

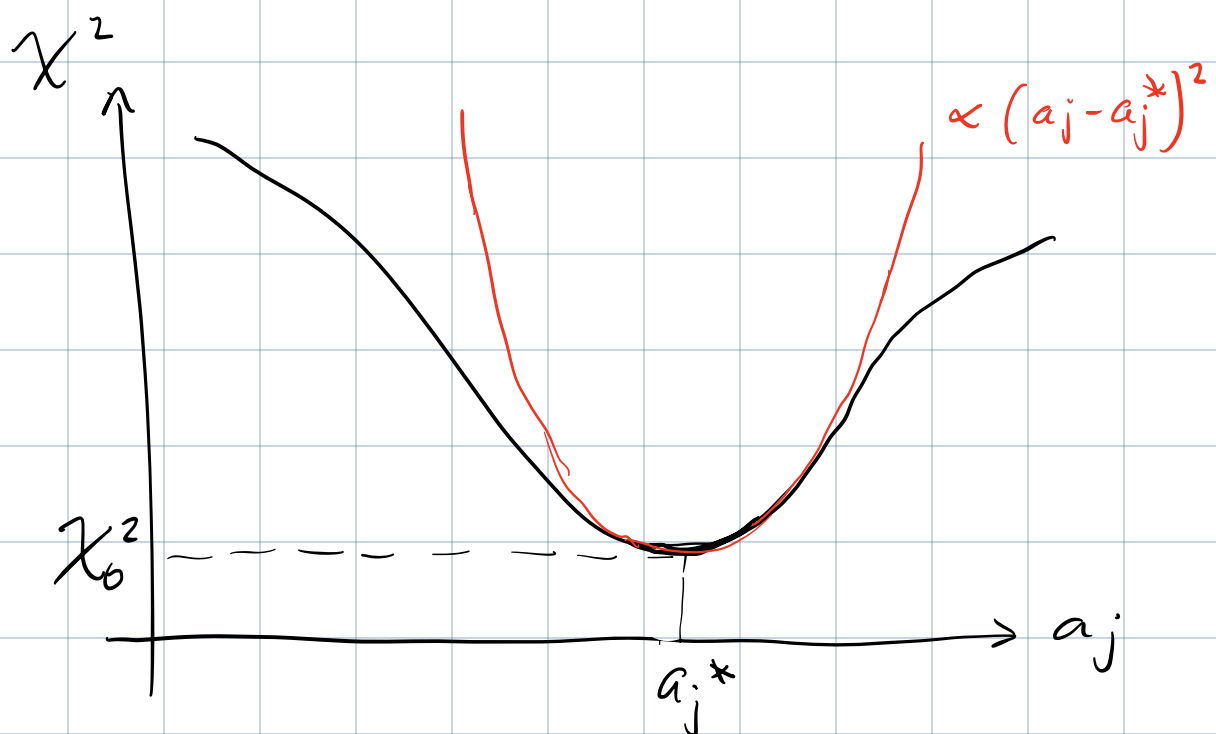
PHYS 232

April 3, 2024

- ✓ Assignment #6 due April 5 @ 11am
- ✓ Formal Report due Apr. 10 @ 11am

Last Time:

Near its minimum, χ^2 varies quadratically w.r.t. changes in the a_j parameter values



$$\chi^2 \approx \chi_0^2 + B (a_j - a_j^*)^2$$

By finding χ^2 at 3 pts. near the minimum, can find:

$$B = \frac{\chi_1^2 - 2\chi_2^2 + \chi_3^2}{2(\Delta a)^2}$$

The uncertainty in parameter a_j is given by

$$\sigma_j = \frac{1}{\sqrt{B_j}}$$

\therefore best-fit value for a_j is:

$$a_j = a_j^* \pm \frac{1}{\sqrt{B_j}}$$

Today Develop some intuition about meaning of χ^2 .

