

自動微分可能なスペクトル 第一原理フィットツールの開発と 高温下の分子線幅測定計画

Hajime Kawahara, Yui Kawashima, Kento Masuda, Ian Crossfield, Erwan Pannier, van den Bekerom, Takayuki Kotani, Kazuo Yoshioka, Tako Ishikawa, Stevanus Nugroho, and REACH collaboration

Kawahara+ ApJS in press 2105.14782

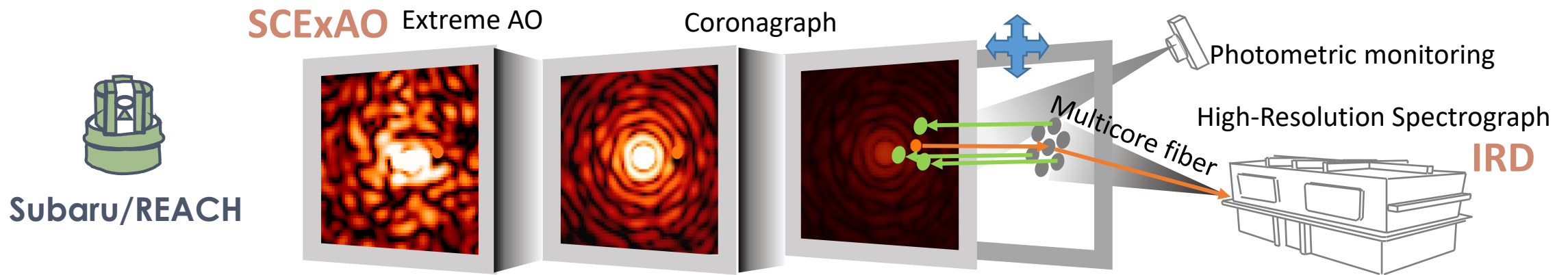
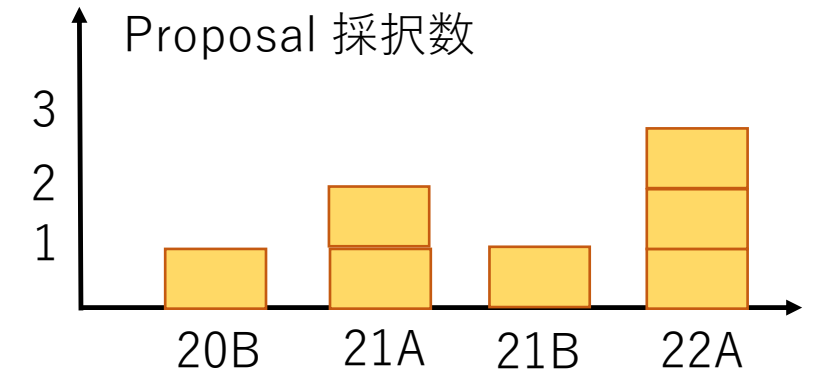
REACH: High-Dispersion Spectrum for High-Contrast Instruments

High-Dispersion Coronagraphy on Subaru

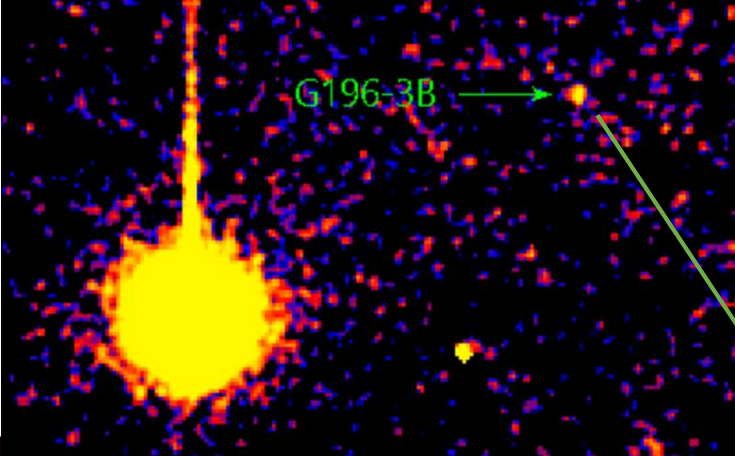
Kawahara, Murakami, Matsuo, and Kotani (2014) ApJS

Late 2014: Started to develop such an instrument with SCExAO team (**Guyon, Lozi, Jovanovic** etc) and IRD (**Kotani, Vievard** etc) on Subaru

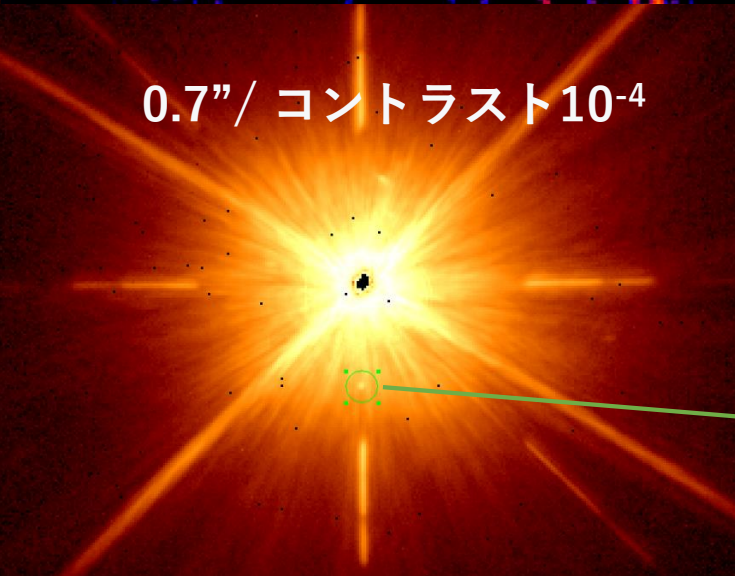
Late 2020: available for science use (open use)



REACH (Subaru):	y, J, H	R=100,000 (SCExAO + IRD)
KPIC (Keck):	K, L, M	R=30,000 (NIRSPEC)
HiRISE (VLT):	Under construction H, K, ...	R=100,000 (SPHERE + CRIRES+)

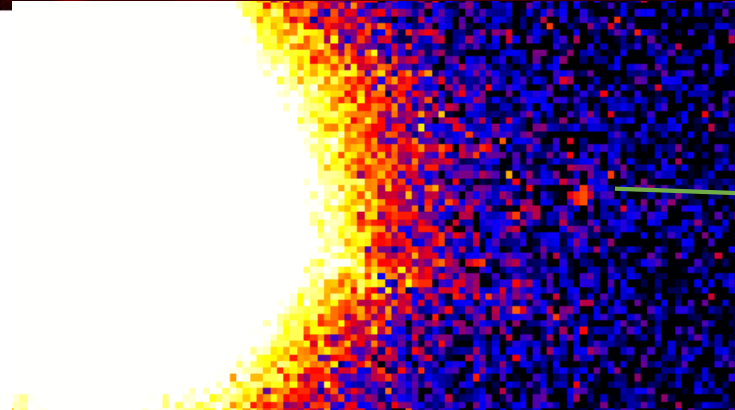


L1

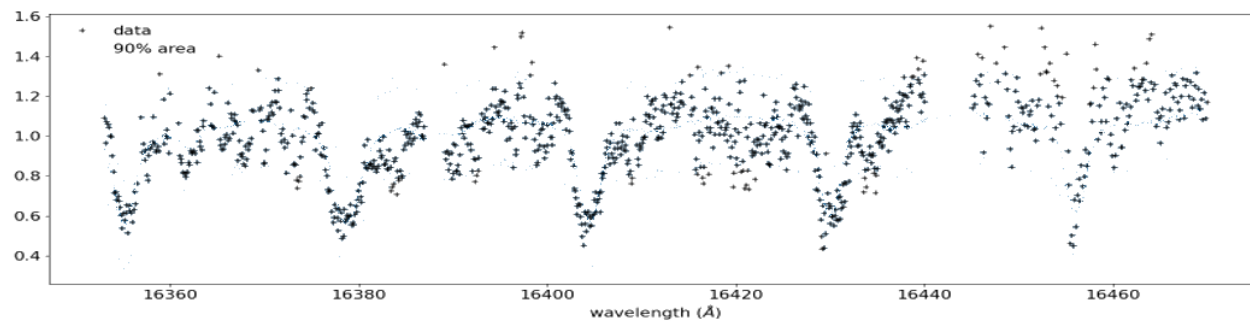
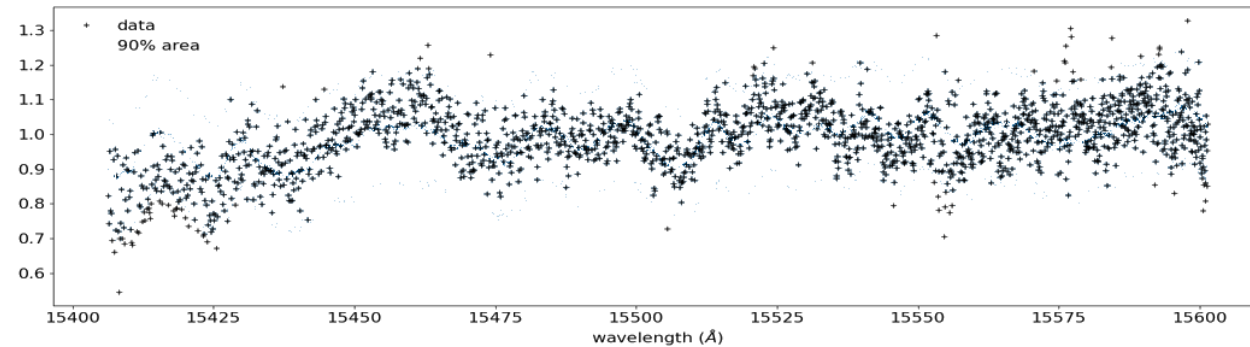
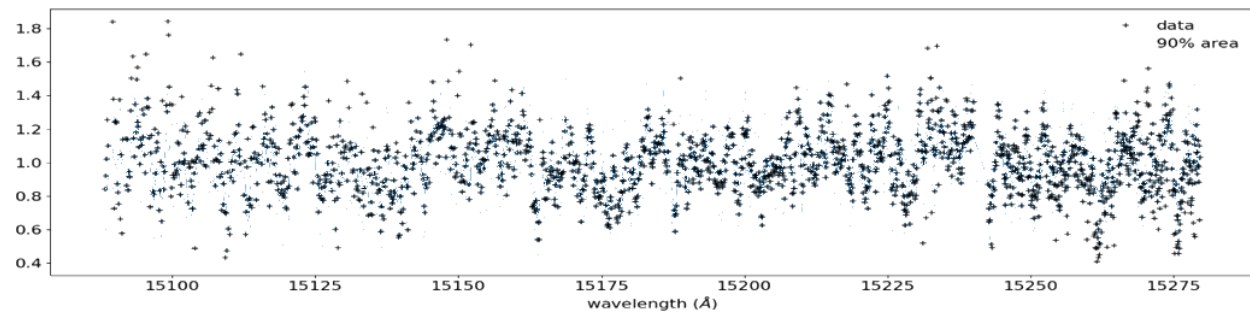
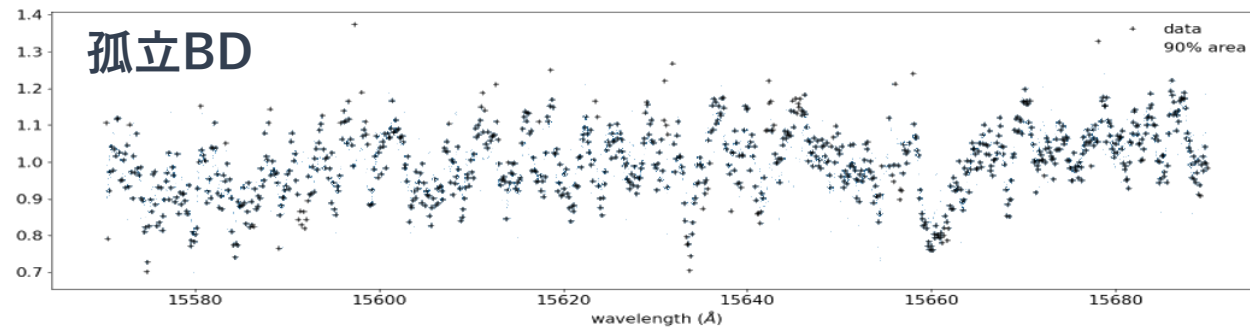


