

Fractional relaxation noises, motions and the fractional energy balance equation

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Abstract

We consider the statistical properties of solutions of the stochastic fractional relaxation equation that has been proposed as a model for the earth's energy balance. In this equation, the (scaling) fractional derivative term models energy storage processes that occur over a wide range of space and time scales. Up until now, stochastic fractional relaxation processes have only been considered with Riemann-Liouville fractional derivatives in the context of random walk processes where it yields highly nonstationary behaviour. For our purposes we require the stationary processes that are the solutions of the Weyl fractional relaxation equations whose domain is $[-\tau]$ to t rather than 0 to t . We develop a framework for handling fractional equations driven by white noise forcings. To avoid divergences, we follow the approach used in fractional Brownian motion (fBm). The resulting fractional relaxation motions (fRm) and fractional relaxation noises (fRn) generalize the more familiar fBm and fGn (fractional Gaussian noise). We analytically determine both the small and large scale limits and show extensive analytic and numerical results on the autocorrelation functions, Haar fluctuations and spectra. We display sample realizations. Finally, we discuss the prediction of fRn, fRm which – due to long memories – is a past value problem, not an initial value problem. We develop an analytic formula for the fRn forecast skill and compare it to fGn. Although the large scale limit is an (unpredictable) white noise that is attained in a slow power law manner, when the temporal resolution of the series is small compared to the relaxation time, fRn can mimic a long memory process with a wide range of exponents ranging from fGn to fBm and beyond. We discuss the implications for monthly, seasonal, annual forecasts of the earth's temperature.

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The Fractional and Half-Order Energy Balance Equations

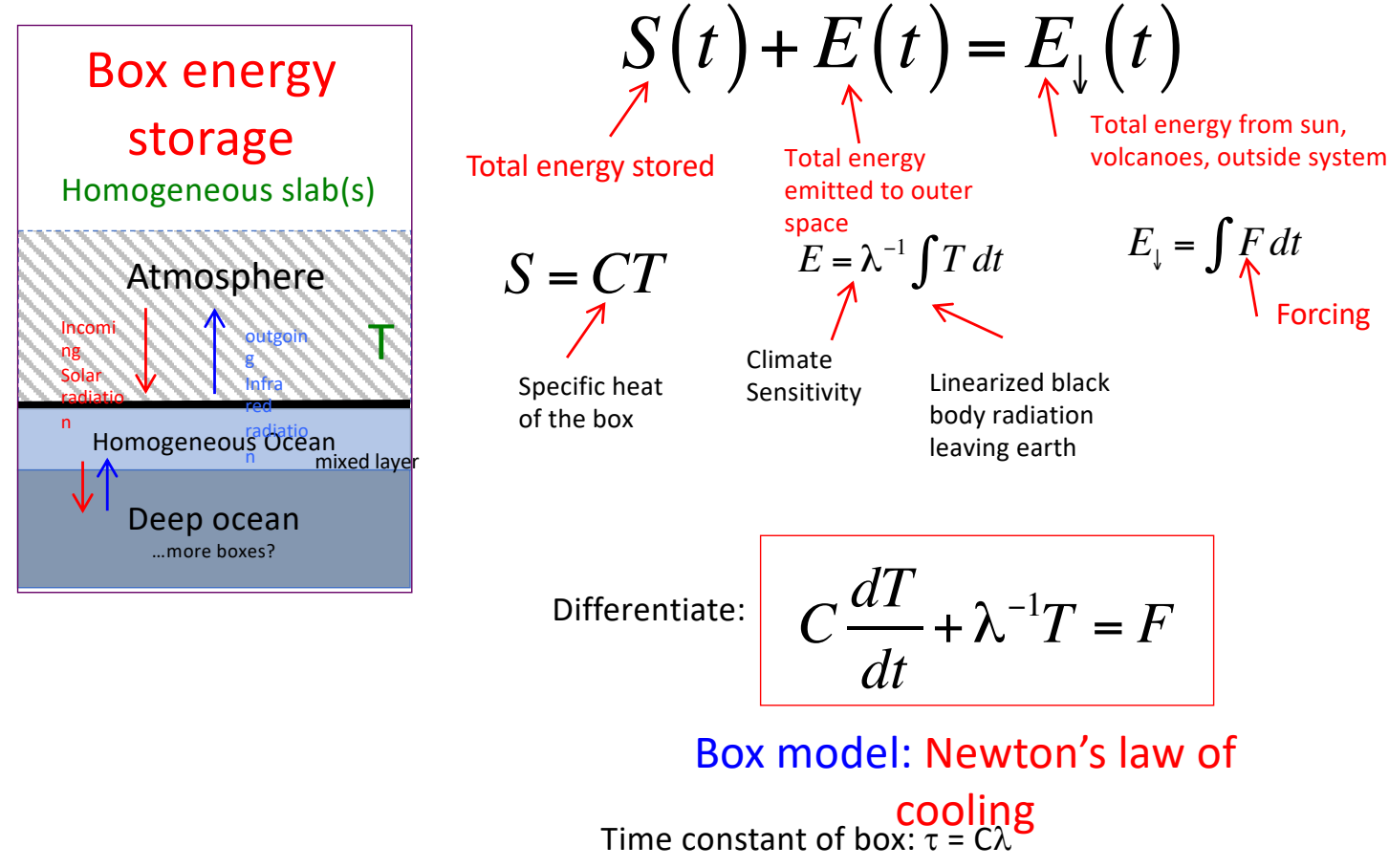
Fractional relaxation noises, motions

The Fractional order Energy Balance Equation (FEBE)

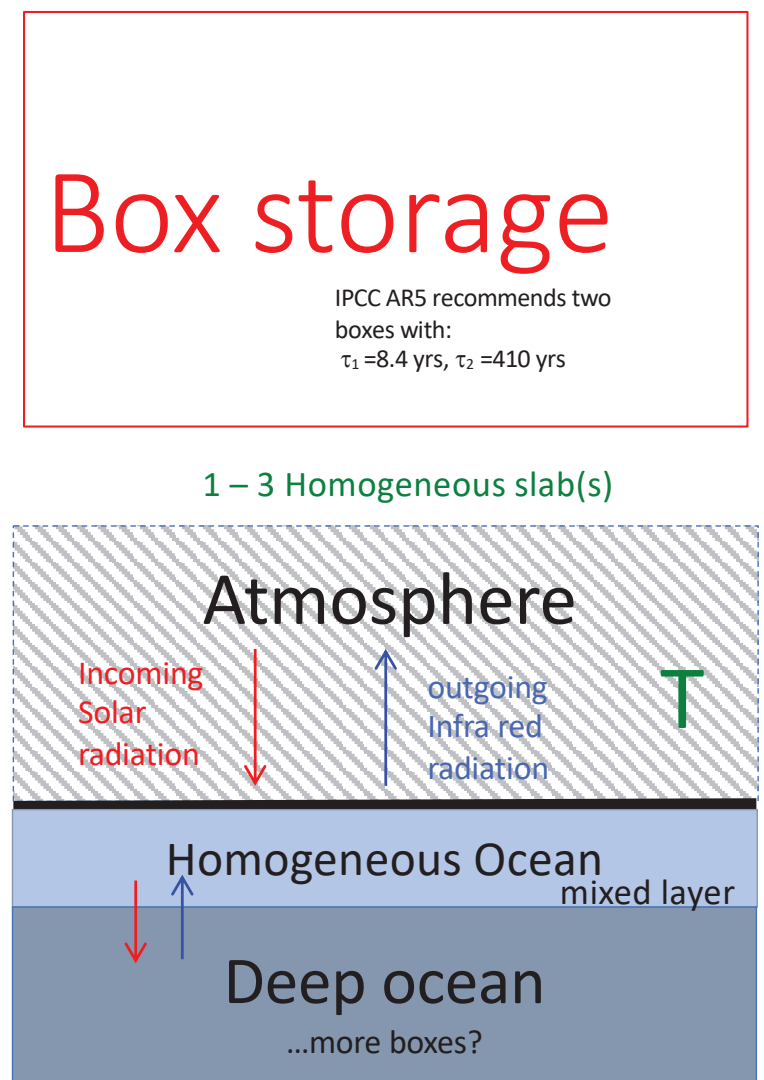
The Half-order Energy Balance Equation (HEBE)

