

# Gravitation Wave Identification

Using Metropolis Hastings Algorithm

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Group 1

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# Problem Statement

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# Problem Statement

Given a time series strain data with added noise with the structure of a gravitational wave as given below

$$h(t) = \alpha e^t [1 - \tanh\{2(t - \beta)\}] \sin(\gamma t)$$

$\alpha, \beta, \gamma$  are parameters that signify the physical properties of the given wave.  
Their value ranges are

$$0 < \alpha < 2$$

$$1 < \beta < 10$$

$$1 < \gamma < 20$$

We need to determine the parameter values using a **Metropolis Hastings** Random walk algorithm in the 3 dimensional space.

# Understanding Wave Parameters

Let us visualize how the parameters  $\alpha$ ,  $\beta$ , and  $\gamma$  influence the waveform

1.  $\alpha$  controls the amplitude of the signal
2.  $\beta$  shifts the signal in time
3.  $\gamma$  controls the oscillation frequency

## **Animation Paramater effects**

# Methodology

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# Random Walks

1. **Initialization:** We start with initial parameter values at the midpoints of the given ranges so

$$\alpha = 1, \beta = 5, \gamma = 10$$

2. **Random Walk:**

For each iteration we propose a new set of parameters using

$$\theta_{\text{new}} = \text{normal}(\theta_{\text{initial}}, \sigma^2) \quad \text{where } \sigma = [0.01, 0.07, 0.07]$$

The new value is discarded or chosen based on an **Acceptance Probability** defined as

$$A(\theta_{\text{new}}, \theta_{\text{initial}}) = \min\left(1, \frac{\text{Posterior}(\theta_{\text{new}})}{\text{Posterior}(\theta_{\text{initial}})}\right)$$

The **Posterior** function is defined as the following

```
def likelihood_reduced(y_data: np.ndarray, y_prior: np.ndarray):
    y_err = 0.1 * np.std(y_data)
    Y = np.mean((y_data - y_prior) ** 2) / y_err**2
    return -0.5 * Y
```

This is different from the function provided in the problem statement, we will explain why this is better in Section 4

# Stochastic Maximum Likelihood Estimation

Hello

# Why not Bayesian Inference ?

Hello

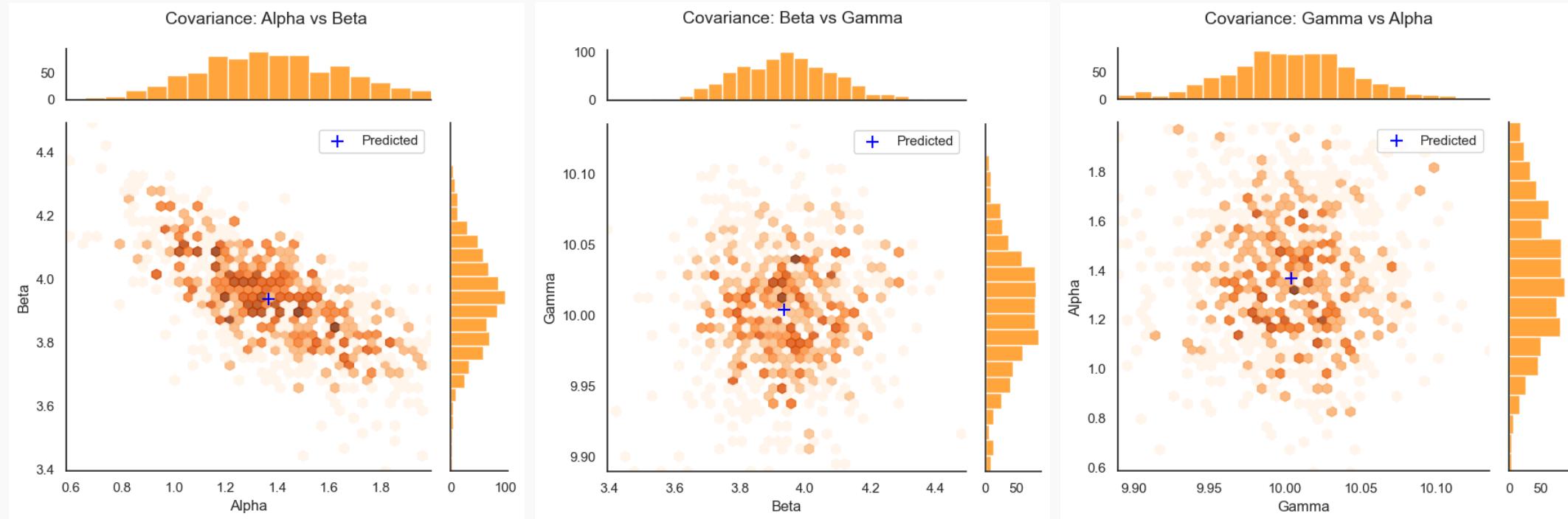
# Results

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# Numerical Analysis

Parameter	$\alpha$ (alpha)	$\beta$ (beta)	$\gamma$ (gamma)
Median Value	1.36	3.94	10.00
95% CI	0.86 - 1.91	N/A	N/A

# Covariance Scatter Plots



# Histograms and Trace Plots

Heloo

# Optimization

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# Scale Selection

Explain step size selection with variance plot

# Likelihood Function Selection

Explain why our likelihood function is better

# Data Generation for Better Inference

Hello

# Conclusion

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# Thank You

Hello