

PEAR

Equal Area Weather Forecasting on the Sphere

Hampus Linander 20/8 - 2025, GeUmetric Deep Learning, Umeå



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PEAR: Equal Area Weather Forecasting on the Sphere



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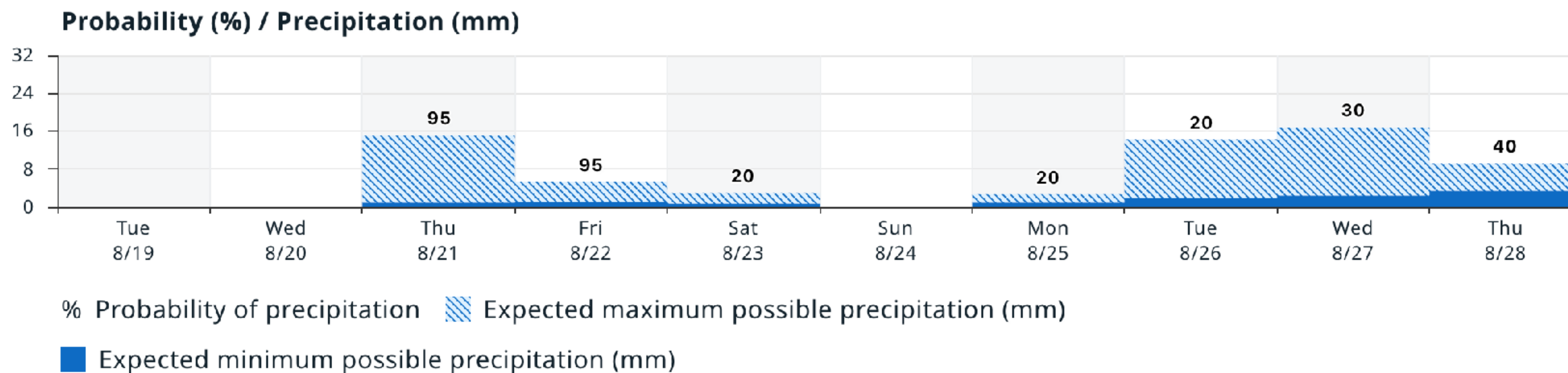
HEAL-SWIN: A Vision Transformer On The Sphere

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Christoffer Petersson^{*e a*} Daniel Persson^{*a*}

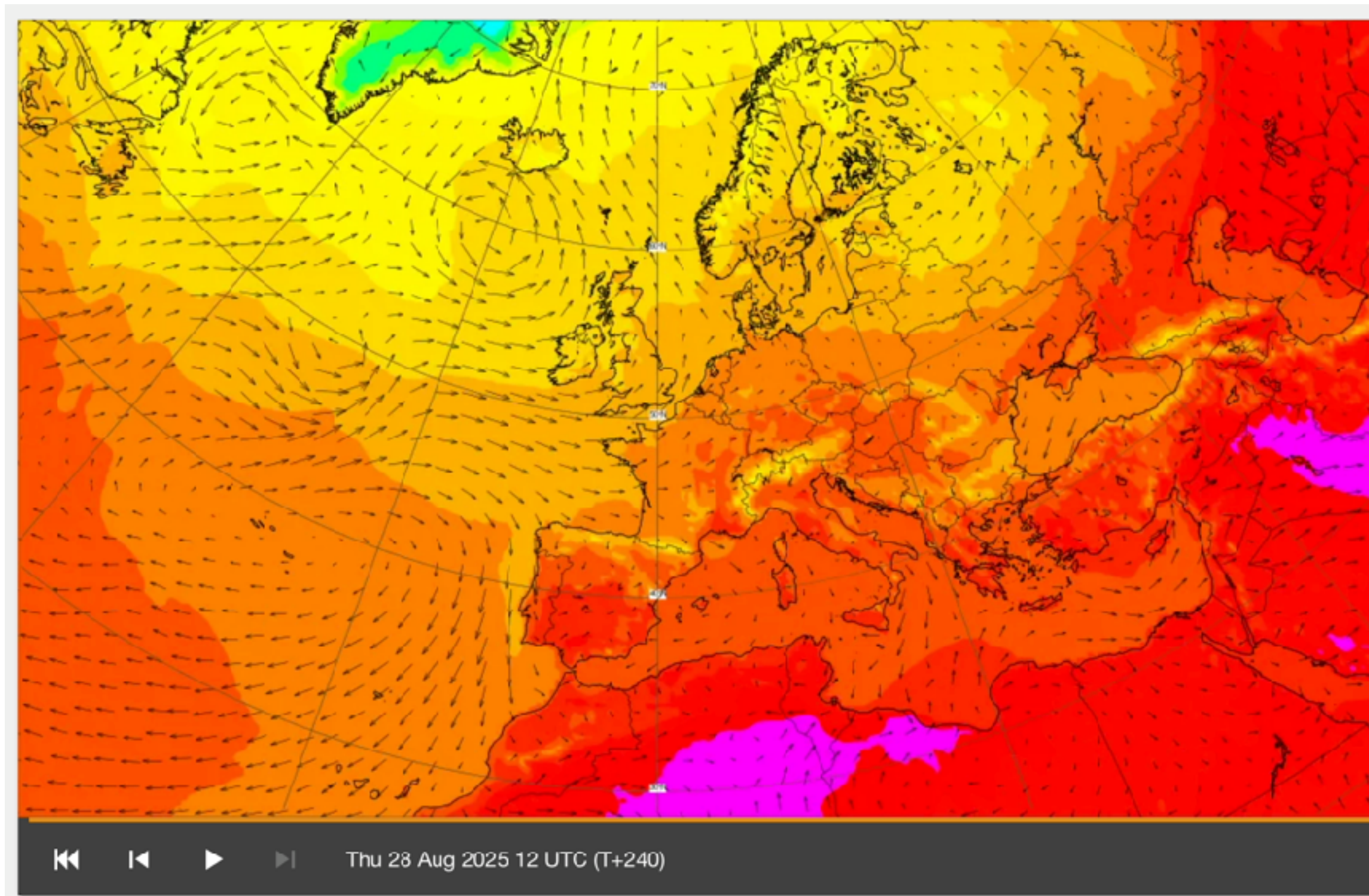
CVPR 2024

Weather forecasting

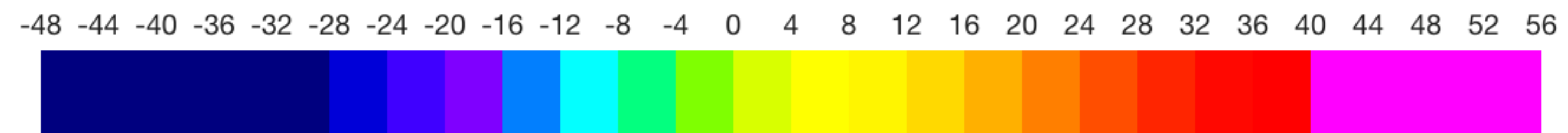
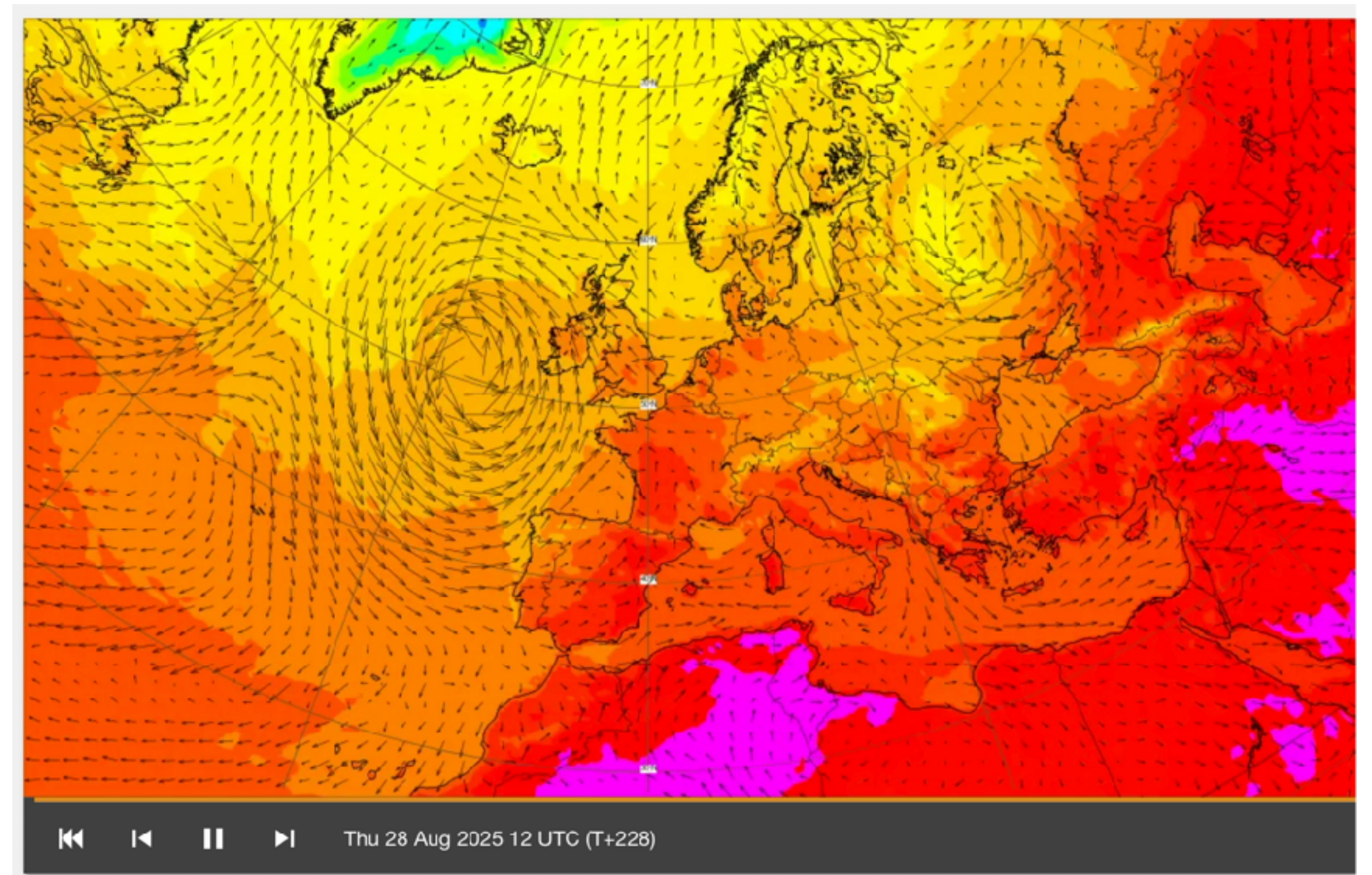


