



# Basics of Event Generators I

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# Outline of Lectures

- ▶ Lecture I: Basics of Monte Carlo methods, the event generator strategy, matrix elements, LO/NLO, . . .
- ▶ Lecture II: Parton showers, initial/final state, (matching/merging), hadronization, decays. . . .
- ▶ Lecture III: Minimum bias, multi-parton interactions, pile-up, summary of general purpose event generators, Idots
- ▶ Lecture IV: Protons vs. heavy ions, Glauber calculations, initial/final-state interactions, . . .

Buckley et al. (MCnet collaboration), *Phys. Rep.* **504** (2011) 145.



# Outline of Lecture I

## Monte Carlo Integration

Importance sampling

Obtaining Suitable Random Distributions

Predicting an Observable

## The Generic Event Generator

Factorization

The Generation Steps

Everything is QCD

## Matrix Element Generation

Tree-Level Matrix Elements

Next-to-Leading Order



How do we numerically estimate an integral of an arbitrary function  $f(\mathbf{x})$ ?

$$I = \int_{\Omega} d^n \mathbf{x} f(\mathbf{x})$$

Simple discretization (Simpsons rule, Gaussian quadrature) can be extremely inefficient if

- ▶  $n$  is large
- ▶  $\Omega$  is complicated
- ▶  $f(\mathbf{x})$  has peaks and divergencies.



## Importance sampling

Assume we are able to generate random variables  $\mathbf{X}_i$  such that

$$P\left(x^{(j)} < X_i^{(j)} < x^{(j)} + dx^{(j)}\right) = p_X(\mathbf{x})$$

if  $p(\mathbf{x}) > 0, \forall \mathbf{x} \in \Omega$  and zero outside, we can rewrite our integral

$$I = \int_{\Omega} d^n \mathbf{x} \frac{f(\mathbf{x})}{p_X(\mathbf{x})} p_X(\mathbf{x}).$$

Now, for any random variable  $Y$  and any function  $g$ , we know that

$$\frac{1}{N} \sum_{i=1}^N g(Y_i) \approx \langle g(Y) \rangle = \int_{-\infty}^{\infty} dy p_Y(y) g(y)$$



Hence

$$\left\langle \frac{f(\mathbf{X})}{p_X(\mathbf{X})} \right\rangle = \int_{\Omega} d^n \mathbf{x} \frac{f(\mathbf{x})}{p_X(\mathbf{x})} p_X(\mathbf{x}) = I$$

So, we can numerically estimate our integral by generating  $N$  points  $\mathbf{X}_i$  and take the average of  $f(\mathbf{X})/p_X(\mathbf{X})$ .

In doing so we will get an error which we can estimate by

$$\delta \approx \sigma \left( \frac{f(\mathbf{X})}{p_X(\mathbf{X})} \right) / \sqrt{N}$$

where the variance is given by  $\sigma^2(Y) = \langle Y^2 \rangle - \langle Y \rangle^2$ .

(cf. Simpsons rule  $\delta \propto 1/N^{4/d}$ )



Clearly if  $p_X(\mathbf{x}) = C|f(\mathbf{x})|$ , we get the smallest possible error (if  $f(x) > 0$  the error is zero).

However, with a bad choice of  $p_X$ , the variance and the error need not even be finite.

Numerically generating points directly according to  $p_X(\mathbf{x}) = C|f(\mathbf{x})|$  is in general difficult, and typically involves analytically solving the integral we want to estimate. But there are some tricks. . .



Normally we only have uniformly distributed (flat) random numbers available on the computer

$$p_R(r) = \begin{cases} 1 & 0 < r < 1 \\ 0 & \text{otherwise} \end{cases}$$

We can transform any distribution into any other by the a transformation using the cumulative distributions

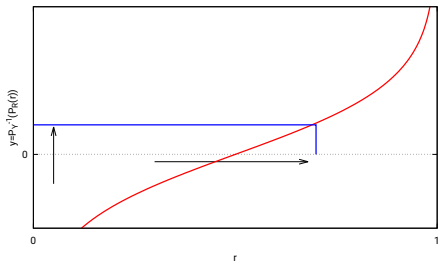
$$P_Y(y) = \int_{-\infty}^y dt p_Y(t) = \int_0^r dt p_R(t) = P_R(r) = r$$

as long as  $P_Y^{-1}(P_R(r))$  is a monotonically increasing function.

