

A Deep Learning Framework for Joint On-Board InSAR Phase Denoising and Compression

Luca Dell'Amore¹, Lorenzo B. Garavelli¹, Nicola Gollin¹, Michele Martone¹, Paola Rizzoli¹, Begüm Demir²

Luca.DellAmore@dlr.de

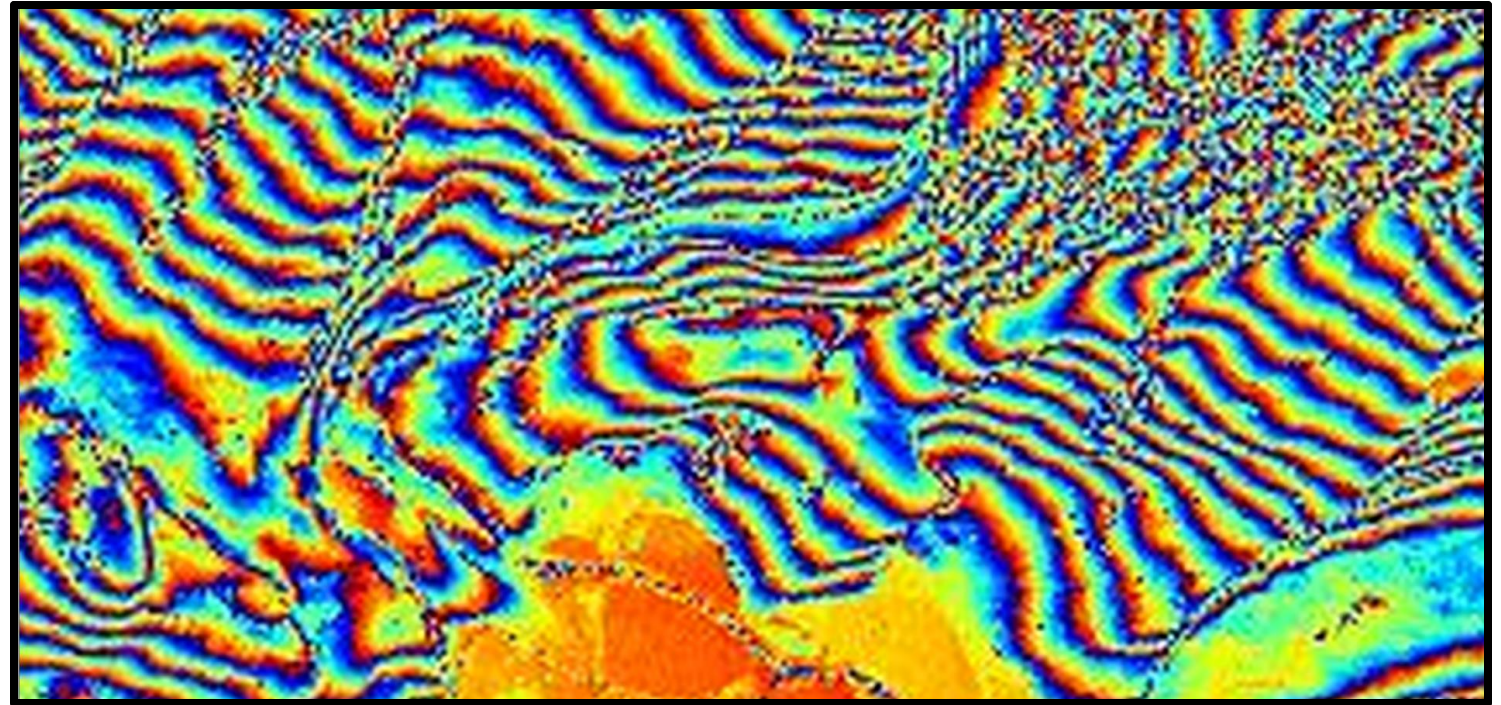
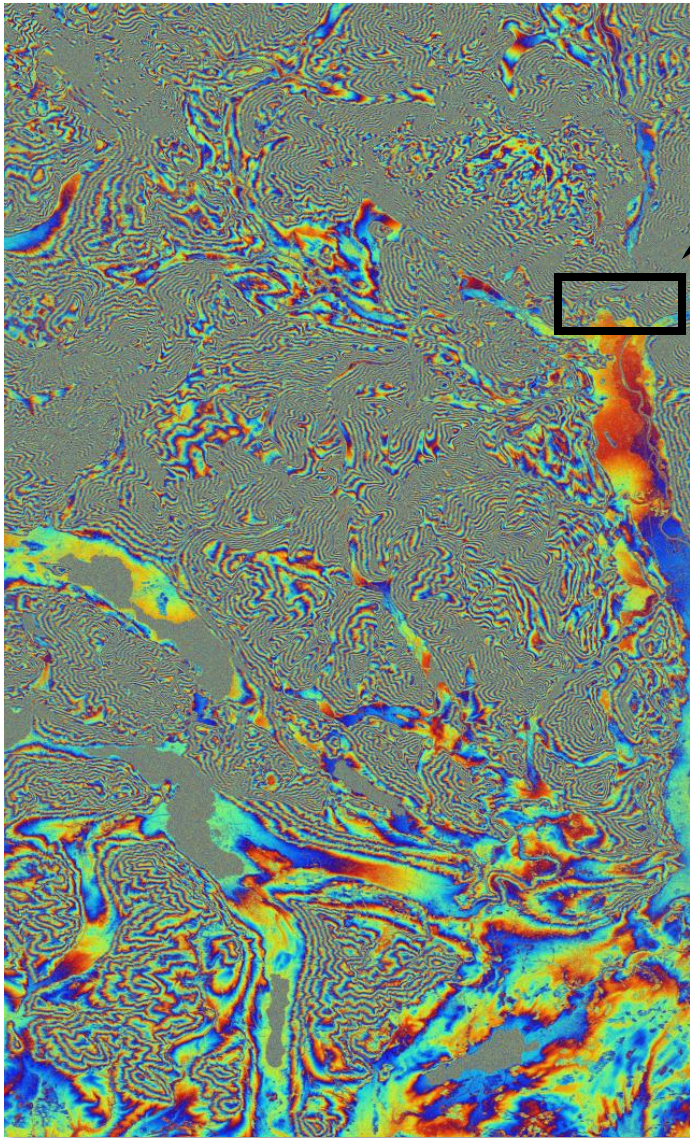
¹ German Aerospace Center (DLR), Germany
Microwaves and Radar Institute (HR)

² Technische Universität Berlin (TUB)



InSAR Phase Denoising vs. Resolution

TDX StripMap: @Singapore [1kx11]

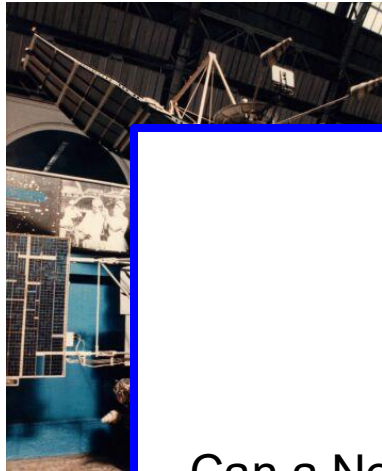


- **Denoising strategies** are required in order to generate reliable and high-quality InSAR products..
- ..common methods estimate the InSAR phase at the **cost of resolution**
- ..**compression** is intrinsically achieved

