Debris disk modeling of Beta Pic and correlations in MATISSE data

Philippe Priolet

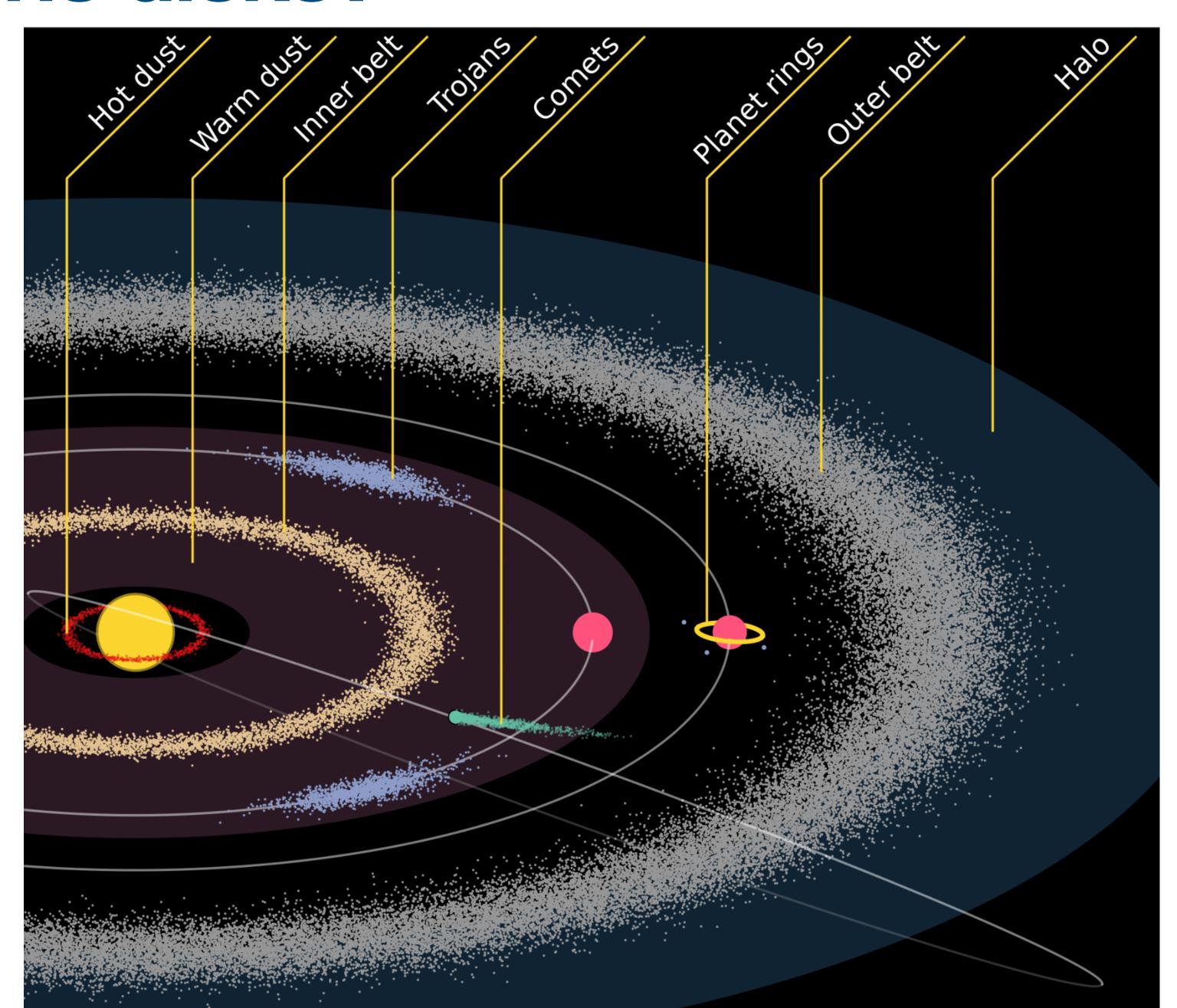
PhD student

Supervised by Jean-Charles Augereau et Julien Milli





What are debris disks?

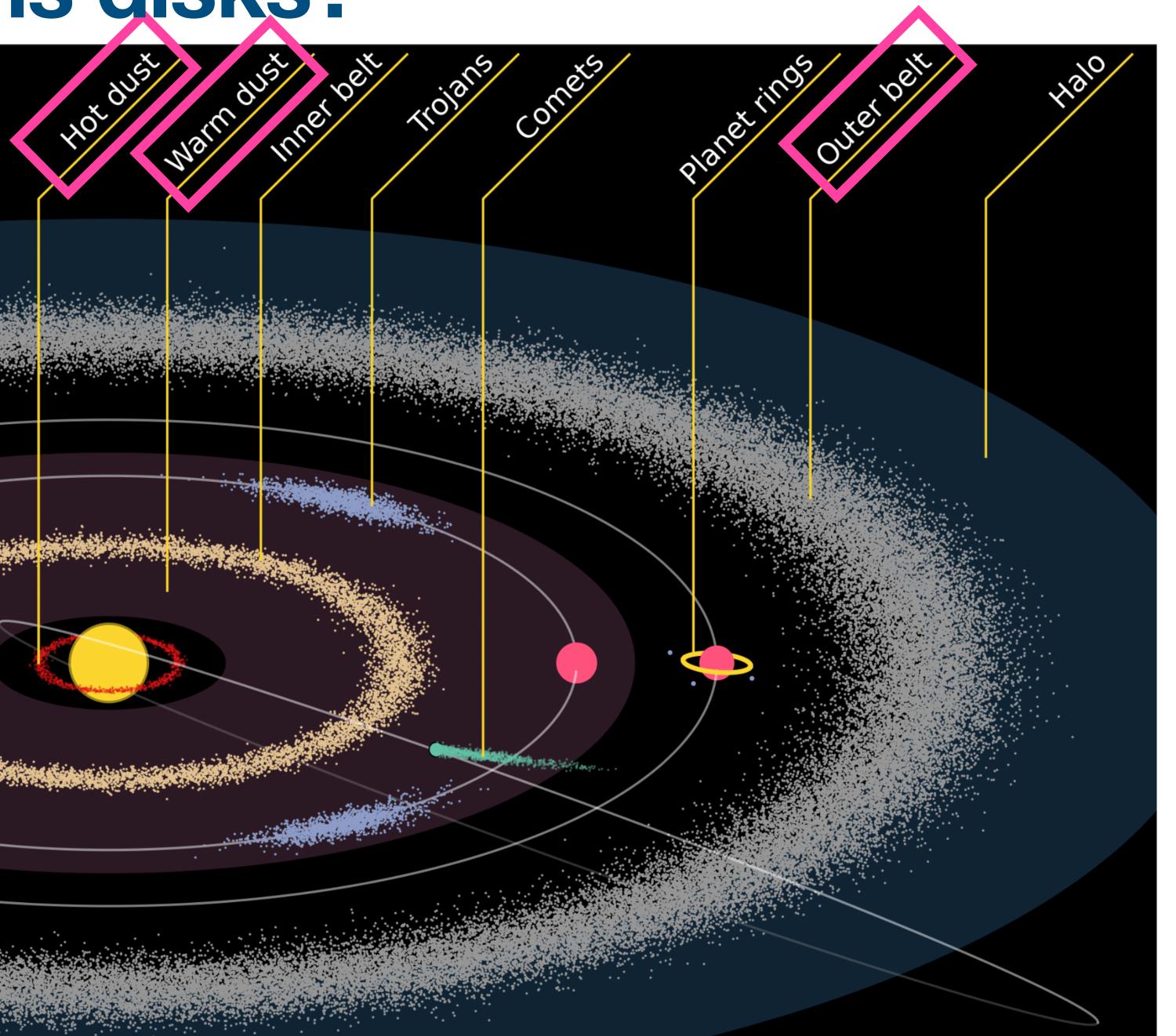


What are debris disks?

