

Data science at EGO/Virgo. Exploring Machine Learning techniques.



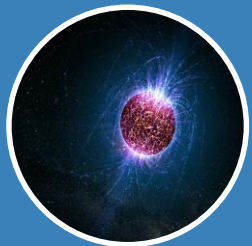
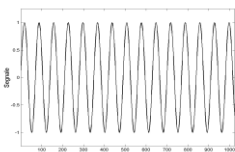
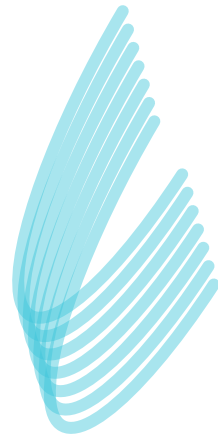
Elena Cuoco

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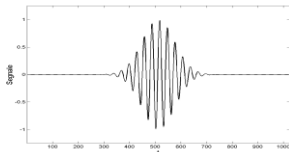


Astrophysical Gravitational wave signals



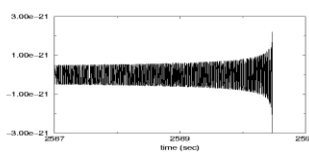
Periodic signals

- Rotating Neutron Stars



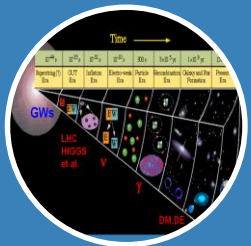
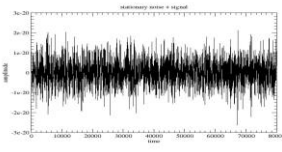
Burst Signals

