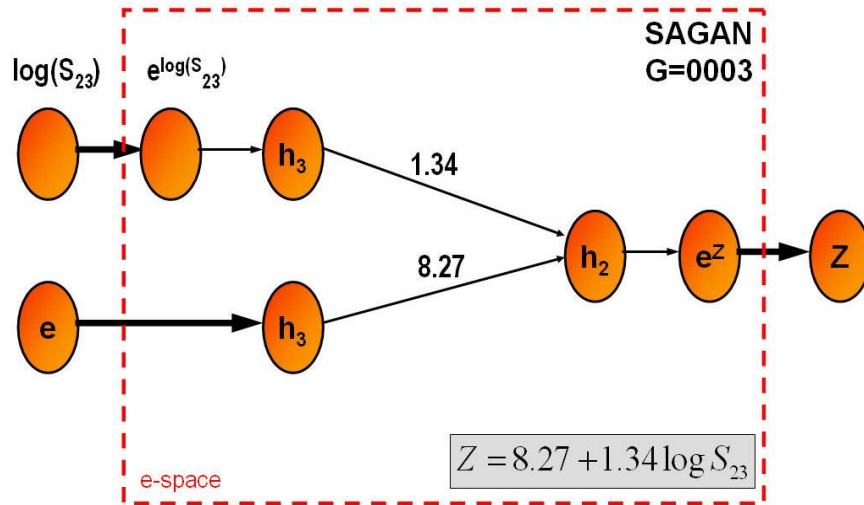
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



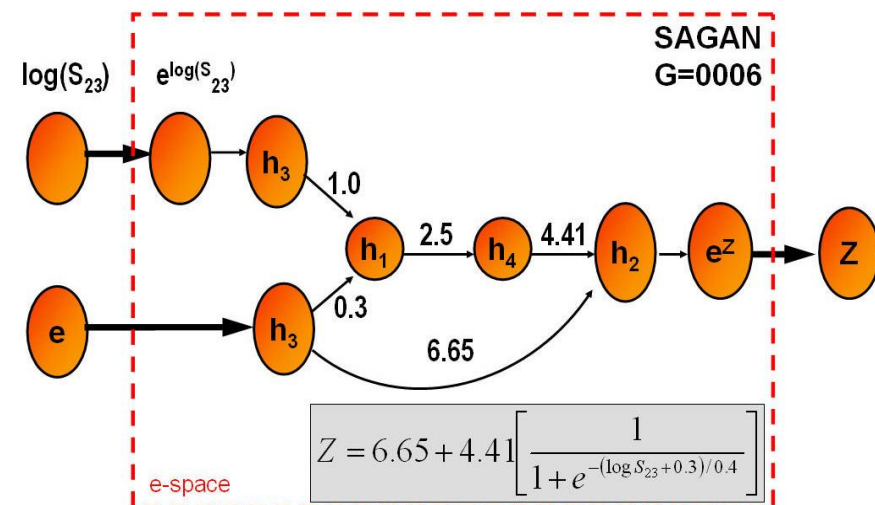
SAGAN Evolution (2 days on a dual 3Ghz PC)

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

a)

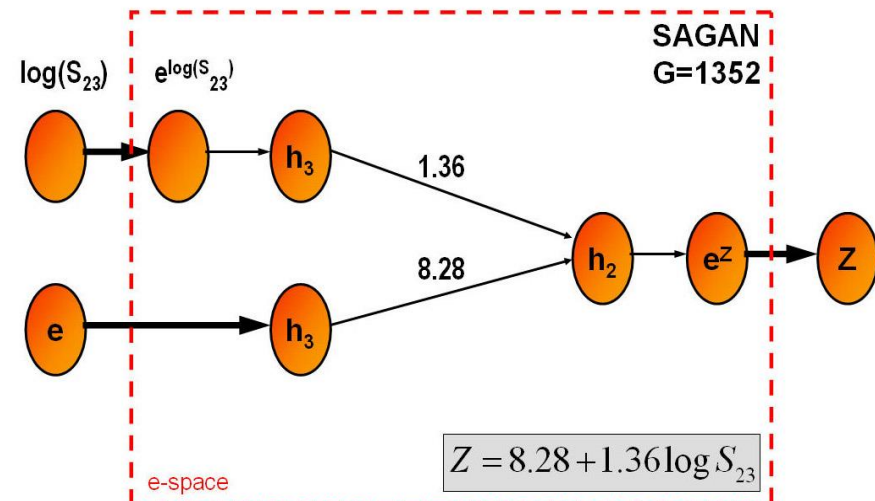
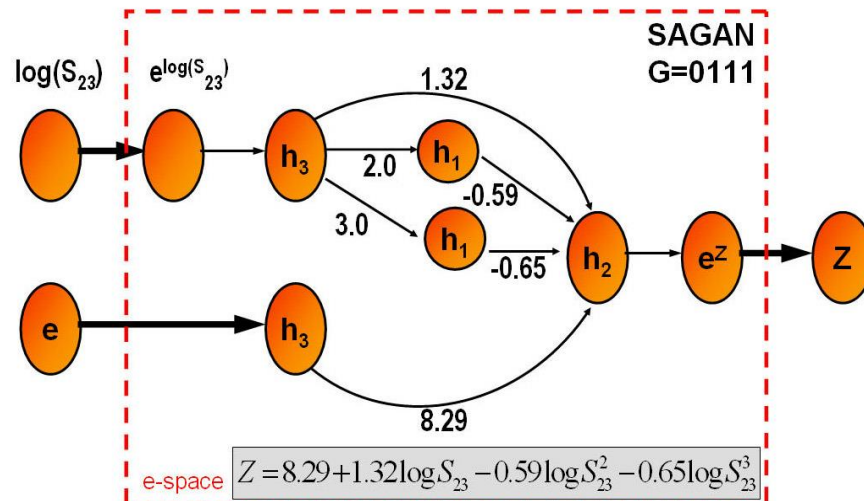


b)



Taylor & Diaz, 2007, PASP, 374, 104

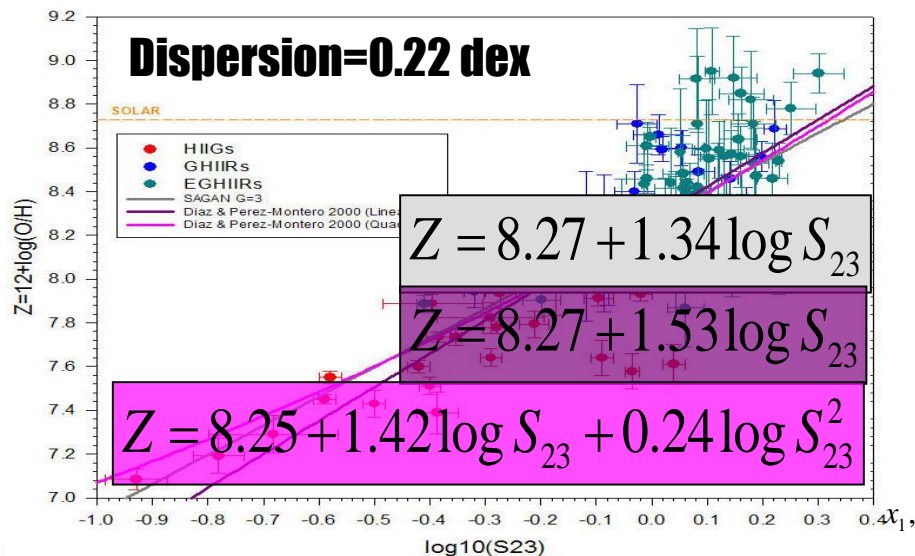
c)



