Hawkins et al. 2017 "Red clump stars and Gaia: calibration of the standard candle using a hierarchical probabilistic model"

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8.11.2022

## Red clump stars



Radek Poleski Hawkins et al. (2017)

# Probabilistic graphical model

- Shaded circles

   observed
   data,
- Open circles model parameters,
- Small filled circles – fixed parameters.





 $r_i$  – distance to star iL – scale-length of the distance prior

 $p(r_i \mid L) = \frac{1}{2L^3}r_i^2 \exp(-r_i/L)$ Bailer Jones (2015)

 $\hat{\varpi}_i$  – measured parallax of star i $\sigma_{\hat{\varpi}_i}$  – measured parallax uncertainty of star i

 $p(\hat{\varpi}_i \mid r_i, \sigma_{\hat{\varpi}_i}) = \mathcal{N}(\hat{\varpi}_i \mid 1/r_i, \sigma_{\hat{\varpi}_i})$  $\mathcal{N}(x \mid x_{\text{mean}}, \sigma_x) - \text{normal distribution}$ 



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#### $M_i$ – predicted mean magnitude of star i

 $M_{
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m RC}}$  – dispersion of RC magnitudes  $\mathcal{N}(M_i \mid M_{
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 $\sigma_{M_{out}}$  – dispersion of magnitudes of outlier population  $f_{out}$  – contamination fraction

 $\theta_{\rm RC} = \{M_{\rm RC}, \sigma_{M_{\rm RC}}, \sigma_{M_{out}}, f_{\rm out}\}$ 

 $p(M_i | \theta_{\rm RC}) = (1 - f_{\rm out}) \mathcal{N}(M_i | M_{\rm RC}, \sigma_{M_{\rm RC}}) + f_{\rm out} \mathcal{N}(M_i | M_{\rm RC}, \sigma_{M_{\rm out}})$ 



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 $E(B - V)_i$  - reddening for star *i* from Green et al. (2015)  $A_i$  - expected extinction  $R_{\lambda}$  - extinction coefficient

$$A_i = R_\lambda \times E(B-V)_i$$

 $m_i$  – expected brightness of star i

$$m_i = M_i + 5 \log_{10}(r_i) - 5 + A_i$$

 $\hat{m}_i$  – measured brightness of star i $\sigma_{\hat{m}_i}$  – measured brightness uncertainty of star i

 $p(\hat{m}_i | \theta_{\mathrm{RC}}, L, r_i, \sigma_{\hat{m}_i}) = \mathcal{N}(\hat{m}_i | m_i, \sigma_{\hat{m}_i})$ 



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$$p(\hat{m}_i | \theta_{\mathrm{RC}}, L, r_i, \sigma_{\hat{m}_i}) = \mathcal{N}(\hat{m}_i | m_i, \sigma_{\hat{m}_i})$$

### Bayesian inference



Data:  

$$\mathcal{D}_i = (\hat{m}_i, \sigma_{\hat{m}_i}, \hat{\varpi}_i, \sigma_{\hat{\varpi}_i}, E(B - V)_i)$$
  
OR:  
 $\mathcal{D}_i = (\hat{m}_i, \hat{\varpi}_i)$ 

# $p(\theta_{\mathrm{RC}}, L \mid \{\mathcal{D}_i\}) \propto p(\theta_{\mathrm{RC}}, L) \prod_i p(\mathcal{D}_i \mid \theta_{\mathrm{RC}}, L, \sigma_{\hat{m}_i}, \sigma_{\hat{\varpi}_i}, E(B-V)_i)$