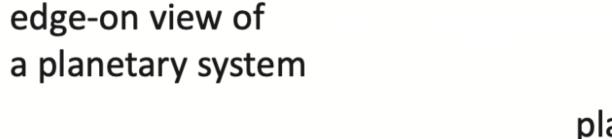
Hot belt ~ 1000K

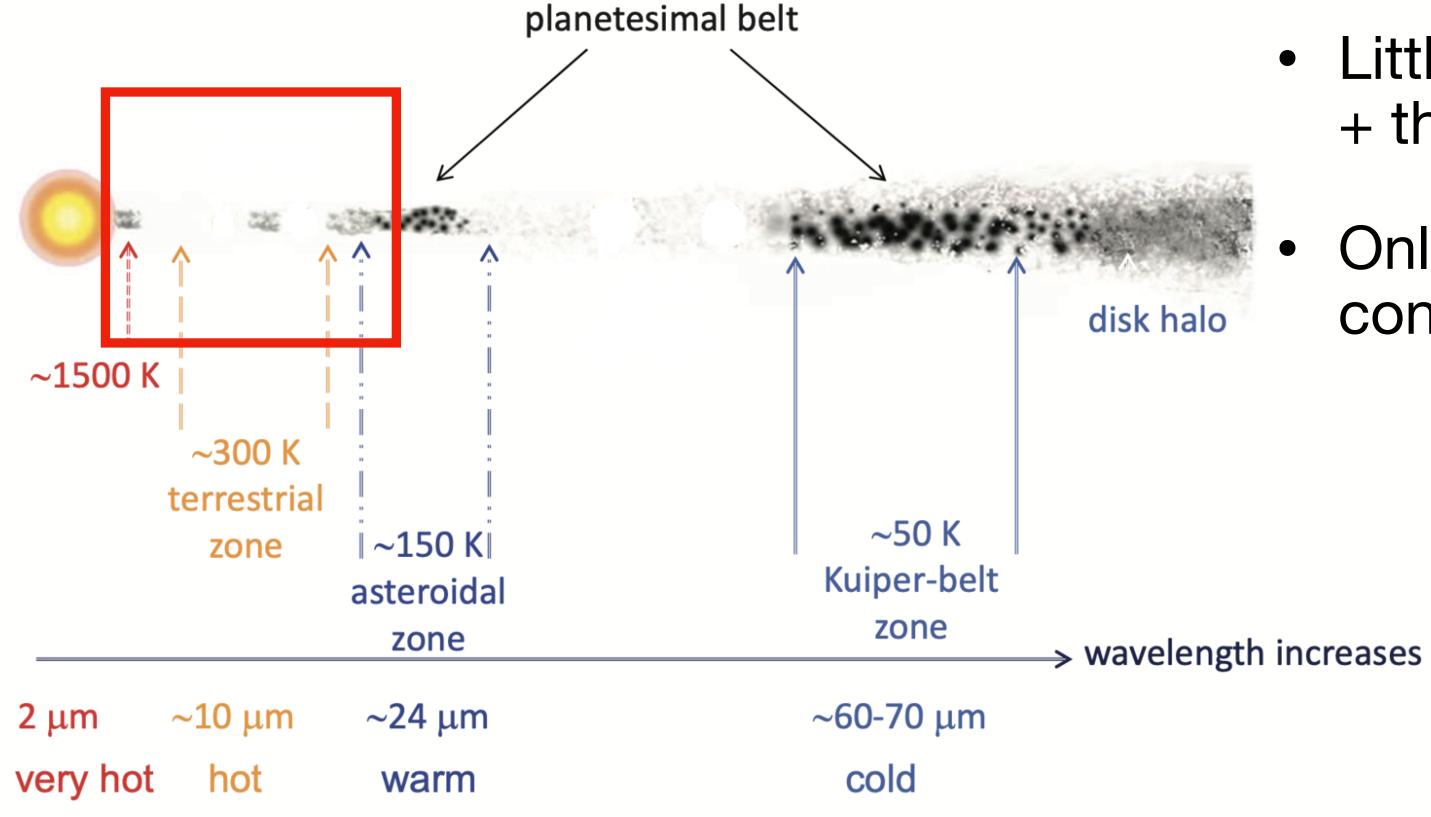
Warm belt ~ 100-300K

Cold belt ~ 50K



Exozodiacal dust

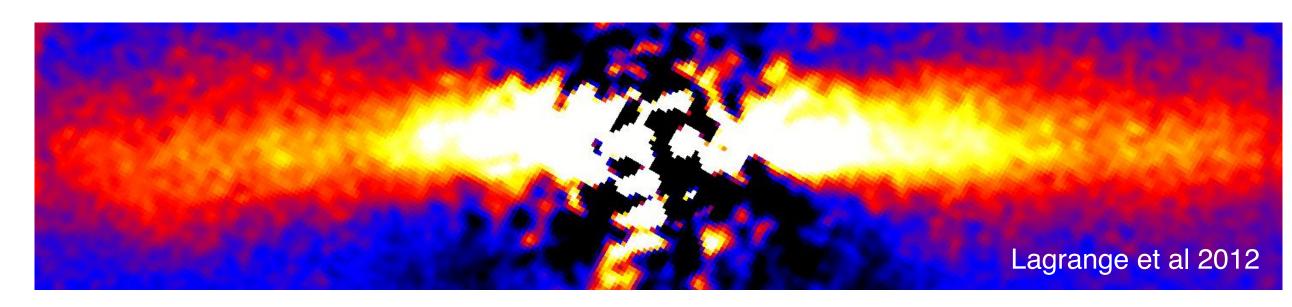




- Internal region: Mostly unconstrained
- Little information on the dust properties
 + their spatial distribution.
- Only the flux coming from this region is constrained.

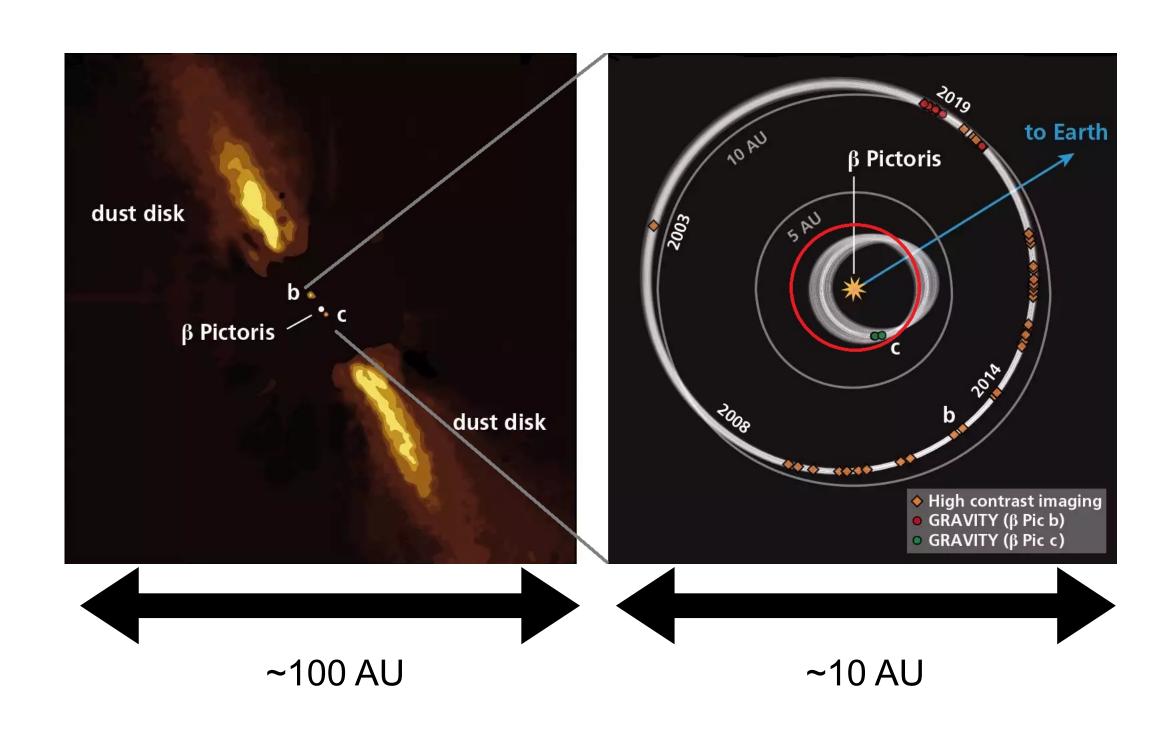


Beta pictoris



GRAVITY Collaboration / Axel M. Quetz, MPIA Graphics Department

- Edge-on debris disk
- Detected exozodi (PIONIER, GRAVITY)
- Exo-cometary activity.

