

Phys 512: PS4

1) a) First we consider Newton's method to find optimal parameters for a given fit

↳ Whereas gradient descent simply follows the gradient ∇f along the χ^2 manifold, Newton's method takes into account the curvature & follows $H^{-1} \nabla f$, where H^{-1} is the Hessian (inverse). Thus our step looks like:

$$x \rightarrow x + \delta x : \delta x = -\eta H^{-1} \nabla f, \quad \eta \in \mathbb{R}$$

The following are the derivatives of the Lorentzian:

$$f(t) = \frac{a}{1 + (t - t_0)^2 / \omega^2} = a [1 + \omega^{-2} (t - t_0)^2]^{-1} \cdot \partial_t f(t) = -[1 + \omega^{-2} (t - t_0)^2]^{-2}$$

$$\partial_{t_0} f(t) = (2a(t - t_0) \omega^2) / (1 + (t - t_0)^2 / \omega^2)^2$$

$$\partial_\omega f(t) = (2a(t - t_0)^2 \omega) / (1 + (t - t_0)^2 / \omega^2)^2$$

$$\partial_{\omega^2} f(t) = -a / (1 + (t - t_0)^2 / \omega^2)^2 = -a \omega^{-2} (1 + (t - t_0)^2 / \omega^2)^{-2}$$

This will be useful for the calculations:

$$f(t; m) \equiv A(m), \quad r \equiv d - A(m) \cdot \chi^2 = r^T N^{-1} r \cdot \nabla \chi^2 = -2A'^T N^{-1} r$$

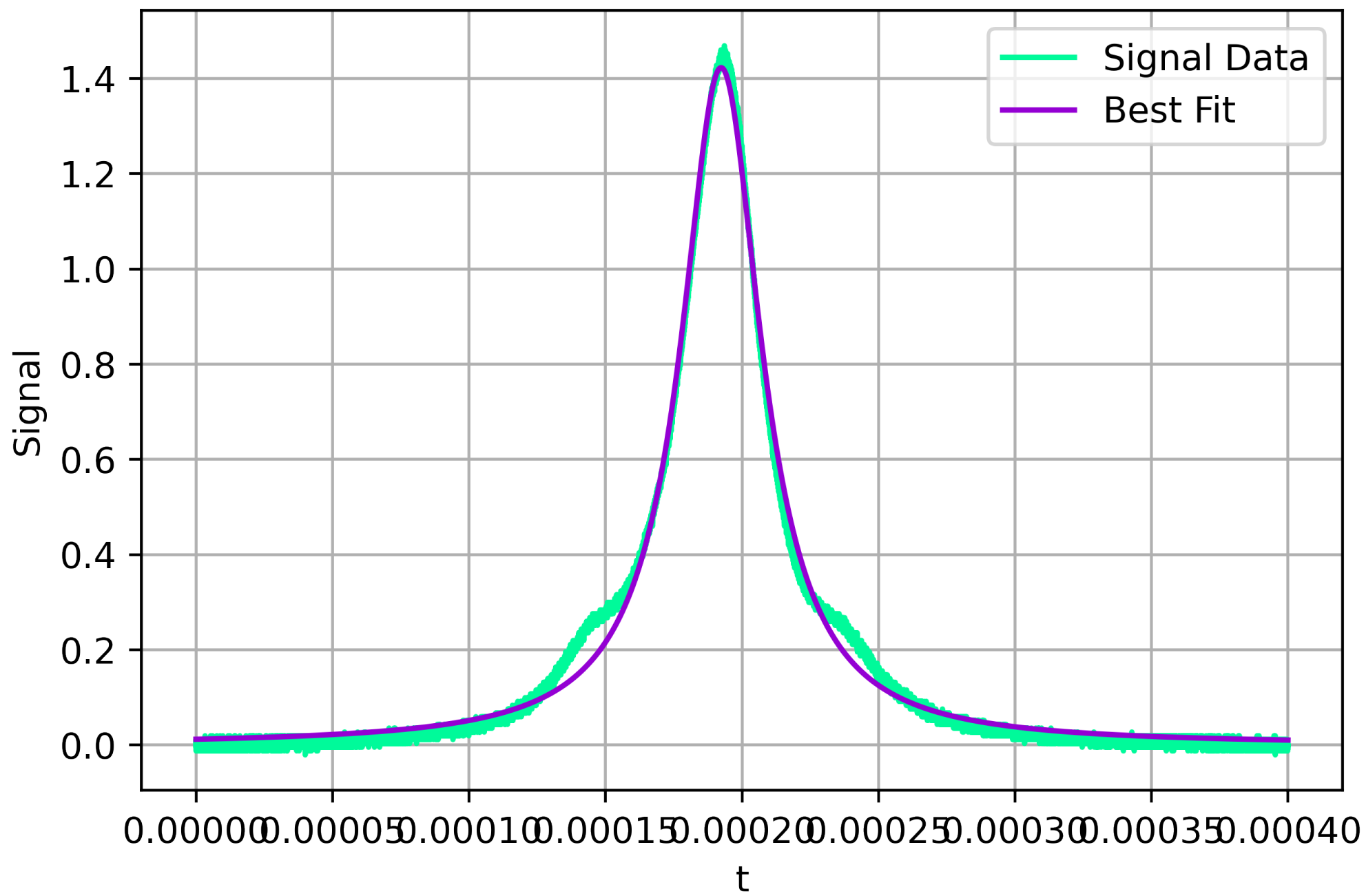
$$\nabla^2 \chi^2 \simeq 2A'^T N^{-1} A' \cdot \delta m = (A'^T N^{-1} A')^{-1} (A'^T N^{-1} r)$$

