

Self-organisation out of disorder – L-H transition

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Thanks: James Heseltine, Schuyler Nicholson,
Hanli Liu, Rainer Hollerbach

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Outline

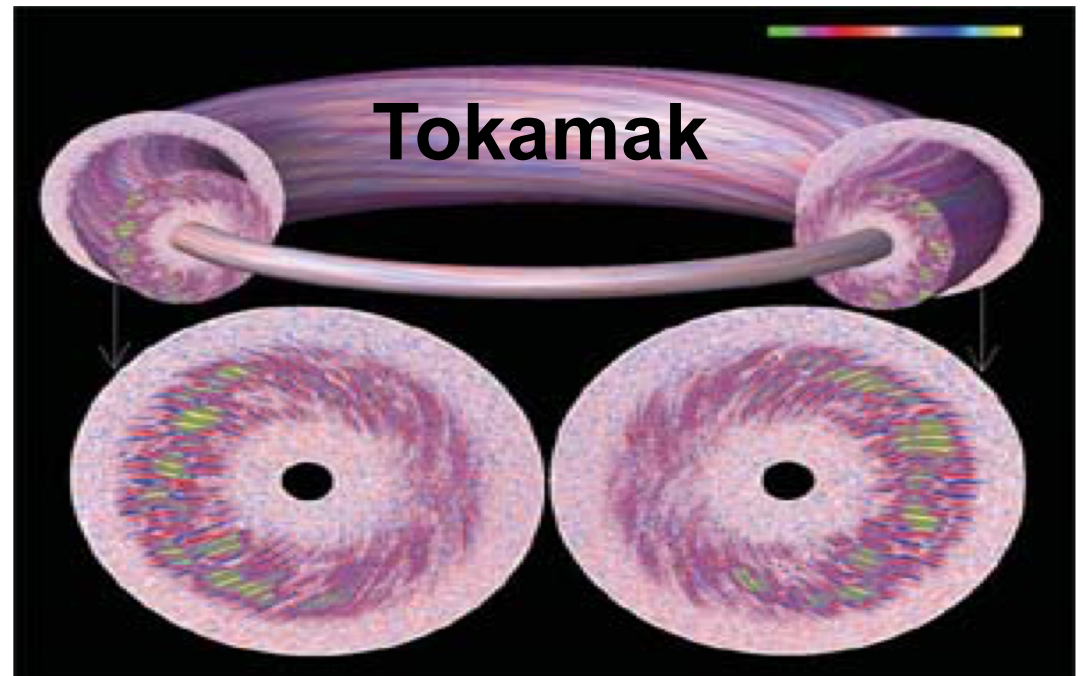
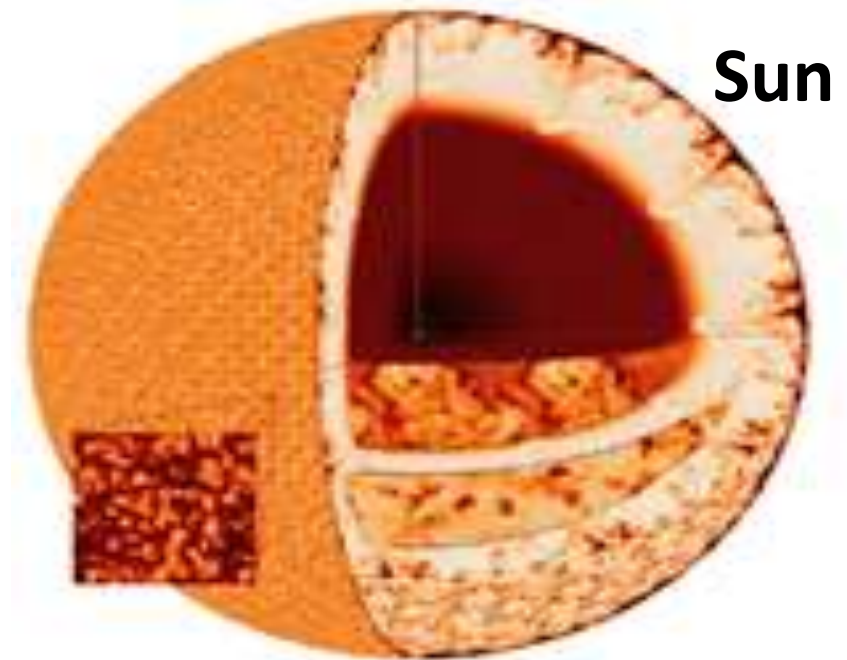
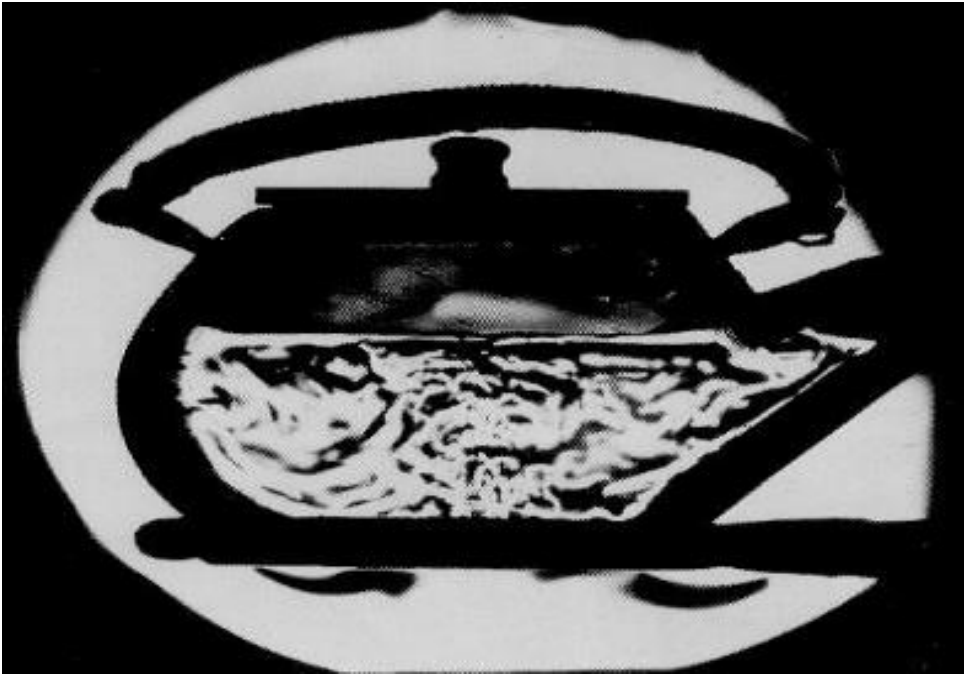
1. Overview
 - 1.1 Order/Structure out of Disorder/Complexity
 - 1.2 Self-organisation
2. The Low-to-High confinement (L-H) transition
 - 2.1 Deterministic model
 - 2.2 Stochastic model (information length)
3. Information length applied to
 - 3.1 Music
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4. Conclusion

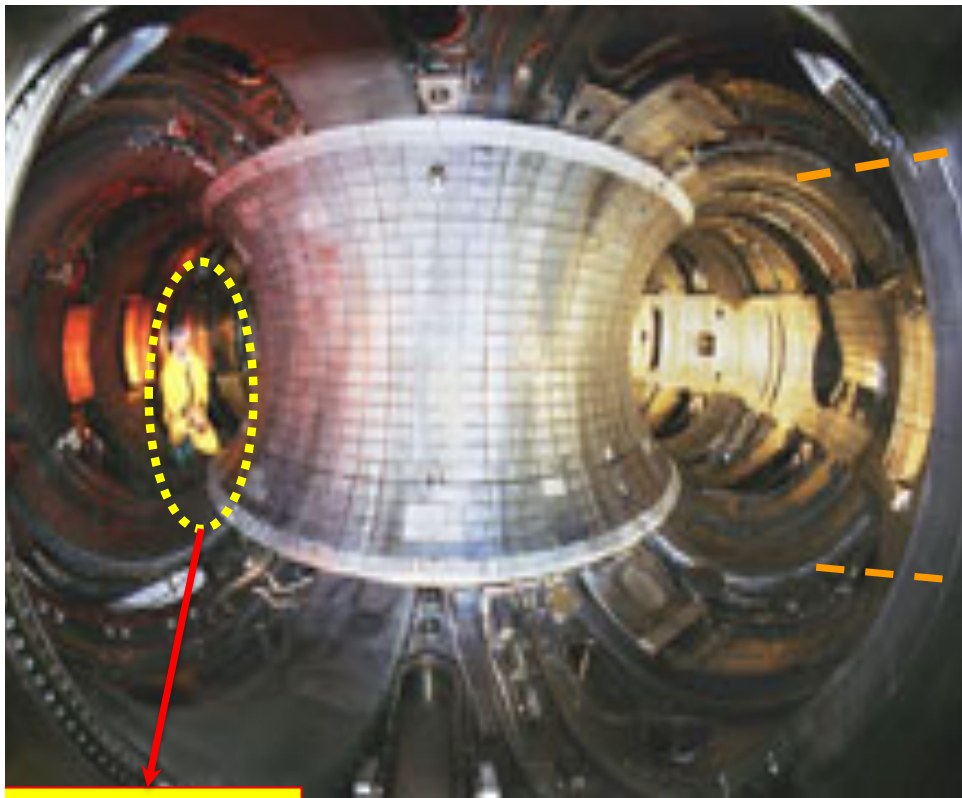
1. Overview

1.1 Complexity/disorder and structure/order

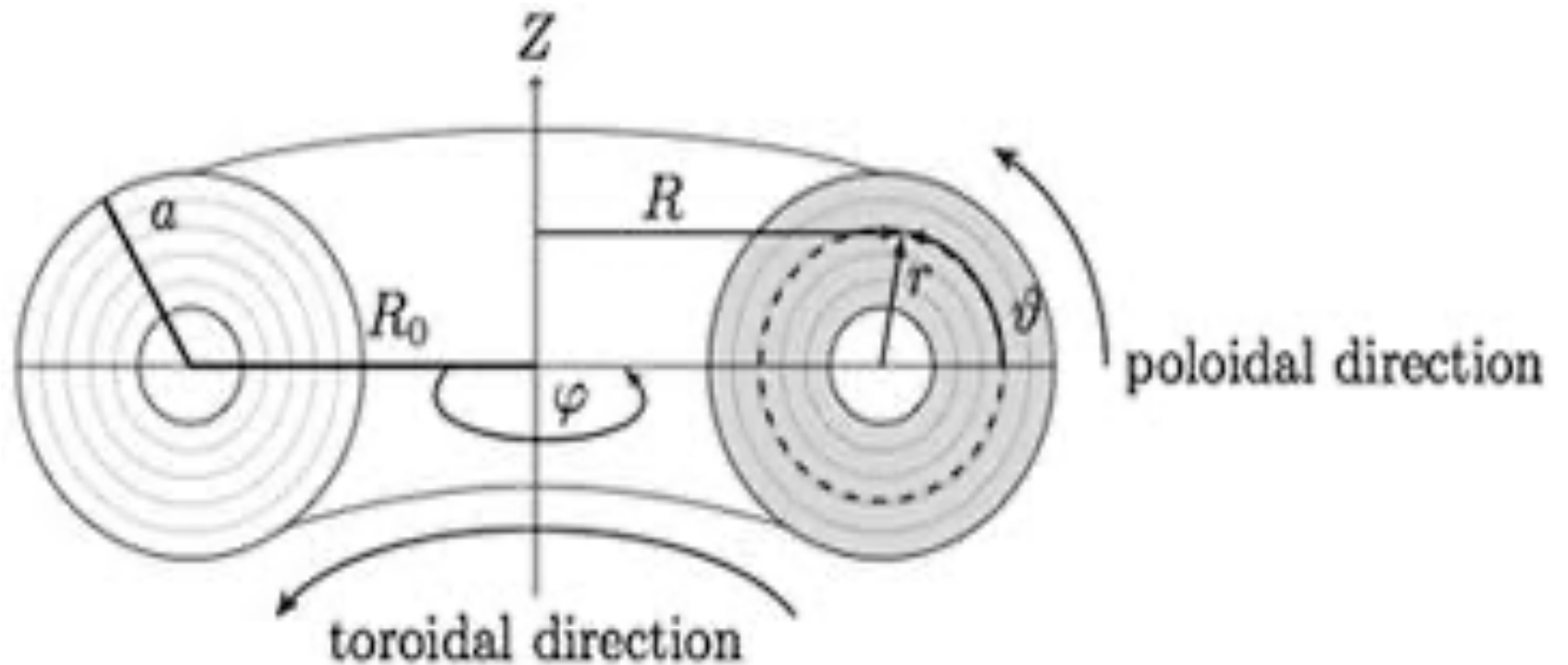
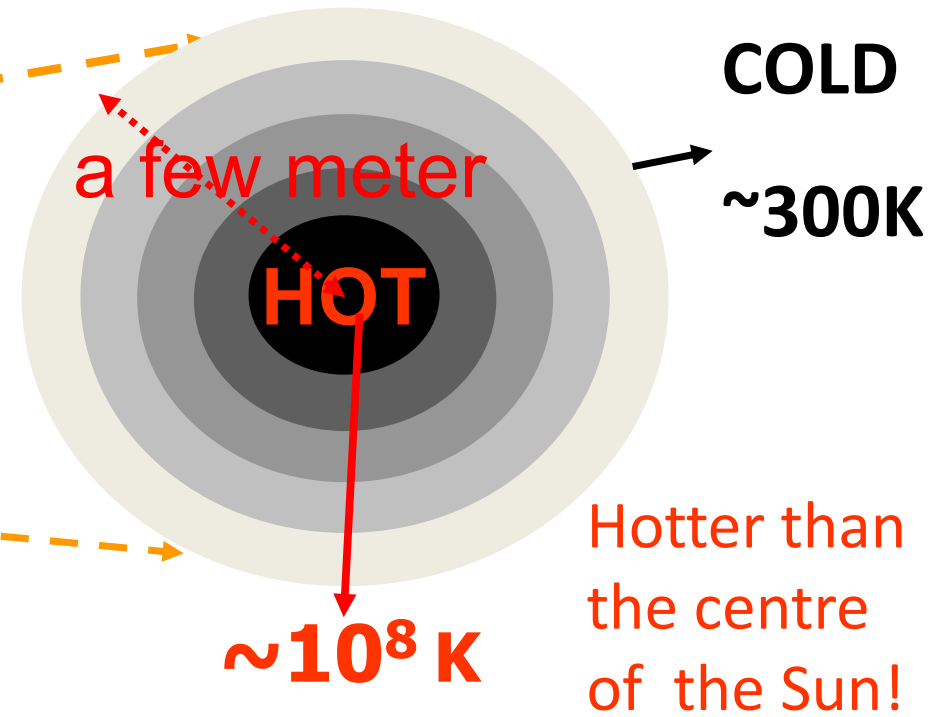
1.2 Self-organisation

1.1 Complexity/disorder (turbulence)

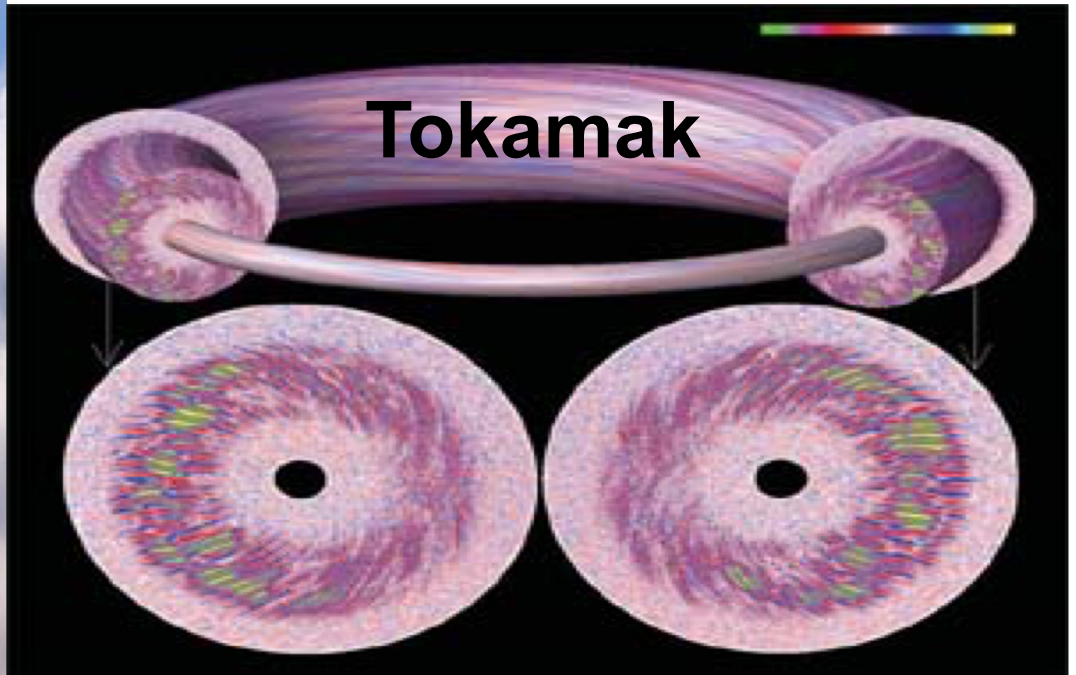
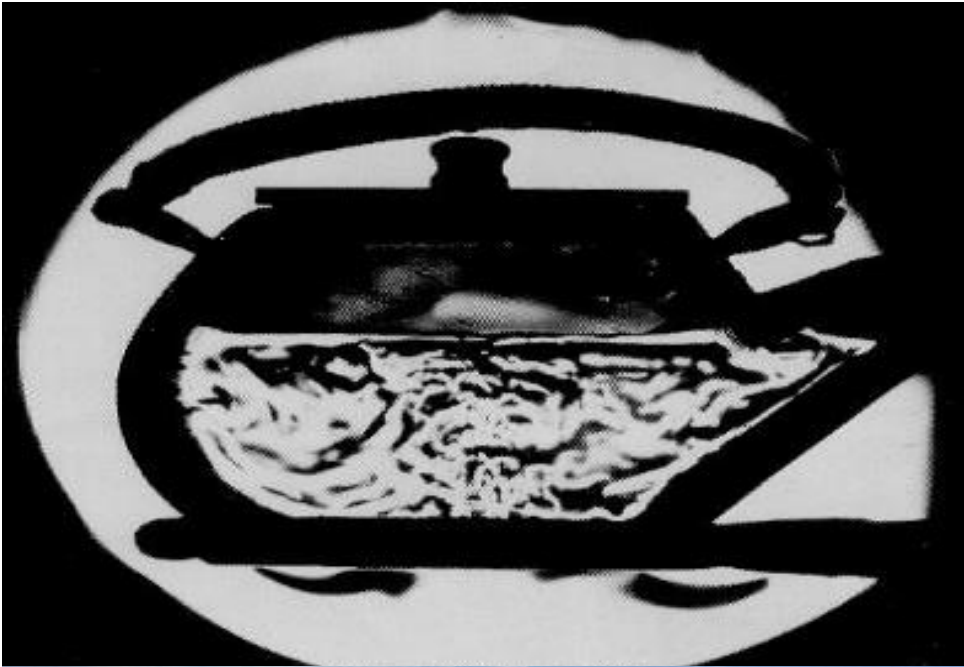




Yasmin
Andrew



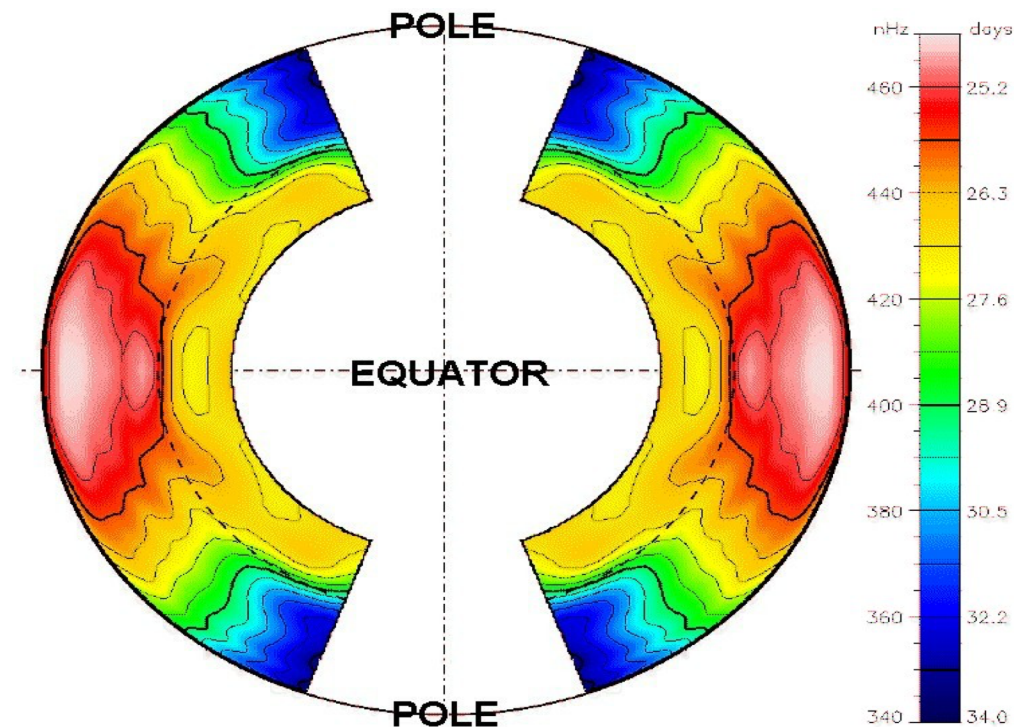
1.1 Complexity/disorder (turbulence)



