



Particle Physics Data Analysis Tools

Gholamhossein Haghight

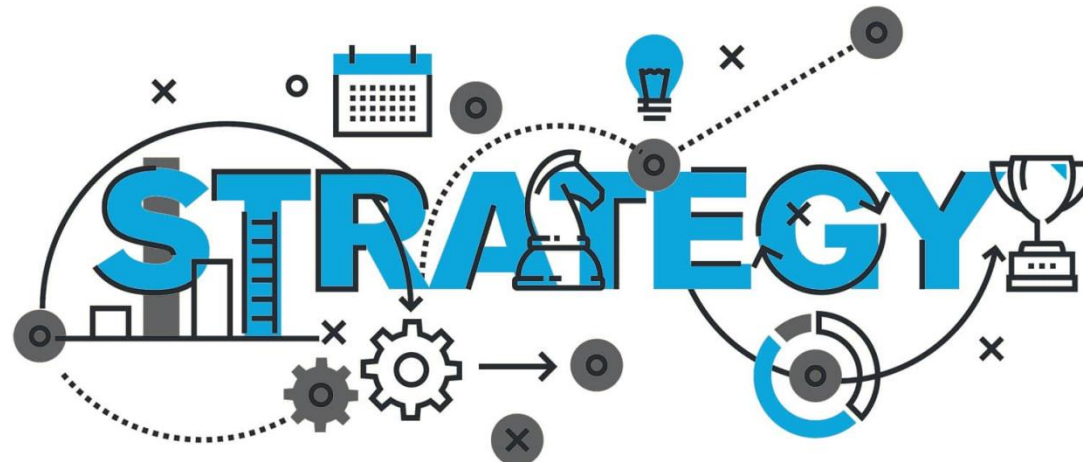
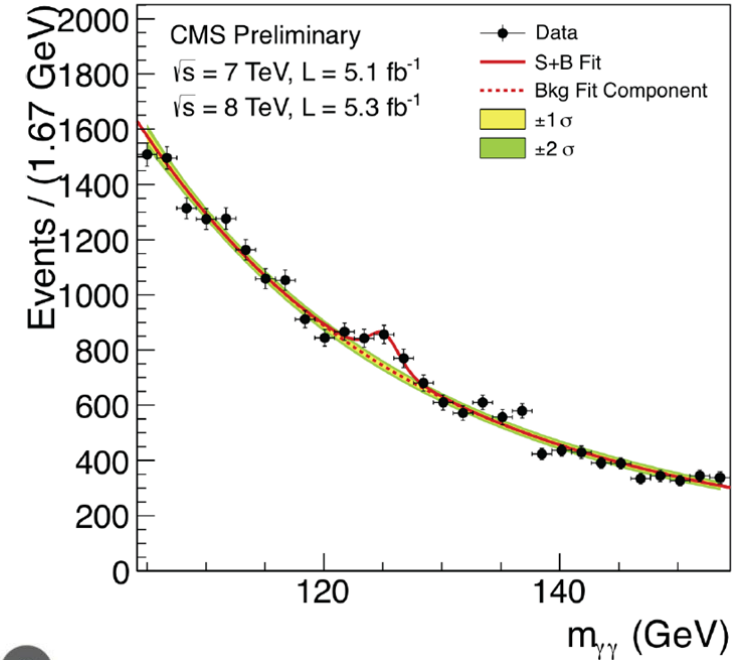
School of Particles and Accelerators, IPM

4th IPM Workshop on
Experimental Particle Physics

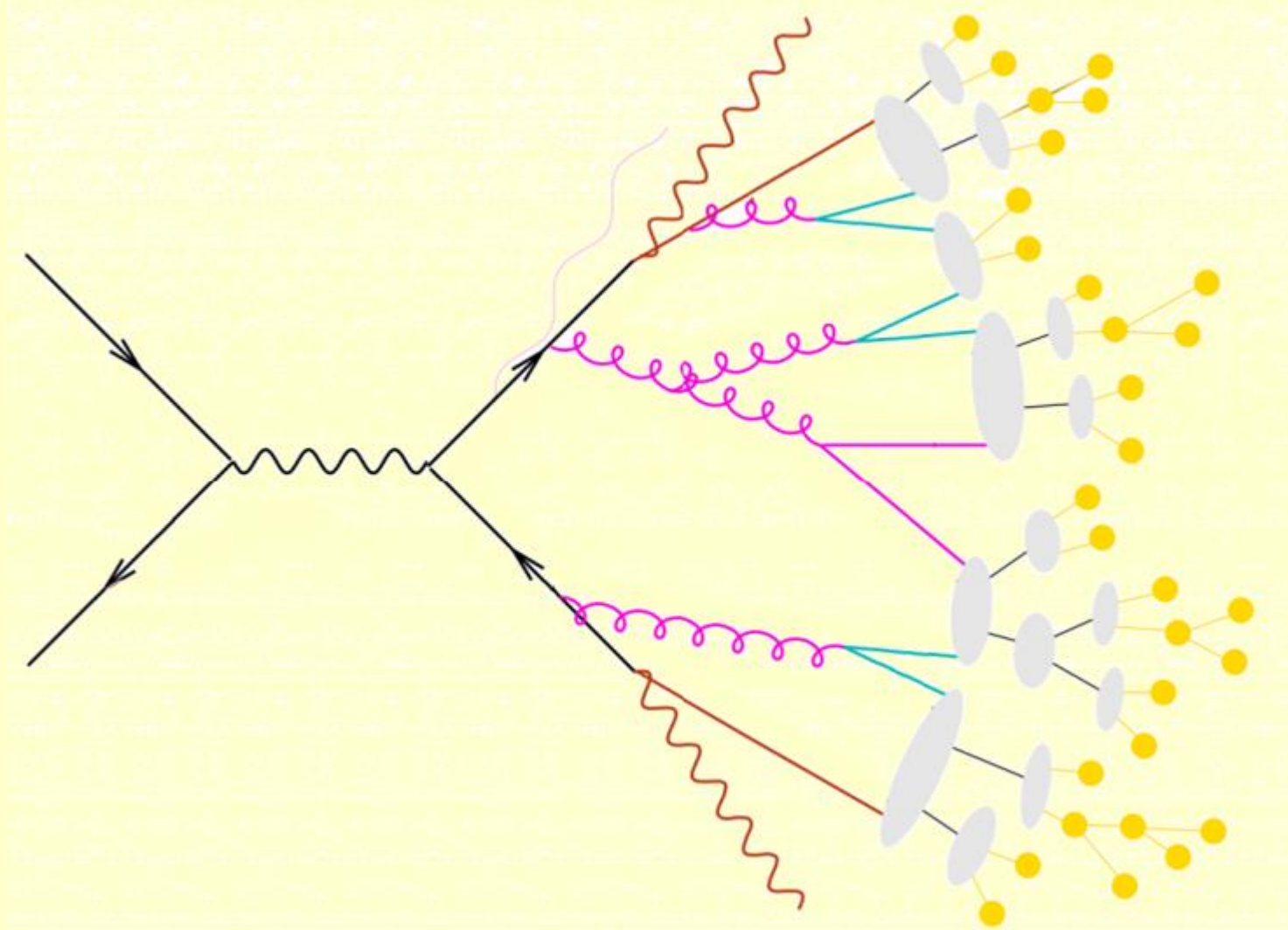
27 Dec 2023

Strategy for particle physics searches

- ❖ Find *excess* over SM background
- ❖ Identify *models* compatible with excess
- ❖ Look for *predicted excesses* in other channels
- ❖ Determine *underlying model*



Simplified LHC event

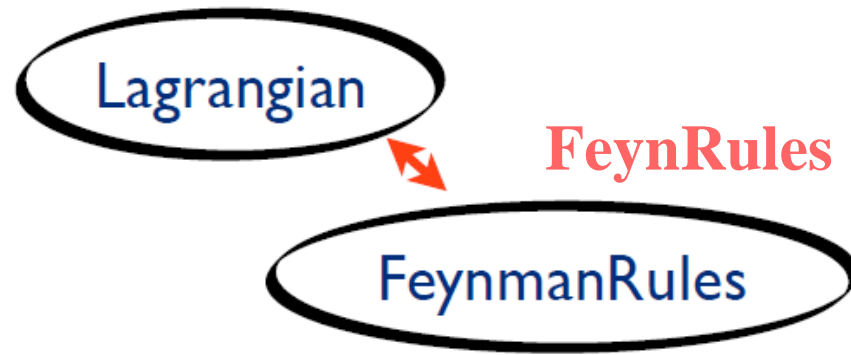


From theory to detector

Lagrangian

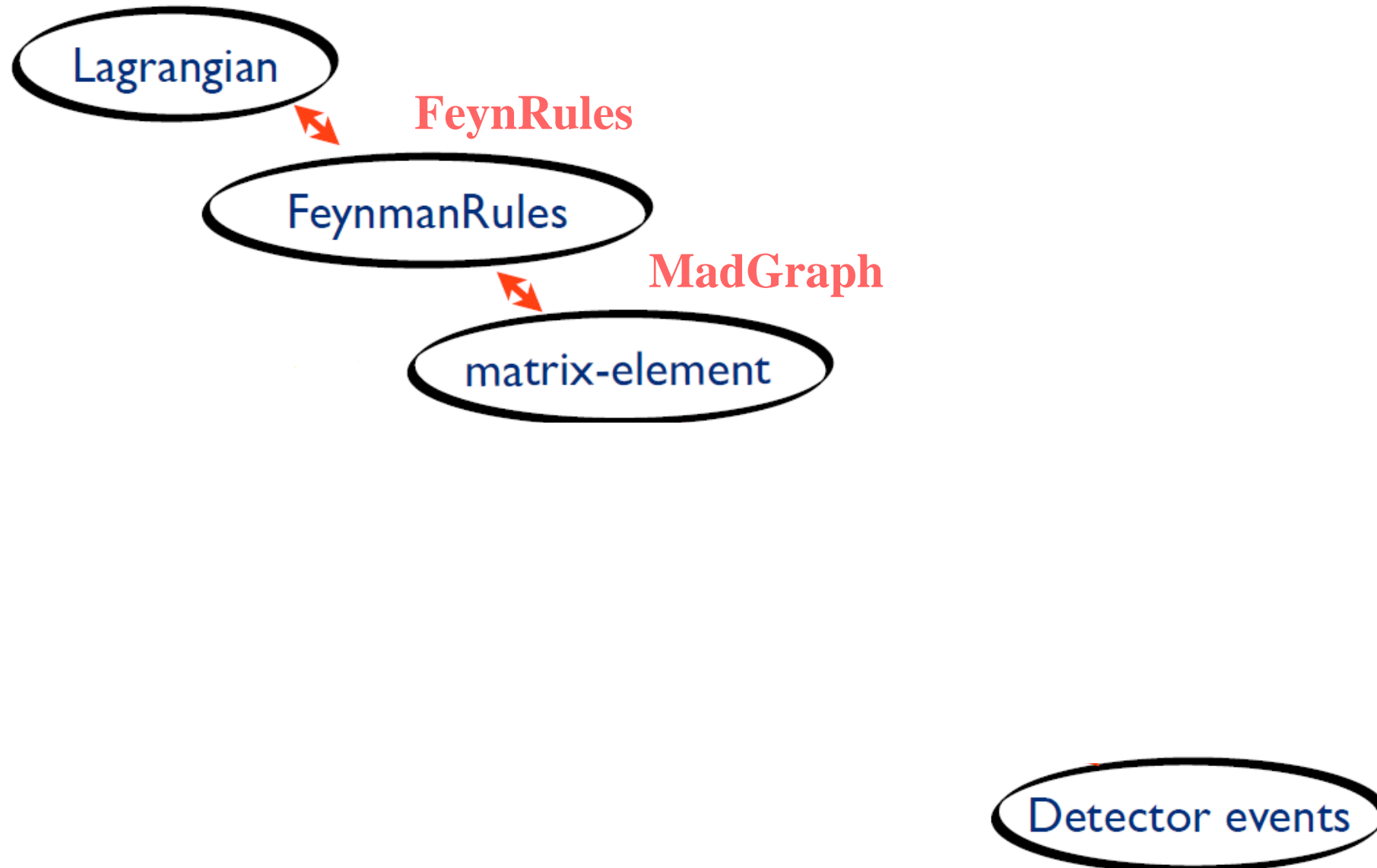
Detector events

From theory to detector

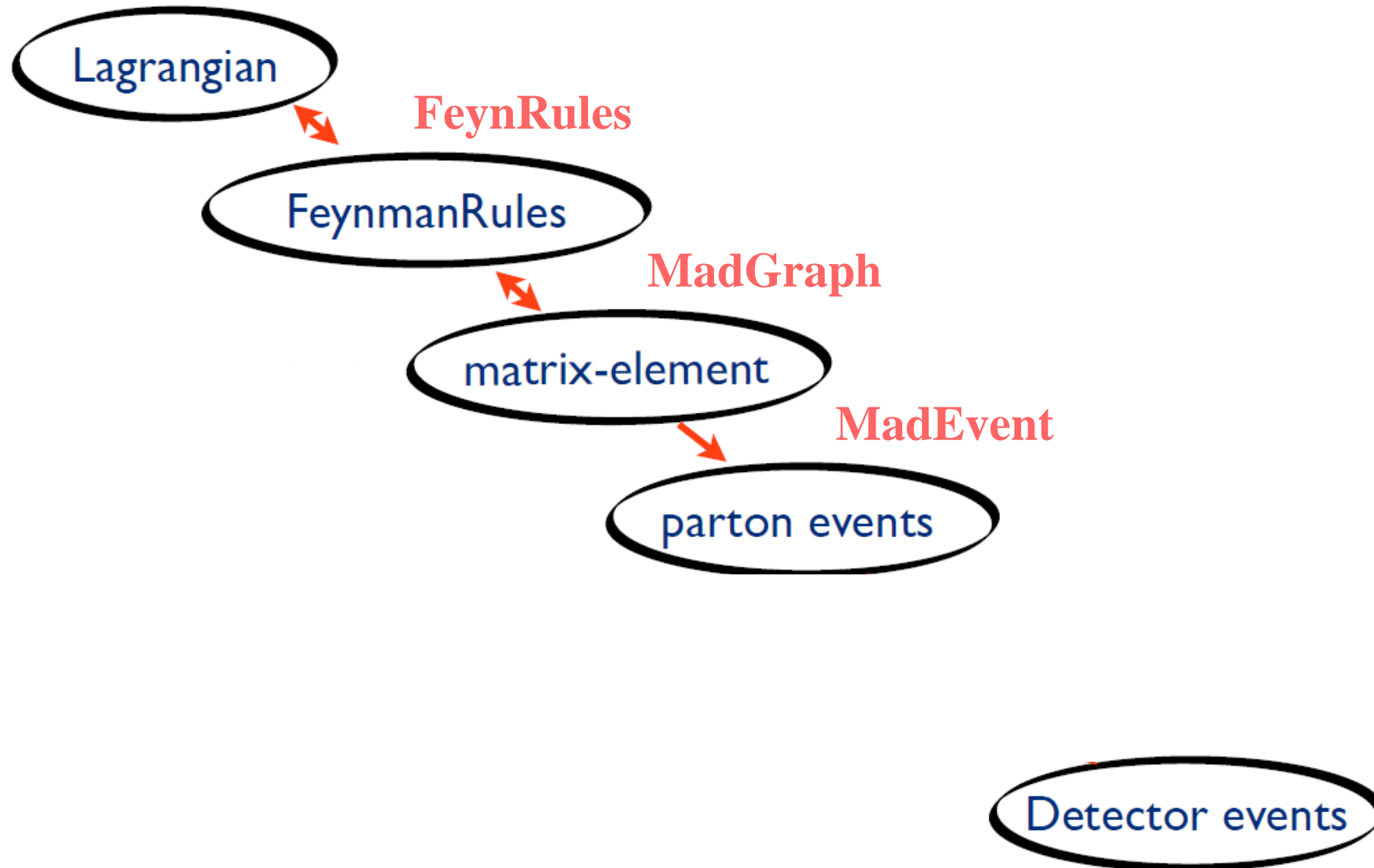


Detector events

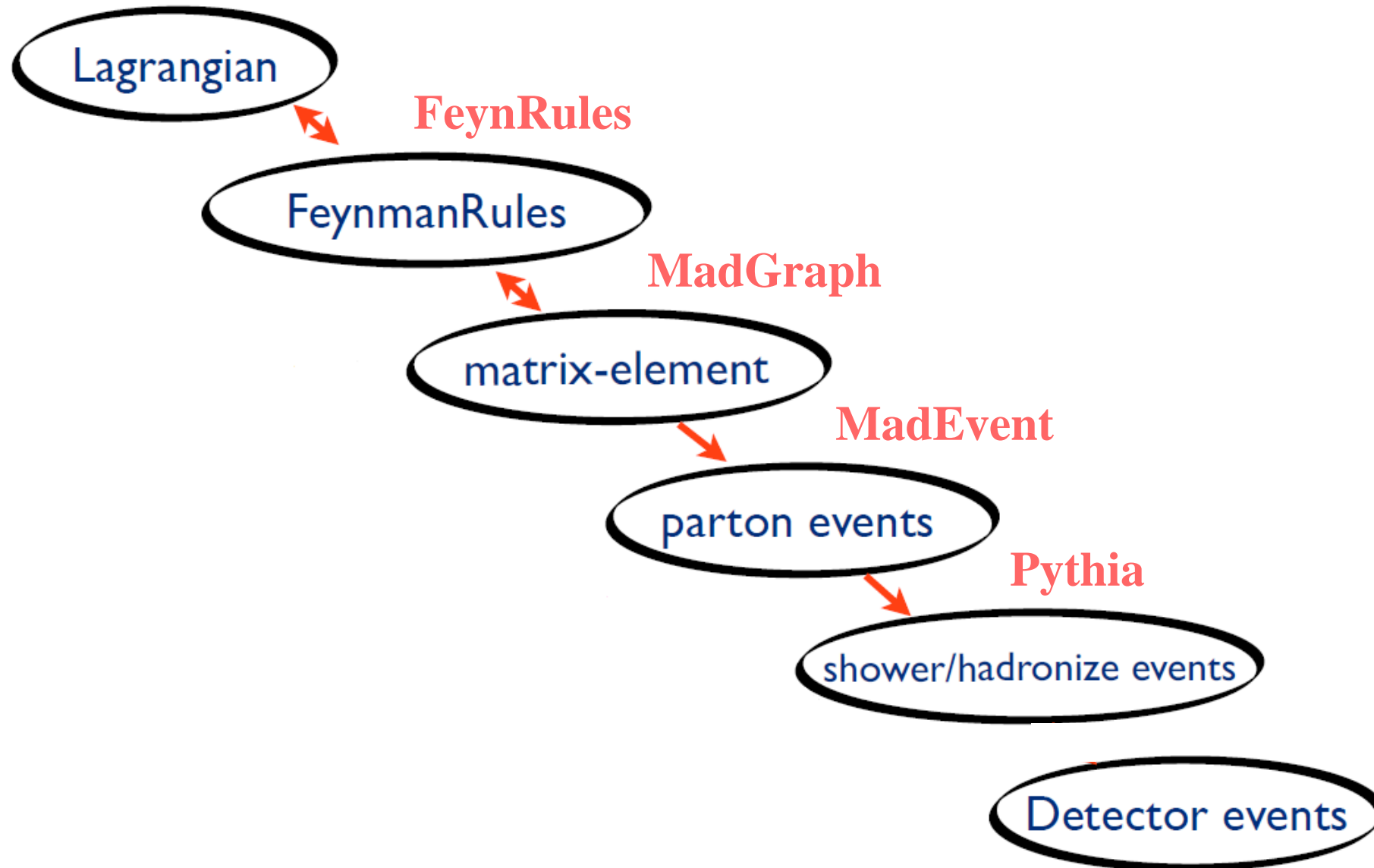
From theory to detector



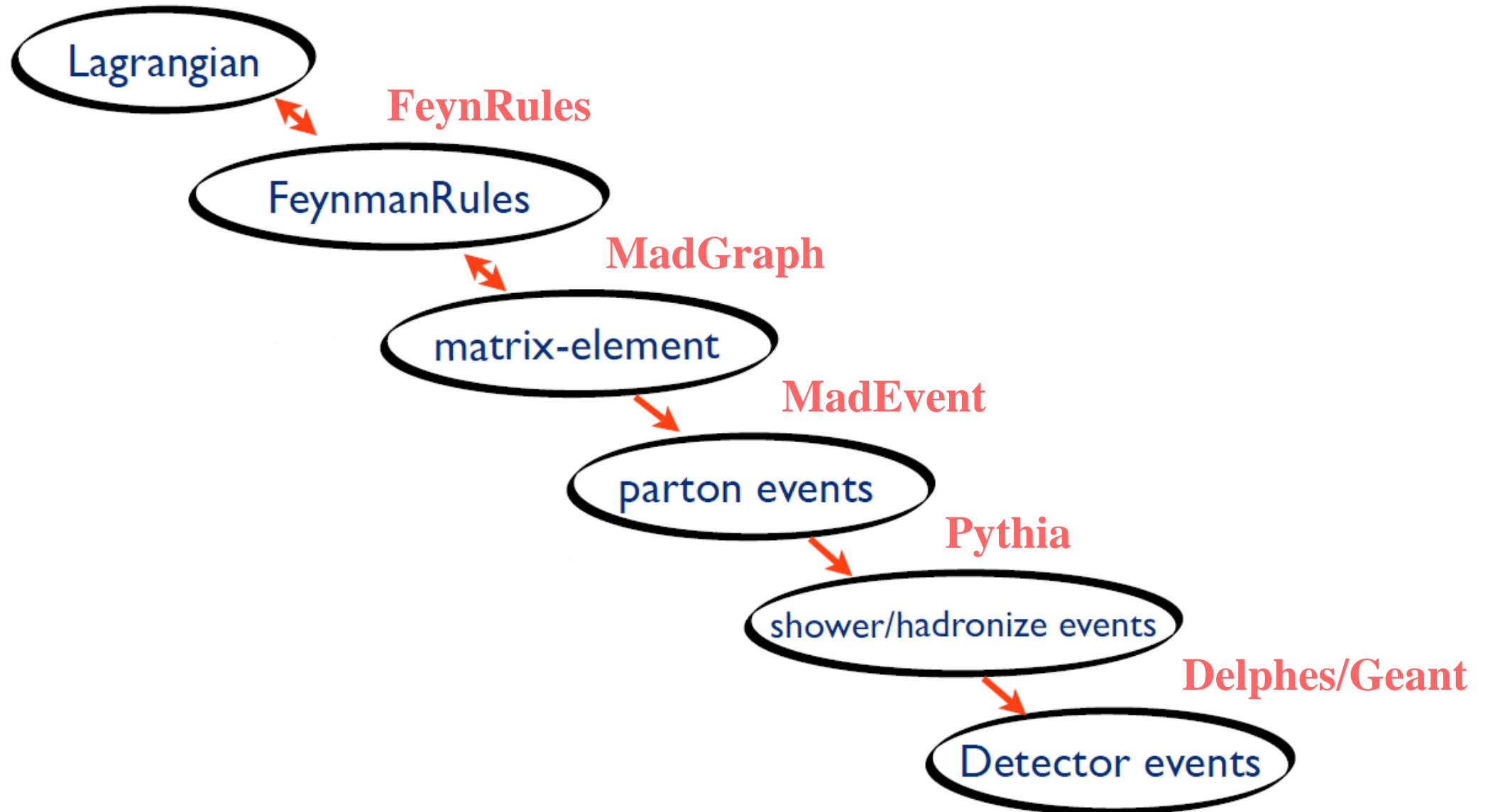
From theory to detector



From theory to detector

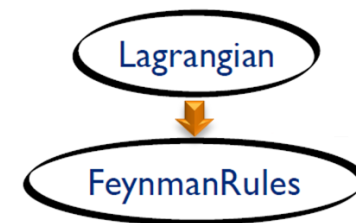


From theory to detector





- ❖ *Mathematica package to derive Feynman rules from a Lagrangian*
- ❖ *Available at feynrules.irmp.ucl.ac.be*



Gauge symmetries Particles Parameters Lagrangian

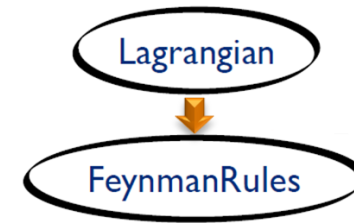
```
M$GaugeGroups = {
  U1Y == {
    Abelian      -> True
  }
  SU2L == {
    Abelian      -> False,
    CouplingConstant -> gw,
    GaugeBoson   -> Wi,
    StructureConstant -> Eps,
    Representations -> {Ta, SU2D},
    Definitions   -> {Ta[a_1, a_2, a_3] := Eps[a_1, a_2, a_3]}
  }
  SU3C == {
    Abelian      -> False,
    CouplingConstant -> gs,
    GaugeBoson   -> G,
    StructureConstant -> f,
    Representations -> {T, Colour},
    SymmetricTensor -> dSUN
  }
  [...]
}
```

FeynRules

[arXiv:1310.1921]



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Gauge symmetries

Particles

Parameters

Lagrangian

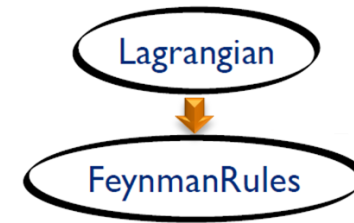
```
M$ClassesDescription = {
  V[1] == {
    ClassName      -> A,
    SelfConjugate
    Mass
    Width
    ParticleName
    PDG
    Propagator
    Propagator
    Propagator
    FullName
  }, [...]
  V[4] == {
    Class
    Self
    Indi
    Mass
    Width
    Part
    PDG
    Prop
    Prop
    Prop
    Full
  }, [...]
  F[1] == {
    ClassName      -> vl,
    ClassMembers  -> {ve,vm,vt},
    Indices        -> {Index[Generation]},
    FlavorIndex    -> Generation,
    SelfConjugate  -> False,
    Mass           -> 0,
    Width          -> 0,
    QuantumNumbers -> {LeptonNumber -> 1},
    PropagatorLabel -> {"v", "ve", "vm", "vt"} ,
    PropagatorType -> S,
    PropagatorArrow -> Forward,
    PDG            -> {12,14,16},
    ParticleName   -> {"ve","vm","vt"},
    AntiParticleName -> {"ve~","vm~","vt~"},
    FullName       -> {"Electron-neutrino", "Mu-neutrino",
                      "Tau-neutrino"}
  }, [...]
```

