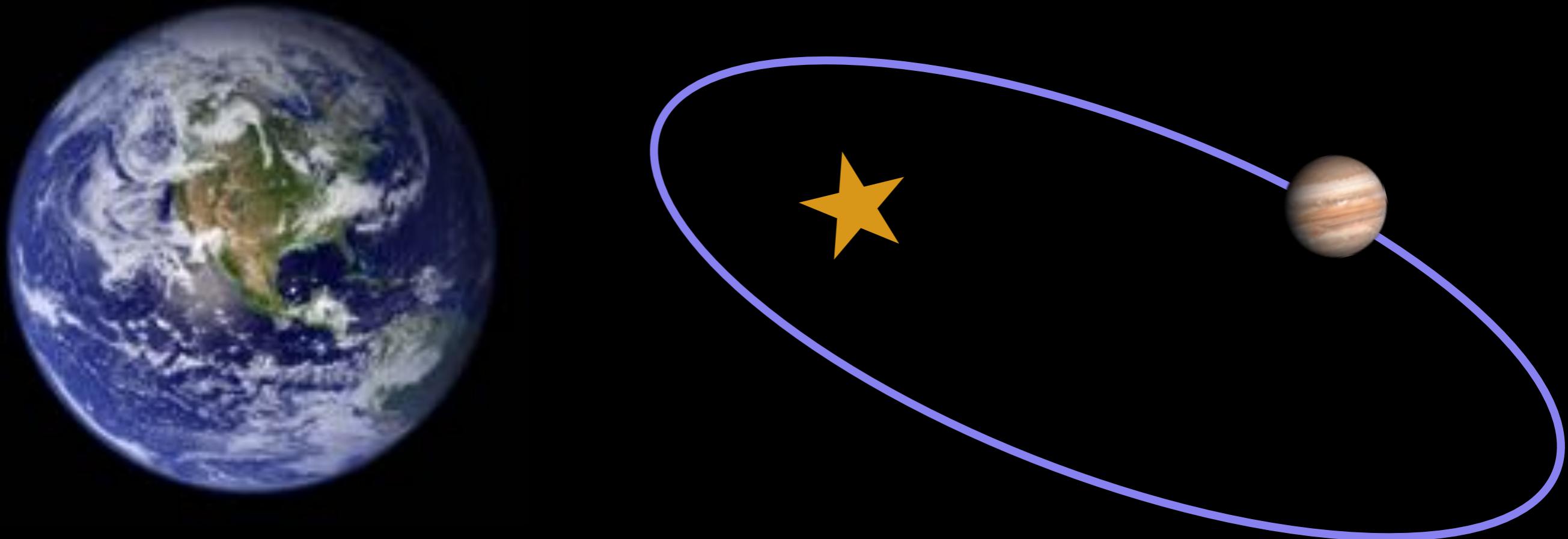


Characterizing Kepler's Transiting Planets in the Presence of Correlated Noise



Rebekah (Bekki) Dawson
Penn State, Center for Exoplanets and Habitable Worlds

Acknowledgments: SAMSI Noise and Detrending Working Group (incl. Daniel Foreman-Mackey, Eric Ford, Ben Montet, Tom Loredo, Ruth Angus, Billy Quarles, Ian Czekala, Robert Wolpert, Jogesh Babu, Tom Barclay, David Hogg); Patricio Cubillos, Josh Carter

Kepler space telescope monitored the brightness of hundreds of thousands of stars



Star dims during planetary transit

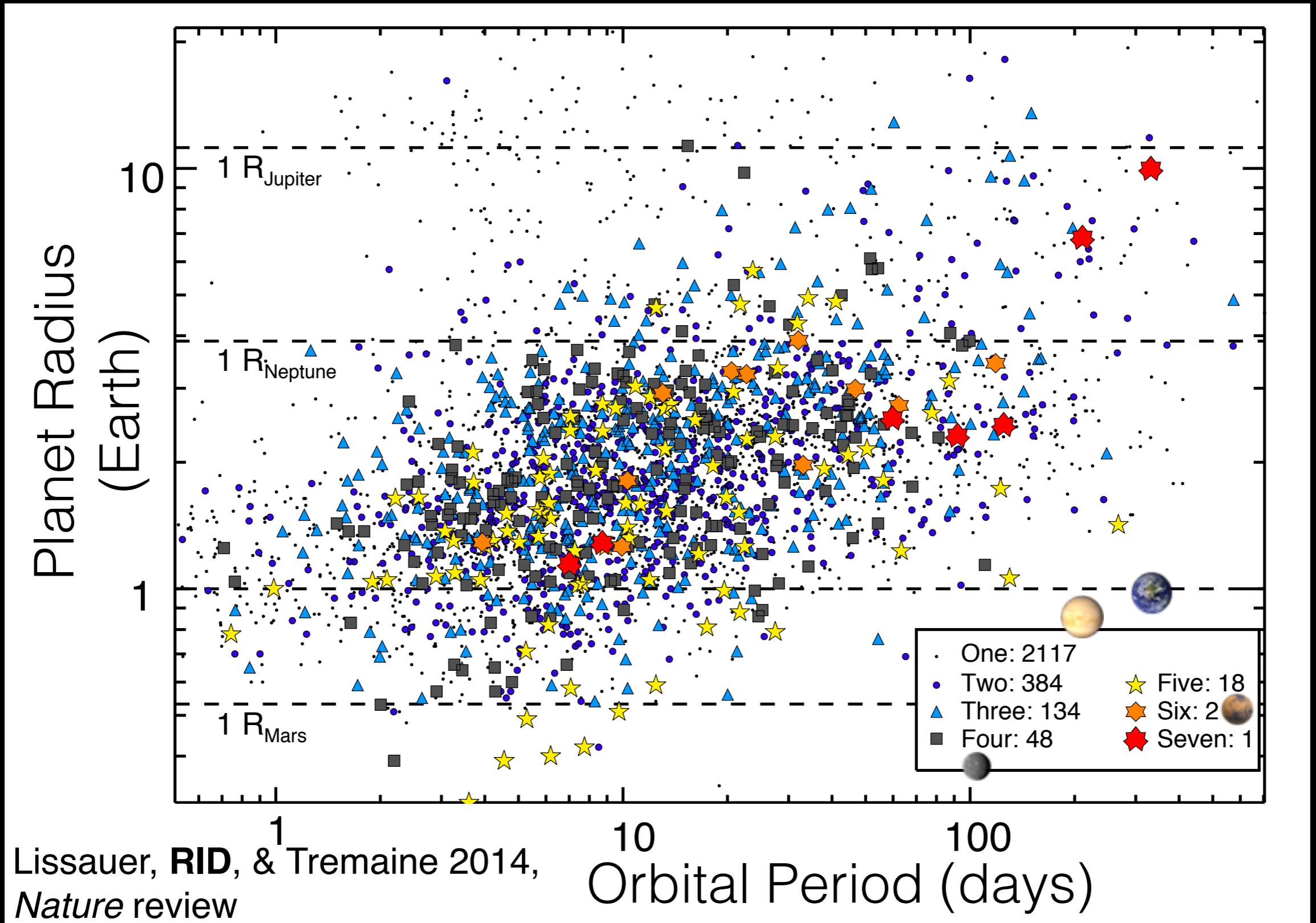


Earth-sized: 80 ppm

Image Credit: NASA

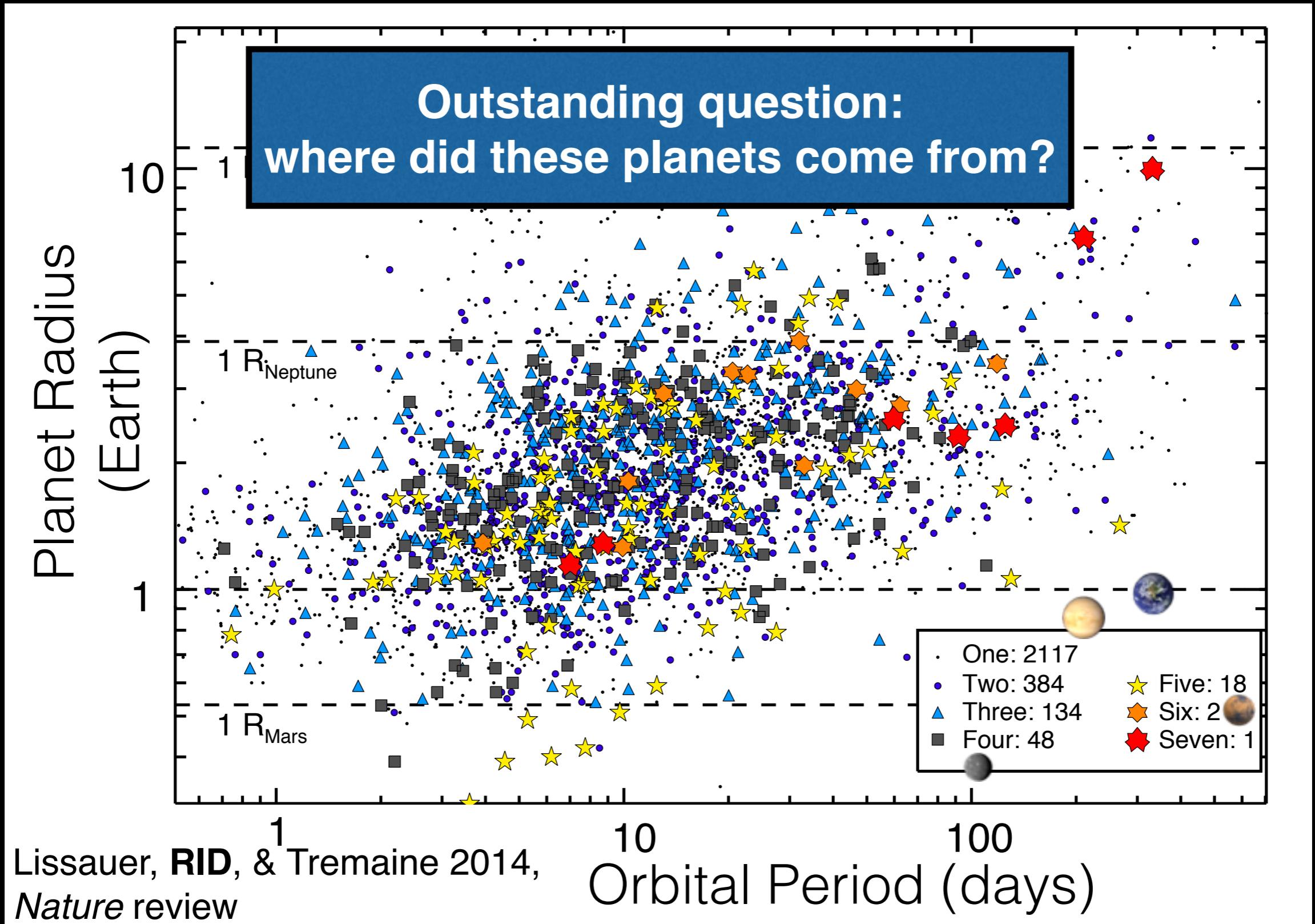
Kepler discovered thousands of close-in planets

Kepler candidate discoveries: Borucki+11ab, Batalha+ 12, Burke+ 14, Mullally+15



Kepler discovered thousands of planets

Kepler candidate discoveries: Borucki+11ab, Batalha+ 12, Burke+ 14, Mullally+15



Gravitational interactions between planets cause transit timing variations

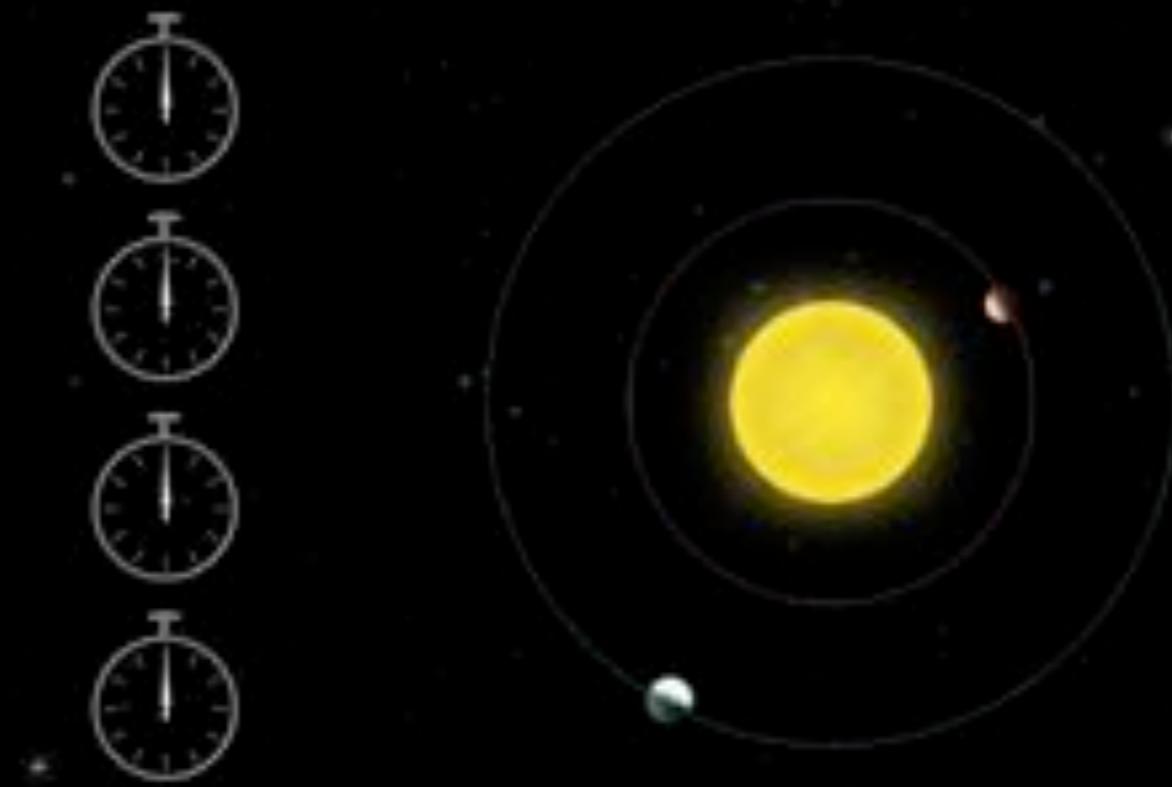
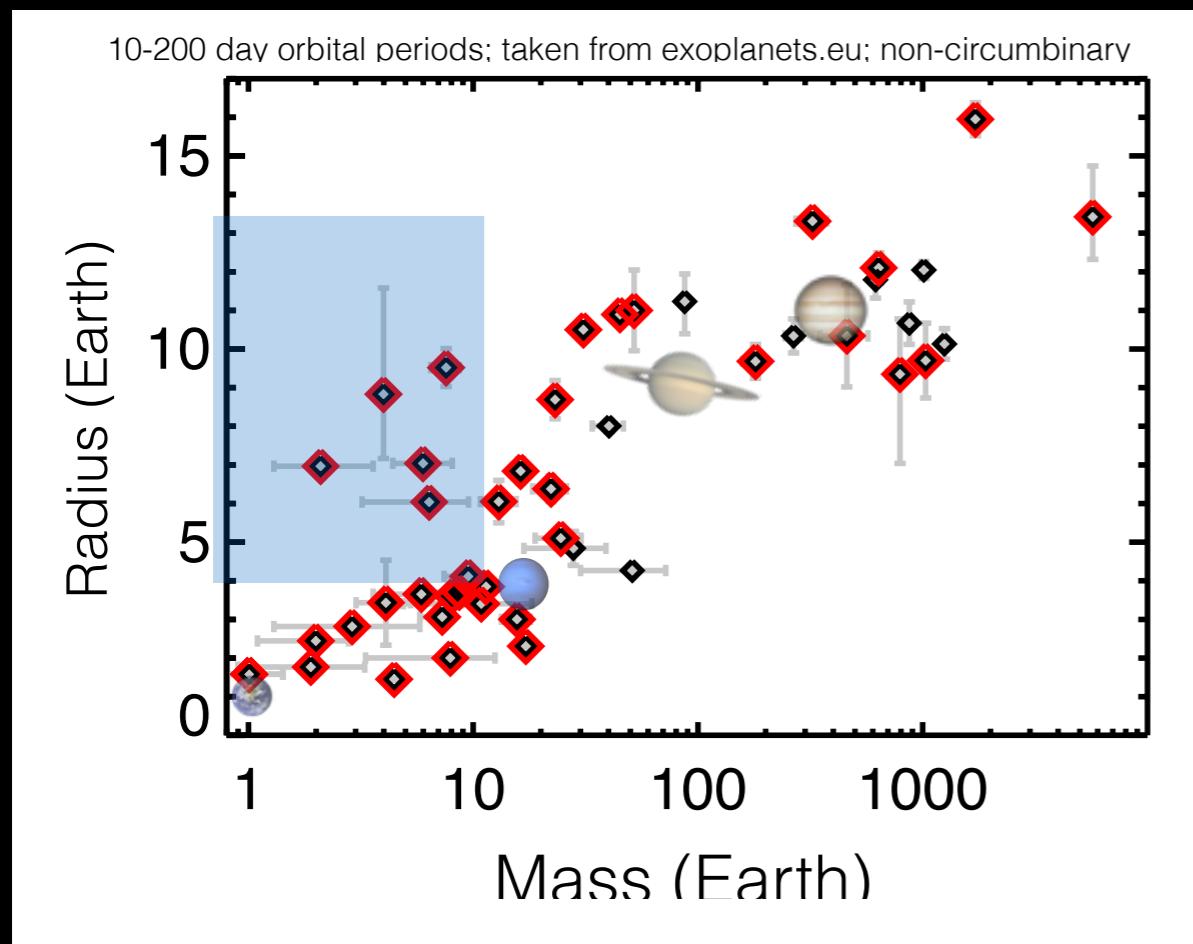


Image Credit: NASA Ames Research Center/Kepler Mission

Transit timing variations essential to understanding where close-in planets come from

e.g.,

Ultra-puffies

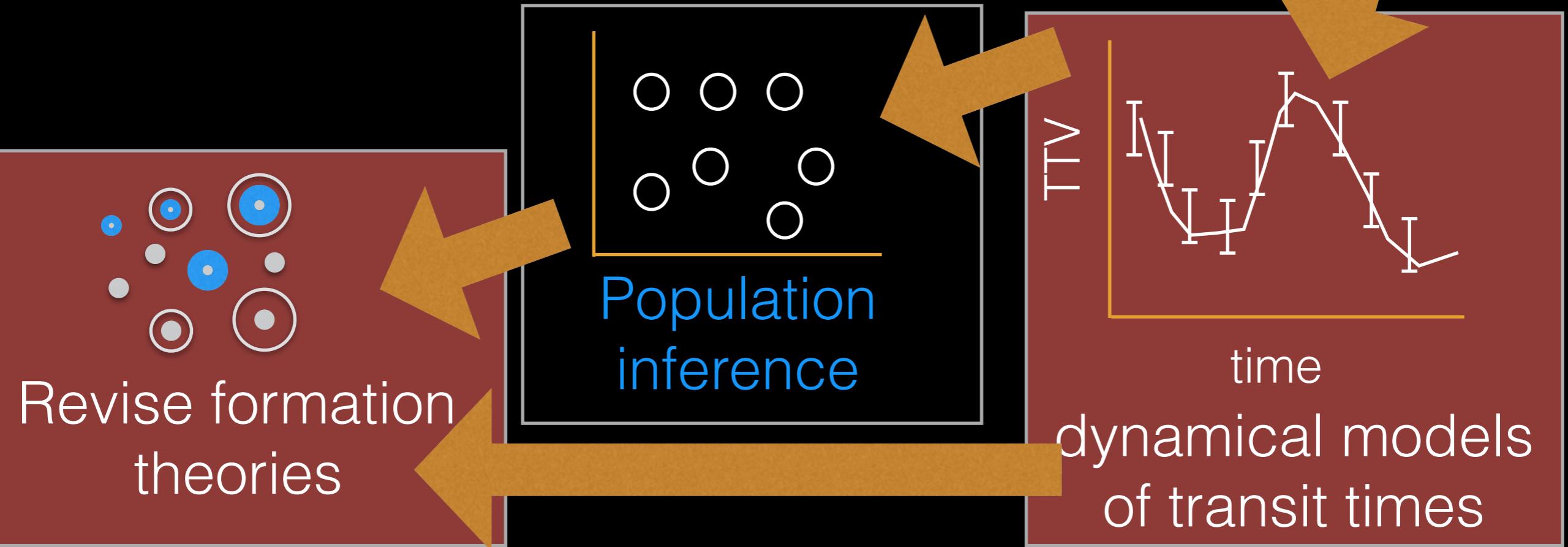
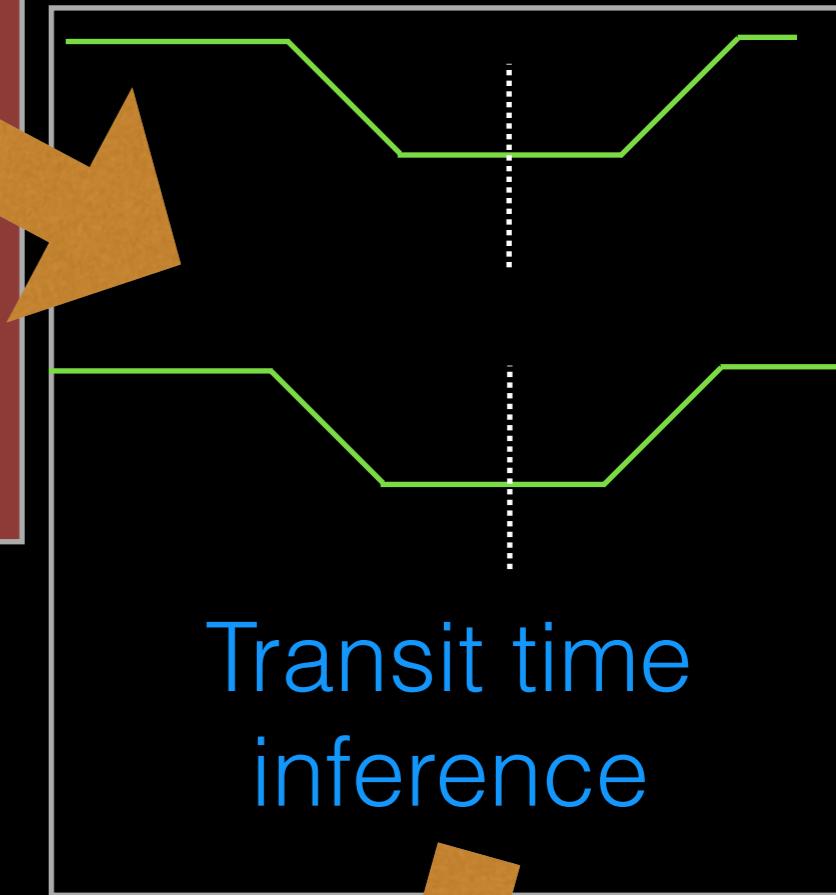
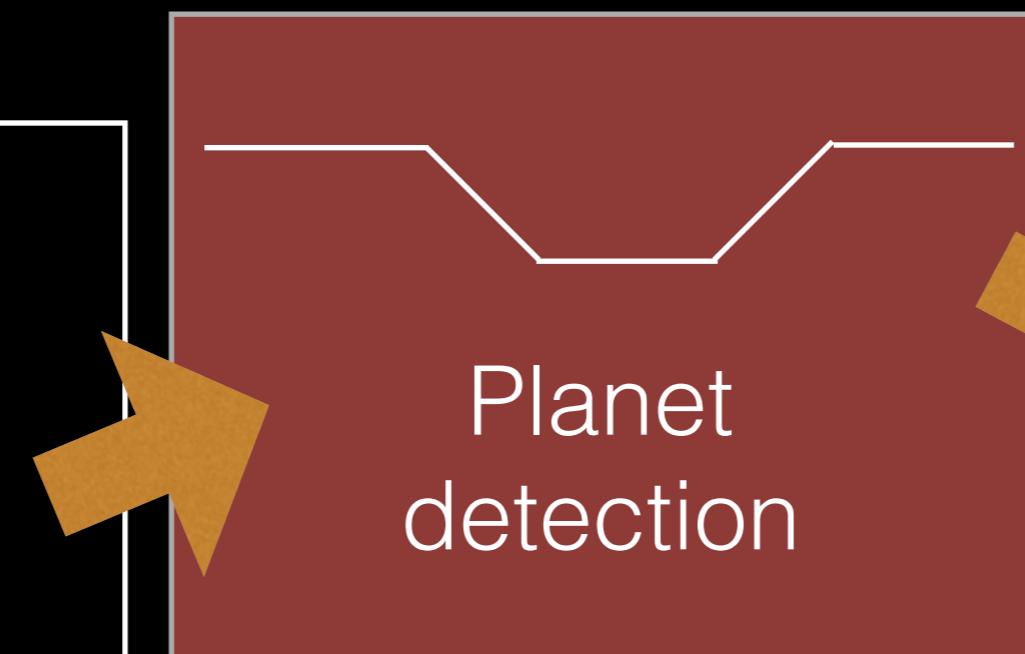
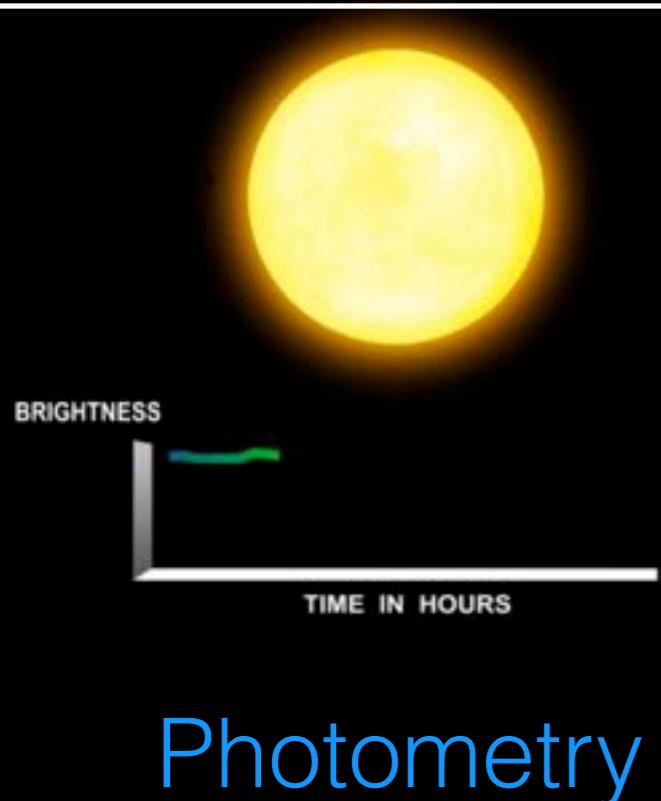


Resonant orbits
[with low libration amplitudes]

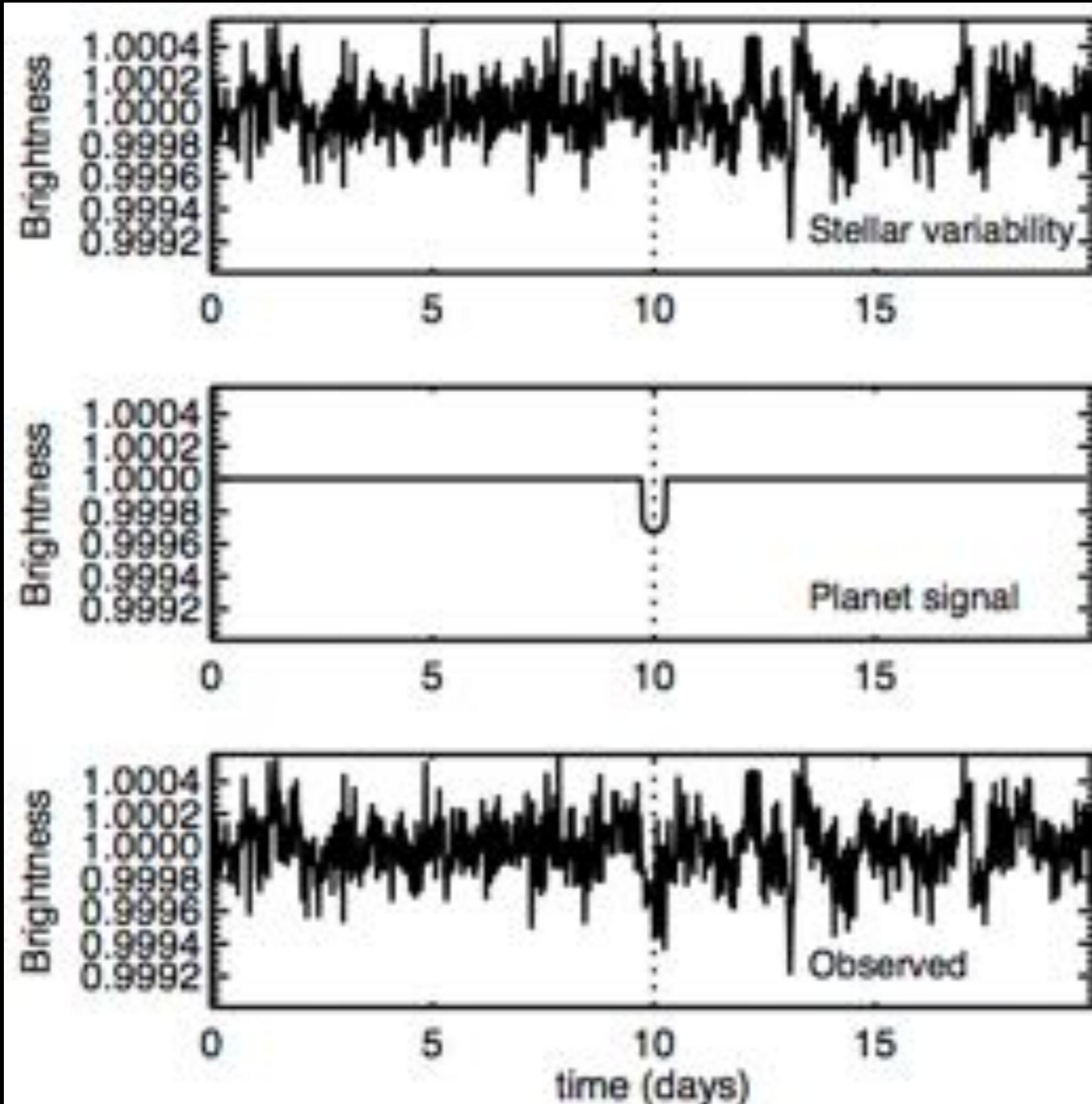


e.g. Jontof-Hutter et al. 2014,
formation: Lee & Chiang 2016

Kepler-223, Mills et al. 2016
Nature



Kepler data key characteristics



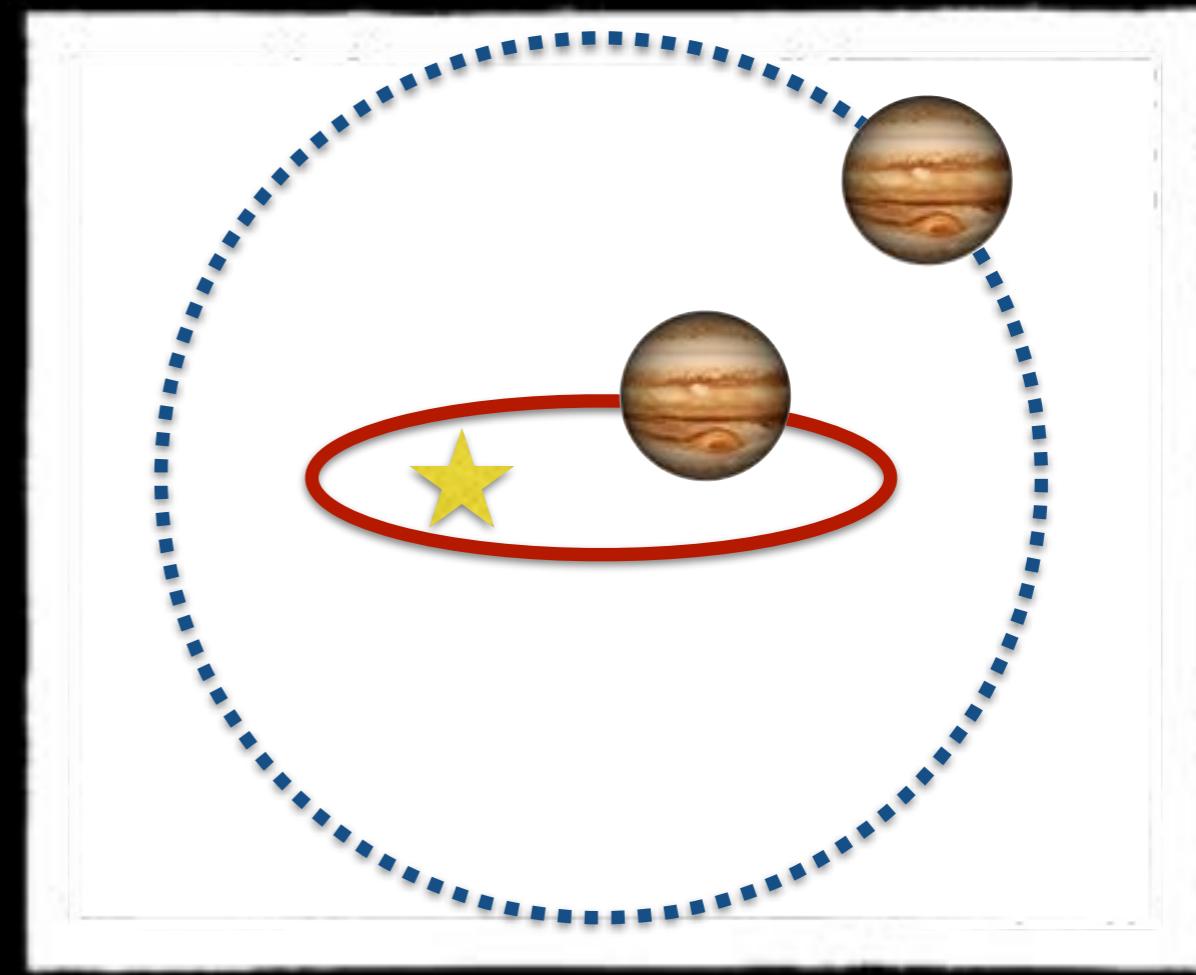
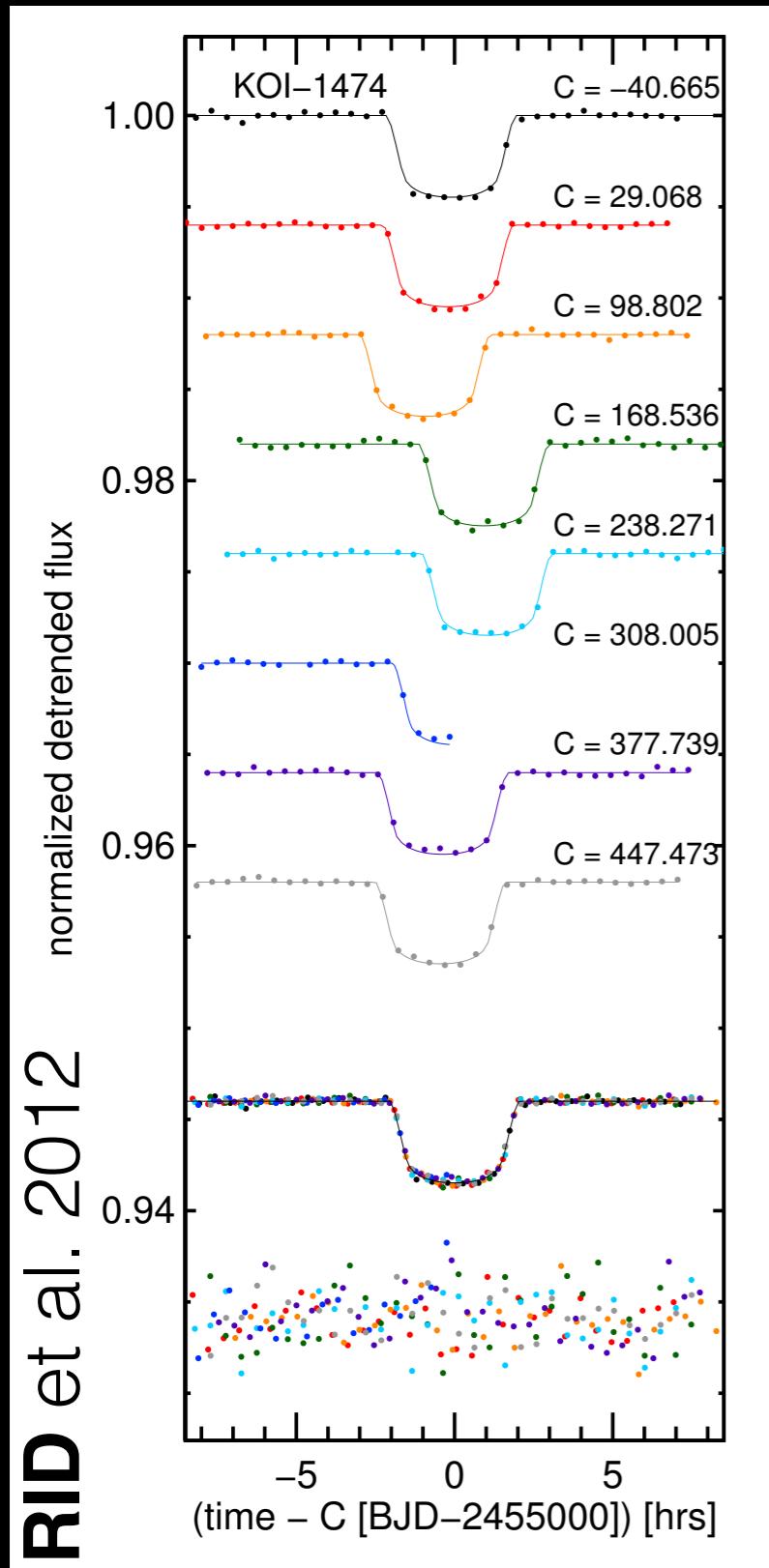
Time span: 4 years