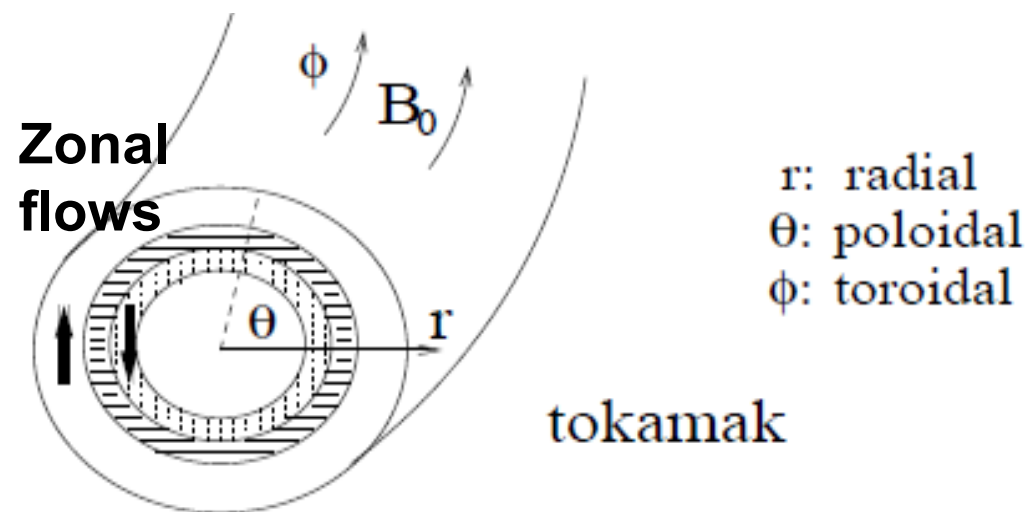
Order/Structure (shear, zonal flows)



Jovian zonal winds



Solar differential rotation

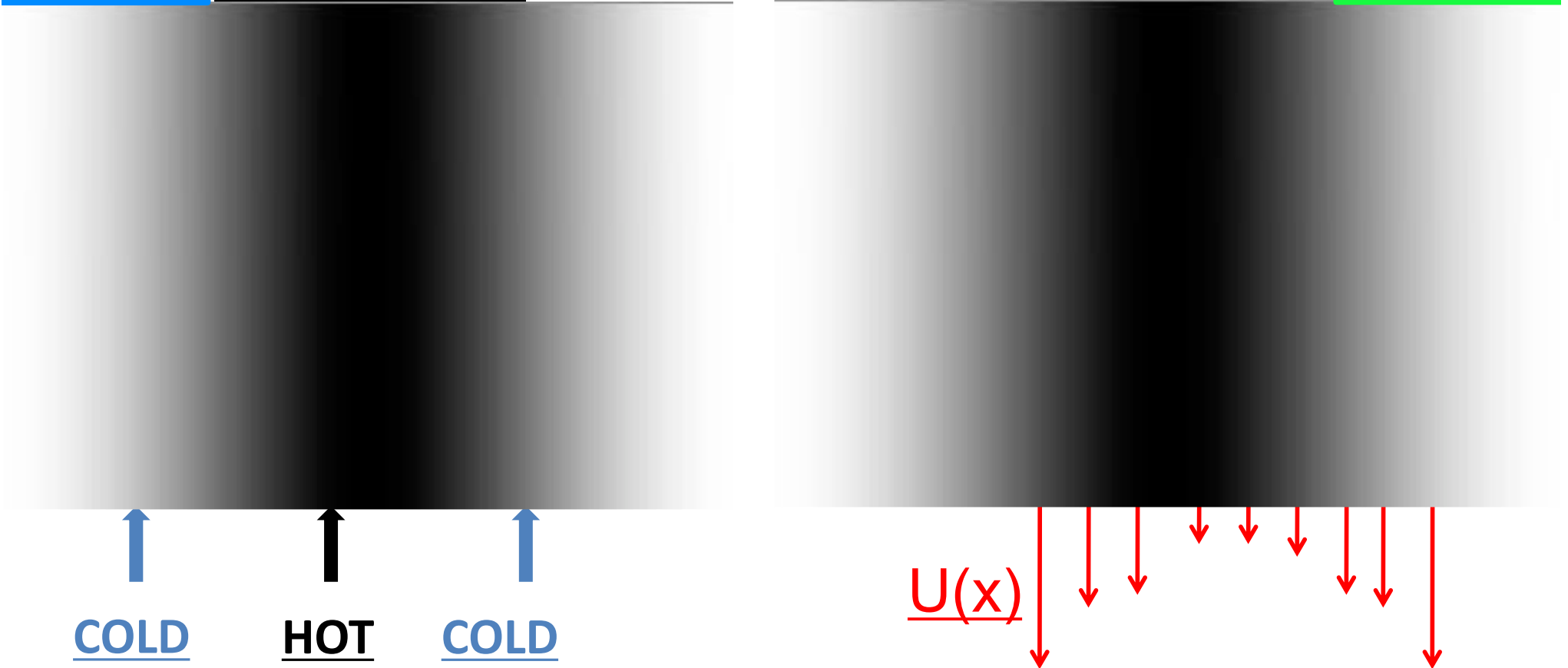




Reduction in turbulent mixing by shear flow

No shear

Strong shear



Turbulence enhances mixing but shear flow eats up turbulence, reducing turbulent mixing rate.

[Kim & Dubrulle 01,02; Kim 04,05,06,07; Kim et al 03,04,05,06; Leprovost & Kim 07,08,09,11; Numerical: Newton & Kim 08,09,11; Courvovsier & Kim 09; Sood, Hollerbach & Kim 16, etc.]

Turbulence grows
from instability

Shear flow grows
from turbulence

Shear flow eats up
turbulence

Less turbulence

Less shear flow

More turbulence

More shear flow

Less turbulence

Less shear flow

Rabbits grow by
eating grasses

Lions grow by
eating rabbits

Lions eat up
rabbits

Less rabbits

Less Lions

More rabbits

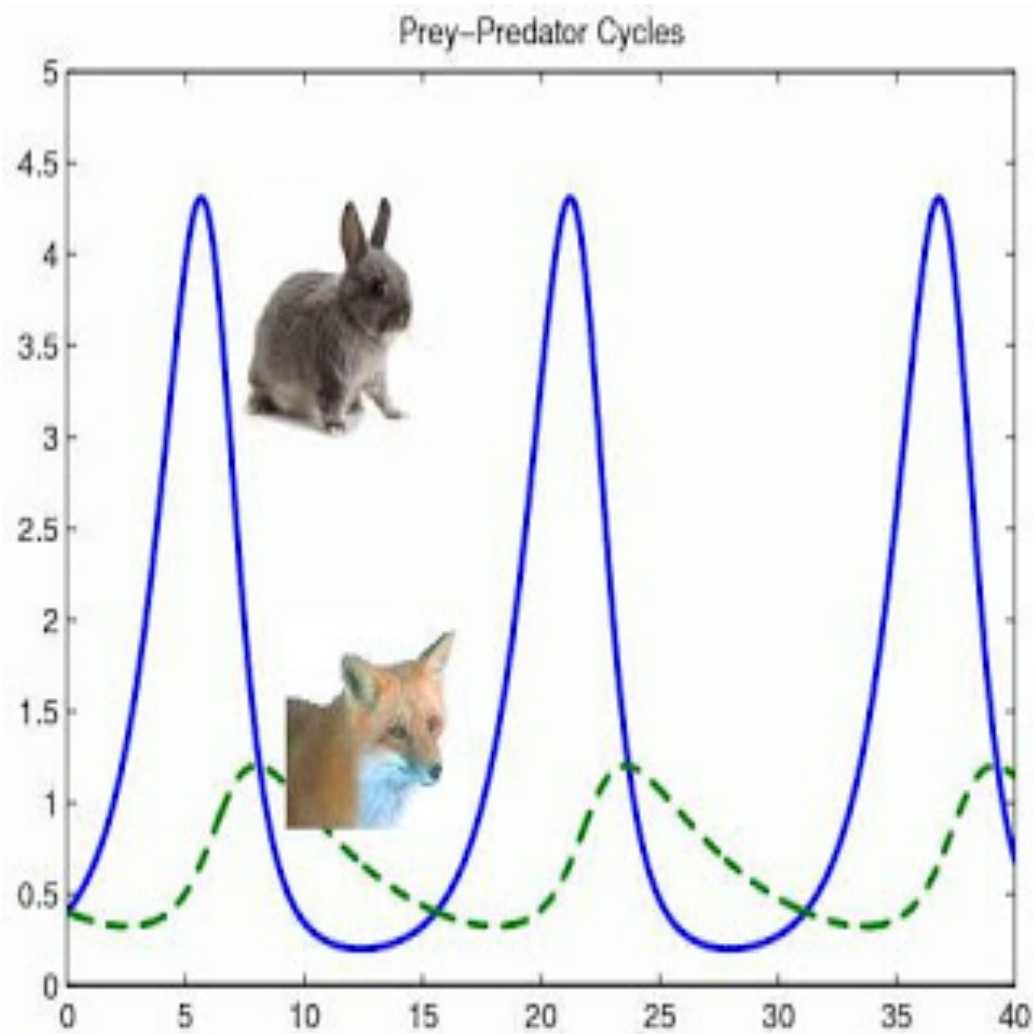
More lions

Less rabbits

Less lions



Prey-Predator



Predator:
Lion (zonal flow)

Prey:
Rabbit (turbulence)

2. The Low-to-High confinement (L-H) transition

2.1 Deterministic model

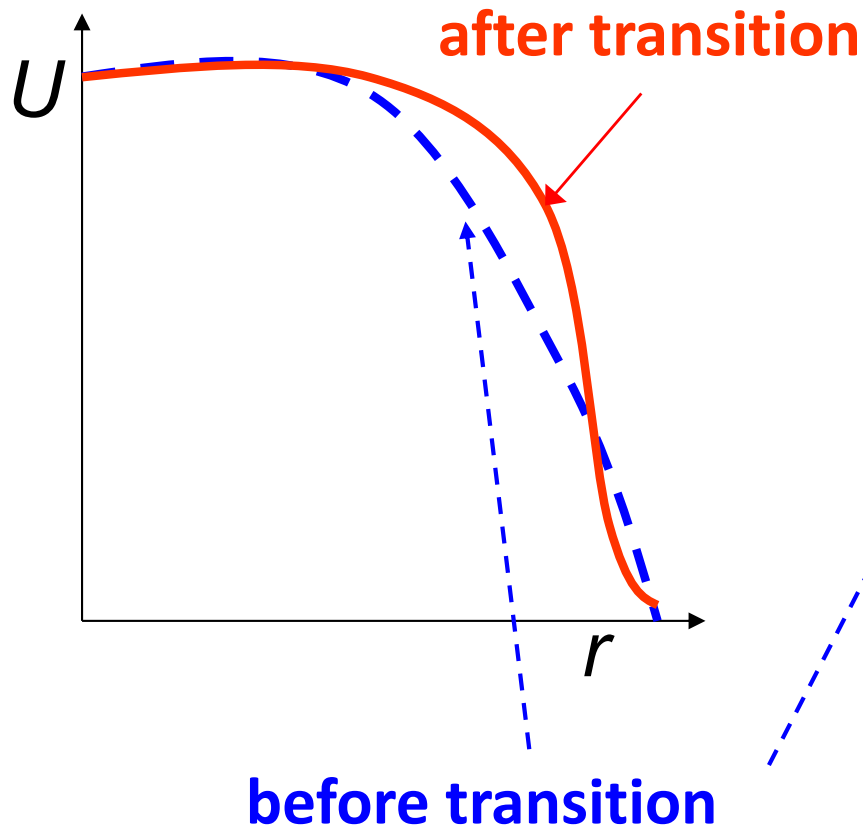
2.2 Stochastic model

2. Low-to-high confinement (L-H) transition

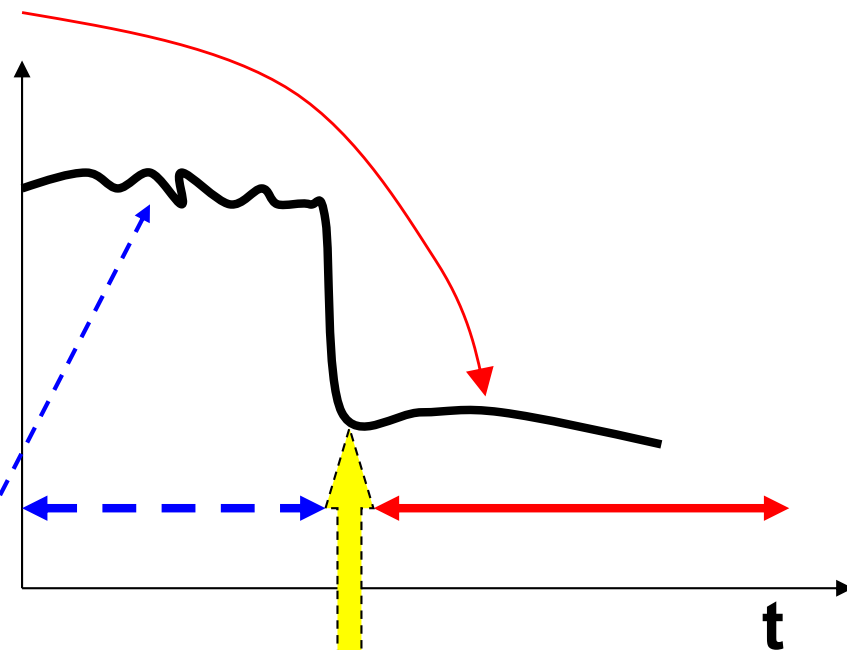
→ Spontaneous formation of shear flow

→ Improvement of confinement!

Radial profile of flow

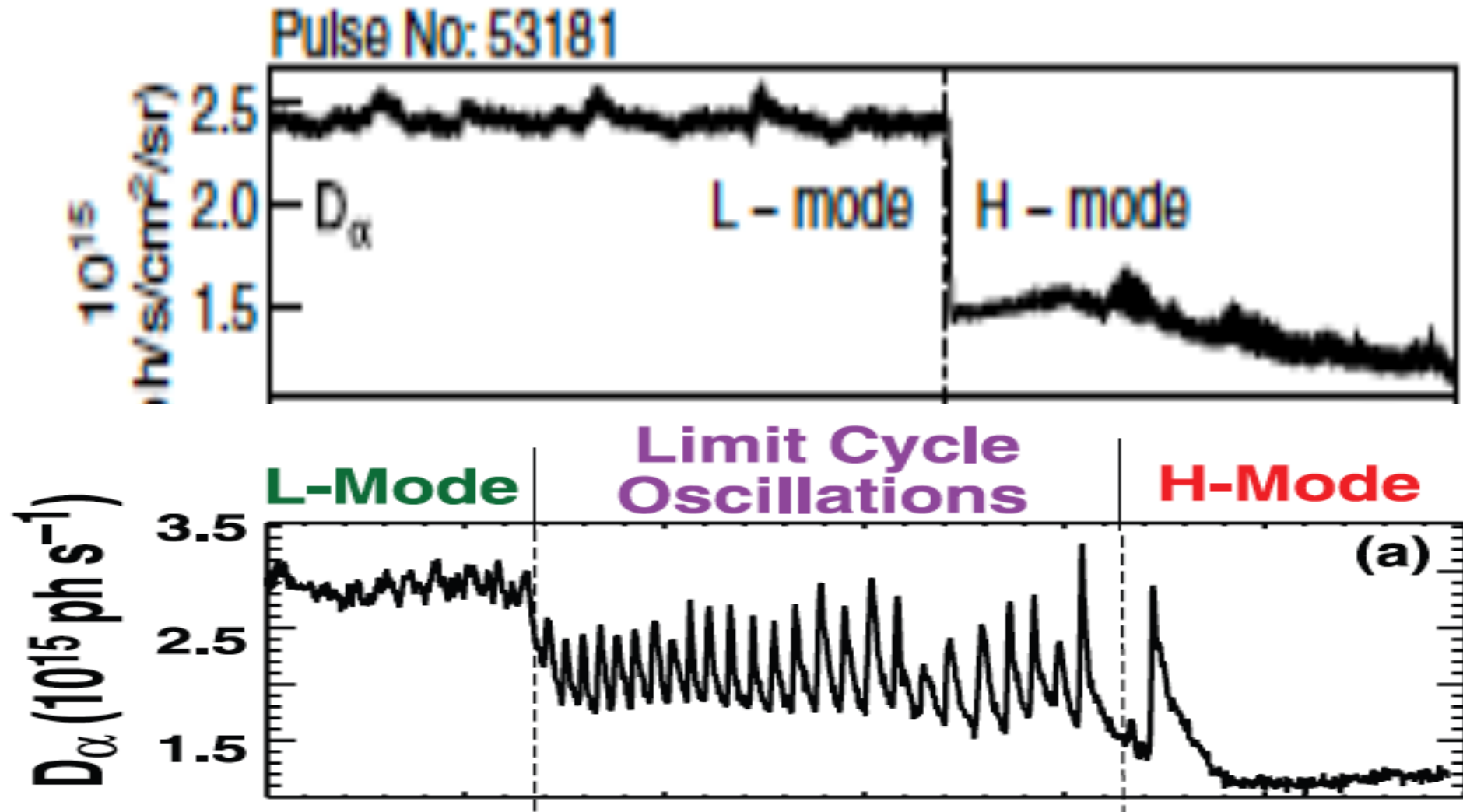


energy/particle loss



transition from low to high
confinement

JET: Andrew et al, 2006, PPCF 48, 479



DII-D: Schmitz et al, 2017, Nuclear Fusion 57, 025005

2.1 Deterministic model (Kim & Diamond, PRL 2003)

$$\partial_t \mathcal{E} = \mathcal{E} \mathcal{N} - a_1 \mathcal{E}^2 - a_2 V^2 \mathcal{E} - a_3 V_{ZF}^2 \mathcal{E},$$

$$\partial_t V_{ZF} = b_1 \frac{\mathcal{E} V_{ZF}}{1 + b_2 V^2} - b_3 V_{ZF},$$

$$\partial_t \mathcal{N} = -c_1 \mathcal{E} \mathcal{N} - c_2 \mathcal{N} + Q.$$

N : Temperature/density gradient (rabbit food: grasses)

Q : External heating (water/sunlight)

\mathcal{E} : Turbulence (prey: rabbits)

V_{ZF} : zonal flows (predator: lions)

$V = dN^2$: mean flows (super predator)

a_i, b_i, c_i, d : constant model parameters

From Kim & Diamond PRL 03

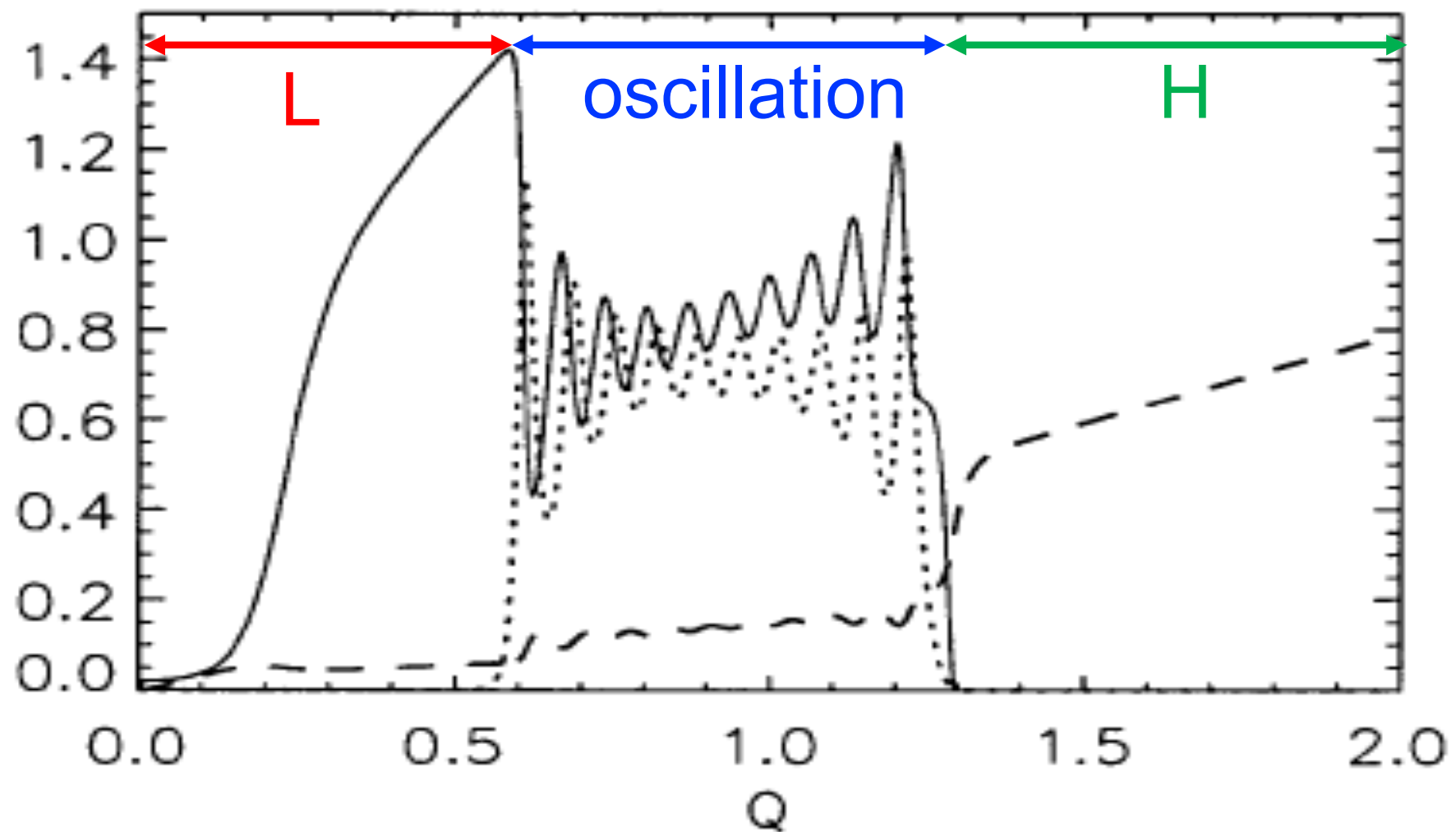


FIG. 1. Evolution of \mathcal{E} (solid line), V_{ZF} (dotted line), and $\mathcal{N}/5$ (dashed line) as a function of input power $Q = 0.01t$. Parameter values are $a_1 = 0.2$, $a_2 = a_3 = 0.7$, $b_1 = 1.5$, $b_2 = b_3 = 1$, $c_1 = 1$, $c_2 = 0.5$, and $d = 1$.

