Studying eclipsing binaries (EBs) in large-scale multi-epoch surveys

Nami Mowlavi

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+ B. Holl, I. Lecoeur, L. Rimoldini, F. Barblan, L. Eyer (Geneva, Switzerland)
+ A. Prša, A. Kochoska (Villanova, USA)
+ T. Mazeh (Tel Aviv, Israel)
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Large scale multi-epoch surveys

→ New perspectives to study binaries/multiples

Aims: - Study specific binary systems

- Identify peculiar binary systems
- From populations of eclipsing binaries
 to populations of binary systems

EBs in large scale surveys

2'878 (Kirk et al. 2016)

40204 (Pawlak et al. 2016)

Number of EBs in some surveys

Hipparcos : 852 (ESA SP-1200, 1997)

: I I '076 (Paczynski et al. 2006)

TrES : 773 (Devor et al. 2008)

OGLE-3 LMC : 26'121 (Graczyk et al. 2011)

OGLE-3 SMC : 6'138 (Pawlak et al. 2013)

OGLE-3 Gal. disks : 11'589 (Pietrukowicz et al. 2013)

Expectations for Gaia (Eyer et al. 2013)

- → 4* (0.5-7) million EBs (with 12*% spectroscopic binaries)
- → 8* million spectroscopic binaries (with 59*% SB2)
- → 30* million astrometric non-single stars

Expectations for LSST (Prsa et al. 2011)

→ ~24 million EBs

8'40 | (Pawlak et al. 2016) OGLE-4 SMC : OGLE-4 Bulge: 450'598 (Soszynski et al. 2016) Photometry

ASAS

Kepler

OGLE-4 LMC:

Based on CU2 simulations of 1% of stars of the Galaxy, using Robin et al. (2012) model of Galaxy

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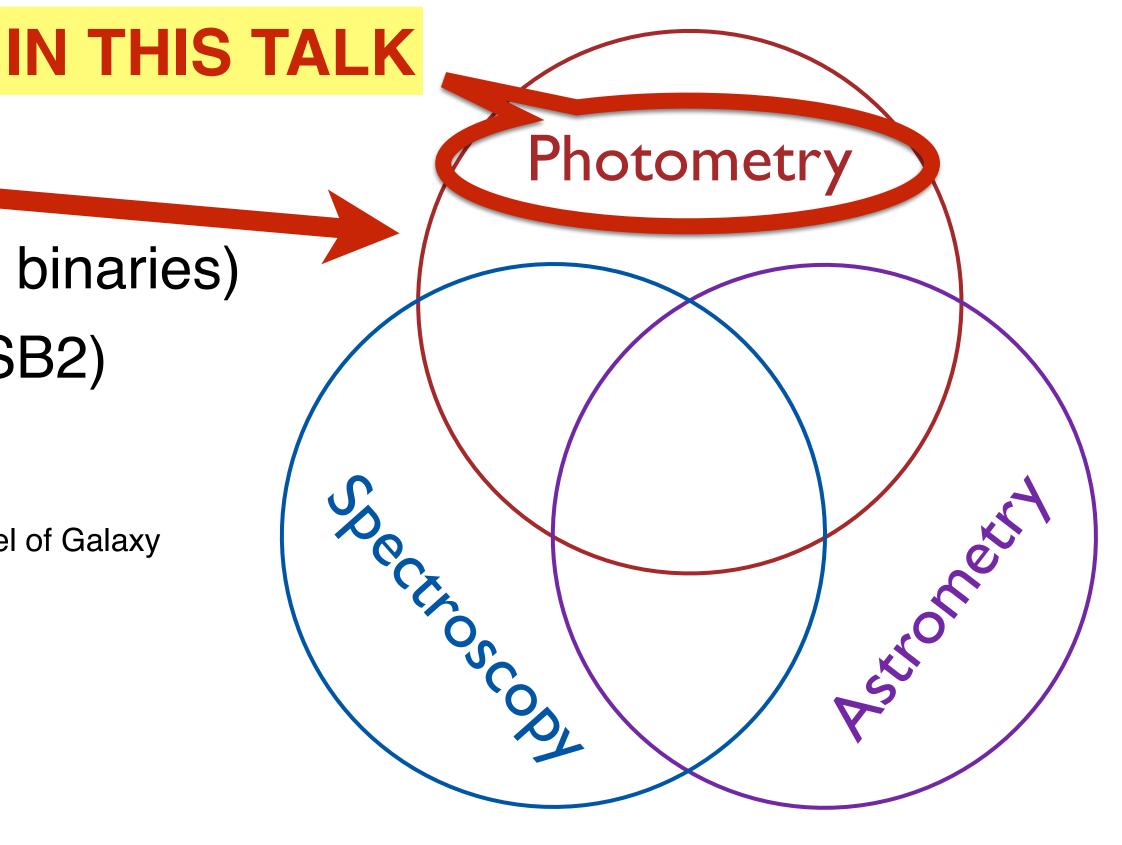
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Based on CU2 simulations of 1% of stars of the Galaxy, using Robin et al. (2012) model of Galaxy

Challenges

1. Identification of EBs

- Variable stars
 - Periodic variables
 - Binary stars

cf. Poster 33 by B. Holl et al.

2. EB characterization based on light curve morphology

3. EB sub-classification in distinct classes

To be done in an automated way

Rewards

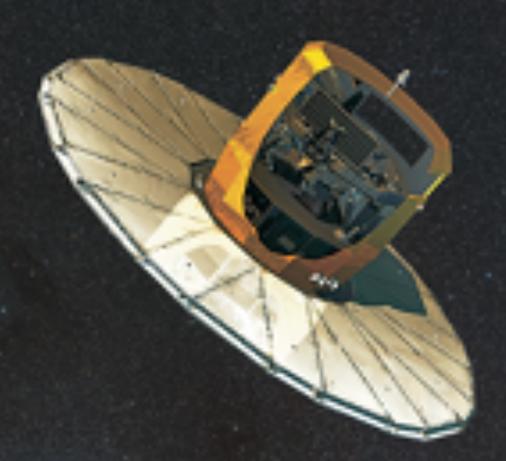
- → Get basic binary properties without full system modeling
- → Study populations of EBs
- → Explore outliers

Poster 33

Automatic determination of eclipsing binary periods, geometric-model and quality ranking in Gaia data

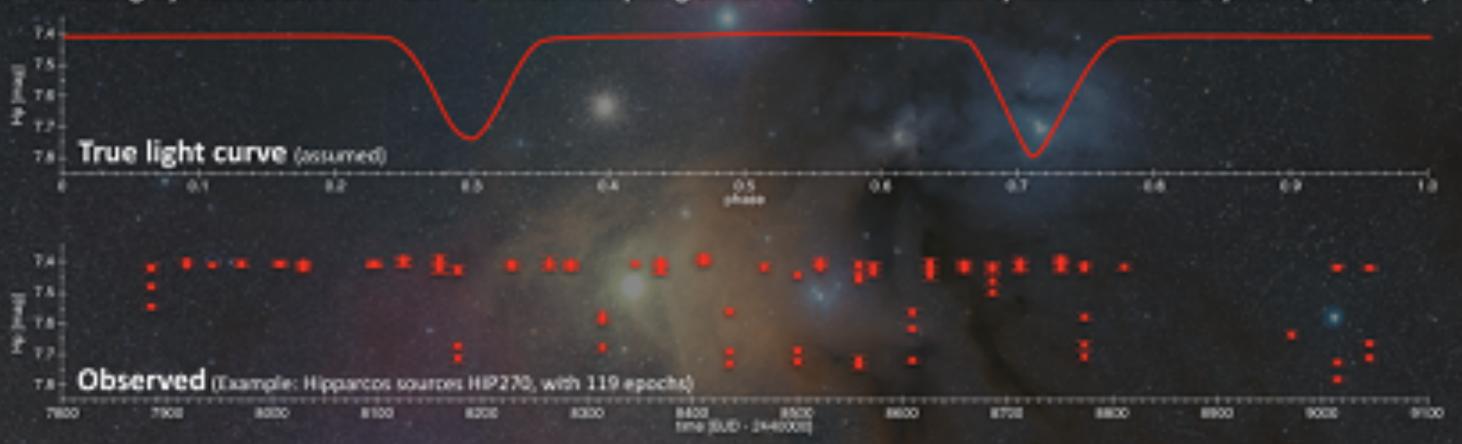


B. Holl, N. Mowlavi, I. Lecoeur-Taibi, F. Barblan, L. Rimoldini, L. Eyer, on behalf of the CU7 Consortium (Department of Astronomy, University of Geneva, Switzerland)



Observing periodic eclipsing binary signal

During Syr Gaia observes ~0.5 - 5 million eclipsing binaries (Holl et al. 2013) with 45 to 300 epochs (70 mean).



