

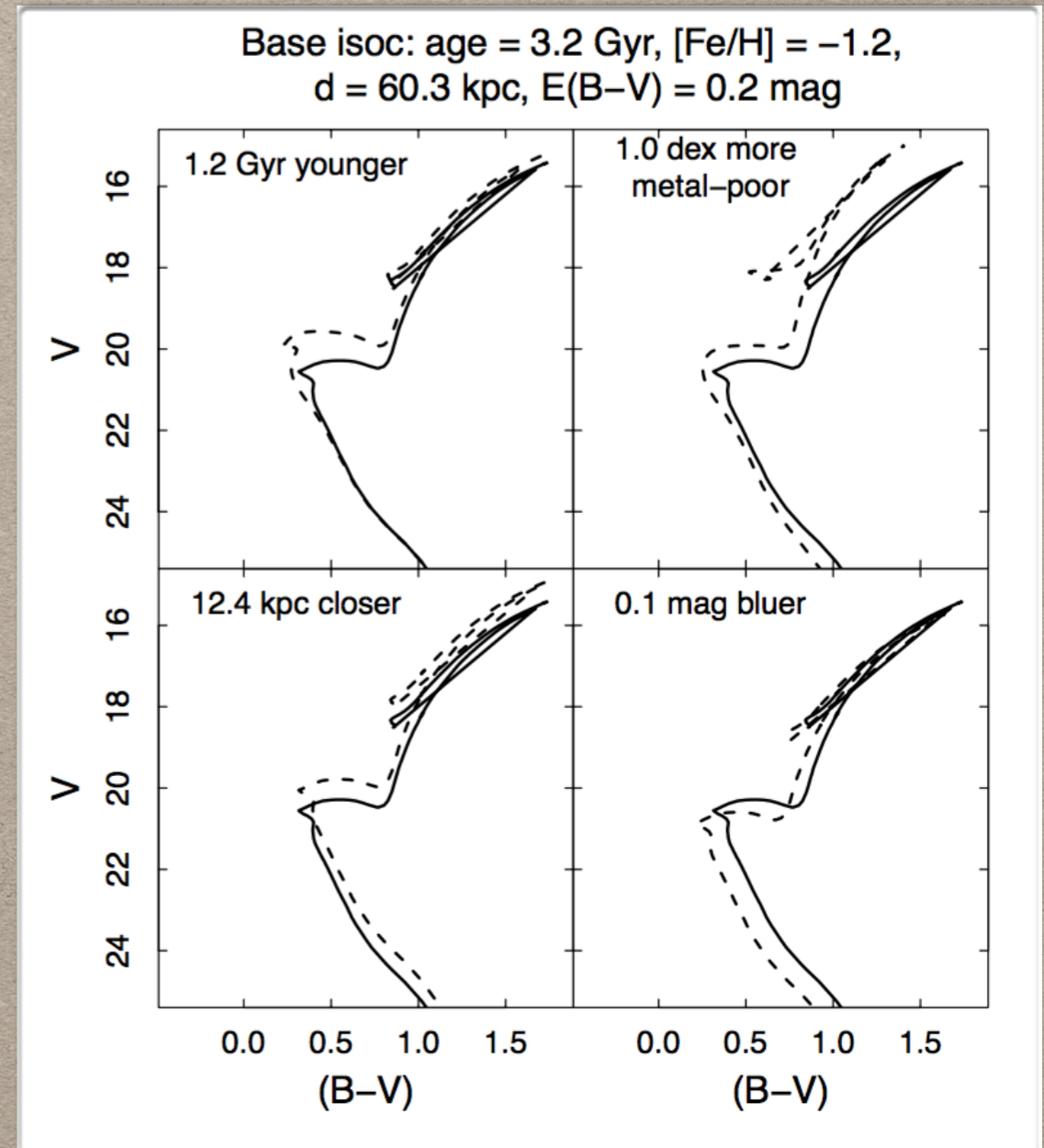
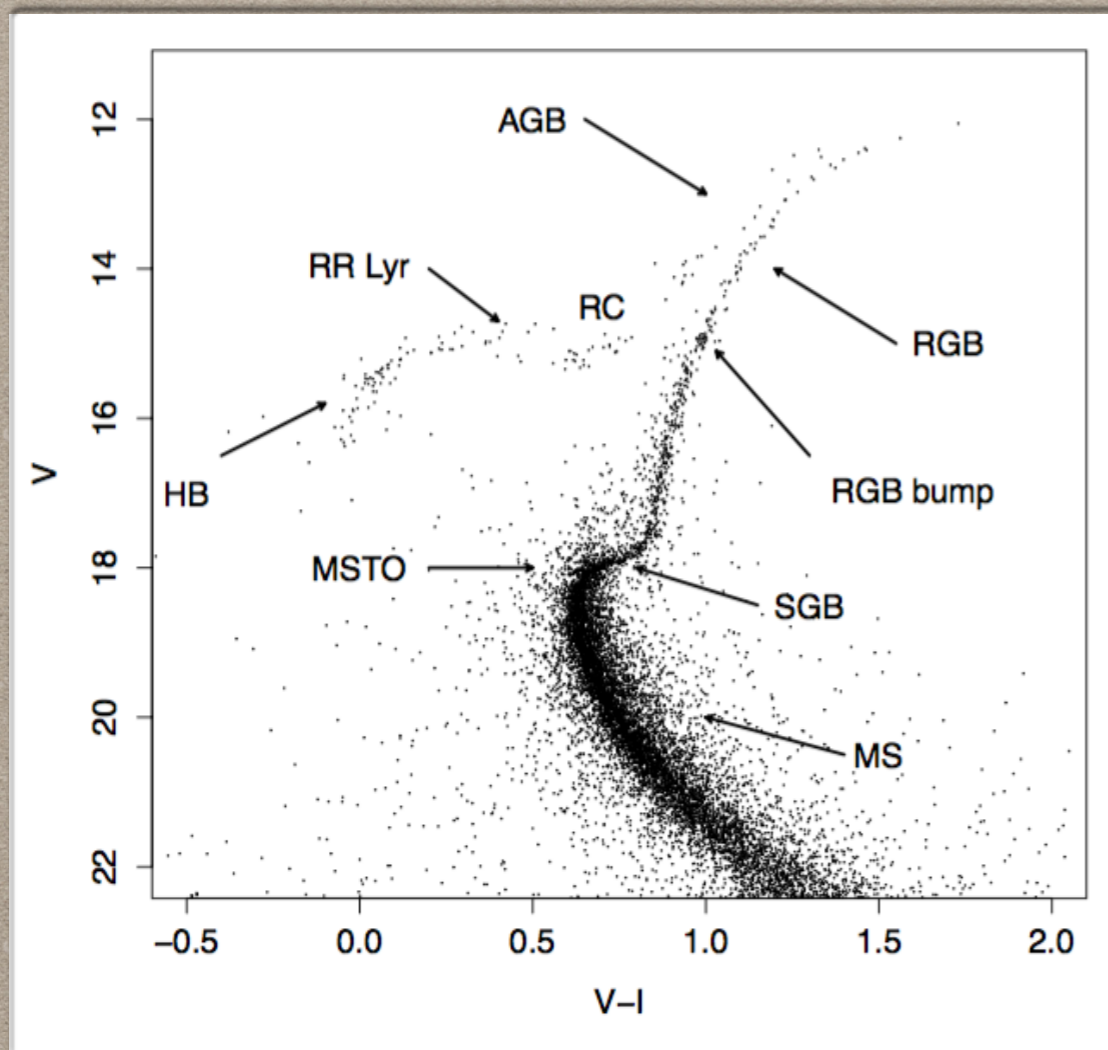
Dias et al. (2014,2016)

