

- ✓ - Assignment #5 due Fri, Mar. 22
- ✓ - Sign up for Experiment #5 by the end of the day.

Last Time: Found best-fit parameters for weighted linear fits.

Intercept: $a \pm \sigma_a$

Slope: $b \pm \sigma_b$

$$a = \frac{1}{\Delta} \left(\sum \frac{x_i^2}{\sigma_i^2} \sum \frac{y_i}{\sigma_i^2} - \sum \frac{x_i}{\sigma_i^2} \sum \frac{x_i y_i}{\sigma_i^2} \right)$$

$$b = \frac{1}{\Delta} \left(\sum \frac{1}{\sigma_i^2} \sum \frac{x_i y_i}{\sigma_i^2} - \sum \frac{x_i}{\sigma_i^2} \sum \frac{y_i}{\sigma_i^2} \right)$$

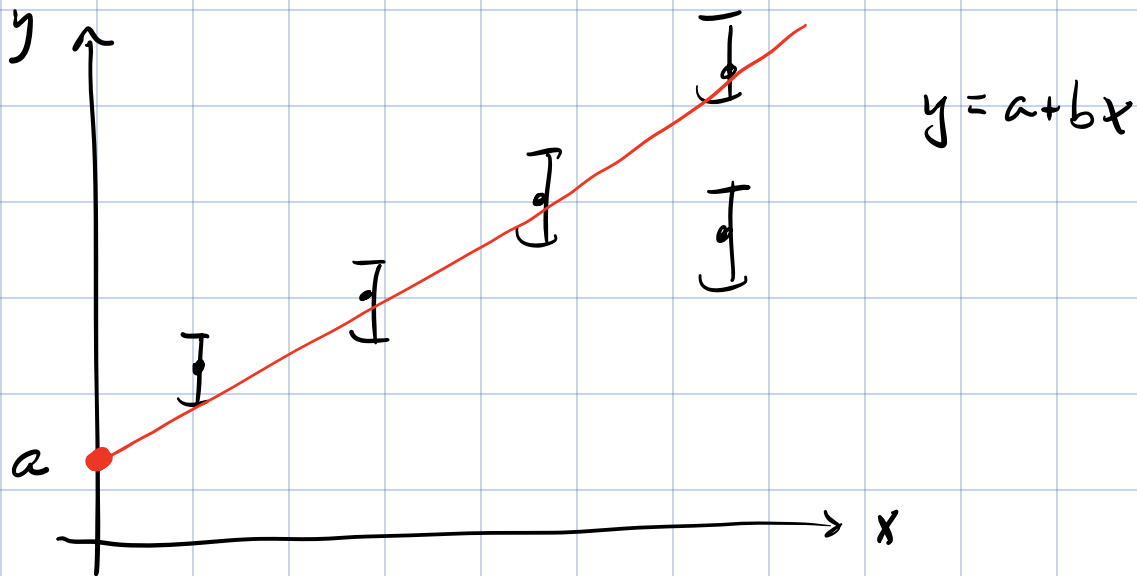
where:

$$\Delta = \sum \frac{1}{\sigma_i^2} \sum \frac{x_i^2}{\sigma_i^2} - \left(\sum \frac{x_i}{\sigma_i^2} \right)^2$$

$$\therefore \sigma_a^2 = \frac{1}{\Delta} \sum \frac{x_i^2}{\sigma_i^2}$$

$$\therefore \sigma_b^2 = \frac{1}{\Delta} \sum \frac{1}{\sigma_i^2}$$

Today: Linearizing Experimental Data.

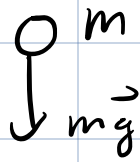


Can determine best-fit values for $a \pm \sigma_a$
& $b \pm \sigma_b$ based on meas. of

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

Easy to use when have obvious linear dependence.