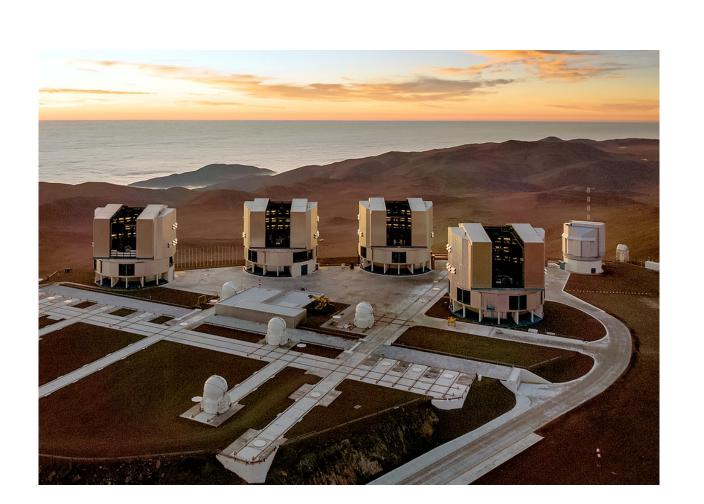


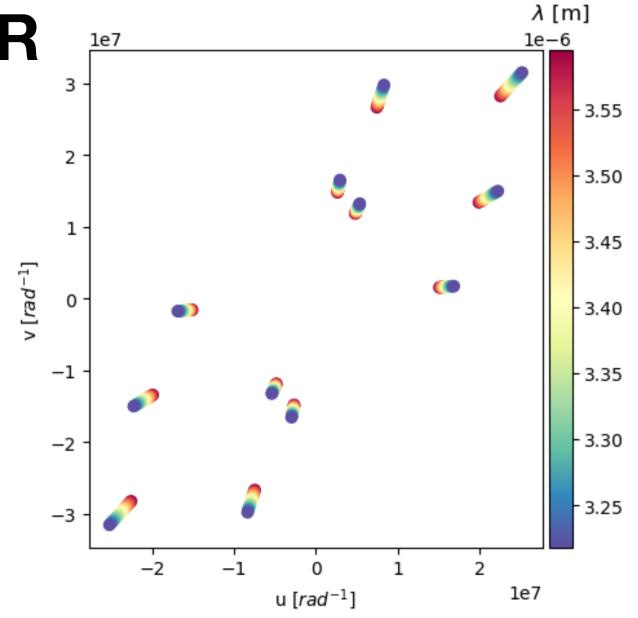
Observations

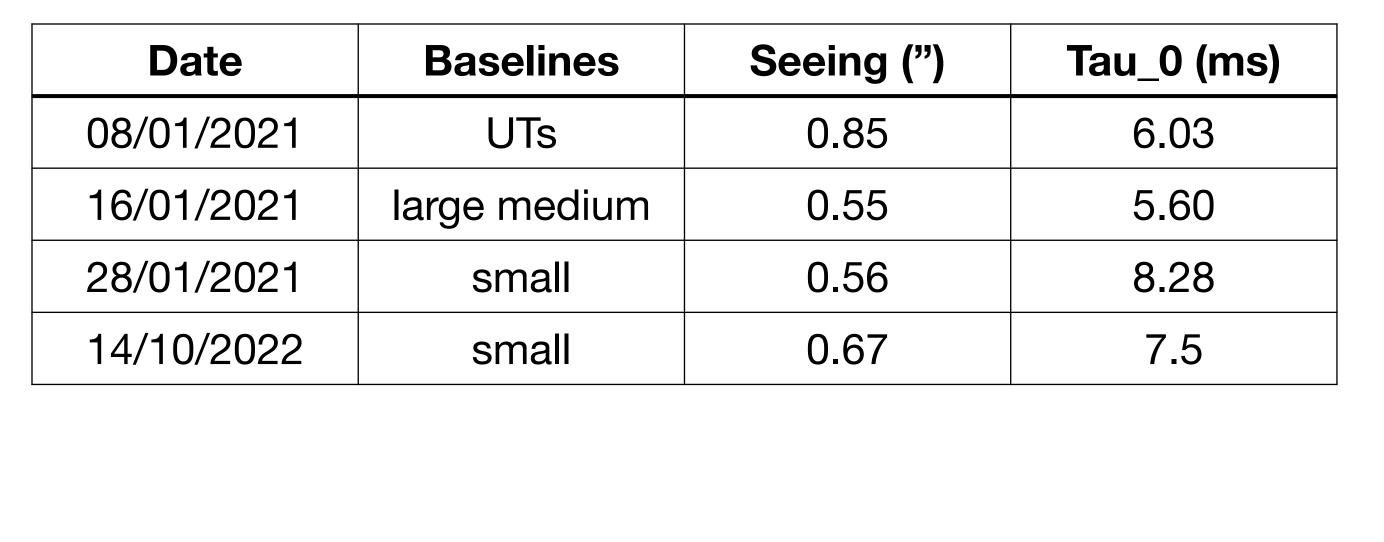
MATISSE

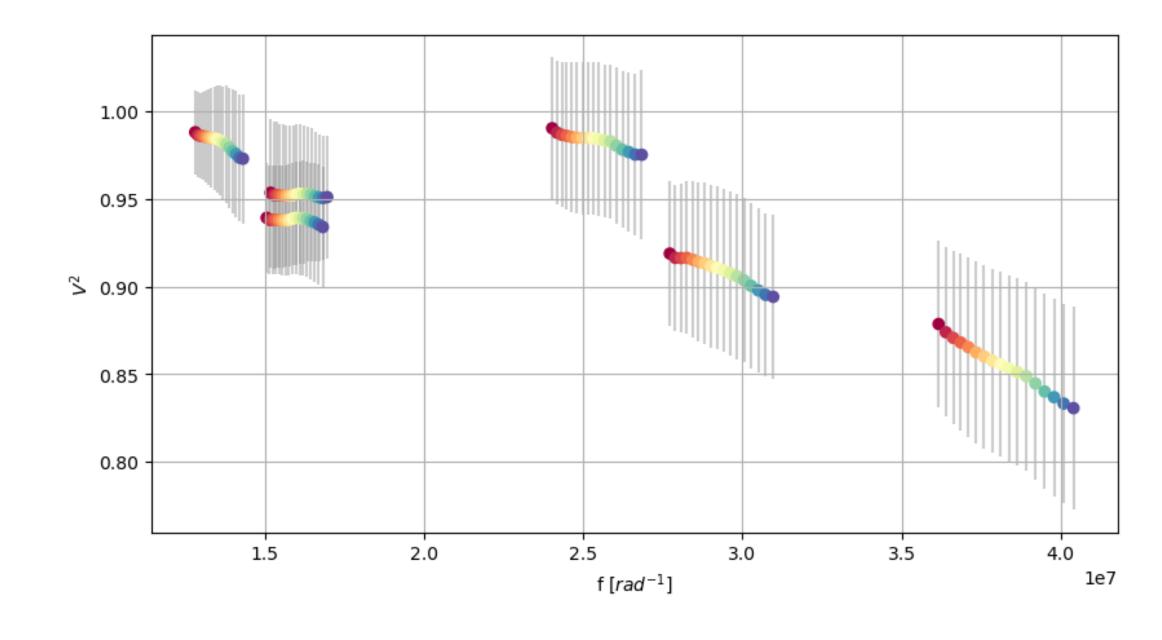
- MATISSE (L-band):
 - 4 observations
 - 3 in 2021
 - 1 in 2022

 We also have PIONIER and GRAVITY observations.



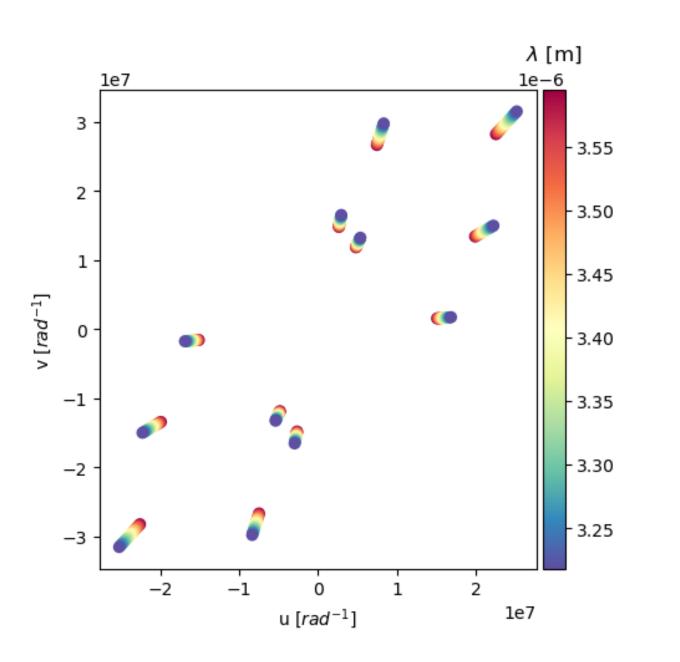


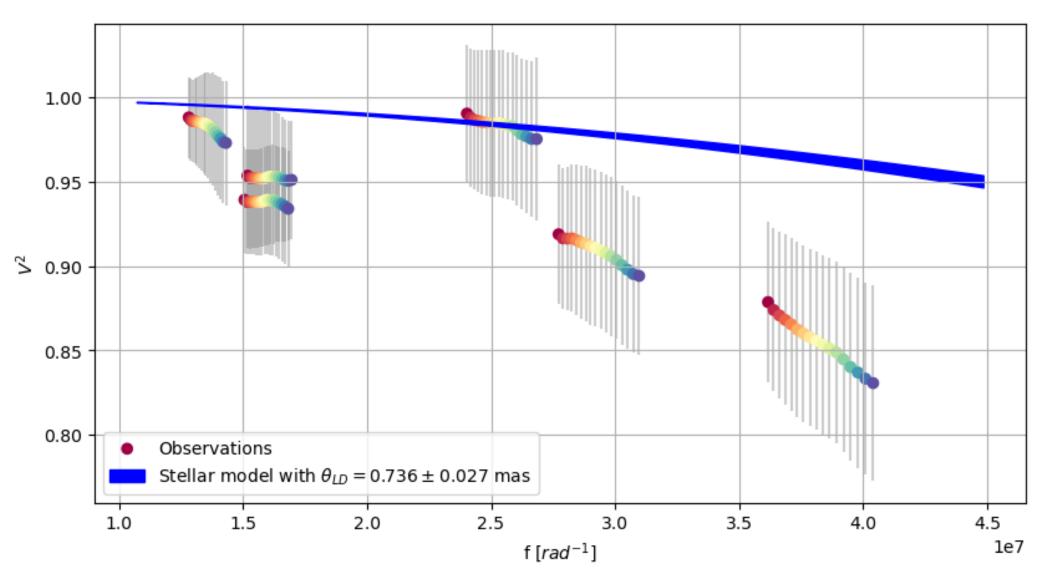




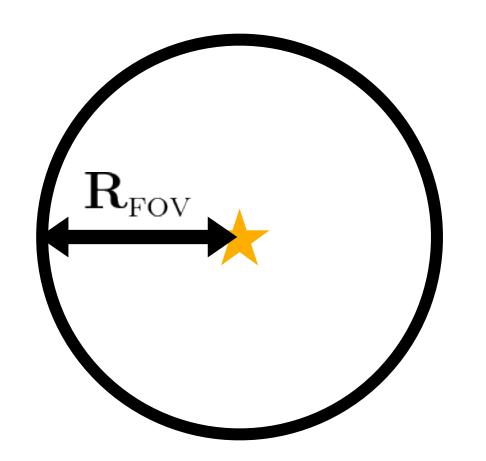
A resolved object around Beta Pic

• Visibility deficit with respect to the stellar visibility.