Cadence: 30 minutes (1 minute), evenly spaced

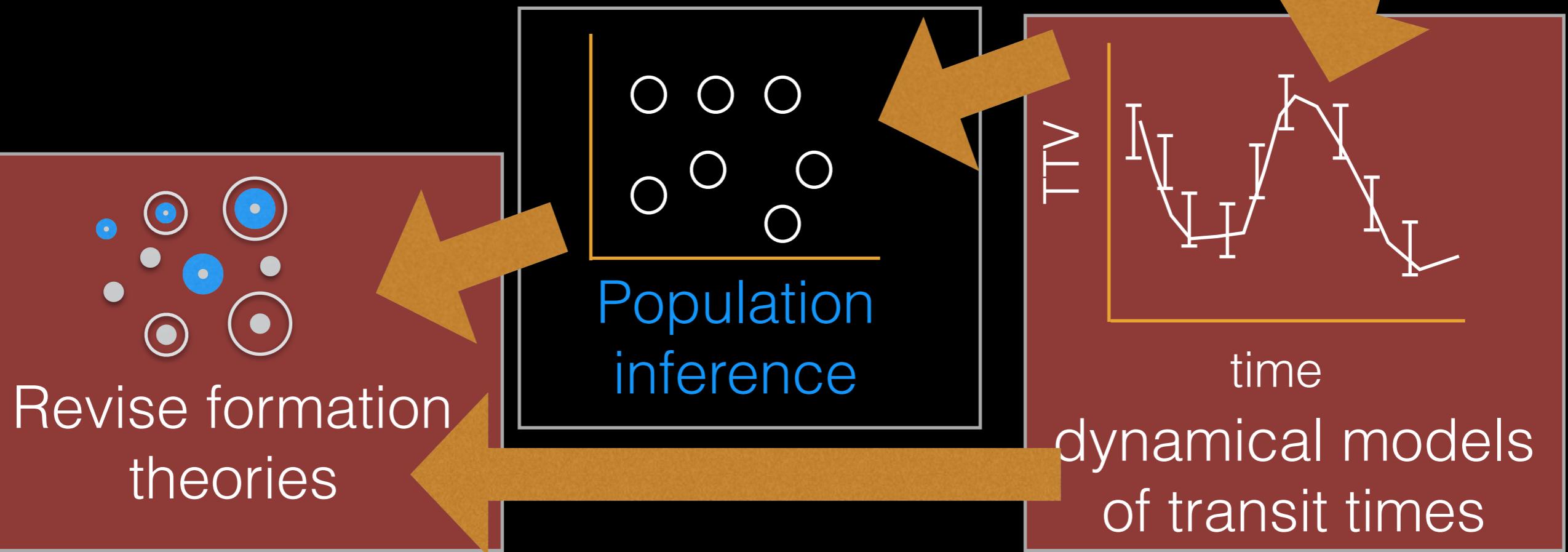
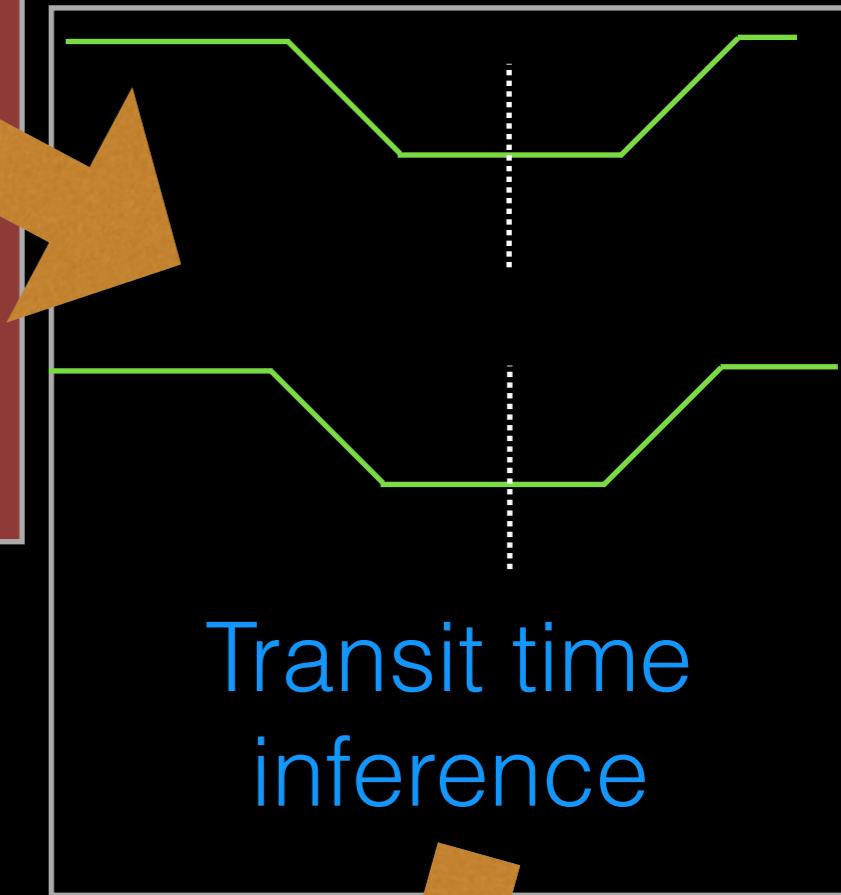
Noise floor: 15 ppm

Earth-like transit signal: 80 ppm, 12 hour duration, 1 year period

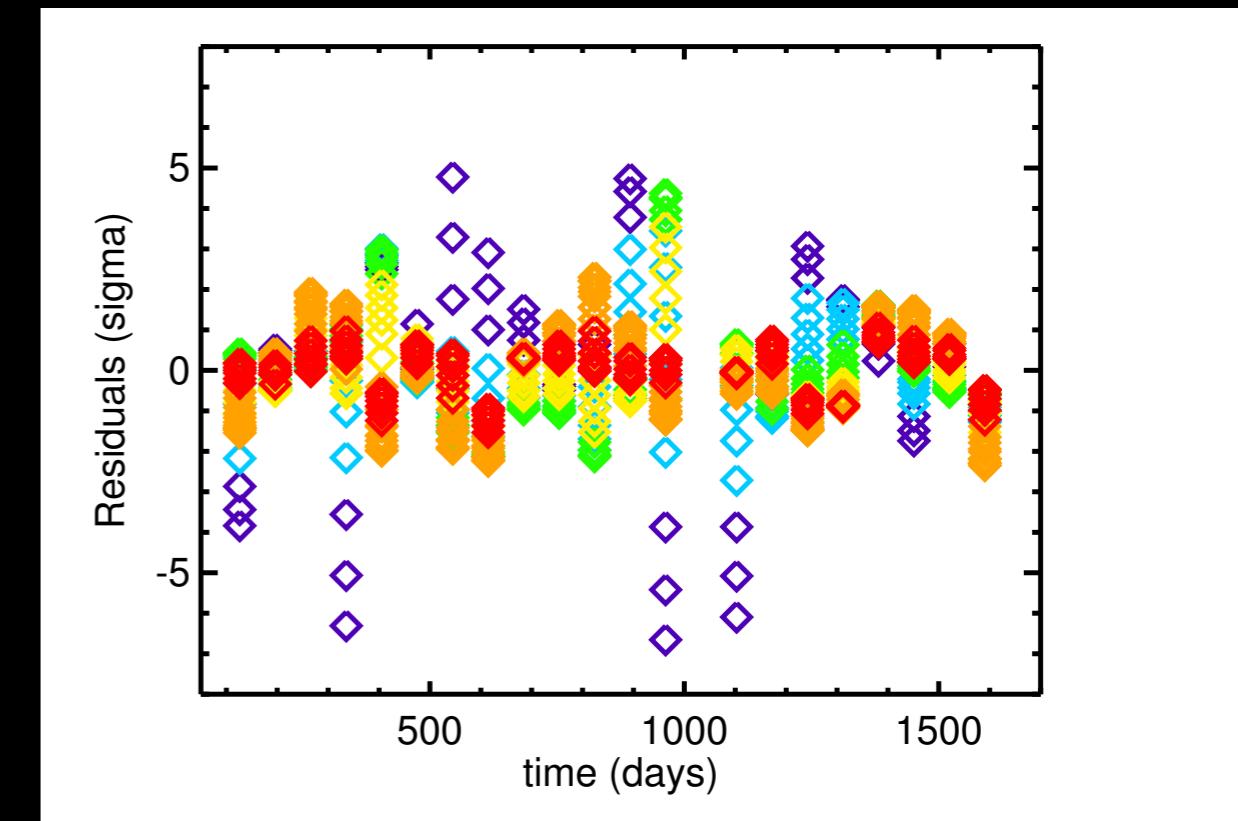
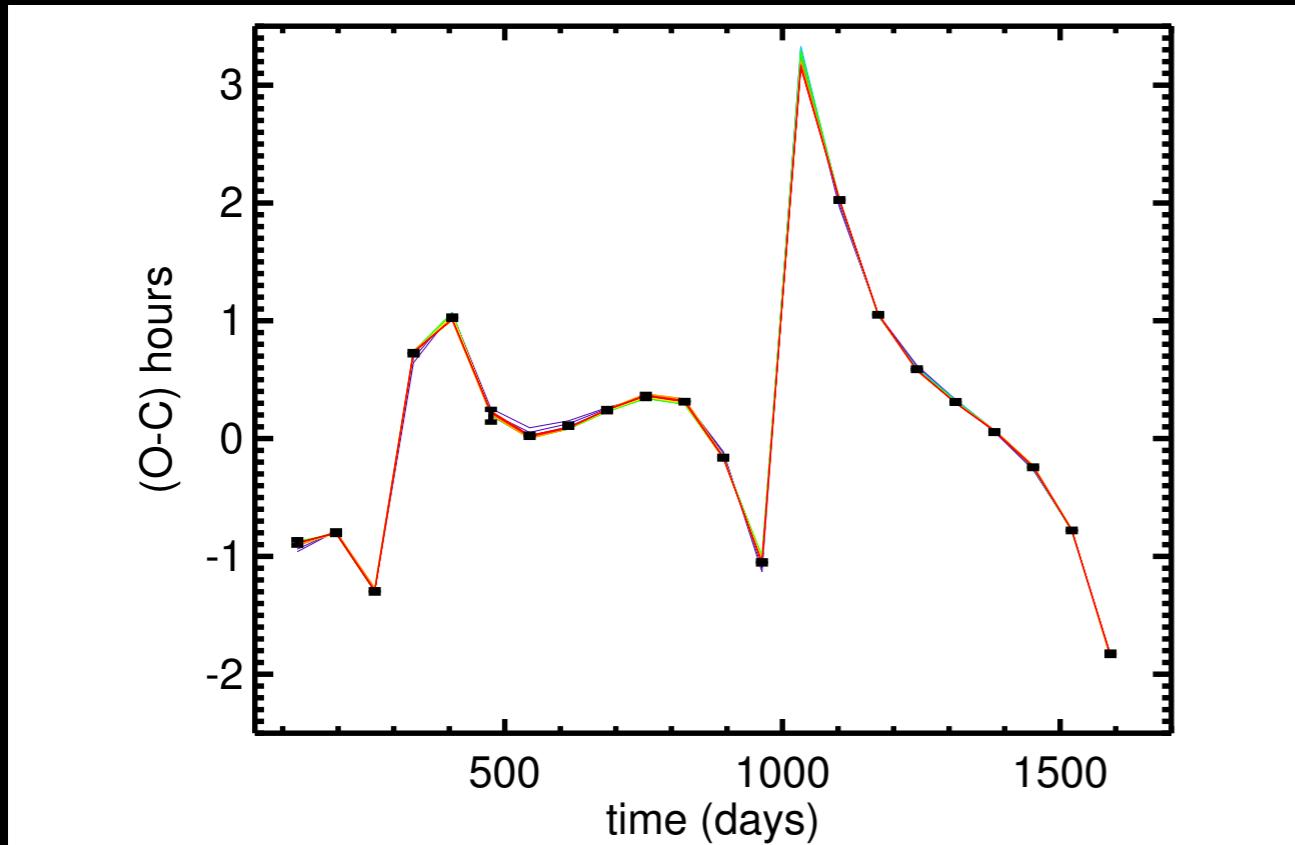
Case study: how did Kepler-419b achieve its close-in, highly elliptical orbit?



Is its non-transiting companion orbiting in the same plane?



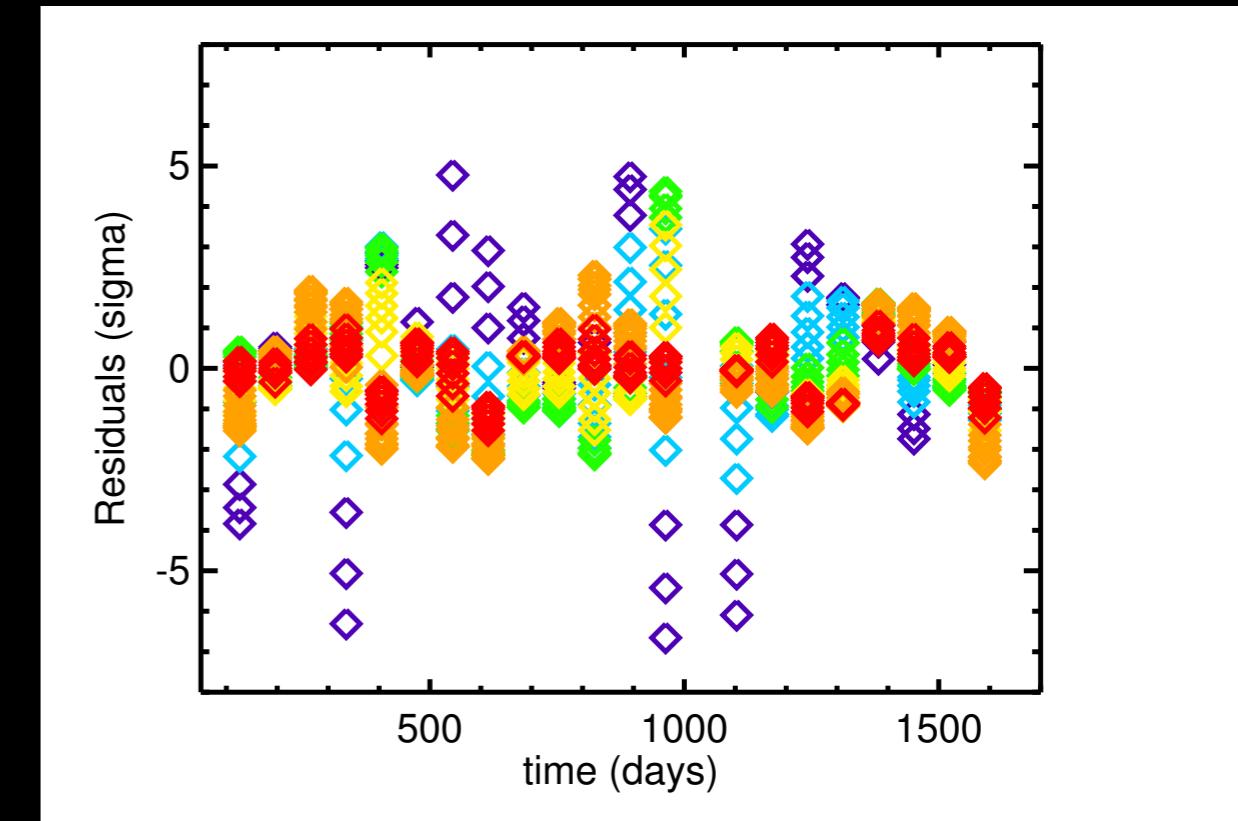
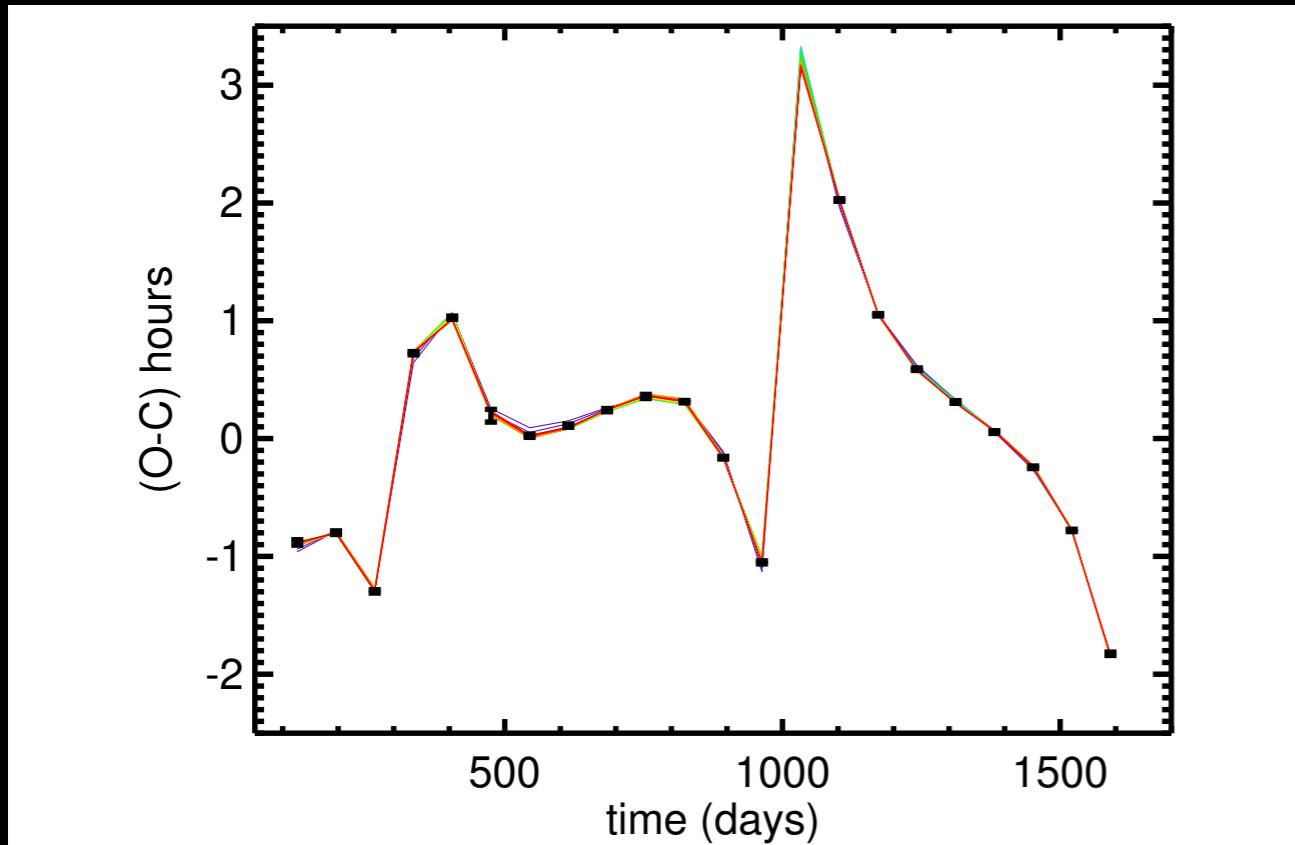
The companion's inclination has a subtle effect on the signal



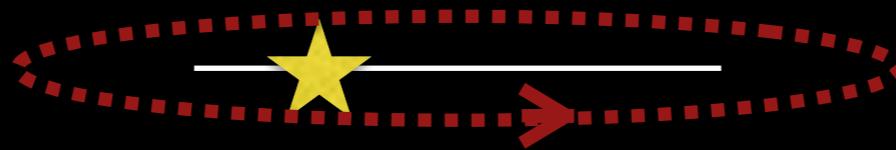
RID et al. 2014



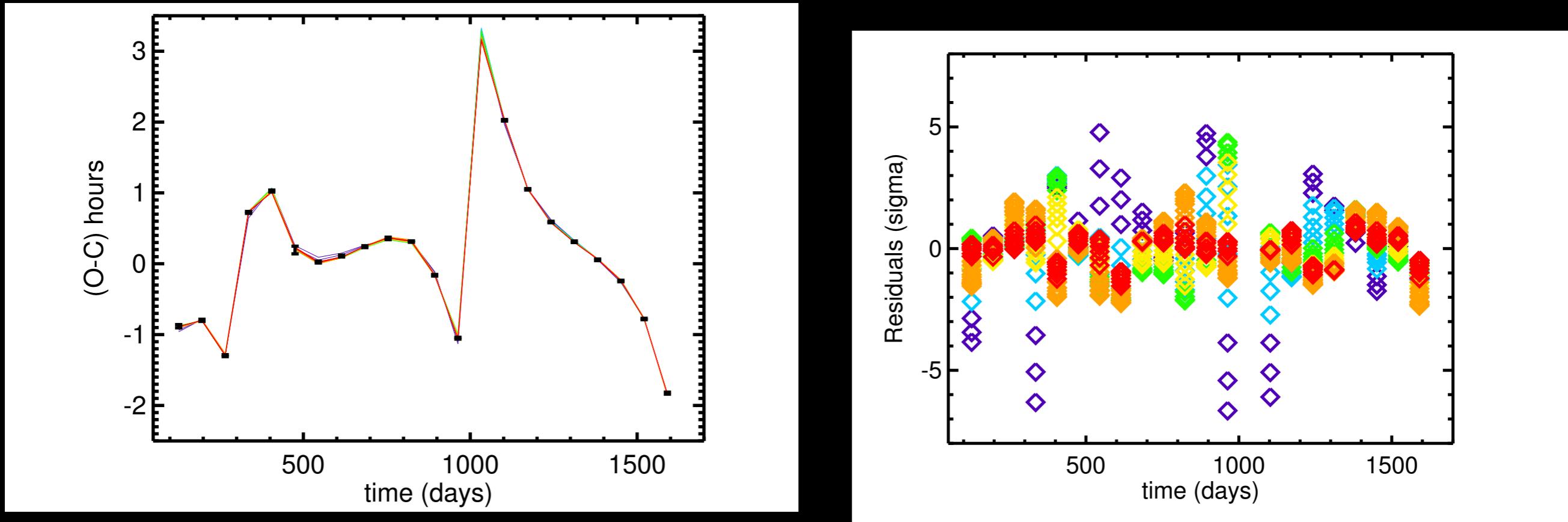
The companion's inclination has a subtle effect on the signal



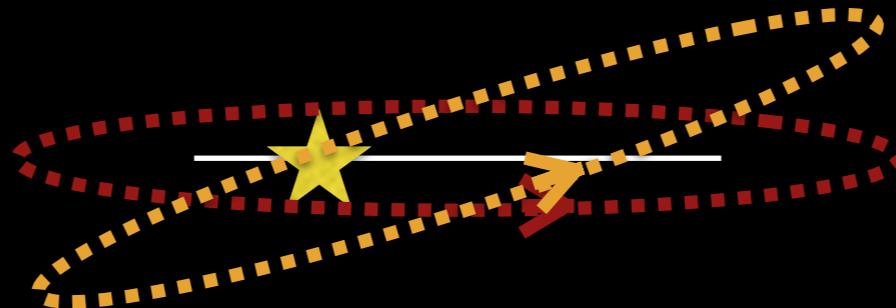
RID et al. 2014



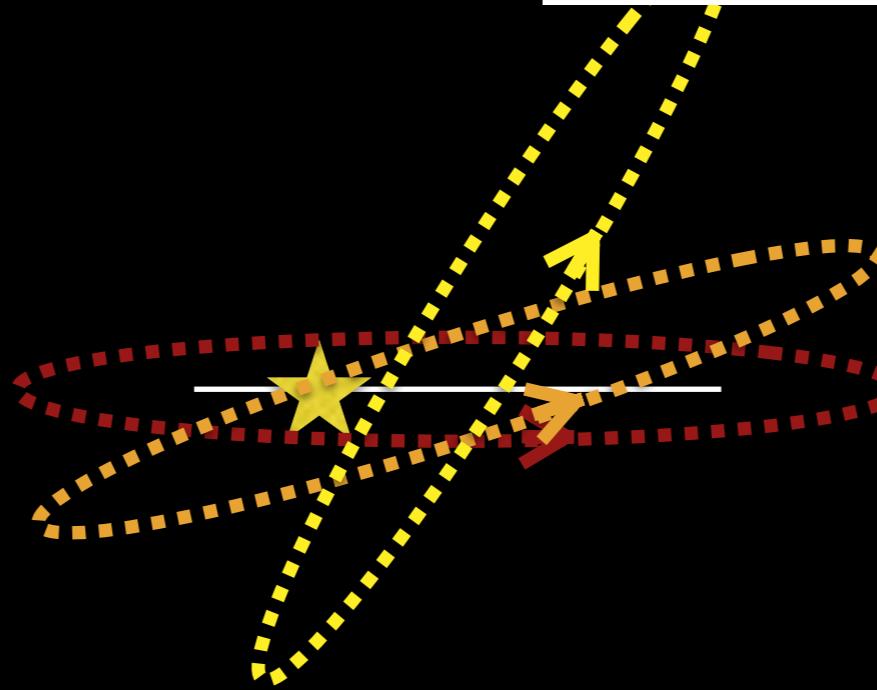
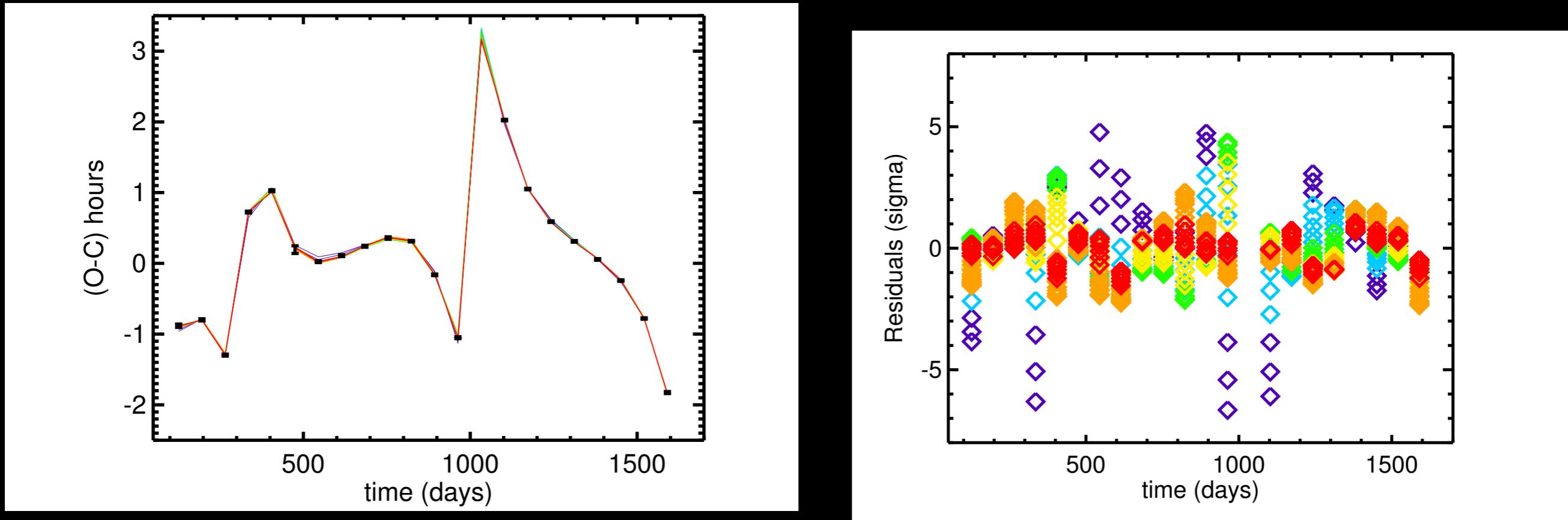
The companion's inclination has a subtle effect on the signal



RID et al. 2014

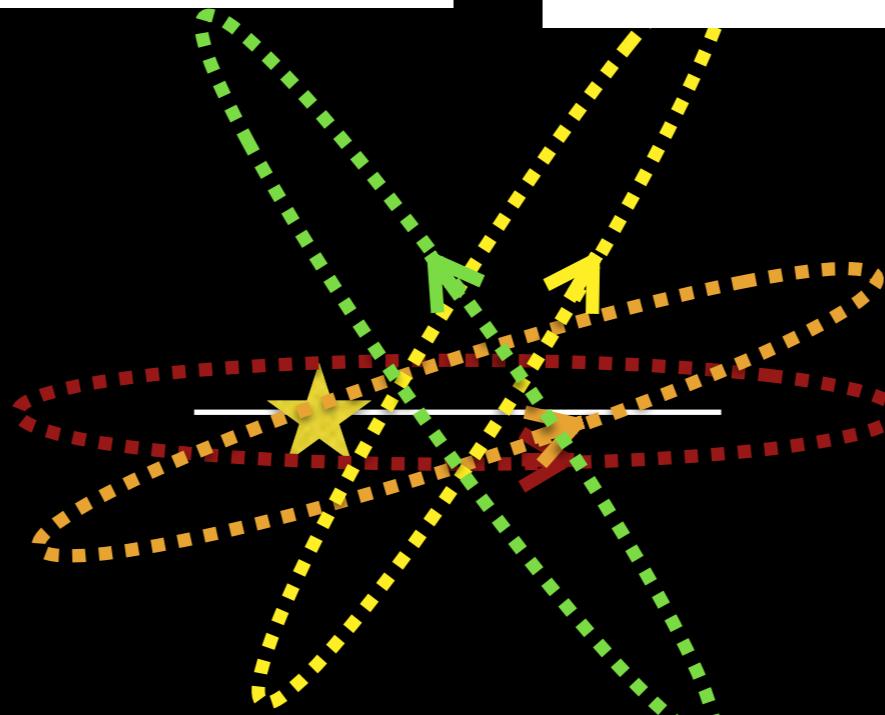
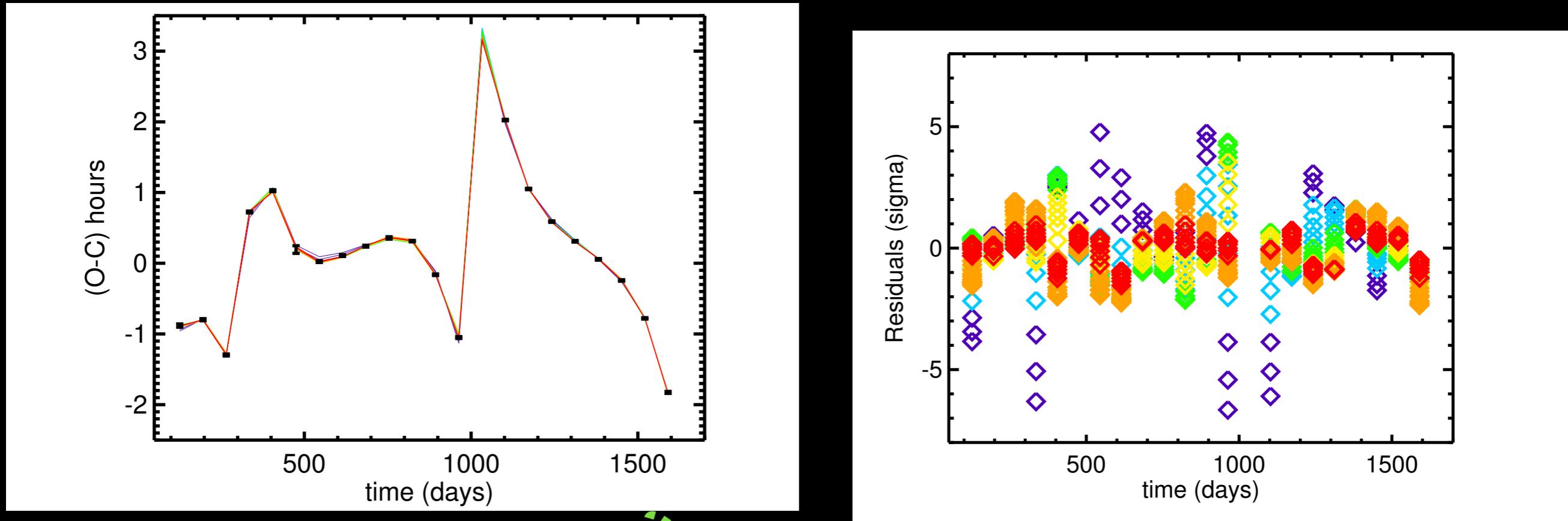


