

# Potential of RNNs for modelling Transit Light Curves

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- Jan-Peter Muller  
- Angelos Tsiaras

More collaborators: Nikolaos Nikolaou, Billy Edwards, Gordon Yip, Quentin Changeat, Giovanna Tinetti





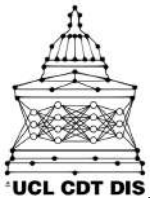
# Introduction

Acronyms:

LC = Light Curve

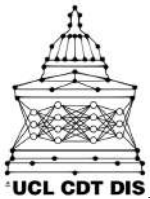
RNN = Recurrent Neural Network

LSTM = Long Short-Term Memory (network)



# Exoplanets Boom

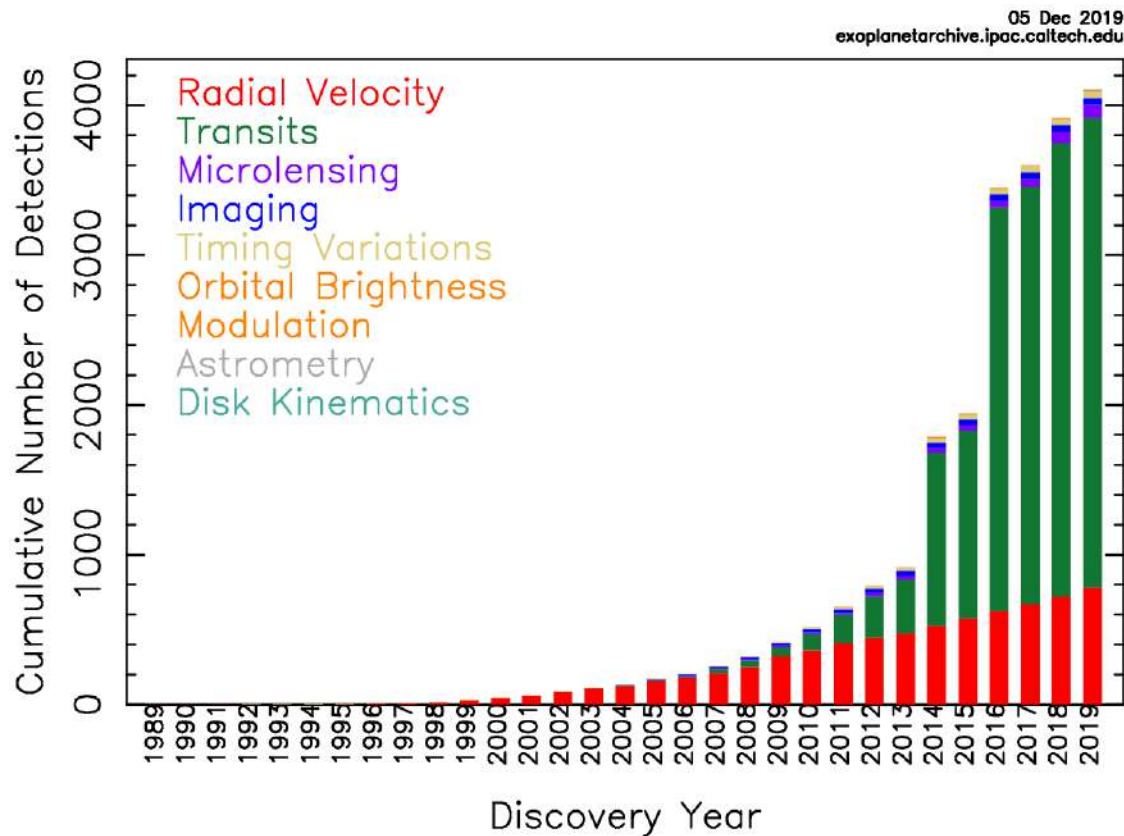
- In our galaxy:  $N_{\text{exoplanets}} \approx N_{\text{stars}}$  [Cassan et al. 2011]

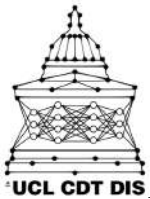


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- Exponential discovery rate?

Cumulative Detections Per Year

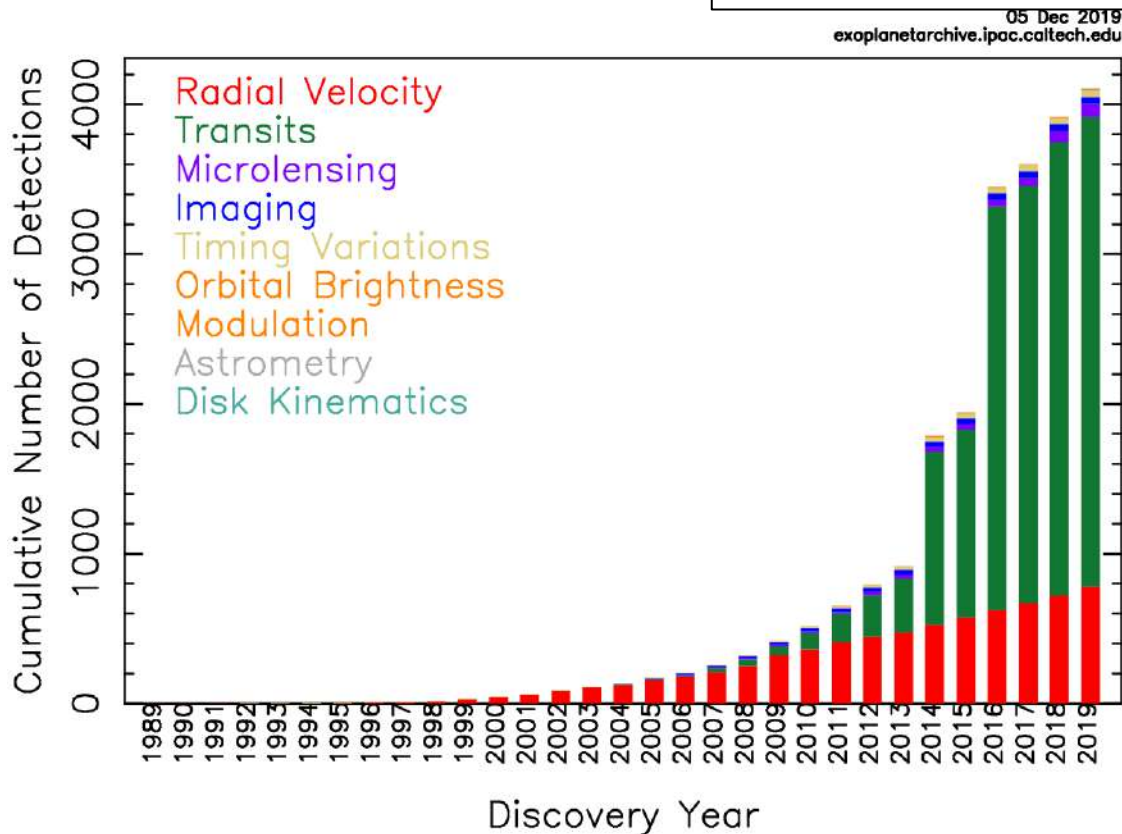


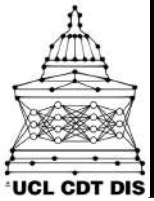


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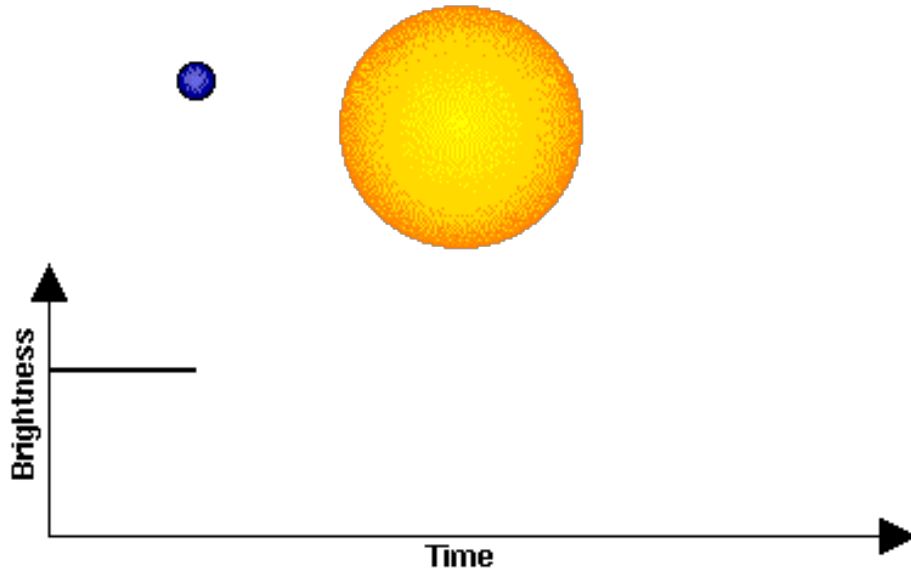
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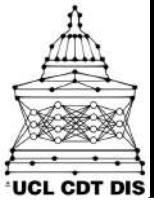
- Exponential discovery rate?  $N(t) = N_0 \cdot 2^{t/4} \rightarrow$  **100 years** to discover all Milky way planets



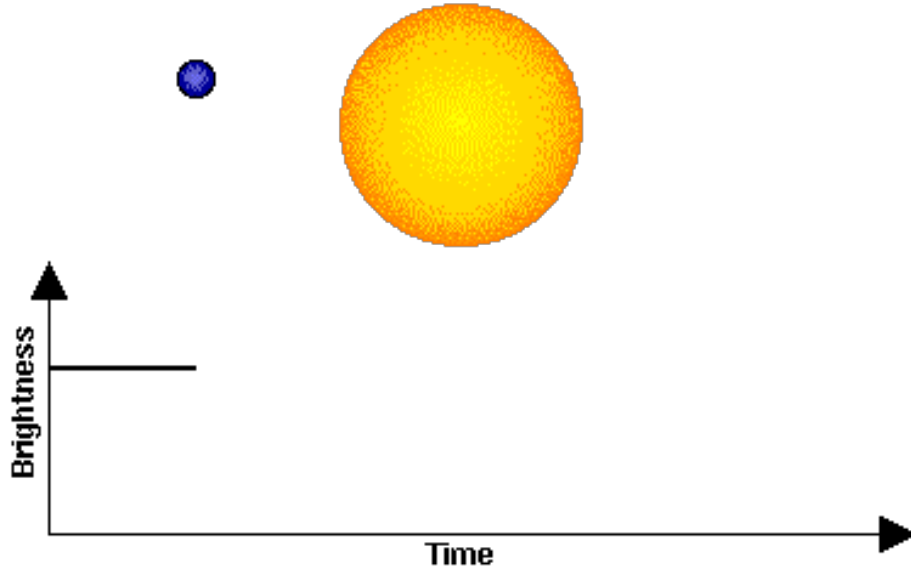


# Transit method

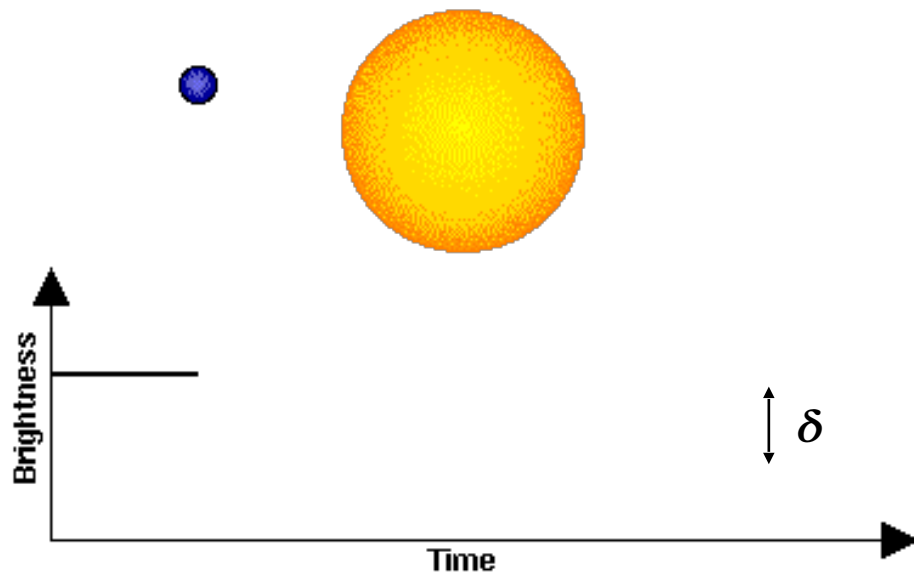




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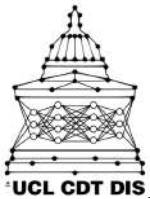
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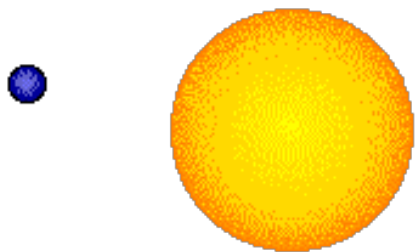
$$F \simeq (1 - \delta) F_{star}$$