Problem Definition

- Problem context: **planetary missions**

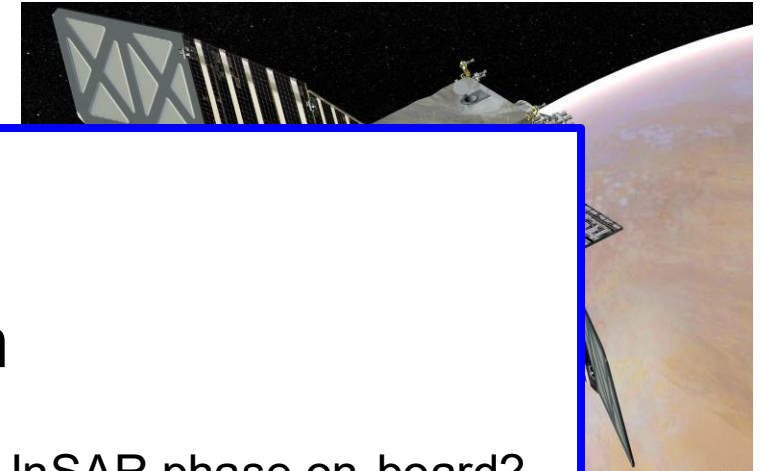
Venera 15&16



NASA's Magellan



NASA's Veritas



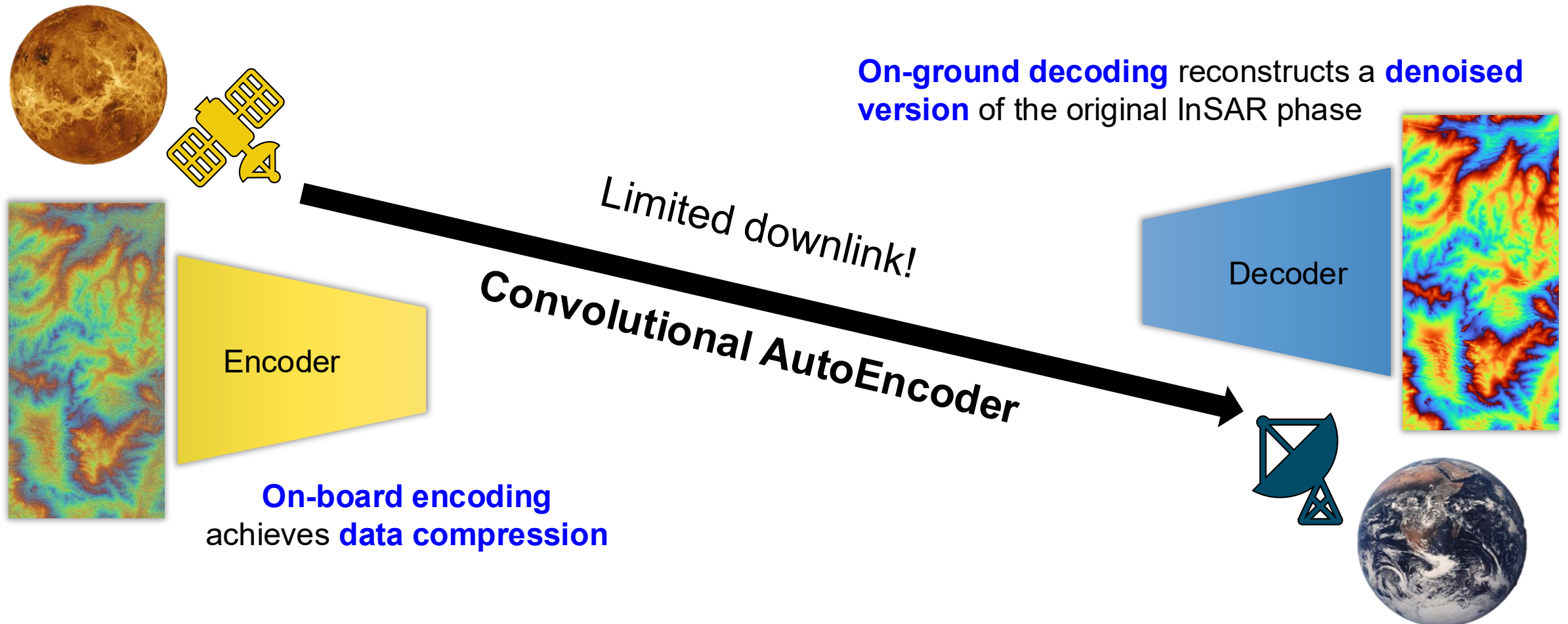
Research Question

Can a Neural Network jointly denoise and compress the InSAR phase on-board?

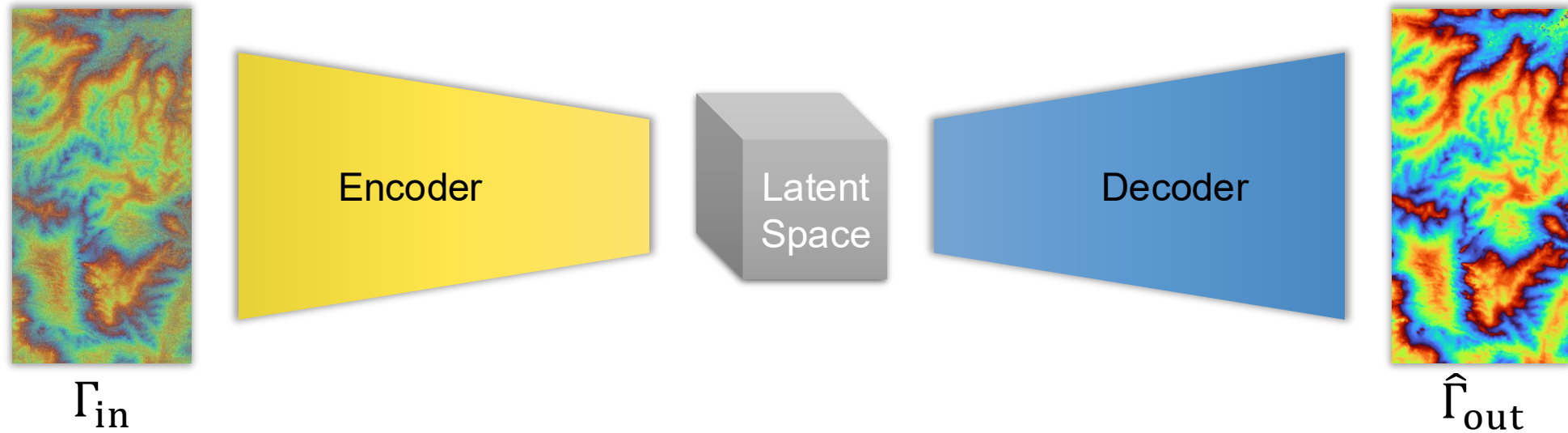
- Planetary missions have **significant downlink constraints**
- NASA's VERITAS objective: generation of a global DEM on Venus
- **On-board** processing (**denoising/compression**) of the **InSAR phase**

Proposed Concept

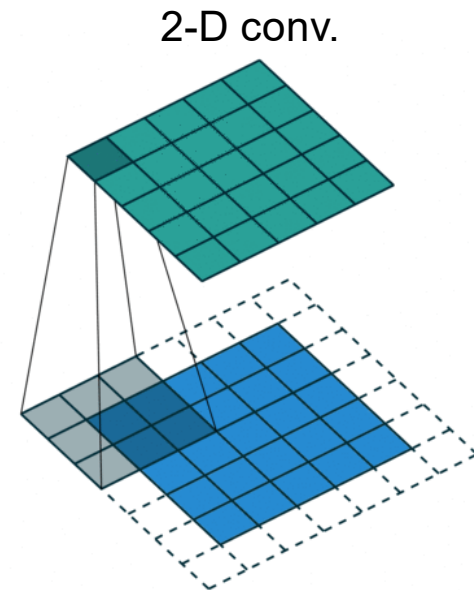
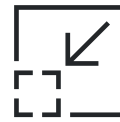
- **Limited downlink** capacity requires effective **compression strategies**
- In this work, **only the InSAR phase is of interest**, while the coherence is neglected



Convolutional AutoEncoder (CAE)