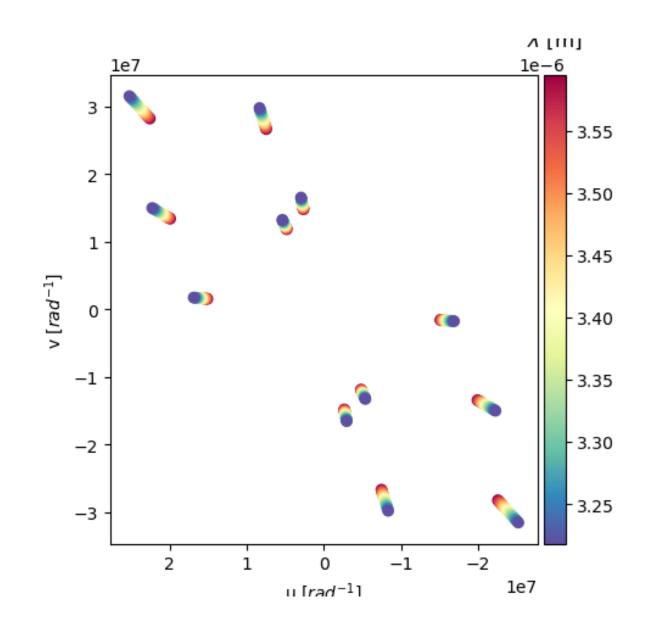


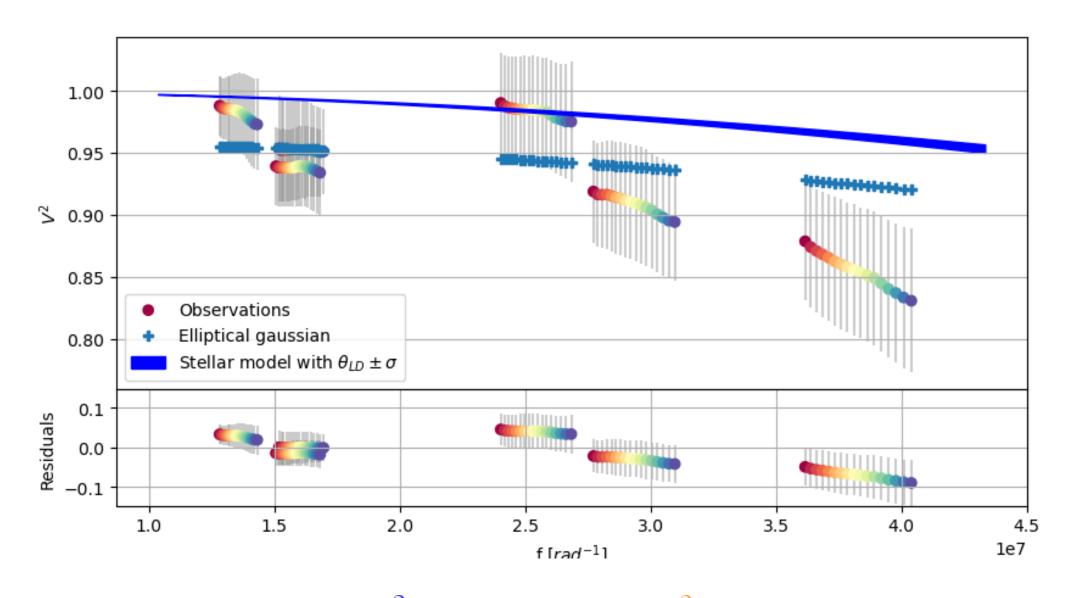


Homogeneous dust

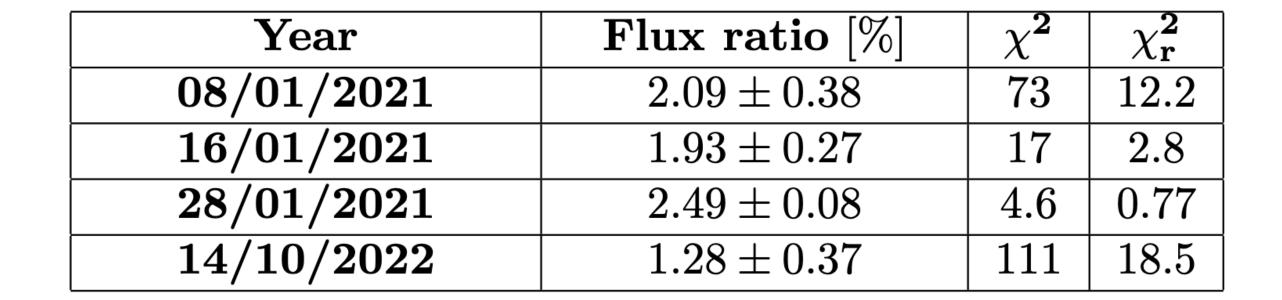
MATISSE

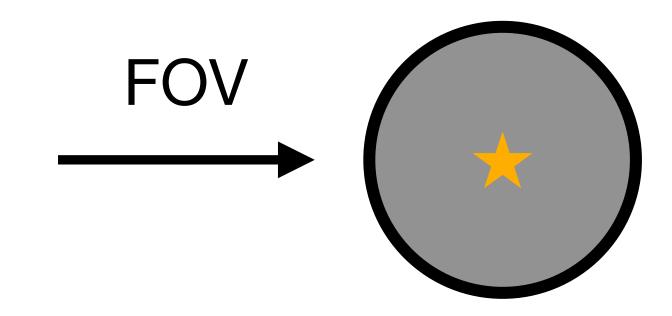
- Close to null residuals
- A circumstellar emission is **detected**.
- Absil et al. 2006, Ertel et al. 2014, etc.





$$V_{obs}^2 = (1 - 2f)V_{\star}^2$$
 $f = 2.09 \pm 0.38\%$





2D elliptical gaussian model

Flux ratio [%]

MATISSE

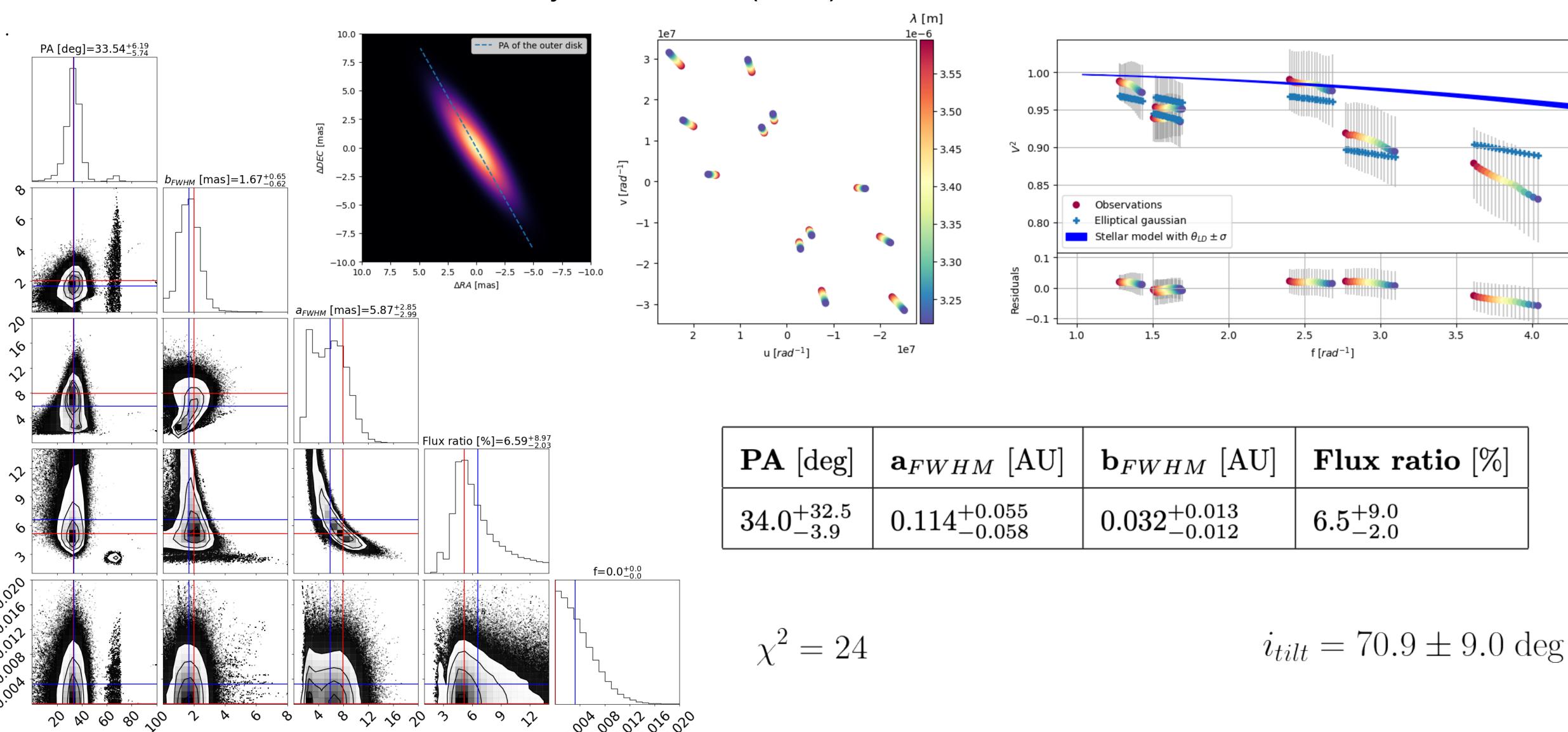
3.5

• Based on zodiacal dust models by Giese et al. (1986)

PA [deg]

 b_{FWHM} [mas]

a_{FWHM} [mas]

