

Advanced LIGO Calibration Uncertainty for Precision Astrophysics

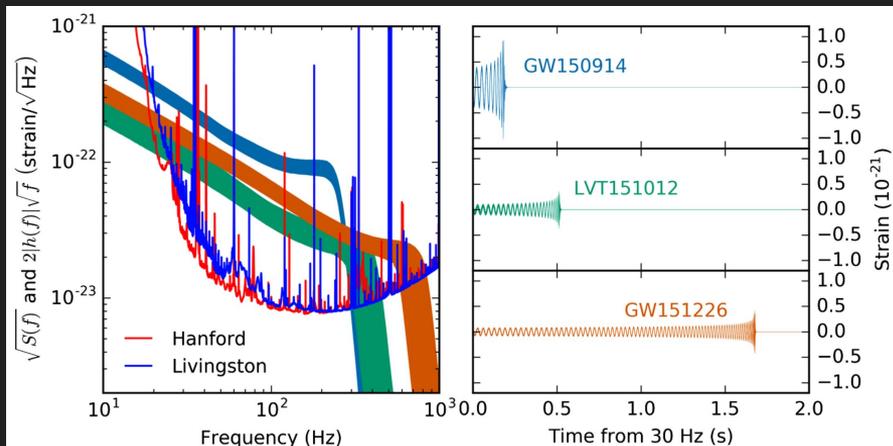
Craig Cahillane
July 10th, 2017

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O1 + O2 Calibration Uncertainty Budgets

- What is calibration?
 - Production of GW strain data from our detector data
- Why calibration uncertainty?
 - It's the project I was handed when I was a first year
 - No one cared until we made a detection
 - Now everyone cares
 - Imperative for precision astrophysics



Calibration
Pipeline

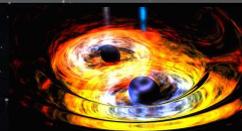


Models

$C(f)$
 $A(f)$



Motivation for Low Calibration Uncertainty

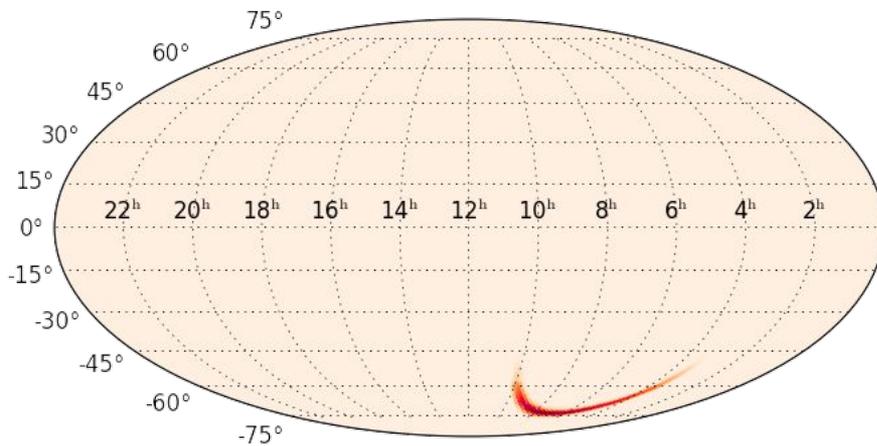
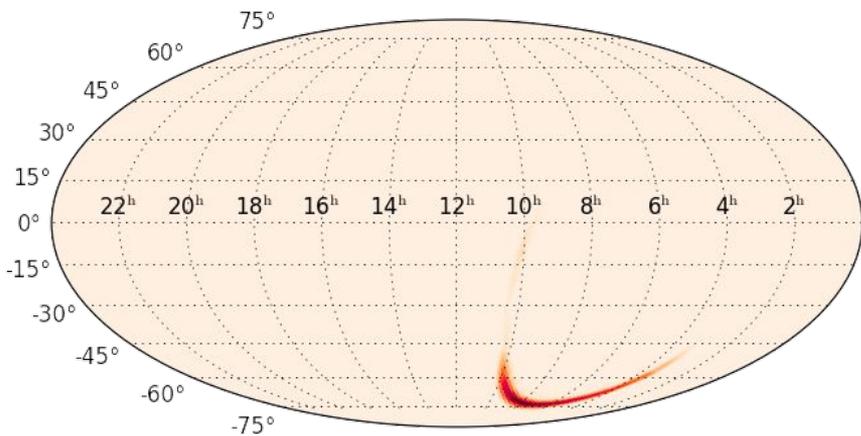


- We don't want to just make detections, we want to do **astrophysics** with these detections
 - Source parameters
 - Black hole masses, spins, luminosity distance, inclination, sky location, etc
 - Merger rates
 - Event rate, universal mass distribution, binary star formation
 - Tests of general relativity
 - Strong-field non-linear regime
 - Cosmology
 - Hubble constant measurements
- The accuracy and precision to which we know GW strain data affects all of the above
- Right now we aren't calibration uncertainty limited, we are SNR limited
 - This won't always be the case, when we start getting SNR ~ 700 detections in Super Advanced LIGO

Impact of Cal Uncertainty on GW150914 Sky Location

GW150914 90% sky area with 10%, 10 degrees cal uncertainty = 231 square degrees

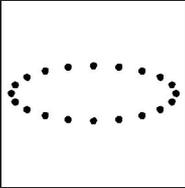
GW150914 90% sky area with NO cal uncertainty = 153 square degrees



Plots from Chris Berry

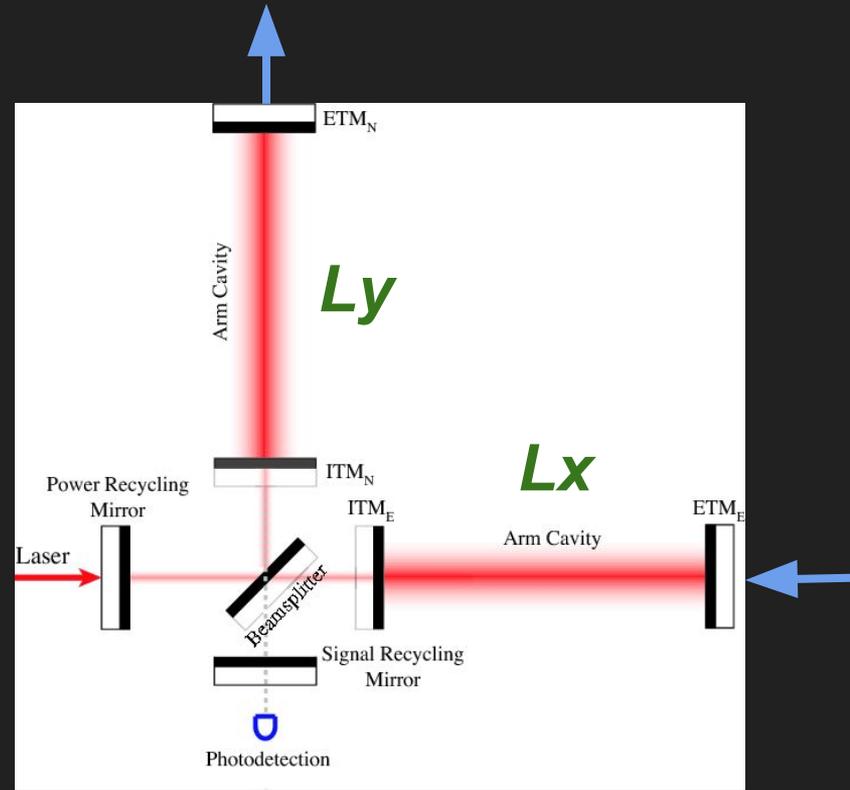
Gravitational Waves and Interferometers

- When a GW hits test particles, it stretches and squeezes them in a quadrupolar way



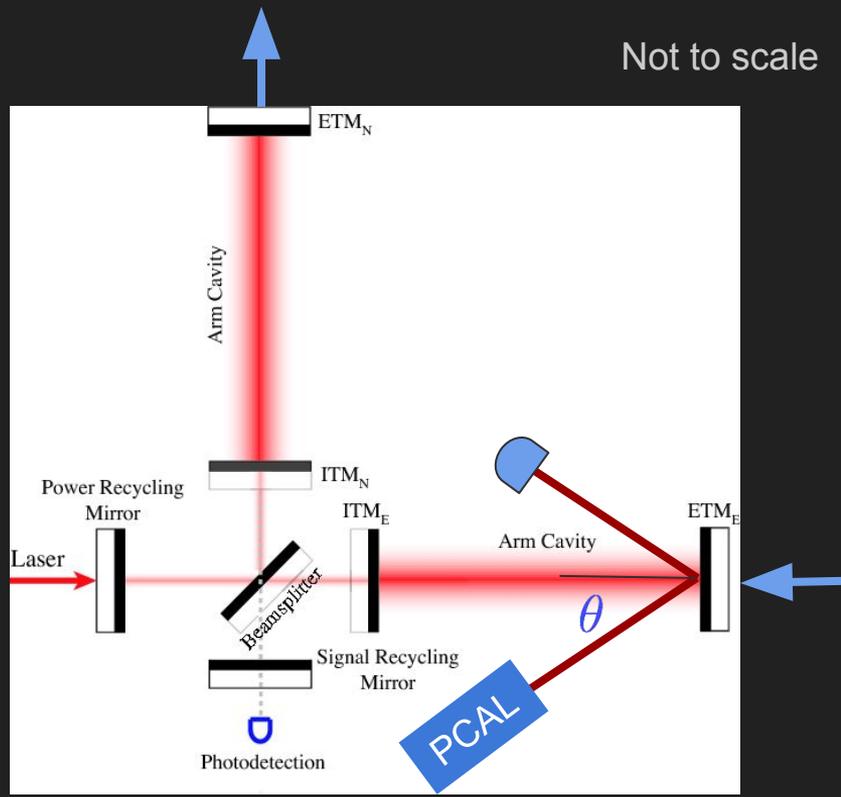
Ring of test particles hit by plus polarized GW

