

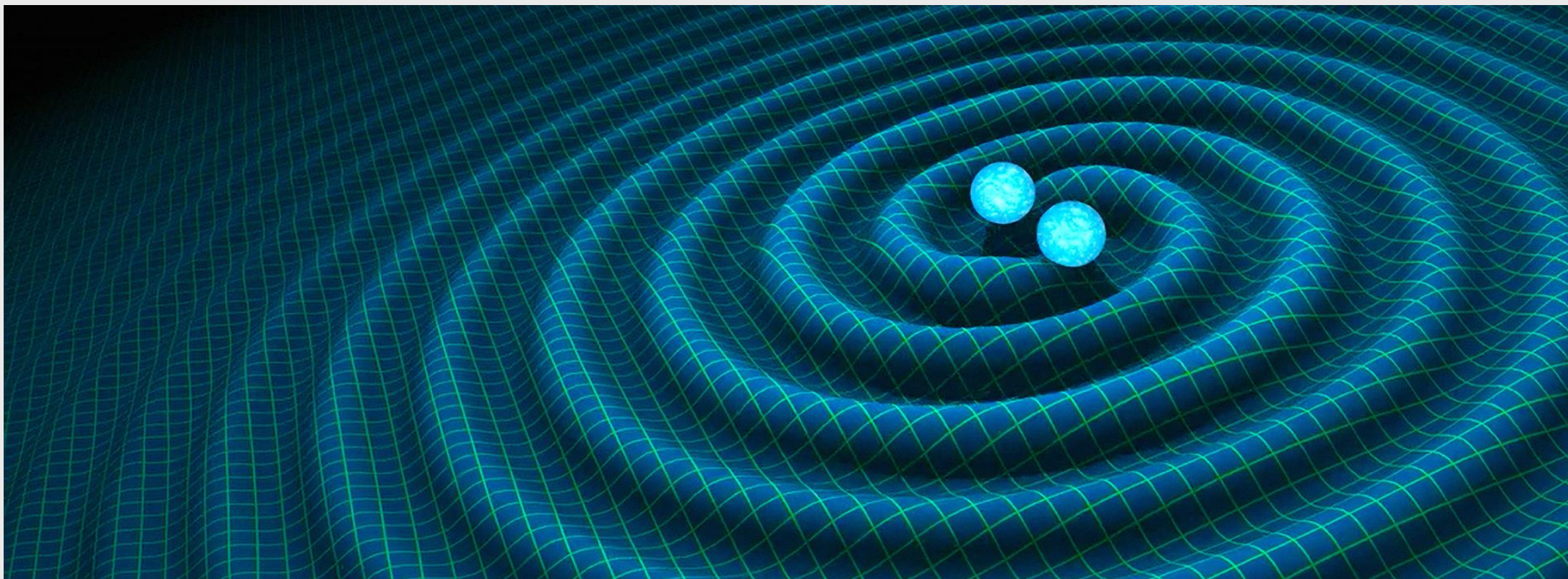


©Johan Jarnestad/The Royal Swedish Academy of Sciences

Einstein Telescope Computing and Challenges

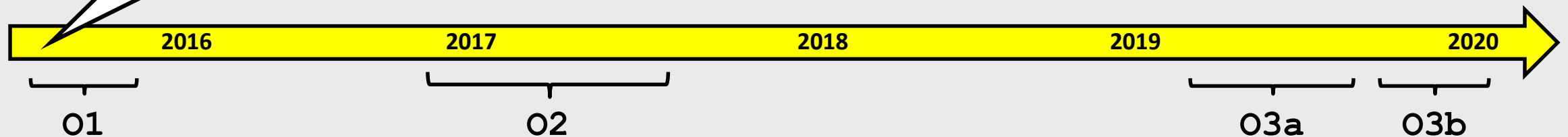
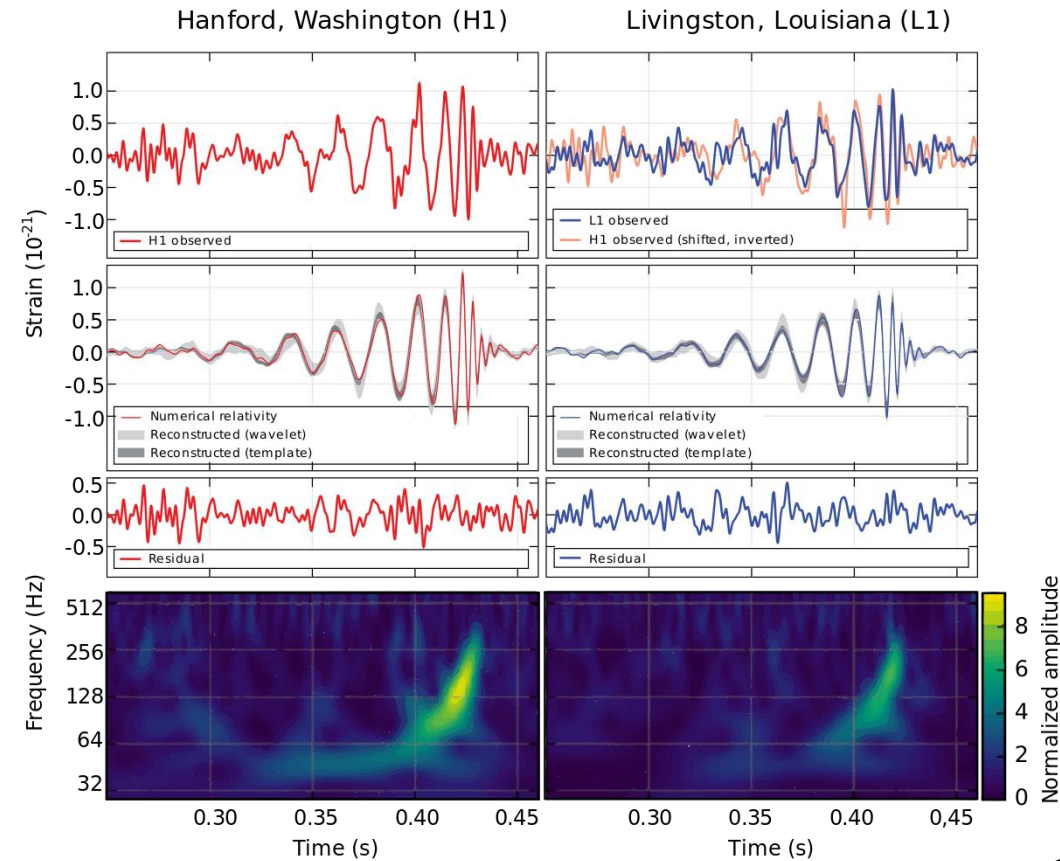
Stefano Bagnasco, INFN

GDB | Jan 12, 2022

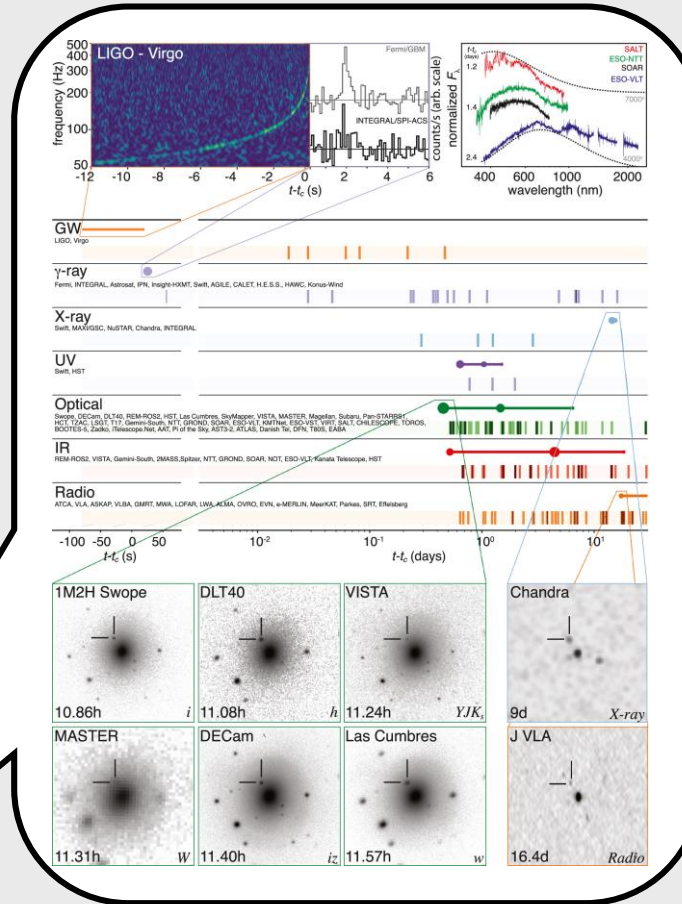
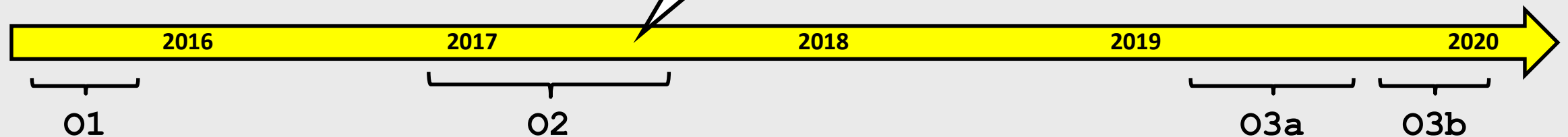


GRAVITATIONAL WAVES RESEARCH

A journey of discovery



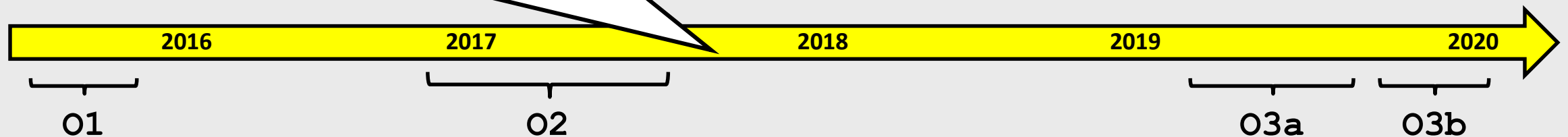
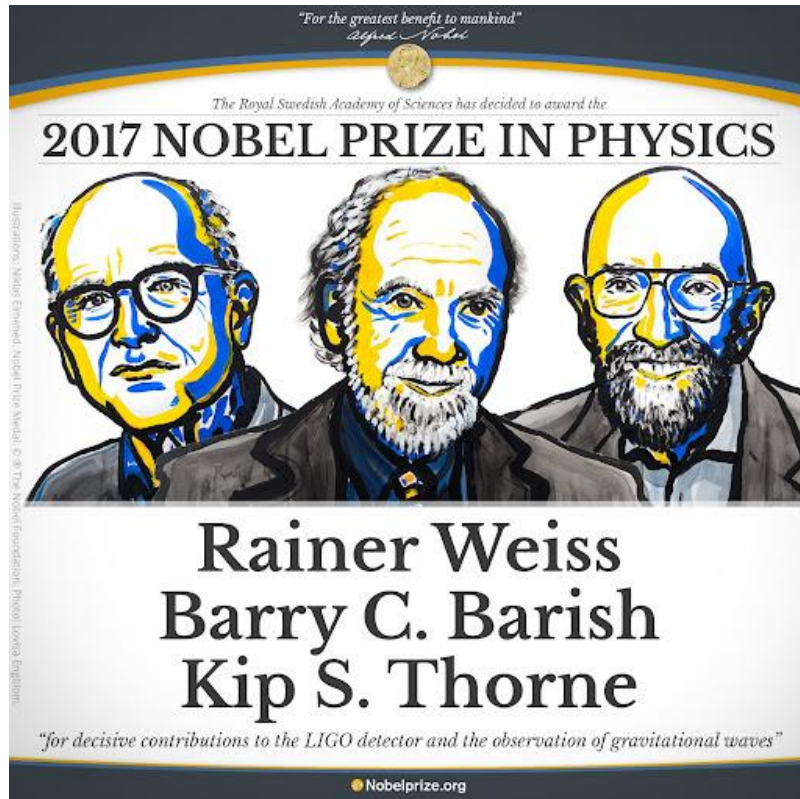
A journey of discovery