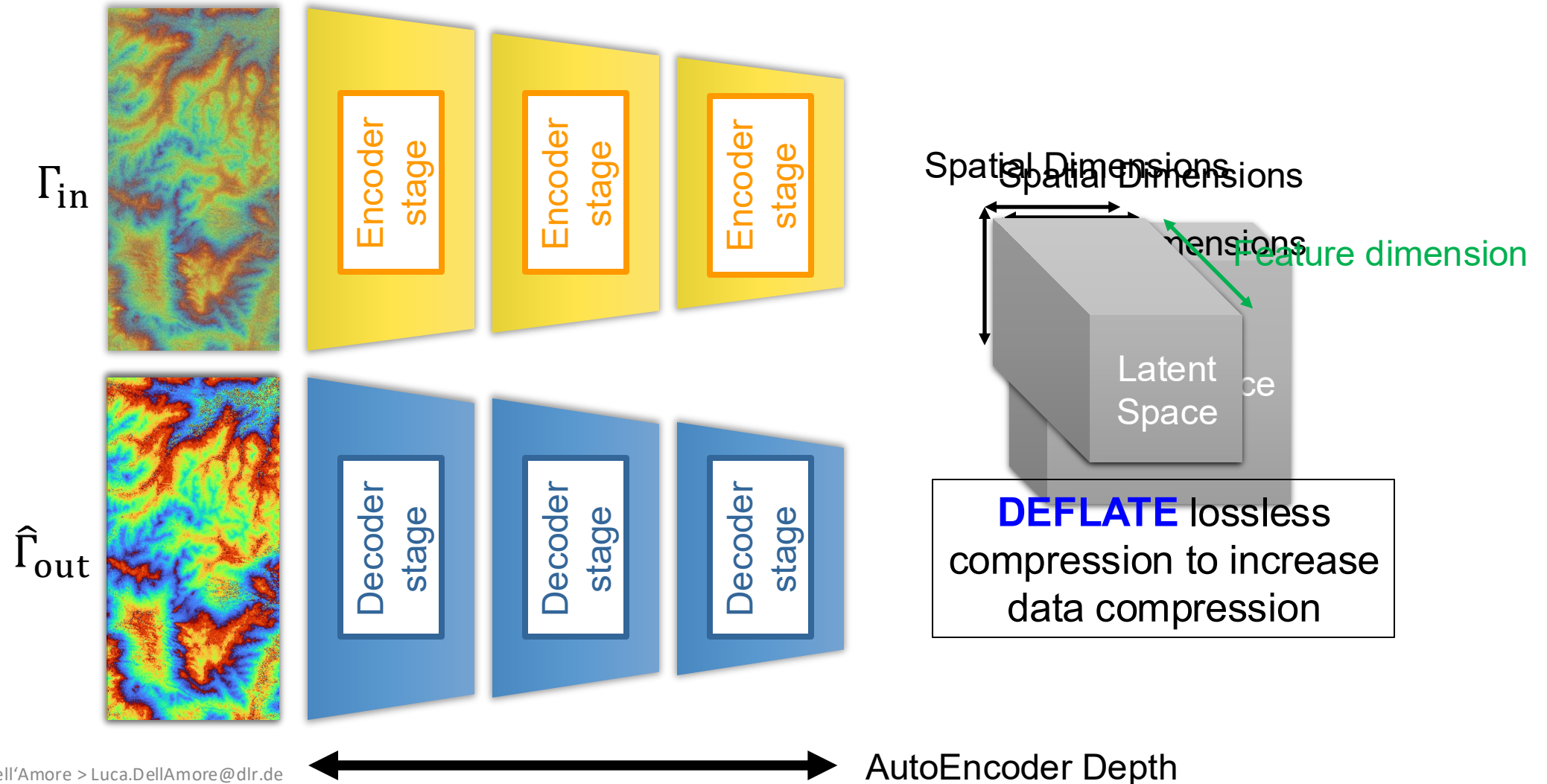


- **Complex interferogram** as **input** (real/imaginary parts): Γ_{in}
- **Encoder** operations:
 - ✓ 2-D convolutions
 - ✓ Pooling operation (**down-sampling**)
- The **coded** and **compressed representation of the input** is called **Latent Space**
- **Decoder** operations:
 - ✓ 2-D convolutions
 - ✓ **Up-sampling** operations reconstruct the original pixel sampling
 - ✓ **High-resolution** phase **details** are partially **lost**
- **Complex InSAR phase** as **output** (real/imaginary parts): $\hat{\Gamma}_{out}$

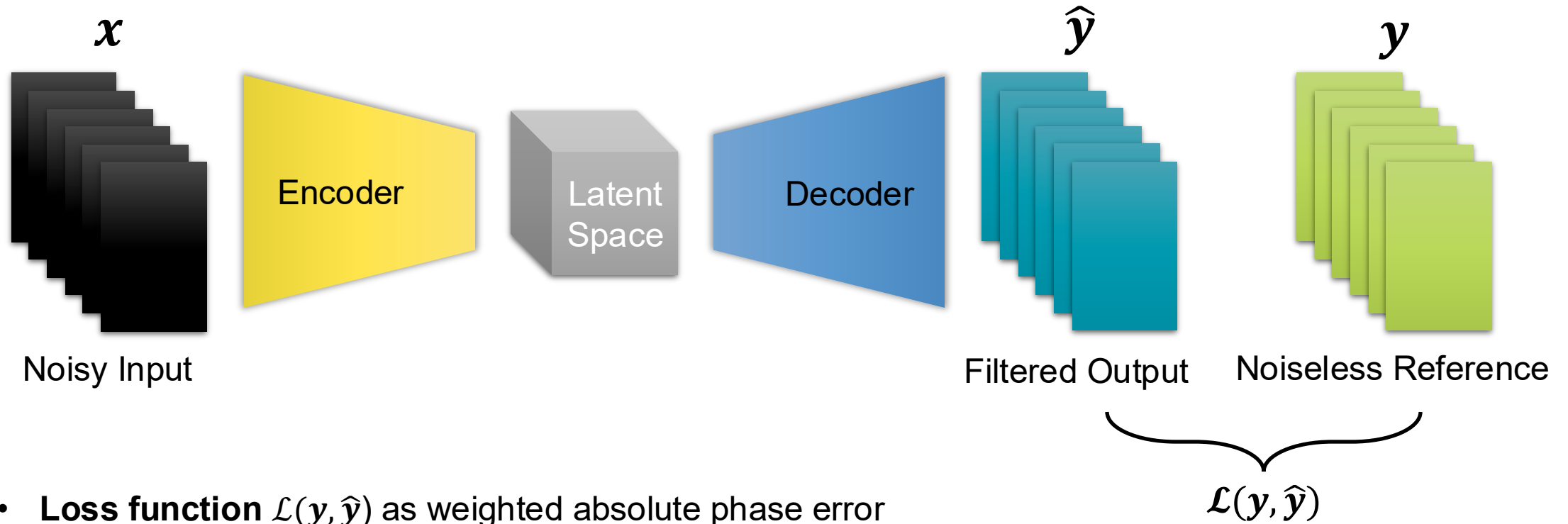


CAE Hyperparameters Tuning

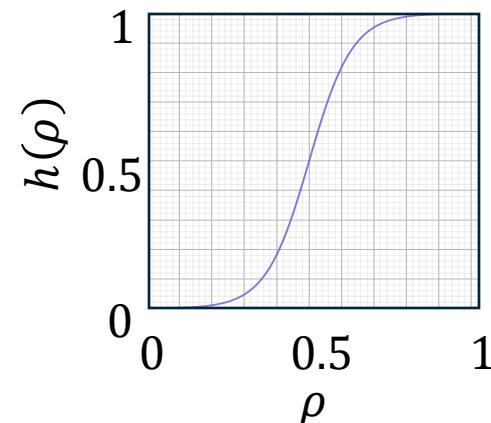
Hyperparameters tuning on different settings allows for exploring different performance trade-offs



Supervised Training Strategy



$$\mathcal{L}(y, \hat{y}) = \sum_i h(\rho_i) \left| \angle \left(\frac{\hat{y}_i}{y_i} \right) \right|$$



Dataset Generation

TanDEM-X NEBZ



$$\Sigma = \mathbb{E} \begin{Bmatrix} |z_1|^2 & z_1 z_2^* \\ z_1^* z_2 & |z_2|^2 \end{Bmatrix} \Rightarrow \Sigma = \Sigma_s + \Sigma_n$$

$z_{1,2}$: complex SLCs

H : signal and noise are statistically independent

TanDEM-X observations

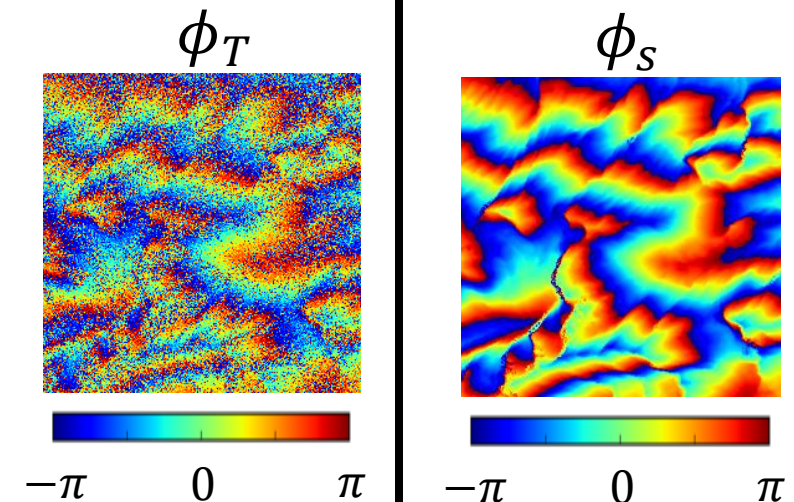
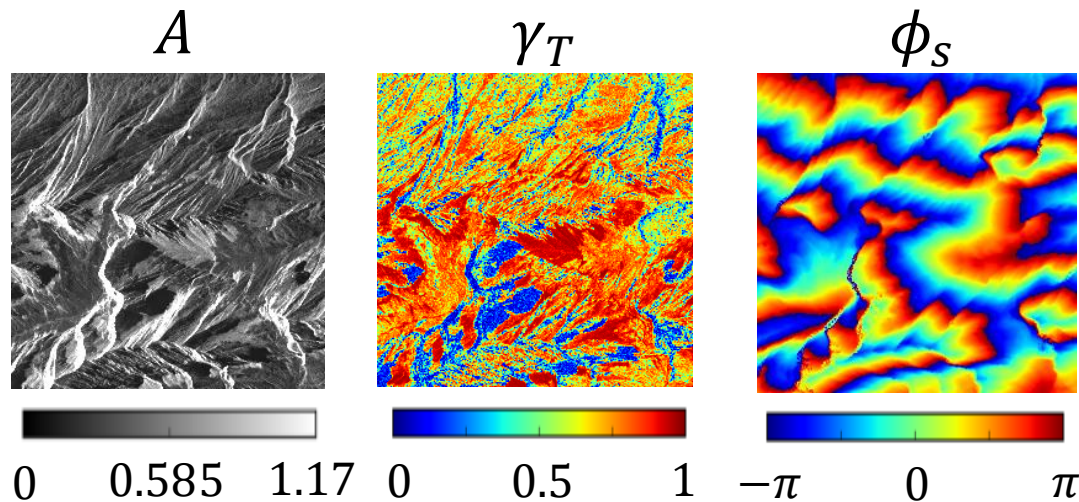
External DEM
back-geocoding



