Simulations in High Energy Physics

H. Jung (DESY)

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Simulations in High Energy Physics

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Simulation:

Oxford advanced dictionary: simulate = pretend to be

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Film Studios





Simulations in High Energy Physics

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Simulation: why?

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Simulations in High Energy Physics

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- Simulation:
 - Oxford advanced dictionary: simulate = pretend to be
- Simulation: why?
 - Can't we just calculate things ????
- Simulation: what?

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Detector response Particle decays ep, e^{*} e^{*}, pp interactions Economy Life

LITE

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Simulation: what?

Detector response
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ep, e* e*, pp interactions
Economy

Life

Simulation: How-to?

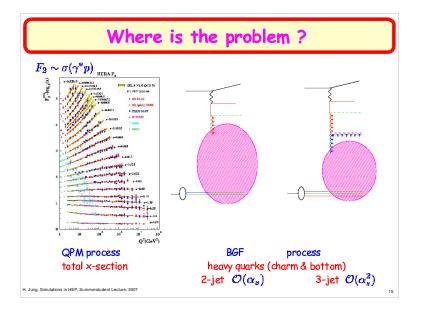
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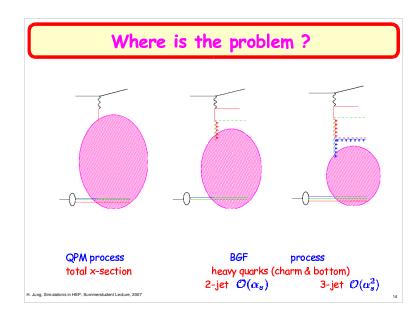


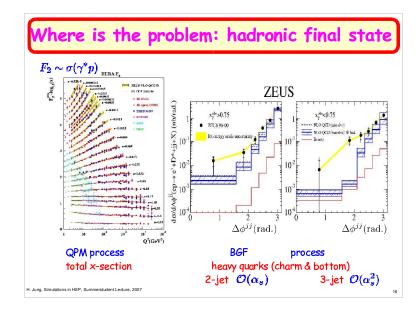
Application in Economy What is monte carlo simulation? montecarlo analysis? DECISIONEERING Six Sigma & DFSS Industries & Applications WHAT IS MONTE CARLO SIMULATION? What is Pisk Getting Started on Our Site What do we mean by "simulation? What is a Model? User Conference Without the aid of simulation, a spreadsheat model will only reveal a single outcome, generally the most likely or average scenario. Spreadsheat risk analysis uses both a spreadsheat model and simulation to automatically analyze the effect of varying inputs on outputs of the modeled system. Armiyosix of Results One type of spreadsheet simulation is **Monte Carlo simulation**, which randomly generates values for uncertain variables over and over to simulate a model. The random behavior in games of chance is similar to how Monte Carlo simulation selects variable values at random to simulate a model. When you roll simulation selects variable values at random to simulate a model. When you ro a die, you know that either a 1, 2, 3, 4, 5, or 6 will come up, but you don't know which for any particular roll. It's the same with the variables that have a known range of values but an uncertain value for any particular time or event (e.g., interest retes, staffing needs, stock prices, inventory, phone calls per minute).

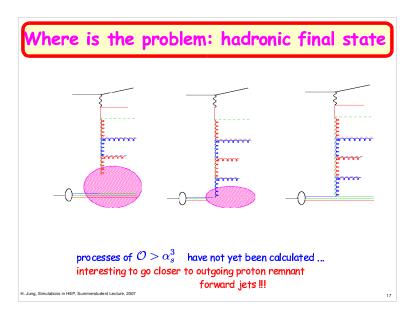


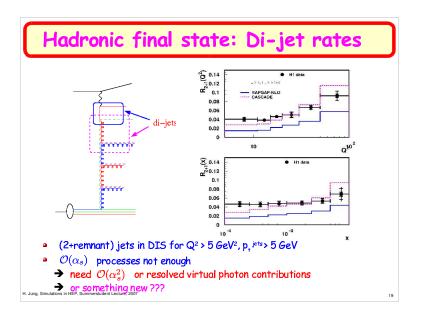


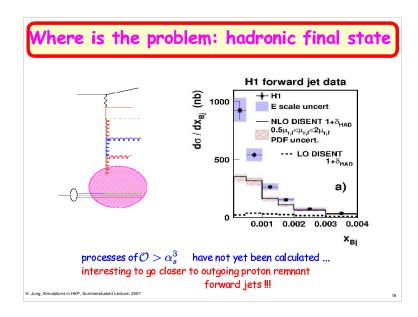


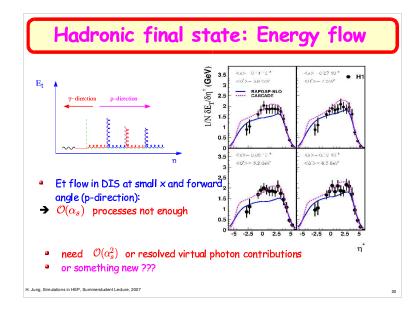












How to simulate these processes?

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Monte Carlo method

- Monte Carlo method
 - refers to any procedure that makes use of random numbers
 - uses probability statistics to solve the problem
- Monte Carlo methods are used in:
 - Simulation of natural phenomena
 - Simulation of experimental apparatus
 - Numerical analysis
- Random number:

one of them is 3

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No such thing as a single random number

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No such thing as a single random number

A sequence of random numbers is a set of numbers that have nothing to do with the other numbers in a sequence

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Random Numbers

- In a uniform distribution of random numbers in [0,1] every number has the same chance of showing up
- Note that 0.000000001 is just as likely as 0.5

To obtain random numbers:

- Use some chaotic system like roulette, lotto, 6-49, ...
- Use a process, inherently random, like radioactive decay
- Tables of a few million truly random numbers exist

(....until a few years ago....)

BUT not enough for most applications

- Hooking up a random machine to a computer is NOT toooooo good, as it leads to irreproducible results, making debugging difficult....
- → Develop Pseudo Random Number generators !!!!

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Going out to Monte Carlo





- Obtain true Random Numbers from Casino in Monte Carlo
- Puhhh... Going out every night ...



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Pseudo means: Oxford Advanced Dict.: False
Quasi means: Oxford Advanced Dict.: almost
BUT here the meaning is different

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...