"Multi-messenger Observations of a
Binary Neutron Star Merger"

B. P. Abbott et al. 2017 *ApJL* 848
L12 doi:10.3847/2041-8213/aa91c9

A journey of discovery



A journey of discovery

GW190425: large BNS coalescence

Total mass exceeds that of known galactic neutron star binaries

GW190412: asymmetric BBH coalescence

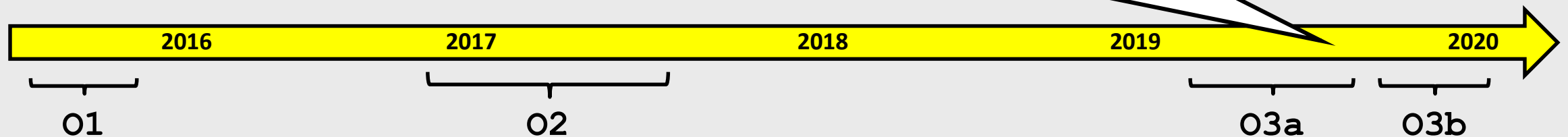
First BBH detection with clear evidence for unequal-mass components and remnant in the "mass gap"

GW190814: most asymmetric CBC

Either the lightest black hole or the heaviest neutron star ever discovered in a double compact-object system

GW190521: first Intermediate Mass BH

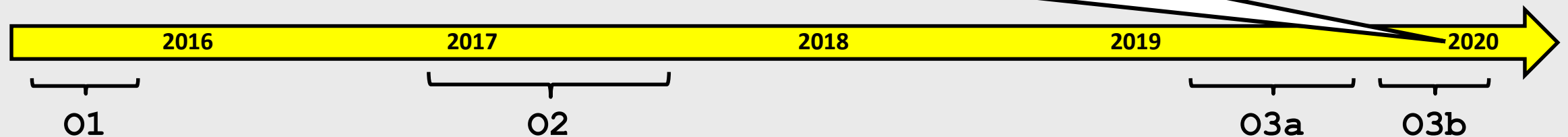
The most massive gravitational wave binary observed to date



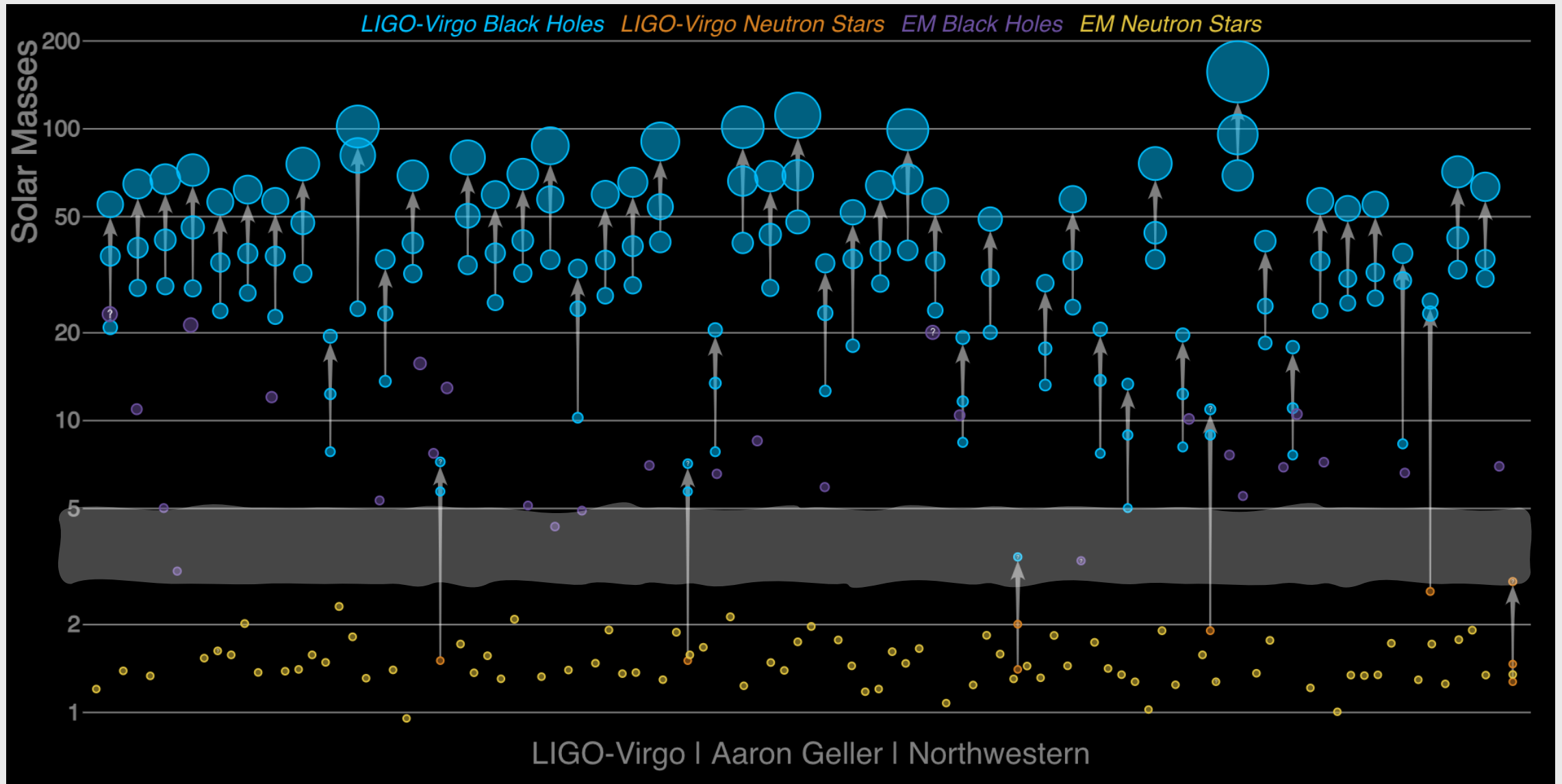
A journey of discovery

GW200105 and GW200115: BHNS

First two (!) coalescences of a neutron star and a black hole



Masses in the stellar graveyard

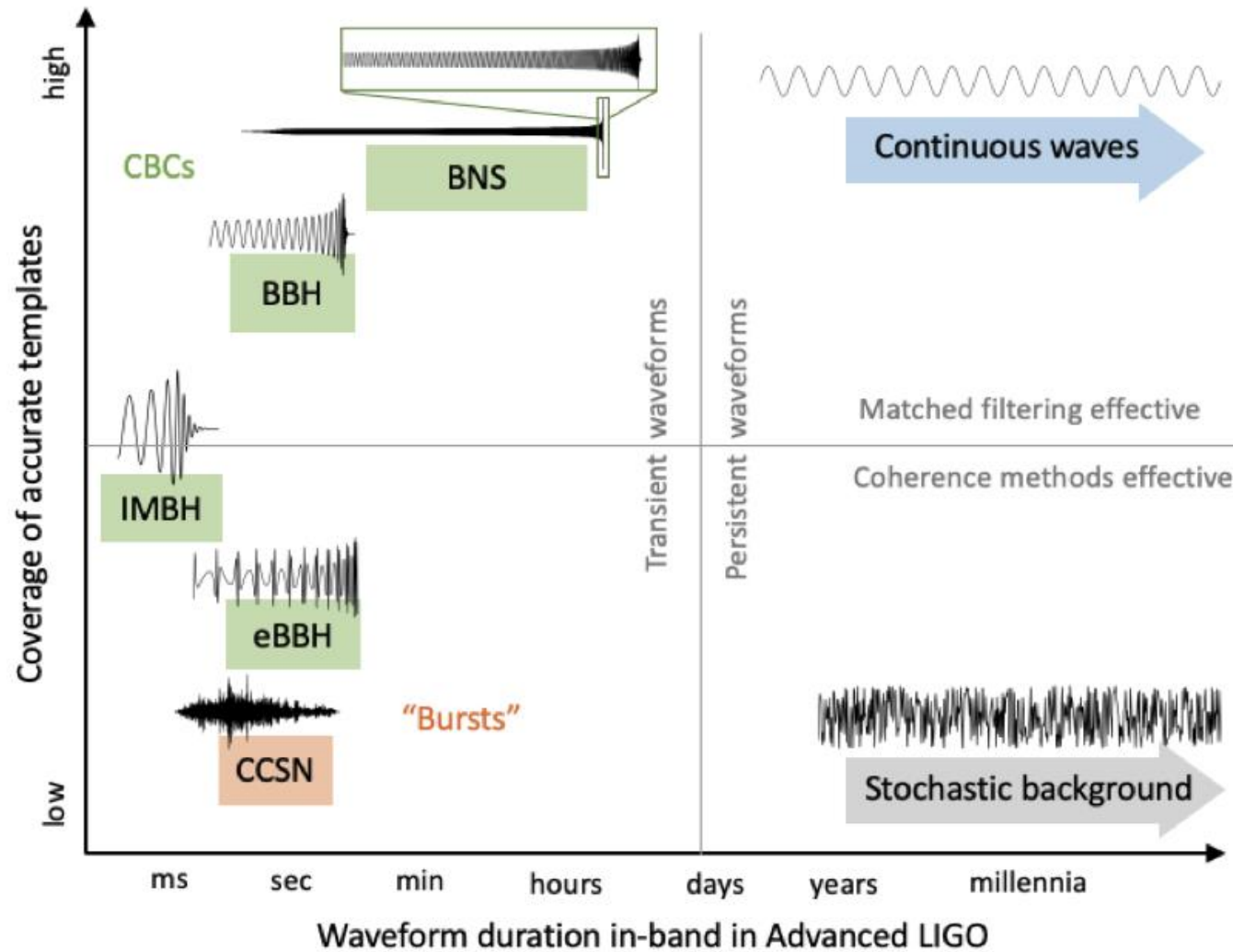


Burst sources:

- CBC: Compact Binary Coalescence
 - Coalescing Compact Binary Systems (Neutron Star-NS, Black Hole-NS, BH-BH): Strong emitters, well modelled for much of the parameter space
- Burst: Unmodeled transient bursts
 - Asymmetric Core Collapse Supernovae: weak emitters, not well-modelled ("bursts"), transient
 - Cosmic strings, soft gamma repeaters, pulsar glitches,...
 - Who knows?

