

THE SUB-LINEAR AND NON- UNIVERSAL KENNICUTT- SCHMIDT RELATIONSHIP

Rahul Shetty,¹

Brandon Kelly,² Frank Bigiel,¹ Nurur Rahman,³

Alberto Bolatto,⁴ Paul Clark,¹ Ralf Klessen,¹

Lukas Konstandin,¹ Erica Hopkins,¹ Camilo Penalzoza¹

¹Institute für Theoretische Astrophysik, Universität Heidelberg, Germany

²U. of California, Santa Barbara, USA

³U. of Johannesburg, South Africa

⁴U. of Maryland, USA



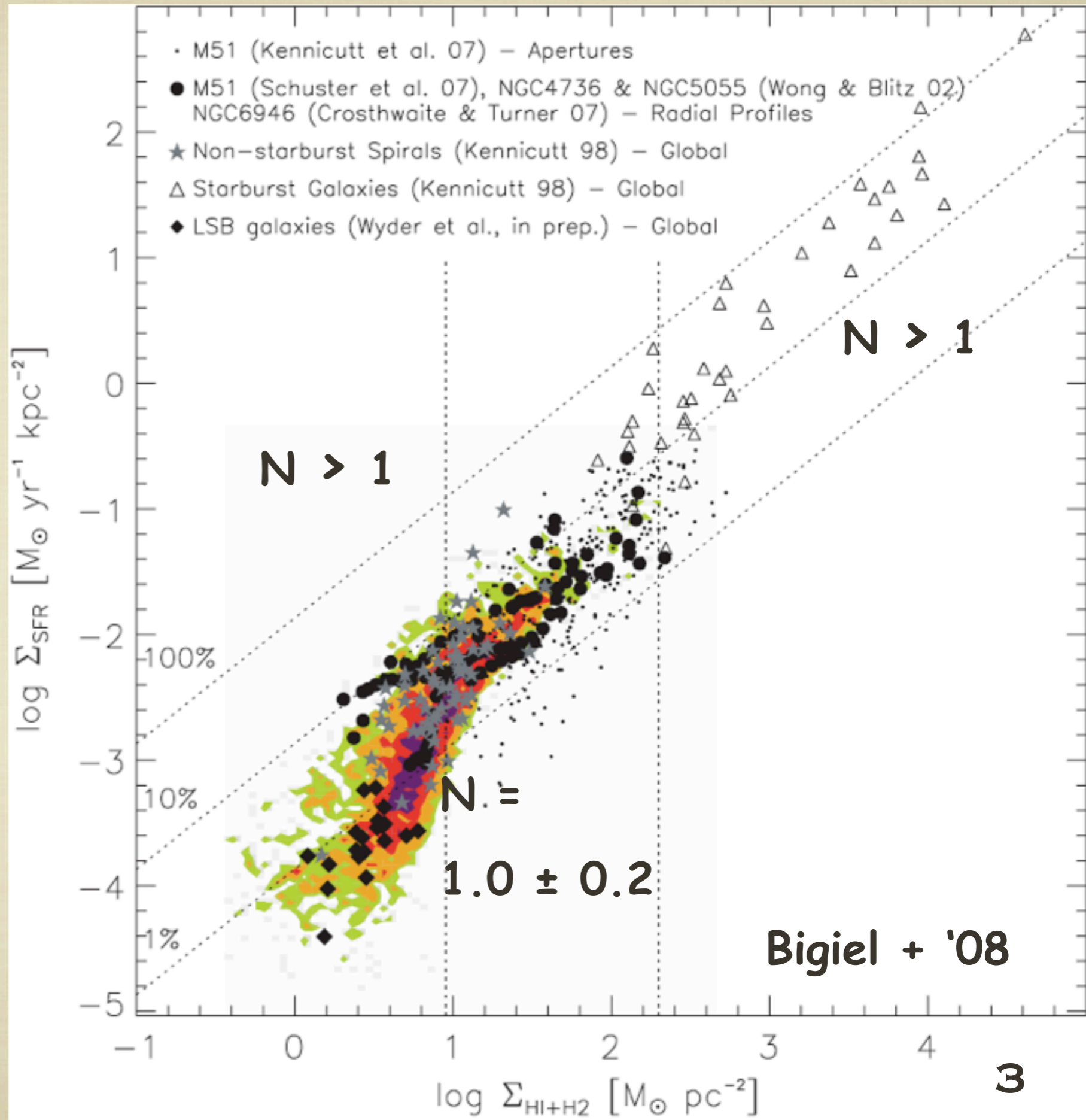
OVERVIEW

- The Kennicutt-Schmidt (KS) Relationship
- Results
- Fitting the KS relationship
 - OLS Methods ($y|x$, $x|y$, bisector)
 - Hierarchical Data
 - Monte Carlo & Bayesian Methods
 - Assessing fits, including “by eye”
- Results and Interpretation of KS parameters
- Caveats, Future Work, and Summary

THE KS RELATIONSHIP

$$\Sigma_{\text{SFR}} \propto \Sigma_{\text{gas}}^N$$

- Schmidt '59
- Buat + '89
- Kennicutt '89
- Kennicutt '98
- Hunter + '98
- Boselli + '02
- Wong & Blitz '02
- Boissier + '03
- Heyer + '04
- Leroy + '05
- Kennicutt + '07
- Leroy + '08
- Bigiel + '08
- Blanc + '09
- Verley + '10
- Daddi + '10
- Genzel + '10
- Liu + '11
- Schruba + '11
- Rahman + '12
- Leroy + '13
- Momose + '13
- ...

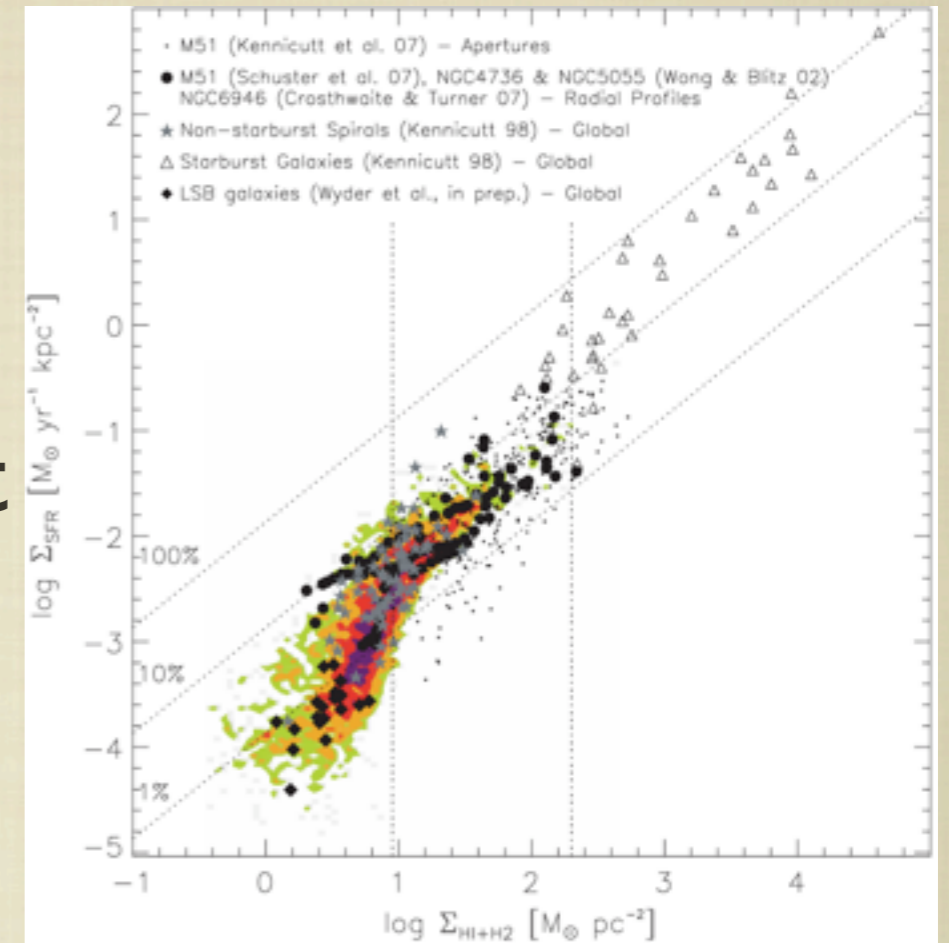


THE KS SLOPE

□ At intermediate $10 M_{\odot} \text{ pc}^{-2} < \Sigma_{\text{gas}} < 100 M_{\odot} \text{ pc}^{-2}$, Bigiel + '08 find $N \sim 1$

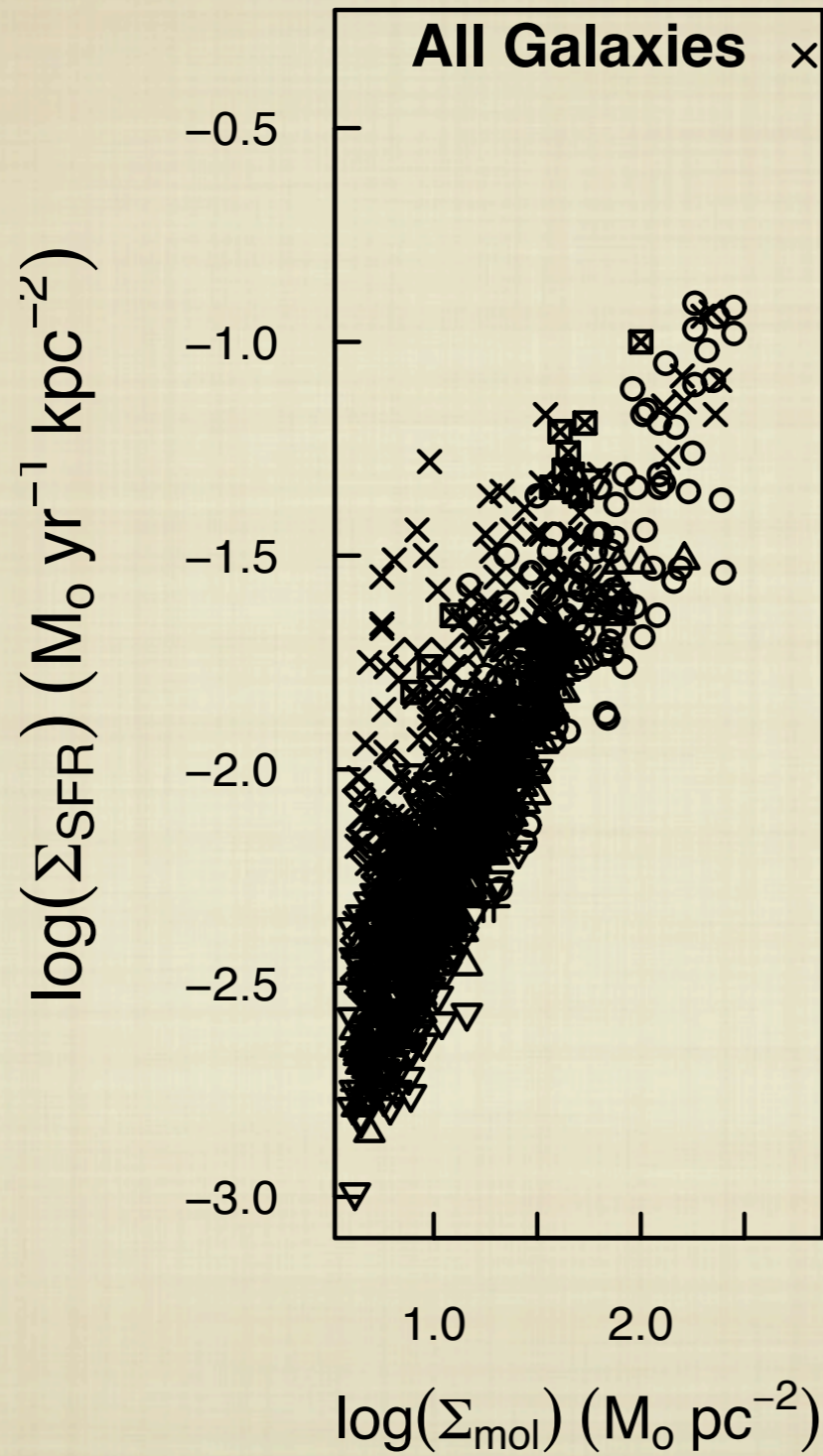
□ A linear relationship suggests a star formation efficiency, or constant depletion time $t_{\text{dep}} \sim 2 \text{ Gyr}$, with significant scatter

(e.g. Bigiel + '08, Rahman + '12, Schruba + '12 Leroy + '13, etc...)



□ Super-linear $N \sim 1.5$ KS relationship indicative of t_{dep} decreasing with surface density (e.g. Kennicutt 98, + '07, Liu + '11, Momose + '13)

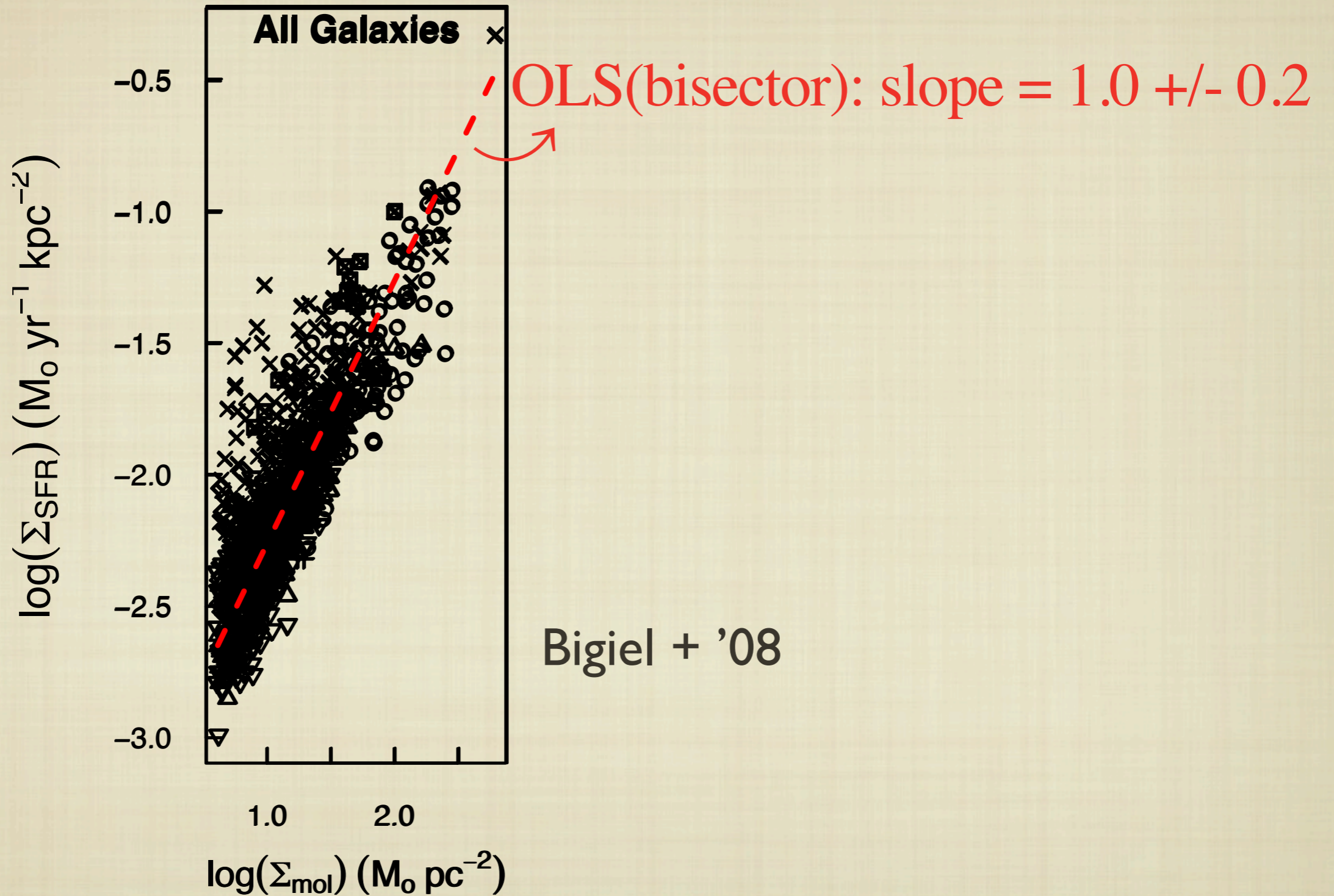
THE KS SLOPE



Bigiel + '08

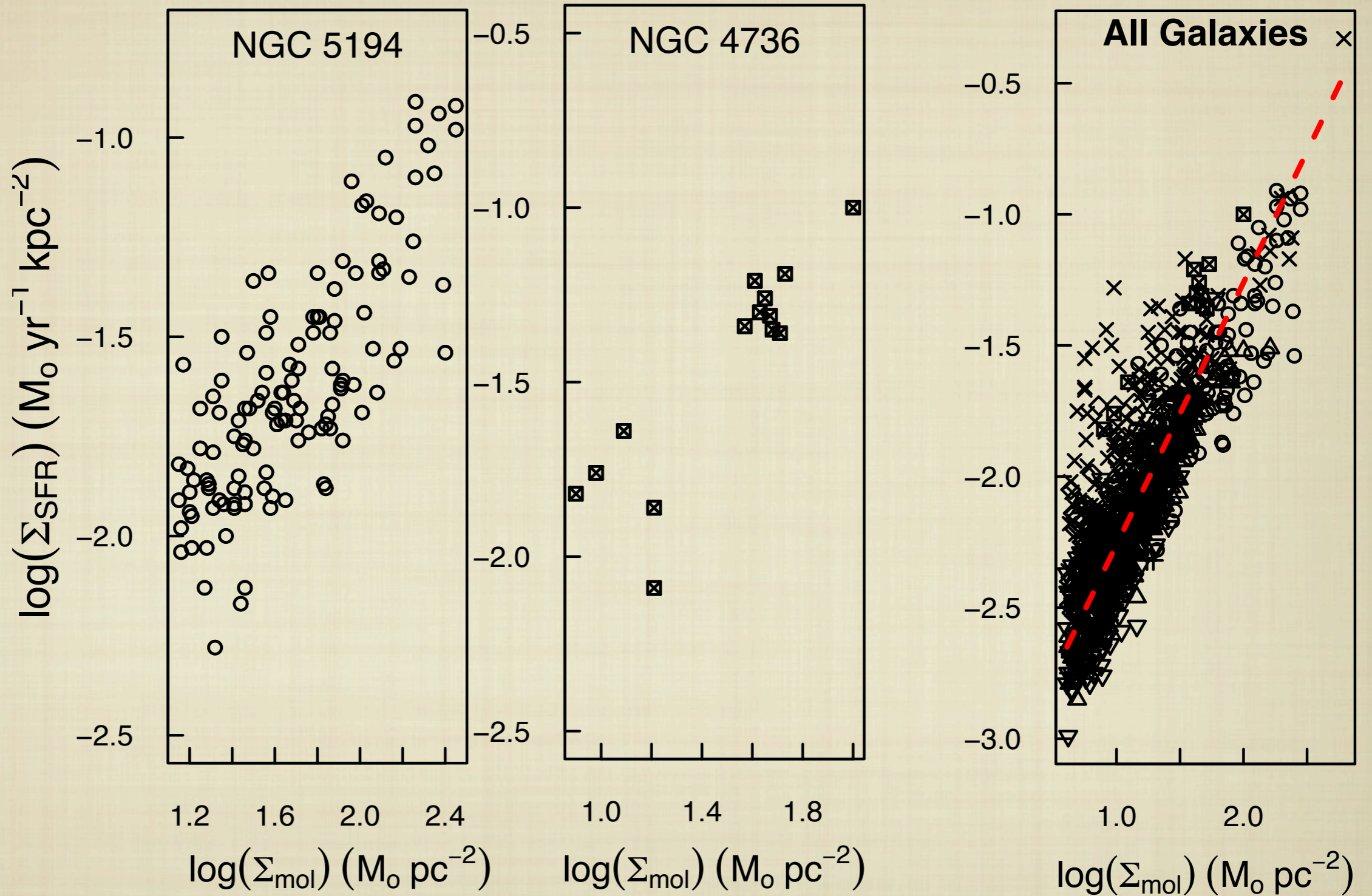
Shetty, Kelly, Bigiel '13

THE KS SLOPE

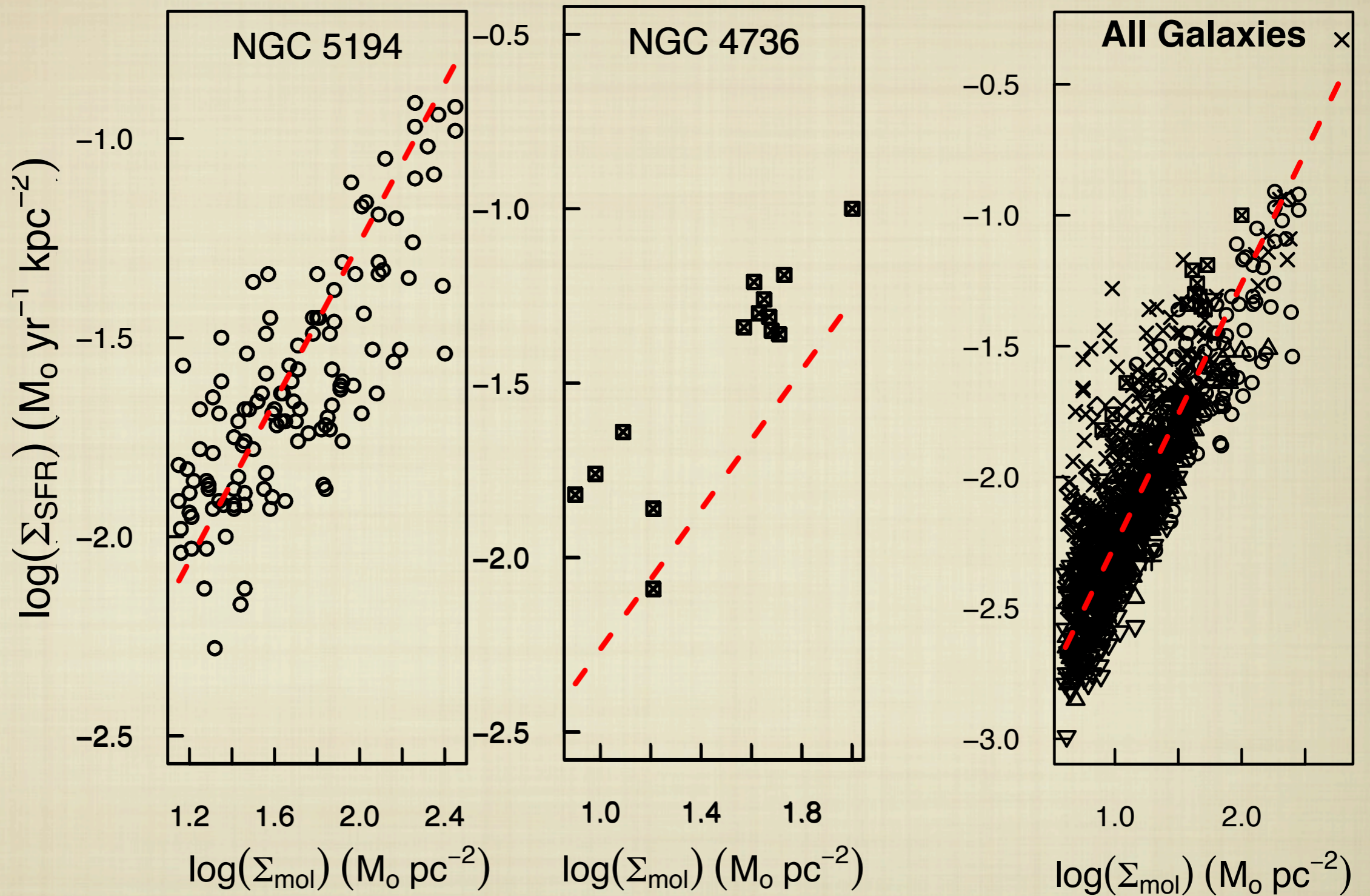


Shetty, Kelly, Bigiel '13

THE KS SLOPE



THE KS SLOPE



RESULTS:

- Analysis of resolved observations using 7 HERACLES (Leroy + '09, Bigiel + '08) galaxies and 13 STING galaxies (Rahman + '12)
- Significant galaxy-to-galaxy variation: Non-universal KS relationship
- Most galaxies portray a sub-linear KS relationship
- Mean KS relationship also sub-linear

FITTING A MODEL TO DATA

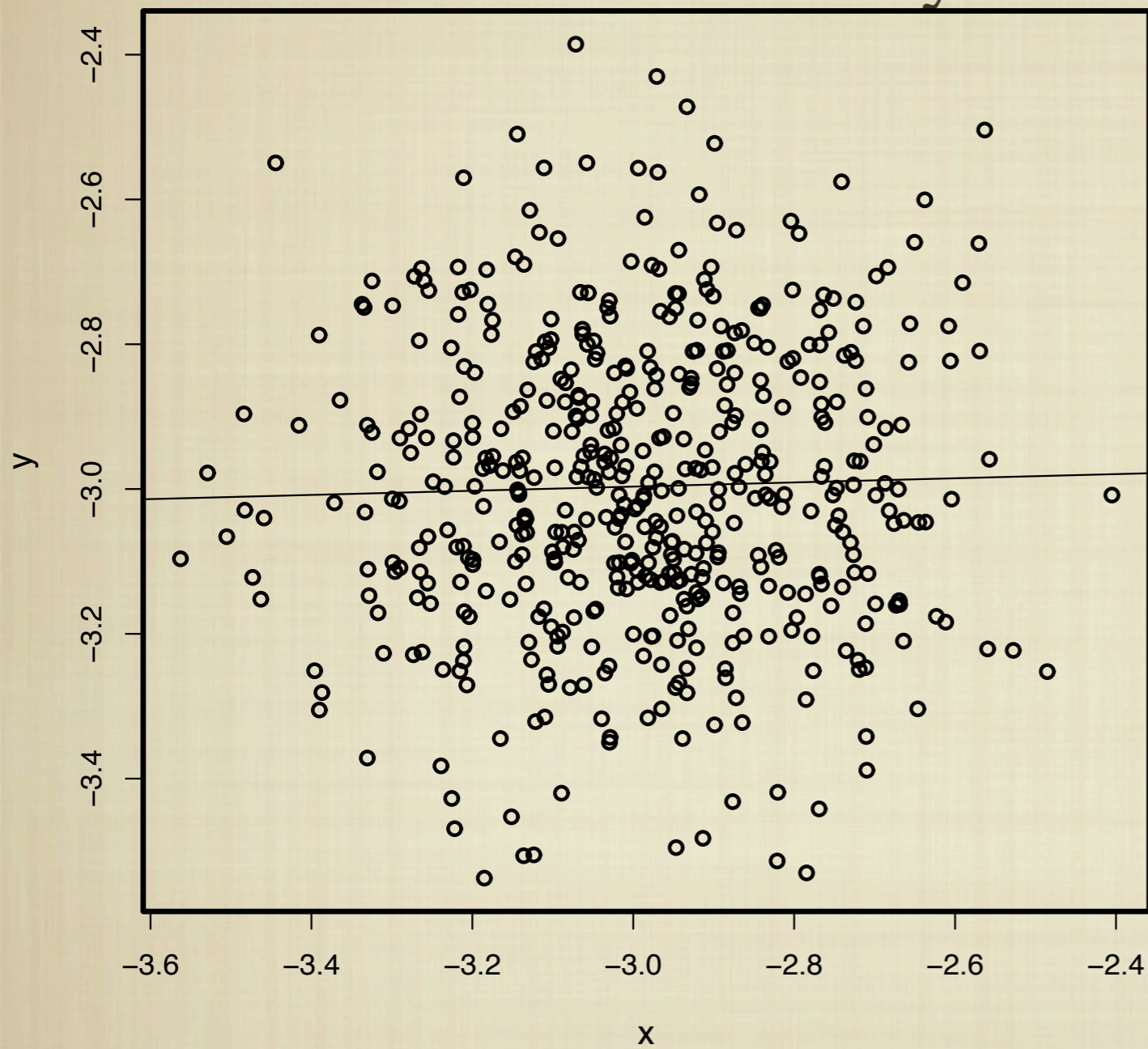
- Measurement uncertainties and intrinsic scatter may lead to biases in fitting a model to data (e.g. Weiner + '06, Kelly '07, Leroy + '12).
- Hierarchical models allow for estimating the model parameters of individual galaxies and for the ensemble
- Monte Carlo methods can account for statistical uncertainties. Bayesian inference is well suited for hierarchical problems, through MCMC methods (Gelman + '04, Gelman & Hill '07, Kruschke '11)

ASSESSING FITTING METHODS

- Important to understand the underlying assumptions and limitations of a fitting method
- Linear fit: $y = \underline{A} + \underline{N}x$. OLS($y|x$), OLS($x|y$) and OLS (bisector) are NOT the same (Isobe + '90), due to treatment of variances (Shetty, Kelly, Bigiel '13)
- Quality control: verify accuracy of fits with synthetic data
- “Seeing is believing”: By eye validity can also be effective

