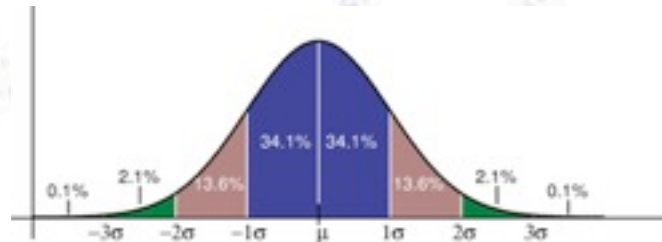# Applied Statistics

The Chi-Square Distribution, Fit & Test

The Chi-Square fit is also (originally) known as Method of Least Squares

Troels C. Petersen (NBI)

"Statistics is merely a quantization of common sense"

The discovery of Ceres

Ophiuchus

Dwarf planet and the largest astroid (r=487km)

Theta Ophiuchi

1st

8th

16th

Ceres

31st

South

Tuesday, September 10, 13

The discovery of Ceres

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8th
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Ceres

On the 1st of January 1801 Giuseppe Piazzi discovered "new light" and could follow this comet/planet until 11th of February. He published the positions, but due to Ceres being behind the sun, it would be out of sight until the following winter. Following the calculations of a 24 year old mathematician/physicist, it was recovered on the 31st of December 1801 by von Zach and H. Olbers.
The young man's name was Carl Friedrich Gauss, and the method he used/invented for this was…

South

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The young man's name was Carl Friedrich Gauss, and the method he used/invented for this was…

…method of least squares!

South

Method of Least Squares

The method of least squares is a standard approach to the approximate solution of **overdetermined systems**, i.e., sets of equations in which there are more equations than unknowns.

"Least squares" means that the overall solution minimizes the sum of the squares of the errors made in solving every single equation.
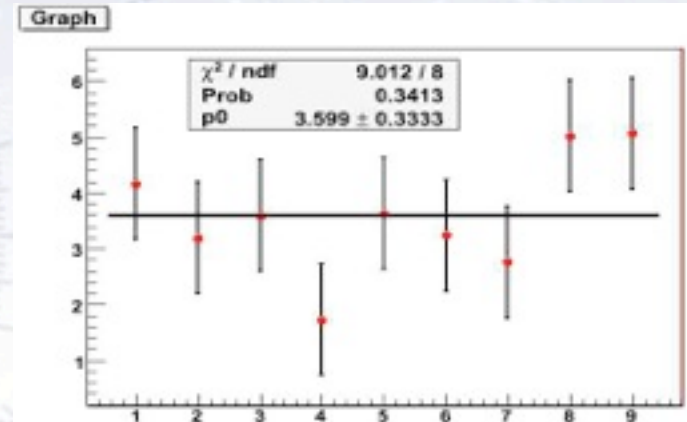
The most important application is in **data fitting**. The best fit in the least-squares sense minimizes the sum of squared residuals, a residual being the difference between an observed value and the fitted value provided by a model.

Least squares corresponds to the maximum likelihood criterion if the experimental errors have a normal distribution.