The companion's inclination has a subtle effect on the signal



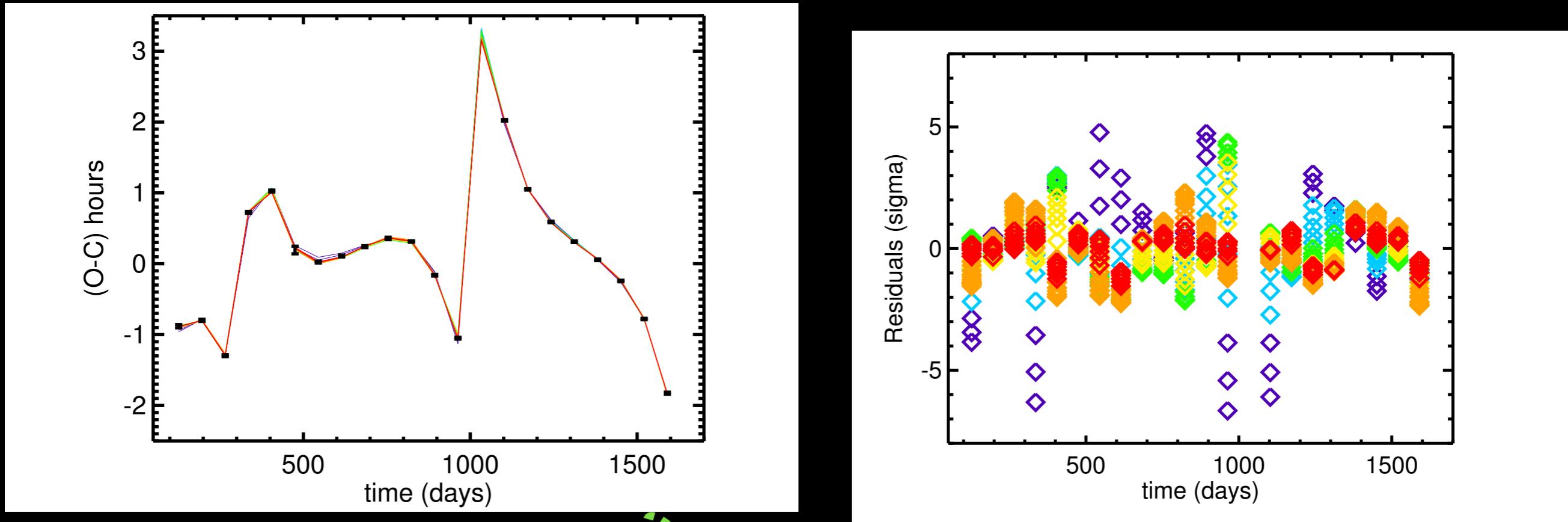
RID et al. 2014

The companion's inclination has a subtle effect on the signal

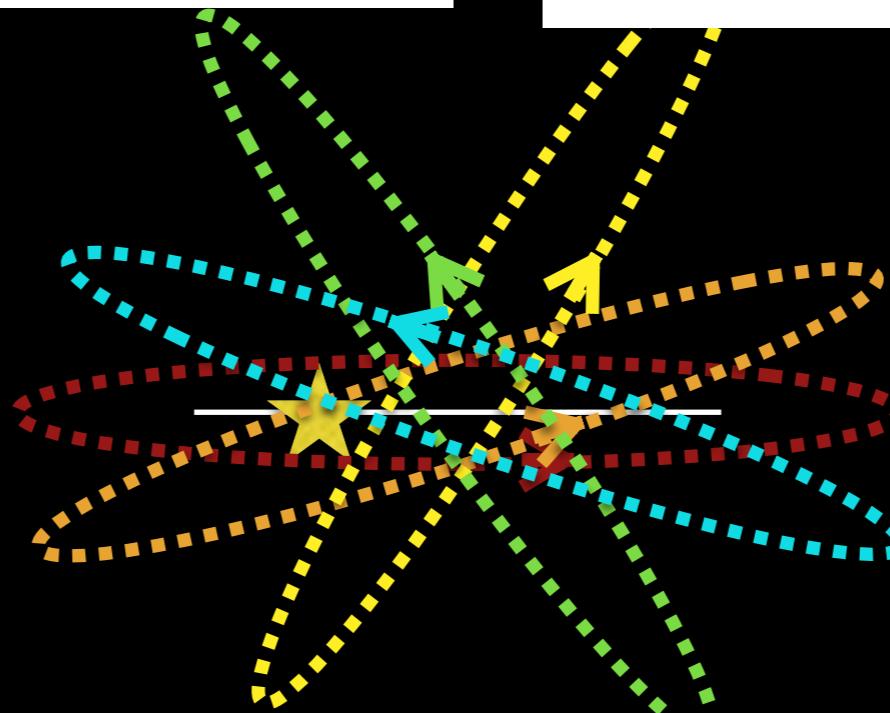


RID et al. 2014

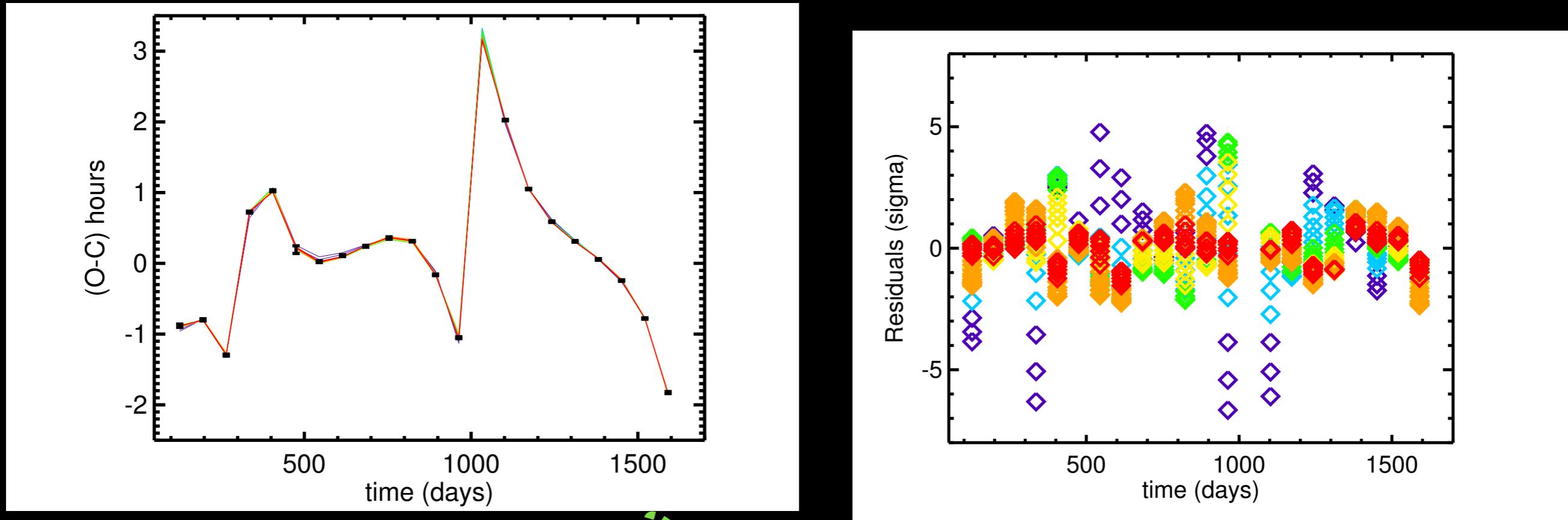
The companion's inclination has a subtle effect on the signal



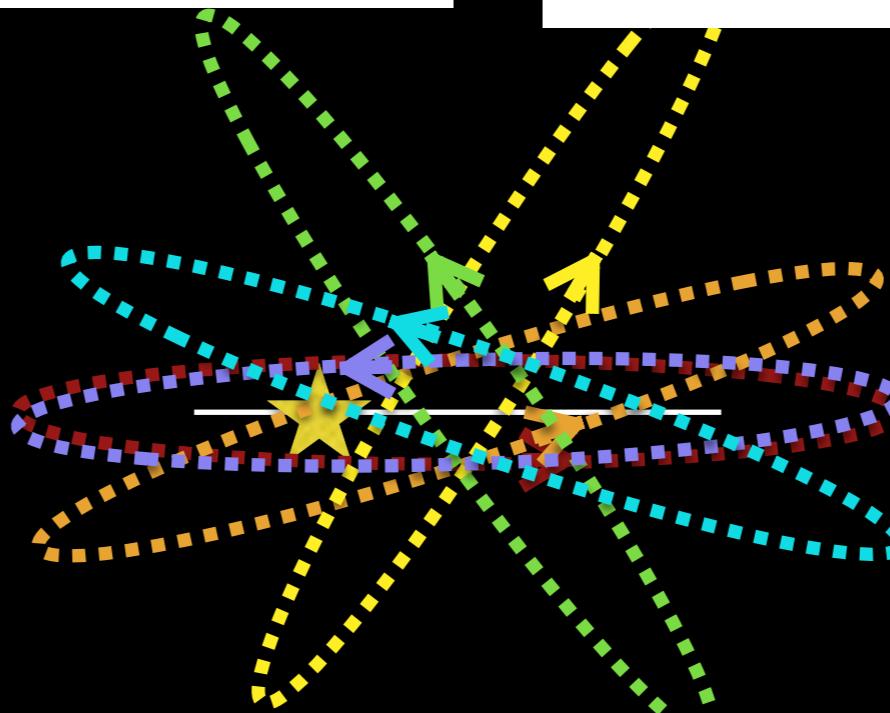
RID et al. 2014

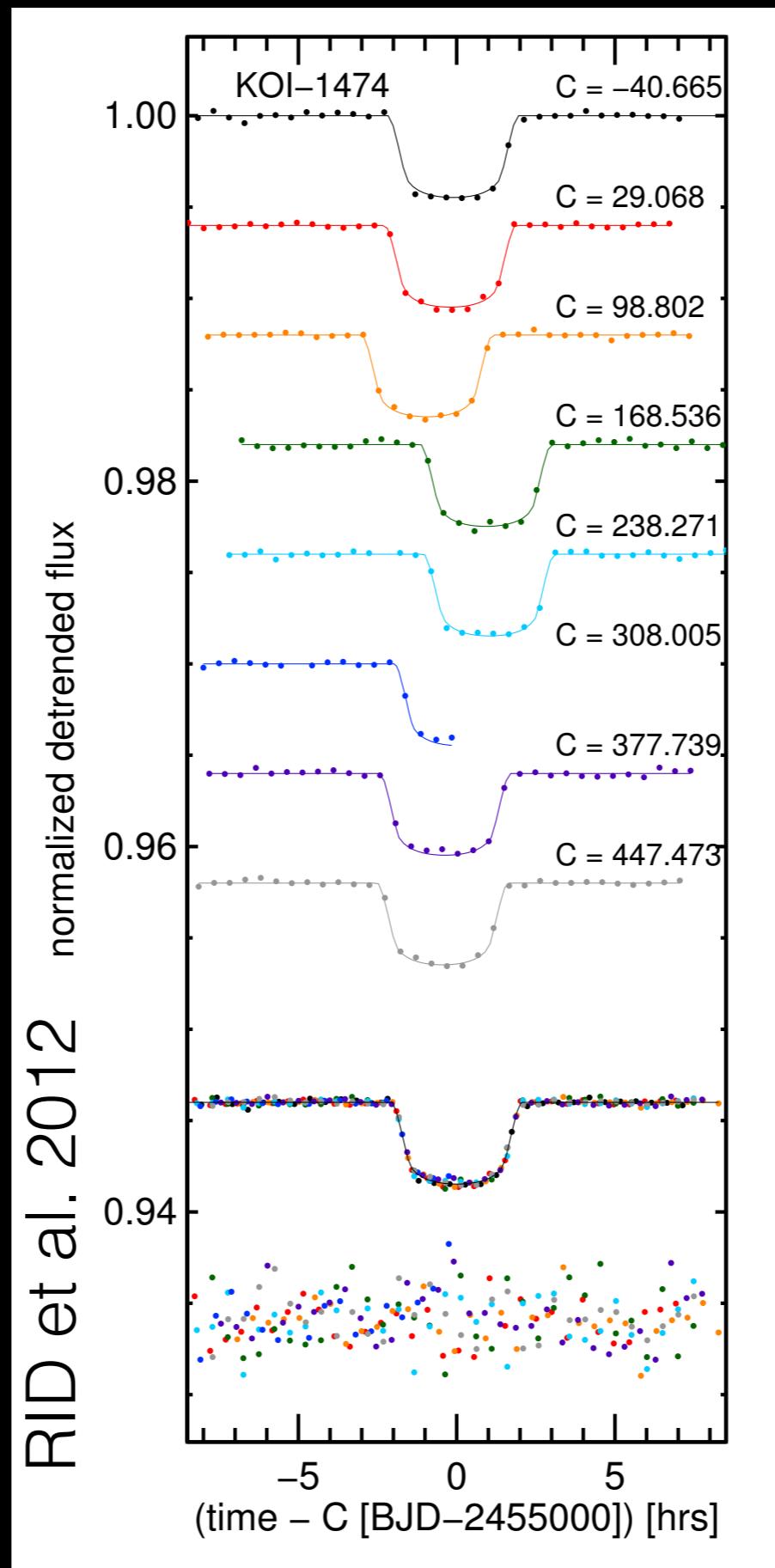


The companion's inclination has a subtle effect on the signal

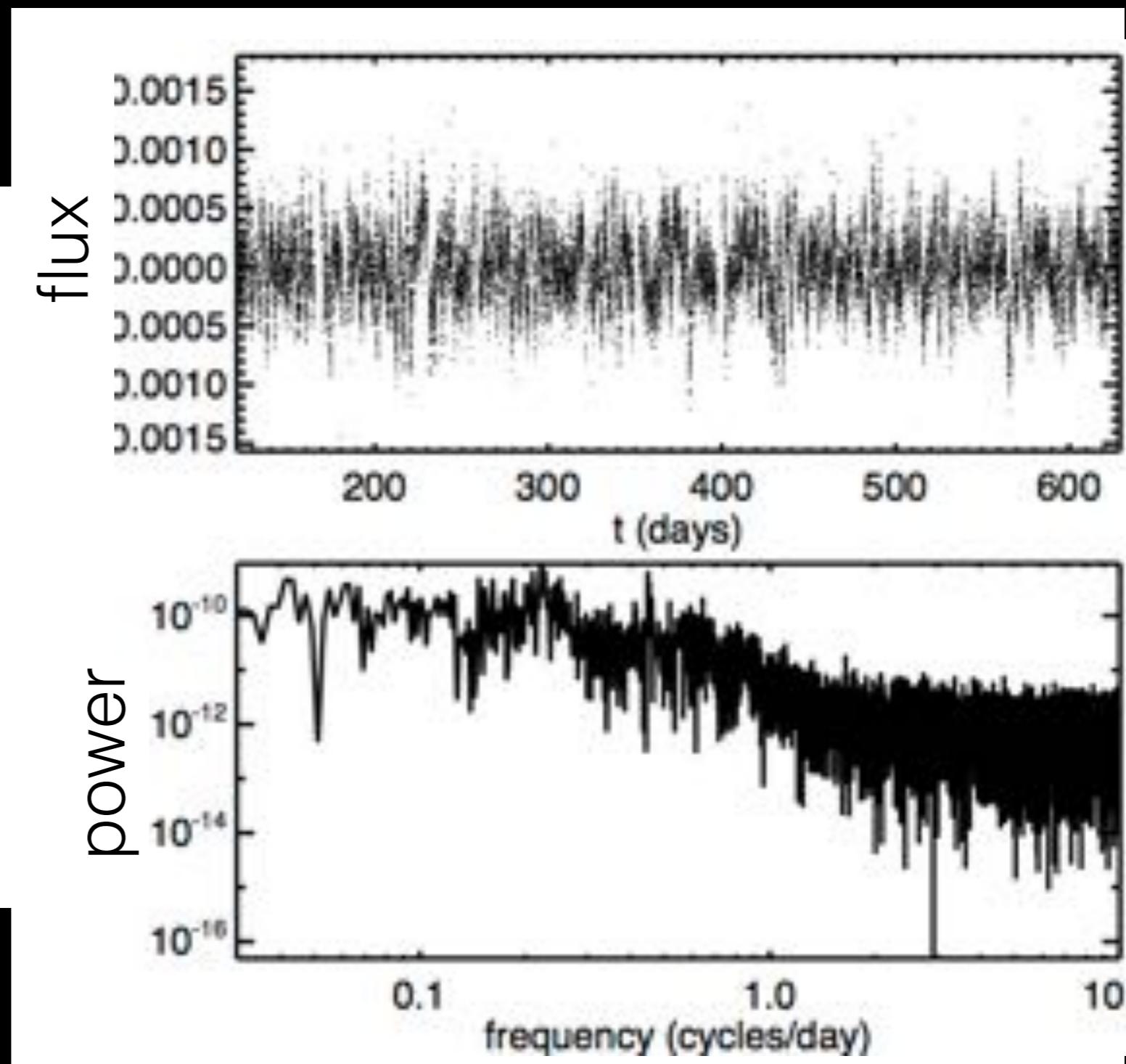
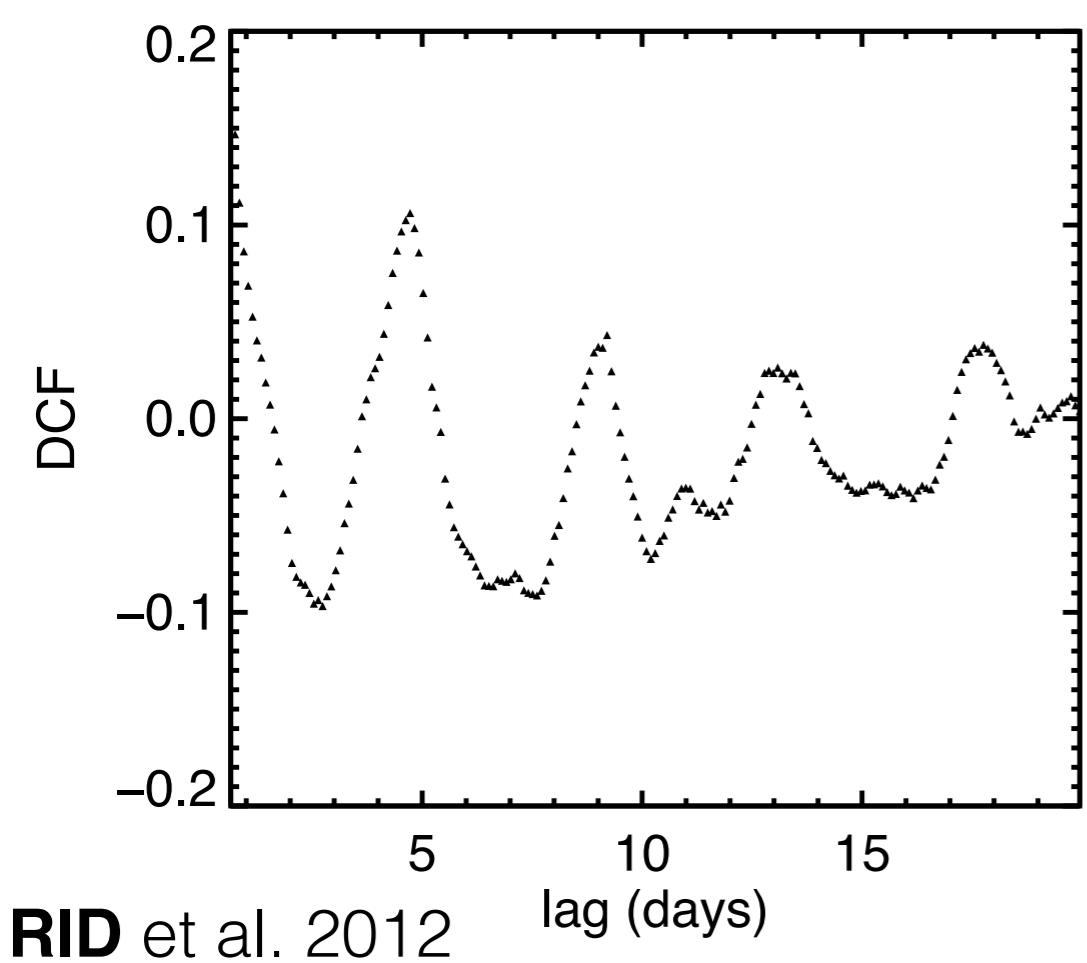


RID et al. 2014

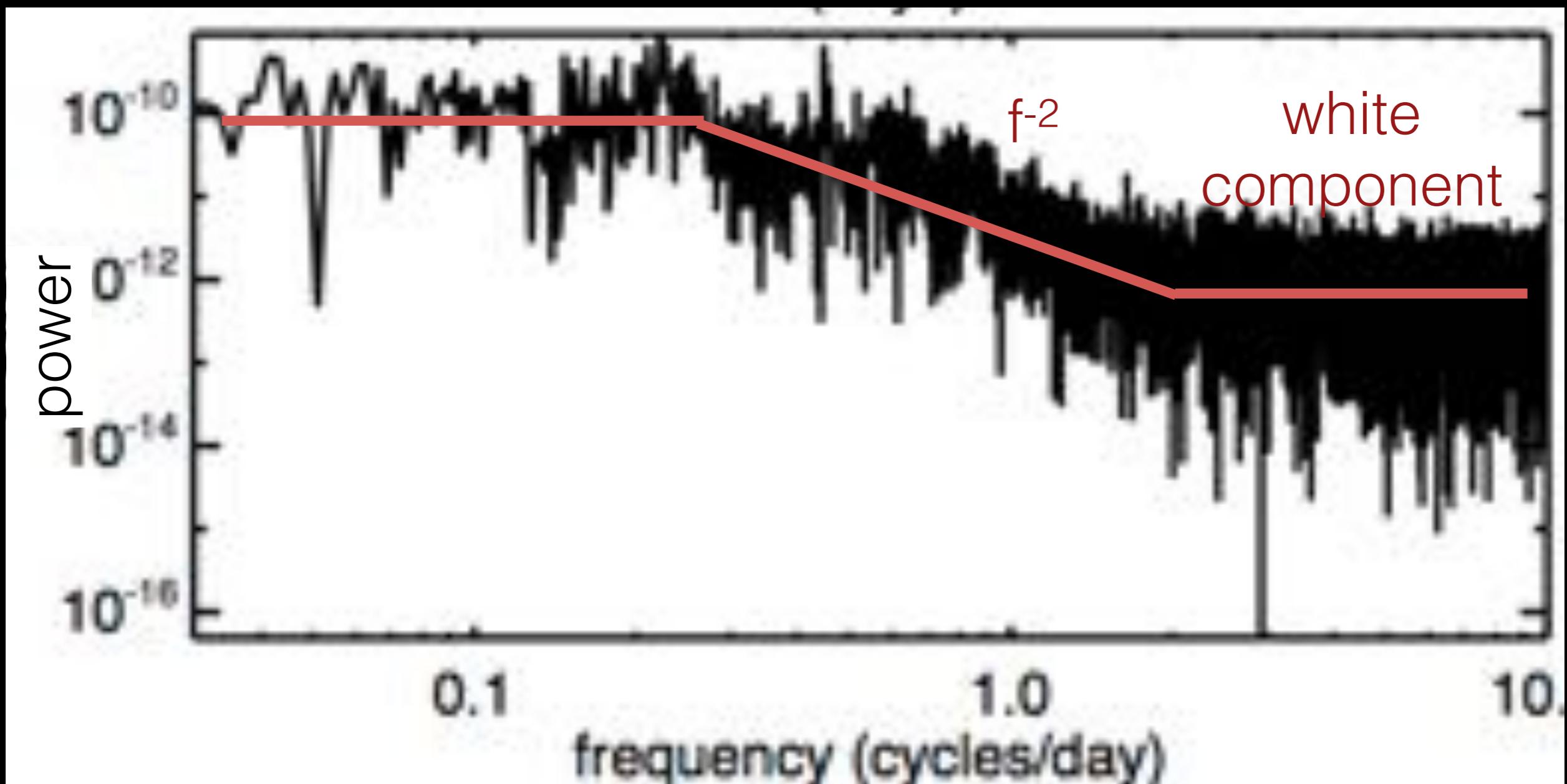




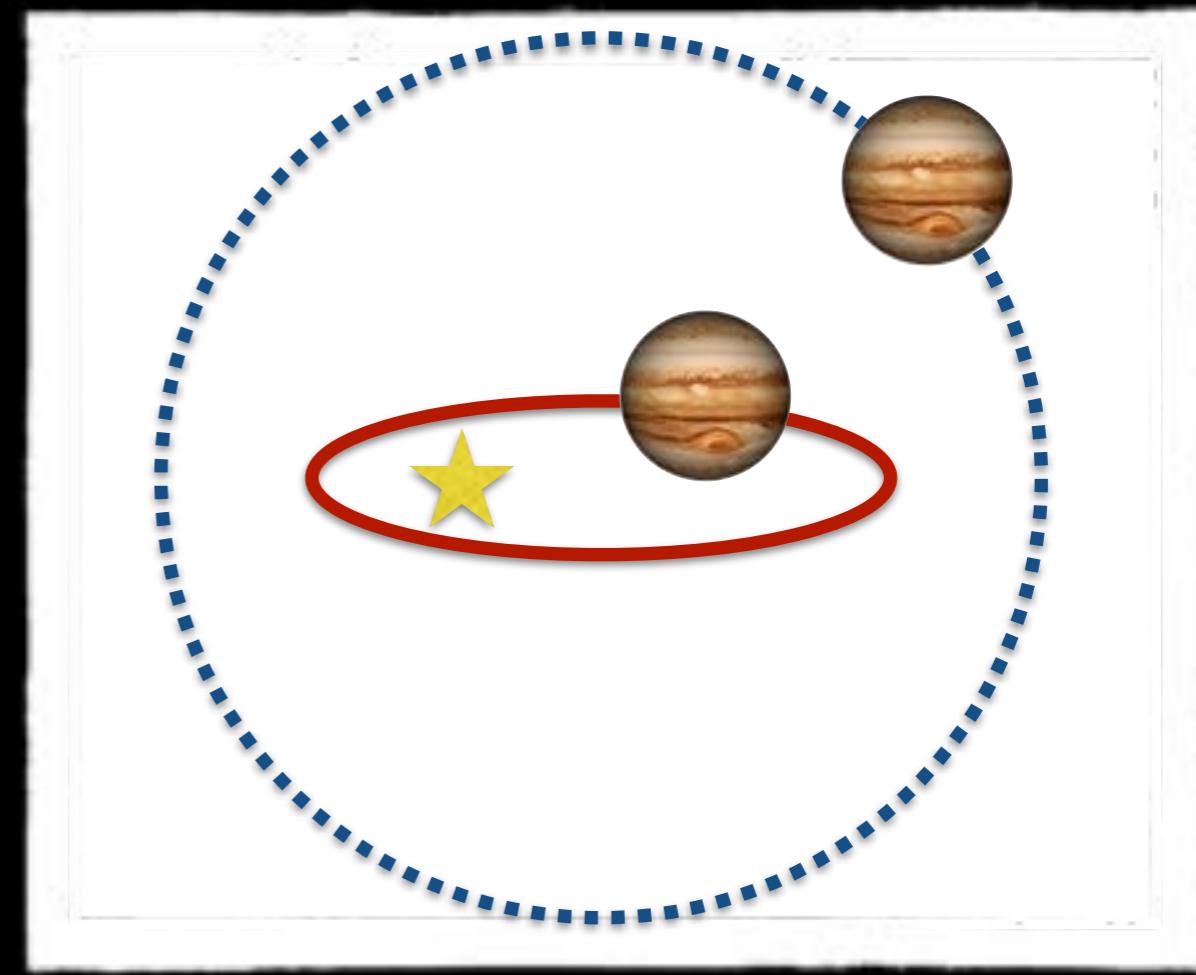
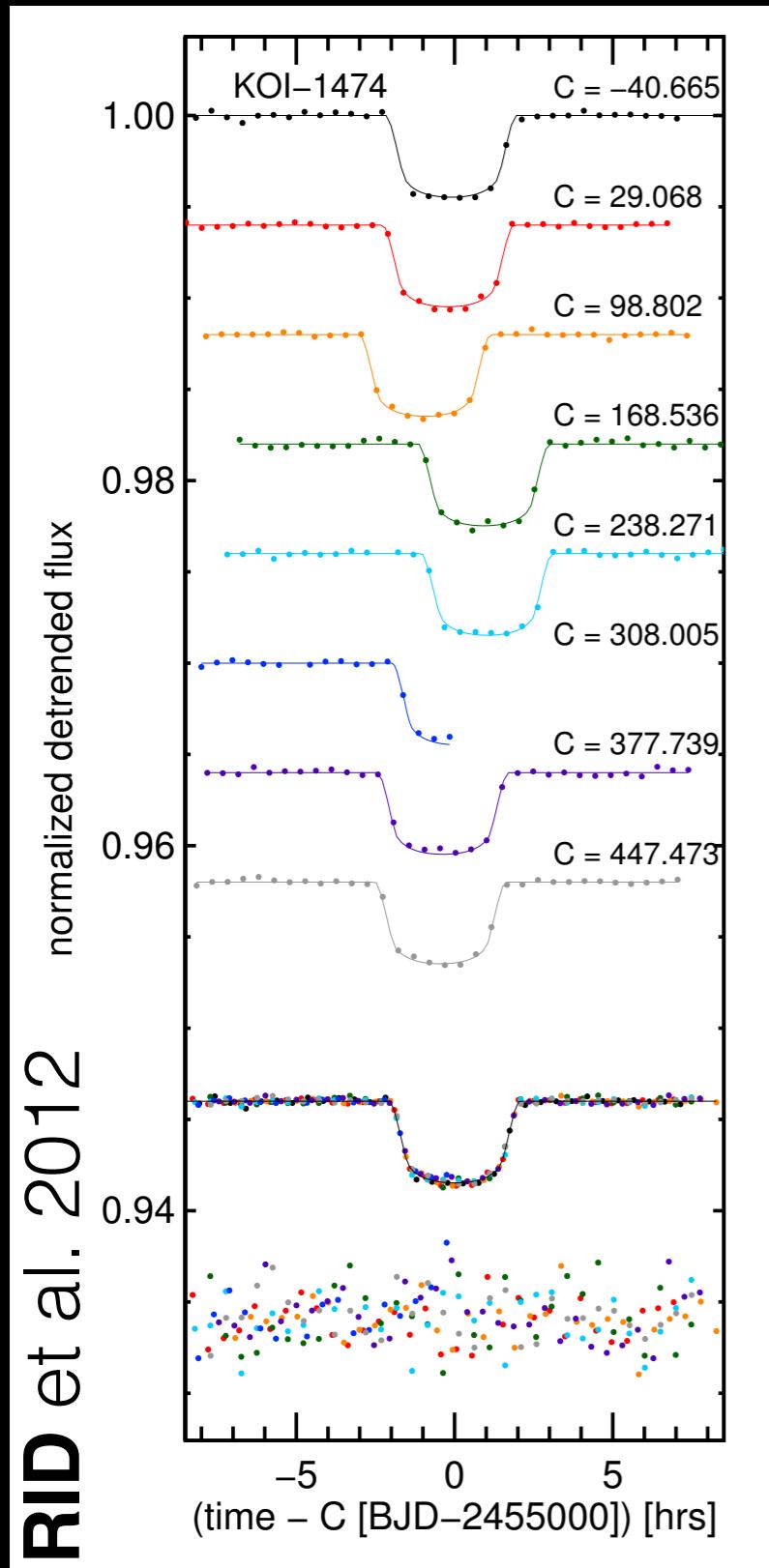
We know the Kepler-419 dataset contains correlated noise



Common behavior: three segment spectrum

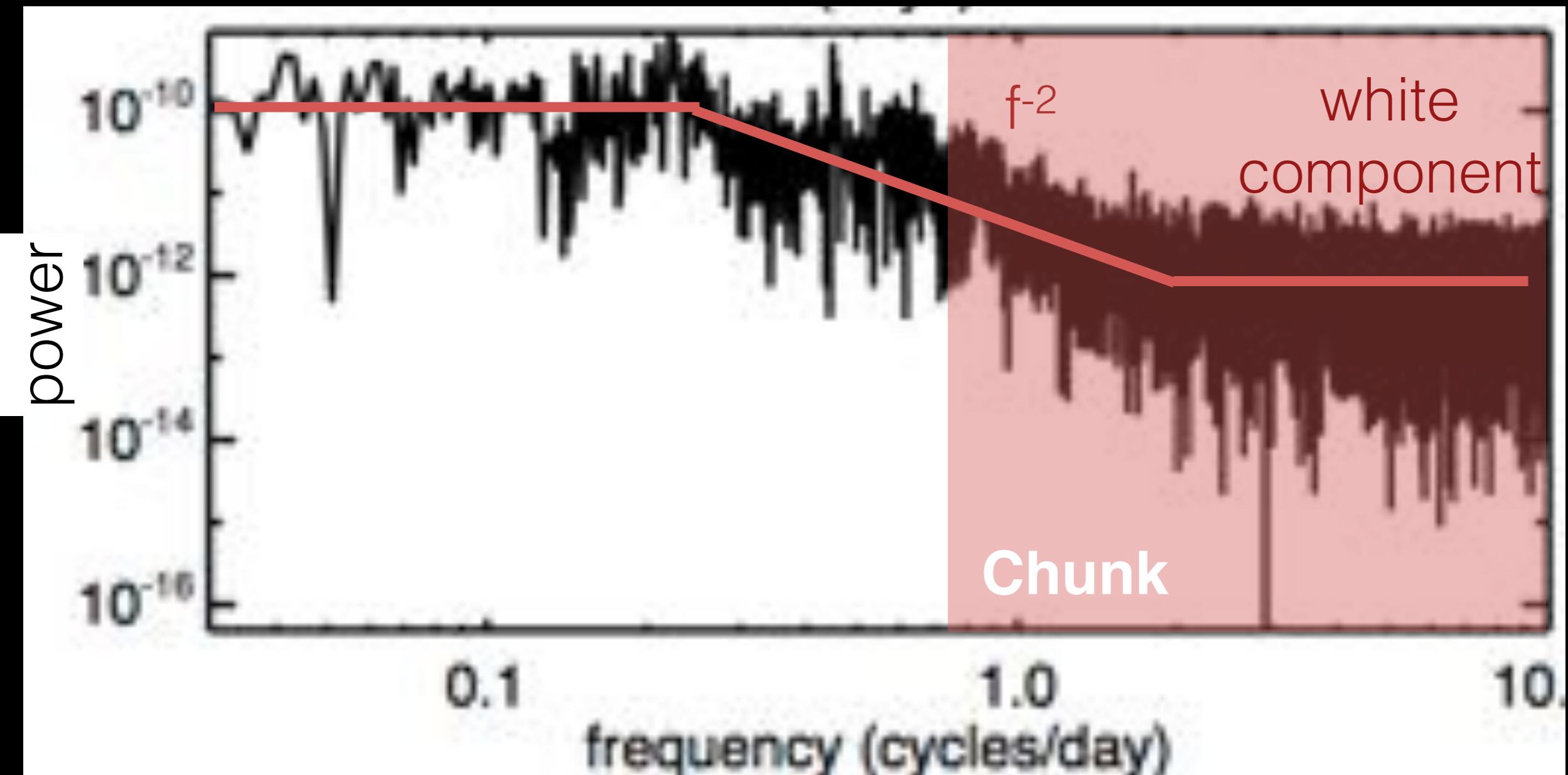


Case study: how did Kepler-419b achieve its close-in, highly elliptical orbit?