FeynRules

[arXiv:1310.1921]



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Gauge symmetries

Particles

Parameters

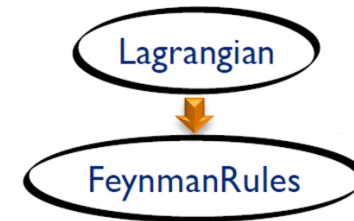
Lagrangian

```
M$Parameters = {
  aS == {
    ParameterType -> External,
    BlockName     -> SMINPUTS,
    OrderBlock    -> 3,
    Value         -> 0.1184,
    InteractionOrder -> {QCD,2},
    TeX           -> Subscript[\[Alpha],s],
    Description   -> "Strong coupling constant at the Z pole"
  },
  gs == {
    ParameterType -> Internal,
    Value         -> Sqrt[4 PI aS],
    InteractionOrder -> {QCD,1},
    TeX           -> Subscript[g,s],
    ParameterName -> G,
    Description   -> "Strong coupling constant at the Z pole"
  }, [...]
```

```
CKM == {
  ParameterType -> Internal,
  Indices       -> {Index[Generation], Index[Generation]},
  Unitary       -> True,
  Value         -> {CKM[1,1] -> Cos[cabi], CKM[1,2] -> Sin[cabi], CKM[1,3] -> 0,
                   CKM[2,1] -> -Sin[cabi], CKM[2,2] -> Cos[cabi], CKM[2,3] -> 0,
                   CKM[3,1] -> 0, CKM[3,2] -> 0, CKM[3,3] -> 1},
  TeX           -> Superscript[V,CKM],
  Description   -> "CKM-Matrix"}
```



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Gauge symmetries

Particles

Parameters

Lagrangian

```
LFermions := Block[{mu},  
  ExpandIndices[I*(  
    QLbar.Ga[mu].DC[QL, mu] + LLbar.Ga[mu].DC[LL, mu] +  
    uRbar.Ga[mu].DC[uR, mu] + dRbar.Ga[mu].DC[dR, mu] +  
    lRbar.Ga[mu].DC[lR, mu]), FlavorExpand->{SU2W, SU2D}]
```

Write **UFO**[LSM] →

UFO becoming the standard

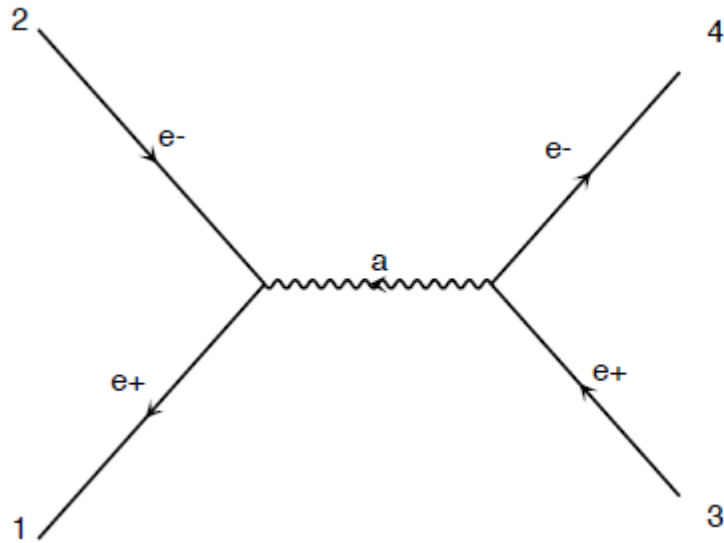
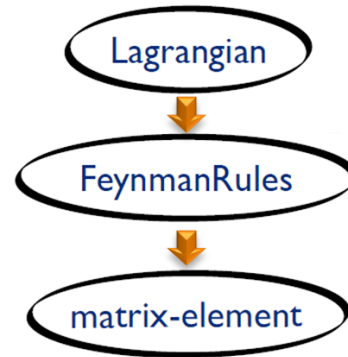




Computing amplitudes with *HELAS*

(*HELicity Amplitude Subroutine*)

Evaluate \mathcal{M} for fixed helicity of external particles



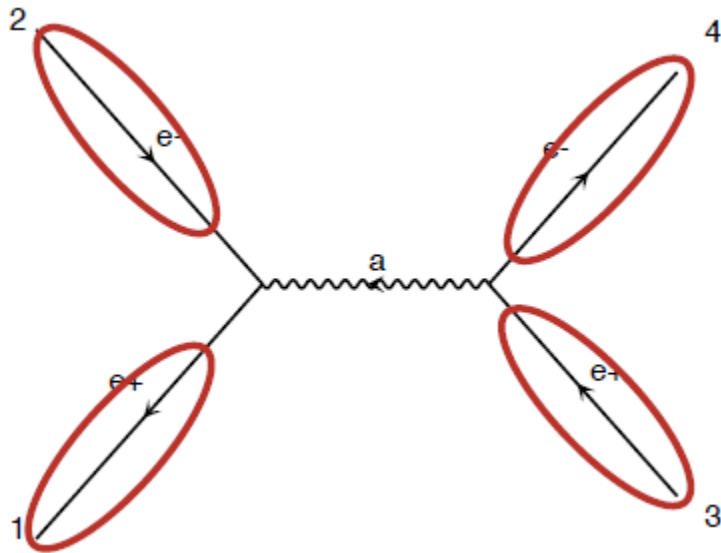
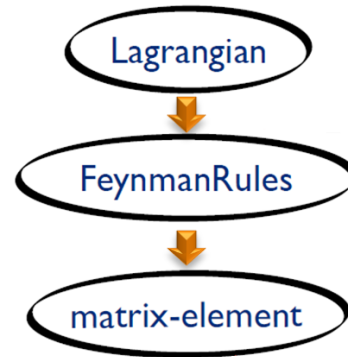
$$\mathcal{M} = \bar{u} \gamma^\mu v P_{\mu\nu} \bar{u} \gamma^\nu v$$



Computing amplitudes with *HELAS*

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$$\mathcal{M} = \bar{u} \gamma^\mu v P_{\mu\nu} \bar{u} \gamma^\nu v$$

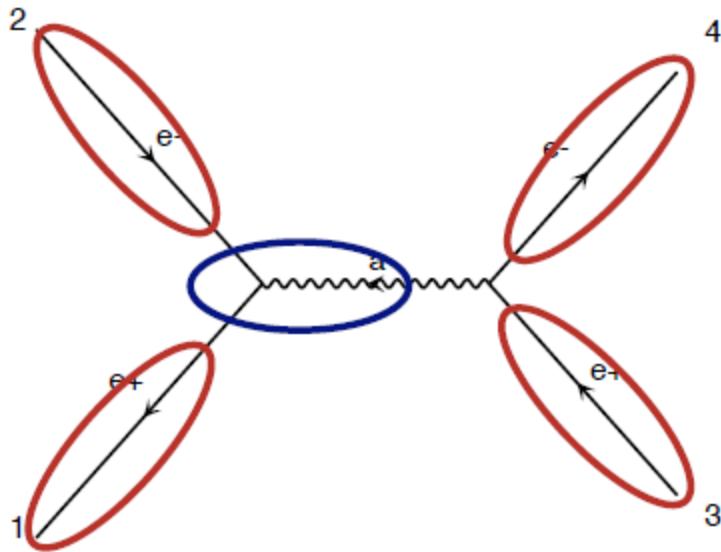
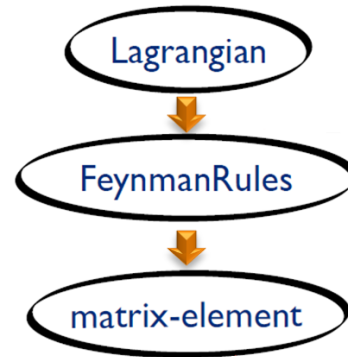
Particles



Computing amplitudes with *HELAS*

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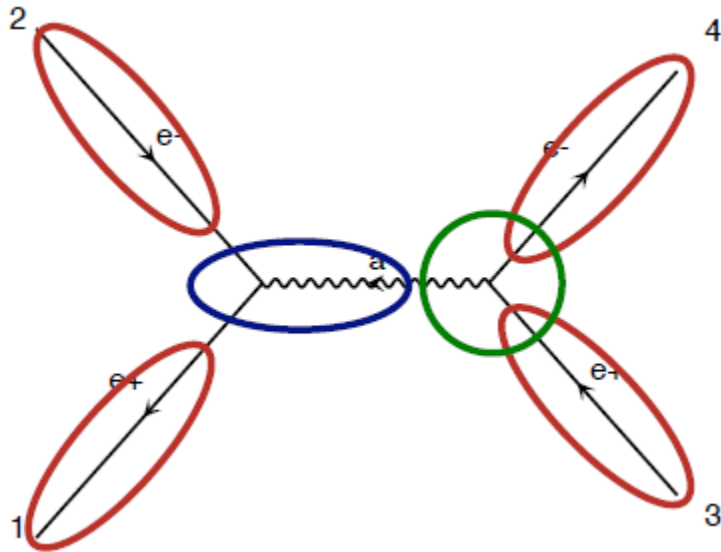
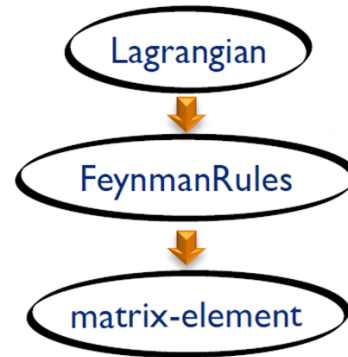
$$\mathcal{M} = \bar{u} \gamma^\mu v \not{P}_{\mu 1} \bar{u} \gamma^\nu v$$

Particles
Propagators



Computing amplitudes with *HELAS* (*HELicity Amplitude Subroutine*)

Evaluate \mathcal{M} for fixed helicity of external particles



$$\mathcal{M} = \bar{u} \gamma^\mu v \not{P}_{\mu\nu} \bar{u} \gamma^\nu v$$

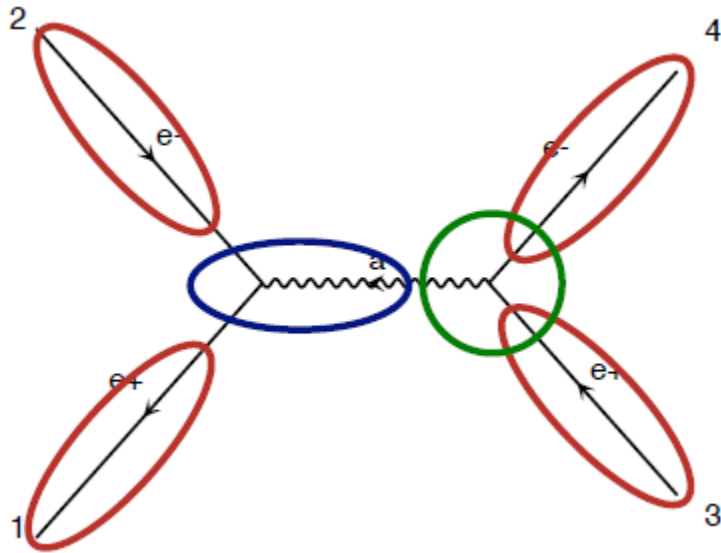
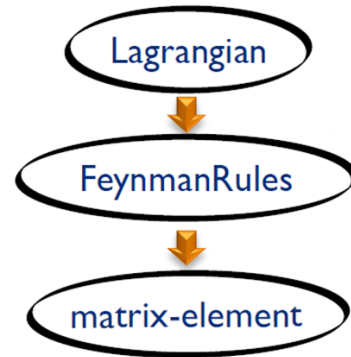
Particles
Propagators
Amplitude

- Helicity amplitude routines needed for the Standard Model, MSSM, ..., in hand-written library
- Any new Lorentz structure needs addition by hand ➡ restriction on types of models that could be implemented in MadGraph



Computing amplitudes with **HELAS** (**HELicity Amplitude Subroutine**)

Evaluate \mathcal{M} for fixed helicity of external particles



$$\mathcal{M} = \bar{u} \gamma^\mu v P_{\mu\nu} \bar{u} \gamma^\nu v$$

Particles
Propagators
Amplitude



(Automatic Libraries Of Helicity Amplitudes)

translates a **UFO** Lorentz structure into pseudo-**HELAS** subroutine

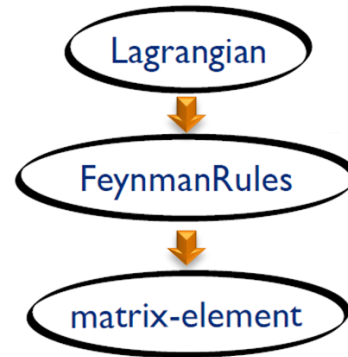


Weighting experimental events with MadWeight

$$P(\mathbf{x}, \alpha) =$$



*Probability of observing \mathbf{x}
predicted by the model α
 \mathbf{x} : experimental measurements*





Weighting experimental events with MadWeight

$$P(\mathbf{x}, \alpha) =$$

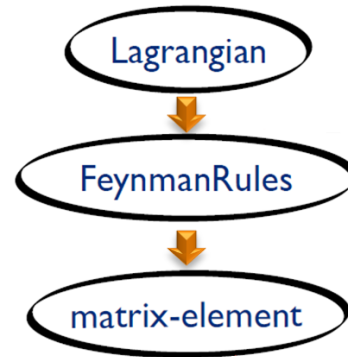


*Probability of observing \mathbf{x}
predicted by the model α
 \mathbf{x} : experimental measurements*

$$|M_\alpha|^2(\mathbf{x})$$

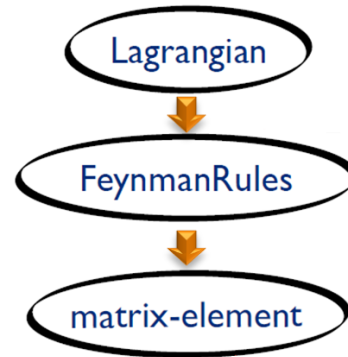


Squared matrix element





Weighting experimental events with MadWeight



$$P(\mathbf{x}, \alpha) =$$



Probability of observing \mathbf{x}
predicted by the model α
 \mathbf{x} : experimental measurements

$$|M_\alpha|^2(\mathbf{y}) W(\mathbf{x}, \mathbf{y})$$



Squared matrix element



Resolution function
 \mathbf{y} : partonic momenta
(experimental extraction)



Weighting experimental events with MadWeight

$$P(\mathbf{x}, \alpha) = \frac{1}{\sigma} \int d\phi(\mathbf{y})$$

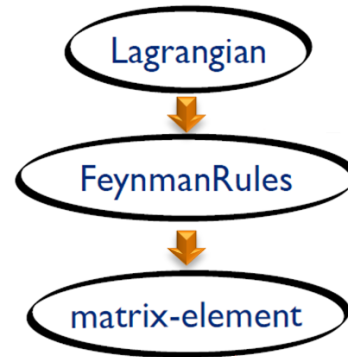
Partonic phase-space measure

Probability of observing \mathbf{x}
predicted by the model α
 \mathbf{x} : experimental measurements

$$|M_\alpha|^2(\mathbf{y}) W(\mathbf{x}, \mathbf{y})$$

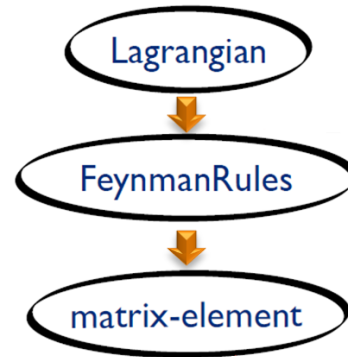
Squared matrix element

Resolution function
 \mathbf{y} : partonic momenta
(experimental extraction)





Weighting experimental events with MadWeight



$$P(\mathbf{x}, \alpha) = \frac{1}{\sigma} \int d\phi(\mathbf{y}) dw_1 dw_2 f_1(w_1) f_2(w_2) |M_\alpha|^2(\mathbf{y}) W(\mathbf{x}, \mathbf{y})$$

Partonic phase-space measure

Squared matrix element

*Probability of observing \mathbf{x}
predicted by the model α*

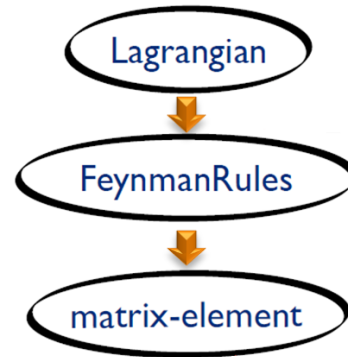
*Parton Distribution
Functions*

*Resolution function
 \mathbf{y} : partonic momenta
(experimental extraction)*

\mathbf{x} : experimental measurements



Weighting experimental events with MadWeight



$$P(\mathbf{x}, \alpha) = \frac{1}{\sigma} \int d\phi(\mathbf{y}) dw_1 dw_2 f_1(w_1) f_2(w_2) |M_\alpha|^2(\mathbf{y}) W(\mathbf{x}, \mathbf{y})$$

Partonic phase-space measure

Squared matrix element

Probability of observing \mathbf{x}
predicted by the model α
 \mathbf{x} : experimental measurements

Parton Distribution
Functions

Resolution function
 \mathbf{y} : partonic momenta
(experimental extraction)

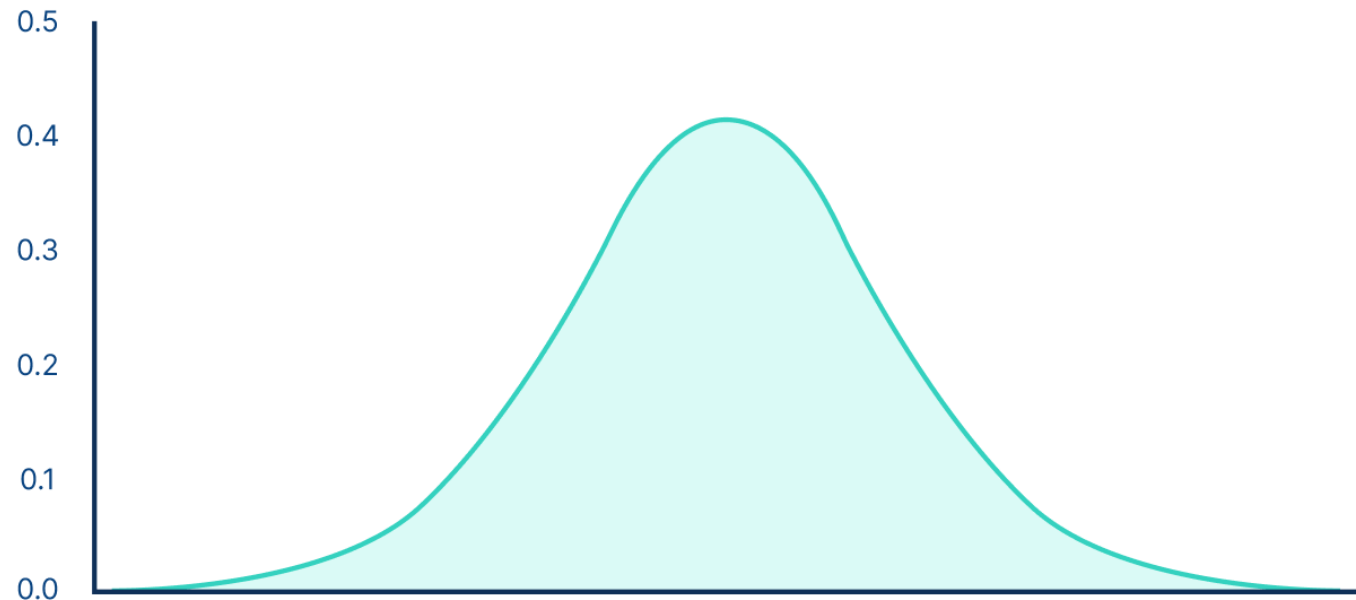
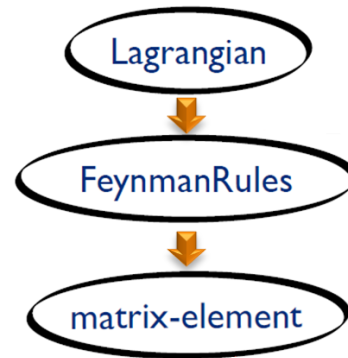
Peaks

$$W(\mathbf{x}, \mathbf{y}) \approx \prod_{i=E,\phi,\eta} \frac{1}{\sqrt{2\pi}\sigma_i} e^{-\frac{(x_i - y_i)^2}{2\sigma_i^2}}$$

Propagators: $\frac{1}{|q^2 - M^2 + iM\Gamma|^2}$



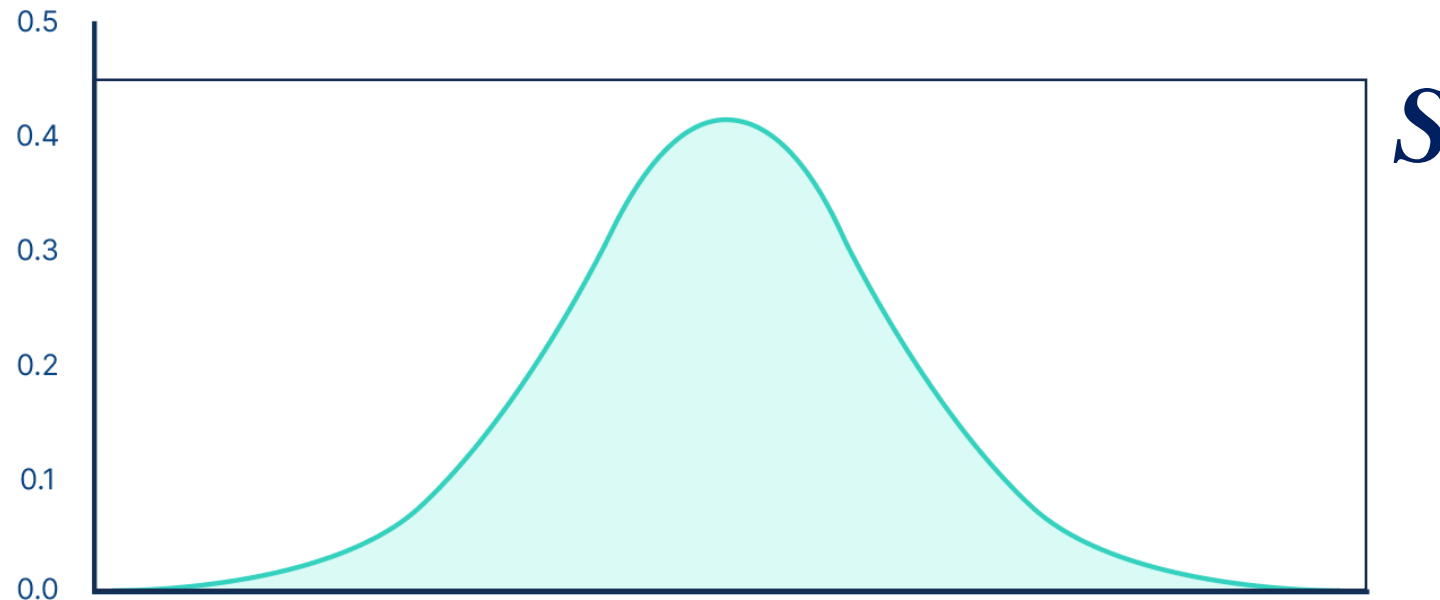
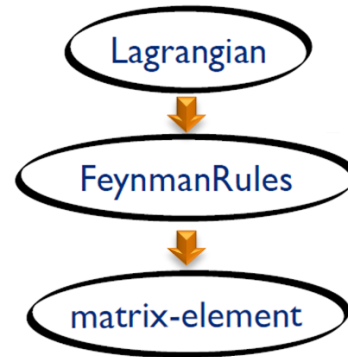
Weighting experimental events with MadWeight Monte-Carlo integration



*Monte Carlo (MC) method:
a method to obtain deterministic results from random values*

