

# Advanced Methods for Modelling Physical Phenomena: from data gathering & analysis to hypothesis testing and parameter estimation

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## Part 1. General approaches and considerations

### Section 1. Experimental data and random variables: representation and statistics

1. Experimental and theoretical cycles and developments in physical studies
  - a. Observation
  - b. Experiment
  - c. Model
2. Representation of experimental data and statistics
  - a. Mean and standard deviation
  - b. Empirical distribution
  - c. Histogram
3. Probabilistic grounds for experimental data analysis
  - a. Random variable
  - b. Probability distribution
  - c. Moments
4. Probability distribution functions
  - a. PMF
  - b. PDF
  - c. CDF
5. Statistics of random variable
  - a. Expectation and Variance
  - b. Co-variance and Correlation
  - c. Characteristic and generating functions
6. Stochastic process, random functions, waveforms and statistics
  - a. From random variable to random function
  - b. Averaging, ergodicity, stationarity
  - c. Transient processes
  - d. Nonlinear processes
7. Examples of stochastic processes
  - a. Poisson
  - b. Markov
  - c. Cox

## Section 2. Analytical modelling of stochastic processes and phenomena

8. Challenges, drawbacks and benefits of analytical modeling
  - a. High risk – high reward area
  - b. Analytics vs Monte Carlo
9. How to build a model of a sequence of stochastic processes
10. Distribution of sum of independent identically distributed random variables
  - a. Moments of the sum
  - b. PMF and PDF of the sum
  - c. Random sum moments: Burgess variance theorem
  - d. Random sum distribution: Laplace transform
11. Distribution of ordered random variables: order statistics approach
  - a. Erlang distribution of arrival times of stationary Poisson process
  - b. Order statistics of time-varying arrival times

## Section 3. Monte Carlo modelling of stochastic processes and phenomena

12. Challenges, drawbacks and benefits of Monte Carlo modeling
13. Verification of MC modelling results

## Section 4. Analysis of experimental and modelling results

14. Measurement errors
  - a. Statistical
  - b. Systematic
  - c. Error propagation
15. Cramer-Rao lower bound on parameter estimation
  - a. Fisher information integral
  - b. Cramer-Rao lower bound
  - c. Random number of events challenge
16. Empirical distribution function vs histogram: in-depth comparison
  - a. Histogram binning challenge and its optimization
  - b. Smirnov-Kolmogorov test of distance between distributions
  - c. Benefits of empirical distribution function approach
17. Approaches to hypothesis testing and parameter estimation
  - d. Least squares fitting
  - e. Chi squared test
  - f. Maximum Likelihood approach
  - g. Maximum Likelihood and “goodness-of-fit”
  - h. Smirnov-Kolmogorov test of “goodness-of-fit”

## Part 2. Probabilistic modelling and analysis of photon-number-resolving detection and Silicon Photomultipliers

### Section 1. Introduction to photon-number-resolving detection and SiPM

1. Physics, concepts, and designs of SiPM
2. Specific features of SiPM: correlated noise, nonlinearity

3. Measurement and characterization of photon-number-resolving detectors
4. Application areas and large projects with SiPM
5. State-of-art, challenges, and perspectives of SiPM technology

## Section 2. Probabilistic modelling of SiPM response and performance in photon number resolution

6. How probability distributions of signal and noise are transformed from input to output
  - a. Photo-conversion (binomial)
  - b. Multiplication (excess noise)
  - c. Dark counts (Poisson)
  - d. Correlated events (crosstalk, afterpulsing)
7. Modelling of probability distributions of correlated events
  - a. Geometric chain model
  - b. Compound Poisson distribution
  - c. Branching Poisson process model
  - d. Generalized Poisson distribution
8. Analysis of experiments using Generalized Poisson distribution model of SiPM response
  - a. Results of S. Vinogradov' studies
  - b. Results of FBK and DESY studies
  - c. Results of CERN CMS studies
  - d. Application of the model to radiation-degraded SiPMs

