

Variable Star Classification

Pierre Dubath

Observatory of the University of Geneva

Gaia CU7 team

Pierre.Dubath@unige.ch

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Random forest automated supervised classification of *Hipparcos* periodic variable stars

P. Dubath,^{1,2*} L. Rimoldini,^{1,2} M. Süveges,^{1,2} J. Blomme,³ M. López,⁴ L. M. Sarro,⁵
J. De Ridder,³ J. Cuypers,⁶ L. Guy,^{1,2} I. Lecoœur,^{1,2} K. Nienartowicz,^{1,2} A. Jan,^{1,2}
M. Beck,^{1,2} N. Mowlavi,^{1,2} P. De Cat,⁶ T. Lebzelter⁷ and L. Eyer^{1,2}

Context

- Work in progress of the CU7 team to get ready for the Gaia data analysis
- Using Hipparcos data as a control sample
 - Establish the best classification strategy
- Current choice: classification in three steps

Variable Star Classification

- Work in progress of the CU7 team to get ready for the Gaia data analysis (major contributions from Isabelle Lecoeur et Lorenzo Rimoldini)
- Random forest classification of Hipparcos periodic variables, Dubath et al. 2011, MNRAS
- Hipparcos unsolved variable classification, Rimoldini et al. in preparation
- Overall performance of a complete Hipparcos variable star classification

Sample of Surveyed Stars

Variable Stars (10% to 30%)

Non-Periodic Stars

Periodic Stars

Constant Stars

L

SR

I

Be

GCAS

SPB

DSCT

ECL

CEP

ACV

RS

BY

LPV

RR

Sample of Surveyed Stars

Variable Stars (10% to 30%)

RS

Periodic

Non-Per

L

Periodic

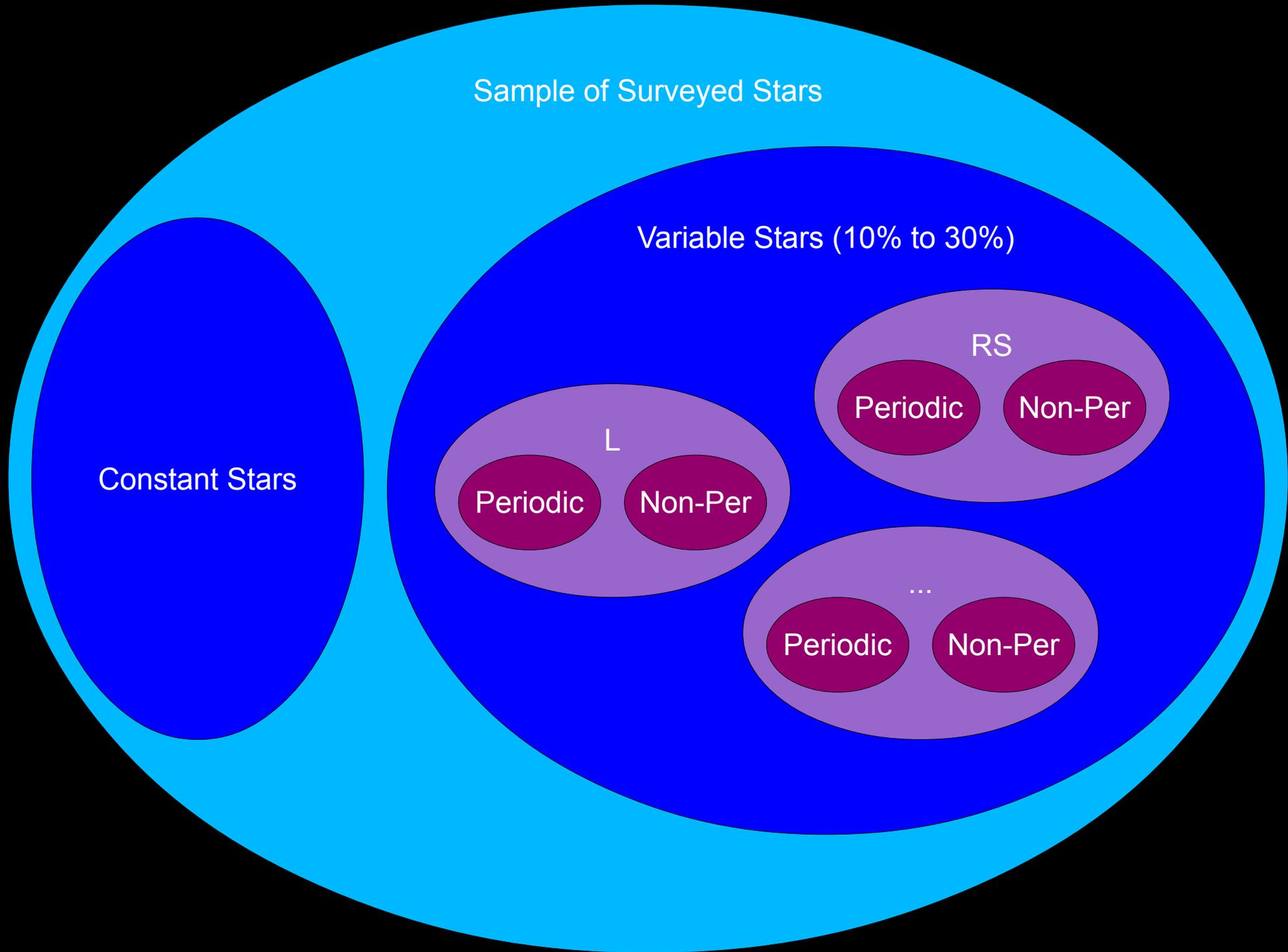
Non-Per

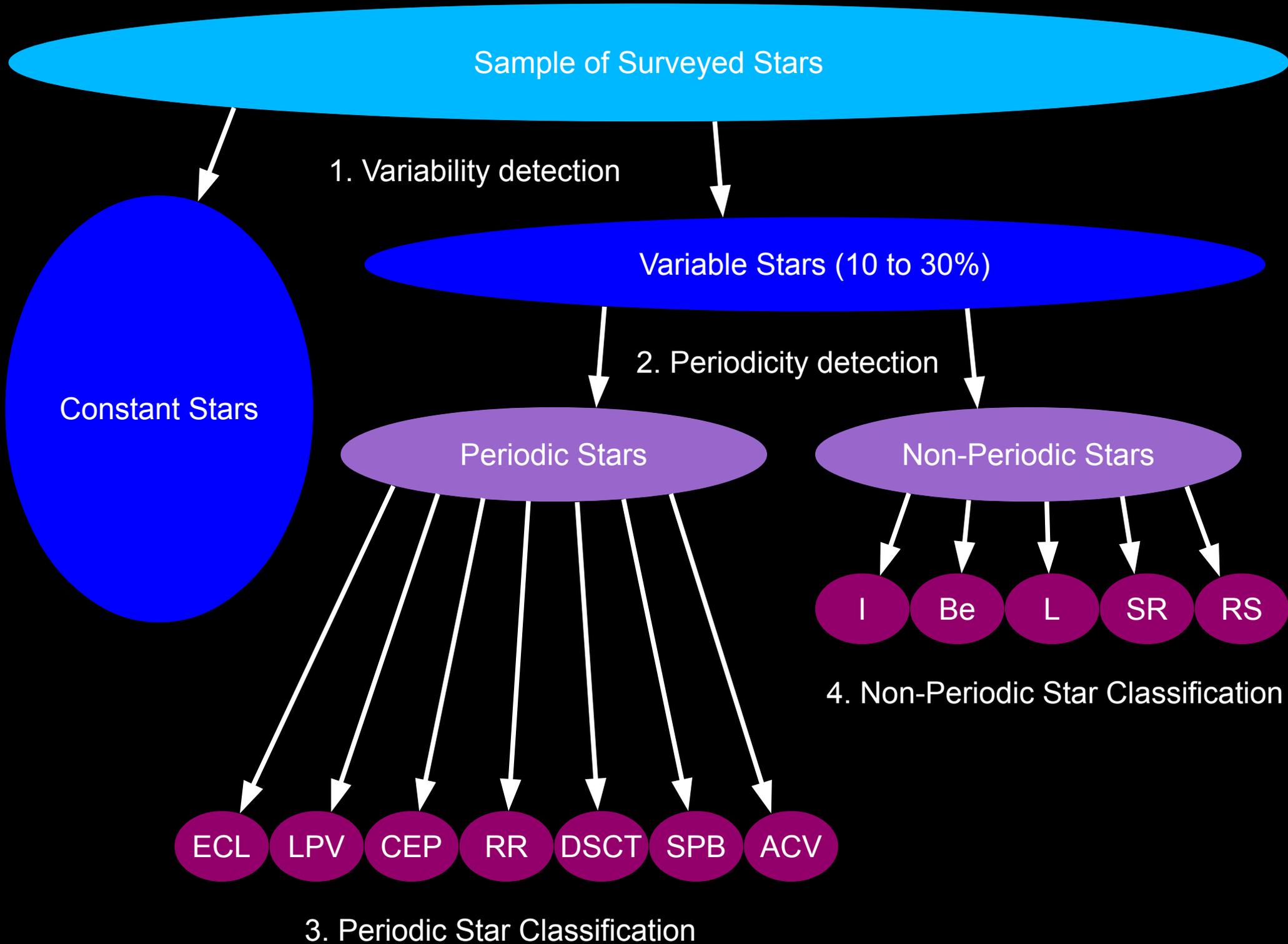
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Periodic

Non-Per

Constant Stars





Sample of Surveyed Stars

1. Variability detection

Variable Stars (10 to 30%)