- Core Supernovae Collapse



Chirp signals

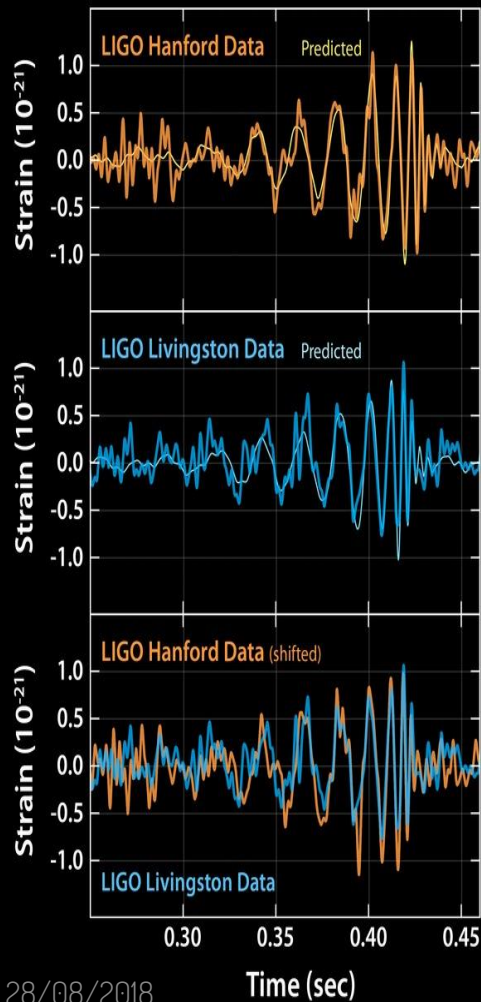
- Compact binary coalescing



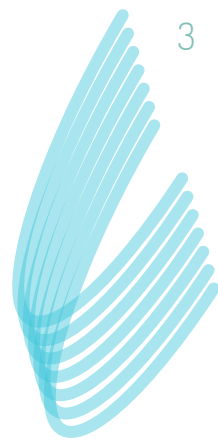
Broad band signals

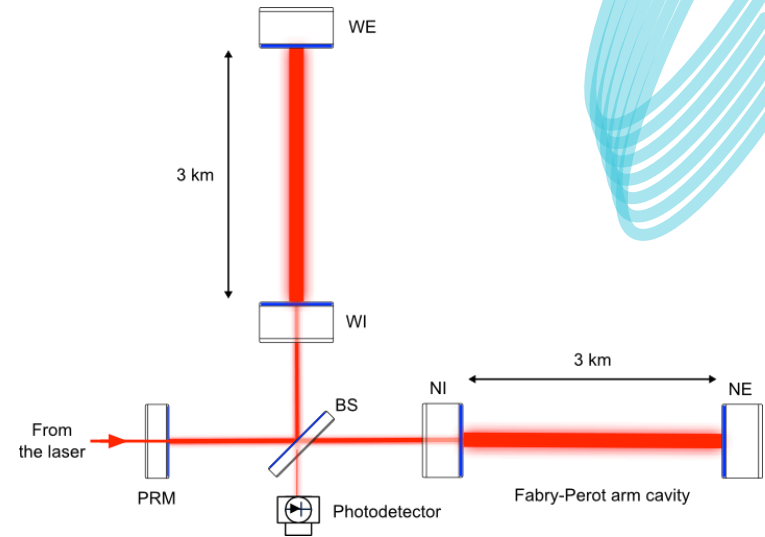
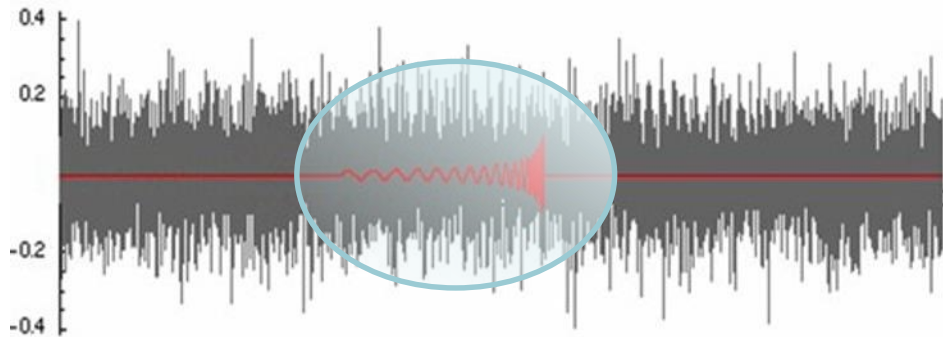
- Cosmic GW background





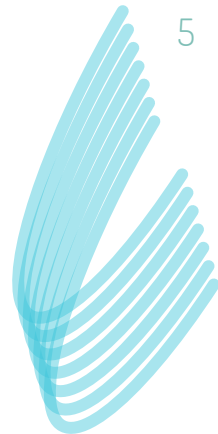
Why Machine Learning in Gravitational Wave research





LIGO/Virgo data

are time series sequences... **noisy time series**
with low amplitude GW signal buried in



Our “signals”

