

The challenges analysing Gaia Time Series

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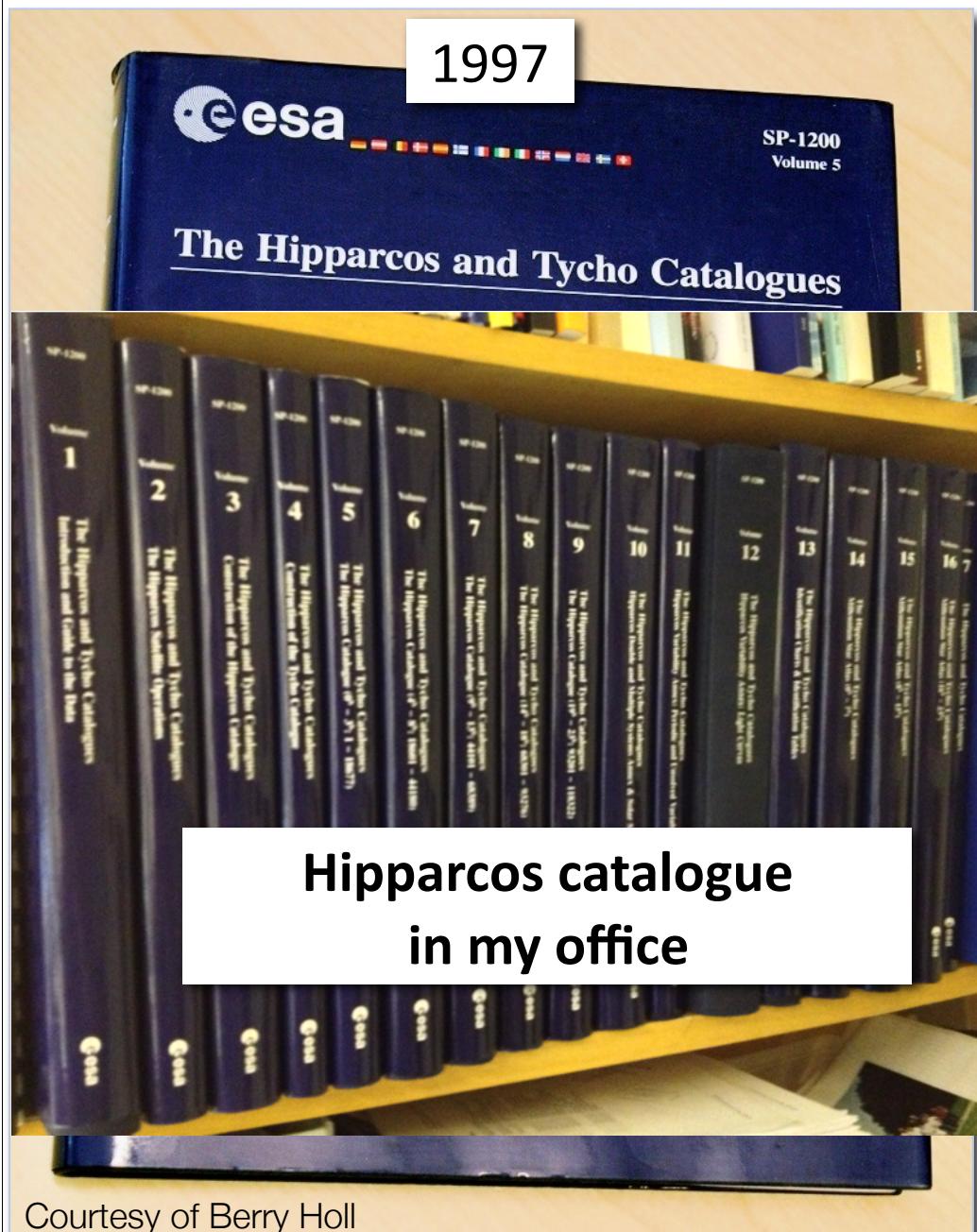
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Statistical Challenges in Modern Astronomy

Carnegie Mellon University, Pittsburgh, USA

Wednesday June 8, 2016

Introduction: Ptolemy vs Hipparchos vs Gaia catalogues



Courtesy of Berry Holl



Almagest
Ptolemy
~150 AD
Here ed. of
1528

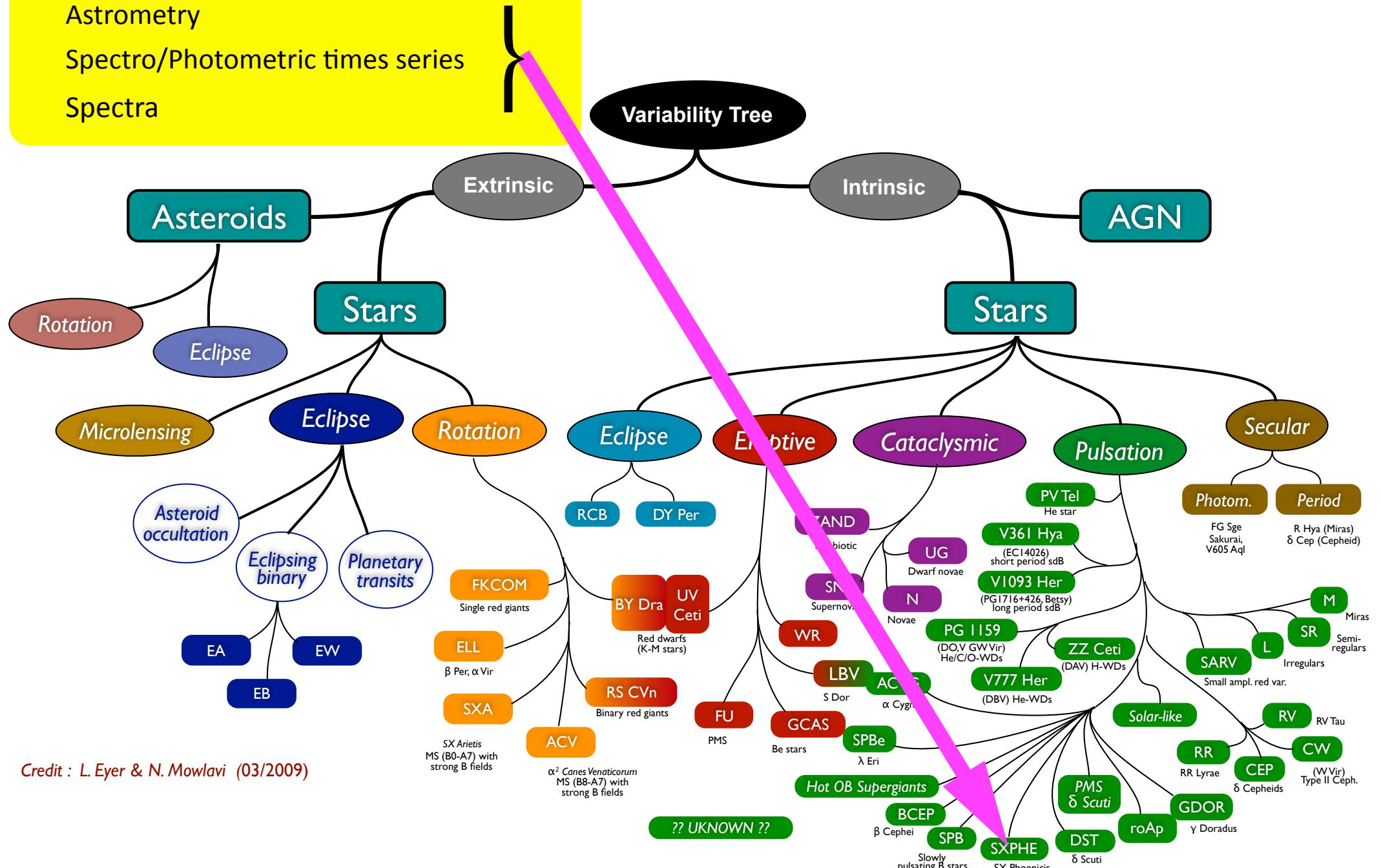
Goal of Variability Processing and Analysis

Gaia measurements for source “i”:

Astrometry

Spectro/Photometric times series

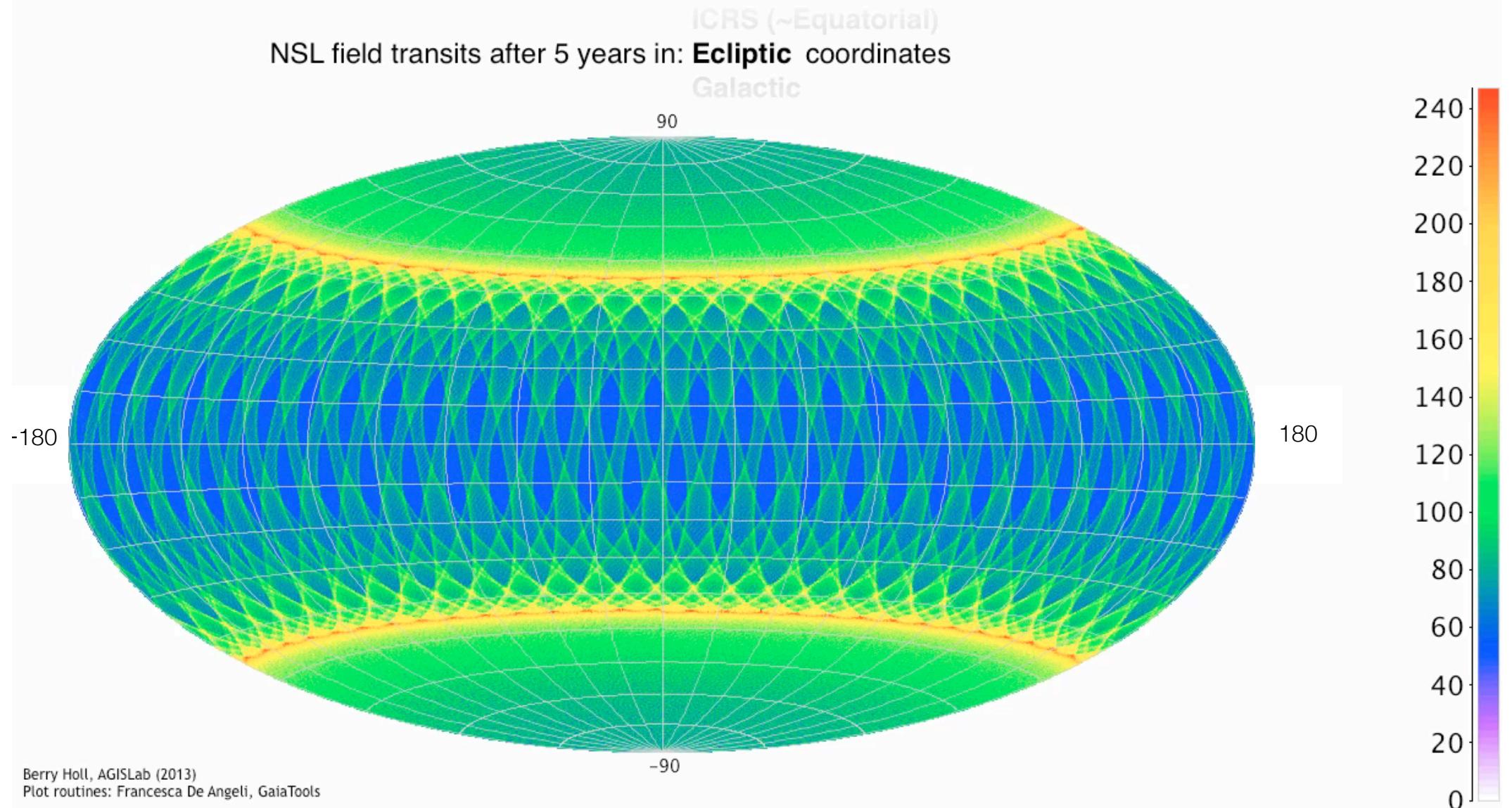
Spectra



Challenges of Gaia in terms of time series analysis

- How to deal with large data set? (e.g. reshuffle data to get time series per source)
- How to gather qualitatively different quantities? (astrometry, photometry, spectra)
- How to handle different number of measurements (from 40 to 250), sparse and different samplings?
- How to deal with heteroscedastic data?
- How to do when there are poor estimates of uncertainties?
- How to search for “small” signals in “noisy” data?
- How to perform model selection?
- How to compute the significance of peaks in a periodogram
- How to deal with the aliasing problem
- How to classify variable objects?
- How do we crossmatch?
- How to rank objects?
- etc...

Five year nominal scanning law (NSL)

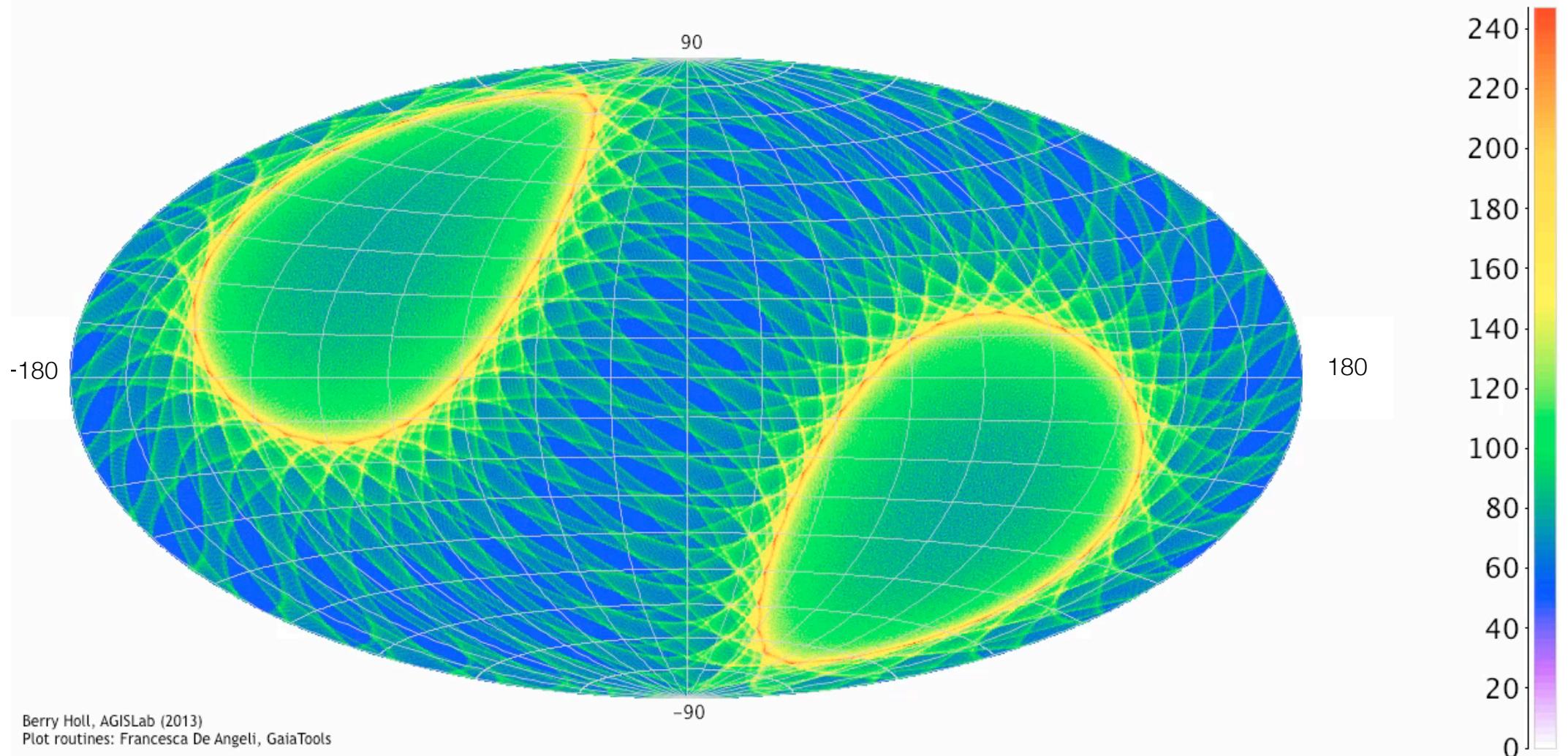


Full scanning law movie on YouTube: <https://www.youtube.com/watch?v=lRhe2grA9wE>

Courtesy of Berry Holl

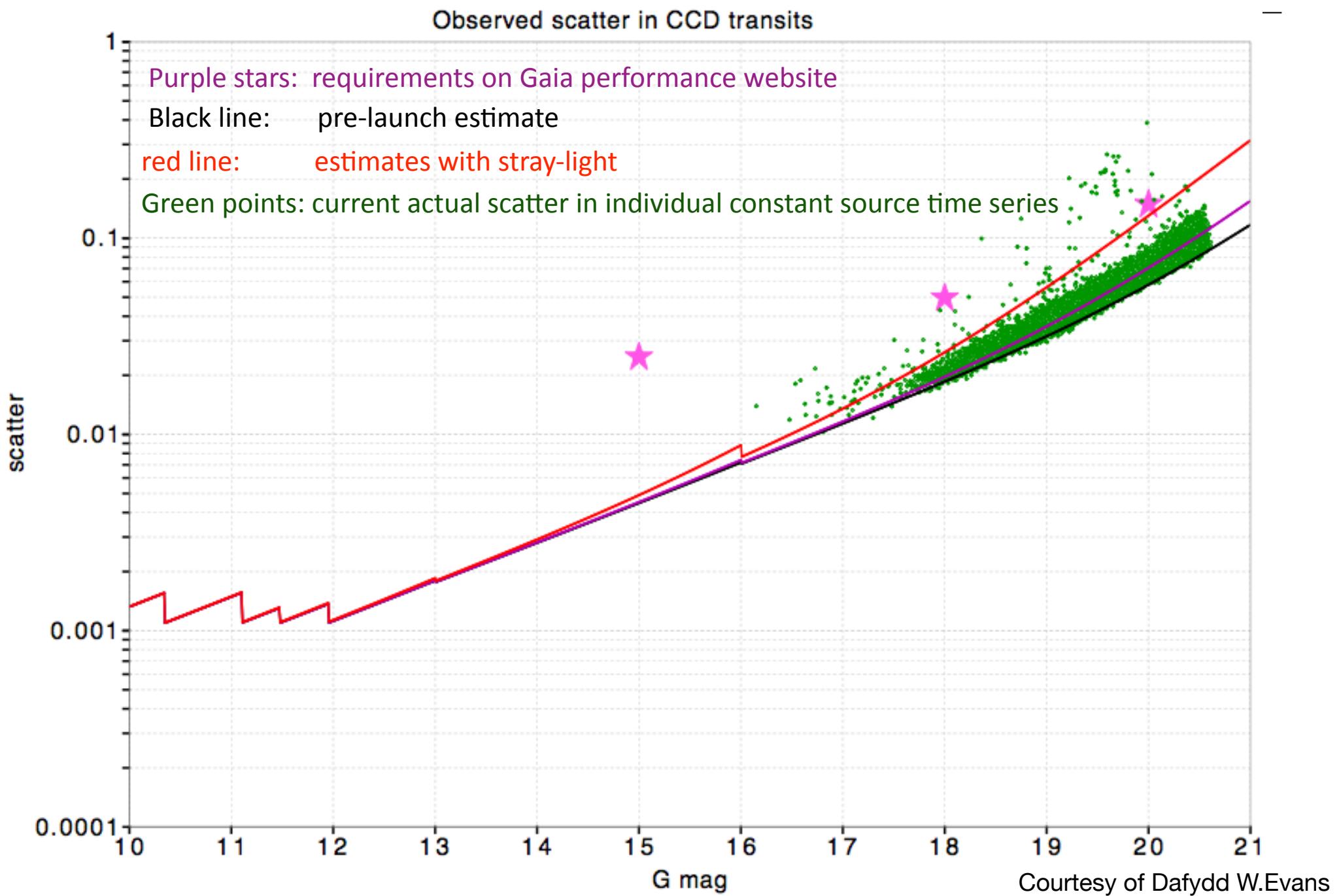
Five year nominal scanning law (NSL)

Ecliptic
NSL field transits after 5 years in: **Galactic** coordinates



Full scanning law movie on YouTube: <https://www.youtube.com/watch?v=lRhe2grA9wE>

