



InTheArt : a discussion group (*started ~1 year ago*)
about *Artificial In*telligence

Valérie Gautard
(CEA – Saclay, IRFU)

Various topics discussed within the group, from Physics to Biology

- IA used in many different topics :
 - Genomics:
 - Classical approach: statistical association between each unique variation of the genome and the disease
 - Human genome: 3 billion nucleotides, 3 billion potential variations + epigenetic factors
 - Today: poorly correlated results with the observation of familial transmission of some diseases
 - Climate (some examples presented in the following slides)
 - Detector

***Idea of the group: share the knowledge, discuss and work together, through and open group.
Not only CEA people, your participation is more than welcome !***

Participation from : CEA (DEN, DRT, DRF-IRFU and CNRGH), CEA-CNRS (IPhT, Neurospin, LSCE) and Paris-Sud university

Some grants from CEA-DRF



Goals:

- ➔ **Organization of seminars** aiming at helping to understand the different "theoretical" problems in AI. And allow the exchange and the definition of a common language in order to solve them.
- ➔ **In a second step, apply this knowledge to some particular problems** (genomics, climate)

Edith Le Floch – CNRGH – Nov. 2017

Can we predict the phenotype of an individual from DNA ?

Özgür Sahin – IRFU – Jan. 2018

Machine learning technics at the LHC experiment

Davide Feranda – LSCE – Avr. 2018

New dynamical systems tools to study atmospheric flows

David Rousseau – LAL – Mai 2018

Advances in Machine learning in High Energy Physics

Etienne Thevenot – LIST – Juil. 2018

Omics data processing and analysis for high-throughput phenotyping

Michalis Vazirgiannis, DASCIM, Ecole Polytechnique – Sep. 2018

Graph Degeneracy for social nets and text mining







Jacques-Henri Sublemonier – Iramis – Nov. 2018

Locality-Sensitive hashing indexing schemes for metagenomics data

Adnane Hamid – IRFU – Feb. 2019

Machine Learning for climate prediction

Themes:

-  Biology, medical app
-  Machine Learning, computing
-  Mathematic
-  Others applications
-  Chemistry
-  Physics

About 60 persons from

- ▶ LSCE, CNRGH, IRFU
- ▶ DEN, DRT
- ▶ Paris-Sud university, etc.

- ▶ Roadmap:
 - Pedagogic presentation
 - Seminars
 - Research project
 - Training

• Training

- ▶ For internship persons from LSCE and IRFU
- ▶ Machine Learning technics for cahotics system

• PhD thesis proposition

- ▶ Artificial Intelligence for a gamma-detector for high resolution PET imaging
(*dir. V. Sharyy, IRFU*)

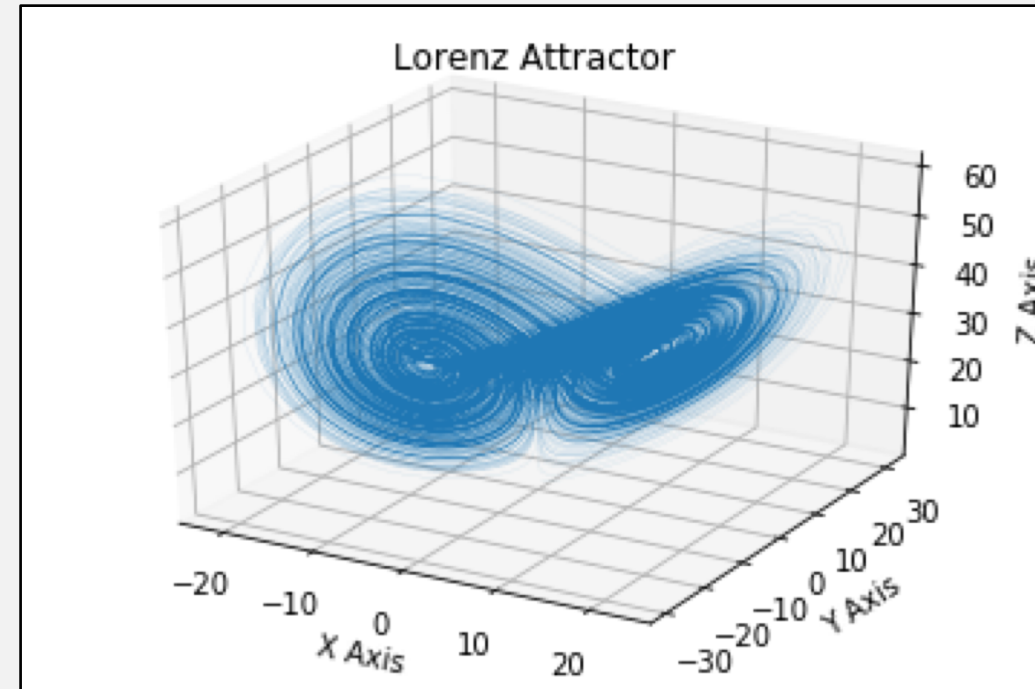
• Web site:

<https://indico.in2p3.fr/event/17858/page/1967-intheart>

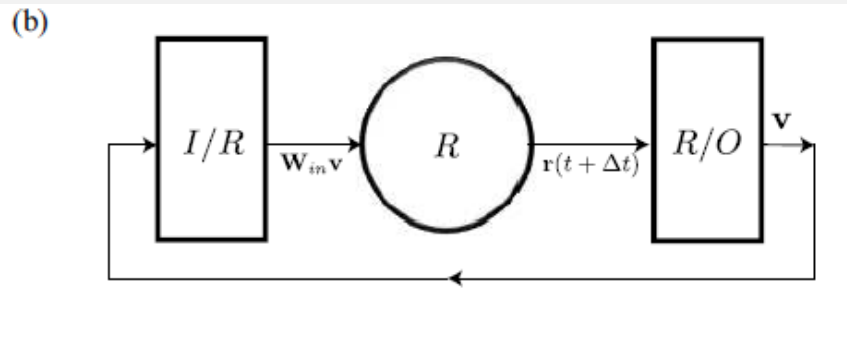
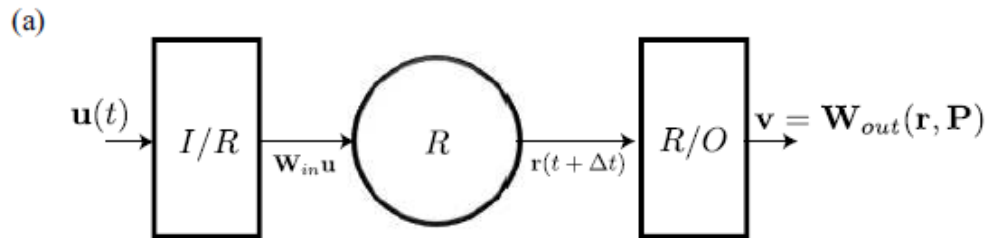
Prediction for chaotic systems

Original paper from Pathak et al., 2017

- **Predict the weather and characterize climate**
 - ▶ Many data recorded, but with some measurement errors
 - ▶ High sensitivity to initial conditions
 - ▶ Contribution of Machine Learning ?
 - ▶ Test on convective cells
 - ▶ Lorenz attractor: atmospheric Toy Model
- **(Re)find the published paper results**
 - ▶ Using Machine Learning technics