2.2 Stochastic model (Kim & Hollerbach, PRL submitted)

$$\frac{dx}{dt} = f + \xi, \quad f = \frac{1}{2} \left[N - a_1 x^2 - a_2 V^2 - a_3 v^2 \right] x,$$

$$\frac{dv}{dt} = g + \eta, \quad g = \frac{b_1 x^2 v}{1 + b_2 V^2} - b_3 v,$$

$$x = \pm \sqrt{\mathcal{E}}, \quad v = V_{ZF}, \quad N \sim \frac{Q}{c_1 x^2 + c_2} \quad \frac{\partial N}{\partial t} = 0$$

N : Temperature (rabbit food: grasses)

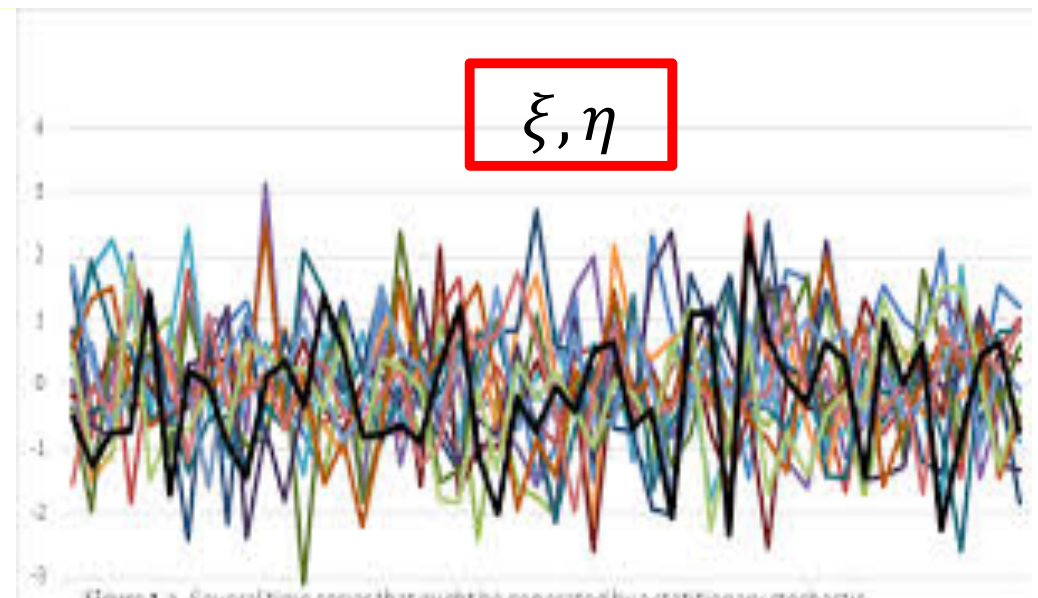
Q : External heating (water/sunlight)

$\mathcal{E} = x^2$: Turbulence (prey: rabbits)

$v = V_{ZF}$: zonal flows (predator: lions)

$V = dN^2$: mean flows (super predator)

a_i, b_i, c_i, d : constant model parameters



ξ and η are two independent Gaussian noise
with a short correlation time

$$\langle \xi(t) \xi(t') \rangle = 2D_x \delta(t - t'), \quad \langle \eta(t) \eta(t') \rangle = 2D_v \delta(t - t')$$

$$\langle \xi(t) \eta(t') \rangle = 0, \quad \langle \xi \rangle = \langle \eta \rangle = 0.$$

Fokker-Planck equation for $p(x, v, t)$

$$\frac{\partial p}{\partial t} = -\frac{\partial}{\partial v}(g p) - \frac{\partial}{\partial x}(f p) + D_x \frac{\partial^2 p}{\partial x^2} + D_v \frac{\partial^2 p}{\partial v^2}$$

Recall: Gaussian PDF

$$p(x,t) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{|x - \mu|^2}{2\sigma^2}\right) = \sqrt{\frac{\beta}{\pi}} \exp(-\beta |x - \mu|^2)$$

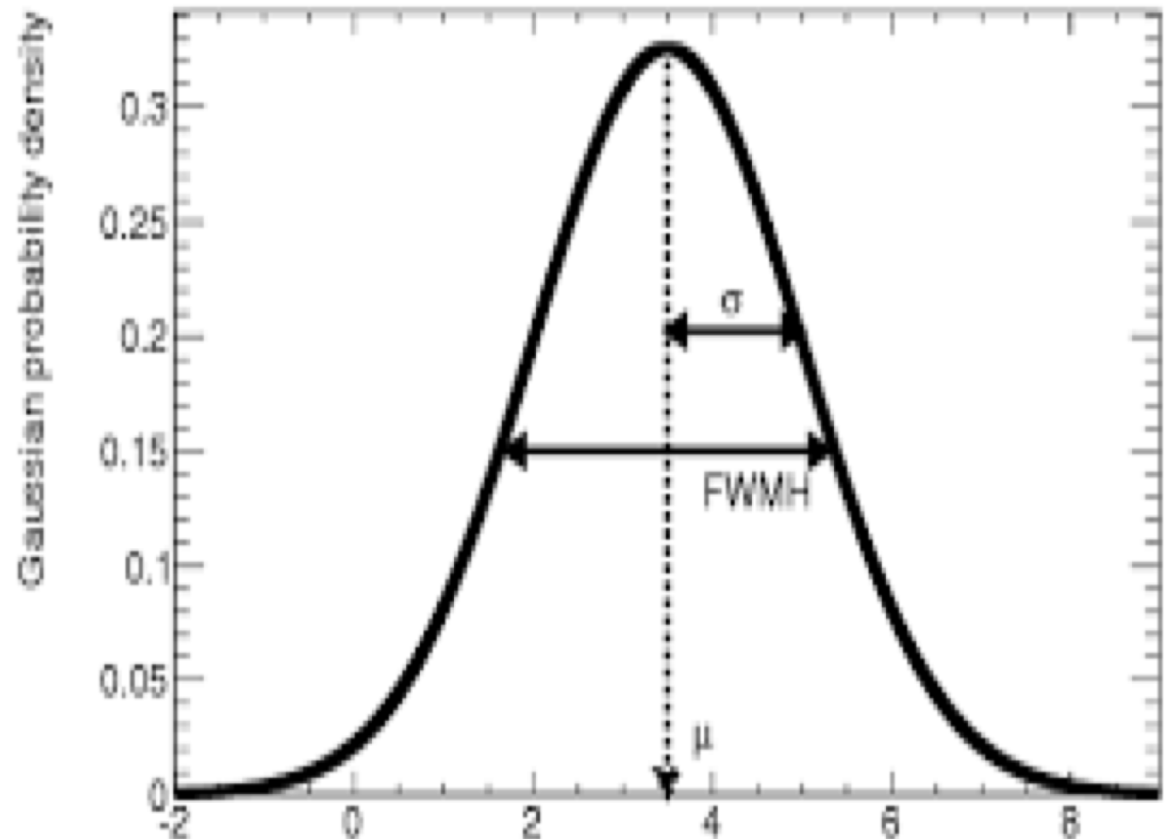
$\lambda^i = (\sigma, \mu)$ parameter

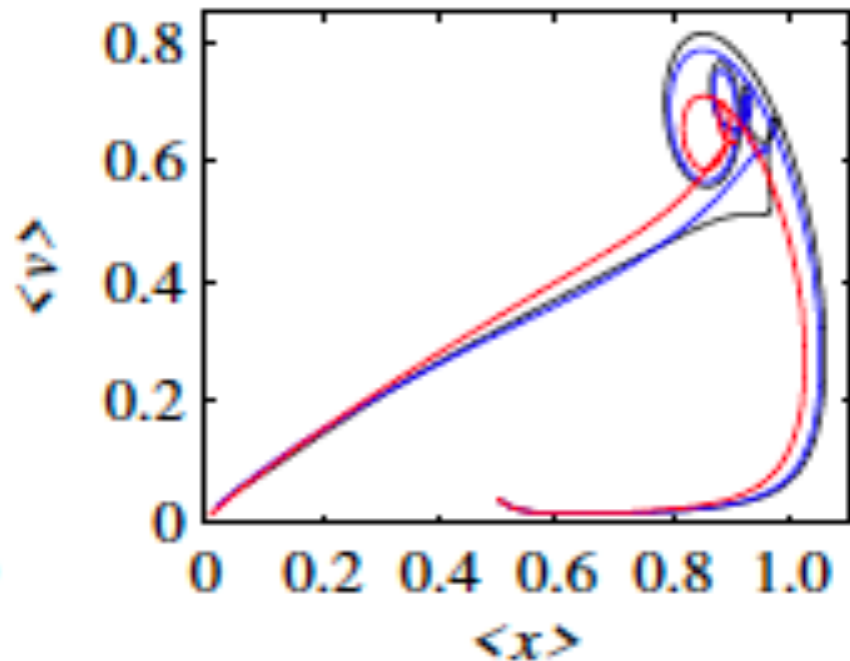
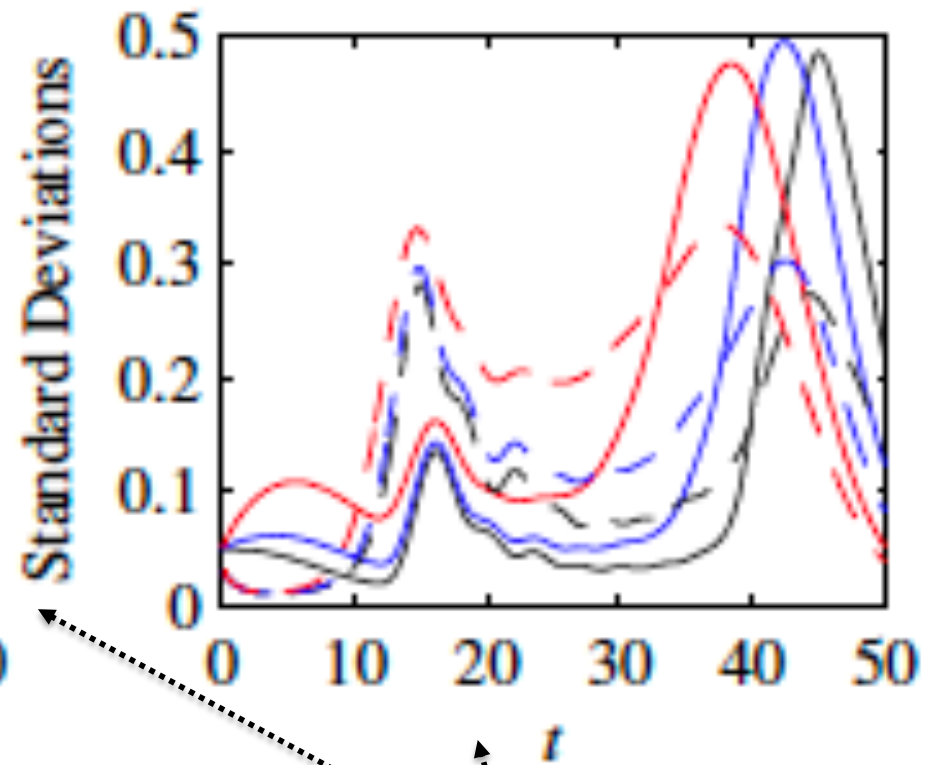
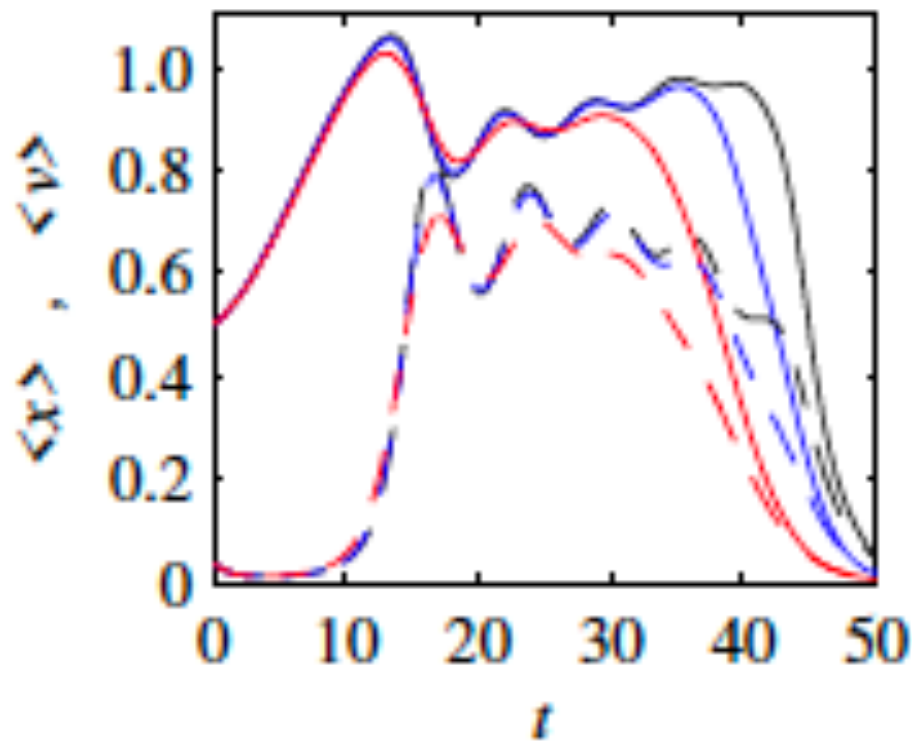
μ = mean value

σ = standard deviation

$$\beta = \frac{1}{2\sigma^2}$$

= inverse temperature



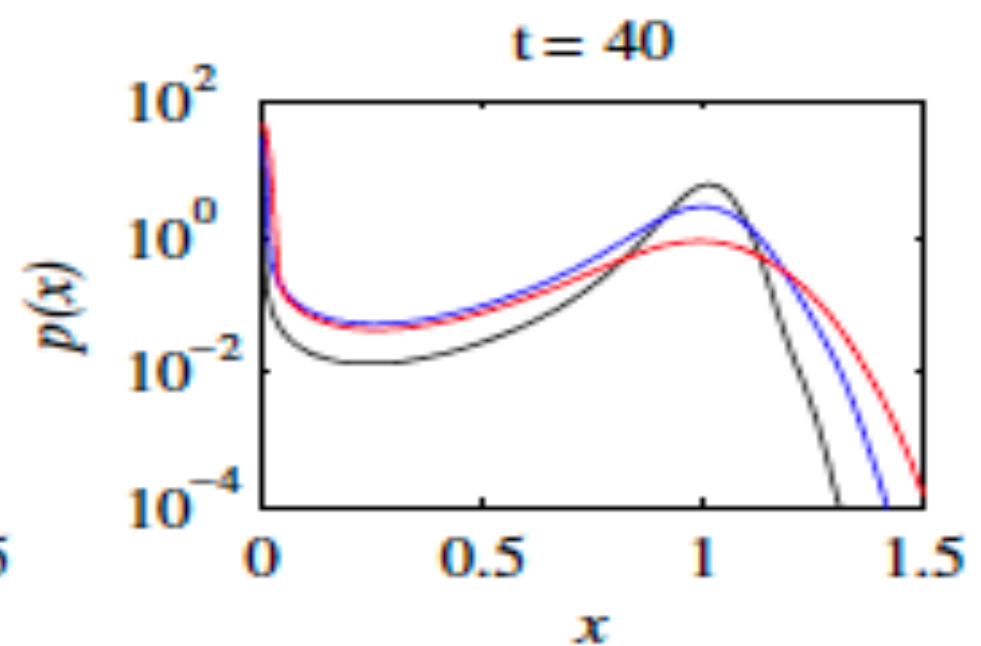
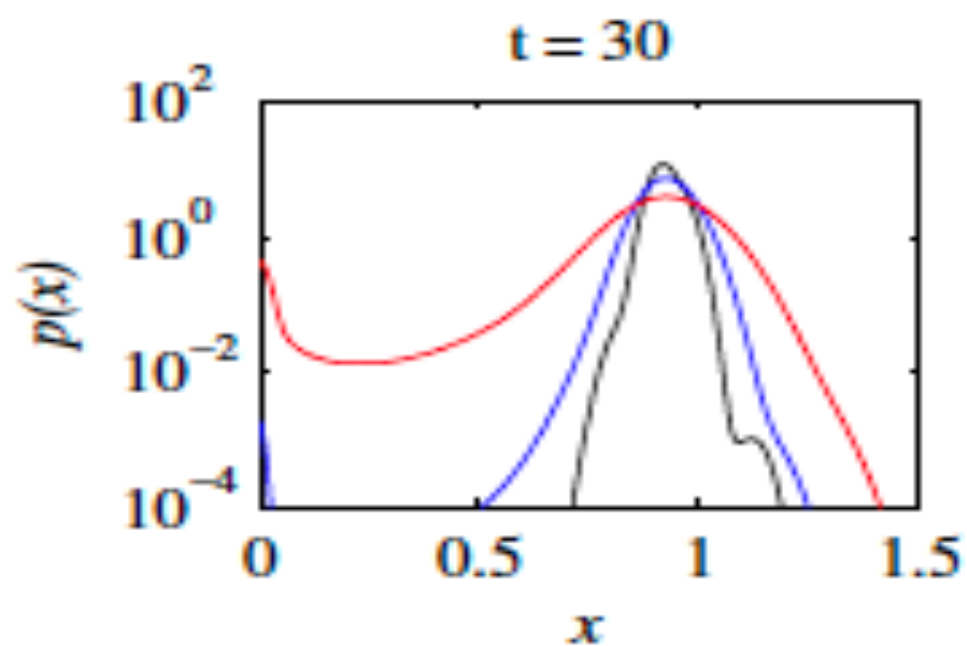
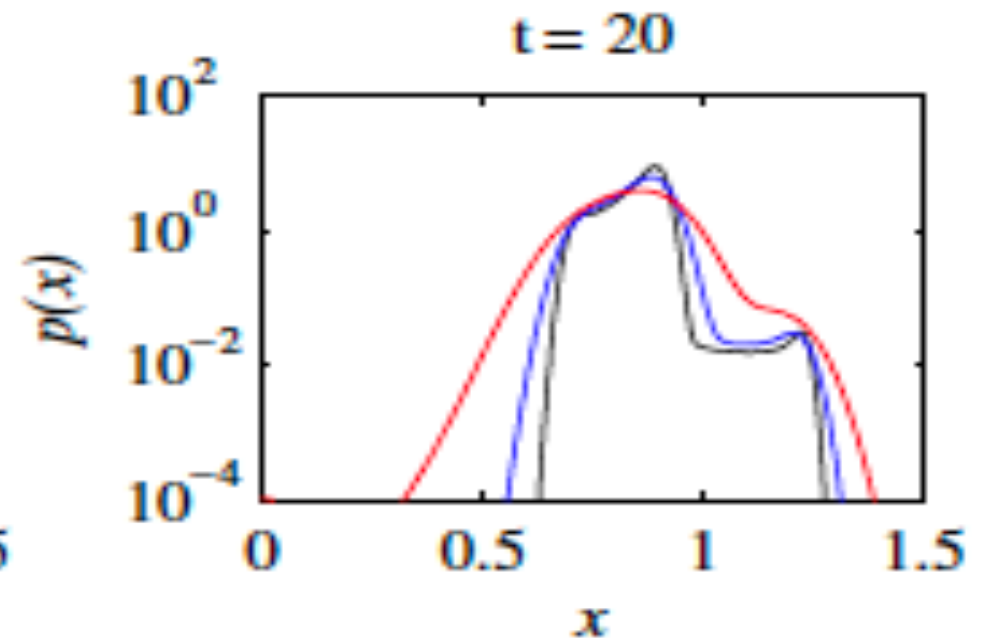
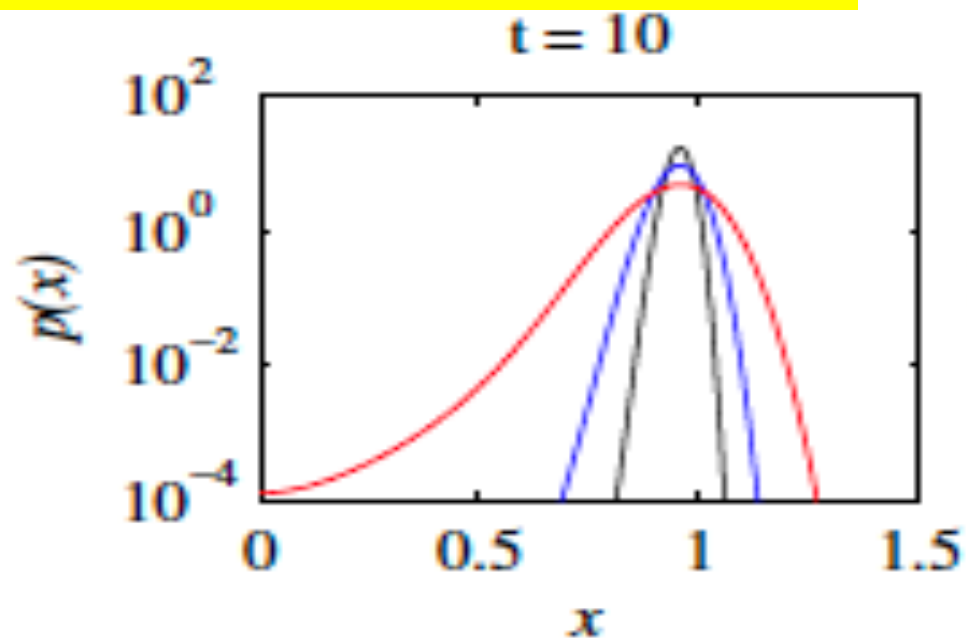


x: solid line (turbulence)
v: dashed line (zonal flow)

