

Spectral Analysis & Statistics

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Overview

1. Ideal experiment / X-ray Astrophysics
 2. Spectral analysis
 3. Questions to be answered by the spectra
 4. Statistical tests
 5. The large view
 6. Golden rules of statistics
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Ideal Experiment

- **expected spectrum: either from theory or short test measurement**
- **resolution of detector \gg intrinsic line-width**
- **effective area: known per resolution element**
- **background: either known or short test measurement**
- **optimization of experiment:**
 - **exposure time per resolution element: such that constant signal/noise ratio is expected**



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X-ray Astrophysics

- spectrum and flux:
 - generally not known with sufficient accuracy to plan observation accurate
- resolution vs. line-width:
 - $pn < MOS \ll RGS$
 - in general:
 - line-width $<$ resolution,
 - **redistribution**
- background:
 - **not well known in advance and variable**
- observation optimization:
 - only partly possible



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Ideal Experiment

- **reconstruction of spectrum:**
 - data [counts]:
 - $Y_1, Y_2, Y_3, Y_4 \dots Y_{2000}$
 - effective area:
 - $a_1, a_2, a_3, a_4 \dots a_{2000}$
 - exposure time:
 - $t_1, t_2, t_3 \dots t_{2000}$
 - spectra [photons/sec/cm²] :
 - $f_1 = Y_1 / t_1 / a_1 + 0 \times Y_2 / t_1 / a_1$
 - $f_2 = Y_2 / t_2 / a_2 + 0 \times Y_1 / t_1 / a_2$
 - $f_3 = Y_3 / t_3 / a_3 + 0 \times Y_1 / t_1 / a_3$
- **start with the spectrum:**
 - input spectrum [photons/sec]:
 - $P_1(E_1), P_2(E_2), P_3, P_4 \dots$
 - registered with a certain probability: effective area
 - $a_1 \times t_1, a_2 \times t_2, \dots$
 - redistributed with respect of counts channels:
 - $Y_1(e_1) = P_1(E_1) \times a_1(E_1) \times t_1(E_1) + P_2(E_2) \times 0$



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X-ray Astrophysics

- **reconstruction of spectrum:**
 - data [counts]:
 - $Y_1, Y_2, Y_3, Y_4 \dots Y_{2000}$
 - effective area [cm²]:
 - $a_1, a_2, a_3, a_4 \dots a_{2000}$
 - exposure time:
 - $t_1, t_2, t_3 \dots t_{2000}$
 - spectra [photons/sec/cm²] :
 - $f_1 = Y_1 / t_1 / a_1 + b \times Y_2 / t_1 / a_1$
 - $f_2 = Y_2 / t_2 / a_2 + c \times Y_1 / t_1 \dots$
 - $f_3 = Y_3 / t_3 / a_3 + c \times Y_1 / \dots$
- **start with the spectrum:**
 - input spectrum [photons/sec/cm²]:
 - $f_1(E_1), f_2(E_2), f_3, f_4 \dots$
 - registered with a certain probability: effective area
 - $a_1 \times t_1, a_2 \times t_2, \dots$
 - redistributed with respect of counts channels:
 - $Y_1(e_1) = f_1(E_1) \times a_1(E_1) \times t_1(E_1) + P_2(E_2) \times b$



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X-ray Astrophysics

reconstruction of spectrum:

- data [counts]:
 - $Y_1, Y_2, Y_3, Y_4 \dots Y_{2000}$
- effective area [cm^2]:
 - $a_1, a_2, a_3, a_4 \dots a_{2000}$
- exposure time:
 - $t_1, t_2, t_3 \dots t_{2000}$
- spectra [counts/sec/ cm^2] :
 - $f_1 = Y_1 / t_1 / a_1 + b \times Y_2 / t_1 / a_1$
 - $f_2 = Y_2 / t_2 / a_2 + c \times Y_1 / t_1 / a_1 \dots$
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start with the spectrum:

- input spectrum [photons/sec/ cm^2]:
 - $P_1(E_1), P_2(E_2), P_3, P_4 \dots$
- registered with a certain probability: effective area
 - $a_1 \times t_1, a_2 \times t_2, a_3, a_4$
- redistributed with respect of counts channels:
- $Y_1(e_1) = P_1(E_1) \times a_1(E_1) \times t_1(E_1) + P_2(E_2) \times b \dots$



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Redistribution and Resolution

- the channels in the spectra reflect a purely mathematical sorting
 - a count in a channel, which corresponds to an energy k , can not be identified with a photon of energy k
 - physical sampling, “real” channels, “independent bins”, resolution:
 - simplest case: (Abbe, 1820-1830) what is the minimum distance between two point which can be resolved with a given resolution?
 - today more complex: Shannon information theorem, Nyquist criteria
 - **information content of a given spectra does not increase if the sampling goes below FWHM/3**
 - most of the spectra provided in X-ray missions are over-sample
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Concept Of Spectral Analysis

