



Università
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A new tool for investigating the magnetism of the solar chromosphere

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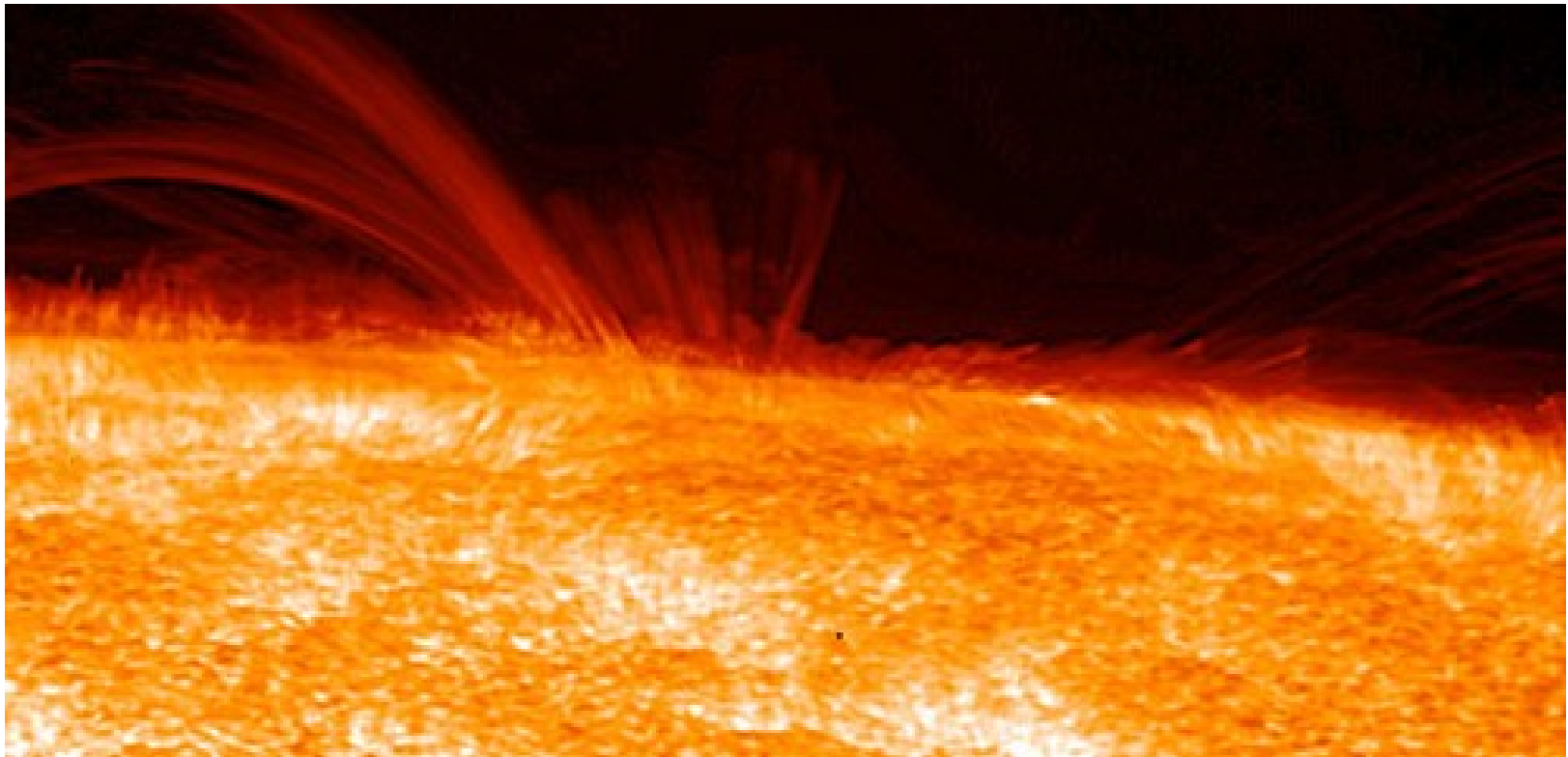
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J. Štěpán (*AI, ASCR, Czech Republic*)

- Introduction
- Computational challenges
- Solution strategy
- First tests

Introduction

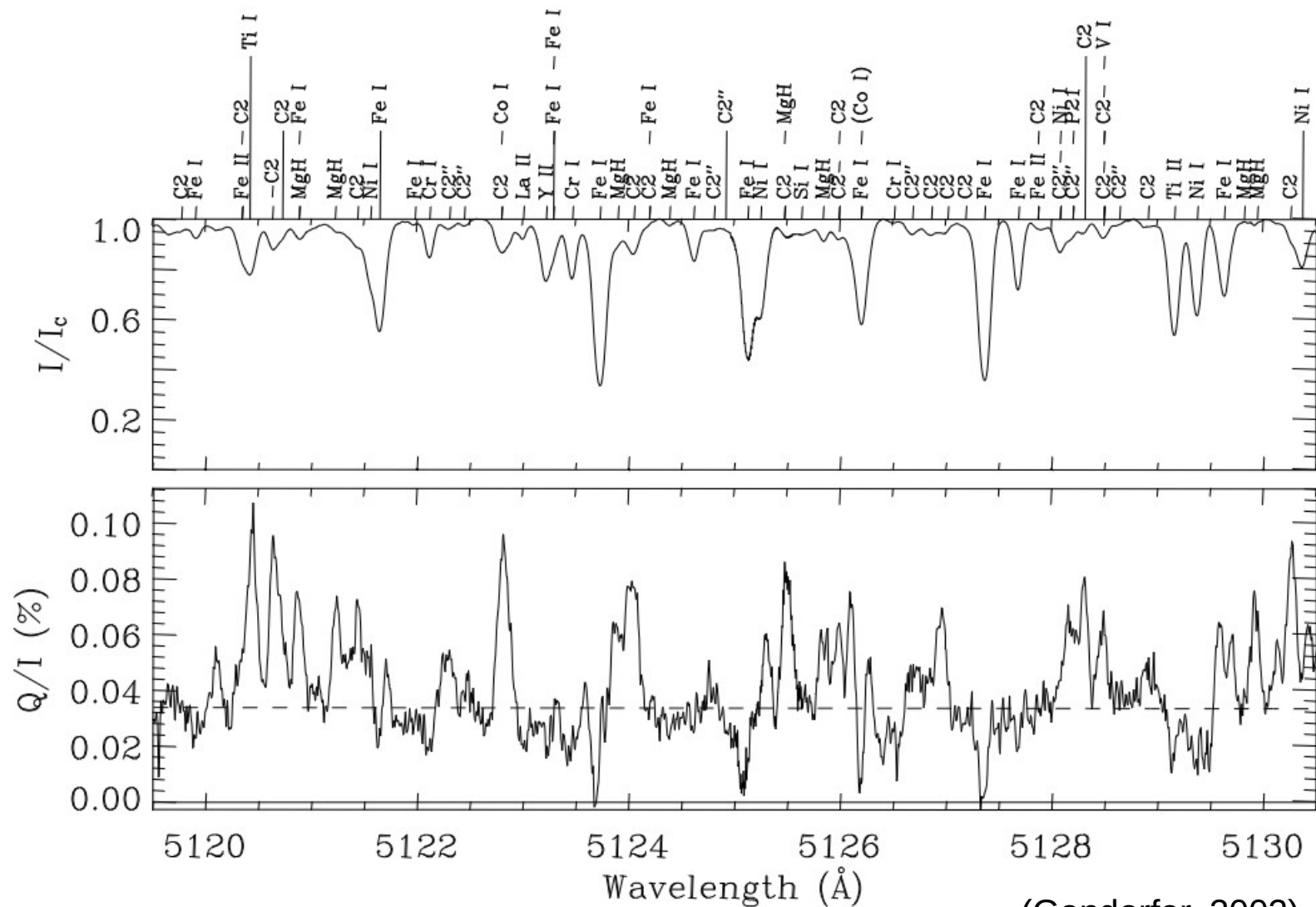
Scientific goal: infer new information on chromospheric magnetic fields