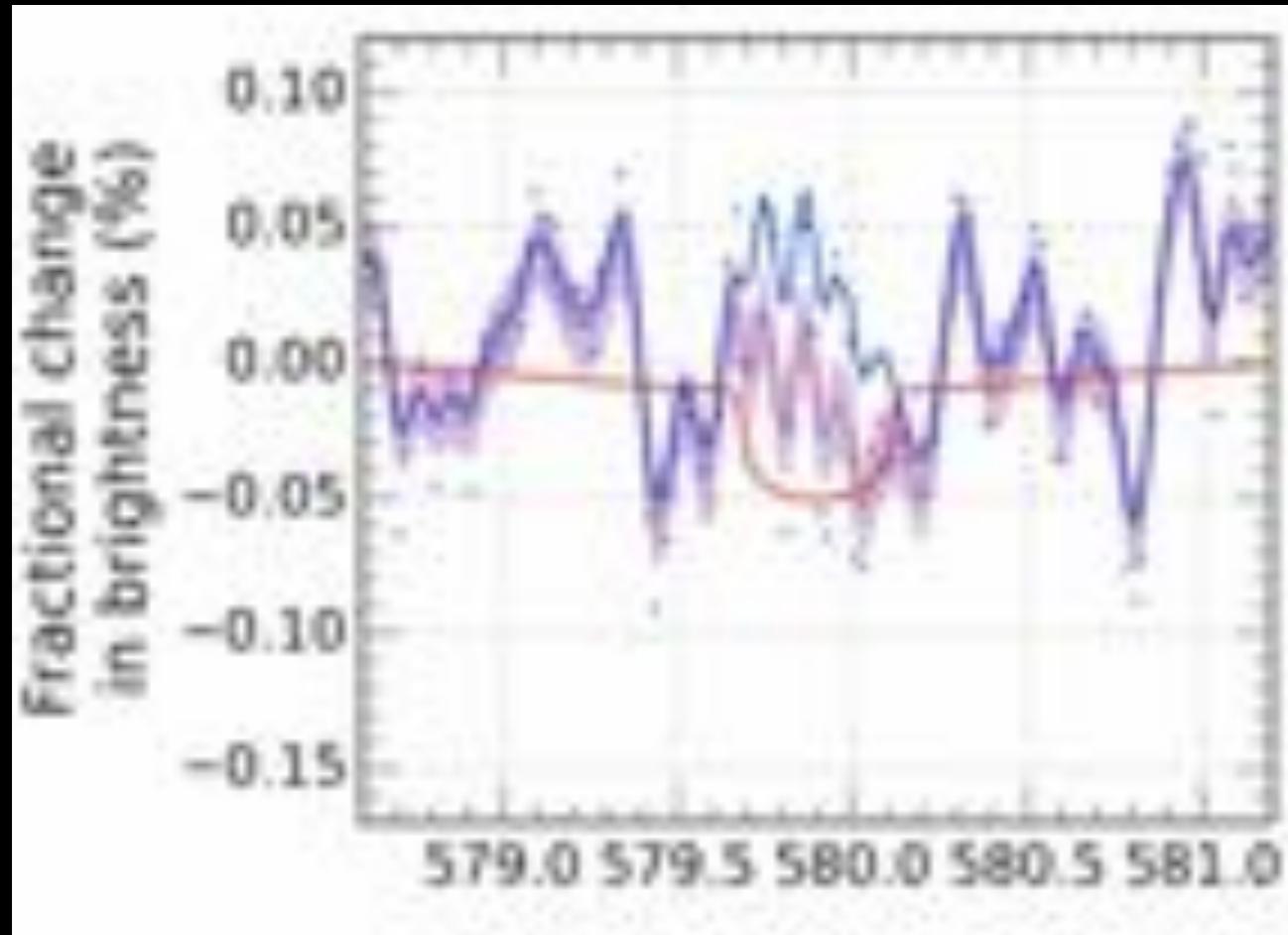
Is its non-transiting companion orbiting in the same plane?

Common behavior: three segment spectrum

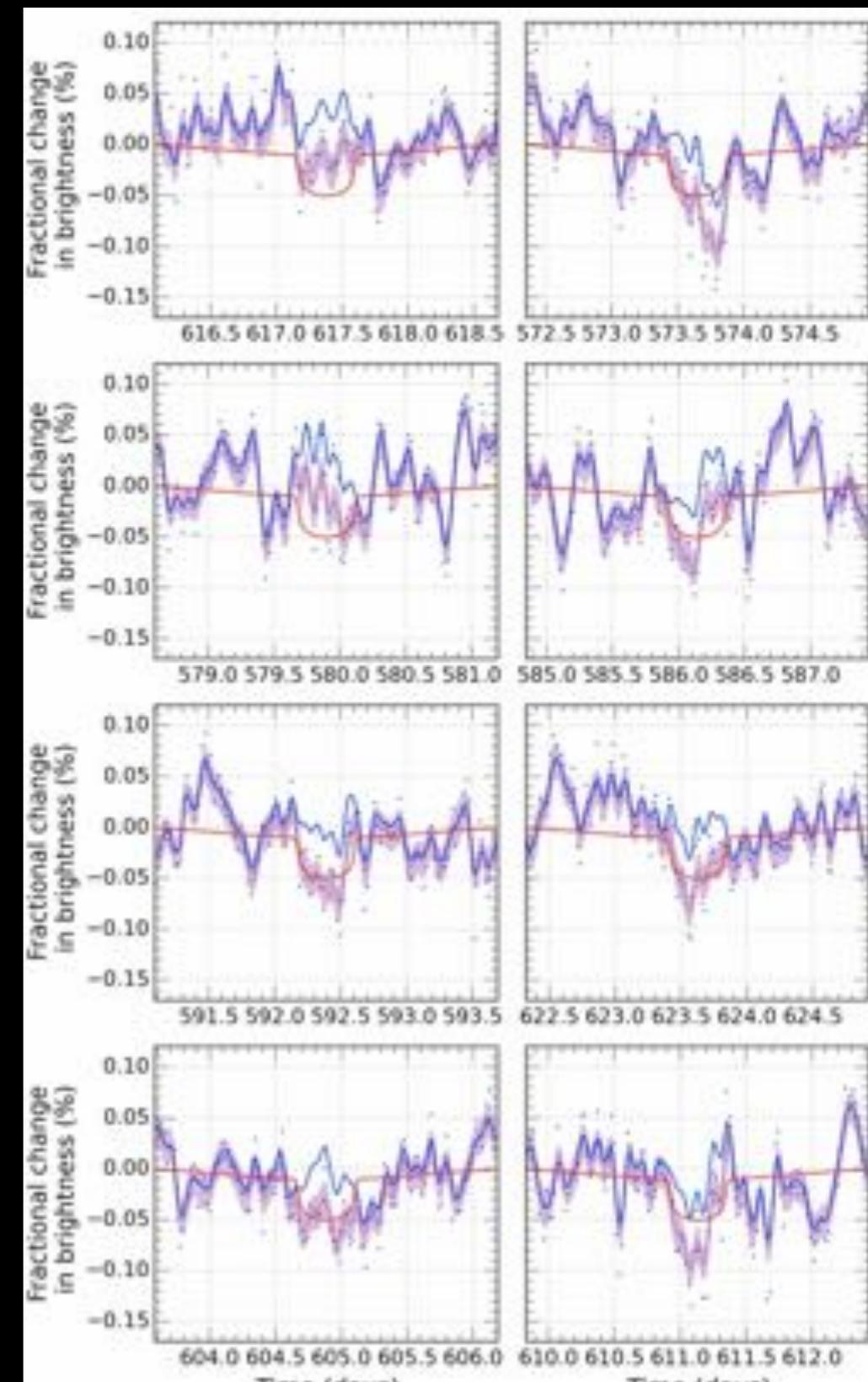


Pre-detrending the data can lead to errors in the inferred planet properties

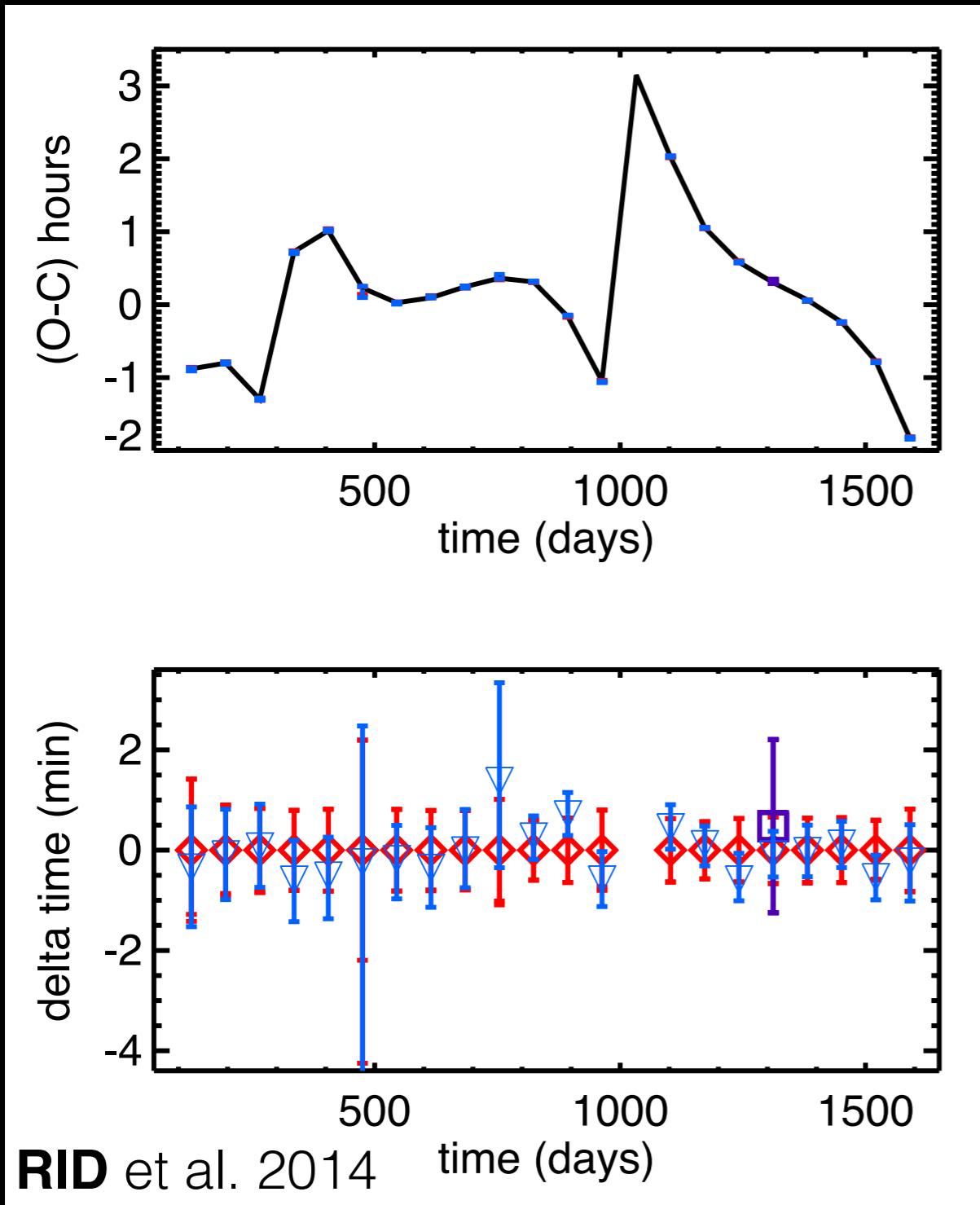
Barclay et al. 15, Kepler-91b
Gaussian process regression correlated noise
using george (Foreman-Mackey et al. in prep)



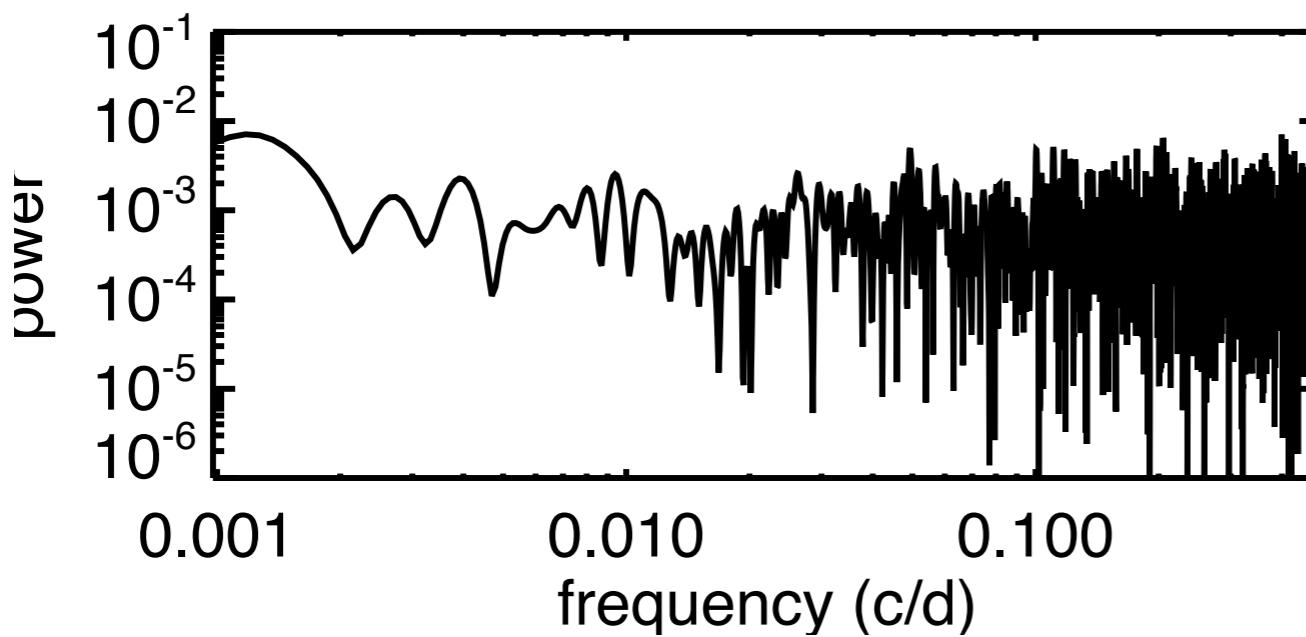
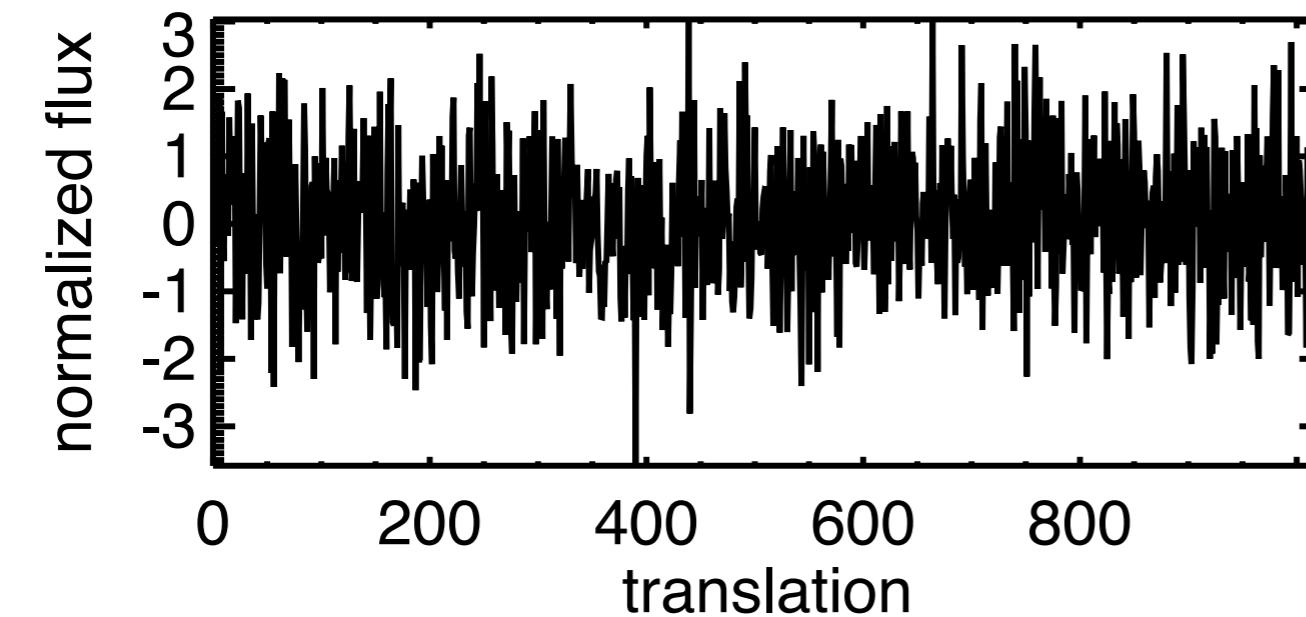
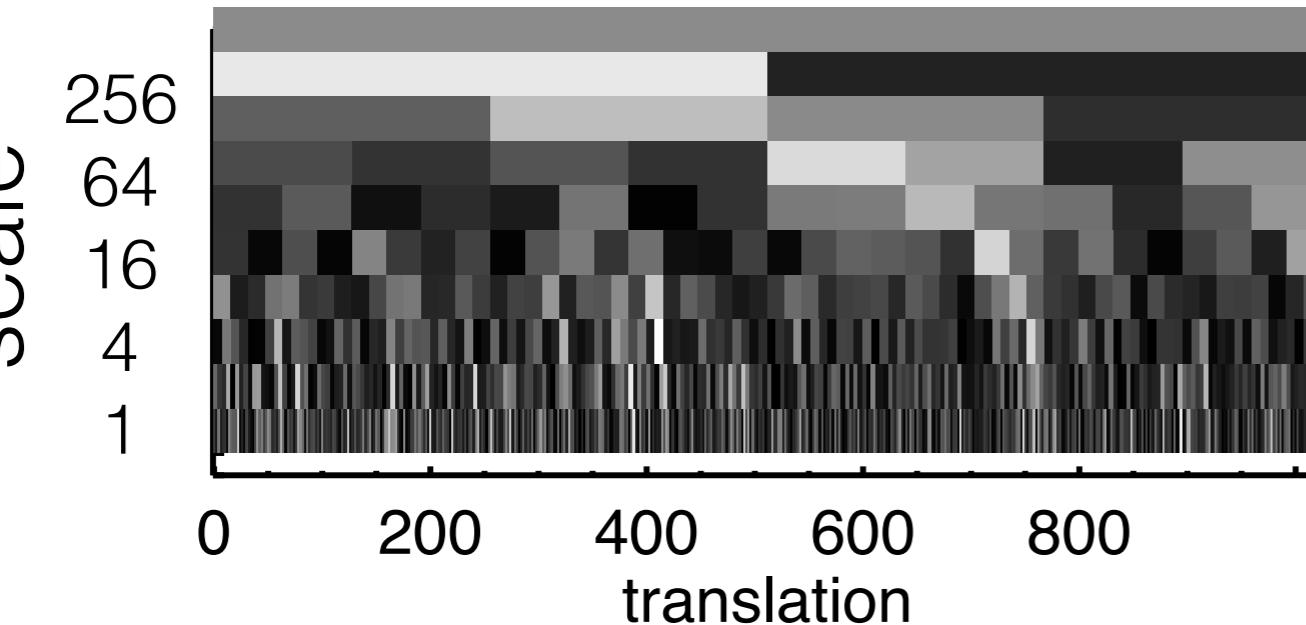
Previous pre-whitening treatment
caused this planet to be
misdiagnosed as an astrophysical
false positive (Sliski & Kipping 14)



Two different correlated noise treatments yield consistent transit times



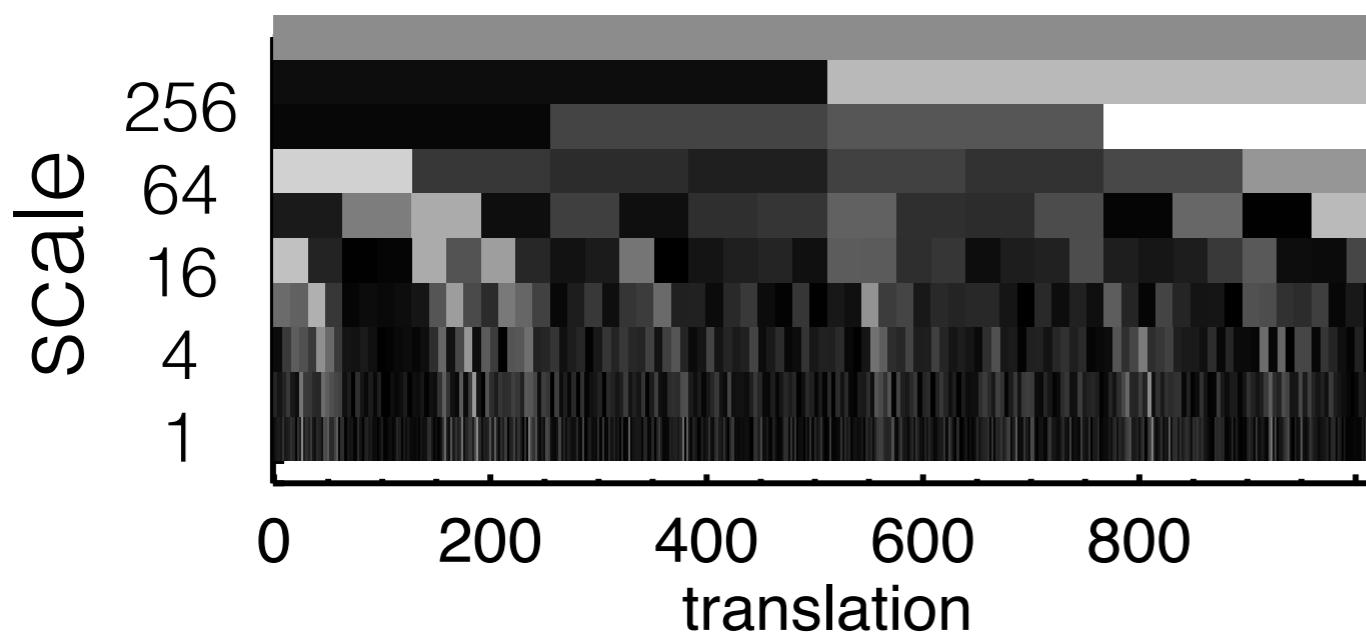
- ◆ Median filter detrending,
Carter & Winn 2009
wavelet likelihood
- ◆ Foreman-Mackey et al.
in prep. Gaussian
process regression
likelihood with squared
exponential covariance
kernel, dan.iel.fm/george



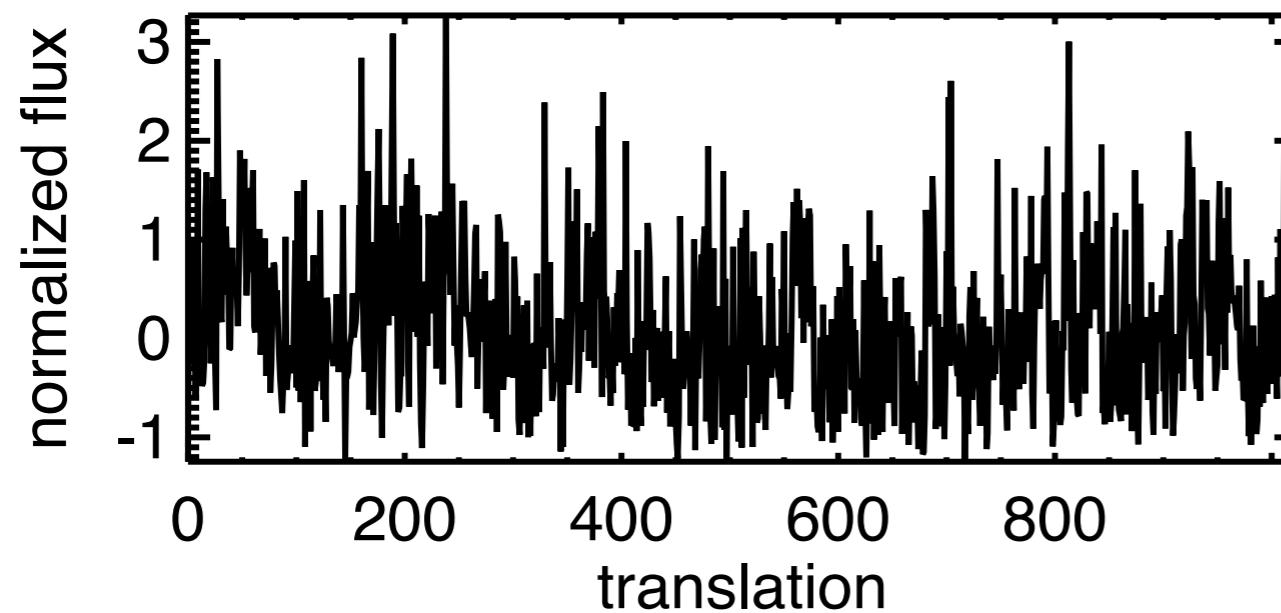
White noise

Wavelet transform
computes power
for different
translations and
scales

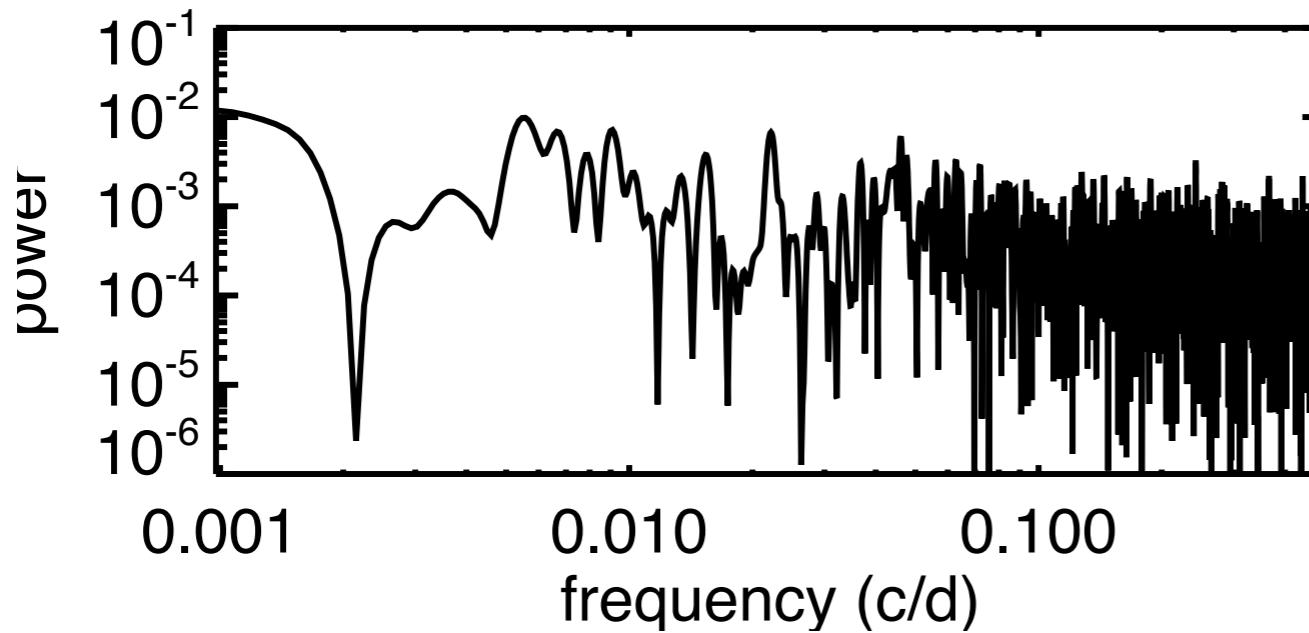
R.