L3

L4



T4



exojax auto-differentiable spectrum model

Atmosphere

Physical quantities

Temperature Profile
Molecular abundance

Gravity
Micro Turbulence



Molecular Database

Opacity Computation

Radiative Transfer

Cross Section

Planet Atmosphere

Macrophysics



Instrument/Dynamics

Instrumental Response

Planet Rotation
Radial Velocity

Bayes inference

Observational Data

+

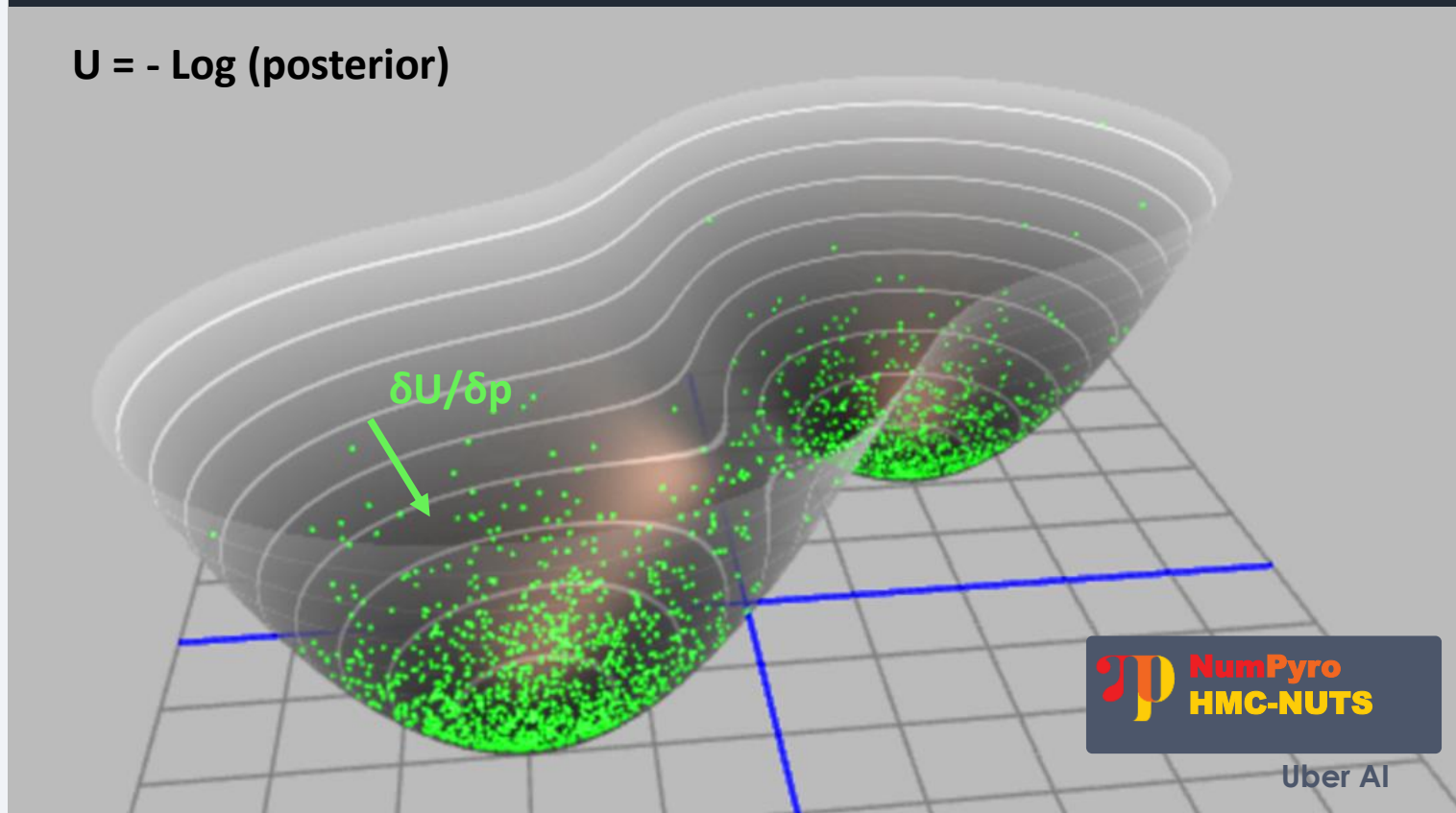
Likelihood & prior

Noise model



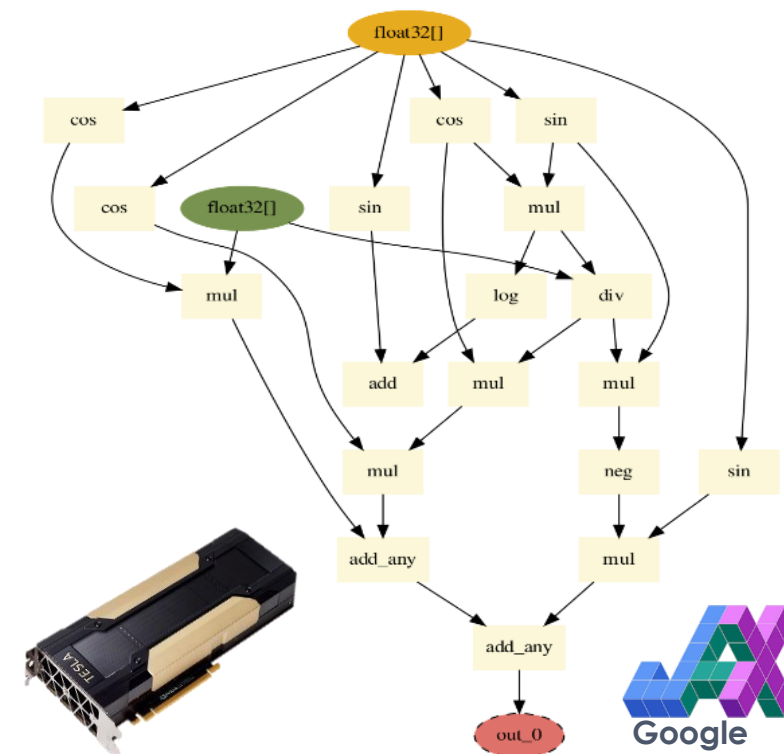
Gradient-based Efficient MCMC Sampling

$U = -\text{Log}(\text{posterior})$



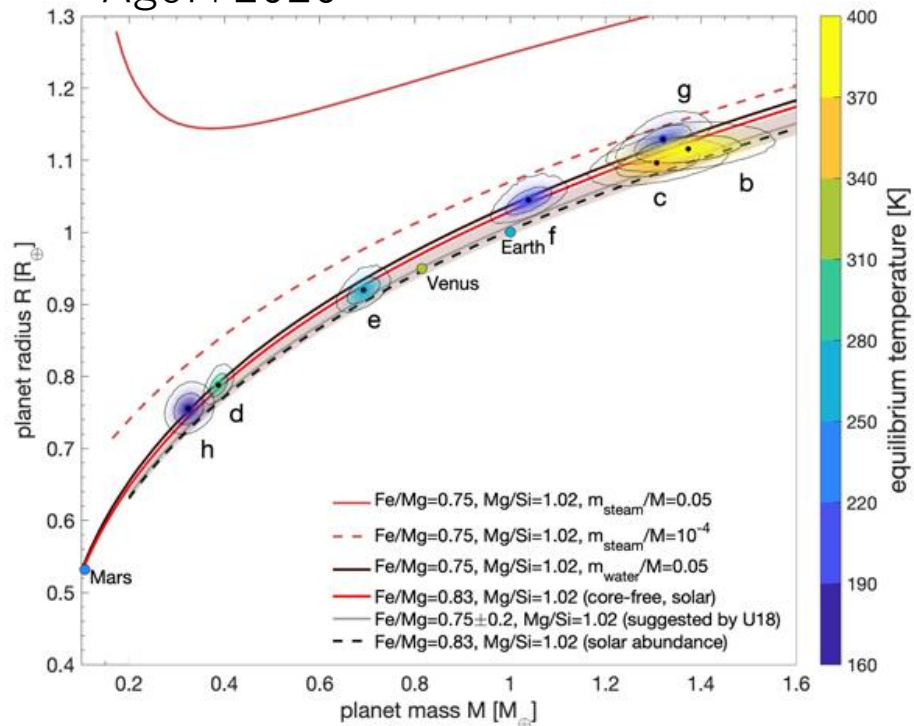
$D^{5/4}$ (HMC) D^2 (random MCMC)