(If  $P_Y^{-1}(P_R(r))$  is not monotonous, we can divide up in intervals.)







Think of it as variable substitution:

$$\int_{y_{\min}}^{y_{\max}} p_Y(y) f(y) dy = \left\{ \begin{array}{l} P_Y(y) = r \\ \frac{dy}{dr} = \frac{1}{p_Y(y)} \\ P_Y(y_{\min}) = 0 \\ P_Y(y_{\max}) = 1 \end{array} \right\} = \int_0^1 f(P_Y^{-1}(r)) dr$$

What if  $P_Y^{-1}$  is hard to find ...



## The Accept/Reject Method

Assume we want to generate random variables,  $Y_i$ , according to some difficult distribution  $p_Y(y)$ . We already know how to generate according to some other distribution,  $p_{Y'}(y)$  such that  $Cp_{Y'}(y) \geq p_Y(y)$  everywhere.

1. Generate  $Y'$  according to  $p_{Y'}(y)$
2. Generate  $R$  according to a flat distribution
3.
  - ▶ If  $\frac{p_Y(Y')}{Cp_{Y'}(Y')} > R$  then accept  $Y = Y'$
  - ▶ otherwise reject  $Y'$  and goto 1

The accepted  $Y$  will be distributed according to  $p_Y(y)$ .  
 We need  $> 2C$  random numbers to get one  $Y$ .



## Multi-channel

Sometimes it is difficult to find an overestimate. But there are many tricks!

Assume  $p(x) \leq g(x) = \sum_i g_i(x)$  where we know how to generate random variables according to each  $g_i$ .

1. select  $i$  with relative probability  $A_i = \int g_i(x) dx$
2. select  $x$  according to  $g_i(x)$
3. throw away  $x$  and  $i$  with probability  $f(x) / \sum_i g_i(x)$

$$\int f(x) dx = \int \frac{f(x)}{g(x)} \sum_i g_i(x) = \sum A_i \int \frac{g_i(x) dx}{A_i} \frac{f(x)}{g(x)}$$

Alternatively we can divide up the integration region into sub-regions, where we can find a suitable overestimate in each.

Again we first choose region according to the integral of the overestimate, and then generate in there.



## How do we get random numbers?

There are ways of getting truly random numbers, but we will use (and actually prefer) **pseudo-random** numbers.

There are many algorithms around for producing pseudo-random numbers.

The simplest one is called **Linear congruential**:

- ▶ Pick integers  $a$ ,  $b$ ,  $m$ , and a seed  $R_0$
- ▶ generate random numbers according to

$$R_i = aR_{i-1} + b \quad (\text{mod } m)$$

**DON'T USE THIS**



## The Marsaglia Effect

Take successive  $d$ -tuplets from a congruential generator with  $t$  bits ( $m = 2^t$ ).

Interpret them as point coordinates in a  $d$ -dimensional hypercube.

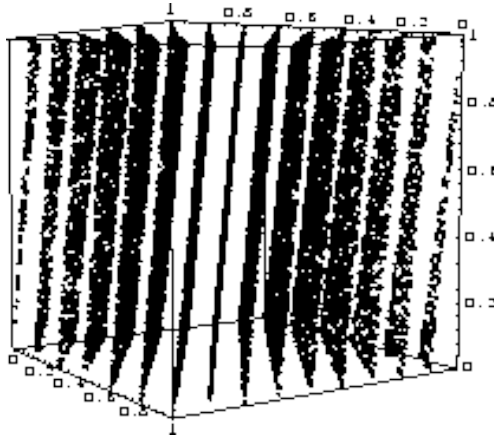
Then they all fall on at most  $(d!2^t)^{1/d}$  parallel hyperplanes.

$t$	$d = 3$	$d = 4$	$d = 6$	$d = 10$
16	73	35	19	13
32	2 953	566	120	41
48	1 190 86	9 065	766	126
64	4 801 280	145 055	4 866	382

**Disastrous** for any repetitive application.



# The Marsaglia Effect



Has lead to explosion of new tests and new generators.