Pink ($1/f$) noise



Wavelet transform
computes power
for different
translations and
scales



Wavelet likelihood method parametrizes noise into red σ_r and white σ_w component

Carter & Winn 2009 (144 citations)
based on Wornell 1996:

Signal Processing with Fractals: A Wavelet-Based Approach

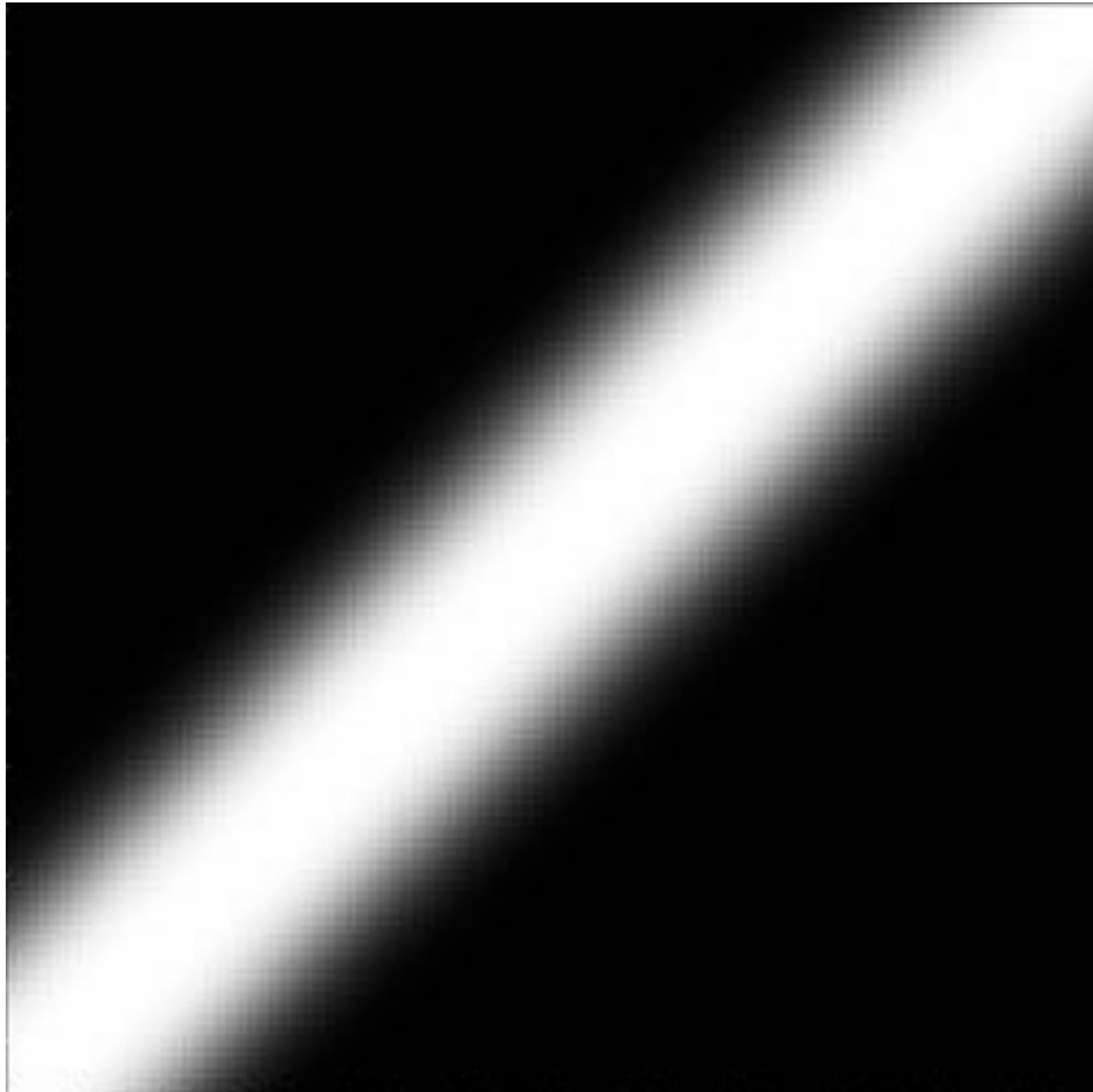
$$\mathcal{L} = \left\{ \prod_{m=2}^M \prod_{n=1}^{R_0 2^{m-1}} \frac{1}{\sqrt{2\pi\sigma_W^2}} \exp \left[-\frac{(r_n^m)^2}{2\sigma_W^2} \right] \right\} \\ \times \left\{ \prod_{n=1}^{R_0} \frac{1}{\sqrt{2\pi\sigma_S^2}} \exp \left[-\frac{(\bar{r}_n^1)^2}{2\sigma_S^2} \right] \right\},$$

$$\sigma_S^2 = \sigma_r^2 2^{-\gamma} g(\gamma) + \sigma_w^2,$$

$$\sigma_W^2 = \sigma_r^2 2^{-\gamma m} + \sigma_w^2,$$

Dictated relationship
between
scale coefficients
for $1/f^\gamma$ noise

translation



translation

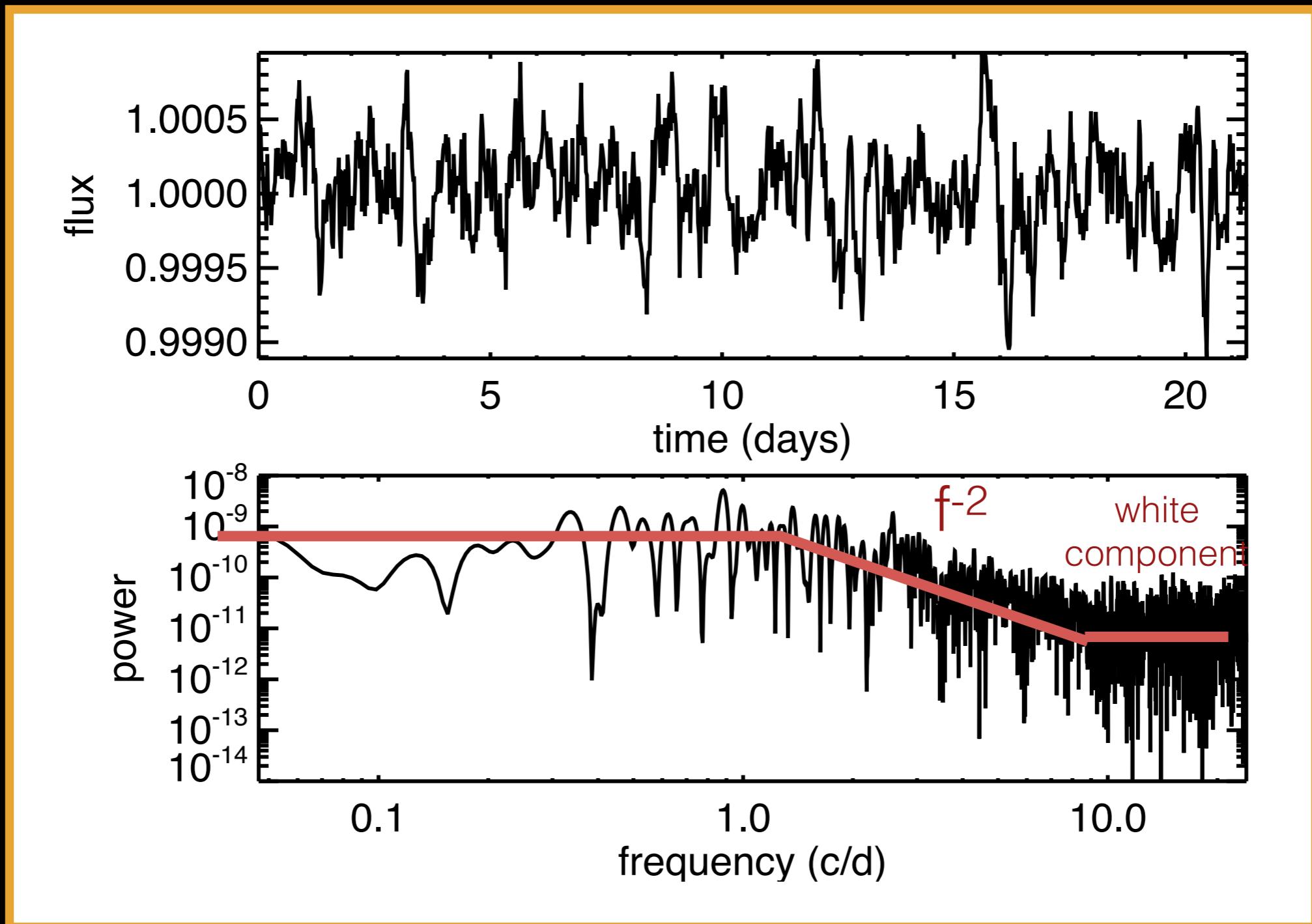
Gaussian process
regression likelihood:
prescription for
covariance matrix

implemented using
dan.iel.fm/george

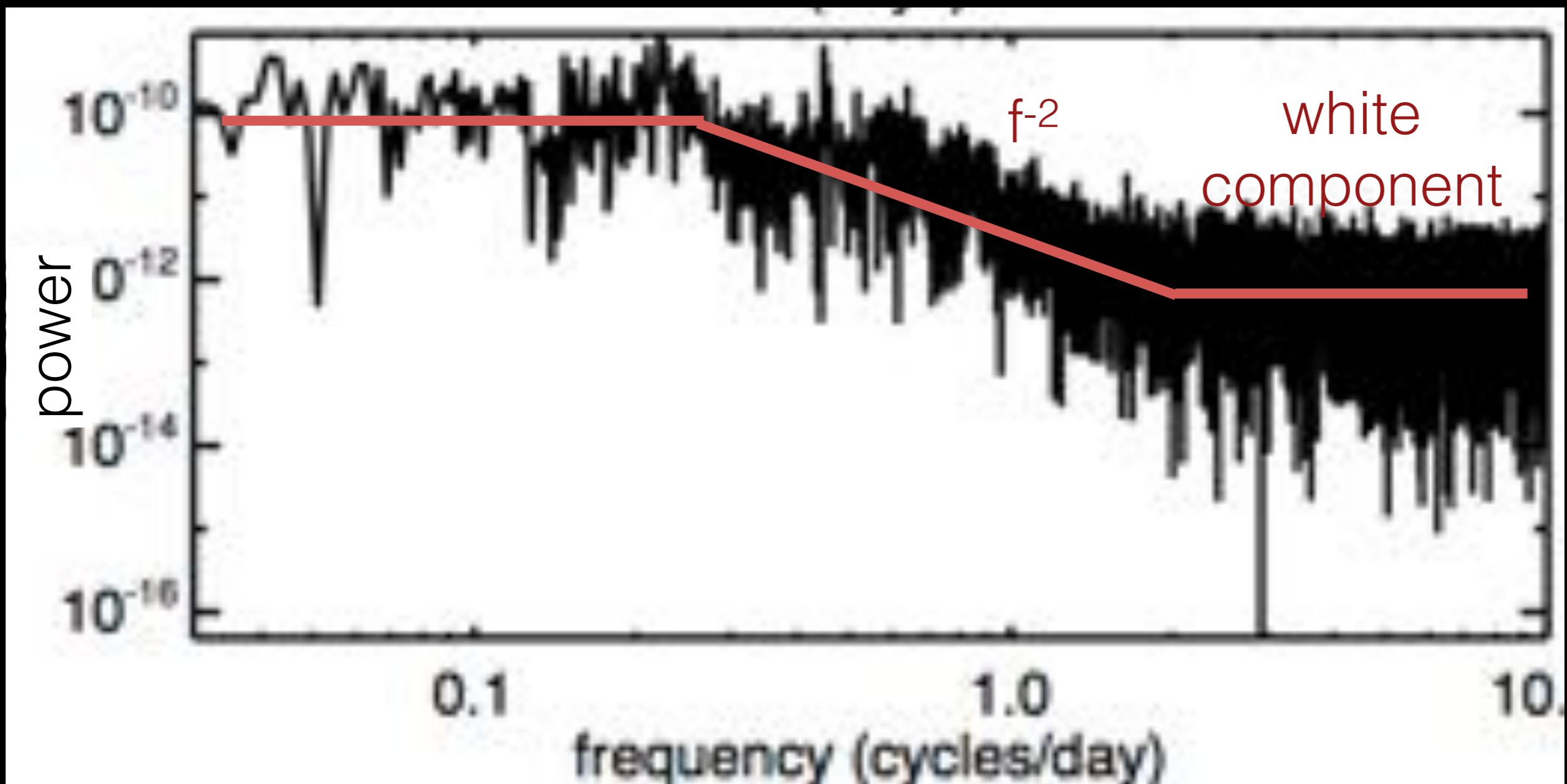
$$k(r^2) = \left(1 + \sqrt{5} r^2 + \frac{5 r^2}{3} \right) \exp\left(-\sqrt{5} r^2\right)$$

Radial Matern
5/2 kernel

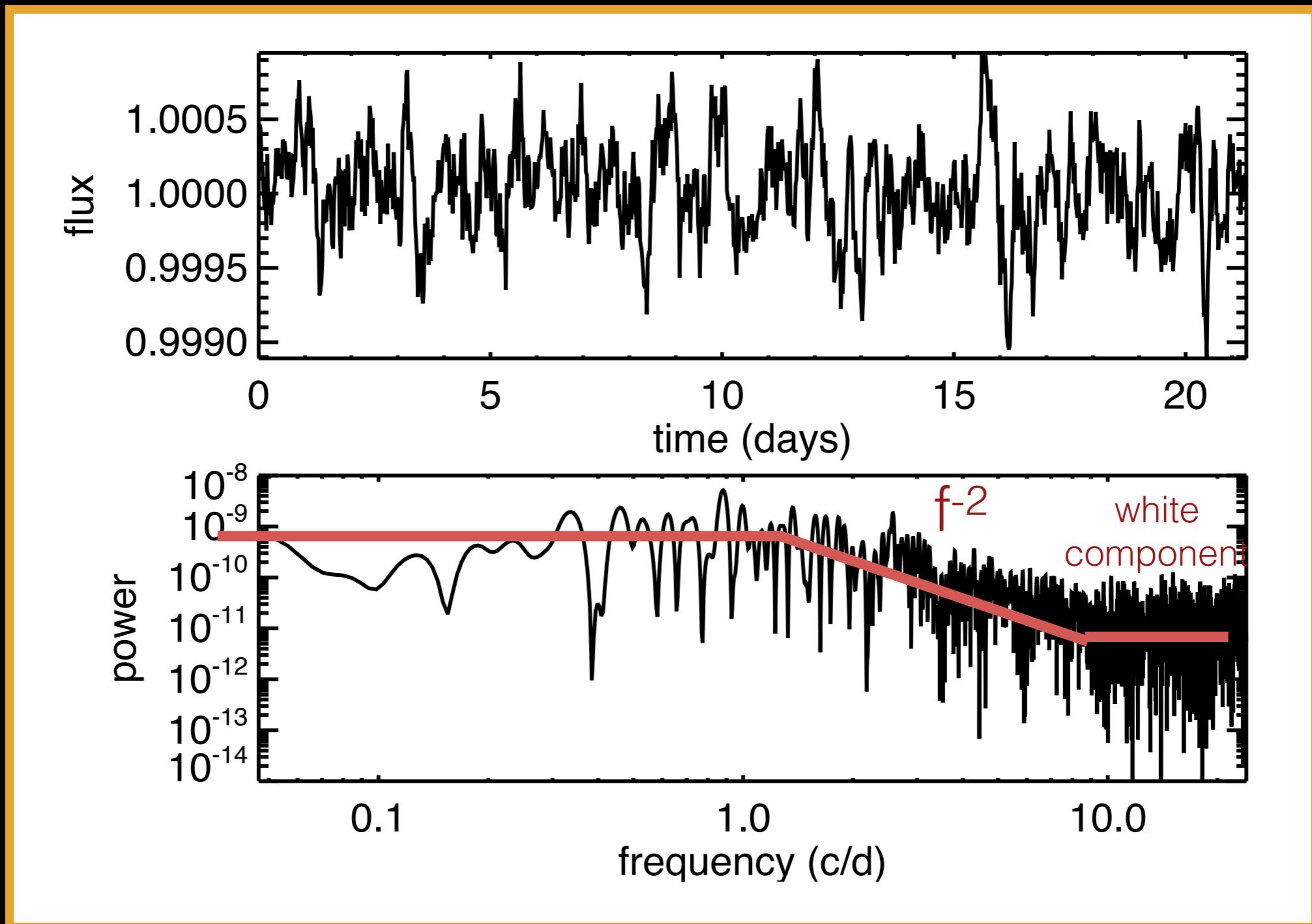
Gaussian process generated with Matern kernel in frequency space



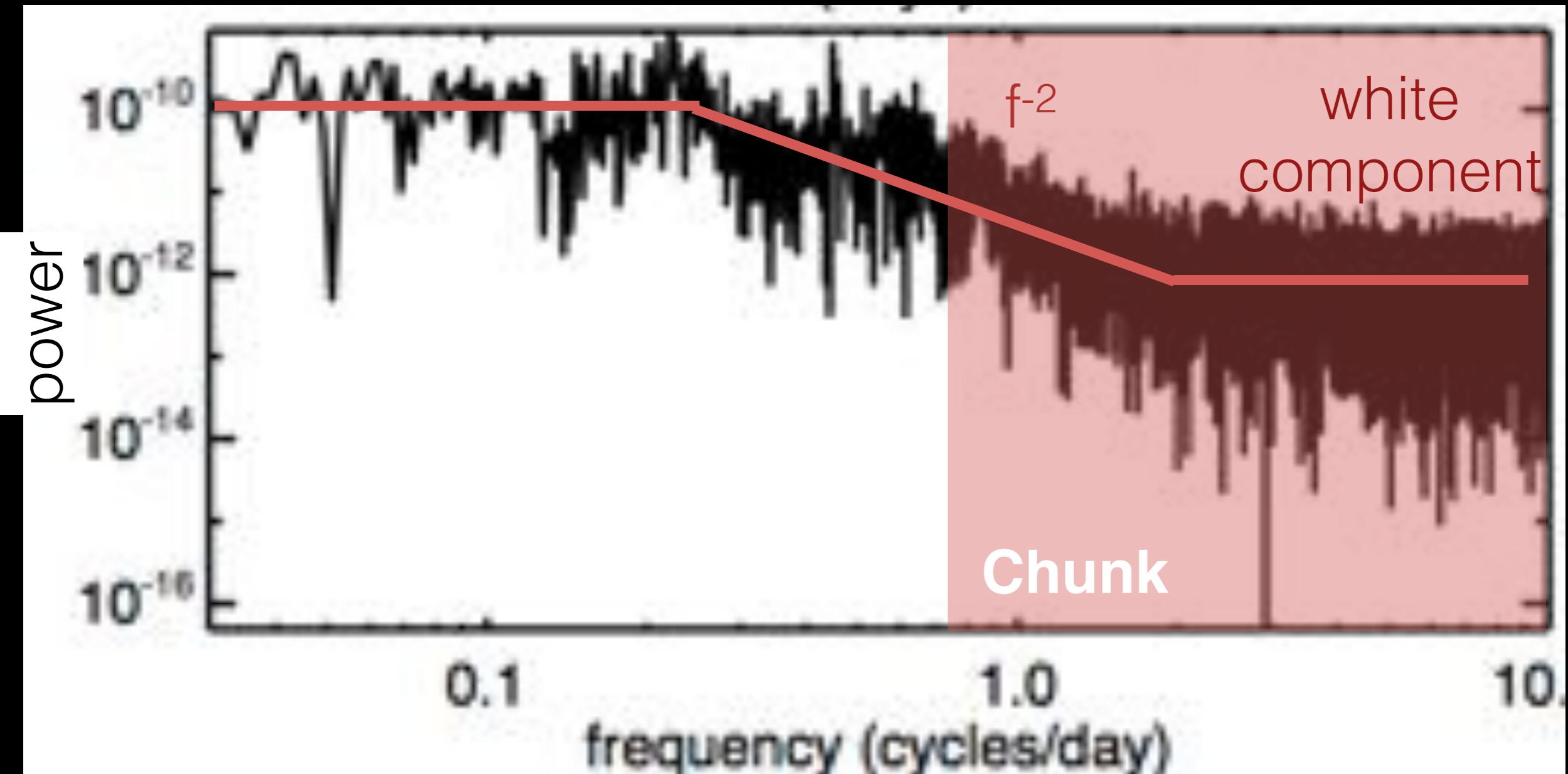
Common behavior: three segment spectrum



Gaussian process generated with Matern kernel in frequency space



Common behavior: three segment spectrum

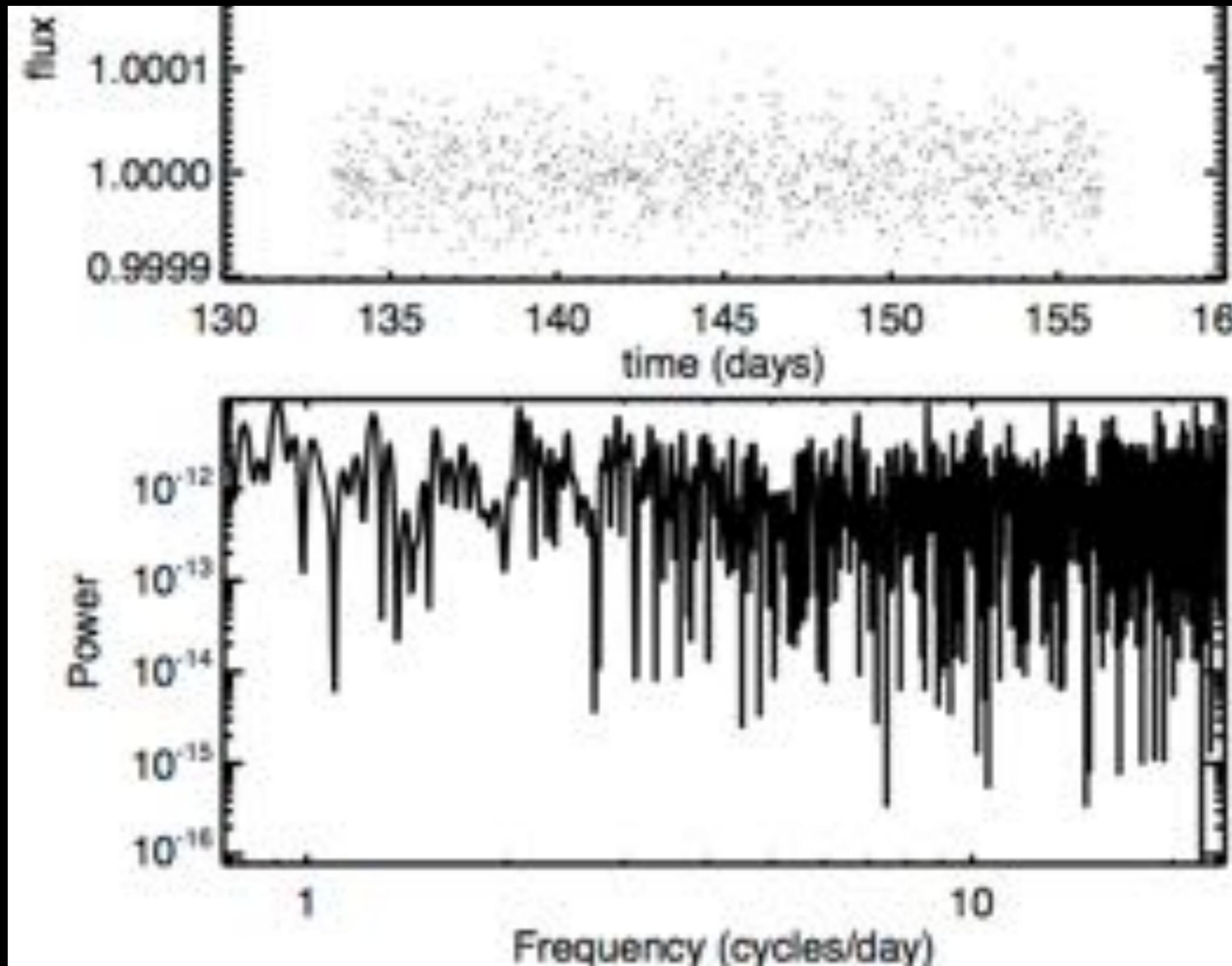


Noise properties of Sun-like *Kepler* stars

Sometimes white-noise dominates

e.g., sun-like star
KIC
12011630

$\sigma_w = 35 \pm 1$
ppm

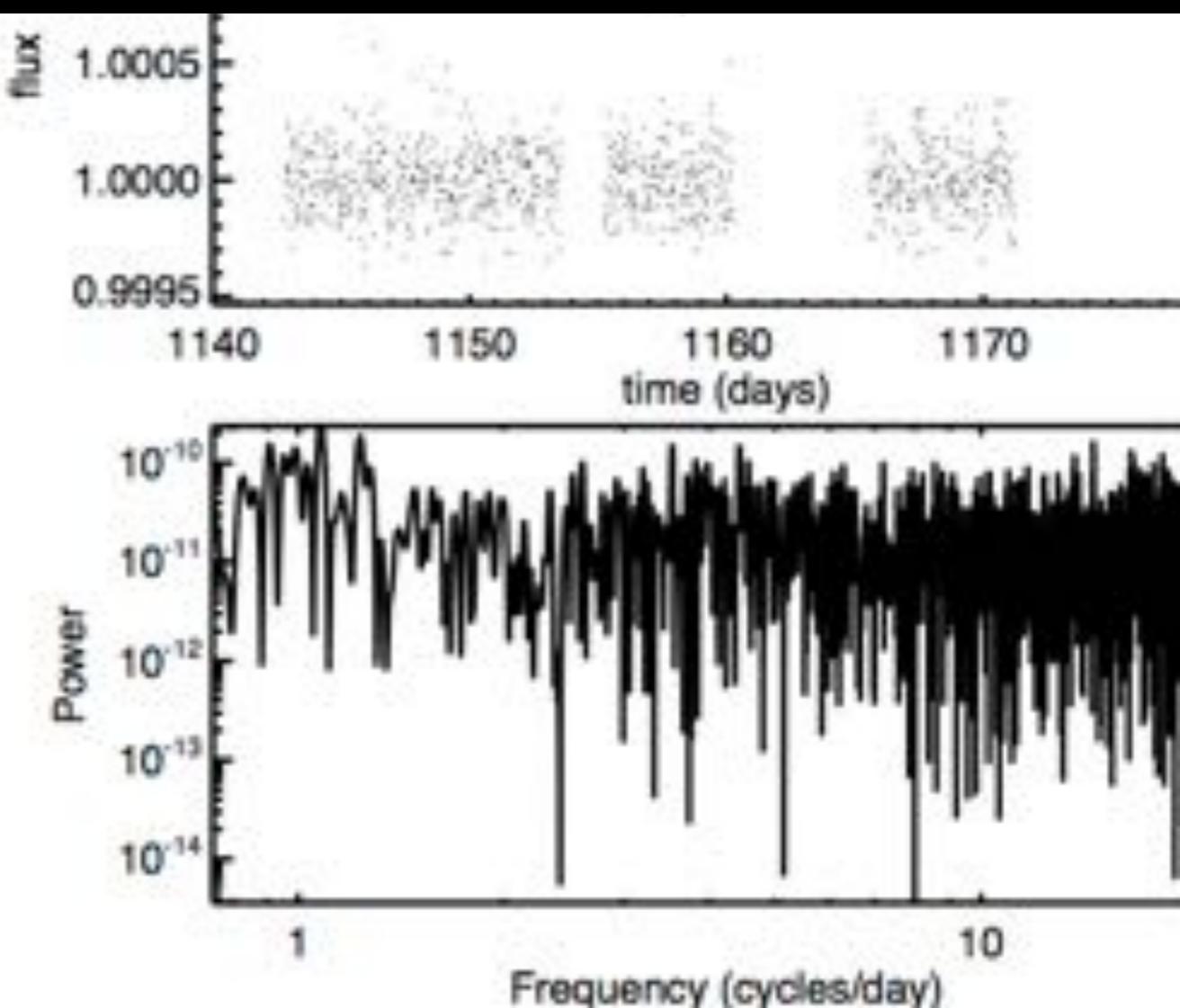
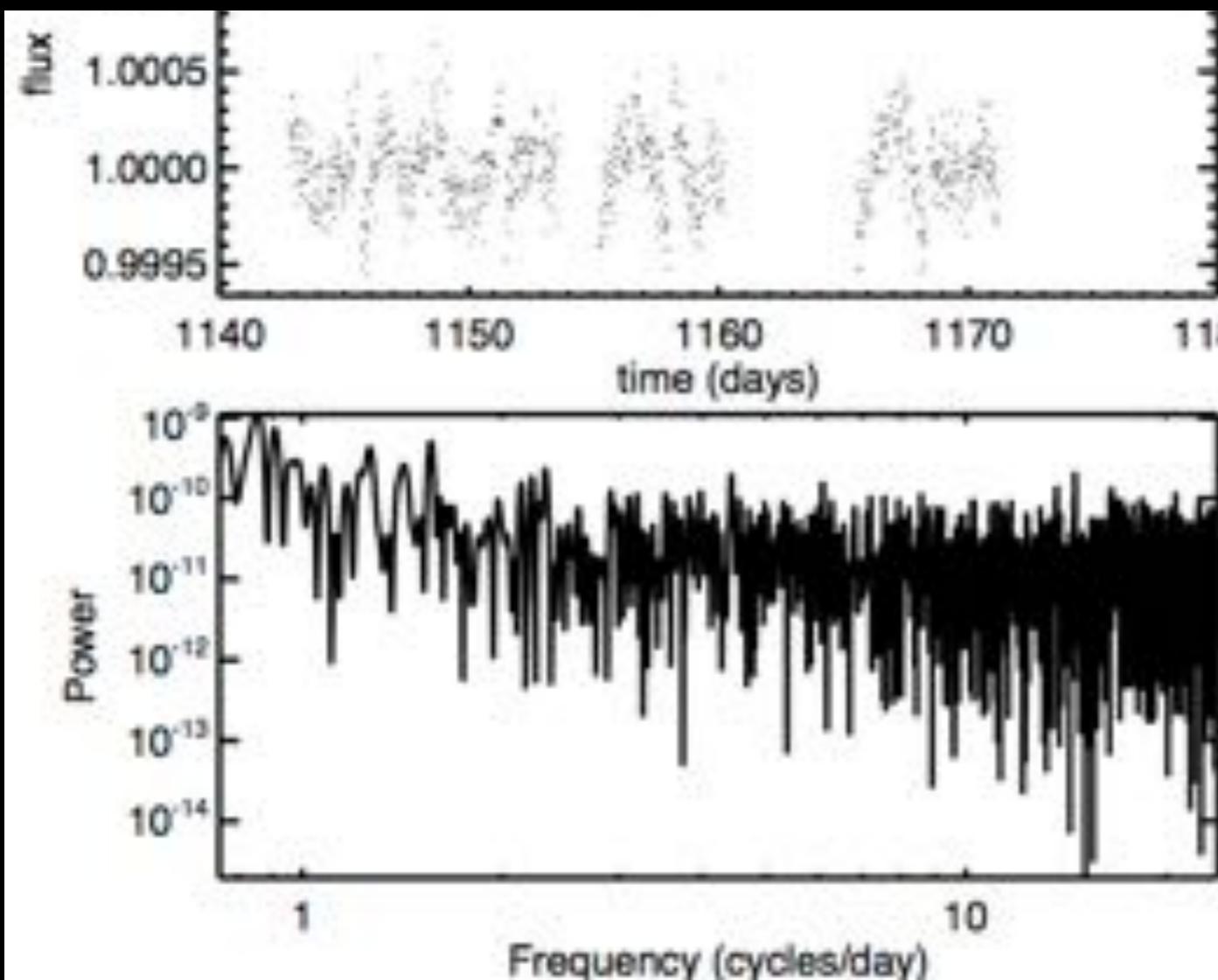


Simultaneous linear fitting: sometimes sufficient

e.g., sun-like star KIC 8374139

$\sigma_w = 106 \pm 8$ ppm,
 $\sigma_r = 700 \pm 40$ ppm

$\sigma_w = 150 \pm 5$ ppm

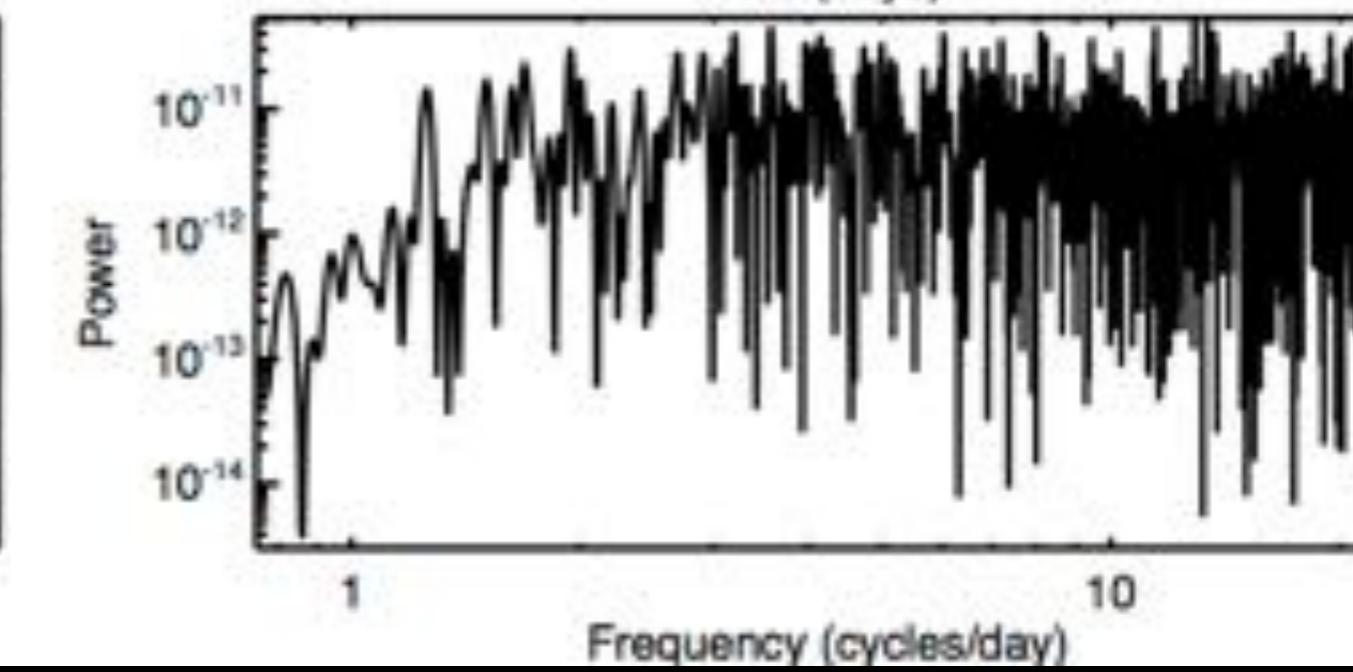
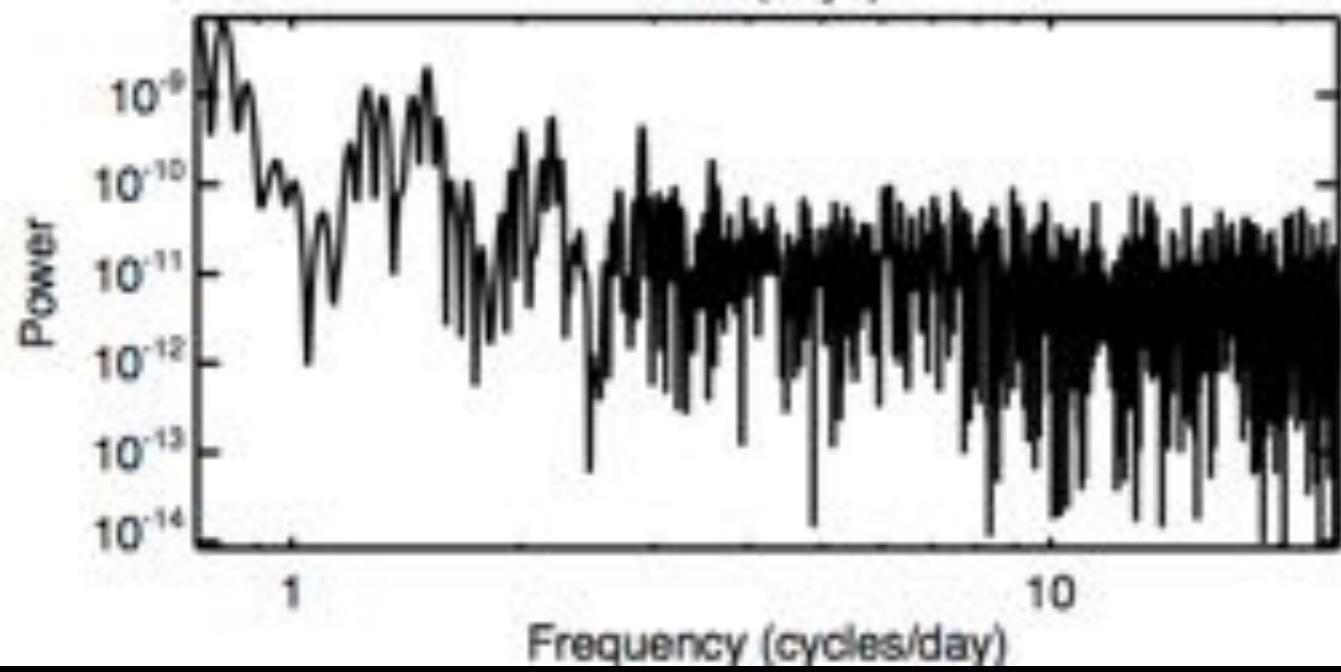
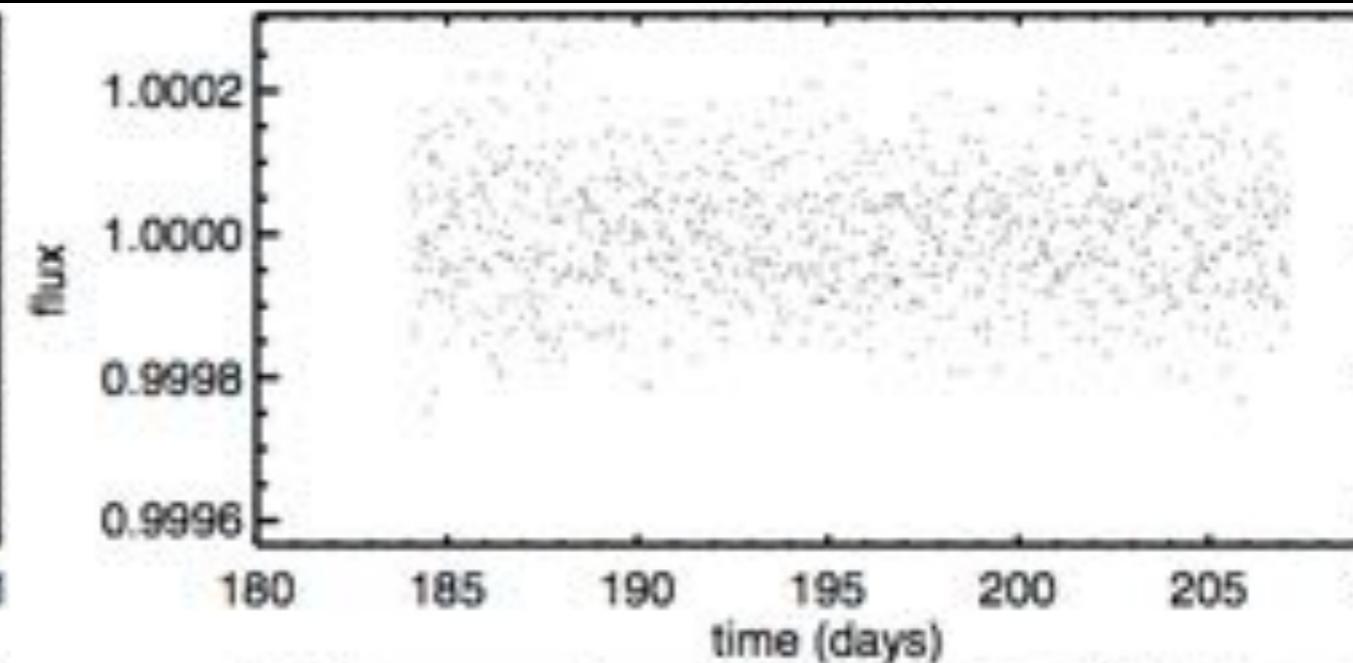
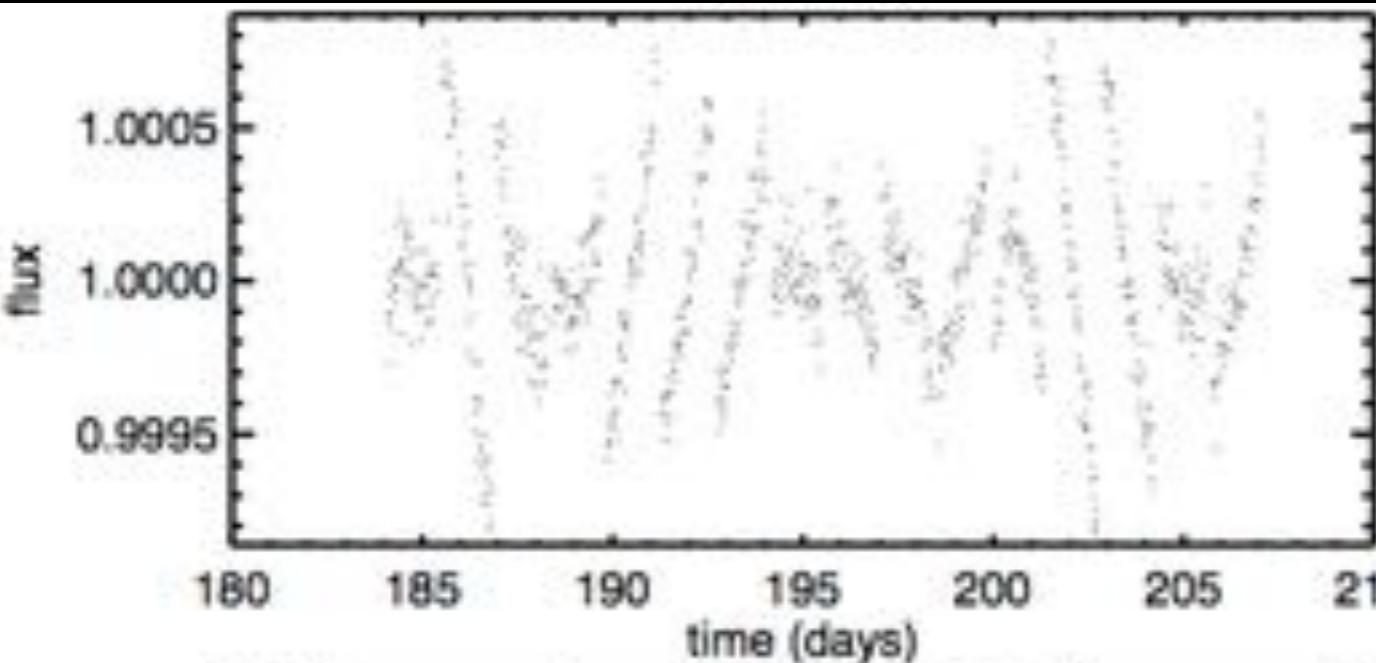