MAXIMUM LIKELIHOOD ESTIMATIONS I: SYNTHETIC COLOUR-MAGNITUDE DIAGRAM FITTING

BRUNO DIAS – MCMC COFFEE, ESO – OCT 5, 2016 (4.2-4.5)

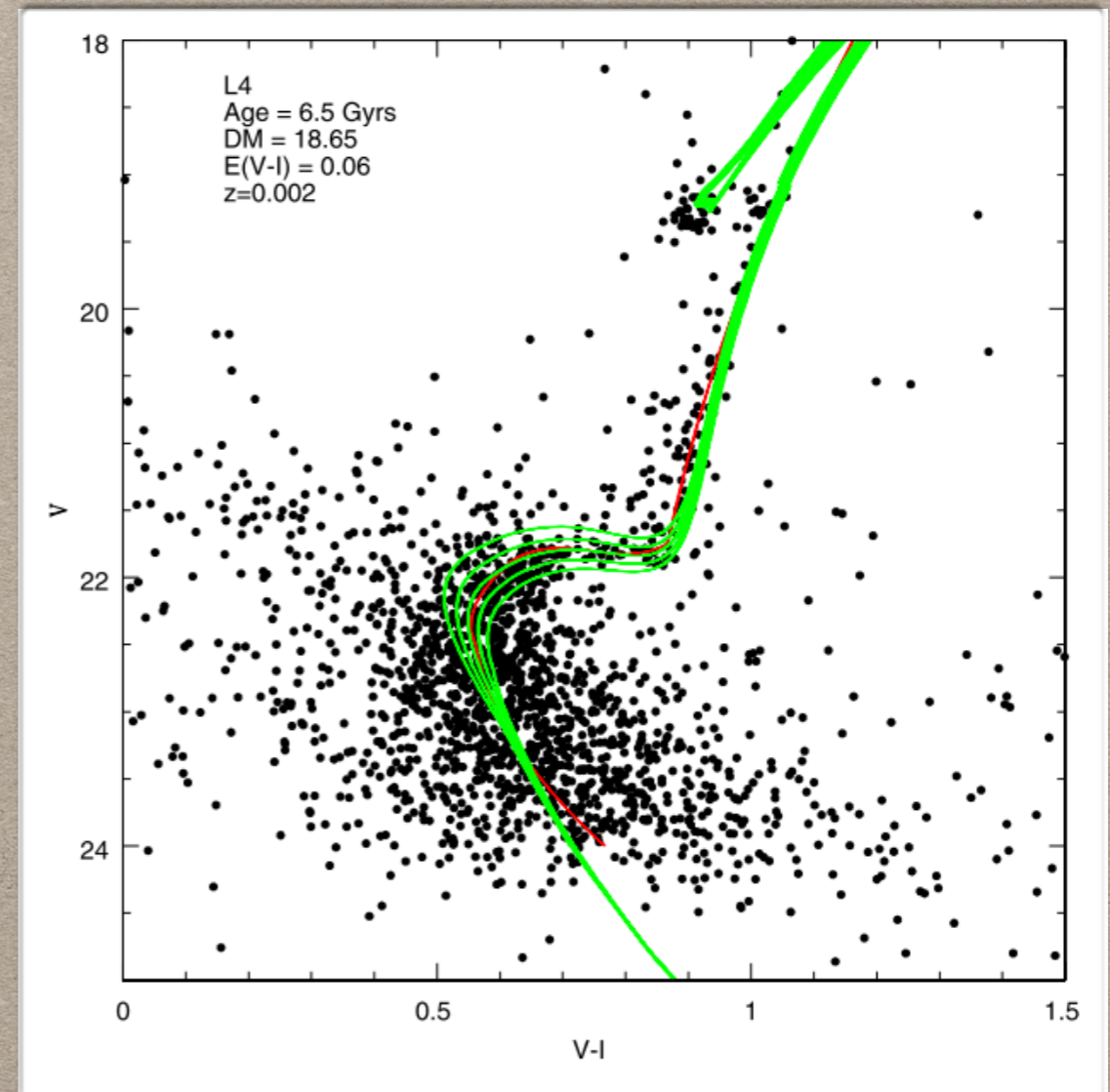
THE CASE: FITTING BY EYE IS TRICKY

Colour-magnitude diagram = CMD



