

Direct Exoplanet Detection Using L1 Norm Low-Rank Approximation

**Hazan Daglayan¹, Simon Vary¹, Valentin
Leplat², Nicolas Gillis³, and P.-A. Absil¹**

¹ ICTEAM/UCLouvain, Belgium

²Center of Artificial Intelligence Technology, Skoltech, Russia

³ University of Mons, Belgium

Exoplanet imaging



Exoplanet imaging

The image displays two news articles side-by-side, both featuring exoplanets.

euronews.next article:

Signs of life? James Webb reveals more about exoplanet K2-18 b's atmosphere

A large image shows a blue exoplanet against a dark background with a small red star visible.

By Luke Wurst
Published on 12/08/2023 - 17:35 • Updated 13/08/2023 - 08:38

BBC NEWS article:

Exoplanet discovered around neighbouring star

A large image shows a brownish exoplanet against a black background.

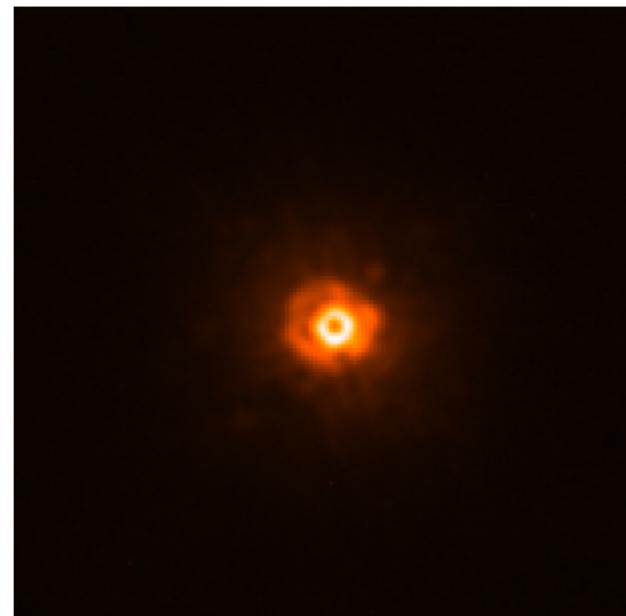
By Paul Rincon
Published on 14 November 2010

Exoplanet imaging



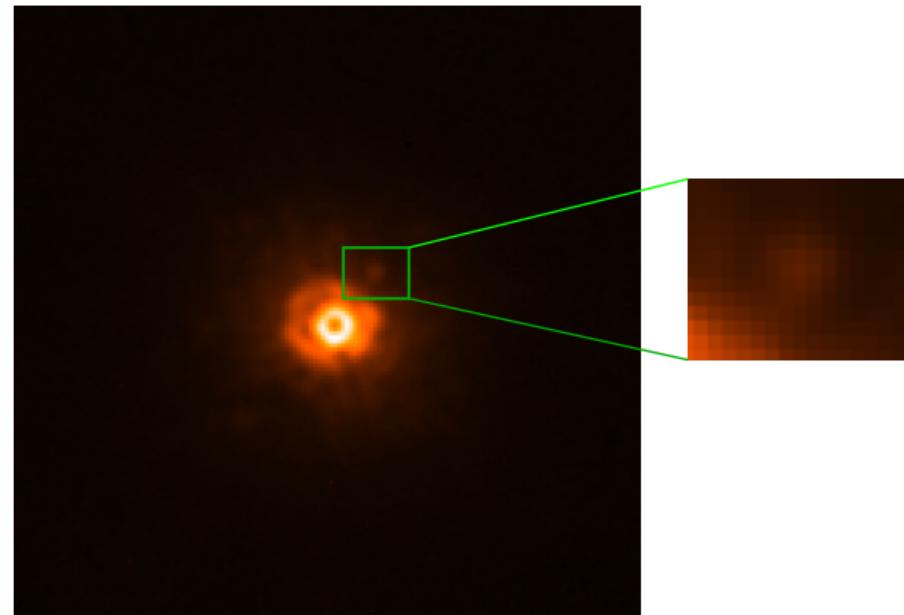
Exoplanet imaging

A real image



Exoplanet imaging

A real image of exoplanet



Direct imaging



Credit: <https://exoplanets.nasa.gov/>

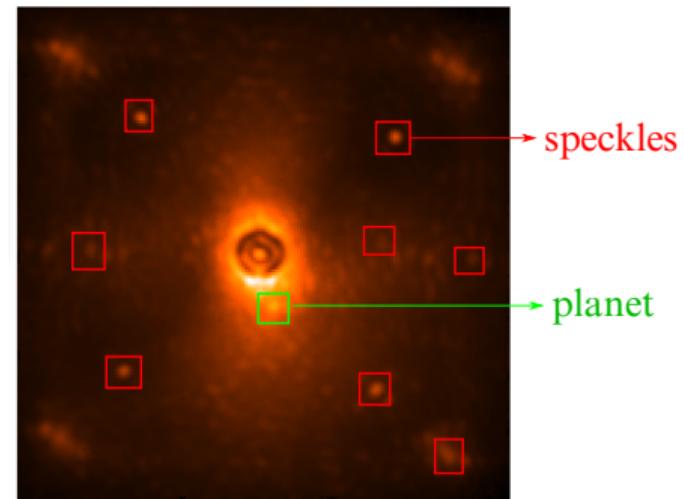
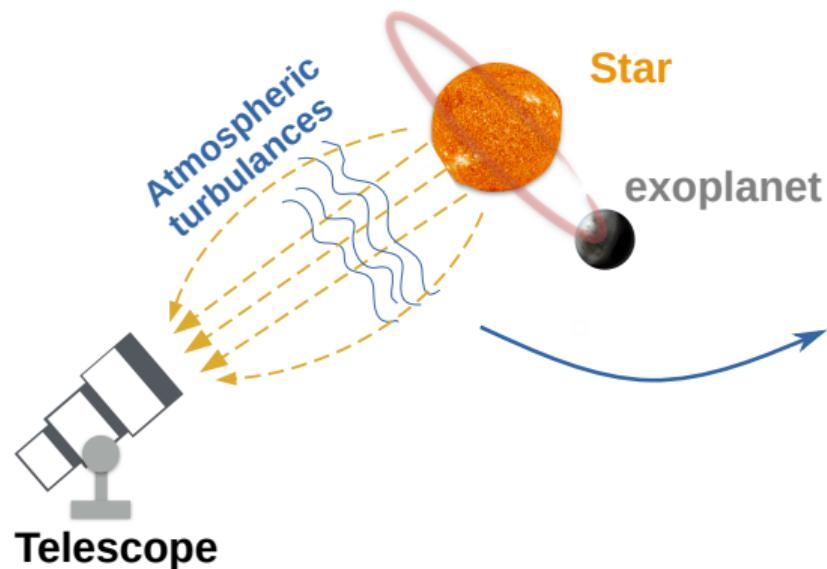
Direct imaging