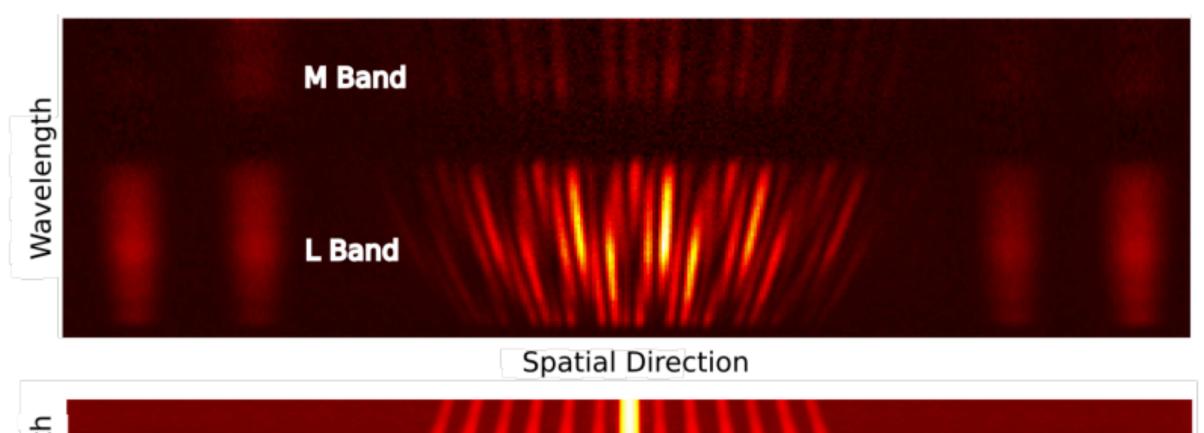


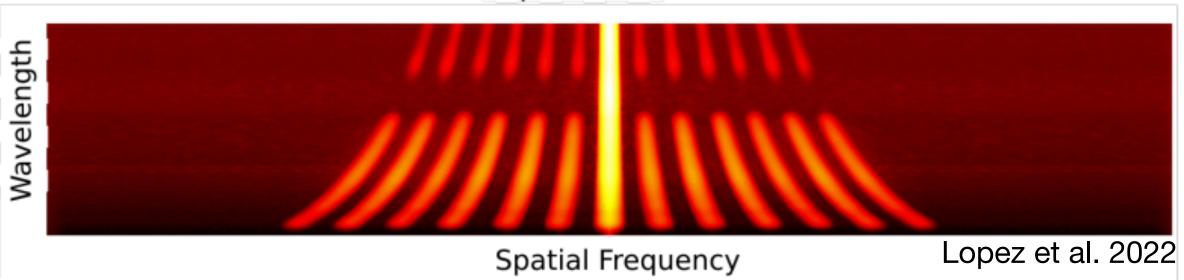
Correlations in MATISSE data

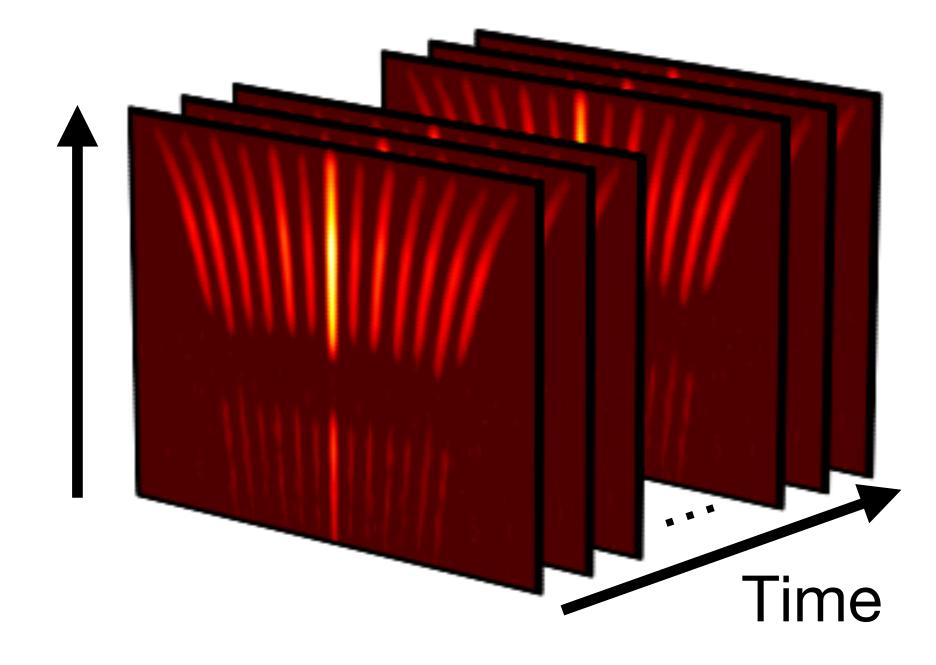
Intermediate files

- Photometry frames
- Correlated flux frames
- This is the last step where observations are handled frame by frame.
- We propagate the correlation matrix to the visibilities.

$$V_{ij}^2(\lambda) = \frac{C_{ij}^2(\lambda)}{\sum_x \left\langle P_{ij}(x,\lambda,t) \right\rangle_t} \qquad C_{ij}^2(\lambda) = \sum_u \left\langle |I(u,\lambda,t)|^2 - \beta \right\rangle_t$$