Note: We initially (for 1a)) assume $N = \mathbb{1}$ & thus get:

$$\delta m = [\nabla f^T \nabla f]^{-1} [\nabla f^T r] = [\nabla f^T \nabla f]^{-1} [\nabla f^T (d - f)]$$

For an initial guess of $a, t_0, \omega = [1.5, 2 \times 10^{-4}, 1 \times 10^{-4}]$ we get the following best fit parameters; the plot looks like:

$$a = 1.4228, \quad t_0 = 1.9236 \times 10^{-4}, \quad \omega = 1.7924 \times 10^{-5}$$



b) Our estimate for computing the error was derived from discussions w/ Guillaume

↳ The procedure for estimating N will be done via taking the standard deviation (at a point x_0) of a small, compact neighborhood of data point surrounding x_0

↳ Moreover we assume the standard error on each data point is $\approx 5 \times 10^{-3} \equiv \sigma$ (\approx the std of 25 points).

With this we can estimate the covariance matrix:

$$N^{-1} = \mathbb{1} / \sigma^2 \Rightarrow N = A^T N^{-1} A = \frac{1}{\sigma^2} \nabla f^T \nabla f$$

Using this, we get the parameter uncertainties:

$$\sigma_a = 8.4304 \times 10^{-5}, \quad \sigma_{t_0} = 1.0618 \times 10^{-9}, \quad \sigma_w = 1.5036 \times 10^{-4}$$

c) We redo part a) w/ numerical derivatives & get the following best-fit parameters:

$$a = 1.4228, \quad t_0 = 1.9236 \times 10^{-4}, \quad w = 1.7923 \times 10^{-5}$$

It is somewhat suspicious that they are the exact same values (on python they agree to a lot more digits), but it suffices to say they are not statistically different from the answers in a)

↳ The difference is less than σ we computed in b)

