

PHYS 232

March 26, 2025

Due dates:

- Assignment #6 : Wed. Apr. 2 @ 09:00
 - Experiment #5 lab notebooks : Fri. Apr. 4 @ 09:00
 - Bring lab notebooks to lecture
 - Formal Report : Mon. Apr. 7 @ 14:00
 - Bring printed and stapled report to SCI 241.
-

Last Time:

Fit data to :

$$y(x) = a_1 f_1(x) + a_2 f_2(x) + \dots + a_m f_m(x)$$

Meas. $(x_i, y_i \pm \sigma_i)$ for $i = 1 \dots N$

Want to determine :

$$a_1 \pm \sigma_{a_1}, a_2 \pm \sigma_{a_2}, \dots, a_m \pm \sigma_{a_m}$$

We found the a_k values from

$$\underline{a} = \underline{\Sigma} \underline{\beta}$$

Diagonal elements of $\underline{\Sigma}$ give the square of the uncertainty in the best-fit parameters

$$a_1 \pm \sqrt{\Sigma_{11}}$$

$$a_2 \pm \sqrt{\Sigma_{22}}$$

$$\vdots$$

$$a_m \pm \sqrt{\Sigma_{mm}}$$

Today: Nonlinear Fits

Examine the case of a non-linear fit,

i.e. $y(x)$ is not of the form $\sum_{k=1}^n a_k f_k(x)$

$$\text{Eg. } y(x) = a_1 \sin(a_2 x)$$

Compare & contrast linear & non-linear fits.

linear case

$$y = a_1 + a_2 x$$

of the form $\sum_k a_k f_k(x)$

$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - y(x_i)}{\sigma_i} \right)^2$$

$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - a_1 - a_2 x_i}{\sigma_i} \right)^2$$

↑
minimize w.r.t. a_1 & a_2

$$\frac{\partial \chi^2}{\partial a_1} = 2 \sum \left(\frac{y_i - a_1 - a_2 x_i}{\sigma_i} \right) \left(-\frac{1}{\sigma_i} \right) = 0$$

$$\therefore \sum \frac{y_i}{\sigma_i^2} = \sum \frac{a_1 + a_2 x_i}{\sigma_i^2} \quad (1)$$

↑
no a_1 or a_2

$$\frac{\partial \chi^2}{\partial a_2} = 2 \sum \left(\frac{y_i - a_1 - a_2 x_i}{\sigma_i} \right) \left(-\frac{x_i}{\sigma_i} \right) = 0$$

$$\sum \frac{x_i y_i}{\sigma_i^2} = \sum \frac{x_i}{\sigma_i^2} (a_1 + a_2 x_i) \quad (2)$$

↑
no a_1 or a_2

non-linear case

$$y = a_1 \sin(a_2 x)$$

$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - y(x_i)}{\sigma_i} \right)^2$$

$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - a_1 \sin(a_2 x_i)}{\sigma_i} \right)^2$$

↑
minimize w.r.t. a_1 & a_2

$$\frac{\partial \chi^2}{\partial a_1} = 2 \sum \frac{y_i - a_1 \sin(a_2 x_i)}{\sigma_i}$$

$$\cdot \left(-\frac{\sin(a_2 x_i)}{\sigma_i} \right) = 0$$

$$\therefore \sum \frac{y_i \sin(a_2 x_i)}{\sigma_i^2} = \sum \frac{a_1 \sin^2(a_2 x_i)}{\sigma_i^2} \quad (1')$$

involves a_2

$$\frac{\partial \chi^2}{\partial a_2} = 2 \sum \frac{y_i - a_1 \sin(a_2 x_i)}{\sigma_i}$$

$$\cdot \left(-\frac{x_i a_1}{\sigma_i} \cos(a_2 x_i) \right) = 0$$

$$\sum \frac{x_i y_i}{\sigma_i^2} a_1 \cos(a_2 x_i) \quad \text{involves } a_1 \text{ \& } a_2 \quad (z')$$

$$= \sum \frac{x_i a_1^2}{\sigma_i^2} \cos(a_2 x_i) \sin(a_2 x_i)$$

For the linear fit, can express ① & ② as a matrix eq'n w/ a_k parameters in a single column matrix