Auto-Gradient + GPU



$$\partial_x (\log \cos x \sin x + \sin x)$$

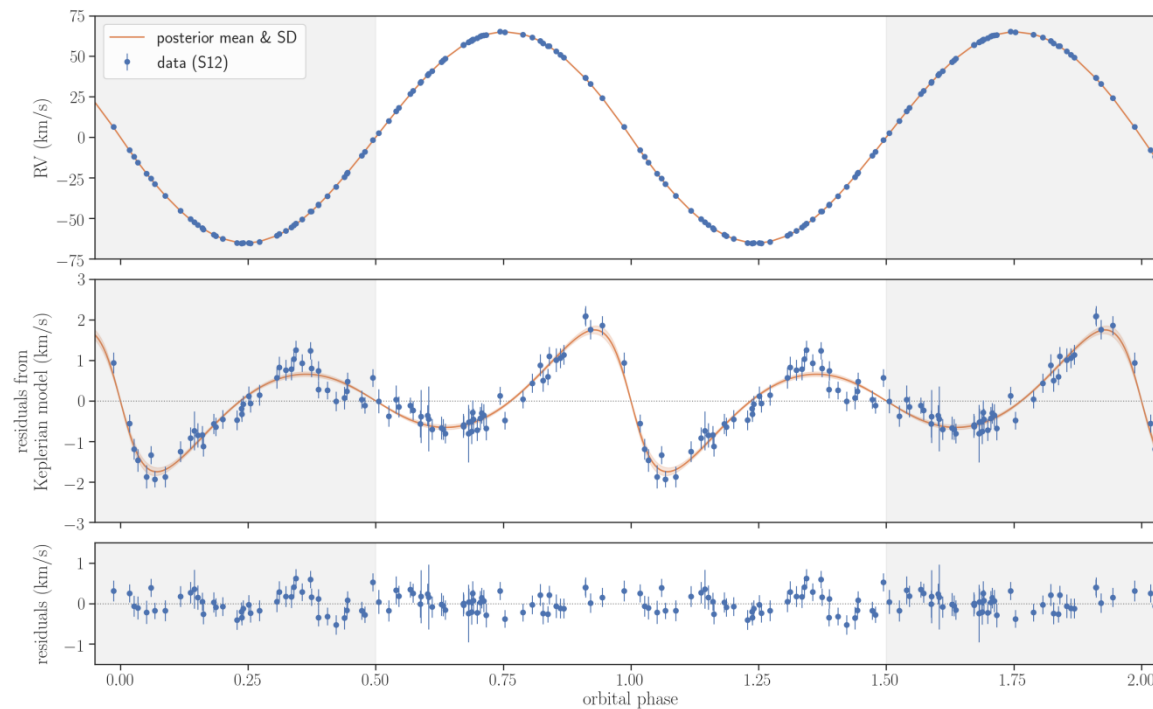
Agol+2020



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MASUDA AND HIRANO

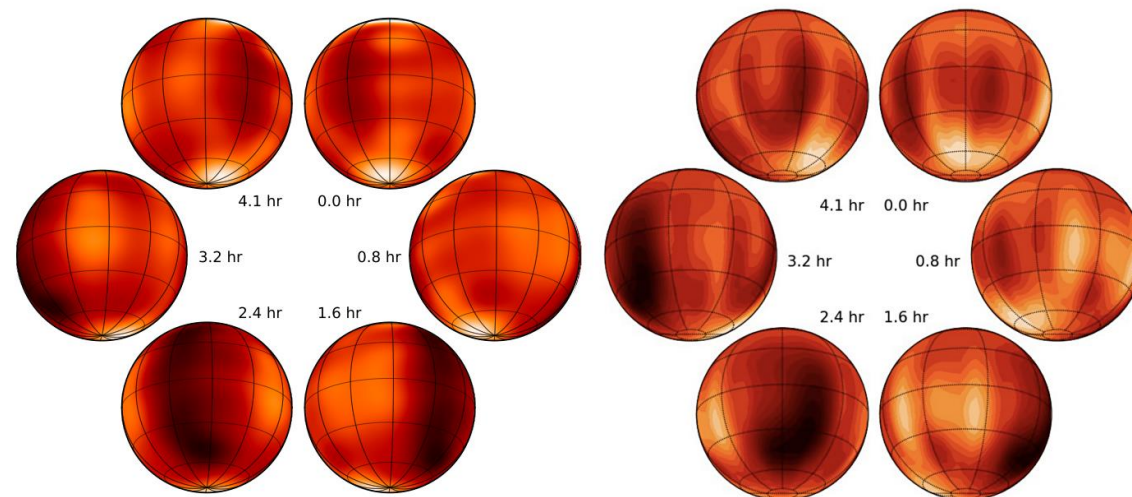
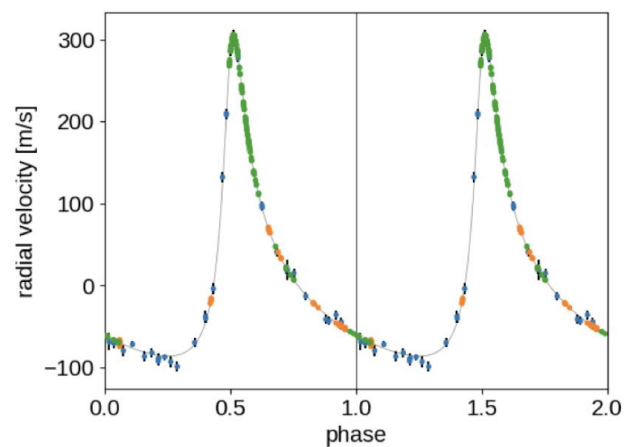
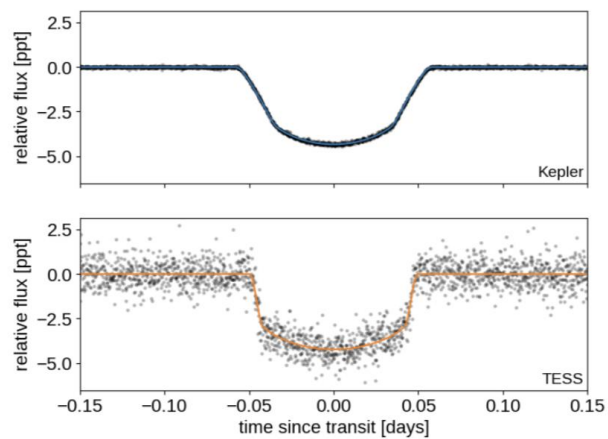
Masuda Hirano 2021



Luger et al. (2021)

Crossfield et al. (2014)

DFM+2021



$$f(x) = |x| + \sin x$$

```
1 import jax.numpy as jnp
```

```
1 def f(x):  
2     return jnp.abs(x) + jnp.sin(x)
```

```
1 from jax import grad
```

```
1 df=grad(f)
```

```
1 df(1.)    # → DeviceArray(1.5403023, dtype=float32)
```


exojax auto-differentiable spectrum model

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Spectrum Model

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Bayes inference

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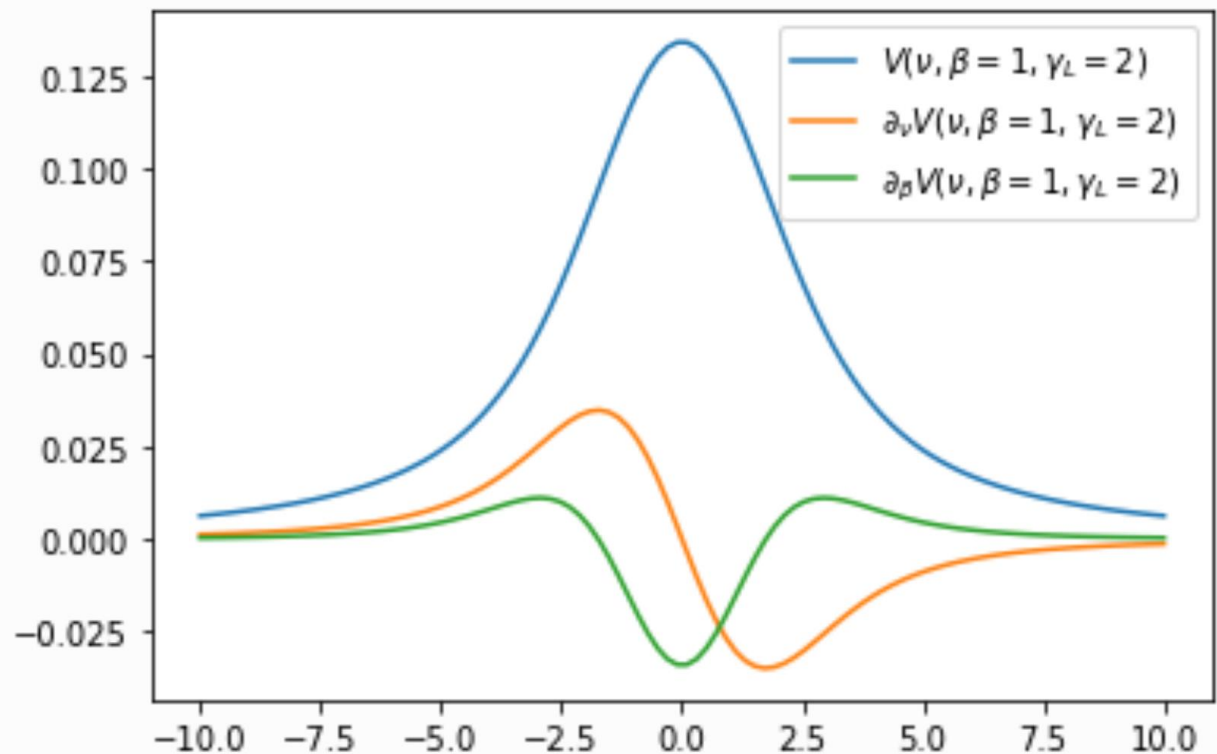
Likelihood & prior

Noise model

Differentiable Voigt Profile

```
from exojax.spec import voigtone
from jax import grad, vmap
```

```
dvoigt_nu=vmap(grad(voigtone,argnums=0),(0, None, None),0) #derivative by nu
dvoigt_beta=vmap(grad(voigtone,argnums=1),(0, None, None),0) #derivative by beta
```

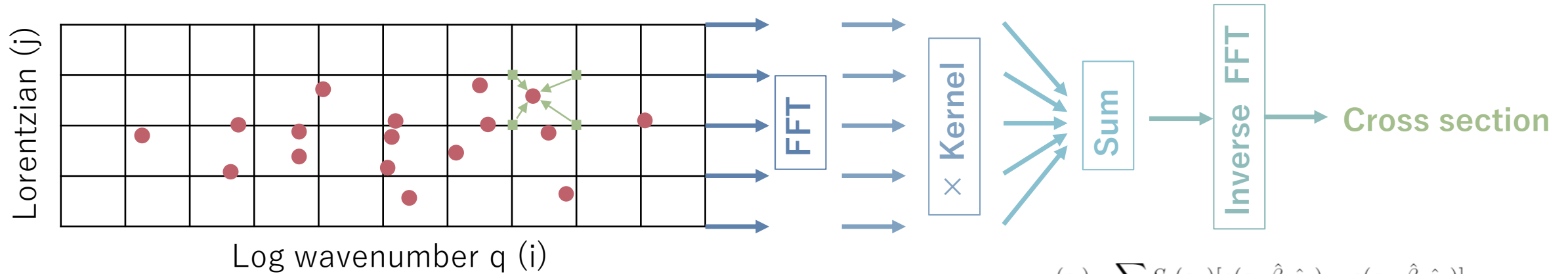


Algorithm 916: Zaghloul & Ali (2011)

Modified Discrete Integral Transform (MODIT)

With van den Bekeroma and Pannier

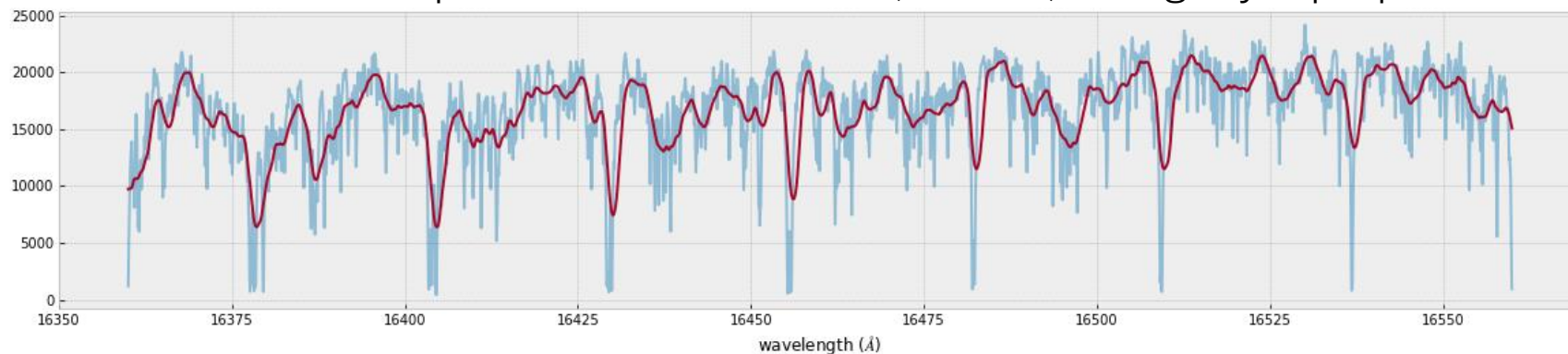
In log wavenumber, a Doppler width is common

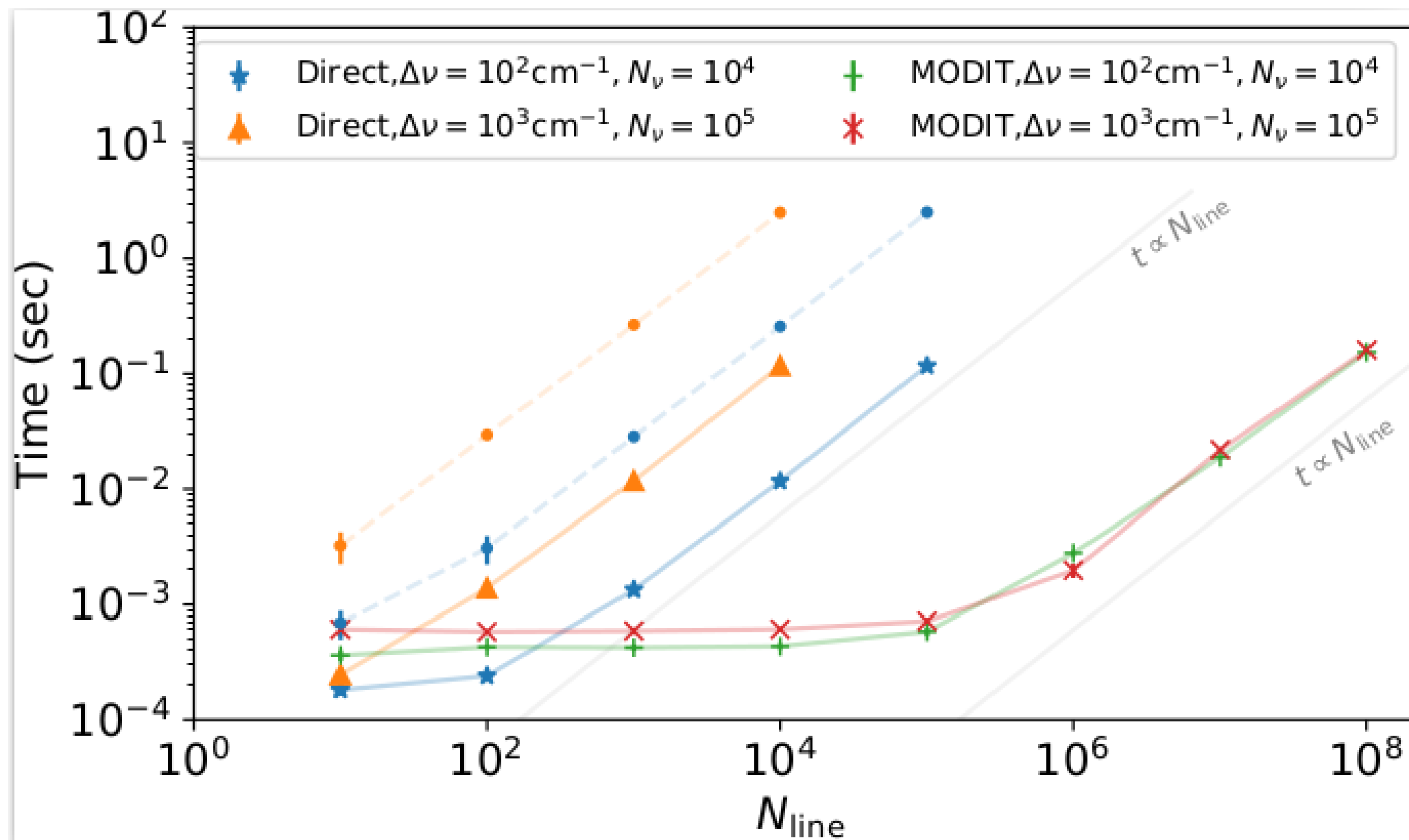


$$\sigma(q_i) = \sum_{i',j} S_j(q_{i'}) [g(q_i; \hat{\beta}, \hat{\gamma}_j) - g(q_{i'}; \hat{\beta}, \hat{\gamma}_j)]$$