Credit: <https://exoplanets.nasa.gov/>

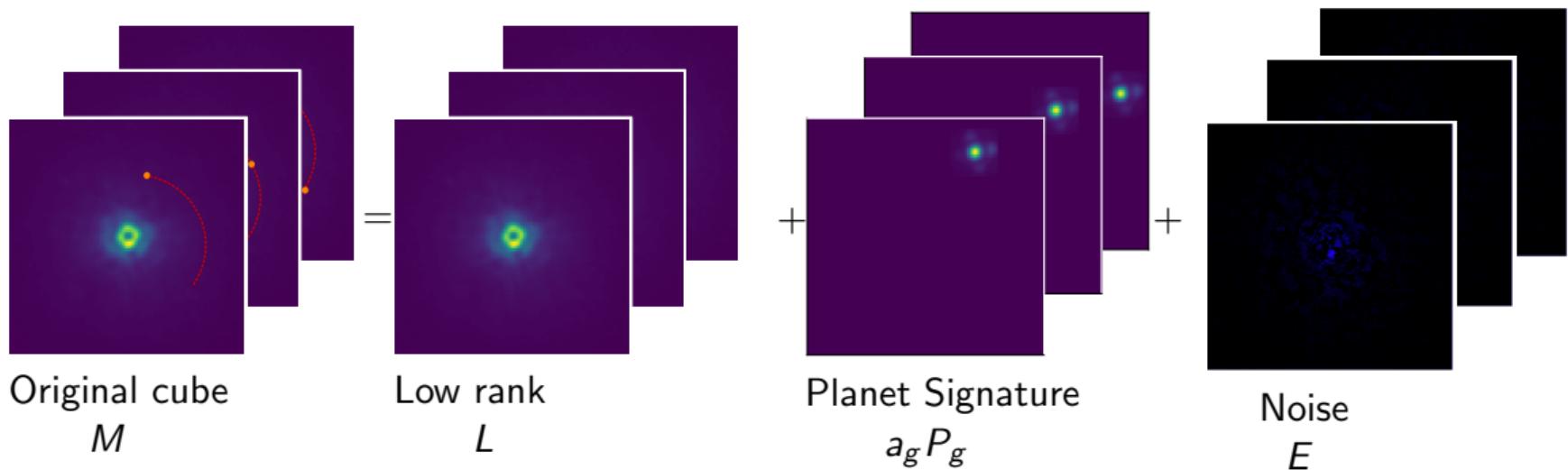
- ▶ firefly → exoplanet
- ▶ lighthouse → star

Direct Imaging

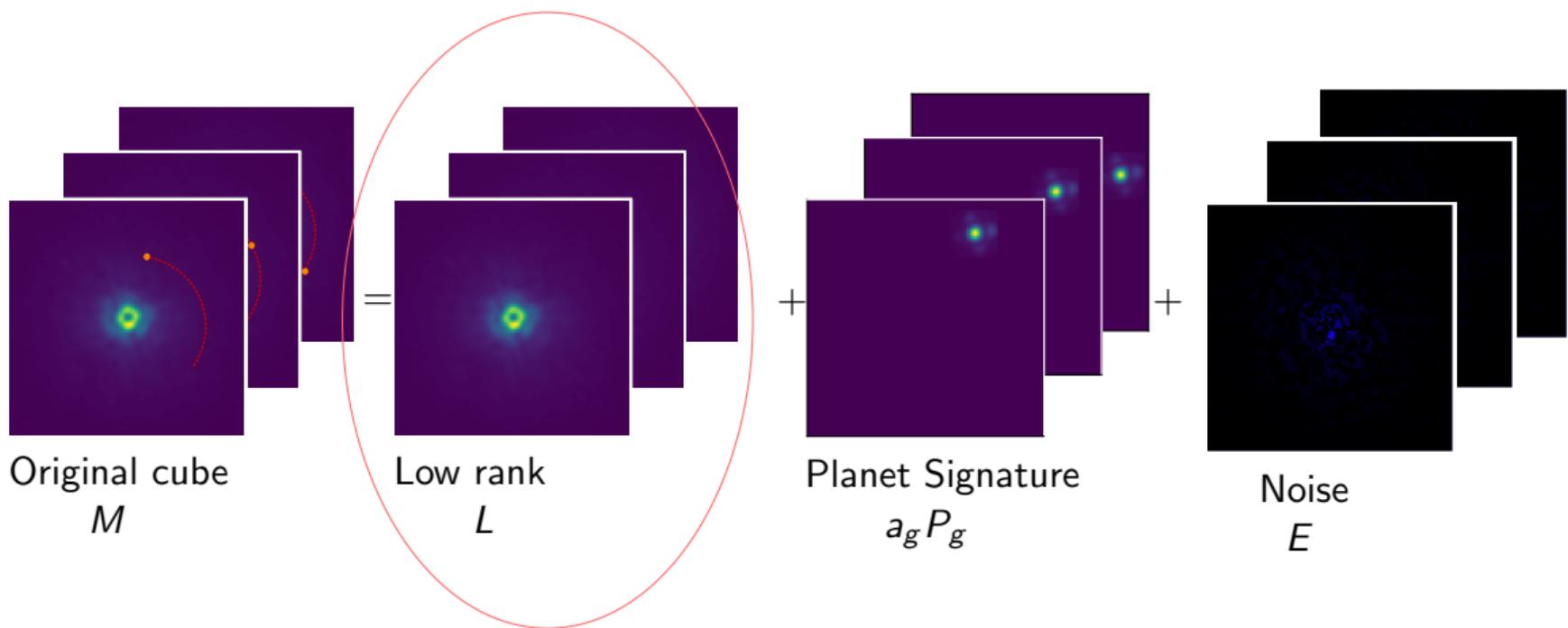


Angular differential imaging

Problem setup & goal



Problem setup & goal



Background: (annular) PCA^{1,2}

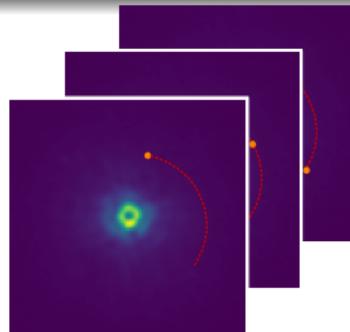
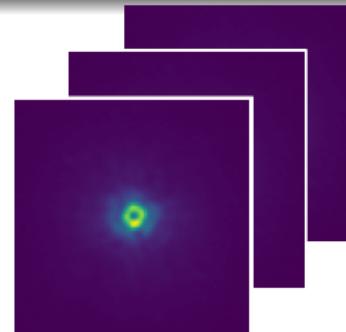