Known GW signals

Compact coalescing binaries has known theoretical waveforms



Optimal filter: Matched filter



Too many templates to test

Unknown GW signals

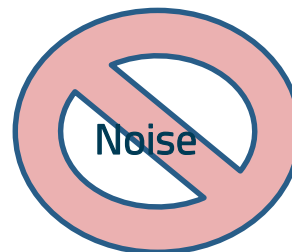
Core collapse supernovae



No Optimal filter



Parameters estimation



Moving lines

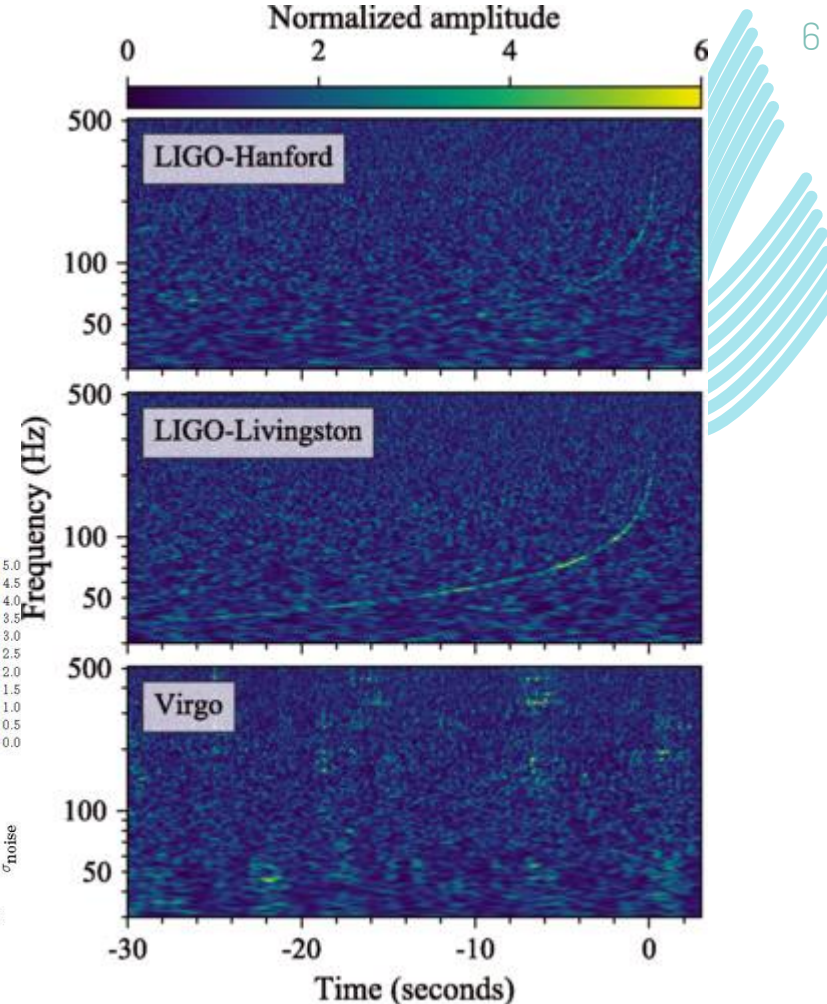
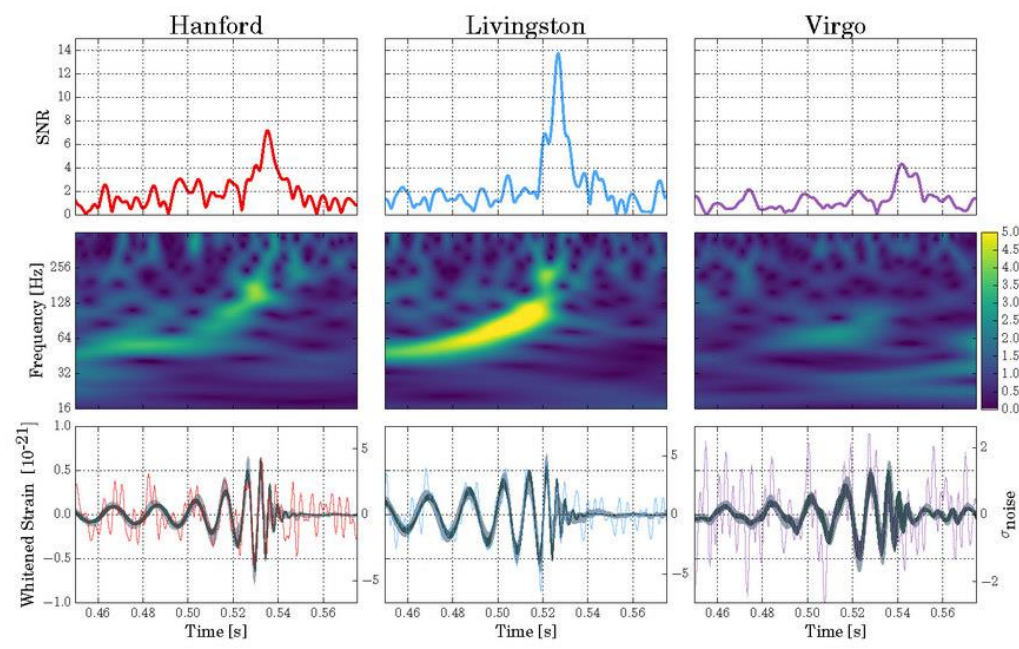
Broad band noise

