

DE LA RECHERCHE À L'INDUSTRIE

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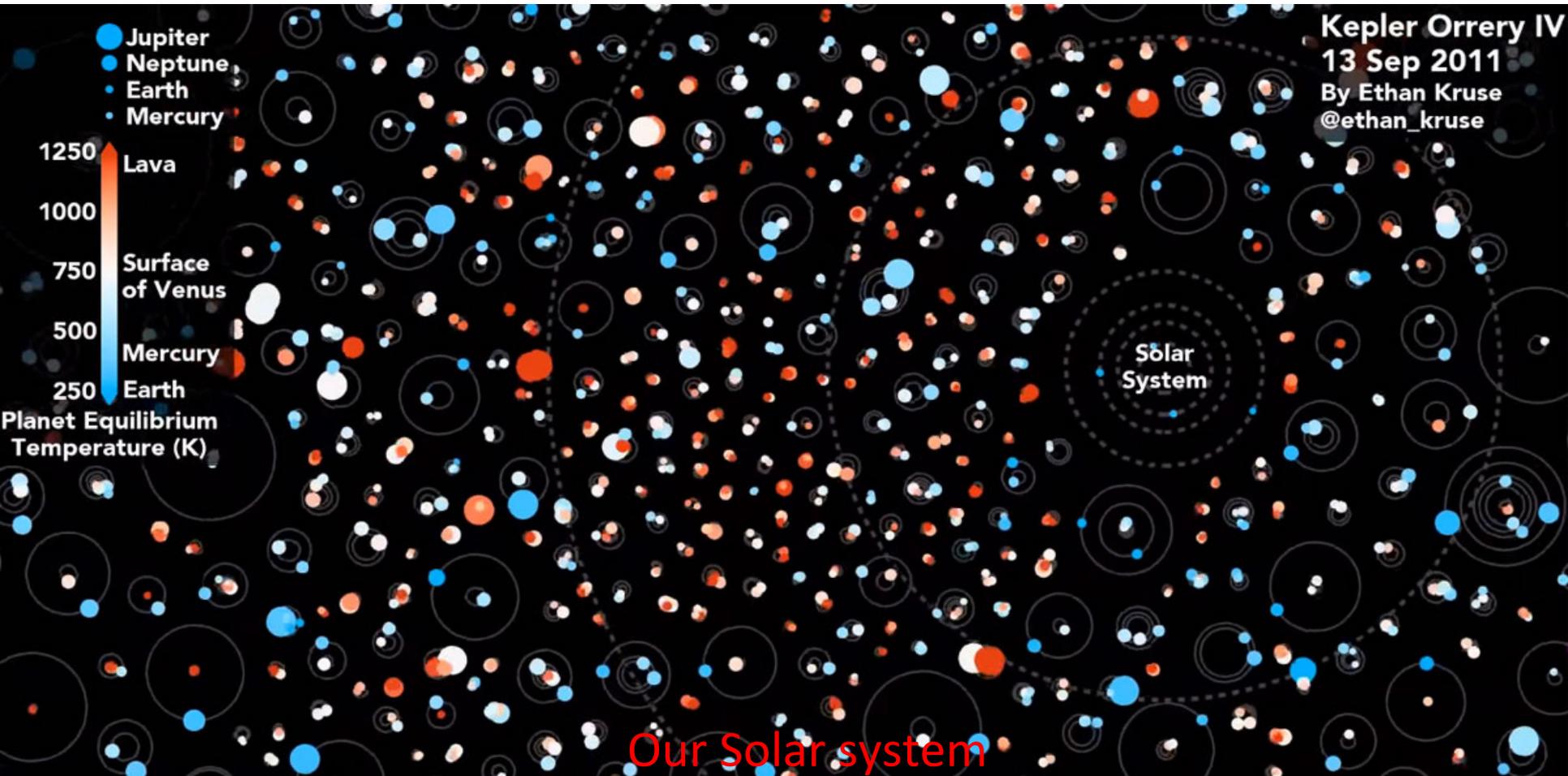
# Data detrending techniques for high-precision exoplanet spectroscopy

Giuseppe Morello  
Bât 703 – P<sup>e</sup> 37D

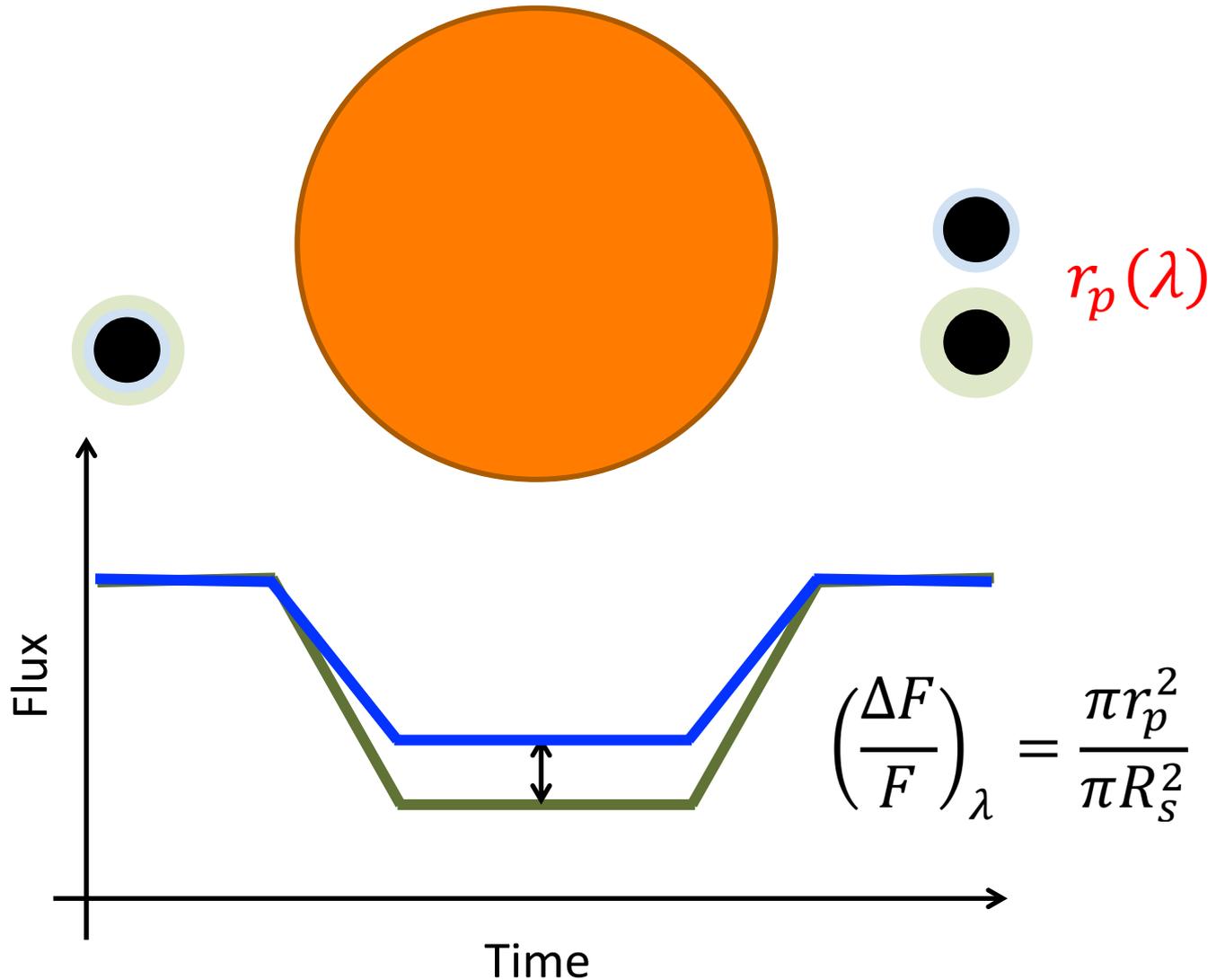
# Scientific career

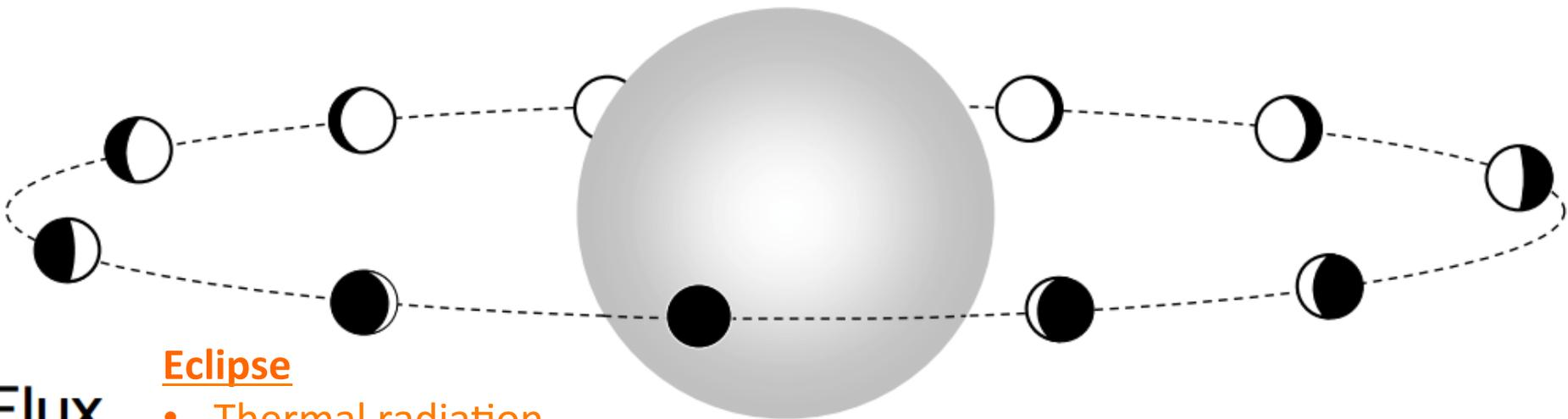
- PhD, UCL , Oct. 2013 – Oct. 2016  
I. D. Howarth & G. Tinetti
- IPAC Fellowship, Caltech, Jan.-Jul. 2016  
P. W. Morris & S. D. Van Dyk  
(automated classification and clustering: kNN, SVM, SOM, ...)
- Post-doc, UCL, Nov. 2016 – Oct. 2017  
G. Tinetti
- Post-doc, CEA Saclay, Nov. 2017 – ongoing  
Pierre-Olivier Lagage

# Exoplanets variety



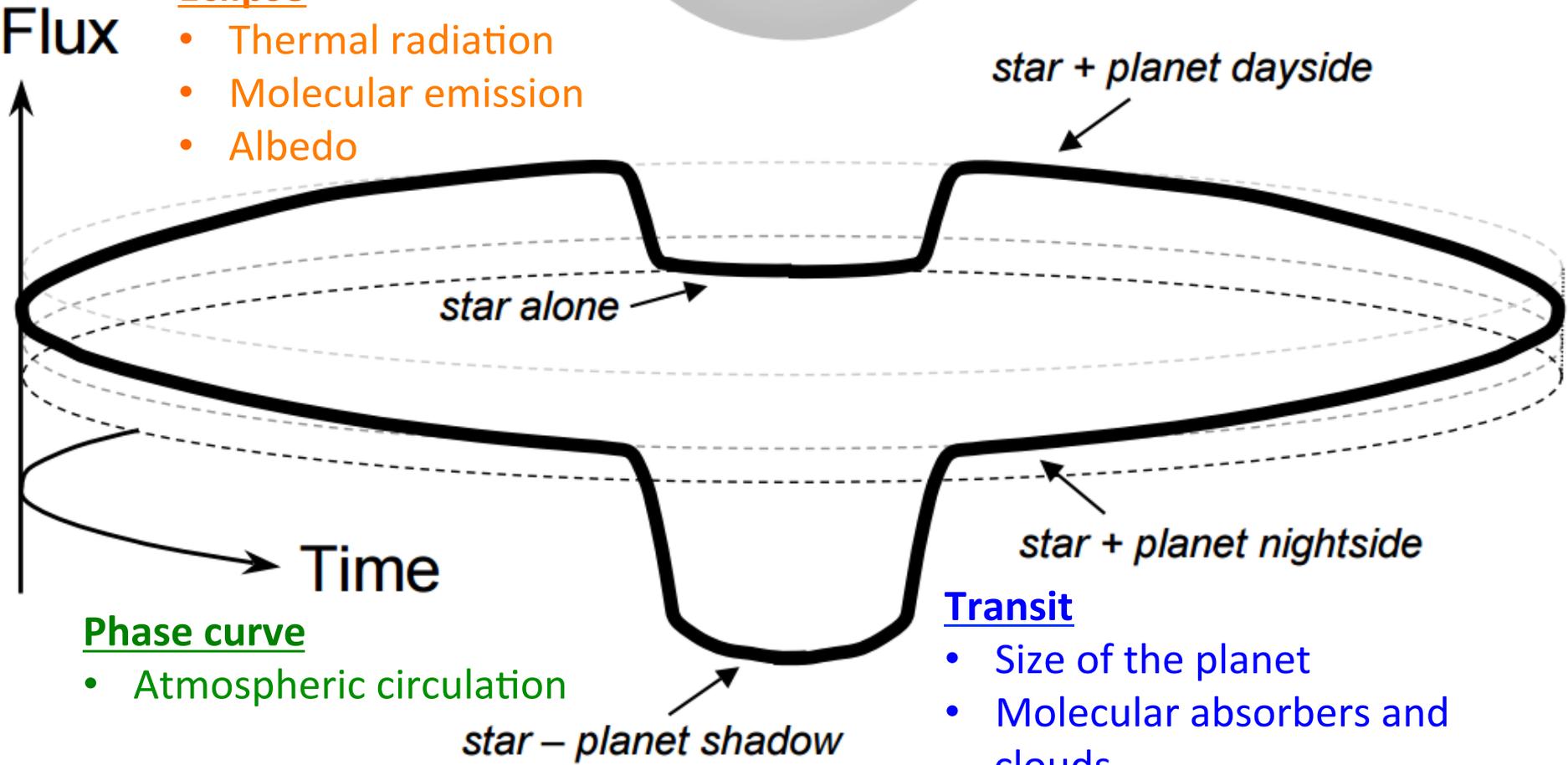
# Transit of an exoplanet with atmosphere





