



**The Weather  
and Climate**

Emergent Laws and Multifractal Cascades

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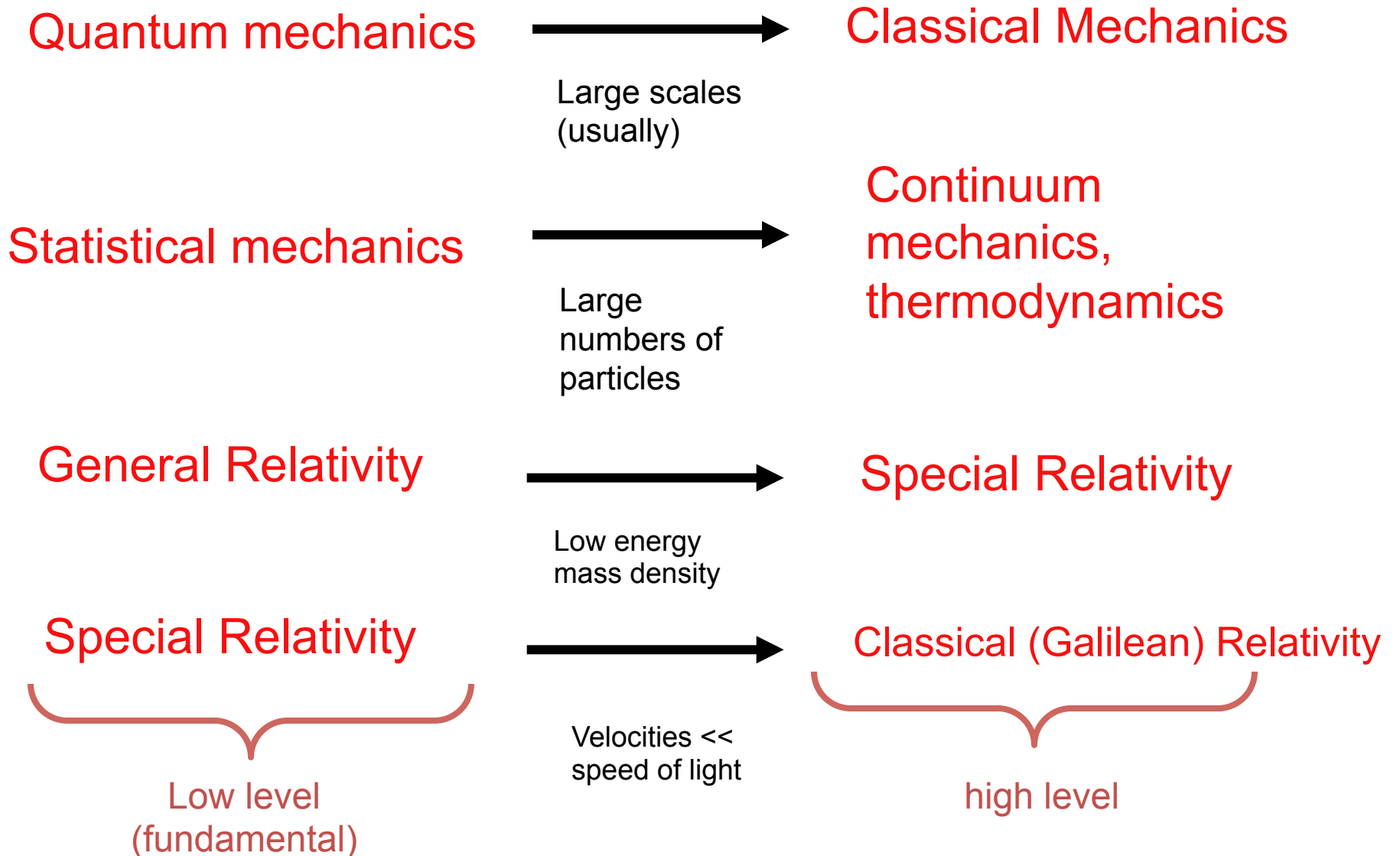
CAMBRIDGE

Emergent atmospheric laws:  
Why the weather and the climate  
are not what you expect

University of Reading,  
15 April, 12:00-13:00, 2013

**S. Lovejoy, McGill**

# The Emergence of physical laws



# The emergence of atmospheric dynamics (Classical)

Continuum mechanics

Low level  
(fundamental)

Large Re



Laws of turbulence

Classical:

Richardson, Kolmogorov,  
Corrsin, Obukhov, Bolgiano

High level

$$\Delta v(\underline{\Delta r}) = \varphi |\underline{\Delta r}|^H$$

e.g. Kolmogorov  $\varphi = \varepsilon^{1/3}, H = 1/3$

## Vortices in strongly turbulent fluid

(M. Wiczek, numerical simulation, 2010)



- a)  $|\underline{\Delta r}| \ll 100m$
- b) isotropic
- c)  $\varphi \approx \text{constant}$ , quasi Gaussian

# Emergence of Atmospheric laws (Modern)

$$\text{Fluctuations} \approx (\text{turbulent flux}) \times (\text{scale})^H$$

Differences,  
tendencies,  
wavelet  
coefficients

Cascading  
Turbulent flux

Anisotropic  
Space-time  
Scale function

Fluctuation  
/conservation  
exponent

Fourier domain:

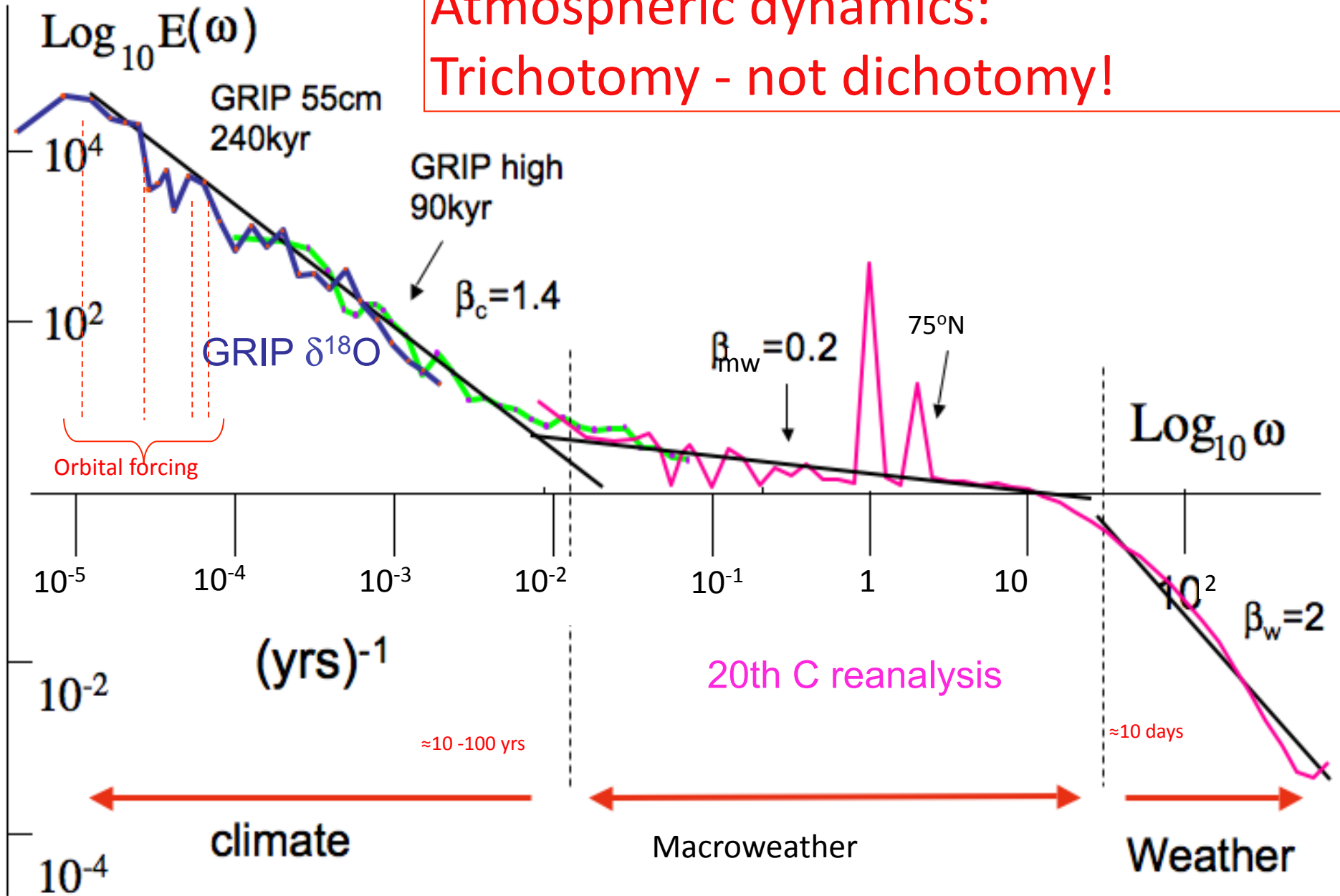
$$\left( \frac{\text{Variance}_{\text{observables}}}{\text{wavenumber}} \right) = \left( \frac{\text{Variance}_{\text{flux}}}{\text{wavenumber}} \right) (\text{wavenumber})^{-2H}$$

$$= (\text{wavenumber})^{-\beta}$$

Space:  $E(k) \approx k^{-\beta}$

Time:  $E(\omega) \approx \omega^{-\beta}$

# Atmospheric dynamics: Trichotomy - not dichotomy!



Two data sources only GRIP, 20CR

# Three regimes: three types of variability

Temperature

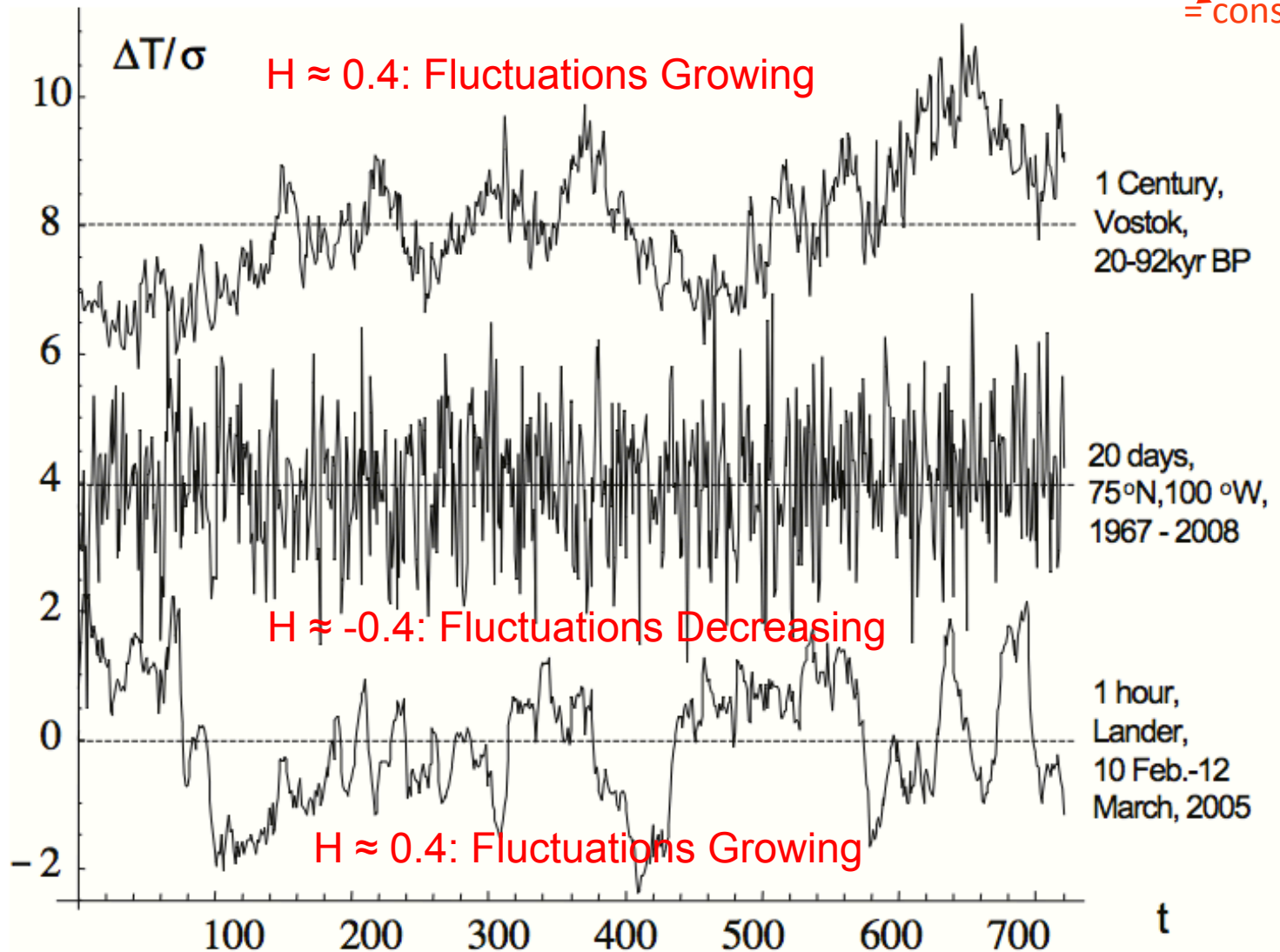
$$\langle \Delta I \rangle = \langle \phi \rangle \Delta t^H$$

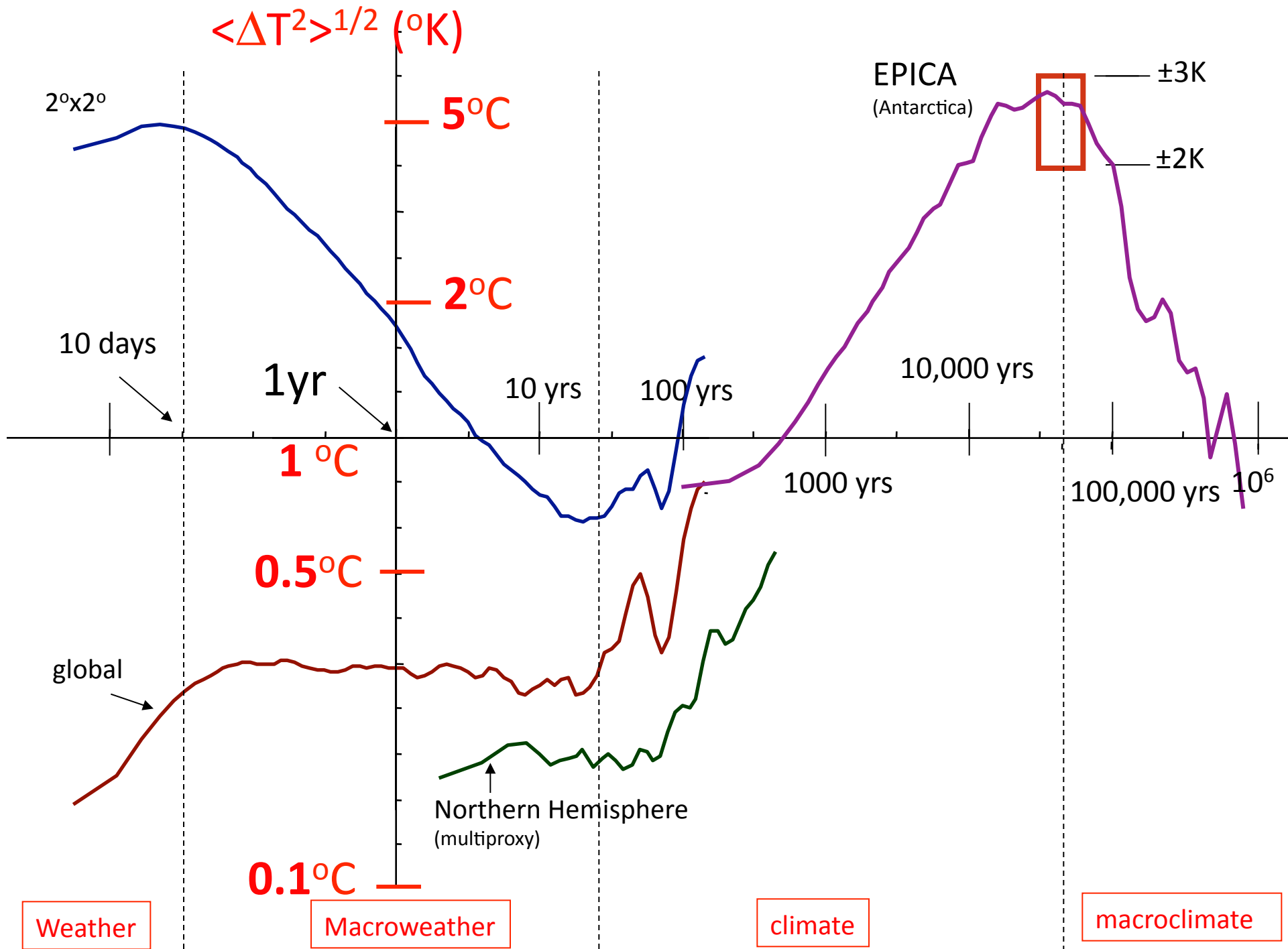
Fluctuation  $\Rightarrow$  constant

Climate  
(10-30 yrs to  
50,000 yrs)

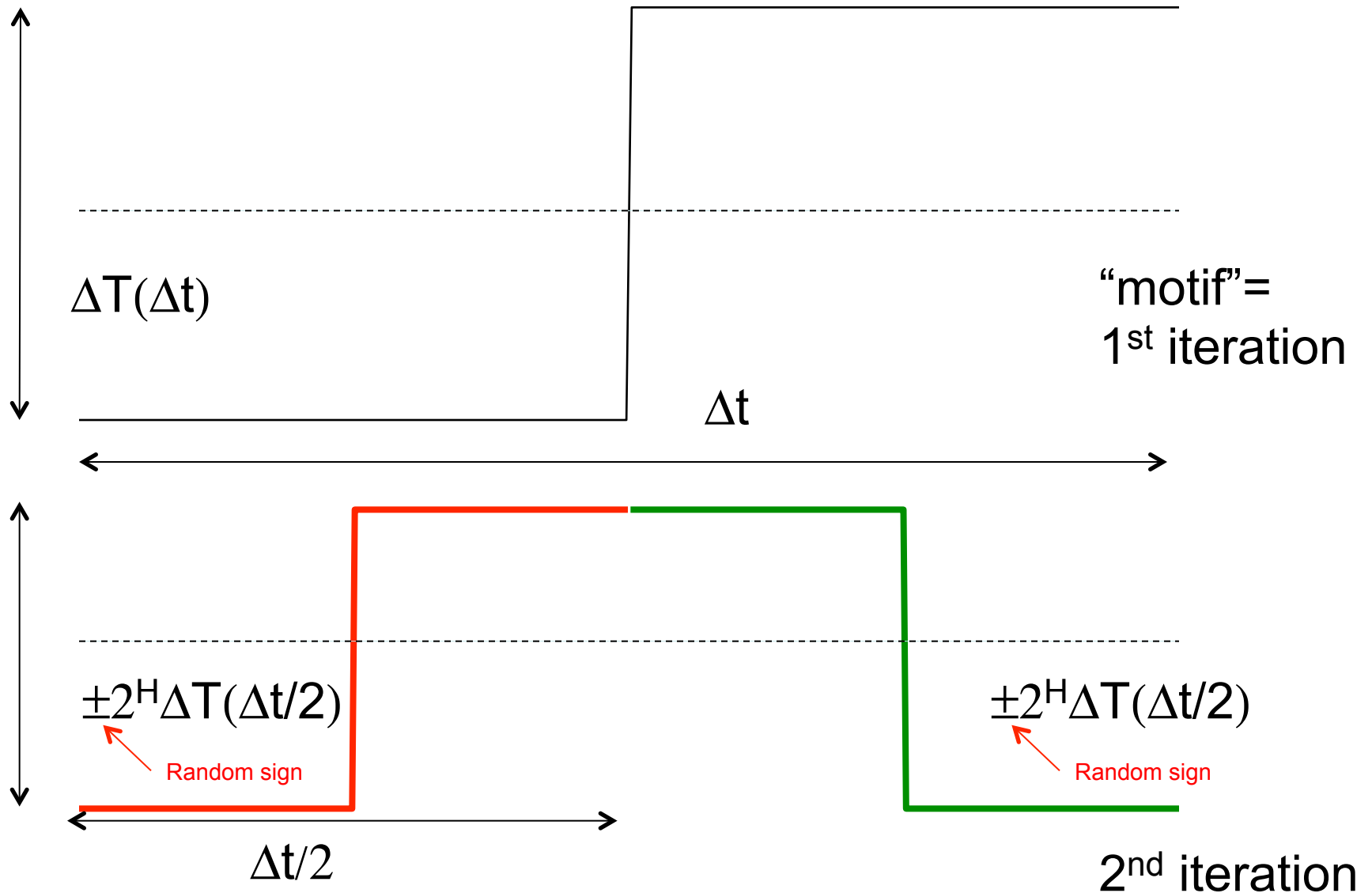
Macroweather  
(10 days to 10-30 yrs)

Weather  
(up to 10 days)





# Additive, fractal “H model”



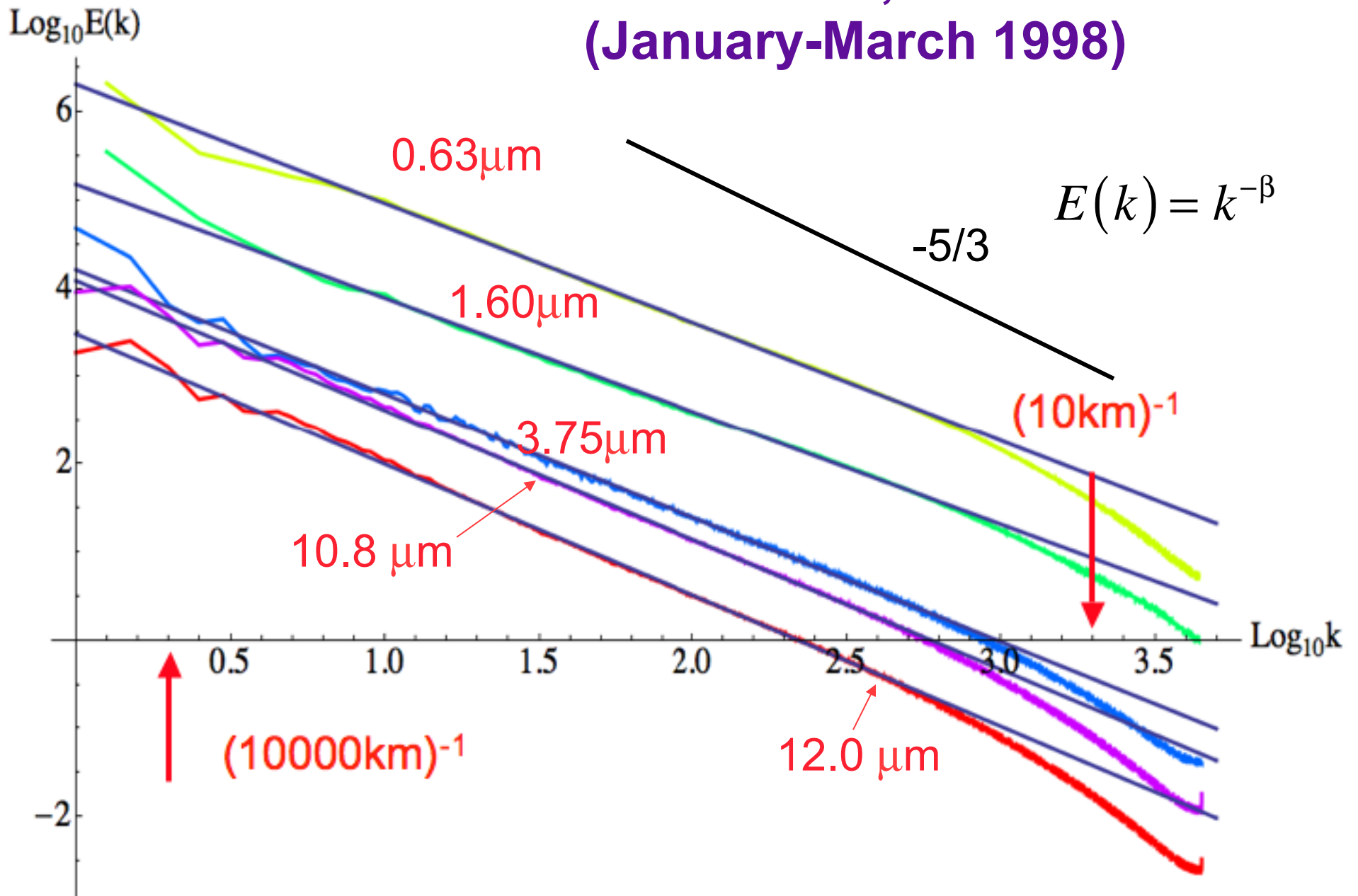




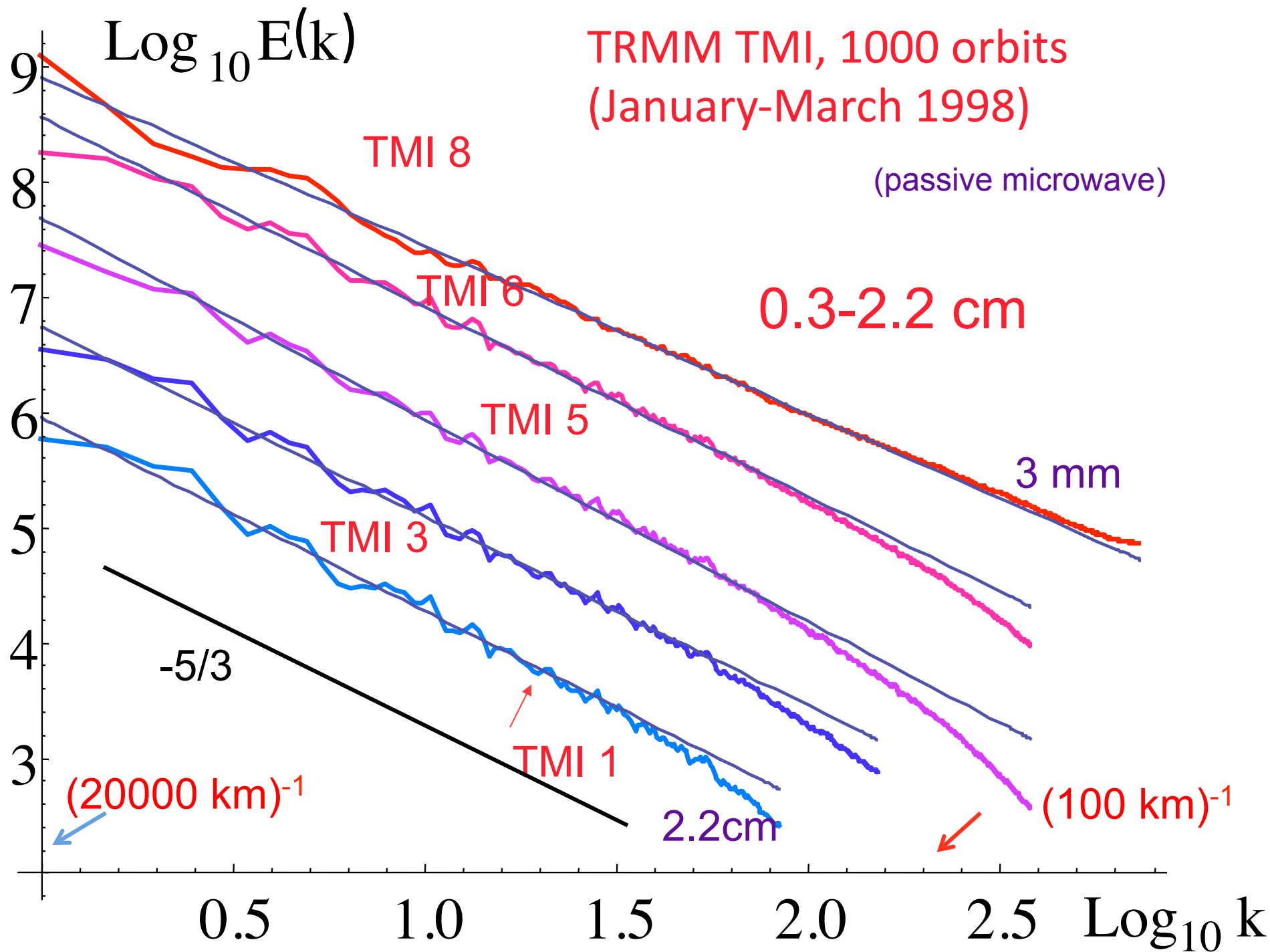
The weather regime:  
The emergent laws hold up to  
planetary scales  
(Horizontal scaling)