1. **physically motivated model spectra which is a function of parameters (x_1, x_2, \dots, x_i) and energy (e_1, e_2, \dots)**
 - **Example:**
 - thermal emission of a blackbody (BB),
 - with parameters: temperature (T) and norm (N)
 - $BB(T, N, e_i)$
2. **multiplication with effective area (a_1, a_2, \dots): expected flux (f) as function of energy and model**
 - $F(BB(T, N), e_i) = BB(T, N, e_i) * a_{ij}, (i=j \ a_{ij} = k_j, i \text{ not } j \ a_{ij}=0)$
3. **folding with redistribution matrix (Detector Response Matrix (b_{ij})) and multiplication with exposure time (t): counts as function of channel (k_i) and model:**
 - $counts(c_j) = F(BB(T, N), e_i) * b_{ij}$
4. **statistical analysis**
 - is model true?
 - hypothesis tests
 - can the description of the spectra be improved?
 - new loop with changed parameter of model

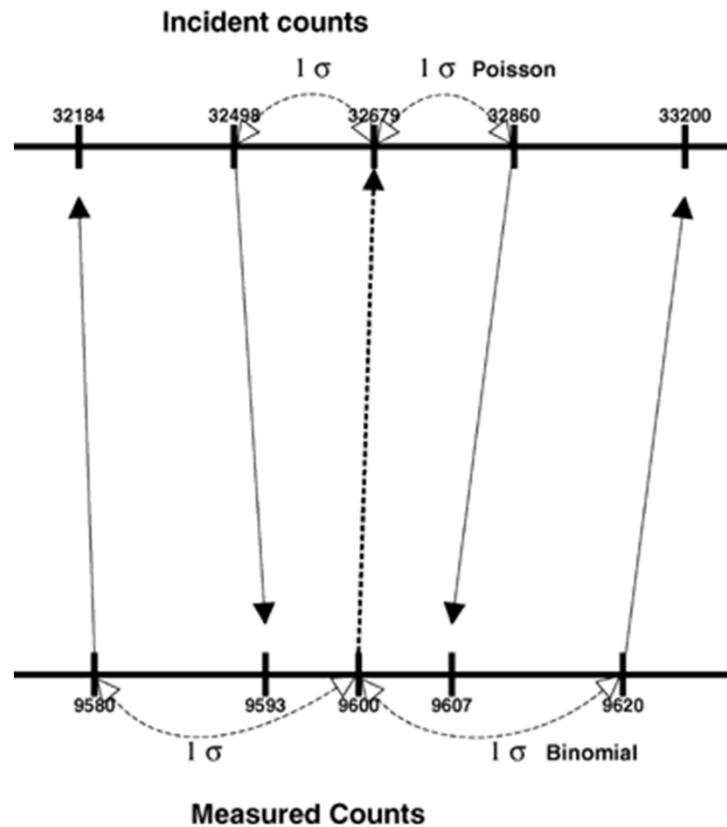
Standard Questions

- given a data set the most common questions are:
 - **distributions**
 - 50 counts are measured. What is there distribution?
 - **hypothesis test:**
 - ASCA showed a power law with $\Gamma=2.1$ and flux =1.5 → does this model describe the pn data?
 - **estimate of parameters:**
 - the data can be described with a power law. What is the index and the normalization?
 - **estimate of confidence level:**
 - what is the error of the power law and what is the error of the normalization?
 - **hypothesis test:**
 - **comparison between two fits:**
 - is a power-law or a power-law + emission line a better description of the data

Statistical Methods

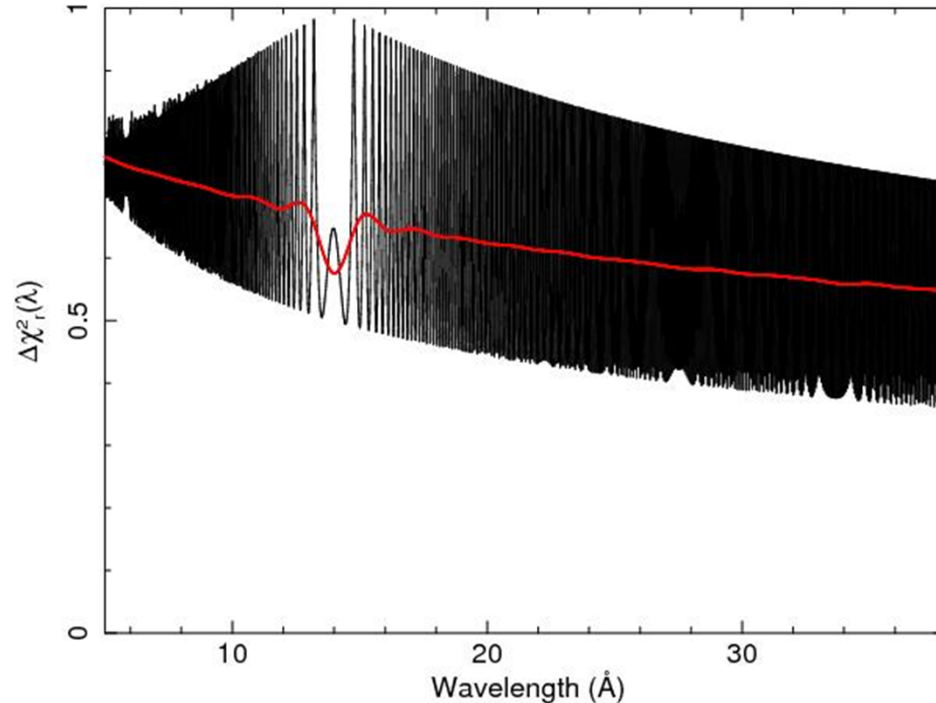
- statistical methods to answer the questions (provided in xspec):
 - **distributions**
 - Poisson distribution versus binomial distribution
 - **hypothesis test:**
 - χ^2 hypothesis test
 - **estimate of parameters**
 - modified minimum χ^2 method
 - minimize C value (C-statistics)
 - **estimate of confidence level**
 - determine parameter for $\chi^2 = \chi^2_{\text{minimum}} + \Delta \chi^2$
 - **comparison between two fits**
 - F-test