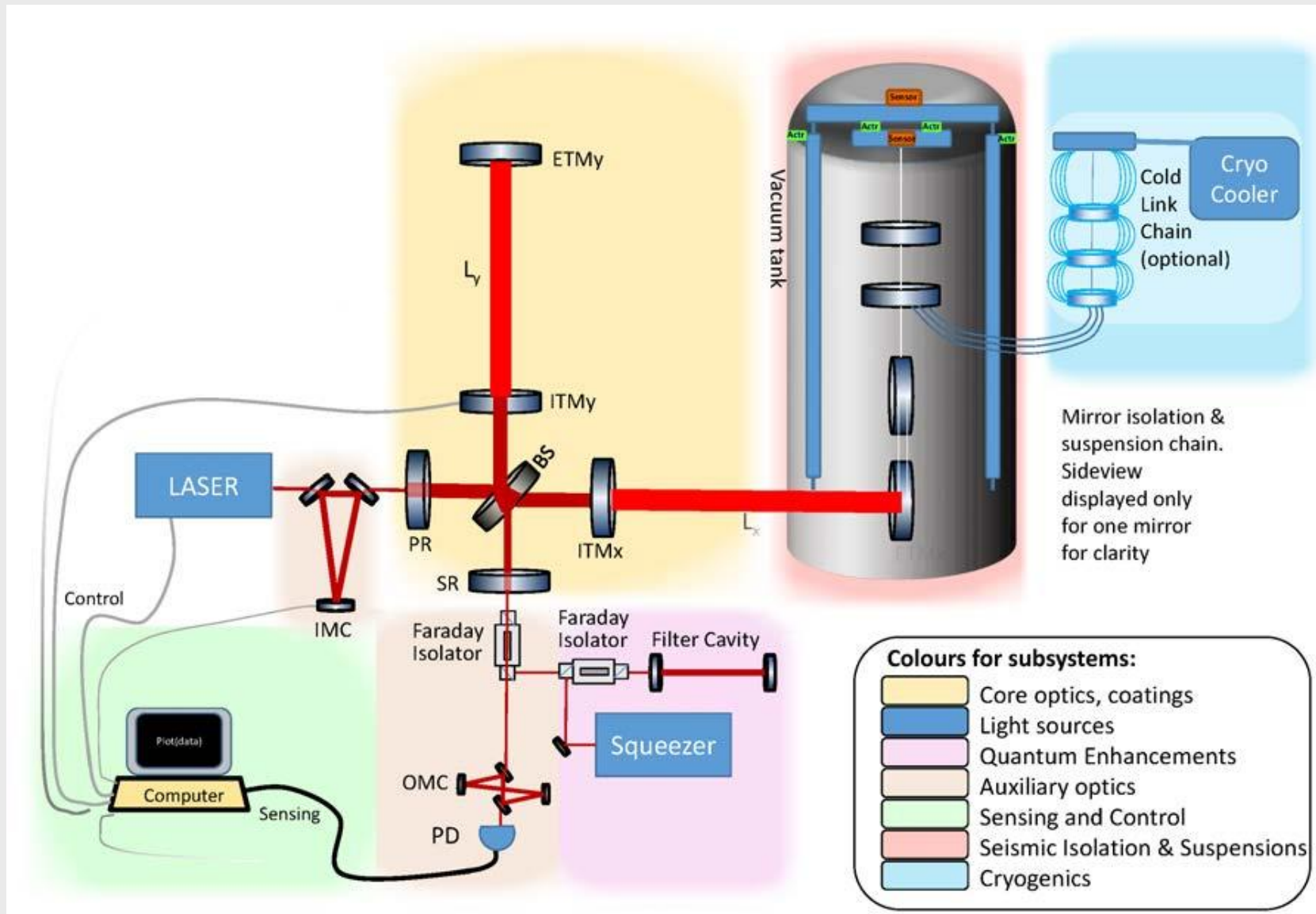
Continuous sources:

- CW: Continuous waves
 - Spinning neutron stars (known waveform, long/continuous duration)
 - All-sky and targeted searches
- SGWB: Continuous stochastic background
 - Cosmological stochastic background (residue of the Big Bang, cosmic GW background, long duration)
 - Astrophysical stochastic background



Jess McIver, D. H. Shoemaker,
*Discovered Gravitational Waves with
 Advanced Ligo, LIGO Document
 P2000530-v1 (2000)*

GW Interferometer

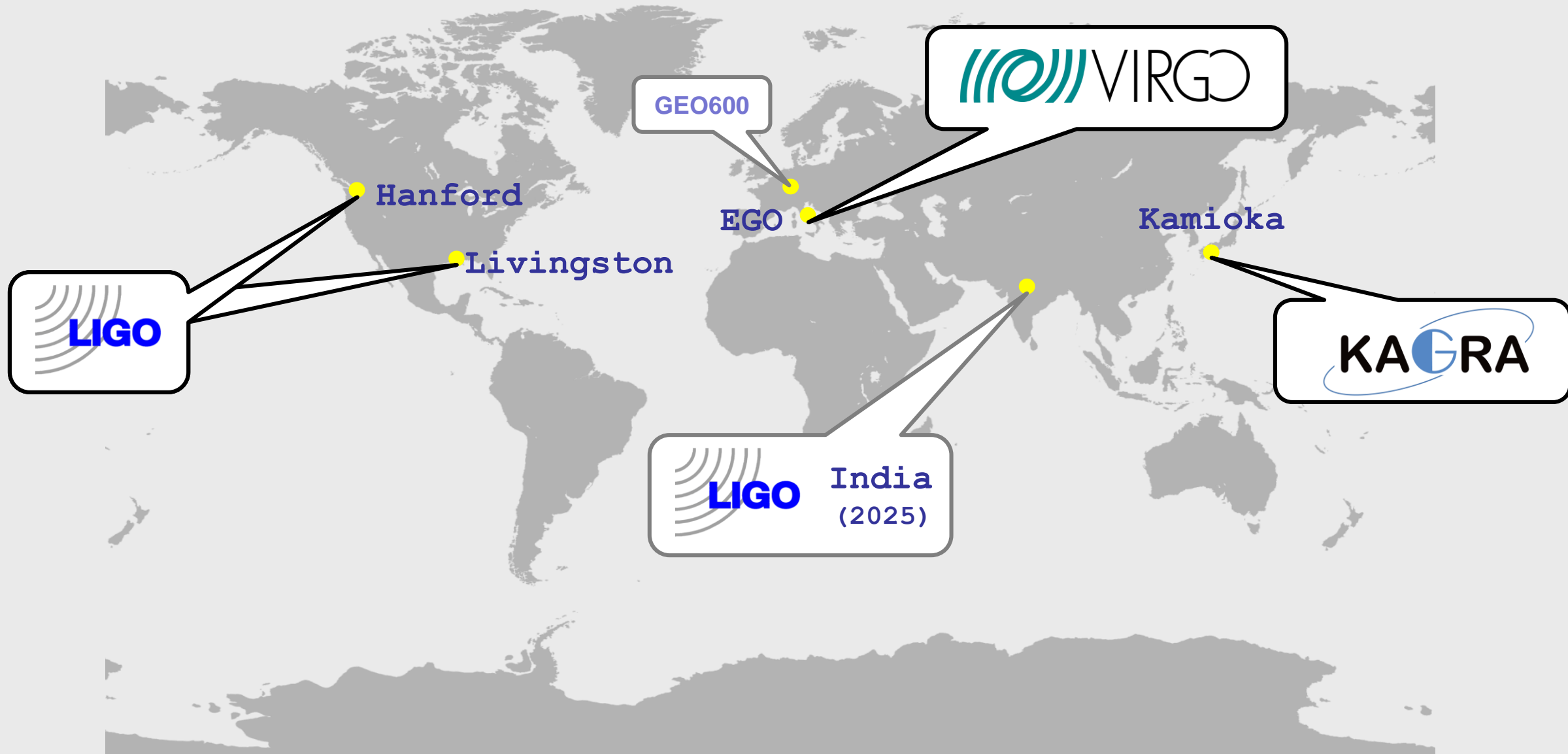


Schematics for information only, not an actual existing instrument!



European Gravitational
Observatory, Cascina (Italy)

A worldwide network



**On-site
infrastructure**

**Plain old HTC
(and some HPC)**

Here's the fun

Online

- Data acquisition and pre-processing
- Instrument control
- Environmental monitoring
- ...

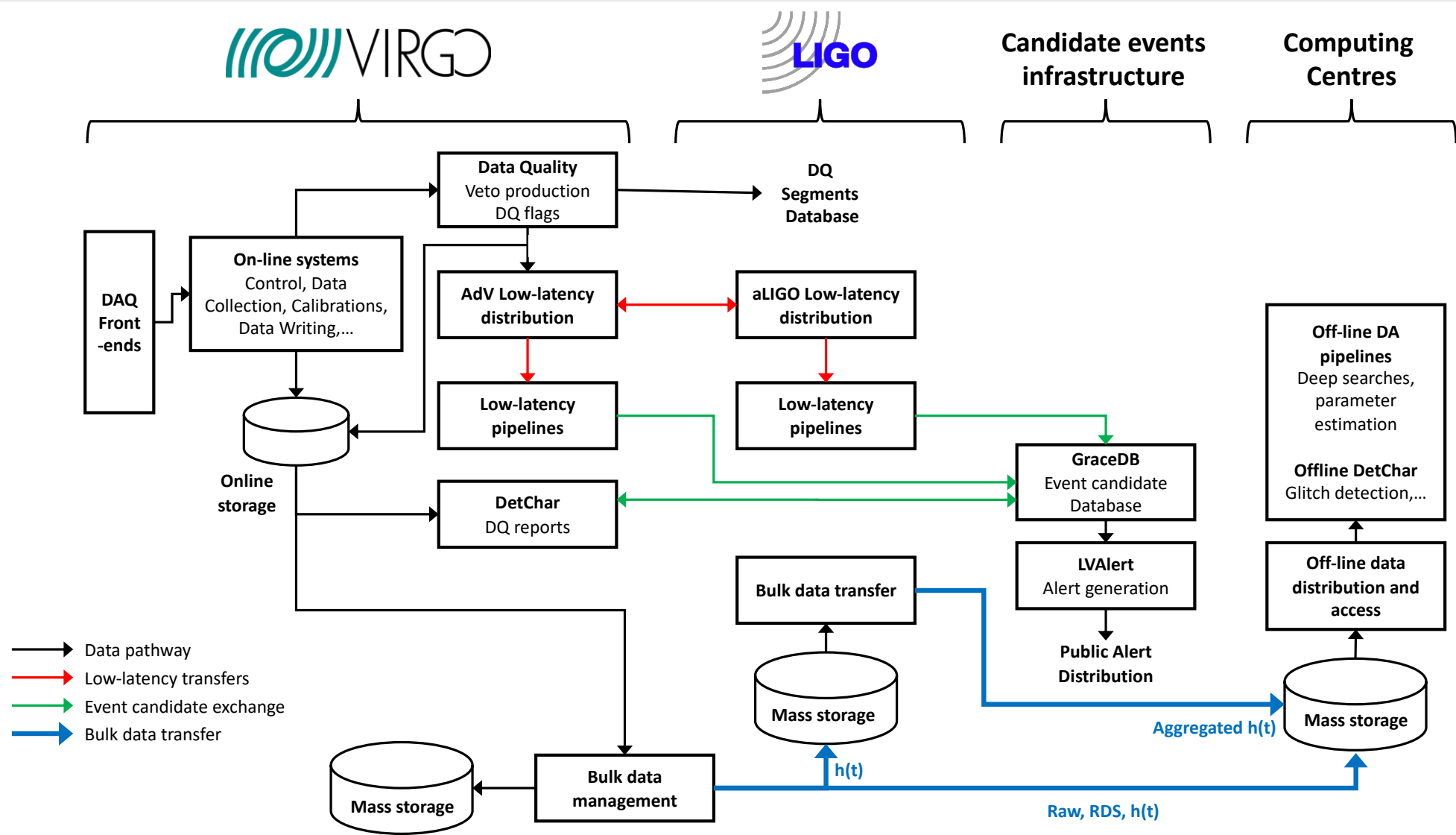
Offline

- Deep searches
- Offline parameter estimation
- (Template bank generation)
- ...

Low-latency

- Candidate search
- Sky localization
- LL parameter estimation
- Alert generation and distribution

Complex overall data flows



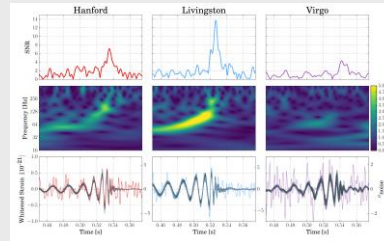
Low-latency searches today

Low-latency searches



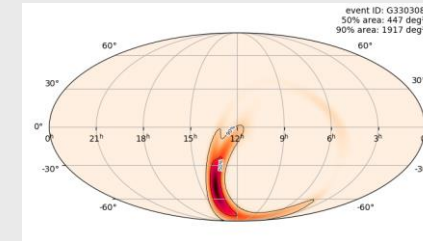
Detector sanity,
Data Quality,
localization,...

