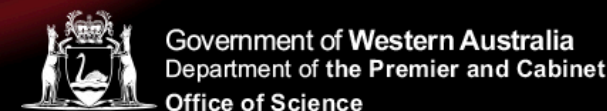




International
Centre for
Radio
Astronomy
Research

EXPLORING THE LIMITS OF THE BAYESIAN UNIVERSE: HOW TO TACKLE BREADTH AND DEPTH



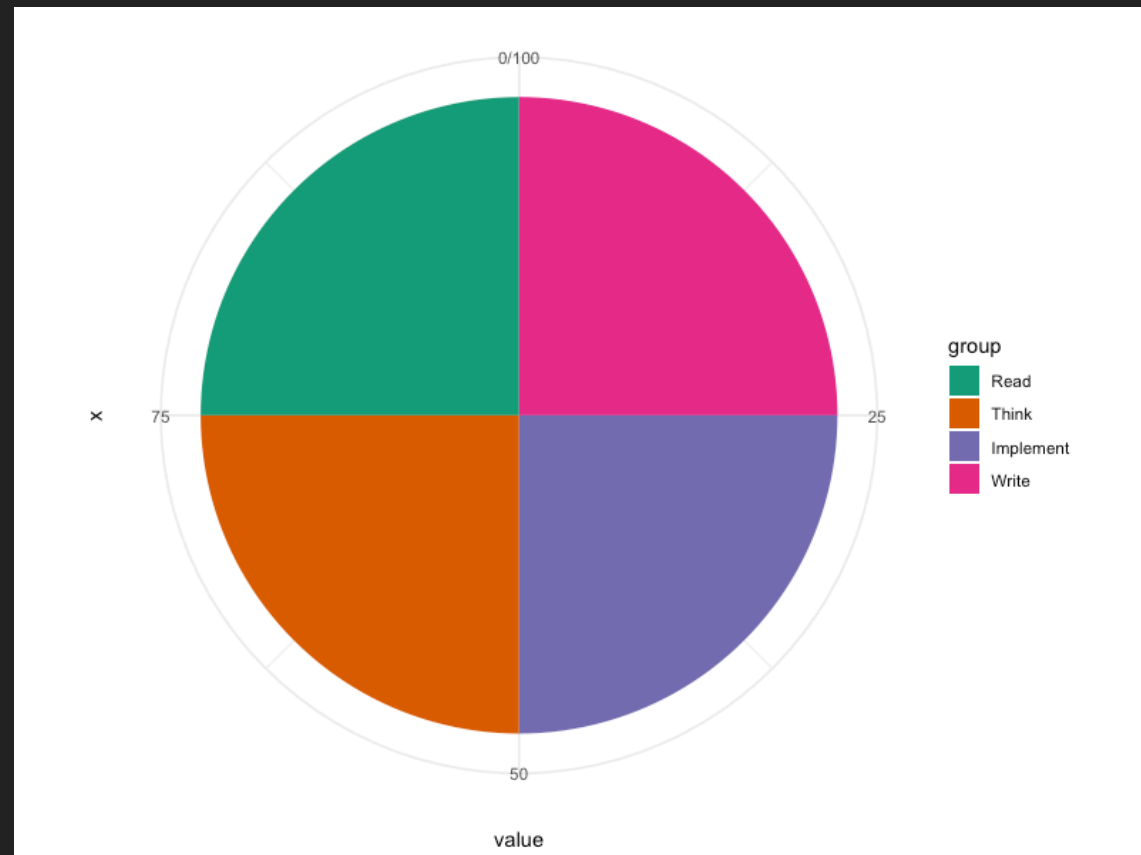
AARON ROBOTHAM

Sabine Bellstedt, Jessica Thorne

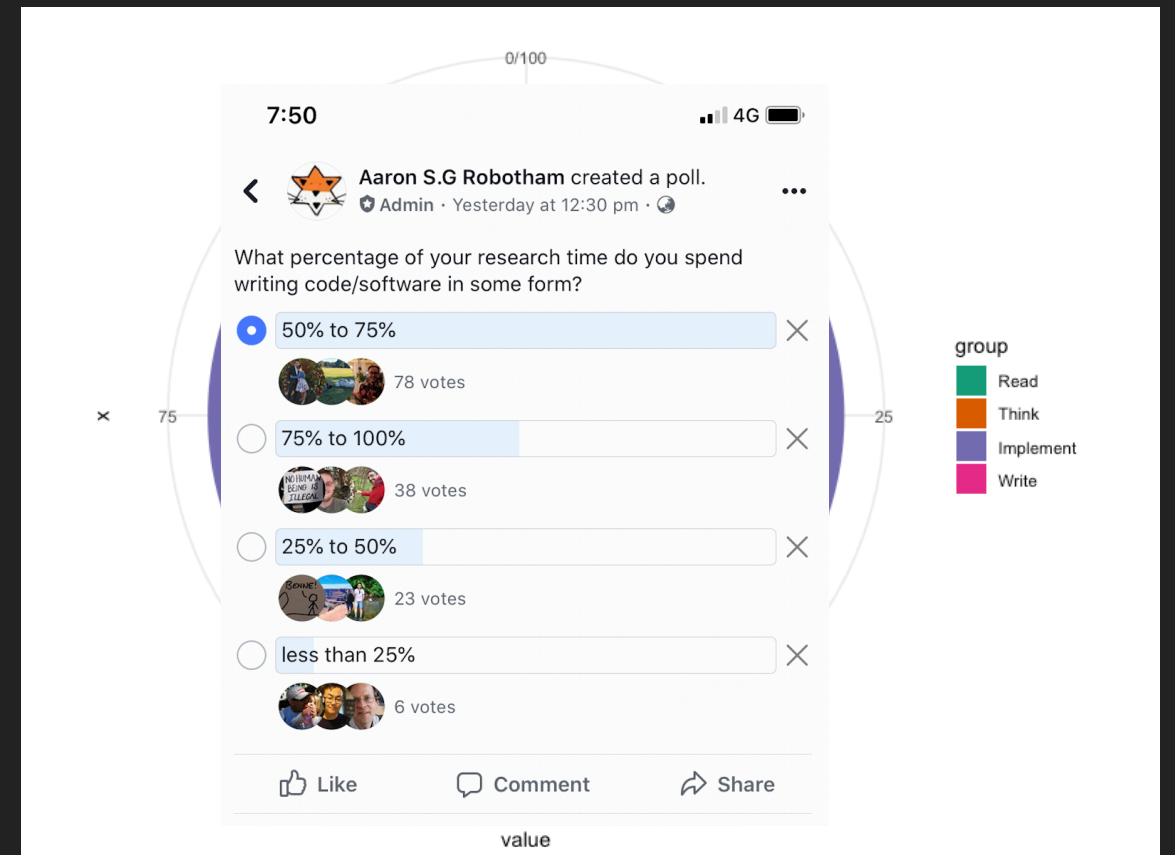
OPEN-SOURCE STATISTICS AND COMPUTING

WHAT WE DO WHEN WE RESEARCH:

The Legend



The Reality

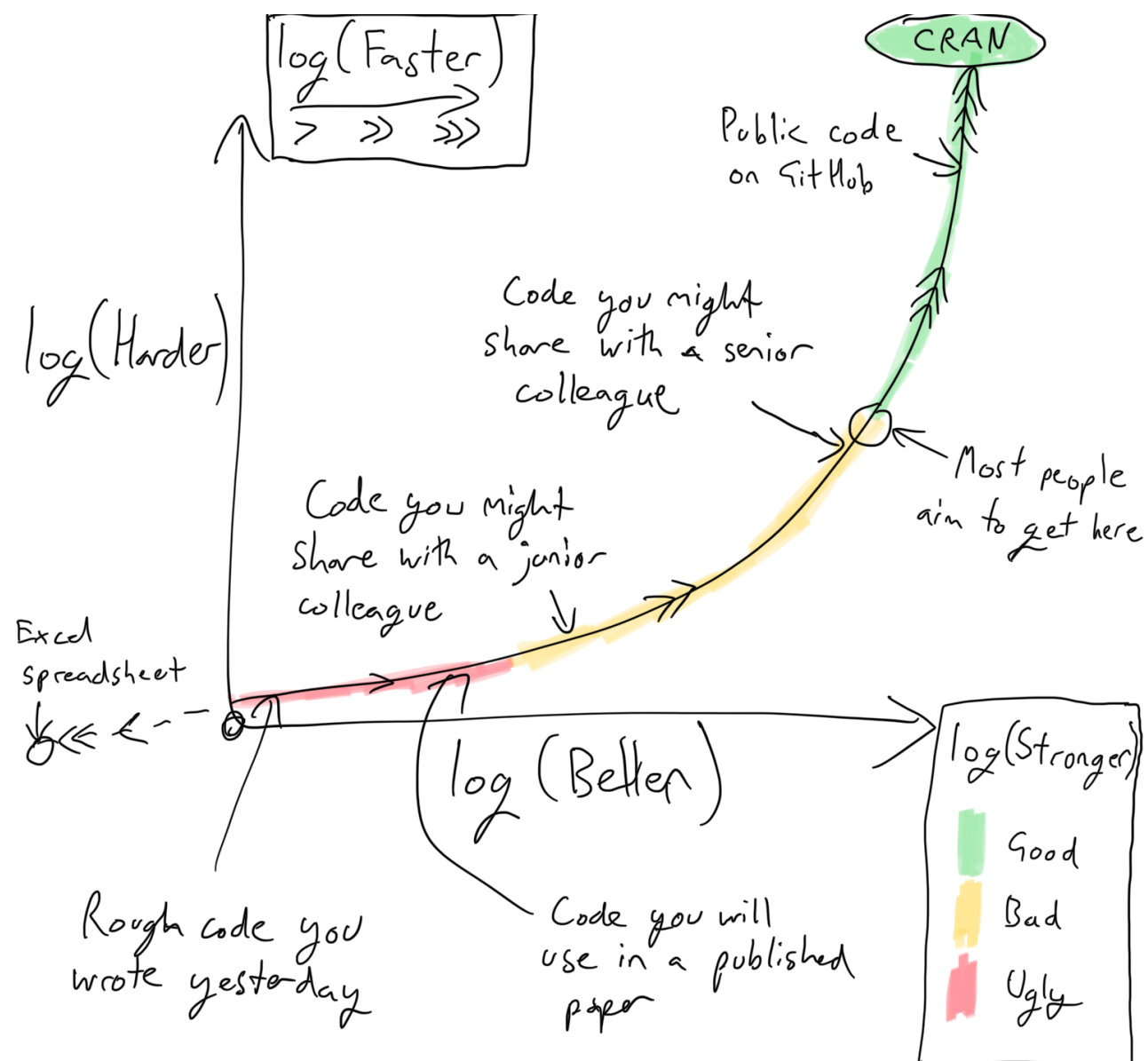


In a poll on the Astrostatistics Facebook group, 80% of people reported spending more than half (average around 70%) of their research time writing code/software on a computer.

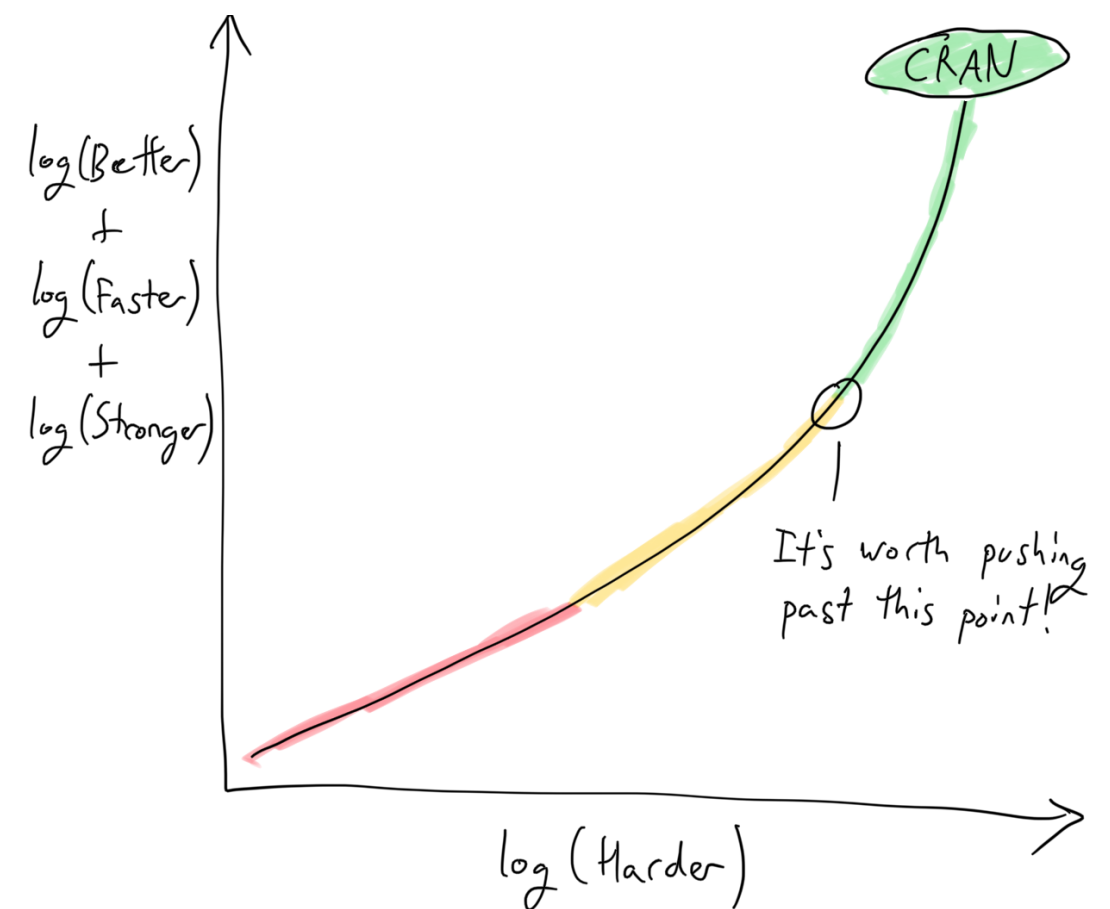
Harder: roughly time taken (mental effort)
Better: commented, maintainable
Faster: does the job faster
Stronger: Reliable, fault tolerant, robust

OPEN-SOURCE STATISTICS AND COMPUTING

SO WE MUST BE GREAT AT THIS, RIGHT? WELL...



But overall it is worth it!



OPEN-SOURCE STATISTICS AND COMPUTING

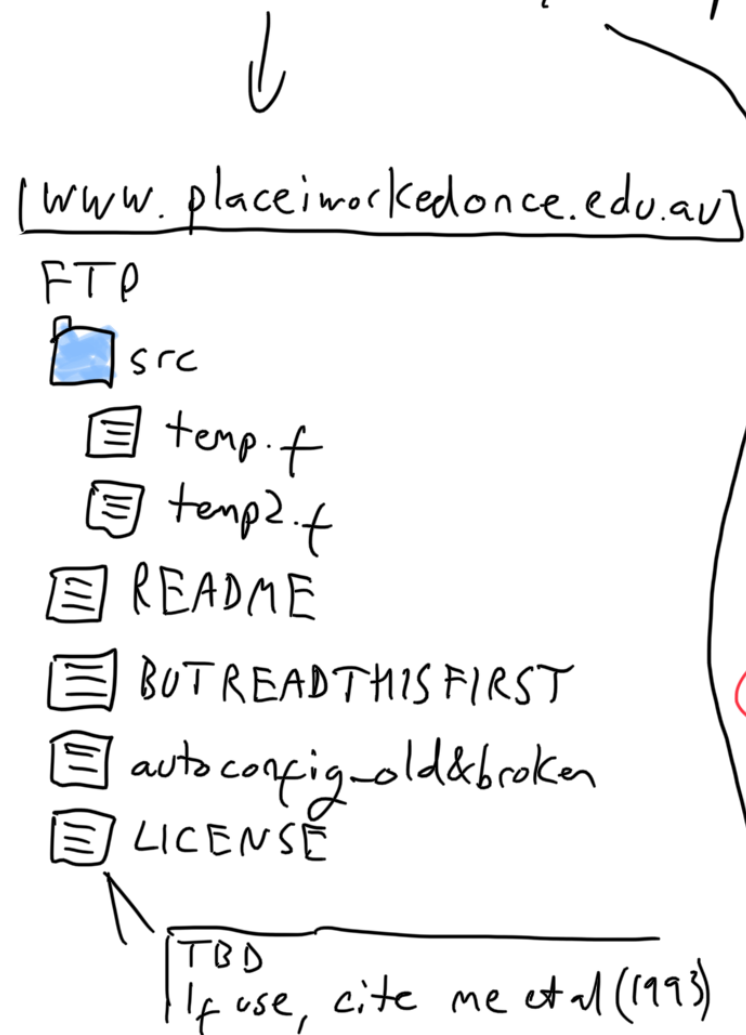
THE SOLUTION

- ▶ We (as a community) bootstrap the “implement” part wherever feasible.
- ▶ Even when we re-invent the wheel, knowing what wheels are available makes this:
 - ▶ A) worthwhile (need a roller-blade wheel in a world of cartwheels)
 - ▶ B) efficient (we can still borrow ideas - make it round!).
- ▶ To help our community this means the following:
 - ▶ Aggressively public code +
 - ▶ Default collaborative efforts =
 - ▶ **Open-source** analysis (statistical methods and software)

OPEN-SOURCE STATISTICS AND COMPUTING

BUT REMEMBER!

Feral Code $\neq$ Open Source!



[www.github.com/me/code]

Code | Commits | Issues

src

Man

README.md

LICENSE

this really matters!!!

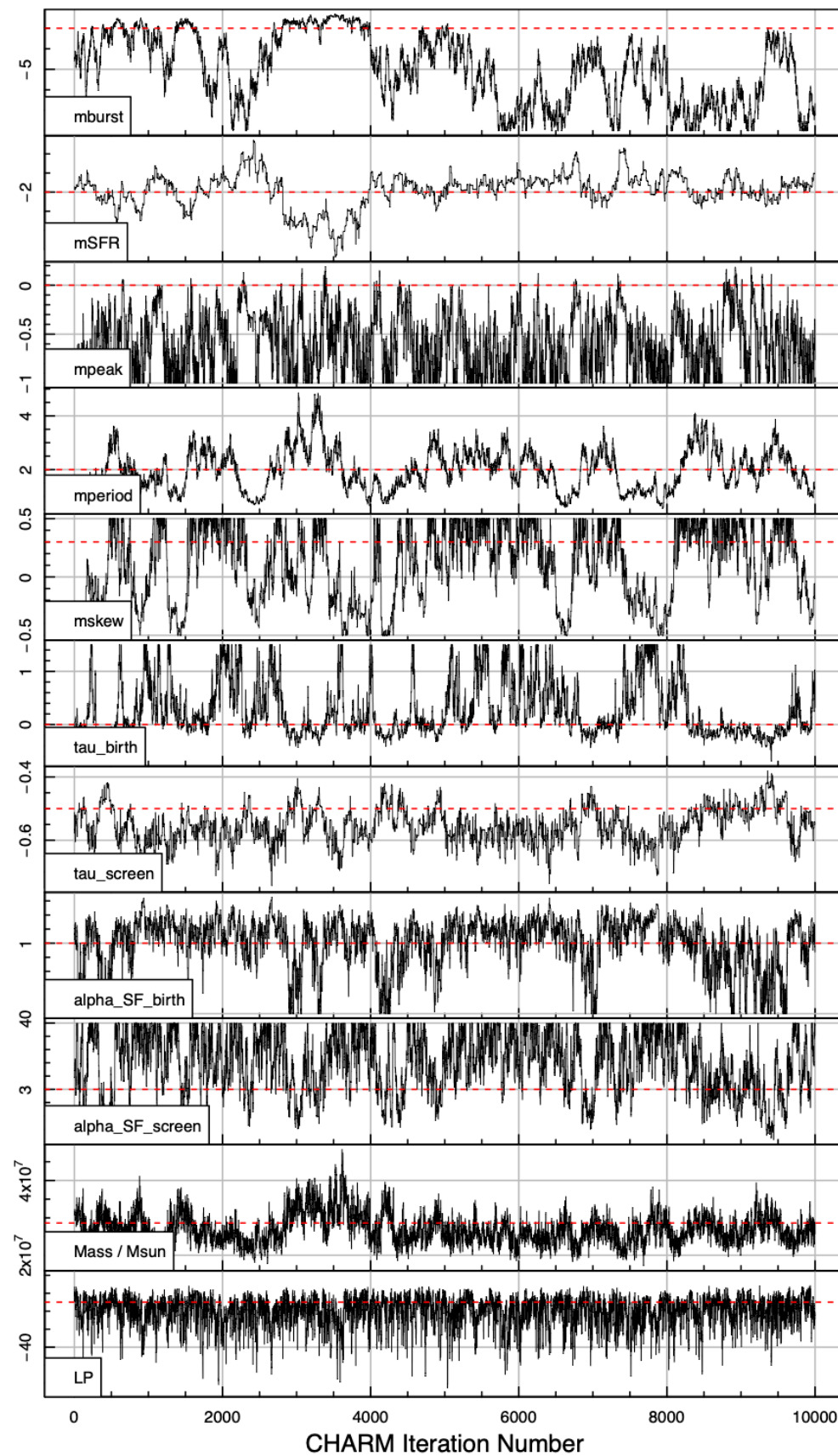
└─┬ GPL/MIT/BSD/APACHE

WHY WE SHOULD DO OPEN-SOURCE STATISTICS IN ACADEMIA?

- ▶ Because it is the new fun thing!
- ▶ I keep seeing talks about GitHub, maybe it will help me get a job...?
- ▶ My grant funding insists.
- ▶ Better than nothing for backing things up.
- ▶ Paying for things sucks.
- ▶ To make the world a better place...
- ▶ These are not the best reasons actually... we can do better by appealing to our more ruthless and selfish instincts!

WHY WE REALLY SHOULD DO OPEN-SOURCE STATISTICS IN ACADEMIA?

- ▶ Only way to efficiently collaborate on complex projects, and drops the bar to entry to almost nothing. If people cannot see it then they will not ask to help.
- ▶ Get a lot more credit and exposure to your work (you will start appearing all over the internet).
- ▶ It instills a much higher level quality of work: if you do not feel comfortable sharing it then you should not feel comfortable publishing it! As Russians say: "Trust, but verify".
- ▶ It provides free (!) wide-coverage quality checking you cannot replicate alone.
- ▶ It allows you to work seamlessly on projects across multiple devices without penalty (I often show people snippets from GitHub on my phone - handy in a pinch).
- ▶ It naturally enforces modern version control by any practical route (Git / GitHub).
- ▶ It helps less senior members of our community. What does this mean...?



STATISTICS AND
COMPUTING

BARRIERS

TOOLS TO BE A MODERN ASTRONOMER

- ▶ A new PhD student in a highly mature field like astronomy (a few thousand years of research and counting) has a lot of catching up to do. They need:
 - ▶ Solid coding skills (no matter what they do).
 - ▶ Relatively sophisticated knowledge of data analytics and statistics.
- ▶ Broad understanding of the current state of the field, and a deep understanding of their own sub-field.
- ▶ Excellent project management.
- ▶ Good scientific writing (an un-taught skill usually)
- ▶ Functional people skills.

A CASE STUDY WITH HYPER-FIT

- ▶ Astronomy is unusual within physics in how expensive it is to get observations, so we tend to have un-ignorable errors, and available techniques like PCA and SVM rarely support the treatment of errors *properly*.
- ▶ Because of some mass-size data, I started wondering about hyperplane fitting with heteroscedastic covariant errors, and published a paper tackling it:

Publications of the Astronomical Society of Australia (PASA), Vol. 32, e033, 14 pages (2015).
© Astronomical Society of Australia 2015; published by Cambridge University Press.
doi:10.1017/pasa.2015.33

Hyper-Fit: Fitting Linear Models to Multidimensional Data with Multivariate Gaussian Uncertainties

A. S. G. Robotham and D. Obreschkow
ICRAR, M468, University of Western Australia, Crawley, WA 6009, Australia
Email: Aaron.robatham@uwa.edu.au

(RECEIVED July 01, 2015; ACCEPTED August 10, 2015)

$$\ln \mathcal{L} = -\frac{1}{2} \sum_{i=1}^N \left[\ln \left(\sigma_{\perp}^2 + \frac{\mathbf{n}^T \mathbf{C}_i \mathbf{n}}{\mathbf{n}^T \mathbf{n}} \right) + \frac{(\mathbf{n}^T [\mathbf{x}_i - \mathbf{n}])^2}{\sigma_{\perp}^2 \mathbf{n}^T \mathbf{n} + \mathbf{n}^T \mathbf{C}_i \mathbf{n}} \right]$$

This involves N-dimensional projections and careful treatment of the data covariance terms. This is all fiddly stuff that took me a lot of sanity checking to implement robustly (e.g. adapting for geometric corner cases). Now the question is, would you confidently trust a new PhD student to implement the above in code you need to use?

OPEN-SOURCE STATISTICS AND COMPUTING

A CASE STUDY WITH HYPER-FIT

- So a paper with an equation is often not enough. How about an R CRAN package?

hyper.fit: Generic N-Dimensional Hyperplane Fitting with Heteroscedastic Covariant Errors and Intrinsic Scatter

Includes two main high level codes for hyperplane fitting (`hyper.fit`) and visualising (`hyper.plot2d` / `hyper.plot3d`). In simple terms this allows the user to produce robust 1D linear fits for 2D x vs y type data, and robust 2D plane fits to 3D x vs y vs z type data. This hyperplane fitting works generically for any N-1 hyperplane model being fit to a N dimension dataset. All fits include intrinsic scatter in the generative model orthogonal to the hyperplane.