Method: Reservoir computing

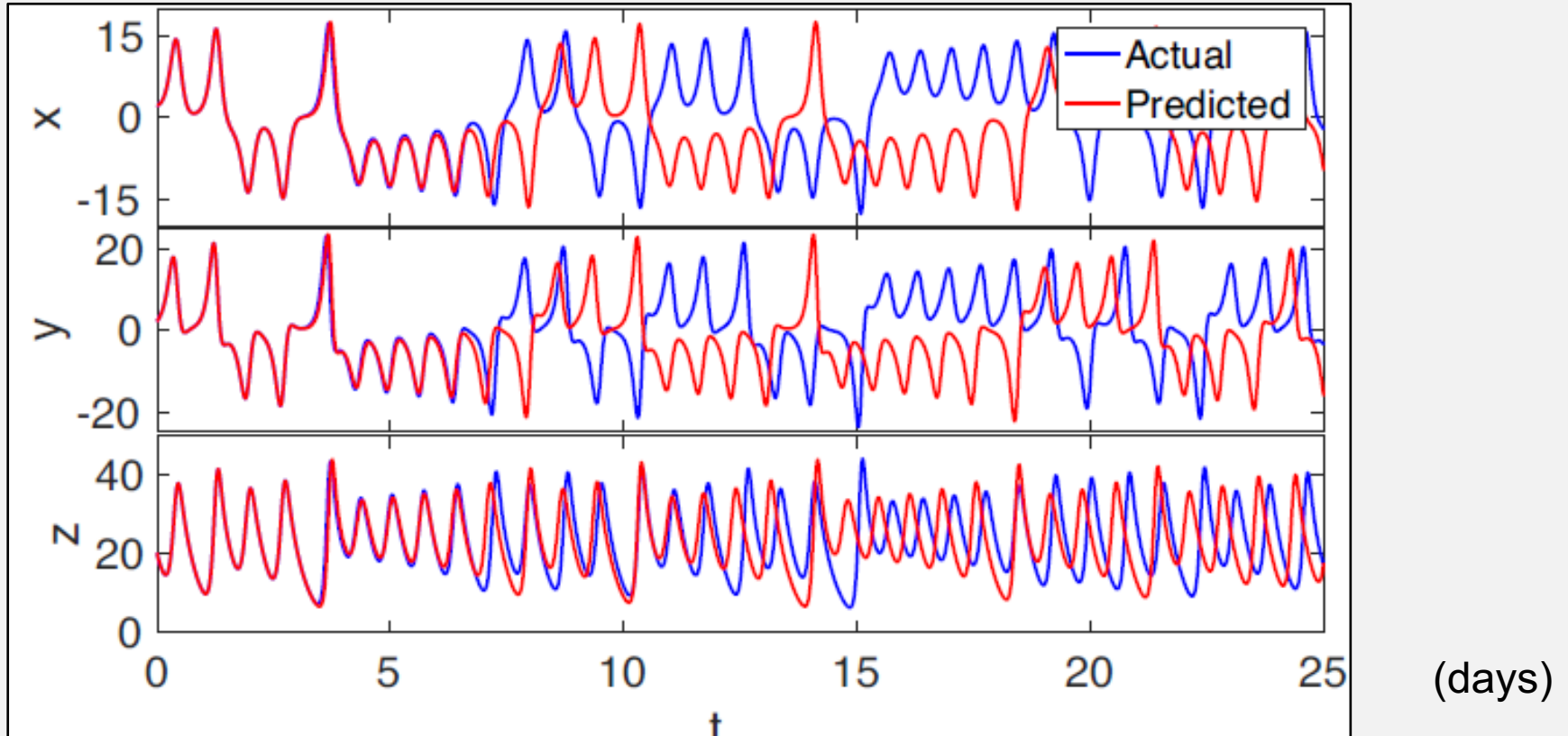


$$\mathbf{r}(t + \Delta t) = \tanh[\mathbf{A}\mathbf{r}(t) + \mathbf{W}_{in}\mathbf{u}(t)]$$

Recurrent neural network

- \mathbf{A} : adjacency matrix randomly generated. Weights the neurons graph.
- $\mathbf{r}(t)$: the states vector at time t . Represents the state of each neuron.
- \mathbf{W}_{in} (matrix): weighting the effects of the inputs on the state.
- $\mathbf{u}(t)$: input at time t , here the real position at time t .
- $\mathbf{v}(t)$: the output at time t , which is the estimation of the position at time t .

