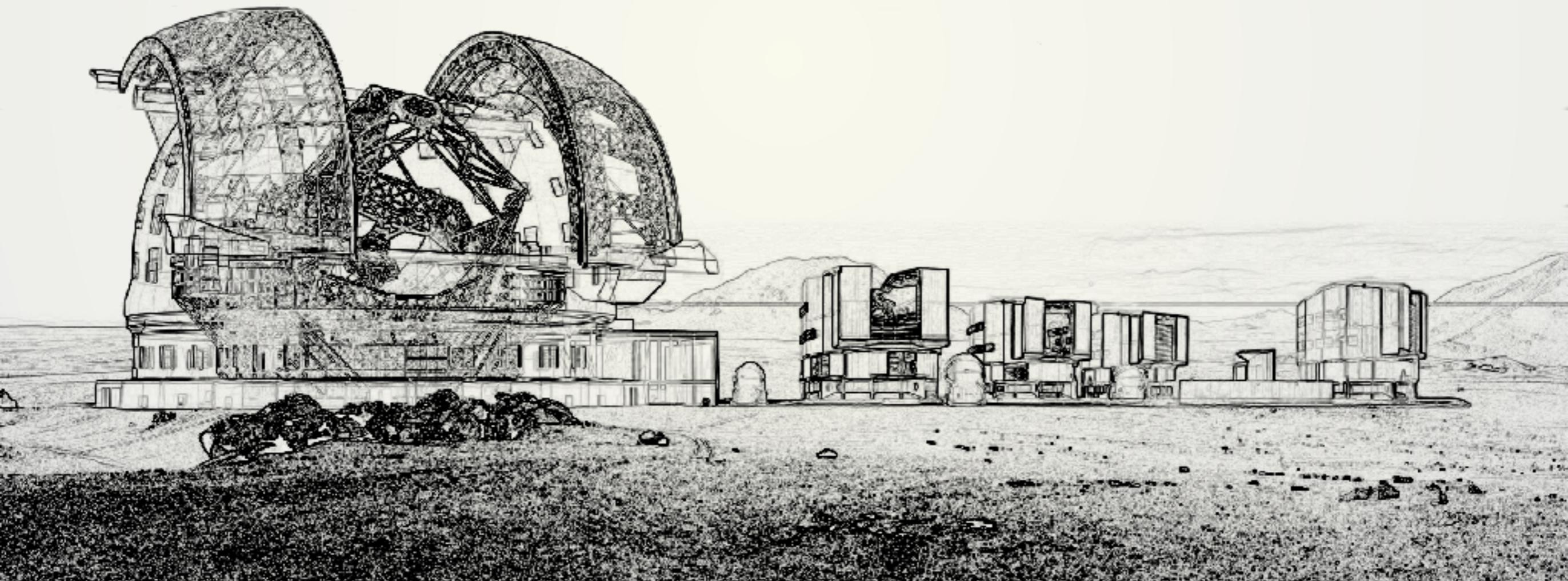


Measuring abundances from high-resolution cross-correlation spectroscopy

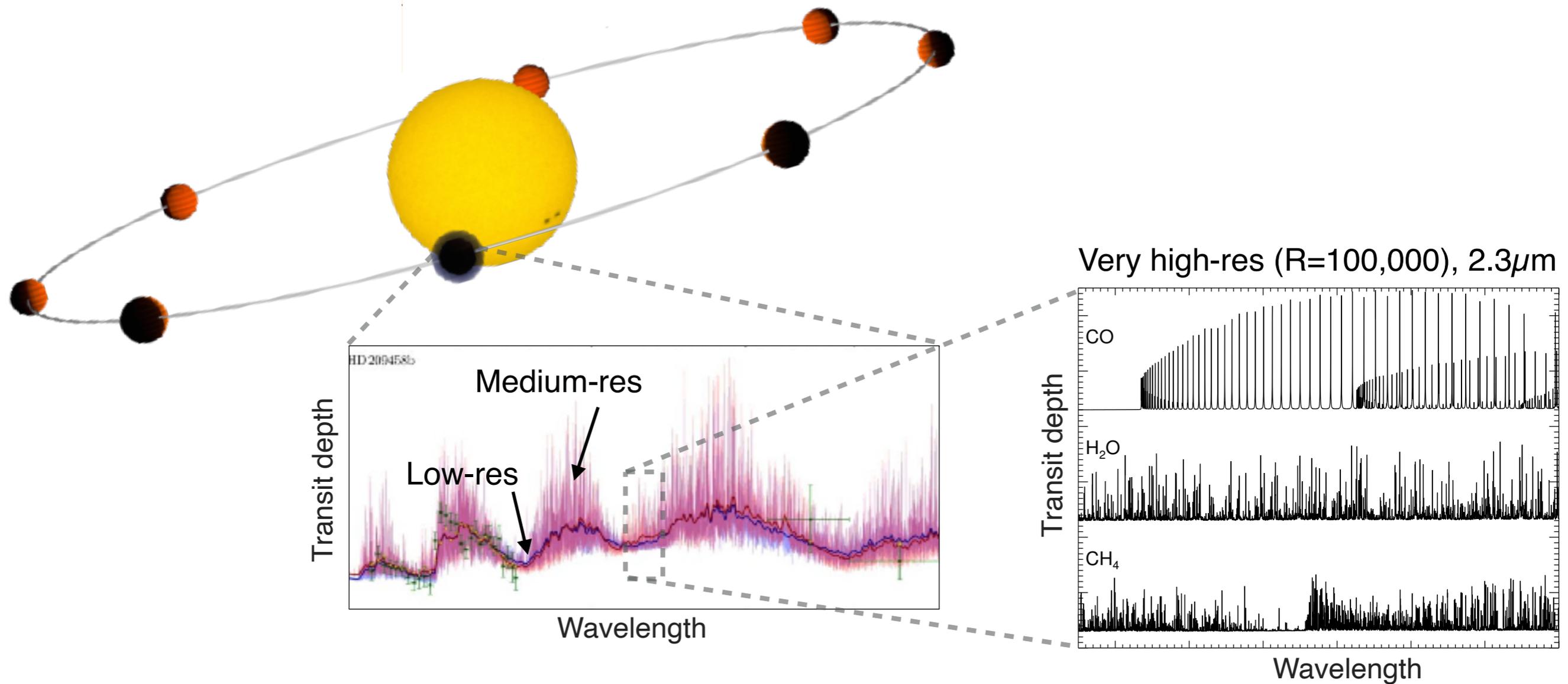
See also Neale Gibson's talk tomorrow 27/08

Matteo Brogi

Assistant Professor, University of Warwick



Exoplanets at high spectral resolution



Each species has a **unique** pattern of spectral lines
Species can be “matched” line by line to templates, e.g. via **cross correlation**

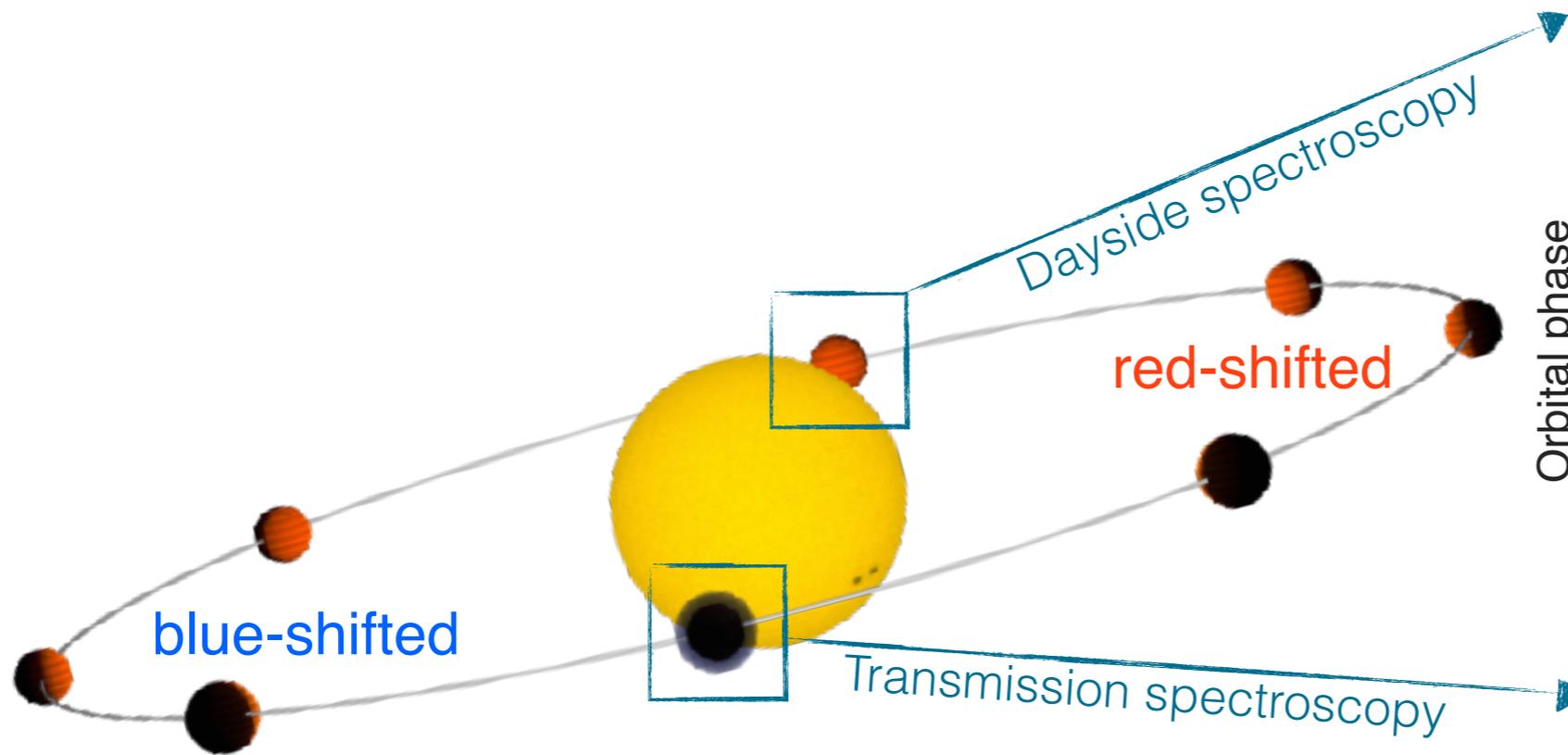
Only possible with ground based telescopes



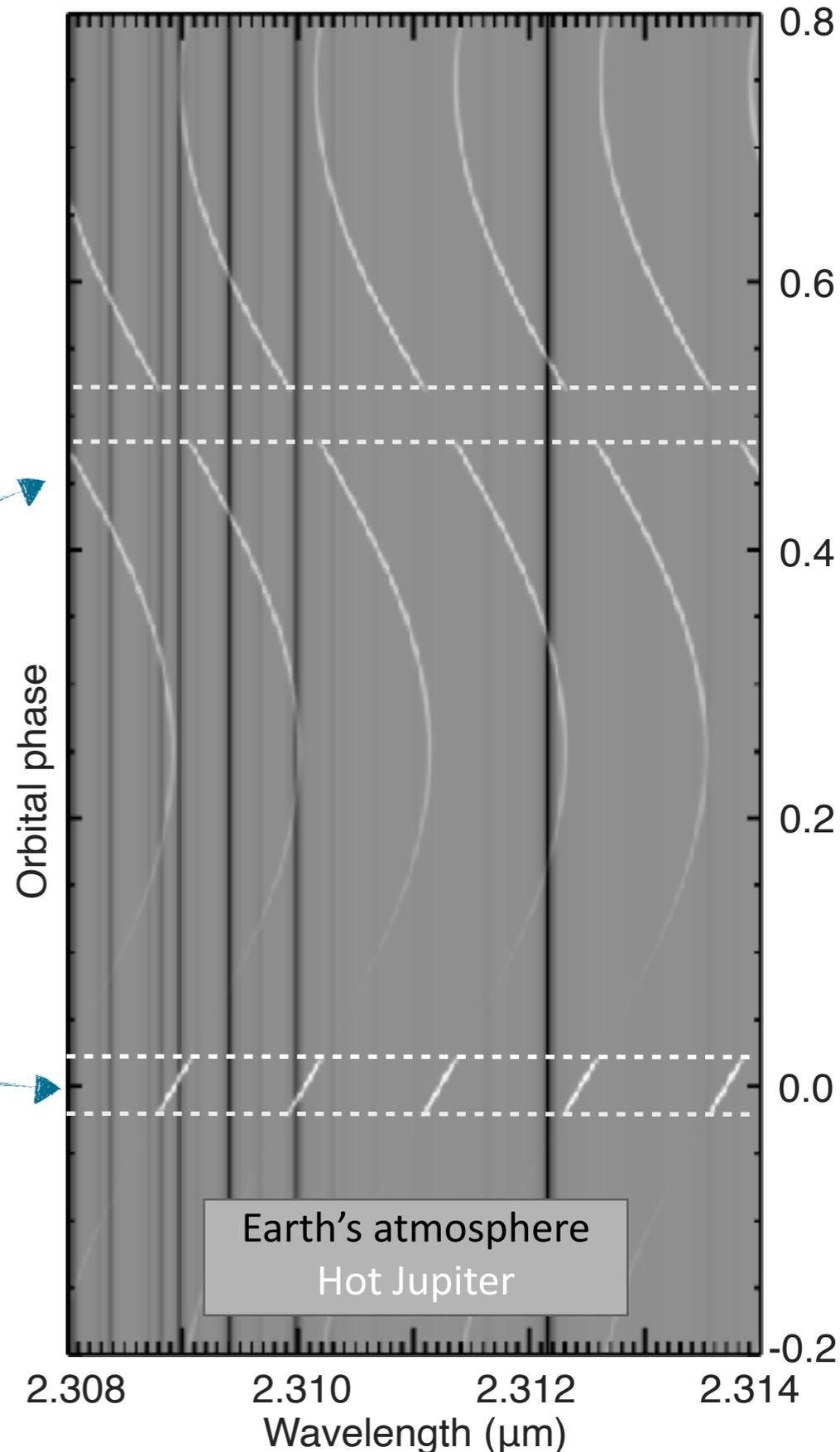
Detecting the orbital motion of close-in planets

Detecting change in planet radial velocity during a few hours of observations
(Planet RV: 10-100 km/s; Stellar RV: 10-100 m/s)

- ⇒ Telluric and planet signal disentangled
- ⇒ Planet radial velocity directly measured



