

# Leveraging Empirical Measurements to Aide Galaxy Modeling

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w/  
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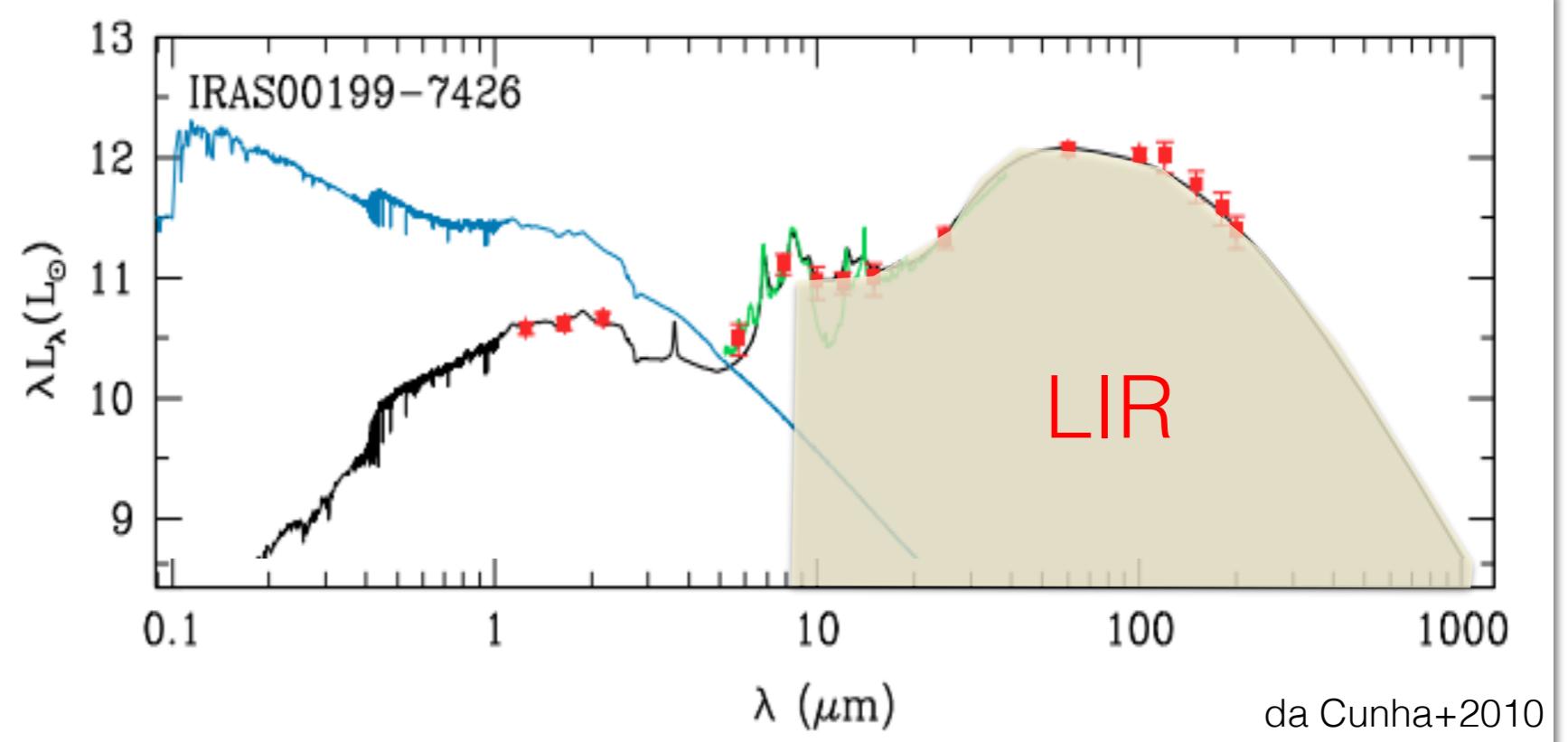
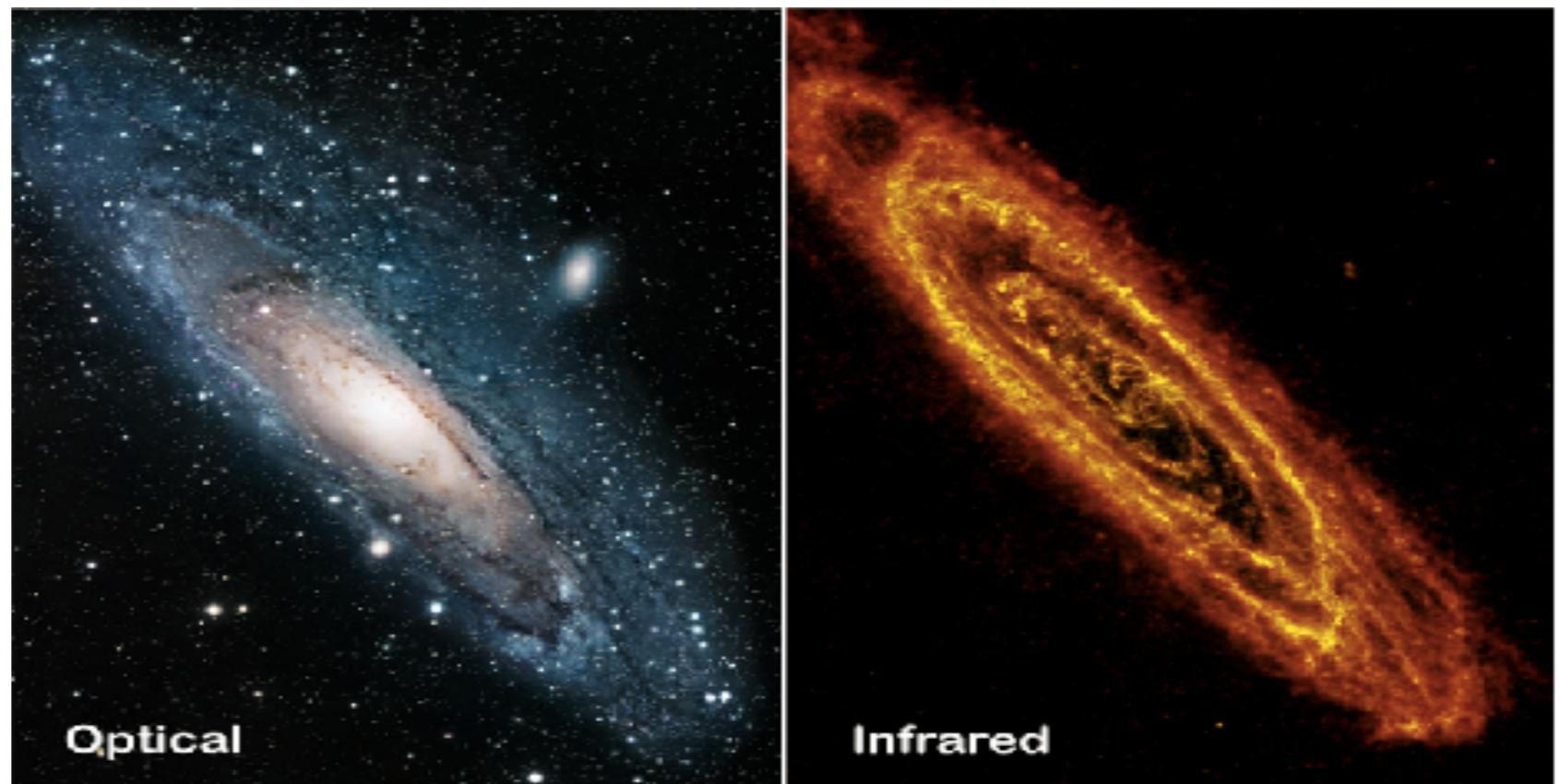
# Outline

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- Models outlined here all need some form of star-formation rate or bolometric infrared luminosity (SFR or LIR).
- Why not pull straight from measurements?
  - Short answer: *it's hard.*
  - Less-short answer: it's not that hard, and look, I've done half of the work for you.

# Challenge

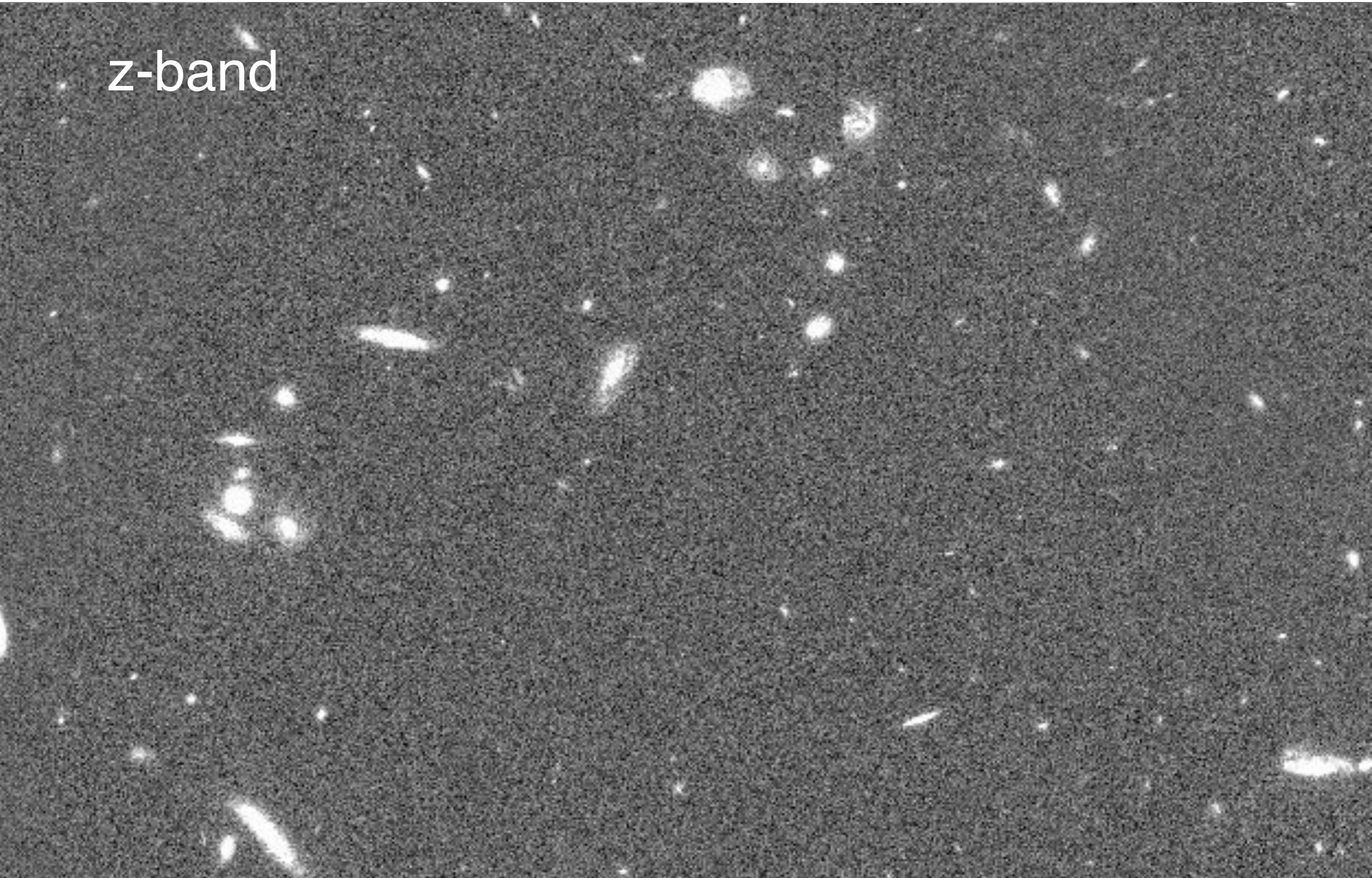
- Infrared/Submillimeter emission reprocessed starlight by dust
- IR/Submm traces star formation
- Half the emission is tied up in dust
  - We want to quantify the optical-LIR connection



