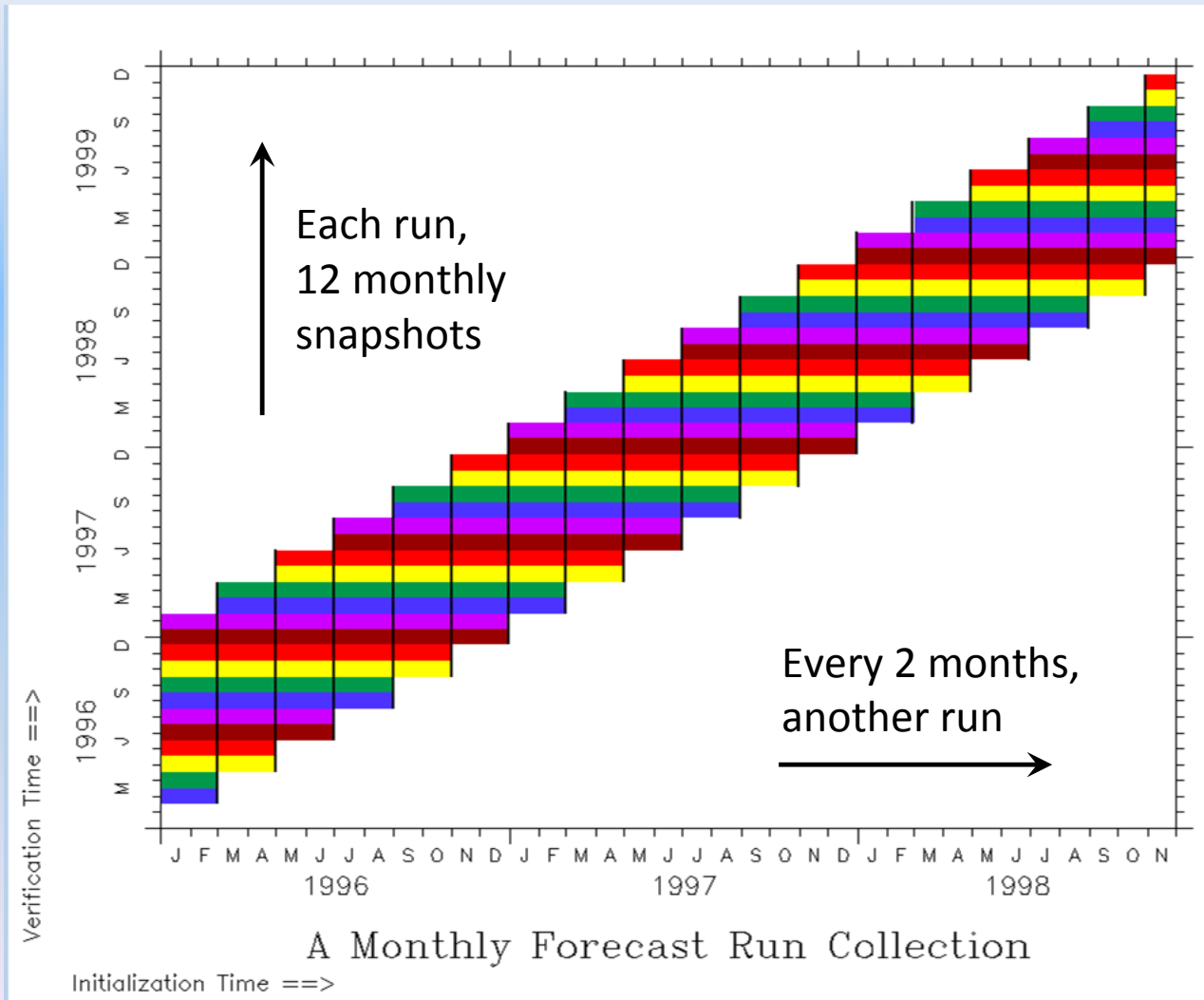


# Working with Forecast Models in PyFerret and Ferret

June 2015

A plot to illustrate the data output by a Forecast Model Run Collection (FMRC). Every colored cell is a model snapshot.



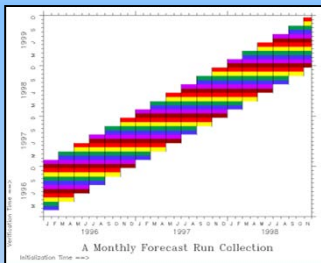
Each model output (starting at one initialization time) is a “dataset”. How to aggregate these many individual datasets into a single dataset of higher dimensionality?

- use the Ferret command  
**DEFINE AGGREGATION/F**  
or
- serve the collection via OPeNDAP using Unidata’s THREDDS Data Server (TDS)