Comments to Distributions I: OM



- Kuin & Rosen, 2008, MNRAS 383, 383
- Fordham et al., 2000, MNRAS 312, 83

Comments to Distributions II: RGS



Relative contribution per wavelength bin to the reduced χ^2 . Calculated and shown is the ratio $\text{Var}[Y_i]/E[Y_i]$ (see text). The thick line is a low-resolution spline approximation showing the trends in the average value. (From Kaastra, J. S. et al., 2011, A&A 534, A37)

- J. S. Kaastra et al. 2011, A&A 534, A37

- Study of 600 ks RGS spectrum of Mrk 509 with excellent quality

- **Fluxed Spectrum:**

- > 400 counts per bin

- Expectation: reduced $\chi^2 \gg 1$

- But: reduced $\chi^2 \approx 0.6$

- Counts are not χ^2 distributed

- use rescaled χ^2

- Small number of counts in fluxed RGS spectra do follow the C-statistic

Statistical Functions

- χ^2 is defined as: $\chi^2 = \sum (Y_m - Y_e)^2 / (Y_e)$
 - for $n \rightarrow \infty$
 - and requires:
 - Gaussian errors of individual data points
 - independence of data points
- modified χ^2 is defined as: $\chi^2_{\text{mod}} = \sum (Y_m - Y_e)^2 / (\Delta Y_m)^2$
 - assumption for $n \rightarrow \infty$ $\chi^2_{\text{mod}} \rightarrow \chi^2$
 - only valid if: $(\Delta Y_m)^2 \approx Y_e$
 - requirements of χ^2 are fulfilled
- C-statistics is defined as: $C = 2 \sum (Y_e - Y_m \ln Y_e)$
 - W. Cash, 1979, ApJ 228, 939
 - assumption for $n \rightarrow \infty$ $C \rightarrow \chi^2 + K$
 - Uses:
 - Poisson errors of individual data points
- F-test is defined as: $F = \chi^2_1 * \text{d.o.f}_2 / \chi^2_2 / \text{d.o.f}_1$



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Independence Of Data Points

- the measurement of each Y_i is statistically independent
- but the Y_i are correlated (over-sampling in provided spectra, resolution)
- statistically independent but correlated data points
- all (presented) test methods require independent data points
 - (small sentence at the beginning of each text book)
- never sample with a bin-size $< \text{FWHM}/3$, this provides
 - complete spectral information
 - avoids over-sampling

Independence Of Data Points

- example: same spectra, same signal/noise for each bin:
1. sampling without considering the resolution:
 - **reduced chi-squared = 0.8191388**
 - for 3364 degrees of freedom
 - **null hypothesis probability = 1.00**
 2. sampling with bin-size $> \text{FWHM}/3$:
 - **reduced chi-squared = 1.094959**
 - for 138 degrees of freedom
 - **null hypothesis probability = 0.210**

Asymptotic Distributed ...

- **all quantities used for the statistical analysis, are itself of statistical nature, i.e. they have an error**
 - **example: $\text{r.m.s.}(\chi^2) = \sqrt{2(\text{d.o.f.})}$**
- **all statistical proofs and theorems are valid for $n \rightarrow \infty$**
- **but we have often not an infinite number of counts**
 - **Example:**
 - **χ^2 tests requires:**
 - **signal to noise ≥ 5 for central Y_i and**
 - **signal to noise ≥ 4 for Y_1 and Y_{last}**
 - **number of $Y \geq 5$**

Parameter Estimation

- intrinsic correlation between error and data!
- example:
 - $\langle Y \rangle = 36$,
 - 10 data points $Y_1, Y_2 \dots$: 30, 42, 24, 48, 33, 39, 34, 38, 36, 36
 - error of $Y_1 = \sqrt{Y_1}$
- mean $Y = 360 / 10 = 36$
- minimum of $\chi^2 = ?$
- minimum of $\chi^2_{\text{mod}} = ?$
- minimum of $C = ?$