Hinode JAXA/NASA

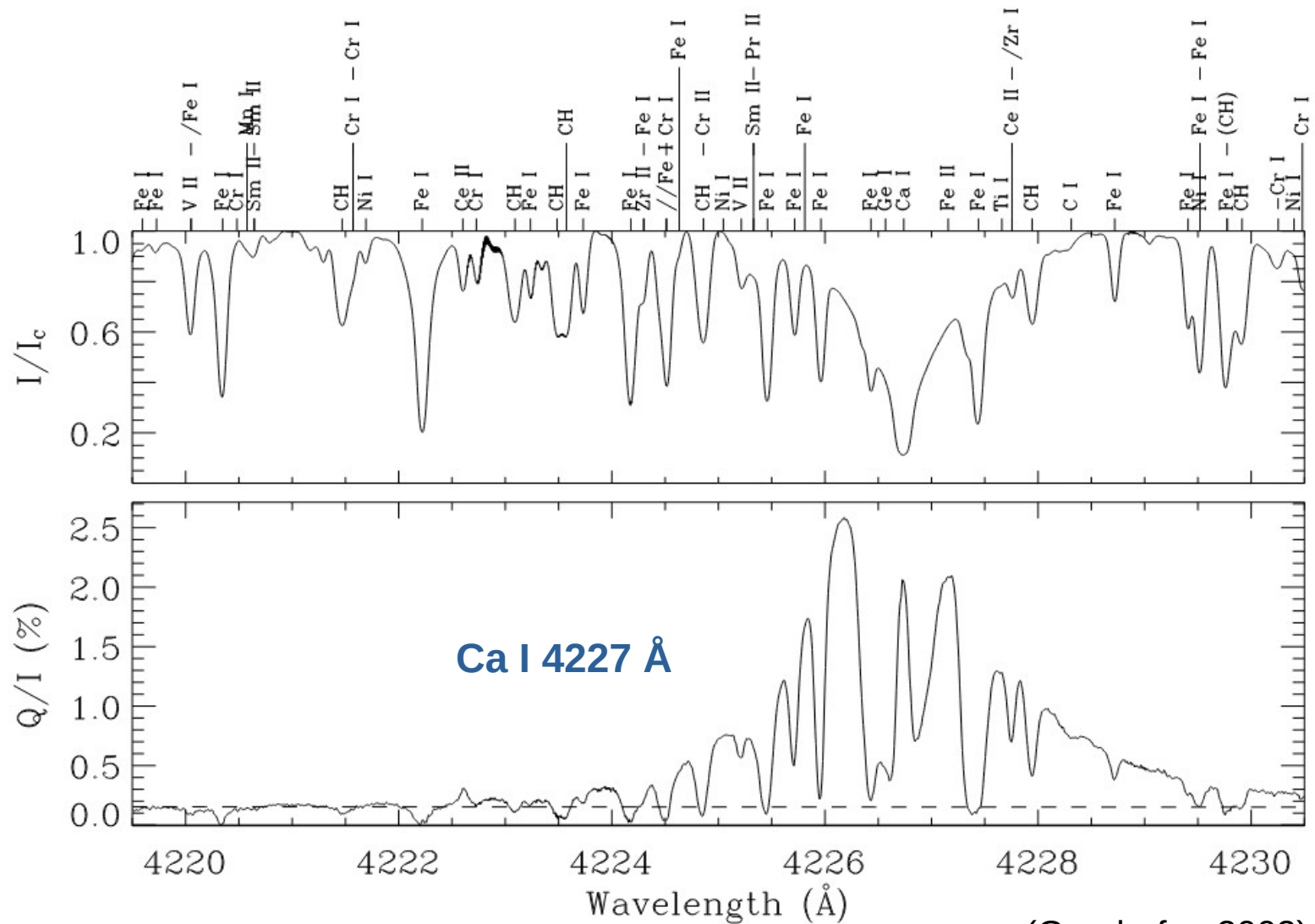
Introduction

Zeeman effect + magnetic sensitivity of **scattering polarization**



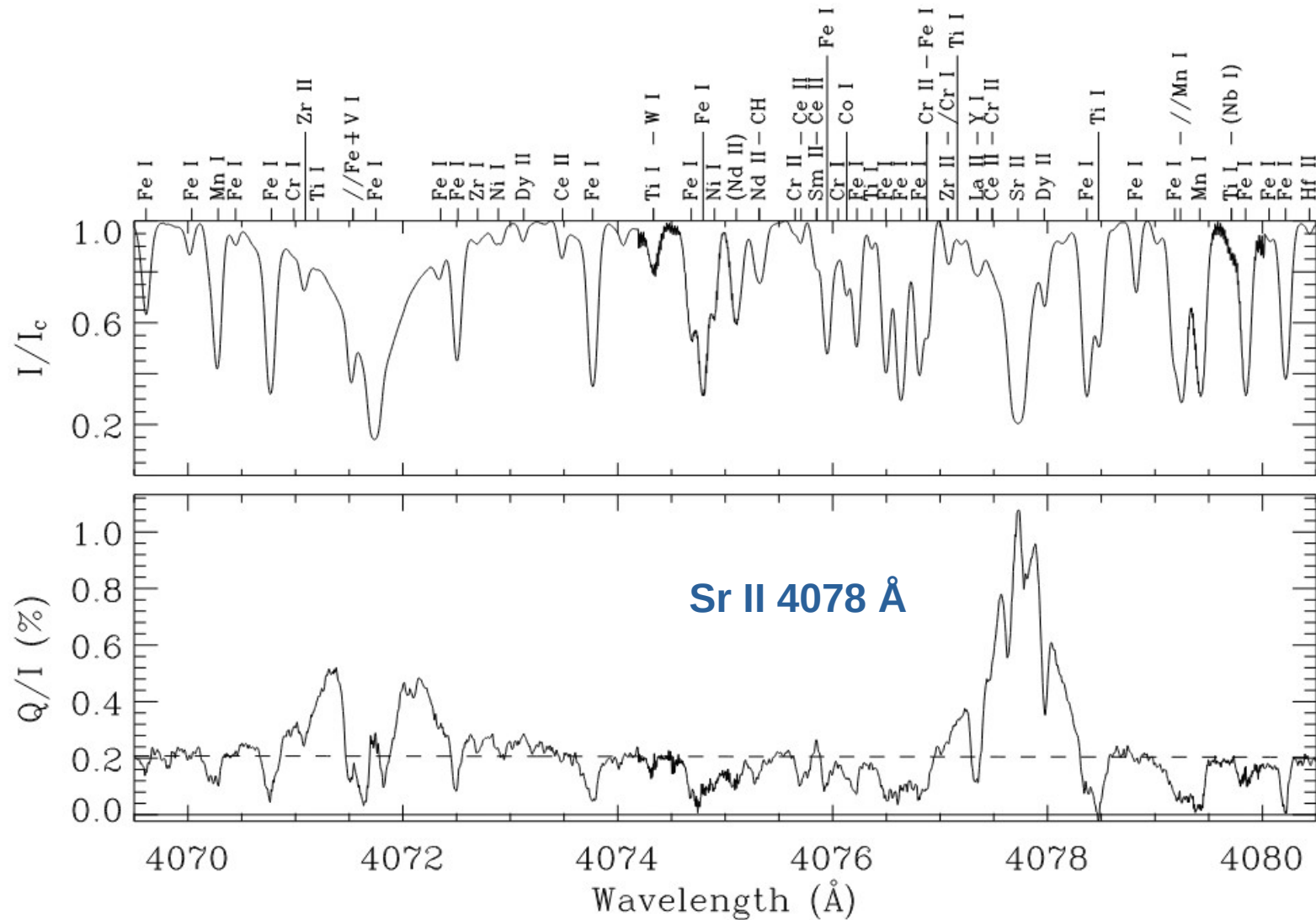
Introduction