**Eclipse**

- Thermal radiation
- Molecular emission
- Albedo



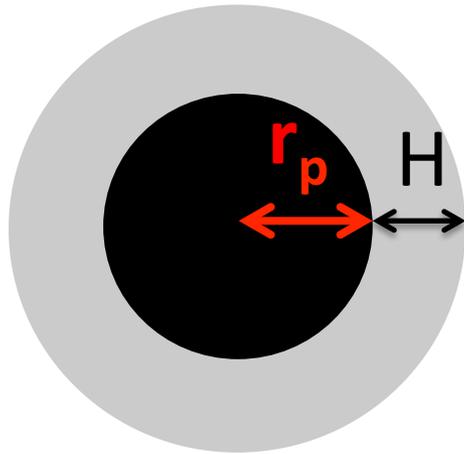
**Phase curve**

- Atmospheric circulation

**Transit**

- Size of the planet
- Molecular absorbers and clouds

# Atmospheric absorption



$$\left(\frac{\Delta F}{F}\right)_{atm} \approx \frac{2\pi r_p H}{\pi R_s^2} \kappa(\lambda)$$

$$H = \frac{KT}{\mu g}$$

Required precision of 10 – 100 parts per million (ppm)!

planet	$r_p$ (km)	$g$ ( $m/s^2$ )	$T$ (K)	$\mu$	$h$ (km)	$\frac{\Delta F_{atm}}{F}$ ( $10^{-4}$ , $\mathfrak{K} = 1$ )
Earth	6400	9,8	300	28 u	9	0,0024
Jupiter	71500	25	140	2 u	23	0,069
HD189733b	90000	18	570	2 u	131	0,84

# Technical and astrophysical challenges

1

Removing the instrument systematics

(Spitzer, Hubble  $\sim 100 - 1000$  ppm)

2

Accurate modeling the astrophysical scenario:

stellar limb-darkening, activity, etc. ( $\sim 100 - 1000$  ppm)

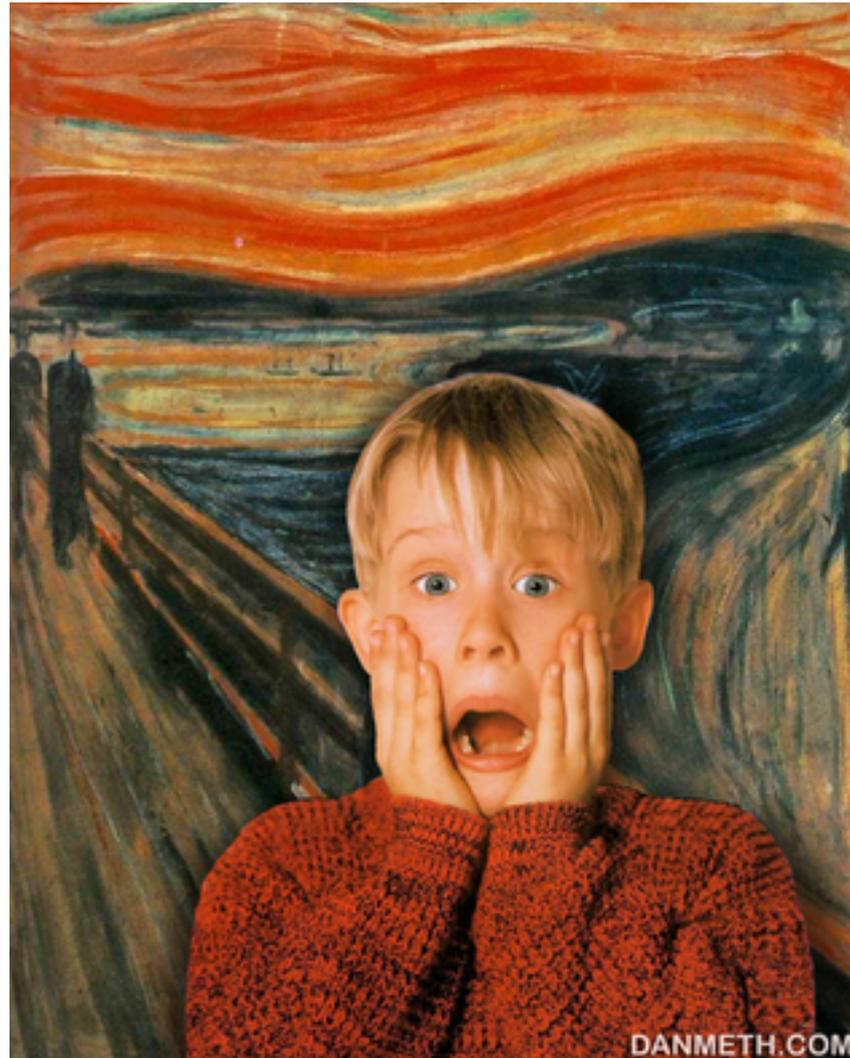
3

Interpreting the atmospheric spectrum:

many chemical species, parameter degeneracies, low signal-to-noise data, sparse wavelength coverage, etc.

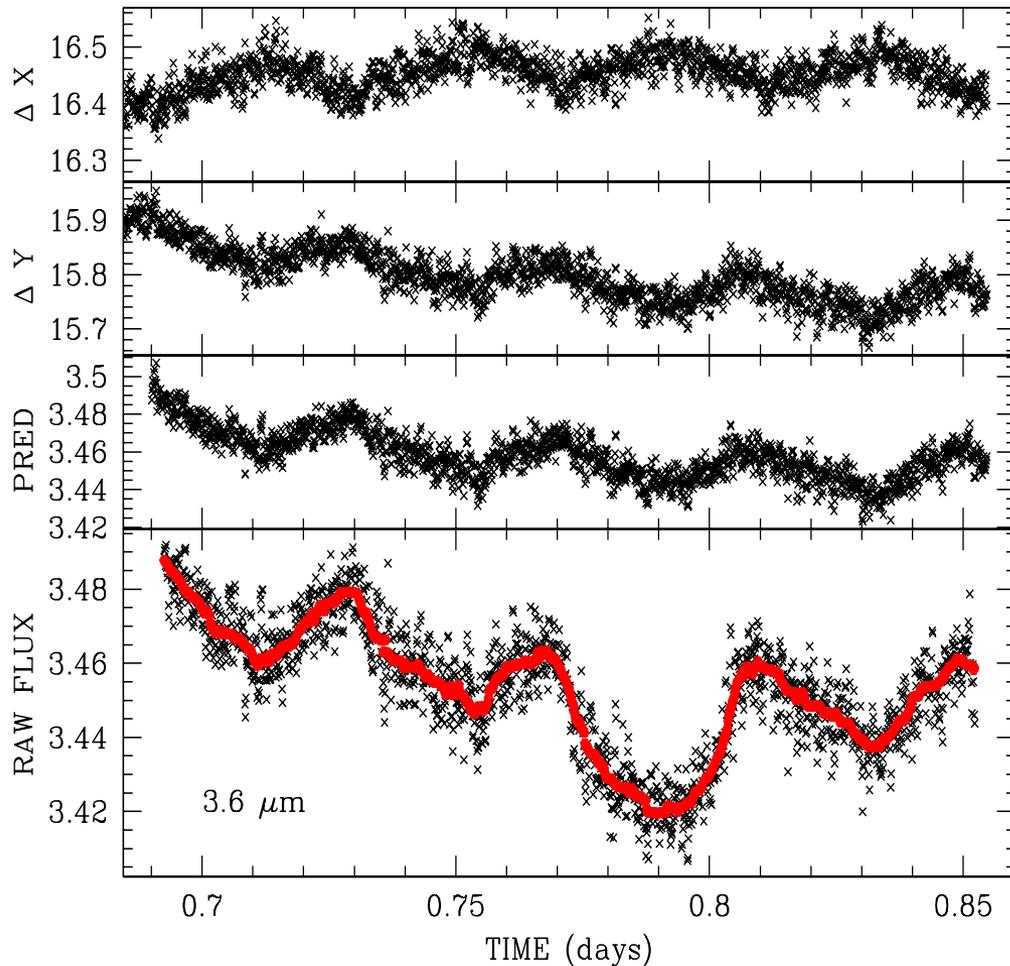
1

# Removing the instrument systematics (Spitzer, Hubble $\sim 100 - 1000$ ppm)



# 1

# Data detrending: parametric



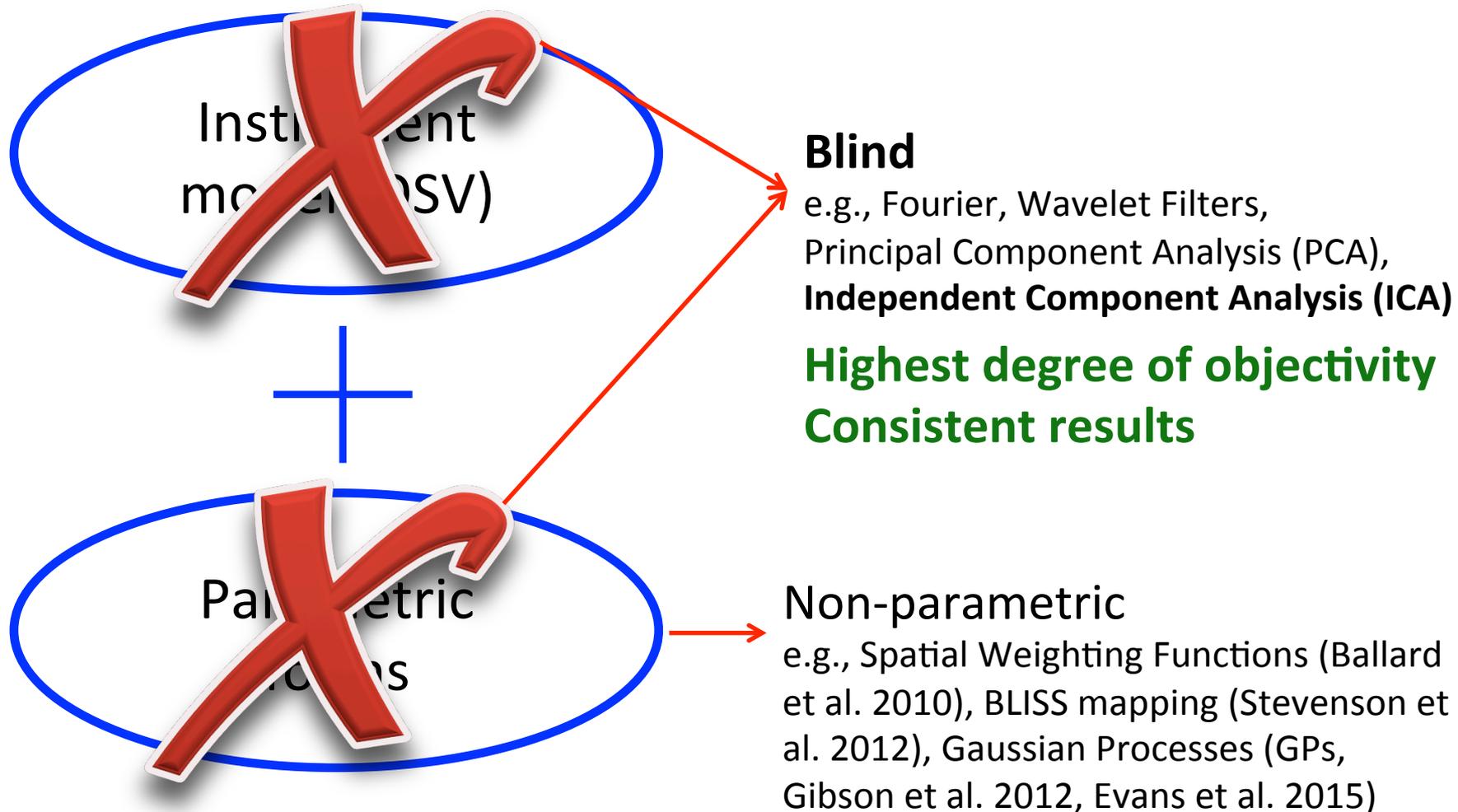
Example: Spitzer/IRAC -> division by a polynomial function of centroid coordinates.

1. Optical State Vector (OSV): detector temperature, inclination, point spread function, etc.
2. Measured flux is correlated with OSV parameters
3. Parametric correction based on those parameters.

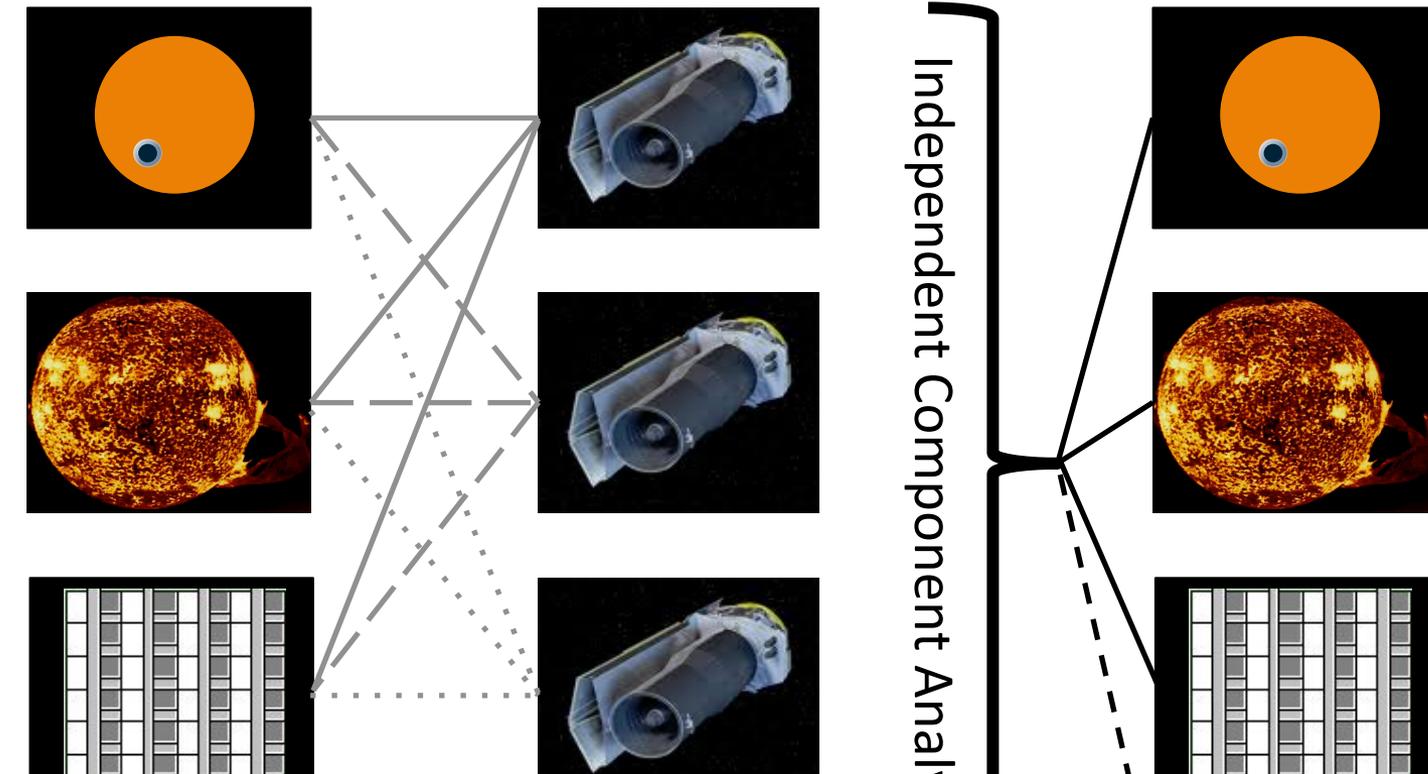


**Some controversial results in the literature** (corrections beyond the well-known instrument response)

# Data detrending: blind



# 1 Independent Component Analysis



observations                      signals

$$\mathbf{X} = \mathbf{A}\mathbf{S}$$

mixing matrix

→  
Minimize  
mutual  
information

$$\mathbf{S} = \mathbf{A}^{-1}\mathbf{X}$$

UNKNOWN

# 1

## ICA: statistics (1)

$$H(\mathbf{y}) = - \sum_k p(\mathbf{y}_k) \log p(\mathbf{y}_k) \quad \text{Shannon entropy}$$