2. Periodicity detection

Constant Stars

Periodic Stars

Non-Periodic Stars

ECL

LPV

CEP

RR

DSCT

SPB

ACV

3. Periodic Star Classification

I

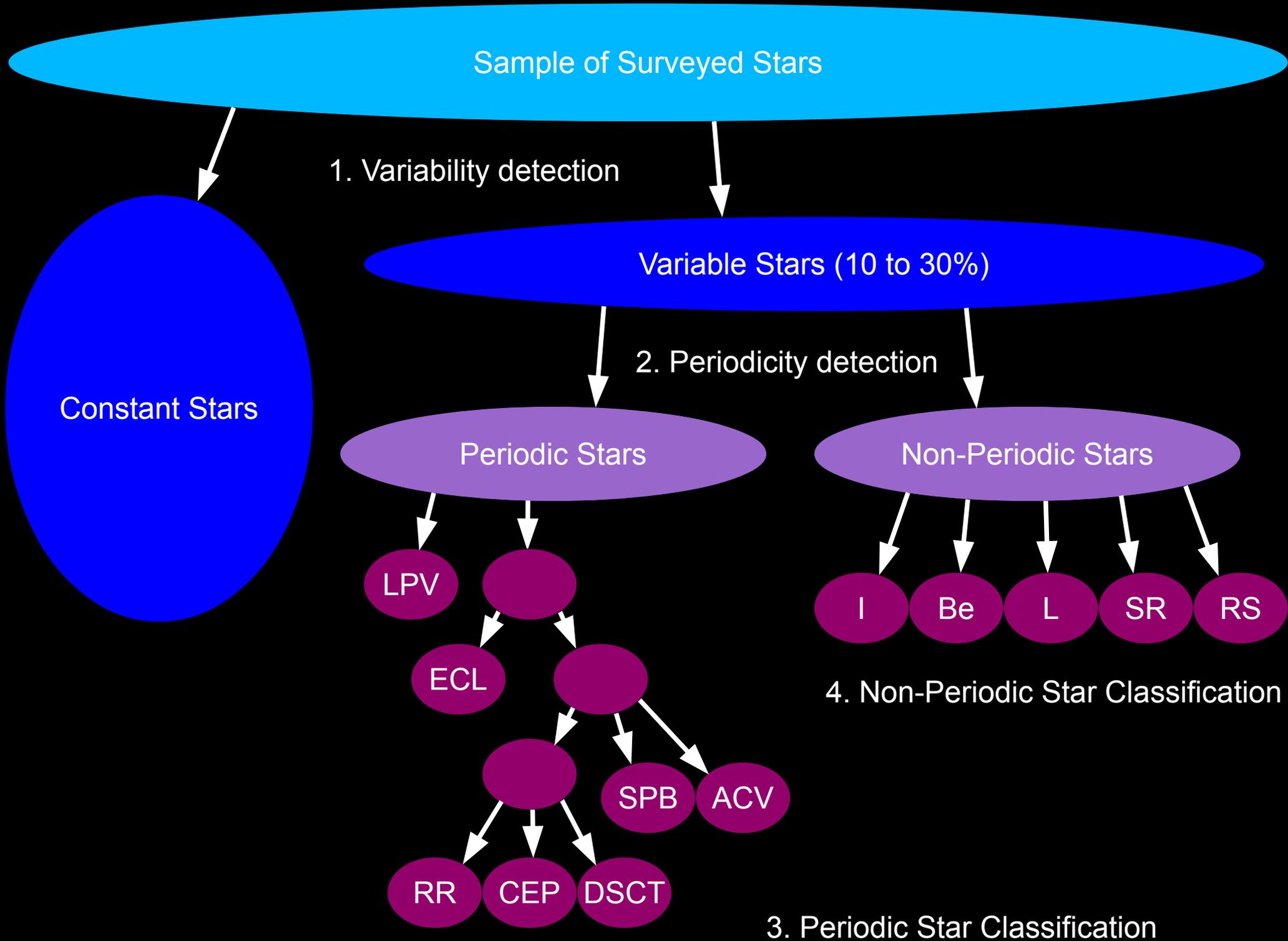
Be

L

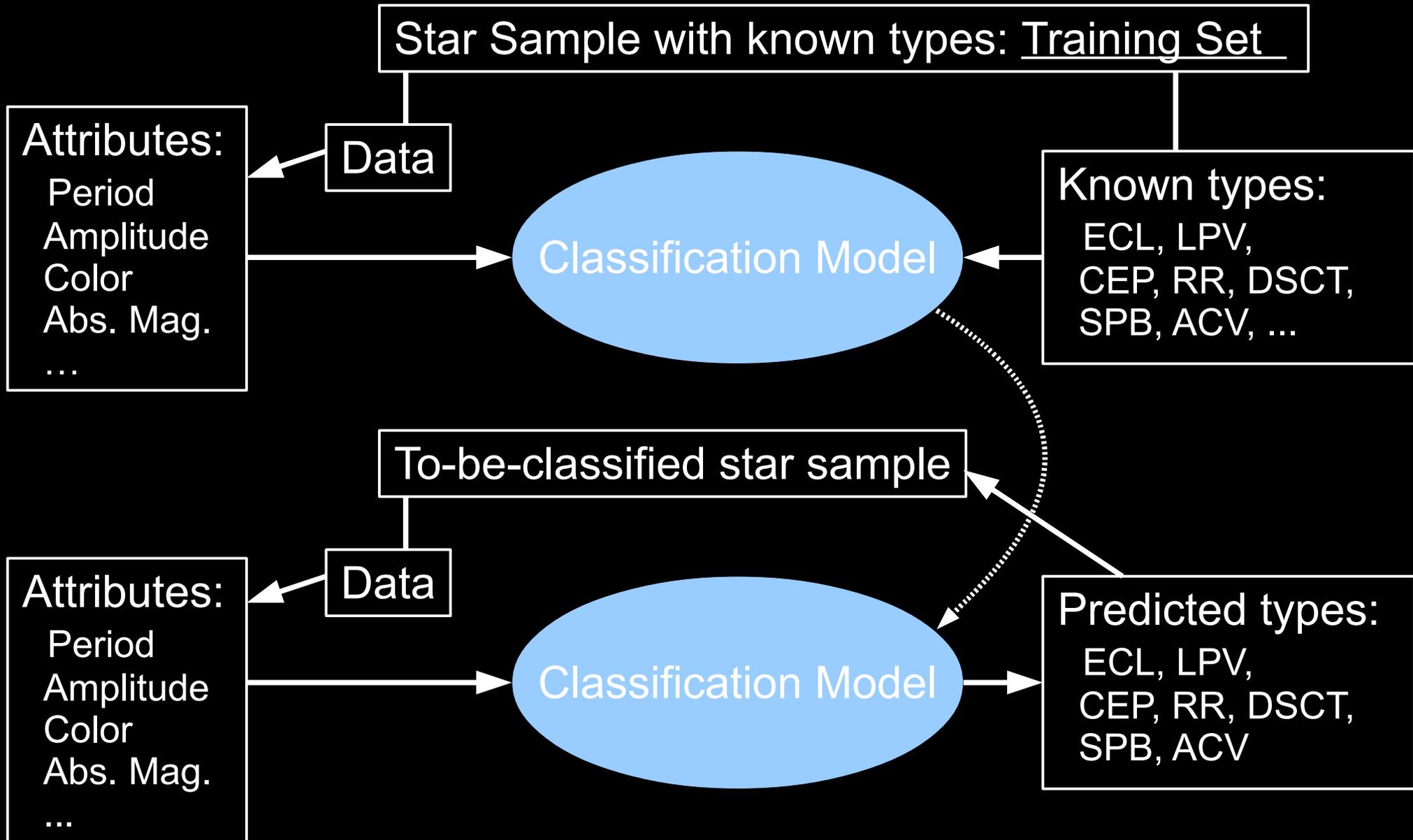
SR

RS

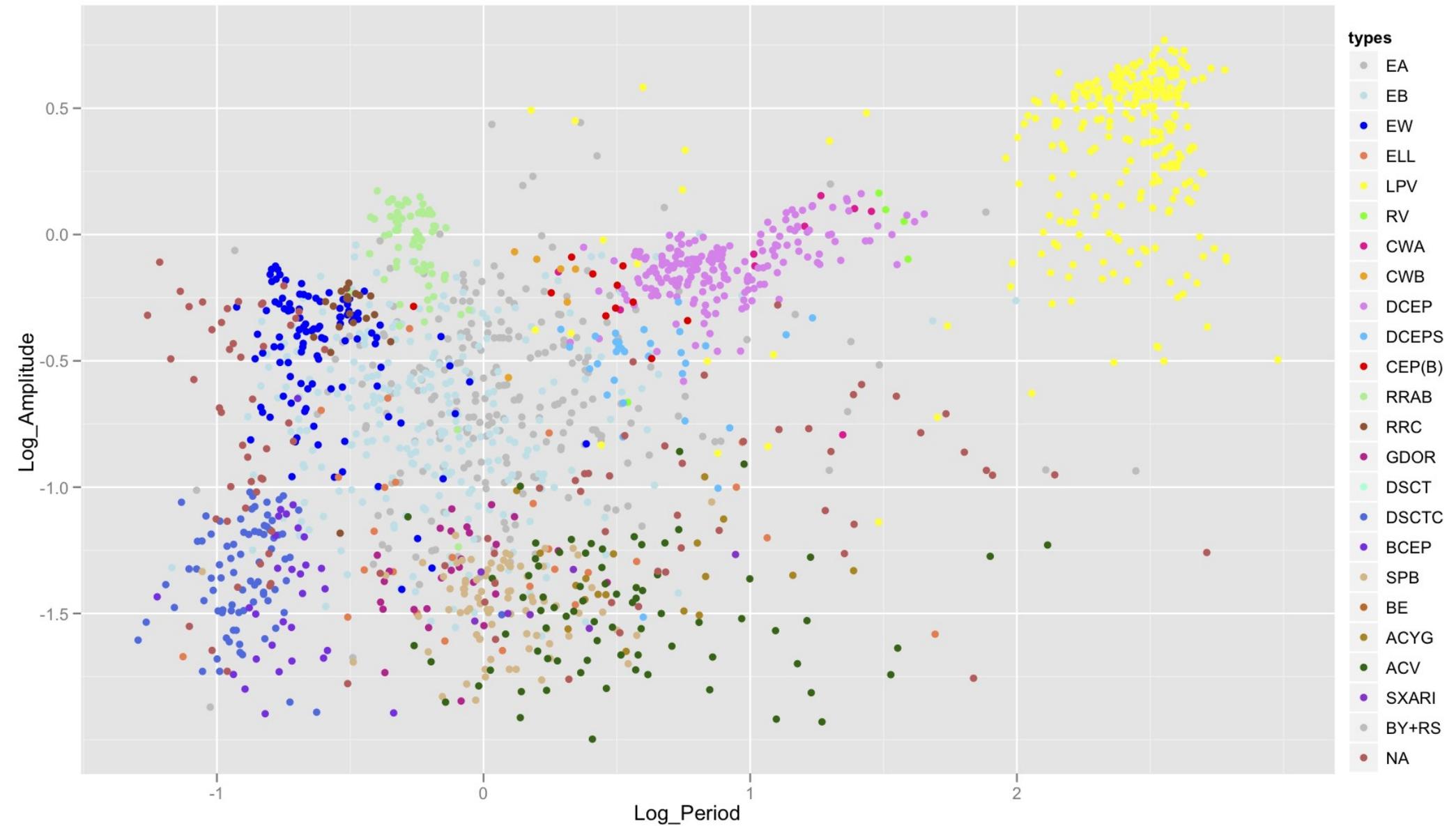
4. Non-Periodic Star Classification



Supervised Classification



Period-Amplitude diagram



Supervised Classification

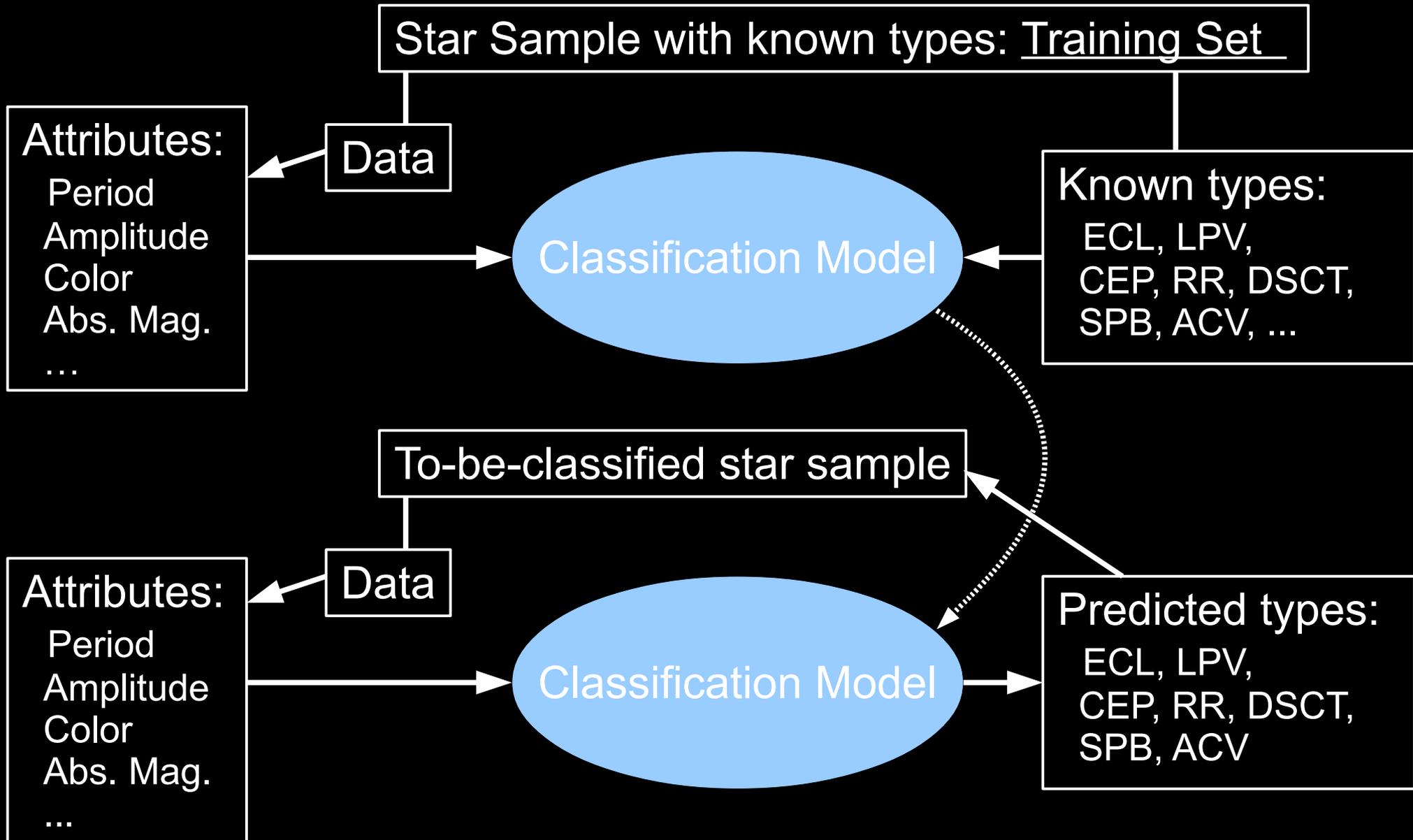
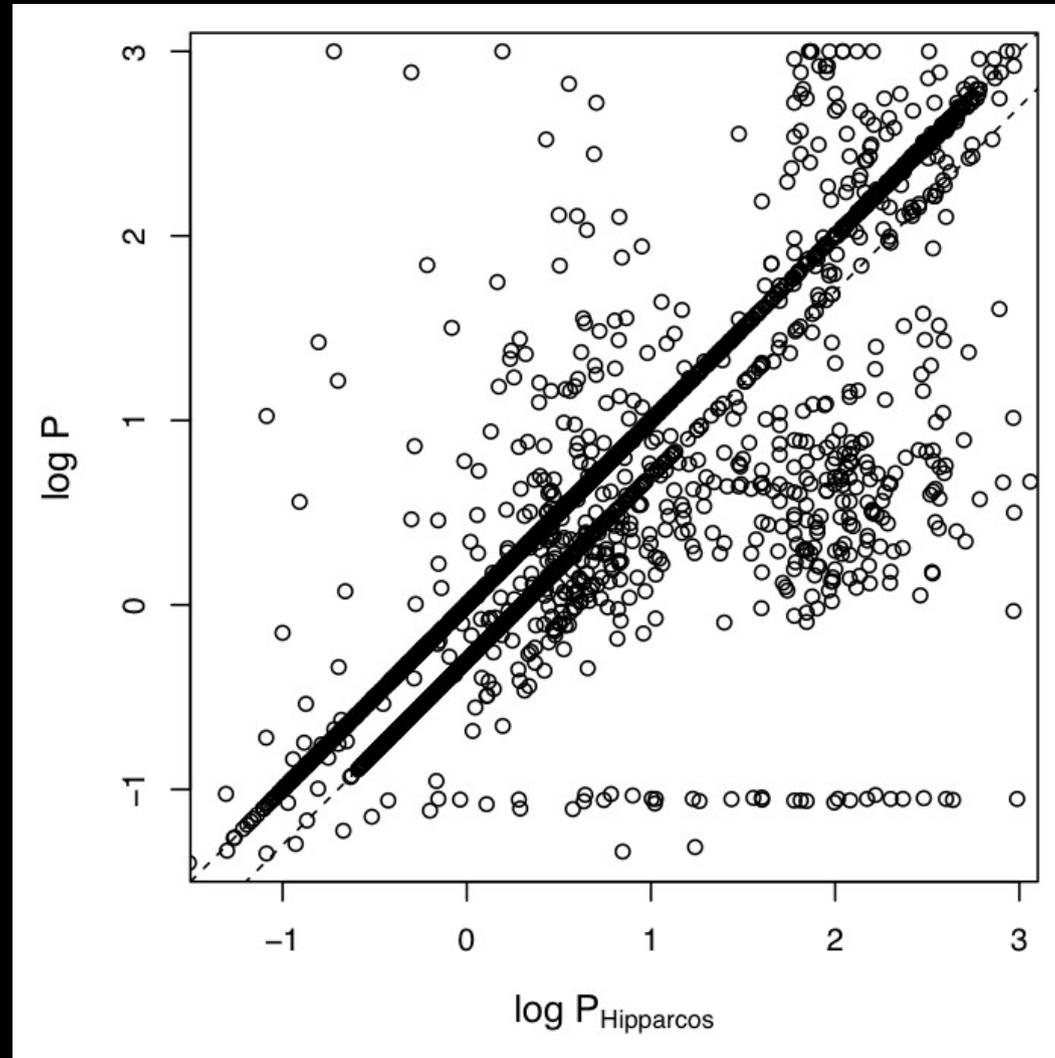


Table 1. Training set composition

Type		Num	Main reference
Eclipsing Binary	<i>EA</i>	228	Hipparcos
	<i>EB</i>	255	Hipparcos
	<i>EW</i>	107	Hipparcos
Ellipsoidal	<i>ELL</i>	27	Hipparcos
Long Period Variable	<i>LPV</i>	285	Lebzelter (p. c.)
RV Tauri	<i>RV</i>	5	AAVSO
W Virginis	<i>CWA</i>	9	AAVSO
	<i>CWB</i>	6	AAVSO
Delta Cepheid (first overtone) (multi mode)	<i>DCEP</i>	189	AAVSO
	<i>DCEPS</i>	31	AAVSO
	<i>CEP(B)</i>	11	AAVSO
RR Lyrae	<i>RRAB</i>	72	AAVSO
	<i>RRC</i>	20	AAVSO
Gamma Doradus	<i>GDOR</i>	27	De Cat (p. c.)
Delta Scuti (low amplitude)	<i>DSCT</i>	43	AAVSO
	<i>DSCTC</i>	81	AAVSO
SX Phoenicis	<i>SXPHE</i>	4	AAVSO
Beta Cephei	<i>BCEP</i>	30	De Cat (p. c.)
Slowly Pulsating B star	<i>SPB</i>	81	De Cat (p. c.)
B emission line star	<i>BE</i>	9	AAVSO
Gamma Cassiopeiae	<i>GCAS</i>	4	AAVSO
Alpha Cygni	<i>ACYG</i>	18	AAVSO
Alpha-2 Canum Venaticorum	<i>ACV</i>	77	Romanyuk (p. c.)
SX Arietis	<i>SXARI</i>	7	Romanyuk (p. c.)
BY Draconis	<i>BY</i>	5	Eker et al. (2008)
RS Canum Venaticorum	<i>RS</i>	30	Eker et al. (2008)
Total:		1661	