Version: 1.1.0
Depends: R (≥ 3.00), [magicaxis](#), [MASS](#), [rgl](#), [LaplacesDemon](#)
Published: 2019-01-31
Author: Aaron Robotham and Danail Obreschkow
Maintainer: Aaron Robotham <aaron.robotham@uwa.edu.au>
License: [GPL-3](#)
NeedsCompilation: no
Citation: [hyper.fit citation info](#)
Materials: [NEWS](#)
CRAN checks: [hyper.fit results](#)

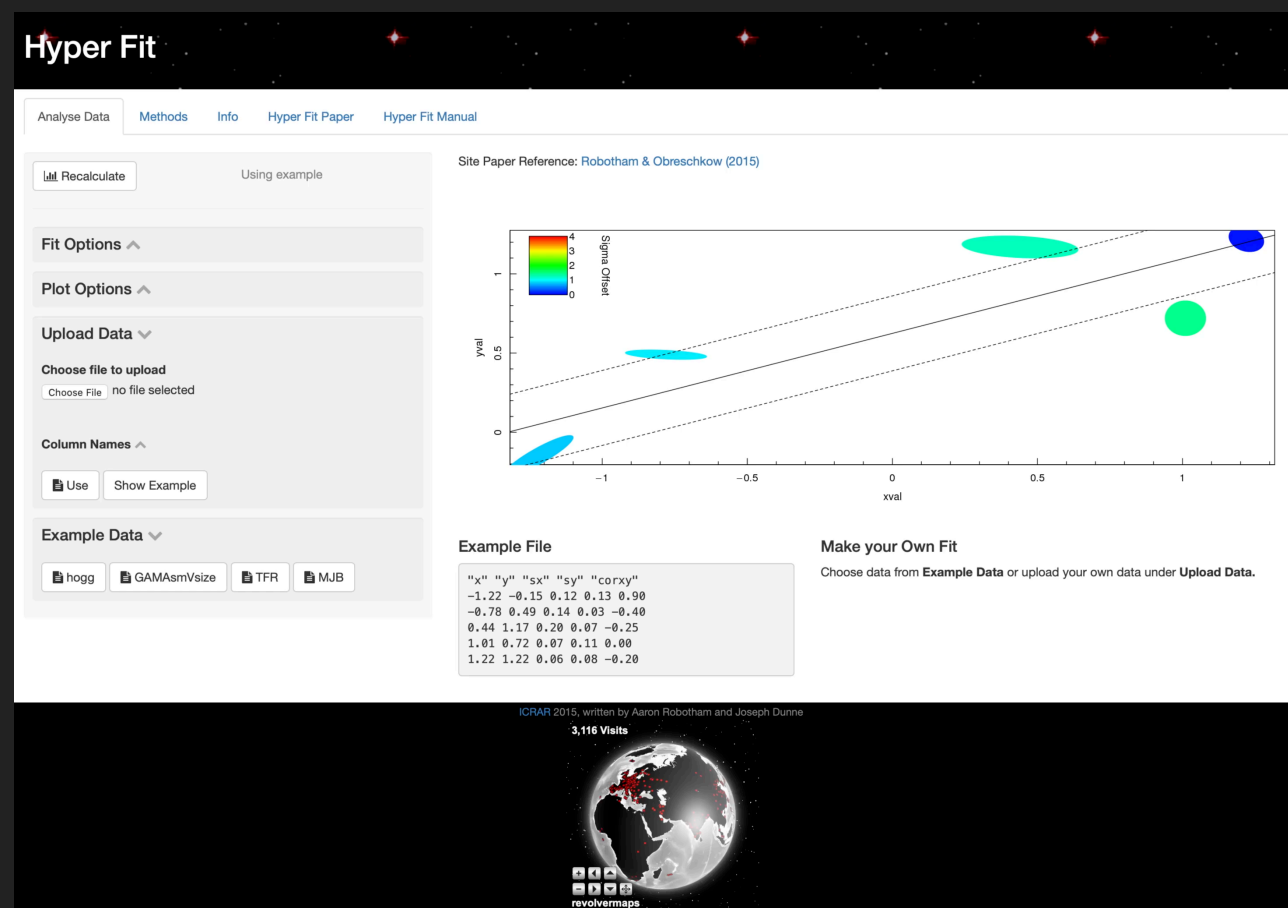
Downloads:

Reference manual: [hyper.fit.pdf](#)
Package source: [hyper.fit_1.1.0.tar.gz](#)
Windows binaries: r-devel: [hyper.fit_1.1.0.zip](#), r-release: [hyper.fit_1.1.0.zip](#), r-oldrel: [hyper.fit_1.1.0.zip](#)
OS X binaries: r-release: [hyper.fit_1.1.0.tgz](#), r-oldrel: [hyper.fit_1.1.0.tgz](#)
Old sources: [hyper.fit archive](#)

- The bar for this is high (*much* higher than a simple code repo like SourceForge or PyPI)- it requires full code documentation (40 pages!) and a full suite of useable examples.

A CASE STUDY WITH HYPER-FIT

- Is that enough? Well lots of people do not use R, fear it, and will not read a manual (arguably they do not deserve my help!). Behold Shiny Apps:



hyperfit.icar.org

<https://www.icrar.org/our-research/tools/>

Cosmology Calculator

cosmocalc.icrar.org

Calculation

Plot

Survey Design

Sky Separation

Info

R Code

Calculate

Set Variables:

z

Sig Figs

Reference Set

H[0]

OmegaM[0]

OmegaL[0]

OmegaR[0]

w[0]

Sigma8[0]

Custom Calc:

Value (Gyr)

Solution to find

Variable

Results:

The redshift z is 1

The expansion factor a is 0.5

The Relative Error is ± 1e-08

Distance:

The Comoving Radial Distance to z is 66.0

The Luminosity Distance to z is 66.0

The Angular Diameter Distance to z is 66.0

The Comoving Radial Distance to z is 66.0

The Distance Modulus to z is 44.10

The Physical Angular Scale at z is 1 arcsec

The Comoving Volume to z is 151.0

Velocity and Expansion:

Hubble's constant H at z is 123.248

Expansion rate $\dot{a}$ at z is 61.6239 (km/s/Mpc)

The cz Velocity at z is 1 (c)

The Cosmological Recession Velocity at z is 1 (c)

The Deceleration parameter (q) at z is 1

What's Up?

whatsup.icrar.org

Day/Night Plan

Now Plan

Year Plan

Finder Chart

Y2019M11D6, RA 125, Dec 19, Lon 116, Lat -32, Mphase 0.67, Msep 150, Ssep 100

Alt / deg

360N

270W

180S

90E

0N

Local Time / Hours

13

14

15

16

17

18

19

20

21

22

23

00

01

02

03

04

05

06

07

08

09

10

11

12

13

Target

Sun

Moon

The HI Fidelity Calculator

hifi.icrar.org

Info

z, Width, Flux

Angles & Beam

Flux/Mass

Flux/nHI

Fluxd/Tb

Noise

SNR

Plot:

Change the parameters below to modify the redshift dependence plot of conversions from the given fixed value:

z Start

z End

z Resolution

solid angle scaling

x Axis

y Axis

Log x axis

Log y axis

Flip x axis

Flip y axis

Save Data

Download as .txt

Download as .csv

Results:

The redshift z is 1

The observed solid angle of the source is 1 arcsec²

The observed flux density is 1e-06 Jy

The area of the source (small angle approx) is 64.14 kpc²

The brightness temperature of the source is 5.491 K

Brightness Temperature vs z

Brightness Temperature (K)

z

HMFcalc

Info

Contact Us

Calculate

View Plots

Plot:

dn/dm

Download

PDF of Current Plot

Mass Function $\left(\frac{dn}{dm}\right) h^4 \text{ Mpc}^{-3} \text{ M}_\odot^{-1}$

Mass ($\text{M}_\odot h^{-1}$)

AstroMap

astromap.icrar.org

Description

Tool

Get Code

Contact

Map Projection

Light Cones

Deep Fields

Update plot

Projection

Projection Type

Long-Cent (0,180)

Coordinate Type

Objects/Paths

Current Wide Surveys

Current Deep Surveys

Upcoming Surveys

User Defined

PDF

Ecliptic

Sun

MW Plane

MW Center

WAVES-Wide

DEVILS

GAMA

WAVES-L

ProSpect SED

prospect.icrar.org

ICRAR

Redshift

0.1

SFH/AGN:

Burst Mass (log10):

Young Mass (log10):

Mid Mass (log10):

Old Mass (log10):

Ancient Mass (log10):

AGN Luminosity (log10):

Aaron Robotham

GitHub: asgr/ProSpect

Dust inputs:

Tau Birth:

Tau Screen:

Tau AGN:

Alpha SF Birth:

Alpha SF Screen:

Alpha SF AGN:

Flux

Luminosity

Info

Flux Density (Jansky)

Wavelength (Angstrom)

FUV_GALEX

NUV_GALEX

u_SDSS

g_SDSS

r_SDSS

i_SDSS

Z_VISTA

Y_VISTA

J_VISTA

H_VISTA

K_VISTA

W1_WISE

W2_WISE

W3_WISE

W4_WISE

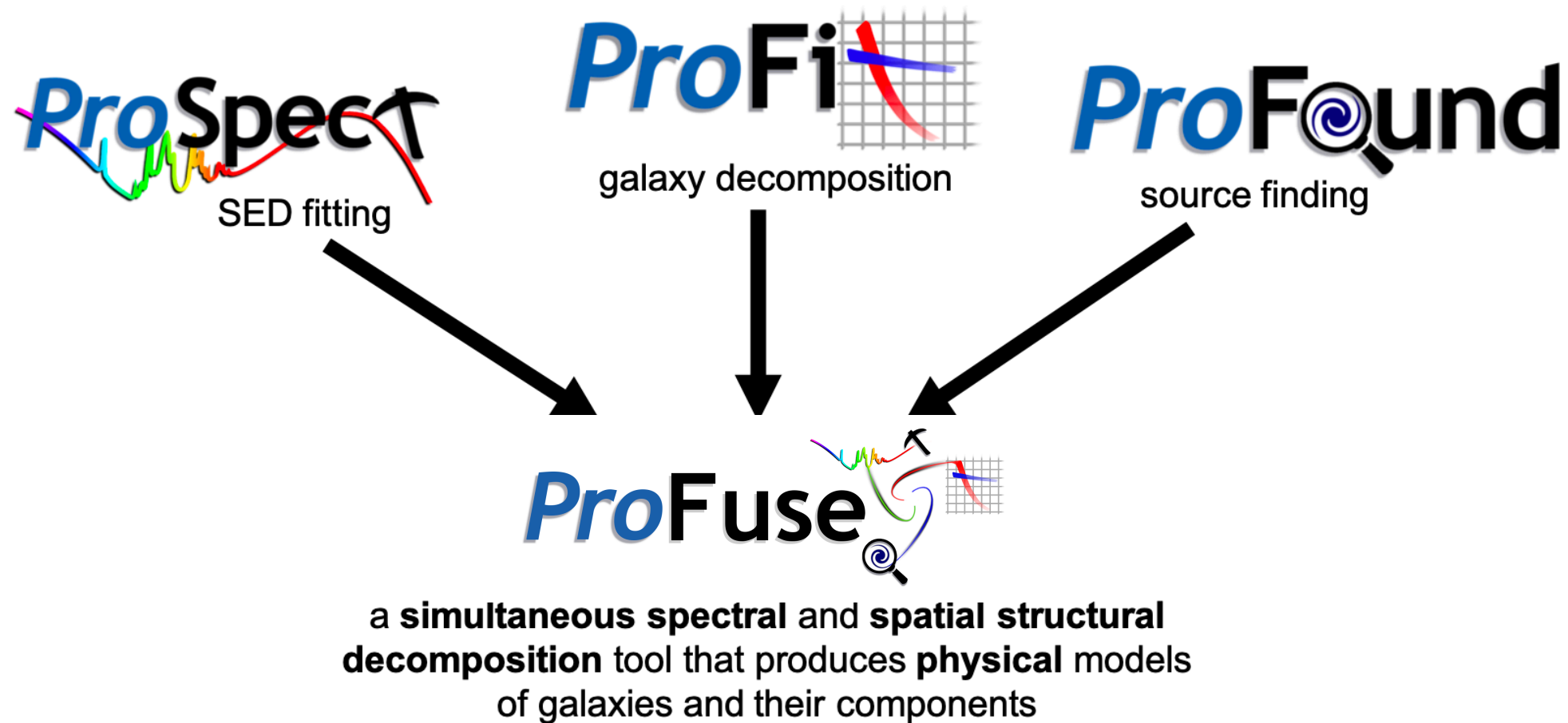
P100_Herschel

P160_Herschel

S250_Herschel

S350_Herschel

S500_Herschel



FROM PIXELS TO SCIENCE

PROFUSE

MNRAS **513**, 2985–3012 (2022)
Advance Access publication 2022 April 15

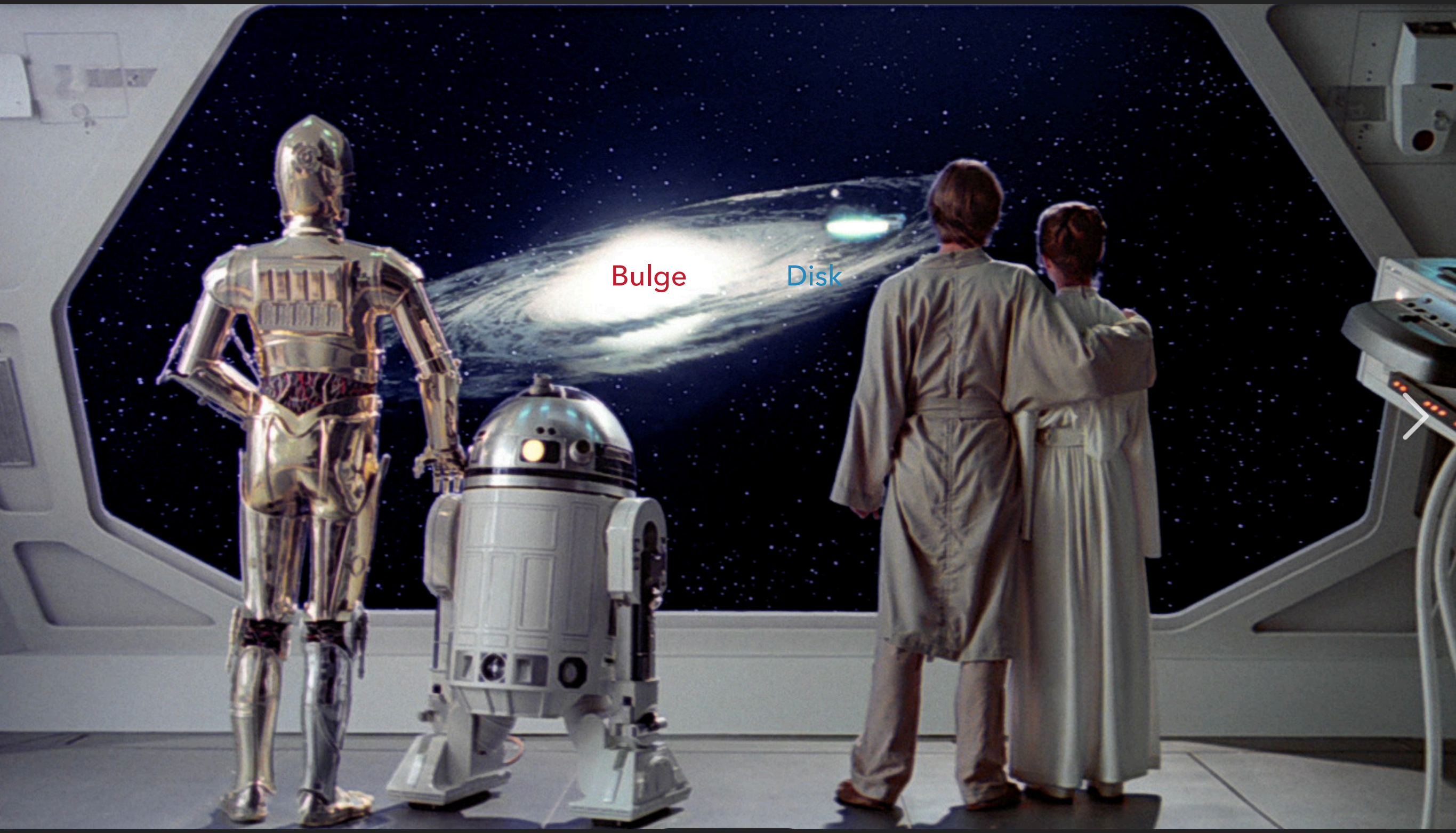
<https://doi.org/10.1093/mnras/stac1032>

ProFuse: physical multiband structural decomposition of galaxies and the mass–size–age plane

A. S. G. Robotham^{1,2}*, S. Bellstedt¹ and S. P. Driver¹

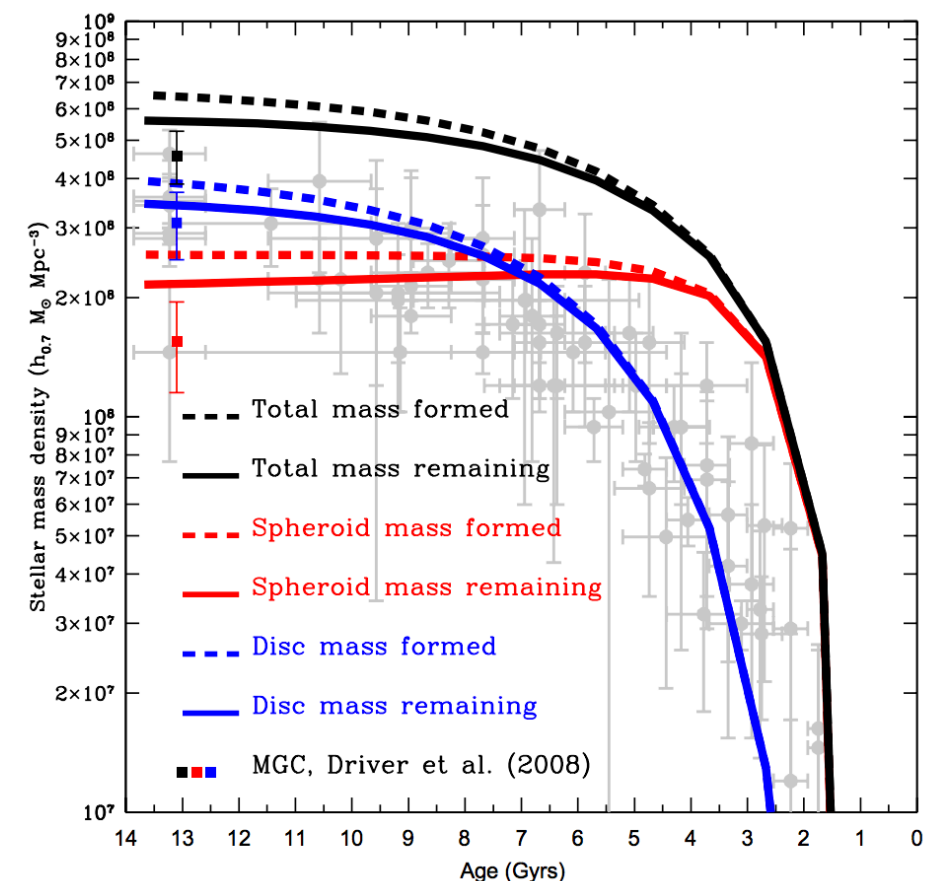
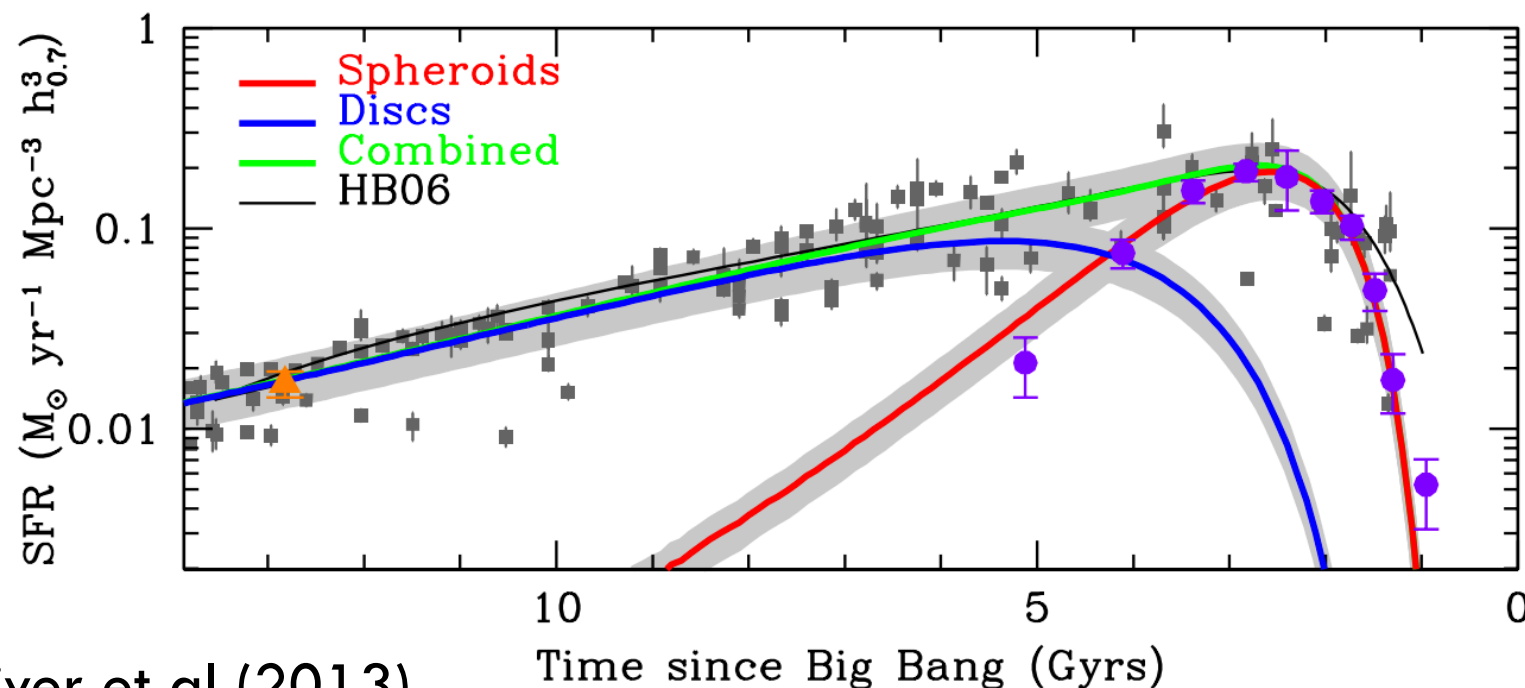
REFRESHER

BULGES AND DISKS



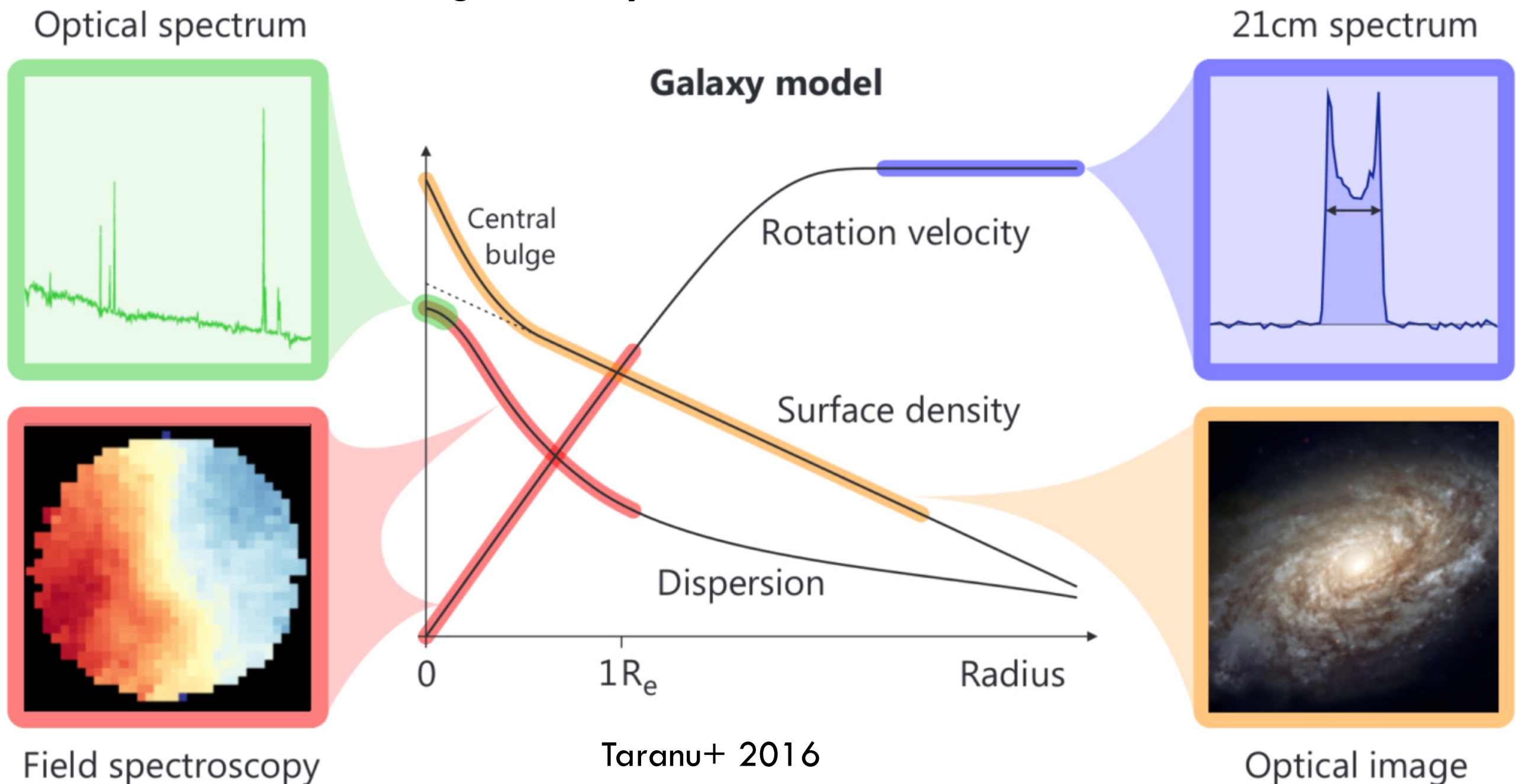
Why Decompose Galaxies?