It is the statistical measure of uncertainty associated with a random variable.

$$I(y_1, y_2, \dots, y_n) = \sum_{i=1}^n H(y_i) - H(\mathbf{y}) \quad \text{mutual information}$$

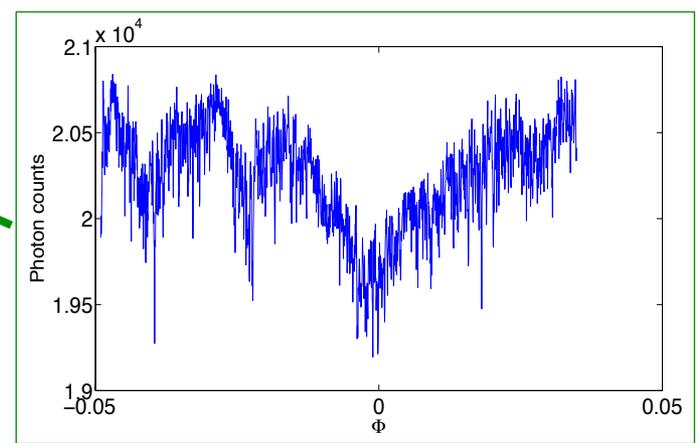
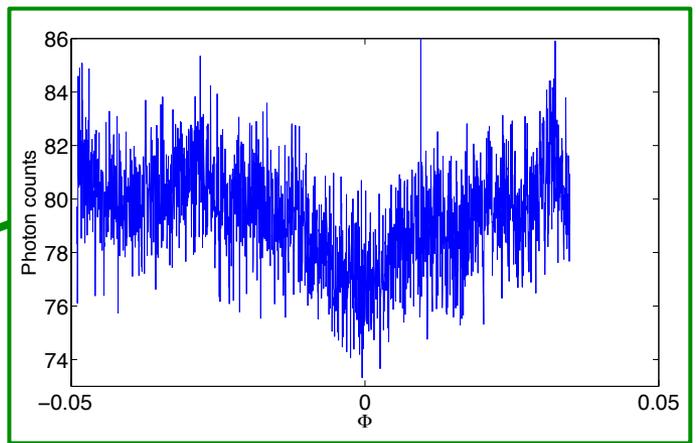
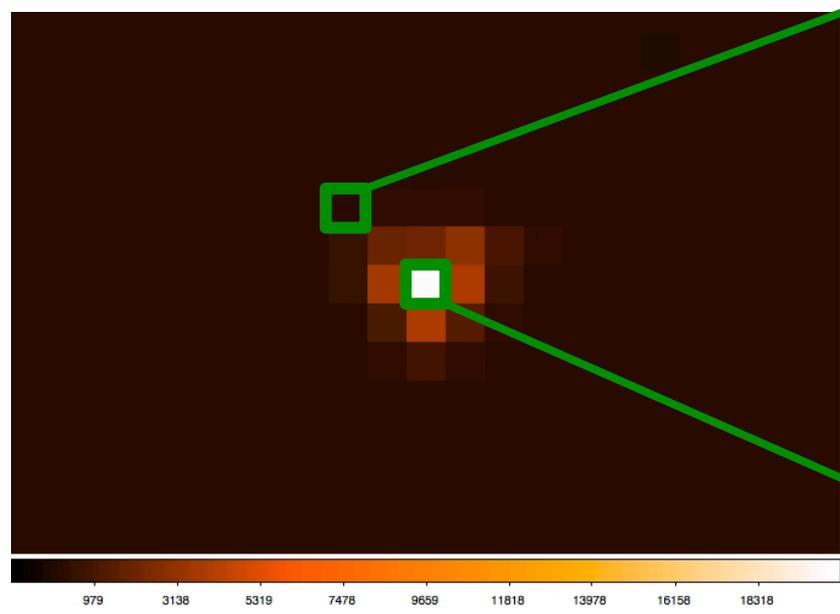
maximum independence = minimum mutual information

$$I(s_1, s_2, \dots, s_n) = \sum_i H(s_i) - H(\mathbf{x}) - \log |\det(\mathbf{W})|$$

1

# Pixel-ICA

Application to Spitzer/IRAC observations

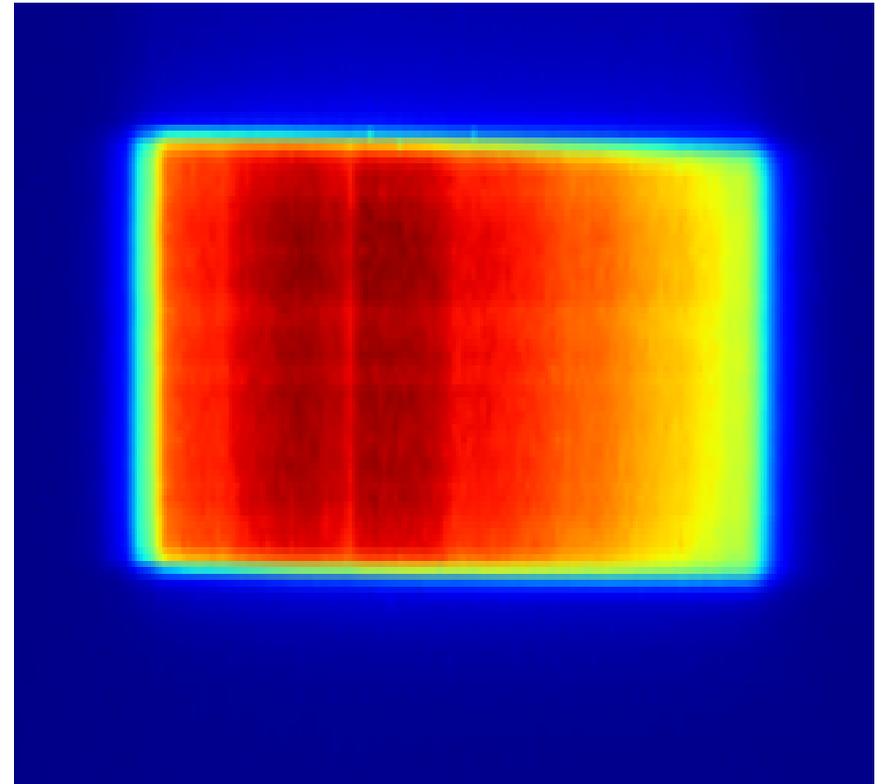
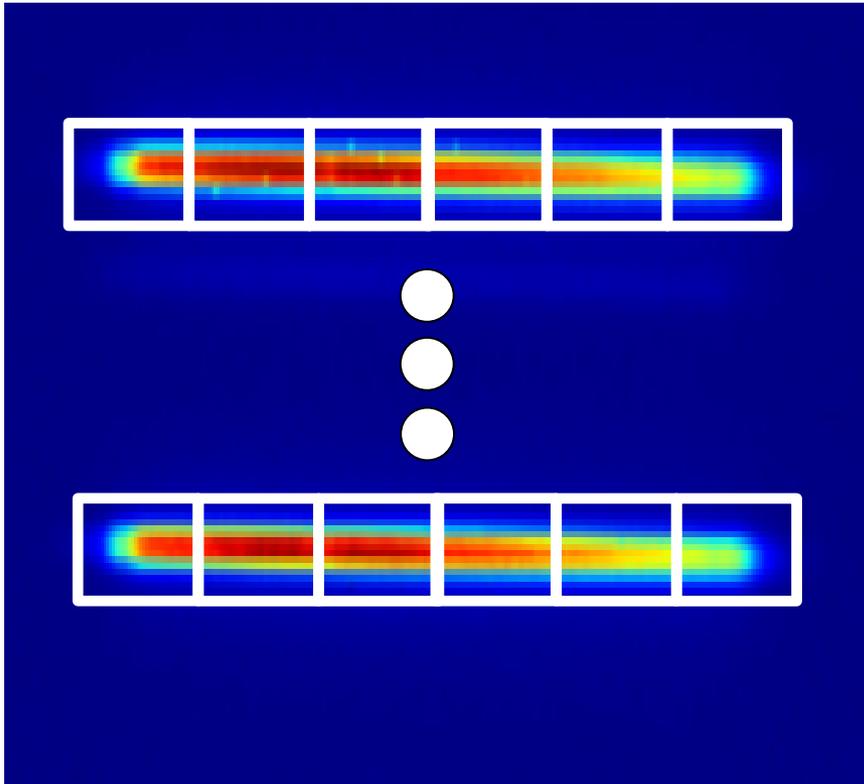


Morello et al. 2014, 2015, 2016; Morello 2015

1

# Stripe-ICA

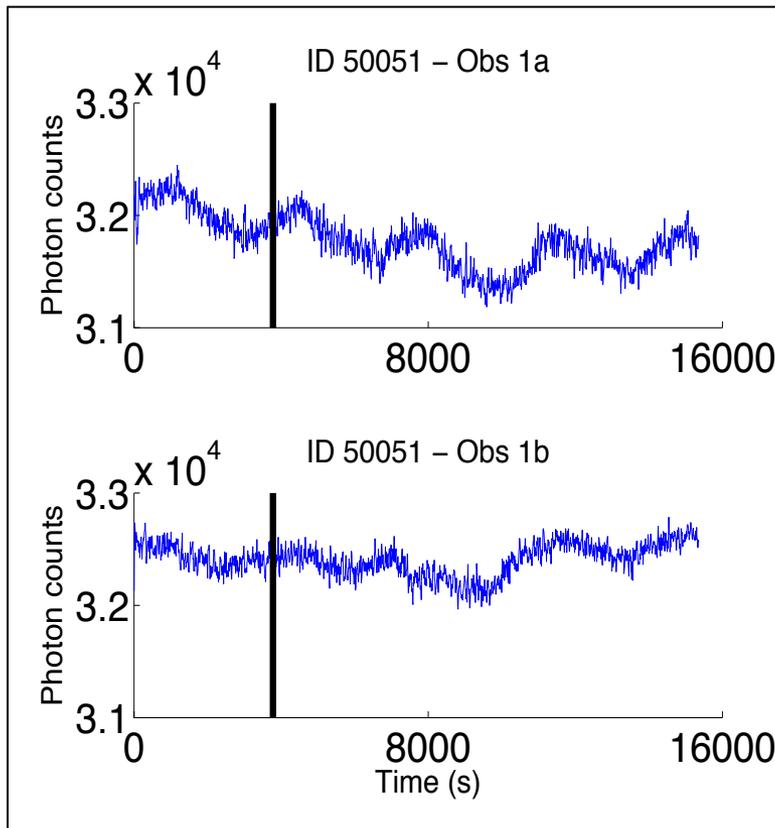
Application to HST/WFC3 observations



**1**

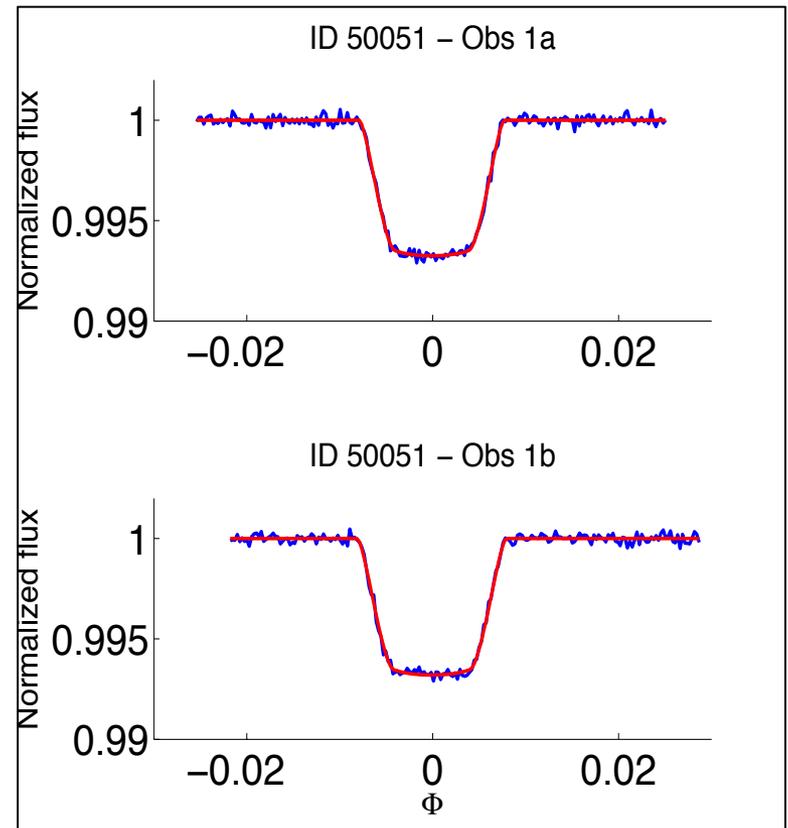
# Spitzer/IRAC observations at 3.6 $\mu\text{m}$ of GJ436b