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Parameter Estimation

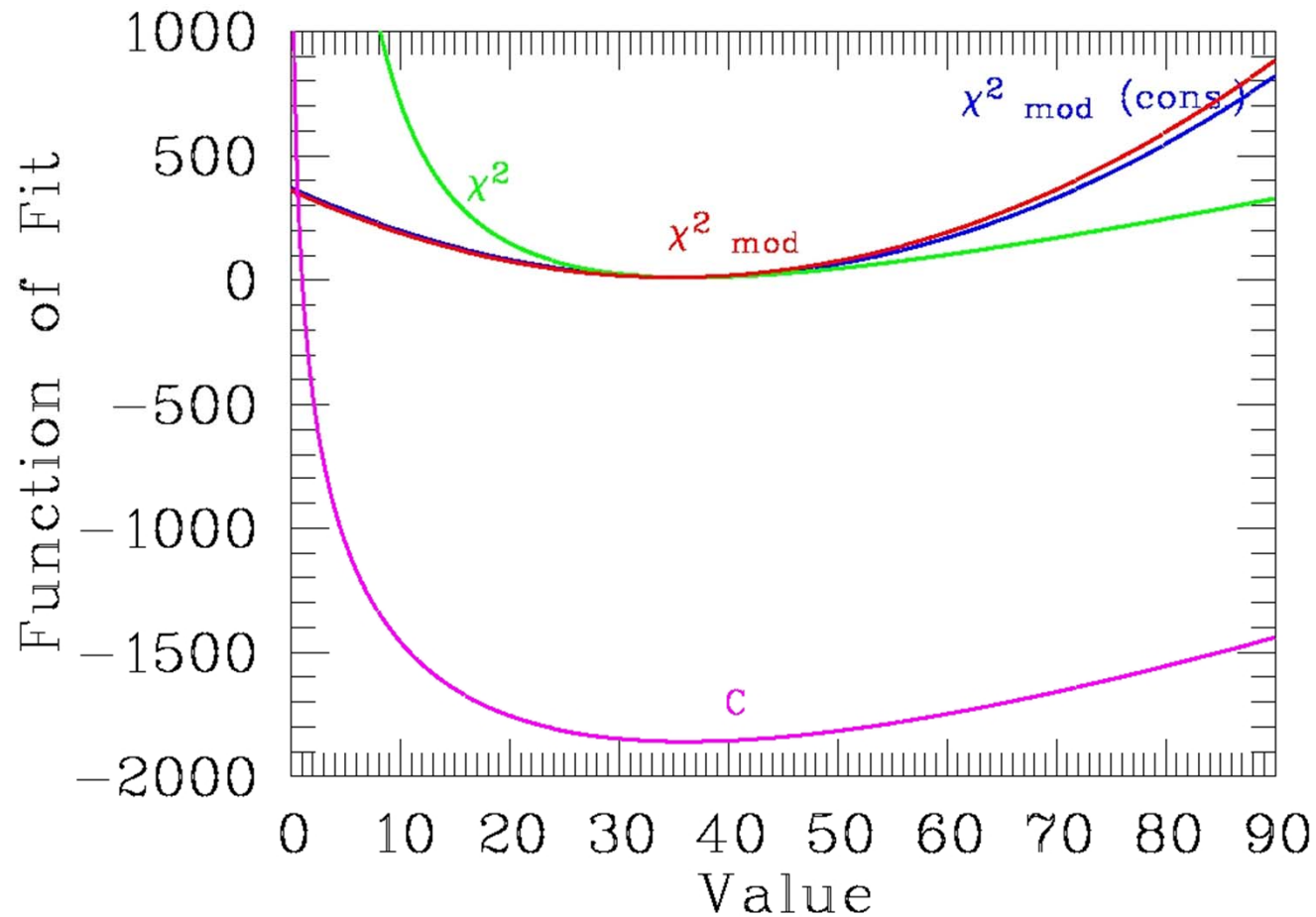
- intrinsic correlation between error and data!
 - example:
 - $\langle Y \rangle = 36$,
 - 10 data points $Y_1, Y_2 \dots$: 30, 42, 24, 48, 33, 39, 34, 38, 36, 36
 - Error of $Y_1 = \sqrt{Y_1}$
- | | | |
|---|---|------|
| – Mean $Y = 360 / 10$ | = | 36.0 |
| – Minimum of χ^2 | = | 36.5 |
| – Minimum of χ^2_{mod} | = | 34.9 |
| – Minimum of C | = | 36.0 |
| – Minimum of χ^2_{mod} (constant error) | = | 36.0 |