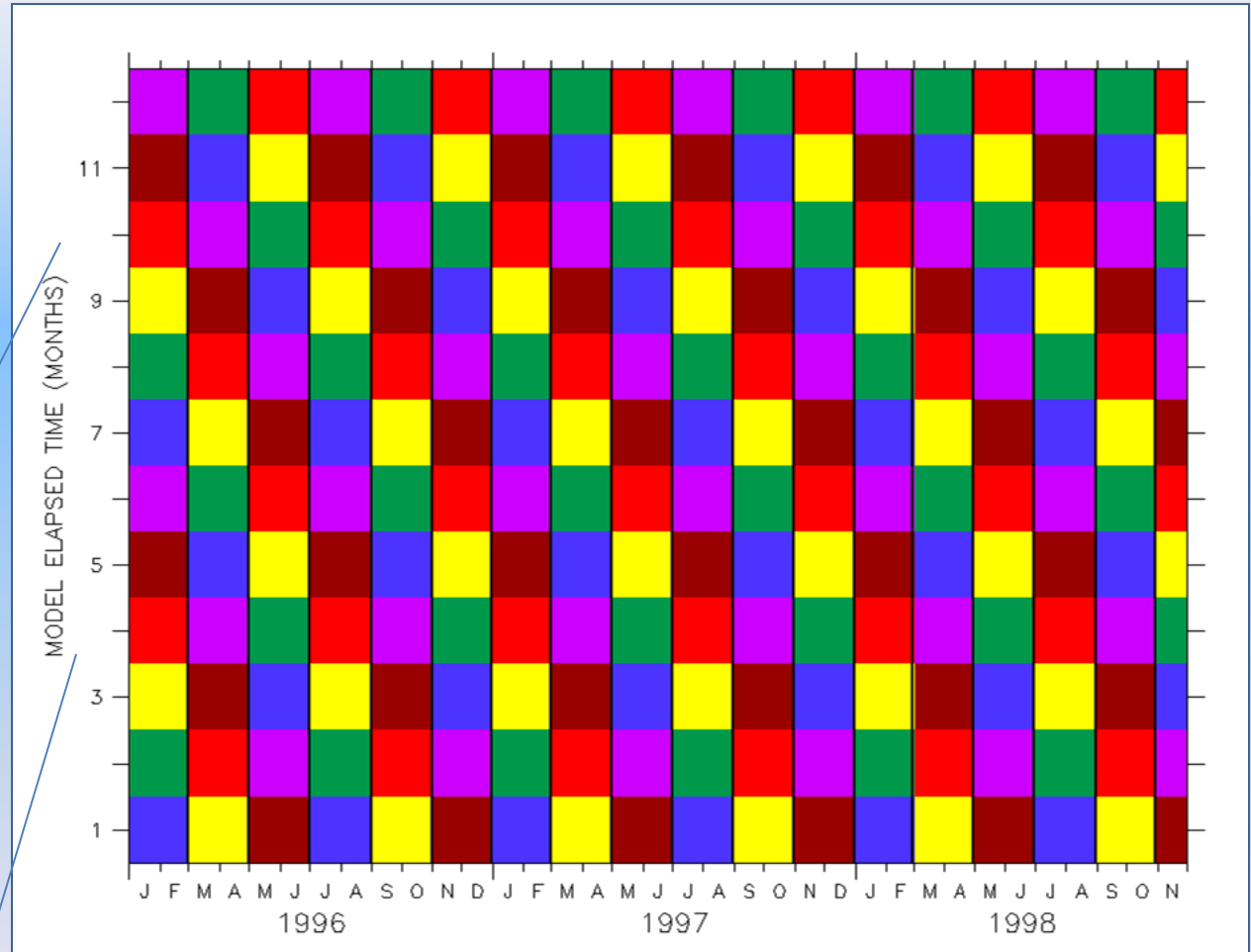
many 4D files → become a single 5D dataset

# Native (“compact”) form of FMRC

as run ...



MODEL ELAPSED TIME (MONTHS)



In the native, compact form the time step values are a 2D variable. Note that the same forecasted date/time occurs at lags in multiple forecast runs.

		Forecast Series Axis →						
Model Run Time Axis ↓		<u>RUN 1</u>	<u>RUN 2</u>	<u>RUN3</u>	<u>RUN4</u>	<u>RUN5</u>	<u>RUN6</u>	<u>RUN7</u>
	1	744.	2184.	3648.	5112.	6576.	8040.	9528.
	2	1440.	2904.	4368.	5856.	7320.	8784.	10200.
	3	2184.	3648.	5112.	6576.	8040.	9528.	10944.
	4	2904.	4368.	5856.	7320.	8784.	10200.	11664.
	5	3648.	5112.	6576.	8040.	9528.	10944.	12408.
	6	4368.	5856.	7320.	8784.	10200.	11664.	13128.
	7	5112.	6576.	8040.	9528.	10944.	12408.	13872.
	8	5856.	7320.	8784.	10200.	11664.	13128.	14616.
	9	6576.	8040.	9528.	10944.	12408.	13872.	15336.
	10	7320.	8784.	10200.	11664.	13128.	14616.	16080.
	11	8040.	9528.	10944.	12408.	13872.	15336.	16800.
	12	8784.	10200.	11664.	13128.	14616.	16080.	17544.

# netCDF CDL markup language showing how 'auxiliary coordinates' are defined in a CF file

```
dimensions:
  ntime = 10
  lev = 15;    // number of levels
float temp(ntime,lev);
  temp: long_name = "temperature";
  temp: coordinates = "depth";
float depth(lev);
  depth: units = "meters";
  depth: axis = "Z";
```

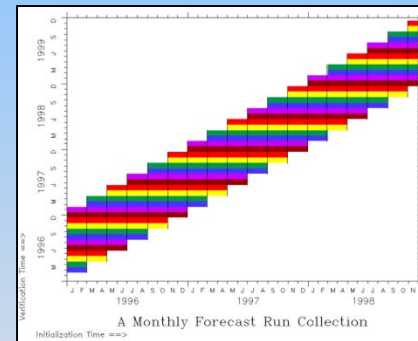
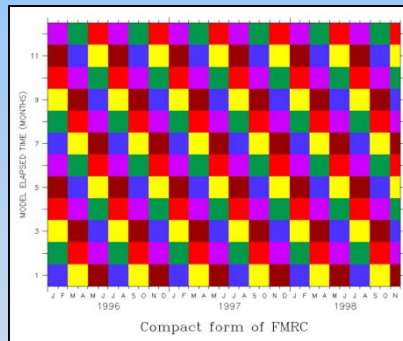
← depth in a layered model

2D times in  
an an FMRC →

```
dimensions:
  ntime = 10
  nrun  = 15;    // number of runs
float temp(nrun,ntime);
  temp: long_name = "temperature";
  temp: coordinates = "times";
float times(nrun, ntime);
  times: units = "days since 01-01-1996";
```