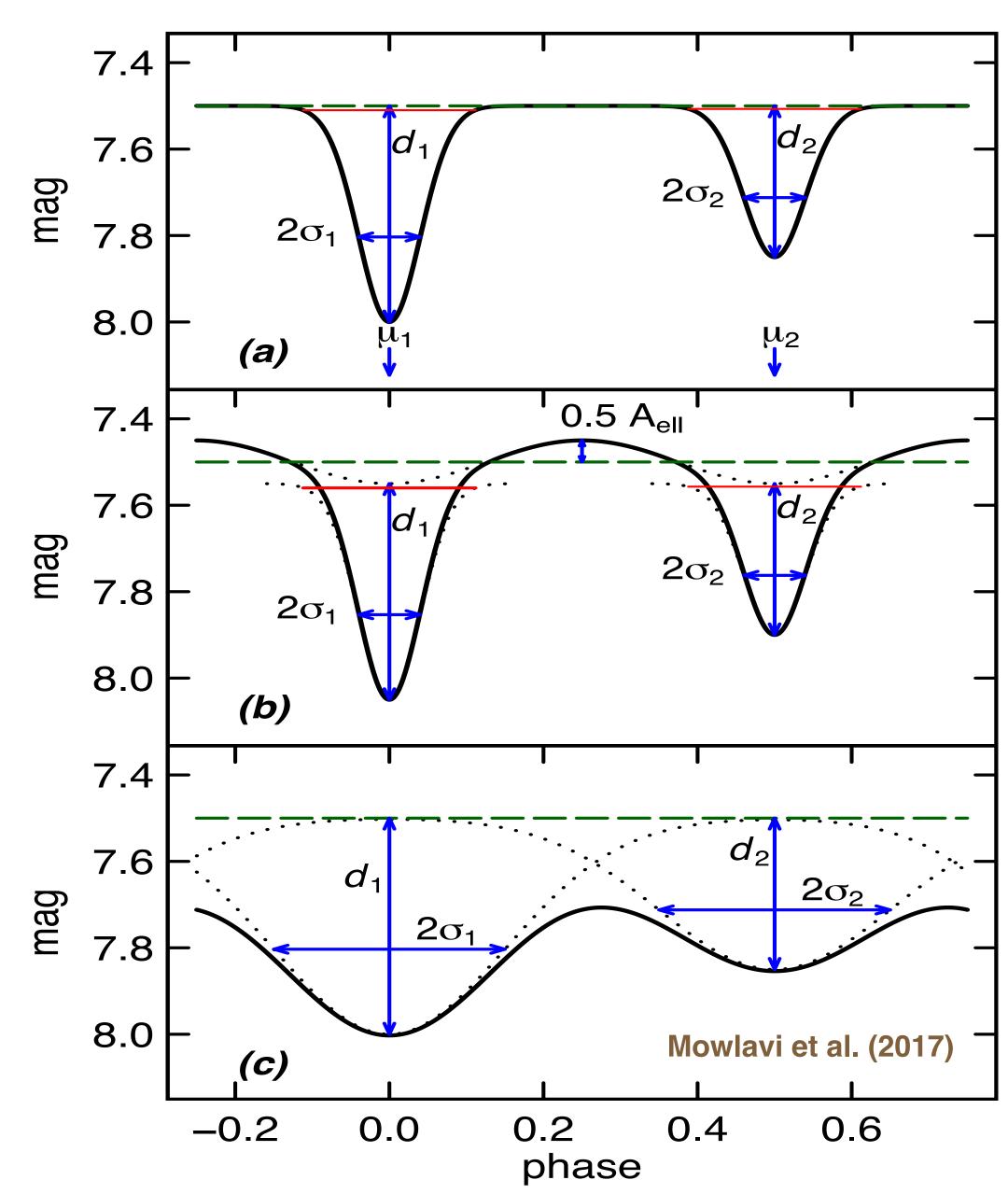
Period recovery from such sparsely sampled data

- An ensemble of period search methods is used to identify candidate periods to cover the wide variety of geometric light curve shapes.
 - From each periodogram we extract several highest peaks and frequently



Identify candidate periods Period search with various methods

'2-Gaussian' models



Components

- Gaussian → eclipses
- Sine half period → ellipsoidal variability
- Sine full period → reflection
- Flat bottom → total eclipses
- beaming, ...

Model selection

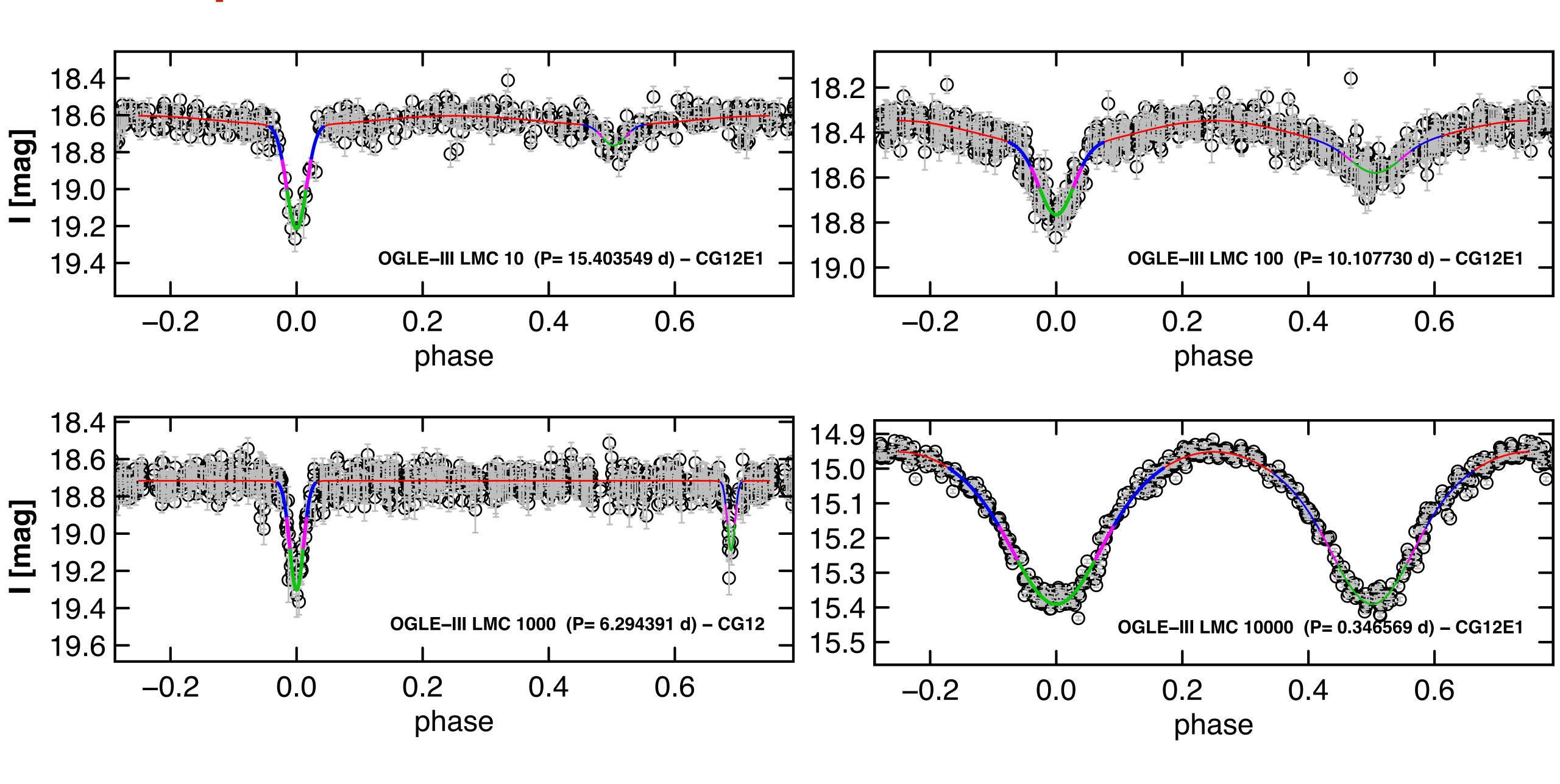
Bayesian Information Criterion

Binary properties

- → eclipse separation
- → eclipse width
- → eclipse depth
- → Ellipsoidal variability amplitude