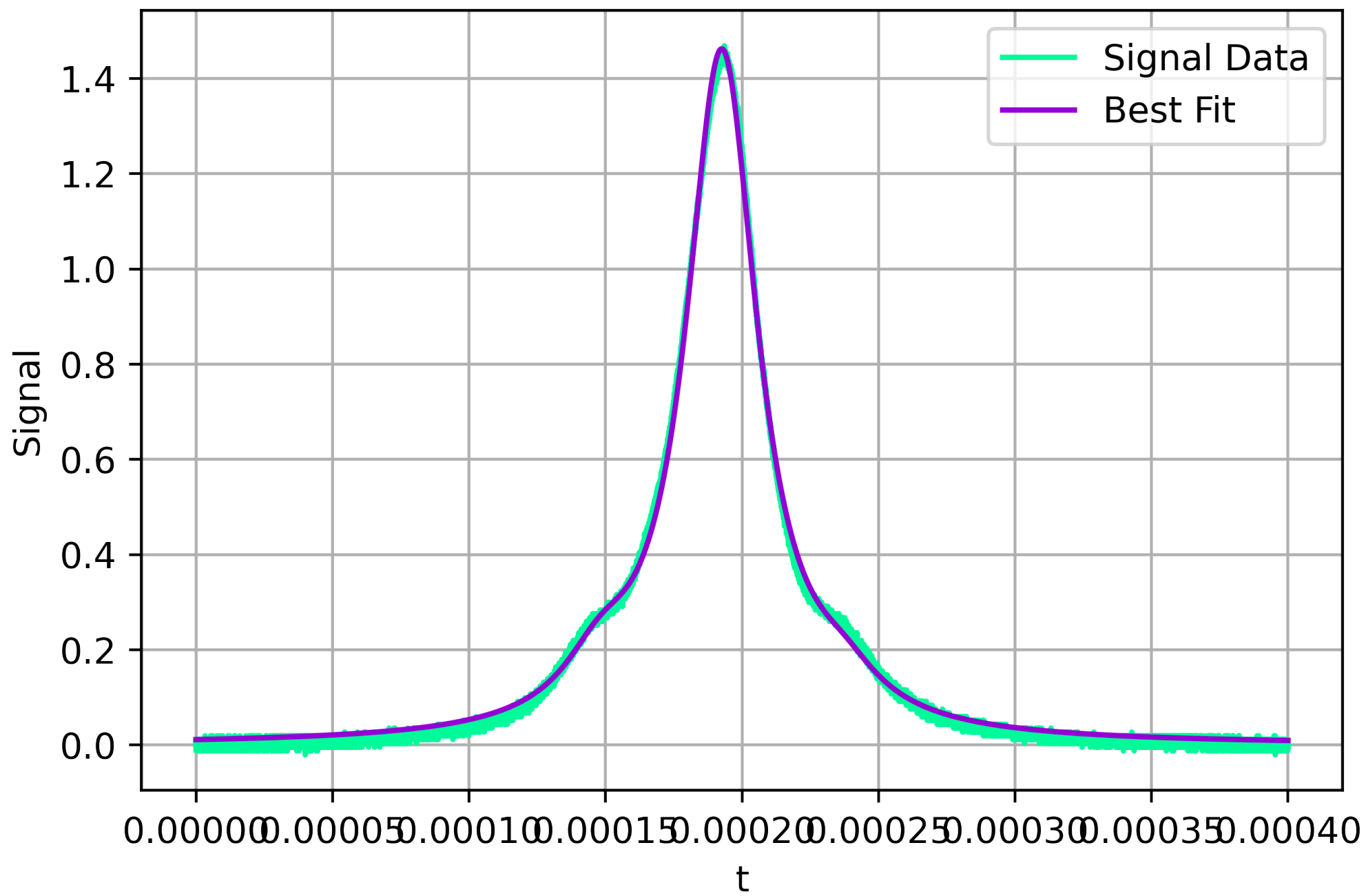
d) We repeat what is done in c) (numerical derivatives), (except now with:


$$f(t) = a[1 + (t - t_0)^2 / \omega^2]^{-1} + b[1 + (t - t_0 + \Delta t)^2 / \omega^2]^{-1} \\ + c[1 + (t - t_0 - \Delta t)^2 / \omega^2]^{-1}$$