- 2 reasons:
 - But the more important reason is that the components of galaxies contain the evolutionary origin of galaxies.
 - Two dominant paradigms:
 - Rapid assembly of bulges (star bursts and mergers)
 - Slower accretion growth of disks



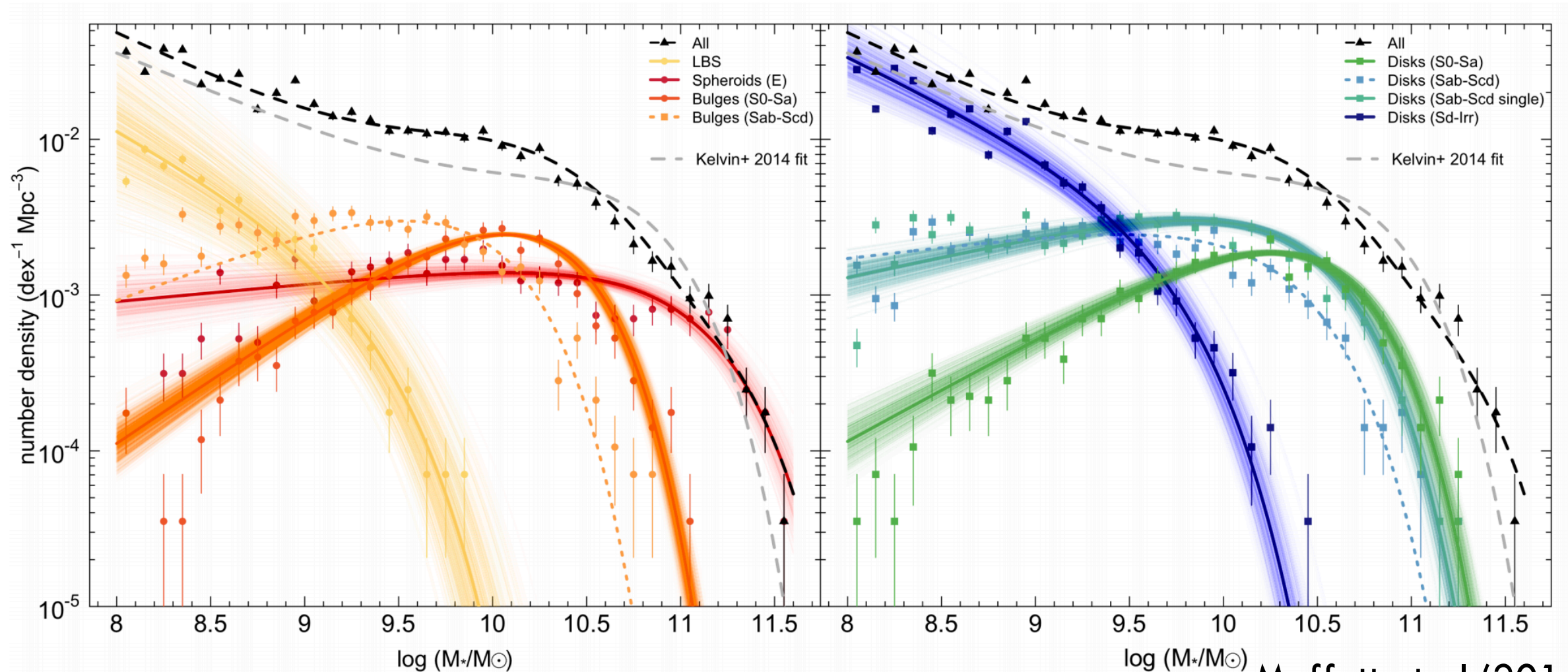
Why Decompose Galaxies?

- The optical properties of galaxies are key to understanding the dynamics:



Why Decompose Galaxies?

- The two major components also show remarkably different levels of significance for different mass galaxies (high mass = bulge dominated, low mass = disk dominated). Even the galaxy stellar mass function (GSMF) is non-trivial in detail:



THE PROFUSE PIPELINE

- ▶ **ProFound** - Detect sources automatically, estimate rough model parameters, create segmentation and error maps. See Robotham+ (18), Bellstedt+ (20), GAMA/DEVILS/WAVES.
- ▶ **ProFit** - Parameter driven single image galaxy decomposition. See Robotham+ (17), Cook+ (18/19), Casura+ (22).
- ▶ **ProSpect** - Multi-band spectral energy distribution generation and fitting. See Robotham+ (20), Bellstedt+ (20/21), Thorne+ (21/22ab).
- ▶ **ProFuse** = **ProFound** + **ProFit** + **ProSpect**
See Robotham+ (22)

ProFound

EXTRACTING SOURCES AND GETTING
READY FOR PROFIT

PROFOUND

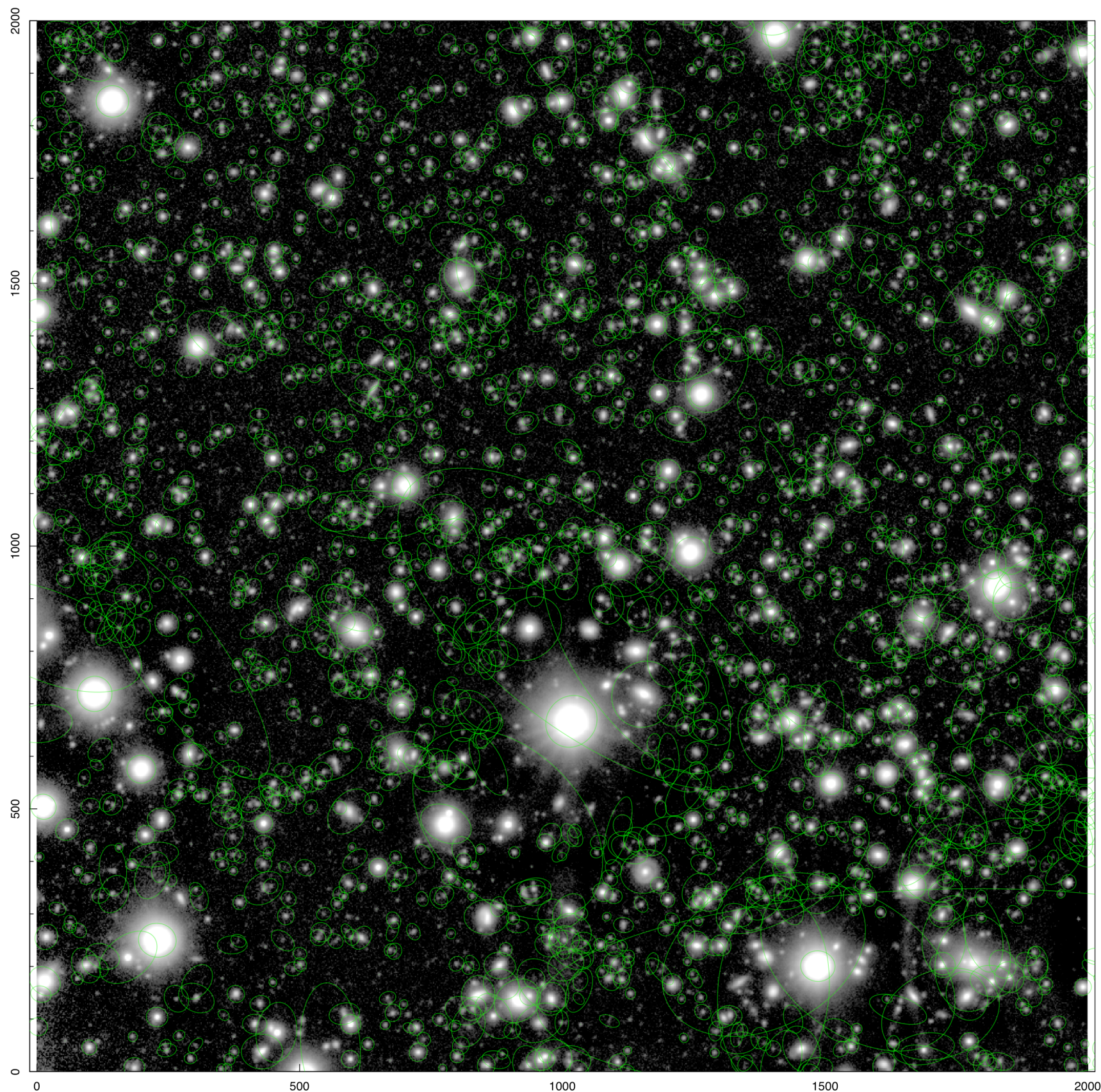
MNRAS **476**, 3137–3159 (2018)
Advance Access publication 2018 February 23

doi:10.1093/mnras/sty440

**PROFOUND: SOURCE EXTRACTION AND APPLICATION TO MODERN SURVEY
DATA**

A. S. G. Robotham,¹★ L. J. M. Davies,¹ S. P. Driver,¹ S. Koushan,¹ D. S. Taranu,^{1,2}
S. Casura³ and I. Liske³

THE PROBLEM WITH SEXTRACTOR APERTURES

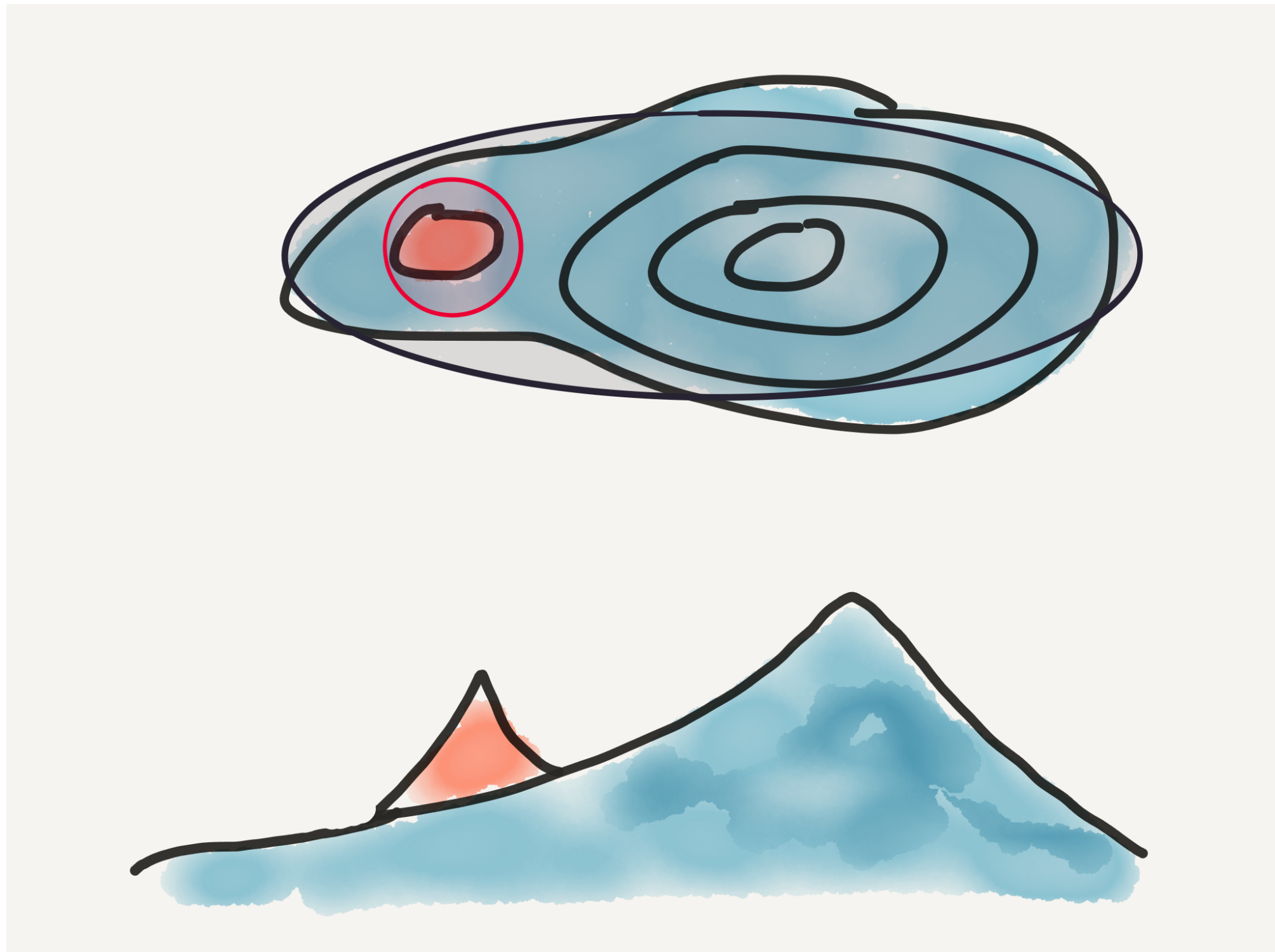


Ultra-VISTA Y-band stack + final release catalogue

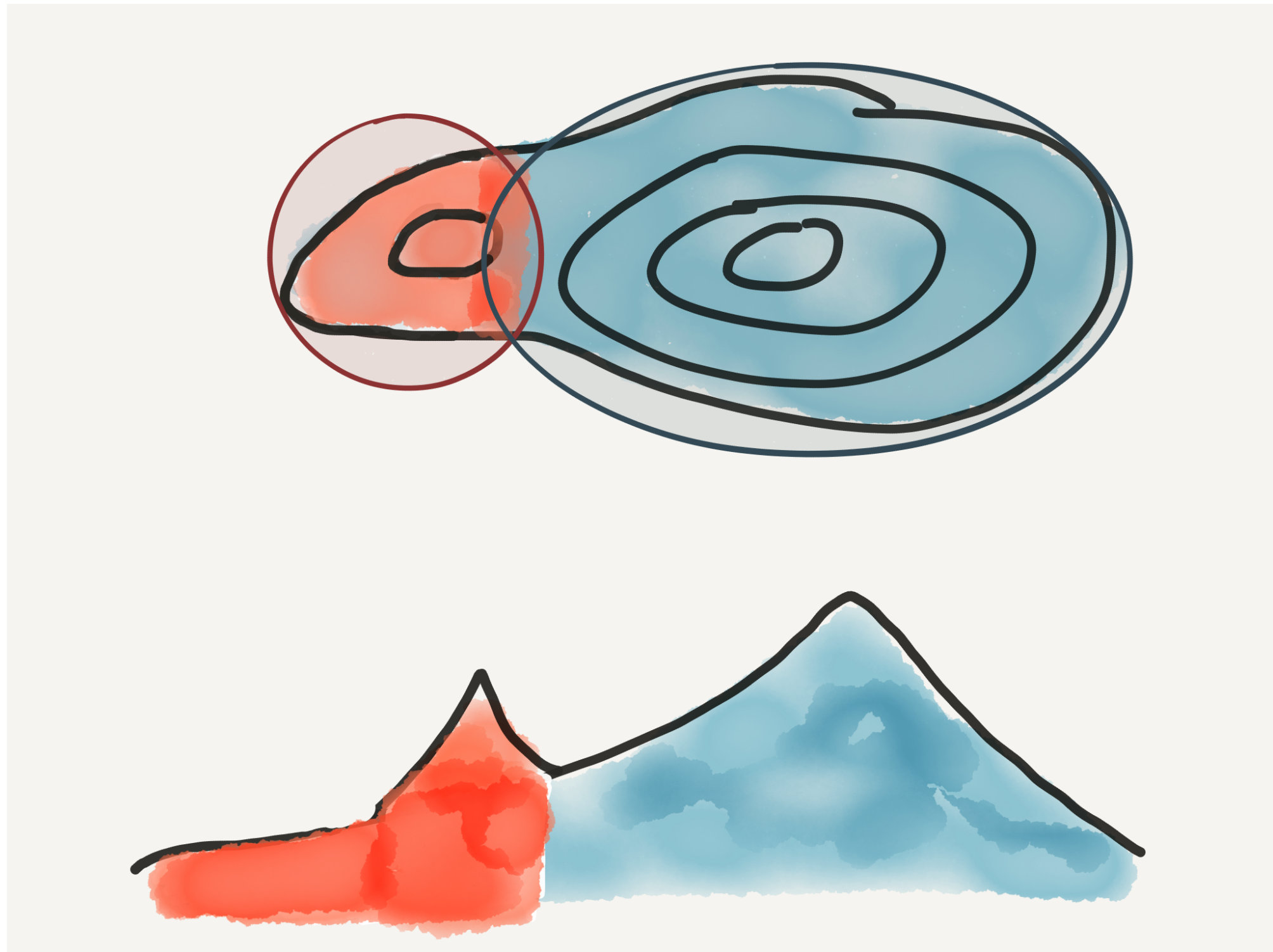
THE NEED TO START AGAIN

- ▶ For GAMA we investigated a lot of effort in manual aperture fixing. This was not scaleable or transferable to other data sets, we sought a better solution for WAVES.
- ▶ In short, I started again with the source extraction.
- ▶ It was not obvious what improvements might be possible over SExtractor (given how well tested and established it is) but two areas quickly came to light:
 - ▶ It does not watershed de-blend optimally (the most common failure we see is due to this). It does coarse island-based deblending.
 - ▶ It uses strictly elliptical apertures and then tries to distribute overlapping flux using a number of opaque internal schemes.

SEXTRACTOR TENDS TO CREATE WATERSHED ISLANDS



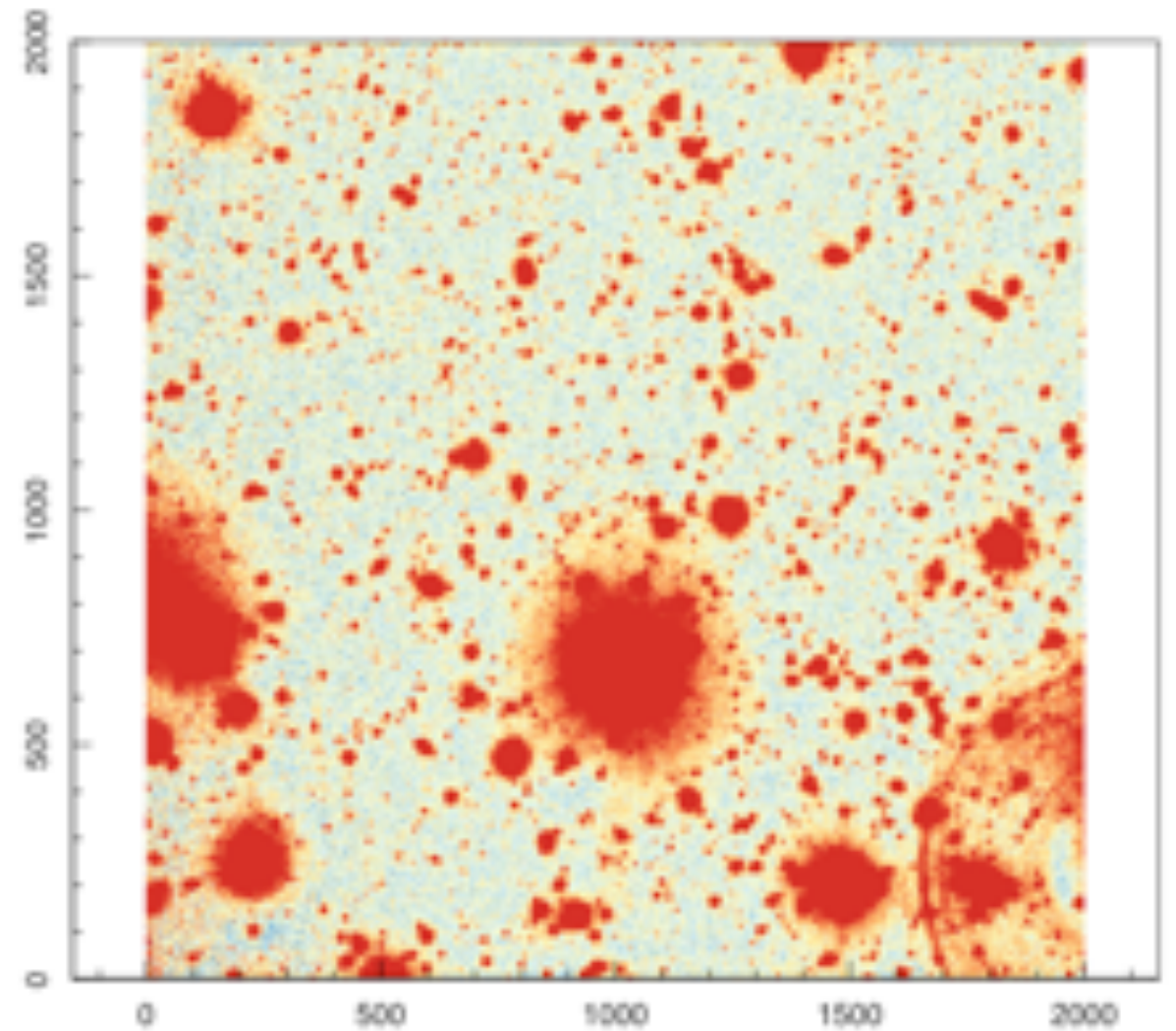
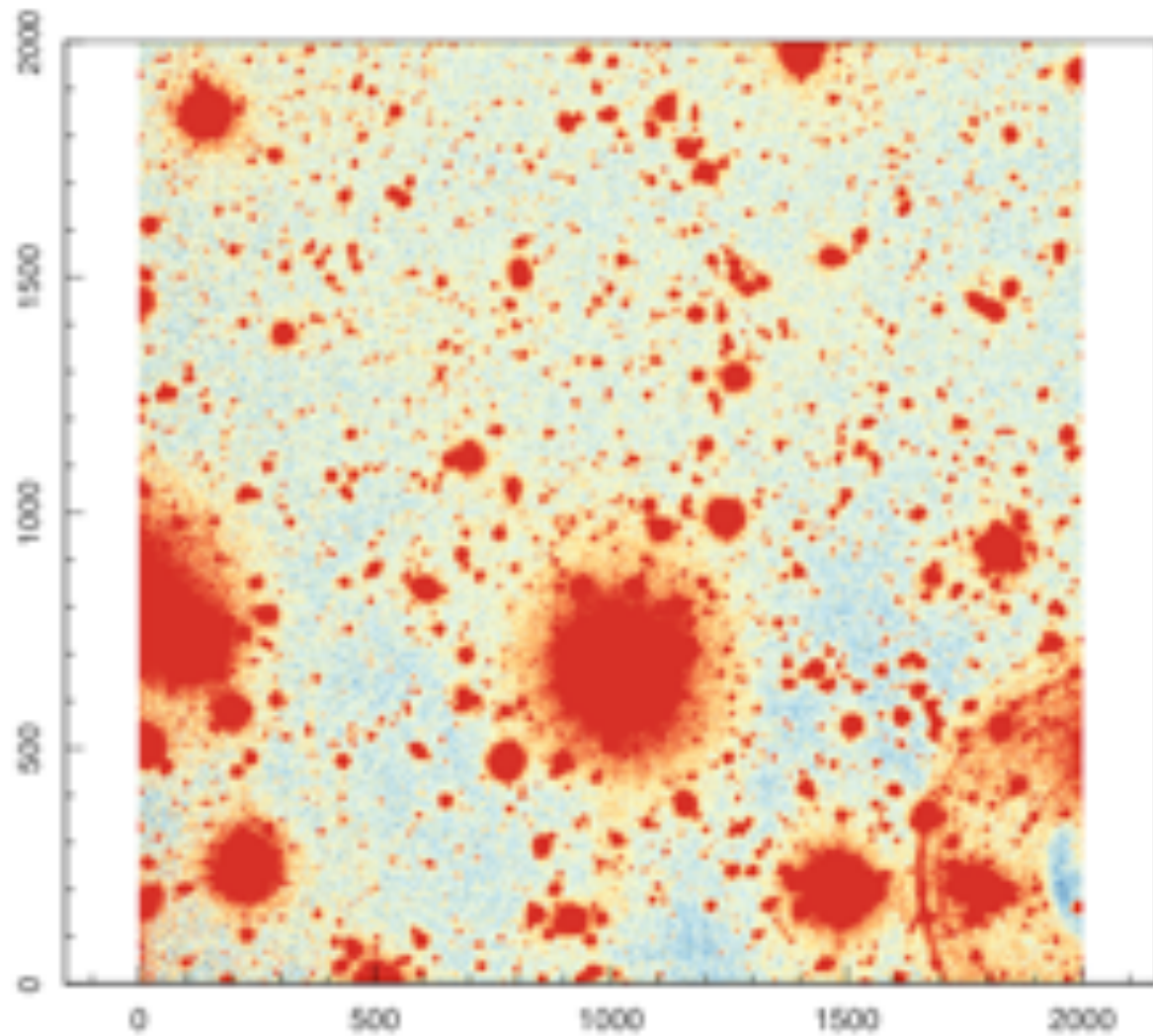
PROFOUND WATERSHEDS THROUGH SADDLE CONTOURS