Example model fits to OGLE-III

EB light curve characterization



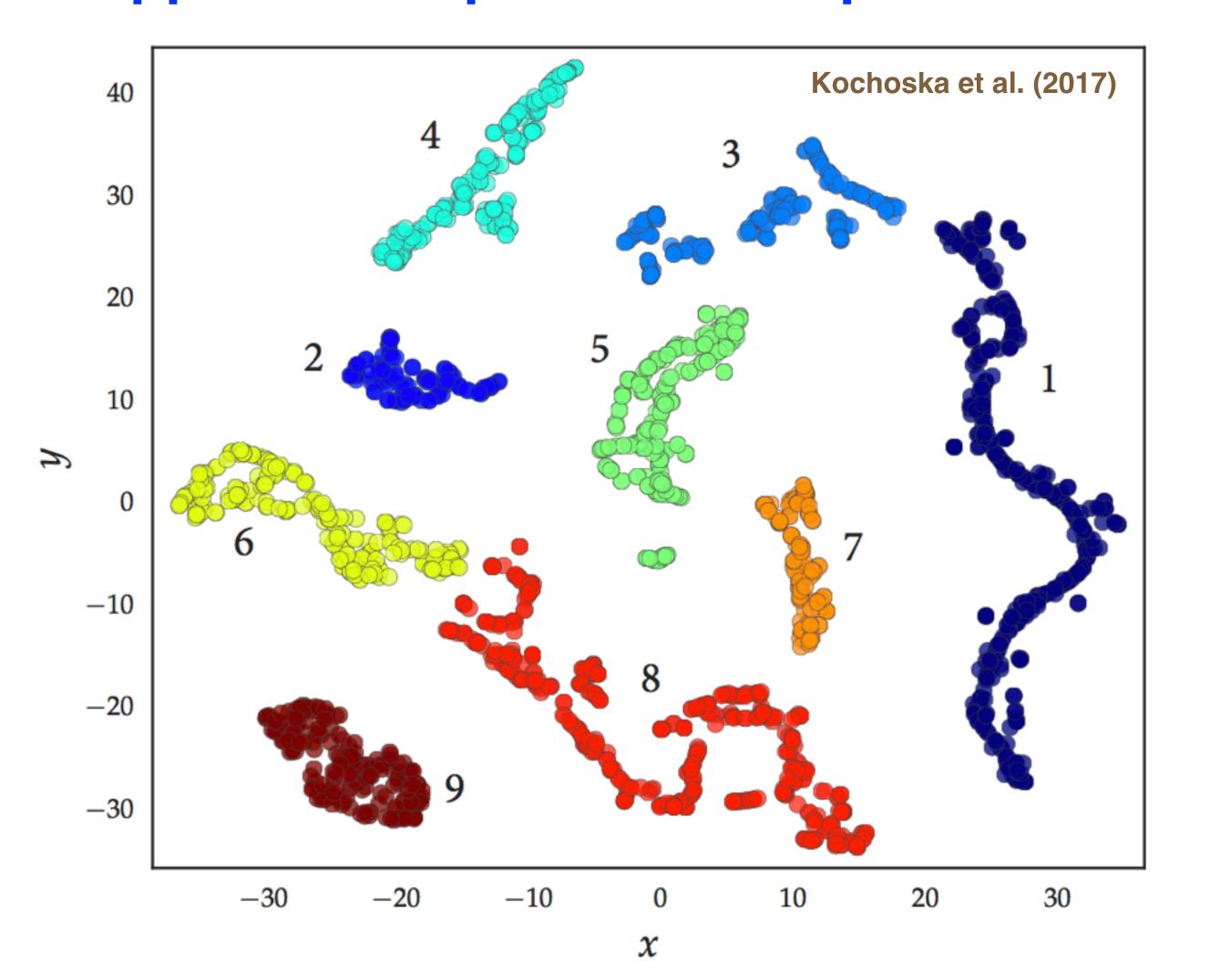
Examples of classification techniques

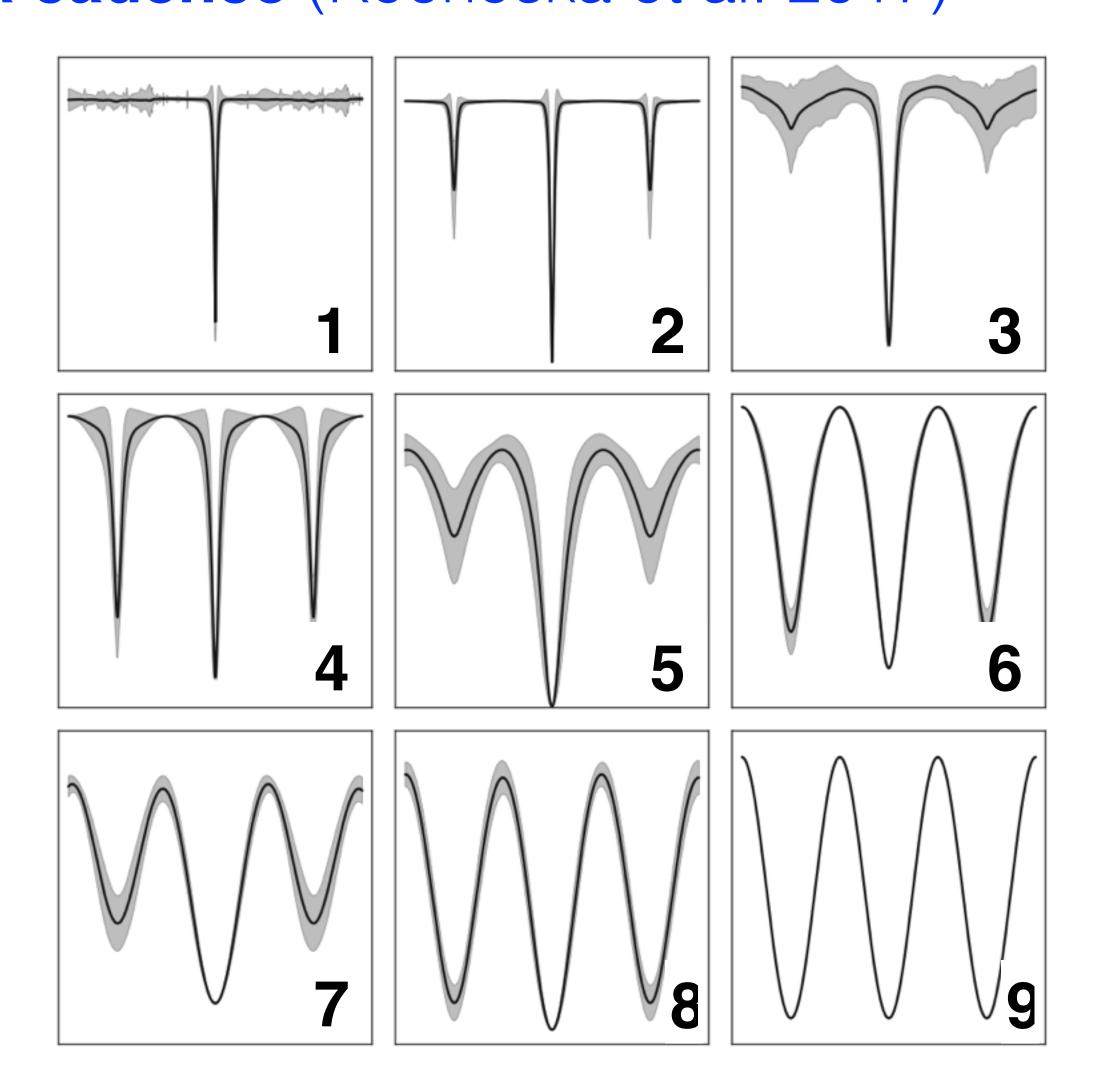
- EA, EB, EW
 GCVS (http://www.sai.msu.su/gcvs/gcvs)
- Detached, Semi-detached, Contact
 Fourier Series parameters (Pojmanski 2002)
- One-parameter LC morphology-based
 Locally Linear Embedding (LLE; e.g. Matijevic et al. 2012)
- t-SNE + DBSCAN
 t-Distributed Stochastic Neighbor Embedding + Density-based spatial clustering (Kochoska et al. 2017)
- Machine learning techniques...
 (e.g. Süveges et al 2017)

EB light curve classification

Ex: Classification based on per-point similarity of phase-folded light curve models using t-SNE (dimensionality reduction) + DBSCAN (clustering algorithm)

Applied to Kepler data sampled with Gaia cadence (Kochoska et al. 2017)





How to detect peculiar cases?

Two quantities:

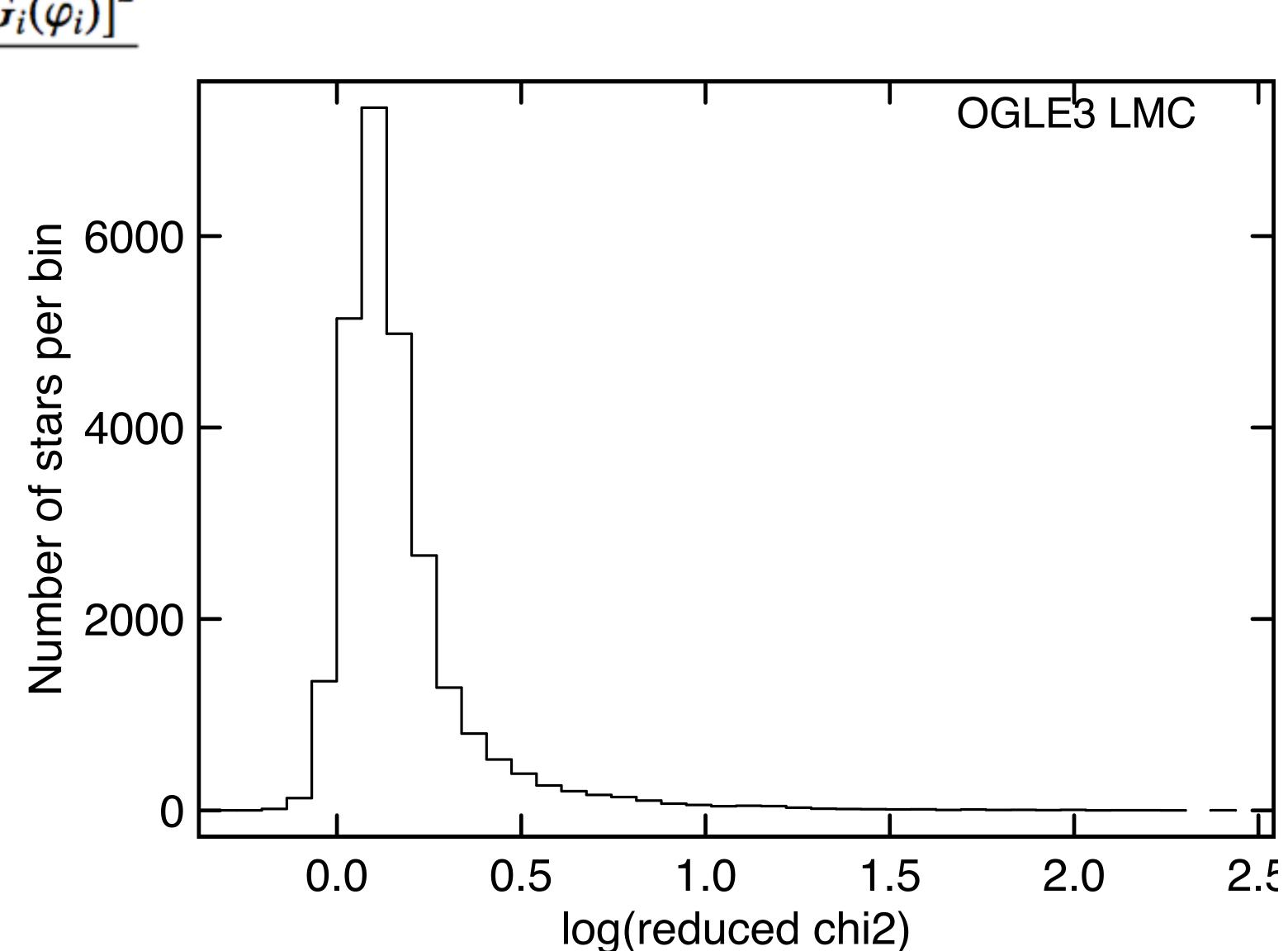
- **Reduced** χ^2 : How well is the light curve modeled?

- Abbe value: How well is the variability pattern described?

Reduced chi2

Test presence of intrinsic variability

$$\chi^{2}_{\text{reduced}} = \frac{1}{(N_{\text{obs}} - N_{\text{param}})} \sum_{i=1}^{N_{\text{obs}}} \frac{\left[y_{i}(\varphi_{i}) - G_{i}(\varphi_{i})\right]^{2}}{\epsilon_{i}^{2}}$$



Abbe value

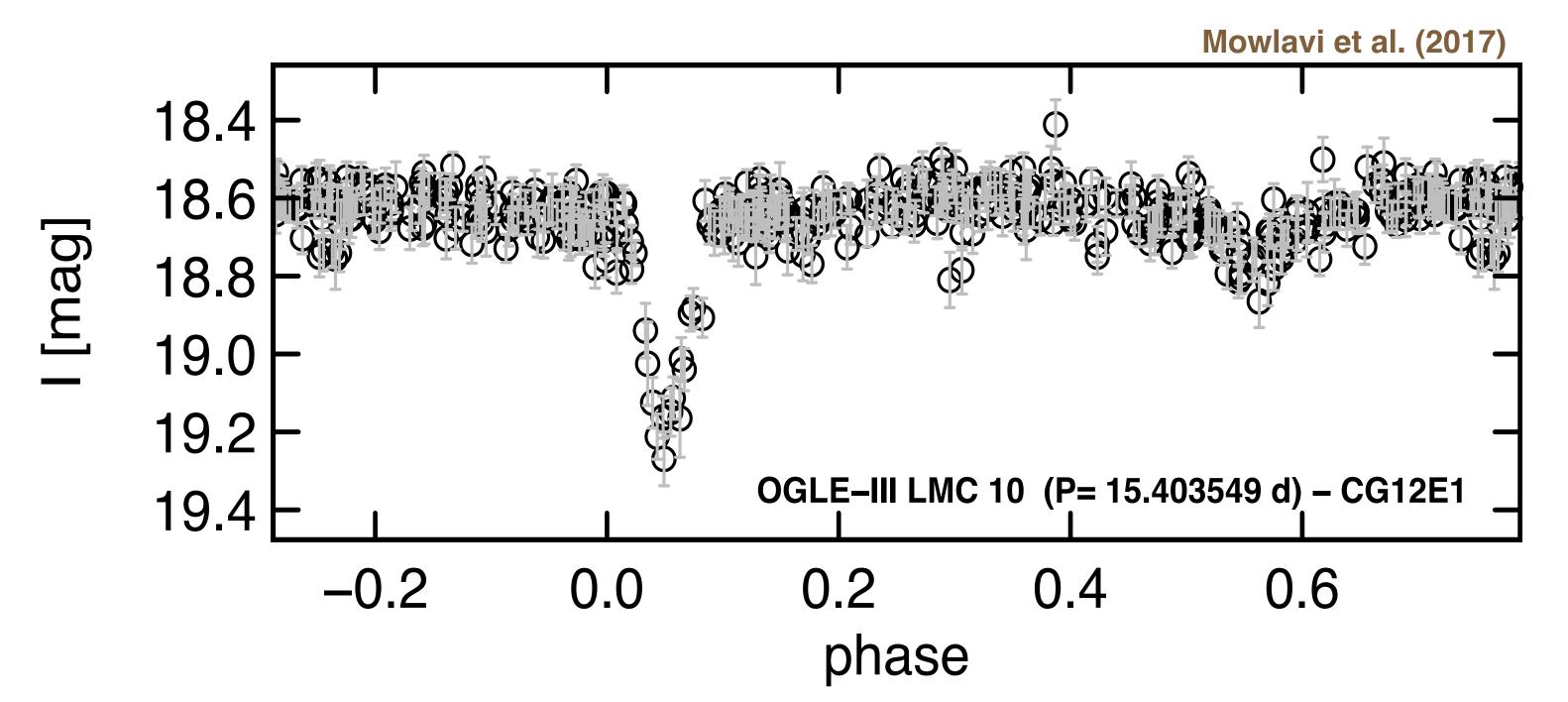
Test presence of a variability pattern

$$\mathcal{A} = \frac{n}{2(n-1)} \frac{\sum_{j=1}^{n-1} (y_{j+1} - y_j)^2}{\sum_{j=1}^{n} (y_j - \bar{y})^2}$$

