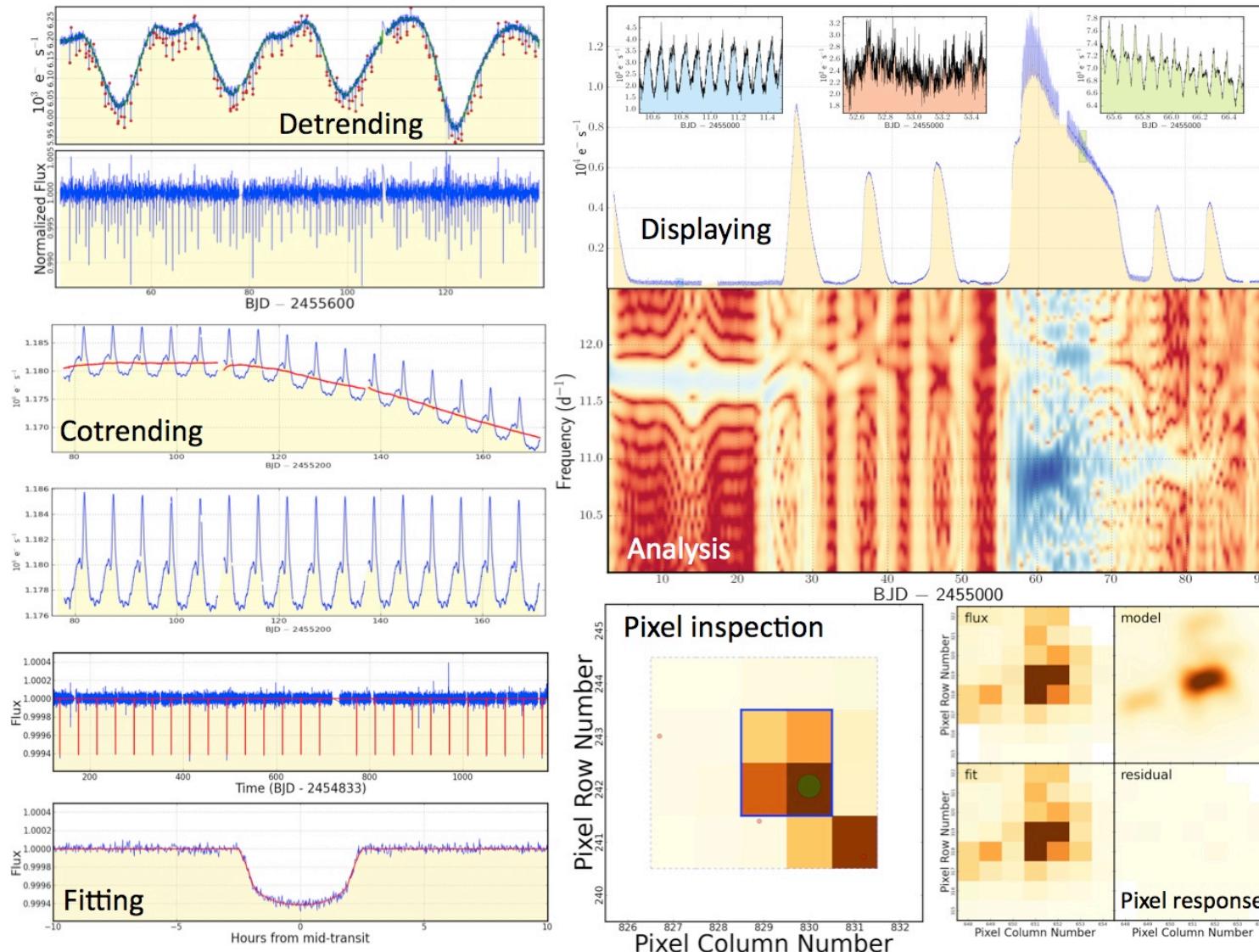


Lecture 3: Ensemble photometry and transit detection

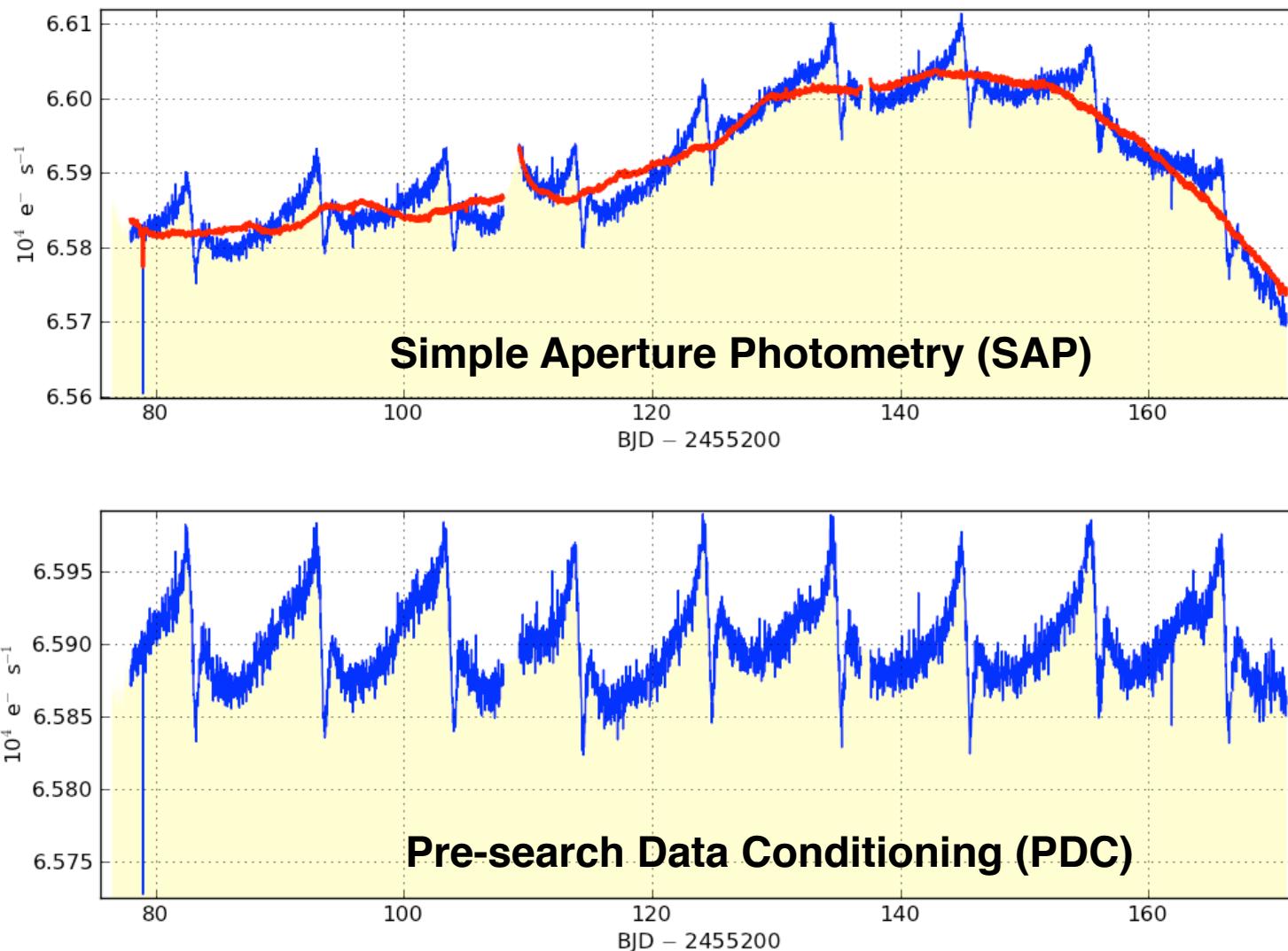
Ensemble differential photometry
Systematics and decorrelation methods
Transit detection by box least-squares
Noise and completeness