$\langle x \rangle$: mean value of x
 $\langle v \rangle$: mean value of v

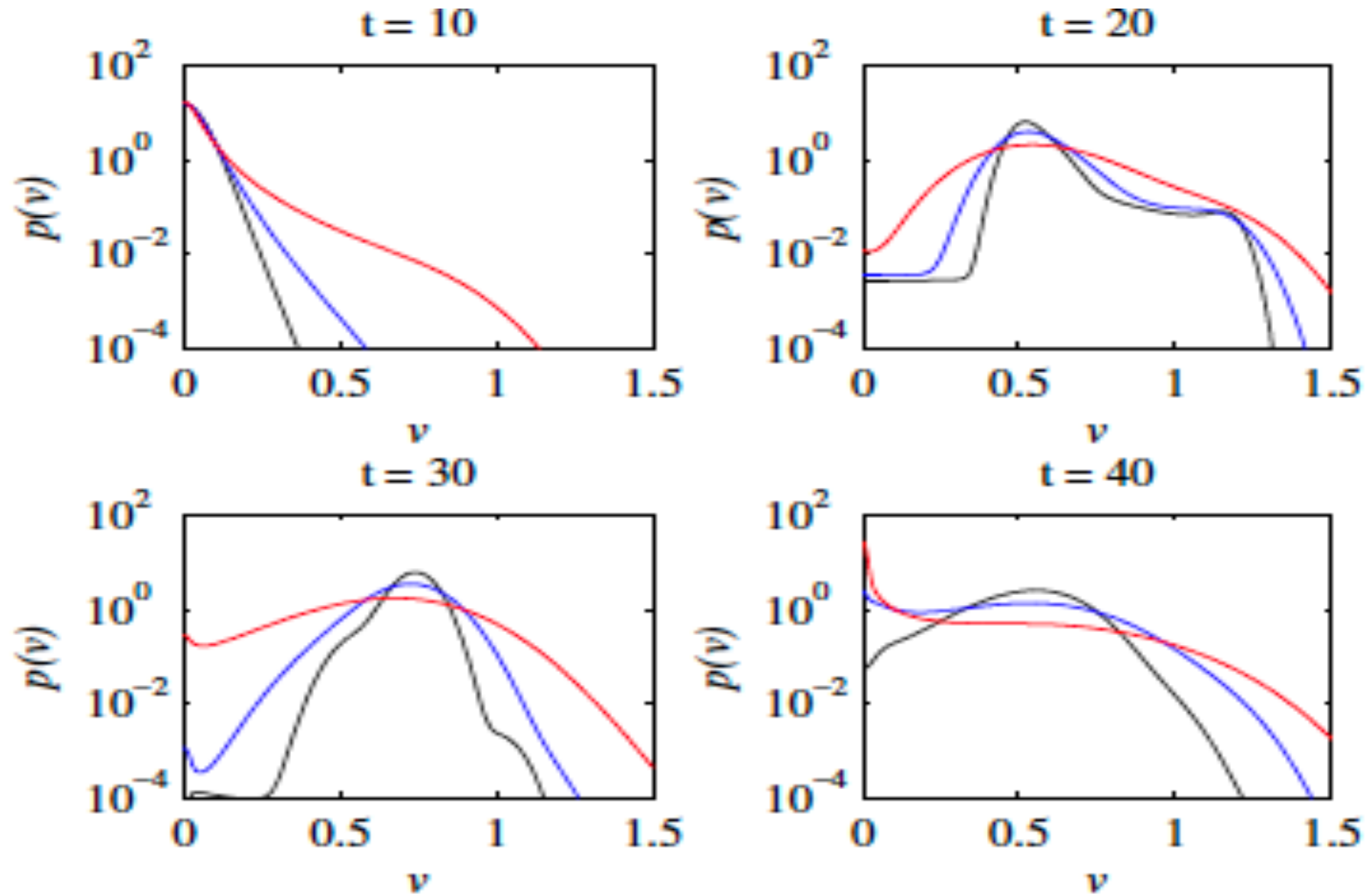
$$p(x, t) = \int dv p(x, v, t)$$

x: turbulence, v: zonal flow

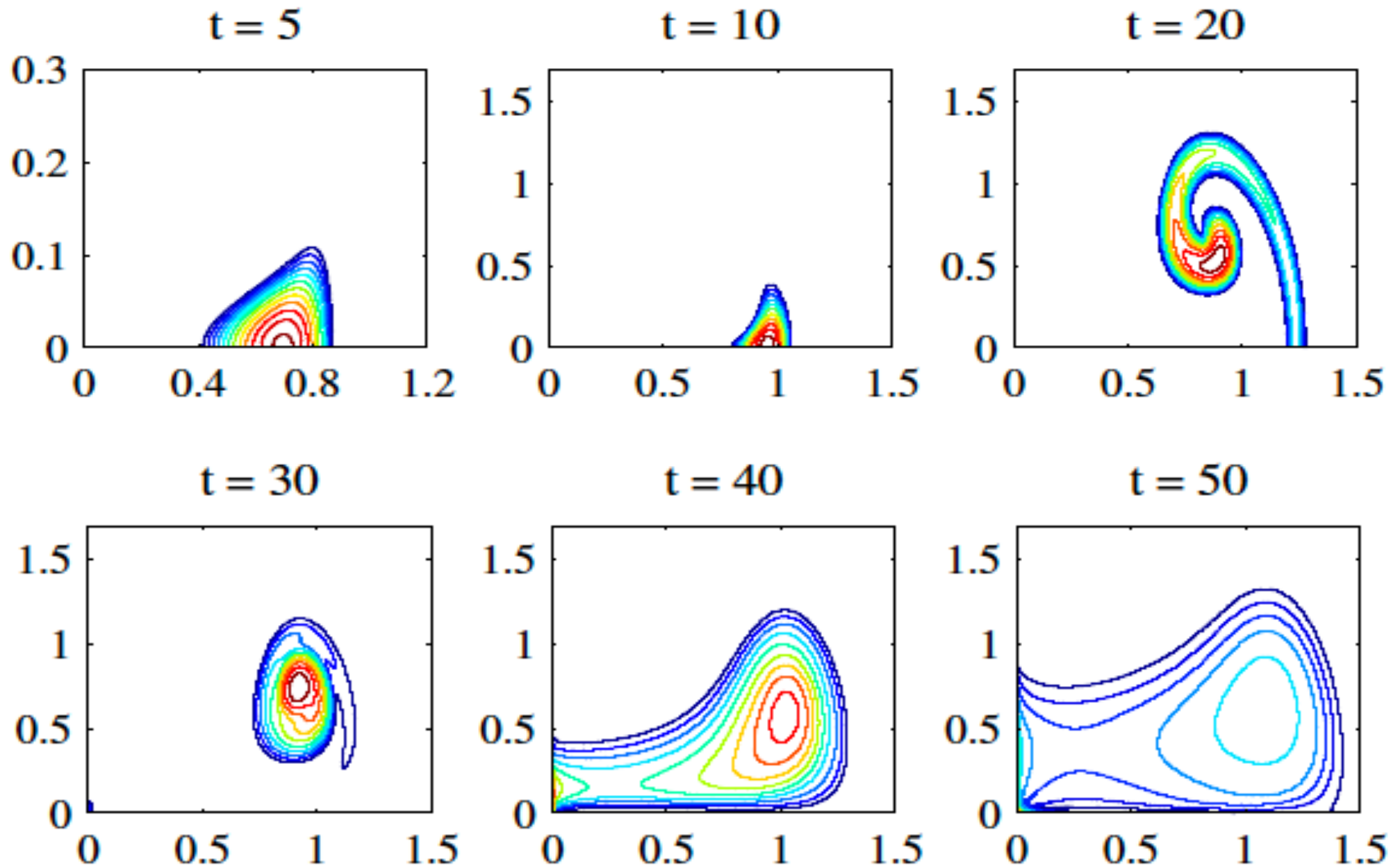
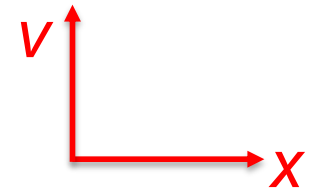


$$p(v, t) = \int dx p(x, v, t)$$

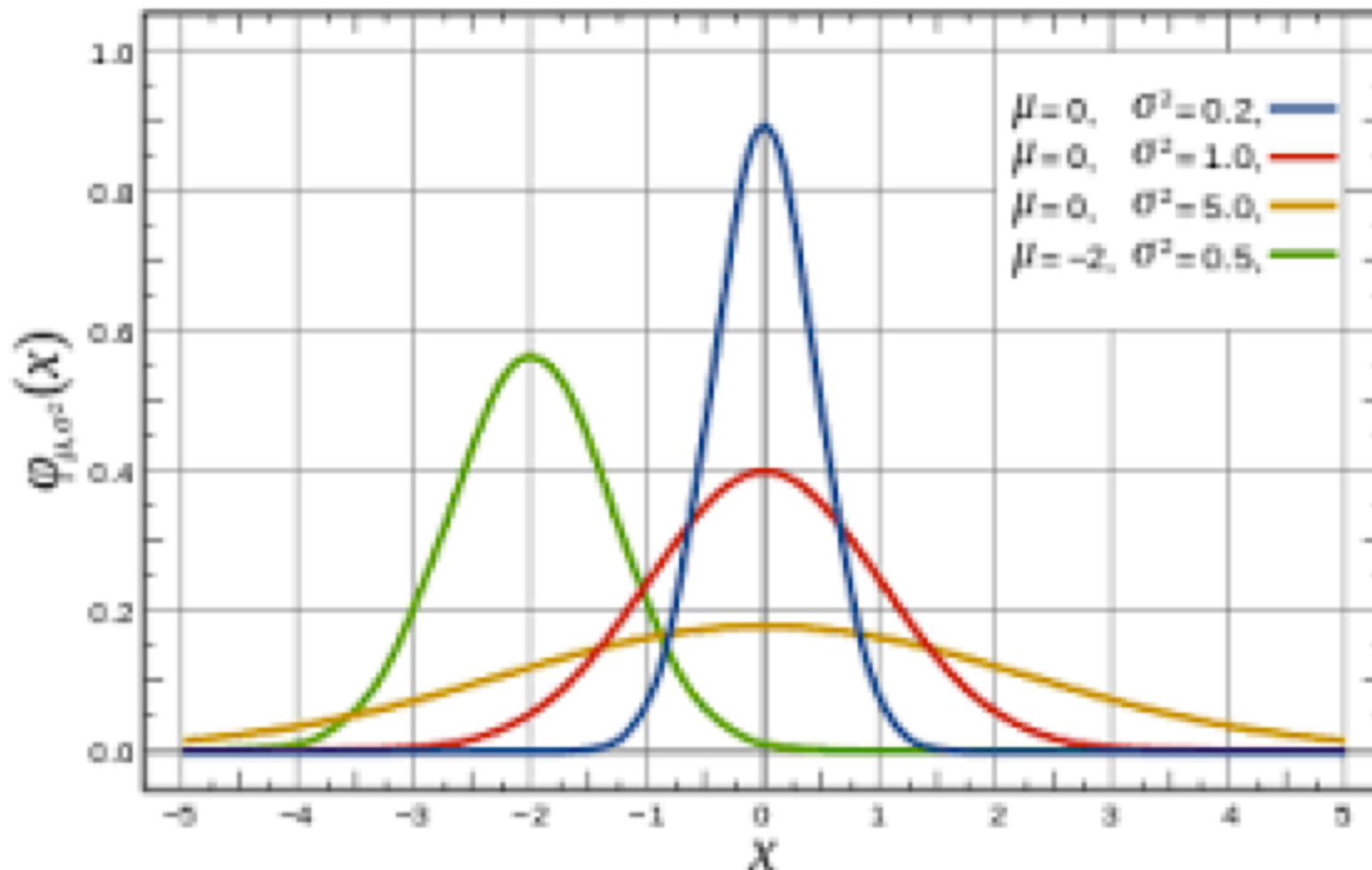
x: turbulence, v: zonal flow



$p(x,v,t)$ in the plane of x and v



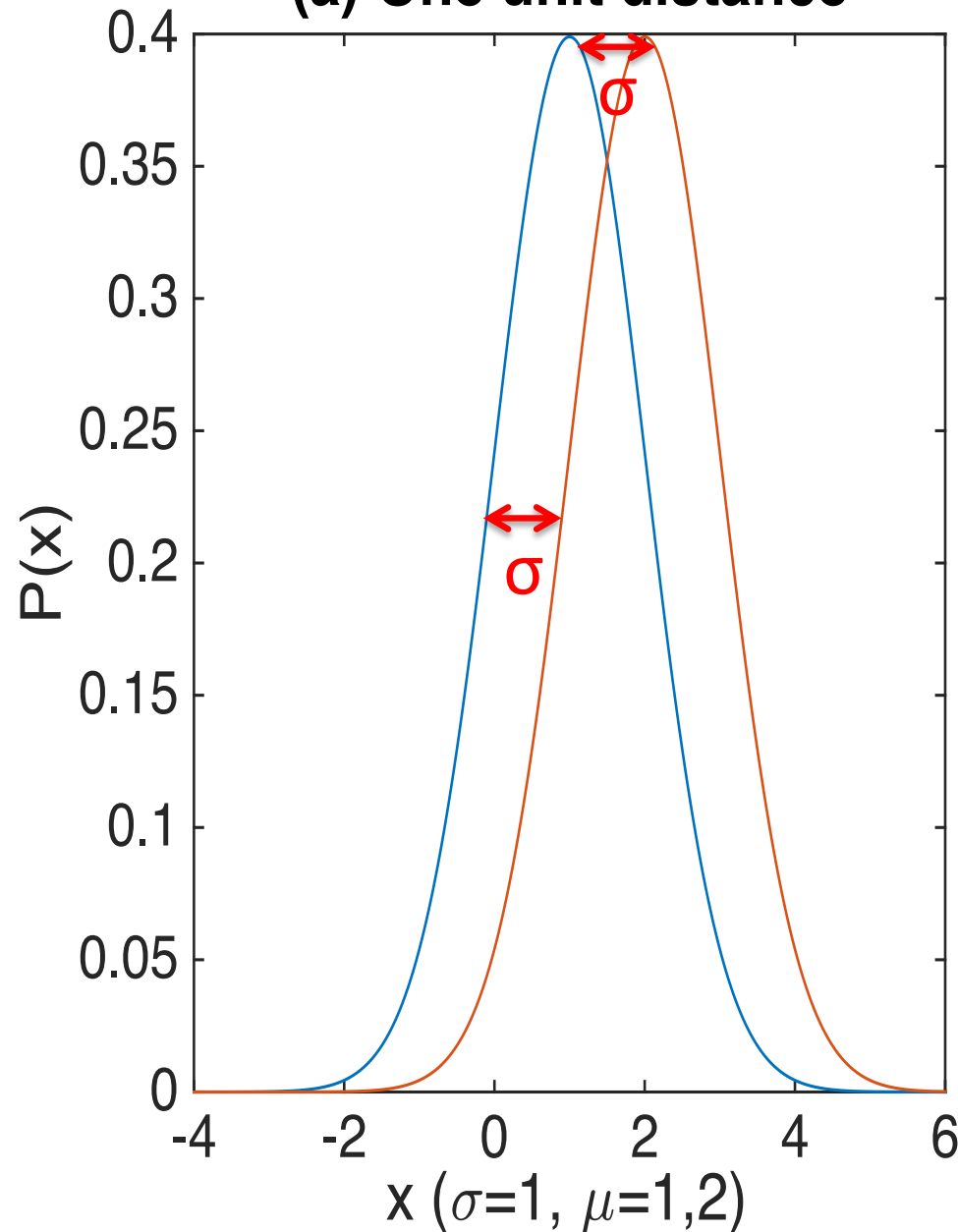
Information length



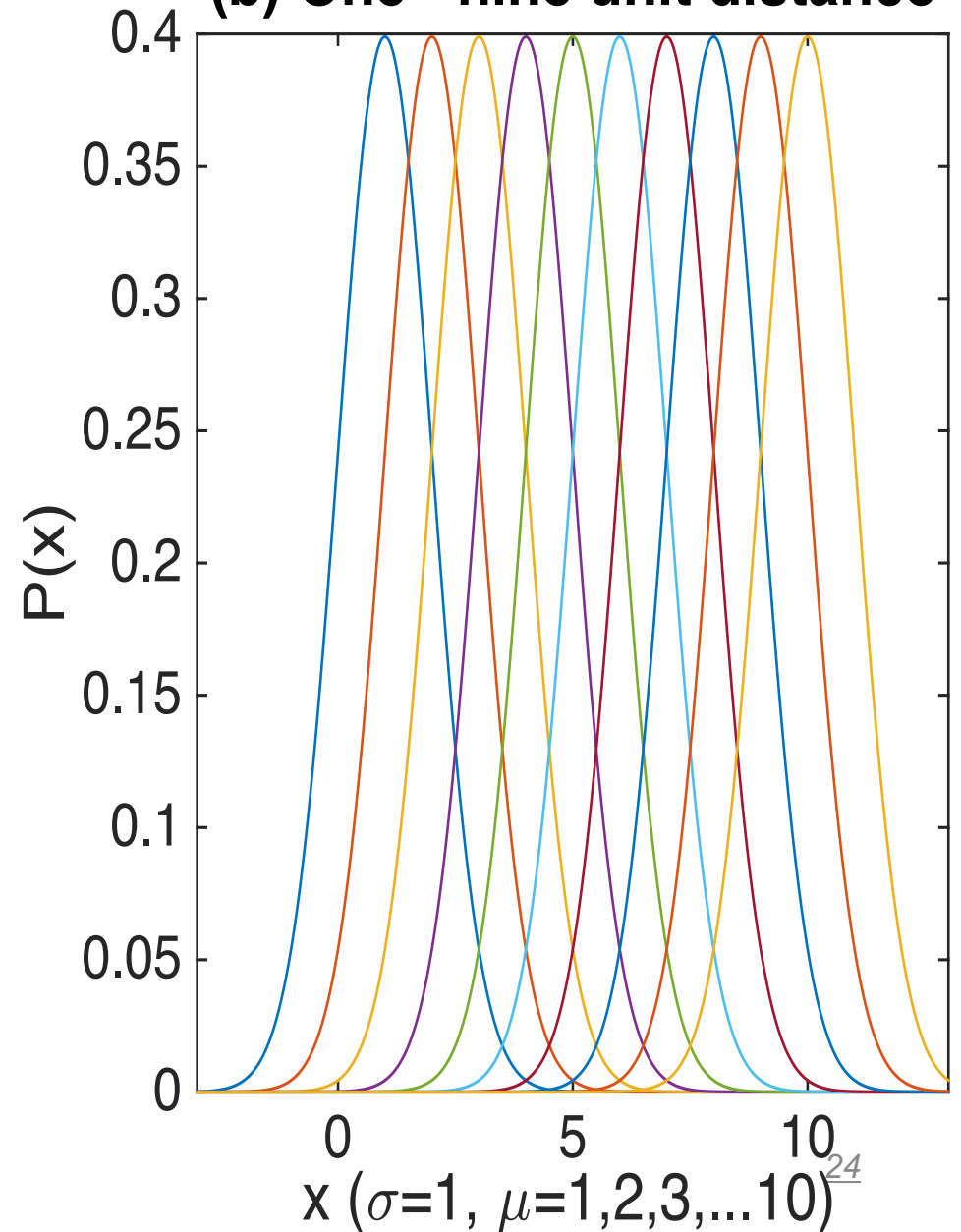
Q: Quantify similarity among PDFs by number (distance):
Smaller distance for similar PDFs
Larger distance for disparate PDFs

Unit of distance = width of PDF (variance σ)