OUR PROFOUND SOLUTION GETS ROUND BOTH THESE ISSUES

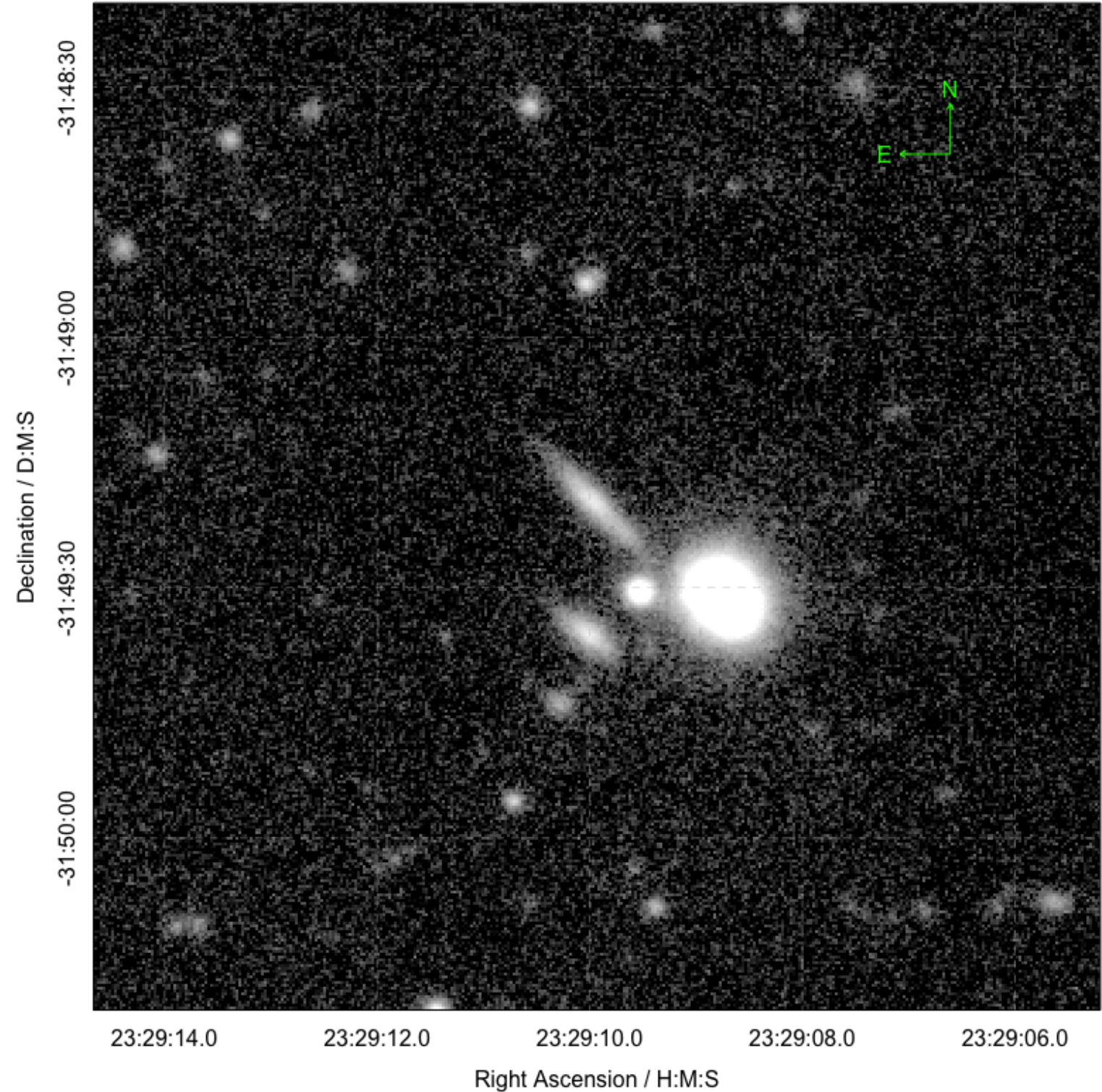
- ▶ We use a similar approach to find the initial high S/N images segments:
 - ▶ Careful sky subtraction (iterative masking and clipping)
 - ▶ S/N=1.5 threshold as standard (can change)
 - ▶ Segments are de-blended to some tolerance (using a different algorithm to SExtractor- non-discretised surface brightness / sky-RMS thresholds and locally adaptive).
 - ▶ Segments are grown organically- apertures never used.

PROFOUND USES AN AGGRESSIVE MESH BASED SCHEME

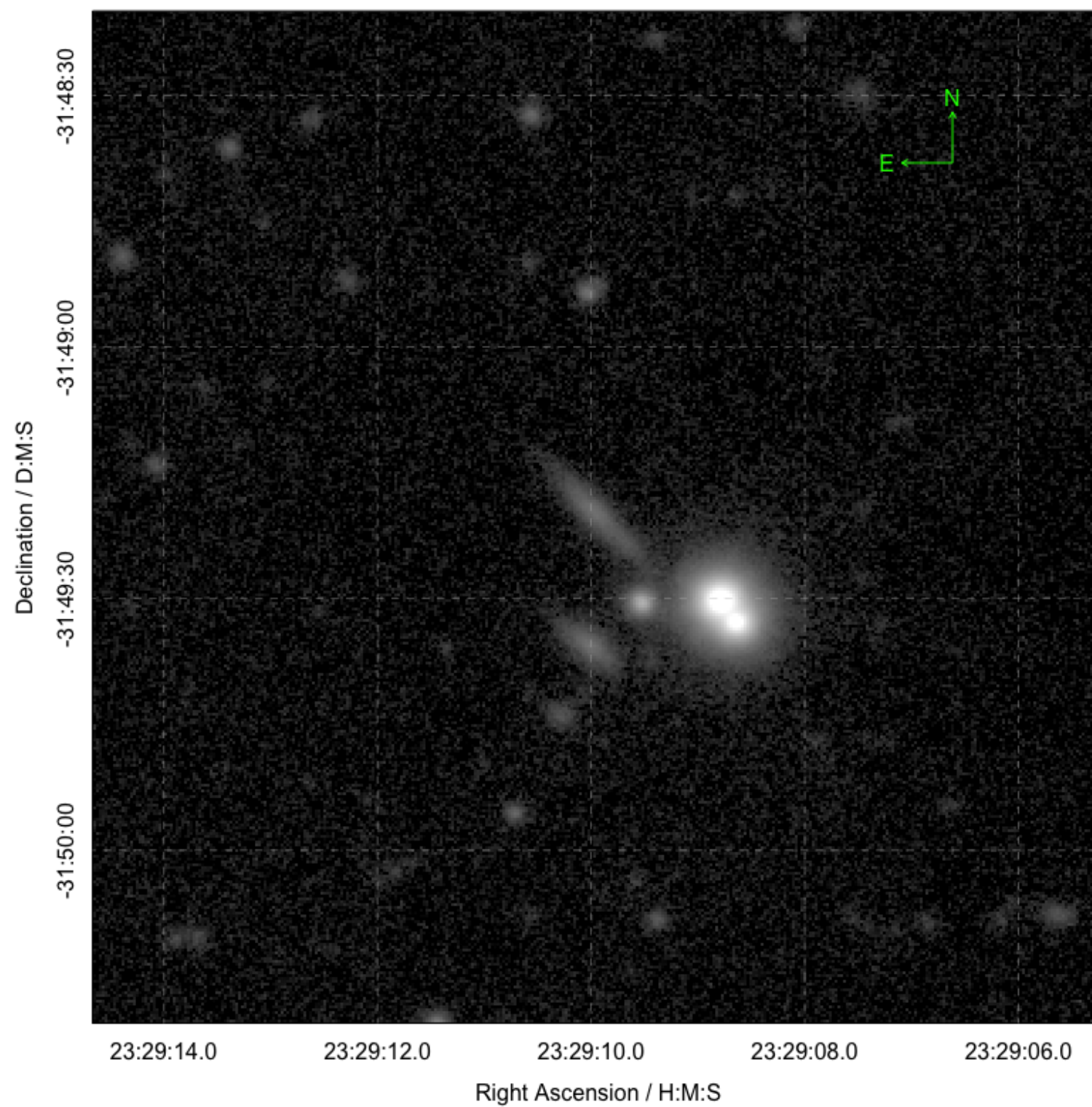


Ultra-VISTA SExtractor sky versus ProFound sky

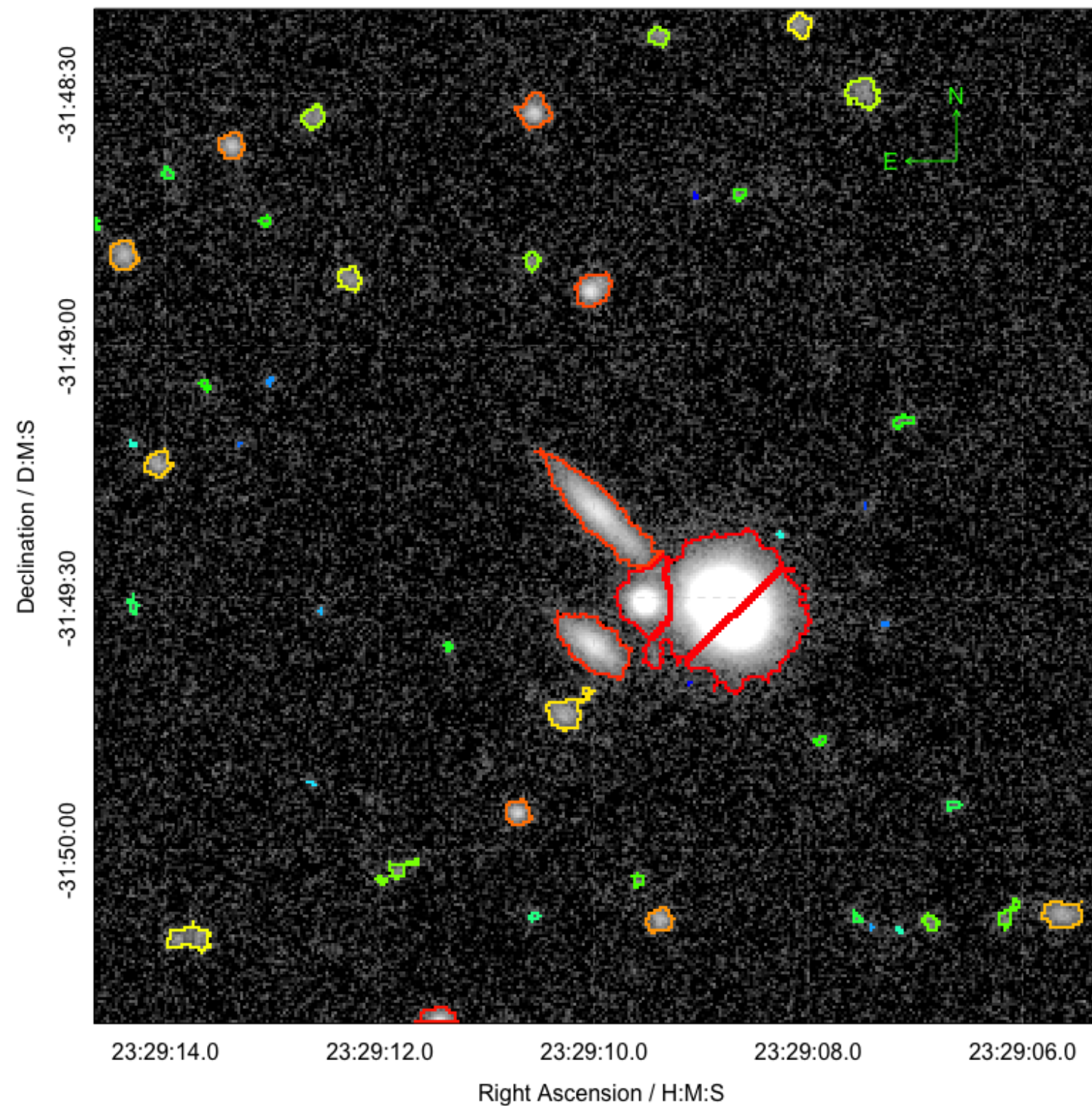
INITIAL VIKING Z-BAND IMAGE



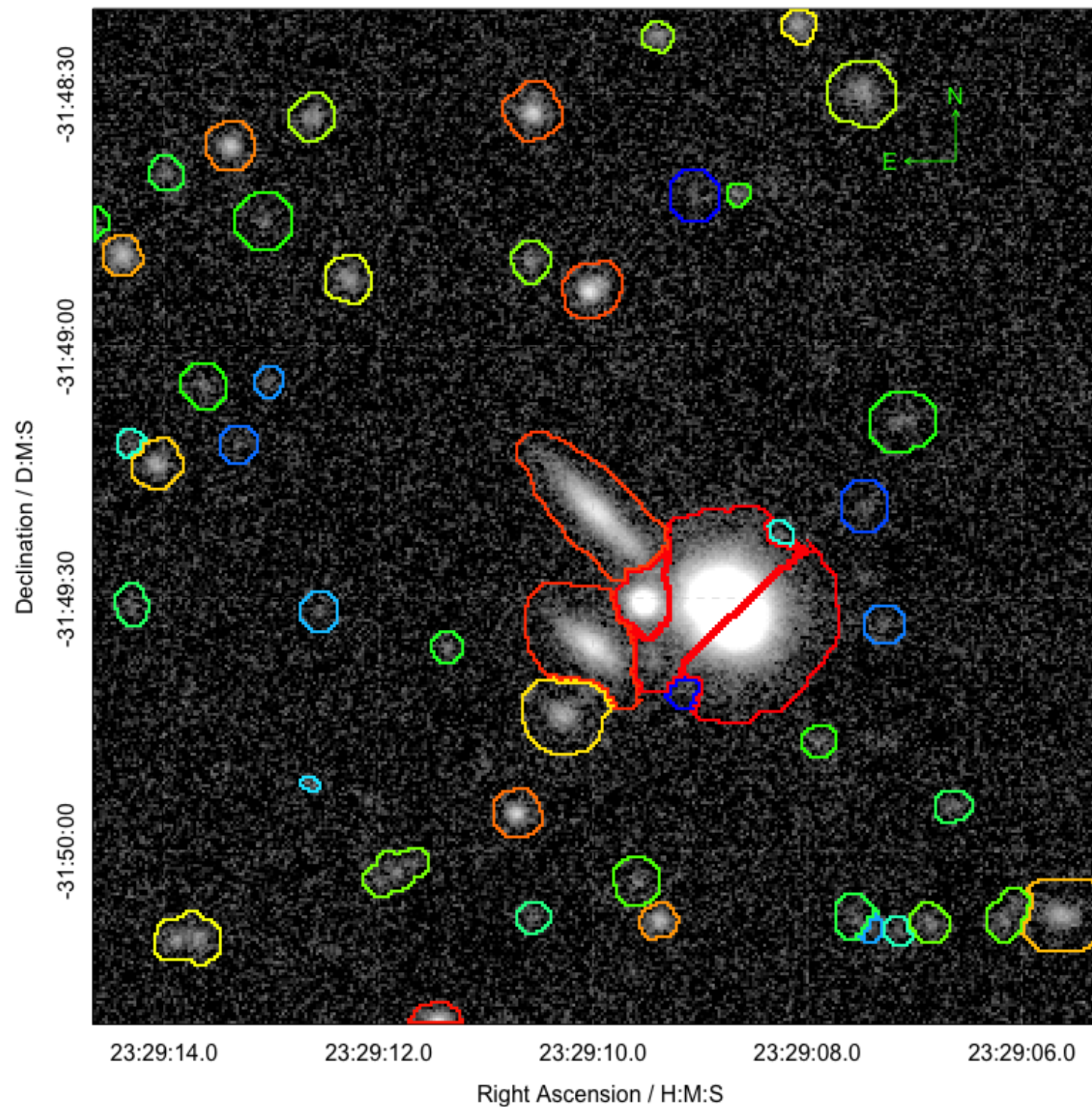
INITIAL VIKING Z-BAND IMAGE



BRIGHT SEGMENTS IDENTIFIED AND DE-BLENDED



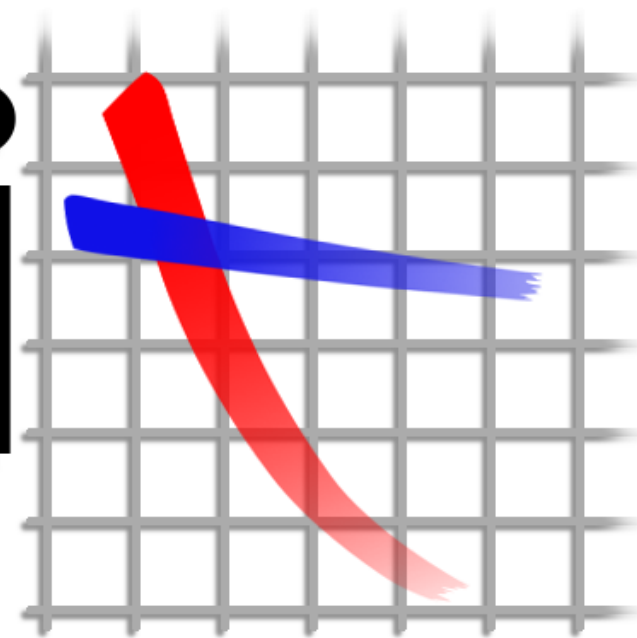
SEGMENTS DILATED UNTIL THE FLUX CONTAINED CONVERGES



PROFOUND → PROFIT

- ▶ Here we have focussed on the general photometry aspects of ProFound, but it is at least equally focussed on creating good inputs for ProFit.
- ▶ It make everything we need:
 - ▶ Source identification (star/gal separation)
 - ▶ Good initial conditions (based on photometric analysis)
 - ▶ Segmentation maps (both tight and dilated to capture *all* flux)
 - ▶ Sky (can be calculated a number of ways as appropriate)
 - ▶ Sigma maps (using sky variance and image gain)

ProFi



TAKING PHOTOMETRY FURTHER

PROFIT

MNRAS **466**, 1513–1541 (2017)
Advance Access publication 2016 November 23

doi:10.1093/mnras/stw3039

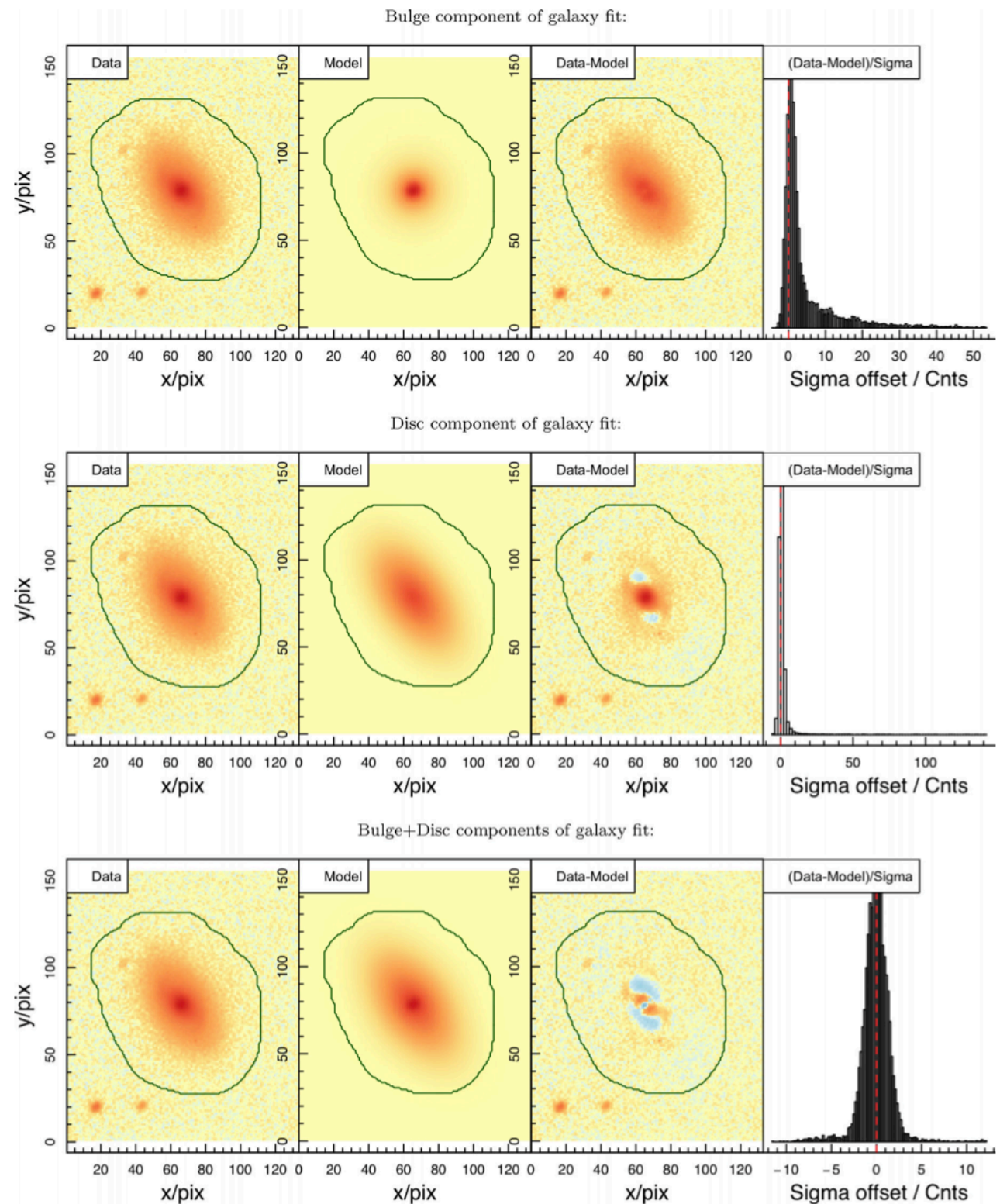
PROFIT: Bayesian profile fitting of galaxy images

GitHub: ICRAR/ProFit

A. S. G. Robotham,^{1★} D. S. Taranu,^{1,2} R. Tobar,¹ A. Moffett¹ and S. P. Driver¹

What's the Aim?

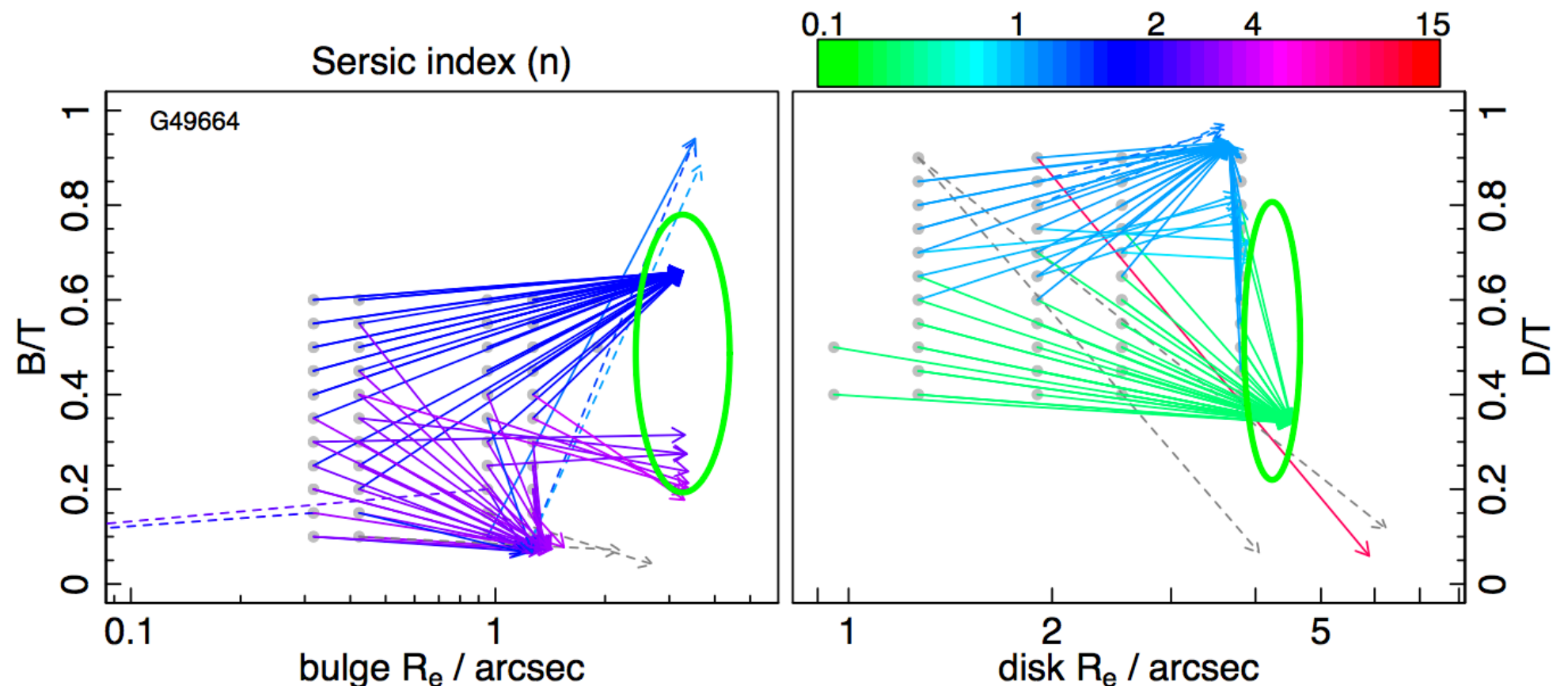
- In simple “single band mode” we want to produce a 2D mixture model of a galaxy image.
- This usually separates a galaxy into a compact bulge and an extended disk.



What's the Problem?

