



Stochastic Gravitational Wave Background from Non-Tensorial Polarizations

Sylvia Biscoveanu
Caltech LIGO SURF 2016
Mentor: Tom Callister

Overview

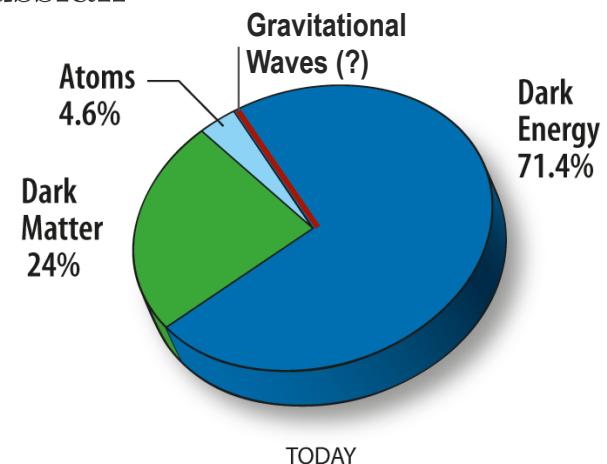
- The Stochastic Background
- Gravitational Wave Polarizations
- Project Description and Motivation
- Cross-Correlated Analysis
- Simulations
- O1 Result
- Conclusion and Outlook

The Stochastic Background

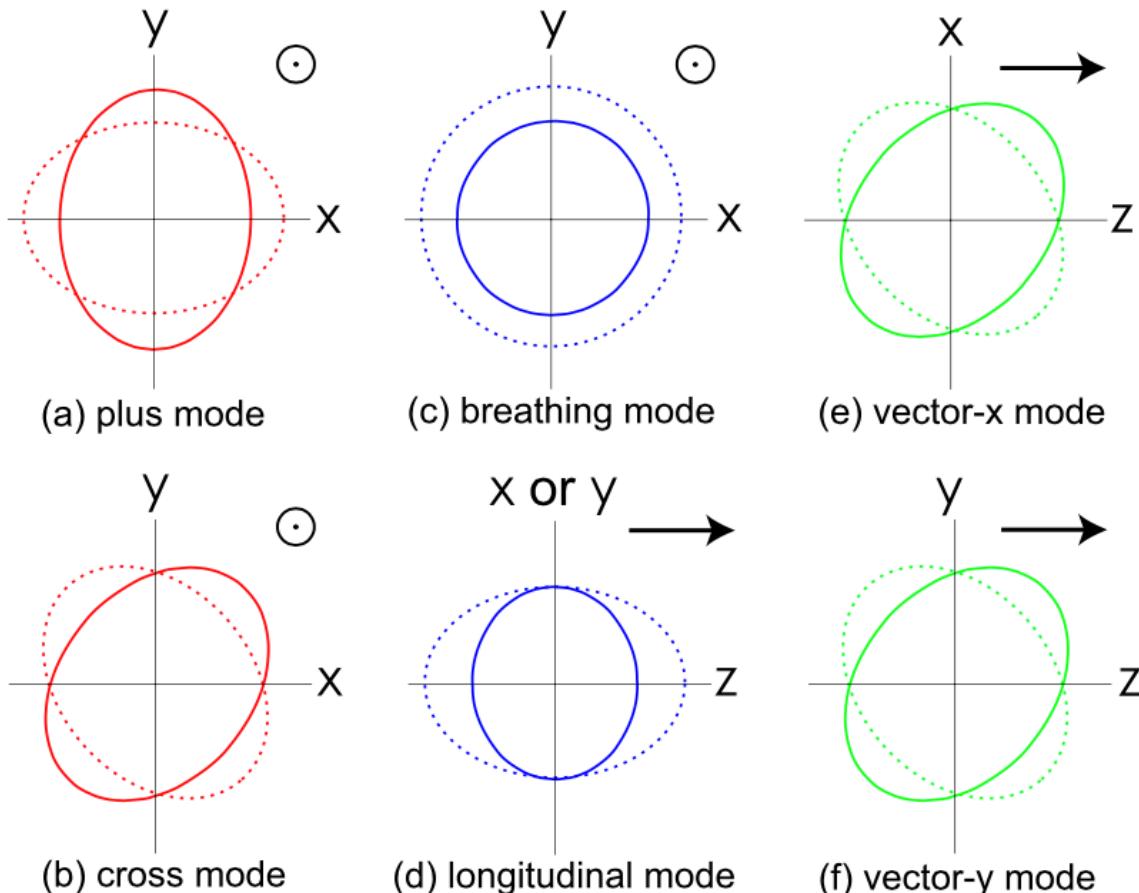
- A random gravitational wave signal produced by many overlapping, individually indistinguishable sources
 - Astrophysical
 - Cosmological
- Isotropic, unpolarized, stationary, and Gaussian

$$\Omega_{\text{gw}}(f) = \frac{1}{\rho_{\text{critical}}} \frac{d\rho_{\text{gw}}}{d \ln f}$$

$$\Omega_{\text{gw}} = \Omega_\alpha (f/f_0)^\alpha$$



Gravitational Wave Polarizations



- Extra polarizations due to extra degrees of freedom associated with scalar fields or massive gravity
- Stochastic search can provide information about gravitation in the early universe and about sources which are too rare for individual detections.

Project Description

- Modify the stochastic search pipeline to allow for searches for vector and scalar polarizations
- Run the search over O1 data
- Use a Bayesian approach to model selection and component separation

Cross Correlated Analysis - GR

$$\tilde{s}(f) = \tilde{h}(f) + \tilde{n}(f)$$

- Correlate the strain signal from two detectors sufficiently far apart to minimize common noise sources

$$Y \propto \langle \tilde{s}_I^*(f) \tilde{s}_J(f) \rangle \quad \sigma^2 \propto \frac{P_I(f) P_J(f)}{\gamma_{IJ}^2(f)}$$

$$\langle Y \rangle = \Omega_{gw} \quad \text{SNR} = \frac{\langle Y \rangle}{\sigma}$$

