Coronal Prediction for the 2024 Total Solar Eclipse: Boundary Conditions

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PSI and Eclipse Predictions

- The team at PSI has been posting eclipse predictions for decades
- They were made using quasi-relaxed MHD simulations of the solar corona, posted ~2 weeks before the event
- Comparing the predictions to observations helps refine the model, and can provide context for observational planning
 This year, we are using a new paradigm: a continually running, data assimilative MHD simulation in real time



The June 21, 2001 Solar Eclipse



The August 11, 1999 Solar Eclips



The February 26, 1998 Solar Eclipse



The March 9, 1997 Solar Eclips



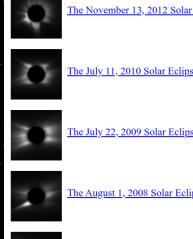
<u>The October 24, 1995 Solar Eclips</u>



<u> he November 3, 1994 Solar Eclipse</u>

predsci.com/

eclipses

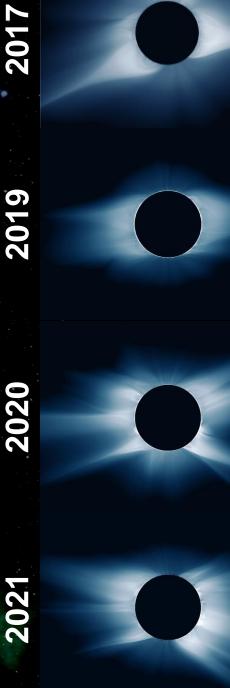


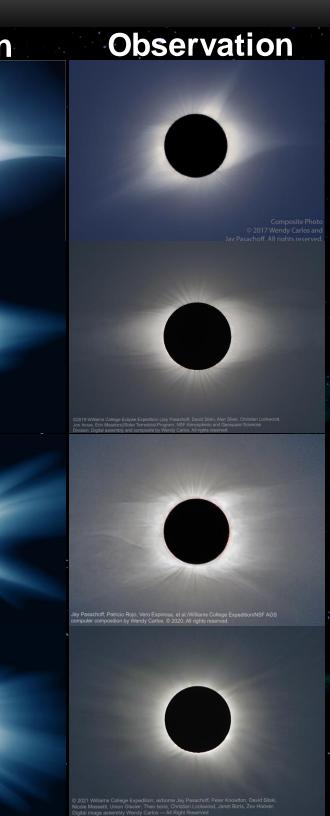




e December 4, 2002 Solar Eclipse

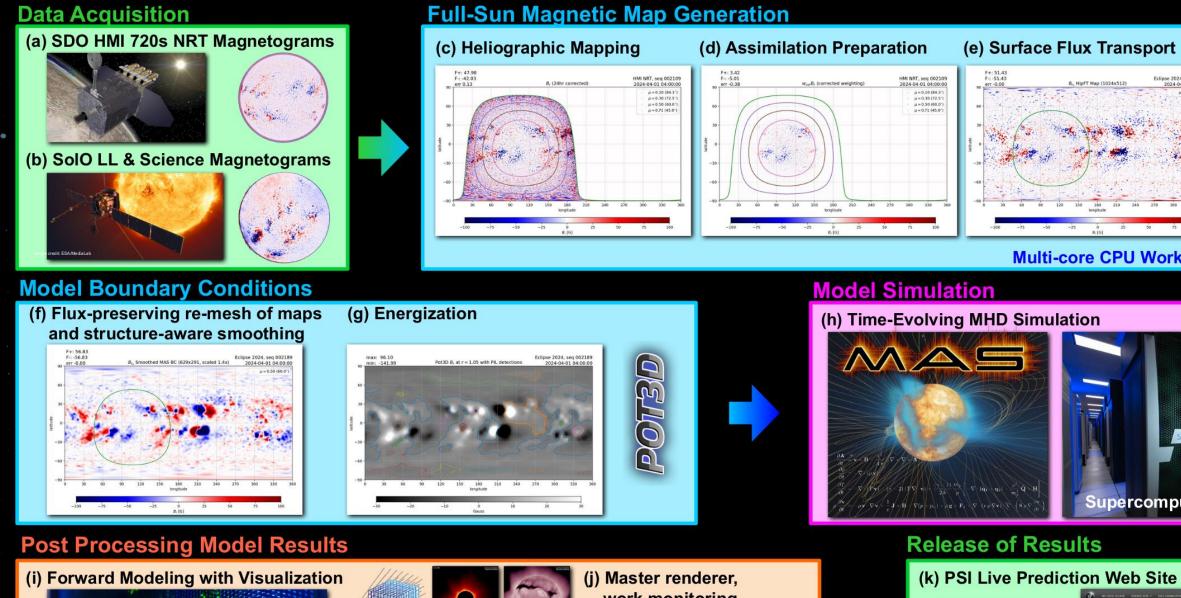
Prediction



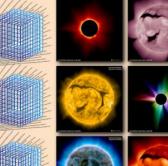




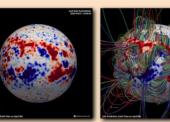
The PSI 2024 Eclipse Prediction Real-Time Pipeline





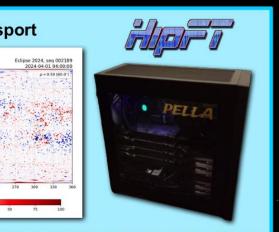


work monitoring, additional visualizations



(k) PSI Live Prediction Web Site





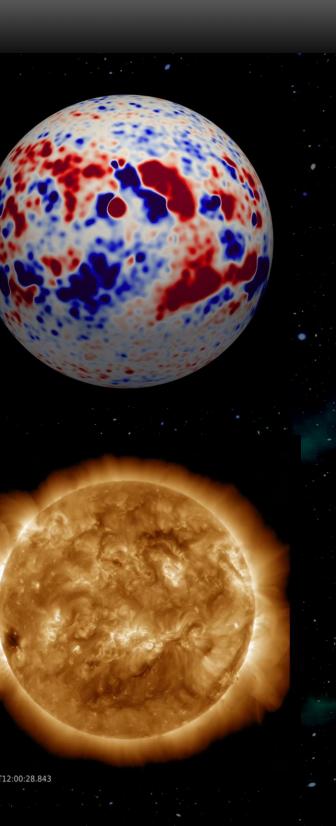
Multi-core CPU Workstation with Multiple GPUs