$$= \mathcal{FT}^{-1} \left[\sum_j \tilde{S}_j(k) \tilde{g}(k; \hat{\beta}, \hat{\gamma}_j) \right]$$

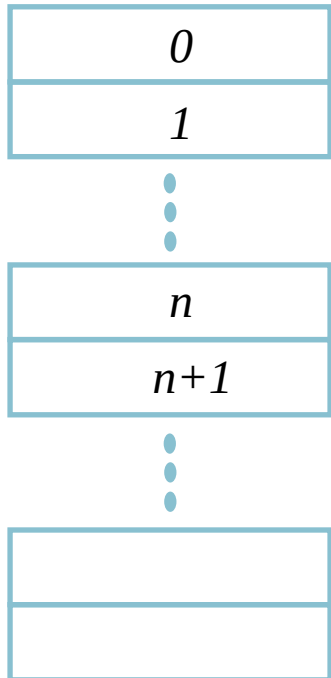
Methane spectrum from scratch (exomol) using my laptop





Differentiable Radiative Transfer

$$F_0 = \text{sum}(Q * \text{cumprod}(T, \text{axis}=0), \text{axis}=0)$$



$$F_0 = \mathcal{T}_0(\mathcal{T}_1(\mathcal{T}_2(\cdots \mathcal{T}_{N-2}(\mathcal{T}_{N-1}F_B + \mathcal{Q}_{N-1}) + \mathcal{Q}_{N-2}) + \cdots + \mathcal{Q}_2) + \mathcal{Q}_1) + \mathcal{Q}_0 = \phi \cdot q$$

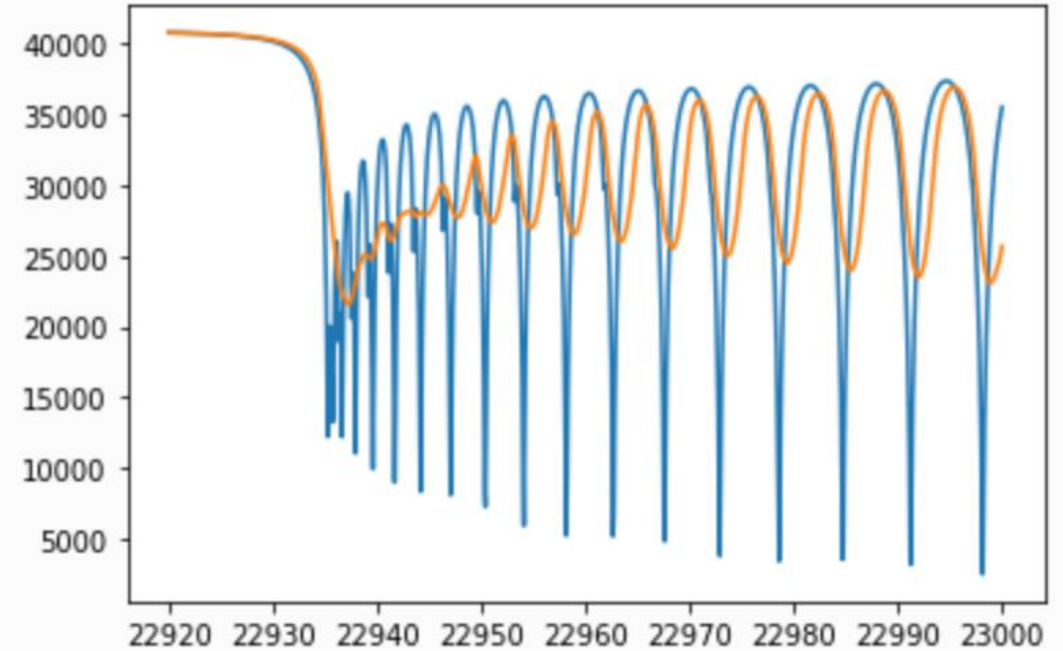
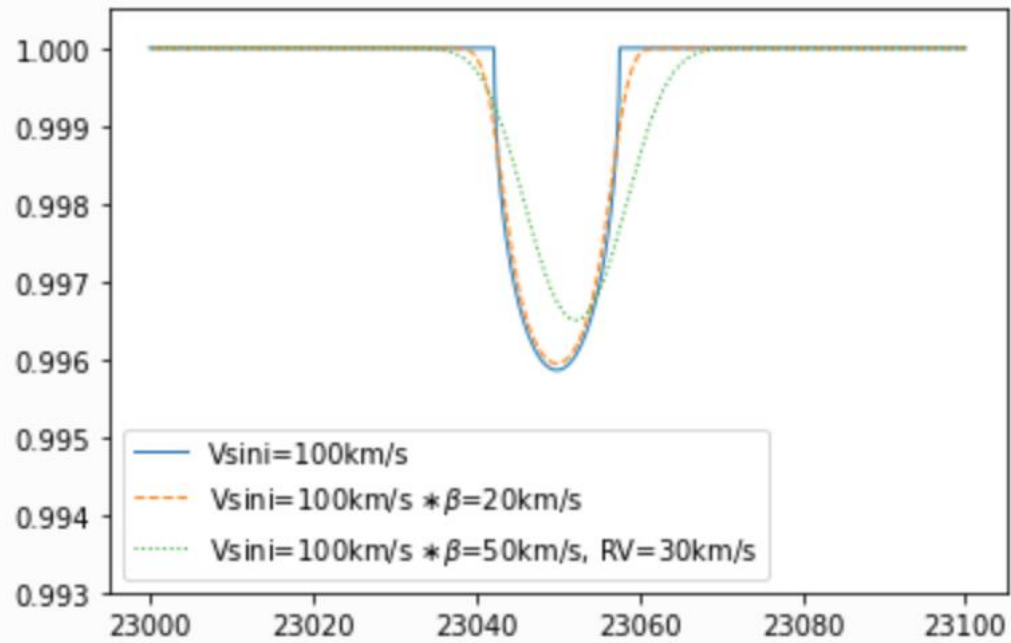
No scattering

$$F_n = \mathcal{T}_n F_{n+1} + \overbrace{(1 - \mathcal{T}_n) \mathcal{Q}_n}^{\mathcal{Q}_n},$$

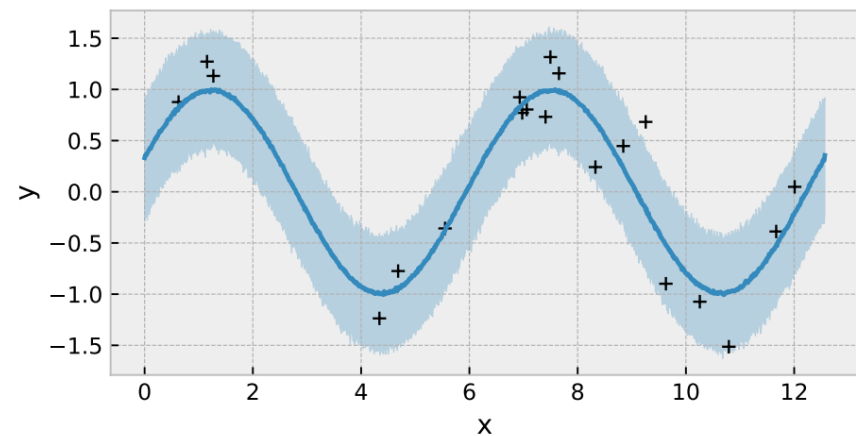
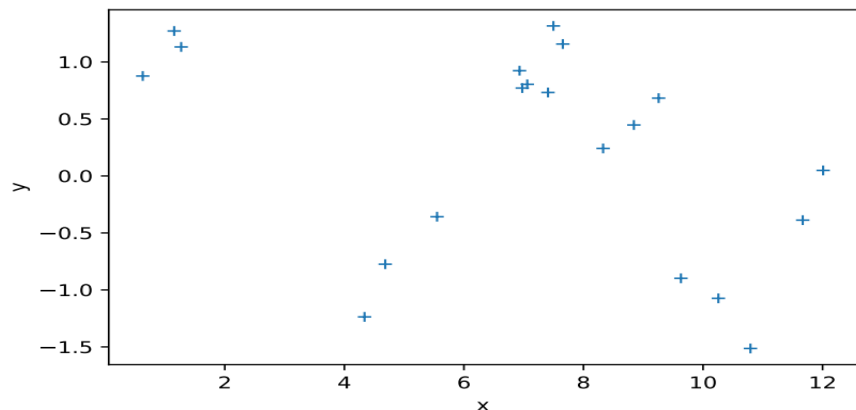
$$\begin{aligned} \phi &= (1, \mathcal{T}_0, \mathcal{T}_0 \mathcal{T}_1, \cdots, \mathcal{T}_0 \mathcal{T}_1 \cdots \mathcal{T}_{N-2}, \mathcal{T}_0 \mathcal{T}_1 \cdots \mathcal{T}_{N-1})^\top \\ q &\equiv (\mathcal{Q}_0, \mathcal{Q}_1, \cdots, \mathcal{Q}_N)^\top \end{aligned}$$

$$\begin{aligned} \mathcal{T}_n &= 2E_3(\Delta\tau_n) \\ &= (1 - \Delta\tau_n) \exp(-\Delta\tau_n) + (\Delta\tau_n)^2 E_1(\Delta\tau_n) \end{aligned}$$

Auto-differentiable Rotation Kernel and Instrumental Profile



簡単NumPyro



```
1 import jax.numpy as jnp
2 import numpyro
3 import numpyro.distributions as dist
4
5 def model(x,y):
6     phase = numpyro.sample('phase', dist.Uniform(-1.0*jnp.pi, 1.0*jnp.pi))
7     sigma = numpyro.sample('sigma', dist.Exponential(1.))
8     mu=jnp.sin(x+phase)
9     numpyro.sample('y', dist.Normal(mu, sigma), obs=y)
```

この例からわかるように、確率分布は `dist` で指定されたものを、`numpyro` の `sample` 内で定義することにより確率変数として扱われる。最後のデータ `y` の部分は式 (4.1,4.2) が