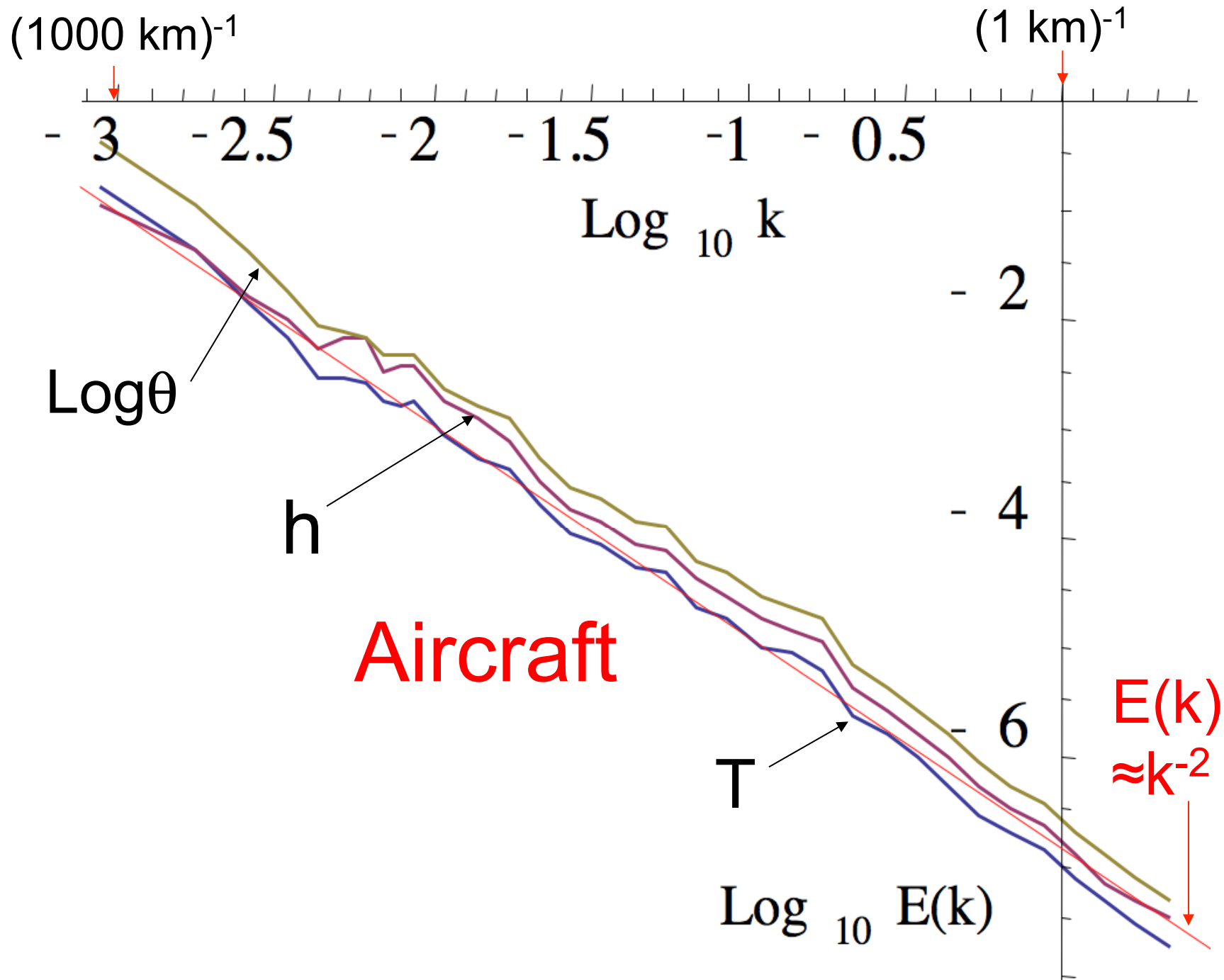
$$E(k) = k^{-\beta}$$

# TRMM VIRS, 1000 orbits (January-March 1998)

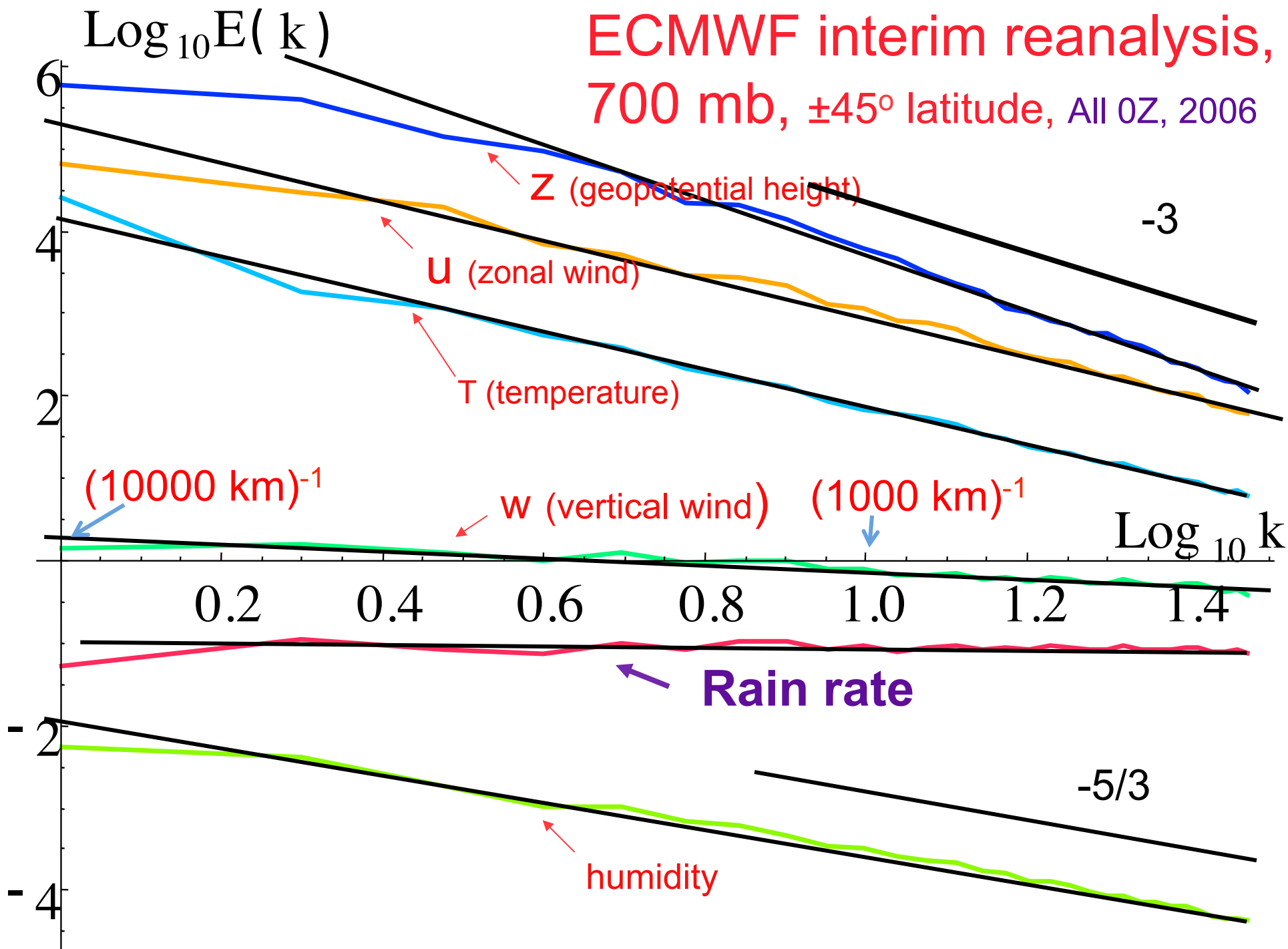


Visible, near infra red, thermal infra red





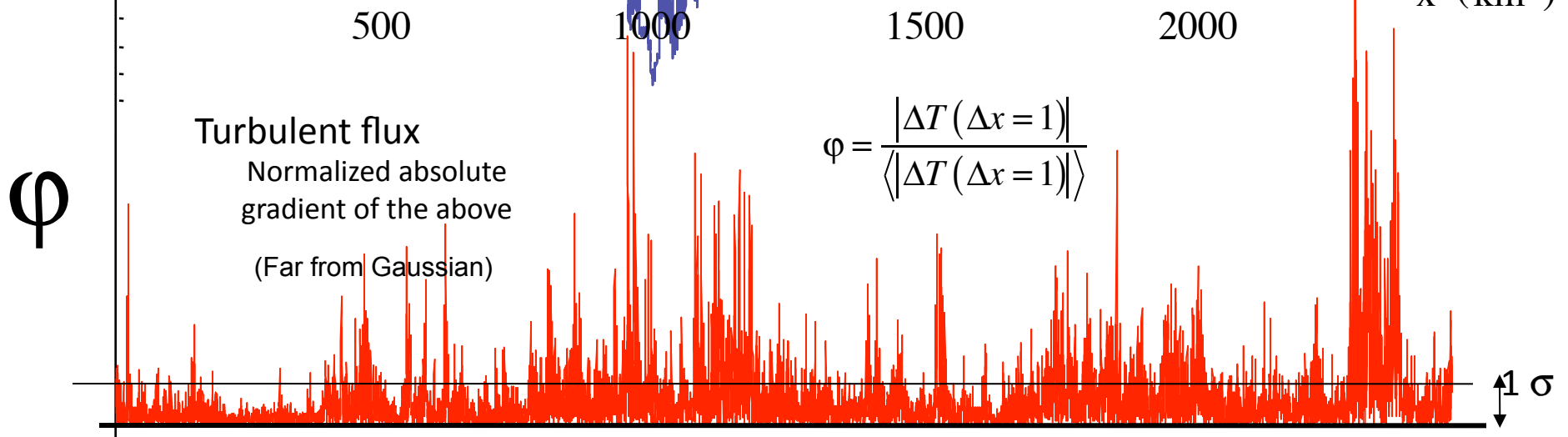
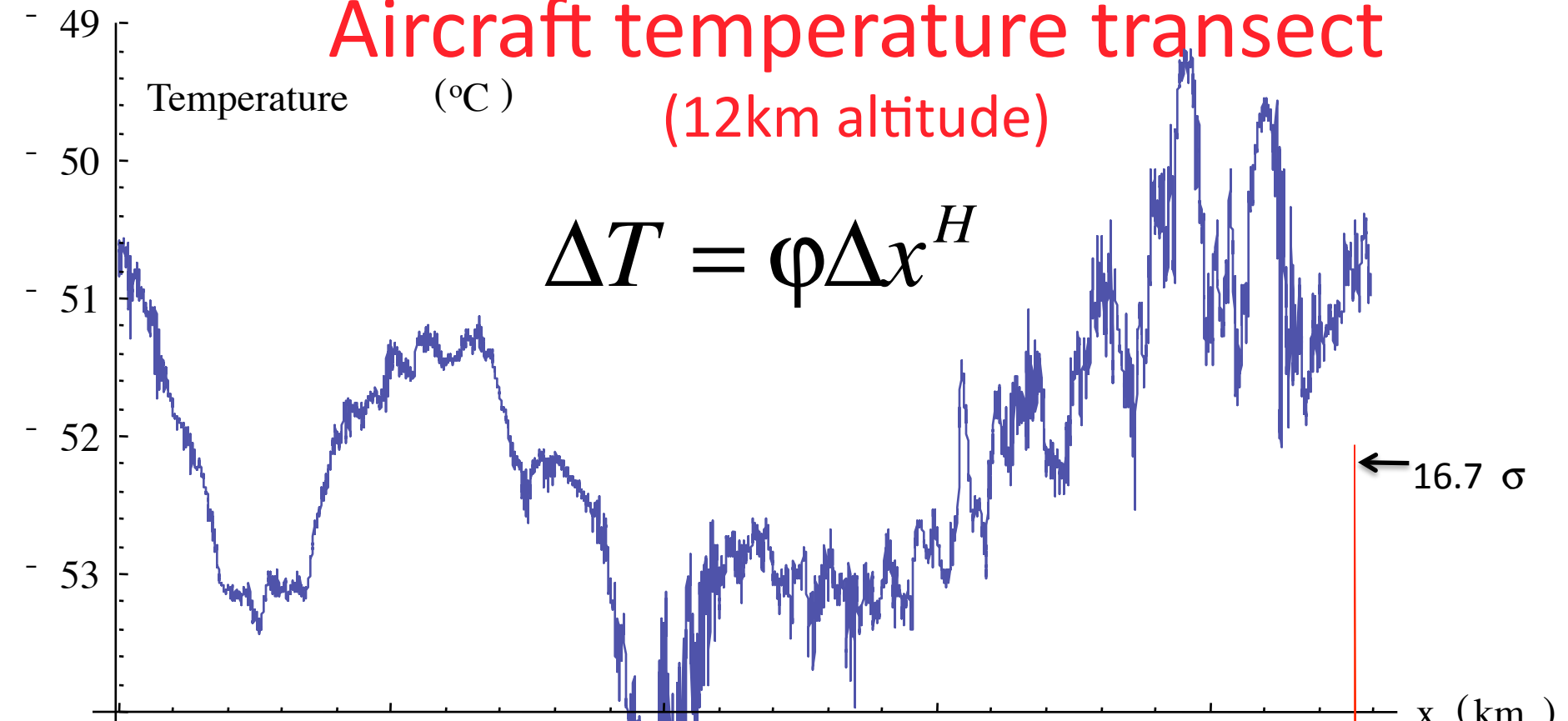
ECMWF interim reanalysis,  
700 mb,  $\pm 45^\circ$  latitude, All OZ, 2006



# Cascades and Multifractals

# Aircraft temperature transect

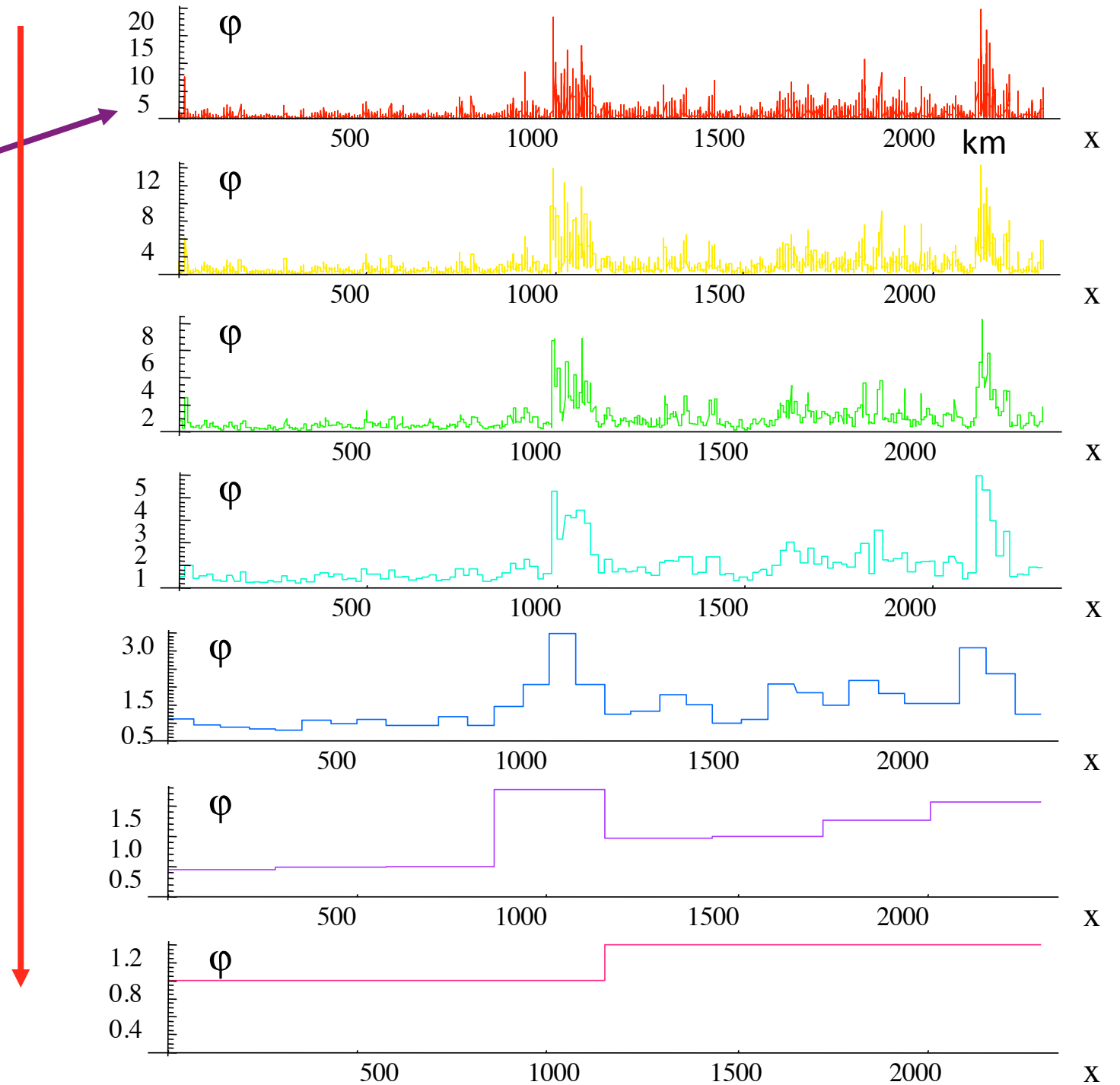
(12km altitude)



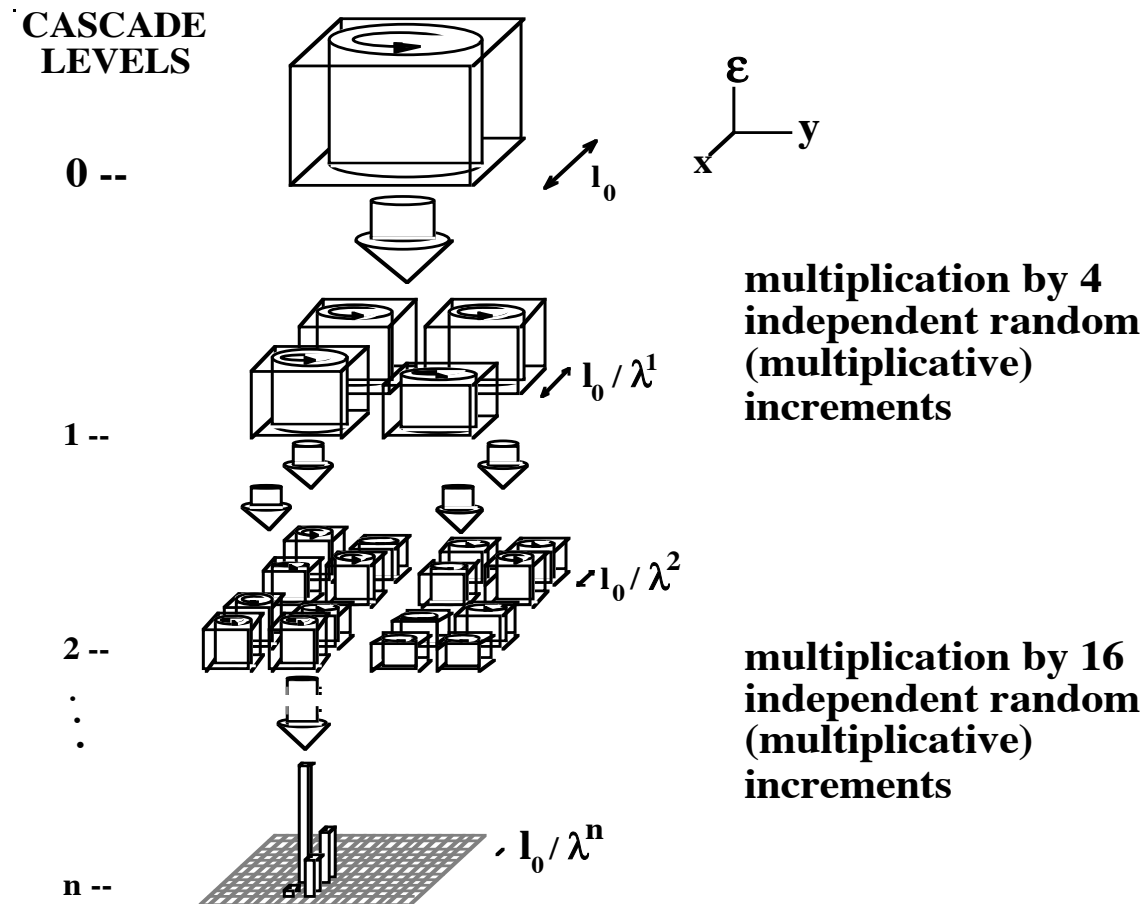


Temperature  
turbulent flux  $\phi$   
at 280m resolution

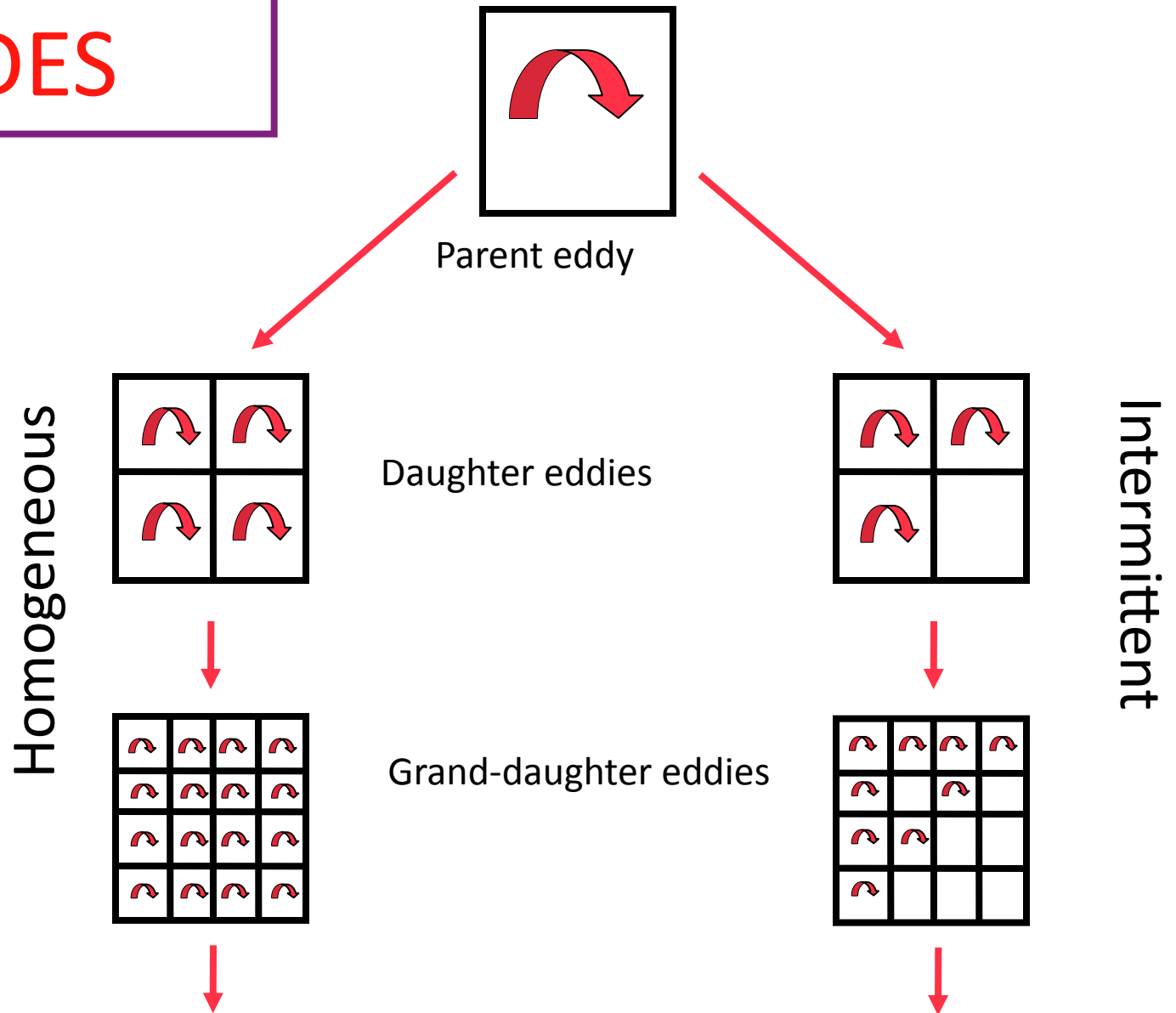
High to low  
Resolution:  
degrading by  
factors of 4



# Scale by scale simplicity: cascades



# CASCADES

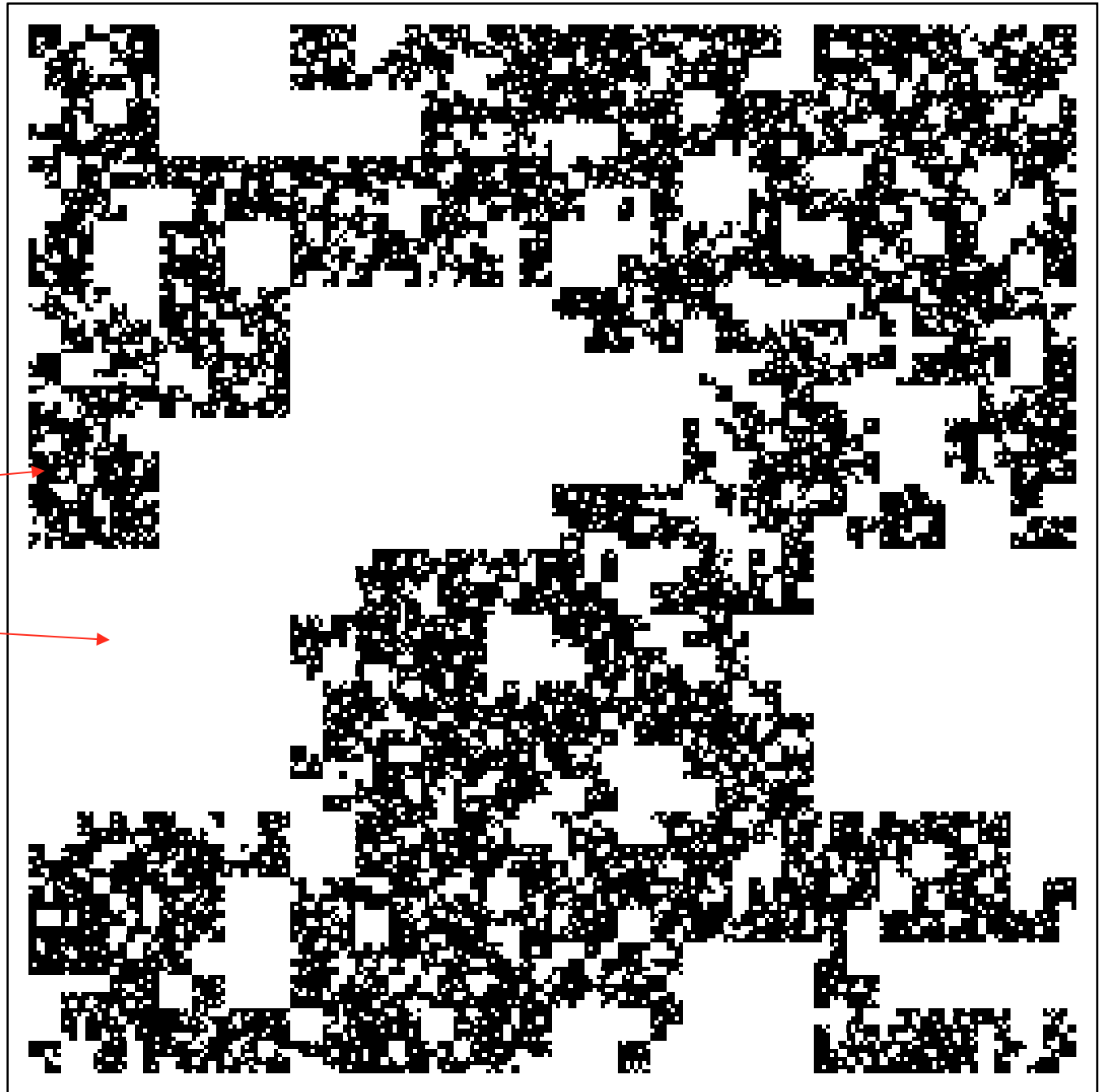


# $\beta$ -model

Fractal set

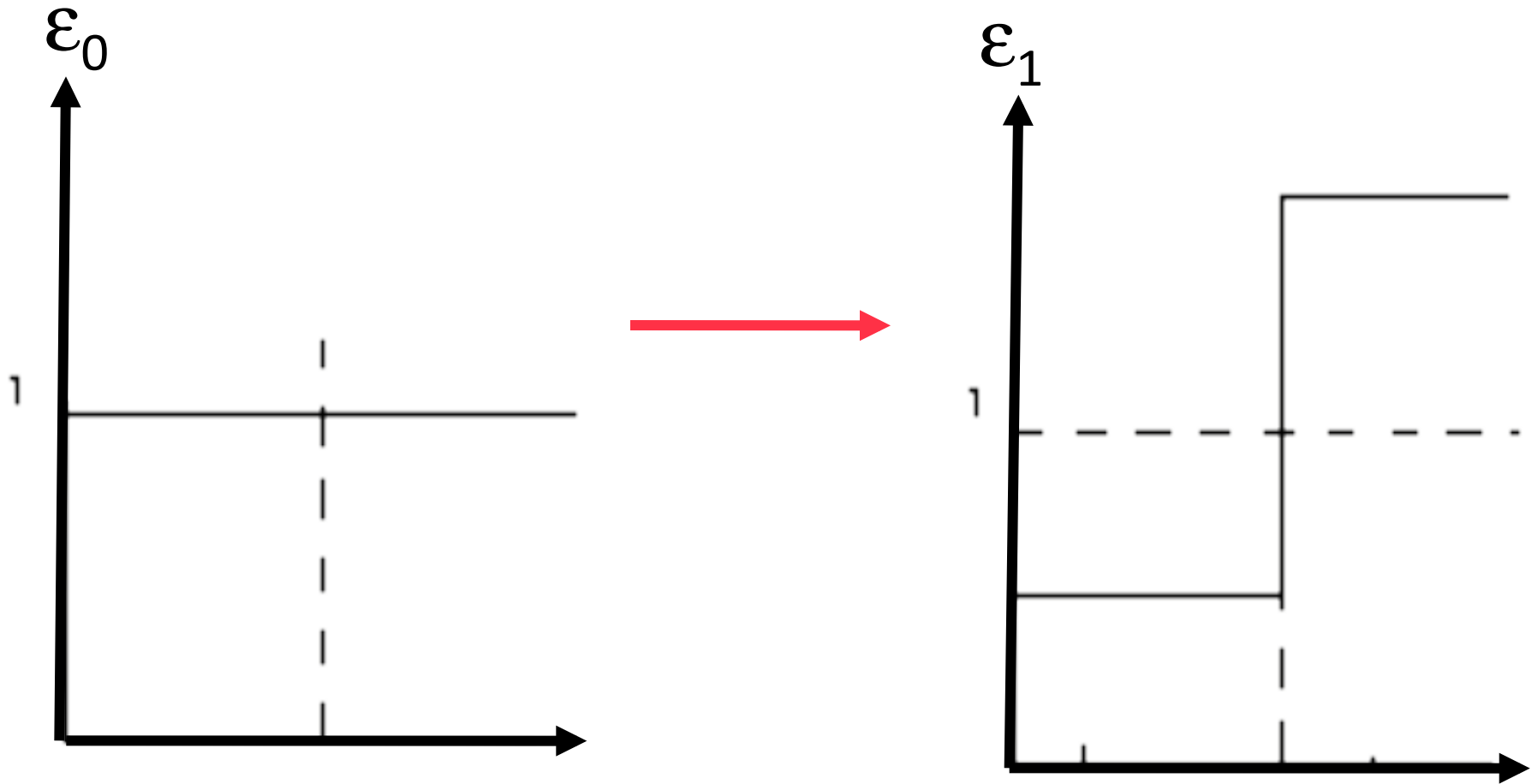
“active”