Fractional relaxation noises, motions and the FEBE

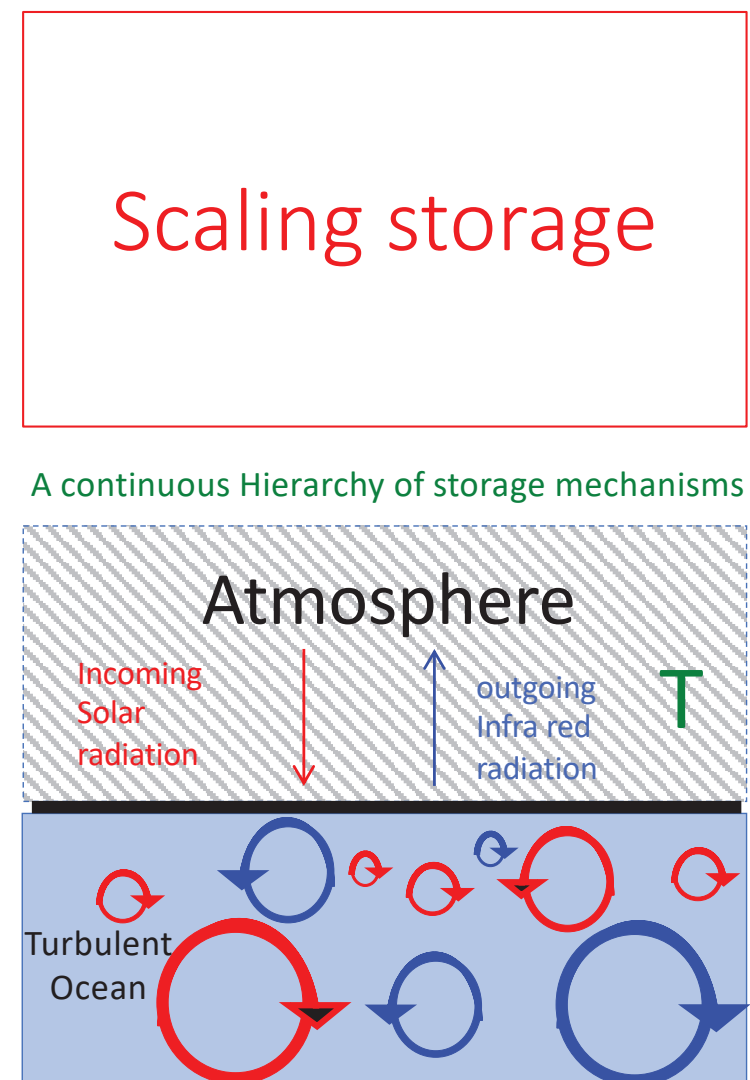
The (conventional) Energy Balance Equation (EBE)



