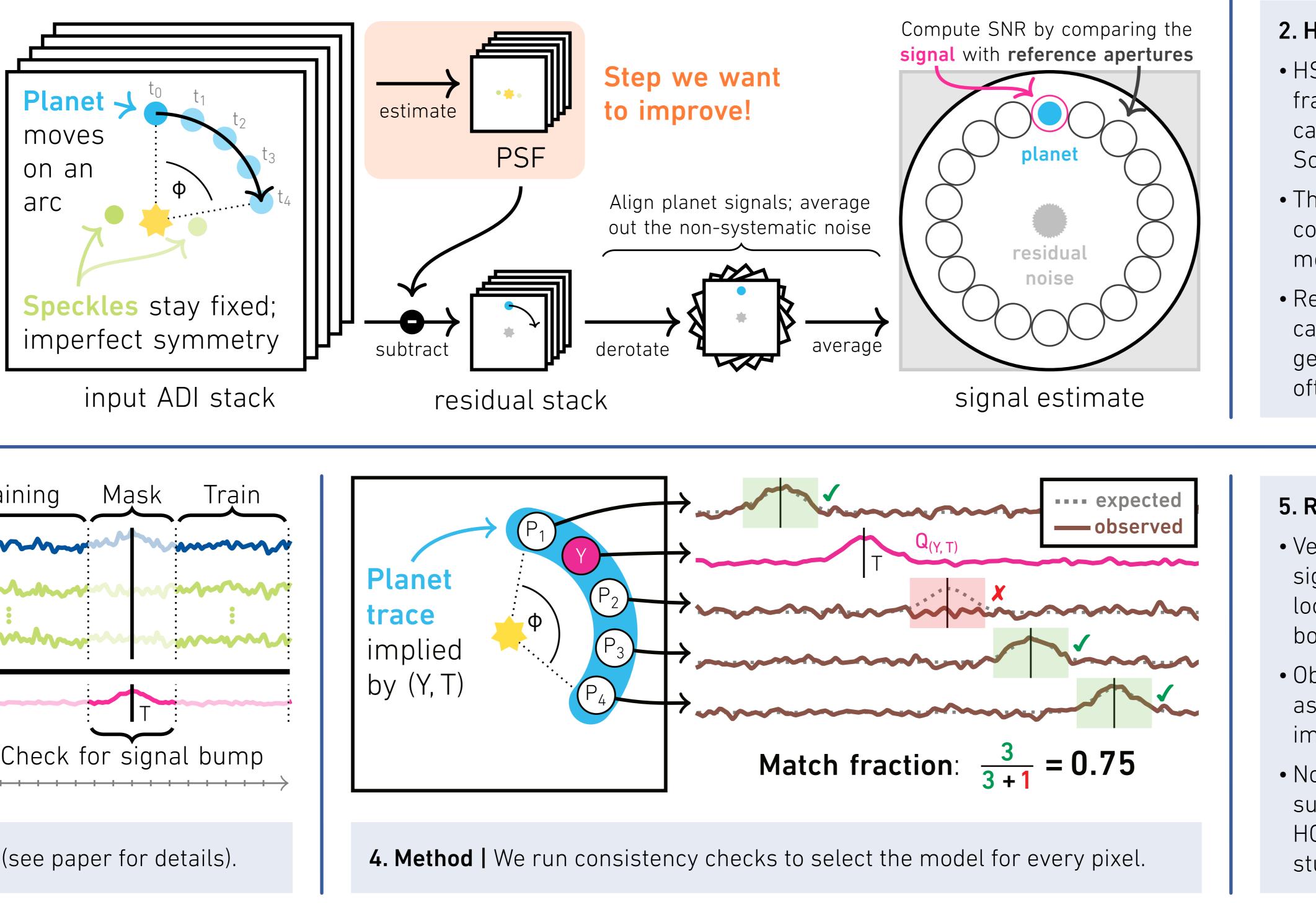
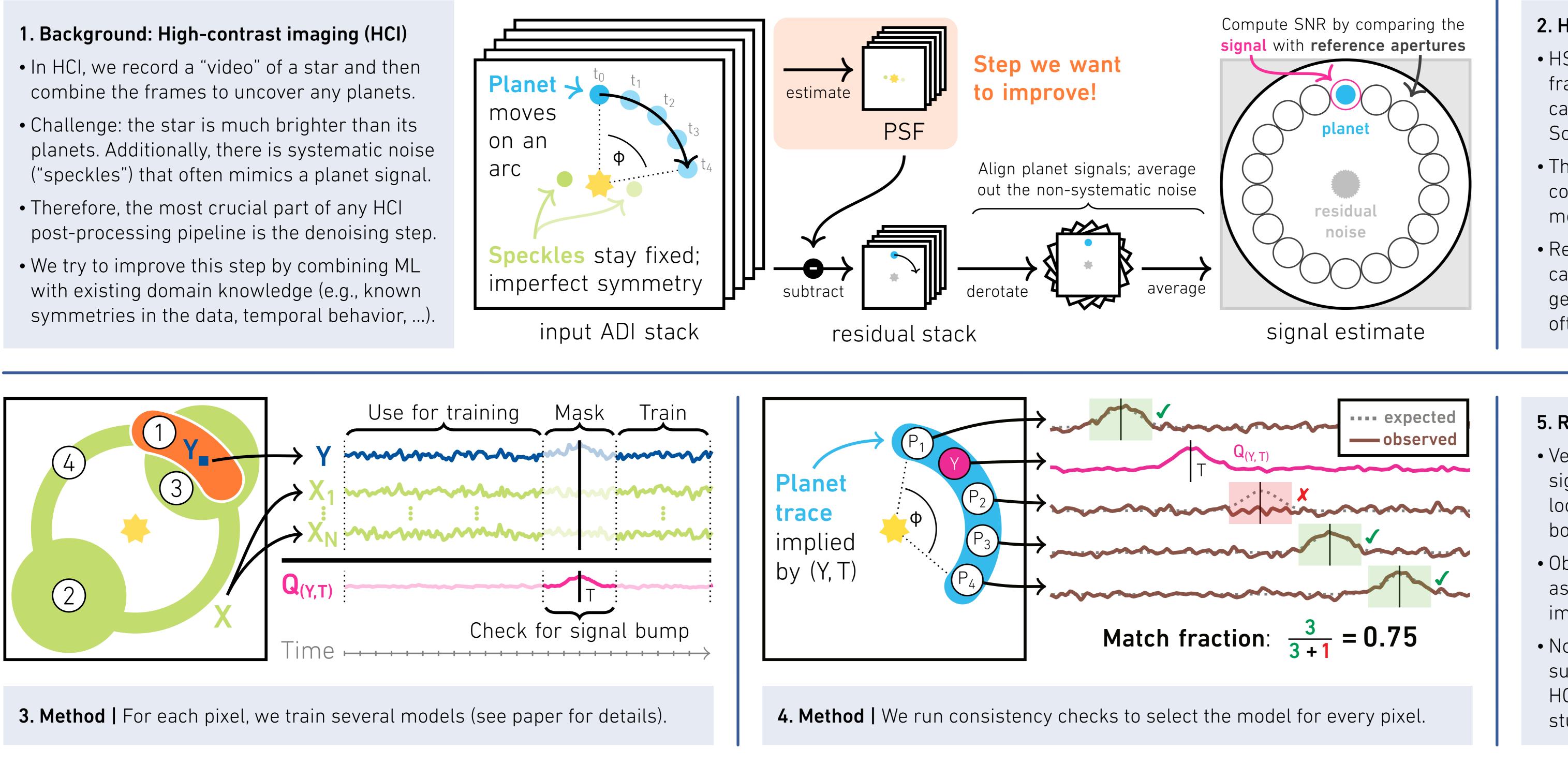
# »We study how to uncover exoplanets in high-contrast imaging data by learning causal, pixel-wise noise models and including scientific domain knowledge.«