## Raw lightcurves



ICA  
➔

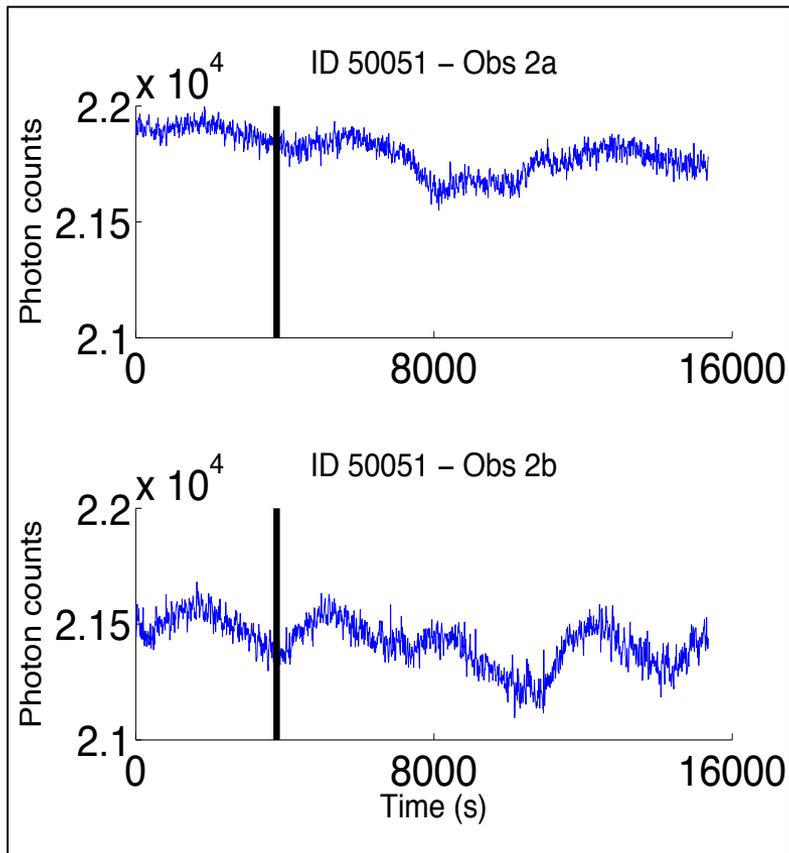
## Detrended lightcurves + models



**1**

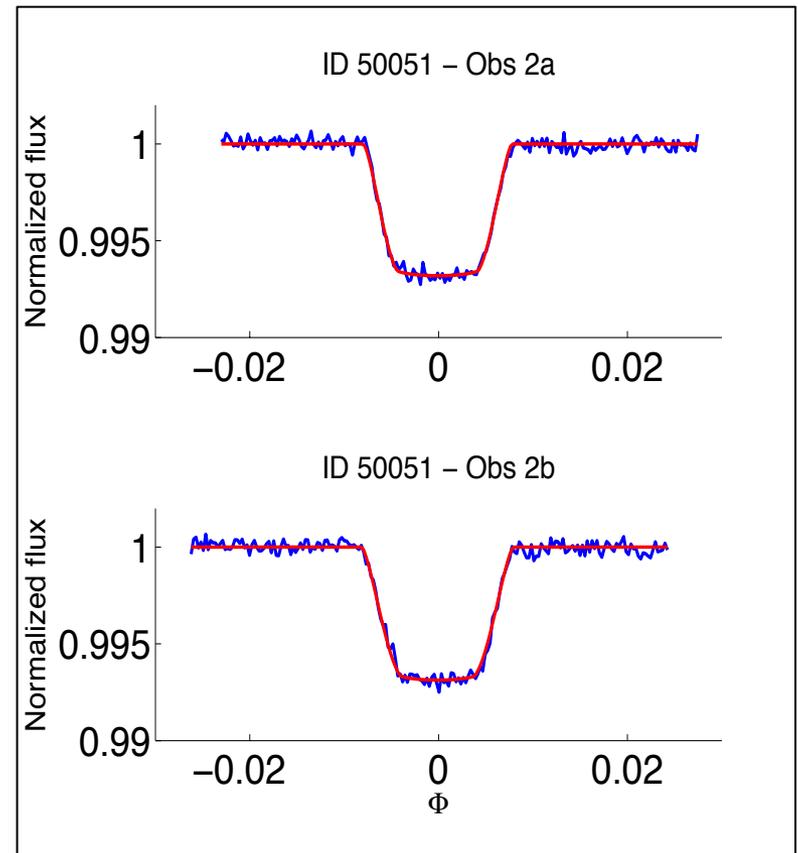
# Spitzer/IRAC observations at 4.5 $\mu\text{m}$ of GJ436b

## Raw lightcurves



ICA  
➔

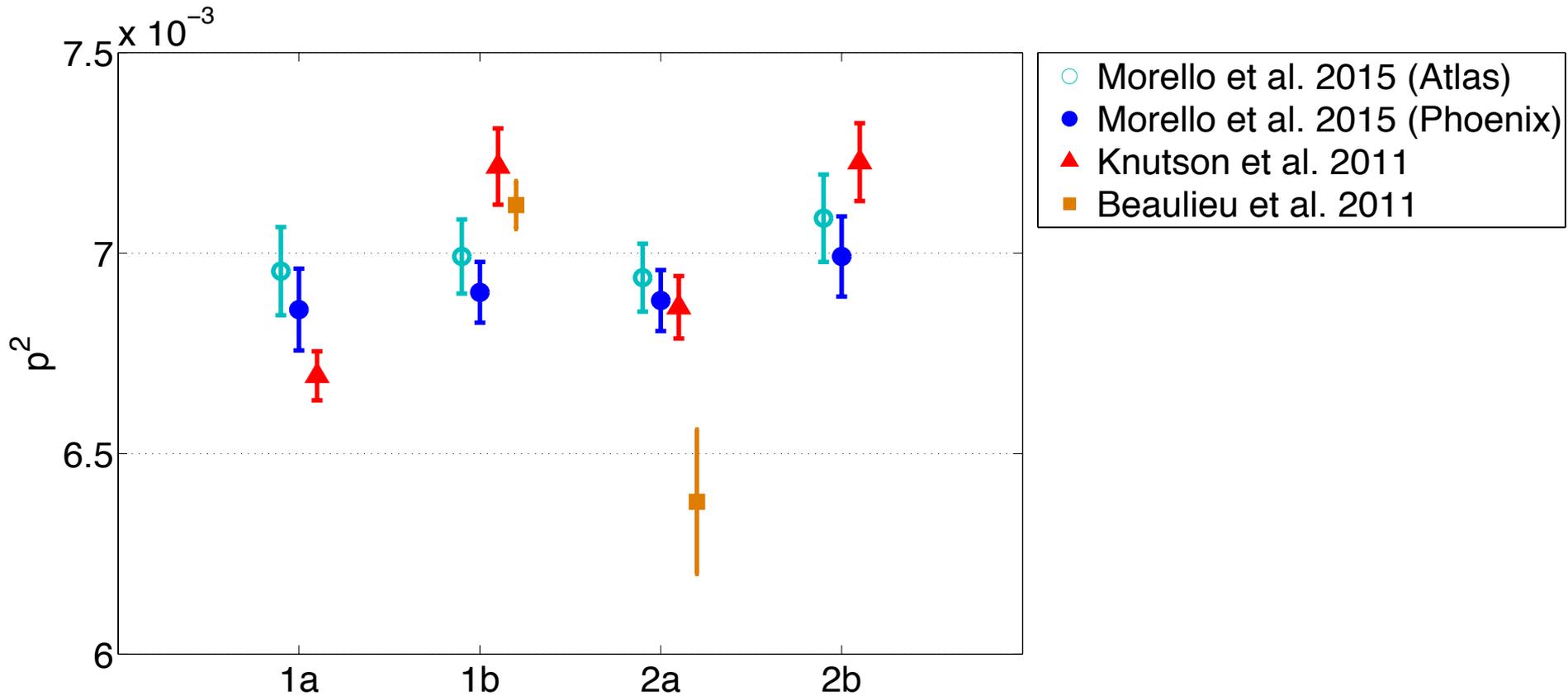
## Detrended lightcurves + models



Morello et al. 2015

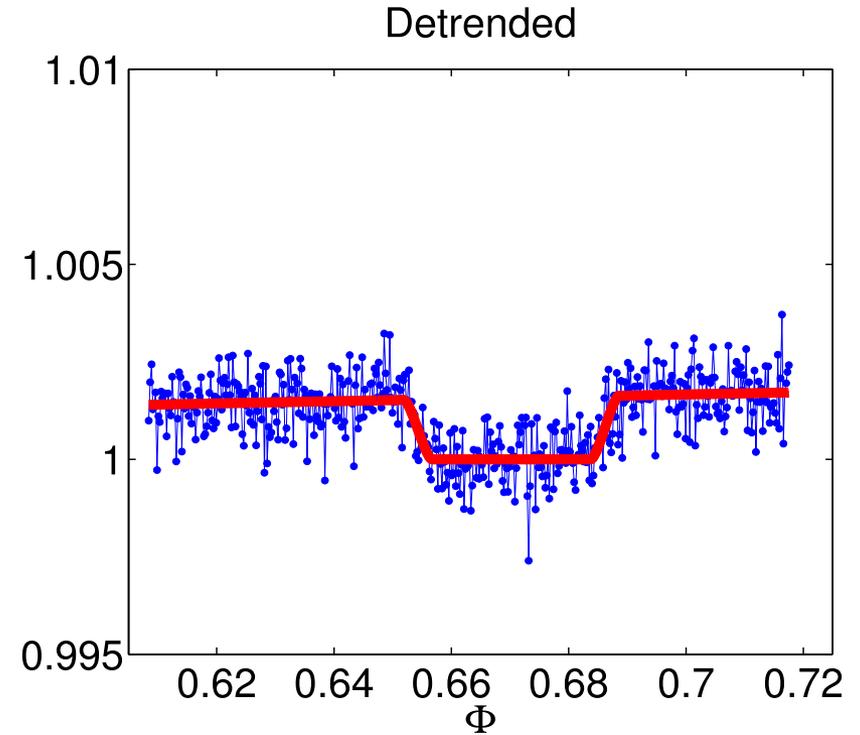
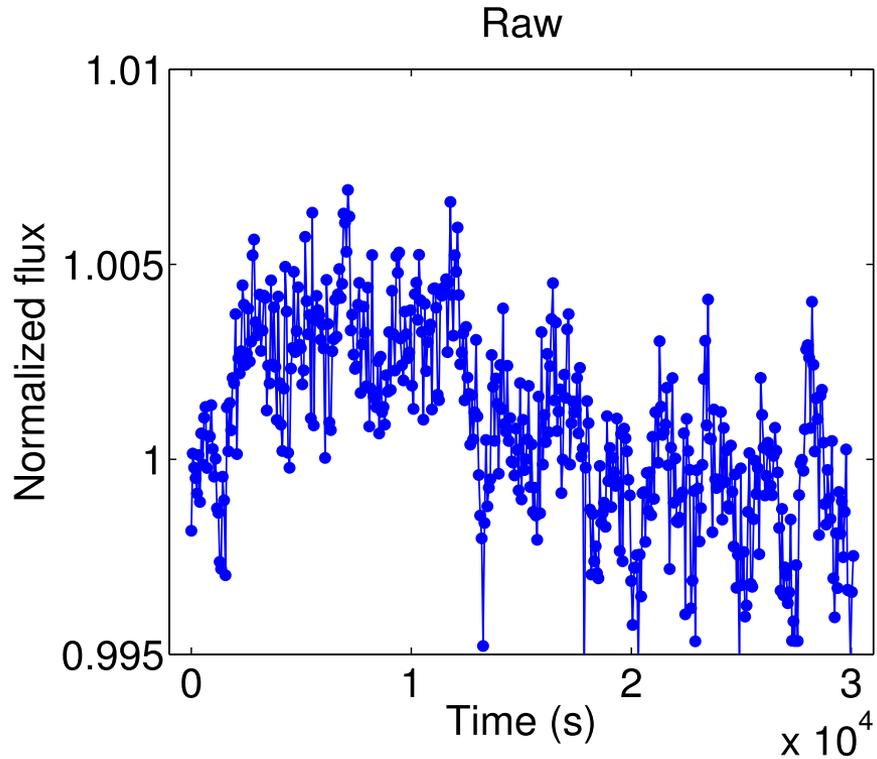
**1**