Detrending – the PyKE way



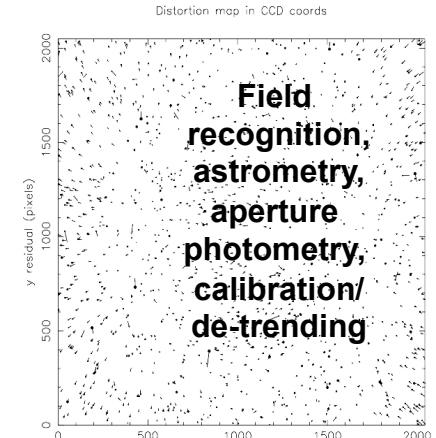
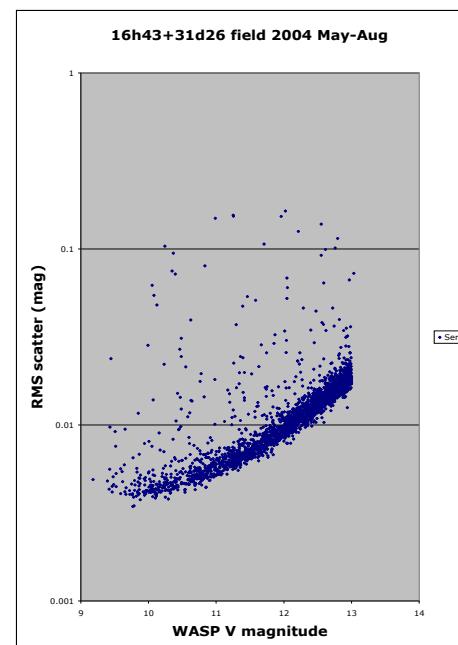
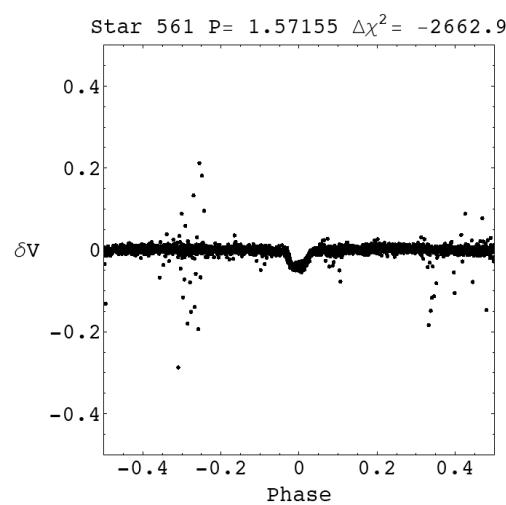
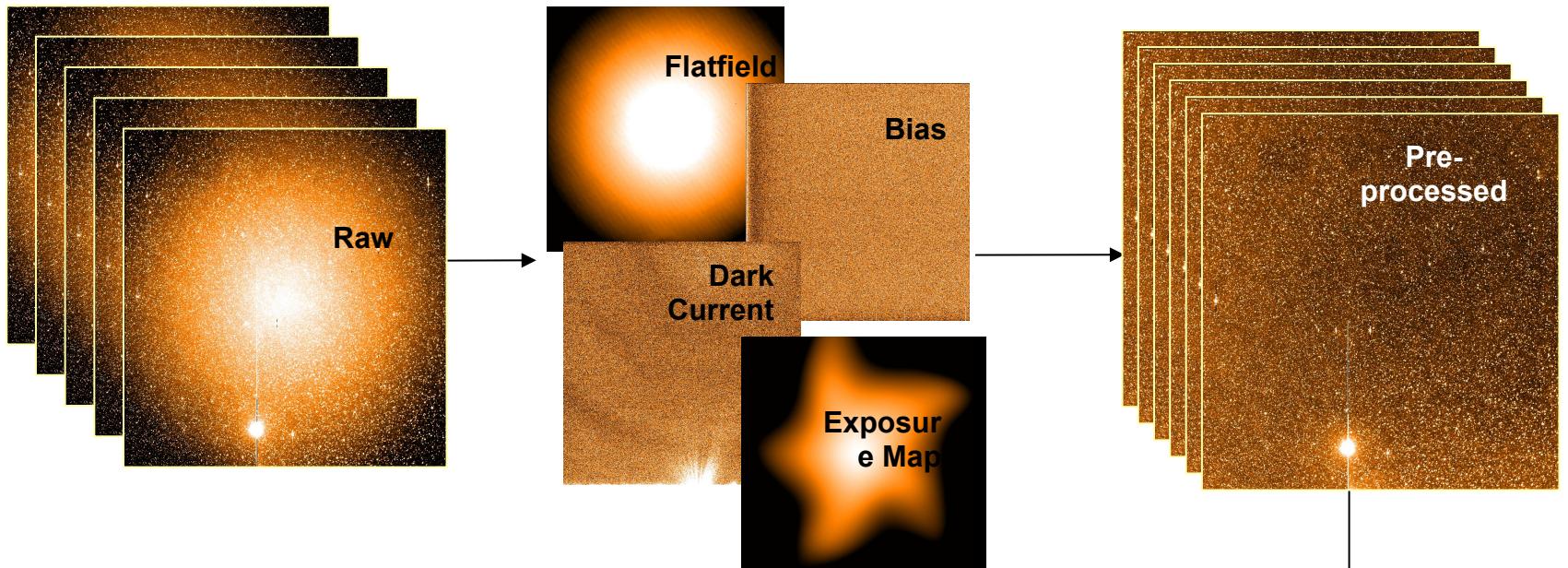
<http://keplerscience.arc.nasa.gov/PyKE.shtml>

Kepler SAP and PDC-SAP



<http://keplerscience.arc.nasa.gov/DataAnalysisProducts.shtml>

WASP data reduction pipeline



Catalogue-driven photometry

- **Step 1: Object detection**
 - e.g (S)EXTRACTOR
 - Catalogue all objects on frame to detection threshold
- **Step 2: Establish astrometric solution**
 - Many standard algorithms available, e.g. astrometry.net
 - Cross-match with astrometric catalogue
 - Write world coordinate system to image.
- **Step 3: Aperture photometry**
 - Optimise aperture – many ways to do this!
 - Use frame catalogue to exclude faint objects in sky annulus

Ensemble differential photometry

- Mean magnitude of star j:

$$\hat{m}_j = \frac{\sum_i m_{ij} w_{ij}}{\sum_i w_{ij}}$$

$$w_{ij} = \frac{1}{\sigma_{ij}^2 + \sigma_{t(i)}^2}$$



Additional intra-frame variance
(downgrades poor images)

- Zero-point correction for frame i:

$$\hat{z}_i = \frac{\sum_j (m_{ij} - \hat{m}_j) u_{ij}}{\sum_j u_{ij}}$$

$$u_{ij} = \frac{1}{\sigma_{ij}^2 + \sigma_{s(j)}^2}$$



Additional stellar variance
(downgrades variable objects)

Frame quality and stellar variability I

- Define data vector X and model μ :

$$X = \{m_{ij}, i = 1...n\} \quad \mu = \{\hat{m}_j + \hat{z}_i, i = 1...n\}$$

- Assuming Gaussian errors:

$$\begin{aligned} P(X_i | \mu_i) &= \frac{1}{\sqrt{2\pi} \sqrt{\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2}} \\ &\times \exp \left\{ -\frac{(m_{ij} - \hat{m}_j - \hat{z}_i)^2}{2[\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2]} \right\} \end{aligned}$$

- Likelihood of entire data vector for star j :

$$L(\mu) = (2\pi)^{-n/2} \prod_i \left[\frac{1}{\sqrt{\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2}} \right] \exp \left(-\frac{1}{2} \chi^2 \right)$$

- Where:

$$\chi^2 = \sum_i \frac{(m_{ij} - \hat{m}_j - \hat{z}_i)^2}{\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2}$$

Frame quality and stellar variability II

- Maximise likelihood w.r.t. $\sigma^2_{s(j)}$ and $\sigma^2_{t(i)}$:

- Solve iteratively for $\sigma^2_{s(j)}$ holding $\sigma^2_{t(i)}$ constant:

$$\sum_i \frac{1}{\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2} - \sum_i \frac{(m_{ij} - \hat{m}_j - \hat{z}_i)^2}{[\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2]^2} = 0$$

- Solve iteratively for $\sigma^2_{t(i)}$ holding $\sigma^2_{s(j)}$ constant:

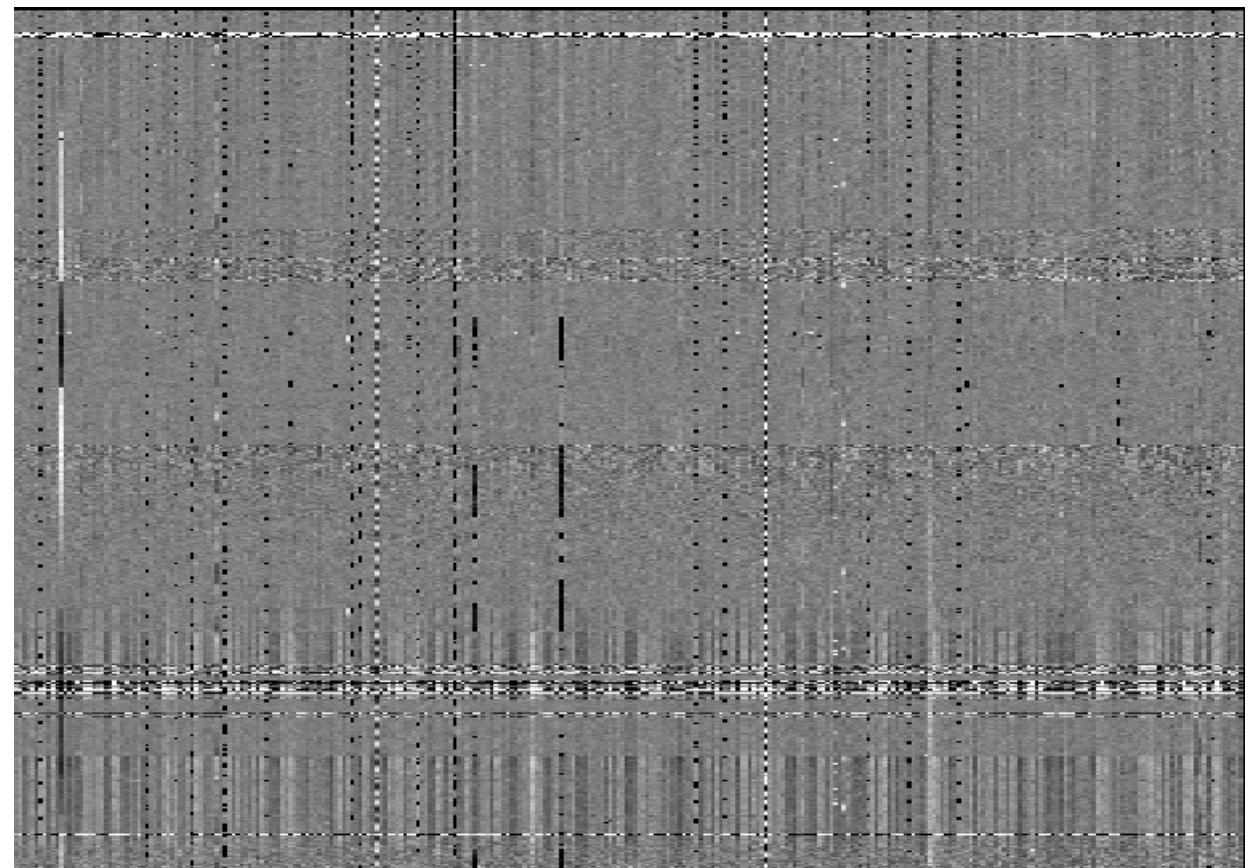
$$\sum_j \frac{1}{\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2} - \sum_j \frac{(m_{ij} - \hat{m}_j - \hat{z}_i)^2}{[\sigma_{ij}^2 + \sigma_{t(i)}^2 + \sigma_{s(j)}^2]^2} = 0$$