“calm”



# Cascades and Multifractals

Simulations: adding small scale details  
(low resolution to high)



(“ $\alpha$  model”)

# Multiplicative Cascades

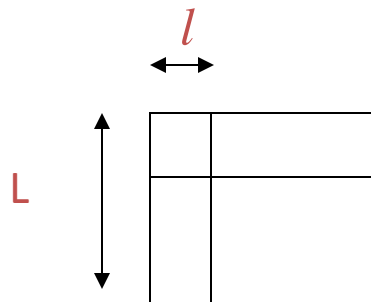
Generic statistical behaviour:

scaling Scale invariant

Turbulent flux

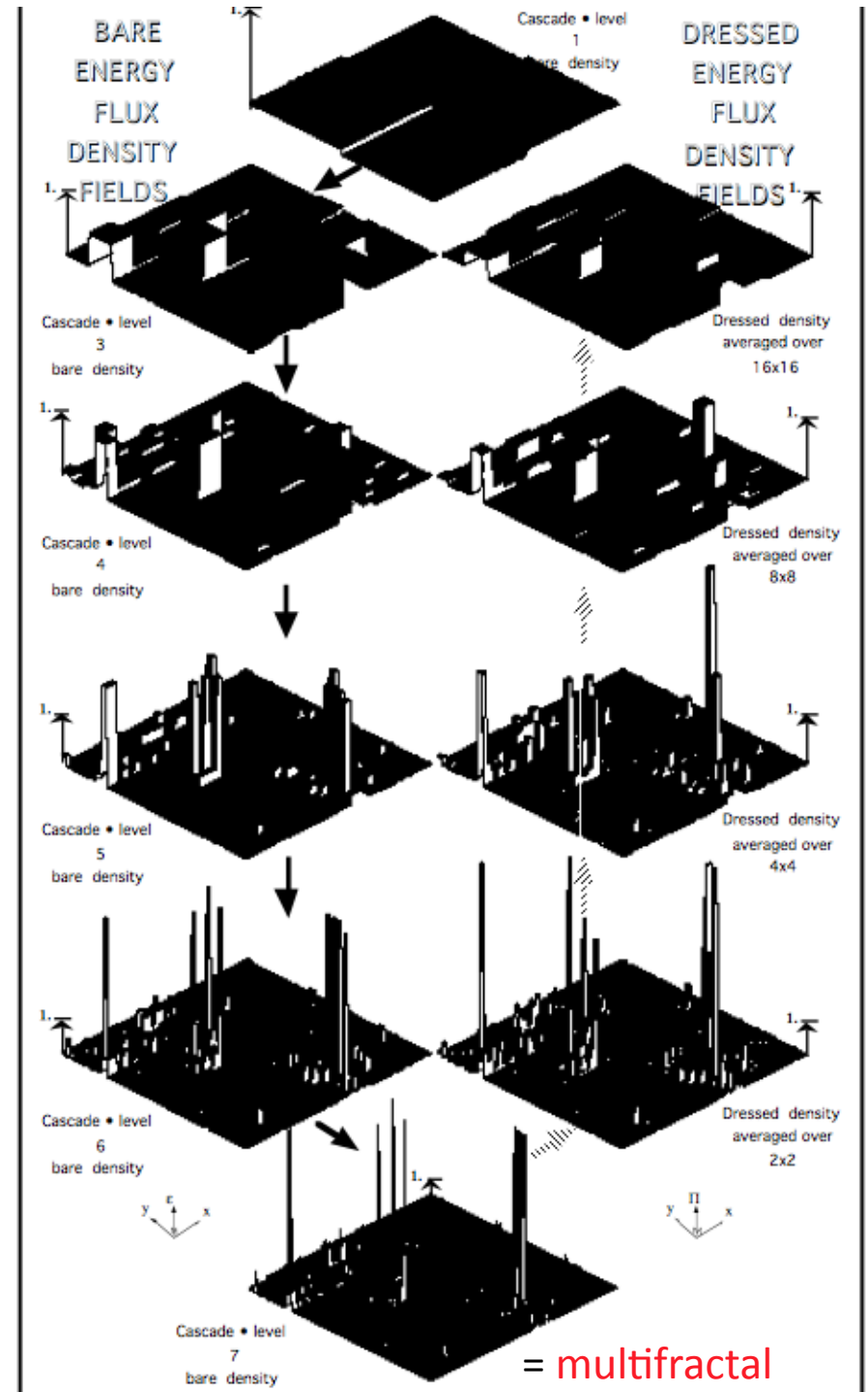
$$\langle \epsilon_\lambda^q \rangle \approx \lambda^{K(q)}$$

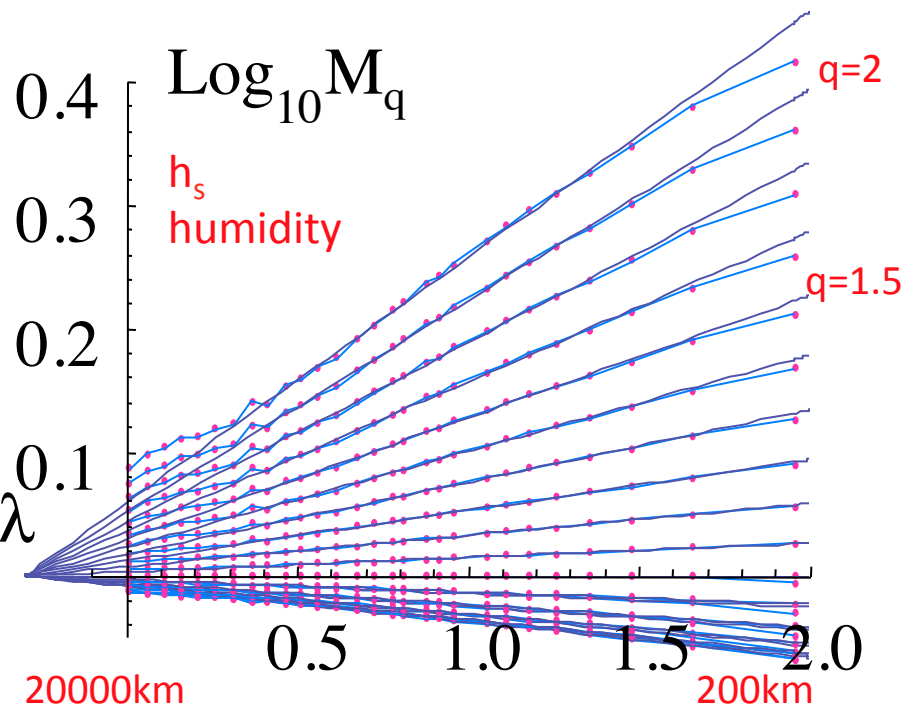
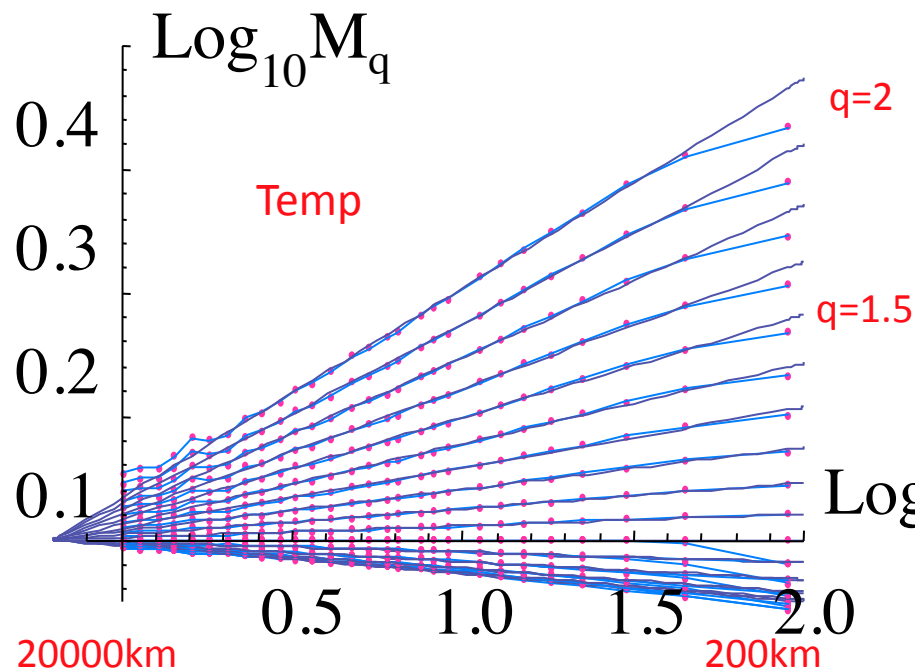
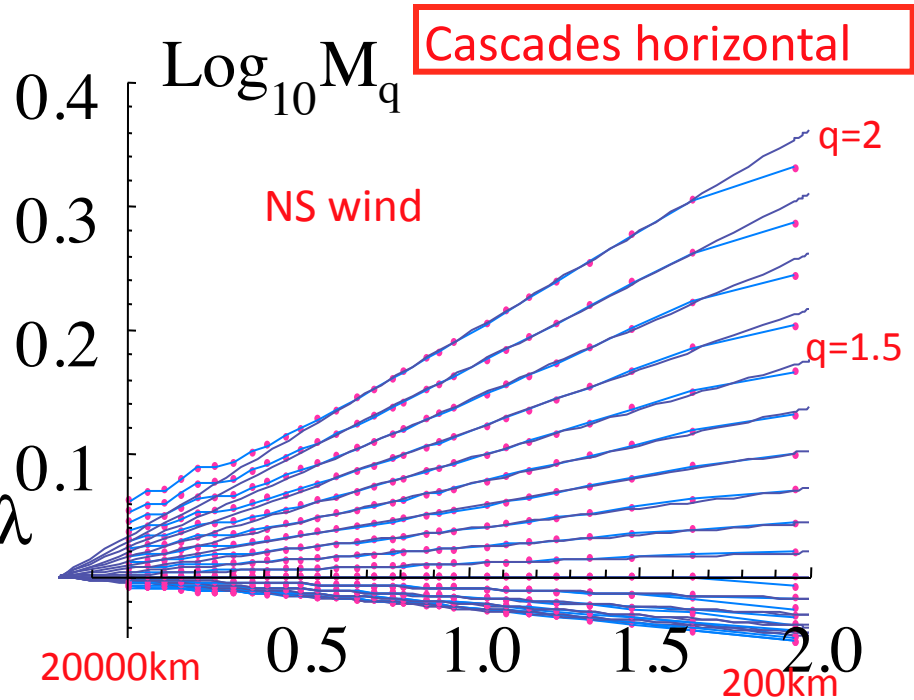
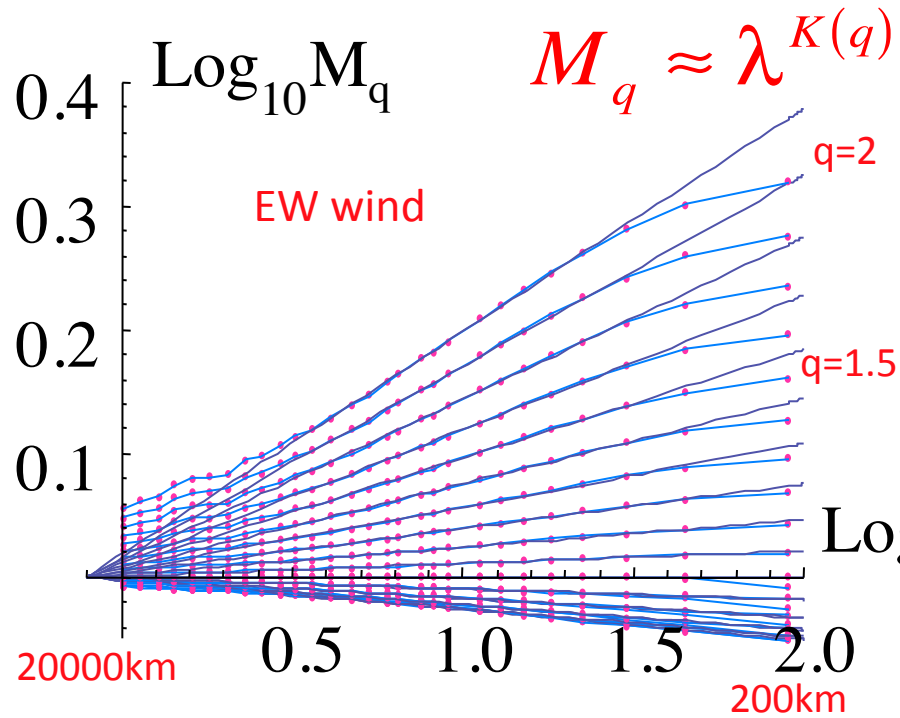
Statistical averaging Resolution: ratio  $\lambda=L/l$



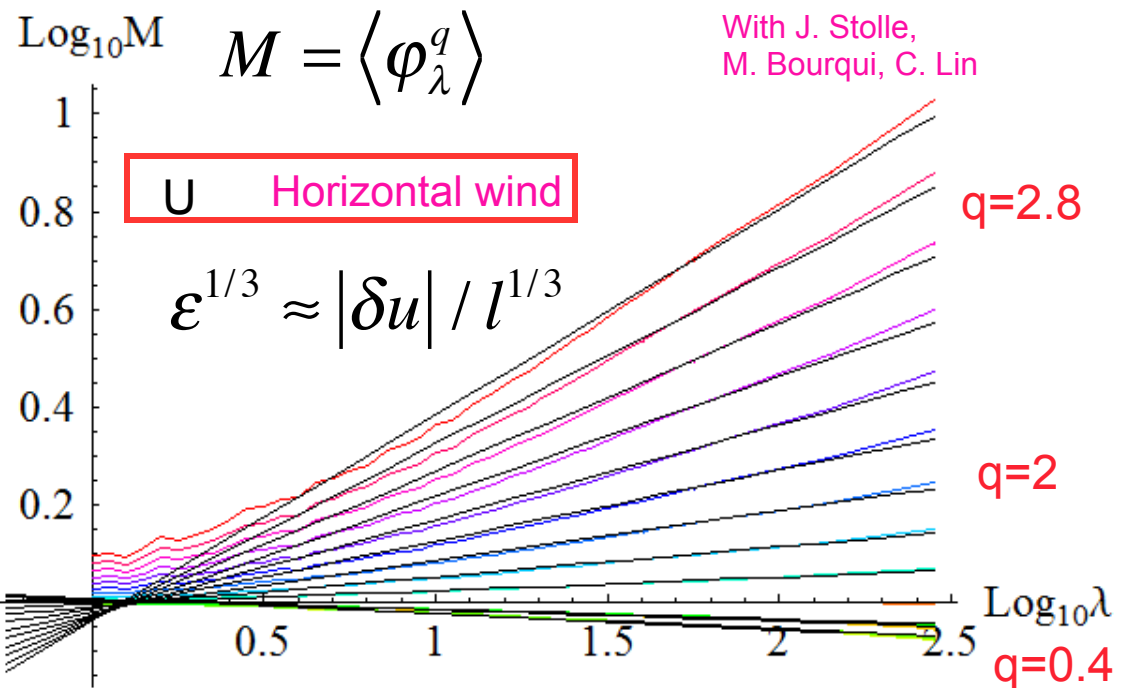
Probabilities:

$$\Pr(\epsilon_\lambda > \lambda^\gamma) \approx \lambda^{-c(\lambda)}$$

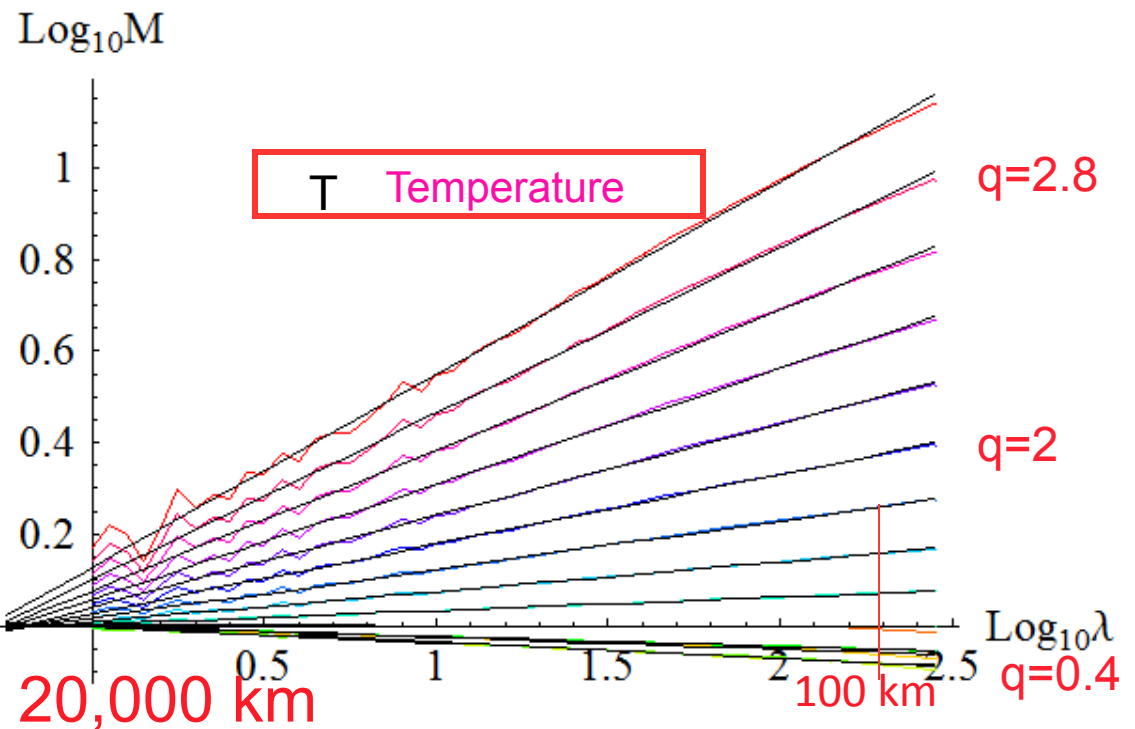




# Global GEMS Model 00h



Analysis of four months  
U,T at 1000 mb

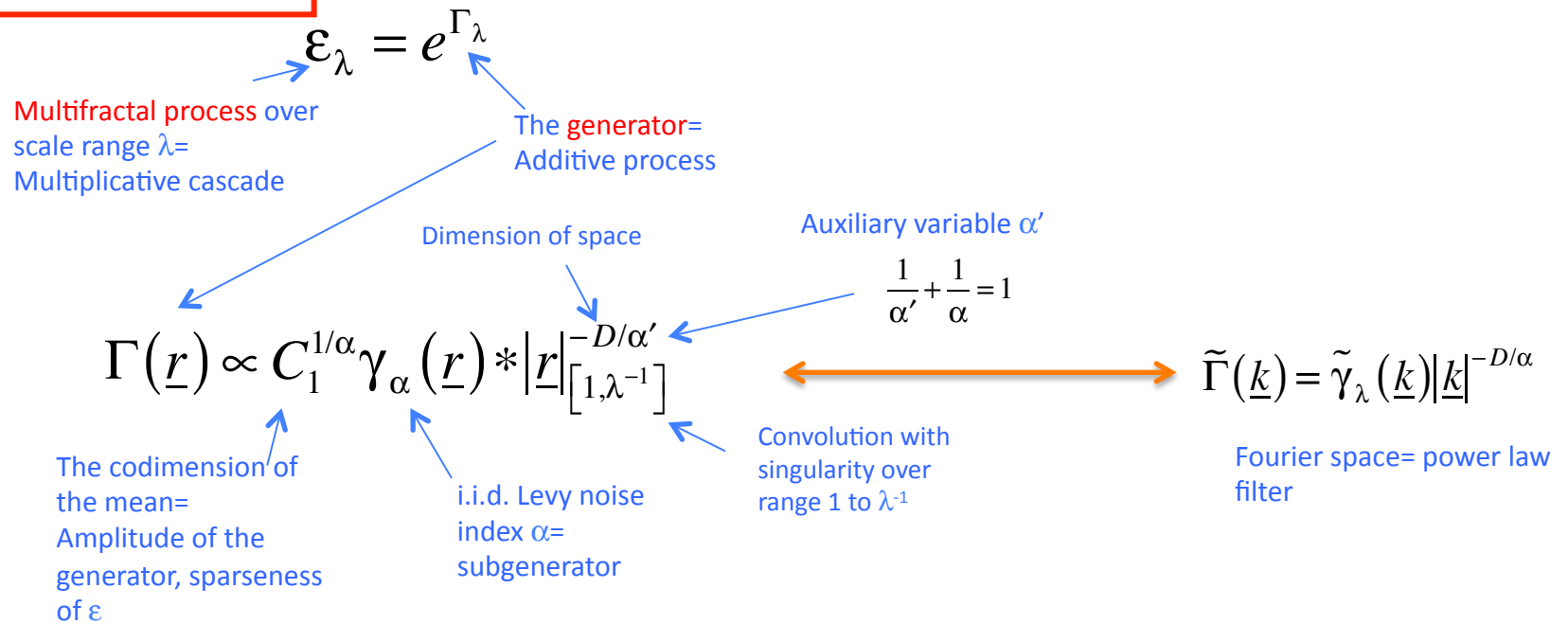


(48 h forecasts are  
almost the same)



# Multiplicative processes

## The process:



## The statistics:

$$\langle \epsilon_\lambda^q \rangle = \lambda^{K(q)}$$

General multifractal statistics, convex  $K(q)$

$$K(q) = \frac{C_1}{\alpha - 1} (q^\alpha - q)$$

Universal multifractals

# Fractionally Integrated Flux (FIF) model (both additive and multiplicative)

## The process

$$I(\underline{r}) = \varepsilon_\lambda(\underline{r}) * |\underline{r}|^{-(D-H)} \longleftrightarrow \tilde{I}(\underline{k}) = \tilde{\varepsilon}_\lambda(\underline{k}) |\underline{k}|^{-H}$$

Convolution=  
fractional integration  
order H

Fourier space= power  
law filter

## The statistics

$$S_q(\underline{\Delta r}) = \langle \Delta I(\underline{\Delta r})^q \rangle = \langle \varepsilon_\lambda^q \rangle |\underline{\Delta r}|^{qH} = |\underline{\Delta r}|^{\xi(q)}$$

q<sup>th</sup> order  
structure  
function

fluctuation

Note:

$$\lambda = L / |\underline{\Delta r}|$$

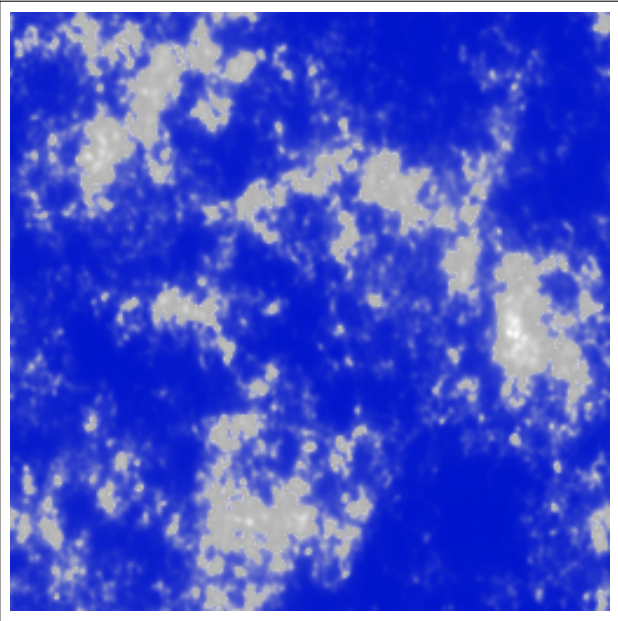
$$\langle \varepsilon_\lambda^q \rangle = \lambda^{K(q)}$$

structure  
function  
exponent

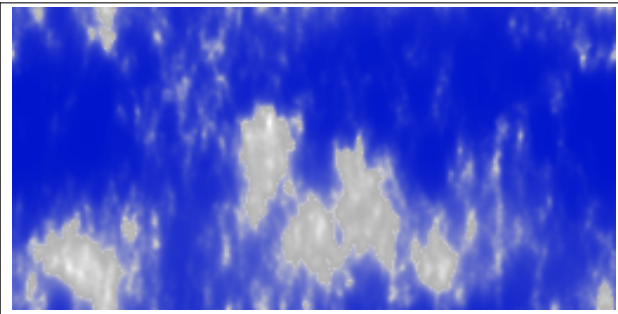
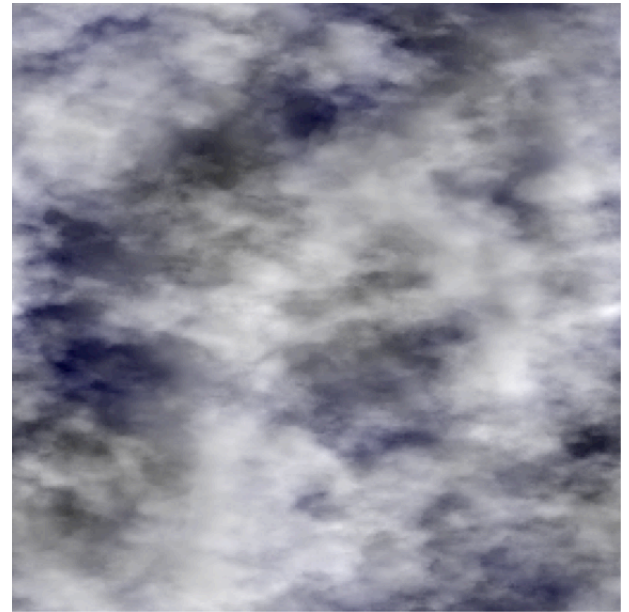
$$\xi(q) = qH - K(q)$$

# FIF modeling: clouds and radiative transfer

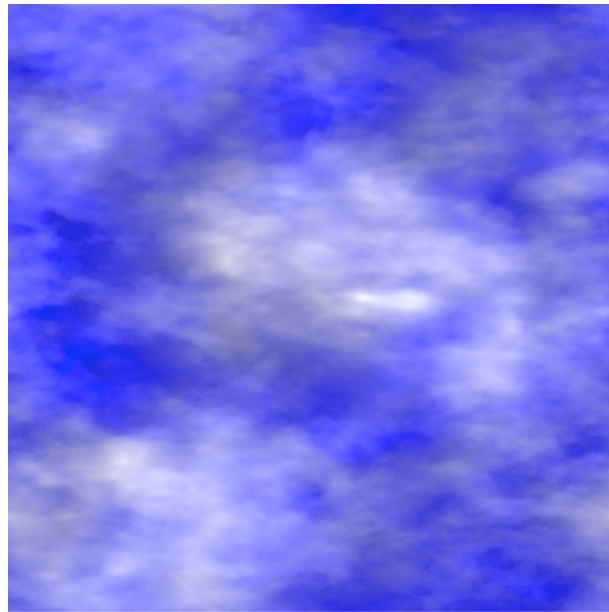
Cloud liquid water (top)