Quasi Random Numbers

- mathematical randomness is not attainable in computer generated random numbers
- more important: assure that the "random" sequence has the necessary properties to produce a desired result ... (hmmmm !!!)
 - examples:
 - in multidimensional integration, each multi-dim point is considered independently of the others, and the order in which they appear plays no role!
 - degree of fluctuations about uniformity: in many cases a "superuniform" distribution is more desirable than a truly random distribution with uniform probability density!
- use of Quasi Random Numbers might lead to faster convergence of the integration but needs to be checked carefully ...

Important in Monte Carlo integrations

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Pseudo Random Numbers

Pseudo Random Numbers

- are a sequence of numbers generated by a computer algorithm, usually uniform in the range [0,1]
- more precisely: algo's generate integers between 0 and M, and then $r_n = I_n/M$
- A very early example: Middles Square (John van Neumann, 1946): generate a sequence, start with a number of 10 digits, square it, then take the middle 10 digits from the answer, as the next number etc.: 57721566492 = 33317792380594909291

Hmmmm, sequence is not random, since each number is determined from the previous, but it appears to be random

• this algorithm has problems

BUT a more complex algo does not necessarily lead to better random sequences

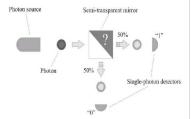
Better us an algo that is well understood

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True Random Numbers

- Random numbers from classical physics: coin tossing evolution of such a system can be predicted, once the initial condition is known... however it is a chaotic process ... extremely sensitive to initial conditions.
- Truly Random numbers used for
- Cryptography
 Confidentiality
 Authentication
- Scientific Calculation
- Lotteries and gambling
 do not allow to increase
 chance of winning by having a
 bias too bad

Random numbers from quantum physics: intrinsic random photons on a semi-transparent mirror



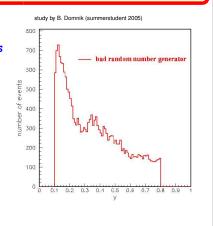
- Available and tested in MC generator by a summer student
- Generator is however very slow...

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Random Number generators

Compare random number generators with physics process

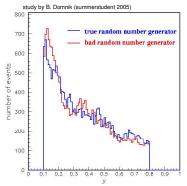
- y spectrum of electron
- observe peaks
- → coming from physics?



Random Number generators

Compare random number generators with physics process

- y spectrum of electron
- observe peaks
- coming from physics?
- BUT coming from bad random number generator



From now on assume:

we have good random number generator

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The first simulation: radioactive decay

- implement algo into a small program
- show results after 3000 sec for:

```
N_0 = 100, \alpha = 0.01 \text{ s}^{-1}

\Delta t = 1\text{ s}

N_0 = 5000, \alpha = 0.03 \text{ s}^{-1}
```

```
\Delta† = 1s
algo:
  alpha1 = 0.01
  N01 = 100
  deltat = 1
  do I=1,300
     it = it + 1
     do j = 1, N01
        x = RN1
        fr = deltat*alpha1
        if(x.lt.fr) then
   reduce number of parents NO1
           N01 = N01 - 1
         endif
   fill for each time it number NO1
      call hfill(400,real(it+0.3),0,1.)
```

Simulating Radioactive Decay

- radioactive decay is a truly random process
- $\bullet \quad dN = -N \alpha dt \text{ i.e. } N = N_0 e^{-\alpha t}$
- probability of decay is constant ... independent of the age of the nuclei: probability that nucleus undergoes radioactive decay in time Δt is p: p = $\alpha \Delta t$ (for $\alpha \Delta t \ll 1$)
- Problem:

consider a system initially having No unstable nuclei.

How does the number of parent nuclei, N, change with time?