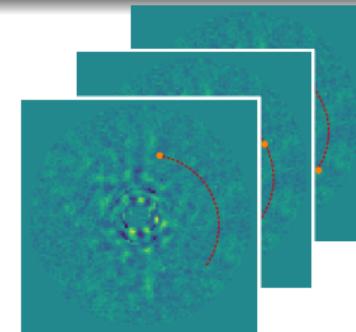
Image sequence
 M

=

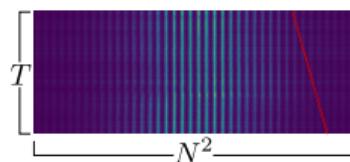


Low rank
 L

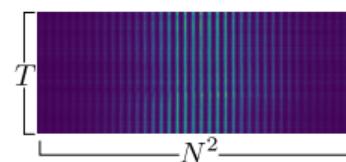
+



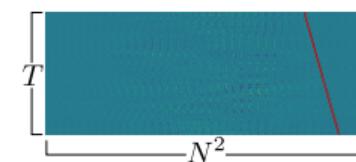
Foreground
 R



=



+

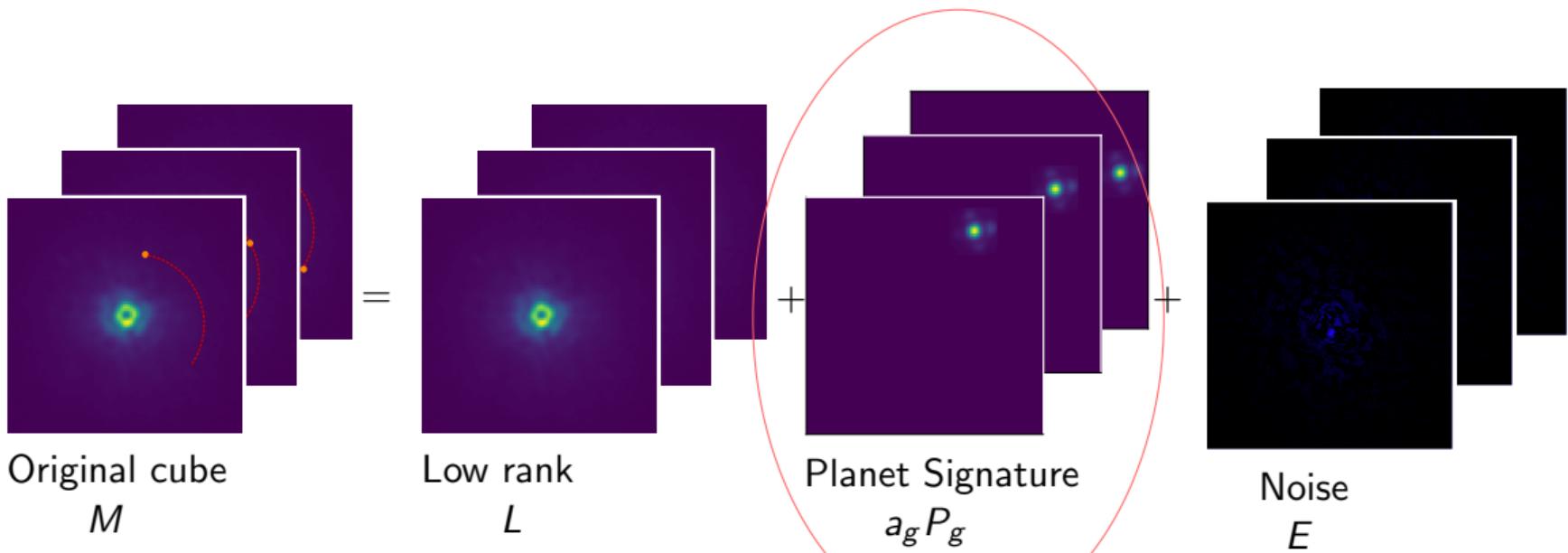


$$\hat{L} = \arg \min_L \|M - L\|_F \quad \text{subject to} \quad \text{rank}(L) \leq k$$