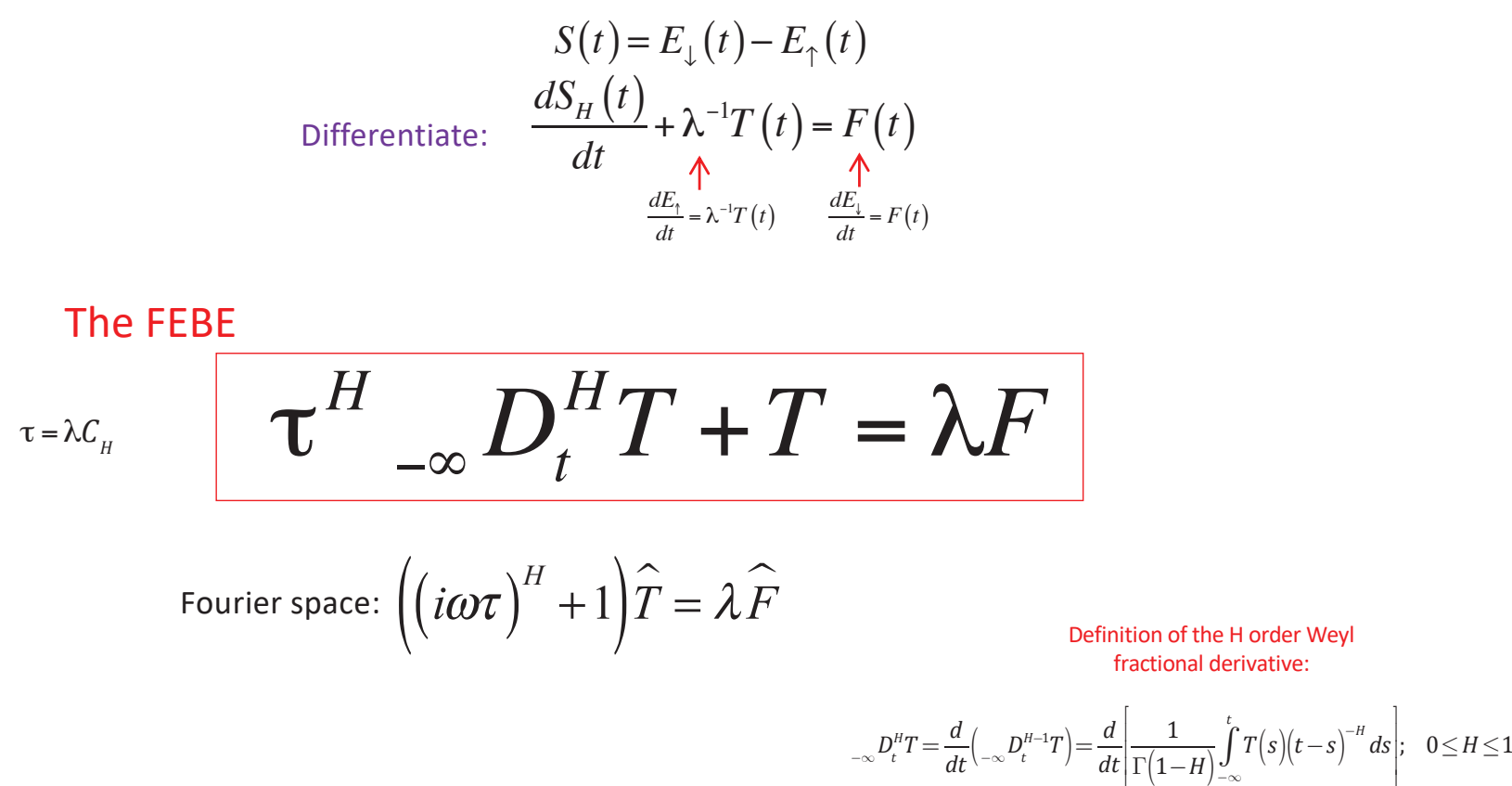
Box storage



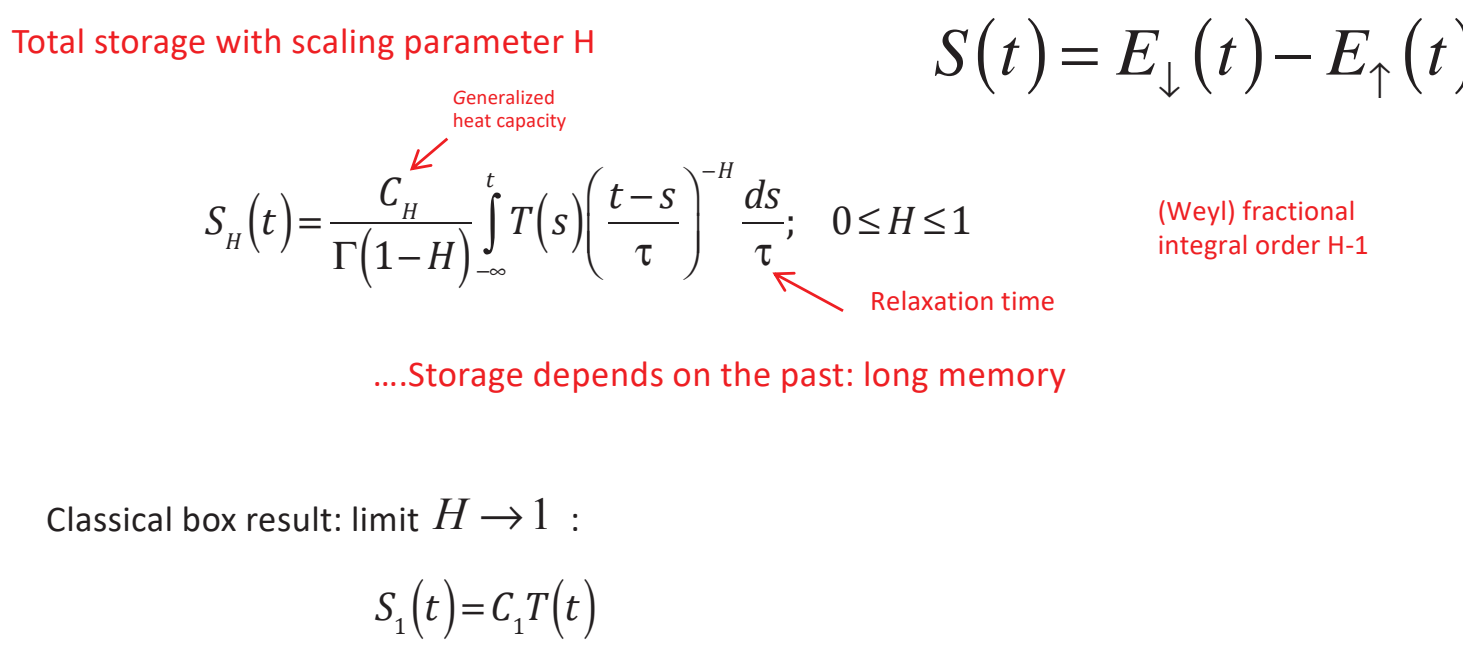
Scaling storage



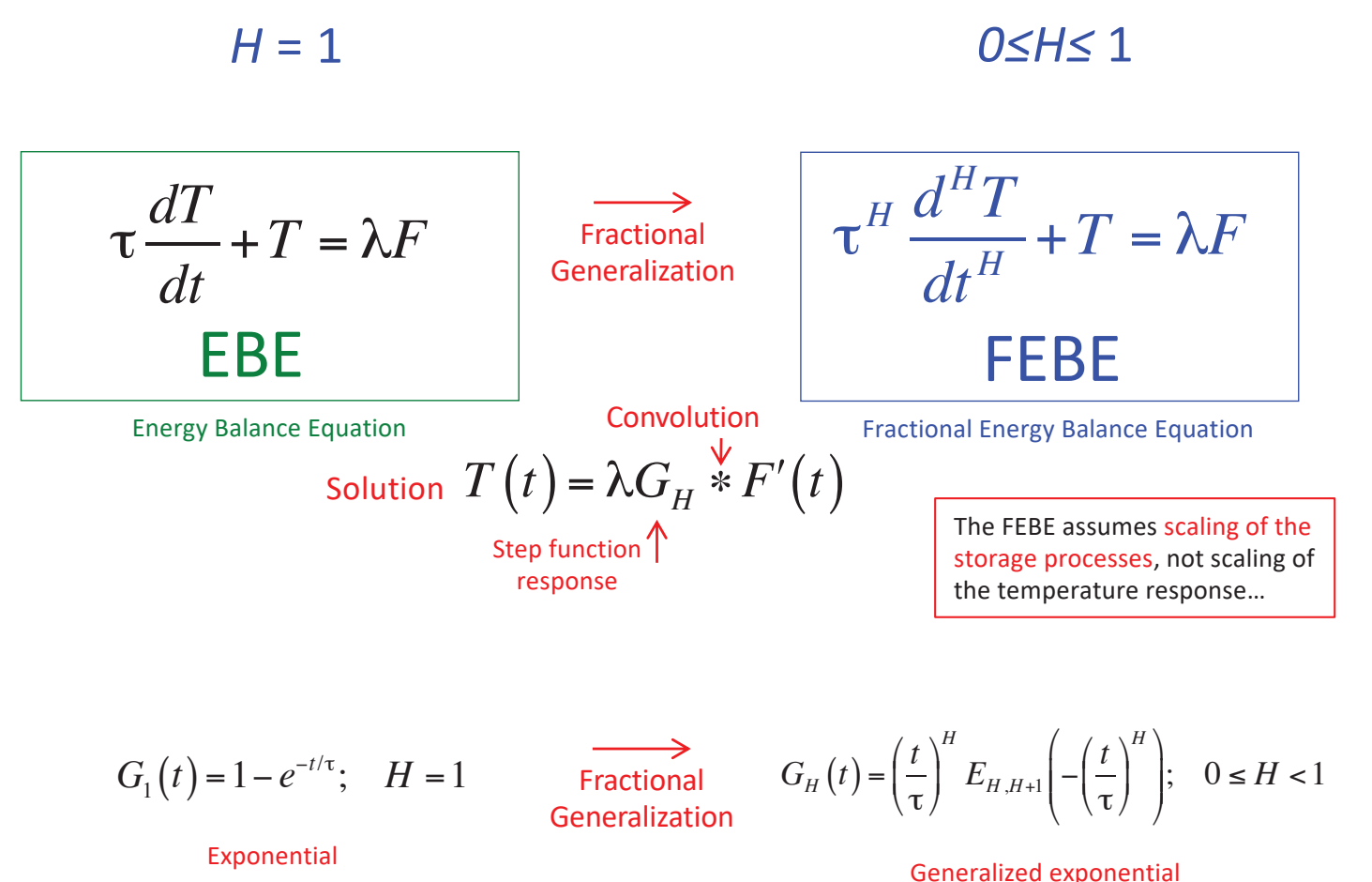
Energy rate equation



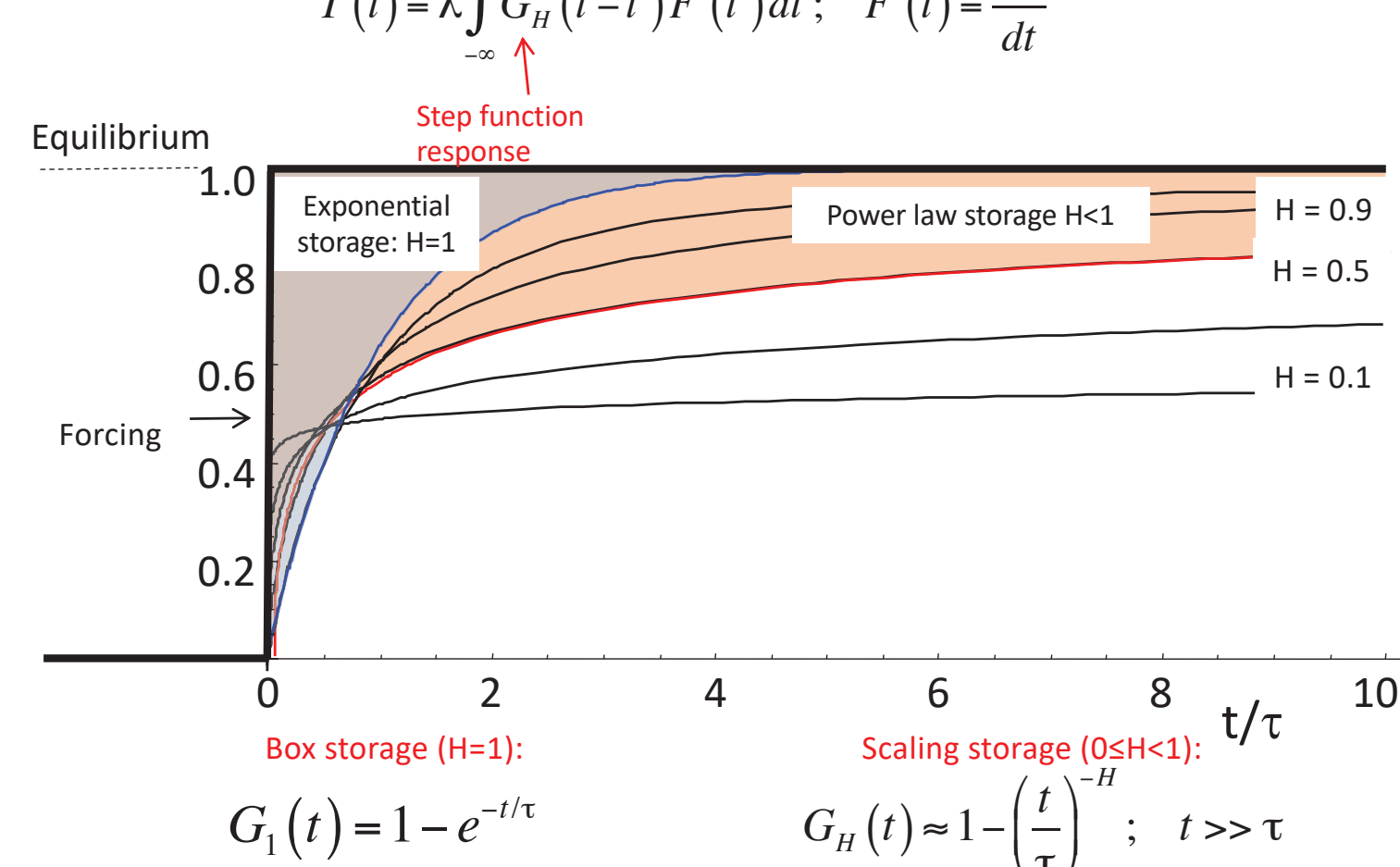