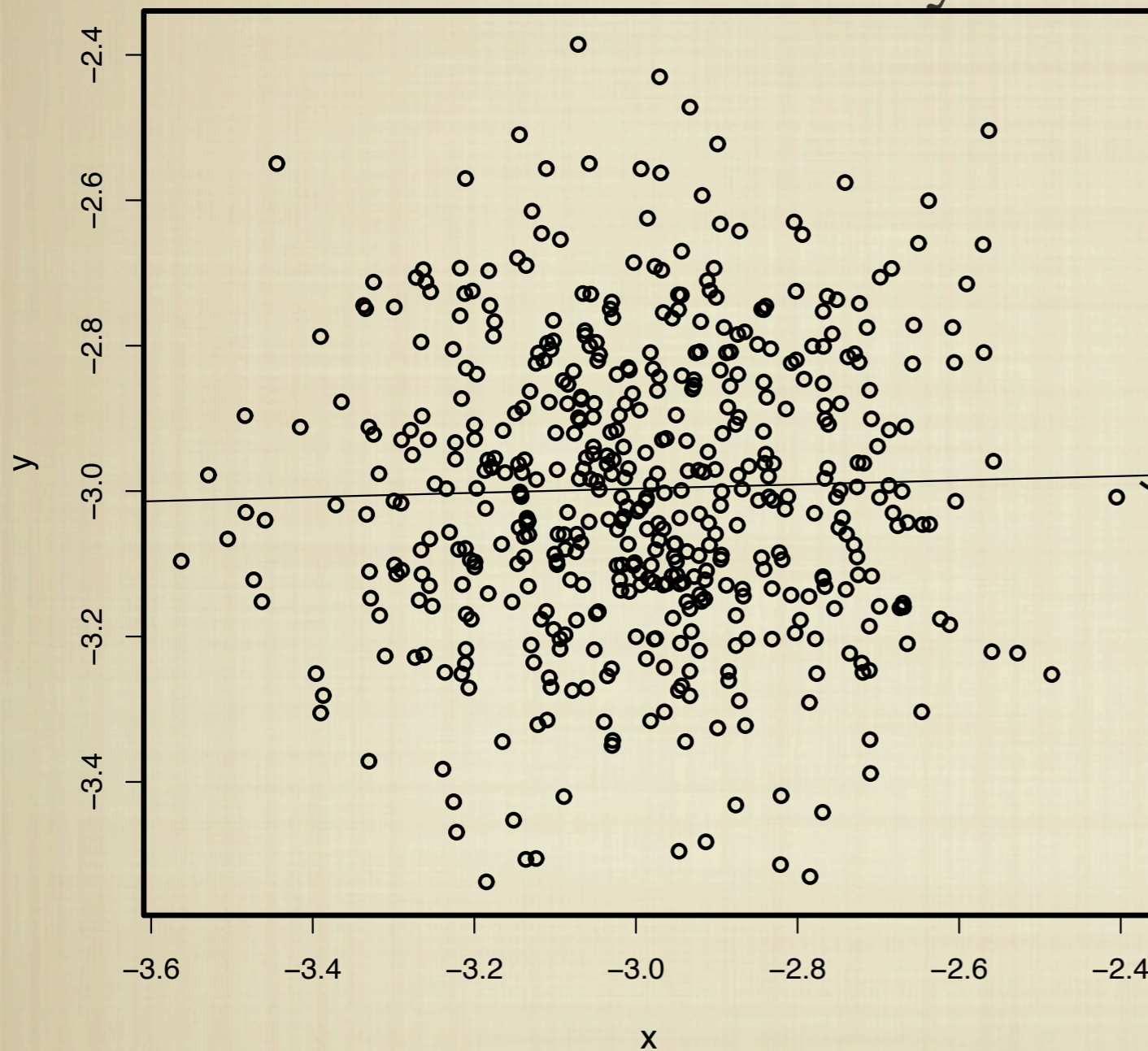
TESTING OLS METHODS

uncorrelated x & y



TESTING OLS METHODS

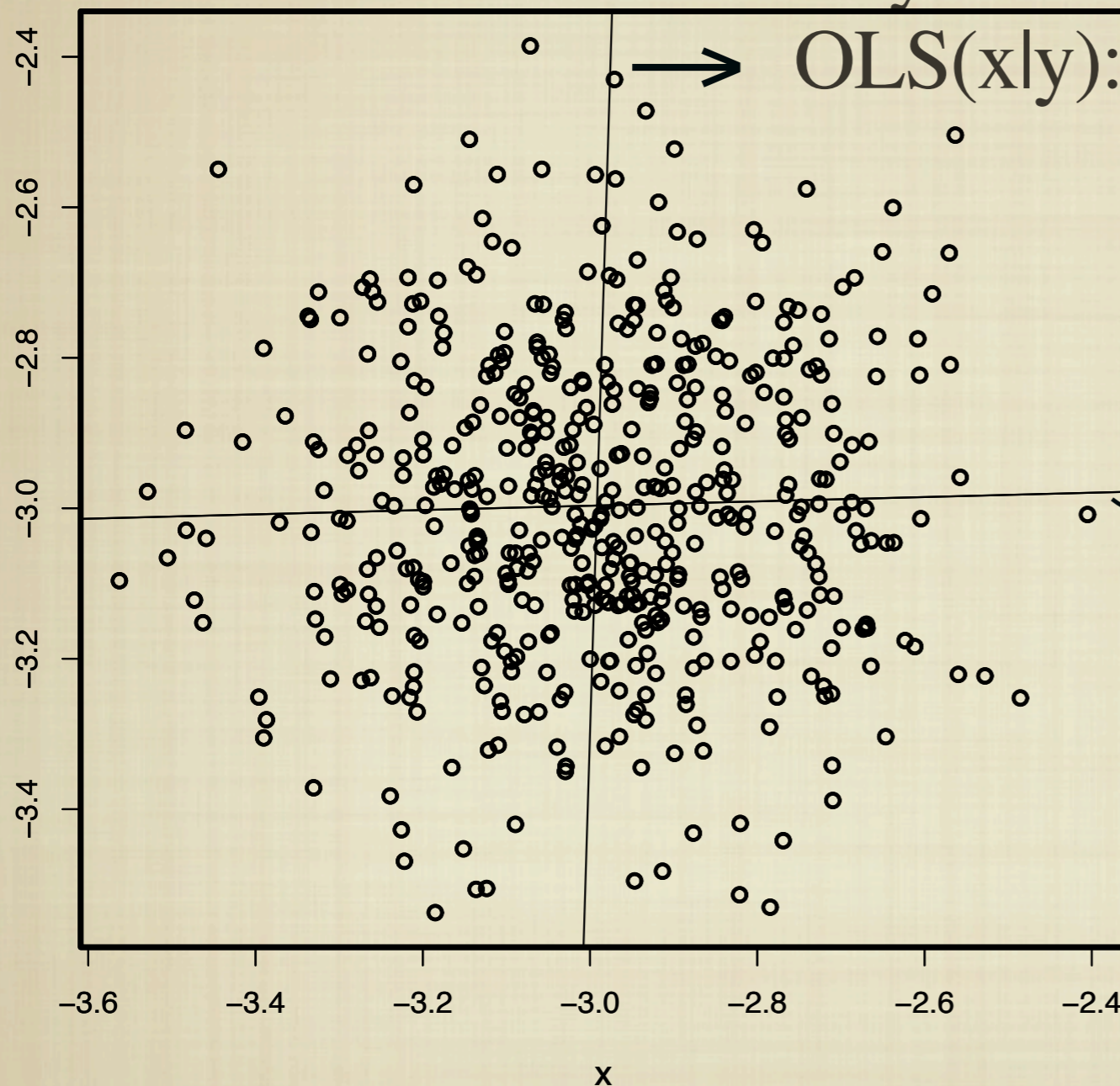
uncorrelated x & y



OLS(y|x): slope = 0.03 +/- 0.05

TESTING OLS METHODS

uncorrelated x & y

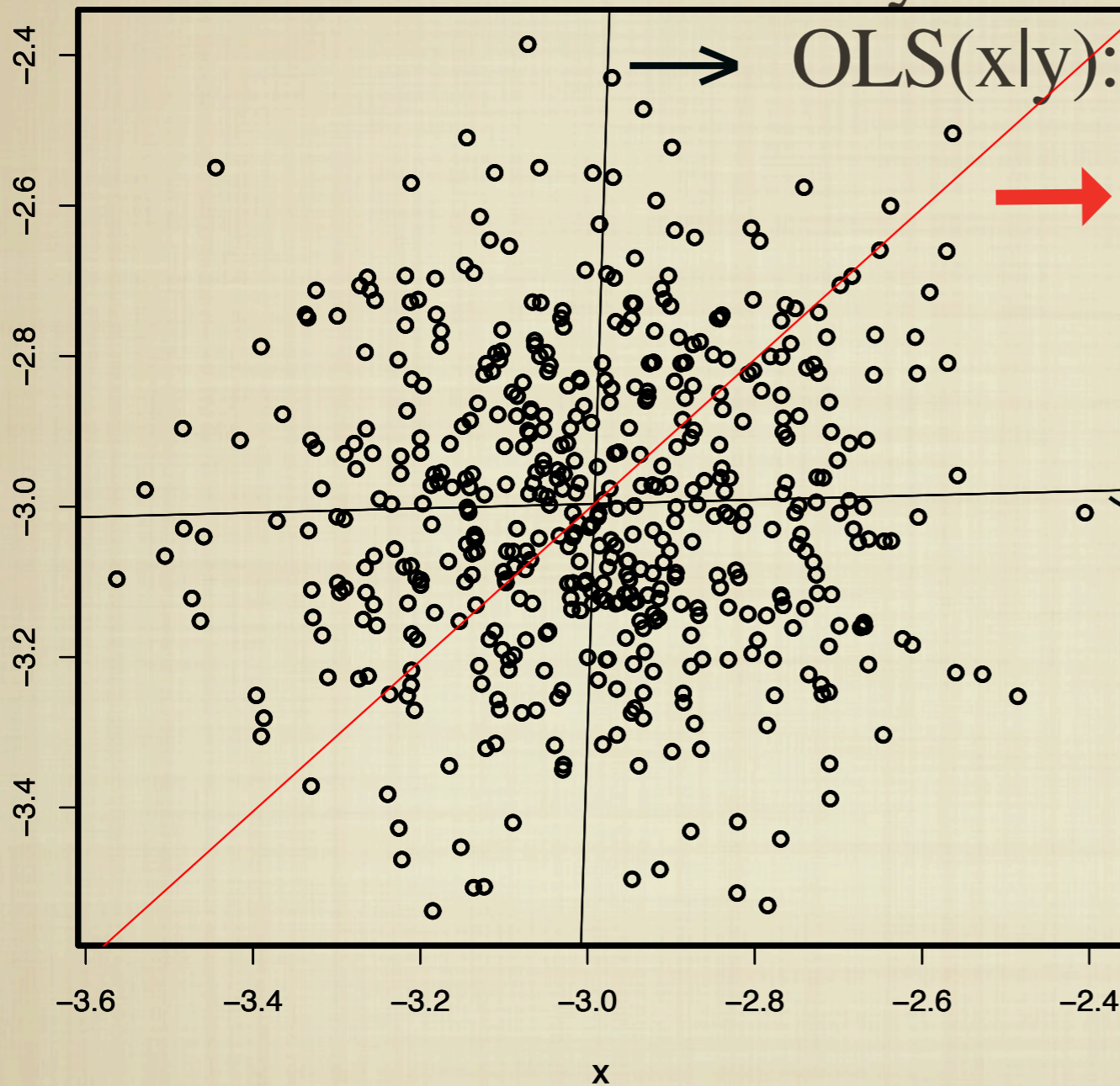


→ OLS(x|y): slope = 36.4 ± 0.04

↘ OLS(y|x): slope = 0.03 ± 0.05

TESTING OLS METHODS

uncorrelated x & y



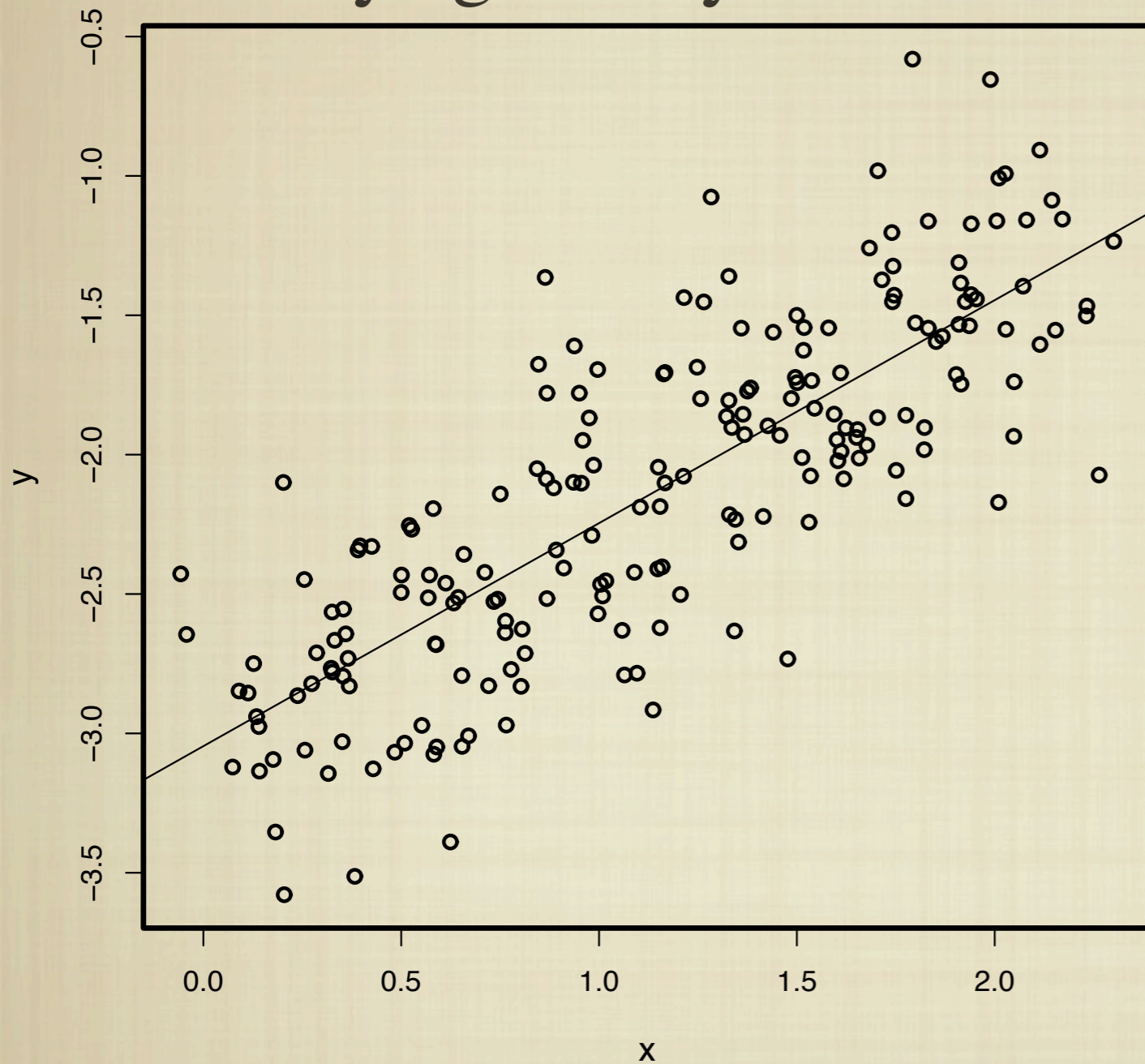
→ OLS(x|y): slope = 36.4 ± 0.04

→ OLS(bisector): slope = 1.0 ± 0.04

↘ OLS(y|x): slope = 0.03 ± 0.05

TESTING OLS METHODS

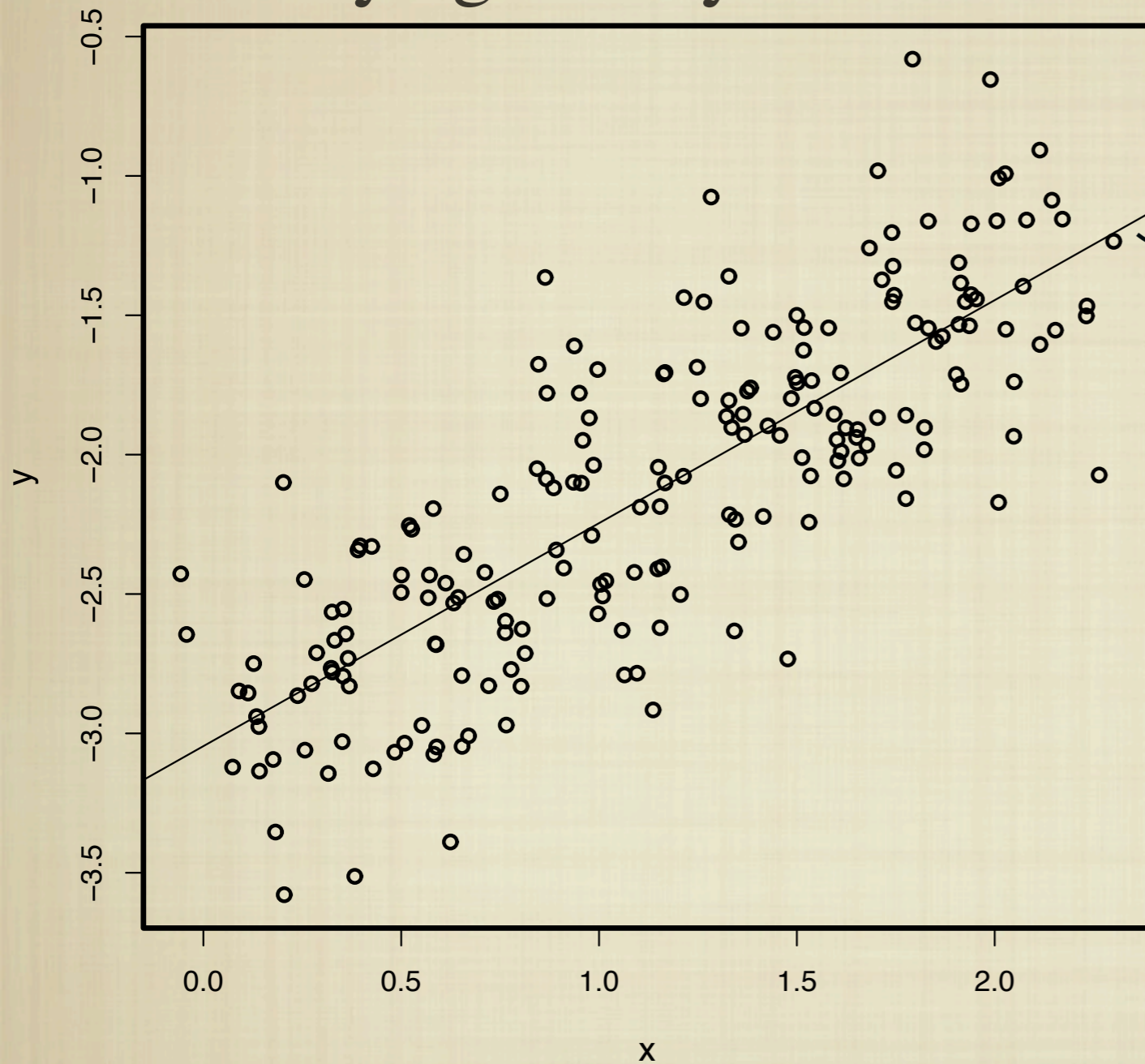
Underlying data: $y = -3 + 0.75x$



Shetty, Kelly, Bigiel '13

TESTING OLS METHODS

Underlying data: $y = -3 + 0.75x$



OLS(y|x): slope = 0.8 +/- 0.04

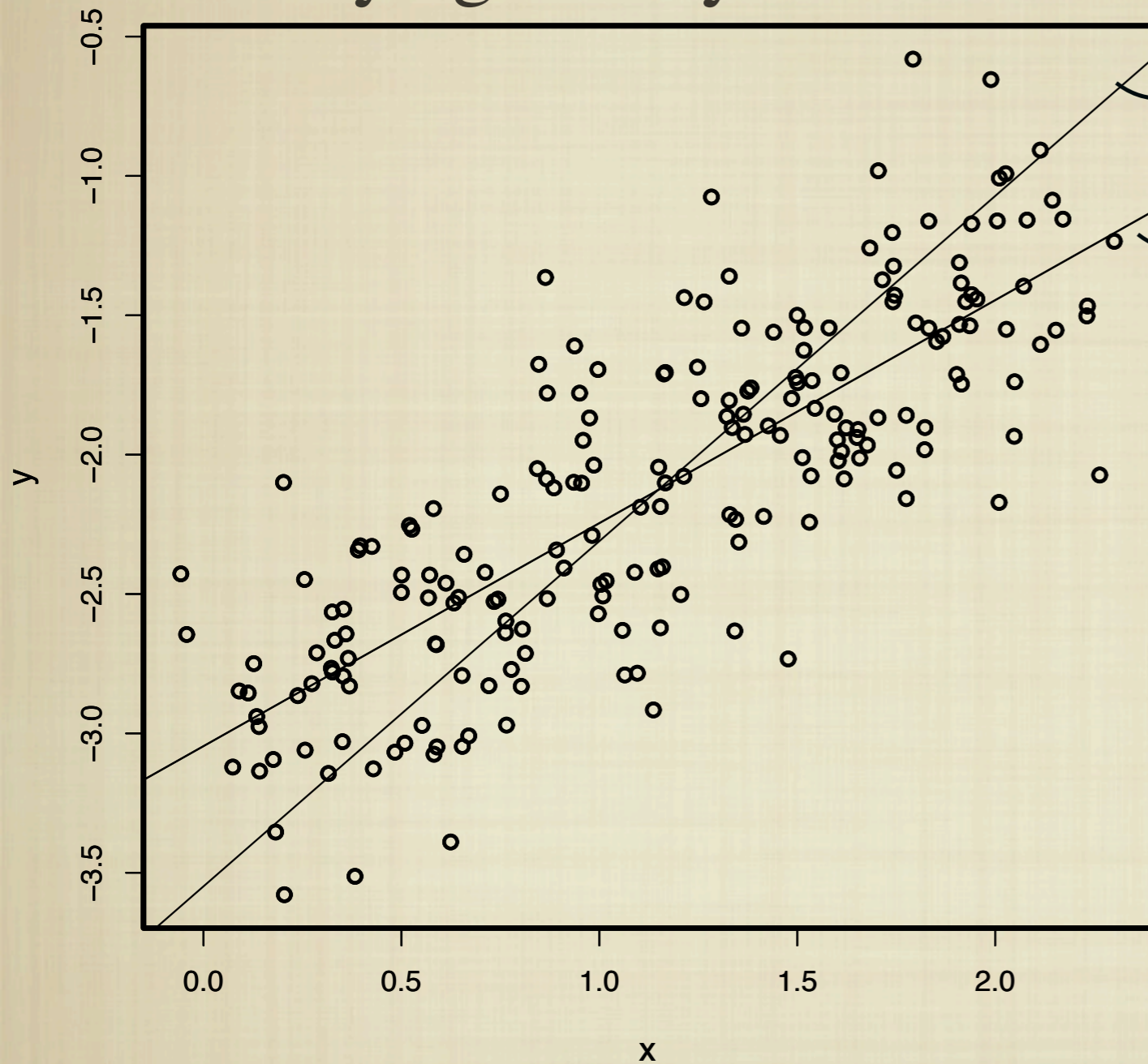
Shetty, Kelly, Bigiel '13

TESTING OLS METHODS

Underlying data: $y = -3 + 0.75x$

OLS(x|y): slope = 1.2 ± 0.04

OLS(y|x): slope = 0.8 ± 0.04



Shetty, Kelly, Bigiel '13

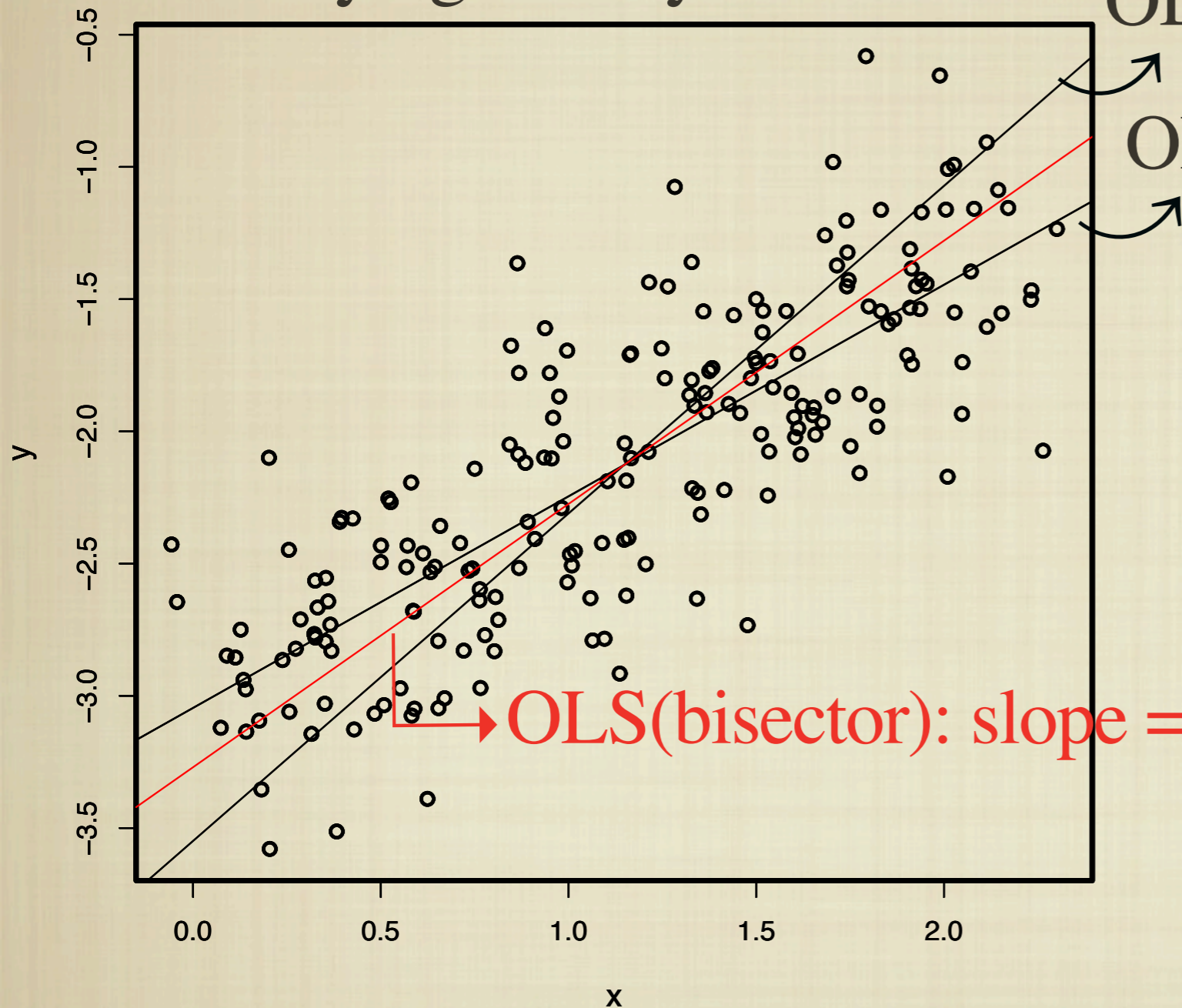
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Shetty, Kelly, Bigiel '13

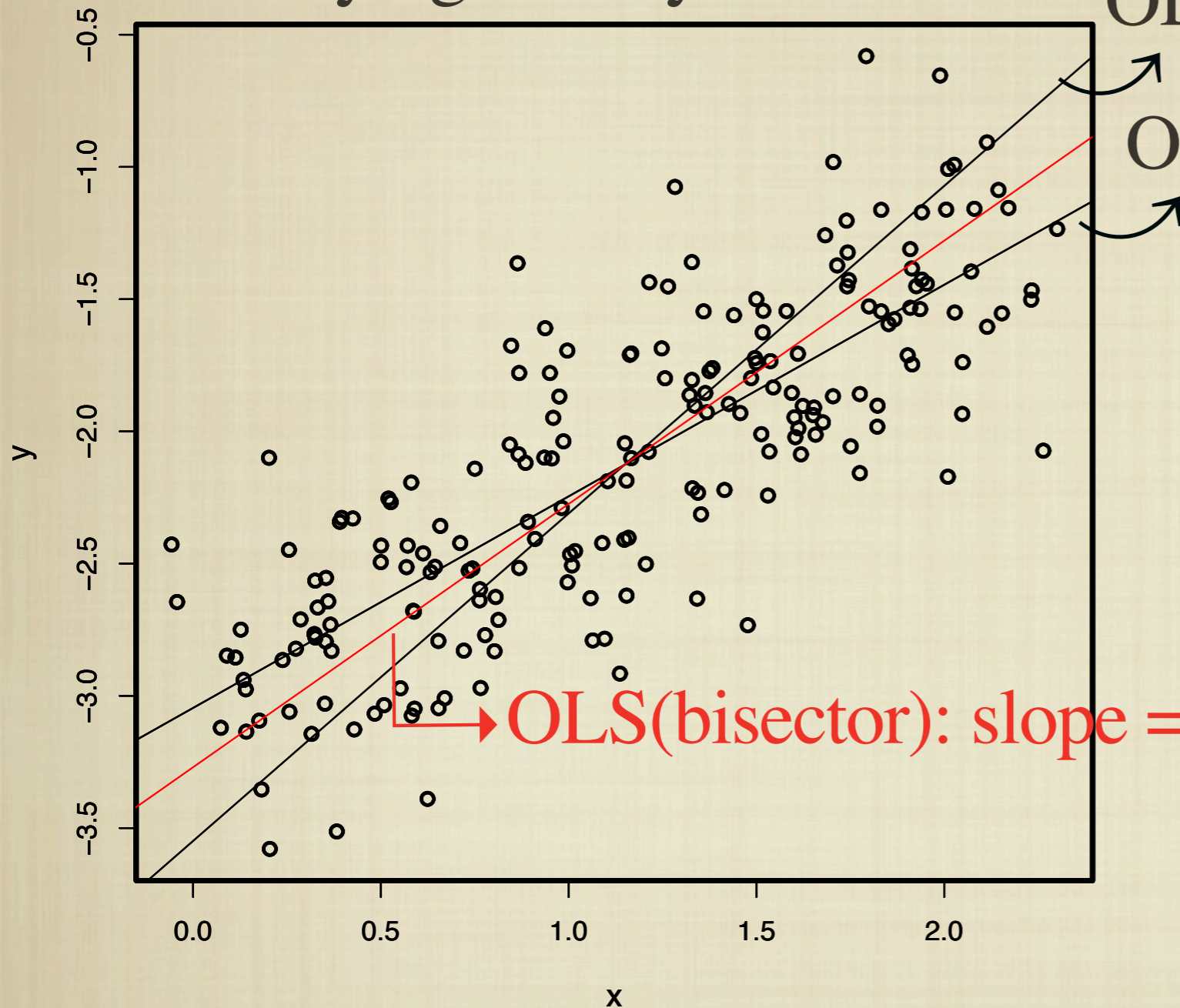
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Shetty, Kelly, Bigiel '13

⇒ **Bisector does not provide accurate parameter estimates of $y = A + Nx$**

⇒ **Bisector usually overestimates slope for $N \sim 1$**

MONTE CARLO METHODS

- Monte Carlo simulations can rigorously account for uncertainties
- fit a number of realizations of the data, given the uncertainties
- final parameter estimates are PDFs

HIERARCHICAL BAYESIAN FITS

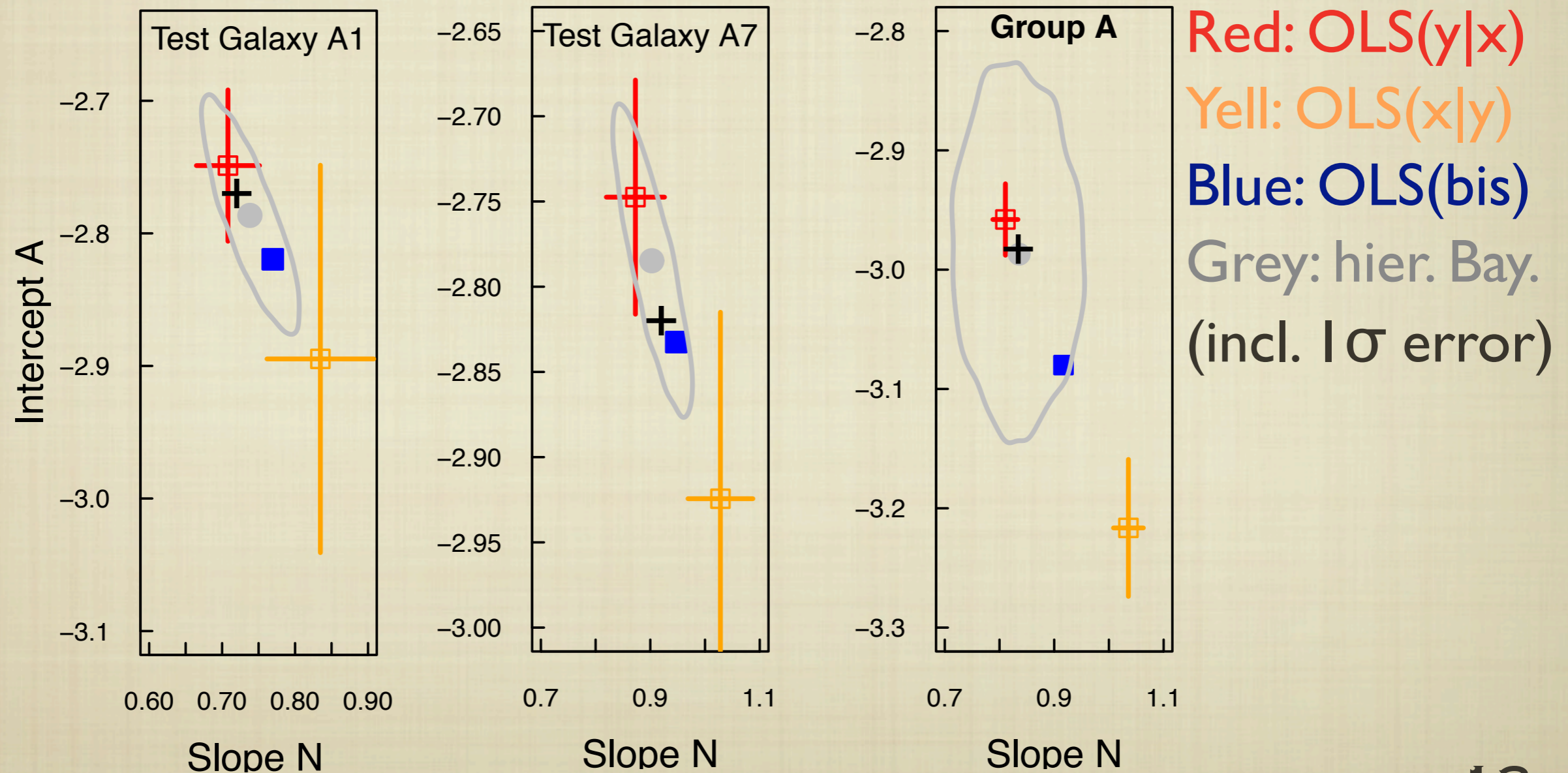
- Hierarchical Bayesian linear regression: $y = \underline{A} + \underline{N}x + \underline{\epsilon}_{\text{scat}}$
 - Estimate parameters for **both** individuals and ensemble. MCMC provides robust uncertainty estimates (convention to use 2σ as plausible range)
- Verify parameter estimates using synthetic data

TESTING THE HIERARCHICAL BAYESIAN FIT

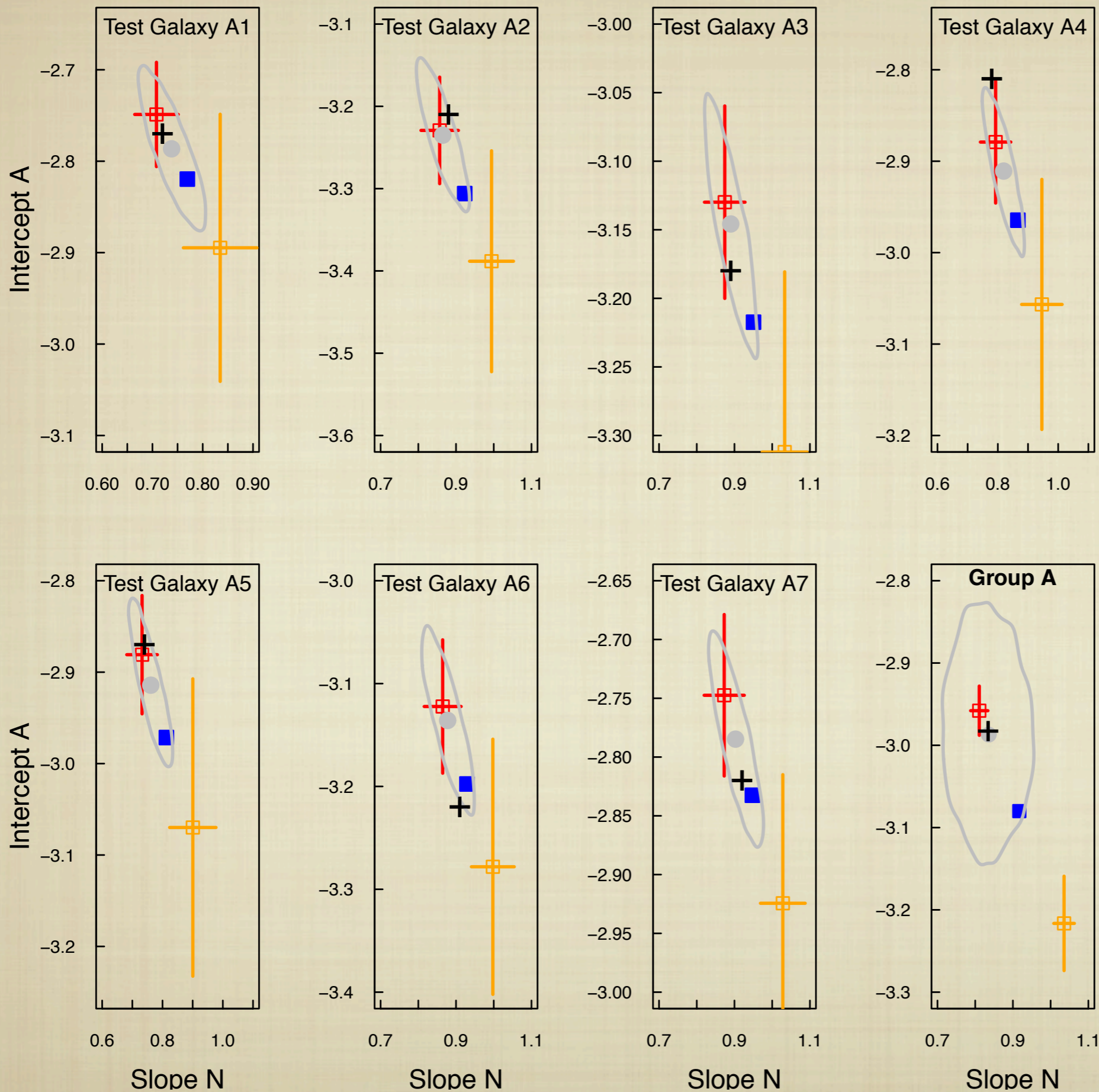
- Use synthetic dataset
- Intrinsic relationship: $\log(\Sigma_{\text{SFR}}) = A + N \log(\Sigma_{\text{mol}}) + \epsilon_{\text{scat}}$
- Include noise in both Σ_{SFR} and Σ_{mol}
- 7 Individuals within group

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TESTING THE HIERARCHICAL BAYESIAN FIT

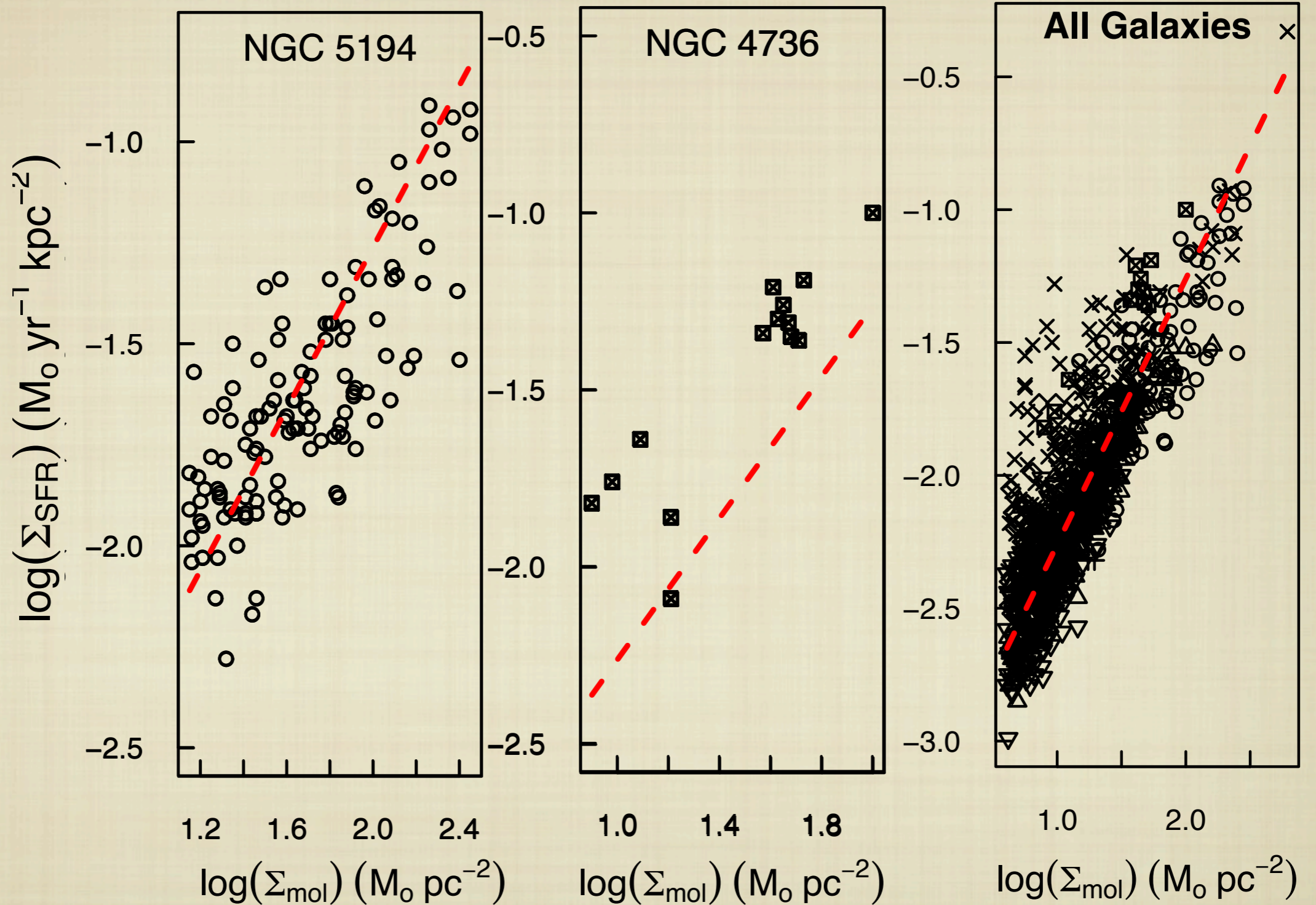


Red: OLS($y|x$)
Yell: OLS($x|y$)
Blue: OLS(bis)
Grey: hier. Bay.
(incl. 1σ error)