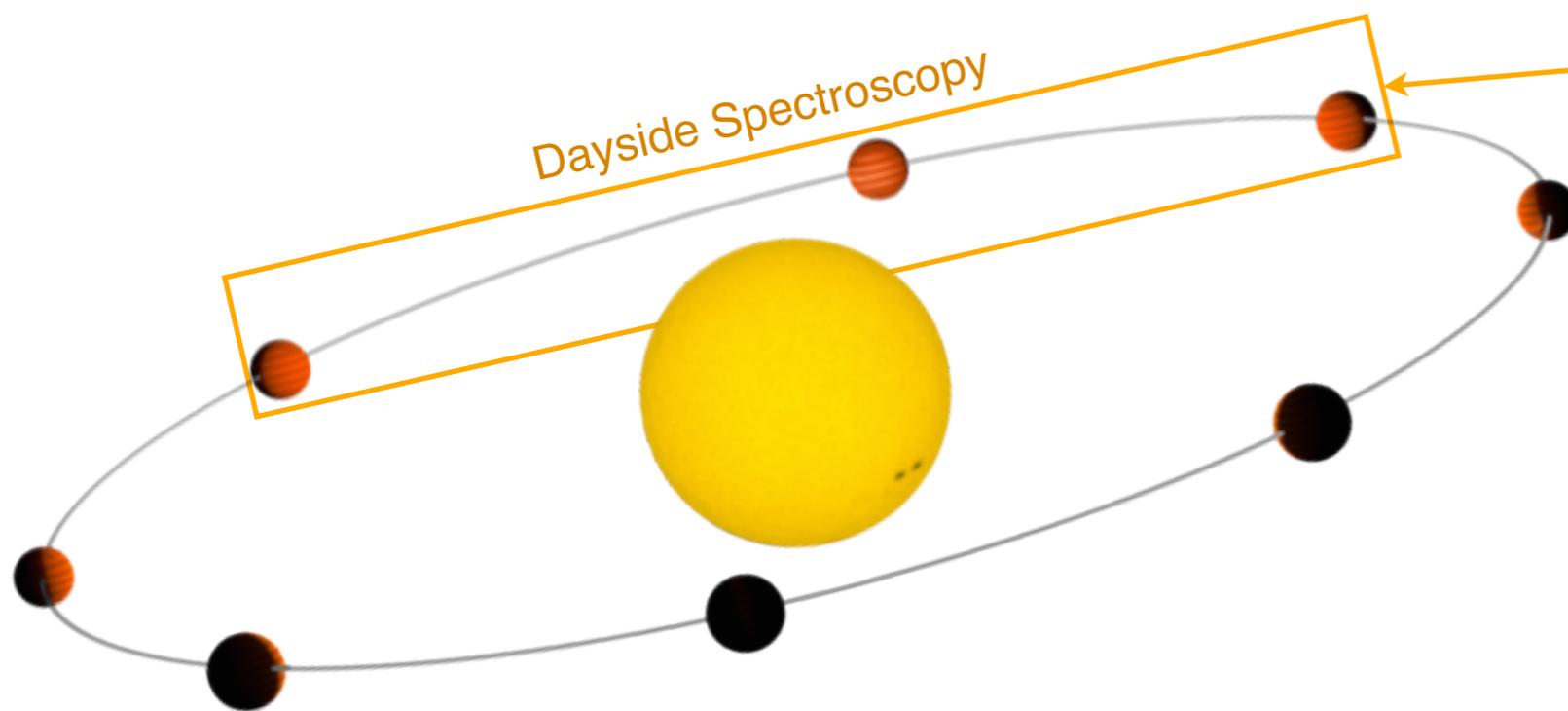
Carbon monoxide - 2.3 μm



High spectral resolution of non-transiting planets

The **thermal spectrum** of the planet is targeted directly

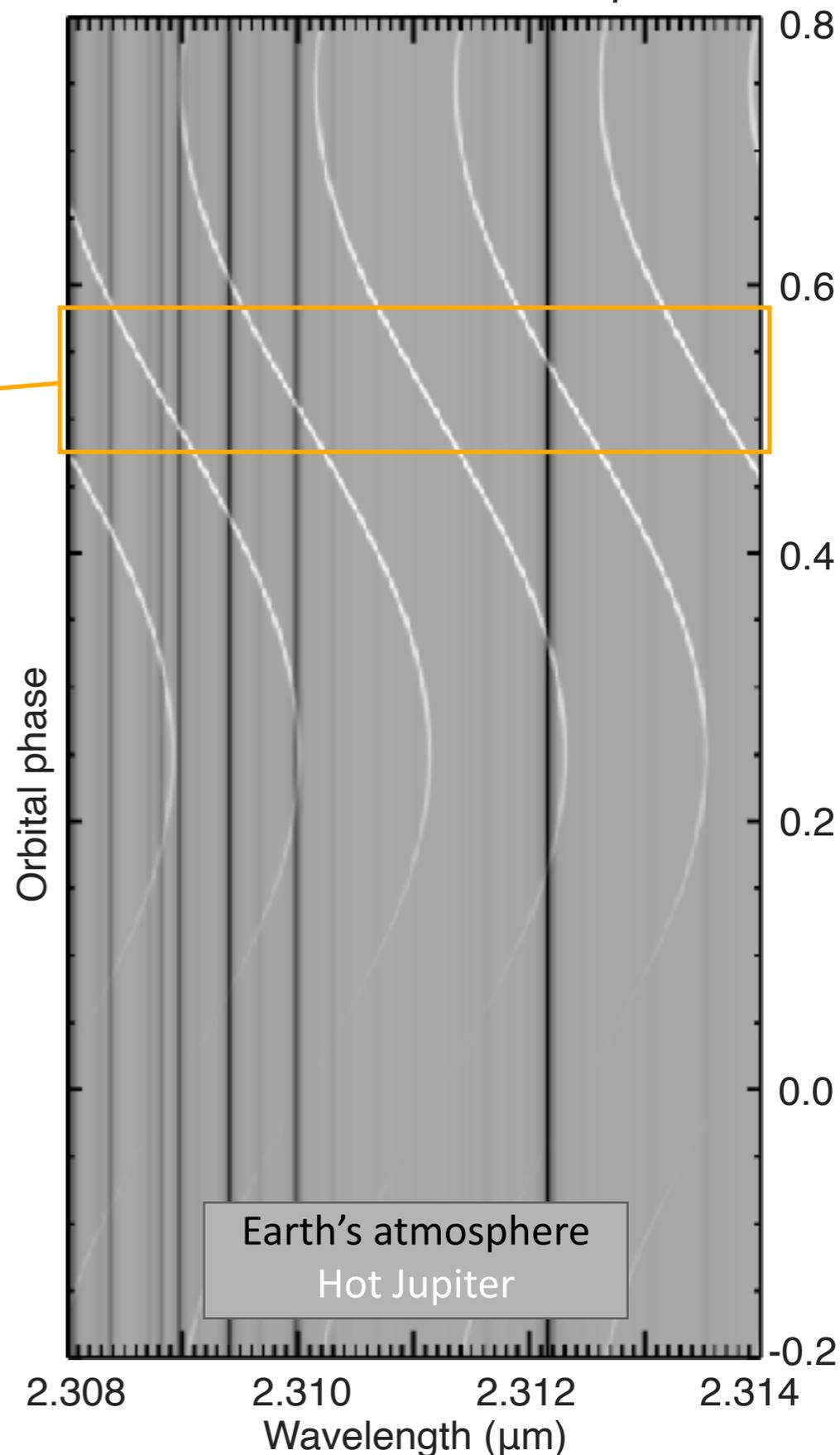
Dayside spectroscopy applicable to **non-transiting planets!**



The first and only method to study atmospheres of most non-transiting planets (evolved, on close-in orbits)

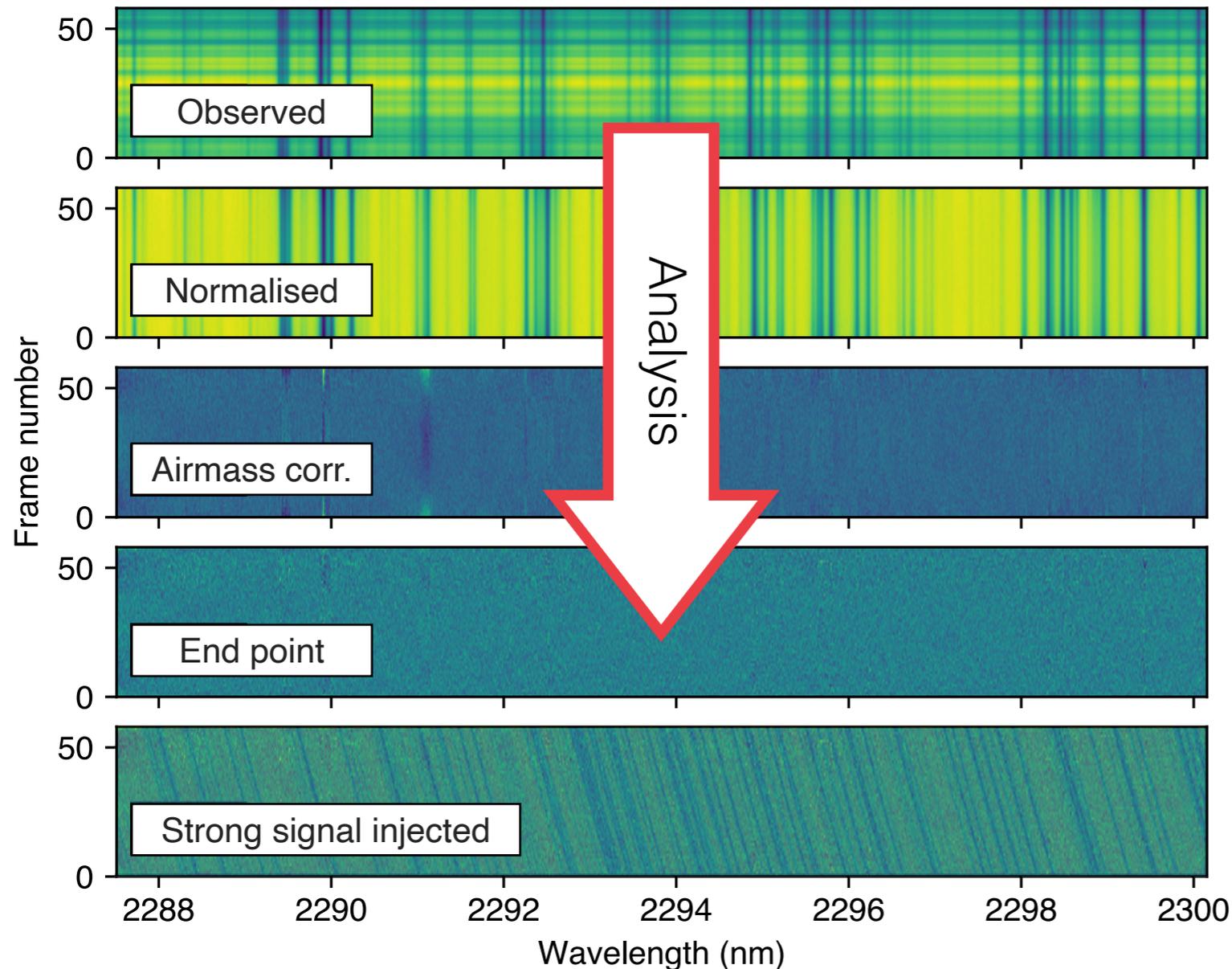
Can solve for **planet mass** and orbital **inclination**

Carbon monoxide - 2.3 μm



Processing ground-based high-res spectroscopy

Every spectral line stationary in wavelength is removed
(check my lecture on Monday)



Time-correlated effects
(transparency, throughput, etc.)
will be “in common-mode”
between spectral channels

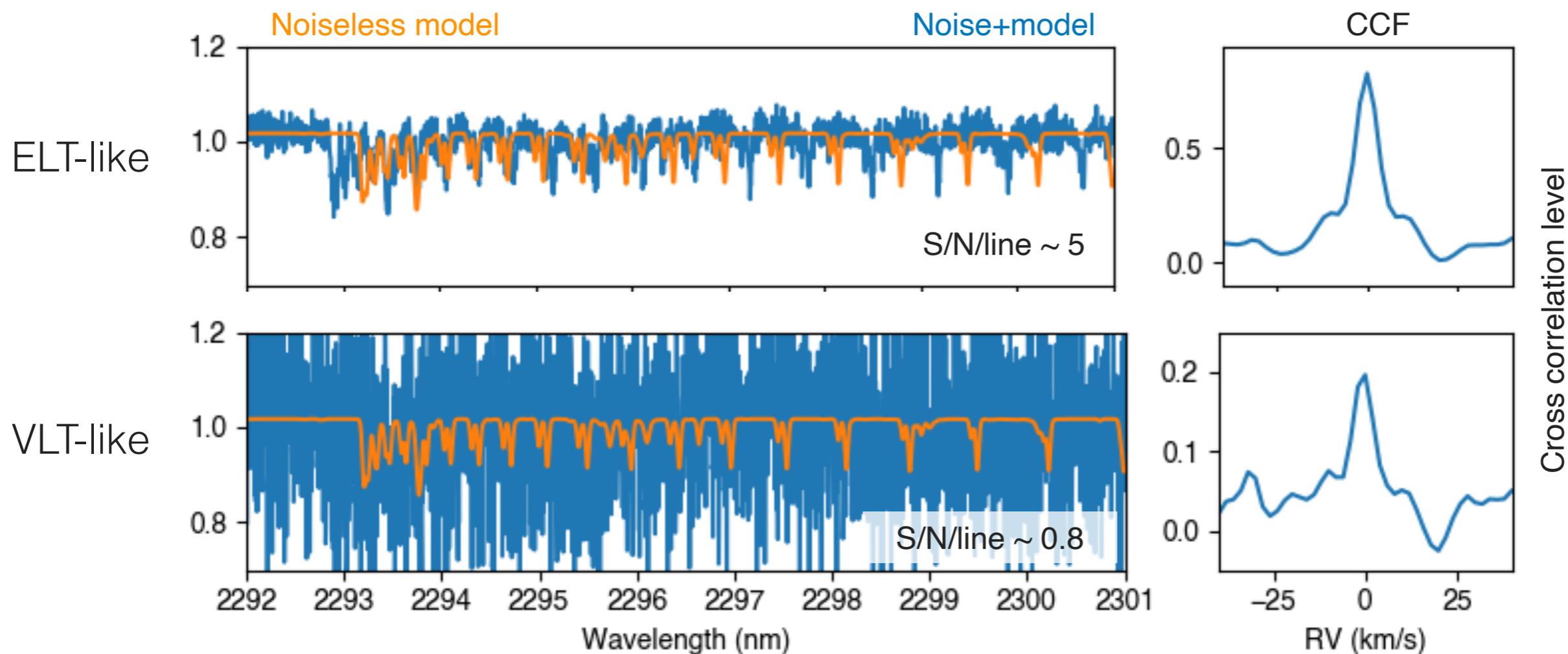
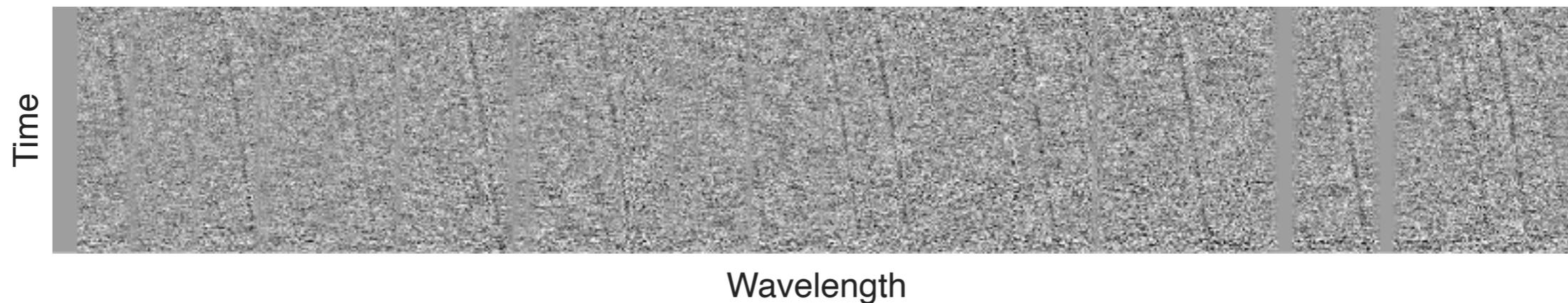
The analysis progressively
“normalises” these effects