Scaling of the storage



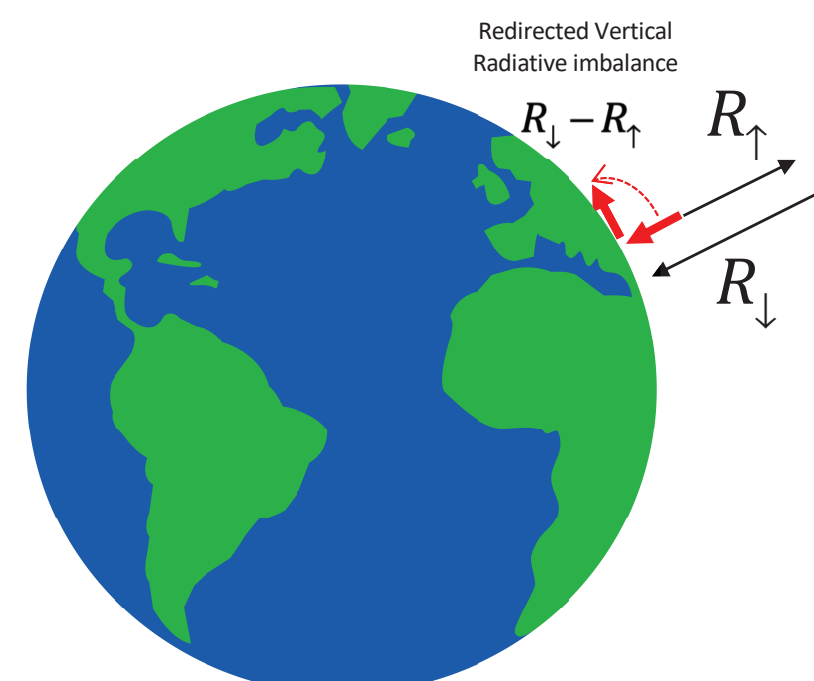
Fractional Generalization: realistic storage



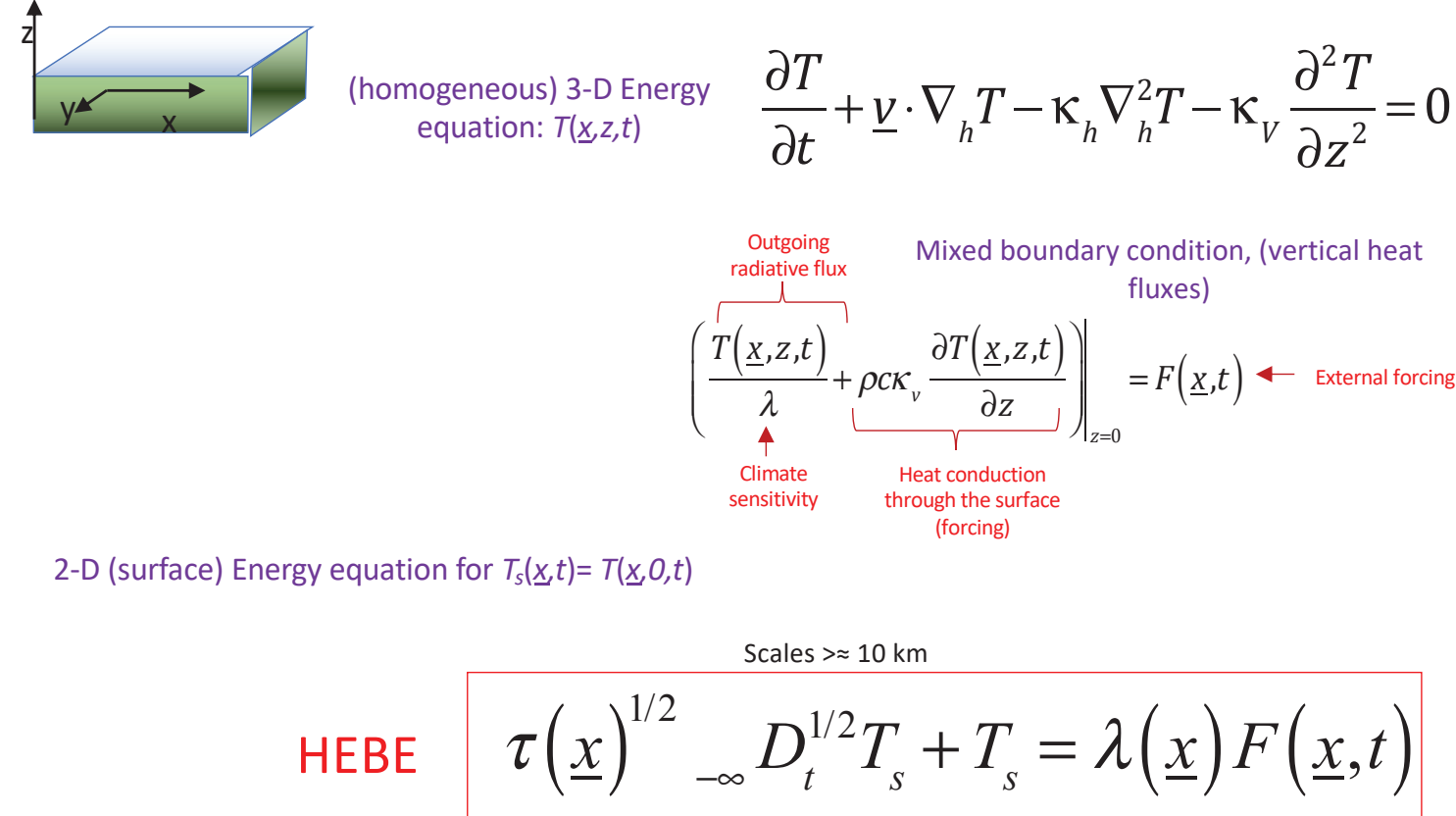
Response: Generalized relaxation processes



