



# Breeding approach for predictability

## CPTEC-INPE Global Model

Luis Fernando Salgueiro Romero\*    Sandra A. Sandri\*\*

Haroldo F. de Campos Velho\*\*

(\*) M.Sc. (CAP-INPE)    (\*\*\*) LAC-INPE

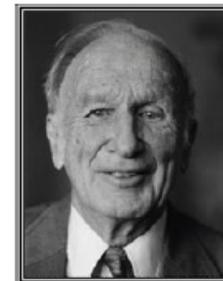
# Summary

- Predictability
- Low order systems
- Bred vector and predictability
- Global model: CPTEC-INPE
- Predictability: Ensemble prediction vs Bred Vector
- ~~Neuro-fuzzy: machine learning for inference~~ (not shown)
- Final Remarks

# Predictability

## History

- Edward N. Lorenz



Reprinted from Transactions of The New York Academy of Sciences  
Ser. II, Volume 25, No. 4, Pages 409-432  
February 1963

**THE PREDICTABILITY OF HYDRODYNAMIC FLOW\*†**

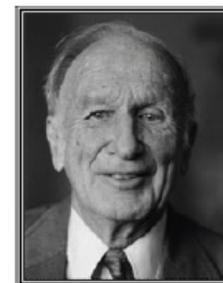
**Edward N. Lorenz**

*Massachusetts Institute of Technology, Cambridge, Mass.*

# Predictability

## History

- Edward N. Lorenz



A study of the predictability of a 28-variable  
 atmospheric model

By EDWARD N. LORENZ, *Det Norske Meteorologiske Institutt*<sup>1,2</sup>

(Manuscript received December 22; revised version February 1)



# Predictability



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## Abstract View

[Volume 20, Issue 2 \(March 1963\)](#)

**Journal of the Atmospheric Sciences**

Article: pp. 130–141 | [Abstract](#) | [PDF \(1.02M\)](#)

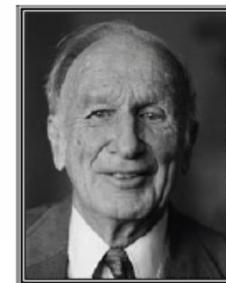
## Deterministic Nonperiodic Flow

**Edward N. Lorenz**

*Massachusetts Institute of Technology*

(Manuscript received November 18, 1962, in final form January 7, 1963)

DOI: 10.1175/1520-0469(1963)020<0130:DNF>2.0.CO;2



# Predictability

## Lorenz's attractor



**Edward N. Lorenz**

Simplest chaotic system:

the Lorenz Attractor.

A metaphor for weather  
unpredictability

$$\frac{dx}{dt} = \sigma(y - x)$$

$$\frac{dy}{dt} = x(\rho - z) - y$$

$$\frac{dz}{dt} = xy - \beta z$$

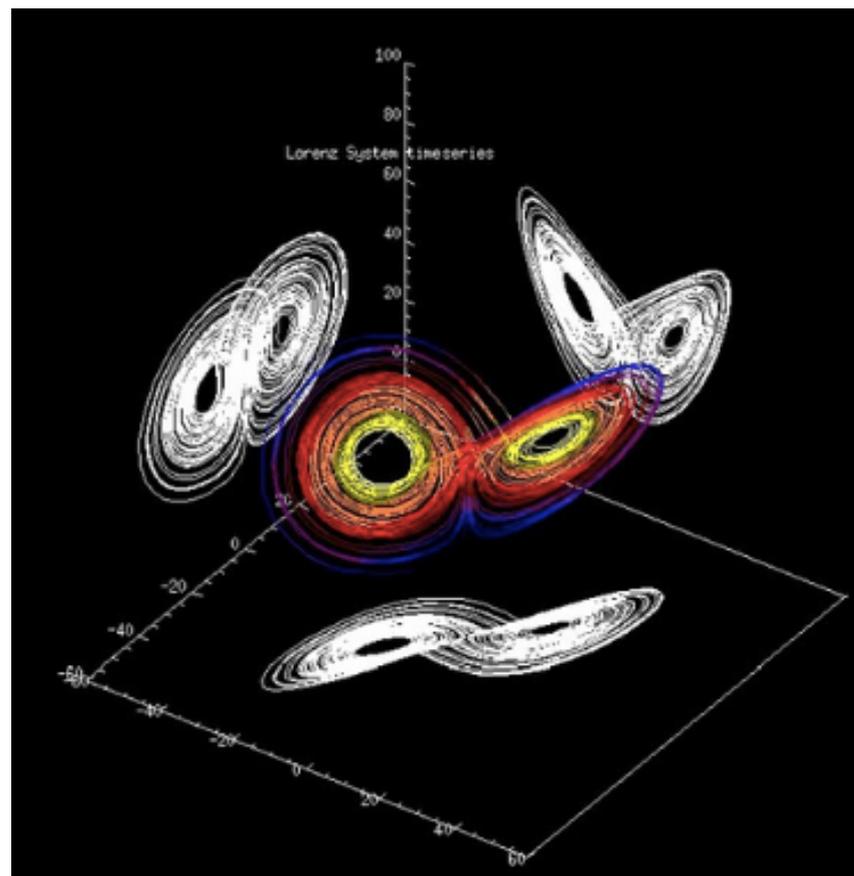
# Predictability

## Lorenz's attractor

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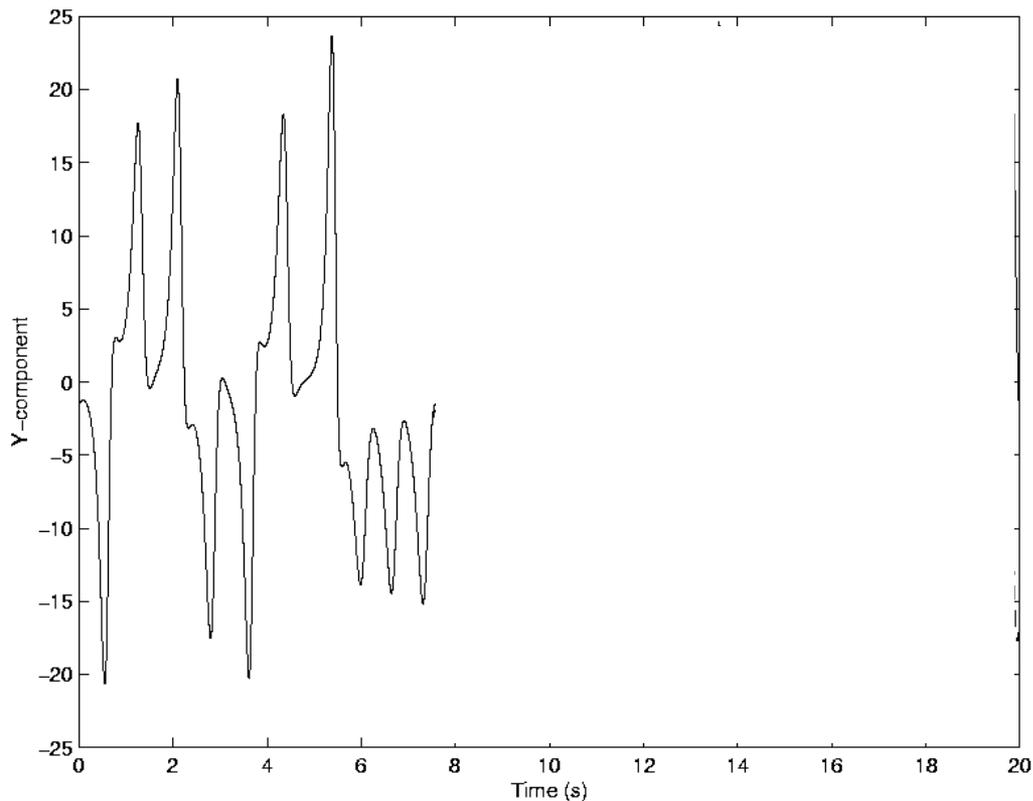
$$\frac{dz}{dt} = xy - \beta z$$



# Predictability

## Lorenz's system

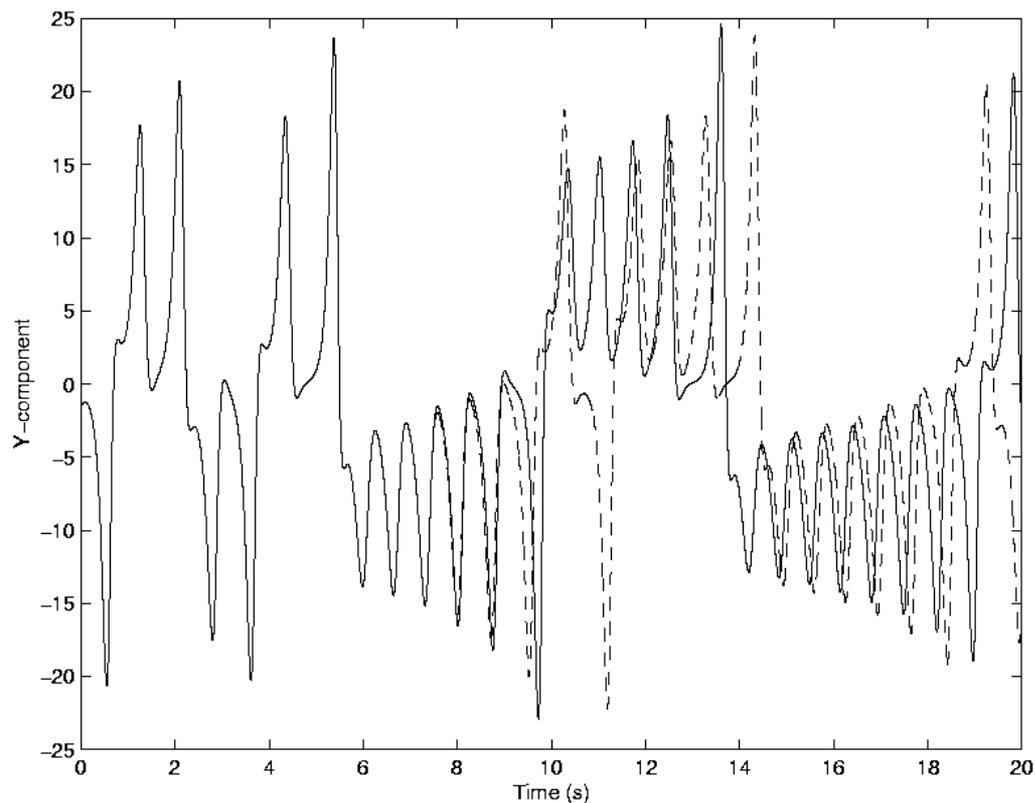
- Two initial conditions (Y component):  $(w_0)$  and  $(w_0 + \Delta w)$



# Predictability

## Lorenz's system

- Two initial conditions (Y component):  $(w_0)$  and  $(w_0 + \Delta w)$



# Predictability

## Are the chaotic systems predictable?

- Chaotic regime: evolution non-linear differential equation
- Chaotic systems: deterministic equations
- A class of chaos system: dynamics is confined in a region
- Strange attractors have fractal dimension
- They have at least one positive Lyapunov exponent

# Predictability

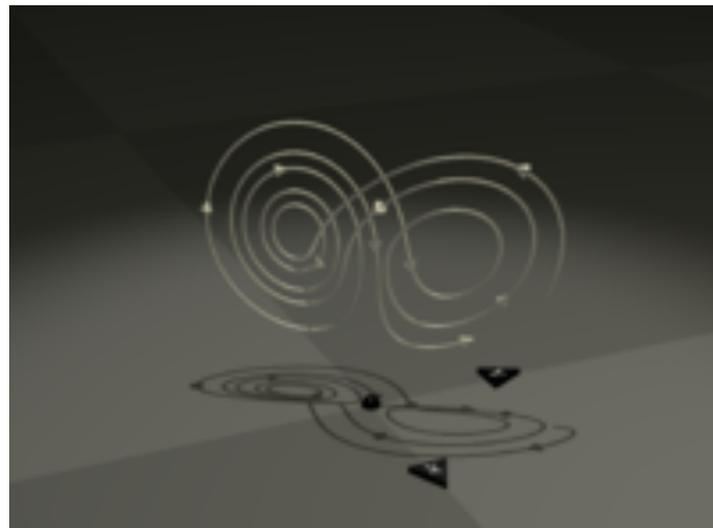
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- Chaotic regime: evolution non-linear differential equation
- Chaotic systems: deterministic equations
- A class of chaos system: dynamics is confined in a region
- Strange attractors have fractal dimension
- They have at least one positive Lyapunov exponent
- **They are strongly dependent on the initial conditions**

# Predictability

## Are the chaotic systems predictable?

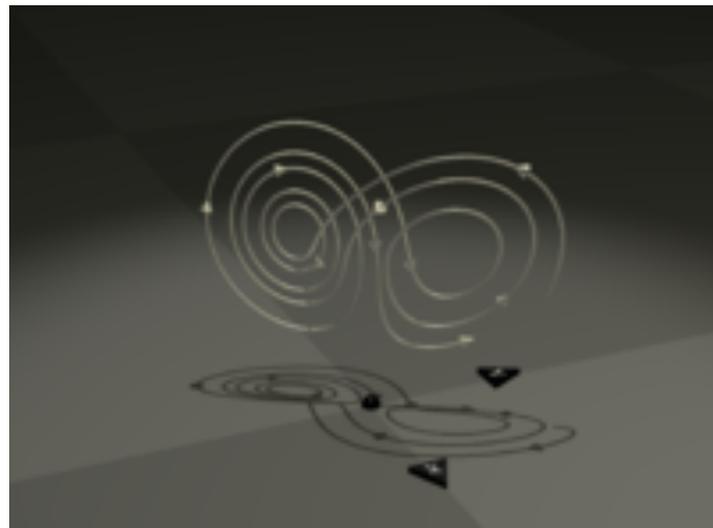
- Questions:
  - When will the changes go to occur in the dynamics?
  - How long will the new period be?