$$\underbrace{\begin{pmatrix} \sum \frac{y_i}{\sigma_i^2} \\ \sum \frac{x_i y_i}{\sigma_i^2} \end{pmatrix}}_{\underline{\beta}} = \underbrace{\begin{pmatrix} \sum \frac{1}{\sigma_i^2} & \sum \frac{x_i}{\sigma_i^2} \\ \sum \frac{x_i}{\sigma_i^2} & \sum \frac{x_i^2}{\sigma_i^2} \end{pmatrix}}_{\underline{\alpha}} \underbrace{\begin{pmatrix} a_1 \\ a_2 \end{pmatrix}}_{\underline{a}}$$

$$\therefore \underline{a} = \underline{\alpha}^{-1} \underline{\beta}$$

For non-linear fit, the a_k 's in (1') & (2') are on both sides of the eq'ns & a_2 is inside a trig. fn. Cannot write down an equiv. matrix eq'n for this case.

$\Rightarrow$ need a new fitting method.

One method for non-linear fits:

Objective is still to minimize

$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - y(x_i)}{\sigma_i} \right)^2$$

For example, want to minimize:

$$\chi^2 = \sum_{i=1}^N \left(\frac{y_i - a_1 \sin(a_2 x_i)}{\sigma_i} \right)^2$$

Strategy:

1. Pick a range of possible values for a_1 & a_2 .

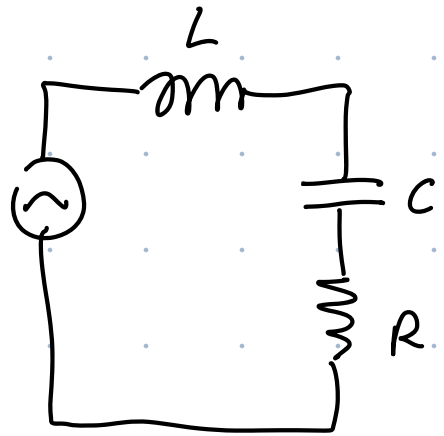
$$\begin{aligned} a_{1,\min} &\leq a_1 \leq a_{1,\max} \\ a_{2,\min} &\leq a_2 \leq a_{2,\max} \end{aligned} \quad \left. \vphantom{\begin{aligned} a_{1,\min} &\leq a_1 \leq a_{1,\max} \\ a_{2,\min} &\leq a_2 \leq a_{2,\max} \end{aligned}} \right\} \begin{array}{l} \text{have to know} \\ \text{something about} \\ \text{physical system to} \\ \text{select reasonable} \\ \text{ranges.} \end{array}$$

Aside

Eg: LRC Circuit.