- Iterate \hat{m}_j , \hat{z}_i , $\sigma^2_{s(j)}$ and $\sigma^2_{t(i)}$ to convergence.

Decorrelation/ systematics removal

Tamuz et al 2005

**16h30+28 field
300 stars
2549 observations
100 days**



Aspen 2005
Vietri Sul Mare 2015

The Transit Method

Systematics and correlated noise

- **References:**

- Tamuz, Mazeh & Zucker 2005, MNRAS 355, 1466 (SysRem)
- Kovacs, Bakos & Noyes 2005, MNRAS 356, 537 (TFA)
- Pont, Zucker & Queloz 2006, MNRAS 373, 231
- Carter & Winn 2009, ApJ 704, 51
- Gibson 2014, MNRAS 445, 3401

SysRem

- Many patterns of systematics are common to all stars
 - Secondary extinction: dependence on stellar colour
 - Sky brightness: Dependence on target brightness
 - Ambient temperature and focus drift: position dependent?
- SysRem: PCA with error bars!
 - Construct common temporal basis functions
 - Compute optimal scaling coefficient per star
 - Remove and repeat.

SysRem algorithm

- **SysRem produces a corrected magnitude given by:**

$$\tilde{x}_{i,j} = x_{i,j} - \sum_{k=1}^M {}^{(k)}c_j {}^{(k)}a_i$$

- **Search for the best c_j that minimises:**

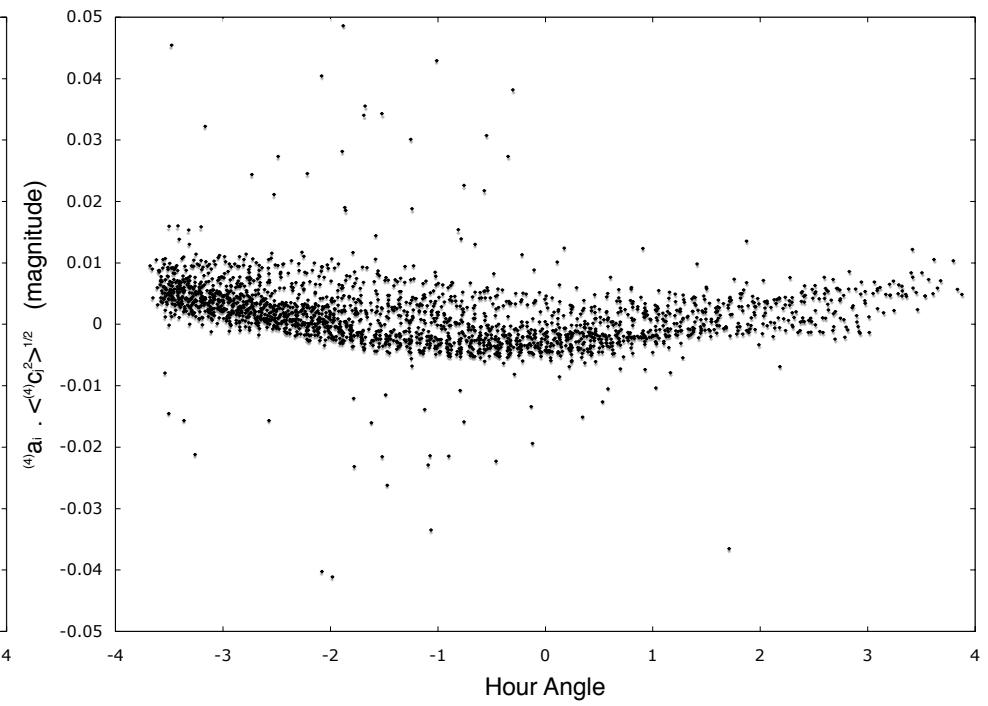
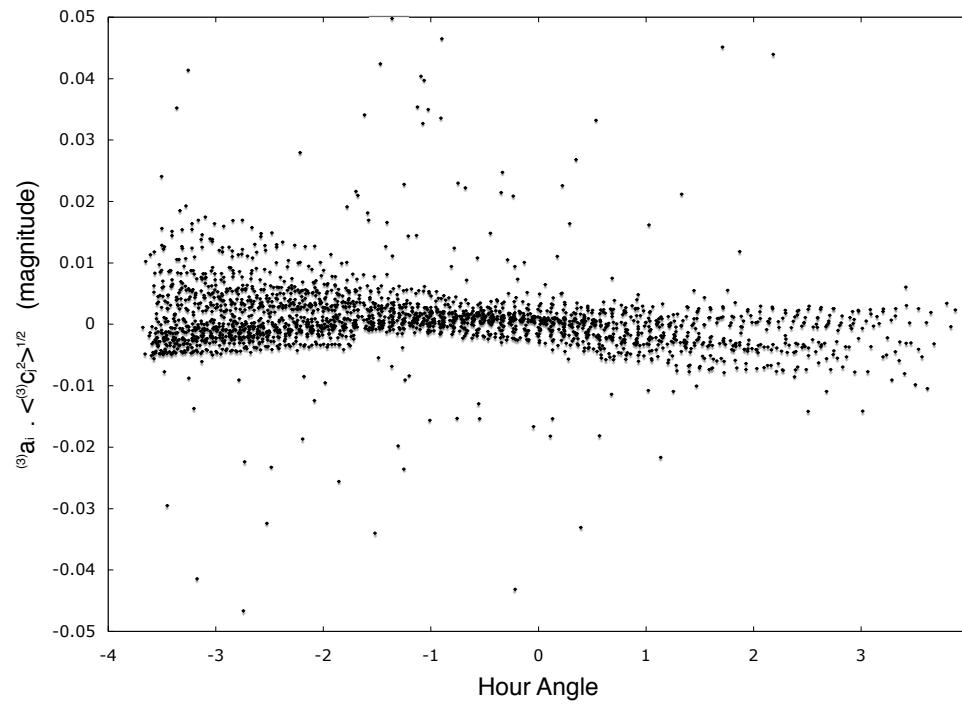
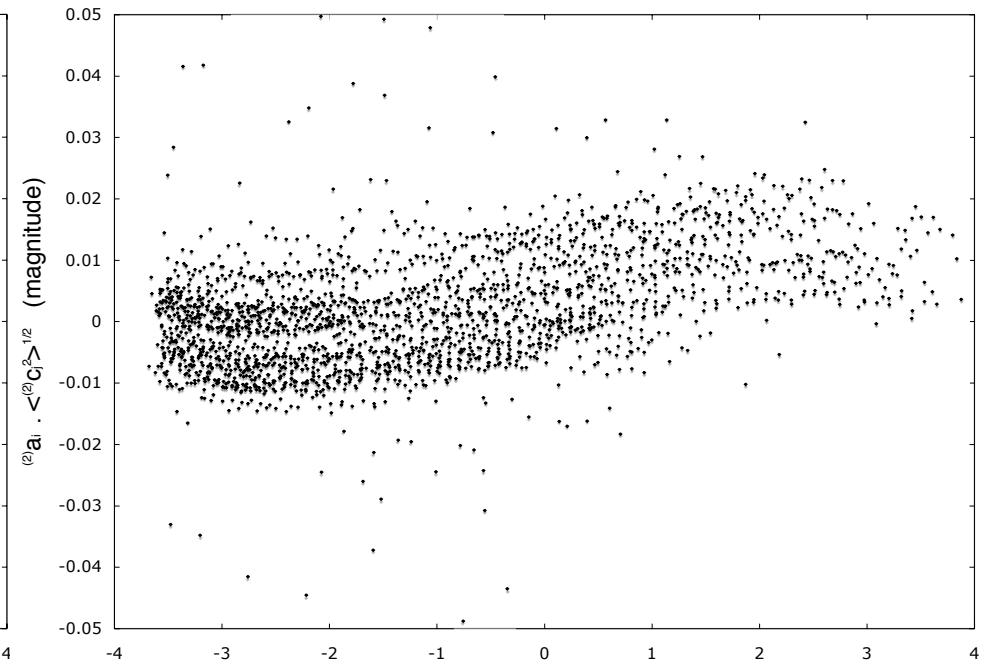
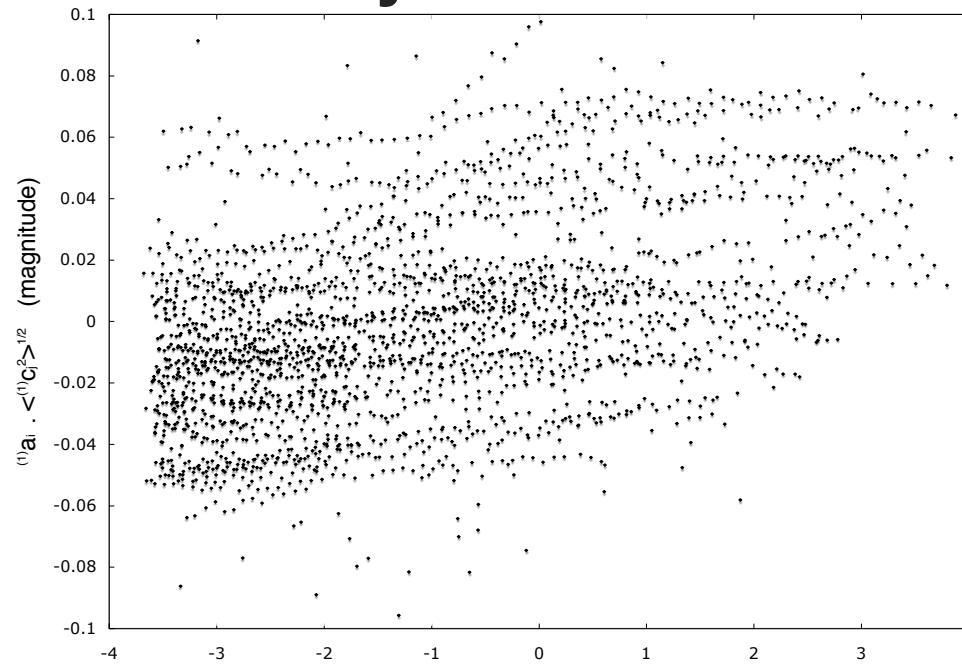
$${}^{(k-1)}S_i^2 = \sum_{i,j} ({}^{(k-1)}\tilde{x}_{i,j} - {}^{(k)}c_i {}^{(k)}a_j)^2 w_{i,j}$$