Weather forecasting

Graphcast



ECMWF ex-HRES



Problem

Medium-range forecasting

Predict future global atmospheric state at medium-range timescales

$$\Delta t \leq 15 \text{ days}$$

Traditional method: Solve Navier-Stokes numerically

Problem

Medium-range forecasting

Predict future global atmospheric state at medium-range timescales

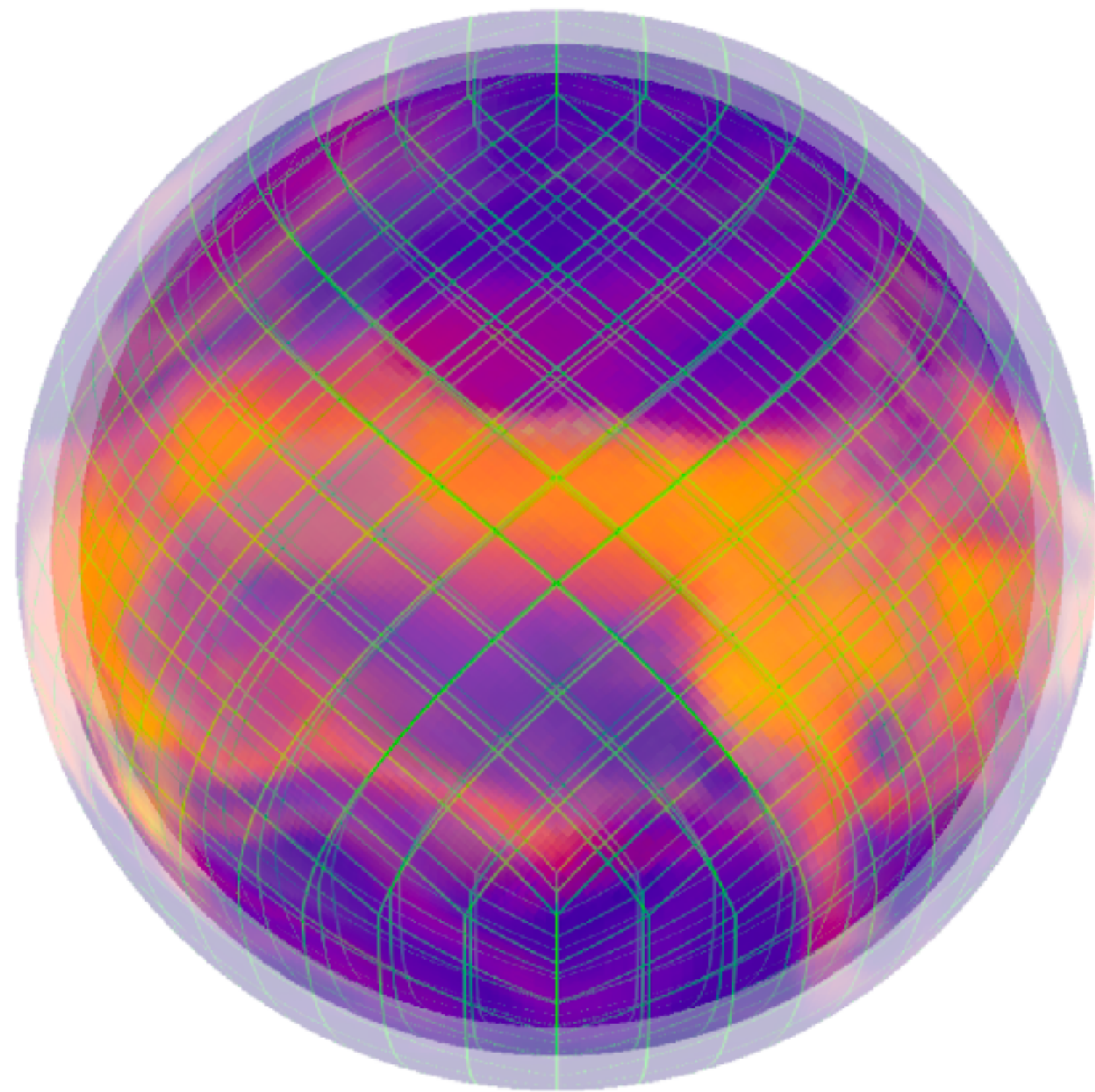
$$\Delta t \leq 15 \text{ days}$$

Traditional method: Solve Navier-Stokes numerically 🥵

Global weather forecasting

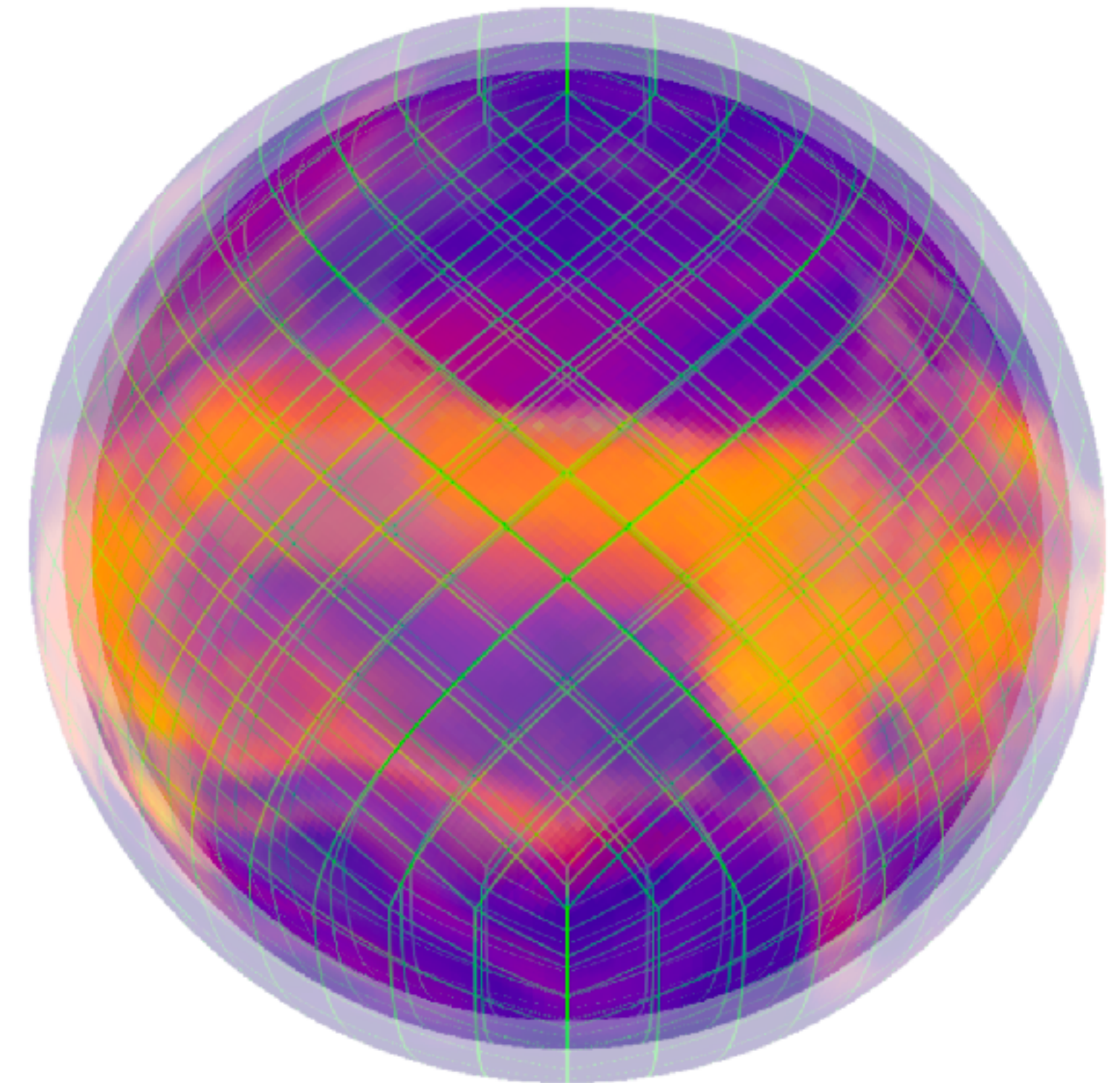
as a regression task

t_0



Input

$t_0 + \Delta t$



Target

ERA5

ECMWF reanalysis 1940-present

Hourly global grid (0.25 degrees, 31km resolution)
from surface up to height of 80km

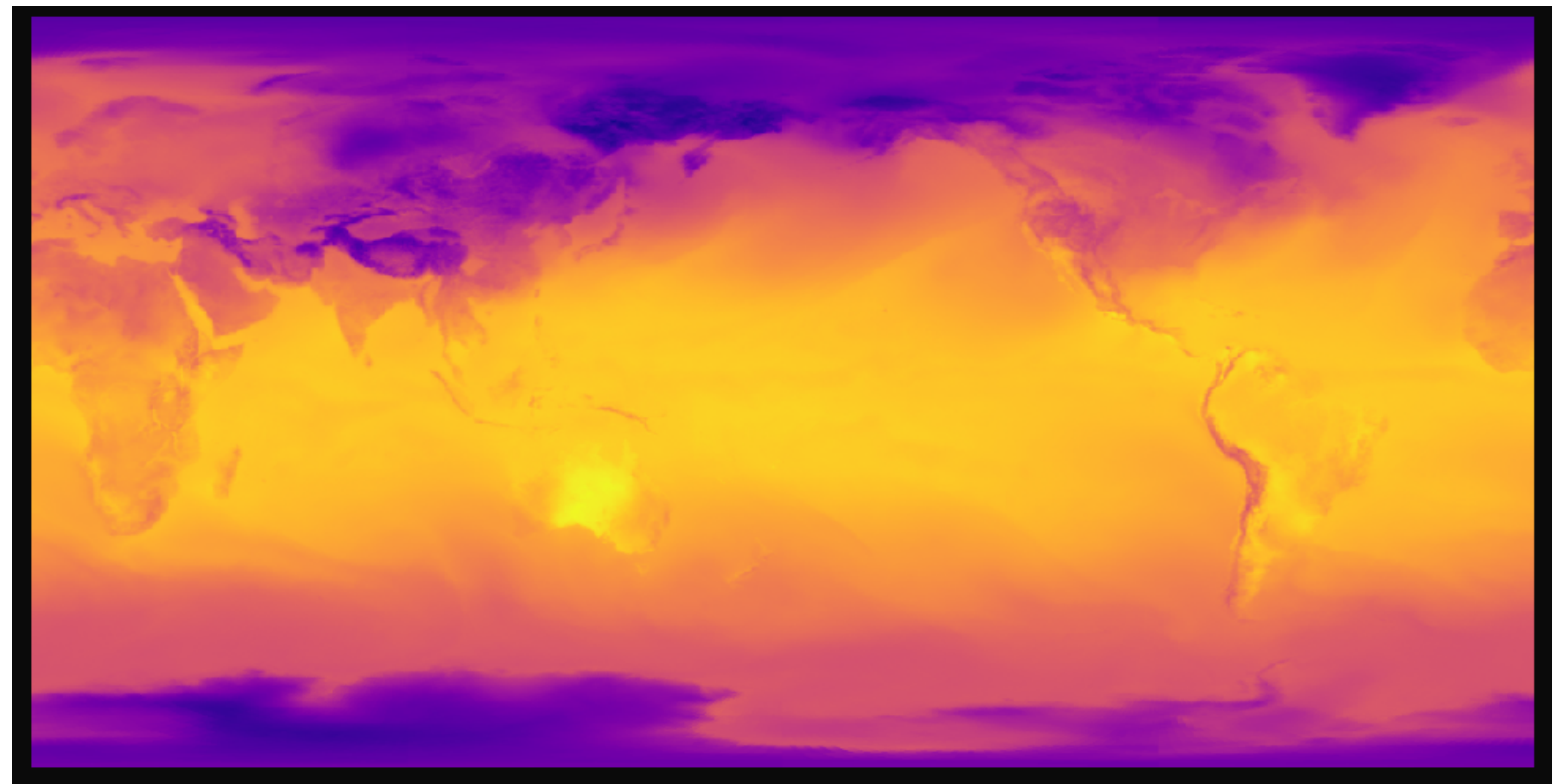
Surface

temperature
pressure
wind velocity
...