Cross Correlated Analysis – Non GR

$$\tilde{s}(f) = \tilde{h}(f) + \tilde{n}(f)$$

- Correlate the strain signal from two detectors sufficiently far apart to minimize common noise sources

$$Y \propto \langle \tilde{s}_I^*(f) \tilde{s}_J(f) \rangle \quad \sigma^2 \propto P_I(f) P_J(f)$$

$$\langle Y \rangle = \sum_A \gamma_A \Omega_A$$

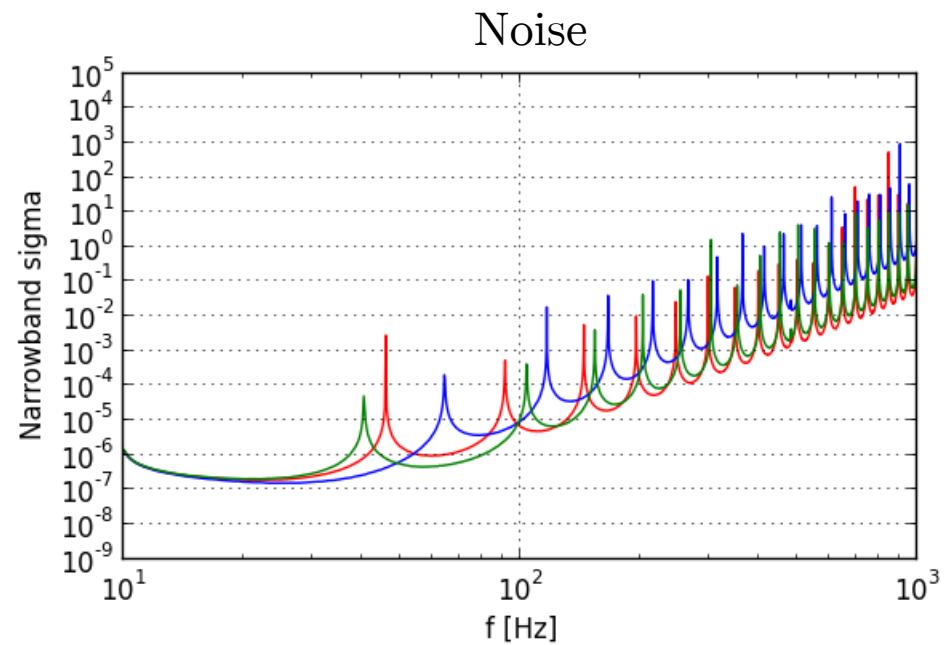
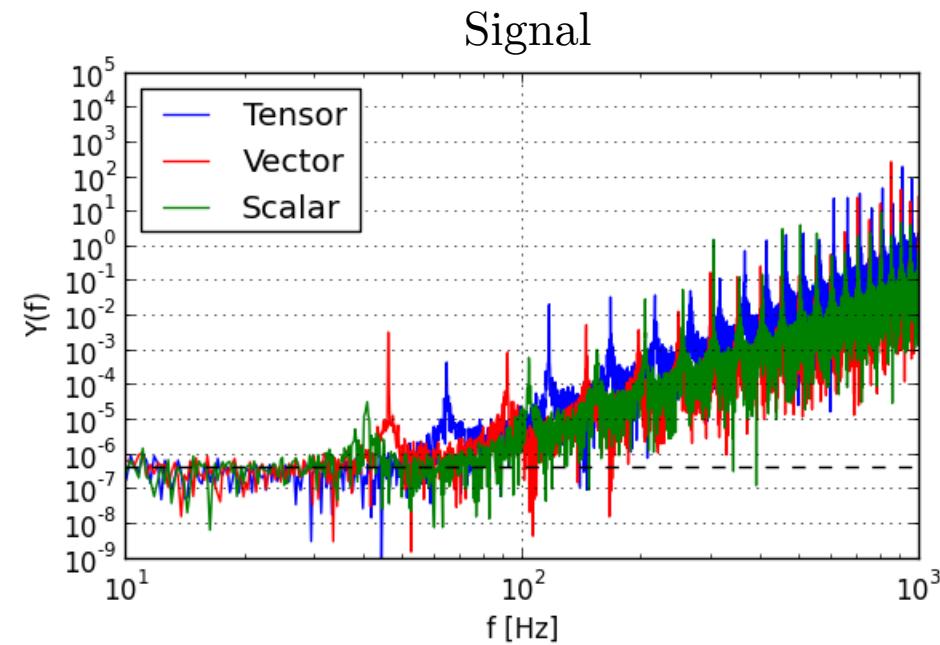
$$\text{SNR}^2 = \frac{\left[\frac{Y \gamma_M \Omega_M}{\sigma^2} \right]^2}{\frac{[\gamma_M \Omega_M]^2}{\sigma^2}}$$

Simulations

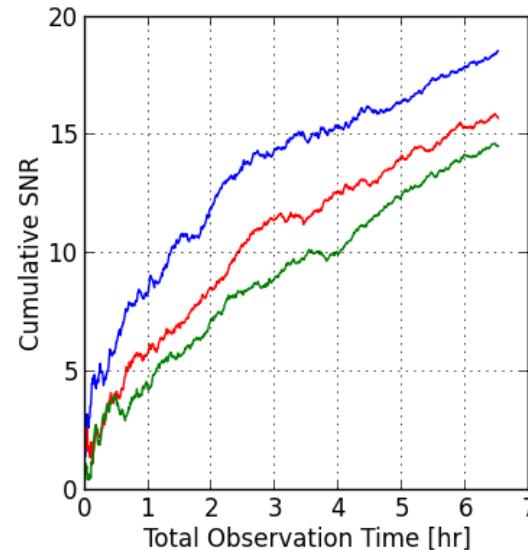
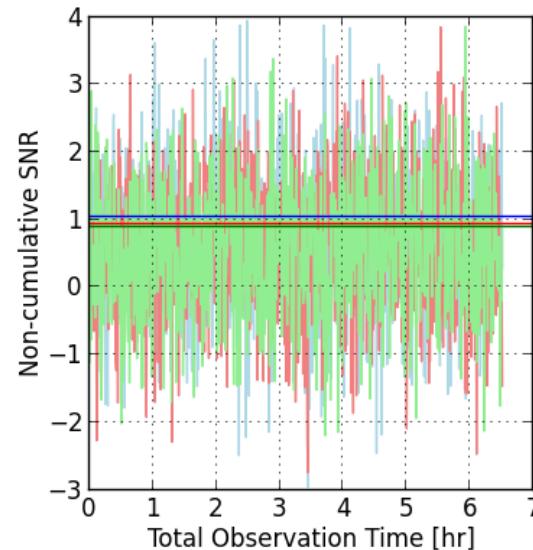
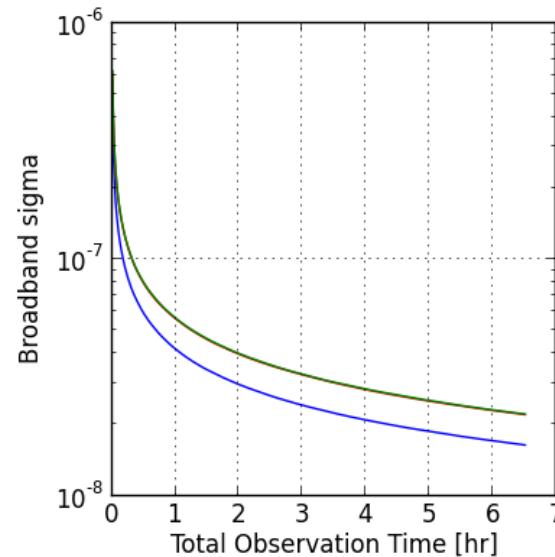
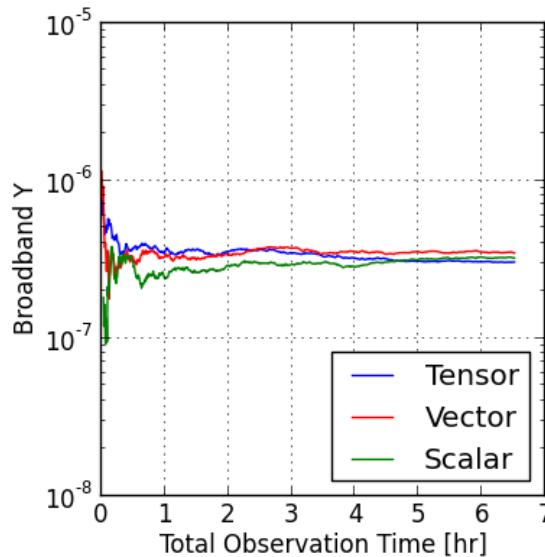
- Flat signal ($\alpha = 0$) with amplitude $\Omega_{gw} = 2 \times 10^{-7}$

