

Bayesian Inference for the White Dwarf Initial-Final Mass Relation

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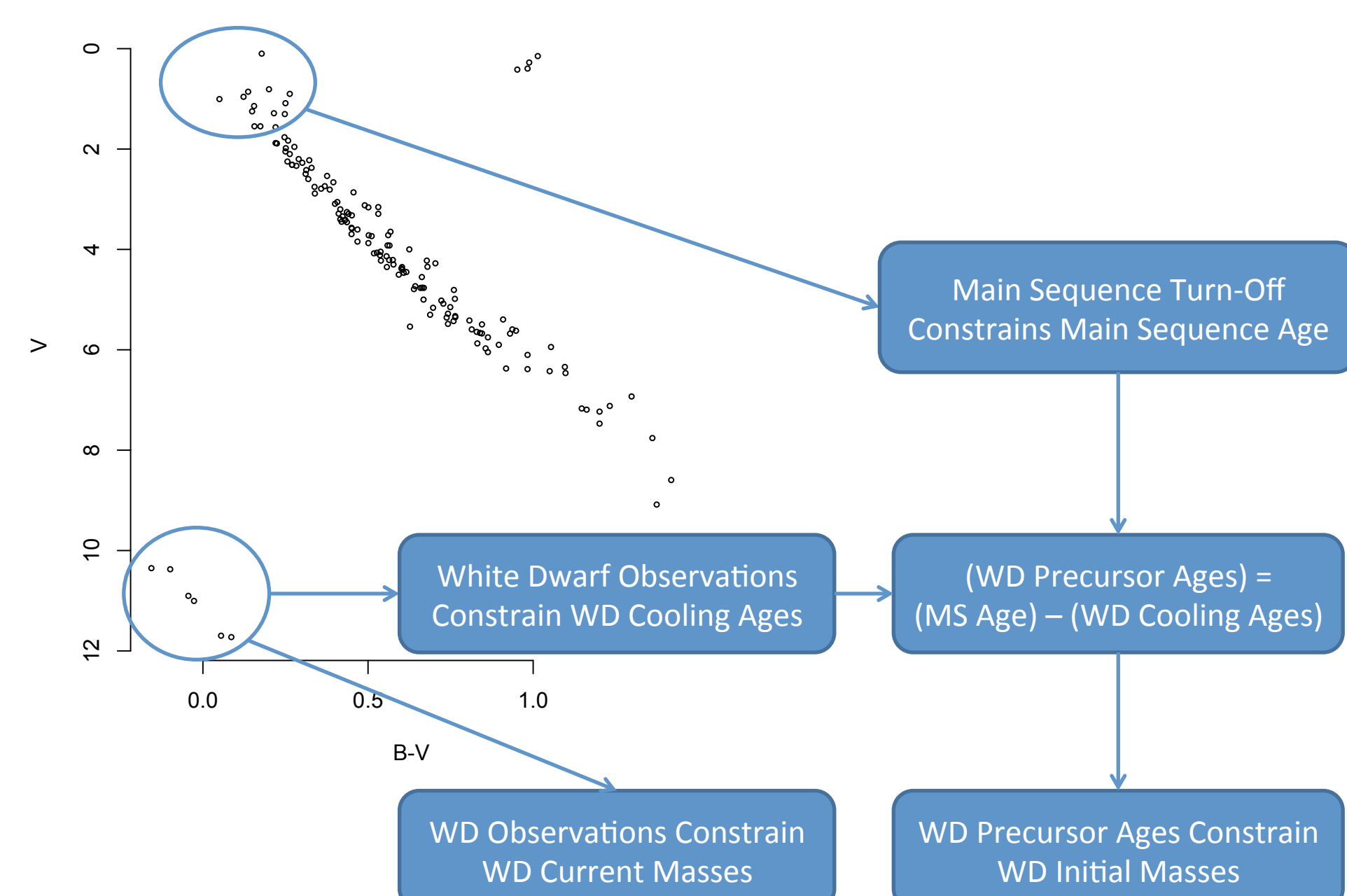
Summary

- Stars lose mass as they age, and understanding mass loss is important for understanding stellar evolution.
- The **initial-final mass relation** (IFMR) is the relationship between a white dwarf's initial mass on the main sequence and its final mass.
- We have developed a new method for fitting the IFMR based on a Bayesian analysis of photometric observations, combining deterministic models of stellar evolution in an internally coherent way. No mass data are used.
- Our method yields precise inferences (with uncertainties) for a parameterized linear IFMR. Our method can also return posterior distributions of white dwarf initial and final masses.

