Fractal analysis of LIGO data a.k.a. How to characterize interferometric noise in low latency

Marco Cavaglià

INSTITUTE FOR MULTI-MESSENGER ASTROPHYSICS AND COSMOLOGY

Today is Juneteenth!





Juneteenth is the oldest nationally celebrated commemoration of the ending of slavery in the United States. Dating back to 1865, it was on June 19th that the Union soldiers, led by Major General Gordon Granger, landed at Galveston, Texas with news that the war had ended and that the enslaved were now free. It is now a federal holiday in the U.S.









Noise monitor concept



Low-latency data quality assessment

Typically a lot of works goes in Data Quality assessment

Superevent In	formatic EM advocate says event is
Superevent ID	okay. S20031 1bg Added by: Geoffrey Mo
Category	Added: March 11, 2020, Production 12:15 p.m.
Labels	DQOK EM_READY ADVOK EM_Selected EMBRIGHT_READY PASTRO_READY SKYMAP_READY GCN_PRELIM_SENT PE_READY
t _{start}	1267963150.37
t ₀	1267963151.40
t _{end}	1267963152.44
Submitted -	2020-03-11 11:59:09 UTC
Links	Data

Rapid Response Team

	Analyst Comments	Data Quality	Sky Localization	External Coincidence	EM Followup		
alyst Comn	alyst Comments						
Log Comme	ent						
RRT minutes as PDF, updated to include chat text. (S200311bg RRT Minutes.pdf)							
— Submittee	d by Jenne Driggers on March 1	1, 2020 17:04:24 UTC					
Log Comme	ent						
RRT minutes: https://docs.google.com/document/d/1TFNX3n7ZfaXmZzTuundu5rYGhmw5pFRhDtM6jcaPuYo/edit							
(S200311bg_RRT.mp3)							
— Submitted by Geoffrey Mo on March 11, 2020 12:34:14 UTC							
Log Comme	ent						
Online DQ after the e) report - H1 and L1 both event at H1 in case anyor) look clean around ne uses data after t	the time of the event. The event time.	nere is some slight noise at 1	.2s (below 100Hz)		
— Submittee	d by Laura Nuttall on March 11	, 2020 12:16:24 UTC					

Rapid Response Team

	And	lyst Comments	Data Quality	Sky Localization	External Coincidence	EM Followup		
Analys	Met	rics for th	ne Detch	ar RRT D	uring O3			
Loc	2.1	2.1 Did the candidate occur at a suspicious GPS time?						
R	2.2	2.1.2 Tech Is there a h	nical solutio iigh probabi	ns	ch was present ne	ar the		
_		candidate b 2.2.1 Exist	ased on stati ting items in picel colutio	stical inference cluded within t	of auxiliary inform his item	nation?		
Loc	2.3	Are known the candidat	sources of notices the sources of th	oise with auxilia	ary witnesses activ	e near		
(5	2.4	2.3.1 Exist 2.3.2 Tech Are known	ting items nical solutio sources of 1	ns	uxiliary witnesses	active		
-		near the car 2.4.1 Exist	ididate?			· · · · ·		
	2.5	2.4.2 Tech Are environ	nical solutio mental moni ting items	ns	the candidate?	· · · · ·		
_	2.6	2.5.2 Tech Was the det	nical solutio ector in a no	ns	· · · · · · · · · · · · · · ·			
		2.6.1 Exist 2.6.2 Tech	ting items nical solutio	ns				

Detector state

	And	Ilyst Comments	Data Quality	Sky Localization	External Coincidence	EM Followup
Analys	Met	rics for th	ne Detch	ar RRT D	uring O3	
Loç	2.1	Did the can 2.1.1 Exist 2.1.2 Tech	didate occur ting items in nical solutio	t at a suspicious acluded within t	GPS time? his item	
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_	2.4	Are known near the car	sources of 1 ndidate?	ns	uxiliary witnesse	s active
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O	2.0	2.5.1 Exis 2.5.2 Tech	ting items nical solutio	nors active hear		
	2.6	Was the det 2.6.1 Exis 2.6.2 Tech	ector in a no ting items nical solutio	ominal state? .)	