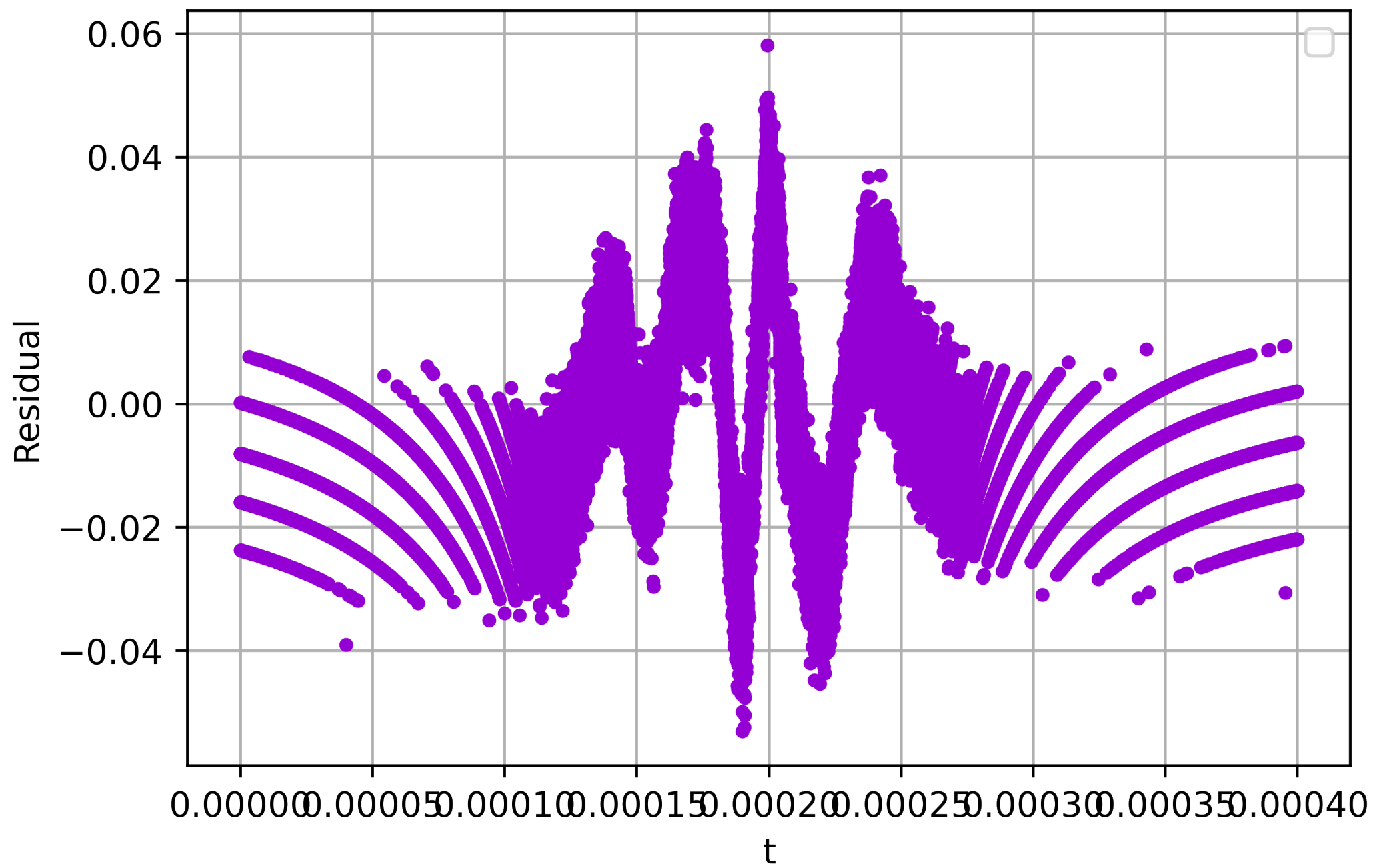
Following the procedure we've done in a-c), we have the following best fit parameters & their uncertainties:

$$a = 1.4430 \pm 9.1390 \times 10^{-5}, \quad b = 1.0391 \times 10^{-1} \pm 8.7167 \times 10^{-5} \\ c = 6.4733 \times 10^{-2} \pm 8.6391 \times 10^{-5}, \quad t_0, t_0 + \Delta t = 1.9258 \times 10^{-4} \pm 1.0820 \times 10^{-9} \\ \omega = 1.6065 \times 10^{-5} \pm 1.9378 \times 10^{-9}, \quad \Delta t = 4.457 \times 10^{-5} \pm 1.3044 \times 10^{-8}$$

Moreover, the prediction fits the data better, as seen in:

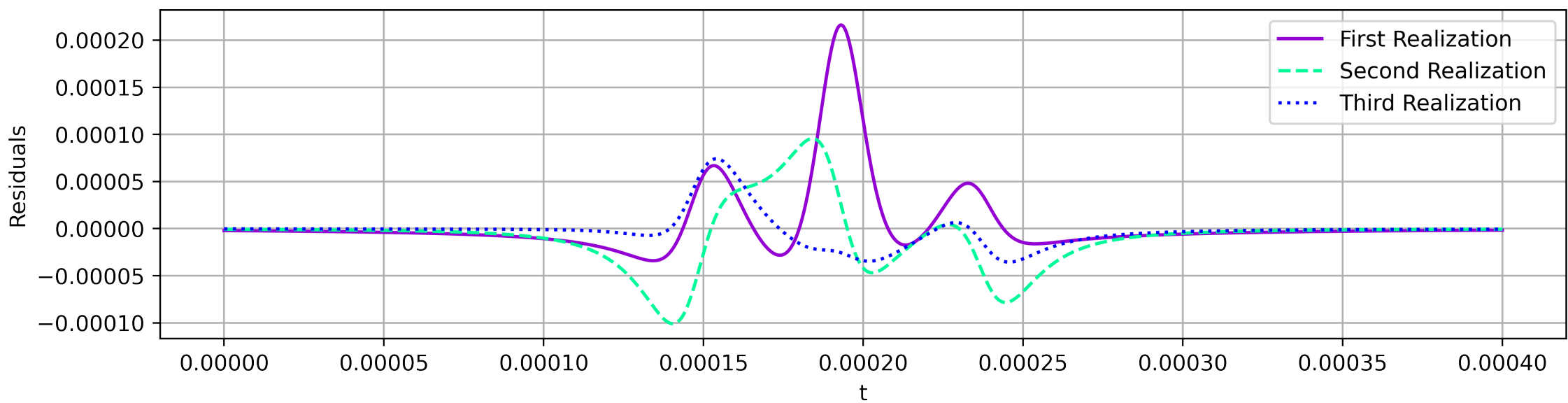
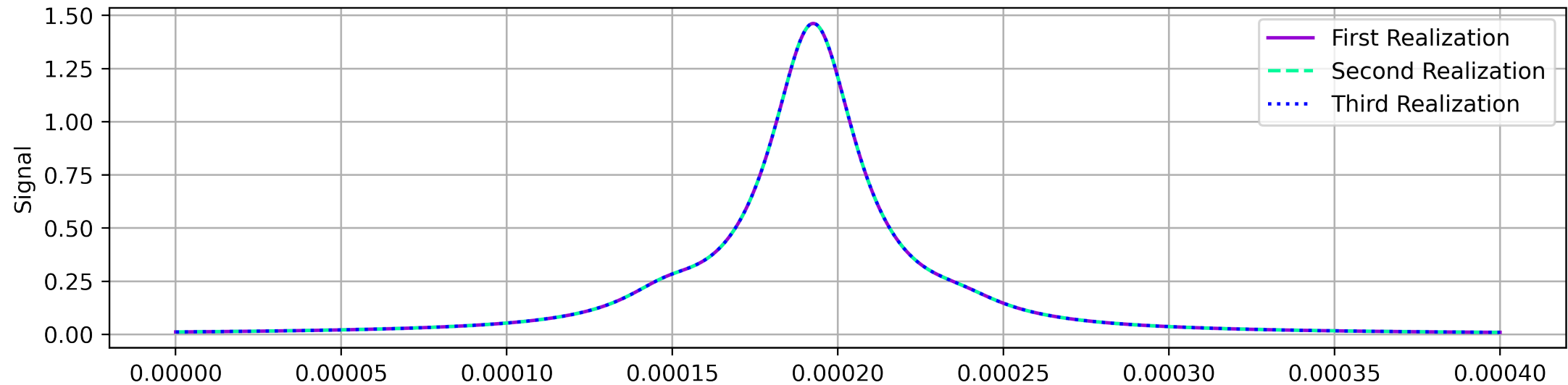


e) Two residuals (data - fit) have the following plot: 