Zeeman effect + magnetic sensitivity of **scattering polarization**



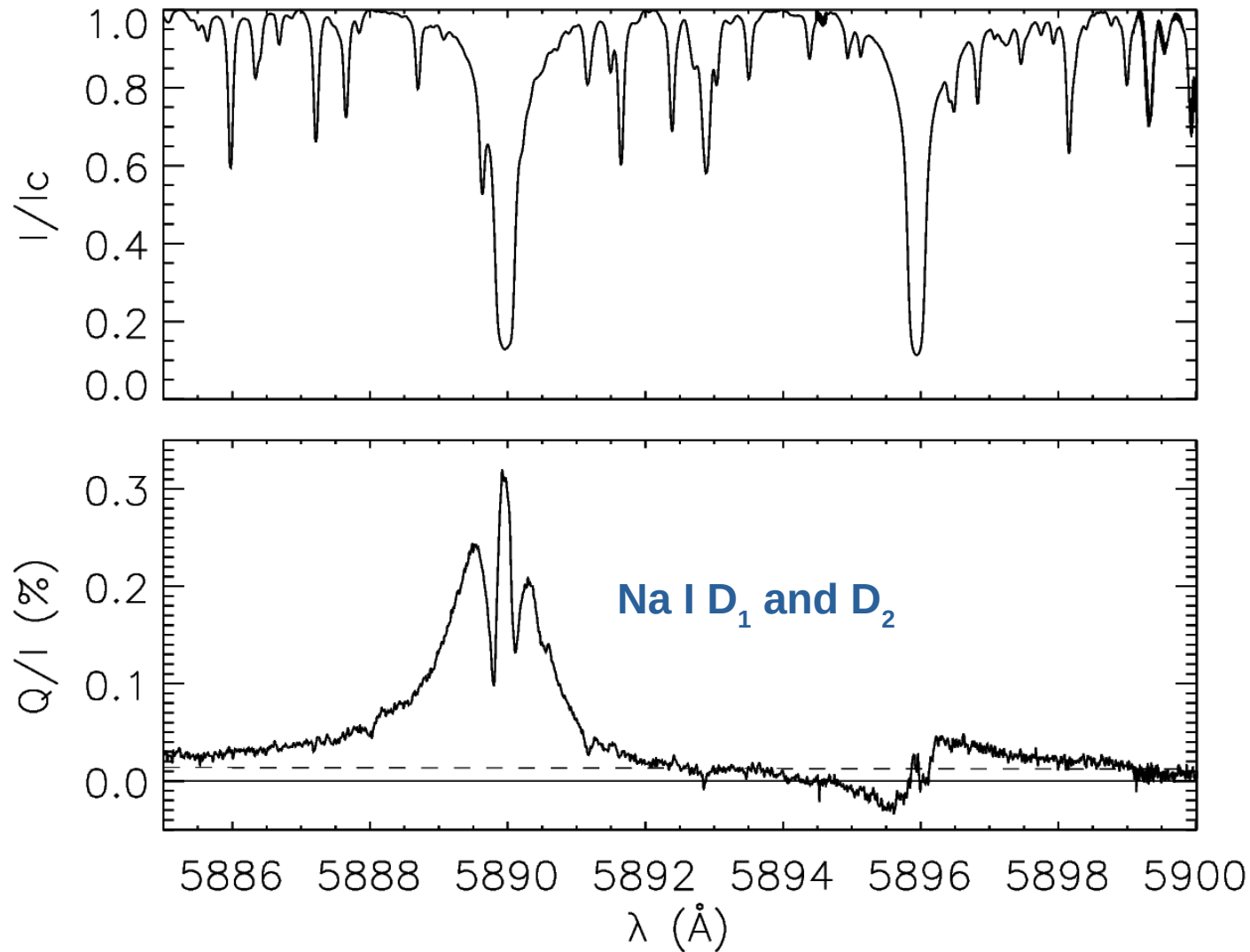
(Gandorfer, 2002)

Zeeman effect + magnetic sensitivity of scattering polarization



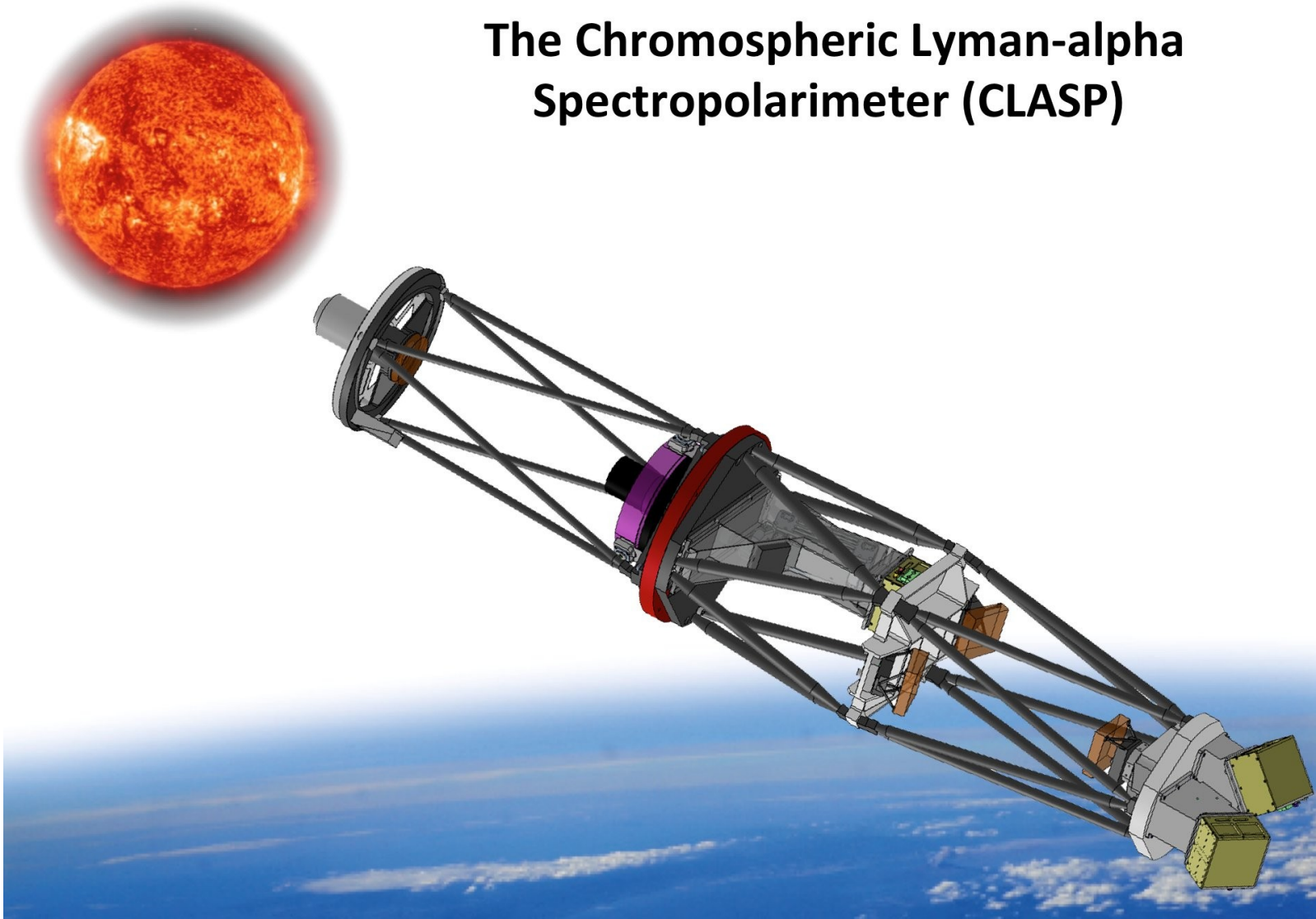
(Gandorfer, 2002)