# Ferret syntax: how to regrid using an auxiliary coordinate

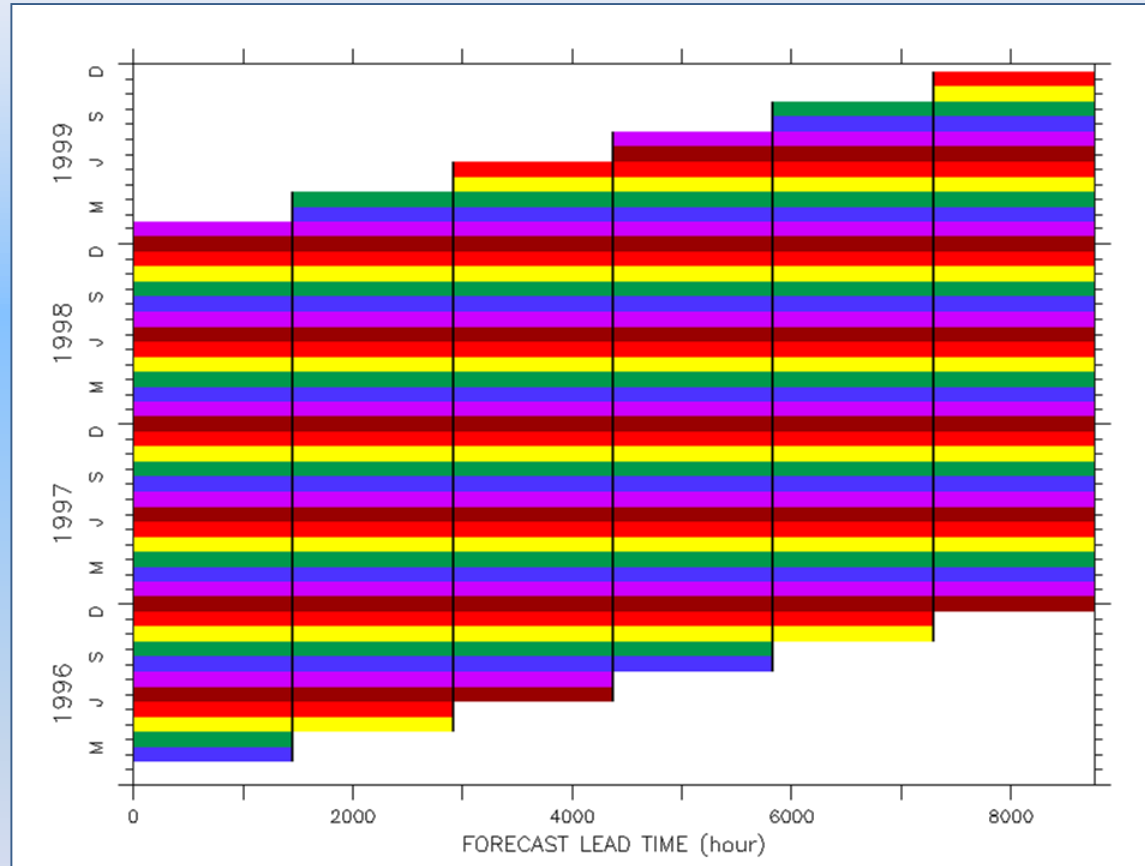
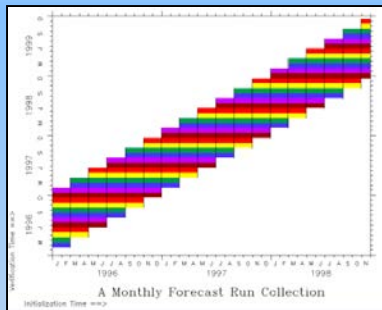
- Depth: `temp[ GZ(density) = My_density_axis]`
- Time2d: `temp[ GT(tf_times)= TF_CAL_T @FMRC]`



Regridding in a manner analogous to a depth-to-density transformation, we convert the FMRC compact form into diagonal form

# Forecast Skill Visualization of an FMRC

Our  
Goal!



Similarly also replace the Forecast time axis  
with a lead (lag) time axis

`temp[GT(tf_times)=TF_CAL_T,GF(tf_times)=TF_LAG_F]`



That's the background.

Now, what does it feel like to analyze  
forecast model outputs using Ferret?

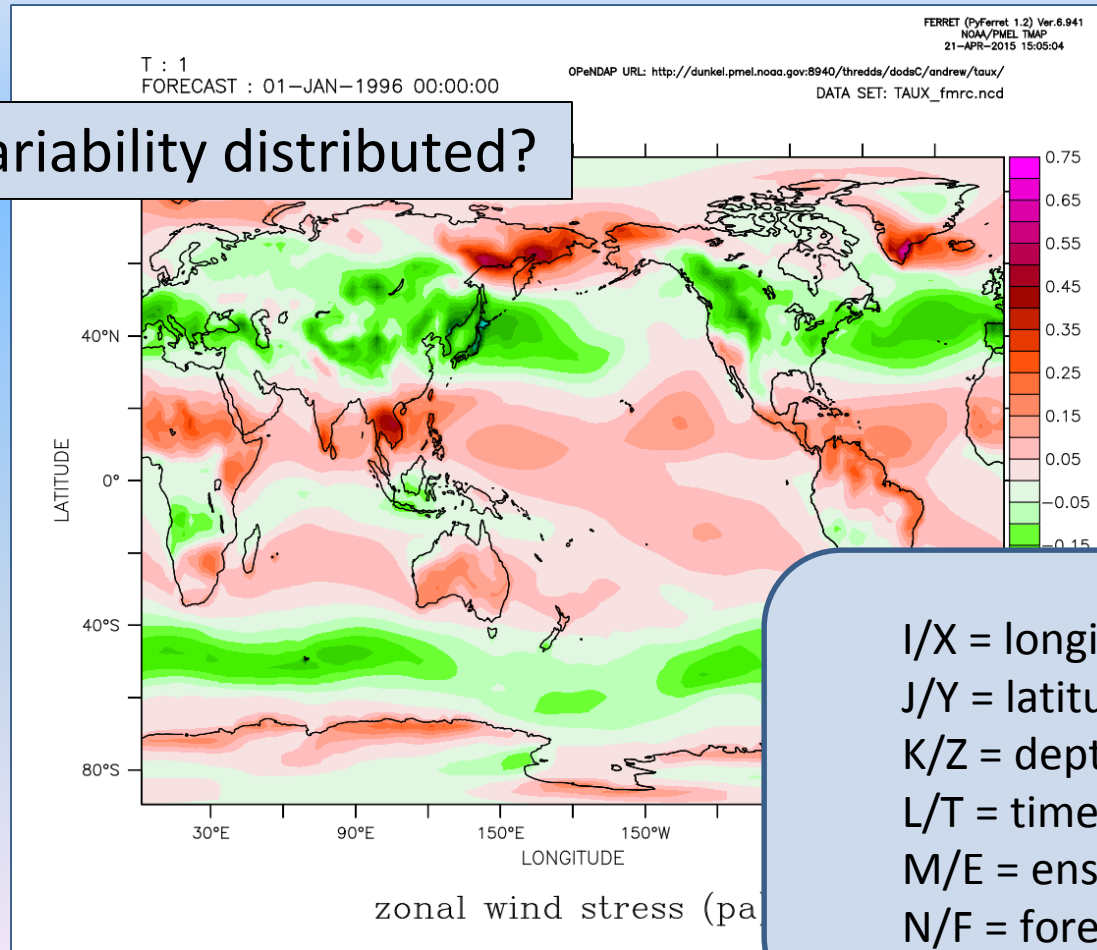
Start up the dataset as you would any other. “N” is the forecast index

```
> pyferret
```