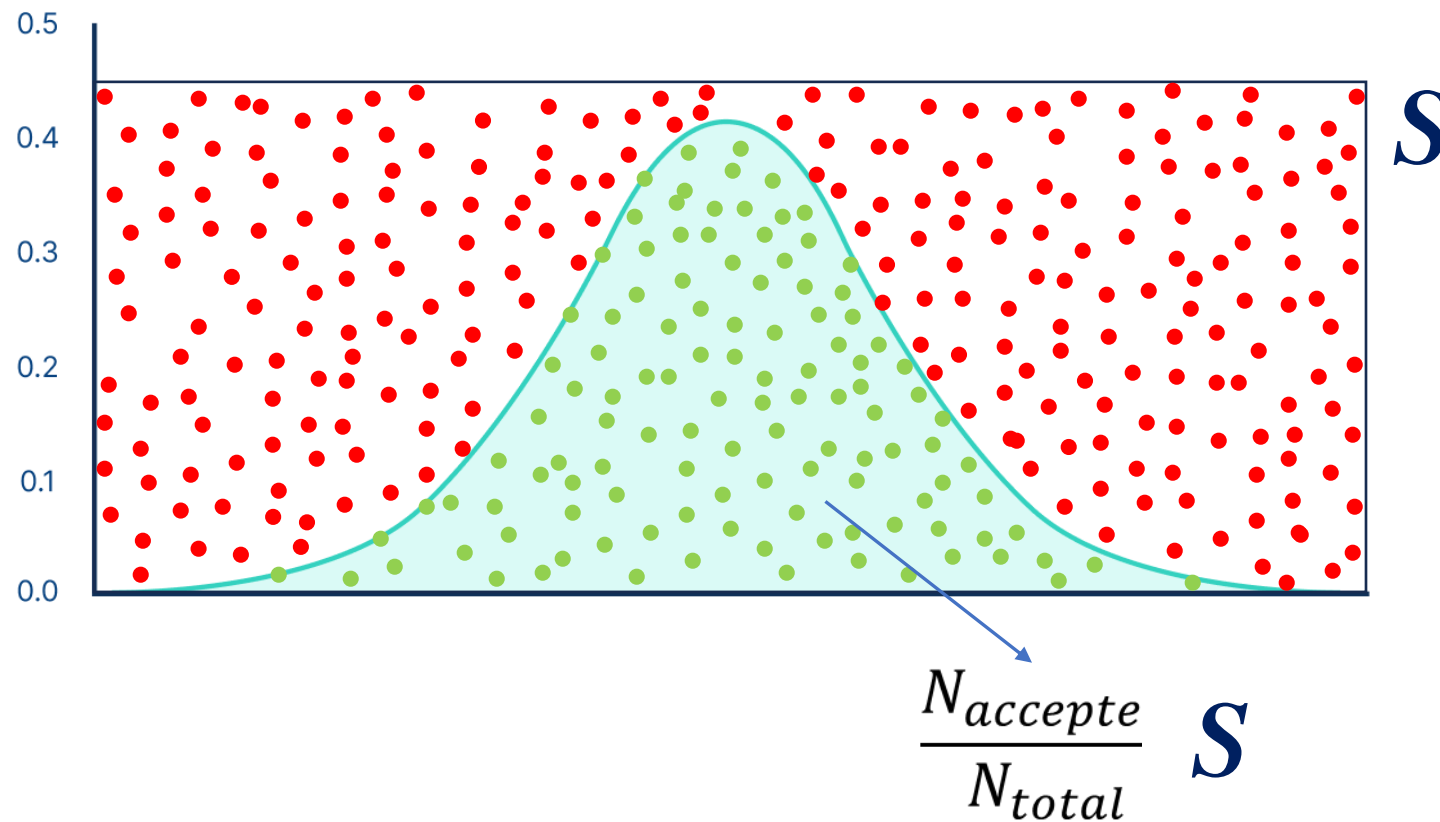
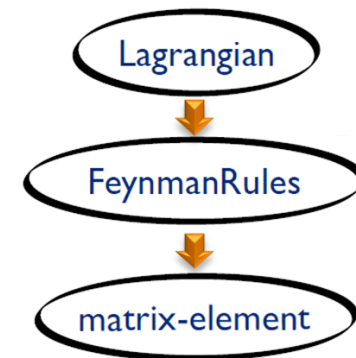


Weighting experimental events with MadWeight Monte-Carlo integration



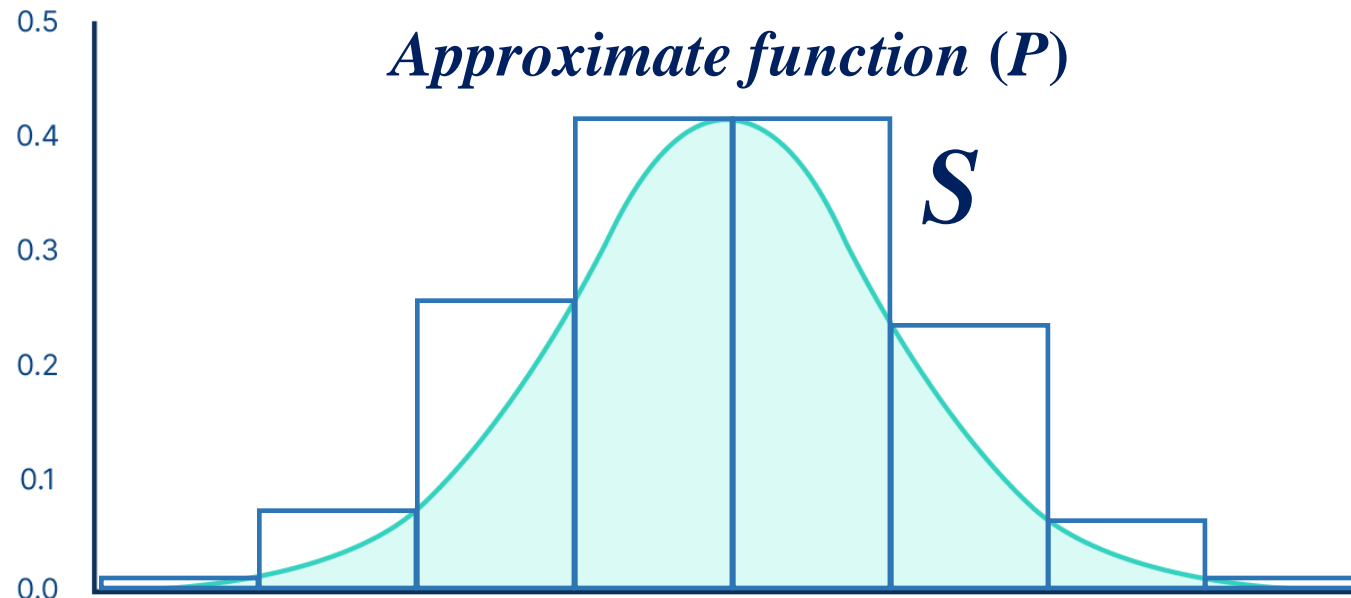
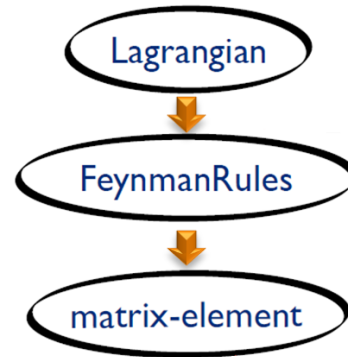


Weighting experimental events with MadWeight Monte-Carlo integration



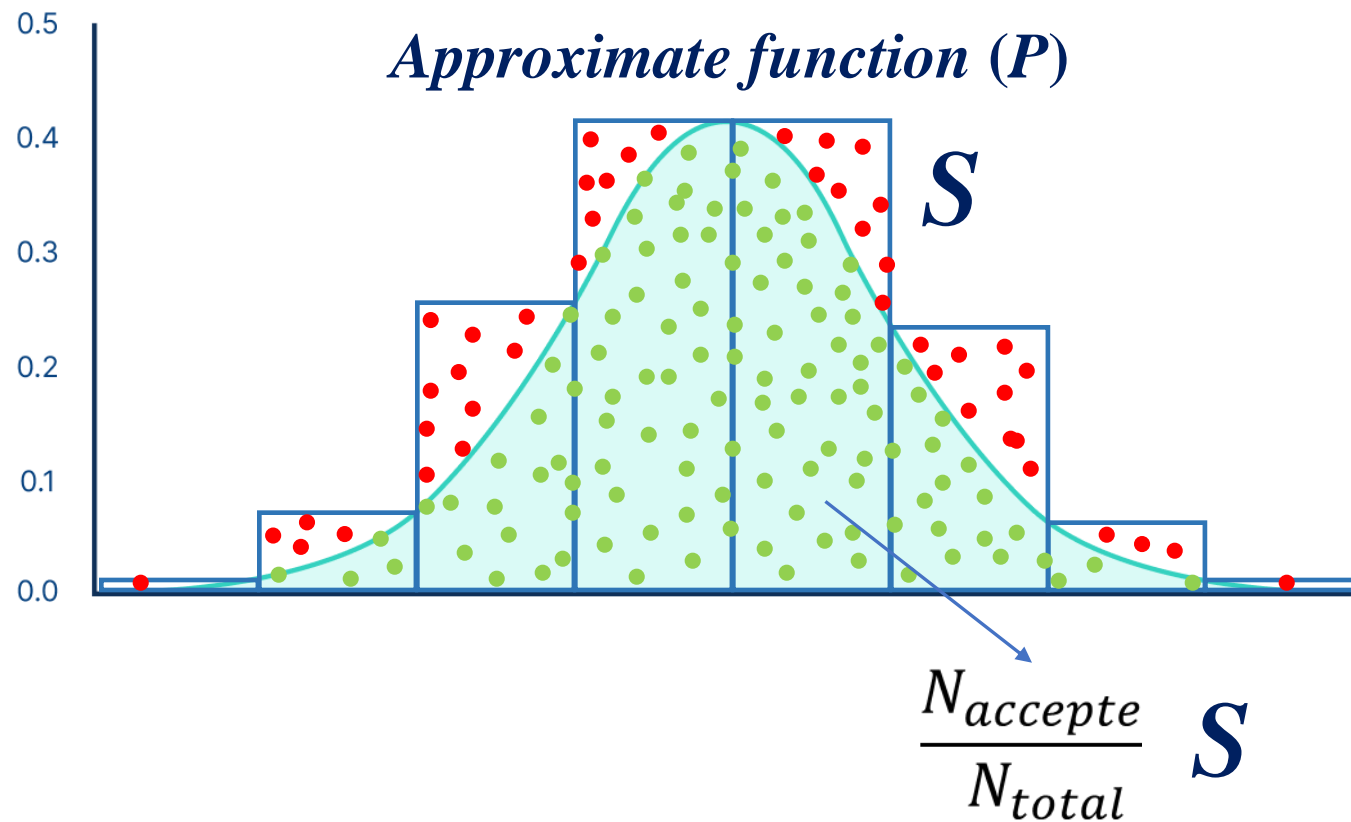
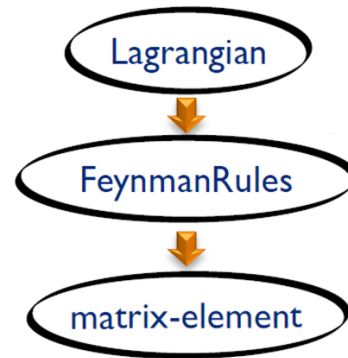


Weighting experimental events with MadWeight Importance sampling



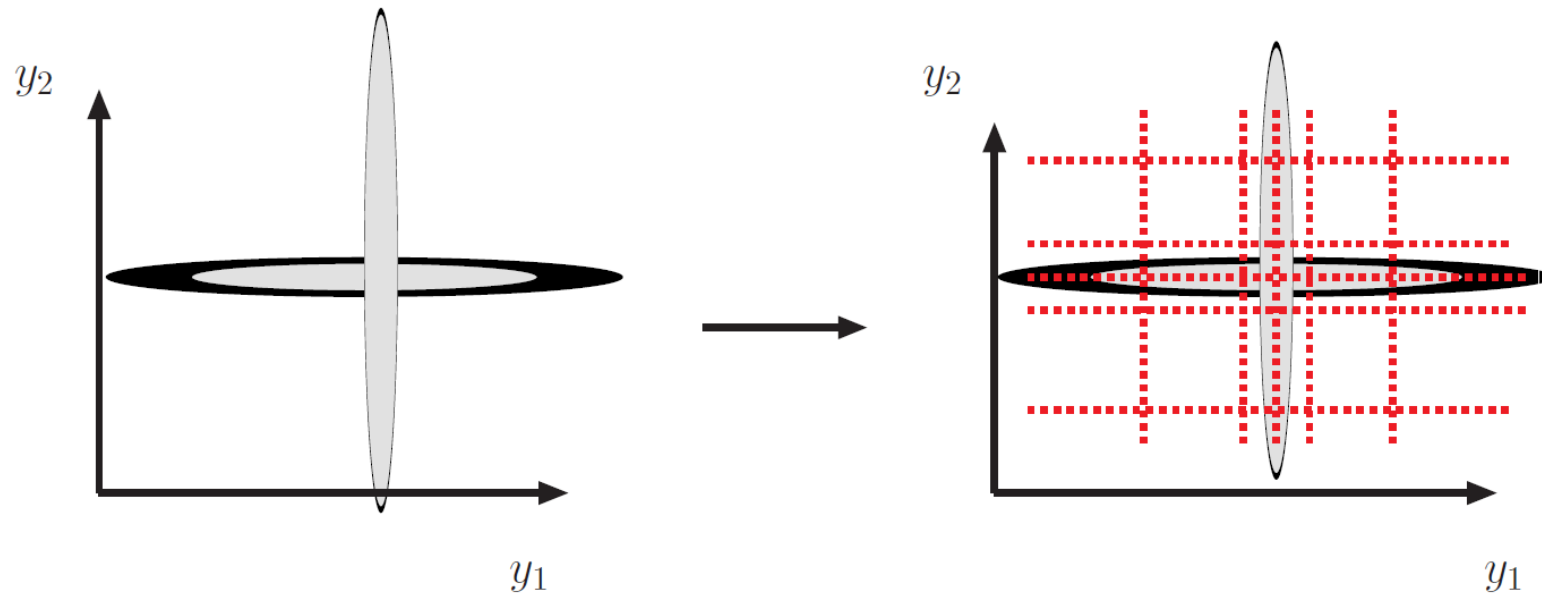
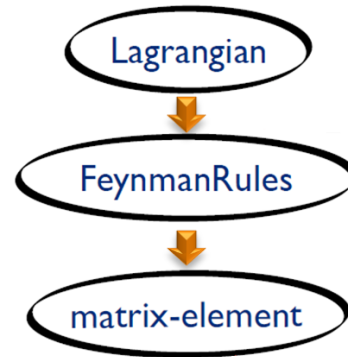


Weighting experimental events with MadWeight Importance sampling





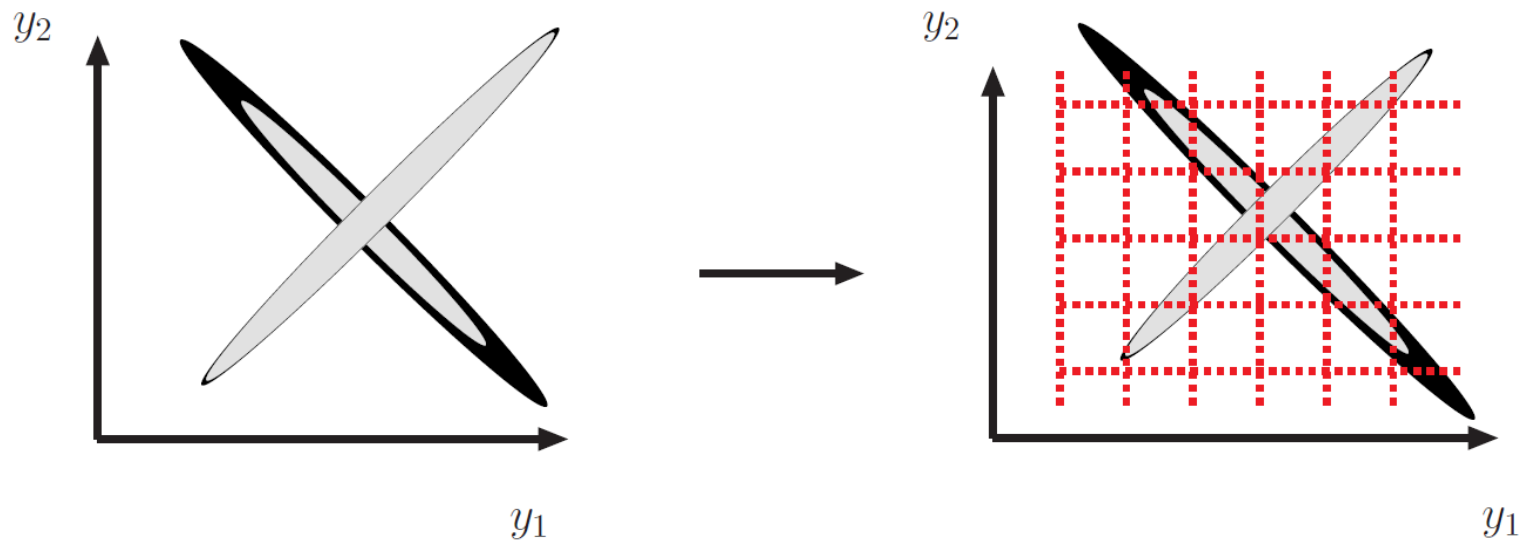
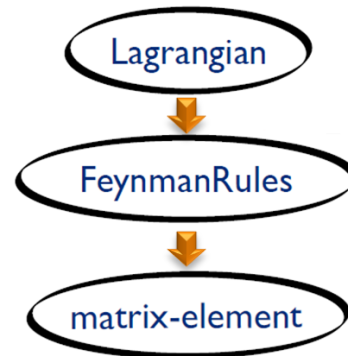
Weighting experimental events with MadWeight VEGAS (Adaptative Monte-Carlo)



*Any peak is aligned along a single direction of the P-S parameterization
Integration is very efficient*



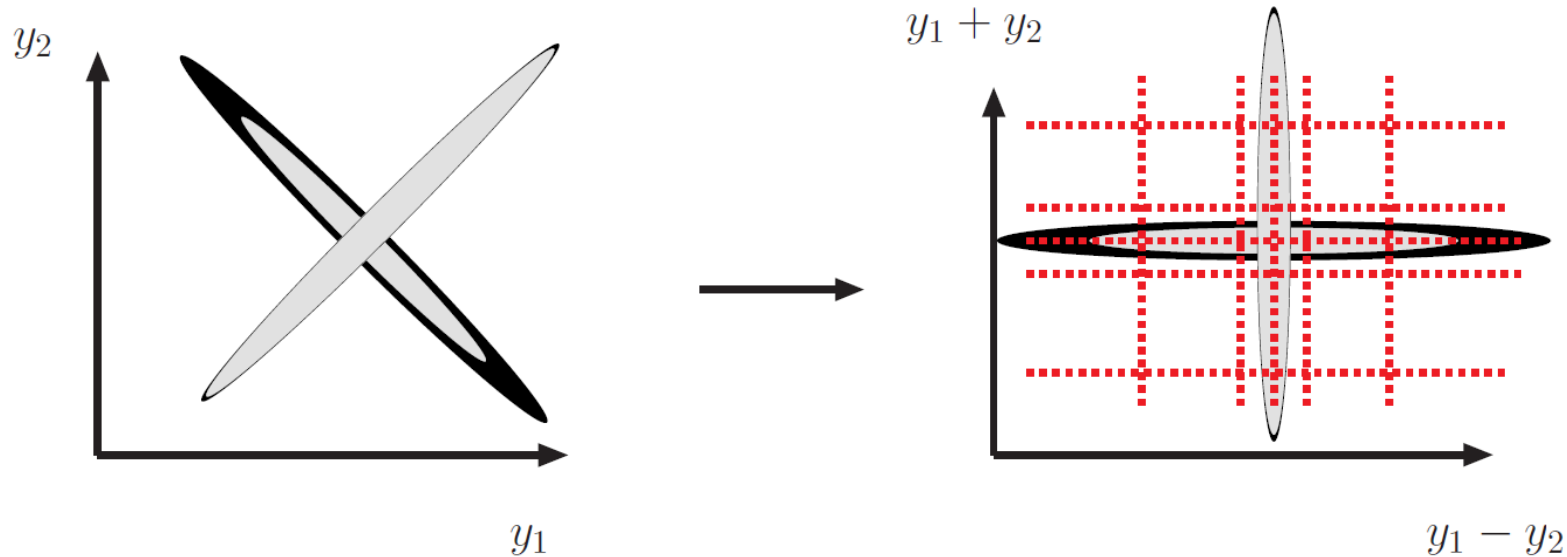
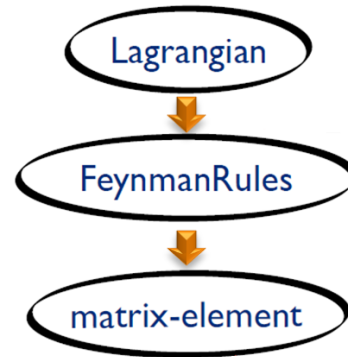
Weighting experimental events with MadWeight VEGAS (Adaptative Monte-Carlo)



*Some peaks are **not aligned** along a single direction of the P-S parameterization
Integration converges slowly*



Weighting experimental events with MadWeight VEGAS (Adaptative Monte-Carlo)



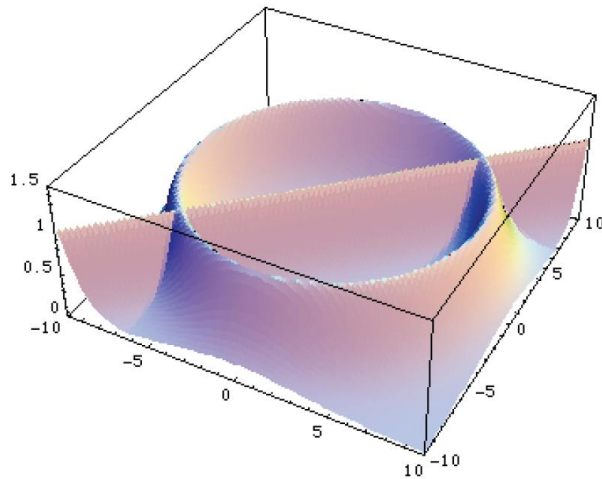
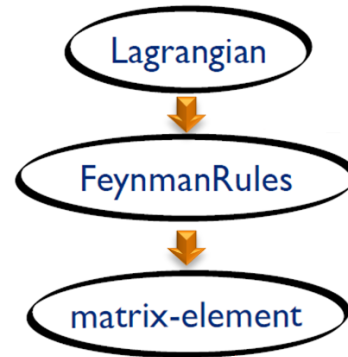
*Some peaks are **not aligned** along a single direction of the P-S parameterization*

Integration converges slowly

Solution: perform a change of variables

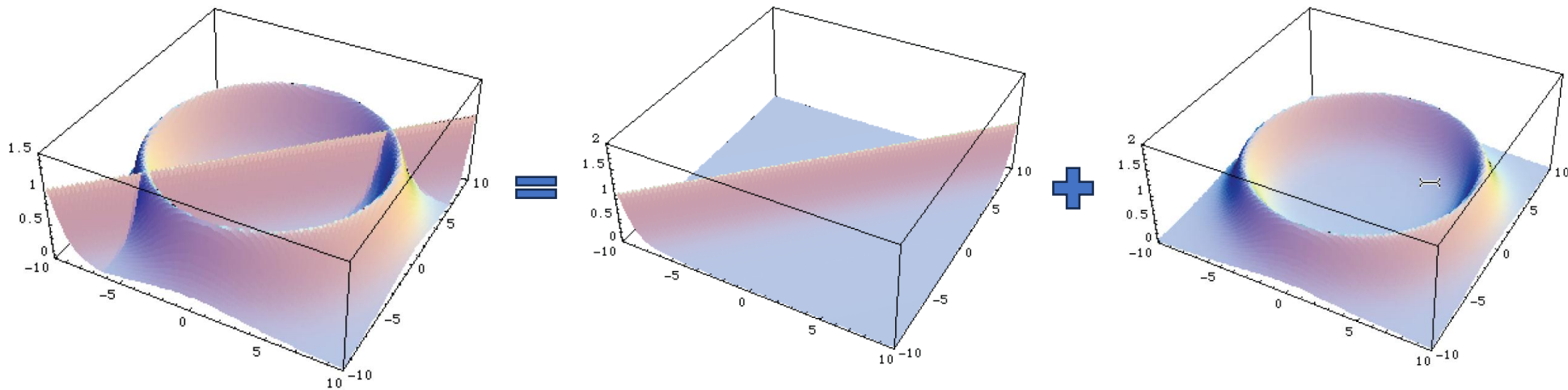
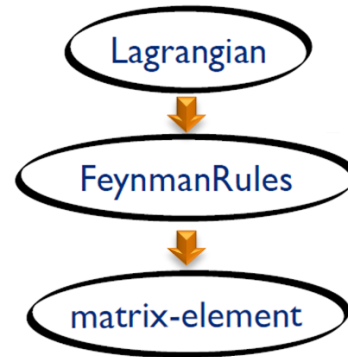


Weighting experimental events with MadWeight Multi-channel Monte-Carlo



What if there is no transformation that aligns all integrand peaks to the chosen axes?

Weighting experimental events with MadWeight Multi-channel Monte-Carlo



What if there is no transformation that aligns all integrand peaks to the chosen axes?