Zeeman effect + magnetic sensitivity of scattering polarization

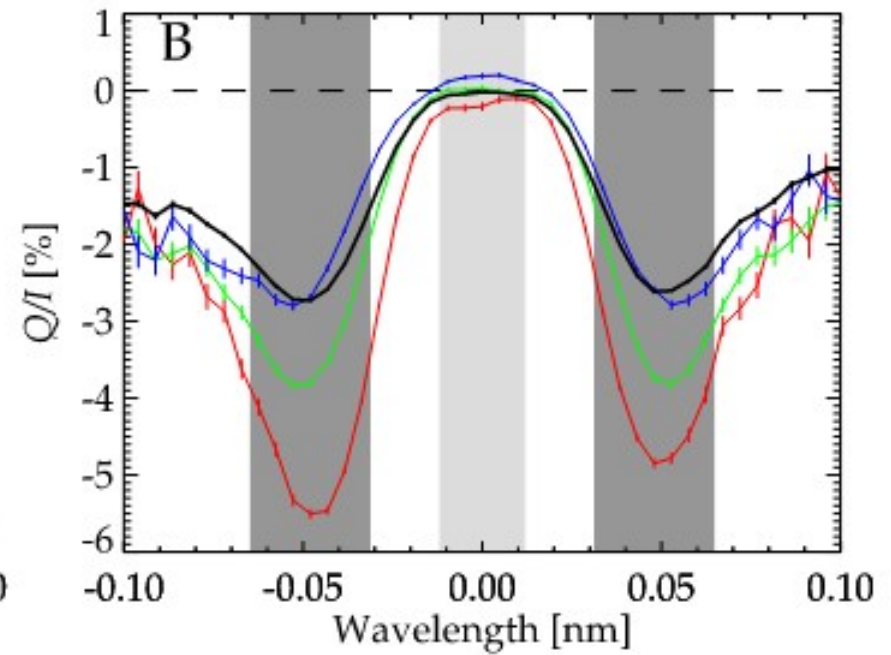
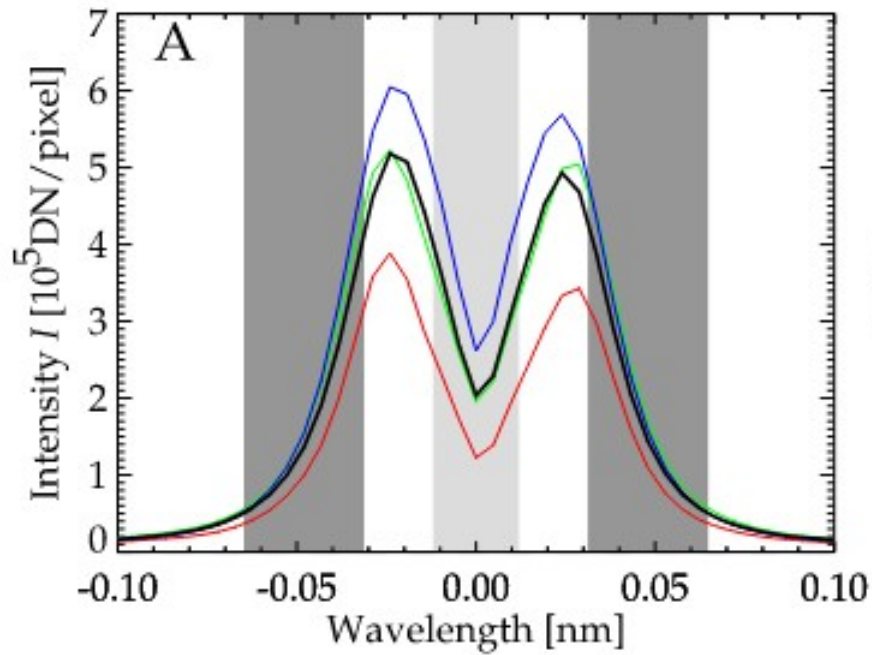


(Gandorfer, 2002)

The Chromospheric Lyman-alpha Spectropolarimeter (CLASP)



H I Ly- α 1215 Å

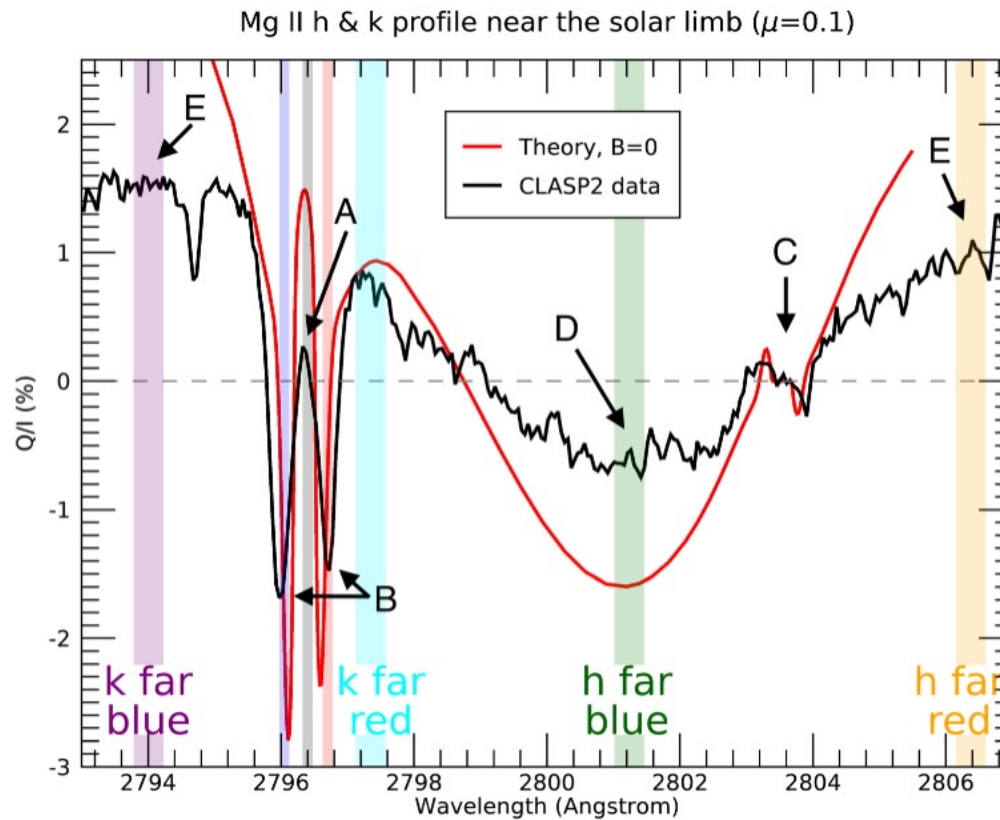


CLASP sounding rocket experiment (2015)

(Kano et al. 2017)

Introduction

Mg II h & k 2800 Å

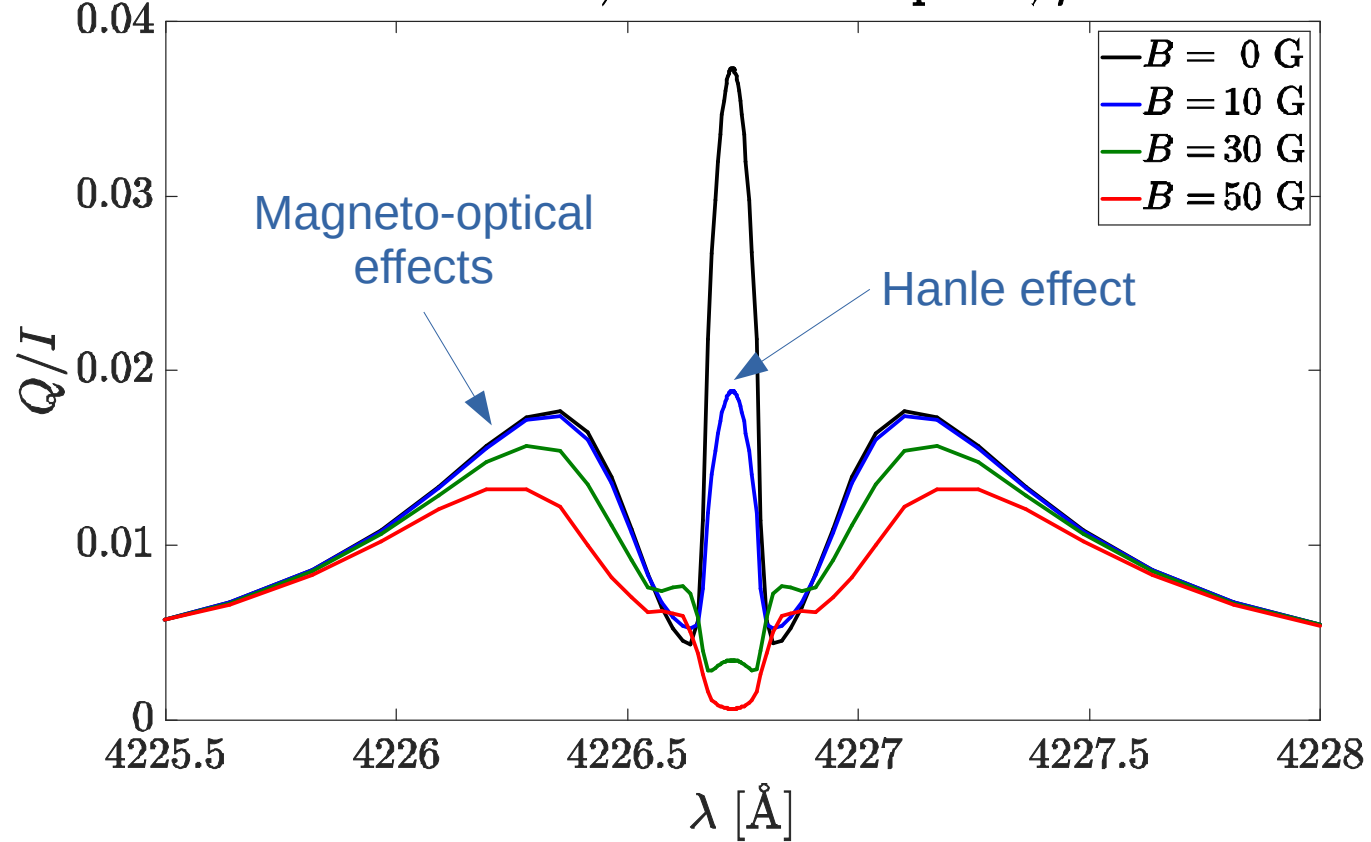


CLASP2 (2019) & CLASP2.1 (2021)