THE HERACLES AND STING SAMPLES

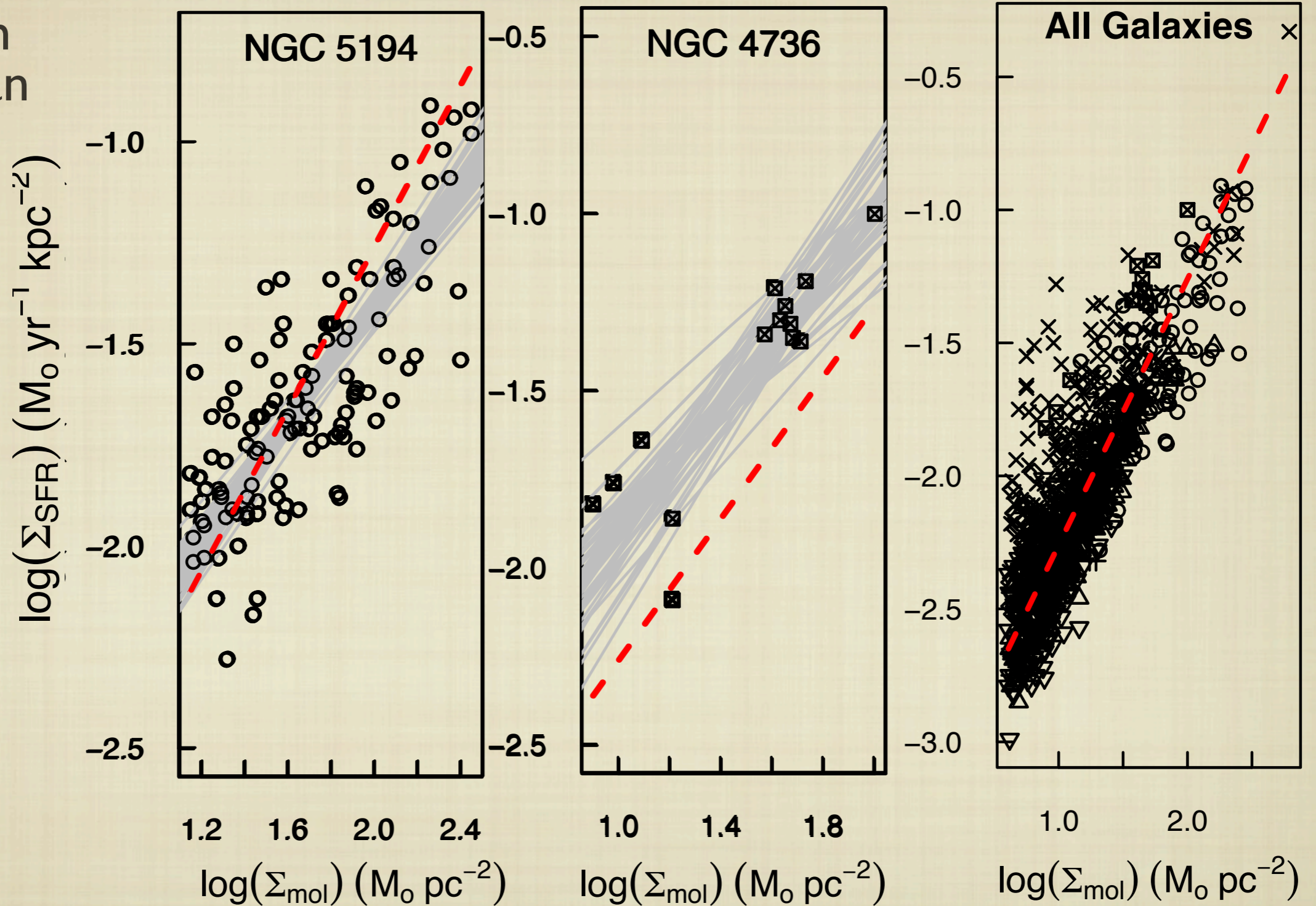
- HERACLES (Leroy + '09): CO [(2-1), but (1-0) for M51] 24 μm SINGS (Kennicutt + '03), and UV from GALEX (Gil de Paz + '07) for 30 Galaxies
- Bigiel + '08 sub-sample of HERACLES: 750 pc regions for 7 Galaxies (Bigiel + '10)
- STING: CARMA CO (1-0) and 24 μm SINGS for 13 Galaxies (Rahman + '11, '12)

The KS Relationship of the Bigiel + '08 Sample



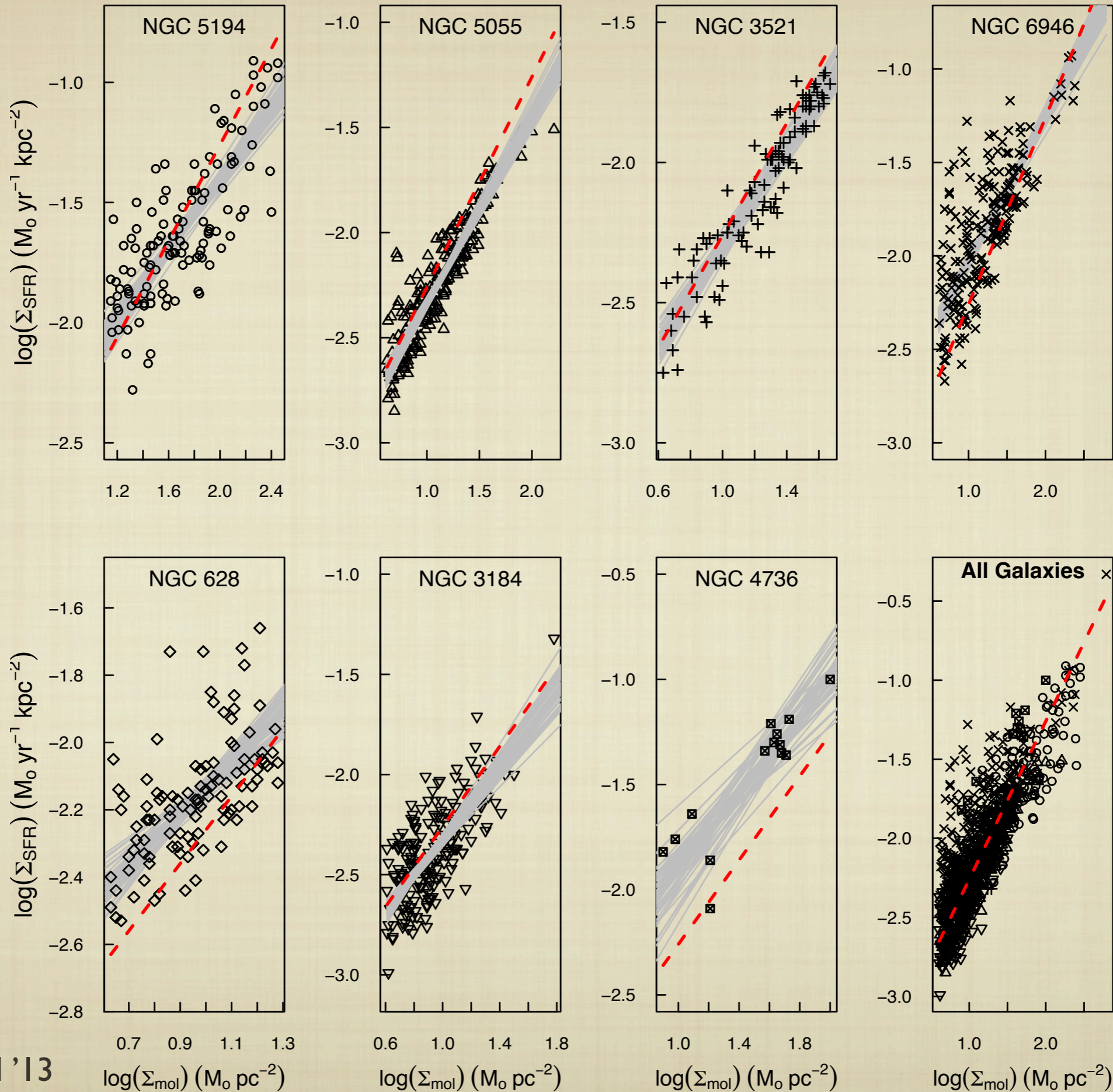
The KS Relationship of the Bigiel + '08 Sample

Gray lines:
50 random
draws from
the Bayesian
posterior



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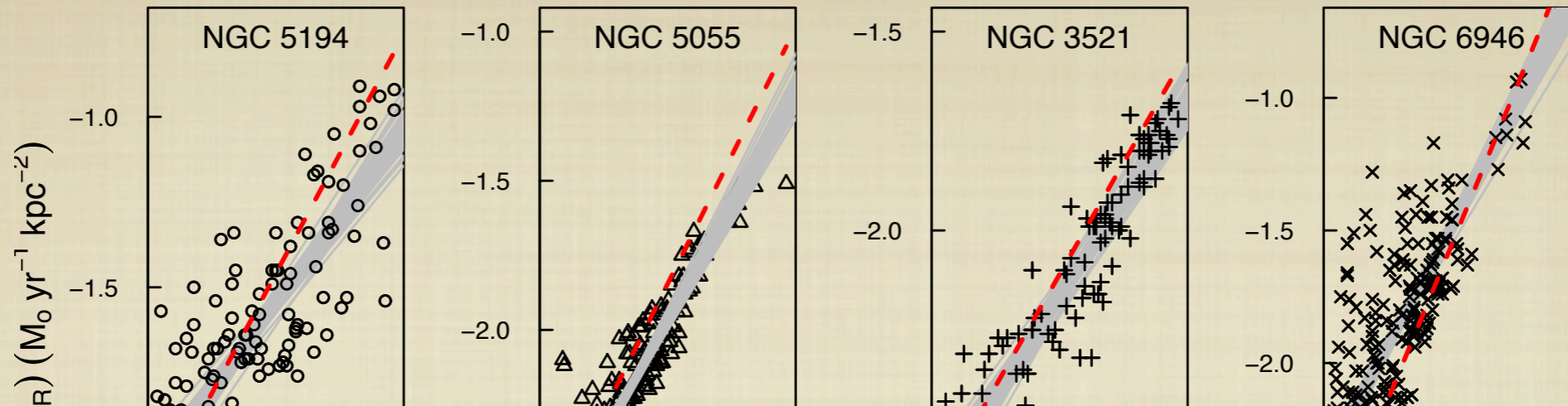
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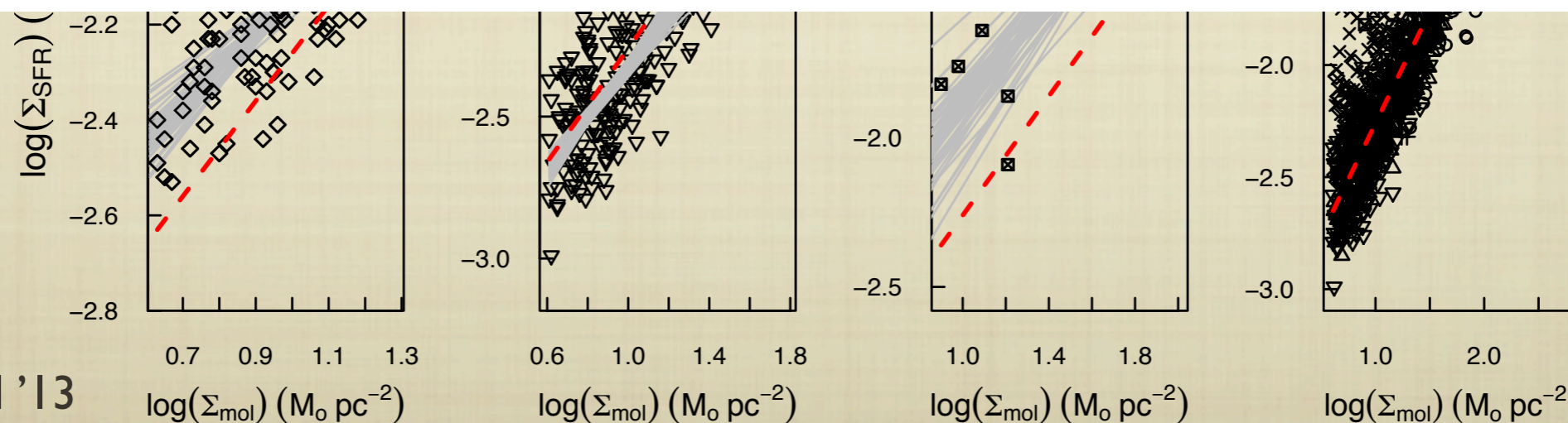
Bigiel + '08
Shetty, Kelly, Bigiel '13

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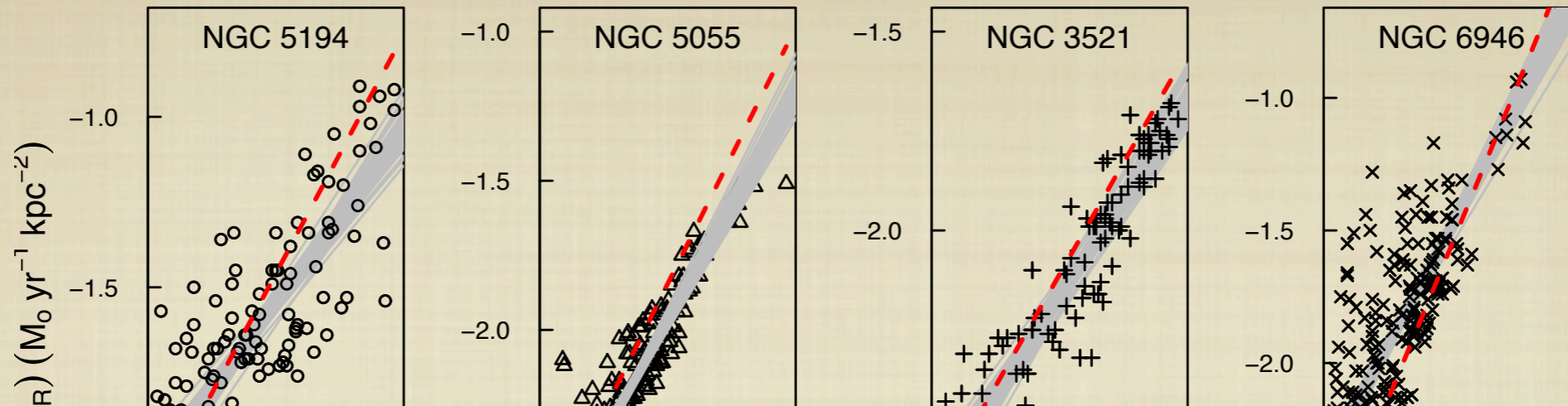


Subject	Bayes A	Bayes $2\sigma_A$	Bayes N	Bayes $2\sigma_N$	Bayes σ_{scat}
NGC 5194 (M51)	-2.84	[-3.0, -2.7]	0.72	[0.62, 0.83]	0.06
NGC 5055	-3.20	[-3.3, -3.1]	0.87	[0.79, 0.95]	0.04
NGC 3521	-3.20	[-3.4, -3.0]	0.90	[0.76, 1.03]	0.05
NGC 6946	-2.81	[-2.9, -2.7]	0.78	[0.70, 0.86]	0.11
NGC 628	-2.89	[-3.1, -2.6]	0.76	[0.51, 0.95]	0.05
NGC 3184	-3.24	[-3.4, -3.1]	0.92	[0.79, 1.10]	0.05
NGC 4736	-2.83	[-3.2, -2.4]	0.92	[0.67, 1.20]	0.08
Group Parameters	-3.00	[-3.3, -2.7]	0.84	[0.63, 1.0]	0.14

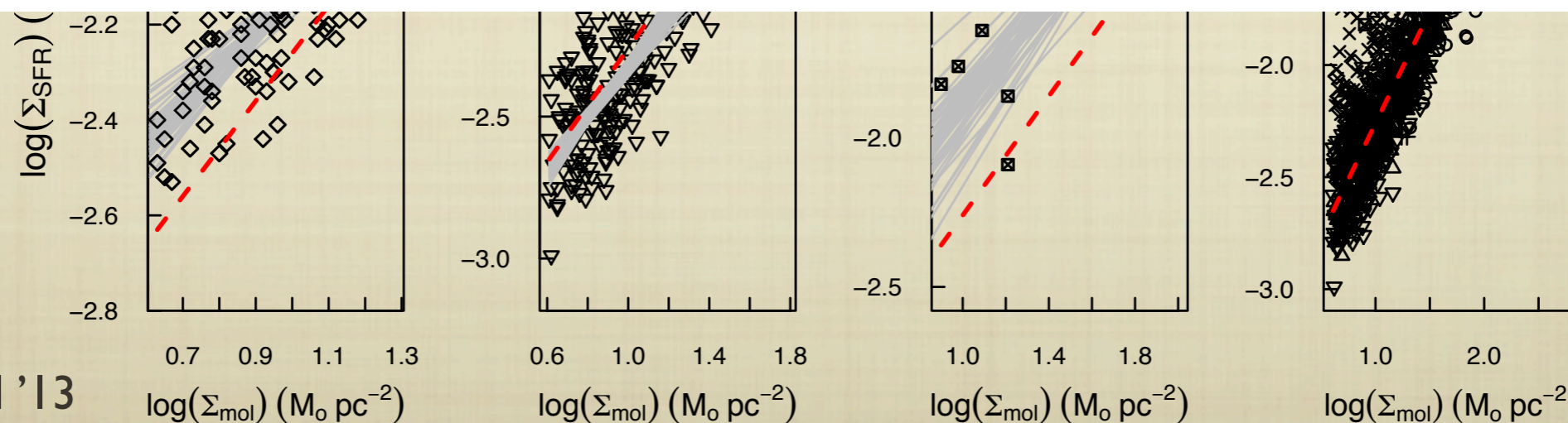


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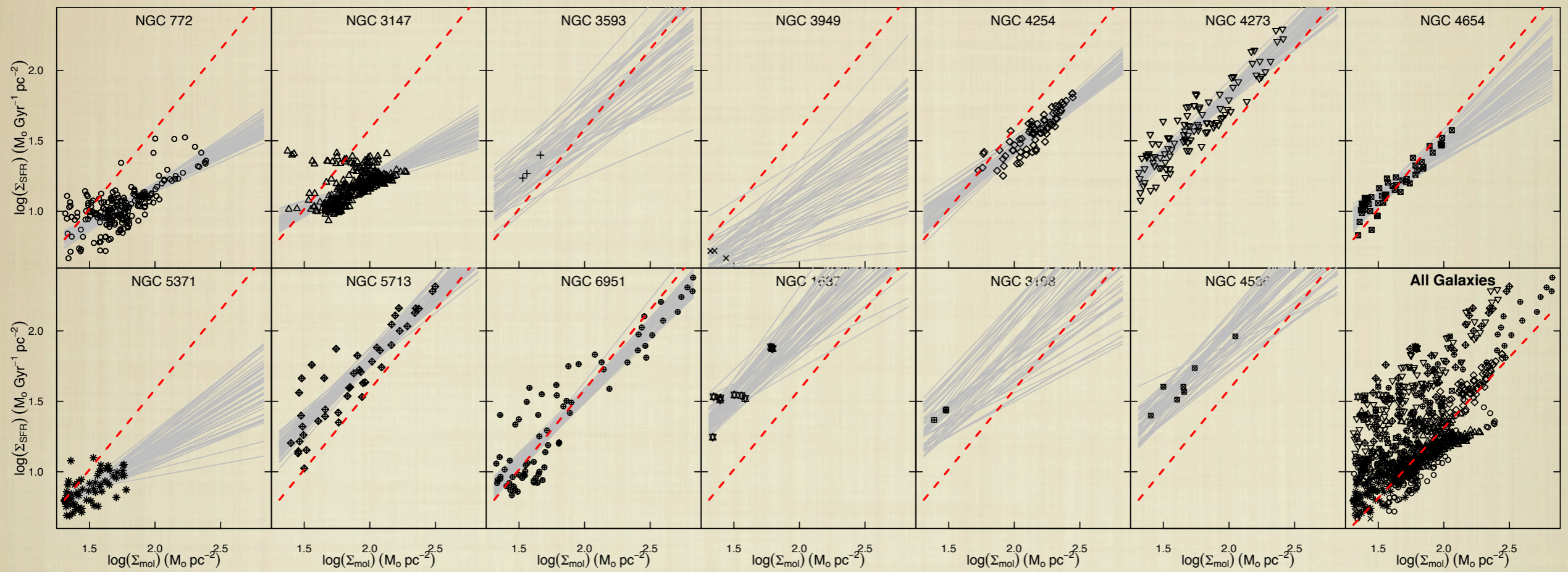
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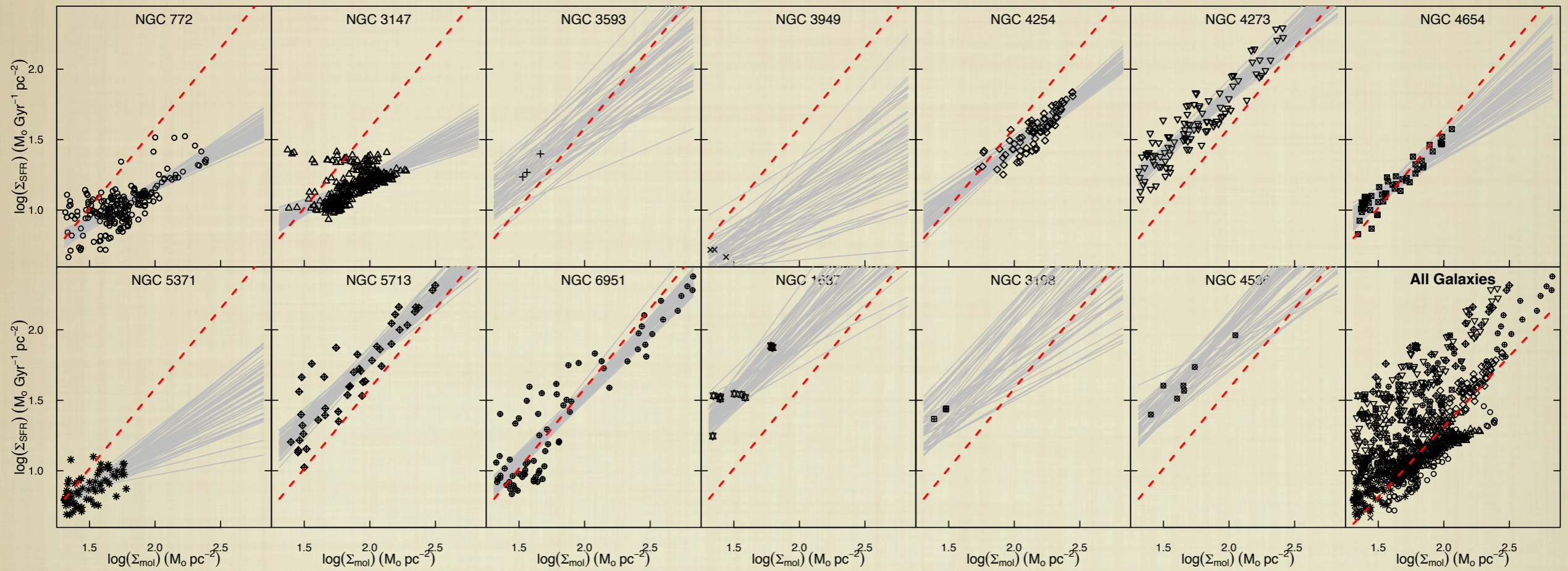