Noisy Input



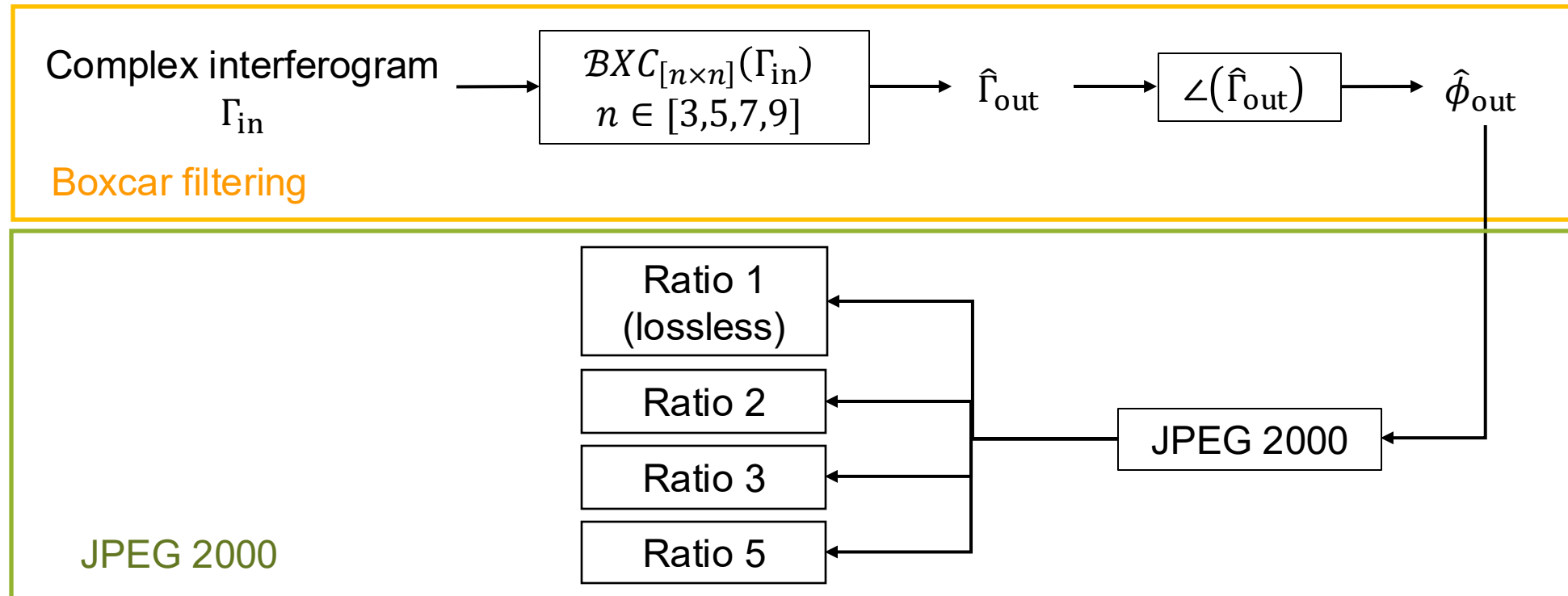
Noiseless Reference



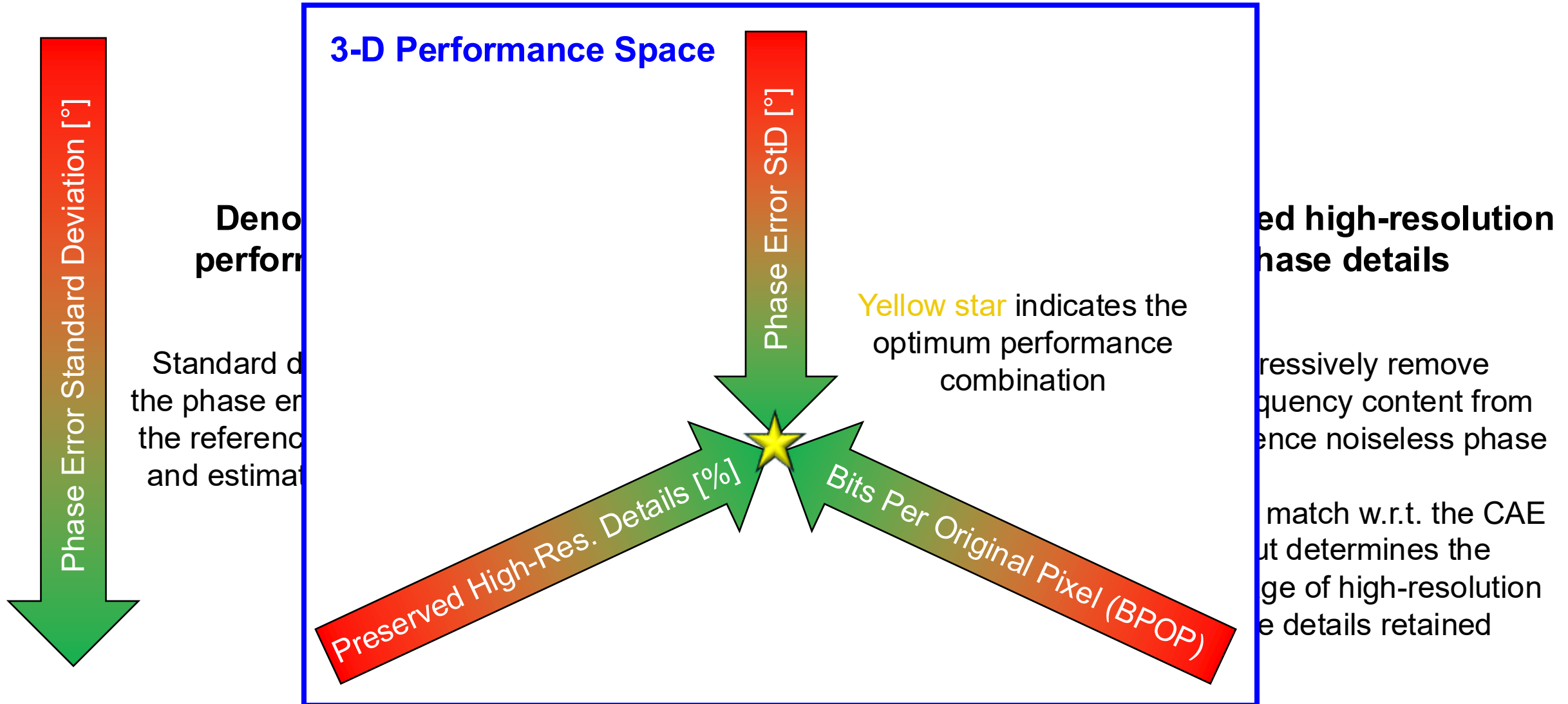
Baseline Method

A combination of traditional methods is used as **performance baseline**:

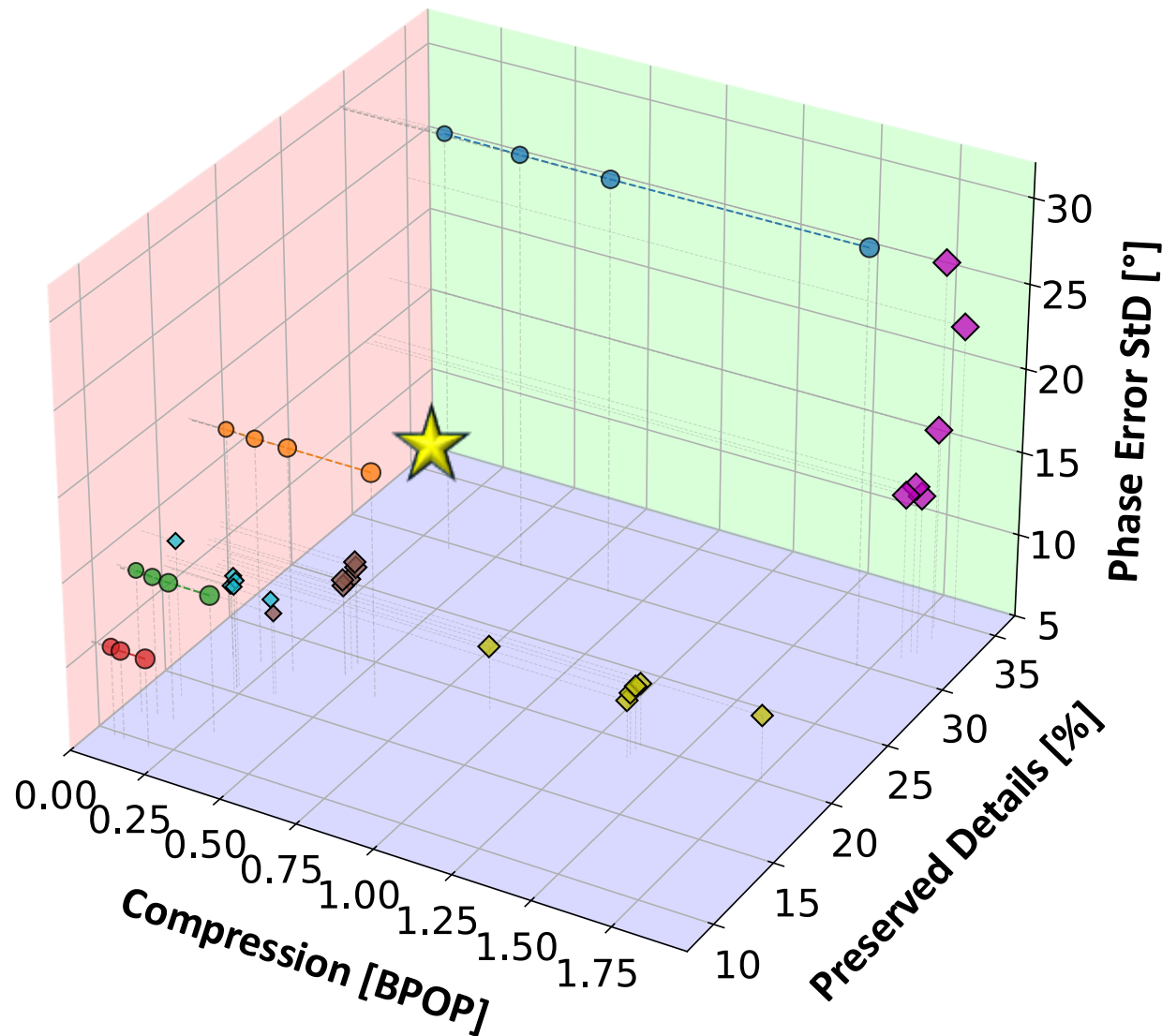
1. **Boxcar filtering** (BXC) for InSAR phase **denoising**
2. **JPEG 2000** for data **compression**