# Predictability

## Are the chaotic systems predictable?

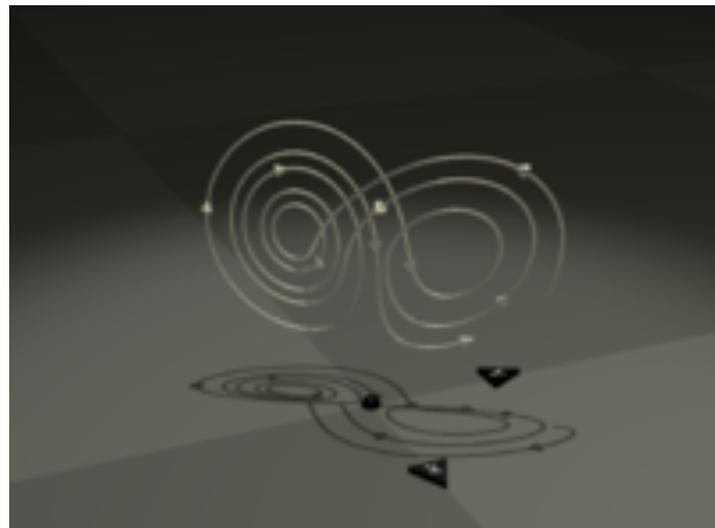
- Questions:
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  - How long will the new period be?
- Bred vectors can help us!



# Predictability

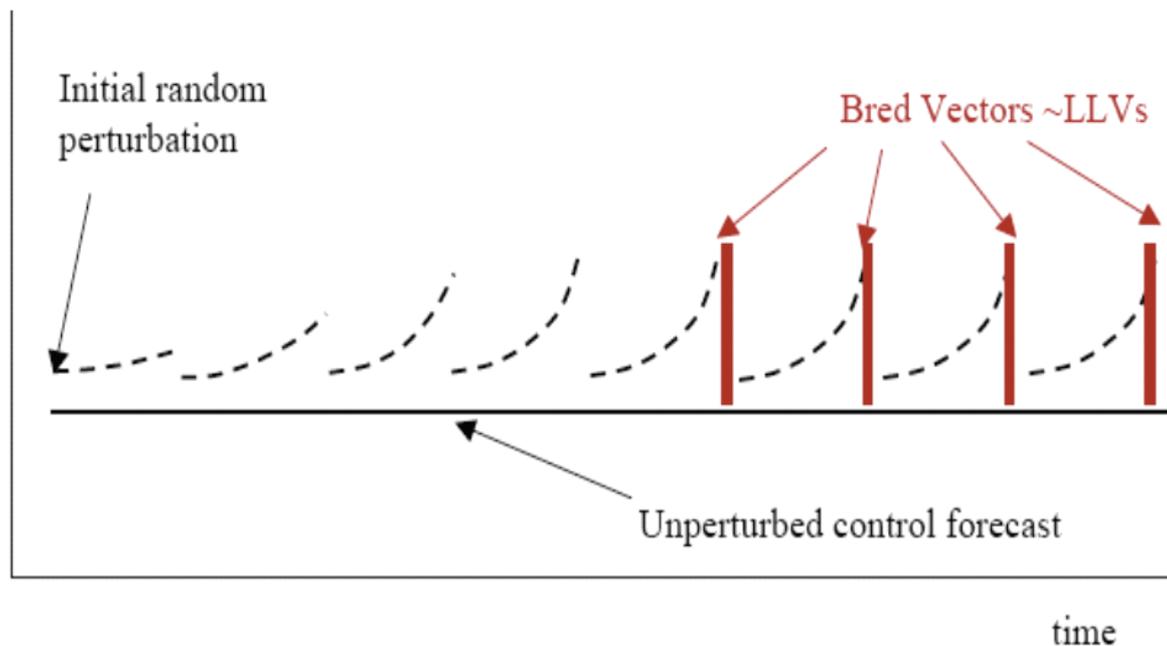
## Are the chaotic systems predictable?

- Questions:
  - When will the changes go to occur in the dynamics?
  - How long will the new period be?
- Bred vectors can help us!
- What is bred vector?



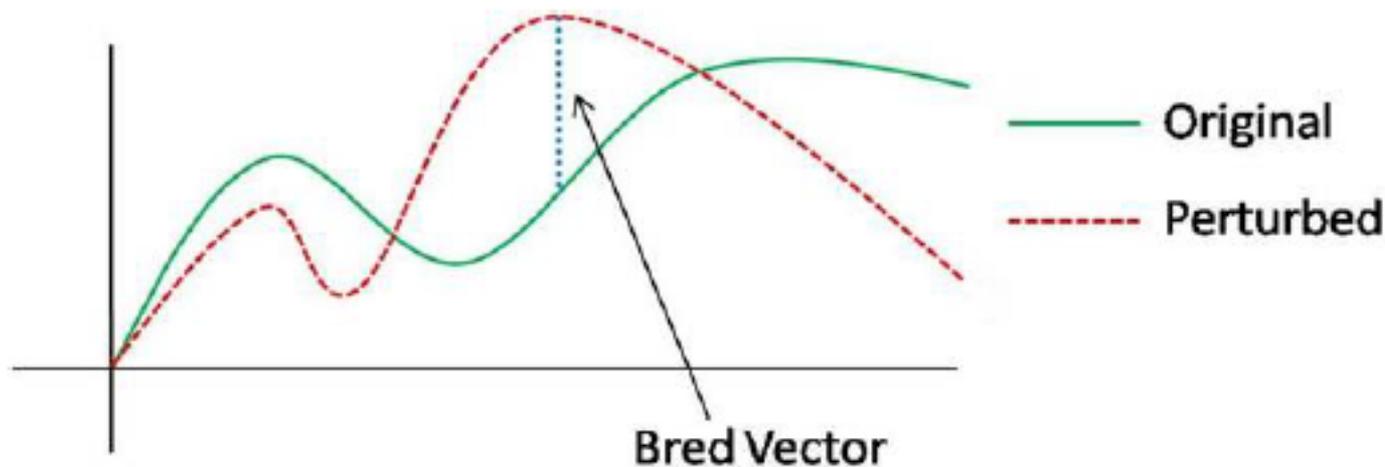
# Bred vector

- Bredding is a method to estimate forecast errors.
- It is applied in weather models (operational centers).
  - Bred vectors are simply the difference between two model runs:
  - The second originating from slightly perturbed initial conditions (periodically rescaled)



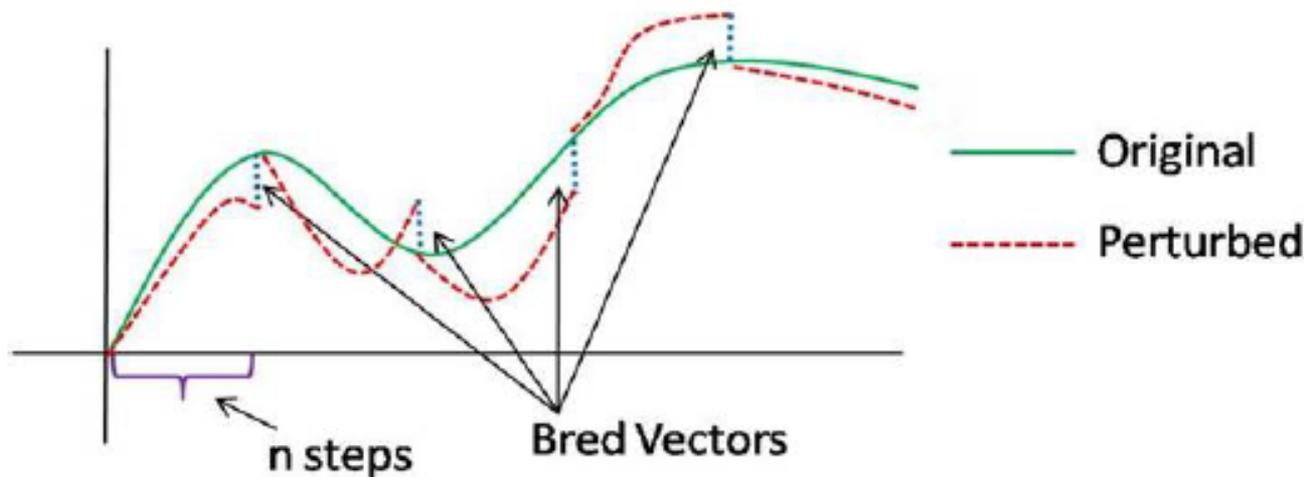
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# Bred vector

## Mathematical implementation

1. Start with a initial perturbation:  $\delta_{r_0} = \vec{r}_0 + \delta\vec{r}_0$
2. Evaluate de perturbation:  $\|\delta\vec{r}_0\|$
3. Evaluate perturbation propagation, after n time-steps

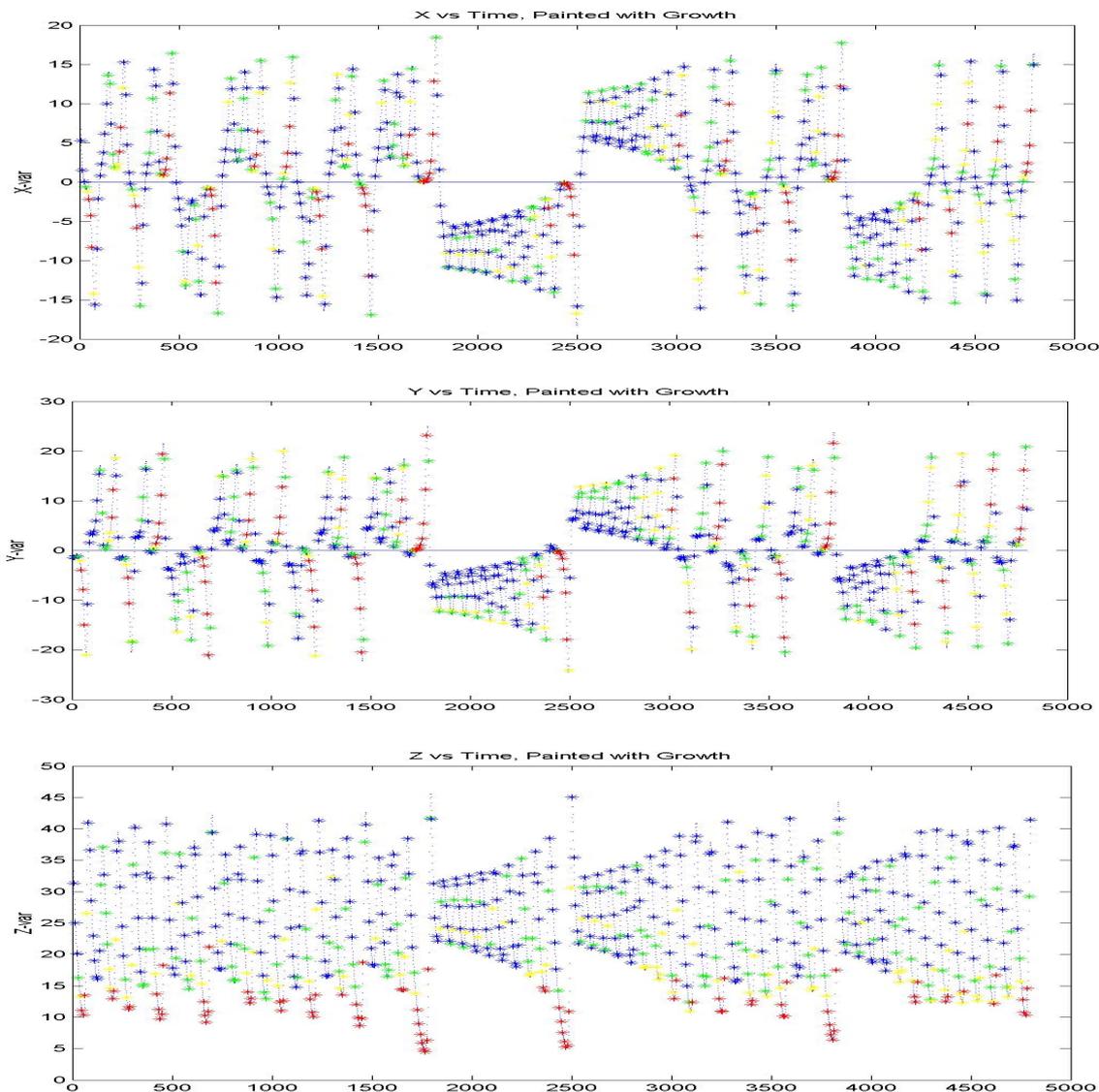
$$\delta\vec{r}(t) = \Phi(\vec{r}_0 + \delta\vec{r}_0, t_0 + n\Delta t) - \Phi(\vec{r}_0, t_0 + n\Delta t)$$

4. Evalute the growth ratio:

$$g = \frac{1}{n} \log \left( \frac{\|\delta\vec{r}(t)\|}{\|\delta\vec{r}_0\|} \right)$$