# Spitzer/IRAC observations of GJ436b - Results



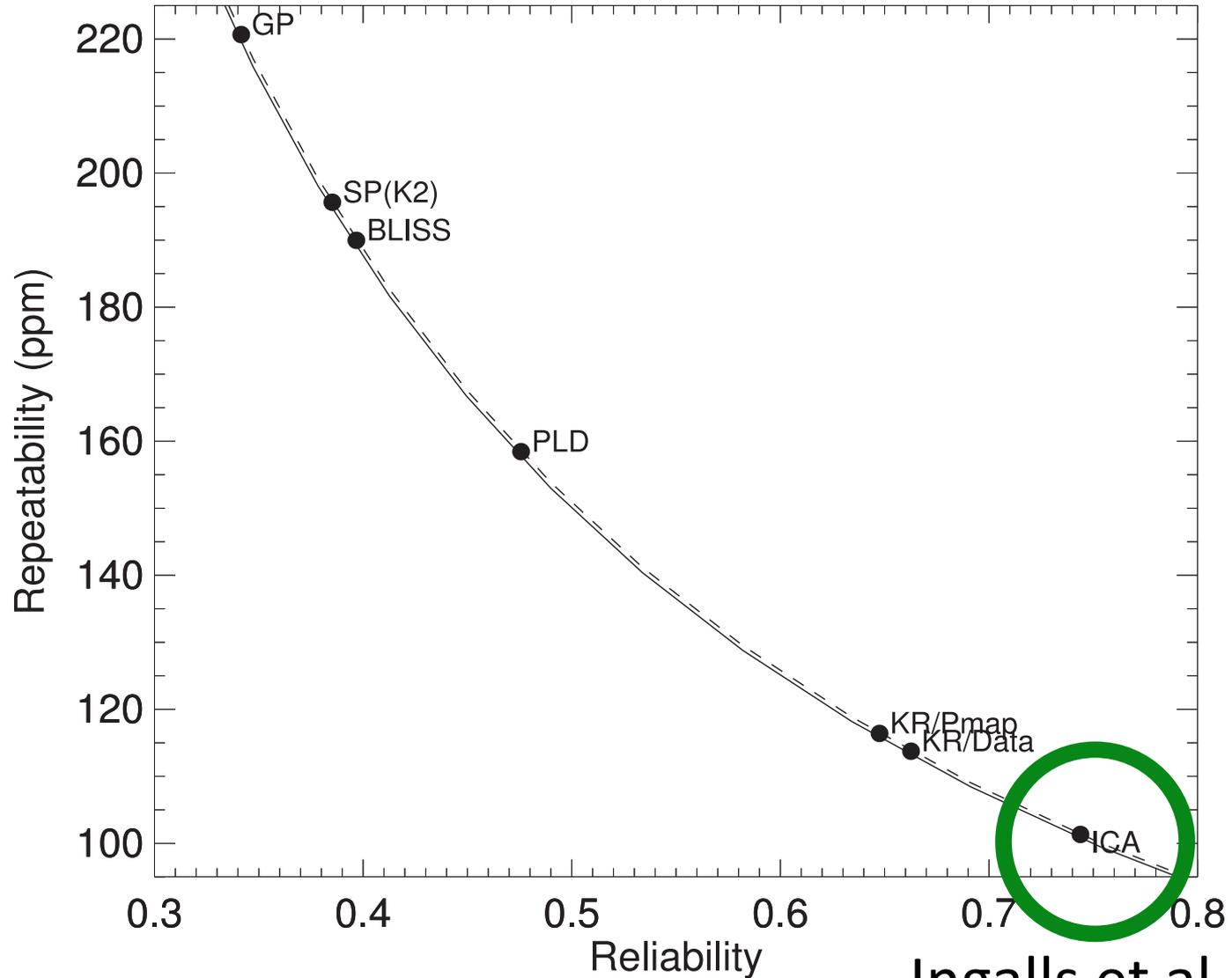
Morello et al. 2015

# Secondary eclipse of XO3b



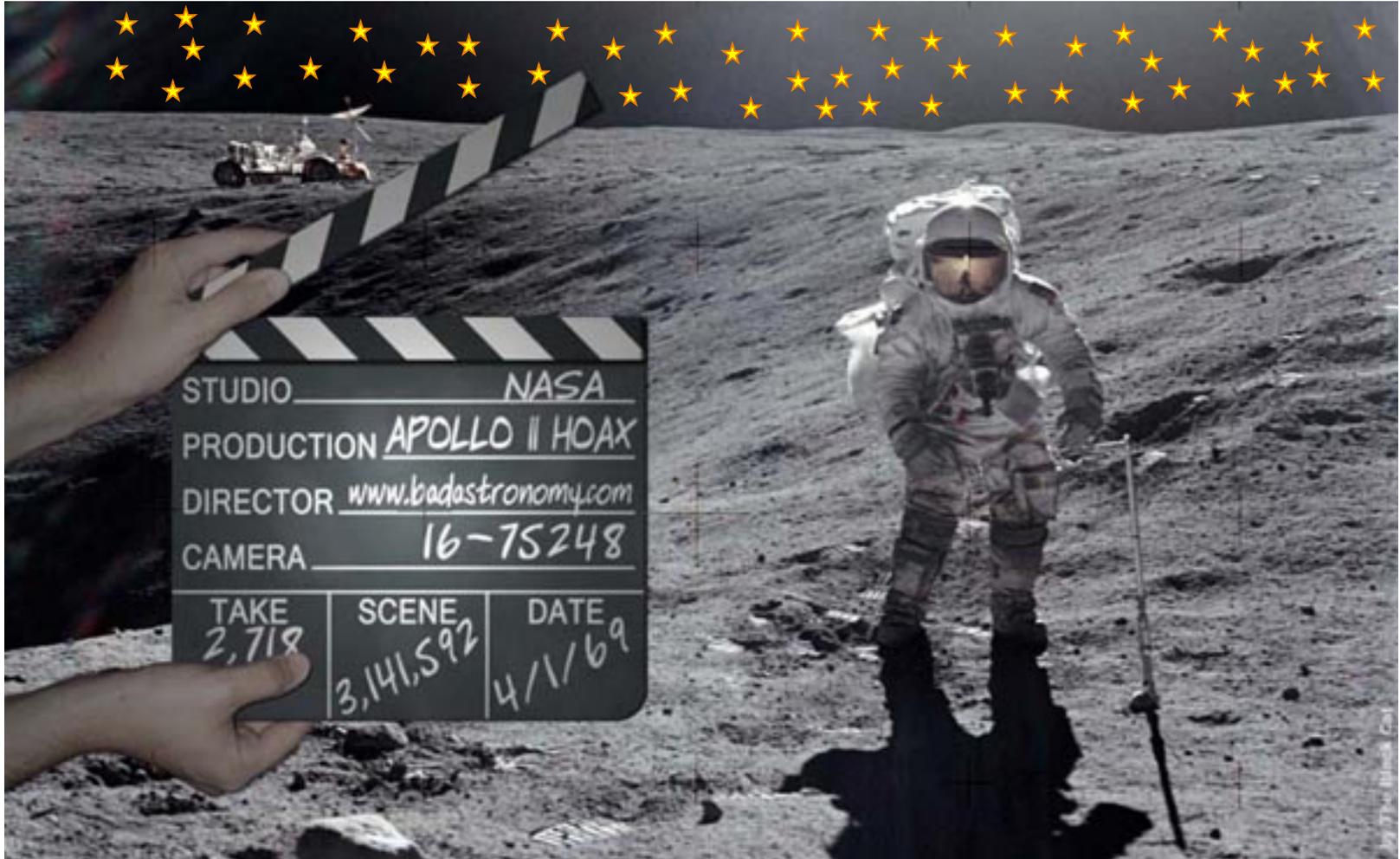
# 1 IRAC data challenge: eclipses of XO3b

THE ASTRONOMICAL JOURNAL, 152:44 (27pp), 2016 August

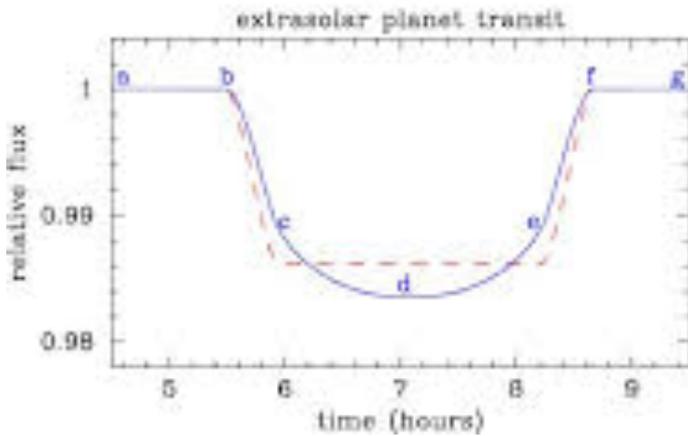
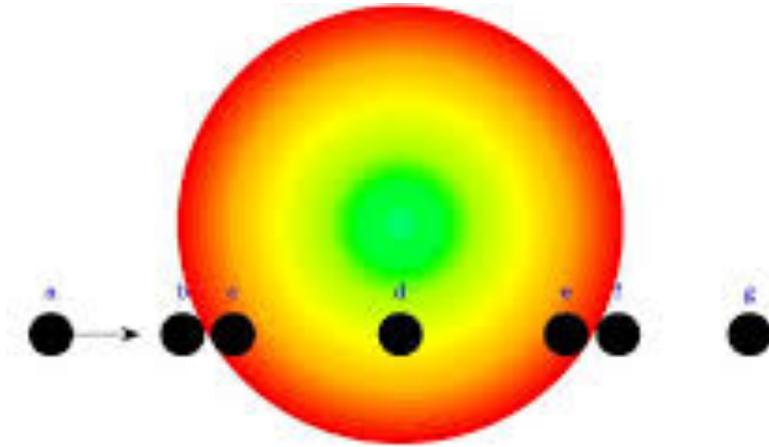


Ingalls et al. 2016

# 2 Accurate modeling the astrophysical scenario: stellar limb-darkening, activity, etc. ( $\sim 100 - 1000$ ppm)



# Stellar limb darkening (1)



$$F(p^2, z(t), I_{\lambda}^*(\mu))$$

$$\left(\frac{R_p}{R_*}\right)^2, \frac{a}{R_*}, i, P, t_0, e, \omega, c_1, c_2, c_3, c_4$$

Four-coefficient law (Claret 2000)

$$\frac{I_{\lambda}(\mu)}{I_{\lambda}(1)} = 1 - \sum_{n=1}^4 c_n (1 - \mu^{n/2})$$

$$\mu = \sqrt{1 - r^2}$$

## 2

# Stellar limb darkening (2)

PROBLEM:

parameter degeneracies!

$$F(p^2, z(t), I_{\lambda}^*(\mu))$$

$$\left(\frac{R_p}{R_*}\right)^2, \frac{a}{R_*}, i, P, t_0, e, \omega, c_1, c_2, c_3, c_4$$

SOLUTIONS (to date):

Limb-darkening coefficients

1. from stellar-atmosphere models

**No empirical verification**

2. free parameters in light-curve fits,  
but linear or two-coefficient laws

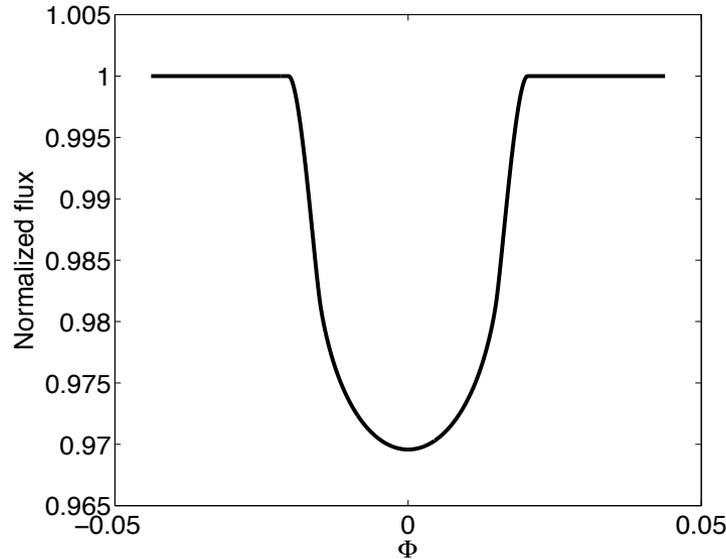
**Inadequate for certain passbands**

**Potential biases**

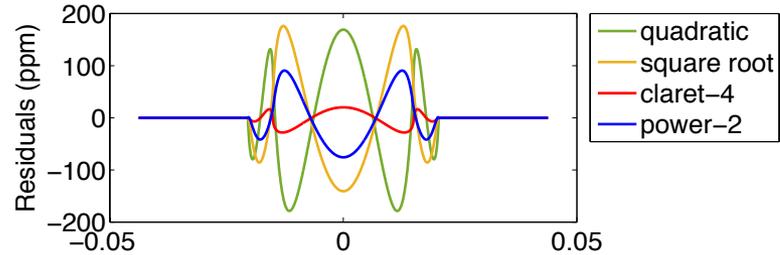
**in transit depth**

# Stellar limb-darkening (3)

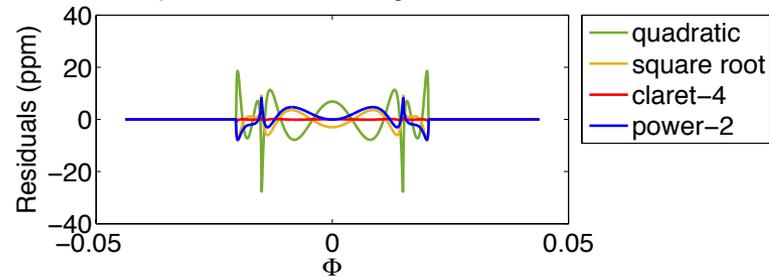
star M5V ( $T_{\text{eff}} = 3084$ ,  $\log g = 5.25$ ), STIS/G430L



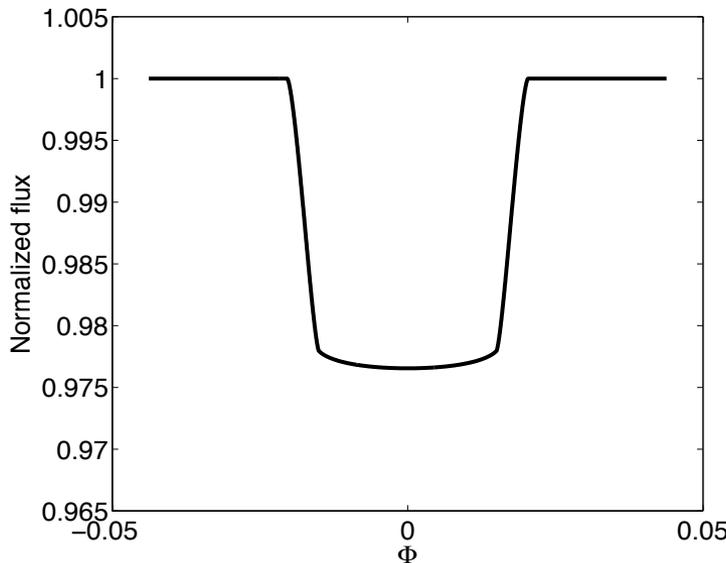
Theoretical limb darkening coefficients



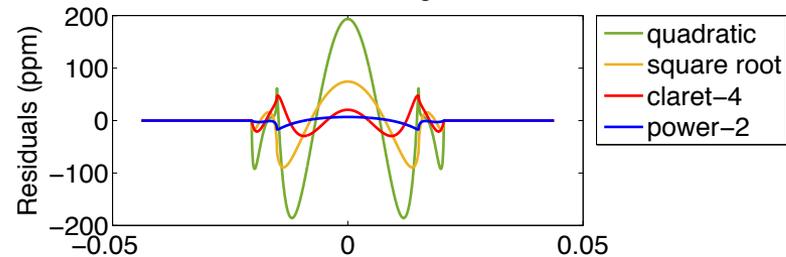
Empirical limb darkening coefficients



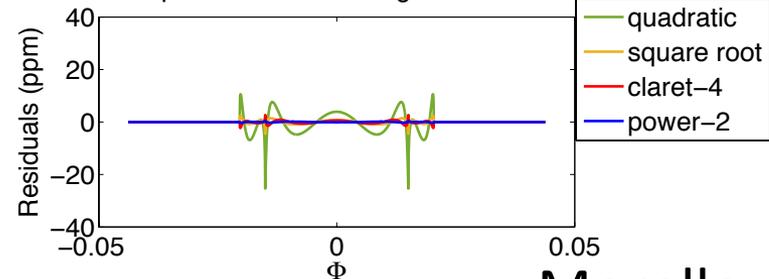
star M5V ( $T_{\text{eff}} = 3084$ ,  $\log g = 5.25$ ), IRAC/ch4