Don't worry, there are several good pseudo random generators out there:

[http://en.wikipedia.org/wiki/List\\_of\\_random\\_number\\_generators](http://en.wikipedia.org/wiki/List_of_random_number_generators)



## Predicting an Observable

To calculate the expectation value of an observable,  $\mathcal{O}$ , in a  $pp \rightarrow X$  collision we need to evaluate an integral looking like

$$\langle \mathcal{O} \rangle = \sum_n \sum_{\mathbf{Q}} \int d^{4n} \mathbf{p} |\mathcal{M}_n(\mathbf{Q}, \mathbf{p})|^2 \mathcal{O}_n(\mathbf{Q}, \mathbf{p}) \Phi_n(\mathbf{p})$$

- ▶  $\mathbf{p}$  are the momenta of the  $n$  particles
- ▶  $\mathbf{Q}$  are their quantum numbers
- ▶  $\mathcal{M}$  is the matrix element
- ▶  $\Phi_n$  is the phase space density etc.





So now, all we need to do is to find a probability distribution  $p(n, \mathbf{Q}, \mathbf{p})$  such that

$$C p(n, \mathbf{Q}, \mathbf{p}) = |\mathcal{M}_n(\mathbf{Q}, \mathbf{p})|^2 \Phi_n(\mathbf{p})$$

Then we generate  $N$  points,  $(n_i, \mathbf{Q}_i, \mathbf{p}_i)$  according to this and get

$$\langle \mathcal{O} \rangle = \frac{C}{N} \sum_i^N \mathcal{O}_n(\mathbf{Q}_i, \mathbf{p}_i)$$

In the same way as we do when measuring the observable experimentally.

We are generating events. And we can measure several observables in one go. Life is simple!



## There are no free lunches

- ▶  $\mathcal{M}$  can typically only be calculated perturbatively to leading and maybe next-to-leading order for a small number of particles.
- ▶  $\Phi_n$  is not trivial
- ▶ finding  $p(n, \mathbf{Q}, \mathbf{p})$  may be **very** difficult



## Weighted vs. Unweighted Events

We can, of course, use any probability distribution and get

$$\langle \mathcal{O} \rangle = \frac{C}{N} \sum_i^N \frac{|\mathcal{M}_n(\mathbf{Q}_i, \mathbf{p}_i)|^2 \Phi_n(\mathbf{p}_i)}{p(n_i, \mathbf{Q}_i, \mathbf{p}_i)} \mathcal{O}_n(\mathbf{Q}_i, \mathbf{p}_i)$$

which means we get weighted events.

This is OK as long as the variance is not too big.



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This is OK as long as the variance is not too big.

We could also try the tricks we looked at before  
 (accept/reject, multi-channel, ...)

But do we even know what  $|\mathcal{M}_n(\mathbf{Q}_i, \mathbf{p}_i)|^2$  is?



## Factorization - Divide and conquer!

$$\langle \mathcal{O} \rangle = \sum_{n_q, \mathbf{Q}_q} \int d^{4n_q} \mathbf{q} |\mathcal{M}_{n_q}(\mathbf{Q}_q, \mathbf{q})|^2 \Phi_{n_q}(\mathbf{q}) \times \left[ \sum_{n_k, \mathbf{Q}_k} \int d^{4n_k} \mathbf{k} PS(\mathbf{Q}_q, \mathbf{q}; \mathbf{Q}_k, \mathbf{k}) \times \left\{ \sum_{n_p, \mathbf{Q}_p} \int d^{4n_p} \mathbf{p} H(\mathbf{Q}_k, \mathbf{k}; \mathbf{Q}_p, \mathbf{p}) \mathcal{O}_{n_p}(\mathbf{Q}_p, \mathbf{p}) \right\} \right]$$

- ▶  $\mathcal{M}$  now only gives a few partons
- ▶  $PS$  is a parton shower giving more partons with unit probability
- ▶  $H$  is hadronization and decays giving final state hadrons with unit probability



# Factorization

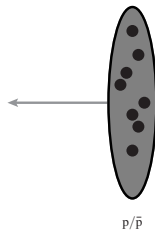
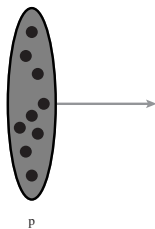
Relies on the factorization ansatz.