Method of Least Squares

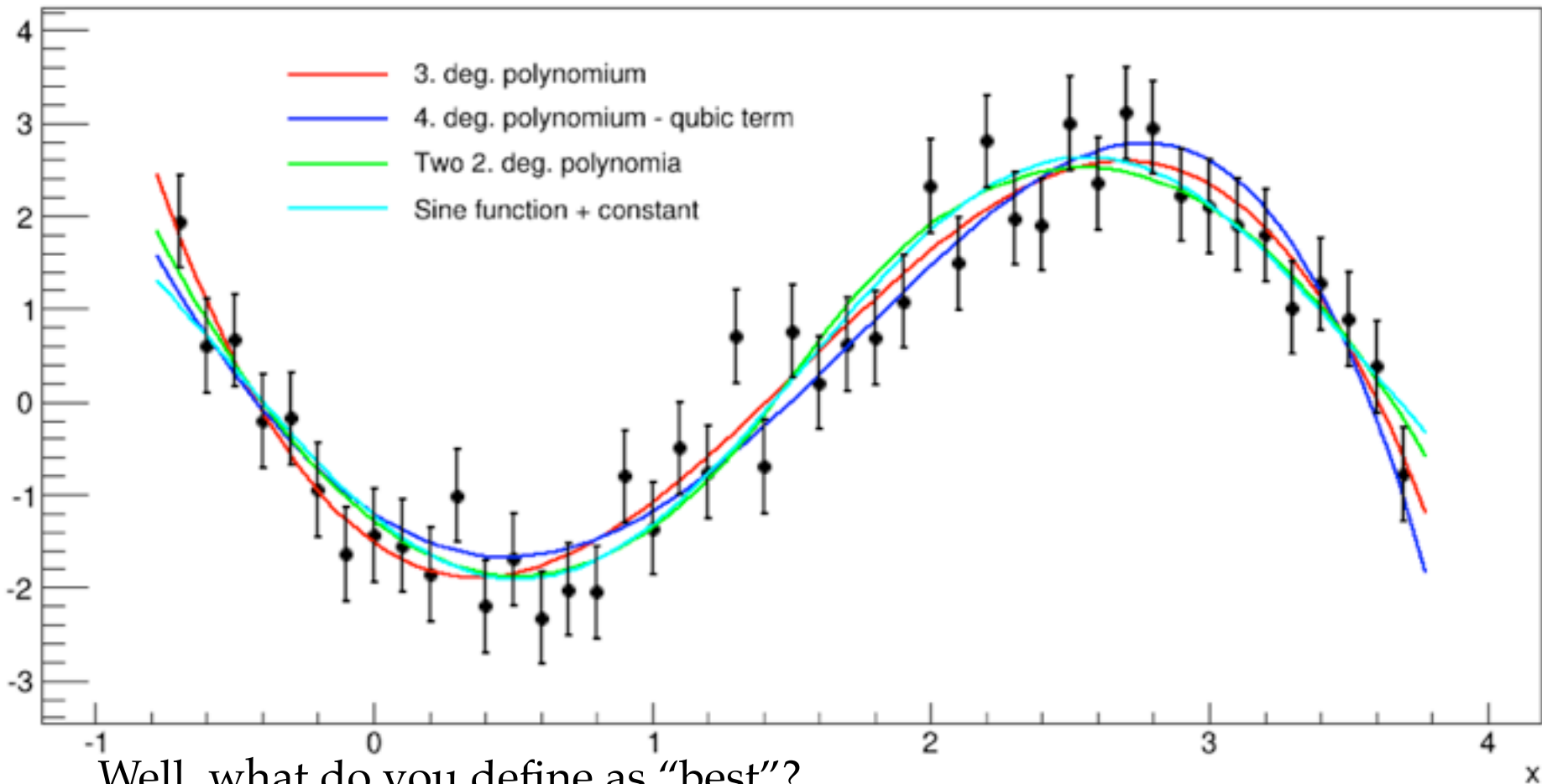The problem at hand was determining the curve that best fitted data:
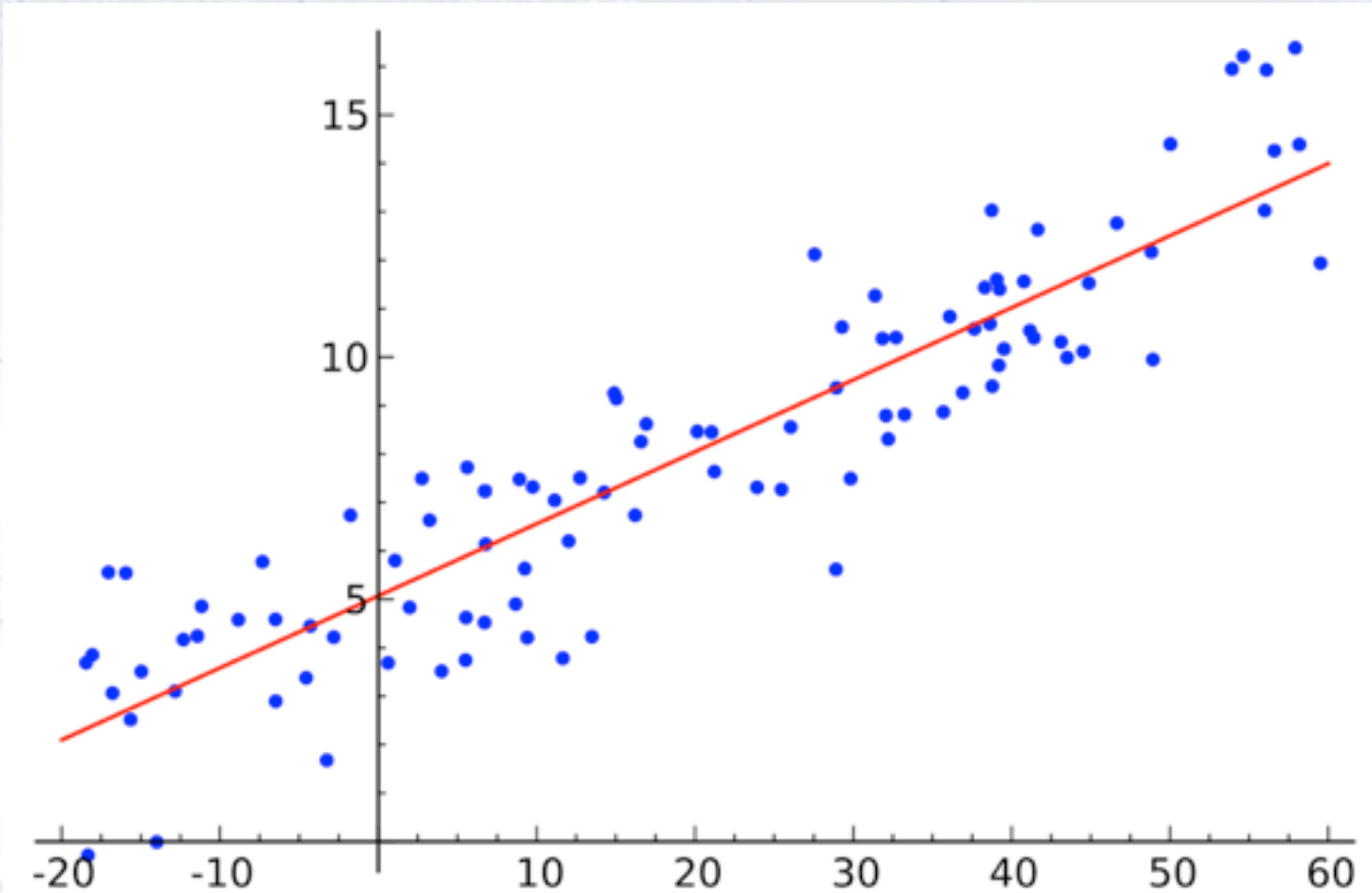
Originally, uncertainties were not included (not "invented" yet!)

Method of Least Squares

Look at the figure below, and determine which curve fits best...

Illustration of Method of Least Squares

— 3. deg. polynomium
— 4. deg. polynomium - qubic term
— Two 2. deg. polynomia
— Sine function + constant

Well, what do you define as "best"?

Defining the Chi-Square

Problem Statement: Given N data points (x,y), adjust the parameters θ of a model, such that it fits data best.