$${}^{(k)}c_j = \sum_i \frac{{}^{(k-1)}\tilde{x}_{i,j} {}^{(k-1)}a_i w_{i,j}}{({}^{(k-1)}a_i)^2 w_{i,j}}$$

- **Similarly**

$${}^{(k)}a_i = \sum_j \frac{{}^{(k-1)}\tilde{x}_{i,j} {}^{(k-1)}c_j w_{i,j}}{({}^{(k-1)}c_j)^2 w_{i,j}}$$

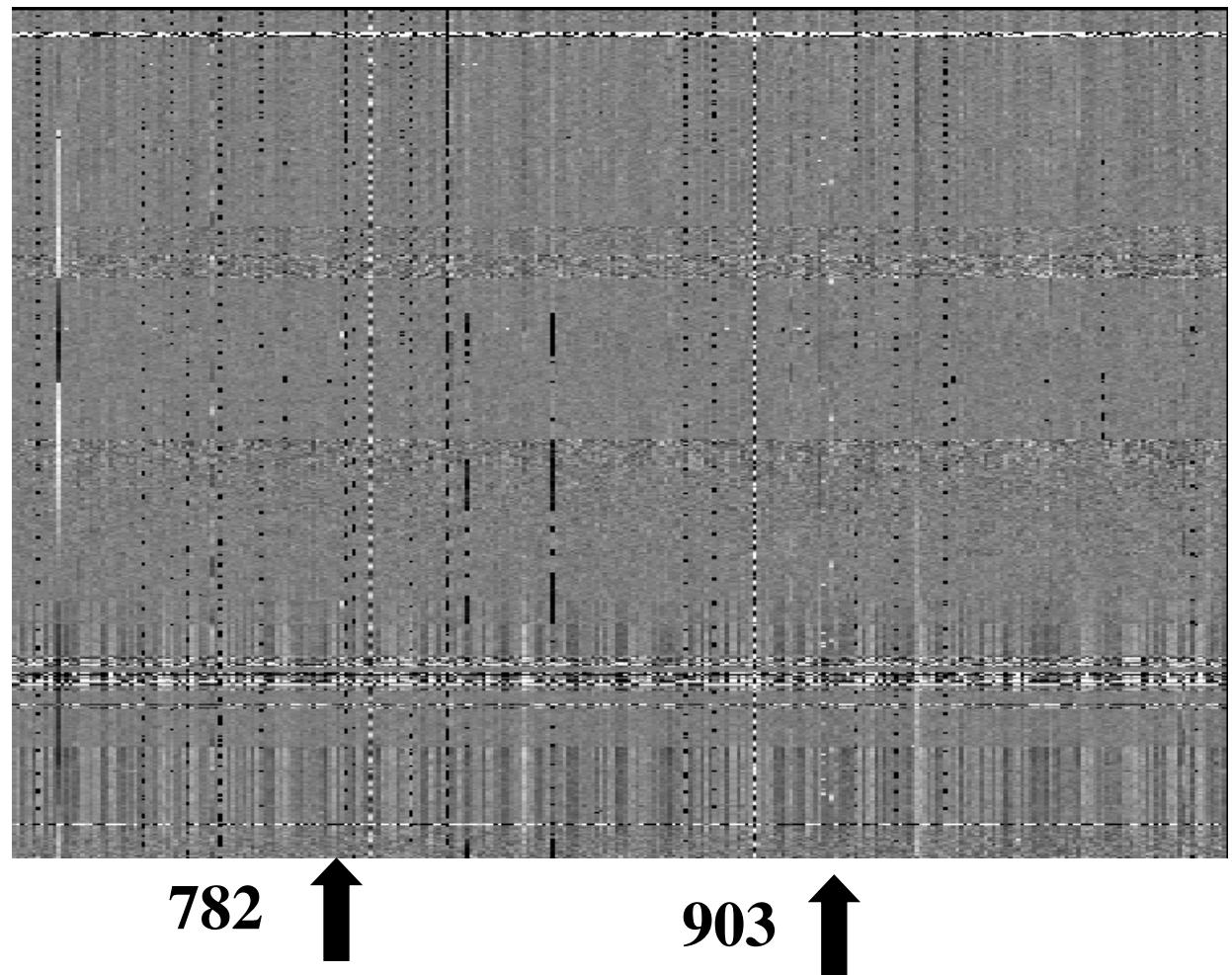
SysRem basis functions vs HA



Decorrelation/ systematics removal

Tamuz et al 2005

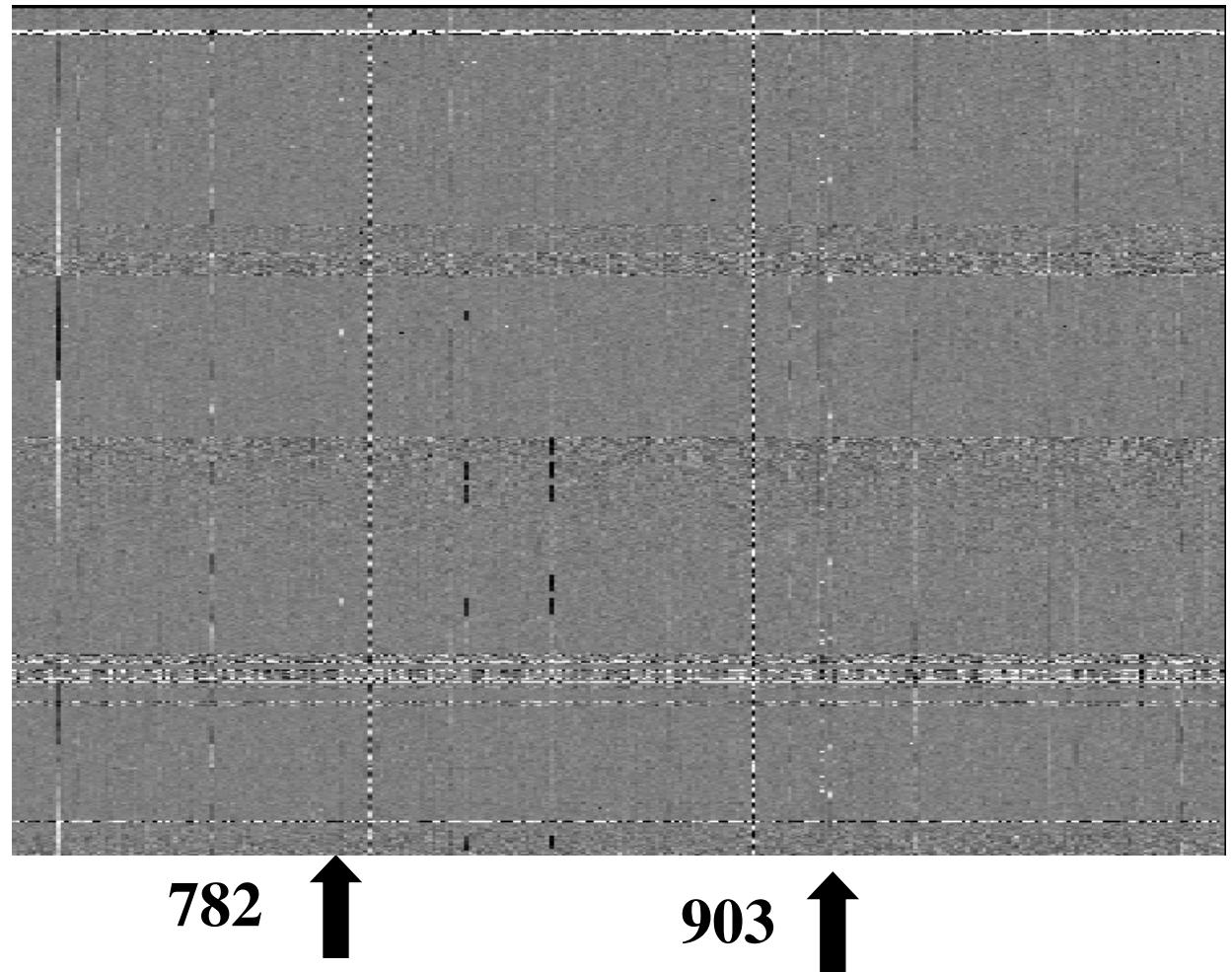
**16h30+28 field
300 stars
2549 observations
100 days**