The cross section and main structure of the event is determined by the **hard** partonic sub process.

Parton showers and hadronization happens at lower (**softer**) scales and *dresses* the events without influencing the cross section.

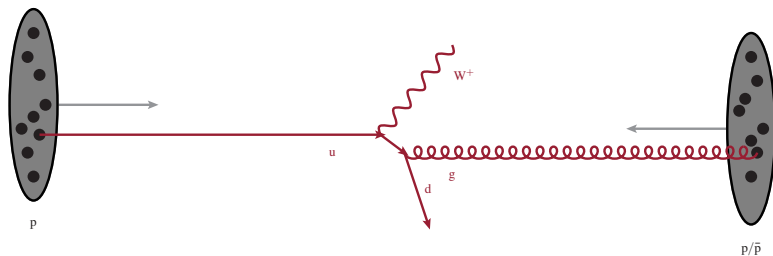


# The structure of a proton collision

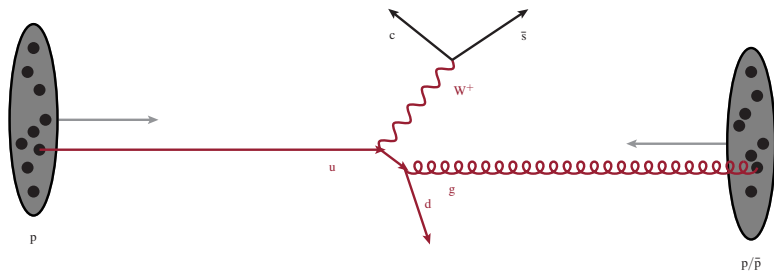




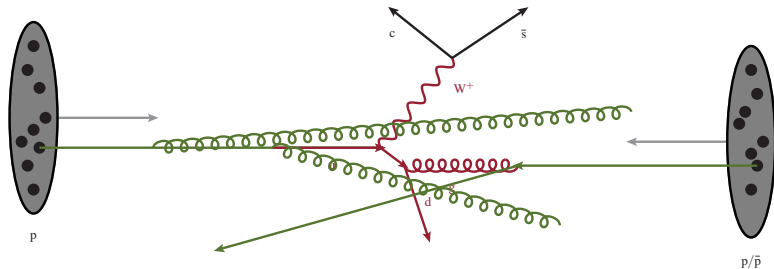
# The hard/primary scattering



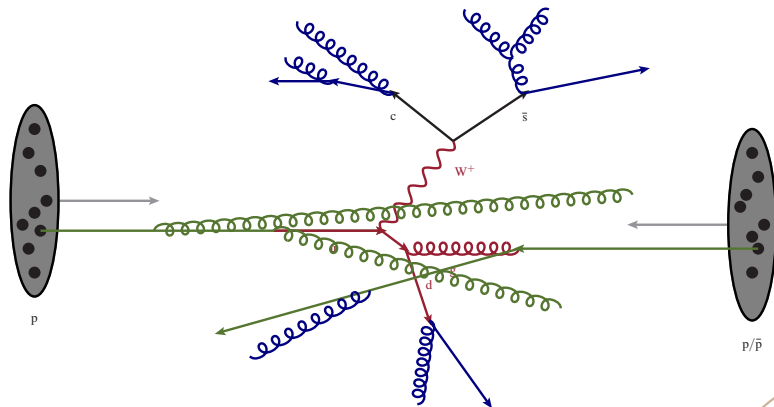
# Immediate decay of unstable elementary particles



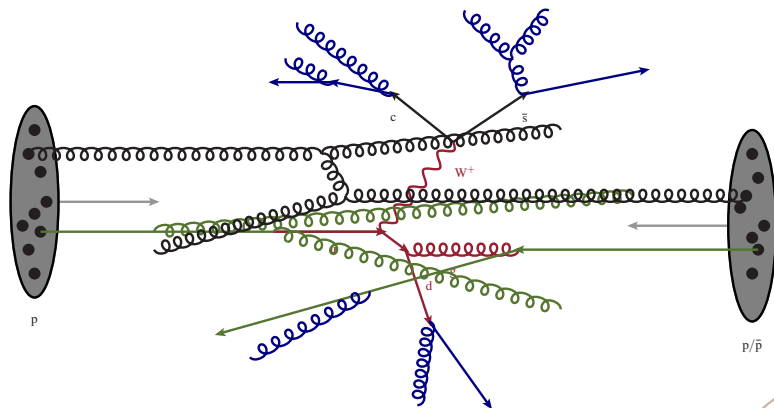
# Radiation from particles before primary interaction