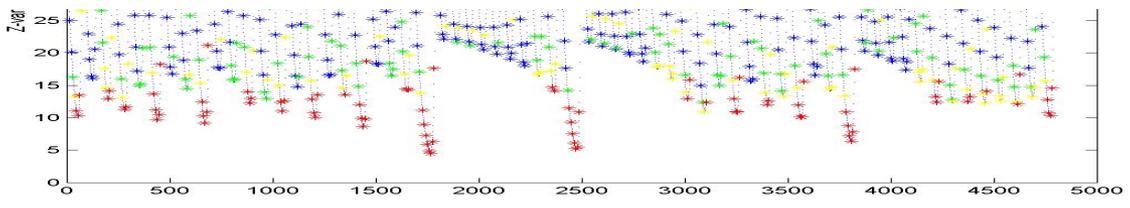
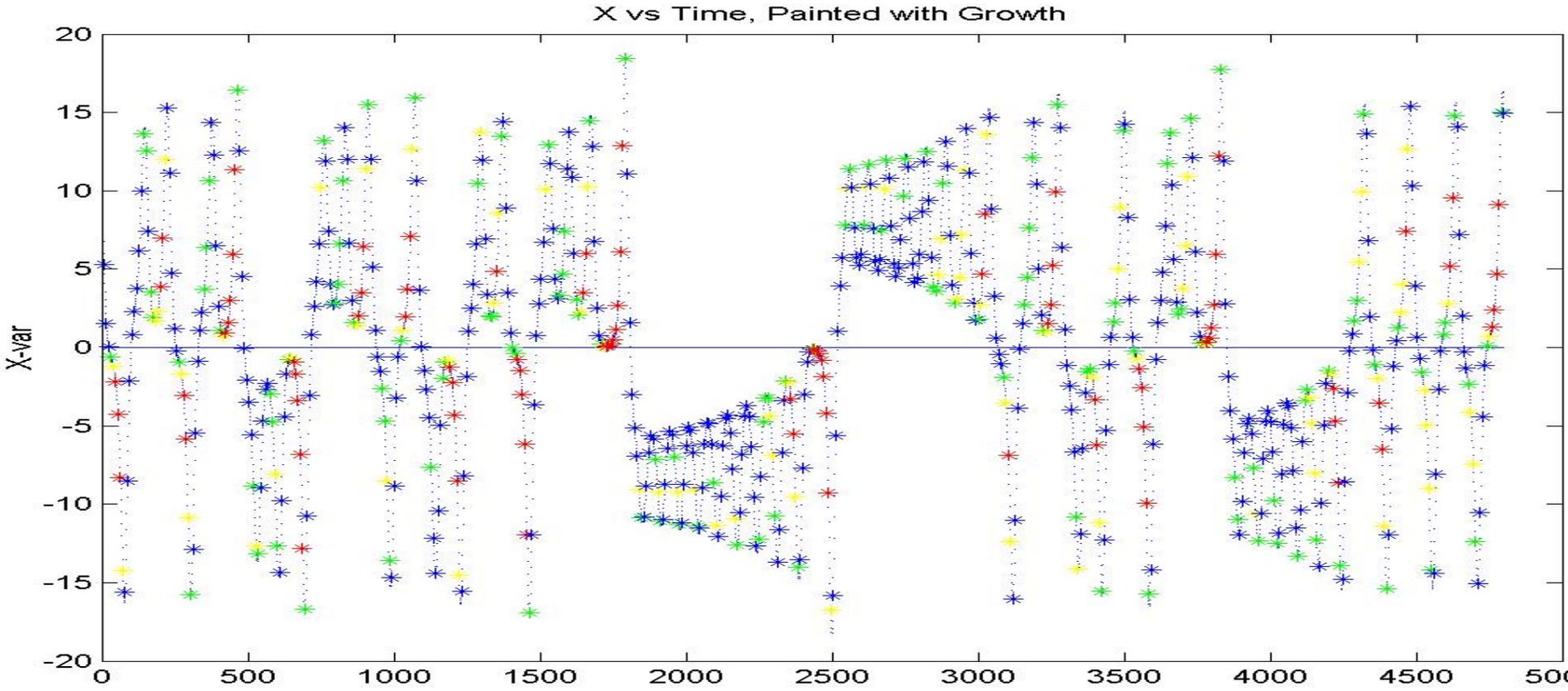
5. Re-scale the perturbation, and repete the process.

# Bred vector: Lorenz system

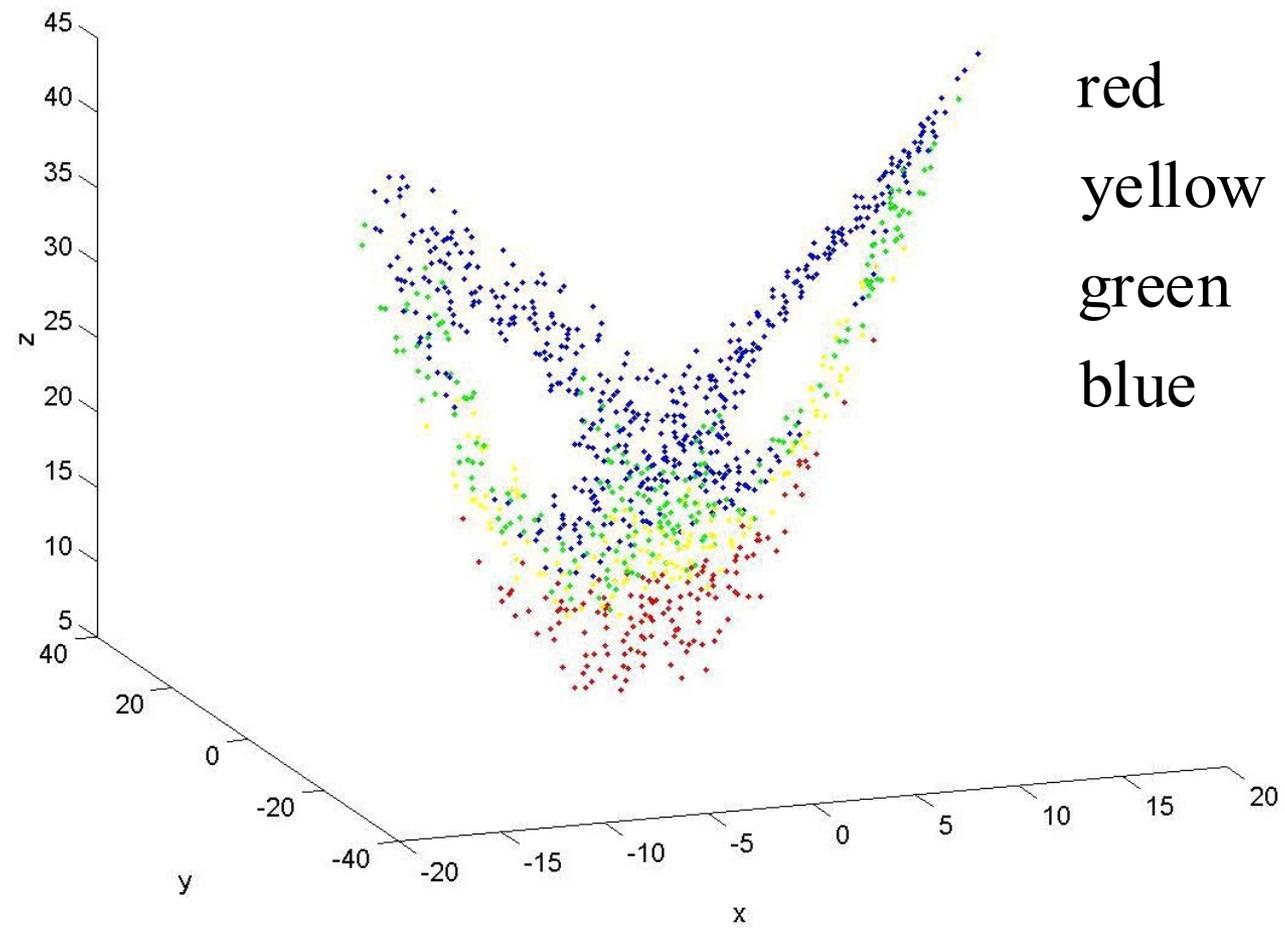


$$g = \frac{1}{n} \log \left( \frac{\|\delta \vec{r}(t)\|}{\|\delta \vec{r}_0\|} \right)$$

# Bred vector: Lorenz system



# Bred vector: Lorenz system



- red  $g > 0.064$
- yellow  $g \in [0.032, 0.064]$
- green  $g \leq 0.032$
- blue  $g < 0$

# Bred vector: Lorenz system

## Breeding method for predictability

- Evan et al. (Bulletin of AMS, 85, 520-524, 2004) have investigated the bred vector ratio growth to infer some “prediction rules”.



RISE UNDERGRADUATES FIND THAT REGIME  
 CHANGES IN LORENZ'S MODEL ARE PREDICTABLE

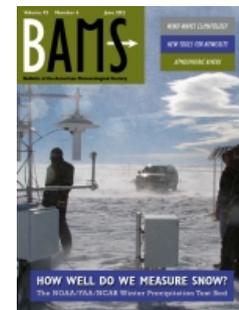
BY ERIN EVANS, NADIA BHATTI, JACKI KINNEY, LISA PANN, MALAQUIAS  
 PEÑA, SHU-CHIH YANG, EUGENIA KALNAY, AND JAMES HANSEN

UNDERGRADUATE

# Bred vector: Lorenz system

## Breeding method for predictability

- Evan et al. (Bulletin of AMS, 85, 520-524, 2004)
- Using bred vector approach two “prediction rules”:



1. When the growth rate exceeds 0.064 over a period of eight steps, as indicated by the presence of one (or more) red stars, the current regime will end after it completes the current orbit.
2. The length of the new regime is proportional to the number of red stars.

# Predictability

## How good is the prediction?

- Ensemble prediction
  - Data for statistical properties
  - Statistical tendencies
  
- Confidence interval
  - Large confidence interval: low predictability
  - Short confidence interval: high predictability

# Predictability

## How good is the prediction?

- Ensemble prediction
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- Predictability: degree of confidence in forecast

# Predictability

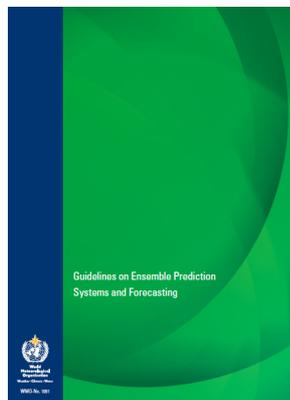
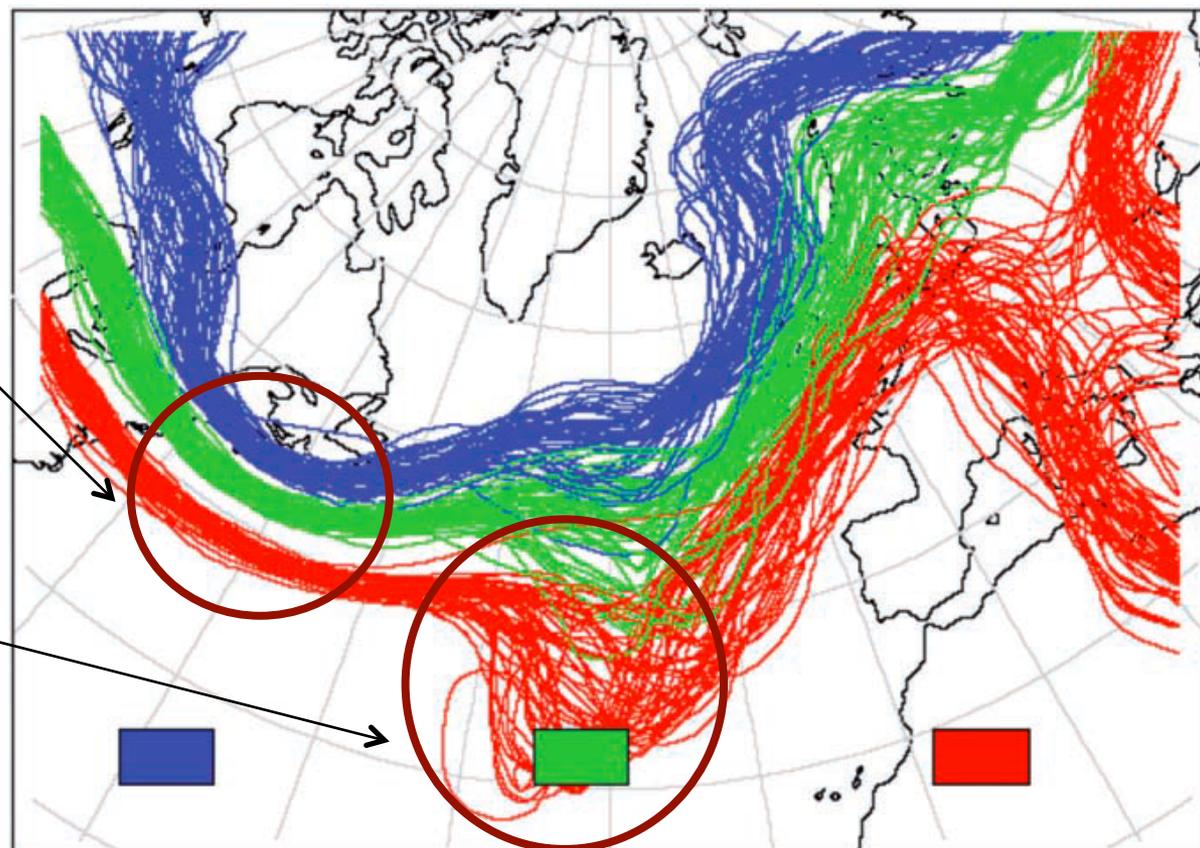
## How good is the prediction?