yes? USE “[http://server/path/myfiles/TAUX\\_fmrc.ncd](http://server/path/myfiles/TAUX_fmrc.ncd)” ! aggregation via TDS

yes? FILL tau\_x[L=1,N=1]

How is the variability distributed?



I/X = longitude

J/Y = latitude

K/Z = depth

L/T = time

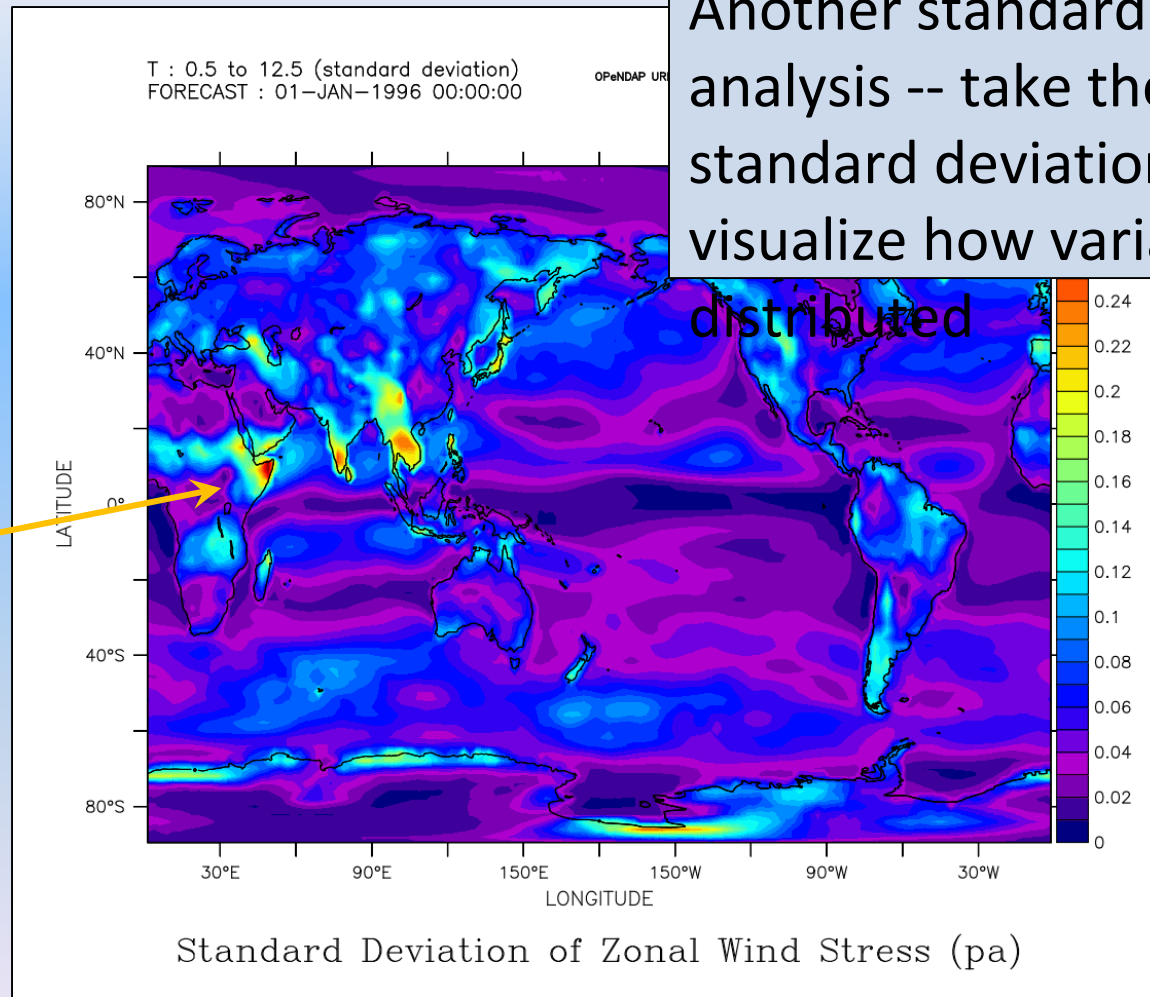
M/E = ensemble member

N/F = forecast run

yes? FILL/N=1 tau\_x[L=@std]