Epoch photometric precision



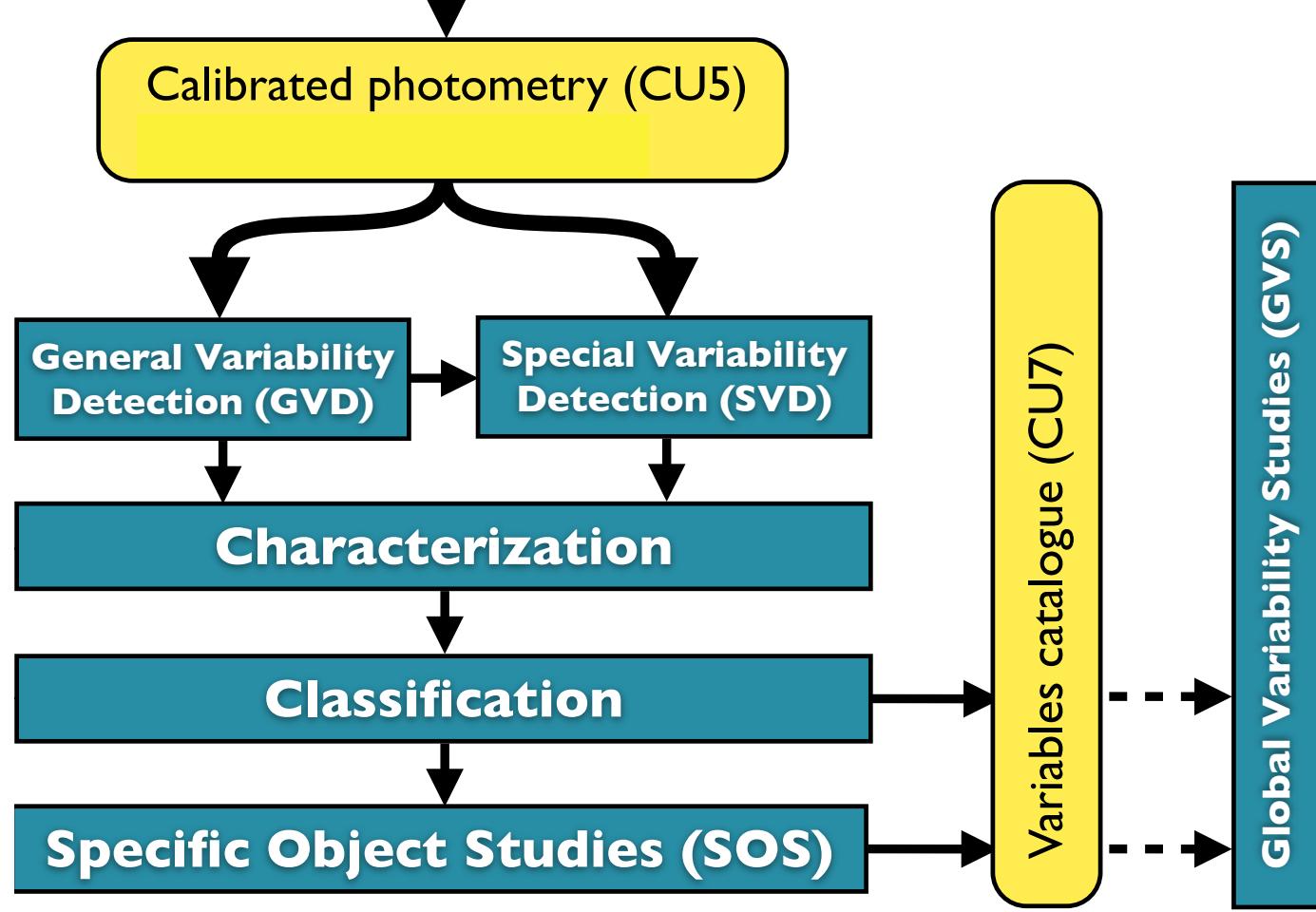
Pragmatic approach

- Develop a “method”
 - Test on simulated data (often very simple)
 - Test on real data
- Apply it to real Gaia data
 - Start on “easy” signal, e.g. large signal to noise ratio
 - Start with law numbers of objects
- Interact with the data (subsamples) at all levels of the analysis

Gaia Variability Processing and Analysis

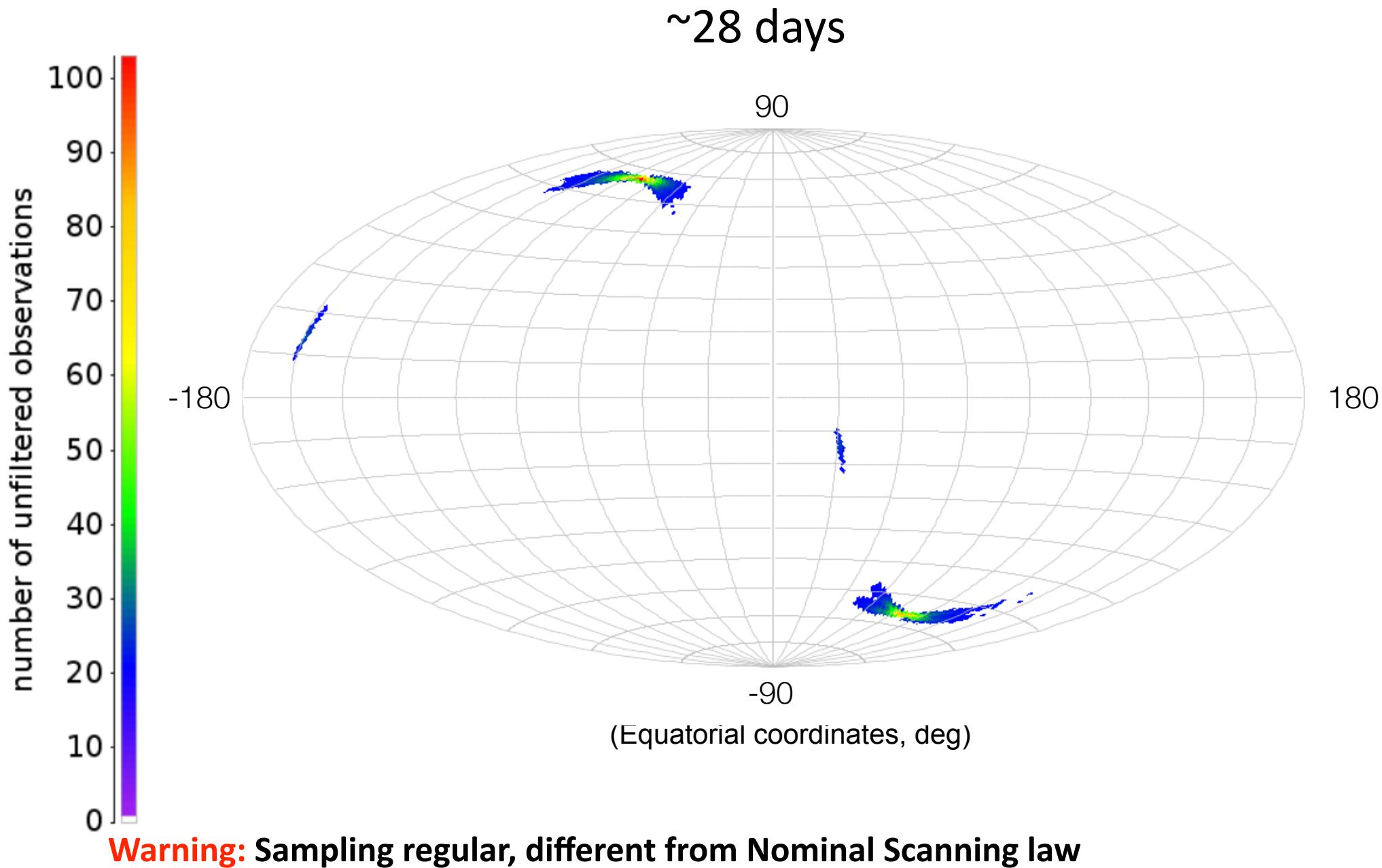
1 billion sources observed by Gaia
790,000 REAL Gaia data from Ecliptic Poles

Time series of 70 (40-250) measurements over 5 years
Time series up to 170 measurements over 28 days

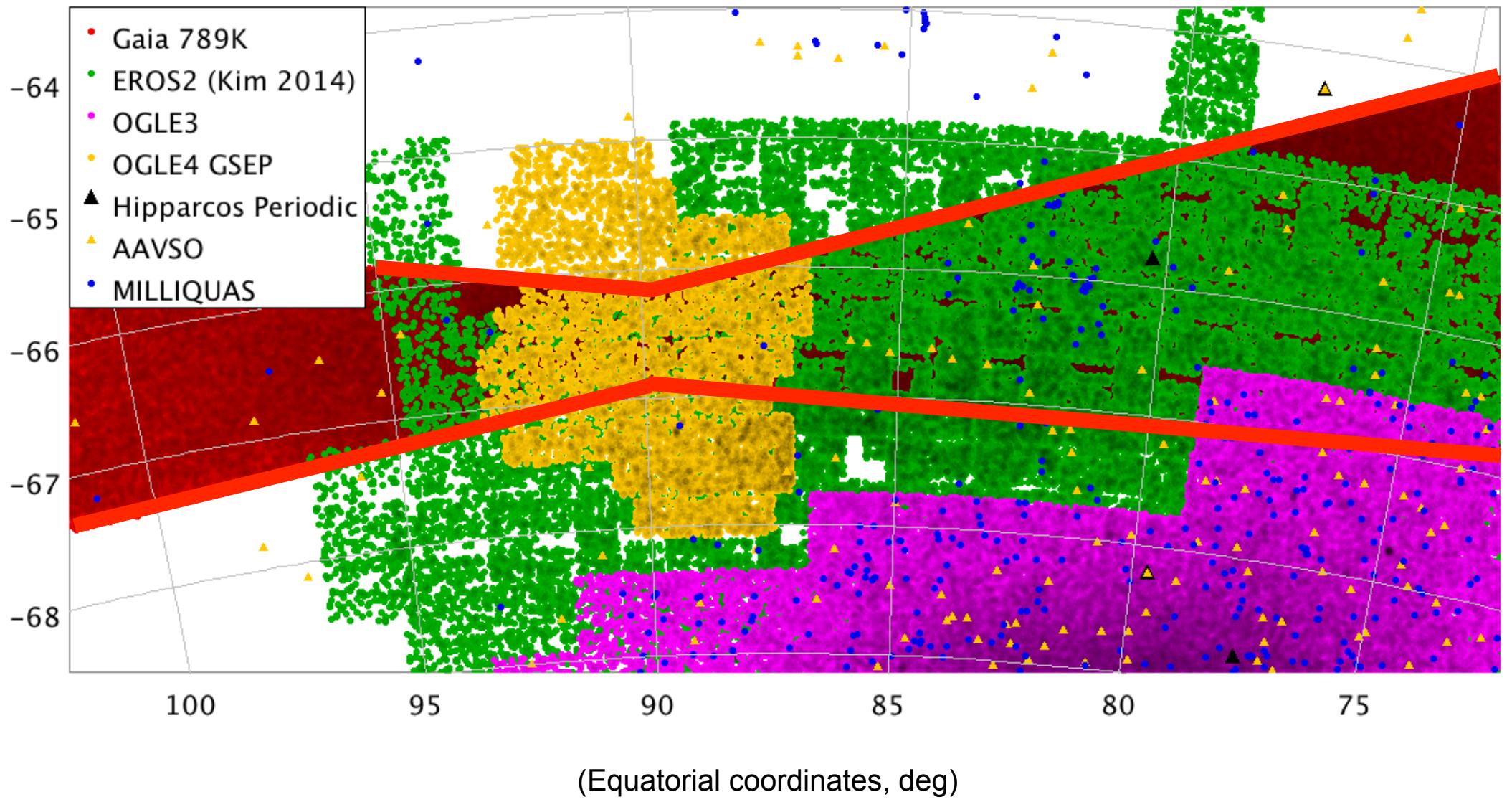


Unexpected Features Analyses

Selection of sources observed by Gaia from Ecliptic Pole Scanning Law (790,000 sources)