Decorrelation/ systematics removal

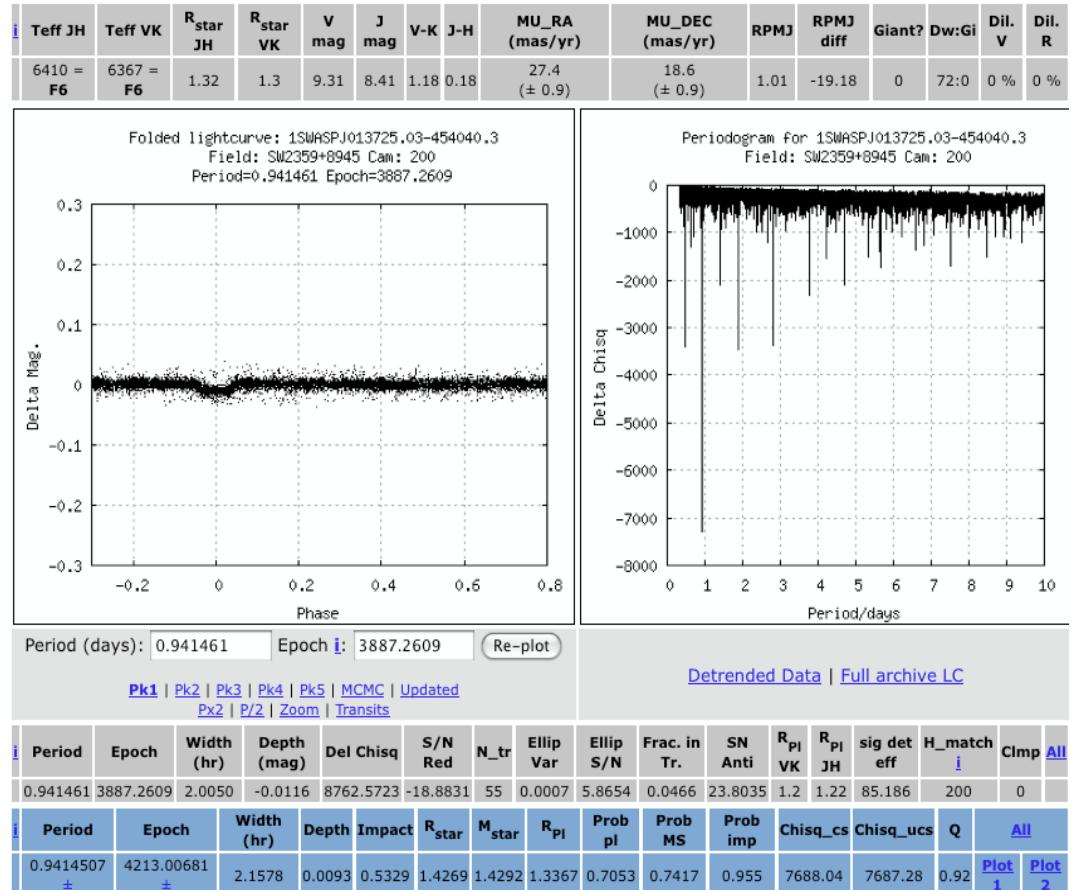
Tamuz et al 2005

**16h30+28 field
300 stars
2549 observations
100 days**



Decorrelation & Detection

- **Decorrelation:**
 - 4-stage SysRem
 - TFA with reconstruction
- **Detection: Cameron et al 2006 MNRAS 373, 799**
 - Accelerated BLS algorithm
 - Coarse grid search
 - Newton-Raphson peak-up on 5 strongest peaks
 - Box width varies with period
- **Example: WASP-1**



Box Least-Squares transit search

- **References**
 - Kovacs, Zucker & Mazeh 2002, A&A 391, 369
 - Aigrain & Irwin 2004, MNRAS 350, 331
 - Collier Cameron et al 2006, MNRAS 373, 799
- **Step 1: Compute inverse variance-weighted mean flux, subtract and precompute χ^2_0 , t :**

$$\hat{x} = \frac{\sum_i \tilde{x}_i w_i}{\sum_i w_i} \quad x_i = \tilde{x}_i - \hat{x}$$
$$\chi^2_0 = \sum_i x_i^2 w_i \quad t = \sum_i w_i$$

- **Step 2: set up frequency grid**
 - phase of each observation must change by less than transit duration between adjacent frequencies
- **Step 3: estimate range of transit duration**
 - Use stellar density

Box Least-Squares transit search

- **Step 3: Phase-fold and partition:**
 - Mean light level in-transit (L):

$$s = \sum_{i \in \ell} x_i w_i, \quad r = \sum_{i \in \ell} w_i, \quad q = \sum_{i \in \ell} x_i^2 w_i.$$

$$L = \frac{s}{r}$$

$$\text{Var}(L) = \frac{1}{r}$$

- Mean light level out of transit (H):

$$H = \frac{-s}{t - r}$$

$$\text{Var}(H) = \frac{1}{t - r}$$

- Fitted transit depth (δ):

$$\delta = L - H = \frac{st}{r(t - r)}$$

$$\text{Var}(\delta) = \frac{t}{r(t - r)}$$

Box Least-Squares transit search

- **Step 4: Badness-of-fit:**
 - Signal-to-noise ratio of transit depth

$$S/N = s \sqrt{\frac{t}{r(t-r)}}$$

- Improvement in fit relative to constant model is $(S/N)^2$:

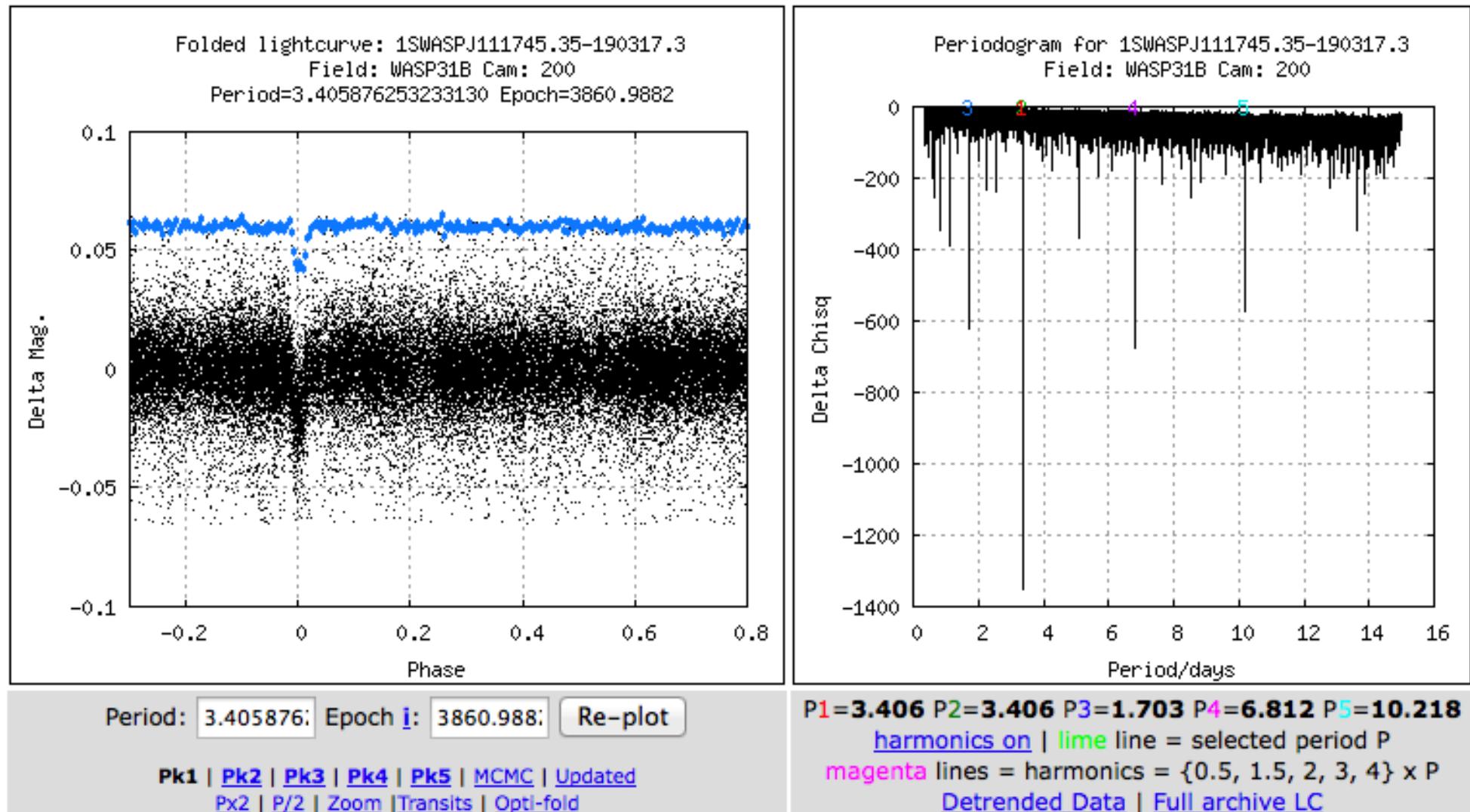
$$\Delta\chi^2 = \frac{s^2 t}{r(t-r)}$$

- Goodness of fit outside transit:

$$\chi_h^2 = \chi_0^2 - \frac{s^2}{(t-r)} - q$$

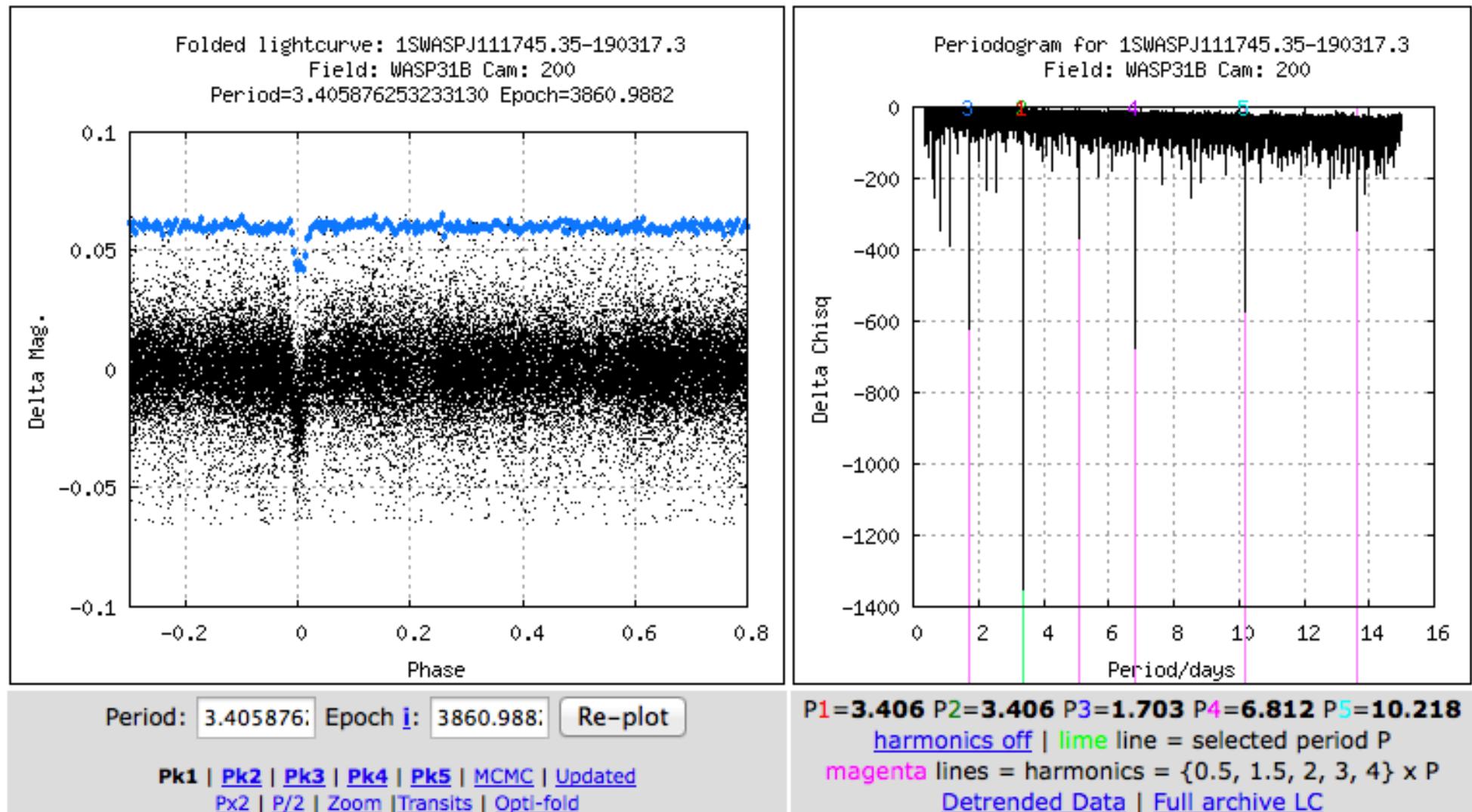
Box Least-Squares transit search

- Step 5: BLS periodogram



Box Least-Squares transit search

- Step 5: BLS periodogram



1SWASPJ022837.22-070338.4

Field = [WASP77B](#), Cam = 200, H_run = ORKP_TAMUZ

SW Vt=10.3391 | Pts_gd=9759 | TSTART=2008-07-30 00:36:17
| TSTOP=2009-12-14 23:26:45 | Pmin=0.35 | Pmax=10

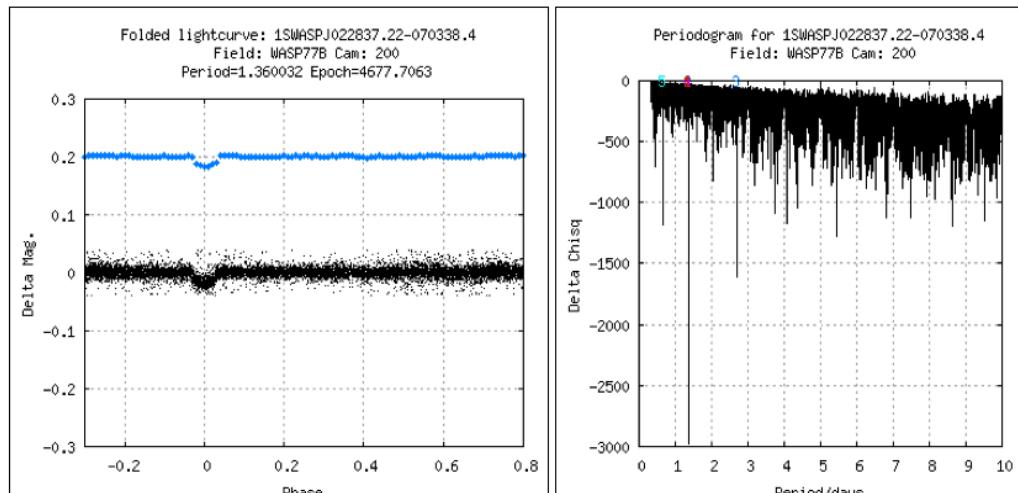
[Thumbnails](#) | [VSI2.0](#) | [Aperture Blends?](#) | [Nearby SW objects?](#) | [RPM](#) | [RPM\(new\)](#)

[Blends](#)

[Param. Fit](#) | [Phase predictor](#) | [Transit scheduler](#) | [Hunt1star](#) | [SWBLS](#) | [SWFOLD](#)

[Alternative Fields/Cams](#) i
[J022837 200 9769 \(ORFG_TAMTFA\)](#)
[J022837 200 9759 \(ORFG_TAMUZ\)](#)
[J022837 200 9737 \(ORFG_TFA\)](#)
[SW0217-0453 224 5594 \(OBOR_TAMTFA\)](#)
[SW0217-0453 224 3316 \(OBOR_TAMUZ\)](#)
[WASP77B 200 9770 \(ORKP_TAMTFA\)](#)
[WASP77B 200 9759 \(ORKP_TAMUZ\)](#)
[WASP77B 200 9736 \(ORKP_TFA\)](#)

Teff JH	Teff VK	Rstar JH	Rstar VK	V mag	J mag	V-K	J-H	MU_RA (mas/yr)	MU_DEC (mas/yr)	RPMJ	RPMJ diff	Giant?	Dw:Gi	Dil. V	Dil. R
5475 = G9	5458 = G9	0.89	0.88	10.30	8.77	1.90	0.39	89.8 (± 1.8)	9.9 (± 1.1)	3.55	-3.2	0	158:1	0 %	0 %