# Challenge – Source Confusion

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z-band



# Solution

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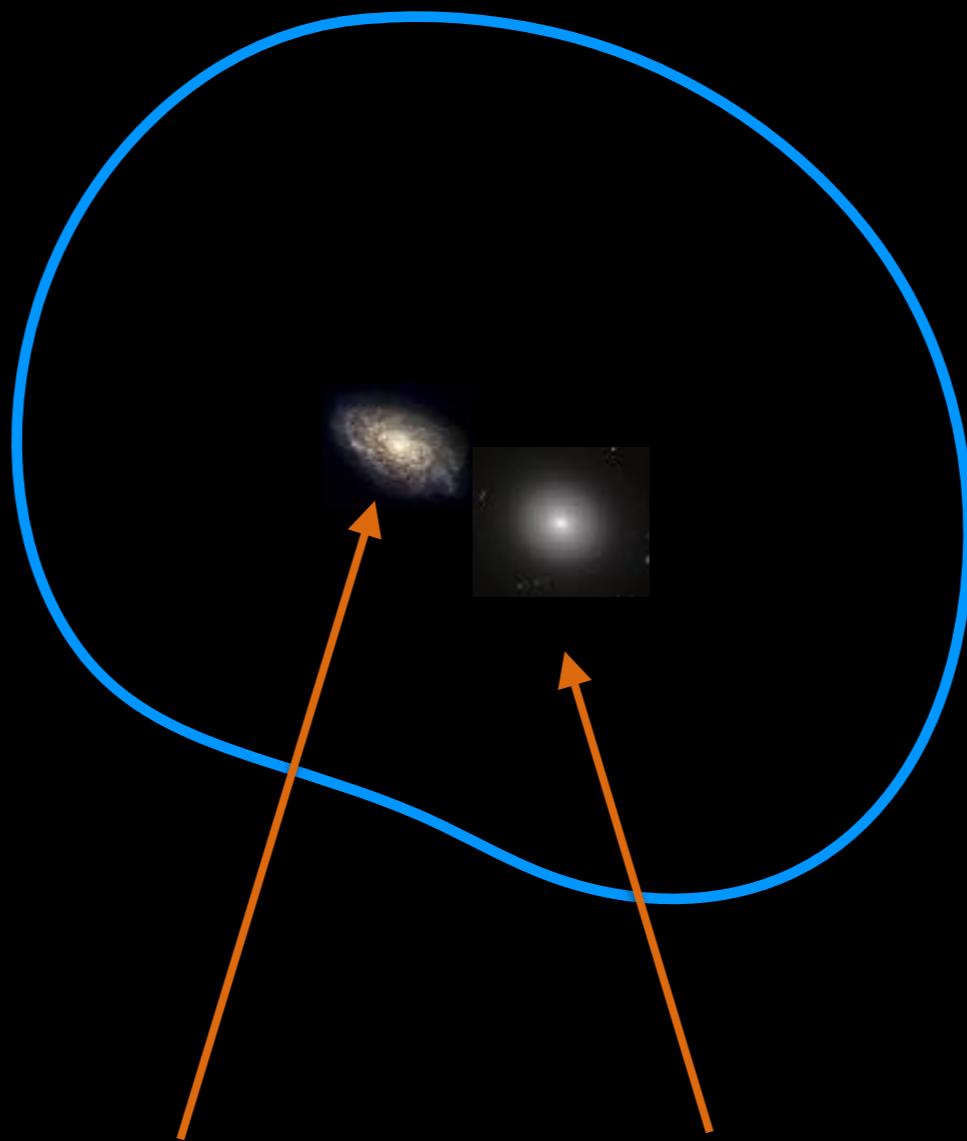
Use:

- The fact that intensity fluctuations contain signal
- *Ancillary Data*
- Creativity + Statistics

GOODS-S  
Half 1

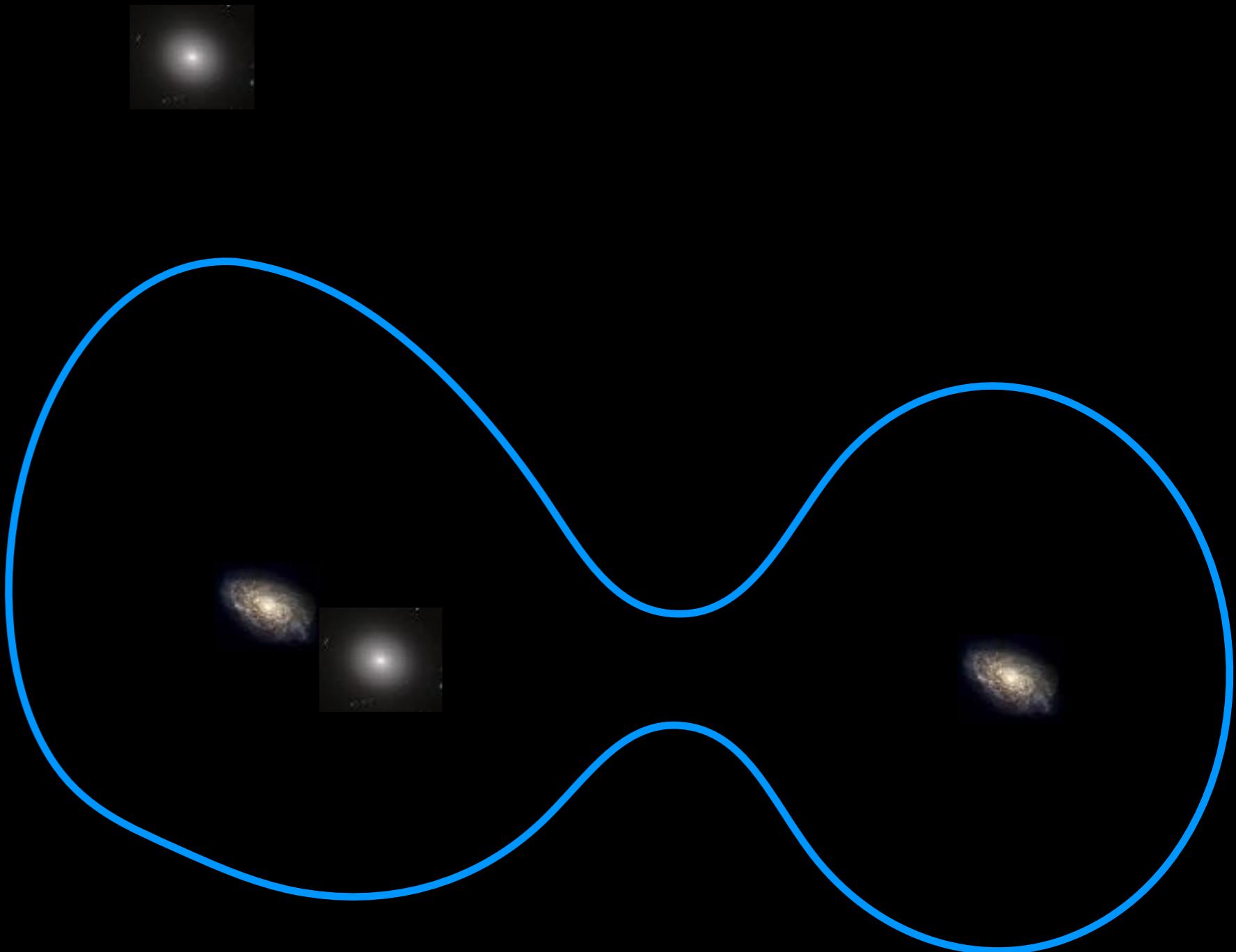
GOODS-S  
Half 2

SPIRE Contour



- What if you know these galaxies are here?

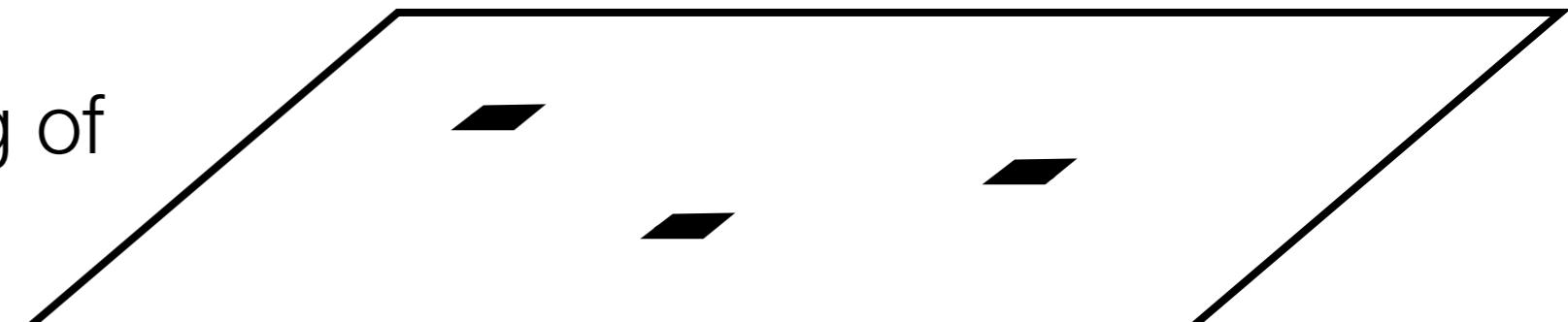
SPIRE Contour



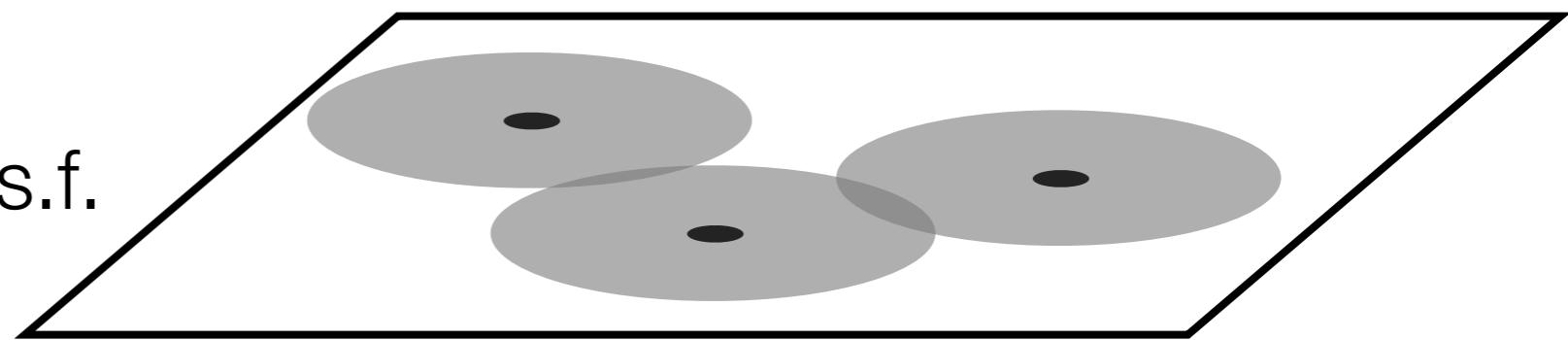
- Key is to identify and **group** galaxies with similar *physical* properties, and then rely on **statistics to fit fluctuations**

# SIMSTACK: Simultaneous Stacking Algorithm

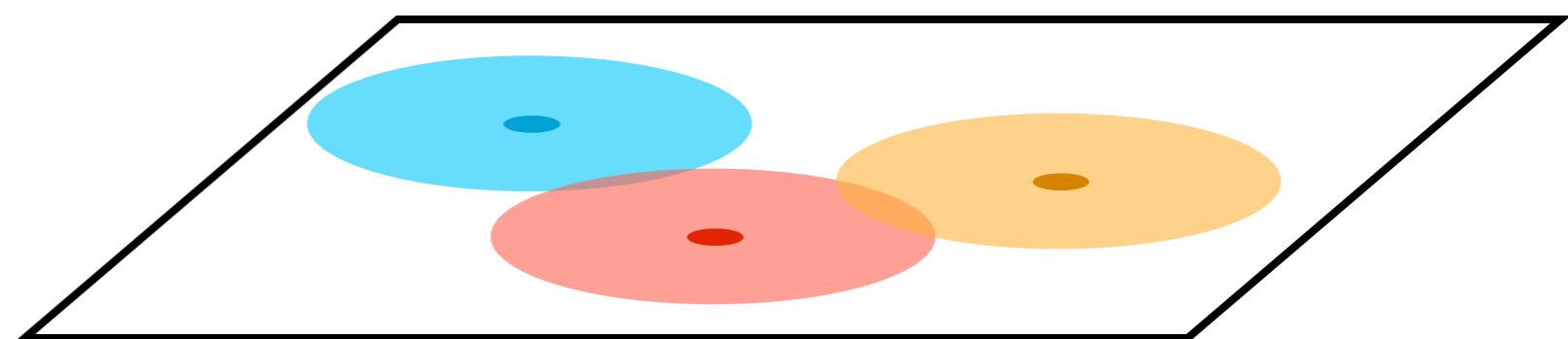
make hits map from catalog of similar objects



convolve with instrument p.s.f.