	Recover Tensor	Recover Vector	Recover Scalar
Inject Tensor	18.56	15.72	14.55
	18.90	16.76	14.00
Inject Vector	12.88	12.89	12.10
	12.58	14.19	13.73
Inject Scalar	10.70	13.53	14.84
	10.30	13.45	13.91

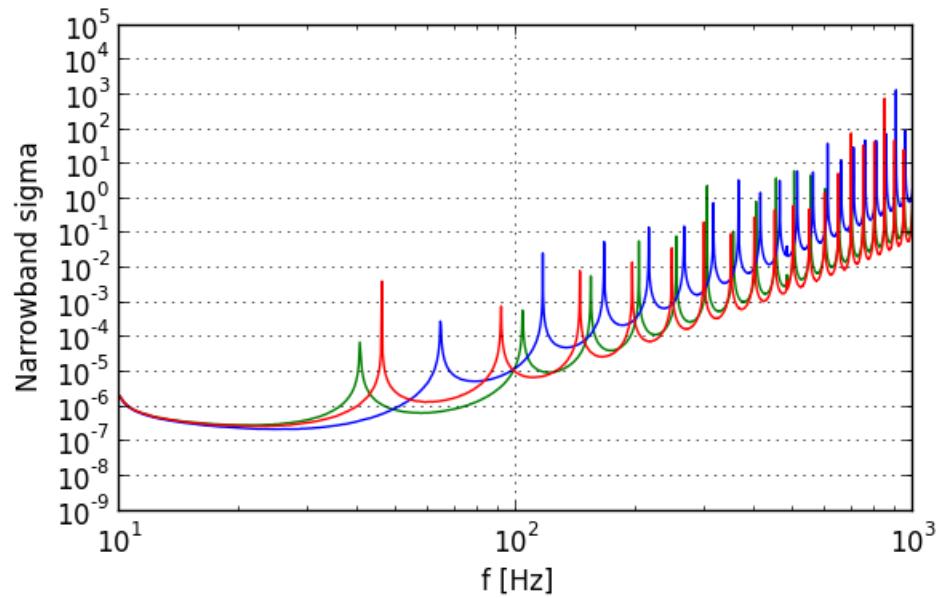
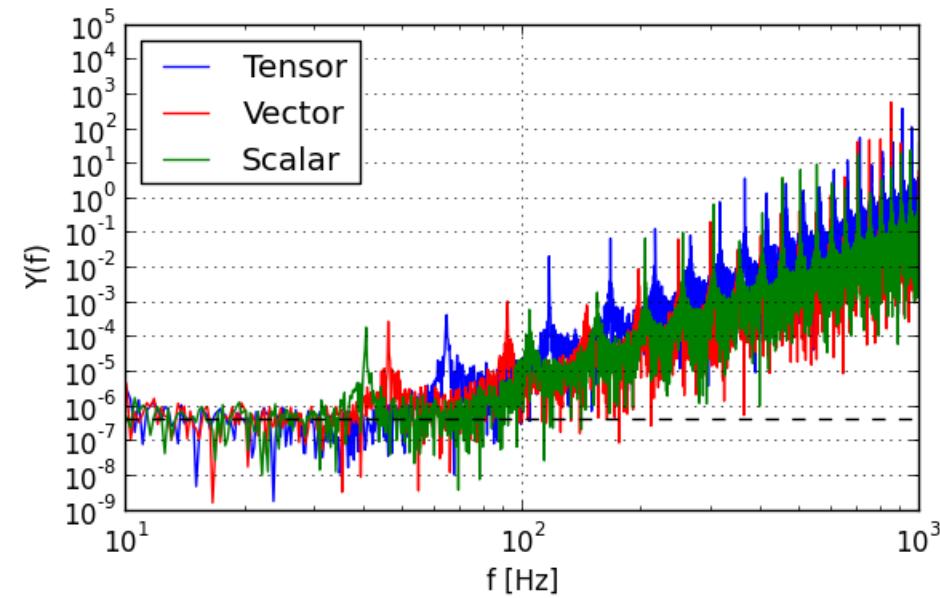
Injecting a Tensor Signal