Performance Metrics



Results – 3-D Performance Space



- 8 scenes considered for testing
- **Synthetic interferograms**
(proposed theoretical noise model)
- Focus is on **mid- and high-coherence** ($\rho > 0.4$)

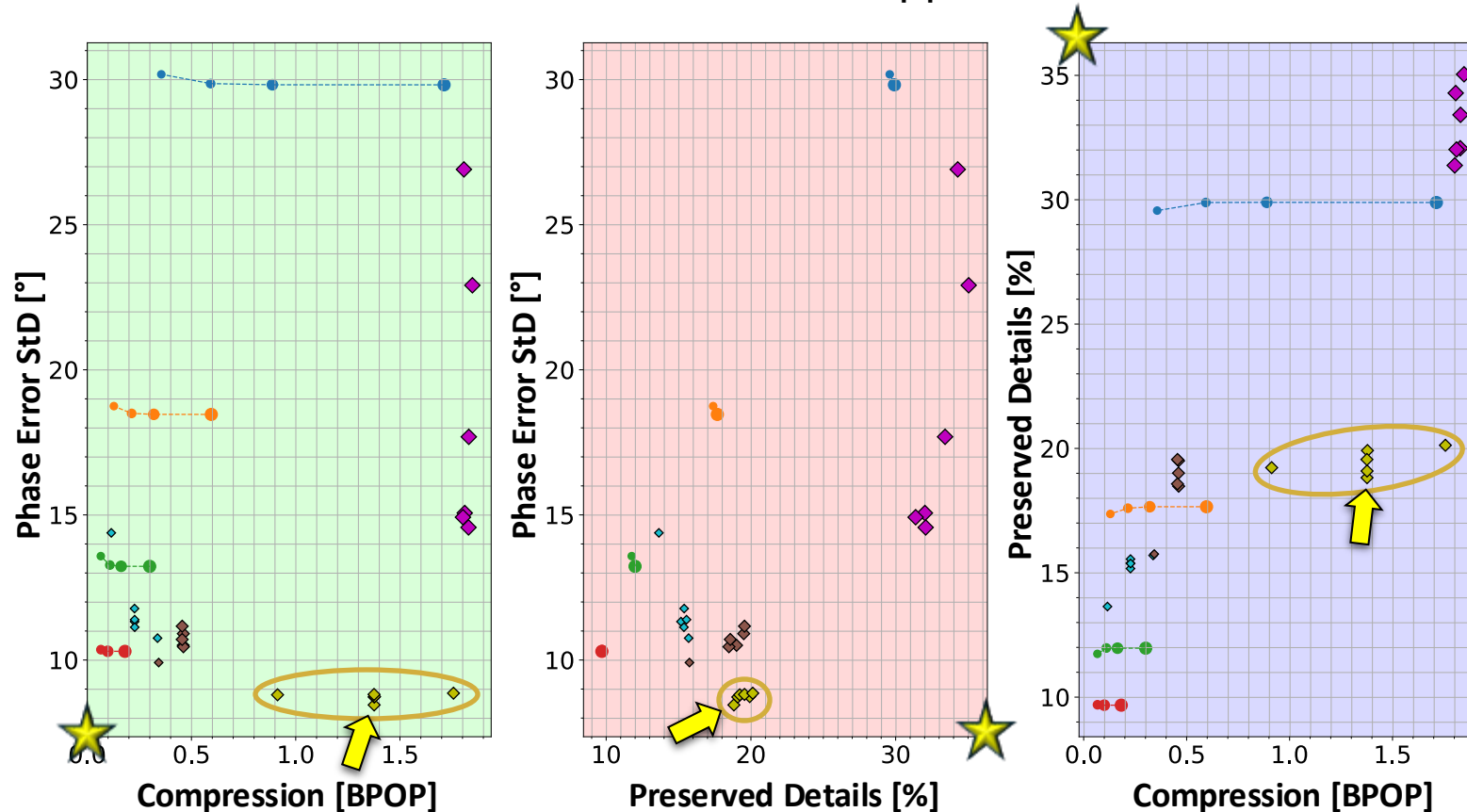
Baseline

- Boxcar [3x3] (r5,r3,r2,r1)
 - Boxcar [5x5] (r5,r3,r2,r1)
 - Boxcar [7x7] (r5,r3,r2,r1)
 - Boxcar [9x9] (r3,r2,r1)
-
- Best denoising
 - Best compression
 - Best details preservation
 - Best compromise

Best CAE solutions

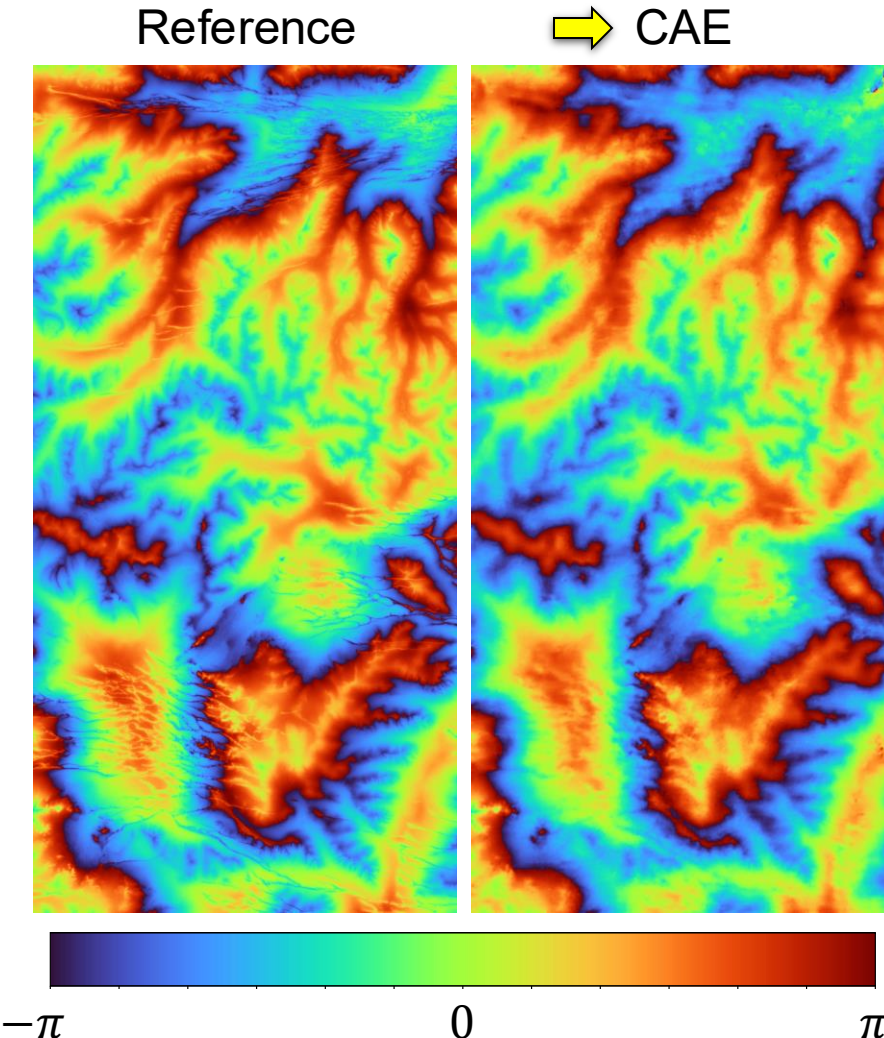
Results – 2-D Cross-Sections

- **Enhanced** overall **flexibility** w.r.t. baseline method, in terms of **performance scalability**
- **Valid alternative** to most traditional approaches in terms of **absolute performance** as well



