

REINFORCE REsearch INfrastructures FOR Citizens in Europe

Glitch Hunting

with Citizen Science and Machine Learning

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Machine Learning in GW searches EGO 38 – 30 Sept 2022





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REINFORCE How to detect gravitational waves

• Extremely tiny signals

- Typical GW sources induce a deformation of 10⁻¹⁸ m over a length of ~ few km
- High background noise!

Laser interferometers

- Exploiting interference between orthogonal laser beams
- Typical km-long scale + Fabry-Perot cavities
- Frequency range ~20-20000 Hz
- Advanced methods to reduce noise
- Detectors working as a network







Upgrades



- Low frequencies: Newtonian, seismic
- Mid frequencies: thermal processes
- High frequencies: quantum noise

Observing Run

Commissioning





Noise glitches

Noise is not stationary

- Transient events can happen
- Not related to astrophysical source, but local disturbances
- Different timescales/frequency ranges
- Affect data quality, stability and GW detection

Noise hunting & characterization is critical

- Detect and classify glitches to find their origin and remove them
- Hardware/software origin
- Glitches have complex time-frequency morphologies
- Data from auxiliary sensors important to understand origin
- Machine learning offers promising approach (e.g. George&Huerta2017, Razzano&Cuoco 2018)



Glitch in LIGO L1 detector during GW170817 Abbott et al 2017



10

Glitch morphologies sample from Virgo

14 12



-0.8 -0.6 -0.4 -0.2 0 0.2 0.4 0.6 0.8

Time [s]

Scattered Light

1



Virgo strain channel



Many Glitches





Machine Learning approach

- Promising to classify complex time-frequency patterns of glitches (timeseries/images)
- Large input required to train machine learning models
- Various approaches (supervised classification, GANs, clustering...)

• Citizen scientists can help

- Preparing labeled dataset (à la GalaxyZoo)
- Time-frequency spectrograms (images)
- Look at glitches & other noise sources and help characterizing them
- Unveil novel glitch classes
- Success story: Gravity Spy on Zooniverse (2016) by LIGO team and NSF



https://www.zooniverse.org/projects/zooniverse/gravity-spy



- Horizon 2020 SWAFS "Science with and for Society" work program
- Cutting-edge citizen science projects on Frontier Physics Research
- Engage >100k citizens in Large Research Infrastructures in Europe (Virgo, KM3Net, LHC)
- PI S. Katsanevas (EGO)
- Participatory design methodology taking into account the special characteristics of different target groups, their barriers and constraints, their perceptions and biases and their attitudes and knowledge regarding science
- Sonification methodologies to increase the senses & inclusion
- Engaging Activities (workshops, schools, etc)
- Started in November 2019





REINFORCE and **GW**

- Citizen science is already supporting gravitational wave science (GravitySpy)
- REINFORCE focus on 4 projects (aka "demonstrators")



https://www.reinforceeu.eu/

- Demonstrators developed on Zooniverse, the leading platform for citizen science
- Our goal: engaging citizens as active participants
- Output used to train machine learning models, but not only that



Our Challenges

• Be engaging and innovative

Attract GravitySpy fans and more
Offer new challenges in GW science

More data, more fun!

 Auxiliary channels offer new insights into how detectors work

Auxiliary channels are not public.
 Prepared ad hoc REINFORCE-EGO MoA

• As a result, an **updated & more rich dataset** than first dataset

Go beyond glitches

• Flexible workflow, can accomodate other noise features (lines/slowly-varying noise)





GW demonstrator development





Highlights of GWitchHunters

• Introduce a new, original way to power GW research with citizen science

- *GravitySpy* as a success story
- Can we expand this approach?

New frontiers

- $\circ\,$ Go beyond classification tasks \rightarrow Noise hunting
- Not only **glitches**
- Include signals from sensors in the detector ("auxiliary channels")
- Run on mobile devices

GWitchHunters Levels





Citizens Tasks - Classification

• Classify noise features

- $\circ\,$ Similar to GravitySpy
- Data presented as time-frequency spectrograms
- Meant mainly to introduce participants to the problem and train them





Citizens Tasks - Detection

• Localize the noise

- $\circ\,$ Draw rectangles around noise features
- $\circ\,$ Can be extended to non-glitch features





Citizens Tasks – Auxiliary Channels

• Study data from auxiliary sensors

- \circ Find similar morphologies between h(t) and aux
- $\circ\,$ Initial set of 8 channels, discuss with Virgo experts on others to add