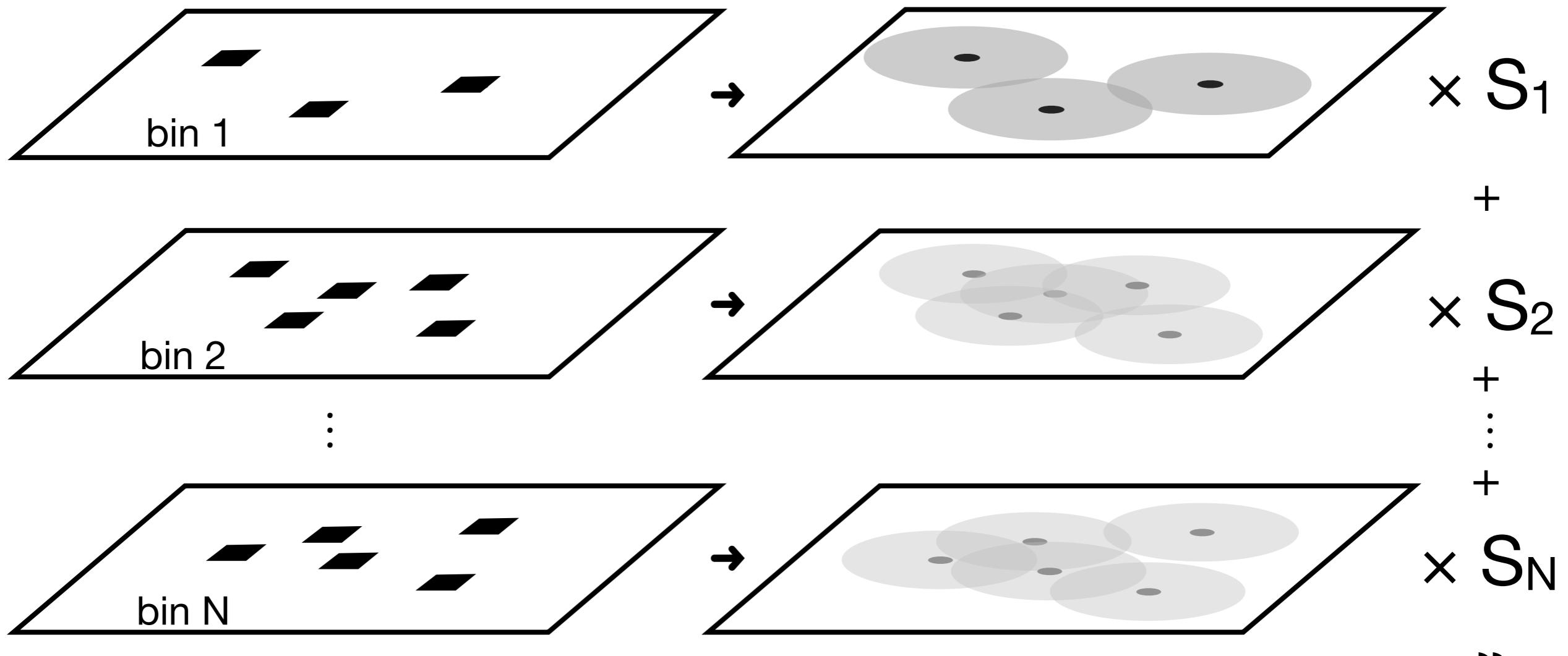
regress to find *mean* flux density, S



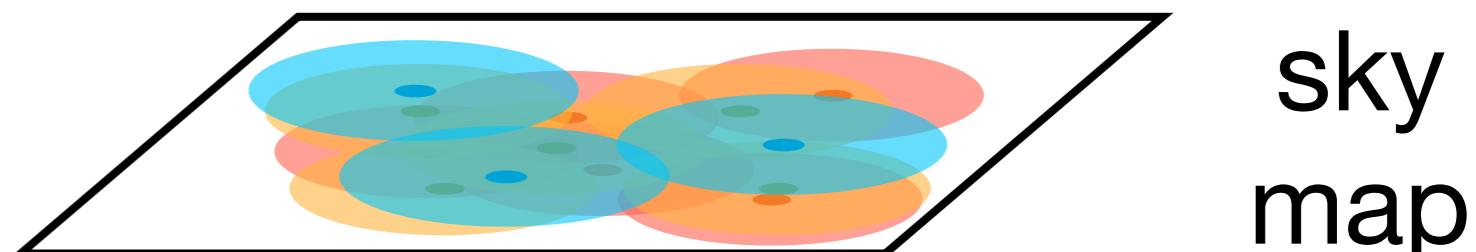
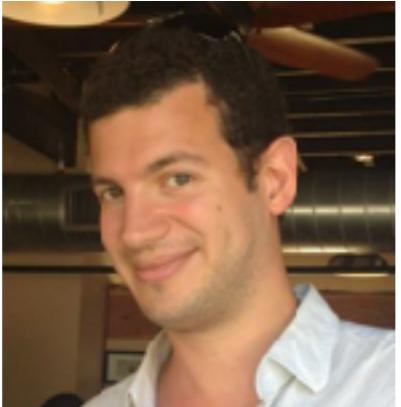
Formalism developed w/ Lorenzo Moncelsi (Caltech);  
also see Kurczynski & Gawiser (2010), Roseboom et al. (2010)

**SIMSTACK code publicly available (see arXiv:1304.0446):**  
**IDL (old) – <https://web.stanford.edu/~viero/downloads.html>**  
**Python – <https://github.com/marcoviero/simstack>**

# SIMSTACK: Simultaneous Stacking Algorithm



Formalism  
developed w/  
Lorenzo Moncelsi  
(Caltech)



**SIMSTACK code publicly available (see arXiv:1304.0446):**

**Python – <https://github.com/marcoviero/simstack>**

# SIMSTACK is simple to use (Python and IDL)

- Define type of stack, and where everything is, **in config file**.  
→ ./run\_simstack\_cmd\_line.py config\_file\_name.cfg

```
; Example parameter file for simstack code
;
; Contact: Marco Viero (marco.viero@stanford.edu)
;
[general]
;populations chooses how the catalog is split into groups with like-properties
;classifying_scheme chooses how the catalog is split into groups with like-properties
;Options are: cf-gt ; general ; uv] ;
classification_scheme = general
bootstrap      = False 0 2 ; True/False, initial number, number of iterations
;Catalog specific names for redshift, stellar mass, RA, and DEC
zkey = PHOTZ
mkey = MASS_MED
ra_key = ALPHA_J2000
dec_key = DELTA_J2000

[populations]
;Name_of_sub-population = index, [conditions]
;Here conditions are: feature, greater than, less than, equal to
;False when one of those does not apply
af = 1 CLASS False False 1
dhad = 0 CLASS False False 0

[cosmology] ; Cosmology - Planck15
omega_m = 0.3075
omega_l = 0.6913
omega_k = 0.
h     = 0.6774

[io] ; Input/output
;output_folder will contain the directories:
;- simstack_fluxes
;- bootstrapped_fluxes
;If they don't exist the code will create them!
output_folder      = PICKLESPATH simstack/stacked_flux_densities/
flux_densities_filename = simstack_flux_densities
shortname          = UVista_Laigle_v1.1_af_gt_z_pina_in_slices_test

[catalogs]
catalog_path = CATSPATH UVista/
catalog_file = COSMOS2016_Laigle+_Simplified_v1.1.csv

[binning]
optimal_binning = False ; Not yet working
bin_in_lookbackt= False ; Not yet working from command line, and requires NPpredict be installed
all_z_at_once   = False
; If binning in lookback time, redshift_nodes should be in Gyr from present day.
redshiftt_nodes = 0.01 0.5 1.0 1.5 2.0 2.5 3.0 3.5 4.0
mass_nodes      = 0.5 0.6 0.7 0.8 10.0 10.5 11.0 12.0
```

```
[maps_to_stack]
; True/False represents whether to stack them
nips_24      = 24.0 False
pacs_green   = 130.0 False
pacs_red     = 150.0 False
suirr_PSW    = 250.0 True
suirr_PMW    = 350.0 False
suirr_PLW    = 550.0 False
scuba_450    = 450.0 False
scuba_850    = 850.0 False

[map_path]
nips_24 = MAPSPATH /data/cutouts/
pacs_green = MAPSPATH /data/cutouts/
pacs_red = MAPSPATH /data/cutouts/
suirr_PSW = MAPSPATH /data/cutouts/
suirr_PMW = MAPSPATH /data/cutouts/
suirr_PLW = MAPSPATH /data/cutouts/
scuba_450 = MAPSPATH /data/cutouts/
scuba_850 = MAPSPATH /data/cutouts/

[map_file]
; Nans need to be in Jy/beam. If they are not, use second element in [beams] below to convert them.
nips_24      = mips_24_G03_sci_10.cutout.fits
pacs_green   = pco_COSMOS_pgreen_Map.DR1.sci.cutout.fits
pacs_red     = pco_COSMOS_red_Map.DR1.sci.cutout.fits
suirr_PSW    = cosmos-uvista-hipe12_itemap_10_iterations_6.0_arcsec_pixels_PSW.signal.cutout.fits
suirr_PMW    = cosmos-uvista-hipe12_itemap_10_iterations_6.0_arcsec_pixels_PMW.signal.cutout.fits
suirr_PLW    = cosmos-uvista-hipe12_itemap_10_iterations_6.0_arcsec_pixels_PLW.signal.cutout.fits
scuba_450    = mas450_new_header.cutout.fits
scuba_850    = S2CLS_COSMOS_NMF_DR1_new_header.cutout.signal.fits

[noise_file]
; If fits file contains noisevar in second extension, has same name as signal map
nips_24      = mips_24_G03_err_10.cutout.fits
pacs_green   = pco_COSMOS_pgreen_Map.DR1.err.cutout.fits
pacs_red     = pco_COSMOS_red_Map.DR1.err.cutout.fits
suirr_PSW    = cosmos-uvista_hipe12_itemap_10_iterations_6.0_arcsec_pixels_PSW.noise.cutout.fits
suirr_PMW    = cosmos-uvista-hipe12_itemap_10_iterations_6.0_arcsec_pixels_PMW.noise.cutout.fits
suirr_PLW    = cosmos-uvista-hipe12_itemap_10_iterations_6.0_arcsec_pixels_PLW.noise.cutout.fits
scuba_450    = mas450_new_header_rms.cutout.fits
scuba_850    = S2CLS_COSMOS_NMF_DR1_new_header.noise.fits

[beams]
;1- PSF file path+names, or effective FWHM
;2- Beam area in sr. Should be 1.0 if maps are in Jy/beam, otherwise actual effective area if Jy/sr
nips_24      = 6.32 1.05e-09
pacs_green   = 6.7 2.0271e-07 ; MJy/sr to Jy/beam
pacs_red     = 11.2 4.5593e-07 ; MJy/sr to Jy/beam
suirr_PSW    = 17.6 1.0
suirr_PMW    = 23.9 1.0
suirr_PLW    = 35.2 1.0
scuba_450    = 7.3 1.0
scuba_850    = 12.1 1.0
```