No variability pattern : $A \approx 1$

Clear variability pattern: $\mathcal{A} \rightarrow 0$

Application 1: Identification of peculiar cases



Original LC: A=0.27

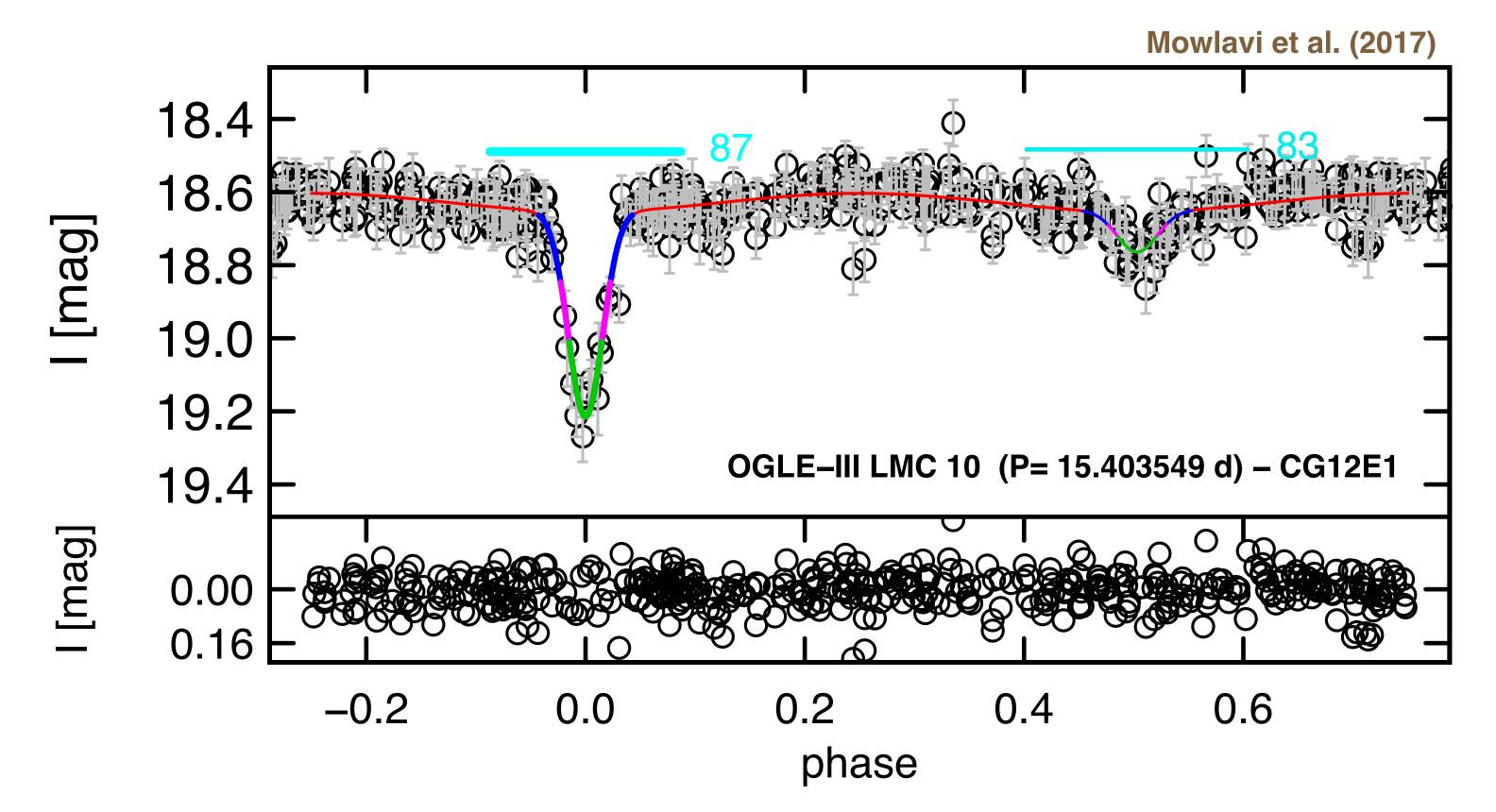
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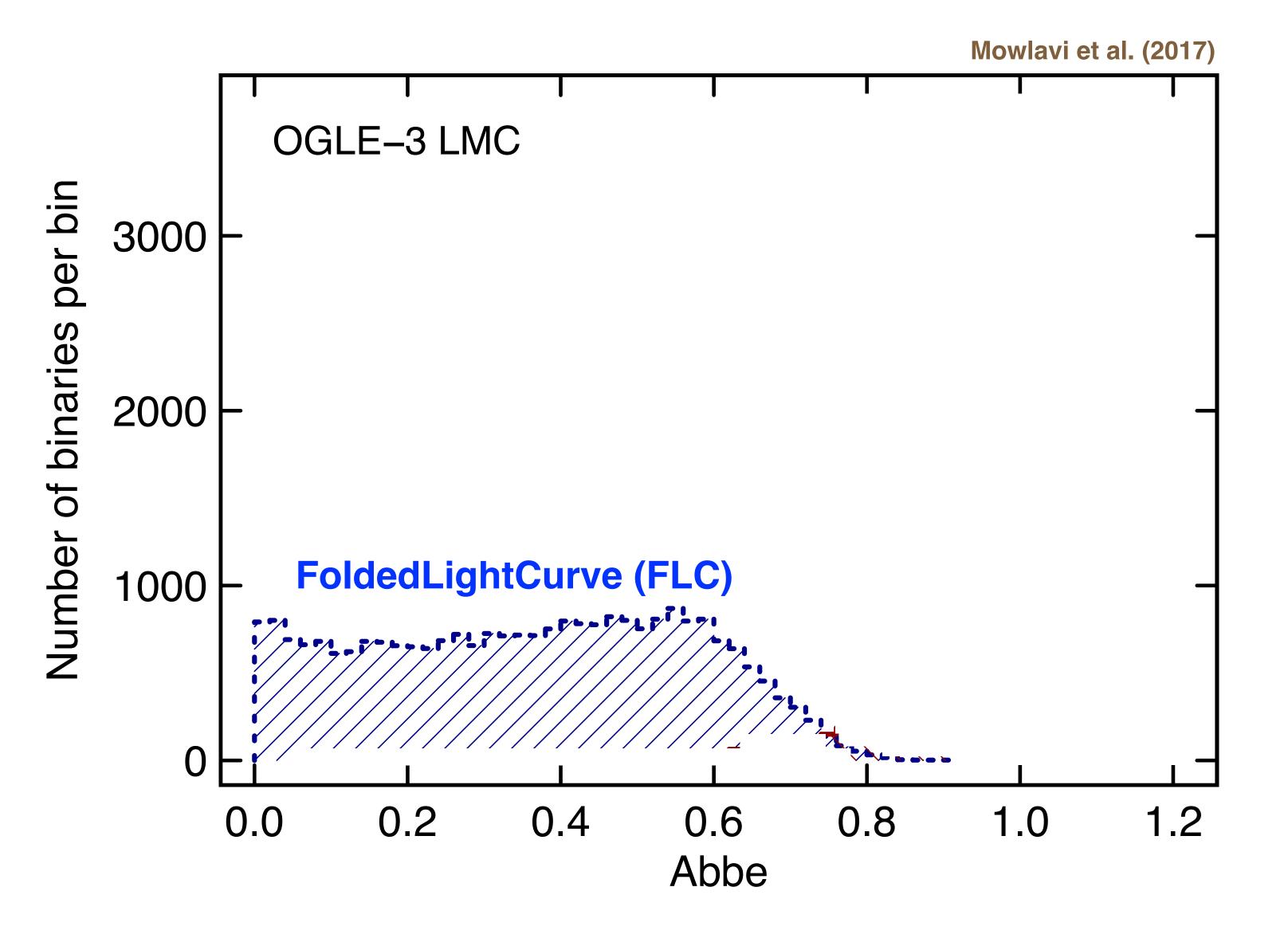
Residual LC: A=0.93

Reduced chi2

Test presence of intrinsic variability

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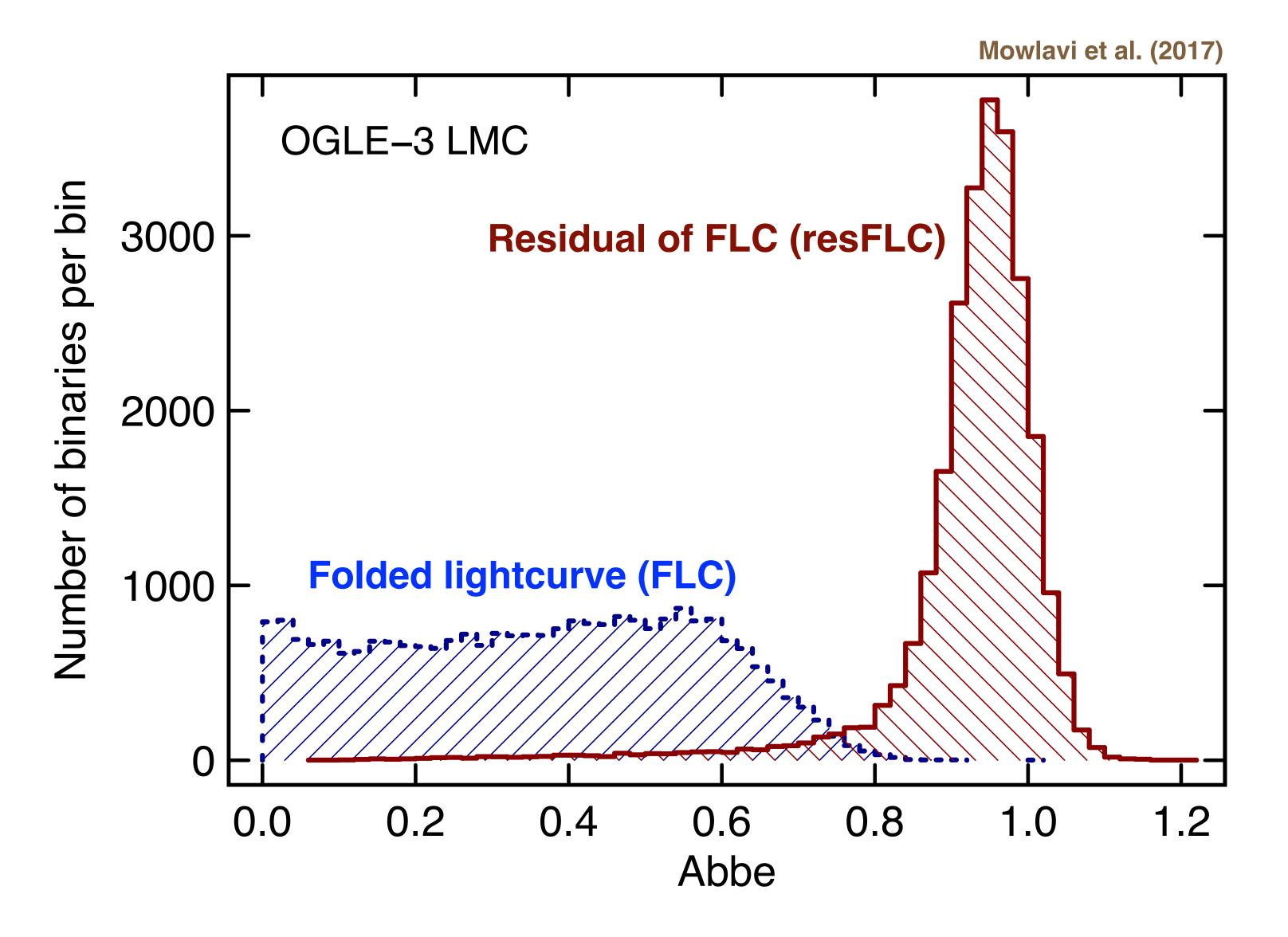
Reduced chi2

