

# Introduction to Data Assimilation, Subgrid-Scale Parameterization and Predictability

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# Outline for Today

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## Linear Stability and Lyapunov exponents



# Lyapunov exponents

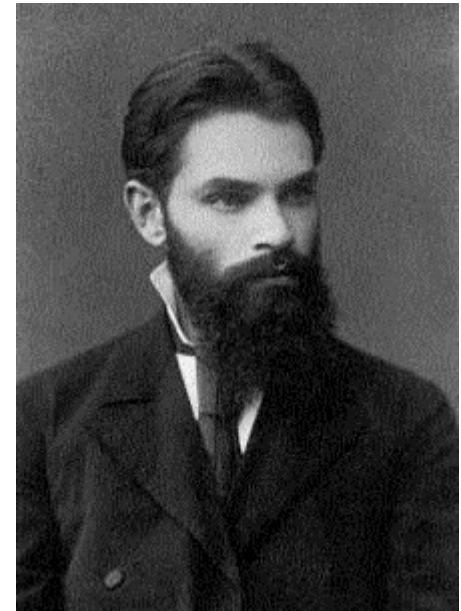
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## Linear versus non-linear dynamical systems

Linear System:

Evolution of basic solutions

→ time independent linear combinations  
span solution space



Aleksandr Mikhailovich Lyapunov  
(1857 – 1918)

# Lyapunov exponents

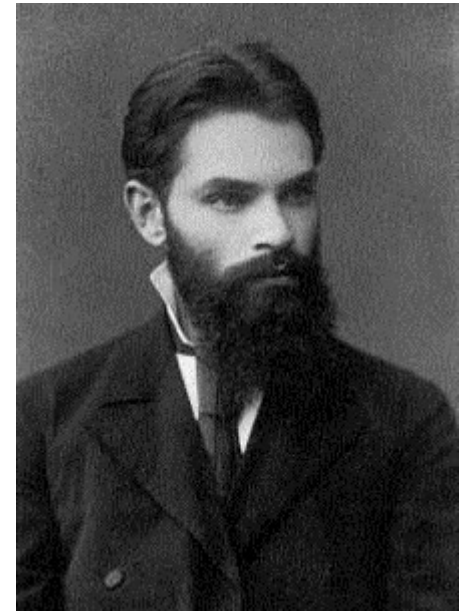
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## Linear versus non-linear dynamical system

Non-Linear System:

No obvious basic solutions

→ there are no time independent linear combinations that span solution space



Aleksandr Mikhailovich Lyapunov  
(1857 – 1918)

# Lyapunov exponents

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**What is the property that makes these systems so different?**



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Non-Linear Systems are very often chaotic



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# Lyapunov exponents

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**What is the property that makes these systems so different?**