# SIMSTACK

- It will save stacked results in a folder you define.
- You can access the results with an iPython Notebook
  - [https://github.com/marcoviero/simstack/blob/master/notebooks/plot\\_simstack\\_output.ipynb](https://github.com/marcoviero/simstack/blob/master/notebooks/plot_simstack_output.ipynb)

```
In [1]: import pdb
import numpy as np
import pandas as pd
import os
import pylab as plt
from utils import clean_args
from utils import clean_nans
from utils import fast_sed
from utils import fast_sed_fitter
from utils import fast_lir
from utils import stagger_n
from utils import subset_averages_from_ids
from utils import main_sequence_s15
from binocatologs import Field_catalogs
from astropy.cosmology import Planck15 as cosmo
import astropy.units as u
try:
    from simstack import PickledStacksReader, measure_cib
except:
    from simstack.simstack import PickledStacksReader, measure_cib
%matplotlib inline
```

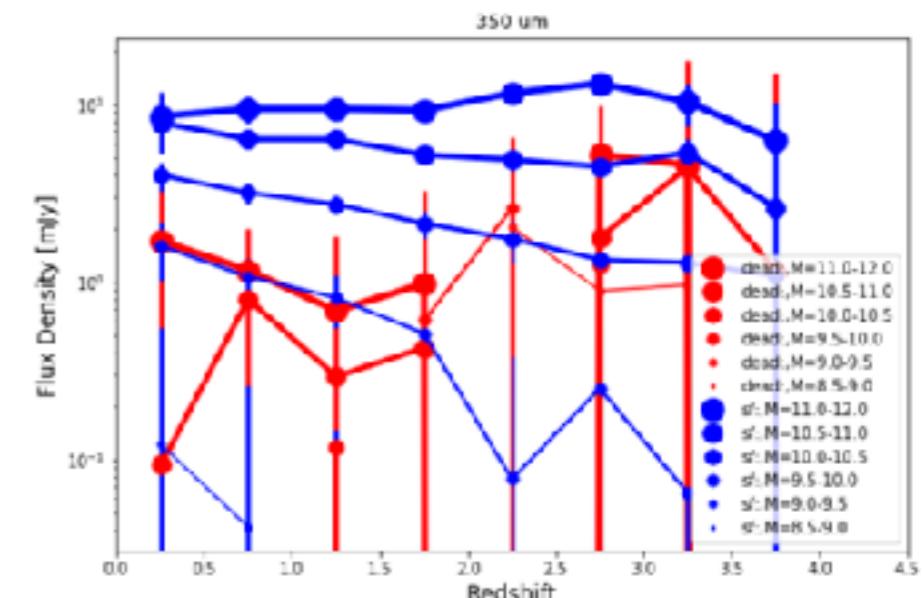
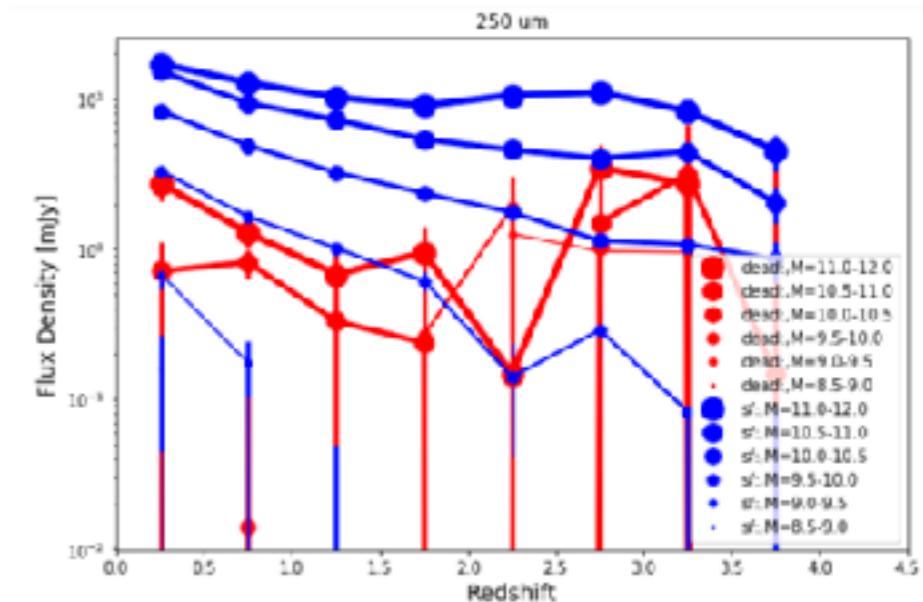
```
In [2]: conv_luv_to_sfr = 2.17e-10
conv_lir_to_sfr = 1.72e-10
L_sun = 3.839e26 # W
c = 299792458.0 # m/s
a_mu_flux_to_mass = 6.7e19
h = 6.62607004e-34 #J s kg / s #4.13e-15 J eV/s
k = 1.38064052e-23 #J2 kg s-2 K-1 0.517e-5 #eV/K
```

```
In [3]: popcolor=['red','blue','green','orange','black','grey','chocolate','darkviolet','pink','magenta','dodgerblue','lavender']
```

```
In [4]: path_pickles = os.environ['PICKLESPATH']
path_maps    = os.environ['MAPSPATH']
path_catalogs= os.environ['CATSPATH']
```

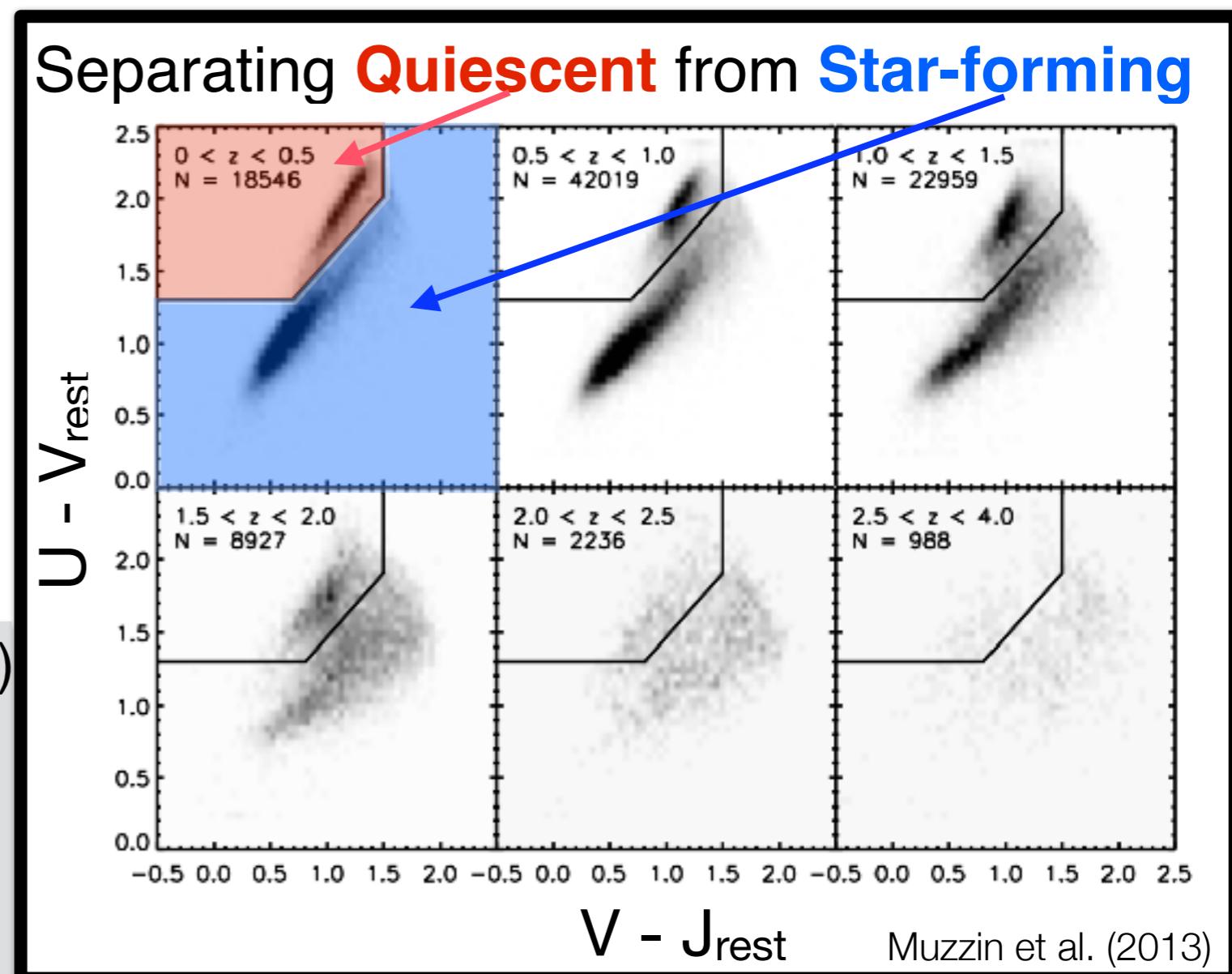
```
In [5]: #Location of the stacked parameter file
shortname = 'uvista_Lsquia_v1.1_sf-qt_z_bins_in_slices_test'
path_config = path_pickles + '/simstack/stacked_flux_densities/simstack_fluxes/' + shortname + '/'
file_config = 'example.cfg'
if os.path.exists(path_config+file_config) == True:
    print path_config+file_config
```

```
/data/pickles//simstack/stacked_flux_densities/simstack_fluxes/uvista_Lsquia_v1.1_sf-qt_z_bins_in_slices_test/example.cfg
```