- Ensemble prediction
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- Confidence interval
  - Large confidence interval: low predictability
  - Short confidence interval: high predictability
- Predictability: degree of confidence in forecast
- Predictability: quantifying uncertainty

# Ensemble prediction

High predictability  
(ensemble convergence)

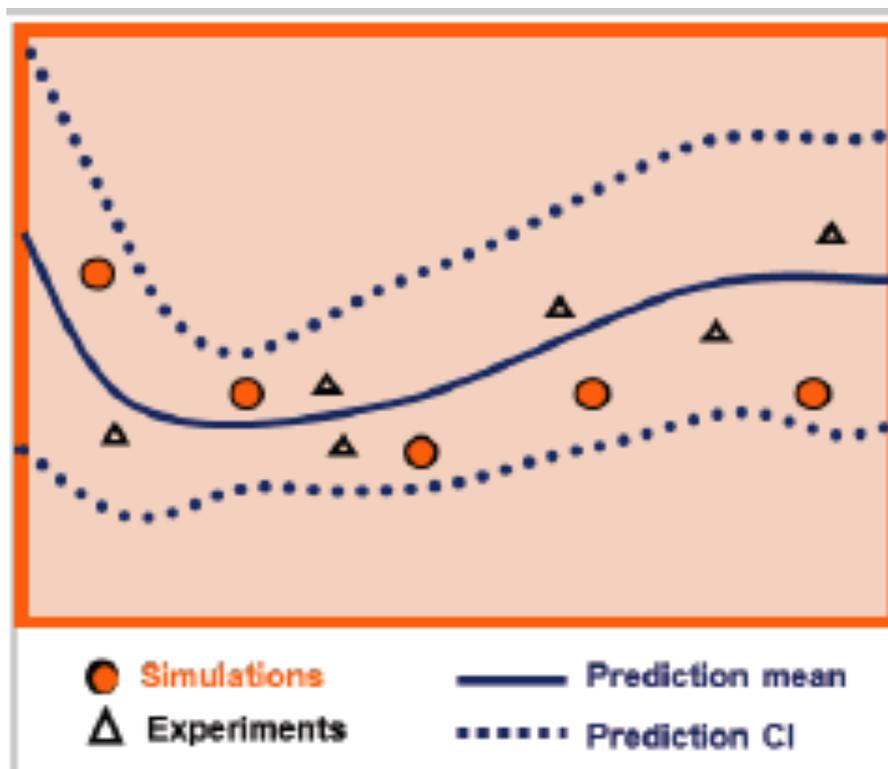
Low predictability  
(ensemble dispersion)



WMO's report describing/suggesting ensemble prediction

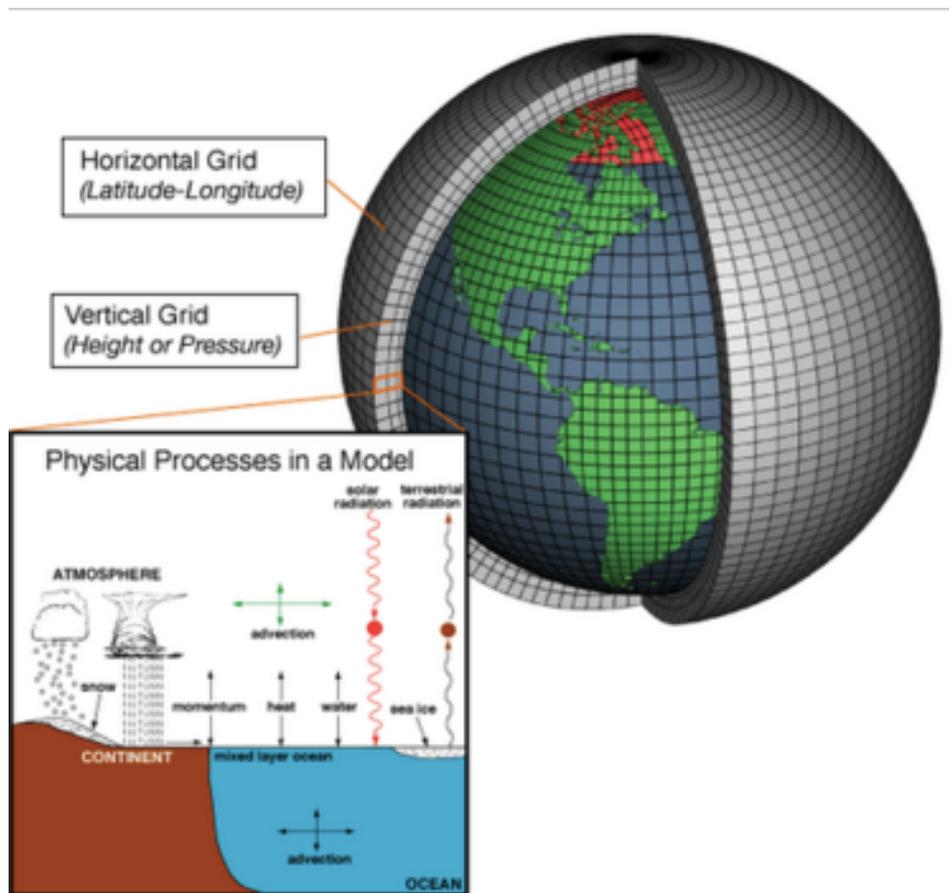
# Ensemble prediction

## Ensemble prediction and confidence interval



# Ensemble prediction

- Numerical Weather Prediction
- Relevant and computer intensive simulation



# Ensemble prediction

- **Numerical Weather Prediction: equations**

## Movement Equation (momentum)

$$\frac{du}{dt} - fv + \frac{1}{\rho} \frac{\partial p}{\partial x} = 0$$

$$\frac{dh}{dt} + g + \frac{1}{\rho} \frac{\partial p}{\partial z} = 0$$

$$\frac{dv}{dt} + fu + \frac{1}{\rho} \frac{\partial p}{\partial y} = 0$$

## Continuity Equation (mass)

$$\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x} (\rho u) + \frac{\partial}{\partial y} (\rho v) + \frac{\partial}{\partial z} (\rho h) = 0$$

## Thermodynamic equation (energy)

$$p = f(T)\rho \Rightarrow \frac{f(T)}{T} = \frac{g(\rho)}{\rho} \equiv R(cte) \quad p = g(\rho)T$$

$$p = \rho RT$$

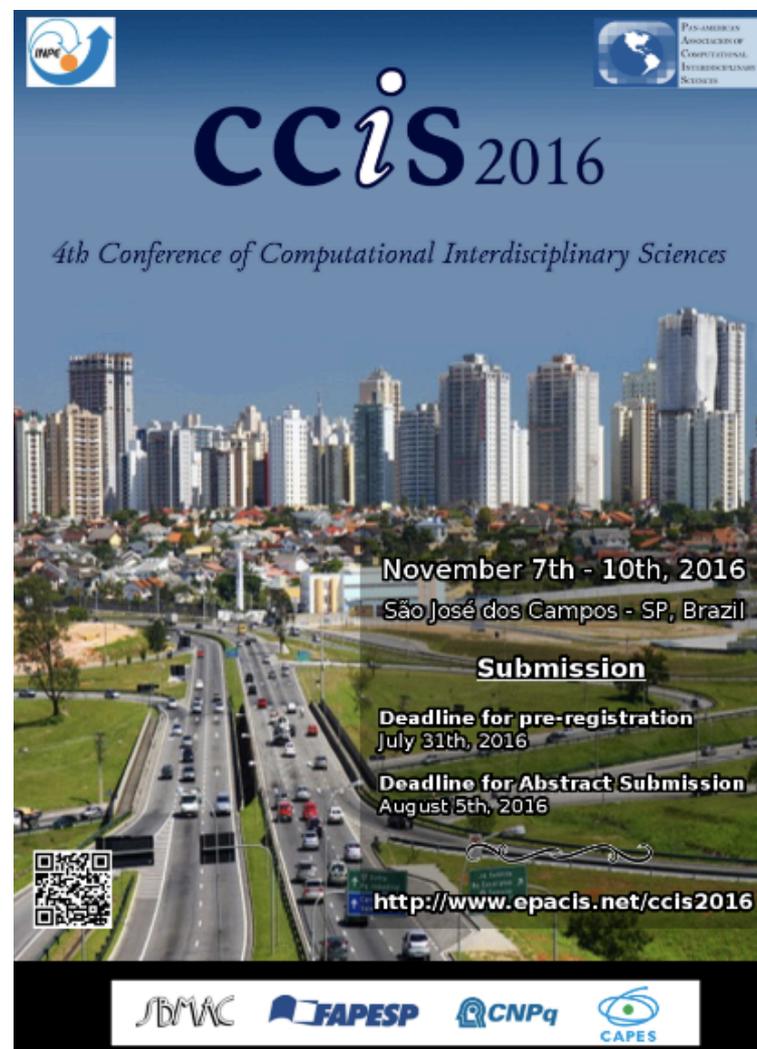
$$C_v \frac{dT}{dt} + p_{30} \frac{d(1/\rho)}{dt} = \frac{dq}{dt}$$

# 1 Analysis of the Breeding Technique applied to the 2 CPTEC-AGCM Model

3 Luis F. Salgueiro Romero<sup>1</sup>, Sandra A. Sandri and Haroldo F. de Campos Velho

CCIS'16 Proceedings

[www.epacis.net/ccis2016/en/](http://www.epacis.net/ccis2016/en/)



The poster for the 4th Conference of Computational Interdisciplinary Sciences (CCIS 2016) features a background image of a city skyline with a highway in the foreground. The text on the poster includes the conference title, dates, location, submission deadlines, and a QR code. Logos for INPE and PAS-AMERICAS are in the top corners, and logos for SBMAC, FAPESP, CNPq, and CAPES are at the bottom.

