

# PhaseNet for Video Frame Interpolation

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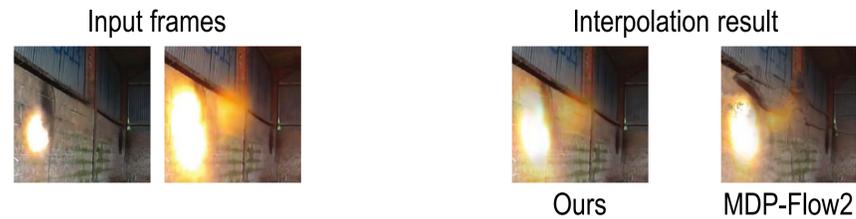
Joint work with Abdelaziz Djelouah, Brian McWilliams, Alexander Sorkine-Hornung, Christopher Schroers

Work done at ETH Zurich and Disney Research



## Motivation

Frame interpolation method without explicit correspondence handling  
 » Suitable for challenging scenarios (brightness change, motion blur)

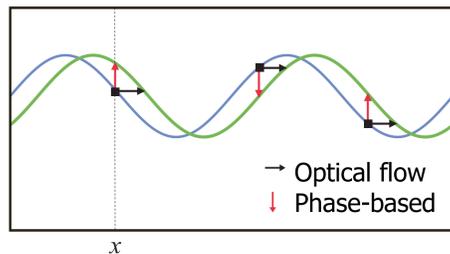


## Contributions

» PhaseNet: Novel architecture which operates on steerable pyramid decomposition of images using phase-based motion representation  
 » Phase loss speeds up training and improves sharpness of result

## Phase-based Motion Representation

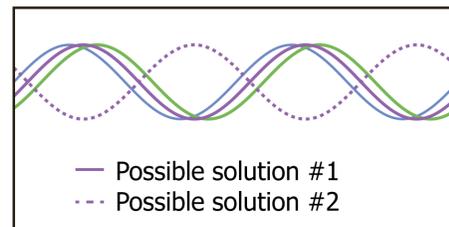
Optical flow vs. phase-based approach



Decomposition in 1 dimension

$$f(x) = \sum_{\omega=-\infty}^{\omega=+\infty} A_{\omega} e^{i\phi_{\omega}}$$

Problem: Periodicity of phase values



Decomposition in 2 dimensions

» With complex steerable filters [3]

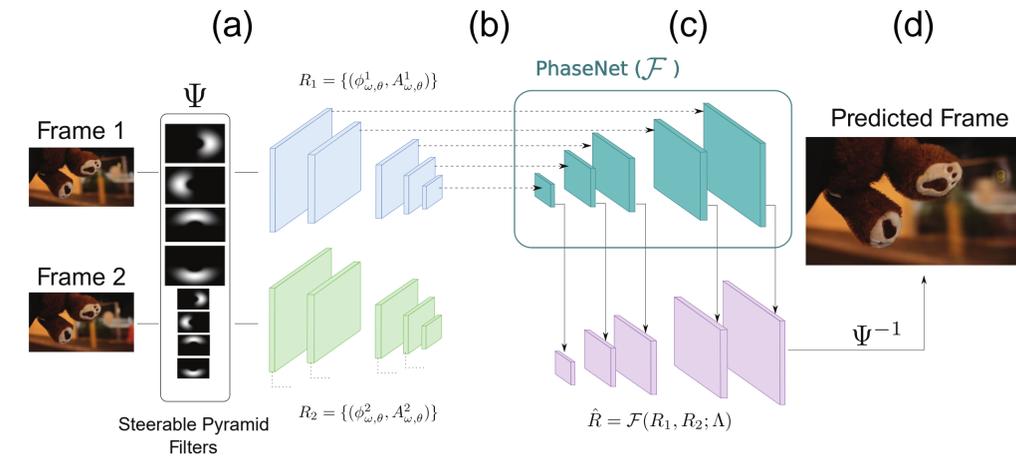
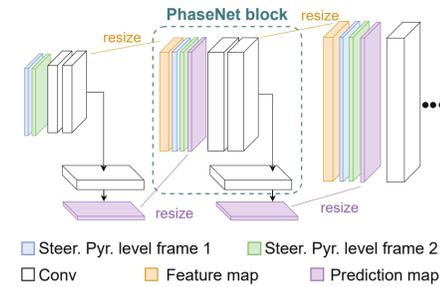
$$R_{\omega, \theta}(x, y) = (I * \Psi_{\omega, \theta})(x, y) \\ = C_{\omega, \theta}(x, y) + i S_{\omega, \theta}(x, y) \\ = A_{\omega, \theta}(x, y) e^{i\phi_{\omega, \theta}(x, y)}$$