Atmosphere

temperature
geopotential
wind velocity
humidity
...

5 petabytes



ERA5 lite

reduced resolution

Daily global grid (0.92 degrees, 300km resolution)
with 13 height levels

2007 - 2017

Surface

temperature

pressure

wind velocity

Atmosphere

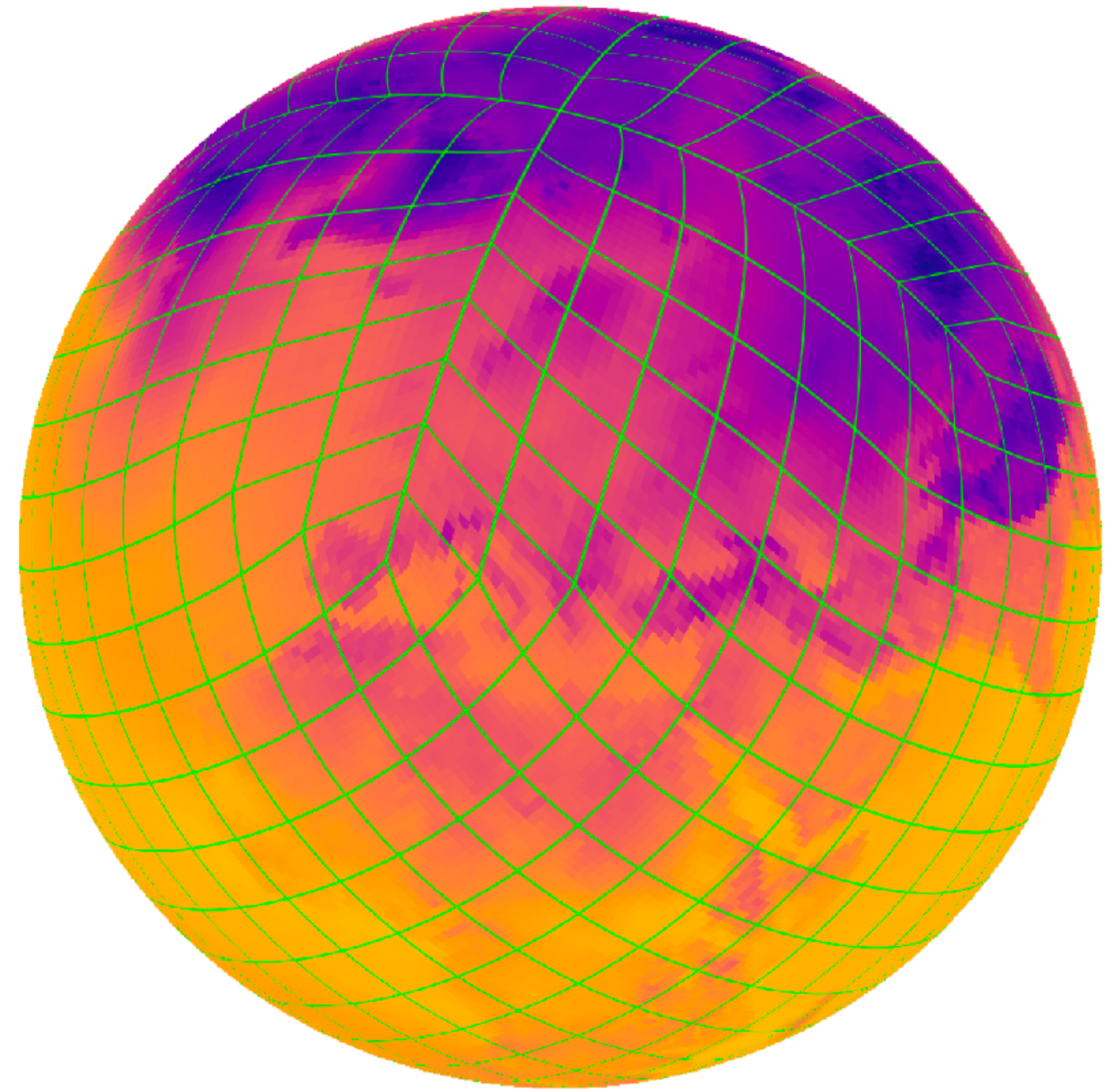
temperature

geopotential

wind velocity