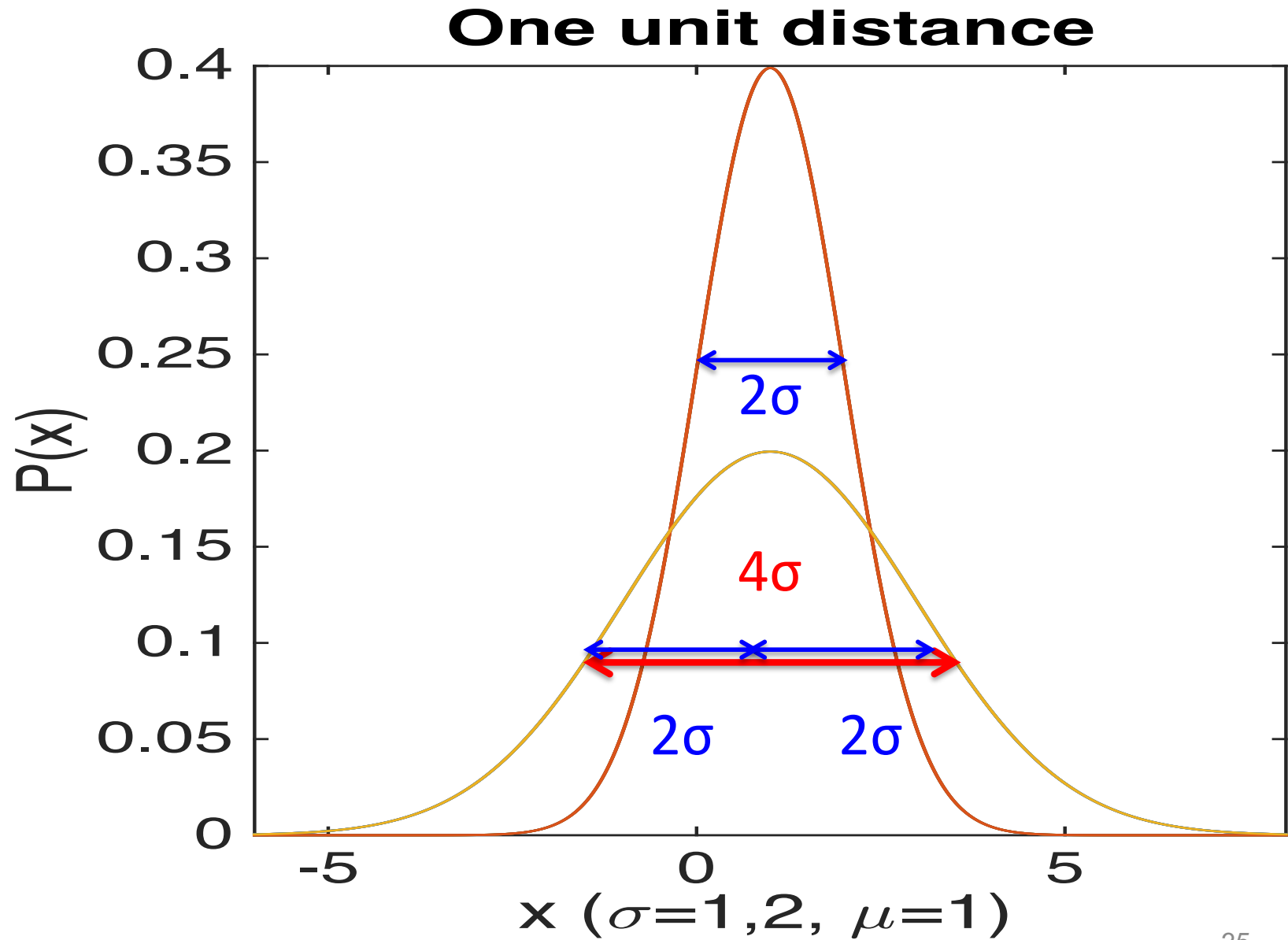
(a) One unit distance



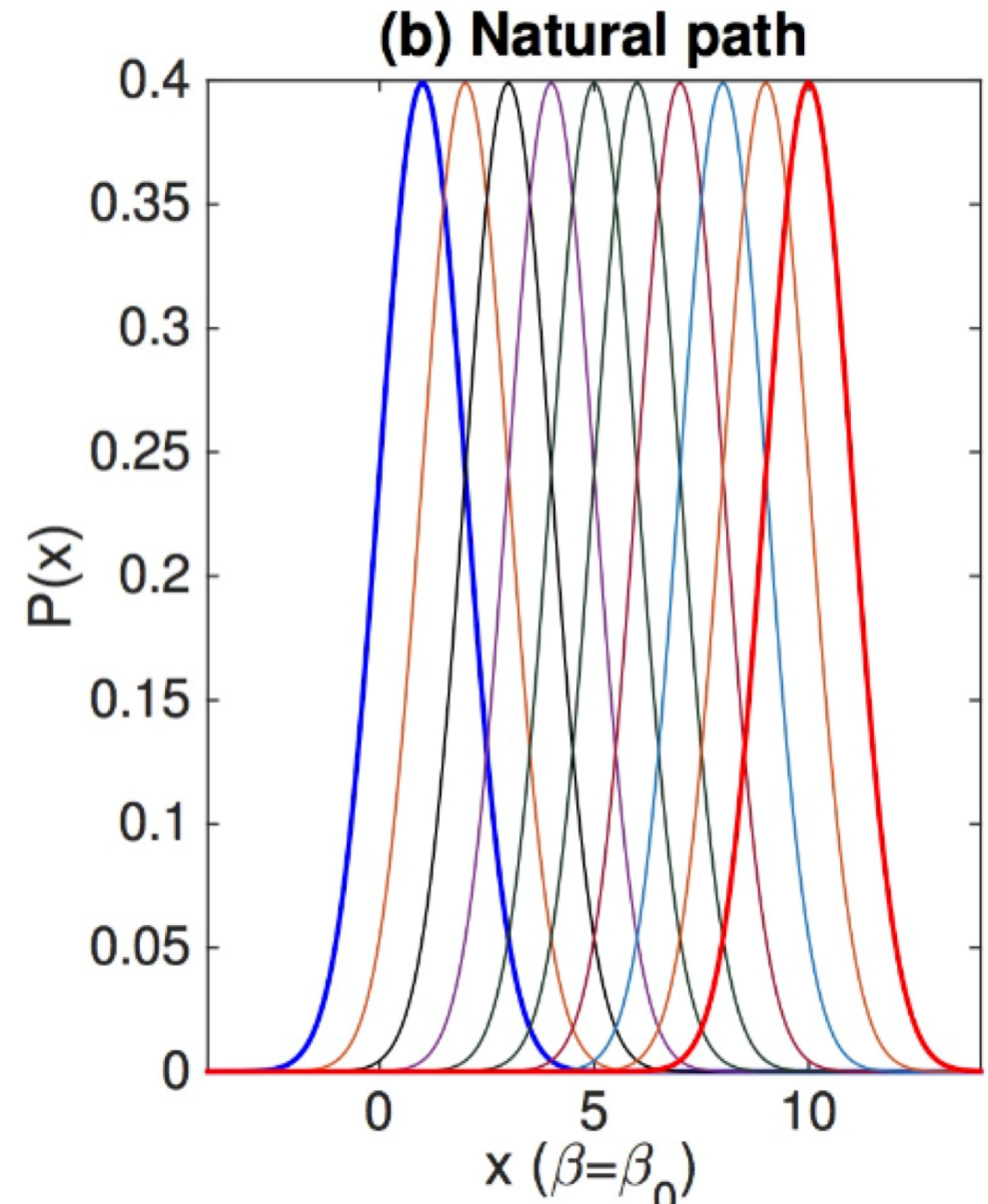
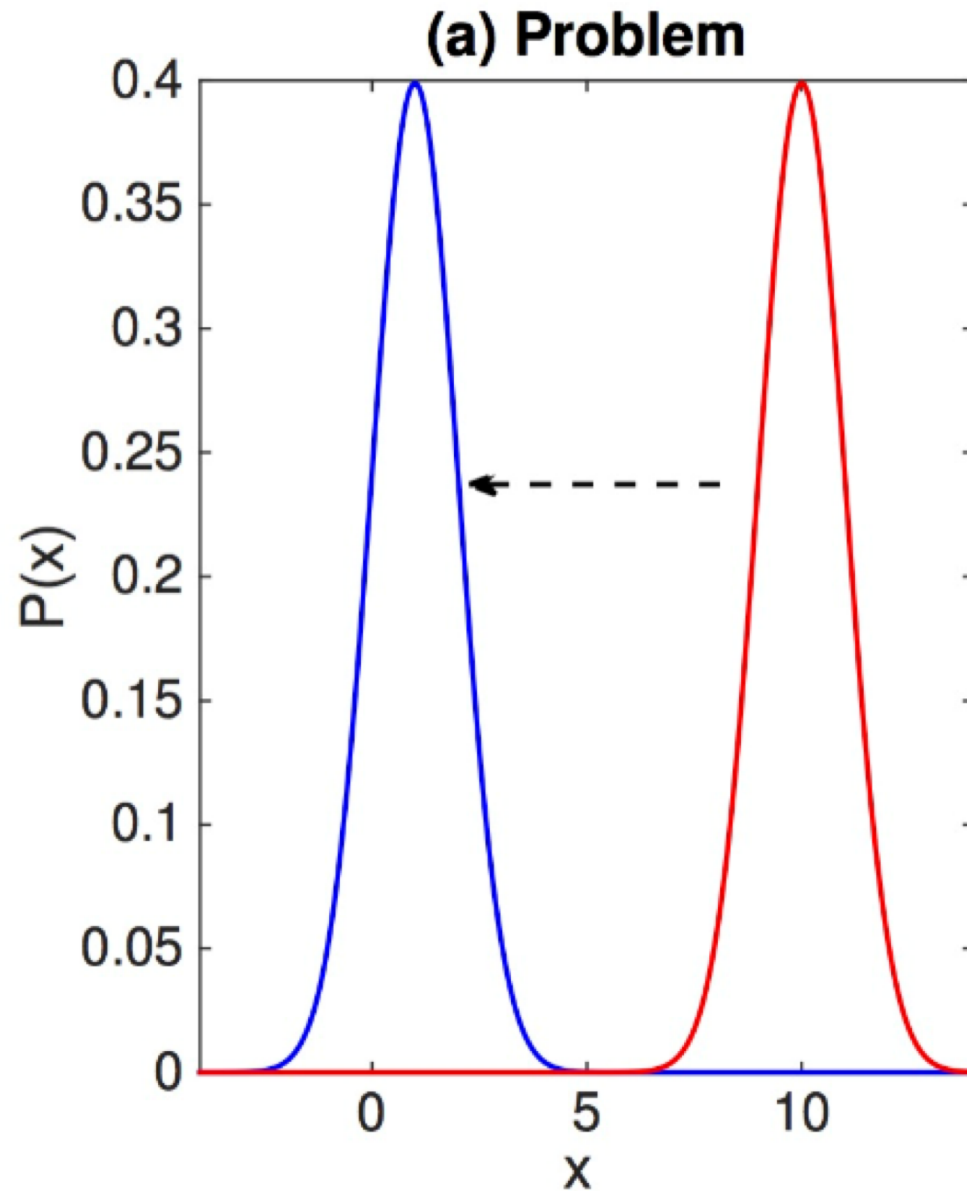
(b) One - nine unit distance



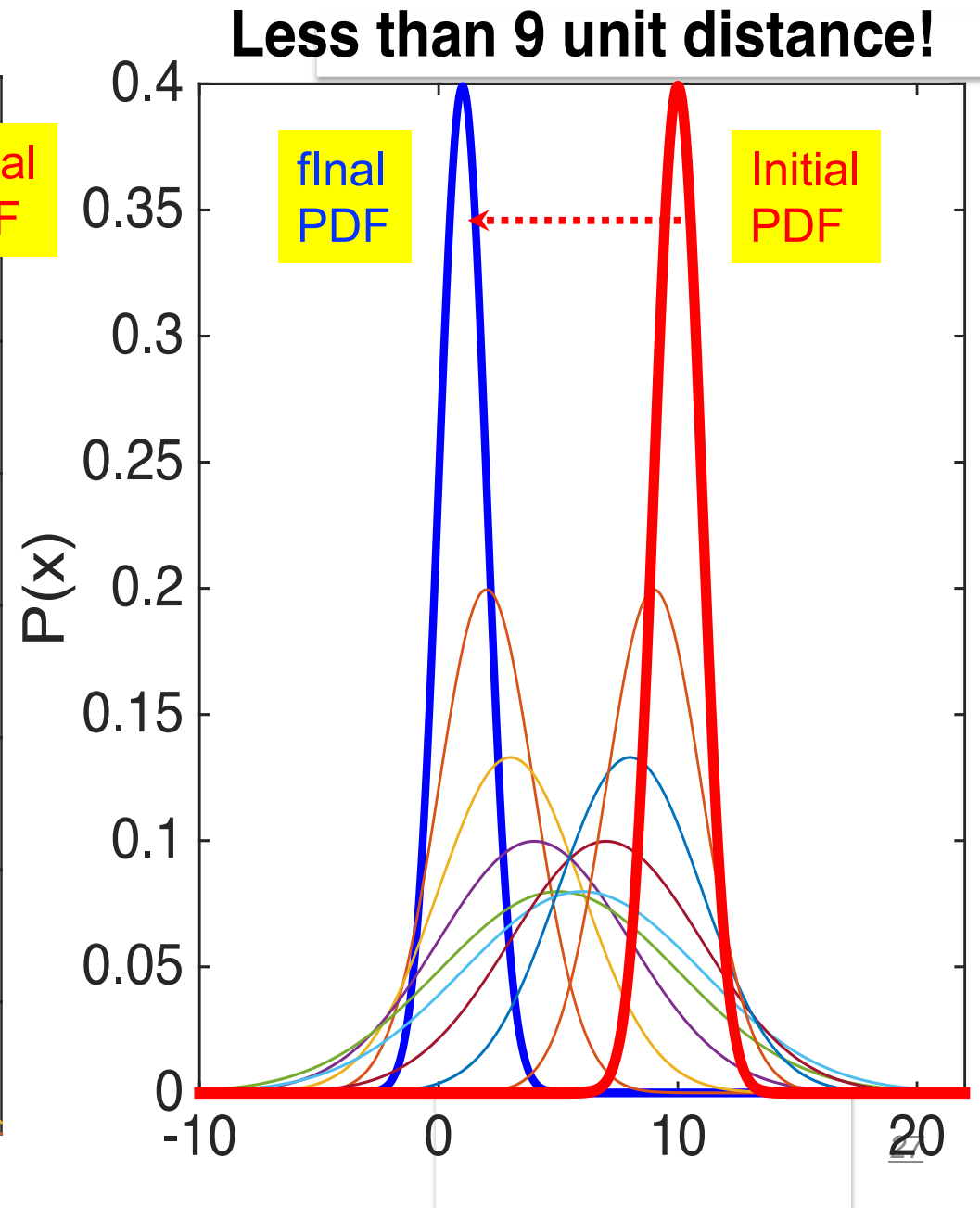
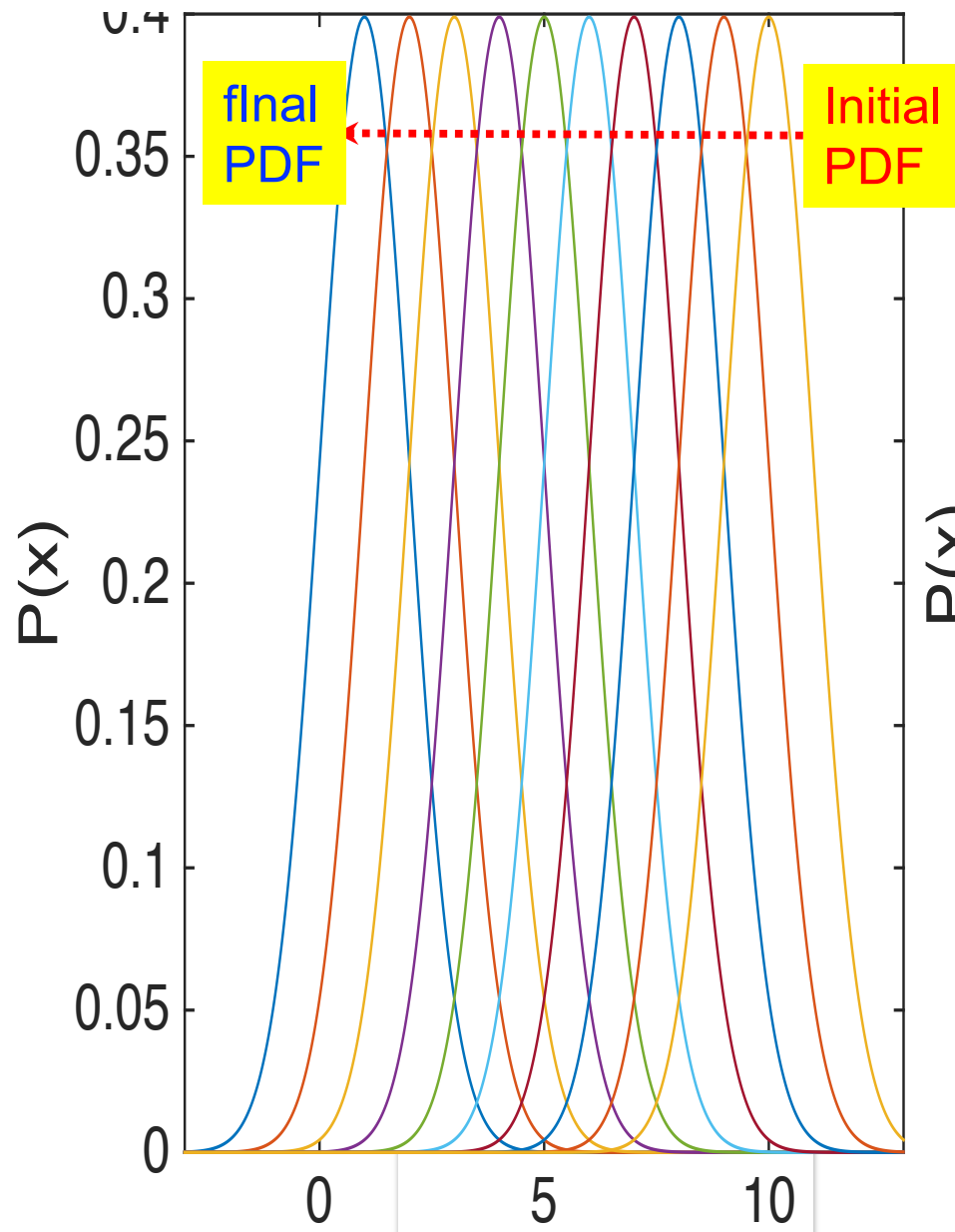
For different σ



Control experiment: Reducing x by constant σ



Time dependent problem: $\mu(t)$ and $\sigma(t)$



Time-dependent PDF

[Kim 18; Nicholson & Kim 15,16; Heseltine & Kim 16; Kim & Hollerbach 20]

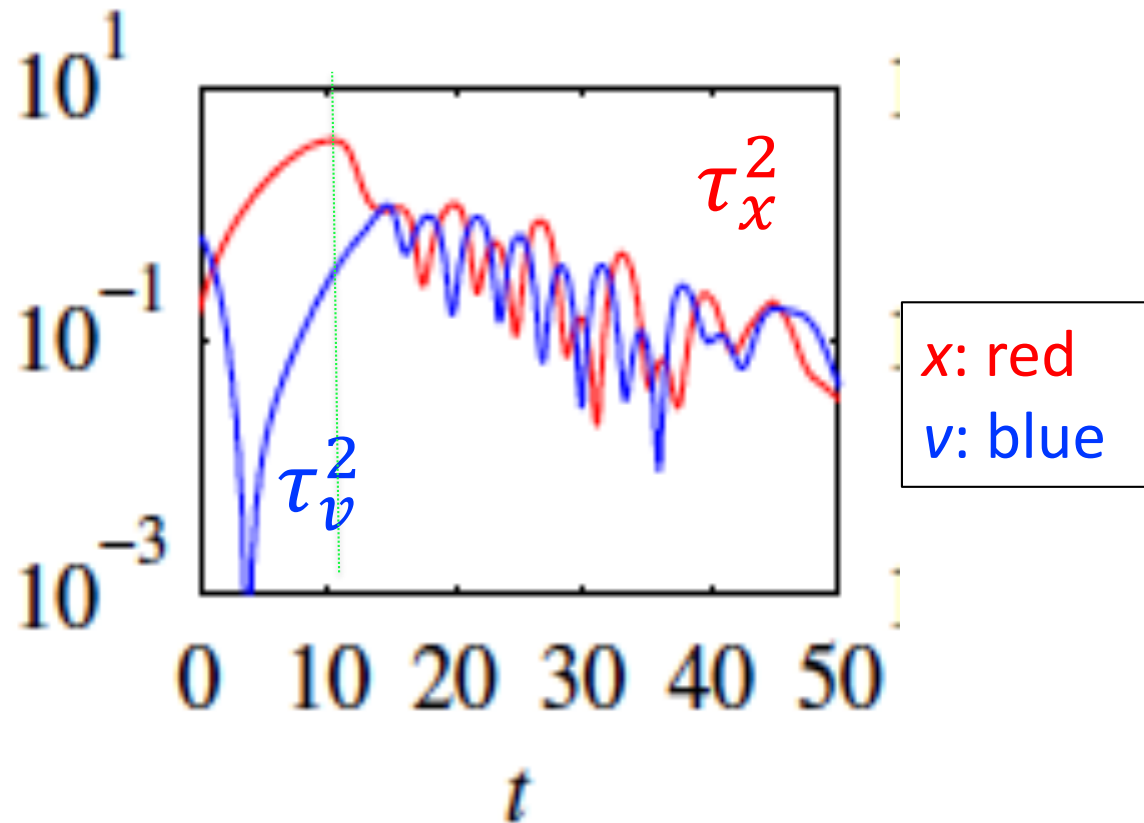
$$\left(\frac{dL}{dt}\right)^2 = \frac{1}{\tau^2(t)} = \int dx \, p(x,t) \left(\frac{\partial \ln p(x,t)}{\partial t}\right)^2$$

rate of information change: $\tau^{-1} = \left|\frac{dL}{dt}\right|$

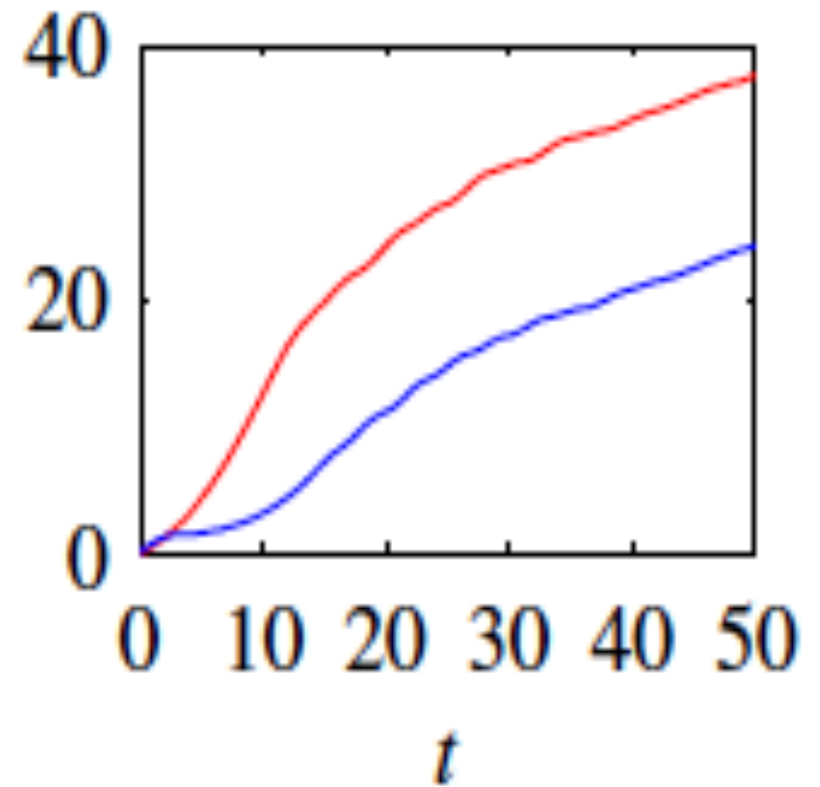
Information length $L(t)$: dimensionless total number of different statistical states that a system evolves through in time $(0,t)$

$$L(t) = \int_0^t dL = \int_0^t \frac{dt_1}{\tau(t_1)} = \int_0^t dt_1 \sqrt{\int dx \, p(x,t_1) \left(\frac{\partial \ln p(x,t_1)}{\partial t_1}\right)^2}$$