These steps can be “automated”
by algorithms decomposing data
into a linear combination of
eigenvectors (e.g. PCA)

The process “auto-calibrates” the data: no reference star needed
However, broad-band variations are removed

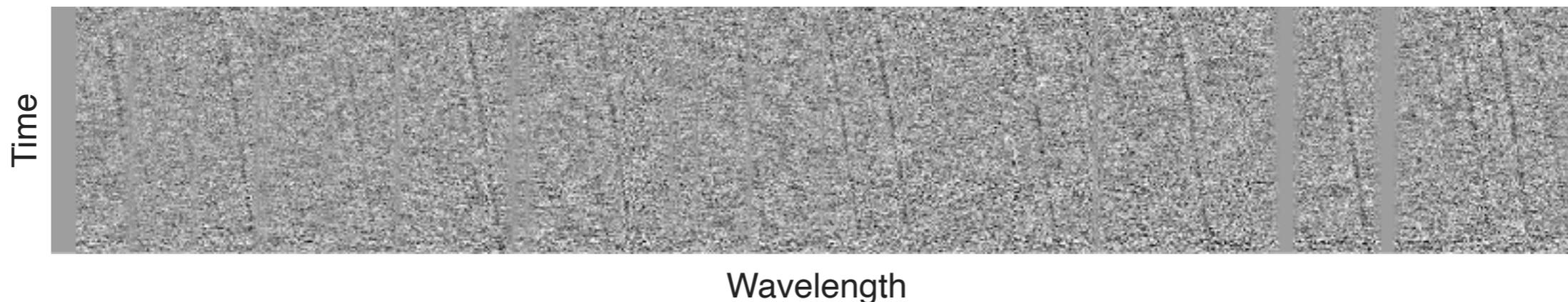
Extracting the (faint) planet signal: cross correlation

5 hours of real data + 20x planet signal (CO)



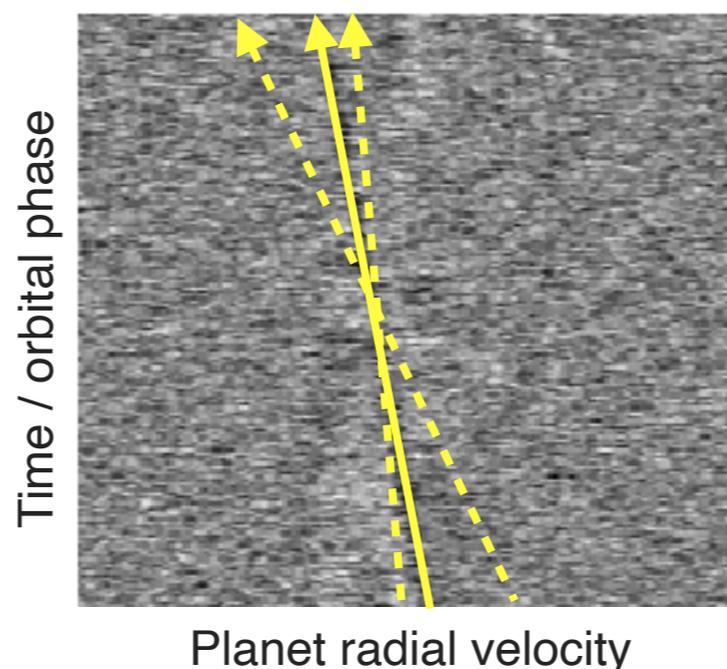
Extracting the (faint) planet signal: cross correlation

5 hours of real data + 20x planet signal (CO)



Cross-correlation with model spectra

Cross-correlation matrix
 $CC(RV, t)$



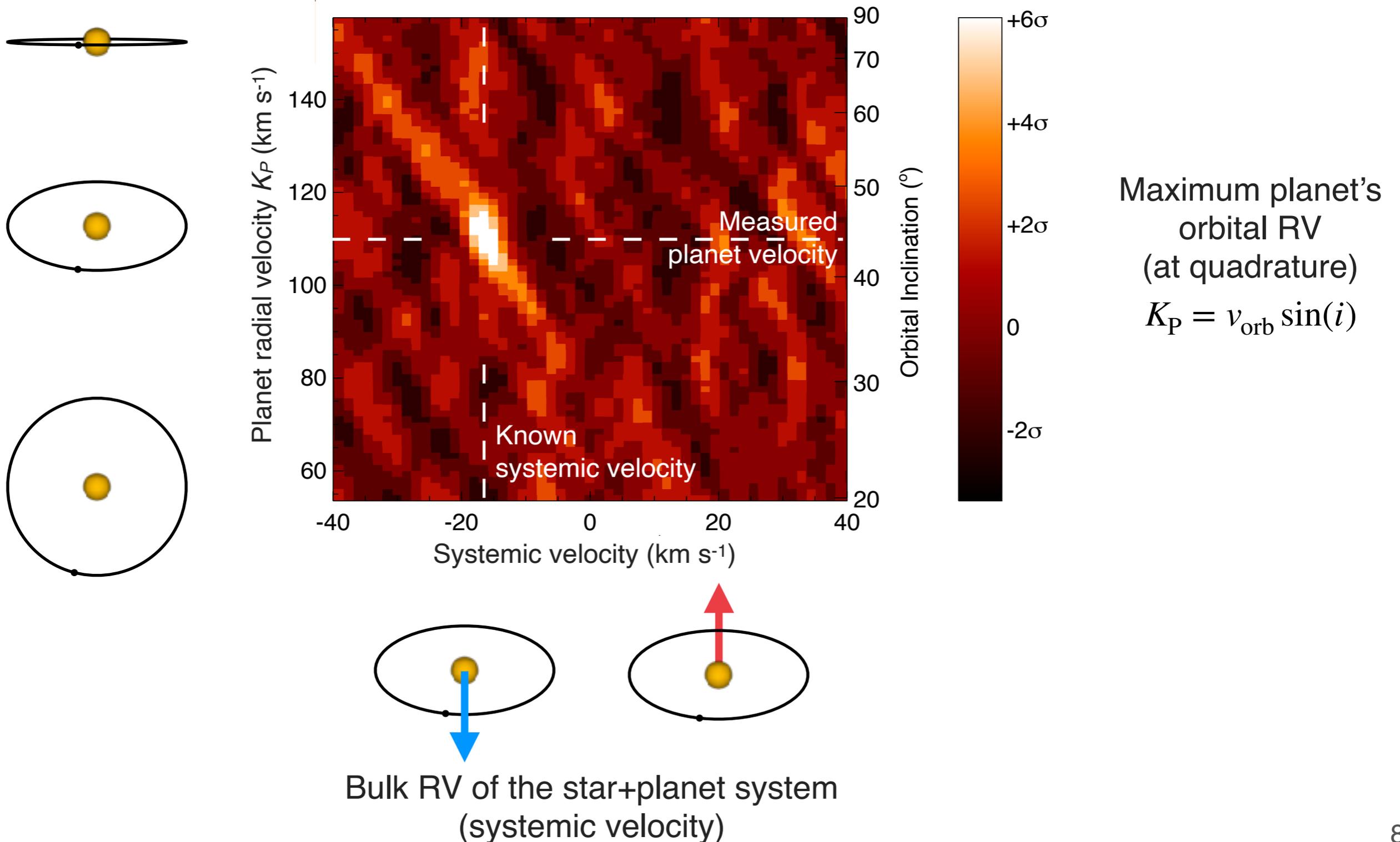
The peak CC tracks
the planet radial
velocity in time

Shifting and co-adding to planet rest-frame
requires knowledge of planet orbital velocity
(two parameters: **slope** and **shift**)

Detections and velocity maps

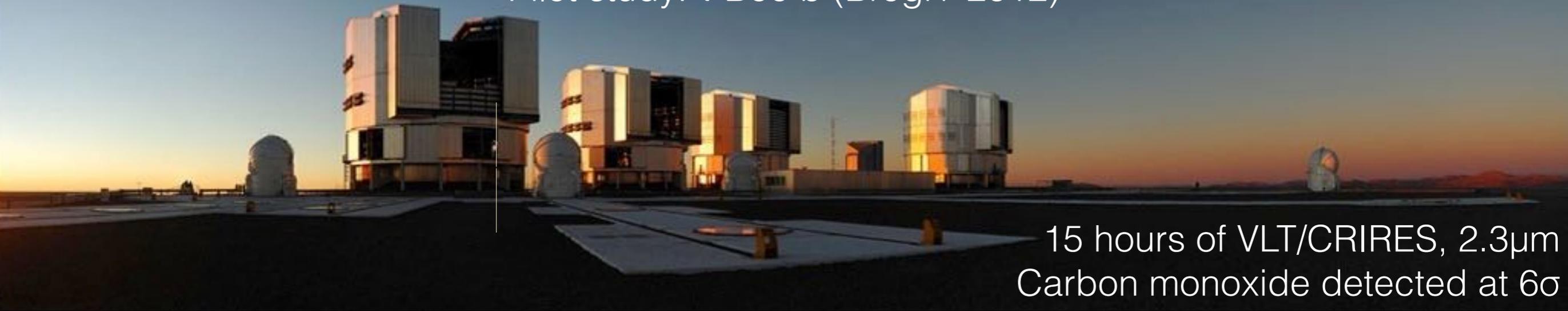
Maximising the cross correlation value as a function of orbital parameters

2 velocities to describe a circular orbit \Rightarrow a 2-dimensional map

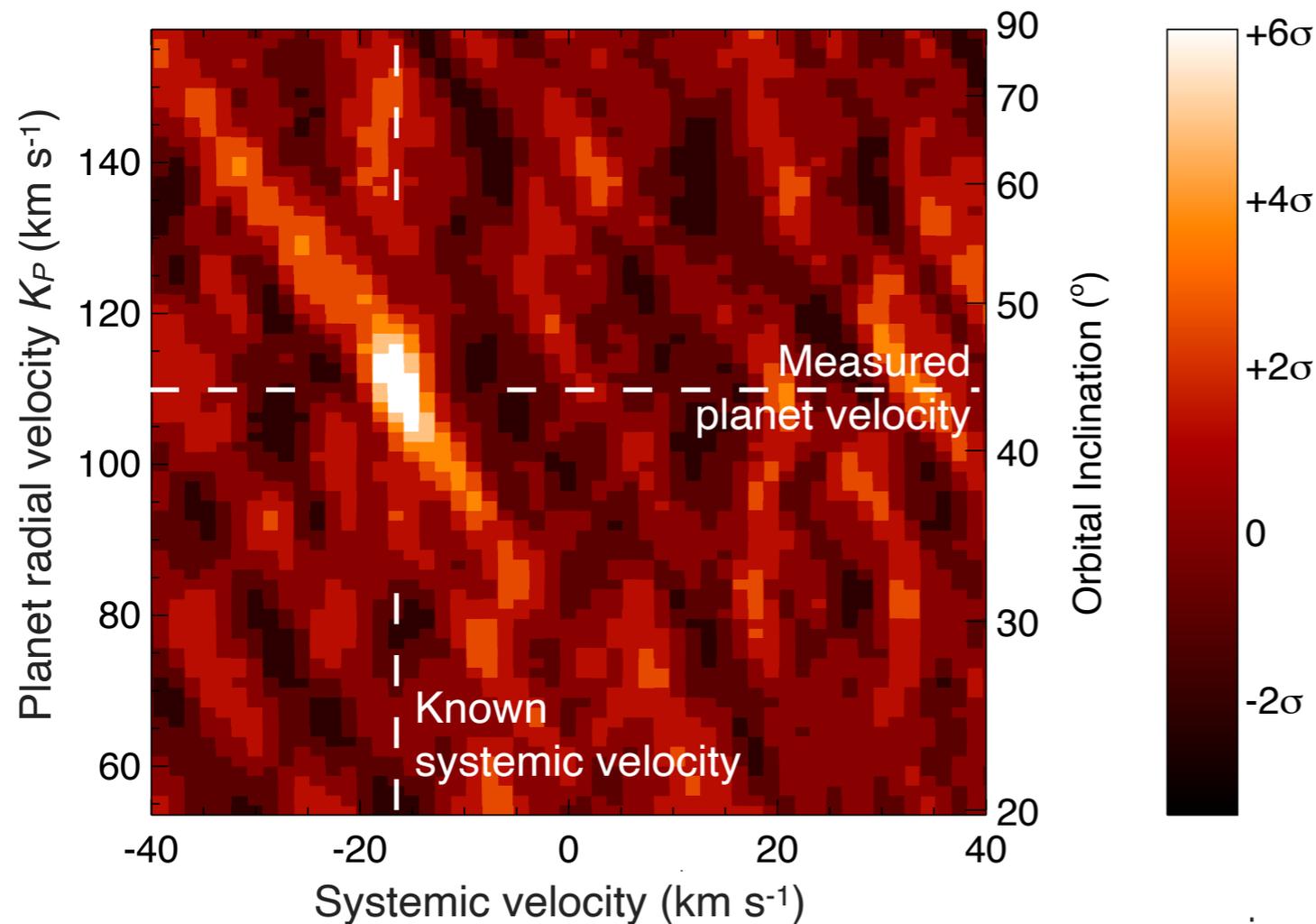


Star and planet as spectroscopic binaries

Pilot study: τ Boo b (Brogi+ 2012)



15 hours of VLT/CRILES, $2.3\mu\text{m}$
Carbon monoxide detected at 6σ



Measured:

RV semi-amplitude ratio: K_P/K_S
 \Rightarrow Mass ratio: M_P/M_S

Inferred:

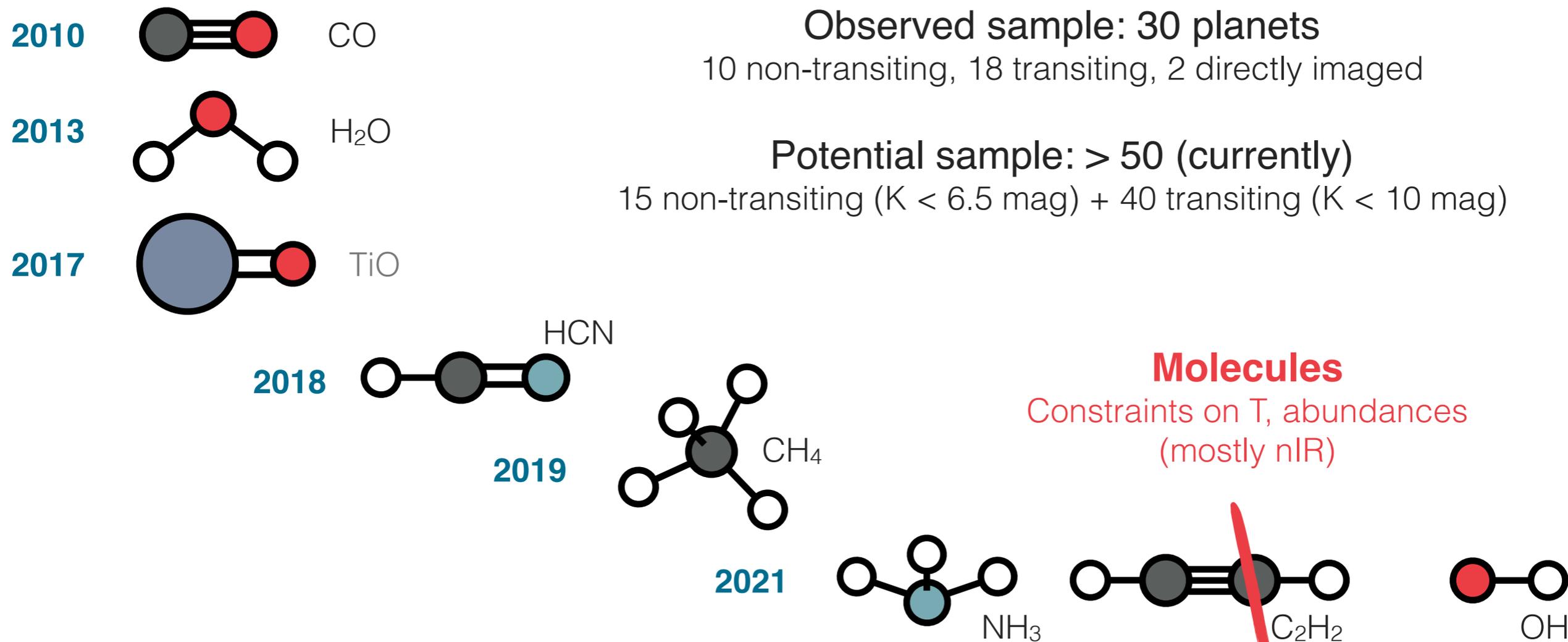
Orbital inclination i
Planet mass $M_P = f(M_S)$

Uncertainties in planet mass
dominated by uncertainties in
stellar mass.