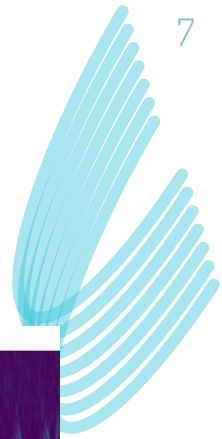
Glitch noise



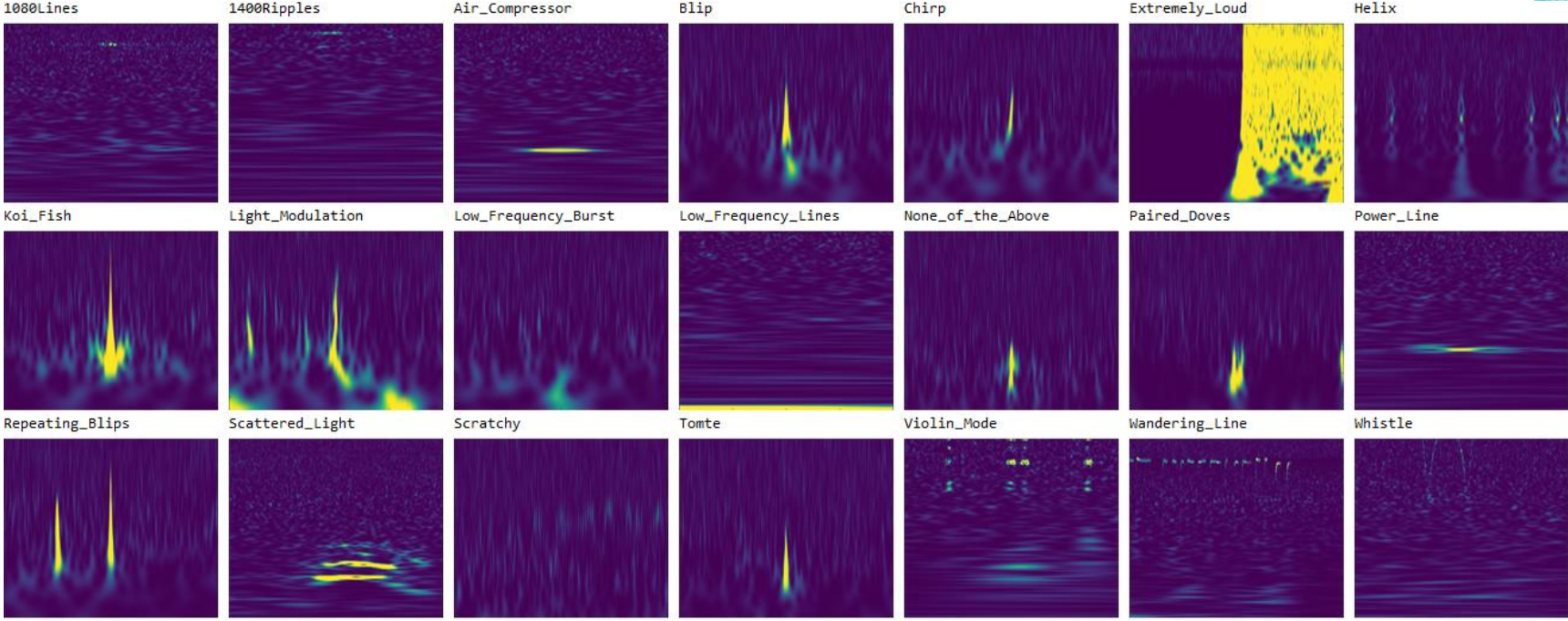
“Pattern recognition”
by visual inspection