- MHD (and other) global models require solar surface magnetic field data as input boundary conditions
- While observed by multiple instruments, routinely only from the Sun-Earth line of sight
- In order to make a global map, old data from the Sun-Earth line can be used (e.g. Carrington/"synoptic" maps)
- The older data in the maps makes predicting the eclipse corona difficult, especially during solar maximum when the Sun is changing rapidly
- A way to mitigate this problem is to run a data-assimilative surface flux transport model (SFT) that models the Sun's surface flows to transport the field
 - Although the SFT model misses new far-side flux emergence, it can accurately predict how the most recently assimilated data will change over time on the back of the Sun



Open Source Flux Transport

Space Weather with Quantified Uncertainty

 Developed as part of the "Improving Space Weather Predictions with Data-Driven Models of the Solar Atmosphere and Inner Heliosphere" SWQU project

github.com/ predsci/hipft



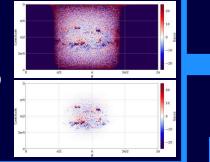
Open Source Flux Transport A Tool for Generating Full-Sun Synchronic Magnetic Field Maps

MagMAP

Magnetic Mapping And Processing

Acquire, process, map, and bin observational magnetic field data into data-assimilation database

ConFlow



Convective Flow generator

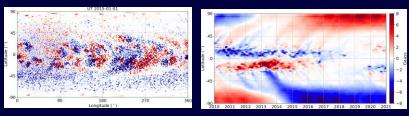
Generate sequence of supergranular flow fields