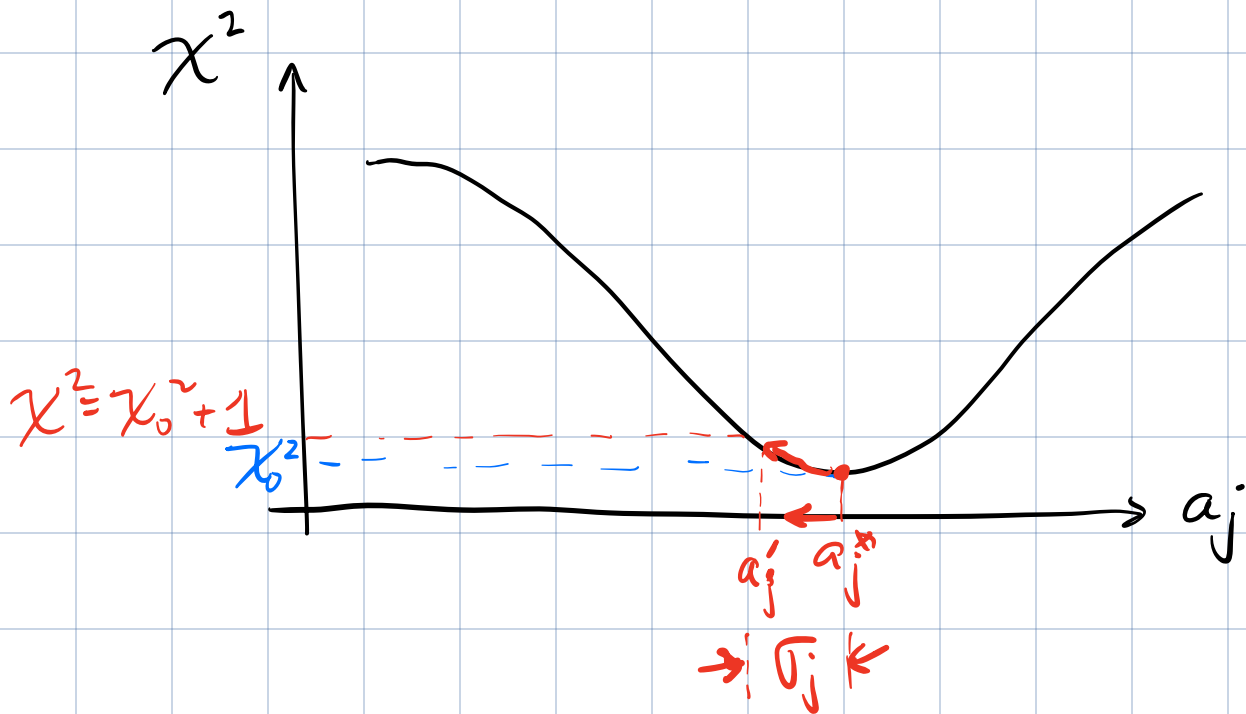
$$\text{if } \sigma_j = \frac{1}{\sqrt{B}} \Rightarrow B = \frac{1}{\sigma_j^2}$$

Notice if:

$$\chi^2 = \frac{(a_j - a_j^*)^2}{\sigma_j^2} + \chi_0^2$$

near the minimum of χ^2

Then, know that χ^2 will increase as we move a_j away from a_j^* .



Imagine moving along a_j axis starting at a_j^* & ending once χ^2 has increased from χ_0^2 to $\chi_0^2 + 1$.

$$\chi^2 = \frac{(a_j - a_j^*)^2}{\sigma_j^2} + \chi_0^2$$

\Downarrow

take $a_j \rightarrow a_j'$ s.t. $\chi^2 = \chi_0^2 + 1$

$$\underbrace{\cancel{\chi_0^2} + 1}_{\chi^2} = \frac{(a_j' - a_j^*)^2}{\sigma_j^2} + \cancel{\chi_0^2}$$