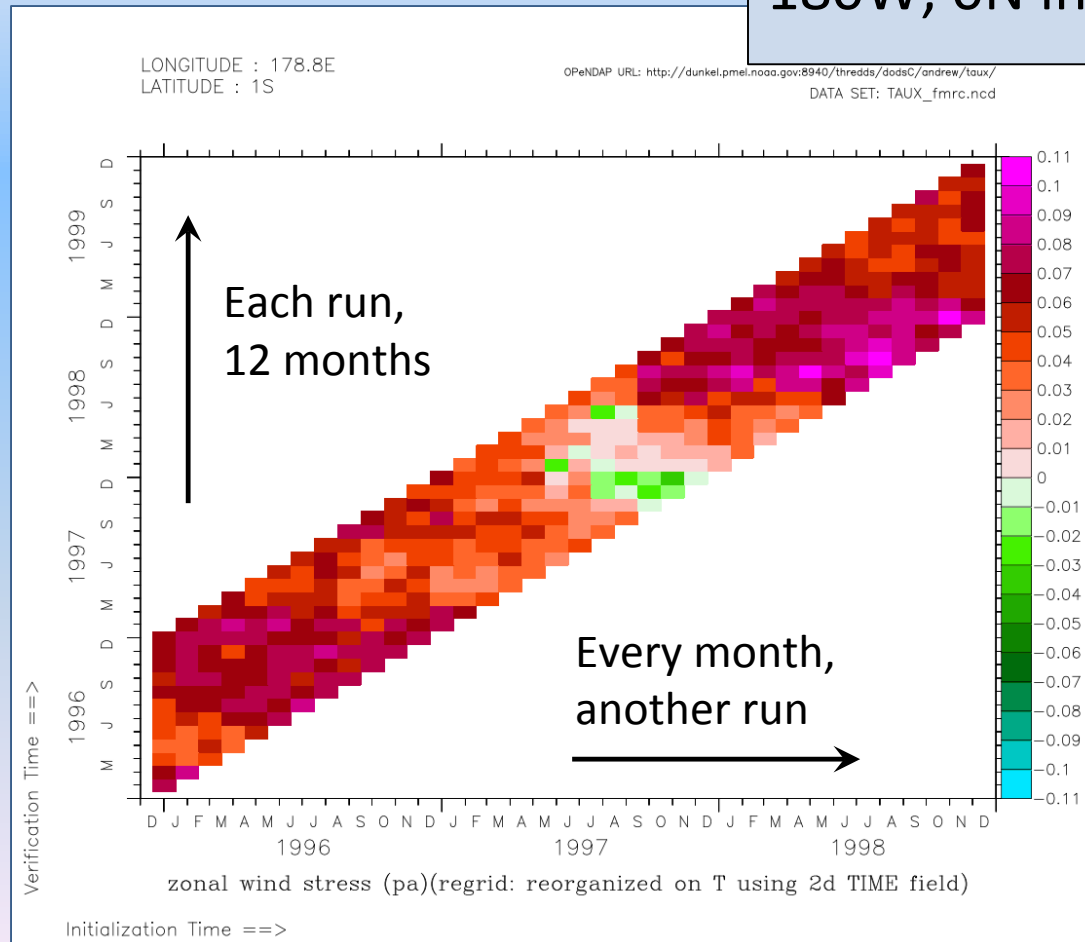
Another standard Ferret analysis -- take the time-standard deviation to visualize how variability is distributed