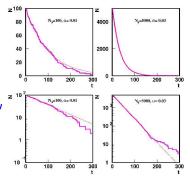
Algorithm:

```
LOOP from t=0 to t, step \Delta t
LOOP over each remaining parent nucleus Decide if nucleus decays:

IF ( random # < \alpha \Delta t) then
reduce number of parents by 1
ENDIF
END LOOP over nuclei
Plot or record N vrs t
END LOOP over time
END
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```

The first simulation: radioactive decay

- implement algo into a small program
- show results after 3000 sec for: N₀ = 100, α = 0.01 s⁻¹
 Δt = 1s
 N₀ = 5000, α = 0.03 s⁻¹
 Δt = 1s
- MC experiment does not exactly reproduce theory
- results from MC experiment show statistical fluctuations ...
-as expected



Monte Carlo technique: basics

Law of large numbers chose N numbers u randomly, with probability density uniform in [a,b], evaluate f(u,) for each u, :

$$rac{1}{N}\sum_{i=1}^N f(u_i)
ightarrow rac{1}{b-a}\int_a^b f(u)du$$

for large enough N Monte Carlo estimate of integral converges to correct answer.

Convergence

is given with a certain probability ...

THIS is a mathematically serious and precise statement !!!!

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Central Limit Theorem

 Central Limit Theorem for large N the sum of independent random variables is always normally (Gaussian) distributed:

$$f(x) = \frac{1}{s\sqrt{2\pi}} \exp\left[-\frac{(x-a)^2}{2s^2}\right]$$

example: take sum of uniformly distributed random numbers:

$$\begin{split} R_n &= \sum_{i=1}^n R_i \\ E[R_1] &= \int u du = 1/2, \\ V[R_1] &= \int (u - 1/2)^2 du = 1/12 \\ E[R_n] &= n/2 \\ V[R_n] &= n/12 \end{split}$$

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Expectation values and variance

• Expectation value (defined as the average or mean value of function f):

$$E[f] = \int f(u)dG(u) = \left(\frac{1}{b-a} \int_a^b f(u)du\right) = \frac{1}{N} \sum_{i=1}^N f(u_i)$$

for uniformly distributed u in [a,b] then dG(u) = du/(b-a)

rules for expectation values:

$$E[cx+y] = cE[x] + E[y]$$

$$V[f] = \int (f - E[f])^{2} dG = \left(\frac{1}{b - a} \int_{a}^{b} (f(u) - E[f])^{2} du\right)$$

rules for variance:

if x,y correlated

if x,y uncorrelated:
$$V[cx+y]=c^2V[x]+V[y]$$

 $V[cx + y] = c^{2}V[x] + V[y] + 2cE[(y - E[y])(x - E[x])]$

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 $E[R_1] = \int u du = 1/2,$
 $V[R_1] = \int (u - 1/2)^2 du = 1/12$
 $E[R_n] = n/2$
 $V[R_n] = n/12$

0.1

0.05

• for Gaussian with mean=0 and

variance=1, take for n=12:

 $N(0,1) o rac{R_n - n/2}{n/12}$

0.15

0.1

0.05



Resumee: Monte Carlo technique

Law of large numbers

$$\frac{1}{N} \sum_{i=1}^{N} f(u_i) \to \frac{1}{b-a} \int_{a}^{b} f(u) du$$

MC estimate converges to true integral

Central limit theorem

MC estimate is asymptotically normally distributed it approaches a Gaussian density

$$\sigma = rac{\sqrt{V[f]}}{\sqrt{N}} \sim rac{1}{\sqrt{N}}$$

with effective variance V(f)

- \rightarrow to decrease σ , either reduce V(f) or increase N
- advantages for n-dimensional integral ...
 i.e. phase space integrals 2 → n processes

is where other approaches tend to fail

Buffons Needle: Crude Monte Carlo

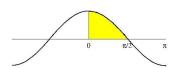
 Buffons needle (Buffon 1777) is essentially integration of

$$\int_0^{\pi/2} \cos(\alpha) d\alpha$$

apply Law of large numbers:

$$\frac{1}{N} \sum_{i=1}^{N} f(u_i) \to \frac{1}{b-a} \int_a^b f(u) du$$

- compare Hit & Miss with Integration
- 1st example of true Monte Carlo experiment
- equivalence of integration and MC event generation



trials	π (hit&miss)	π (integral)		
100	3.27869	3.12265		
1000	3.36700	3.11833		
10000	3.14218	3.15129		
100000	3.13087	3.13416		
1000000	3.14127	3.14337		
10000000	3.14154	3.14168		
100000000	3.12174	3.14156		

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Monte Carlo: Buffons Needle – Hit & Miss

- Buffons needle (Buffon 1777)
 pattern of parallel lines with
 distance d,
 randomly throw needle with
 length d onto stripes,
 count hit, when needle crosses
 strip count miss, if not
- probability for hit is:

$$\frac{d\cos(\alpha)}{d} = \cos(\alpha)$$

all angles are equally likely:

$$\frac{\int_0^{\pi/2} \cos(\alpha) d\alpha}{\pi/2} = \frac{2}{\pi}$$

http://www.angelfire.com/wa/hurben/buff.htm

op over ntrials x=RN(1) * d

trials	π	error
100	2.9850	0.2374
1000	3.2733	0.0749
10000	3.1645	0.0237
100000	3.1483	0.0075
1000000	3.1401	0.0024
10000000	3.1422	0.0008

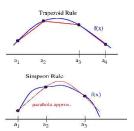
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Integration: Monte Carlo versus others

One dimensional quadrature

$$I = \int f(x)dx = \sum_{i=1}^{n} w_i f(x_i)$$

- Monte Carlo: Hit & Miss
 w = 1 and x, chosen randomly
- Trapezoidal Rule: approximate integral in subinterval by area of trapezoid below (above) curve
- Simpson quadrature approximate by parabola
- Gauss quadrature
 approximate by higher order polynomial



method	err (1d)	error
MC	$n^{-1/2}$	$n^{-1/2}$
Trapez	n^{-2}	$n^{-2/d}$
Simpson	n^{-4}	$n^{-4/d}$
Gauss	n^{-2m+1}	$n^{-(2m-1)/d}$

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MC method: advantage of hit & miss

- integration ➤ weighting events large fluctuations from large weights weights also to errors applied difficult to apply further hadronization