-90 0	90	180 Longitude (*)	270	360 -1000
				-750
-45				-500
1.				-250
0				0
1				250
45				500
5.00	S. 1. 1. 1.	1. A. S. S.		750
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High-performance Flux Transport

Integrate surface flux transport model with differential rotation, meridional flows, diffusion, and flux sources using high-order numerical schemes and CPU/GPU parallelism over multiple realizations



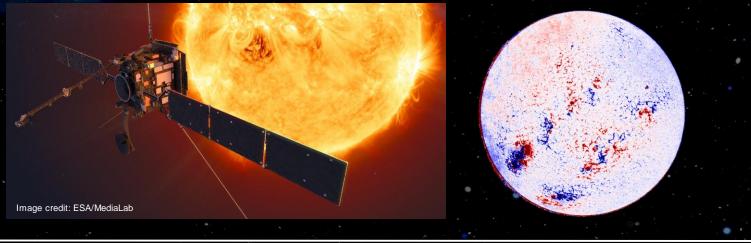
etic Field Maps

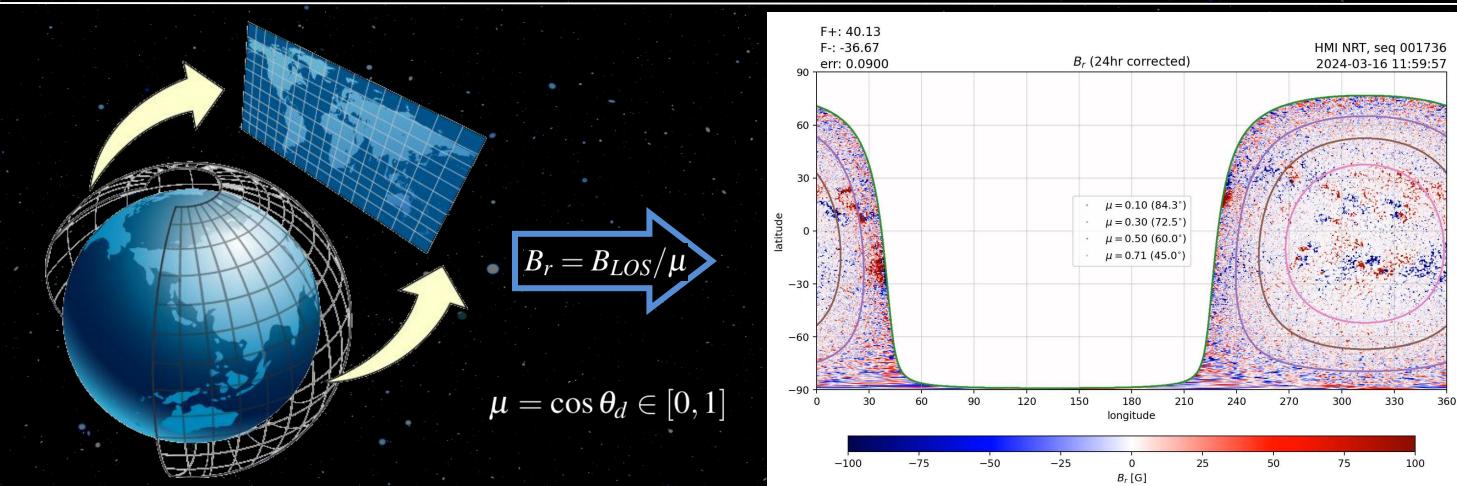
Data Acquisition and Heliographic Mapping

SDO HMI 720s NRT Magnetograms

SO PHI LL & Science Magnetograms







25	50	75	1



Data Assimilation

- Data is assimilated based on a weighting between the new data and the pre-existing model
- For each disk magnetogram, the center-to-limb angle is used to create the assimilation weights
- In order to avoid issues such as strong half-active regions, the weights are modified so that data entering the edge of the assimilation window takes longer to emerge
- To avoid flux imbalance, the added change in the field is flux balanced