#### $p(\mathcal{D}_i|\ldots) = p(\hat{\varpi}_i \mid 1/r_i, \sigma_{\hat{\varpi}_i}) \times p(\hat{m}_i|\ldots)$

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## Example posterior



Band	M <sub>RC</sub> (mag)	$\sigma_{M_{ m RC}}$ (mag)	$\sigma_{M_{ m out}}$ (mag)	L (pc)	$f_{\rm out}$	Ν	$R_{\lambda} = \frac{A_{\lambda}}{E(B-V)}$
J	$-0.93 \pm 0.01$	$0.20 \pm 0.02$	$0.72 \pm 0.09$	$213.5 \pm 4.0$	$0.13 \pm 0.05$	972	0.72
H	$-1.46 \pm 0.01$	$0.17\pm0.02$	$0.71 \pm 0.09$	$213.3^{+4.1}_{-3.9}$	$0.18\pm0.05$	972	0.46
$K_{\rm s}$	$-1.61 \pm 0.01$	$0.17 \pm 0.02$	$0.70^{+0.10}_{-0.08}$	$222.7 \pm 4.3$	$0.18 \pm 0.05$	972	0.30
W1	$-1.68 \pm 0.02$	$0.10 \pm 0.04$	$0.73_{-0.09}^{+0.12}$	$231.5 \pm 4.8$	$0.15 \pm 0.04$	936	0.18
W2	$-1.69 \pm 0.02$	$0.20 \pm 0.03$	$0.84 \pm 0.10$	$237.8 \pm 4.8$	$0.15 \pm 0.04$	934	0.16
W3	$-1.67 \pm 0.01$	$0.17 \pm 0.02$	$0.74 \pm 0.08$	$228.3 \pm 4.6$	$0.18 \pm 0.05$	936	0.16
W4	$-1.76\pm0.01$	$0.16\pm0.02$	$0.73^{+0.09}_{-0.07}$	$221.1\pm4.5$	$0.18\pm0.05$	910	0.11

Table 1. Red clump model parameters for the J, H, K, G, W1, W2, W3 and W4 bands.

Note. The bandpass is shown in column 1 while the absolute magnitude and dispersion in the absolute magnitude of the RC and 'contaminate' population in that bandpass is listed in columns 2, 3, 4 and 5, respectively. The inferred scalelength of the distance prior is tabulated in column 6 and the contaminate fraction, *f*<sub>out</sub> can be found in column 7. The number of stars used in the inference and the assumed extinction coefficient for each band is tabulated in column 8 and 9, respectively.

```
 \begin{array}{l} \text{Sampling of:} \\ \theta_{\text{RC}} = \{ M_{\text{RC}}, \sigma_{M_{\text{RC}}}, \sigma_{M_{out}}, f_{\text{out}} \} \\ L \\ \{ r_i \} \end{array}
```

## Output

```
Sampling of:

\theta_{\rm RC} = \{M_{\rm RC}, \sigma_{M_{\rm RC}}, \sigma_{M_{out}}, f_{\rm out}\}

L

\{r_i\}
```

Quote 1:

"We only select those stars which have an inferred probability of belonging to the RC population that is larger than or equal to 80 per cent."

Quote 2:

"Probable RC stars are defined as those which have probabilities of being attributed to the RC component greater than or equal to 80 per cent. In this case, the probability for each star belonging to the RC is computed for every MCMC chain and the median is taken."

```
Sampling of:

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L

\{r_i\}
```

What else they calculated?

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What else they calculated? I'm guessing:  $M_i = \hat{m}_i - 5 \log_{10}(r_i) + 5 - A_i$   $\begin{array}{l} \text{Sampling of:} \\ \theta_{\text{RC}} = \{ M_{\text{RC}}, \sigma_{M_{\text{RC}}}, \sigma_{M_{out}}, f_{\text{out}} \} \\ L \\ \{ r_i \} \end{array}$ 

What else they calculated? I'm guessing:  $M_i = \hat{m}_i - 5 \log_{10}(r_i) + 5 - A_i$ 

Which allows calculating probability that given star belongs to RC:  $p(\text{RC}, i) = \frac{(1-f_{\text{out}})\mathcal{N}(M_i \mid M_{\text{RC}}, \sigma_{M_{\text{RC}}})}{(1-f_{\text{out}})\mathcal{N}(M_i \mid M_{\text{RC}}, \sigma_{M_{\text{RC}}}) + f_{\text{out}}\mathcal{N}(M_i \mid M_{\text{RC}}, \sigma_{M_{\text{out}}})}$ 

## Error shrinkage



#### • Inconsistent notation throughout the paper.

- Unclear p(RC, i).
- Hard to read histograms.
- No ticks on the axes.
- L and r<sub>i</sub> values change between bands.
- Fixed  $R_{\lambda}$ .

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