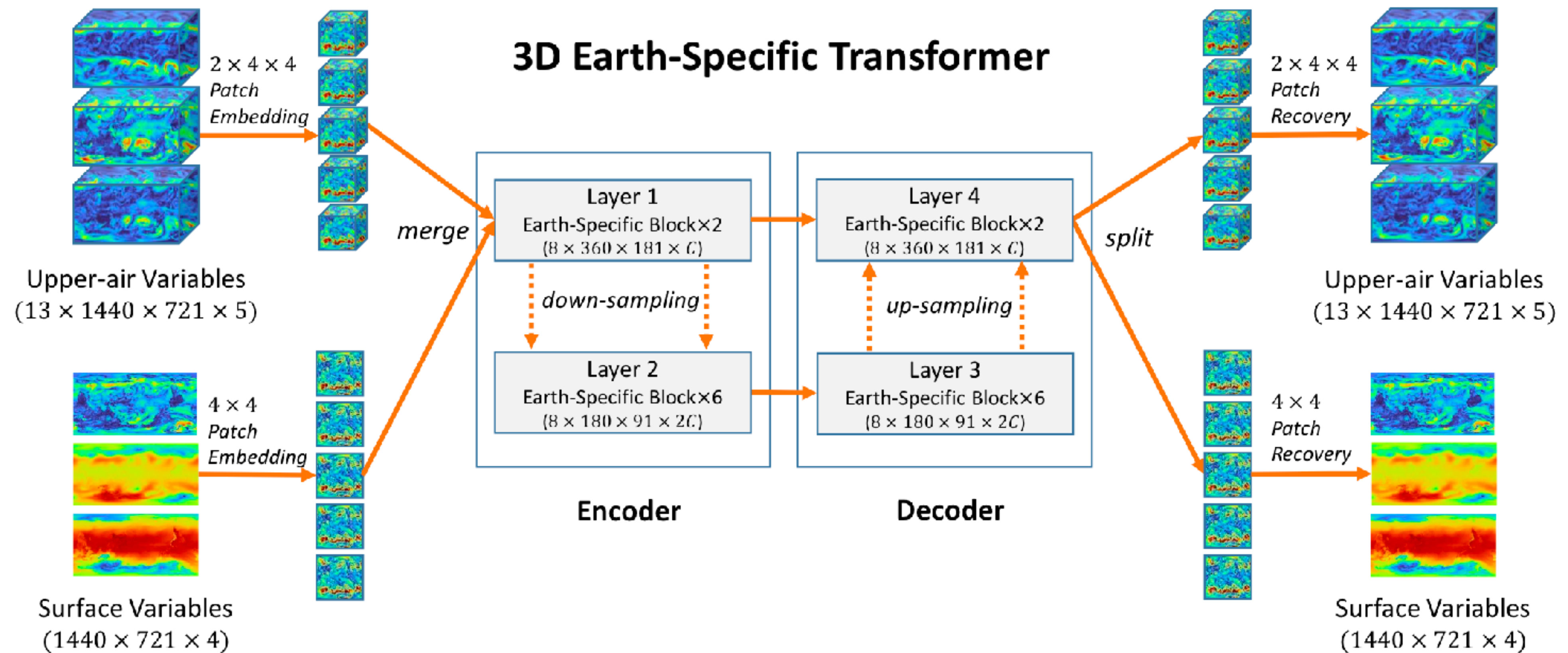
humidity

1 terrabyte



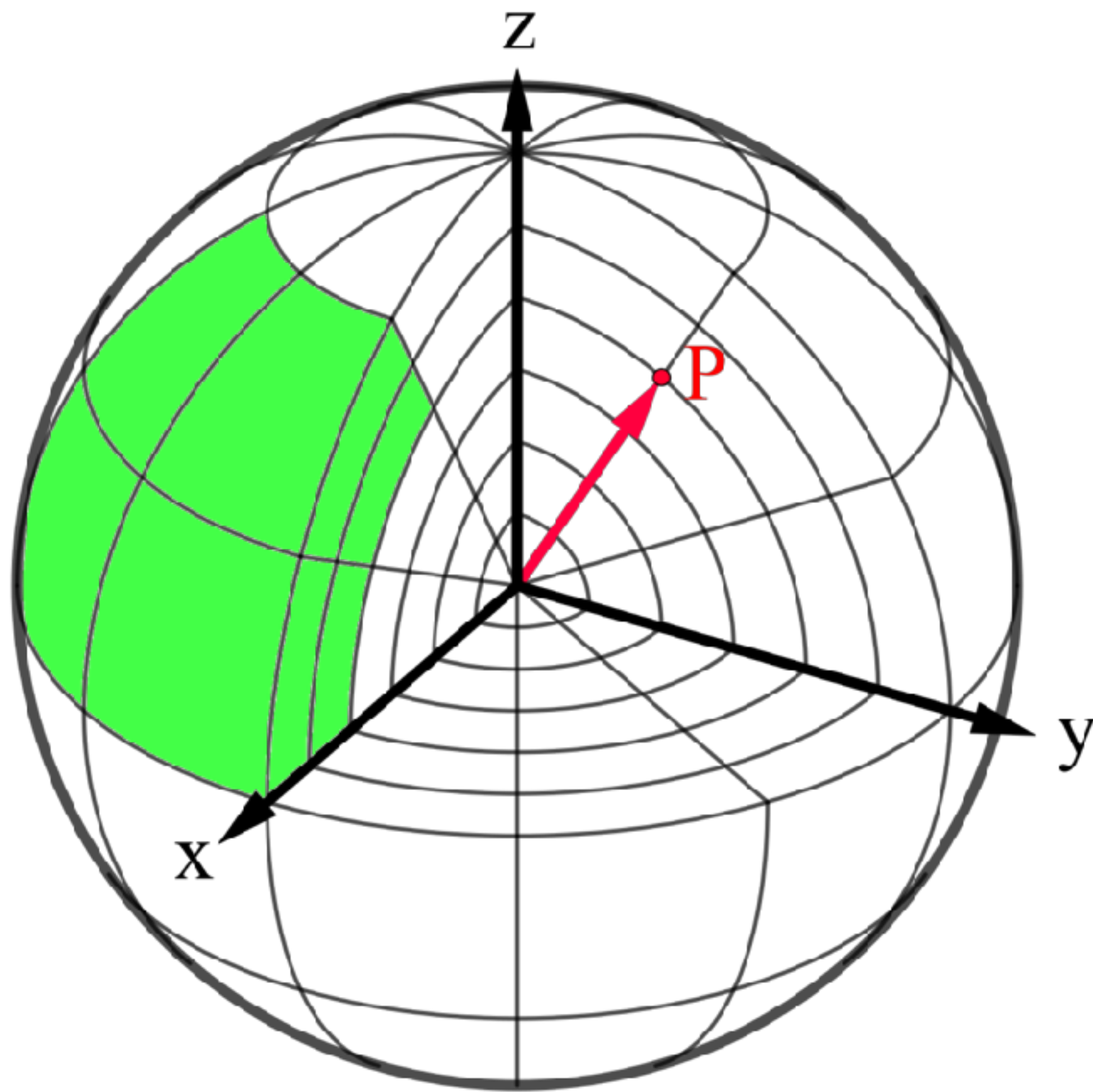
Pangu Weather

volumetric transformer

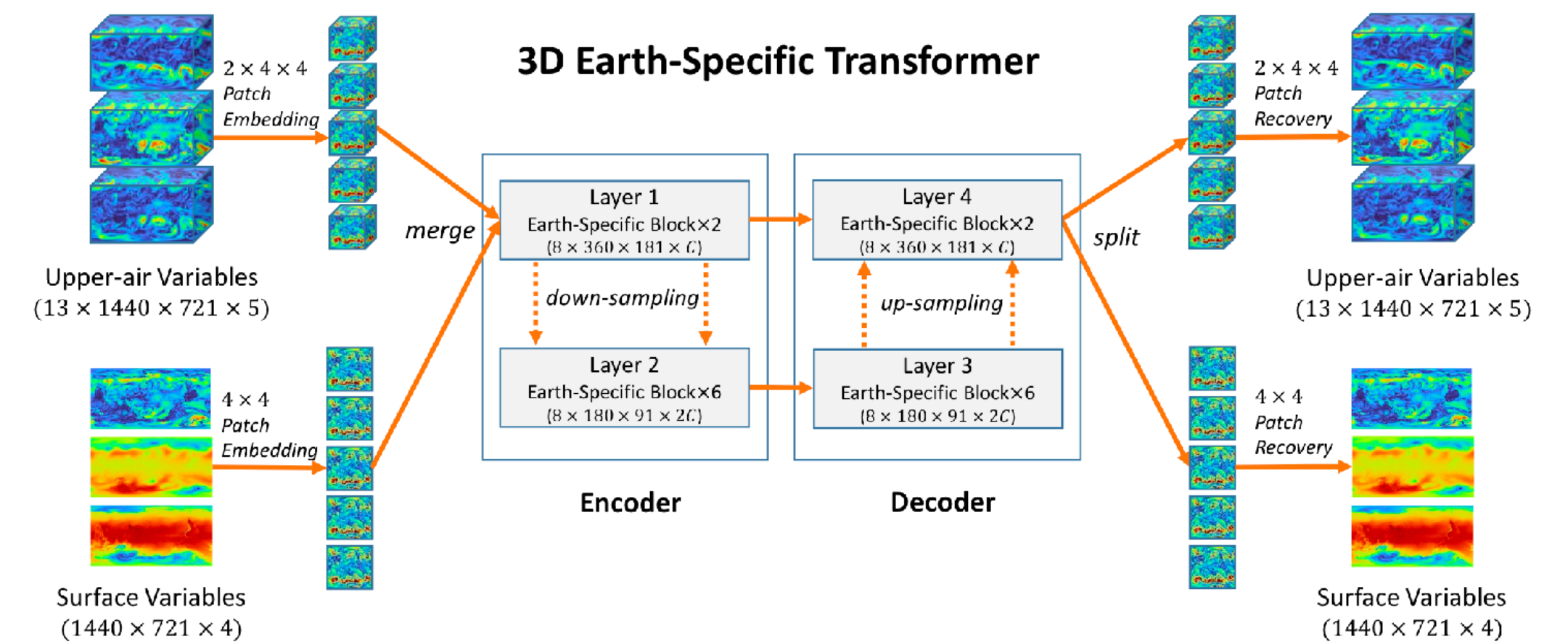


Pangu weather

Shifted windowed attention

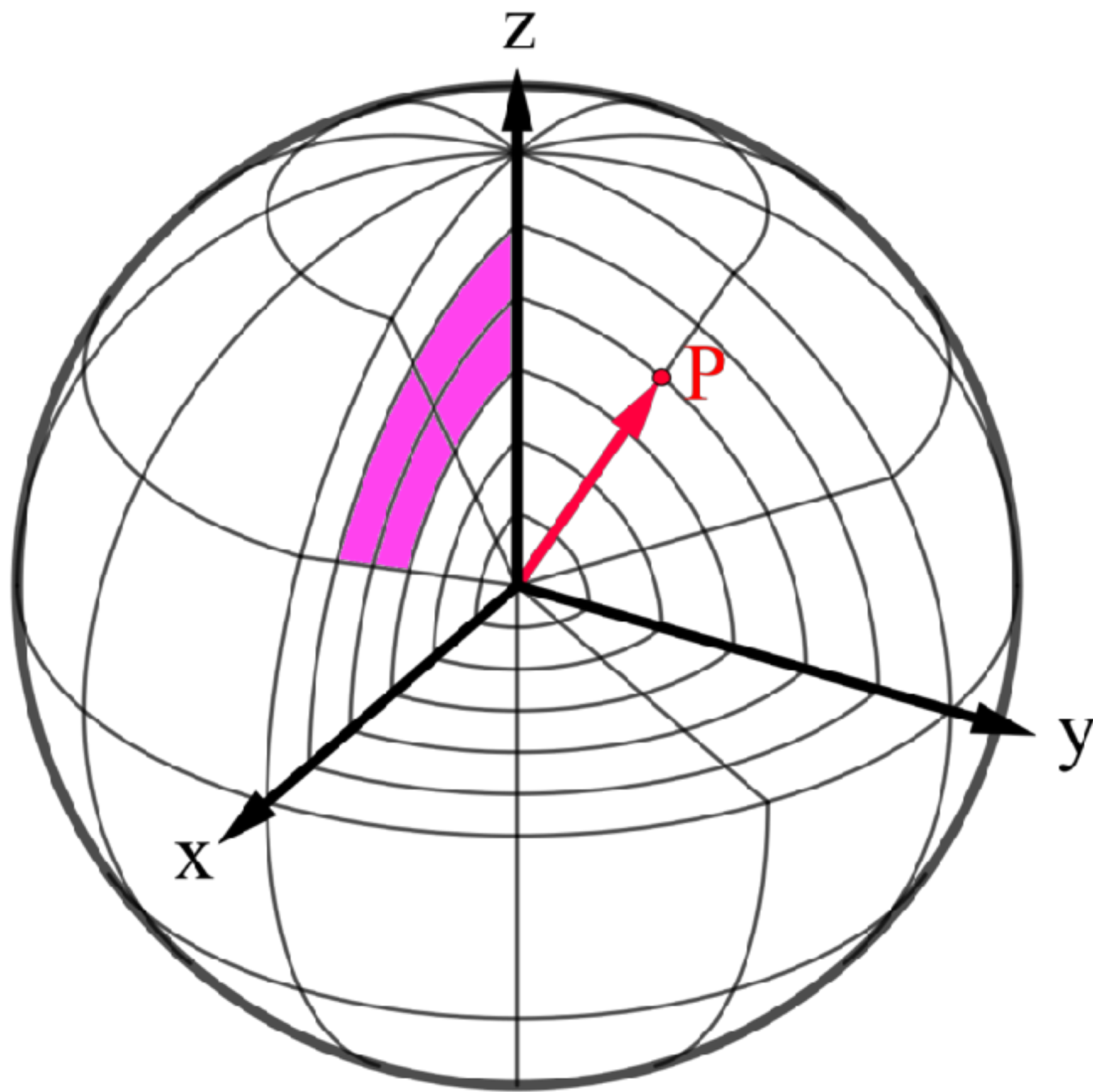


Liu et.al. "Swin Transformer: Hierarchical Vision Transformer using Shifted Windows", CVPR 2021