 $B_r^{new} = F B_{r:data} + (1 - F) B_r^{old}$ $F = \begin{cases} \mu^4, & \mu \ge 0.1 \\ 0, & \mu < 0.1 \end{cases}$

 $B_r^{new} = B_r^{old} + \Delta B_r$

 $\Delta B_r =$

$$\Delta B_r = F(B_r)$$

$$\Delta B_r/\sqrt{|\Phi_+|}$$

$$\Delta B_r \sqrt{|\Phi_+/|}$$

 $r;data - B_r^{old})$

 $/\Phi_{-}$

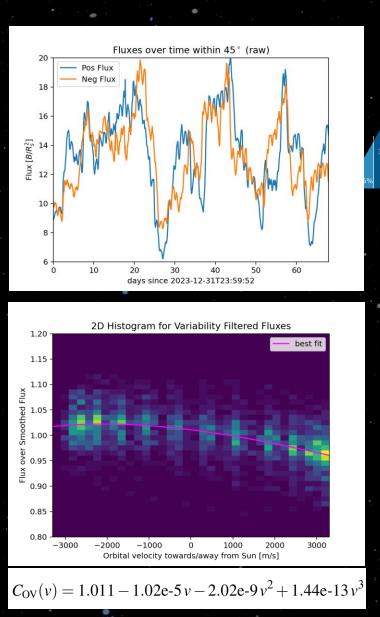
 $|\Phi_{-}|,$

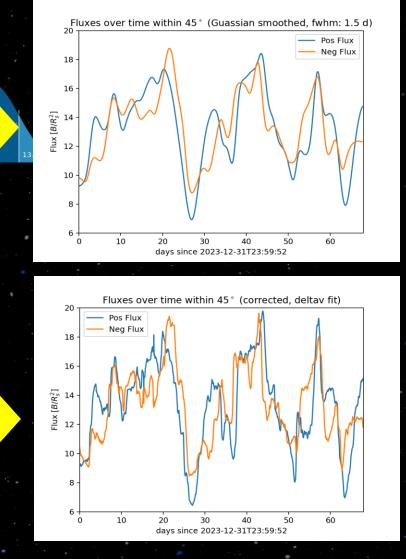
 $\Delta B_r > 0$

 $\overline{\Delta B_r} \leq 0$

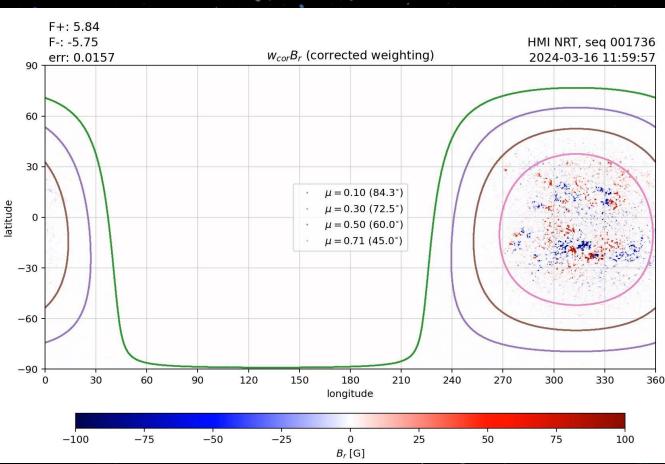
Data Assimilation Cont.