- An interferometer's End Test Masses (ETMs) are like the above test particles
 - A GW changes the distances L_x and L_y
- When on resonance, or "locked", an interferometer is hyper-sensitive to differential arm motion
 - DARM = Differential Arm Motion
 - $L_{\text{DARM}} = L_x - L_y$
- We control this motion with the DARM control loop



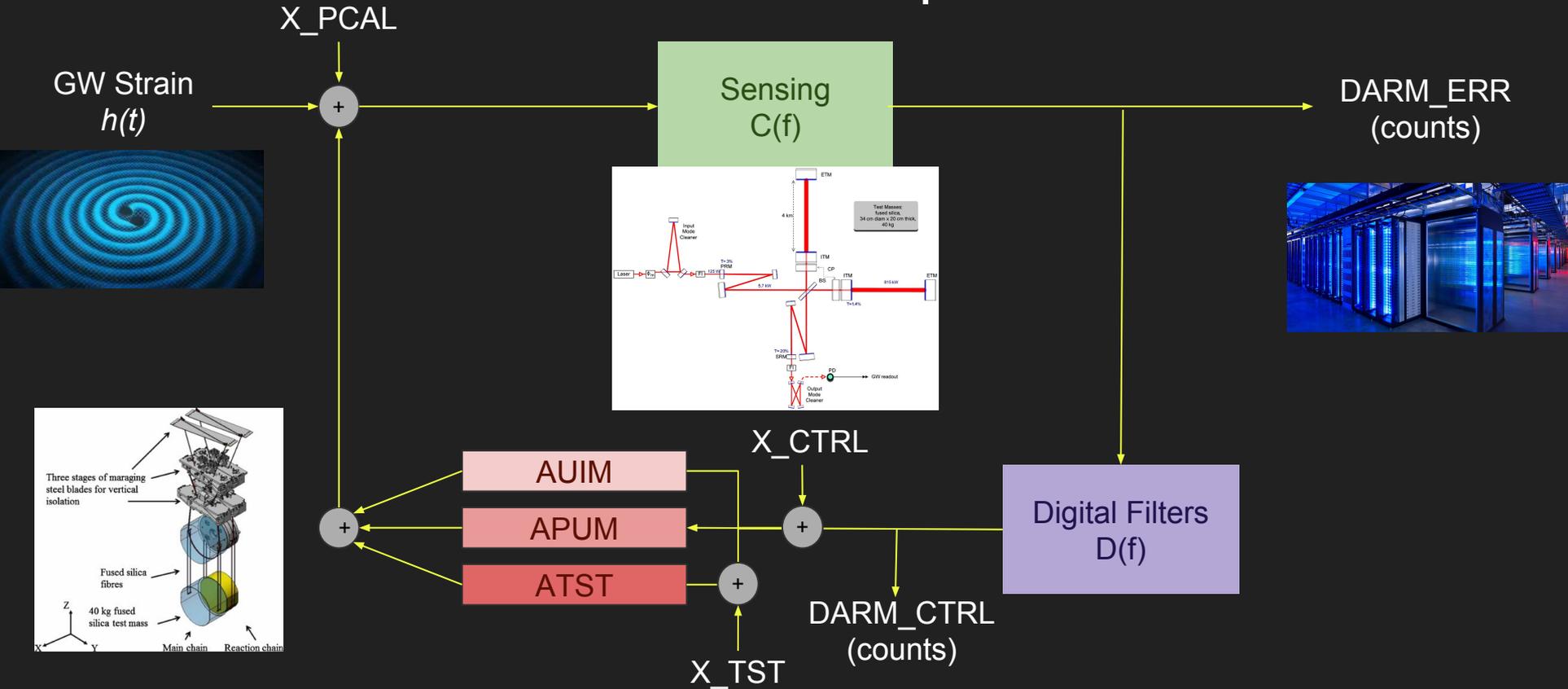
What is Calibration?

- Push on end mirrors by known amount with the **photon calibrator laser (PCAL)** [8]
 - This laser's power is extremely well known (~2 Watts)
 - Imposes a fundamental limit on our test mass motion uncertainty of **~0.8%**
- When we push on one end test mass, it simulates a gravitational wave incident on our detector
 - Light in the cavity is phase shifted into the antisymmetric port onto our photodetector
- This photodetector readout gives us our calibration from meters of test mass motion to arbitrary counts



$$\theta = 8.75^\circ$$
$$\lambda = 1047 \text{ nm}^7$$

DARM Loop



$$G(f) = C(f) D(f) A(f)$$

$$R^{-1}(f) = \frac{1 + G(f)}{C(f)}$$

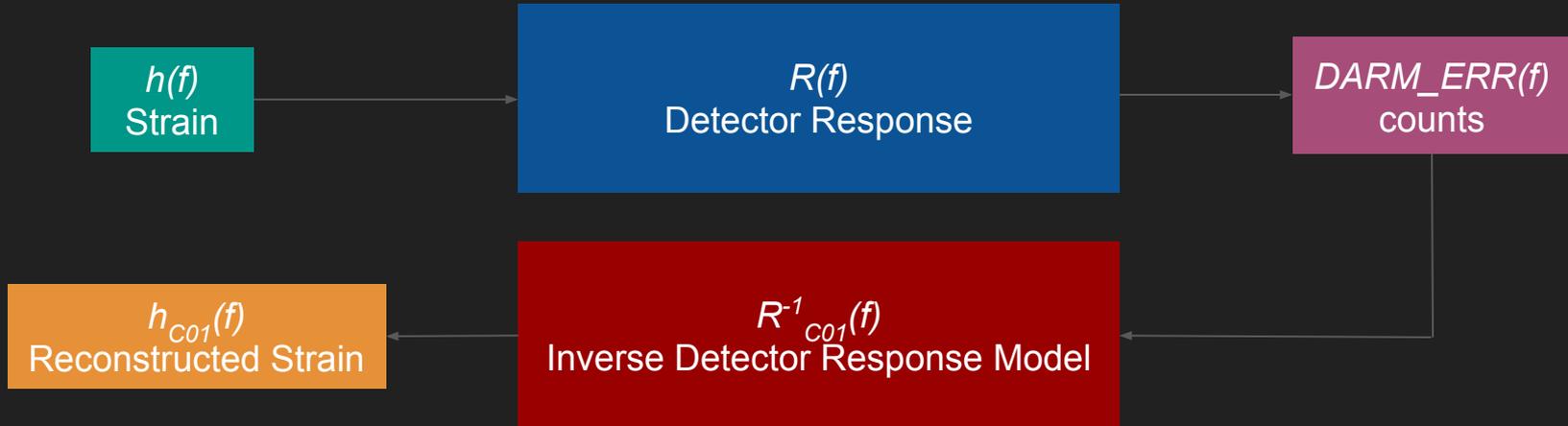
DARM Response

- The inverse detector response function $R^{-1}(f)$ is the transfer function from GW strain to DARM_ERR counts:

$$h(f) = R^{-1}(f) d_{err}(f)$$

- This means that uncertainty in strain is equivalent to uncertainty in the response:

$$\sigma_h(f) = \sigma_{R^{-1}}(f)$$



Sensing Function $C(f)$ Model

- Through the work of Buonanno and Chen, Robert Ward, Evan Hall, and Kiwamu and myself, the calibration group has a physical model for our interferometer
 - Buonanno and Chen modeled a signal-recycled interferometer using quantum optics [2].
 - Robert Ward converted the above into dual-recycled Fabry-Perot interferometer model [3].
 - Evan Hall showed the above model described detuning of the interferometer [4].
 - Kiwamu and I simplified the model down to the simple pole and optical spring we have today.