GW
candidate

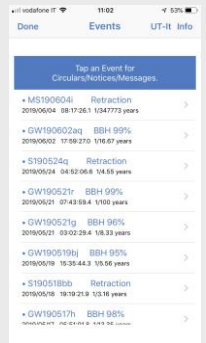


Event
validation

Sky
localization



Public
Alert



A few
minutes

1/2 hour

- Parameter estimation
- GW Candidate Update

Hours,
days

On-site

Off-site

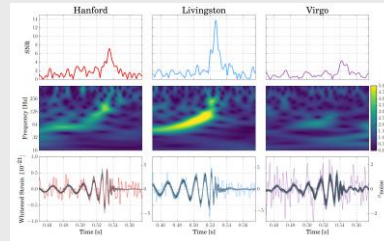
Low-latency searches today

Low-latency searches



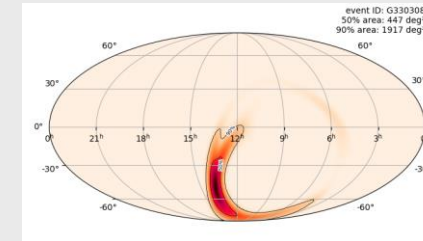
Detector sanity,
Data Quality,
localization,...

GW
candidate

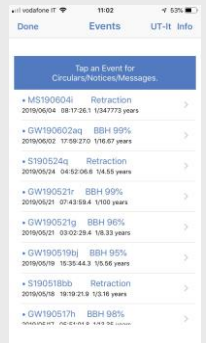


Event
validation

Sky
localization



Public
Alert



A few
minutes

1/2 hour

- Parameter estimation
- GW Candidate Update

Hours,
days

On-site

Off-site



First Demonstration of Early Warning Gravitational-wave Alerts

Ryan Magee^{1,2,3,27}, Deep Chatterjee^{4,5,6,27}, Leo P. Singer^{7,27}, Surabhi Sachdev^{2,3,27}, Manoj Kovalan^{8,9}, Geoffrey Mo^{10,11}, Stuart Anderson¹, Patrick Brady⁶, Patrick Brockill¹, Kipp Cannon¹², Tito Dal Canton¹³, Qi Chu^{8,9}, Patrick Clearwater^{8,14}, Alex Codoreanu^{8,14}, Marco Drago^{15,16}, Patrick Godwin^{2,3}, Shaon Ghosh¹⁷, Giuseppe Greco^{18,19}, Chad Hanna^{2,3,20}, Shashvath J. Kapadia²¹, Erik Katsavounidis^{10,11}, Victor Oloworaran^{8,9}, Alexander E. Pace^{2,3}, Fiona Panther^{8,9}, Anwarul Patwary^{8,9}, Roberto De Pietri^{22,23}, Brandon Piotrzkowski⁶, Tanner Prestegard⁶, Luca Rei²⁴, Anala K. Sreekumar^{8,9}, Marek J. Szczepańczyk²⁵, Vinaya Valsan¹, Aaron Viets²⁶, Madeline Wade²⁶, Linqing Wen^{8,9}, and John Zweizig¹

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Abstract

Gravitational-wave observations became commonplace in Advanced LIGO-Virgo's recently concluded third observing run. 56 nonretracted candidates were identified and publicly announced in near real time. Gravitational waves from binary neutron star mergers, however, remain of special interest since they can be precursors to high-energy astrophysical phenomena like γ -ray bursts and kilonovae. While late-time electromagnetic emissions provide important information about the astrophysical processes within, the prompt emission along with gravitational waves uniquely reveals the extreme matter and gravity during—and in the seconds following—merger. Rapid communication of source location and properties from the gravitational-wave data is crucial to facilitate multimessenger follow-up of such sources. This is especially enabled if the partner facilities are forewarned via an early warning (pre-merger) alert. Here we describe the commissioning and performance of such a low-latency infrastructure within LIGO-Virgo. We present results from an end-to-end mock data challenge that detects binary neutron star mergers and alerts partner facilities before merger. We set expectations for these alerts in future observing runs.

Unified Astronomy Thesaurus concepts: Gravitational waves (678); Gravitational wave astronomy (675); Neutron stars (1108); High energy astrophysics (739)

1. Introduction

The field of gravitational-wave astronomy has exploded in the years following the first direct observation of gravitational waves (GWs) from a binary black hole (BBH) merger (Abbott et al. 2016). Since then, LIGO-Virgo have published 49 candidate events, many of which were identified in low-latency; these include two binary neutron star (BNS) and two neutron

star–black hole (NSBH) candidates (Abbott et al. 2020a). The detection of GWs from compact binaries, especially from BBHs, has become routine. GWs from BNS and NSBH mergers, however, remain rare. BNS and NSBH mergers are of special interest due to the possibility of counterpart electromagnetic (EM) signals. For BNS mergers, in particular, it has long been hypothesized that the central engine (post merger) can launch short gamma-ray bursts (SGRBs; Lattimer & Schramm 1976; Lee & Ramirez-Ruiz 2007), kilonovae (Li & Paczyński 1998; Metzger et al. 2010), and radio waves and X-rays post merger (Nakar & Piran 2011; Metzger & Berger 2012). In the special case of the

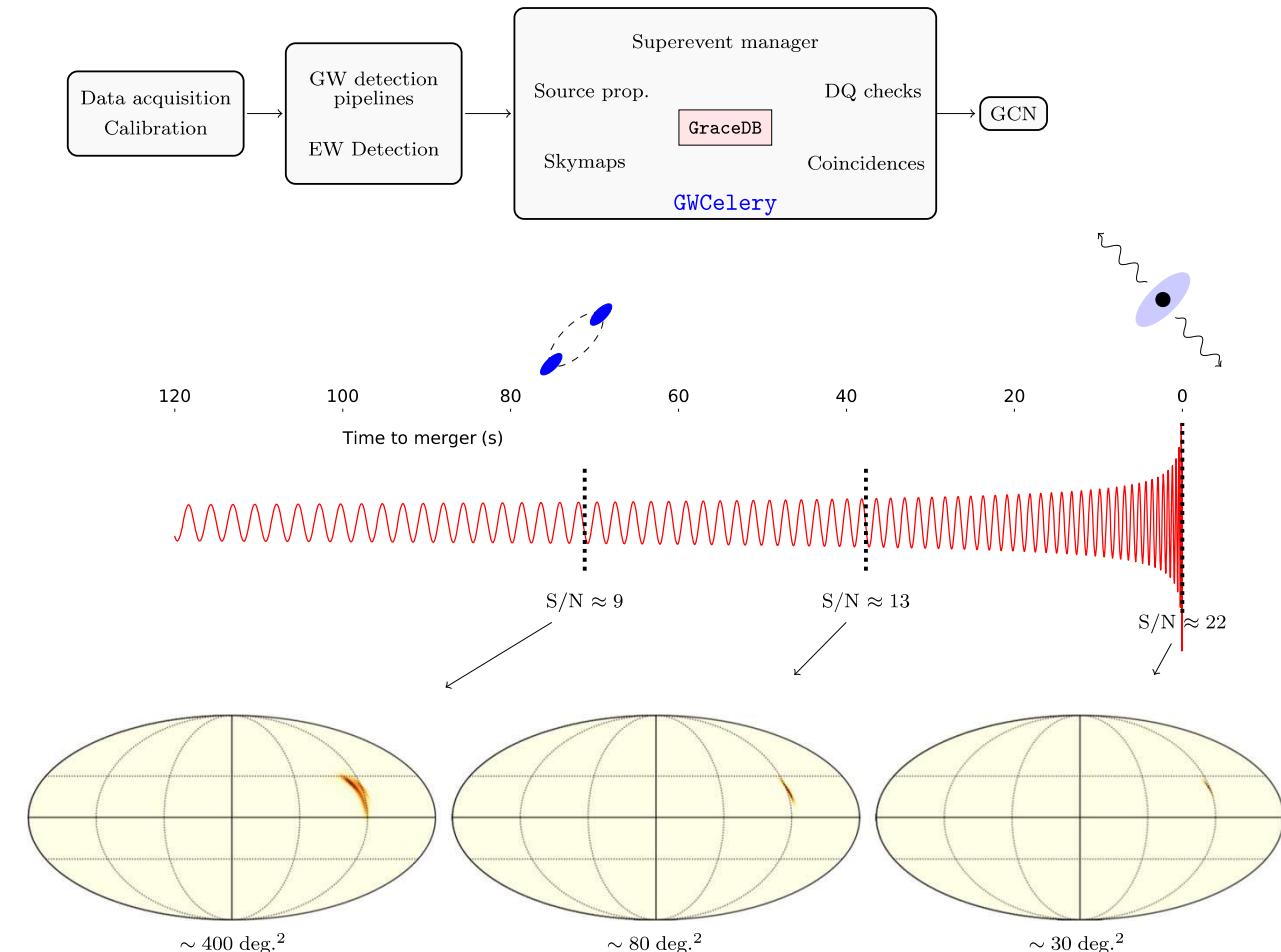
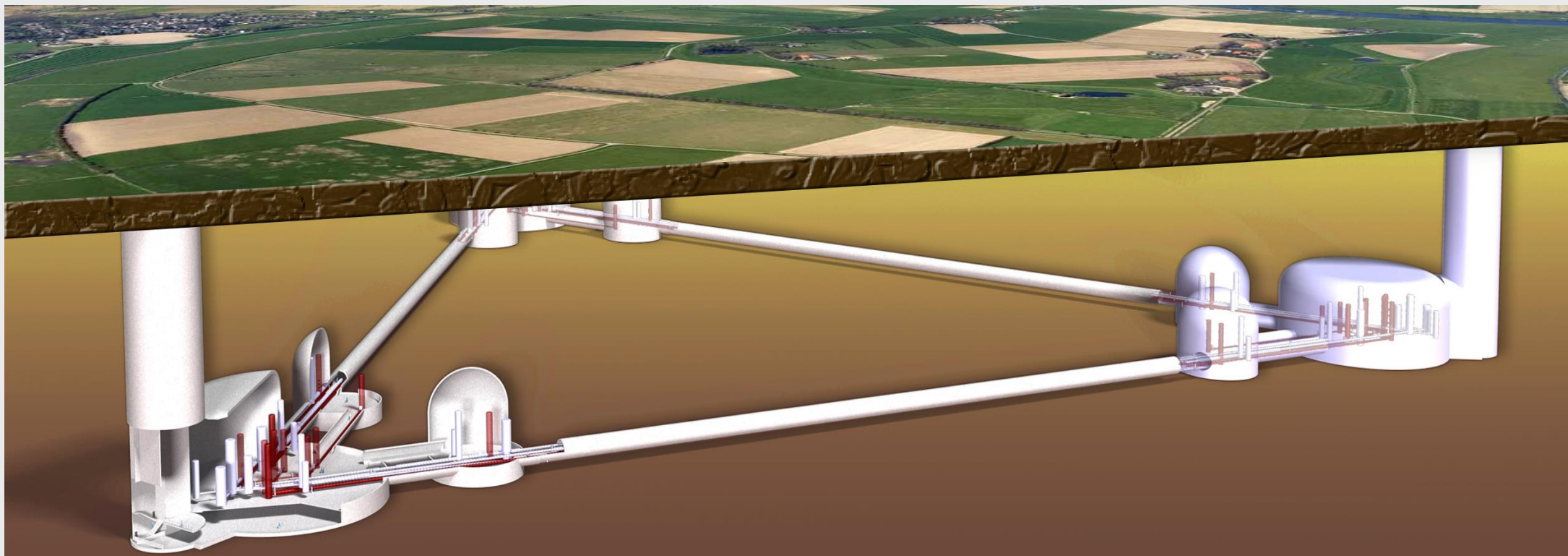


Figure 1. The upper half of the figure illustrates the complete pipeline and interaction of the various (sub)systems, mentioned in Section 2, responsible for disseminating early warning alerts. The waveform evolution with time is shown in the bottom half along with the dependence of the sky-localization area on the cutoff time of the early warning templates and the accumulated S/N during the binary inspiral. The waveforms, time to merger, S/N, and localizations in this figure are qualitative.