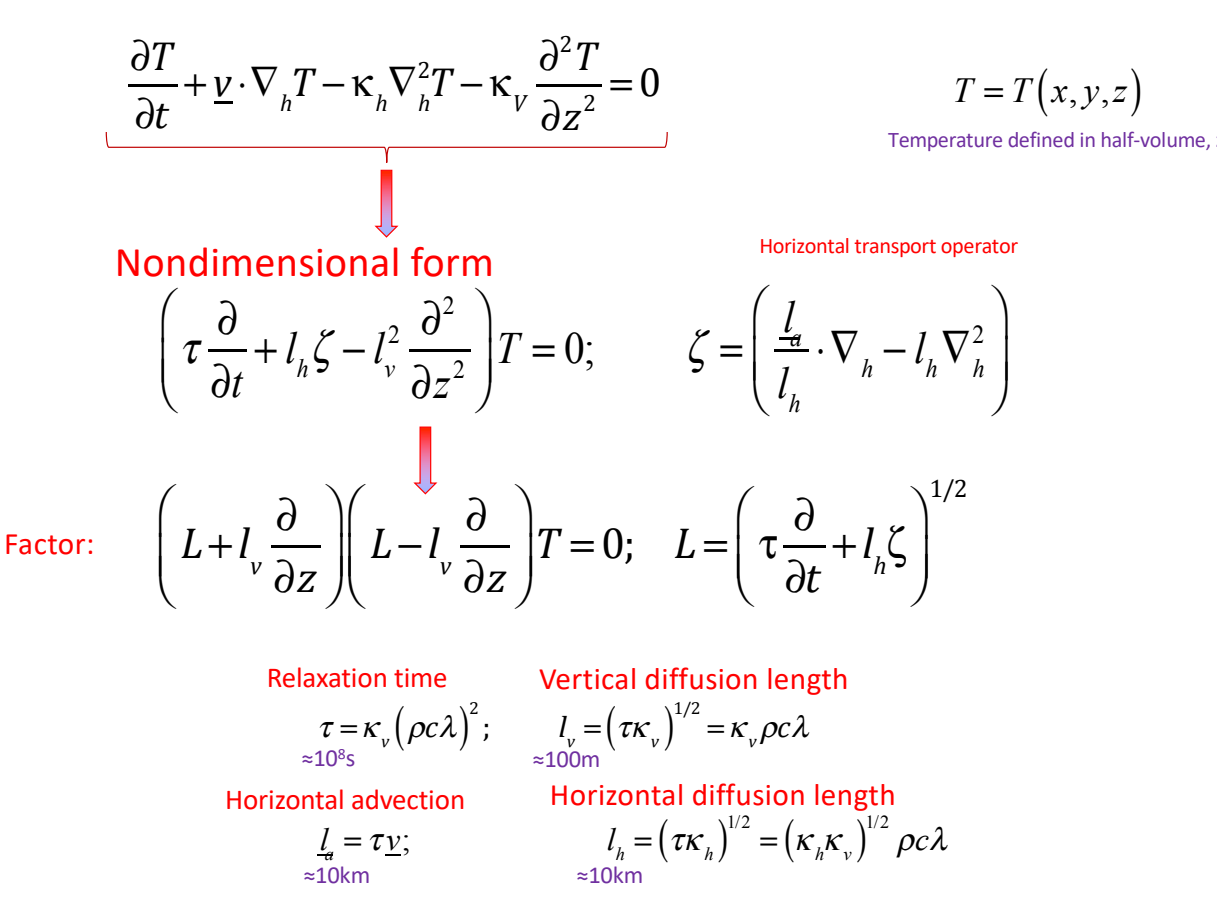
Regional energy balance



Correcting Budyko-Seller's boundary conditions:



Babenko's method



Conclusions

We have proposed a new 2D energy balance equation for macroclimate scales, ten days and longer. By introducing a vertical coordinate, we were able to rigorously treat the forcing of the entire system as the basic surface boundary condition: local vertical radiative flux imbalances that force heat into the earth. A consequence is that the apparently difficult "mixed" surface boundary conditions are avoided and the latitudinal and longitudinal (2D) variation of the time-independent climatological temperatures can be determined. More importantly, we were able to obtain an equation for the time-varying anomalies. In comparison, the usual (e.g. Budyko-Seller) type energy balance models have no vertical coordinate and instead redirect the local imbalances in an ad hoc fashion, away from the equator, meridionally.

Since the forcing is via the vertical boundary condition, the equation remains homogeneous and Babenko's method can be applied. Babenko's method elegantly transforms the mixed boundary conditions directly into a simple equation for the surface temperature: the Generalized Half-Order Energy Balance Equation (GHEBE). The detailed vertical structure turns out to only be important over a thin surface layer. A key novelty of the GHEBE is that instead of using classical first-order time derivatives (the Energy Balance Equation, EBE) it is of half order in both temporal and horizontal operators. However, for spatial scales larger than ten kilometers or so, and time scales less than centennial, the temporal (inherited) processes dominate over the advective and/or diffusive horizontal transport processes, this yields the Half-Order EBE or HEBE. Under fairly mild assumptions we obtained a full analytic solution to the HEBE surface temperature anomalies $T_s(t)$.

The EBE and HEBE are the $H=1$ and $H=1/2$ special cases of the Fractional EBE (FEBE) that we recently introduced as a phenomenological macroclimate model [Lovejoy, 2019a], with empirical estimates $H \approx 0.4 - 0.5$, i.e. very close to the HEBE. A feature of the HEBE is that its Green's function is a power law rather than an exponential and this implies a long memory: indeed the Gaussian white noise driven HEBE has a high frequency 1/f spectrum that is cut-off at the relaxation time (temporally of the order 4 years). By extending energy balance models to 2D, it allows us to treat regional temporal anomalies, and this at significantly shorter time scales than were previously possible, perhaps down to the ~ 10 day weather-macroclimate transition scale. Depending on the space-time statistics of the anomaly forcing, the HEBE justifies the current Fractional EBE (FEBE) based macroclimate (monthly, seasonal) temperature forecasts [Lovejoy et al., 2015]. [See also Amador and Lovejoy, 2019].

At this transition scale, GCMs are beyond the predictability limits of their atmospheric components, they become stochastic. Analyses of 32 CMIP5 GCMs showed that although each GCM had a distinct climate, that each responded nearly linearly to the climate forcing scenarios considered in the IPCC AR5 (Inghel and Lovejoy, 2018). This implies that macroclimate dynamics are plausibly linear, consistent with the HEBE. The regional HEBE – when stochastically forced – is thus a promising macroclimate temperature model. Indeed, the high frequency part of its FEBE generalization can already be used for monthly, seasonal forecasts and the overall FEBE with $H \approx 1/2$ can produce climate projections with significantly lower uncertainties than current GCM based alternatives (work in progress with R. Procyk).