Calibration Sensing Model $C(f)$

$$\frac{f^2}{f^2 + f_S^2 - i f f_S Q^{-1}} \frac{\kappa_C(t) H_C}{1 + i f / f_{CC}(t)}$$

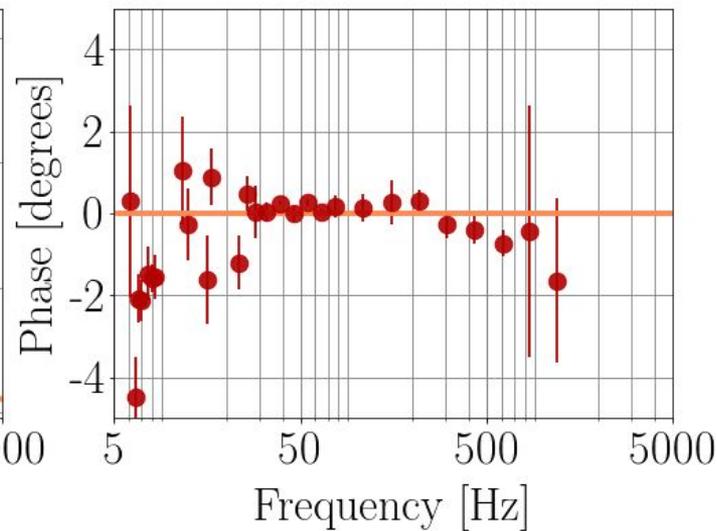
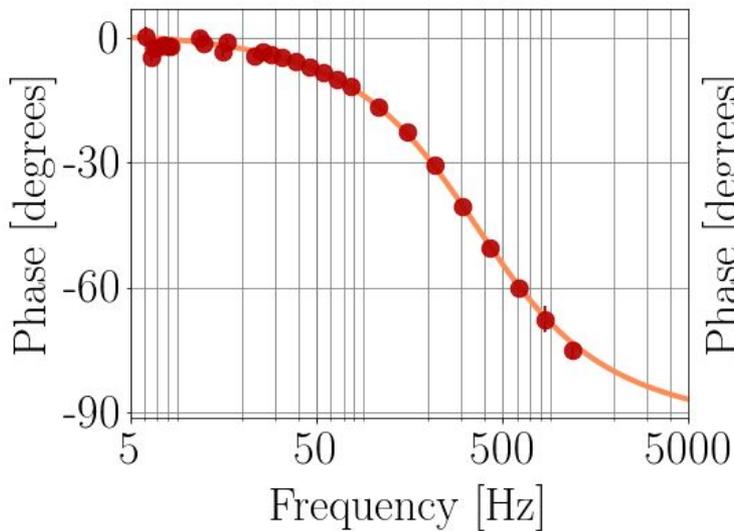
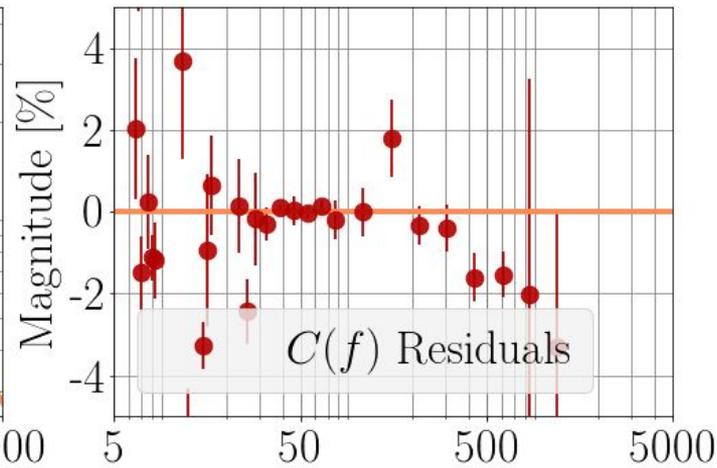
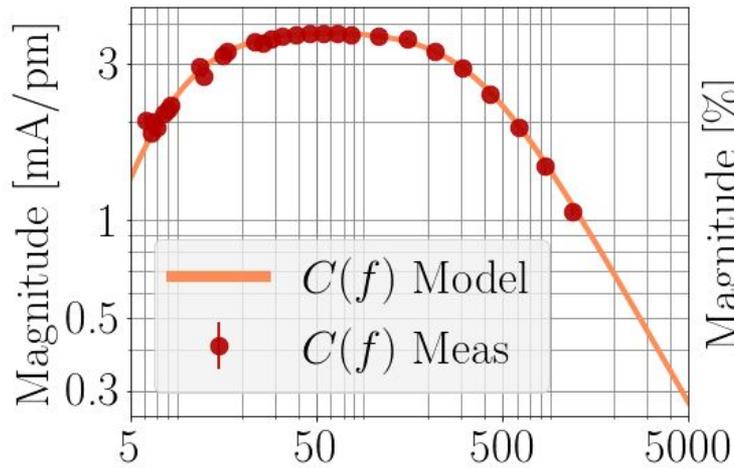
H_c = Optical Gain

$\kappa_C(t)$ = Gain Time-Dependent Scalar

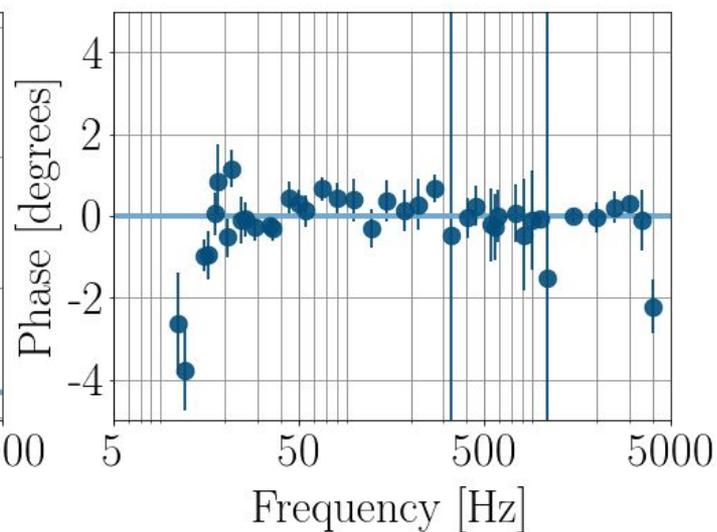
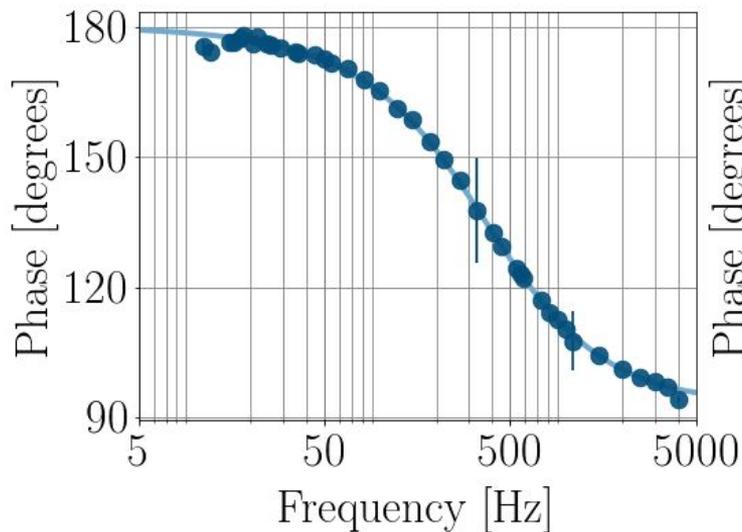
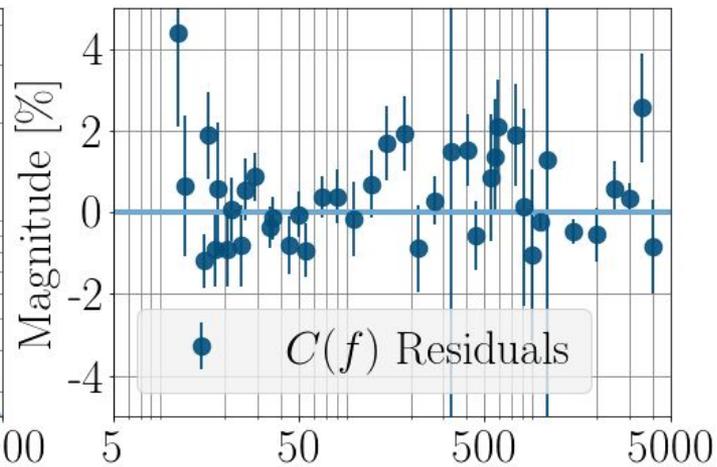
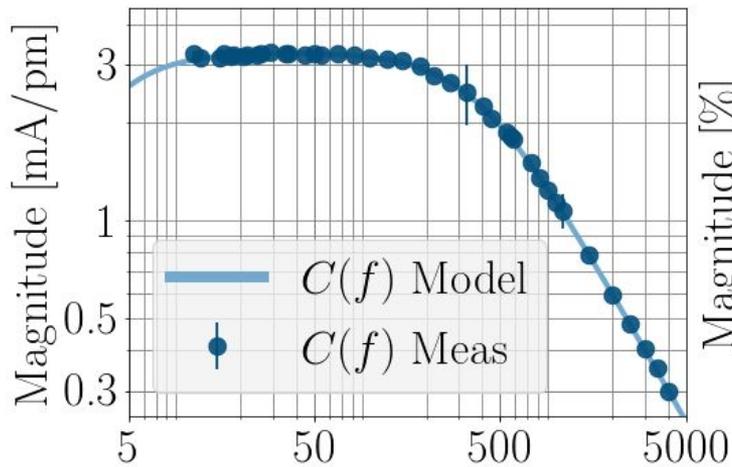
$f_{cc}(t)$ = Coupled Cavity Pole