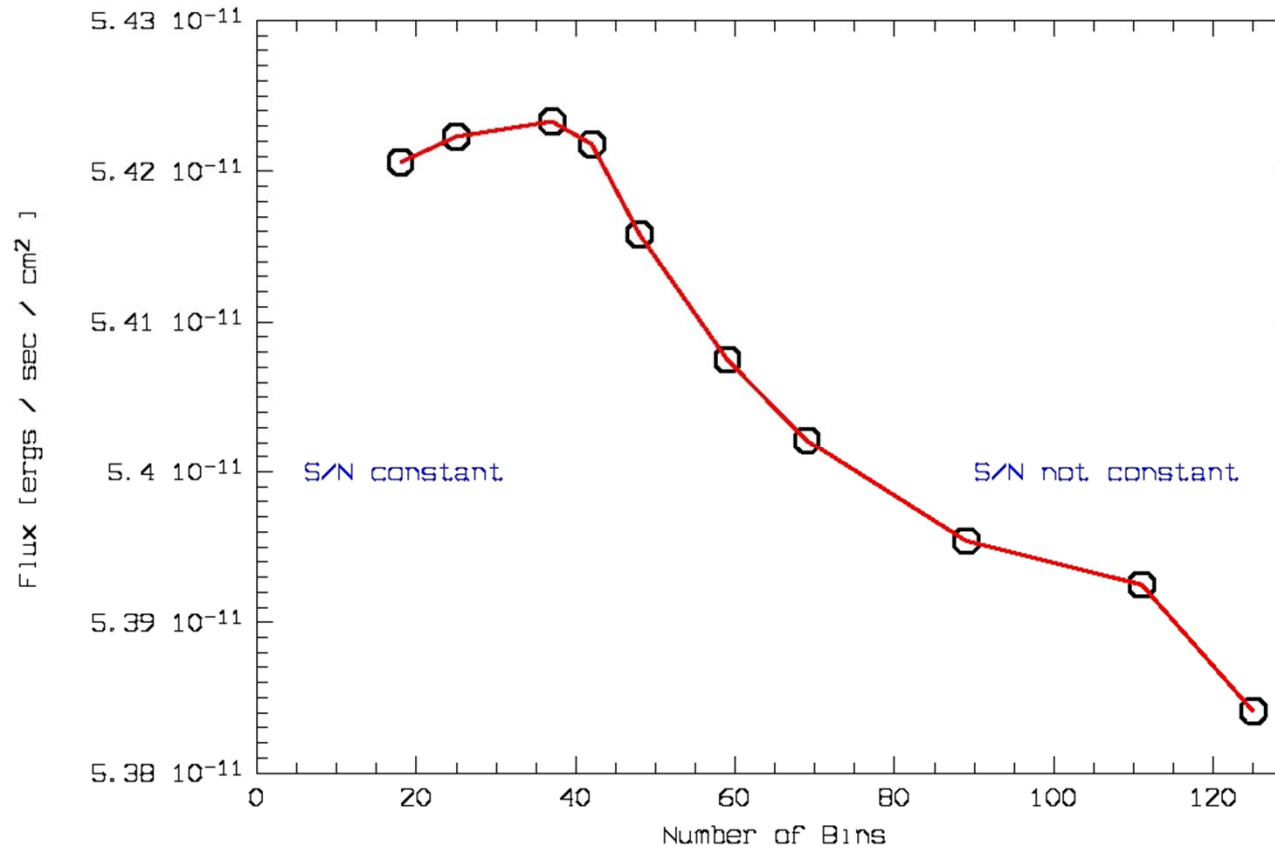
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Parameter Estimation



Real Example



Parameter Estimation

our data show a correlation between data points and their error: $\Delta Y \approx \sqrt{Y}$

bias for parameter estimation!

possible solutions:

- **estimator which is independent of error,**
 - e.g. C-statistics
- **ensure, that data points have equal errors,**
 - e.g. binning with constant signal/noise

the higher the numbers of counts the less important are bias effects!

