Kamodo Analysis Suite

A Functional API for Space Weather Models and Data

Asher Pembroke Lutz Rastaetter Darren Dezeeuw Katherine Garcia-Sage





SERVING THE SPACE WEATHER COMMUNITY

Mission Statement: To enable, support, and perform research for next generation space science and operational space weather models through an interagency partnership.



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A few of the models hosted at CCMC

SOLAR

PFSS J. LUHMAN @UC, BERKELEY

WSA-PF+CS N. ARGE @AFRL

ANMHD B. ABBET, ET AL @BERKELEY

MAS LINKER ET AL @SAIC, CA

HELIOSPHERE

ENLIL, ENLIL+CONE D. ODSTRCIL @ UC BOULDER

BATSRUS GOMBOSI, TOTH ET AL @ CSEM, U. MICHIGAN

HELIOSPHERIC TOMOGRAPHY + IPS/SMEI JACKSON, HICK @CASS, UCSD

EXOSPHERIC SOLAR WIND H. LAMY, V. PIERRARD @IASB-BIRA

GLOBAL MAGNETOSPHERE

LANL* NEURAL NETWORK YIQUN, KOLLER @LANL OPEN GGCM RAEDER, FULLER-ROWELL @SSC, UNH

BATSRUS GOMBOSI, TOTH ET AL @ CSEM, U. MICHIGAN GUMICS JANHUNEN @FINISH METEROLOGICAL INST

WINDMI HORTON ET AL @ U TEXAS CMIT LFM-MIX LYON ET AL @DARTMOUTH, NCAR-HAO, JHU-APL,CISM

THERMOSPHERE/IONOSPHERE

IRI

BILITZA

@NASA/GSFC

COSGROVE-PF

COSGROVE

@SRI INTL., CA

SAMI-3 HUBA ET AL @NRL, ICARUS RES. INC.

CTIP-E

ROWELL ET AL

@CASS/USU

ABBYNORMAL

ECCLES ET AL

@CASS/USU

@UTAH STATE

USU-GAIM GITM SCHUNK ET AL RIDLEY @UTAH STATE

A PBMOD EY RETTERER

OVATION PRIME NEWELL @JHU APL

WEIMER MODEL @VIRGINIA TECH

ATMOSPHERE: MSISE HEDIN

INNER MAGNETOSPHERE

LFM-RCM-MIX-TIEGCM LYON, ET AL @DARTMOUTH,NCAR/ HAO,APL,RICE

S RICE CONVECTION MODEL WOLF, SAZYKIN @RICE, TX

> CRCM MEI-CHING FOK @NASA/GSFC

ATSRUS

m

PLASMASPHERE MODEL V PIERRARD

@IASB-BIRA

TSYGANENKO TSYGANENKO @ U ST-PETERSBURG

IGRF MACMILLAN, MAUS @IAGA FOK RADIATION BELT ELECTRON MEI-CHING FOK @NASA/GSFC

CIMI FOK, BUZULUKOVA @NASA/GSFC

Kamodo Design Goals

Endeavor to support a wide variety of users, models, and data sources:

- Quickly integrate new models and data
- LaTeX APIs for scientists and educators who don't code
- Model-agnostic interpolation API
- Format-agnostic
- Transparent, Permissive Metadata
- Automatic unit conversion
- Compatibility with helio-python ecosystem and support PyHC standards
- Instant visualization

Kamodo Architecture



Kamodo Architecture



Kamodo Functional API

Scientists work with models and data through Kamodo objects, which map symbols to interpolating functions or mathematical expressions.



Kamodo converts each expression to a highly optimized python function capable of operating on large arrays.

Kamodo Functional API

Existing Kamodo objects can be updated with new formulas using dictionary syntax. Function composition is applied automatically.