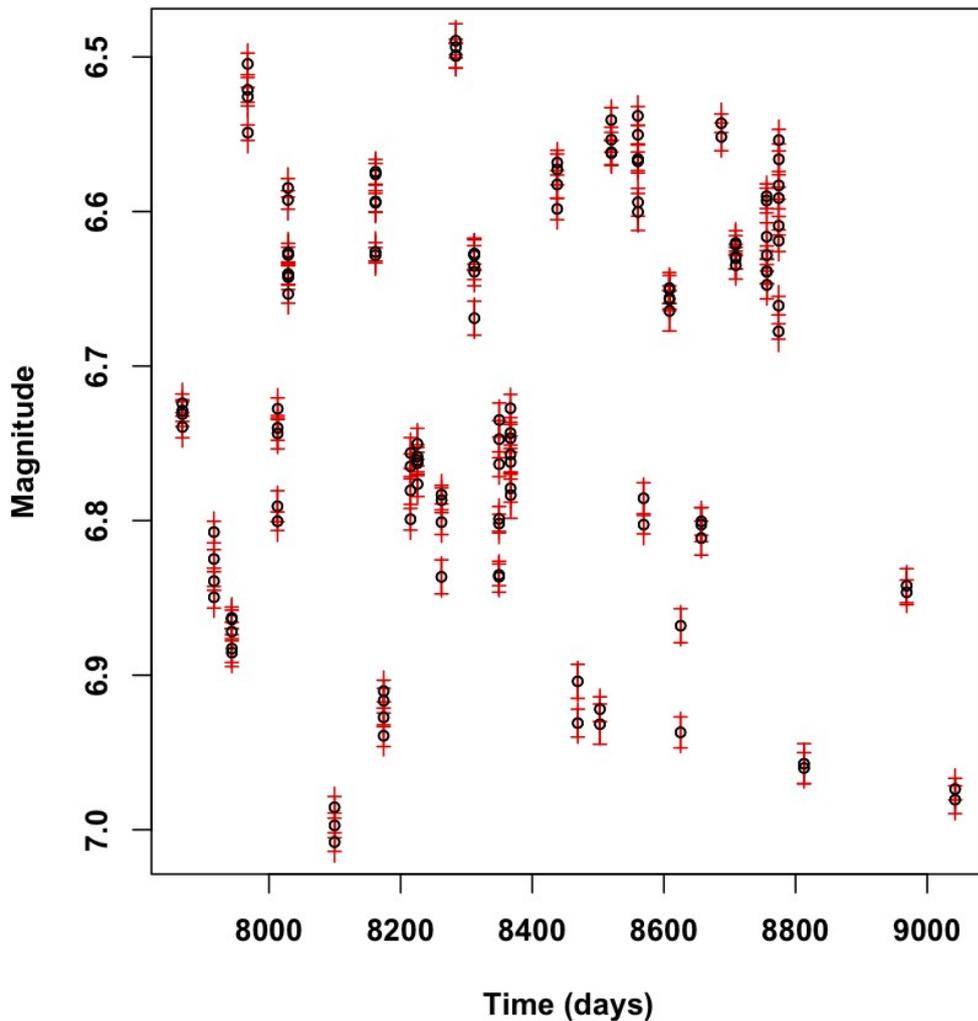
Lomb-Scargle period search



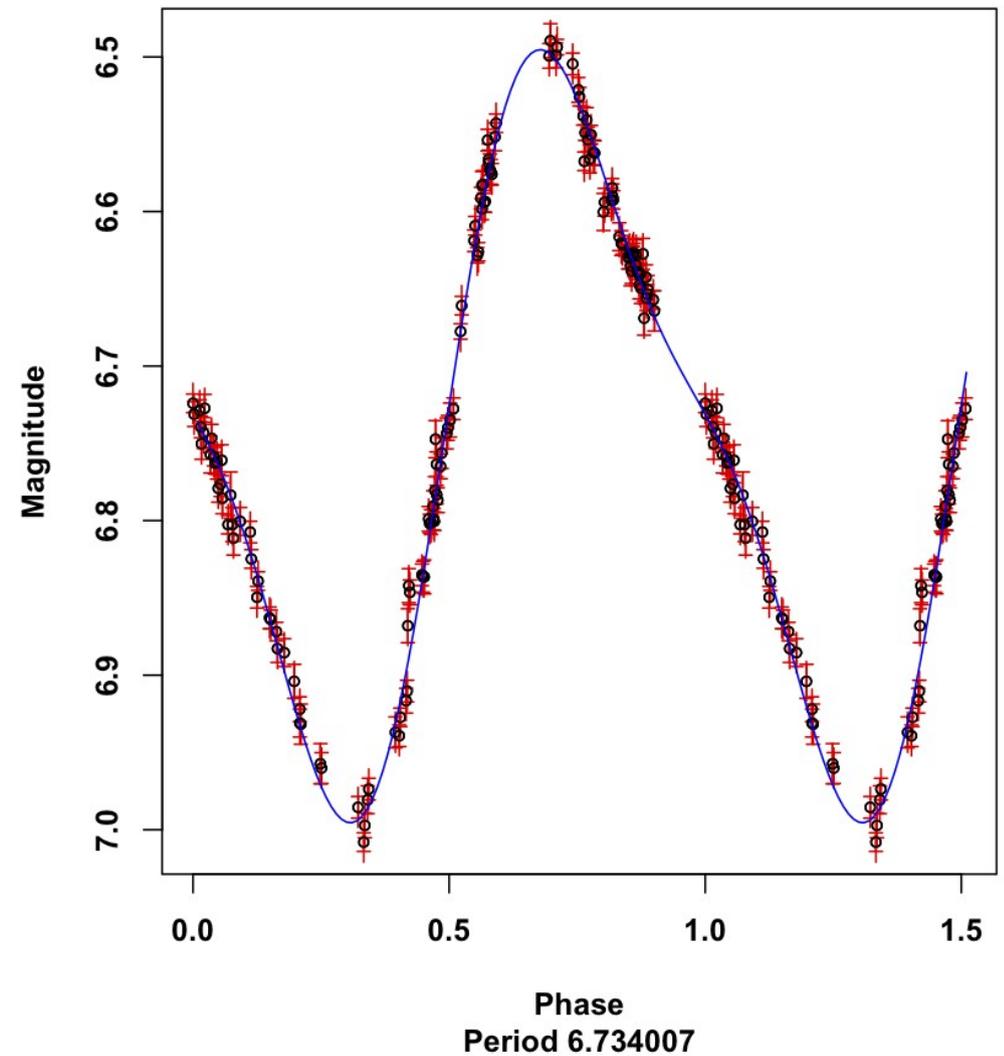
- (Zechmeister & Kürster 2009)

Fourier series modeling

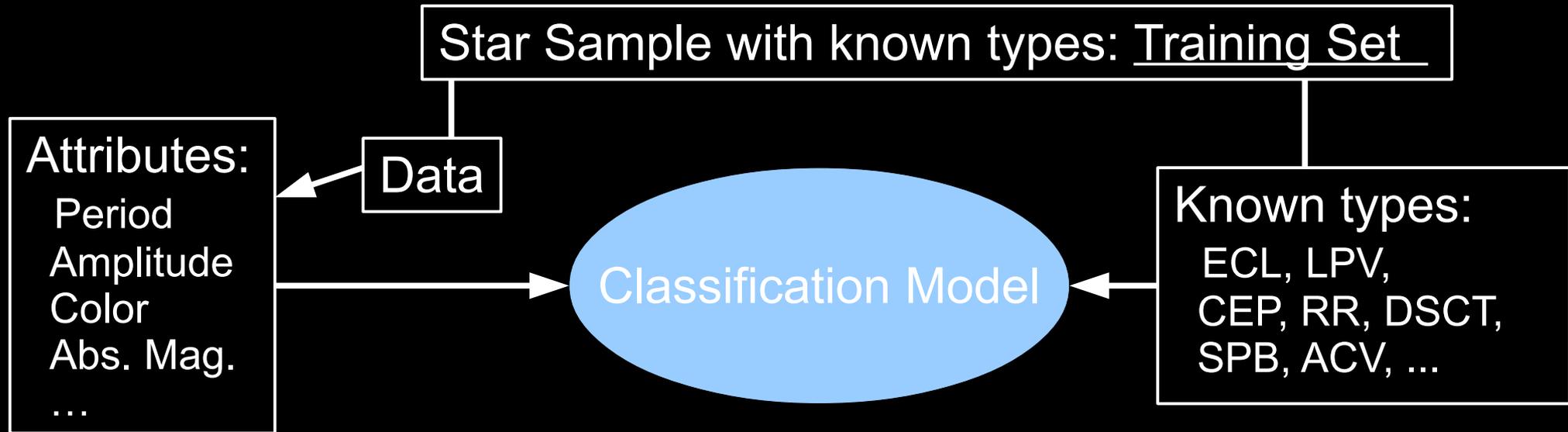
Original Time Series



Srv: hipparcos, Src: 60259

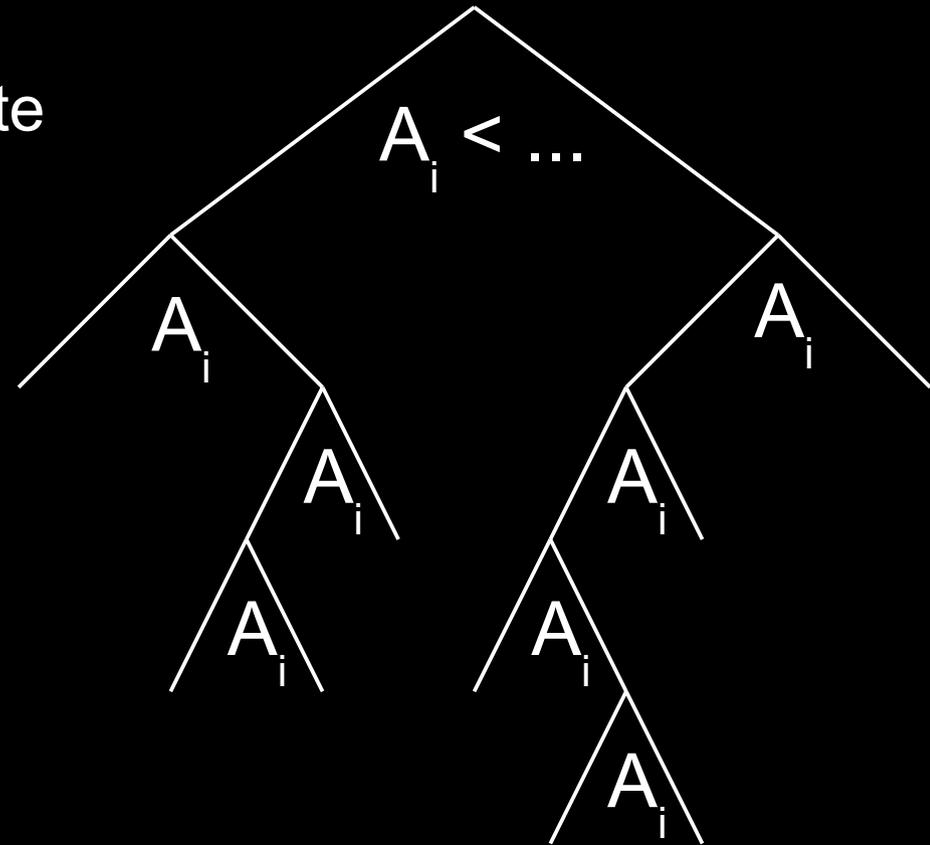


Supervised Classification

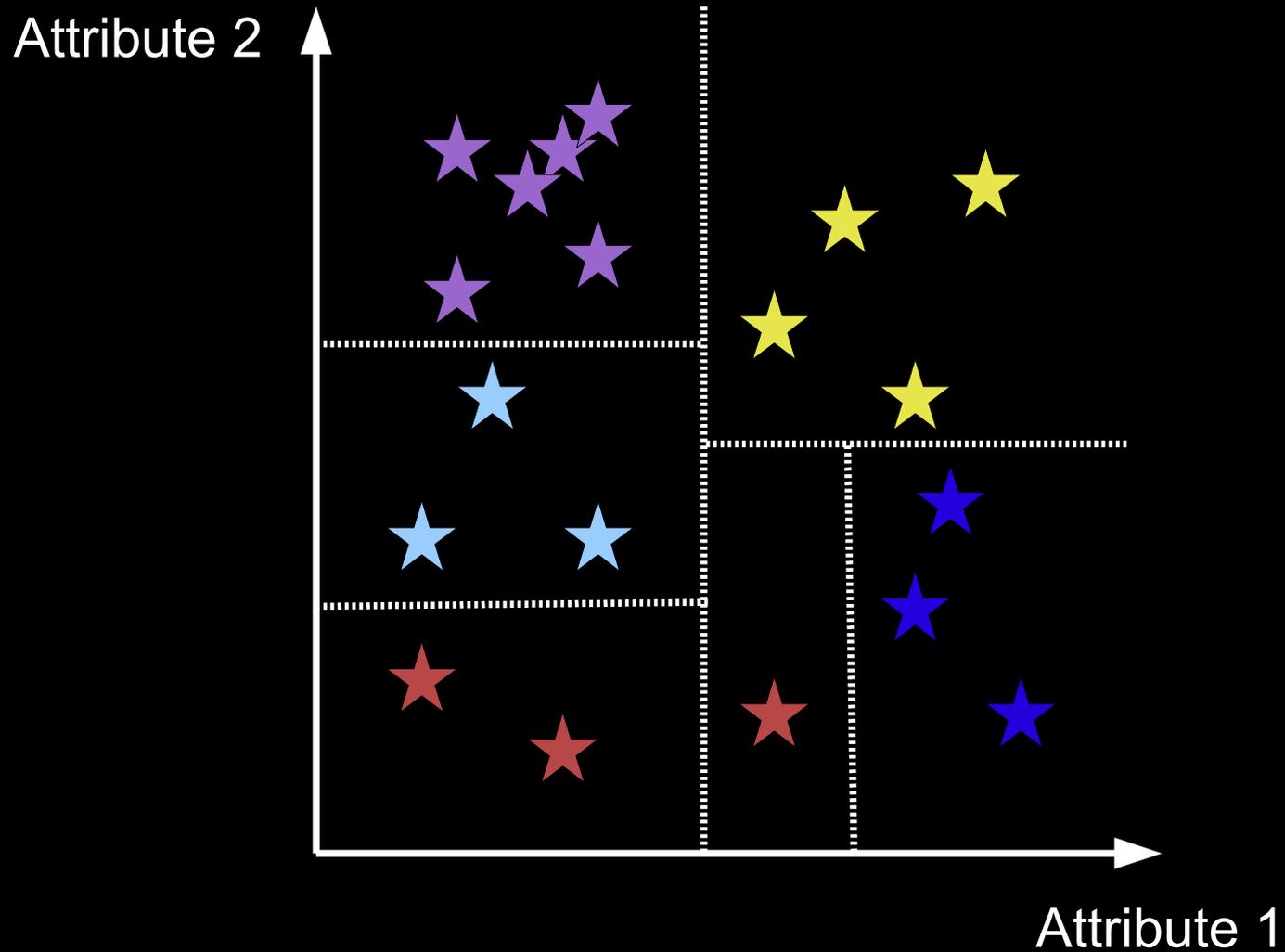


Random Forest (1/3)

- Classification trees
- Binary partitions using one attribute
- Each split minimize impurity