- real events all have weight = 1
- Hit & Miss method:

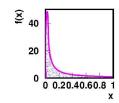
MC for function f(x): get random number: R1 in (0,1) and R2 in (0,1) calculate x = R1 reject event if: $f_x < f_{max} R2$

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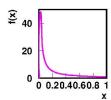


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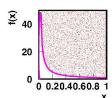


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MC method: advantage of hit & miss

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 difficult to apply further hadronization
- real events all have weight = 1 !!!
- Hit & Miss method:

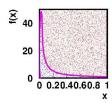
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MC for function f(x):
get random number:
R1 in (0,1) and R2 in (0,1)
calculate x = R1
reject event if: f_x < f_{max} R2



BUT: Hit & Miss method inefficient for peaked f(x)

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MC method: do even better ...

Importance sampling

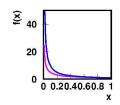
generate x according to $g(\mathbf{x})$: $\int_{x_{min}}^{x} g(x')dx' = R1 \int_{x_{min}}^{x_{max}} g(x')dx'$ example: $f(x) = 1/x^{0.7}$ g(x) = 1/x $x = x_{min} \cdot \left(\frac{x_{max}}{x_{min}}\right)^{R1}$

reject event if: f(x) < g(x) R2

MC for function f(x)

approximate $f(x) \sim g(x)$

with g(x) > f(x) simple and integrable



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MC method: do even better ...

Importance sampling

MC for function f(x) approximate $f(x) \sim g(x)$ with g(x) > f(x) simple and integrable generate x according to g(x): $\int_{x_{min}}^{x} g(x')dx' = R1 \int_{x_{min}}^{x_{max}} g(x')dx'$ example: $f(x) = 1/x^{0.7}$ g(x) = 1/x $x = x_{min} \cdot \left(\frac{x_{max}}{x_{min}}\right)^{R1}$ reject event if: f(x) < g(x) R2

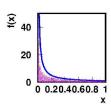
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MC method: do even better ...

Importance sampling

approximate $f(\mathbf{x}) \sim g(\mathbf{x})$ with $g(\mathbf{x}) > f(\mathbf{x})$ simple and integrable generate \mathbf{x} according to $g(\mathbf{x})$: $\int_{x_{min}}^{x} g(x')dx' = R1 \int_{x_{min}}^{x_{max}} g(x')dx'$ example: $f(x) = 1/x^{0.7}$ g(x) = 1/x $x = x_{min} \cdot \left(\frac{x_{max}}{x_{min}}\right)^{R1}$ reject event if: $f(\mathbf{x}) < g(\mathbf{x}) \in \mathbb{R}2$

MC for function f(x)



MC method: do even better ...

Importance sampling

MC for function f(x)

approximate $f(x) \sim g(x)$ with g(x) > f(x) simple and integrable generate x according to g(x):

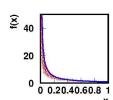
$$\int_{x_{min}}^{x} g(x')dx' = R1 \int_{x_{min}}^{x_{max}} g(x')dx'$$

example: $f(x) = 1/x^{0.7}$

$$g(x) = 1/x$$

$$x = x_{min} \cdot \left(\frac{x_{max}}{x_{min}}\right)^{R1}$$

reject event if: f(x) < g(x) R2



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Importance Sampling

Importance sampling

sample important regions

 MC calculations most efficient for small weight fluctuations:

 $f(x)dx \rightarrow f(x) dG(x)/g(x)$

- chose point according to g(x) instead of uniformly
- f is divided by g(x) = dG(x)/dx
- generate x according to:

$$R \int_{a}^{b} g(x')dx' = \int_{a}^{x} g(x')dx'$$

- relevant variance is now V(f/g): small if $g(x) \sim f(x)$
- how-to get g(x)
 - (1) g(x) is probability: g(x) > 0 and $\int dG(x) = 1$
 - (2) integral $\int dG(x)$ is known analytically
 - (3) G(x) can be inverted (solved for x)
 - (4) f(x)/g(x) is nearly constant, so that V(f/g) is small compared to V(f)

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MC method: do even better ...

Importance sampling

MC for function f(x)

approximate $f(x) \sim g(x)$ with g(x) > f(x) simple and integrable generate x according to g(x):

$$\int_{x_{min}}^{x} g(x')dx' = R1 \int_{x_{min}}^{x_{max}} g(x')dx'$$

example: $f(x) = 1/x^{0.7}$

$$g(x) = 1/x$$

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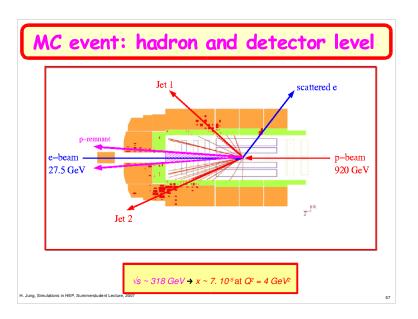
20 0 0.20.40.60.8 1

WOW !!! very efficient even for peaked f(x)

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Applications in High Energy Physics

- Simulation of detector response
- Apply MC method to ete-
- what about hadronization
- what about QCD radiation
- going even further: initial state radiation
- how-to do a DIS Monte Carlo event generator
- some examples



MC generators - different applications ..

- calculate x-section of various processes complicated integrals multi - differential, in any variable
- MC simulation of detector response

input: hadron level events - output: detector level events

Calorimeter ADC hits

Tracker hits

need knowledge of particle passage through matter, x-section ...

need knowledge of actual detector

x-section on parton level

multipurpose MC event generators:

x-section on parton level

including multi-parton (initial & final state) radiation

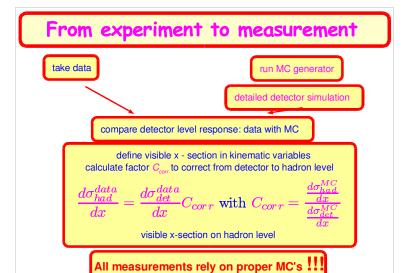
remnant treatment (proton remnant, electron remnant)

hadronisation/fragmentation (more than simple fragmentation functions...)

fixed order parton level theorists like it