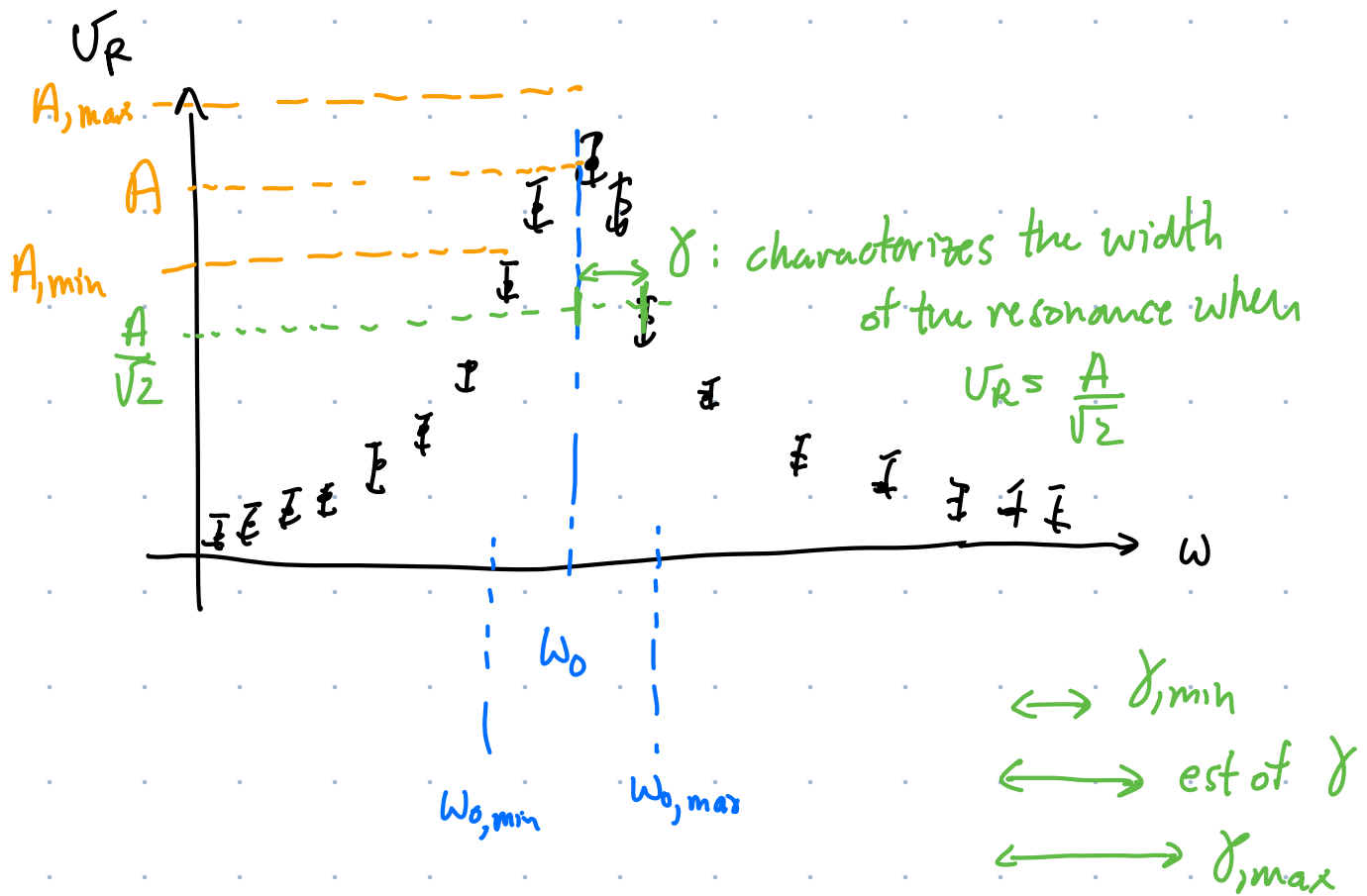
Voltage across R in

$$V_R = \frac{A}{\sqrt{1 + \frac{\omega^2}{\gamma^2} \left(1 - \frac{\omega_0^2}{\omega^2}\right)^2}}$$



meas. V_R vs $\omega \rightarrow$ fit parameters are: A, γ, ω_0

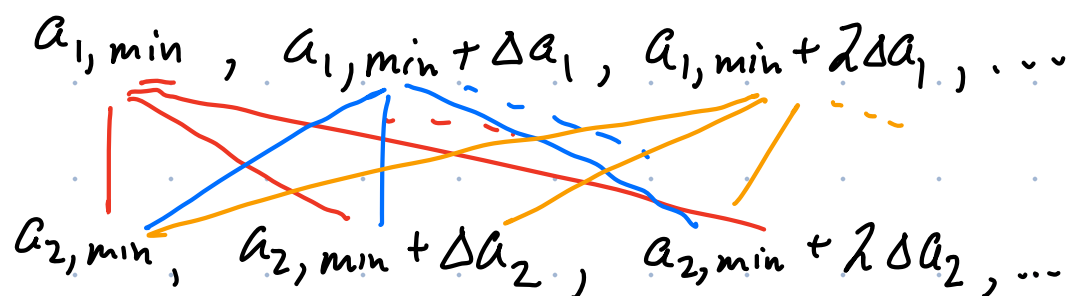
Need to use the meas. data to est. reasonable ranges for the fit parameters.