- Work by ICRAR PhD R. Lange showed that the code of popular community choice (GALFIT) struggles with convergence for many galaxies:

Convergence plots



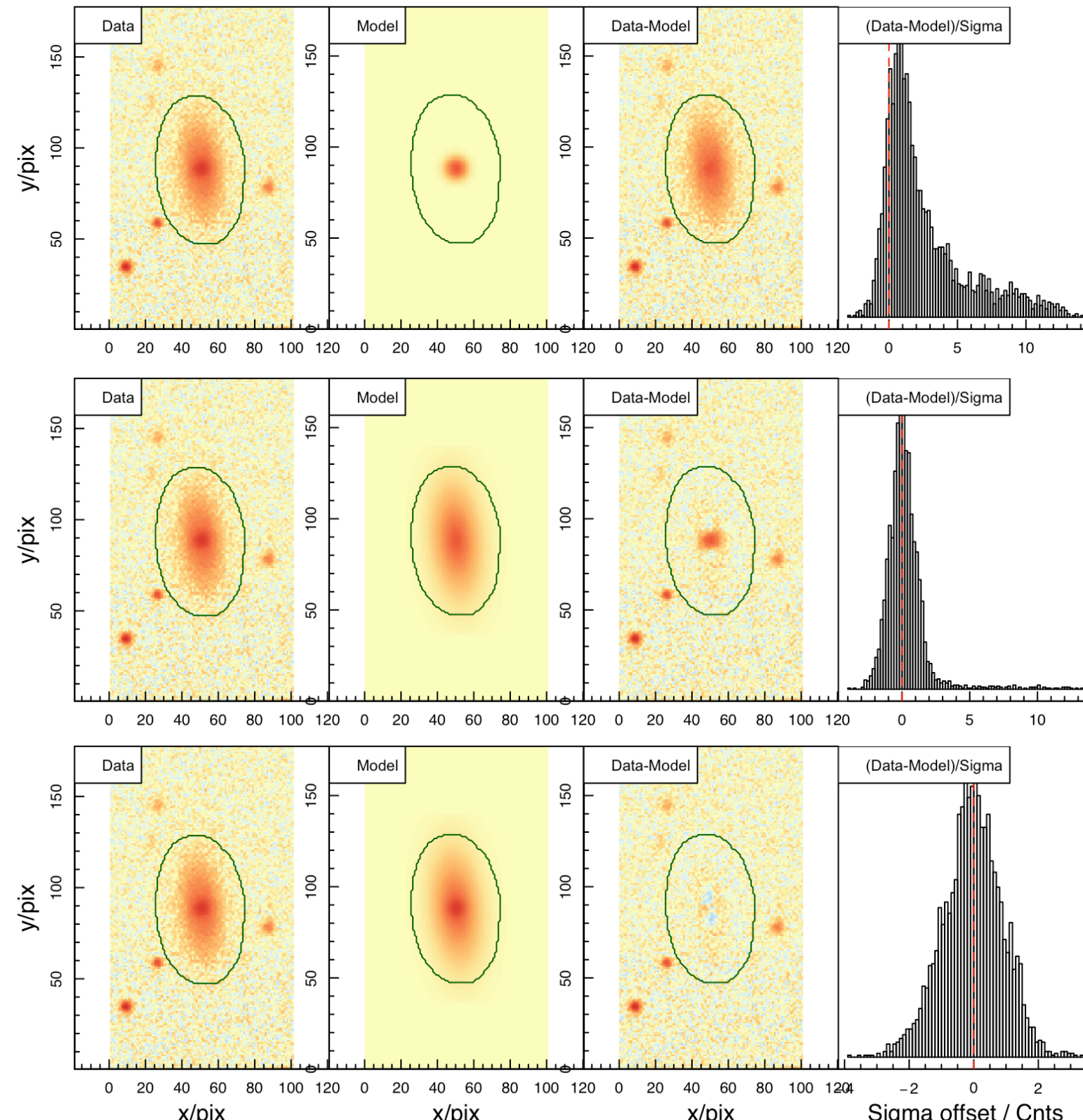
What to do?

- Options were to modify existing code to use more sophisticated samplers, but GALFIT is closed license (binaries only) and other codes available at the time were not battle-tested.
- Vitally for ProFuse (already in the planning in 2017), we need image generation capability, not just ‘fitting’.
- Best option all round was to start again and develop a rapid image generation library written in C++ (libprofit) and higher level interfaces to achieve the fitting (ProFit, initially Python and R).
- Initial version was written by me, subsequently R. Tobar (libprofit / pyprofit) and D. Taranu (ProFit) have added a lot of additional features and code re-factors.

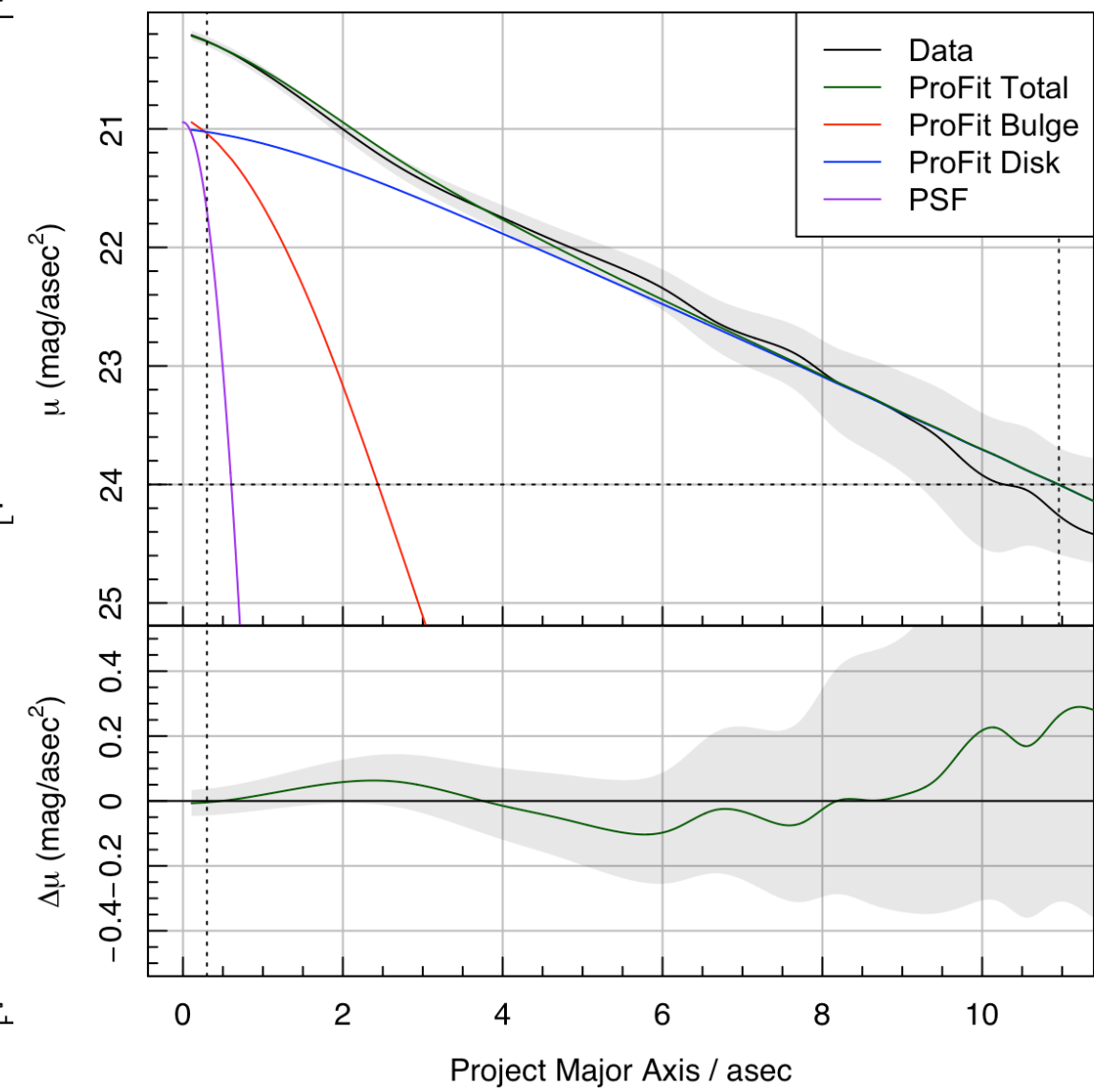
What Does libprofit Do For You?

- Profiles supported in libprofit:
 - ▣ Sersic: Popular for galaxy components.
 - ▣ Core-Sersic: Popular for early-type galaxies.
 - ▣ Moffat: Popular for stars / PSFs.
 - ▣ Ferrer/s: Popular for galaxy bars.
 - ▣ King: Popular for globular clusters.
 - ▣ Empirical PSF: Generic point spread function.
 - ▣ Sky: The sky background.
- Easy (relatively) to add more profiles: ~30 lines of C++ code to describe the 1D radial profile shape and total flux calculation.

Decomposing Bulges and Disks with ProFit

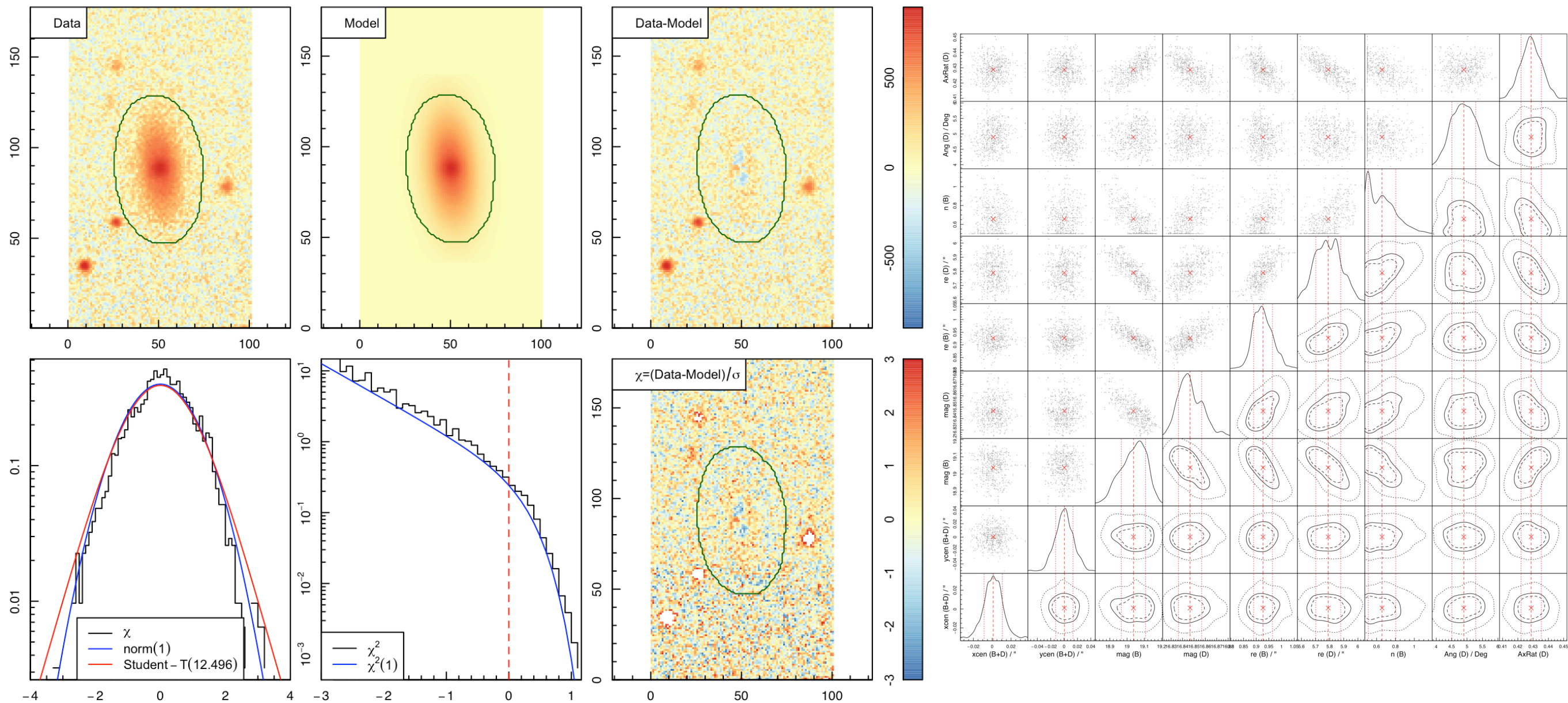


Can achieve converged MCMC fits with useful errors!



Decomposing Bulges and Disks with ProFit

- Errors are provided via full covariance information.



ProSpect

TAKING SPECTRA TO THE EDGE

PROSPECT

MNRAS **495**, 905–931 (2020)
Advance Access publication 2020 April 26

doi:10.1093/mnras/staa1116

PROSPECT: generating spectral energy distributions with complex star formation and metallicity histories

A. S. G. Robotham^{1,2}★, S. Bellstedt¹, C. del P. Lagos^{1,2}, J. E. Thorne¹,
L. J. Davies¹, S. P. Driver¹ and M. Bravo¹

GitHub: [asgr/ProSpect](https://github.com/asgr/ProSpect)

PROSPECT CODE GOAL

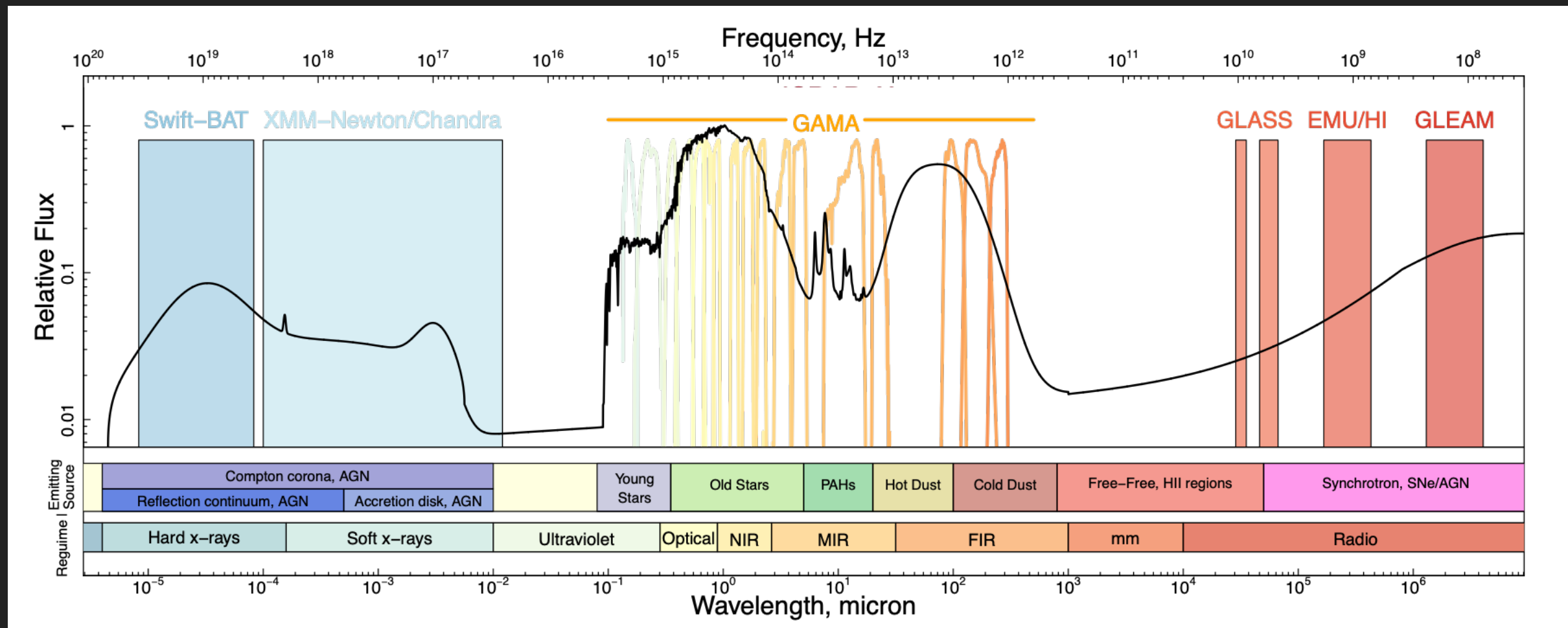
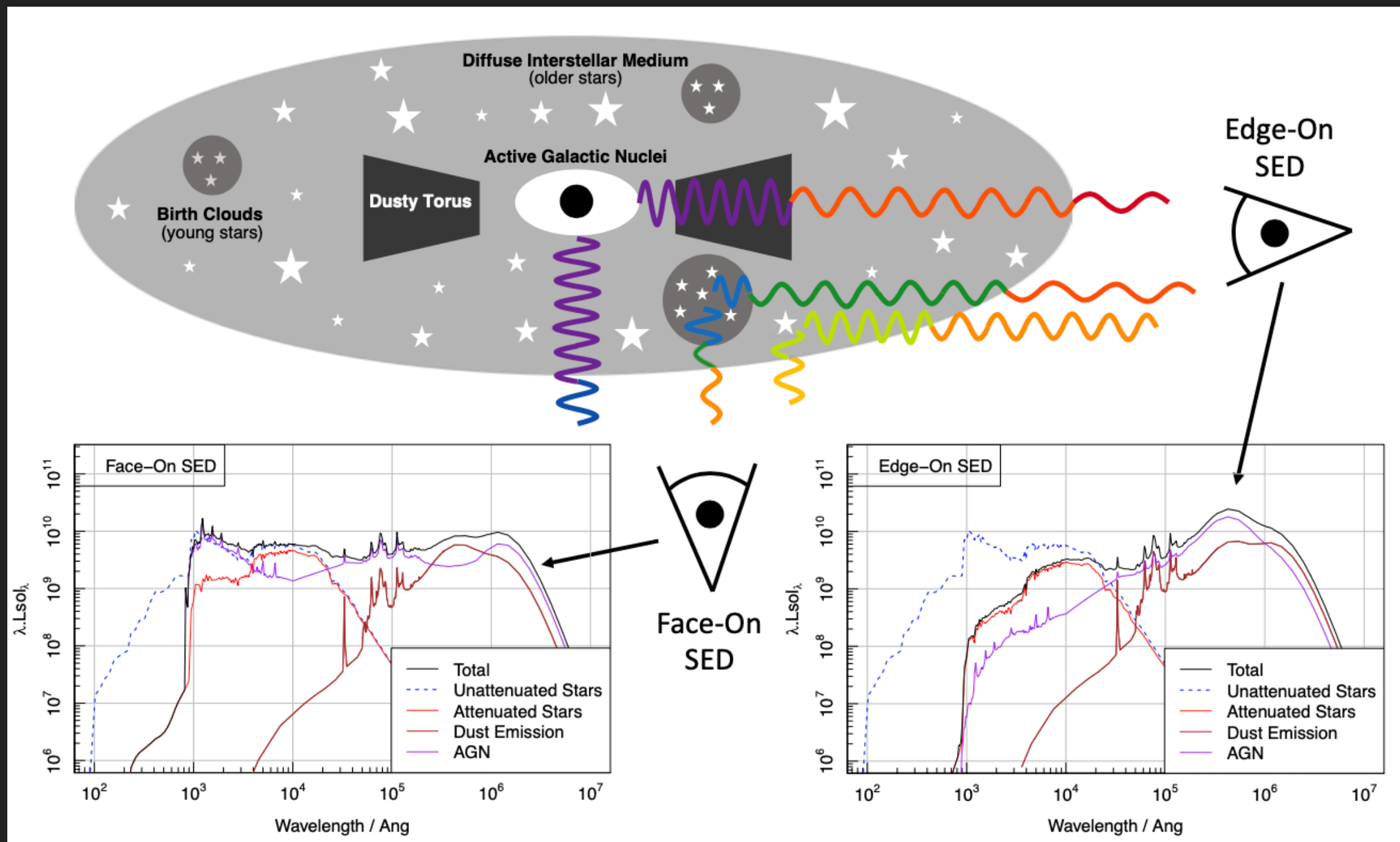
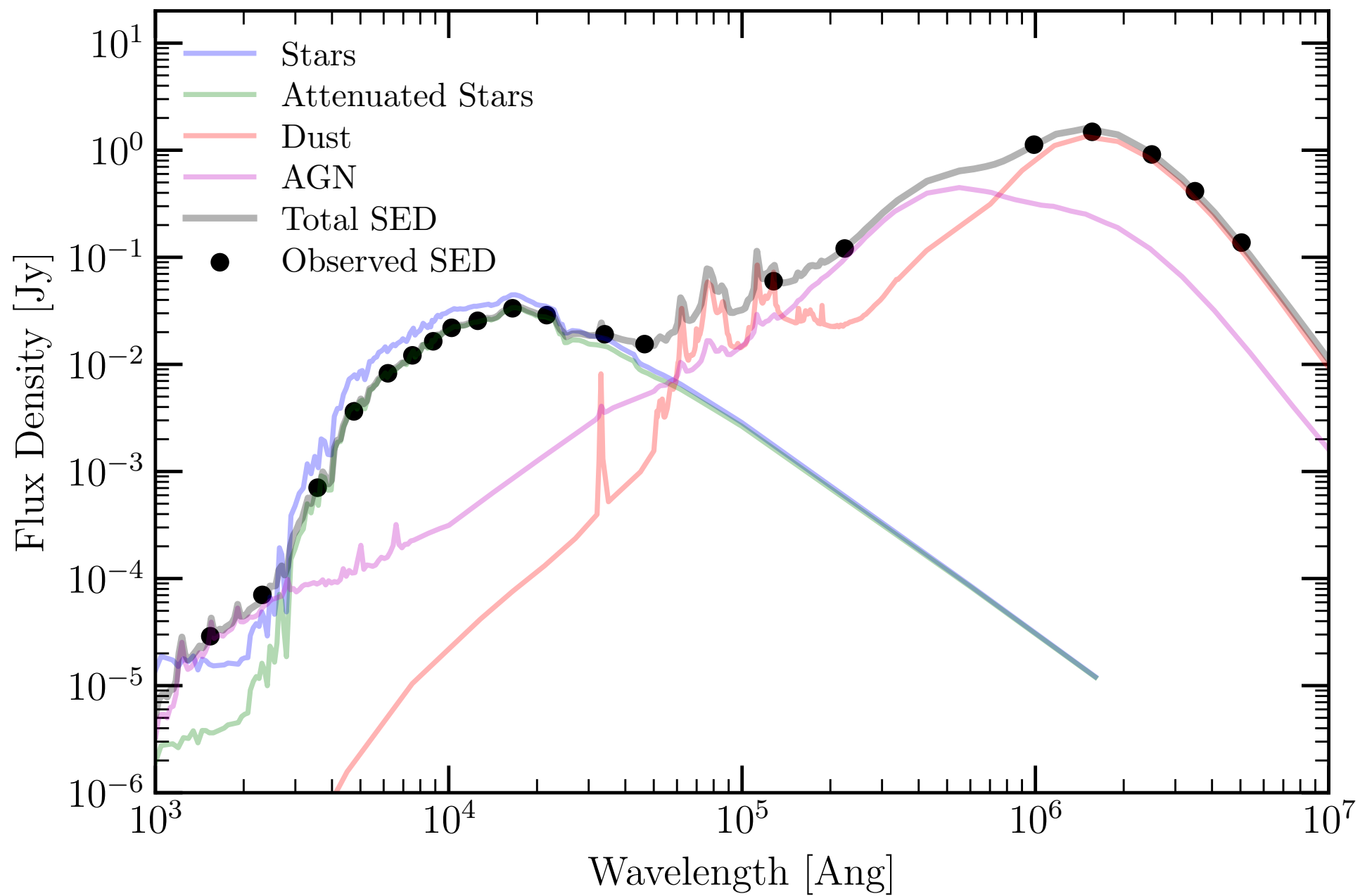