$1/\tau^2$ vs time



Information length vs time



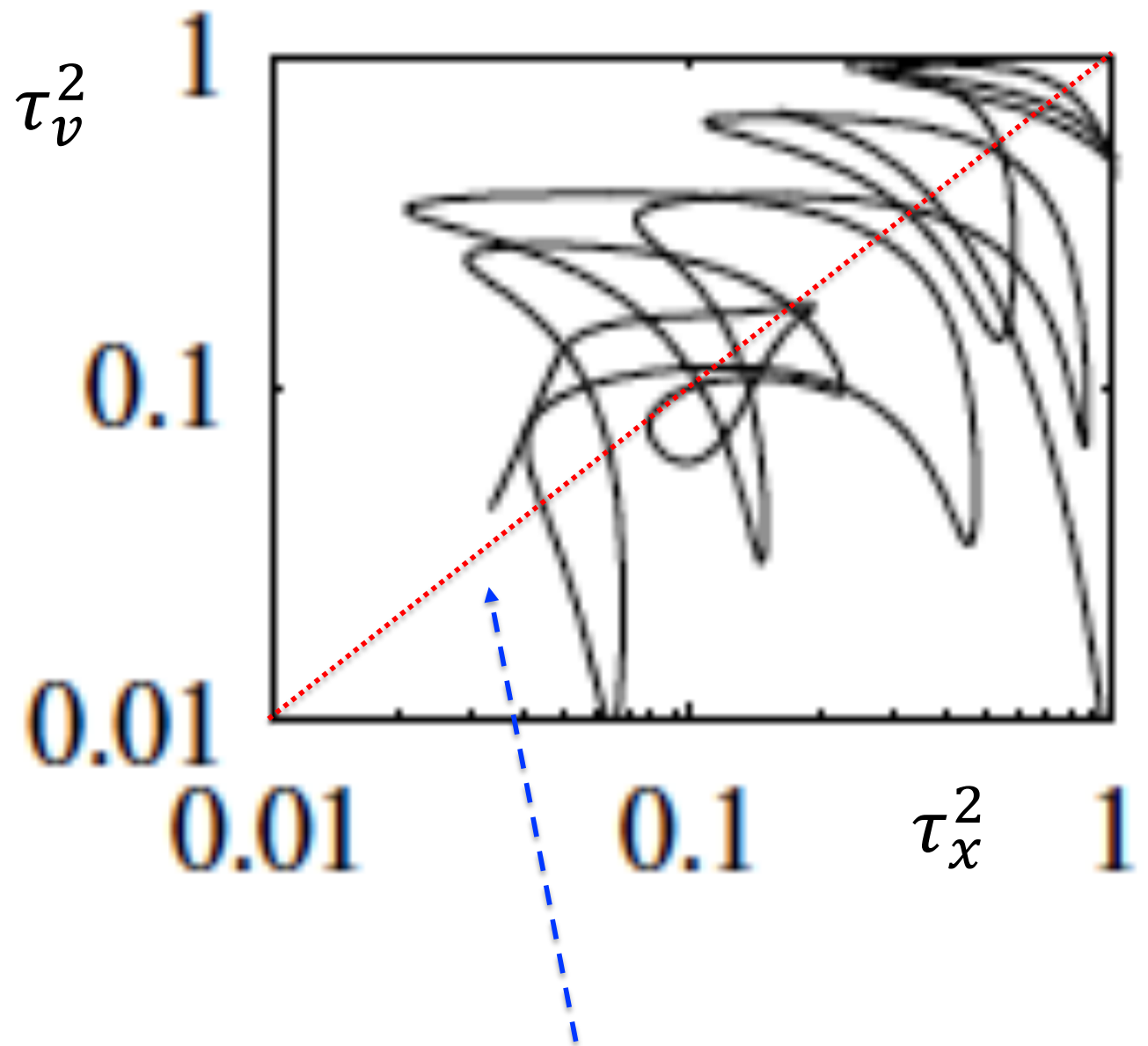
$$\frac{1}{\tau_x(t)^2} = \int_0^t dt' \frac{1}{p(x,t')} \left[\frac{\partial p(x,t')}{\partial t} \right]^2$$

$$L_x = \int_0^t dt' \frac{1}{\tau_x(t')}$$

$$\frac{1}{\tau_v(t)^2} = \int_0^t dt' \frac{1}{p(v,t')} \left[\frac{\partial p(v,t')}{\partial t} \right]^2$$

$$L_v = \int_0^t dt' \frac{1}{\tau_v(t')}$$

Information
plane:
 τ_x^2 vs τ_v^2



Self-organisation: Oscillation around $\tau_x = \tau_y$

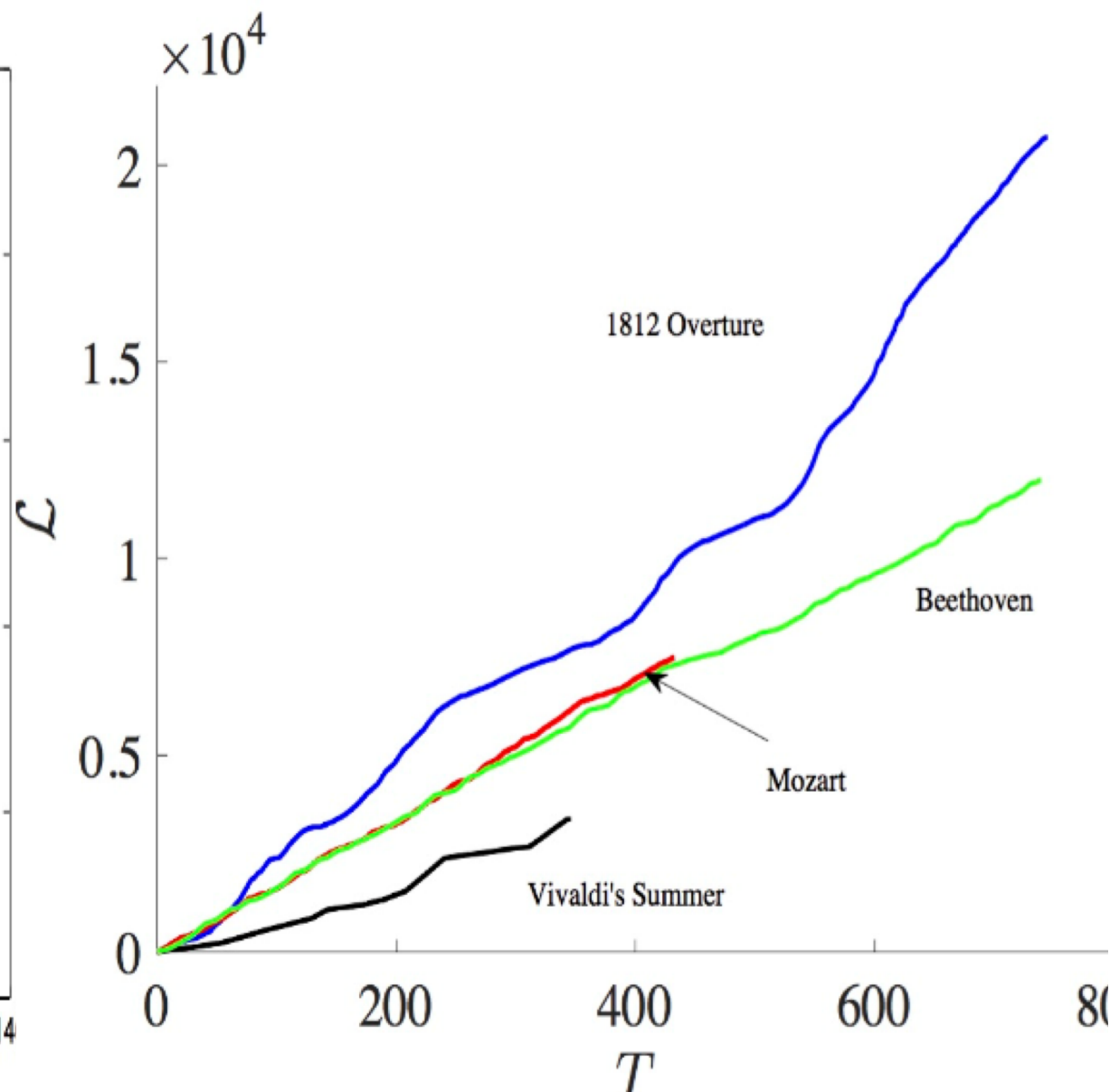
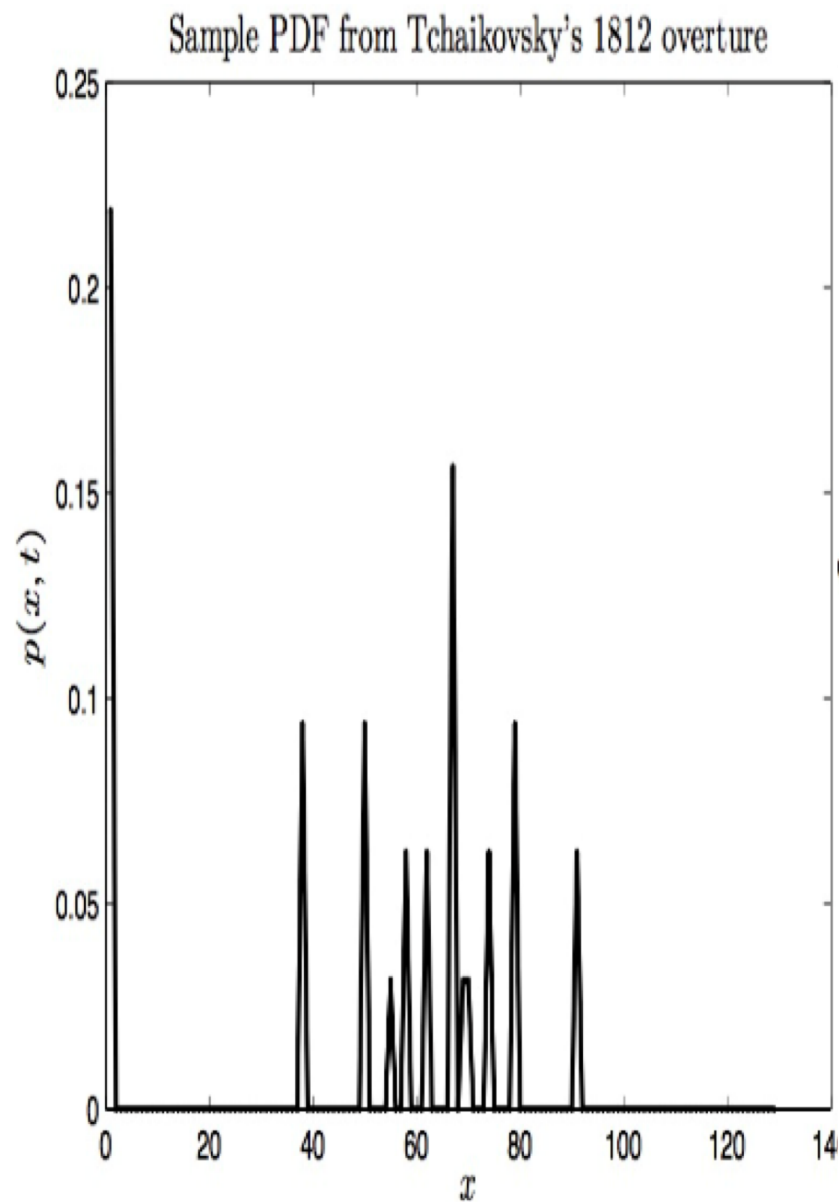
3. Other examples:

3.1 Music

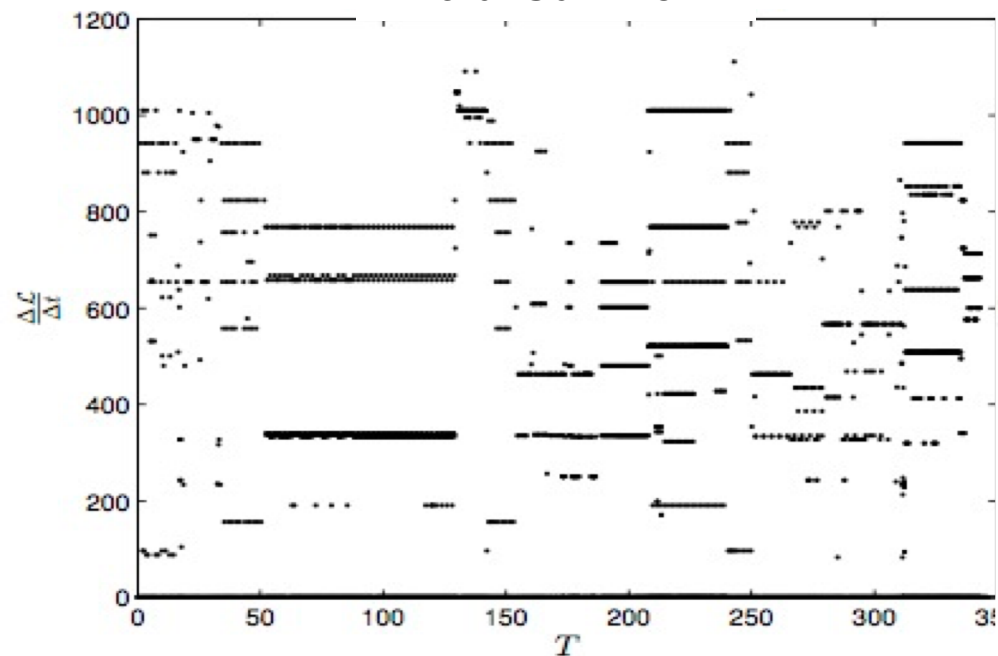
3.2 Global circulations

3.1 Music: can we see music?

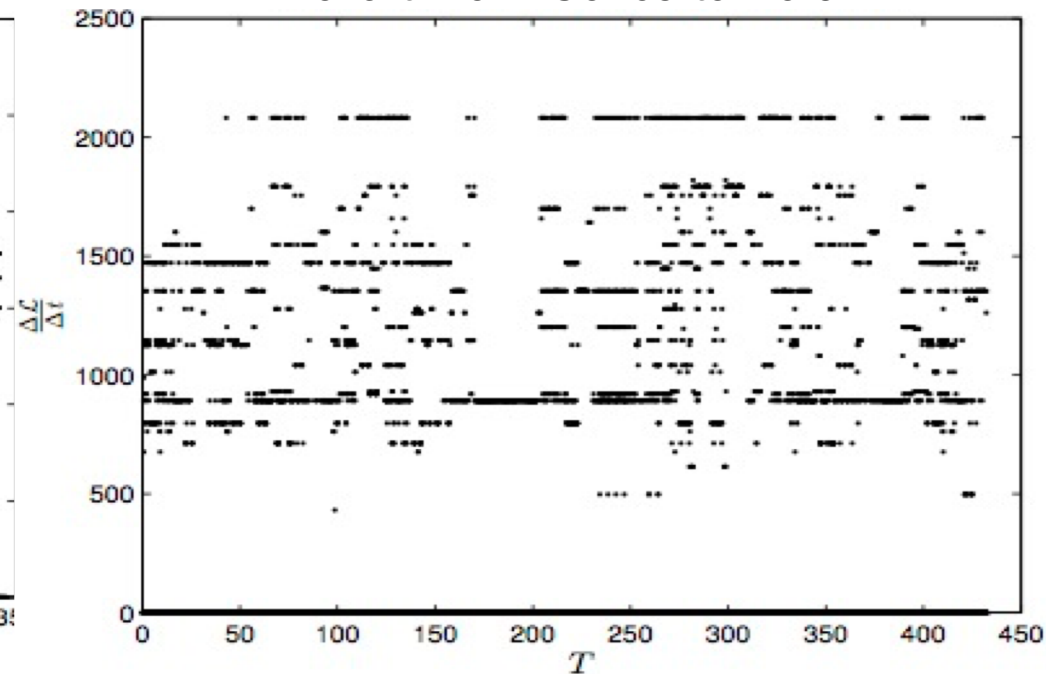
[Nicholson & Kim Entropy 2016]