COMPUTING FOR ET

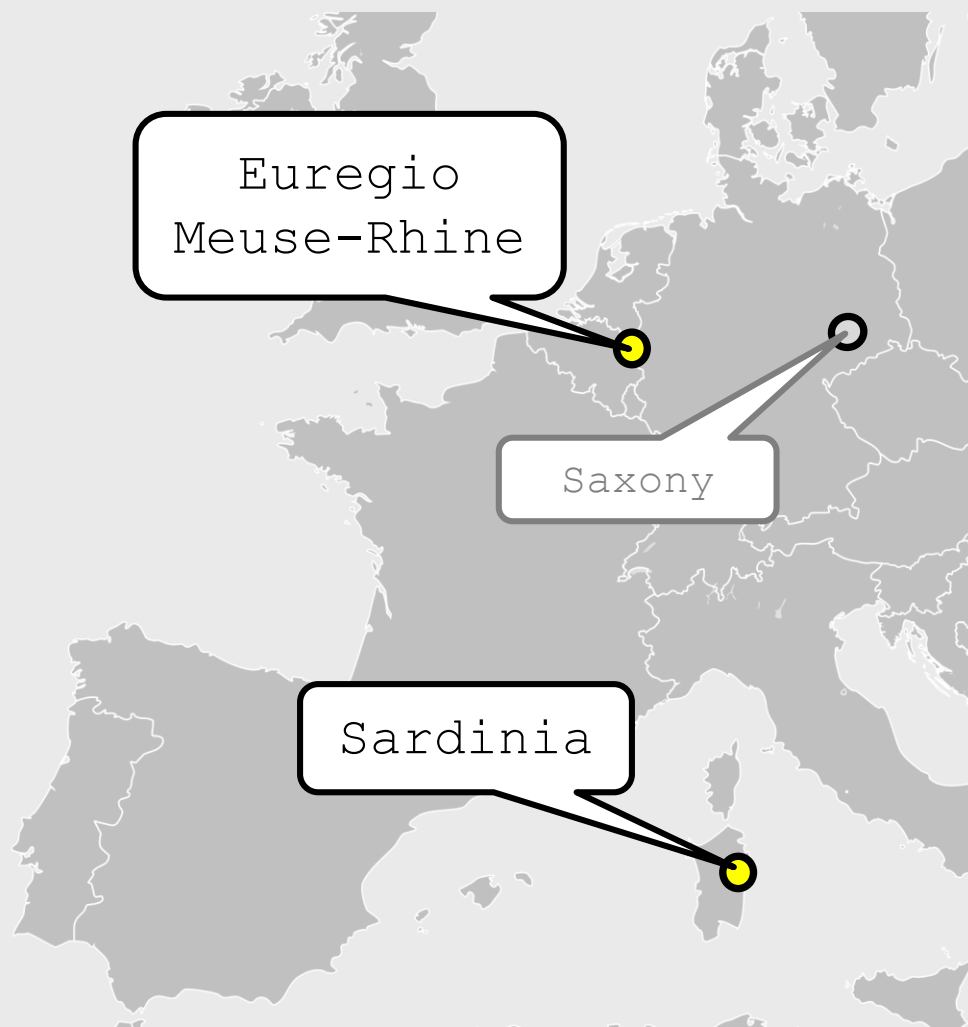
- ET is the project aiming to realise the **European 3rd Generation Gravitational Wave observatory**
- ET defined the concept of 3rd generation GW observatory:
 - A sensitivity at least 10 times better than the (nominal) advanced detectors on a large fraction of the detection frequency band
 - Wideband (possibly wider than the current detectors) accessing the frequency band below 10Hz
 - High reliability and improved observation capability
- After being included in the ESFRI Roadmap, ET is now becoming also a (formal) scientific collaboration

ASTROPHYSICS

- Black hole properties
 - origin (stellar vs. primordial)
 - evolution, demography
- Neutron star properties
 - interior structure (QCD at ultra-high densities, exotic states of matter)
 - demography
- Multi-band and -messenger astronomy
 - joint GW/EM observations (GRB, kilonova,...)
 - multiband GW detection (LISA)
 - neutrinos
- Detection of new astrophysical sources
 - core collapse supernovae
 - isolated neutron stars
 - stochastic background of astrophysical origin

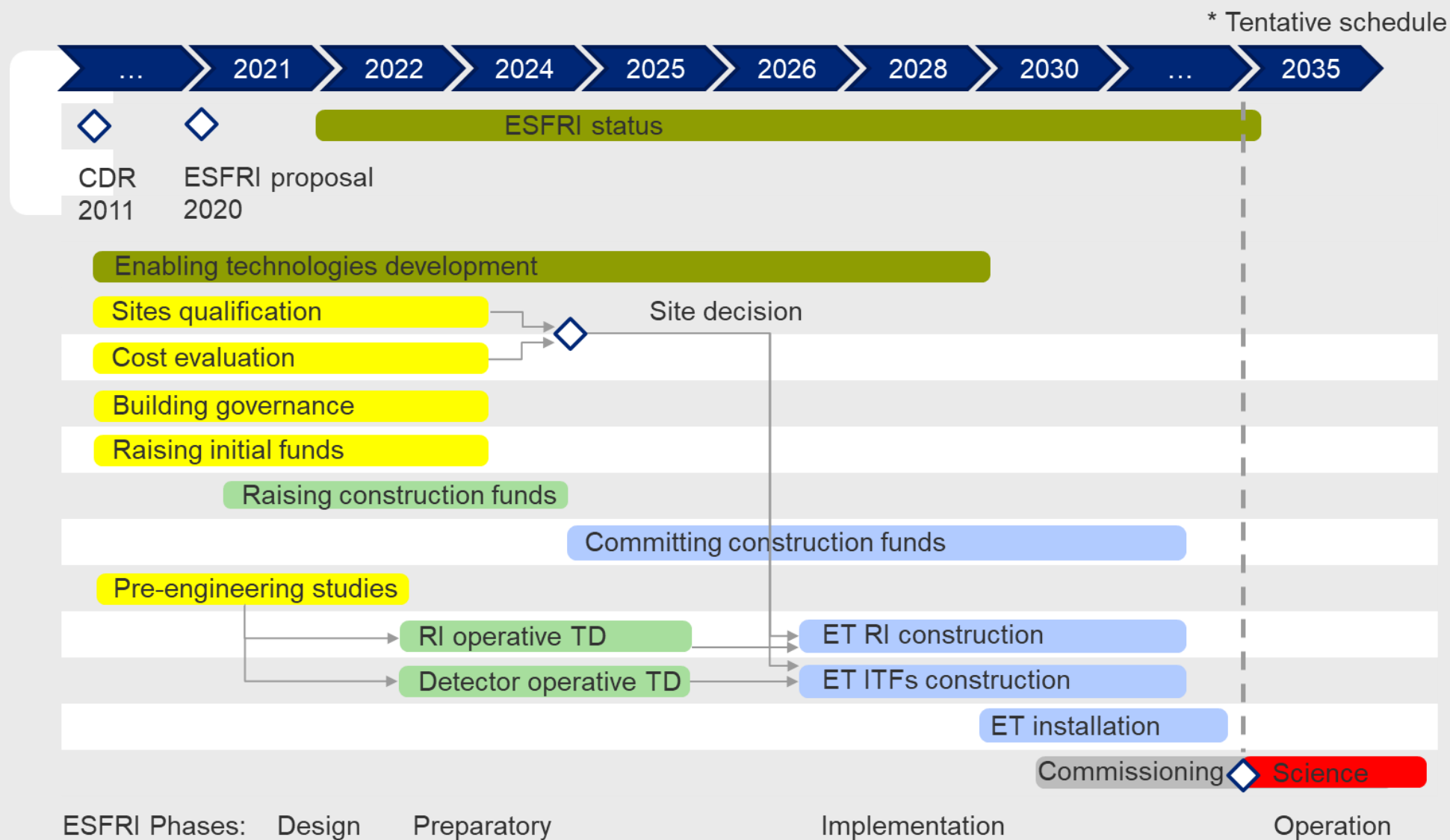
FUNDAMENTAL PHYSICS AND COSMOLOGY

- The nature of compact objects
 - near-horizon physics
 - tests of no-hair theorem
 - exotic compact objects
- Tests of General Relativity
 - post-Newtonian expansion
 - strong field regime
- Dark matter
 - primordial BHs
 - axion clouds, dark matter accreting on compact objects
- Dark energy and modifications of gravity on cosmological scales
 - dark energy equation of state
 - modified GW propagation
- Stochastic backgrounds of cosmological origin
 - inflation, phase transitions, cosmic strings



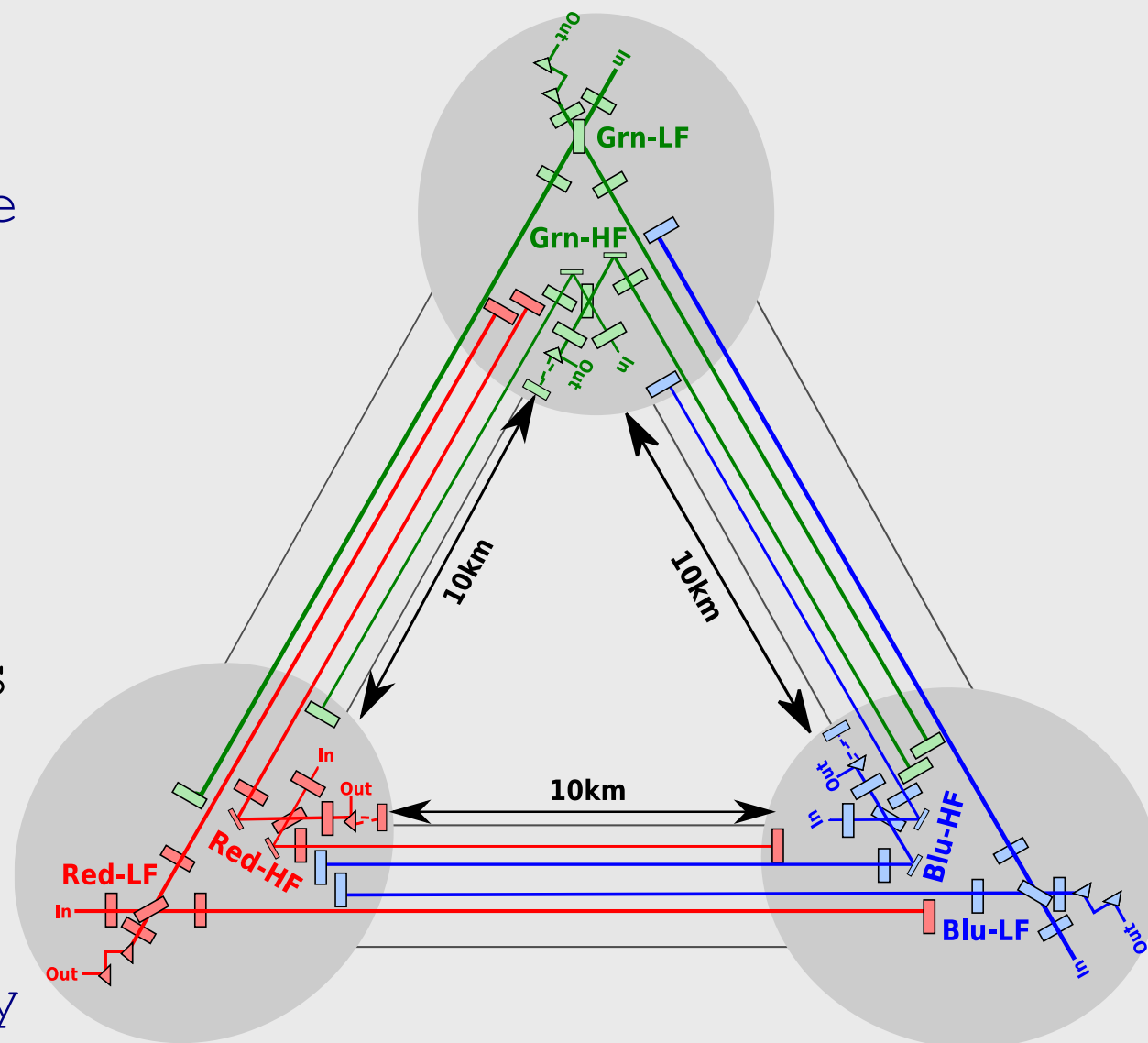
- Currently there are two candidate sites being characterized to host ET:

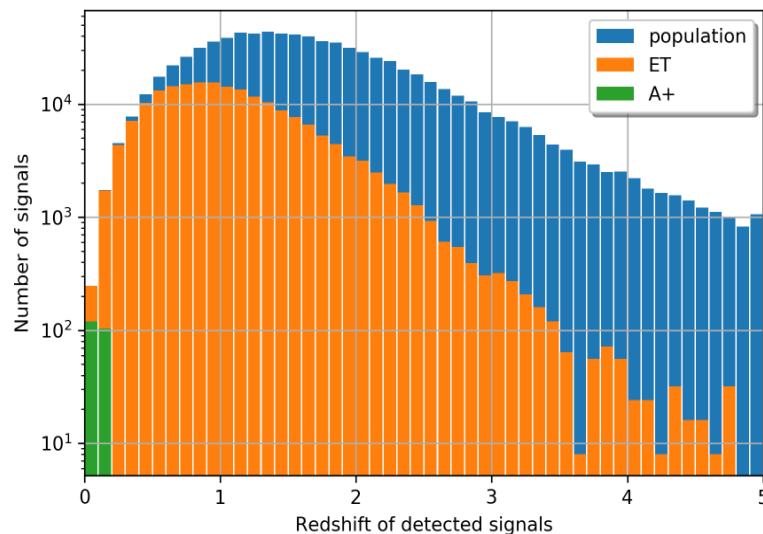
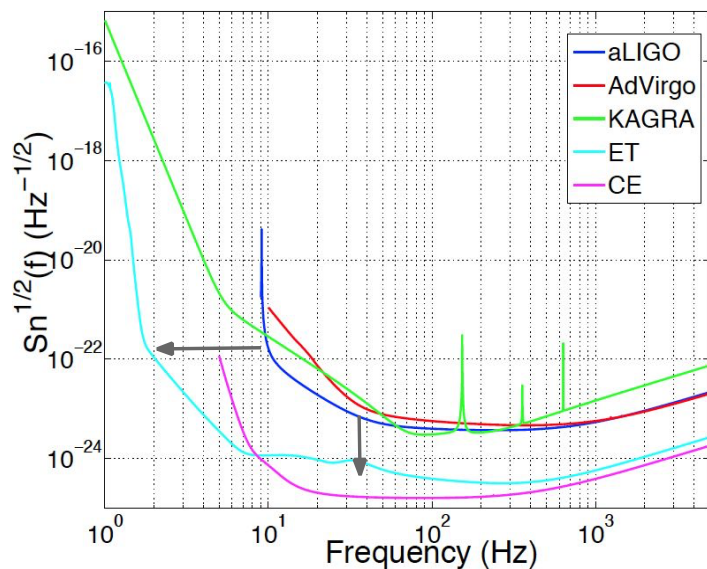
- The Sardinia site, close to the Sos Enattos mine
- The Euregio Meuse-Rhine site, close to the NL-B-D border
- A third option in Saxony (Germany) was recently proposed and is under discussion



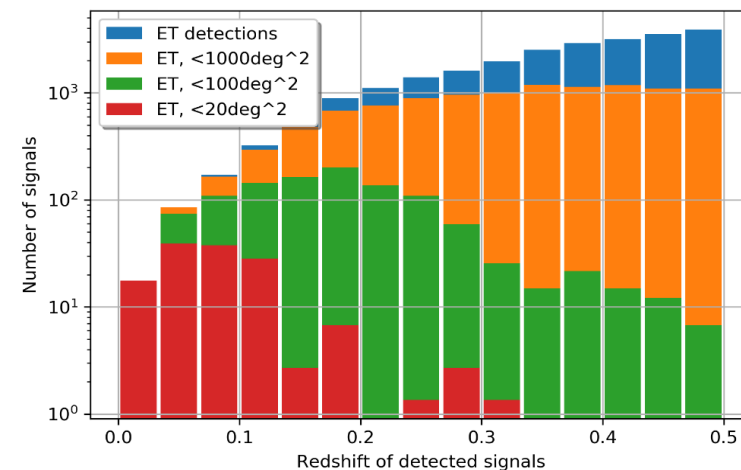
A triangular xylophone

- Three detectors in a **triangular** structure
 - Closed geometry allows the use of the null data stream
 - The third detector makes up for 60° angle
- Each detector (red, green and blue) consists of **two** Michelson interferometers
 - High-frequency and (more challenging) Low-frequency





ET sky-localization capabilities



marica.branchesi@gssi.it

- Lower frequencies (down to 1-5Hz)

- Much higher sensitivity

- 10^5 BBH detections per year

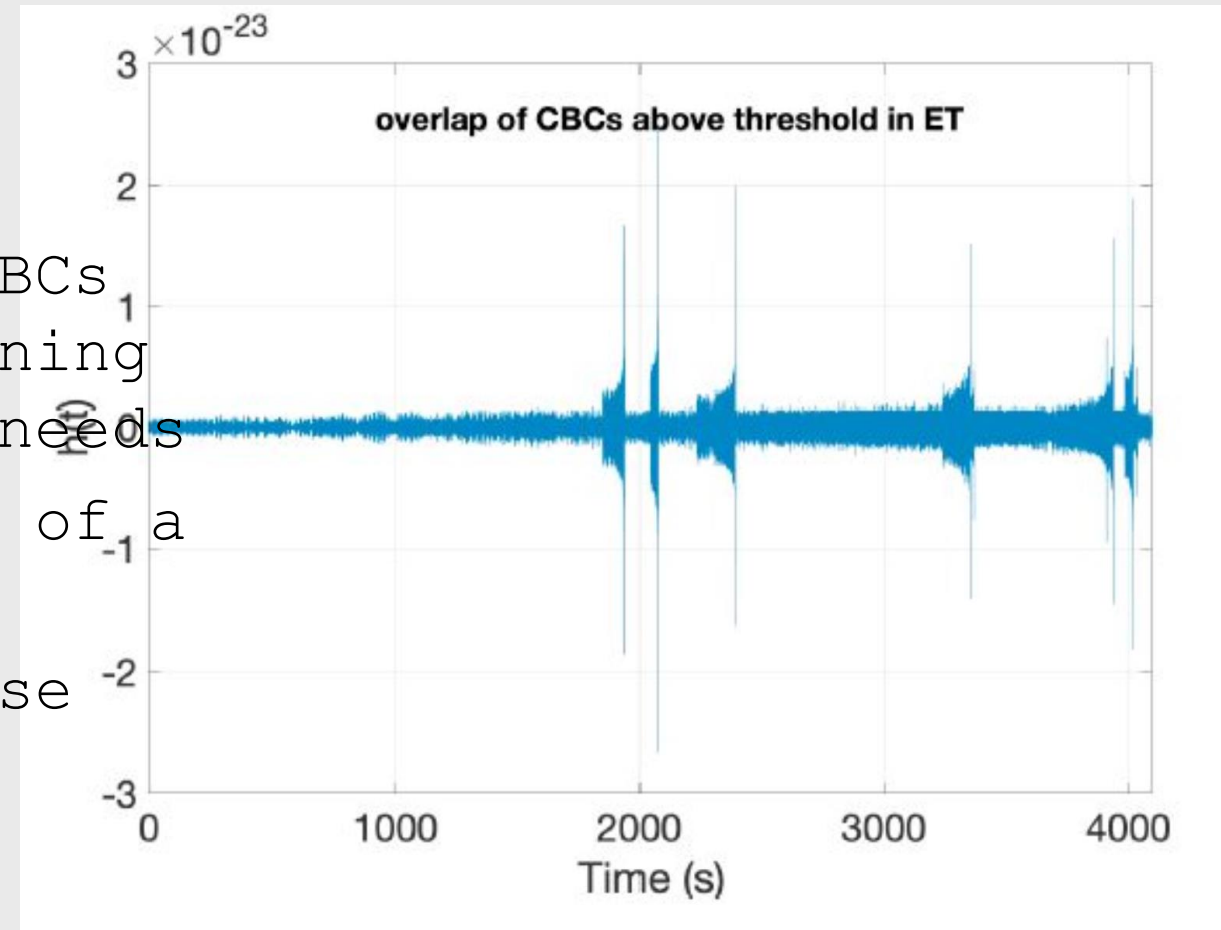
- 10^5 BNS detections per year

- ~ 100 detections per year with $<20\text{deg}^2$

- Early warning by minutes (hours)

Some challenges

- High alert rates
- Overlapping signals
- Long duration waveform for CBCs (and a moving detector), meaning large increase in computing needs
- FAR estimate in the presence of a strong foreground
- Environmental correlated noise (non-independent colocated detectors)
- ...



- Luckily, raw interferometer data don't grow much with increasing instrument sensitivity
 - We're not exploding like HL-LHC was!
 - Current detectors write $\mathcal{O}(2\text{PB})/\text{year}$ of raw data per detector
 - $h(t)$ (or "strain", the physics channel) + $\mathcal{O}(10^5)$ control channels
 - Pre-processed data for final user analysis is more than 1 order of magnitude smaller
 - In ET we expect about few tens of PB of raw data per year (baseline 6-interferometer design, more control channels,...)
 - No big deal today, piece of cake by 2035
- However, the amount of useful scientific information encoded in the data does grow a lot
 - And the computing power needed to wring it out
 - It's a difficult task in itself to precisely estimate the computing power needs

1/10th of an LHC experiment

- Current computing needs of the entire GW network are roughly $\sim 10\%$ of an LHC experiment of today
- In ET the event rate will be $10^3 - 10^4$ times the current one
 - Analysis of the “golden” events (EM counterparts, high SNR or “special” events) would already be within reach using current technologies
 - $\mathcal{O}(500)$ events per year = 12.5MHS06-y per year, the same order of magnitude of a LHC experiment in Run 4
 - Target: 1/10th of an LHC experiment in Run 4
- But: low-latency!

- Down-sampling of the data stream for long duration events
- Hierarchical methods and “decimation”
- Technology tracking of leading-edge technologies
 - Artificial Intelligence and Machine Learning
 - CUDA GPUs and HPC (FPGA and fancier architectures such as TPUs still to be tested)
 - Role of HPC is expected to grow with the SNR (Numerical Relativity, template bank production) and role of ML
 - Quantum computing!
- Early Mock Data Challenges to develop and validate everything

What will ML look like 10 years from now?

nature
astronomy

ARTICLES

<https://doi.org/10.1038/s41550-021-01405-0>



Accelerated, scalable and reproducible AI-driven gravitational wave detection

E. A. Huerta^{1,2}✉, Asad Khan³, Xiaobo Huang³, Minyang Tian³, Maksim Levental², Ryan Chard¹, Wei Wei³, Maeve Heflin³, Daniel S. Katz³, Volodymyr Kindratenko³, Dawei Mu³, Ben Blaiszik^{1,2} and Ian Foster^{1,2}

The development of reusable artificial intelligence (AI) models for wider use and rigorous validation by the community promises to unlock new opportunities in multi-messenger astrophysics. Here we develop a workflow that connects the Data and Learning Hub for Science, a repository for publishing AI models, with the Hardware-Accelerated Learning (HAL) cluster, using funcX as a universal distributed computing service. Using this workflow, an ensemble of four openly available AI models can be run on HAL to process an entire month's worth (August 2017) of advanced Laser Interferometer Gravitational-Wave Observatory data in just seven minutes, identifying all four binary black hole mergers previously identified in this dataset and reporting no misclassifications. This approach combines advances in AI, distributed computing and scientific data infrastructure to open new pathways to conduct reproducible, accelerated, data-driven discovery.

Gravitational waves were added to the growing set of detectable cosmic messengers in the fall of 2015 when the advanced Laser Interferometer Gravitational-Wave Observatory (LIGO) detectors reported the observation of gravitational waves consistent with the collision of two massive, stellar-mass black holes¹. Over the last five years, the advanced LIGO and advanced Virgo detectors have completed three observing runs, reporting over 50 gravitational wave sources^{2–5}. As advanced LIGO and advanced Virgo continue to enhance their detection capabilities and other detectors join the international array of gravitational wave detectors, it is expected that gravitational wave sources will be observed at a rate of several per day⁶.