Theoretical limb darkening coefficients



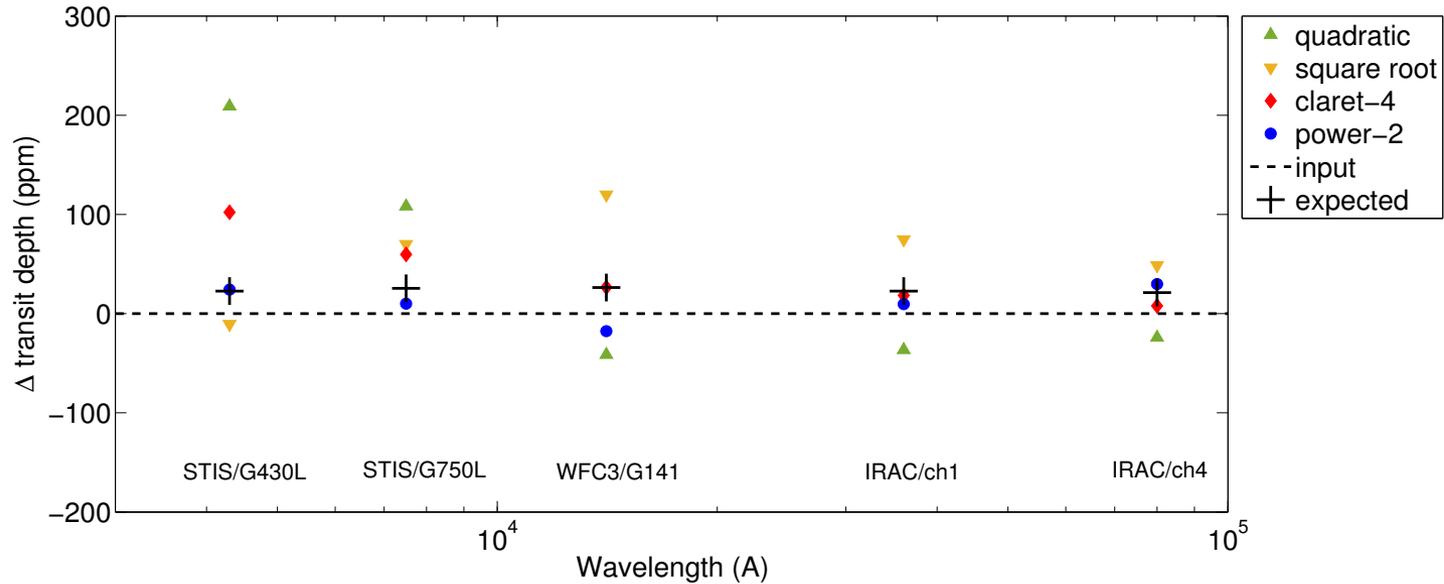
Empirical limb darkening coefficients



# Stellar limb-darkening (4)

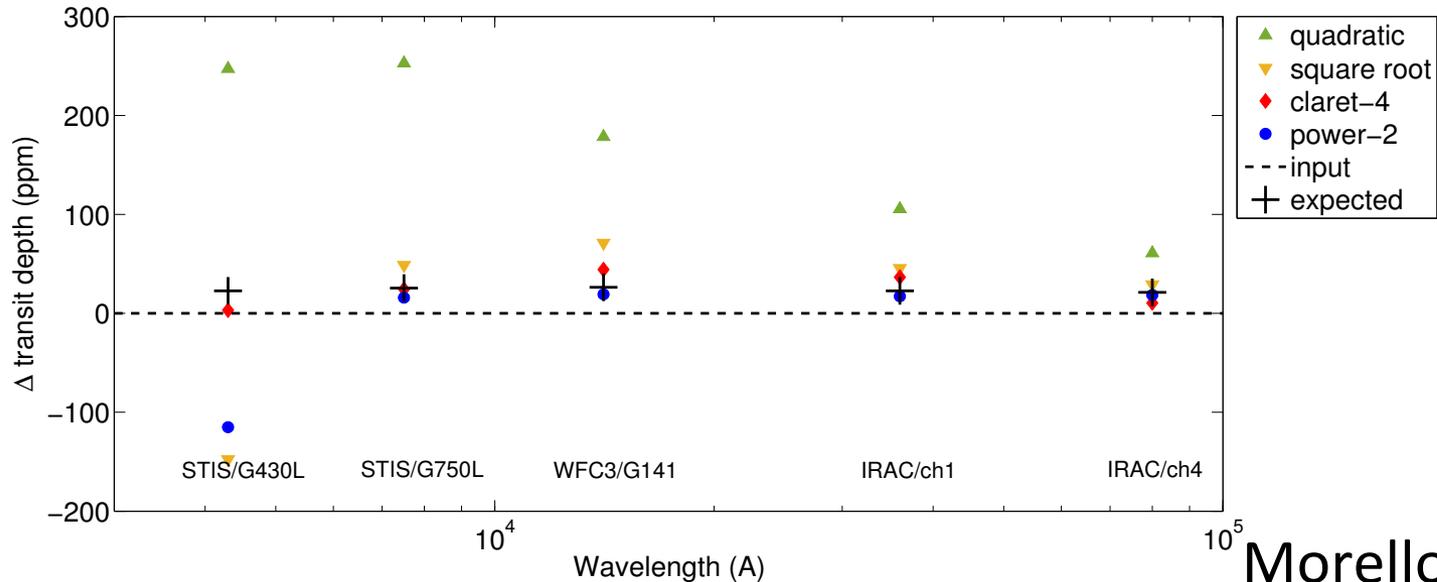
star M5V ( $T_{\text{eff}} = 3084$ ,  $\log g = 5.25$ )

(stellar model-atmosphere)



star M5V ( $T_{\text{eff}} = 3084$ ,  $\log g = 5.25$ )

(light-curve fit)



# 2

# SEA BASS

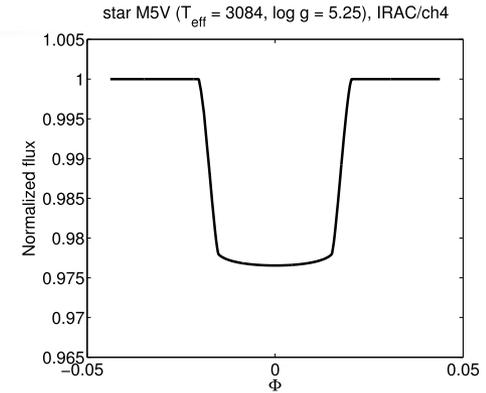
## Stellar and Exoplanetary Atmospheres Bayesian Analysis Simultaneous Spectroscopy



### INFRARED TRANSITS

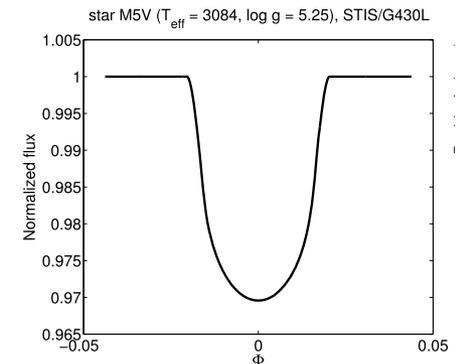
- ✓ Two-coefficient limb-darkening laws,  
or
- ✓ Fixed limb-darkening coefficients

=> geometric parameters (orbital, transit duration)



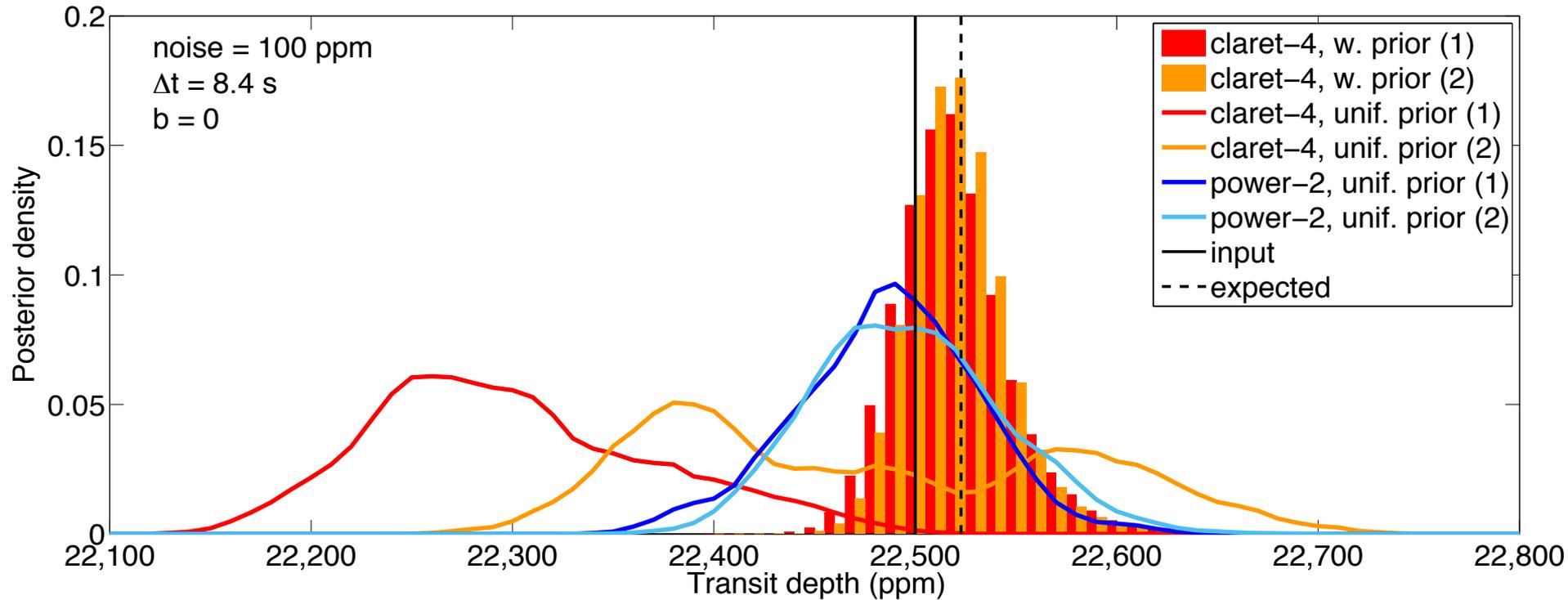
### VISIBLE TRANSITS

- ✓ Informative Bayesian priors on the geometric parameters
- ✓ Fully empirical four-coefficient limb-darkening