Cloud top visible

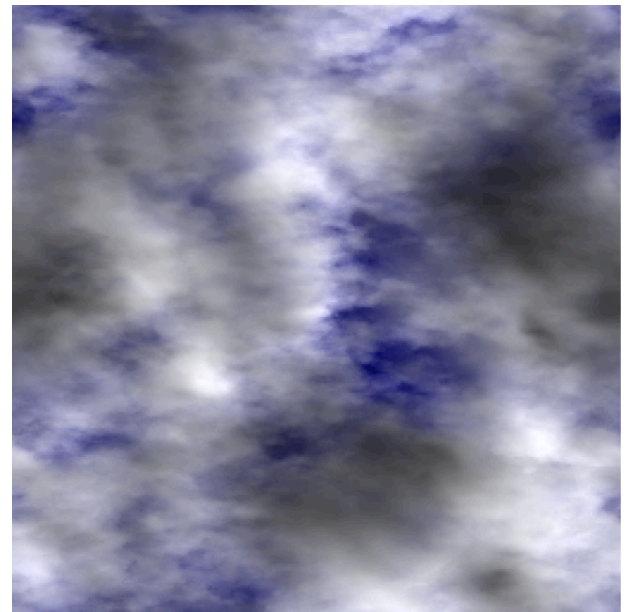


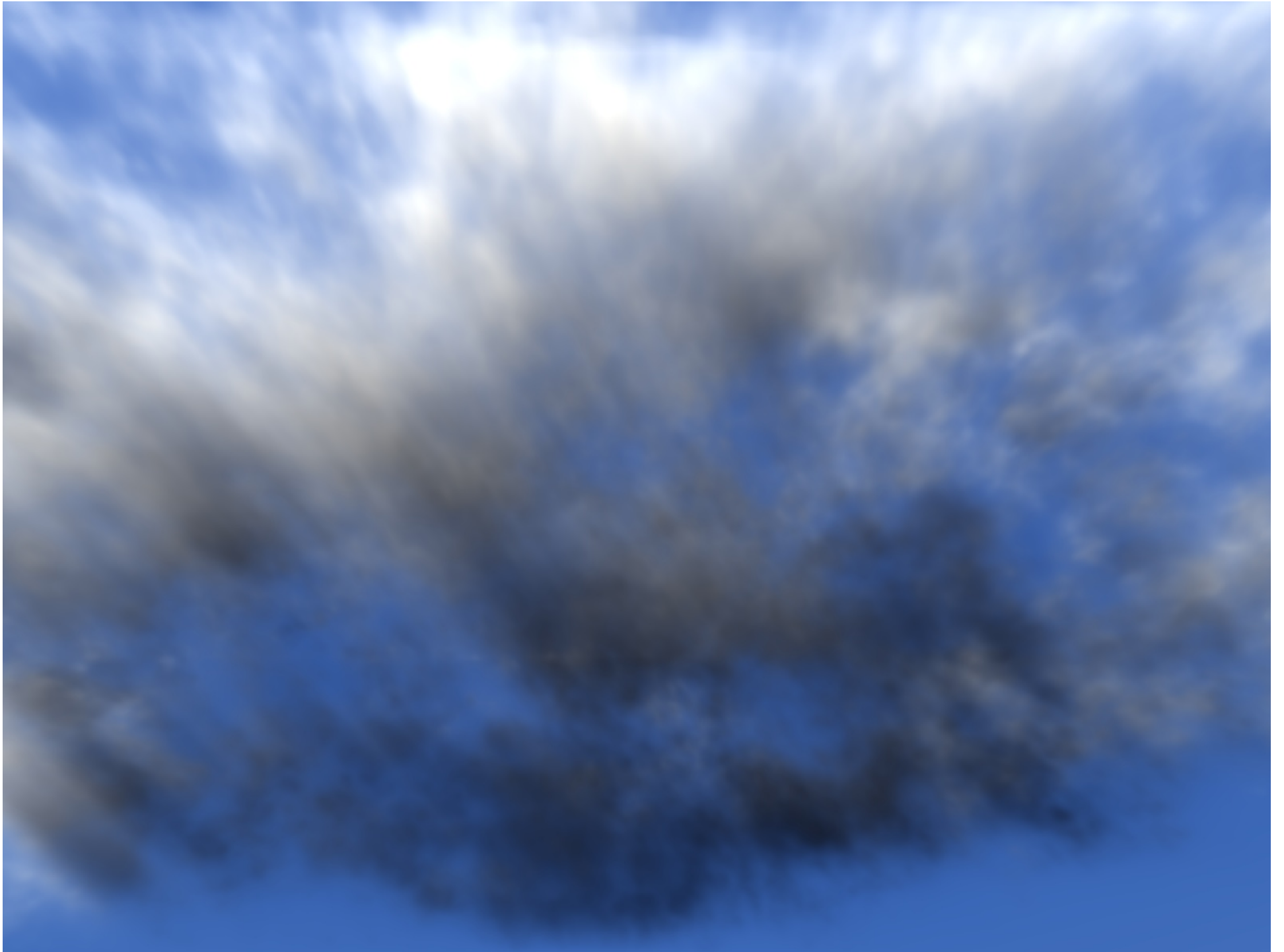
Cloud liquid water (side)



Cloud top, infra red

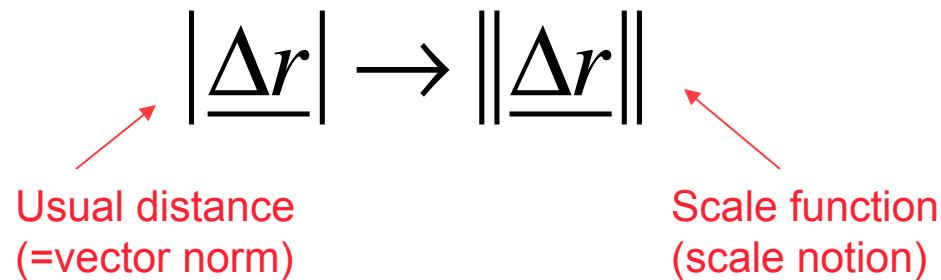
Cloud bottom visible





Extensions to the vertical  
(scaling stratification)

# The physical scale function and differential scaling

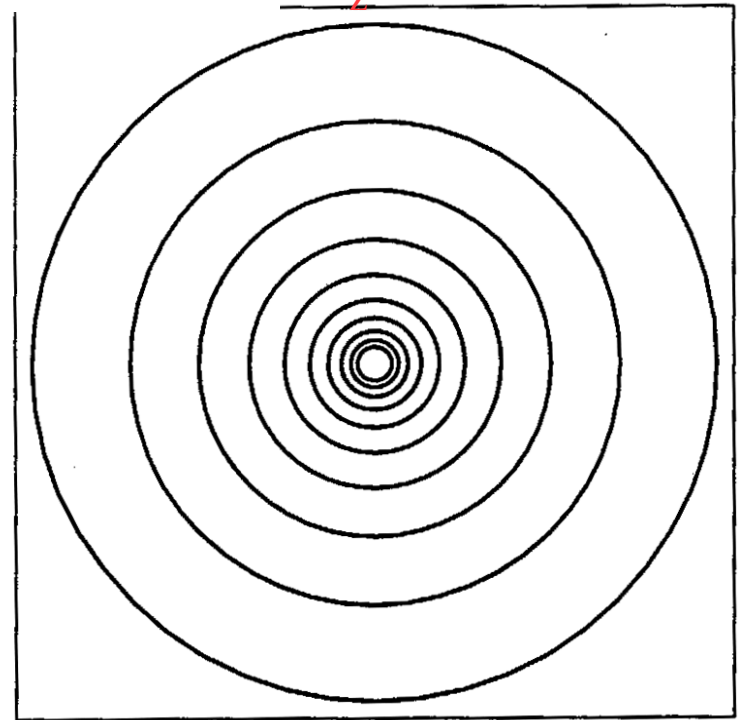


“canonical” scale function:

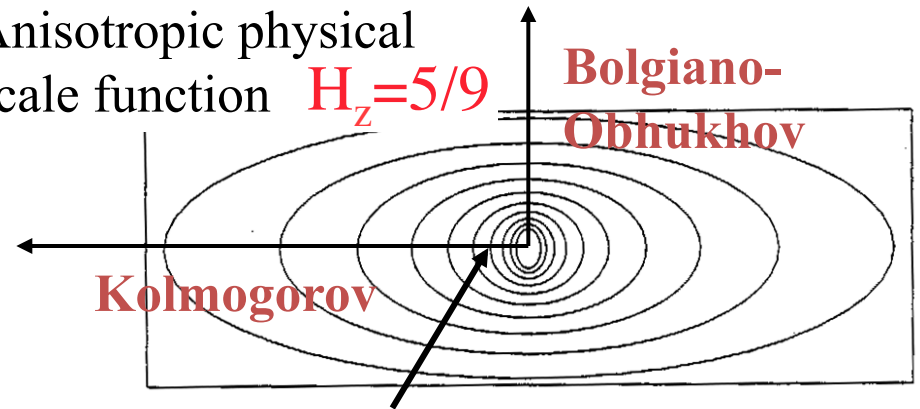
$$\|(\Delta x, \Delta z)\| = l_s \left( \left( \frac{\Delta x}{l_s} \right)^2 + \left( \frac{\Delta z}{l_s} \right)^{2/H_z} \right)^{1/2}$$

## Vertical sections

Isotropic function  $H_z=1$

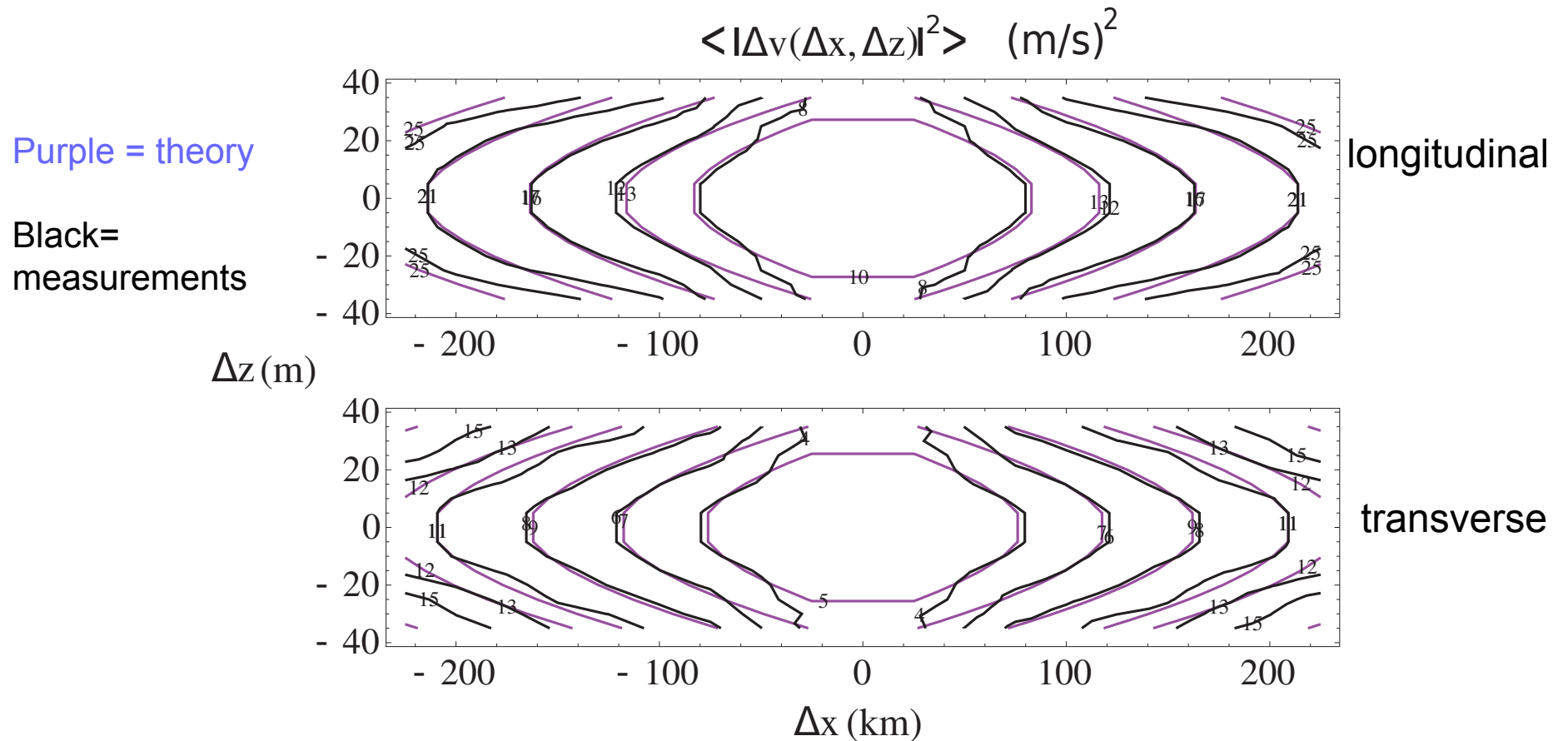


Anisotropic physical scale function  $H_z=5/9$



Sphero-scale

# 14500 aircraft flights: 5-5.5km altitude, 2009, US (TAMDAR data)



Velocity structure function

$$\langle \Delta v^2 (\Delta x, \Delta z) \rangle = C \|(\Delta x, \Delta z)\|^{\xi(2)}$$

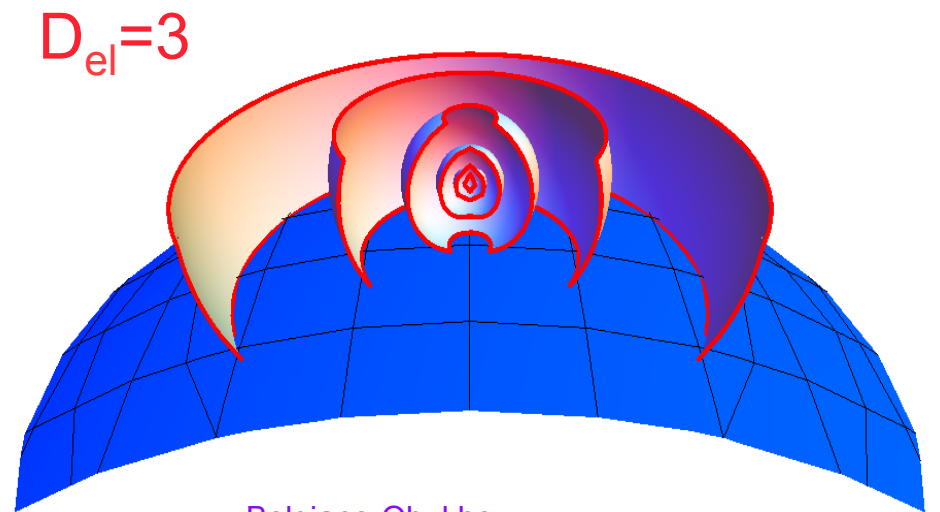
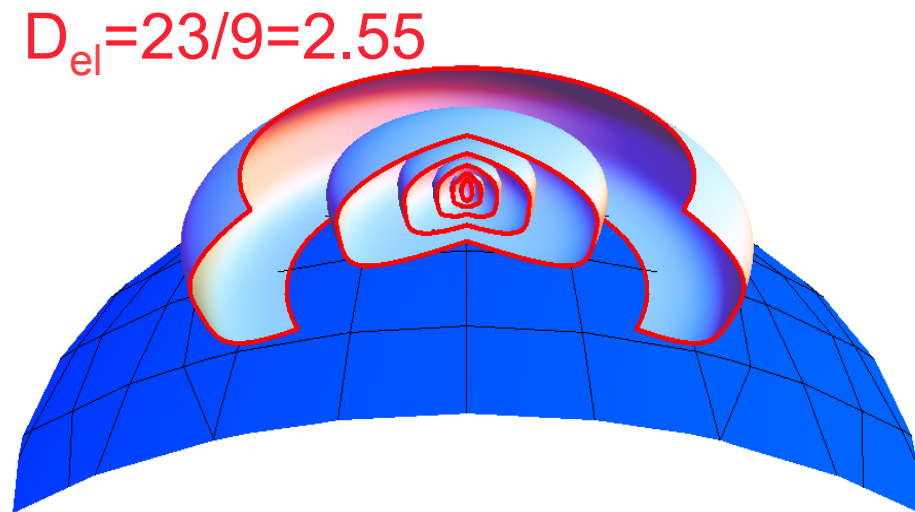
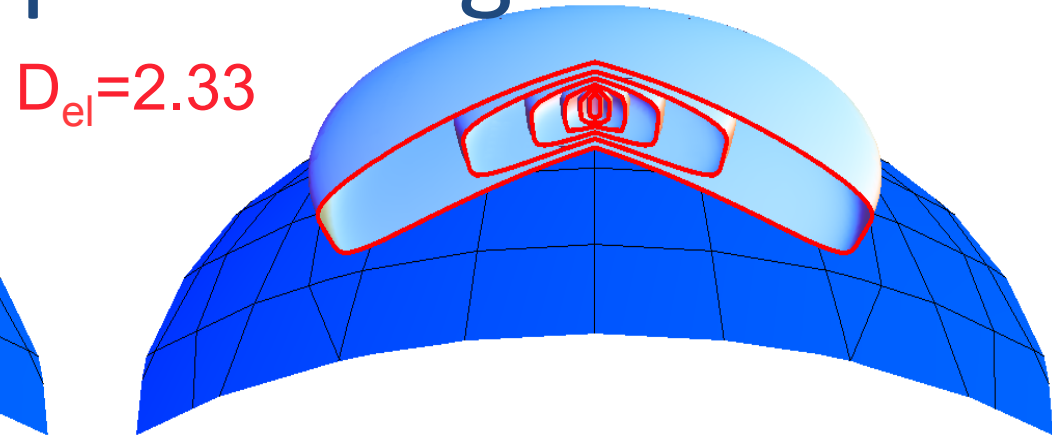
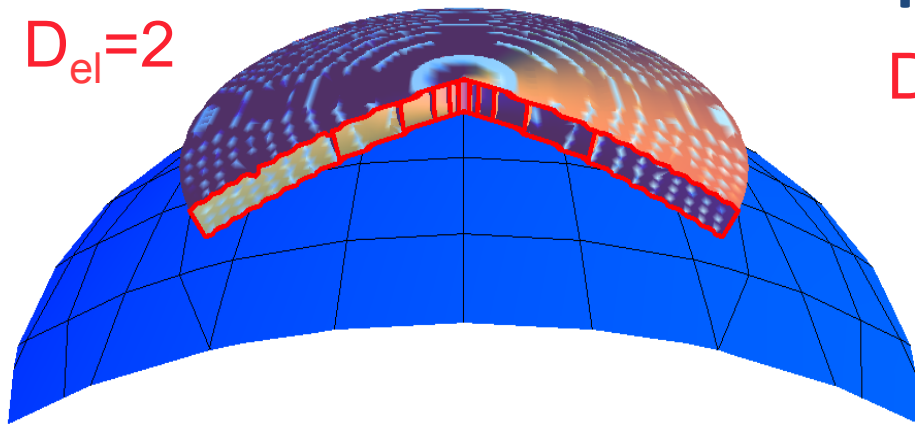
$$\xi(2) \approx 0.80$$

Canonical scale function

$$\|(\Delta x, \Delta z)\| = \left( \left( \frac{\Delta x}{l_s} \right)^2 + \left( \frac{\Delta z}{l_s} \right)^{2/H_z} \right)^{1/2}$$

$$H_z \approx 0.57 \pm 0.01 \quad (\text{Theory: } 5/9=0.555\dots)$$

# Anisotropic Scaling



**The 23/9D model:**

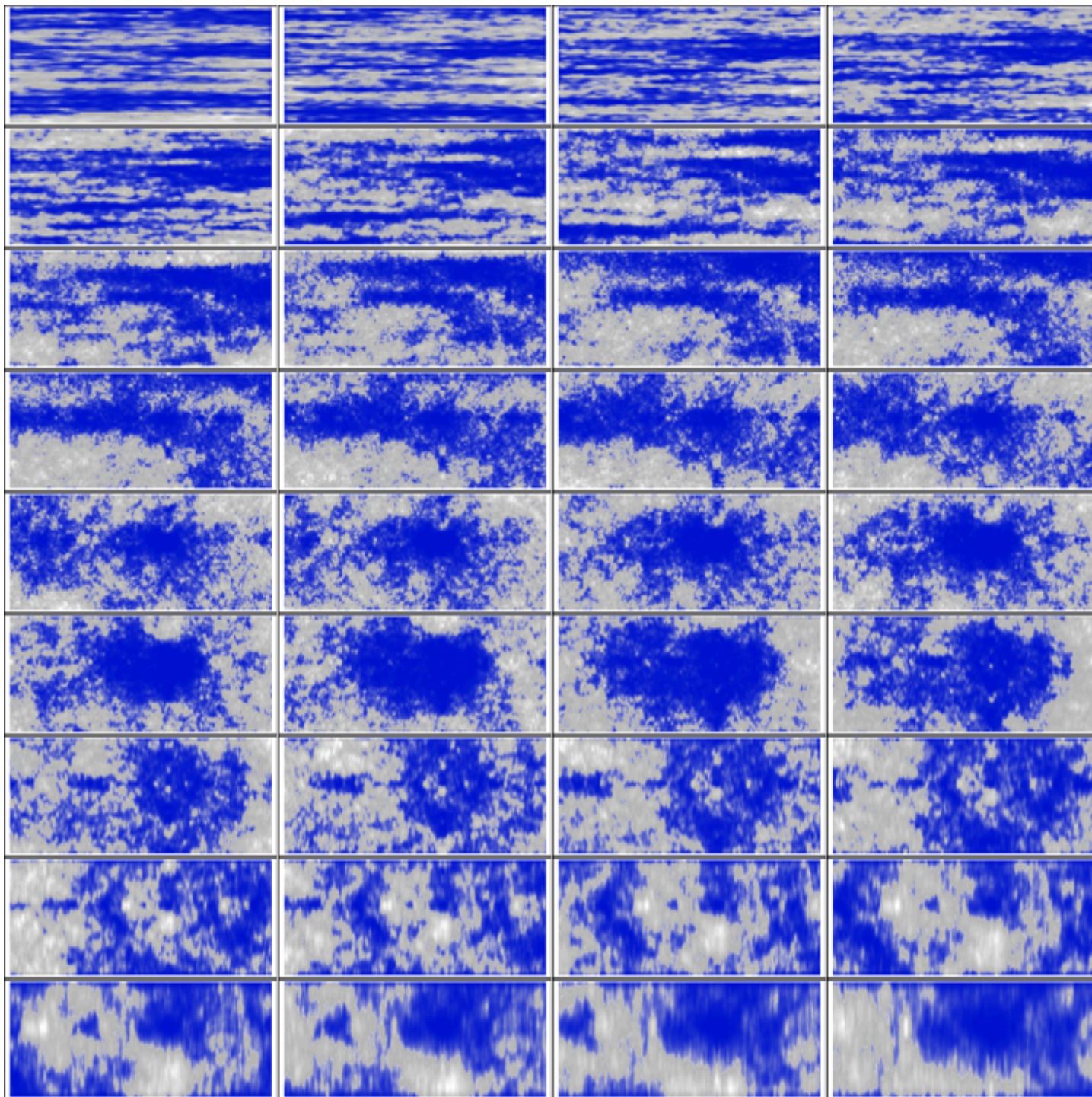
$$\underbrace{\Delta v(\Delta x) = \varepsilon^{1/3} \Delta x^{1/3}}_{\text{Kolmogorov}}; \quad \underbrace{\Delta v(\Delta z) = \phi^{1/5} \Delta z^{3/5}}_{\text{Bolgiano-Obukhov}} \quad H_z = (1/3)/(3/5) = 5/9$$

$$\text{Volume} \approx L_x L_y L_z \approx L^{D_{el}} \quad D_{el} = 2 + H_z = 23/9$$



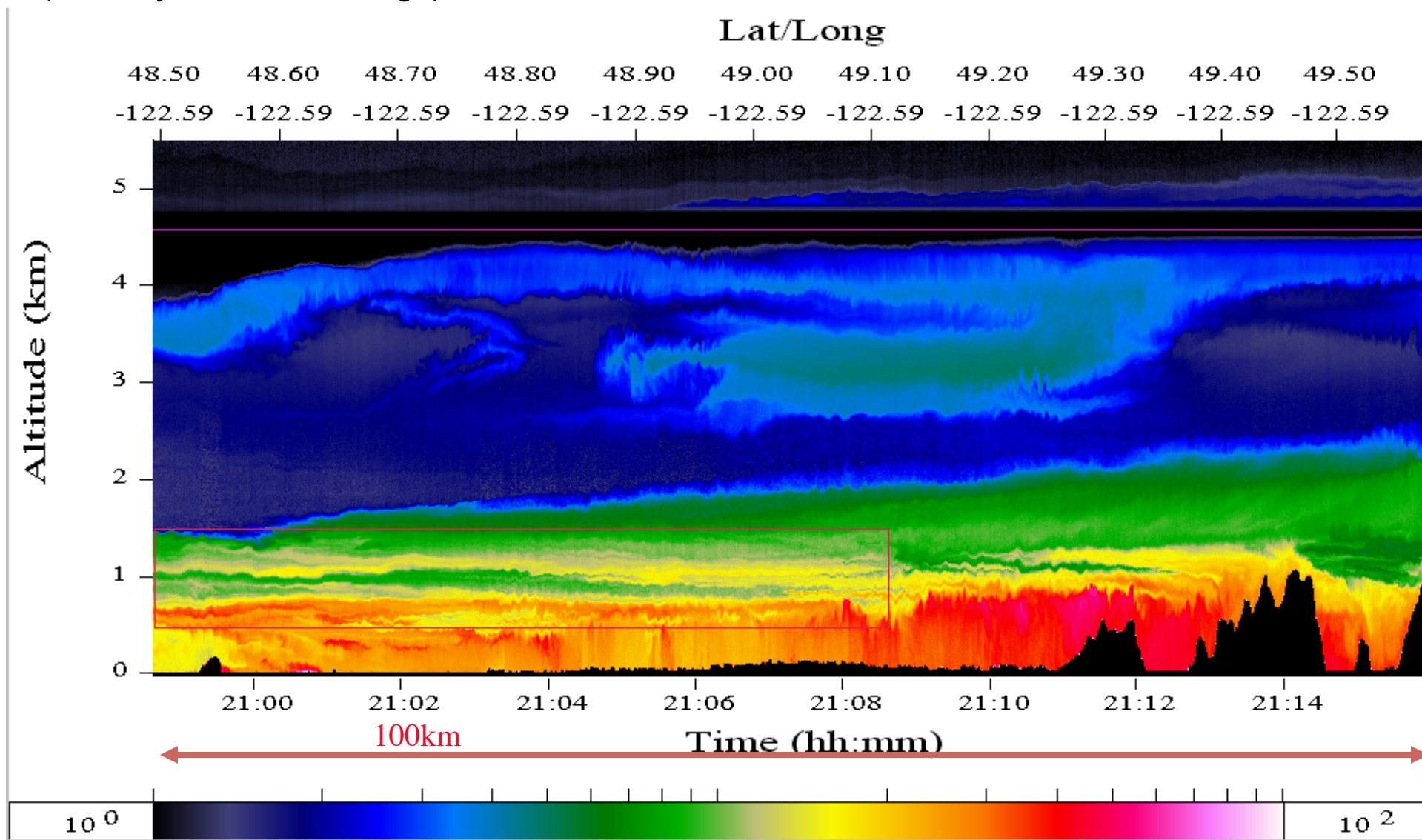
Zoom  
factor  
1000

Vertical cross-  
section



# AERIAL Lidar Data

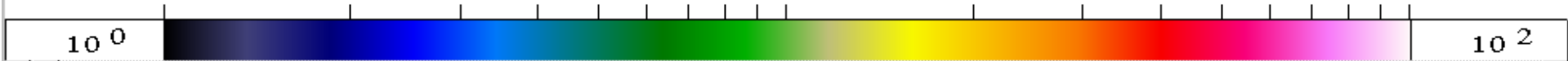
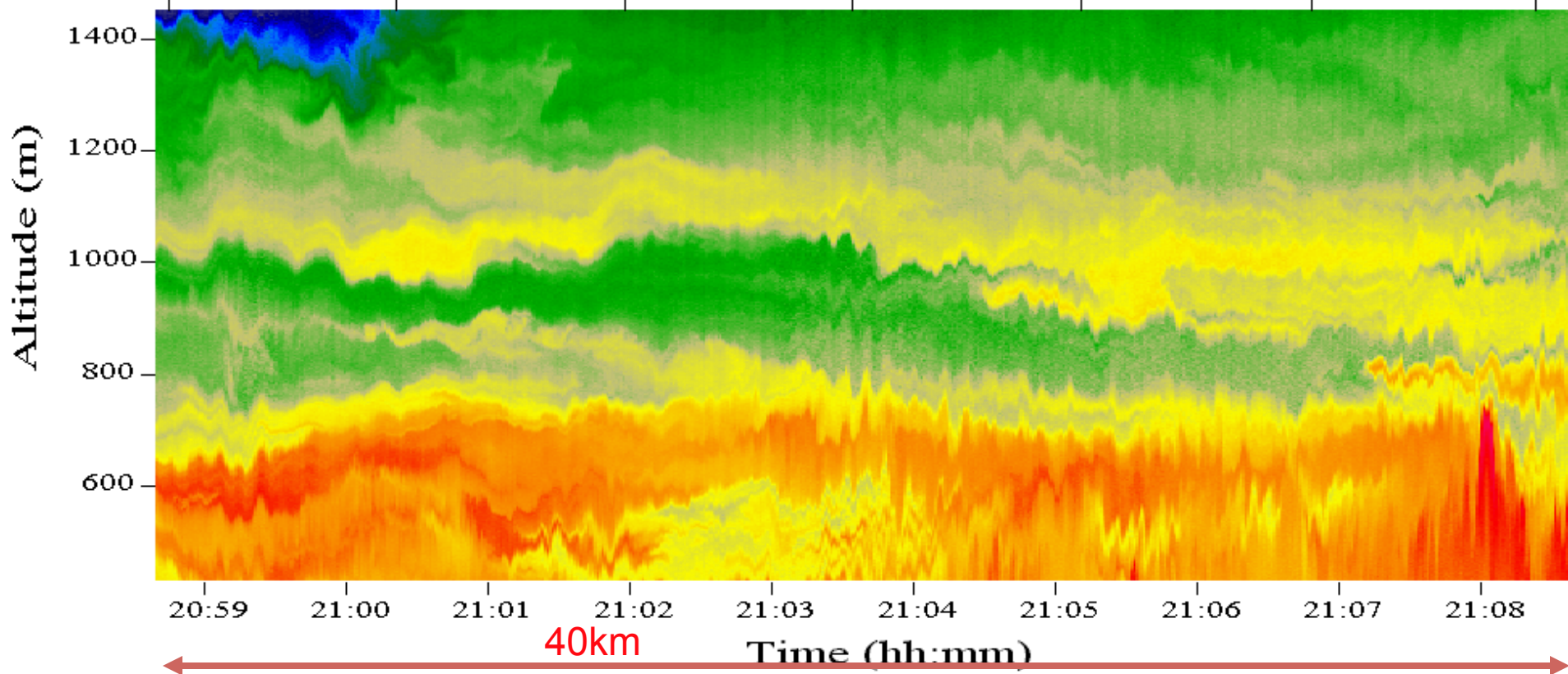
(courtesy of K. Strawbridge)