$$\delta(\lambda) \simeq \left( \frac{R_P(\lambda)}{R_S} \right)^2$$



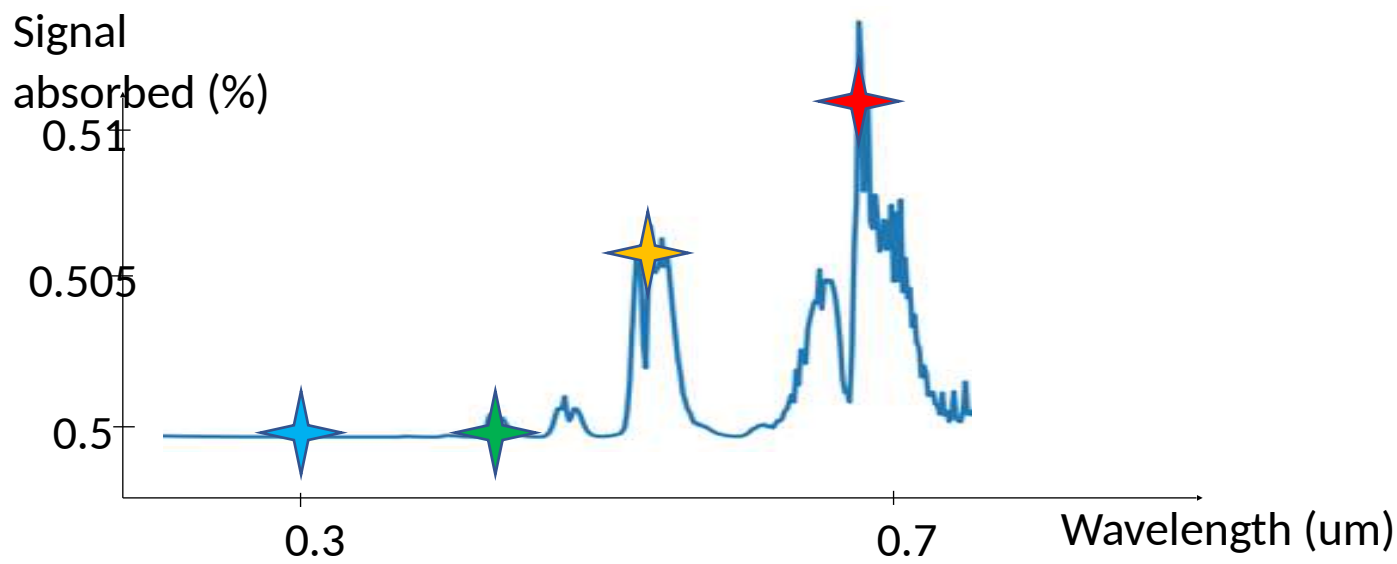
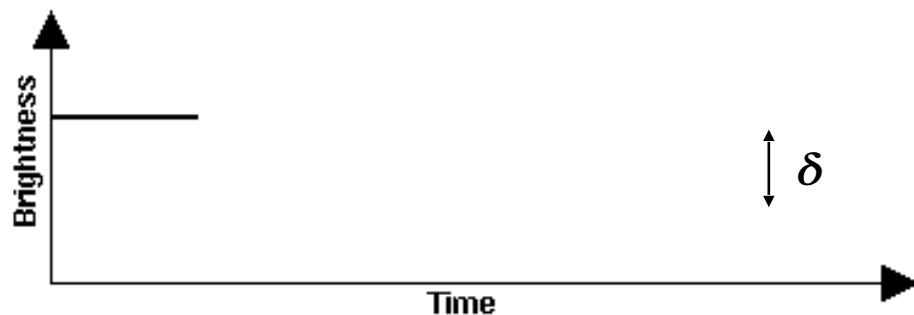


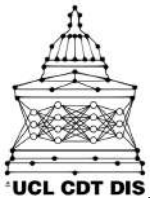
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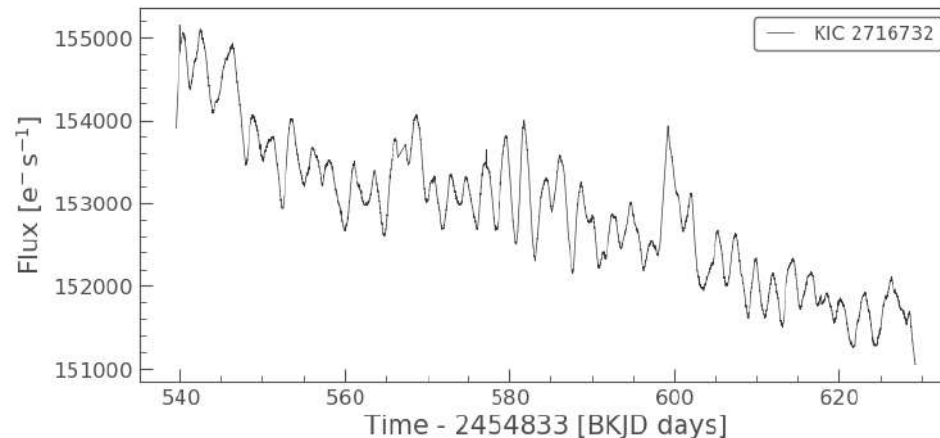
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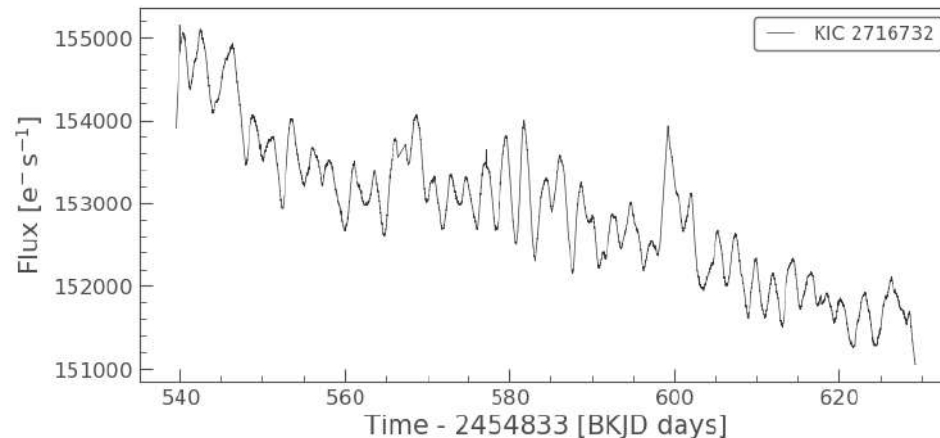
# What's the problem?

- Low SNR: stars are much bigger than planets
- Complex Physics: limb-darkening, phase-curves, multi-planets...
- Variable stars: Spots, flares...
- Instrument systematics: ramps, jitter, bumps...




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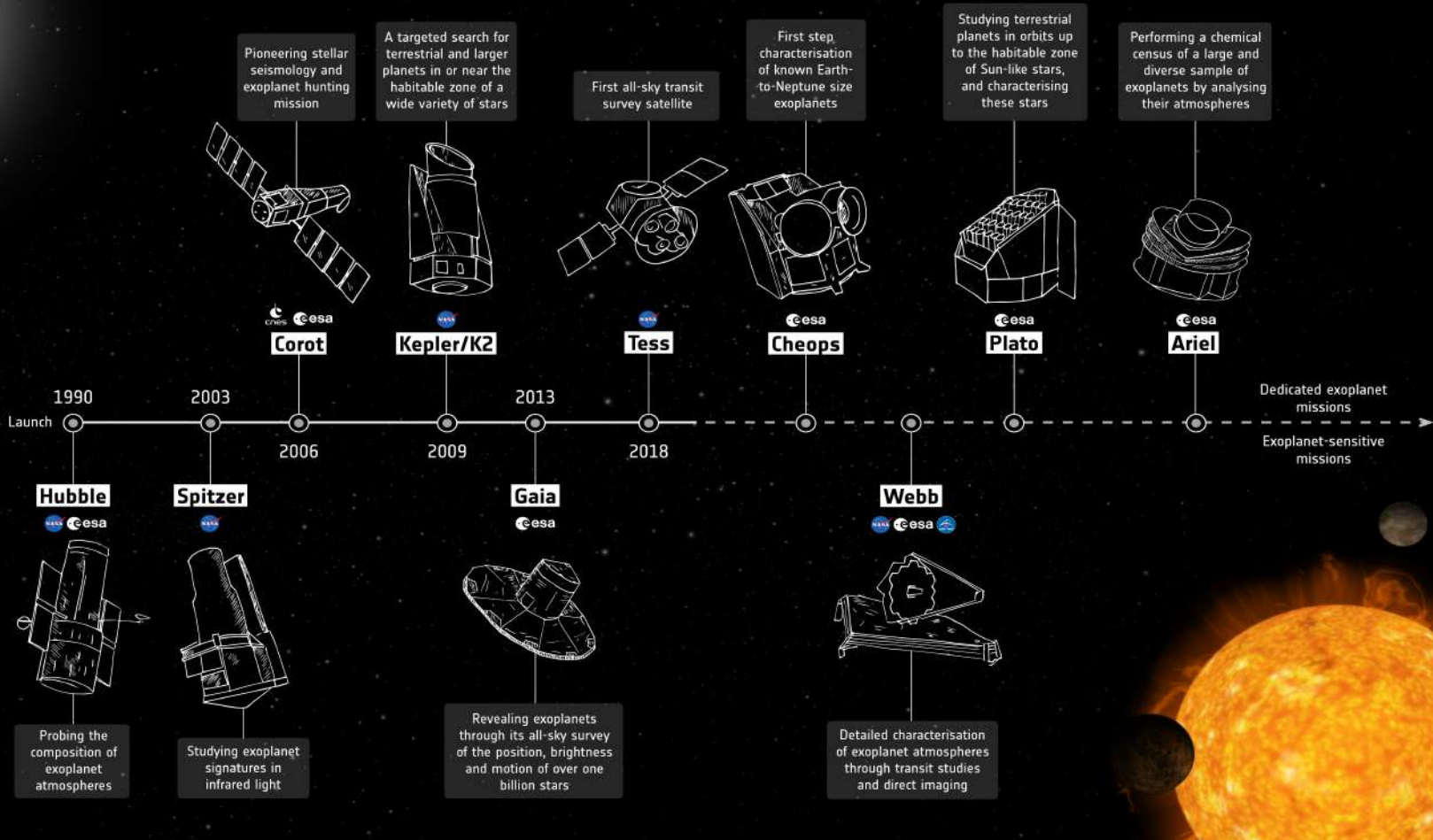
→ Observed Light Curves are complex stochastic Time-Series  
They can't be modelled fully analytically