THE CASE: LINE VS. LINE

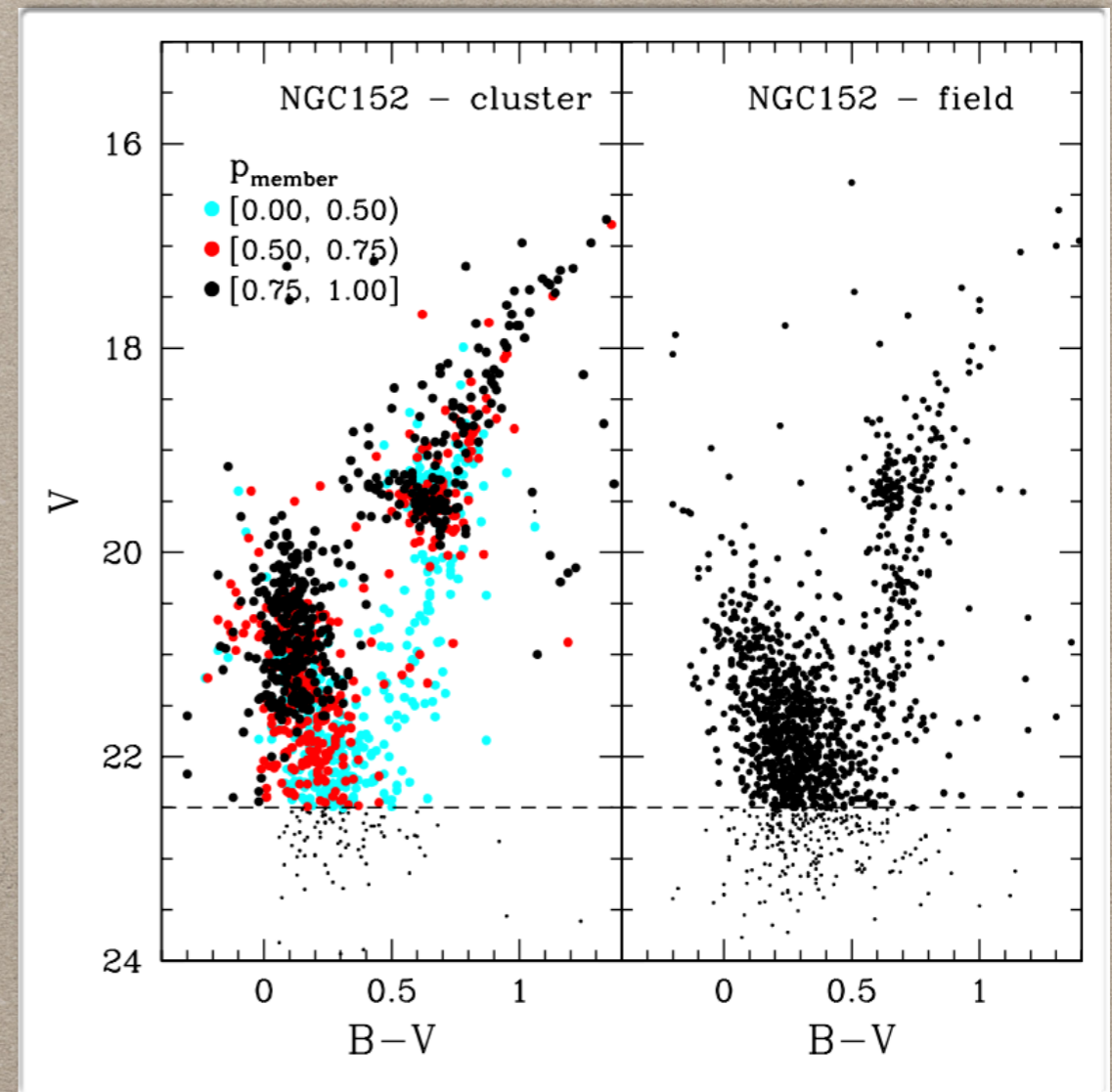
- Red: ridgeline
- Green: BaSTI isochrones (Pietrinferni et al. 2013)
- $(m-M) =$ literature
 $\Delta = 0.2 \sim 10\text{kpc}$
- $[\text{Fe}/\text{H}] =$ literature
- $E(V-I) =$ literature