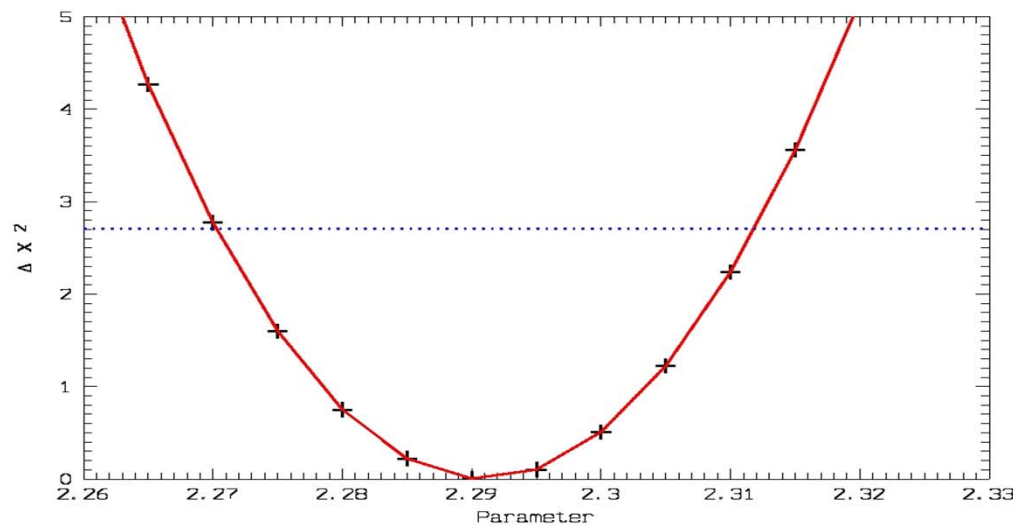


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Estimation Of Errors

- **method:** search for the value of the parameter which corresponds to a certain confidence level, i.e. $\Delta \chi^2$
 - **example:**
 - 90 % confidence level of one interesting parameter =>
 - $\Delta \chi^2 = 2.71$ ($\chi^2 = \chi^2_{\text{minimum}} + 2.71$)



Estimation Of Errors

- be careful:
- $\Delta \chi^2$ (parameter) is in general not a smooth function with a single minimum
 - plot $\Delta \chi^2$ as function of the parameter
- the interesting parameter is often not independent from other parameters
 - plot two-dimensional $\Delta \chi^2$ contours

Hypothesis Test

- Literature: J. Kaastra in X-ray spectroscopy in astrophysics, 1999, p 276
- correct model:
 - $\chi^2 = n - m$ (n: number of data points, m:=number of free parameters)
 - $\text{r.m.s.}(\chi^2) = \sqrt{2(n-m)}$
- false model:
 - $\chi^2 = n - m + r$
 - $r \sim \text{number of photons}$ (but independent from n)
- rejection of model
 - $\chi^2 \geq n - m + f \sqrt{2(n-m)} \rightarrow r > \sqrt{2(n-m)}$