In addition to the anomalies, our approach opens the door to the determination of the full 2-D climate state – generalizations of the 1-D Budyko-Seller type climate – to determine past and future climates. This could done first by applying the method to the existing climate by fixing the forcing at current values and solving the time independent transport equations. Then, the long term effects of changes – such as step function increases in forcing – could be determined from the GHEBE anomaly equation (section 3.5) which regionally corrects the local climate anomalies for (slow) horizontal energy transport effects. Miklandov (spatial) forcing are linear and can easily be introduced, and generalizations to account for albedo feedbacks and other nonlinear effects could easily be made in order to study glacial cycles. The power law relaxation processes implied by the GHEBE suggests straightforward explanations for the observed power law climate regime spanning the range from centennial to Millennium scales.

fRn, fRm

The stochastic FEBE, fractional relaxation motion (fRm), fractional relaxation noise (fRn)

In the previous section we considered the deterministic response to a simple piecewise linear model of external forcing. We noted that since the FEBE is a linear equation that we could separately model this deterministic response and the stochastic, internal variability. In this section therefore we turn to the pure stochastic case driven by white noise "innovations", we briefly summarize some of the results in [Lovejoy, 2019a].

The physical problem that we wish to solve is the noise driven FEBE with noise amplitude σ and sensitivity λ .

$\tau^H \frac{d^H T}{dt^H} + T = \lambda(T) + \sigma(t)$ (1)

with initial conditions $T(-\infty) = 0$, $d^H T/dt^H = 0$, and $\sigma(t)$ a unit Gaussian white noise so that $\langle \sigma(t) \sigma(t') \rangle = \sigma^2 \delta(t-t')$, $\langle \sigma(t) \rangle = 0$, $\langle \sigma(t) \rangle = 0$, $\langle \sigma(t) \rangle = 0$.

The integral of the response, it suffices to study the nondimensional equation: $\frac{d^H T}{dt^H} + T = \lambda(T) + \sigma(t)$ (2)

where λ is the normalization constant in eq. 35 and the nondimensional $\lambda(T)$ function is called the fractional relaxation noise (fRn) since it generalizes fGn. Using U we can obtain the solution to the dimensional eq. 46:

$T(t) = \frac{\lambda(T)}{\Gamma(1-H)} \int_0^t \tau^{H-1} U(\tau) d\tau$ (3)

Since the fRn process is the solution of the fractional relaxation equation with a stationary, Gaussian, zero mean, white noise forcing, it is also stationary, Gaussian with zero mean. Its statistics are therefore fully characterized by its autocorrelation function.

We can now calculate the correlation function relevant for the fRn statistics. The main complication is that in the small limit, the fractional term dominates so that we obtain the fGn limit. The solution of eq. 47 is therefore – like $\gamma(t)$ – a generalized function: to obtain solutions with finite variances, we must take averages over finite resolutions τ_n .

The resulting τ_n resolution autocorrelation function at lag Δt is:

$R_{fRn}(\Delta t) = \left(\frac{\tau_n}{\tau_n + \Delta t} \right)^{1-H} \frac{\Gamma(1-H)}{\Gamma(1-H)}$ (4)

where τ_n is the resolution of the fRn process. In the limit $\tau_n \rightarrow 0$, we can obtain the high frequency fRn approximation valid for $\Delta t \ll \tau_n$ corresponding to fGn in the dimensional equation [Lovejoy, 2019a], where we used the normalization K given above. From this, we can obtain the high frequency fRn approximation valid for $\Delta t \ll \tau_n$ corresponding to fGn in the dimensional equation [Lovejoy, 2019a].

References

Lovejoy, S., Fractional Relaxation noises, motions and the stochastic fractional relaxation equation *Nonlinear Proc. in Geophys. Disc.*, <https://doi.org/10.1016/j.nonrpro.2019.07.001>, 2019a.

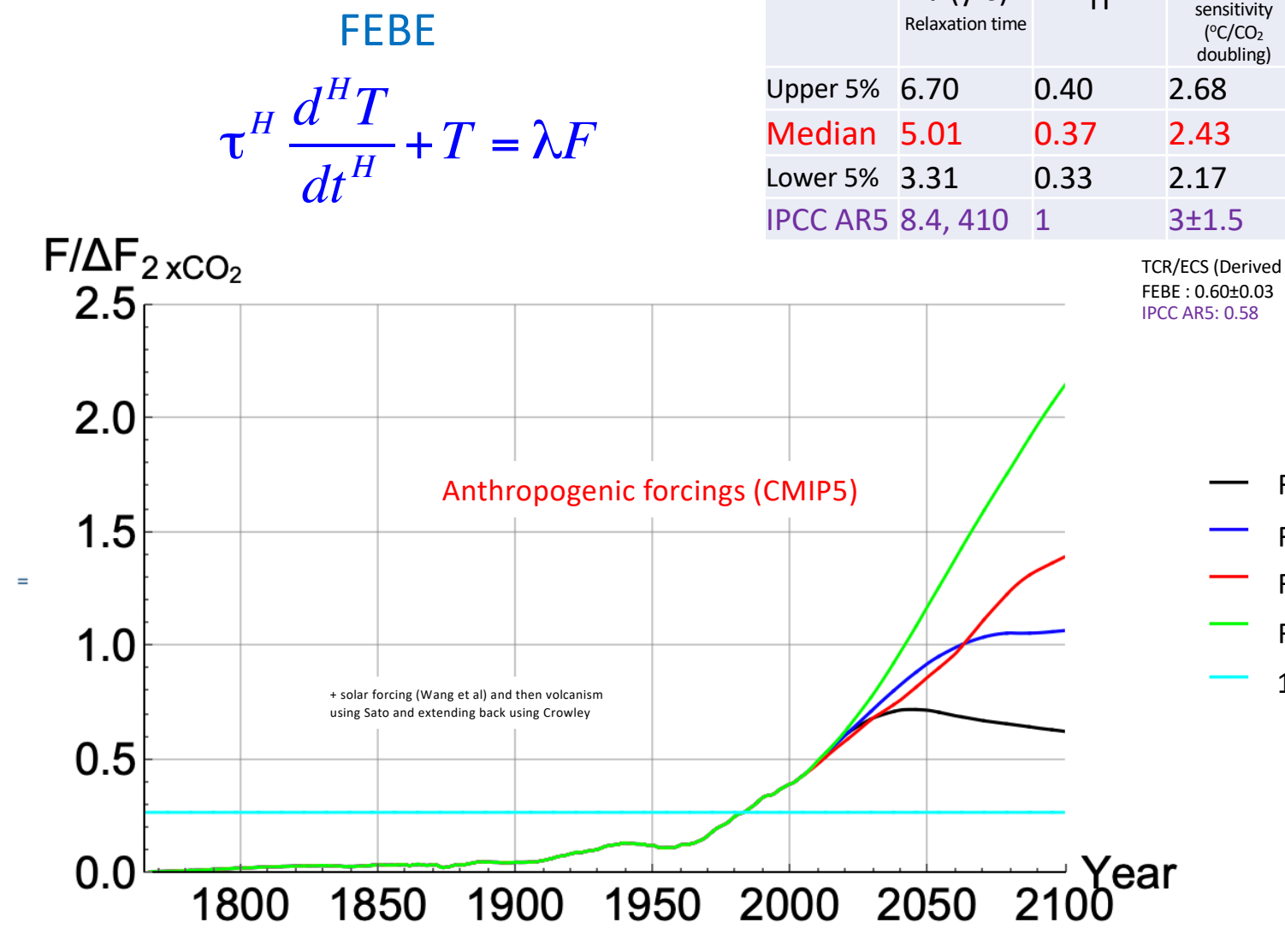
Lovejoy, S., *Weather, Macroclimate and Climate: our random yet predictable atmosphere*, 334 pp., Oxford U. Press, 2019b.

Lovejoy, S., The half-order energy balance equation, *J. Geophys. Res. (Atmos.)*, (submitted, Nov. 2019), 2019c.

