

MODELLING SFR-M* AND M*-Z RELATIONS TO BE OBSERVED AT HIGH-REDSHIFT WITH JWST

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MOTIVATION



Image credit: Harry Ferguson <u>http://candels-collaboration.blogspot.fr/2013/02/star-formation-in-mountains.html</u>

MOTIVATION



Stark+ 2013

MOTIVATION



Pacifici+ 2014

PREVIOUS WORK



Salmon+ 2015

PREVIOUS WORK



Kurczynski+ 2016

PREVIOUS WORK







(d) $2.0 < z \le 2.5$

Kurczynski+ 2016

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QUESTION

How well can recover the intrinsic scatter from simulated, self-consistent star formation and chemical enrichment histories?



CF00 Dust attenuation

CF00 Dust attenuation



Charlot & Fall 2000



See Brett Salmon's poster



Charlot & Fall 2000



BEAGLE mock catalogue production BEAGLE mock catalogue production

 Includes self-consistent nebular emission based on wide grid of CLOUDY models - see Julia Gutkin's poster!



BEAGLE SED fitting

BEAGLE SED fitting

At this stage we get to test impact of different fitting parameters on derivation of individual object physical parameters, as well as the M*-SFR relation as a whole.

Fitting parameters include:

- Sampling of SED (different filter sets)
- Depth of the data
- Dust prescriptions
- SFH prescriptions

MEASURING THE M*-SFR RELATION

NOTATION

For galaxy *i*:

- SFR_i is $\log_{10}(SFR/M_{\odot}yr^{-1})$
- M_i is $log_{10}(M_{\star}/M_{\odot})$
- Θ_i are the remaining parameters (e.g., metallicity, z)
- \mathbf{y}_i are "observed" photometry, with known errors $\boldsymbol{\Sigma}_i$
- $\mu_i = \mathbf{G}(SFR_i, M_i, \Theta_i)$ is predicted photometry from forward-model \mathbf{G}

MEASURING THE M*-SFR RELATION

MULTILEVEL MODEL

First Level: $\mathbf{y}_i \mid \mathrm{SFR}_i, \mathrm{M}_i, \boldsymbol{\Theta}_i \sim N(\mu_i, \boldsymbol{\Sigma}_i)$

Second Level: $SFR_i | M_i \sim \beta_0 + \beta_1 M_i + N(0, \sigma)$ $M_i \sim Uniform(a, b)$

$$P(\mathbf{SFR}, \mathbf{M}, \mathbf{\Theta}, \sigma, \beta_0, \beta_1 | \mathbf{Y}, \mathbf{\Sigma})$$

$$\propto P(\sigma) P(\beta_0, \beta_1) \prod_{i=1}^{N} P(\mathbf{y}_i | \mathrm{SFR}_i, \mathrm{M}_i, \mathbf{\Theta}_i, \mathbf{\Sigma}_i)$$

$$\times \prod_{i=1}^{N} P(\mathrm{SFR}_i | \mathrm{M}_i, \beta_0, \beta_1, \sigma) P(\mathrm{M}_i) P(\mathbf{\Theta}_i),$$

where $\mathbf{SFR} = (SFR_1, \dots, SFR_N), \mathbf{M} = (\mathbf{M}_1, \dots, \mathbf{M}_N),$ $\mathbf{\Theta} = (\mathbf{\Theta}_1, \dots, \mathbf{\Theta}_N), \mathbf{Y} = (\mathbf{y}_1, \dots, \mathbf{y}_N),$ $\mathbf{\Sigma} = (\mathbf{\Sigma}_1, \dots, \mathbf{\Sigma}_N),$ and N is the number of galaxies.

 $P(\mathbf{SFR}, \mathbf{M}, \mathbf{\Theta}, \sigma, \beta_0, \beta_1 | \mathbf{Y}, \mathbf{\Sigma})$ $\propto P(\sigma)P(\beta_0, \beta_1) \prod_{i=1}^{N} P(\mathbf{y}_i | \mathbf{SFR}_i, \mathbf{M}_i, \mathbf{\Theta}_i, \mathbf{\Sigma}_i)$ $\times \prod_{i=1}^{N} P(\mathbf{SFR}_i | \mathbf{M}_i, \beta_0, \beta_1, \sigma)P(\mathbf{M}_i)P(\mathbf{\Theta}_i),$

where $\mathbf{SFR} = (\mathrm{SFR}_1, \dots, \mathrm{SFR}_N), \mathbf{M} = (\mathrm{M}_1, \dots, \mathrm{M}_N),$ $\mathbf{\Theta} = (\mathbf{\Theta}_1, \dots, \mathbf{\Theta}_N), \mathbf{Y} = (\mathbf{y}_1, \dots, \mathbf{y}_N),$ $\mathbf{\Sigma} = (\mathbf{\Sigma}_1, \dots, \mathbf{\Sigma}_N),$ and N is the number of galaxies.