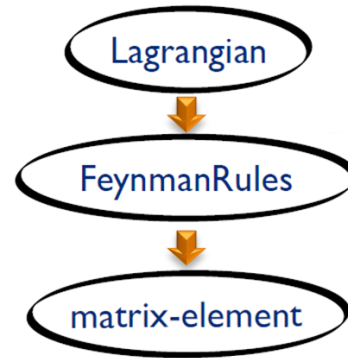
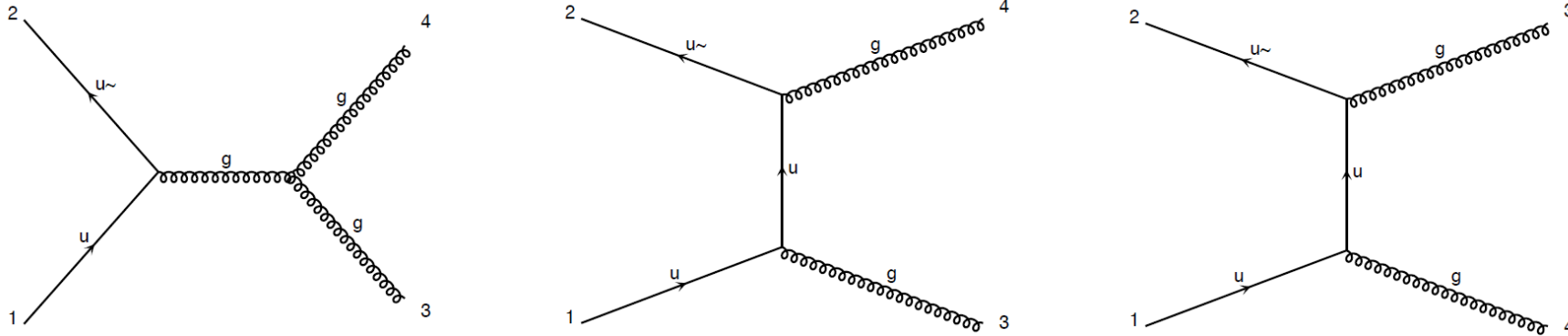
Solution: use different transformations (channels)

$$p(x) = \sum_{i=1}^n \alpha_i p_i(x) \quad \sum_{i=1}^n \alpha_i = 1$$



Weighting experimental events with MadWeight

Example



Three very different pole structures contributing to the same matrix element

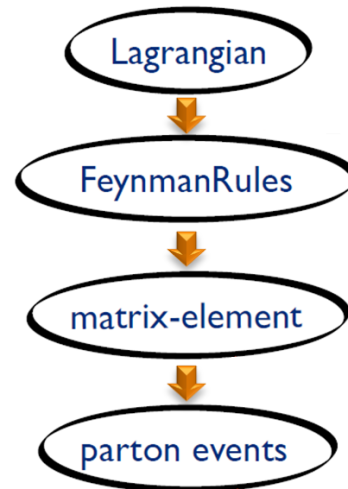
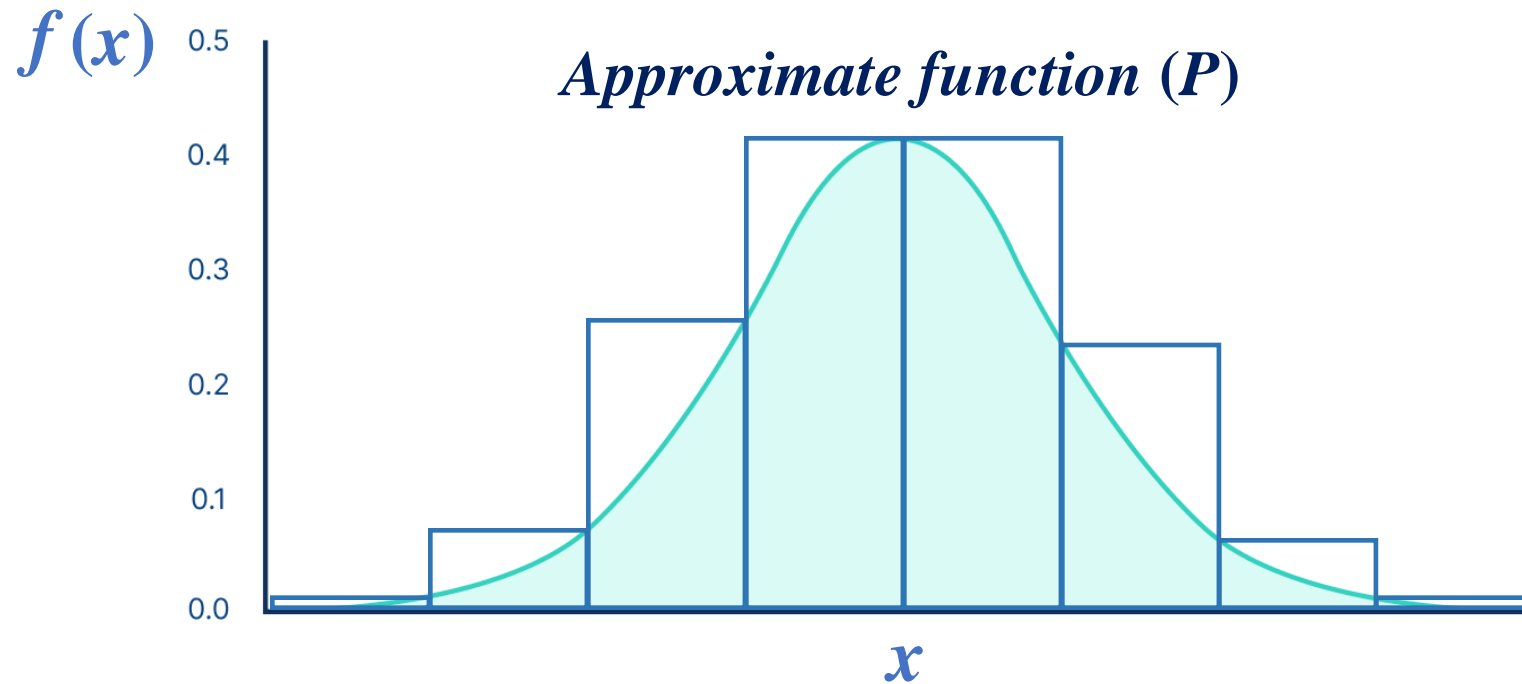
$$p(x) = \sum_{i=1}^n \alpha_i p_i(x) \quad \sum_{i=1}^n \alpha_i = 1$$

$$\int |M_{tot}|^2 = \int \frac{\sum_i |M_i|^2}{\sum_j |M_j|^2} |M_{tot}|^2 = \sum_i \int \frac{|M_i|^2}{\sum_j |M_j|^2} |M_{tot}|^2$$

≈ 1

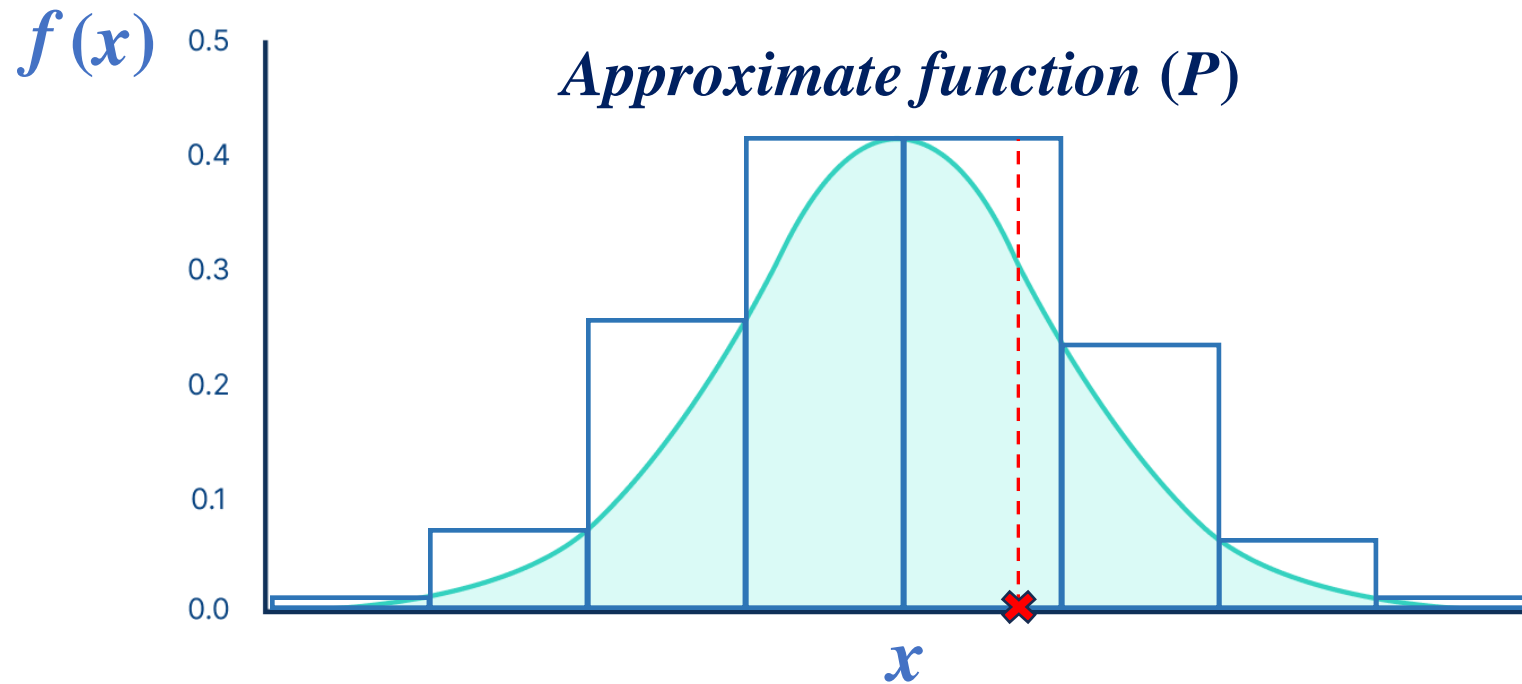


Generating events with MadEvent

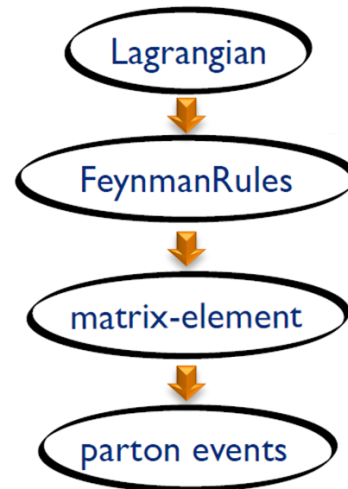




Generating events with MadEvent

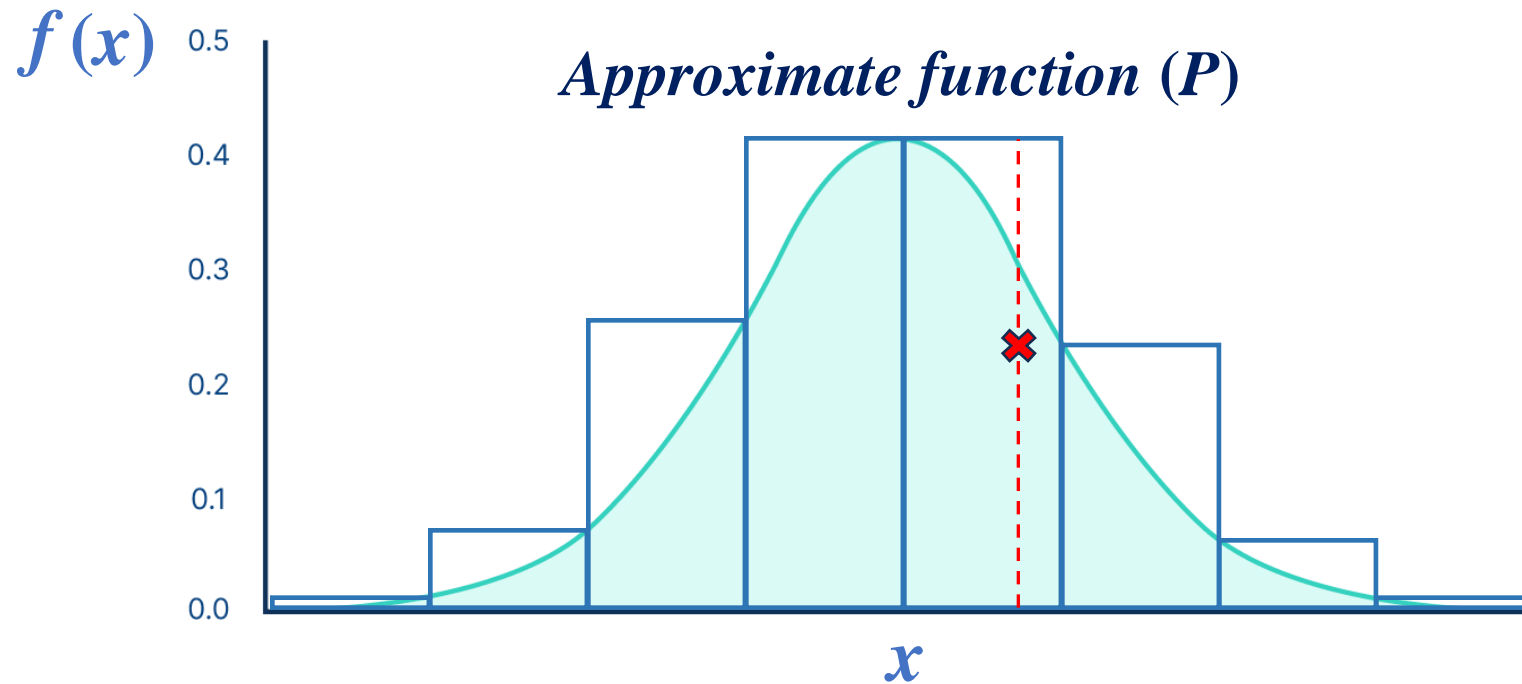


1. pick x distributed as $p(x)$



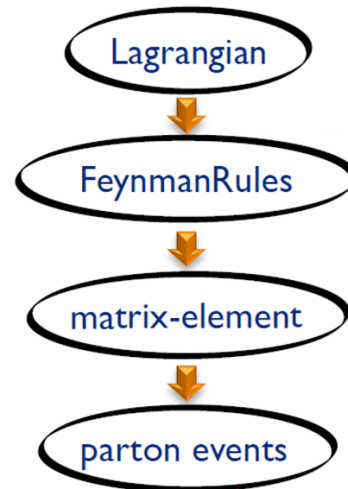


Generating events with MadEvent



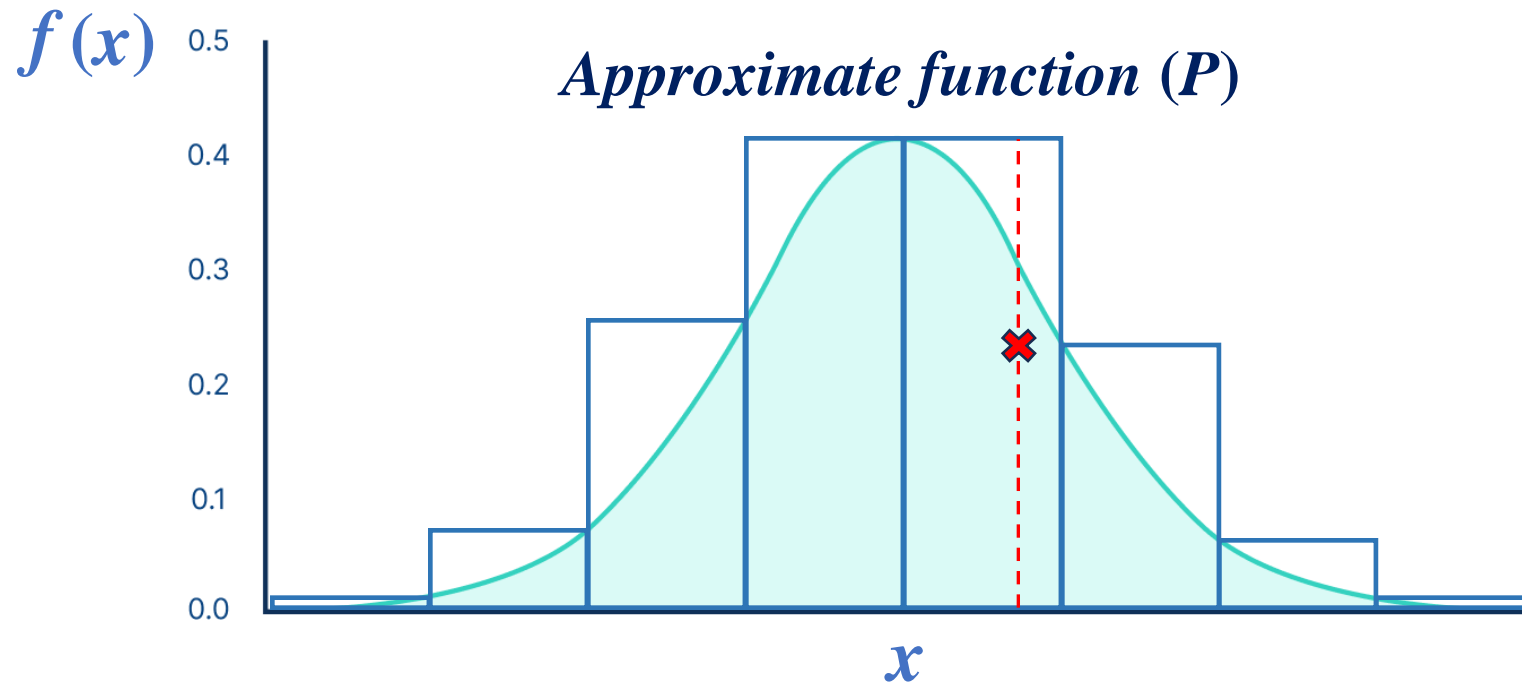
1. pick x distributed as $p(x)$

2. pick $0 < y < p(x)$

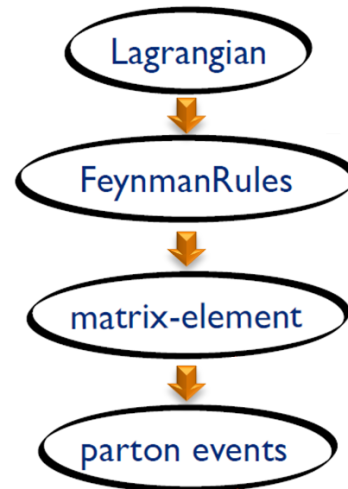




Generating events with MadEvent



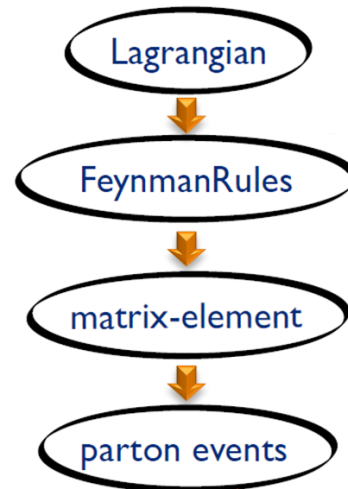
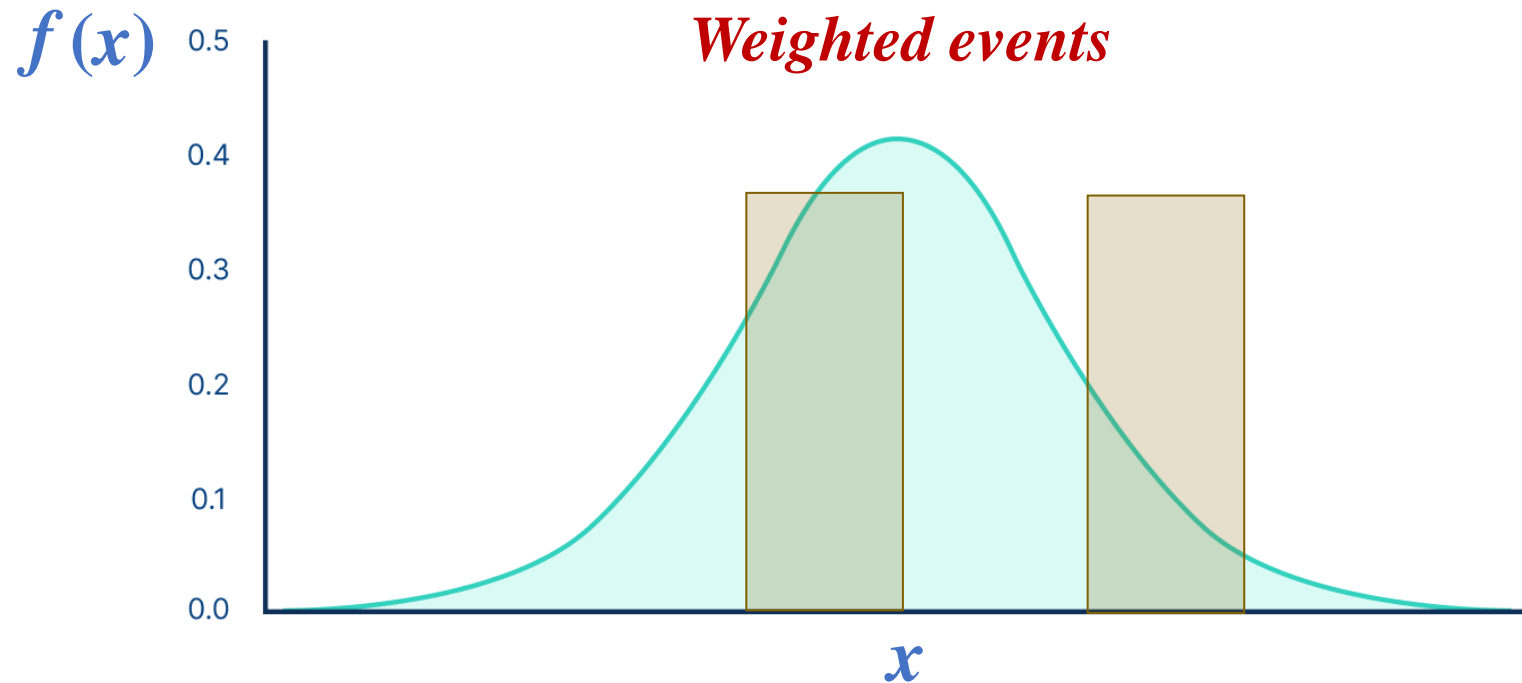
1. pick x distributed as $p(x)$
2. pick $0 < y < p(x)$
3. if $y < f(x)$ accept event, else reject it





Generating events with MadEvent

Weighted events



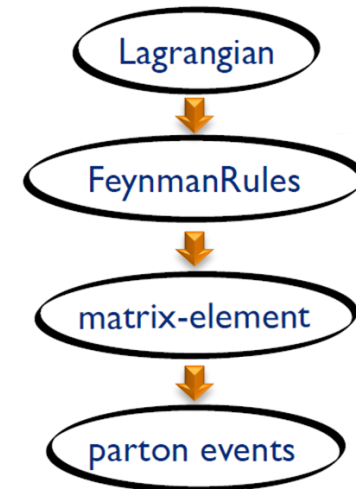
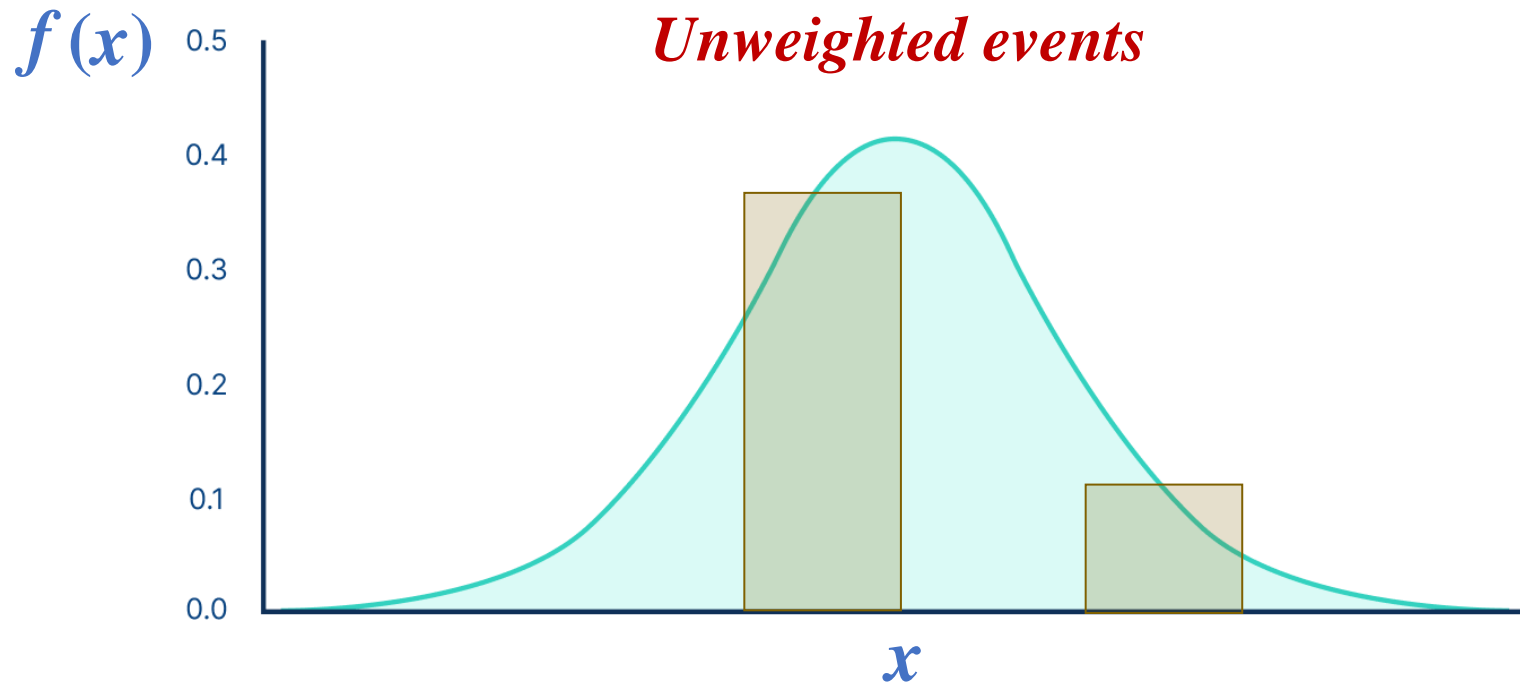
Same number of events in areas of phase space with different probabilities

Events must have different weights (weighted)



Generating events with MadEvent

Unweighted events



Number of events is proportional to the probability of areas of phase space

Events have the same weight (unweighted)