Eg. Object in free fall

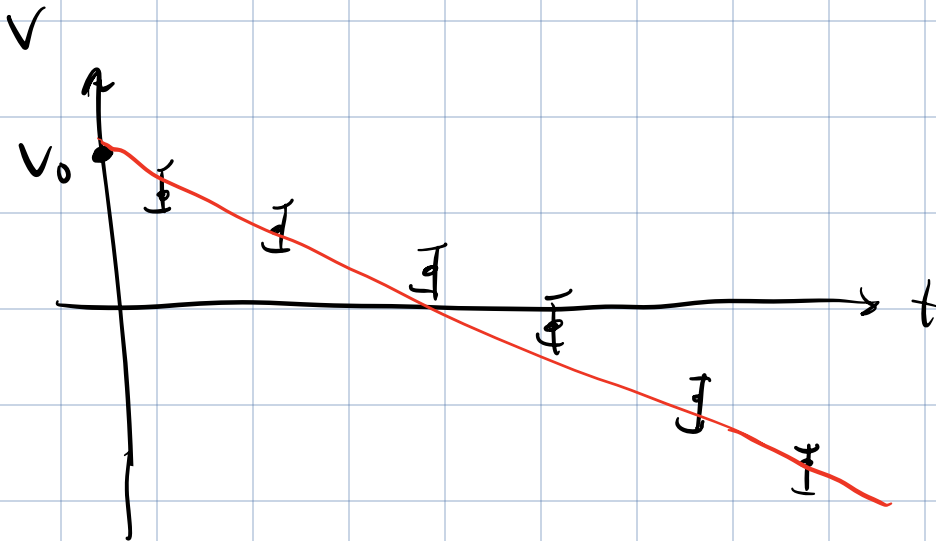


$$v = v_0 - gt$$

$$y = a + bx$$

meas. $v \pm \sigma_v$ as a fun of t .

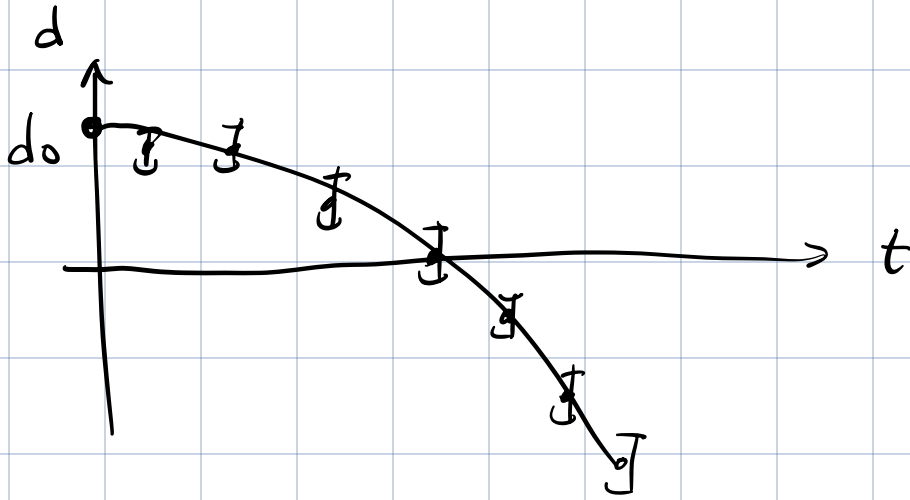
$$\begin{array}{l} \text{set } y = v \\ x = t \end{array} \Rightarrow \begin{array}{l} a = v_0 \\ b = -g \end{array}$$



Sometimes a pair of variables may not be strictly linearly related, but often can still use a linear fit to interpret the data.

Eg. Object in free fall (start rest)