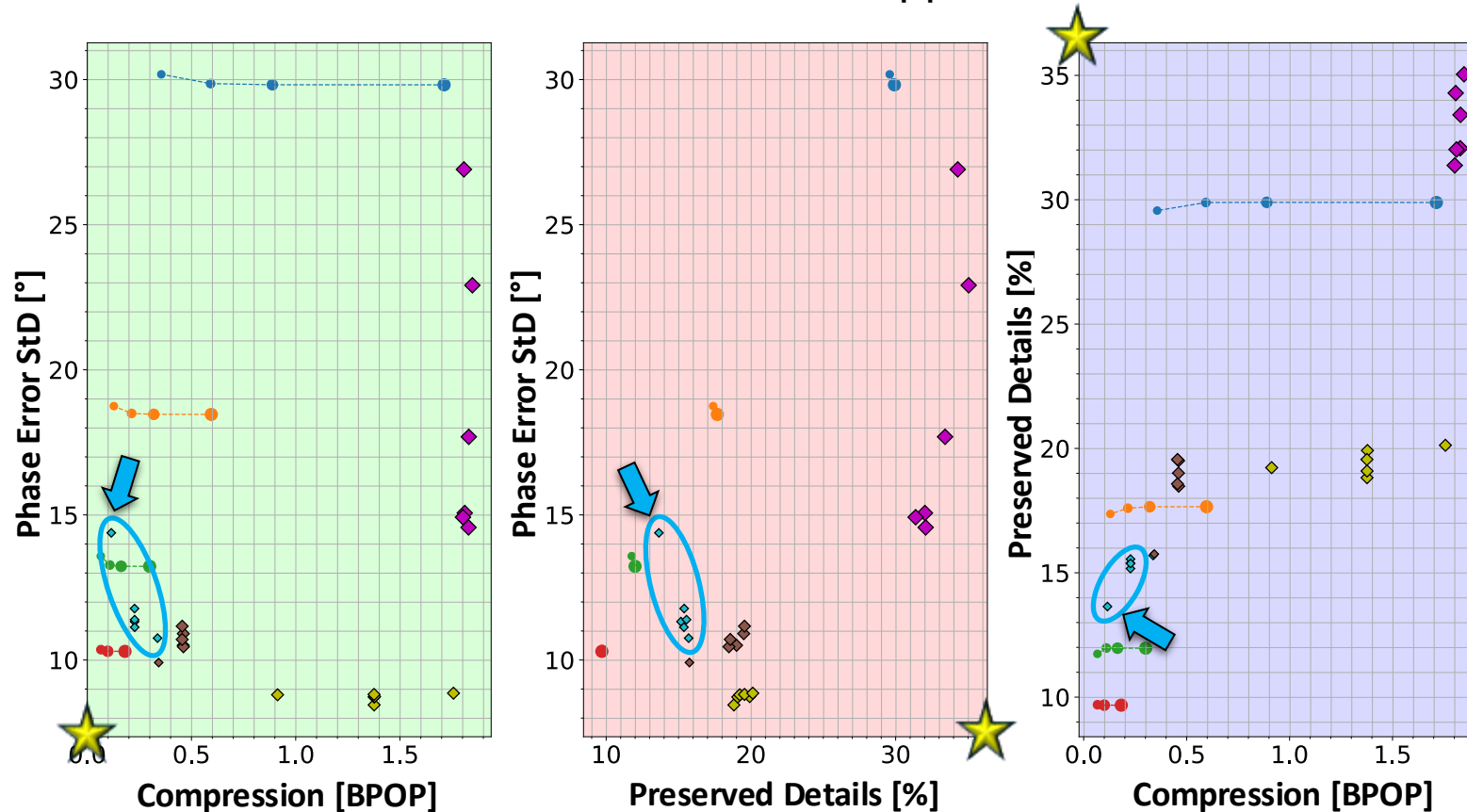
- Boxcar [3x3] (r5,r3,r2,r1)
- Boxcar [5x5] (r5,r3,r2,r1)
- Boxcar [7x7] (r5,r3,r2,r1)
- Boxcar [9x9] (r3,r2,r1)
- Best denoising
- Best compression
- Best details preservation
- Best compromise

→ **CAE depth: 3**
CAE feature space: 3



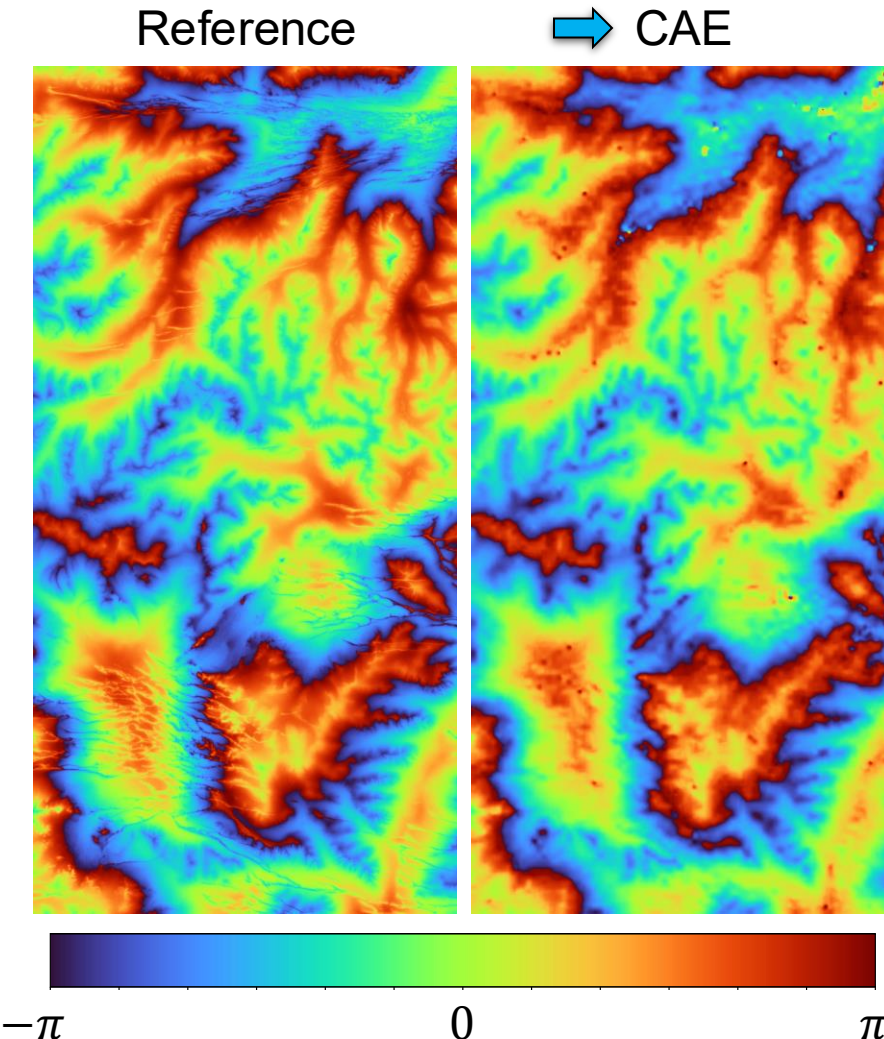
Results – 2-D Cross-Sections

- **Enhanced** overall **flexibility** w.r.t. baseline method, in terms of **performance scalability**
- **Valid alternative** to most traditional approaches in terms of **absolute performance** as well