Non-Linear Systems are very often chaotic

**What is a chaotic dynamical system?**

Exponential divergence for small perturbations



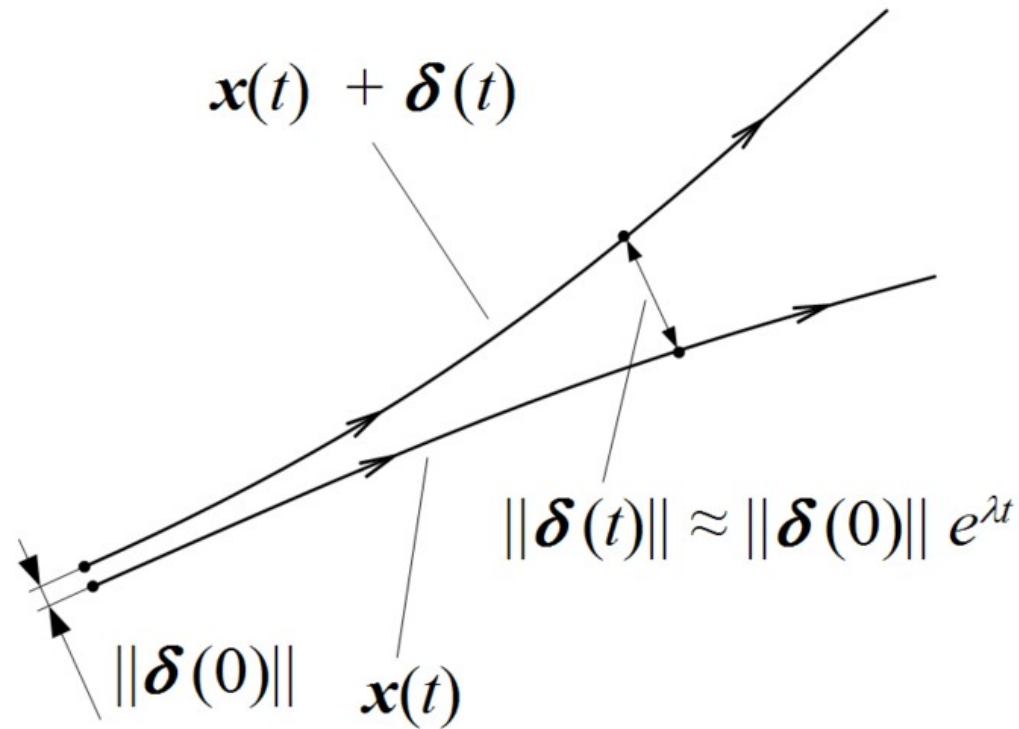
# Lyapunov exponents

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- $x(t)$  evolves freely
- The perturbation grows exponentially
- If the perturbation is rescaled at regular intervals, we get the maximum Lyapunov exponent
- In a high dimensional system there are many directions for growth
- If we repeat this procedure iteratively in orthogonal space we get successively all Lyapunov exponents (Lyapunov spectrum)

# Lyapunov exponents

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# What has this to do with meteorology?

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- Classical Lifecycle of Cyclones (Bjerknes & Solberg 1922)
- Small perturbations create variability
- What kind of perturbations are amplifying and which are decaying?

# What has this to do with meteorology?

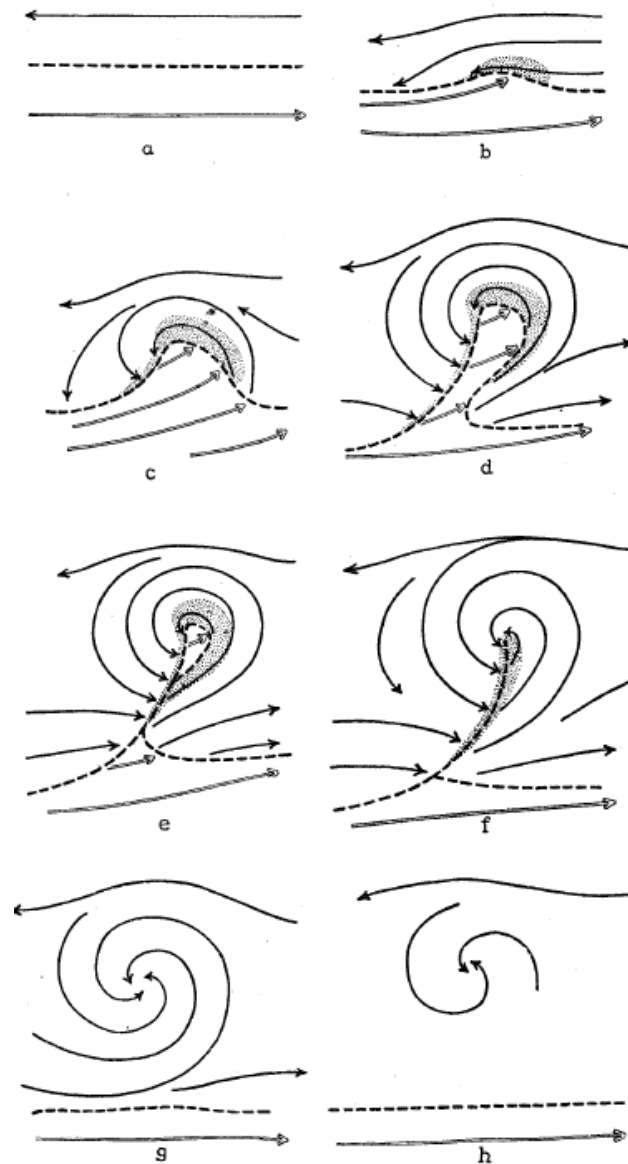


Fig. 2. The «Life cycle» of cyclones.