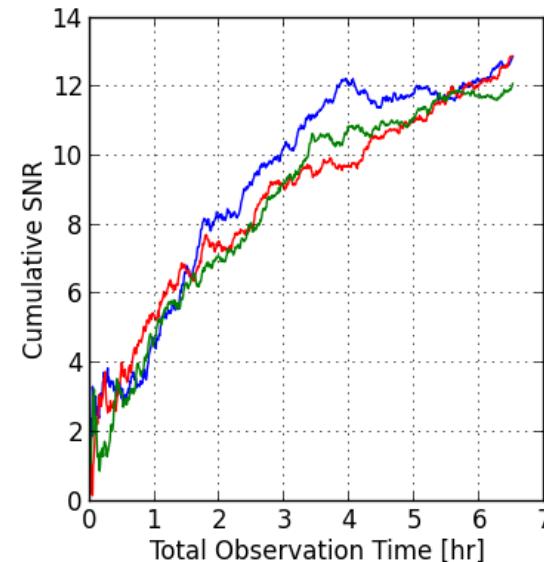
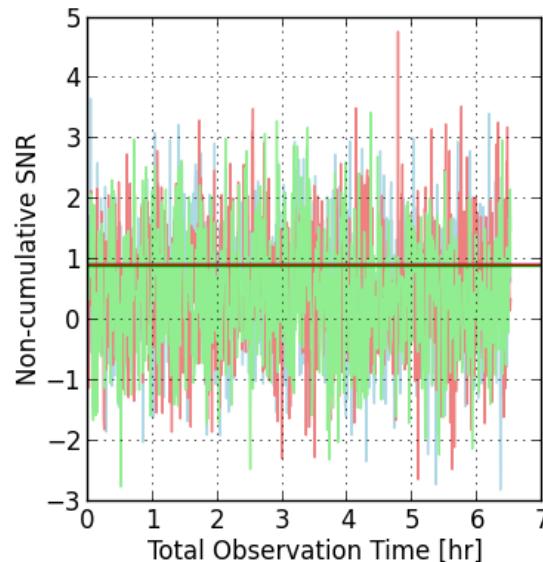
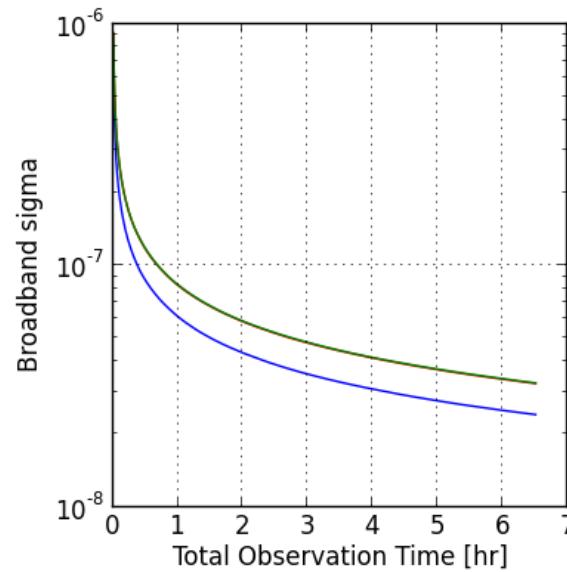
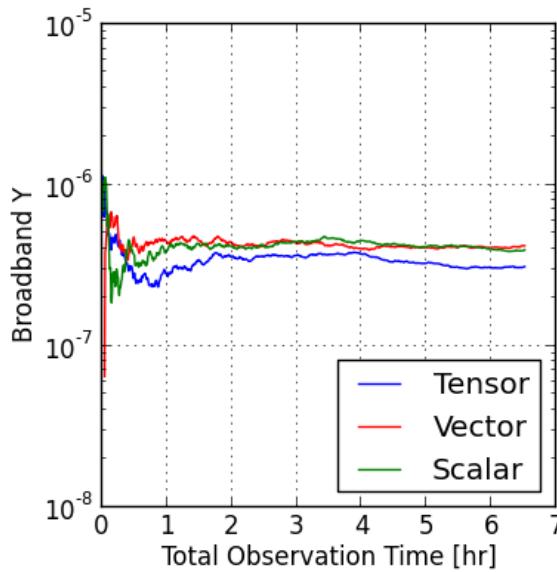
Injecting a Tensor Signal



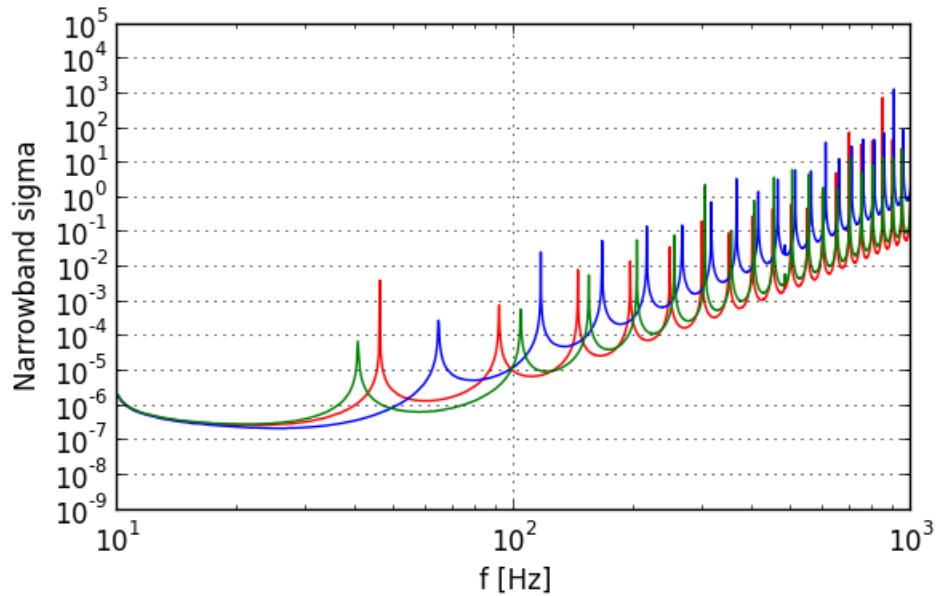
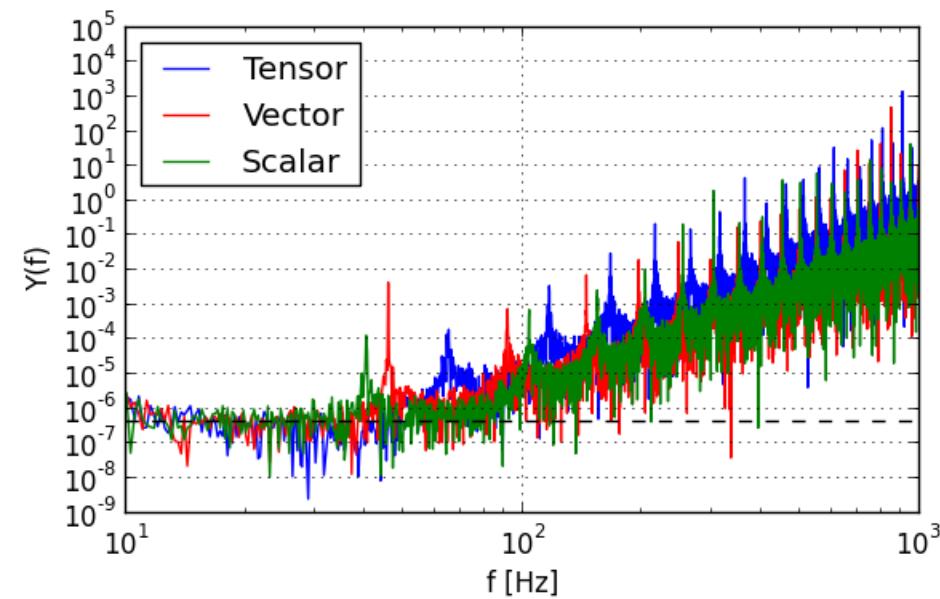
Injecting a Vector Signal