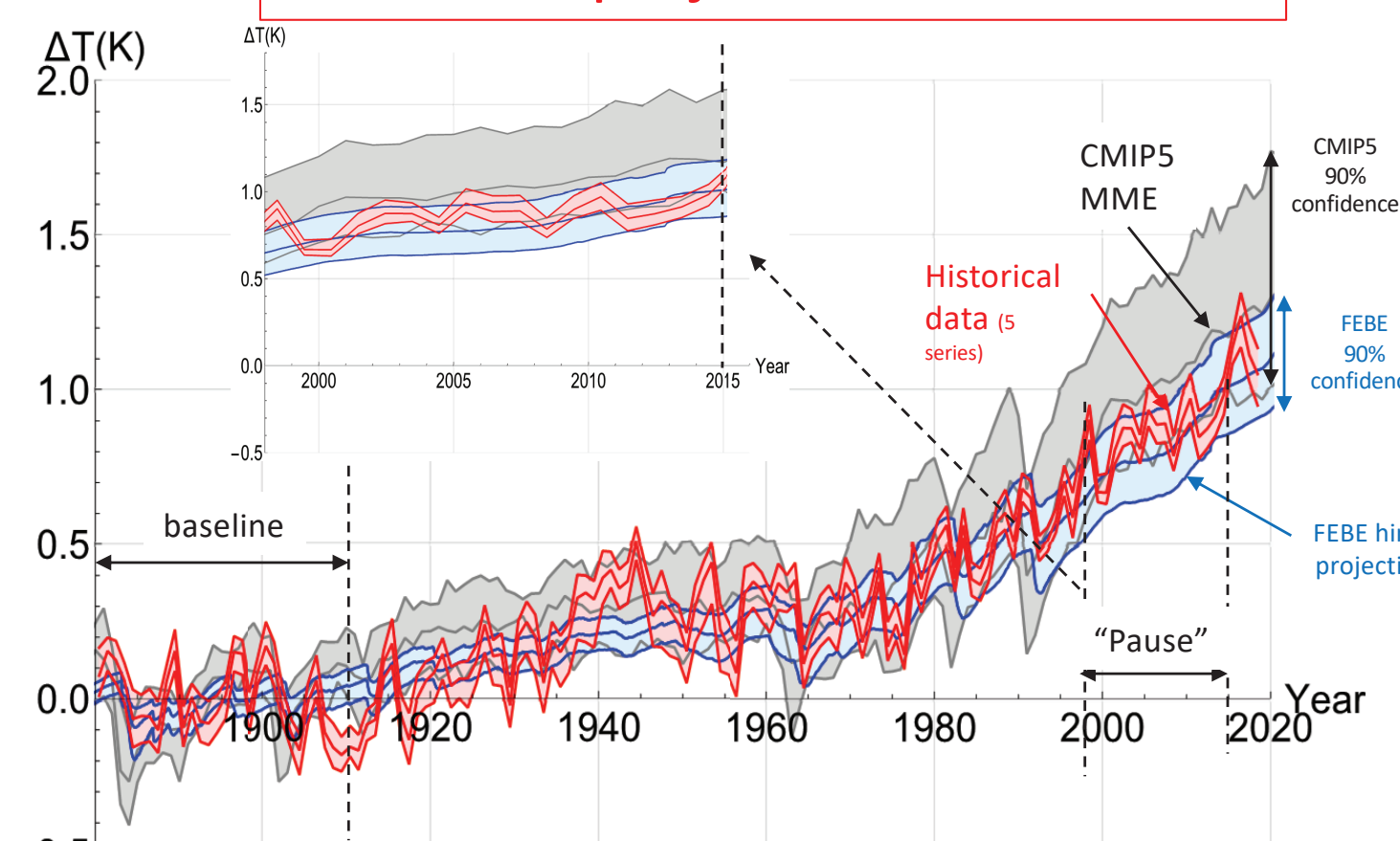
Lovejoy, S., Procyk, R., del Rio Amador, L., and Hübner, R., The fractional Energy Balance Equation, *Earth Syst. Dyn. Disc.*, (in preparation), 2019.

Projecting the temperature to 2100 using FEBE

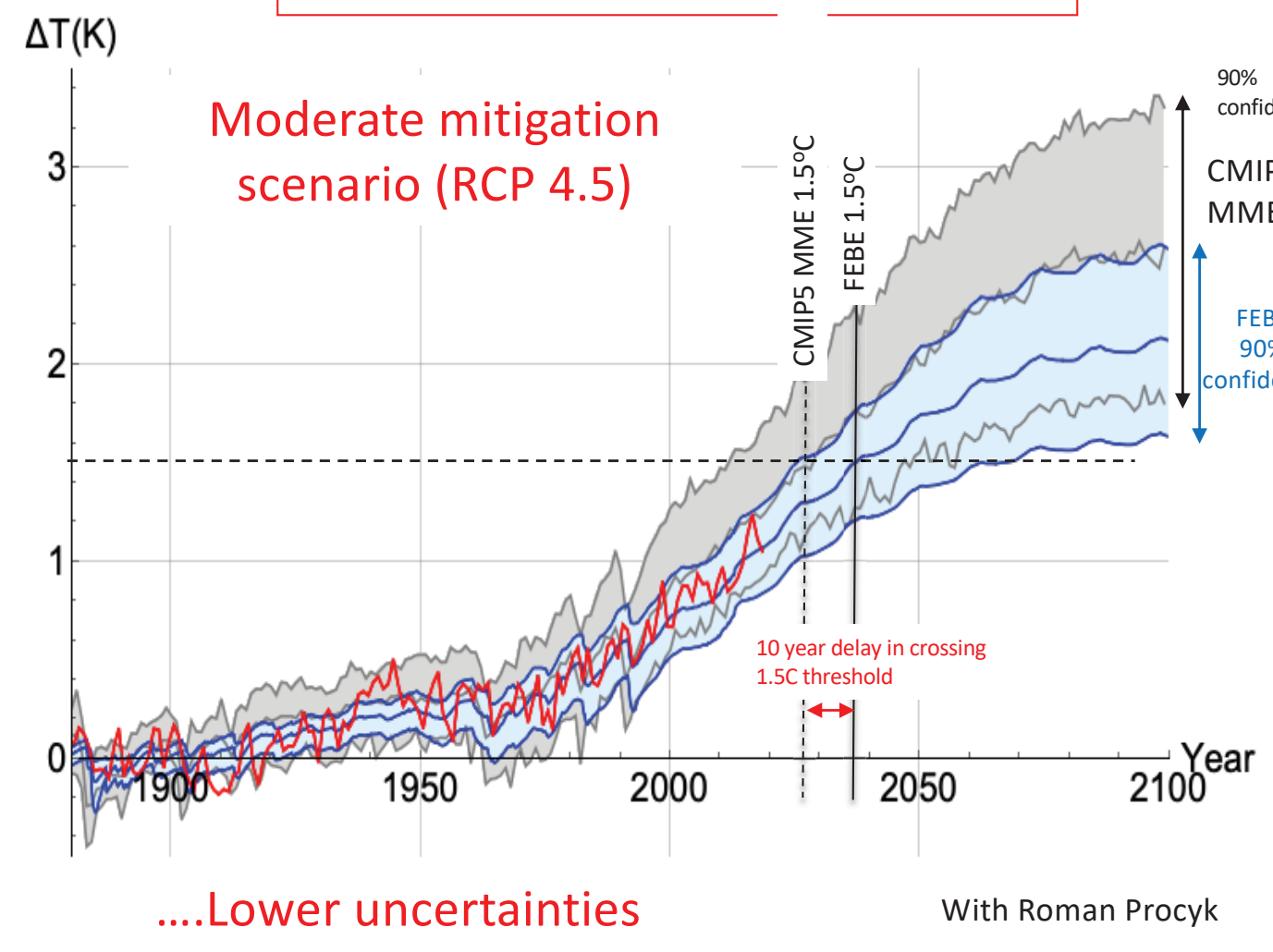
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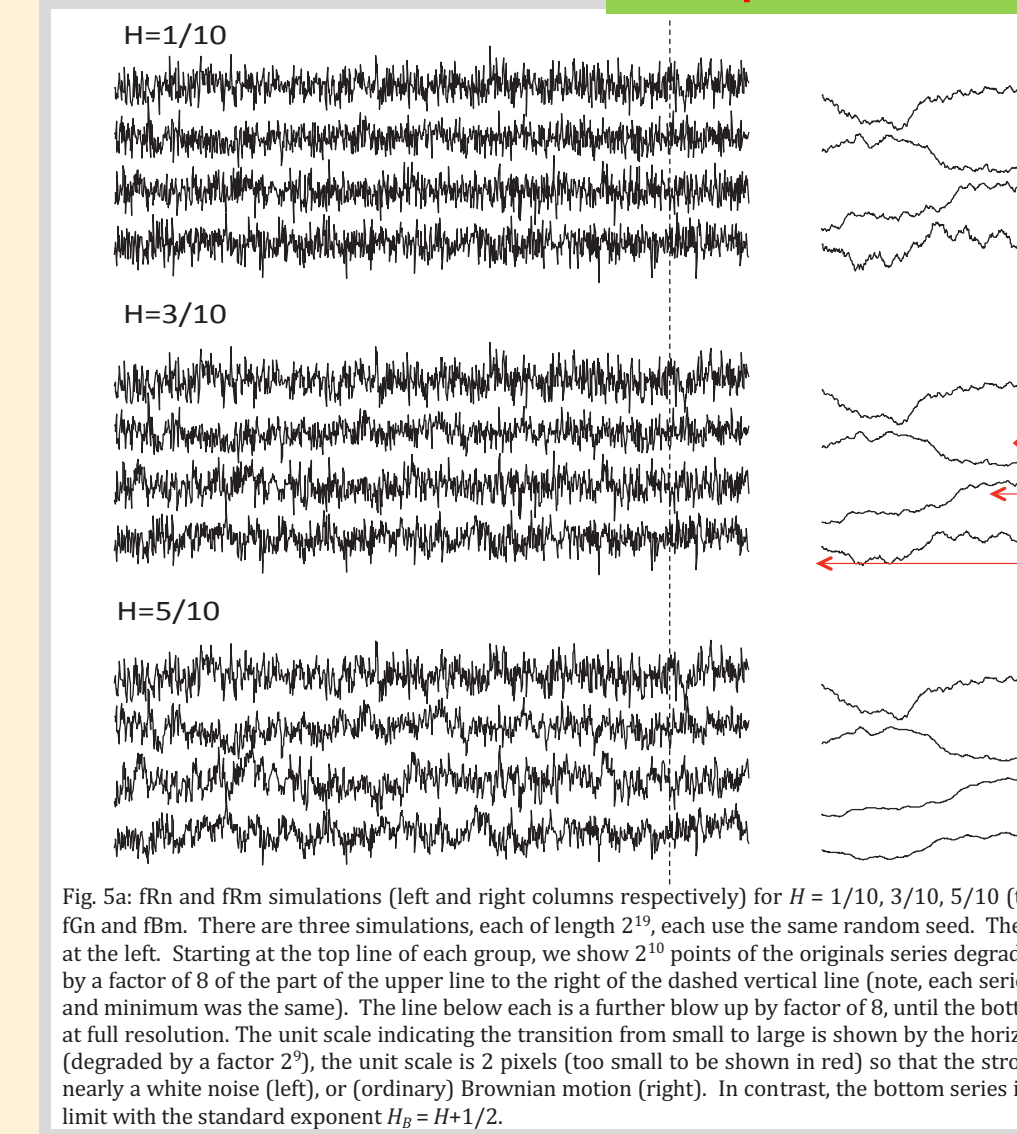
FEBE Hind projections: from 1880