¹Amara and Quanz, 2012

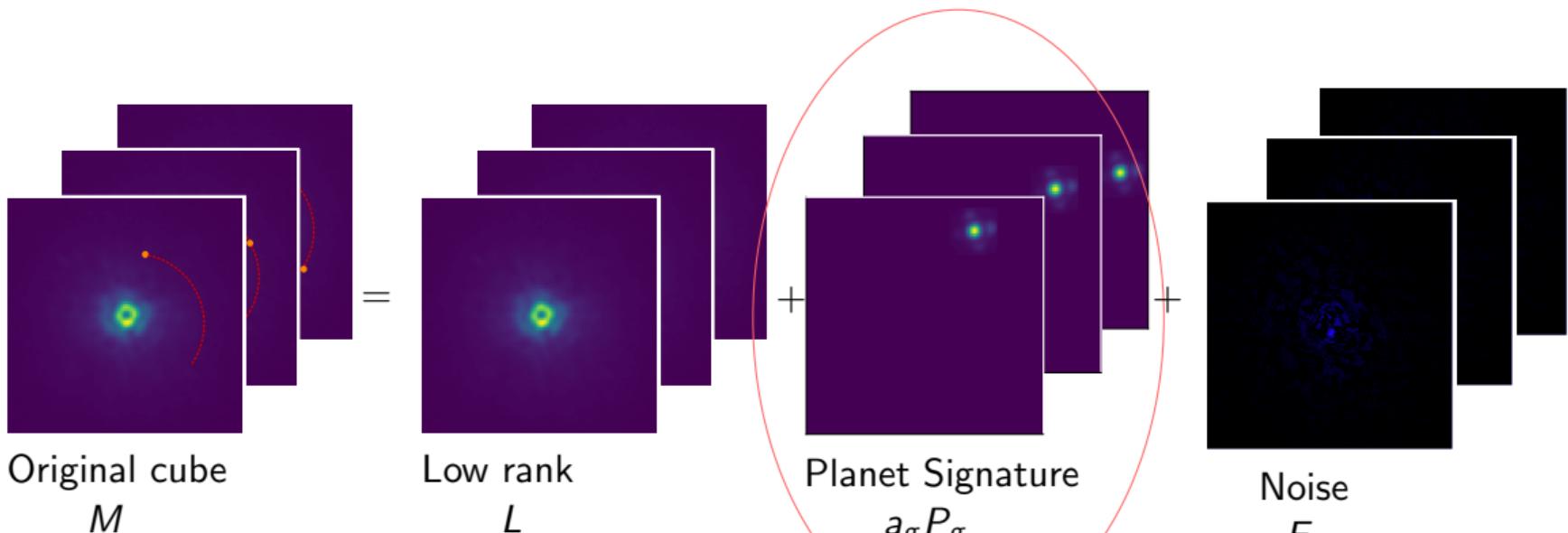
²Soummer, et al., 2012

The intensity of planet signature



$$\hat{a}_g = \arg \min_{a_g > 0} \|M - \hat{L} - a_g P_g\|_2,$$

The intensity of planet signature



$$\hat{a}_g = \arg \min_{a_g > 0} \|M - \hat{L} - a_g P_g\|_2,$$

$$\hat{a}_g = \frac{\langle P_g, M - \hat{L} \rangle}{\|P_g\|_F^2}$$

The intensity of planet signature

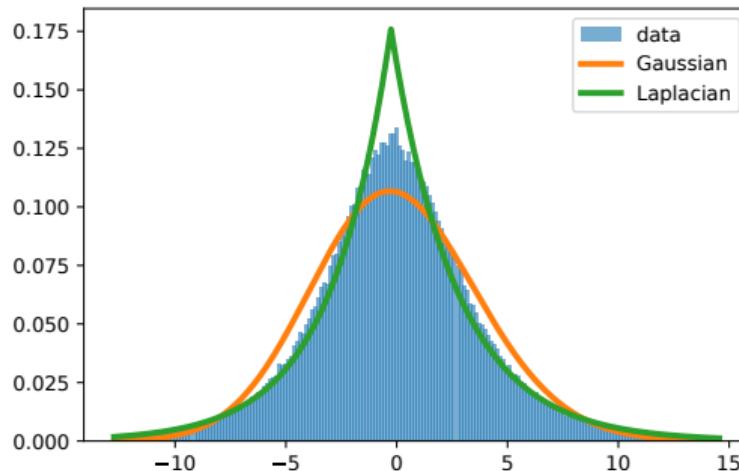
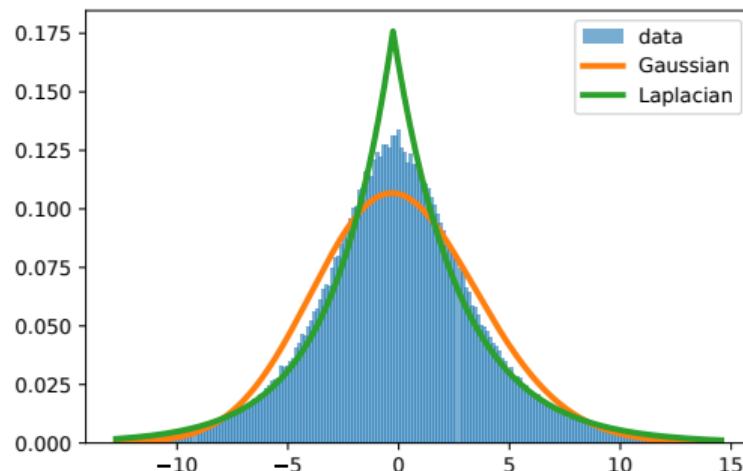


Figure: Residual data at $10\lambda/D$ ^{3,4}

³Pairet, et al., 2019

⁴Daglayan, et al., 2022

The intensity of planet signature



$$\hat{a}_g = \arg \min_{a_g > 0} \| M - \hat{L} - a_g P_g \|_1$$

Figure: Residual data at $10\lambda/D$ ^{3,4}

³Pairet, et al., 2019

⁴Daglayan, et al., 2022

The intensity of planet signature

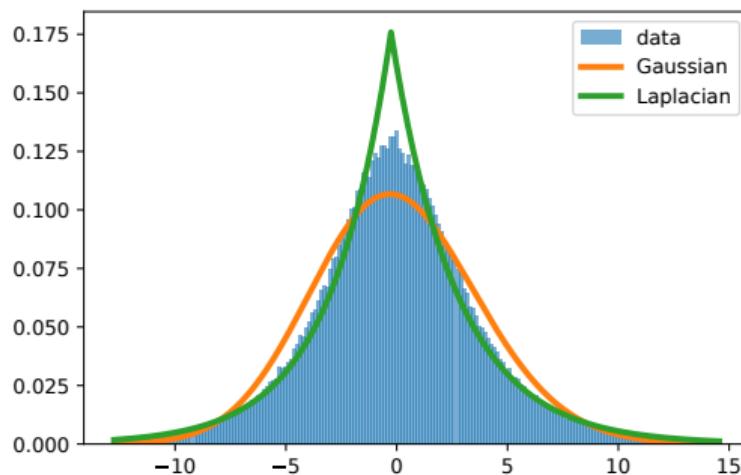


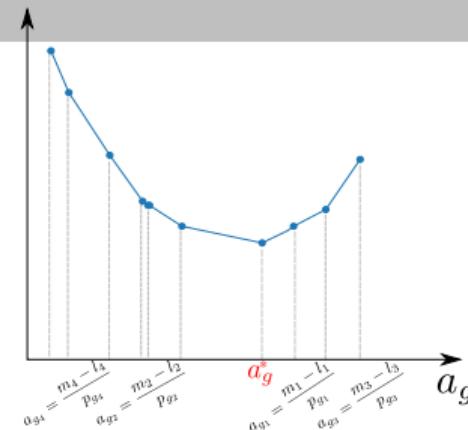
Figure: Residual data at $10\lambda/D$ ^{3,4}

