

Young Physicists' Forum UZH, April 22, 2017



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Overview

Part I

What are we trying to measure ? What do our experiments look like ?

Part II "Online" event selection

Part III "Offline" analysis

April 22, 2017 Event Selection and Data Analysis in Particle Physics

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The "Standard Model" of Particle Physics describes elementary particles and their interactions through three of the four known fundamental forces

> electromagnetic interaction weak interaction strong interaction

Formulated in the 1960's, it is a Quantum Field Theory based on

relativistic kinematics quantum mechanics group theory

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Describes many observed phenomena with stunning precision

e.g. electric dipole moment of the electron $(g^{SM}-2)/2 = 0.00115965218164(76)$ $(g^{exp}-2)/2 = 0.00115965218073(28)$

Many of it's predictions have been confirmed by experiments

e.g. observation of Higgs Boson at CERN, 40 years after its existence was predicted to accommodate mass of elementary particles

Does not explain some fundamental observations, e.g.

Matter – Antimatter asymmetry in the Universe Dark Matter Does not incorporate gravity

Holy grail of particle physics today: Search for phenomena "Beyond the Standard Model"

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Two approaches:

"Direct Searches" New elementary particles

"Indirect Searches"

Confront precise predictions with precise measurements

In both cases, studying very rare phenomena:

Need effective and efficient selection algorithms to separate signal from large backgrounds

Need to handle huge data volumes

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Elementary particles

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Higgs Boson \rightarrow mass

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Three "generations" of matter particles 2nd and 3rd generation are heavier siblings of 1st generation

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Fundamental Questions



WHY THREE GENERATIONS ?

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For each matter particle there is a corresponding antiparticle

• Same mass as the particle, but opposite charge

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Fundamental Questions



WHY DO WE SEE SO MUCH MORE MATTER THAN ANTIMATTER IN THE UNIVERSE ?



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Fundamental Questions



WHY DOES ANYTHING EXIST AT ALL ?



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Quarks are not observed as free particles

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Particles that are observed consist of

- three quarks (e.g. proton), or
 - a quark and an antiquark

("exotic" combinations: Tetraquarks, Pentaquarks)

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Many possible combinations: "particle zoo"

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If I could remember the names of all these particles, I would have been a botanist.

(Enrico Fermi)

izquotes.com

Most particles in the particle zoo are very short-lived

Very few are stable or live long enough to leave a trace in a particle detector

electrons and muons

protons (*uud*) pions (*ud*) and kaons (*us*)

> photons neutrons (*udd*)

... and their antiparticles

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charged

neutral

Short-lived particles can be reconstructed indirectly by measuring their long-lived decay products

Relativistic kinematics

$$\mathbf{E} = \mathbf{m} \cdot \mathbf{c}^2$$

Short-lived particles can be reconstructed indirectly by measuring their long-lived decay products

Relativistic kinematics



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Relativistic kinematics

$$\mathbf{E}^2 = \mathbf{m}^2 + \mathbf{p}^2$$

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Relativistic kinematics

$$m^2 = \boldsymbol{E}^2 - \boldsymbol{p}^2$$

Short-lived particles can be reconstructed indirectly by measuring their long-lived decay products

Relativistic kinematics

$$m^2 = E^2 - p^2$$

Energy and momentum conservation in the decay

$$\left(\boldsymbol{M}^{2} = \left(\sum_{i} \boldsymbol{E}_{i}\right)^{2} - \left|\sum_{i} \boldsymbol{\vec{p}}_{i}\right|^{2}\right)$$

with

$$E_i^2 = m_i^2 + p_i^2$$

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Discovery of \Upsilon Particles in 1977



OBSERVATION OF A DIMUON RESONANCE AT 9.5 GeV IN 400 GeV PROTON-NUCLEUS COLLISIONS

S. W. Herb, D. C. Hom, L. M. Lederman, J. C. Sens, H. D. Snyder, and J. K. Yoh Columbia University, New York, New York 10027

and

J. A. Appel, B. C. Brown, C. N. Brown W. R. Innes, K. Ueno, and T. Yamanouchi Fermi National Accelerator Laboratory, Batavia, Illinois 60510

and

A. S. Ito, H. Jöstlein, D. M. Kaplan, and R. D. Kephart State University of New York at Stony Brook Stony Brook, New York 11794

July 1977



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Discovery of Y Particles in 1977



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Yesterday's sensation is today's calibration channel

(Richard P. Feynman)



... Today's Calibration Channel



Distribution should peak at the known mass values: Verify / calibrate your momentum reconstruction