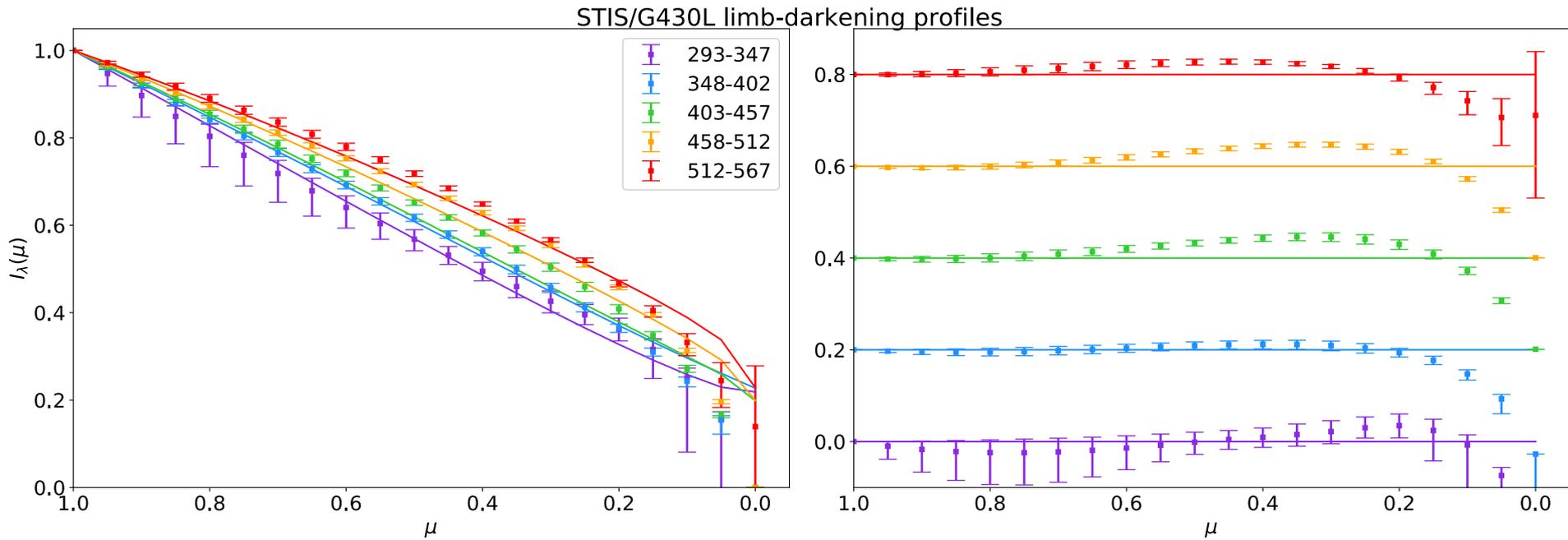
# SEA BASS: simulations

star M5V ( $T_{\text{eff}} = 3084$ ,  $\log g = 5.25$ ), STIS/G430L



2

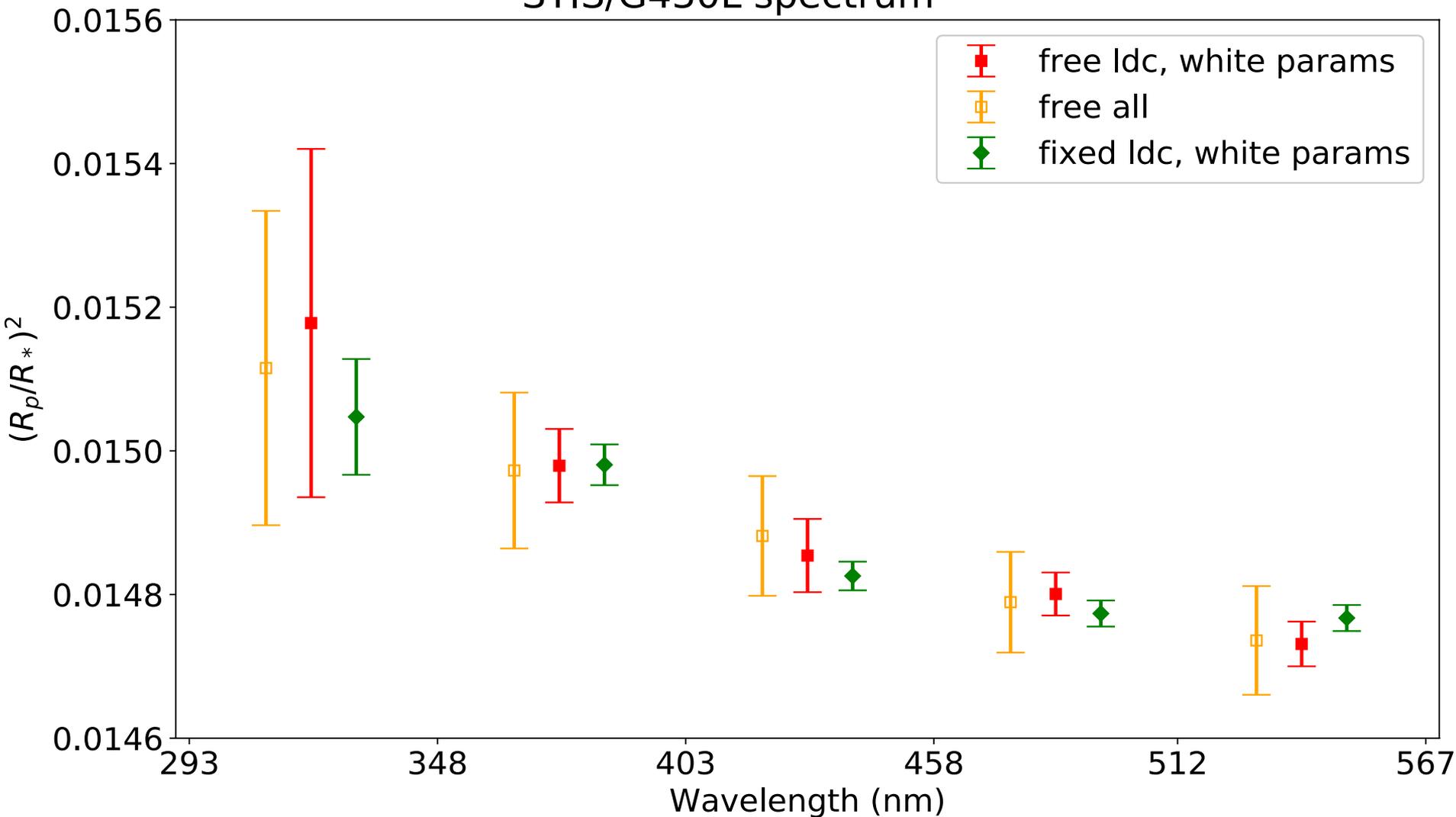
# SEA BASS: HD20458 I-d profile



2

# SEA BASS: HD20458b spectrum

STIS/G430L spectrum



# State-of-the-art scientific results



# Rapid rotators: KOI13

- Stellar oblateness
- Gravity darkening
- Rossiter–McLaughlin effect

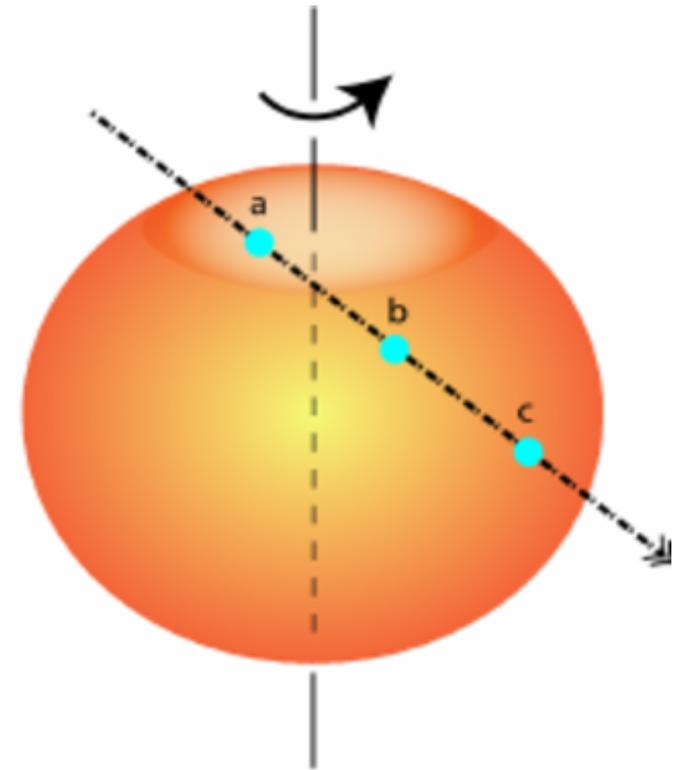
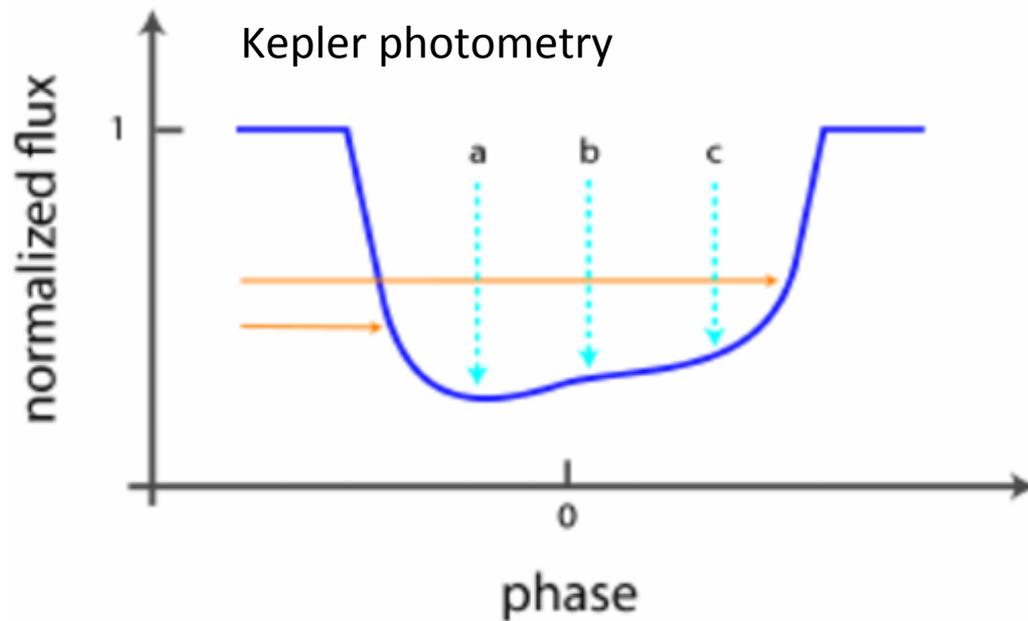


Image credit: Szabo et al. 2013

Absolute system dimensions determined from  
star's projected equatorial rotational speed

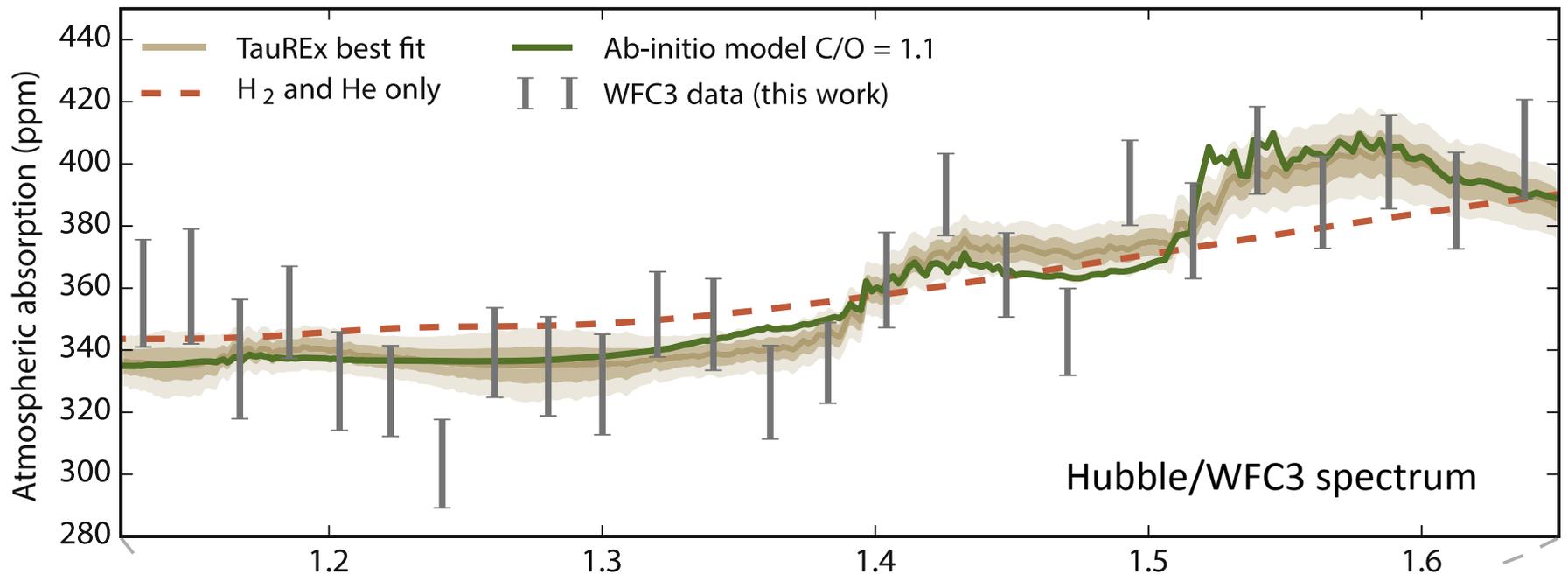
Howarth & Morello 2017

# Atmosphere around super-Earth 55 Cnc e

Star:  $T_{\text{eff}} = 5200 \text{ K}$ ,  $M_* = 0.91 M_{\odot}$ ,  $R_* = 0.94 R_{\odot}$

Planet:  $T_{\text{eq}} = 1950 \text{ K}$ ,  $M_p = 8.1 M_{\oplus}$ ,  $R_p = 2.0 M_{\oplus}$

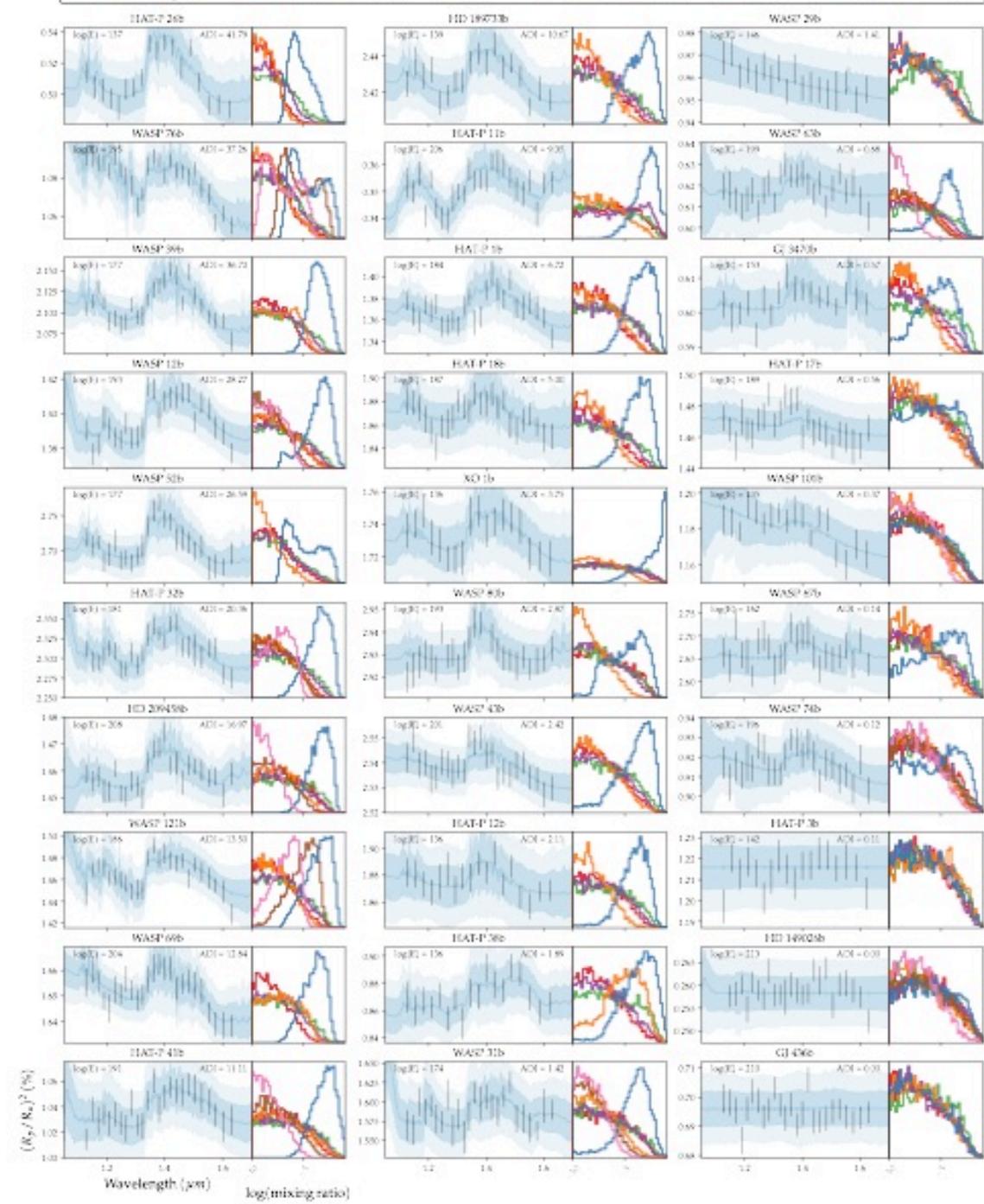
Orbital:  $a = 0.01545 \text{ AU} = 3.523 R_*$ ,  $P = 0.7365417 \text{ days}$





# A POPULATION STUDY OF HOT JUPITER ATMOSPHERES

- Largest catalogue of exoplanet atmospheres (30 planets);
- Observed with Hubble/WFC3;
- H<sub>2</sub>O, TiO, VO detections;
- No clouds in very hot planets;
- Atmospheric Detectability Index (ADI) not correlated with S/N