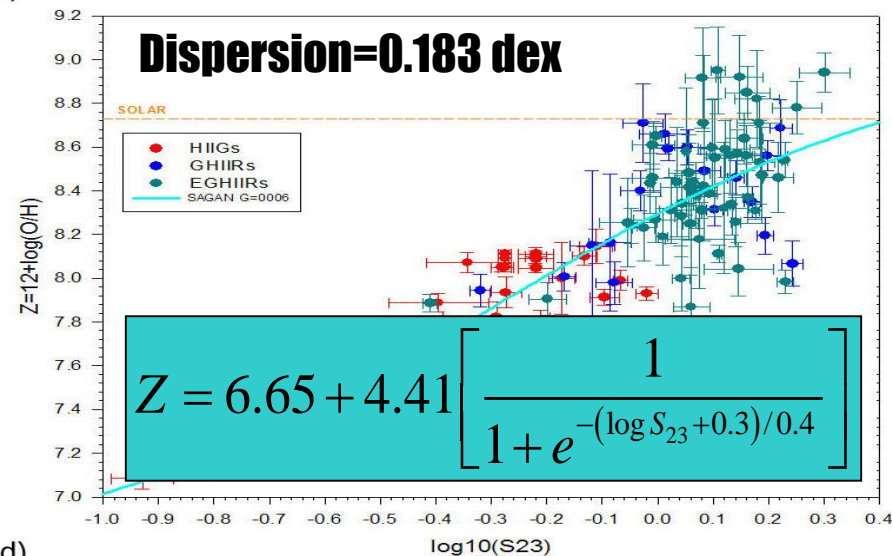
Evolution of the calibration: $Z=f(S23)$

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

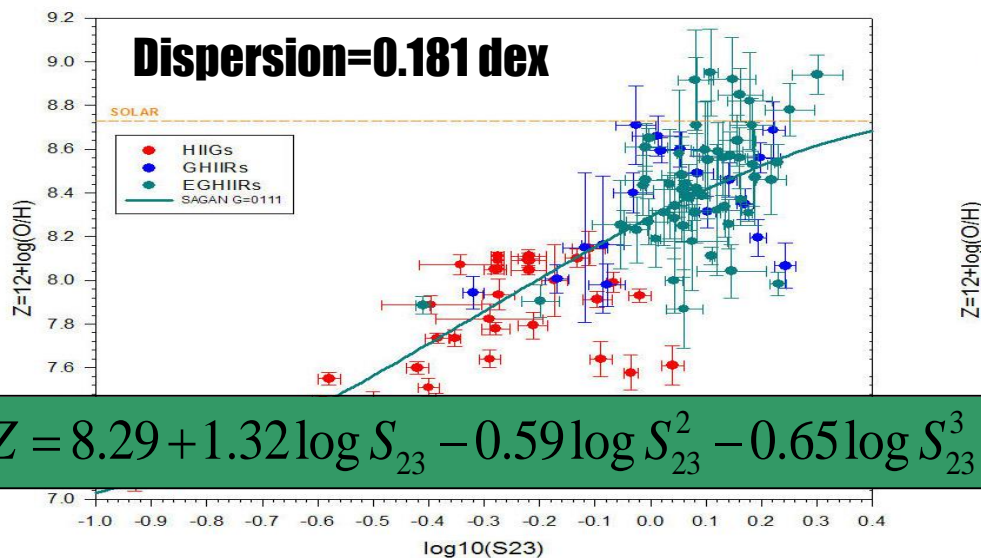
a)



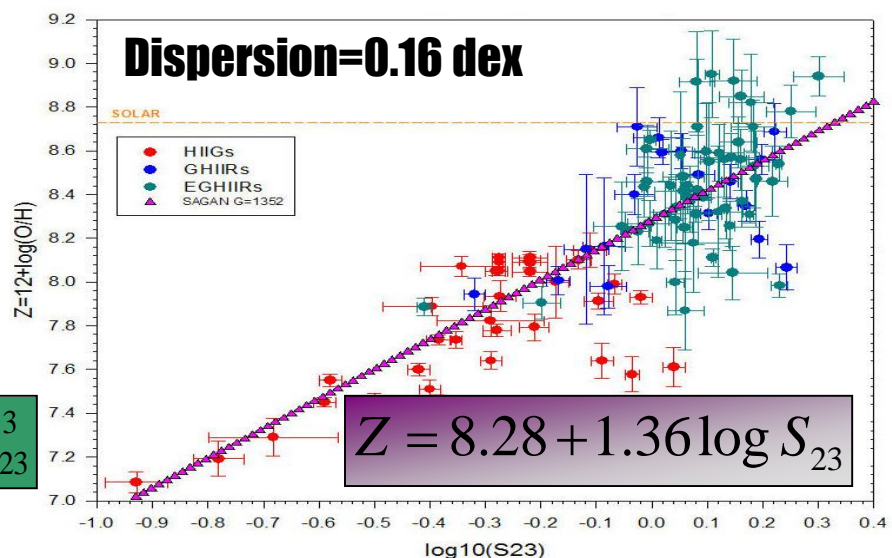
b)



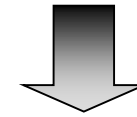
c)



d)



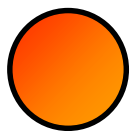
$$Z(O/H) = f(c, \underbrace{x_1, x_2, \dots, x_n}_{\text{Dimensionless Line Ratios}})$$



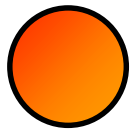
e.g. Dimensionless Line Ratios

$$x_1 = O_{23} = ([OII]_{\lambda 3727 + \lambda 3739} + [OIII]_{\lambda 4959 + \lambda 5007}) / H_{\beta}$$

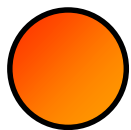
$$x_2 = S_{23} = ([SII]_{\lambda 6717 + \lambda 6731} + [SIII]_{\lambda 9069 + \lambda 9532}) / H_{\beta}$$



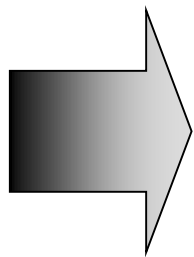
Needs a big and homogeneous data-set



Can potentially reduce subjectivity



Can potentially identify new indicators



A sample for n-D studies

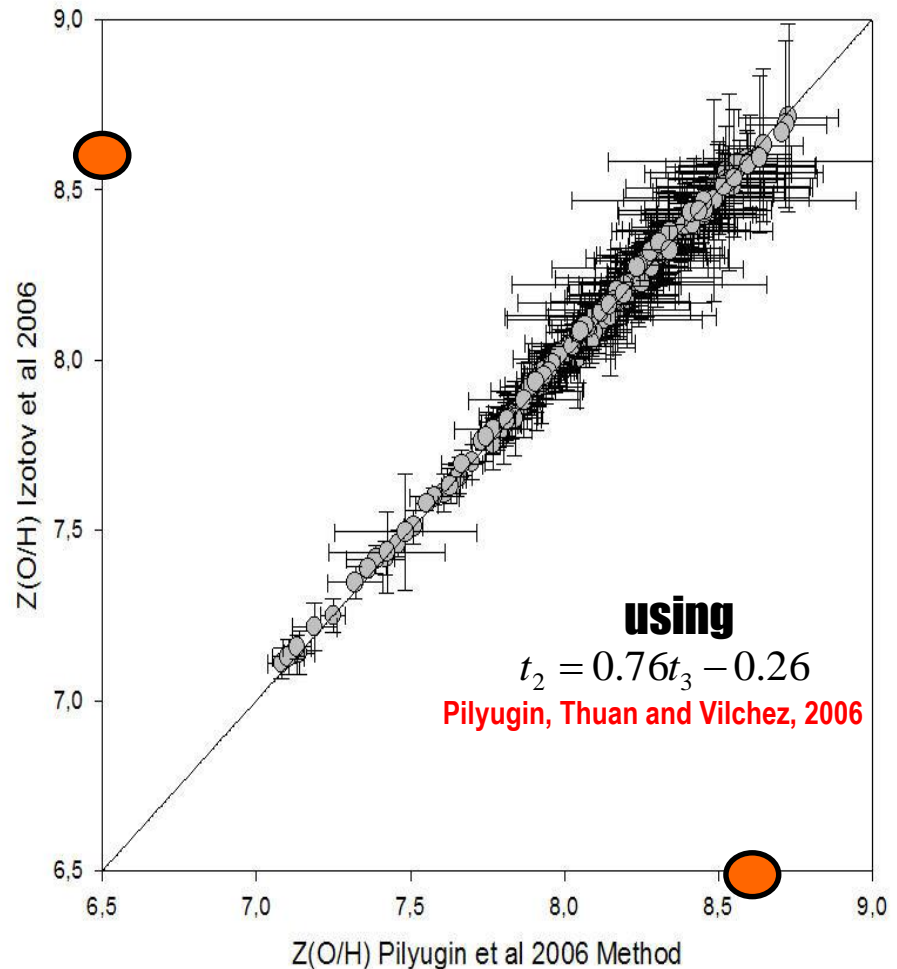
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