$$1 = \frac{(a_j' - a_j^*)^2}{\sigma_j^2}$$

$$\therefore \sigma_j = |a_j' - a_j^*|$$

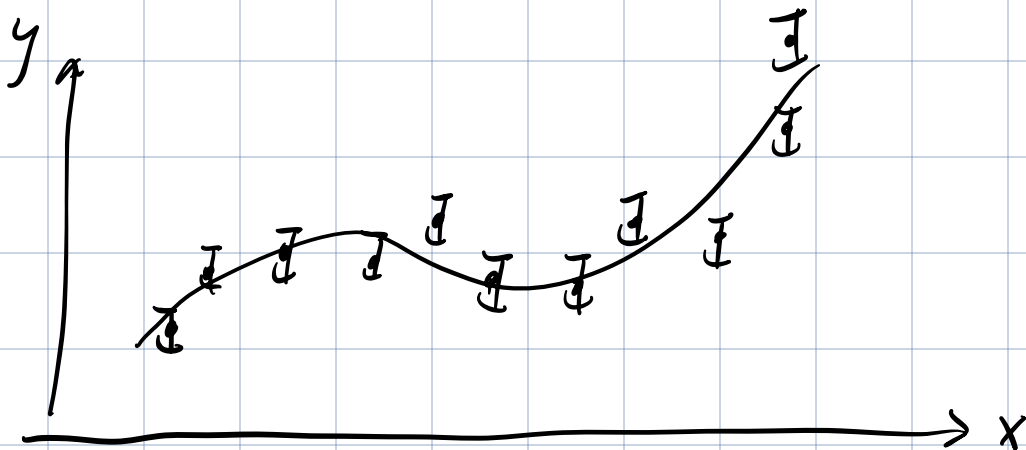
If we move away from the minimum of χ^2 along the a_j axis s.t. $\chi^2 \rightarrow \chi_0^2 + 1$, then the distance moved $|a_j' - a_j^*|$ is an estimate of the uncertainty in a_j^* .

What should we expect for a value of χ^2 after going a "good" fit?

Definition of χ^2 :

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

assume we've measured $(x_i, y_i \pm \sigma_i)$
for $i=1..N$.



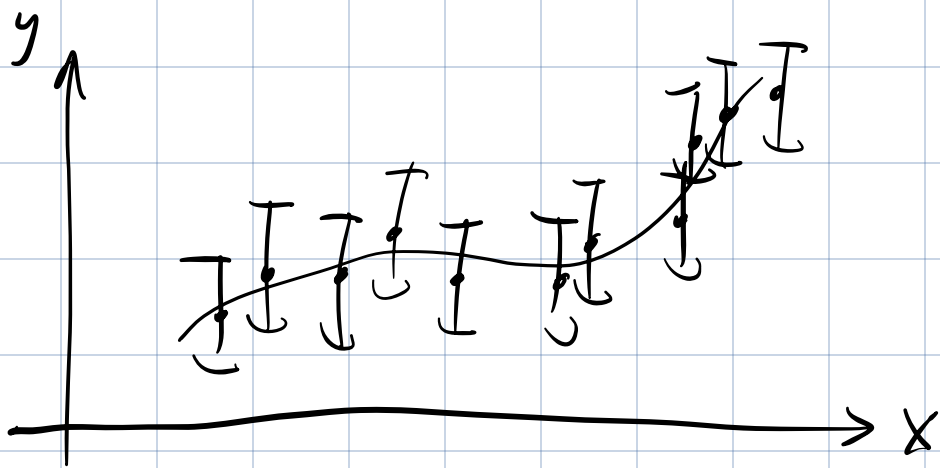
On average, we expect the deviation
 $y_i - y(x_i)$ to approximately

equal to $\pm \sigma_i$ if we've made reasonable
estimates of the uncertainties.