## Catalogs

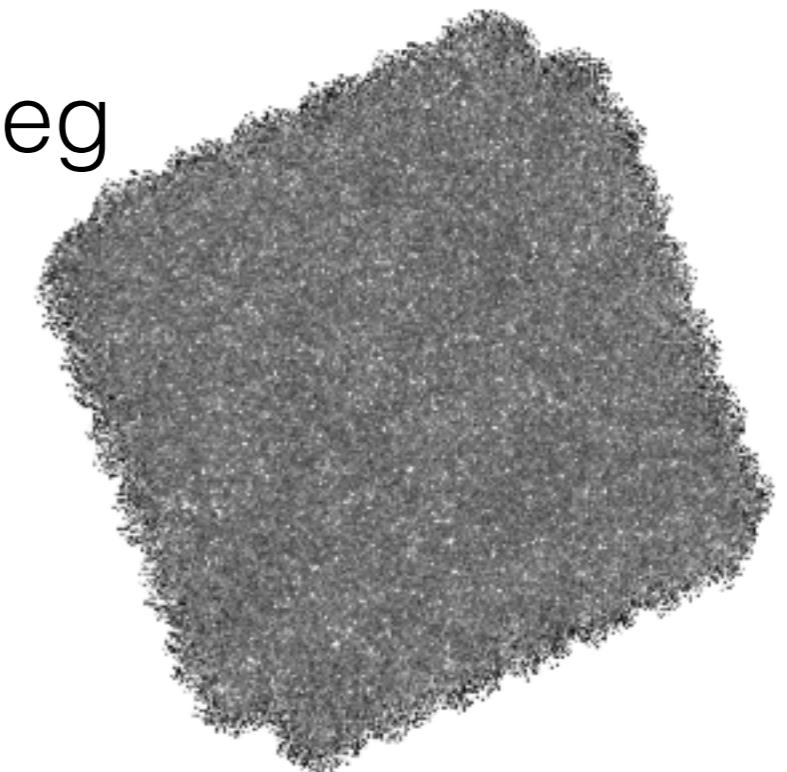
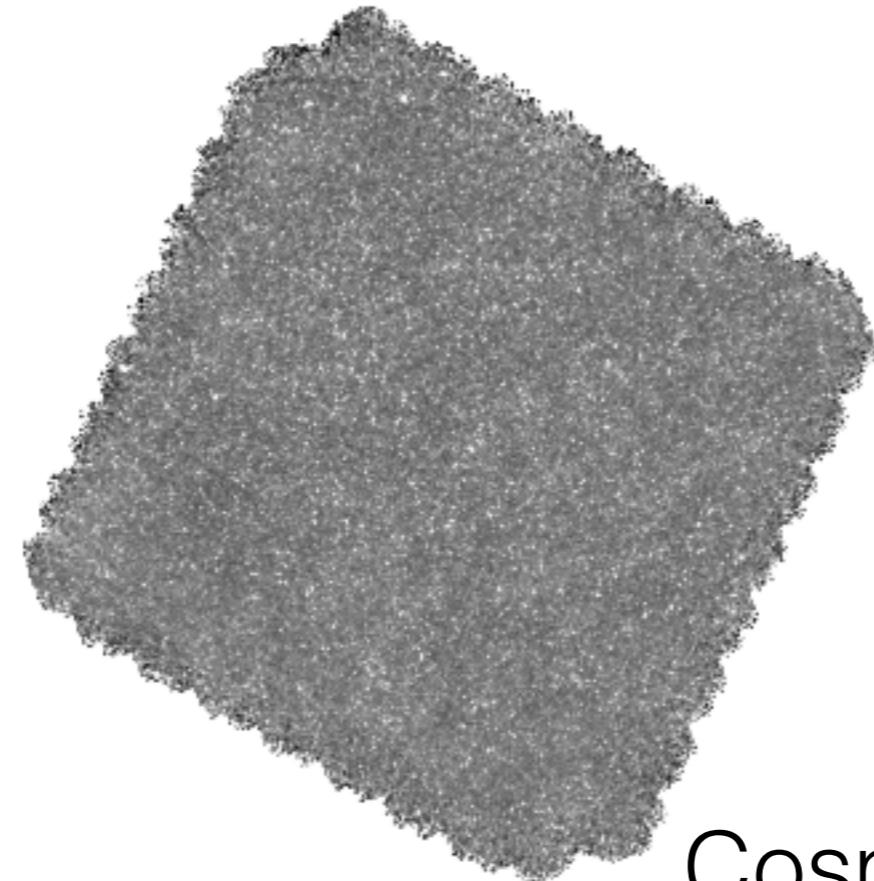
- UKIDSS/UDS [2/3 deg<sup>2</sup>] / COSMOS [1.6 deg<sup>2</sup>]
  - uBVRizJHK + IRAC ch1234
  - K-band cut 23.4 / 24 AB
  - 80,000 / 120,000 sources
- **Redshifts** - EAZY (Brammer 2008)
- **Masses** - FAST (Kriek 2009)
- **Colors** - UVJ (Williams 2009)



## Maps

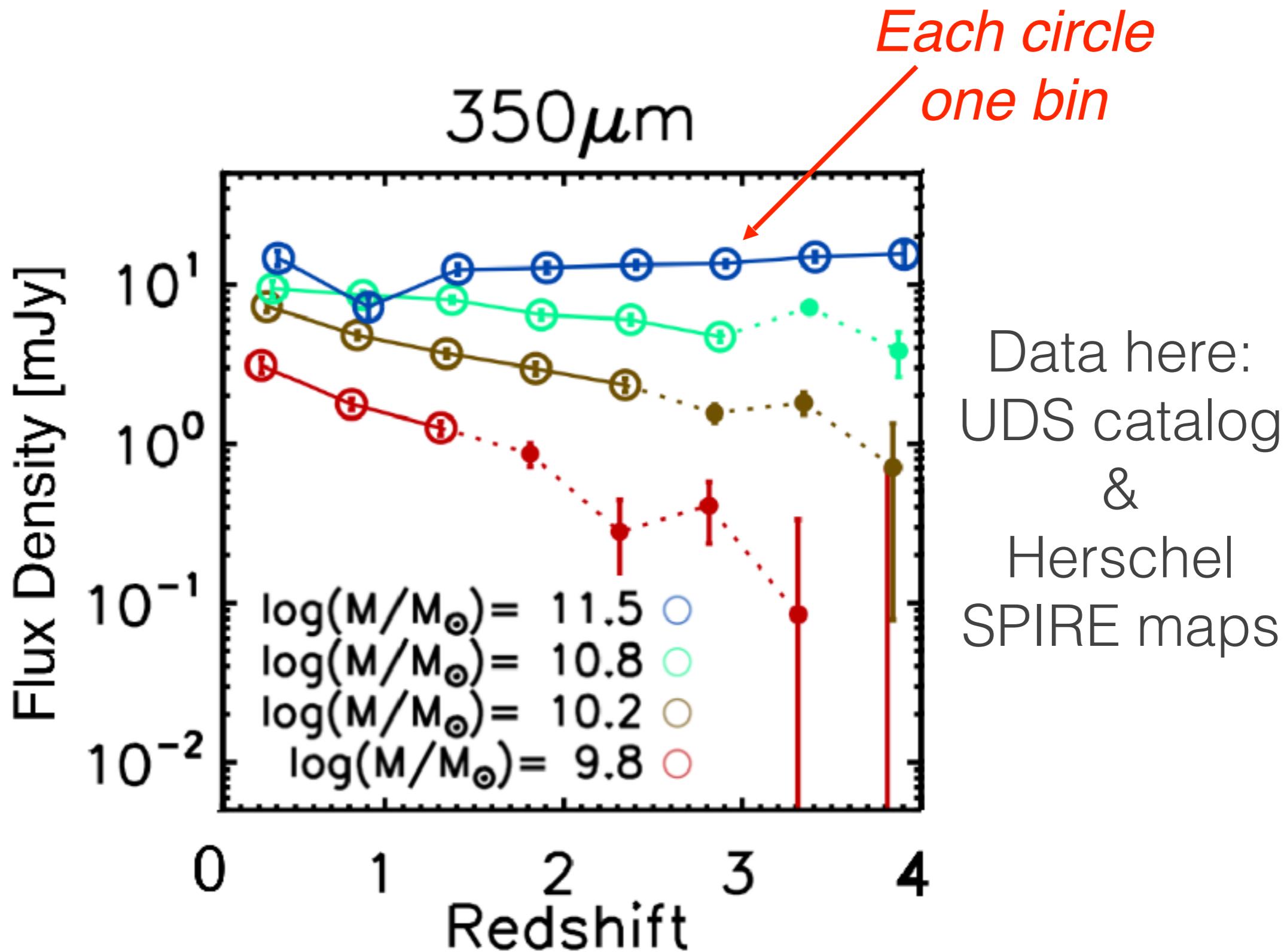
- *Spitzer/MIPS*
  - 24, 70 $\mu$ m
- *Herschel/PACS*
  - 100, 160 $\mu$ m
- *Herschel/SPIRE*
  - 250, 350, 500 $\mu$ m
- ASTE/AzTEC
  - 1100 $\mu$ m