high variability

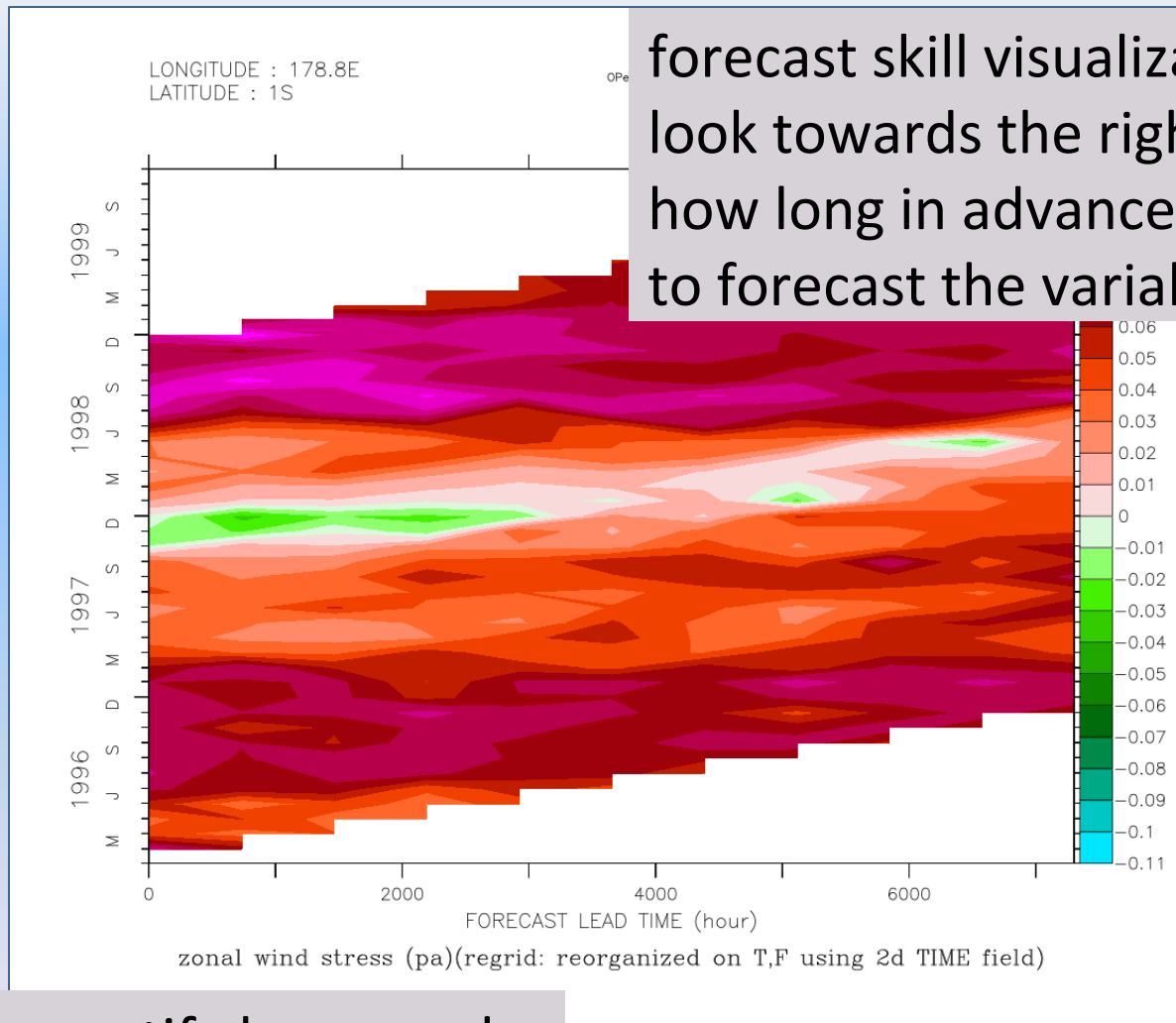


SHADE/X=180/Y=0 tau\_x[GT(times)=TF\_CAL\_T]

View the forecast series at  
180W, 0N in 'diagonal' form



FILL/X=180/Y=0 tau\_x[GT(times)=TF\_CAL\_T,GF(times)=TF\_LAG\_F]



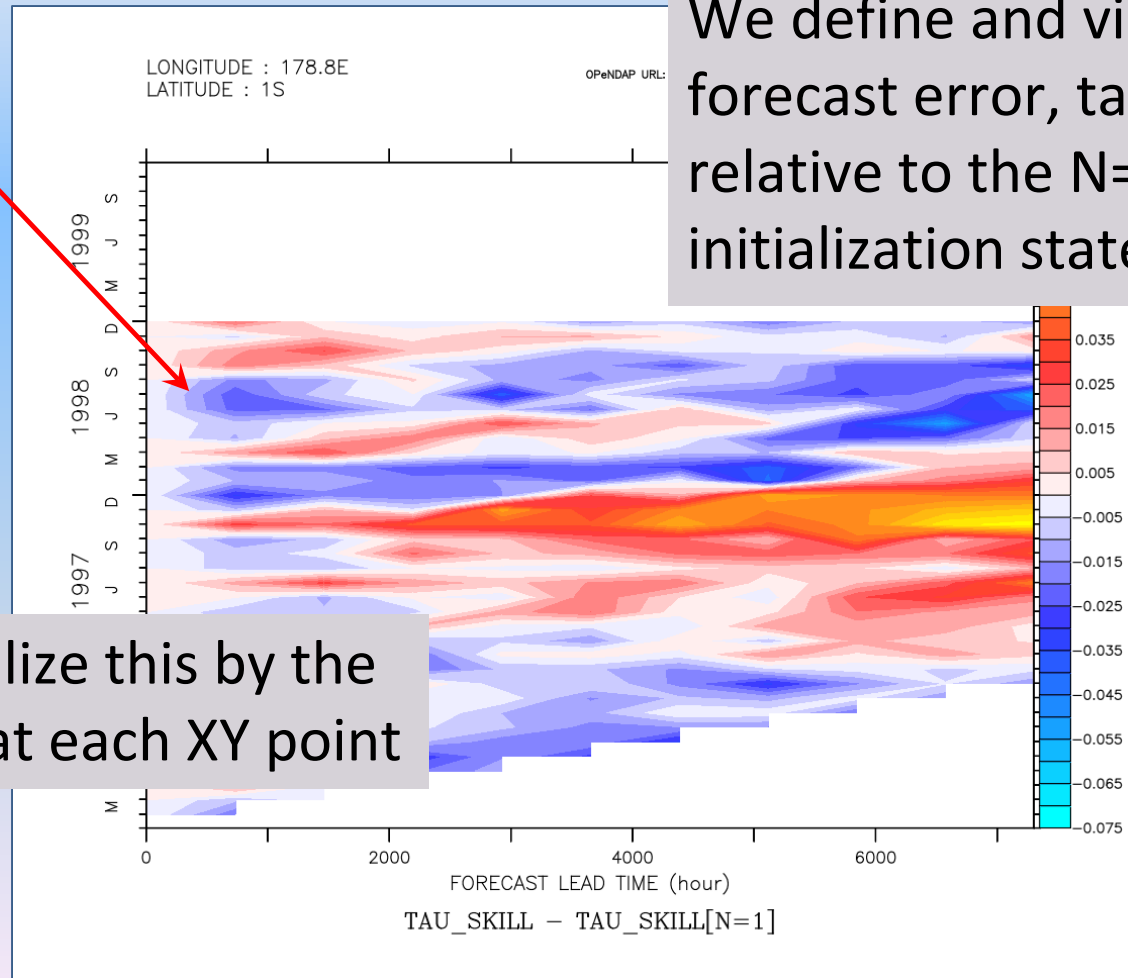
Can we quantify how good our forecasts were?

```
LET tau_tf = tau_x[gt(times)=TF_CAL_T,gf(times)=TF_LAG_F]
LET tau_fe = tau_tf - tau_tf[N=1]
FILL/X=180/Y=0 tau_fe
```

Is there a  
season cycle  
to the errors?