1 Izotov et al. 2006, A&A, 448, 955]

- 2 Díaz, Terlevich, Pagel, Vilchez & Edmunds, 1987, MNRAS, 226, 19.
- 3 Pastoriza, Dottori, Terlevich, Terlevich & Díaz, 1993, MNRAS, 260, 177.
- 4 González-Delgado et al., 1994, ApJ, 437, 239.
- 5 Castellanos, Díaz & Terlevich, 2002, MNRAS, 329, 315.
- 6 Skillman, Terlevich, Kennicutt, Garnett & Terlevich, 1994, ApJ, 431, 172
- 7 Skillman & Kennicutt, 1993, ApJ, 411, 655
- 8 González-Delgado et al., 1995, ApJ, 439, 604.
- 9 Bresolin, Schaerer, González-Delgado & Stasinska, 2005, A&A, 441, 981.
- 10 Lequeux, Peimbert, Rayo, Serrano & Torres-Peimbert, 1979, A&A, 80, 155.
- 11 Kunth & Sargent, 1983, ApJ, 273, 81
- 12 French, 1980, ApJ, 240, 41
- 13 Izotov & Thuan, 1998, ApJ, 500, 188.
- 14 Pagel, Simonson, Terlevich & Edmunds, 1992, MNRAS, 255, 325.
- 15 Dinerstein & Shields, 1986, ApJ, 311, 45.
- 16 Kennicutt, Bresolin & Garnett, 2003, ApJ, 591, 801.
- 17 Garnett & Kennicutt, 1994, ApJ, 426, 123.
- 18 Shields & Searle, 1978, ApJ, 222, 821
- 19 Vilchez, Pagel, Díaz, Terlevich & Edmunds, 1988, MNRAS, 235, 633
- 20 Kwitter & Aller, 1981, MNRAS, 195, 939.
- 21 Vilchez, Pagel, Díaz, Terlevich & Edmunds, 1988, MNRAS, 235, 633
- 22 Garnett, Kennicutt & Bresolin, 2004, ApJ, 607, L21.
- 23 Bresolin, Garnett & Kennicutt, 2004, ApJ, 615, 228.
- 24 Peimbert, Peña & Torres-Peimbert, 1986, A&A, 158, 266
- 25 Garnett, Shields, Skillman, Sagan & Dufour, 1997, ApJ, 489, 63.
- 26 Díaz, Terlevich, Pagel, Vilchez & Edmunds, 1987, MNRAS, 226, 19.
- 27 Vermeij, Damour, van der Hulst & Baluteau, 2002, A&A, 390, 649
- 28 Vilchez & Esteban, 1996, MNRAS, 280, 720
- 29 Dennefeld & Stasińska, 1983, A&A, 118, 255
- 30 Hägele, Pérez-Montero, Díaz, Terlevich & Terlevich, 2006, MNRAS,
- 31 Kehrig, Vilchez, Telles, Cuisinier & Pérez-Montero, 2006 A&A
- 32 Izotov, Thuan & Lipovetsky, 1994, ApJ, 435, 647
- 33 Terlevich, Melnick, Masegosa, Moles & Copetti, 1991, A&ASS, 91, 285
- 34 Guseva, Izotov & Thuan, 2000, ApJ, 531, 776.
- 35 Kniazev et al, 2004, ApJS, 153, 429K.

1	H12
2	H11
3	H10
4	H9
5	H-delta
6	H-gamma
7	H-beta
8	H-alpha
9	[He I] 5876
10	[He II] 4686
11	[O II] 3727+ 3739
12	[O III] 4363
13	[O III] 4959
14	[O III] 5007
15	[O I] 6300
16	[O II] 7325
17	[N II] 6548
18	[N II] 6584
19	[S II] 6312
20	[S II] 6717 + 6731
21	[S III] 9069
22	[S III] 9532
23	[Ne III] 3868
24	[Fe III] 4658
25	[Fe III] 4988
26	[Ar IV] 4740
27	[Ar IV] 7136

Izotov et al 2006 → 144 objects
(another 163 from other sources)