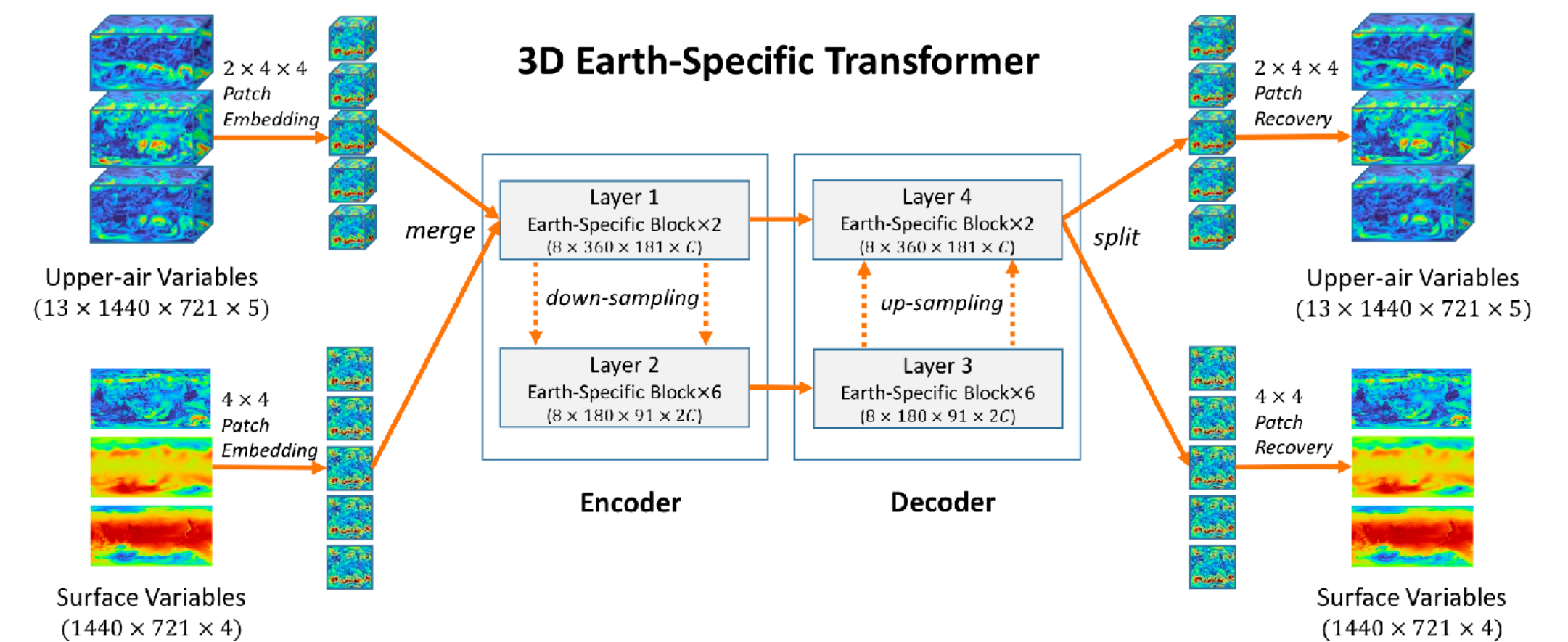


Pangu weather

Shifted windowed attention

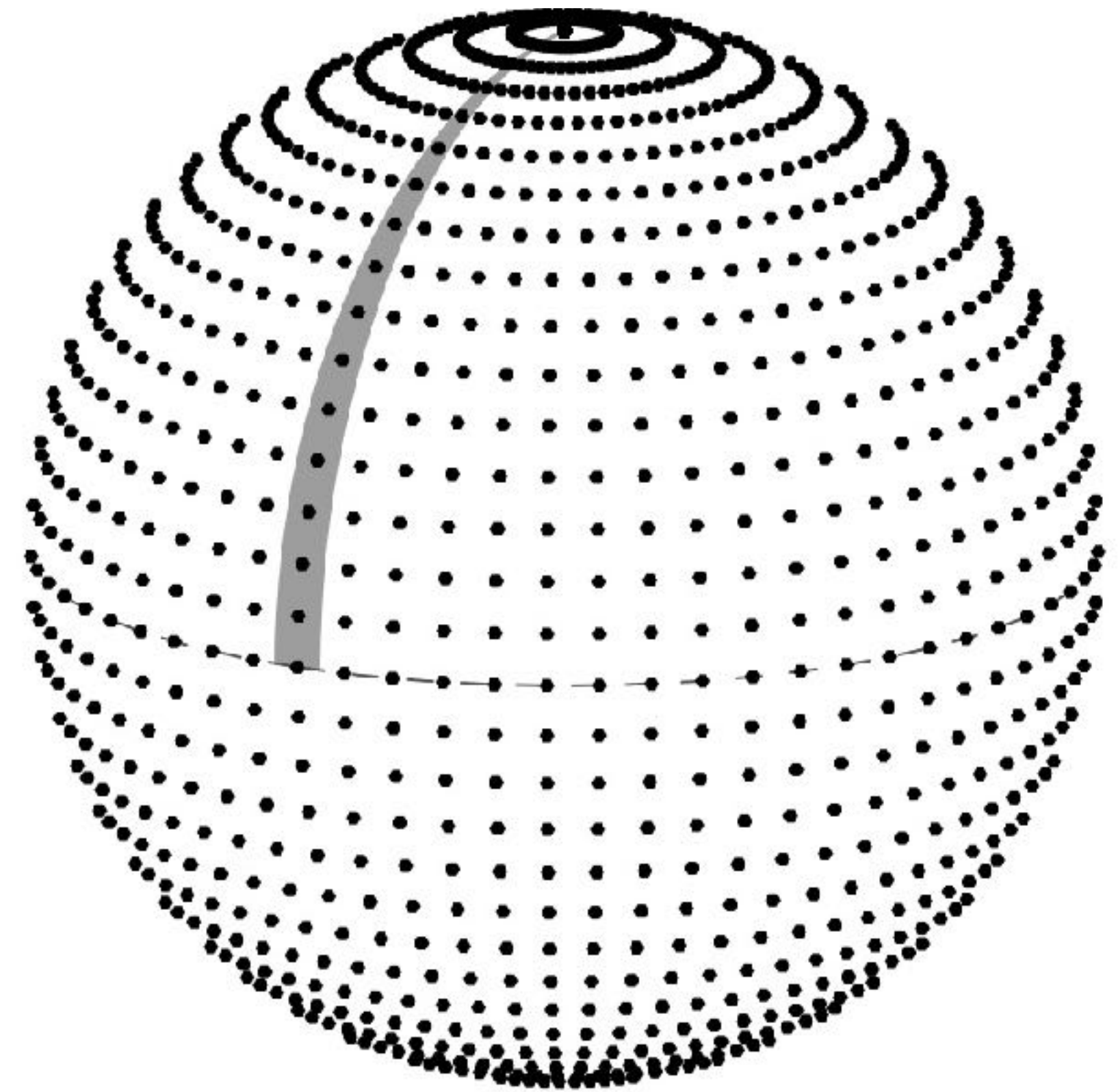
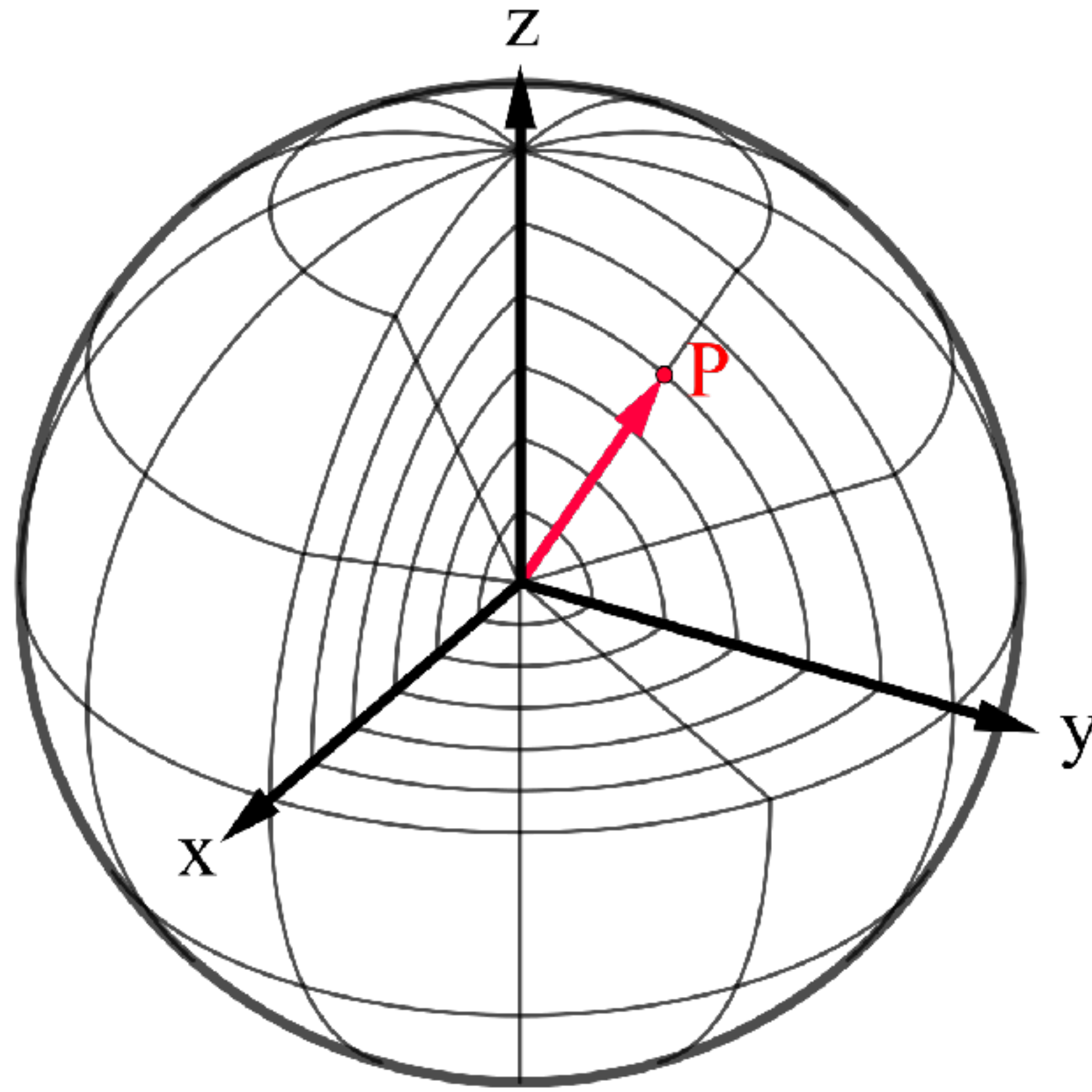


Liu et.al. "Swin Transformer: Hierarchical Vision Transformer using Shifted Windows", CVPR 2021



Grids on the sphere

Equiangular grids

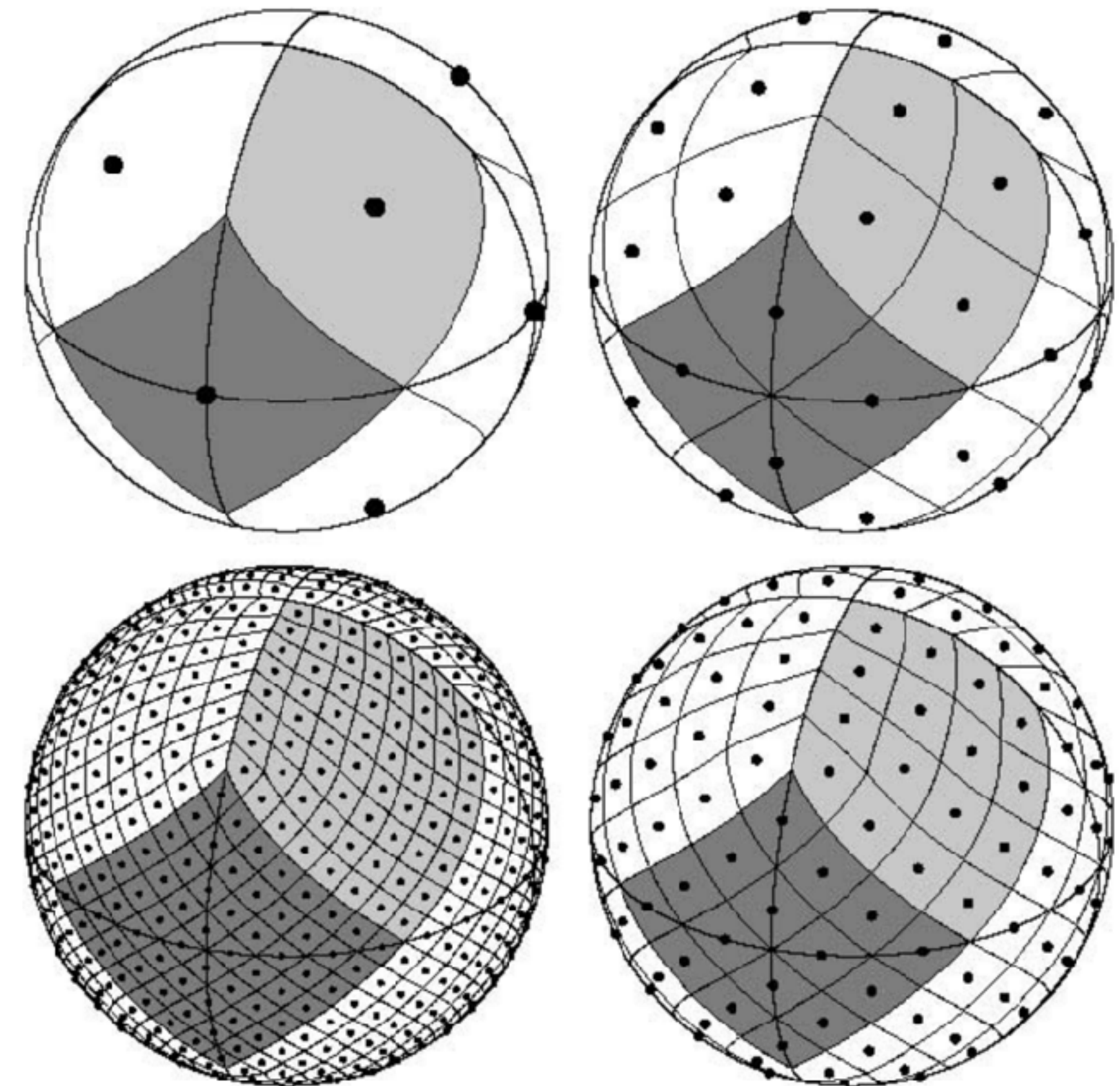


Grids on the sphere

Hierarchical Equal Area isoLatitude Pixelation (HEALPix)