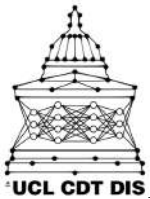
# Transits observation



**Ground-based observatories**

First discoveries of exoplanets in the 1990s opened up the field of exoplanet research. New innovations and discoveries continue to this day





# Data Volumes

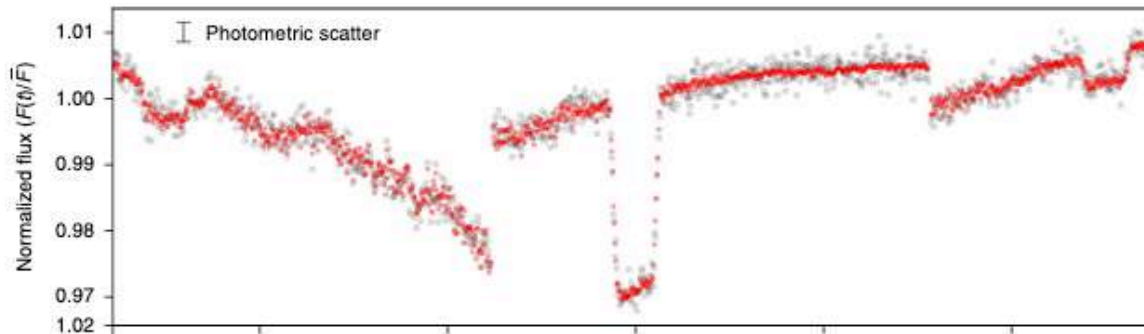
- Large amount of unprocessed data

Space Missions	Data Volume [GB/day]
Kepler	1.10
Spitzer	~1-6
Hubble	3
Tess	27.21
JWST	57.5



# From one transit LC...

- Precision fitting of a physics model
  - 1<sup>ary</sup> Transit, 2<sup>ary</sup> transit, phase curves, timing...

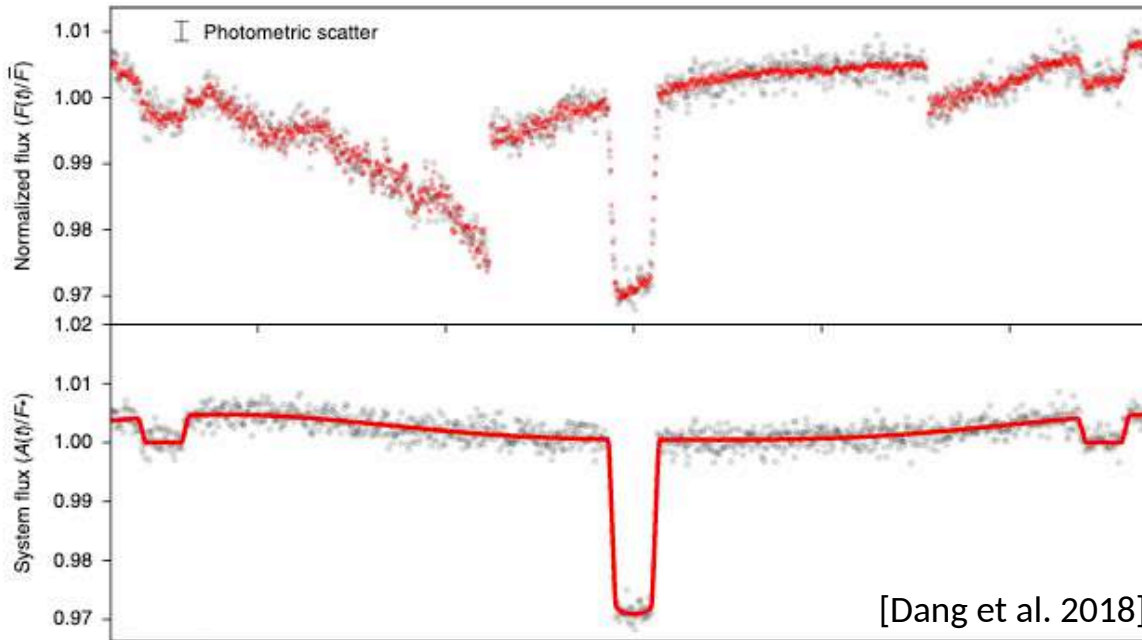


**Not detrended**

[Dang et al. 2018]

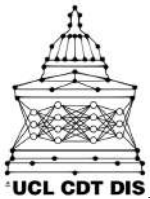
# From one transit LC...

- Precision fitting of a physics model
  - 1<sup>ary</sup> Transit, 2<sup>ary</sup> transit, phase curves, timing...
- Detrending is necessary



Not detrended

- Detrended
- Physical model fitted

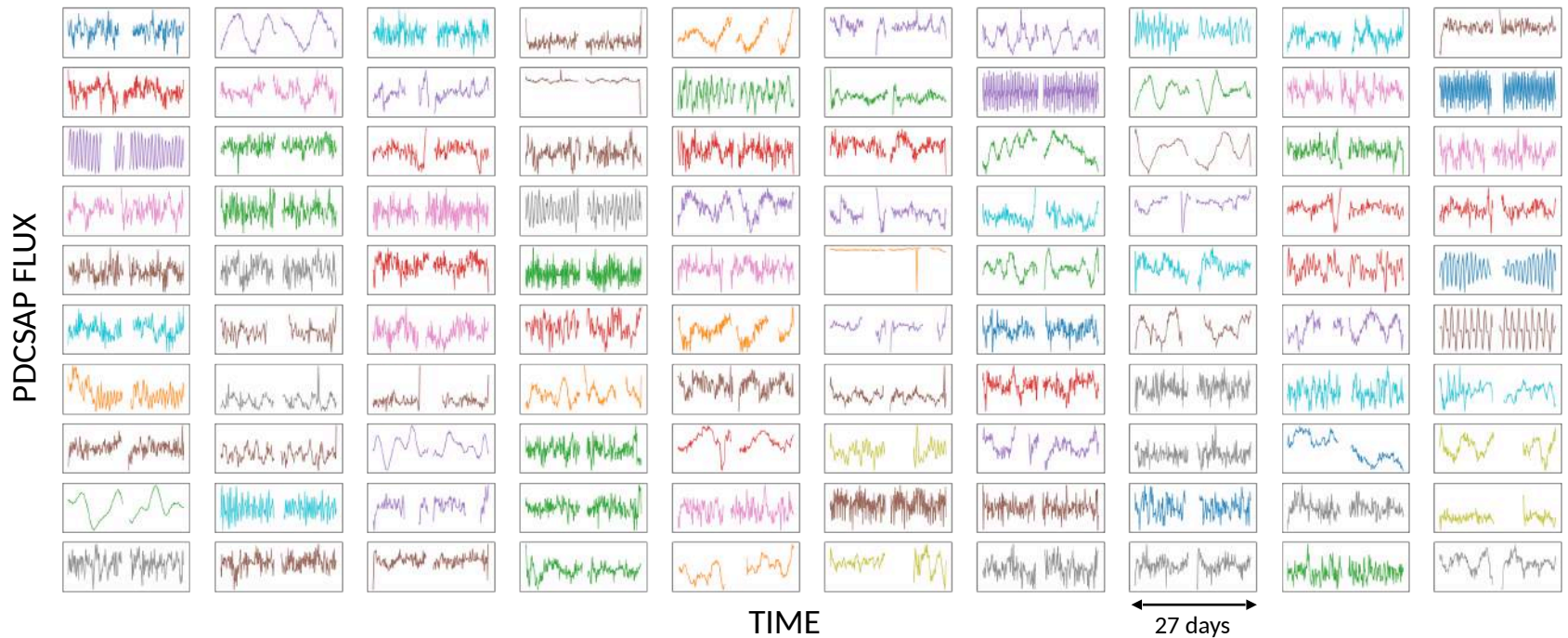


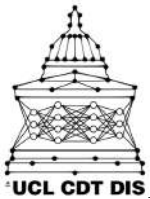
# ...to many many LCs!

- Stellar LCs:

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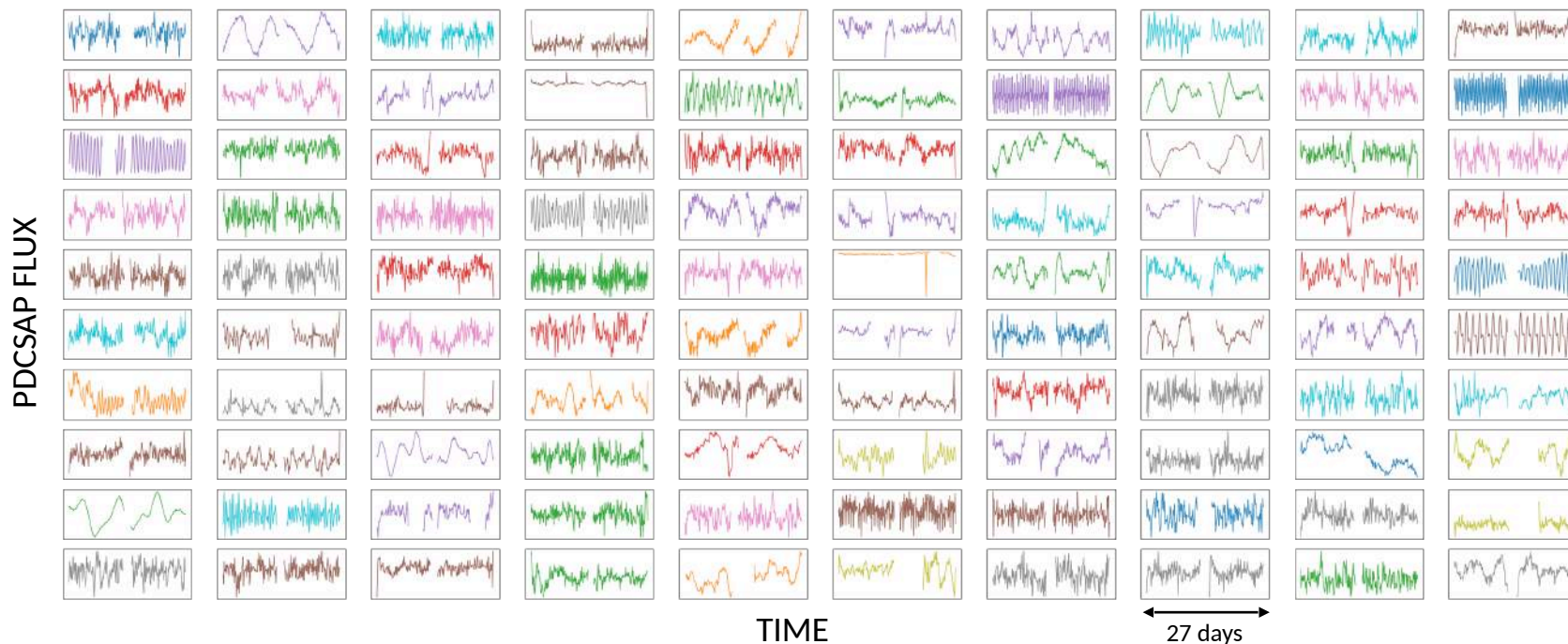
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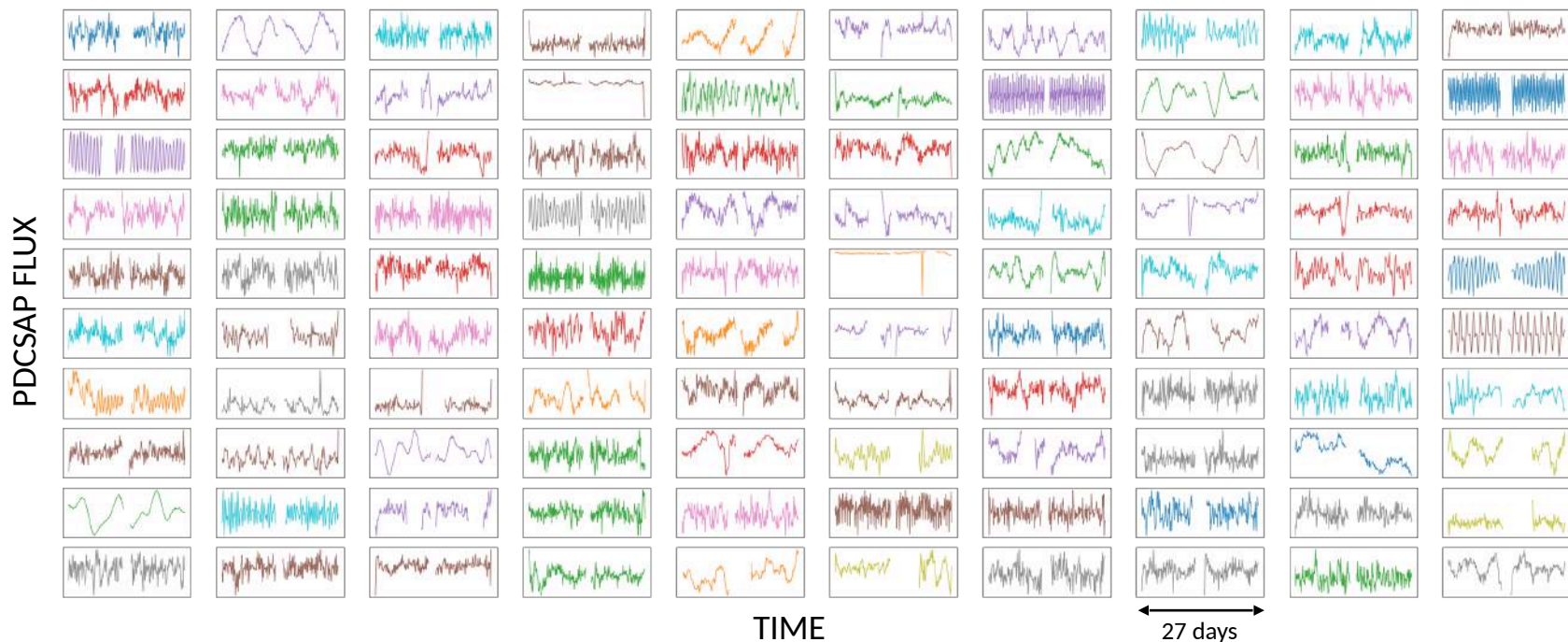
- Stellar LCs:
  - Clustering





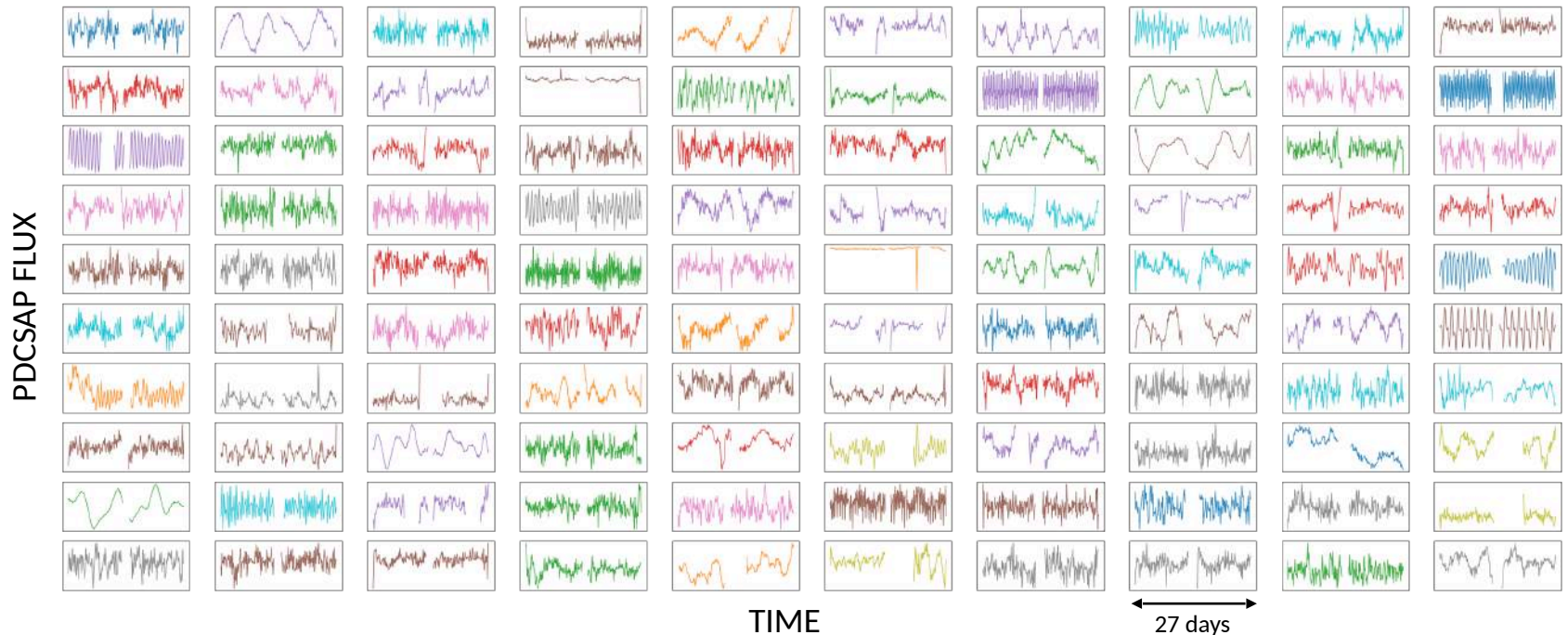
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- Stellar LCs:
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  - Regression of Stellar parameters



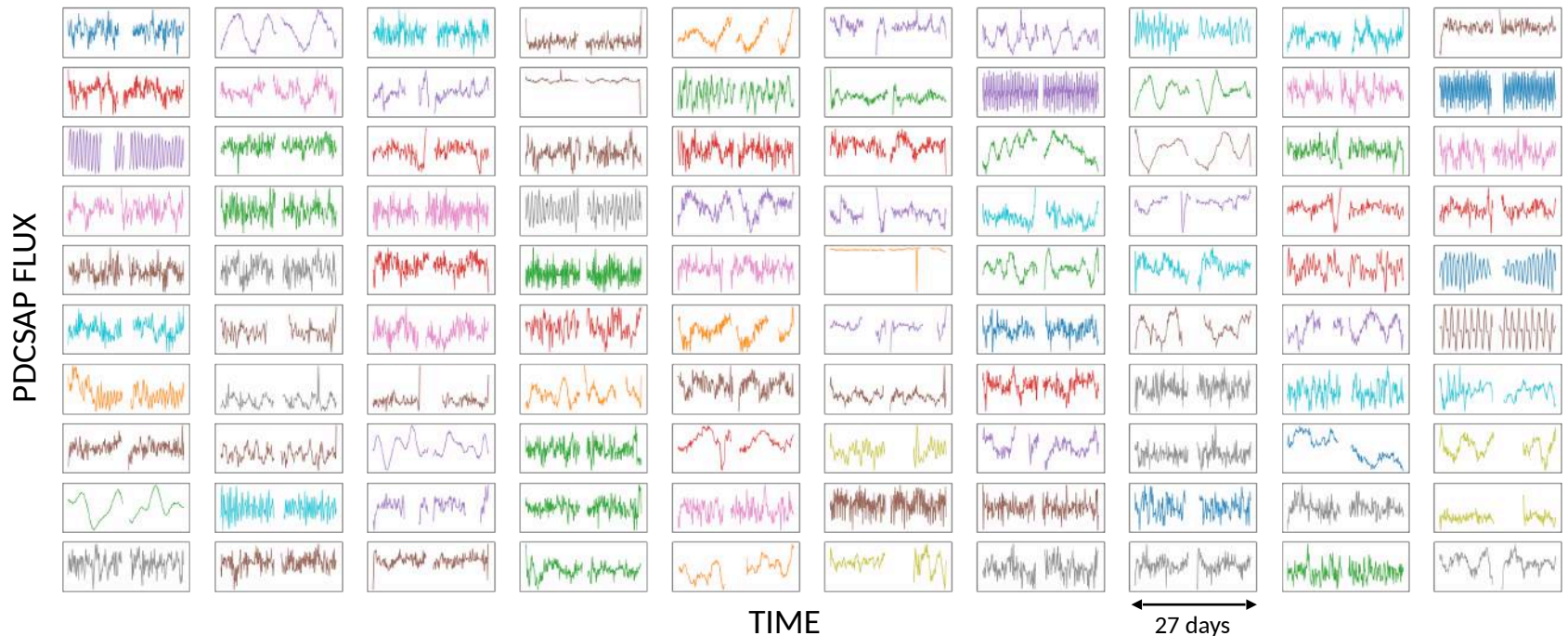
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  - Anomaly detection