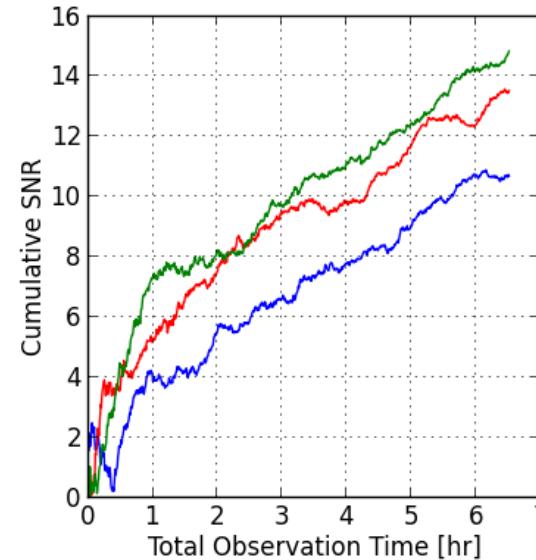
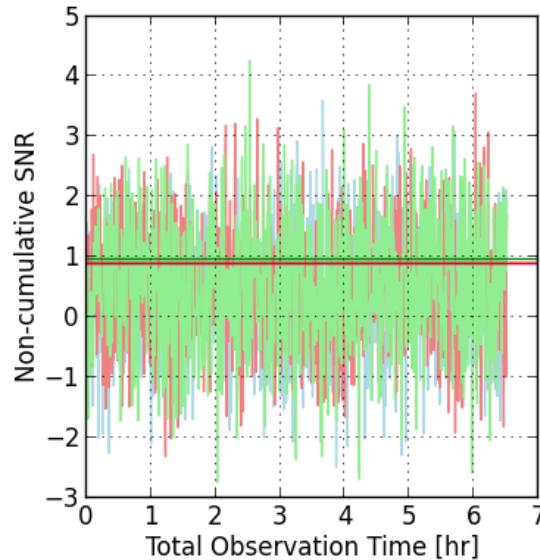
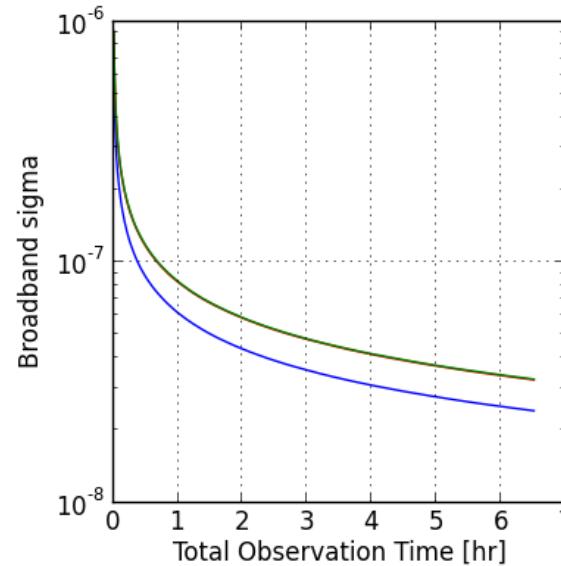
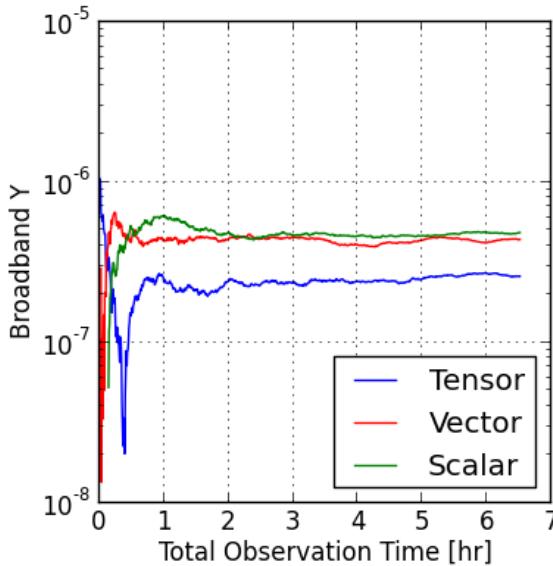
Injecting a Vector Signal