The best way to do this, given uncertainties σ_i on y_i is by minimizing:

$$\chi^2(\theta) = \sum_i^N \frac{(y_i - f(x_i, \theta))^2}{\sigma_i^2}$$

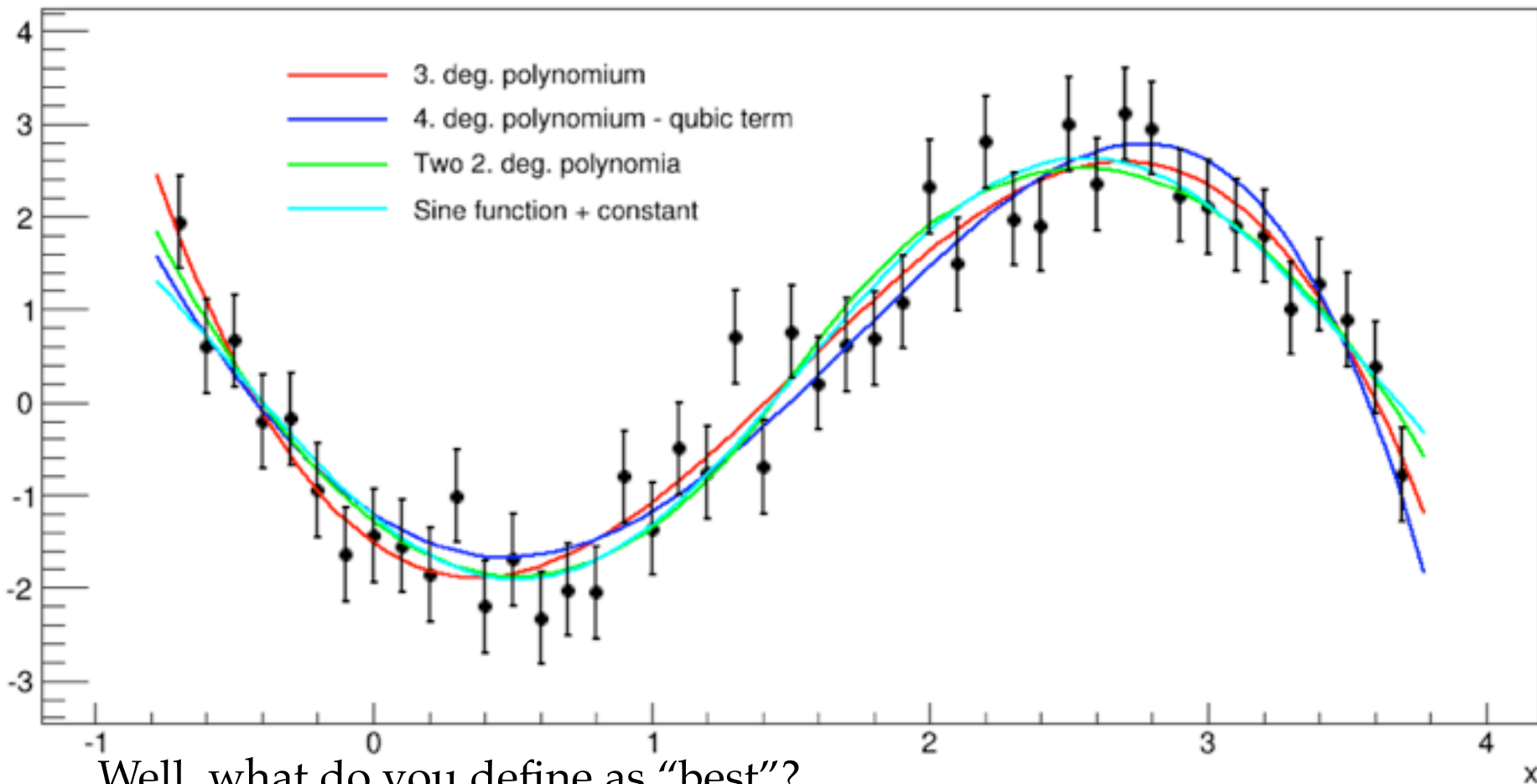## The power of this method is hard to overstate!

Not only does it provide a simple, elegant and unique way of fitting data, but more importantly it provides a **goodness-of-fit measure**.

This is the Chi-Square test!

Method of Least Squares

Look at the figure below, and determine which curve fits best...
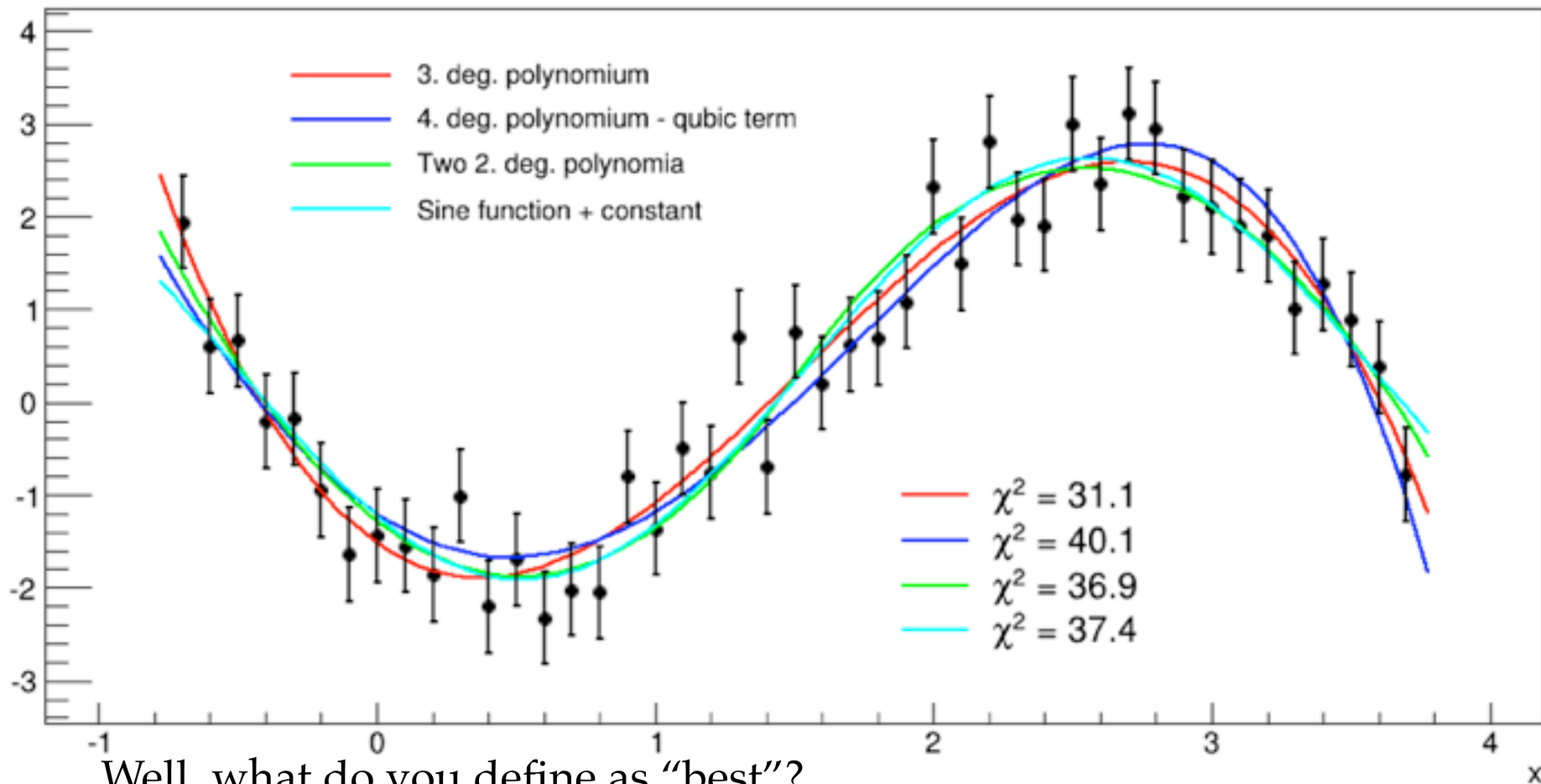