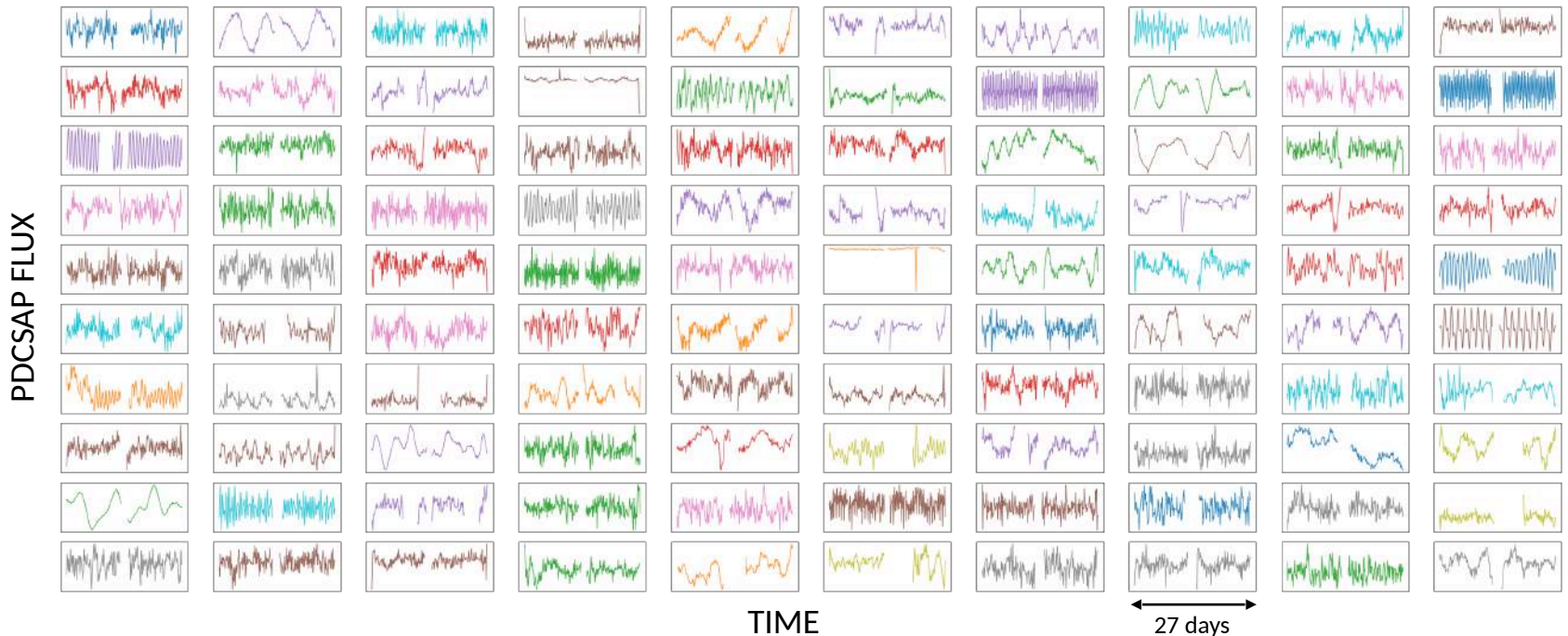
# ...to many many LCs!

- **Stellar LCs:**

- *Clustering*
- *Regression of Stellar parameters*
- *Forecasting – Interpolating*
- *Anomaly detection*

- **Transits LCs (*labels avail.*)**

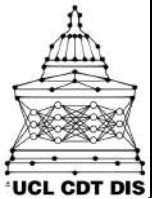
- *Detection*
- *Candidate vetting*





# Detrending (a few) Spitzer Light Curves

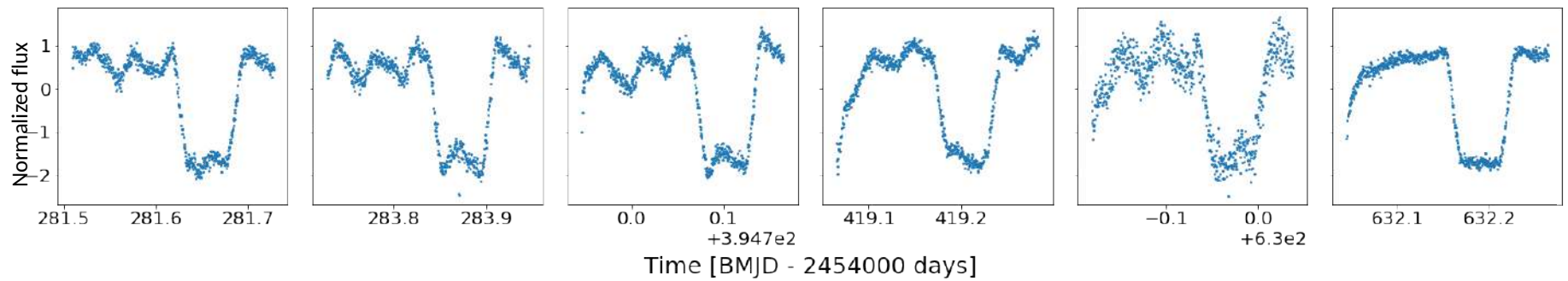


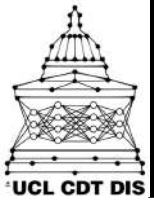


# The problem

Data:

- 6 observations of HD 189733 b
- Detector = IRAC 8  $\mu\text{m}$

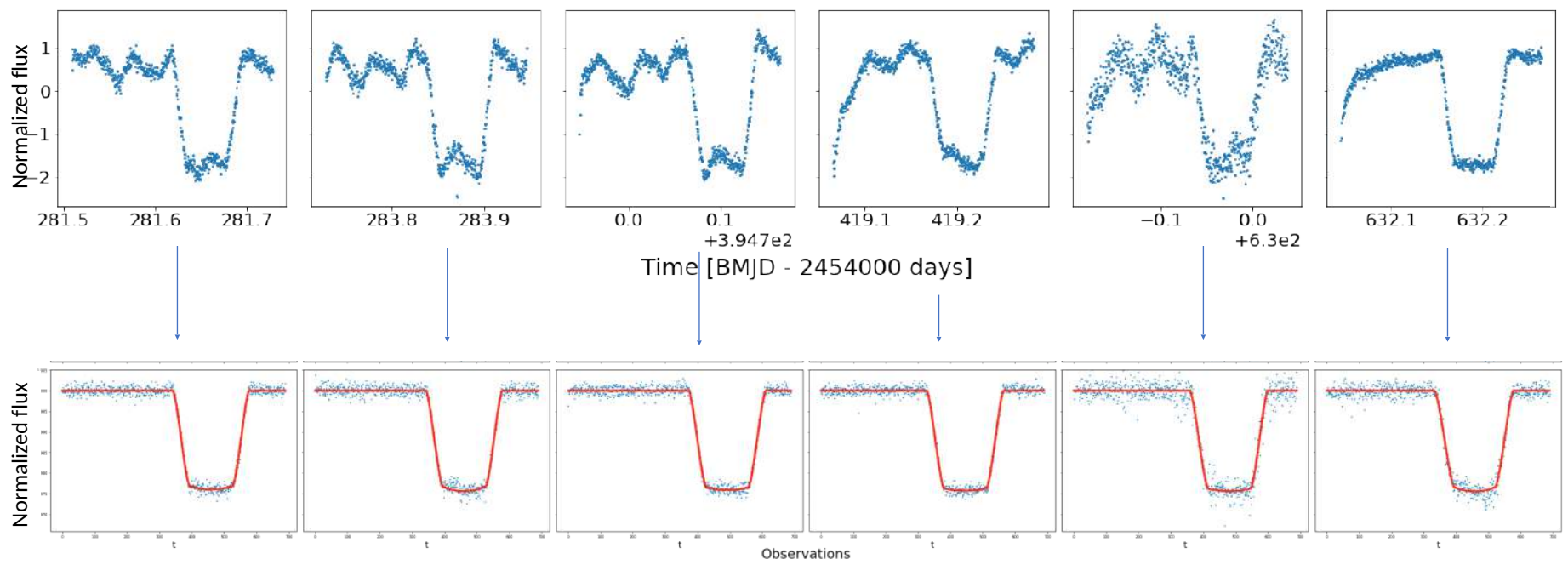




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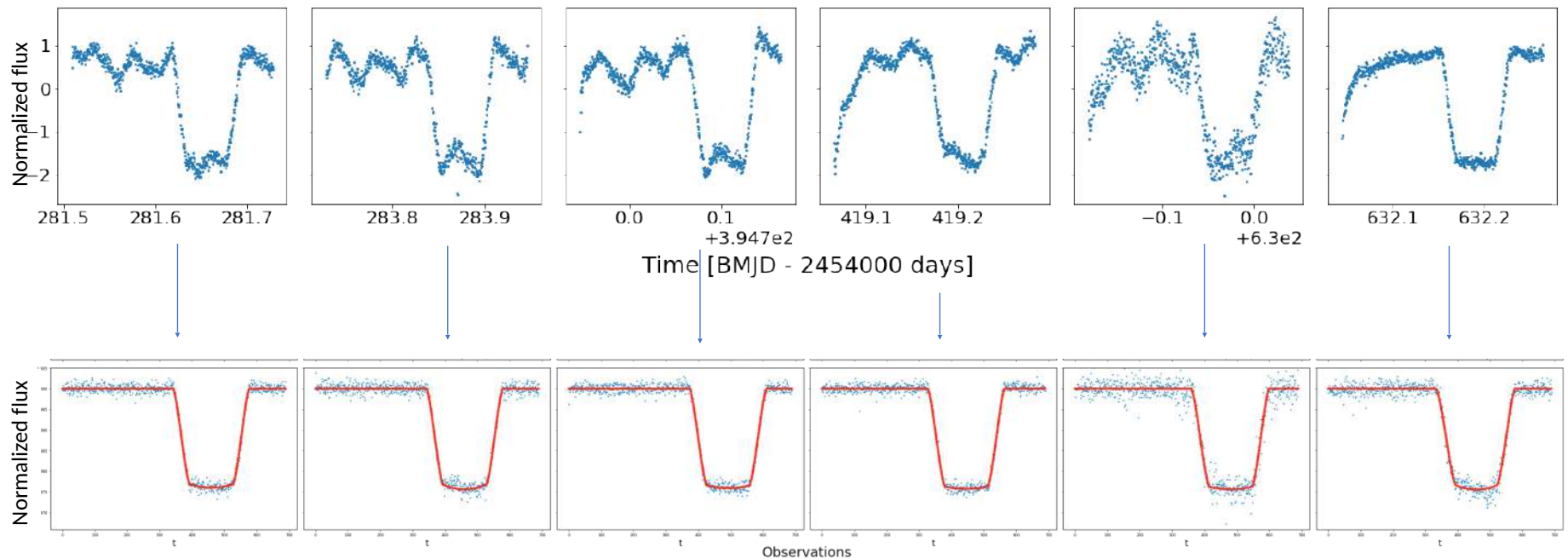


**Physical model = Transit**( $R_p/R_s$ ,  $a/R_s$ ,  $P$ ,  $i$  ...)

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$$\text{Physical model} = \text{Transit}(R_p/R_s, a/R_s, P, i \dots)$$