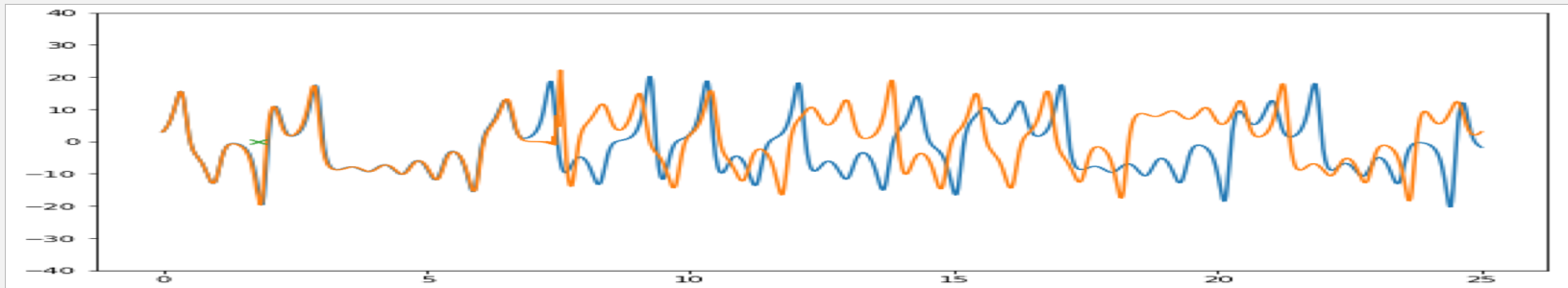
Results from original paper



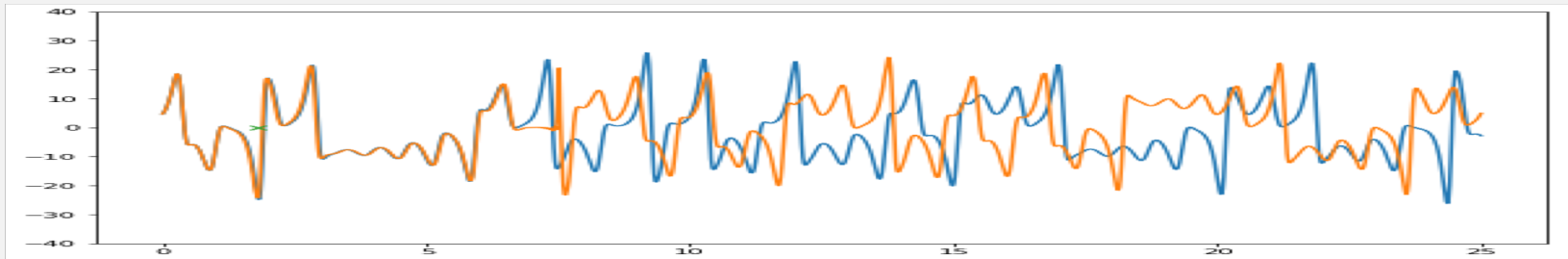
Thanks to the reservoir computing, trajectory approximation of the Lorenz detector