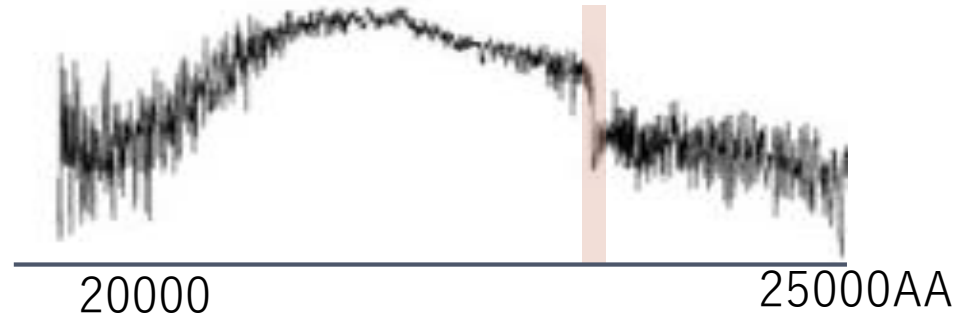
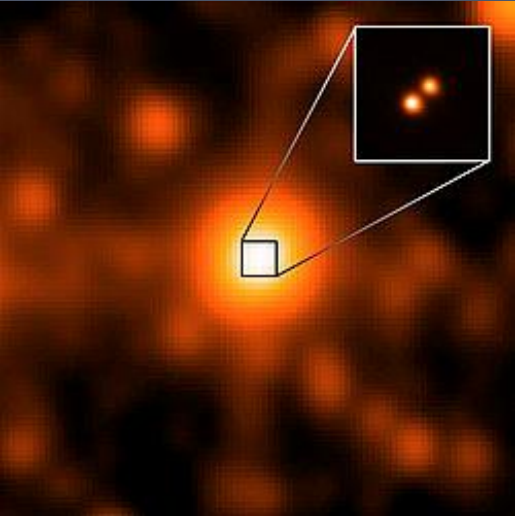
$$y \sim \mathcal{N}(\sin(x + \phi), \sigma) \quad (4.5)$$

と書き直されることよりわかる。さてこれを `numpyro` の HMC-NUTS にかける。

```
1 from jax import random
2 from numpyro.infer import MCMC, NUTS
3 rng_key = random.PRNGKey(0)
4 rng_key, rng_key_ = random.split(rng_key)
5 num_warmup, num_samples = 1000, 2000
```

```
1 kernel = NUTS(model)
2 mcmc = MCMC(kernel, num_warmup=num_warmup, num_samples=num_samples)
3 mcmc.run(rng_key_, x=x, y=y)
4 mcmc.print_summary()
```

Demonstration using Luhman 16A as observed by CRILES

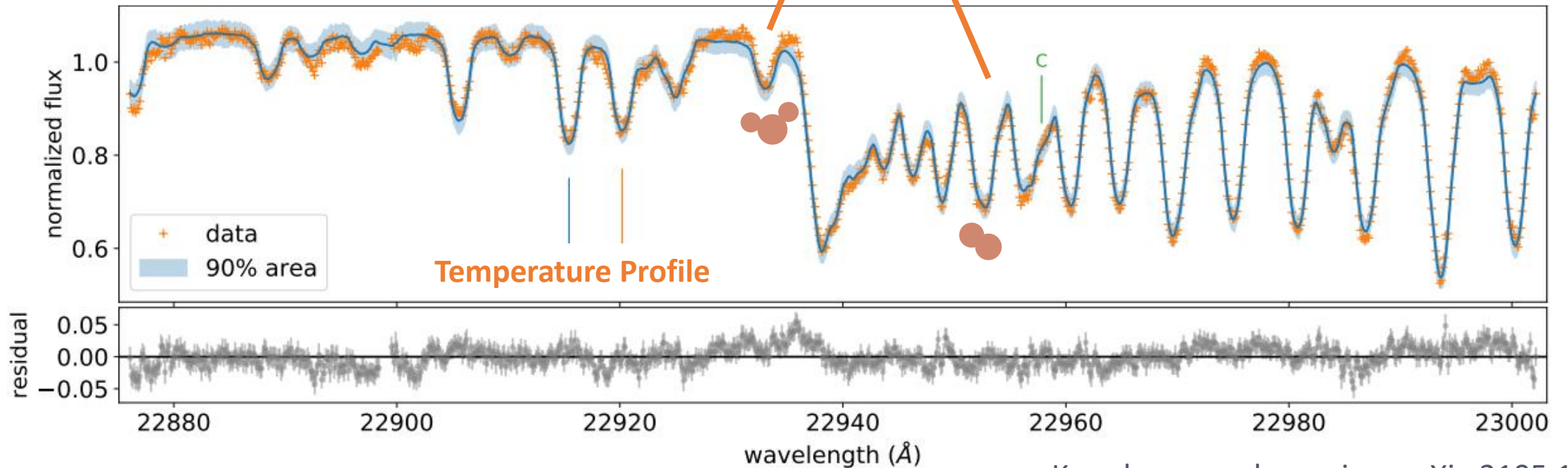


CO+H₂O+CIA

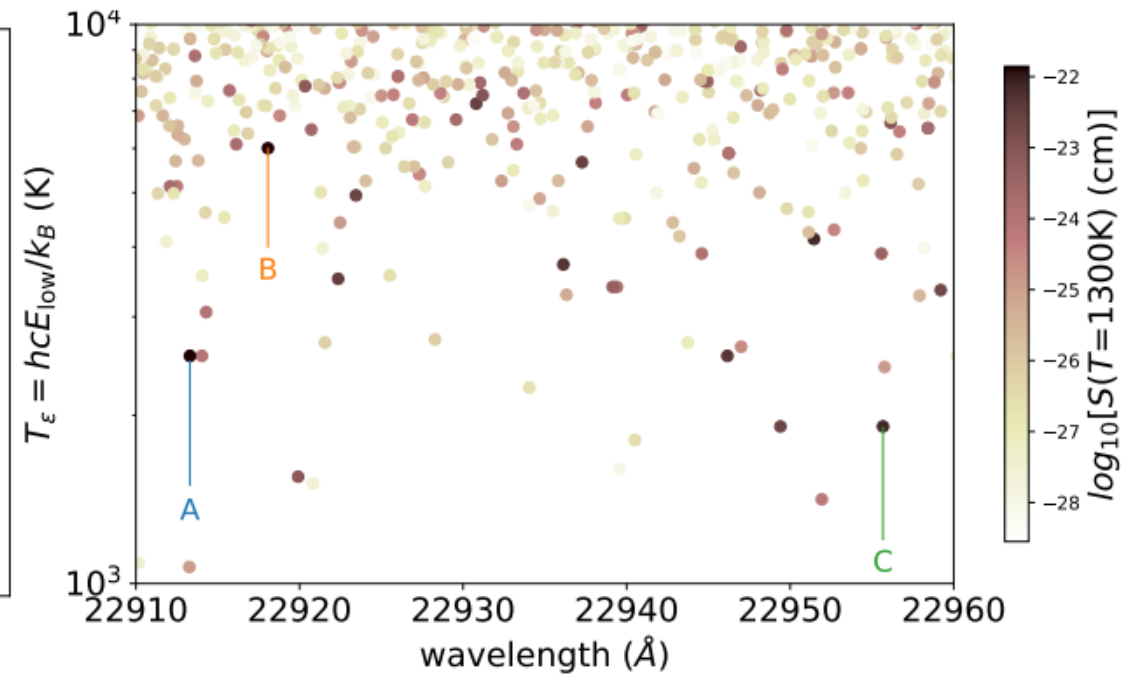
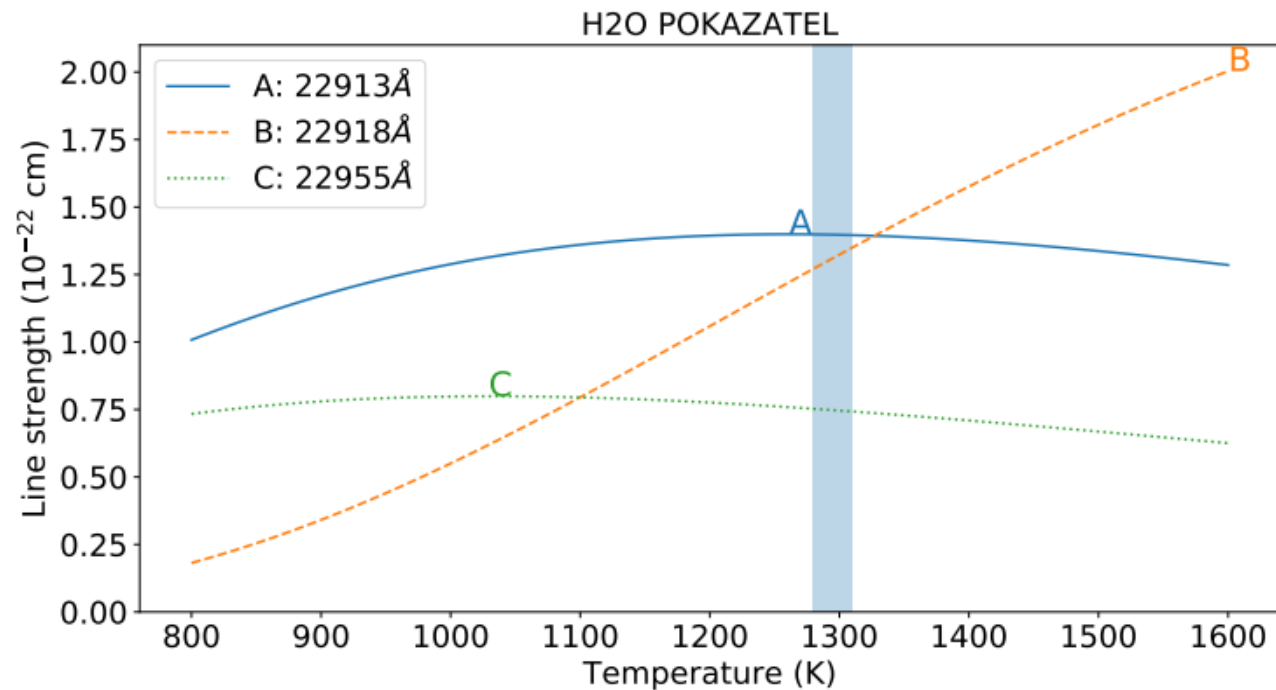
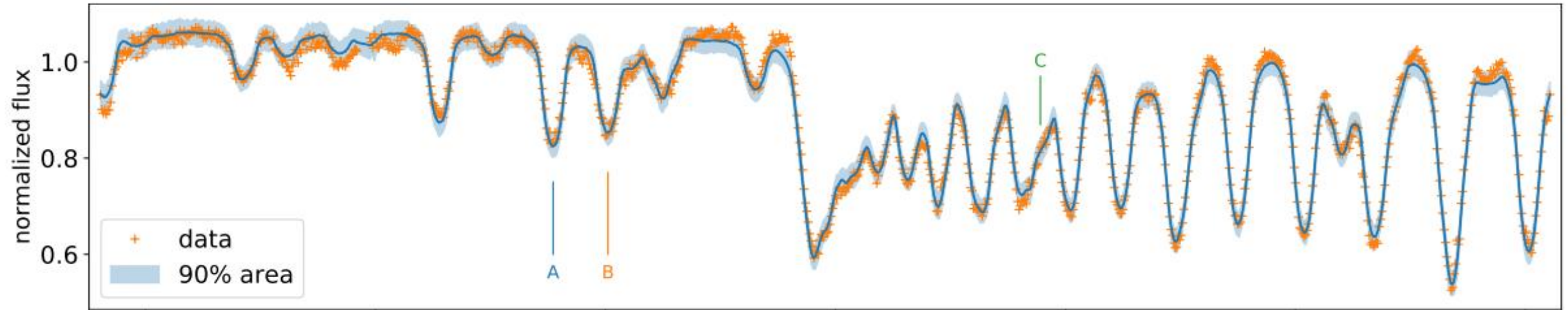
Rotation