³Pairet, et al., 2019

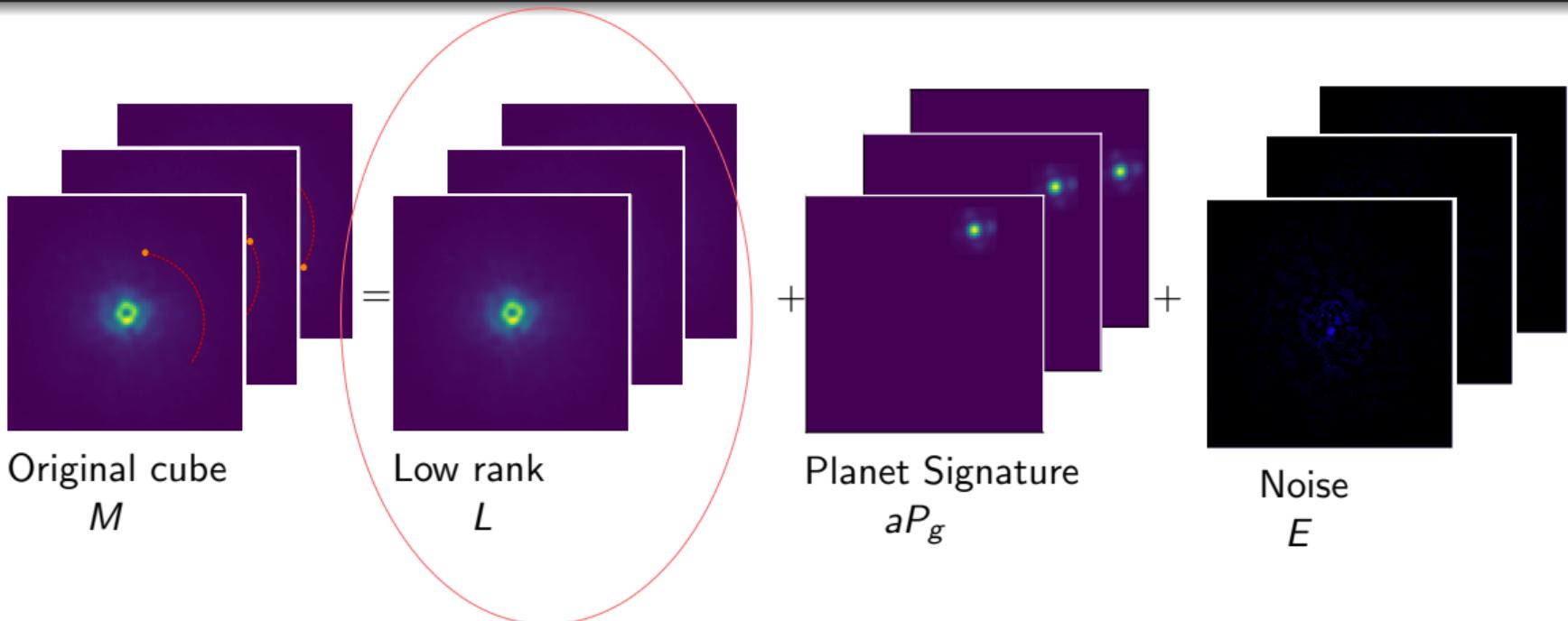
⁴Daglayan, et al., 2022

$$\hat{a}_g = \arg \min_{a_g > 0} \| M - \hat{L} - a_g P_g \|_1$$

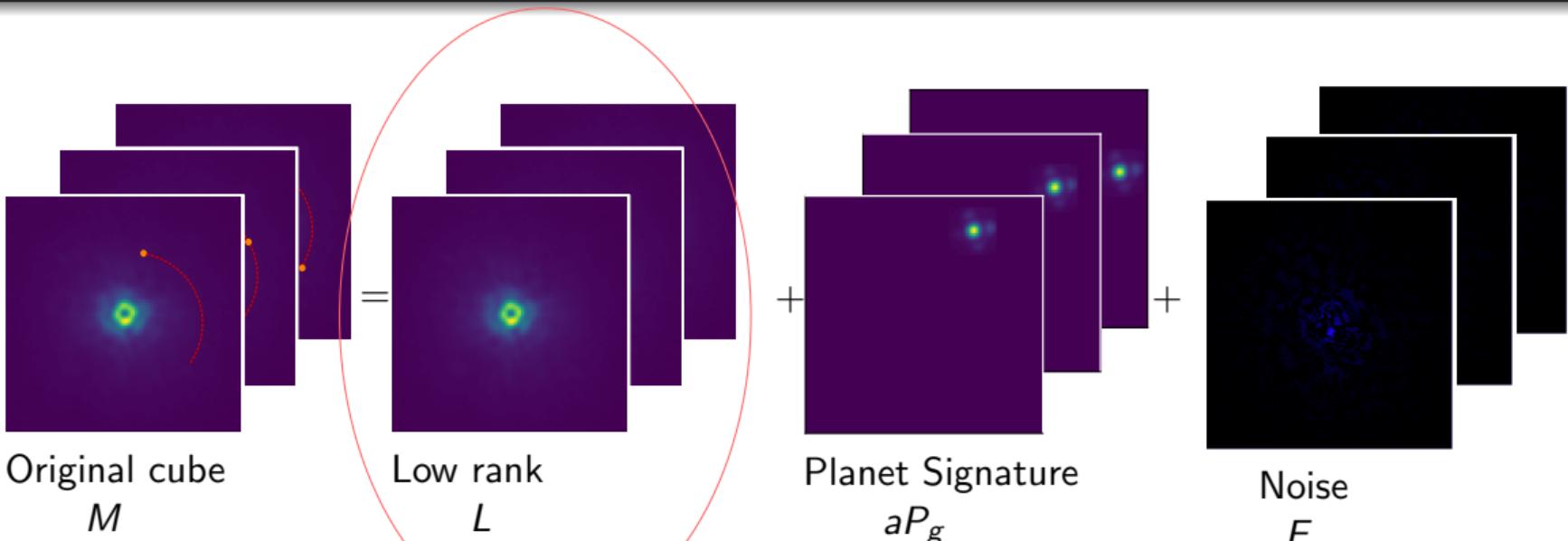
$$\hat{a}_g = \arg \min_{a_g \in \mathbb{R}} \sum_j |m_j - l_j - a_g p_{gj}|$$



Background: (annular) L1-LRA



Background: (annular) L1-LRA



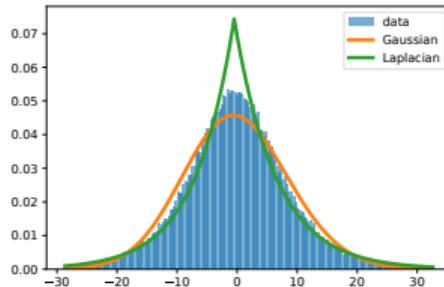
$$\hat{L} = \arg \min_L \|M - L\|_1 \quad \text{subject to} \quad \text{rank}(L) \leq k$$

- Solved using an exact block-cyclic coordinate descent method⁵.

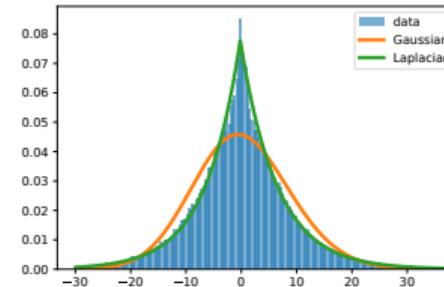
⁵Gillis and Plemmons, 2011

Data distributions of residual matrix after PCA/L1-LRA

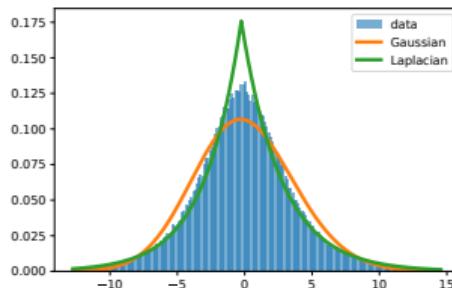
PCA
at $4\lambda/D$



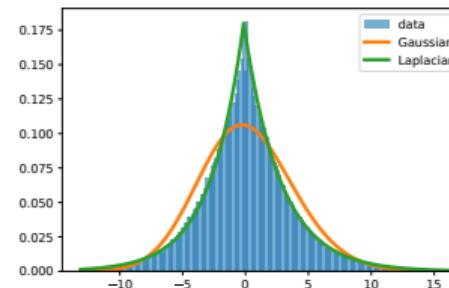
L1-LRA
at $4\lambda/D$



PCA
at $10\lambda/D$



L1-LRA
at $10\lambda/D$



Coefficient of determination

$$\rho^2 = \frac{(\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}))^2}{(\sum_{i=1}^n (x_i - \bar{x})^2)(\sum_{i=1}^n (y_i - \bar{y})^2)}$$

- ▶ x_i : the height of the bins of the data histogram
- ▶ y_i : the values of the probability density function within these bins
- ▶ \bar{x} and \bar{y} : mean of x_i and y_i , respectively

Table: The coefficient of determination ρ^2 for large separation $10\lambda/D$

Rank	PCA		L1-LRA	
	Gaussian	Laplacian	Gaussian	Laplacian
5	0.9866	0.9920	0.9894	0.9940
10	0.9859	0.9936	0.9872	0.9948
15	0.9912	0.9954	0.9922	0.9960

Likelihood ratio map

L1 Norm⁶

$$\log \Lambda_g(R) = - \sum_{(t,r) \in \Omega_g} \frac{|R(t,r) - \hat{a}_g P_g(t,r)| - |R(t,r)|}{\sigma_{R(r)}} \quad (1)$$