Detector state

Analyst Comments Data Quality Sky Localization External Coincidence EM Followup							
Analys	Analys Metrics for the Detchar RRT During O3						
Loc	2.1	Did the candidate occur at a suspicio 2.1.1 Existing items included withir 2.1.2 Technical solutions	 Identify nearest hardware injection 				
R	2.2	Is there a high probability that a g candidate based on statistical inferen	• Bruco https://git.ligo.org/gabriele-vajente/bruco				
Loc	2.3	2.2.1 Existing items included withir2.2.2 Technical solutionsAre known sources of noise with aux	• Stochmon https://dcc.ligo.org/LIGO-T1400205/public				
R (S		the candidate?2.3.1Existing items2.3.2Technical solutions	 Stationarity 				
_	2.4	Are known sources of noise without near the candidate?	 Check calibration kappas 				
	2.5	2.4.2 Technical solutions Are environmental monitors active ne 2.5.1 Existing items	 Calibration state vector check 				
	2.6	 2.5.2 Technical solutions					

Building a nominal state metric from strain data



Noise and fractals



"In 1961 IBM was involved in transmitting computer data over phone lines, but a kind of white noise kept disturbing the flow of information—breaking the signal—and IBM looked to Mandelbrot to provide a new perspective on the problem."

https://www.ibm.com/ibm/history/ibm100/us/en/icons/fractal/

Fractal analysis



Figure credit: https://wallup.net/

- Fractal analysis can be used to characterize the degree of complexity of a set.
- The concept has been applied to different physical phenomena met in various fields from materials science to chemistry, biology, etc.

Topological dimension of a set

A topological space is an ordered pair (X, τ), where X is a set and τ is a collection of subsets of X, satisfying the following axioms:

- The empty set and X itself belong to τ .
- Any arbitrary (finite or infinite) union of members of τ belongs to τ .
- The intersection of any finite number of members of τ belongs to τ .

The elements of τ are called open sets and the collection τ is called a topology on X.

A topological space can be covered by open sets. The topological (Lebesgue covering) dimension is the smallest number n such that for every cover, there is a refinement in which every point in X lies in the intersection of no more than n + 1 covering sets. The topological dimension is an integer number that does not change as the space is continuously deformed under an homeomorphism.

M. Yamaguti et al. "Mathematics of Fractal" Mathematical monographs, vol.167 AMS 1997s

Topological dimension of a set

Examples:

A circle requires a set of two (or more) open arcs to be covered. The topological dimension of the circle is one.

A disk requires at least three open sets to be covered. The topological dimension of the disk is two.

In general an *n*-dimensional Euclidean space has topological dimension n.







Fractal dimension

A fractal is a subset of an Euclidean space with a fractal dimension that strictly exceeds its topological dimension.

The fractal dimension D is defined with the number of covering areas necessary to cover the fractal structure F.

 ϵ = radius of the covering balls

 $m_d = s$ -dimensional Hausdorff-outer measure

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D is a measure of roughness.

For an *n*-dimensional Euclidean space, the Hausdorff dimension coincides with the topological dimension: Single point = 0, line segment = 1, square = 2, etc.

M. Yamaguti et al. "Mathematics of Fractal" Mathematical monographs, vol.167 AMS 1997s

Examples

Takagi-Landsberg curve \sim

$$T(x) = \sum_{n=0}^{\infty} w^n s(2^n x) \quad s(x) = \min_{n \in \mathbb{Z}} |x - n|$$
$$D = 2 + \log_2(w)$$



Weierstrass function

$$f(x) = \sum_{k=1}^{\infty} a^{-k} \sin(b^k x), \ 1 < a < 2, \ b > 1$$
$$D = 2 - \log_b(a)$$



Examples

Brownian motion (Wiener process)

$$D = 3/2$$

White noise

$$D=2$$





Calculation of fractal dimension

ANAM method

$$M_{\tau}^{\alpha}(f,x) = \left[\frac{1}{\tau^2} \int_{t_1=0}^{\tau} \int_{t_2=0}^{\tau} |f(x+t_1) - f(x-t_2)|^{\alpha} dt_1 dt_2\right]^{1/\alpha}$$

$$K_{\tau}^{\alpha}(f,a,b) = \frac{1}{b-a} \int_{x=a}^{x=b} \left[\frac{1}{\tau^2} \int_{t_1=0}^{\tau} \int_{t_2=0}^{\tau} |f(x+t_1) - f(x-t_2)|^{\alpha} dt_1 dt_2 \right]^{1/\alpha} dx$$

$$D = \lim_{\tau \to 0} \left(2 - \frac{\log K^{\alpha}_{\tau}(f, a, b)}{\log \tau} \right)$$