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... Today's Calibration Channel



Width of the peaks measures precision of the momentum reconstruction

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... Today's Calibration Channel



Width of the peaks measures precision of the momentum reconstruction

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(2013) 052004]

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Phys.Rev.D

(ATLAS Collaboration)

G.Aad et al

To reconstruct an event, need to:

- Measure the flight directions of long-lived particles
 - Measure the magnitudes of their momenta
 - Determine which type of particle they are (to know their mass and energy)

Reconstructing an Event

To reconstruct an event, need to:

- Measure the flight directions of long-lived particles
 - Measure the magnitudes of their momenta

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Measuring Momentum

Apply a magnetic field \rightarrow Lorentz force on moving particle

$$\vec{F}_L = \boldsymbol{q} \cdot \vec{\boldsymbol{v}} \times \vec{\boldsymbol{B}}$$



Measuring Momentum

Apply a magnetic field \rightarrow Lorentz force on moving particle

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Particle forced onto a circular trajectory

$$\frac{\mathbf{m}\cdot\mathbf{v}^2}{\mathbf{r}} = \mathbf{q}\cdot\mathbf{v}\cdot\mathbf{B} \quad \Rightarrow \quad \mathbf{p} = \mathbf{q}\cdot\mathbf{B}\cdot\mathbf{r}$$

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Precision of the momentum measurement

$$\frac{\Delta p}{p} \propto \frac{\sigma_x}{\sqrt{N}} + \frac{p}{B \cdot L^2}$$
precision & number length over which
of position measurements the trajectory is measured
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Ap

Large Hadron Collider



27 km long ring in a tunnel 100 m below ground Two proton beams: clockwise and anti-clockwise → proton collisions every 25 ns (40 million times per second) Four collision points: four large experiments → more than 10'000 physicists, more than 100 nationalities April 22, 2017 Event Selection and Data Analysis in Particle Physics 0. St



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Readout



particle deposits energy in detector → voltage pulse

amplification

storage, digitization reconstruction, analysis

Millions of detector channels \rightarrow large amount of data

ATLAS / CMS: 1-2 MB per event

LHCb: about 100 kB per event

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1.5 MB per event × 40 million events per second

≈ 60 TB / sec

> 10'000 DVDs every second
 ≈ 6 billion phone calls
 Facebook: ≈ 600 TB / day

 \rightarrow Can't afford to store everything !



(1 PB = 1000 TB)

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(1 PB = 1000 TB)

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(1 PB = 1000 TB)

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But: we are interested in studying rare processes !

e.g. On average about one Higgs boson / sec



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But: we are interested in studying rare processes !

e.g. On average about one Higgs boson / sec

Some more events are useful as "today's calibration channel"

> But most events are just "today's background"



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But: we are interested in studying rare processes !

e.g. On average about one Higgs boson / sec

Some more events are useful as "today's calibration channel"

But most events are just "today's background"

 \rightarrow No need to store everything !



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plot: <http://www.hep.</pre>

Trigger Requirements

Needs to have high efficiency for selecting the interesting events

Obviously: want to minimize losses in statistics

Also: losses can cause biases on physics results

Needs to provide <u>high rejection</u> for uninteresting high-rate processes Determined by how much data you can afford to store

Needs to be <u>flexible</u>

Operating conditions can change

Want to be able to implement new ideas

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Trigger Requirements

Needs to be **fast**

Data need to be stored temporarily while the decision is made: the more time the algorithm needs to come to a decision, the more events need to be stored in parallel

 \rightarrow cost of storage space

Events come in at a constant rate of 40 MHz: the more time the algorithm needs to come to a decision, the more copies of the algorithm have to be executed in parallel

 \rightarrow cost of computing

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Trigger Levels



Low-level triggers:

Full input rate \rightarrow fast, simple, crude decisions Usually implemented in custom-made hardware

Higher-level triggers:

Reduced input rate \rightarrow more time for more sophisticated decisions

Usually implemented in software running on large CPU farms

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Trigger Levels



example: LHCb | CMS

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Low-Level Triggers

Need to be fast and simple

The particles we are interest in studying are usually heavy ATLAS/CMS: Higgs boson 125 GeV = $125 \times$ the mass of a proton LHCb: particles containing *b* quarks ≈ 5 GeV = $5 \times$ the mass of a proton

→ Their decay products tend to have large energy and momenta

Typical signatures to look for: Large energy deposits in the calorimeters Muons with large momentum

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Typical signatures to look for: Large energy deposits in the calorimeters Muons with large momentum
LHCb "L0 Muon" Trigger