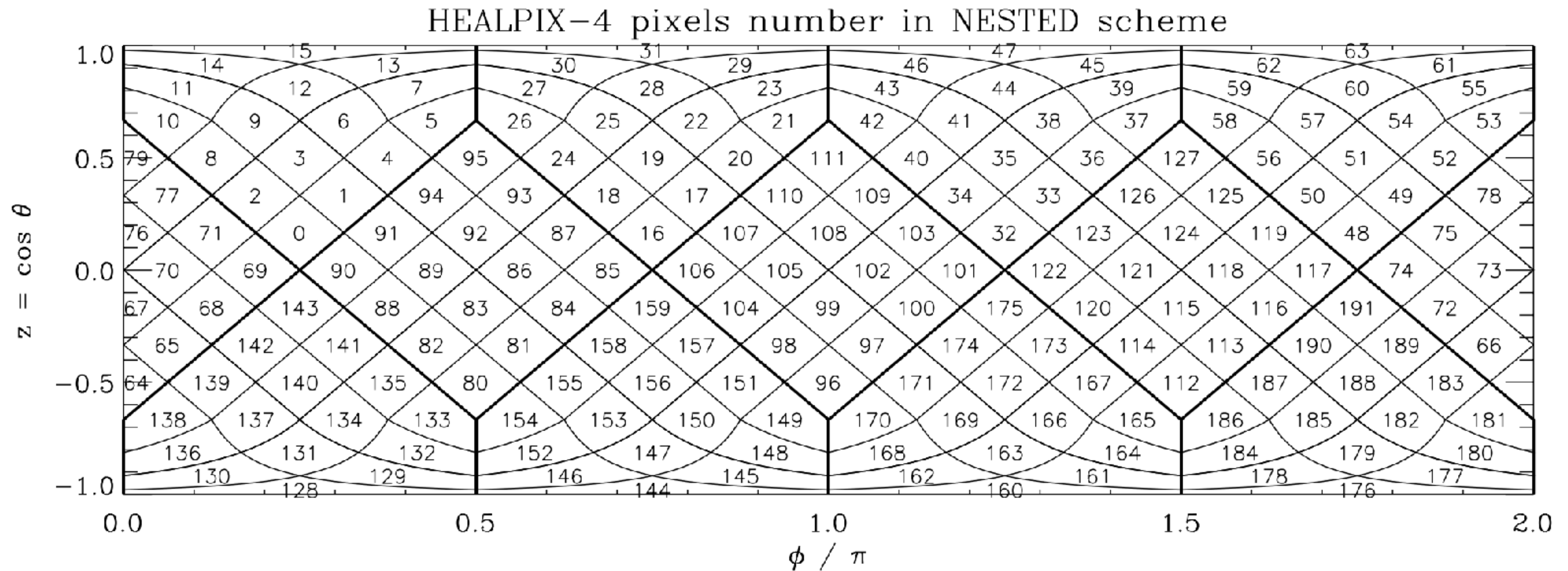
Equal area grid with pixels on isolatitude bands