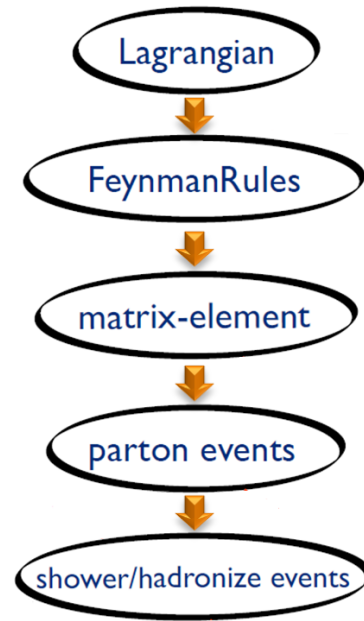
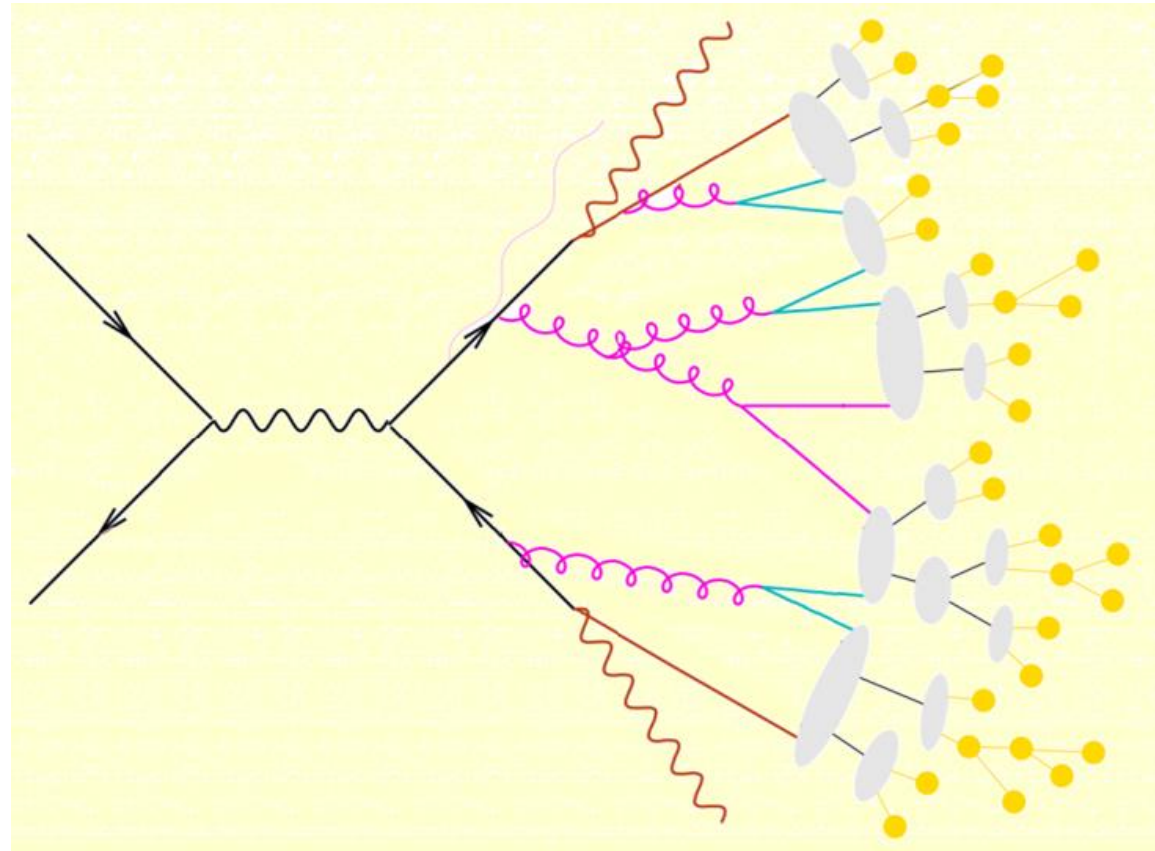
Events distributed as in nature

Pythia

[arXiv:2203.11601]



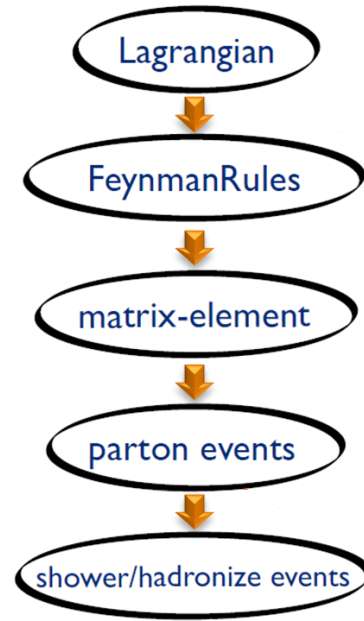
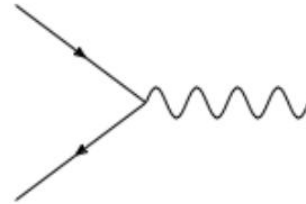
Pythia adds parton showers, multiparton interactions, hadronization and decay





Matching to parton shower

PARTON EVENTS



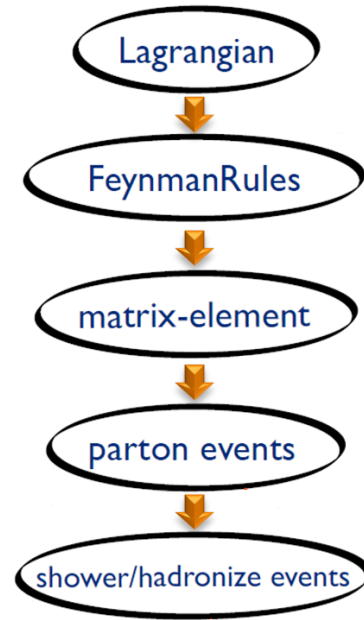
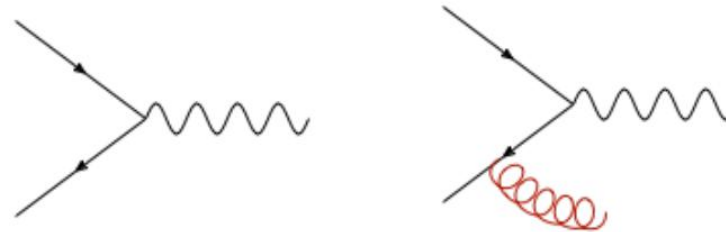


Matching to parton shower

PARTON SHOWER



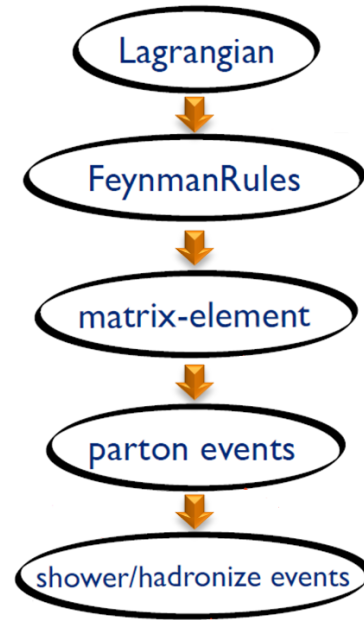
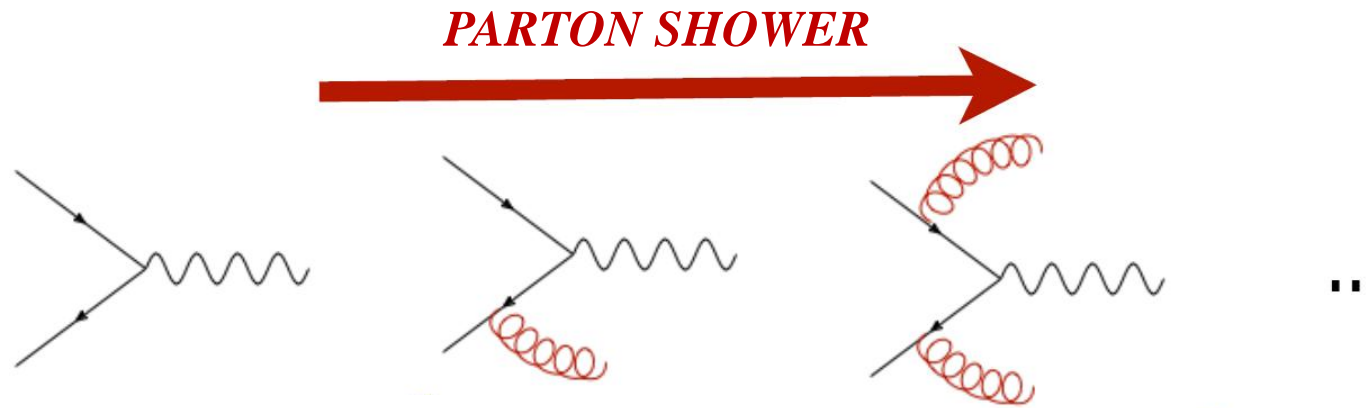
PARTON EVENTS





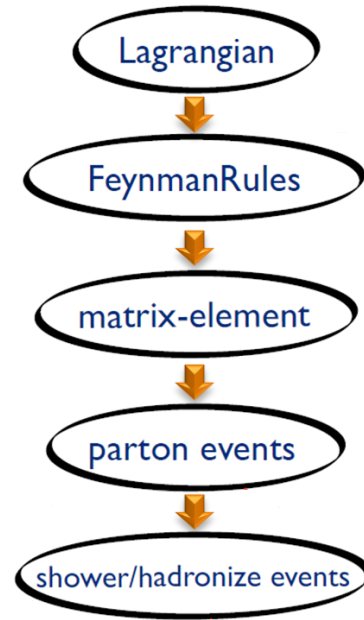
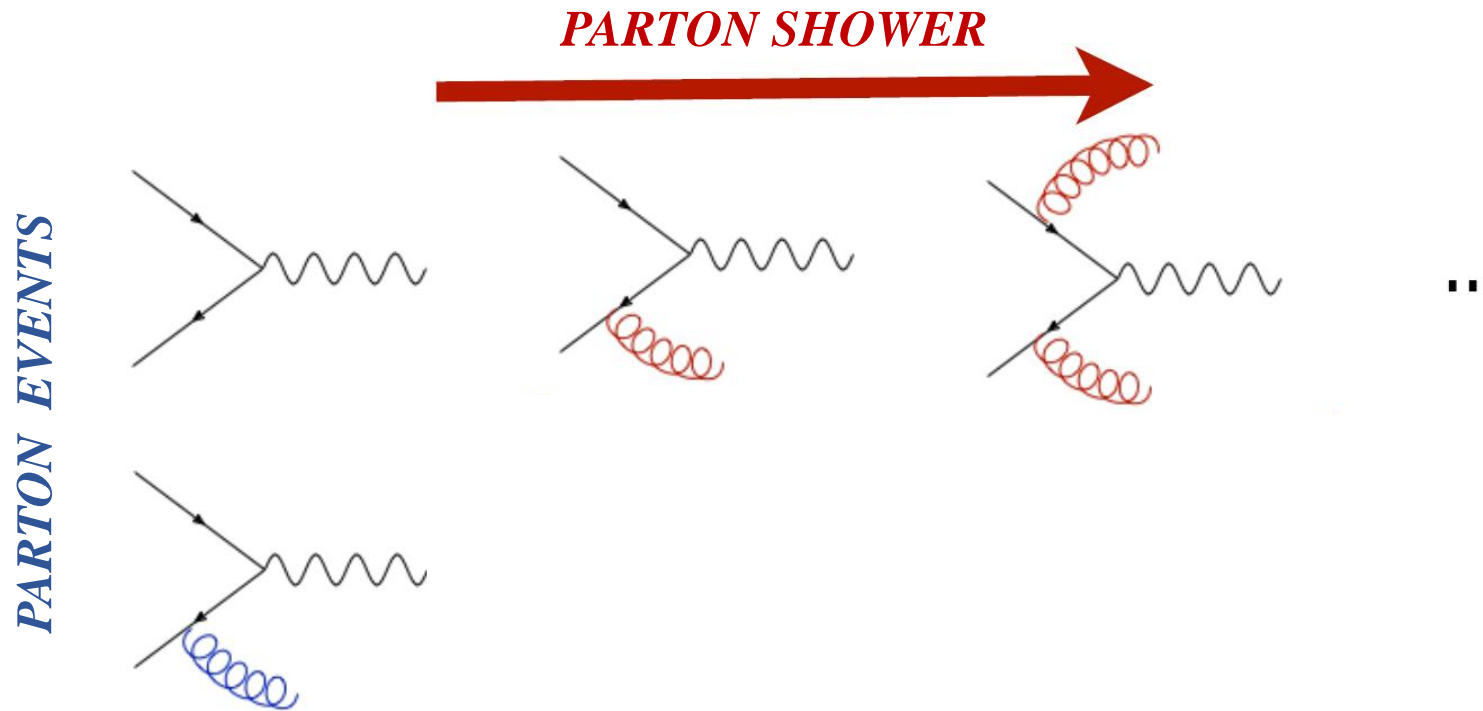
Matching to parton shower

PARTON EVENTS



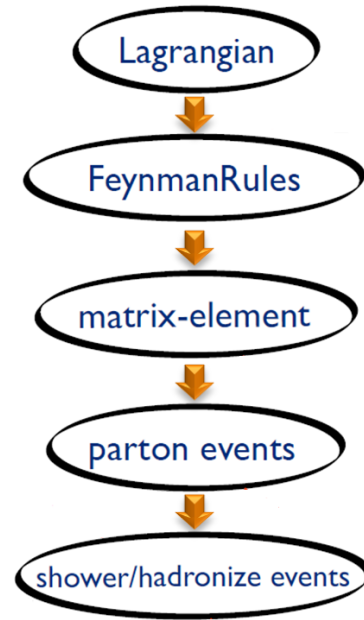
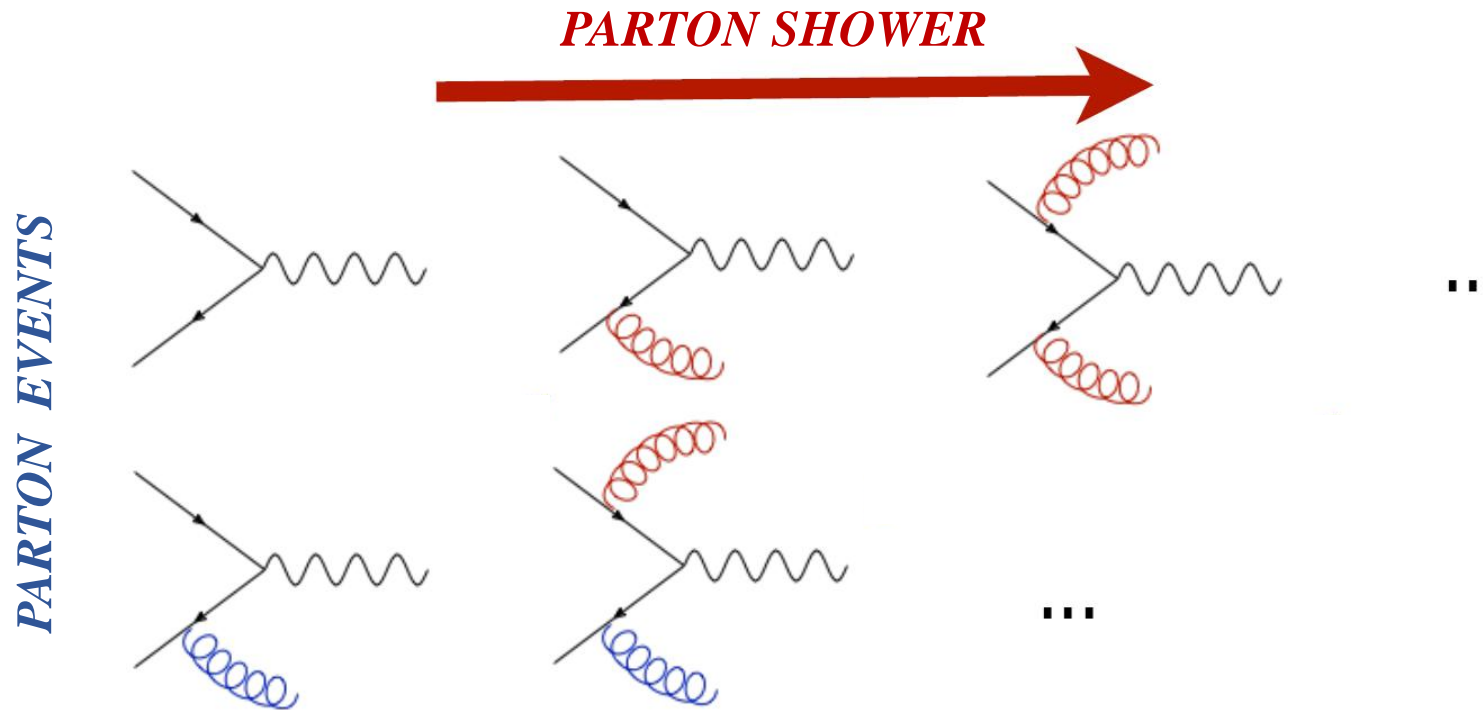


Matching to parton shower



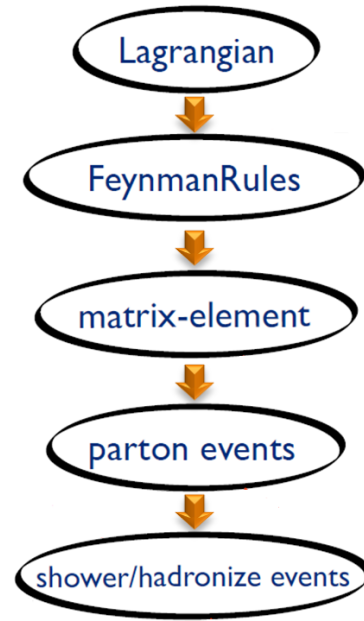
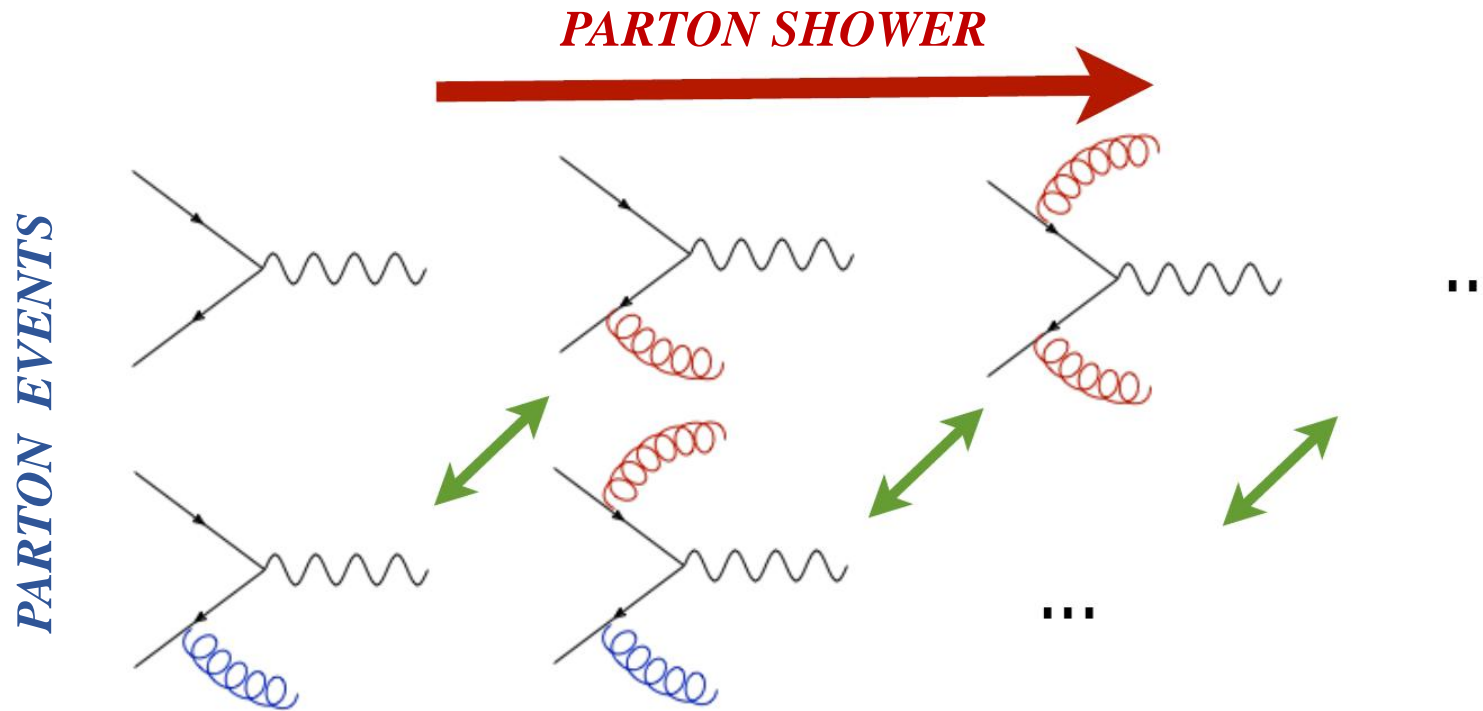


Matching to parton shower





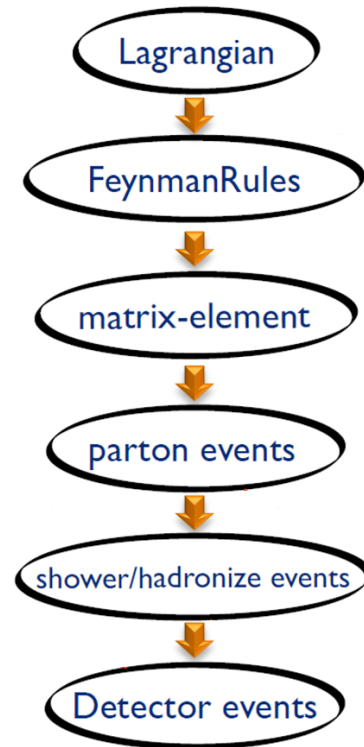
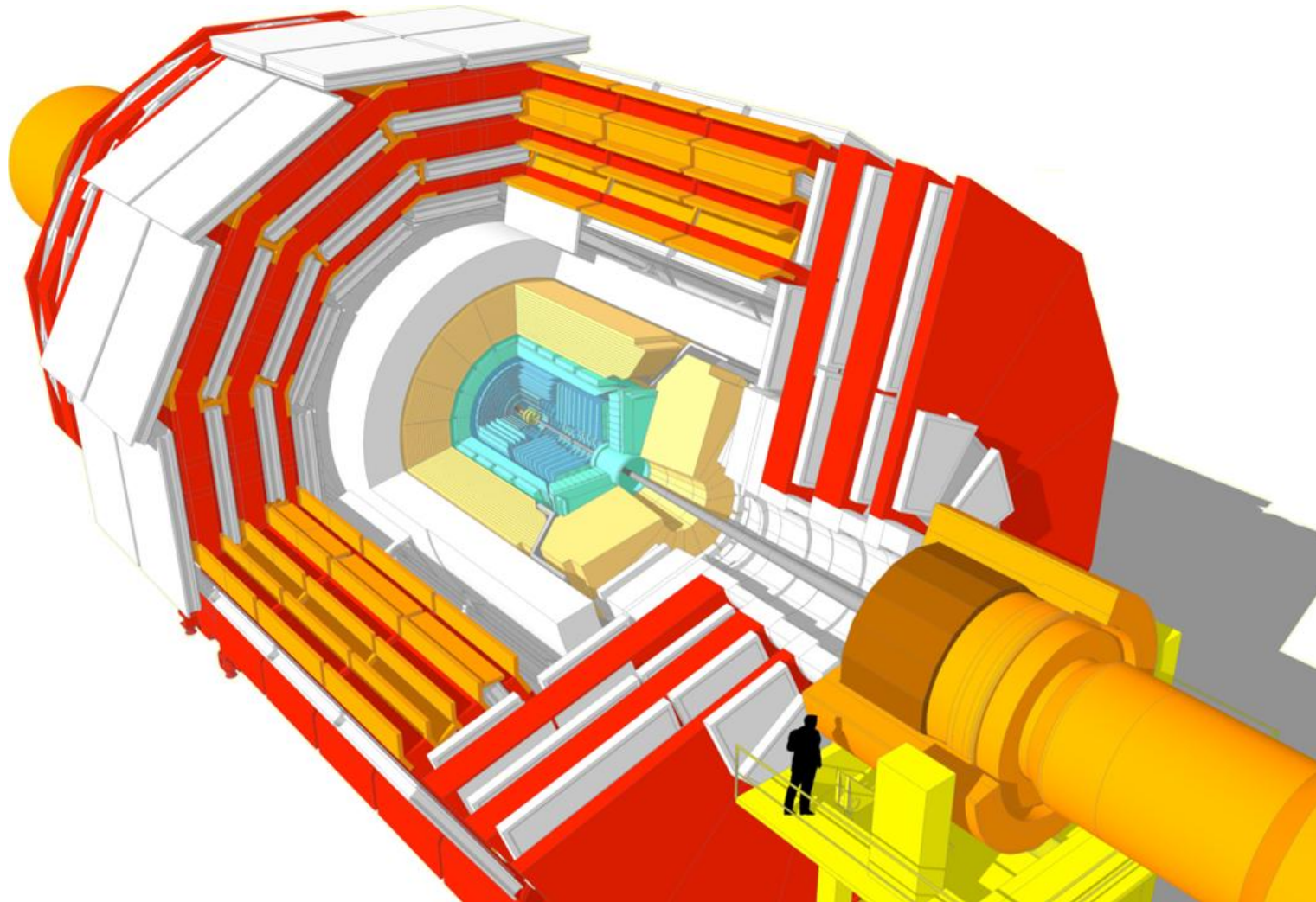
Matching to parton shower



Double counting occurs!