Results obtained with ML technics

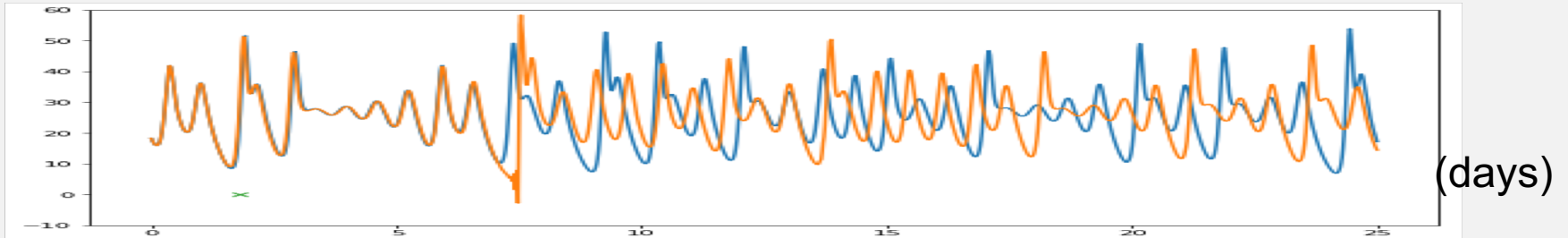
Convection



Temperature



Symmetry

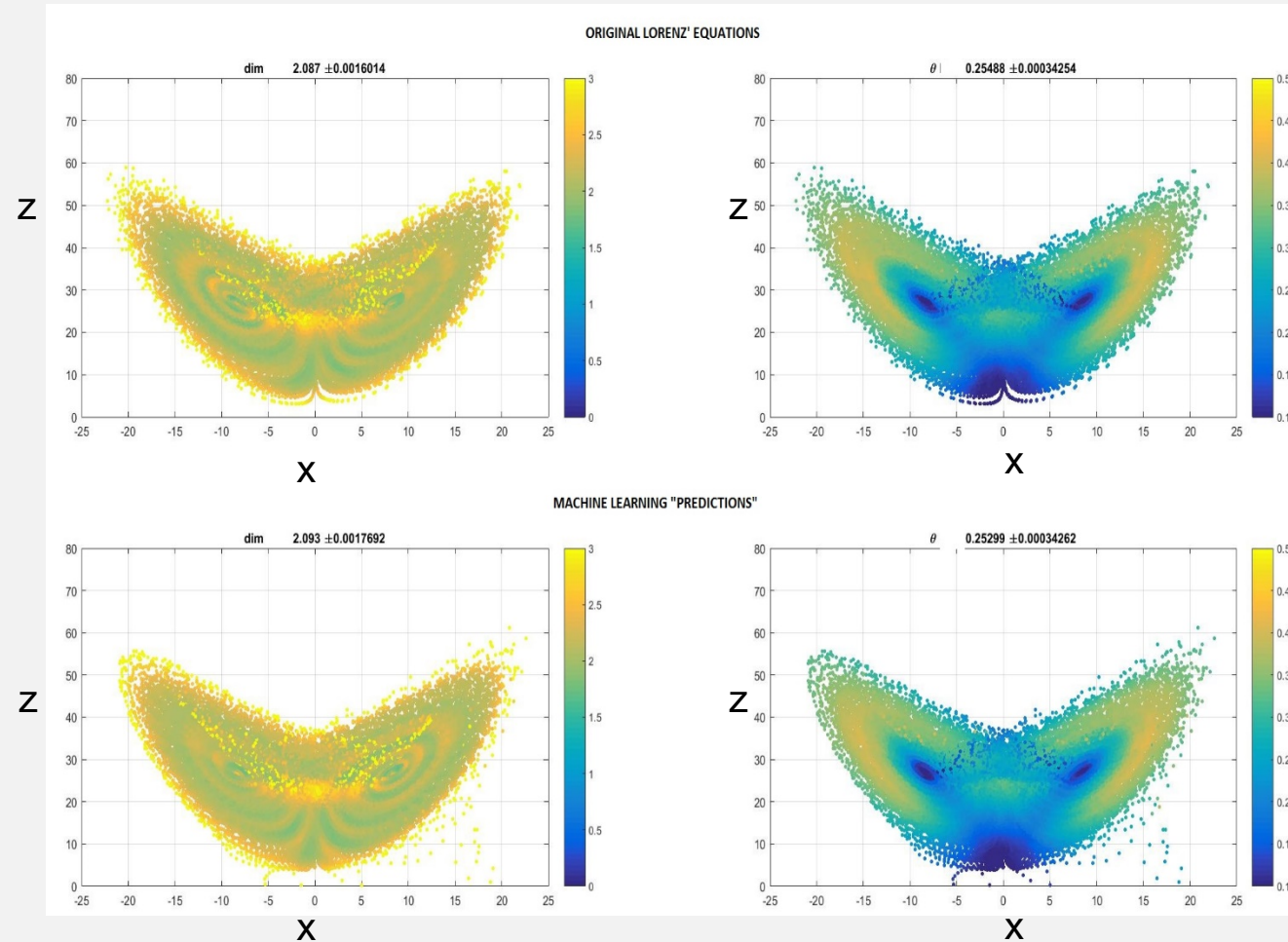


Long term behaviour:

Neural network running several steps after going away from the “real” trajectory

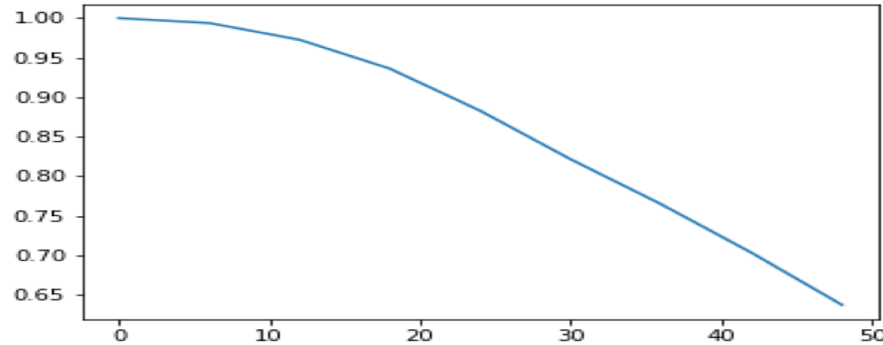
Coherent trajectory observed for X and Z variables

Could we predict long term climatic behavior using Machine Learning ?