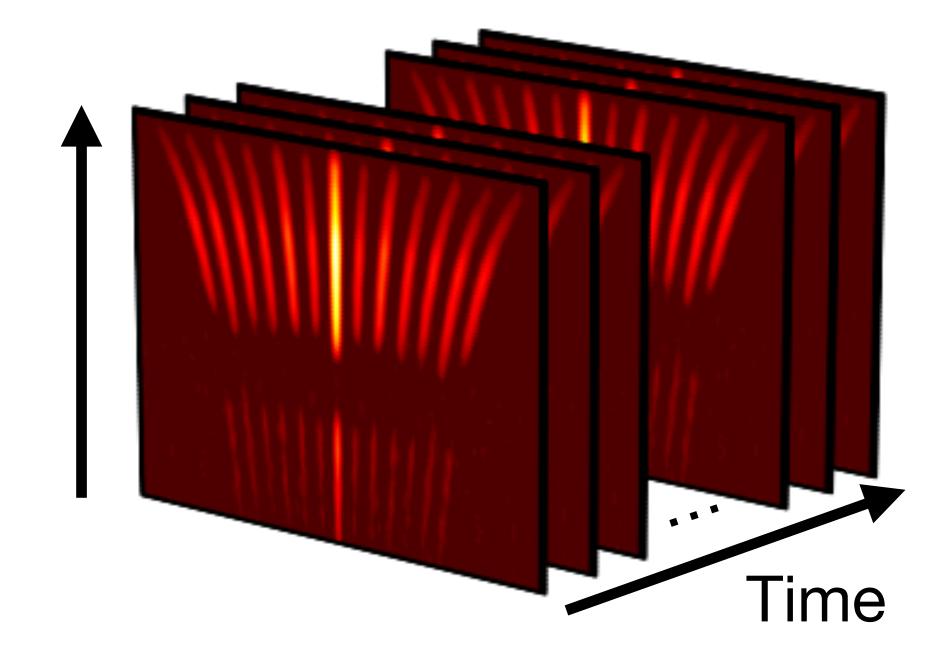


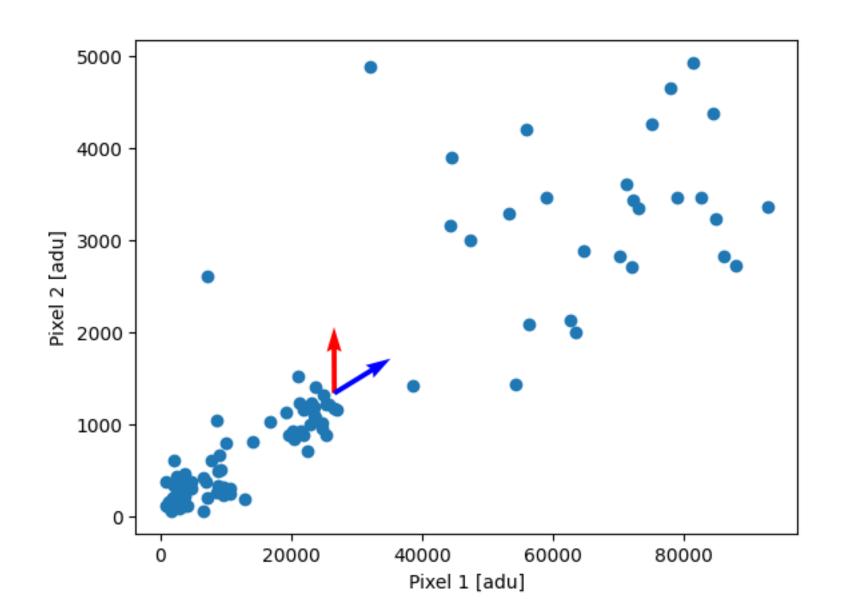


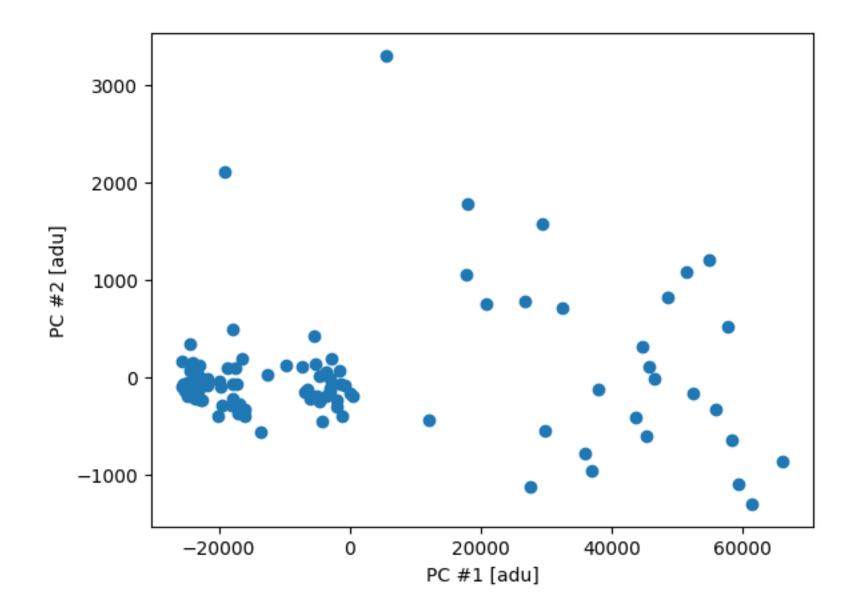
Propagating the correlation matrix

 1. Deriving the probability distribution that each pixel follows.

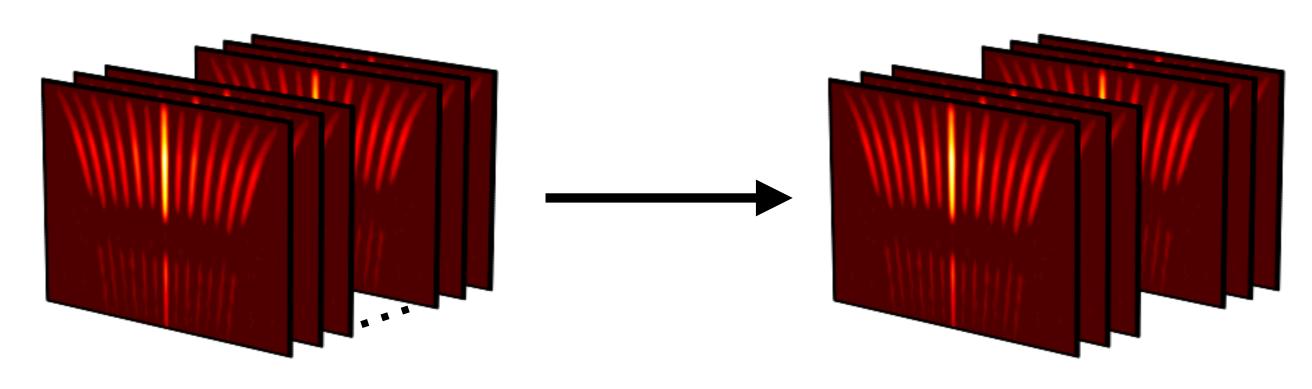
PCA







Propagating the correlation matrix

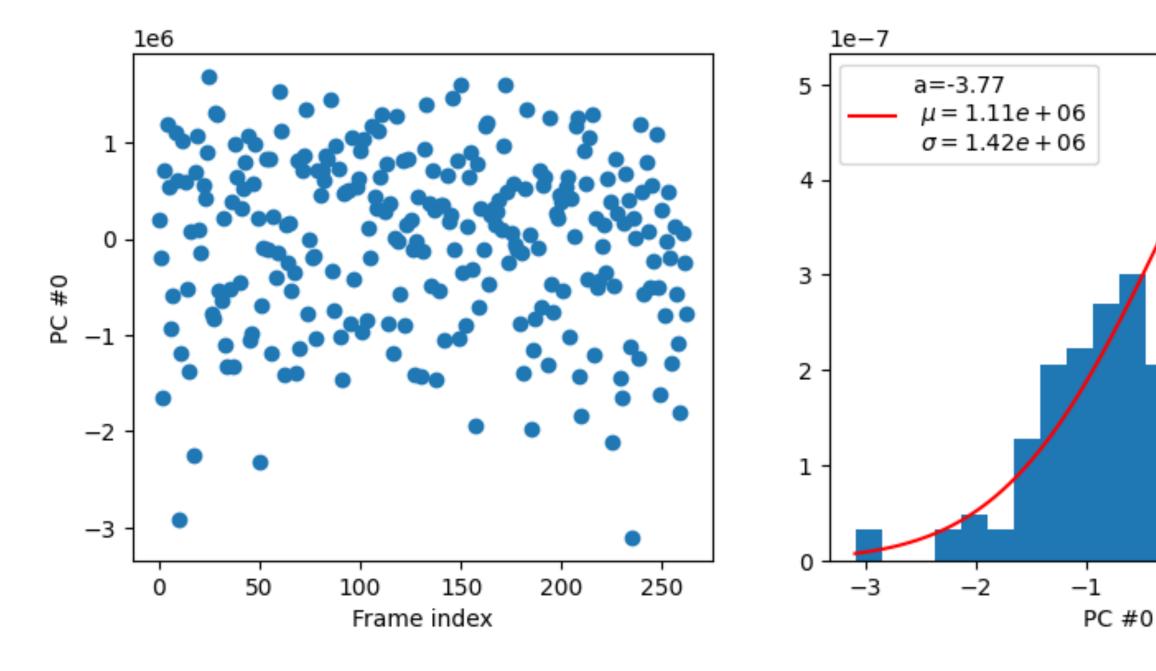


• 2. Generating new synthetic frames that follow the same distribution.