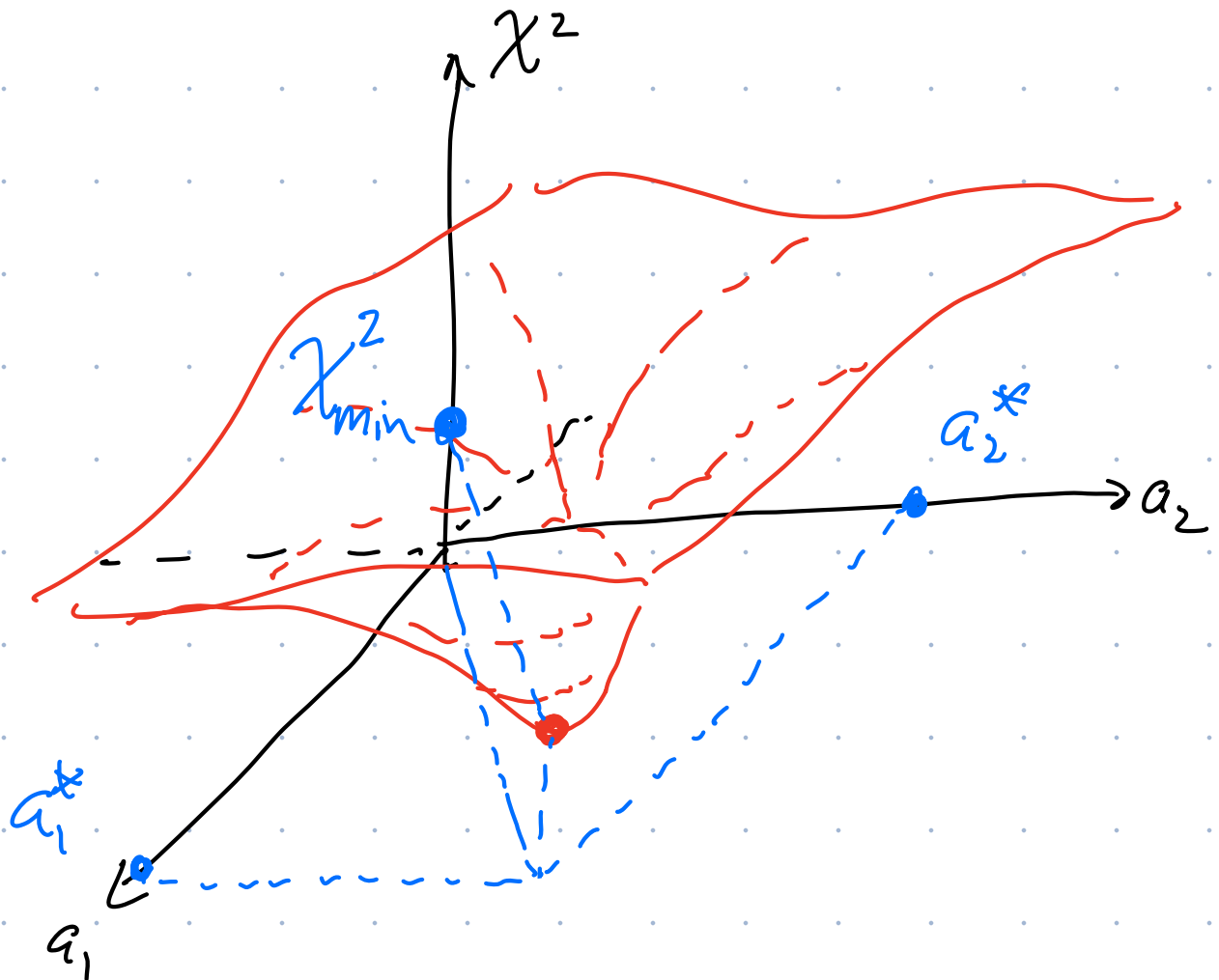
2. Pick a step size Δa_1 & Δa_2 for each parameter.

- small Δa_1 , Δa_2 good for high resolution, but takes longer to calc. all χ^2 values.

3. Evaluate χ^2 at all possible combos of (a_1, a_2) .



Best-fit parameters are the values of a_1 & a_2 that give the minimum for χ^2 .



Plot the χ^2 surface vs a_1 & a_2 .

If (a_1^*, a_2^*) give the minimum value of χ^2 , then they are our estimates of the best-fit parameters.