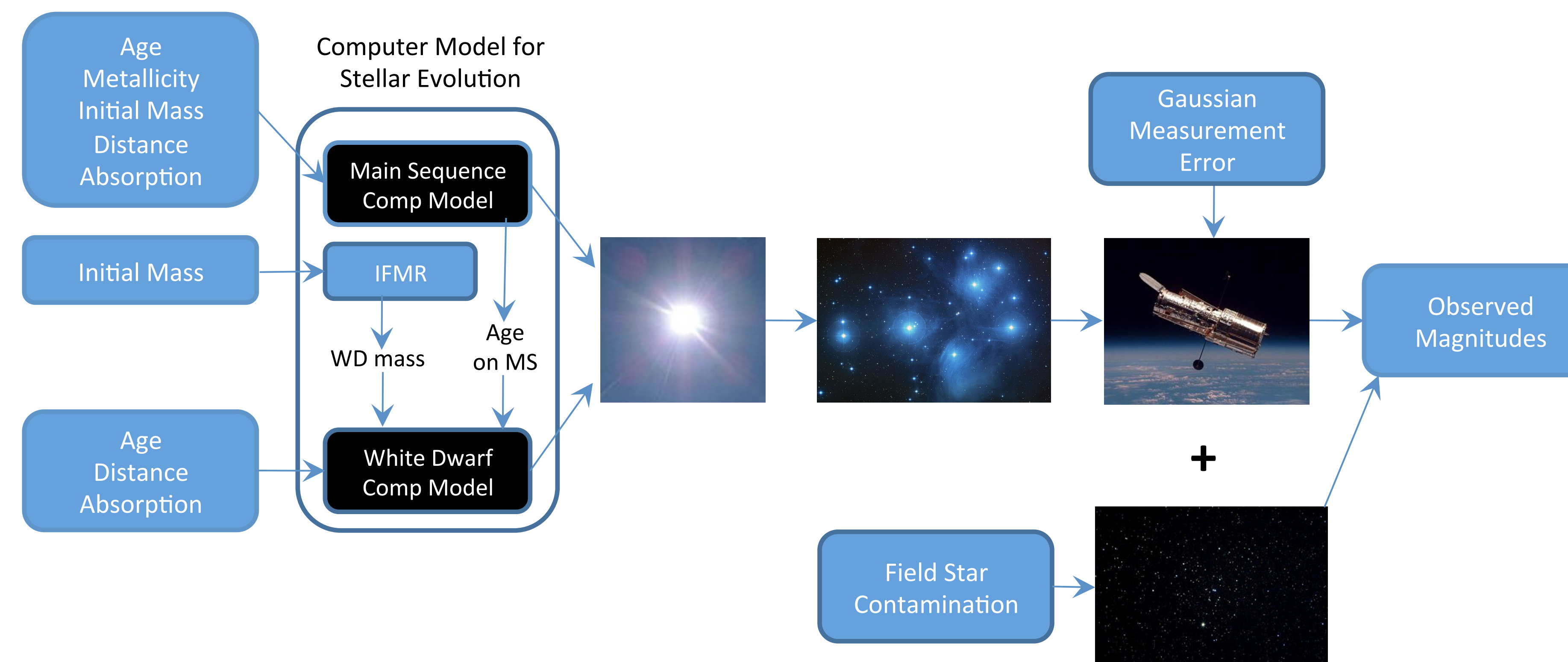
Background: Color-Magnitude Diagrams

- Observe stars' luminosities through different filters
- For the star clusters we study, several parameters are common to all stars:
 - Chemical composition (metallicity)
 - Age
 - Distance
 - Absorption
- Initial masses vary star to star
- Color-magnitude diagrams** show the temperature (horizontal axis) and brightness (vertical axis) of stars in different evolutionary states
- For single-age clusters, these different evolutionary states are determined by stars' initial masses