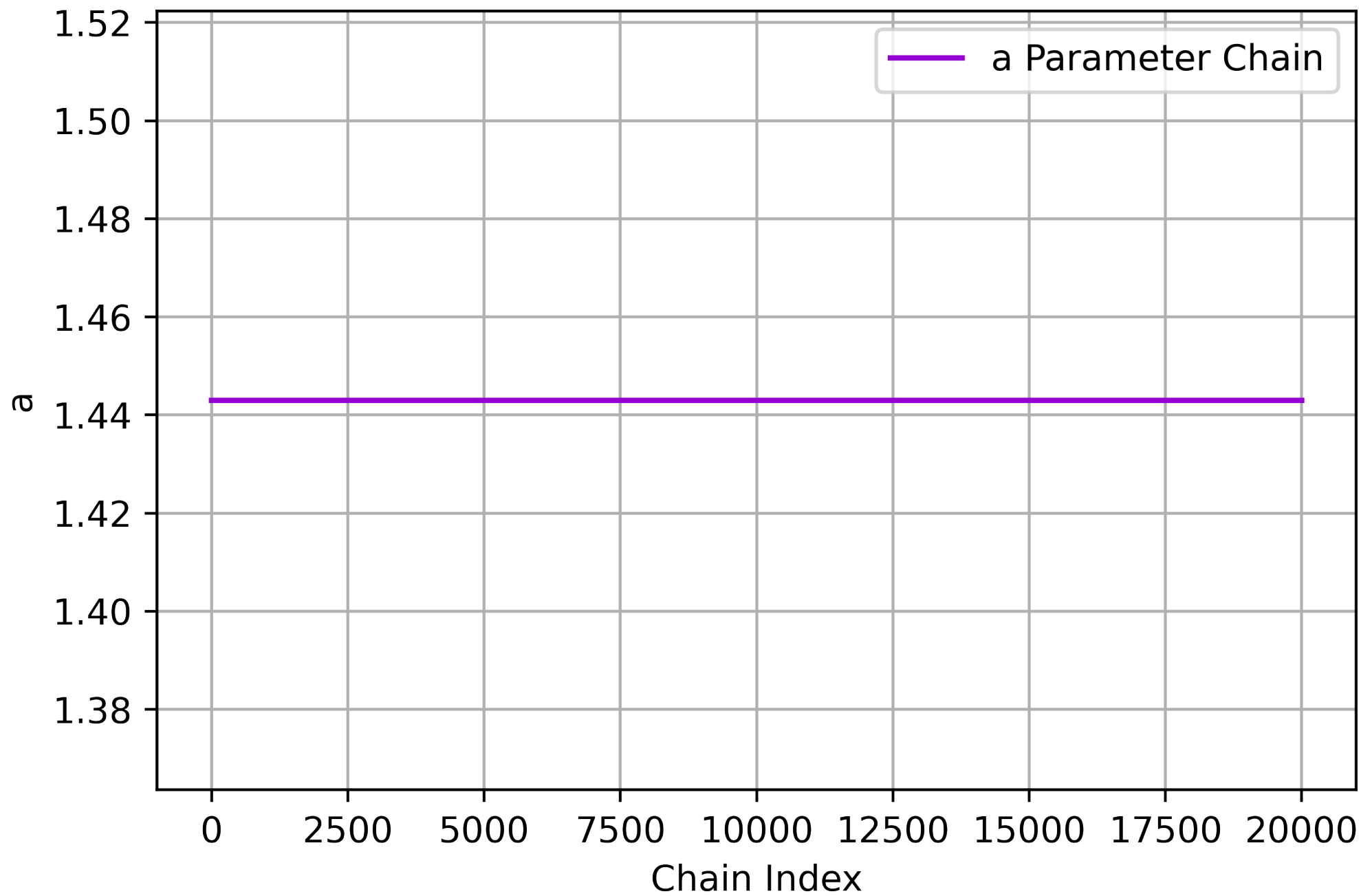
The residuals clearly contain a structure & are thus correlated. This tells us the error term we got by assuming independence is not a full description of the data.

f) We consider three different realizations for randomly ϵ_i (a perturbed parameter) & it has the following plot (of their corresponding residuals):



Computing the typical difference in χ^2 gives us 8.5200×10^{-5} which is reasonable as it is much less than 1 & we only perturbed the parameters by a small, random amount.

g) After computing the MCMC algorithm for 20,000 steps, we have the chain plot for the 'c' parameter:



As we can see from above, the chain converges as there is no structure on the scale of the full chain*.

↳ Moreover, the best-fit parameters & their uncertainty is:

$$\begin{aligned} \bullet a &= 1.4430 \pm 6.6613 \times 10^{-16} & \bullet b &= 0.1039 \pm 4.1633 \times 10^{-17} \\ \bullet c &= 0.0647 \pm 1.3878 \times 10^{-17} & \bullet t_0 &= 0.0002 \pm 5.4210 \times 10^{-20} \\ \bullet w &= 1.6065 \times 10^{-5} \pm 1.8061 \times 10^{-16} & \bullet dt &= 4.4567 \times 10^{-5} \pm 1.2351 \times 10^{-17} \end{aligned}$$

As we can see, the error bars are much, much smaller. This is a very good result but it makes me think there is some error w/ my code as my peers have much higher errors (I have not yet found the bug).

*The fact that it is constant raises some red flags but I cannot find the bug in my code that would produce such a result.

h) Using our best fit parameters, we can compute k_c (7)
width μ of the resonance cavity:

$$\mu = \Delta t / \omega \cdot 96 \text{ Hz} = (24.9674 \pm 0.0280) \text{ GHz}$$