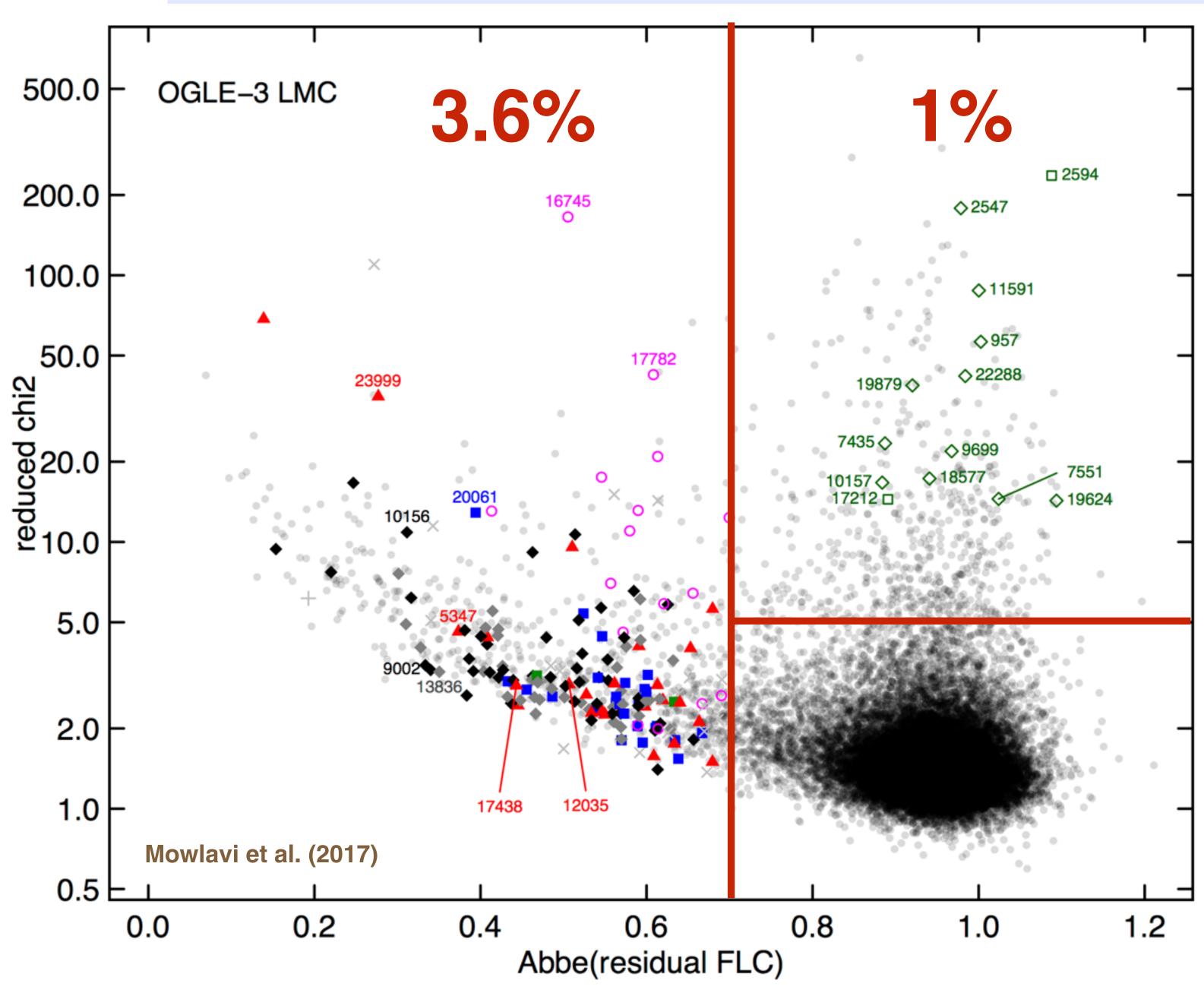
Test presence of intrinsic variability

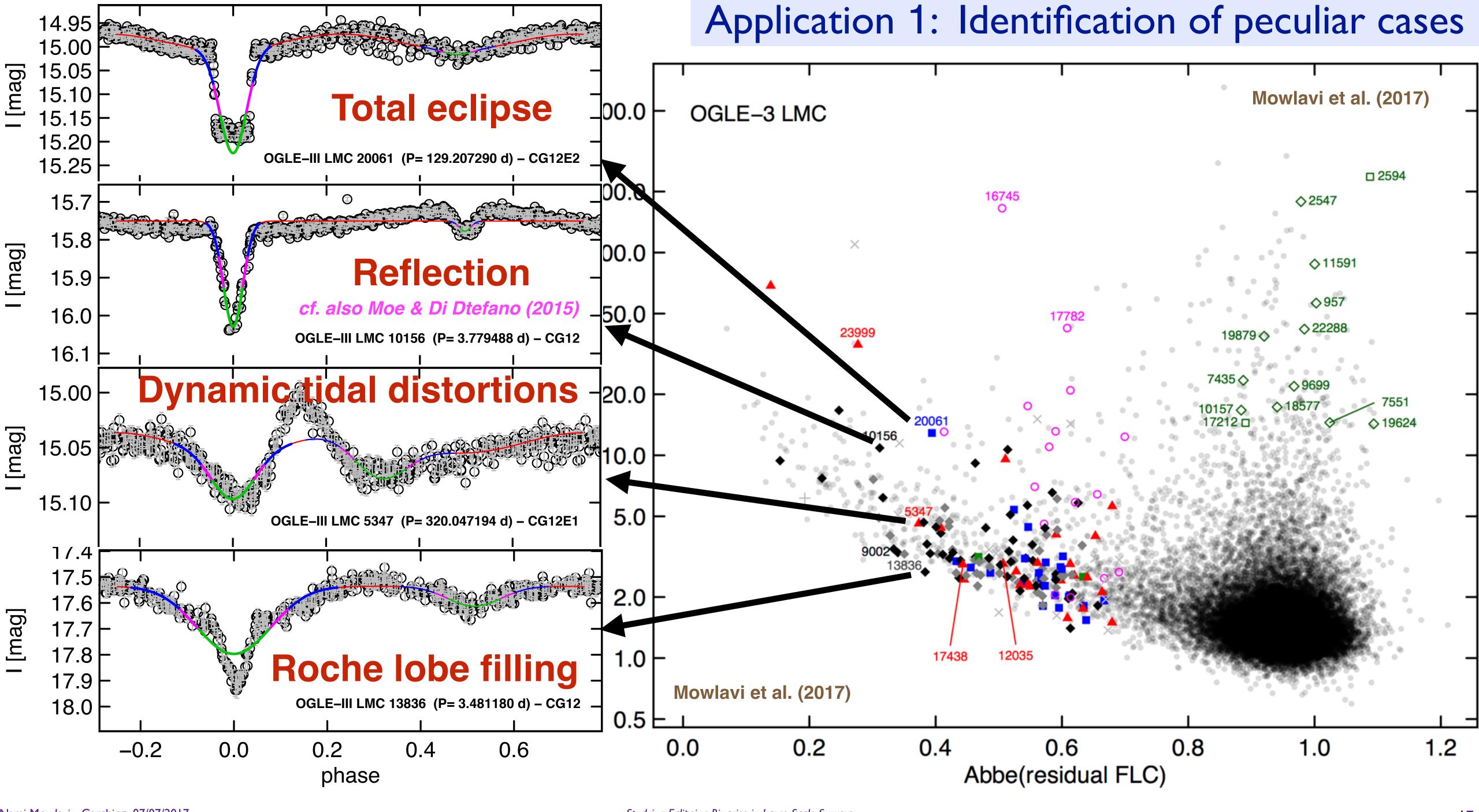
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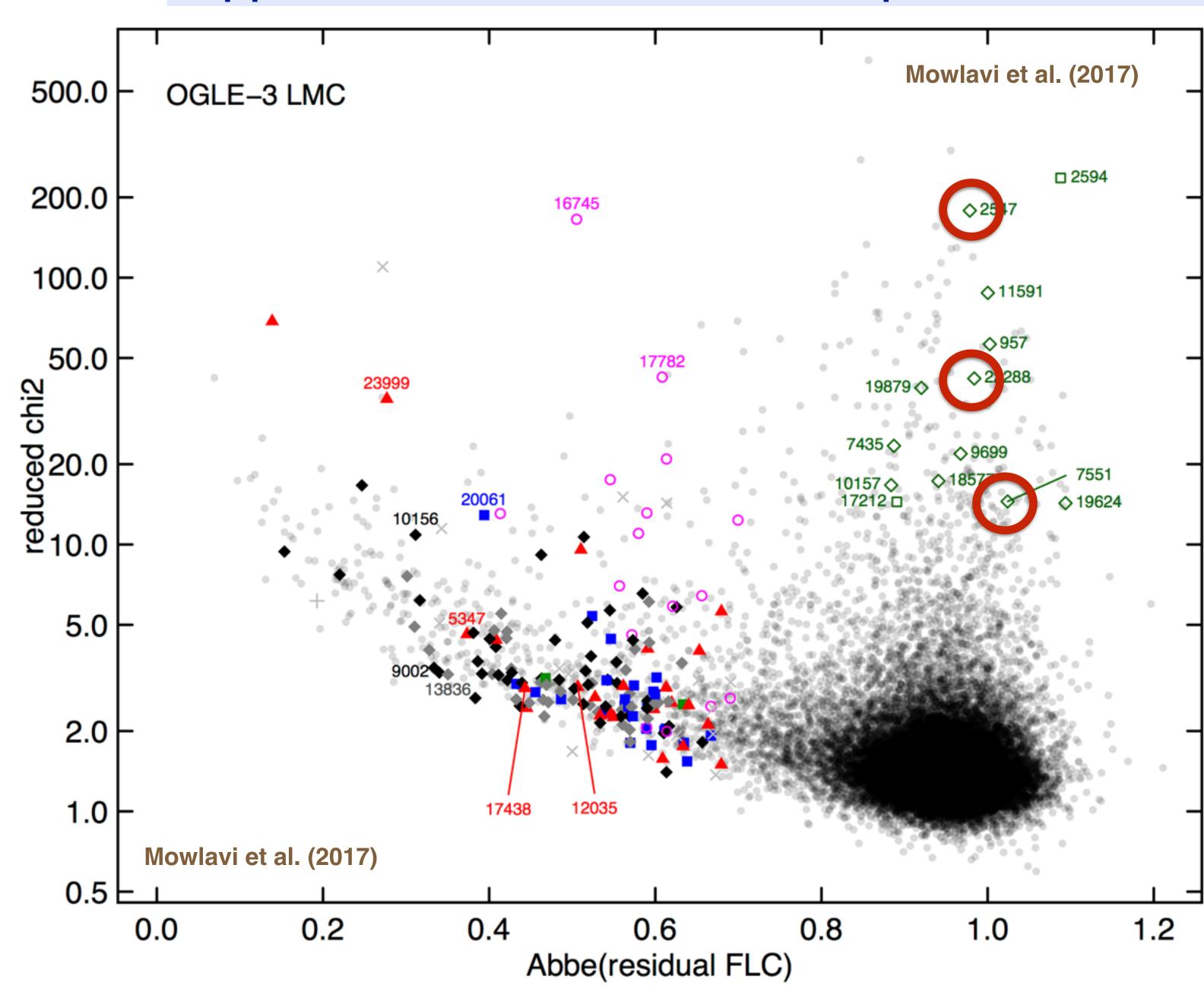