## Section 3. Probabilistic modelling of SiPM timing and performance in time resolution

9. How probability distribution of signal and noise event times are transformed by photon detection processes
  - a. Poisson point process
  - b. Super-Poisson (bunched) process affected by SiPM correlated events
  - c. Sub-Poisson (anti-bunched) process affected by SiPM pixel recovery time
10. Analysis of interarrival time distribution for dark and correlated events
  - a. Histogram approach
  - b. Empirical Complimentary CDF approach
11. Single-photon time resolution measurement using multi-photon transit time histogram
12. Calibration of gain of noisy low gain photomultipliers using transit time histogram
13. Modelling of SiPM time resolution
  - a. Filtered marked point process approach
  - b. Scintillation pulse detection (TOF PET)
  - c. Laser / LED pulse detection (LIDAR)
  - d. Contributions of baseline fluctuations due to electronic noise and DCR

## Section 4. Probabilistic modelling of SiPM nonlinearity and saturation

14. General consideration of nonlinearity as a source of random losses and noises
15. Binomial nonlinearity – detection of short light pulses
  - a. Conventional model: balls (Poisson photons) in bins (pixels)
  - b. Adjustments to crosstalk
  - c. Adjustments to recovery
16. Recovery nonlinearity – detection of long light pulses

- a. Conventional model: non-paralizable counting with dead time
- b. Advanced model of exponential recovery process
- c. Advanced model of Markov reward-renewal process
- d. Recovery nonlinearity – detection of long light pulses

17. Excess noise factor of nonlinearity

## Section 5. Analysis of applicability and competitiveness of photodetectors

18. Total ENF and Detective Quantum Efficiency as FOM for detector performance

- a. Noise-affected lower bound
- b. Nonlinearity-affected upper bound
- c. Optimal operating range
- d. Dynamic range and detective range

19. Analysis of performance of modern photodetectors and SiPMs

- a. PIN
- b. PMT
- c. APD
- d. SiPM
- e. SiPMs of various designs and vendors

20. Application analysis of SiPM competitiveness and potential

- a. TOF PET
- b. LIDAR

# Big Data Analytics: tools & infrastructure

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## Lezione 1 (4 ore in tutto, 2+2)

- i) Introduzione ai BigData. Definizione ed esempi pratici sia in campo industriale che delle fisica
- ii) Tecniche di gestione dei dati in regime BigData
- iii ) Maggiori framework conosciuti (Hadoop, Hive, Spark)
- iv) Applicazione a un dataset reale in ambiente locale

## Lezione 2 (4 ore in tutto, 2+2)

- i) Approccio distribuito e cloud Computing. Accenno a gestione dei dati attraverso database non relazionali.
- ii) Servizi offerti dalle piattaforme commerciali (Google ad Amazon) per i problemi di BigData e di Machine Learning.
- iii) Architetture e servizi cloud per data preparation e data cleaning.
- iv) Esempi di applicazioni. Il caso di Google DataPrep

## Lezione 3 (4 ore in tutto, 2+2)

- i) Dalla data preparation al data mining e analytics. Il caso di Google BigQuery e Google Analytics. Applicazioni in ambito di Astrofisica delle alte energie.
- ii) Calcolo Distribuito con Amazon. Il caso di Hubble Space Telescope. Analizzare un catalogo di una missione spaziale in 15 minuti.

## Lezione 4 (4 ore in tutto)

- i) Machine Learning. Accenni e prime proprietà
- ii) Il problema della classificazione. Classificatori lineari Bayesiani e a massima verosimiglianza. Classificatori basati sulla distanza e su vettori di supporto (Support Vector Machine)
- iii) Alberi di decisione e random forests. Adaboost.
- iv) Introduzione al concetto di rete neurale
- v) I tools messi a disposizione dalle piattaforme Cloud commerciali per Machine-Learning-off the shelf.