The KS Relationship of the STING Sample



Rahman + '11, '12,
Shetty + '13b

The KS Relationship of the STING Sample



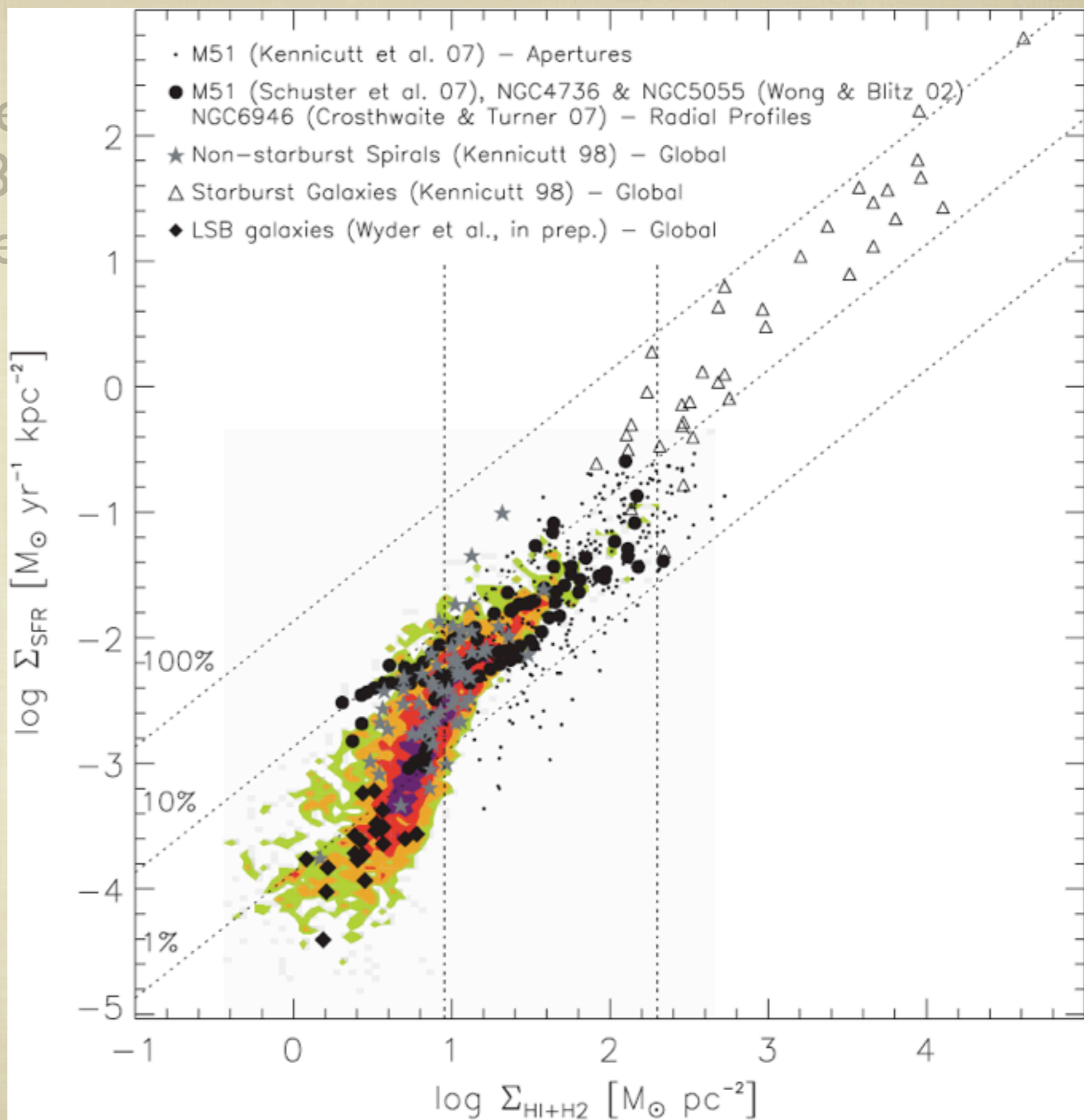
Individual slopes range from 0.42 - 0.95
Mean Slope = 0.76; $2\sigma = [0.58 - 0.94]$

IMPORTANCE OF HIERARCHICAL FITS

- Even without hierarchical fit, Bayesian fit slope of Bigiel + '08 sample [0.88 - 0.95], indicating sub-linear KS relationship (Shetty, Kelly, Bigiel '13)
- By eye fit...?

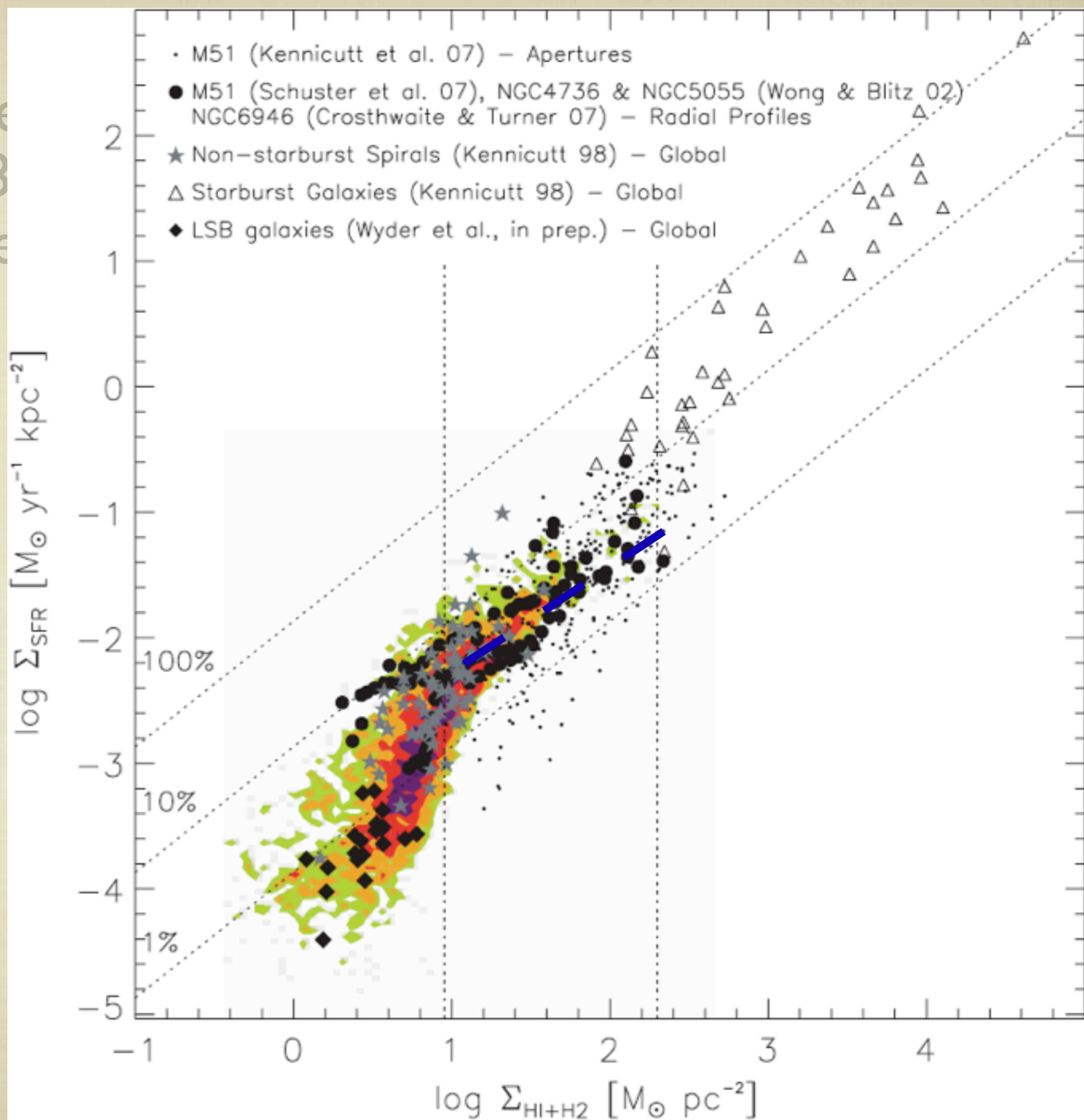
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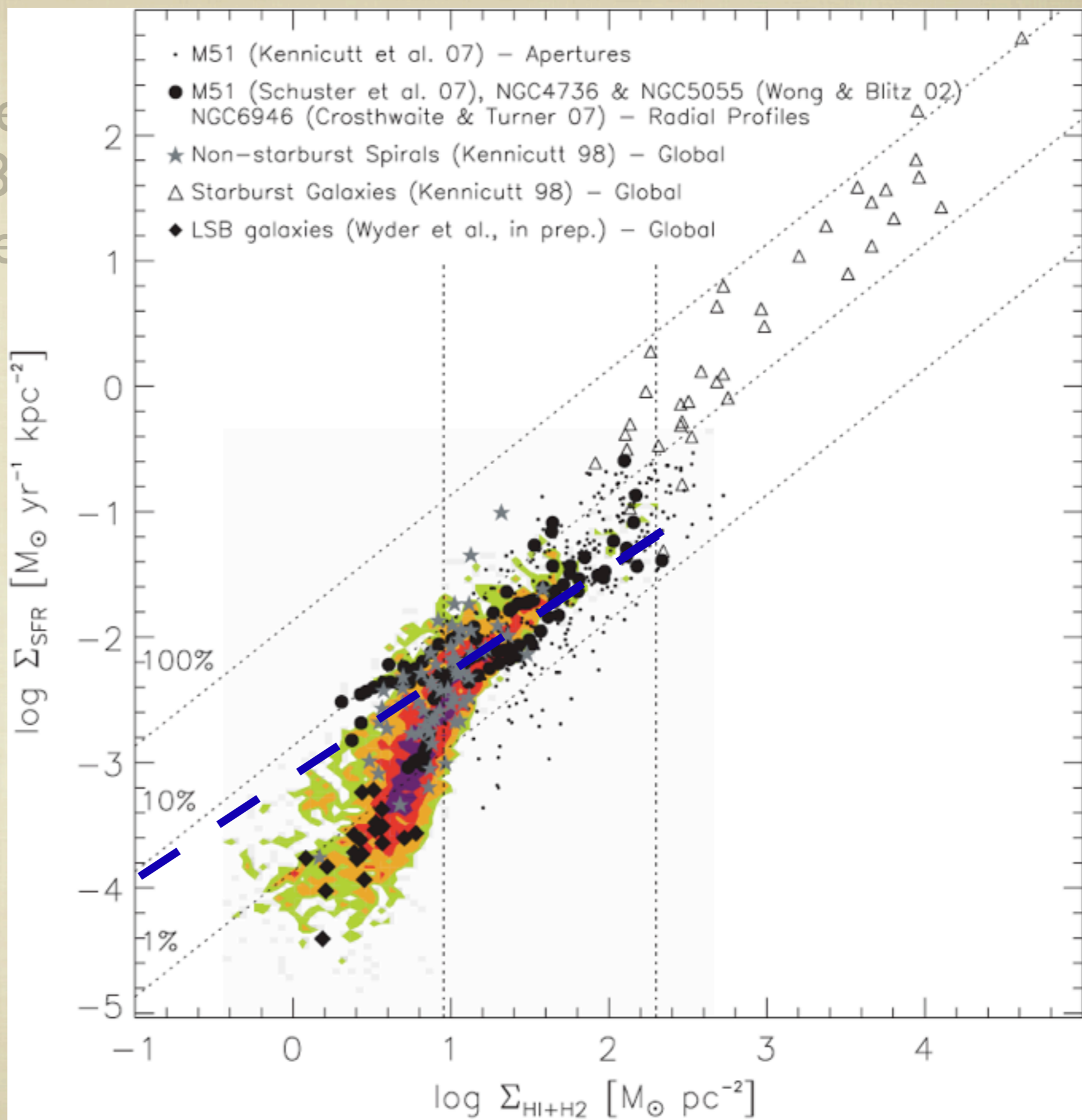
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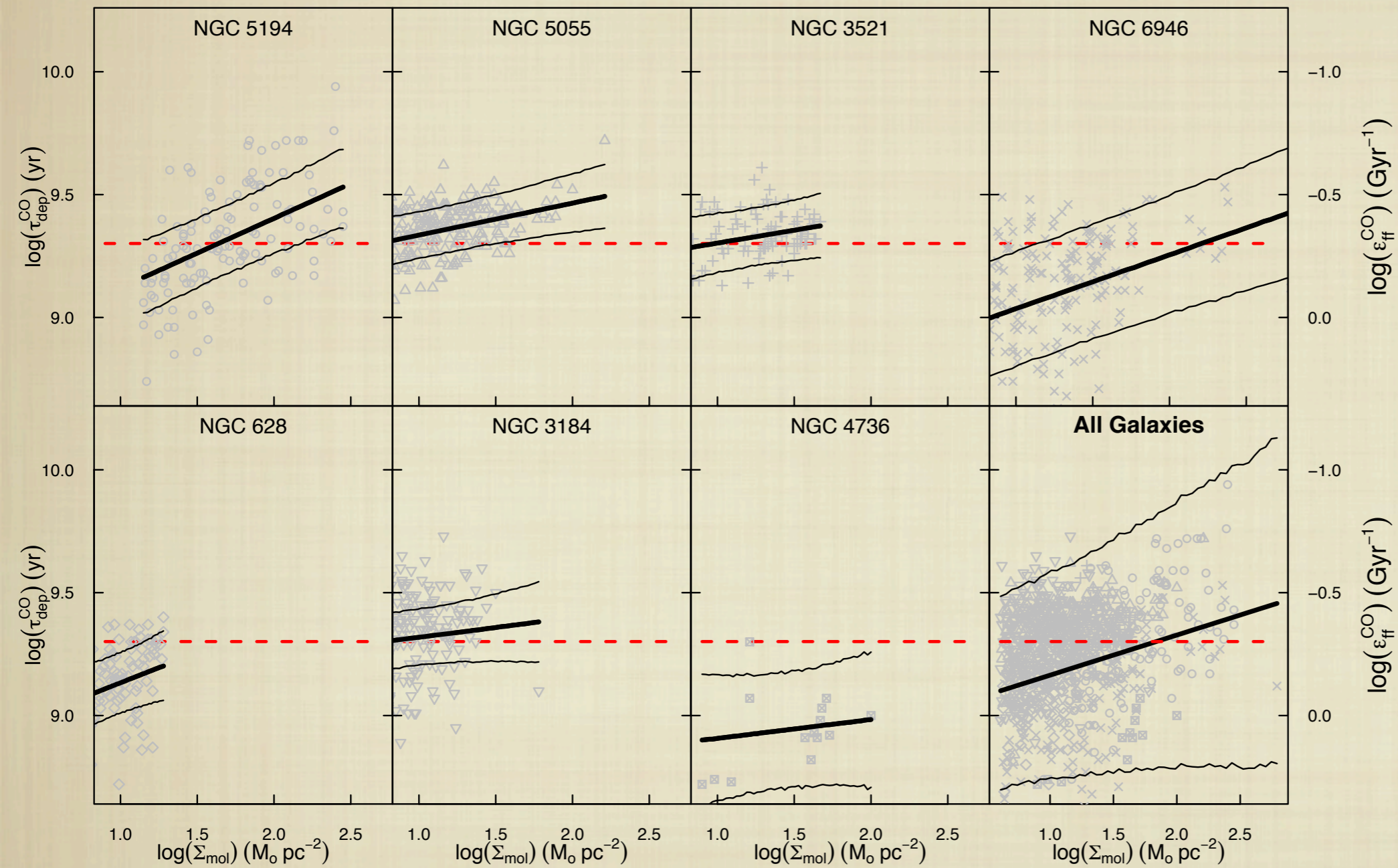


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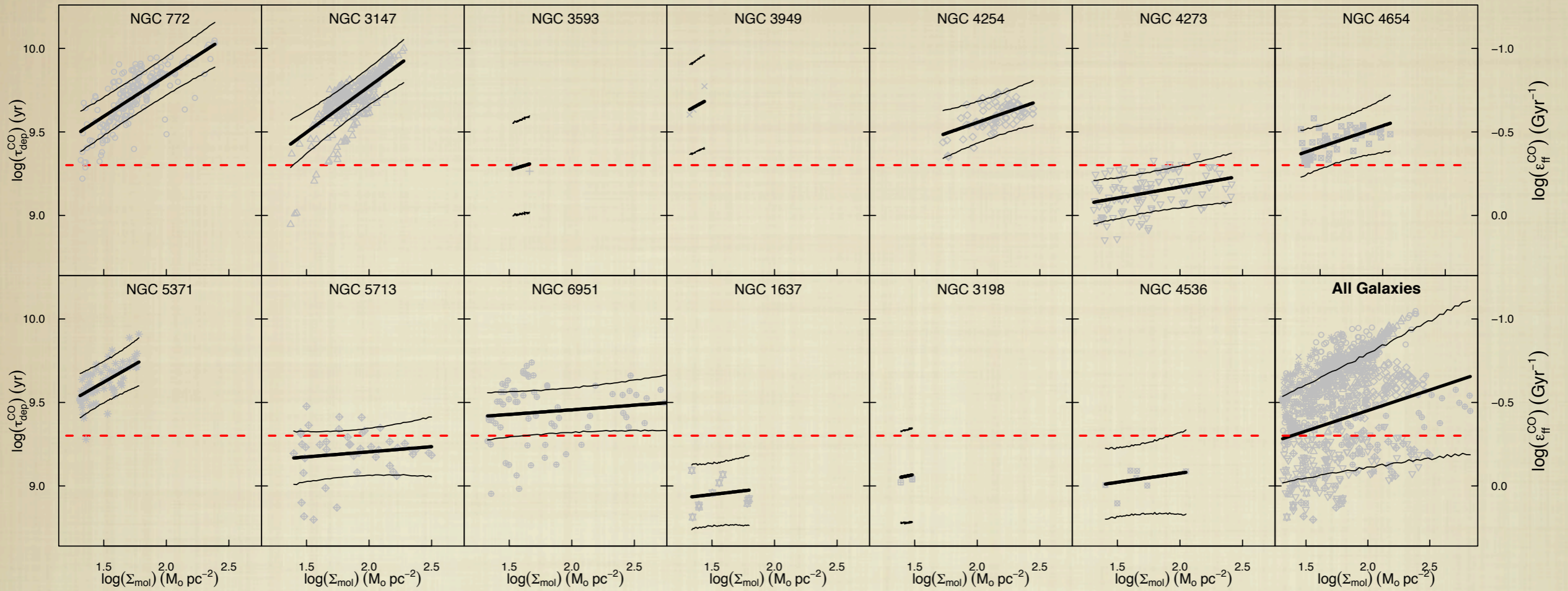


DEPLETION TIMES $\tau_{\text{DEP}} = \Sigma_{\text{GAS}} / \Sigma_{\text{SFR}}$



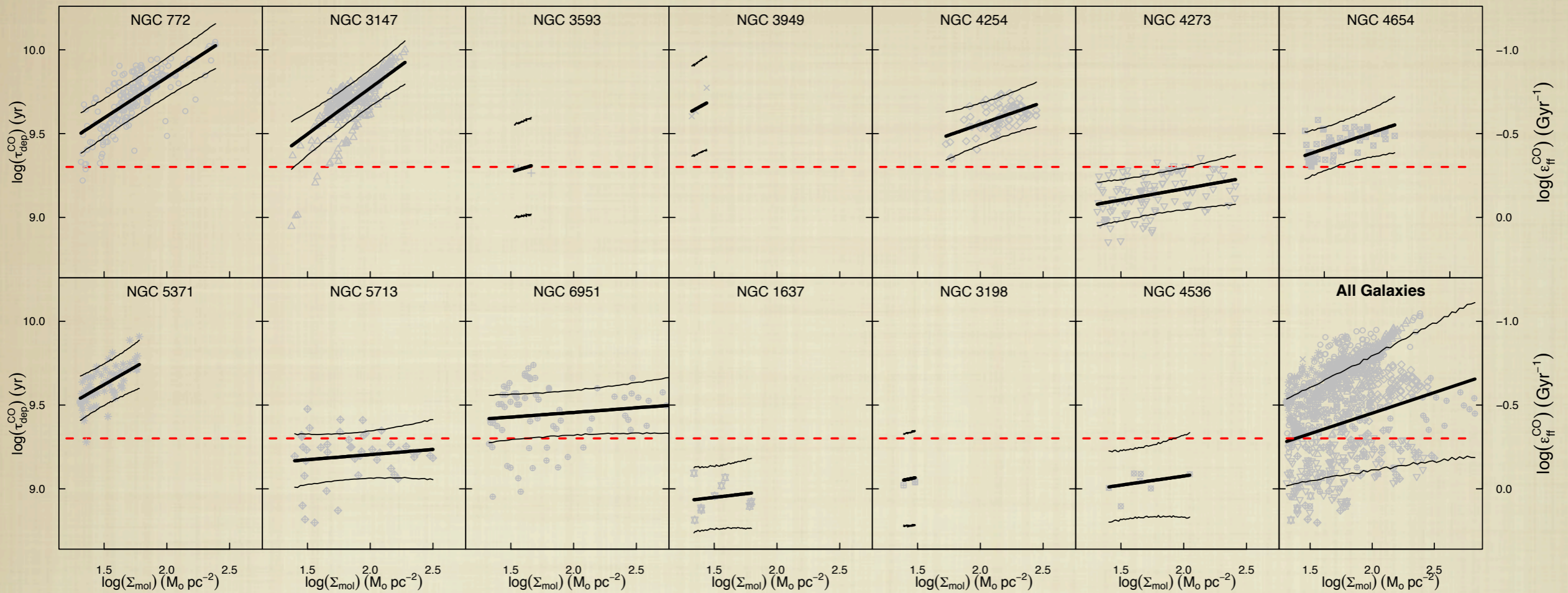
Bigiel + '08 HERACLES sub-Sample

DEPLETION TIMES $T_{\text{DEP}} = \Sigma_{\text{GAS}} / \Sigma_{\text{SFR}}$



Rahman + '12 STING Sample

DEPLETION TIMES $T_{\text{DEP}} = \Sigma_{\text{GAS}} / \Sigma_{\text{SFR}}$



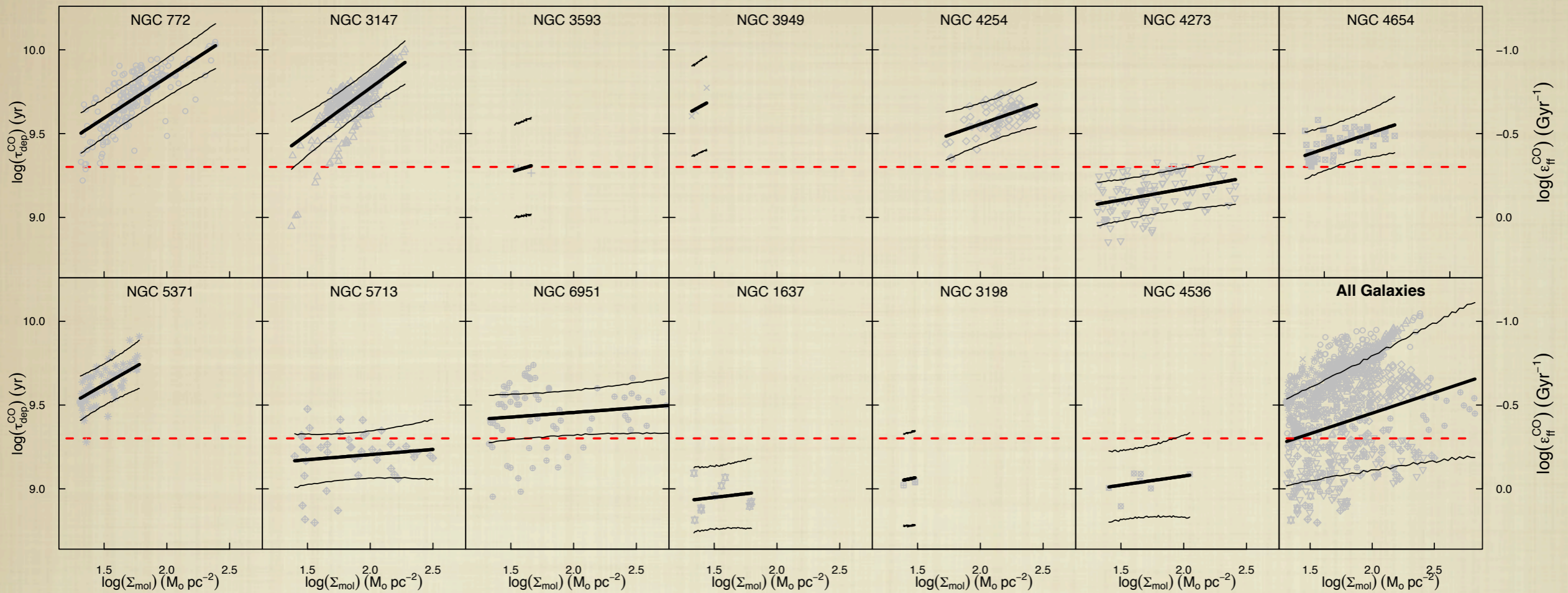
Rahman + '12 STING Sample

⇒ **No Universal depletion time**

(Bigiel+'08, Leroy+'09, Schruba+'11, Rahman+'12, etc...)