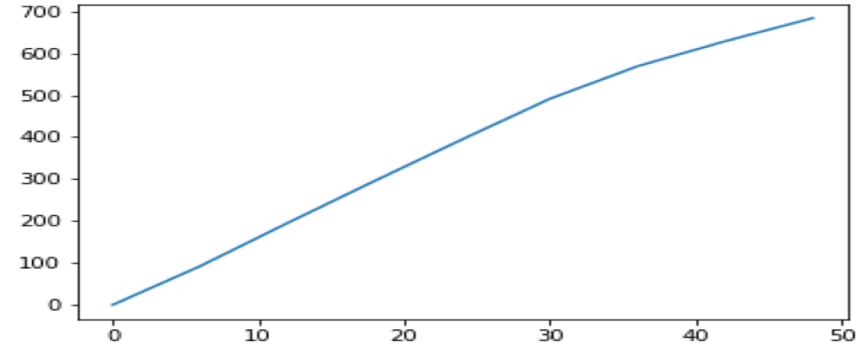
Pressure only

Spatial correlation



Hours

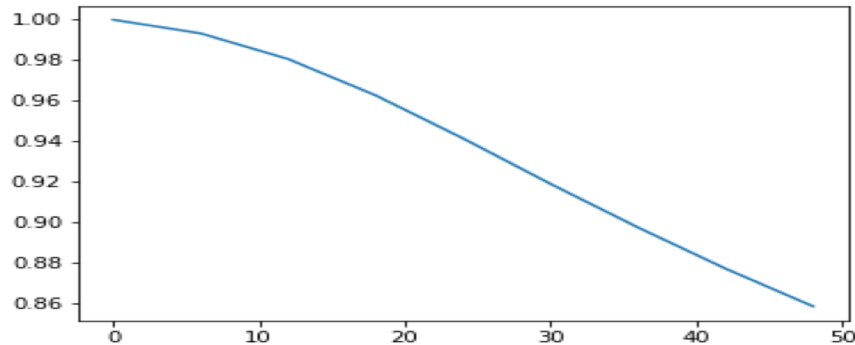
Rmse



Hours

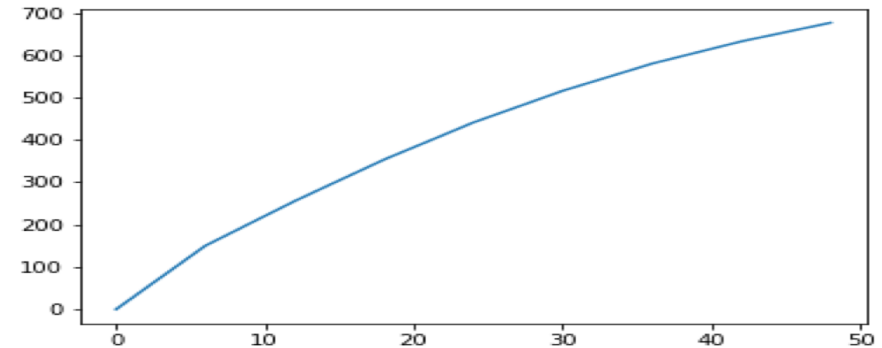
Pressure + temperature

Spatial correlation



Hours

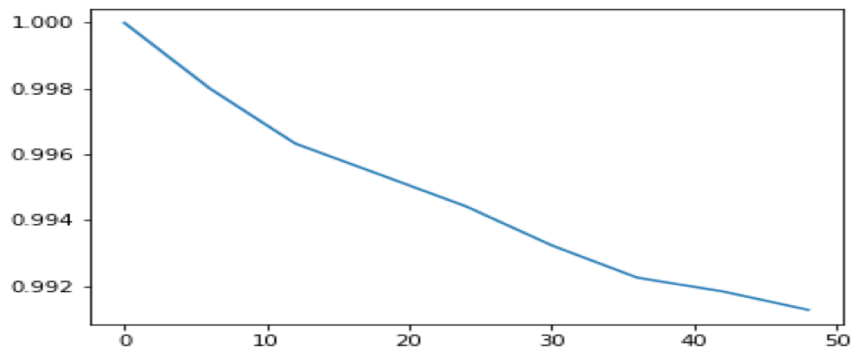
Rmse



Hours

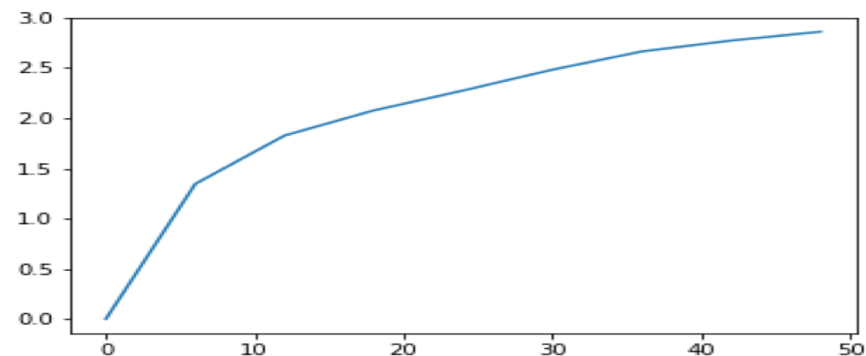
Temperature only

Spatial correlation



Hours

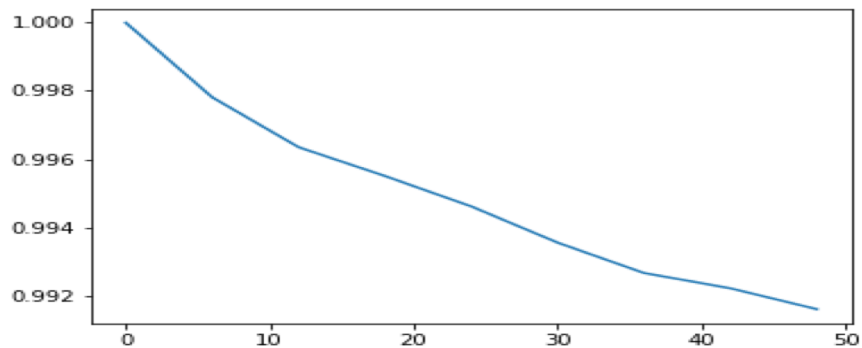
Rmse



Hours

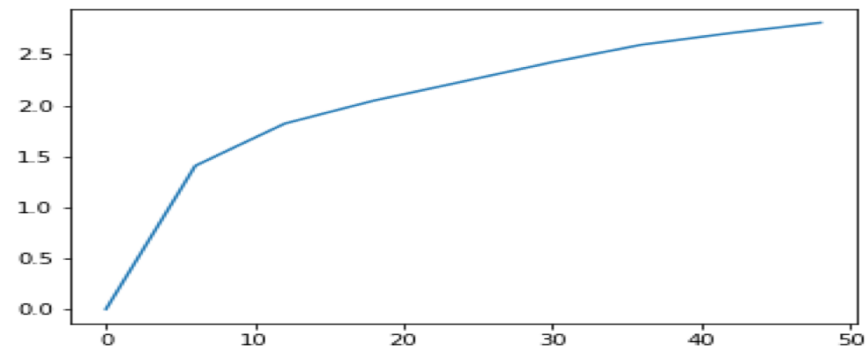