Fitting the IFMR



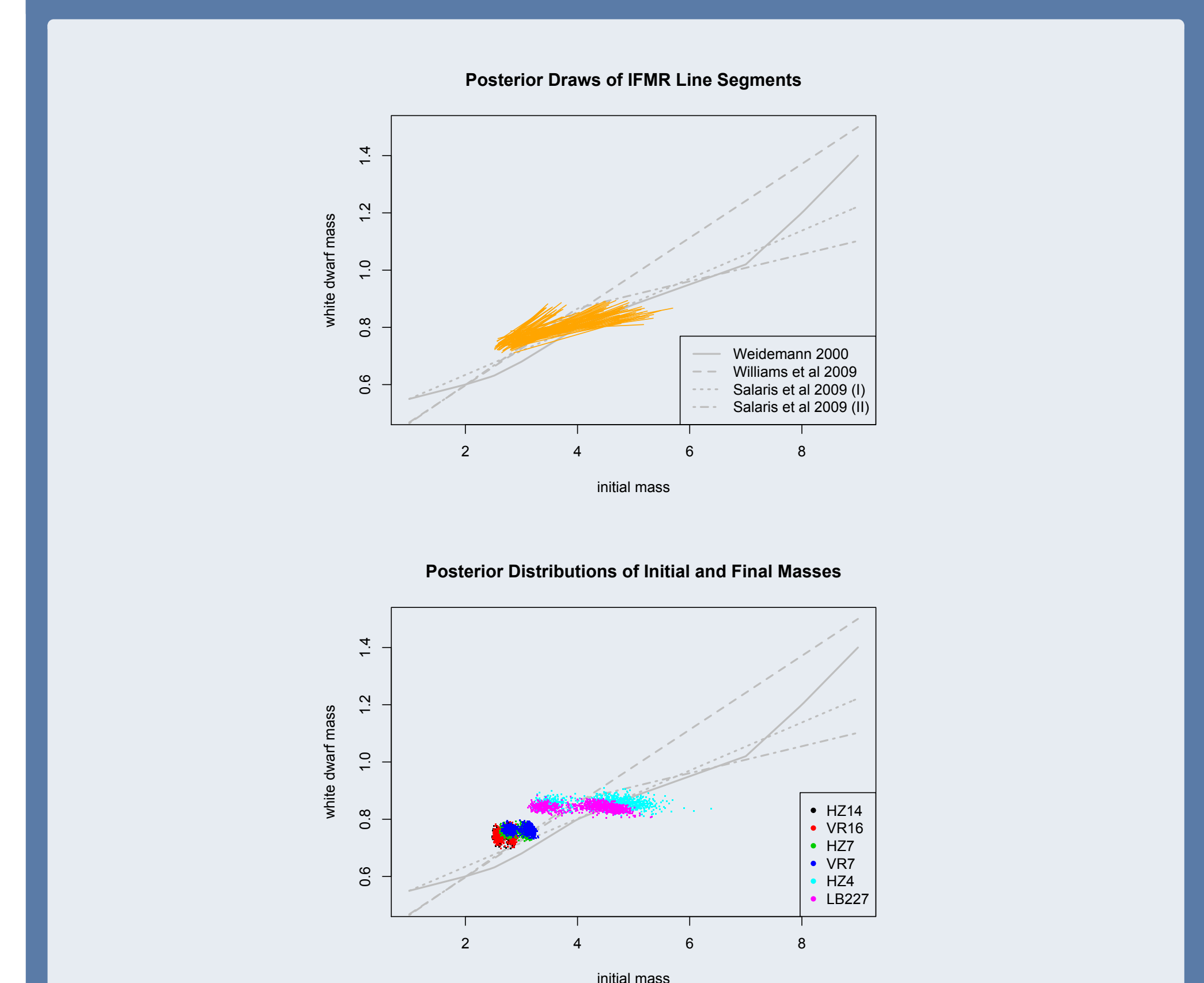
Statistical Model



Model Fitting

- Unknown parameters: $M_1, R, Z, \theta, \alpha$
- MCMC (Metropolis algorithm) on lower-dimensional marginal distribution $p(\theta, \alpha | Y)$, where it is more reliable
- Numerical integration to marginalize over (M_1, R)
- Because of conditional independence, $2N$ -dimensional integral factors into N 2-dimensional integrals that can be evaluated in parallel within each MCMC iteration

Results: Hyades



- Analyzed Hyades data after adjusting for different distances to individual cluster members.
- Inferences agree with IFMRs from the literature, without using white dwarf mass data.
- Bimodality due to two possible age solutions, at approximately 525 Myr and 665 Myr.

Acknowledgements

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Cluster Star Likelihood

- Gaussian errors:

$$Y_i | M_i, \theta, \alpha, \Sigma_i \stackrel{indep}{\sim} N(\mu_i, \Sigma_i)$$
- Y_i = vector of observations of magnitudes through different filters
- $M_i = (M_{i1}, M_{i2})$ = the primary and secondary mass of star i
- θ = vector of cluster parameters, including age, metallicity, distance, and absorption
- Observational uncertainties Σ_i are assumed known
- Means μ_i are functions of unknown parameters and depend on deterministic stellar evolution models G_{ms} and G_{wd}
- If star i is a main sequence star, we model it as a binary system:

$$\mu_{ij} = -2.5 \log_{10} \left(10^{-G_{ms,j}(M_{i1}, \theta)/2.5} + 10^{-G_{ms,j}(M_{i2}, \theta)/2.5} \right)$$

Single star systems will have small M_{i2} , with negligible effect on modeled luminosity