f_S = Optical Spring Frequency

Q = Optical Spring Q



Meas Date:
Jan 4, 2017



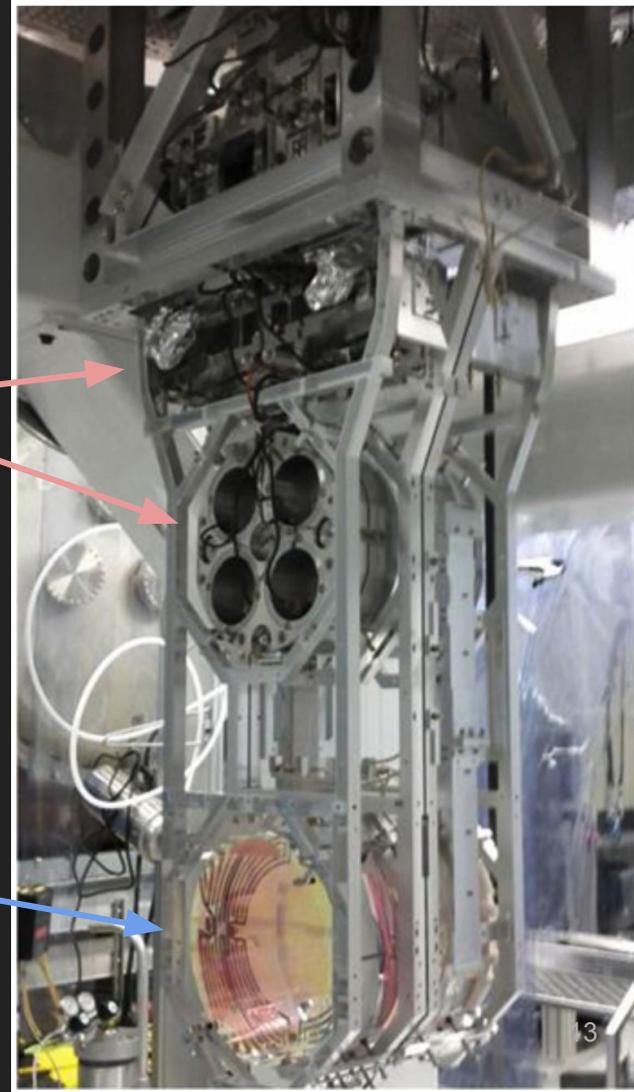
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Nov 26, 2016

Actuation Function $A(f)$ Model

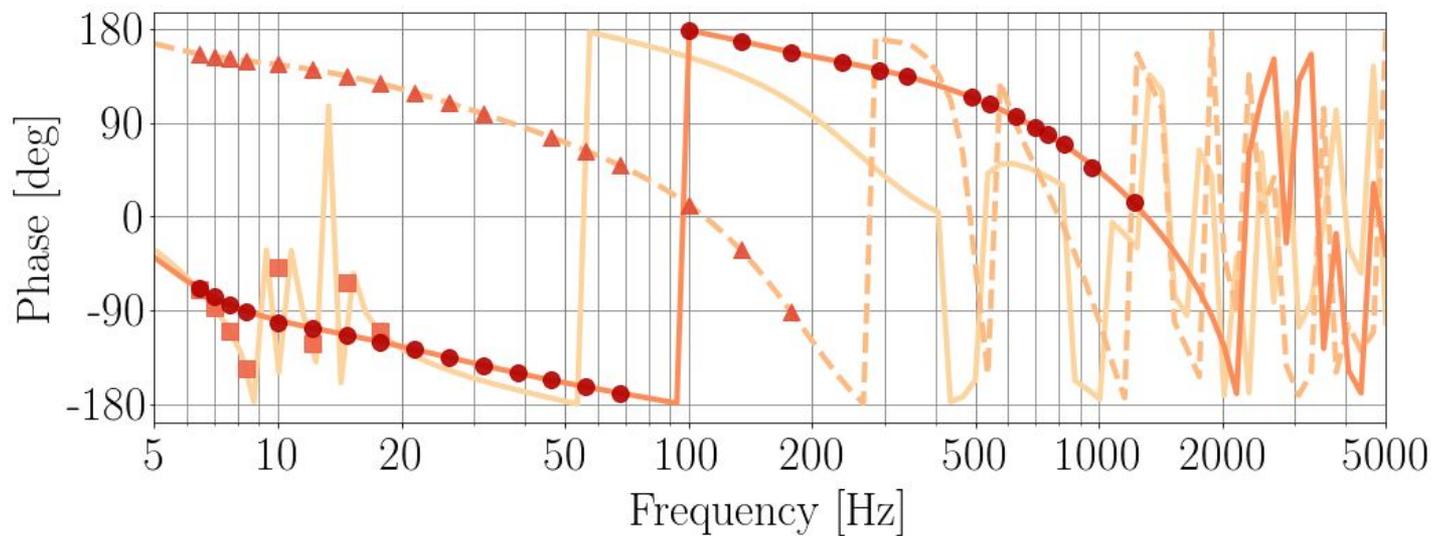
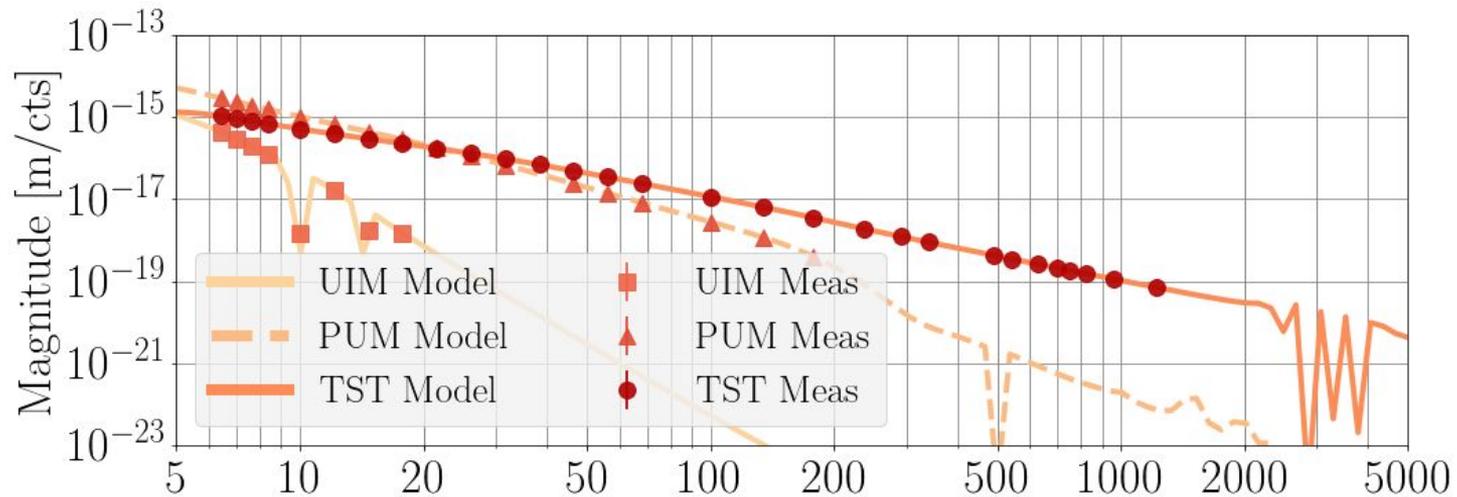
- We also have a complete model of our suspensions.
 - This is important because we actuate on our suspensions to keep the interferometer locked
 - The photon calibrator (PCAL), actuates on end optics using radiation pressure
 - This laser is our fundamental limit on calibration uncertainty
- With the model of the suspensions and the model of the interferometer, we have a complete physical model of our detector DARM control loop.