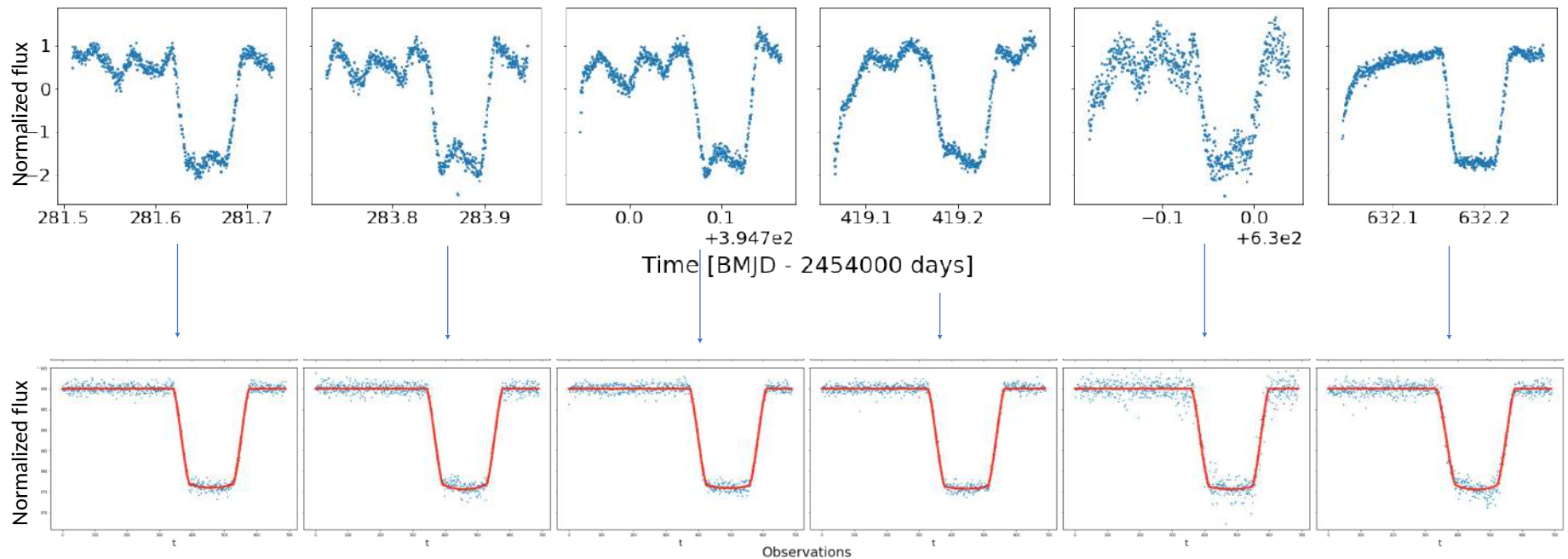
**Regression Problem.**

Often decomposed in: **Detrending** + **Fitting**

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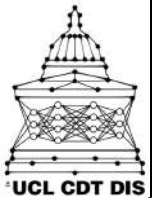
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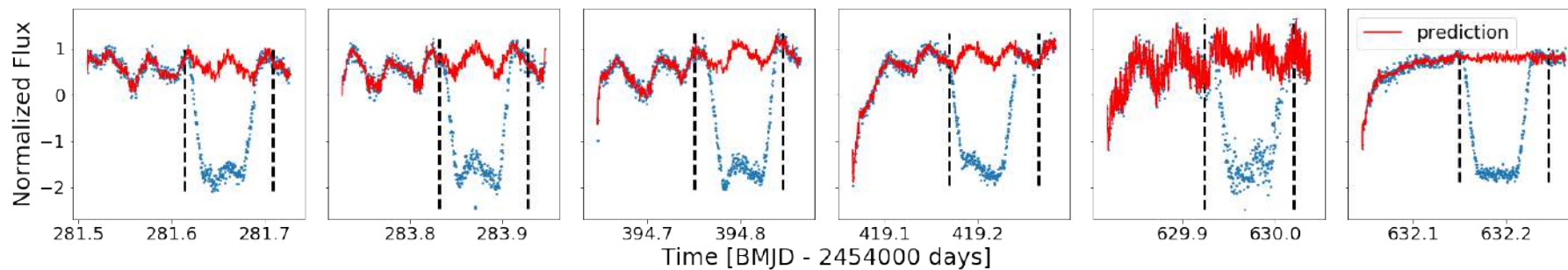
**Existing works:** *parametric, PLDs, ICA, GPs...*

Anything more automatic and generalizable?



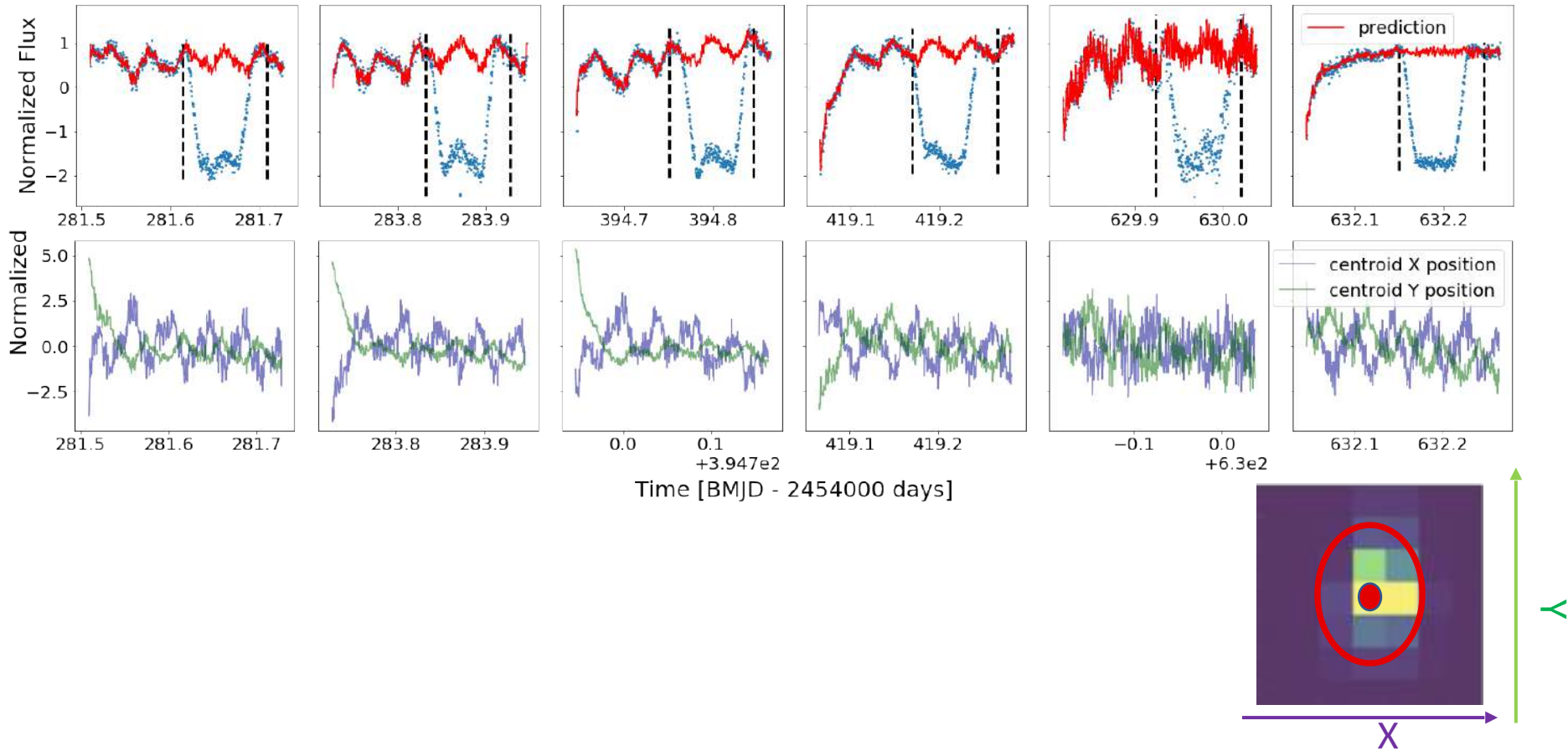
# Our approach: interpolation

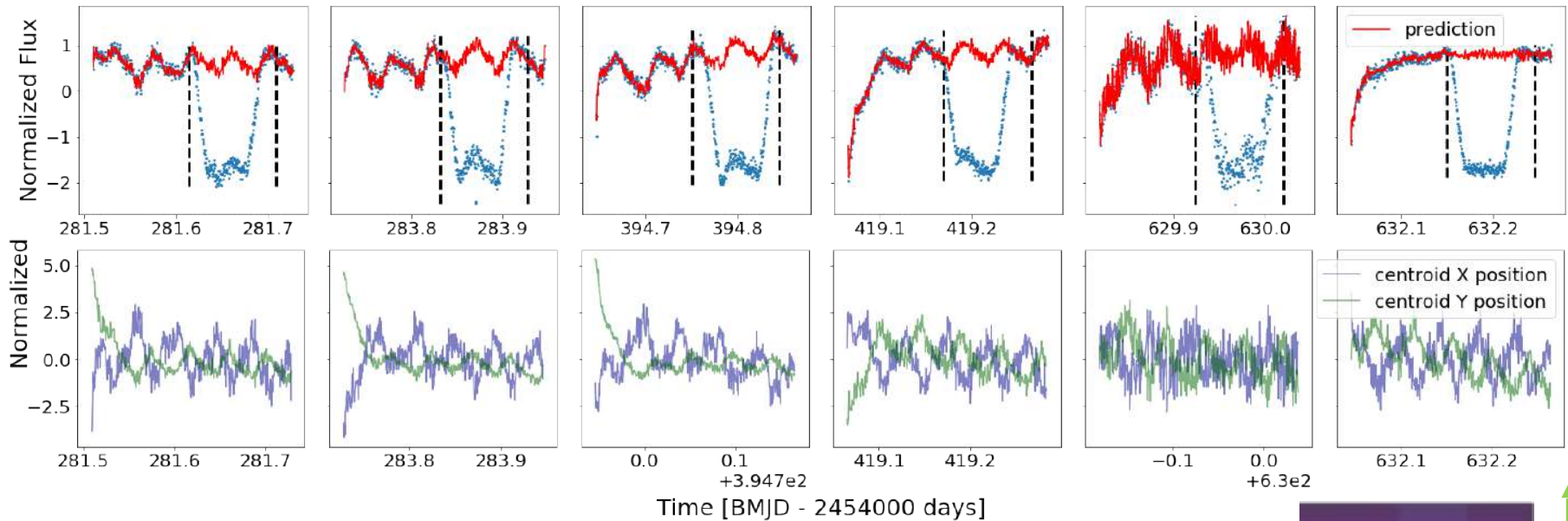
Correcting Spitzer  
Light Curves





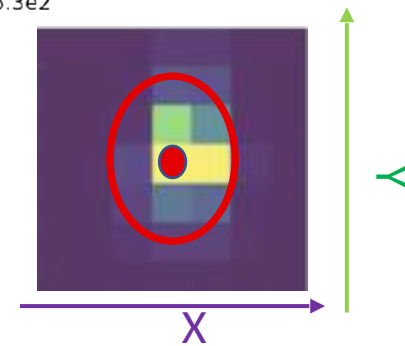
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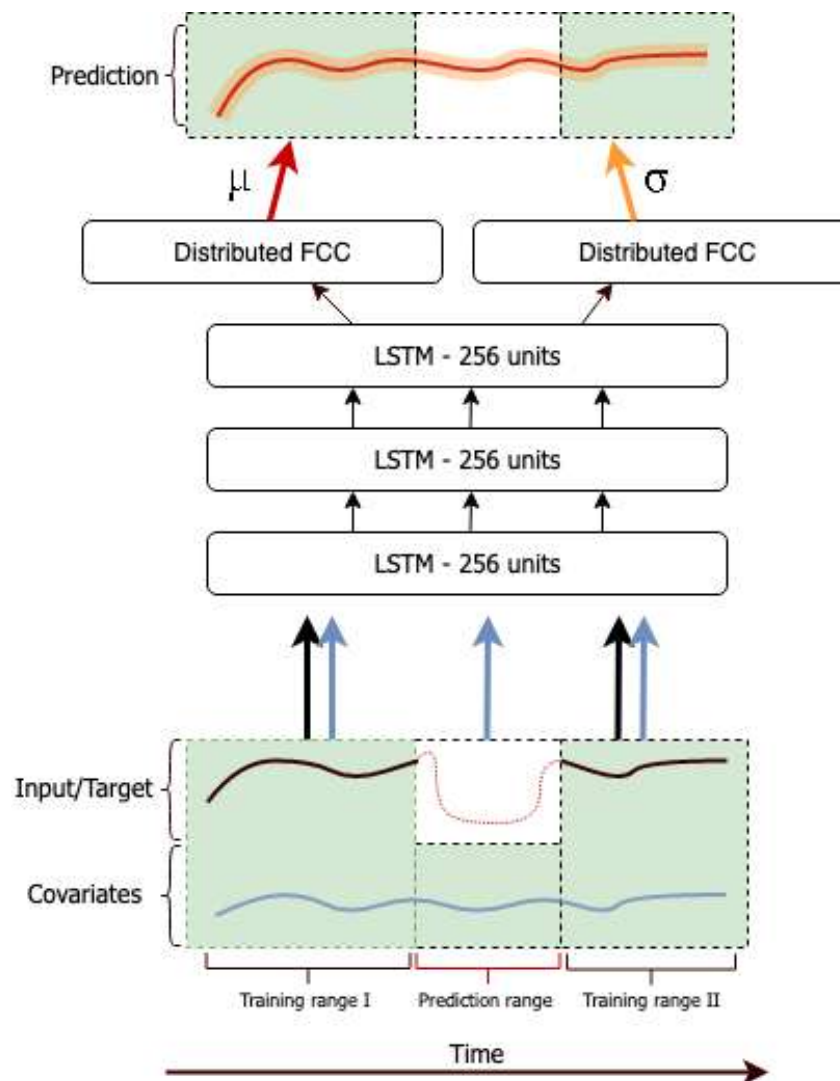