Simultaneous polynomial fitting: sometimes sufficient

e.g., sun-like star KIC 3970397

$$\sigma_w = 88 \pm 2 \text{ ppm}, \sigma_r = 53 \pm 5 \text{ ppm}$$

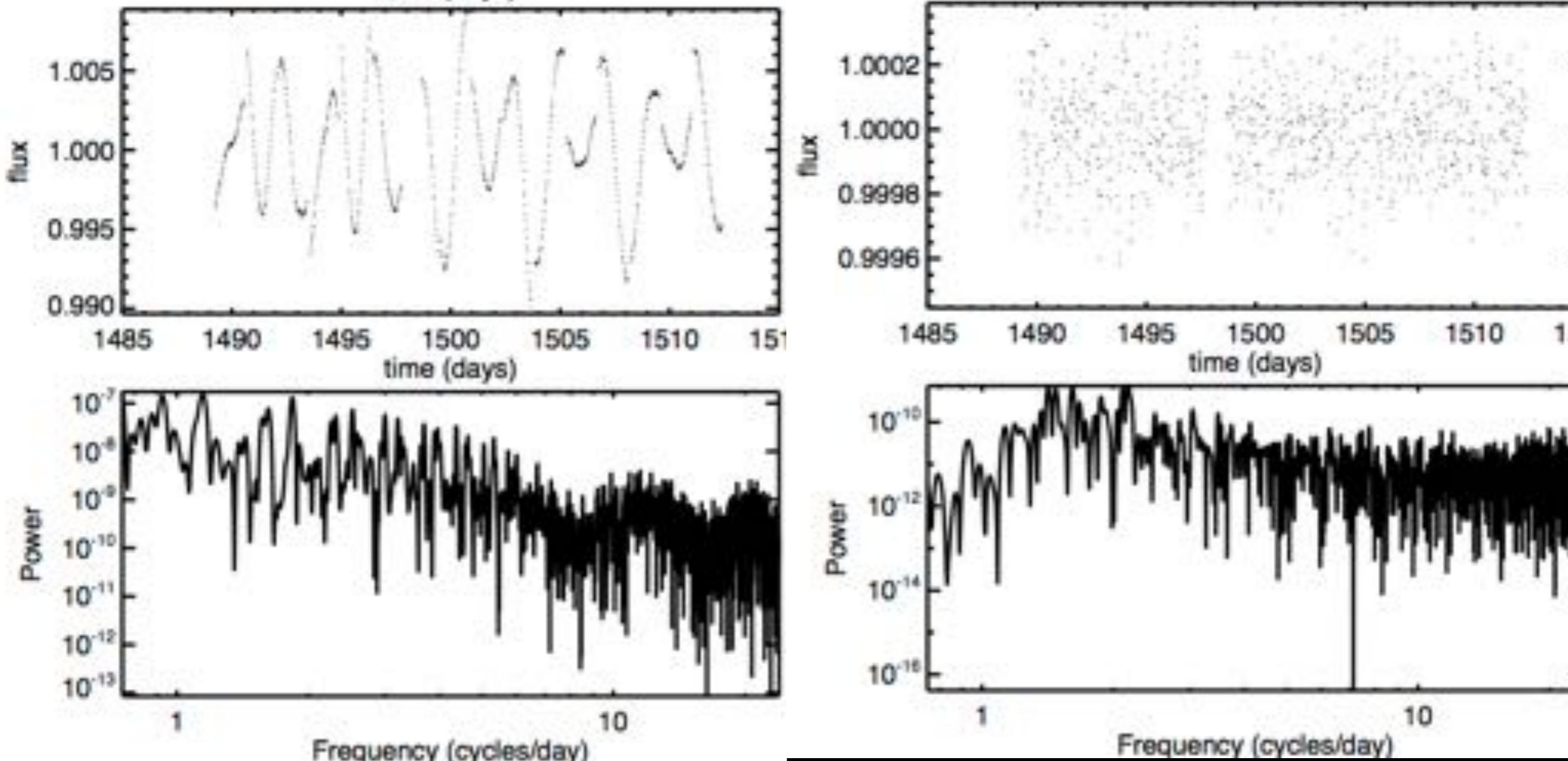
$$\sigma_w = 86 \pm 3 \text{ ppm}$$



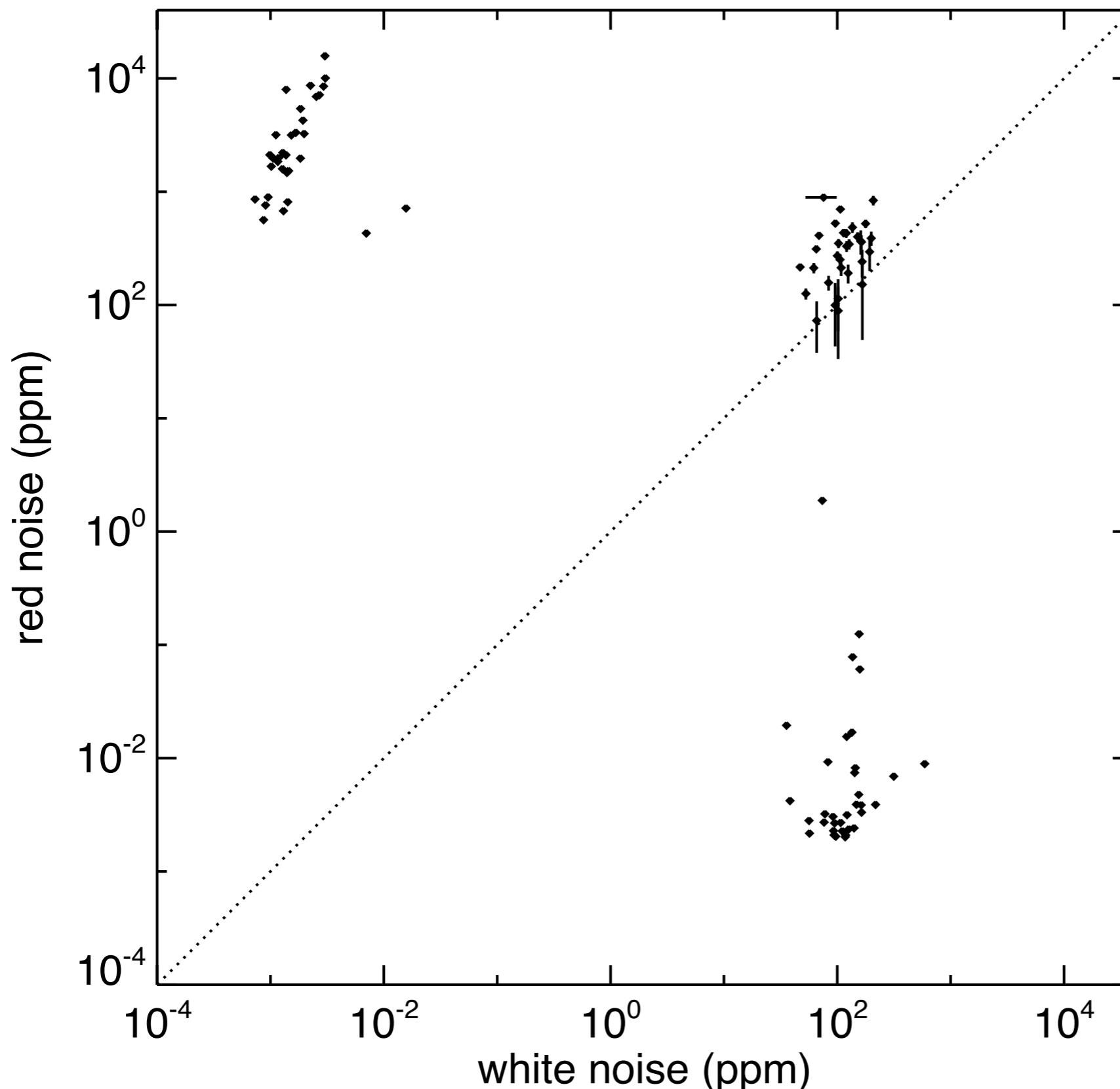
Simultaneous polynomial fitting: sometimes insufficient

e.g., sun-like star KIC 4819602

$\sigma_w = 0 \pm 10$ ppm, $\sigma_r = 8799 \pm 200$ ppm $\sigma_w = 64 \pm 8$ ppm, $\sigma_r = 580 \pm 30$ ppm

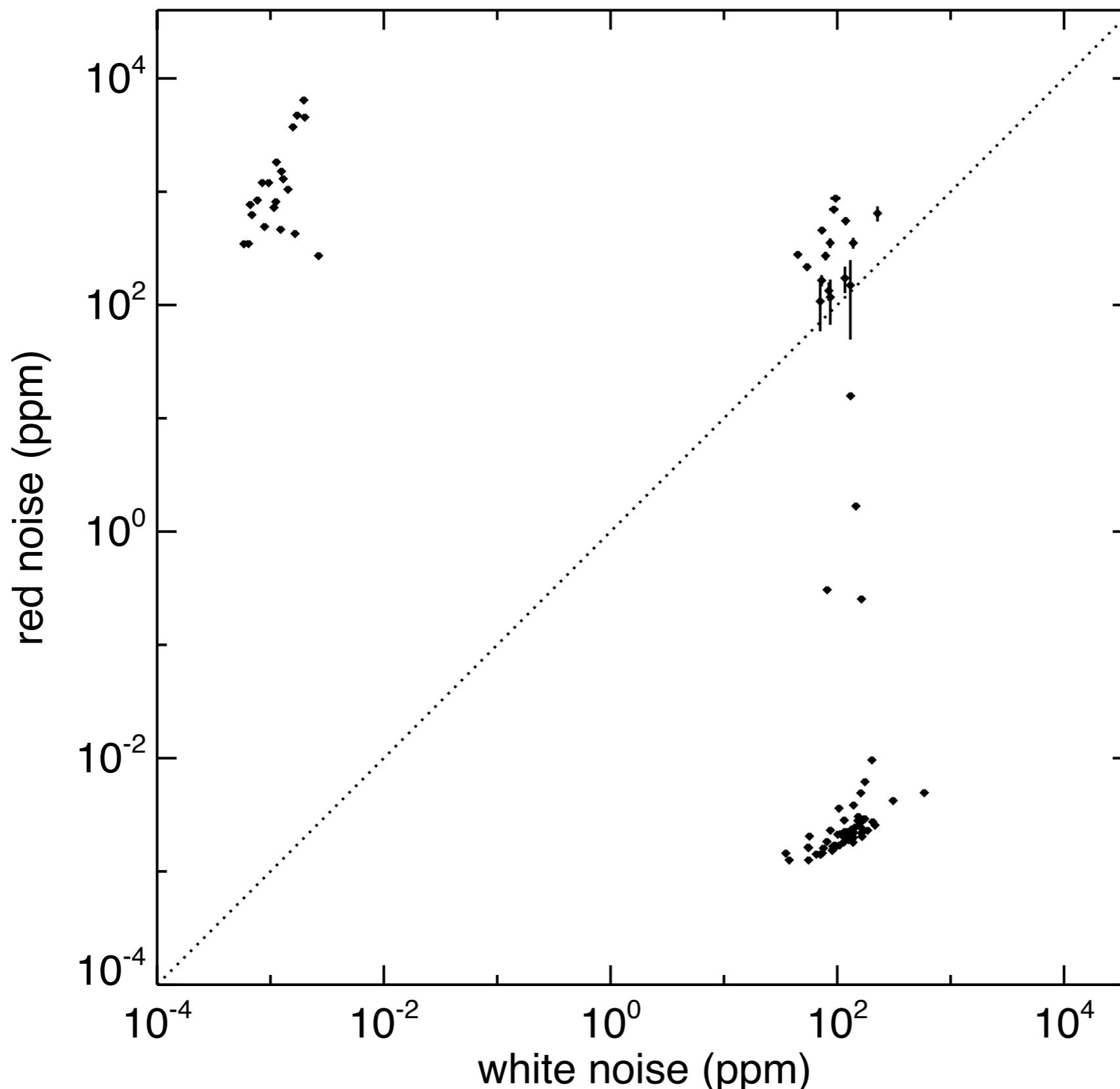


Kepler sun-like star properties: wavelet likelihood



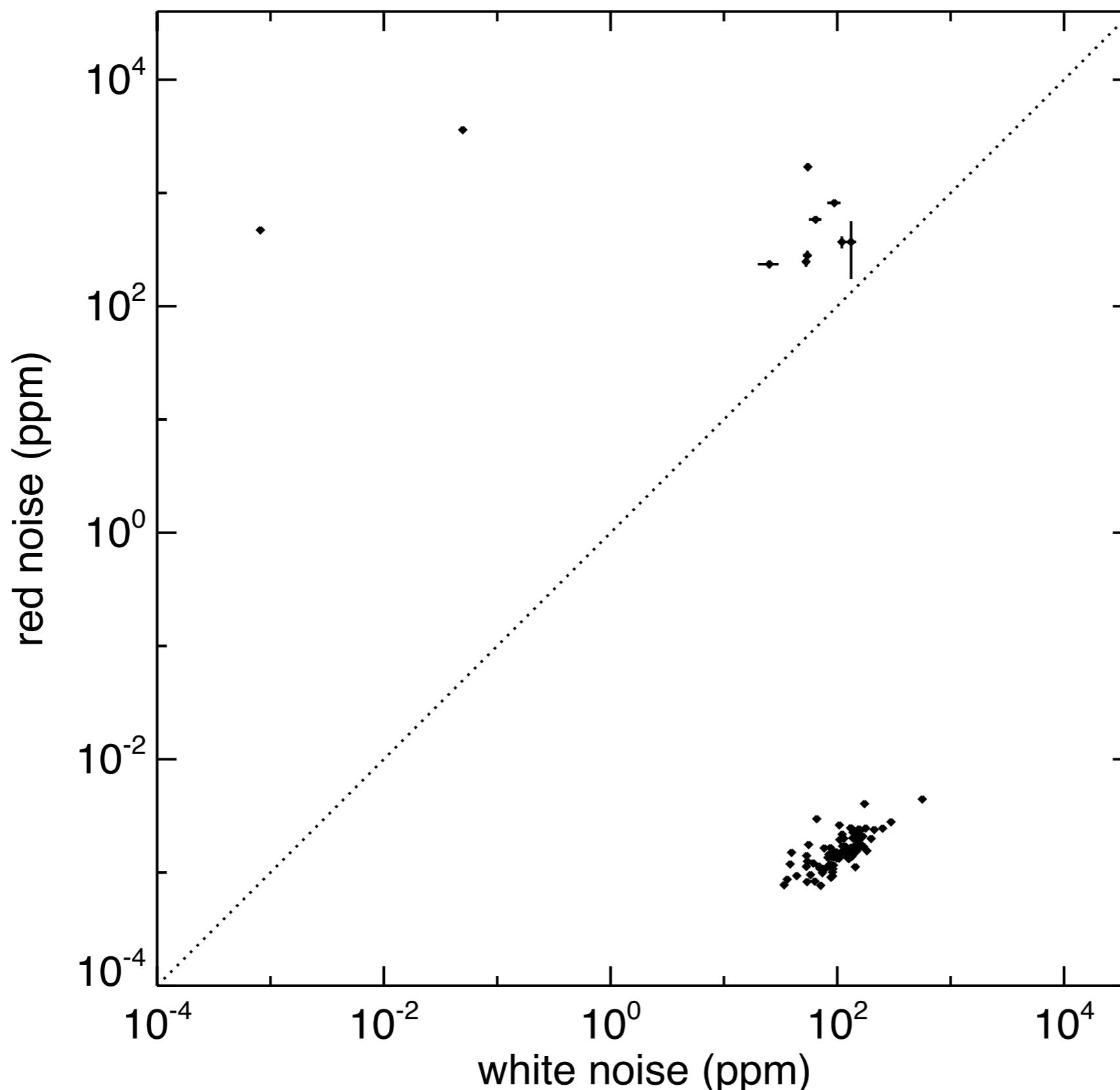
no polynomial

Kepler sun-like star properties: wavelet likelihood



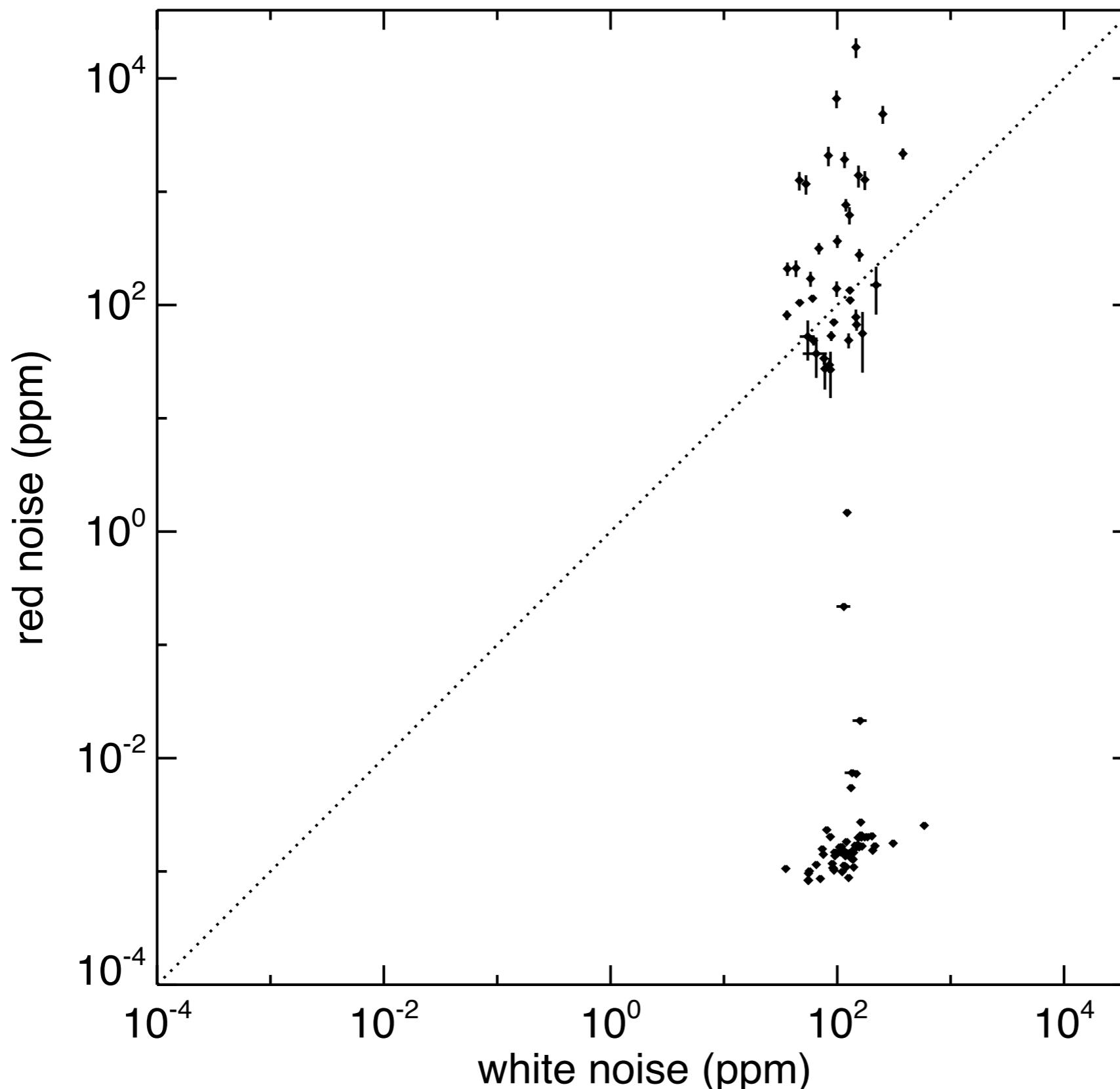
simultaneous
line

Kepler sun-like star properties: wavelet likelihood



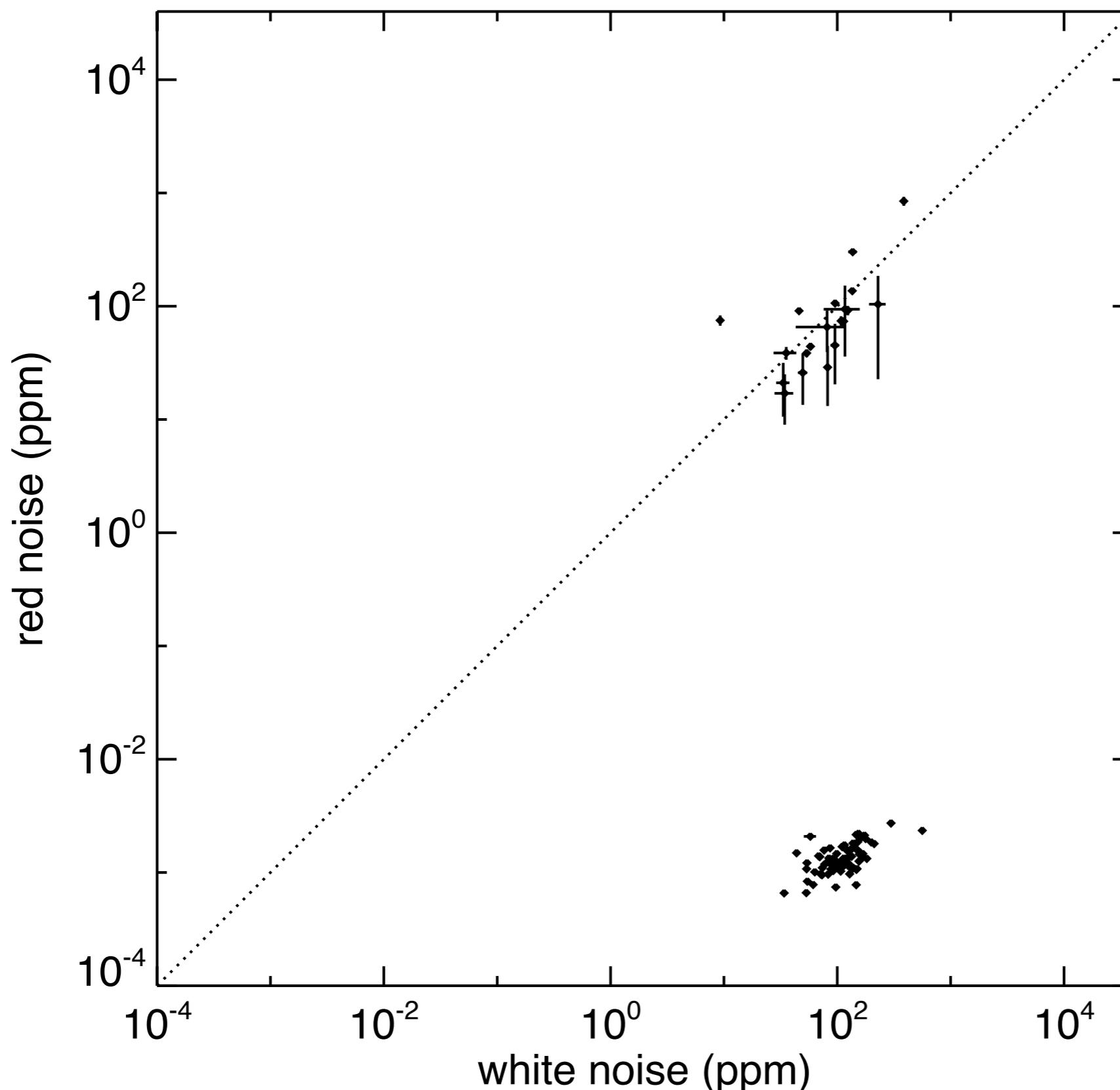
simultaneous polynomial

Kepler sun-like star properties: GPR likelihood



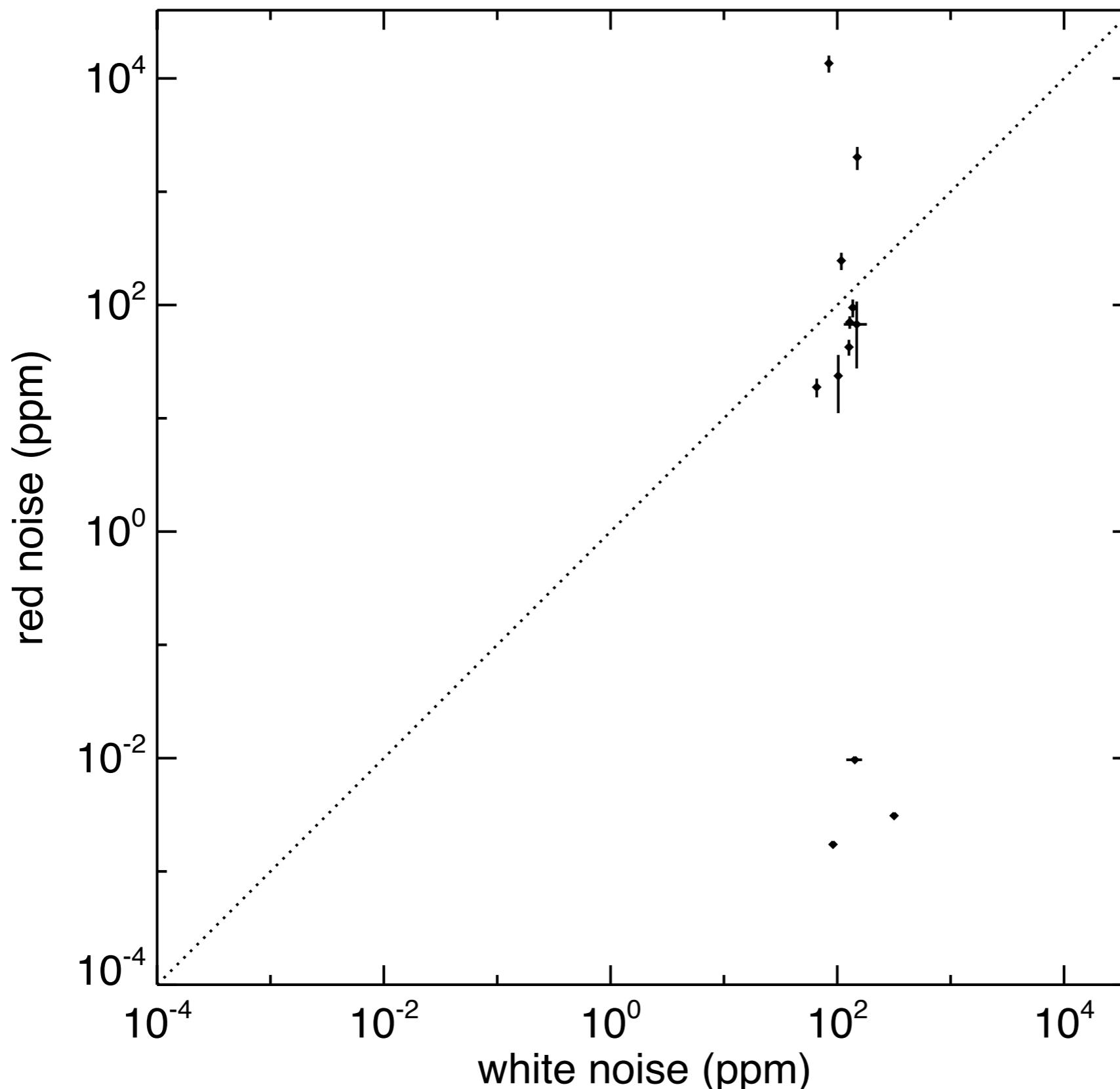
simultaneous
line

Kepler sun-like star properties: GPR likelihood



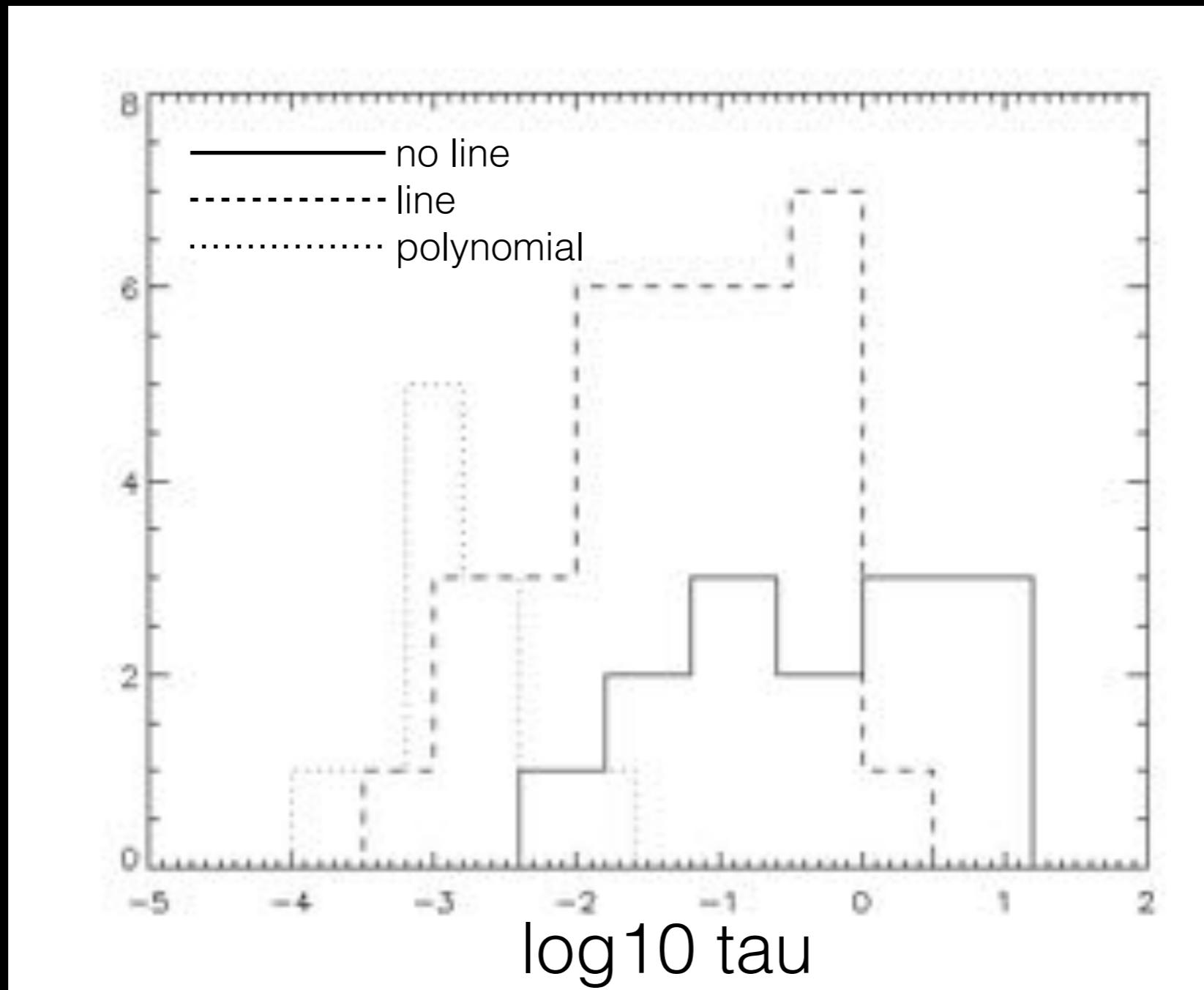
simultaneous
polynomial

Kepler sun-like star properties: GPR likelihood



no polynomial

Kepler sun-like star properties: Gaussian process regression, timescale

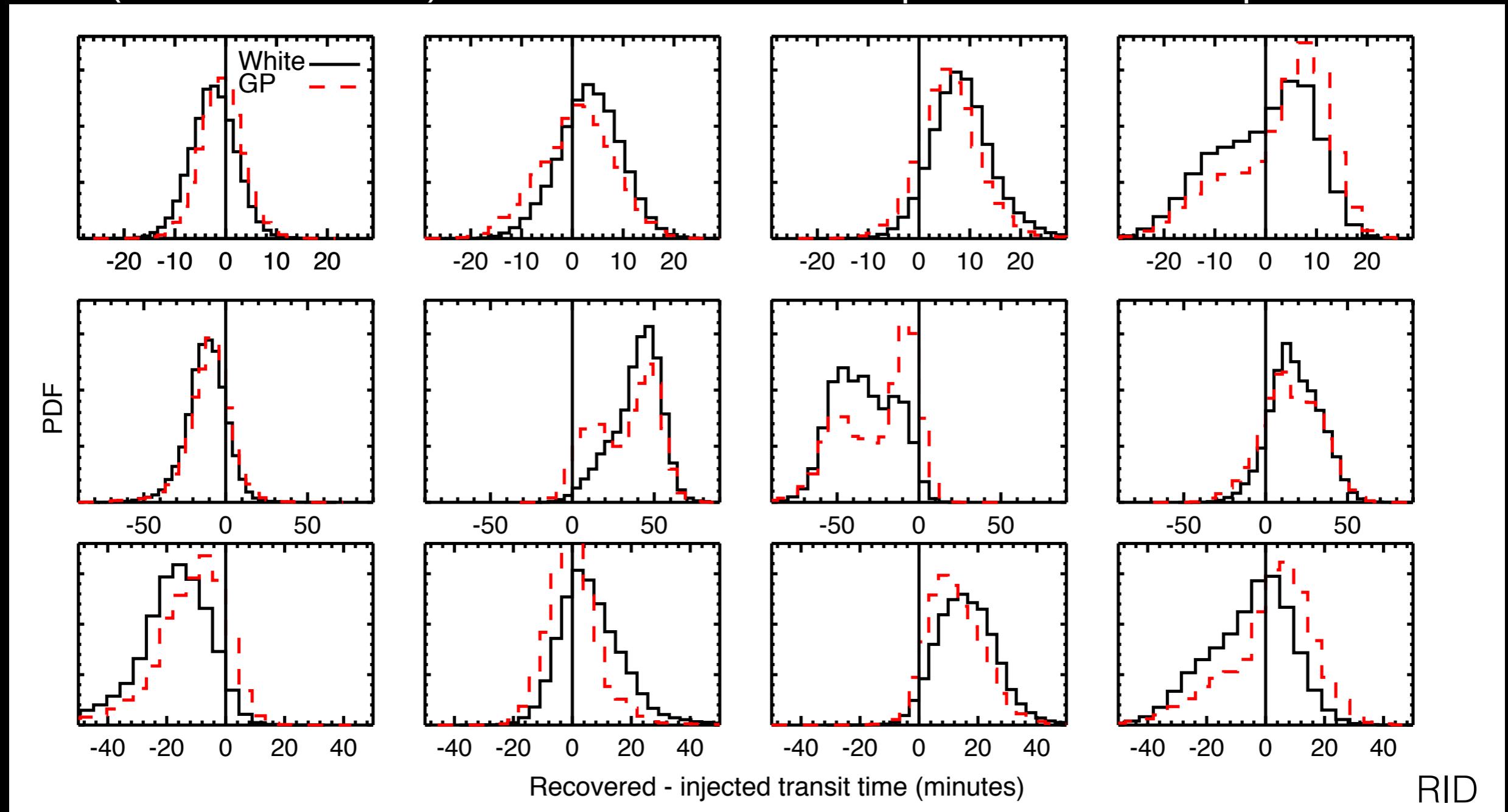


Correlated noise treatment: key questions

- Which stars merit a correlated noise treatment?
- How do we optimize the use of out-of-transit data to infer noise hyperparameters?
- Do wavelet likelihood functions or Gaussian process regression likelihood functions perform better? Which wavelet families and noise power law (wavelets) or kernels (GP regression) is best suited?
- What degree polynomial, if any, should be simultaneously fit to each data chunk?
- How do correlated noise treatments perform on short cadence data? (1 min vs. 30 min cadence)