Removing Orbital Variations:





Data and weights stored in a sequence of indexed files ready for use in the SFT model





The High-performance Flux Transport Model

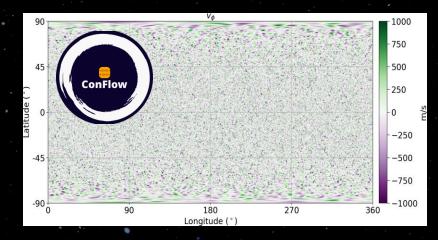
Implements advection, diffusion, data assimilation, and flux emergence over multiple realizations using high-accuracy numerical methods and CPU/GPU parallelism

github.com/ predsci/hipft

$$\frac{\partial B_r}{\partial t} = -\nabla_s \cdot (B_r \mathbf{v}) + \nabla_s \cdot (\nu \nabla_s B_r) + S,$$

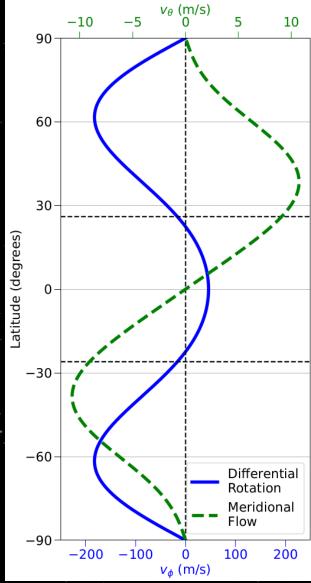
Flows

- Analytic observationally-derived differential rotation and meridional flows
- Time-dependent super-granular convective flows generated by ConFlow
- Flow attenuation based on field strength Diffusion
 - Used to capture flux cancellations at smaller scales than the super granular flows



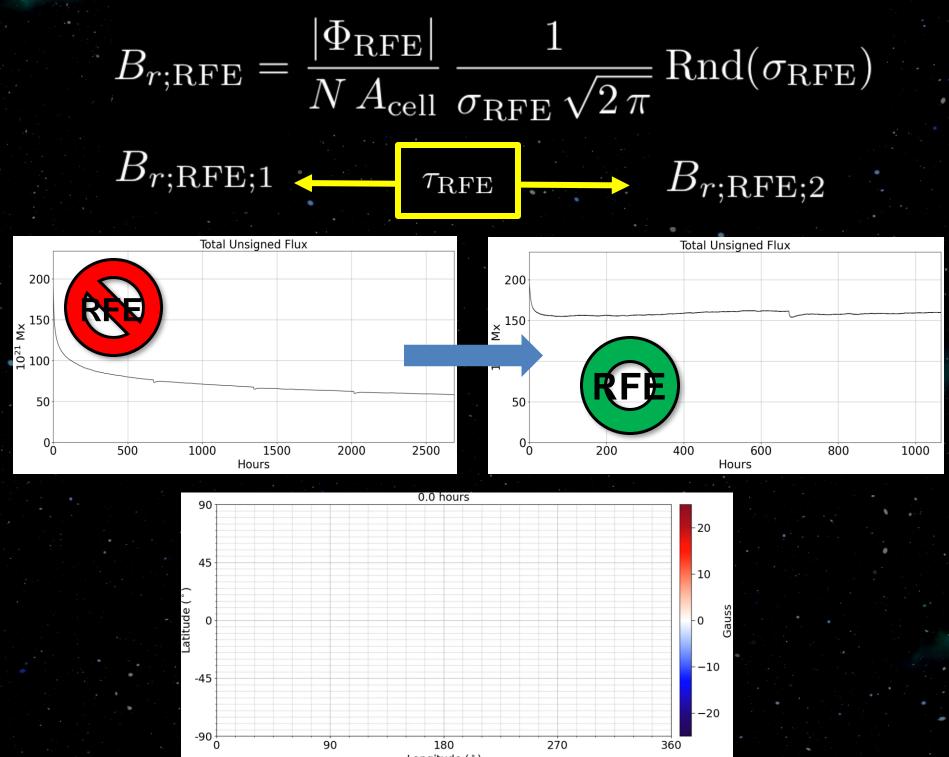
$$v_{\theta/\phi} \to v_{\theta/\phi} \left[1.0 - \tanh\left(\frac{|B_r|}{B_0}\right) \right]$$

