



Test & Analysis Project

Statistical Testing Physics Testing

http://www.ge.infn.it/geant4/analysis/TandA





Maria Grazia Pia, INFN Genova on behalf of the T&A team

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Test & Analysis Project

Test & Analysis is a project to develop a statistical analysis system for usage in Geant4 testing

- Main application areas •
- physics validation
 regression testing
 system testing
- Provide tools to compare Geant4 simulation results with reference data
 - equivalent reference distributions (for instance, regression testing)
 - experimental measurements
 - data libraries from reference distribution sources
 - functions deriving from theoretical calculations or from fits

Users



- provide and document requirements
- provide feedback on prototypes
- perform beta testing
- provide use cases for acceptance testing



- provide and document requirements
 perform formal acceptance testing

Other potential users:

users of the Geant4 Toolkit, to verify the results of their applications with respect to reference data or their own experimental results

+ users of the standalone Statistical Toolkit

History

- "Statistical testing" agreed as a collaboration-wide goal 2001-2002
- Initial ideas for this project presented at a TSB meeting, ~end 2001
- Informal discussions, spring summer 2002
- Test & Analysis Project launched at Geant4 Workshop 2002





Provide statistics algorithms to compare various kinds of distributions *i.e. binned, unbinned, continuous, multi-dimensional, affected by experimental errors, ...* Maria Grazia Pia, INFN Genova Provide distributions of physical quantities of interest, to be compared to reference ones

Developers







- Pablo Cirrone (INFN)
- Stefania Donadio (INFN)
- Susanna Guatelli (INFN)
- Alfonso Mantero (INFN)
- Barbara Mascialino (INFN)
- Luciano Pandola (INFN)
- Sandra Parlati (INFN)
- Andreas Pfeiffer (CERN)
- MG Pia (INFN)
- Alberto Ribon (CERN)
- Simona Saliceti (INFN)
- Paolo Viarengo (IST)

HEPstatistics Team

- S. Donadio
- F. Fabozzi
- S. Guatelli
- L. Lista
- B. Mascialino
- A. Pfeiffer
- MG Pia
- A. Ribon
- P. Viarengo
- discussions with Fred James, Louis Lyons, Giovanni Punzi

Collaboration: G. Cosmo, V. Ivanchenko, M. Maire, S. Sadilov, L. Urban Production resources: Gran Sasso Laboratory

One year later...

Statistics

- Scope
- Architecture
- Software process
- Statistical algorithms
- Current status
- HEPstatistics

Physics

- Scope
- Physics tests
- Results
- Resources needed

This workshop: discuss results, open problems discuss integration in regular Geant4 testing agree on future plans synergy with Geant4 Physics Book

Statistical Testing Project

GoF component of HEPstatistics

What is?

A project to develop a **Statistical comparison system**

Provide tools for the statistical comparison of distributions

- equivalent reference distributions
- experimental measurements
- data from reference sources
- functions deriving from theoretical calculations or fits

Main application areas:

Physics analysis Simulation validation Detector monitoring Regression testing <u>Reconstruction vs. expectation</u>

Vision: the basics



- Motivated by Geant4
- First core of a statistics toolkit for HEP





Who are the developers?

Rigorous software process

Software quality

• Build on a solid architecture

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Flexible, extensible, maintainable system

User Requirements Document

http://www.ge.infn.it/geant4/analysis/HEPstatistics/

Specific requirements: capability requirements

Comparing distributions (χ^2 , KSG, KS, AD, CVM, Kuiper, Lilliefors, list of histograms, select the proper algorithm)

Converting distributions (binned ⇔ unbinned)

Confidence levels

Handling distributions (mono/many dim. distributions , reduce dimensionality, filters, toy MonteCarlo)

Treatment of errors (handle statistical and systematic experimental errors)

Plotting (original, normalised, cumulative distributions)

Architectural guidelines

- The project adopts a **solid architectural approach**
 - to offer the *functionality* and the *quality* needed by the users
 - to be *maintainable* over a large time scale
 - to be *extensible*, to accommodate future evolutions of the requirements
- Component-based approach
 - to facilitate re-use and integration in diverse frameworks
- AIDA
 - adopt a (HEP) standard
 - no dependence on any specific analysis tool
- Python
 - for interactivity
- The approach adopted is compatible with the recommendations of the LCG Architecture Blueprint RTAG

Software process guidelines

• USDP, specifically tailored to the project

- practical guidance and tools from the RUP
- both rigorous and lightweight
- mapping onto ISO 15504

Guidance from ISO 15504

- standard!
- Incremental and iterative life cycle model
- Various software process artifacts available on the web
 - Vision
 - User Requirements
 - Architecture and Design model
 - Traceability matrix
 - etc.