Injecting a Scalar Signal

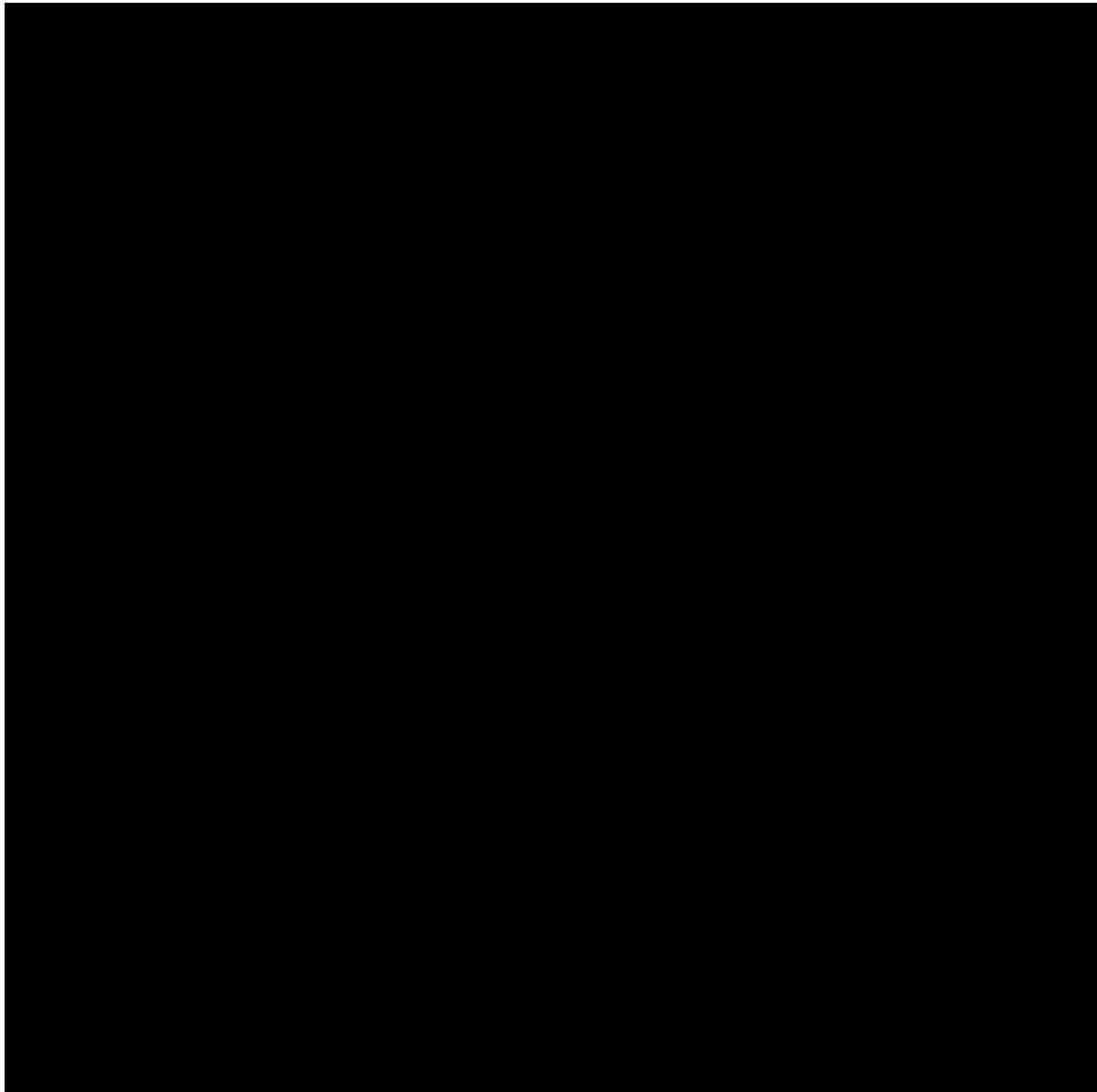


Injecting a Scalar Signal



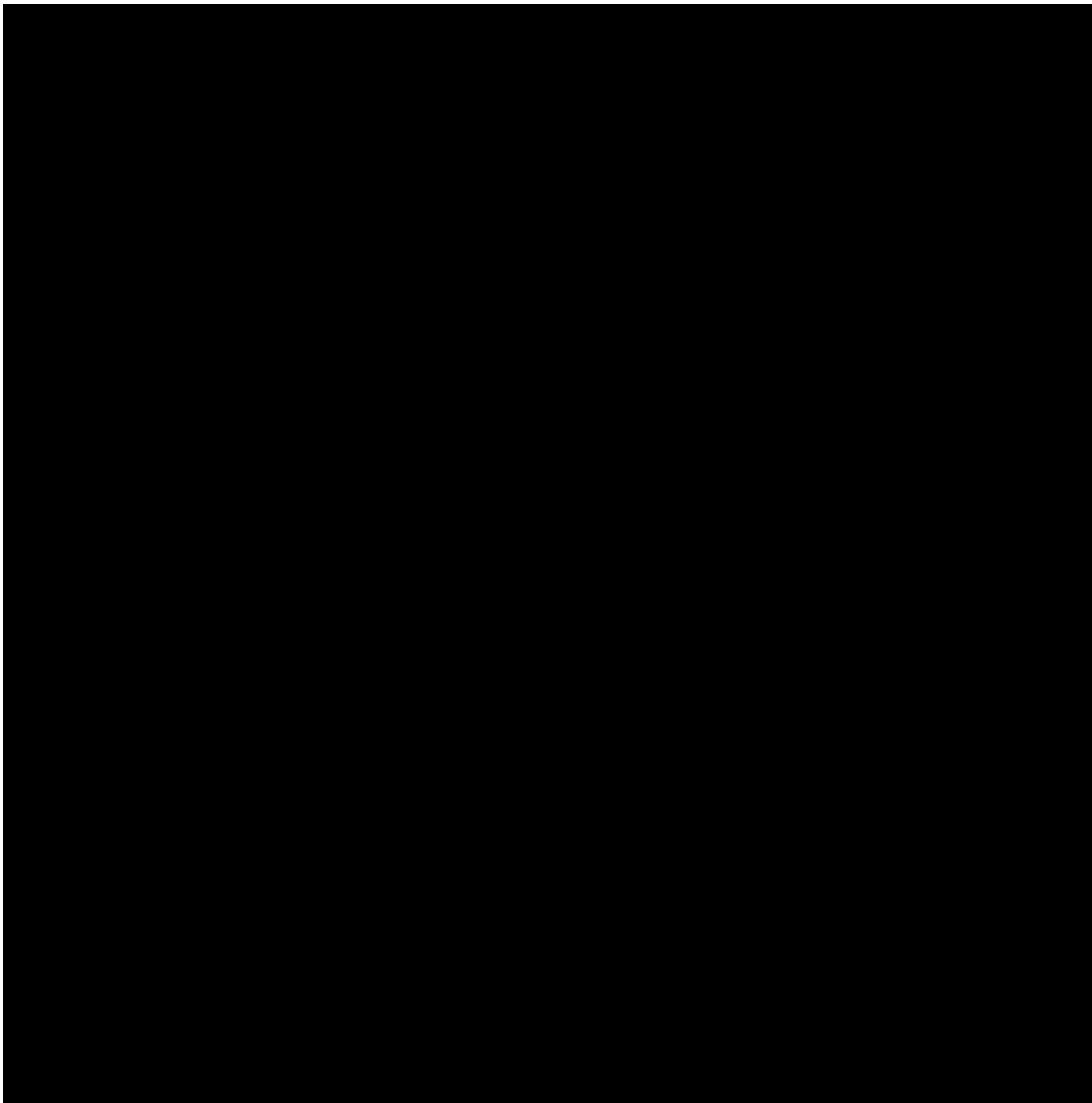
O1 Vector Result

O1 Vector Result



O1 Scalar Result

O1 Scalar Result



Conclusion and Outlook

- [REDACTED]
- [REDACTED]
- [REDACTED]
- Long term goal: Prepare a component separation scheme for a three detector system to be implemented when Virgo comes live.

Acknowledgements

- Thank you to Alan, the LIGO stochastic group, and my mentor Tom for the patience, guidance, and support.
- Thank you to the NSF and the CIT LIGO Lab for sponsoring and organizing this project.

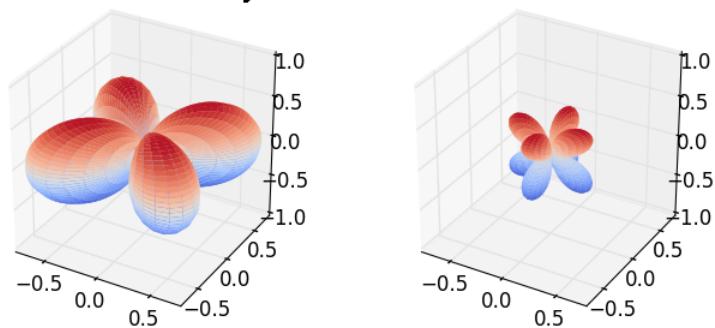
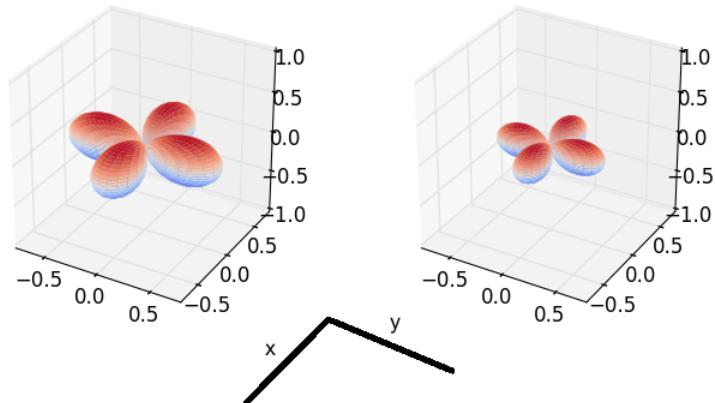
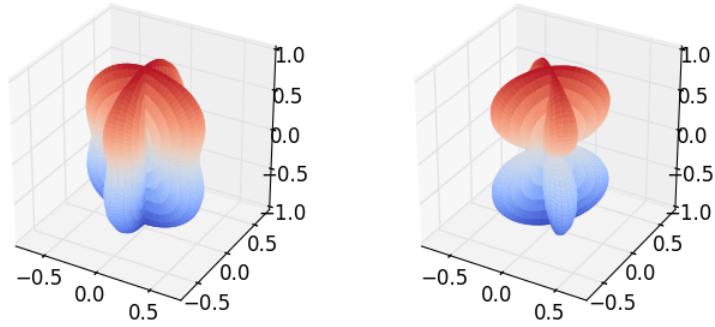
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- [15] Correspondence with Joe Romano.