An ever-increasing catalogue of gravitational waves will enable systematic studies to advance our understanding of stellar evolution, cosmology, alternative theories of gravity, the nature of supranuclear matter in neutron stars, and the formation and evolution of black holes and neutron stars, among other phenomena^{7–11}. Although these science goals are feasible in principle given the proven detection capabilities of astronomical observatories, it is equally true that established algorithms for the observation of multi-messenger sources, such as template-matching and nearest-neighbour algorithms, are compute-intensive and poorly scalable^{12–14}. Furthermore, available computational resources will remain oversubscribed, and planned enhancements will be outstripped rapidly with the advent of next-generation detectors within the next couple of years¹⁵. Thus, an urgent rethink is critical if we are to realize the multi-messenger astrophysics program in the big-data era¹⁶.

To contend with these challenges, a number of researchers have been exploring the application of deep learning and of computing accelerated by graphics processing units (GPUs). Co-authors of this article pioneered the use of deep learning and high-performance computing to accelerate the detection of gravitational waves^{17–19}. The first generation of these algorithms targeted a shallow signal manifold (the masses of the binary components) and required only tens

of thousands of modelled waveforms for training, but these models served the purpose of demonstrating that an alternative method for gravitational wave detection is as sensitive as template matching and significantly faster, at a fraction of the computational cost.

Research and development in deep learning is moving at an incredible pace^{20–22} (see also ref. ²³ for a review of machine-learning applications in gravitational wave astrophysics). Specific milestones in the development of artificial intelligence (AI) tools for gravitational wave astrophysics include the construction of neural networks that describe the four-dimensional (4D) signal manifold of established gravitational wave detection pipelines, that is, the masses of the binary components and the z component of the three-dimensional spin vector in $(m_1, m_2, \hat{s}_1, \hat{s}_2)$. This requires the combination of distributed training algorithms and extreme-scale computing to train these AI models with millions of modelled waveforms in a reasonable amount of time²⁴. Another milestone concerns the creation of AI models that enable gravitational wave searches over hour-long datasets, keeping the number of misclassifications at a minimum²⁵.

In this article, we introduce an AI ensemble, designed to cover the 4D signal manifold $(m_1, m_2, \hat{s}_1, \hat{s}_2)$, to search for and find binary black hole mergers over the entire month of August 2017 in advanced LIGO data²⁶. Our findings indicate that this approach clearly identifies all black hole mergers contained in that data batch with no misclassifications. To conduct this analysis we used the Hardware-Accelerated Learning (HAL) cluster deployed and operated by the Innovative Systems Laboratory at the National Center for Supercomputing Applications. This cluster consists of 16 IBM S3922 POWER9 nodes, with four NVIDIA V100 GPUs per node²⁷. The nodes are interconnected with an EDR InfiniBand network, and the storage system is made of two DataDirect Networks all-flash arrays with SpectrumScale file system, providing 250TB of usable space. Job scheduling and resource allocation are managed by the SLURM (Simple Linux Utility for Resource Management) system. As we show below, we can process data from the entire month of

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ML is not yet a mainstream “tool of the trade” in GW, but a huge lot of R&D is already ongoing

- Efficiency & speed
 - Signal Classification
 - Parameter estimation
 - Noise glitch hunting
 - (Template bank generation)
- Technology exploitation
 - Use advanced hardware (GPU, TPU...)
 - FPGAs / custom hardware
- Automatization
 - Automatize standard procedure for Data Quality
 - Automated de-noising with synthetic noise from GANs?