We define and view the  
forecast error,  $\tau_{fe}$ ,  
relative to the N=1  
initialization state

Lets normalize this by the  
variability at each XY point

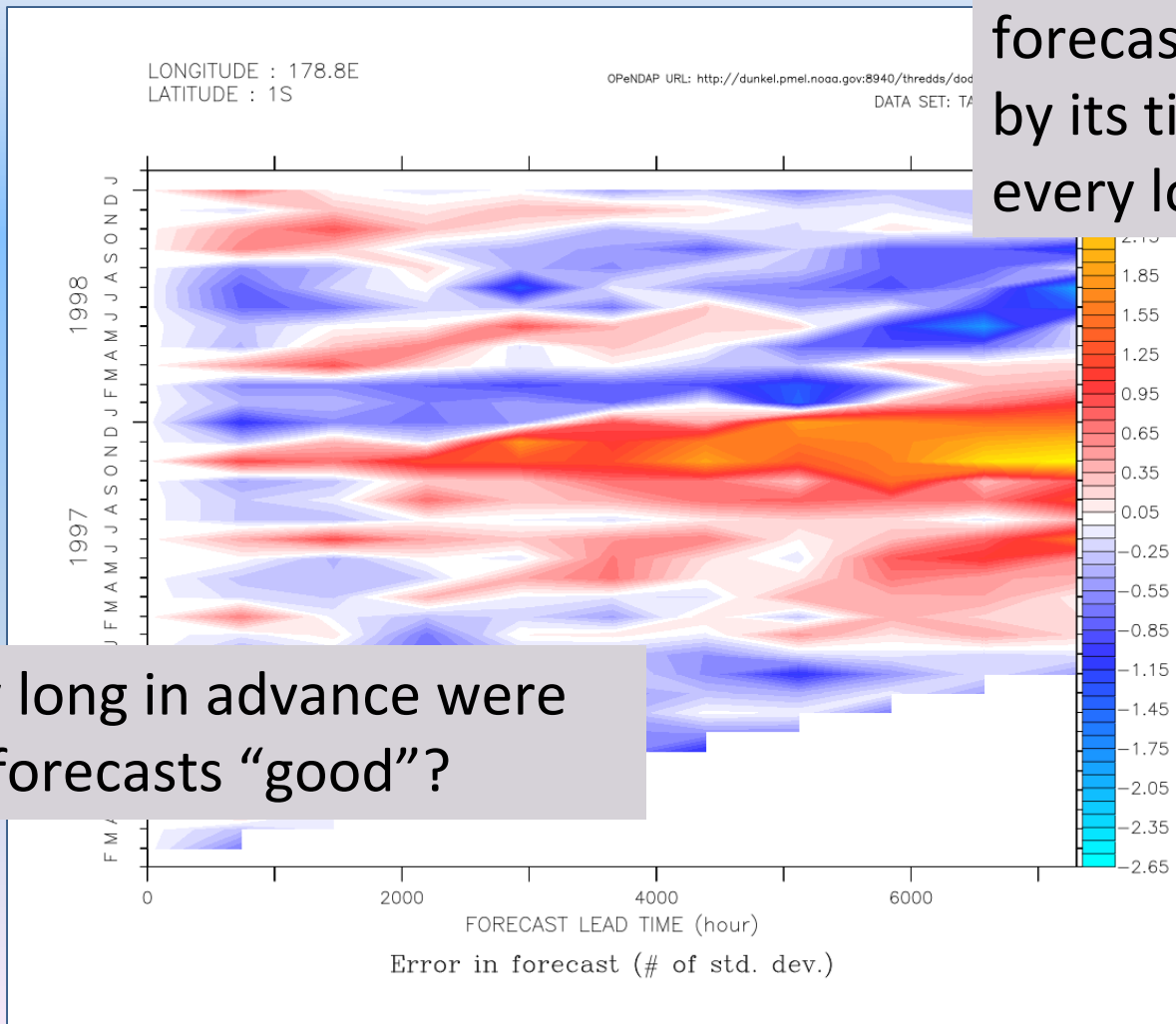


LET tau\_stddev = tau\_tf[N=1,L=@std]

LET/TITLE=... tau\_nfe = tau\_fe/tau\_stddev

FILL/Y=180/Y=0/T=... tau\_nfe

We normalize the forecast error, dividing by its time-std dev at every location



How long in advance were our forecasts “good”?

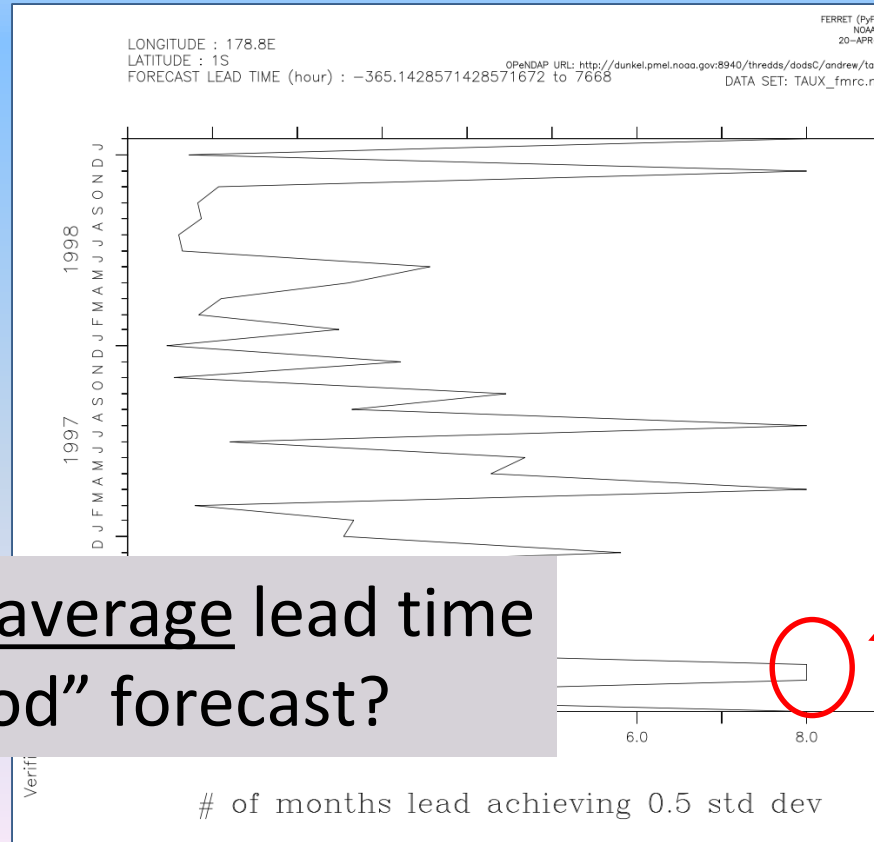
Lets say “good” == abs. val. of error within 0.5 std dev of final