Parisi et al. (2014)

THE SOLUTION: 1. CLEANING THE CMD

- Control field nearby
- Hypothesis: the positions of the off-cluster stars represent the most likely positions for field stars on any similar CMD.
- p_{member} = probability of a given (observed) star be member of a given cluster

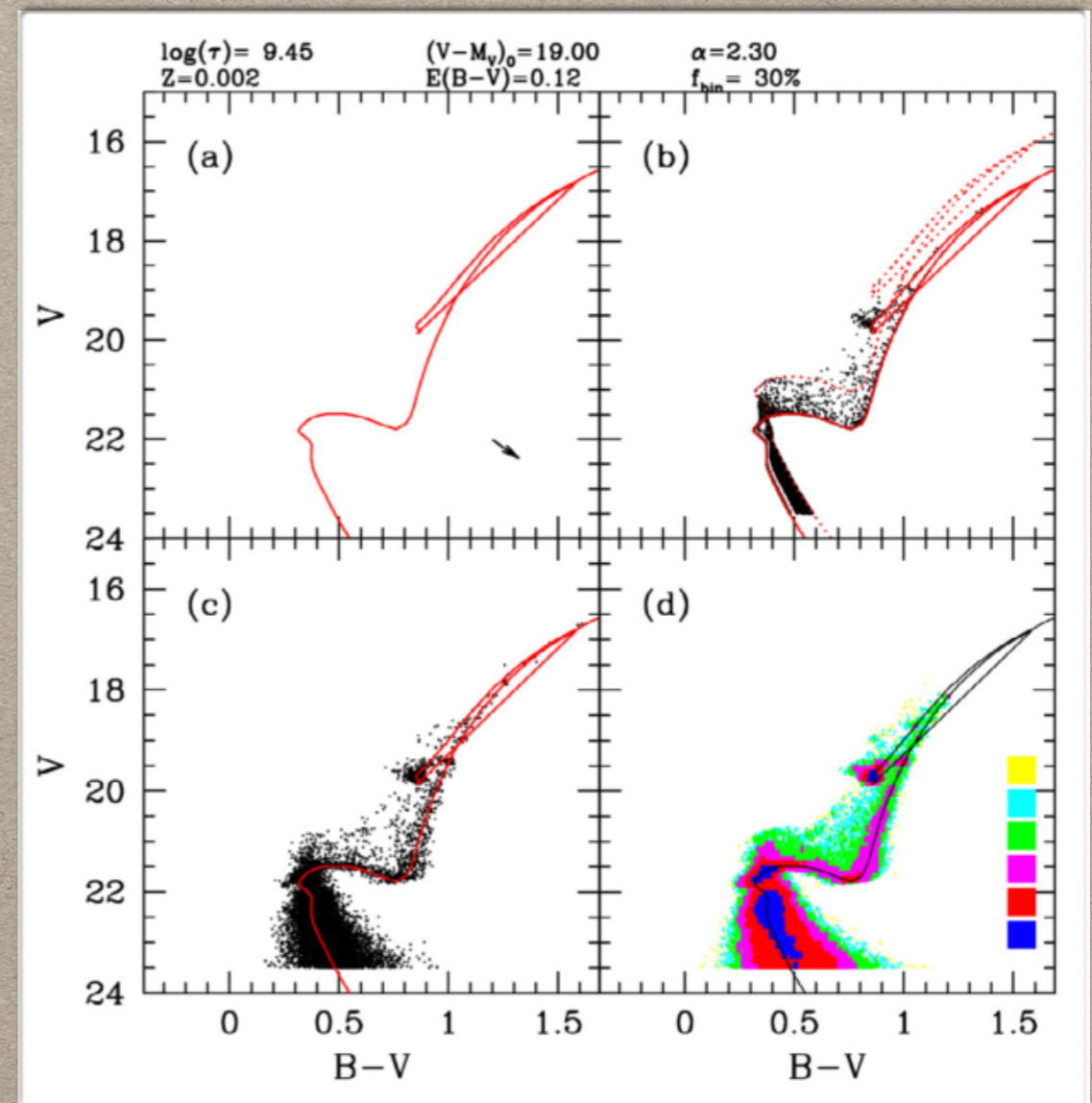


Dias et al. (2016)

THE SOLUTION:

2. CREATING SYNTHETIC CMDs

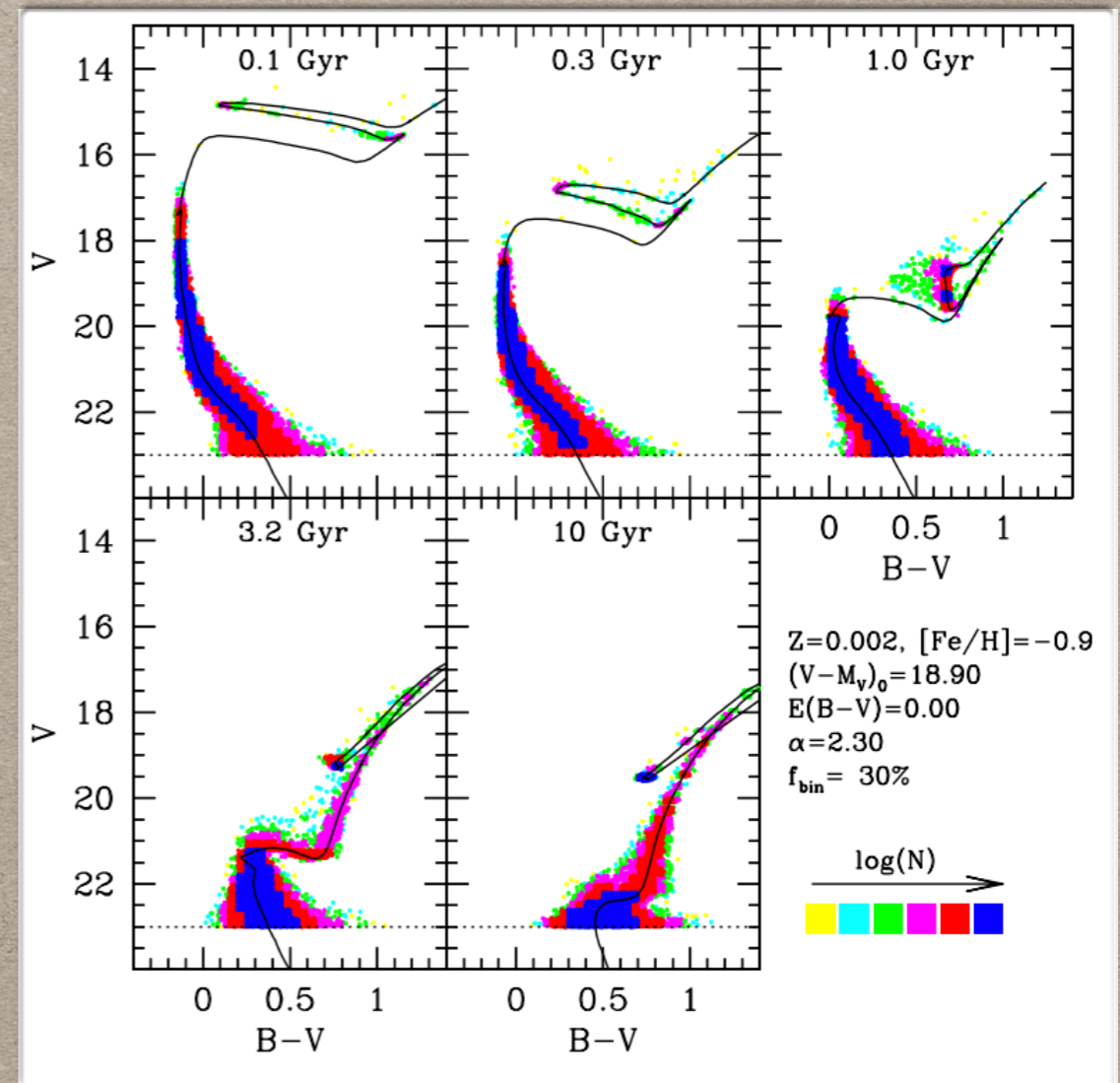
- Isochrone:
age, Z , $E(B-V)$, $(m-M)$
- IMF
- binary fraction
- photometric errors
- photometric completeness
- magnitude range
- Hess diagram: weights!



Dias et al. (2014)

THE SOLUTION: 3. FITTING THE CMD

- "make use of as much information provided by the bidimensional CMD plane as possible."
(Kerber et al. 2002)
- p_{CMD} = probability of a given (synthetic) star at $[(B-V), V]$ be of a given SSP \sim density in the Hess diagram



Dias et al. (2016)