Illustration of Method of Least Squares

Legend:
- 3. deg. polynomium
- 4. deg. polynomium - qubic term
- Two 2. deg. polynomia
- Sine function + constant

Well, what do you define as "best"?

Method of Least Squares

Look at the figure below, and determine which curve fits best...

Illustration of Method of Least Squares

Legend:
- 3. deg. polynomium (red)
- 4. deg. polynomium - qubic term (blue)
- Two 2. deg. polynomia (green)
- Sine function + constant (cyan)

$\chi^2 = 31.1$ (red)
$\chi^2 = 40.1$ (blue)
$\chi^2 = 36.9$ (green)
$\chi^2 = 37.4$ (cyan)

Well, what do you define as "best"?

The Chi-Square distribution and test

The **Chi-Square distribution** for N_{dof} degrees of freedom is the distribution of the sum of the squares of N_{dof} normally distributed random variables.

$$\frac{1}{2^{\frac{k}{2}}\Gamma\left(\frac{k}{2}\right)}\, x^{\frac{k}{2}-1}e^{-\frac{x}{2}}$$
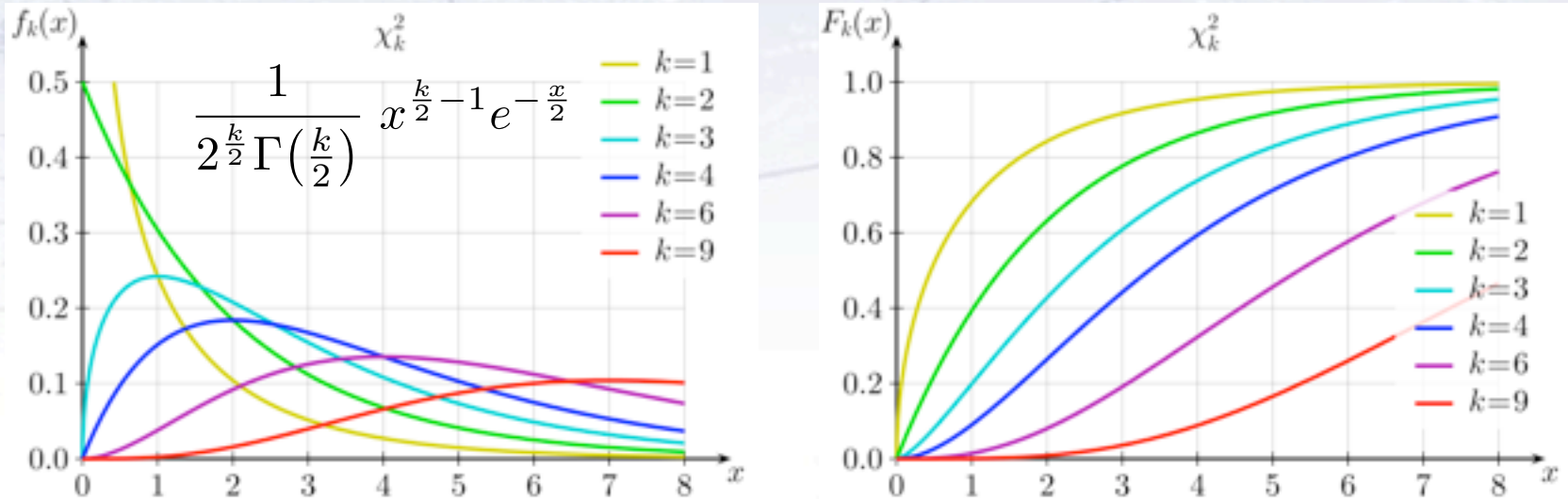
The **Chi-Square test** consists of comparing the Chi-Square value obtained from a fit with the PDF of expected Chi-Square values. This allows the calculation of the *probability* of observing something with the same Chi-Square value or higher...