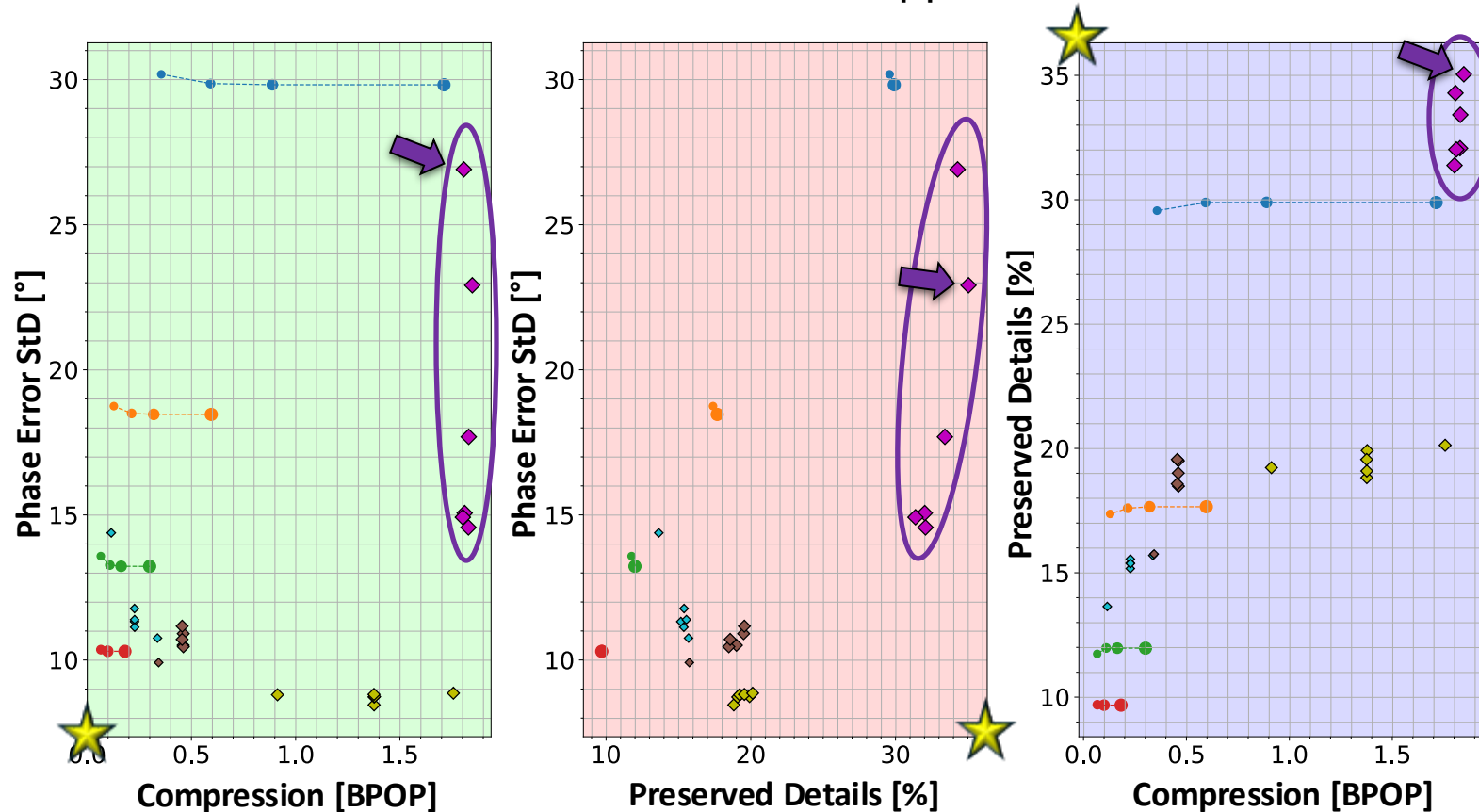
- Boxcar [3x3] (r5,r3,r2,r1)
- Boxcar [5x5] (r5,r3,r2,r1)
- Boxcar [7x7] (r5,r3,r2,r1)
- Boxcar [9x9] (r3,r2,r1)
- Best denoising
- Best compression
- Best details preservation
- Best compromise

→ **CAE depth: 4**
CAE feature space: 1

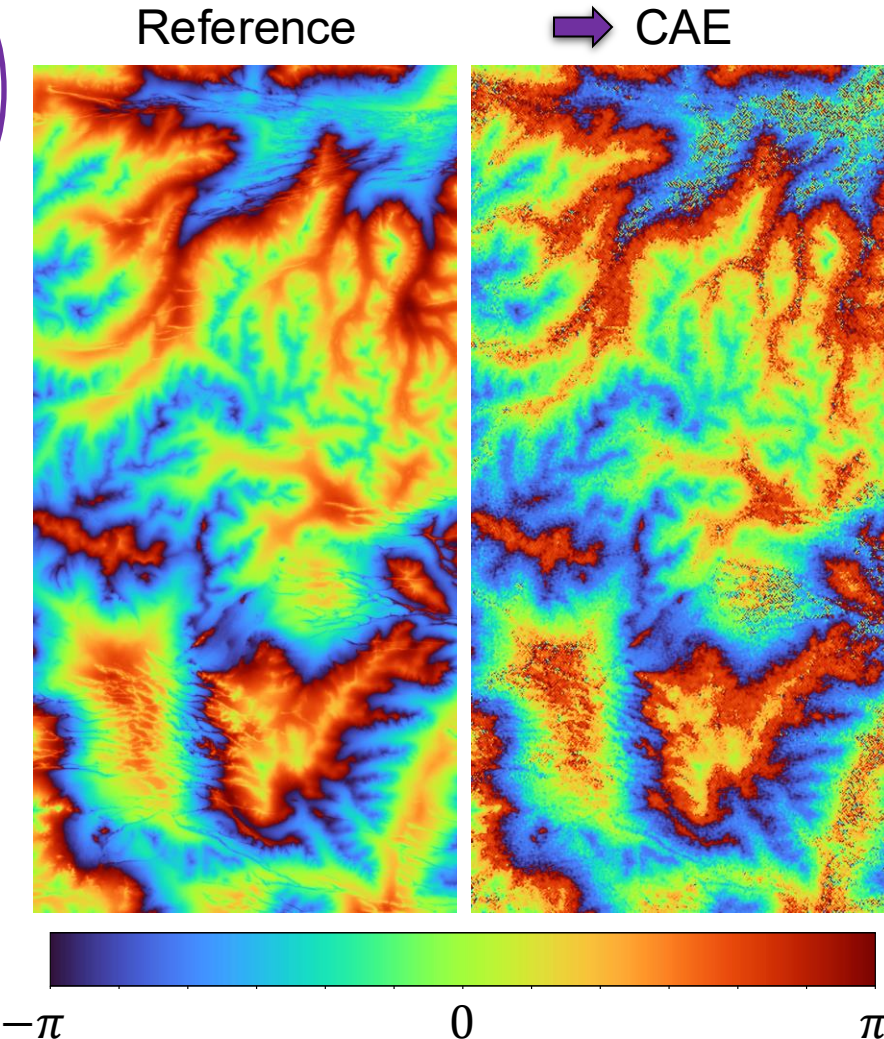


Results – 2-D Cross-Sections

- **Enhanced** overall **flexibility** w.r.t. baseline method, in terms of **performance scalability**
- **Valid alternative** to most traditional approaches in terms of **absolute performance** as well