(Rachmeler et al. 2022)

Hanle and magneto-optical effects

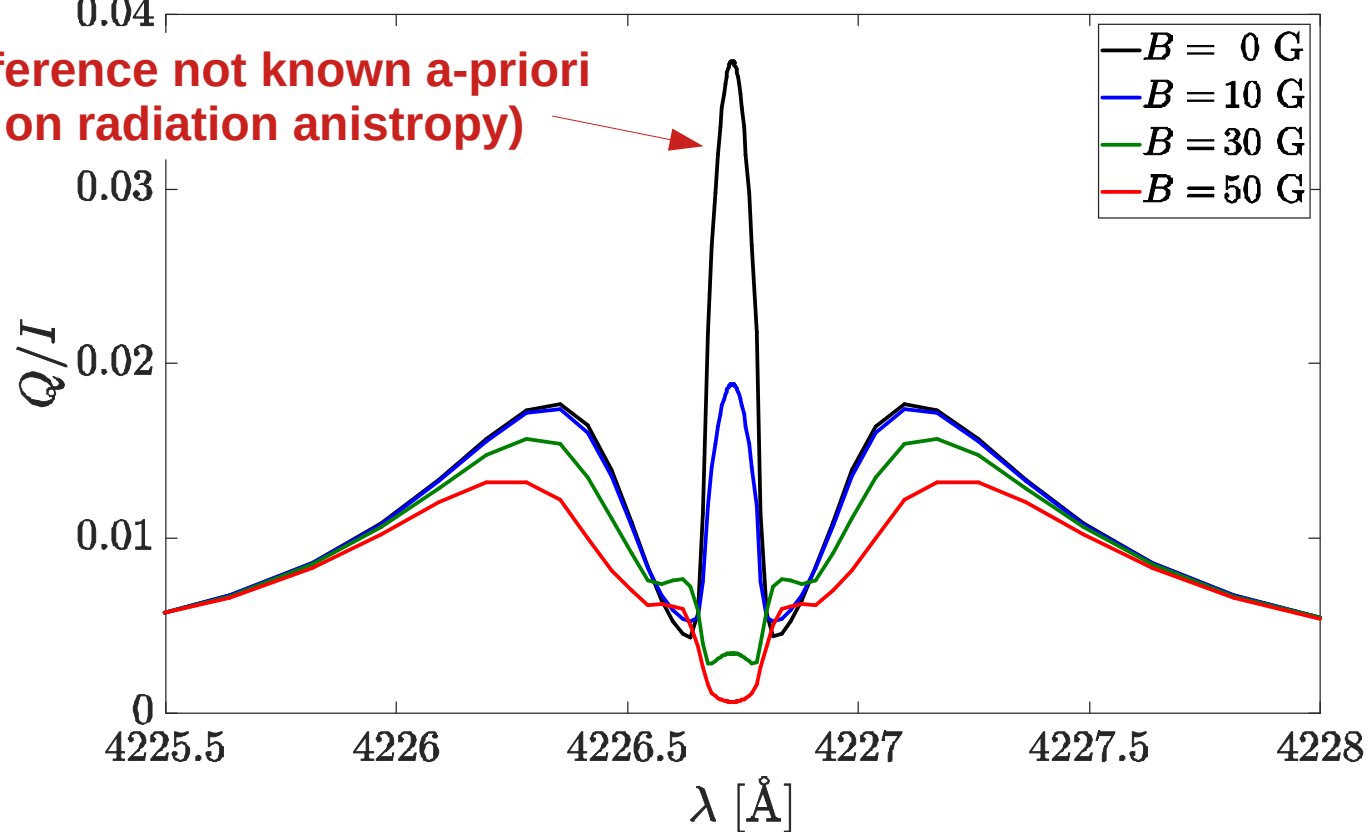
Ca I 4227 Å line, FAL-C atmosphere, $\mu = 0.17$



Hanle and magneto-optical effects

Ca I 4227 Å line, FAL-C atmosphere, $\mu = 0.17$

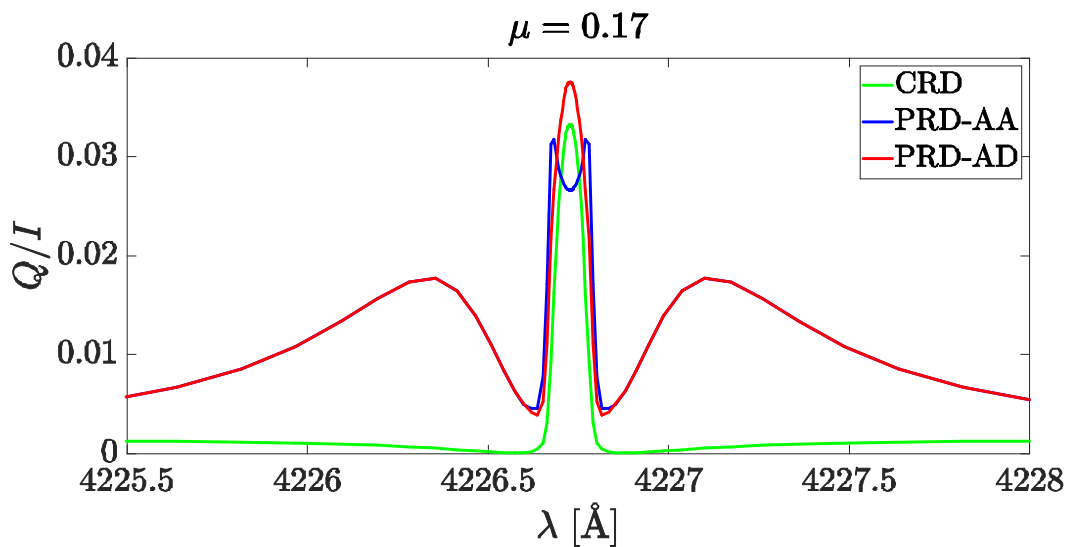
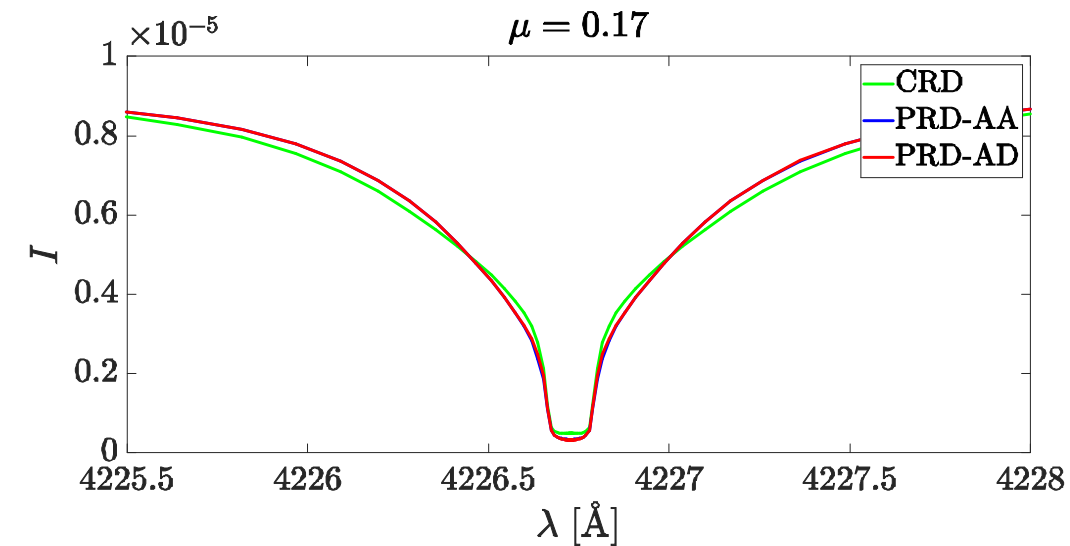
Zero-field reference not known a-priori
(depends on radiation anisotropy)