**CCIS** 2016  
*4th Conference of Computational Interdisciplinary Sciences*

November 7th - 10th, 2016  
São José dos Campos - SP, Brazil

**Submission**

**Deadline for pre-registration**  
July 31th, 2016

**Deadline for Abstract Submission**  
August 5th, 2016

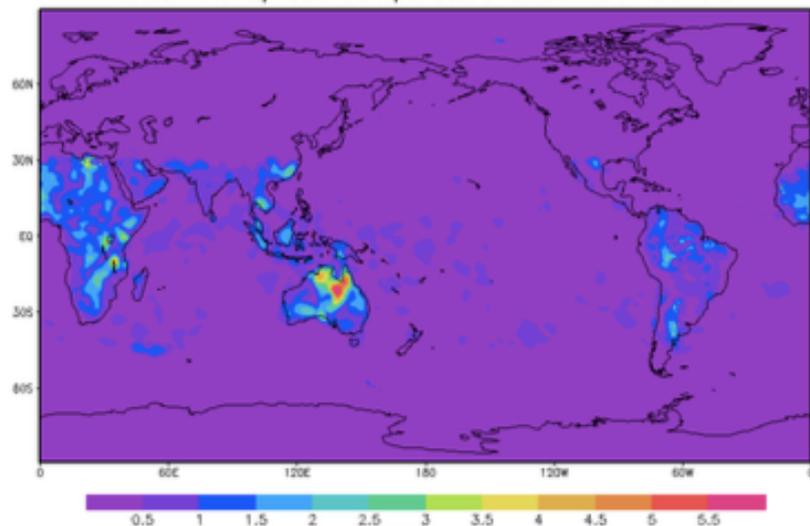
<http://www.epacis.net/ccis2016>

SBMAC FAPESP CNPq CAPES

# Ensemble prediction

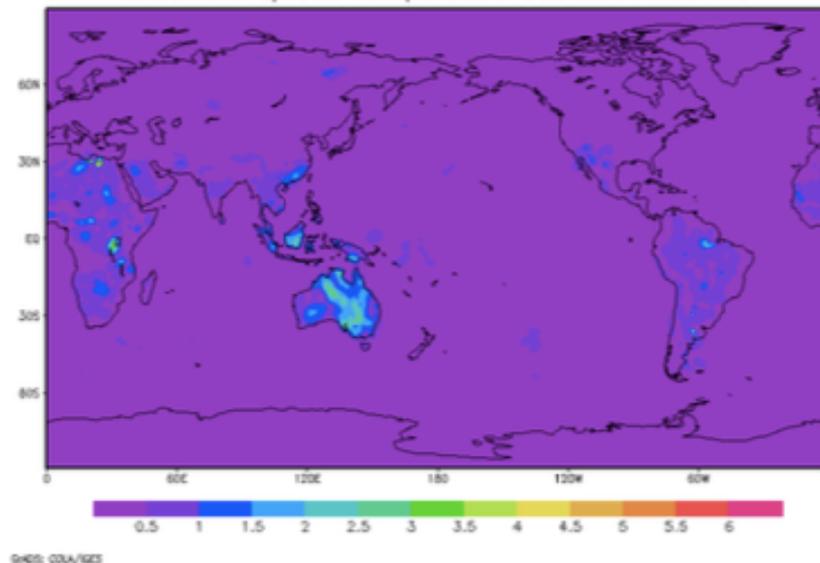
Ensemble: 15 members

Ensemble Spread Temperature 02Dez2014:06-Hs



(a) 6 Horas de previsão

Ensemble Spread Temperature 03Dez2014:06-Hs

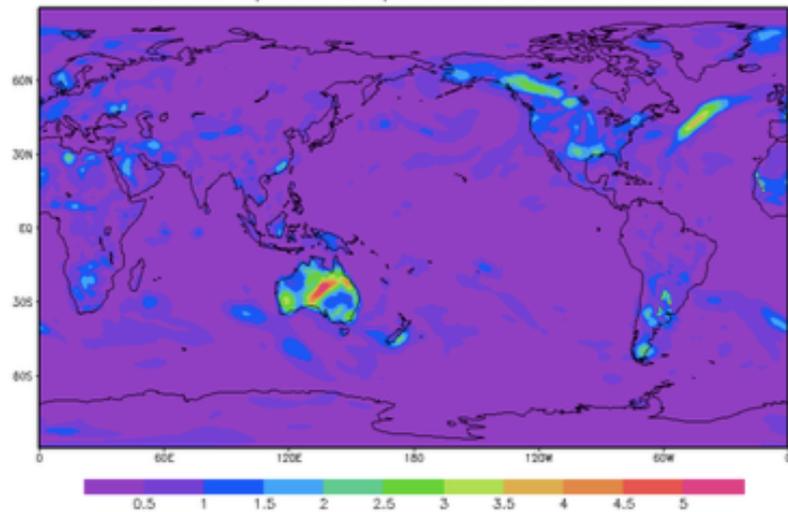


(b) 1 dia de previsão

# Ensemble prediction

Ensemble: 15 members

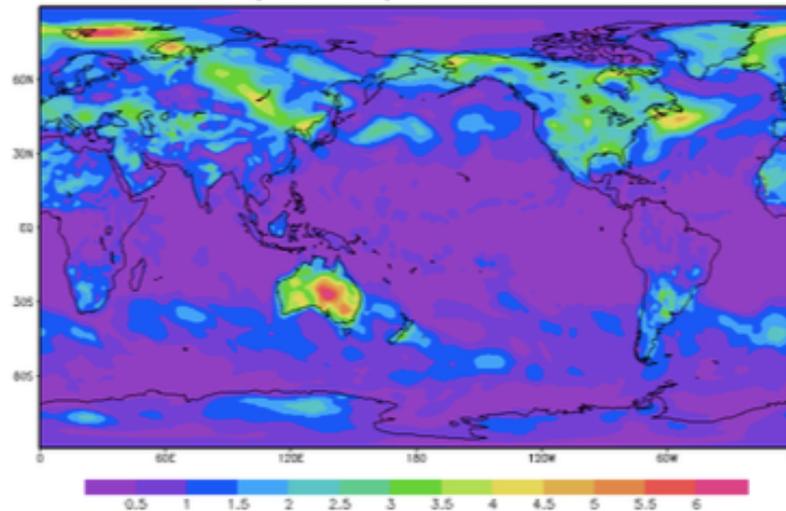
Ensemble Spread Temperature 07Dez2014:06-Hs



GHRS: ODU/IGES

(c) 5 dias de previsão

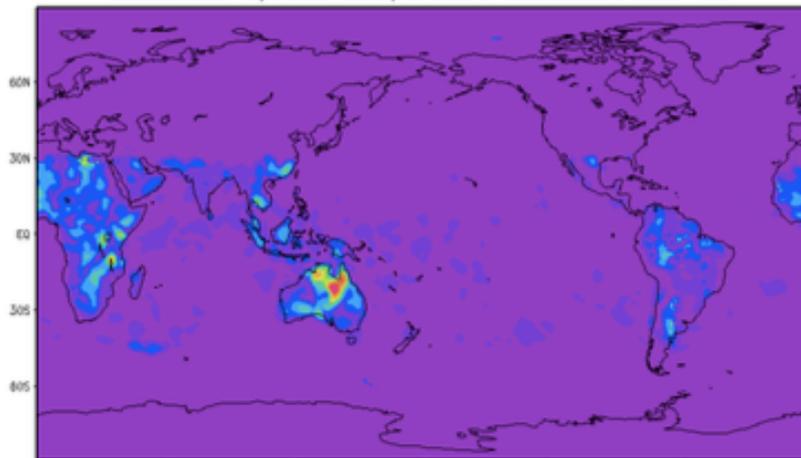
Ensemble Spread Temperature 12Dez2014:06-Hs



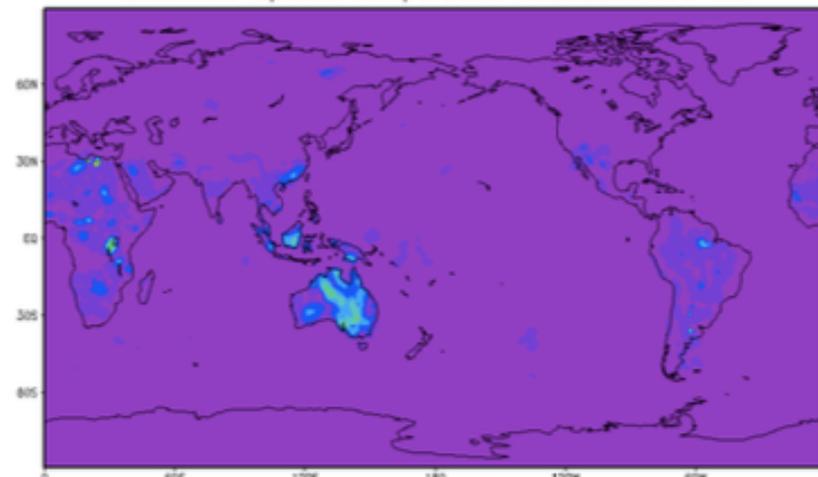
(d) 10 dias de previsão

# Ensemble prediction

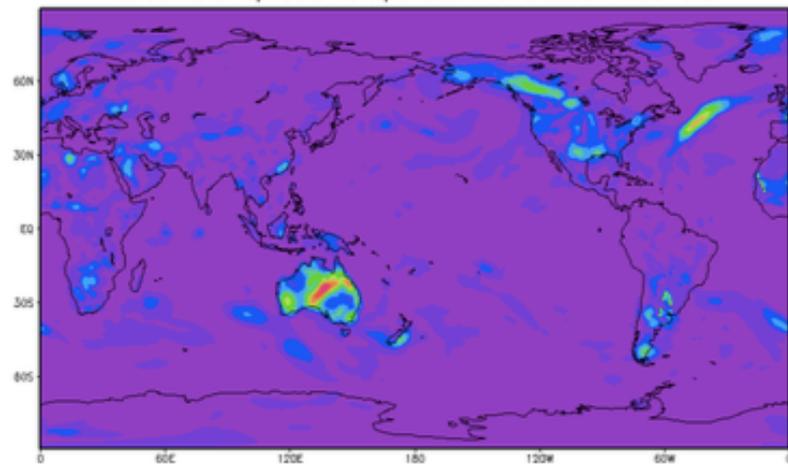
Ensemble Spread Temperature 02Dez2014:06-Hs



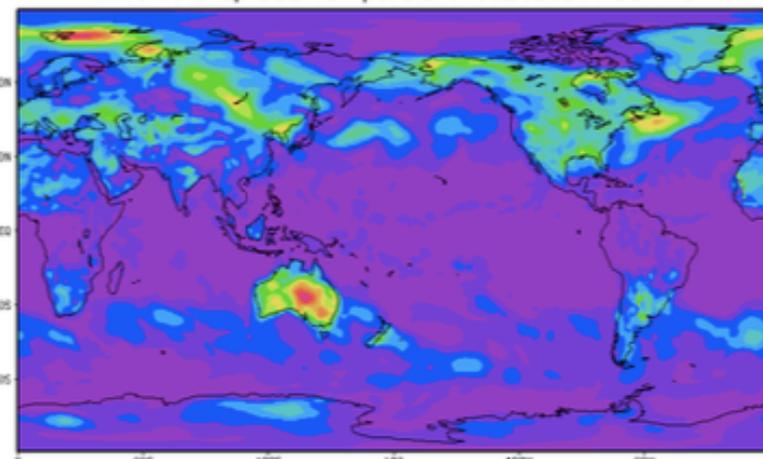
Ensemble Spread Temperature 03Dez2014:06-Hs



Ensemble Spread Temperature 07Dez2014:06-Hs



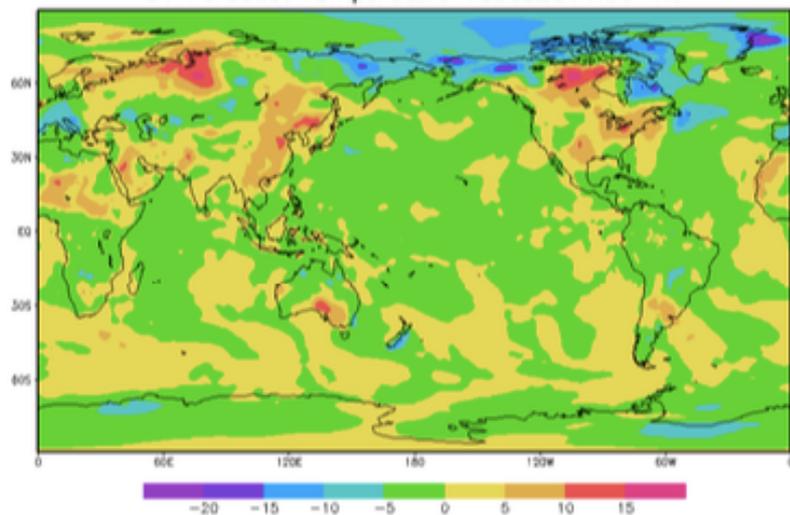
Ensemble Spread Temperature 12Dez2014:06-Hs



# Bred vector

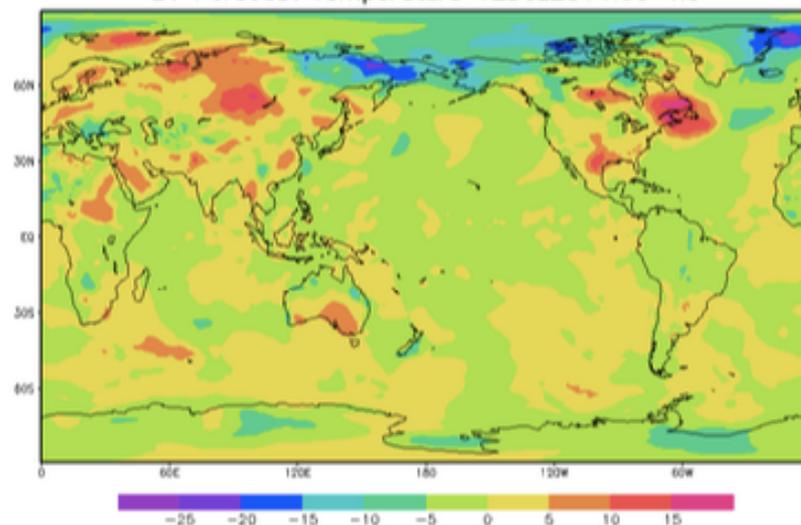
## Re-scaling at each 6 hours

BV Forecast Temperature 10Dez2014:06-Hs



(a) 8 dias de Previsão

BV Forecast Temperature 12Dez2014:06-Hs

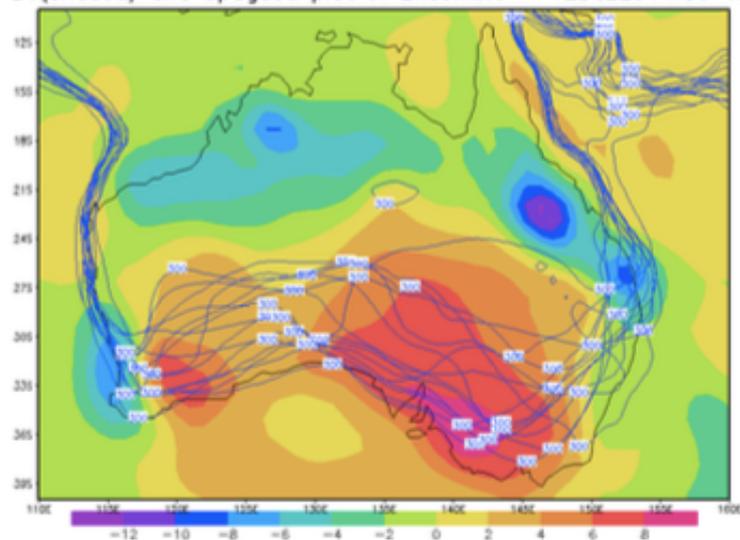


(b) 10 dias de Previsão

# Ensemble x Bred vector

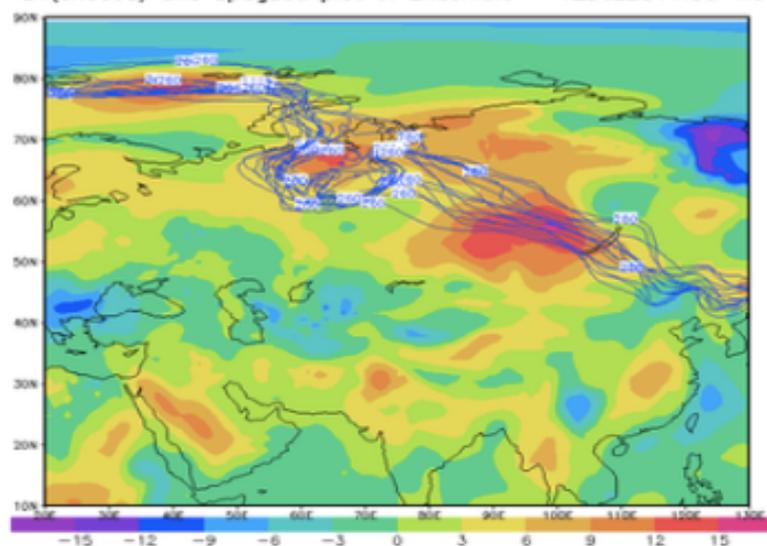
Re-scaling at each 6 hours (12/Dez/2014)