Many analysis and visualization problems can be framed in terms of *function composition*. Since most scientists are comfortable with function manipulation, this makes Kamodo ideal for their workflow.

from kamodo import kamodofy, Kamodo
import numpy as np

@kamodofy(units = 'kg/m^3', citation = 'Pembroke et. al, 2018')
~ def rho(x = np.array([3,4,5]), y = np.array([1,2,3])):
 """A function that computes density"""
 return x+y

Any python function can be "Kamodofied" using the @kamodofy decorator, adds metadata to a function.



 $\rho(x, y)[kg/m^3] = \lambda(x, y)$ den (x, y)[g/cm^3] = $\frac{\rho(x, y)}{1000}$ Any python function can be "Kamodofied" using the @kamodofy decorator, adds metadata to a function.

Lambda indicates kamodofied function



kamodo.den(3, 4)

0.007

Any python function can be "Kamodofied" using the @kamodofy decorator, adds metadata to a function.

Lambda indicates kamodofied function

Function composition & unit conversion

<pre>from kamodo import kamodofy, Kamodo import numpy as np @kamodofy(units = 'kg/m^3', citation = 'Pembroke et. al, 2018') • def rho(x = np.array([3,4,5]), y = np.array([1,2,3])): """A function that computes density""" return x+y</pre>	An the fur
<pre>kamodo = Kamodo(rho = rho) kamodo['den[g/cm^3]'] = 'rho' kamodo</pre>	
$ \rho(x, y)[kg/m^3] = \lambda(x, y) $ den $(x, y)[g/cm^3] = \frac{\rho(x, y)}{1000}$	La
kamodo.den(3,4)	Fu
0.007	
kamodo.rho.meta <i># PyHC standard</i>	P
<pre>{'citation': 'Pembroke et. al, 2018', 'units': 'kg/m^3'}</pre>	re
kamodo.rho.data # PyHC standard	00

array([4, 6, 8])

Any python function can be "Kamodofied" using the @kamodofy decorator, adds metadata to a function.

Lambda indicates kamodofied function

Function composition & unit conversion

Python-in-Heliophysics Community Standards require a data attribute - we satisfy this by calling functions with their default arguments.

<pre>from kamodo import kamodofy, Kamodo import numpy as np @kamodofy(units = 'kg/m^3', citation = 'Pembroke et. al, 2018') v def rho(x = np.array([3,4,5]), y = np.array([1,2,3])): """A function that computes density""" return x+y</pre>	Any pyth the @ka function	non moe	function can be "Karr dofy decorator, adds r	10dofie netada	d" using ta to a
<pre>kamodo = Kamodo(rho = rho) kamodo['den[g/cm^3]'] = 'rho' kamodo</pre>					
$\rho(x, y)[kg/m^3] = \lambda(x, y)$ den (x, y)[g/cm^3] = $\frac{\rho(x, y)}{1000}$	Lambda	a in	dicates kamodofied fu	Inction	
kamodo.den(3,4)	Functio	n co	omposition & unit con	version	l
0.007					
kamodo.rho.meta # PyHC standard	Python	-in-	Heliophysics Commu	nity Sta	andards
{'citation': 'Pembroke et. al, 2018', 'units': 'kg/m^3'}	require	a d fund	ata attribute - we satis	sfy this	by ments
kamodo.rho.data <i># PyHC standard</i>	Cannig			n argai	nemo.
array([4, 6, 8])	Suppo	rting	g Documentation		
help(kamodo.rho)	kamodo	.det	ail <u>()</u>		
Help on function rho in modulemain:		lhs	rhs	symbol	units
<pre>rho(x=array([3, 4, 5]), y=array([1, 2, 3])) </pre>	rho(x, y)	rho	<pre><function 0x122bcde60="" at="" rho=""></function></pre>	rho(x, y)	kg/m^3

den(x, y) den rho(x, y)/1000

den(x, y) g/cm^3

A function that computes density

Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	Ν	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	



1D Time Series

Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

2D Parametric Curve



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	N, M	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

3D Parametric Curve



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	N, M	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

Colored 3D curve



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	N	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	N	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

2-D Vector Curve



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	Ν	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	N	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