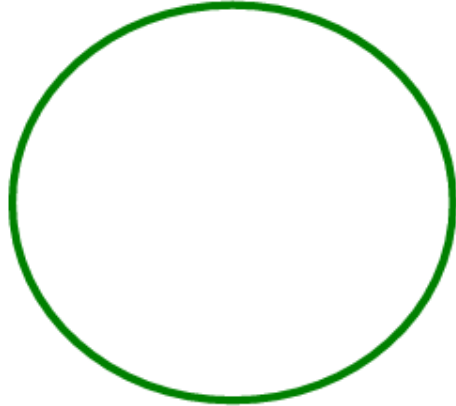
Possible solution: MLM merging

Detector simulation tools



Detector simulation tools

realism

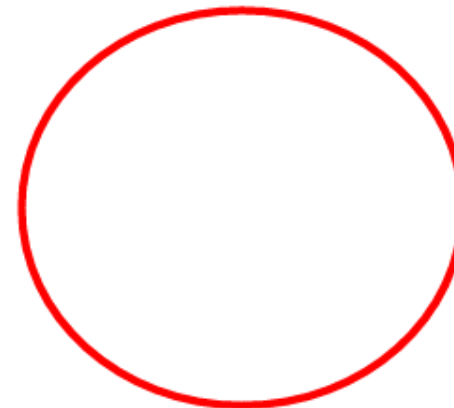


Full simulation (GEANT4):

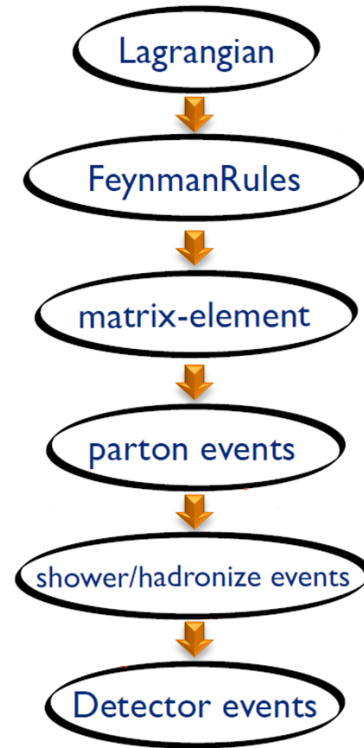
- *Particle-matter interactions*
- *Reconstruction algorithms*

Fast simulation (Delphes):

- *Functions between particles and reconstructed objects*

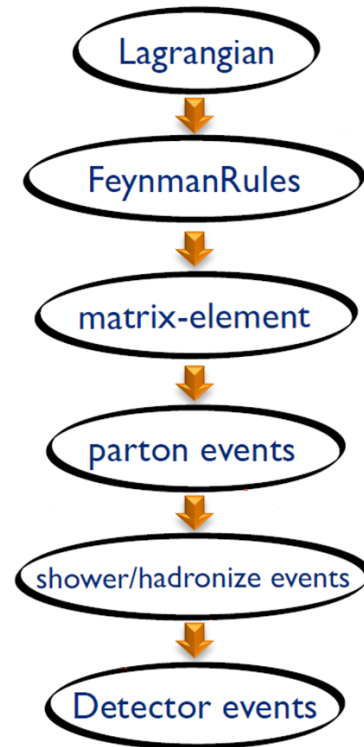
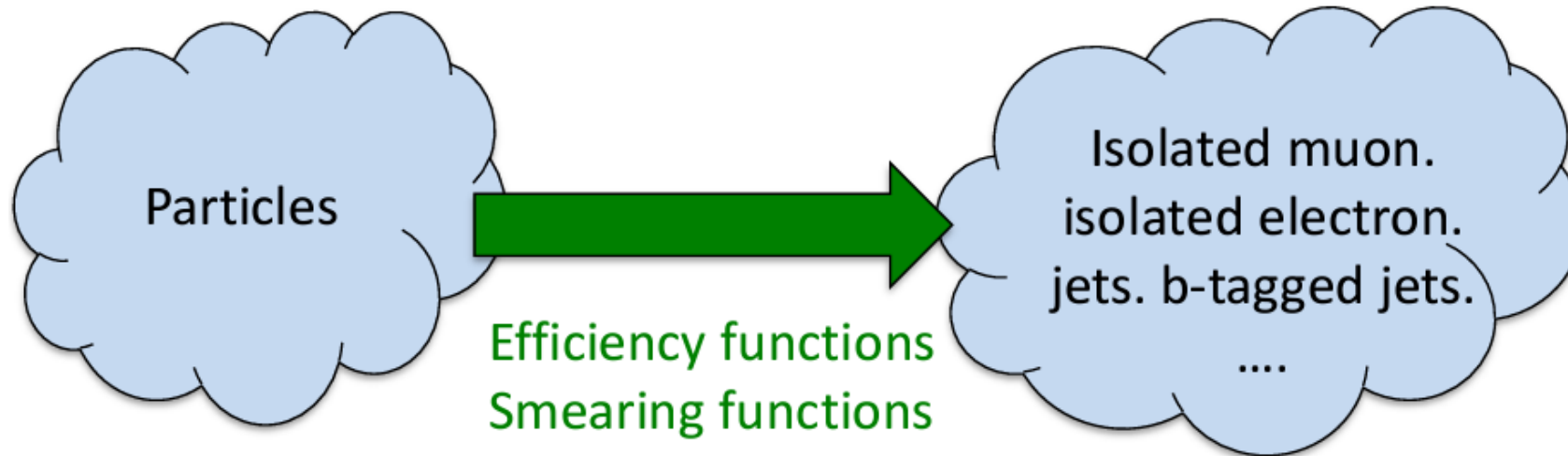


speed





*Fast detector simulation for phenomenological studies
Based on the parameterization of the detector response*

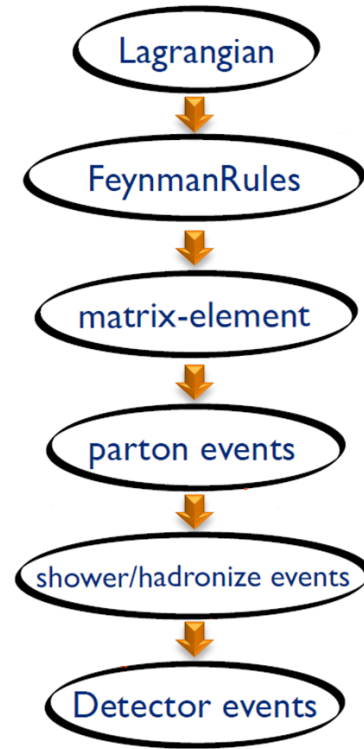
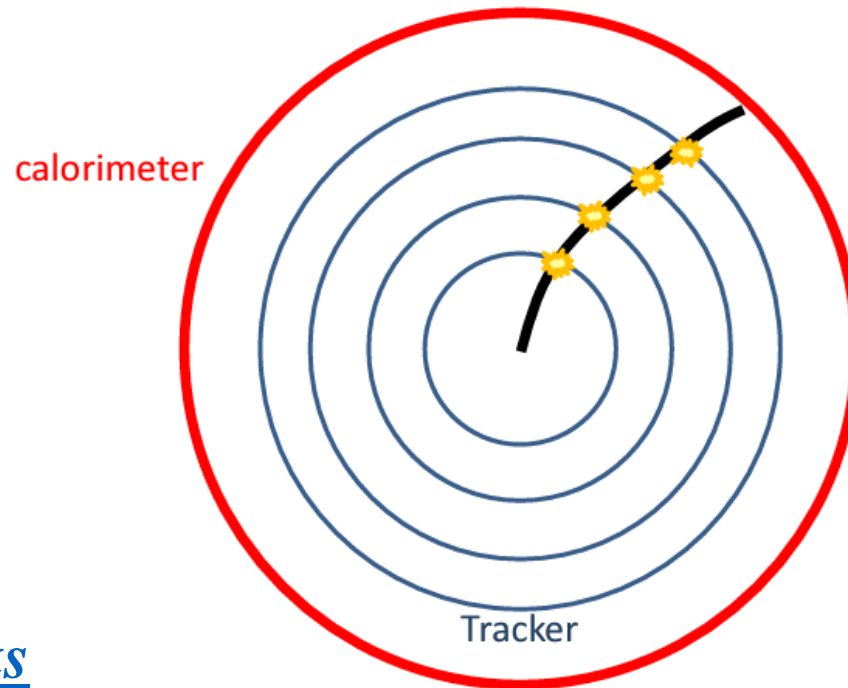


Delphes

[arXiv:1307.6346]



*Fast detector simulation for phenomenological studies
Based on the parameterization of the detector response*



[See how Delphes works](#)

No real tracking in Delphes

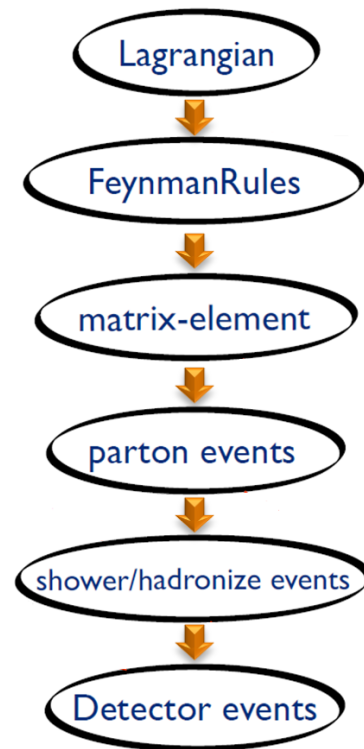
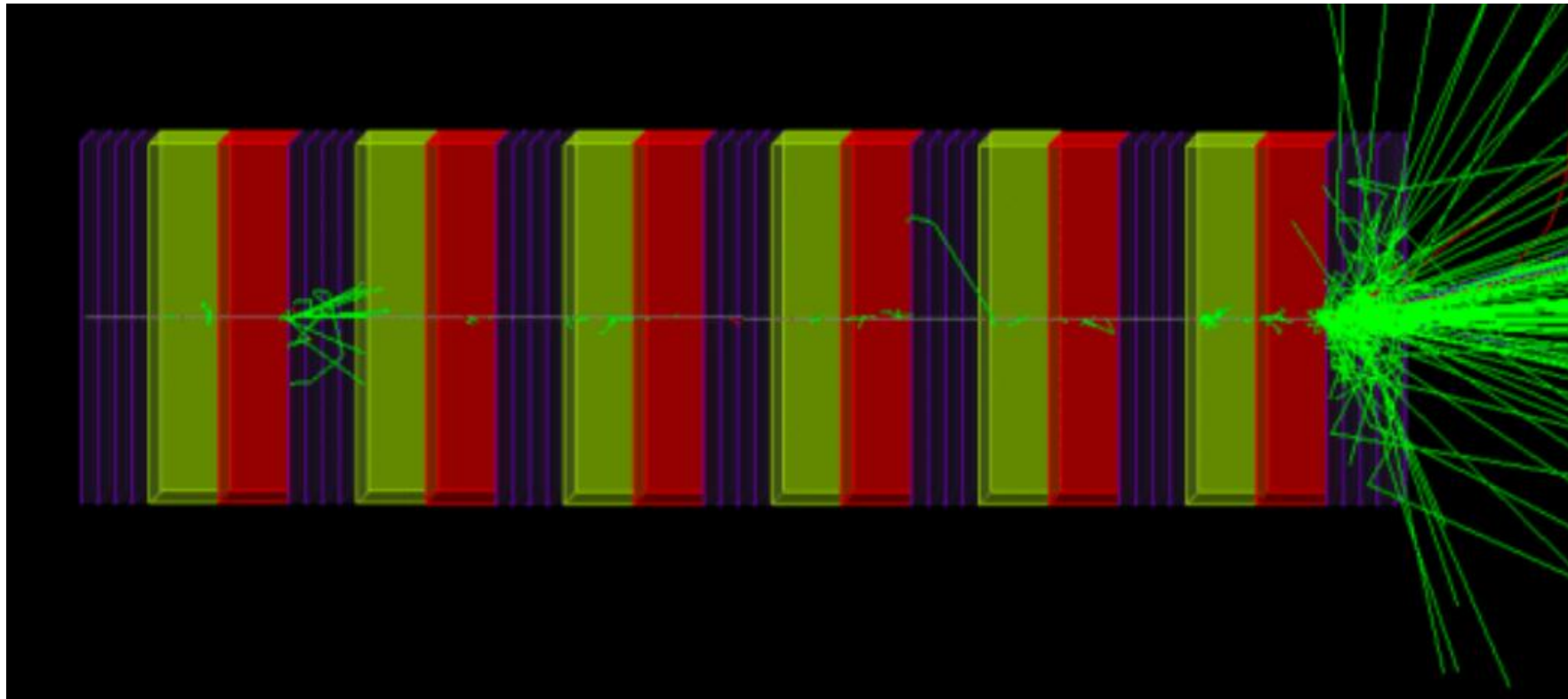
GEANT

[0.1109/TNS.2006.869826]

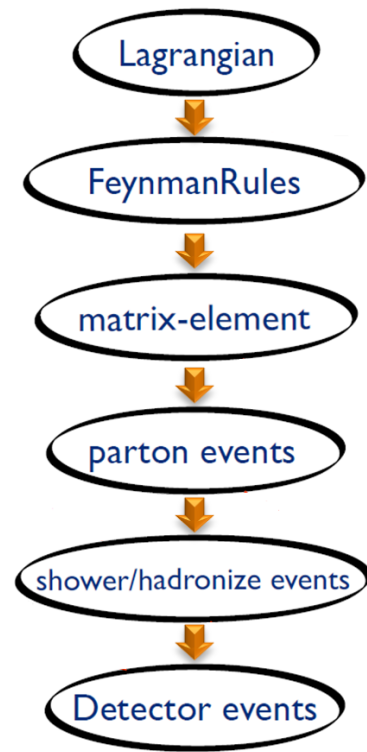
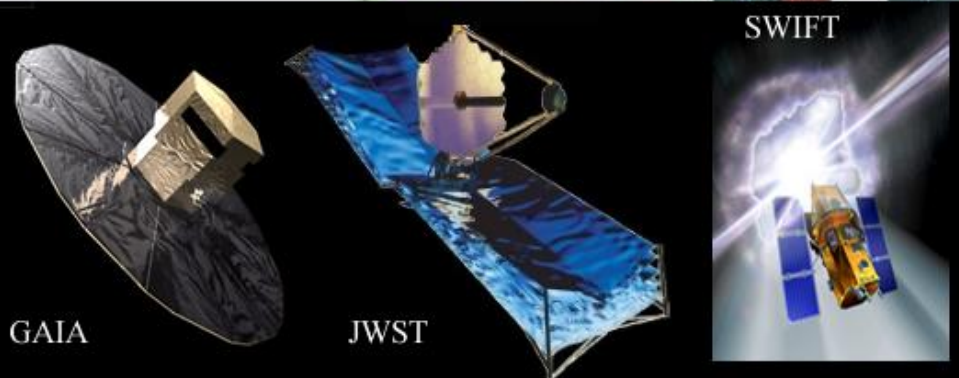
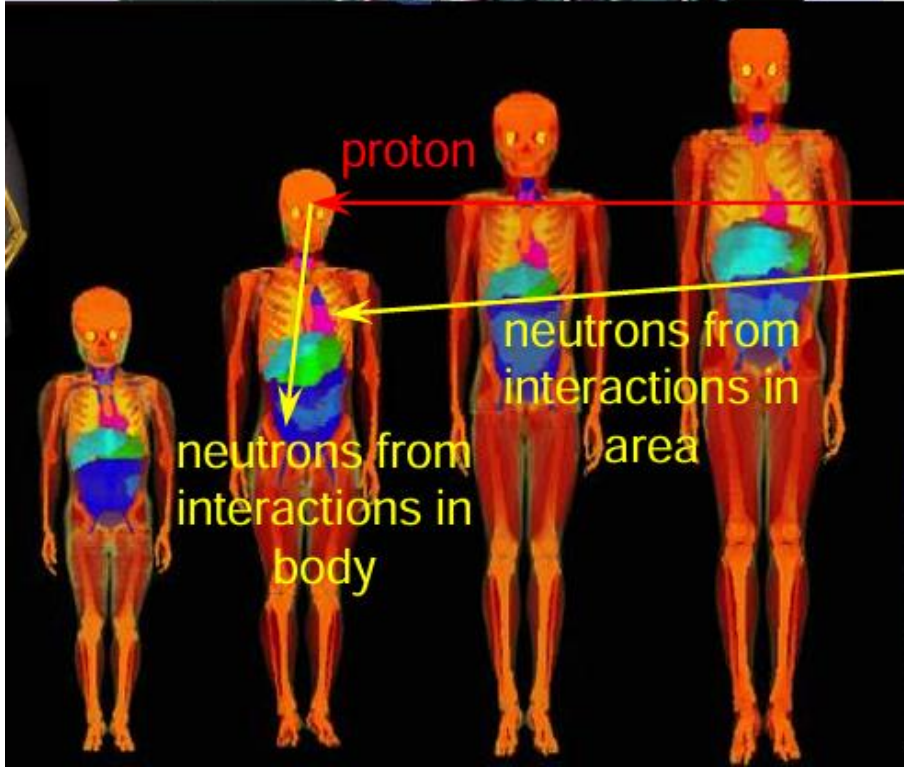
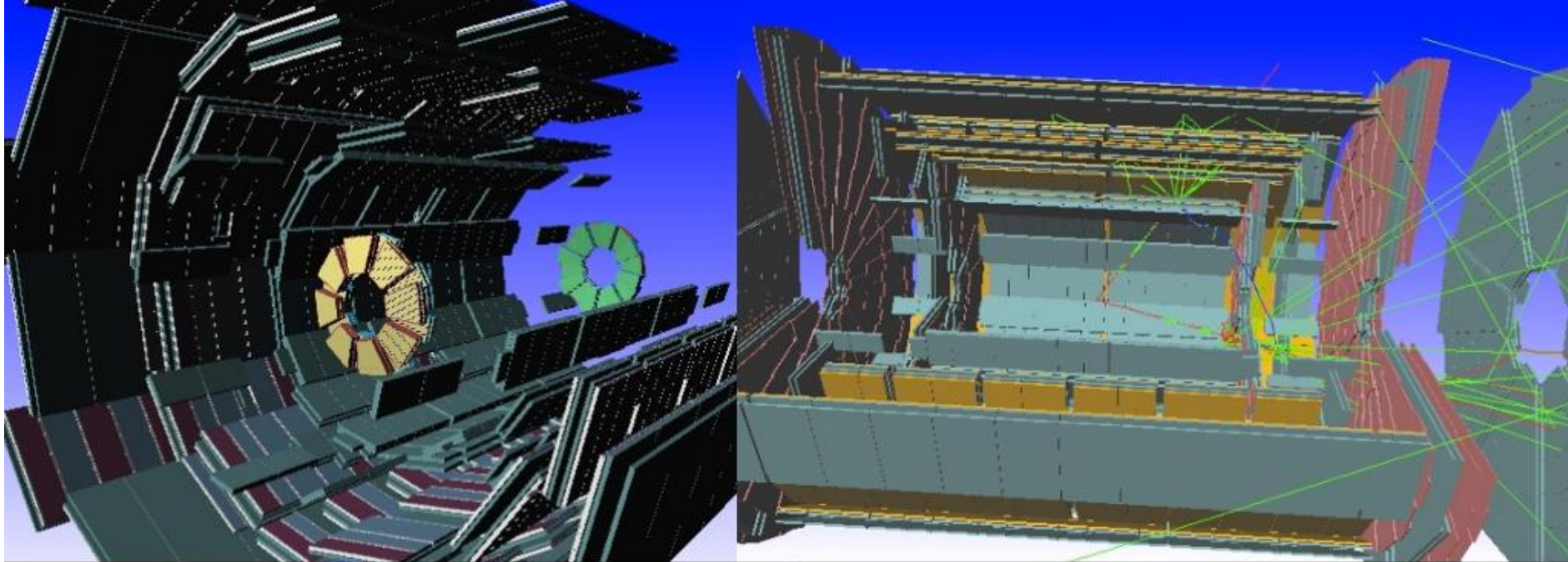


GEometry ANd Tracking

*A Monte Carlo software toolkit
to simulate the passage of particles through matter*

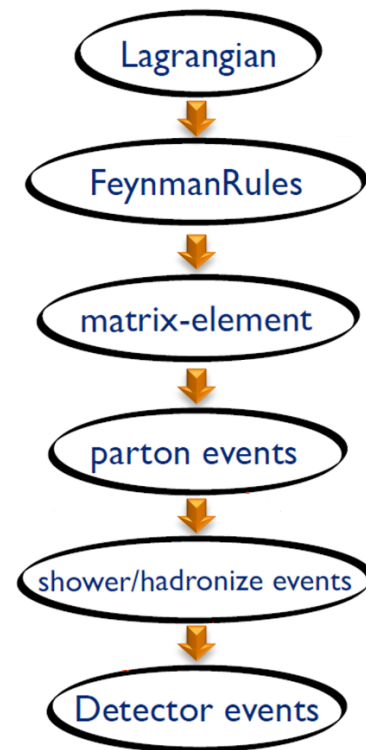


*Widely used
A gift from particle physics*



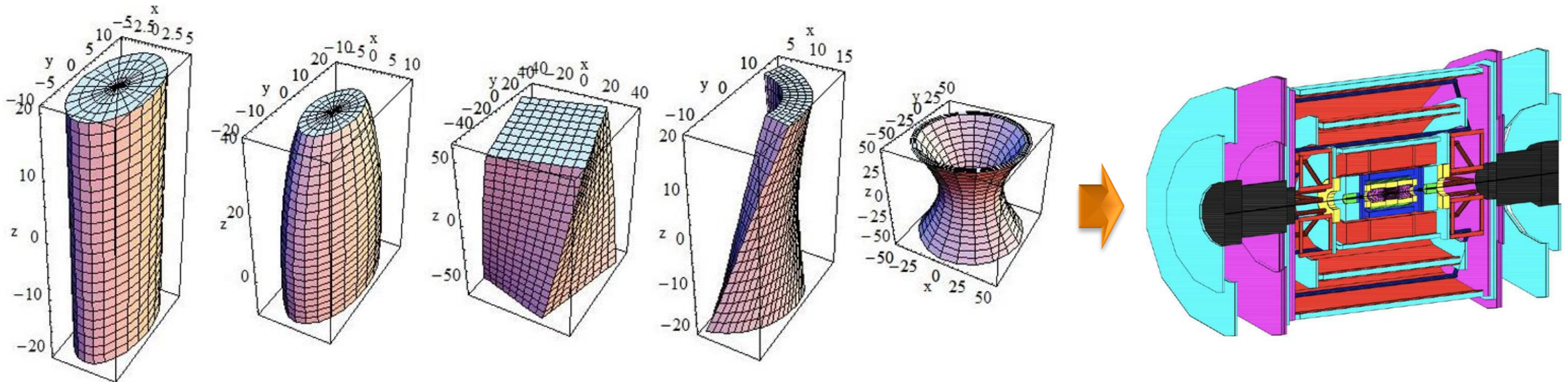
Particle-Matter interaction

Digitization



Particle-Matter interaction

➤ Describe the geometry and the material of the detector

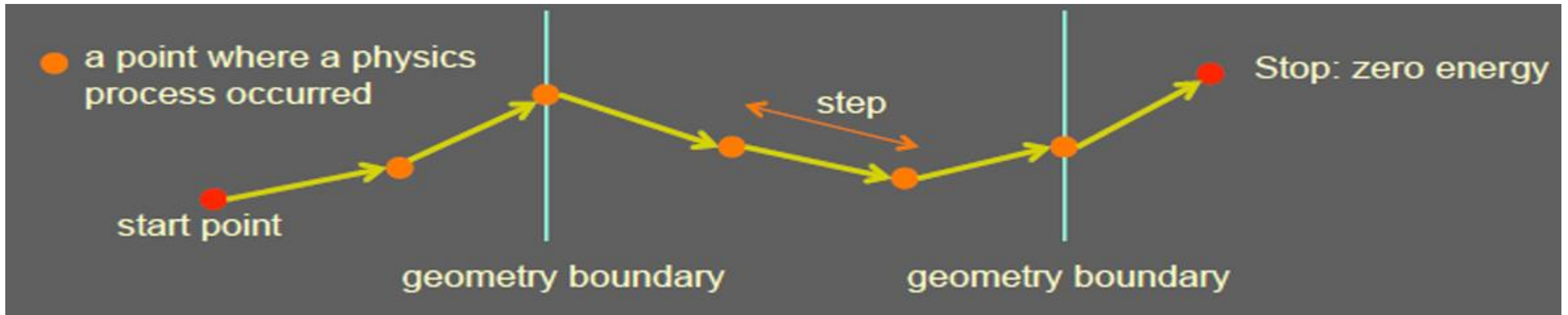


Particle-Matter interaction

- *Describe the geometry and the material of the detector*
- *Treat a particle at a time*

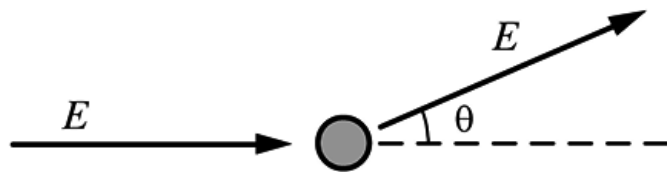
Particle-Matter interaction

- Describe the geometry and the material of the detector
- Treat a particle at a time
- Trajectory of the particle is split in steps (finite displacements)

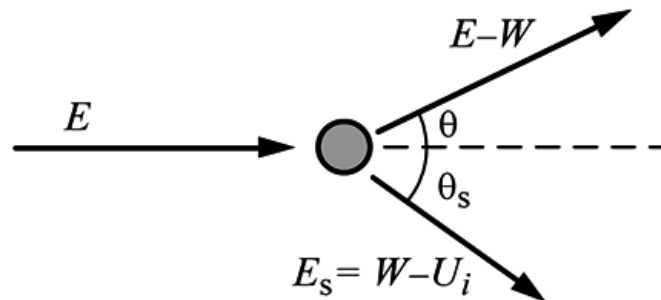


Particle-Matter interaction

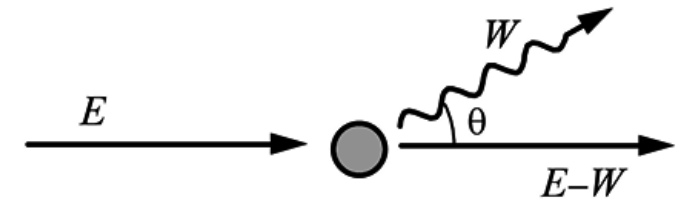
- Describe the geometry and the material of the detector
- Treat a particle at a time
- Trajectory of the particle is split in steps (finite displacements)
- Simulate the physics along a step and at the end of each step



Elastic scattering



Inelastic scattering



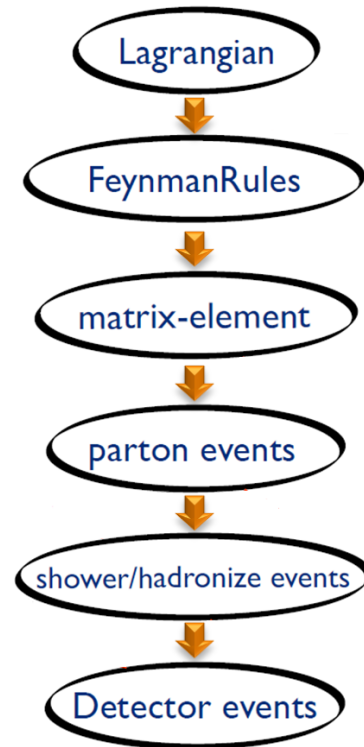
Bremsstrahlung emission

Particle-Matter interaction

- Describe the geometry and the material of the detector
- Treat a particle at a time
- Trajectory of the particle is split in steps (finite displacements)
- Simulate the physics along a step and at the end of each step