integration of multidimensional phase space

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Constructing a MC for ete: the simple case

process: e⁺e⁻ → μ⁺ μ⁻



 $d\cos\theta d\phi$

$$\frac{d\sigma}{d\cos\theta d\phi} = \frac{\alpha_{em}^2}{4s} \left(1 + \cos^2\theta\right)$$

• goal: generate 4-momenta of μ 's, need cm energy s, $\cos \theta$, ϕ

random number R1(0,1) ϕ = 2 π R1 random number R2(0,1) cos θ = -1 + 2 R2

for every R1, R2 use weight with repeat many times

Z 4000 3000 1000 -1 -0.5 0 0.5 1

after 100000 events



Constructing a MC for ete: the simple case

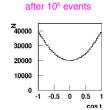
process: e⁺e⁻ → μ⁺ μ⁻



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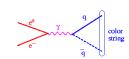
- for every R1, R2 use weight with
- repeat many times

 $\frac{d\sigma}{d\cos\theta\,d\phi}$

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Constructing a MC for $e^+e^- o q \bar q$

- process $e^+e^- o q \overline{q}$
- $\frac{d\sigma}{d\cos\theta d\phi} = \frac{\alpha_{em}^2}{4s} \left(1 + \cos^2\theta\right)$



- generate scattering as for e⁺e⁻ → μ⁺ μ⁻
- BUT what about fragmentation/hadronization ???
- use concept of local parton-hadron duality

Different approaches to fragmentation/hadronization:

- → independent fragmentation
- string fragmentation (Lund Model)
- cluster fragmentation (HERWIG model)

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Example event: $e^+e^- \rightarrow \mu^+ \mu^-$

example from PYTHIA: Event listing

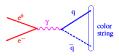
I	particle/jet	KS	KF	orig	p_x	р_у	p_z	E	m		
	!e+! !e-!	21 21	-11	0	0.000	0.000	30.000 -30.000	30.000 30.000	0.001 0.001		
4 5 6 7 8	!e+! !e-! !e+! !e-! !ZO! !mu-!	21 21 21 21 21 21	- 11 - 11 - 11 11 23 13	2 3 4 0 7	0.000 0.000 0.143 0.000 0.143 -9.510	0.000 0.000 0.040 0.000 0.040 1.741	30.000 -30.000 26.460 -29.998 -3.539 24.722	30.000 30.000 26.460 29.998 56.458 26.546	0.000 0.000 0.000 0.000 56.347 0.106	e [†] (3) ζ (11)	μ ⁺ (12)
10 11 12	(ZO) gamma mu- mu+	21 11 1 1 1 sum:	-13 23 22 13 -13	7 3 8 9	9.653 0.143 -0.143 -9.510 9.653	0.040 -0.040 1.741 -1.700	-28.261 -3.539 3.539 24.722 -28.261 0.000	29.913 56.458 3.542 26.546 29.913 60.000	0.106 56.347 0.000 0.106 0.106	e (4)	μ-(13)

- technicalities/advantages
- can work in any frame
- → Lorentz-boost 4-vectors back and forth
- > can calculate any kinematic variable
- history of event process

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Constructing a MC for $e^+e^- o q \bar q$

- process $e^+e^- o q \overline{q}$
- $\frac{d\sigma}{d\cos\theta d\phi} = \frac{\alpha_{em}^2}{4s} \left(1 + \cos^2\theta \right)$



- generate scattering as for e⁺e⁻ → μ⁺ μ⁻
- BUT what about fragmentation/hadronization ???
- use concept of local parton-hadron duality

linear confinement potential: $V(r) \sim -1/r + \kappa r$

with κ ~ 1 GeV/fm

qq connected via color flux tube of transverse size of hadrons (~1 fm) color tube: uniform along its length → linearly rising potential

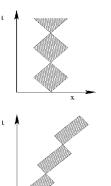
→ Lund string fragmentation

Lund string fragmentation

- in a color neutral qq-pair, a color force is created in between
- color lines of the force are concentrated in a narrow tube connecting q and q, with a string tension of:

κ~1GeV/fm~0.2 GeV²

- as q and q are moving apart in qq rest frame, they are de-accelerated by string tension, accelerated back etc ... (periodic oscillation)
- viewed in a moving system, the string is



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Hadronization: particle masses and decays

- particle masses
 - → taken from PDG, where known, otherwise from constituent masses
- \Rightarrow in hard scattering production process short lived particles (ρ,Δ) have nominal mass, without mass broadening
- → in hadronization use Breit-Wigner:

$$\mathcal{P}(m)dm \propto rac{1}{(m-m_0)^2 + \Gamma^2/4}$$

- lifetimes
 - → related to widths ... but for practical purpose separated
 - \rightarrow $P(\tau)d\tau \sim exp(-\tau/\tau)d\tau$
 - \rightarrow calculate new vertex position $v' = v + \tau p/m$
- → taken from PDG, where known
- \Rightarrow assume momentum distribution given by phase space only exceptions, like $\omega,\phi\to\pi^+\pi^-\pi^0$, or $D\to K\pi,\,D^*\to K\pi\pi$ and some semileptonic decays use matrix elements

Fragmentation in the String Model

longitudinal frag.

transverse frag.

hadronization: iterative process

string breaks in ag pairs (still respecting color flow)

select transverse motion with m=m_{aa} (and flavor)

$$P \sim \exp\left(-\frac{\pi m_t^2}{\chi}\right) = \exp\left(-\frac{\pi m^2}{\chi}\right) \exp\left(-\frac{\pi p_t^2}{\chi}\right)$$

suppression of heavy quark production u:d:s:c~1:1:1:0.37:10-10 actually leave it as a free parameter

 longitudinal fragmentation symmetric fragmentation function (from either g or g) $f(z) \sim z^{-1}(1-z)^{\alpha} \exp(-b m^2/z)$ harder spectrum for heavy quarks

- start from q or q
- repeat until cutoff is reached
- heavy use of random numbers and importance sampling method

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Example event e⁺e⁻ →qq