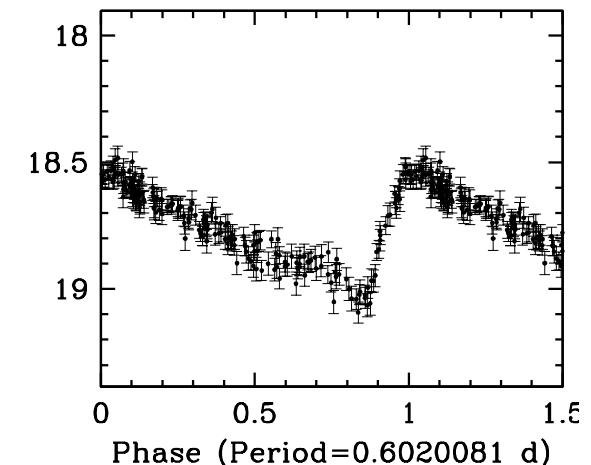
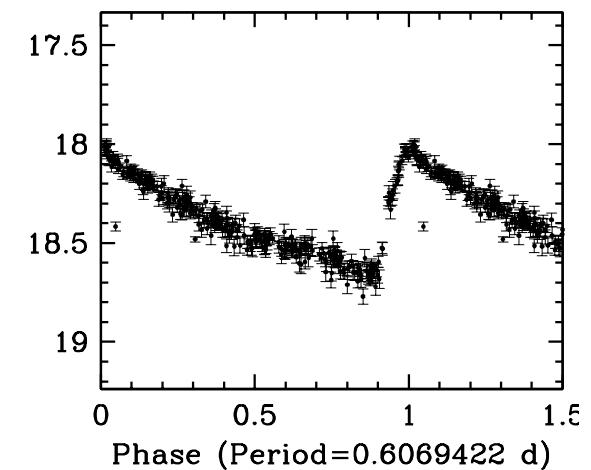
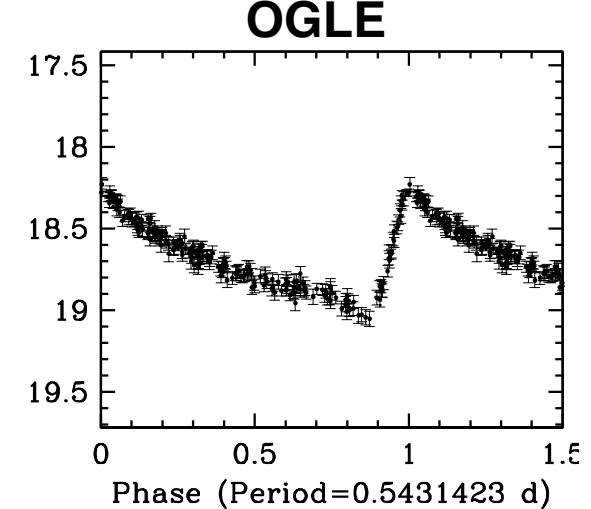
South Ecliptic Pole region (part of Large Magellanic Cloud): Gaia and other surveys



To get the data flavour Comparison with OGLE

Image of the Week (March 05, 2015):
RR Lyrae stars

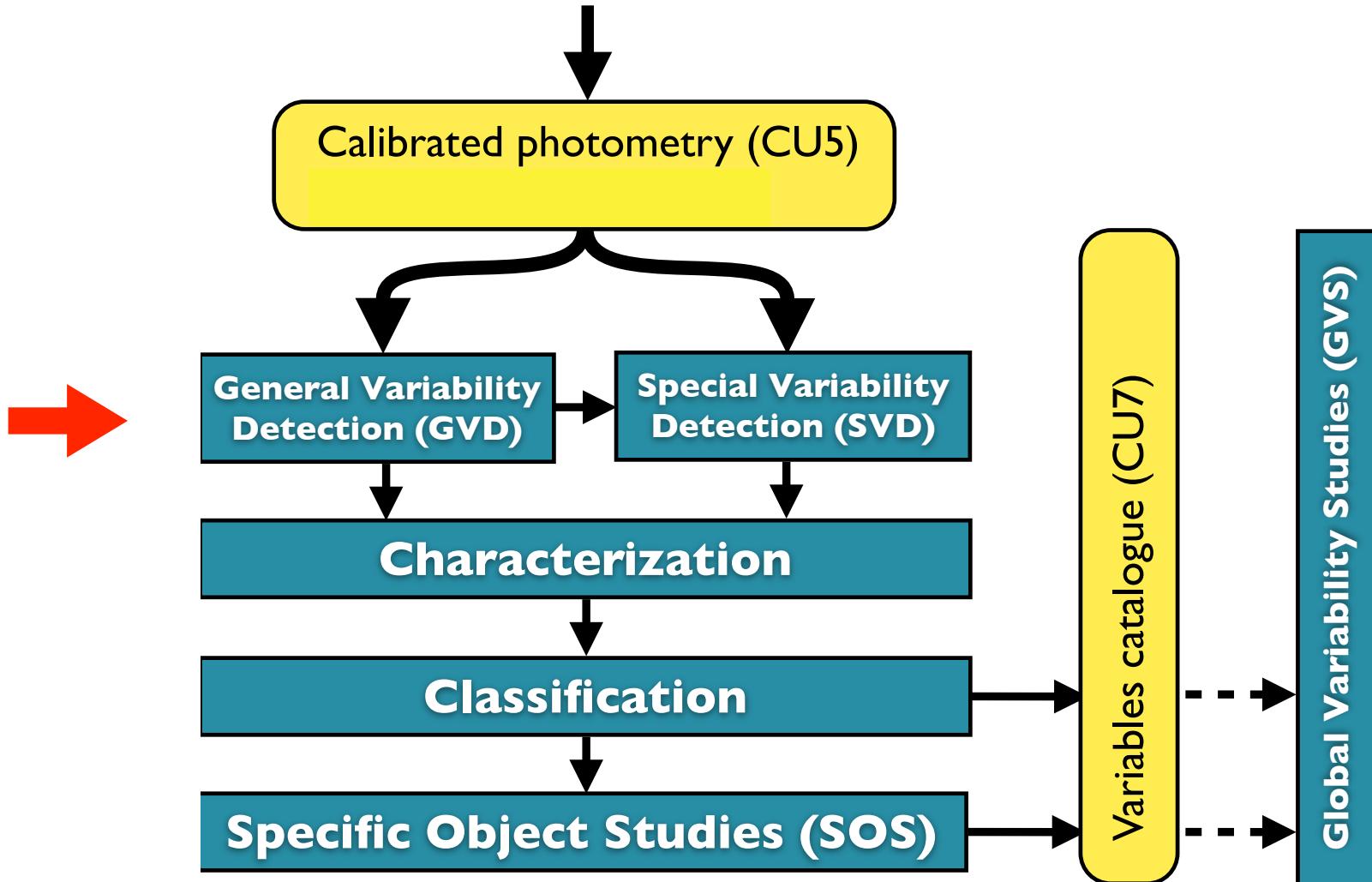
Credits: ESA/Gaia/DPAC/CU5/CU7/INAF-OABo, Gisella Clementini, Dafydd Evans, Laurent Eyer, Krzysztof Nienartowicz, Lorenzo Rimoldini and the Geneva CU7/DPCG and CU7/INAF-OACN teams.



Gaia Variability Processing and Analysis

790,000 REAL Gaia data from Ecliptic Poles

Time series up to 170 measurements over 28 days



General Variability Detection

Classical hypothesis testing can't be applied because of "poor" estimates of uncertainties

Detection was done with a classifier (Random Forest)
attributes were computed
a training set was defined (based on OGLE)

Two fundamental quantities to estimate:

- Completeness
- Contamination

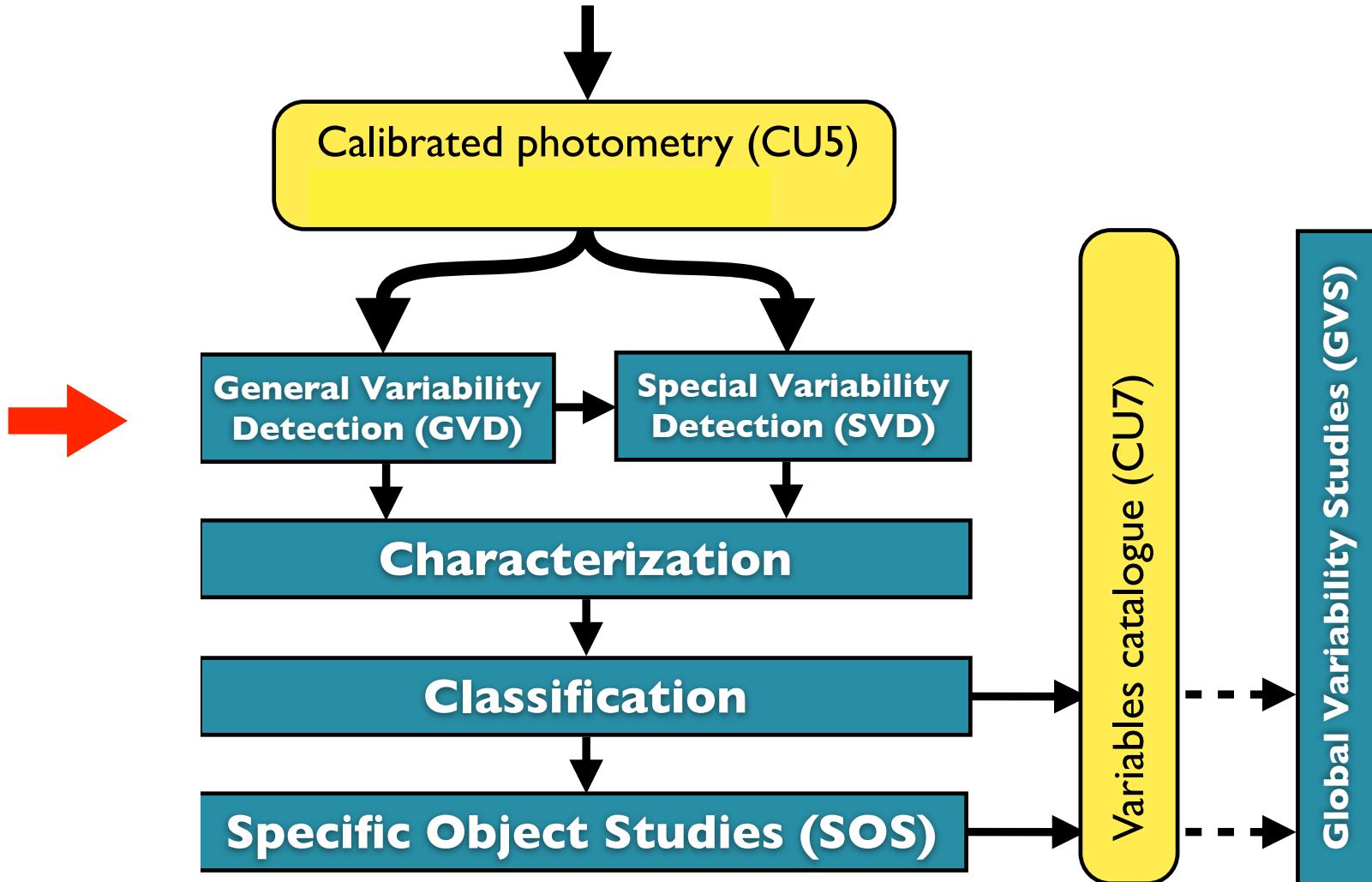
Classifier result: The confusion matrix

| | | VARIABLE | CONSTANT |
|---------------|----------|----------|----------|
| 376 | VARIABLE | 80 | 20 |
| 546 | CONSTANT | 5 | 95 |
| Contamination | | 8 | 13 |