Example of GW signals

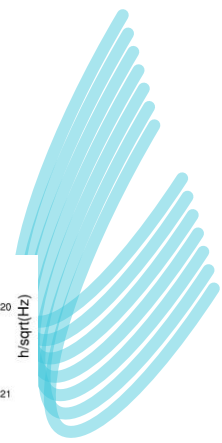




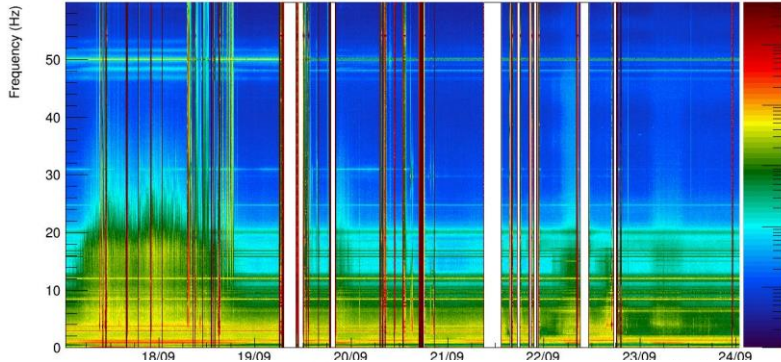
Example of Glitch signals



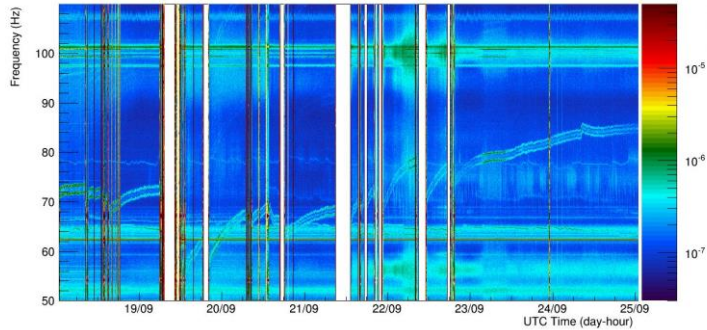
Example of other noise signals



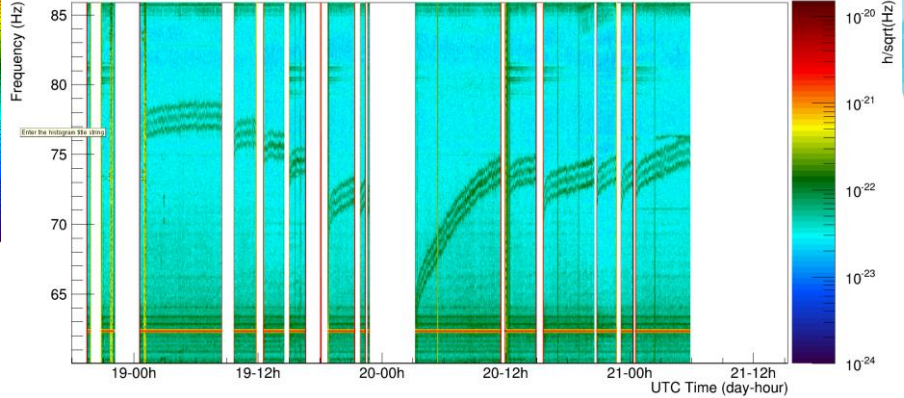
Spectrogram of V1:spectro_LSC_DARM_300_100_0_0 : start=1189644747.000000 (Sun Sep 17 00:52:09 2017 UTC)



Spectrogram of V1:spectro_LSC_PRCL_300_100_0_0 : start=1189731268.000000 (Mon Sep 18 00:54:10 2017 UTC)