2-D Vector Field



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

3-D Vector Field/Curve



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	Ν	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

Contour/Smooth



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes	Skew/Carpet
nargs	arg shapes	out shape			
1	N	N	1-d line		$\mathrm{f}_{\mathrm{NM}}\left(x_{NM},y_{NM} ight)[cm^{2}]=\lambda(x_{NM},y_{NM})$
		Nx2	2-d line		3.5
		Nx3	3-d line		3
	Nx2	Nx2	2-d vector field		25
	Nx3	Nx3	3-d vector field		
2	N, M	NxM	2-d contour	indexing	2
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing	W 1.5
3	1, M, N	MxN	Map-to-plane	indexing*	1
	L, 1, N	LxN	Map-to-plane	indexing*	0.5
	L, M, 1	LxM	Map-to-plane	indexing*	
	N, N, N	Ν	3-d colored line		
	NxM, NxM, NxM	1	Parametric Surface		-0.5
		NxM	Coloured Parametric Surface		\vec{x}

3

0

4

Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	N, M	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
'	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

Map-to-plane



Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	Ν	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

$iplot(kamodo.plot(p = \{\}), show_link = False)$ $p(x_{NM}, y_{NM}, z_{NM})[cm] = \lambda(x_{NM}, y_{NM}, z_{NM})$ $u_{a} = \frac{1}{2 - NM} \frac{1}{2 - N} \frac{1}{2 - N}$

Parametric Surface

Kamodo generates assets ready for visualization based on the array shapes of field inputs and outputs. Kamodo currently outputs Plotly (json) for interactivity and easy web deployment.

			Plot Type	notes
nargs	arg shapes	out shape		
1	N	Ν	1-d line	
		Nx2	2-d line	
		Nx3	3-d line	
	Nx2	Nx2	2-d vector field	
	Nx3	Nx3	3-d vector field	
2	Ν, Μ	NxM	2-d contour	indexing
	NxM, NxM	NxM	2-d contour (skew/carpet)	indexing
3	1, M, N	MxN	Map-to-plane	indexing*
	L, 1, N	LxN	Map-to-plane	indexing*
	L, M, 1	LxM	Map-to-plane	indexing*
	N, N, N	Ν	3-d colored line	
	NxM, NxM, NxM	1	Parametric Surface	
		NxM	Coloured Parametric Surface	

Colored Parametric Surface



Kamodofication

- the process of exposing models and data to Kamodo

A model or data source is considered "kamodofied" when all scientifically relevant variables are exposed as Kamodo objects.

Kamodofication requirements:

- 1. Model must be accessible from python
- 2. Model must provide an interpolating function for each variable
- 3.Interpolating functions should supply default values as arguments, indicating the valid domain for their inputs.
- 4. Variable names should follow Kamodo's naming specification for LaTeX legibility.
- 5. Interpolating functions must contain the following metadata as attributes:
 - 1. meta dictionary of {'units': 'kg', 'citation' : 'Doe, J. et. al'}
 - 2. data array
- 6. Class Methods should use "self" as the first argument.

TIEGCM

Thermosphere Ionosphere Electrodynamics General Circulation Model R. G. Roble et al., High Altitude Observatory, National Center for Atmospheric Research