Muons are easy to identify: put lots of material,

whatever you see in a detector behind this material must be a muon

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LHCb "L0 Muon" Trigger





LHCb muon system:

Five detection layers, separated by 80 cm thick walls of iron

First-level muon trigger:

Require coincidence of hits in several detection layers,

pointing back to the pp collision point

 \rightarrow Small deviation in magnetic field \rightarrow Muon with high momentum

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LHCb "L0 Muon" Trigger





Algorithm implemented in custom-designed electronics boards, using FPGA and RAM chips and high-speed data links

Start with a hit in station M3

Extrapolate straight line to M2, M4, M5

Define search windows around the extrapolated position

If matching hits are found, look up the corresponding momentum (stored in large look-up tables in RAM chips)

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Usually a simplified version of the offline reconstruction software, running on a large computer farm, close to the experiment

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LHCb: 58'000 CPU cores 40'000 reconstruction jobs running in parallel 400 different trigger selections

for different physics analyses

LHCb has so far published 370 physics papers, with many more to come



Usually a simplified version of the offline reconstruction software, running on a large computer farm, close to the experiment

Simplifications necessary to meet CPU time constraints, but result in poorer resolution

Usually a simplified version of the offline reconstruction software, running on a large computer farm, close to the experiment Simplifications necessary to meet CPU time constraints, but result in poorer resolution



≈ factor 2 improvement due to better calibration

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LHCb High-Level Trigger



LHC is colliding protons only \approx 35 % of the time

Inject fresh proton beams, maintenance work, technical problems, ...

LHCb High-Level Trigger



LHC is colliding protons only ≈ 35 % of the time Inject fresh proton beams, maintenance work, technical problems, ...

Split High-Level Trigger into two parts

Buffer events that pass a first selection on local disks

Use CPU farm to perform final selection when there are no collisions and no new data is coming in

Calibration with "offline quality" before final selection

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10'000 TByte disk space

→ buffer data up to a week

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After trigger selections, experiments are saving of the order of 1'000 events per second for offline analysis

Reconstructing one event takes about 10-20 sec

Might need to run the reconstruction more than once, if improved reconstruction algorithms become available

In addition: need to generate huge numbers of simulated events to study reconstruction and selection efficiencies etc.

Generating one simulated event can take up to 15 minutes

In addition: physics analyses

usually fast, but many users, many analysis jobs

 \rightarrow need tens of thousands of processors

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After trigger selections, experiments are saving of the order of 1'000 events per second for offline analysis

≈ 10¹⁰ events per year × 1.5 MB per event

 \rightarrow need storage space for \approx 15'000 TB of data per experiment per year



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Hierarchical structure

Tier 0: CERN + Wigner centre Budapest

- \rightarrow first copy of the raw data
- \rightarrow first pass reconstruction
- \rightarrow reprocessing during LHC down-times

Tier 1: Fourteen large computer centres

- \rightarrow fraction of raw and reconstructed data
- \rightarrow large-scale reprocessing

Tier 2: Large clusters at Scientific Institutions

- \rightarrow specific analysis tasks
- \rightarrow production/reconstruction of simulated events

Tier 3: Local clusters in a University Department

 \rightarrow what the user connects to





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travels \approx 1 cm in LHCb before it decays

Decay into $\mu^+\mu^-$ strongly suppressed in the Standard Model

Predicted "Branching Fraction" ≈ 3.5 × 10⁻⁹

(i.e. only about three out of a billion B_s^0 mesons should decay into $\mu^+\mu^-$)

But other theories (e.g. SuperSymmetry) predict this to be much larger



(also $B^0 \rightarrow \mu^+ \mu^-$, even more suppressed)

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Example: $B_s^0 \rightarrow \mu^+ \mu^-$



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signal

background

Use twelve discriminating variables

to distinguish between signal and background

(but do NOT use $m(\mu^+\mu^-)$ yet)

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At each step, determine the most discriminating cut for the given sample The same variable can be used several times

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Boosted Decision Tree

Advantage: exploits correlations between variables

Disadvantage: not stable

Small fluctuations in the data can make a big difference in the outcome

Solution: "Boosting"

Build many different trees (O(1000)), calculate weighted average over all trees

Automated algorithm:

Apply larger weights to misclassified events, build new Decision Tree on reweighted events

AdaBoost Algorithm

For Tree *m*

 $W_{m}(x_{i}) = \text{weight of event } x_{i}$ $\operatorname{err}_{m} = \sum_{x_{i} \text{ misclassified}} W_{m}(x_{i})$ $\alpha_{m} = \operatorname{const.} \times \ln\left(\frac{1 - \operatorname{err}_{m}}{\operatorname{err}_{m}}\right)$