Approaches:

- Differential Hanle effect (e.g., with molecular lines)
- Synthetic reference from realistic 3D models of the solar atmosphere (e.g., PORTA code by Stepan & Trujillo Bueno 2013)

Importance of PRD effects



Ca I 4227 Å line

Run times

CRD: ~ 10 s

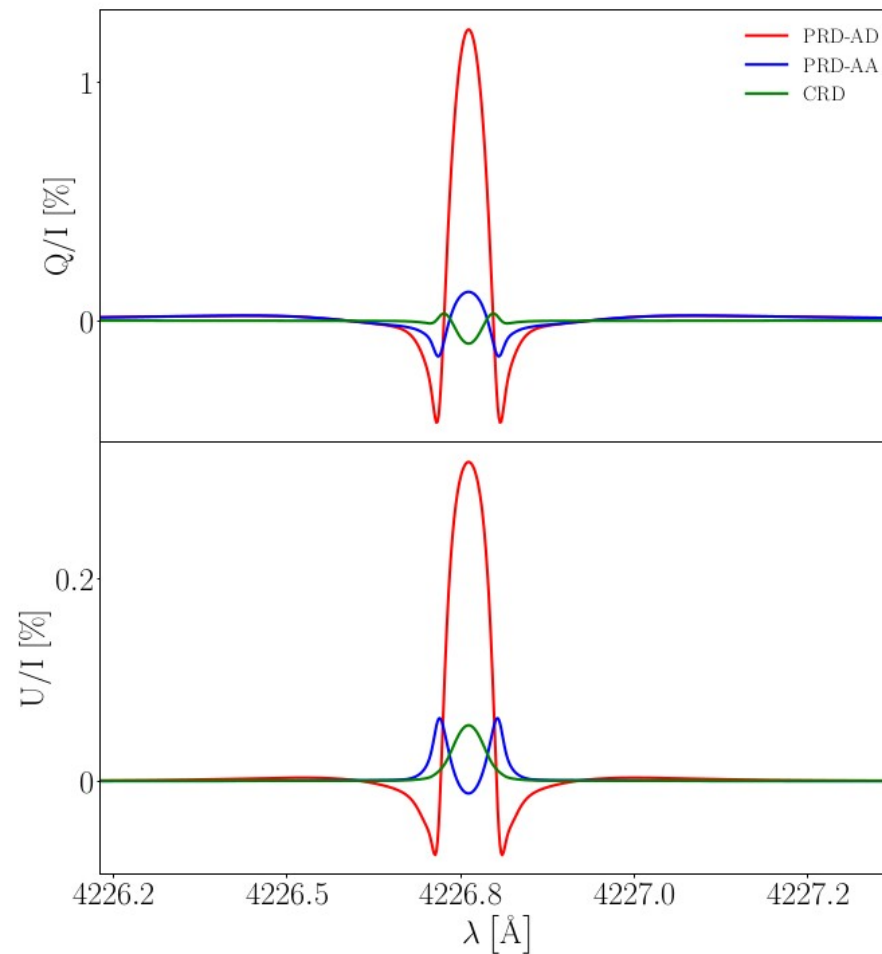
PRD: ~ 1 hour

PRD-AA: ~ 1 min

Non-LTE RT calculations in FAL-C, no magnetic fields

Importance of PRD effects

$\mu = 0.966$, horizontal magnetic field (20 G)
(forward scattering Hanle effect)



(Belluzzi et al., in prep.)

Goal: solve the non-LTE RT problem for polarized radiation in **3D atmospheric** models accounting for **scattering polarization** and **angle-dependent PRD** effects.

Problem computationally challenging already in 1D!

SNSF *Sinergia* project:

“HPC techniques for 3D modeling of resonance line polarization with PRD”

- Physics of polarization
- Numerical radiative transfer
- Computational sciences

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Model problem

- **Two-level atom** with unpolarized and infinitely-sharp lower level
- **PRD theory of Bommier (1997a,b)**
- **Redistribution matrix formalism**

$$R(\mathbf{r}, \Omega', \Omega, \nu', \nu) = R^{\text{II}}(\mathbf{r}, \Omega', \Omega, \nu', \nu) + R^{\text{III}}(\mathbf{r}, \Omega', \Omega, \nu', \nu)$$

$R^{\text{II}}(\mathbf{r}, \Omega', \Omega, \nu', \nu)$: **exact angle-dependent** expression in the observer's frame

$R^{\text{III}}(\mathbf{r}, \Omega', \Omega, \nu', \nu)$: **assumption of totally incoherent scattering** in the observer's frame

Arbitrary magnetic fields accounted for

Computational challenges

Memory requirements

Need to store the full radiation field: $I(\mathbf{r}, \Omega, \nu)$

Problem discretization

$\mathbf{r} \rightarrow N_p \sim 500^3$ (3D)	\longrightarrow	40 TB	(CRD: ~ 10 GB)
$\Omega \rightarrow N_\Omega \sim 100$			
$\nu \rightarrow N_\nu \sim 100$			

Solution method

Iterative solution methods successfully applied in CRD or PRD-AA
cannot be easily generalized to PRD-AD (not efficient or too expensive)

Evaluation of scattering integral

$$\varepsilon^{\text{sc}}(\mathbf{r}, \Omega, \nu) = k_L(\mathbf{r}) \int d\nu' \oint \frac{d\Omega'}{4\pi} R(\mathbf{r}, \Omega', \Omega, \nu', \nu) I(\mathbf{r}, \Omega', \nu')$$

Computational complexity in the AD case: $O(N_p N_\Omega^2 N_\nu^2)$

Algebraic formulation

Continuous problem

$$\boldsymbol{\Omega} \cdot \nabla I(\mathbf{r}, \boldsymbol{\Omega}, \nu) = -K(\mathbf{r}, \boldsymbol{\Omega}, \nu) I(\mathbf{r}, \boldsymbol{\Omega}, \nu) + \epsilon(\mathbf{r}, \boldsymbol{\Omega}, \nu)$$

$$\epsilon(\mathbf{r}, \boldsymbol{\Omega}, \nu) = k_L(\mathbf{r}) \int d\nu' \oint \frac{d\boldsymbol{\Omega}'}{4\pi} R(\mathbf{r}, \boldsymbol{\Omega}', \boldsymbol{\Omega}, \nu', \nu) I(\mathbf{r}, \boldsymbol{\Omega}', \nu') + \epsilon^{\text{th}}(\mathbf{r}, \boldsymbol{\Omega}, \nu)$$

Discretized problem

$$\mathbf{I} = \Lambda \boldsymbol{\epsilon} + \mathbf{t}$$

$\Lambda : \mathbb{R}^N \rightarrow \mathbb{R}^N$ Transfer operator (encodes formal solver and K)

$$\boldsymbol{\epsilon} = \Sigma \mathbf{I} + \boldsymbol{\epsilon}^{\text{th}}$$

$\Sigma : \mathbb{R}^N \rightarrow \mathbb{R}^N$ Scattering operator (encodes quadratures)

$$N = 4 N_p N_\Omega N_\nu$$

$$(\text{Id} - \Lambda \Sigma) \mathbf{I} = \mathbf{b}$$

$$\mathbf{b} = \Lambda \boldsymbol{\epsilon}^{\text{th}} + \mathbf{t} \in \mathbb{R}^N$$

References

- Janett et al. 2021, A&A, 655, 87
- Benedusi et al. 2022, A&A, 664, 197

Main assumption: lower level population known a-priori and kept fixed

(e.g., pre-calculated with an independent RT code, possibly neglecting polarization)