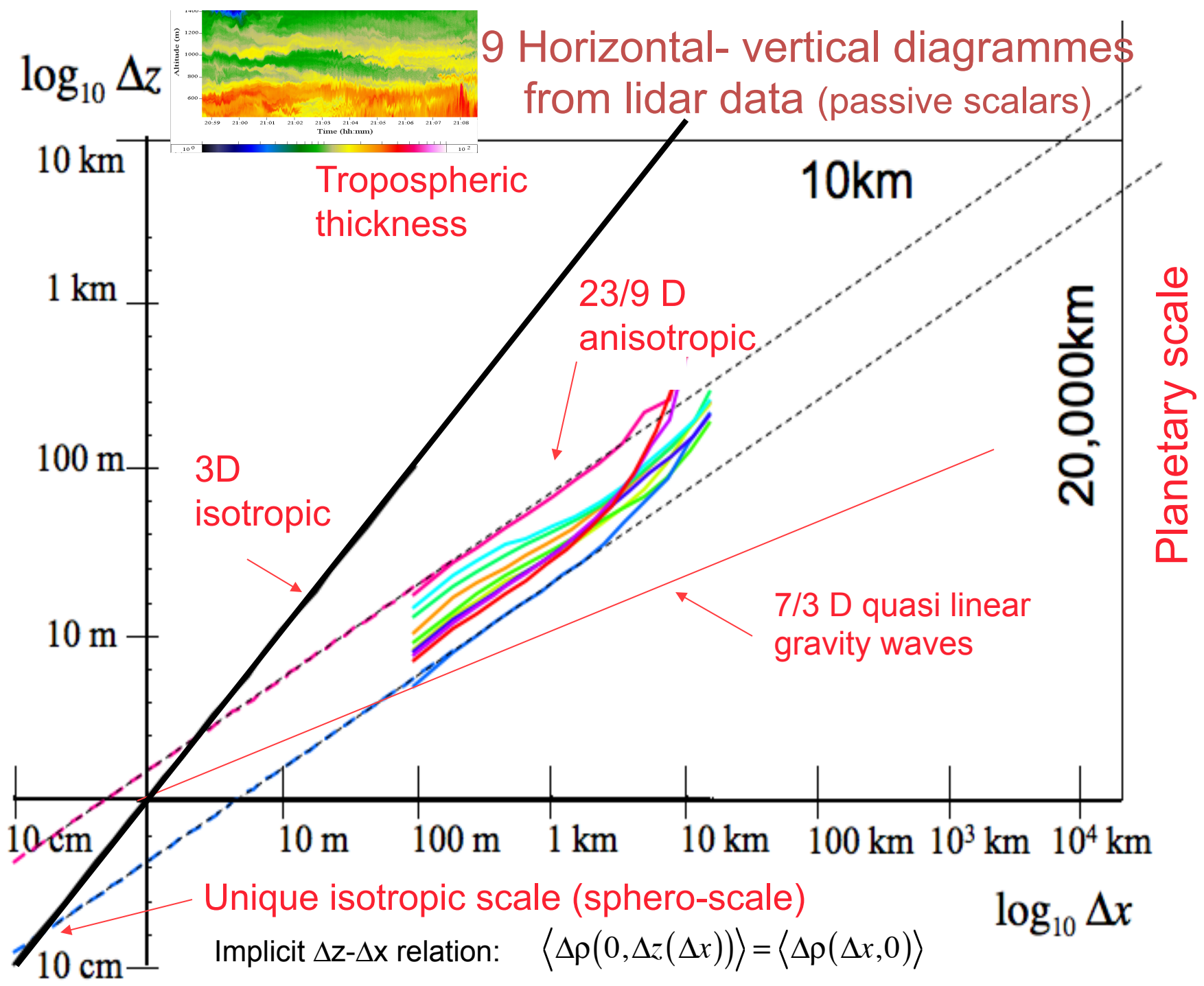


**Zoomed**

Lat/Long

48.50      48.60      48.70      48.80      48.90      49.00      49.10  
-122.59    -122.59    -122.59    -122.59    -122.59    -122.59    -122.59

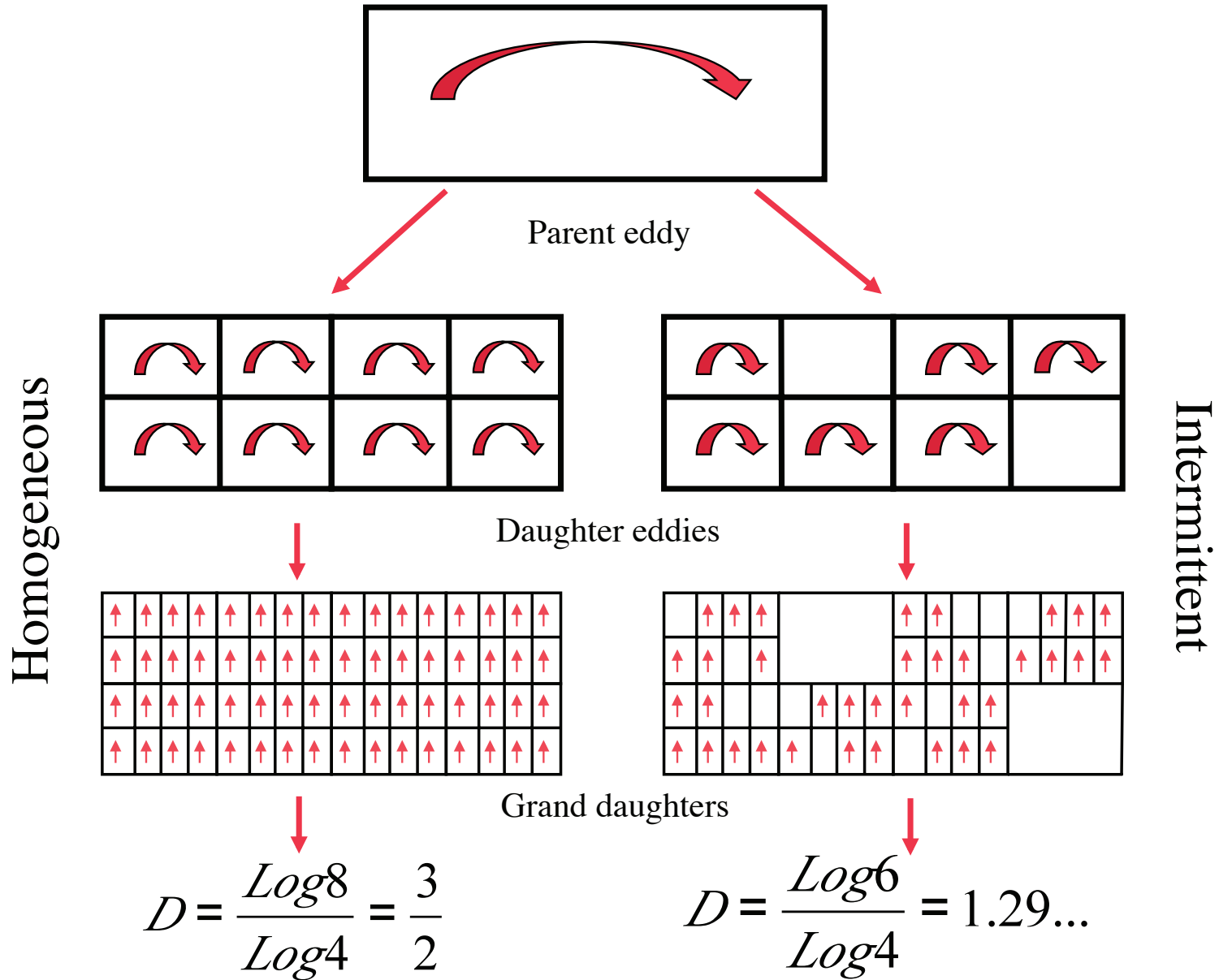




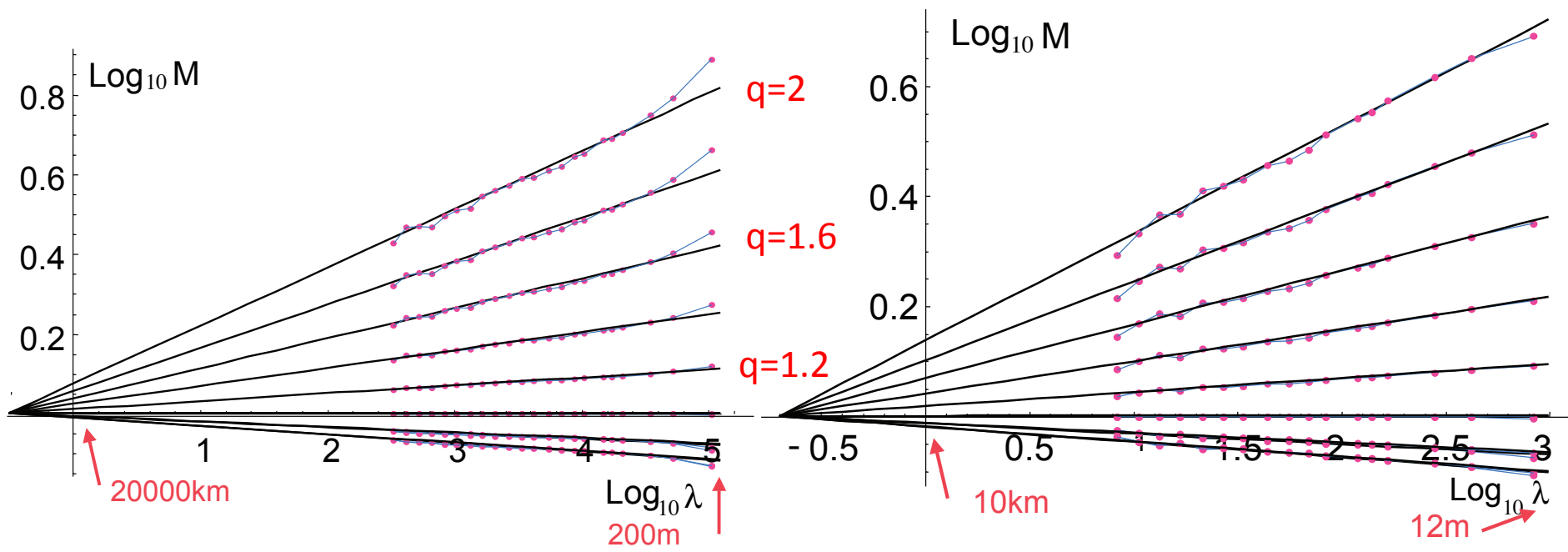
Fly by of anisotropic (multifractal,  
cascade) cloud



# Stratified CASCADES



# Lidar Backscatter



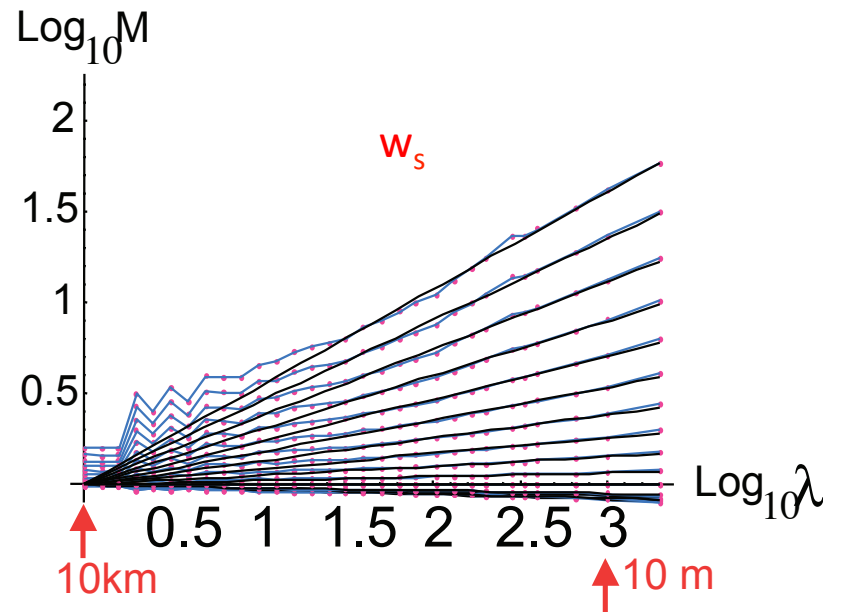
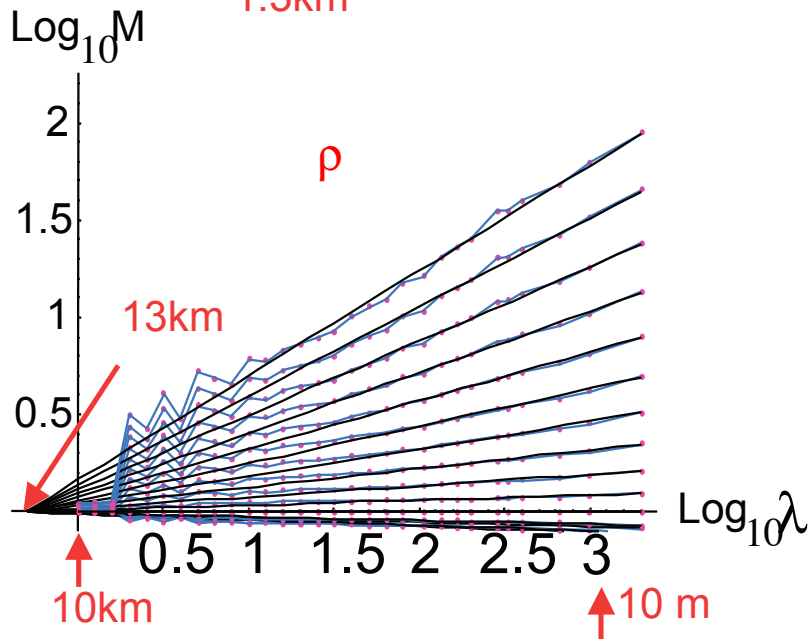
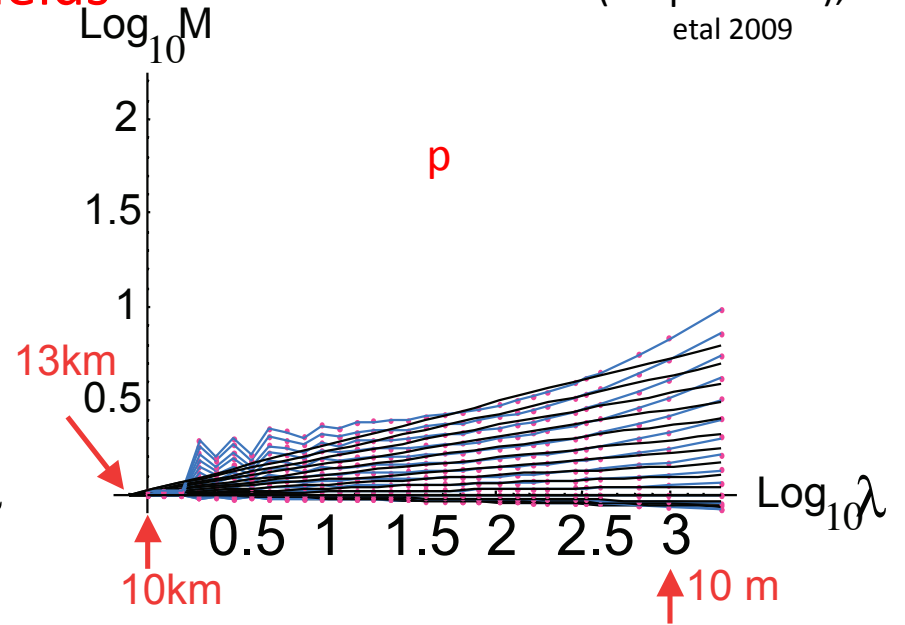
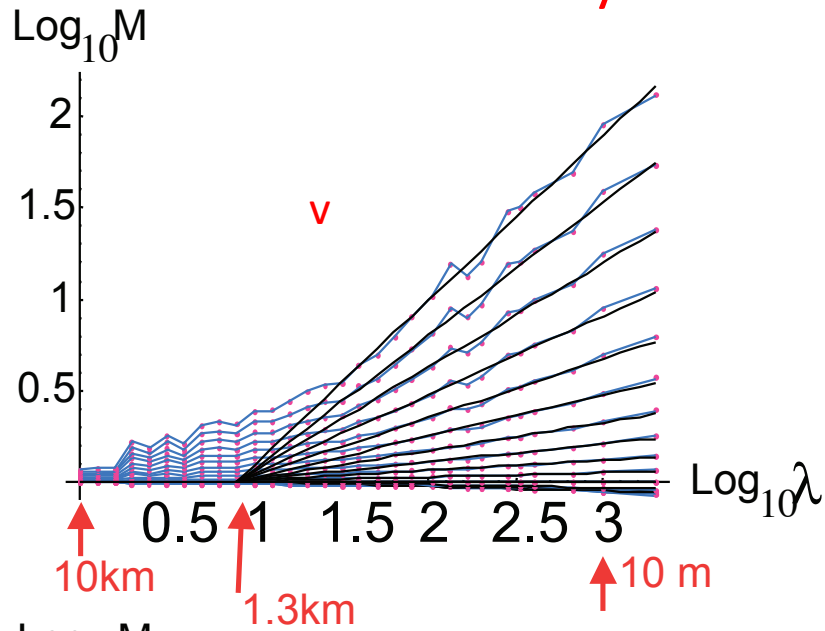
Horizontal

Vertical

# Vertical cascades:

dynamic fields

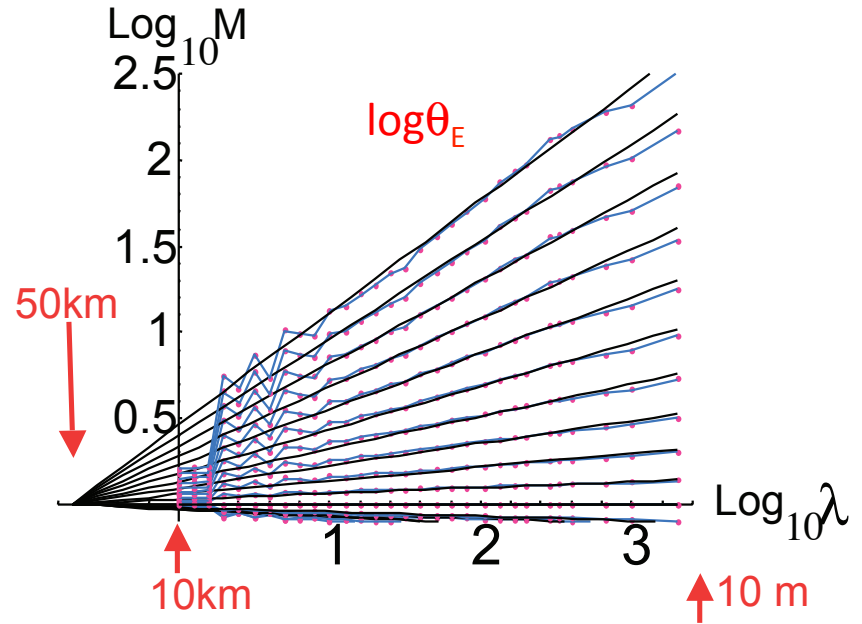
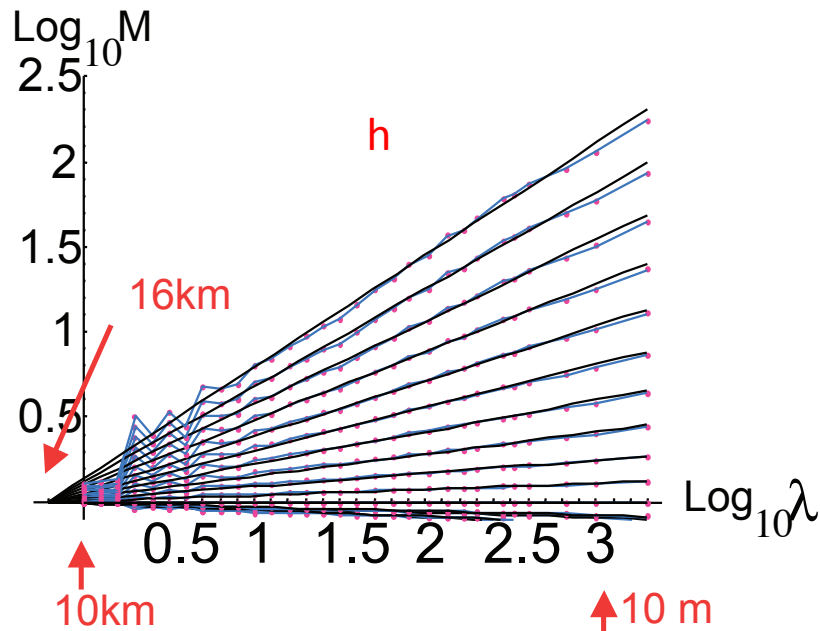
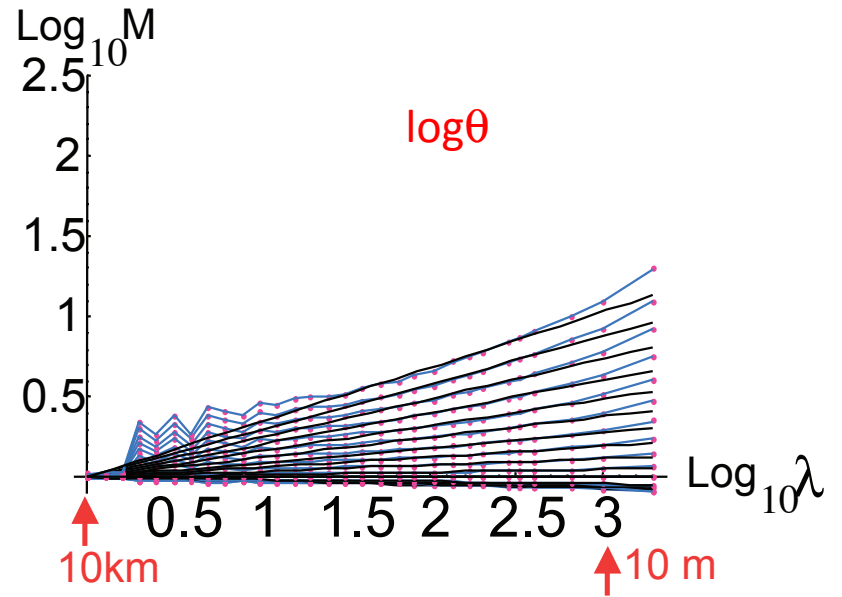
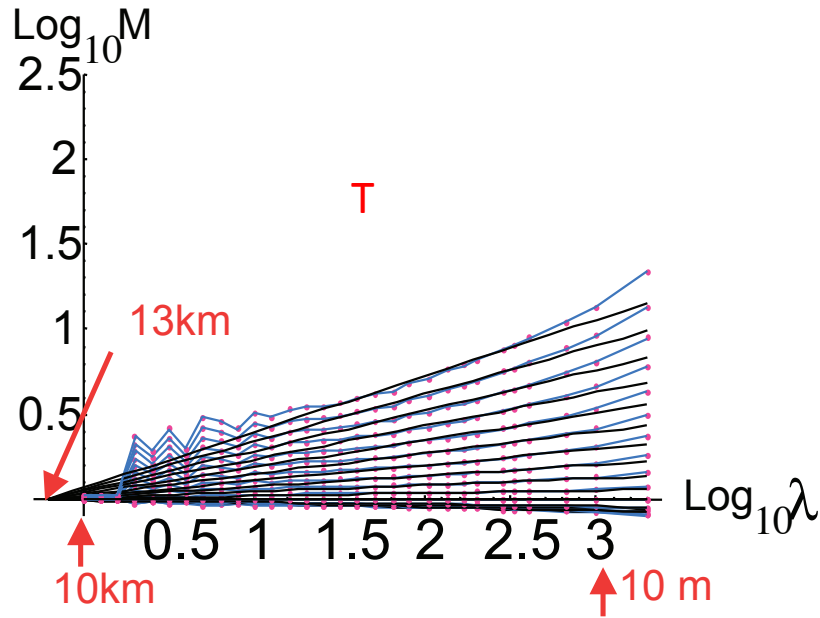
(drop sondes), L  
etal 2009





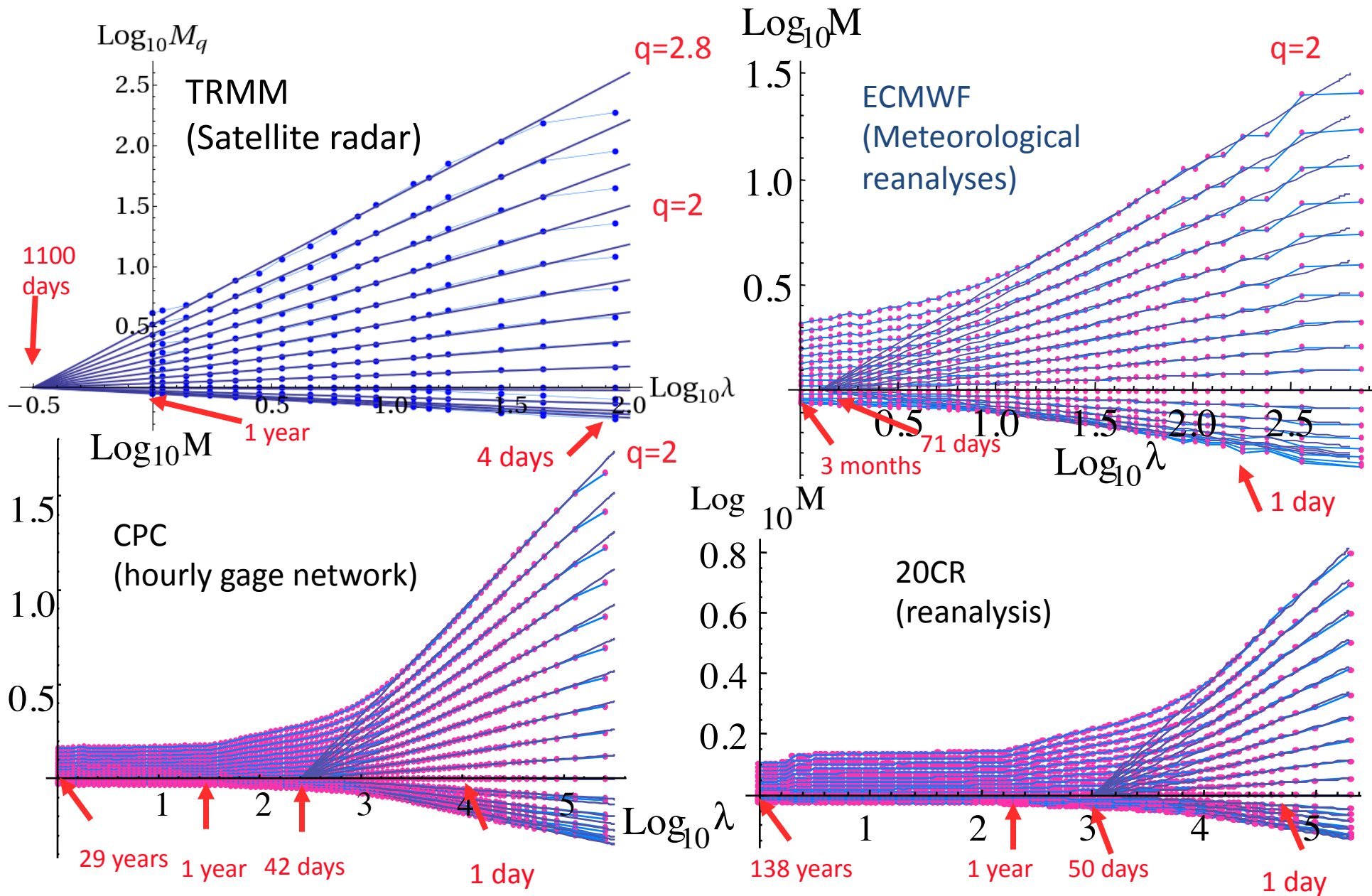
# Vertical cascades:

Thermodynamic fields



$$M = \langle \phi_\lambda^q \rangle / \langle \phi \rangle^q$$

# Rainrate Moments: (time)



Extension from space  
to space-time  
(including waves)

# Turbulence in Space-time (horizontal)

Theory (assuming largest eddies “sweep” smaller ones)

Observable  $\rightarrow$  Turbulent flux forcing

$$g_I^{-1}(\underline{r}, t) * I(\underline{r}, t) = \varphi(\underline{r}, t)$$

$$g_I(\underline{r}, t) \overset{F.T. \sim}{\leftrightarrow} \tilde{g}_I(\underline{k}, \omega)$$

propagator

$$\tilde{I}(\underline{k}, \omega) = \tilde{g}_I(\underline{k}, \omega) \tilde{\varphi}(\underline{k}, \omega)$$

$$\tilde{g}_{tur}(\underline{k}, \omega) = \left( i\omega' + \|\underline{k}\| \right)^{-H_{tur}}$$

$$\omega' = (\omega + \underline{k} \cdot \underline{\mu}) \sigma^{-1} \quad \|\underline{k}\| = (k_x^2 + k_y^2 / a^2)^{1/2}$$

$$\sigma = \left( 1 - (\mu_x^2 + a^2 \mu_y^2) \right)^{1/2}$$

$$\underline{\mu} = (\overline{v_x}, \overline{v_y}) / V_w \quad V_w = \epsilon_{L_e} L_e^{1/3}$$

EW/NS aspect ratio =  $a$

mean horizontal wind =  $(\overline{v_x}, \overline{v_y})$

Mean planetary scale energy flux  $\epsilon_{L_e}$


Planet size:  $L_e = 20000$  km

Pure (localized) turbulence propagator

# Turbulence and waves

Turbulence forcing

$$\tilde{I}(\underline{k}, \omega) = \tilde{g}_I(\underline{k}, \omega) \tilde{\varphi}(\underline{k}, \omega)$$

Turbulent flux  


Turbulence-waves

$$\tilde{g}_I(\underline{k}, \omega) = \underbrace{\tilde{g}_{wav}(\underline{k}, \omega) \tilde{g}_{tur}(\underline{k}, \omega)}_{\text{Wheeler Kiladis 1999 factorization}}$$

Wheeler Kiladis 1999  
factorization

Propagator symmetries, constraints

Reality

$$\tilde{g}(\underline{k}, \omega) = \tilde{g}^*(-\underline{k}, -\omega)$$

Space-time scaling

$$\tilde{g}(\lambda^{-1}(\underline{k}, \omega)) = \lambda^{-H} \tilde{g}(\underline{k}, \omega)$$

Causality

Poles of g in  $\omega$  plane are below real axis:

$$\omega' = -i\|\underline{k}\| \quad \text{OK since} \quad \|\underline{k}\| \geq 0$$

# Simple wave ansatz

## Simple scaling wave propagator

$$\tilde{g}_{wav}(\underline{k}, \omega) = \left( \omega'^2 / v_{wav}^2 - \|\underline{k}\|^2 \right)^{-H_{wav}/2}$$

Fractional (and anisotropic) wave equation propagator

$$H = H_{tur} + H_{wav}$$

## Dispersion relation:

$$\omega = -\underline{k} \cdot \underline{\mu} \pm \sigma v_{wav} \|\underline{k}\| \quad \leftarrow \omega' = \pm v_{wav} \|\underline{k}\|$$

## Spectral density

$$P_I(\underline{k}, \omega) = P_\varphi(\underline{k}, \omega) \left| \tilde{g}_I \right|^2$$

Turbulent part                      Wave part

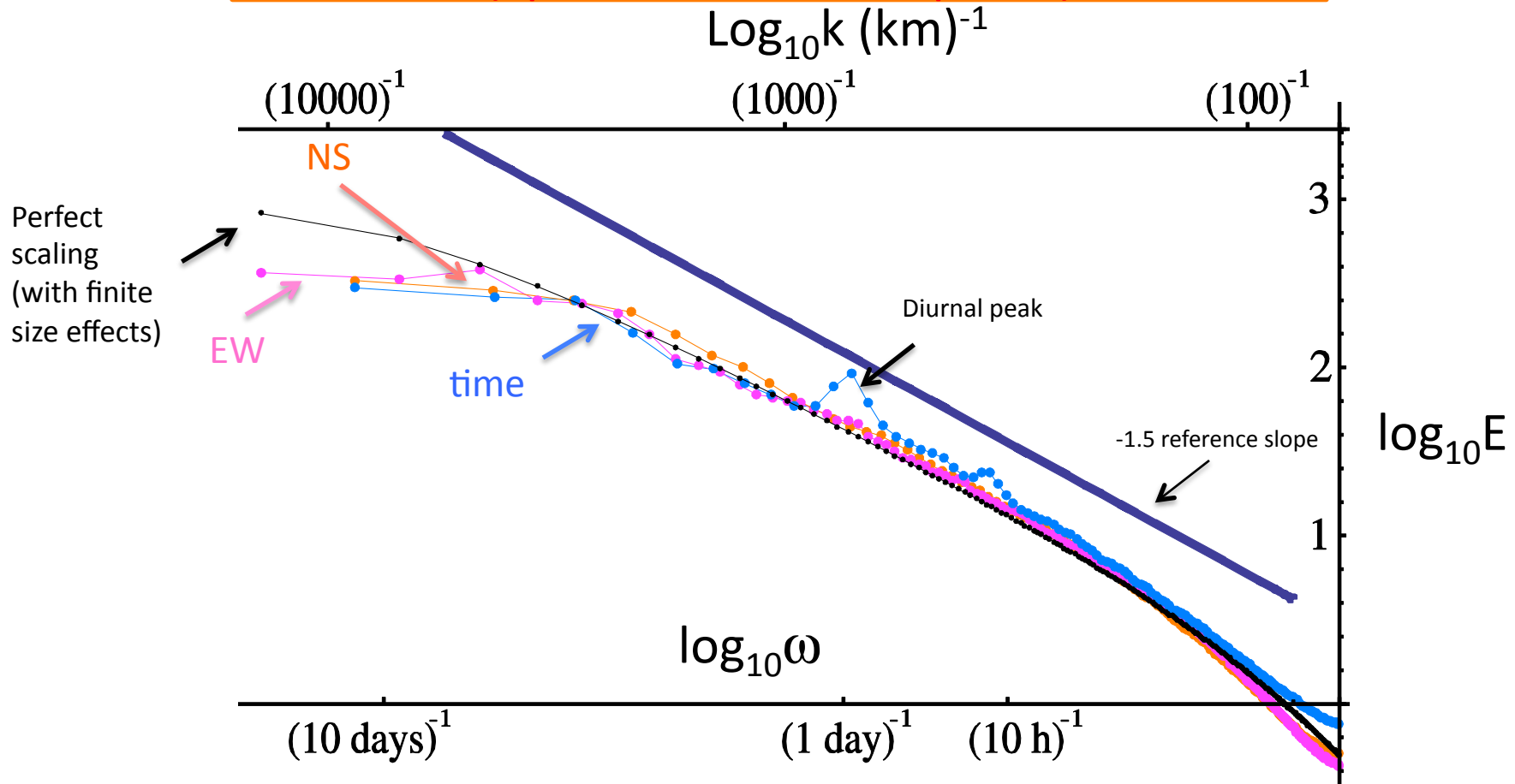
$$\left| \tilde{g}_I \right|^2 = \left| \tilde{g}_{tub} \right|^2 \left| \tilde{g}_{wav} \right|^2 = \left( \omega'^2 + \|\underline{k}\|^2 \right)^{-H_{tur}} \left( \omega'^2 / v_{wav}^2 - \|\underline{k}\|^2 \right)^{-H_{wav}/2}$$

$$P_\varphi(\underline{k}, \omega) = P_0 \left( \omega'^2 + \|\underline{k}\|^2 \right)^{-s_\varphi/2} \quad \leftarrow \text{Spectrum of turbulence forcing}$$

# 1400 MTSAT IR images -30° to

+40° latitude, Pacific  
(Spectrum, 1-D subspaces)

1 hour, 10 km  
resolution



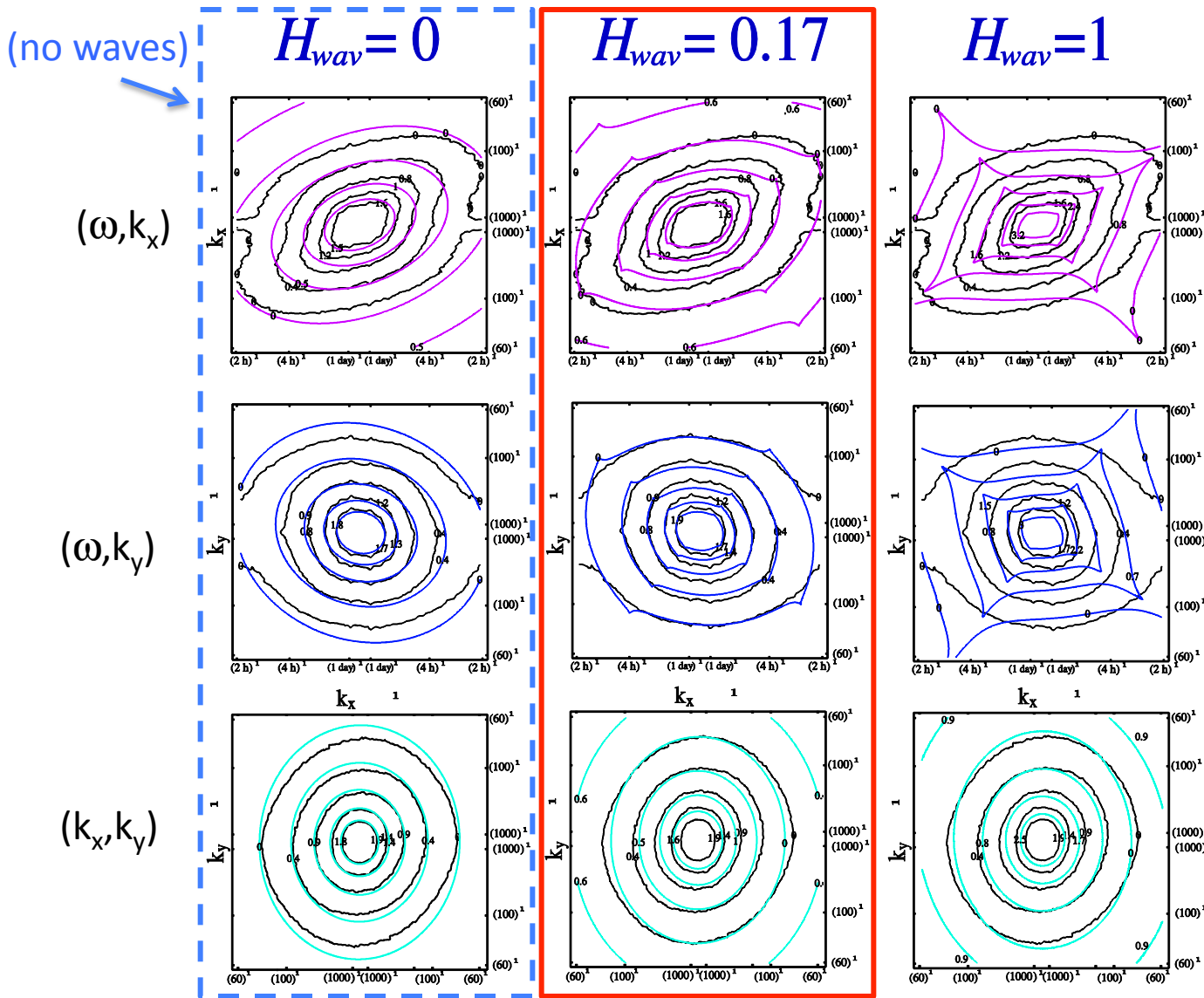
Space-time scaling is accurately respected:

$$\tilde{g}(\lambda^{-1}(\underline{k}, \omega)) = \lambda^{-H} \tilde{g}(\underline{k}, \omega)$$

implies

$$E(k_x) \approx k_x^{-\beta}; \quad E(k_y) \approx k_y^{-\beta}; \quad E(\omega) \approx \omega^{-\beta}$$

# Spectrum, 2-D subspaces



(Classical wave equation:  $H_{wav}=2$ )

## Parameters

$$H_{wav} = 0.17 \pm 0.04$$

$$H_{tur} = 0.09 \pm 0.05$$

$$s_\phi = 2.88 \pm 0.01$$

$$V_w = 41 \pm 3 \text{ km/h}$$

$$\tau_w = L_e / V_w \approx 20 \pm 1 \text{ days}$$

$$a \approx 1.2 \pm 0.1$$

$$\mu_x \approx -0.3 \pm 0.1; (v_x - 12 \pm 4 \text{ km/h})$$

$$\mu_y \approx 0.10 \pm 0.08; (v_y \approx 4 \pm 3 \text{ km/h})$$

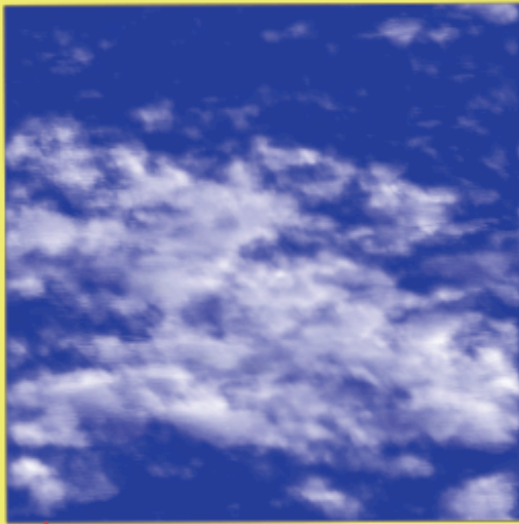
Spectral density: 
$$P_l(\underline{k}, \omega) \propto \left( \omega'^2 + \|\underline{k}\|^2 \right)^{-H_{tur} - s_\phi} \left( \omega'^2 / v_{wav}^2 - \|\underline{k}\|^2 \right)^{-H_{wav}/2};$$

$$H = H_{tur} + H_{wav}$$

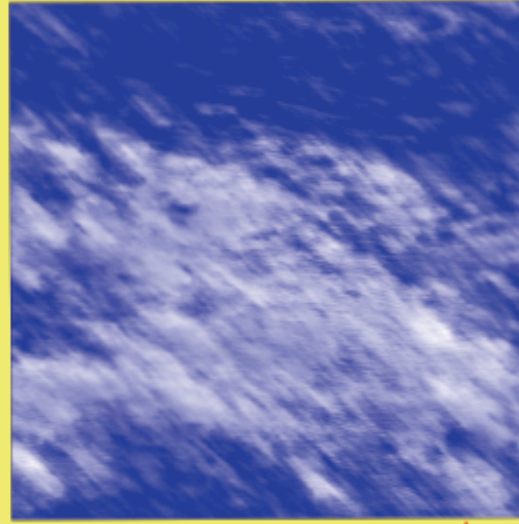


Cascades from localized to increasingly unlocalized structures:

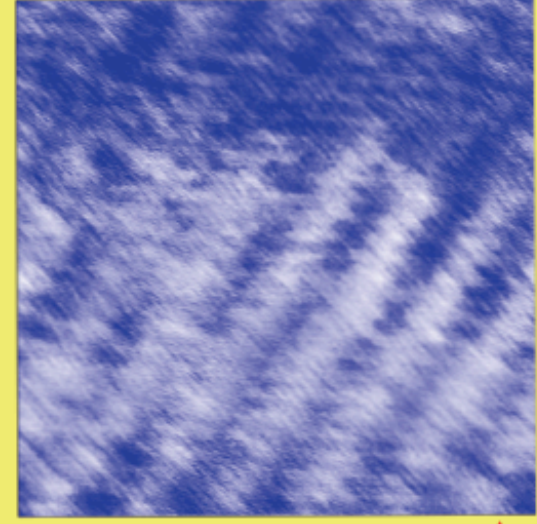
$$H_{\text{wav}} = 1/3 - H_{\text{tur}}$$



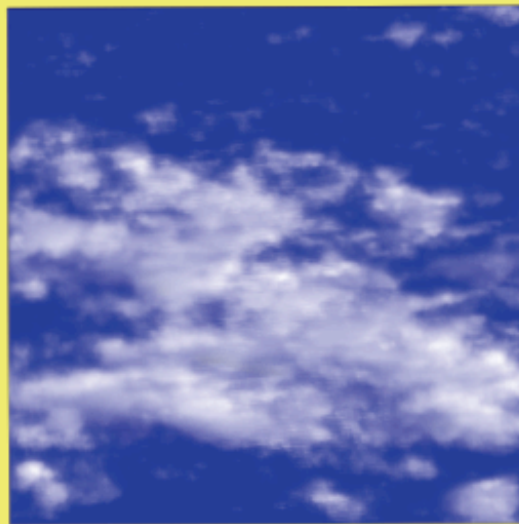
↑  $H_{\text{wav}} = 0.22$



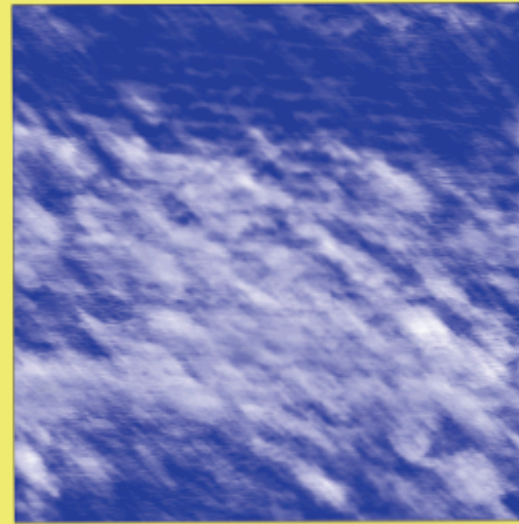
$H_{\text{wav}} = 0.37$  ↑



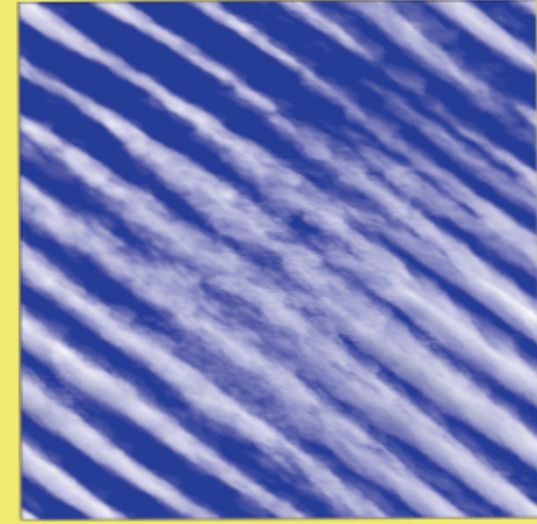
$H_{\text{wav}} = 0.52$  ↑



$H_{\text{wav}} = 0.0$

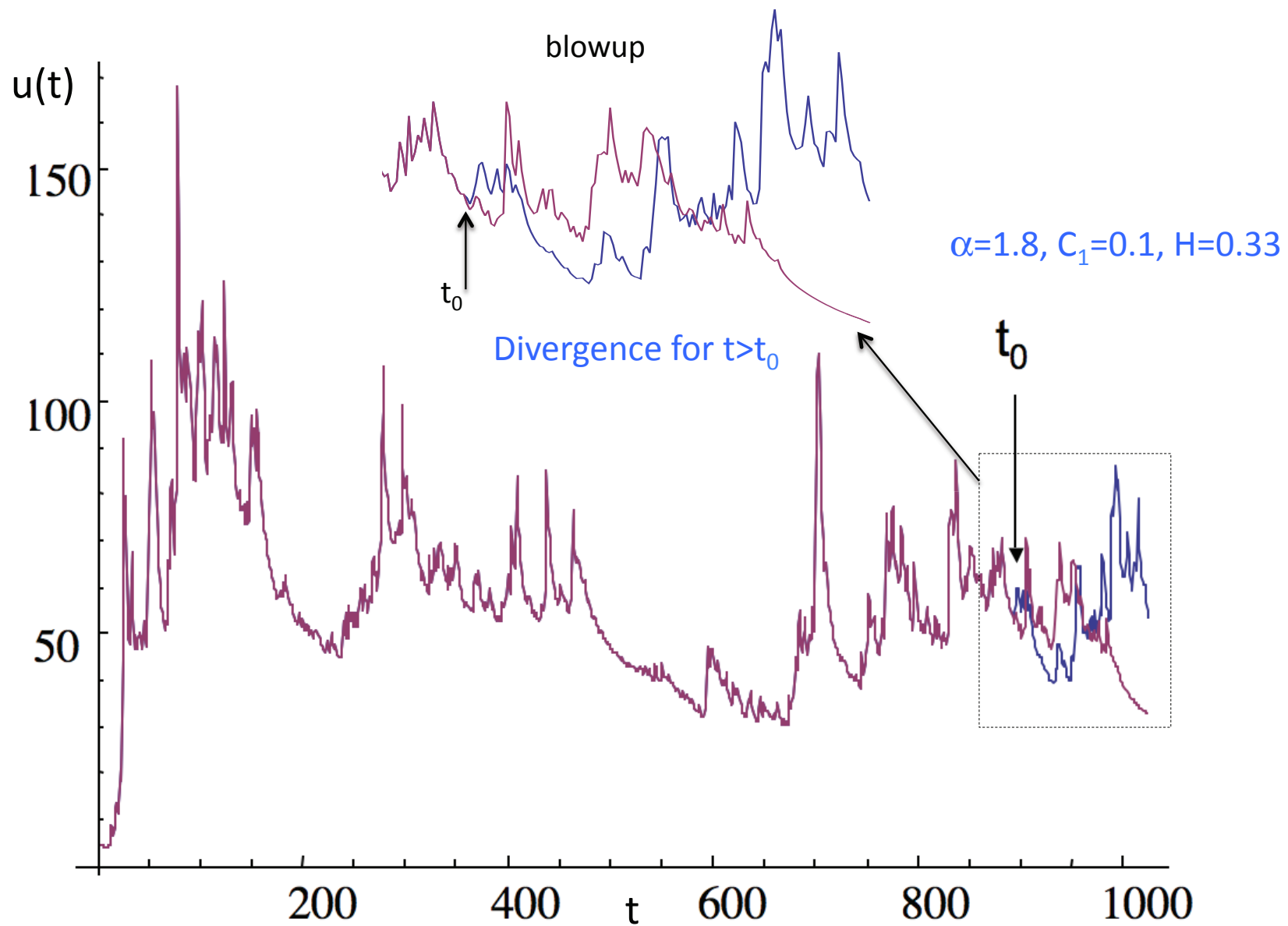


$H_{\text{wav}} = 0.33$



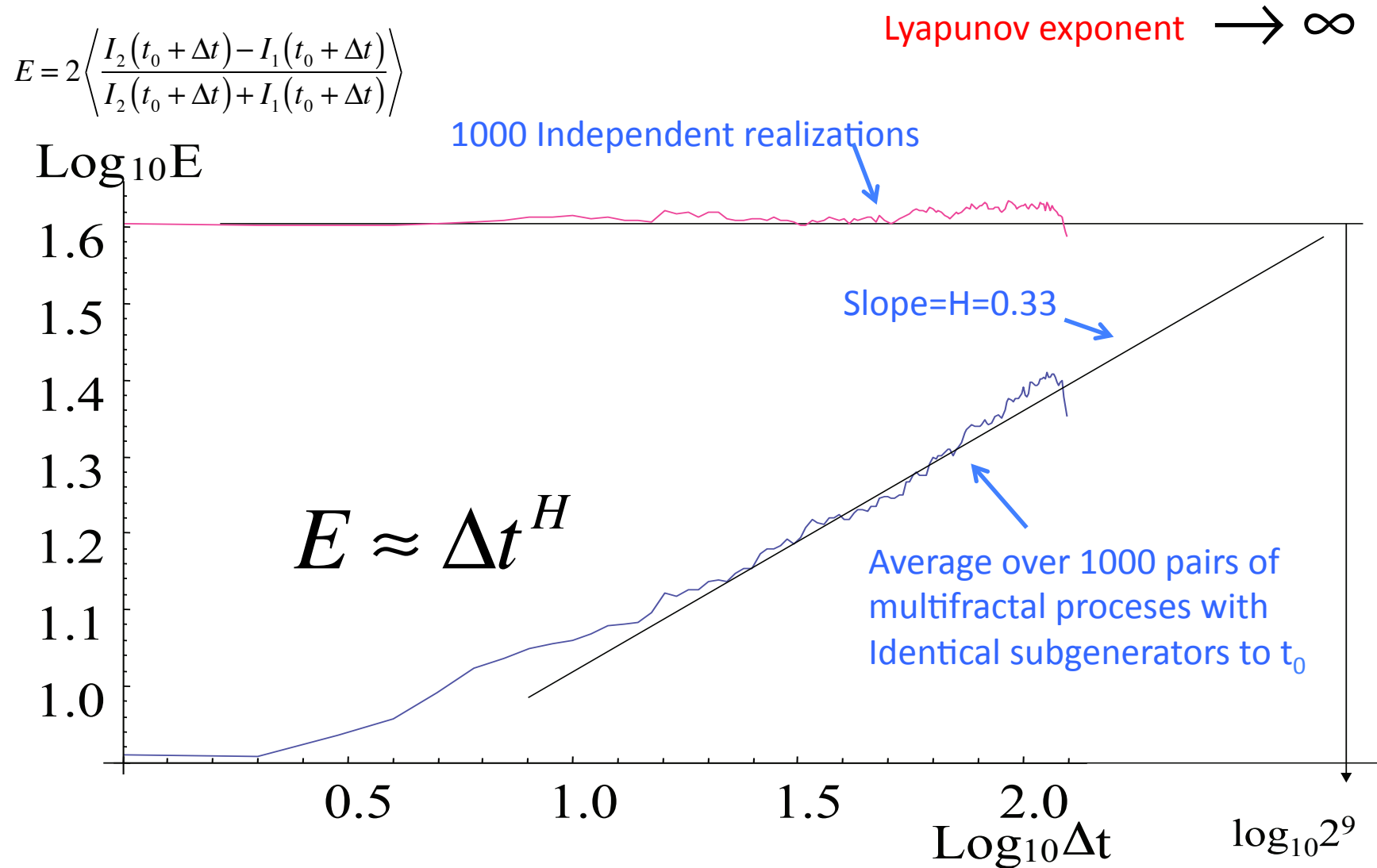
$H_{\text{wav}} = 0.47$

# Predictability and stochastic forecasting

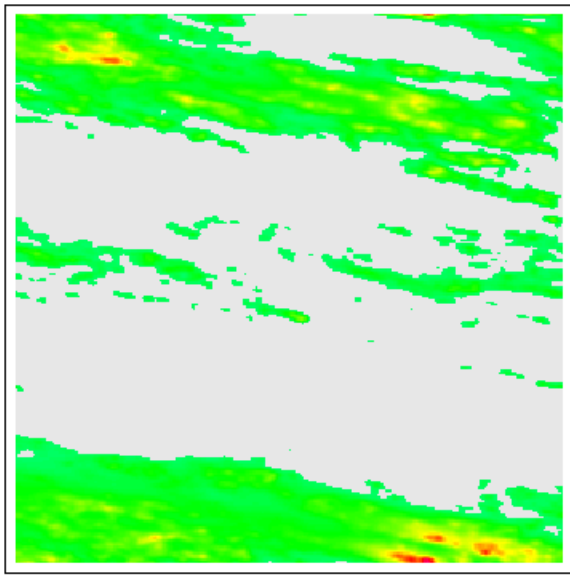


Two multifractal processes with Identical subgenerators to  $t_0$

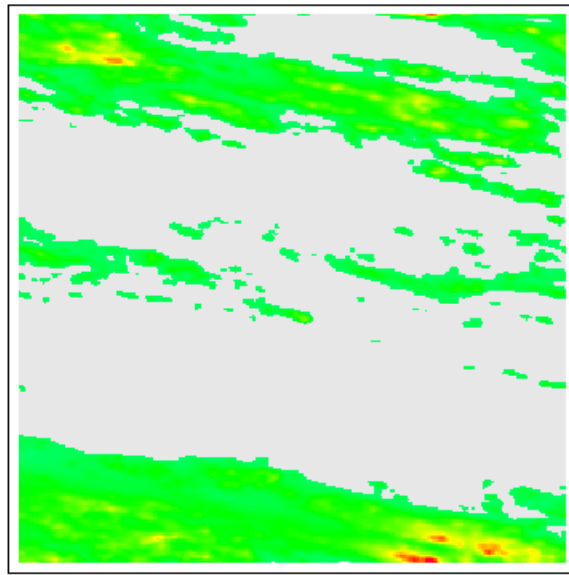
# Algebraic divergence of realizations



# Space-time Cascades, stochastic nowcasting (rain)

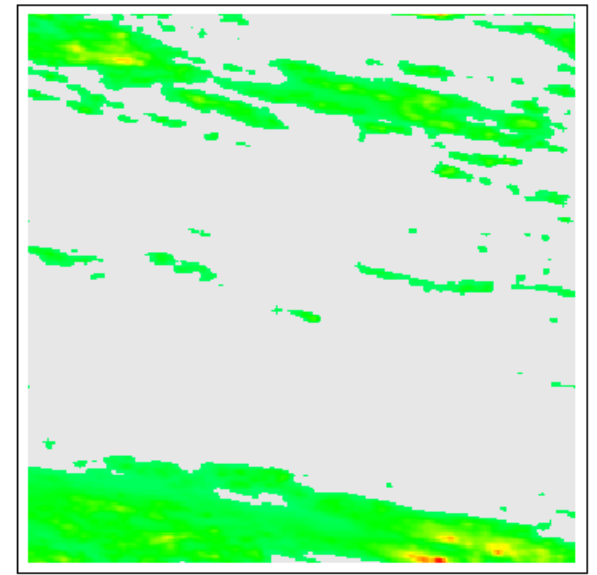


Realization A



Realization B

(all same initially)