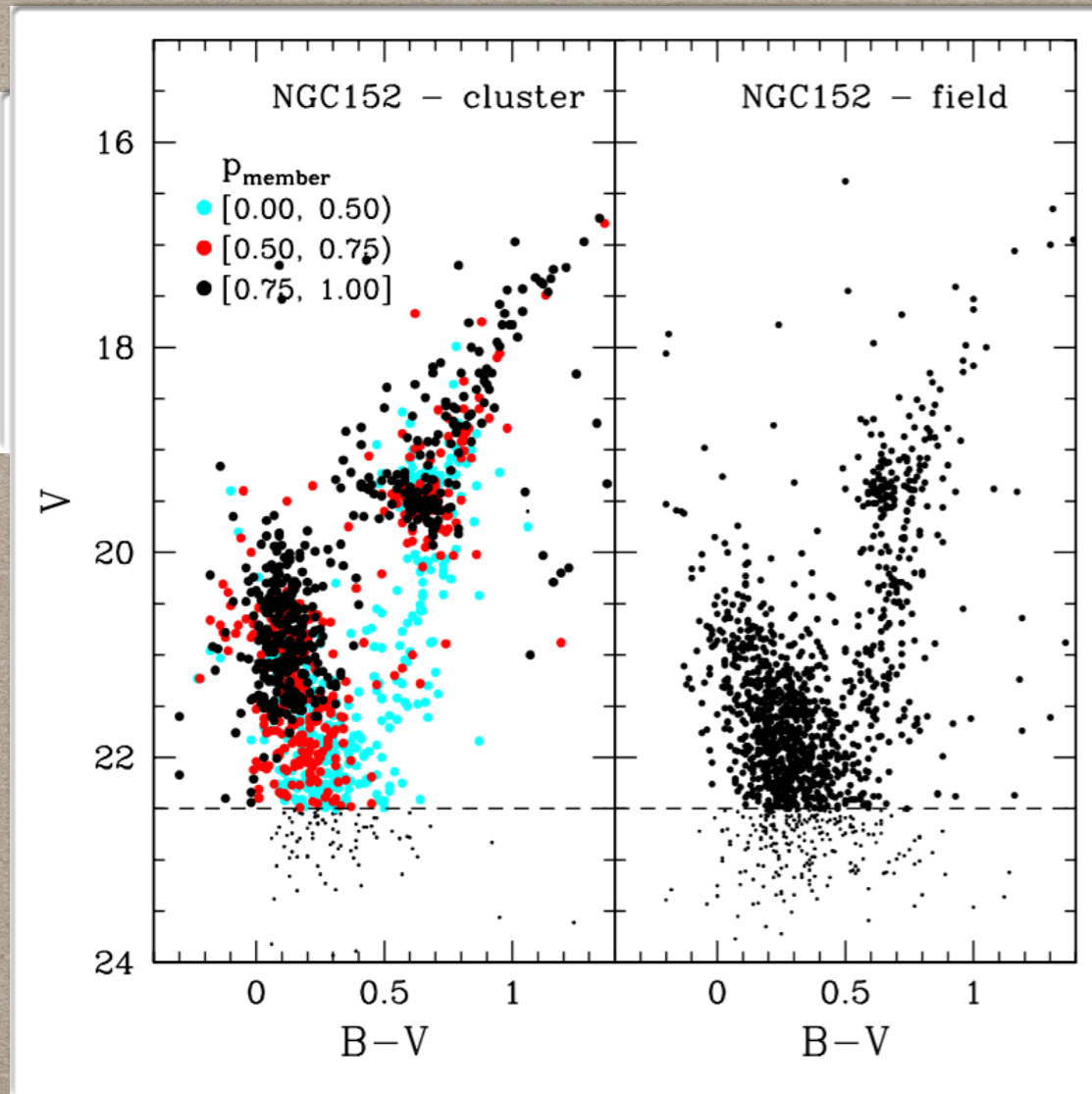
THE SOLUTION:

4. MAXIMUM LIKELIHOOD (EMPIRICAL)

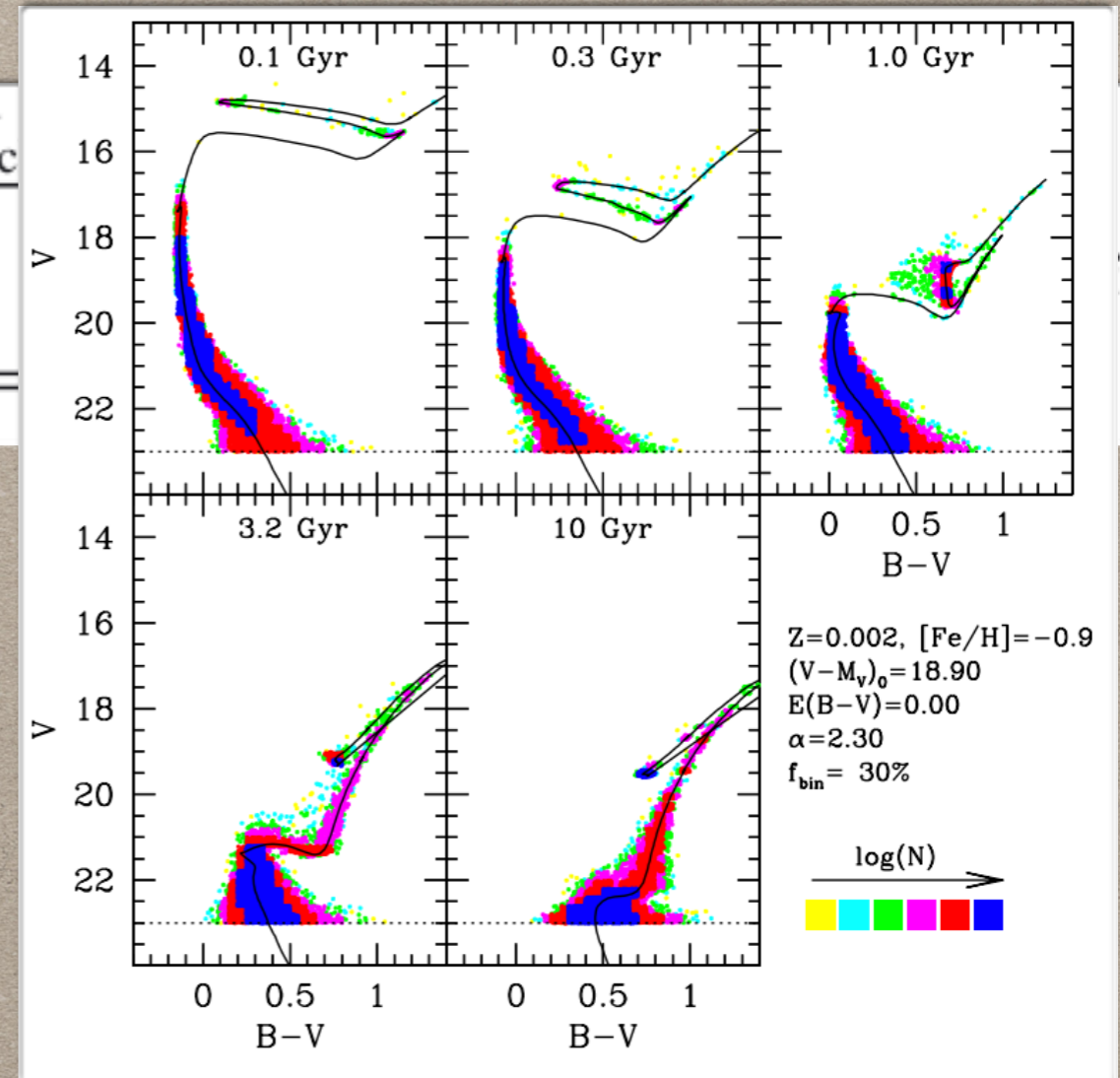
$$L \propto \prod_{i=1}^{N_{\text{clus}}} p_{\text{CMD},i} \times p_{\text{member},i} \propto \prod_{i=1}^{N_{\text{clus}}} N[V_i, (B - V)_i] \times p_{\text{member},i}$$