L2 Norm

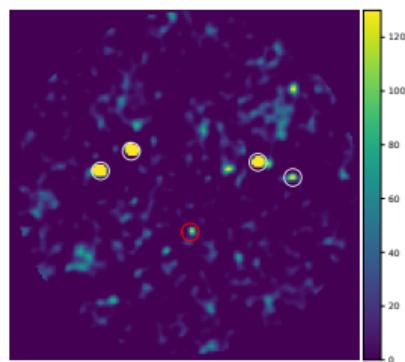
$$\log \Lambda_g(R) = - \frac{1}{2} \sum_{(t,r) \in \Omega_g} \frac{|R(t,r) - \hat{a}_g P_g(t,r)|^2 - |R(t,r)|^2}{\sigma_{R(r)}^2} \quad (2)$$

Background subtraction	Planet detection	
	L1 (2)	L2 (1)
L1-LRA	L1L1	L1L2
PCA	L2L1	L2L2

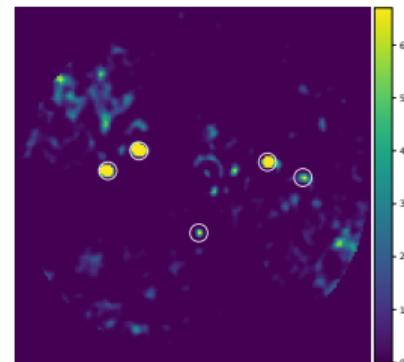
⁶Daglayan, et al., 2022

Likelihood ratio map results

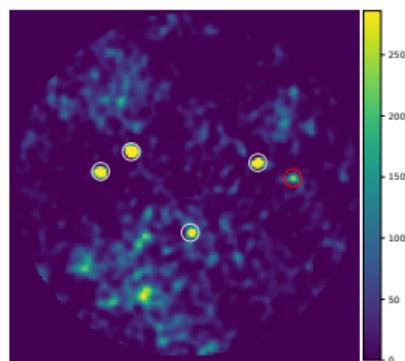
L2L2



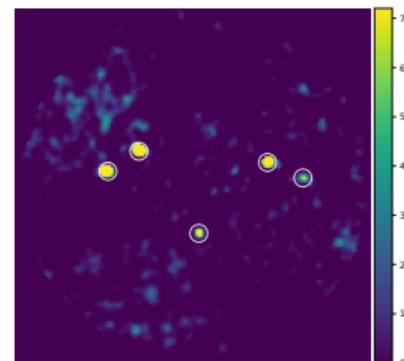
L2L1



L1L2

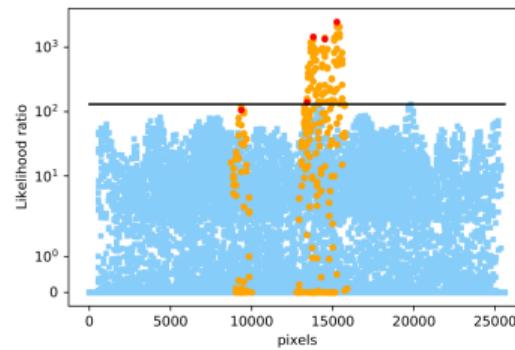


L1L1

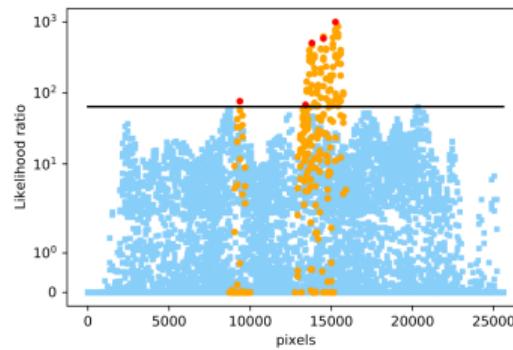


Likelihood ratio map results

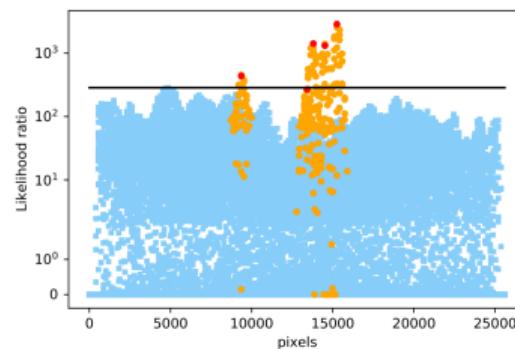
L2L2



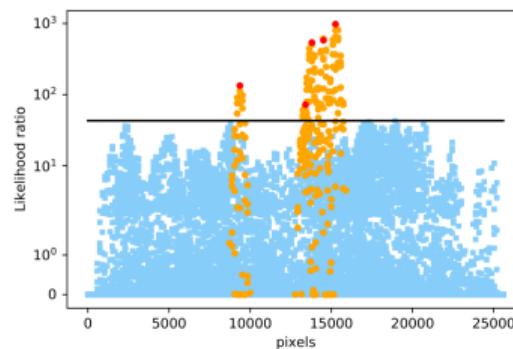
L2L1



L1L2



L1L1



ROC curve results

- ▶ $\sqrt{\text{TPR}}$ & $\sqrt{\text{FPR}}$ are used instead of TPR & FPR.
- ▶ 50 different datasets are used by injecting two synthetic planets in each, 180 degrees apart.
- ▶ Planets are placed at the separation $10\lambda/D$.

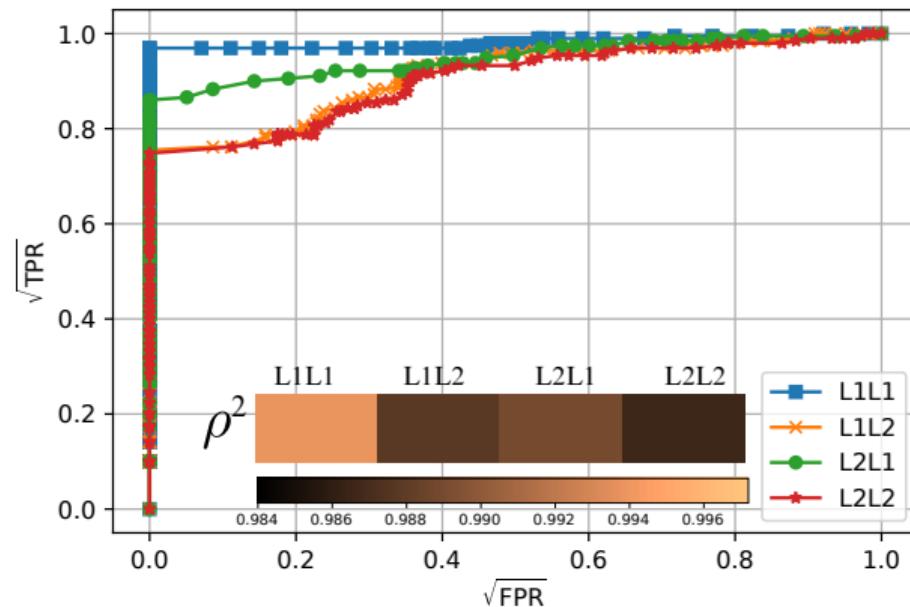


Figure: ROC Curves - Likelihood ratio map

ROC curve results

- ▶ $\sqrt{\text{TPR}}$ & $\sqrt{\text{FPR}}$ are used instead of TPR & FPR.
- ▶ 50 different datasets are used by injecting two synthetic planets in each, 180 degrees apart.
- ▶ Planets are placed at the separation $4\lambda/D$.