For τ Boo b:

$$i = (45.5 \pm 1.5) \text{ deg}, M_P = (5.95 \pm 0.28) M_{\text{Jup}}$$

The chemical inventory at high spectral resolution



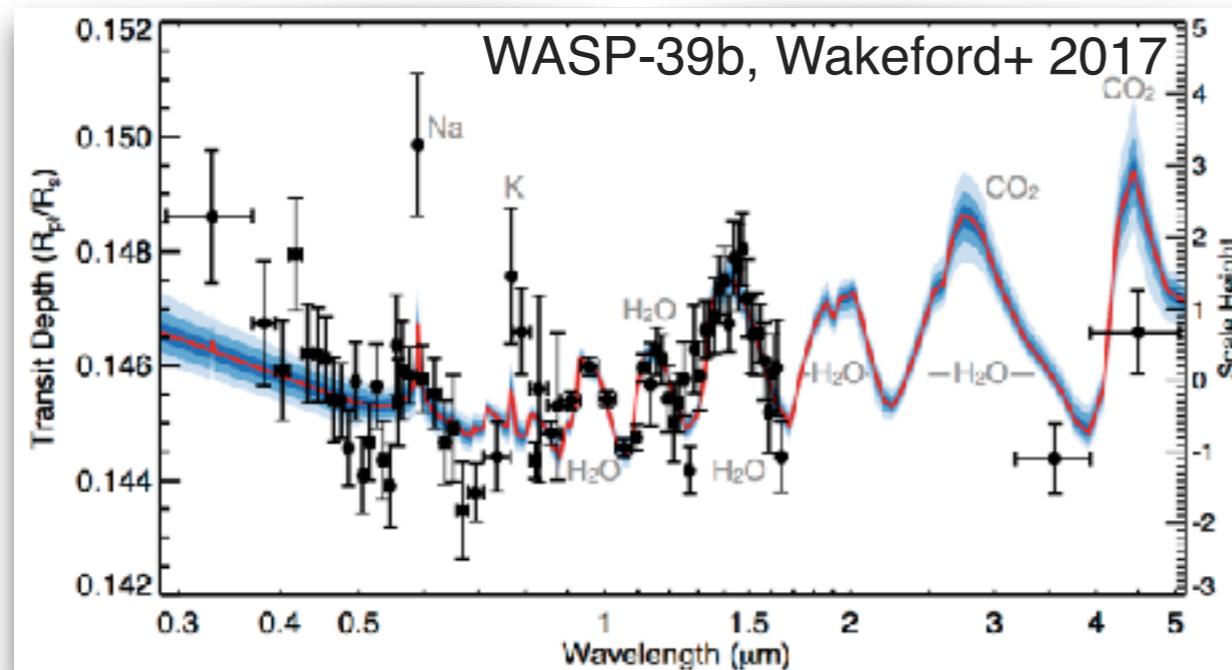
Atoms
Probing higher altitude incl. exospheres, escape, etc. (mostly optical)

1 IA H Hydrogen 1.008	2 IIA Li Lithium 6.94	Be Beryllium 9.0121831	3 IIIB Sc Scandium 44.955908	4 IVB Ti Titanium 47.887	5 VB V Vanadium 50.9415	6 VIB Cr Chromium 51.9961	7 VIIB Mn Manganese 54.938044	8 VIIB Fe Iron 55.845	9 VIIB Co Cobalt 58.933194	10 VIIB Ni Nickel 58.6934	11 IB Cu Copper 63.546	12 IIB Zn Zinc 65.38	13 IIIA B Boron 10.81	14 IVA C Carbon 12.011	15 VA N Nitrogen 14.007	16 VIA O Oxygen 15.999	17 VIIA F Fluorine 18.99840323	18 VIIA Ne Neon 20.1797	19 K Potassium 39.0983	20 Ca Calcium 40.078	21 Sc Scandium 44.955908	22 Ti Titanium 47.887	23 V Vanadium 50.9415	24 Cr Chromium 51.9961	25 Mn Manganese 54.938044	26 Fe Iron 55.845	27 Co Cobalt 58.933194	28 Ni Nickel 58.6934	29 Cu Copper 63.546	30 Zn Zinc 65.38	31 Ga Gallium 69.723	32 Ge Germanium 72.630	33 As Arsenic 74.921595	34 Se Selenium 78.971	35 Br Bromine 79.904	36 Kr Krypton 83.798	18 VIIIA He Helium 4.002602
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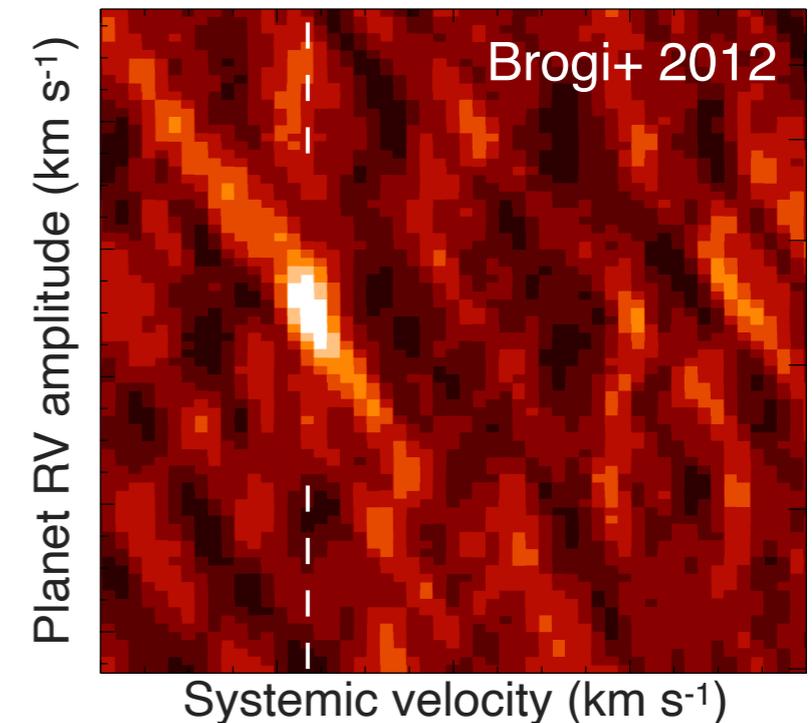
From detecting to measuring: detection significance

Quantifying the “goodness of fit” of a model is not (yet) possible at high-res

Low-res spectroscopy



High-res spectroscopy



Low-res spectroscopy recovers an actual **spectrum**

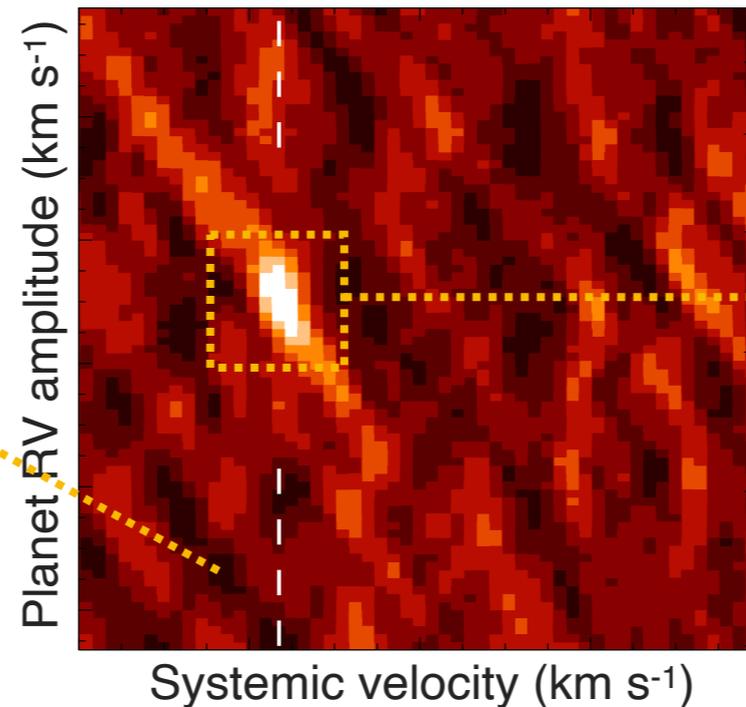
Models can be matched to observations via chi-square fitting (also in a Bayesian way)

High-resolution spectroscopy measures a **level of correlation**

***How do we even quantify significance?
How do we “select” models?***

S/N as a proxy for detection significance

Noise: the standard deviation of all the other cross correlation values



Signal: the peak value of the total cross correlation

$$S/N = \text{Peak CC} / \text{stdev}(\text{CC})$$

Immediate and intuitive quantity to compute

Some of the “noise” is actually auto-correlation / aliasing signal

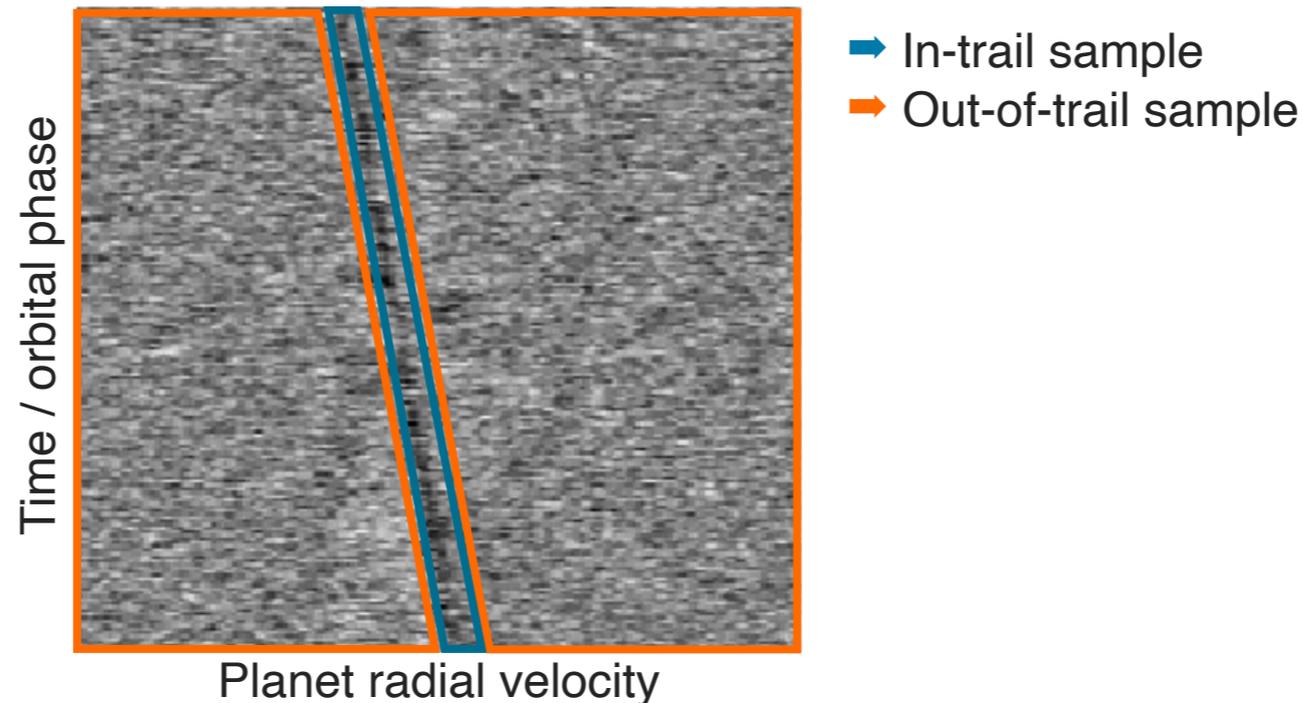
Some (V_{sys}, K_P) values will have increased noise due e.g. to residual telluric or stellar lines

At low SNR peaks can arise by just noise fluctuations

Error bars are usually defined by (V_{sys}, K_P) values corresponding to $(S/N)_{\text{max}} - 1$

Detection significance from statistical tests on the CCFs

Testing the means of the in-trail and out of trail cross-correlation values



Null hypothesis H_0 : in-trail and out-of-trail sample have the same mean

Welch t-test (data samples can have \neq size and variance) used to reject H_0
p-value \Rightarrow detection significance σ

Hp #1: the cross correlation values follow a Gaussian distribution (usually true)

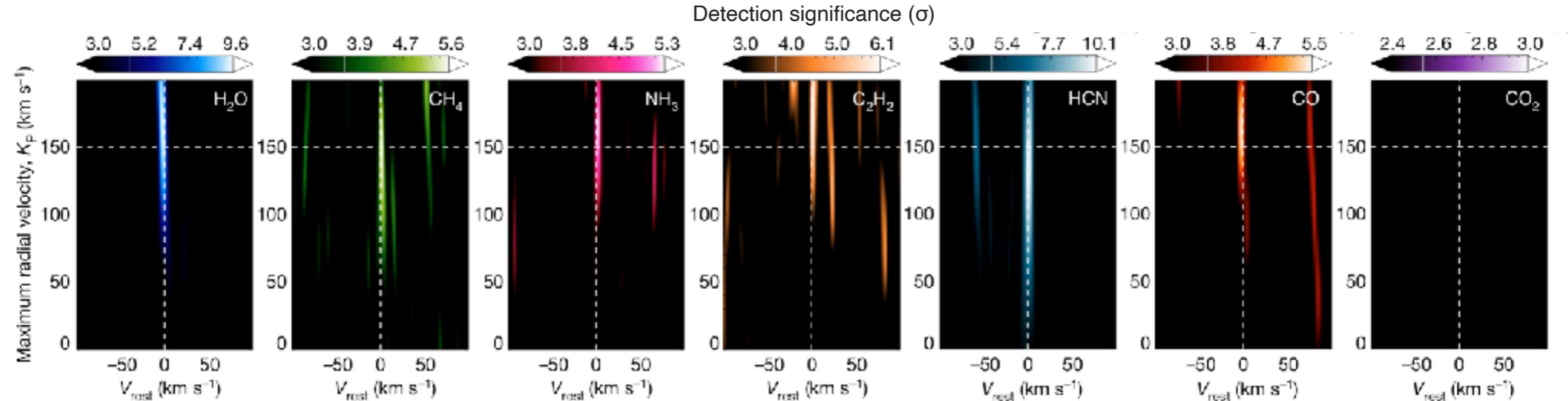
Hp #2: the cross correlation values are independent (depends on RV sampling)