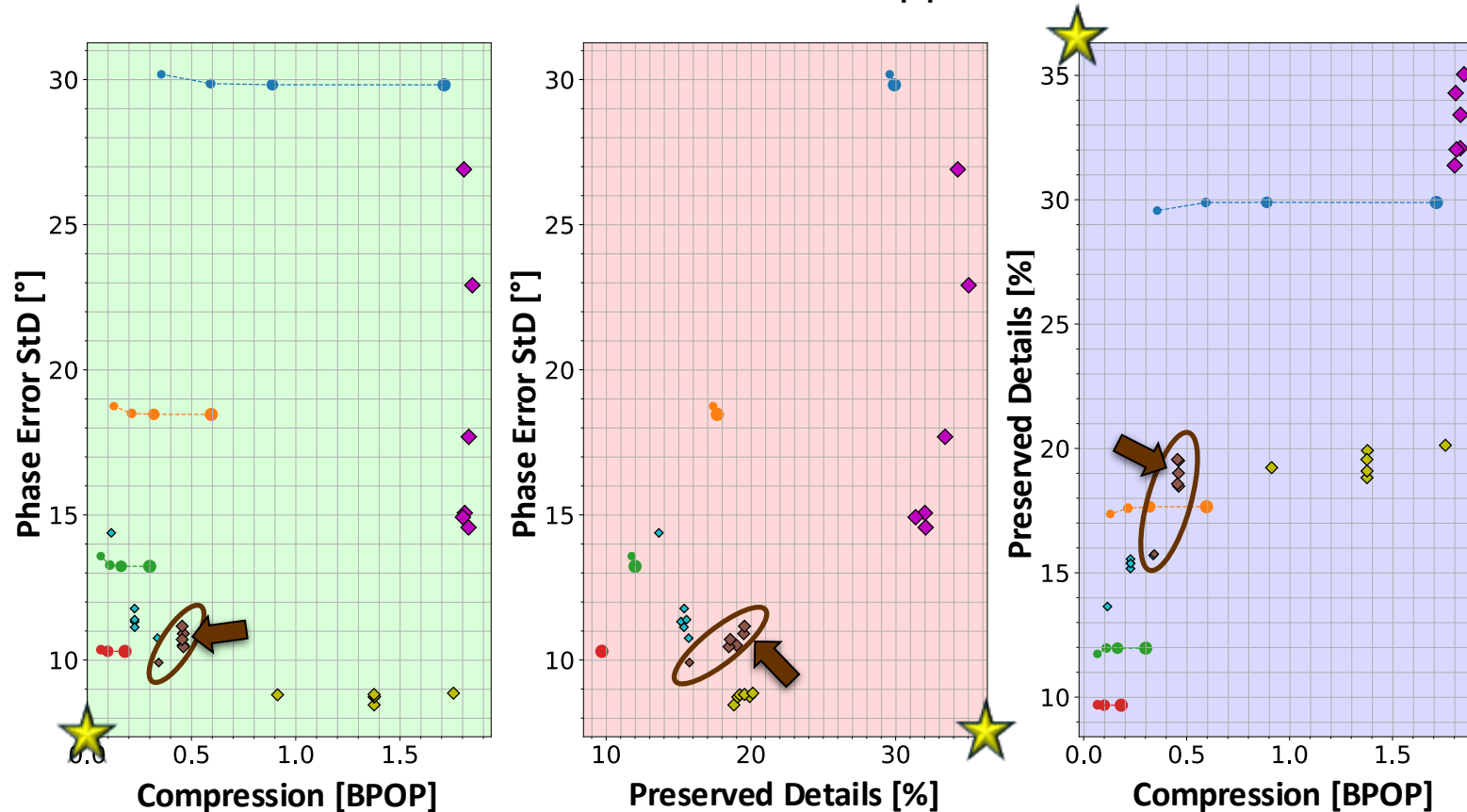


➔ **CAE depth: 2**
CAE feature space: 1



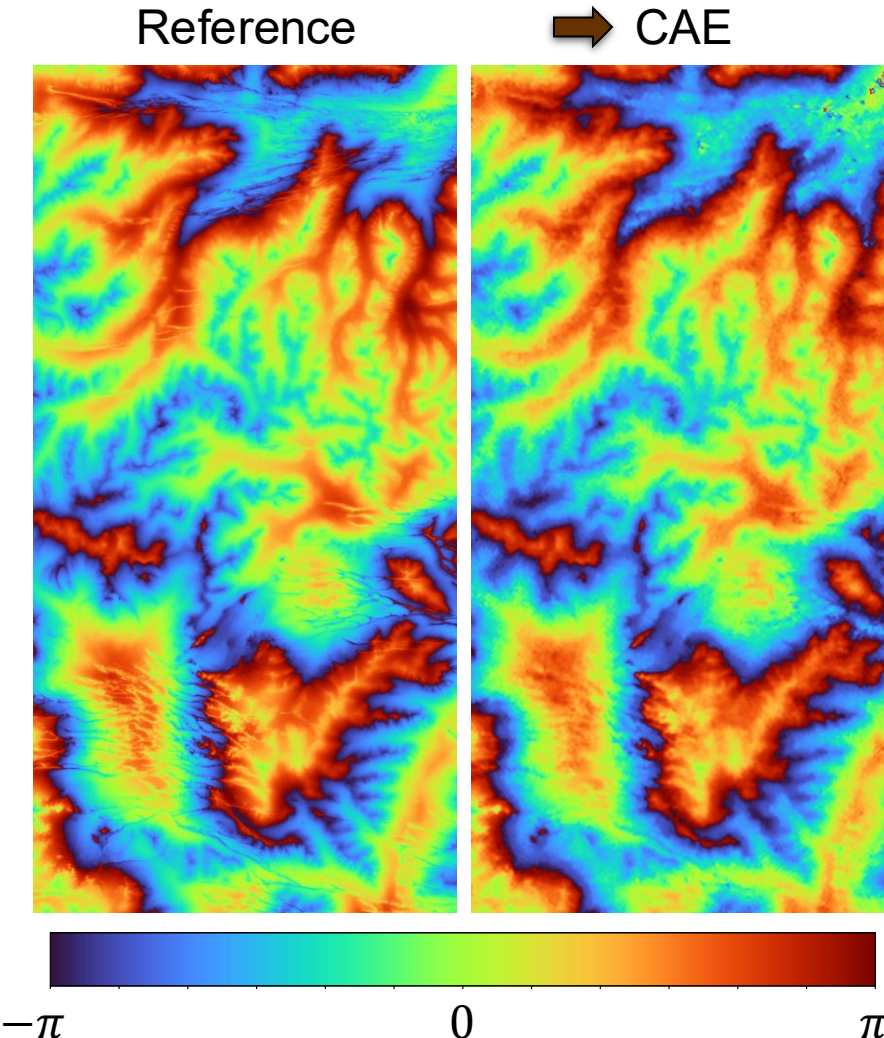
Results – 2-D Cross-Sections

- **Enhanced** overall **flexibility** w.r.t. baseline method, in terms of **performance scalability**
- **Valid alternative** to most traditional approaches in terms of **absolute performance** as well



- Boxcar [3x3] (r5,r3,r2,r1)
- Boxcar [5x5] (r5,r3,r2,r1)
- Boxcar [7x7] (r5,r3,r2,r1)
- Boxcar [9x9] (r3,r2,r1)
- Best denoising
- Best compression
- Best details preservation
- Best compromise

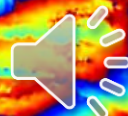
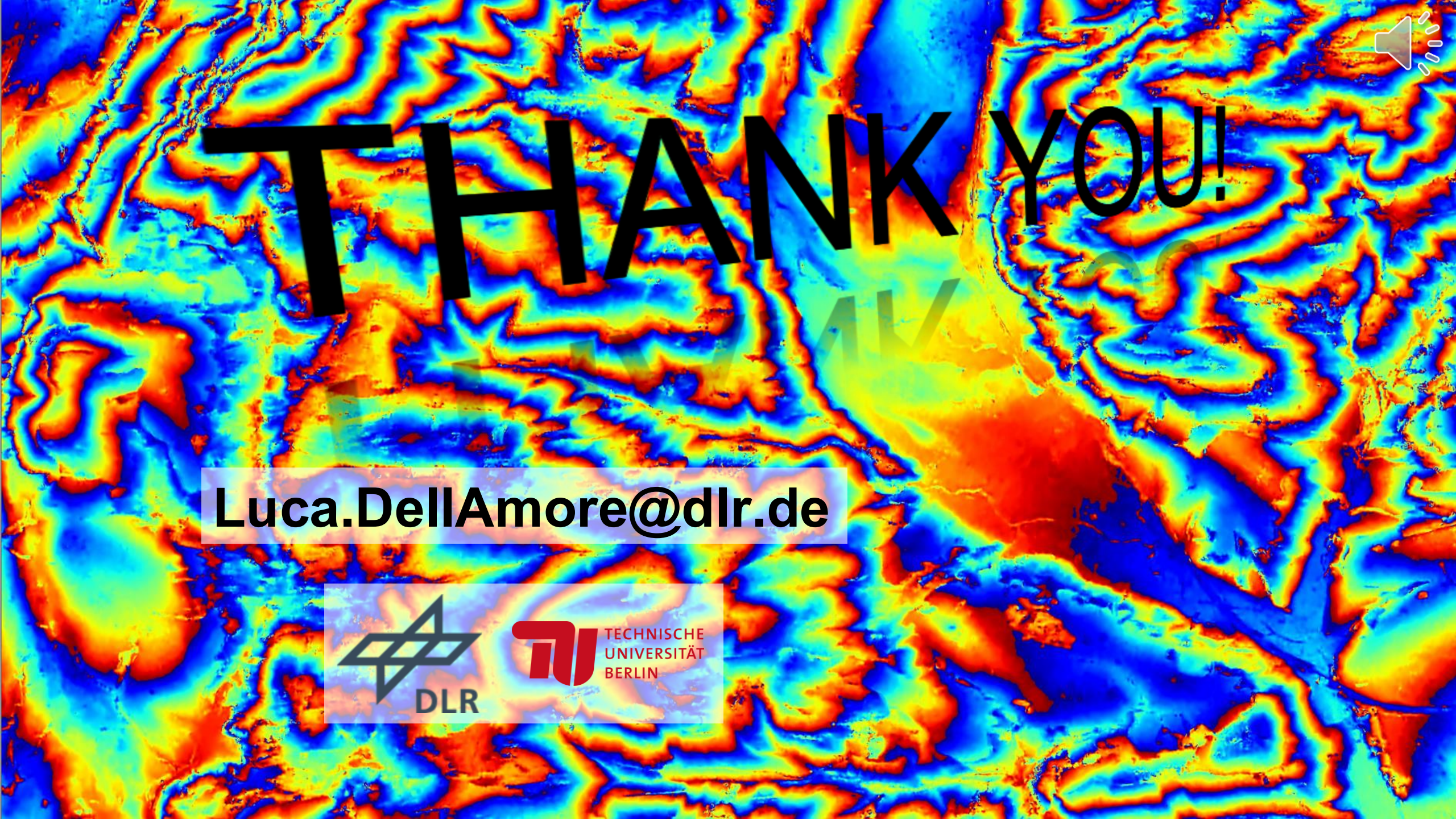
➔ **CAE depth: 3**
CAE feature space: 1



Conclusions & Outlook



- Planetary missions are significantly constrained by **downlink capability**, requiring **effective compression strategies**
- In this work, we target the **on-board processing of the InSAR phase**, while the coherence is neglected
- We proposed a split **Convolutional AutoEncoder (CAE)**:
 - ✓ **On-board encoding** achieves **data compression**
 - ✓ **On-ground decoding** reconstructs a **denoised version** of the original InSAR phase
 - ✓ Due to standard up-sampling operations in the decoder, **high-resolution** phase **details** are partially **lost**
- An **Hyperparameters tuning** on different settings allows for exploring different performance trade-offs
- **Novel mathematical noise model** fully describing a real-case-like interferometric pair, **prior to assumptions**
- A combination of traditional methods (**boxcar filtering + JPEG 2000**) is used as **performance baseline**
- Results are assessed in terms of: **denoising capability**, **compression ratio** and **high-resolution phase details preservation**
- **Enhanced overall flexibility** w.r.t. baseline method, in terms of **performance scalability**
- Valid **alternative** to most traditional approaches in terms of **absolute performance** as well, mainly targeting denoising capability and high-resolution details preservation
- **Outlook**:
 - ✓ The **interferometric coherence** might be estimated as well
 - ✓ **Performance assessment on final DEMs**



THANK YOU!

Luca.DellAmore@dlr.de