Reweight misclassified events for next Tree

$$\boldsymbol{W}_{m+1}(\boldsymbol{x}_i) = \boldsymbol{W}_m(\boldsymbol{x}_i) \times \boldsymbol{e}^{\alpha_m}$$

Final classifier for event x_i

$$T(\mathbf{x}_i) = \sum_{m=1}^{N_{\text{trees}}} \alpha_m \times T_m(\mathbf{x}_i)$$

 $T_{m}(x_{i}) = \begin{cases} 1 & \text{if } x_{i} \text{ is classified as signal in Tree } m \\ 0 & \text{if } x_{i} \text{ is classified as background in Tree } m \end{cases}$

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Train BDT on simulated samples of signal and background events

(where we know for each event whether it is signal or background)

Recalibrate BDT response on control channels in data

 $B^0 \rightarrow K^+ \pi^-$ as proxy for signal

Events with $m(\mu^+\mu^-) > m(B_s^0)$ as proxy for background

Aim at flat response for signal, peaking at 0 for background



Fit two-dimensional distribution of BDT classifier and $m(\mu^+\mu^-)$ to extract the number of signal candidates LHCb 3fb 6000 $m_{\mu^+\mu^-}$ [MeV/ c^2] 5800 5600 B_s^0 mass 5400 region 5200 5000 0.2 0.8 0.60.4BDT

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Need to model the expected distribution in $m(\mu^+\mu^-)$

Background from random $\mu^+\mu^-$ combinations \rightarrow Exponential distribution

Background from wrongly reconstructed *B* meson decays \rightarrow from simulated events



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Need to model the expected distribution in $m(\mu^+\mu^-)$

Signal: expected position of the peak from large samples of reconstructed $B_s^0 \rightarrow K^+ K^-$ and similar decays



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Need to model the expected distribution in $m(\mu^+\mu^-)$

Signal: expected width of the peak from large samples of reconstructed $\Upsilon \to \mu^+ \mu^-$ decays and others

 $\Upsilon \rightarrow \mu^+ \mu^-$



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Yesterday's sensation is today's calibration channel

(Richard P. Feynman)

Have models for the expected signal and background distributions in the BDT classifier and in $m(\mu^+\mu^-)$

→ Maximum Likelihood fit to the measured distribution to estimate the number of signal events in our sample



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Hurray, find an excess of events around B_s^0 mass at high BDT



Hurray, find an excess of events around B_{s}^{0} mass at high BDT



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Two remaining questions:

Is the excess statistically significant?

i.e. how large is the probability that it could be caused by a random fluctuation in the distribution of background events ?

If the excess is "real", how large is the Branching Fraction ?

i.e. how does the extracted number of $B_s^0 \rightarrow \mu^+ \mu^-$ candidates translate into a probability for a B_s^0 to decay into $\mu^+ \mu^-$?

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what we Number of selected signal events have Number of *pp* collisions analysed х Probability that a pp collision produces a B^o X Probability that the B_s^0 decays to $\mu^+\mu^-$ X Probability that the μ^+ and the μ^- leave a trace in the detector X Efficiency of the trigger selection X **Efficiency of the reconstruction algorithms** X Efficiency of the offline selection criteria

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Select a reference decay mode for which the

Branching Fraction (decay probability) is large and well known from earlier measurements

Trigger, reconstruction and selection efficiencies are as similar as possible to those for $B_s^0 \rightarrow \mu^+ \mu^-$



Select a reference decay mode for which

The Branching Fraction (decay probability) is large and well known from earlier measurements

The trigger, reconstruction and selection efficiencies are as similar as possible to those for $B_s^{\ 0} \rightarrow \mu^+ \mu^-$

Actually use two reference decays

 $B^{*} \to J/\psi \: K^{*}$ with $J/\psi \to \mu^{*} \: \mu^{-}$ similar trigger but an additional particle

 $B^0 \rightarrow K^+ \pi^-$

same number of particles, but different trigger

The two give consistent results



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Maximum Likelihood fit:

Finds optimum values for the fit parameters such that the probability to obtain the observed distribution is maximized

Likelihood profile:

Change the values of the parameters from those found by the fit and re-calculate the probability



Sign up for my course "Datenanalyse" here at UZH in Fall semester



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BR ($B_s^0 \rightarrow \mu^+ \mu^-$) = 0 is excluded with 99.999999999997 % probability

Result of the measurement is in good agreement with the prediction from the Standard Model

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