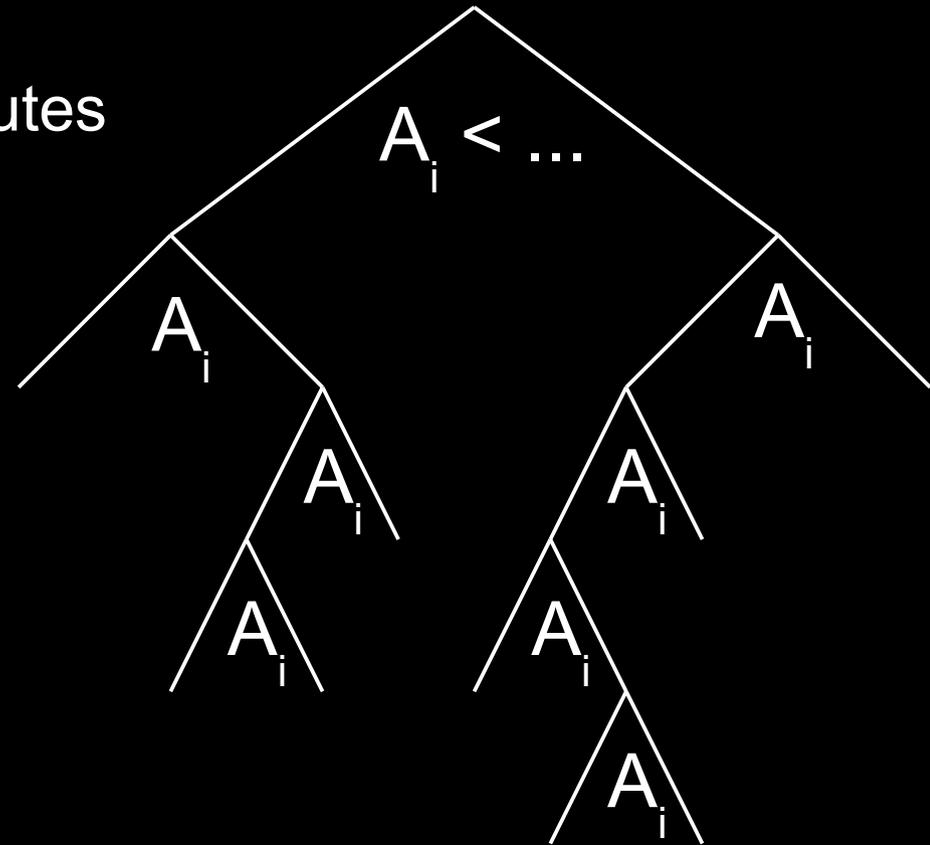


Random Forest (2/3)

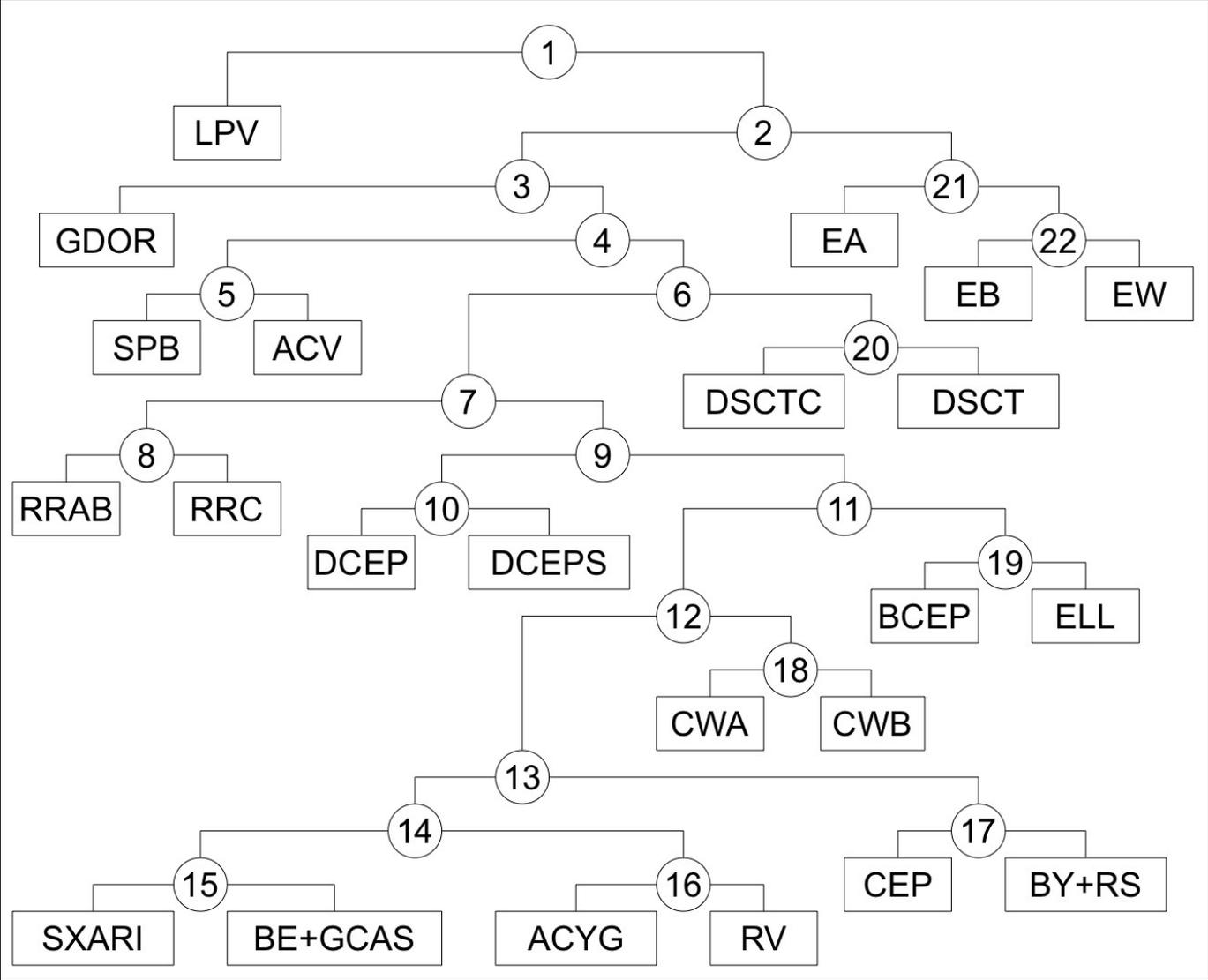


Random Forest (3/3)

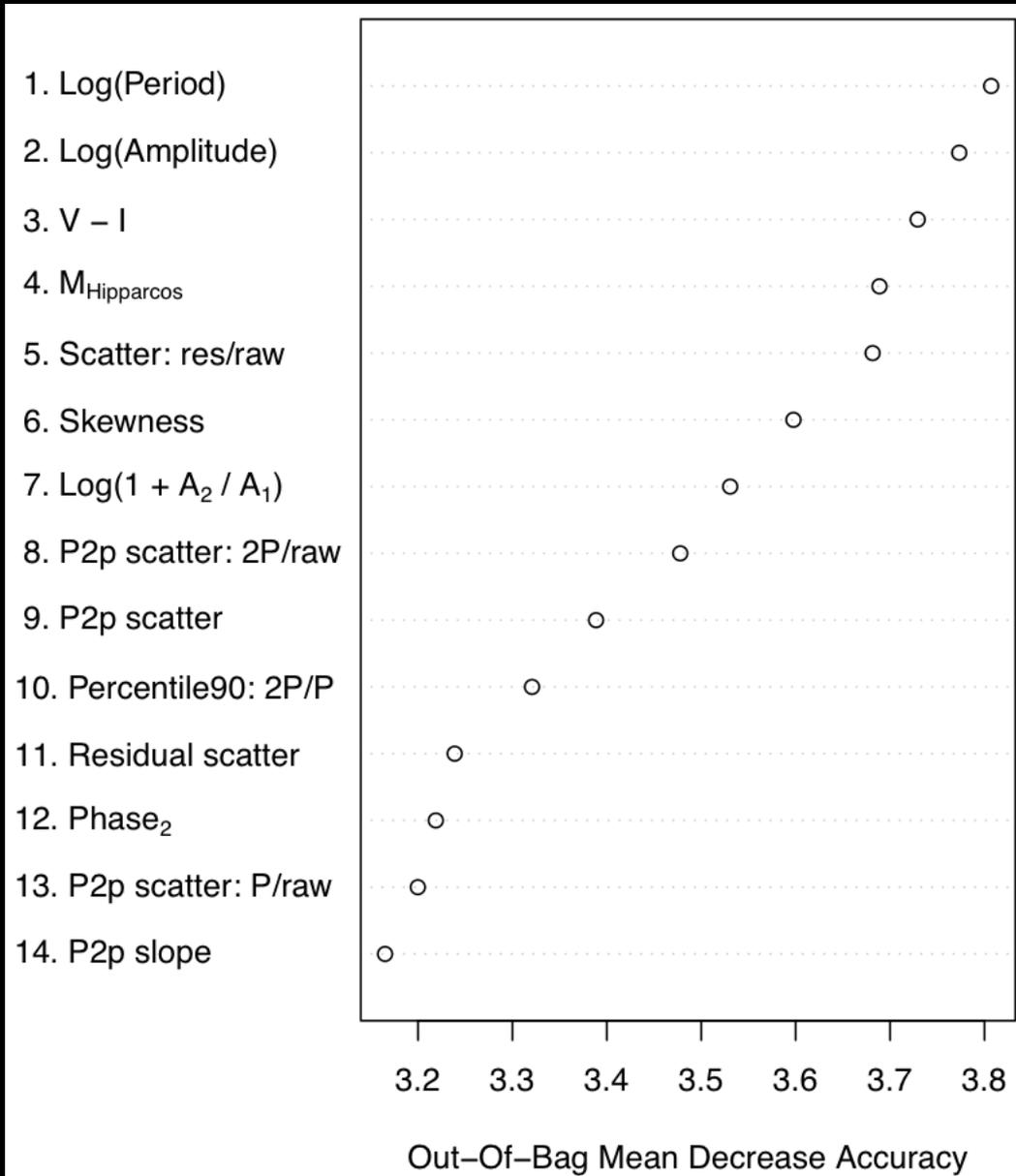
- Bootstrapping
- A_i from a random sub-set of attributes
- Average many trees