Data classification with the BPT plot

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

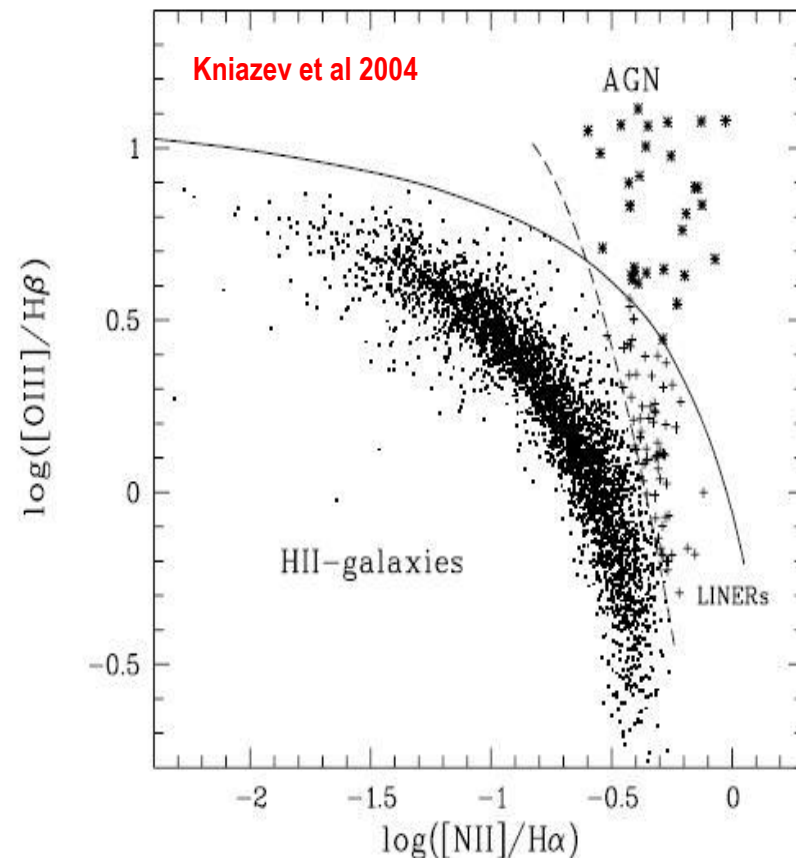
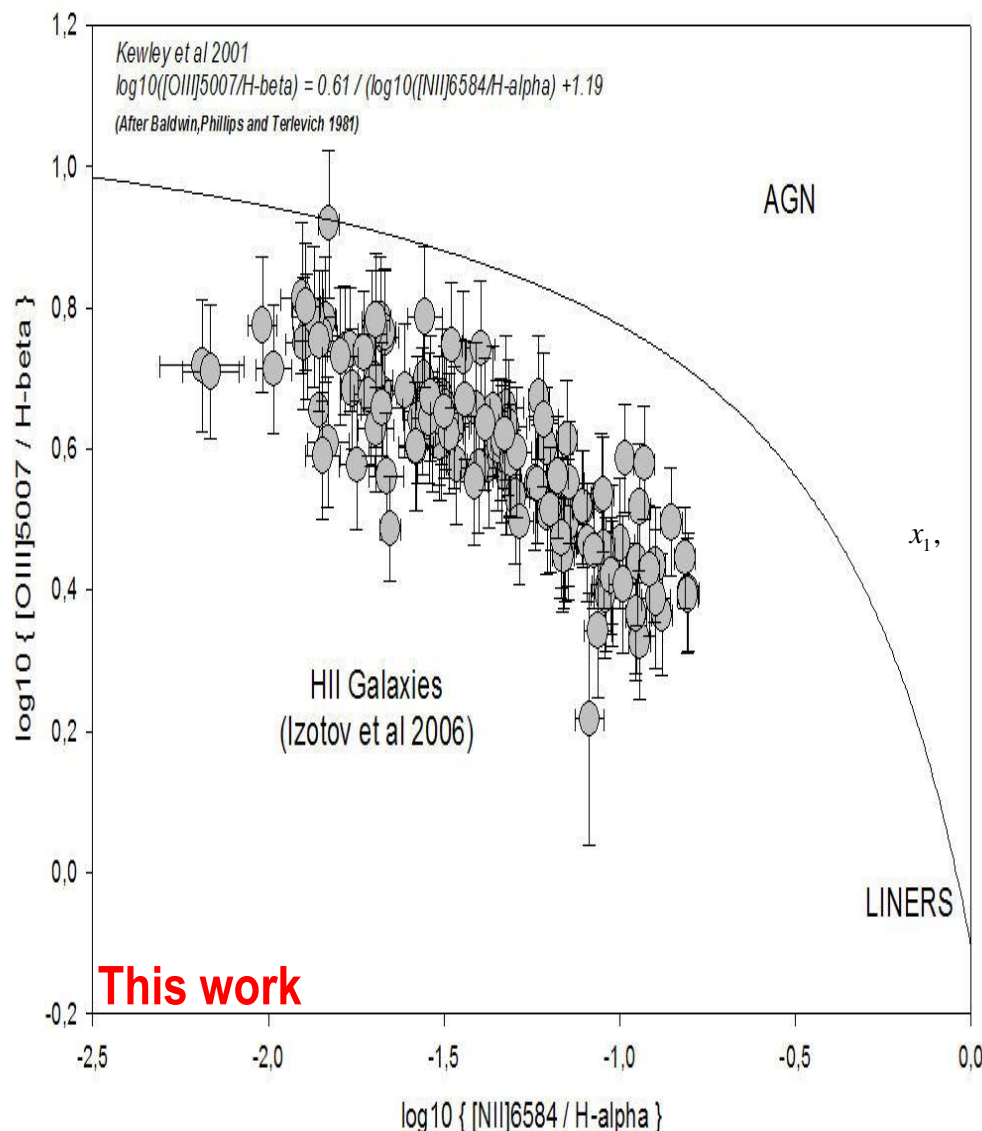
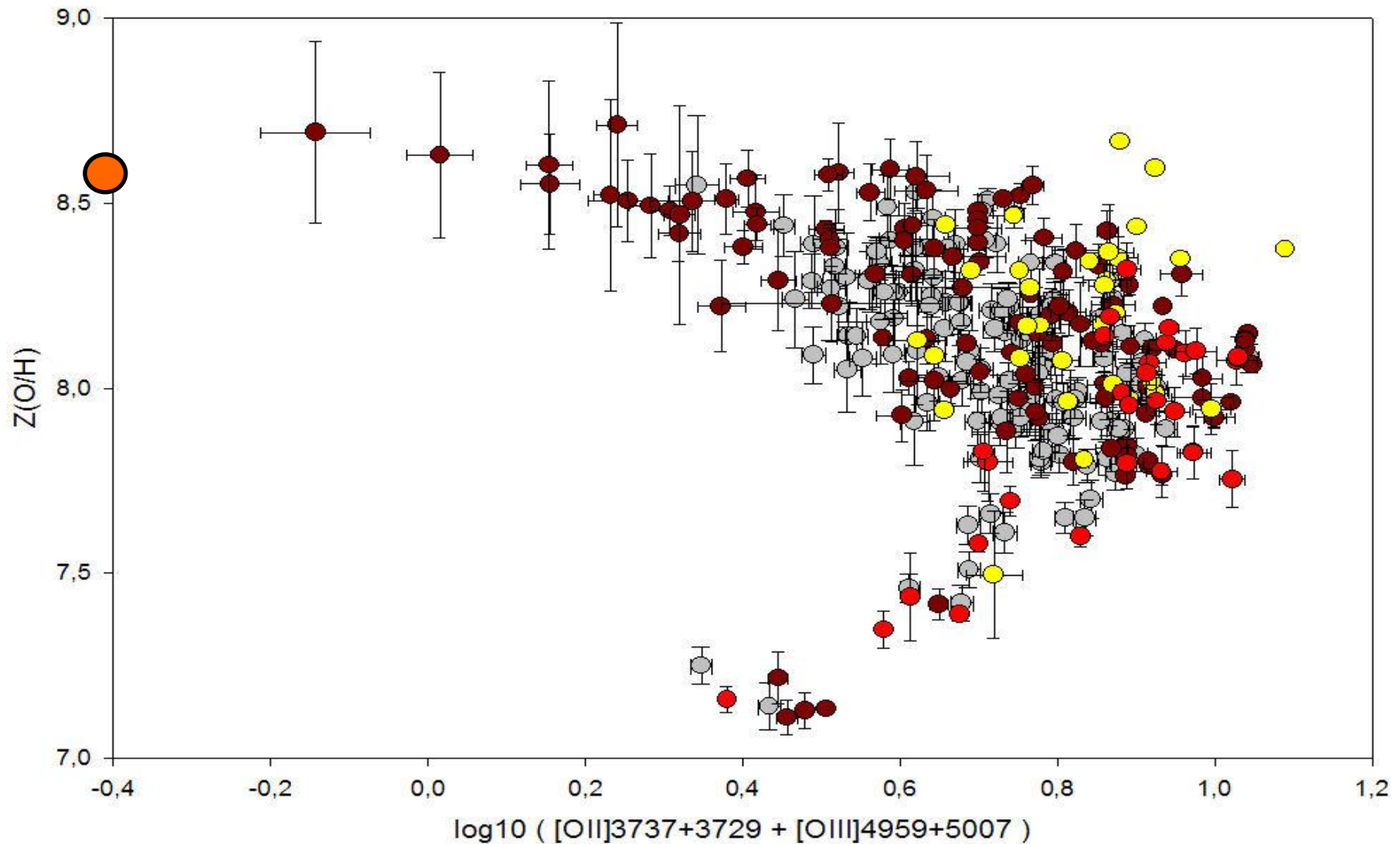


FIG. 2.—Classification diagram for all preselected ELGs (~ 5000 spectra). Galaxies identified as AGNs are shown as asterisks, and galaxies identified as LINERS are shown as crosses. The other ELGs are plotted as filled circles. The dashed line separates regions of H η -type and AGN/LINER spectra following Veilleux & Osterbrock (1987) and Baldwin et al. (1981). The solid line shows models from Kewley et al. (2001) that were used for AGN/LINER separation.