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Characterisation

Time series per object:

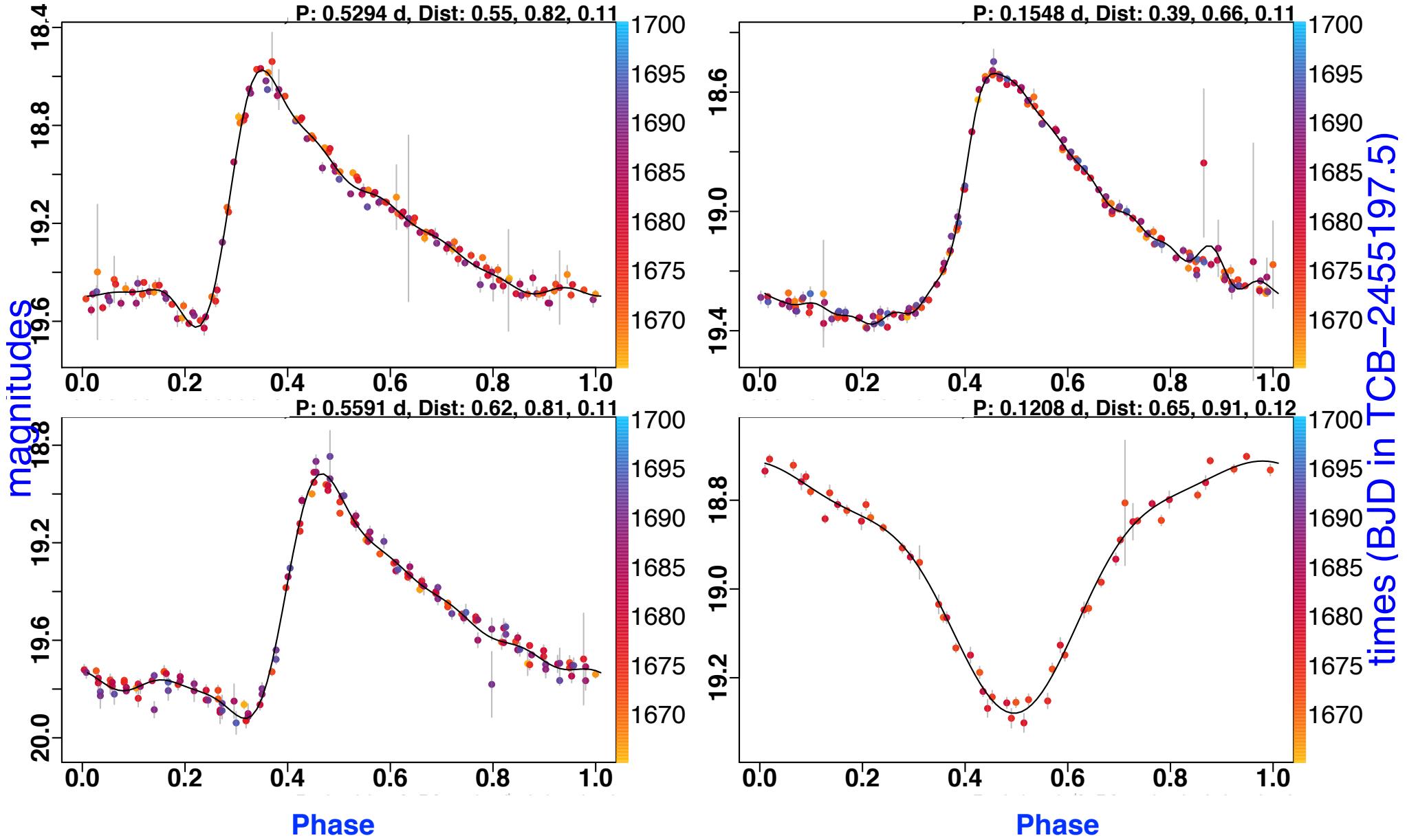
Time_(i), G-, BP-, RP- mag_(i) [or radial velocity_(i)]

i=1,..., number of measurements

Goal: To define attributes

- statistical parameters
- Modelling
 - Period search
 - Fourier Series and polynomial fit

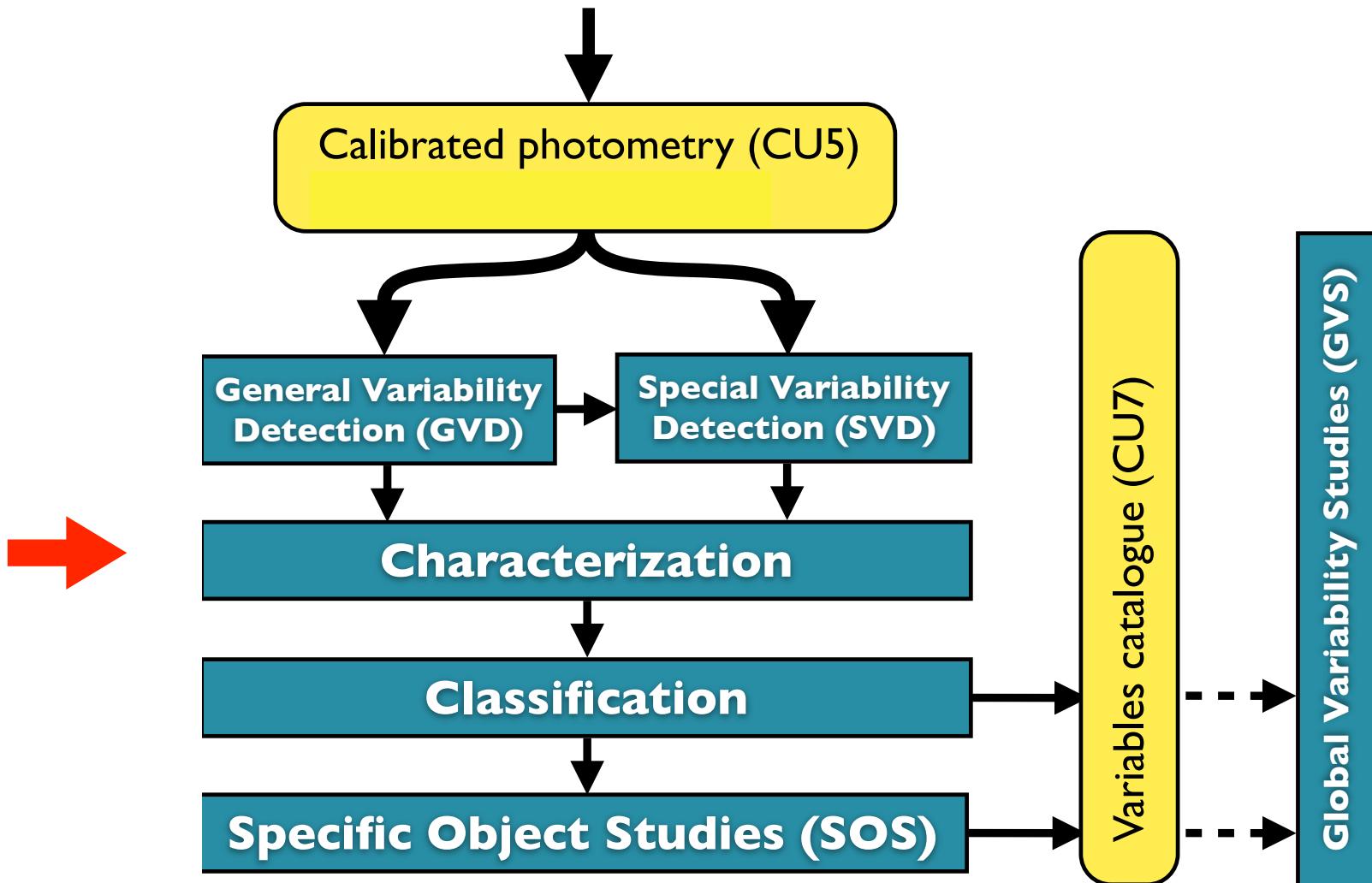
Characterisation: few examples of modelling



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Classification

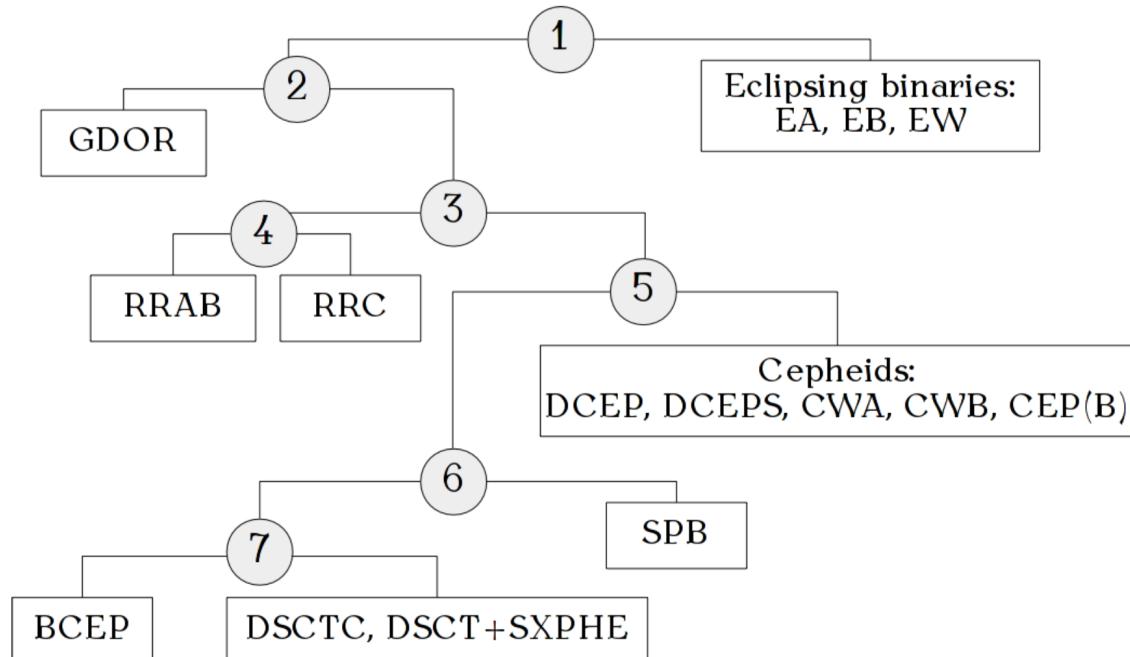
Supervised classification (several methods):