UDS - 1.4 x 1.4 deg

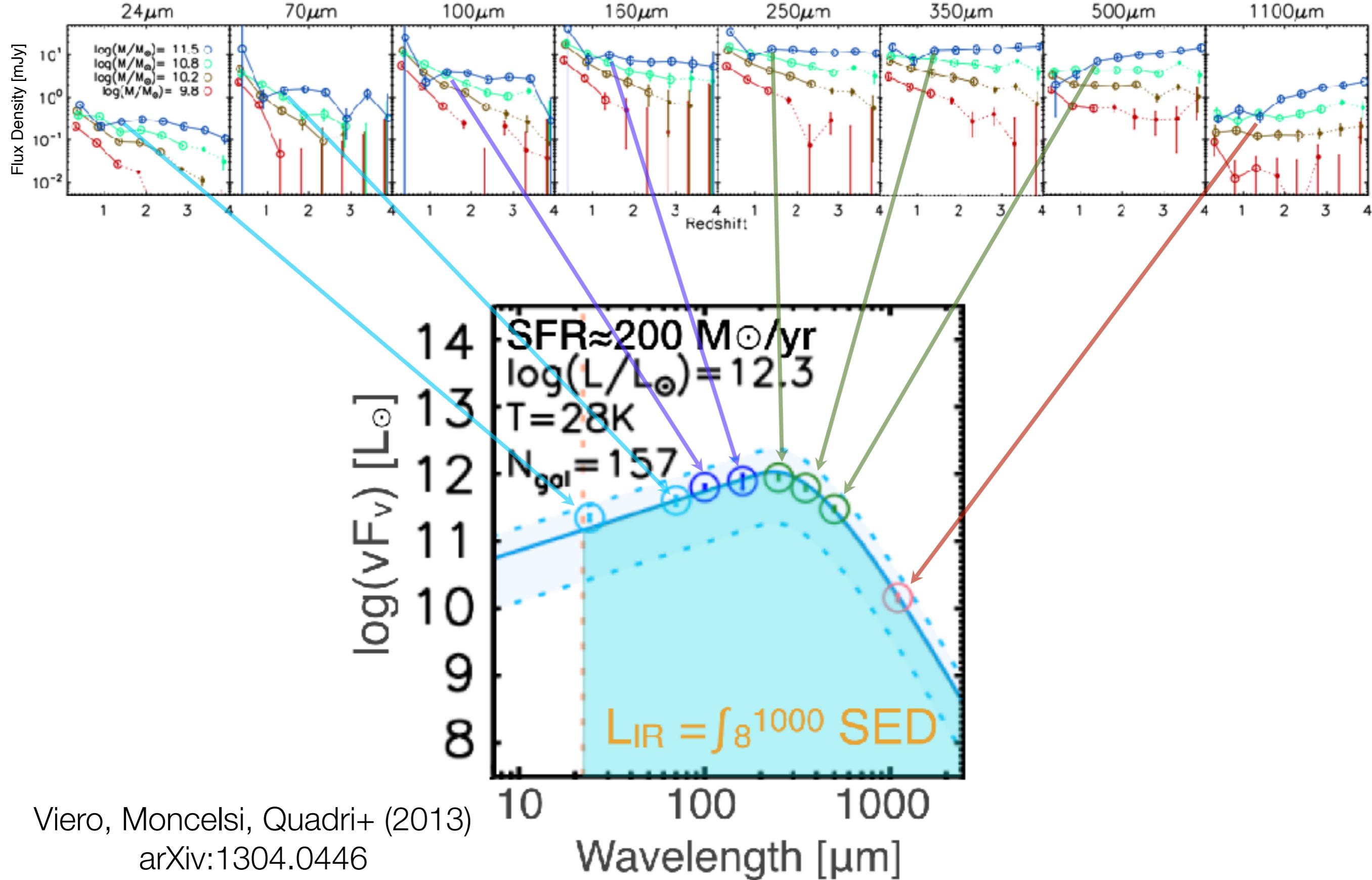


Cosmos - 1.8 x 1.8 deg

# SIMSTACK: Flux Densities (M,z)



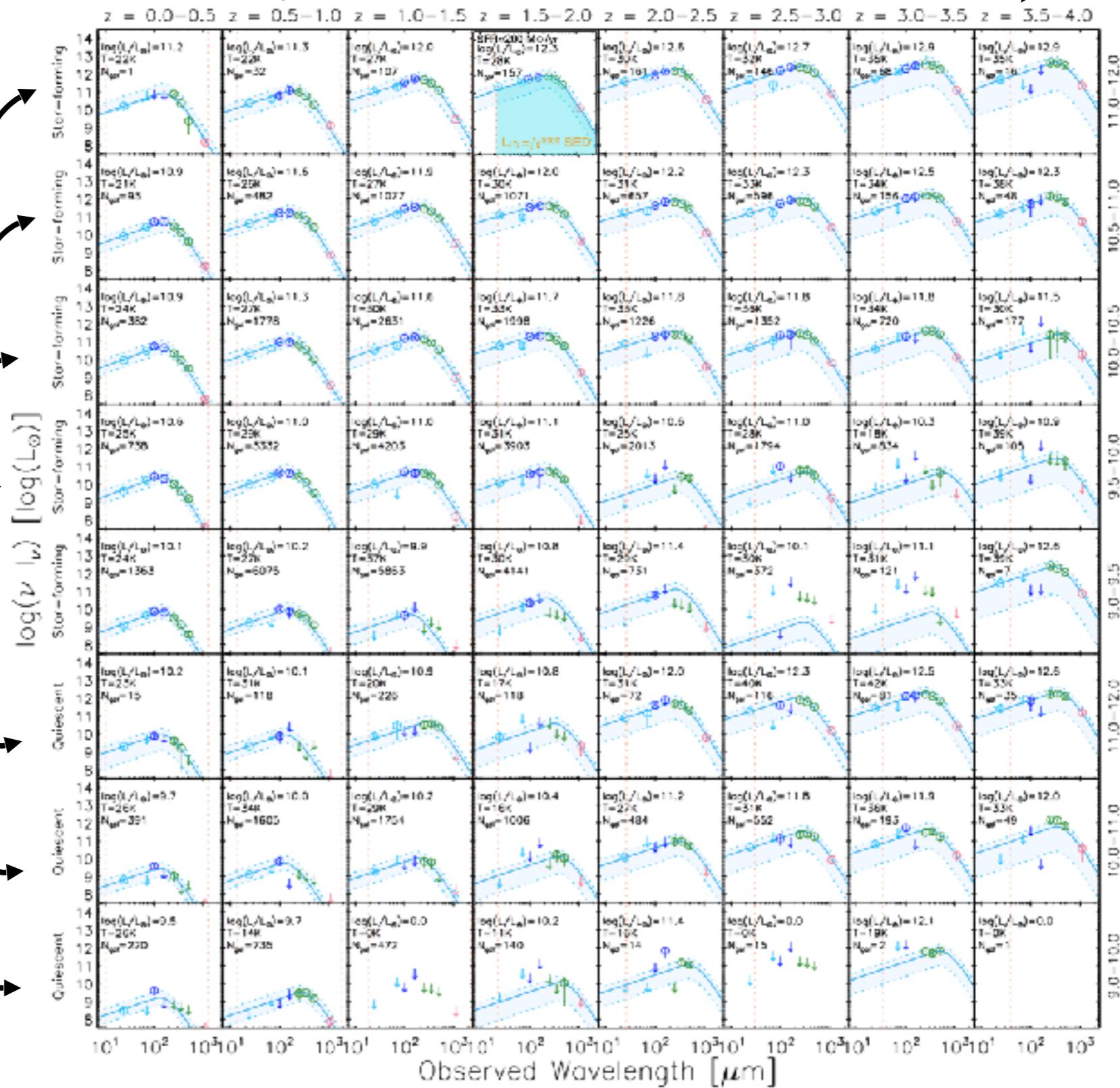
# SIMSTACK: SEDs



# SIMSTACK: SEDs

stellar  
mass  
slices

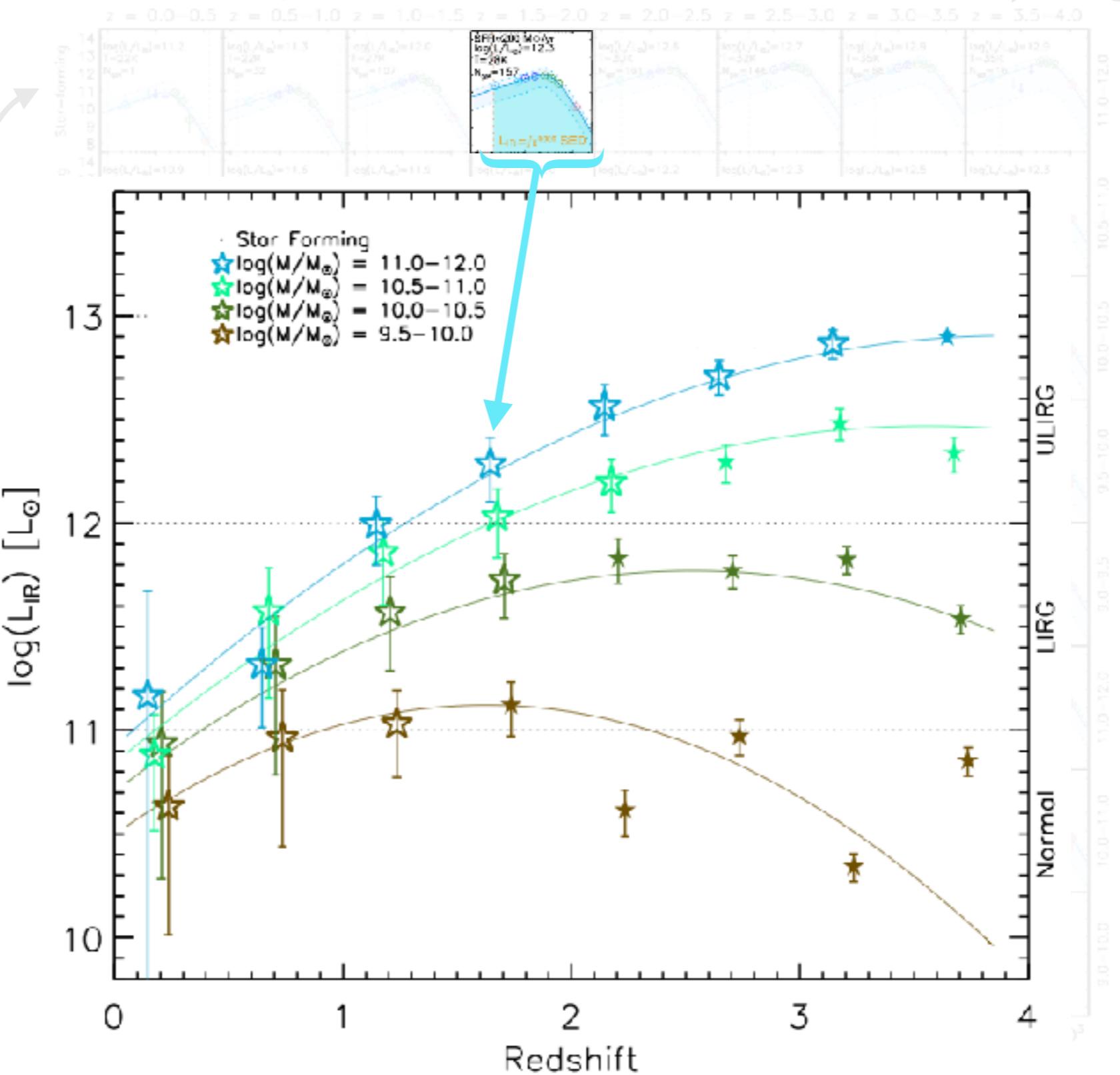
redshift  
slices



# SIMSTACK: $L_{\text{IR}}(M, z)$

stellar  
mass  
slices

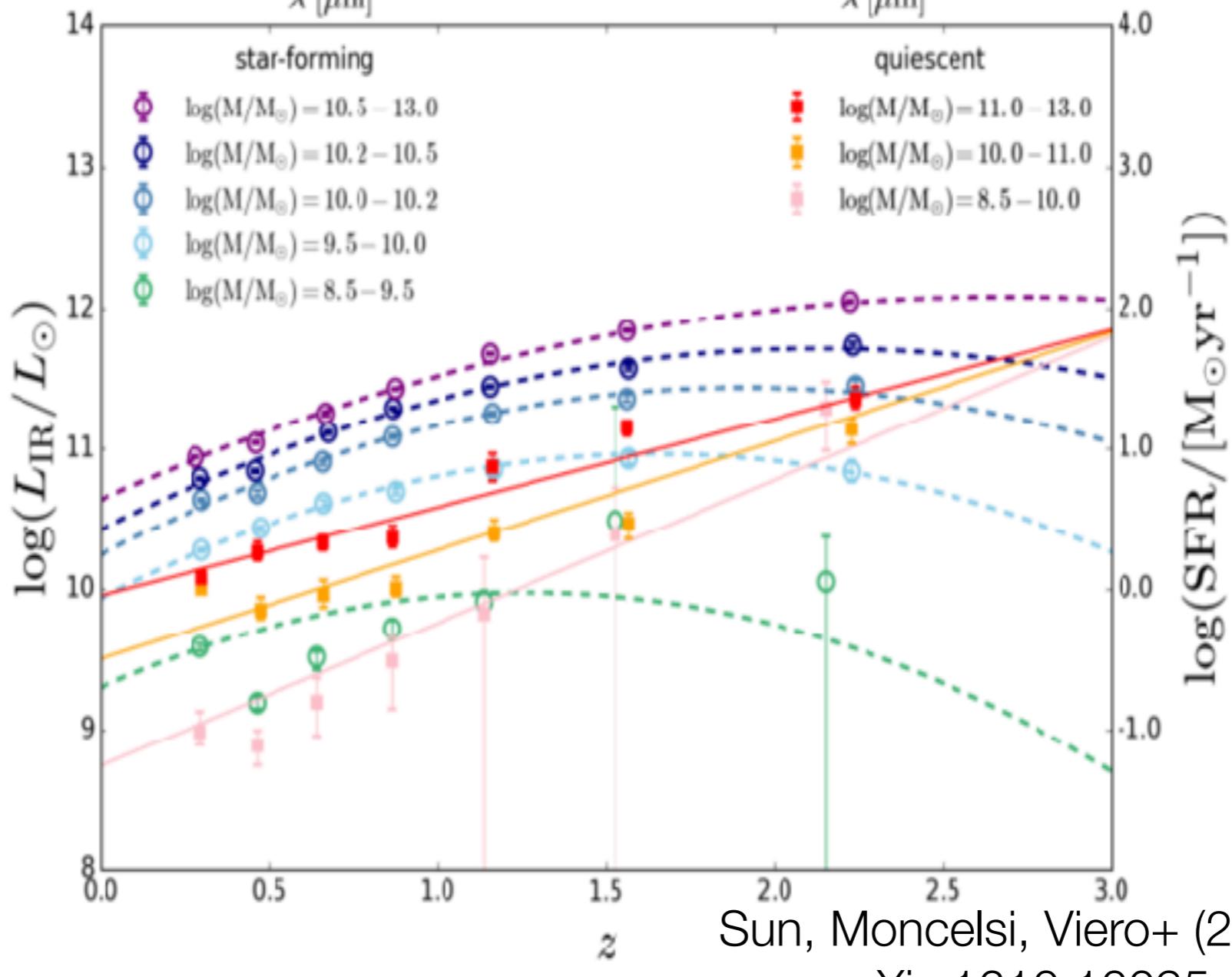
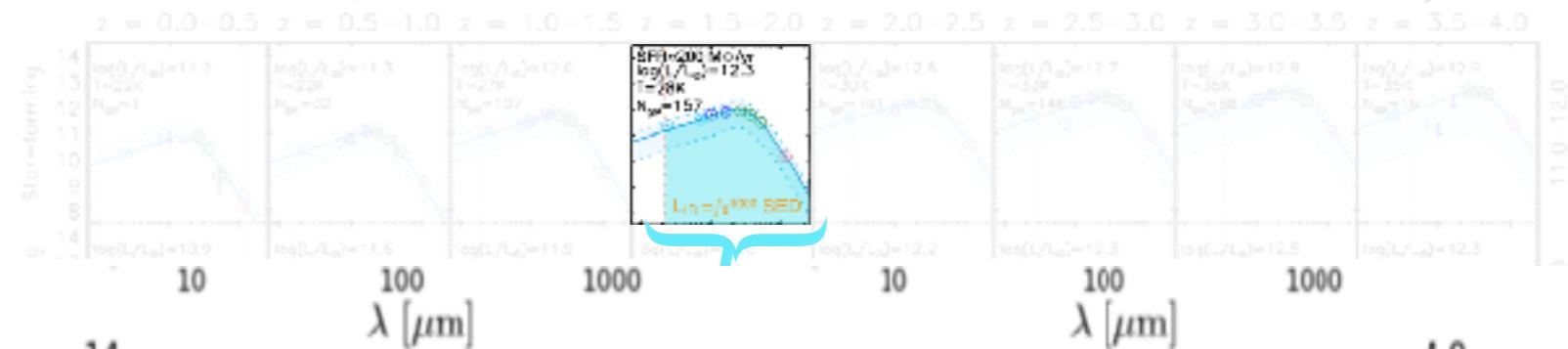
redshift  
slices