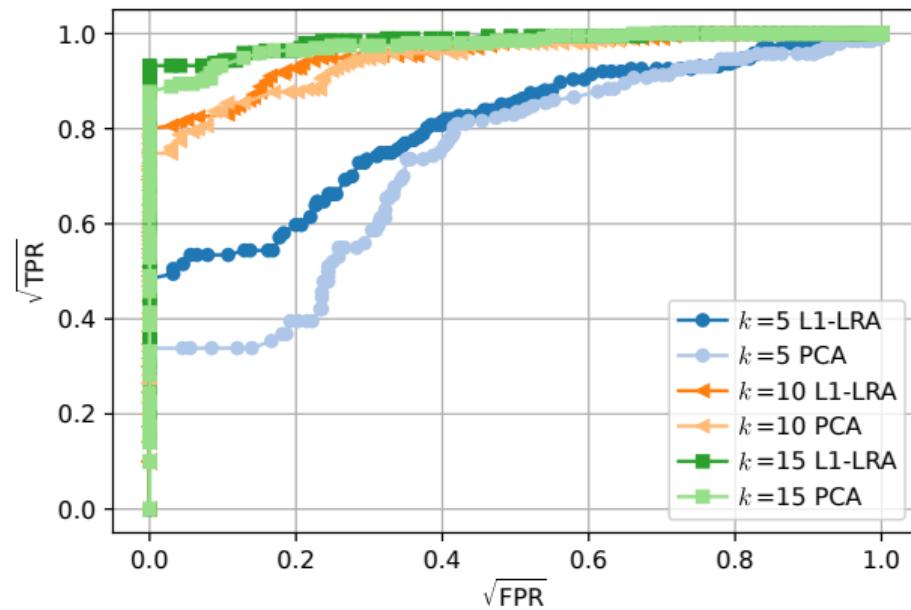


Figure: ROC Curves - SNR map

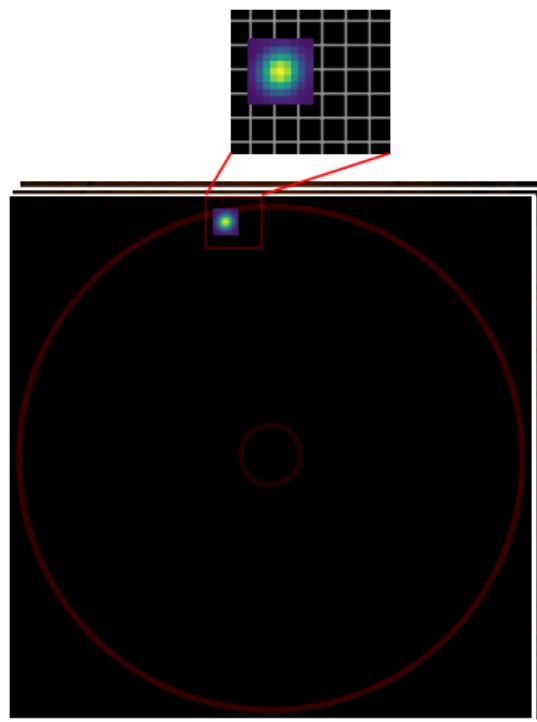
Thank you for your attention!
Any questions?

hazan.daglayan@uclouvain.be

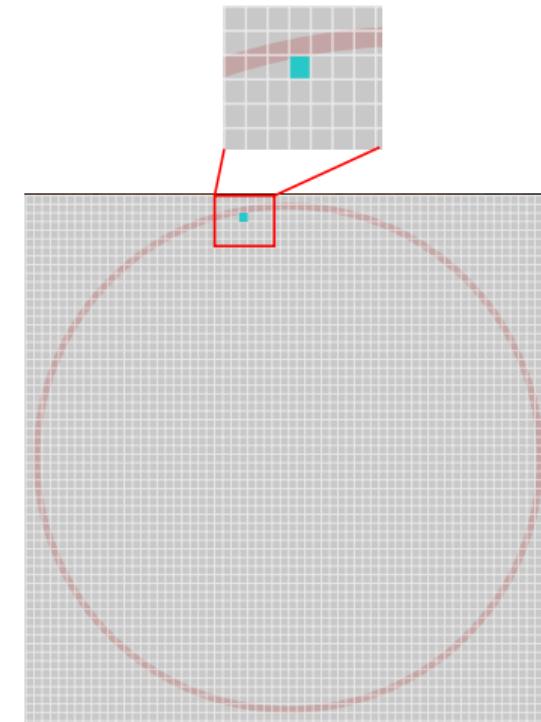
GitHub: [hazandaglayan/l1lra_for_exoplanets](https://github.com/hazandaglayan/l1lra_for_exoplanets)