⇒ **$t_{\text{dep}} = 2 \text{ Gyr}$ is not a representative timescale**

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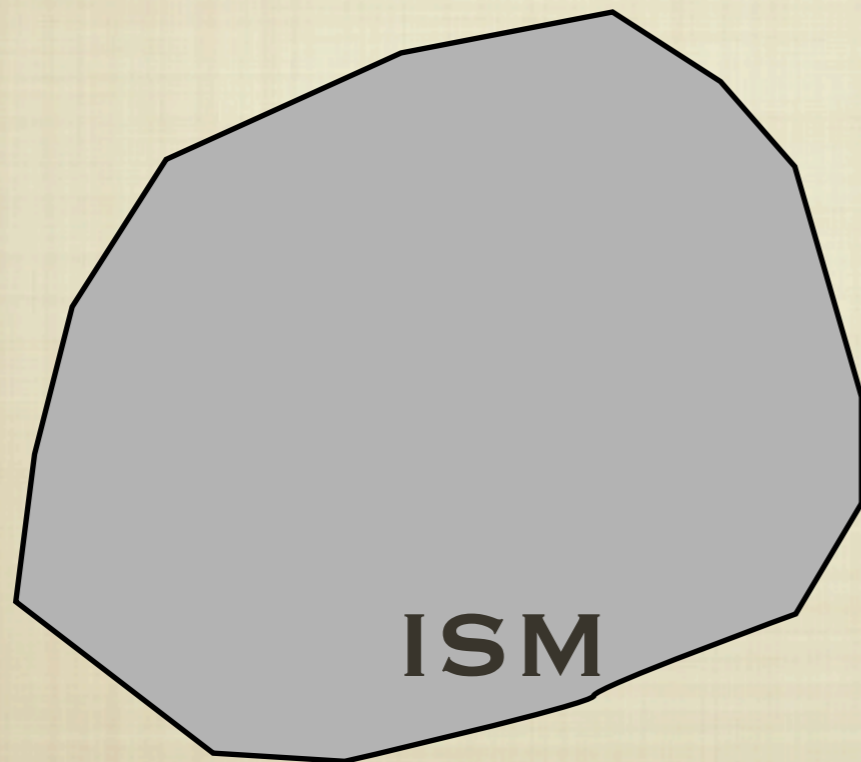
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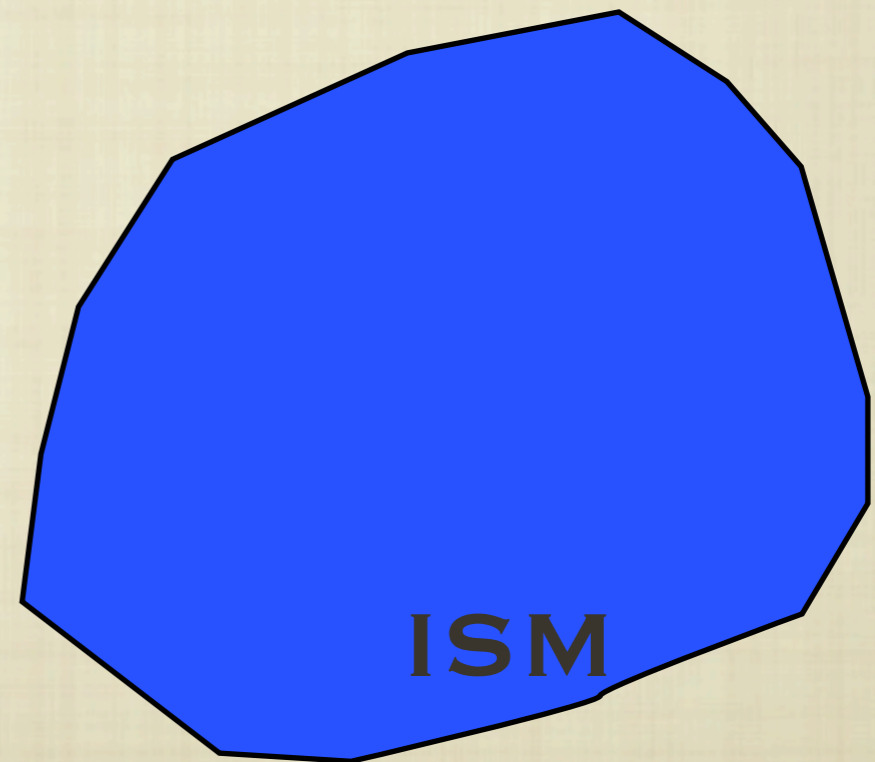
⇒ **τ_{dep} increases with Σ_{gas} , as traced by CO**

INTERPRETING THE KS SLOPE

- No evidence for an “Universal” KS slope: magnetic fields, stellar content, metallicity, molecular gas fraction all affect SF properties of given galaxy
- Significant evidence for sub-linear relationship, for most individual galaxies and for both ensembles: CO tracing some gas that is not associated with star formation?



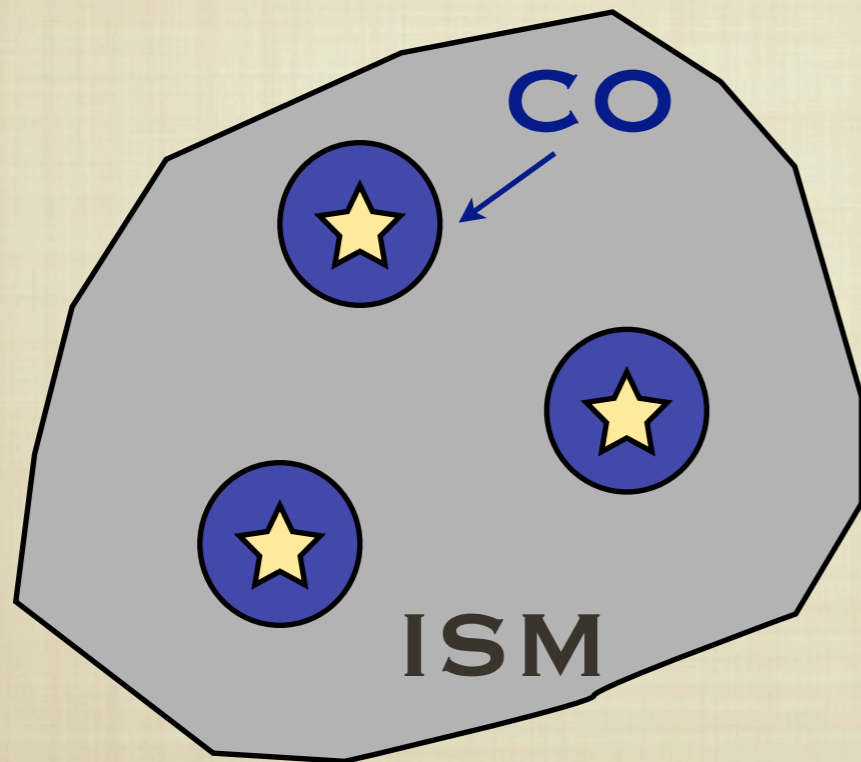
OUTER DISK OR LOW
METALLICITY



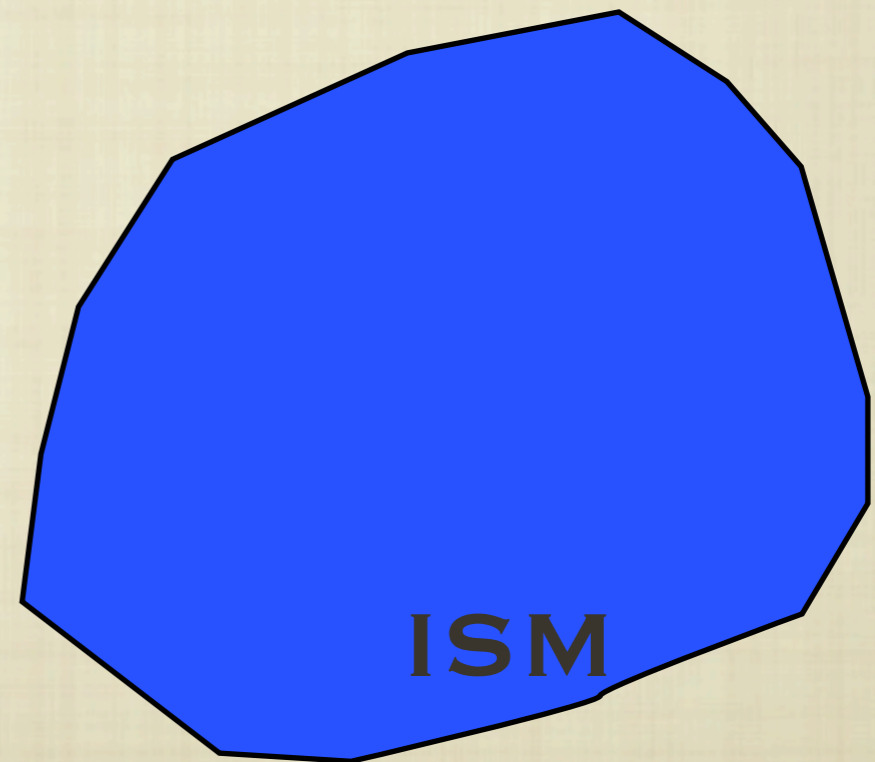
INNER DISK OR HIGH
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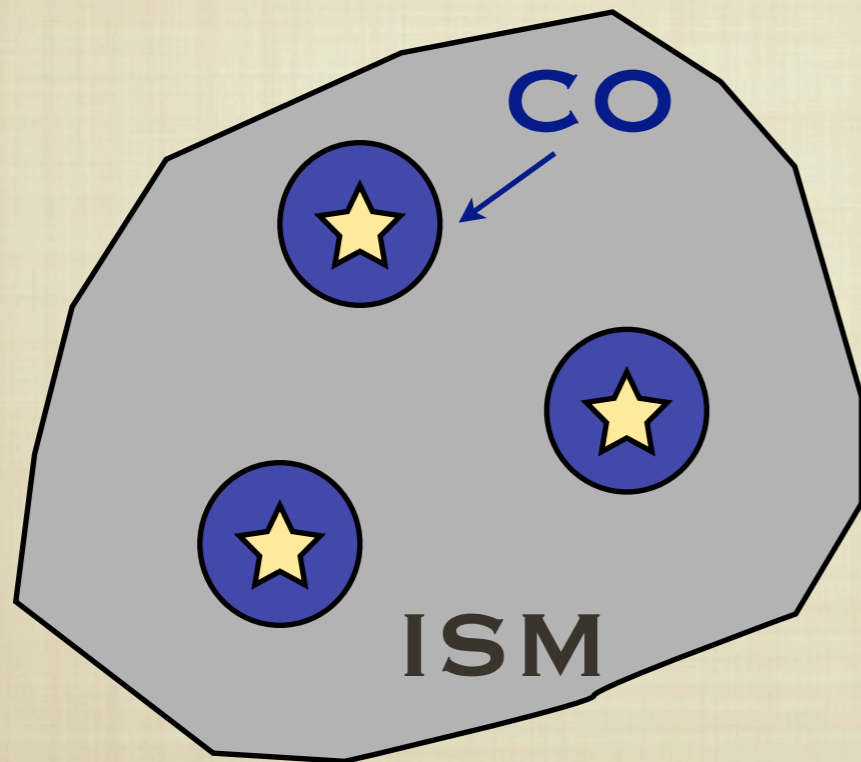
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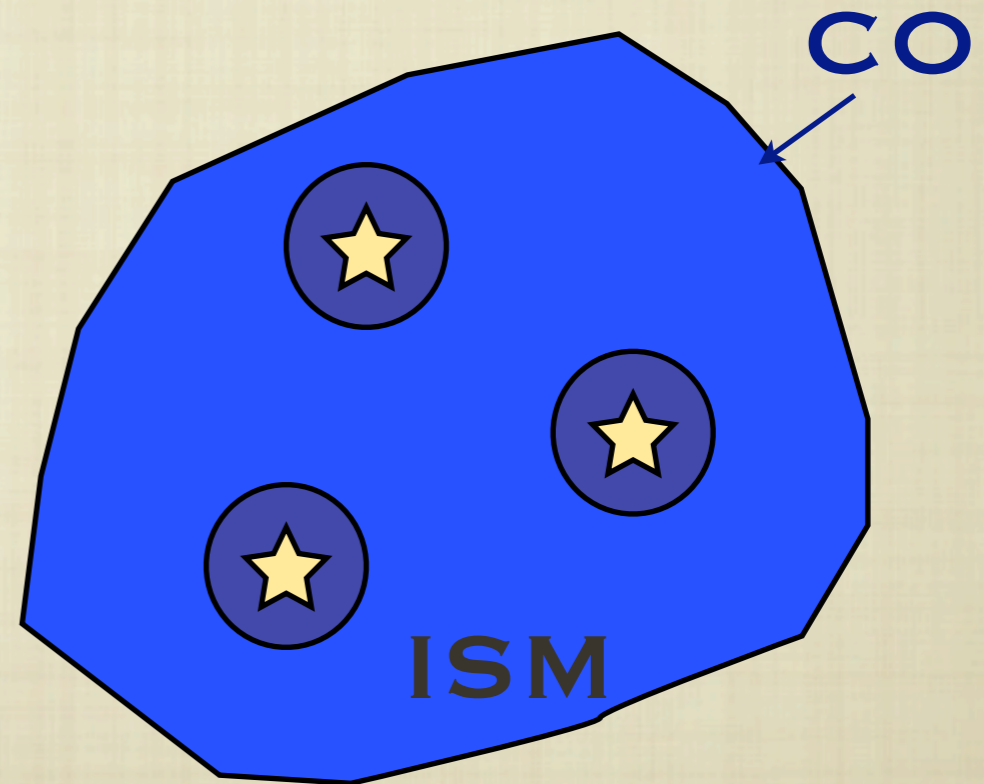
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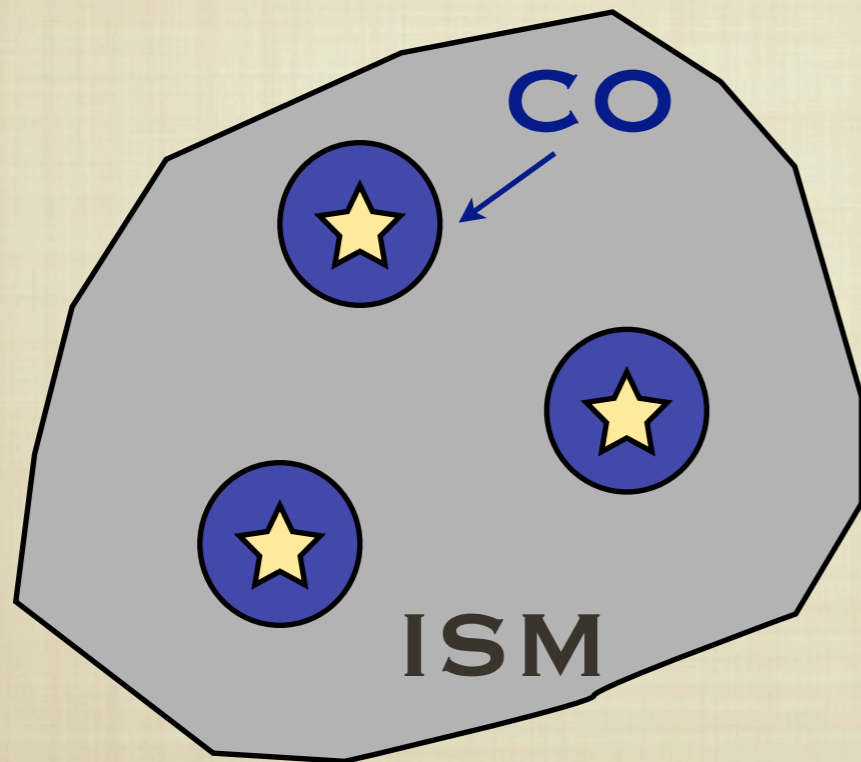
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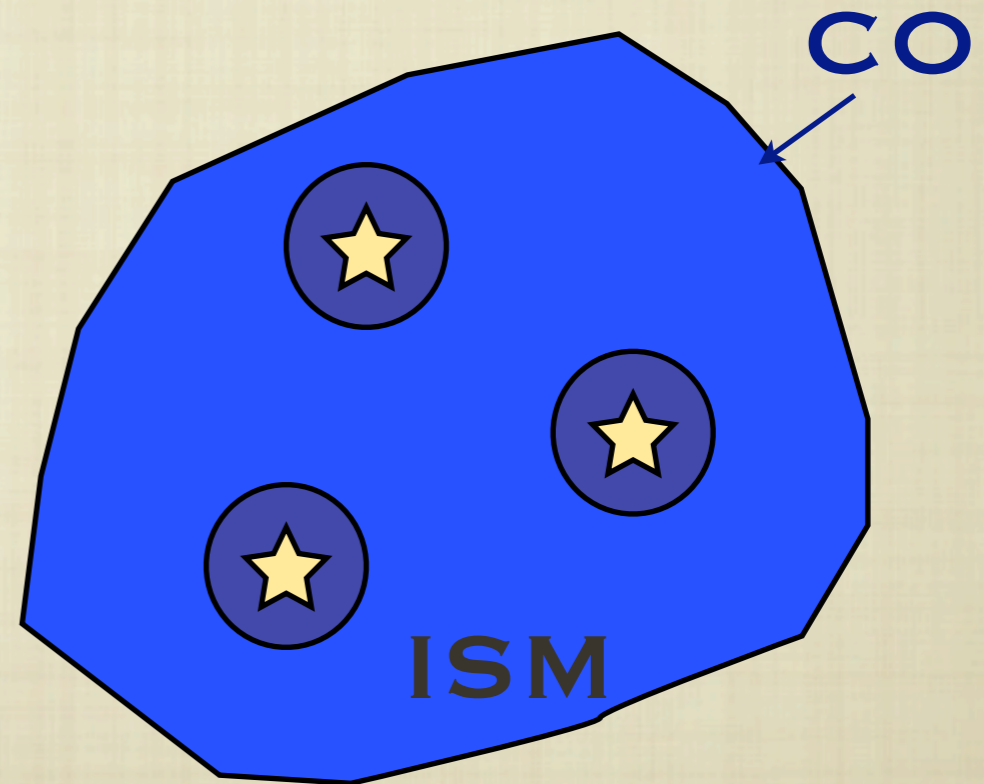
INNER DISK OR HIGH
METALLICITY

INTERPRETING THE KS SLOPE

- M51 the most sublinear slope (0.72) in Bigiel + '08 sample (see also Blanc + '09): diffuse CO emission? (PAWS, Pety, Hughes, Meidt, Schinnerer)



OUTER DISK OR LOW
METALLICITY



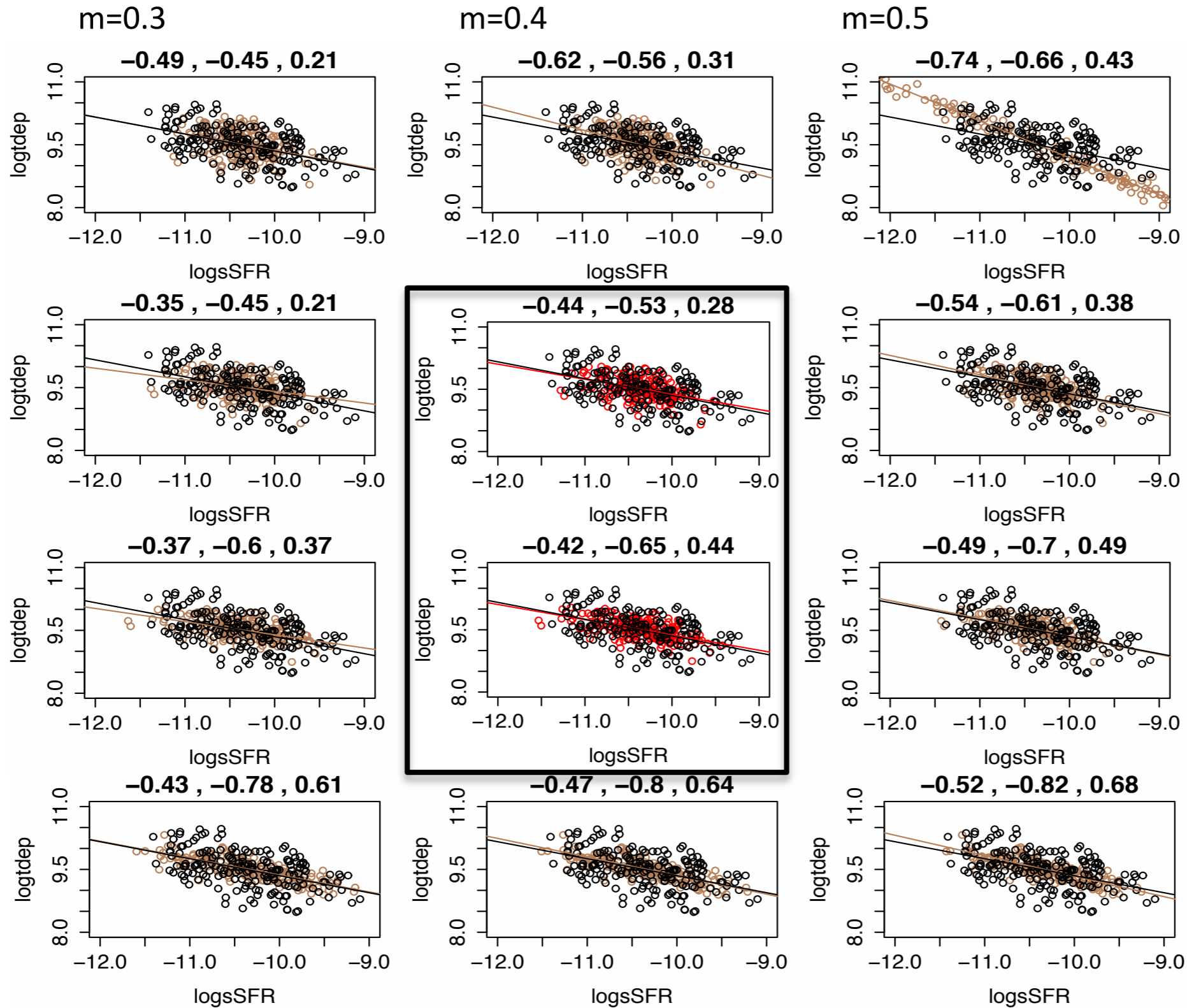
INNER DISK OR HIGH
METALLICITY

CAVEATS

- Diffuse emission? (Kennicutt + '07, Liu + '11)
- Conversion factors...?
 - e.g. X factor, old stellar population, IMF...?
 - Correlations b/w parameters (Shetty, Kelly, Bigiel '13)?
- slope and scatter depend on scale
 - Schruba + '11, Kruijssen & Longmore In Prep
- Single dish for STING (checked NGC 4254 in Shetty + '13b)