Optical Sensor,
Electromagnetic
Coil Actuators

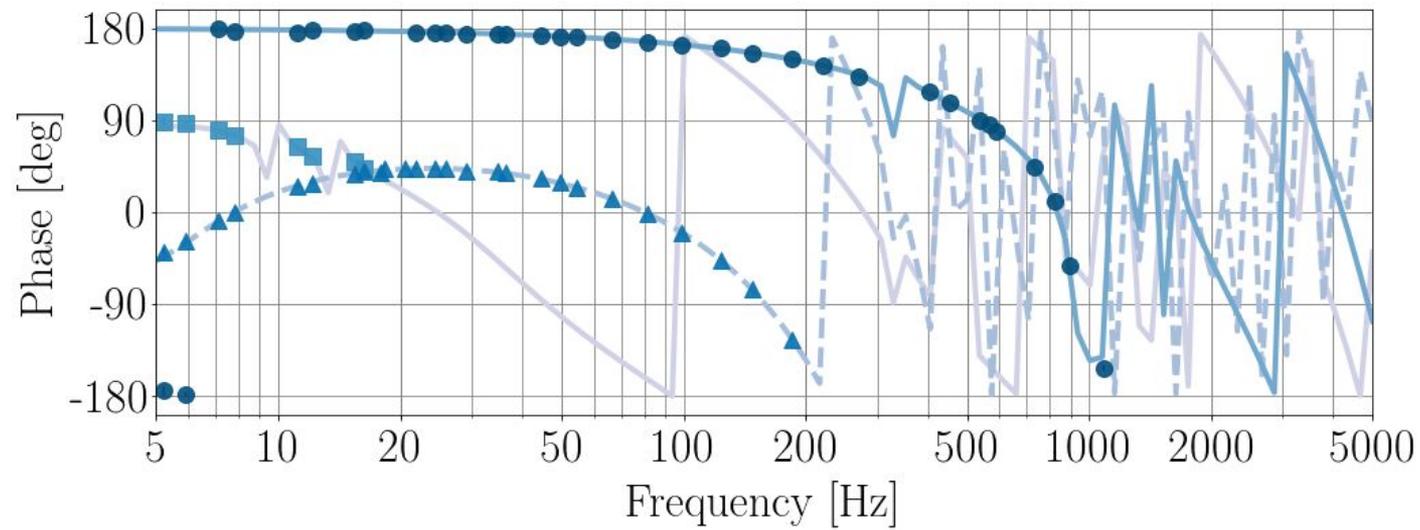
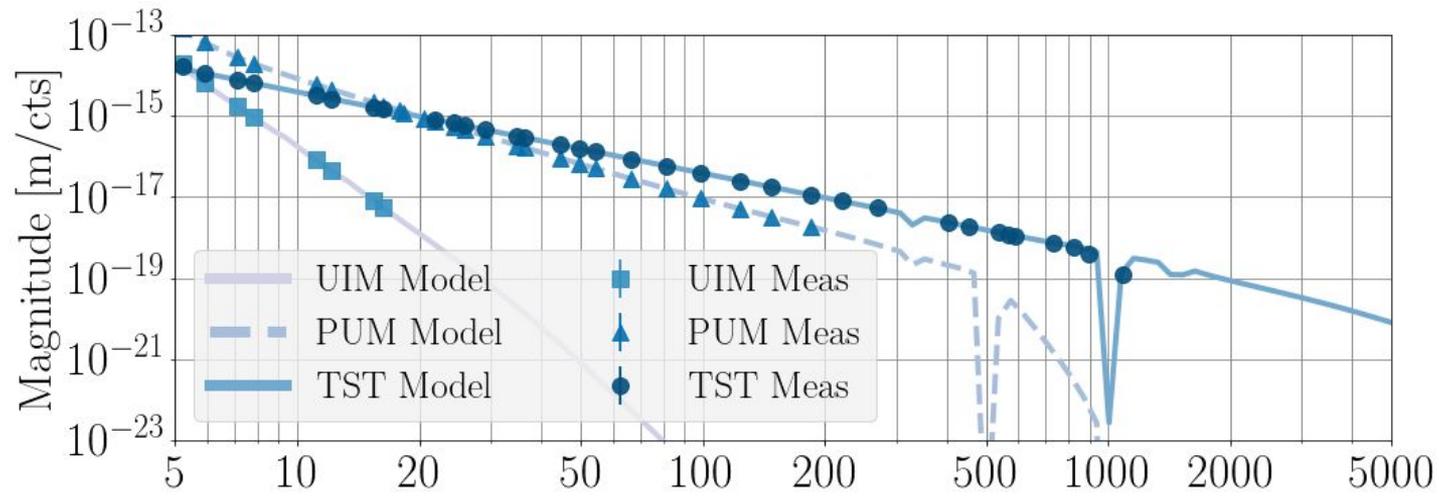
Electrostatic
Drive



From [5]



Meas Date:
Jan 4, 2017



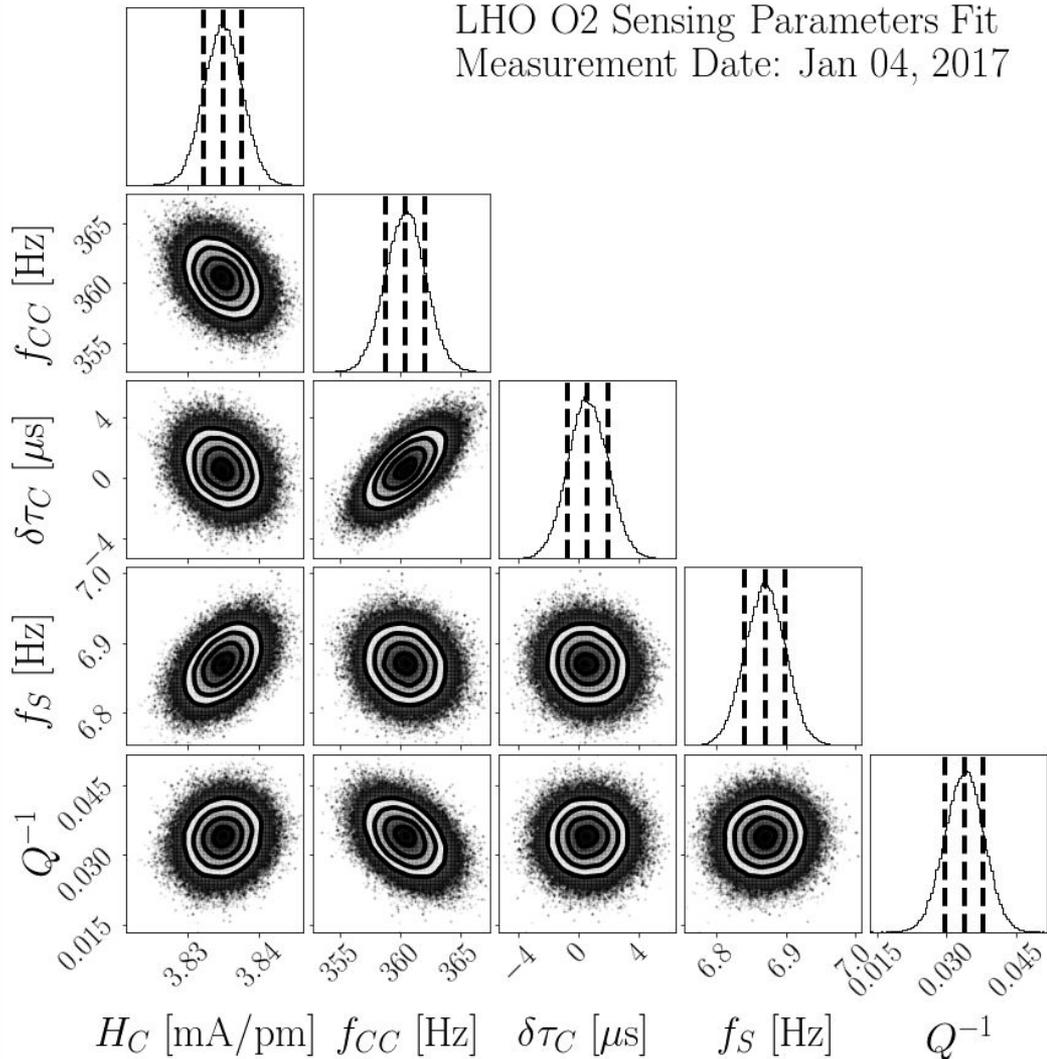
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Nov 26, 2016