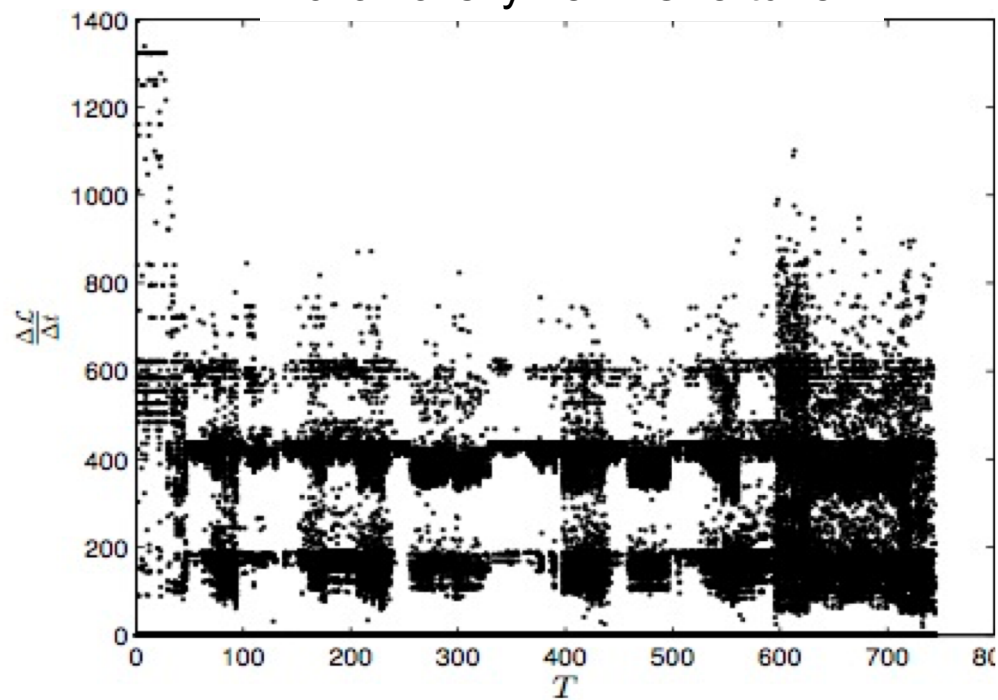
Vivaldi Summer



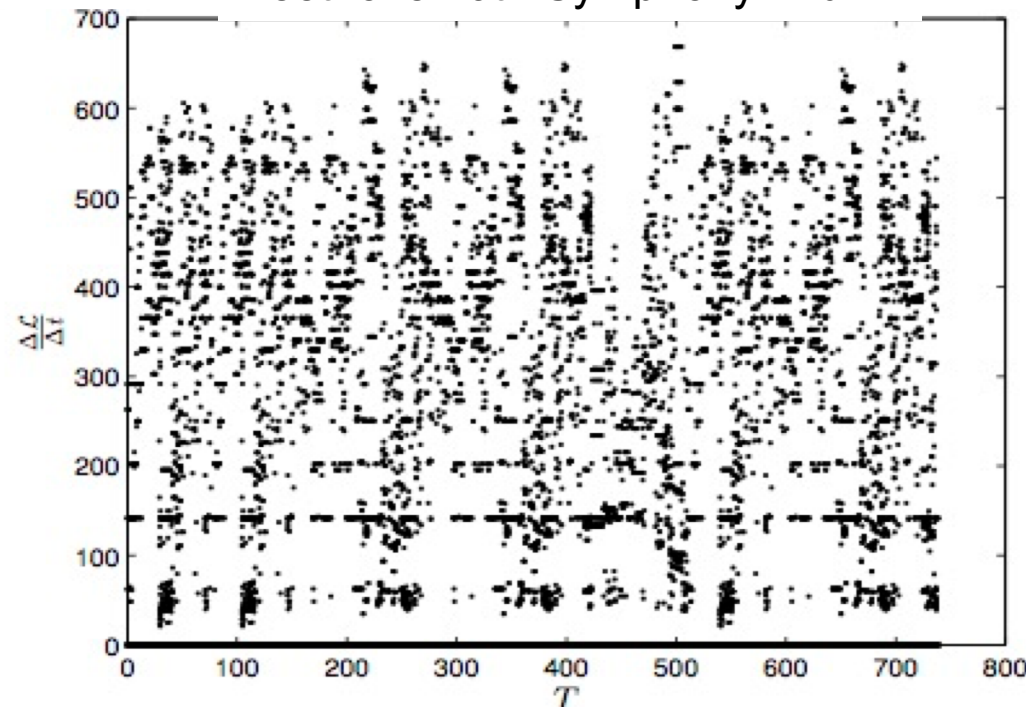
Mozart Violin Concerto No 3



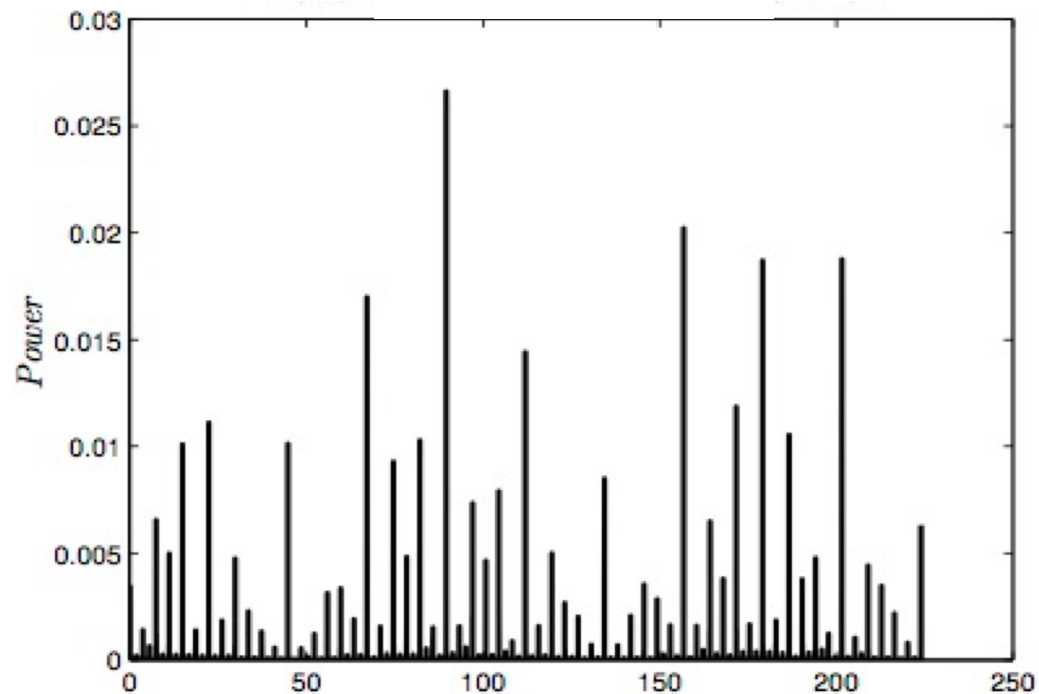
Tchaikovsky 1812 Overture



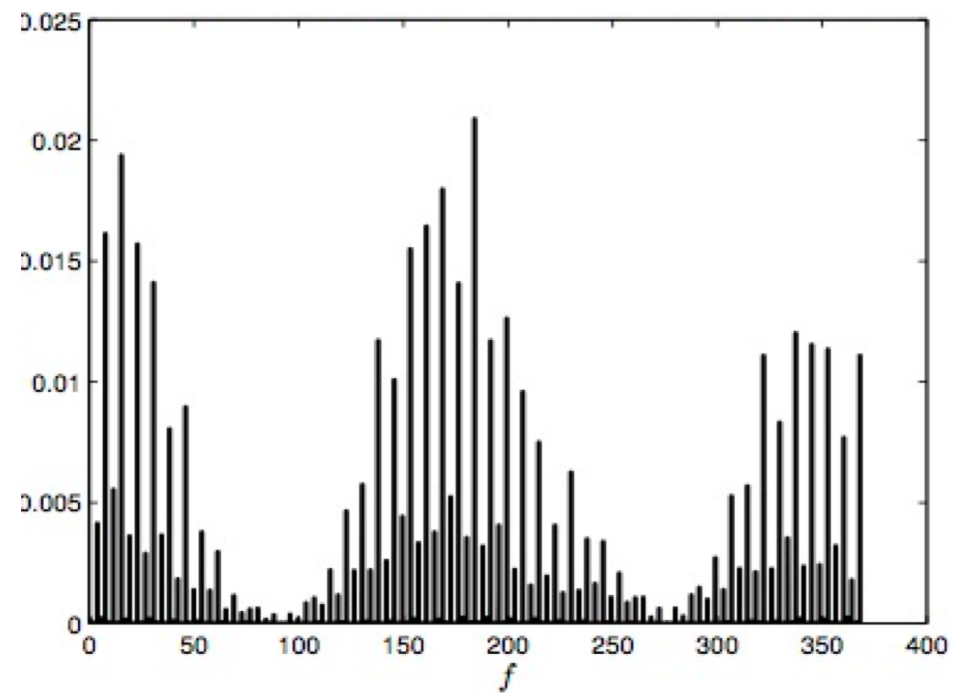
Beethoven 9th Symphony 2nd



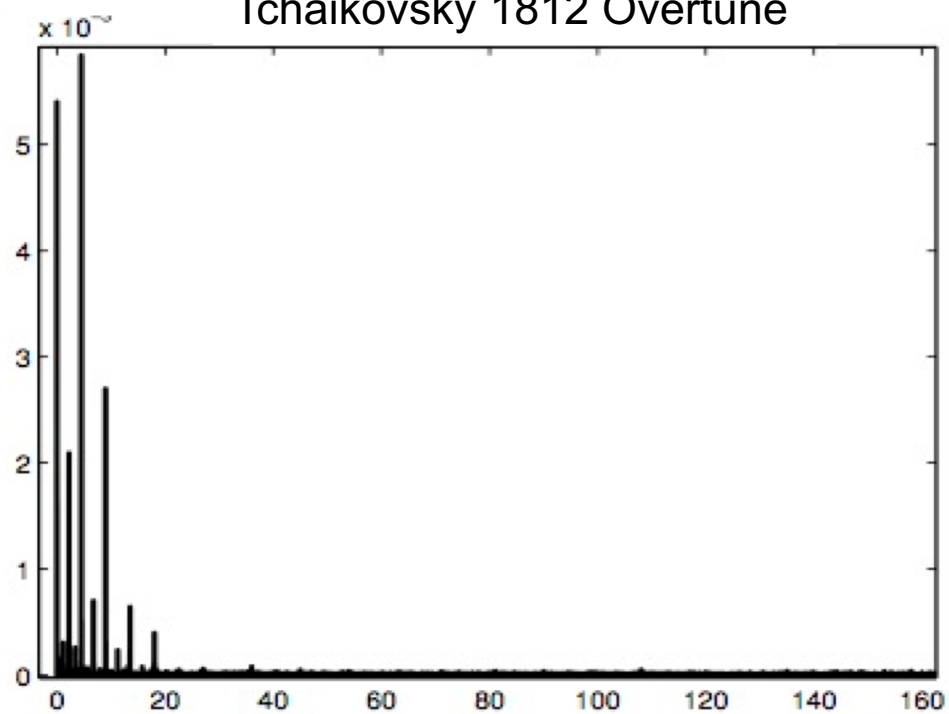
Vivaldi Summer



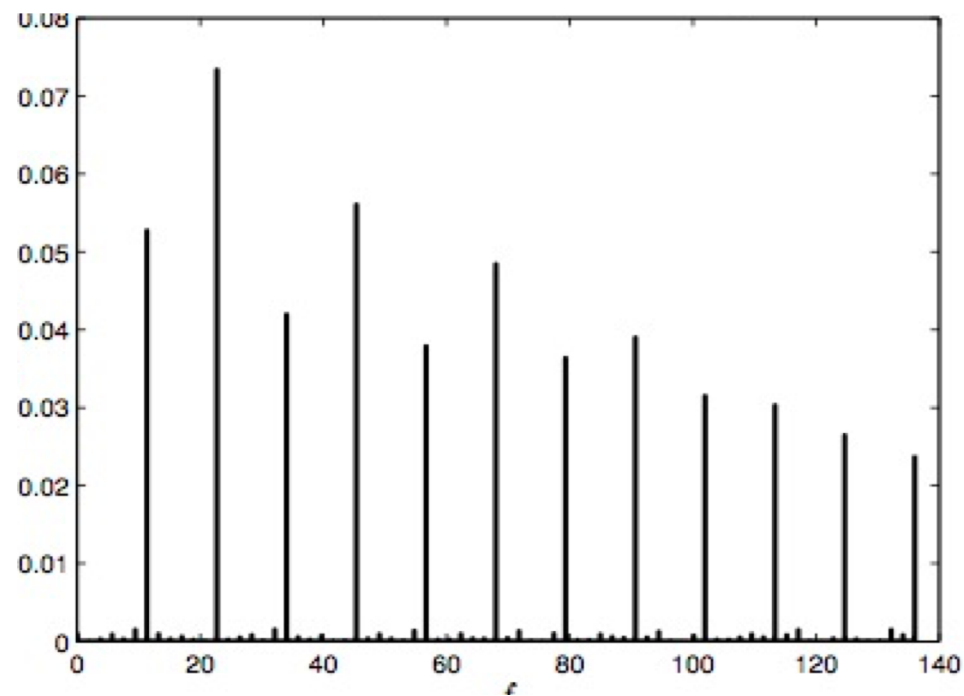
Mozart Violin Concerto No 3



Tchaikovsky 1812 Overture

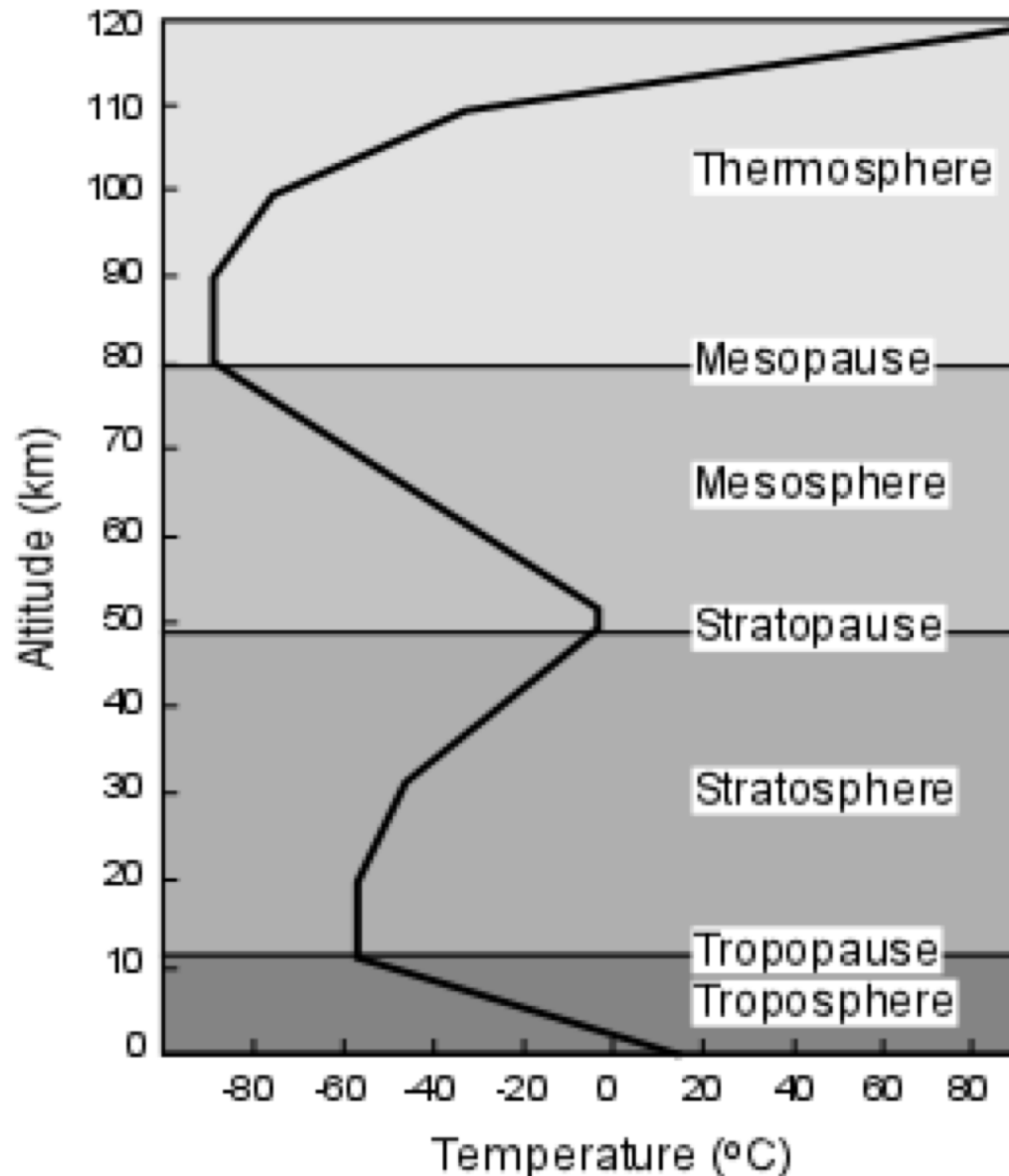


Beethoven 9th Symphony 2nd



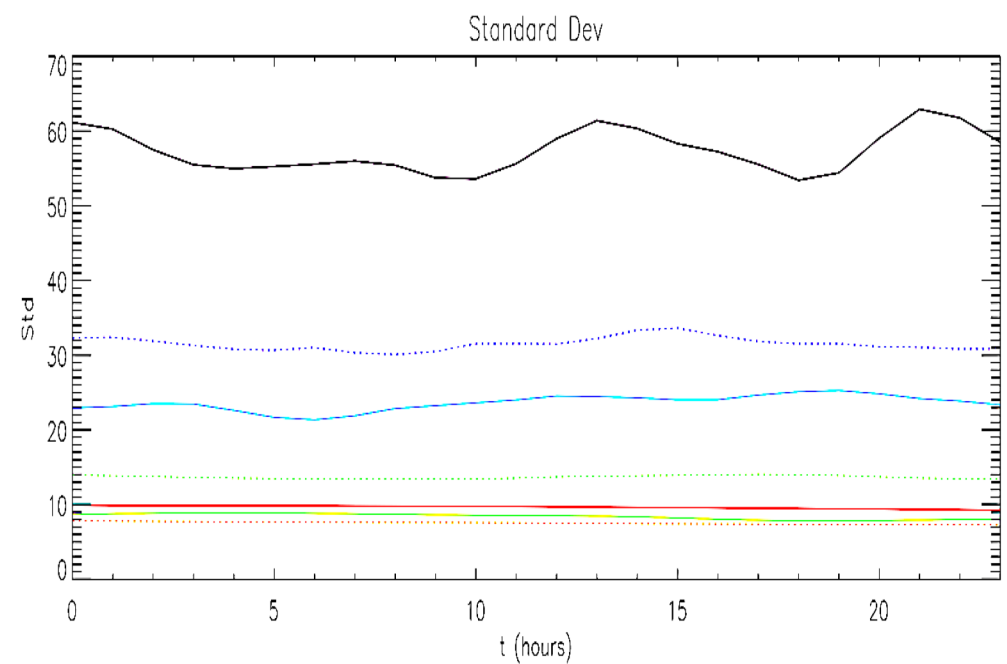
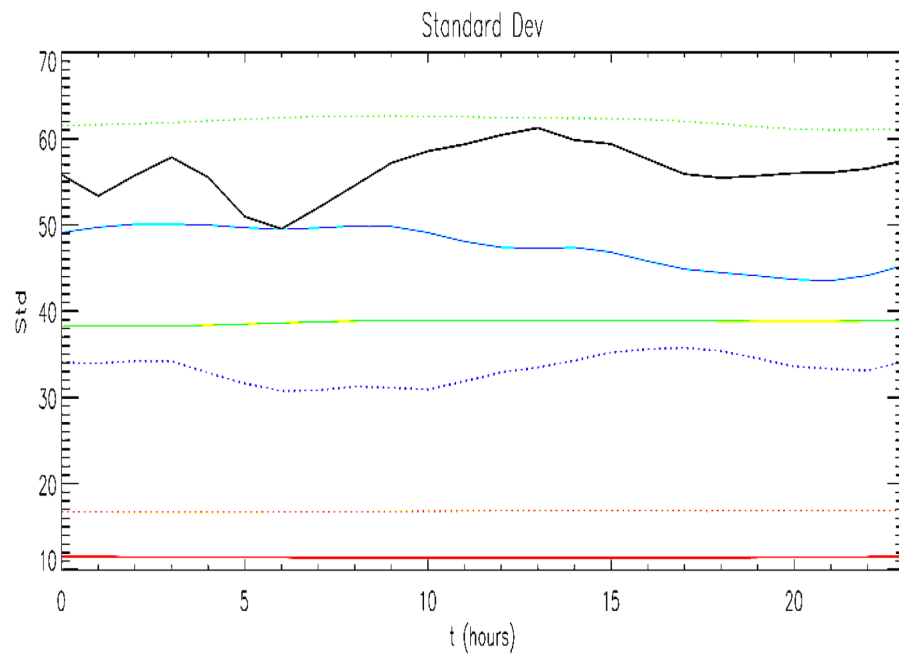
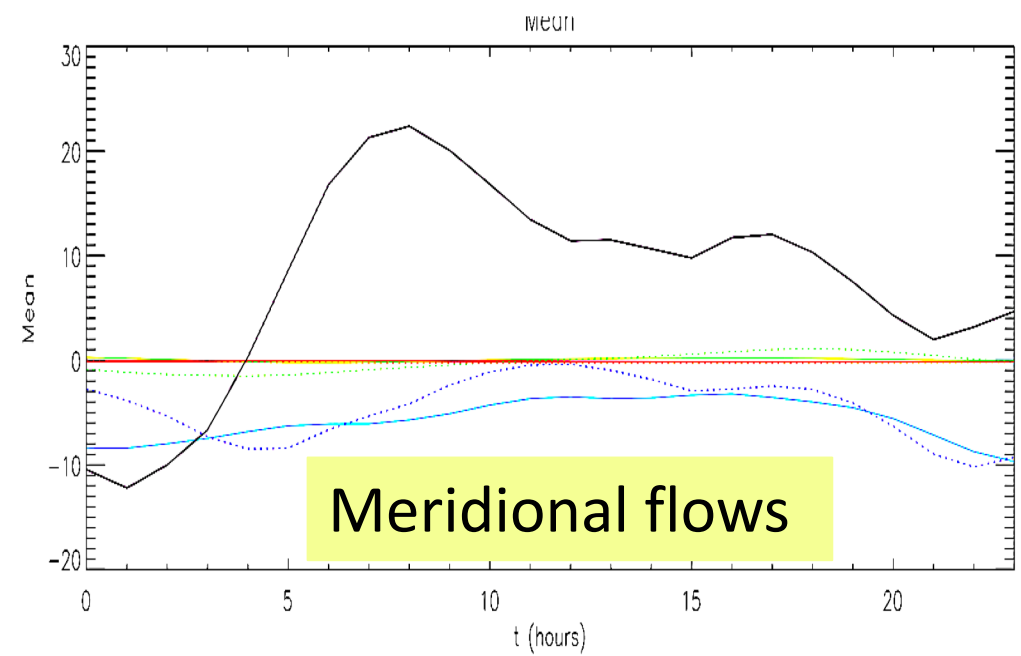
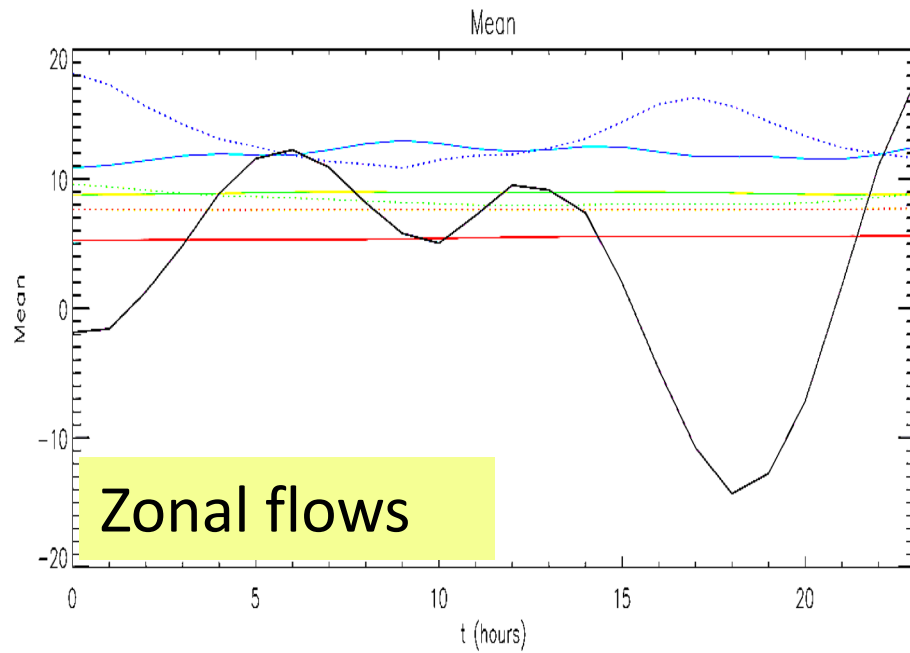
3.2 Global circulation model

[Kim, Liu & Heseltine 2020]