example from PYTHIA Monte Carlo generator including hadronization

										$e^{+}(1) \gtrsim (11)$ c (15)
I	particle/je	t KS	KF	orig	p_x	P_Y	p_z	E	m	γ , , , , , , , , , , , , , , , , , , ,
1	1e+1	21	-11	0	0.000	0.000	30.000	30.000	0.001	>vvv
2	!e-!	21	11	0	0.000	0.000	-30.000	30.000	0.001	(10)
5	!e+!	21	-11	3	0.018	0.040	0.702	0.703	0.000	₹ (2) ₹ (16)
	!e-!	21	11	4	0.000	0.000	-29.998	29.998	0.000	- (10)
10		11	23	7	0.018	0.040	-29.297	30.701	9.180	
	qamma	1	22	í	-0.018	-0.040	29.298	29.298	0.000	
15	(c) 1	12	4	10	-1.950	-3.529	-19.752	20.215	1.500	
16	(cbar) \	/ 11	- 4	10	1.967	3.569	-9.545	10.486	1.500	
	(-1				0.010	0.040	20 207	20.701		apply fragmentation
17		11	92 421	15 17	0.018	-1.495	-29.297 -9.002	30.701 9.325	9.180 1.865	6 4 5
19		11	223	17	-0.300	-0.076	-3.228	3.338	0.793	directly to parton
	pi+	1	211	17	-0.168	-0.172	-0.861	0.904		
21	(rho-)	11	-213	17	-0.114	-0.513	-4.992	5.106		all covered by
22		11	223	17	-0.173	0.118	-2.022	2.180	0.789	· ·
	pi+	1	211	17	0.226	0.925	-2.593	2.766	0.140	hadronization soft
	(D*-)	11	-413	17	1.001	1.253	-6.599	7.082	2.010	
25		1	-11 12	18 18	-0.191 -0.154	0.241	-1.261 -4.174	1.297	0.001	• where is
	nu_e	1	12	18	-0.154	-0.789	-4.1/4	4.250	0.000	- Muele 12
- ::										
										QCD ???
	pi-	1	-211	47	0.318	-0.061	-1.293	1.340	0.140	don
		sum:	0.00		0.000	0.000	0.000	60.000	60.000	

Doing things better: e⁺e⁻ →qqg

- process ete →qqg
- full matrix element calculation
- watch out color flow !!!
- gluons act as kicks on strings

_	partitle/je		1.0	011	19	P_X	Py	P_2	-	200	6. (1
											~
	!e+!		21 -1		0	0.000	0.000	30.000	30.000	0.001	
2	!e-!		21 1	1	0	0.000	0.000	-30.000	30.000	0.001	
5	!e+!		21 -1	1	1	0.000	0.000	29.699	29.699	0.000	ē (2
6	!e-!	- 1	21 1	1	2	-1.319	-1.236	-26.950	27.011	0.000	6 (2
7	!Z0!		21 2	3	0	-1.319	-1.236	2.748	56.710	56.614	
8	!c!		21	4	7	-15.986	16.072	18.293	29.167	1.500	
9	!cbar!		21 -	4	7	14.667	-17.308	-15.545	27.542	1.500	
 											-
11	gamma		1 2	2	2	1.320	1.236	-2.744	3.286	0.000	
15	(c) A	. :	12	4	8	-11.291	11.550	13.219	20.926	1.500	
16	(g) I		12 2		8	-3.992	3.139	4.805	6.991	0.000	
17	(g) I		12 2	1	8	-0.279	0.951	0.179	1.007	0.000	On the
18	(g) I		12 2	1	8	0.122	-0.178	-0.505	0.550	0.000	•
19	(g) I		12 2		9	0.128	-0.237	0.146	0.307	0.000	
20	(g) I		12 2	1	9	-0.093	-0.746	-0.364	0.835	0.000	
21	(g) I		12 2		9	8.331	-6.743	-6.396	12.482	0.000	
22	(cbar) V	1	11 -	4	9	5.754	-8.971	-8.335	13.613	1.500	
											_

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	m	البيويو
·	q 💘	7.



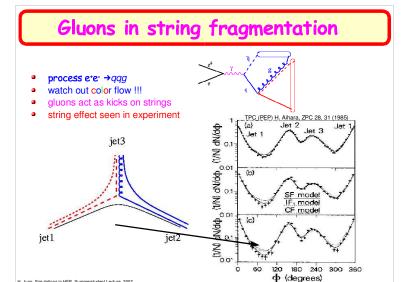
- more large p. emissions
- not all covered by fixed order calculations
- doing much better needed
- parton shower approach

Approximations to higher orders: parton showers

- Approximation to higher orders.....
- fragmentation functions
- parton density functions
- since alphas is not small, higher orders contributions are important Approximations:

DGLAP (Dokshitzer, Gribov, Lipatov, Altarelli, Parisi) BFKL (Balitski, Fadin, Kuraev, Lipatov)

CCFM (Catani, Ciafaloni, Fiorani, Marchesini)



DGLAP equation

- differential form $q rac{\partial}{\partial q} f(x,q) = \int rac{dz}{z} rac{lpha_s}{2\pi} P_+(z) \, f\left(rac{x}{z},q
 ight)$
- modified differential form using "Sudakov form factor"

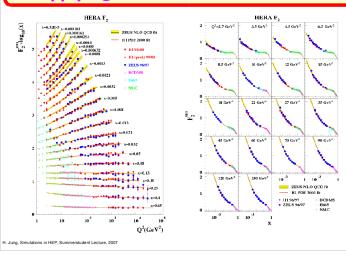
$$\Delta_s(q_0,q) = \exp\left(-\bar{lpha}_s \int rac{dz}{z} \int_{q_0}^q rac{dq'}{q'} \tilde{P}(z)
ight)$$

$$q \frac{\partial}{\partial q} \frac{f(x,q)}{\Delta_s(q,q_0)} = \int \frac{dz}{z} \frac{\alpha_s}{2\pi} \frac{\tilde{P}(z)}{\Delta_s(q,q_0)} f\left(\frac{x}{z},q\right)$$

$$f(x,q) = f_0(x,q)\Delta_s(q,q_0) + \int \frac{dz}{z} \int \frac{dq'}{q'} \cdot \Delta_s(q',q_0)\tilde{P}(z)f\left(\frac{x}{z},q\right)$$

no-branching probability form q₀ to q

Applying DGLAP to DIS data



Solving integral equations

$$\phi_1(x) = f(x) + \lambda \int_a^b K(x,y) f(y) dy$$

$$\phi_2(x) = f(x) + \lambda \int_a^b K(x,y_1) f(y_1) dy_1 + \lambda^2 \int_a^b \int_a^b K(x,y_1) K(y_1,y_2) f(y_2) dy_2 dy_1$$

$$\phi_n(x) = \sum_{i=0}^n \lambda^i u_i(x)$$

$$u_0(x) = f(x)$$

$$u_1(x) = \int_a^b K(x,y)f(y)dy$$

$$u_n(x) = \int_a^b \int_a^b K(x, y_1) K(y_1, y_2) \cdots K(y_{n-1}, y_n) f(y_n) dy_2 \cdots dy_n$$

with the solution:
$$\phi(x)=\lim_{n o\infty}q_n(x)=\lim_{n o\infty}\sum_{i=0}^n\lambda^iu_i(x)$$

Solving DGLAP equations ...

- Different methods to solve integro-differential equations
 - brute-force (BF) method (M. Miyama, S. Kumano CPC 94 (1996) 185)

$$rac{df(x)}{dx} = rac{f(x)_{m+1} - f(x)_m}{\Delta x_m} \int f(x) dx = \sum f(x)_m \Delta x_m$$

- Laquerre method (5. Kumano J.T. Londergan OPC 69 (1992) 373, and L. Schoeffel Nucl. Instrum. Meth., A423:439-445, 1999)
- Mellin transforms (M. Glueck, E. Reya, A. Vogt Z. Phys. C48 (1990) 471)
- QCDNUM: calculation in a grid in x,Q2 space (M. Botje EurPhys, J. C14 (2000) 285-297)
- CTEQ evolution program in x,Q2 space: http://www.phys.psuedu/~cteq
- QCDFIT program in x,Q2 space (C. Passaud, F. Zomer, LAL preprint LAL/94-02, H109/94-404 H109/94-
- MC method using Markov chains (S. Jadach, M. Skrzypek hep-ph/0504205)
- Monte Carlo method from iterative procedure