$$d = d_0 - \frac{1}{2} g t^2$$



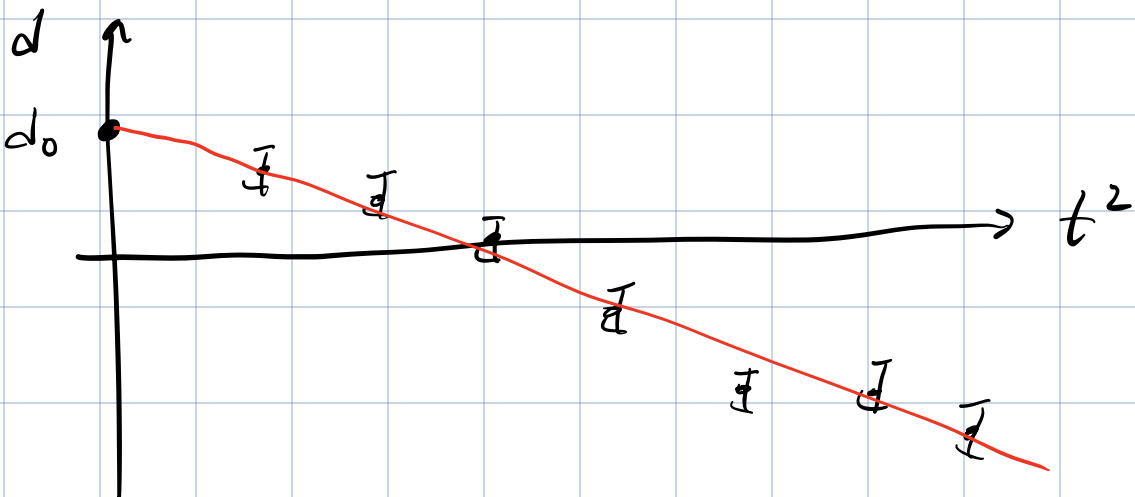
Instead, can linearize data by plotting

$$y = d \quad \text{vs} \quad x = t^2$$

$$d = d_0 - \frac{1}{2} g t^2$$

\uparrow y $= a + b x$ \uparrow

In this case a plot of d vs t^2 has intercept $a = d_0$ & slope $b = -g/2$.



Another Example. In thermal waves, the amplitude of fund. Fourier component is

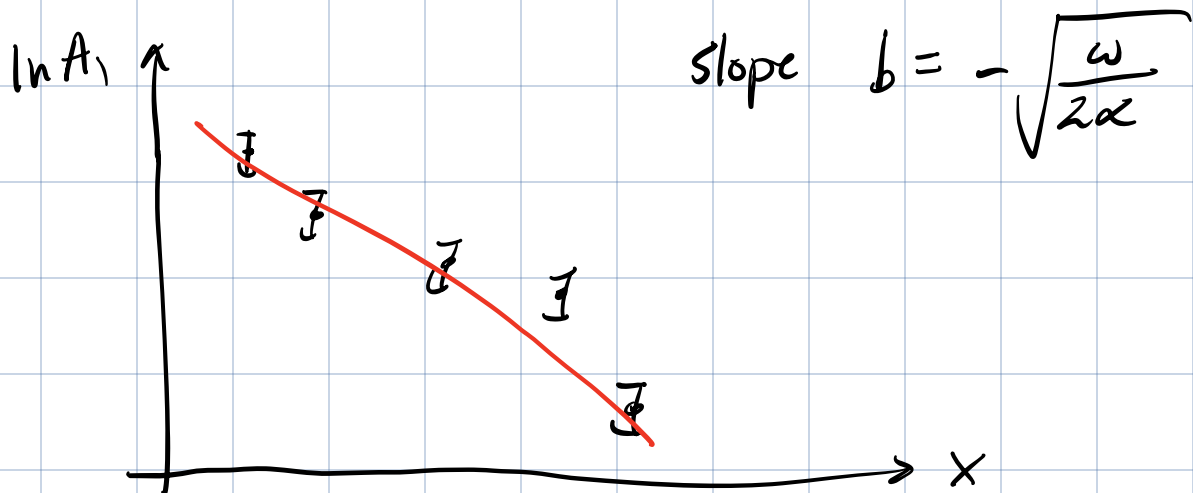
$$A_1 = \left(\frac{4T_0}{\pi} \right) \exp \left(-\sqrt{\frac{\omega}{2\alpha}} x \right)$$

To linearize, take \ln of both sides:

$$\ln A_1 = \ln \left(\frac{4T_0}{\pi} \right) - \sqrt{\frac{\omega}{2\alpha}} x$$

$y = a + b x$

Plot $\ln A_1$ vs x , get intercept $a = \ln \left(\frac{4T_0}{\pi} \right)$



In thermal waves exp't use the slope b to find the thermal diffusivity α of copper

$$b = -\sqrt{\frac{\omega}{2\alpha}} \Rightarrow b^2 = \frac{\omega}{2\alpha}$$

$$\Rightarrow \alpha = \frac{\omega}{2b^2}$$

$$\sigma_\alpha^2 = \left(\frac{\partial \alpha}{\partial b} \sigma_b \right)^2 + \left(\frac{\partial \alpha}{\partial \omega} \sigma_\omega \right)^2$$

$$= \left(-\frac{2\omega}{2b^3} \sigma_b \right)^2 + \left(\frac{1}{2b^2} \sigma_\omega \right)^2$$

$$\therefore \sigma_\alpha = \sqrt{\left(\frac{\omega \sigma_b}{b^3} \right)^2 + \left(\frac{\sigma_\omega}{2b^2} \right)^2}$$