Multistage classification

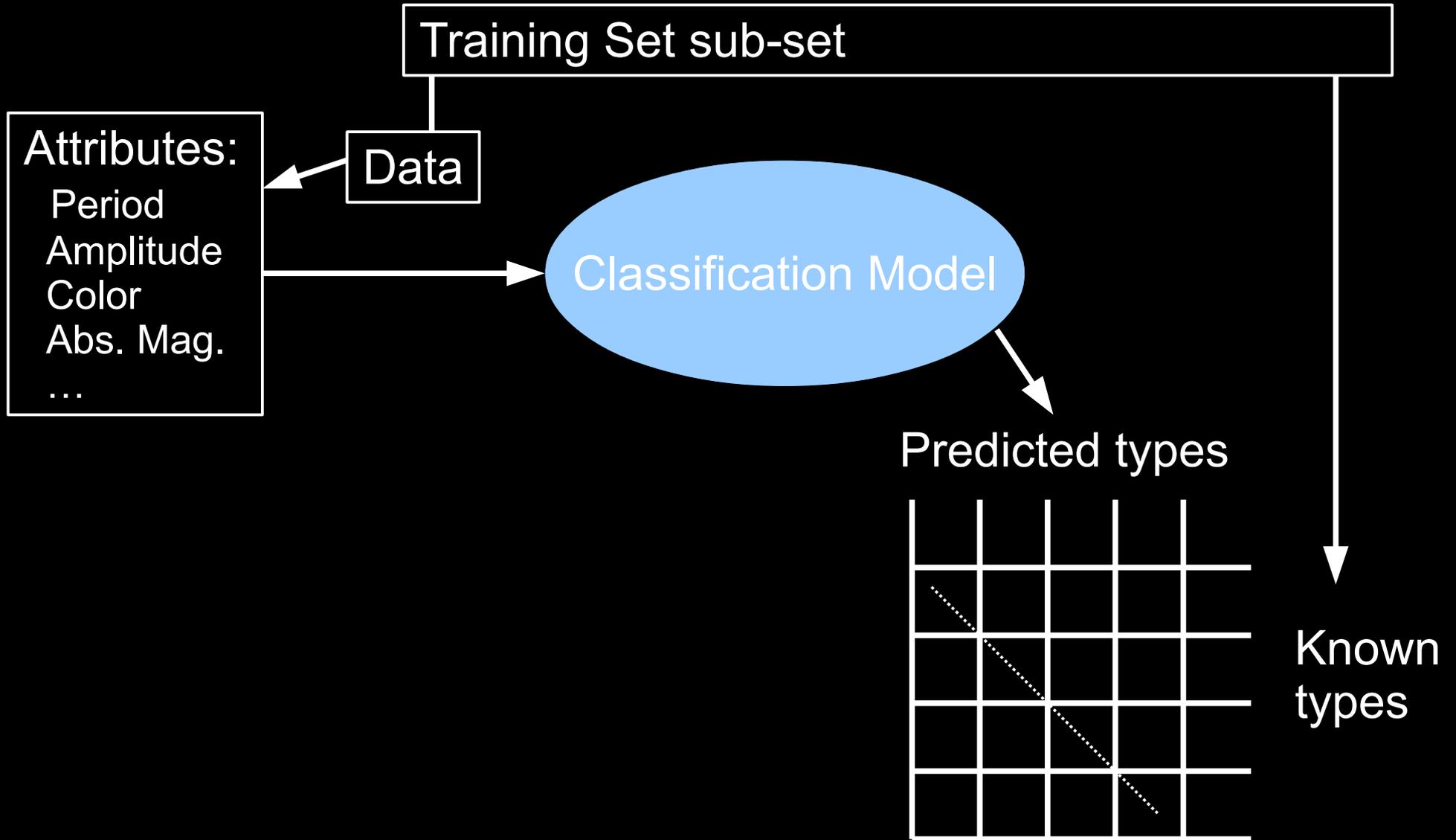


Random Forest Attribute Importance

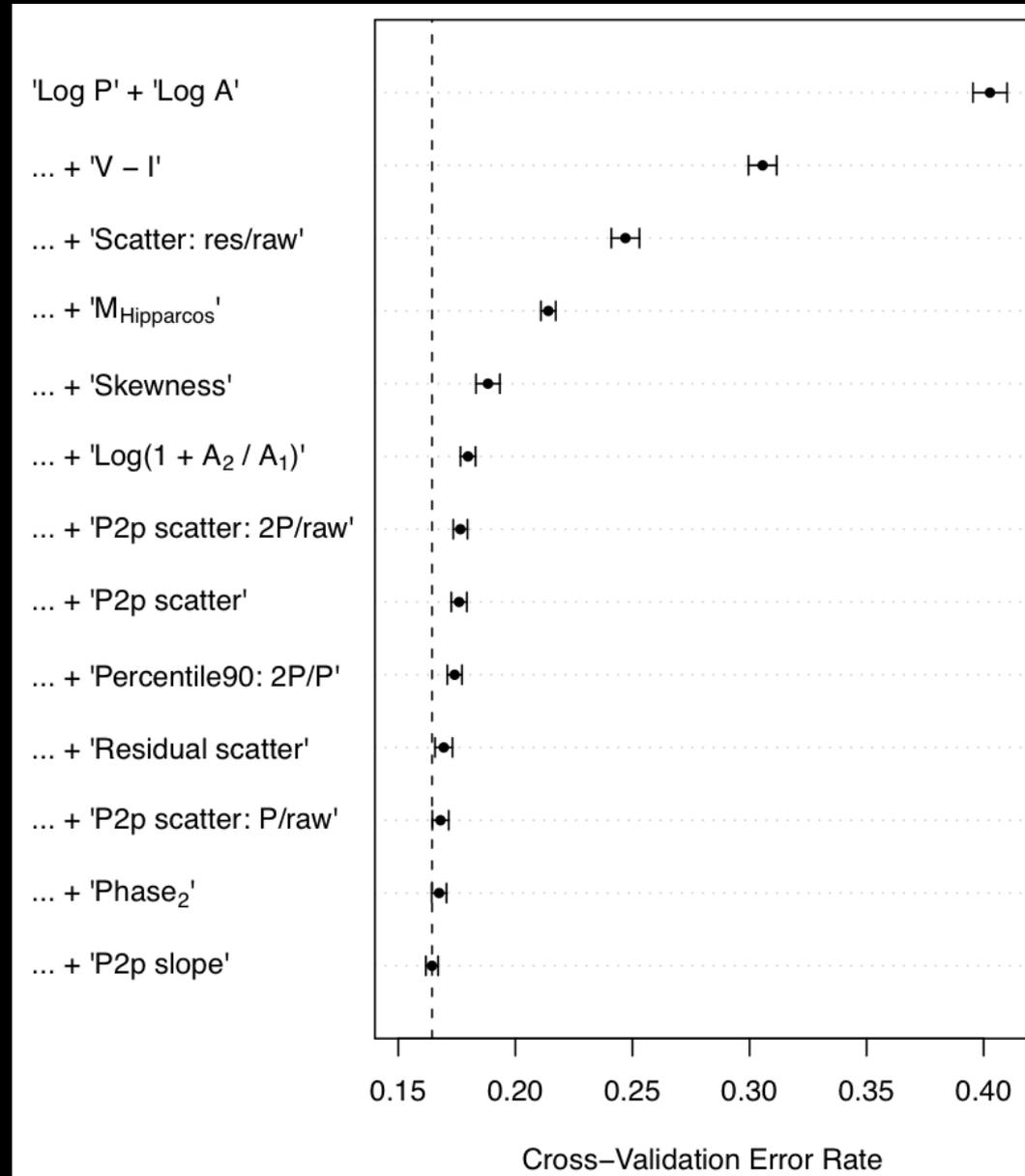


- Attributes with Spearman correlation larger than 80% are trimmed

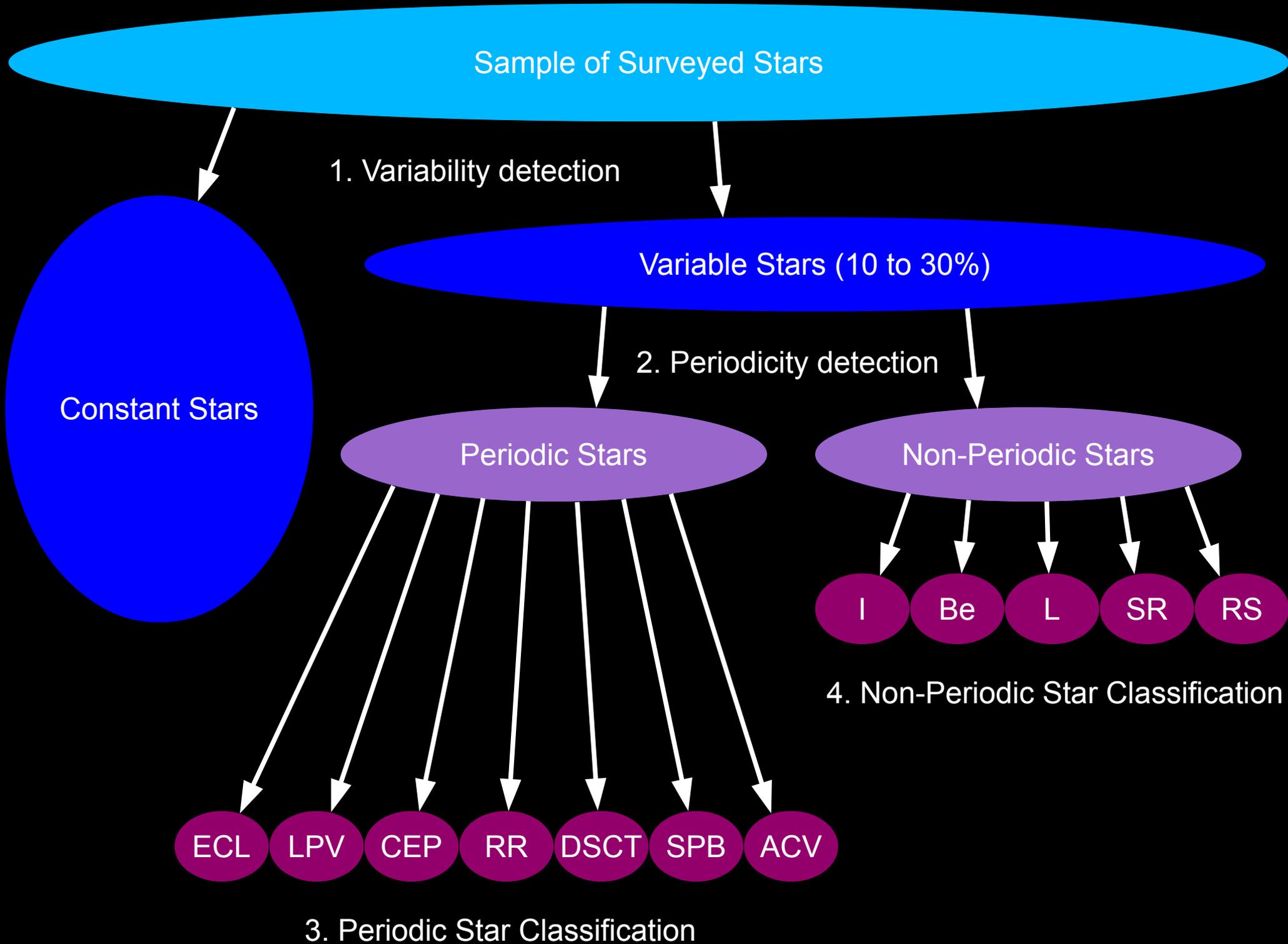
Cross-Validation



Random Forest CV error rates



EA	EB	EW	ELL	LPV	RV	CWA	CWB	DCEP	DCEPS	CEP(B)	RRAB	RRC	GDOR	DSCT	DSCTC	BCEP	SPB	BE+GCAS	ACYG	ACV	SXARI	BY+RS		
214	13									1													EA	
19	191	28	2	1				2					1		4		3		2	2			EB	
	30	76							1														EW	
	14			1									1		1		3				5	2	ELL	
				285																			LPV	
	1			1				2	1														RV	
	2				1			5															1	CWA
	1						2	2	1															CWB
								183	5	1														DCEP
	1							11	17														2	DCEPS
	1							4		6														CEP(B)
	1									69	1						1							RRAB
	2	4								1	12			1										RRC
													27											GDOR
	1	1									1			32	12									DSCT
	1													1	77						2			DSCTC
	1	1													1	26	1							BCEP
			1													1	74		1	4				SPB
1									1								5		2	4				BE+GCAS
	1																	1	13	2		1		ACYG
	3								1				1				6				66			ACV
	2																2				3			SXARI
	1							1															33	BY+RS



Variability detection

- Variability criteria

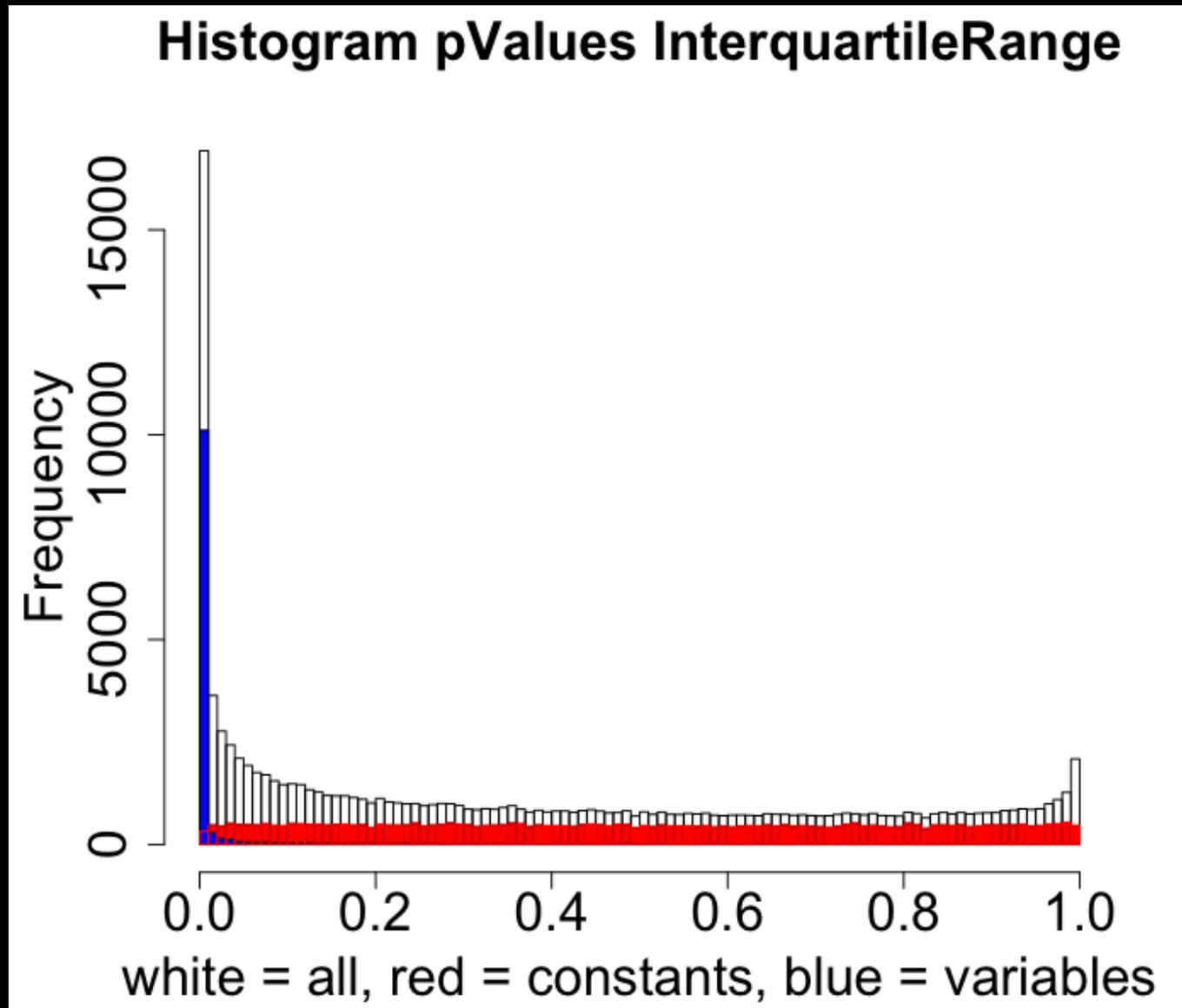
<u>Using errors</u>	<u>Not using errors</u>
Chi square	Abbe
Skewness	Skewness
Kurtosis	Kurtosis
Stetson	Inter-quartile range
Outlier median	B/R Correlation

$$\chi^2 = \sum_{i=1}^N \left(\frac{x_i - \bar{x}}{\sigma} \right)^2$$

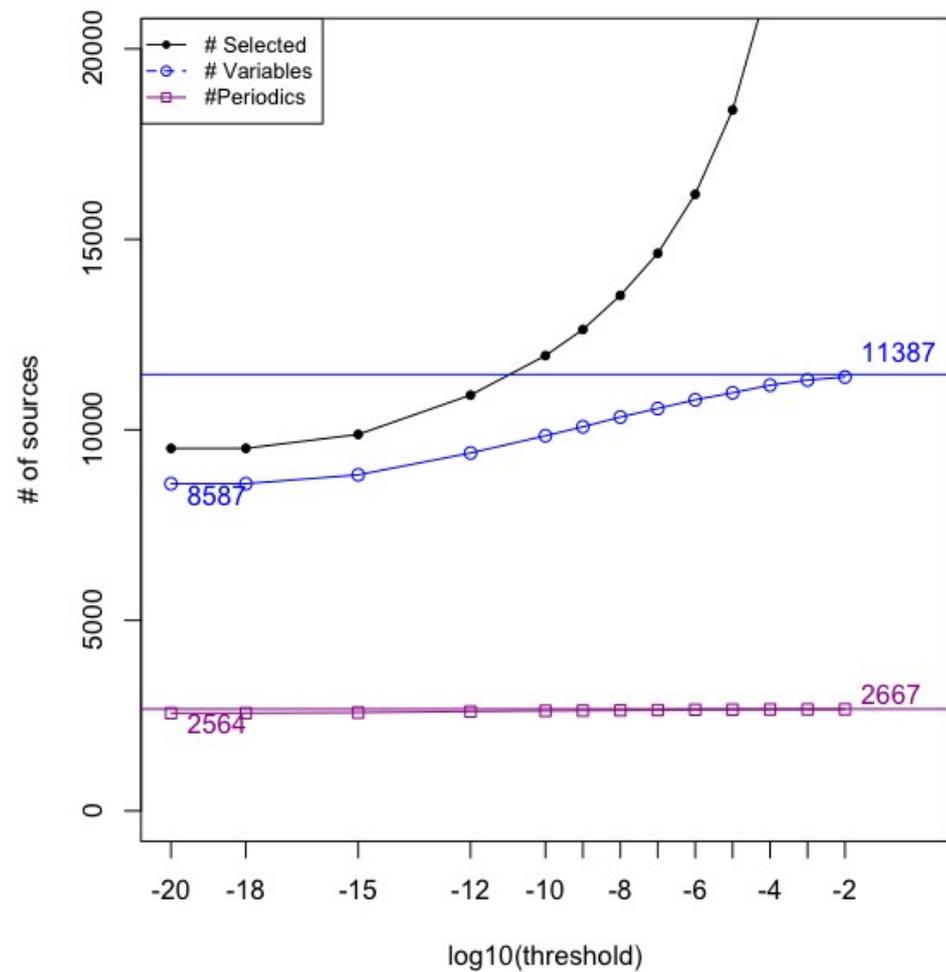
$$Abbe = \frac{1}{2} \frac{\sum_{i=1}^{n-1} (x_{i+1} - x_i)^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

- Compute pValues: probability of the null hypothesis H_0
= constant star
→ If pValue < 1e-4 → variable star