Forecast based on first  
16 time steps

# The macroweather regime

Low frequency cascades

Time scales  $\approx > 10$  days ( $\tau > \tau_w$ )

...Predicting the spectral plateau /  
macroweather regime

# “First principles” predictions Atmosphere:

The large scale winds and weather-climate transition scale

## Power:

Solar heating, top of the atmosphere:  $\approx 10^3 \text{ W/m}^2$

Absorbed  $\approx 2 \times 10^2 \text{ W/m}^2$

$\approx 2\%$  Converted to K.E.  $\approx 4 \text{ W/m}^2$

## Energy flux:

If power is distributed over the troposphere,  $10^4 \text{ m}$  thick, density,  $0.75 \text{ Kg/m}^3$

$\epsilon \approx 5 \times 10^{-4} \text{ m}^2 \text{ s}^{-3}$

**c.f. modern value  $10^{-3} \text{ m}^2 \text{ s}^{-3}$**

Prediction using horizontal relation:

$$\Delta v = \epsilon^{1/3} \Delta x^{1/3}$$

## Scales:

Length:  $L \approx 2 \times 10^7 \text{ m}$

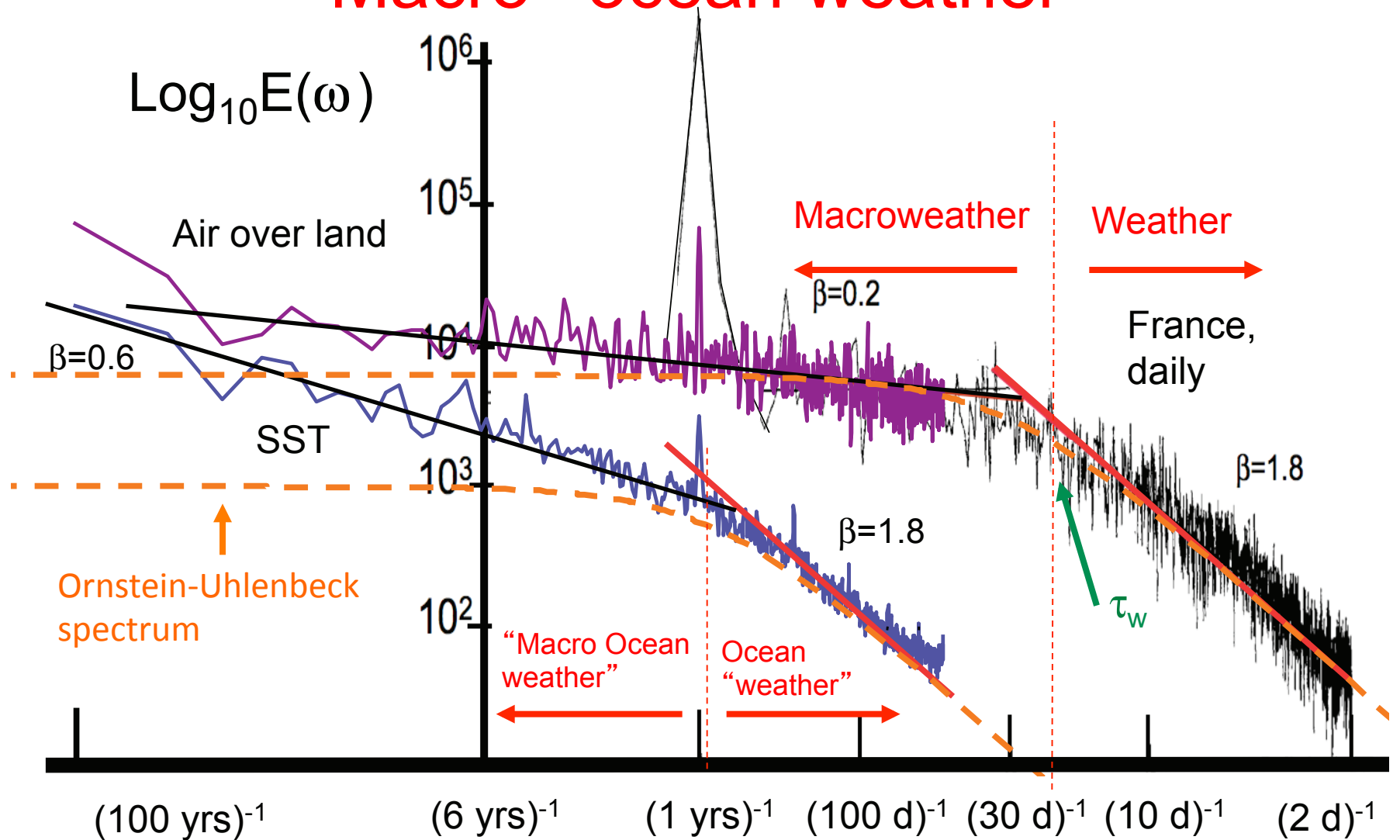
Velocity:  $V \approx \epsilon^{1/3} L^{1/3} \approx 21 \text{ m/s}$

Time:  $T = L/V \approx 10^6 \text{ s} = 11 \text{ days}$

c.f. empirical antipodes velocity difference

$17.4 \pm 5.7 \text{ m/s}$

# Macroweather, Macro “ocean weather”



**Ocean Drifter data:**  $\epsilon_o$   
 $\approx 10^{-8} \text{ m}^2/\text{s}^3$   
 $\tau_o \approx \epsilon_o^{1/3} L^{-2/3} \approx 1 \text{ yr}$

**Reanalyses:**  
 $\epsilon_w \approx 10^{-3} \text{ m}^2/\text{s}^3$   
 $\tau_w \approx \epsilon_w^{1/3} L^{-2/3} \approx 10 \text{ dys}$



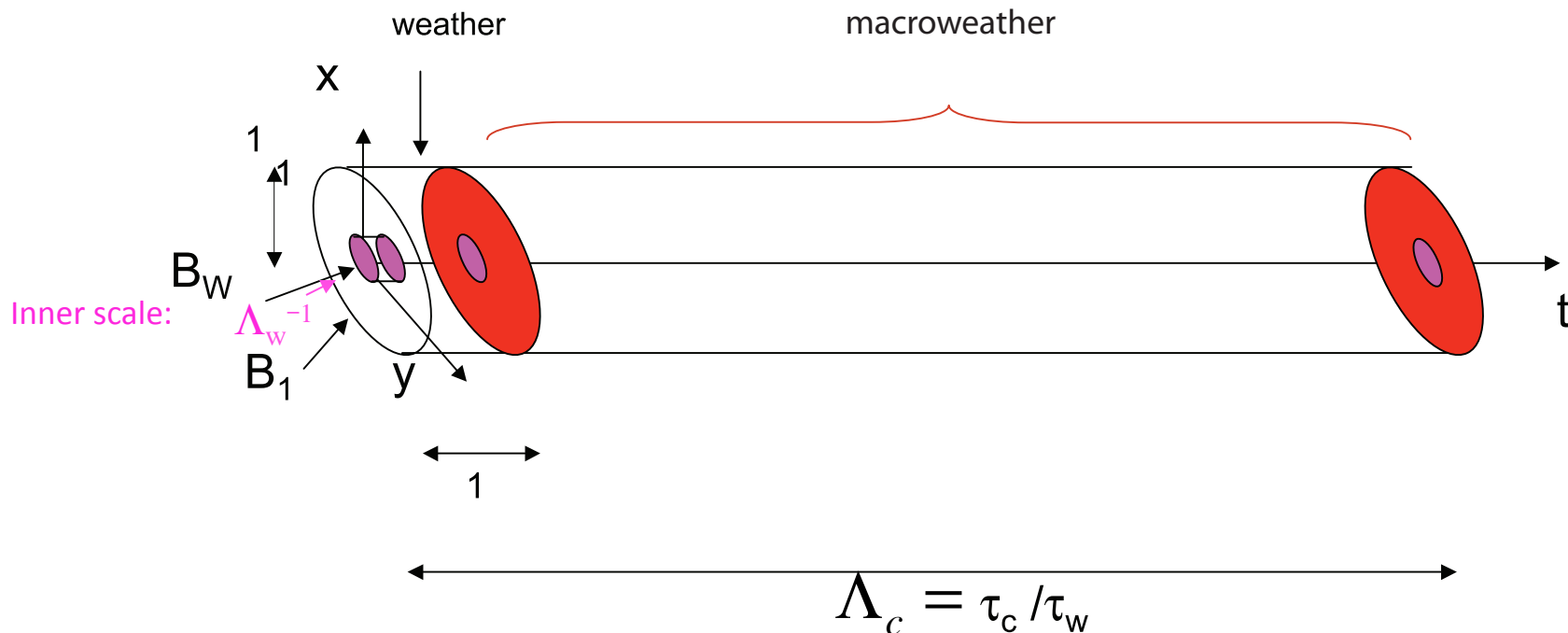
# From the Weather to macroweather: a dimensional transition

$$\Gamma(\underline{r}, t) = \underbrace{\int_{\Lambda_w^{-1} B_w}^1 \int_{B_1}^{B_w} \gamma(\underline{r} - \underline{r}', t - t') g(\underline{r}', t') d\underline{r}' dt'}_{\Gamma_w} + \underbrace{\int_1^{\Lambda_c} \int_{B_w}^{B_1} \gamma(\underline{r} - \underline{r}', t - t') g(\underline{r}', t') d\underline{r}' dt'}_{\Gamma_{mw}}$$

$$\varepsilon_\lambda = e^{\Gamma_\lambda}$$

The generator=  
Additive process

Space-time region of generator interactions:



For  $\tau_c \gg \tau_w$ , space-time interaction domain becomes pencil-like (1-D), not 3-D:

**Dimensional transition**

# Implications of the FIF model for $\tau > \tau_w$

(the macroweather regime)

$$\varepsilon_{\Lambda_w, \Lambda_c}(\underline{r}, t) \approx e^{\Gamma_w(\underline{r}, t) + \Gamma_{mw}(\underline{r}, t)} = \varepsilon_{\Lambda_w}(\underline{r}, t) \varepsilon_{\Lambda_c}(t)$$

Weather-macroweather factorization

Observable:

$$I = \varepsilon * [[\Delta r, \Delta t]]^{-(D-H)}$$

Autocorrelations:

$$\langle \Delta I_w(\Delta t)^2 \rangle \propto \Delta t^{2H-K(2)}$$

Weather regime

$$\langle I_{mw}(t) I_{mw}(t - \Delta t) \rangle \propto \Delta t^{-1}$$

macroweather regime

Low frequency divergence

Spectra:

$$E(k) \approx k^{-\beta_w}; \quad k > L_w^{-1}$$

$$E(\omega) \approx \omega^{-\beta_w}; \quad \omega > \tau_w^{-1}$$

Weather regime

$$E(\omega) \approx \omega^{-\beta_{mw}}; \quad \tau_c^{-1} < \omega < \tau_w^{-1}$$

macroweather regime

Exponents:

$$\beta_w = 1 + 2H - K(2)$$

$$0.2 < \beta_{mw} < 0.4$$

Independent of H, C<sub>1</sub>, weak dependence on  $\alpha$ , scale range  $\Lambda_c$

# Comparing the data at 75°N (20thC reanalysis) and the cascade simulations

The cascade simulations depend on the following parameters for the temperatures:

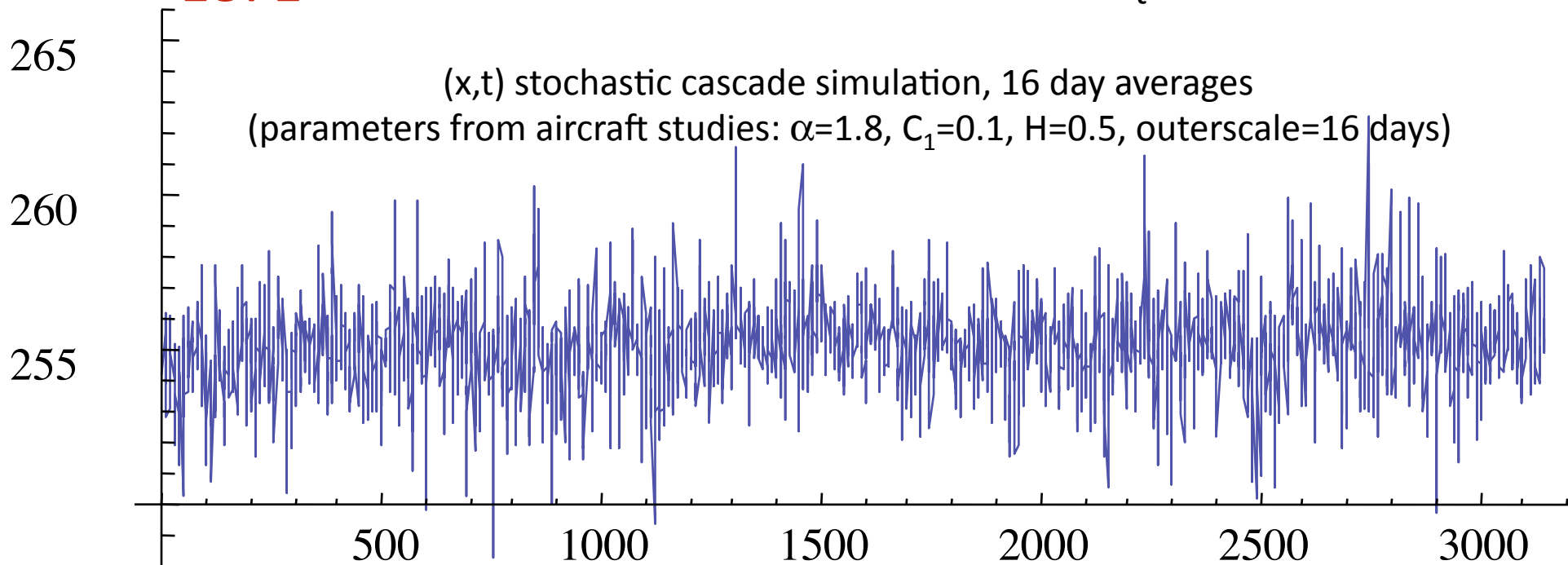
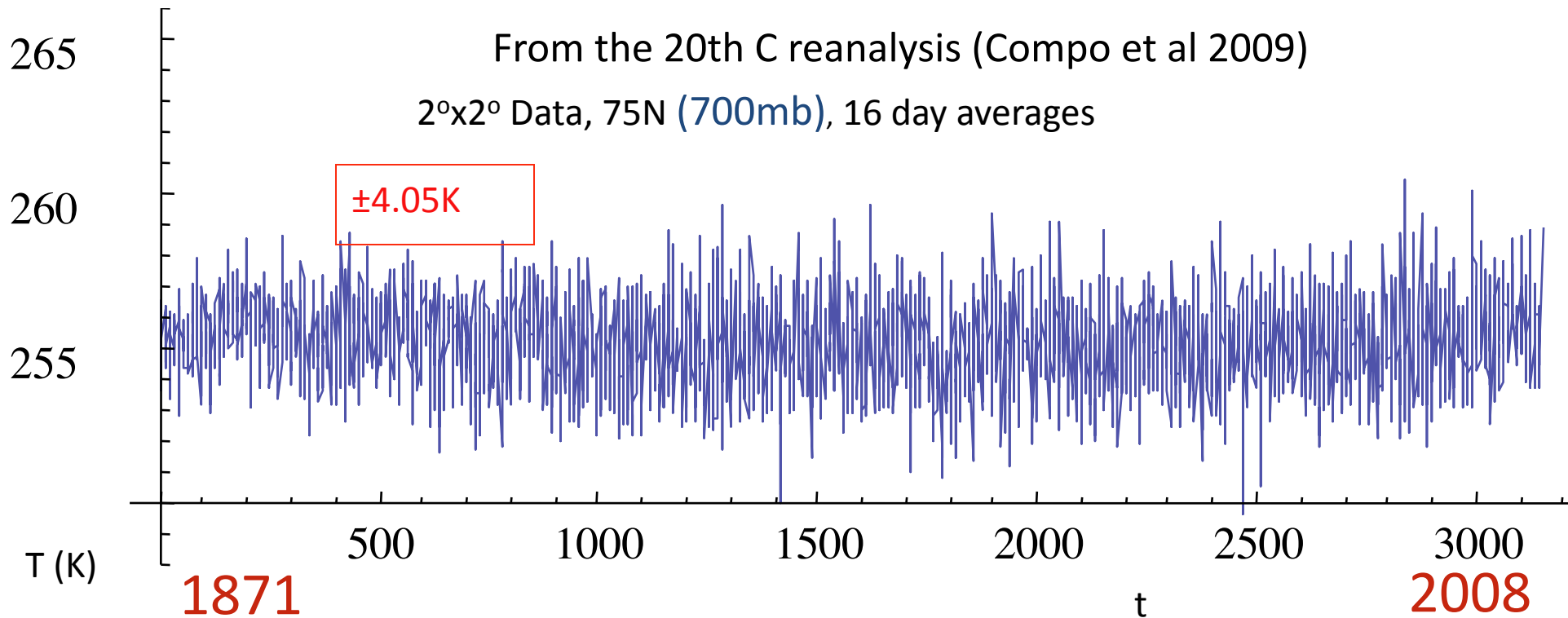
These following turbulent quantities were measured by the Pacific 2004 experiment using the Gulfstream 4 aircraft over the Northern Pacific ocean, at 200mb. The data were taken at 4Hz ( $\approx 280\text{m}$ ) resolution:

Cascade exponents:  $\alpha = 1.8$ ,  $C_1 = 0.1$

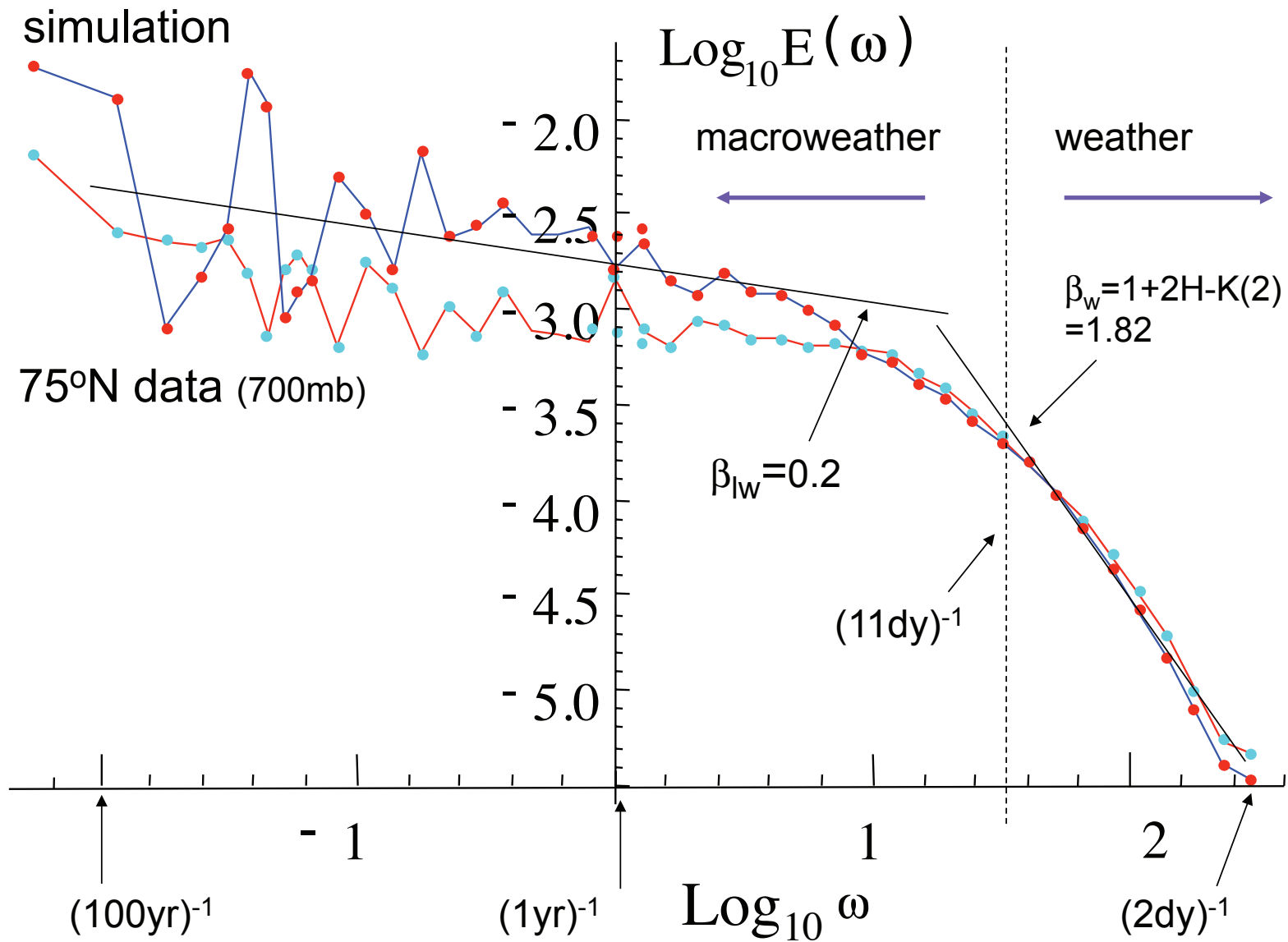
Scale by scale nonconservation exponent:  $H = 0.5$

Average energy flux:  $5 \times 10^{-4} \text{m}^2/\text{s}^3$

The only parameter measured by the reanalysis was the standard deviation of the daily temperature at 75°N, 700 mb:  $\pm 4.05\text{K}$



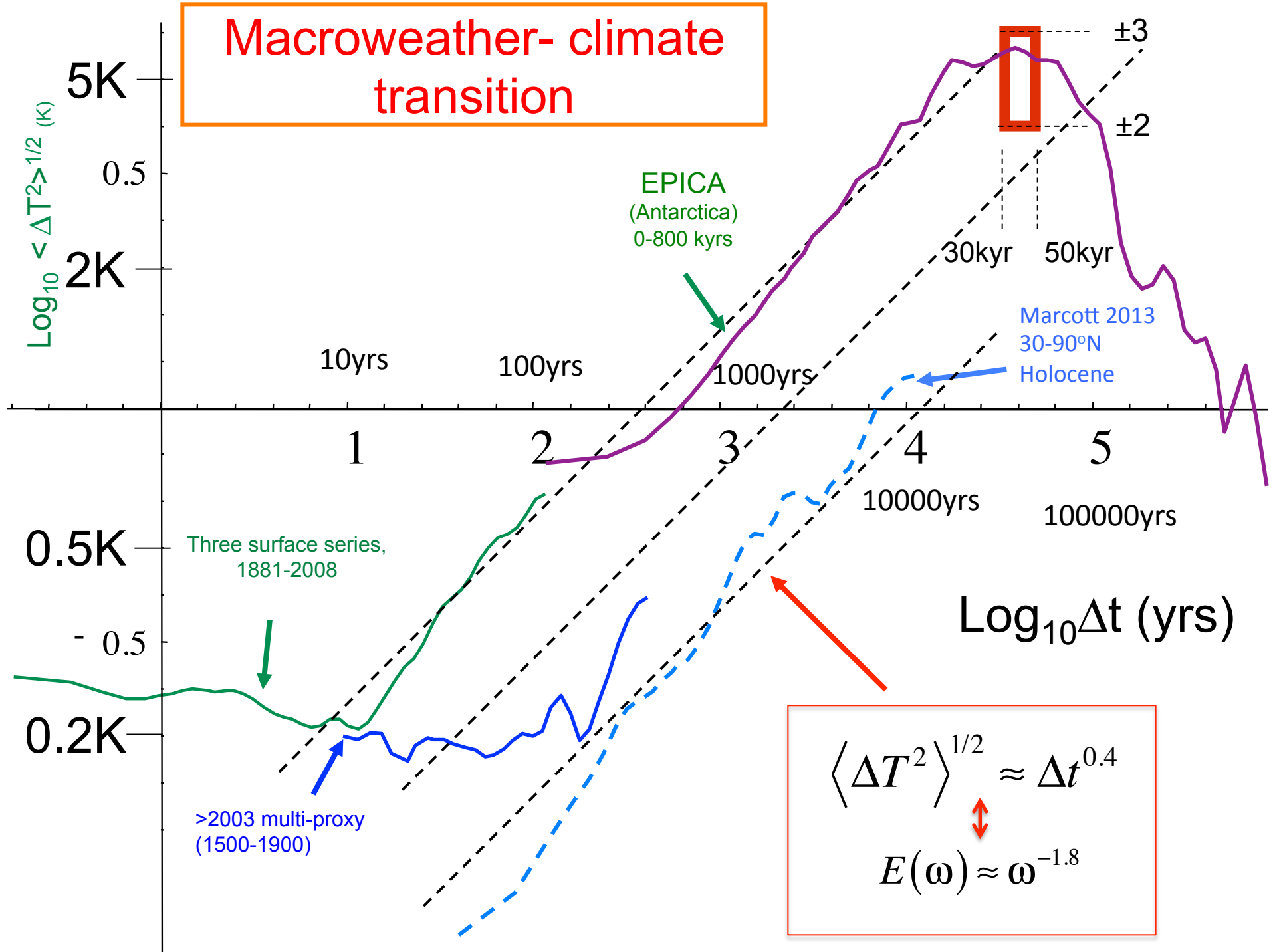
# The weather cascade spectra calibrated with aircraft data