TN (*time*, *ilev*, *lat*, *lon*)[K] = λ (*time*, *ilev*, *lat*, *lon*) UN (time, ilev, lat, lon) $[cm/s] = \lambda$ (time, ilev, lat, lon) VN (*time*, *ilev*, *lat*, *lon*)[*cm/s*] = λ (*time*, *ilev*, *lat*, *lon*) O_1 (*time*, *ilev*, *lat*, *lon*)[1] = λ (*time*, *ilev*, *lat*, *lon*) NO (time, ilev, lat, lon)[1] = λ (time, ilev, lat, lon) N4S (*time*, *ilev*, *lat*, *lon*)[1] = λ (*time*, *ilev*, *lat*, *lon*) HE (time, ilev, lat, lon)[1] = λ (time, ilev, lat, lon) NE (time, ilev, lat, lon) $[cm * * - 3] = \lambda$ (time, ilev, lat, lon) TE (time, ilev, lat, lon)[K] = λ (time, ilev, lat, lon) TI (*time*, *ilev*, *lat*, *lon*)[K] = λ (*time*, *ilev*, *lat*, *lon*) O_2 (*time*, *ilev*, *lat*, *lon*)[1] = λ (*time*, *ilev*, *lat*, *lon*) $O2P_{ELD}$ (time, ilev, lat, lon) = λ (time, ilev, lat, lon) $\omega(time, ilev, lat, lon)[s * * - 1] = \lambda(time, ilev, lat, lon)$ POTEN (*time*, *ilev*, *lat*, *lon*)[*volts*] = λ (*time*, *ilev*, *lat*, *lon*) UI_{ExB} (time, ilev, lat, lon)[cm/s] = λ (time, ilev, lat, lon) VI_{ExB} (time, ilev, lat, lon)[cm/s] = λ (time, ilev, lat, lon) WI_{ExB} (time, ilev, lat, lon)[cm/s] = λ (time, ilev, lat, lon) OP (*time*, *ilev*, *lat*, *lon*)[cm * * - 3] = λ (*time*, *ilev*, *lat*, *lon*) N2P_{ELD} (*time*, *ilev*, *lat*, *lon*) = λ (*time*, *ilev*, *lat*, *lon*) NPLUS (time, ilev, lat, lon) = λ (time, ilev, lat, lon) NOP_{FLD} (*time*, *ilev*, *lat*, *lon*) = λ (*time*, *ilev*, *lat*, *lon*) $\sigma_{PFD}(time, ilev, lat, lon)[ohm * * - 1/m] = \lambda(time, ilev, lat, lon)$ σ_{HAI} (time, ilev, lat, lon) [ohm ** - 1/m] = λ (time, ilev, lat, lon) DEN (*time*, *ilev*, *lat*, *lon*)[g/cm * *3] = λ (*time*, *ilev*, *lat*, *lon*) QJOULE (time, ilev, lat, lon)[erg/(g * s)] = λ (time, ilev, lat, lon) $Z(ilev, time, lat, lon, z_{level})[cm] = \lambda(ilev, time, lat, lon, z_{level})$ $ZG(time, ilev, lat, lon)[cm] = \lambda(time, ilev, lat, lon)$ O_{N2} (time, ilev, lat, lon)[1] = λ (time, ilev, lat, lon) $N2D_{ELD}$ (time, ilev, lat, lon) = λ (time, ilev, lat, lon) O2N (time, ilev, lat, lon) = λ (time, ilev, lat, lon) N2N (time, ilev, lat, lon) = λ (time, ilev, lat, lon) TEC (*time*, *lat*, *lon*) = λ (*time*, *lat*, *lon*) QJOULE_{INTEG} (*time*, *lat*, *lon*) = λ (*time*, *lat*, *lon*) EFLUX (time, lat, lon) = λ (time, lat, lon) HMF₂ (time, lat, lon) = λ (time, lat, lon) NMF₂ (time, lat, lon) = λ (time, lat, lon) TLBC (*time*, *lat*, *lon*) = λ (*time*, *lat*, *lon*) ULBC (time, lat, lon) = λ (time, lat, lon) VLBC (time, lat, lon) = λ (time, lat, lon) TLBC_{NM} (time, lat, lon) = λ (time, lat, lon) ULBC_{NM} (time, lat, lon) = λ (time, lat, lon) VLBC_{NM} (time, lat, lon) = λ (time, lat, lon) latitude (*time*, *lat*, *lon*) = λ (*time*, *lat*, *lon*) longitude (*time*, *lat*, *lon*) = λ (*time*, *lat*, *lon*)

Kamodofied TIEGCM fields. Interpolator functions provided by scipy. Dimension reduction is achieved by specifying a value for time.