Dependence on the “width” of the in-trail sample (at least 1 FWHM)

n- σ error bars can correctly be determined as $\sigma_{\max} - n$

Five carbon- and nitrogen-bearing species in a hot giant planet's atmosphere

P. Giacobbe, M. Brogi, S. Gandhi et al., *Nature* **592**, 205-208 (2021)



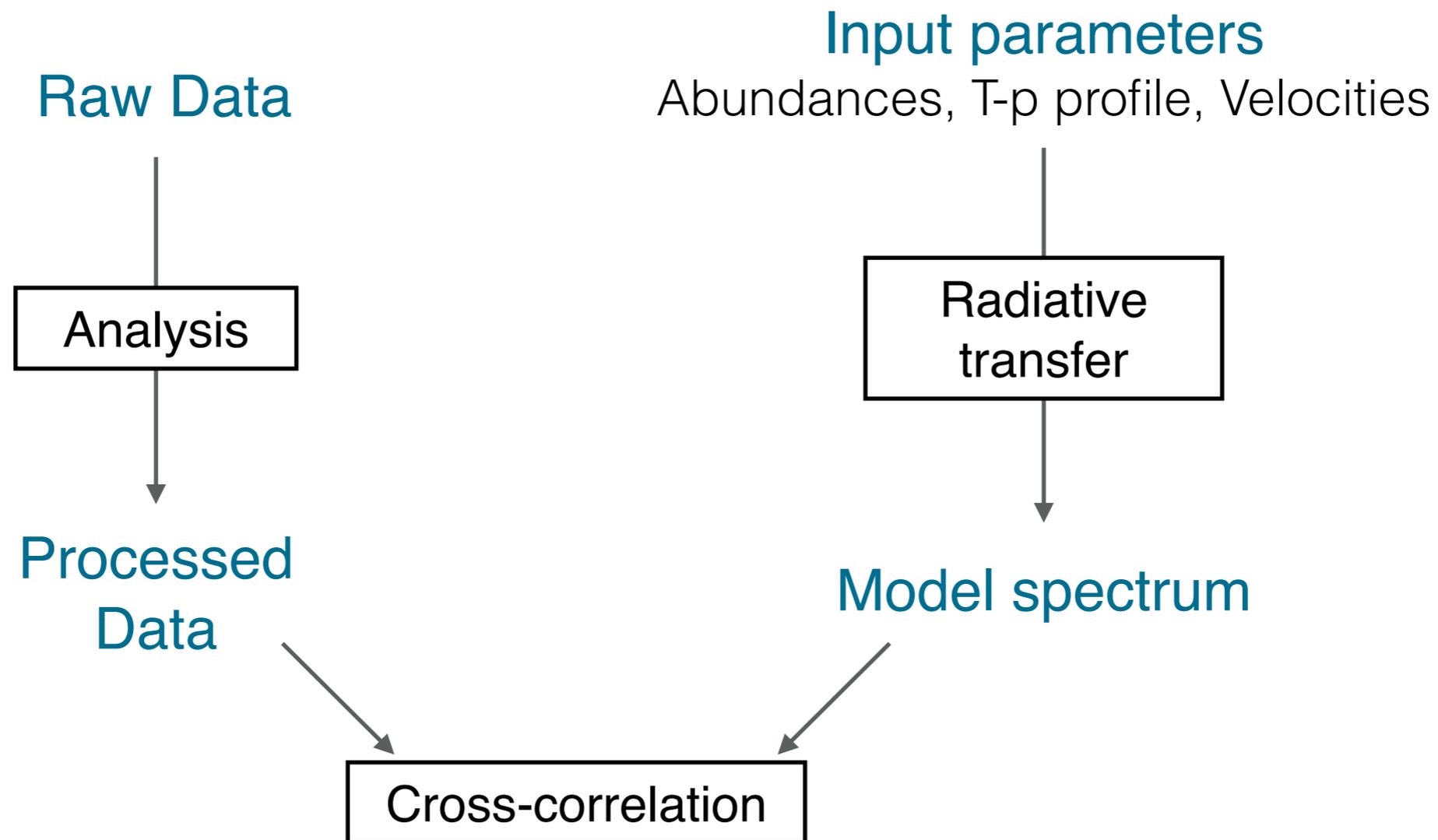
4 transits of hot Jupiter HD 209458b (1,500K) \Rightarrow H₂O + 5 species simultaneously detected



What does it mean for the atmosphere of HD 209548 b?

Need to move *beyond detecting* and towards *measuring*
(Just hold on for a few more slides)

From detecting to measuring: our checklist

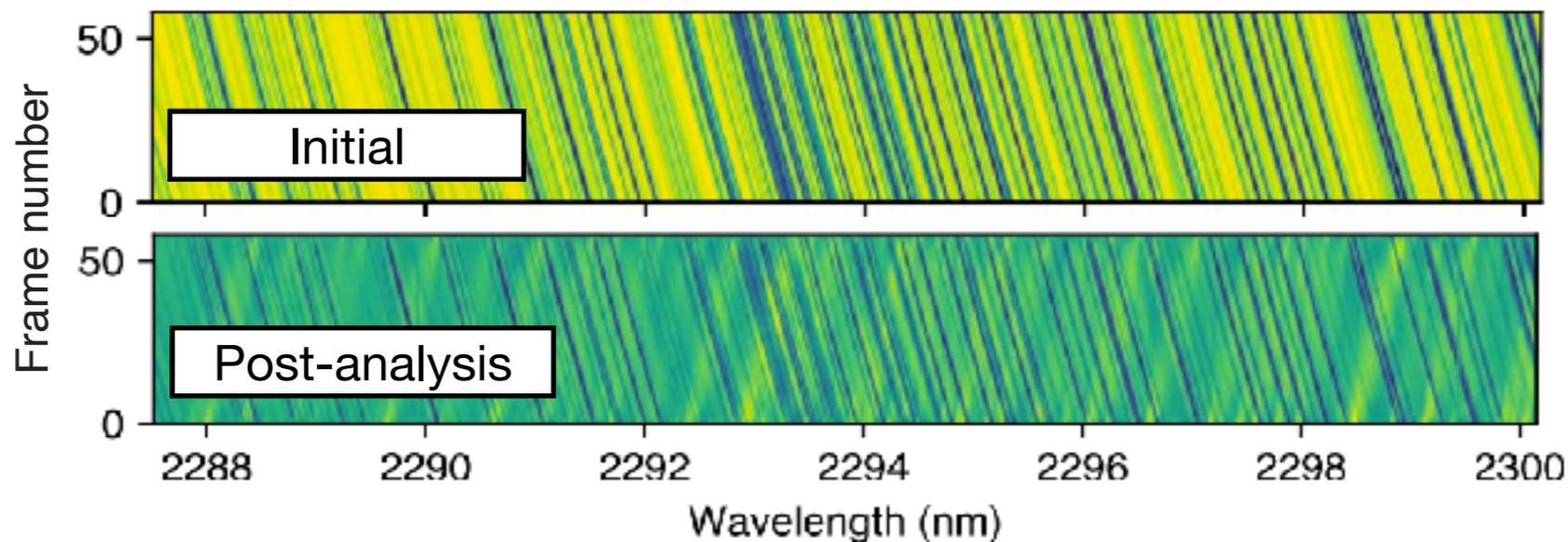


- Need to:**
- account for any biases of the analysis
 - understand what's the information content at high-res
 - design a method to select the best model within a grid
 - explore the whole parameter space to understand degeneracies

The data analysis is not completely harmless

The removal of telluric and stellar lines affects exoplanet lines

Shown by Brogi & Line (2019) on simulated data - easy to see in the noiseless case



Different telluric removal techniques show different biases

see Gibson's talk

(e.g. airmass de-trending, PCA, Sysrem)

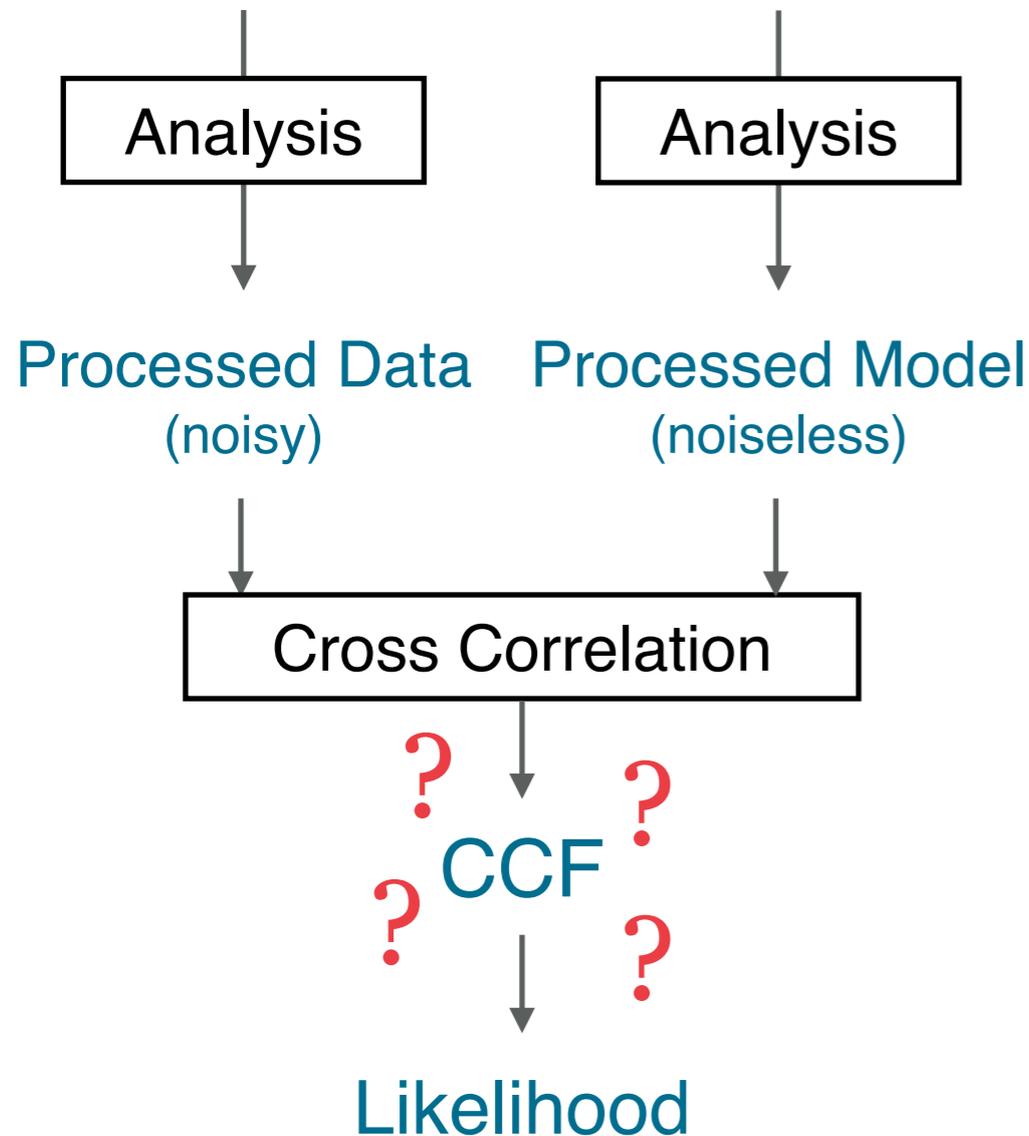
Altered shape & depth of spectral lines \Rightarrow biased abundances and T

Model reprocessing is **unavoidable** to obtain **unbiased measurements** from HRCCS

Model reprocessing: an unavoidable step

The model planet spectrum is **injected** in the data or a **synthetic sequence** is created

Observed Data Modelled Data ← Model spectrum



Can we translate cross correlation into a statistically meaningful quantity (a likelihood)?

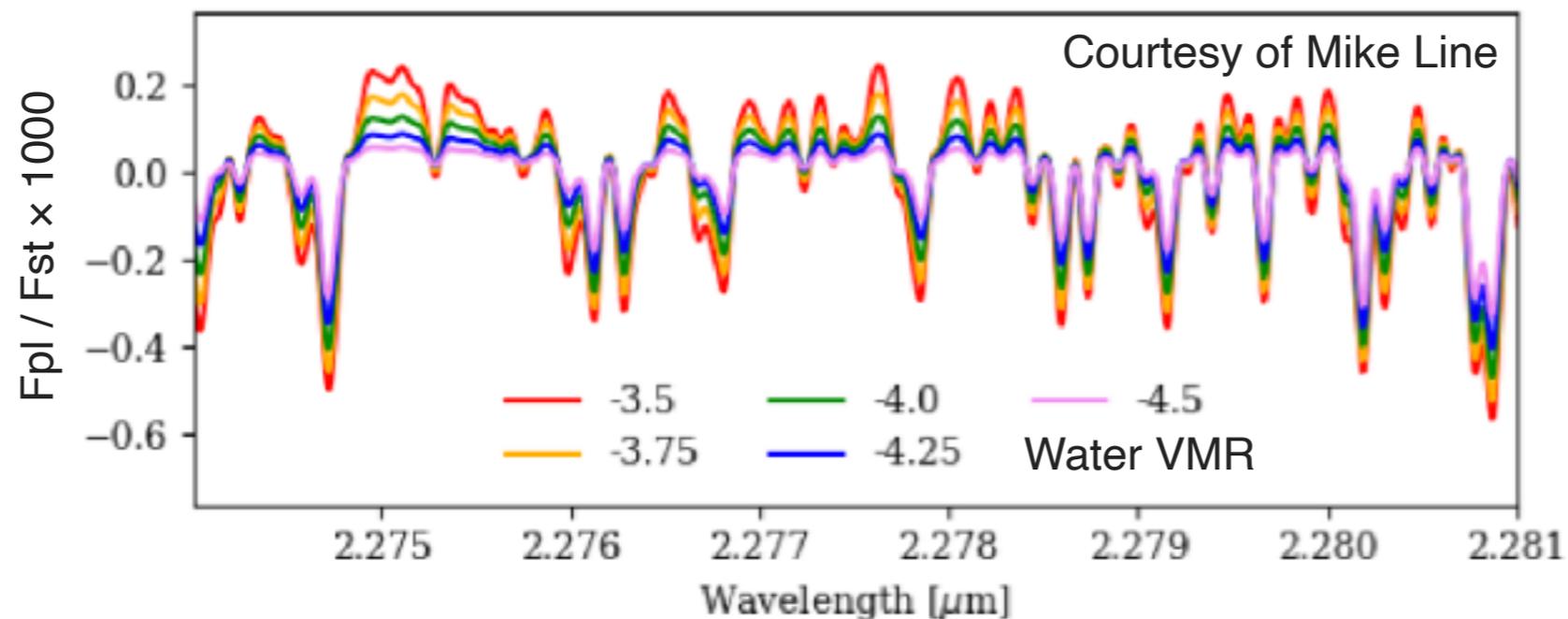
What is the information content in high-res data?