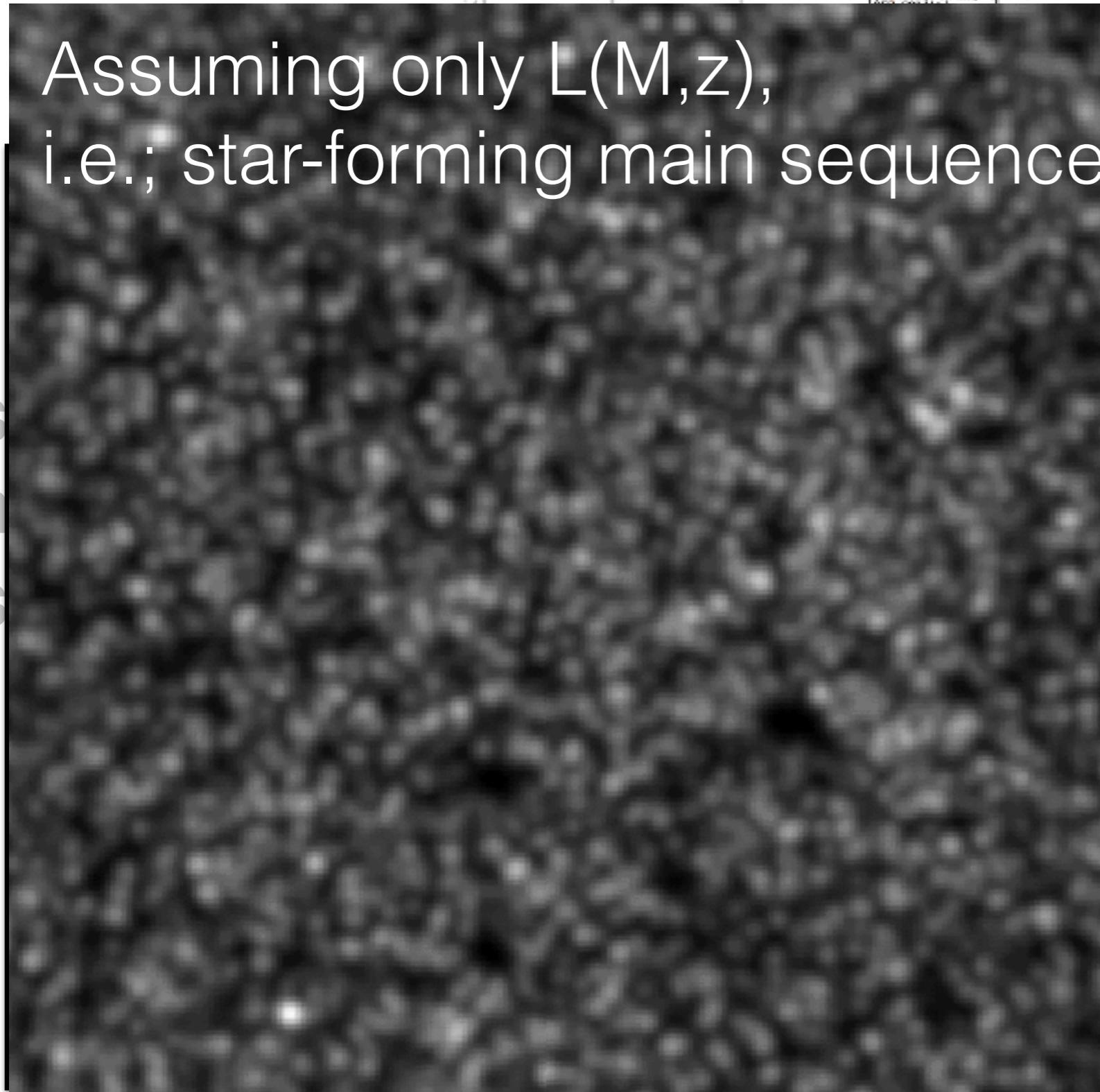
# SIMSTACK: $L_{\text{IR}}(M, z)$

stellar  
mass  
slices

redshift  
slices



# SIMSTACK: $L_{\text{IR}}(M, z, \dots)$



Assuming only  $L(M, z)$ ,  
i.e.; star-forming main sequence

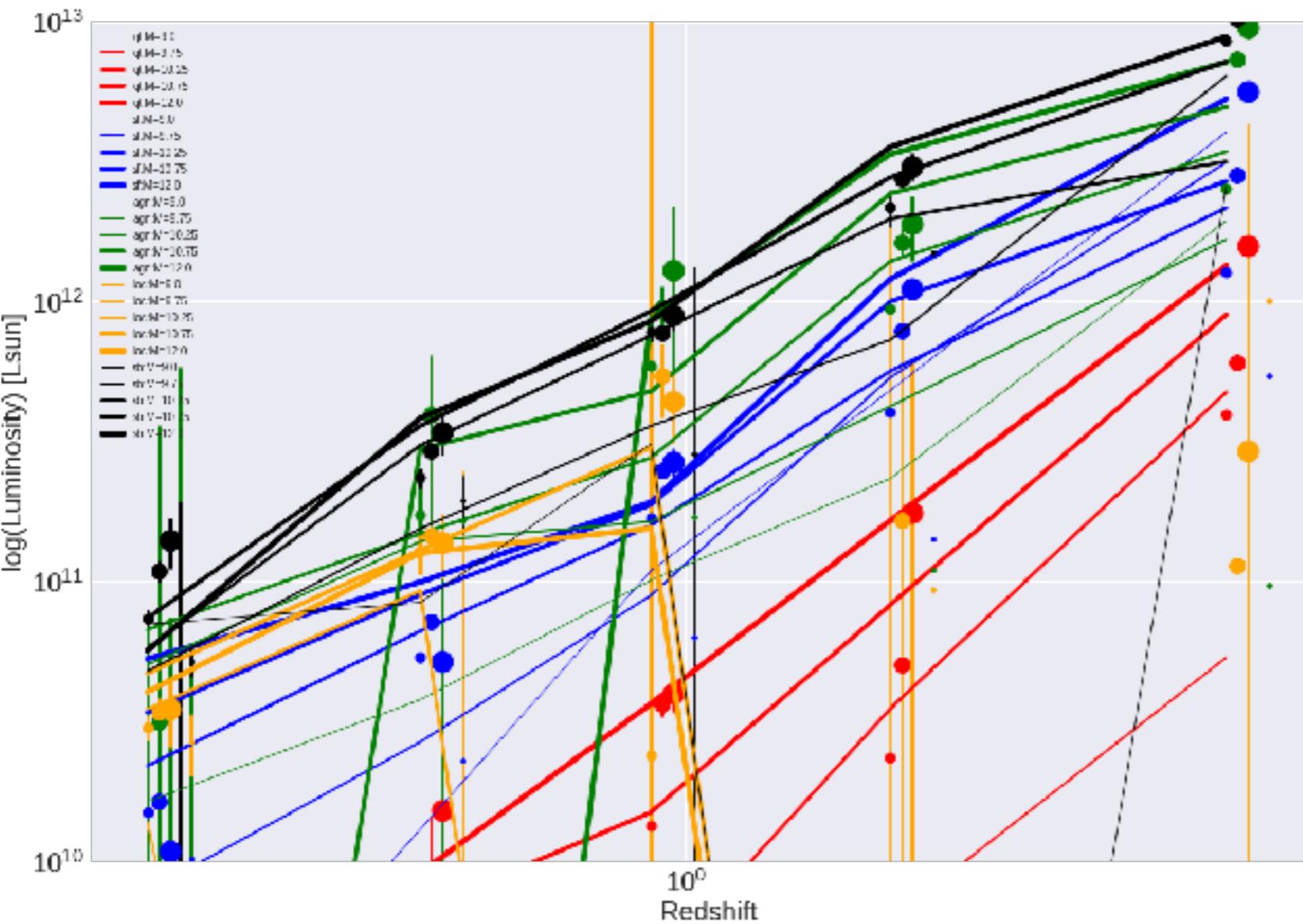
- Deep ancillary data can be fit with SED models, providing:

- Stellar Mass
- Redshift
- Extinction/UV slope
- AGN fraction
- Age/Tau...

Each bin therefore has  $\langle M \rangle, \langle z \rangle, \langle Av \rangle, \langle F_{agn} \rangle$ , etc., which can be fit with function of form:

$$\bullet L_{\text{IR}} = P(z)^\alpha P(M)^\beta P(Av)^\gamma \dots$$

# SIMSTACK: $L_{\text{IR}}(M, z, \text{Av}, \text{Fagn})$

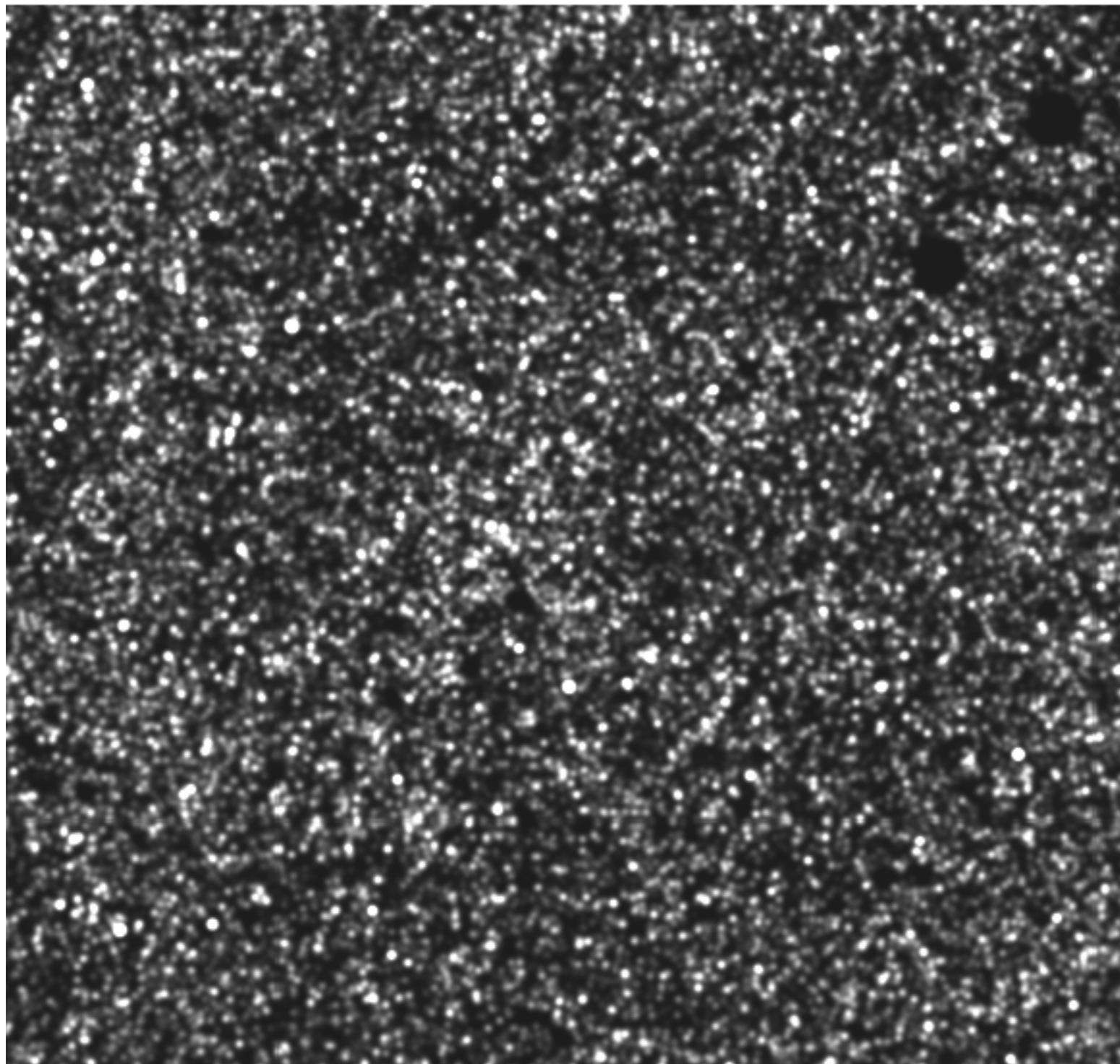


Fit can be improved by splitting the sample into finer subsamples, isolating e.g.;

- Star-forming/Quiescent
- AGN
- Starbursts

# SIMSTACK: $L_{\text{IR}}(M, z, Av, F_{agn})$

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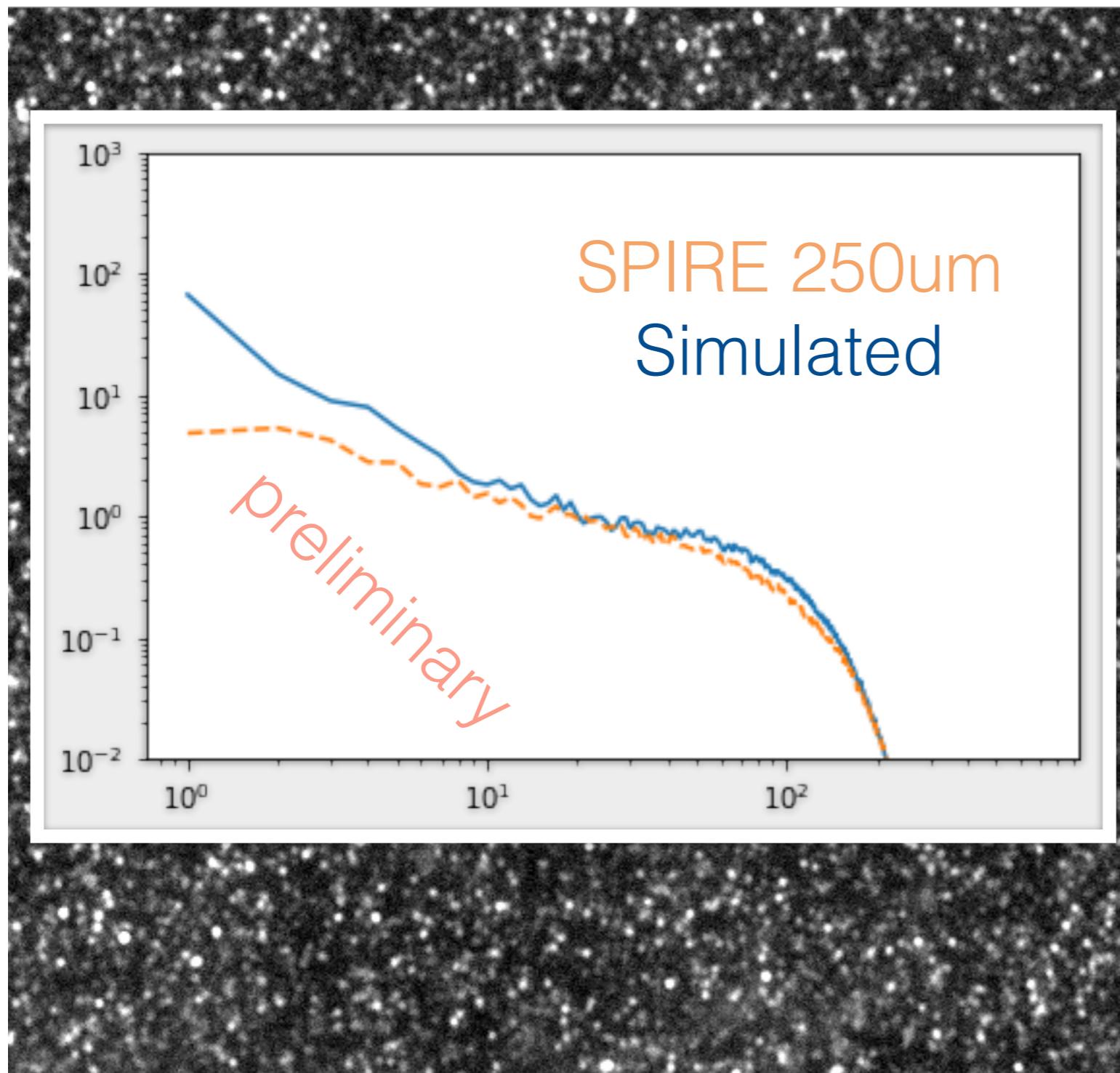
Fit can be improved by splitting the sample into finer subsamples, isolating e.g.;

- Star-forming/Quiescent
- AGN
- Starbursts

We find features most influential are, for 4 subsamples:

- $\log(L_{\text{IR}}) = C + \alpha(z) \times \log(1+z) + \beta(z) \times \log(M) + \gamma(z) \times \log(Av) + \delta(z) \times \log(F_{agn})$

# SIMSTACK: $L_{\text{IR}}(M, z, Av, F_{agn})$



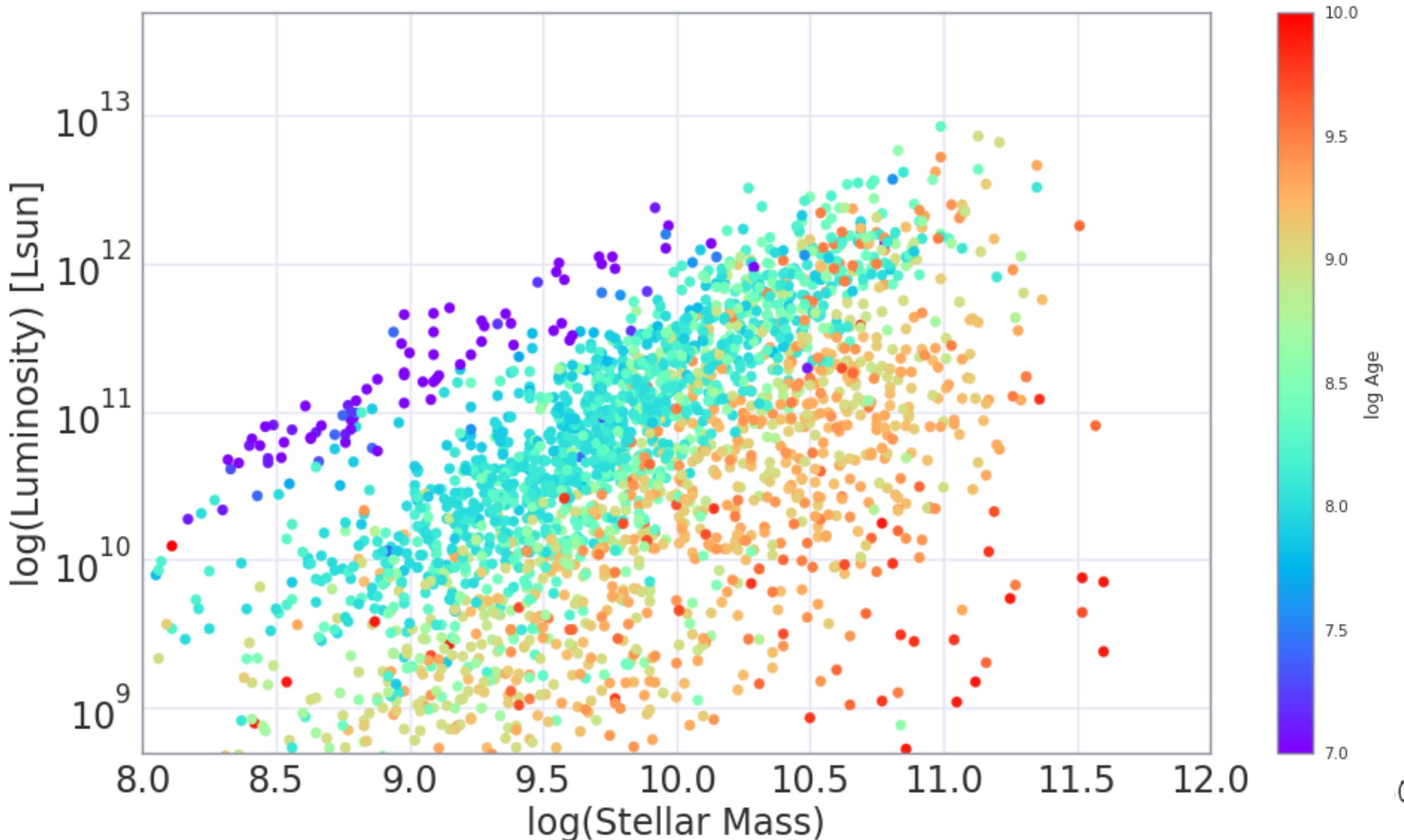
Fit can be improved by splitting the sample into finer subsamples, isolating e.g.;

- Star-forming/Quiescent
- AGN
- Starbursts

We find features most influential are, for 4 subsamples:

$$\begin{aligned} \log(L_{\text{IR}}) = & C + \\ & \alpha(z) \times \log(1+z) + \\ & \beta(z) \times \log(M) + \\ & \gamma(z) \times \log(Av) + \\ & \delta(z) \times \log(F_{agn}) \end{aligned}$$

# SIMSTACK: $L_{\text{IR}}(M, z, A_v, \text{Fagn})$



# Applications

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- Signal
  - Connect to Halo properties (including assembly bias) to:
    - ▶ estimate CO levels,
    - ▶ construct covariances,
    - ▶ test different estimators (i.e., beyond power spectrum),
    - ▶ Details being discussed during this meeting!
  - Extend to other lines that correlate with thermal dust SED
    - ▶ CII, OII, OIII, NII
    - ▶ r.f. 850um as tracer of ISM Mass.
- Foregrounds
  - Predict CO contamination in CII data cubes (e.g, Sun and the TIME collaboration, 2017)

# Summary

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- CIB continuum intensities are key to empirically connecting optical features of typical galaxies to their FIR/submm components
- Applications for this model include:
  - Forecasting CO power for:
    - ▶ Survey design
    - ▶ Covariance construction
    - ▶ Testing Estimators
    - ▶ Measurement Interpretation
- SIMSTACK is easy to use, and available at:
  - <https://github.com/marcoviero/simstack>