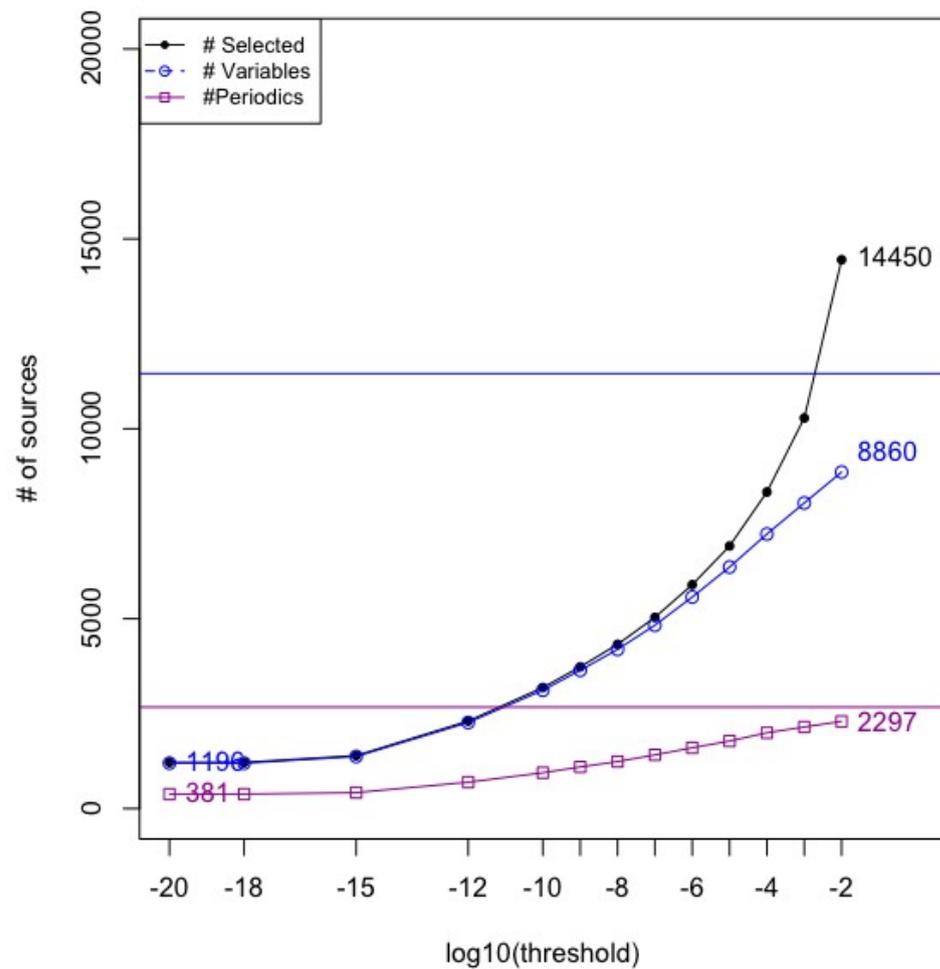
Inter-quartile range Hypparcos pValues



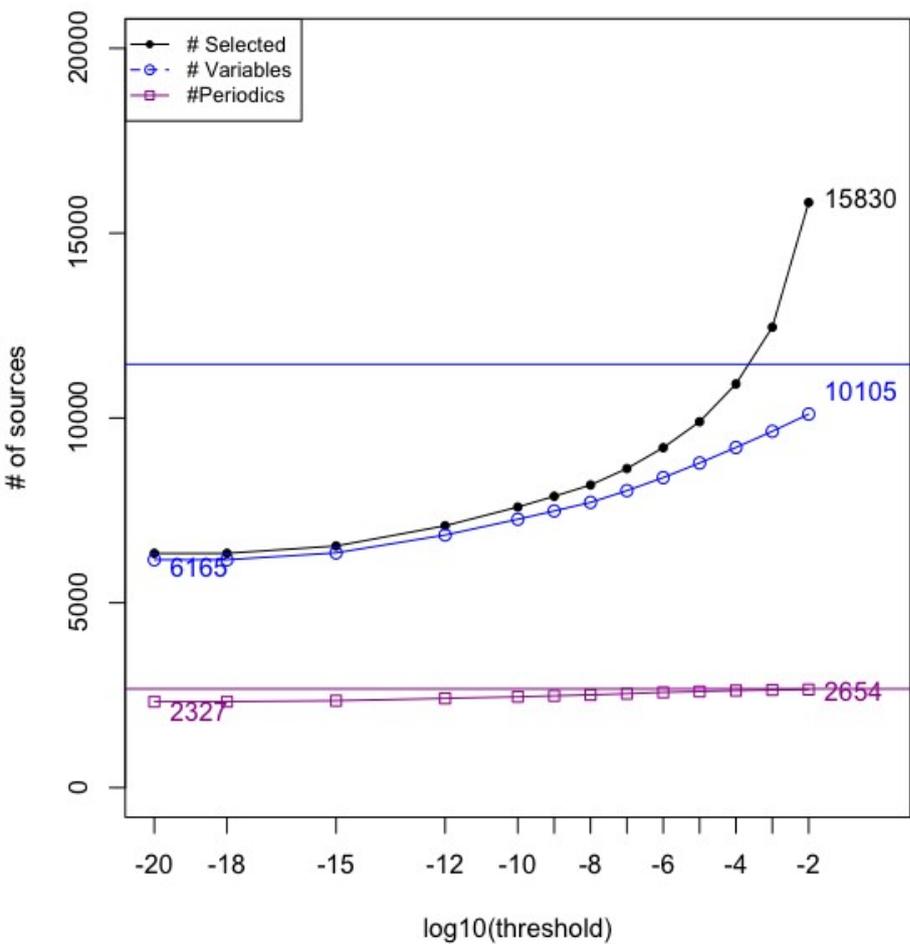
Chi2 Test



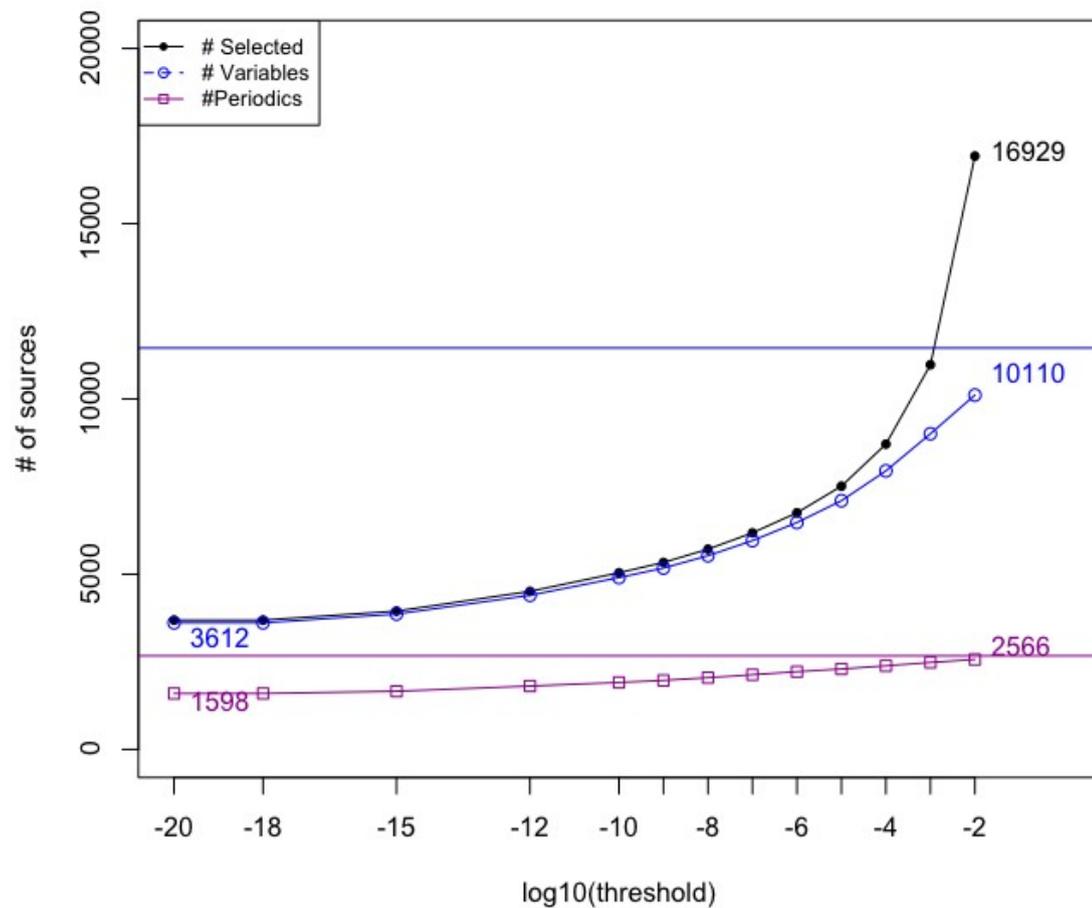
Abbe Test



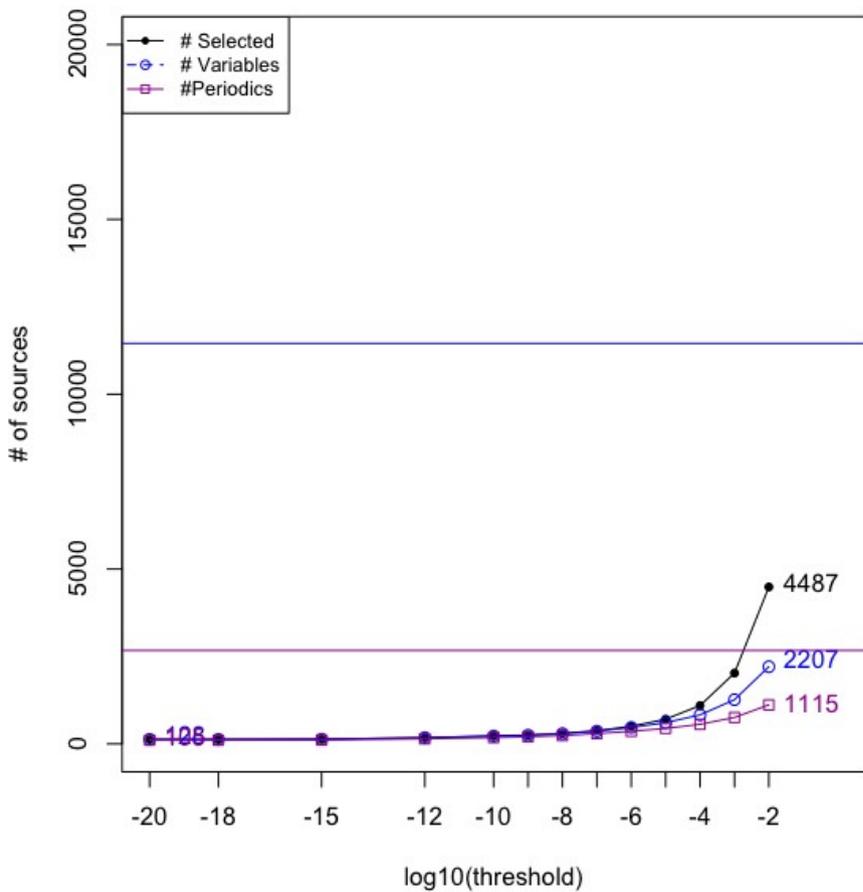
Stetson Test



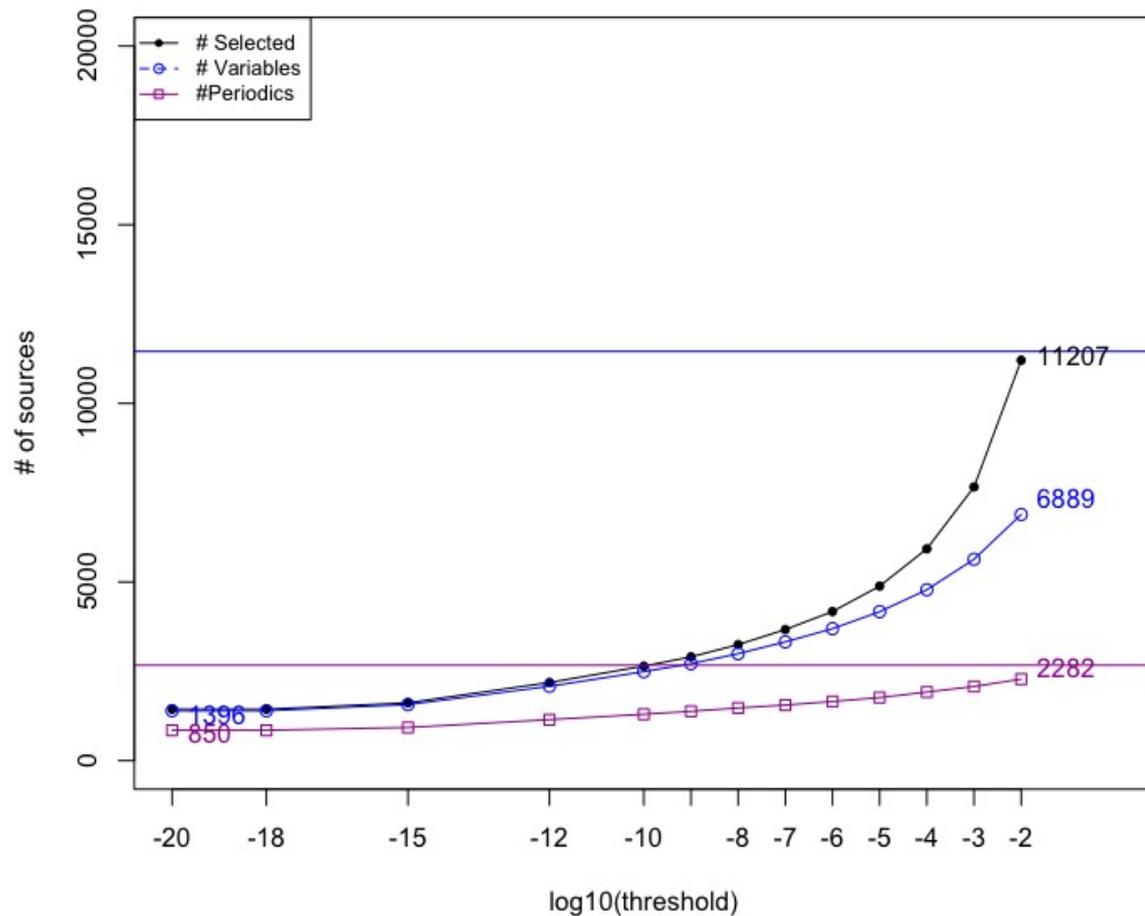
Interquartile Range Test



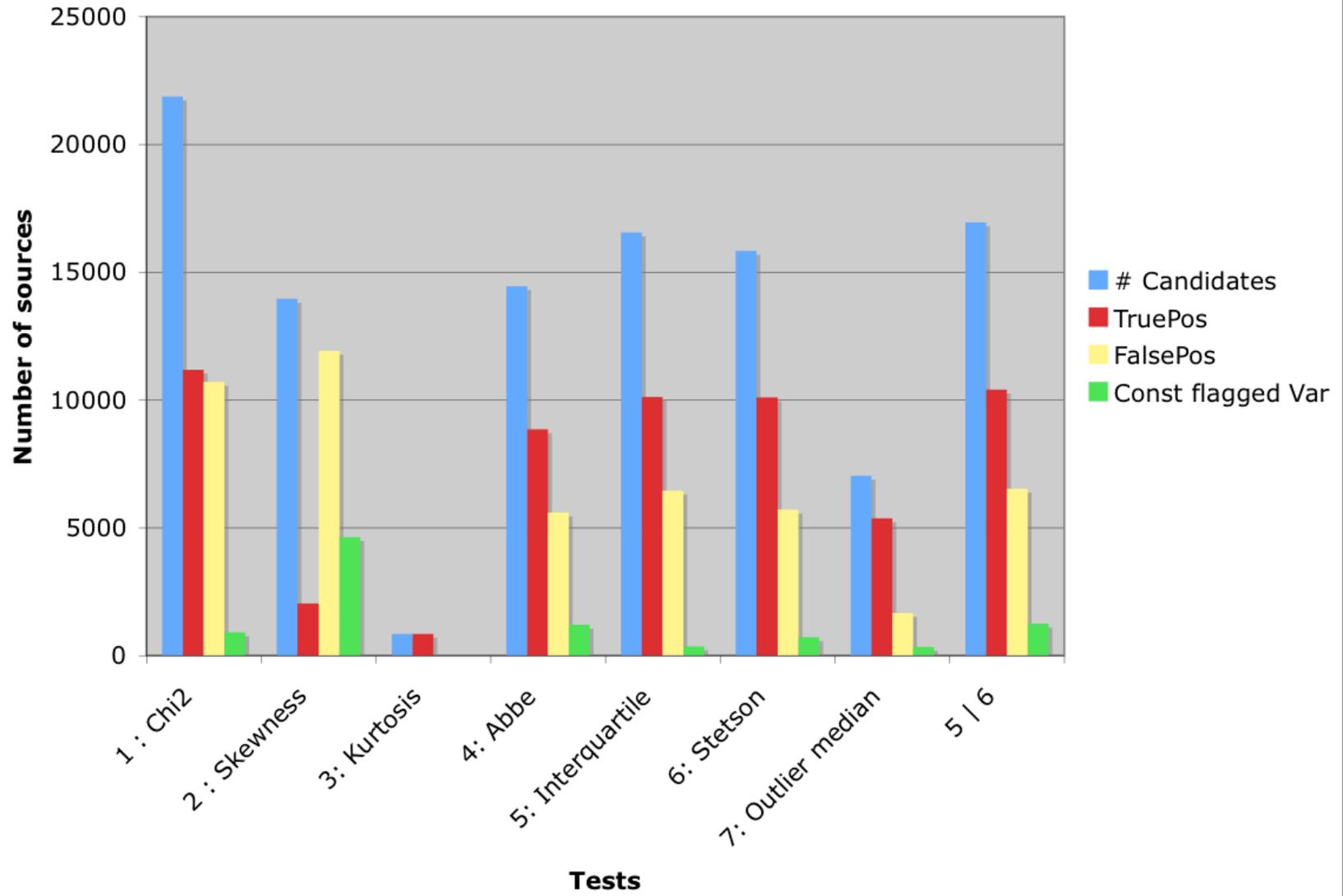
Kurtosis Test



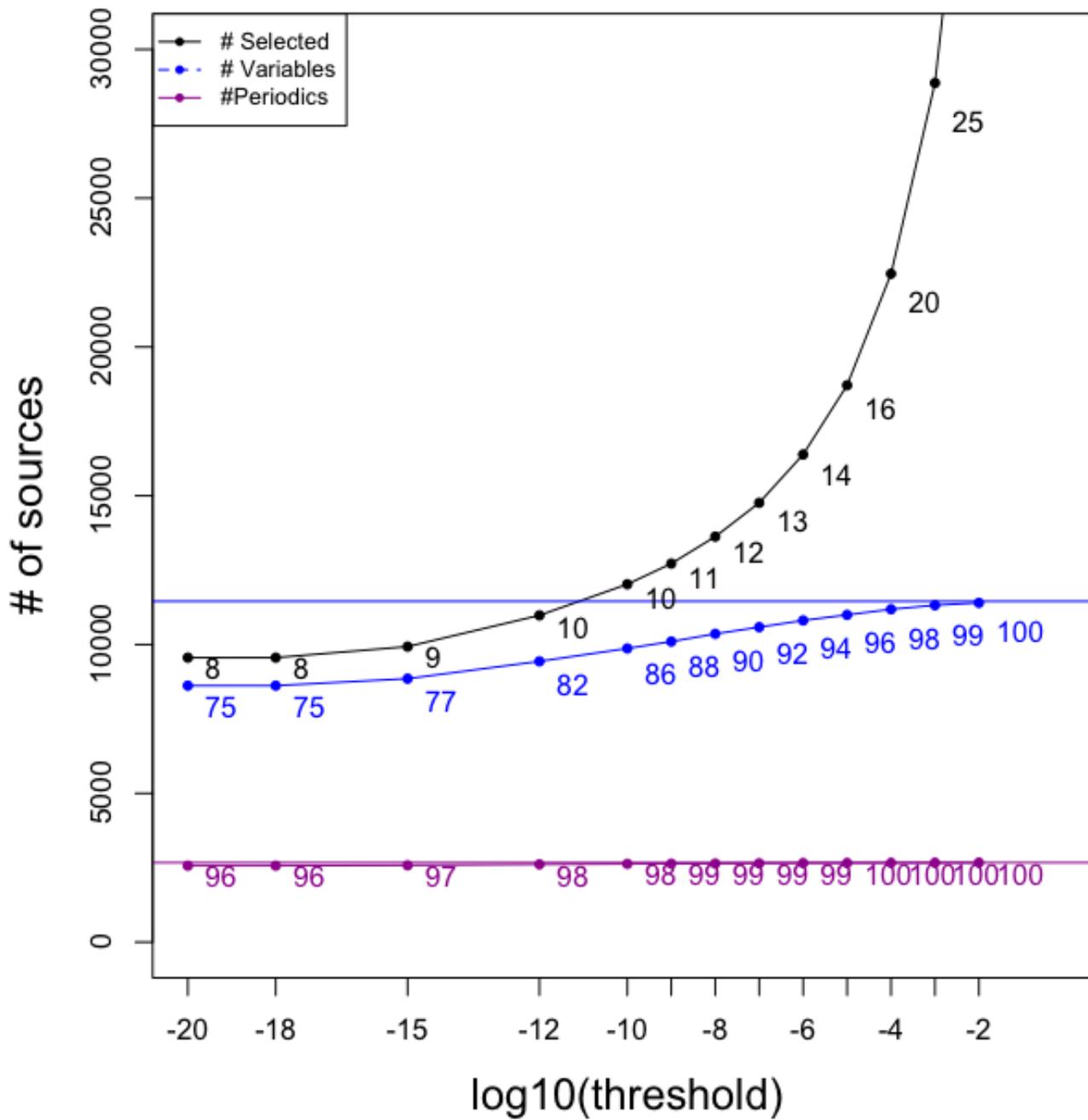
Outliermedian Test



Comparison of tests to select variables

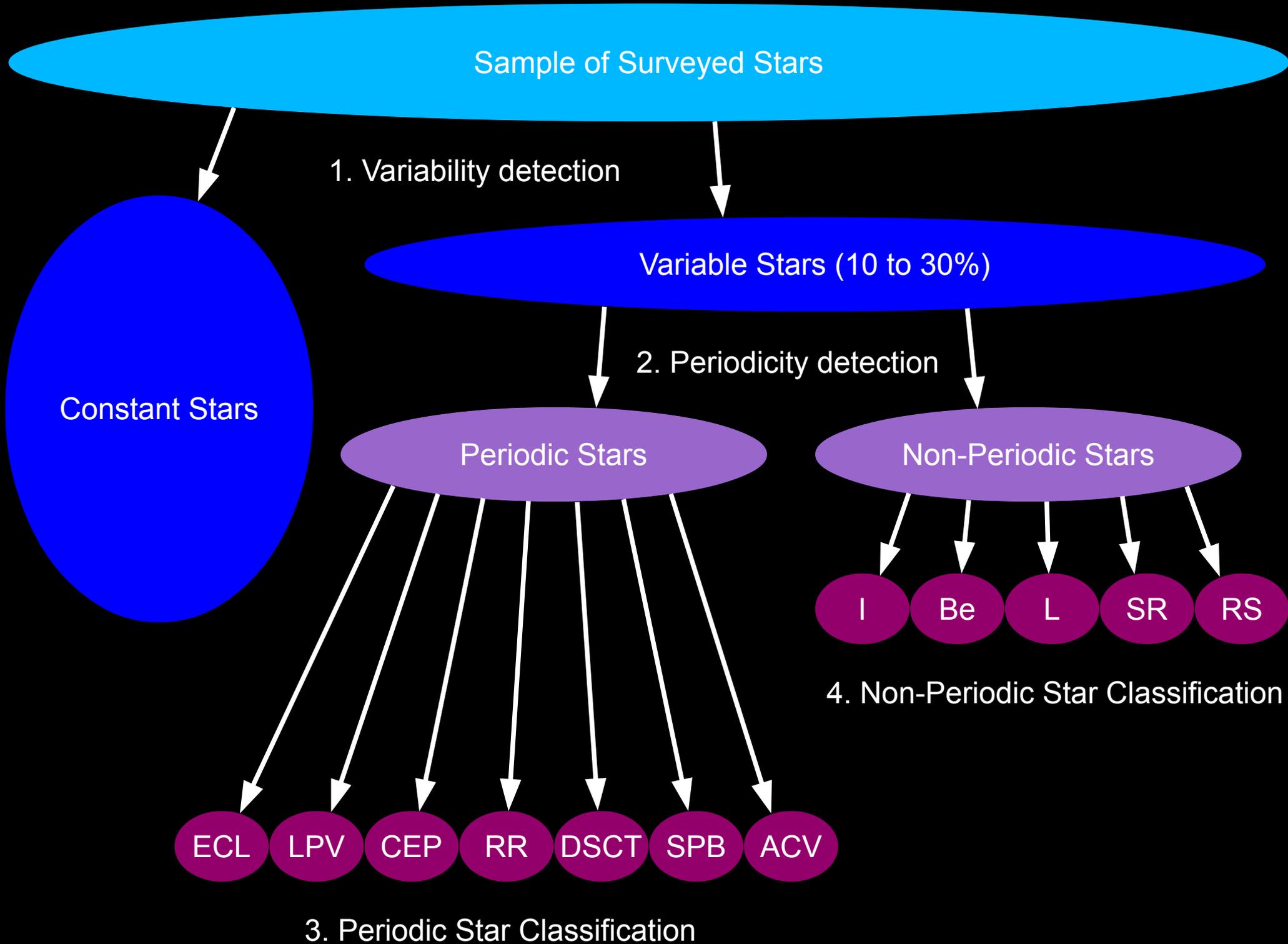


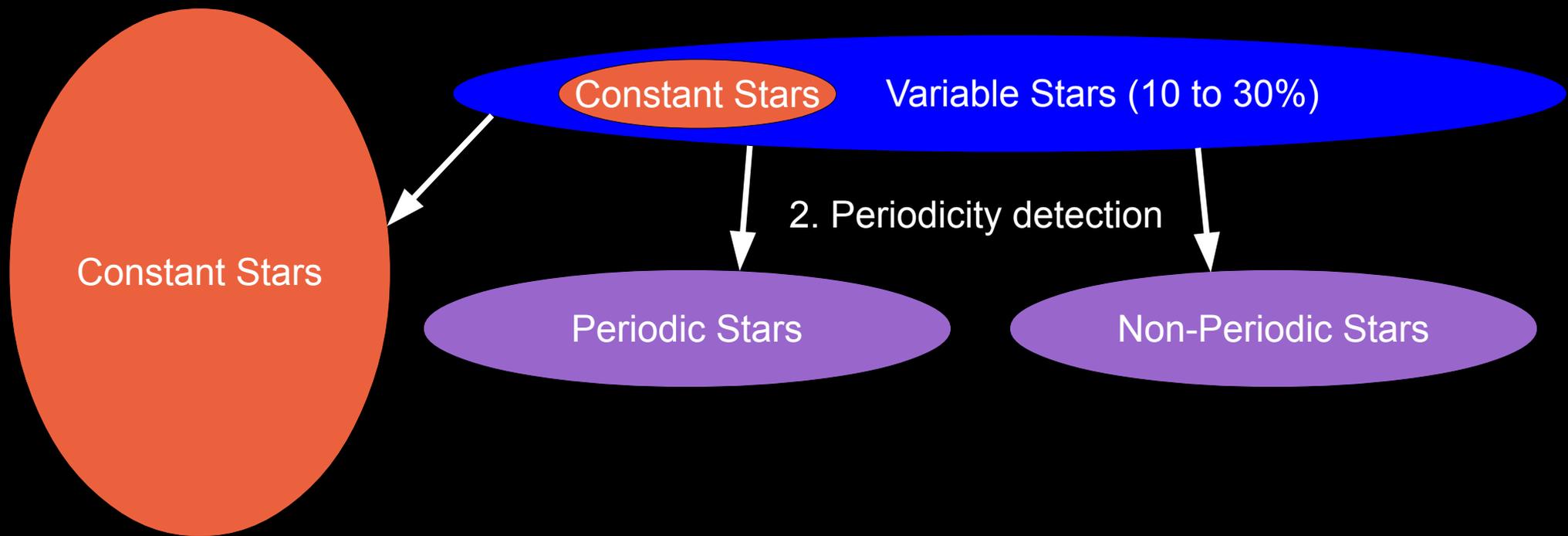
Union (Abbe/Stet/IQR/Chi2/Kur/Out)Test



Choice for variability detection

- Union of Stetson with pValues $< 1e-2$ and Inter-quartile range with pValues $< 1e-3$
 - 17'006 candidates (14.8 % of total)





Constant Stars

Constant Stars

Variable Stars (10 to 30%)