# Conclusions

## TO DATE

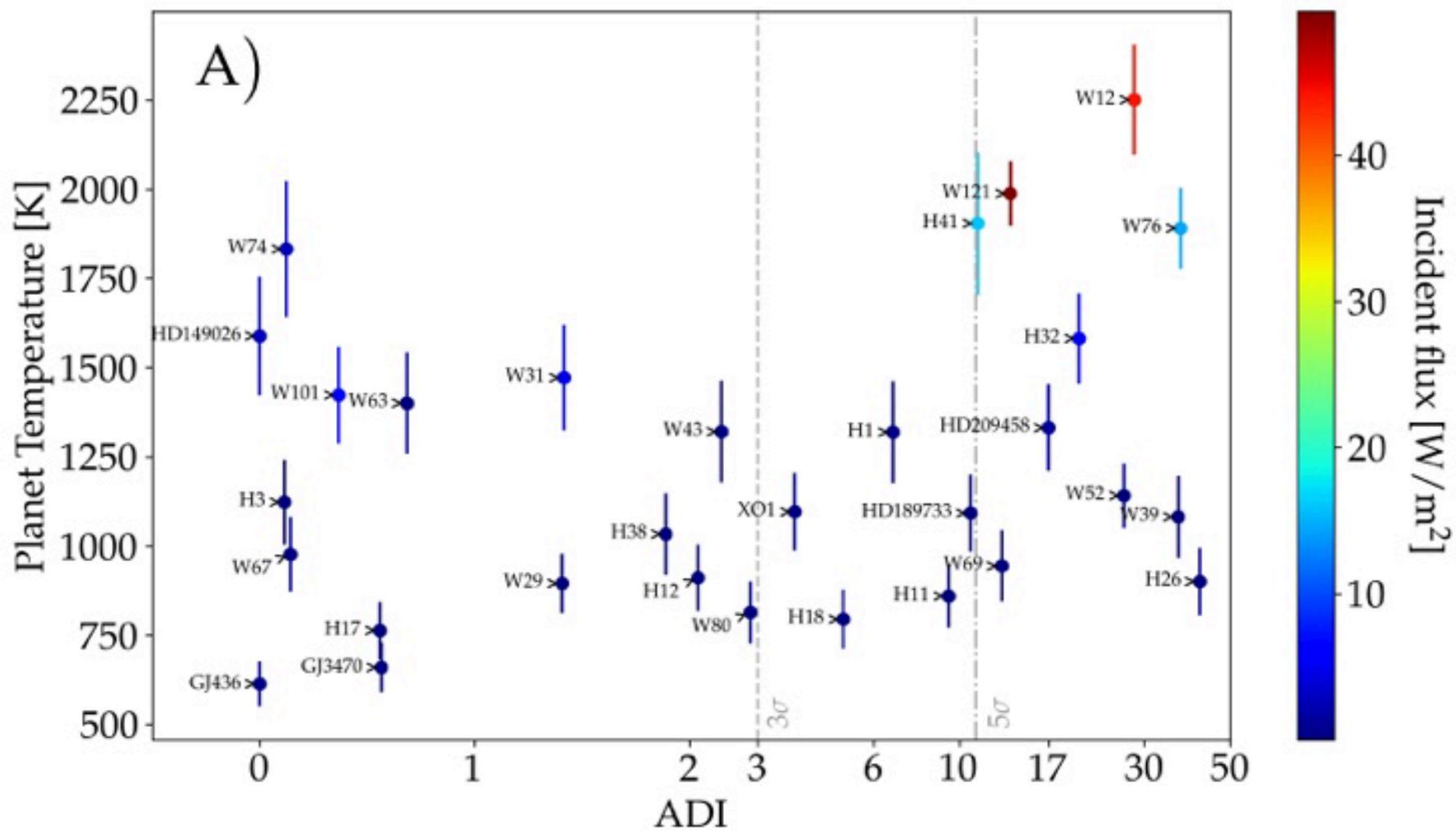
- Transit/eclipse/phase curve spectroscopy for exoplanets atmospheric characterisation;
- High-performance, objective (blind) data detrending methods: pixel-ICA, stripe-ICA;
- High-precision and accuracy empirical star's limb-darkening: SEA BASS;
- Scientific results: individual targets (KOI13, 55 Cnc e), catalog (30 planets)

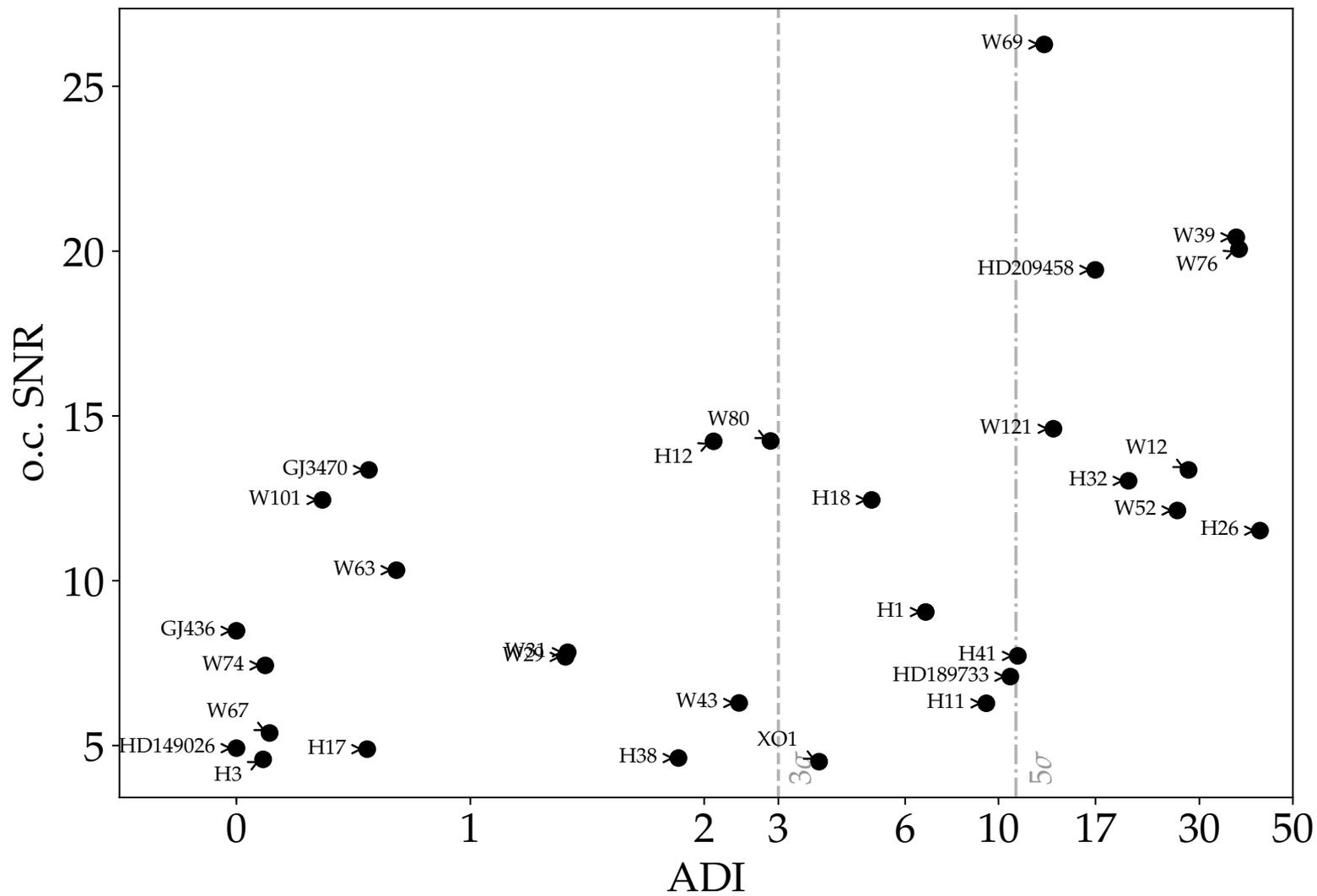
## FUTURE PROJECTS

- JWST data detrending pipelines
- Observational techniques for stellar activity, non-spherical bodies, etc.
- Catalog: higher-quality data, larger wavelength coverage and sample of exoplanets;
- Correlations, link with formation and evolution.

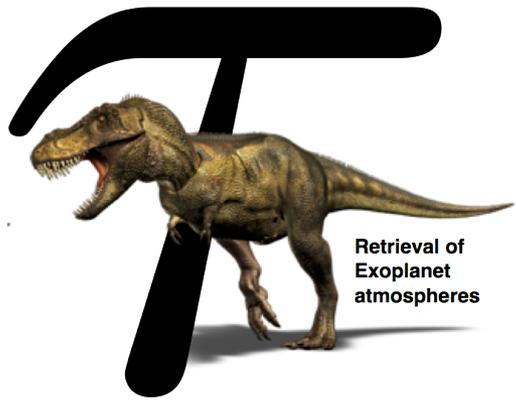


MERCI



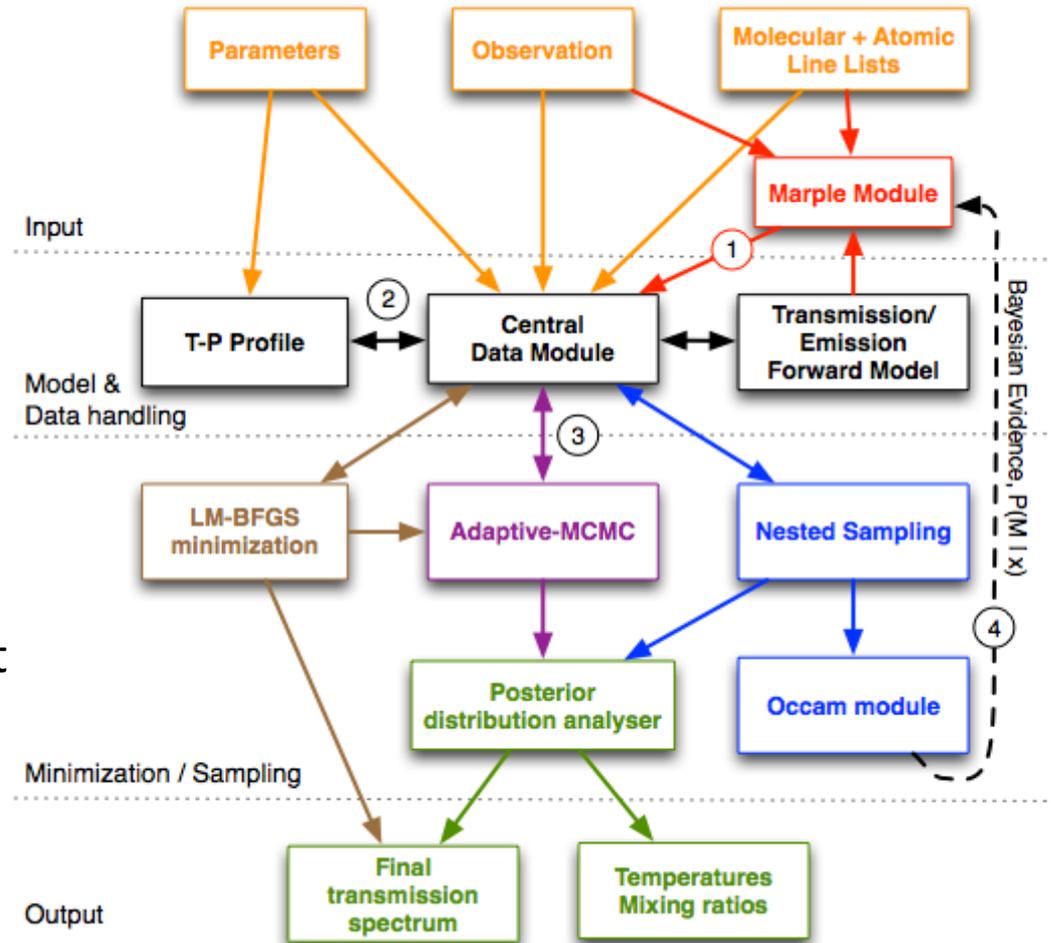


3



Retrieval of  
Exoplanet  
atmospheres

- **Fully Bayesian Retrieval**
  - MCMC
  - Nested Sampling
  - Maximum Likelihood
- **Cross-sections** from Hitran/  
Hitemp and the ExoMol project
- Prior composition selection  
through **pattern recognition  
software**
- **Full parallelisation for cluster  
computing**



# PCA vs ICA

- Linear transformation of the observations;
- Find a basis of orthogonal components.

## PCA

- Uncorrelatedness:  
 $E\{xy\} = E\{x\} E\{y\}$
- Up to 2<sup>nd</sup>-order statistics
- Pre-processing step for ICA

## ICA

- Independence:  
 $E\{f(x)g(y)\} = E\{f(x)\} E\{g(y)\}$
- Higher-order statistics

# ICA: statistics (2)

Among all the distributions with fixed mean and covariance, the gaussian distribution has the maximum entropy.

$$J(\mathbf{y}) = H(\mathbf{y}_{gauss}) - H(\mathbf{y}) \quad \text{negentropy}$$

- Mutual information and negentropy are hard computing.
- Alternatively, we can maximize non-gaussianity of the source signals, through different estimators.

$$kurt(y) = E(y^4) - 3E(y^2)^2 \quad \text{kurtosis}$$

$$J(y) \approx \sum_{i=1}^p k_i [E\{G_i(y)\} - E\{G_i(\nu)\}]^2 \quad \text{approximated negentropy}$$

# Negentropy approximations

$$J(y) \approx \sum_{i=1}^p k_i [E\{G_i(y)\} - E\{G_i(\nu)\}]^2$$



contrast functions



random gaussian variable

$$G_i(s) = \log(\cosh s)$$

$$G_i(s) = e^{-\frac{s^2}{2}}$$

$$G_i(s) = s^4$$

# Time-correlated signals

Joint diagonalization of several time-lagged covariance matrices

$$\mathbf{R}_{\mathbf{x}}(\tau_k) = E(\mathbf{x}(t)\mathbf{x}^T(t + \tau_k)), \quad k = 0, \dots, M - 1$$

$$\mathbf{R}_{\mathbf{s}}(\tau_k) = \mathbf{A}^T \mathbf{R}_{\mathbf{x}}(\tau_k) \mathbf{A}$$

We adopt MULTICOMBI code (Tichavsky et al. 2006) to perform ICA decomposition.

# Interference-to-Signal Ratio

- If the source signals and the true mixing matrix are known, it is possible to test the goodness of the separation:

Normalized gain matrix  $\tilde{\mathbf{G}} = \hat{\mathbf{W}} \mathbf{A} \mathbf{D}^{\frac{1}{2}}$  Diagonal matrix of the variances of the estimated source signals

Estimated inverse mixing matrix      True mixing matrix

- In case of perfect demixing, the normalized gain matrix is the identity.

$$\text{ISR}_{ij} = \frac{\tilde{\mathbf{G}}_{ij}^2}{\tilde{\mathbf{G}}_{ii}^2} \approx \tilde{\mathbf{G}}_{ij}^2$$

- For certain algorithms, it is possible to calculate asymptotical expressions for the **ISR** matrix, which are independent on the mixing matrix.

# Error bars

$$\sigma_{par} = \sigma_{par,0} \sqrt{\frac{\sigma_0^2 + \sigma_{ICA}^2}{\sigma_0^2}}$$

$$\sigma_{ICA}^2 = f^2 \left( \sum_j o_j^2 \mathbf{ISR}_j + \sigma_{ntc-fit}^2 \right)$$

MULTICOMBI:

$$\mathbf{ISR} = \frac{\mathbf{ISR}^{EF} + \mathbf{ISR}^{WA}}{2}$$

$$\mathbf{ISR}_{i,j} = \min(\mathbf{ISR}_{i,j}^{EF}, \mathbf{ISR}_{i,j}^{WA})$$