Another example (used in current research)

Low-Temperature thermal conductivity of a conductor is typically of the form:

$$K = \underbrace{C_1 T}_{\text{electron contribution}} + \underbrace{C_3 T^3}_{\text{lattice/phonon contribution}}$$

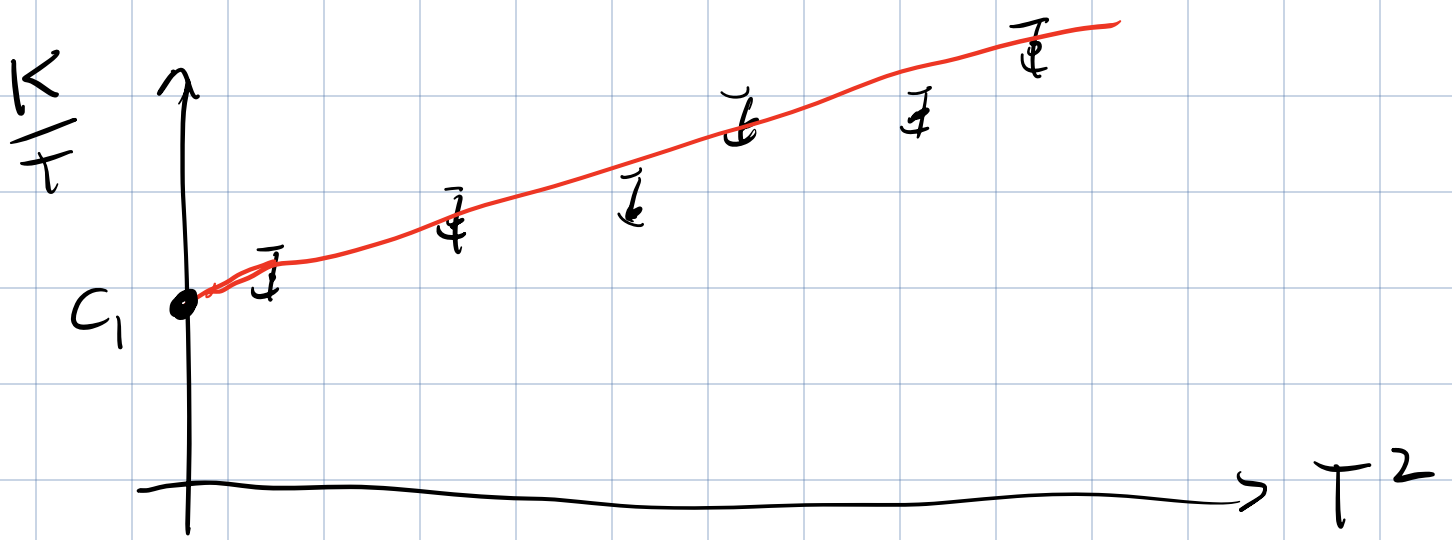
Meas. K as a fun of T .

Does not appear to be easily linearized at first glance.

$$\text{Try } \frac{K}{T}. \quad \underbrace{y}_{\frac{K}{T}} = a + b \underbrace{x}_{T^2}$$

Plot $\frac{K}{T}$ vs T^2

\Rightarrow intercept of $a = C_1$
slope of $b = C_3$



Another Example. Sometimes a pair of variables cannot be strictly linearized.

In Blackbody radiation

$$I_{\lambda} = \frac{2c^2h}{\lambda^5} \left[\frac{1}{e^{hc/k_B T \lambda} - 1} \right]$$

meas. I_{λ} vs T which is not linear.

However, under certain conditions, can make valid approximations that allow for the data to be linearized.

In this case, if $\frac{hc}{k_B T \lambda} \gg 1$

then $e^{hc/k_B T \lambda} - 1 \approx e^{hc/\lambda k_B T}$

$$I_\lambda \approx \frac{2c^2 h}{\lambda^5} e^{-hc/\lambda k_B T}$$

Now it makes sense to take \ln of both sides

$$\underbrace{\ln I_\lambda}_{y} \approx \ln\left(\frac{2c^2 h}{\lambda^5}\right) - \left(\frac{hc}{\lambda k_B}\right) \underbrace{\frac{1}{T}}_x$$