Digitization

- convert the energy deposit into electric signal

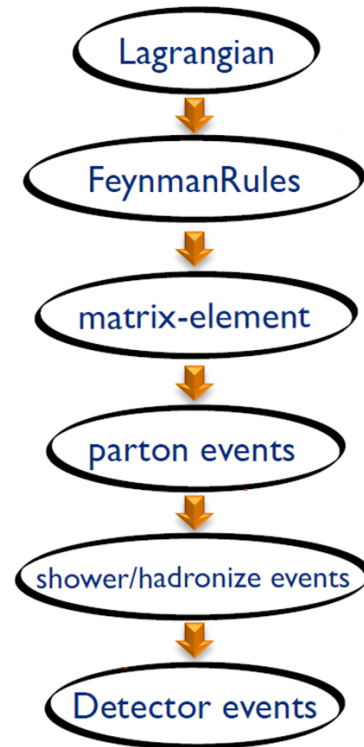


Particle-Matter interaction

- Describe the geometry and the material of the detector
- Treat a particle at a time
- Trajectory of the particle is split in steps (finite displacements)
- Simulate the physics along a step and at the end of each step

Digitization

- convert the energy deposit into electric signal
- Identify the sensitive part of the detector

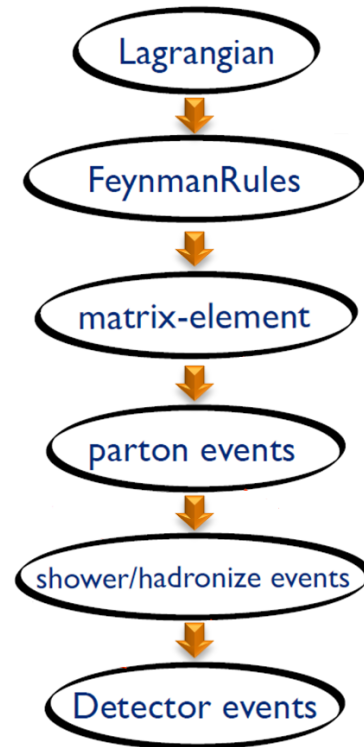


Particle-Matter interaction

- Describe the geometry and the material of the detector
- Treat a particle at a time
- Trajectory of the particle is split in steps (finite displacements)
- Simulate the physics along a step and at the end of each step

Digitization

- convert the energy deposit into electric signal
- Identify the sensitive part of the detector
- Modelize detector answer



MC Simulation of Particle Interactions with Matter

➤ The exponential law:

$P(x)$: probability of not having an interaction after a distance x

$w dx$: probability of having an interaction between x and $x+dx$

↓
depends on
material and
physical process

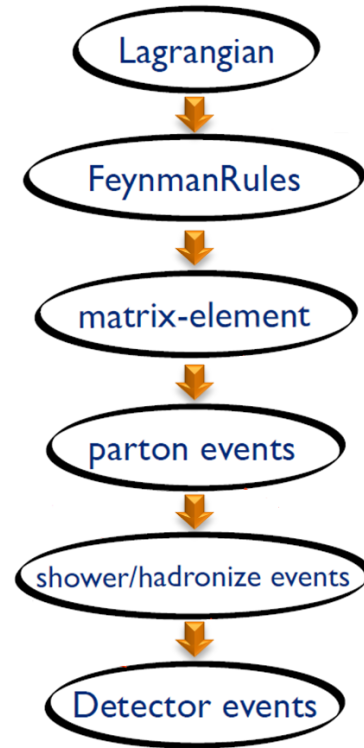
$$P(x+dx) = P(x)(1-w dx)$$

$$P(x) = e^{-wx}$$

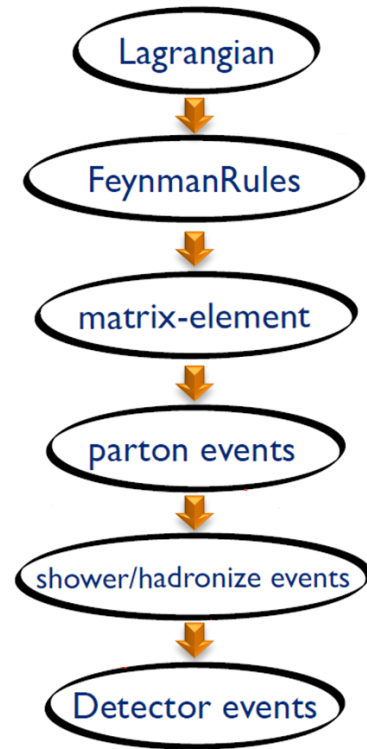
$$P_{int}(x) = 1 - e^{-wx}$$

Now use the inverse method to generate an interaction:

$$P_{int} = \alpha : \text{uniform random number of } [0,1] \Rightarrow xw = -\ln(1-\alpha)$$

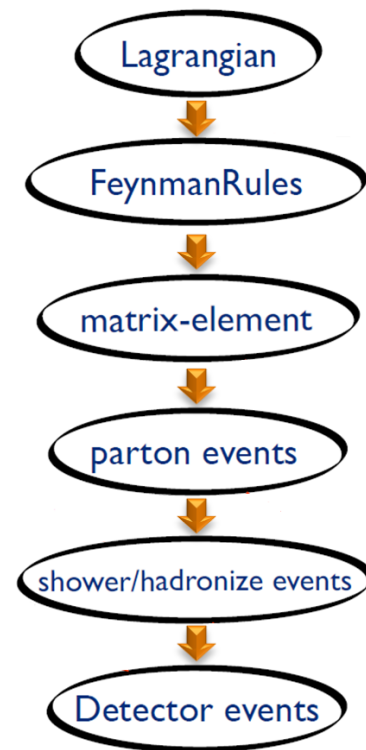


Particle Transportation: How to Determine a Step



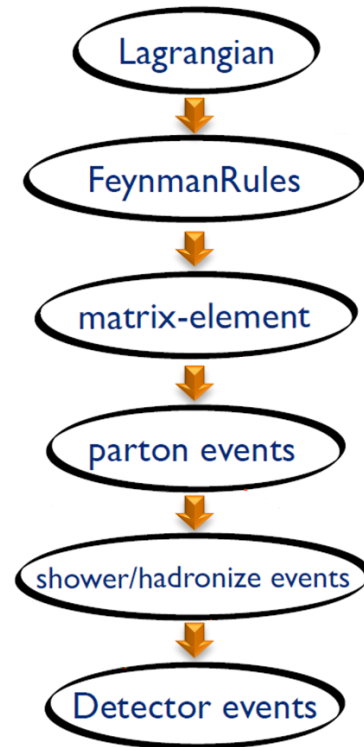
Particle Transportation: How to Determine a Step

1) Evaluate x using α, w for each physical process independently ($xw = -\ln(1-\alpha)$)



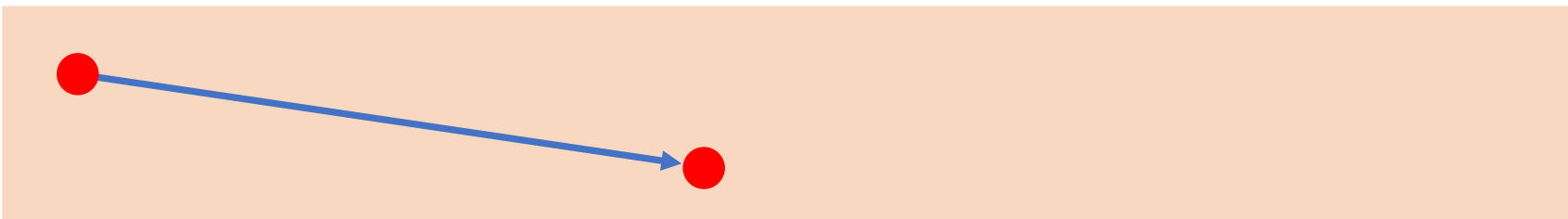
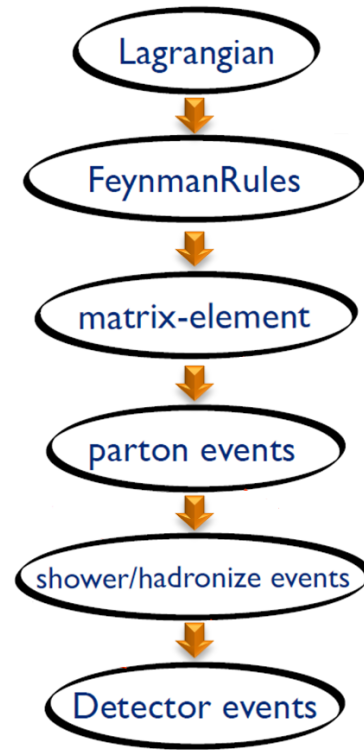
Particle Transportation: How to Determine a Step

- 1) Evaluate x using α, w for each physical process independently ($xw = -\ln(1-\alpha)$)
- 2) Compare: process with **minimum** x determines the step length



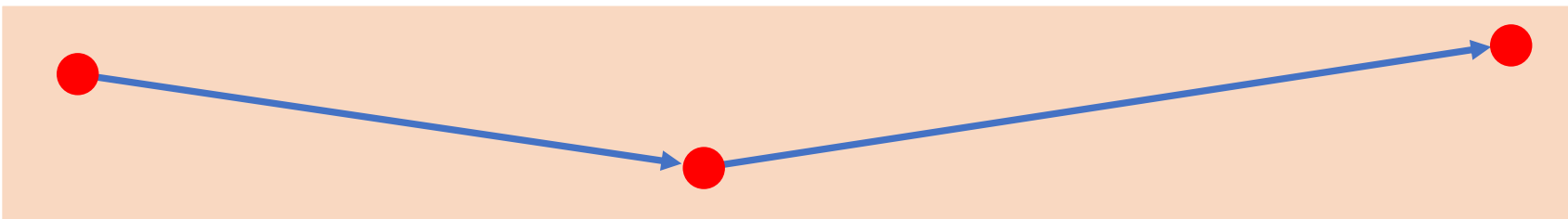
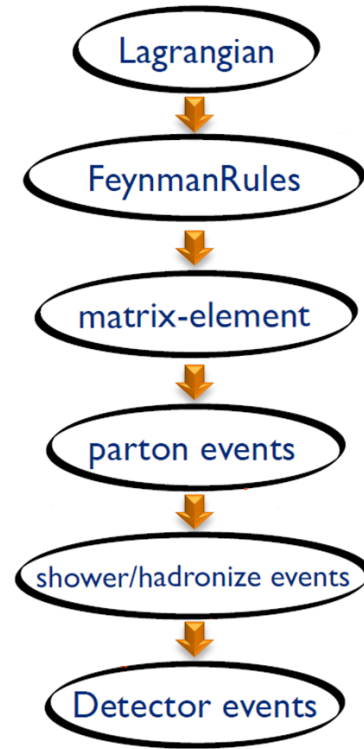
Particle Transportation: How to Determine a Step

- 1) Evaluate x using α, w for each physical process independently ($xw = -\ln(1-\alpha)$)
- 2) Compare: process with **minimum** x determines the step length
- 3) Transport particle for the determined step



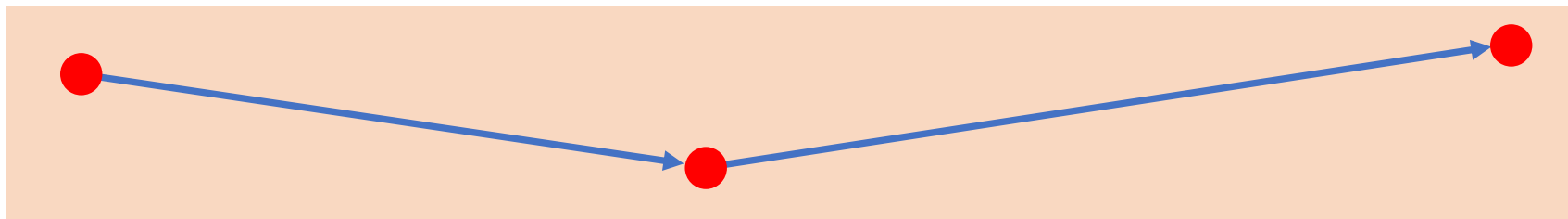
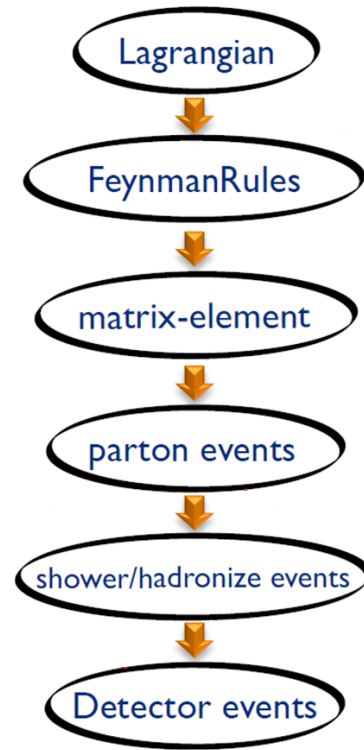
Particle Transportation: How to Determine a Step

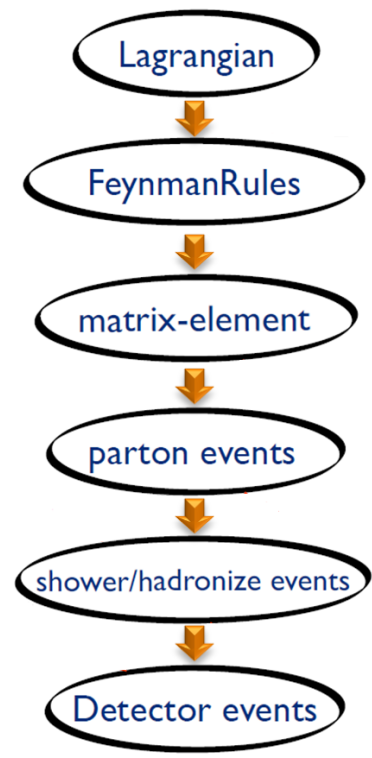
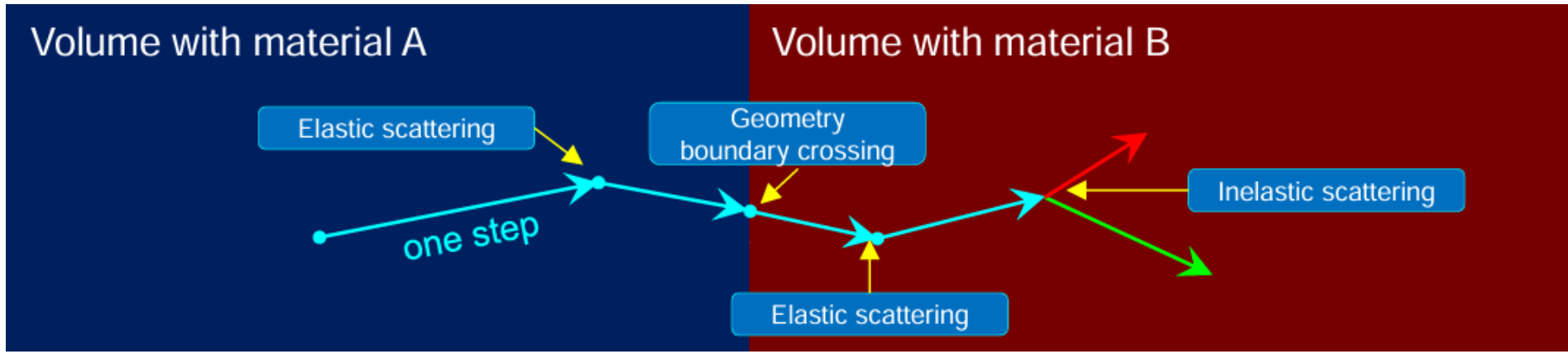
- 1) Evaluate x using α, w for each physical process independently ($xw = -\ln(1-\alpha)$)
- 2) Compare: process with **minimum x** determines the step length
- 3) Transport particle for the determined step
- 4) If the particle is still alive after the interaction, do the **sampling again and continue transportation**



Particle Transportation: How to Determine a Step

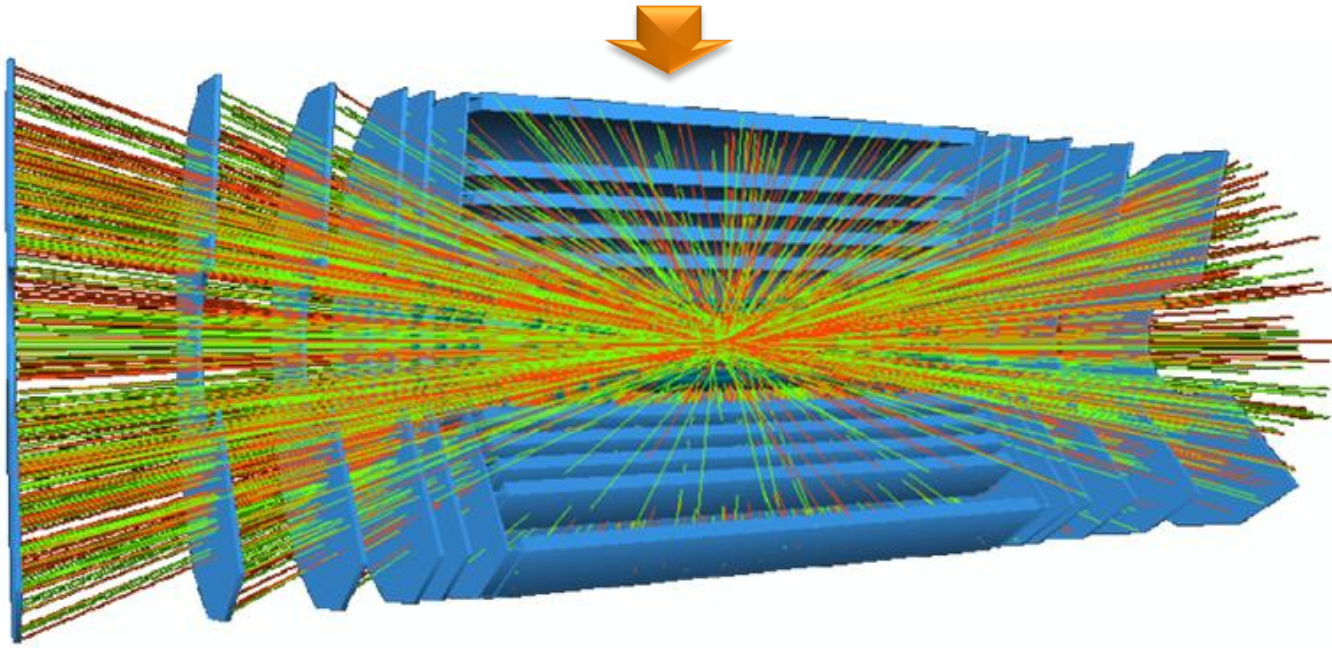
- 1) Evaluate x using α, w for each physical process independently ($xw = -\ln(1-\alpha)$)
- 2) Compare: process with **minimum x** determines the step length
- 3) Transport particle for the determined step
- 4) If the particle is still alive after the interaction, do the **sampling again and continue transportation**
- 5) If the particle disappears after the interaction, then the **transportation is terminated**





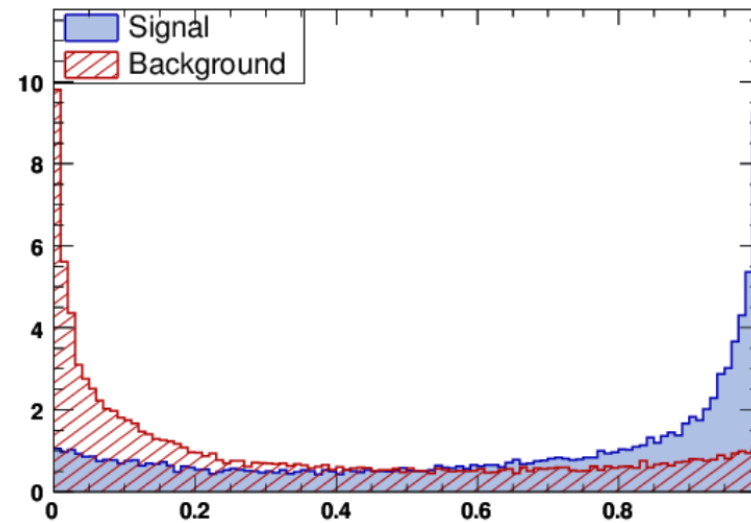
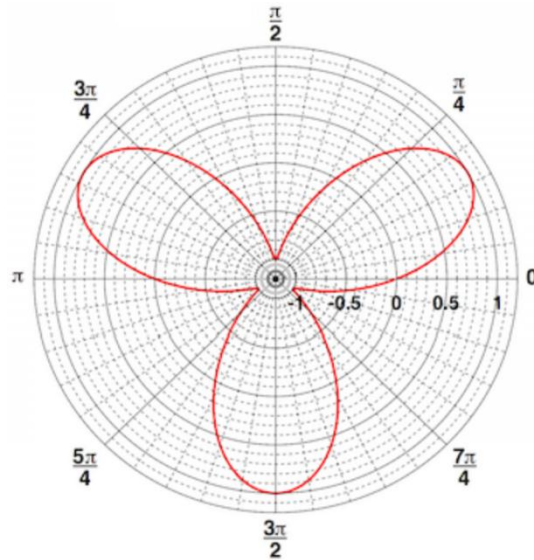
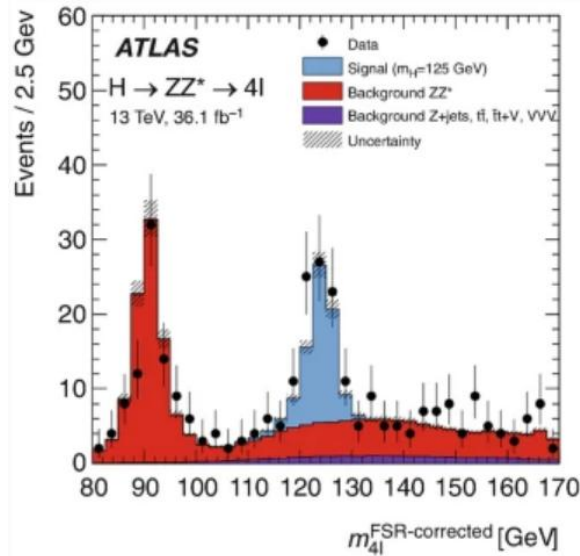
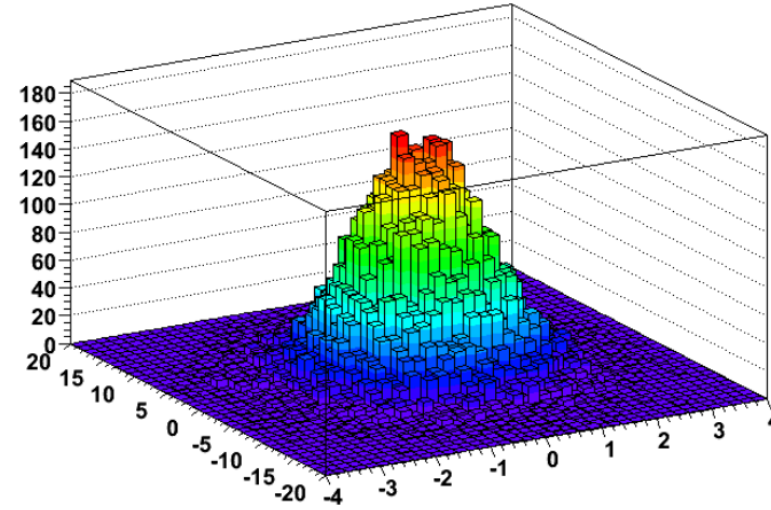
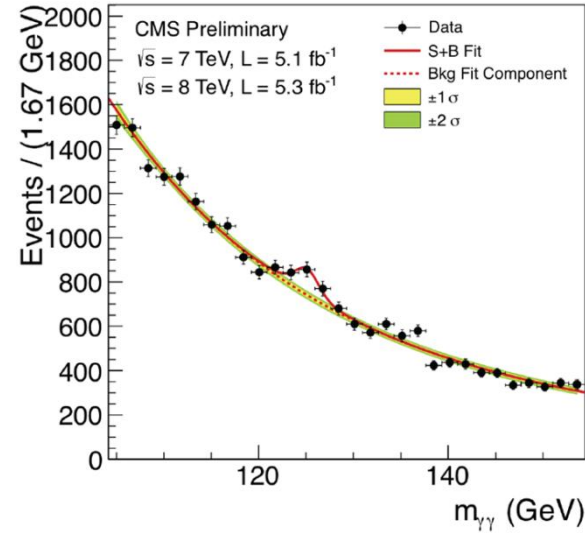
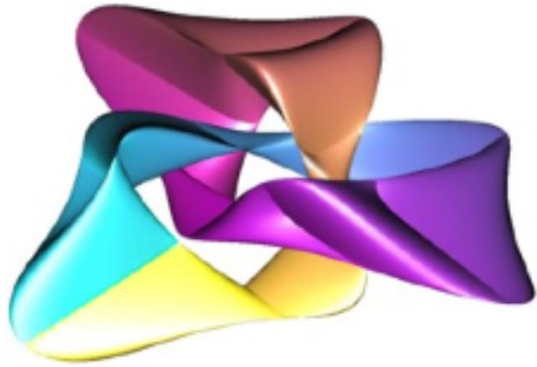
*Then
modelize the
detector response
using a digitizer

GEANT is done*



ROOT

[10.1016/S0168-9002(97)00048-X]



❖ *ROOT is written in C++ and is designed for*

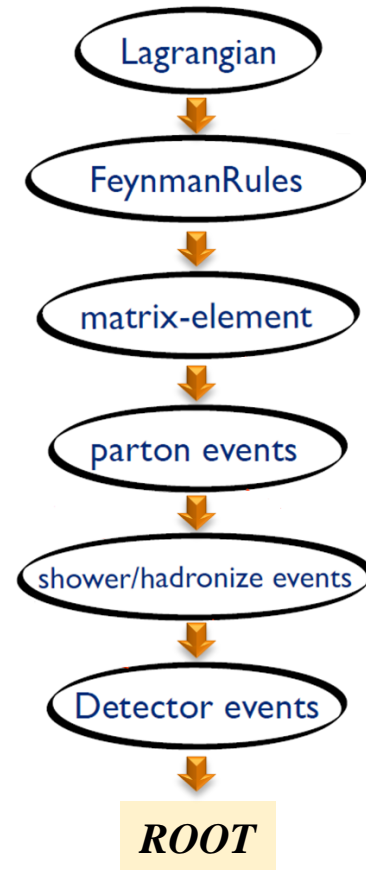
- *Data processing*
- *Data analysis*
- *Data visualization*
- *Data storage*