kamodo['DEN_250(lat, lon, ilev)'] = 'DEN(250, ilev, lat, lon)'

 DEN_{250} (*lat*, *lon*, *ilev*) = DEN (250, *ilev*, *lat*, *lon*)

llat, llon = np.meshgrid(lat, lon)
iplot(kamodo.plot(DEN_250 = dict(lat = llat, lon = llon, ilev = 0)))

 DEN_{250} (*lat*, *lon*, *ilev*) = DEN (250, *ilev*, *lat*, *lon*)



GITM

Global lonosphere Thermosphere Model

A.J. Ridley, Department of Atmosphere, Oceanic and Space Sciences, University of Michigan

Access to GITM is provided by SpacePy, interpolation methods built on SciPy.

 $\rho(lon, lat, alt)[kg/m^3] = \lambda(lon, lat, alt)$ Altitude (*lon*, *lat*, *alt*) = λ (*lon*, *lat*, *alt*) NeutralTemperature $(lon, lat, alt)[K] = \lambda(lon, lat, alt)$ $V_n^{up}(lon, lat, alt)[m/s] = \lambda(lon, lat, alt)$ Latitude (*lon*, *lat*, *alt*) = λ (*lon*, *lat*, *alt*) V_{i}^{east} (lon, lat, alt)[m/s] = λ (lon, lat, alt) NO (lon, lat, alt) $[1/m^3] = \lambda(lon, lat, alt)$ $N_2(lon, lat, alt)[1/m^3] = \lambda(lon, lat, alt)$ NO⁺ (lon, lat, alt) $[1/m^3] = \lambda(lon, lat, alt)$ O (3P) (lon, lat, alt) $[1/m^3] = \lambda(lon, lat, alt)$ lt (lon, lat, alt)[hours] = λ (lon, lat, alt) O^+_{ASD} (lon, lat, alt)[1/m³] = λ (lon, lat, alt) dLon (*lon*, *lat*, *alt*) = λ (*lon*, *lat*, *alt*) e- (lon, lat, alt) $[1/m^3] = \lambda(lon, lat, alt)$ Longitude (*lon*, *lat*, *alt*) = λ (*lon*, *lat*, *alt*) $V_{i}^{up}(lon, lat, alt)[m/s] = \lambda(lon, lat, alt)$ N_2^+ (lon, lat, alt) $[1/m^3] = \lambda(lon, lat, alt)$ V_i^{north} (lon, lat, alt)[m/s] = λ (lon, lat, alt) V_n^{east} (lon, lat, alt)[m/s] = λ (lon, lat, alt) O_2^+ (lon, lat, alt) $[1/m^3] = \lambda(lon, lat, alt)$ dLat $(lon, lat, alt) = \lambda(lon, lat, alt)$ $O_2(lon, lat, alt)[1/m^3] = \lambda(lon, lat, alt)$ V_n^{north} (lon, lat, alt)[m/s] = λ (lon, lat, alt)

 $O(3P)(lon, lat, alt)[1/m^3] = \lambda(lon, lat, alt)$



GITM

Global Ionosphere Thermosphere Model

A.J. Ridley, Department of Atmosphere, Oceanic and Space Sciences, University of Michigan

Transformation to cartesian coordinates is achieved through 6M function composition. 4M 2M $ALT(r)[m] = \lambda(r)$ z -2M $LON(\phi)[degrees] = \lambda(\phi)$ _AM -6M LAT $(\theta)[degrees] = \lambda(\theta)$ _6M $r(x, y, z) = \sqrt{x^2 + y^2 + z^2}$ AM .2M `On PA. $\theta(x, y, z) = \operatorname{asin}\left(\frac{z}{r(x, y, z)}\right)$ RA. ZA, ×1, GM х S2 $\phi(x, y) = \operatorname{atan}_2(y, x)$