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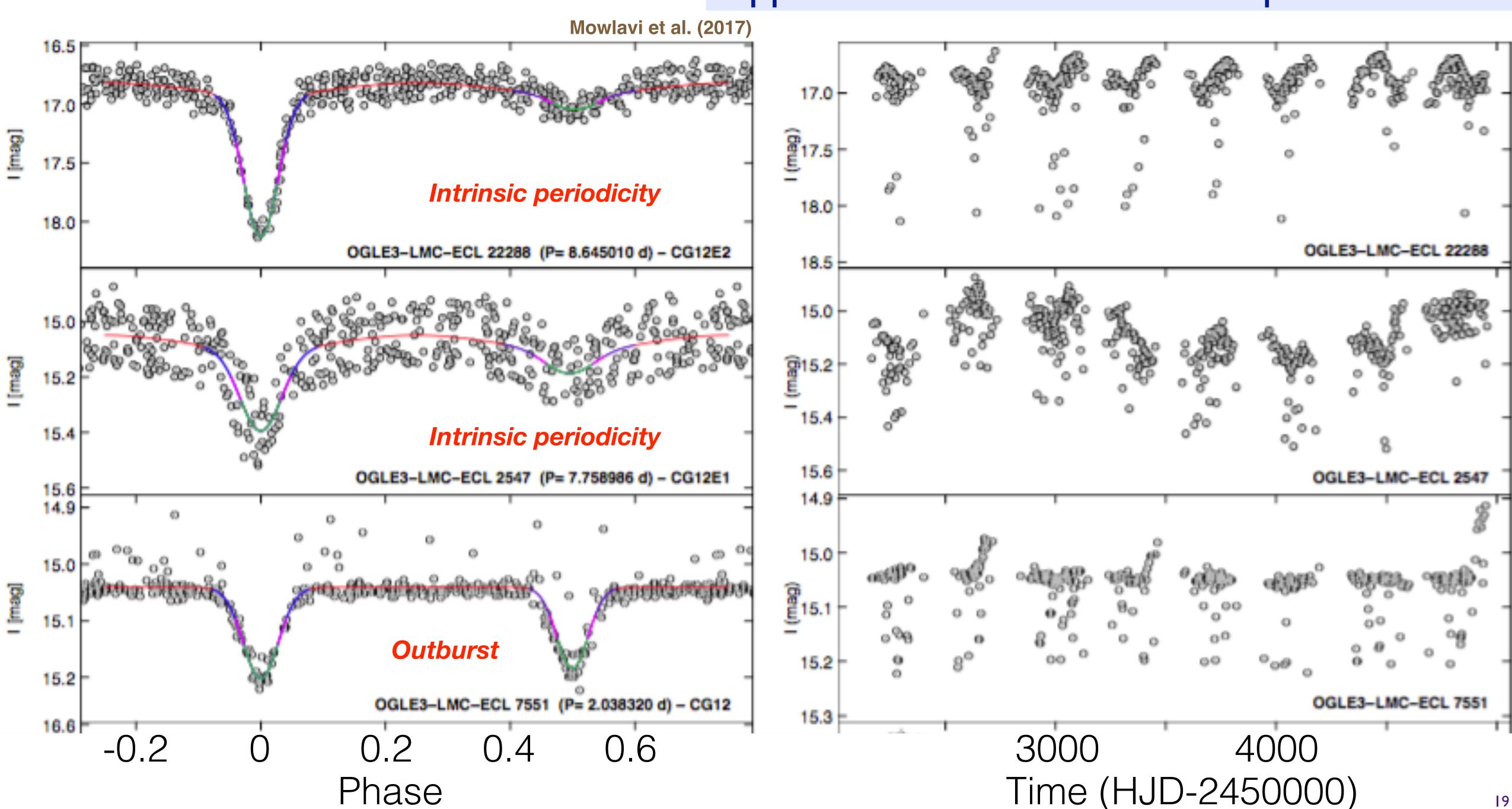
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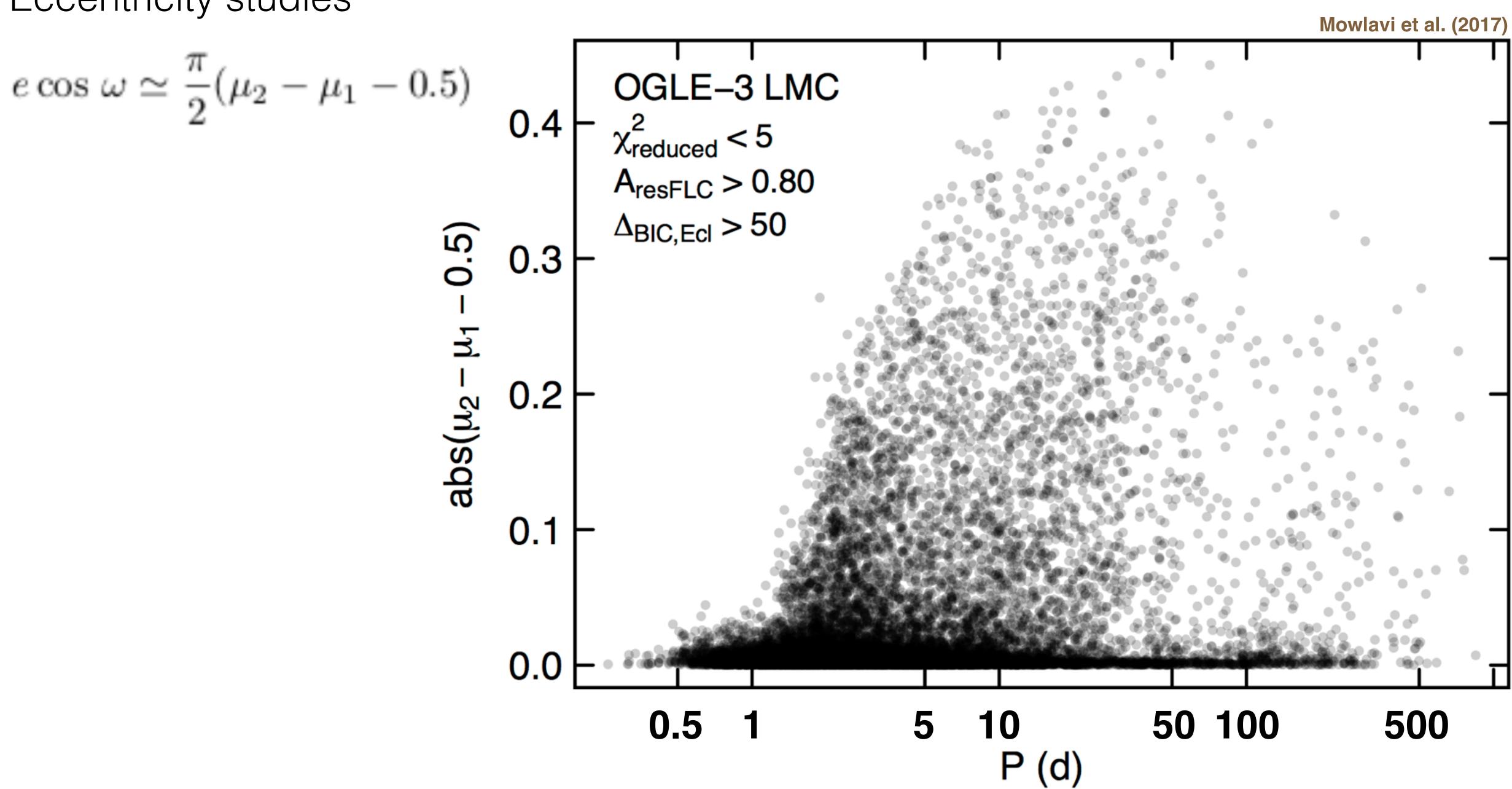