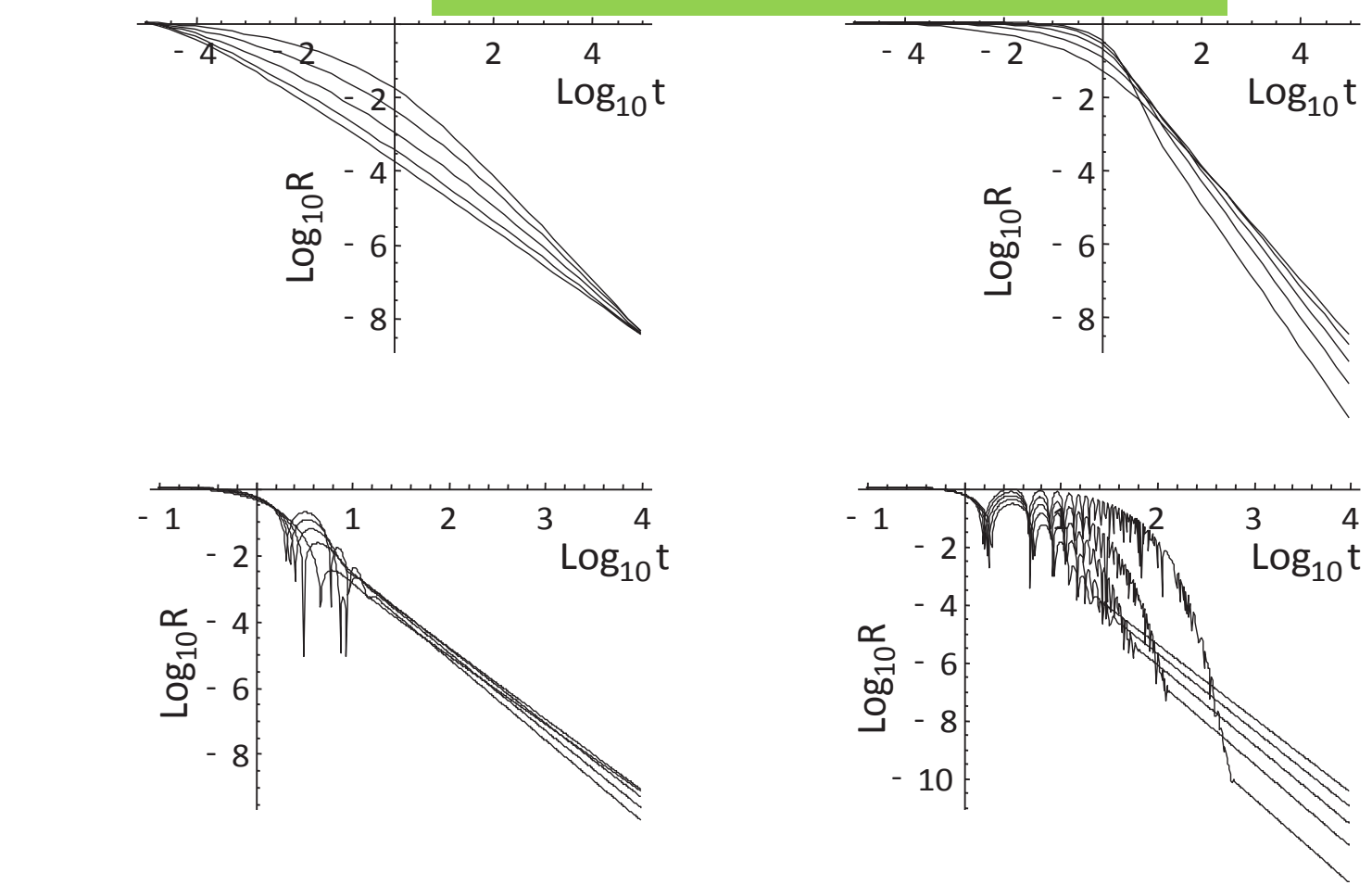
FEBE versus CMIP5 models



fRn processes



fRn autocorrelation functions



fRn predictability skill