Figure credit: Luke Davies

GALAXY MODEL



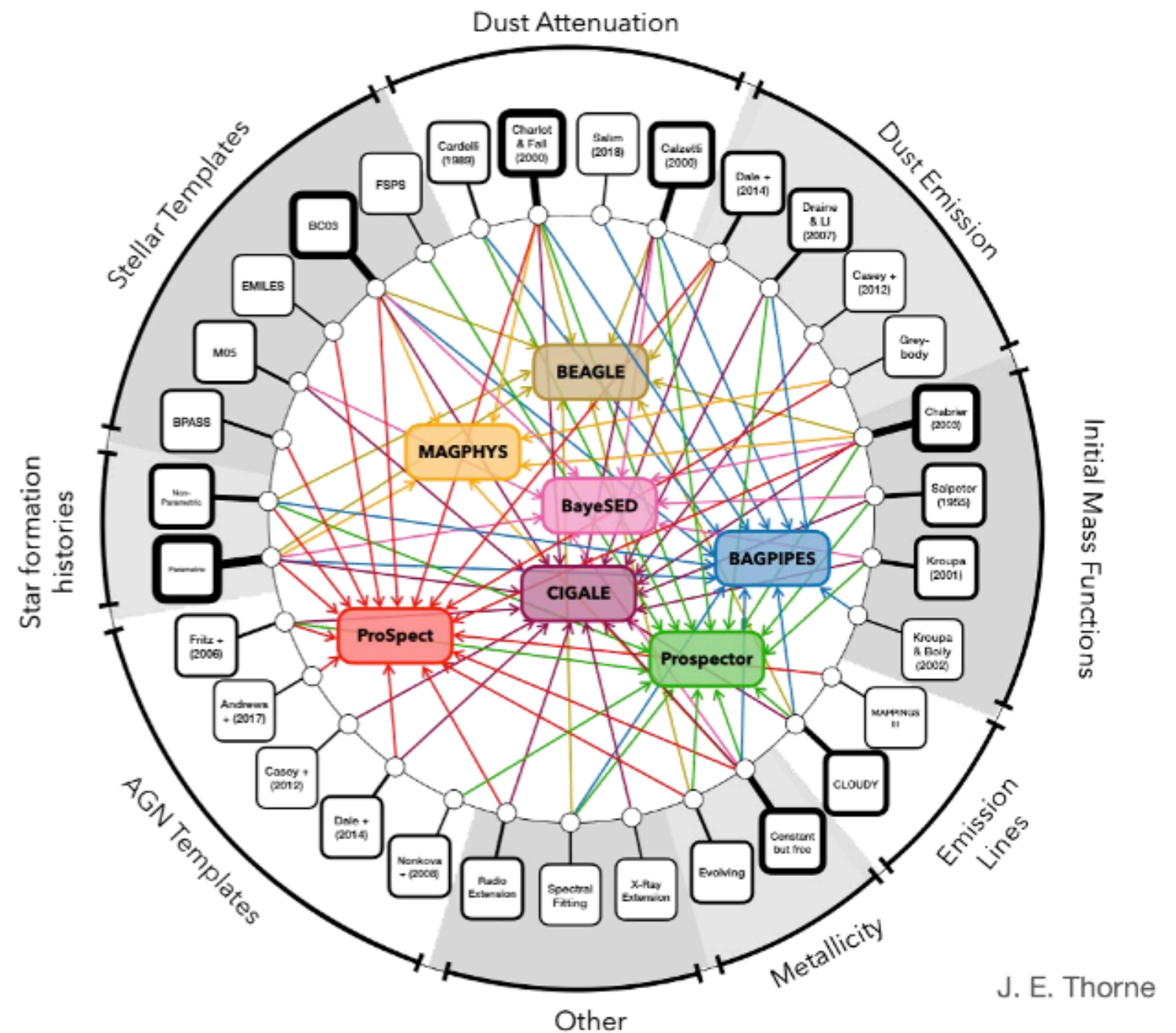
UV - FIR



PROSPECT VERSUS

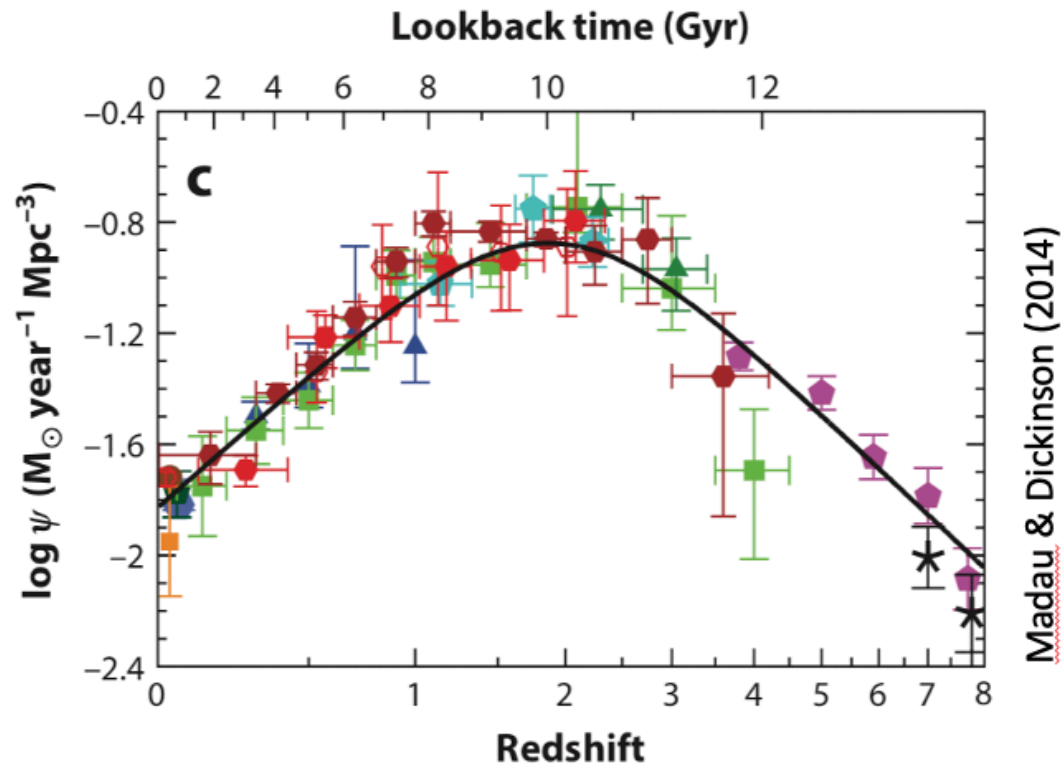
Many SED-fitting tools exist.

They all do roughly the same thing (model an SED via its stellar and dust components), but all in slightly different ways...

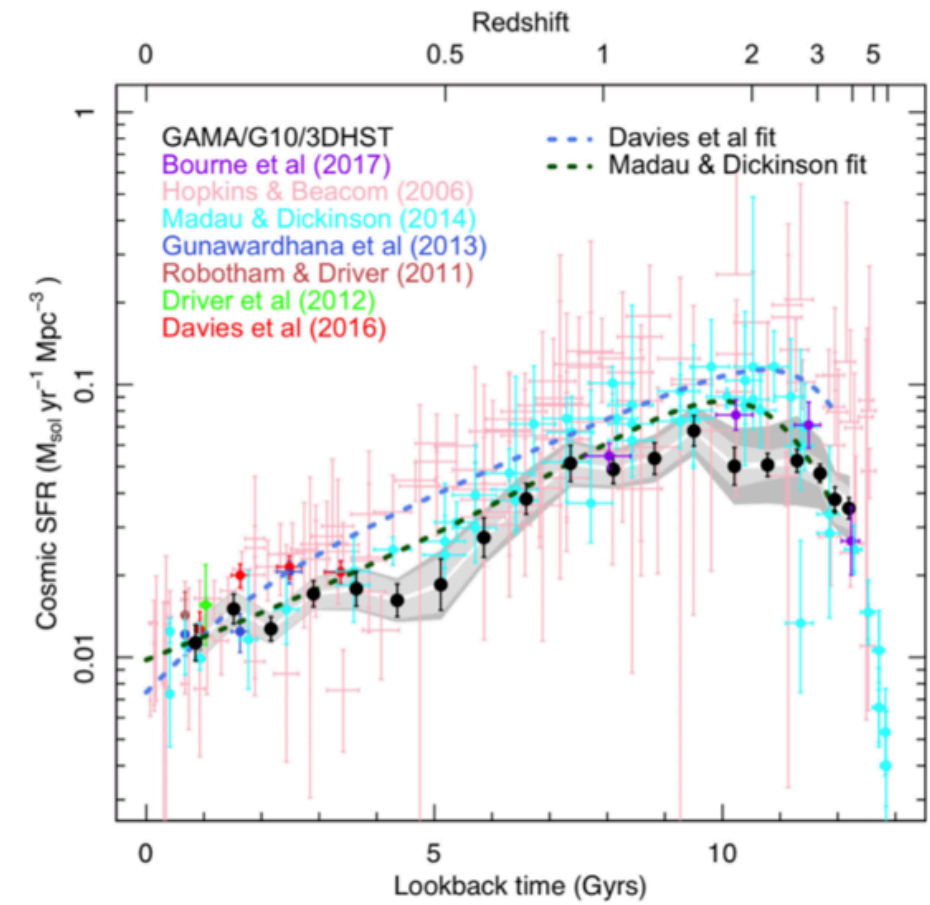


Again, with ProFuse in mind we do not just want an SED 'fitting' code, we need a generative model.

PROSPECT ON GAMA (BELLSTEDT+ 2020)

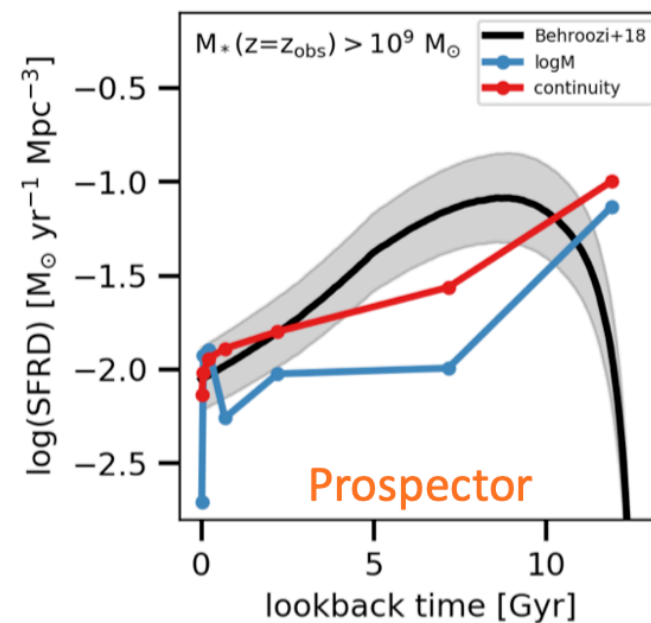


Madau & Dickinson (2014)

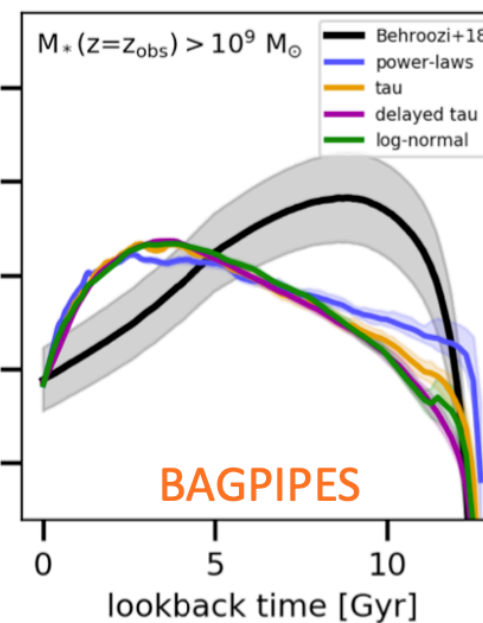


Driver et al. (2018)

A challenge for SED-fitting techniques

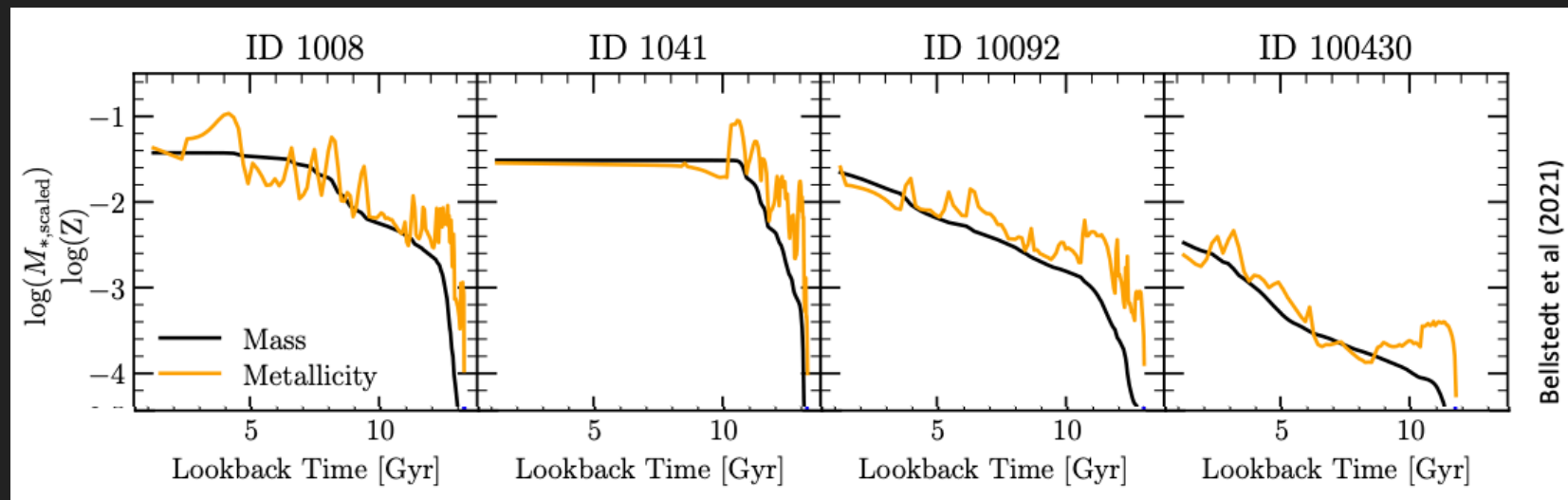


Leja et al. (2018)

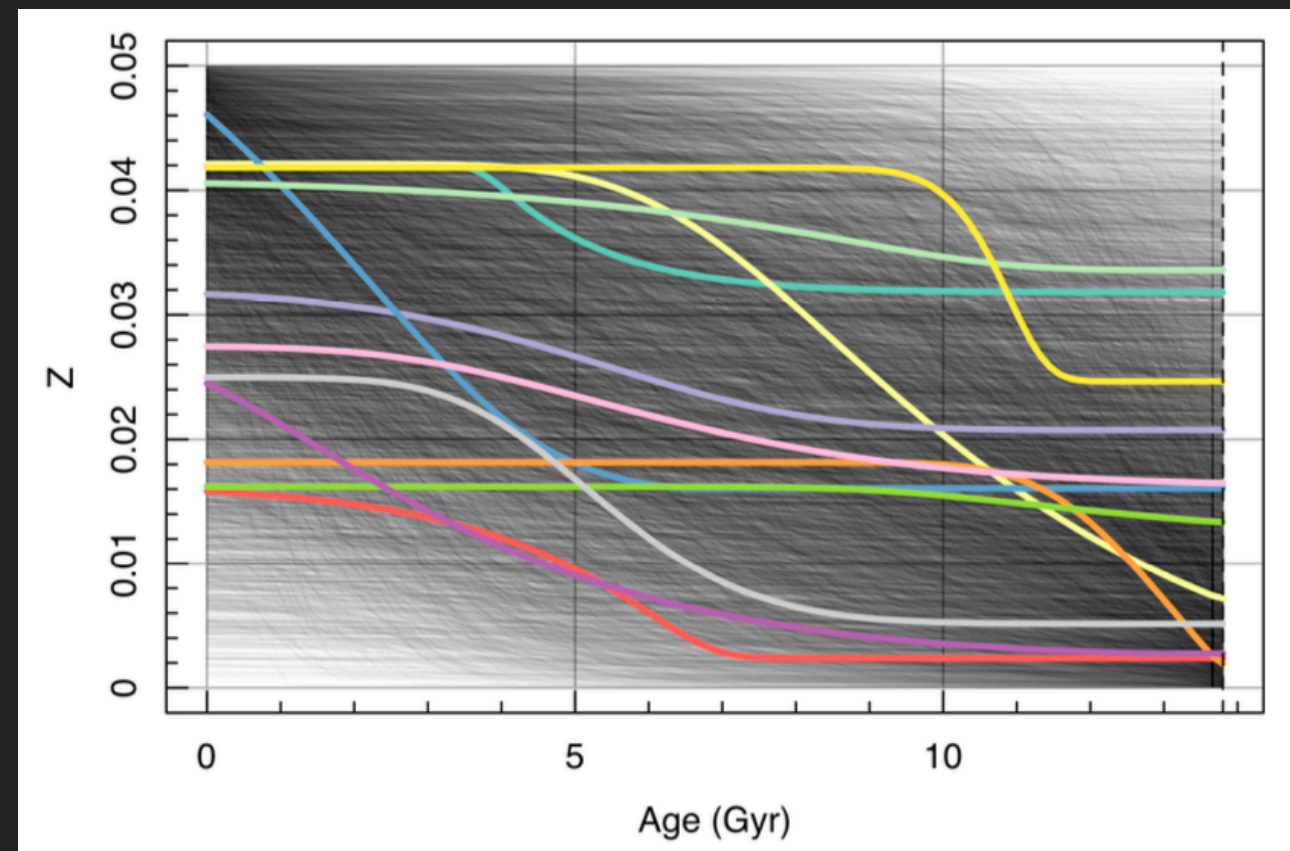
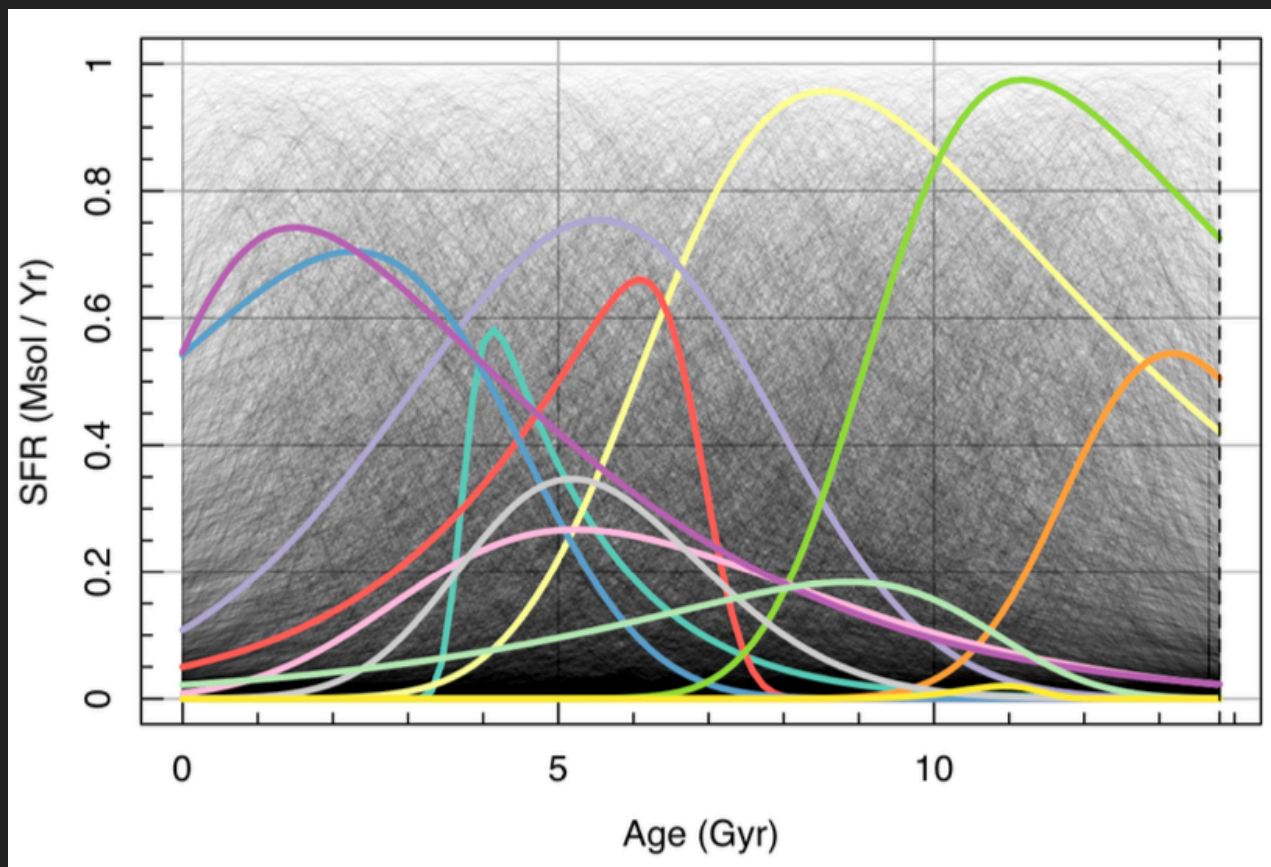


Carnall et al. (2018)

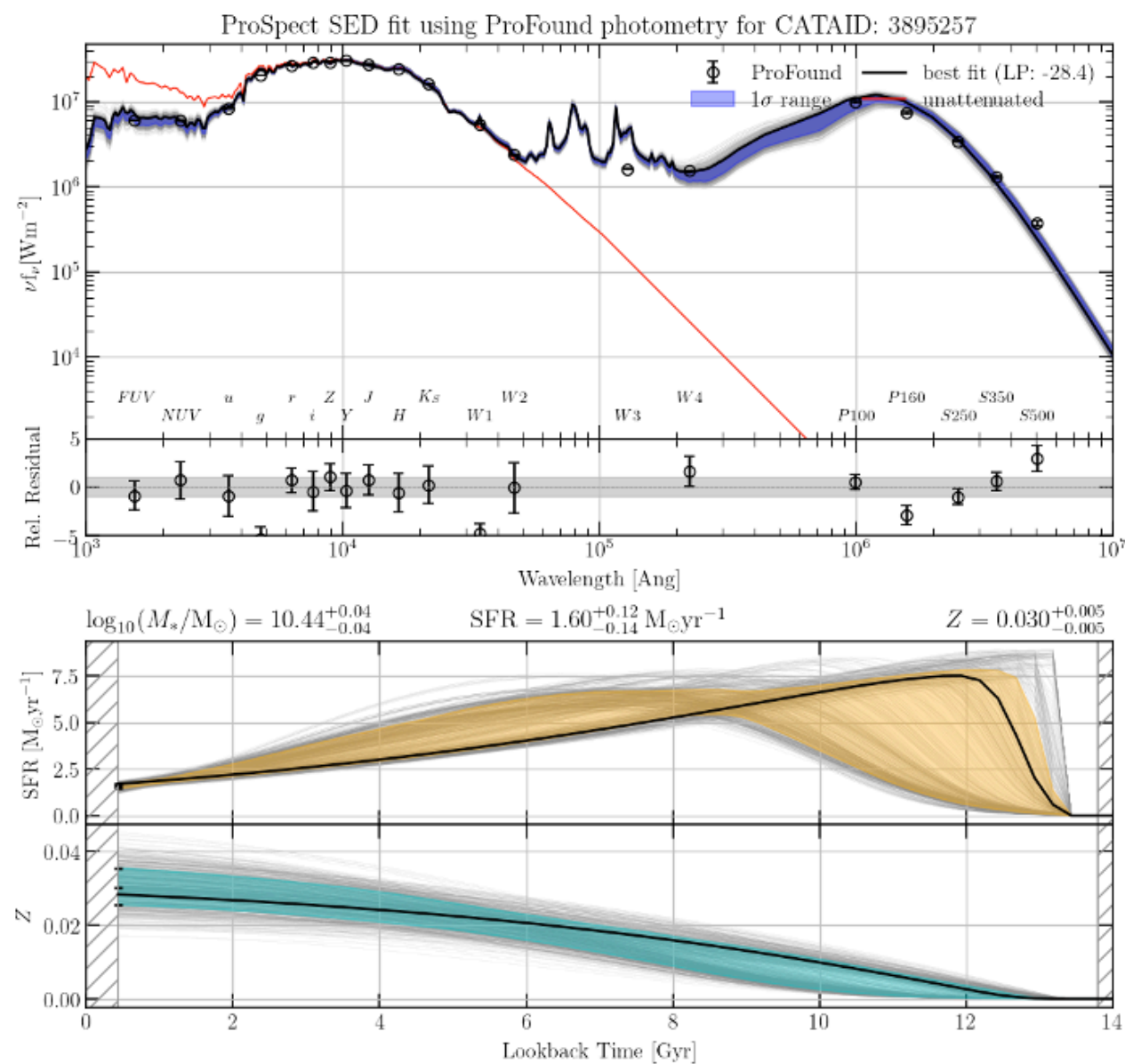
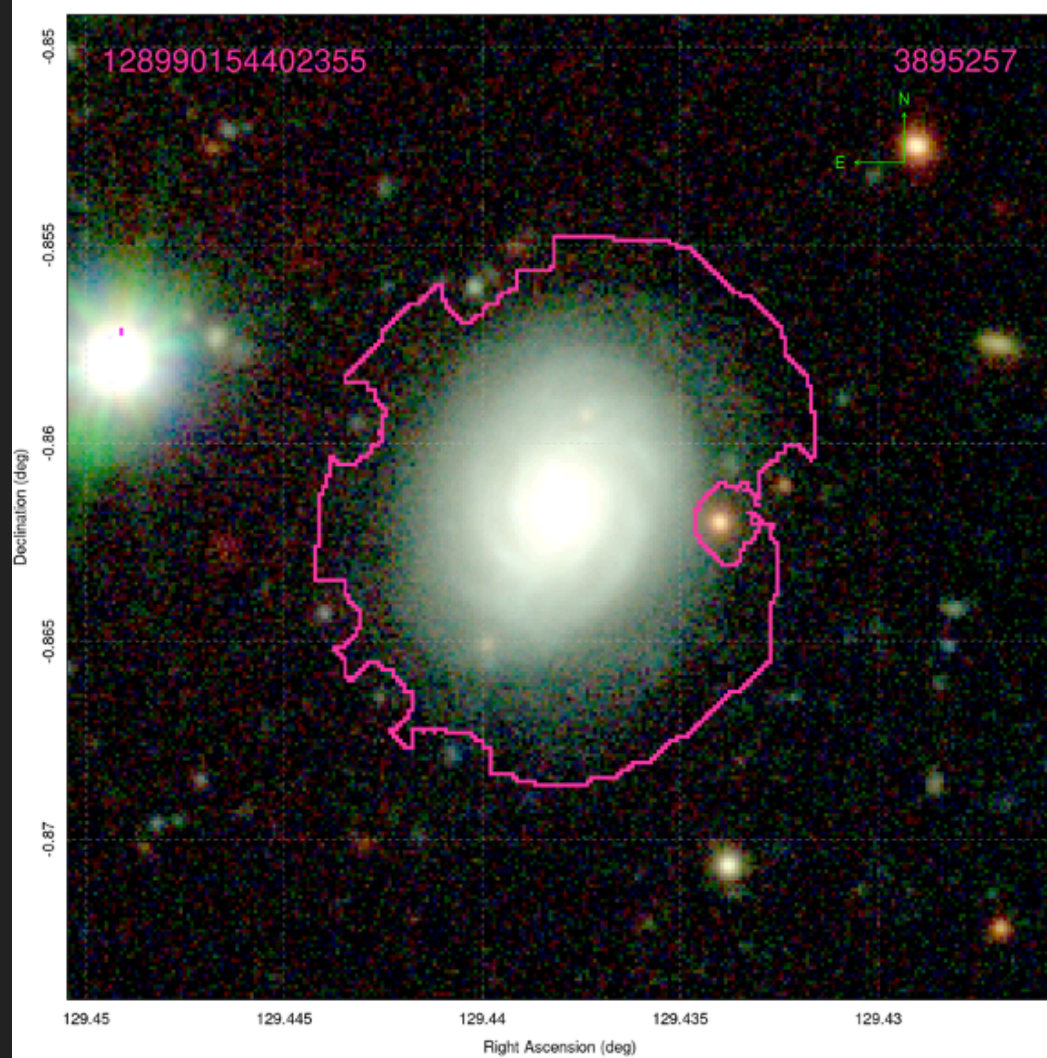
PROSPECT SFH AND ZH