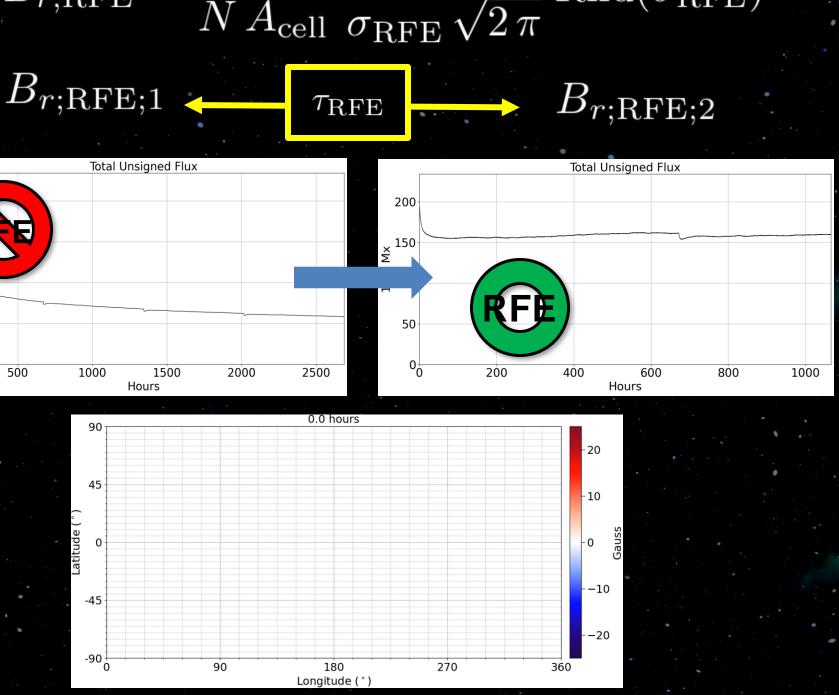
$$\nu = 300 \, \mathrm{km}^2/s$$



Random Flux Emergence (RFE)

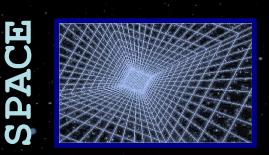
- The flux-canceling processes in SFT models reduces the unsigned flux (UF) compared to that of the assimilated data
- This leads to unrealistic localized \bigcirc low UF regions and a variable average UF away from the assimilation region
- As our MHD model uses UF in \mathbf{O} the heating model, this can adversely affect the simulations
- We therefore add random flux \bigcirc emergence as a source term
- We tune the parameters to yield \bigcirc a constant average UF in the quiet Sun regions calibrated to the current time period and resolution of the model





HipFT Numerical Methods





ADVECTION: 3rd-order WENO3-CS(h) DIFFUSION: 2nd-order CD

DIFFUSION: 2nd-order RKG2(3/2) + PTL

ADVECTION: 3rd-order SSPRK(4,3)

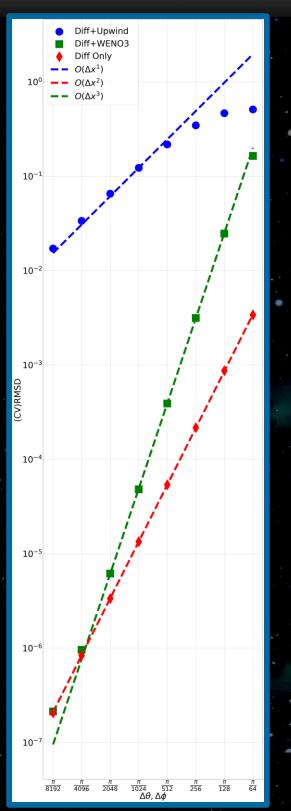


do concurrent (i=1:N,j=1:M) Computation enddo



!\$omp target enter data map(to:a) !\$omp target exit data map(from:a)