- p_{member} = probability of a given (observed) star be member of a given cluster
- p_{CMD} = probability of a given (synthetic) star at $[(B-V), V]$ be of a given SSP ~ density in the Hess diagram

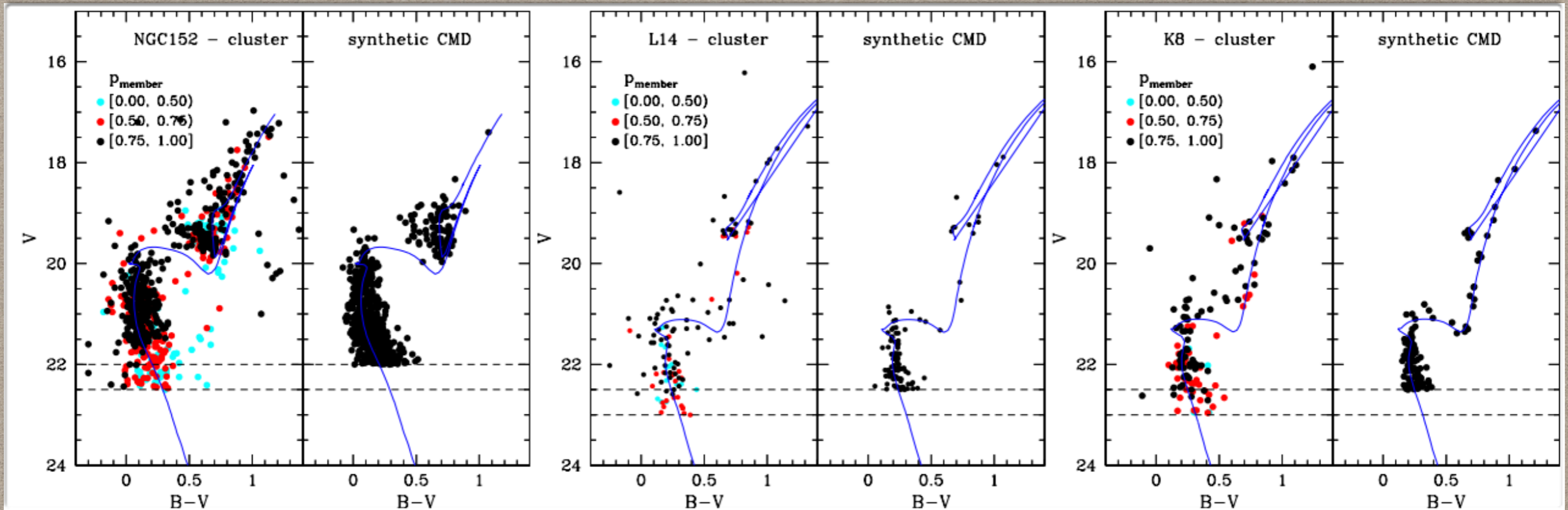
THE SOLUTION: 4. MAXIMUM LIKELIHOOD



$$N_c = \prod_{i=1}^n$$



THE SOLUTION: 5. RESULTS



THE THEORY BEHIND

Cleaned Synthetic
CMD CMD

$$L \equiv p(\{x_i\} | M(\theta)) = \prod_{i=1}^n p(x_i | M(\theta))$$

“Likelihood: the probability of the data given the model”

- Not normalized, i.e., not P.D.F.*
- Function of the data or the model*
- MLE is not probability of parameters θ*
- Usually $\log(L)$ is used: same maximum and $\Pi \rightarrow \Sigma$*

THE THEORY BEHIND

- Critical assumption: the data truly come from the specified model class.