Backup Slides

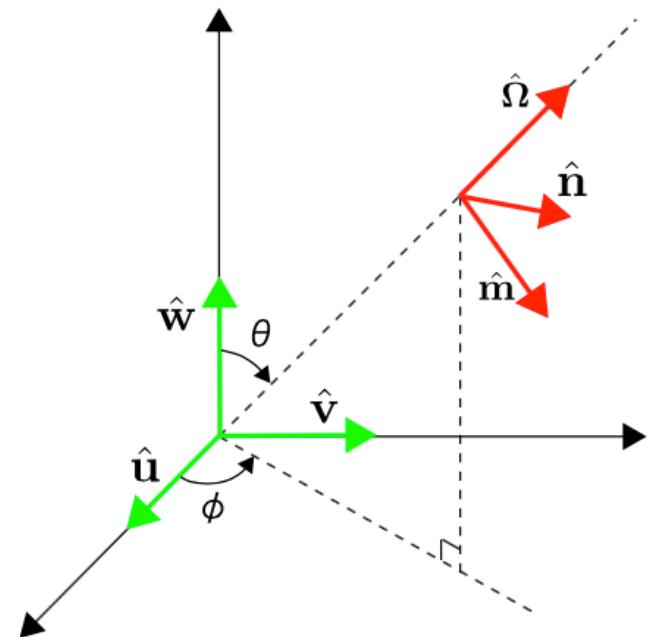
Single Detector Response

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$$F_A(\hat{\Omega}) = \mathbf{D} : \tilde{\mathbf{e}}_A(\hat{\Omega})$$

$$\mathbf{D} = \frac{1}{2}(\hat{\mathbf{u}} \otimes \hat{\mathbf{u}} - \hat{\mathbf{v}} \otimes \hat{\mathbf{v}})$$



Overlap Reduction Functions

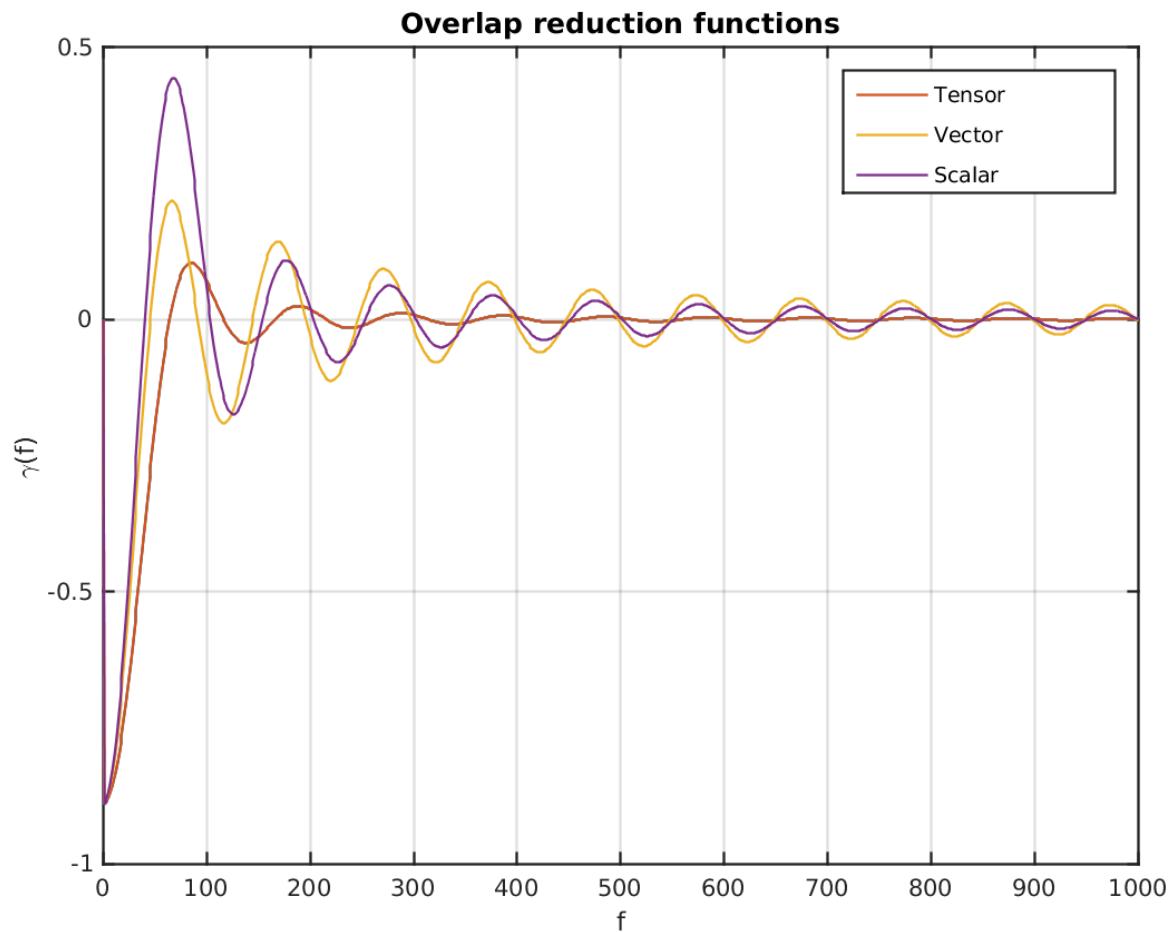
$$\gamma_{IJ}^T = \frac{5}{8\pi} \int_{S^2} e^{2\pi i f \hat{\Omega} \cdot \Delta \vec{X}/c} (F_I^+ F_J^+ + F_I^\times F_J^\times) d\Omega$$

$$\gamma_{IJ}^V = \frac{5}{8\pi} \int_{S^2} e^{2\pi i f \hat{\Omega} \cdot \Delta \vec{X}/c} (F_I^x F_J^x + F_I^y F_J^y) d\Omega$$