Multistage tree:
Bayesian networks

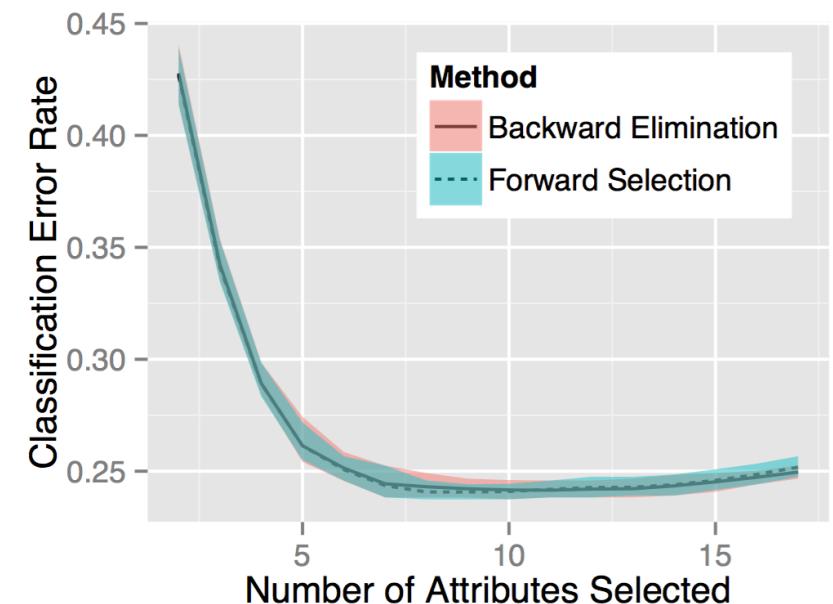
Multistage tree:
Gaussian mixture

Random Forest

Tree for Gaussian Mixture:



Furnish training set
built from Crossmatched data



Classification

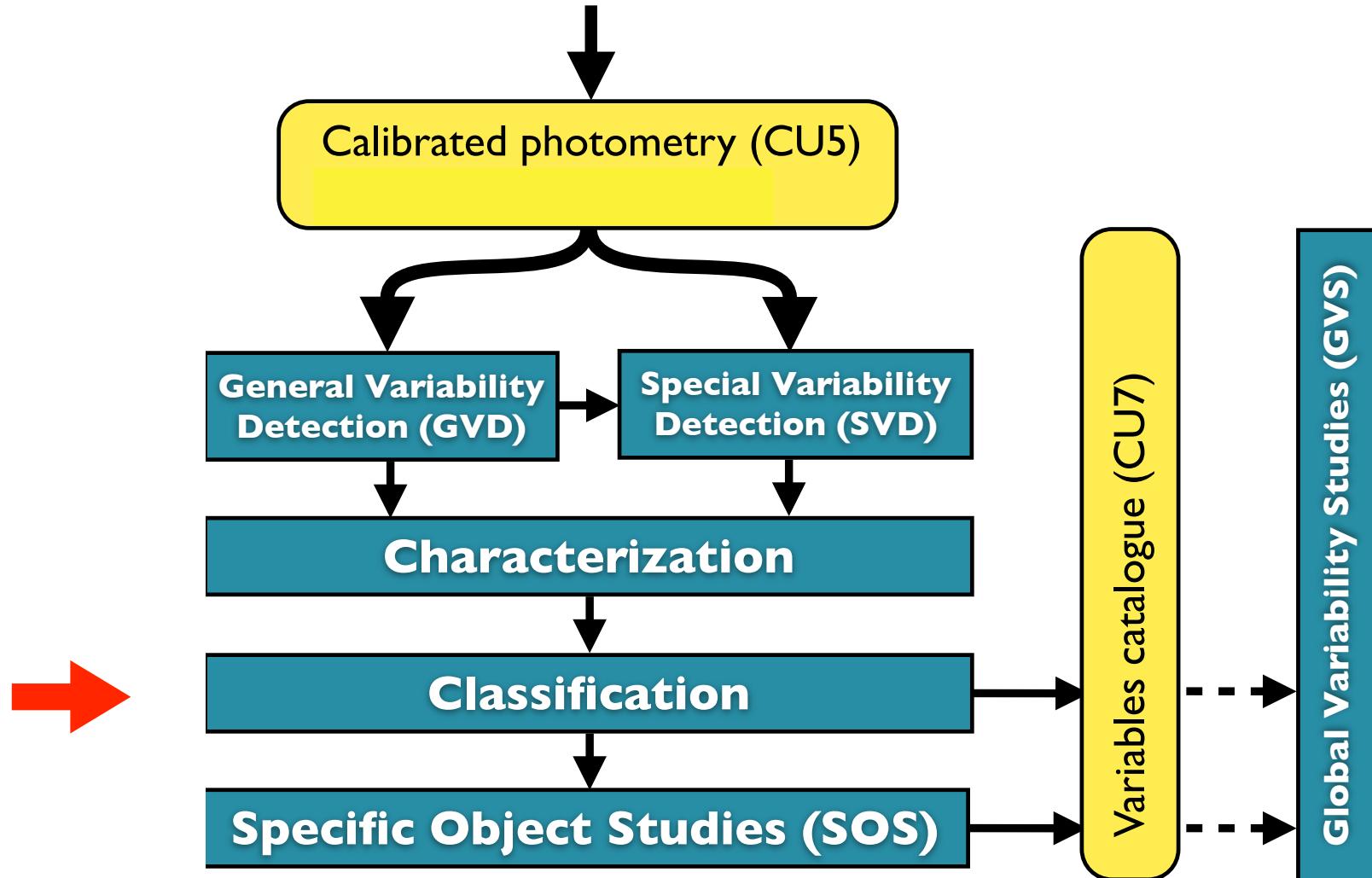
Confusion matrix of Random Forest
using cross-matched data (OGLE, Hipparcos, AAVSO, Milliquas)

| | | CONSTANT | QSO | ECL | OTHER | RRLYR | LPV | ELL | DSCT | DCEP | UG |
|---------------|----------|----------|-----|-----|-------|-------|-----|-----|------|------|----|
| 103 | CONSTANT | 79 | 8 | 5 | 9 | | | | | | |
| | QSO | 7 | 88 | 2 | 3 | | | | | | |
| 100 | ECL | 4 | 3 | 84 | 7 | 1 | 1 | 1 | | | |
| 135 | OTHER | 13 | 4 | 12 | 69 | | 2 | 1 | | | |
| 108 | RRLYR | 3 | | 1 | 2 | 94 | | | | | |
| 27 | LPV | 30 | 19 | 7 | 30 | | 15 | | | | |
| 8 | ELL | 25 | 12 | | 62 | | | | | | |
| 7 | DSCT | 14 | | 29 | 29 | 29 | | | | | |
| 2 | DCEP | 50 | | 50 | | | | | | | |
| 2 | UG | | | | | 100 | | | | | |
| Contamination | | 34 | 20 | 19 | 34 | 5 | 43 | 100 | - | - | - |

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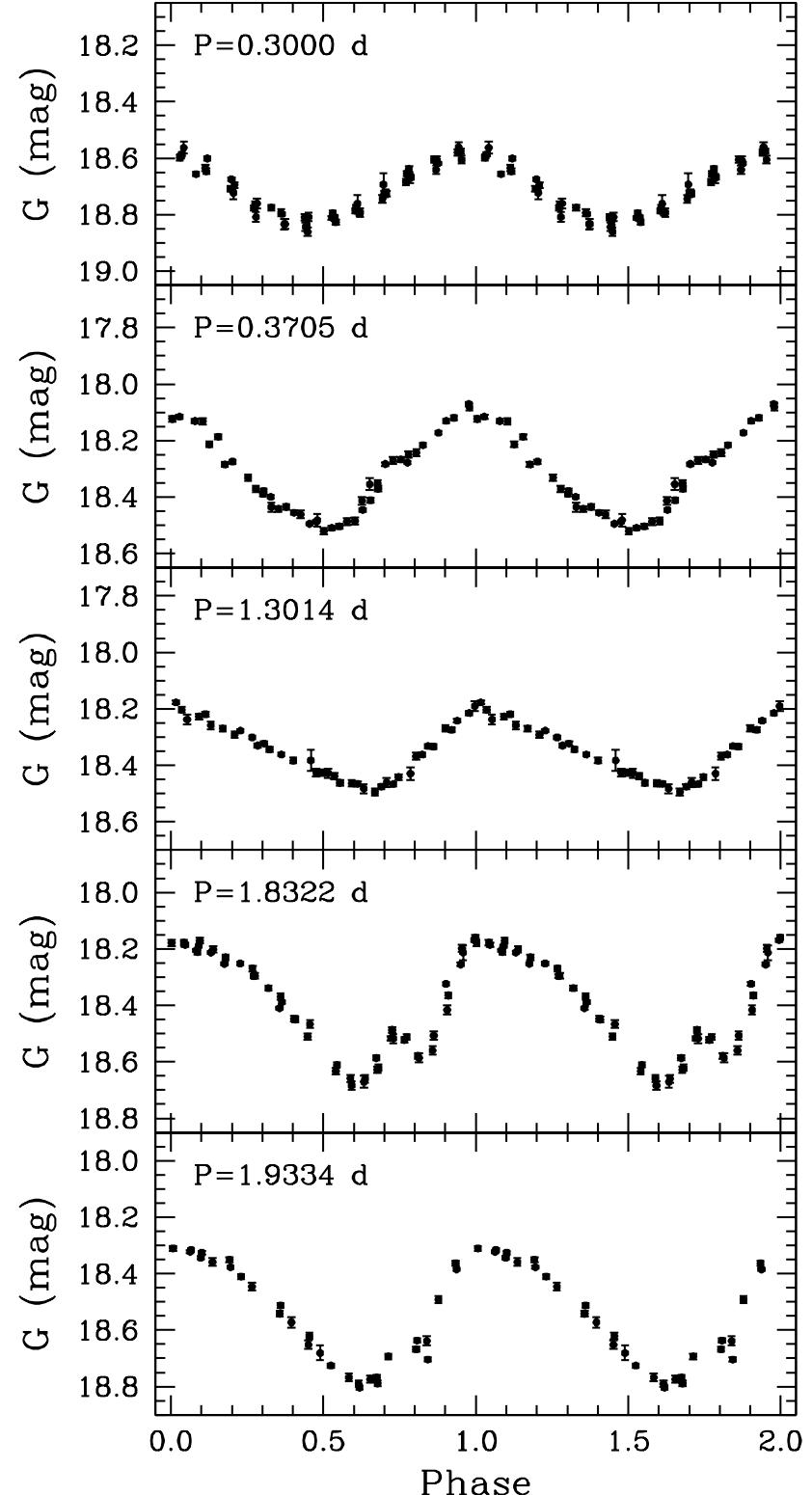