The mandatory slide with boxes and arrows

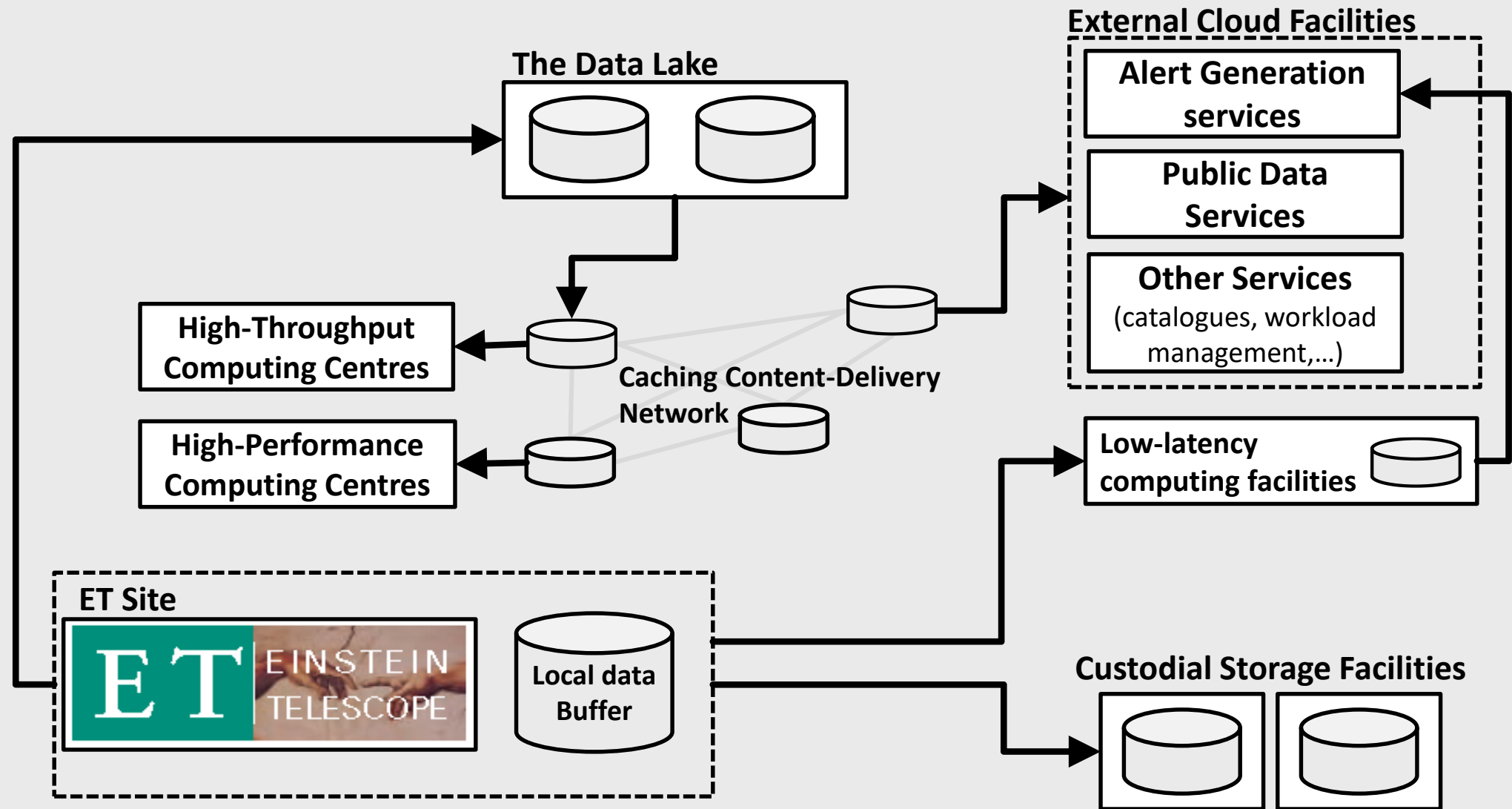


Figure from the ET ESFRI proposal

The mandatory slide with boxes and arrows

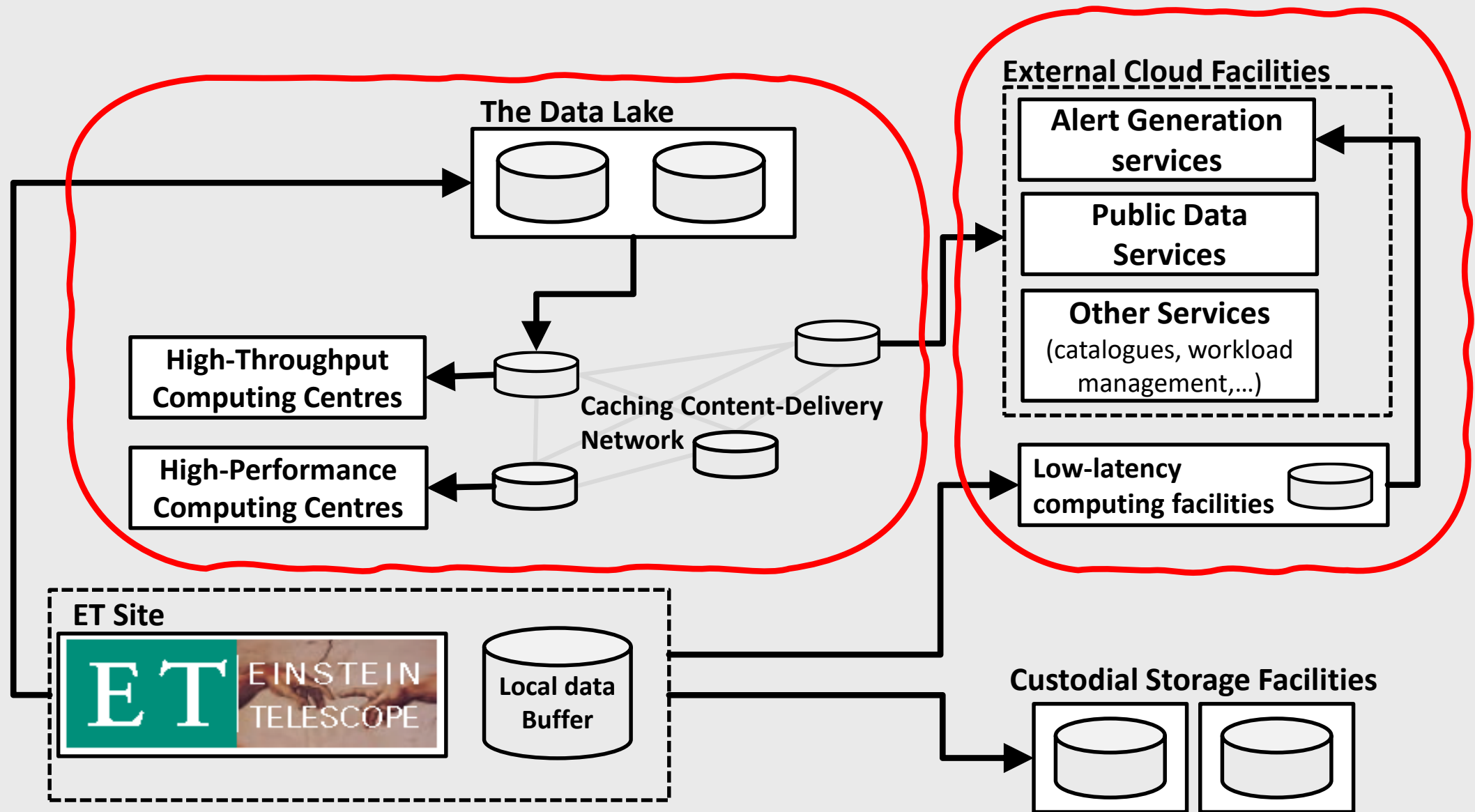
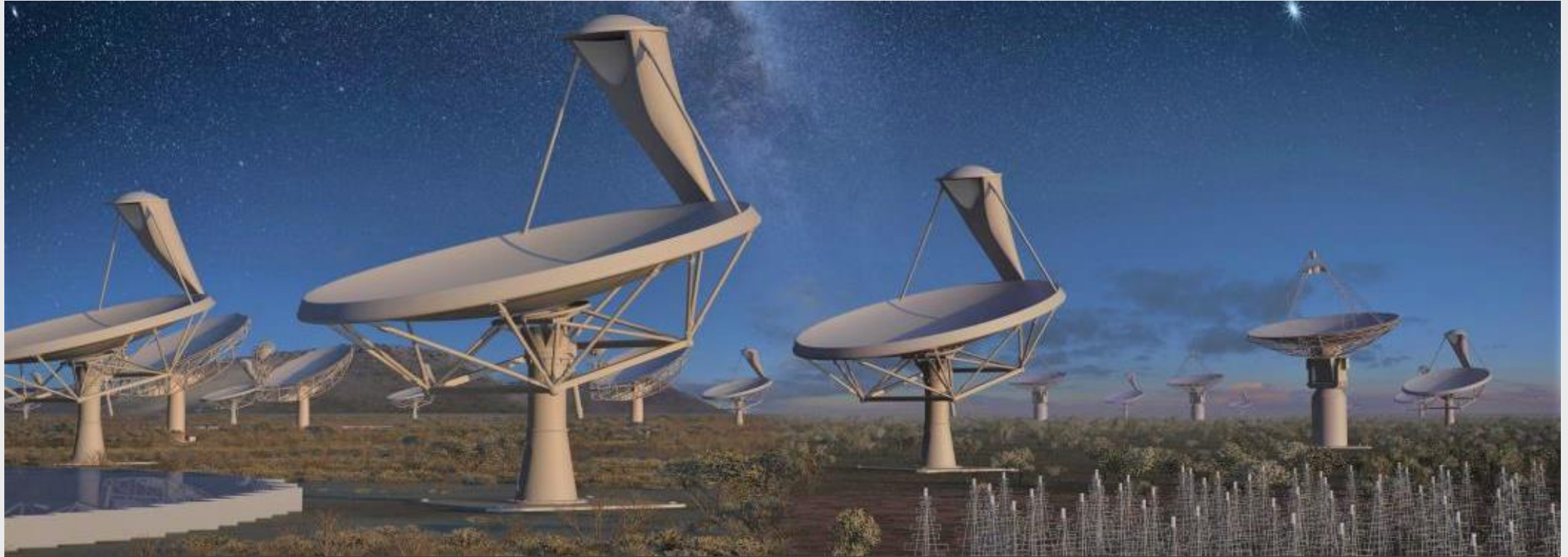
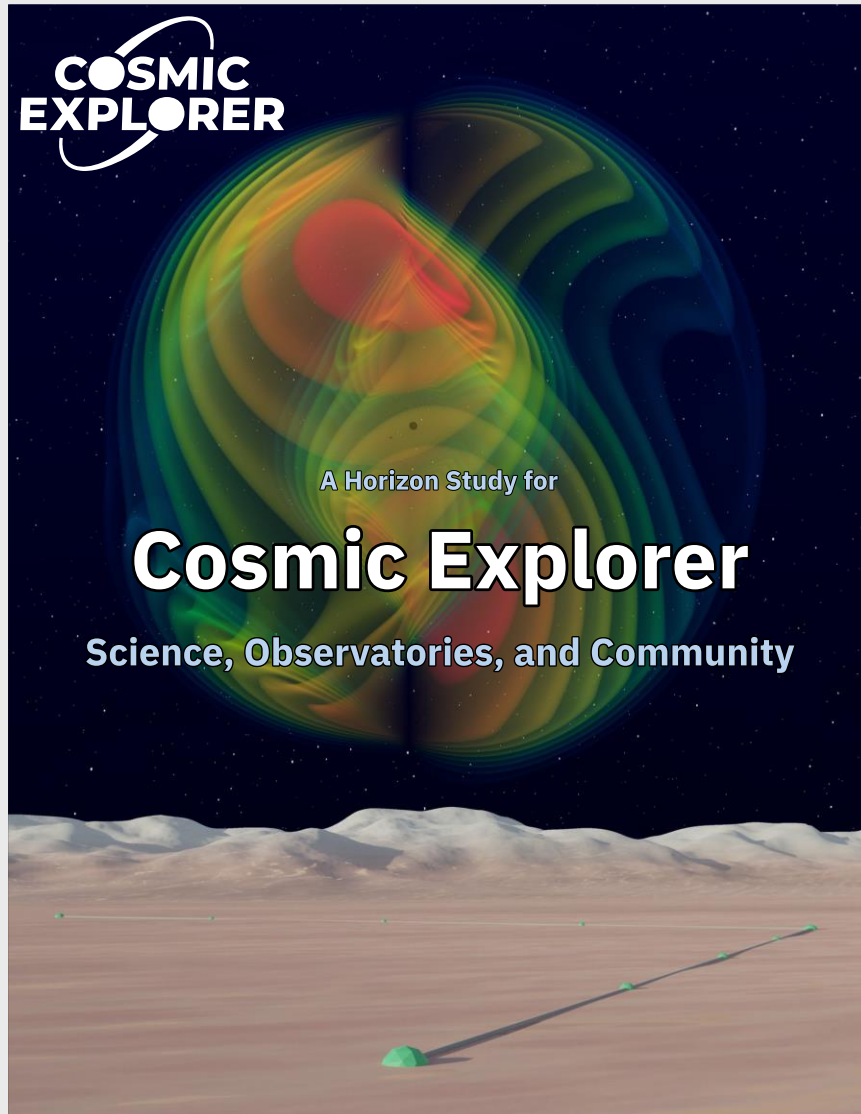


Figure from the ET ESFRI proposal

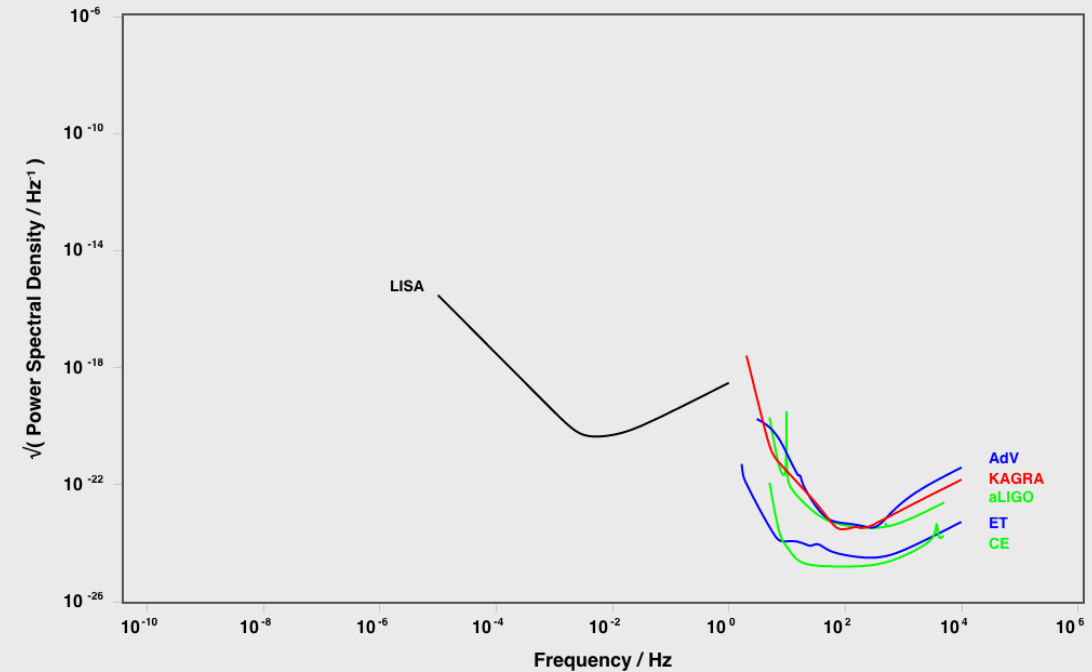
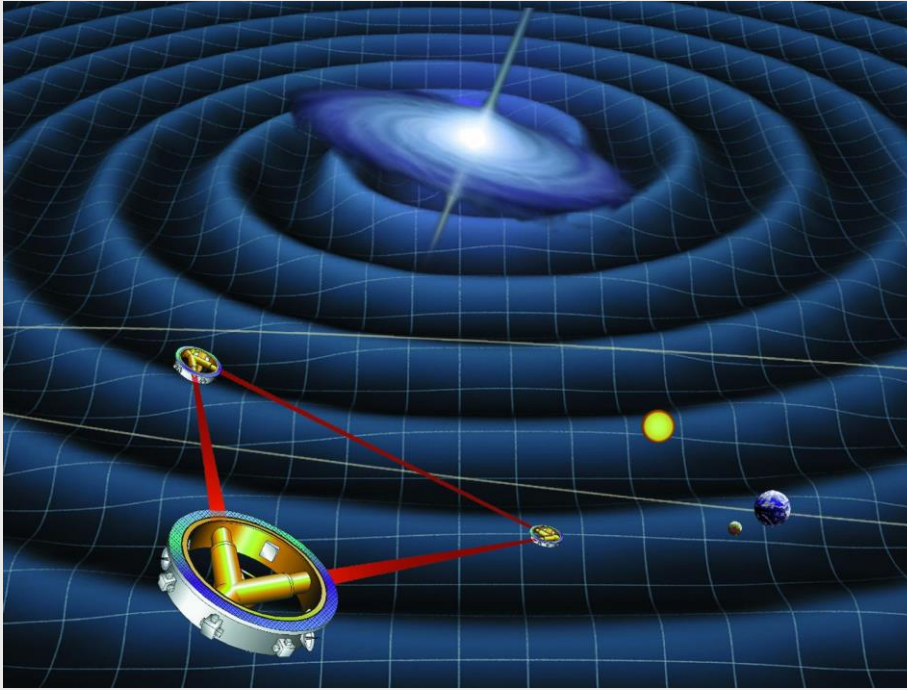


MULTIMESSENGER ASTRONOMY IN THE 2030'S



- Cosmic Explorer: proposed 3G facility in the US
- More conventional (and less risky) surface L-shaped design
- 40km + 20km arms
- Better sensitivity in mid-frequency range
- Higher (worse) low-frequency cutoff
- We'll have a 3G ground-based network!
- Will there be something like Einstein Telescope Computing and Challenges | Stefano Bagnasco, INFN
IGWN for 3G?

...and farther away: LISA



- Space-based (solar orbit) triangular laser interferometer
- 2.5 million km arms, launch planned in 2034
- Very low frequencies: complementary to ground-based interferometers
- Very massive (astrophysical) black holes, very early alerts for BNS, ultra-compact binaries, extreme mass ratio inspirals, precision tests of GR, ...

The multimessenger ecosystem

A large number of existing and future facilities will produce and consume alert triggers. For example, ESFRI-only:

Groud-based optical telescopes, like the Extremely Large Telescope: 5-mirror 39m optical telescope for the ESO on Cerro Armazones, Chile.

Large radiotelescope arrays, like the Square Kilometer Array: huge multi-band radiotelescope arrays in Africa and Australia

Facilities for cosmic ray astronomy, like the Cherenkov Telescope Array: Cherenkov telescopes for highest-energy gamma-ray astronomy, in the Canary Islands and Chile

Neutrino detectors, like KM3NeT: underwater network of neutrino detectors in the Mediterranean

The multimessenger ecosystem

- Several new, large EM and astroparticle facilities coming of age in roughly the same time frame
 - Several neutrino facilities (JUNO, DUNE, Hyperkamiokande,...)
 - Space and ground-based EM instruments (LOFAR, QTT, Vera Rubin, James Webb, Euclid, Nancy Roman, gamma and x instruments on satellites,...)
 - They will all have very similar low-latency alert requirements, as producers or consumers (or both)
 - High rates will imply extreme automation in the generation and selection of triggers, and sophisticated scheduling algorithms
 - How will the 2030's heir to today's NASA GCN work?

- Gravitational-wave discovery opened a whole new window on the universe
 - We are now progressing from the discovery to actual astronomy
- Computing for GW started small, but is coming of age
 - Upgraded detectors will raise the stakes
- Computing for the 3G era will be challenging
 - Not for data size, but for the analysis challenges
 - Much work will be needed to quantify and optimize the computing needs
 - Low-latency networks will be an important part of the challenge
- GW research will be a major player in the 2030's computing landscape
 - Even if not one of the largest
 - But multimessenger astronomy at large will be!