 $\rho(x, y, z)[kilogram/meter * *3] = \rho(\text{LON}(\phi(x, y)), \text{LAT}(\theta(x, y, z)), \text{ALT}(r(x, y, z)))$ $O(3P)(x, y, z)[meter * *(-3)] = O(3P)(\text{LON}(\phi(x, y)), \text{LAT}(\theta(x, y, z)), \text{ALT}(r(x, y, z)))$

Adaptively Refined MHD Solver (ARMS)

Antiochos, S. K., Masson, S. DeVore, Goddard GSFC

ARMS is a 3D solar physics model capable of capturing transient solar eruptions. We have kamodofied the following fields from ARMS, using a custom octree interpolator written in python. Kamodo is used to transform from the model coordinates to cartesian.

$$\begin{split} \rho(logr,\theta,\phi)[g/cm^3] &= \lambda(logr,\theta,\phi) \\ v_r(logr,\theta,\phi)[cm/s] &= \lambda(logr,\theta,\phi) \\ v_\theta(logr,\theta,\phi)[cm/s] &= \lambda(logr,\theta,\phi) \\ v_\phi(logr,\theta,\phi)[cm/s] &= \lambda(logr,\theta,\phi) \\ P(logr,\theta,\phi)[Pa] &= \lambda(logr,\theta,\phi) \\ T(logr,\theta,\phi)[K] &= \lambda(logr,\theta,\phi) \\ B_r(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ B_\theta(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ B_\phi(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ J_r(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ J_\theta(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ J_\theta(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ J_\theta(logr,\theta,\phi)[G] &= \lambda(logr,\theta,\phi) \\ \beta(logr,\theta,\phi)[1] &= \lambda(logr,\theta,\phi) \\ S_{alfven}(logr,\theta,\phi)[1] &= \lambda(logr,\theta,\phi) \\ \Phi(logr,\theta,\phi)[1] &= \lambda(logr,\theta,\phi) \end{split}$$



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$$r(x, y, z) = \sqrt{x^2 + y^2 + z^2}$$

$$\log r(x, y, z) [log(Dimension(1))] = \log (r(x, y, z))$$

$$\theta(x, y, z) = asin\left(\frac{z}{r(x, y, z)}\right)$$

$$\phi(x, y) = atan_2 (y, x)$$

$$\rho(x, y, z) [gram/centimeter * *3] = \rho(\log r(x, y, z), \theta(x, y, z), \phi(x, y))$$

Trick: rho and RHO are unique python names that render as the same greek letter in LaTeX, so we are free to define both rho(x,y,z) and RHO(logr,theta,phi), different signatures for the same variable.

$\begin{array}{c} 2 \\ 1.5 \\ 1 \\ 0.5 \\ 2 \\ 0 \\ -0.5 \\ -1 \\ -1.5 \\ -2 \\ -1.5 \\ 0 \\ y \\ 0.5 \\ 1.5 \\ z \\ y \\ 0 \\ x \\ \end{array}$

iplot(arms.plot(RHO = dict(x = 0, y = yy))

z = zz)))

Kamodofied Data: DISCOVR

The Kamodofied discovr feed provides interpolated plasma and field variables. The time interpolator is built on pandas' time series interpolators.



Kamodofied Data: HAPI

A work-in-progress, the Kamodofied HAPI api is built on the python hapiclient from Bob Weigel. Here, the variable *xvec* is built from the position components of ACE spacecraft as a function of time.



-20

-30 _20

,10

Y[R_E]

0

\$

 $\sqrt[n]{2}$

30

0

X[R_E]

Summary/Future Plans

tldr: Kamodo provides a functional, publication-focused interface for space weather models and data.

Further Resources:

Project Page - <u>https://ccmc.gsfc.nasa.gov/Kamodo/</u>

Code - https://sed-gitlab.gsfc.nasa.gov/ccmc/Kamodo

Future:

Output formats - An extension of visualization API (csv, OpenSpace, json)

Komodo-Live - Automatically generate web-based interfaces similar to Kameleon-Live

Packaging - Provide conda environments for easy distribution

OpenSource - NASA Software Release Process is underway

Thank You!