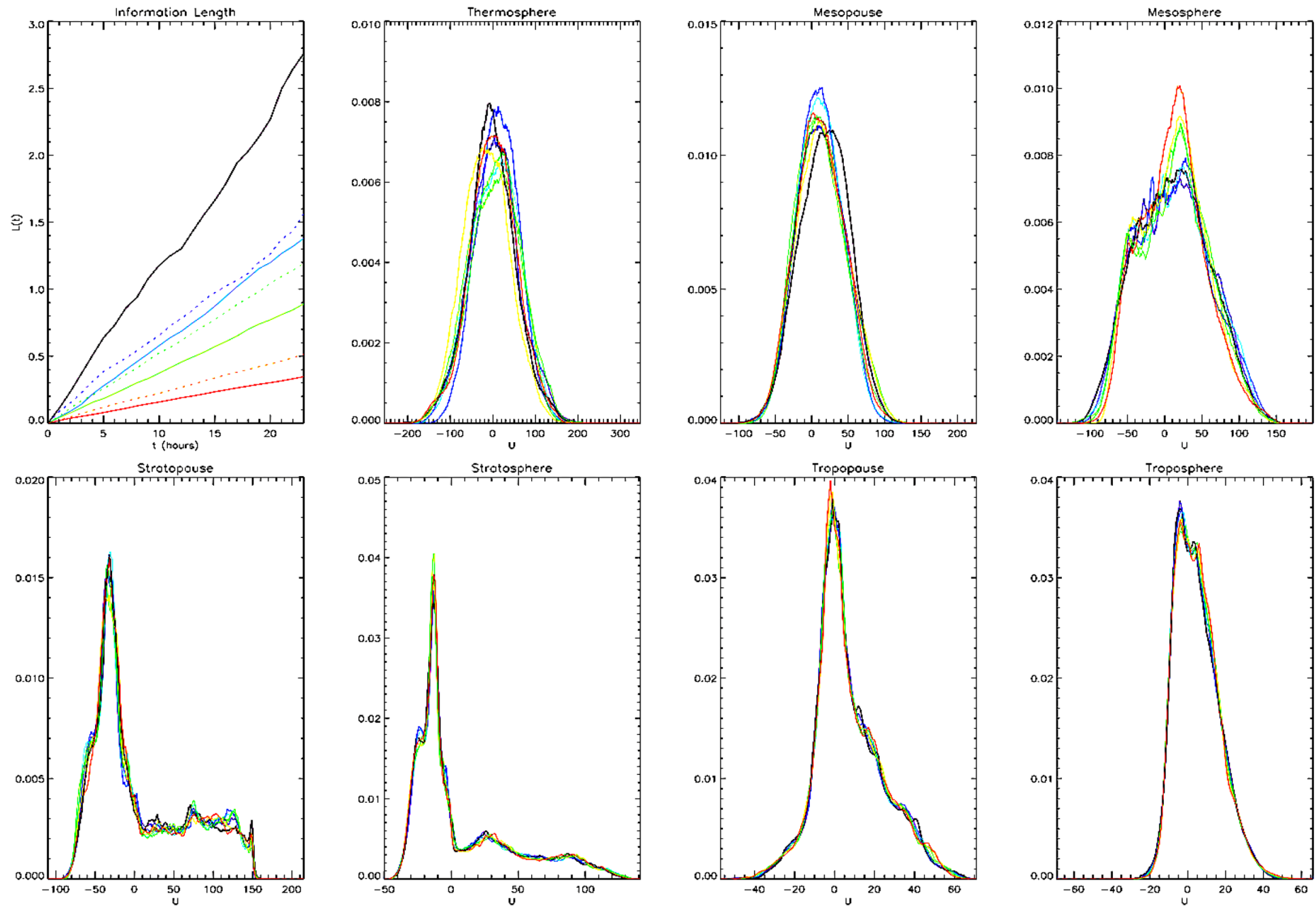


- Data from WACCM
- A few points around the middle of each sphere and pause
- Time dependent PDF from using data at these points, all longitude and latitude
- Information length $L(t)$

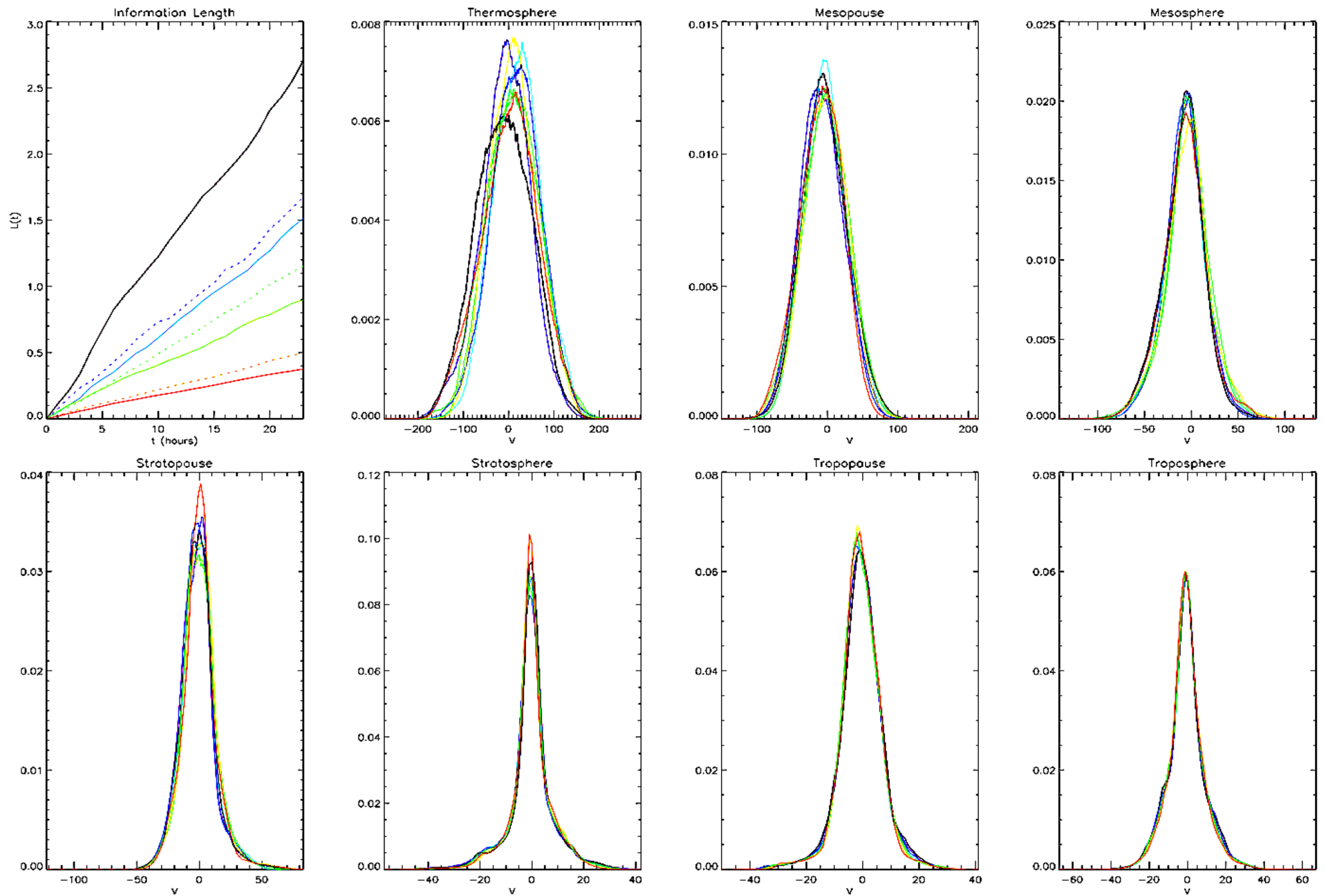
Mean and Standard deviation: zonal & meridional flows



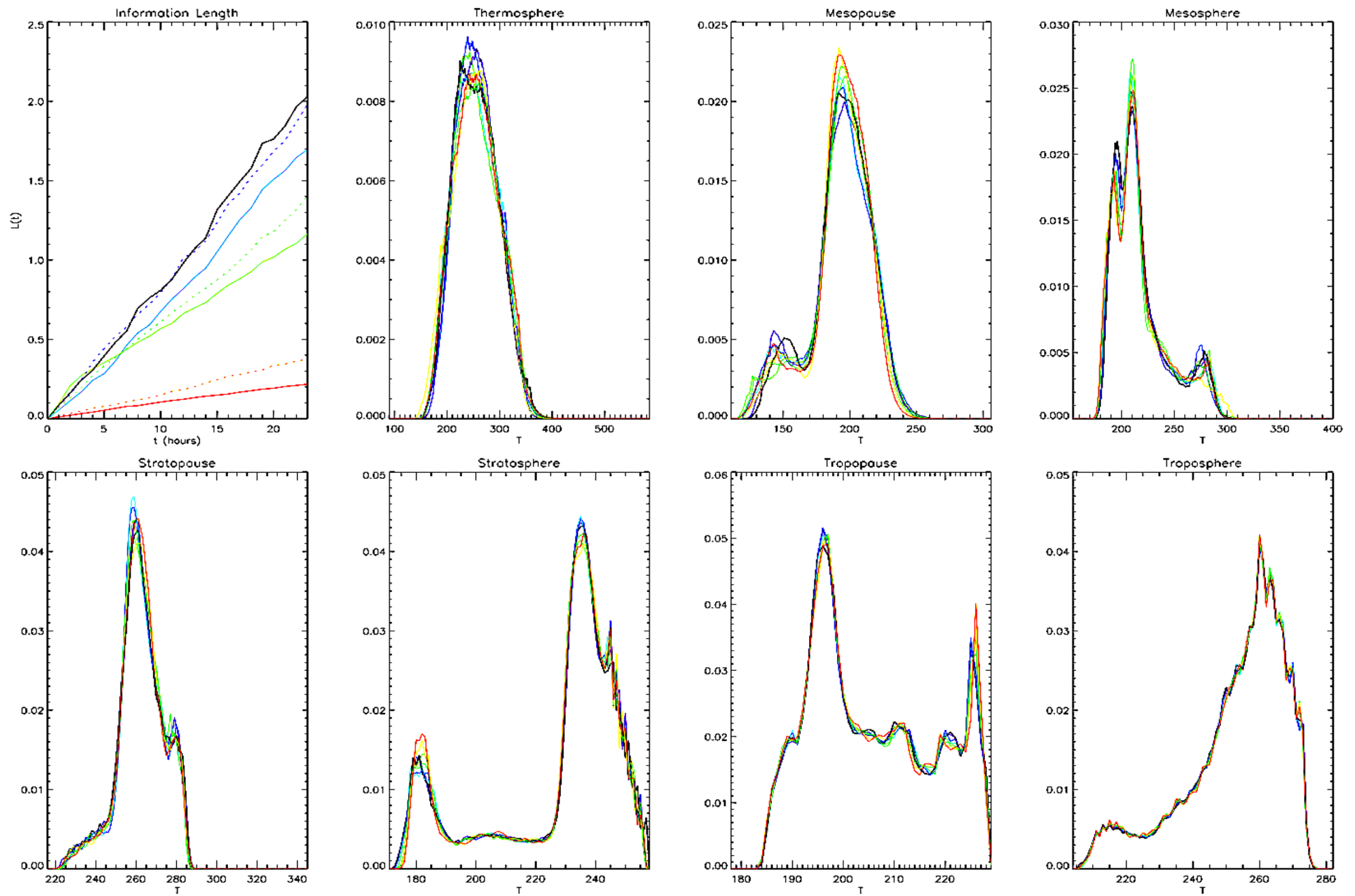
Information length and PDFs: Zonal flows

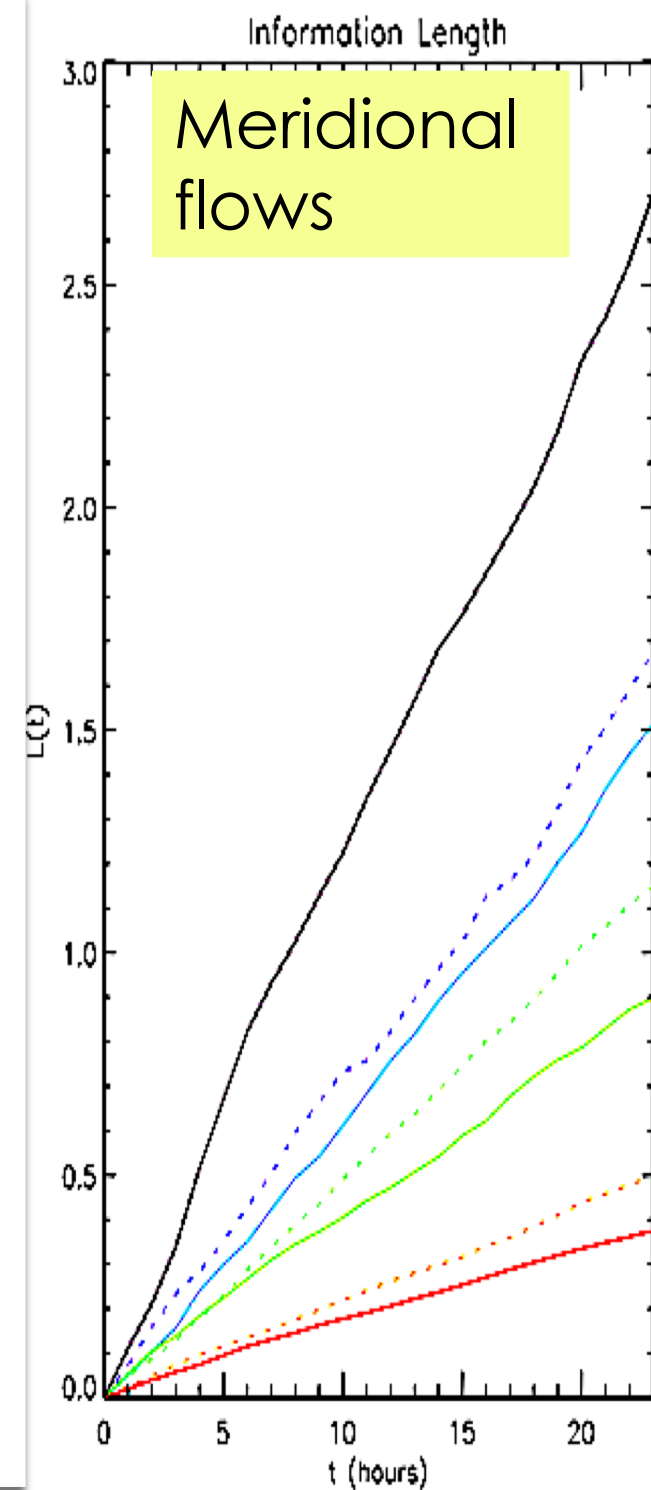
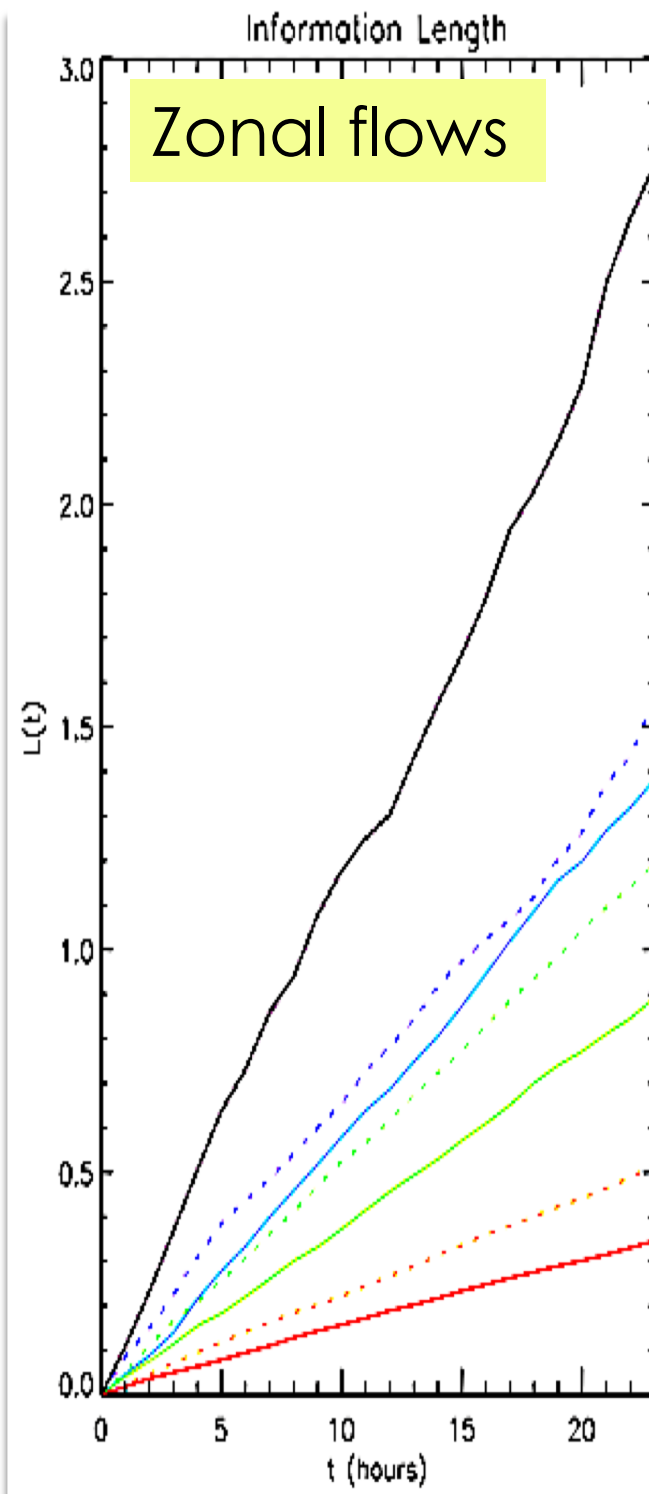
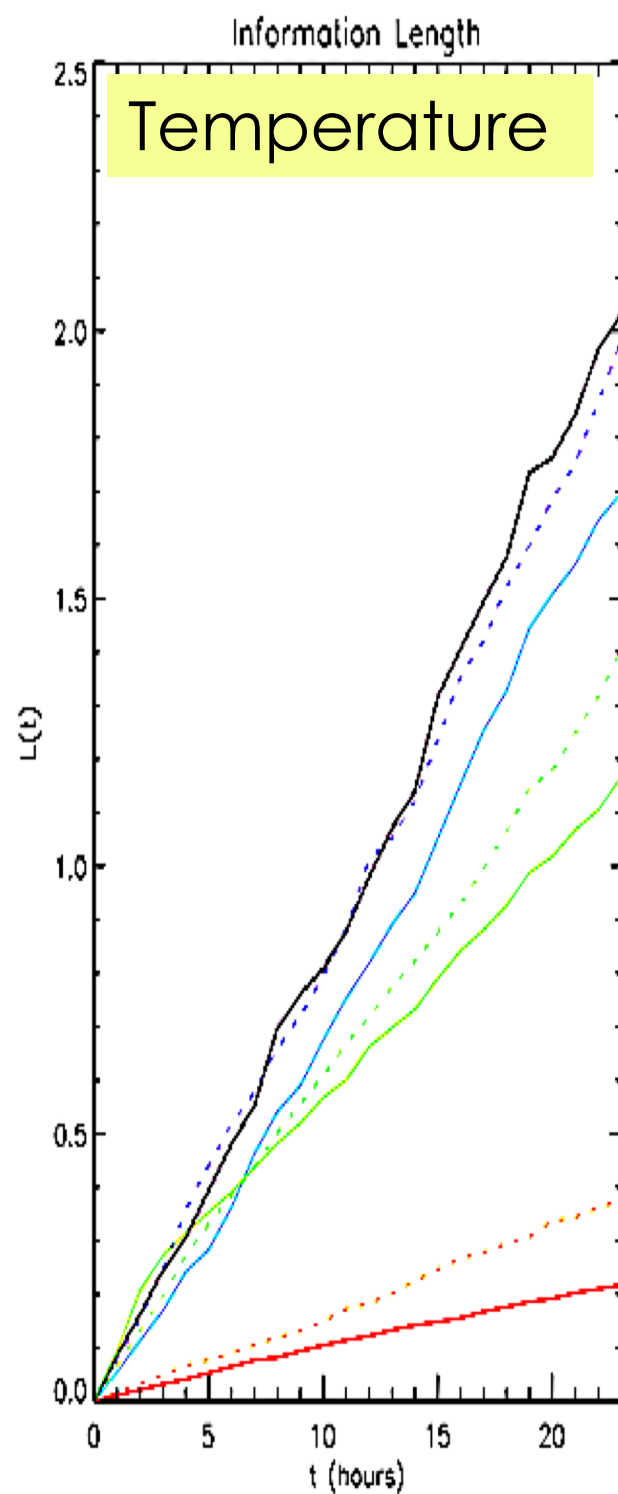


Information length and PDFs: meridional flows



Information length and PDFs: Temperature





4. Conclusions

- Much scope for research on the L-H transition
- Limitation of mean value, standard deviation, Gaussian PDF
- Information length: the number of statistically different states that a system evolves through in time.
- It is dimensionless and invariant under (time-independent) change of variable.
- Useful to understand correlation in self-organising process.
- Applicability to different processes.
- Useful index to classify a growing number of data.

Thank you

- **Supported Summer Internships at Coventry**
- **Funded PhD studentship on turbulent plasmas**

<https://www.findaphd.com/phds/project/turbulent-plasma-in-laboratory-and-space/?p118985>

or

<https://warwick.ac.uk/fac/sci/physics/prospective/postgraduate/pgintro/resourcesforapplicants>

Contact: Prof Eun-jin Kim at ejk92122@gmail.com

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