Synthetic CMD = stellar evolution and SSP

- Important property: MLE converges to the true parameter value as the number of data point increases

Even with low statistics we recover well the parameters

THE THEORY BEHIND

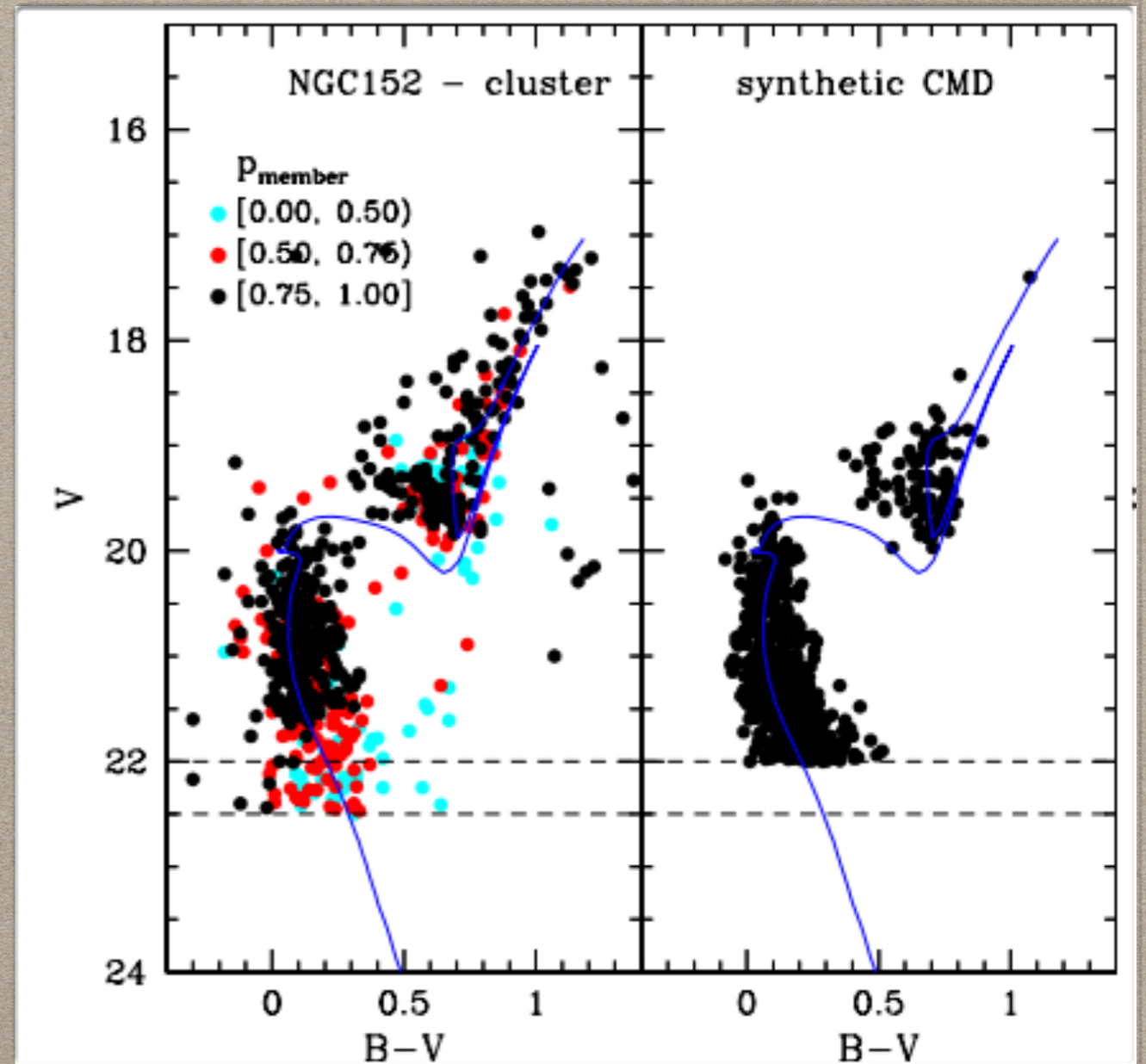
- Goodness of fit: if it is very unlikely to obtain L_{\max} by randomly drawing data from the best-fit distribution, then the model does not represent well the data
- To compare models and expected L_{\max} , we should know the expected distribution of L_{\max} (for Gaussian it is the χ^2 distribution)
- To compare L_{\max} the models should have the same number of parameters

THE THEORY BEHIND

- Uncertainties: analytically with correlation matrix
What if we use an empirical model?

THE SOLUTION: 6. UNCERTAINTIES

- 300 synthetic CMDs with same number of stars as observed
- Compute L distribution
- Take the standard deviation



THE THEORY BEHIND: BOOTSTRAP