- If star i is a white dwarf, then μ_i depends on the IFMR:

$$\mu_{ij} = G_{wd,j}(M_i, \theta, f, \alpha)$$

- f is the IFMR, which we parameterize as a linear relationship:

$$f(M_i, \alpha) = \alpha_0 + \alpha_1(M_{i1} - M^*)$$

where M^* is a fixed value for centering the white dwarf initial masses.

- $M_{i2} = 0$ for all white dwarfs (we do not model binary systems involving white dwarfs)

Mixture Model for Field Stars

- Field stars appear in observational field of view, but are not part of the cluster.
- For simplicity, field stars are assumed uniformly distributed in magnitude space.
- Mixture model

$$Z_i \sim \text{Bernoulli}(\pi_i)$$

$$(M_{i1}, R_i) | Z_i \sim p(M_{i1}, R_i | Z_i)$$

$$Y_i | M_{i1}, R_i, Z_i \sim \begin{cases} p_1(Y_i | M_{i1}, R_i, \theta, \alpha) & \text{if } Z_i = 1 \\ p_0(Y_i) = \text{constant} & \text{if } Z_i = 0 \end{cases}$$

- $Z_i = 1$ if star i is a cluster member, $Z_i = 0$ otherwise
- π_i = prior probability of cluster membership for star i
- $R_i = M_{i2}/M_{i1}$ = ratio of secondary to primary mass
- p_1 = cluster star likelihood
- p_0 = field star likelihood

Prior Distributions

- Primary mass:

$$\log_{10}(M_{i1}) \sim N(-1.02, 0.677^2), 0.1M_{\odot} < M_{i1} < 8.0M_{\odot}$$
 based on Miller and Scalo's initial mass function
- Cluster membership prior probabilities come from external information when available
- Uniform on R_i , $\log_{10}(\text{age})$, and α , with appropriate boundaries
- Gaussian prior distributions on metallicity and distance
- Truncated Gaussian prior distribution on absorption (absorption is positive)