Application 2: Binary population study

Eccentricity studies



Eccentricity studies

$$e\cos\omega \simeq \frac{\pi}{2}(\mu_2 - \mu_1 - 0.5)$$

- Select a given population
 - → MS stars
- Derive eccentricities

$$e \sin \omega = \frac{w_2 - w_1}{w_2 + w_1}$$

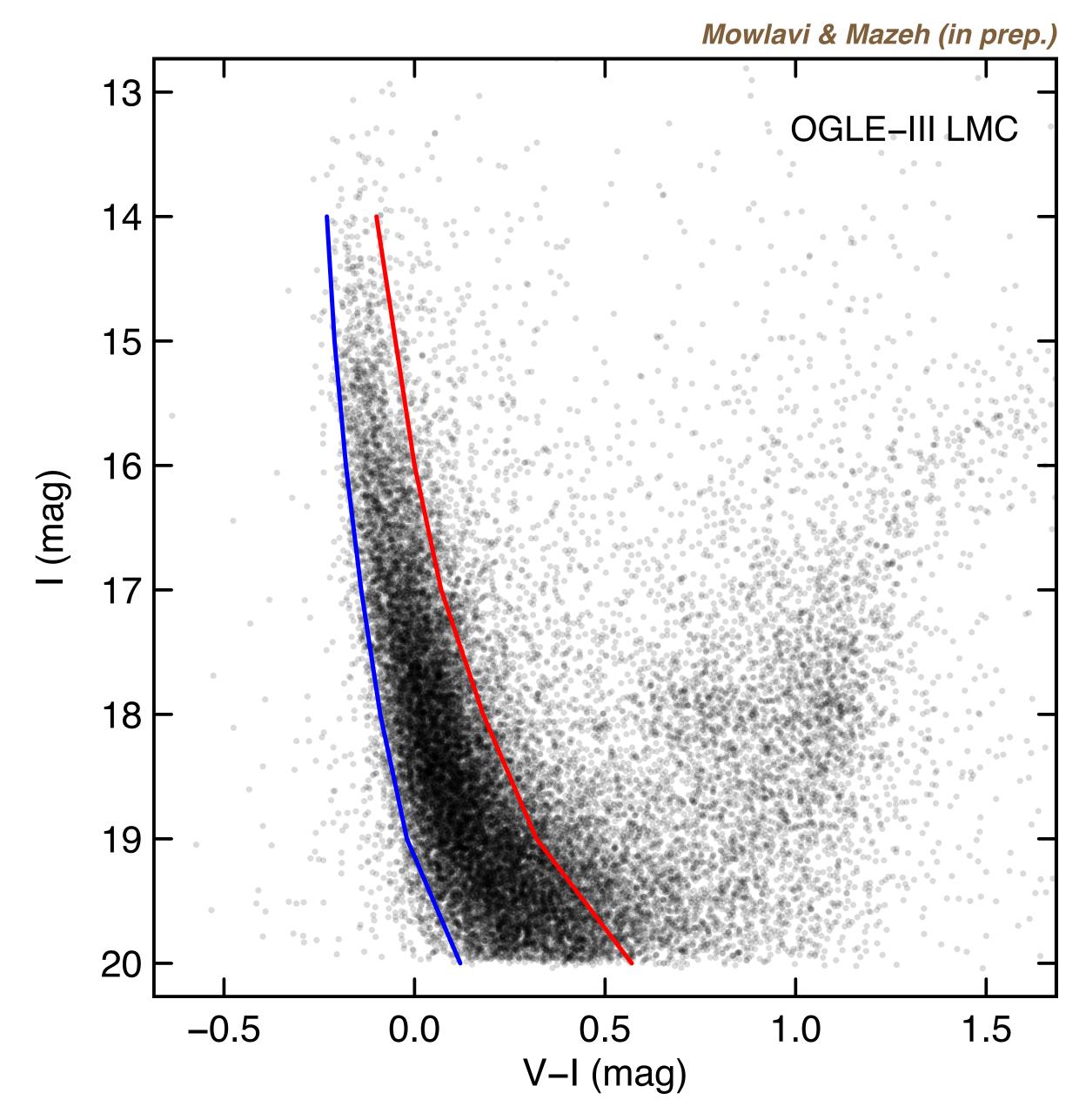
$$\rightarrow e \simeq \left[\frac{\pi^2}{4} (\mu_2 - \mu_1 - 0.5)^2 + \left(\frac{w_2 - w_1}{w_2 + w_1} \right)^2 \right]^{1/2}$$

→ Tidal circularization studies

Work in progress...

cf. also Poster 67 by A. Prša

Application 2: Binary population study



From populations of eclipsing binaries ...

RA (deg)

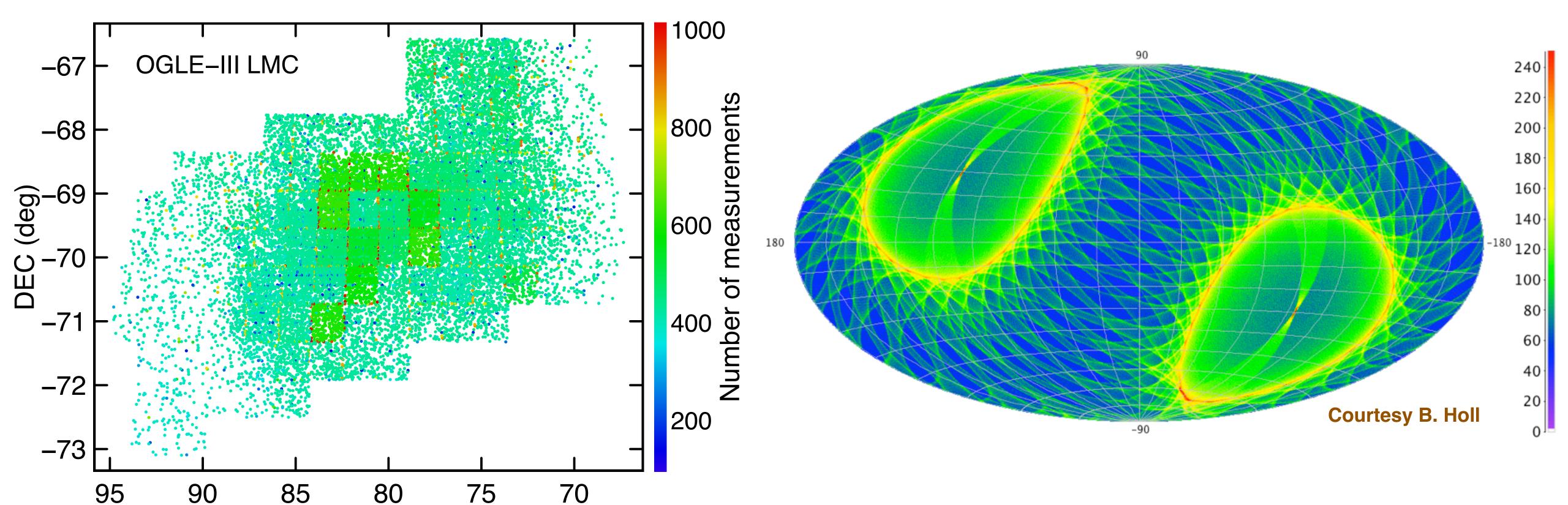
... to populations of binary systems

Selection effects!

For example, ~70% of Kepler EBs potentially detectable with Gaia

(due to time sampling effects, Kochoska et al. 2017)

→ Evaluate transfer functions to convert from pop. of EBs to underlying pop. of binaries



Conclusion

Large scale multi-epoch surveys

→ New perspectives to study binaries/multiples

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Hipparcos: ~0.001 million EBs 1997
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Ogle 3 : ~0.04 million EBs 2013

Ogle 4 : ~0.5 million EBs 2016

Predicted

Gaia : ~4 million EBs 2022 --- + Spectroscopy and astrometry

LSST : ~24 million EBs 2021-2031