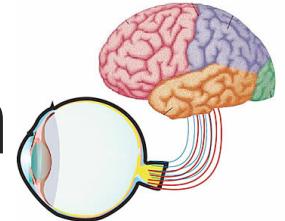
Period	Epoch	Width (hr)	Depth (mag)	Del Chisq	S/N Red	N_tr	Ellip Var	Ellip S/N	Frac. in Tr.	SN Anti	Rpi VK	Rpi JH	sig det eff	H_match	Climp	All
1.360032	4677.7063	1.9748	-0.0188	3591.0742	-15.6899	30	0.0005	4.3669	0.0300	5.0434	1.03	1.04	29.4944	10	0	
Period	Epoch	Width (hr)	Depth	Impact	Rstar	Mstar	Rpi	Prob pi	Prob MS	Prob imp	Chisq_cs	Chisq_ucs	Q		All	
1.3600363	4889.87036	2.1445	0.0136	0.106	0.915	0.868	1.039	1	0.965	1	9756.92	9758.48	0.18	Plot 1	Plot 2	

Updated ephemeris: Period = 1.360032 Epoch = 4677.7063 Source = ORFG_TAMUZ (Mar 11) [Update](#)

[Request Observations](#)

Comments		Flags	
2012-02-04 20:47:46	PFLM	Checked for rotational modulation. P~15d comes up in both seasons with FAP = 0.01 and 0.1. Combined analysis of both seasons gives P=15.3, amp=3mmag and FAP=0.0055. Plot uploaded. (J022837/ORFG_TAMTFA)	<input type="radio"/> A <input type="radio"/> B <input type="radio"/> C <input type="radio"/> D <input checked="" type="radio"/> P <input type="radio"/> EB <input type="radio"/> BI <input type="radio"/> X <input type="radio"/> EBLM <input type="radio"/> V <input type="radio"/> RAF <input type="radio"/> Q
2012-02-03 11:08:50	CH	Flag changed to P (J022837/ORFG_TAMTFA)	
2012-02-03 11:08:50	CH	WASP-77 (J022837/ORFG_TAMTFA)	
2012-02-03 09:57:51	BS	WDS 02286-0704 mags 10.38+13.40, sep 2.9" (SW0217-0453/0BOR_TAMTFA)	
New comments:		Followup Flag =	
<input type="text"/> <input type="button" value="Submit"/>			
Initials (required) ACC <input type="button" value="Add comment / Change Flag"/> User uploaded Files = 15 files			

Planet search



SysRem+BLS

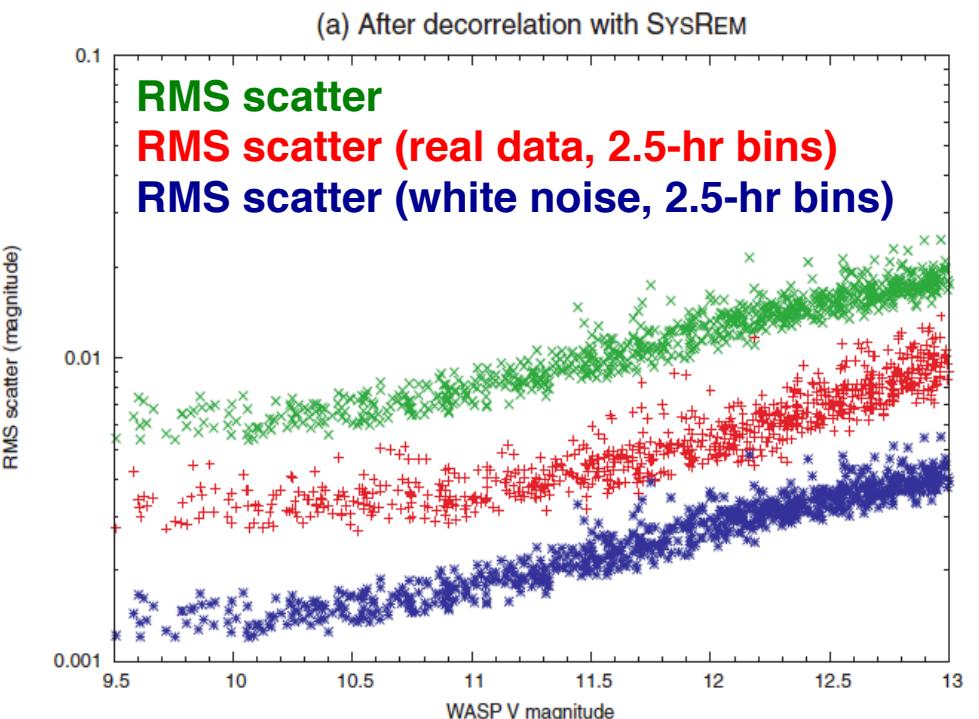
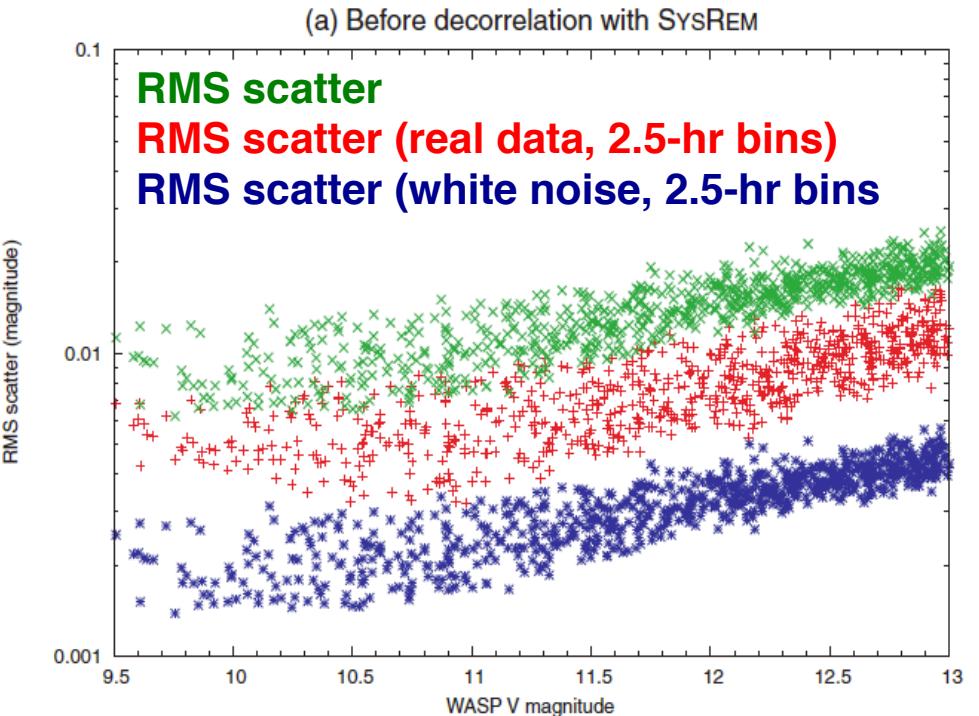
- Tamuz, Mazeh & Zucker 2005
- Kovacs, Zucker & Mazeh 2002
- Skype-linked network of human neuro-visual processors.
- Each processor allocated several thousand candidate light curves + associated statistics from main database.
- Motivated by promises of food, postdoc employment, fame, glory, etc.