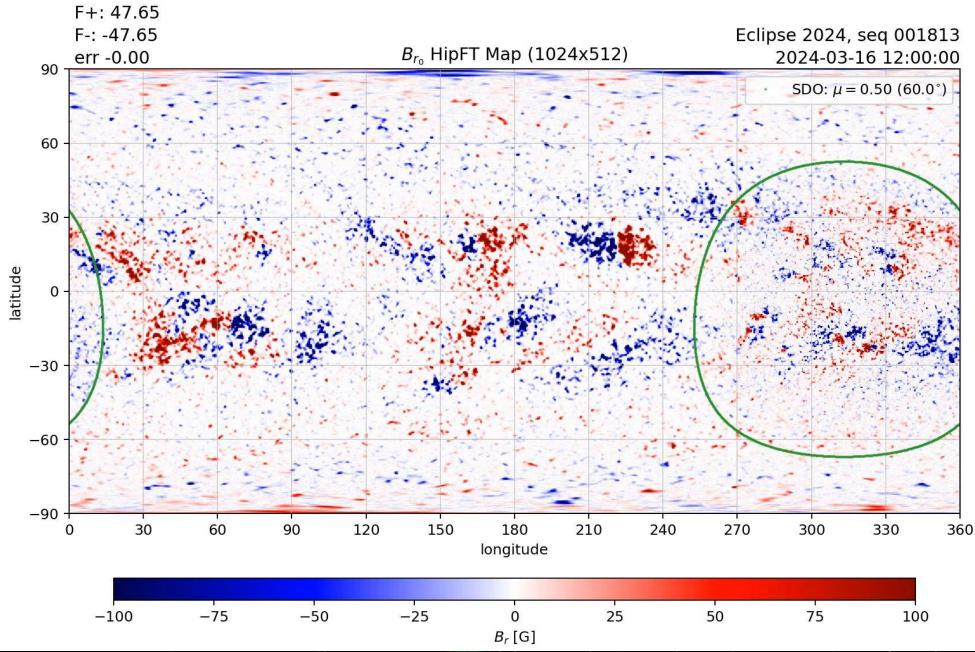




HipFT Eclipse Run



In-house workstation: EPYC 7702P 64-core CPU 4x NVIDIA 2080Ti GPUs



Map Processing for MHD Model

- Scale by 1.4x to make HMI similar to MDI
- Flux-preserving integral re-mesh from HipFT resolution to MHD run resolution
- Smoothing \bigcirc

 $\nu = f_b \nu_{\rm grid}$

- Grid-based diffusivity (resolvability)
- Very strong fields can strain model, so we also smooth such regions locally
- The smoothing mask is made by computing a potential field solution and taking a radial slice at 1.05 Rs
- The diffusivity is modified proportional to 0 the magnitude of B

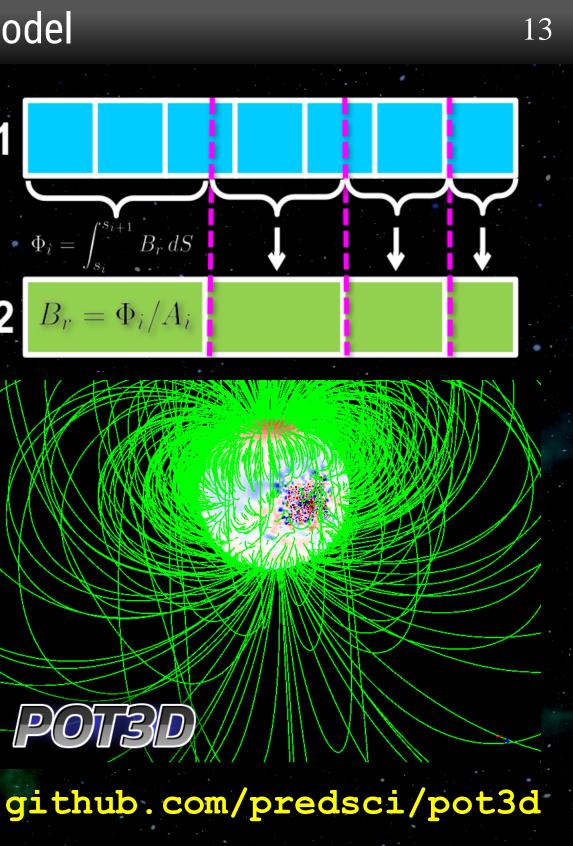
$$f_b = 0.5 + 7.5 \frac{|\vec{B}|}{200 \,\text{Gauss}}$$