COLD GASS (Saintonge + '12)

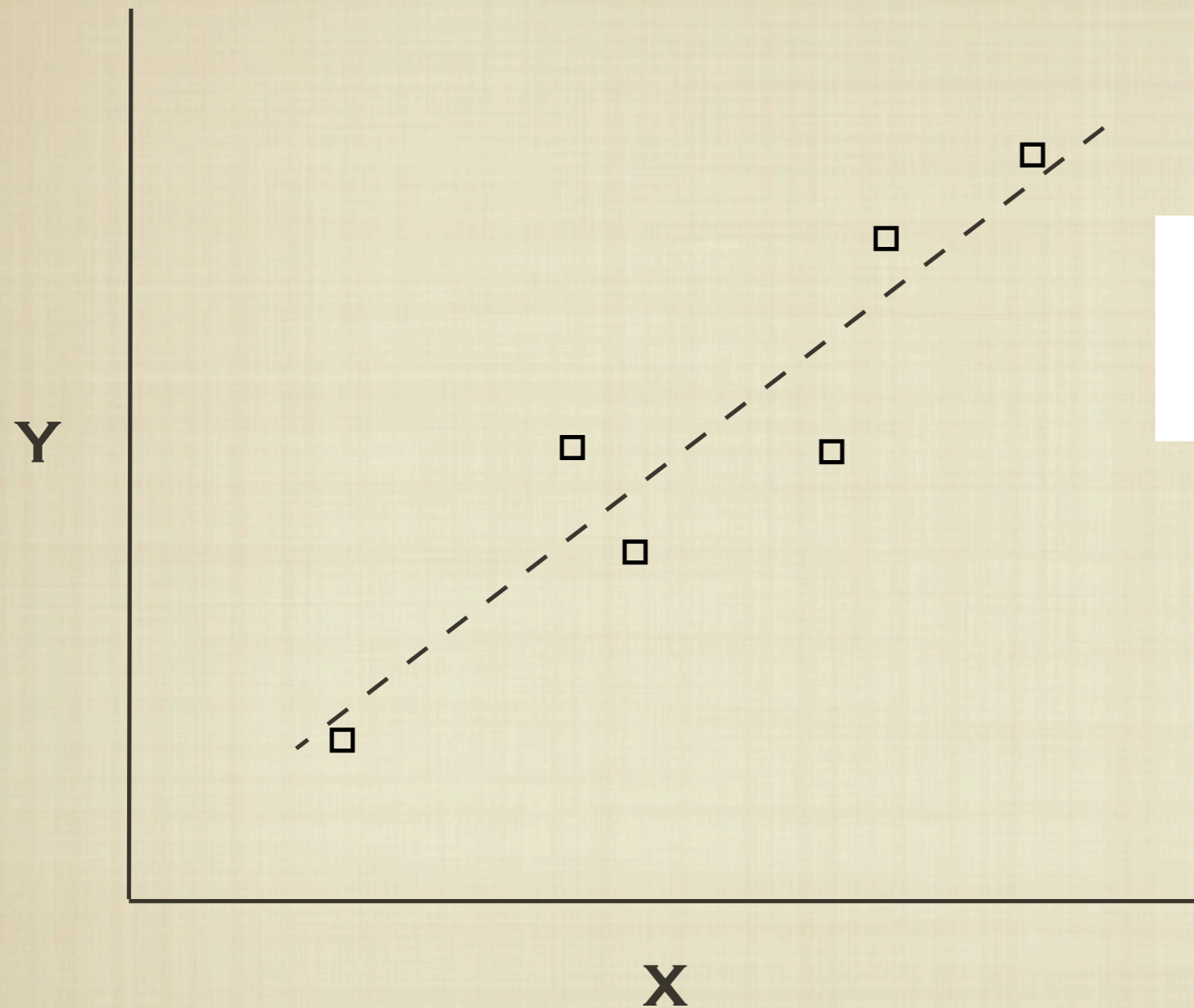
Hopkins + '13,
In Prep
(see poster)



SUMMARY

- Need to fit data carefully
 - noise leads to biases
 - bisector provides inaccurate estimates
 - hierarchical fits for hierarchical data
 - Monte Carlo (& Bayesian) provide accurate estimates
- 2 datasets strongly indicate a mean sub-linear KS relationship, and sub-linear slopes for most galaxies
- No “Universal” KS law
 - $t_{\text{dep}} = 2 \text{ Gyr}$ not representative
 - t_{dep} increases with CO traced surface density
 - indication of diffuse CO component?

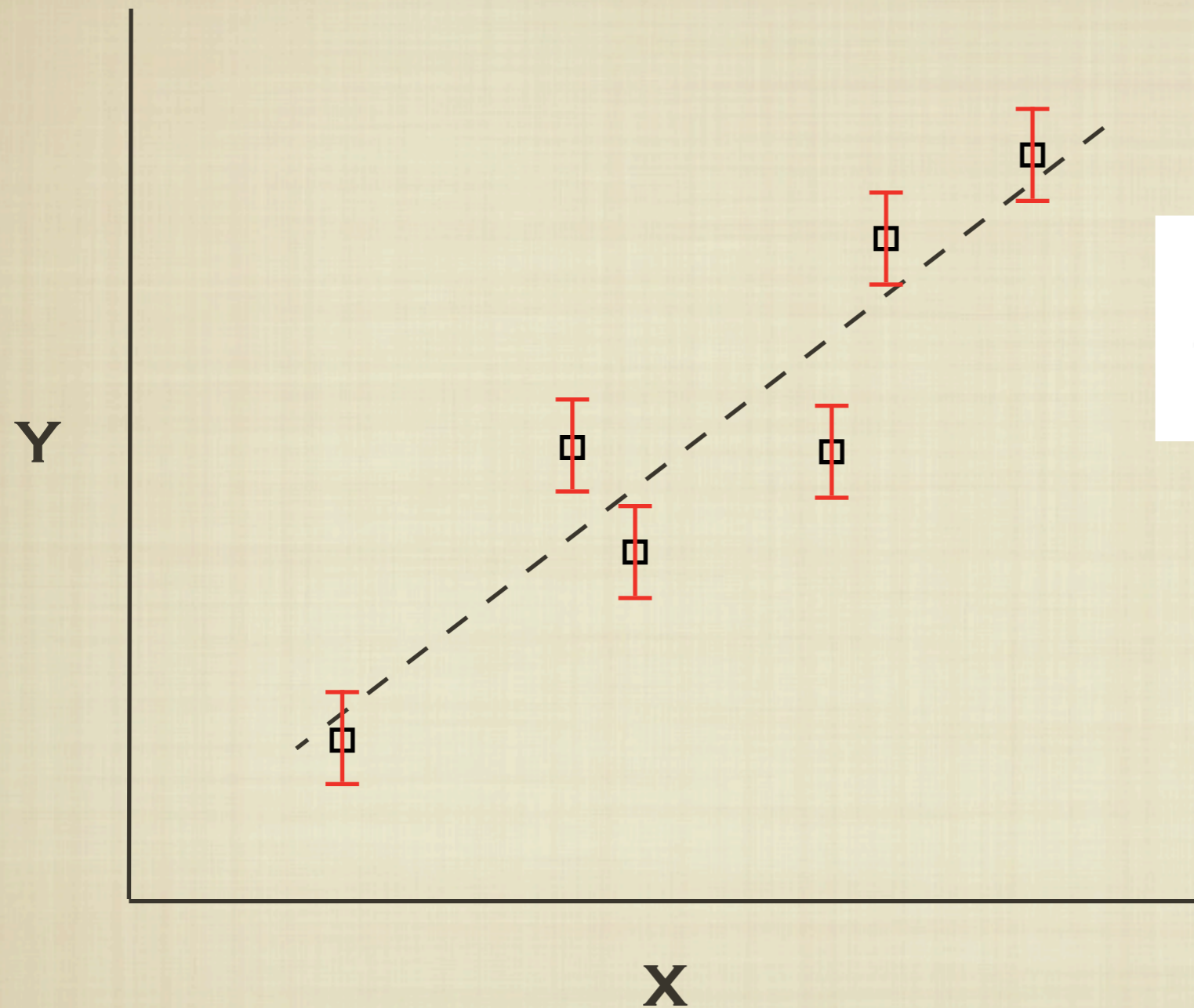
MONTE CARLO METHODS



$$y_i = A + Nx_i$$

$$\chi^2 = \sum_{i=1}^m \left[\frac{\hat{y}_i - y_i(A, N)}{\sigma_y} \right]^2$$

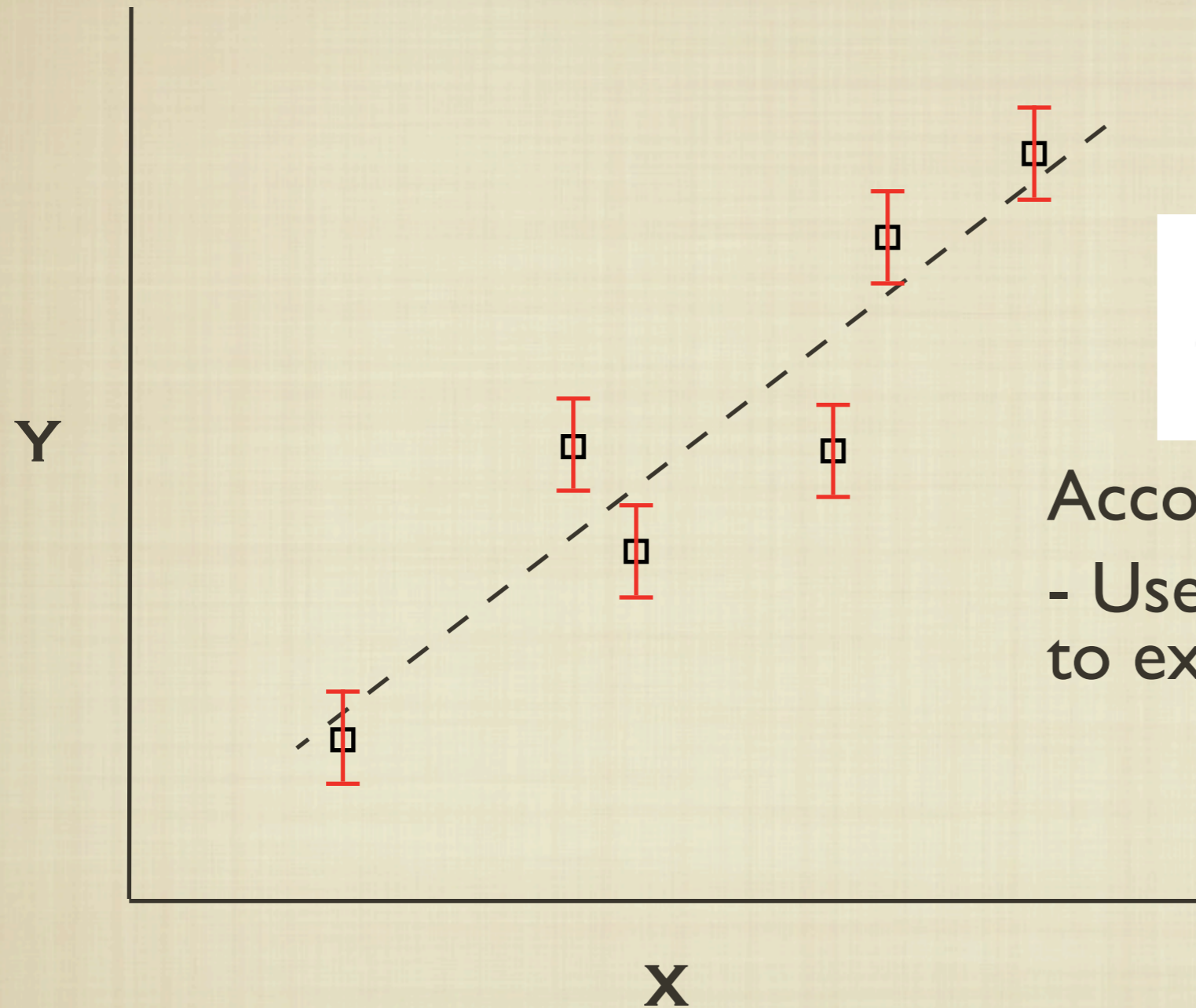
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MONTE CARLO METHODS

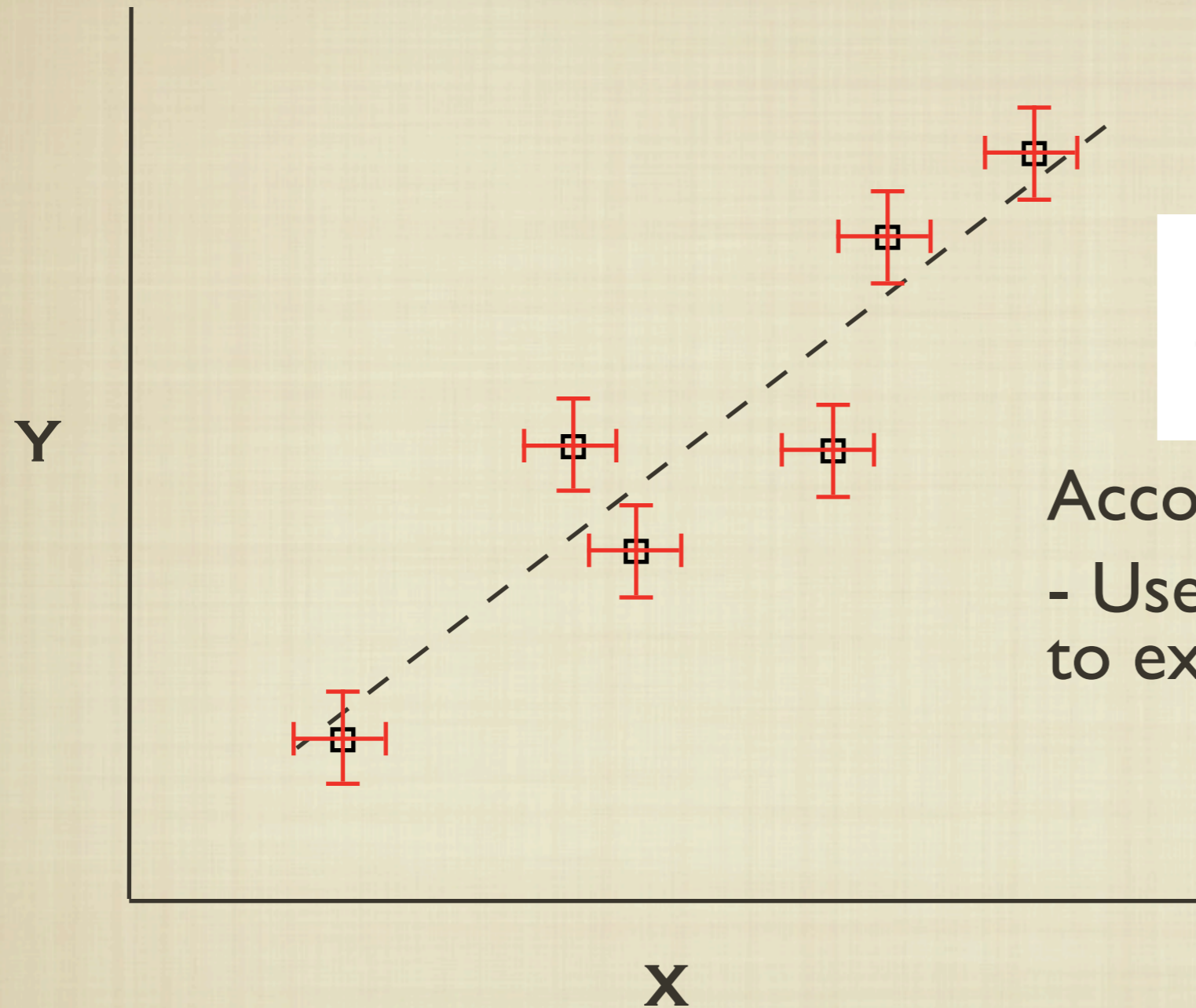


$$y_i = A + Nx_i$$

$$\chi^2 = \sum_{i=1}^m \left[\frac{\hat{y}_i - y_i(A, N)}{\sigma_y} \right]^2$$

Account for Uncertainty in y ?
- Use Monte Carlo simulations to explore possible y values.

MONTE CARLO METHODS



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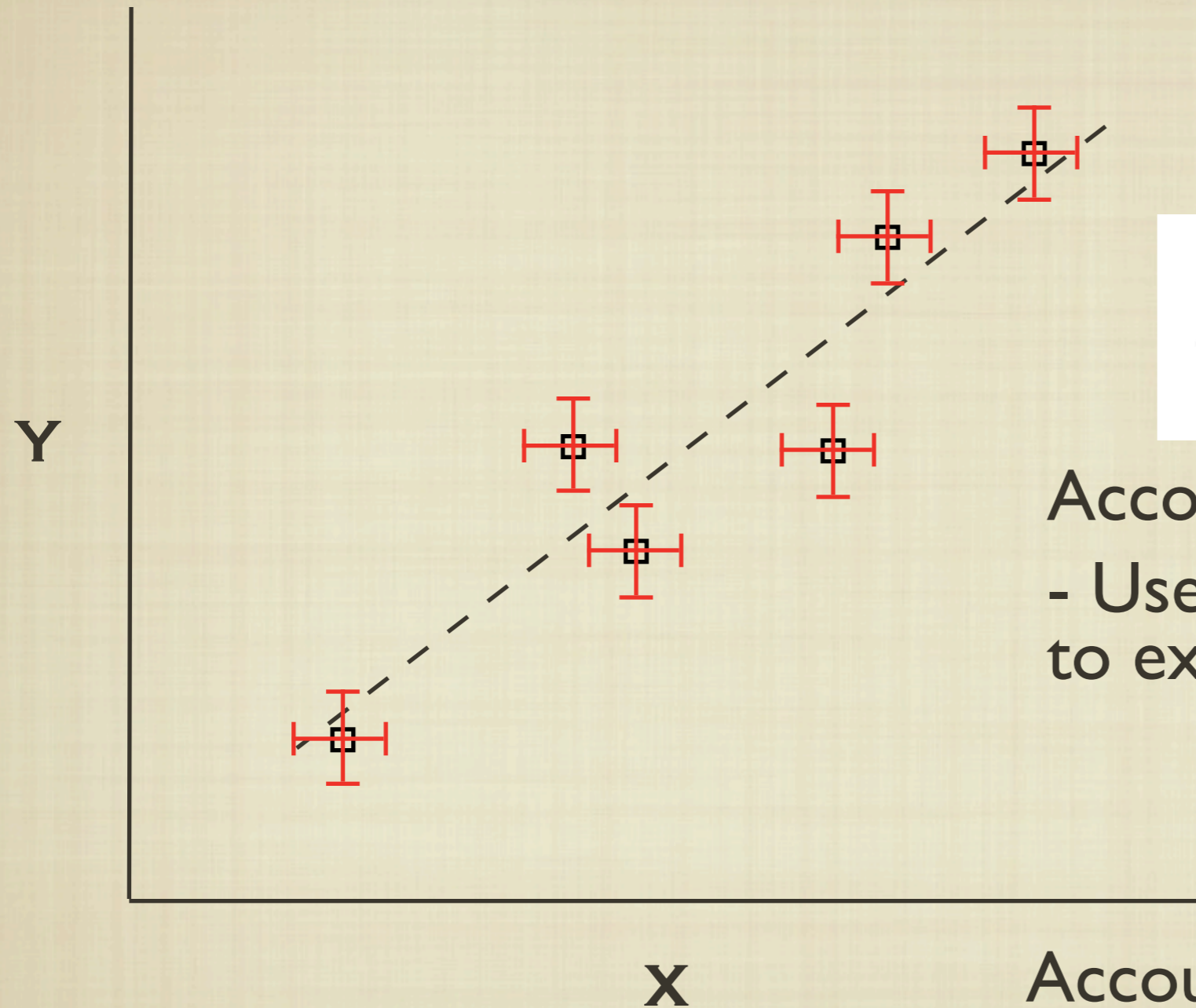
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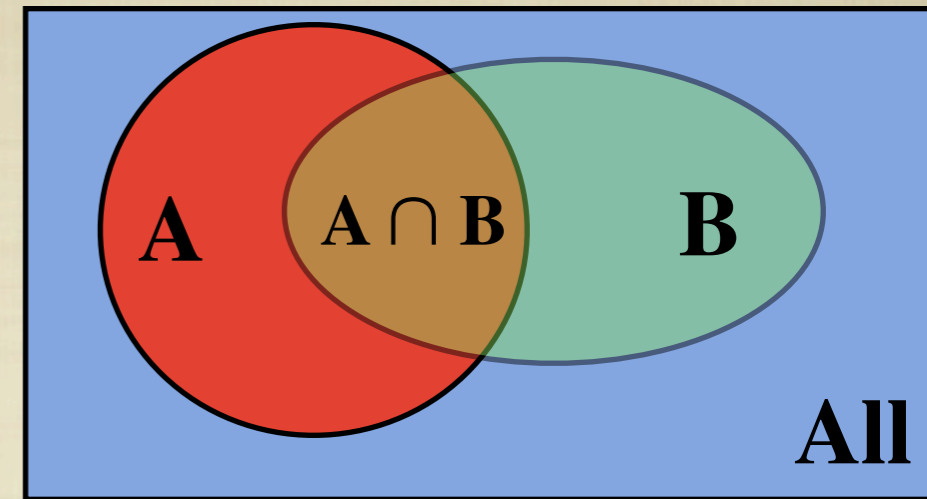


Account for Uncertainty in y ?
- Use Monte Carlo simulations to explore possible y values.

Account for Uncertainty in x ?
- Monte Carlo simulations explore uncertainties in both x and y values.