Classification of RR Lyrae and Cepheid stars

Gisella Clementini, Silvio Leccia, Vincenzo Ripepi, Nami Mowlavi, Isabelle Lecoeur

Classical overtone Cepheid
3 candidate anomalous Cepheids
Type 2 Cepheid

Credits: *ESA/Gaia/DPAC/CU5/DPCI/CU7/INAF-OAB/INAF-OACn Gisella Clementini, Vincenzo Ripepi, Silvio Leccia, Laurent Eyer, Lorenzo Rimoldini, Isabelle Lecoeur-Taibi, Nami Mowlavi, Dafydd Evans, Geneva CU7/DPCG and the whole CU7 team. The photometric data reduction was done with the PhotPipe pipeline at DPCI; processing data were received from the IDT pipeline at DPCE.*

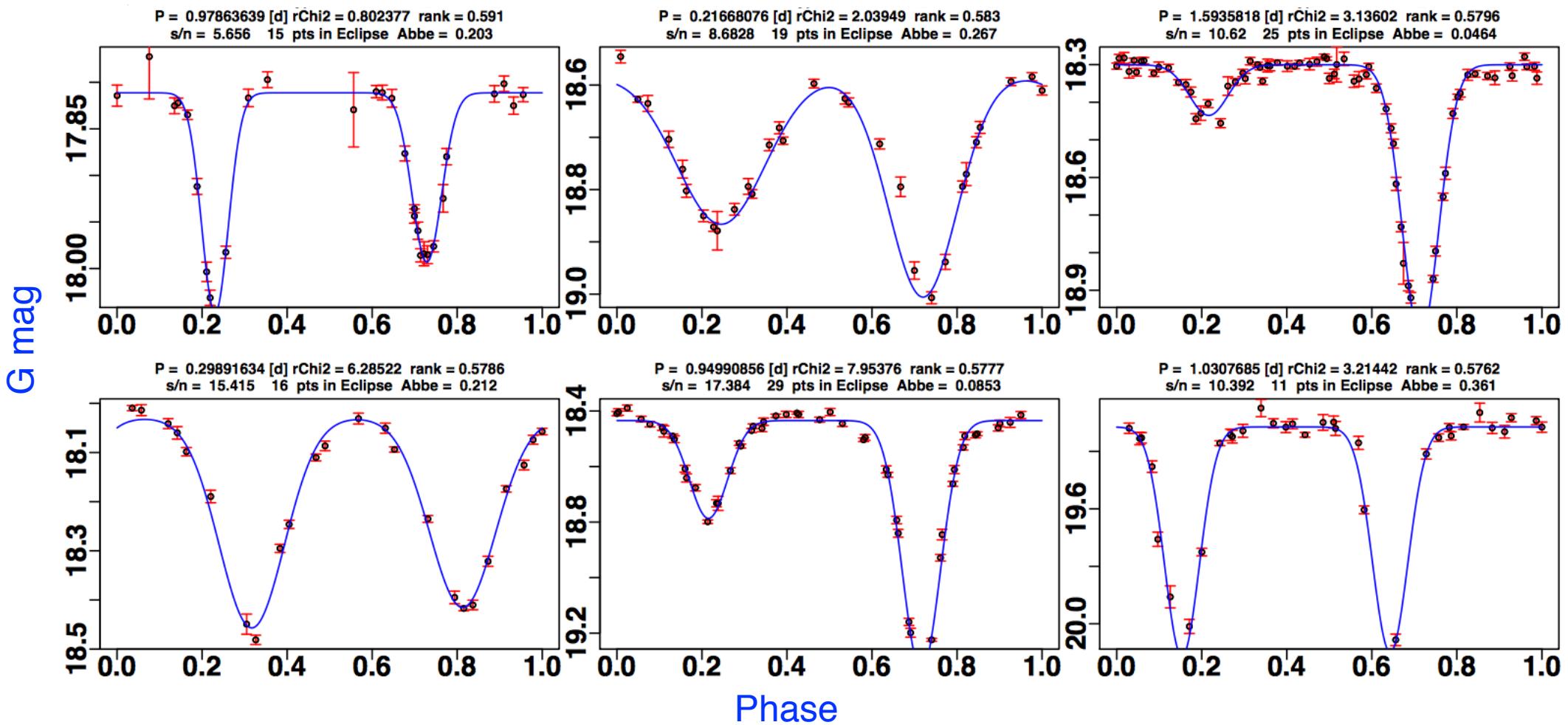


Specific Object Studies: Eclipsing binaries

Eclipsing binaries go to a dedicated treatment (Université Libre de Bruxelles) for a full modelling

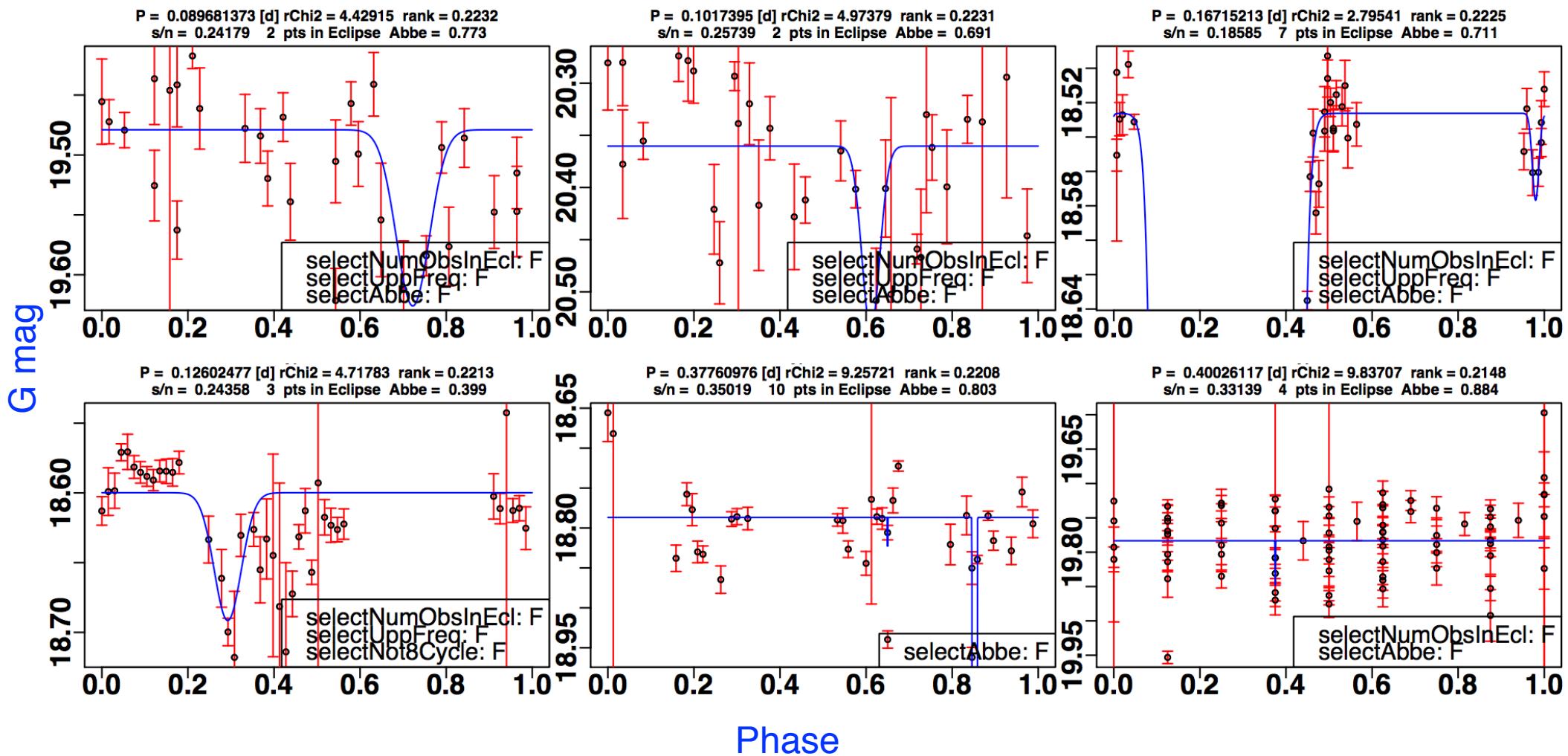
Here, simple modelling (fit with “two Gaussians”) are made
The solutions enable a ranking