Bigerelle M, Iost A., C.R. Acad. Sci. Paris, t. 323, Serie II b, 1996:669±74

Calculation of fractal dimension

Variation (VAR) method

 $f: [a,b] \to \mathbb{R},$ $OSC_{\tau}(f,x) = \left| \max_{\substack{|x-t| < \tau}} (f(t)) - \min_{\substack{|x-t| < \tau}} (f(t)) \right|$

$$\operatorname{VAR}_{\tau}(f, a, b) = \frac{1}{b-a} \int_{a}^{b} \operatorname{OSC}_{\tau}(f, x) \, \mathrm{d}x$$

$$D = \lim_{\tau \to 0} \left(2 - \frac{\log \mathbf{VAR}_{\tau}(f, a, b)}{\log \tau} \right)$$

Algorithm

- Discretize either the $VAR_{\tau}(f, a, b)$ or the $K^{\alpha}_{\tau}(f, a, b)$ estimator
- The fractal dimension of the time series is the slope of 2 – log(estimator) / log(τ)
- Theoretically this a straight line. In practice, do a fit of the log(estimator) as function of log(τ).



Simple tests

Data length: 1 s, sampling rate: 4096 Hz. Decimate factor: 1/128 (VAR/ANAM).

Theory	VAR	ANAM
1.000	1.020 (+2%)	1.002 (+0.2%)
1.485	1.448 (-2.5%)	1.503 (+1.2%)
1.678	1.748 (+4.2%)	1.668 (-0.6%)
1.838	1.747 (-5.0%)	1.804 (-1.9%)
1.721	1.638 (-4.8%)	1.662 (-3.4%)
	Theory 1.000 1.485 1.678 1.838 1.721	TheoryVAR1.0001.020 (+2%)1.4851.448 (-2.5%)1.6781.748 (+4.2%)1.8381.747 (-5.0%)1.7211.638 (-4.8%)

VAR	ANAM
Faster	Slower (x100 VAR)
Less accurate	More accurate

Tests on fake noise: White noise

Data length: 1 s, sampling rate: 4096 Hz. Decimate factor: 16/128 (VAR/ANAM) Runs: 100. Theoretical fractal dimension: 2.000



Tests on fake noise: Brownian noise

Data length: 1 s, sampling rate: 4096 Hz. Decimate factor: 16/128 (VAR/ANAM) Brownian motion speed: 2. Runs: 100. Theoretical fractal dimension: 1.500













LIGO Hanford O2 data. Sampling rate: 4096 Hz. Fractal dimension over 1 sec. Method: VAR. Decimate factor: 16



Category 1 data (interferometer in low noise, but locked)

LIGO Hanford O2 data. Sampling rate: 4096 Hz. Fractal dimension over 1 sec. Method: VAR. Decimate factor: 16



Category 1 data (interferometer in low noise, but locked)













Background noise is not the same across the run!





Deviations > 2.5 σ



Deviations > 2.5 σ



Deviations > 2.5 σ



Deviations > 2.5 σ



Deviations > 2.5 σ







Rolling mean + anomaly detection with Local Outlier Factor (LOF)





Rolling mean + anomaly detection with Local Outlier Factor (LOF)



Rolling mean + anomaly detection with Local Outlier Factor (LOF)



Long-term noise variation



Long-term noise variation



Long-term noise variation



Conclusions

- Simple algorithms that can run on GPUs with numba \rightarrow speed x 100
- One second of O2 open data at 4kHz decimated by a factor 16 is processed with the VAR algorithm on Caltech's GPU's pcdev11 in less than 0.6 s (including I/O)
- ~ 1/3 of O2 open data processed in a few days: https://ldas-jobs.ligo.caltech.edu/~marco.cavaglia/Fractals/
- Fractal dimension can characterize the noise and can be processed in real time!

Useful references:

- M. Bigerelle, I. Alain, *Fractal dimension and classification of music*, Chaos Solitons & Fractals 11(14):2179-2192 (November 2000), DOI:10.1016/S0960-0779(99)00137-X
- P. Maragos, A. Potamianos, Fractal dimensions of speech sounds: Computation and application to automatic speech recognition, The Journal of the Acoustical Society of America 105(3):1925-32 (April 1999) DOI:10.1121/1.426738.
- M. Yamaguti et al. "Mathematics of Fractal" Mathematical monographs, vol.167 AMS 1997.

Thank you! Questions (†) ?

(†) "If you ask me a question I do not know, I'm not going to answer it" – Yogi Berra



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