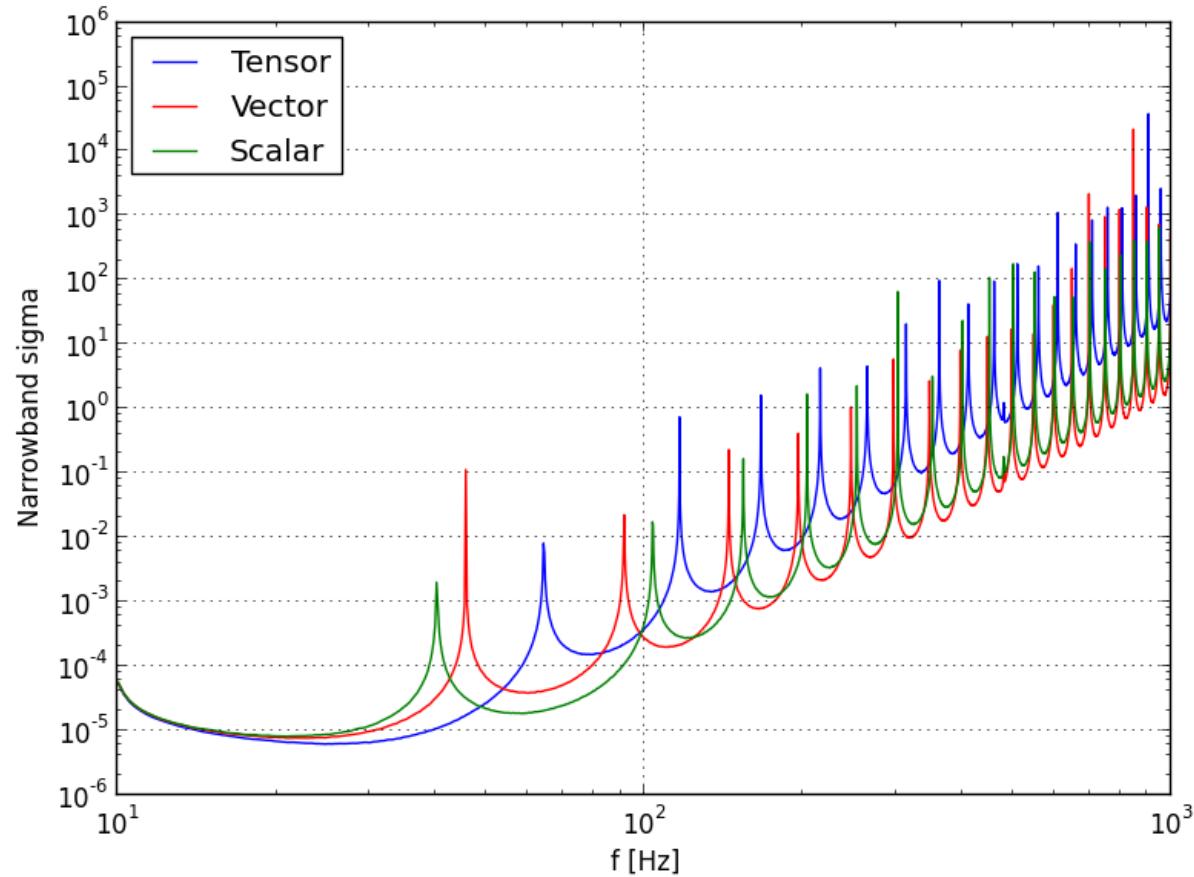
$$\gamma_{IJ}^S = \frac{5}{4\pi} \int_{S^2} e^{2\pi i f \hat{\Omega} \cdot \Delta \vec{X}/c} (F_I^l F_J^l + F_I^b F_J^b) d\Omega$$

- Normalized to 1 for coincident and coaligned detectors
- Separation and lack of perfect arm alignment reduces sensitivity in cross-correlated analysis

Overlap Reduction Functions



Simulated Gaussian Noise



Broadband Estimators

$$Y_i = \frac{\sum_f Y(f) \sigma^{-2}(f)}{\sum_f \sigma^{-2}(f)}$$

$$\sigma_i^{-2} = \sum_f \sigma^{-2}(f)$$

$$Y_C = \frac{\sum_i Y_i \sigma_i^{-2}}{\sum_i \sigma_i^{-2}}$$

$$\sigma_C^{-2} = \sum_i \sigma_i^{-2}$$

Signal to Noise Ratio

- Optimal Filter found by maximizing the SNR

$$\mu \equiv \langle Y \rangle = \frac{3H_0^2}{10\pi^2} T \sum \frac{\gamma(f)\Omega_{gw}(f)\tilde{Q}(f)}{f^3} df,$$

$$\sigma^2 = \frac{T}{2} \sum \tilde{Q}^2(f) P_1(f) P_2(f) df$$

$$\begin{aligned} \tilde{Q}(f) &= K \frac{\gamma(f)\Omega_{gw}(|f|)}{|f|^3 P_I(|f|) P_J(|f|)} \\ \text{SNR}^2 &= \frac{\mu^2}{\sigma^2} = \frac{\left[\frac{3H_0^2}{10\pi^2} T \sum \frac{\gamma(f)\Omega_{gw}(f)\tilde{Q}(f)}{f^3} df \right]^2}{\frac{T}{2} \sum \tilde{Q}^2(f) P_1(f) P_2(f) df} \end{aligned}$$

Signal to Noise Ratio

- Redefine the filter such that $\mu = \gamma_A \Omega_A$

$$\tilde{Q}(f) = \left(\frac{3H_0^2}{10\pi^2} \right) \frac{2}{f^3 P_1(f) P_2(f)}$$

$$\text{SNR}^2 = \left(\frac{3H_0^2}{10\pi^2} \right)^2 2T \frac{\left[\sum \frac{\gamma(f) \Omega_{gw}(f) \gamma_M(f) \Omega_M(f)}{f^6 P_1(f) P_2(f)} df \right]^2}{\sum \frac{(\gamma_M(f) \Omega_M(f))^2}{f^6 P_1(f) P_2(f) df}}$$

Component Separation

$$p(Y|\mathcal{A}) \propto \exp \left[-\frac{1}{2}(Y - M\mathcal{A})^T \mathcal{N}^{-1}(Y - M\mathcal{A}) \right]$$

$$M = \begin{bmatrix} \gamma^T(f_1) \left(\frac{f_1}{f_0}\right)^\alpha & \gamma^V(f_1) \left(\frac{f_1}{f_0}\right)^\alpha & \gamma^S(f_1) \left(\frac{f_1}{f_0}\right)^\alpha \\ \gamma^T(f_2) \left(\frac{f_2}{f_0}\right)^\alpha & \gamma^V(f_2) \left(\frac{f_2}{f_0}\right)^\alpha & \gamma^S(f_2) \left(\frac{f_2}{f_0}\right)^\alpha \\ \vdots & \vdots & \vdots \\ \gamma^T(f_N) \left(\frac{f_N}{f_0}\right)^\alpha & \gamma^V(f_N) \left(\frac{f_N}{f_0}\right)^\alpha & \gamma^S(f_N) \left(\frac{f_N}{f_0}\right)^\alpha \end{bmatrix} \quad \mathcal{A} = \begin{bmatrix} \Omega^T \\ \Omega^V \\ \Omega^S \end{bmatrix}$$

$$\mathcal{A} = F^{-1}X, \quad F \equiv M^T \mathcal{N}^{-1} M, \quad X \equiv M^T \mathcal{N}^{-1} Y$$