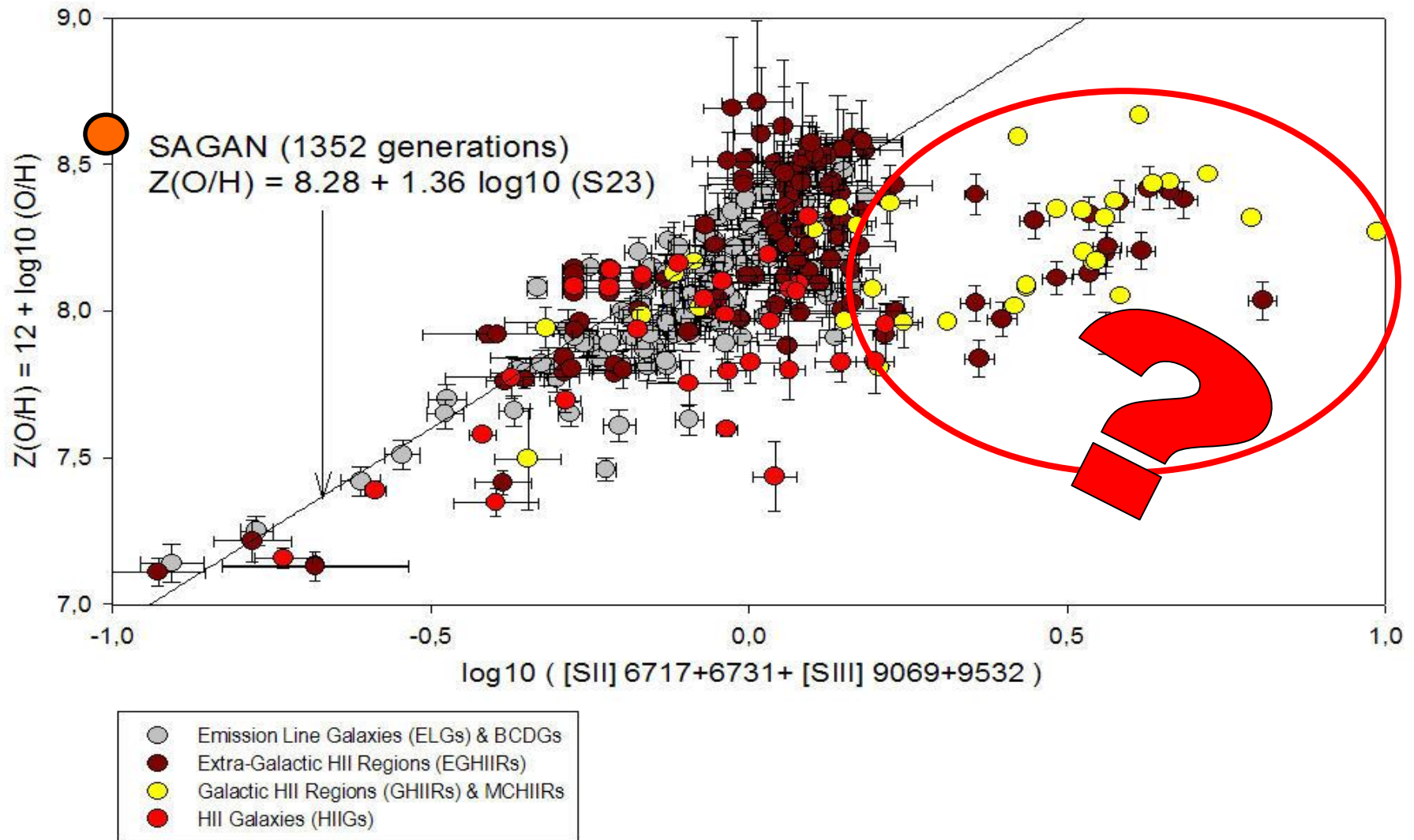
Metallicity and O_{23}

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



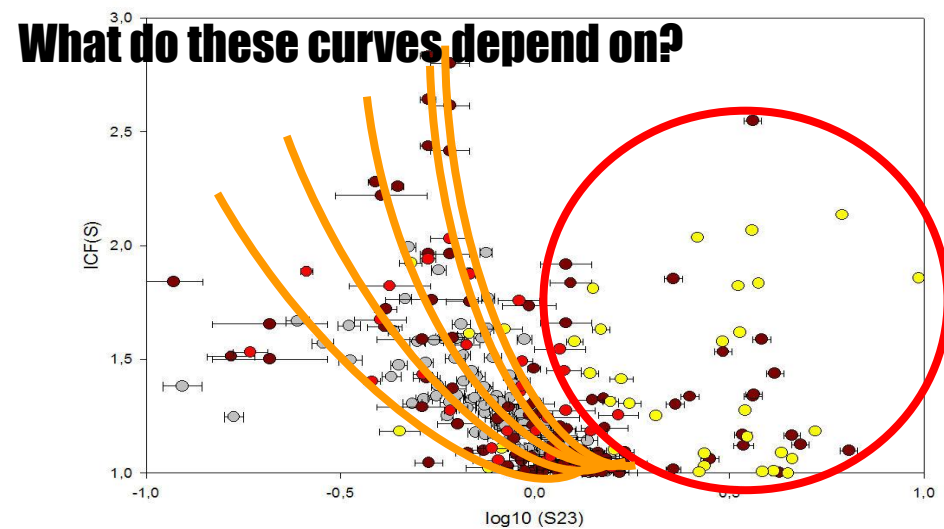
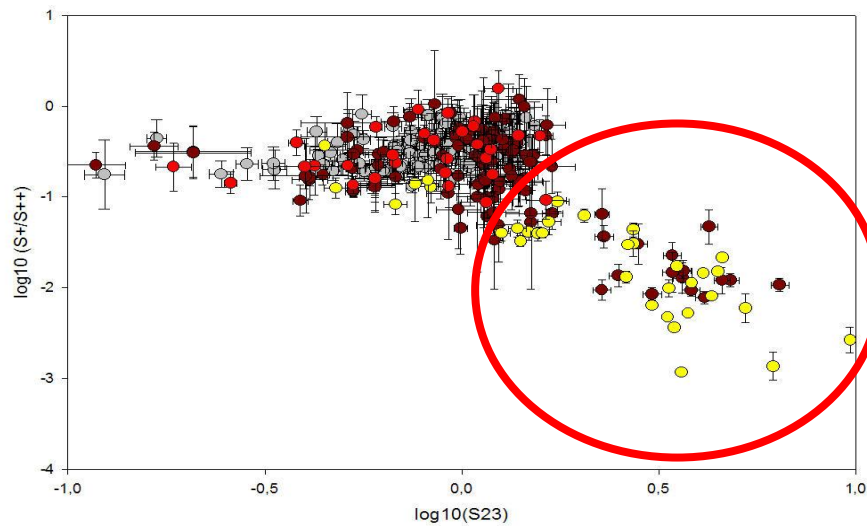
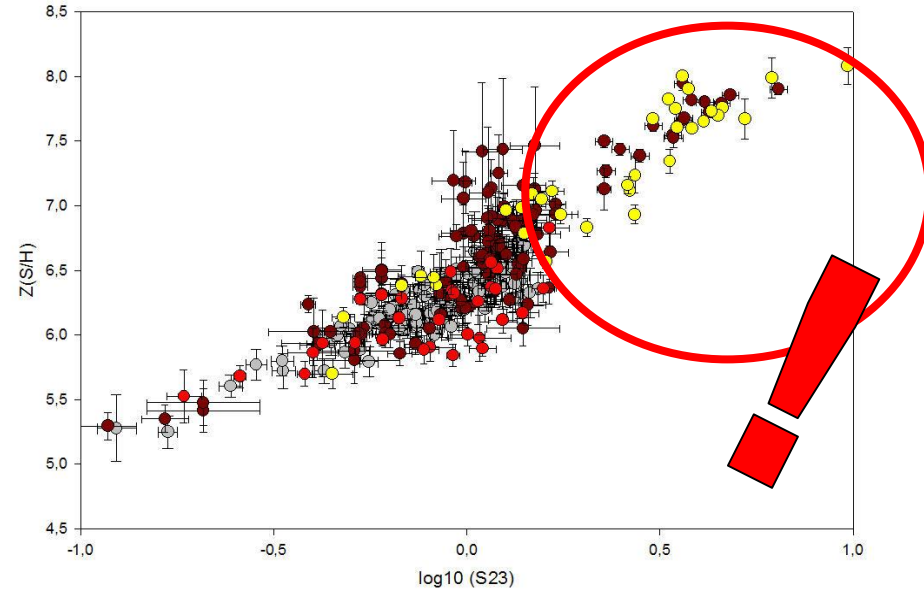
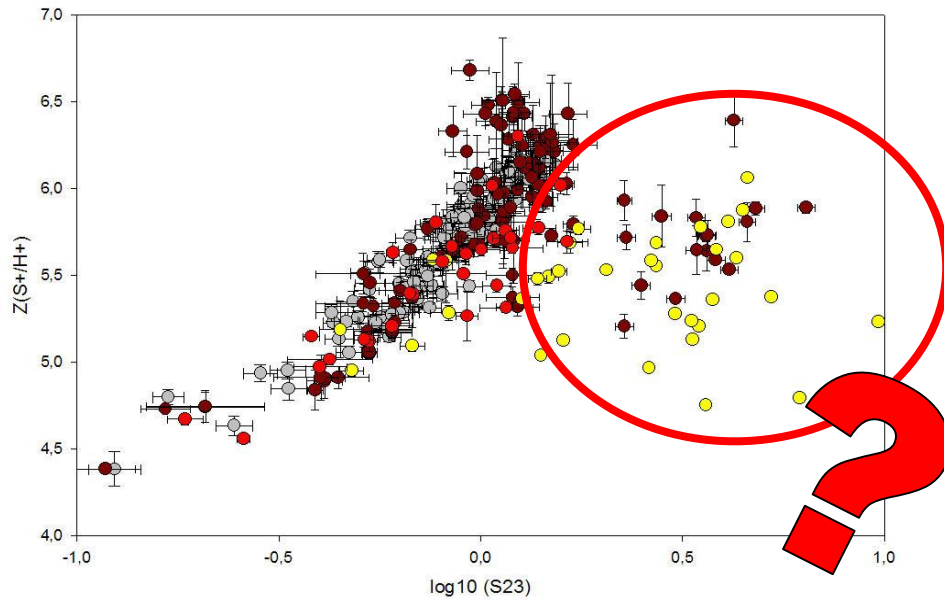
Metallicity and S_{23}