Carbon/Oxygen Ratio

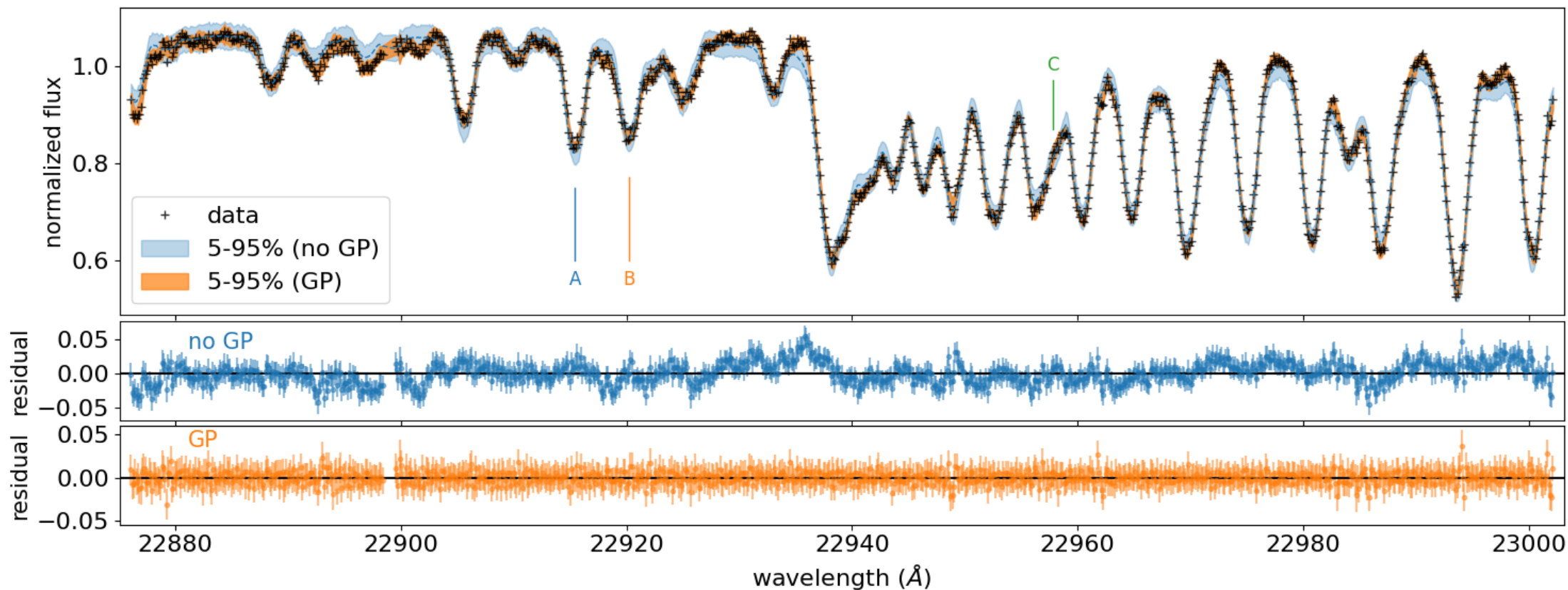
CRILES data (Crossfield+2014)



Line diagnostics of Temperature



GPもノイズモデルに簡単に入ります



```
numpyro.sample(tag, dist.MultivariateNormal(loc=mu, covariance_matrix=cov), obs=y)
```


GPも温度圧力モデルに簡単に入ります

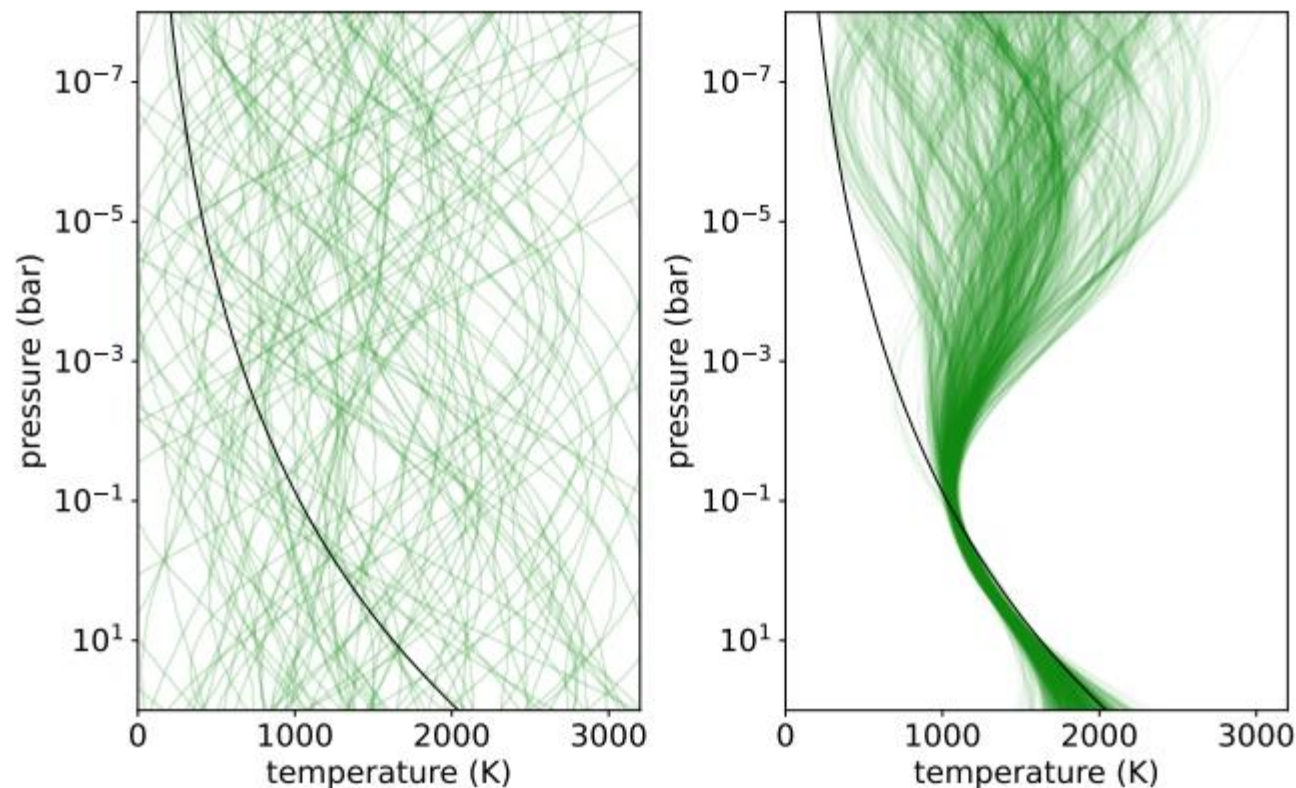
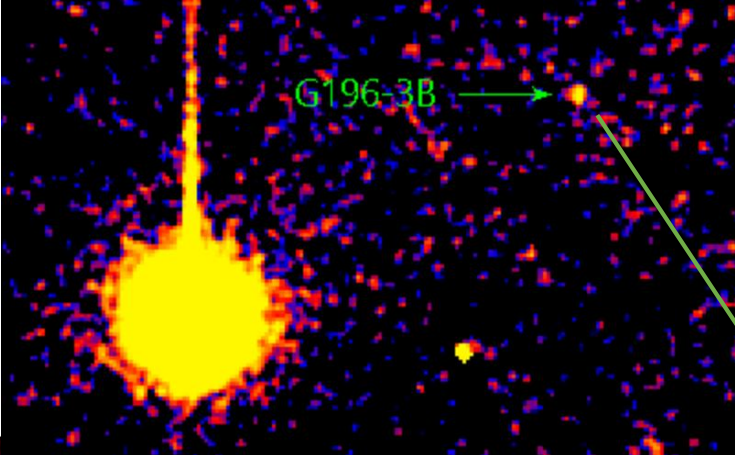
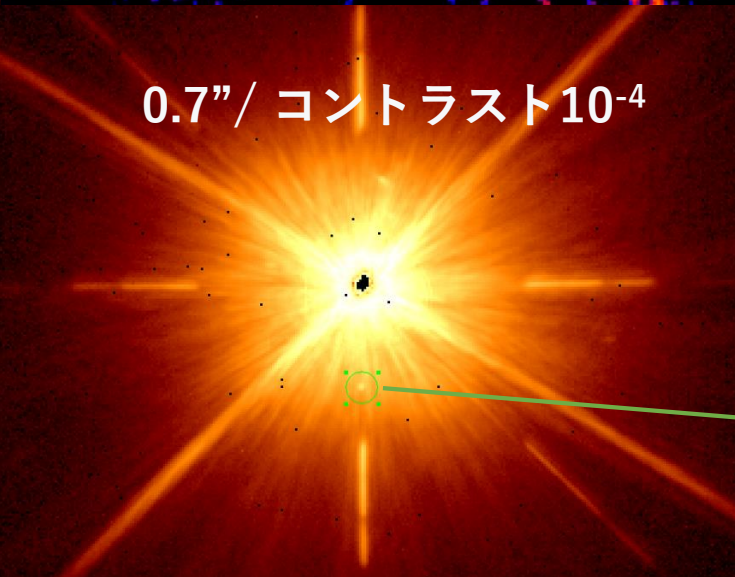


Figure 11. Prior (left) and posterior (right) sampling of the layer-by-layer temperature model (green lines). The solid black lines indicate the best-fit power law model in Section 5.