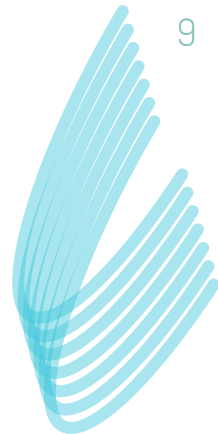


Spectrogram of V1:spectro_Hrec_hoft_20000Hz_300_100_0_0 : start=1210701379.000000 (Fri May 18 17:56:01 2018 UTC)

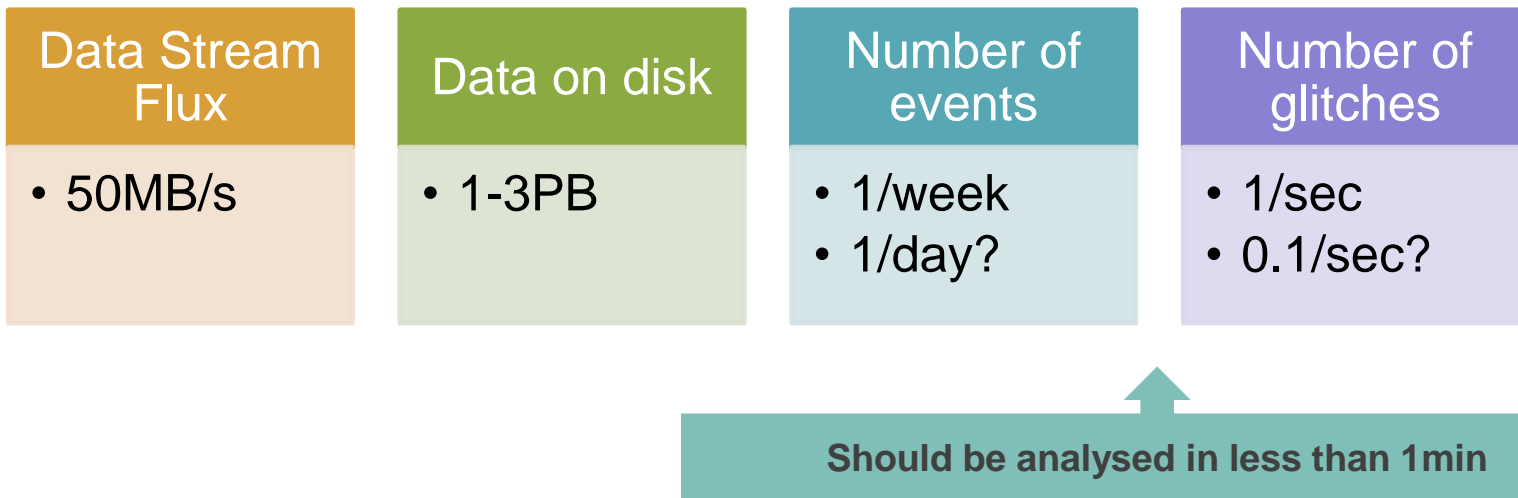


I. Fiori courtesy

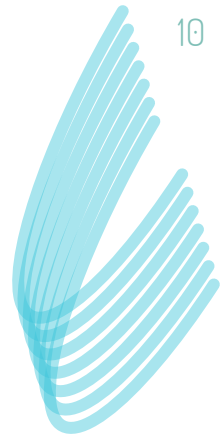
28/08/2018



Numbers about data



How Machine Learning can help

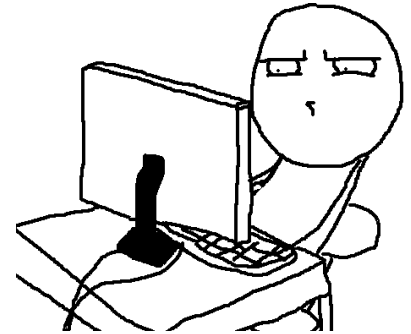


Data conditioning

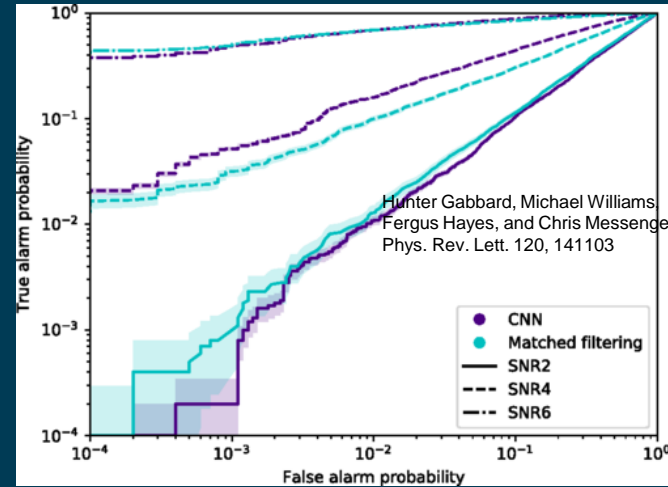
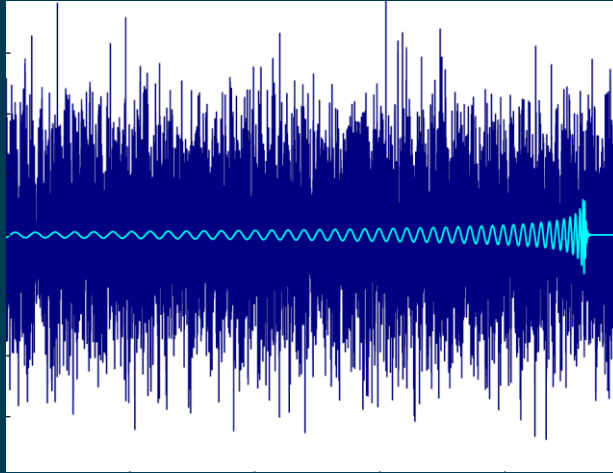