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



The role of S (S+, S2+, S, ICF)

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



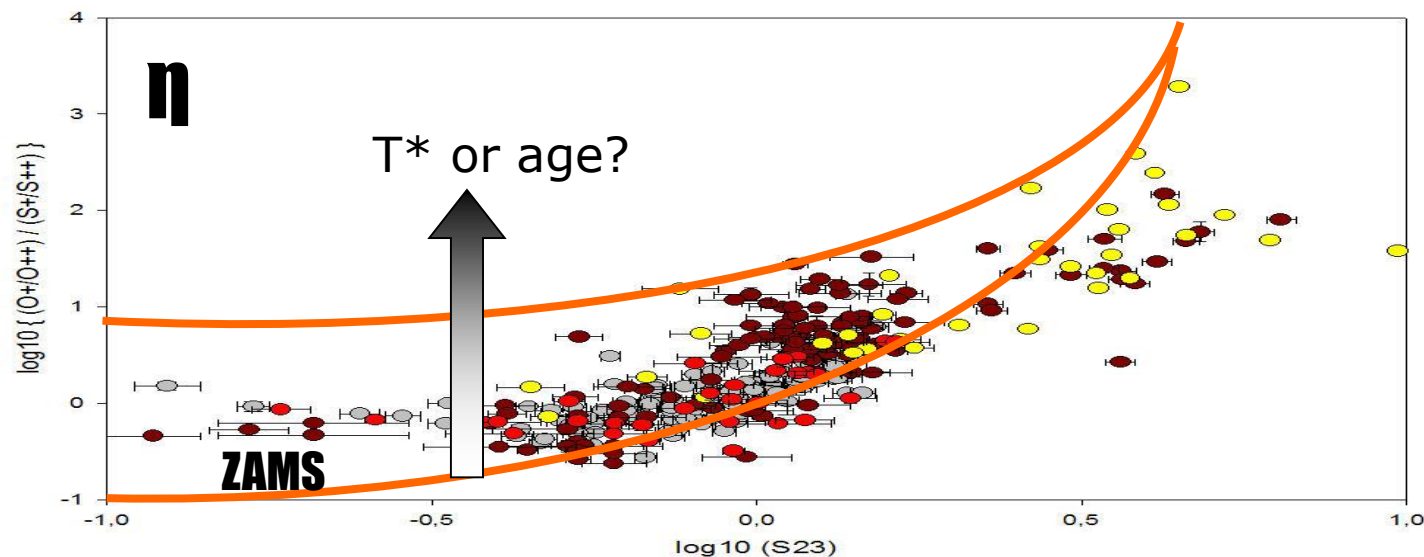
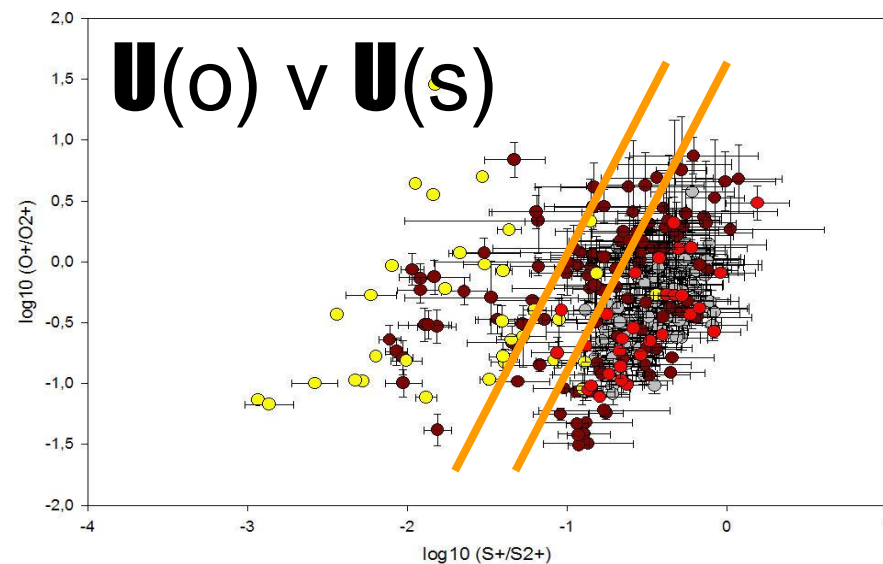
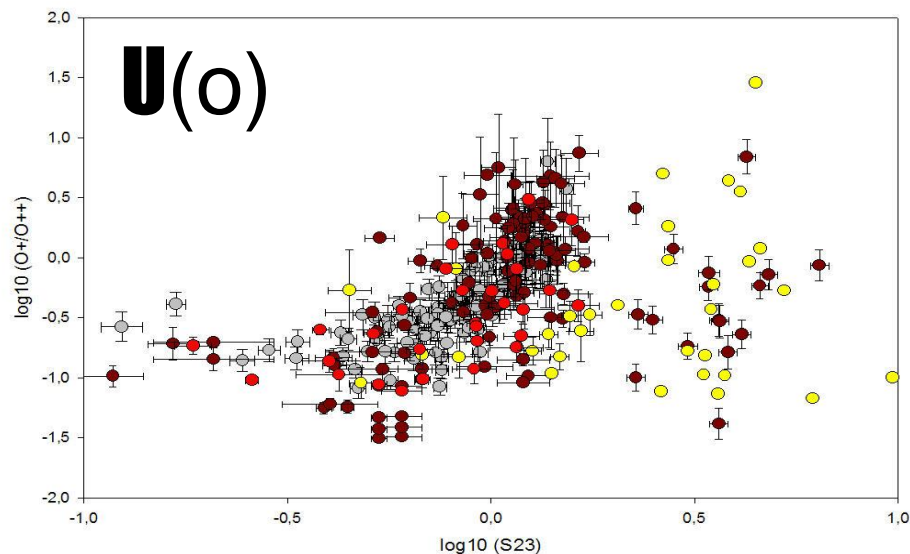
What do these curves depend on?