# What has this to do with meteorology?

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## Formal Decomposition

- Decompose atmospheric field  $x$  into eddy field and zonal mean

## Classification of solutions

- Stationary balance
- Turbulent balance
- Climate Mean



# “Classical” Linear Stability Analysis

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## Success in classical linear theory with normal modes

(e.g. Charney 1947, Eady 1949, Pedlosky 1964)

Fast growing **linear instabilities** (normal modes) of stationary/mean states qualitatively explain

- **energy conversions** between  $x^E$  and  $[x]$  (**baroclinic** and **barotropic** conversions)
- **heat and momentum transports** of  $x^E$  → **Lorenz Energy Cycle**
- most important **spatial scales** and **life times** of mid latitudes cyclones
- Compare **normal modes** vs **empirical orthogonal functions**



## Limitations of Classical Approach

- Atmosphere is chaotic and turbulent
  - Cyclones/eddies develop not on stationary state
- Air Parcels are not simply perturbed by themselves (they have collective oscillations)
- Variability is at least initially created by linear instabilities of the turbulent flow

# Classical Stability analysis vs Lyapunov exponents

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## TODAY

- **Lyapunov exponents (LEs)**: the average growth rates of linear perturbations

## New & Extended Approach

- Covariant Lyapunov Vectors (CLVs): **actual perturbations** which grow with the LEs





# Lyapunov exponents – Definiton

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## Full Solution of dynamical system

$$\frac{\partial}{\partial t} u_A = f(u_A)$$

## Perturbed Solution

$$\frac{\partial}{\partial t} u_B = f(u_B), \quad u_B = u_A + v$$

# Lyapunov exponents – Oseledec's Theorem

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## Essential Content of the Theorem

- $n$  solutions to  $n$  dimensional tangent linear equation ( $n$  vectors dependent on time)
- Each solution grows exponentially  $\sim \exp\left(\int_{t_0}^t \lambda(\tau) d\tau\right)$
- The mean of  $\lambda(t)$  are the Lyapunov exponents

We for now are only interested in rate of separation, hence the Lyapunov exponents

# Lyapunov exponents – Computational Algorithm

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## Continuous Equation

$$\frac{\partial}{\partial t} \mathbf{v}(t) = \frac{\partial F}{\partial u}(u_A) \mathbf{v}(t)$$

## Discretized Version

$$\mathbf{v}_{n+1} = \mathcal{F}_n \mathbf{v}_n = ((\mathbb{1} + J(x_n)) \mathbf{v}_n)$$

Bennetin et al. 1996 algorithm

$$\mathcal{F}_n \cdot Q_n = V_{n+1} = \underbrace{Q_{n+1} R_{n+1}}_{\text{QR decomposition}}$$

**The diagonal of the R matrix contains  
 $\exp(\lambda_n * \Delta t)$**

# Tangent Linear Model

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$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{x}) \quad \mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad \mathbf{F} = \begin{bmatrix} F_1 \\ \vdots \\ F_n \end{bmatrix}$$

$$\mathbf{x}^{n+1} = \mathbf{x}^n + \Delta t \mathbf{F} \left( \frac{\mathbf{x}^n + \mathbf{x}^{n+1}}{2} \right)$$

Nonlinear model that only depends on initial conditions

$$\mathbf{x}(t) = M[\mathbf{x}(t_0)]$$

# Tangent Linear Model

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Add small perturbation

$$\begin{aligned} M[\mathbf{x}(t_0) + \mathbf{y}(t_0)] &= M[\mathbf{x}(t_0)] + \frac{\partial M}{\partial \mathbf{x}} \mathbf{y}(t_0) + O[\mathbf{y}(t_0)^2] \\ &= \mathbf{x}(t) + \mathbf{y}(t) + O[\mathbf{y}(t_0)^2] \end{aligned}$$

Evolution model of small perturbation

$$\frac{d\mathbf{y}}{dt} = \mathbf{J}\mathbf{y}$$

where  $\mathbf{J} = \partial \mathbf{F} / \partial \mathbf{x}$  is the Jacobian of  $\mathbf{F}$ .

# Tangent Linear Model

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$$\frac{d\mathbf{y}}{dt} = \mathbf{J}\mathbf{y}$$

This system of linear ordinary differential equations is the tangent linear model in differential form. Its solution between  $t_0$  and  $t$  can be obtained by integrating (6.3.5) in time using the same time difference scheme used in the nonlinear model (6.3.3):

$$\mathbf{y}(t) = \mathbf{L}(t_0, t)\mathbf{y}(t_0) \tag{6.3.6}$$

# Exercise

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Compute largest Lyapunov exponent:

- 1) Start with initial condition on attractor
- 2) Select nearby point (separated by  $d_0 \sim 10^{-9}$ )
- 3) Advance both orbits by one time step and calculate new separation  $d_1$ .
- 4) Plot separations over a few time steps
- 5) Compute linear regression slope
- 6) Average also over many different initial conditions.