The climate

# Macroweather- climate transition

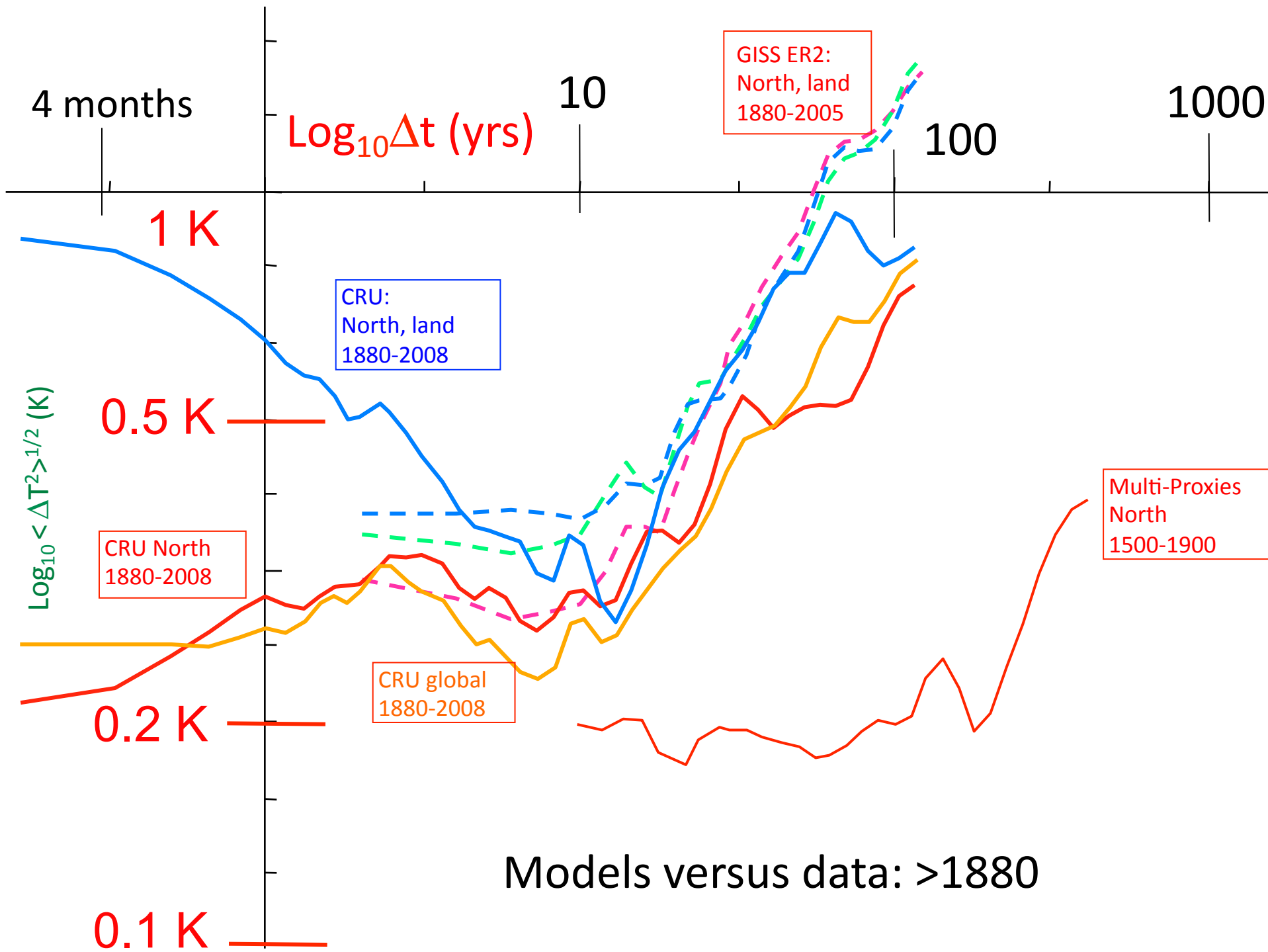


$$\langle \Delta T^2 \rangle^{1/2} \approx \Delta t^{0.4}$$

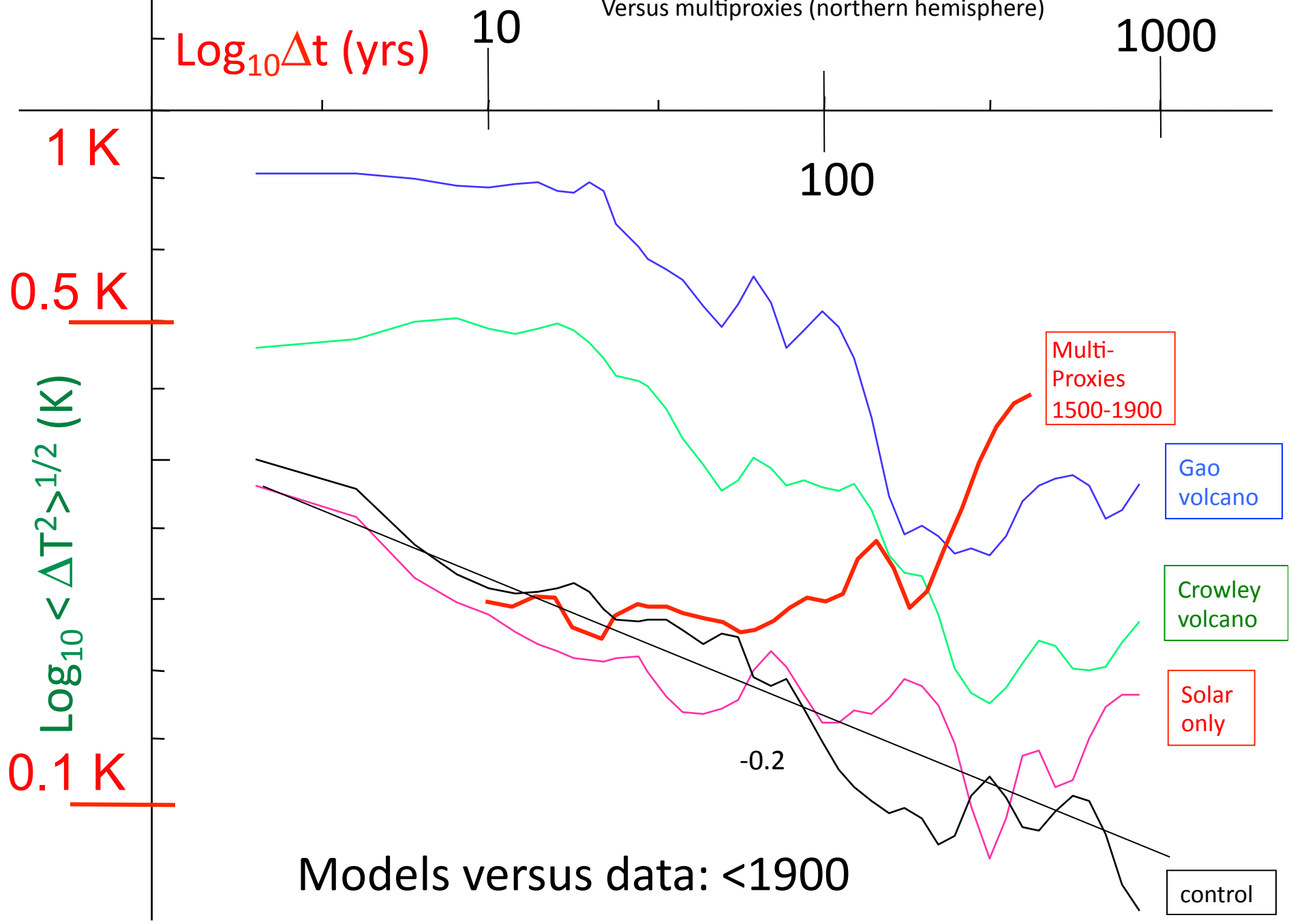
$$E(\omega) \approx \omega^{-1.8}$$



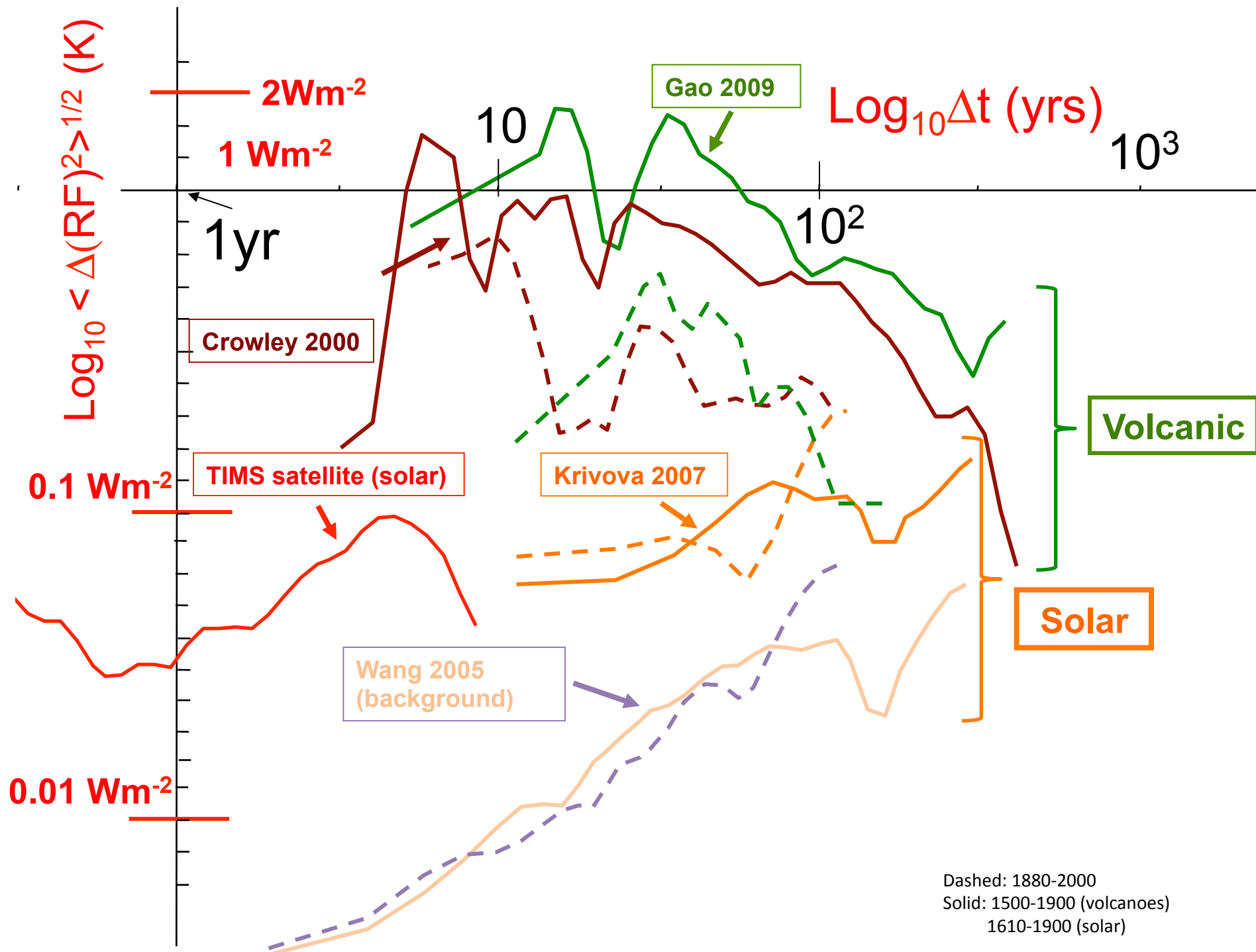
Do GCM' s predict the  
climate.... Or Macroweather?



GISS-ER2 simulations 1000-1900 (Northern Hemisphere, land only)  
Versus multiproxies (northern hemisphere)



Models versus data: <1900



# Climate

Overall weather – climate process

$$\varepsilon_{w,c}(\underline{r}, t) = \varepsilon_{w,mw}(\underline{r}, t) \varepsilon_c(\underline{r}, t)$$

Weather- macroweather cascade process (previous)

Low frequency space-time climate flux (new)

Weather/macroweather flux factorization:

$$\varepsilon_{w,mw}(\underline{r}, t) \approx \varepsilon_w(\underline{r}, t) \varepsilon_{mw}(t)$$

Macroweather: temporal variability only

# Space-time Macroweather-climate statistical factorization

$$\varepsilon_{\tau}(\underline{r}, t) = \frac{1}{\tau} \int_t^{t+\tau} \varepsilon(\underline{r}, t') dt' \quad \leftarrow \text{The flux at resolution } \tau$$

Weather-climate process at resolution  $\tau$

$$\varepsilon_{w,c,\tau}(\underline{r}, t) = \overbrace{\varepsilon_{w,mw,\tau}(\underline{r}, t)}^{\text{Weather-climate process at resolution } \tau} = \varepsilon_{w,\tau}(\underline{r}, t) \varepsilon_{mw,\tau}(t) \varepsilon_{c,\tau}(\underline{r}, t)$$

Weather variability averaged out

$$\varepsilon_{w,\tau}(\underline{r}, t) \approx 1; \quad \tau > \tau_w$$

$$\varepsilon_{c,\tau}(\underline{r}, t) \approx \varepsilon_{c,\tau}(\underline{r}); \quad \tau < \tau_c$$

climate too slow

Spatial variability: climatic zones

**Prediction: Space-time Statistical factorization in the macroweather regime**

$$\varepsilon_{w,c,\tau}(\underline{r}, t) \approx \begin{cases} \varepsilon_{mw,\tau}(t) \varepsilon_{c,\tau_c}(\underline{r}); & \tau_w < \tau < \tau_c & \text{macroweather} \\ \varepsilon_{c,\tau}(\underline{r}, t); & \tau > \tau_c & \text{climate} \end{cases}$$

Macroweather spectral Prediction:

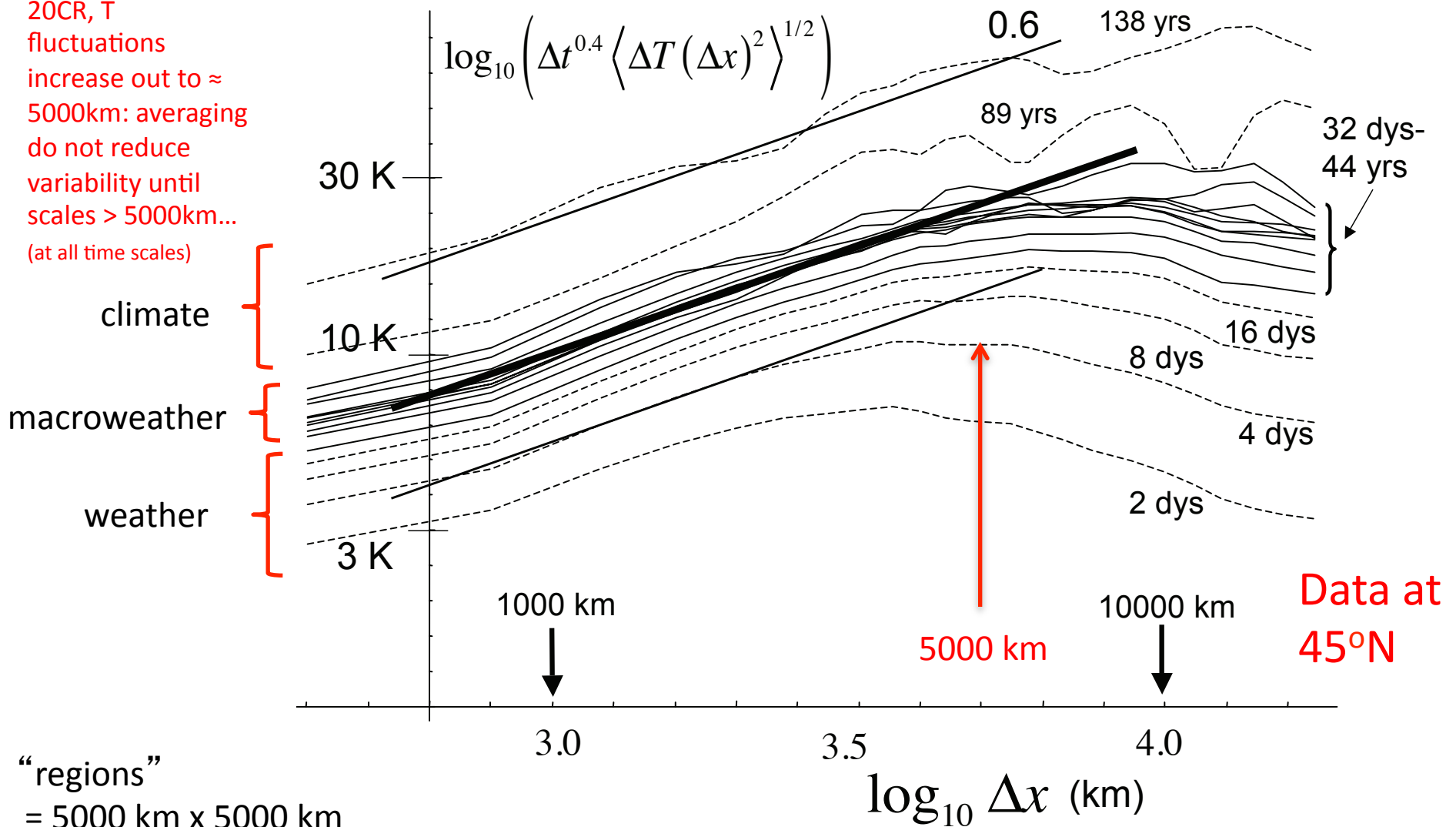
$$P_{w,c,\tau_w}(k_x, \omega) \approx E_c(k_x) E_{mw}(\omega)$$

$$\tau_w < \tau < \tau_c$$



# Space-time scaling of fluctuations

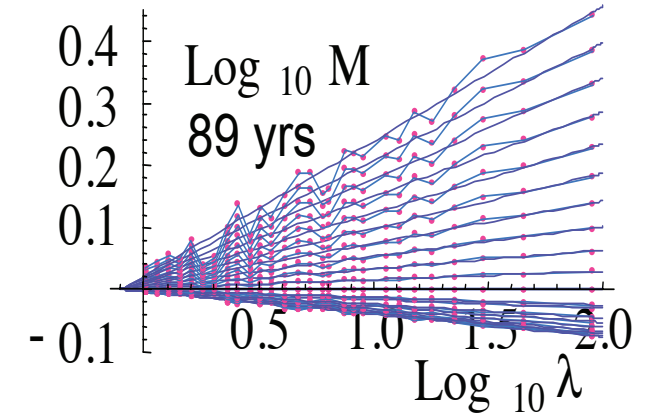
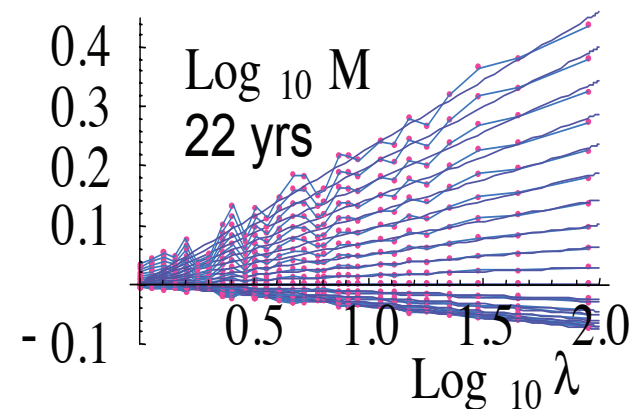
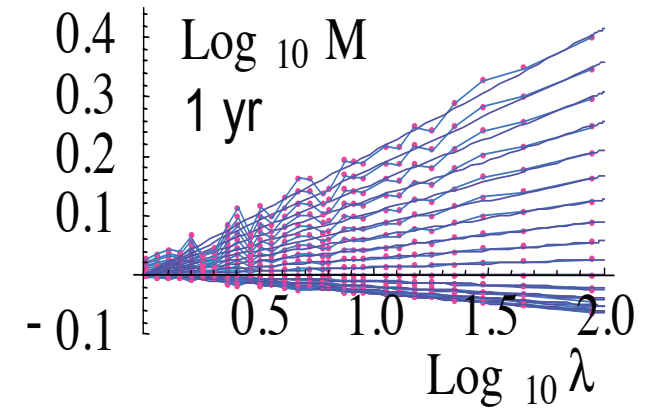
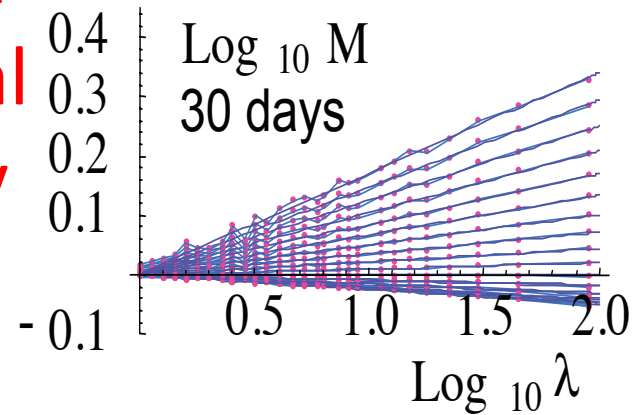
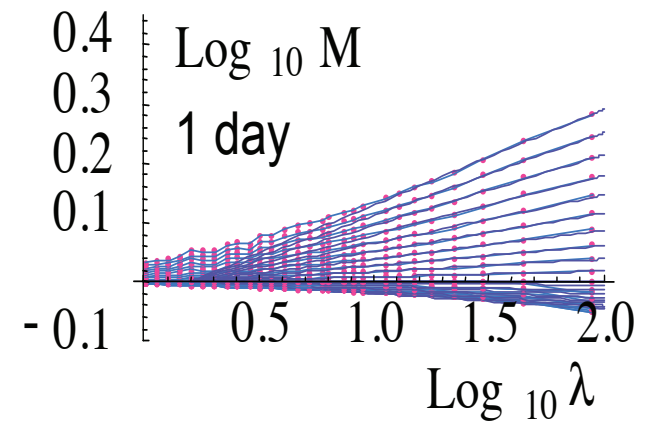
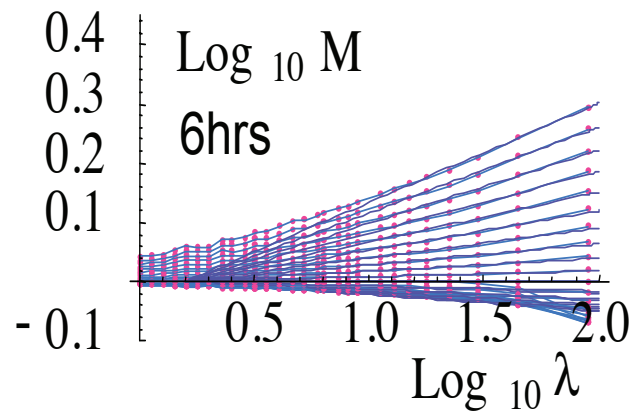
20CR, T  
 fluctuations  
 increase out to  $\approx$   
 5000km: averaging  
 do not reduce  
 variability until  
 scales  $>$  5000km...  
 (at all time scales)



“regions”  
 = 5000 km x 5000 km  
 =North America



# Macroweather- Climate spatial Intermittency



	$C_1$	$\alpha$	$L_{eff}$
6 hrs	0.095	1.65	6400 km
1 day	0.091	1.71	6400 km
30 dys	0.093	1.81	10000 km
1 yrs	0.118	1.75	10000 km
27 yrs	0.136	1.67	10000 km
89 yrs	0.130	1.63	13000 km

# Atmospheric Statistics in a nutshell

## Weather

$$S_{q,w}(\underline{\Delta r}, \Delta t) = (s_{q,w})^q \left[ \left[ (\underline{\Delta r}, \Delta t) \right]_w \right]^{qH_w - K_w(q)} ; \quad \tau_i < \Delta t < \tau_w$$

$$\left[ (\Delta x, 0, 0, \Delta t) \right]_{w,can} = \left( \left( \frac{\Delta x}{L_e^*} \right)^2 + \left( \frac{\Delta t}{\tau_w} \right)^2 \right)^{1/2}$$

## Macroweather

$$S_{q,mw}(\underline{\Delta r}, \Delta t) = (s_{q,mw})^q \left\| \underline{\Delta r} \right\|_c^{qH_c - K_c(q)} \left( \frac{\Delta t}{\tau_w} \right)^{qH_{mw}} ; \quad \tau_w < \Delta t < \tau_c$$

$$\left\| (\Delta x, 0, 0) \right\|_{c,can} = \left| \frac{\Delta x}{L_e^*} \right|$$

## Climate

$$S_{q,c}(\underline{\Delta r}, \Delta t) = \left\langle \Delta T(\Delta x, \Delta t)^q \right\rangle_c = (s_{q,c})^q \left[ \left[ (\underline{\Delta r}, \Delta t) \right]_c \right]^{qH_c - K_c(q)}$$

$$\left[ (\Delta x, 0, 0, \Delta t) \right]_{c,can} = \left( \left( \frac{\Delta x}{L_e^*} \right)^2 + \left( \frac{\Delta t}{\tau_{lc}} \right)^{2/H_{c,t}} \right)^{1/2} ; \quad H_{c,t} = H_c / H_{c,\tau} ; \quad \tau_{lc} > \Delta t > \tau_c$$

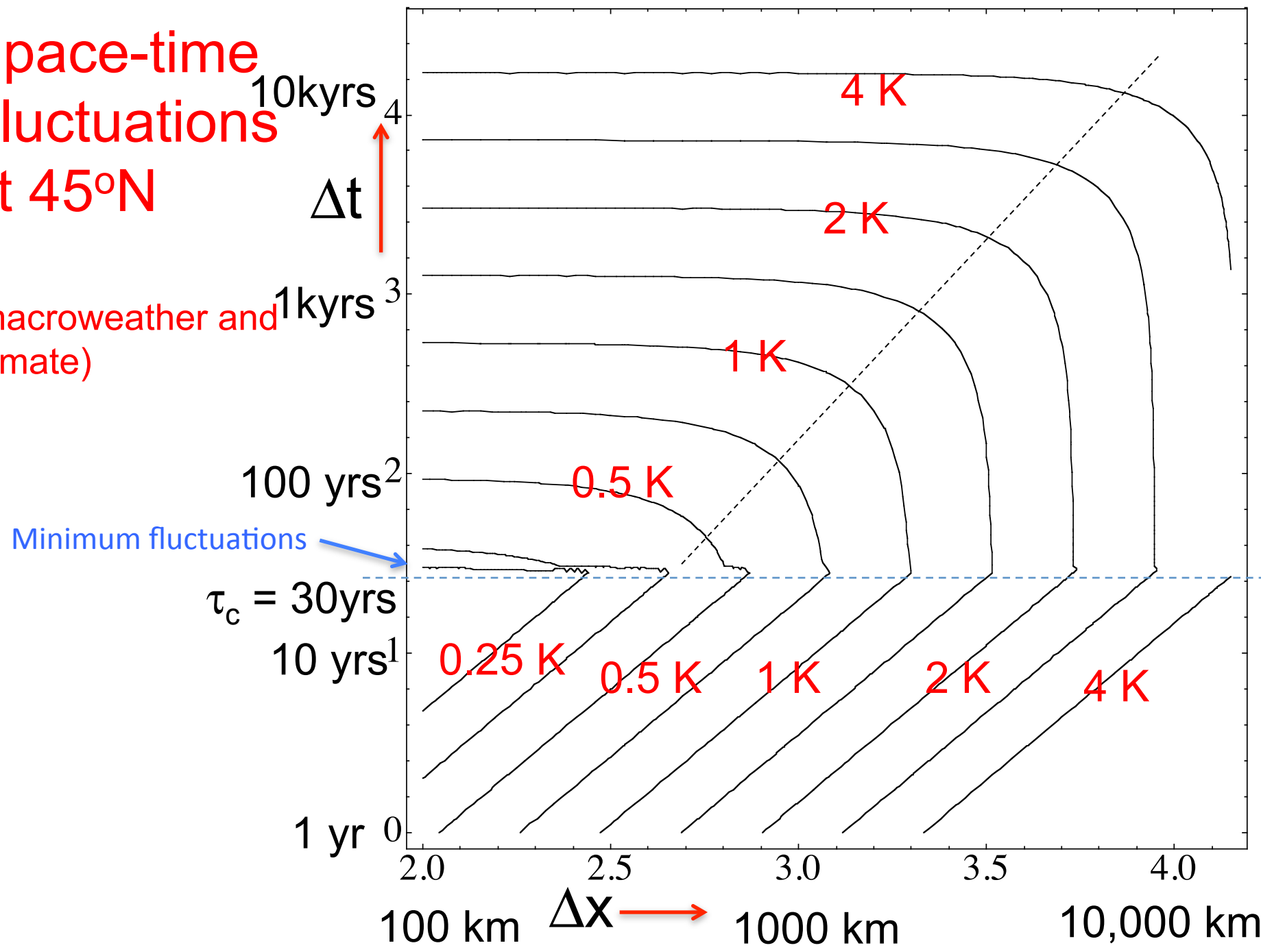
## Parameters

$$K(q) = \frac{C_1}{\alpha - 1} (q^\alpha - q)$$

Regime	H		C <sub>1</sub>		α		Outer time scale
	Space	Time	Space	Time	Space	Time	
Weather	0.51		0.087		1.61		5-20 dys
Macro weather		-0.4		0		-	20-40 yrs
Climate	0.7	0.4	0.11	0.065	1.4	1.5	30-50 kyrs

# Space-time Fluctuations at 45°N

(macroweather and  
climate)



# Conclusions

1. High level stochastic turbulence laws emerge from (deterministic) continuum mechanics at strong nonlinearity
2. Regimes: Weather, macroweather, climate
3. Generalize classical laws: Intermittency using cascades
4. Generalize classical laws: wide range of scales using anisotropic scaling, stratification
5. Modelling with Fractionally Integrated Flux model
6. Consequences for space-time scaling: turbulent propagators and turbulence driven waves
7. Predictability and stochastic forecasting
8. At scales  $>\approx 30$  years new scaling processes (anthropogenic at  $\approx 10$  yrs, natural at  $\approx 100$  yrs) with  $H>0$  dominate up to  $\approx 100$  kyrs. Can be modelled in FIF framework.