

# Probabilistic templates for periodic sources

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Joint work with Sujit Ghosh, Ashish Mahabal, Ana-Maria Staicu,  
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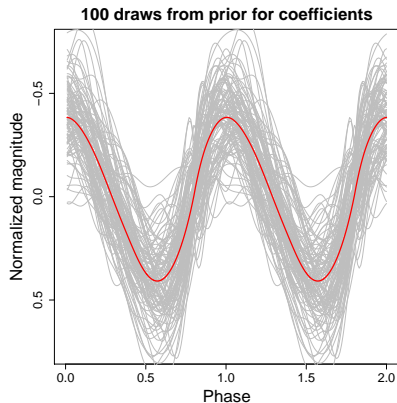
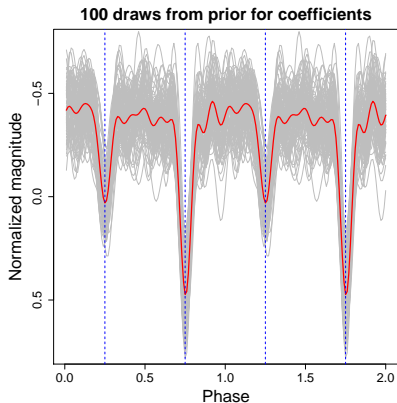
March 20, 2017

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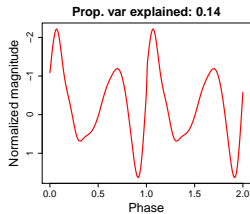
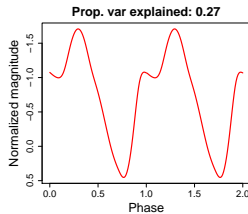
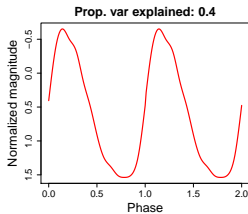
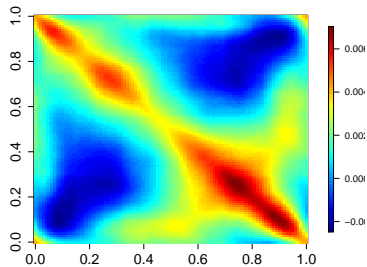
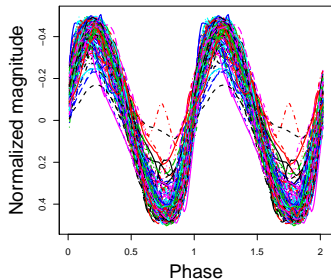
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## Probabilistic templates for periodic sources

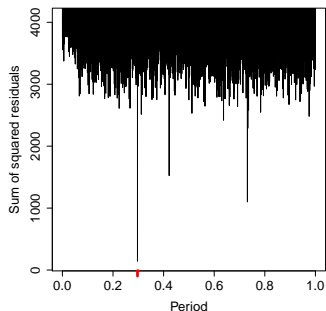
- ▶ Catalina Real-Time Transient Survey data - mainly eclipsing binaries and RR Lyrae
- ▶ Non-parametric Bayesian hierarchical model
- ▶ Fourier and B-splines bases



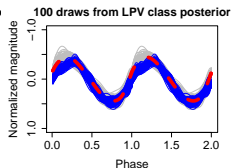
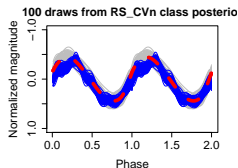
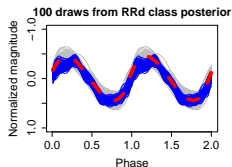
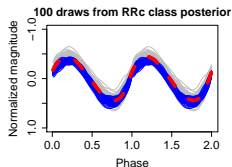
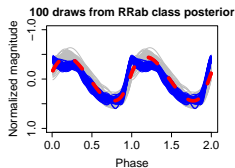
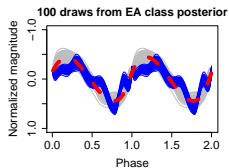
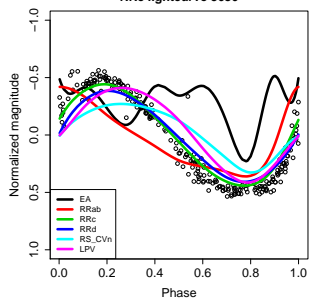
# Probabilistic templates using FDA eigen-functions



# Period finding and classification (ideally joint)



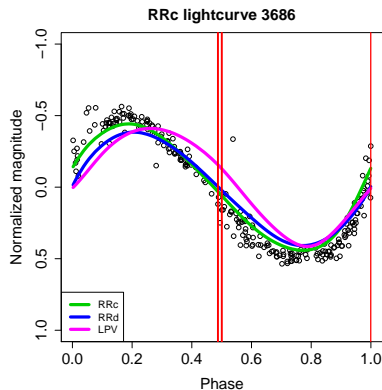
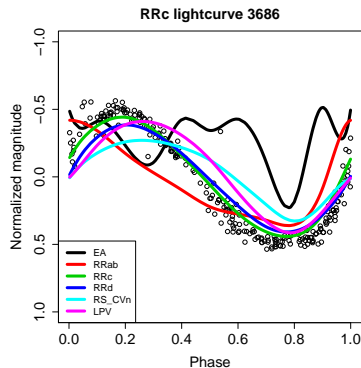
RRc lightcurve 3686



|            | EA   | RRAb | RRc  | RRd  | RS_CVn | LPV  |
|------------|------|------|------|------|--------|------|
| Class prob | 0.00 | 0.00 | 0.83 | 0.17 | 0.00   | 0.00 |

Note: not using period priors

# Scheduling follow-up observations



Full version will find posterior distribution for

$$\hat{t}_{c_1, c_2} = \operatorname{argmax}_{t \in [T_1, T_2]} \left| \sum_{j=0}^{(J-1)} \alpha_{lj}^{(C=c_1)} \psi_j^{(C=c_1)}(t) - \sum_{j=0}^{(J-1)} \alpha_{lj}^{(C=c_2)} \psi_j^{(C=c_2)}(t) \right|$$

- ▶ Use of templates for feature extraction / identification / imputation? Perhaps in multi-band setting.
- ▶ Use of bases mitigates dependence among “features”. Methods for when features are dependent e.g. projections?
- ▶ Feature quality / benefits of reducing to a small number of features?
- ▶ Computational issues? In template setting, FDA approach may reduce the number of basis functions.
- ▶ What if the lightcurve is not from any of the classes?
- ▶ Model misspecification / training set contamination.