High-res data is normalised to remove stellar & telluric spectrum
(loss of absolute level of continuum in both emission and transmission)

No actual “spectrum” is visible
(no ground truth - consequences for goodness of fit)

Data is still expressed in units of stellar spectrum
(absolute line-to-line and line-to-continuum depths can still be recovered)

Line ratios and line shape change with absolute abundances and temperatures



HRCCS can measure absolute and relative abundances with the right framework

Building a likelihood function for high-res data

Brogi & Line (2019), but also Zucker (2003) and Gibson et al. (2020)

We would like to:

- use the match in line position
- distinguish between +ve and -ve correlation
- use information about line shape and amplitude

$$\log(L) = -\frac{N}{2} \log [s_f^2 - 2R(s) + s_g^2].$$

Length of array

Data Variance Cross-covariance Model Variance

Cross correlation

$$C(s) = \frac{R(s)}{\sqrt{s_f^2 s_g^2}}.$$

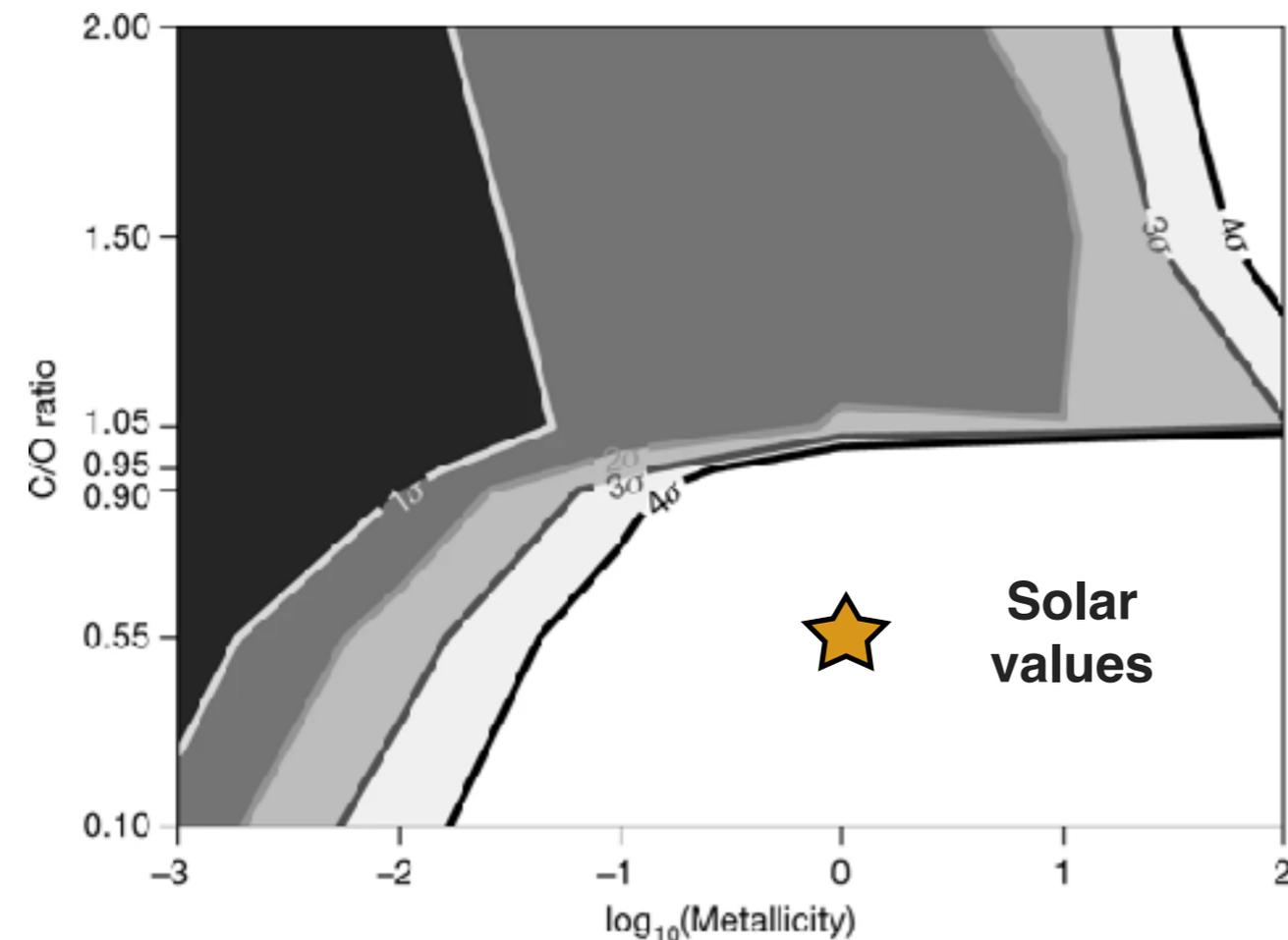
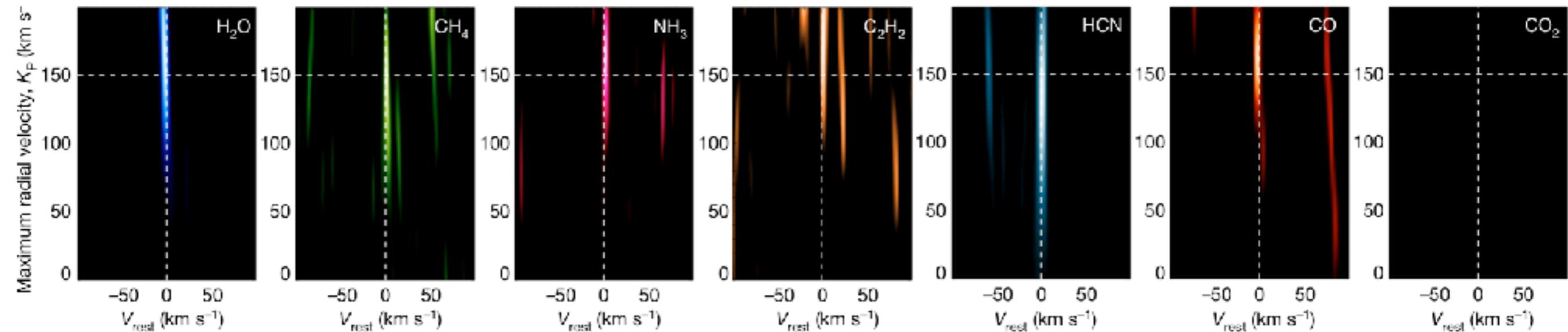
logL contains the **model** and **data variances** s^2
(it accounts for the amplitude of lines)

logL contains the **cross covariance** R
(not normalised - accounts for amplitude of lines)
(penalises anti-correlation - accounts for emission/absorption)

Model selection through likelihood-ratio tests

Exploring a grid of equilibrium models by varying metallicity and C/O

Giacobbe, Brogi, Gandhi et al., *Nature* (2021)



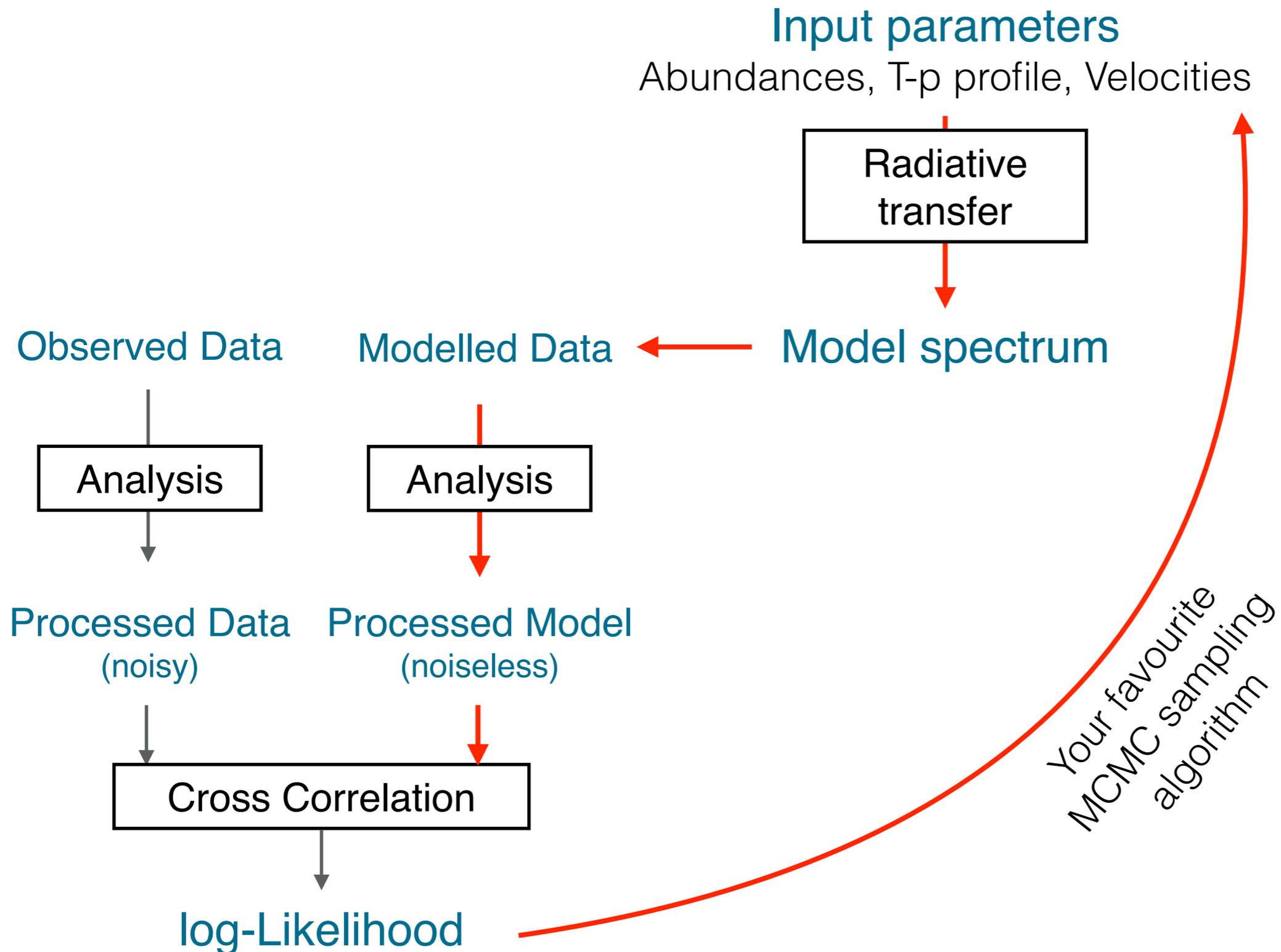
Addition of clouds (with LR parameters)
highly favoured (17 sigma)

Disequilibrium chemistry disfavoured

**HD 209458b formed beyond the snow line and
subsequently migrated w/o accreting ice
planetesimals**

Running a Bayesian retrieval on HR data

Letting the data “inform” model selection to explore full parameter space

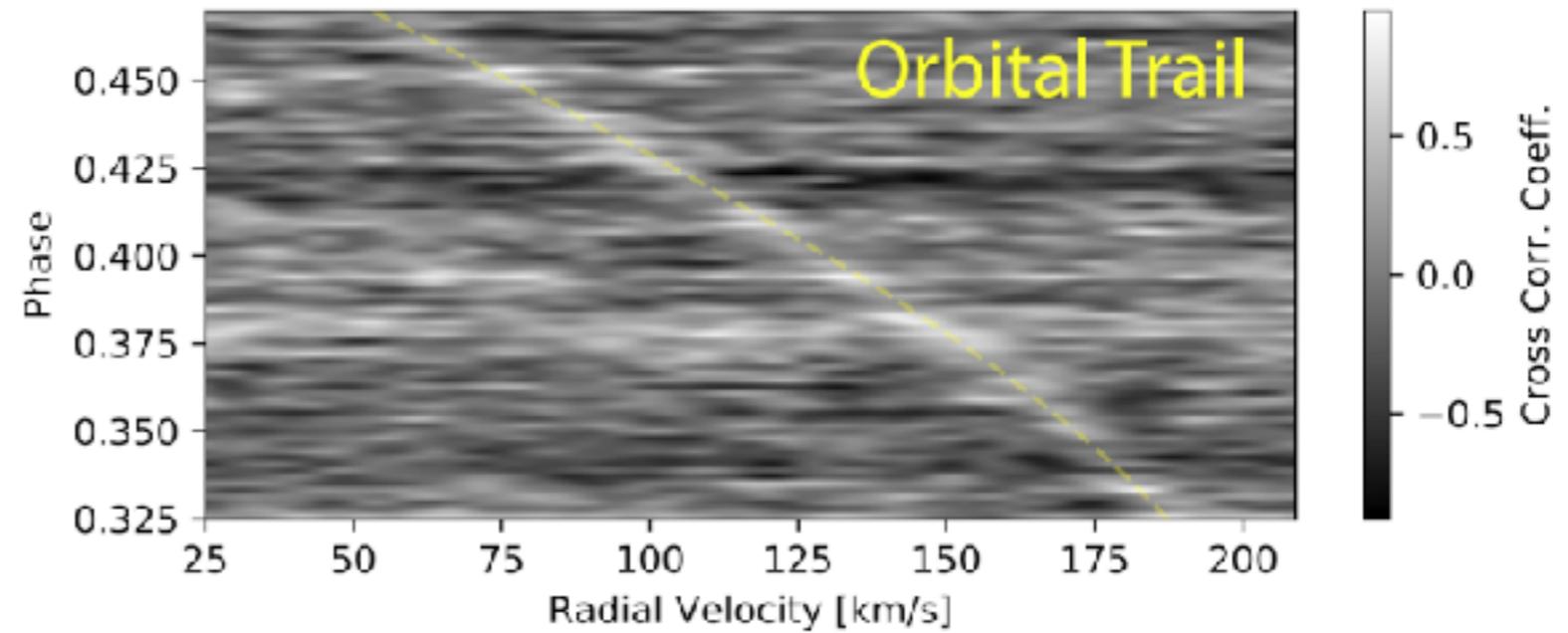


The emission spectrum of WASP-77 A b

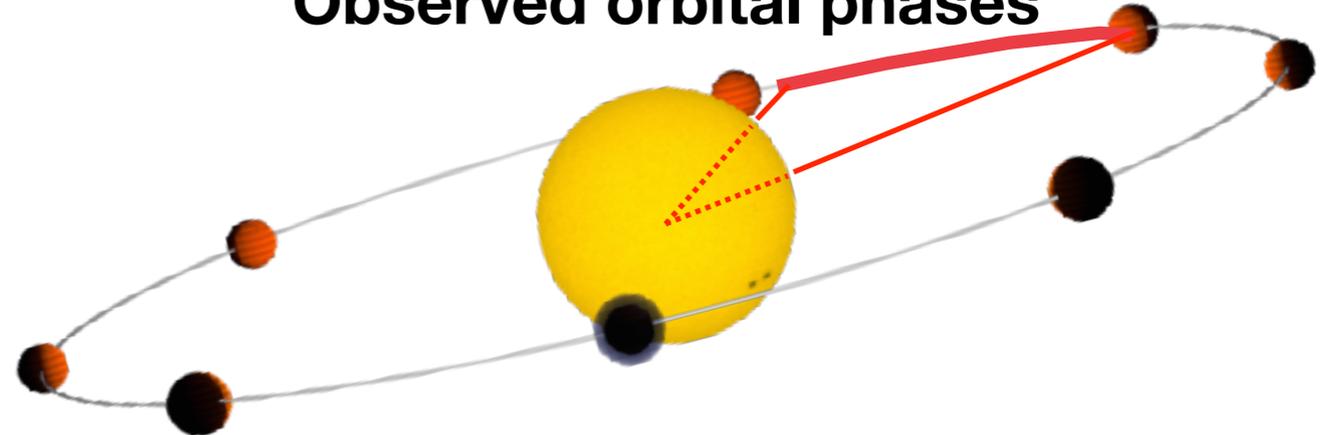
Line, Brogi, Gandhi et al., *Nature*, accepted (coming soon!)