$y = a + b x$

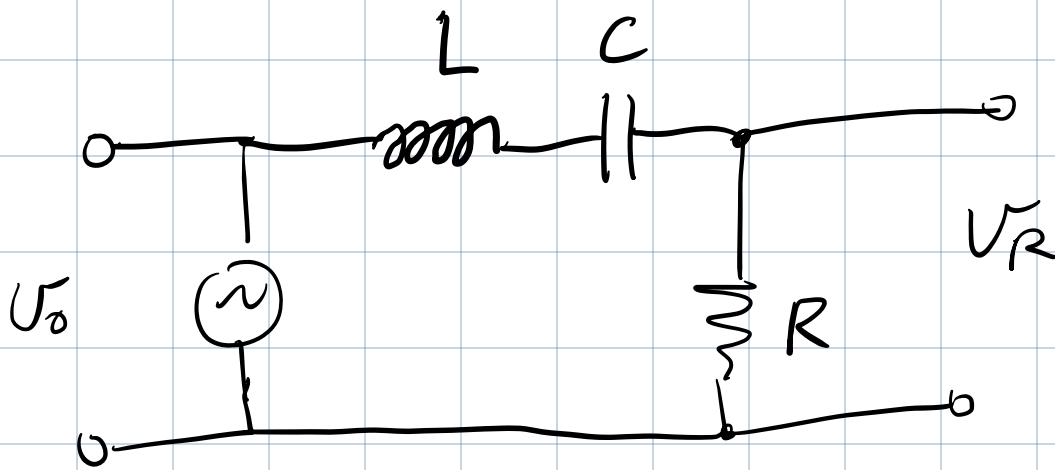
Plot of $\ln I_\lambda$ vs $\frac{1}{T}$ has intercept $a = \ln\left(\frac{2c^2 h}{\lambda^5}\right)$

$$b = -\frac{hc}{\lambda k_B}$$

$$\text{Find } h = \frac{-\lambda k_B b}{c}$$

Sometimes it's just not possible to linearize a data set.

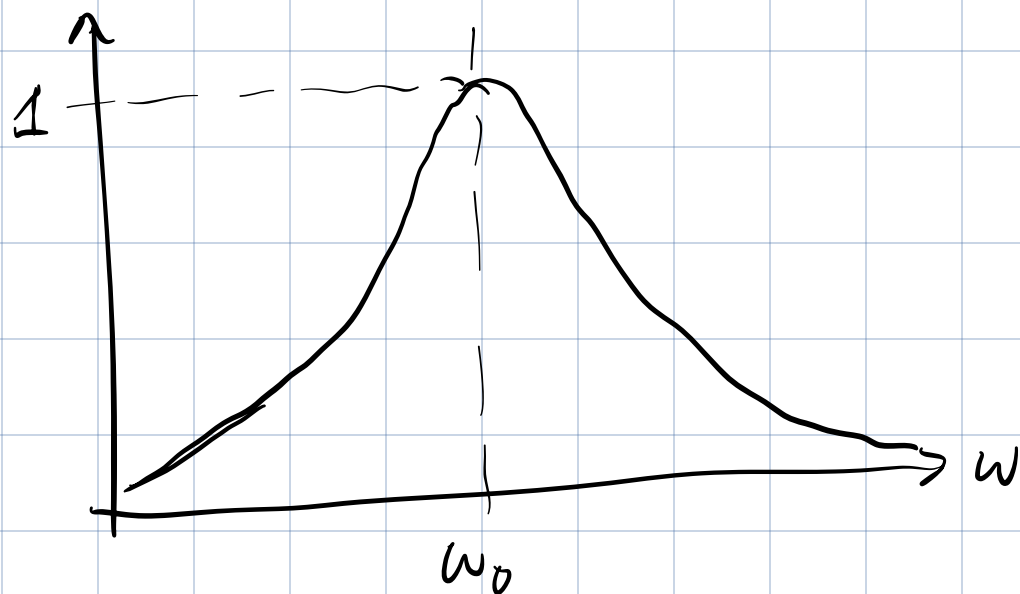
Eg. LRC resonance



meas. $\left| \frac{U_R}{U_0} \right|$ vs ω ← freq.

$$\left| \frac{U_R}{U_0} \right| = \frac{1}{\sqrt{1 + \left(\omega \frac{L}{R} - \frac{1}{\omega RC} \right)^2}}$$

$$\left| \frac{V_R}{V_0} \right|$$



Need different fitting techniques to extract desired parameters from non-linear datasets.