BV(shaded) and Spagueti plot of Ensemble - 12Dez2014:06-Hs



(a) Região de Austrália

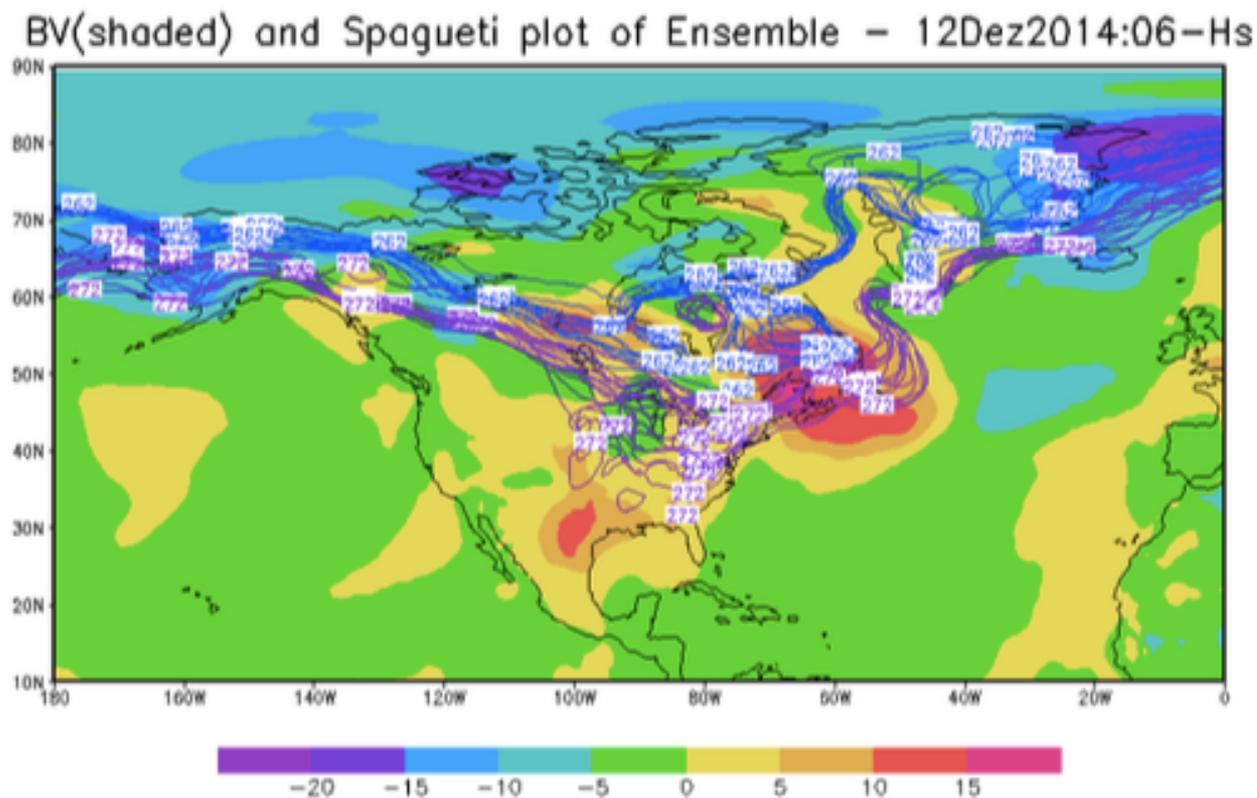
BV(shaded) and Spagueti plot of Ensemble - 12Dez2014:06-Hs



(b) Região Norte de Europa e Asia

# Ensemble x Bred vector

Re-scaling at each 6 hours (12/Dez/2014)

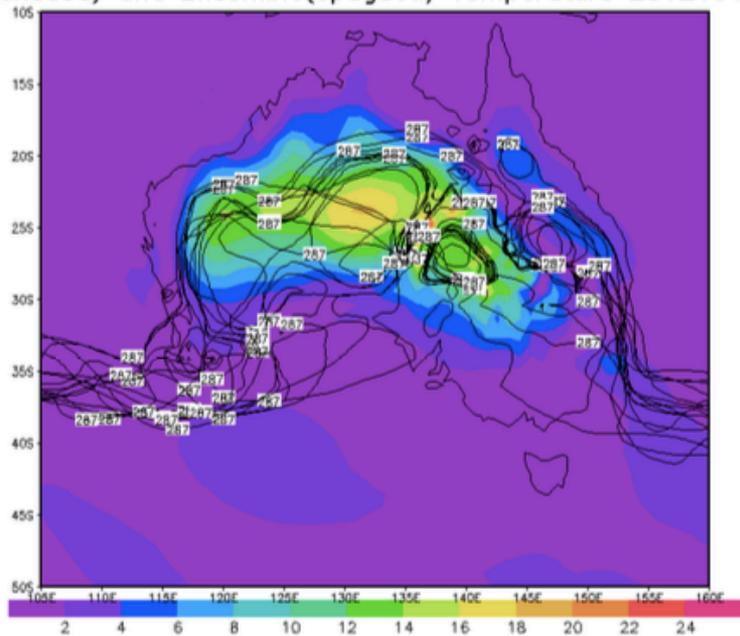


(c) Região de América do Norte

# Ensemble x Bred vector

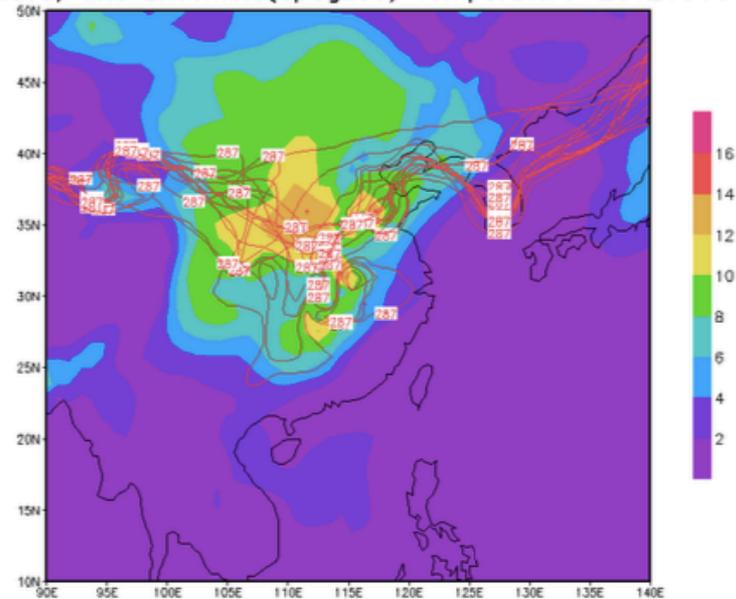
## Re-scaling at each 6 hours (16/out/2012)