Trajectories

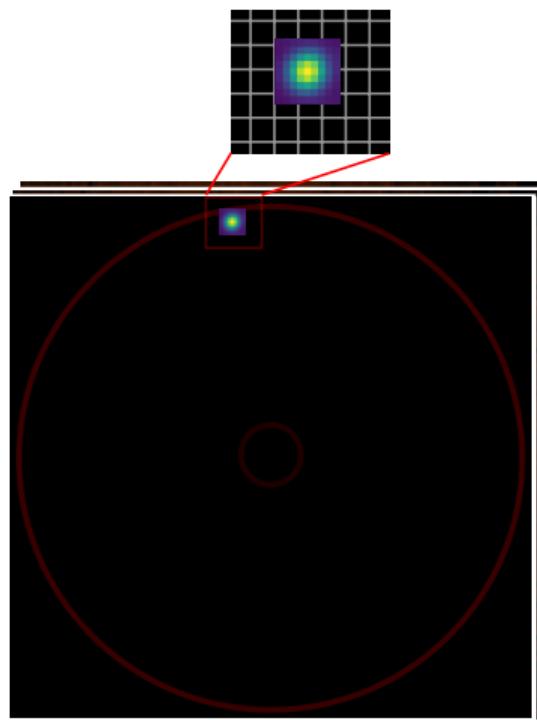


P_g

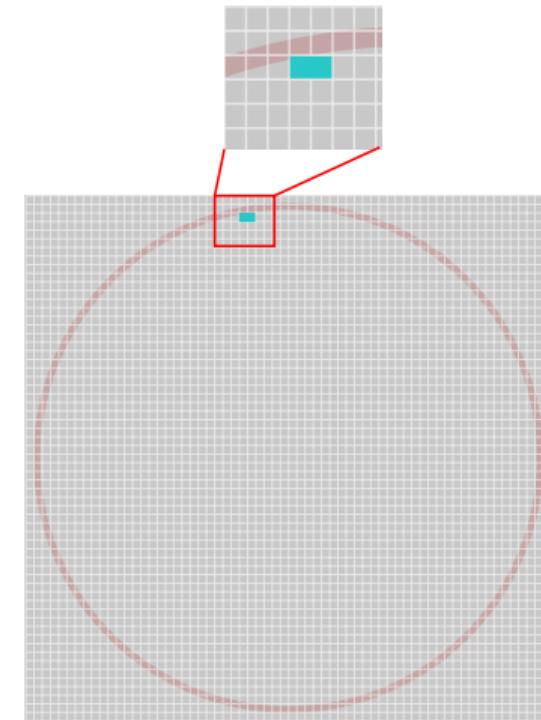


Flux map

Trajectories

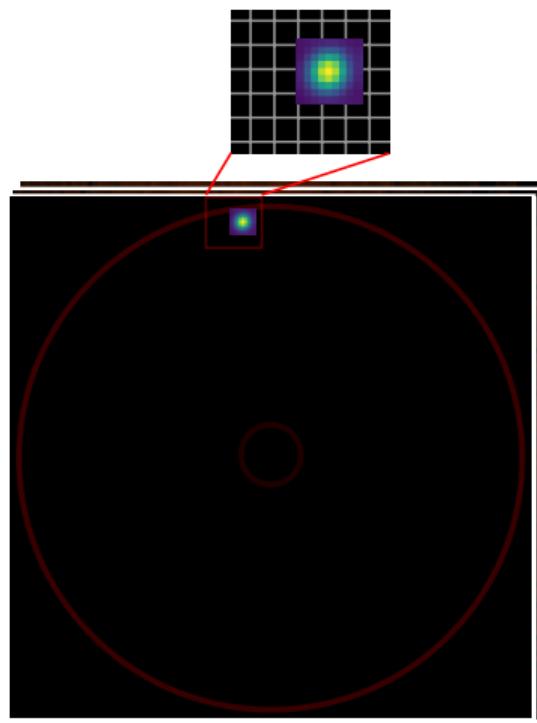


P_g

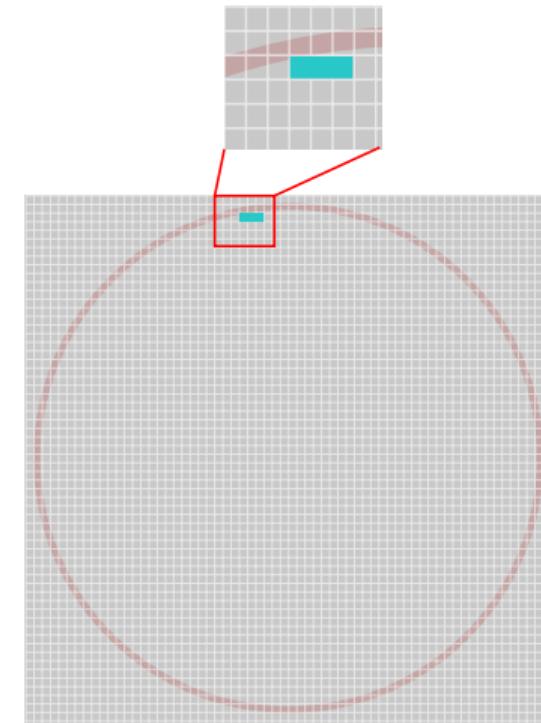


Flux map

Trajectories

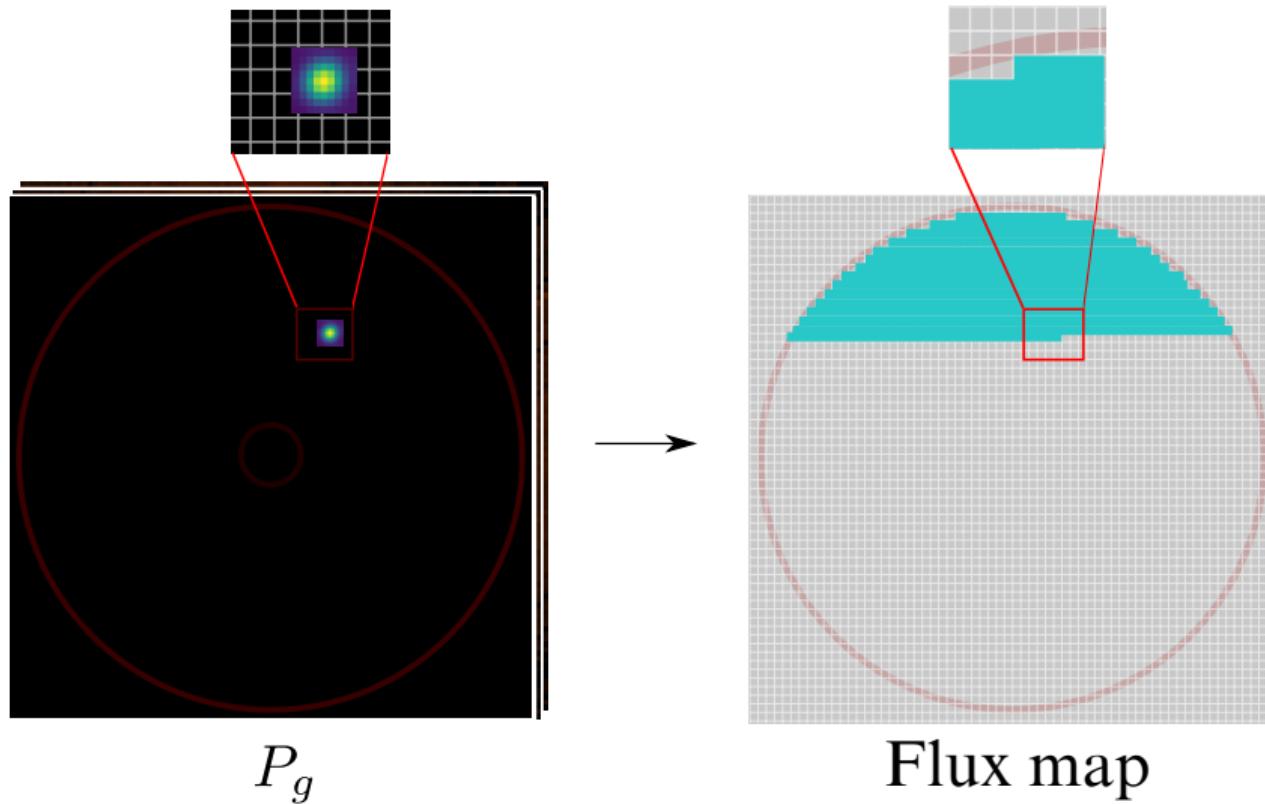


P_g



Flux map

Trajectories



More ROC curves