Kamodo Functional API

Kamodo will interpret certain strings as LaTeX for "syntactic sugar":



Corresponding LaTeX output is available for use in publications:

kamodo.to_latex()

u"\begin{equation}\rho{\\left (\\alpha,\\beta,\\gamma \\right)} = \\alpha + \\beta + \\gamma\\end{equation}\\begin {equation}\vec{r}{\\left (t \\right)} = t\\end{equation}\\begin{equation}\\operatorname{{f}'}{\\left (x \\right)} = x\\end{equation}\\begin{equation}\\vec{x}_{i}{\\left (\\vec{x}_{i-1} \\right)} = \\vec{x}_{i-1} + 1\\end{equation} \\begin{equation}\\operatorname{F^{gravity}}{\\left (G,M,R,m \\right)} = \\frac{G M m}{R^{2}}\\end{equation}"

Fortran -> Kamodo

Models must be accessible as python modules to be compatible with Kamodo. Here, a Fortran reader is used to subclass Kamodo.

Python Kamodo-compatible class

```
import read ascii
from scipy.interpolate import RegularGridInterpolator
import numpy as np
class ColumnReader(Kamodo):
    def __init__(self, filename, *args, **kwargs):
        super(ColumnReader, self). init (*args, **kwargs)
        read ascii.ascii.read file(filename)
        self.lon = read ascii.ascii.lons
        self.lat = read ascii.ascii.lats
        self.alt = read ascii.ascii.alts
        self.data = read ascii.ascii.data
        bounds error = kwargs.get('bounds error', False)
        fill_value = kwargs.get('missing_value', np.nan)
        self.interpolator = RegularGridInterpolator((self.alt, self.lat, self.lon),
                                                    self.data, bounds error = False)
        try:
            self['rho'] = self.density
        except:
            print self.signatures
            raise
    @kamodofy(units = '1/cm^3')
    def density(self, alt, lat, lon):
        points = np.hstack((alt, lat, lon)).reshape(3, -1).T
        return self.interpolator(points)
```

filename = 'sample data/SampleFortran/Sophia Schwalbe 082718 IT 1 TIE-GCM 20120709 002000.dat'

Fortran Reader

```
! FILE: read_ascii.f90
module ascii
  INTEGER :: nlon,nlat,nalt
  REAL*8, allocatable, dimension(:) ::
ALTS,LATS,LONS
  REAL*8, allocatable, dimension(:,:,:) ::
data
contains
  SUBROUTINE read_file(FILENAME)
!
  READ ASCII FILE with interpolated data
from TIE-GCM
!
```

Fortran readers and interpolators may be accessed through the popular F2Py utility

Custom readers inherit all

the functionality of Kamodo

```
Jser
```

column

 $\rho(alt, lat, lon)[1/cm^3] = \lambda(alt, lat, lon)$

column = ColumnReader(filename)

column.rho(331, 33.5, -106.35)

array([2526809.55244755])

If numerical solutions are not available, as in the case of field line integration, Kamodo can use scipy to solve initial value problems, which are a class of ordinary differential equations.

 $f'(t, x) = \lambda(t, x)$ $f(t) = \lambda(t)$

A stop condition may be triggered by returning a negative value

If numerical solutions are not available, as in the case of field line integration, Kamodo can use scipy to solve initial value problems, which are a class of ordinary differential equations.



For vector-valued problems, treat each component separately. This framework will allow us to generate integral curves, including fieldlines, streaklines, particle trajectories, etc.



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Next Steps with OpenSpace

Now that we can quickly expose space weather models and data to scientists and educators, we can leverage this in the OpenSpace astrovisualization engine.

We need a target format for assets optimized in OpenSpace. Either

- have Kamodo output to that format
- convert from Plotly's json
- lupa (python wrapper for Lua/LuaJit)

We would like a discussion of pros and cons of either approach.