$$\nu_{\text{grid}} = (\Delta \theta)^2 + (\Delta \phi \, \sin \theta)^2$$

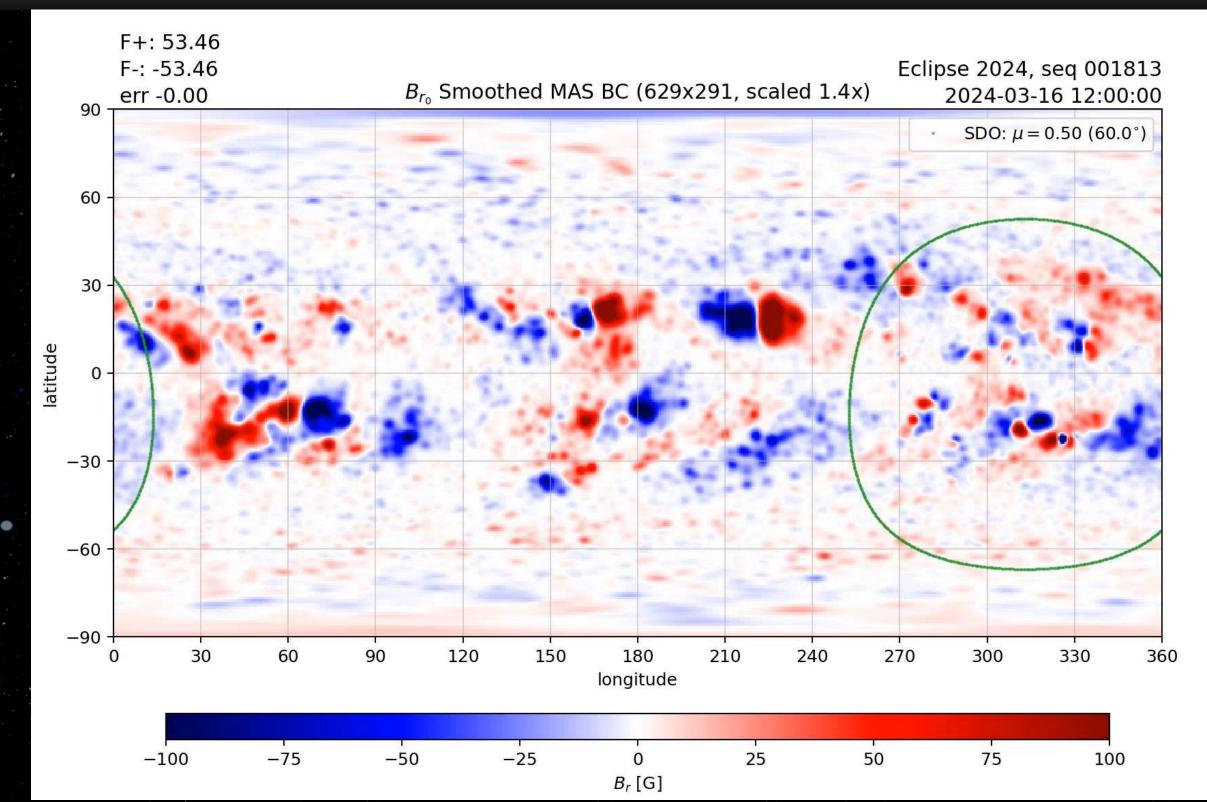
GRID 1

$$\Phi_i = \int_{s_i}^{s_{i+1}} B_r \, dS$$

GRID 2 $B_r = \Phi_i / A_i$



Maps Ready for Time-Dependent MHD Simulation





The PSI 2024 Eclipse Prediction Real-Time Pipeline