Originally developed for cosmic microwave background data



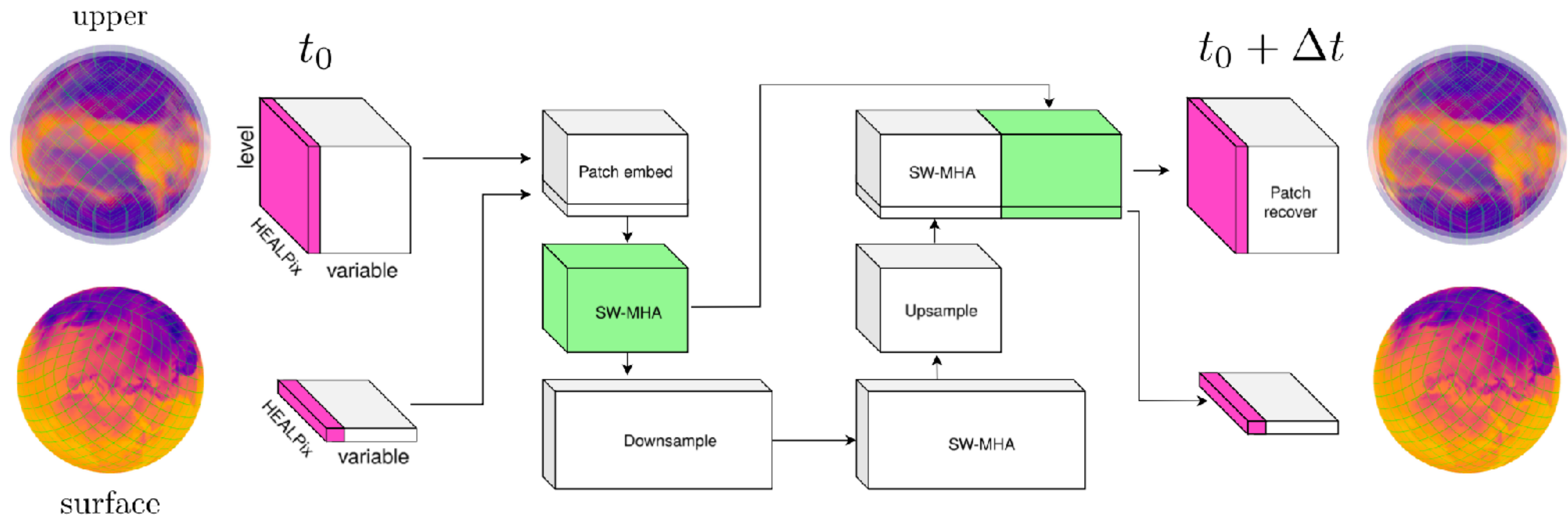
HEALPix

Nested indexing



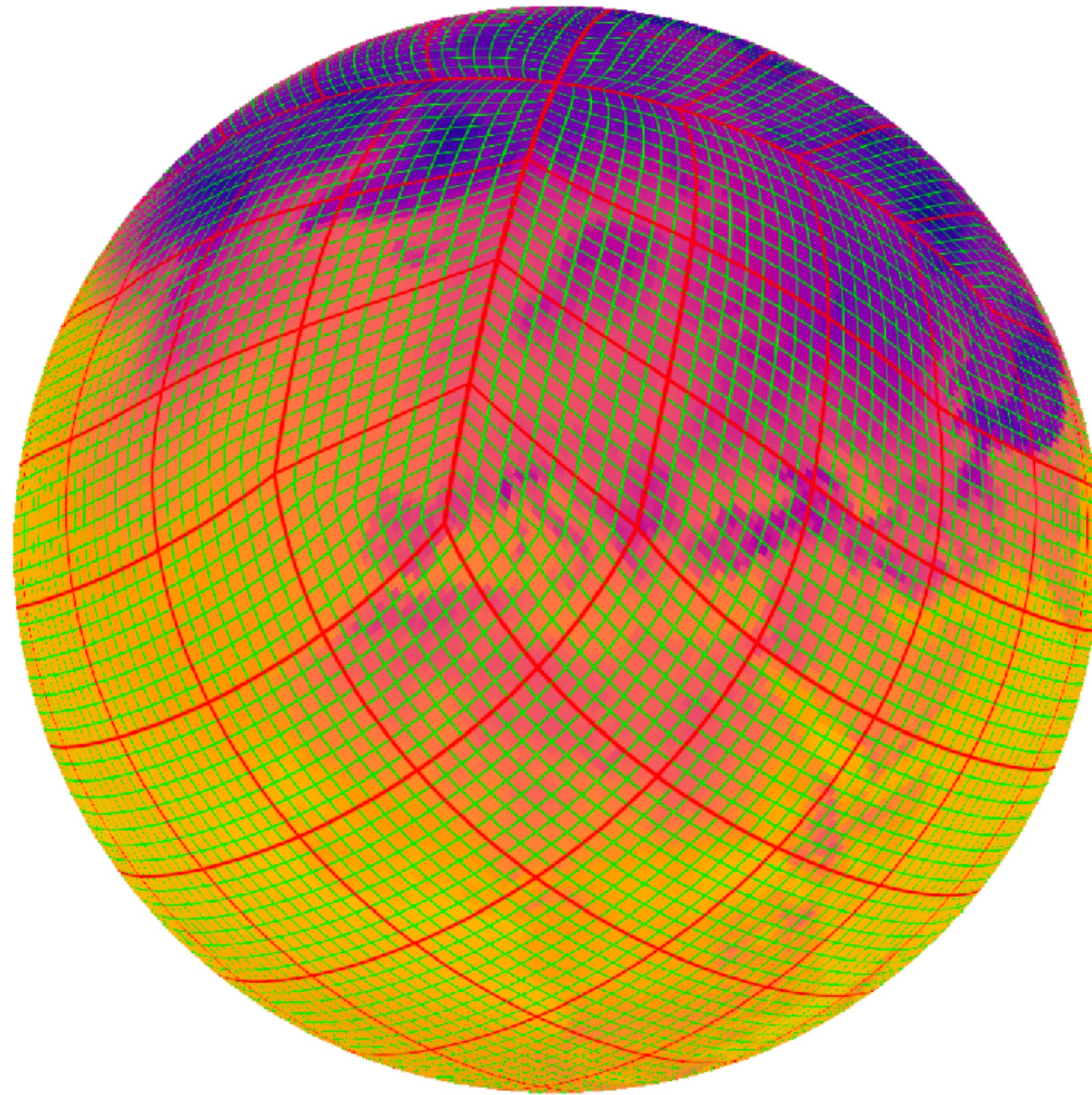


Pangu Equal ARea model



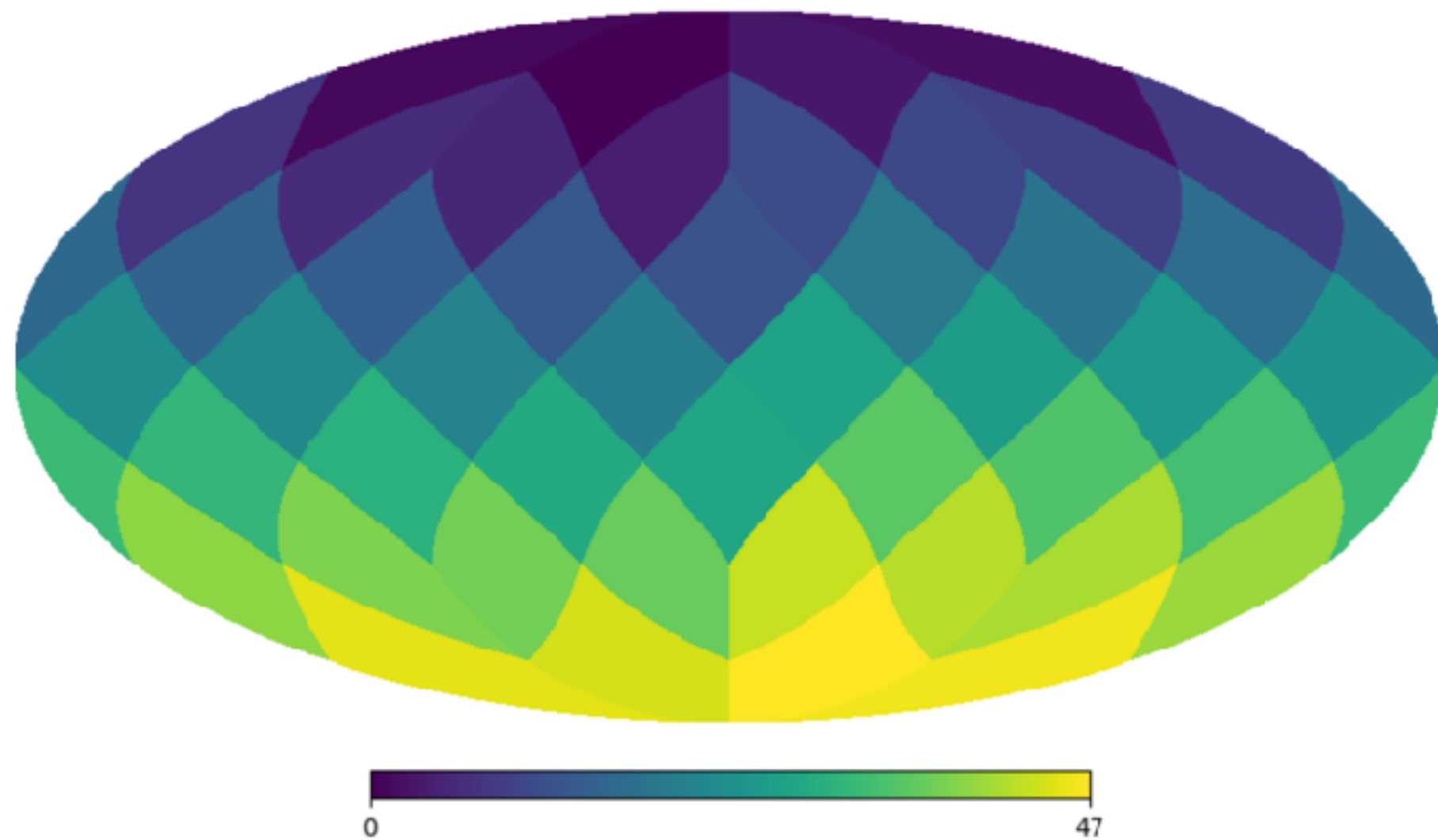
PEAR

Windows and patches



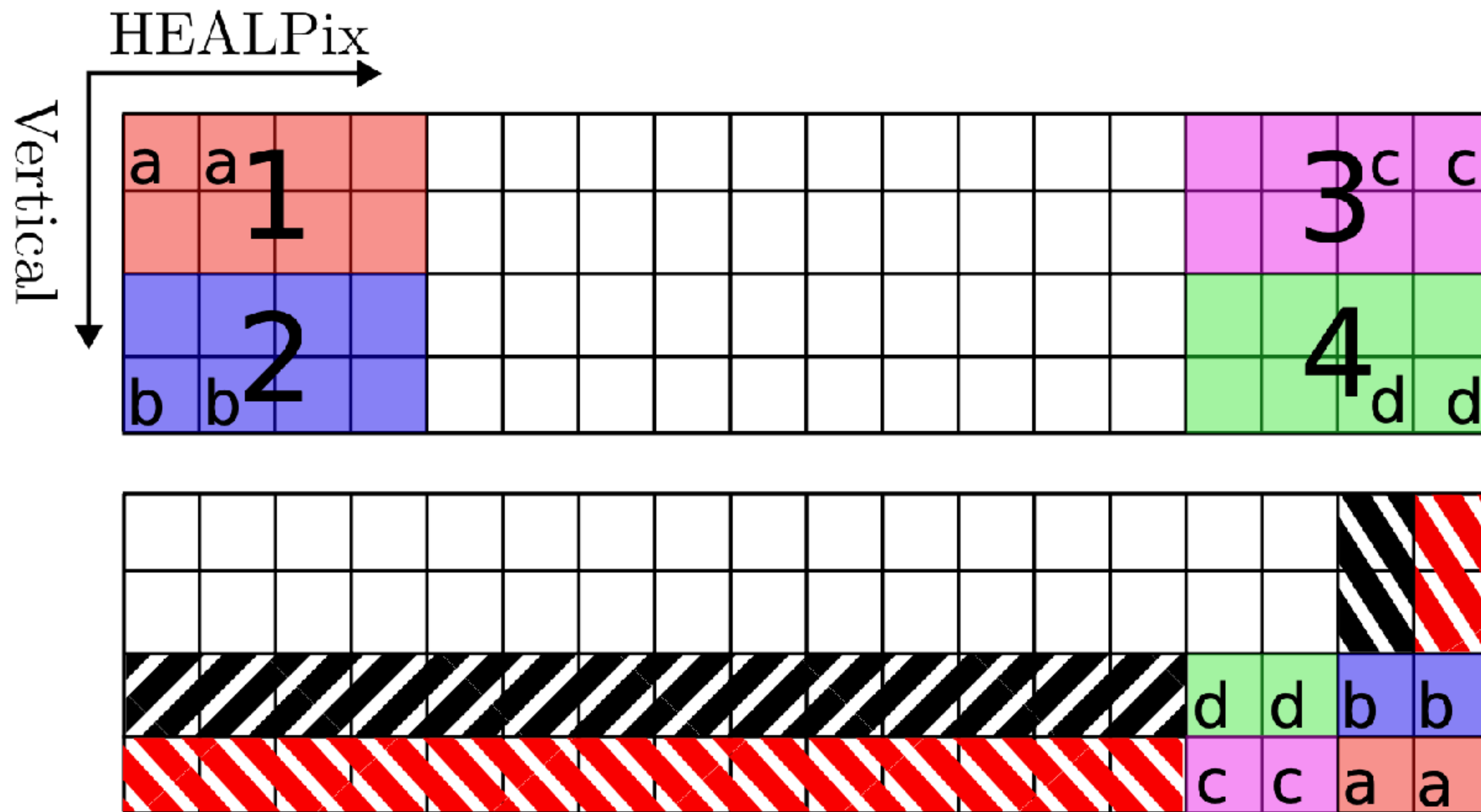
PEAR

Window shifting



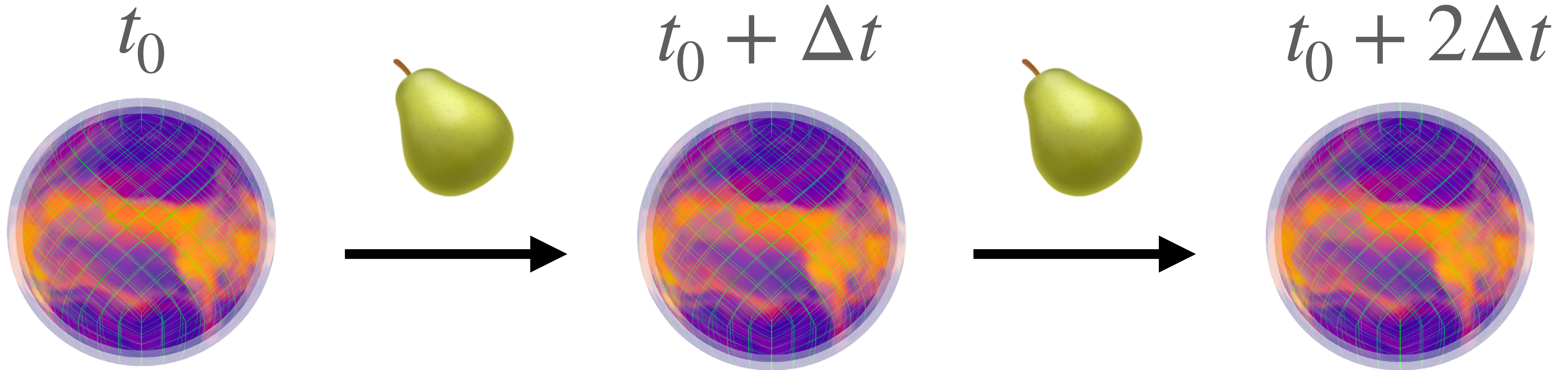
PEAR

Window shifting



PEAR

Longer forecast time



Evaluation

Metrics

HEALPix

$$\text{RMSE}(y, \hat{y}) = \sqrt{\frac{1}{12n_{\text{side}}^2} \sum_{i=0}^{12n_{\text{side}}^2} (y^i - \hat{y}^i)^2}$$

Equiangular

$$\text{RMSE}(v, t) = \sqrt{\frac{\sum_{i=1}^{N_{\text{lat}}} \sum_{j=1}^{N_{\text{lon}}} L(i) (\hat{\mathbf{A}}_{i,j,t}^v - \mathbf{A}_{i,j,t}^v)^2}{N_{\text{lat}} \times N_{\text{lon}}}}$$

$$L(i) = N_{\text{lat}} \times \frac{\cos \phi_i}{\sum_{i'=1}^{N_{\text{lat}}} \cos \phi_{i'}}$$

Evaluation

Metrics

Anomaly Correlation Coefficient (ACC)

$$\text{ACC}(y, \hat{y}) = \frac{\sum_i^{12n_{\text{side}}^2} \Delta y^i \Delta \hat{y}^i}{\sqrt{\left(\sum_i^{12n_{\text{side}}^2} (\Delta y^i)^2 \right) \left(\sum_i^{12n_{\text{side}}^2} (\Delta \hat{y}^i)^2 \right)}}$$

$$\Delta y = E_{\text{yearly}} [\hat{y}] - y$$

Correlation between deviation from “climate average” between data and model predictions

PEAR

Quantitative evaluation

Variable	Δt (days)	ACC		RMSE		unit
		PEAR	Pangu	PEAR	Pangu	
msl surface	1	0.984	0.979	84.6	97.9	Pa
	3	0.924	0.860	175	230	
	5	0.790	0.675	282	355	
t2m surface	1	0.886	0.876	0.792	0.846	K
	3	0.779	0.728	1.11	1.31	
	5	0.650	0.551	1.41	1.77	
u10 surface	1	0.930	0.923	1.05	1.1	$\frac{\text{m}}{\text{s}}$
	3	0.819	0.759	1.59	1.8	
	5	0.647	0.557	2.17	2.42	
v10 surface	1	0.932	0.922	1.07	1.13	$\frac{\text{m}}{\text{s}}$
	3	0.820	0.750	1.64	1.87	
	5	0.638	0.528	2.26	2.54	