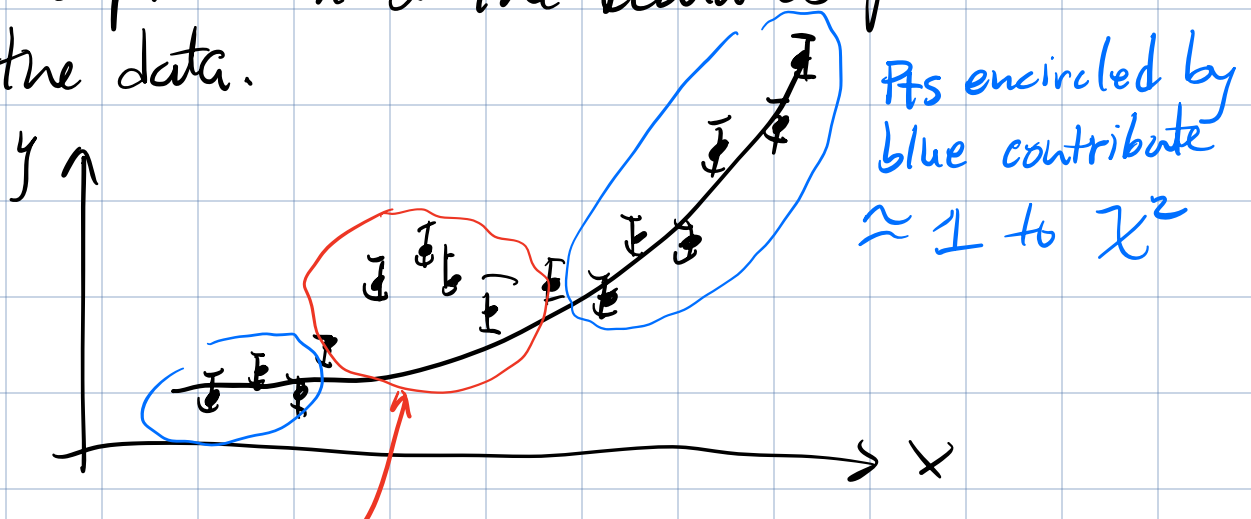
$$\therefore \chi^2 \approx \sum_{i=1}^N \left(\frac{\pm \cancel{\sigma_i}}{\cancel{\sigma_i}} \right)^2 \approx N$$

For N large, a good fit to a data set will yield $\chi^2 \approx N$.

If you find that χ^2 is significantly less than N , it prob. means that σ_i has been over estimated.



If χ^2 is larger than N , it could be an indication that your model does not capture all of the features present in the data.



✓
Pts encircled by red,
contribute more than 1
to χ^2

Overall $\chi^2 > N \rightarrow$ indicates a poor fit
(assumes that reasonable
estimate of σ_i values)

$\rightarrow \chi^2$ is a "goodness of fit" parameter
when σ_i estimated appropriately.

Expecting $\chi^2 \approx N$ is correct when N is large.
However, consider the following extreme cases.