## Idea:

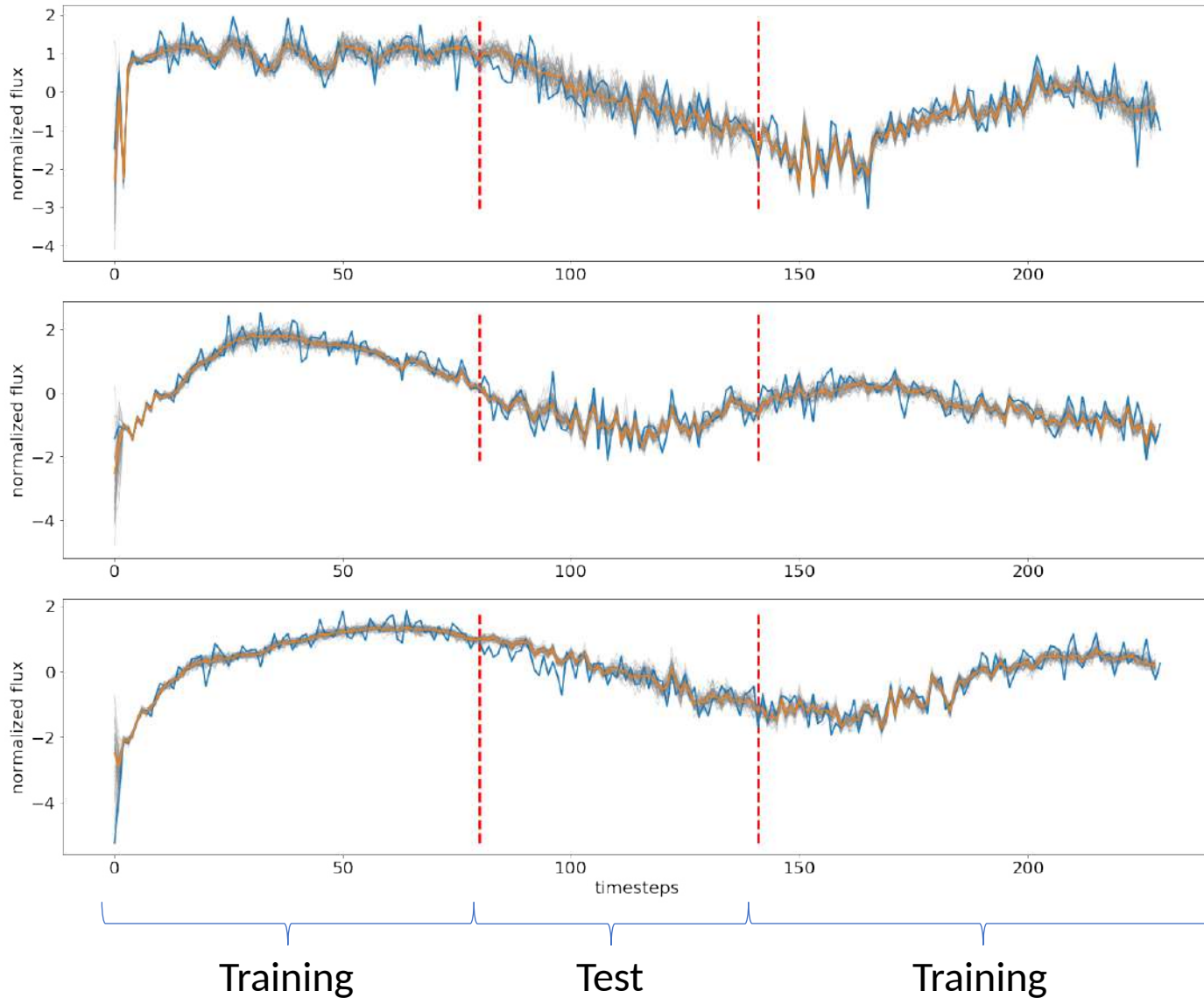
- Use a RNN to learn the temporal dependency in the LC
  - Note: it can also learn correlation with centroids
- Train the RNN at each epoch to:
  - forecast the next step on the out-of-transit
  - Still predict on transit range: input=previous prediction, no loss computed