BV(shaded) and Ensemble(spagueti) Temperature 2012101618



(a) Região de Australia

BV(shaded) and Ensemble(Spagueti) Temperature-2012101618

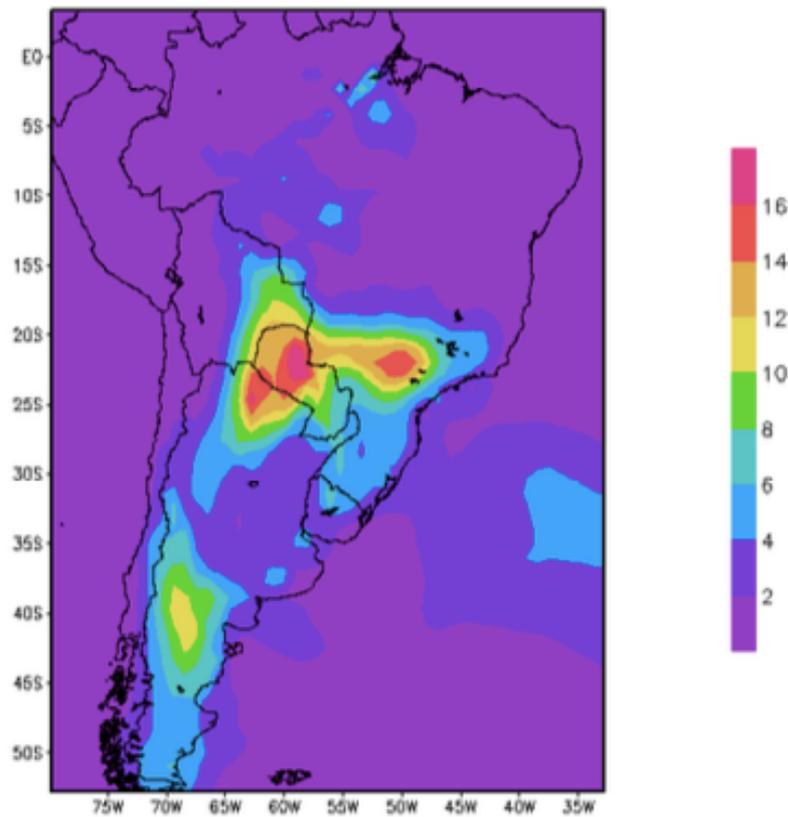


(b) Região de América do Norte

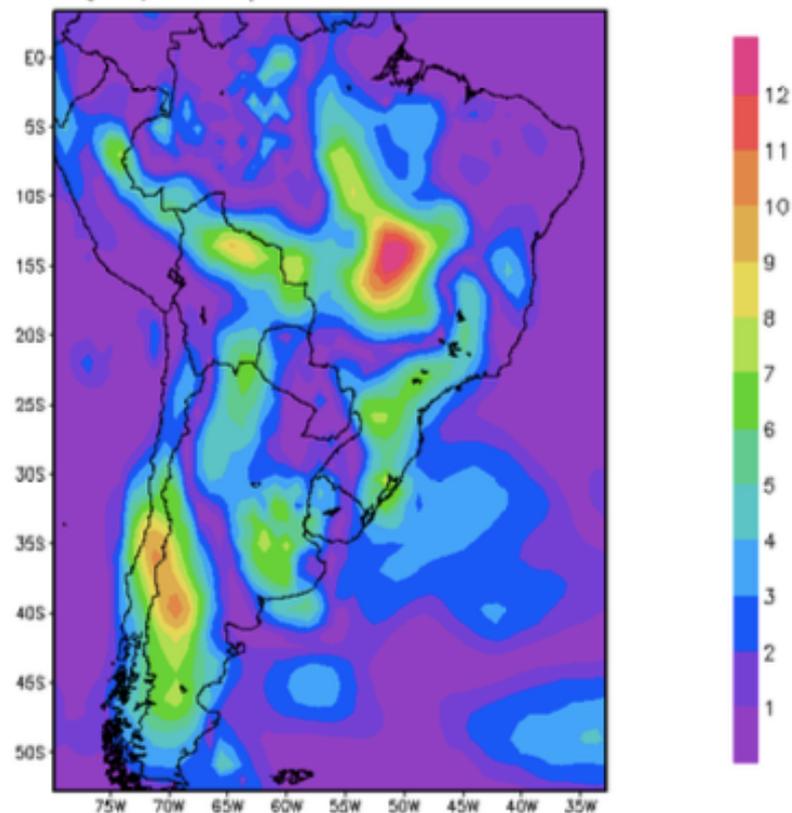
# Ensemble x Bred vector

Re-scaling at each 6 hours

Variância do Ensemble 2012101700

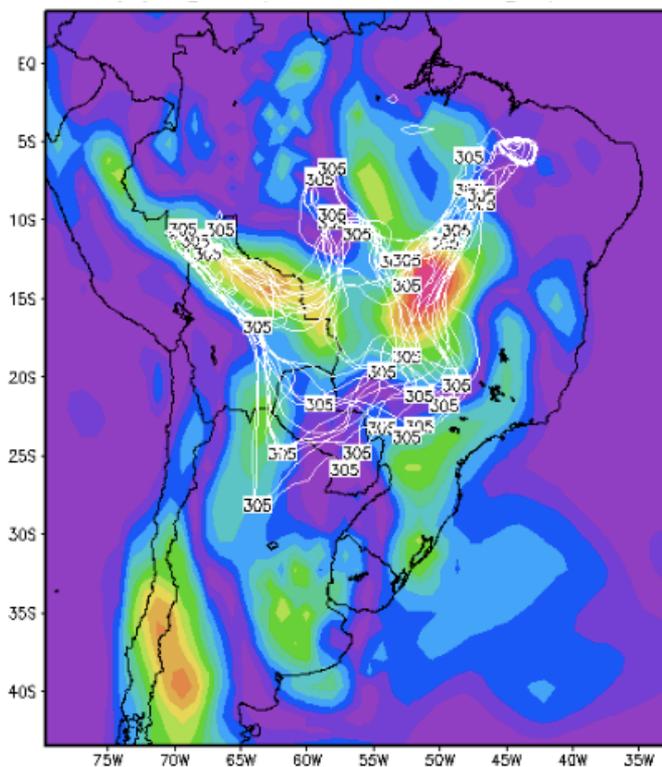


Abs(BV) Temperatura 2012101700

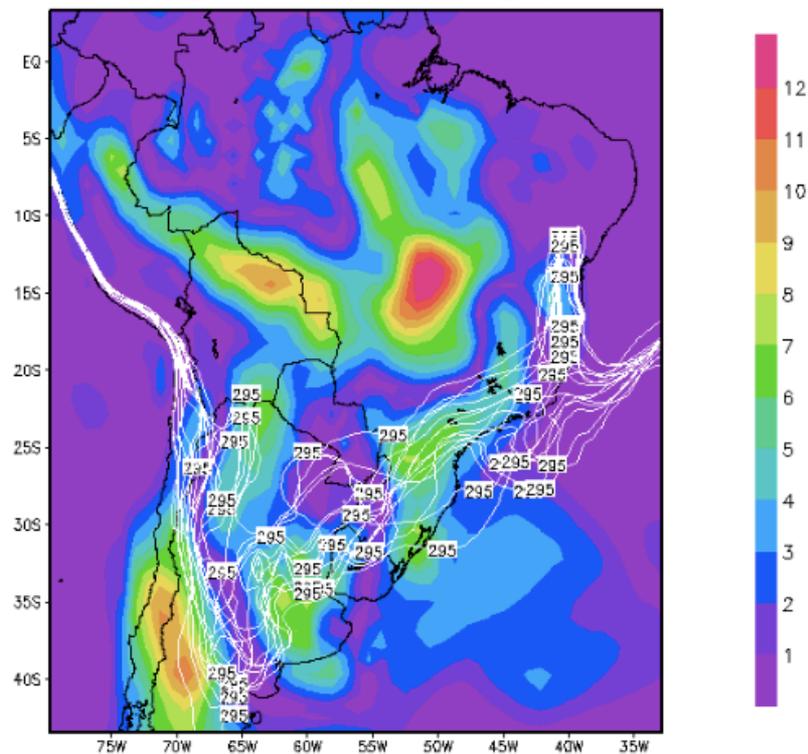


# Ensemble x Bred vector

## Ensemble: isoTemp 305



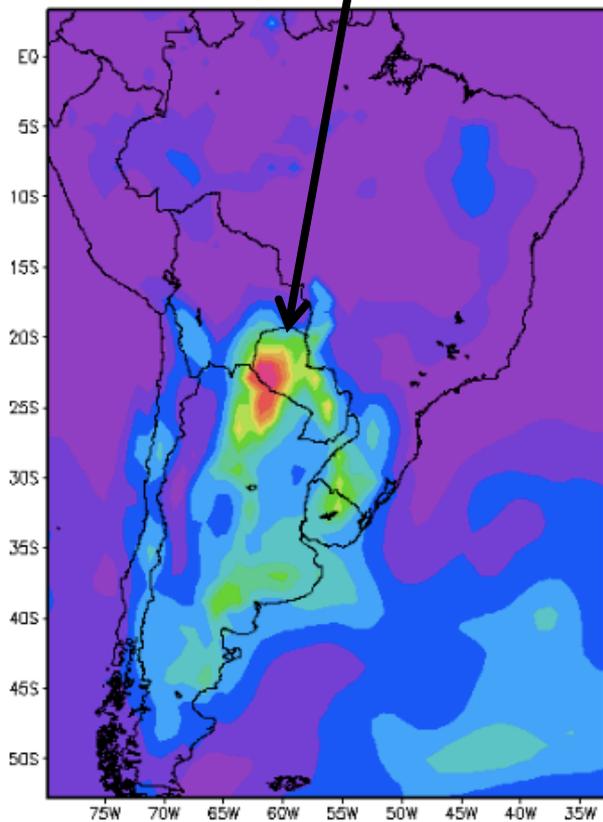
## Ensemble: isoTemp 295



# Ensemble x Bred vector

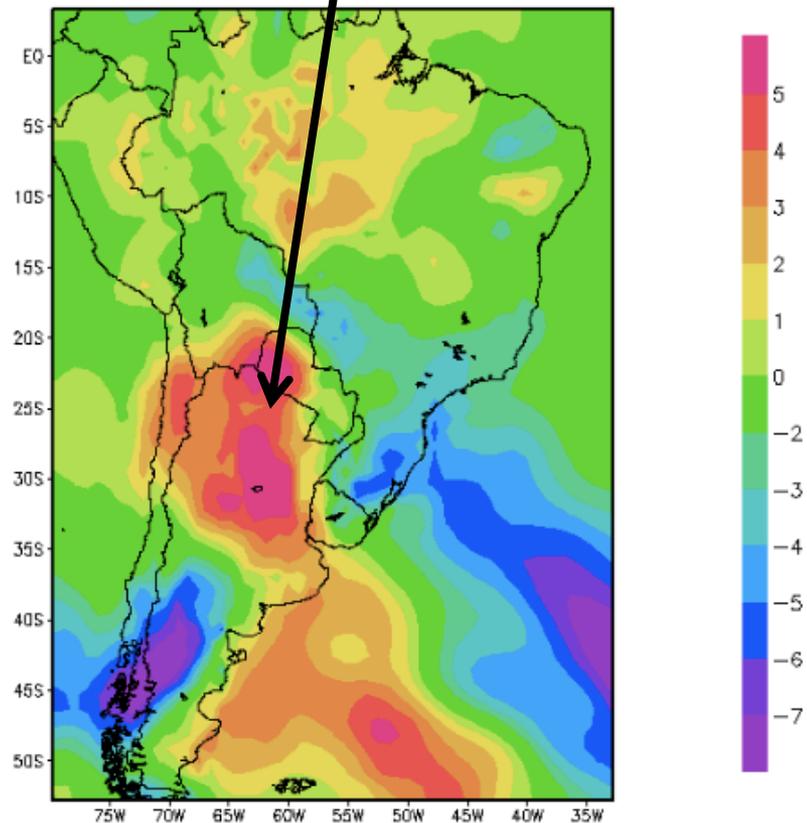
**Ensemble: max-Var**

Variância Temperatura 2012051700



**BV: max value**

Bred vector Temperatura 2012051700



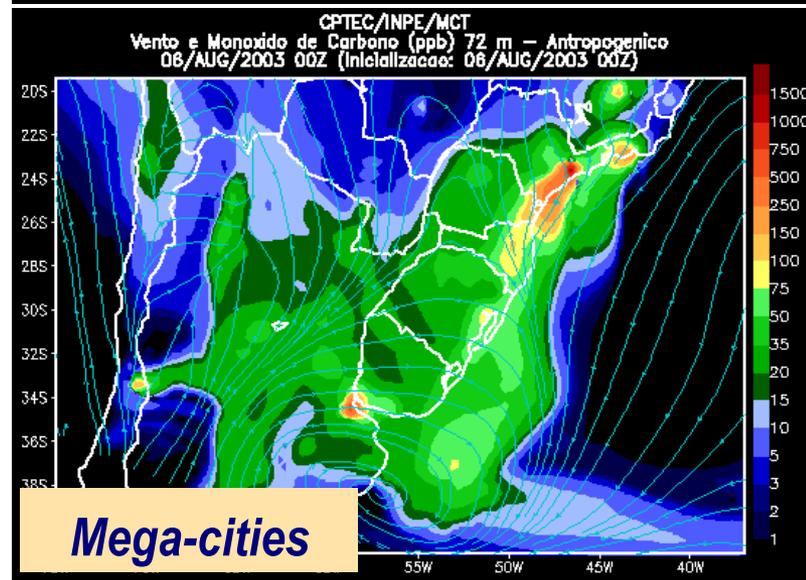
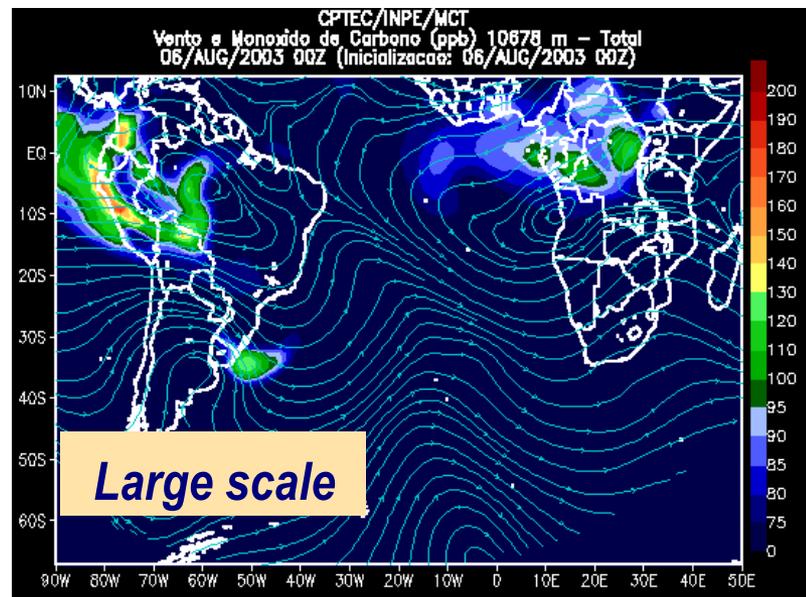
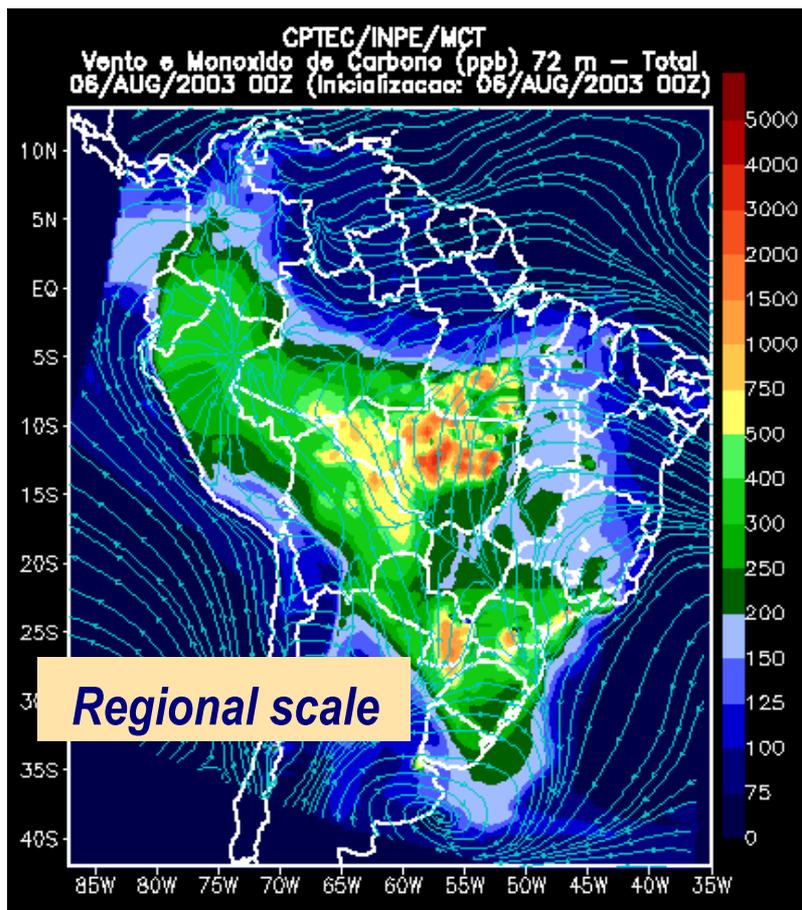
# Ensemble x Bred vector

## Predictability: some remarks

1. Bred vector demands less computer effort than ensemble prediction.
2. There is a good agreement between ensemble maximum variance and higher magnitude of BV.
3. When a disagreement is noted, the BV indicates a kind of low predictability – see South America case.
4. The BV technique can be employed to the CPTEC-INPE mesoscale prediction (BRAMS) and for SUPIM prediction.

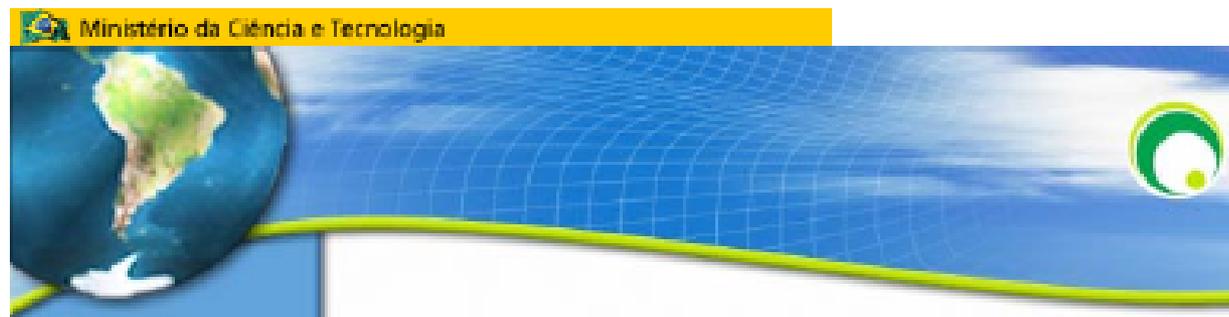
# BRAMS model

Pollutant emission by forest fires and urban-industries



# BRAMS model

<http://brams.cptec.inpe.br>



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## Model Description

### Brazilian Regional Atmospheric Modeling System (BRAMS)

BRAMS (Brazilian Regional Atmospheric Modeling System) is a joint project of [ATMET](#), [IME/USP](#), [IAG/USP](#) and [CPTEC/INPE](#), funded by [FAPESP](#) (Funding Agency), aimed to produce a new version of [RAMS](#) in the tropics. The main objective is to provide a single model to Brazilian Weather Centers. The BRAMS/RAMS model is a multipurpose prediction model designed to simulate atmospheric circulation scale from hemispheric scales down to large eddy simulations: planetary boundary layer.