Amplitude  $A_{\omega, \theta}(x, y) = |R_{\omega, \theta}(x, y)|$

Phase  $\phi_{\omega, \theta}(x, y) = \text{Im}(\log(R_{\omega, \theta}(x, y)))$

## Algorithm

- Decompose images
- Input decomposition into PhaseNet
- Predict pyramid decomposition
- Reconstruct image from prediction



## Loss Function

Total loss  $\mathcal{L} = \mathcal{L}_1 + \nu \mathcal{L}_{\text{phase}}$

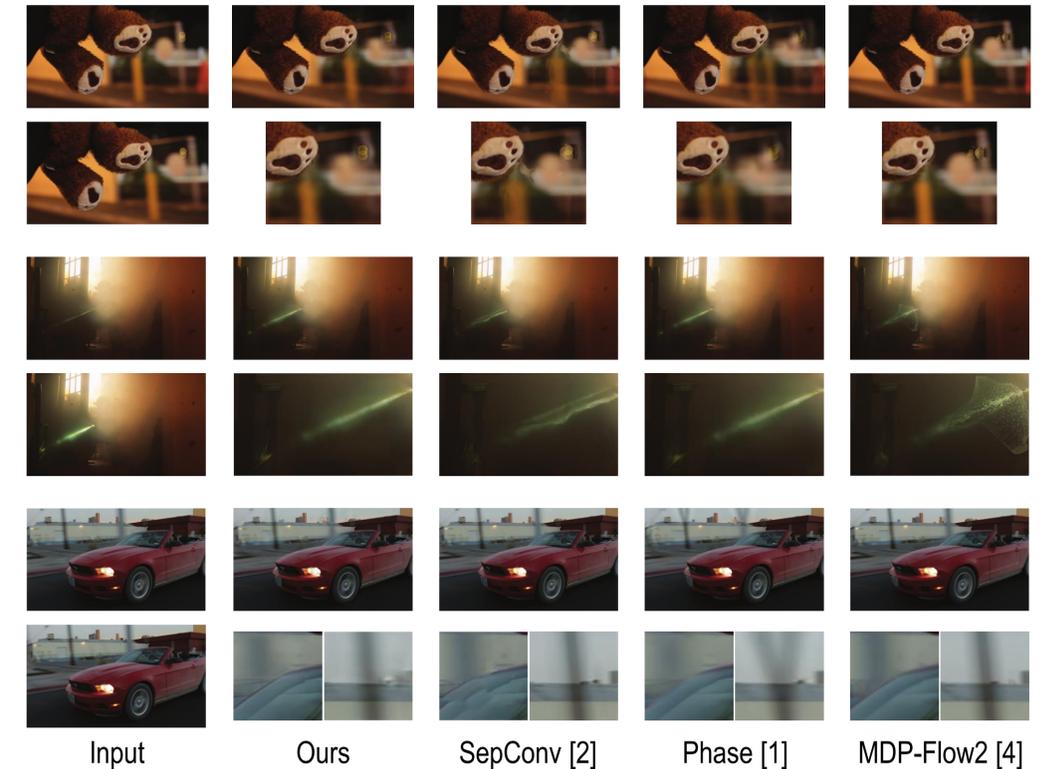
» Image loss  $\mathcal{L}_1 = \|I - \hat{I}\|_1$

» Phase loss  $\mathcal{L}_{\text{phase}} = \sum_{\omega, \theta} \|\Delta\phi_{\omega, \theta}\|_1$

Measures phase difference  $\Delta\phi_{\omega, \theta}$  to the ground truth pyramid decomposition



## Results



## Conclusions

PhaseNet is well suited for interpolating challenging scenes with motion blur and brightness changes at the expense of preserving some fine detail.

## References

[1] S. Meyer et al. Phase-based frame interpolation for video. CVPR, 2015.  
 [2] S. Niklaus et al. Video frame interpolation via adaptive separable convolution. ICCV, 2017.  
 [3] J. Portilla and E. P. Simoncelli. A parametric texture model based on joint statistics of complex wavelet coefficients. IJCV, 2000.  
 [4] L. Xu et al. Motion detail preserving optical flow estimation. PAMI, 2012.