Highest rank



Specific Object Studies: Eclipsing binaries

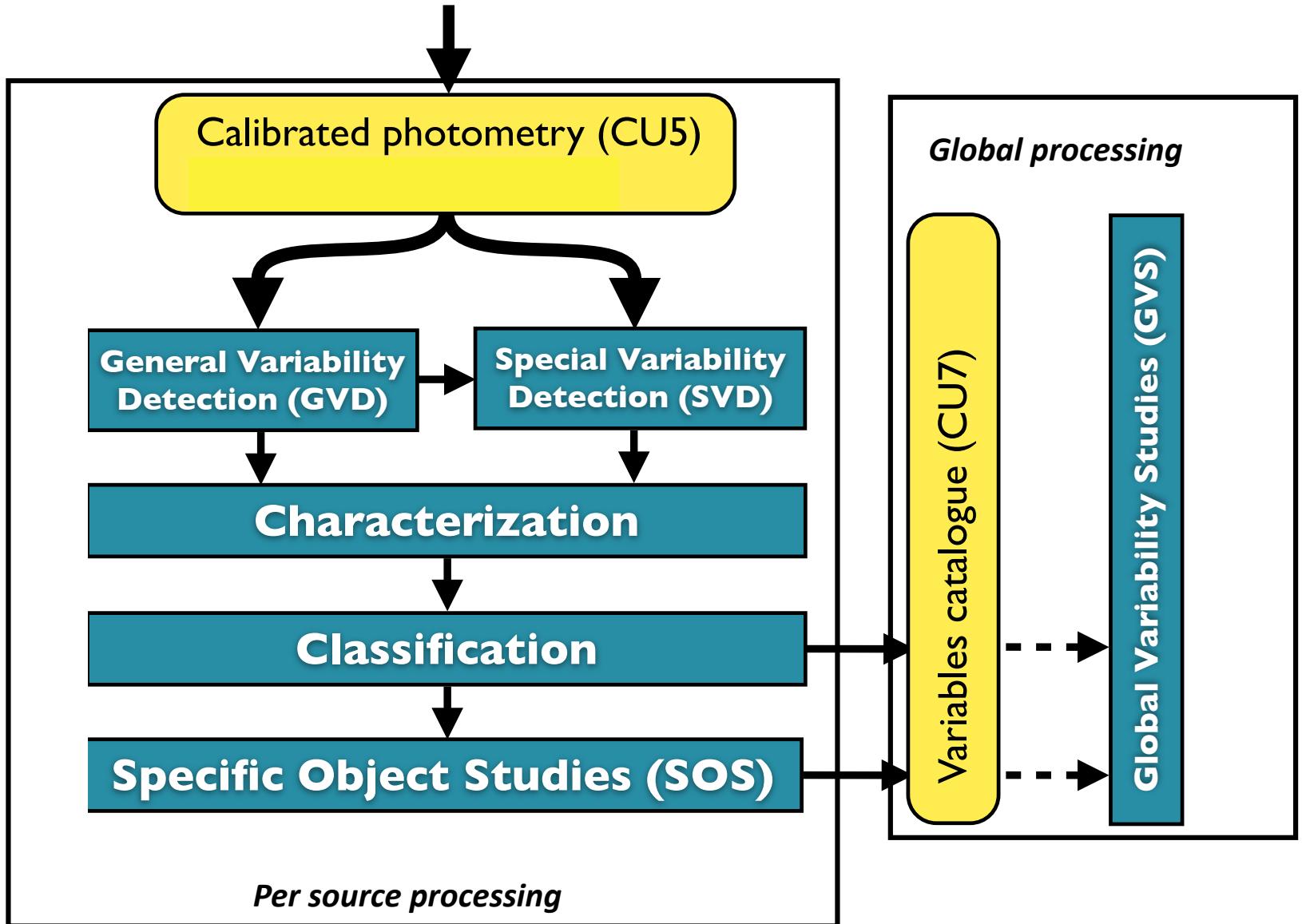
Lowest rank



Gaia Variability Processing and Analysis

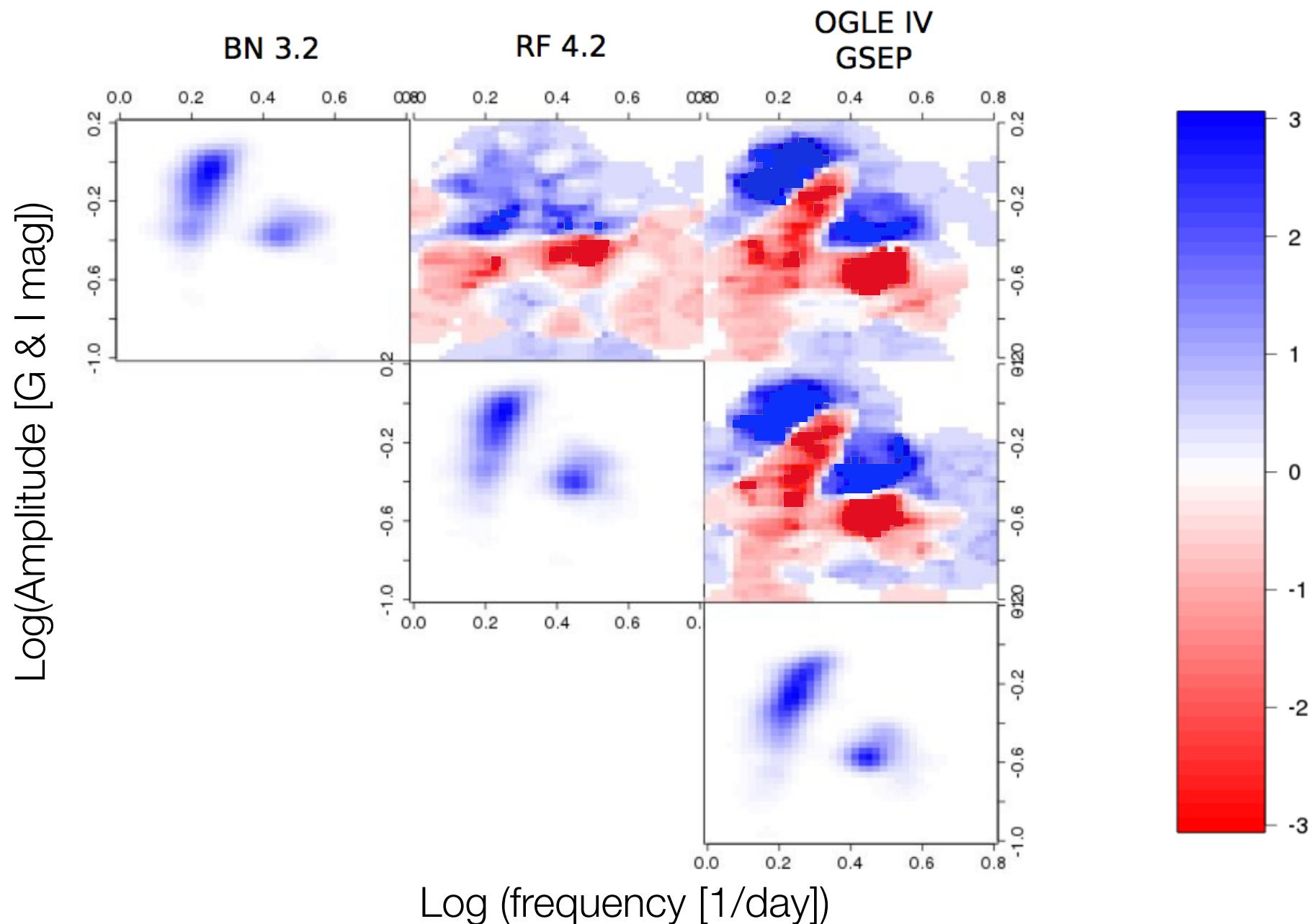
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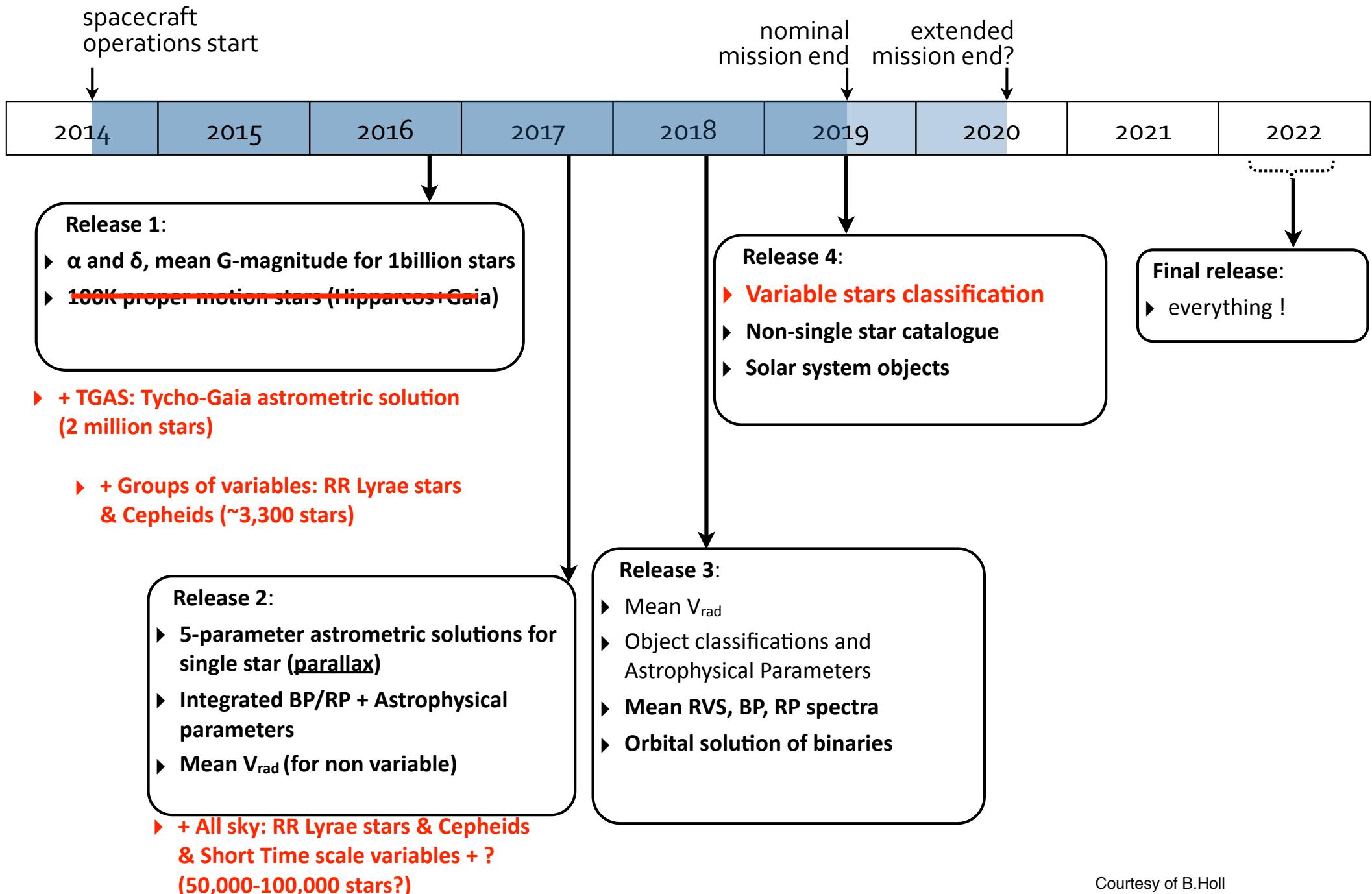


Global Variability studies

Comparison of distribution functions of RR Lyrae stars



The data releases



Data release caveats/limitations

- All sources have been treated as single stars
- Some filtering has been applied, e.g.;
 - Omit sources with too few observations, without astrometry and/or photometry, with very elongated positional error ellipses, ...
 - Upper limit on errors in parallax, position, and photometry
- No high-proper-motion stars ($\mu > 3.5 \text{ arcsec/year}$)
- Various unmodeled effects left in the data (chromaticity, CTI, micro-meteoroid hits, micro-clanks, ...)
- Basic-angle-variation correction derived from on-board metrology
- Cross-matching limited by crude attitude, IGSL, and spurious sources
- Cyclic processing loops not yet closed
- All of these issues will be addressed in upcoming releases!

Warning: the above weaknesses will lead to spatially-correlated systematics in DR1: do not blindly average astrometric quantities