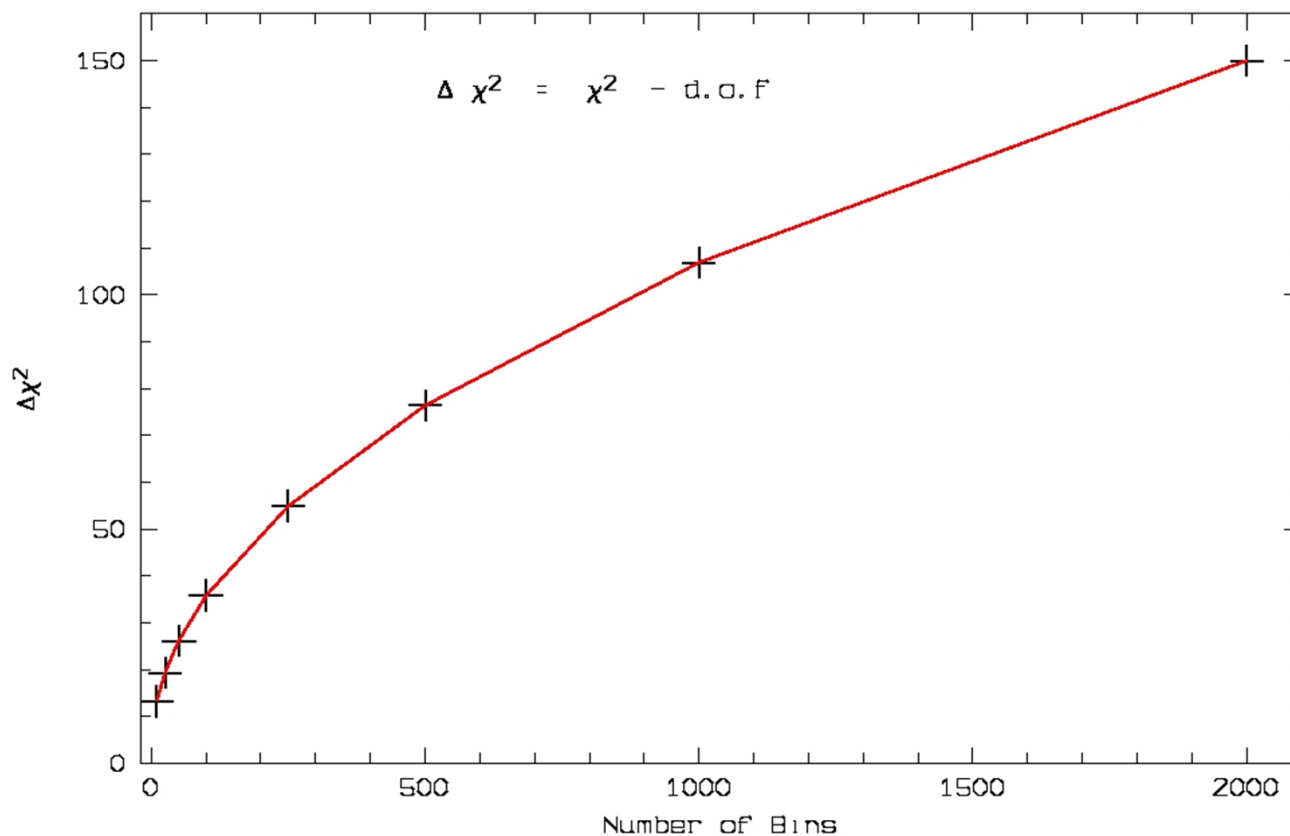


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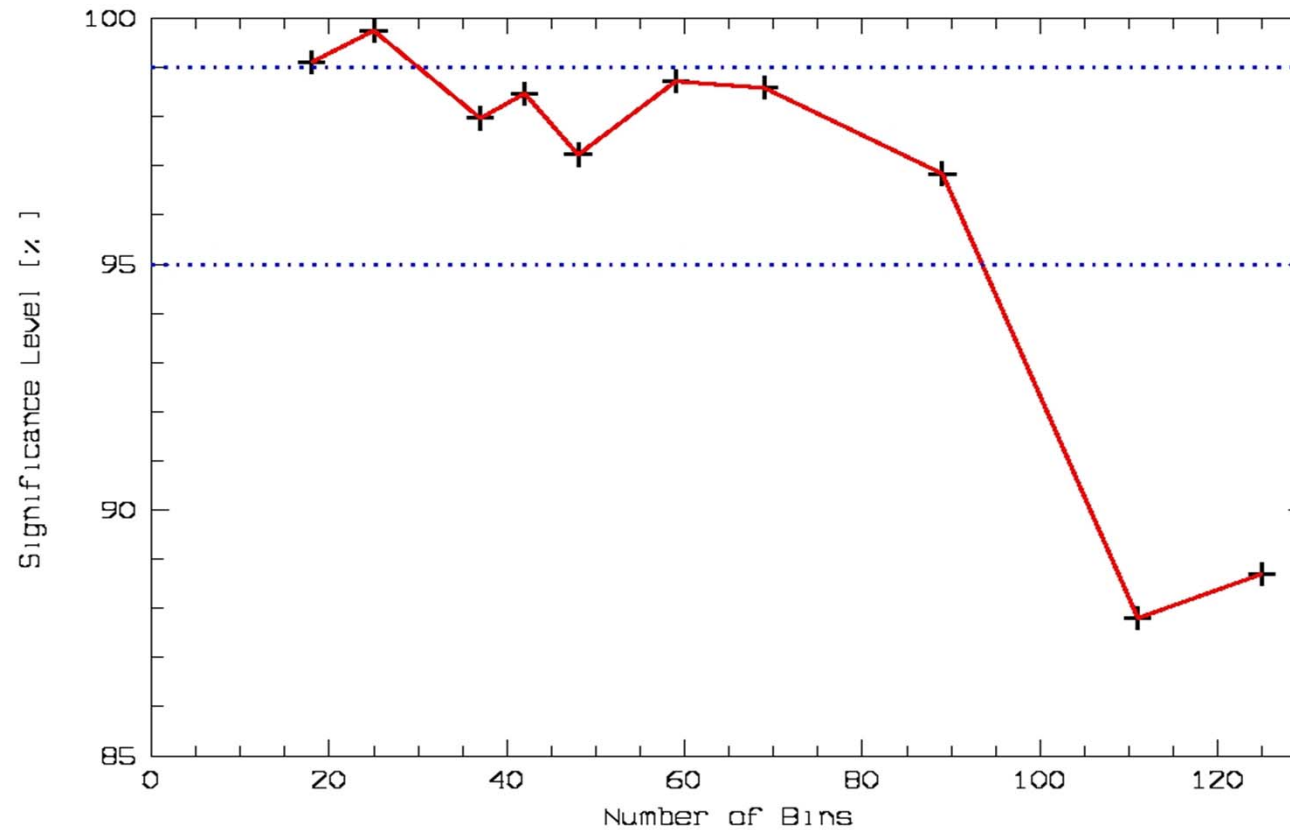
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Hypothesis Test

99% confidence level for hypothesis rejection



Real Example



F-Test

- literature: J. Kaastra in X-ray spectroscopy in astrophysics, 1999, p 276
- F is defined as:
 - $F = \chi^2_1 \text{ d.o.f}_2 / \chi^2_2 \text{ d.o.f}_1$
 - $F = \chi^2_1(n-m_2) / \chi^2_2(n-m_1)$
 - where
 - n number of data points
 - m number of fitted parameters
- F is a **statistical quantity!**
 - For large n:
 - **r.m.s (F) = $2 / \sqrt{n}$**
 - **$F = 1 + (r_1+r_2)/n$**