Pros:

- $(Id - \Lambda\Sigma) \mathbf{I} = \mathbf{b}$ linear system \longrightarrow Linearization
- Accurate estimate of lower-level population (e.g., from multi-level atomic models)

Cons:

- Neglect feedback of polarization on lower level population

References

- Janett et al. 2021, A&A, 655, 87

$$(Id - \Lambda\Sigma) \mathbf{I} = \mathbf{b}$$

Preconditioned matrix-free GMRES Krylov solver

(e.g., Nagendra 2009, Anusha & Nagendra 2011, 2012, 2013; Sampurna 2019)

Preconditioning: $P^{-1} (Id - \Lambda\Sigma) \mathbf{I} = P^{-1} \mathbf{b}$

Methods successfully applied to CRD and PRD-AA (Jacobi, block-Jacobi, Gauss-Siedel, SOR)
cannot be easily generalized to PRD-AD case (P^{-1} not cheap, or not effective)

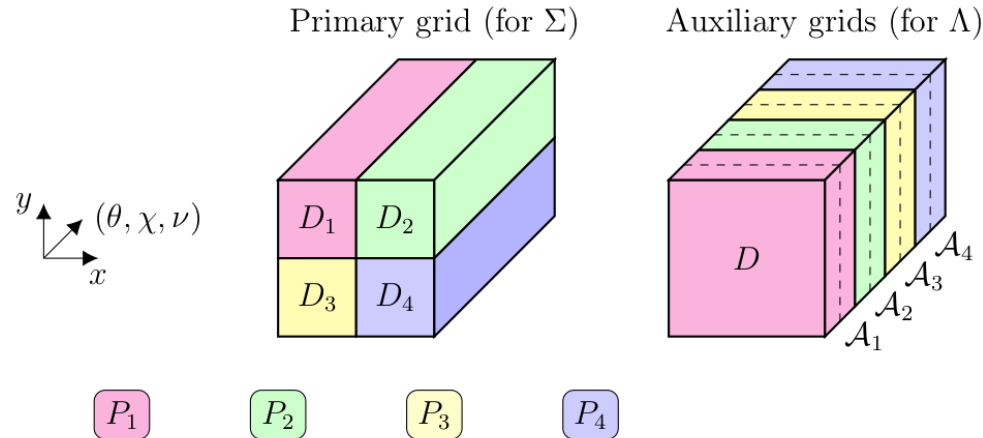
Physics-based preconditioning: $P = Id - \Lambda\Sigma^{\text{CRD}}$

References

- Janett et al. 2021, A&A, 655, 87
- Benedusi et al. 2021, A&A, 655, 88
- Benedusi et al. 2022, A&A, 664, 197

Parallel implementation

Primary – auxiliary grid approach



(Benedusi et al. 2022, JCP, 479, 112013)

Primary grid:
space-distributed

**All-to-all communication
between grids**

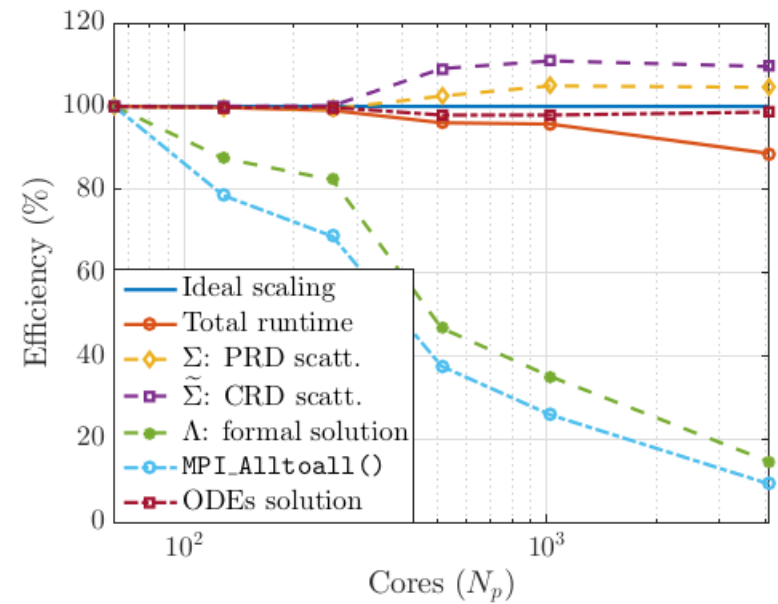
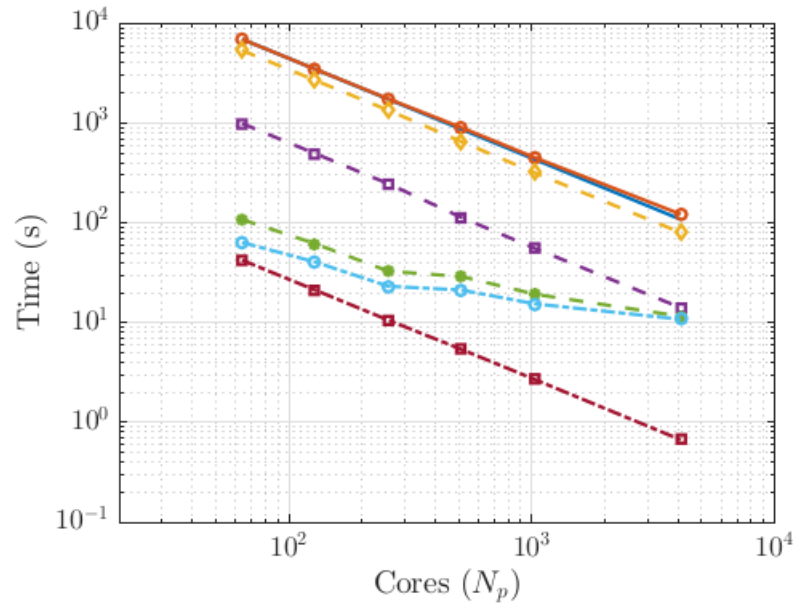
Auxiliary grid:
ray-distributed

$\epsilon^{\text{sc}} = \Sigma \mathbf{I}$ Parallel calculation at
different spatial points

$\mathbf{I} = \Lambda \epsilon$ Parallel calculation along
different “rays” (Ω_l, ν_m)

Scaling experiments

Strong scaling experiment



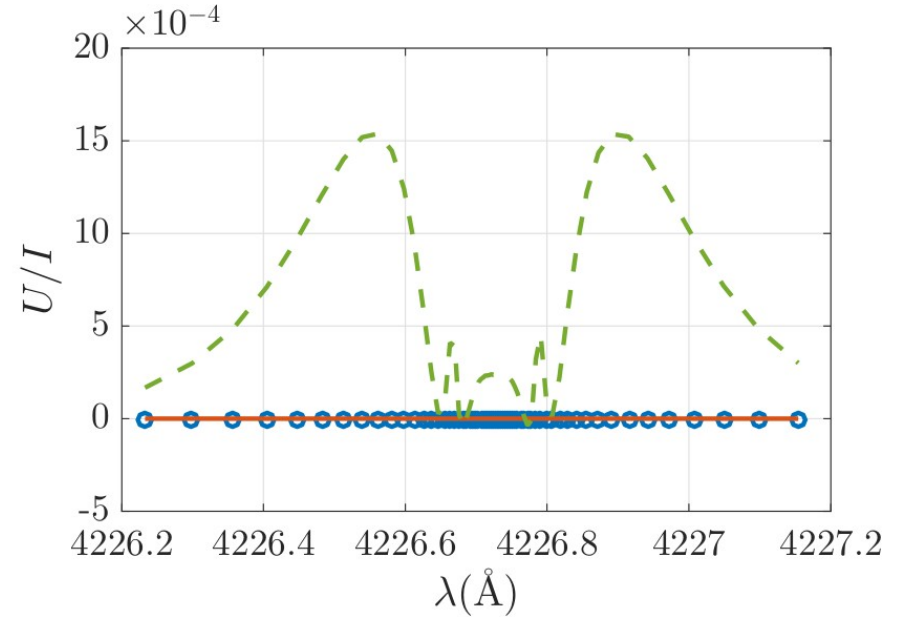
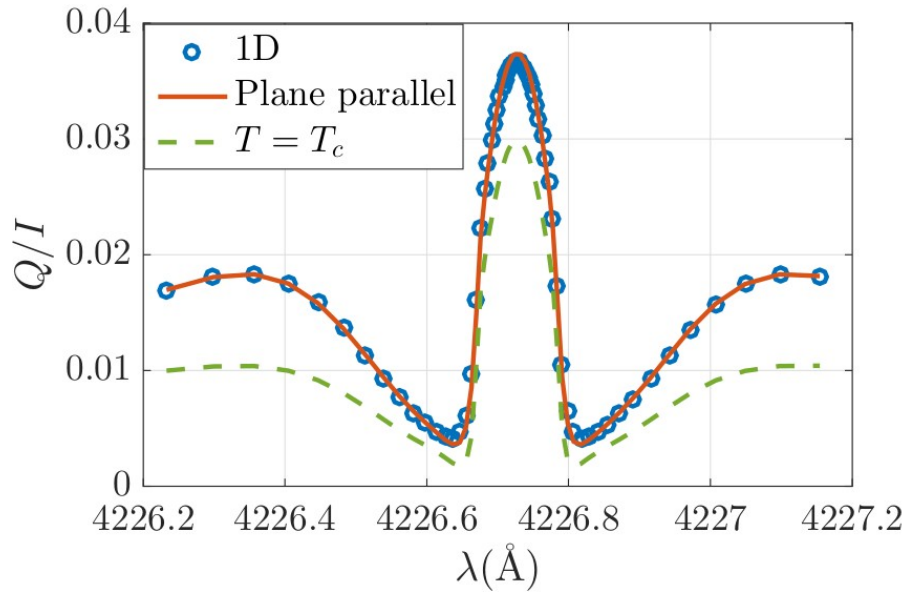
$N_x = 8, N_y = 8, N_z = 64, N_\Omega = 64, N_v = 64$