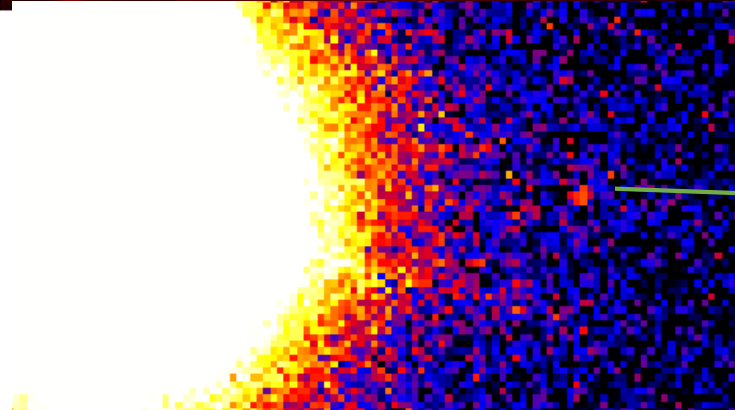


L1

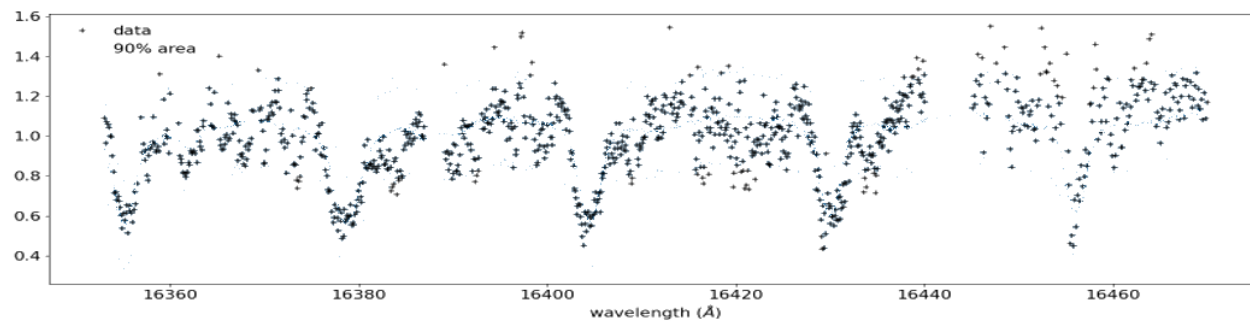
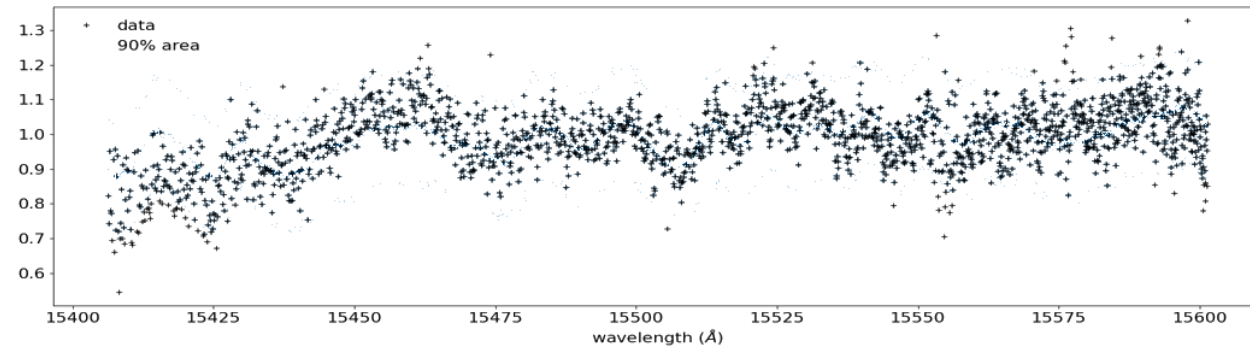
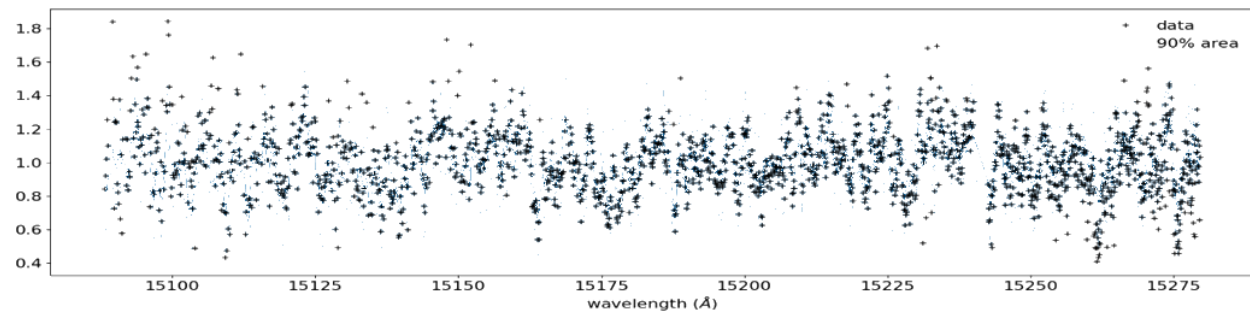
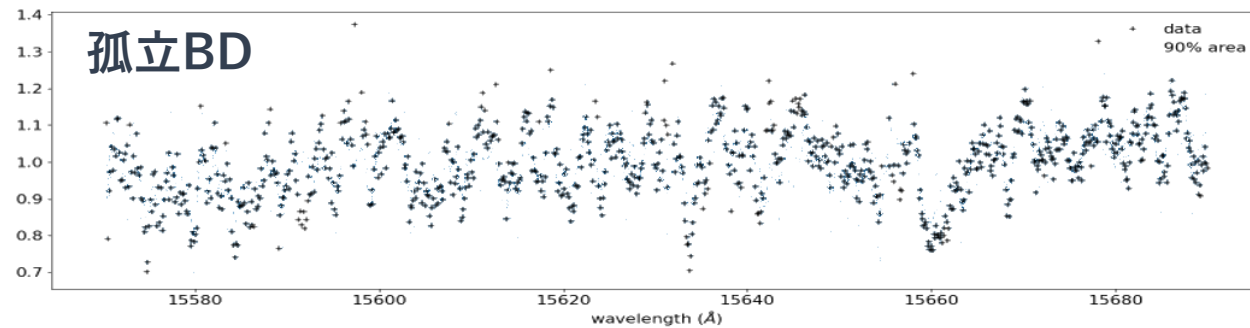


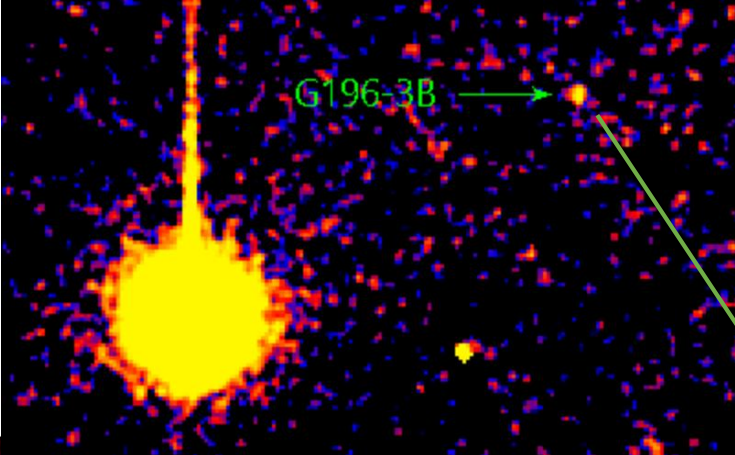
L3

L4



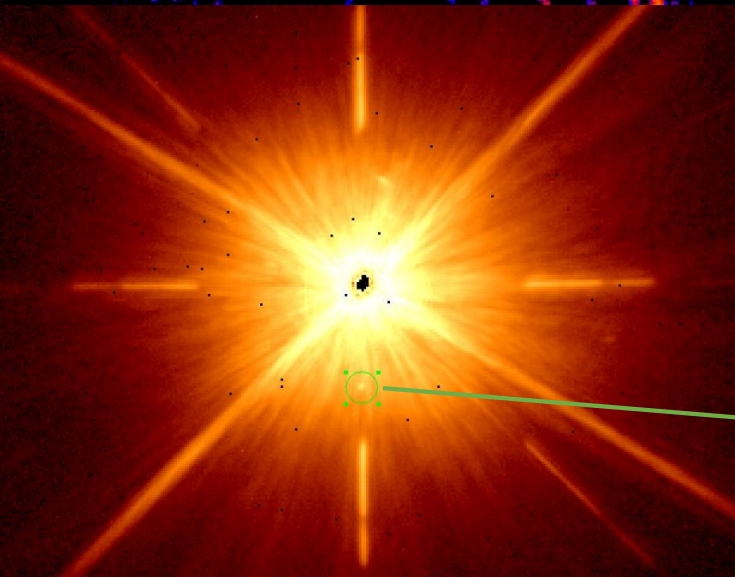
T4



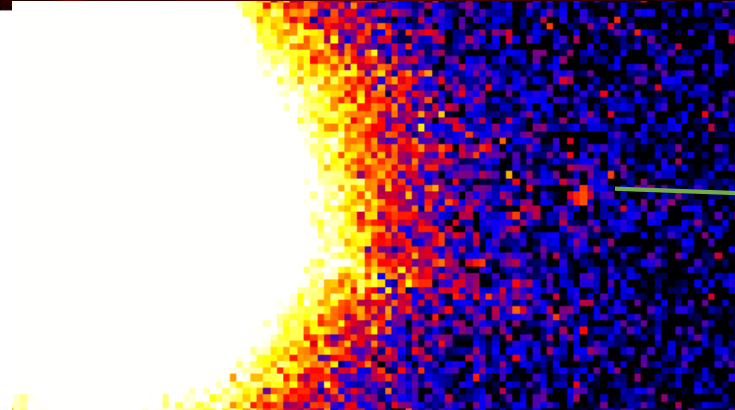


G196-3B

L1

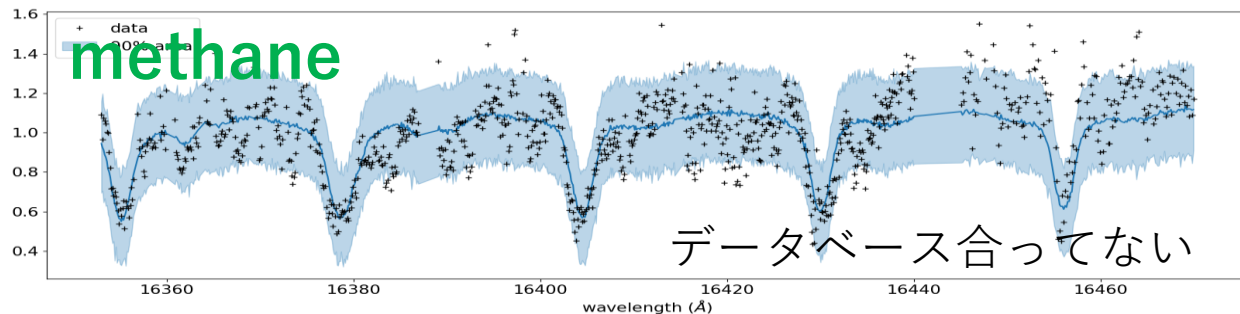
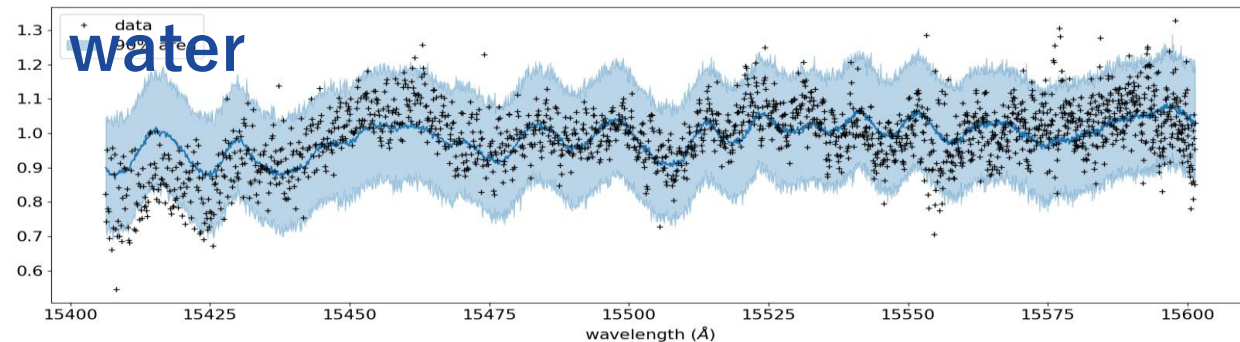
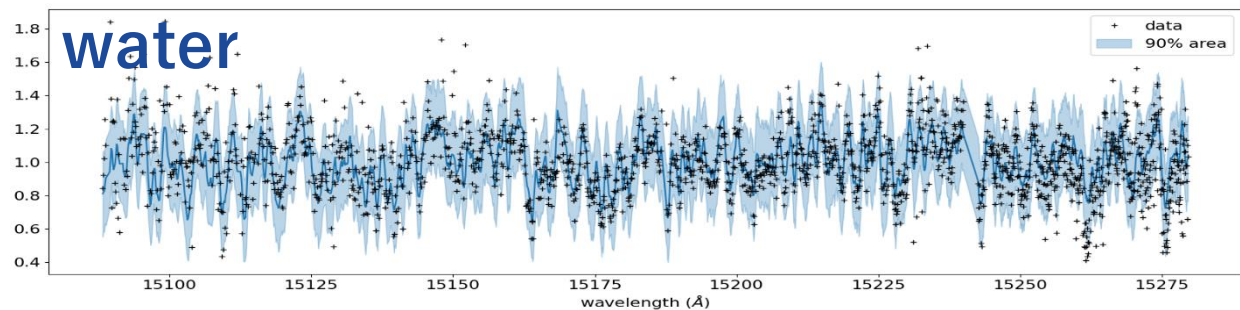
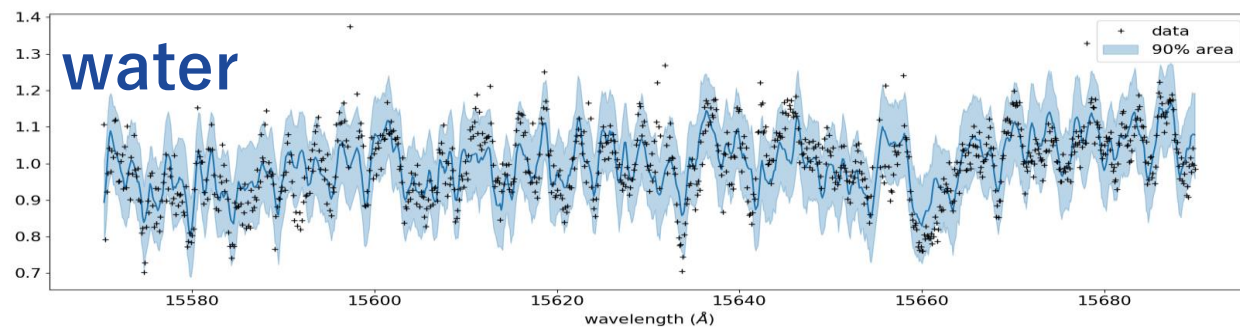