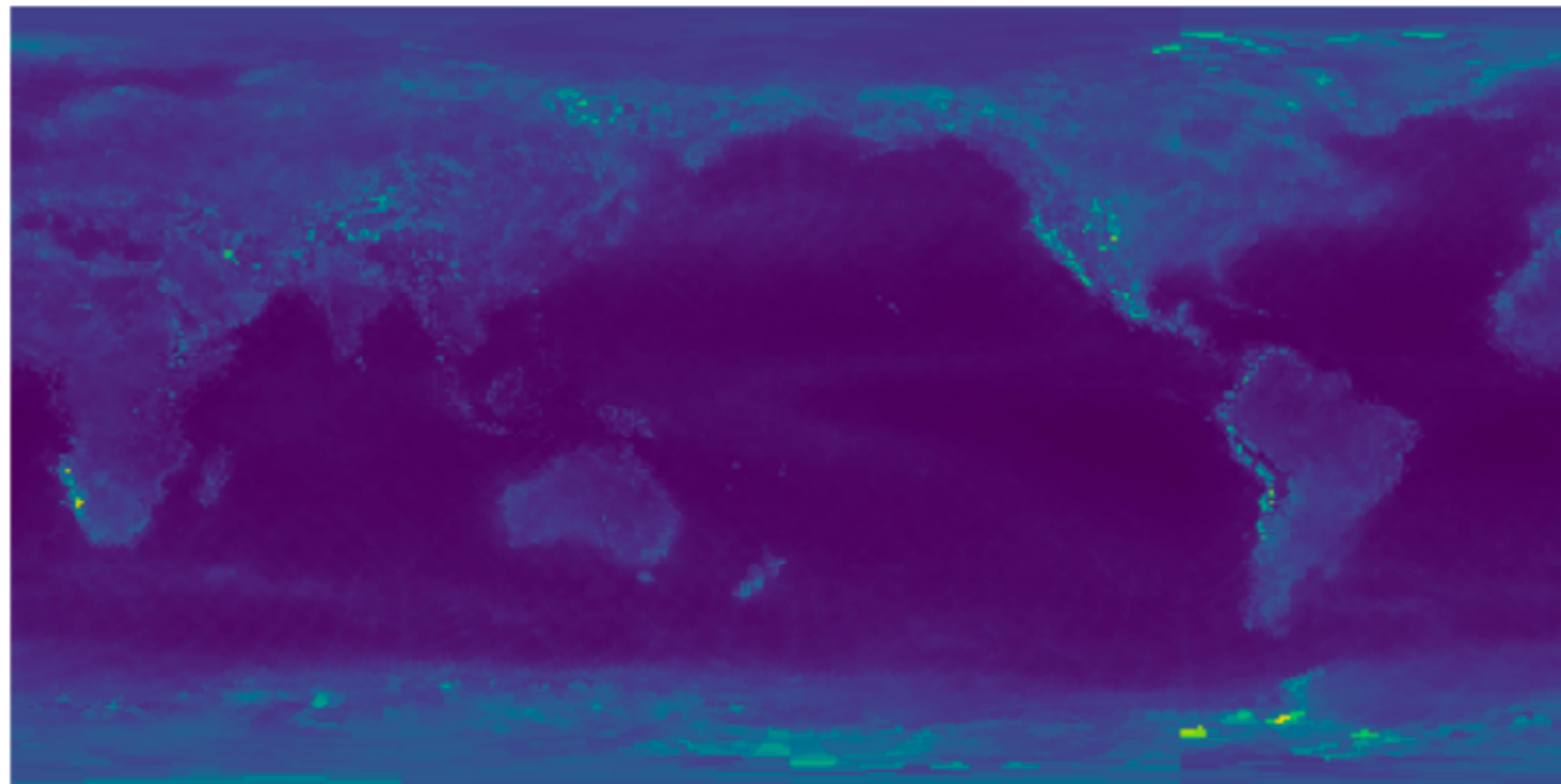
PEAR

Quantitative evaluation

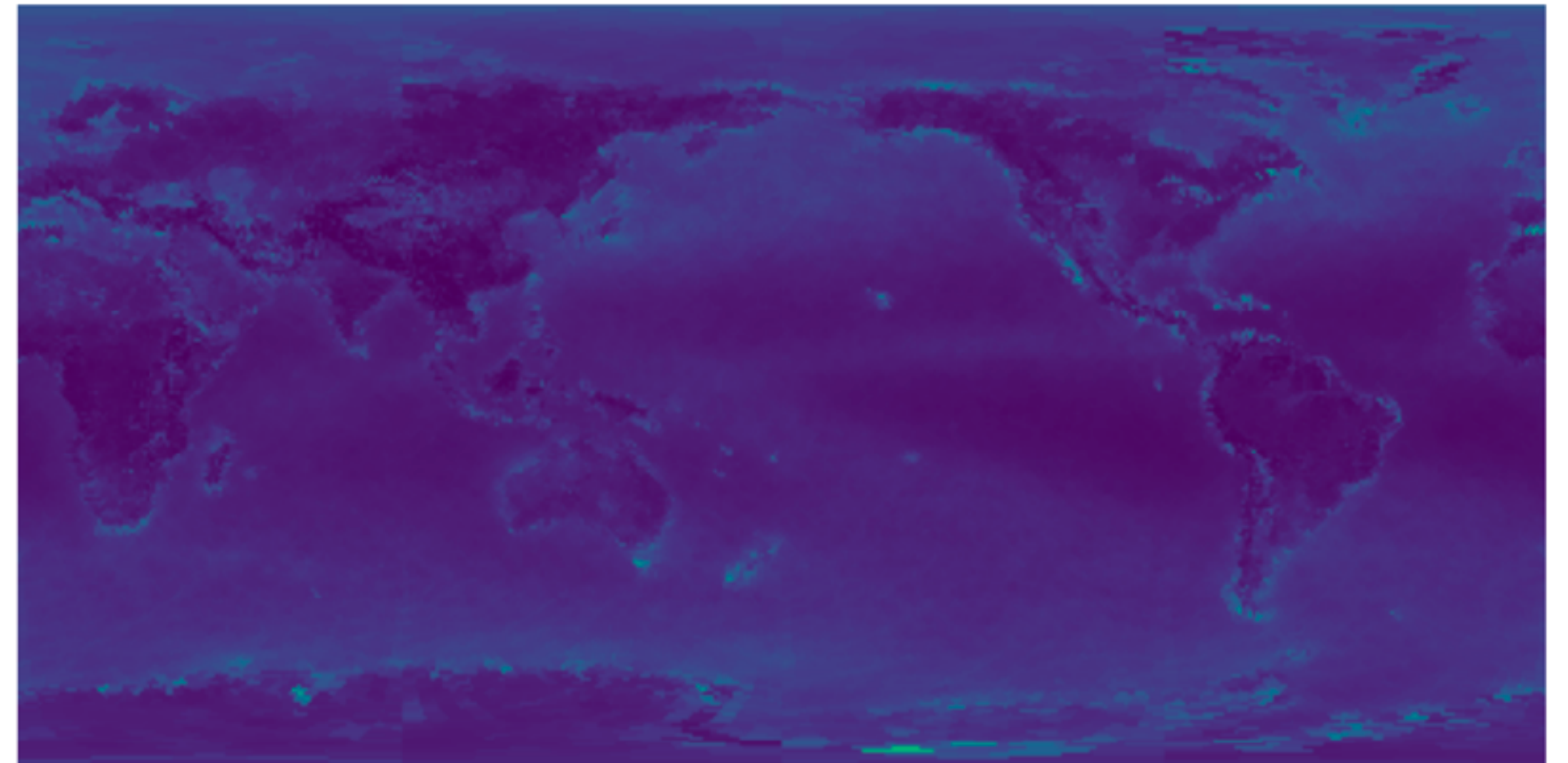
Variable	Δt (days)	ACC		RMSE		unit
		PEAR	Pangu	PEAR	Pangu	
q_{upper}	1	0.824	0.792	2.87×10^{-4}	3.1×10^{-4}	$\frac{\text{g}}{\text{kg}}$
	3	0.706	0.647	3.69×10^{-4}	4.09×10^{-4}	
	5	0.584	0.510	4.52×10^{-4}	5.03×10^{-4}	
t_{upper}	1	0.945	0.938	0.756	0.809	K
	3	0.877	0.822	1.11	1.3	
	5	0.763	0.664	1.51	1.77	
u_{upper}	1	0.944	0.937	2.04	2.17	$\frac{\text{m}}{\text{s}}$
	3	0.866	0.812	3.12	3.6	
	5	0.739	0.648	4.31	4.92	
v_{upper}	1	0.941	0.931	2.02	2.16	$\frac{\text{m}}{\text{s}}$
	3	0.859	0.787	3.07	3.61	
	5	0.717	0.595	4.26	4.92	
z_{upper}	1	0.991	0.985	79	98.8	gpm
	3	0.949	0.889	175	251	
	5	0.850	0.734	291	393	

PEAR

spatially resolved prediction errors



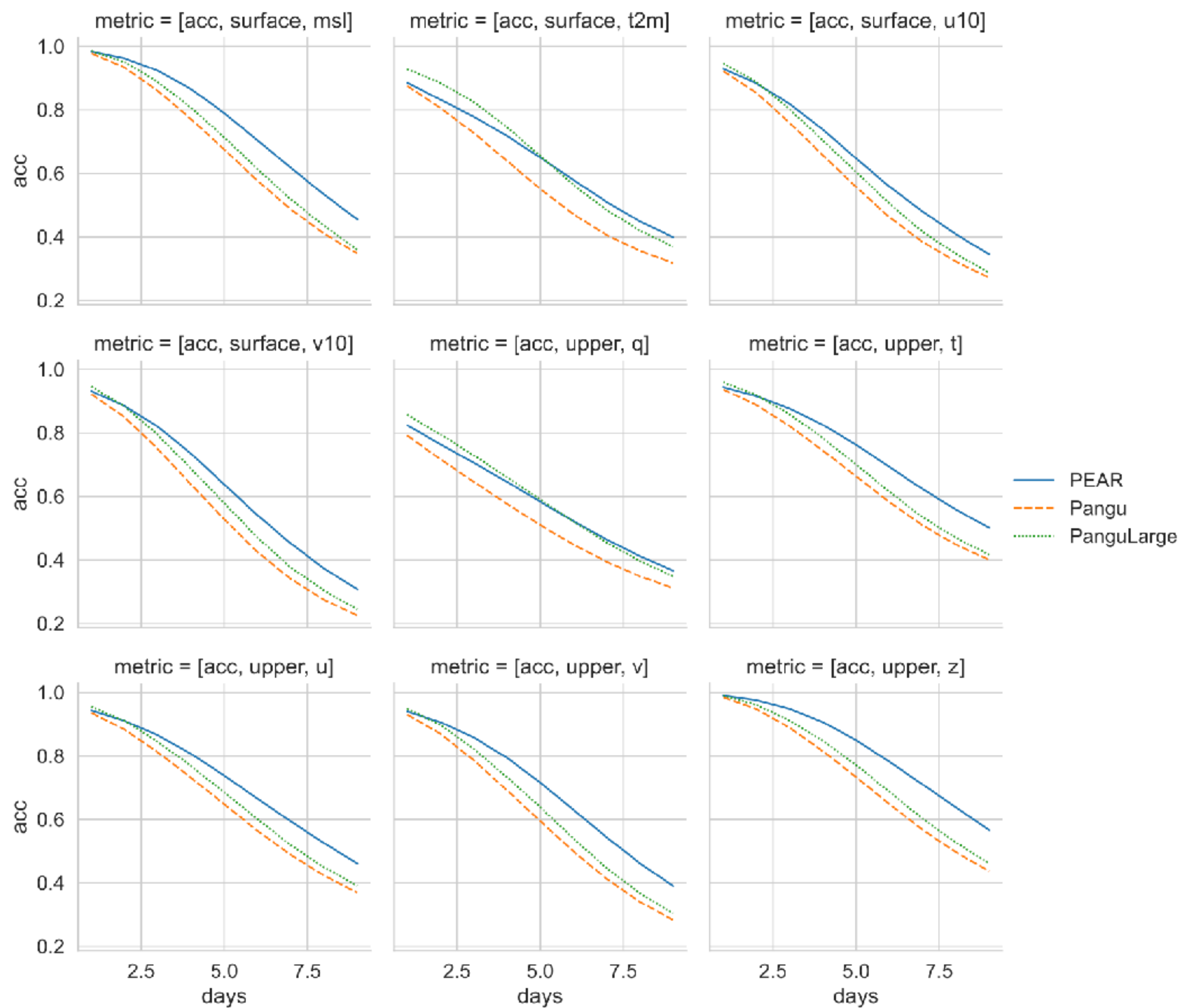
temperature
error (RMSE)



wind speed
error (RMSE)

PEAR

ACC



PEAR

compute

Model	Trainable parameters (M)	Inference time (ms)
PEAR	4.3	17 ± 0.12
Pangu	11.4	25 ± 0.05
Pangu-Large	33.7	54 ± 0.02

Outlook

Azimuthal equivariance



Full ERA5 dataset

Increased resolution



gapindnns.github.io

PEAR

Data shape

Tensor	Shape
surface	(49152, 4)
upper	(49152, 13, 5)