Unravelling the nonlinear physics of high-Z regions with neural networks

Mike Taylor, Pepe Vilchez
& Angeles Diaz

Trends of U and η with S_{23}

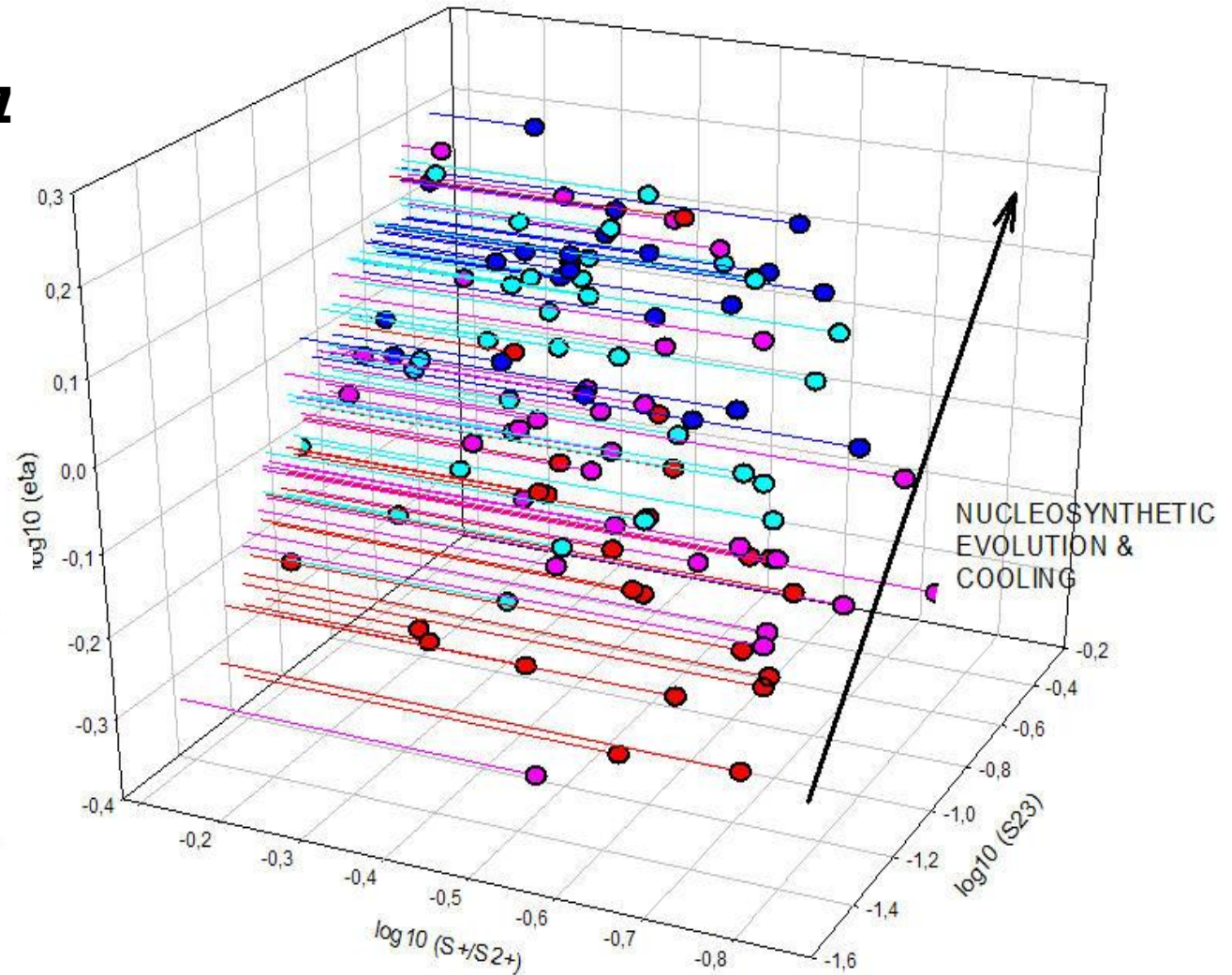
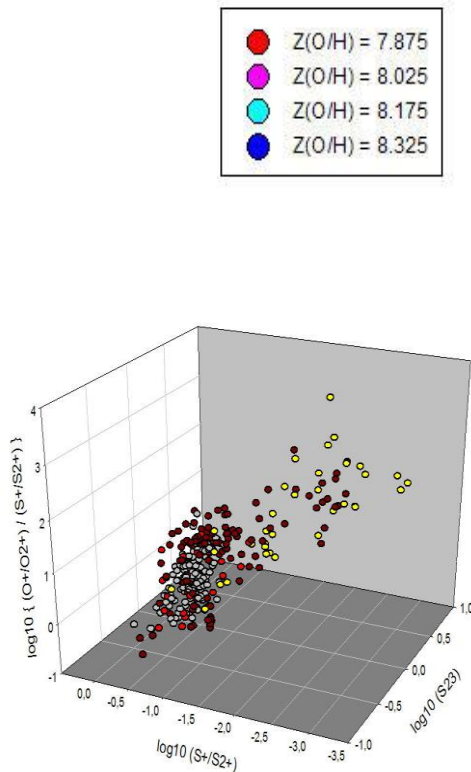
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