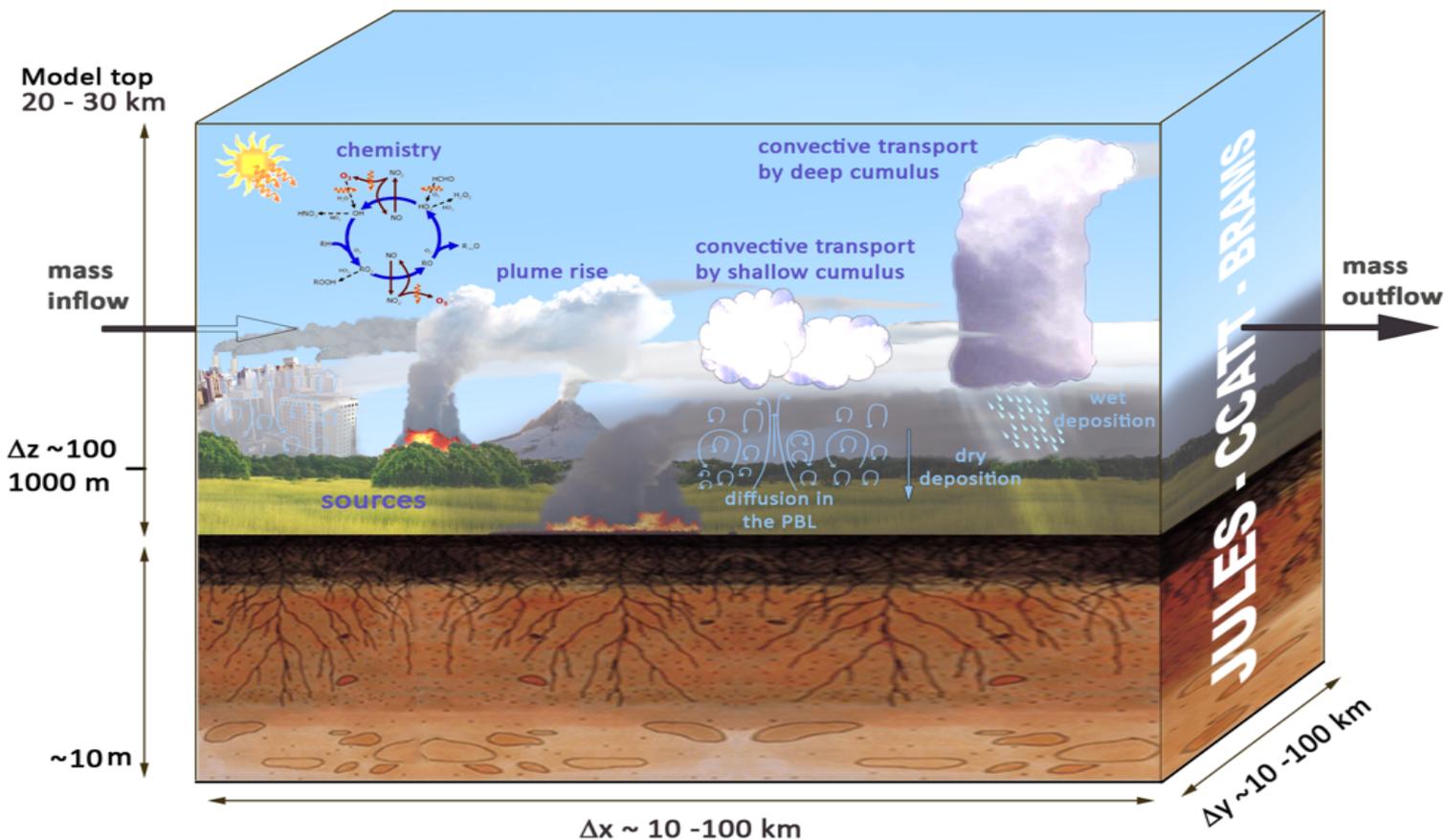
BRAMS is licensed under the [CC-GNU GPL](#).

## BRAMS Version 3.2 is RAMS Version 5.0.4 plus:

- Shallow Cumulus and New Deep Convection (mass flux several closures, based on Grell et al., 2002)

# BRAMS model

## BRAMS 5.2 (new version) Air quality and weather prediction



# BRAMS model

Geosci. Model Dev. Discuss., doi:10.5194/gmd-2016-130, 2016

Manuscript under review for journal Geosci. Model Dev.

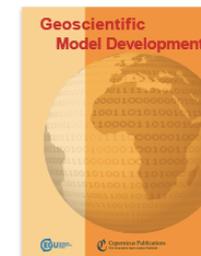
Published: 7 June 2016

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## The Brazilian developments on the Regional Atmospheric Modeling System (BRAMS 5.2): an integrated environmental model tuned for tropical areas

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# SUPIM model



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*Advances in Space Research* 54 (2014) 22–36

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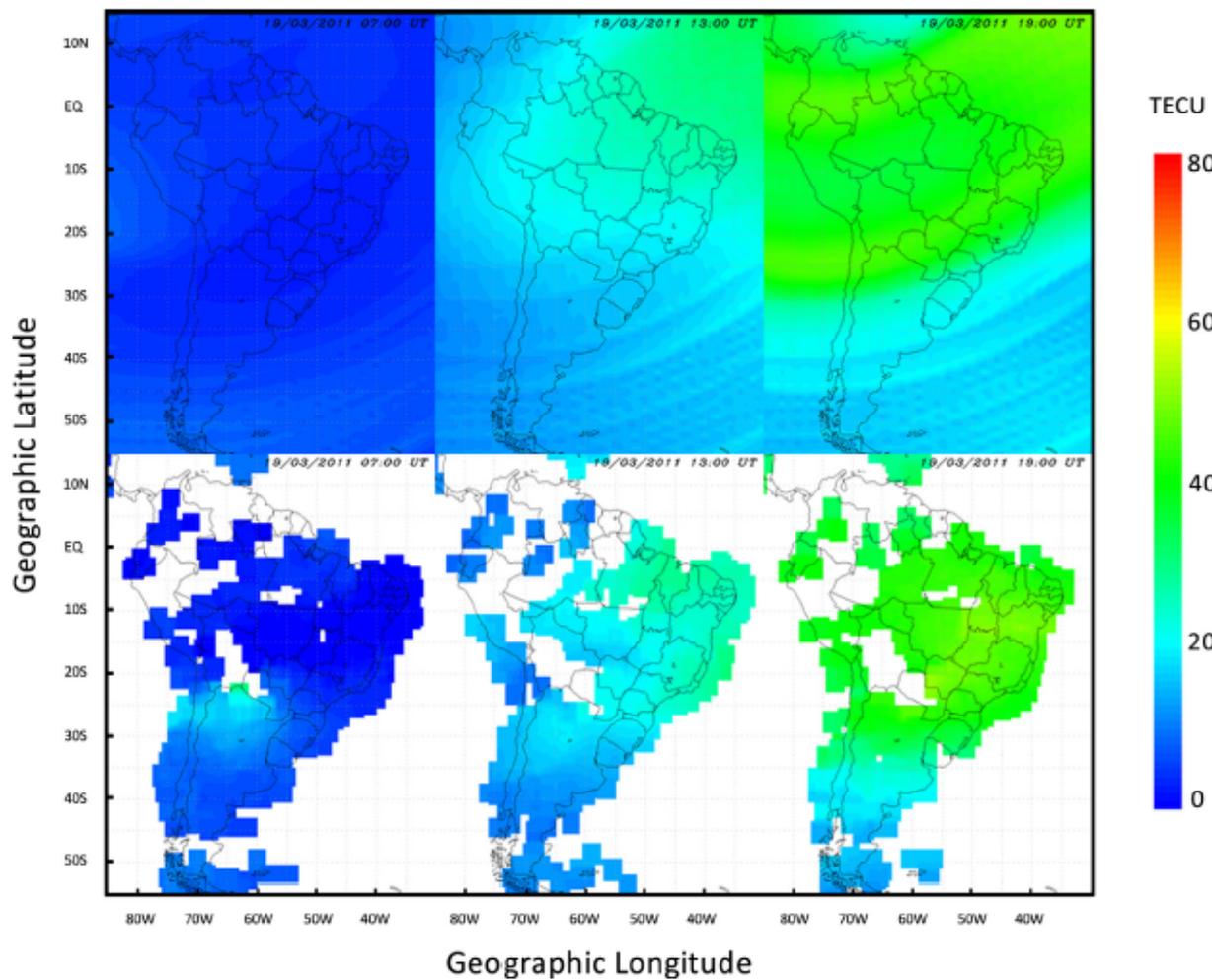
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## First results of operational ionospheric dynamics prediction for the Brazilian Space Weather program

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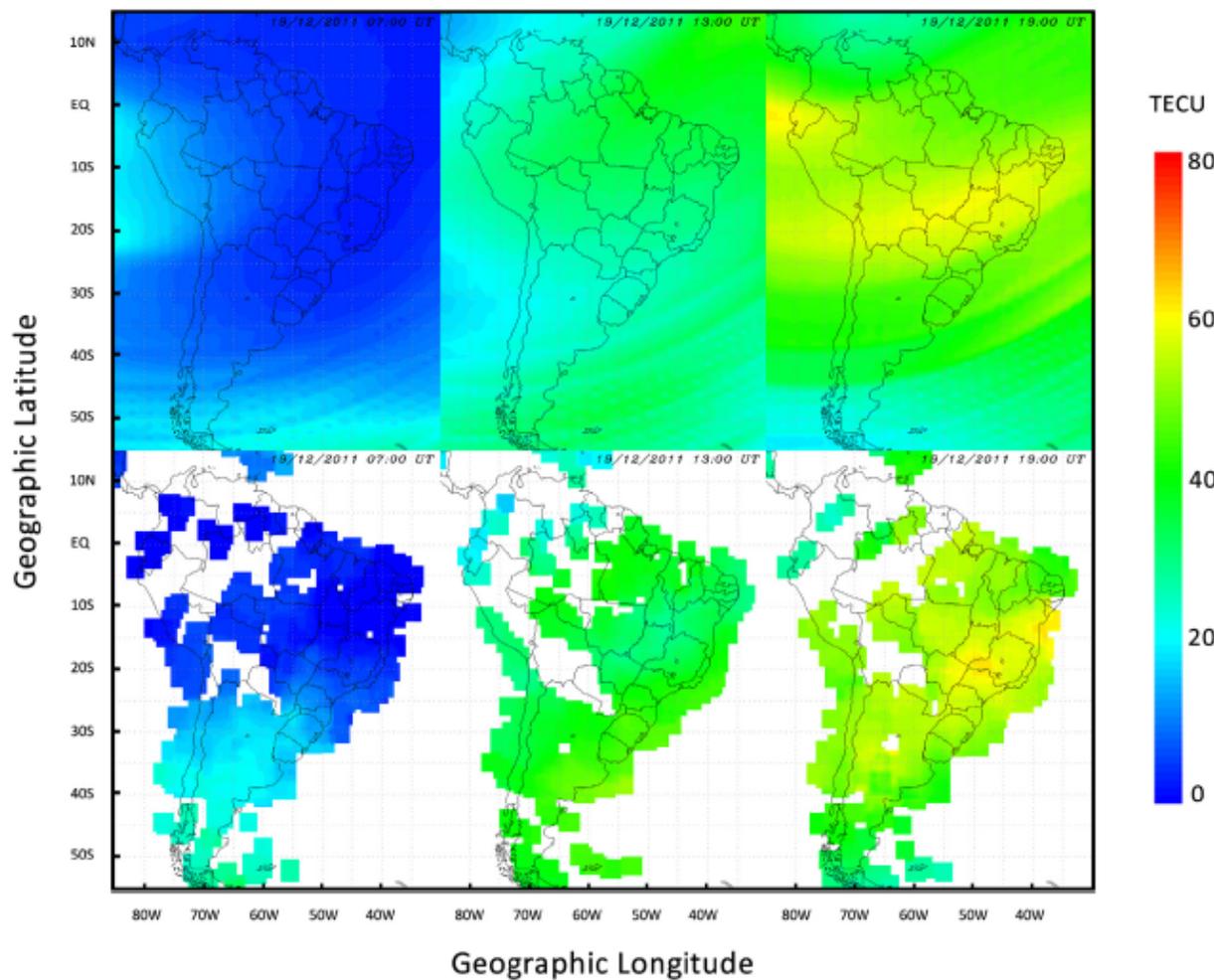
# SUPIM model

- 7, 13, 19 UT: March 19<sup>th</sup>, 2011



# SUPIM model

- 7, 13, 19 UT: December 19<sup>th</sup>, 2011





**Obrigado!**