2. Periodicity detection

Periodic Stars

Non-Periodic Stars

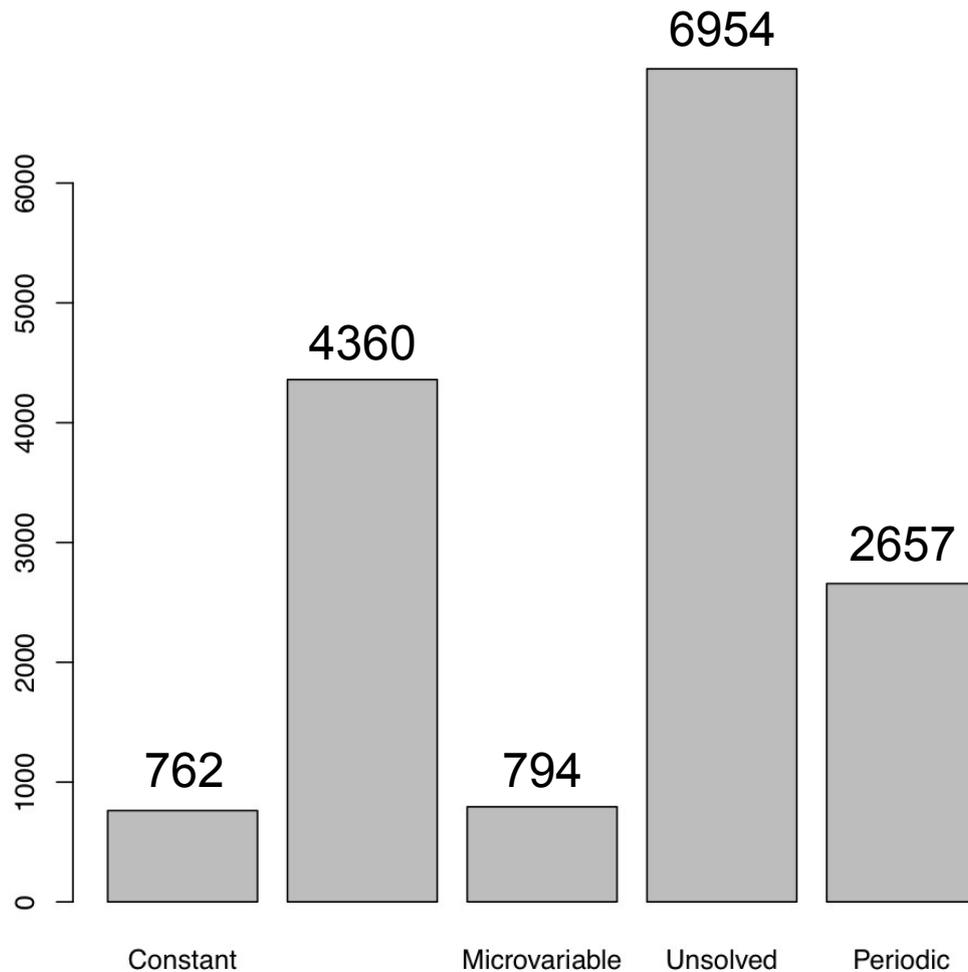
Periodicity detection through a supervised classification

- Compute a number of “attributes” characterizing objects and their light-curve
- Use attributes (features) as predicators (variables) in a supervised classifiers
- Train the classifier with a set of stars of known types
- Use a 10-fold Cross-Validation (CV) to evaluate the performance of this approach

Period search

- Generalized Lomb–Scargle method (Zechmeister & Kürster 2009)
- Our Sample of 15'527 stars includes 3022 stars with known periods ($3022 = 2657 P + 365 U$)
- Recovery rate of 77 % (2323 out of 3022)
 - 1644 with correct period
 - 679 with twice the period values
- Recovery rate for the 2657 periodic = 86 % (i.e., 2300)