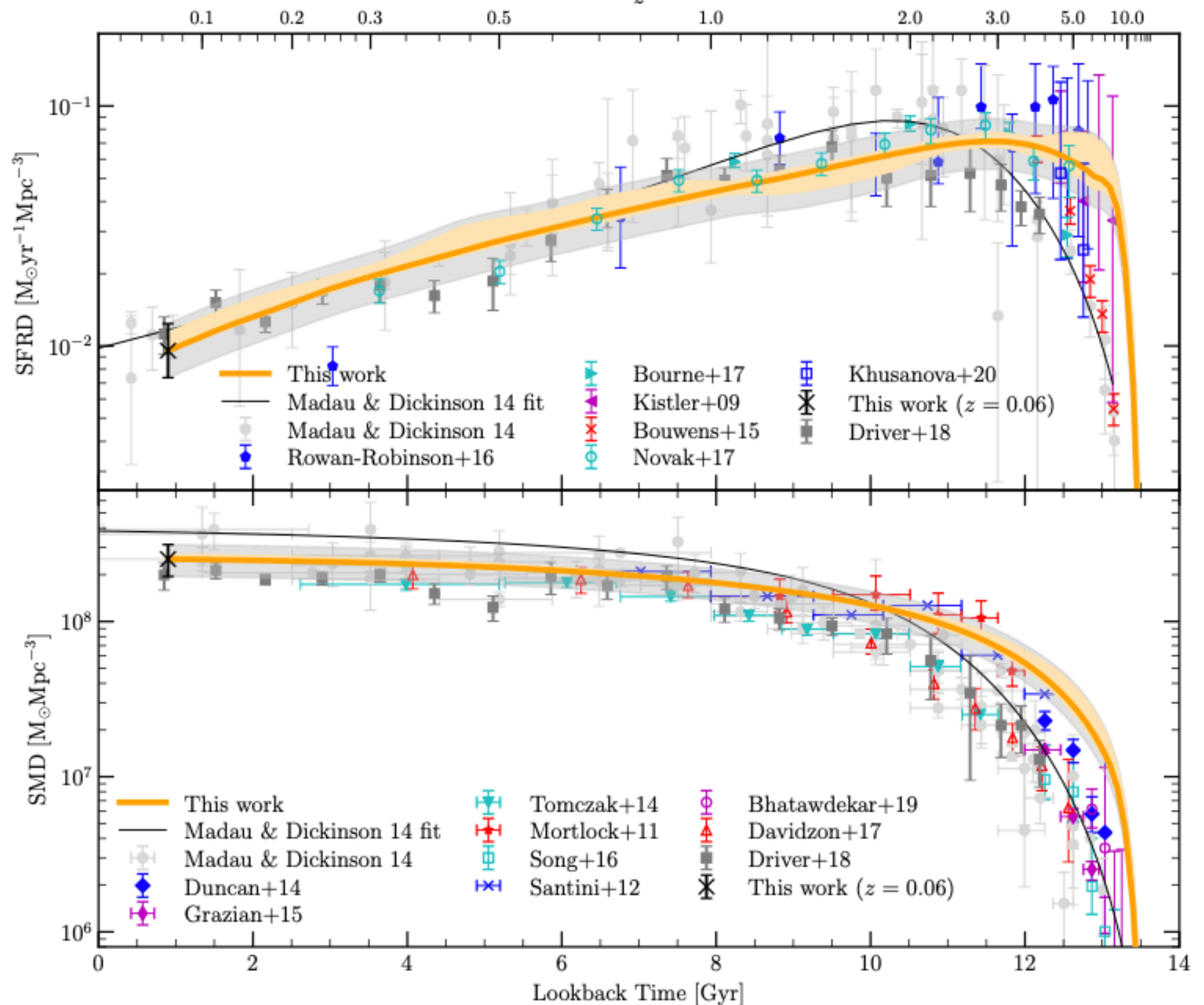
Allows almost any, but we focus on skewed Normal SFH, and linearly mapped ZH:



PROSPECT ON GAMA (BELLSTEDT+ 2020)

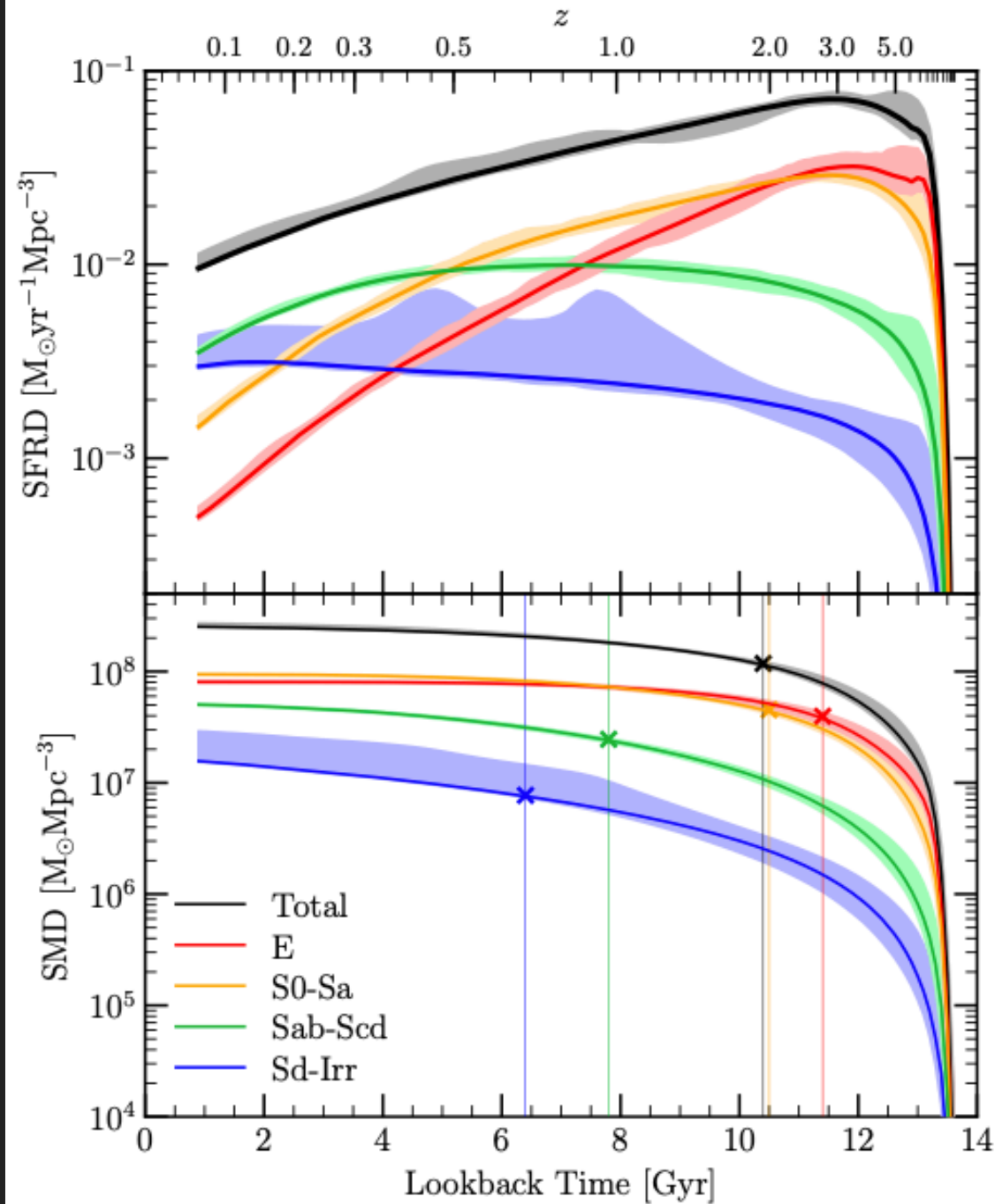


cosmic star
formation
history



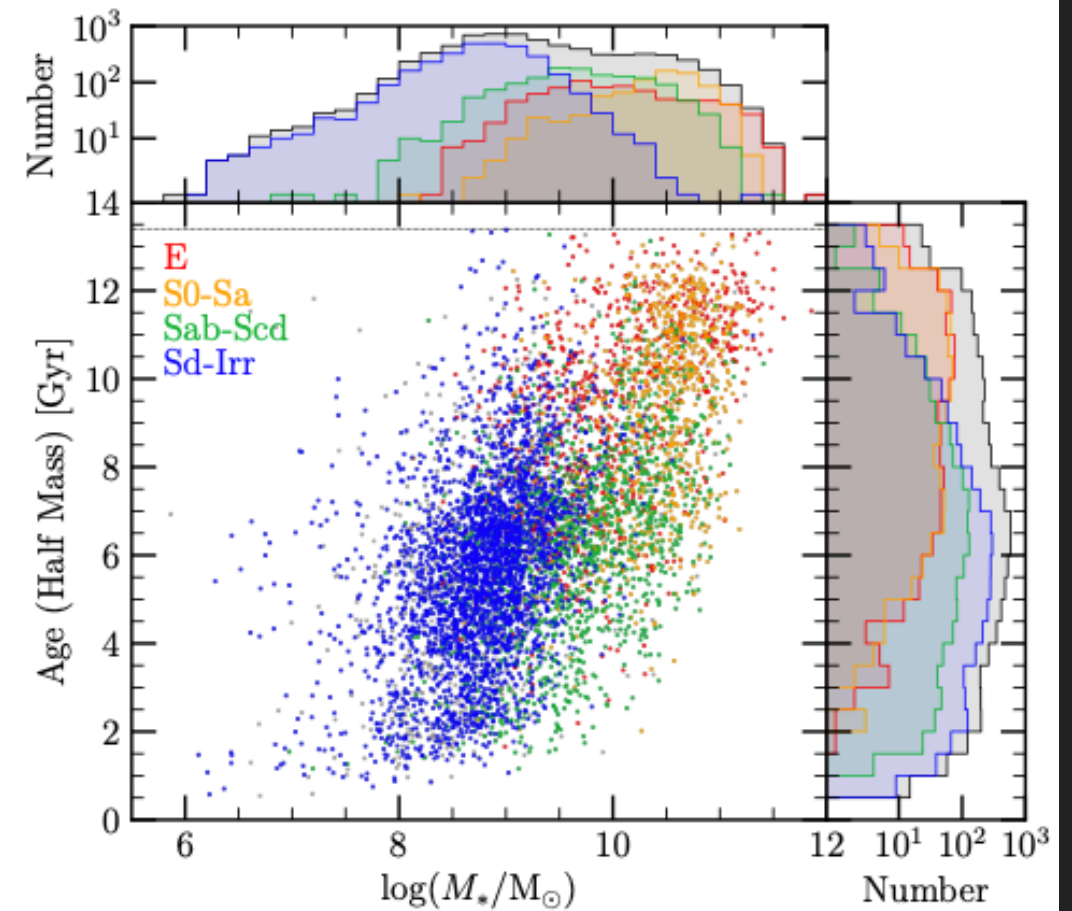
cosmic
stellar mass
build-up

PROSPECT ON GAMA (BELLSTEDT+ 2020)

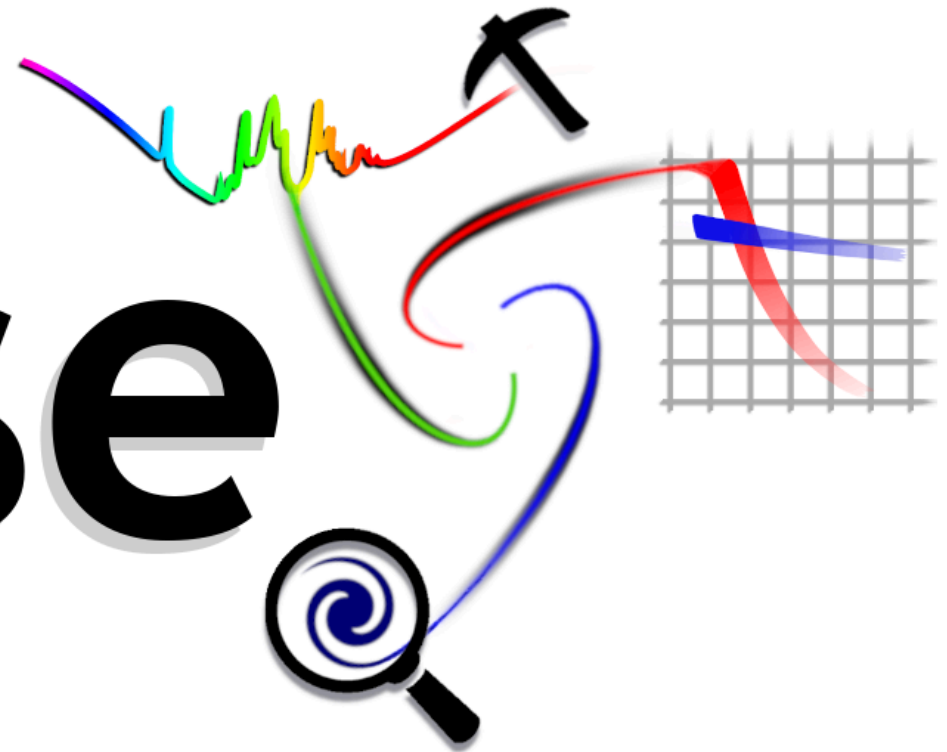


By morphology:

Early-type galaxies are older,
whereas **late-type** galaxies were
formed more recently.



*Pro*Fuse



PUTTING THIS ALTOGETHER

PROFUSE

MNRAS **513**, 2985–3012 (2022)
Advance Access publication 2022 April 15

<https://doi.org/10.1093/mnras/stac1032>

ProFuse: physical multiband structural decomposition of galaxies and the mass–size–age plane

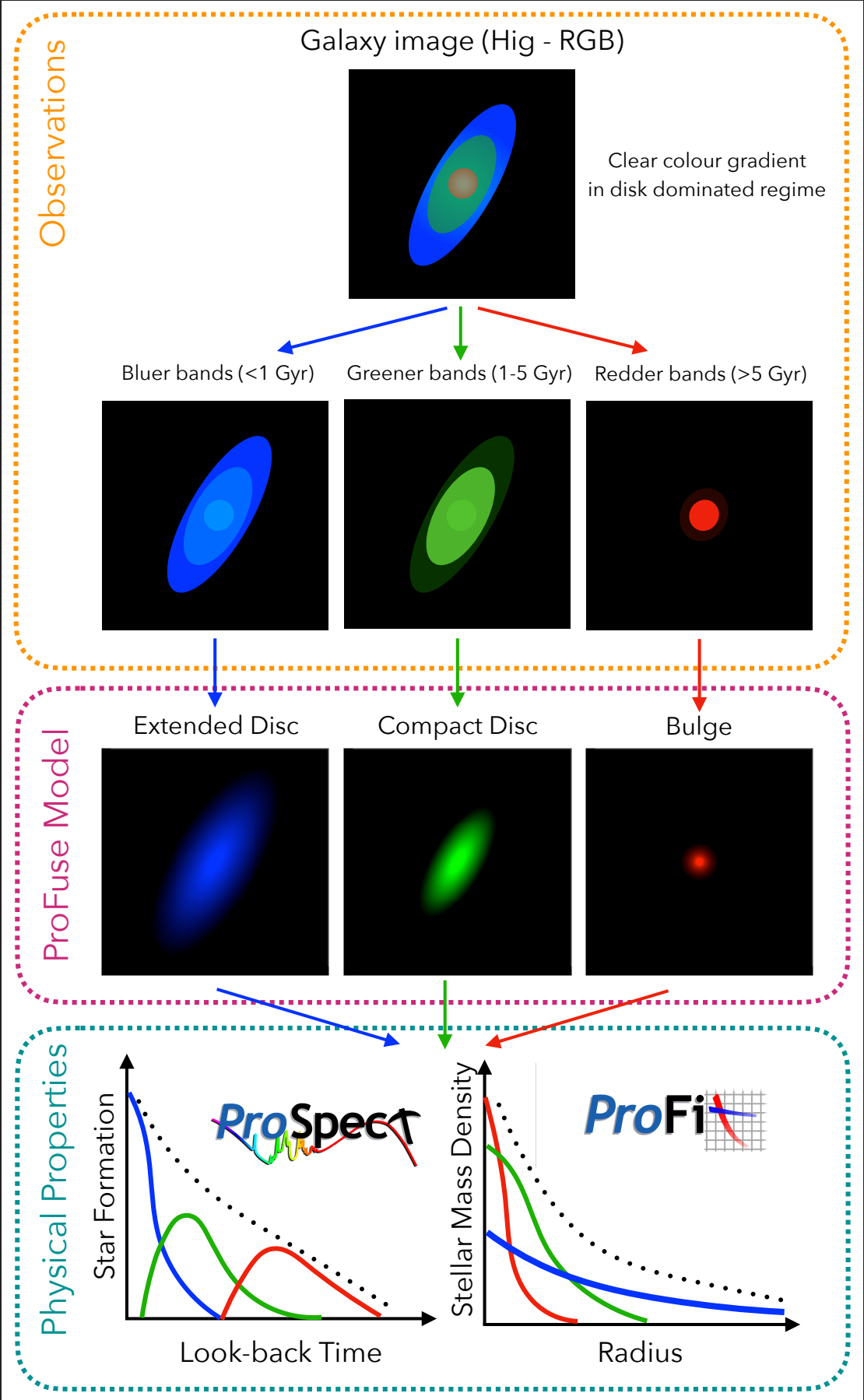
GitHub: [asgr/ProFuse](https://github.com/asgr/ProFuse)

A. S. G. Robotham ^{1,2}★ S. Bellstedt ¹ and S. P. Driver ¹

- ▶ Starting from just multi-band image data we want to:
 - ▶ Detect object
 - ▶ Estimate sky
 - ▶ Find PSFs
 - ▶ Fit bulge + disk model with distinct SED for each component

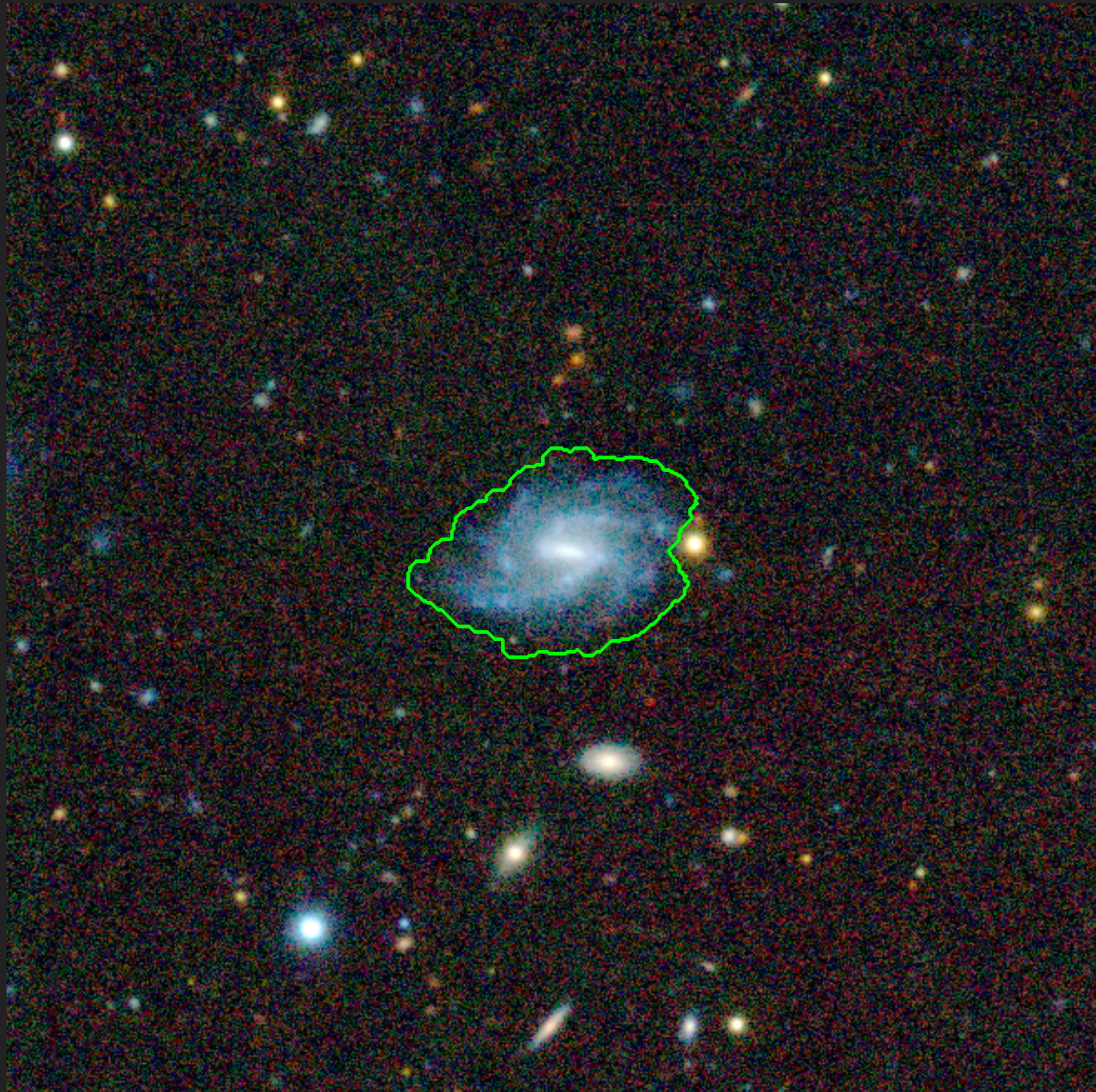


PROFUSE IN ONE SCHEMATIC

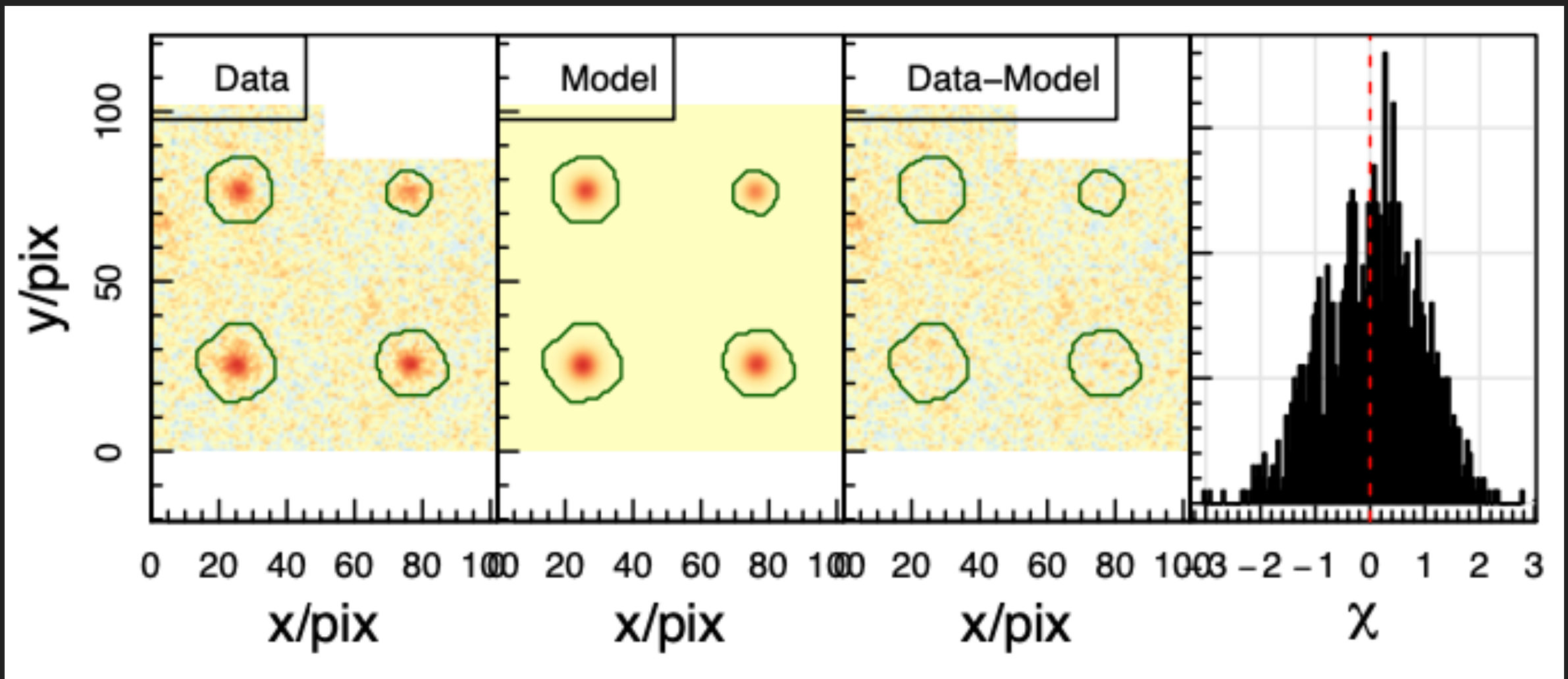


PROFUSE IN ANGER - IMAGE SEGMENTATION AND STAR DETECTION (PROFOUND)