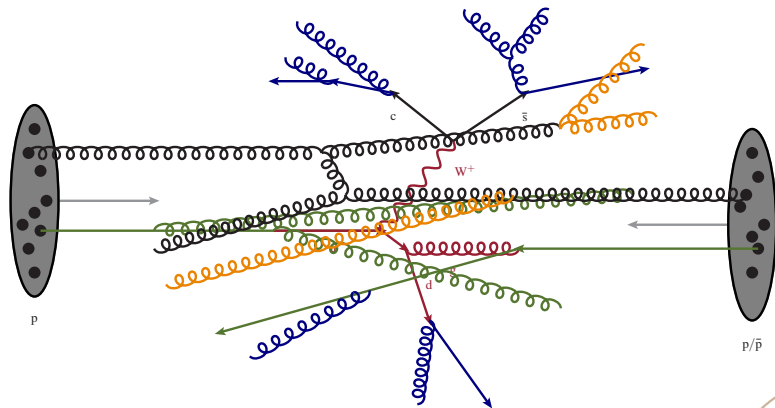
# Radiation from produced particles



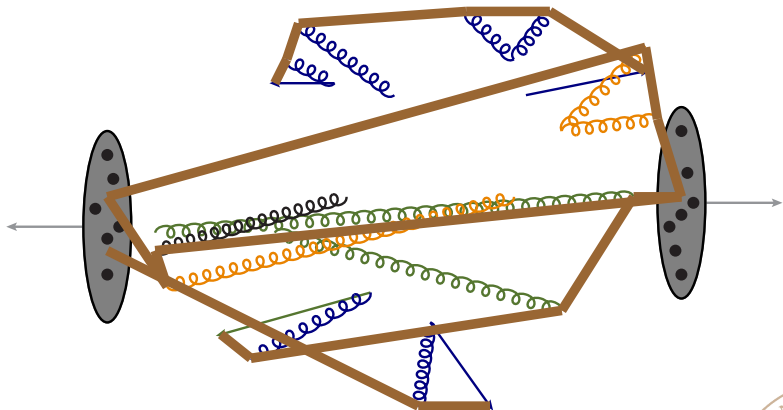
# Additional sub-scatterings



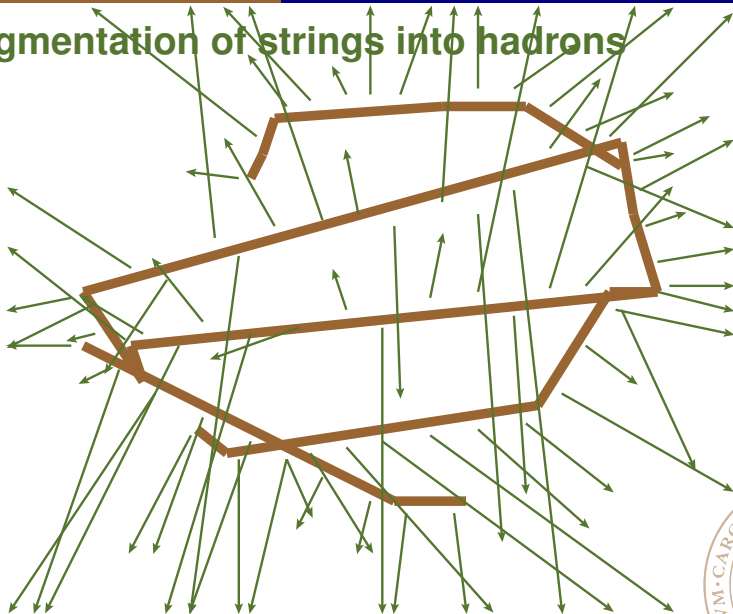
## ... with accompanying radiation



## Formation of *colour strings*

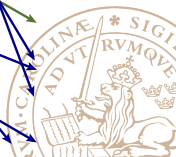
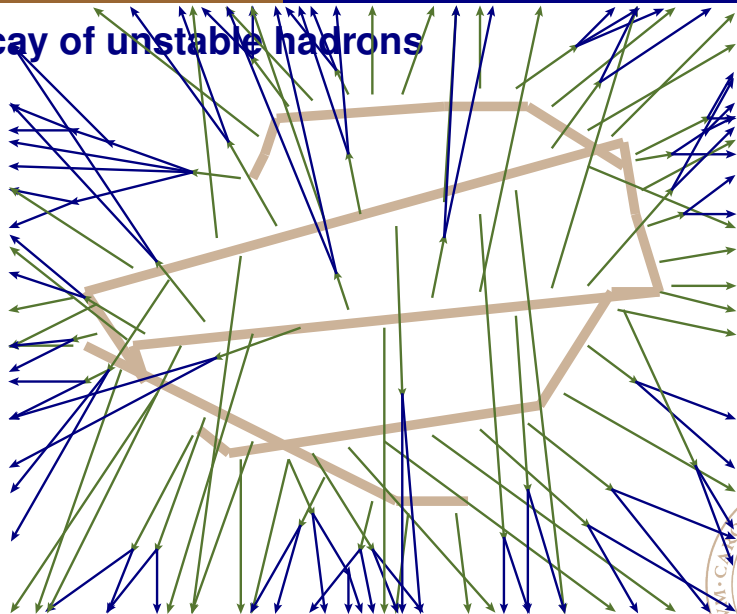


# Fragmentation of strings into hadrons





# Decay of unstable hadrons



# What happens at LHC? (13 TeV)

Total	100 mb
Non-diffractive	56 mb
Elastic	22 mb
Diffractive	22 mb
Jets $p_{\perp} > 150 \text{ GeV}$	220 nb
W+Z	200 nb
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<b>Higgs</b>	<b>30 pb</b>



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Almost everything at LHC is pure QCD



# Everything at the LHC is QCD

- ▶ Any measurement at the LHC requires understanding of QCD
- ▶ Electro-weak processes or BSM processes are easy (although sometimes tedious)
- ▶ Even **golden** signals such as  $H \rightarrow 4\mu$  are influenced by QCD
- ▶ Any observable prediction will have QCD corrections  
$$\langle \mathcal{O} \rangle = \sigma_0(1 + \alpha_s \mathcal{C}_1 + \alpha_s^2 \mathcal{C}_2 + \dots)$$
- ▶ Any signal will have a QCD background
- ▶ QCD is difficult



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Event Generators are all about QCD.



## Why is QCD difficult?

- ▶  $\alpha_s$  is not very small ( $\gtrsim 0.1$ )
- ▶ The gluon has a self-coupling and we get a lot of gluons