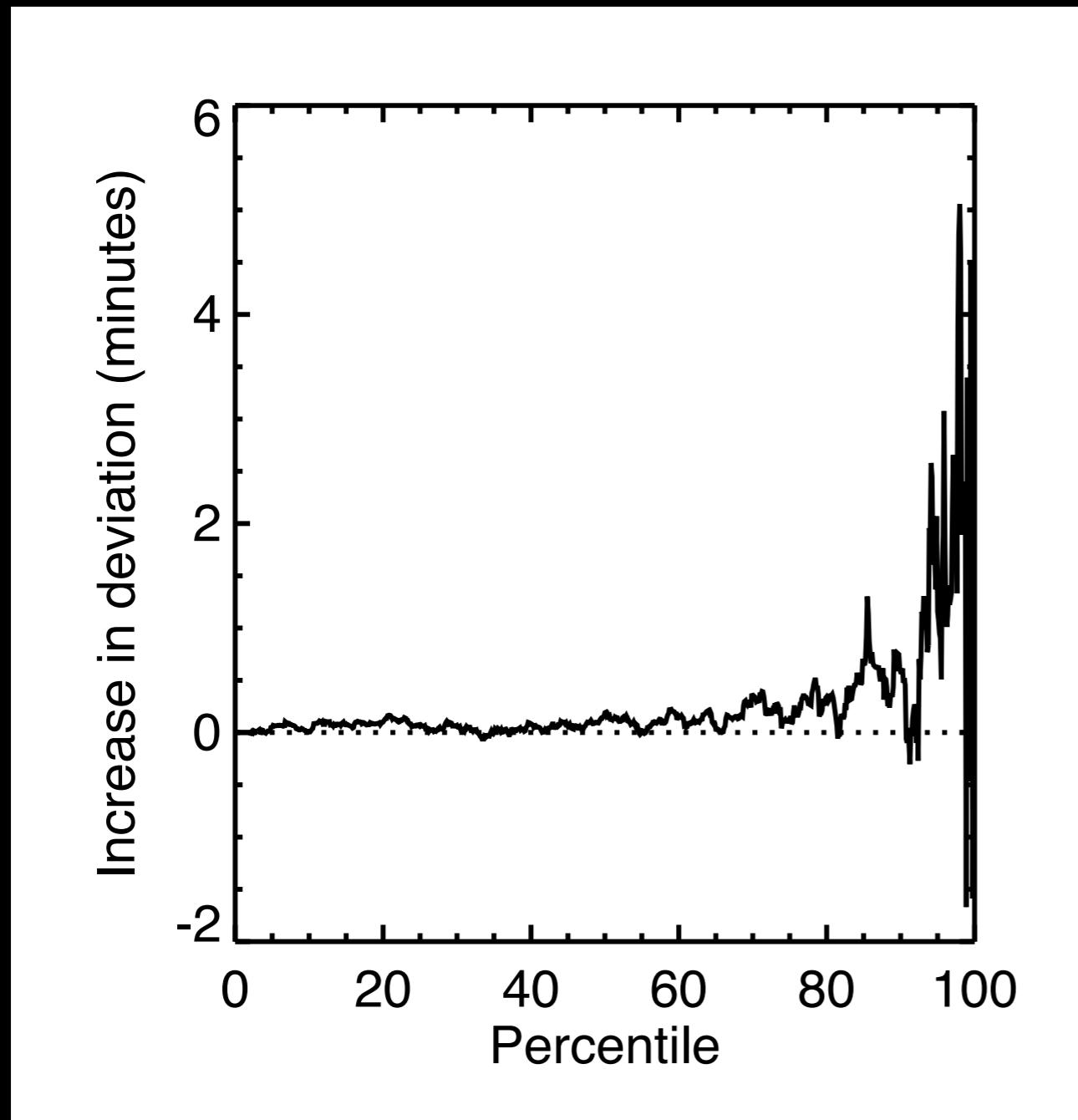
Example transit time posteriors

Better recovery when accounting for correlated noise
(red dashed) and multi-modal posteriors captured



Model 1: Joint modeling of transits + line with white noise likelihood
Model 2: Joint modeling of transits + line with Gaussian process likelihood

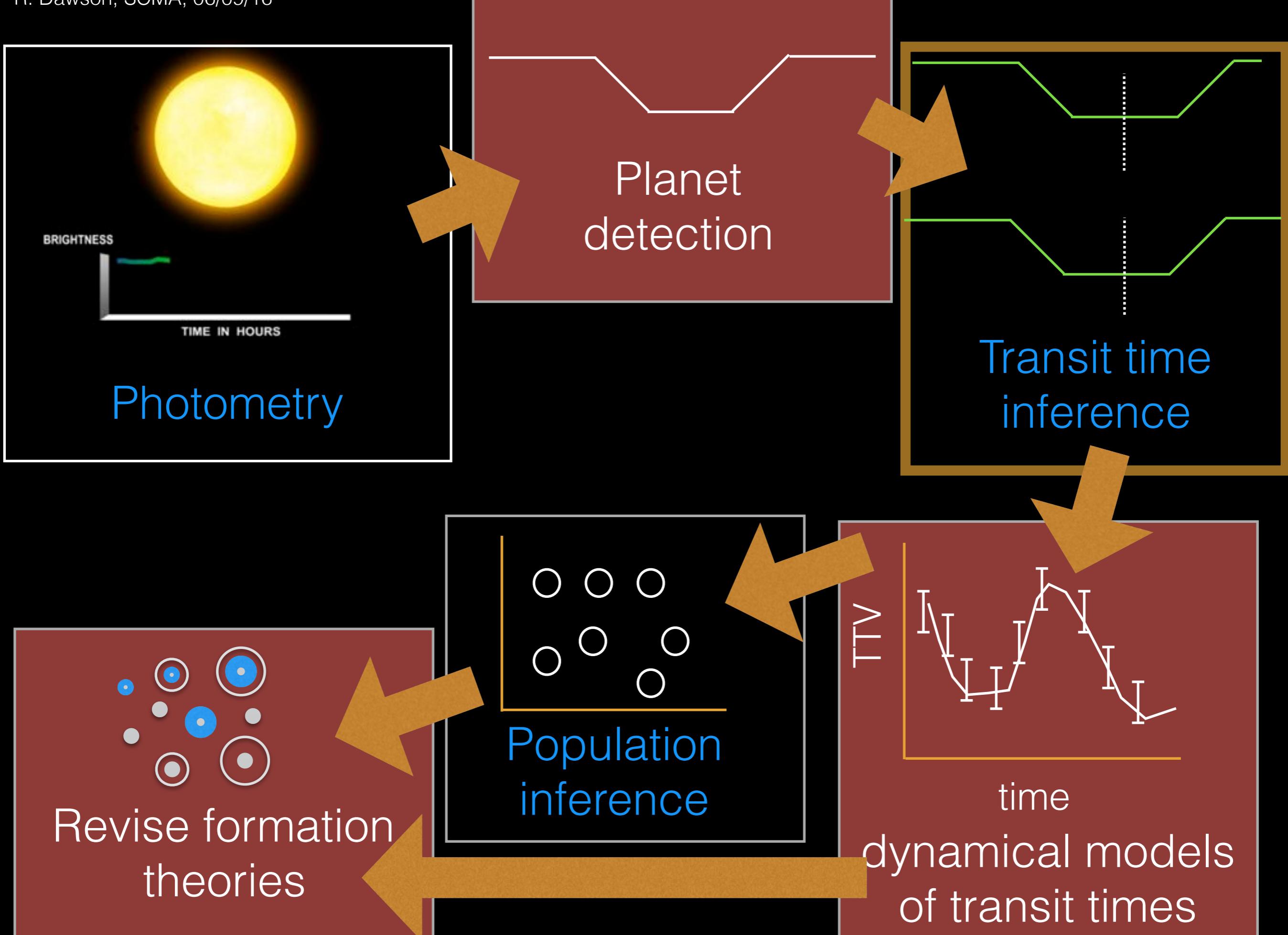
Worse recovery when correlated noise is not accounted for in the likelihood



Based on fits to 50 sets of 16 injected transits for Sun-like star with significant correlated noise

Summary and future work

- Systematic study of correlated noise treatment for inferring transit times is underway; will only be relevant for subset of stars
- Correlated noise treatment needs to be assessed for its impact on other key transit observables, e.g., depth, duration
- Correlated noise is an even more severe problem for radial velocity method of planet detection and characterization, including interplay of noise and aliasing due to gaps in time sampling

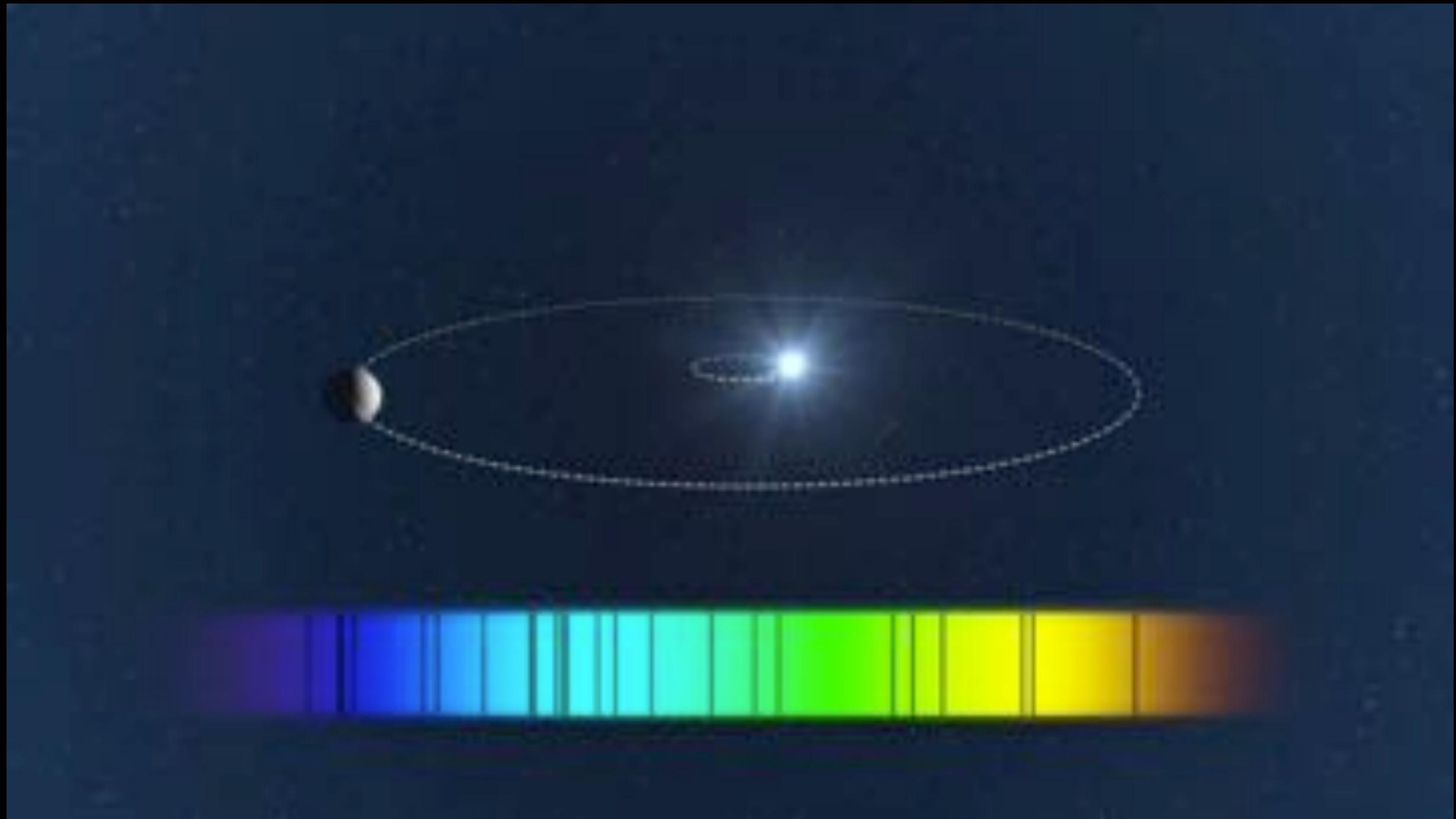
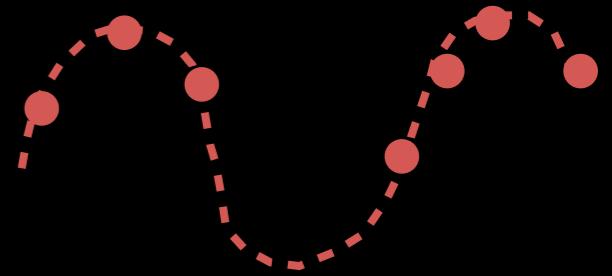


Summary and future work

- Systematic study of correlated noise treatment for inferring transit times is underway; will only be relevant for subset of stars
- Correlated noise treatment needs to be assessed for its impact on other key transit observables, e.g., depth, duration
- Correlated noise is an even more severe problem for radial velocity method of planet detection and characterization, including interplay of noise and aliasing due to gaps in time sampling

Extra slides

The radial-velocity technique



Earth twin: 10 cm/s

Image Credit: ESO/L. Calçada

Transit vs. radial- velocity challenges

Transit (mostly space)

Planet duty cycle

Low (0.3% for Earth twin)

A priori planet
probability

Low (~few percent)

Datapoints

~100,000 or more

Signal repetition

Detectable changes in
period, duration

Time sampling

Even, continuous

Radial-velocity (ground)

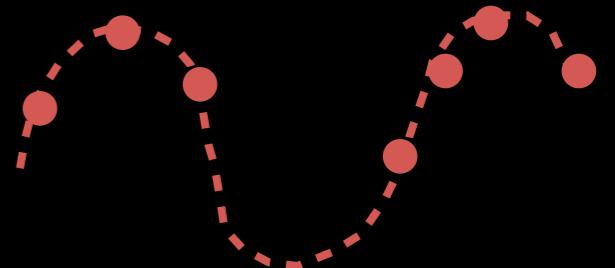
100% (but 0% for other
line diagnostics)

High ($\geq\sim 50\%$)

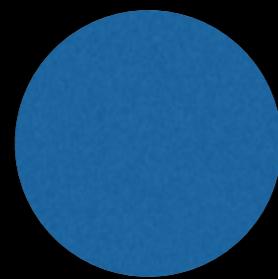
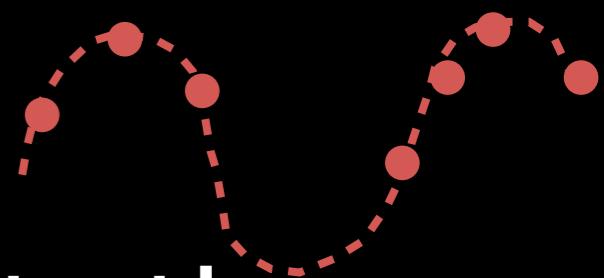
~100

Undetectable for
most planets

Uneven, gaps

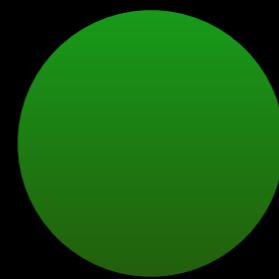


Time sampling complicates RV interpretation



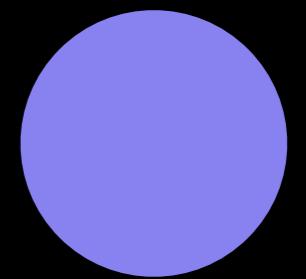
55 Cnc e

One of first
mini-Neptunes
discovered



GJ 581d

Habitable zone
super-Earth

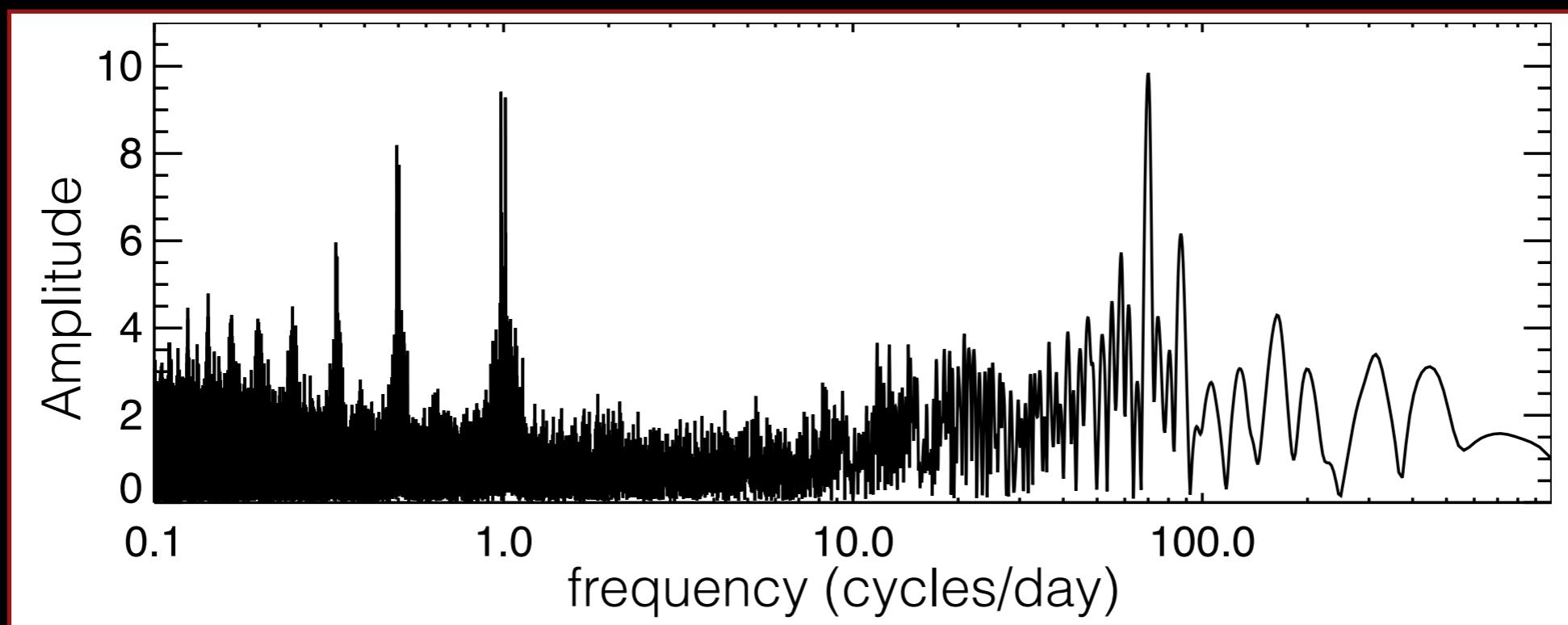
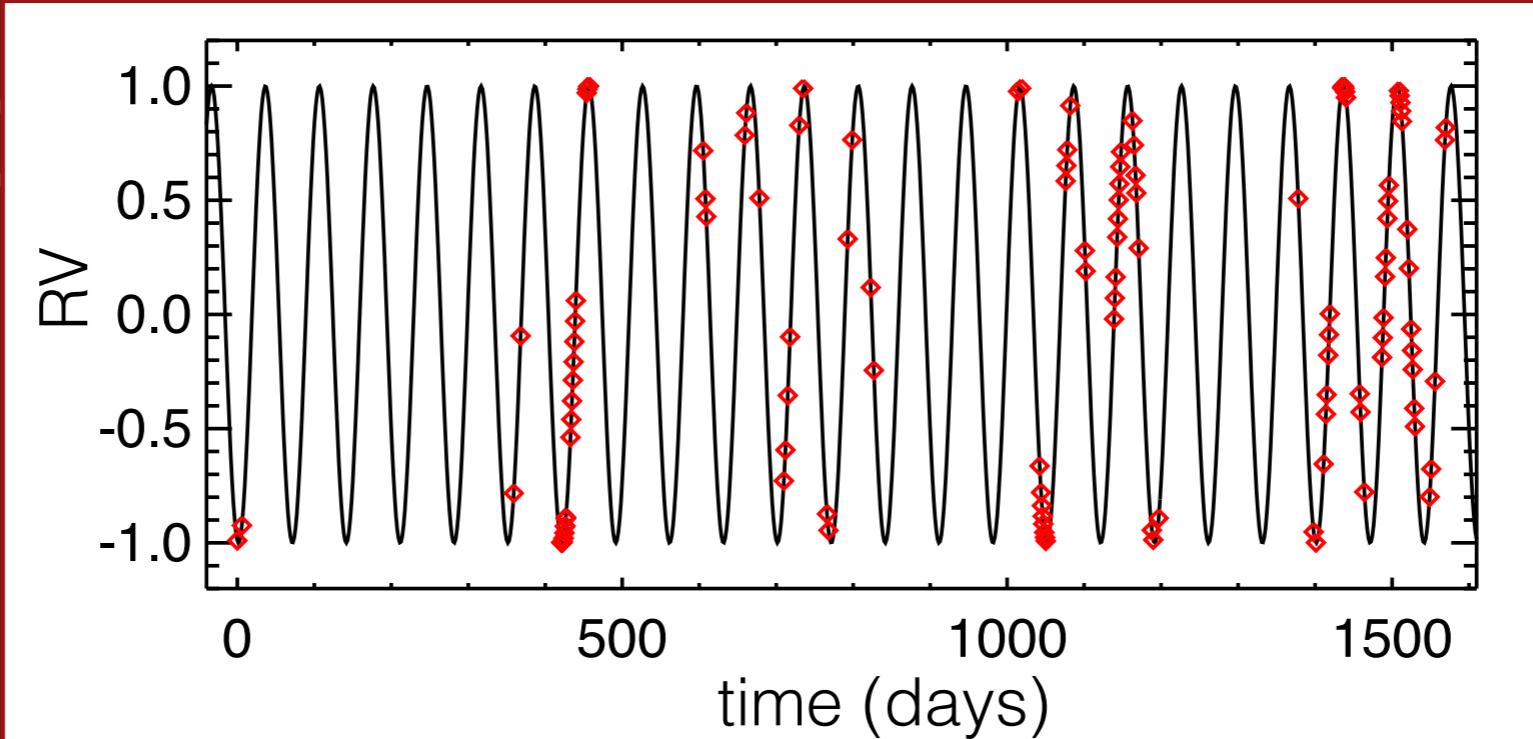


Alpha Cen b

Orbits nearby
star

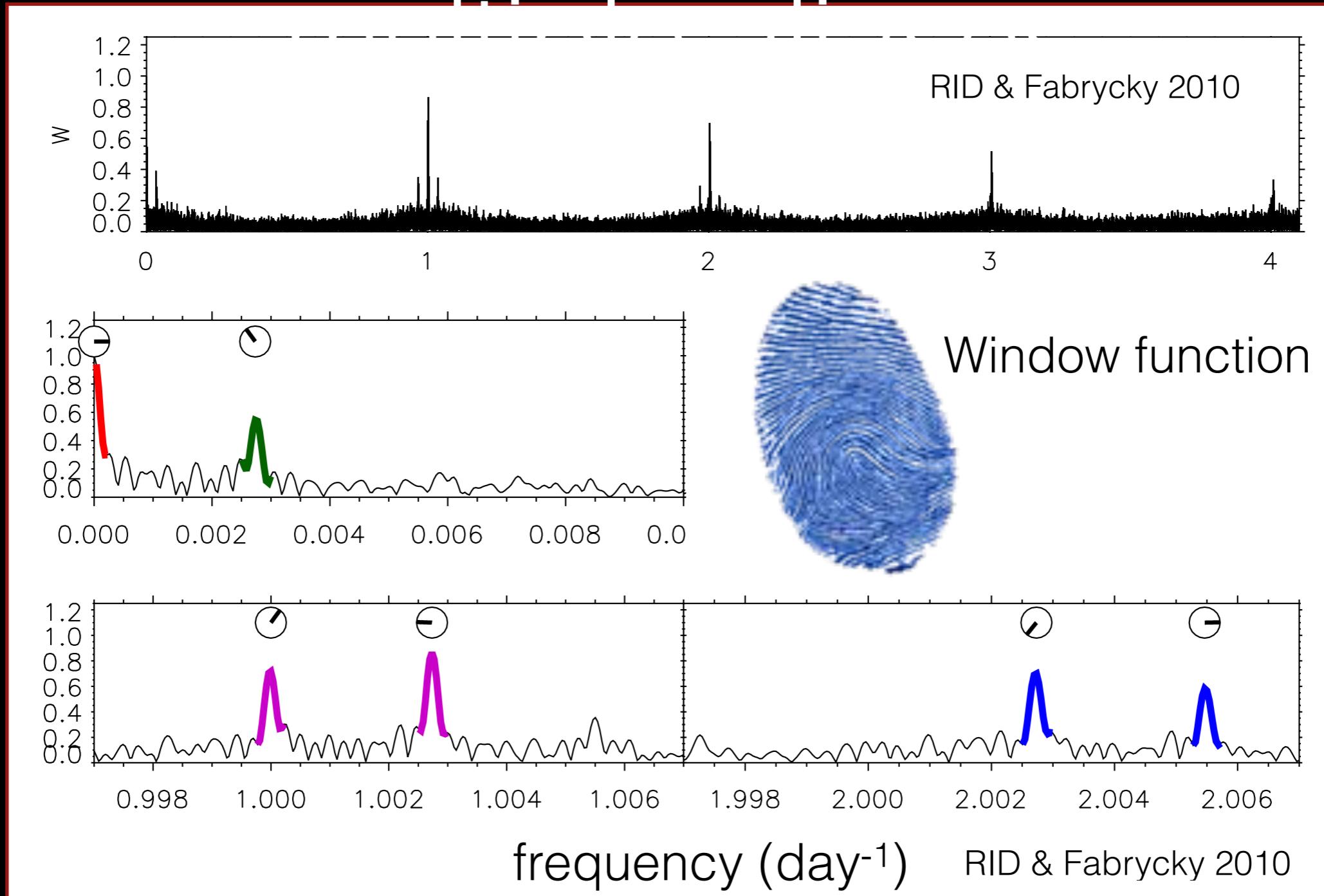
Radial velocity sampling

Noise-free
sinusoid with
Gl 581 HARPS
sampling



$$f_{\text{alias}} = | f_{\text{true}} \pm f_{\text{sample}} |$$

Use the fingerprint of

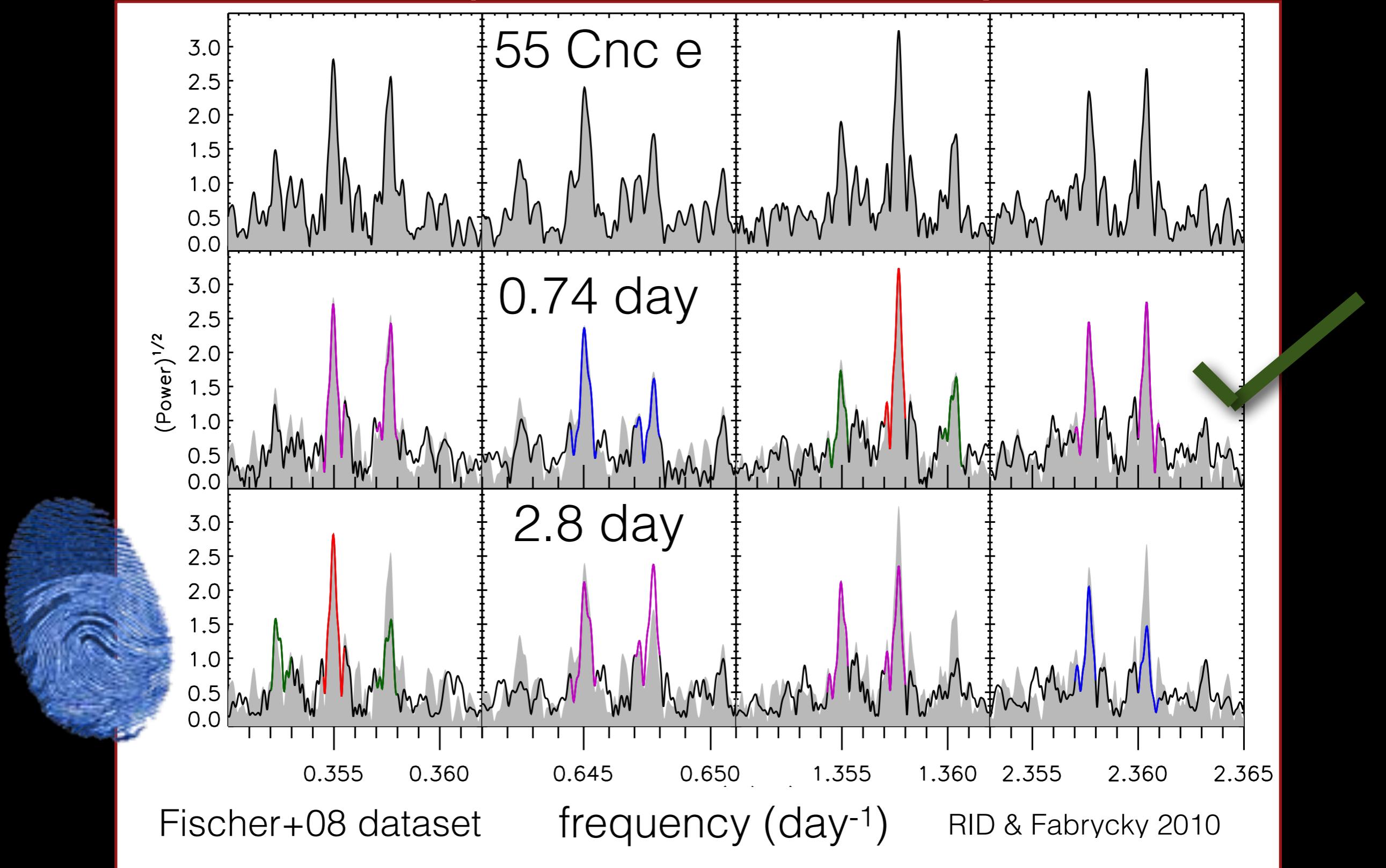


$$f_{\text{alias}} = | f_{\text{true}} \pm f_{\text{sample}} |$$

Fischer+08 dataset

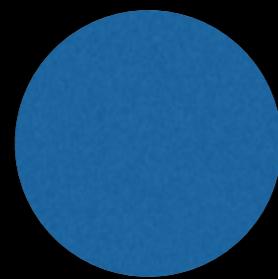
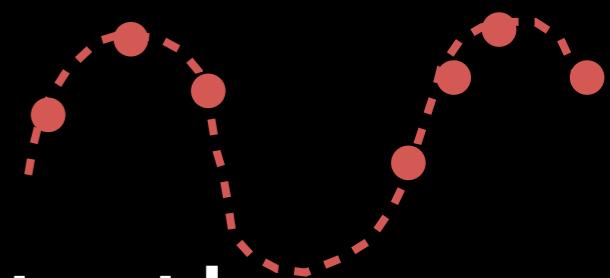
55 Cnc