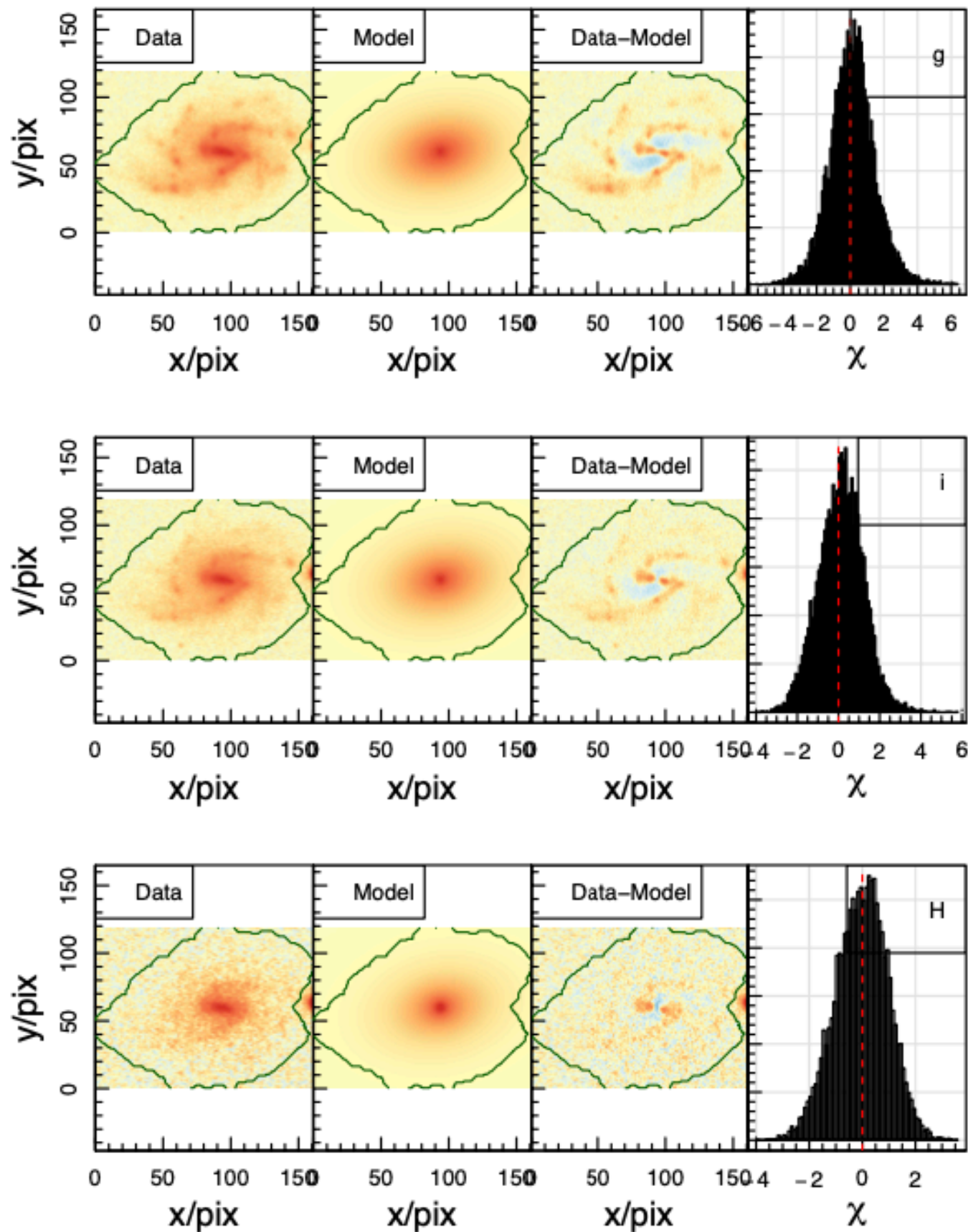
- ▶ Objects are roughly detected in all bands, and sky / sky-RMS estimated.
- ▶ An inverse variance stack is created, and a global detection is carried out (deep).
- ▶ Good candidate stars are detected per band, and Moffat PSFs made.



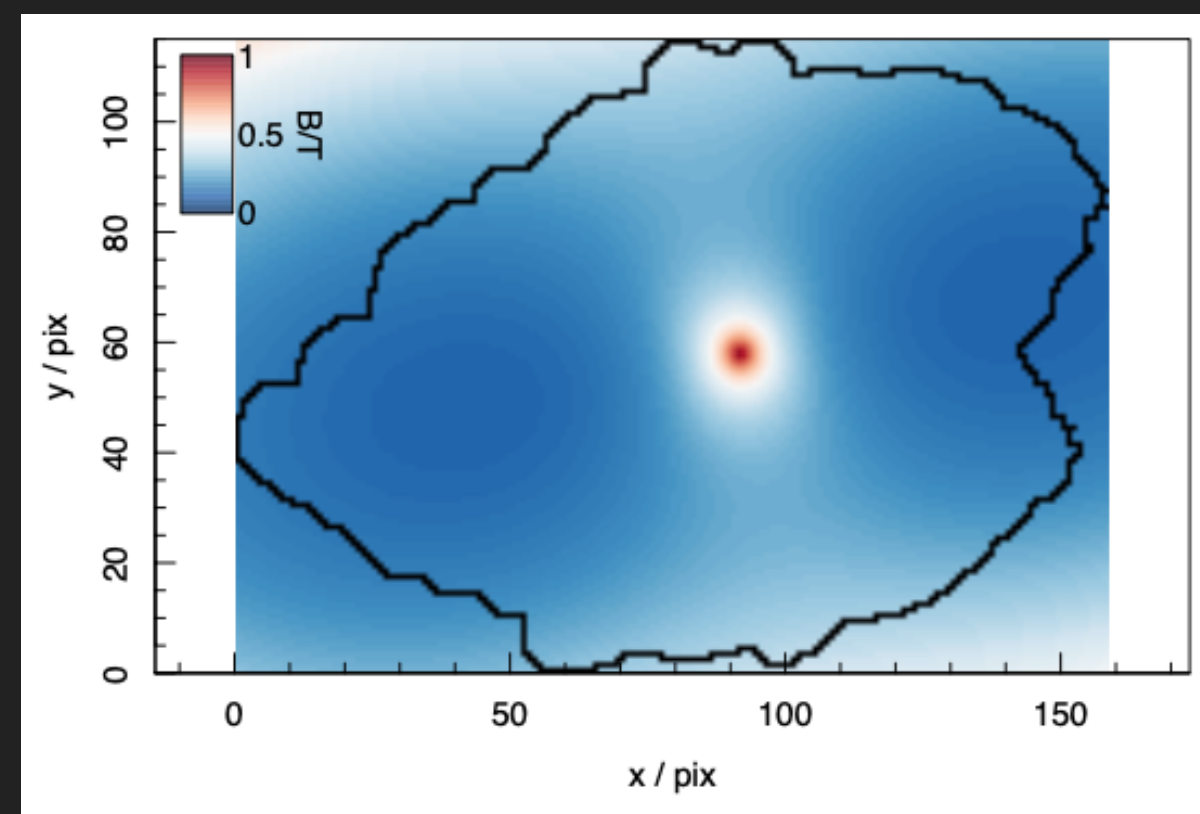
PROFUSE IN ANGER - AUTOMATIC PSF IN EVERY BAND



PROFUSE IN ANGER - SPECTRAL + SPATIAL DECOMPOSITION (PROSPECT + PROFIT)

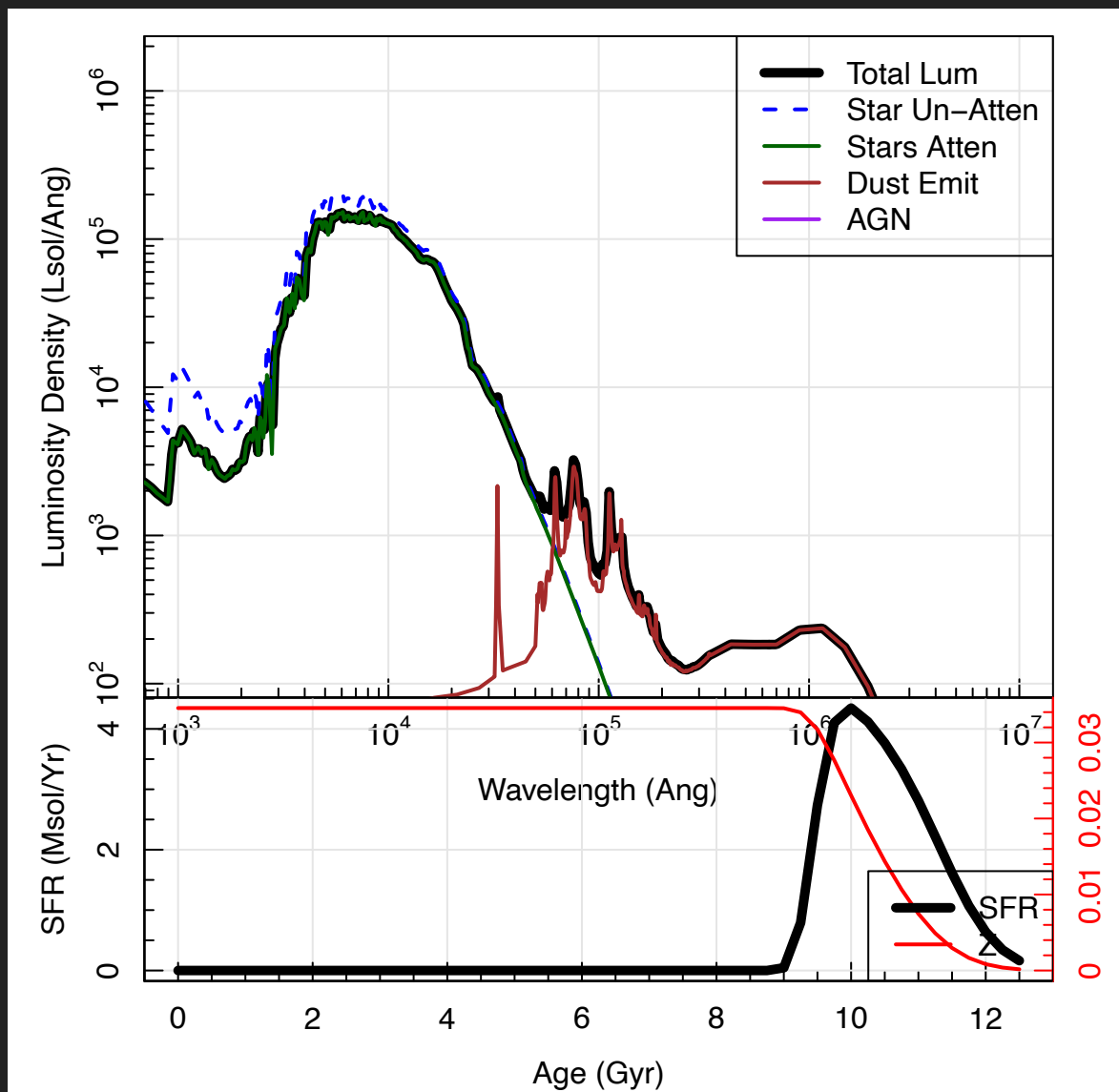


B/T Map

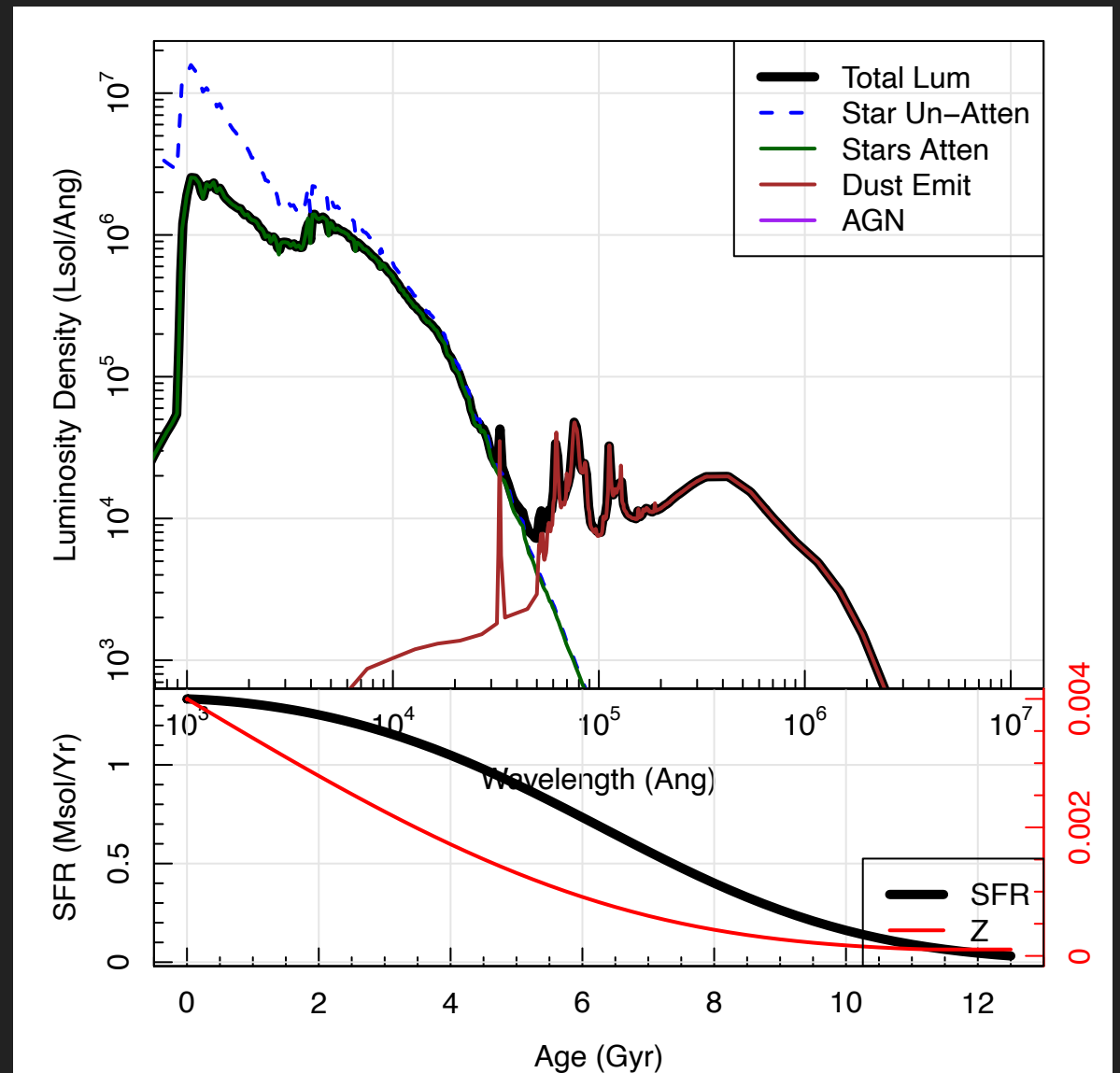


PROFUSE IN ANGER - INVESTING BULGE / DISK SEDS (PROSPECT)

Bulge



Disk

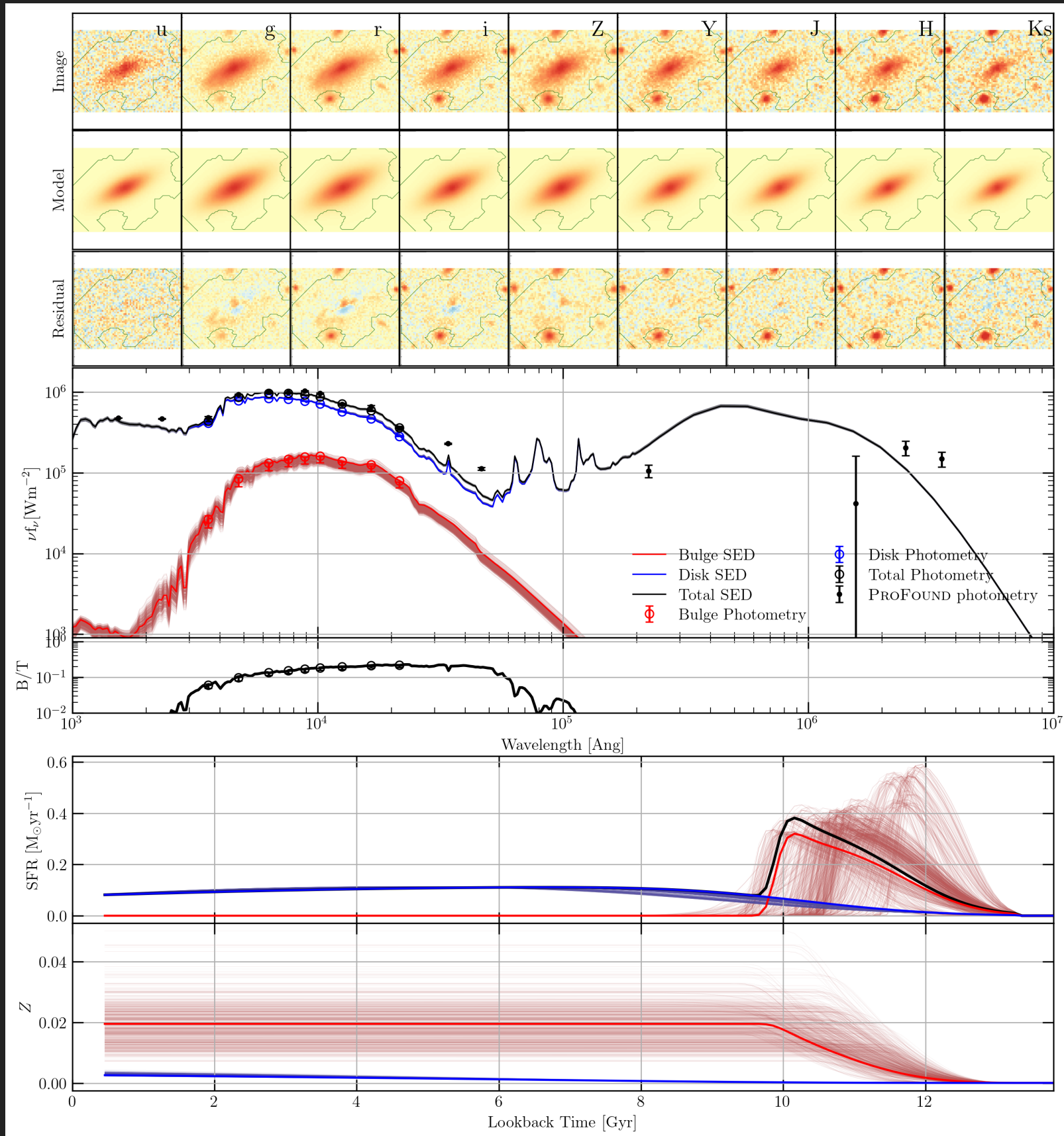


- ▶ We get all the usual outputs of ProSpect (SM, SFR, SFH, ZH, and lots more), but per component rather than for the ensemble galaxy.
- ▶ This allows us to generate multi-epoch stellar populations that are not possible in ProSpect with a single skewed Normal SFH.

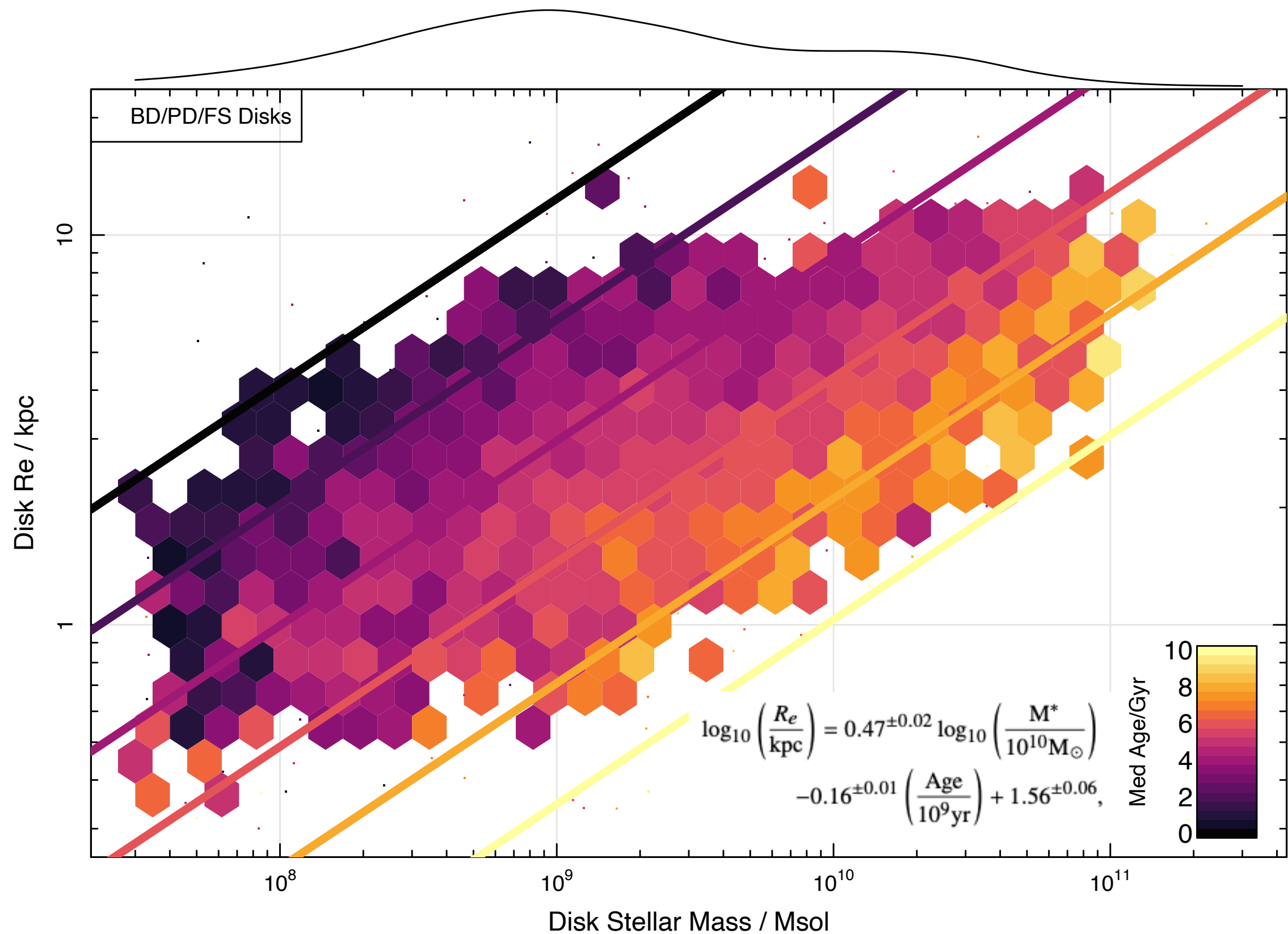
WITH GREAT POWER...

- ▶ In this first application we restrict ourselves to relatively simple models:
 - ▶ Free Sersic [FS] - a single profile that better suits single component systems with a dominant SFH/ZH (elliptical).
 - ▶ Bulge ($n=4$) + Disk ($n=1$) [BD] - a bulge-disk fit that allows for an extended bulge.
 - ▶ Bulge (PSF) + Disk ($n=1$) [PD] - a bulge-disk fit that allows for a PSF (unresolved) bulge.
 - ▶ Bulge ($n=4$) + Disk ($n=1$) + Disk ($n=1$) [BDD] - a bulge-double-disk fit that allows for an extended bulge and two disk components. These interact to create SFH/ZH and colour gradients.

PROFUSE ON GAMA (~7K GALAXIES, $Z < 0.06$) - ROUGHLY 2-4 HOURS PER FIT



PROFUSE ON GAMA (~7K GALAXIES, Z<0.06) - THE MASS-SIZE-AGE PLANE



PROFUSE LIVES!

- ▶ ProFuse combines all of the ProTools developed to date to provide a fully automated Bayesian Inference engine for the spectral-spatial decomposition of galaxies:
 - ▶ ProFound detects sources, segments the galaxy of interest and identifies stars.
 - ▶ Stars are then modelled with ProFuse in single-band mode to provide per-band PSFs.
 - ▶ ProSpect and ProFit are then run in tandem to produce SFHs and ZHs of bulges and disks simultaneously.
 - ▶ In principle arbitrarily complex models are allowed, but with the first application with GAMA we only push as far as a Bulge + Disk +Disk (BDD) model. Mostly FS and BD models are preferred.

PROFUSE EXPERIENCE

- ▶ Working to build a code as complicated as ProFuse has taken a significant fraction of my time over the last 7 years.
- ▶ Whilst it has supported many students (~10) and post docs (~2) along the way, the core task itself was too major and longterm to be supported via traditional academic means, i.e.:
 - ▶ PhDs (still only 3 years in Australia)
 - ▶ Post docs (usually only 3 years)
 - ▶ Grants (max typically ~4) and my current Future Fellowship position is not to develop any tools- the proposal to develop ProFuse was actually rejected in 2018 (too 'risky' and not directly 'scientific' enough).