- ▶ Even if  $\alpha_s$  is small the phase space for emitting gluons is large. In any  $\alpha_s$  expansion the coefficients may be large.
- ▶ In the end we need hadrons, which are produced in a non-perturbative process.

We need **models** for parton showers and hadronization



# Questions!

- ▶ What are the questions the lecturer is trying to answer?
- ▶ Is he answering these questions satisfactorily?
- ▶ Which questions would you like the lecturer to answer?



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# Matrix Element Generation

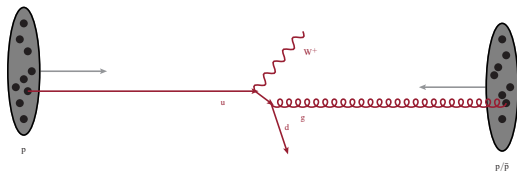
We always need to start with a  $2 \rightarrow n$  matrix element. This can in principle be obtained from the standard model (or BSM) Lagrange density in a straight-forward manner.

However,

- ▶ On tree-level we have divergencies if the scale ( $\sim p_{\perp}$ ) is small. Soft or collinear partons.
- ▶ Beyond leading order we get nasty loops and infinities
- ▶ If  $n$  is large, the number of diagrams grows factorially
- ▶ If  $n$  is large, it is difficult to find a suitable probability distribution for the momenta



## Simple 2 $\rightarrow$ 2 Matrix Elements



Can in principle be written down by hand from relevant Feynman diagrams.

$$\sigma = \int dx_1 dx_2 d\hat{t} f_i(x_1, Q^2) f_j(x_2, Q^2) \frac{d\hat{\sigma}_{ij}}{d\hat{t}}$$

With the parton densities sampled at a scale  $Q^2 \sim |\hat{t}| \sim p_{\perp}^2$ .