IGRINS@Gemini-S (8.1m)



Observed orbital phases

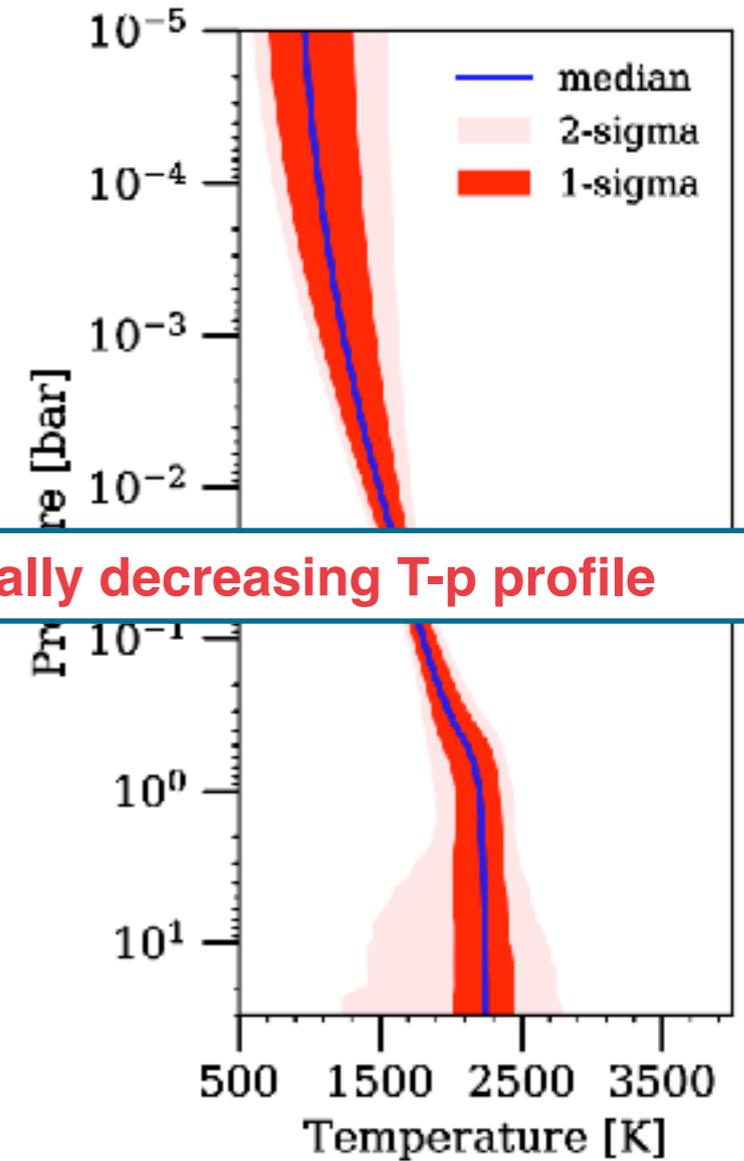
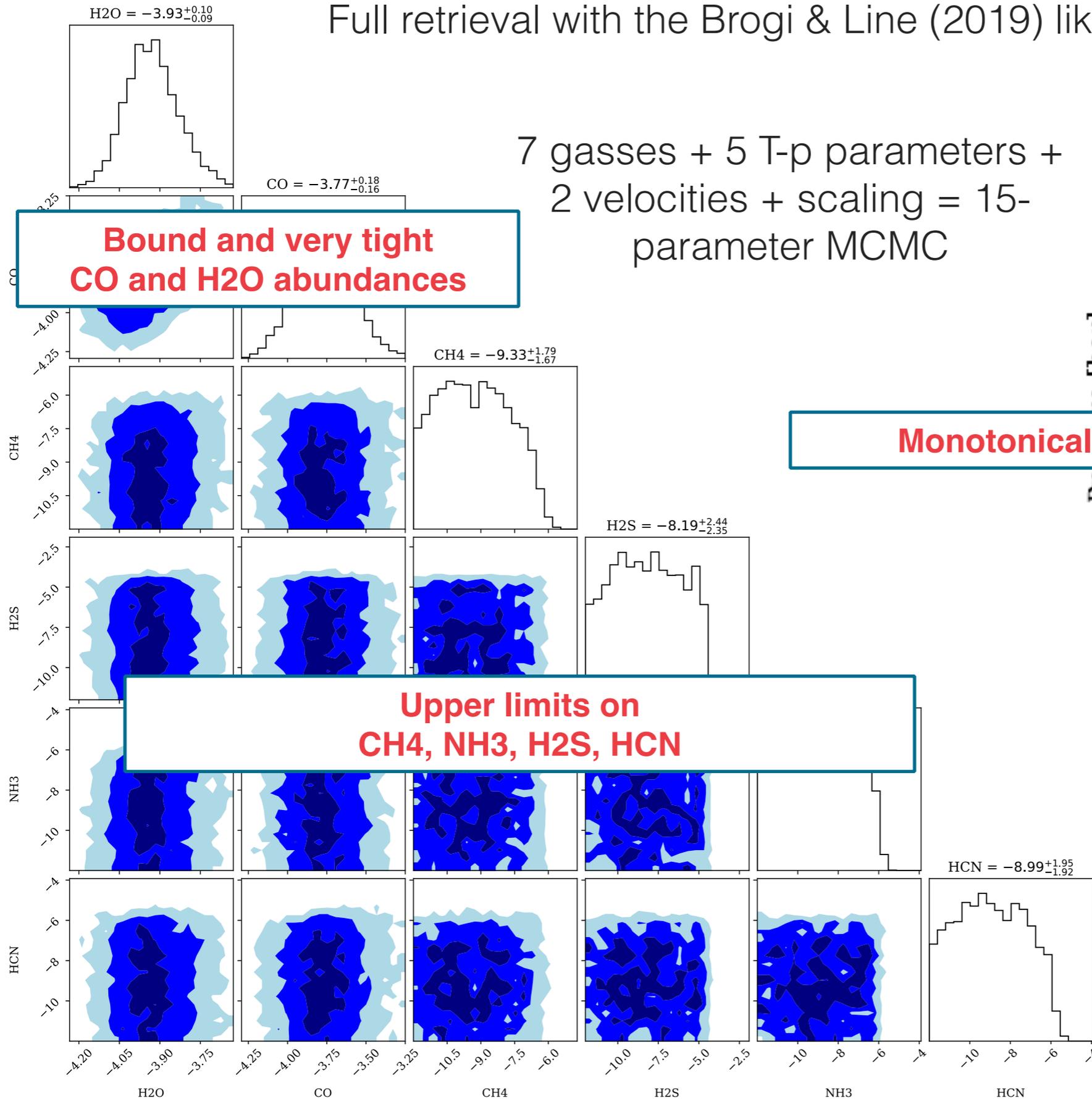


R~45,000
1.45 - 2.45 μm simultaneously
Silicon immersion grating
(keeping the instrument compact)

Achieving “solar system” precisions in the chemistry

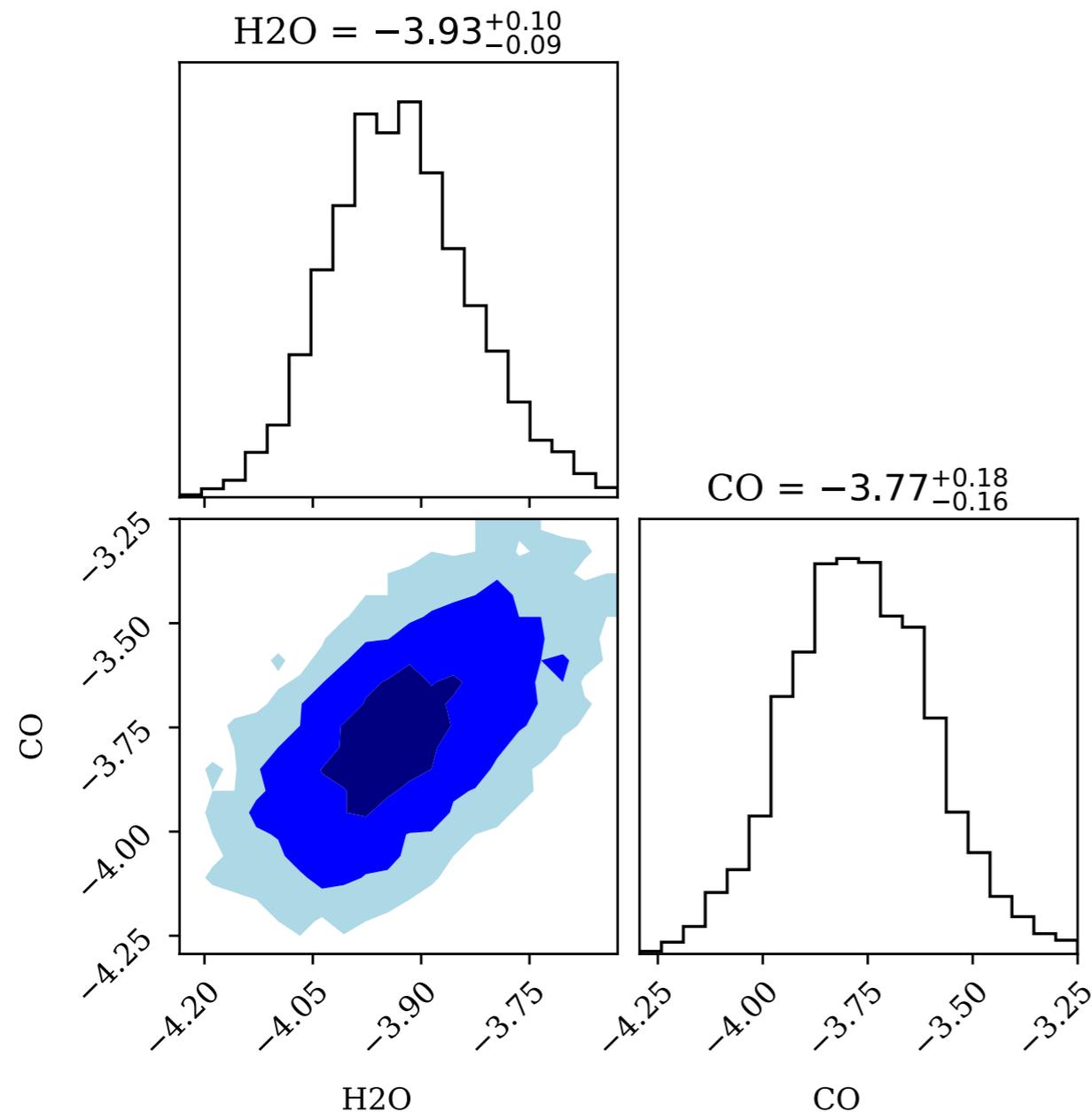
Full retrieval with the Brogi & Line (2019) likelihood

7 gasses + 5 T-p parameters +
2 velocities + scaling = 15-
parameter MCMC



Achieving JWST precisions in the chemistry

0.1-0.2 dex precision in **absolute abundance** for H₂O and CO



Validated independently with
2 retrieval frameworks
CHIMERA (Line)
GENESIS/HyDRA-H (Gandhi)

Accuracy tested by changing:
Data processing
T-p parametrisation
Choice of line lists

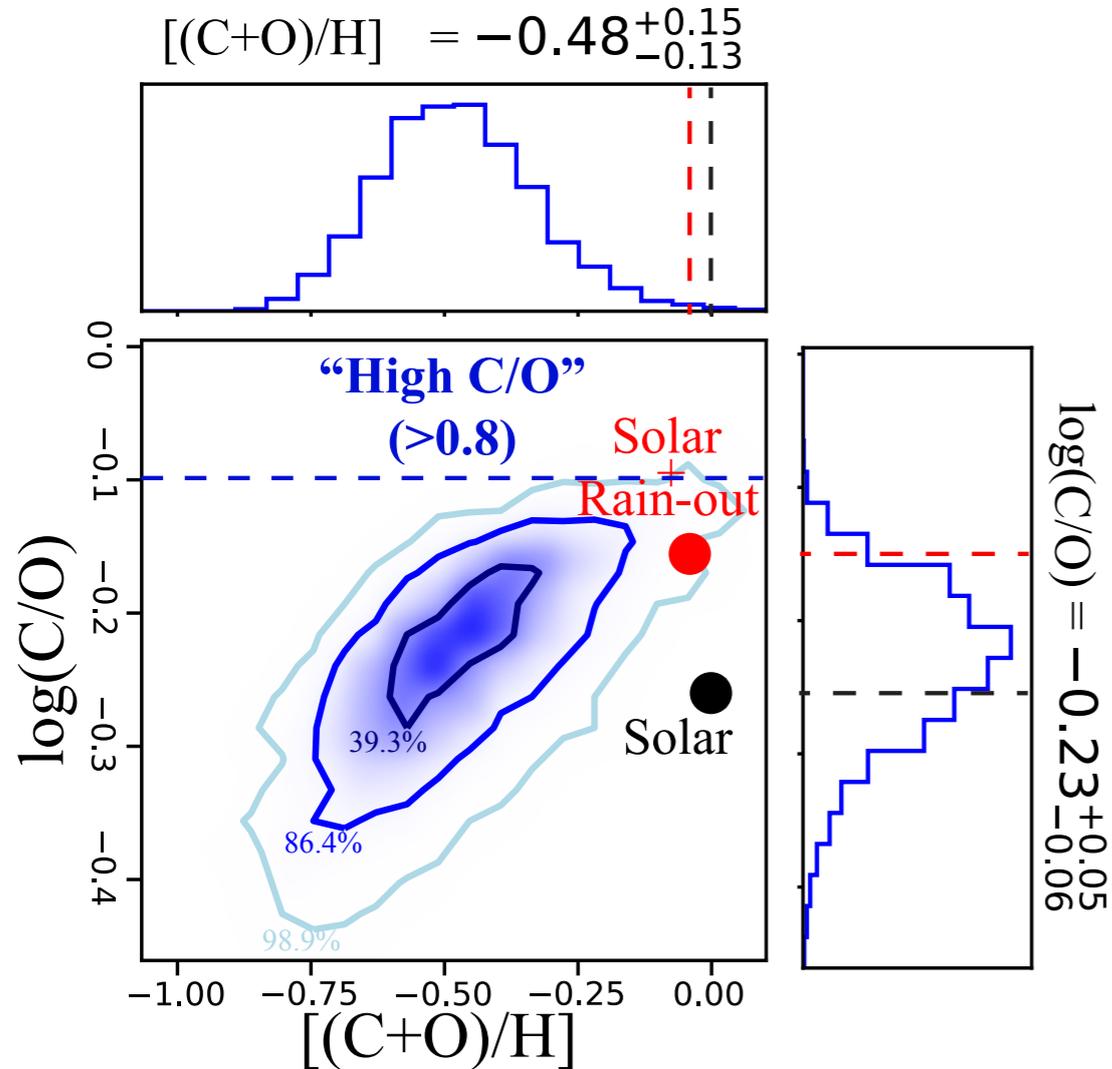
Computationally intensive
1 model evaluation =
5-10s on a single CPU core
(GPU+parallel computing)

What can we do with such precision?

