

# A physics-based deep learning approach for focal-plane wavefront sensing

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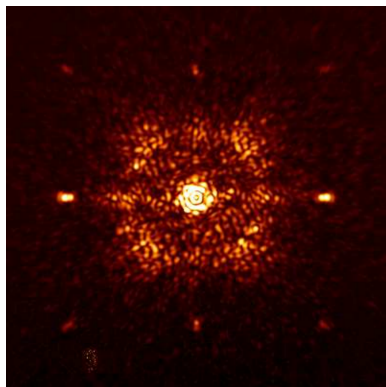
ESA/ESO SciOps workshop

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# Exoplanet imaging: Limitation

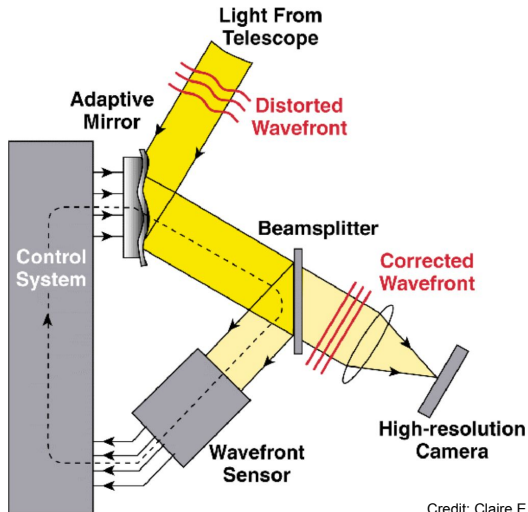
The problem: **SPECKLES**



Martinez et al. 2013

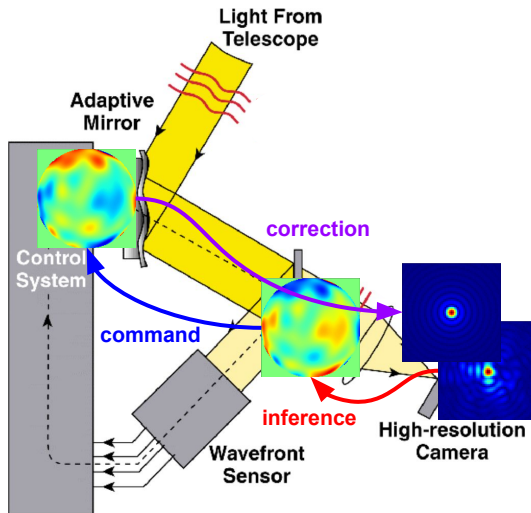
Especially the **quasi-static** ones

# Non-common path aberrations



Credit: Claire E. Max, UCSC

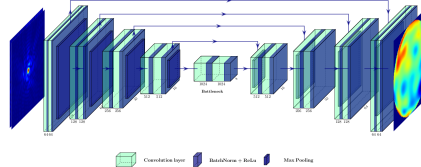
# Focal-plane wavefront sensing: Principle



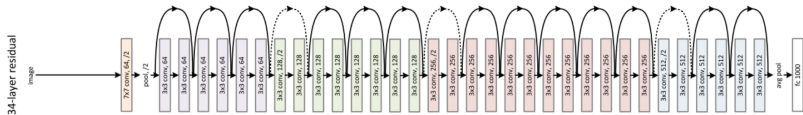
# Deep Convolutional Neural Networks

- Motivation: fast predictions, higher performance, better robustness.
- Deep convolutional neural networks: U-Net, ResNet, EfficientNet.
- Applied on (post-coronagraphic) simulated data and in-lab data.

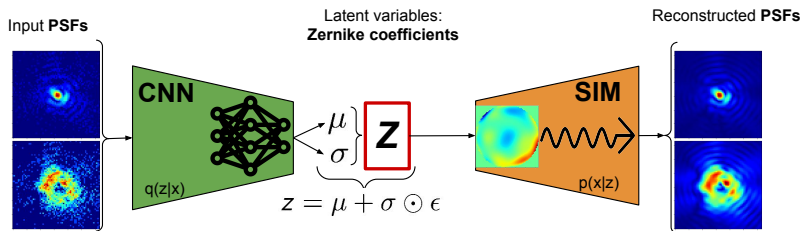
U-Net:



ResNet:



# Simulator-based autoencoder: Unsupervised learning



## Loss function: ELBO

Reconstruction term:

$$\alpha \mathbb{E}_{q(z|x; \phi)} [\log(p(x|z))]$$

$$p(x|z) := \text{Pois}(\lambda)$$

K-L divergence term:

$$-\beta \text{KL}(q(z|x; \phi) || p(z))$$

$$q(z|x) := \mathcal{N}(\mu, \sigma^2)$$

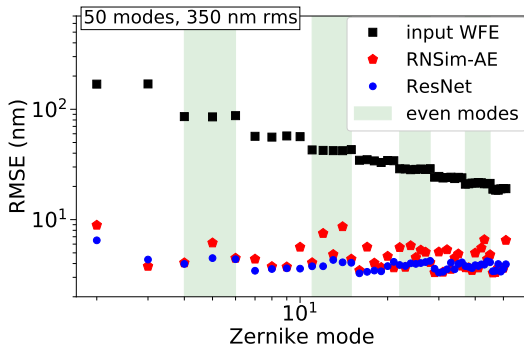
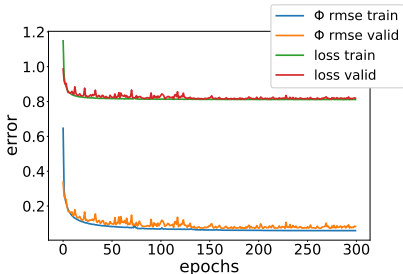
$$p(z) := \mathcal{N}(0, 1)$$

# Sim-VAE: Performance

- Simple simulator:

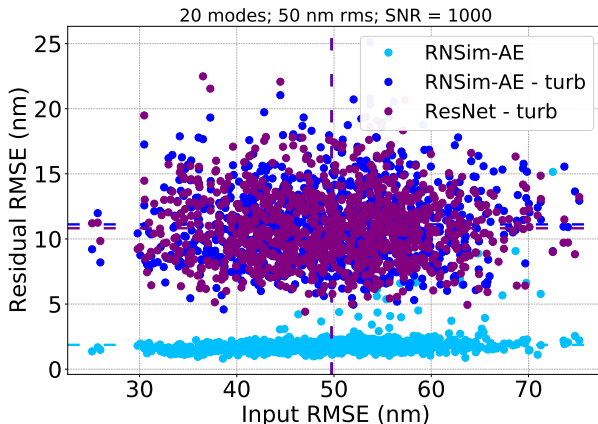
$$\text{PSF}(\mathbf{x}, \mathbf{y}) = |\mathcal{F}[A(\mathbf{x}, \mathbf{y})e^{i\theta(\mathbf{x}, \mathbf{y})}]|^2$$

- K-L divergence term factor:  $\beta = 0$
- Learning rate =  $10^{-4}$
- SNR =  $100 \pm 20$

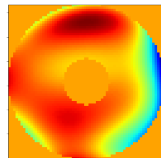


# Sim-VAE: Robustness to AO residuals

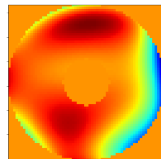
- Adding phase turbulence residuals: 50nm rms,  $t = 100$ ms.
- Information not included in labels (CNN) nor simulator (AE).
- WFS telemetry: could add AO residuals into simulator.



Predicted phase:



True phase:





# Application to real data: SCE<sub>x</sub>AO instrument

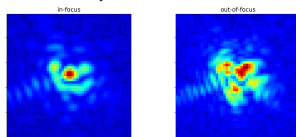


Credits: NAOJ

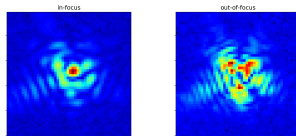
- SCE<sub>x</sub>AO: science and technology development.
- Labelled datasets + control of the instrument.
- Simulations close enough to in-lab PSFs.
- SimVAE: **Morphine** package (Poppy + JAX).  
[github.com/benjaminpope/morphine](https://github.com/benjaminpope/morphine)



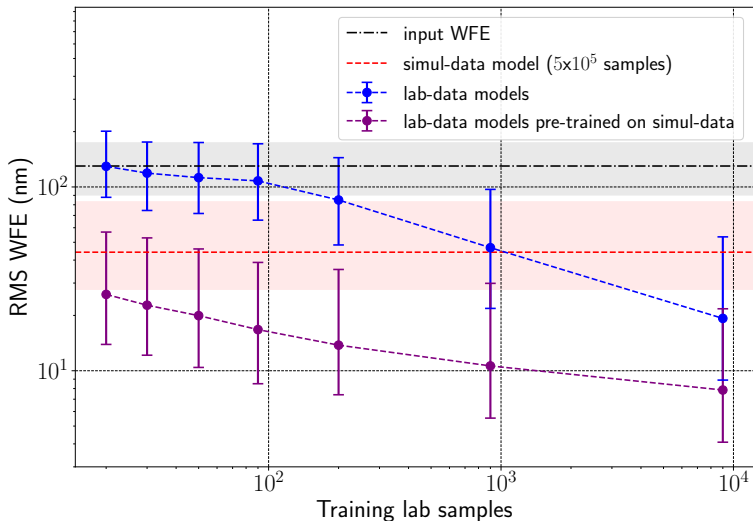
Experimental PSFs:



Simulated PSFs:



# Transfer learning with SCE<sub>x</sub>AO data



# Conclusions