- Non linear noise coupling
- Use Neural Network to learn noise
- Use Neural Network to remove noise

Signal Detection/Classification/PE

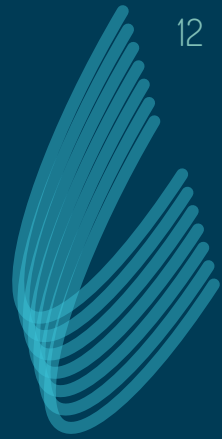
- A lot of fake signals due to noise
- Fast alert system
- Manage parameter estimation



Signal detection



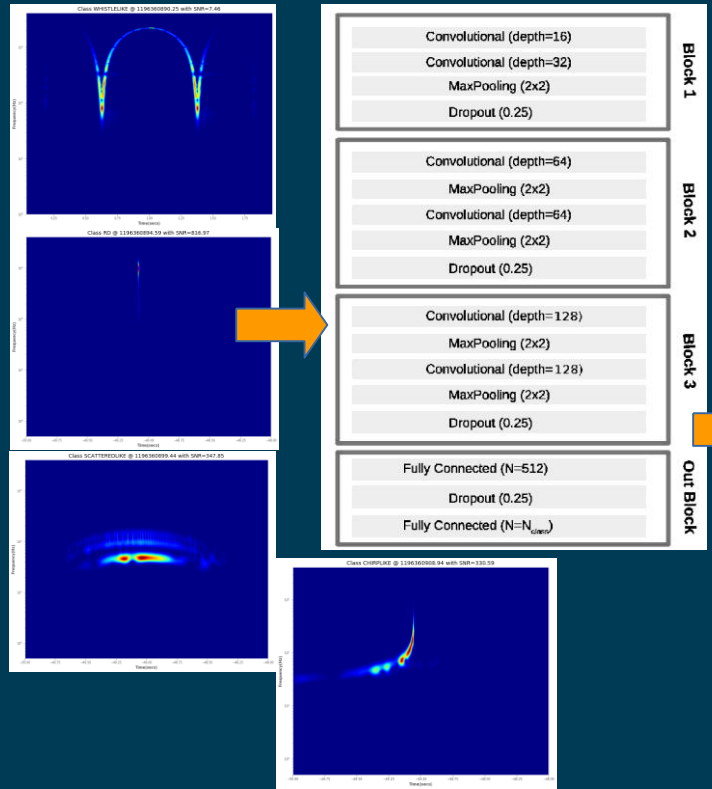
- Deep learning procedure requiring only the raw data time series as input with minimal signal pre-processing.
- Performance similar to Optimal Wiener Filter