- brute-force method and MC method are best suited for detailed studies of branching processes !!!

Solution of DGLAP equation

$$f(x,t) = f(x,t_0)\Delta_s(t) + \int rac{dz}{z} \int rac{dt'}{t'} \cdot rac{\Delta_s(t)}{\Delta_s(t')} ilde{P}(z) f\left(rac{x}{z},t'
ight)$$

solve integral equation via explicit iteration:

$$f_0(x,t) = f(x,t_0)\Delta(t) \qquad \begin{array}{c} \text{from } t\text{ to } t\\ \text{wo branching} \end{array} \text{ branching at } t' \qquad \begin{array}{c} \text{form } t_0\text{ to } t'\\ \text{wo branching} \end{array}$$

$$f_1(x,t) = f(x,t_0)\Delta(t) + \int_{t_0}^t \frac{dt'}{t'} \frac{\Delta(t)}{\Delta(t')} \int \frac{dz}{z} \tilde{P}(z) f(x/z,t_0)\Delta(t')$$

DGLAP re-sums leading logs...

$$f(x,t) = f(x,t_0)\Delta_s(t) + \int rac{dz}{z} \int rac{dt'}{t'} \cdot rac{\Delta_s(t)}{\Delta_s(t')} ilde{P}(z) f\left(rac{x}{z},t'
ight)$$

solve integral equation via iteration:

$$\begin{array}{lll} f_{0}(x,t) & = & f(x,t_{0})\Delta(t) & \text{from } t' \text{ to } t \text{ wo branching at } t' \text{ wo branching } \\ f_{1}(x,t) & = & f(x,t_{0})\Delta(t) + \int_{t_{0}}^{t} \frac{dt'}{t'} \frac{\Delta(t)}{\Delta(t')} \int \frac{dz}{z} \tilde{P}(z) f(x/z,t_{0})\Delta(t') \\ & = & f(x,t_{0})\Delta(t) + \log \frac{t}{t_{0}} A \otimes \Delta(t) f(x/z,t_{0}) \\ f_{2}(x,t) & = & f(x,t_{0})\Delta(t) + \log \frac{t}{t_{0}} A \otimes \Delta(t) f(x/z,t_{0}) + \\ & \quad \frac{1}{2} \log^{2} \frac{t}{t_{0}} A \otimes A \otimes \Delta(t) f(x/z,t_{0}) \\ f(x,t) & = & \lim_{n \to \infty} f_{n}(x,t) = \lim_{n \to \infty} \sum_{x = 1}^{n} \frac{1}{n!} \log^{n} \left(\frac{t}{t_{0}}\right) A^{n} \otimes \Delta(t) f(x/z,t_{0}) \end{array}$$

Parton showers for the initial state

spacelike (Q<0) parton shower evolution

starting from hadron (fwd evolution)
 or from hard scattering (bwd evolution)



• select q, from Sudakov form factor



select z, from splitting function



- select q₂ from Sudakov form factor
- $_{1}=\mathbf{z}_{1}\mathbf{x}_{0}$

- select z, from splitting function
- stop evolution if q₂>Q_{1,111}

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Parton showers to solve DGLAP evolution of for fixed x and Q2 chains with different branchings contribute iterative procedure, spacelike parton showering of the space of th

Parton Showers for the final state

timelike parton shower evolution

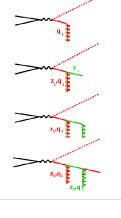
starting with hard scattering

• select q, from Sudakov form factor

• select z, from splitting function

select q₂ from Sudakov form factor

- select z₂ from splitting function
- stop evolution if q₂<q₀



Parton Shower

- Evolution equation with <u>Sudakov form factor</u> recovers exactly evolution equation (with prescription)
- Sudakov form factor particularly suited form Monte Carlo approach
- Sudakov form factor
 - → gives probability for no-branching between q₀ and q
 - → sums virtual contributions to all orders (via unitarity)
 - → virtual (parton loop) and
 - → real (non-resolvable) parton emissions
 - need to specify scale of hard process (matrix element) Q ~ p.
- need to specify cutoff scale Q₀ ~ 1 GeV

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From F₂ to Heavy Quarks in pp

The superior of the entire control of the contro

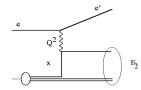
The DIS process ep → epX

cross section

$$\frac{d\sigma(ep\rightarrow e'X)}{dy\,dQ^2} = \frac{4\pi\alpha^2}{yQ^4} \left(\left(1-y+\frac{y^2}{2}\right)F_2^p(x,Q^2) - \frac{y^2}{2}F_L^p(x,Q^2) \right)$$

with
$$F_2^p(x,Q^2) = \sum_f e_f^2 \left(x q_f(x,Q^2) + x \overline{q}_f(x,Q^2)
ight)$$

- Exercise: how to simulate this process?
 - integration of x-section
 - simulation of events
 - using parton densities and neglect F_L



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Heavy Quark production in pp

x-section

$$\sigma(\mathrm{pp}
ightarrow \mathrm{Q} \mathrm{ar{Q}} \mathrm{X}) \;\;\; = \;\; \int rac{dx_1}{x_1} rac{dx_2}{x_2} x_1 G(x_1, ar{q}) x_2 G(x_2, ar{q}) imes \hat{\sigma}(\hat{s}, ar{q})$$

with gluon densities $xG(x,ar{q})$

• hard x-section:



Heavy Quarks in pQCD

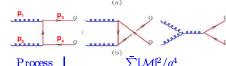
Light Quarks

$$\hat{s} = (p_1 + p_2)^2$$
 $\hat{t} = (p_1 - p_3)^2$
 $\hat{u} = (p_2 - p_3)^2$

Process $\begin{array}{c|c}
\hline
Process & \overline{\sum} |\mathcal{M}|^2/g^4 \\
\hline
q\bar{q} \to q'\bar{q}' & \frac{4}{9} \frac{\hat{t}^2 + \hat{a}^2}{\hat{s}^2} \\
gg \to q\bar{q} & \frac{1}{6} \frac{\hat{t}^2 + \hat{a}^2}{\hat{t}\hat{a}} - \frac{3}{8} \frac{\hat{t}^2 + \hat{a}^2}{\hat{s}^2}
\end{array}$

Heavy Quarks with

$$au_1 = -rac{\hat{t} - m^2}{\hat{s}},$$
 $au_2 = -rac{\hat{u} - m^2}{\hat{s}},$ $ho = rac{4m^2}{\hat{s}}$



Ellis, Stirling, Webber QCD & Collider physics p348

1 100000	Z V 1 / 9
q ar q o Q ar Q	$rac{4}{9}\left(au_1^2+ au_2^2+rac{ ho}{2} ight)$
gg o Qar Q	$\left(\frac{1}{6\tau_1\tau_2} - \frac{3}{8}\right) \left(\tau_1^2 + \tau_2^2 + \rho - \frac{\rho^2}{4\tau_1\tau_2}\right)$

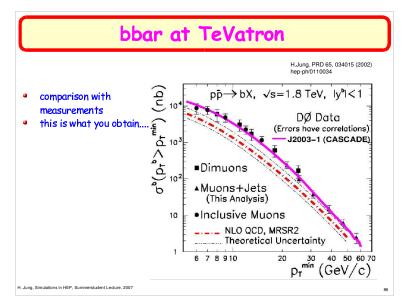
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Conclusion

- Monte Carlo event generators are needed to calculate multi-parton cross sections