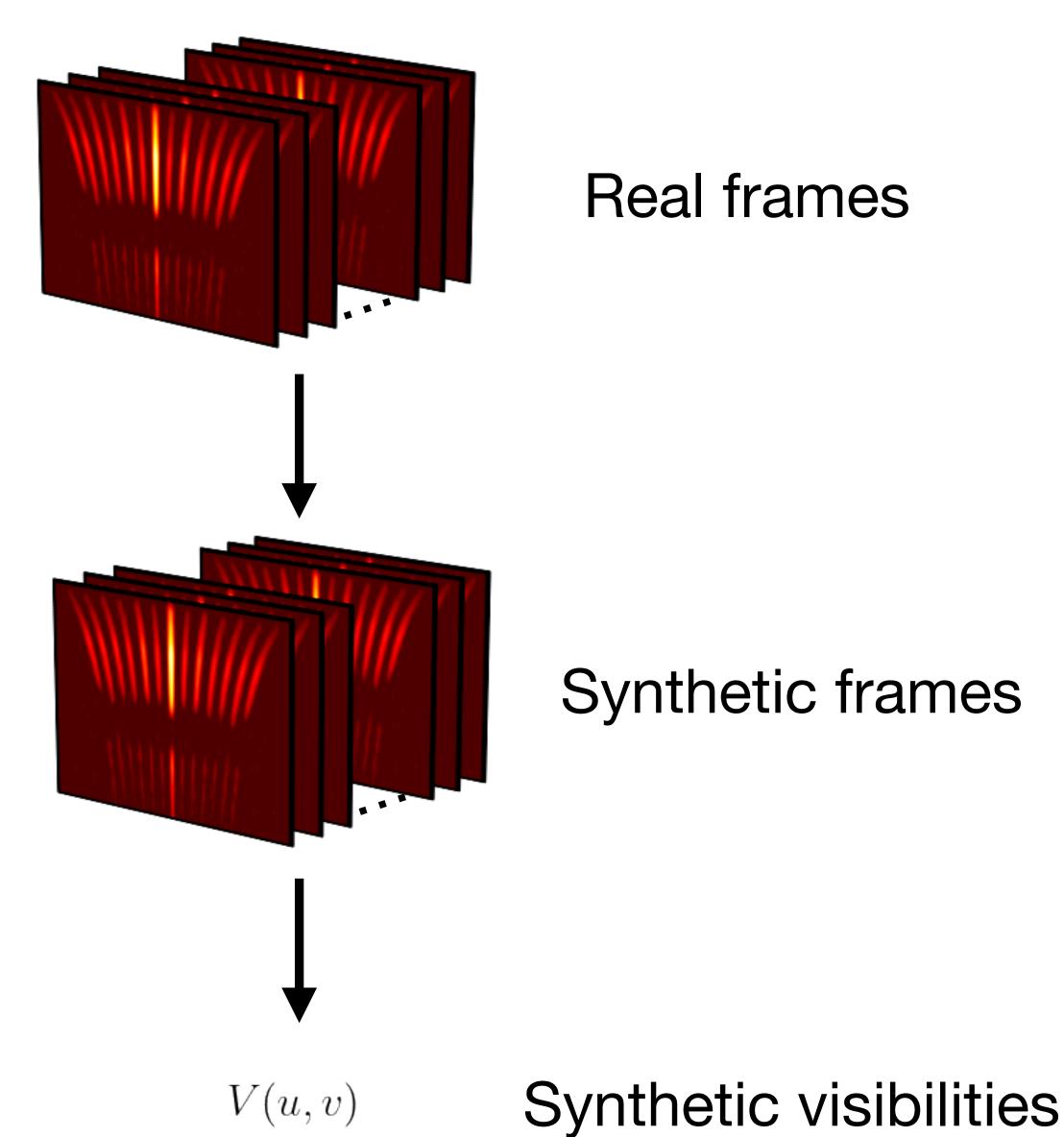
Synthetic frames

1e6

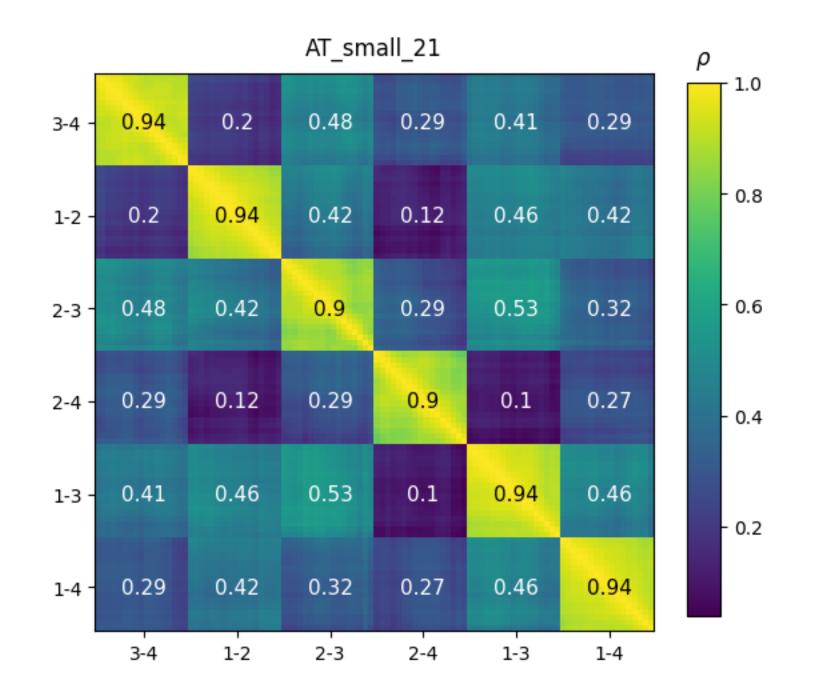


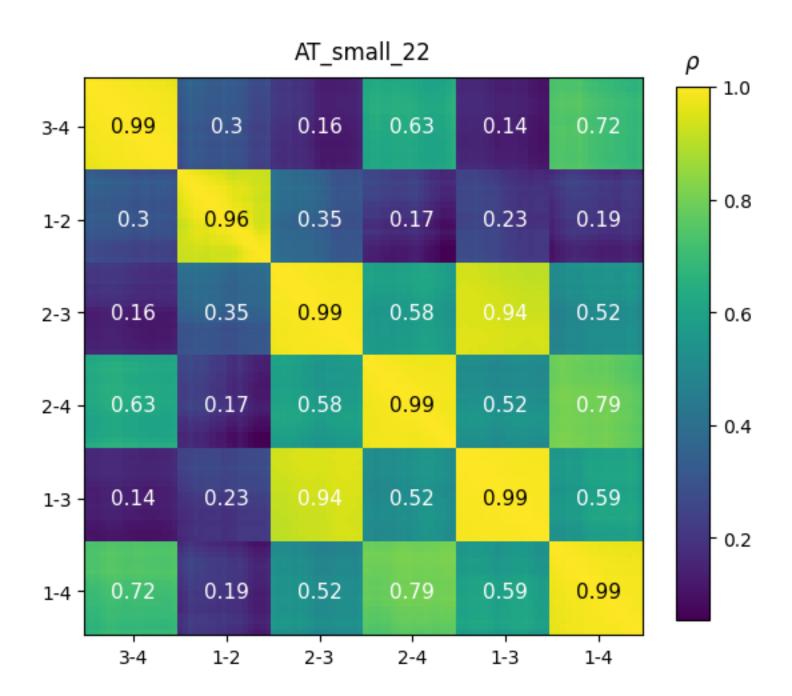
Propagating the correlation matrix

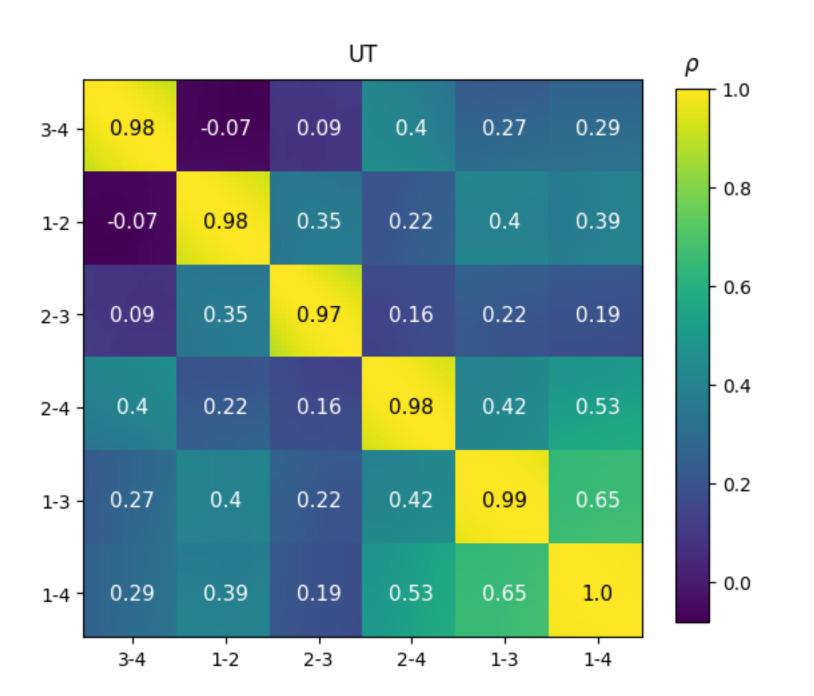
- 1. Deriving the probability distribution that each pixel follows.
- 2. Generating new synthetic frames that follow the same distribution.
- 3. Processing with DRS to obtain visibilities.
- 4. Calculating correlation matrices from the set of visibilities.