Also rather easy to generate as the integrand is fairly flat in

$$d \ln(x_1) d \ln(x_2) d \ln(p_{\perp}^2)$$

Note however that  $\hat{\sigma}$  may be divergent as  $\hat{t} \rightarrow 0$ .

Eg. Standard QCD ME:

$$\begin{aligned} \frac{\hat{\sigma}_{gg \rightarrow gg}}{d\hat{t}} = & \frac{9\pi\alpha_s^2}{4\hat{s}^2} \left( \frac{\hat{s}^2}{\hat{t}^2} + 2\frac{\hat{s}}{\hat{t}} + 3 + 2\frac{\hat{t}}{\hat{s}} + \frac{\hat{t}^2}{\hat{s}^2} \right. \\ & + \frac{\hat{u}^2}{\hat{s}^2} + 2\frac{\hat{u}}{\hat{s}} + 3 + 2\frac{\hat{s}}{\hat{u}} + \frac{\hat{s}^2}{\hat{u}^2} \\ & \left. + \frac{\hat{t}^2}{\hat{u}^2} + 2\frac{\hat{t}}{\hat{u}} + 3 + 2\frac{\hat{u}}{\hat{t}} + \frac{\hat{u}^2}{\hat{t}^2} \right) \end{aligned}$$



We clearly need a cutoff.

Typically this is given as a jet resolution scale, which for this simple process typically means a  $p_{\perp}$ -cut of some sort.

Eg. the  $k_{\perp}$ -algorithm:

Find the pair of particles with smallest

$$k_{\perp ij} = \frac{\min(k_{\perp i}, k_{\perp j})}{R} \sqrt{\Delta\phi_{ij}^2 + \Delta\eta_{ij}^2}$$

and cluster them together into one. Or if any  $k_{\perp i}$  is smaller cluster it to the beam.

Continue until all *clusters* have  $k_{\perp ij}$  and  $k_{\perp i}$  above some cut.

These remaining **jets** are then “close” to the original partons.



# BUT A JET IS NOT A PARTON



BUT A JET IS NOT A PARTON (or a jet)



## Higher Order Tree-Level Matrix Elements

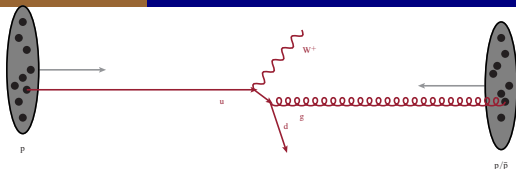
We can go on to higher order  $2 \rightarrow n$  Matrix Elements. This is in principle straight forward and can even be automated. However the number of diagrams grows  $\propto n!$  which makes generation of events forbiddingly slow for  $n \sim 10$ .

Remember also the difficulty in constructing a reasonable probability distribution for the momenta to sample the phase space, especially since there are divergencies everywhere.

Multi-channel sampling helps:

$$\sigma \propto \left| \sum_i \mathcal{M}_i \right|^2 = \sum_i |\mathcal{M}_i|^2 \frac{|\sum_j \mathcal{M}_j|^2}{\sum_j |\mathcal{M}_j|^2}$$





We can use a tree-level  $2 \rightarrow 2$  ME to predict an observable such as the rapidity distribution of a jet in a  $W$ -event.

We can try to get a better estimate by going to higher order tree-level MEs

$$\langle \mathcal{O} \rangle_{1j} = \sigma_{\rightarrow W+1j}(\mu) \otimes \mathcal{O}(W+j)$$

$$\langle \mathcal{O} \rangle_{2j} = \sigma_{\rightarrow W+2j}(\mu) \otimes \mathcal{O}(W+j)$$

$$\langle \mathcal{O} \rangle_{3j} = \sigma_{\rightarrow W+3j}(\mu) \otimes \mathcal{O}(W+j)$$

$\vdots$

Where we use some jet-resolution scale  $\mu$  to cut off divergencies.