 $P(\mathbf{SFR}, \mathbf{M}, \mathbf{\Theta}, \sigma, \beta_0, \beta_1 | \mathbf{Y}, \mathbf{\Sigma})$ $\propto P(\sigma)P(\beta_0, \beta_1) \prod_{i=1}^{N} P(\mathbf{y}_i | \mathbf{SFR}_i, \mathbf{M}_i, \mathbf{\Theta}_i, \mathbf{\Sigma}_i)$ Prior Distribution $\times \prod_{i=1}^{N} P(\mathbf{SFR}_i | \mathbf{M}_i, \beta_0, \beta_1, \sigma)P(\mathbf{M}_i)P(\mathbf{\Theta}_i),$

where $\mathbf{SFR} = (SFR_1, \dots, SFR_N), \mathbf{M} = (\mathbf{M}_1, \dots, \mathbf{M}_N),$ $\mathbf{\Theta} = (\mathbf{\Theta}_1, \dots, \mathbf{\Theta}_N), \mathbf{Y} = (\mathbf{y}_1, \dots, \mathbf{y}_N),$ $\mathbf{\Sigma} = (\mathbf{\Sigma}_1, \dots, \mathbf{\Sigma}_N),$ and N is the number of galaxies.

 $P(\mathbf{SFR}, \mathbf{M}, \mathbf{\Theta}, \sigma, \beta_0, \beta_1 | \mathbf{Y}, \mathbf{\Sigma})$ Hyperprior $\propto P(\sigma)P(\beta_0, \beta_1) \prod_{i=1}^{N} P(\mathbf{y}_i | \mathbf{SFR}_i, \mathbf{M}_i, \mathbf{\Theta}_i, \mathbf{\Sigma}_i)$ Prior Distribution $\times \prod_{i=1}^{N} P(\mathbf{SFR}_i | \mathbf{M}_i, \beta_0, \beta_1, \sigma)P(\mathbf{M}_i)P(\mathbf{\Theta}_i),$

where $\mathbf{SFR} = (SFR_1, \dots, SFR_N), \mathbf{M} = (\mathbf{M}_1, \dots, \mathbf{M}_N),$ $\mathbf{\Theta} = (\mathbf{\Theta}_1, \dots, \mathbf{\Theta}_N), \mathbf{Y} = (\mathbf{y}_1, \dots, \mathbf{y}_N),$ $\mathbf{\Sigma} = (\mathbf{\Sigma}_1, \dots, \mathbf{\Sigma}_N),$ and N is the number of galaxies.

MEASURING THE M*-SFR RELATION



MEASURING THE M*-SFR RELATION



$$P(\mathbf{SFR}, \mathbf{M}, \mathbf{\Theta}, \sigma, \beta_0, \beta_1 | \mathbf{Y}, \mathbf{\Sigma})$$

$$\propto P(\sigma) P(\beta_0, \beta_1) \prod_{i=1}^{N} P(\mathbf{y}_i | \mathrm{SFR}_i, \mathrm{M}_i, \mathbf{\Theta}_i, \mathbf{\Sigma}_i)$$

$$\times \prod_{i=1}^{N} P(\mathrm{SFR}_i | \mathrm{M}_i, \beta_0, \beta_1, \sigma) P(\mathrm{M}_i) P(\mathbf{\Theta}_i),$$

where $\mathbf{SFR} = (SFR_1, \dots, SFR_N), \mathbf{M} = (\mathbf{M}_1, \dots, \mathbf{M}_N),$ $\mathbf{\Theta} = (\mathbf{\Theta}_1, \dots, \mathbf{\Theta}_N), \mathbf{Y} = (\mathbf{y}_1, \dots, \mathbf{y}_N),$ $\mathbf{\Sigma} = (\mathbf{\Sigma}_1, \dots, \mathbf{\Sigma}_N),$ and N is the number of galaxies.

Input:

Constant SFHs drawn from M*-SFR relation with known parameters (intercept, slope and scatter).

- •No Dust
- Single metallicity
- Single Redshift
- Including nebular emission
- CANDELS Deep depths/filters + IRAC CH1 + CH2
- SED Fitting:
- Constant SFHs
- •No Dust
- Redshift left free
- Metallicity left free

Test:

• Can we recover the input M*-SFR relation parameters?













SUMMARY

BEAGLE allows us to both produce mock photometry with selfconsistent SF and chemical enrichment histories and nebular emission as well as performing full Bayesian SED fitting (see Jacopo Chevallard's talk and Julia Gutkin's poster).

Our Bayesian Hierarchical modelling can recover and provide the full uncertainties in slope, intercept and intrinsic scatter of the M*-SFR relation.

- No assumption of shape of joint uncertainties in M*-SFR for individual objects.
- Allows us to assess and compare different selection and SED fitting strategies.

For more results - see my talk at the Malta "Signals from the Deep Past" conference