LET tau\_abs = ABS(tau\_nfe) value

LET tau\_fcst\_lead = tau\_abs[F=@loc:.5] / 730

LET/TITLE=“...” tau\_skill = MISSING(tau\_fcst\_lead,8) ! cap at 8 mo

PLOT/X=180/Y=0 tau\_skill



Lead time (months)  
for achieving “good”  
forecast error

What is the average lead time  
to get a “good” forecast?

cap



The average lead time at which the forecast  
becomes accurate to 0.5 std dev  
at X=180, Y=0

```
LIST /X=180/Y=0 tau_skill[L=@ave]
```

```
VARIABLE : # of months lead achieving 0.5 std dev
```

```
FILENAME : TAUX_fmrc.ncd
```

```
FILEPATH : http://dunkel.pmel.noaa.gov:8940/thredds/dodsC/andrew/taux/
```

```
LONGITUDE: 178.8E
```

```
LATITUDE : 1S
```

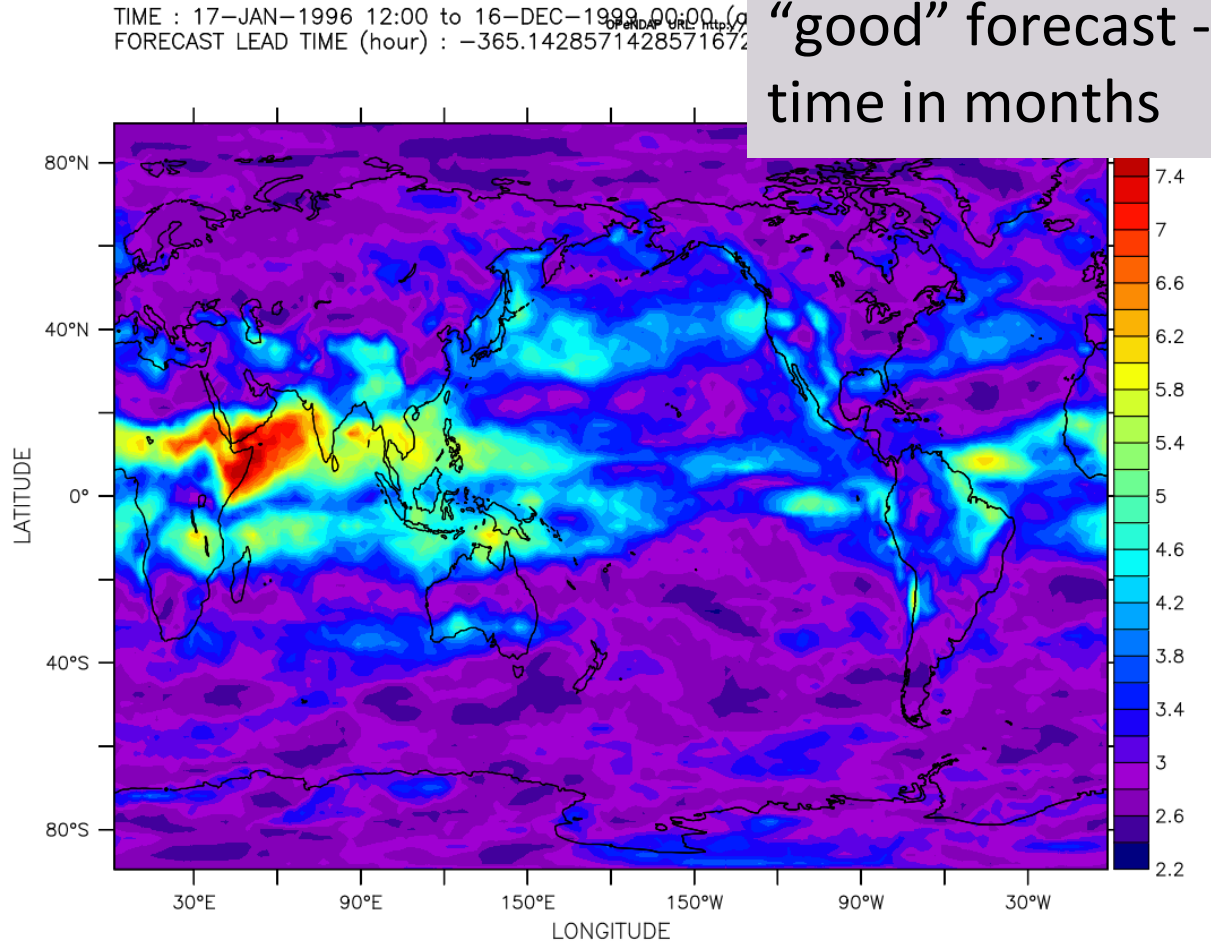
```
TIME      : 17-JAN-1996 12:00 to 16-DEC-1999 00:00 (averaged)
```

4.165

Now lets see how our skill is distributed globally

yes? FILL tau\_skill[L=@ave]

Global distribution of  
“good” forecast -- lead  
time in months



# of months lead achieving 0.5 std dev