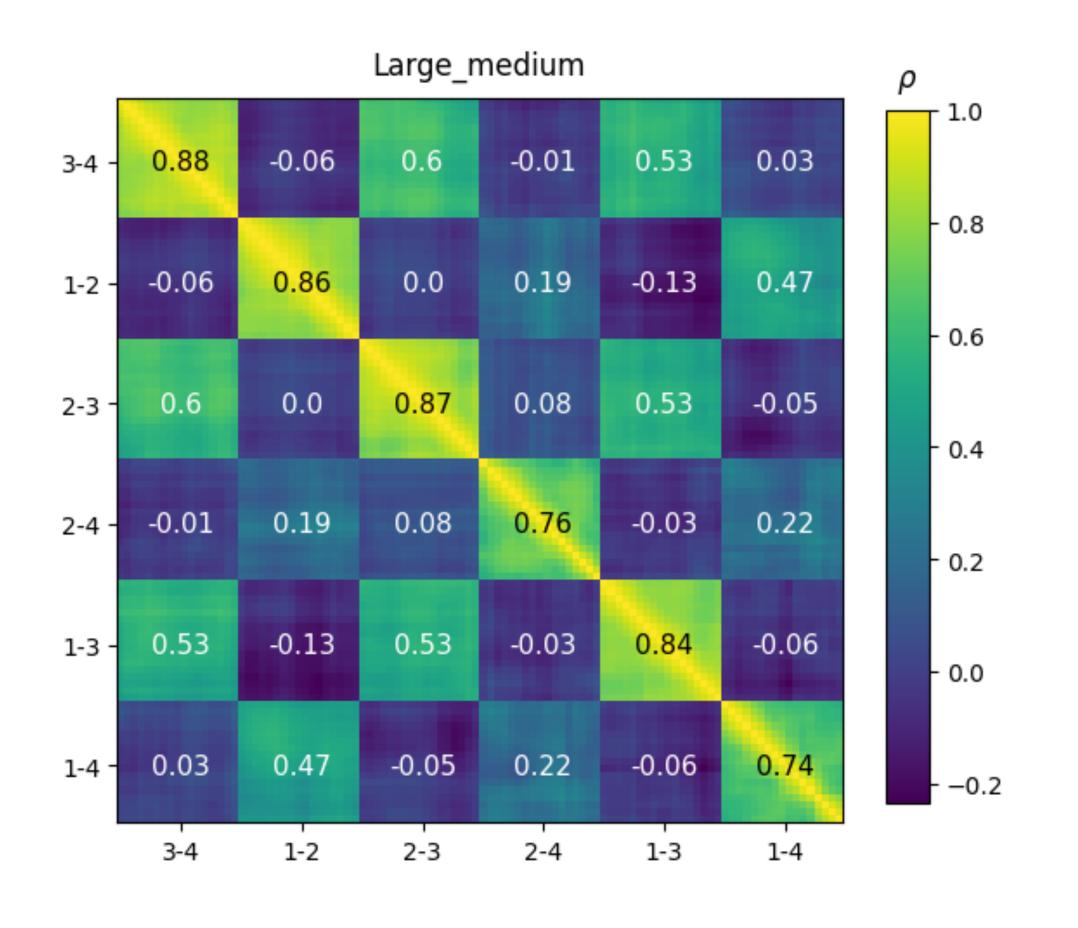
Results: Correlation matrices



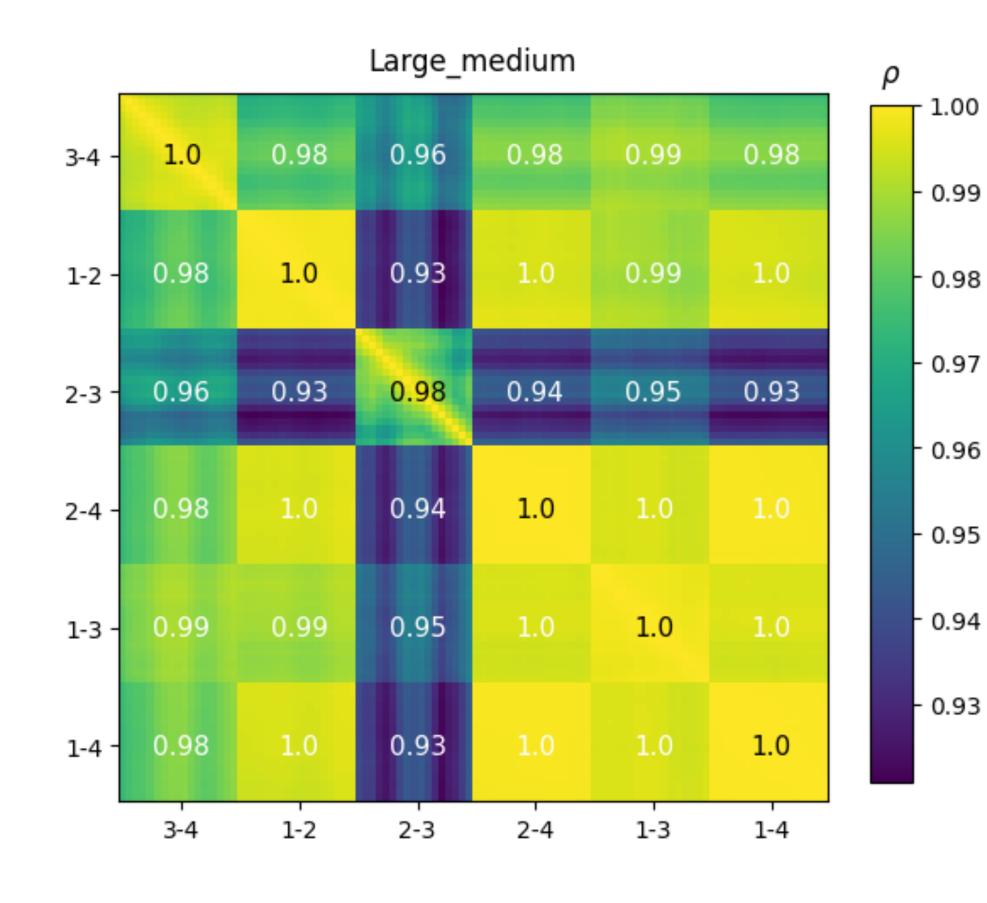




Impact of calibration error

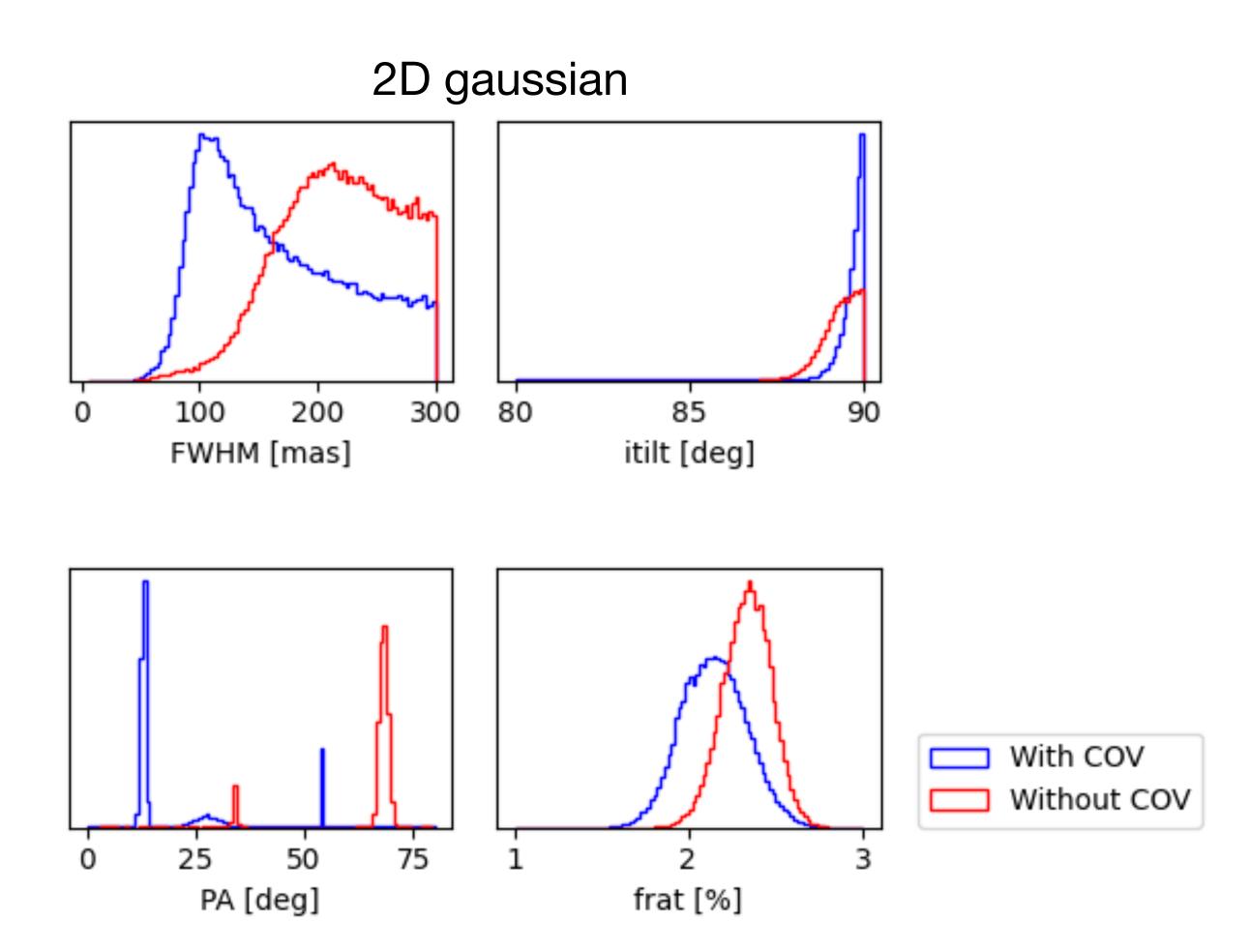


Propagating the error on the angular size of the calibrator.



Results: Impact on exozodi model fitting

- Impact on the median value of the distribution
- Impact on the shape/ variance of the distribution.
- Especially important for low u-v coverage



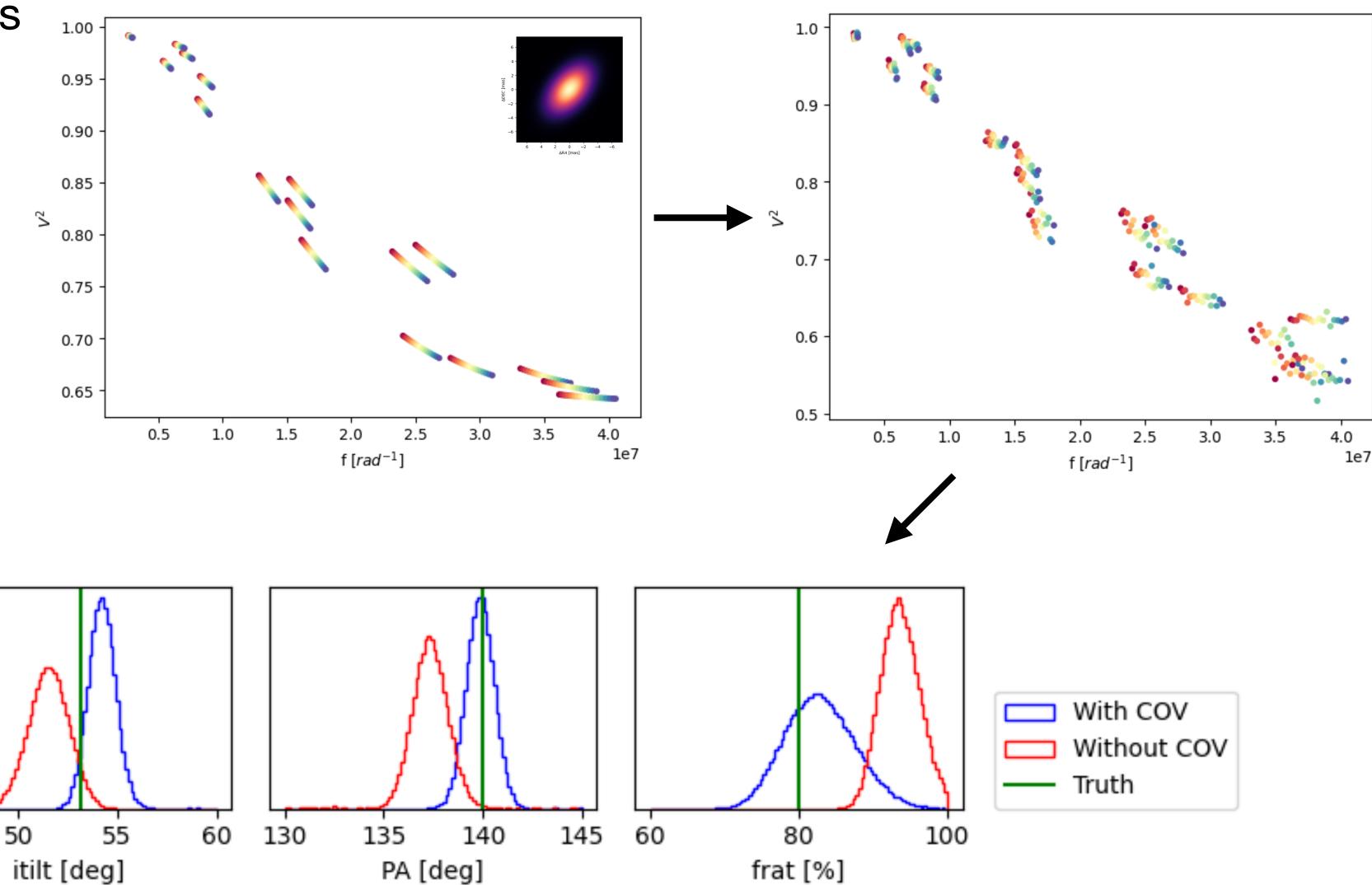
Results: Impact on model fitting

45

- We generate model visibilities using the covariance matrices of the real observations.
- We try to **retrieve** the input parameters in two ways:
 - Assuming the errors are independent.

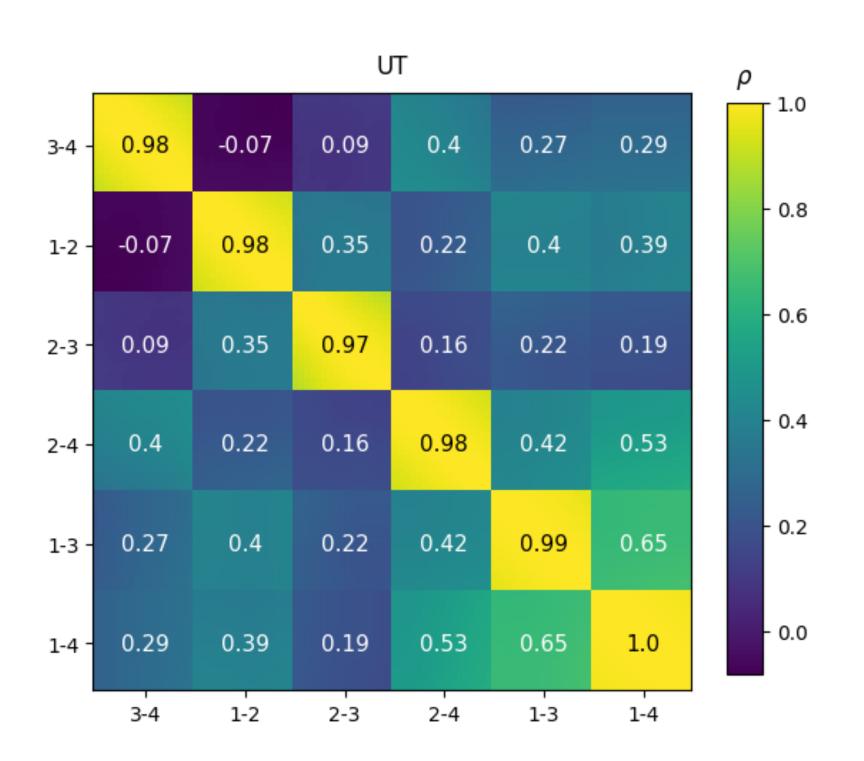
FWHM [mas]

 Assuming they are correlated.



Conclusions and perspectives

- We developed a method to systematically generate correlation matrices of MATISSE data
 - Correlations have a strong impact on model fitting.
 - MATISSE data might be highly correlated, especially between spectral channels.
 - Calibration induces strong correlations.



Thank you!

Comments and questions are always appreciated