❖ *Widely used in High Energy Physics and other sciences/industry*

❖ *Can be used for petabytes/year rates of data*

❖ *Provides Python Bindings* C++ 

❖ *I/O: row-wise, column-wise storage of any C++ object*

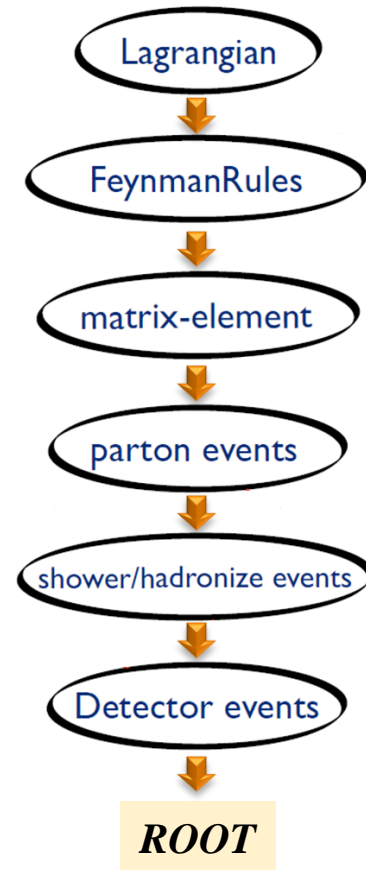


❖ Modes of work:

- *Interactive (ROOT prompt with CLING interpreter)*
 - *interpreted C++ commands*
 - *Macros : interpreted or (Just In Time) compiled*
- *As compilable C++ code : using Root libraries*

```
[haghighat@SuperMicro10 SR1]$ root
-----
| Welcome to ROOT 6.22/06                               https://root.cern |
| (c) 1995-2020, The ROOT Team; conception: R. Brun, F. Rademakers |
| Built for linuxx8664gcc on Nov 27 2020, 15:14:08      |
| From tags/v6-22-06@v6-22-06                          |
| Try '.help', '.demo', '.license', '.credits', '.quit'/'.'q' |
-----

root [0] █
```



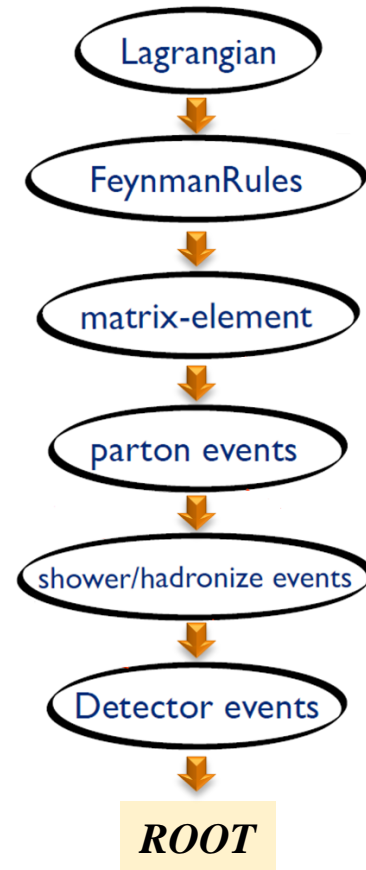
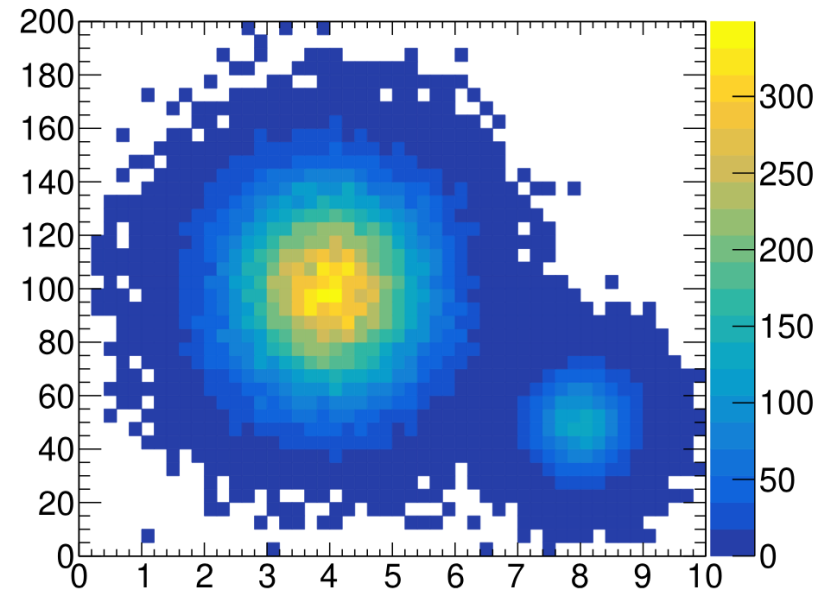
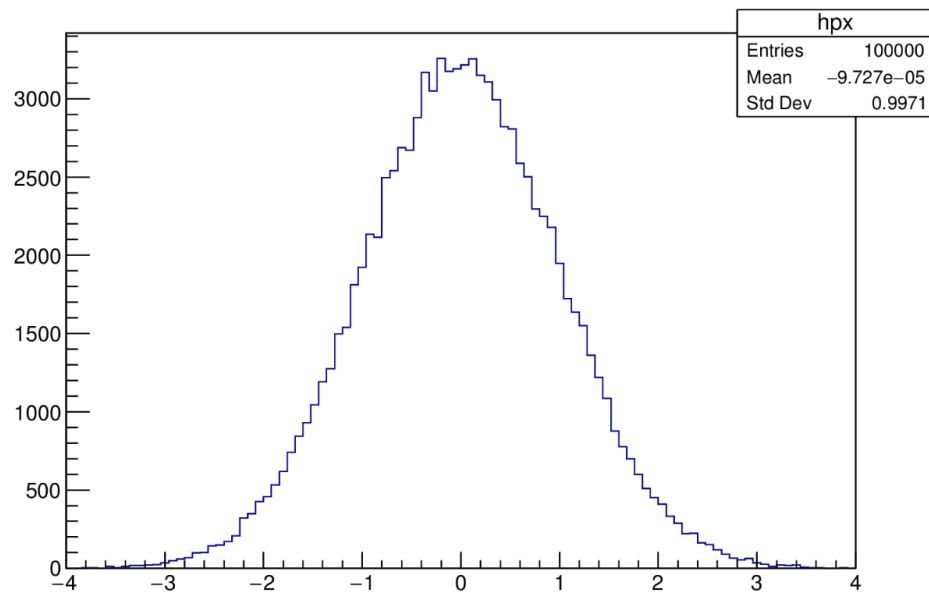
❖ *Examples of what ROOT provides:*



❖ Examples of what ROOT provides:

➤ Histograms, graphs, trees, ntuples: *TH1, TGraph, TTree, TNtuple*

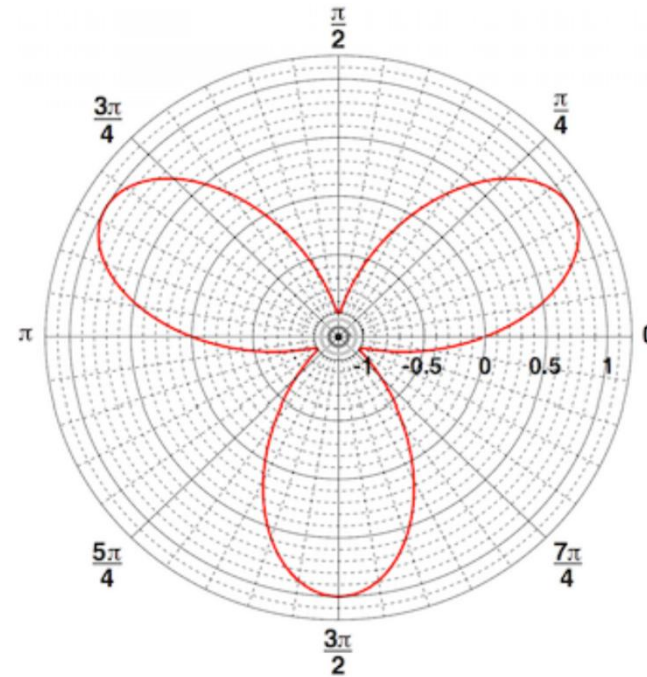
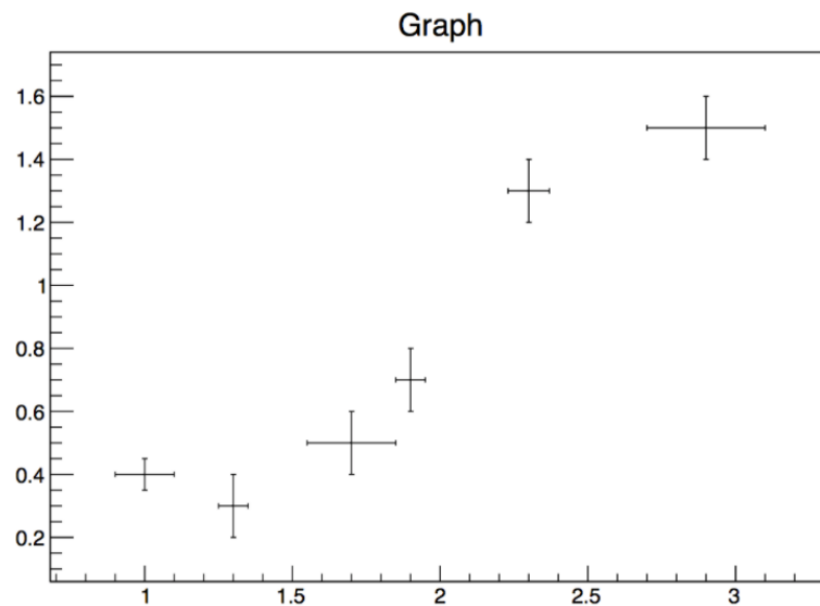
Histograms



❖ Examples of what ROOT provides:

➤ Histograms, graphs, trees, ntuples: *TH1, TGraph, TTree, TNtuple*

Graphs

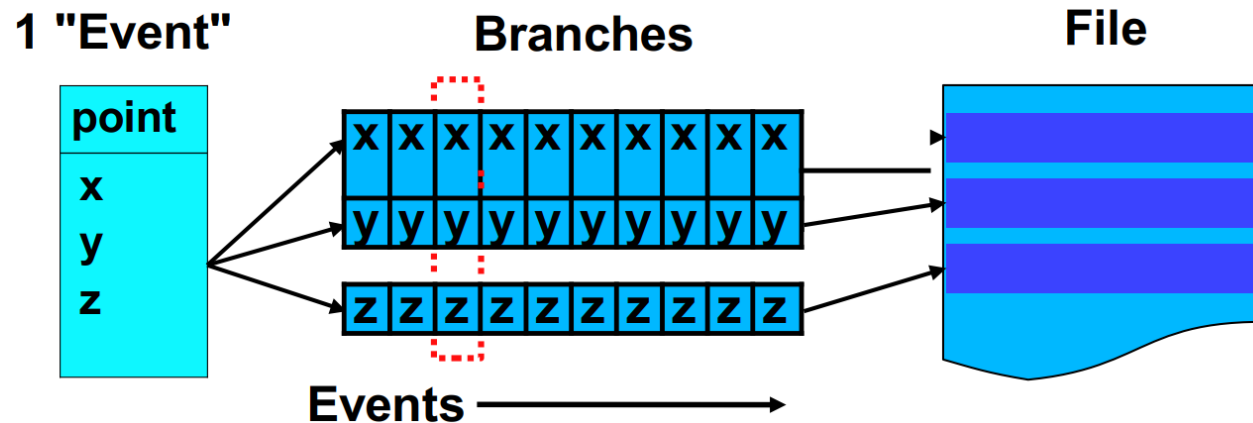


❖ Examples of what ROOT provides:

➤ Histograms, graphs, trees, ntuples: *TH1, TGraph, TTree, TNtuple*

Trees

- Data structure provided to store large quantities of objects
- Organized in branches, each one holding objects
- Organized in independent events, e.g. collision events
- Efficient disk space usage, optimized I/O runtime

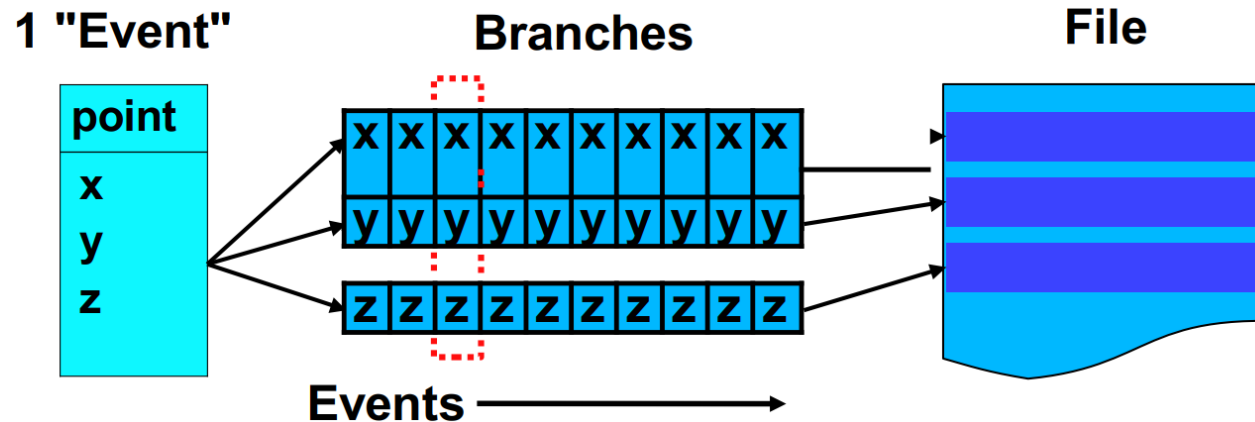


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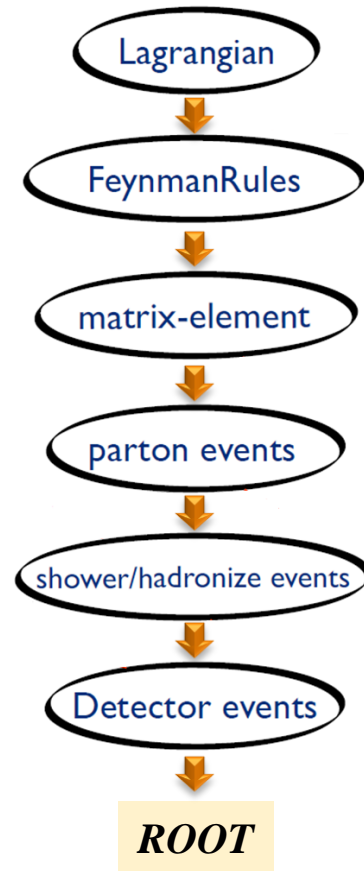
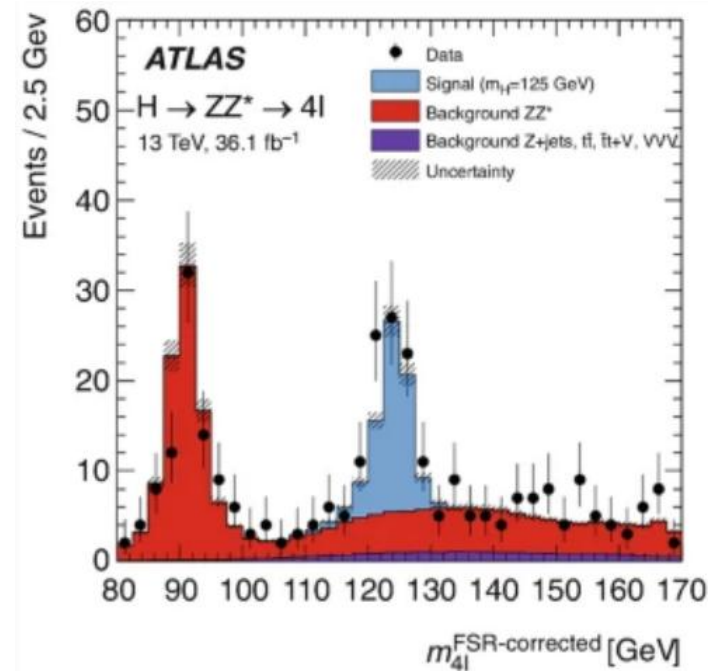
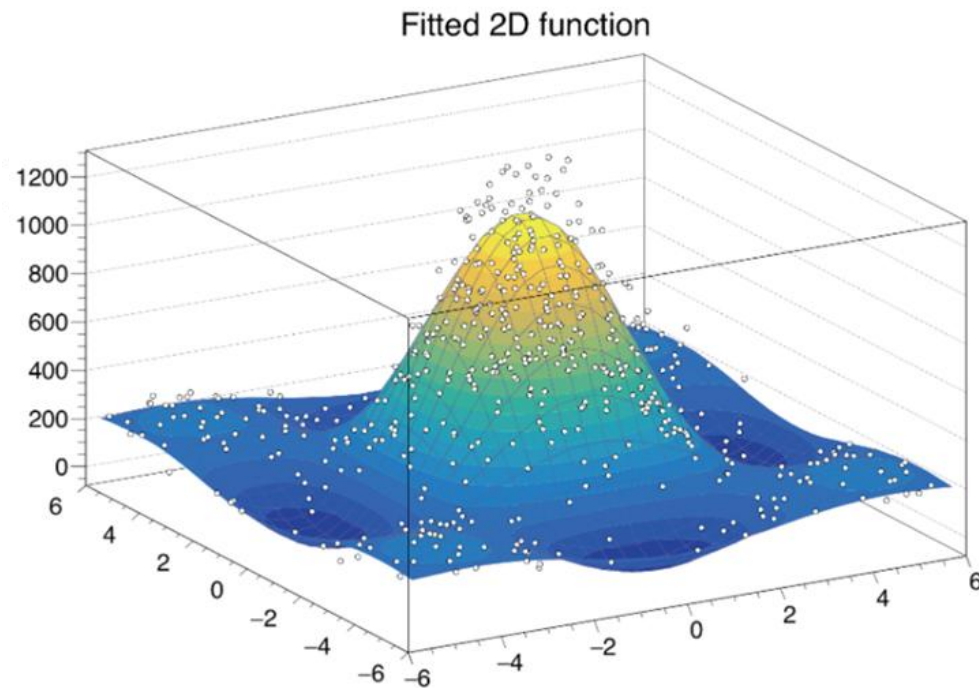
Ntuples

A simplified version of the TTree: store only floating point numbers



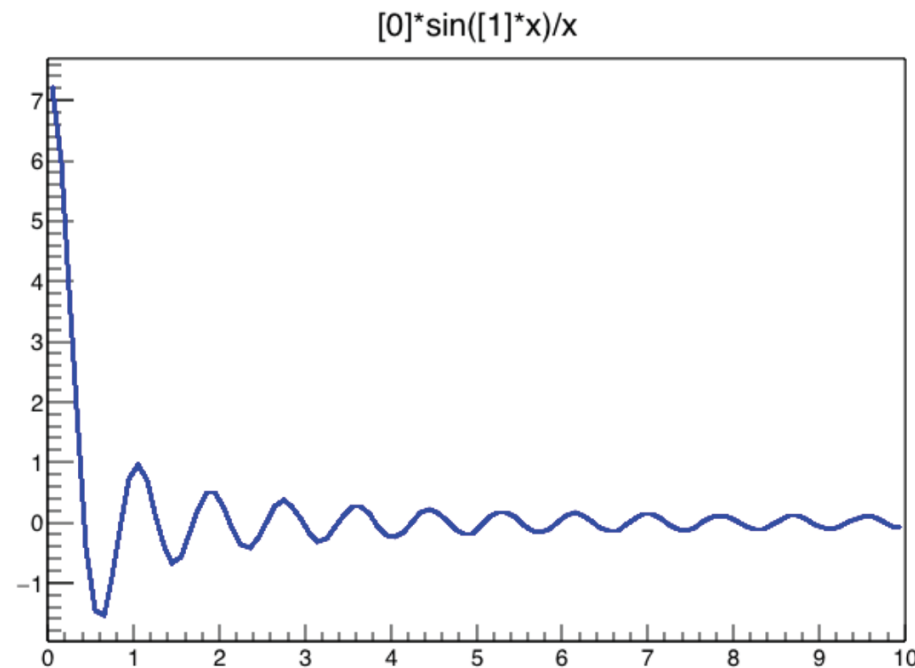
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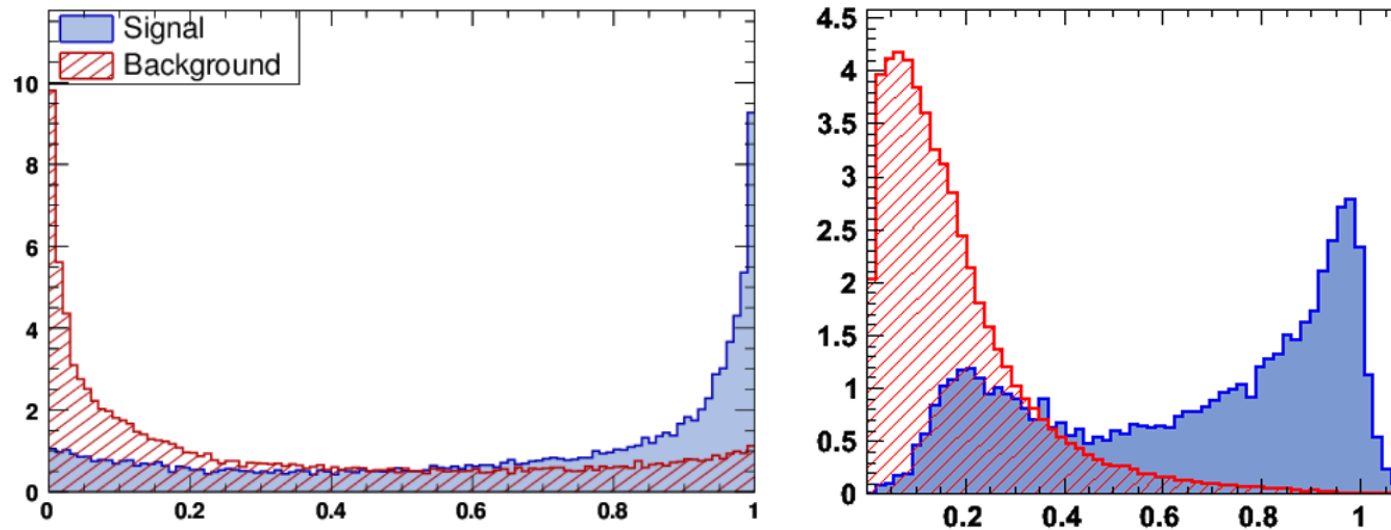
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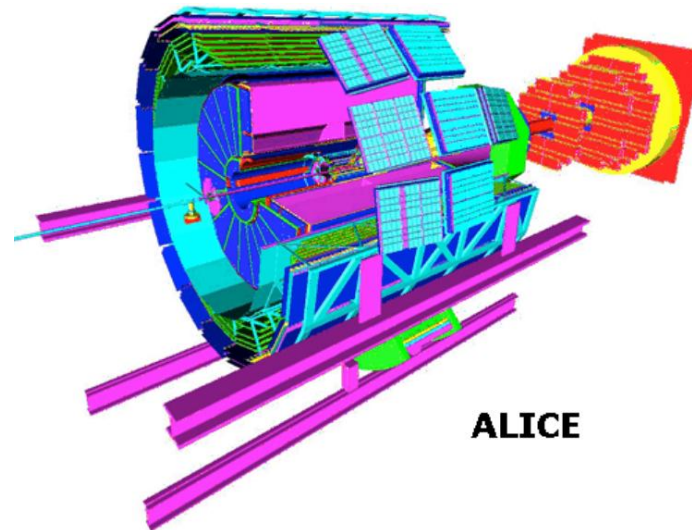
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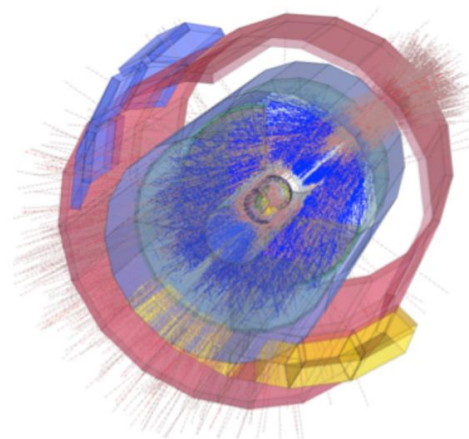
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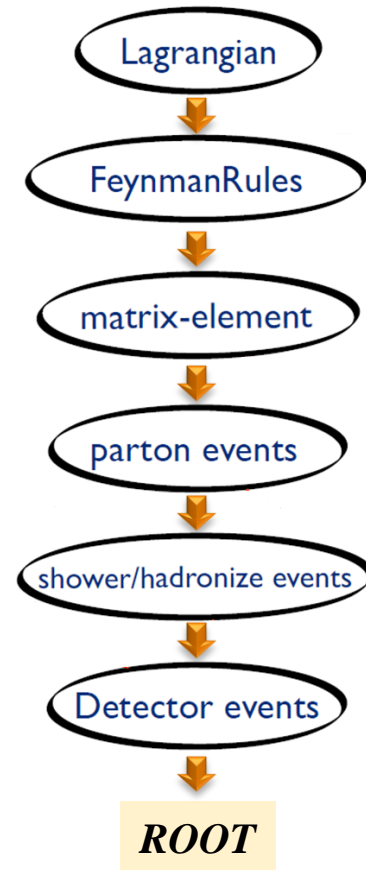
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- ***PyROOT**: bindings to interface to Python*
- ***PROOF**: parallel analysis facility*
 - *Run in parallel on a large number of computers*
 - *Proof-lite: use multiple cores to run on a desktop machine*



*We will see how to use the packages
in the hands on session*



Thank you