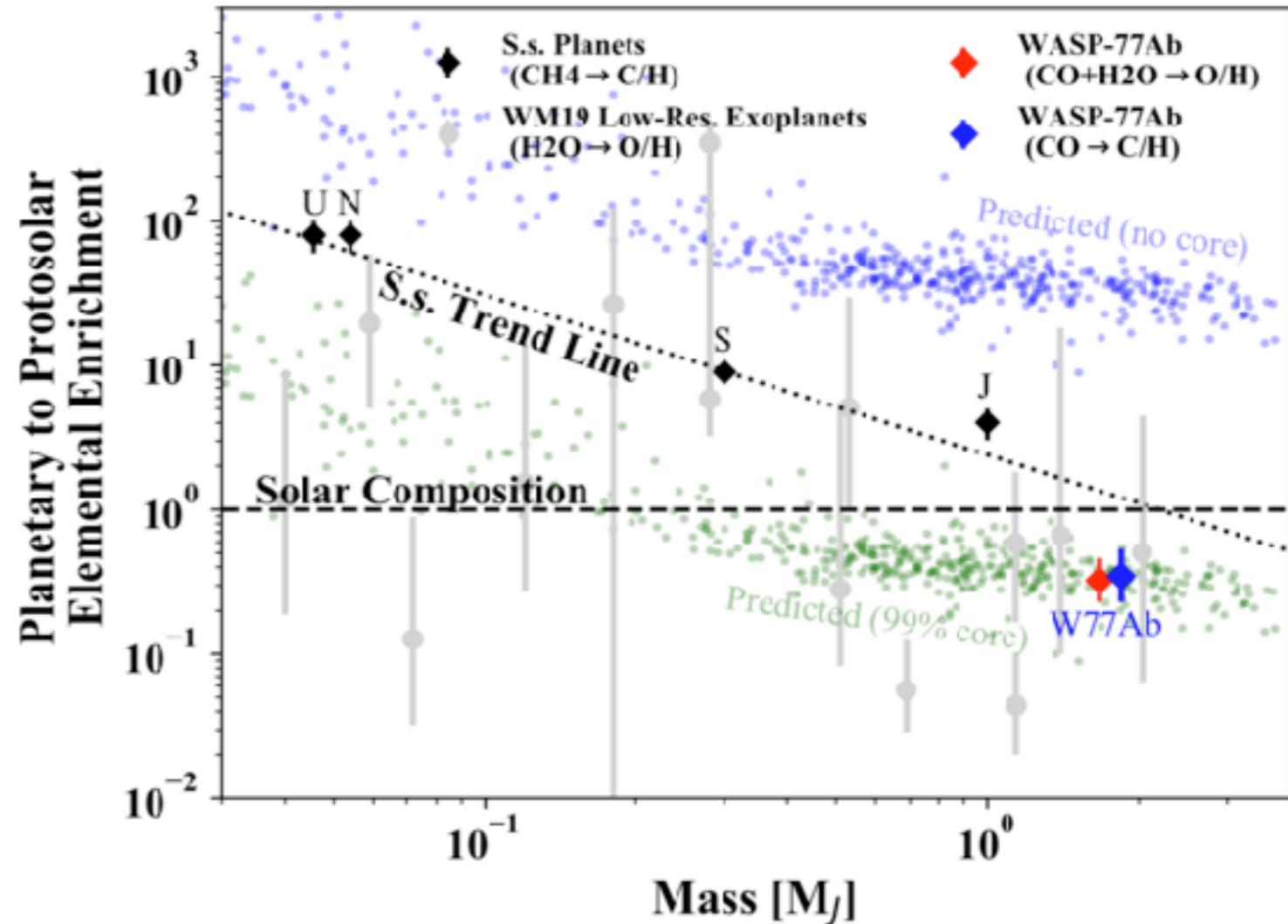
Constraints in the chemistry

WASP-77 A b has sub-solar metallicity but solar C/O

CO+H2O \Rightarrow Metallicity, C/O



Mass-metallicity relation



C/O & metallicity of hot Jupiters can be connected to formation and early evolution scenarios

2021 has seen three measurements of C/O & metallicity (Giacobbe+21; Line+21; Pelletier+21 - see talk!)

A joint analysis of low- and high-resolution spectra

Low res information

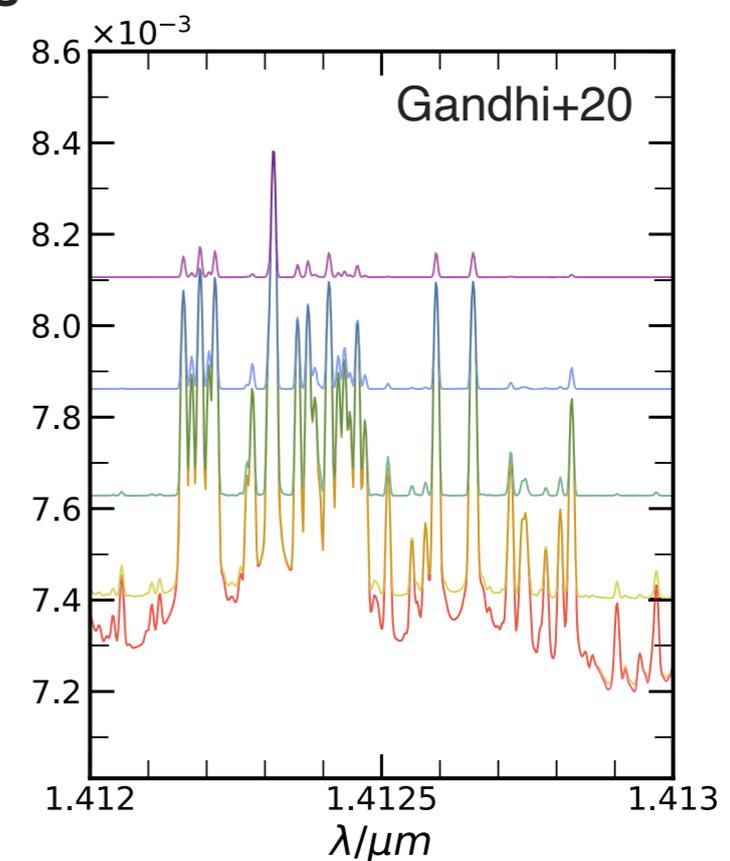
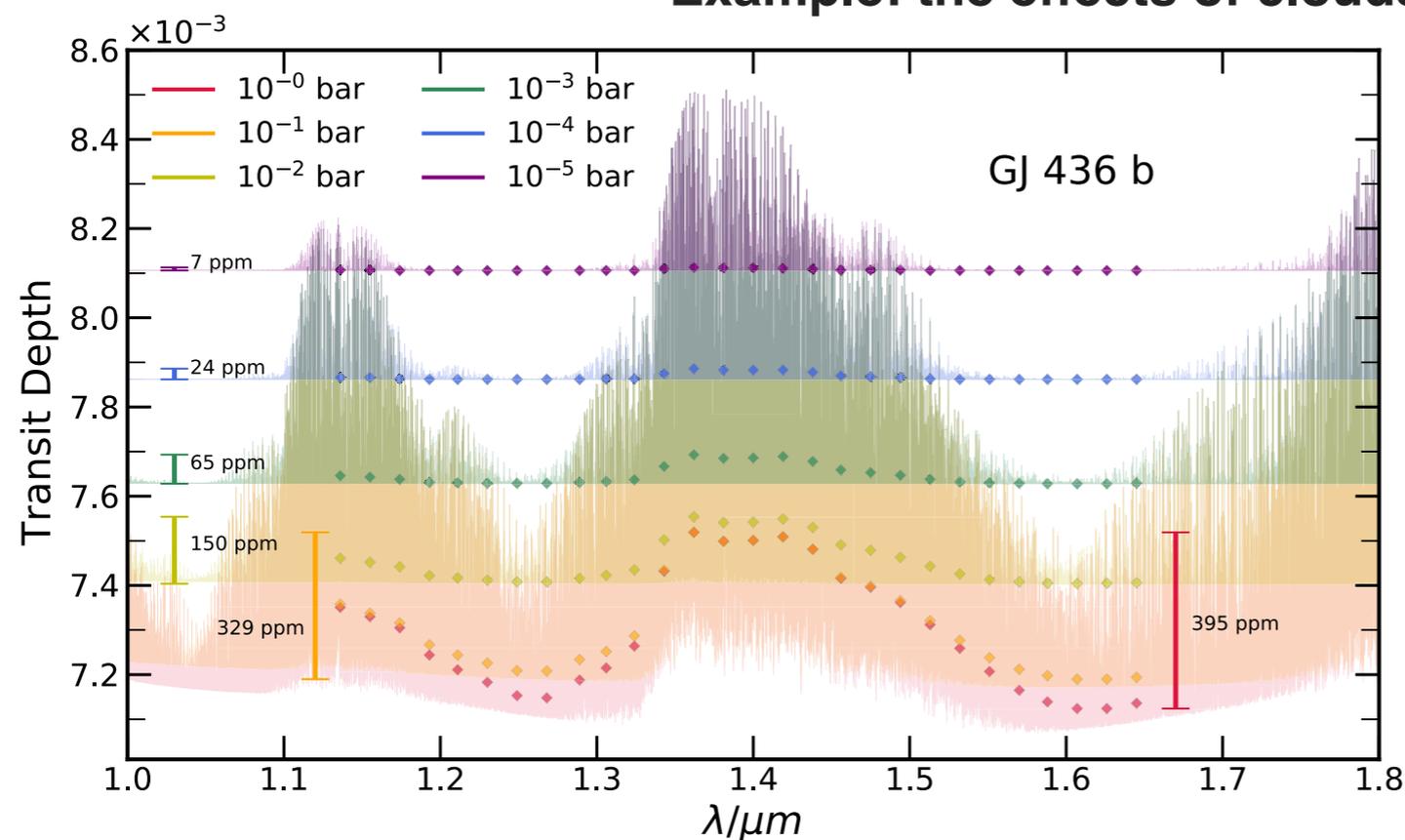
Broad-band variations (the “wiggles” in the spectrum)
Overall transit depth of planet flux (the “level of continuum”)
see Natasha’s and Ryan’s lecture on Tuesday

High-res information

Line-to-line variations, line-to-continuum variations, line shape
see Sid’s lecture on Tuesday

Independently encoding opacity sources, temperature vs pressure, gravity, etc.

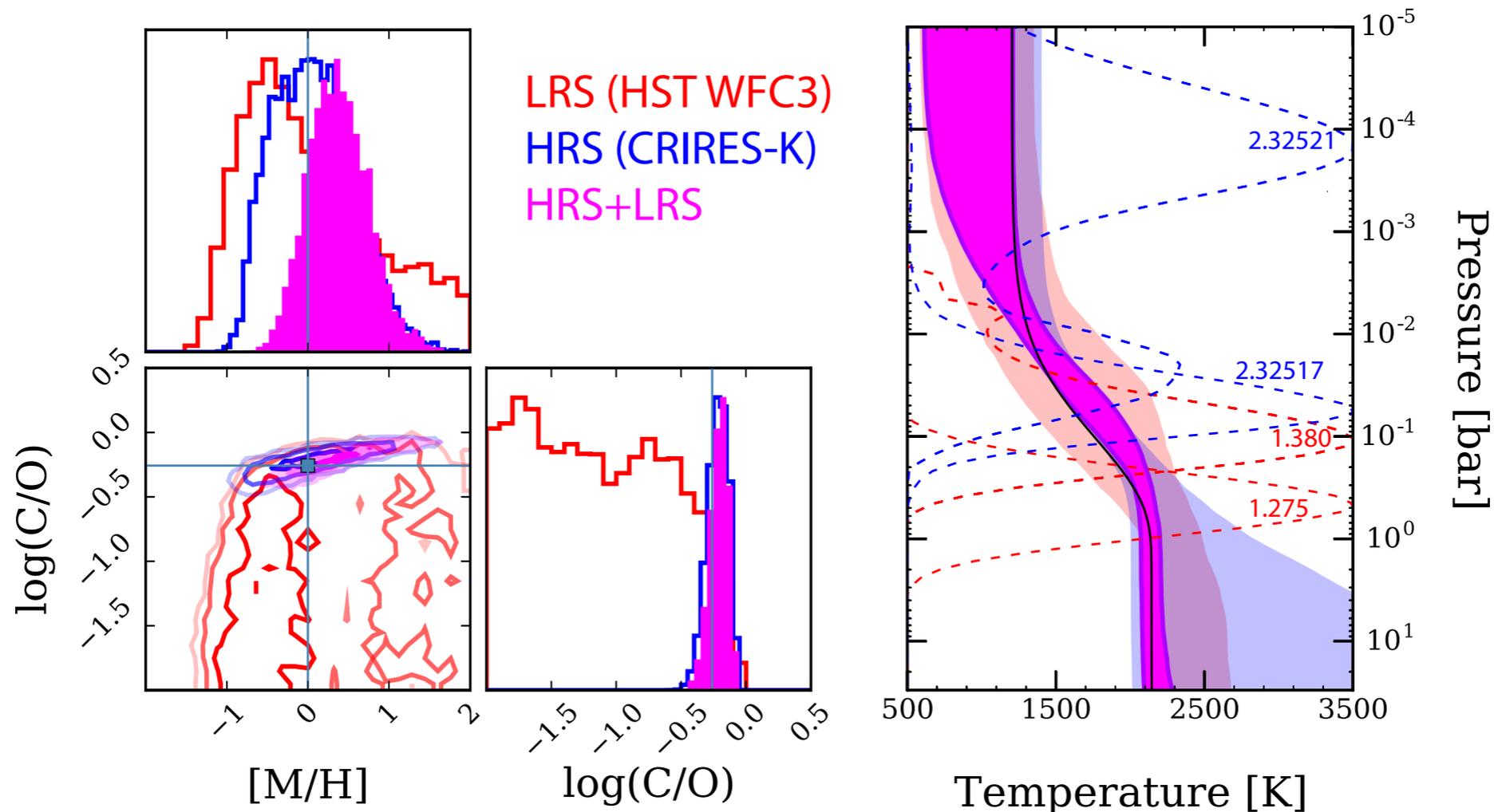
Example: the effects of clouds



Combining LR and HR within the likelihood framework

Brogi & Line (2019): simulated HST + VLT/CRIRES observations
Noise level of **current** observatories (1 eclipse / 5 hours)

$$\text{total logL} = \text{logL (low-res)} + \text{logL (high-res)}$$



Published LR+HR on real data are still rare
(Piskorz+18, Gandhi+ 2019, Gandhi+ in prep.)

**Why should we even care with JWST incoming
and such high-quality ground-based observations?**

4 good reasons to combine low- and high-resolution

Robustness

Independent method, information, and instruments
Test for validity of model assumptions

Errors

Confidence intervals shrink across the whole parameter space

Reducing biases and degeneracy

Aerosols and 3-dimensional effects have different impact on HR and LR

Optimisation

Use the predictive capability of a HR dataset
to inform JWST observations

Ground and space observations are in synergy, not in competition