Tutorials and accessibility

• Introduce the project to non scientists

- Detailed tutorials
- Field Guide on GW detectors, Virgo, glitches

• Promote engangement

- Implement mobile tasks using Zooniverse app
- Forum to discuss and interact with research team and among citizens
- Multilanguage support
- Self-training
 - Playground level with real-time feedback





GWitchHunters: The big picture





Where we are

Launched in Nov 2021, very good feedback from the citizens community

Some numbers:



Mobile Challenges done quickly: first data release completed in mid Feb



Status of the tasks

Some tasks are in a very good stage



https://www.zooniverse.org/projects/reinforce/gwitchhunters



Engaging citizens

Many initiatives to promote the project

- Winter/Easter challenges
- Dedicated Workshops
- Events
- Courses





Processed data







Machine Learning processing

CNN for automatic classification

• First step of a multi-stage pipeline (including other tasks)

Checks and development on simulations

- Experience from previous works
- Deep, multi-layer structure
- Developed in Python + standard libraries (tensorflow, Keras)
- Run on GPUs





Machine Learning processing

2D vs 1D

- 2D good results, both on simulations and previously labeled real data
- Convolutions robust for multi-glitch situations
- Image-building (little more) time consuming, tried 1D CNN







Razzano&Cuoco, 2018

Talpini&Razzano, 2022

M. Razzano



Preparing the analysis: Citizens scientists performance

- Putative class assigned on majority vote
- Results good for some classes, for other requires more data

Confusion Matrix of Level 3 - Watch out the sensors!

Others	56%	6%	4%	4%	2%	3%	14%	6%	3%	2%	
Blip	7%	63%	4%	10%	8%	0%	0%	0%	7%	0%	
Extremely Loud	6%	12%	65%	3%	12%	0%	0%	0%	1%	0%	
Helix	9%	16%	2%	49%	6%	1%	1%	0%	16%	0%	
Koi Fish	3%	12%	12%	4%	66%	0%	0%	0%	3%	0%	
w Frequency Burst	14%	0%	3%	1%	0%	61%	20%	0%	0%	1%	
Scattered Light	8%	0%	1%	1%	0%	7%	80%	0%	1%	0%	
Scratchy	13%	0%	0%	0%	0%	2%	3%	71%	0%	12%	
Tomte	5%	6%	1%	15%	5%	1%	2%	0%	66%	0%	
Violin Mode	5%	0%	0%	0%	0%	0%	0%	14%	0%	81%	
	Othe	Blip ars	Ettr	Helij emelij	t toi,	Low	Scall Freque	Scra tered	tom	te Violi	n Mc
	'Y Loud Light										

Data from May 18,2021 to Jul 15 2022

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Citizens performance (II)

Distributions of Scattered Light parameters

- Useful to find glitch parameter distribution
- Better discrimination among glitch classes



Data from May 18,2021 to Jul 15 2022



Citizens performance (III)

- Only 8 channels, many info already
- DARM correlation is a known fact
- Useful to show how glitch with similar morphology correlate with different aux channels (e.g. Scattered Light and Low Frequency Bursts)



Data from May 18,2021 to Jul 15 2022



The road ahead

GWitchHunters project on Zooniverse

- Successfully launched: we keep monitoring the data inflow
- New datasets included, more challenges to come

Data Analysis

- We are gathering data and new datasets will be prepared
- Validating and assessing citizens performance (comparison CNN/human)
- Standard CNN (2D/1D) producing good results
- Working on automatically processing pipeline
- Stay Tuned (& classify glitches!) Not only classifications, exploring regression and localization

Next steps

- Add new glitch datasets & new auxiliary channels
- Offer novel challenges for data exploration & noise hunting 0



REINFORCE REsearch INfrastructures FOR Citizens in Europe

Join the community





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The era of Gravitational Waves

O4 expected to start in December 2022





GW Transient Catalogs

GWTC-1: Abbott et al 2019, Physical Review X, 9, 3 GWTC-2: Abbott et al 2021, Physical Review X, 11, 2 GWTC-2.1: Abbott et al 2021, arXiv: 2108.01045 GTWC-3: Abbott et al 2022, arXiv:2111.03606



Few words on data

Source data

- Real data from Advanced Virgo O3 (for the launch, next steps we plan to include LIGO)
- Main "strain" channel (aka h(t)), also publicly available
- Auxiliary channels, not public but available thanks to Virgo-EGO-REINFORCE MoA

• Datasets

- Timeseries (+ preprocessing, whitening) to Time-Frequency image maps
- Transitioning from "pathfinder" to "production" phase: 2 Data Releases so far
 - DR01 (2021, launch) 2000 glitches from O3a + 8 aux channels
 - DR02 (2022, follow-up) +4000 glitches from O3b + 8 aux channels