Pressure + temperature

Spatial correlation



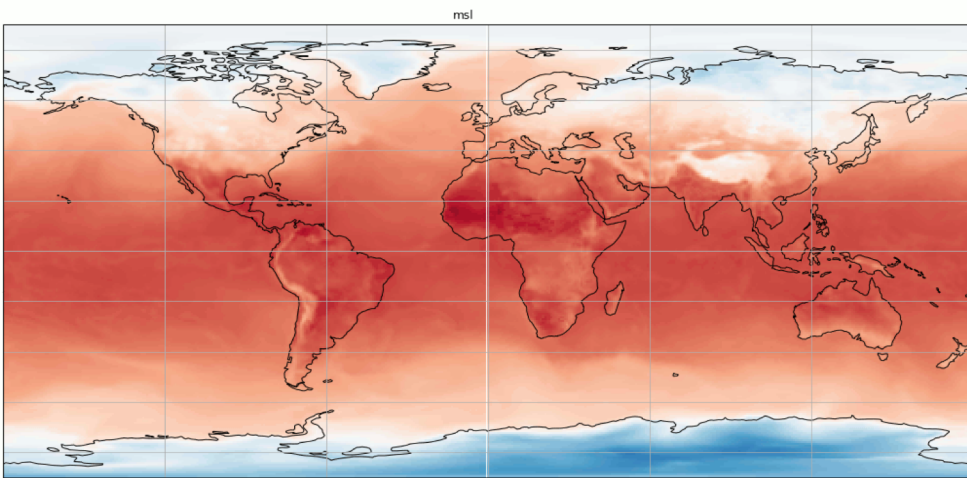
Hours

Rmse

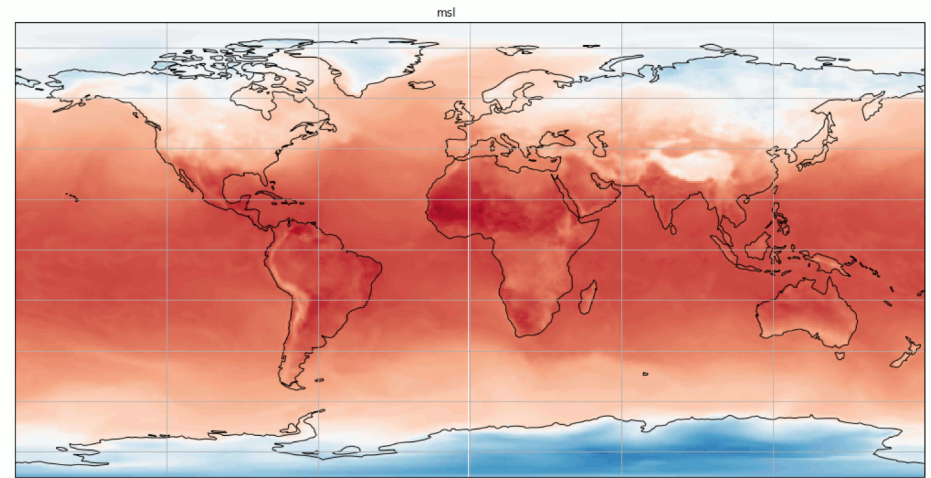


Hours

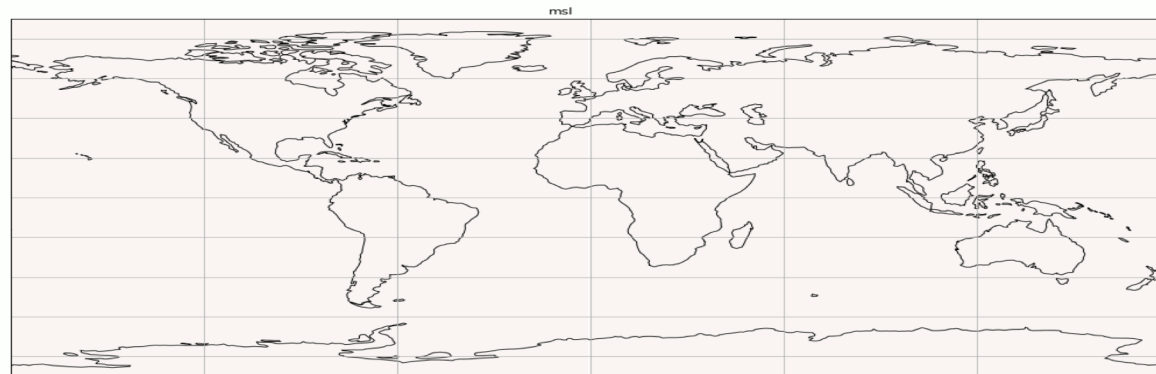
Target



Prediction



Difference: Prediction - Target



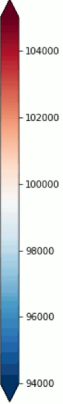
Unit: Kelvins
Duration: 5 days
Data: Temperature

Target

msl

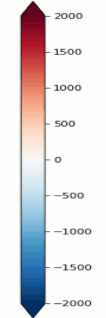
Prediction

msl



Difference: Prediction - Target

msl



Unit: Pascals
Duration: 48h
Data: Pressure

- A motivated and active group from CEA, CNRS and Paris-Sud university
- You are welcome to participate to the group. Contact us !
(to be in our mailing list, please send an email to: valerie.gautard@cea.fr)
- Study of different “use case”
- Try to attract and develop the group on complementary activities:
 - Involvement of geneticists, mathematicians, statisticians, physicists
but lack of chemists and biologists
- Web site: <https://indico.in2p3.fr/event/17858/page/1967-intheart>