BAYESIAN INFERENCE

- Bayes' Theorem: $P(B|A) = \frac{P(A|B)P(B)}{P(A)}$



- Consider data (D), model (M): $P(M|D)$

- Model M can set the range of plausible parameters:
 $M = (A_j, N_j, \mu_A, \mu_N \dots)$

- Modeling should also account for uncertainties in data

$$\log \hat{\Sigma}_{\text{SFR}} = \log \Sigma_{\text{SFR}} + \epsilon_{\text{SFR}}$$
$$\log \hat{\Sigma}_{\text{mol}} = \log \Sigma_{\text{mol}} + \epsilon_{\text{mol}}$$

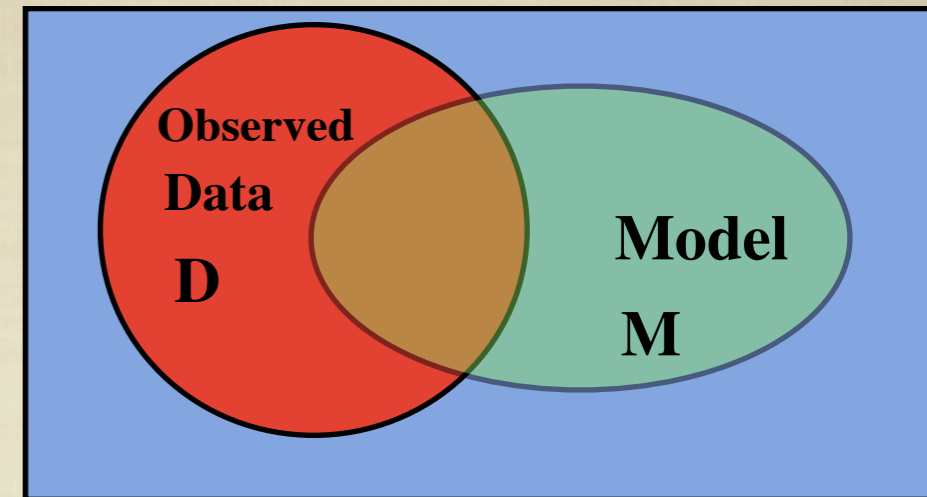
- Relate M and D by KS equation & uncertainties

$$\log(\Sigma_{\text{SFR}}) = A + N \log(\Sigma_{\text{mol}}) + \epsilon_{\text{scat}}$$

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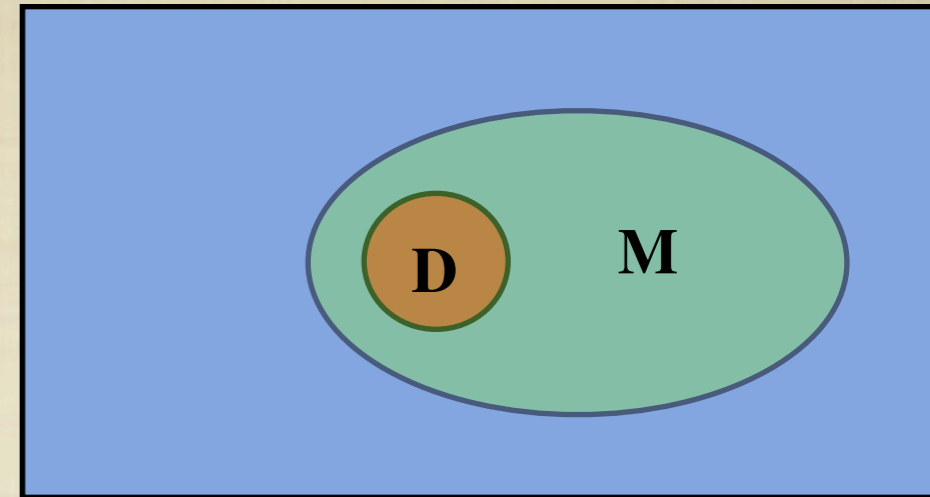
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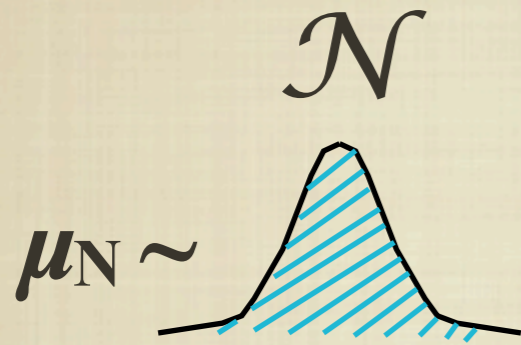
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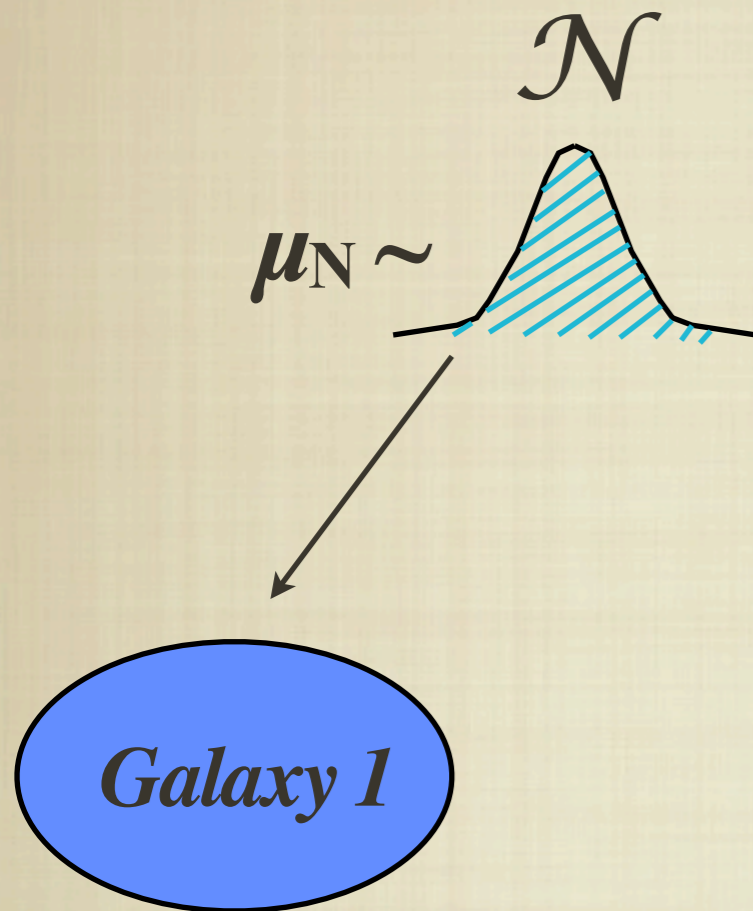
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HIERARCHICAL MODELING



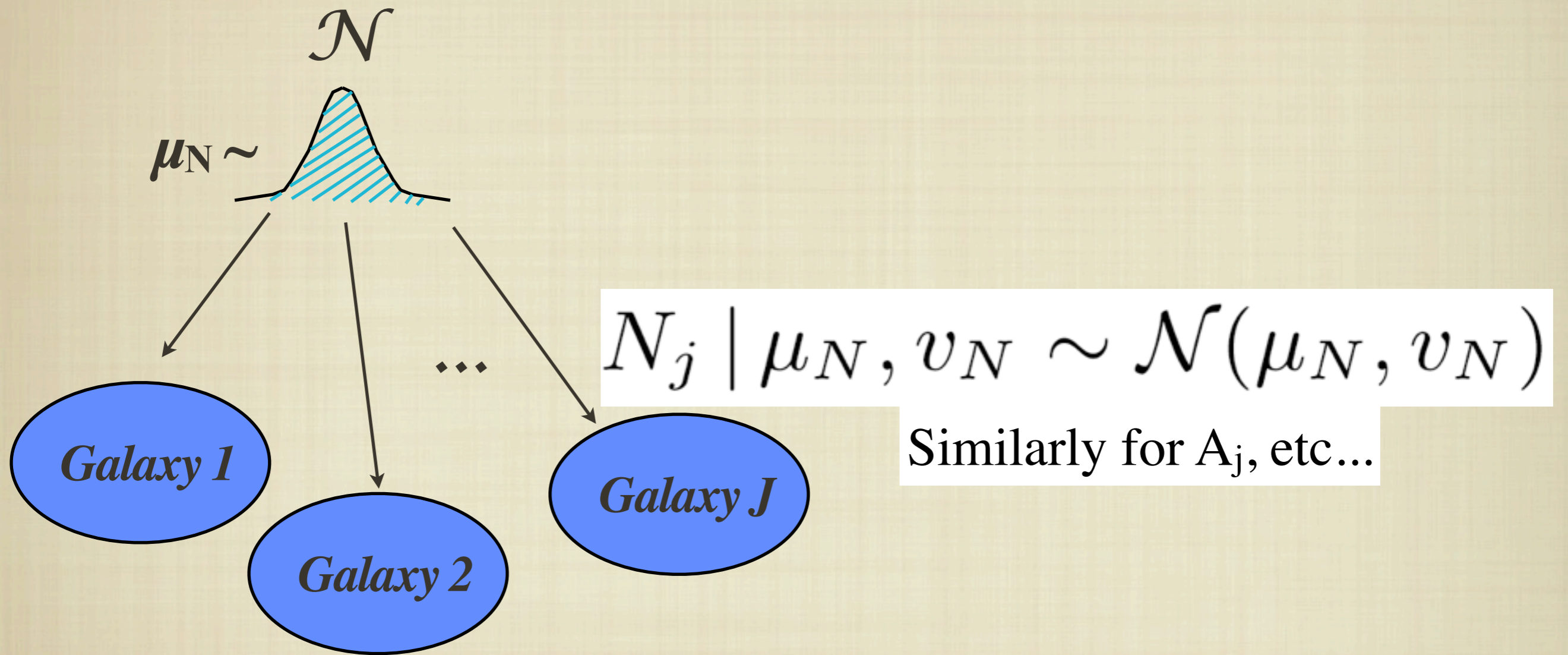
HIERARCHICAL MODELING



$$N_j \mid \mu_N, v_N \sim \mathcal{N}(\mu_N, v_N)$$

Similarly for A_j , etc...

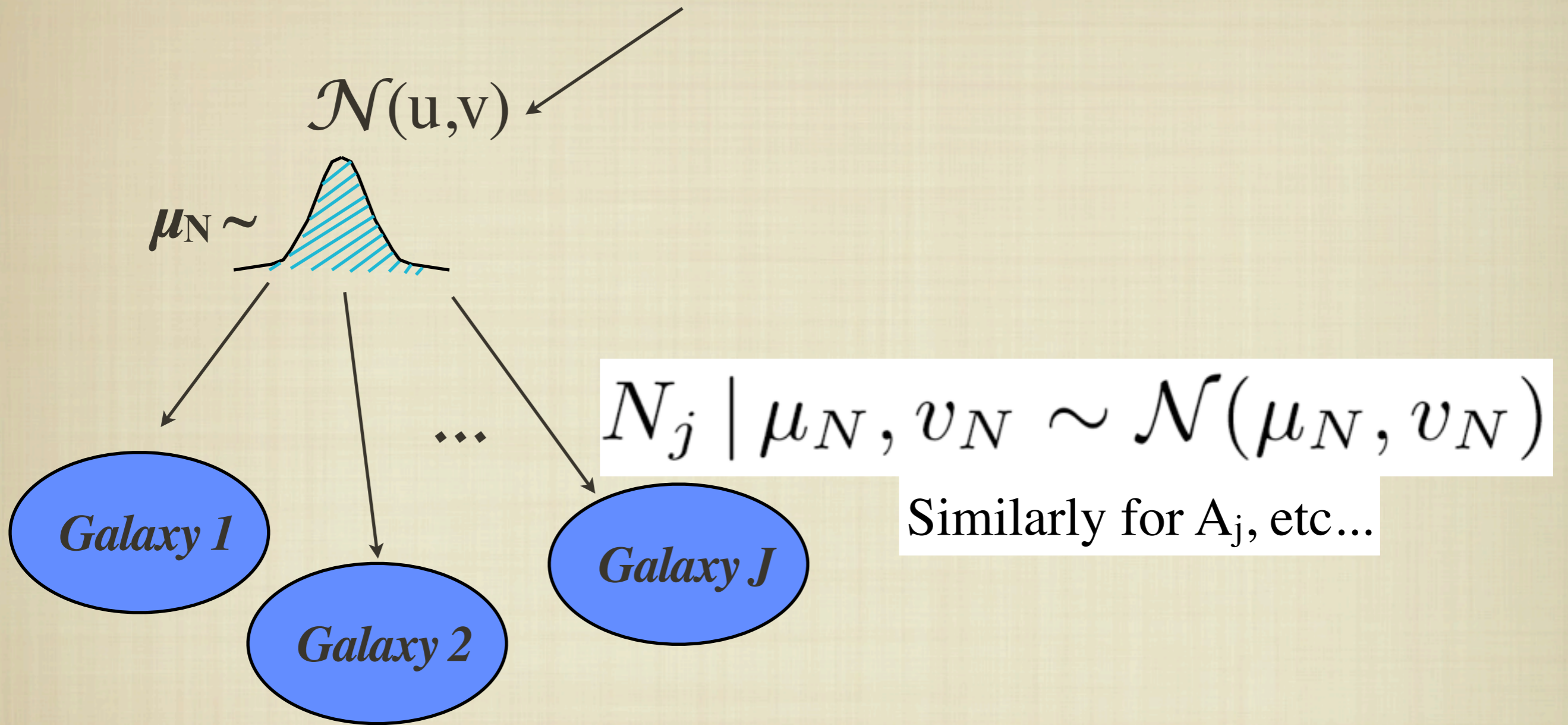
HIERARCHICAL MODELING



$$\log \Sigma_{\text{SFR},ij} | A_j, N_j, \Sigma_{\text{mol},ij}, \sigma_{\text{scat},j}^2 \sim \mathcal{N}(A_j + N_j \log \Sigma_{\text{mol},ij}, \sigma_{\text{scat},j}^2)$$

HIERARCHICAL MODELING

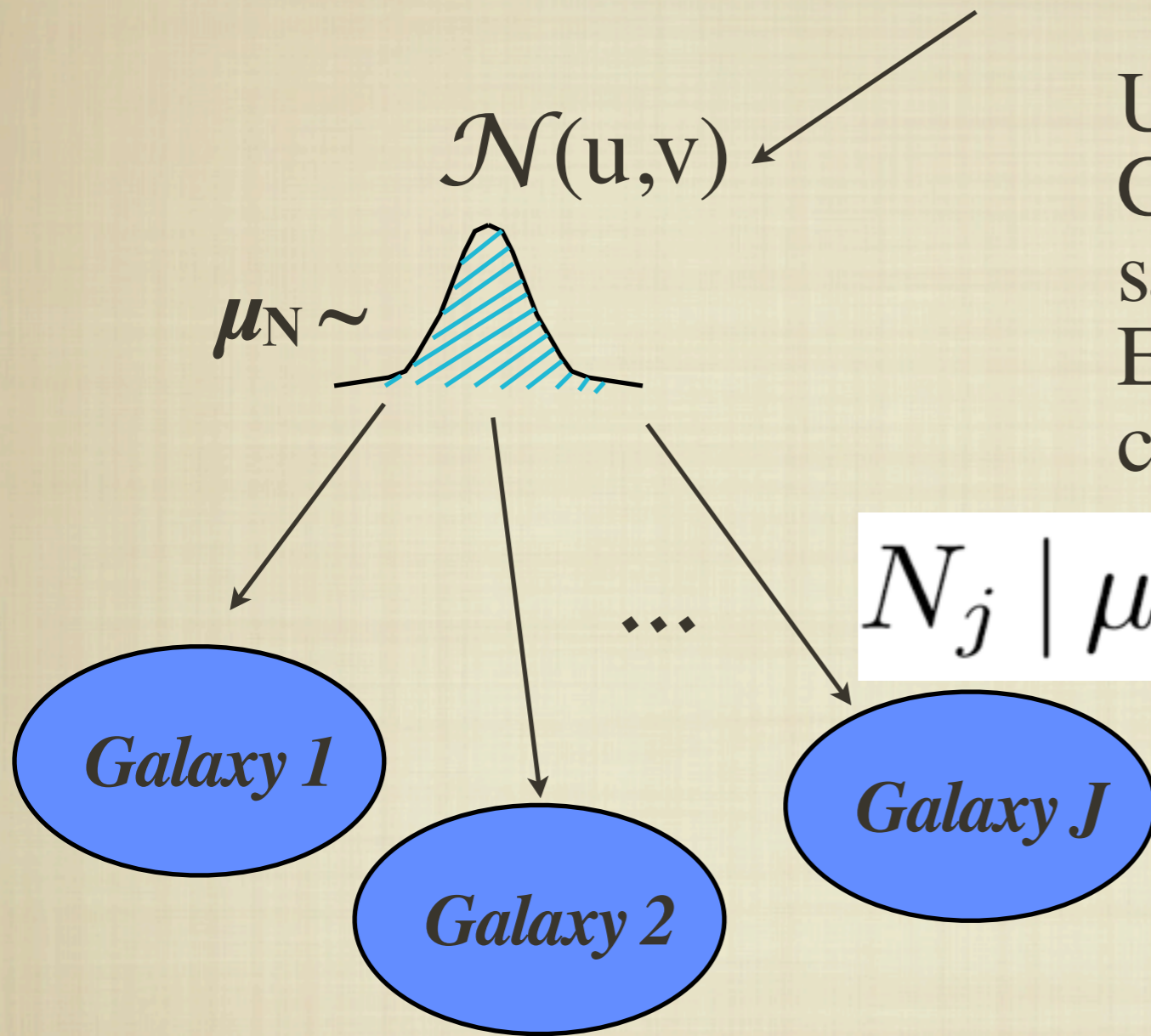
Additional level for population parameters



$$\log \Sigma_{\text{SFR},ij} | A_j, N_j, \Sigma_{\text{mol},ij}, \sigma_{\text{scat},j}^2 \sim \mathcal{N}(A_j + N_j \log \Sigma_{\text{mol},ij}, \sigma_{\text{scat},j}^2)$$

HIERARCHICAL MODELING

Additional level for population parameters



Using Markov Chain Monte Carlo (MCMC) techniques to sample all distributions. Evaluate the likelihood, and construct the posterior PDF.

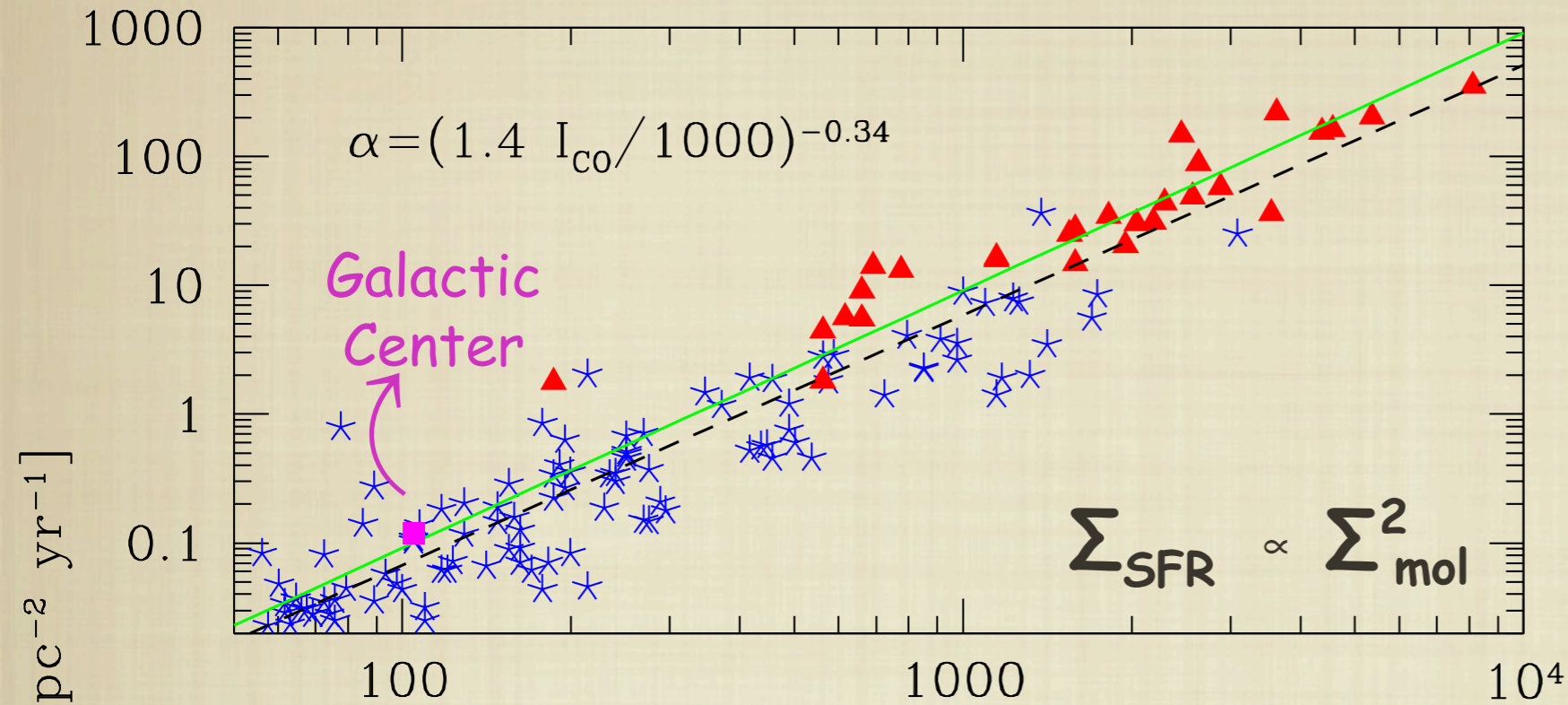
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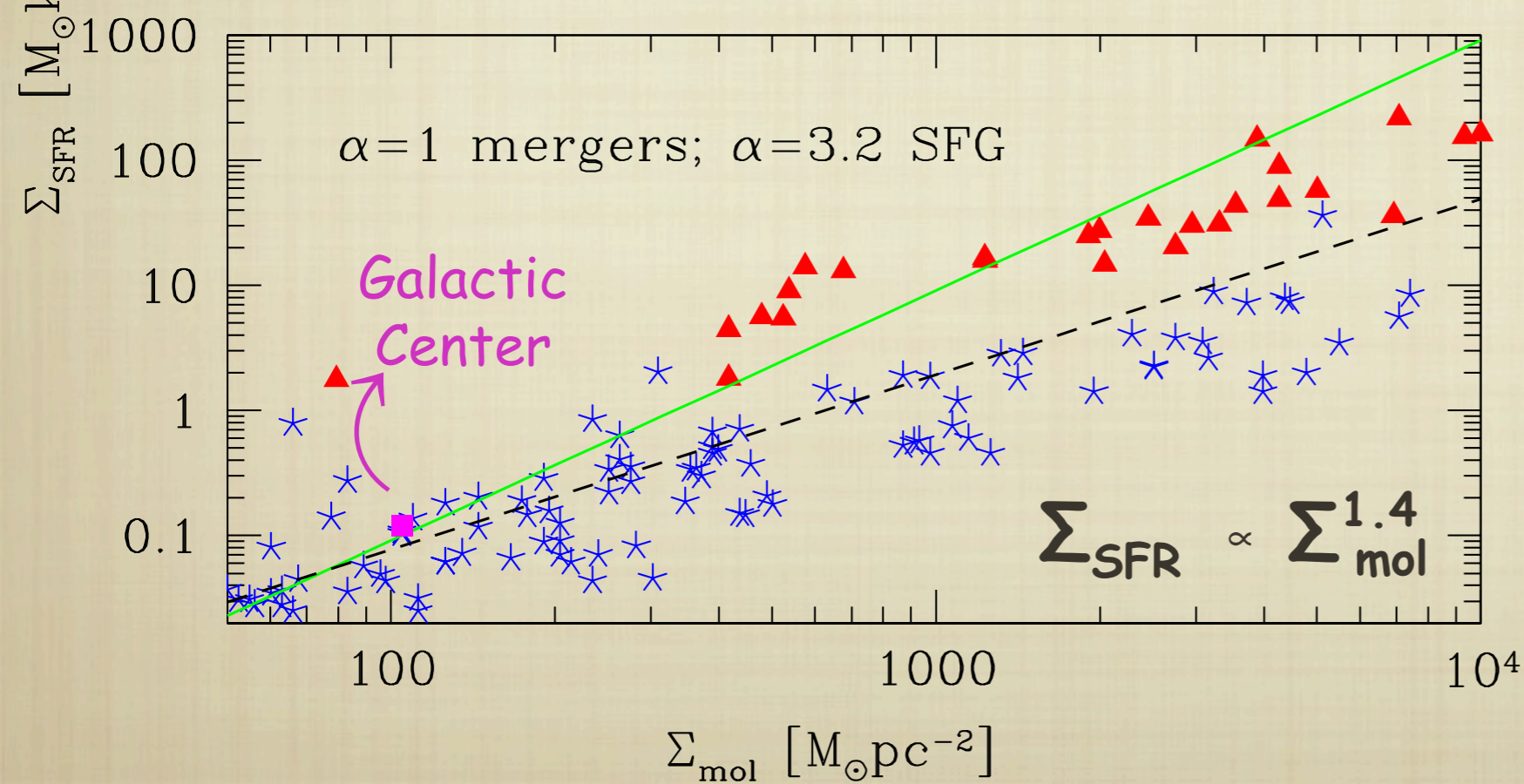
SELF-REGULATION IN OBSERVED SYSTEMS?

Ostriker & Shetty '11, Shetty & Ostriker '12



Star forming galaxies and merger systems from Genzel + '10

Galactic Center from Yusef-Zadeh + '09



Continuous X factor? (Shetty + I I a,b, Narayanan + I I, I 2)