Rule of thumb: Chi-Square should roughly match N_{dof}

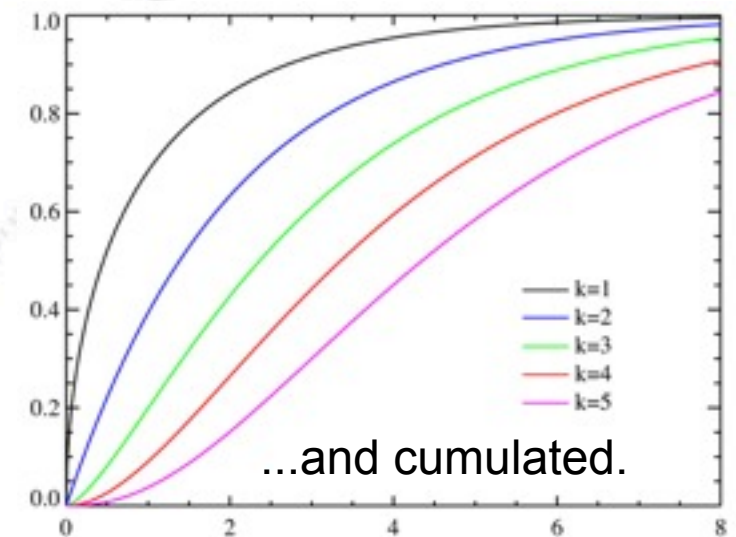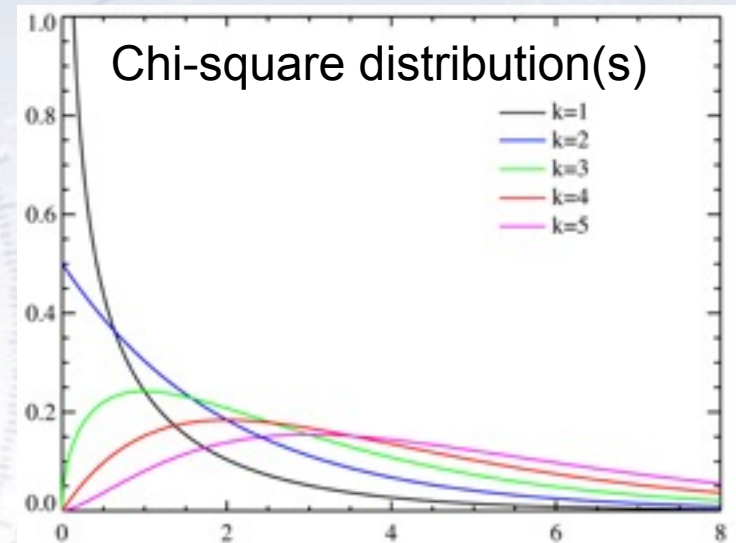# Chi-Square probability calculation

Given a **Chi-square value** and a **number of degrees of freedom** (Ndof), one can obtain a **"goodness-of-fit"**.

It is known, what Chi-square values to expect given the Ndof. One can therefore compare to this (Chi-square) distribution, and see...

what is the probability of getting this Chi-square value or something worse!

Example:
A fit gave the Chi-square 7.1 with 5 dof. The chance of getting this Chi-square or worse is... (reading the pink bottom curve (Ndof = k = 5) at 7.1)...

Chi-square distribution(s)
k=1
k=2
k=3
k=4
k=5

...and cumulated.
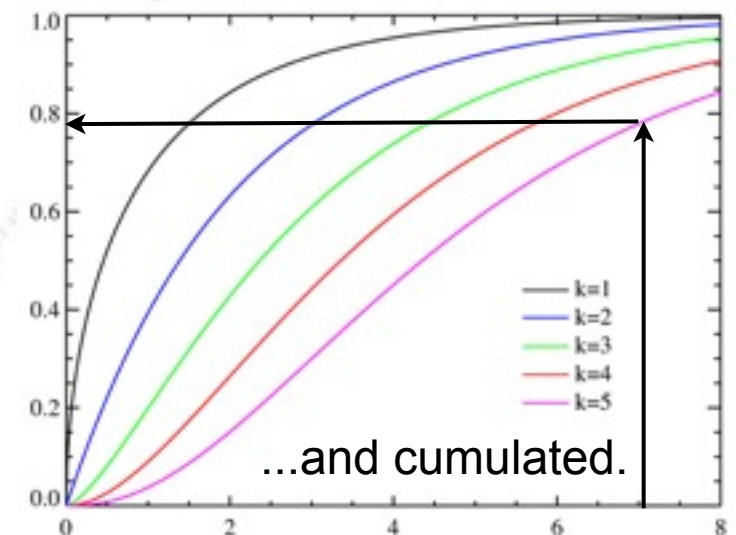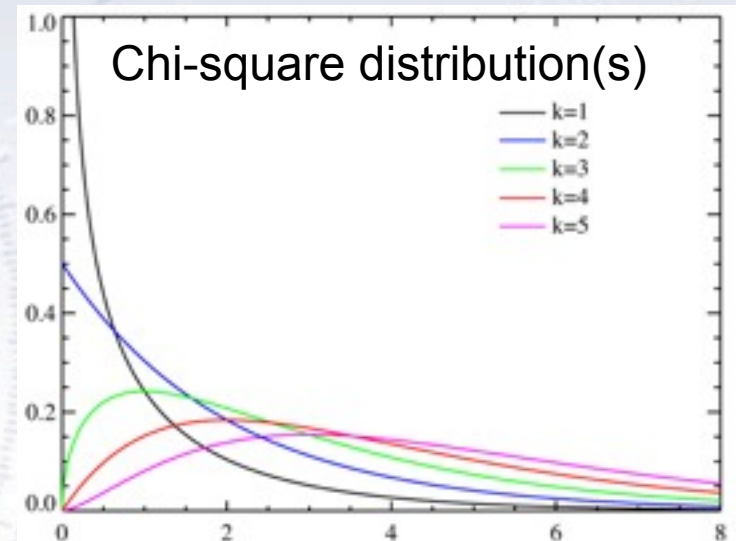
Chi-Square probability calculation

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Example:

A fit gave the Chi-square 7.1 with 5 dof. The chance of getting this Chi-square or worse is... (reading the pink bottom curve (Ndof = k = 5) at 7.1)... 1 - 0.78 = **22%**

Chi-square distribution(s)

...and cumulated.

Chi-Square probability calculation

In the table below, one can get a quick estimate for low N_{dof}.