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F-Test

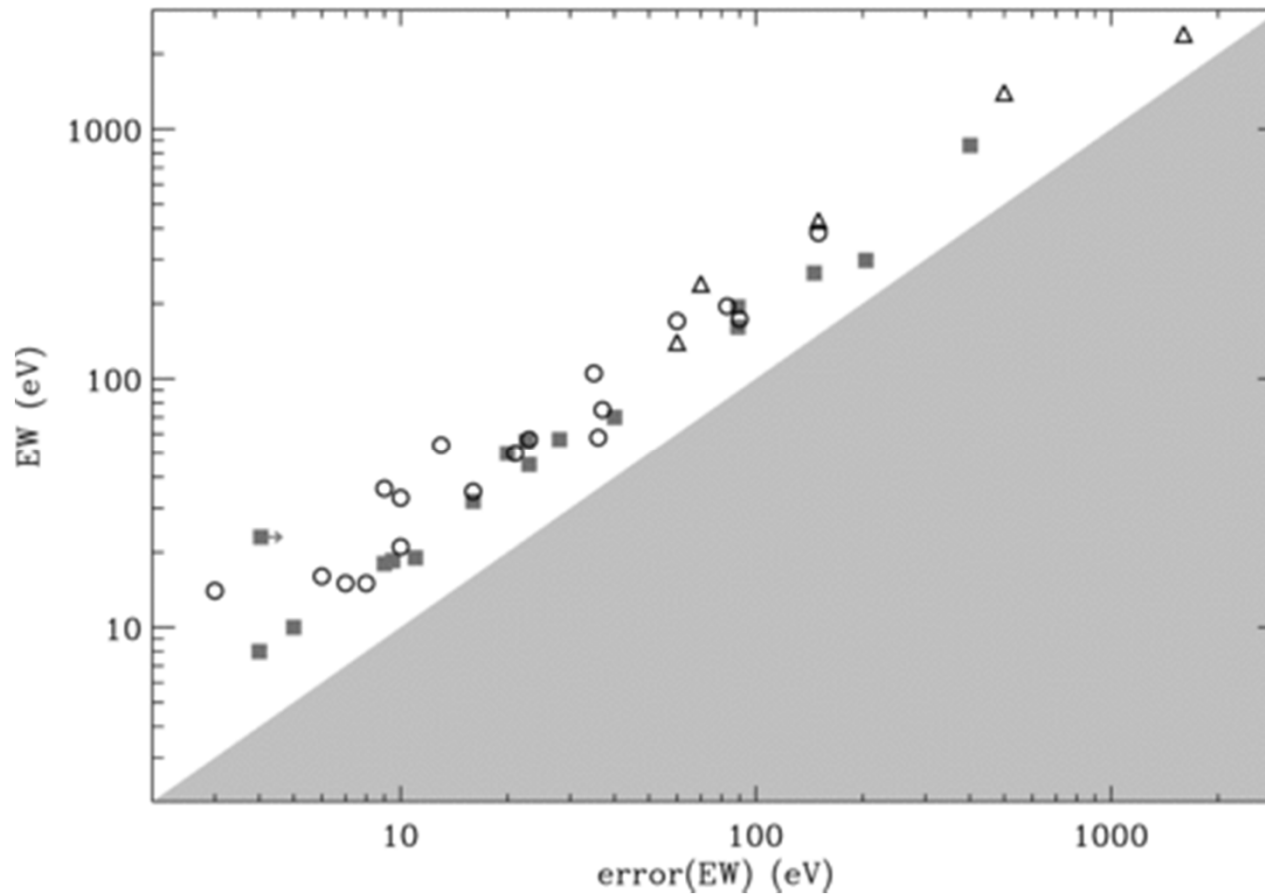
- **Example:**
 - $r_1 + r_2 = 15$ (independent from n)
 - $m_1=3$,
 - $m_2=6$,
- $n = 40$,
 - d.o.f.₁=37, $\chi^2_1 = 52$, d.o.f.₂=34 $\chi^2_2=34$
 - R.m.s.(F) = 0.3162
 - Probability of improvement:
 - $F \Rightarrow 99.951897\%$
 - $F+R.m.s(F) \Rightarrow 99.997780\%$
 - $F-R.M.S(F) \Rightarrow 98.051537\%$
- $n = 100$,
 - d.o.f.₁=97, $\chi^2_1 = 112$, d.o.f.₂=94 $\chi^2_2=94$
 - R.m.s.(F) = 0.2
 - Probability of improvement:
 - $F \Rightarrow 99.912766\%$
 - $F+R.m.s(F) \Rightarrow 99.999939\%$
 - $F-R.M.S(F) \Rightarrow \text{NO, model 1 is better than model 2}$



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The Large View



Vaughan, S.; Uttley, P., 2008, MNRAS 390, 421



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Golden Recipe Of The Statistical Analysis

1. golden Recipe of the statistical analysis does not exist.
2. every scientific question is a new statistical challenge and often requires a new concepts
3. In general:
 1. never over-sample a spectra
 2. a bin-width $< 1/3$ FWHM does not add any spectral information, but increases the d.o.f.
 3. if possible bin with constant signal/noise
4. every spectral analysis requires a new optimal solution depending on the analyzed spectra and on the specific scientific question



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