A revised, ultra-short period



$$f_{\text{alias}} = |f_{\text{true}} \pm f_{\text{sample}}|$$

Time sampling complicates RV interpretation

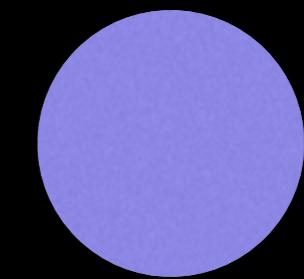
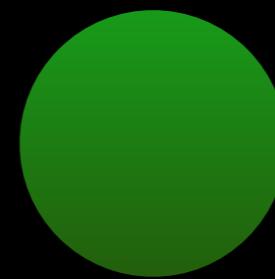


55 Cnc e

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GJ 581d

Habitable zone
super-Earth



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Orbits nearby
star

Gl 581 d: alias ambiguity

Udry: 07:

 $P = 84$ days

Mayor+ 09:

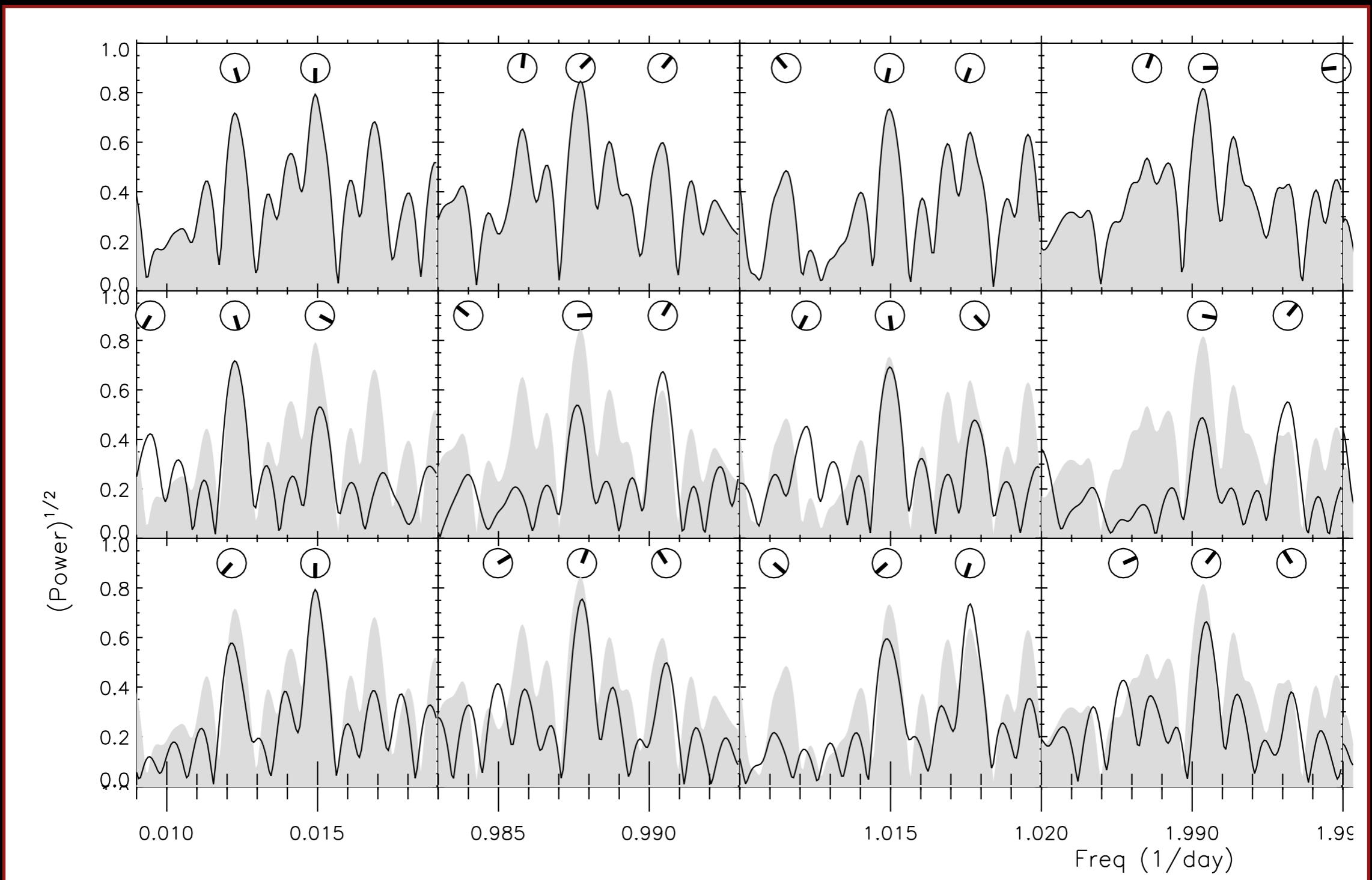
 $P = 67$ days

RID & Fabrycky 10:

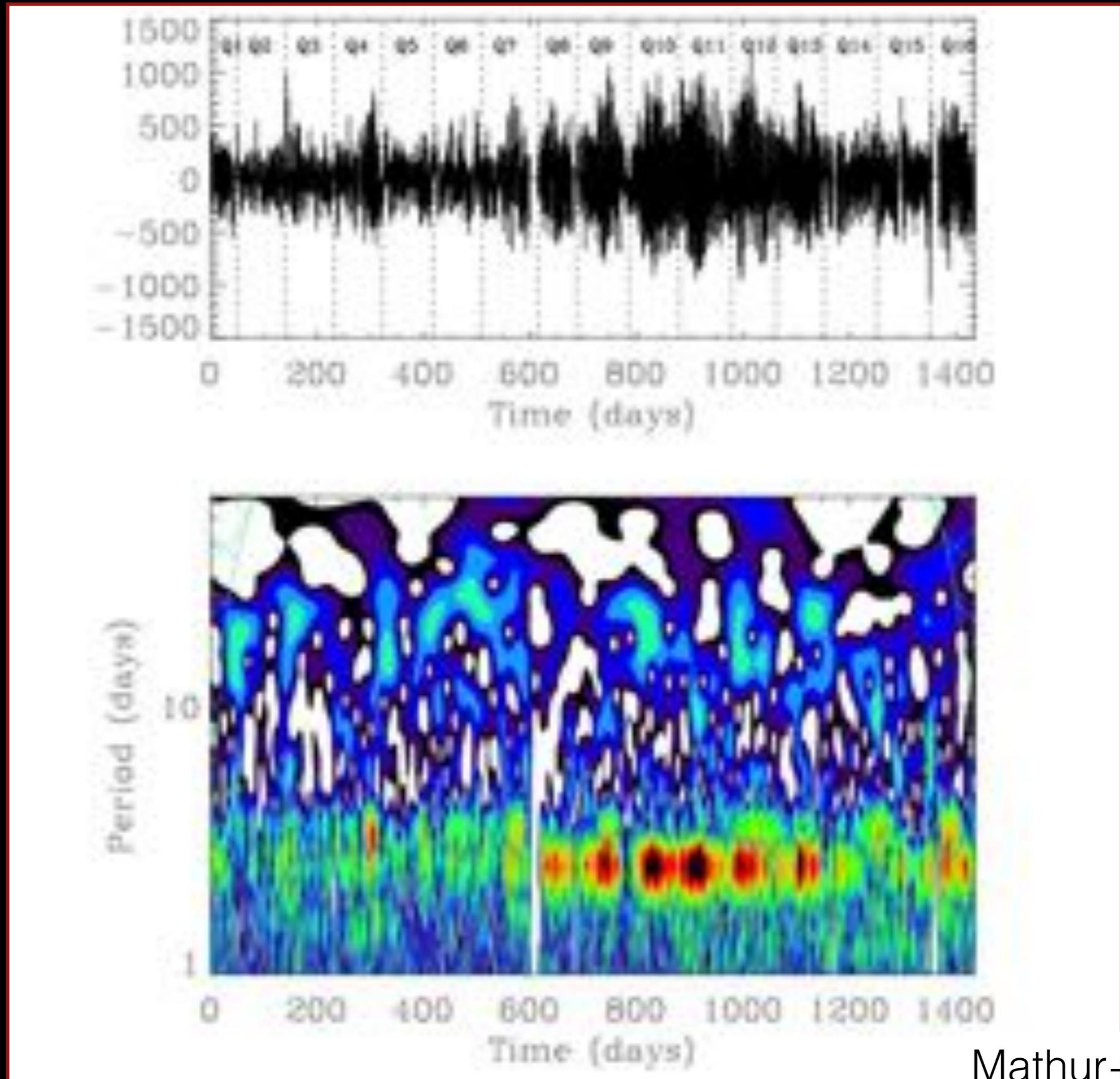
 $P = ?$

Robertson+ 14:

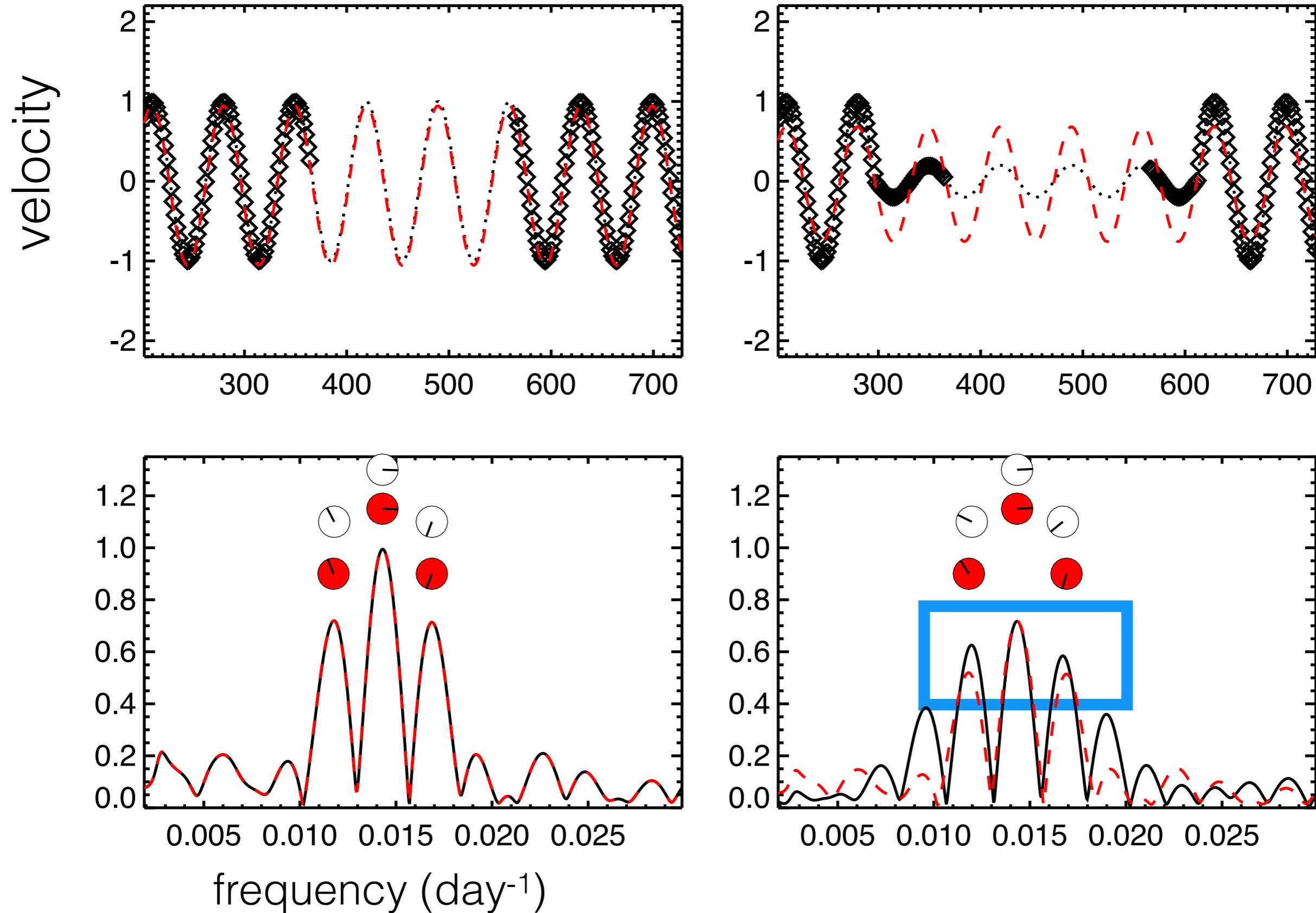
Stellar activity



Stellar activity: a stochastic, quasi-periodic signal

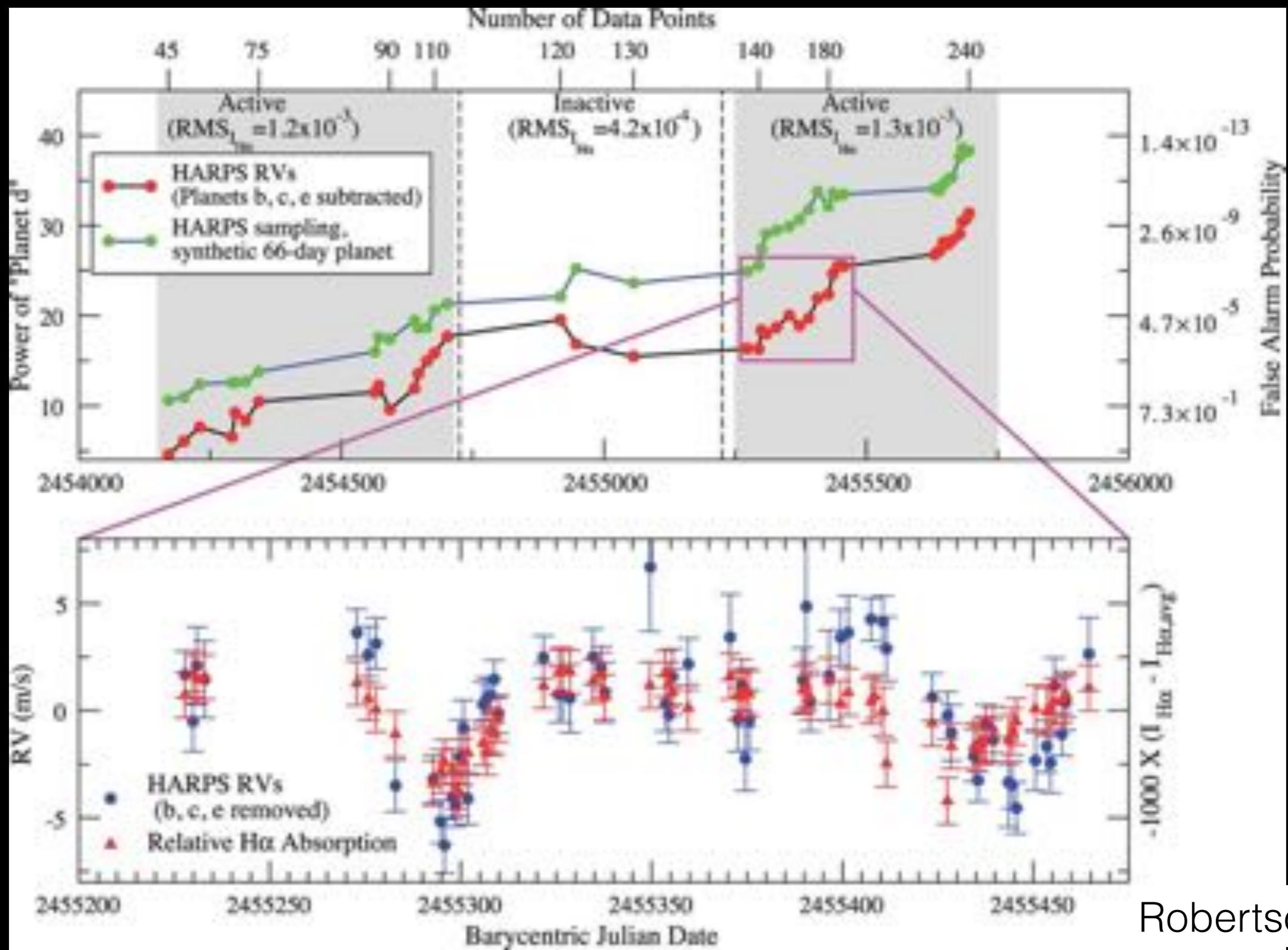


Aliasing and activity cycles

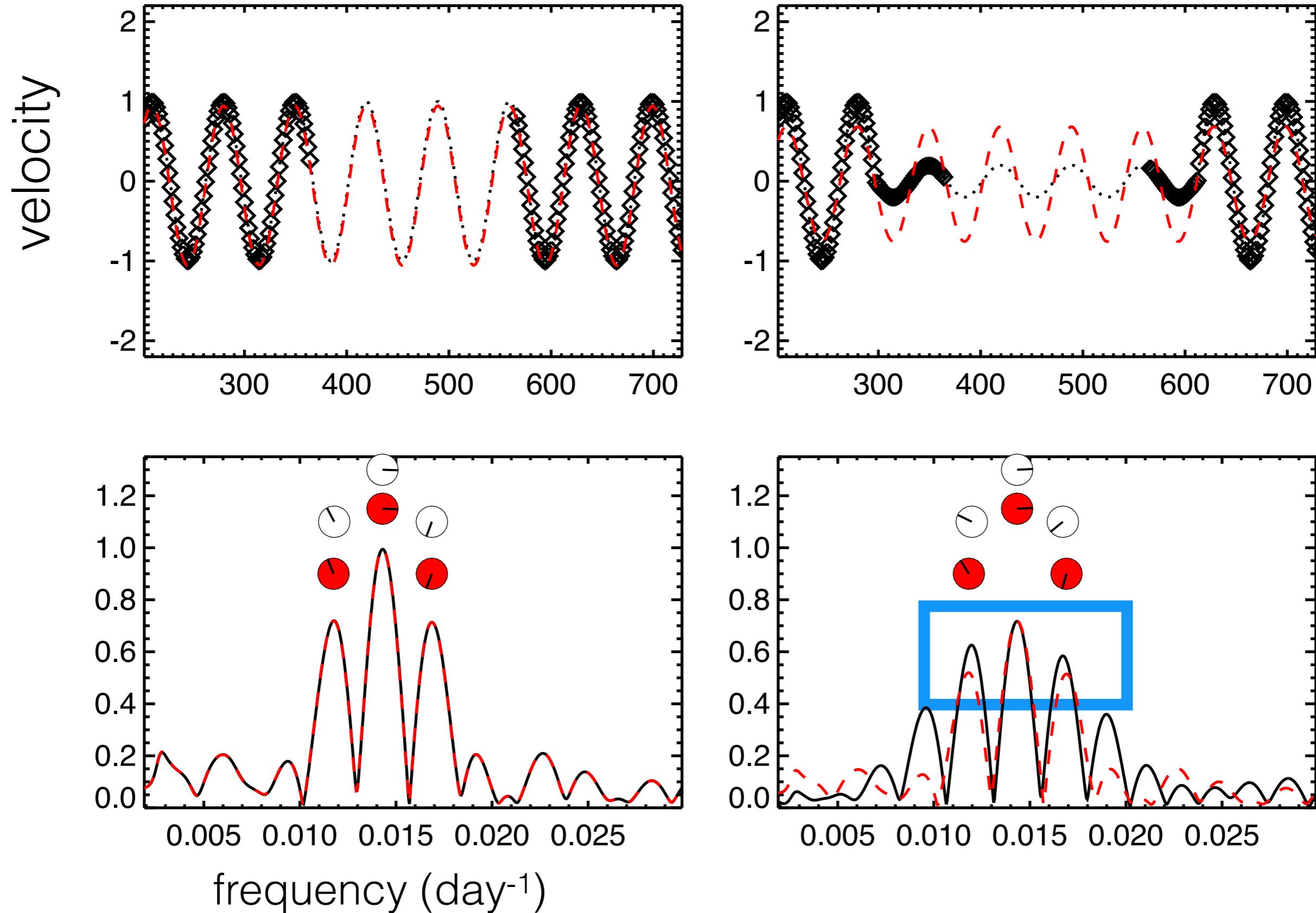


Less time sampling during inactive cycle

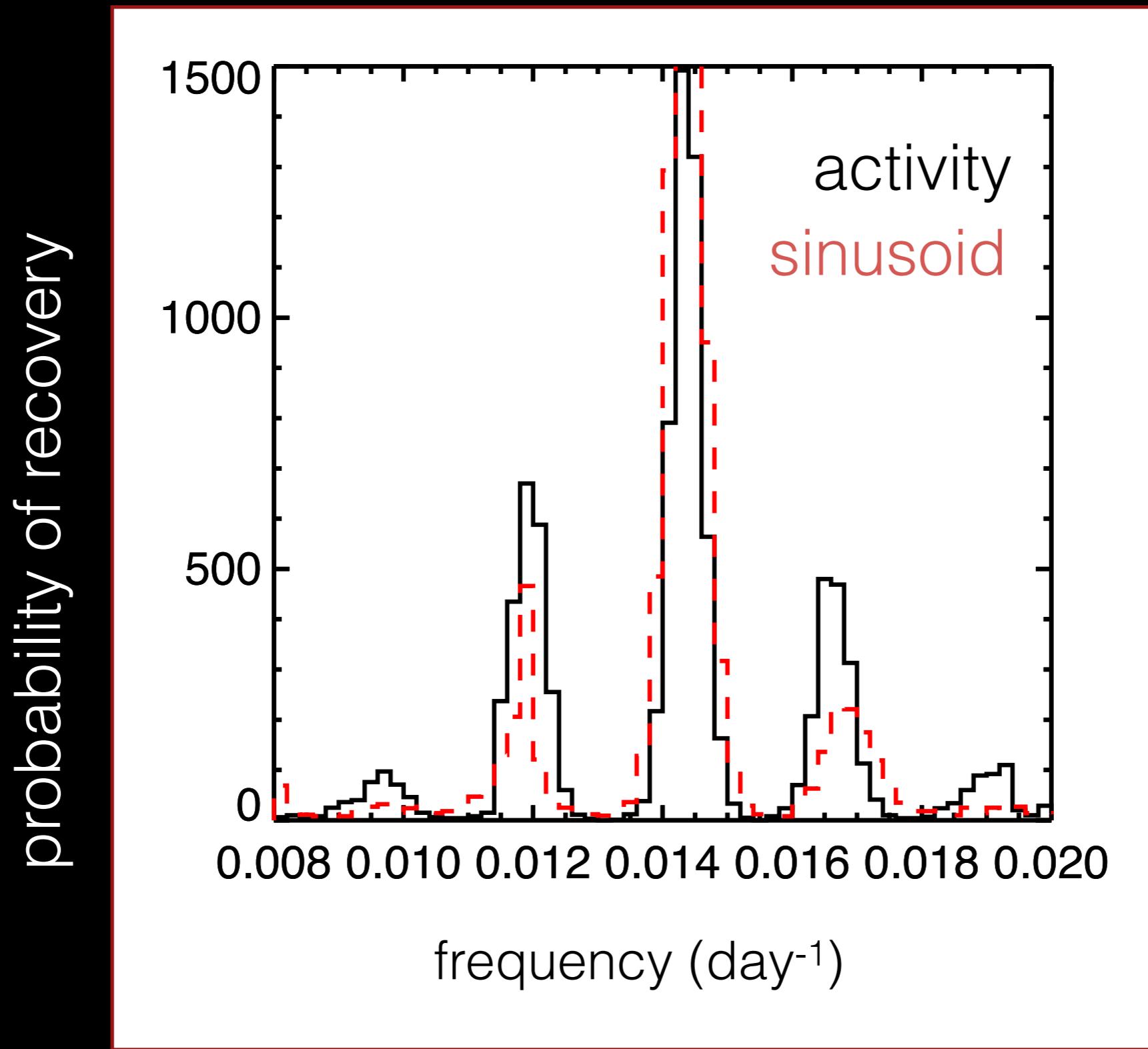
Gl 581



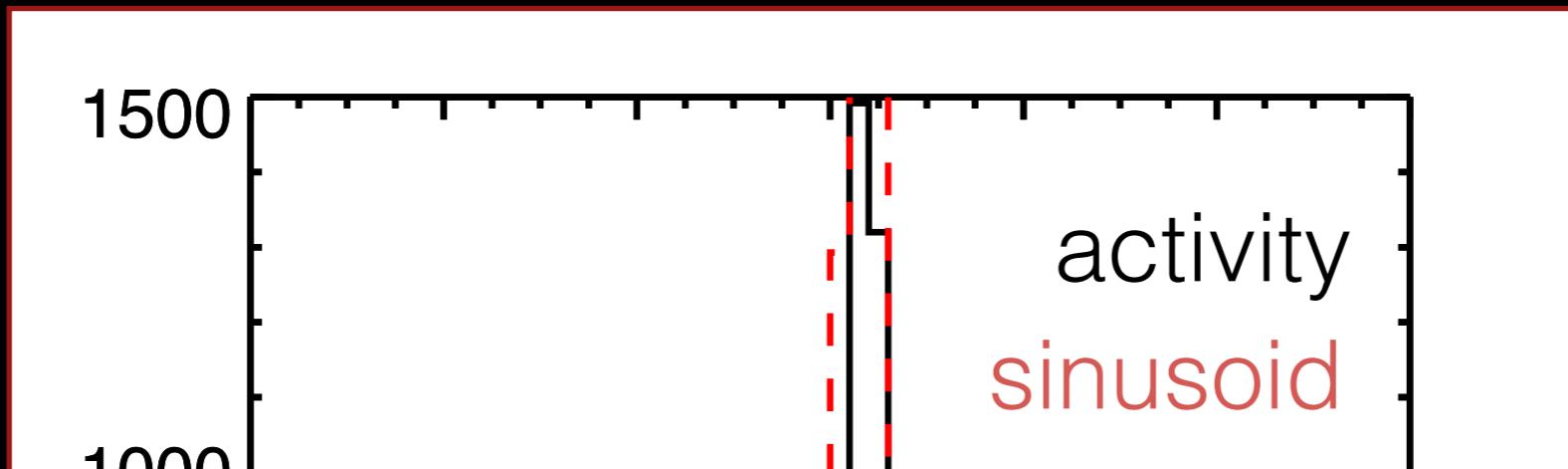
Aliasing and activity cycles



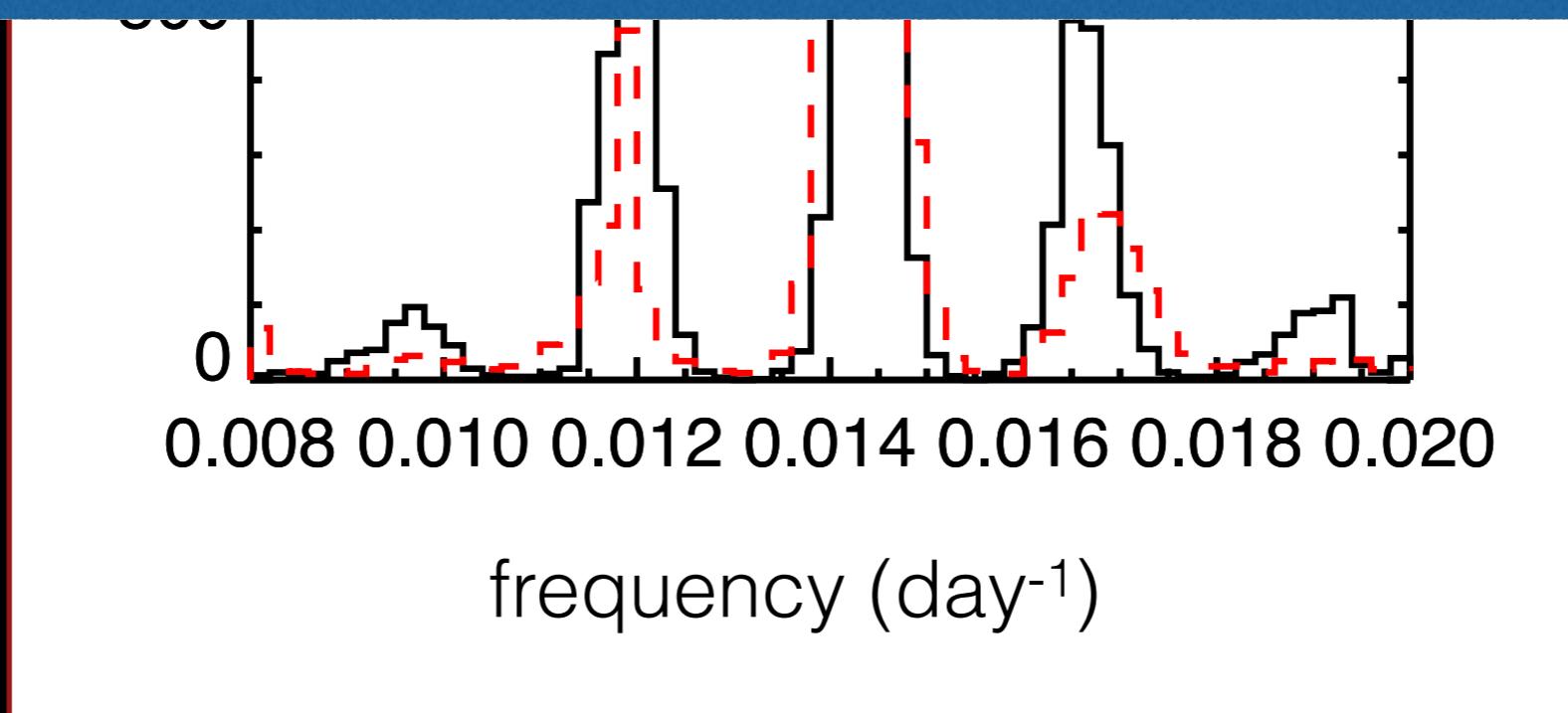
Stellar activity signal experiences extra aliasing when activity is low during sampling gap



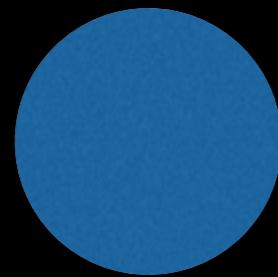
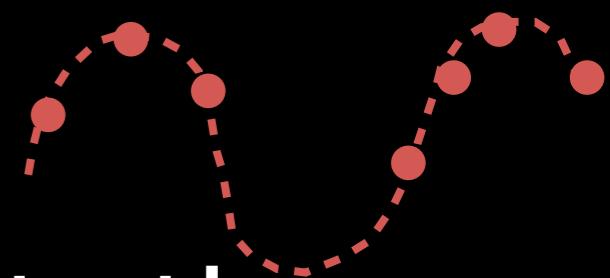
Stellar activity signal experiences extra aliasing when activity is low during sampling gap



Aliasing ambiguities may tip us off about stellar activity

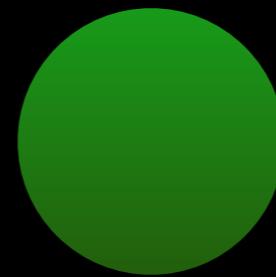


Time sampling complicates RV interpretation



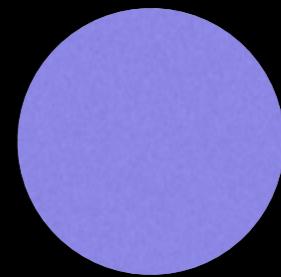
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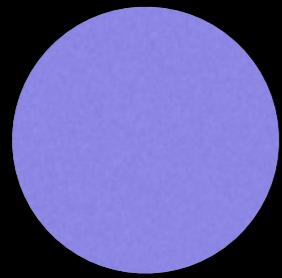
Habitable zone
super-Earth



Alpha Cen b

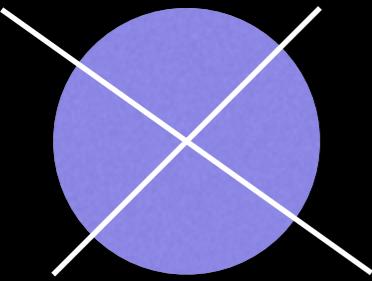
Orbits nearby
star

Alpha Cen b: the danger of pre-whitening



Dumusque+ 12:

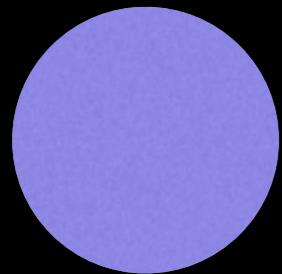
- 1) detrend
- 2) fit planet parameters
- 3) check for aliasing



Rajpaul+15, 16

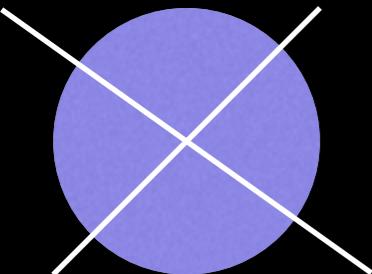
- 1) notice planet frequency in the window function
- 2) account for stellar activity simultaneous with orbit fitting with Gaussian processes

Alpha Cen b: the danger of pre-whitening



Dumusque+ 12:

- 1) detrend
- 2) fit planet parameters
- 3) check for aliasing



Rajpaul+15, 16

- 1) notice **planet frequency in the window function**
- 2) account for stellar activity simultaneous with orbit fitting with Gaussian processes

$$f_{\text{alias}} = | f_{\text{true}} \pm f_{\text{sample}} |$$

long-term activity!