Historical introduction to EDF tests

- In 1933 Kolmogorov published a short but landmark paper on the Italian *Giornale dell'Istituto degli Attuari*. He formally defined the empirical distribution function (EDF) and then enquired how close this would be to the true distribution F(x) when this is continuous.
- It must be noticed that Kolmogorov himself regarded his paper as the solution of an interesting probability problem, following the general interest of the time, rather than a paper on statistical methodology.
- After Kolmogorov article, over a period of about 10 years, the foundations were laid by a number of distinguished mathematicians of methods of testing fit to a distribution based on the EDF (Smirnov, Cramer, Von Mises, Anderson, Darling, ...).
- The ideas in this paper have formed a platform for vast literature, both of interesting and important probability problems, and also concerning methods of using the Kolmogorov statistics for testing fit to a distribution. The literature continues with great strength today showing no sign to diminish.

Goodness-of-fit tests

- Pearson's χ² test
- Kolmogorov test
- Kolmogorov Smirnov test
- Goodman approximation of KS test
- Lilliefors test
- Fisz-Cramer-von Mises test
- Cramer-von Mises test
- Anderson-Darling test
- Kuiper test

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It is a difficult domain...

Implementing algorithms is easy But comparing real-life distributions is not easy

Incremental and iterative software process Collaboration with statistics experts

Patience, humility, time...

System open to extension and evolution

Suggestions welcome!

Statistical test

ComparisonAlgorithm<BinnedDistribution>



ComparisonAlgorithm<UnbinnedDistribution>

Distributions



Statistical comparison: algorithms



Simple user layer

- Shields the user from the complexity of the underlying algorithms and design
- Only deal with AIDA objects and choice of comparison algorithm







http://www.ge.infn.it/geant4/analysis/HEPstatistics/

Requirements
Design
Implementation
Test & test results
Documentation (coming...)



- Applies to **binned** distributions
- It can be useful also in case of unbinned distributions, but the data must be grouped into classes
- Cannot be applied if the counting of the theoretical frequencies in each class is < 5
- When this is not the case, one could try to unify contiguous classes until the minimum theoretical frequency is reached

Kolmogorov test

- The easiest among non-parametric tests
- Verify the adaptation of a sample coming from a random **continuous** variable
- Based on the computation of the maximum distance between an empirical repartition function and the theoretical repartition one
- Test statistics:

$D = \sup | F_O(x) - F_T(x)|$





Kolmogorov-Smirnov test

- Problem of the two samples
 - mathematically similar to Kolmogorov's
- Instead of comparing an empirical distribution with a theoretical one, try to find the maximum difference between the distributions of the two samples F_n and G_m:

 $D_{mn} = \sup |F_n(x) - G_m(x)|$

- Can be applied only to **continuous** random variables
- *Conover (1971)* and *Gibbons and Chakraborti (1992)* tried to extend it to cases of discrete random variables

Goodman approximation of K-S test

• Goodman (1954) demonstrated that the Kolmogorov-Smirnov exact test statistics

 $D_{mn} = \sup |F_n(x) - G_m(x)|$

can be easily converted into a χ^2 :

 $\chi^2 = 4D_{mn}^2 [m^*n / (m+n)]$

- This approximated test statistics follows the χ^2 distribution with 2 degrees of freedom
- Can be applied only to continuous random variables

Lilliefors test

• Similar to Kolmogorov test

• Based on the null hypothesis that the random continuous variable is normally distributed N(m, σ^2), with m and σ^2 unknown

• Performed comparing the empirical repartition function $F(z_1, z_2, ..., z_n)$ with the one of the standardized normal distribution $\Phi(z)$:

 $D^* = \sup |F_O(z) - \Phi(z)|$

Fisz-Cramer-von Mises test

- Problem of the two samples
- The test statistics contains a weight function
- Based on the test statistics:

 $t = n_{1*}n_2 / (n_1+n_2)^2 \Sigma_i [F_1(x_i) - F_2(x_i)]^2$

- Can be performed on **binned** variables
- Satisfactory for symmetric and right-skewed distribution

Cramer-von Mises test

- Based on the test statistics:
 - ω^2 = integral (F_O(x) F_T(x))² dF(x)
- The test statistics contains a weight function
- Can be performed on **unbinned** variables
- Satisfactory for symmetric and right-skewed distributions

Anderson-Darling test

• Performed on the test statistics:

A²= integral { $[F_{O}(x) - F_{T}(x)]^{2} / [F_{T}(x) (1 - F_{T}(X))]$ dF_T(x)

- Can be performed both on **binned** and **unbinned** variables
- The test statistics contains a weight function
- Seems to be suitable to any data-set (Aksenov and Savageau 2002) with any skewness (symmetric distributions, left or right skewed)
- Seems to be sensitive to **fat tail of distributions**

Kuiper test

- Based on a quantity that remains invariant for any shift or re-parameterisation
- Does not work well on tails

$D^* = \max (F_O(x)-F_T(x)) + \max (F_T(x)-F_O(x))$

 It is useful for observation on a circle, because the value of D* does not depend on the choice of the origin. Of course, D* can also be used for data on a line

Power of the tests

In terms of power:

The power of a test is the probability of rejecting the null hypothesis correctly



• χ^2 loses information in a test for continuous distribution by grouping the data into cells

→ Kac, Kiefer and Wolfowitz (1955) showed that D requires $n^{4/5}$ observations compared to n observations for χ^2 to attain the same power

• Cramer-von Mises and Anderson-Darling statistics are expected to be superior to D, since they make a comparison of the two distributions all along the range of x, rather than looking for a marked difference at one point

χ^2 design

UR 1.1 The user shall be able to compare binned distributions by means of χ^2 test.