## Physically constrained causal noise models for high-contrast imaging of exoplanets

- ("speckles") that often mimics a planet signal.
- with existing domain knowledge (e.g., known









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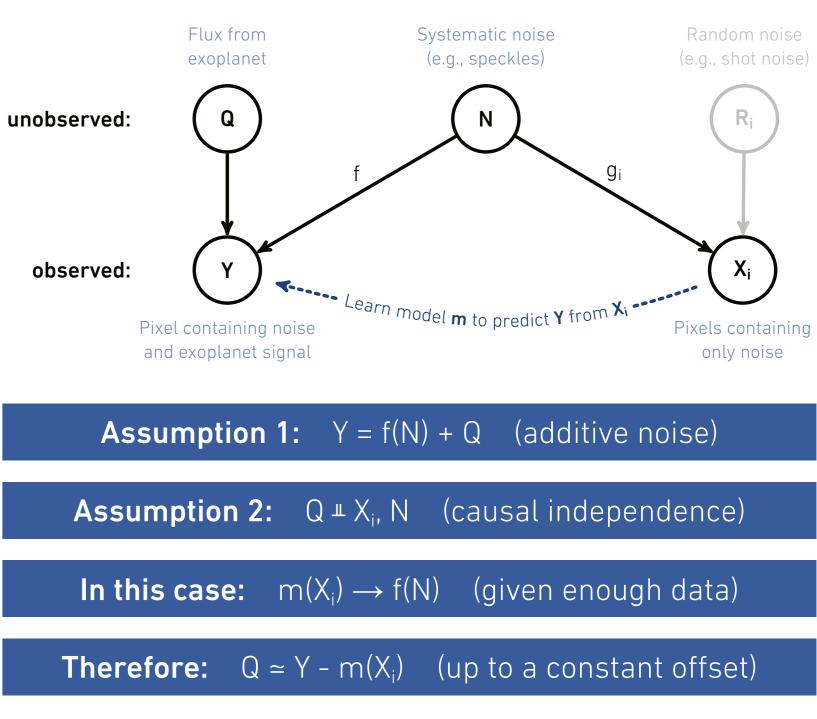
### 2. Half-sibling regression

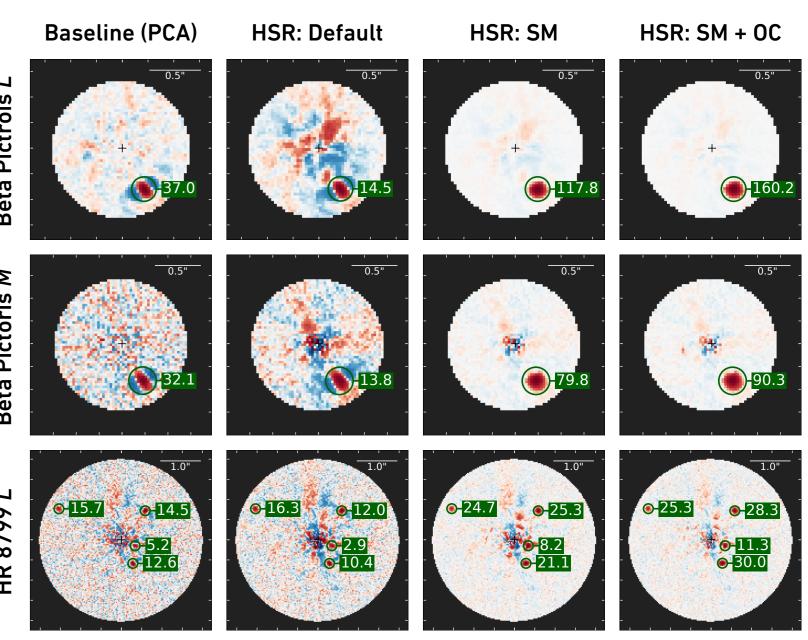
- HSR is a denoising framework based on causality proposed by Schölkopf et al. (2016).
- The key idea is to exploit confounding effects to model systematic noise.
- Relevant: in physics, the causal structure of datagenerating processes is often well-understood.

## 5. Results and outlook

- Very promising: HSR with signal masking (SM) looks better than PCA both visually and in SNR
- Observing conditions (OC) as additional predictors improves results further.
- Note: SNR alone is *not* yet sufficient to characterize HClpp algorithms; further studies are necessary!

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Preliminary, exemplary results for 3 real data sets that contain planets.

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**NEURAL INFORMATION PROCESSING SYSTEMS**