Degrees of freedom (df)	x^2 value [16]										
1	0.004	0.02	0.06	0.15	0.46	1.07	1.64	2.71	3.84	6.64	10.83
2	0.10	0.21	0.45	0.71	1.39	2.41	3.22	4.60	5.99	9.21	13.82
3	0.35	0.58	1.01	1.42	2.37	3.66	4.64	6.25	7.82	11.34	16.27
4	0.71	1.06	1.65	2.20	3.36	4.88	5.99	7.78	9.49	13.28	18.47
5	1.14	1.61	2.34	3.00	4.35	6.06	7.29	9.24	11.07	15.09	20.52
6	1.63	2.20	3.07	3.83	5.35	7.23	8.56	10.64	12.59	16.81	22.46
7	2.17	2.83	3.82	4.67	6.35	8.38	9.80	12.02	14.07	18.48	24.32
8	2.73	3.49	4.59	5.53	7.34	9.52	11.03	13.36	15.51	20.09	26.12
9	3.32	4.17	5.38	6.39	8.34	10.66	12.24	14.68	16.92	21.67	27.88
10	3.94	4.86	6.18	7.27	9.34	11.78	13.44	15.99	18.31	23.21	29.59
P value (Probability)	0.95	0.90	0.80	0.70	0.50	0.30	0.20	0.10	0.05	0.01	0.001
	Non-significant								Significant		

Chi-Square probability interpretation

The Chi-Square probability can be interpreted as follows:

- If χ^2 / **Ndof** ≃ **1** or more precisely if $0.01 < p(\chi^2, \textbf{Ndof}) < 0.99$, then all is good.
- If χ^2 / **Ndof** ≫ **1** or more precisely if $p(\chi^2, \textbf{Ndof}) < 0.01$, then your fit is bad, and your hypothesis is probably not correct.
- If χ^2 / **Ndof** ≪ **1** or more precisely if $0.99 < p(\chi^2, \textbf{Ndof})$, then your fit is TOO good and you probably overestimated the errors.

If the statistics behind the plot is VERY high (great than 10^6), then you might have a hard time finding a model, which truly describes all the features in the plot (as now tiny effects become visible), and one hardly ever gets a good Chi-Square probability.
However, in this case, one should not worry too much, unless very high precision is wanted.

Chi-Square for binned data

If the data is binned (i.e. put into a histogram), then Pearson's Chi-square applies:

Hist_Sum	
Entries	100000
Mean	0.002193
RMS	1.003
χ^2 / ndf	58.35 / 84
Prob	0.9851
Constant	3981± 15.4
Mean	0.001794 ± 0.003169
Sigma	1.002 ± 0.002
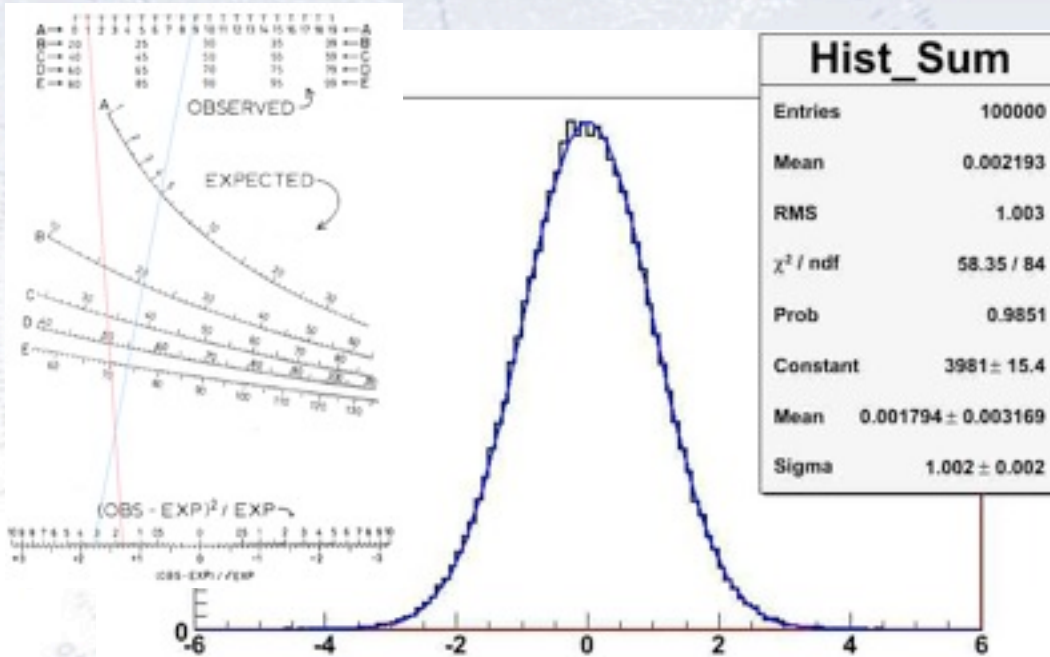
The formula (based on Poisson statistics) is:

$$\chi^2 = \sum_{i \,\in\, \mathrm{bin}} \frac{(O_i - E_i)^2}{E_i}$$

Chi-Square for binned data

While Pearson's Chi-square test is quite useful, it has some limitations, especially when some bins have low statistics.