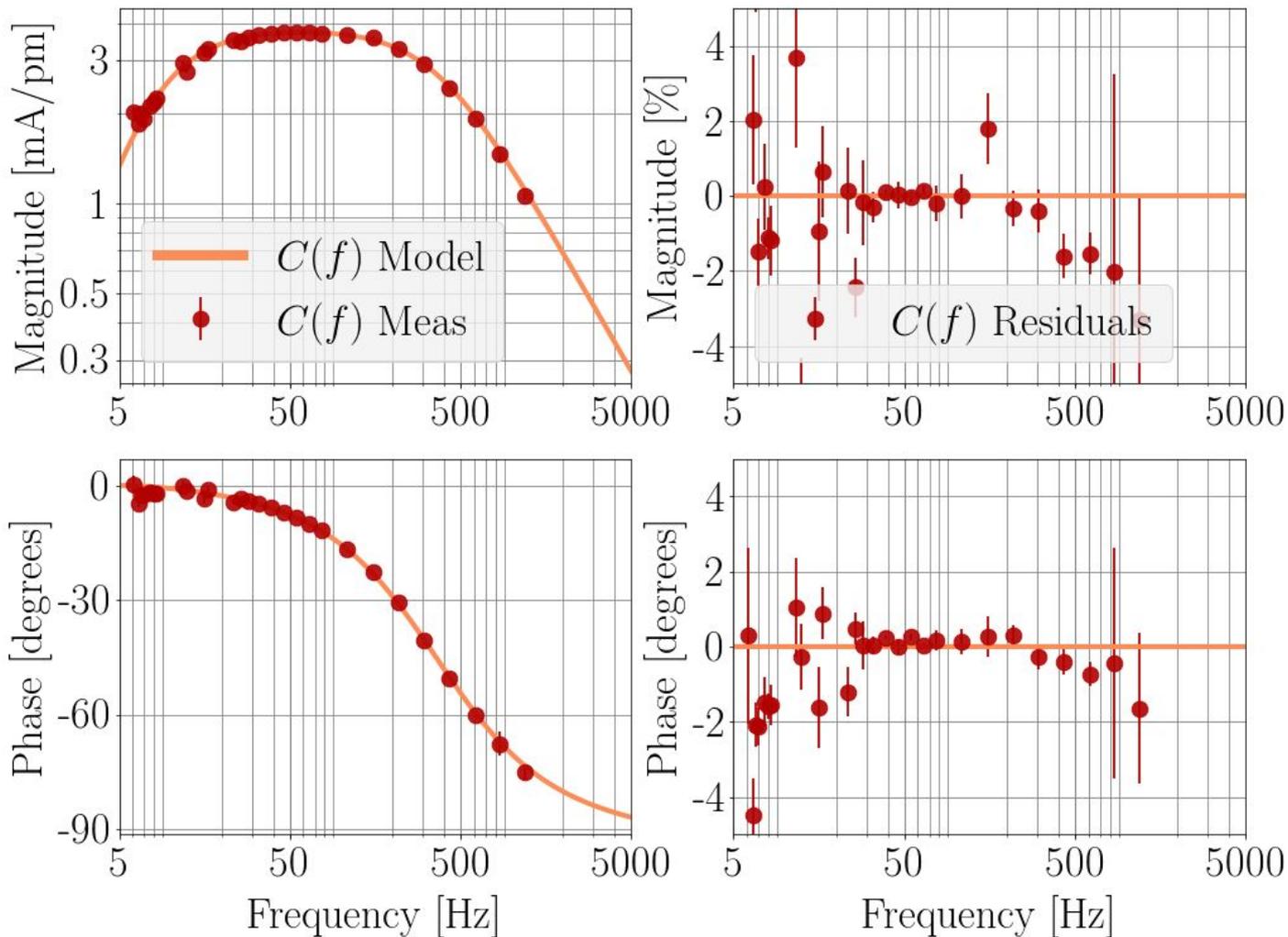
Sensing Model Parameter Estimation

- We have calibration parameters $\vec{\lambda}$ which describe the state of our detector.
 - Optical gain
 - Coupled cavity pole
 - Time delay
 - Optical spring frequency
 - Optical spring inverse Q
- We have a calibration model $M(\vec{\lambda})$ and measurements \vec{d} .
- We use a Markov Chain Monte Carlo (MCMC) method to find the most likely parameter values $\vec{\lambda}$ given our data \vec{d} and model $M(\vec{\lambda})$:

$$\log \mathcal{L}(\vec{d} | M, \vec{\lambda})$$

LHO O2 Sensing Parameters Fit
Measurement Date: Jan 04, 2017

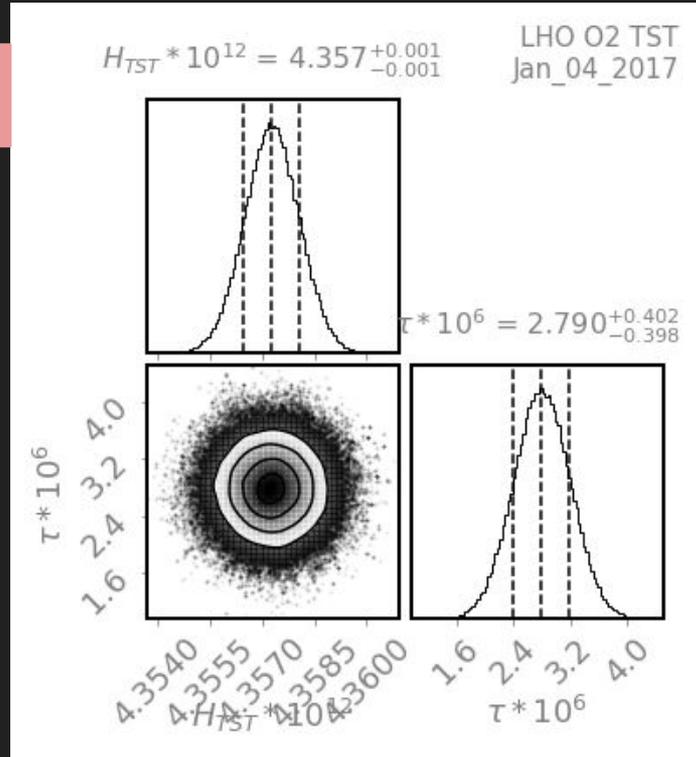




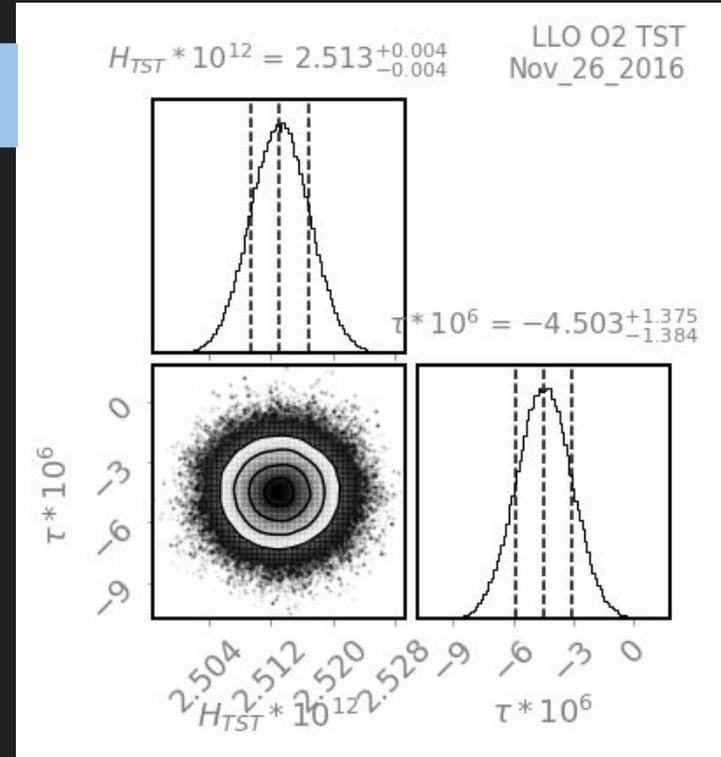
Actuation Model Parameter Estimation

- Just two parameters here: Gain and Delay
- Do this for all three stages of actuation: A_{UIM} , A_{PUM} , and A_{TST} .

LHO



LLO



Estimating Unmodeled Deviations from Measurement

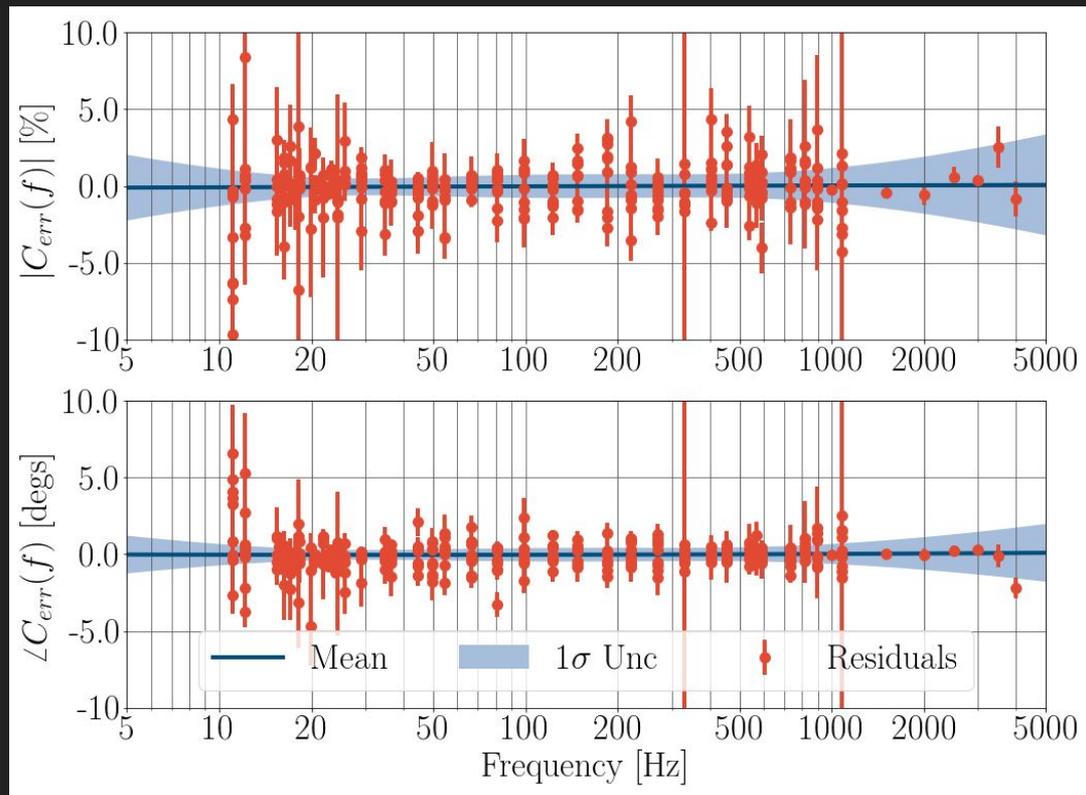
- Want to find deviations from the calibration model for the sensing and actuation functions.
 - Known as systematic biases, or systematic errors
- Also need rigorous uncertainty estimation in this systematic bias

- Gaussian Process Regression $f(\vec{x}) = \mathcal{GP}(m(\vec{x}), k(\vec{x}, \vec{x}'))$
 - Mean Function: $m(\vec{x})$
 - Covariance Kernel: $k(\vec{x}, \vec{x}')$
 - “A Gaussian Process is a collection of random variables, any finite number of which have a joint Gaussian distribution.” [6]
 - Uses training data \vec{x} and covariance kernel $k(\vec{x}, \vec{x}')$ to create a distribution over functions $f(\vec{x})$
 - **With this function distribution, we may rigorously sample to get potential fits to our training data**