But we cannot simply add these together, since each cross section is **inclusive**

The tree-level  $ab \rightarrow W + 1j$  matrix element gives the cross section for the production of a  $W$  plus **at least one jet**.

Hence it includes also a part of the tree-level  $ab \rightarrow W + 2j$  matrix element.



# Next-to-Leading Order

To correctly sum  $W + 1j$  and  $W + 2j$  contributions to an observable, we need to add virtual contributions to the generated  $W + 1j$  states. In that way we get a consistent expansion of the observable.

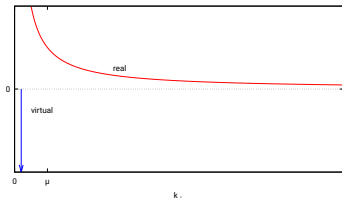
$$\langle \mathcal{O} \rangle_{1j} = \alpha_s \mathcal{C}_{11}(\mu) + \alpha_s^2 \mathcal{C}_{12}(\mu)$$

$$\langle \mathcal{O} \rangle_{2j} = \alpha_s^2 \mathcal{C}_{22}(\mu)$$

$$\langle \mathcal{O} \rangle_{\text{NLO}} = \langle \mathcal{O} \rangle_{1j} + \langle \mathcal{O} \rangle_{2j}$$



Here the jet resolution scale  $\mu$  is essential, since the **virtual corrections** are infinite and negative. But if we add together the **1j virtual terms** and the **unresolved 2j contributions**, (the contributions below  $\mu$ ) the sum,  $\alpha_s^2 C_{12}(\mu)$  is finite.



Today there are several NLO generators available.

They produce few-parton events and you can measure jet observables

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But what happens if  $\Delta\phi < 120^\circ$ ?





- ▶ Leading order is the first order in  $\alpha_s$  which gives a non-zero result for a given observable.
- ▶ If NLO corrections are large, we need NNLO.
- ▶ However, chances are that we have a poorly converging series in  $\alpha_s$ .
- ▶ This means we need to resum.



## All-Order Resummation

Rather than calculating a few terms in the  $\alpha_s$  expansion exactly, we can try to approximate **all** terms.

It turns out that if we just consider the leading divergent part of the cross section, everything exponentiates

$$\begin{aligned}\sigma_{0j} &= C_{00} + \alpha_s C_{01} + \alpha_s^2 C_{02} + \dots \approx C_{00} \exp(\alpha_s C'_{01}/C_{00}) \\ \sigma_{1j} &= \alpha_s C_{11} + \alpha_s^2 C_{12} + \alpha_s^3 C_{13} + \dots \approx \alpha_s C_{11} \exp(\alpha_s C'_{12}/C_{11}) \\ &\vdots\end{aligned}$$

Even if the coefficients diverge as  $\mu \rightarrow 0$  the exponentiation is finite.



The resummation corresponds to obtaining the **leading logarithmic** contributions to the coefficients

$$\propto \alpha_s^n \ln(\mu)^{2n}$$

This can be done analytically even to next-to-leading log  $\propto \alpha_s^n \ln(\mu)^{2n-1}$  and higher.

Or approximately numerically by using parton showers...



# Summary I

Event generators may feel like black boxes, with a lot of magic and trickery going on inside and a lot of incomprehensible switches and knobs on the outside. But in fact, all they do is to numerically integrate a complicated integral using well-known numerical methods.

However, the function we want to integrate is **complicated**.



# Summary I

Event generators may feel like black boxes, with a lot of magic and trickery going on inside and a lot of incomprehensible switches and knobs on the outside. But in fact, all they do is to numerically integrate a complicated integral using well-known numerical methods.

However, the function we want to integrate is **complicated**.



# Questions!



# Outline of Lectures

- ▶ Lecture I: Basics of Monte Carlo methods, the event generator strategy, matrix elements, LO/NLO, . . .
- ▶ Lecture II: Parton showers, initial/final state, (matching/merging), hadronization, decays. . . .
- ▶ Lecture III: Minimum bias, multi-parton interactions, pile-up, summary of general purpose event generators, Idots
- ▶ Lecture IV: Protons vs. heavy ions, Glauber calculations, initial/final-state interactions, . . .

Buckley et al. (MCnet collaboration), *Phys. Rep.* **504** (2011) 145.