Unit test: χ^2 (1)

EXAMPLE FROM PICCOLO BOOK (*STATISTICS* - page 711)

The study concerns monthly birth and death distributions (binned data)

 χ^2 test-statistics = 15.8

Expected $\chi^2 = 15.8$

Exact p-value=0.200758 Expected p-value=0.200757



Unit test: χ^2 (2)

EXAMPLE FROM CRAMER BOOK

(MATHEMATICAL METHODS OF STATISTICS - page 447)

The study concerns the sex distribution of children born in Sweden in 1935



ADB design

UR 1.2 The user shall be able to compare binned distributions by means of Anderson-Darling test.

WORK IN PROGRESS



KSG Design

UR 1.3 The user shall be able to compare unbinned distributions by means of Kolmogorov-Smirnov Goodman test.



Unit test: K-S Goodman (1)

EXAMPLE FROM PICCOLO BOOK (STATISTICS - page 711)

The study concerns monthly birth and death distributions (unbinned data)

 χ^2 test-statistics = 3.9

Expected $\chi^2 = 3.9$

Exact p-value=0.140974 Expected p-value=0.140991



Unit test: K-S Goodman (2)

EXAMPLE FROM LANDENNA BOOK (NONPARAMETRIC TESTS BASED ON FREQUENCIES - page 287)

We consider body lengths of two independent groups of anopheles



KS Design

UR 1.4 The user shall be able to compare unbinned distributions by means of Kolmogorov-Smirnov test.





Unit test: Kolmogorov-Smirnov(1)

EXAMPLE FROM http://www.physics.csbsju.edu/stats/KS-test.html The study concerns how long a bee stays near a particular tree (Redwell/Whitney)



Unit test: Kolmogorov-Smirnov (2)

EXAMPLE FROM LANDENNA BOOK (NONPARAMETRIC STATISTICAL METHODS - page 318-325)

We consider one clinical parameter of two independent groups of patients



ADU Design

UR 1.5 The user shall be able to compare unbinned distributions by means of Anderson-Darling test.



CVMB Design

UR 1.6 The user shall be able to compare binned distributions by means of Cramer-Von Mises test.





CVMU Design

UR 1.7 The user shall be able to compare unbinned distributions by means of Cramer-Von Mises test.



...and more

- No time to illustrate all the algorithms and statistics details...
- Other components in progress, not only GoF
 - PDF
 - Toy Monte Carlo
 - (L. Lista et al., INFN Napoli, BaBar)
- A general purpose, open source tool for statistical data analysis
 - interest in HEP community: LCG, BaBar, CDF etc.
 - 2 talks at PHYSTAT 2003, SLAC, 8-11 September 2003

Do we need such sophisticated algorithms?

ESA test beam at Bessy, Bepi Colombo mission



χ^2 not appropriate

(< 5 entries in some bins, physical information would be lost if rebinned)



Anderson-Darling $A_c (95\%) = 0.752$

Status and plans

- β-release March 2003
- 1st (preliminary) release June 2003
 - basic algorithms
 - unit and system tests
- Recent developments
 - added new algorithms, improved design
- Work in progress
 - filtering
 - treatment of errors (uncertainties)
- In preparation
 - improved documentation
 - user examples (now use system tests as preliminary examples)

Conclusions

A lot of progress since last year's workshop...

– a lot of work, but also a lot of fun!

A group of bright, enthusiastic, hard-working young collaborators...

Ground for Geant4 Physics Book

Physics Testing Project

see talk on Validation, Thursday plenary



http://www.ass-mic.org/2003

Nuclear Science Symposium

Medical Imaging Conference

13th International Workshop on Room-Temperature Semiconductor X- and Gamma-Ray Detectors



Radiation Detectors and Electronics: Applications in Physics, Industry, Space, Biology, Genetics and Medicine Physics, Engineering and Mathematical Aspects of Medical Imaging



Abstract Submission Deadlines: NS5 & MIC: May 16, 2003 RTSD Workshop: June 27, 2003

October 19-25, 2003 Portland, Oregon, USA DoubleTree Hotel – Hayden Island Complex

More at IEEE-NSS,

Portland, 19-25 October 2003

B. Mascialino et al.,

A Toolkit for statistical data analysis

L. Pandola et al.,

Precision validation of Geant4 electromagnetic physics