4 Preconditioned GMRES iterations to converge

(Benedusi et al. 2023, JCP, 479, 112013)

Results

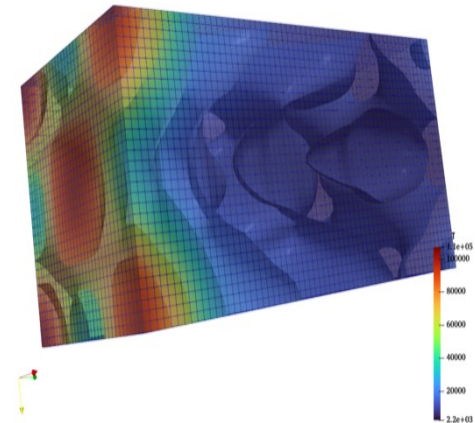
$N_x = 8, N_y = 8, N_z = 64, N_\Omega = 64, N_v = 64$ LOS: $\mu=0.1, \chi=\pi/4$



(Benedusi et al. 2023, JCP, 479, 112013)

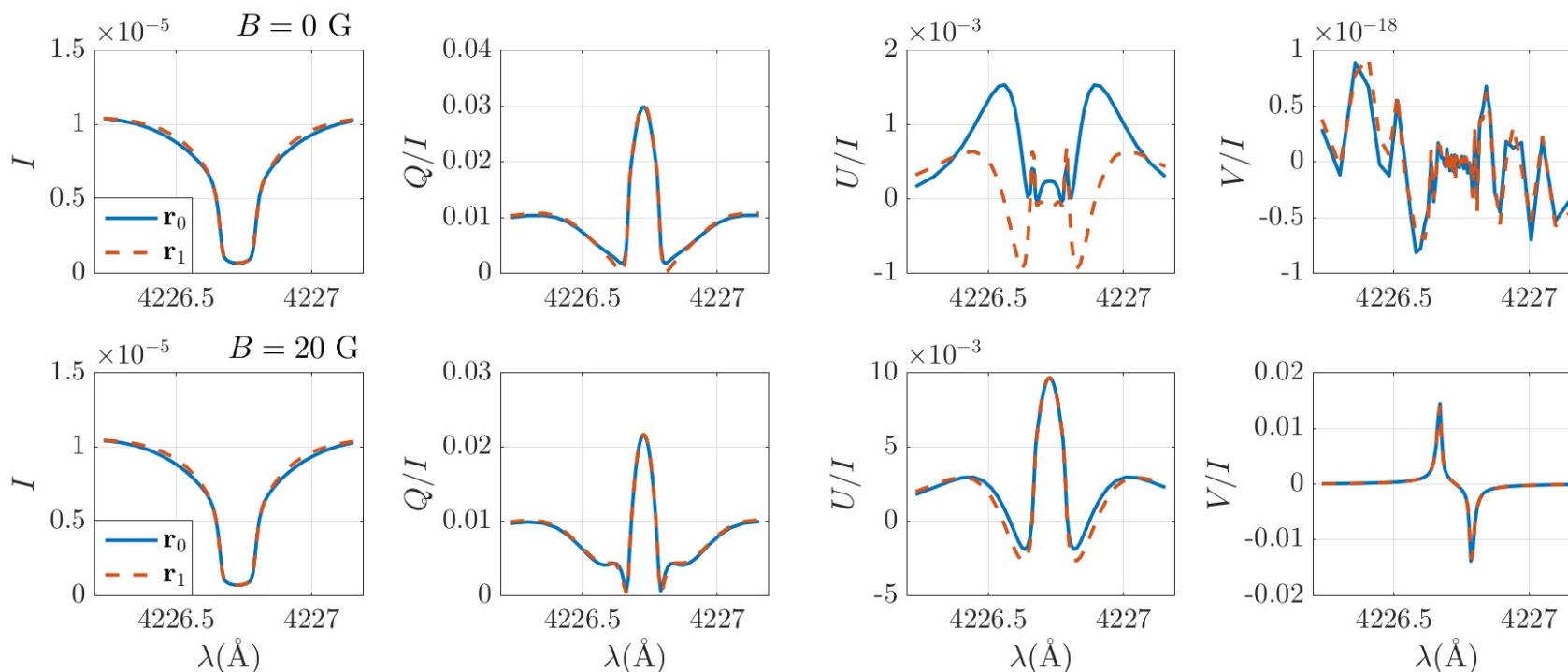
- FAL-C (1D)
- FAL-C (3D)
- FAL-C (3D) + sinusoidal temperature variation

4 Preconditioned GMRES iterations to converge
Run time: 122s (4096 cores)



Results

$N_x = 8, N_y = 8, N_z = 64, N_\Omega = 64, N_v = 64$ LOS: $\mu=0.1, \chi=\pi/4$

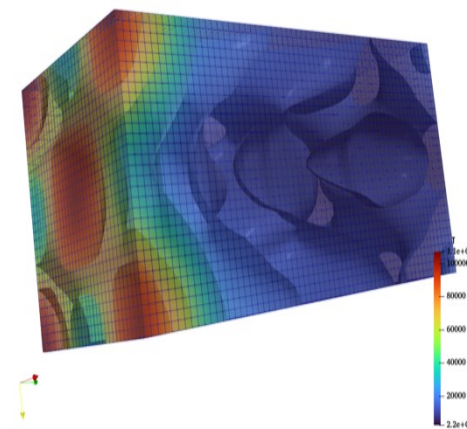


(Benedusi et al. 2023, JCP, 479, 112013)

FAL-C (3D) + sinusoidal temperature variation + **magnetic field**

4 Preconditioned GMRES iterations to converge

Run time: 122s (4096 cores)



Work in progress

- **3D code testing and applications**
- **Generalization to two-term model atom** (Mg II h & k, H I Ly- α , Na I D; Janett et al. in prep.)
- Investigation on the impact of bulk velocities (**Guerreiro et al. submitted**)
- Assessment of CRD approximation for R_{III} (**Riva et al. submitted**)
- Investigation of new preconditioners (**Janett et al. in prep.**)
- 1D modeling of other spectral lines (**Riva et al. in prep.**)

Future

- **Generalization of solution strategy to nonlinear problem**
- New strategies for scattering integral calculation (GPUs, machine learning)

References:

- Janett et al. 2021, A&A, 655, A13
- Janett et al. 2021, A&A, 655, A87
- Benedusi et al. 2021, A&A, 655, A88
- Benedusi et al. 2022, A&A, 664, A197
- Benedusi et al. 2023, JCP, 479, 112013

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