Gaussian Process Regression

- Fit to residuals (meas/model) for our four functions $C(f)$, $A_U(f)$, $A_P(f)$, $A_T(f)$
- Shown: LLO Sensing Gaussian Process Regression Results
- Assumptions
 - Functions are smooth
 - Can be described by simple lines
 - Uncertainty is gaussian
 - Time dependence of measurements is removed
 - Can stack measurements

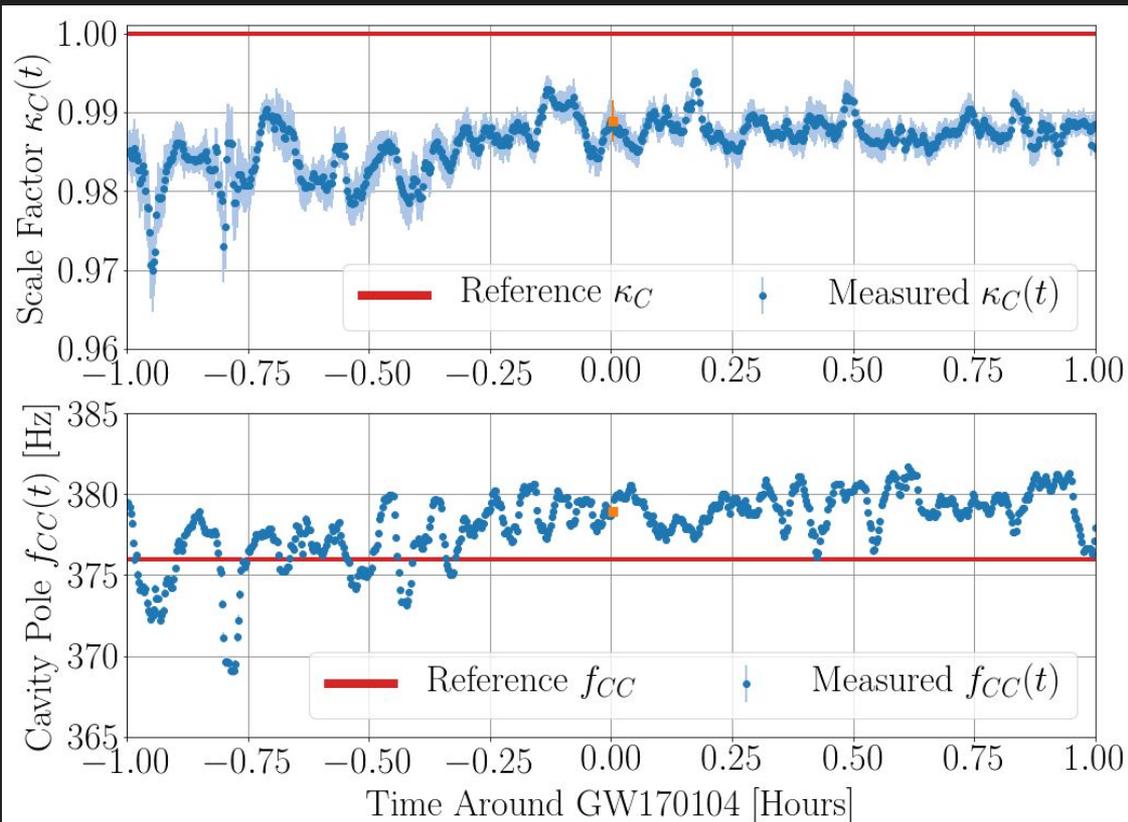


Time Dependent Parameter Uncertainty

We track changes in the interferometer in real time using calibration lines.

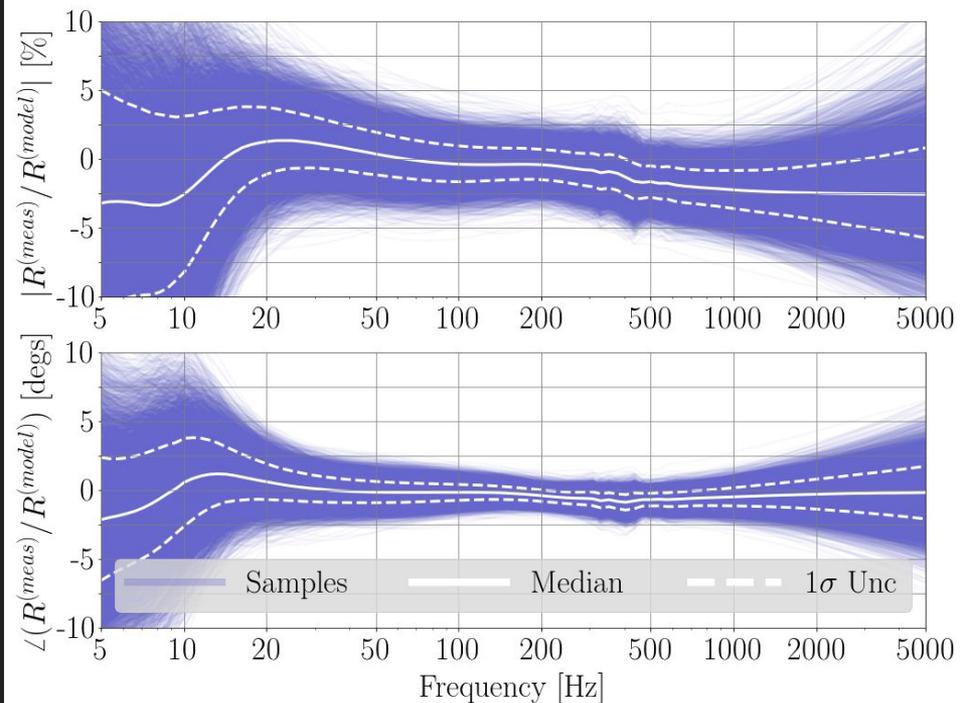
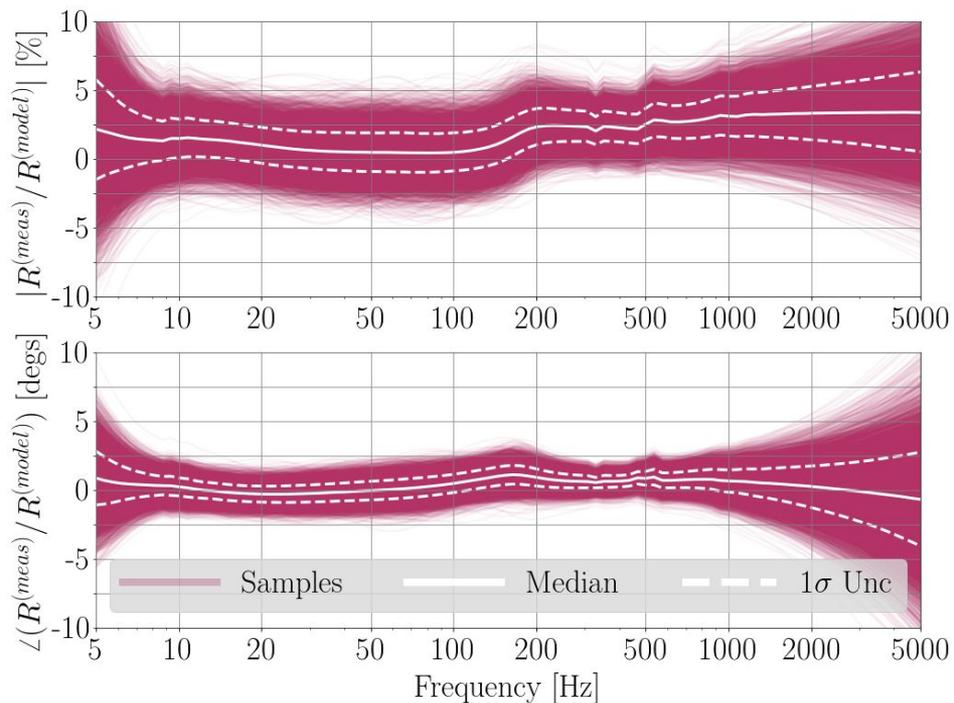
- Optical Gain
- Coupled Cavity Pole
- Electrostatic Drive Strength
- Electromagnetic Coil Drive Strength

Using the coherence of our calibration lines, we can calculate uncertainty in the lines themselves, and propagate forward to the time-dependent detector parameters.