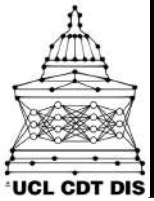


# The architecture

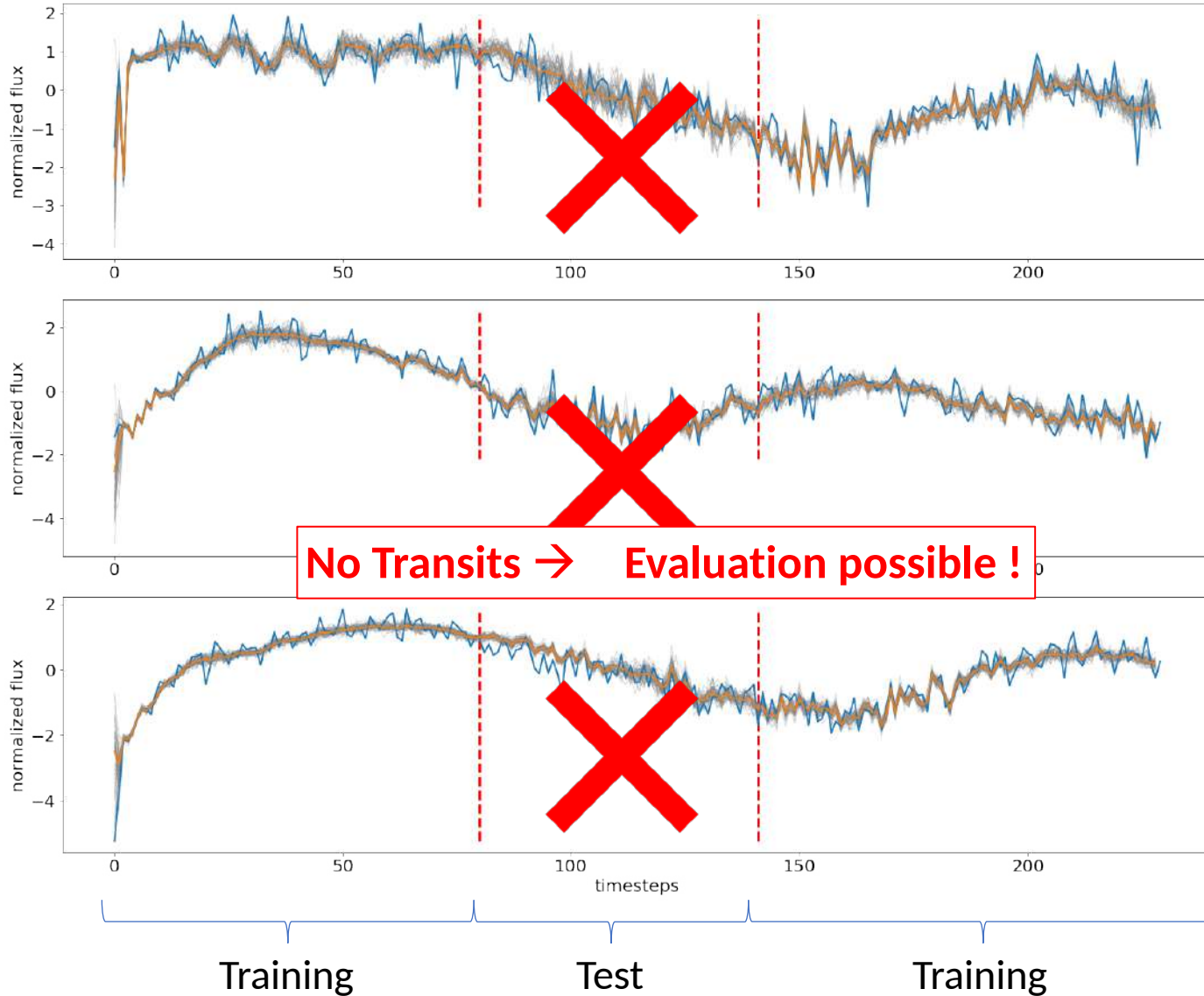


# Model's validation

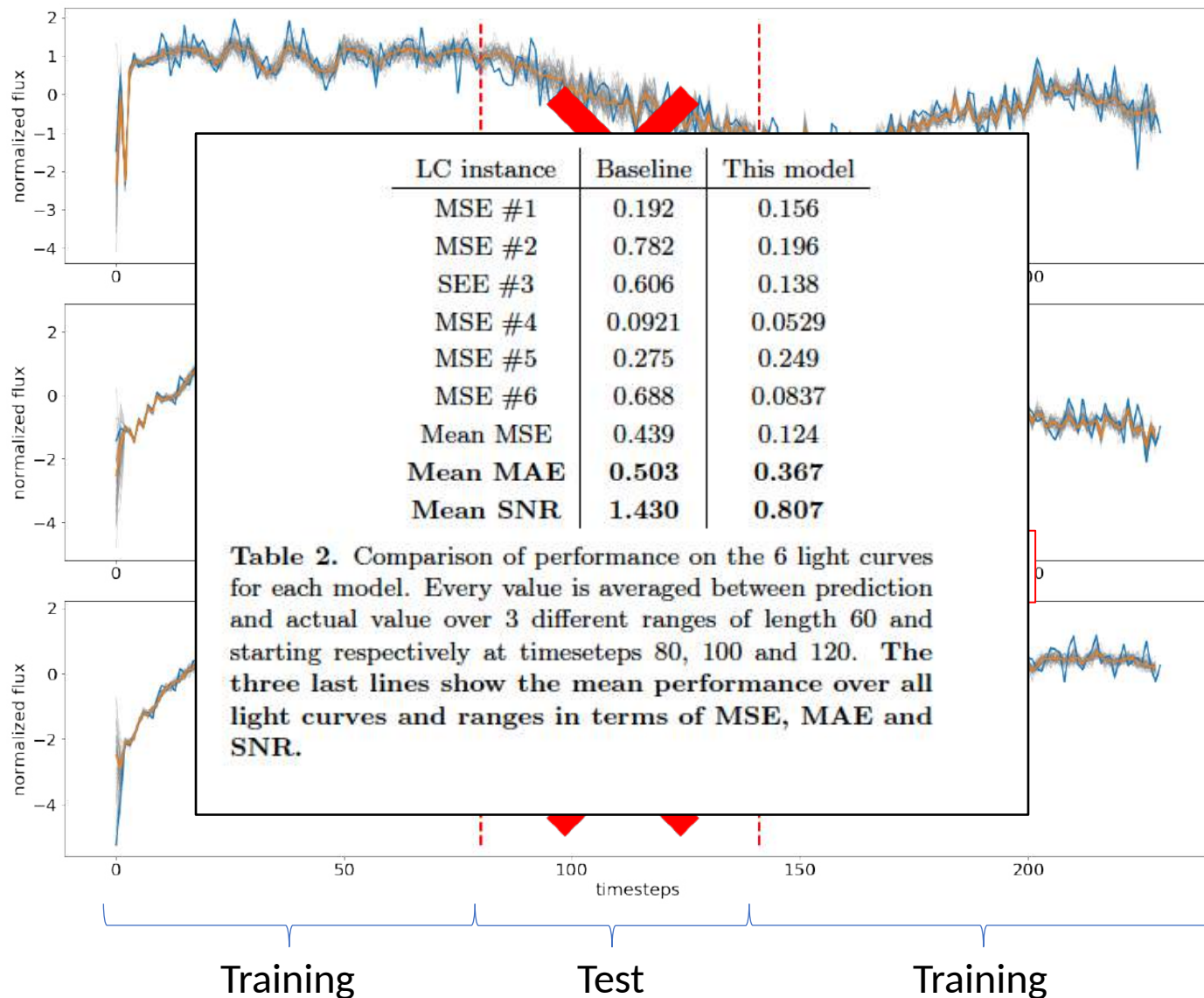


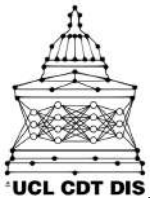


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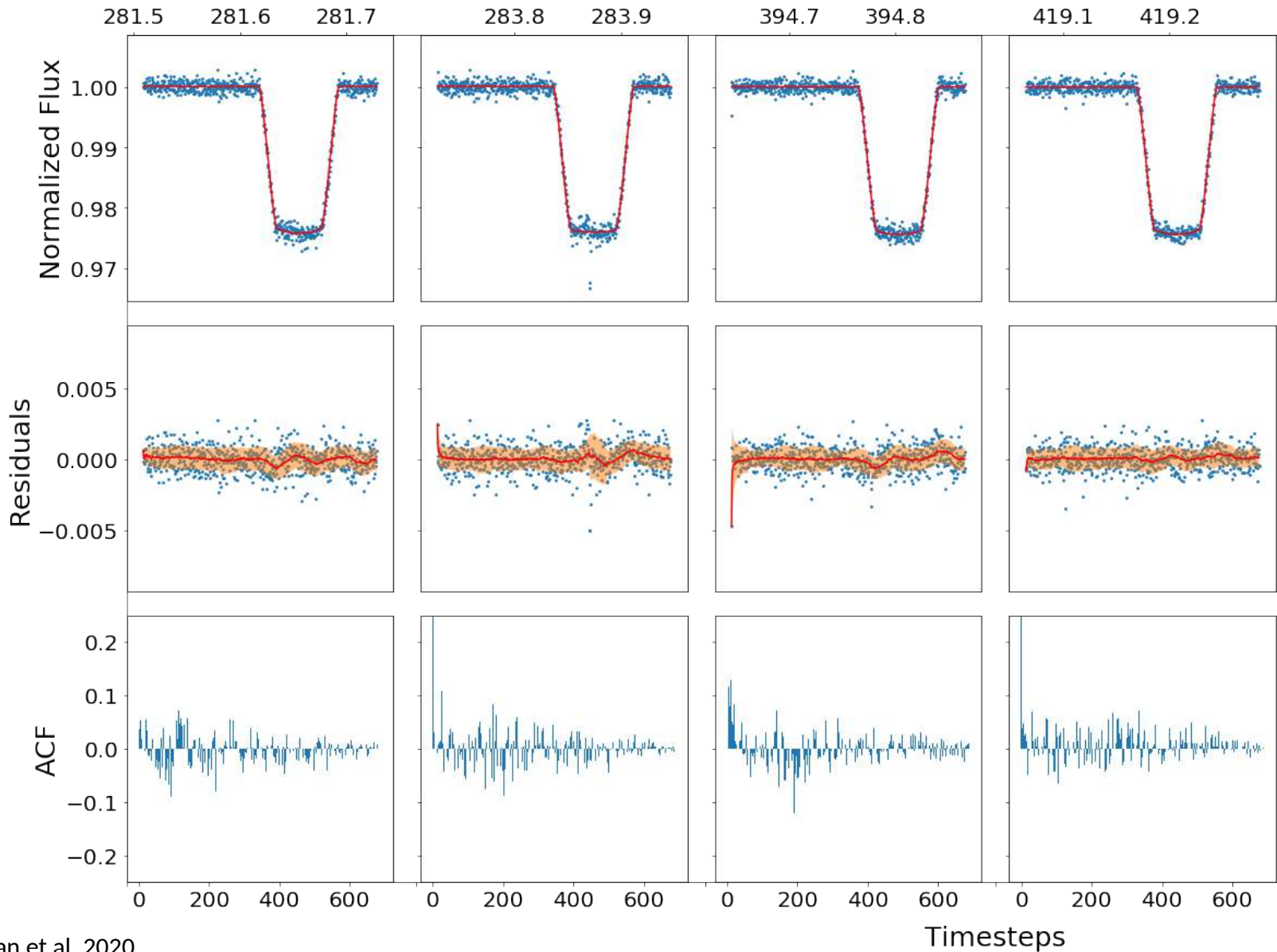


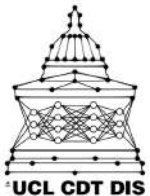


# Prediction + Transit Fit

Correcting Spitzer  
Light Curves

Time [BJD - 2454000 days]

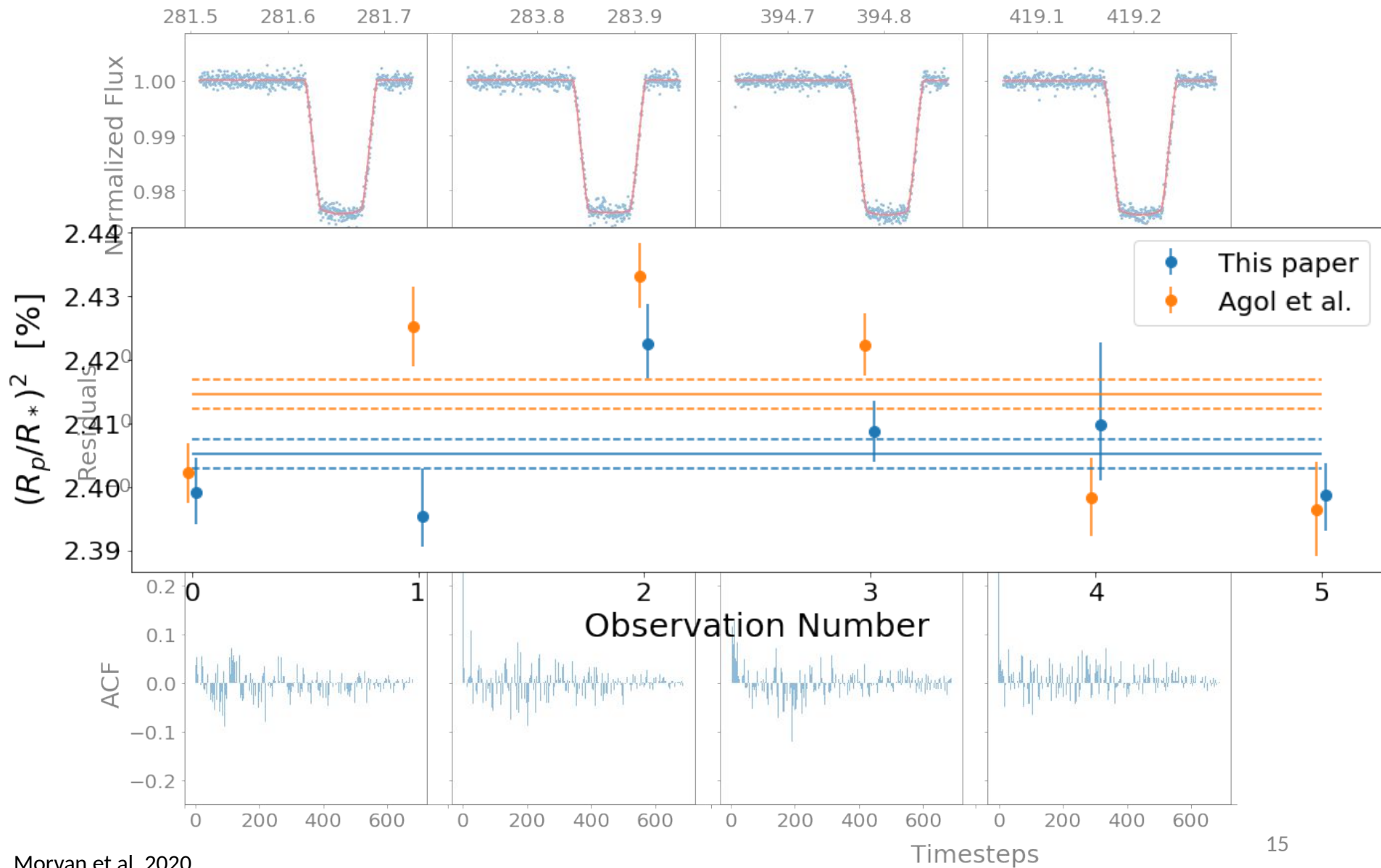




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Time [BJD - 2454000 days]



# Conclusion

- Many exoplanets
- Many many light curves
- Work on Spitzer
  - LSTM models time dependencies in individual LCs
  - Useful for **detrending** LCs.
- Preliminary work on TESS
  - hopes for a **global LC model**
- Towards a global, automatic detrending pipeline?
- Would be great for JWST, Cheops, Plato, Ariel...

## SOFTWARE

DeepARTransit: Morvan, M. 2019. DeepARTransit: A library for interpolating and detrending transit light curves, 1.0.0, Zenodo, doi: 190091225  
PyLightcurve: <https://github.com/ud-exoplanets/pylightcurve/>  
Lightkurve: This research made use of Lightkurve, a Python package for Kepler and TESS data analysis (Lightkurve Collaboration, 2018).

## REFERENCES

Agol E, Cowan NB, Knutson HA, Deming D, Steffen JH, et al. 2010. *The Astrophysical Journal*, 721(2):1861-77

Cassan A, Kubas D, Beaulieu J-P, Dominik M, Horne K, et al. 2012. *Nature*, 481(7380):167-69

Sallinas D, Flunkert V, Gasthaus J. 2017. [arXiv:1704.04110](https://arxiv.org/abs/1704.04110) [cs, stat]

Dang L, Cowan NB, Schwartz JC, Rauscher E, Zhang M, et al. 2018. *Nat Astron*. 2(3):220-27



# Thank you

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