The Transit Method

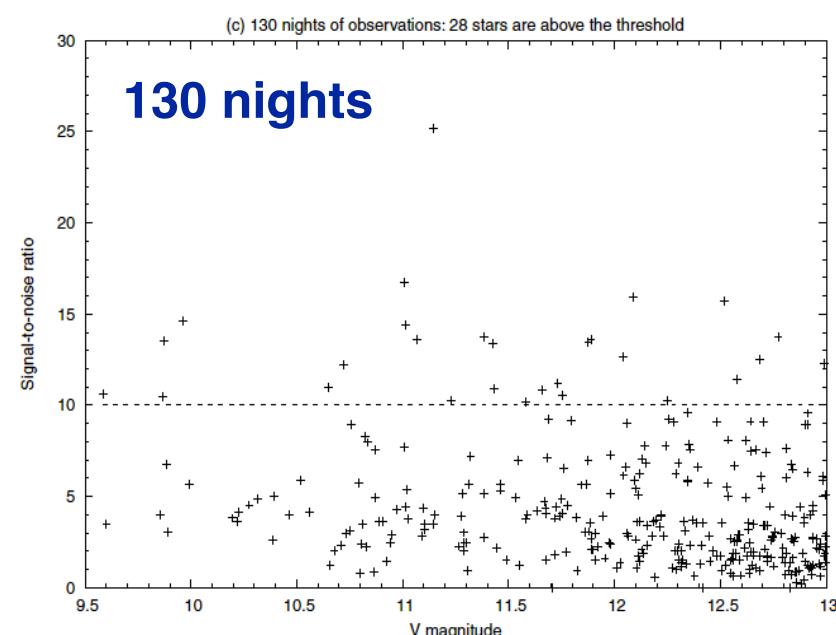
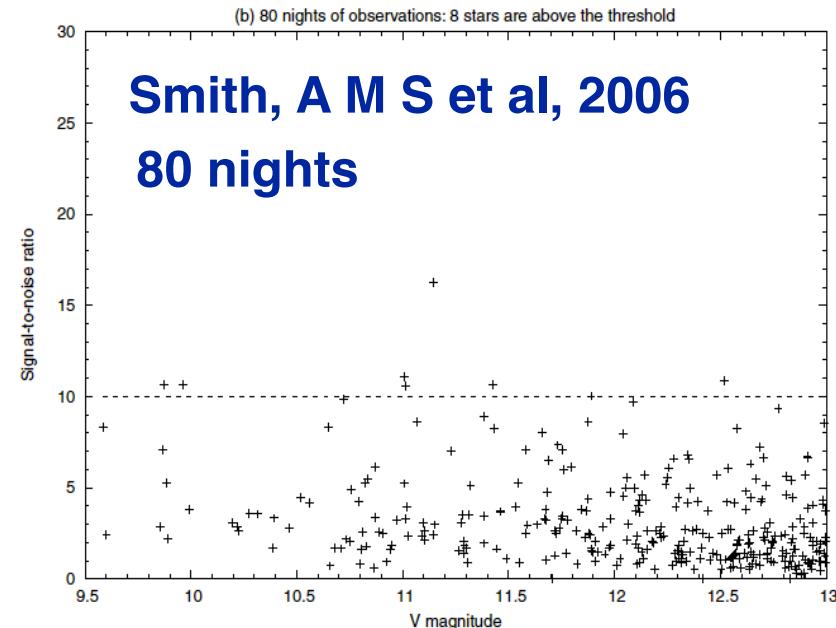
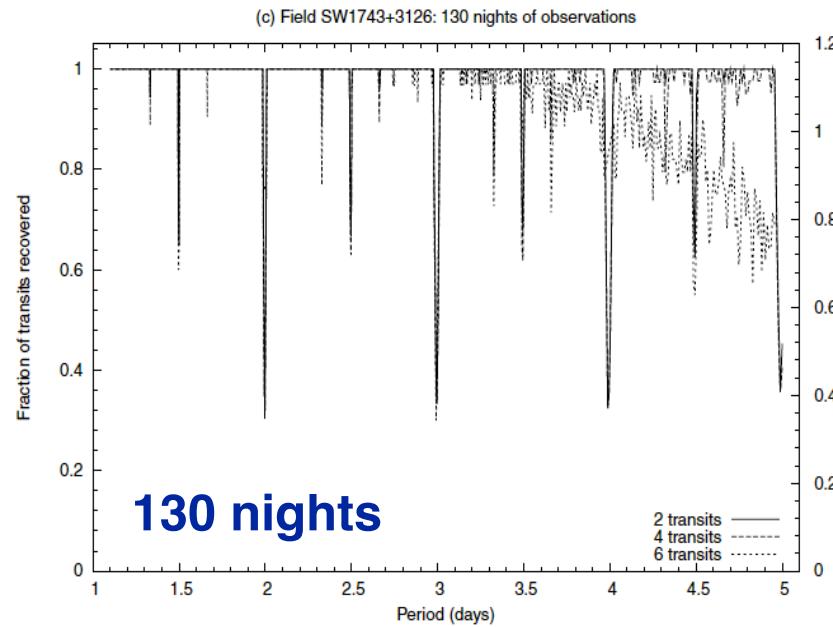
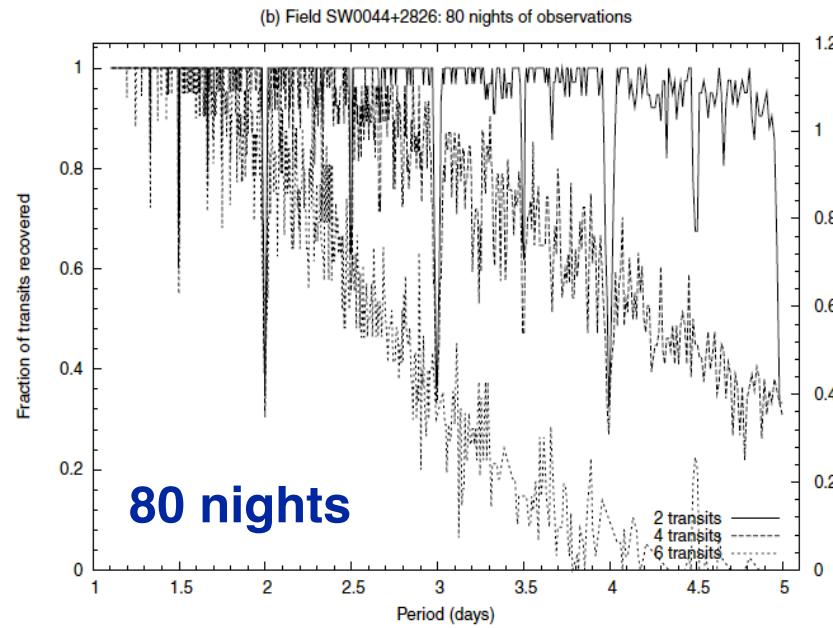
Systematics and red noise

- **Systematics:**
 - Secondary extinction
 - Temperature-dependent focus
 - Sky brightness-dependent bias in background subtraction
 - SysRem: Tamuz et al 2005
 - TFA: Kovacs et al 2005

- **Red noise:**
 - Pont et al 2006
 - Smith et al 2006



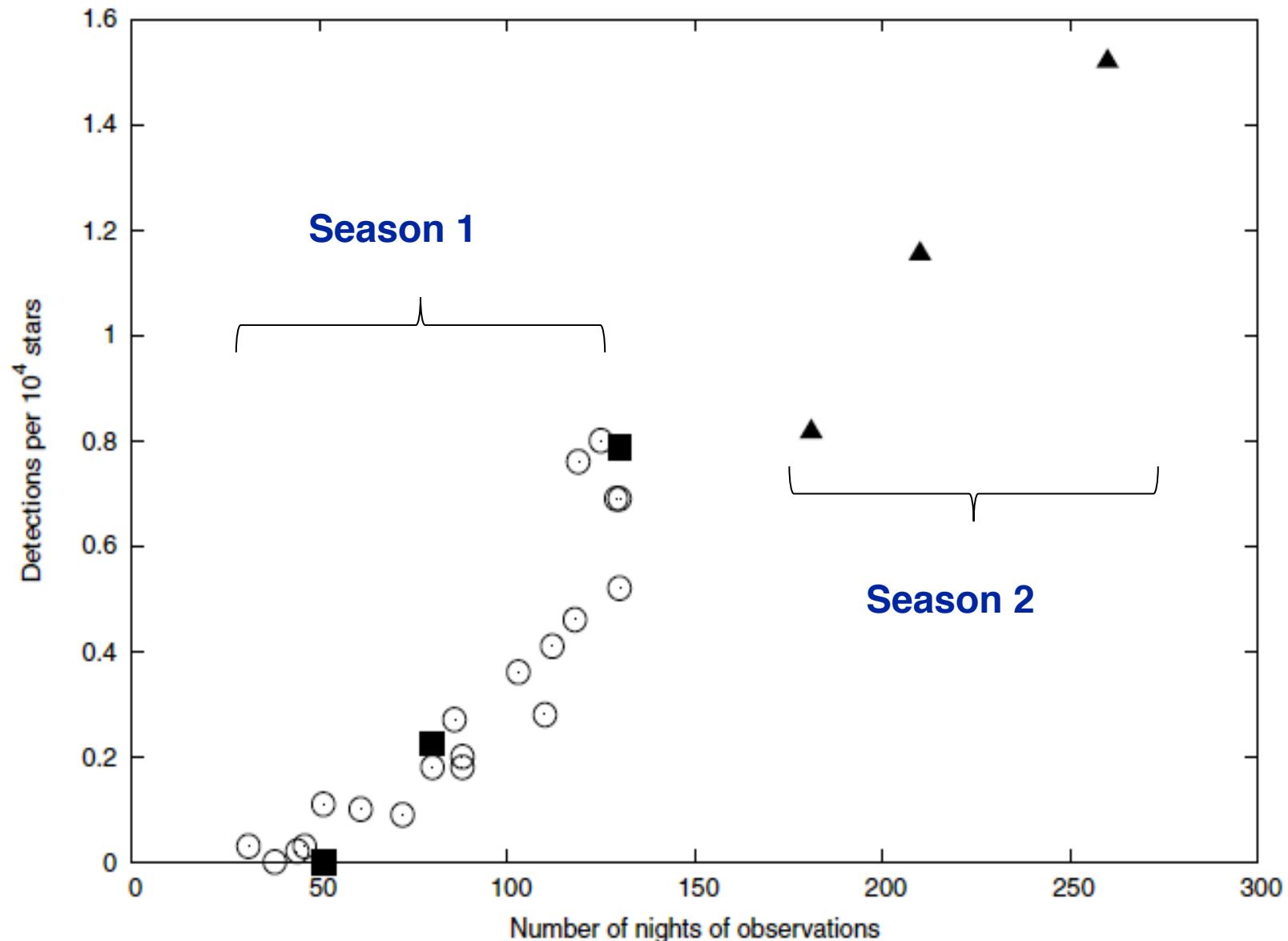
Red noise and detection threshold



Viets

ethod

Red noise and planet catch



Summary

- **Systematics and astrophysical variability mask transits**
- **Not all systematics are understood**
- **PCA-like methods (e.g. SysRem, PDC-MAP) remove systematics while preserving astrophysical variability**
- **More aggressive detrending (e.g. TFA) removes astrophysical variability too**
- **Box least-squares method is efficient**
- **Correlated noise must be taken into account when assessing detection thresholds.**