L3



L4

T4



データベース合っていない

Gravity from line broadening

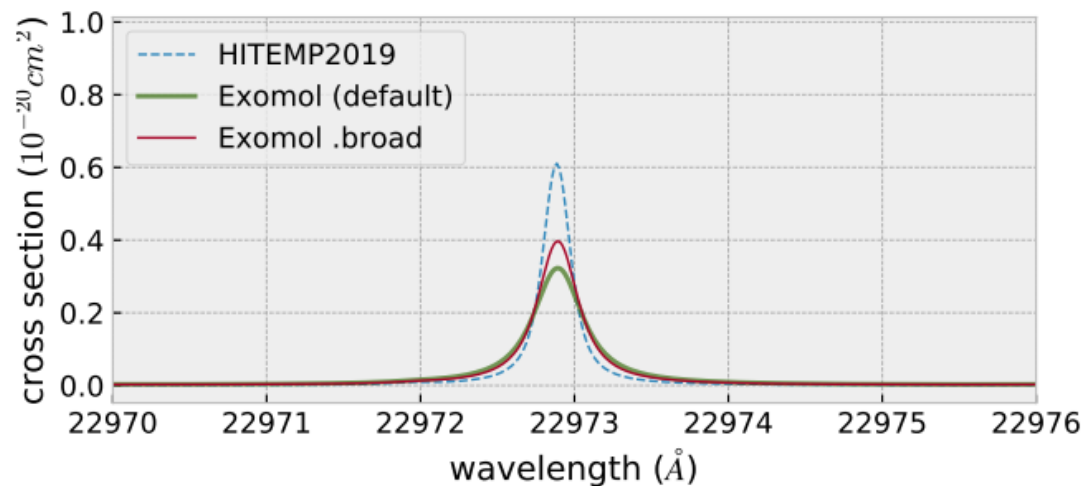


Figure 9. Comparison of the CO cross section between HITEMP2019 and ExoMol/Li2015 (default: constant default values of $\alpha_{\text{ref}} = 0.07$ and $n_{\text{texp}} = 0.5$ in the definition file, .broad: broadening parameters in the “.broad” file). We assume $T = 1300 \text{ K}$, $P = 1 \text{ bar}$ ($P_{\text{air}} = 0.99 \text{ bar}$ and $P_{\text{self}} = 0.1 \text{ bar}$ for HITEMP2019).

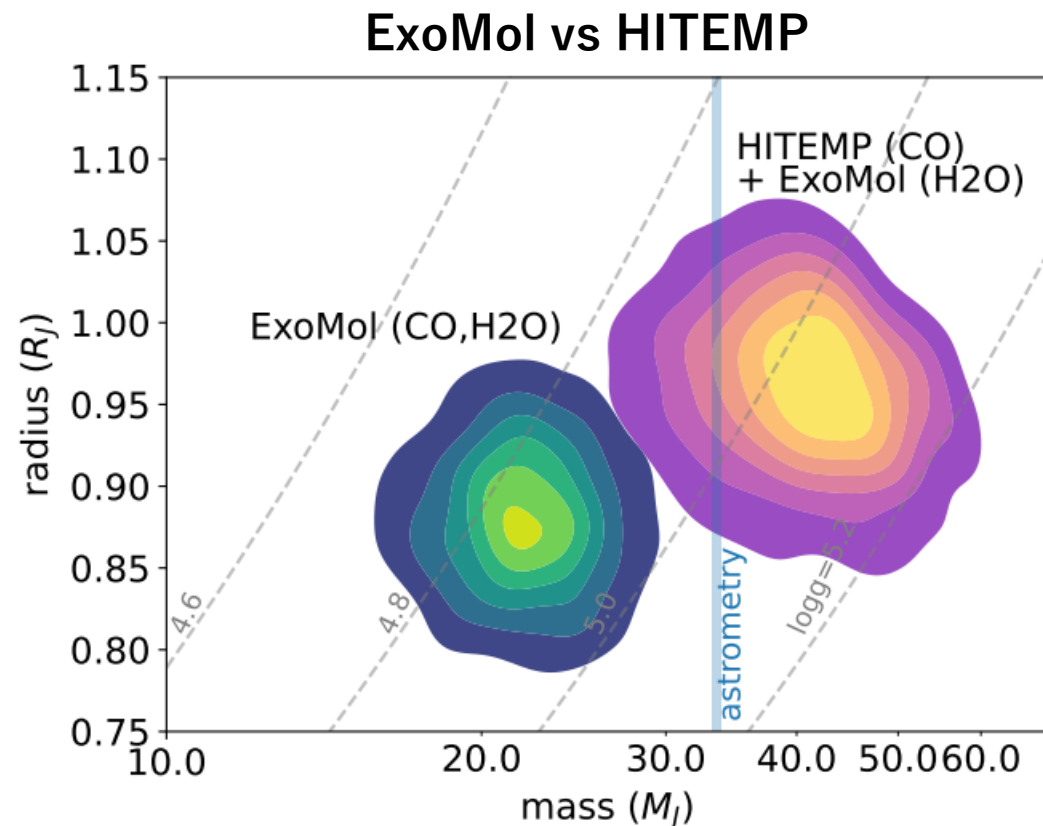
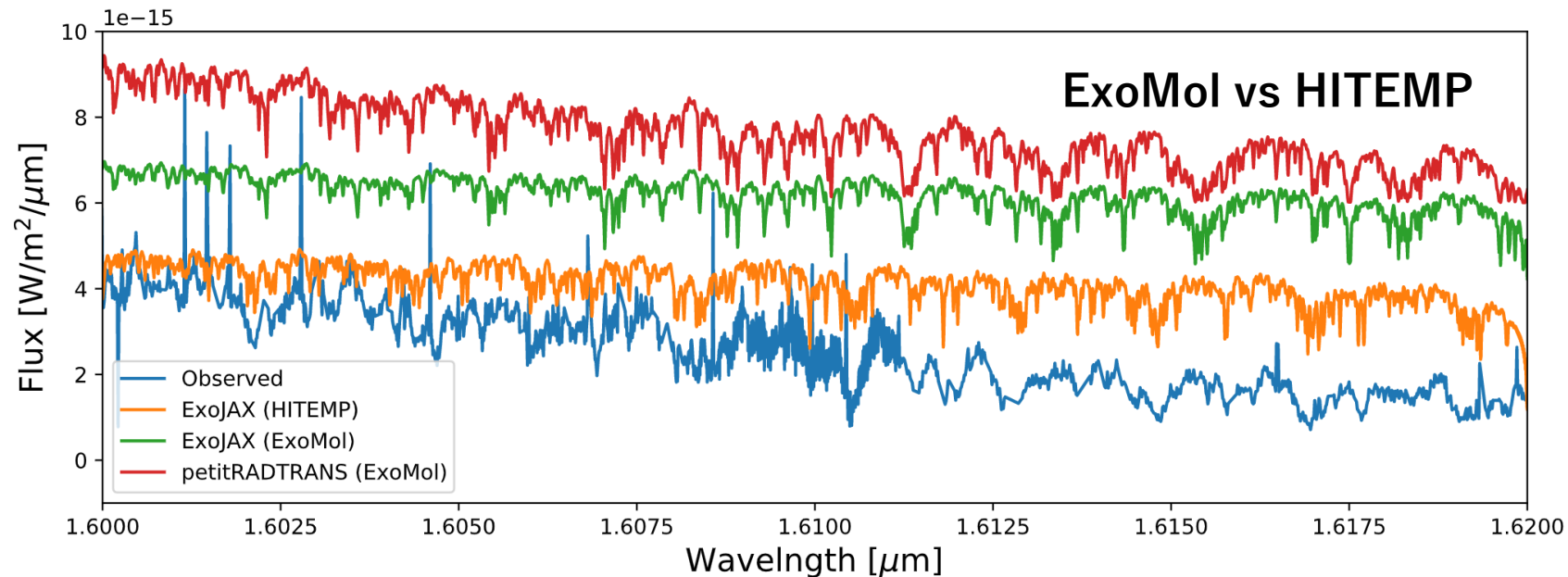


Figure 10. The mass and radius inferred using two different models for CO (ExoMol using the “.broad” file and HITEMP2019). The astrometric mass is shown in the blue shaded region (Lazorenko & Sahlmann 2018). The dashed lines show the iso-gravity lines.

○既存データベースの問題点

ExoMol 水素大気 but 実験での検証が不完全（とくに圧力幅）
HITEMP 地球大気



川島氏提供

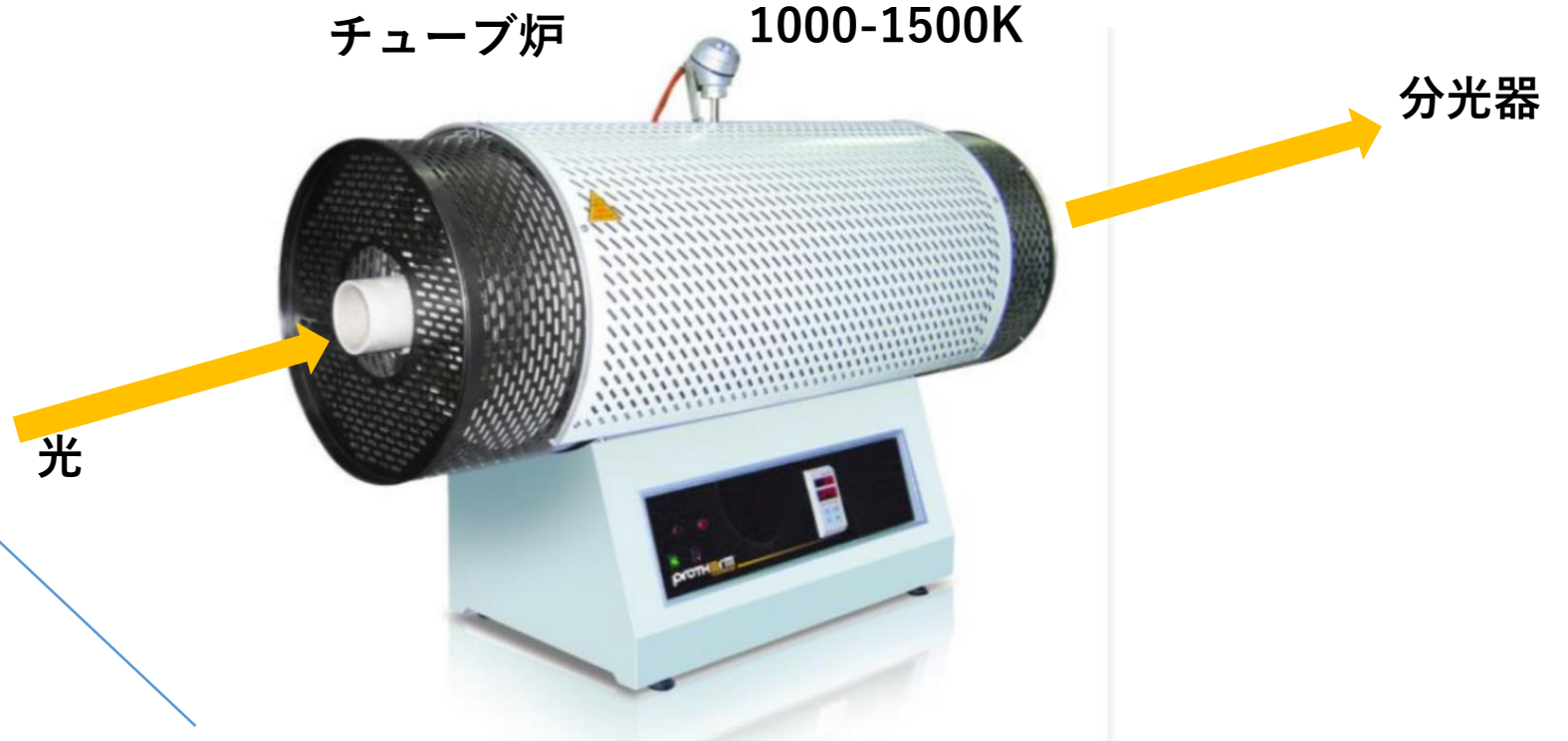
○考えるのがめんどくさいので自分たちで調べたい

ガラス石英管



チューブ炉

1000-1500K



CO (1%) + 水素
CH₄ + 水素
etc を封入 → 分光したい

写真はイメージです



分子データベースから計算でき**ベイズ推定が可能な**
系外惑星・褐色矮星スペクトルモデルをつくった

実際、観測スペクトルを良くフィットできる

基礎データがいまいちなところがあるので、**惑星大気を模擬し**
たガスをいれて分光実験を企画している（共同研究募集中）