- Monte Carlo method is a well defined procedure
- hadronization is needed to compare with measurements
- parton shower (leading log) approach is needed, hadronziation not enough
- MC approach extended from simple e+e- processes to
 - ep processes
 - pp processes
 - and heavy Ion processes
- proper Monte Carlos are essential for any measurement

Monte Carlo event generators contain all our physics knowledge !!!!!

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List of available MC programs

- HERA Monte Carlo workshop: www.desy.de/~heramc
- ARIADNE
- A program for simulation of QOD cascades implementing the color dipole model
- AROMA

Heavy quark production in boson-gluon fusion using full electroweak LO cross-sections (with quark masses) in ep collisions, DIS and photoproduction, Parton showers and Lund hadronization gives full events.

CASCADE

is a full hadron level Monte Carlo generator for \$ep\$ and \$p\bar{p}\$scattering at small \$x\$ build according to the CCFM evolution equation. It is applicable in \$ep\$ to photoproduction and DIS, and for heavy quark production as well as inelastic \$J/\psi\$.

HERWIG

General purpose generator for Hadron Emission Reactions With Interfering Gluons; based on matrix elements, parton showers including color coherence within and between jets, and a cluster model for hadronization.

JETSET

The Lund string model for hadronization of parton systems.

List of available MC programs

LDCMC

A program which implements the Linked Dipole Chain (LDC) model for deeply inelastic scattering within the framework of ARIADNE. The LDC model is a reformulation of the CCFM model.

LEPTO

Deep inelastic lepton-nucleon scattering based on LO electroweak cross sections (incl. lepton polarization), first order QCD matrix elements, parton showers and Lund hadronization giving complete events. Soft color interaction model gives rapidity gap events

PHOJET

Multi-particle production in high energy hadron-hadron, photon-hadron, and photon-photon interactions (hadron = proton, antiproton, neutron, or pion).

POMPYT

Diffractive hard scattering in \$p\bar{p}\$, \$\gamma-p\$ and \$ep\$-collisions, based on pomeron flux and pomeron parton densities (several options included). Also pion exchange is included. Parton showers and Lund hadronization to give complete events.

H. Jung, Simulations in HEP, Summerstudent Lecture, 200

General literature

- Many new books are available in DESY library NEW ... ask at the desk there ...
- Statistische und numerische Methoden der Datenanalyse

V. Blobel & E. Lohrmann

- STATISTICAL DATA ANALYSIS. Glen Cowan.
- Particle Data Book S. Eidelman et al., Physics Letters B592,1 (2004) (http://pdg.lbl.gov/)
- Applications of pQCD

R.D. Field Addison-Wesley 1989

Collider Physics

V.D. Barger & R.J.N. Phillips Addison-Wesley 1987

Deep Inelastic Scattering.

R. Devenish & A. Cooper-Sarkar, Oxford 2

Handbook of pQCD

G. Sterman et al

Quarks and Leptons,

F. Halzen & A.D. Martin, J. Wiley 1984

QCD and collider physics

R.K. Ellis & W.J. Stirling & B.R. Webber Cambridge 1996

QCD: High energy experiments and theory G. Dissertori, I. Knowles, M. Schmelling Oxford 2003

H. Jung, Simulations in HEP, Summerstudent Lecture, 2007

List of available MC programs

PYTHIA

General purpose generator for \$e^+e^-\$, \$p\bar{p}\$ and \$ep\$-interactions, based on LO matrix elements, parton showers and Lund hadronization.

RAPGAP

A full Monte Carlo suited to describe Deep Inelastic Scattering, including diffractive DIS and LO direct and resolved processes. Also applicable for \$\gamma\$-production and partially for \$p\bar(p)\$ scattering.

H. Jung, Simulations in HEP, Summerstudent Lecture, 200

Literature & References

- F. James Rep. Prog. Phys., Vol 43, 1145 (1980)
- Glen Cowan STATISTICAL DATA ANALYSIS. Clarendon, 1998.
- Particle Data Book S, Eidelman et al., Physics Letters B592, 1 (2004) section on: Mathematical Tools (http://pda.lbl.gov/)
- Michael J. Hurben Buffons Needle

(http://www.angelfire.com/wa/hurben/buff.html)

J. Woller (Univ. of Nebraska-Lincoln) Basics of Monte Carlo Simulations

(http://www.chem.unl.edu/zeng/joy/mclab/mcintro.html)

- Hardware Random Number Generators:
 - A Fast and Compact Quantum Random Number Generator

(http://arxiv.org/abs/guant-ph/9912118)

Quantum Random Number Generator

(http://www.idguantique.com/products/quantis.htm)

Hardware random number generator (http://en.wikipedia.org/wiki/)

- Monte Carlo Tutorals
 - (http://www.cooper.edu/engineering/chemechem/MMC/tutor.html)
- History of Monte Carlo Method
 - (http://www.geocities.com/CollegePark/Quad/2435/history.html)
- Google: search for Monte Carlo Simulations

Literature & References (cont'd)

- T. Sjostrand et al PYTHIA/JETSET manual - The Lund Monte Carlos http://www.thep.lu.se/tf2/staff/torbjorn/Pythia.html
- H. Jung RAPGAP manual http://www-h1.desy.de/~jung/rapgap.html CASCADE manual
- http://www-h1.desy.de/~jung/cascade.html
 V. Barger and R. J.N. Phillips Collider Physics Addison-Wesley Publishing Comp. (1987)
- R.K. Ellis, W.J. Stirling and B.R. Webber QCD and collider physics Cambridge University Press (1996)