Data in the η -U- S_{23} space

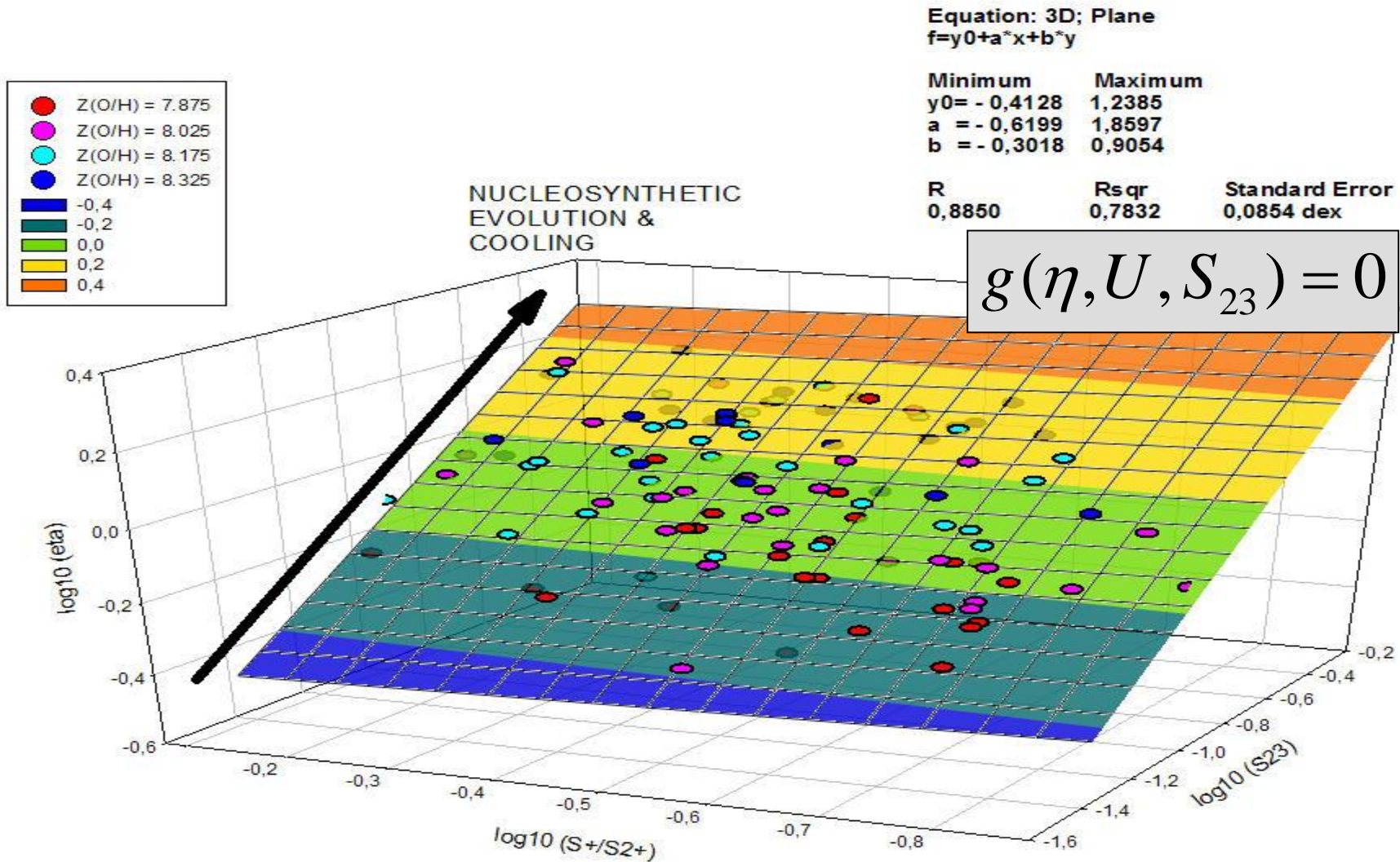
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

Clustering objects with Z



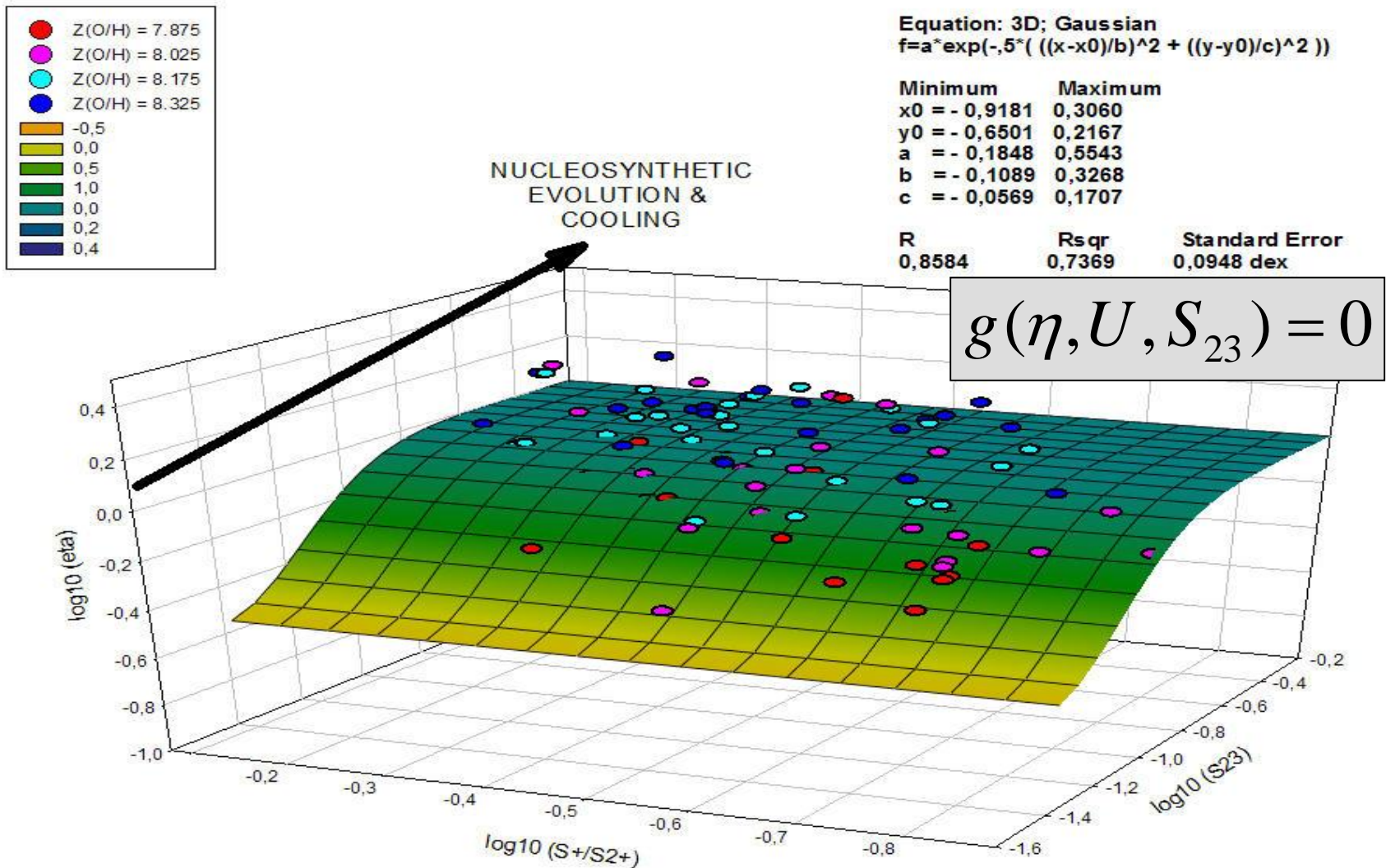
Fitting with a plane

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



Fitting with a Gaussian surface

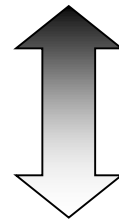
Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future



The biggest question...

Abundances
AI - A new opportunity
n-D Calibrations
Ideas for the future

$$Z(O / H) = f(c, x_1, x_2, \dots, x_n)$$



$$g(\eta, U, S_{23}) = 0$$

This mapping is possible if we can define a scale of effective temperature T^* using empirical measures **ONLY**. But how?



- To find new n-D indicators & improve calibrations of the form:
 $Z=f(c, x_1, x_2 \dots x_n)$ for HII-like regions and then for AGNs and PNs
- Find the relation $g(\eta, U, S_{23}) = 0$ and to test its robustness
- To investigate finding an empirical scale for T^*

On behalf of myself, Pepe and Angeles, I would like to thank Guille Hägele, Enrique Perez-Montero and Mariluz Martin for kindly making observational data available and to everyone who has patiently and carefully taken accurate images

Many thanks