The expected cell count (E_i) should not be too low. Some require 5 or more, and others require 10 or more. A common rule is 5 or more in 80% of bins, but no cells with zero expected count.
When this assumption is not met, Yates's Correction can be applied.
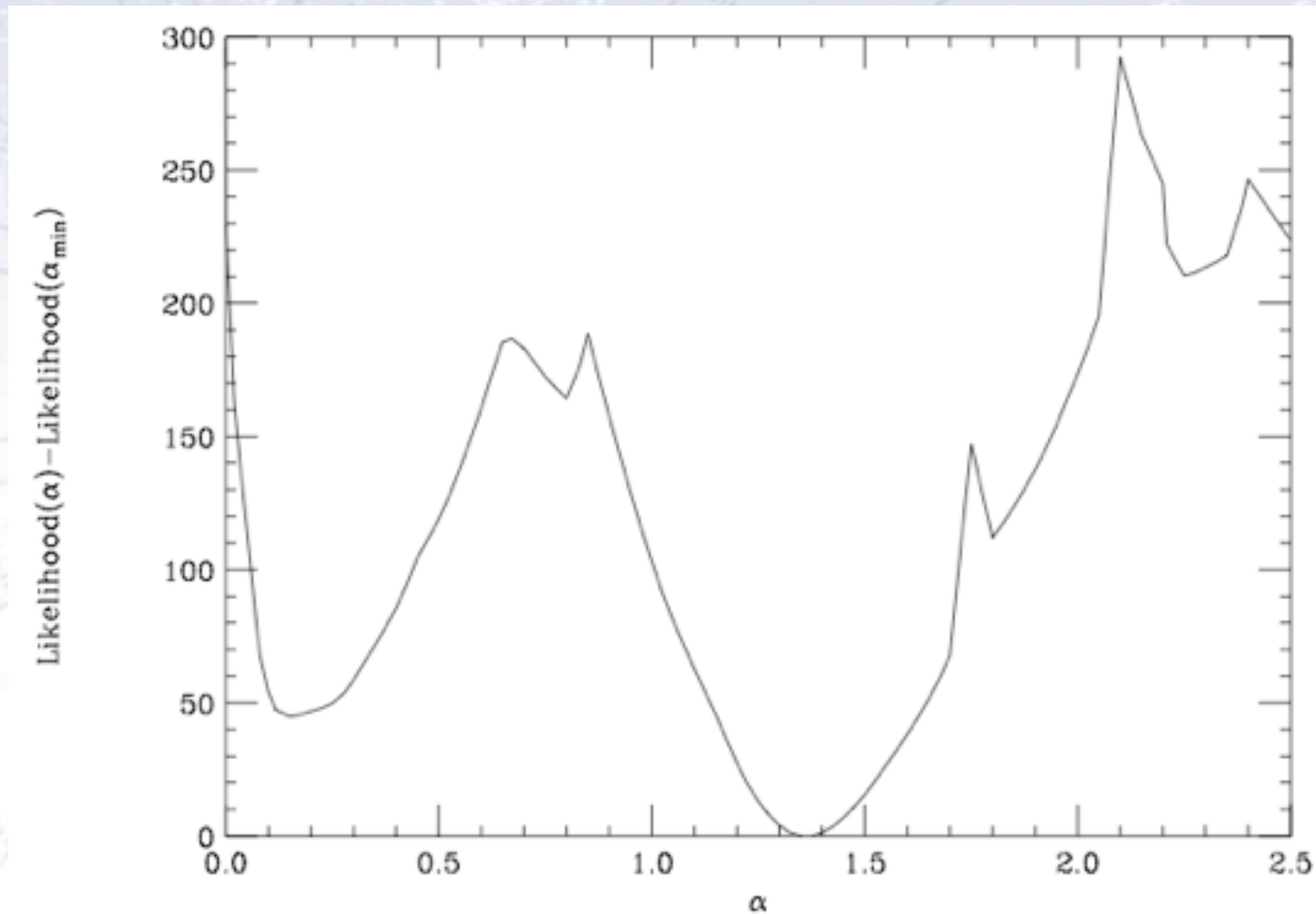
One alternative is the likelihood fit, which does not suffer under low statistics.

$$\chi^2 = \sum_{i \in \text{bin}} \frac{(O_i - E_i)^2}{E_i}$$

Another alternative is the G-test, which is more robust at low statistics. However, I've never seen it in use.

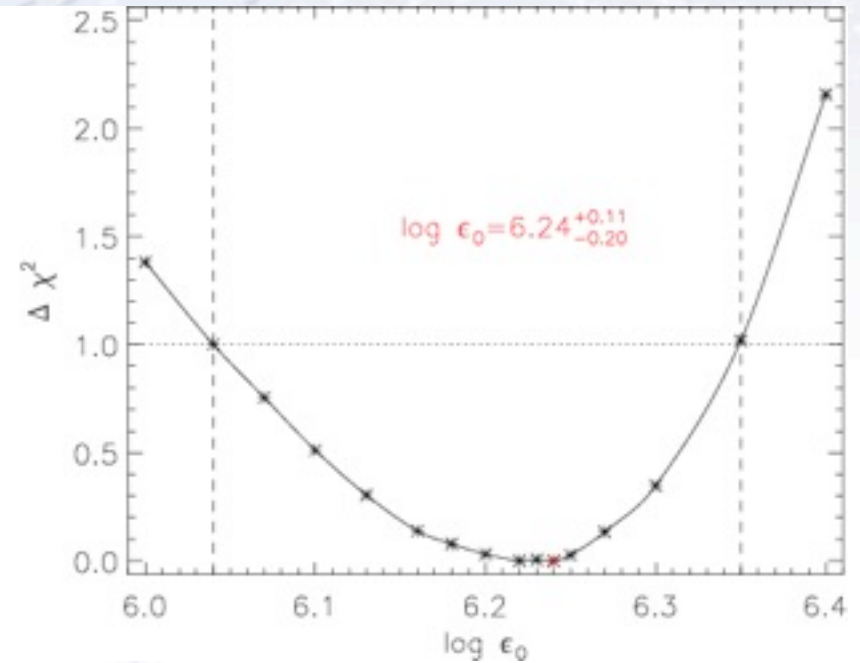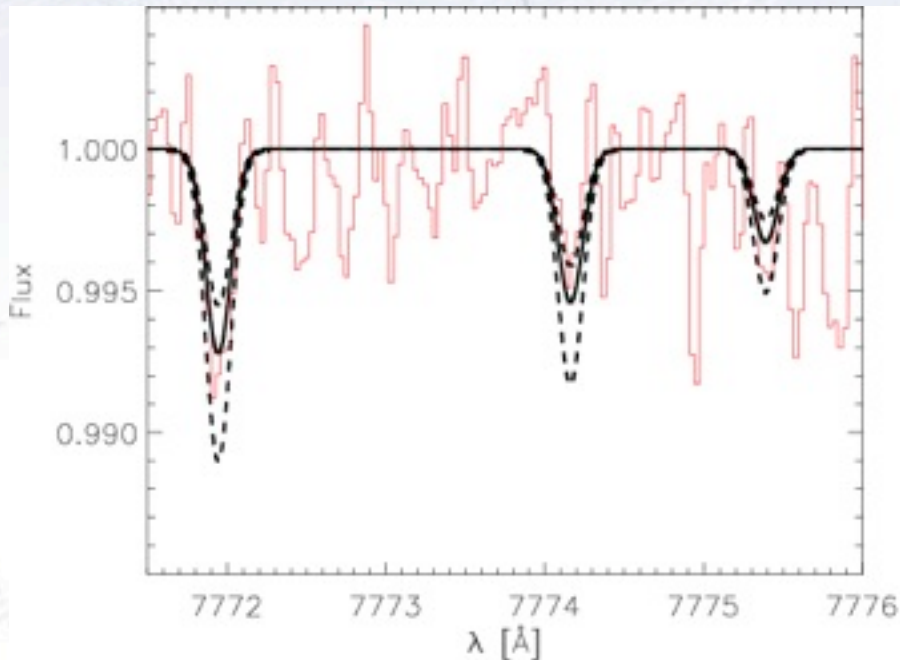$$G = 2 \sum_{i \in \text{bin}} O_i \, \ln(O_i/E_i)$$