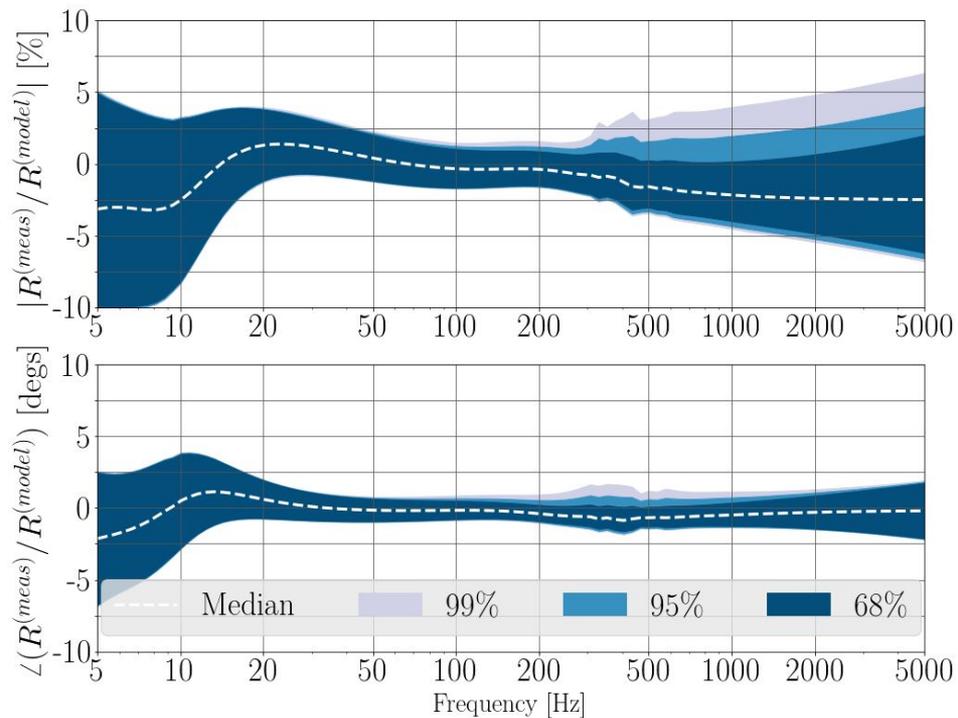
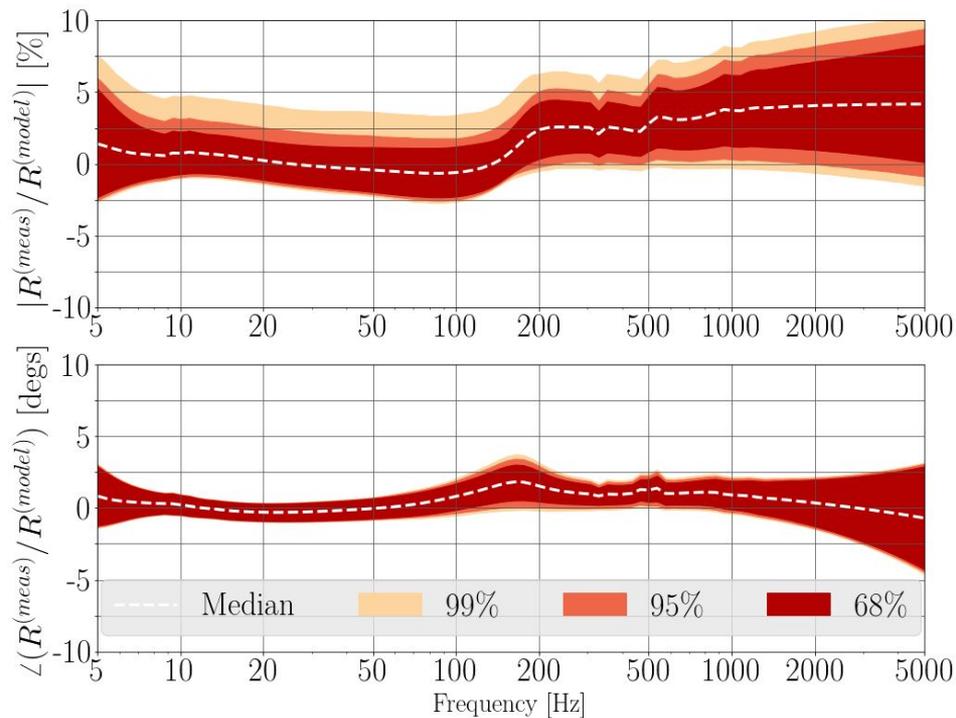
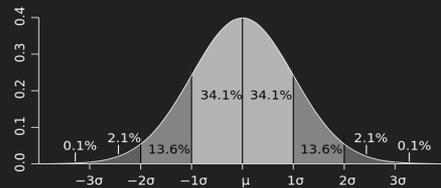


GW170104 Uncertainty Budgets

Extreme Uncertainties	Hanford	Livingston
1σ Magnitude [%]	-1.0 to +4.6	-3.7 to +3.7
1σ Phase [degrees]	-0.9 to +1.8	-1.5 to +1.9

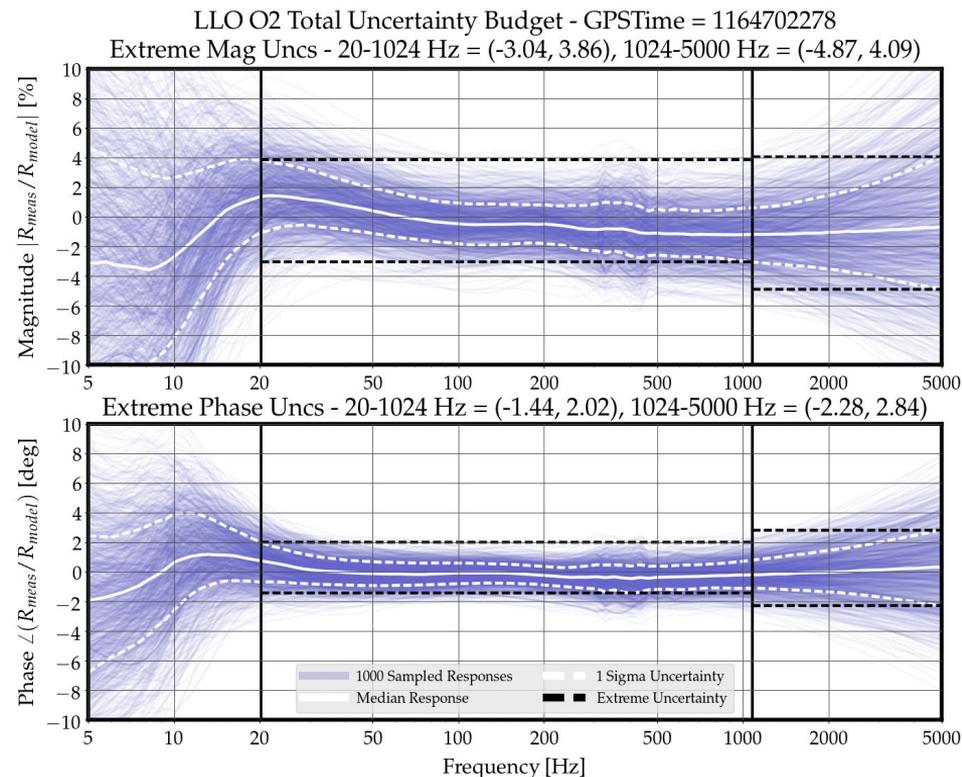
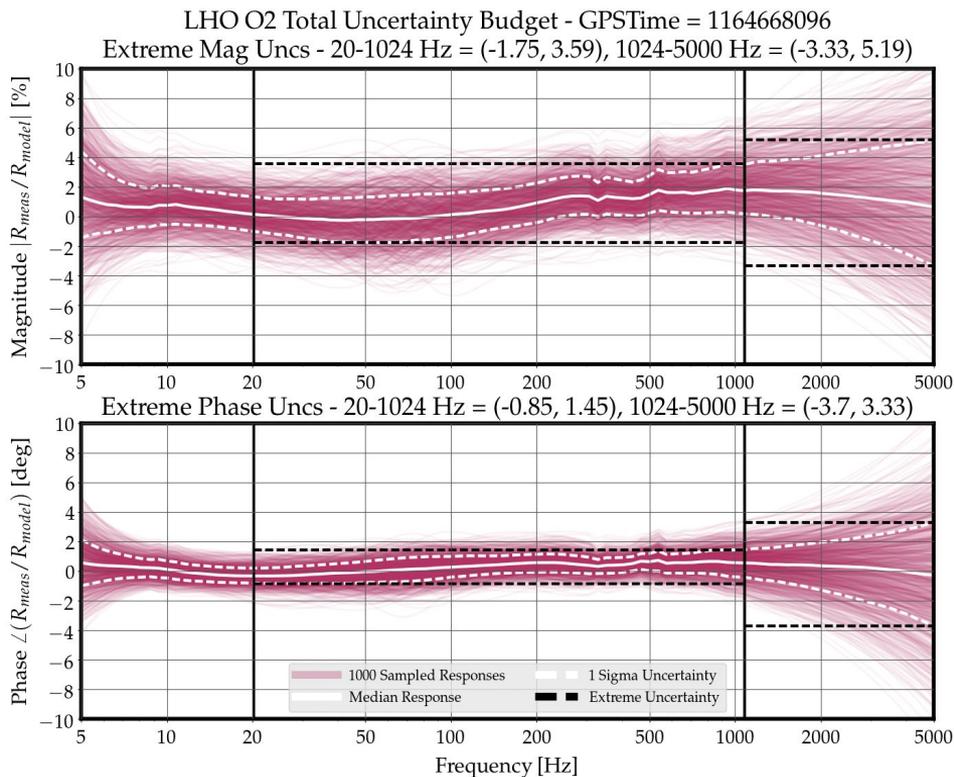


Nov - Jun O2 Uncertainty Budget Percentiles





Nov - Jun O2 Uncertainty Budget Movie



Conclusion

- The uncertainty in gravitational wave strain data is improved from 10% and 10 degrees to ~ 7.4% and 3.4 degrees from 20 to 1024 Hz for both detectors.
- The uncertainty budget is frequency dependent and quantifies known systematic biases
- This information from the uncertainty pipeline is getting incorporated into astrophysical parameter estimation pipelines
- Future work to further push down calibration uncertainty is underway



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