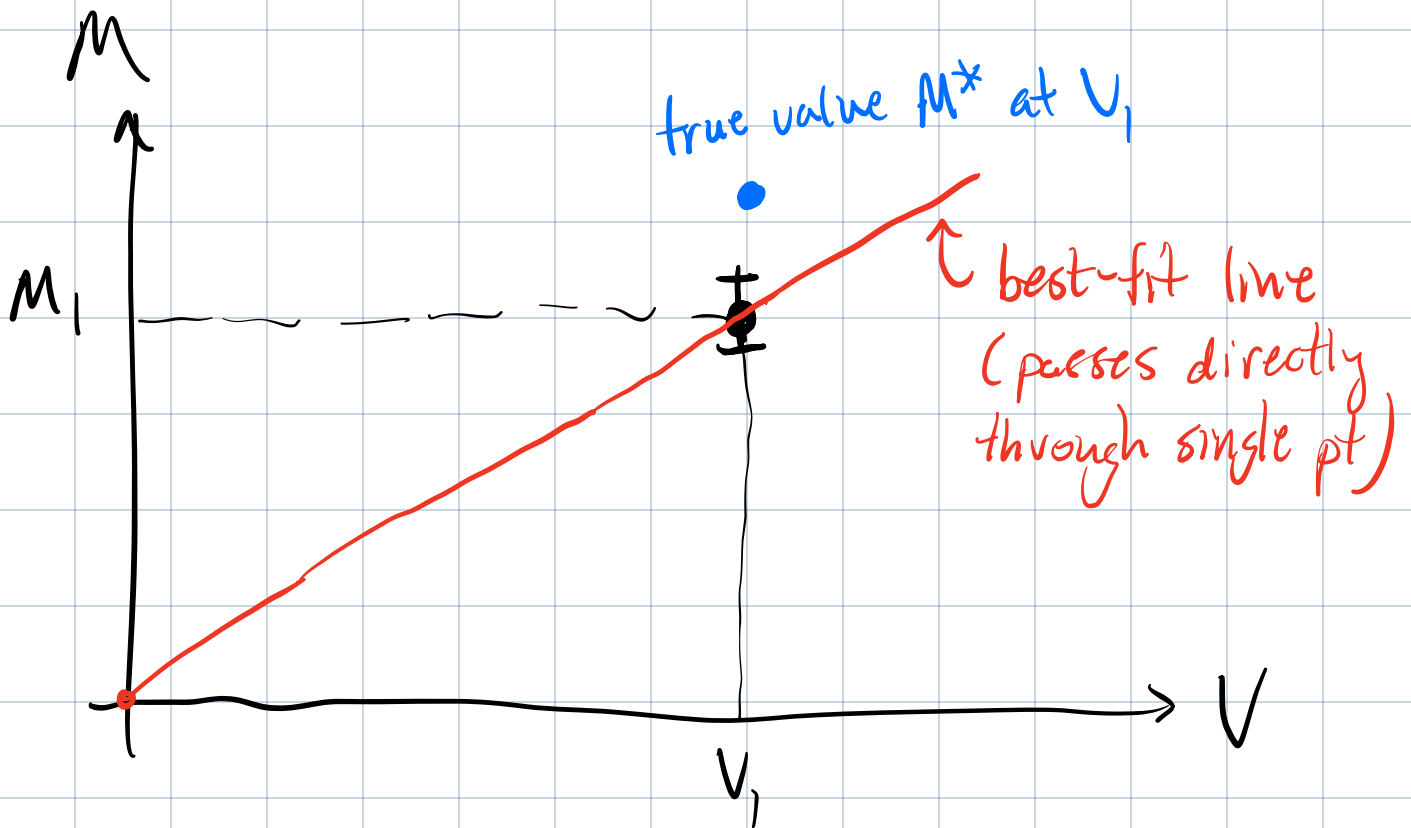
Meas. the volume of an object & measure
its mass. Then, determine the density ρ
from a fit to the data.

$$M = \rho V$$

$y \nearrow$ slope \nearrow x

M vs V has
a slope $b = \rho$
& intercept of zero.

1-parameter fit (ρ) w/ $N=1$ data pts.



If we have only a single pt to fit to,
the deviation $y_i - y(x_i) = 0$
 $\therefore \chi^2 = 0$.

Since our best-fit line is braced to pass exactly through our single measurement, the calculate χ^2 value is artificially low ($\chi^2 = 0$).

If we could calculate χ^2 using the true value of

M (called M^* in figure) then we would get $\chi^2 \approx 1 (=N)$.

Resolution is that expected value of χ^2 is not N , but is $N-m$ where m is the number of parameters in your fit.

For $N=1$ & a one-parameter fit (slope or density ρ) we expect

$$\chi^2 = N - m = 1 - 1 = 0 \checkmark.$$

↑ ↑
no. of data pts. no. of fit parameters.

χ^2 summary:

Expect $\chi^2 \approx N - m$

usually call $N - m = \nu$

"no. of degrees of freedom"

▣ If we make reasonable estimates of σ_i ,
then $\chi^2 > N-m$ indicates a poor fit.

▣ If $\chi^2 < N-m$, then prob. over-estimate
the value of σ_i

Reduced χ^2 is denoted

Expect $\chi^2_v = 1$ for
a good fit.

$\chi^2_v > 1$ is an
indication of a poor fit.

$$\chi^2_v = \frac{\chi^2}{\nu}$$
$$= \frac{\chi^2}{N-m}$$