Hipparcos variability types



Total : 15527

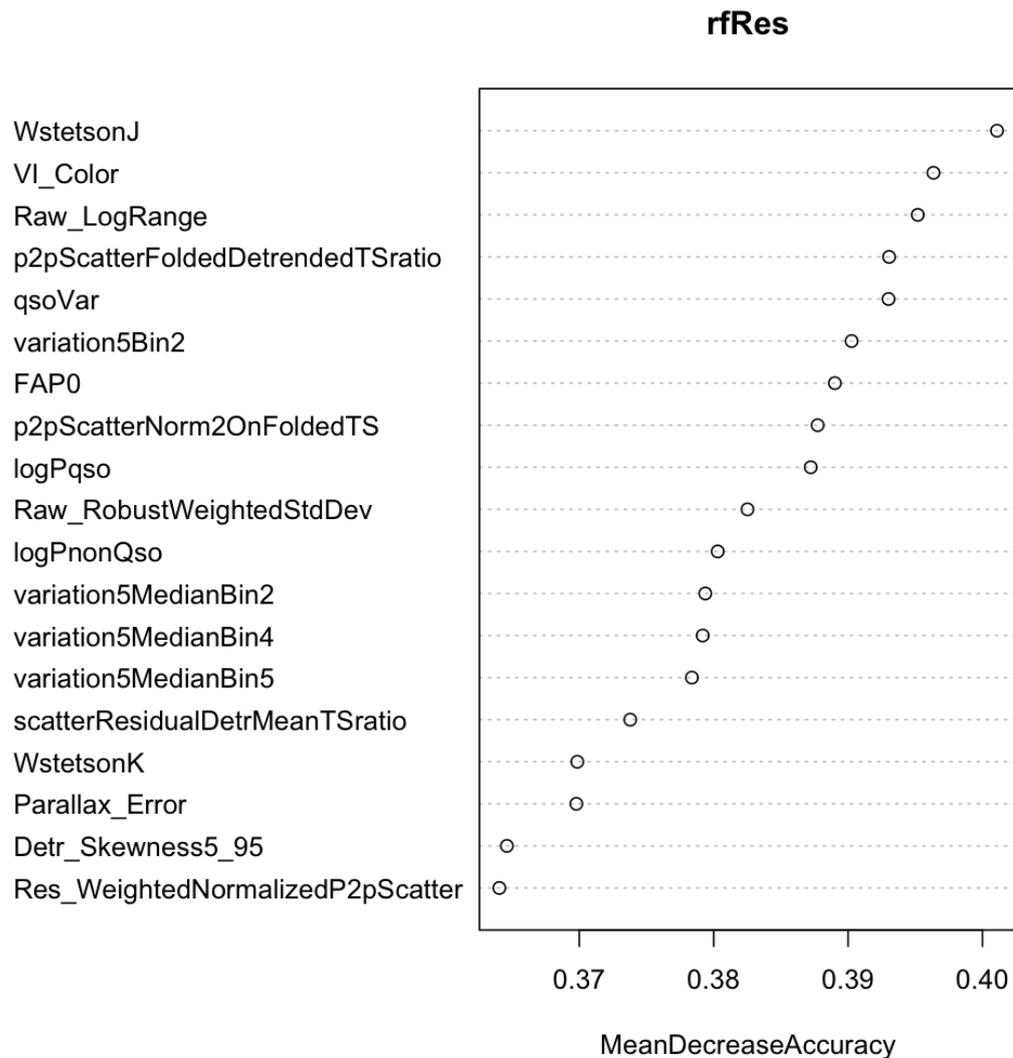
P : 2657 \neq 2712

Hip Types R (662)

D (816)

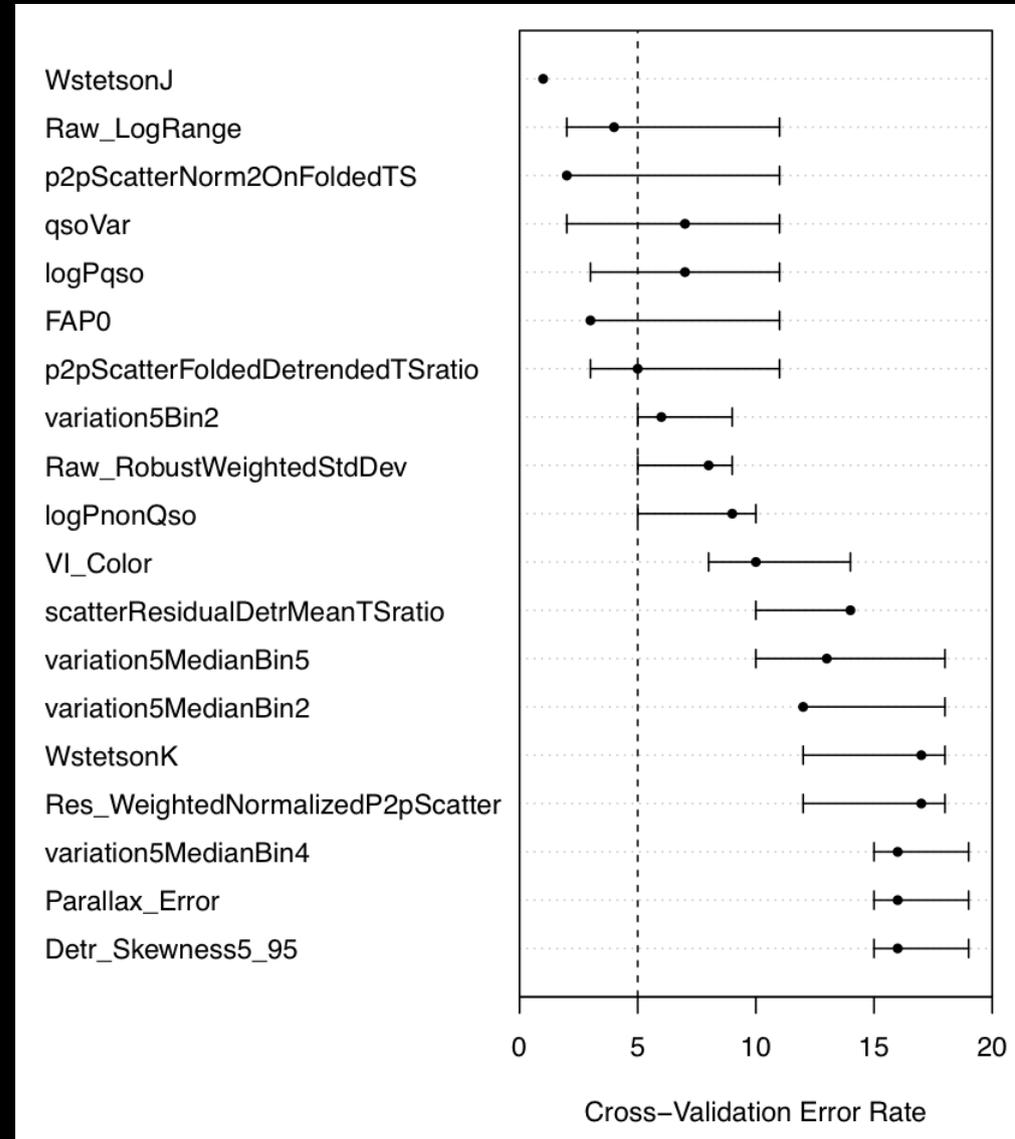
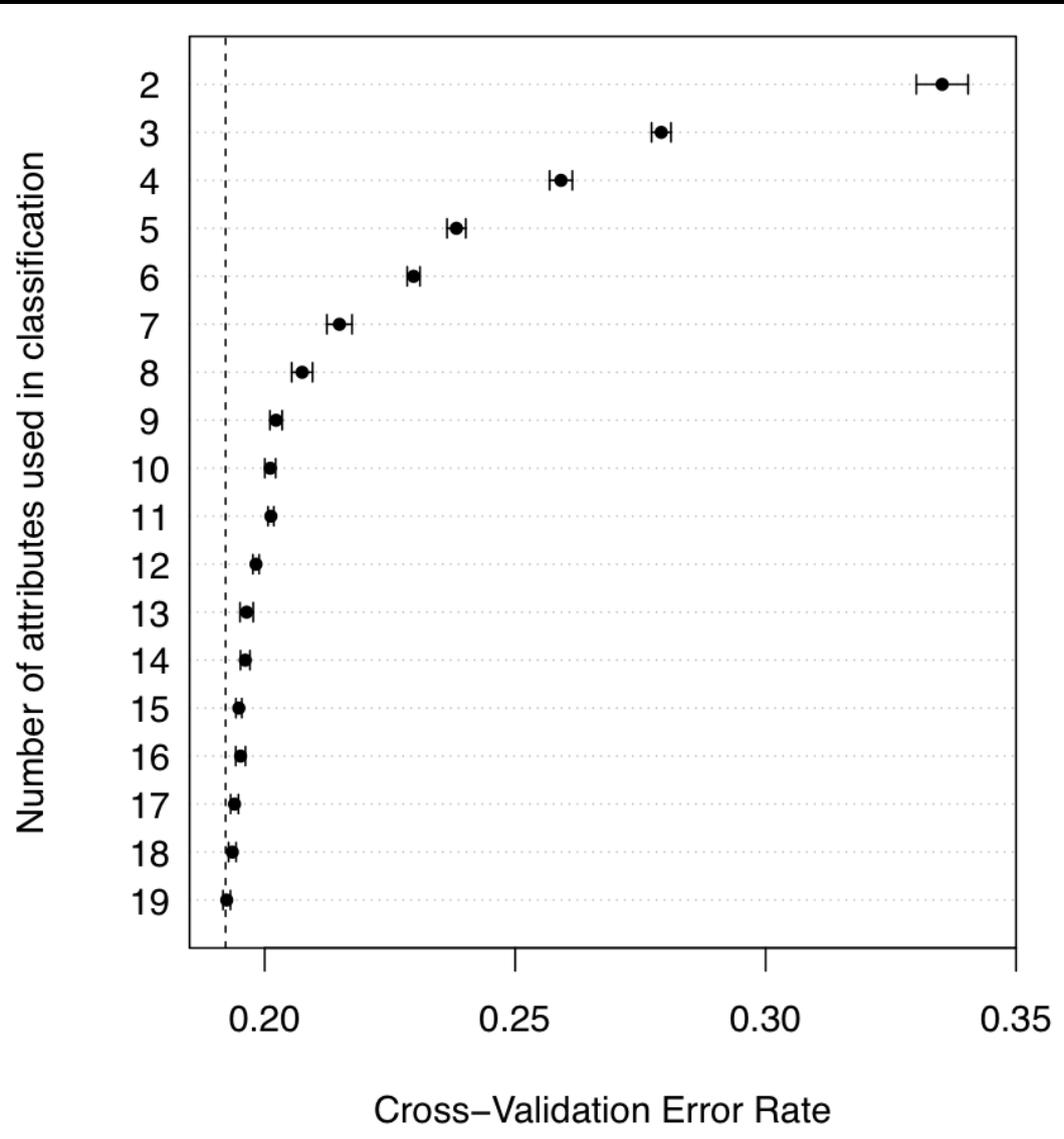
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Random forest attribute importance



- Attributes with Spearman correlation larger than 80% are trimmed

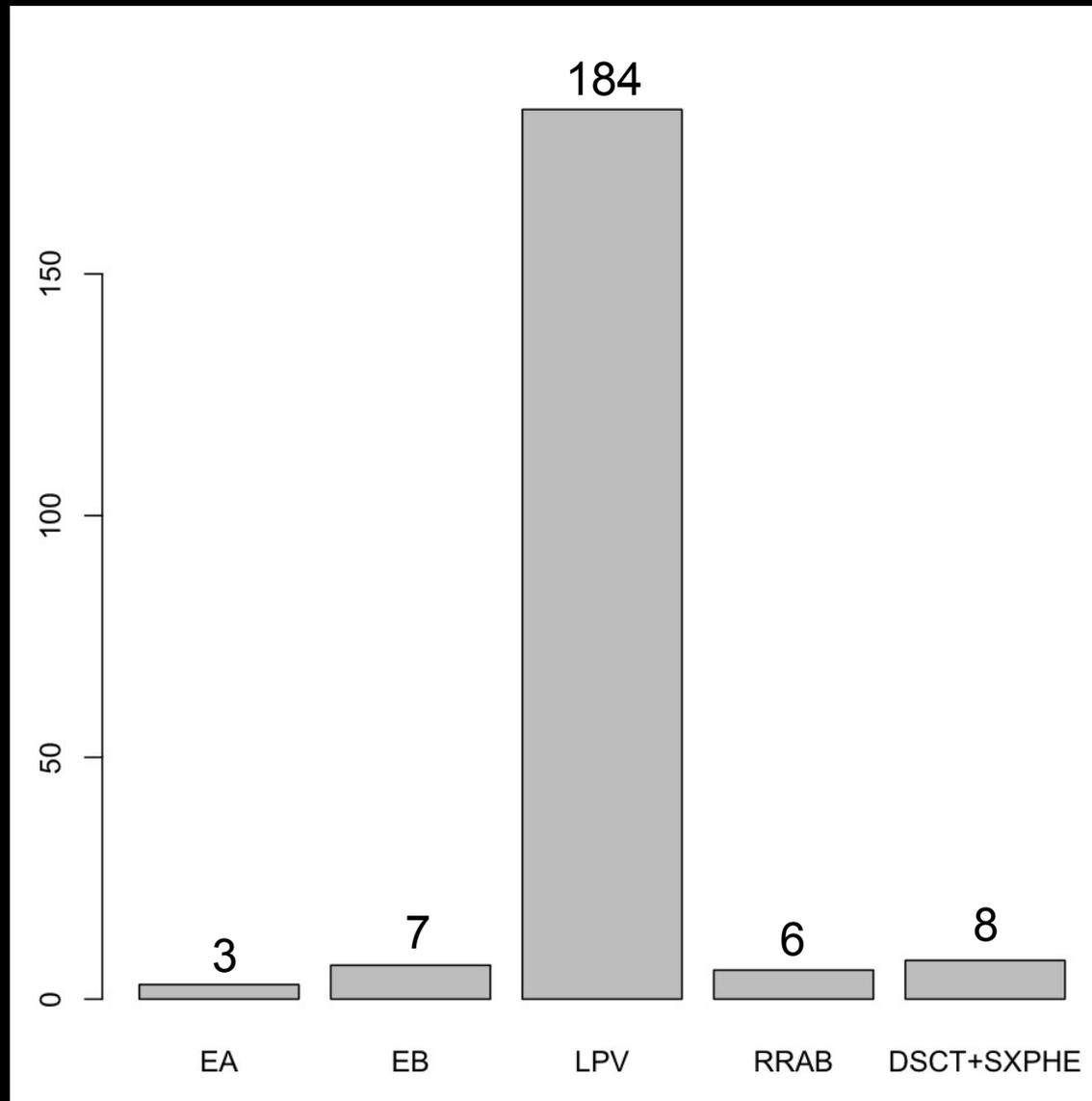
Random forest CV error rates



Random forest confusion matrix

	Constant	Microvariable	Unsolved	Periodic	
Constant	212	549	1		Constant
Microvariable	112	3738	94	405	11
Unsolved		319	313	152	10
Periodic		523	68	6167	196
		33	27	503	2094

Predicted types for non-periodic identified as periodic



Conclusion

- We established a complete scheme for variable star classification
- Optimized for Hipparcos data
 - Training set must be representative of the test set
- Hipparcos classification relatively easy: clean sample and well known stars
- Can be completed with additional information
 - Color light curves
 - Radial velocity time series
- Next step: apply our scheme to other surveys