SpIN4D aims to develop deeplearning models that rapidly infer the 4D solar photosphere structure from a sequence of spectropolarimetric observations made by NSF's Inouye Solar Telescope. This poster provides an overview of the project.

SpIn4D: Spectropolarimetric **Inversion in Four Dimensions** with Deep Learning*

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Background

- ρ , etc.

Motivation

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Fig. 1 | Accumulated data volume of NSF/NASA solar observatories. *DKIST*'s data rate averages 20 TB/day. Courtesy K. Reardon.

Objectives

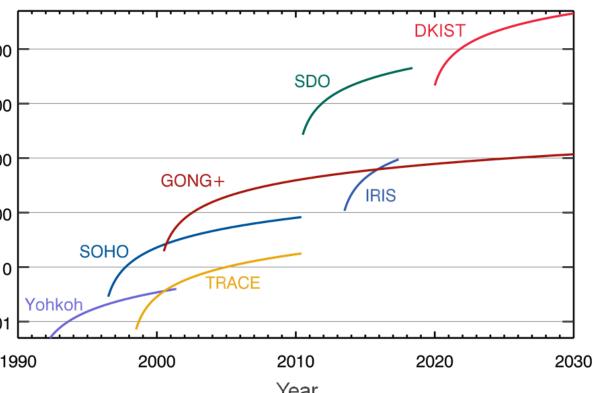
A set of DL models will be trained/tested on MURaM MHD simulations [2] of solar plages (Fig. 2). The SIR code [3] will be used for Stokes synthesis/inversion.

- instrument [4].

• Solar photosphere are well described by *MHD state variables*: magnetic field *B*, temperature *T*, density

• Emergent polarized spectra, known as the *Stokes* profiles, can be used to infer the state variables.

• This *inversion* process requires radiative transfer modeling and can be computationally expensive.



• NSF's *Inouye Solar Telescope* (*DKIST*) will provide high-cadence, high-resolution, multi-line Stokes data with revolutionary diagnostic potential.

• Owing to *DKIST*'s large data rate (Fig. 1), new computational methods are needed to meet the demands of the "big-data" solar physics.

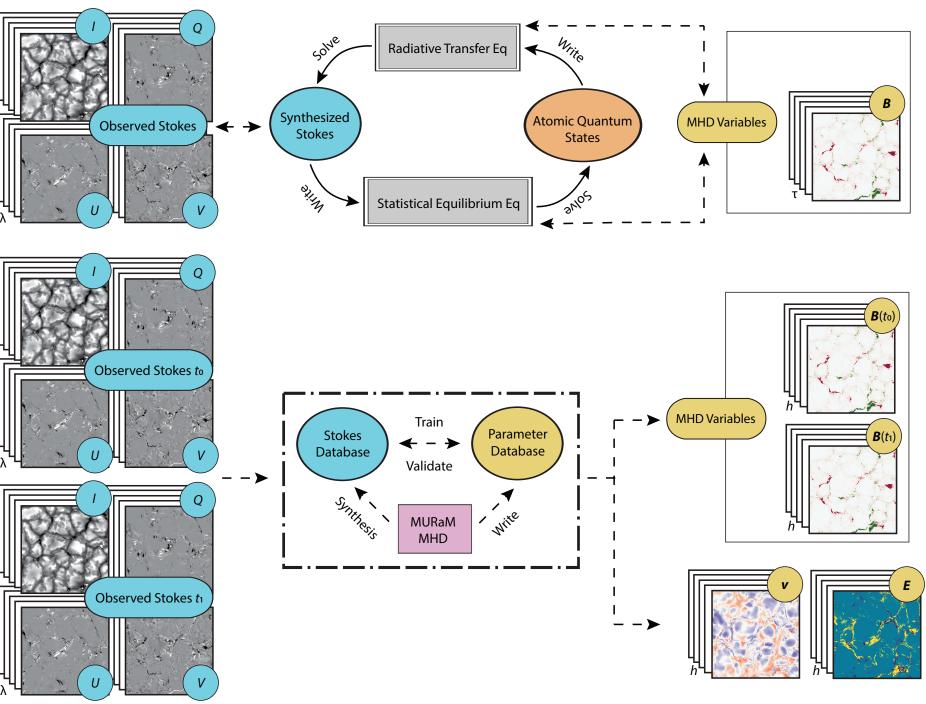
• Advances in deep learning (DL) and MHD simulations allows for faster and more accurate Stokes inversion algorithm, as demonstrated in [1].

• We will use MURaM simulations to create publicly available Stokes data sets that mimic Fe I 630 nm and 1.56 µm observations from *DKIST*/DL-NIRSP

• We will use these data to develop open-source, deep convolutional neural networks that rapidly invert Stokes profiles.

• We will compare our DL models to SIR inversions to benchmark the performance of each.

• We will explore *domain adaptation* methods to reduce potential differences between simulation and observation domains.



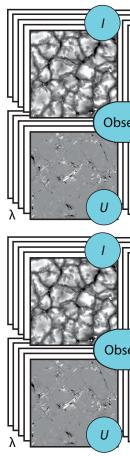


Fig. 2 | Illustration of inversion models. *Top*: traditional methods fit for Stokes profiles (*I*,*Q*,*U*,*V*) at individual pixels. Additional steps are required to derive other parameters (e.g., velocity **v** and Poynting flux **E**). Bottom: SpIN4D, trained on a large library of MHD simulations, will take a temporal sequence of Stokes to utilize the spatial/temporal structure. Higher-level parameters in 4D may be directly estimated.

Highlights

Progress

References

SpIN4D will exploit spatial/temporal (4D) coherence properties in observations (Fig. 2), as well as the implicit physical constraints in MHD simulation.

• SpIN4D will address the 180° azimuthal ambiguity resolution *during* DL inversion.

• SpIN4D will provide the uncertainty over the inferred MHD states using the latest DL methods.

• We performed 10 solar hr of MURaM runs (25 × 25 × 8.2 Mm domain, 16/12 km horizontal/vertical resolution, 40 s cadence) on NCAR's *Cheynne*, totaling 23 TB output, ~20% of all planned.

• We are close to completing the SIR Stokes synthesis pipeline with *DKIST*/DL-NIRSP specs.

• We demonstrated improvement of inversion accuracy using both Fe I lines over a single line with group equivariant neural networks.

• We designed a probabilistic approach to resolving the azimuthal ambiguity.

[1] Asensio Ramos, A., & Díaz Baso, C. J. 2019, *A&A*, **626**, A102 [2] Rempel, M. 2014, *ApJ*, **789**, 132

[3] Ruiz Cobo, B., & del Toro Iniesta, J. C. 1992, *ApJ*, **398**, 375 [4] https://nso.edu/telescopes/dkist/instruments/dl-nirsp/