Another way to visualize the same thing:

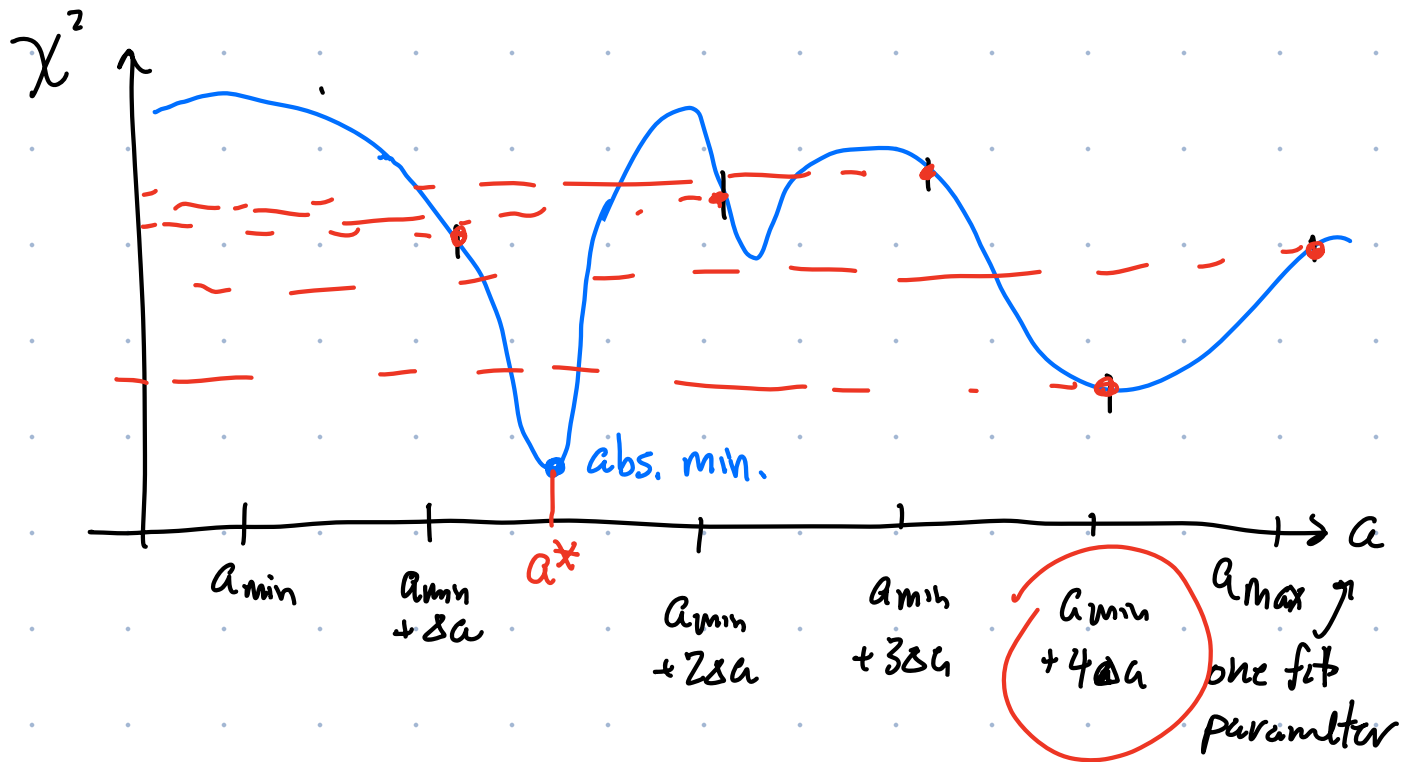
a_1^*

a_1, a_2	$a_{1, \min}$	$a_{1, \min} + \Delta a_1$	$a_{1, \min} + 2\Delta a_1$	...	$a_{1, \max}$
$a_{2, \min}$	36	42	41		68
$a_{2, \min} + \Delta a_2$	28	36	34		90
$a_{2, \min} + 2\Delta a_2$	22	18	24		36
$\vdots$					
$a_{2, \max}$	50	49	62		112

$\chi^2_{\min}$

populate table w/ χ^2 values.

If the non-linear fcn is complicated & changing rapidly, it is possible to miss the absolute global minimum.



In This case, would identify $a_{\min} + 4\delta a$ as our best-fit parameter, missing a^* b/c of our coarse search.