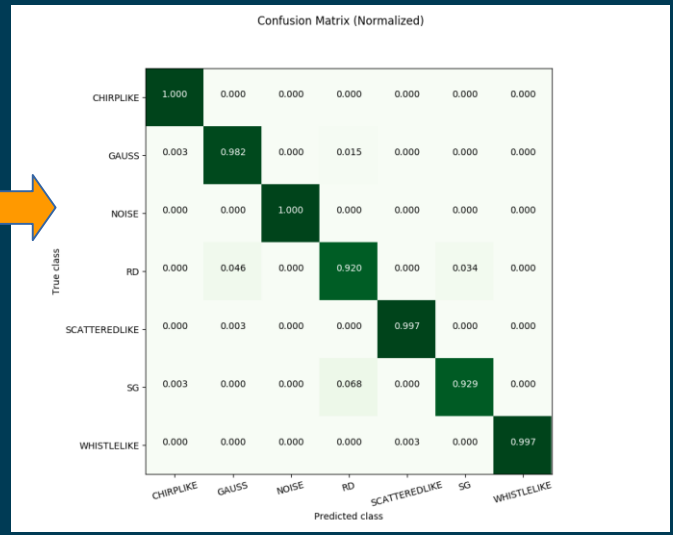
Glitch classification example

Massimiliano Razzano and Elena Cuoco

2018 Class. Quantum Grav. 35 095016



Deep learning with CNN



CA COST Action CA17137

Parties

Action details

MoU	048/18
CSO Approval date	13/04/2018
Start of Action	18/10/2018
End of Action	17/10/2022

Participations

Country	Date	Status
▶ Bosnia and Herzegovina	31/07/2018	Confirmed
▶ Croatia	04/06/2018	Confirmed
▶ France	02/05/2018	Confirmed
▶ Greece	16/05/2018	Confirmed
▶ Hungary	14/05/2018	Confirmed
▶ Ireland	19/07/2018	Confirmed
▶ Italy	09/05/2018	Confirmed
▶ Malta	14/07/2018	Confirmed
▶ Netherlands	06/07/2018	Confirmed
▶ Poland	25/06/2018	Confirmed
▶ Portugal	01/06/2018	Confirmed
▶ Romania	25/06/2018	Confirmed
▶ Serbia	13/07/2018	Confirmed
▶ Spain	13/06/2018	Confirmed
▶ United Kingdom	22/05/2018	Confirmed

Total: 15

Elena Cuoco

COST Association COST Action CA17137

▶ Description

▶ Parties

▶ Management Committee

General Information*

Proposer of the Action:

[Dr Elena Cuoco](#)

Science officer of the Action:

[Dr Ralph STUEBNER](#)

Administrative officer of the Action:

[Ms Rose CRUZ SANTOS](#)

Downloads*

Action Fact Sheet

[Download AFS as .RTF](#)

Memorandum of Understanding

[Download MoU as PDF](#)

G2net: A network for Gravitational Waves, Geophysics and Machine Learning (COST Action 17137)

Main Proposer: E. Cuoco, EGO

G2net: goals of the ACTION

Facilitate conceiving innovative solutions for the analysis of the data of Gravitational Wave (GW) detectors.

Investigate possible solutions to monitor the low-frequency Newtonian noise through the use of adaptive robots.

Train a new generation of young scientists with broad skills in Machine Learning, GW, Control and Robotics.

Investigate new strategies for the handling/suppression of instrumental and environmental noise using Machine Learning techniques.

Bridge the gap between the disciplines of GW physics, geophysics, computer science and robotics