Example of Chi-Square

The fact that there are several minima makes fitting difficult/uncertain!
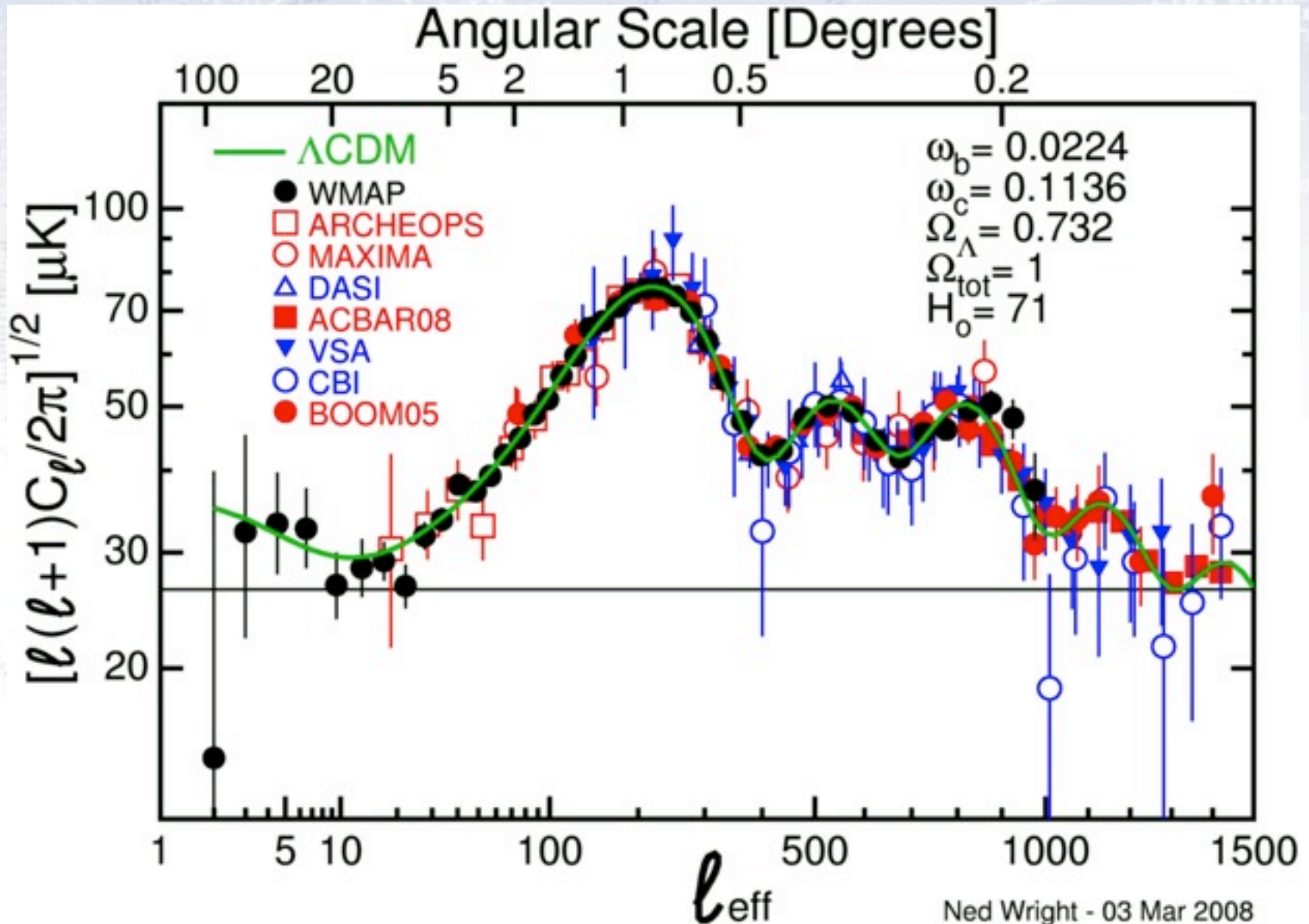*Always give good starting values!!!*

Example of Chi-Square

Uncertainties need not always be symmetric (though that is usually better!)

The uncertainty on a parameter is found where the Chi2 has increased by 1 from the minimum.

Example of Chi-Square

Notes on the ChiSquare method

"It was formerly the custom, and is still so in works on the theory of observations, to derive the method of least squares from certain theoretical considerations, the assumed normality of the errors of the observations being one such.
It is however, more than doubtful whether the conditions for the theoretical validity of the method are realized in statistical practice, and the student would do well to regard the method as recommended chiefly by its comparative simplicity and by the fact that it has **stood the test of experience***".*

[G.U. Yule and M.G. Kendall 1958]