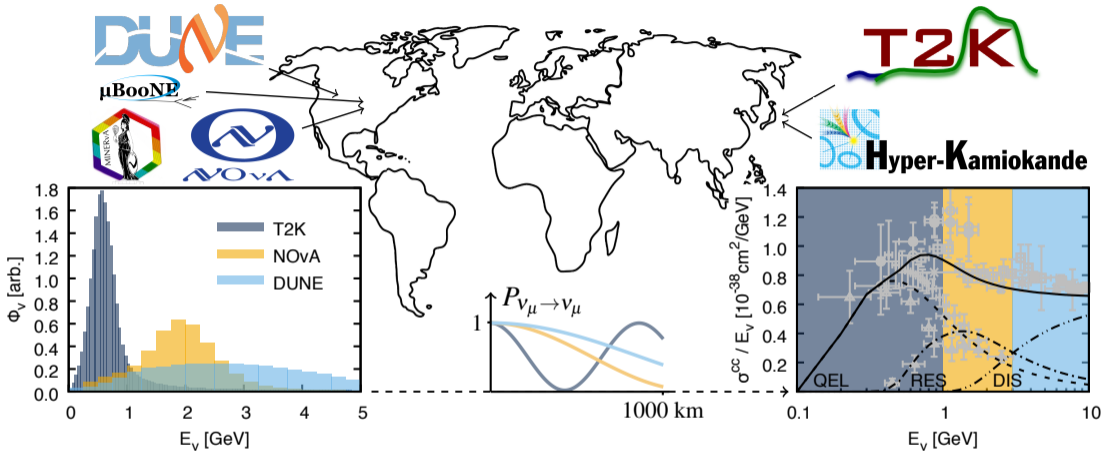


How generators work ...and why it's wrong*

Kajetan Niewczas





$$P(\nu_{\mu} \rightarrow \nu_e) \simeq \sin^2(2\theta) \sin^2\left(1.27 \frac{\Delta m^2 L}{E_{\nu}}\right)$$

↑
oscillation

↑
amplitude

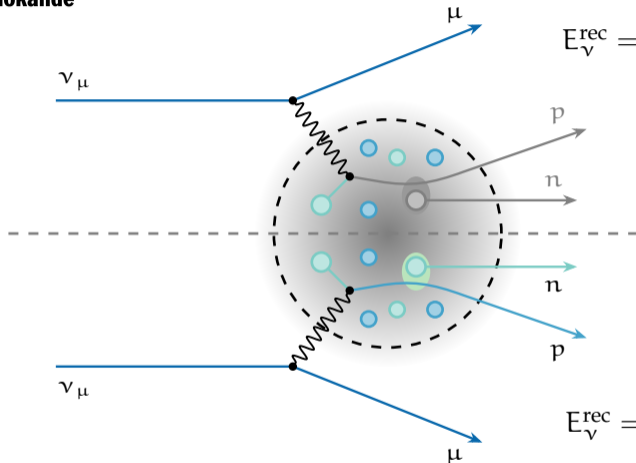
↑
frequency

$$A_{CP} = \frac{P(\nu_{\mu} \rightarrow \nu_e) - P(\bar{\nu}_{\mu} \rightarrow \bar{\nu}_e)}{P(\nu_{\mu} \rightarrow \nu_e) + P(\bar{\nu}_{\mu} \rightarrow \bar{\nu}_e)}$$

↑
asymmetry

↑
oscillation ratio

Kinematical energy reconstruction



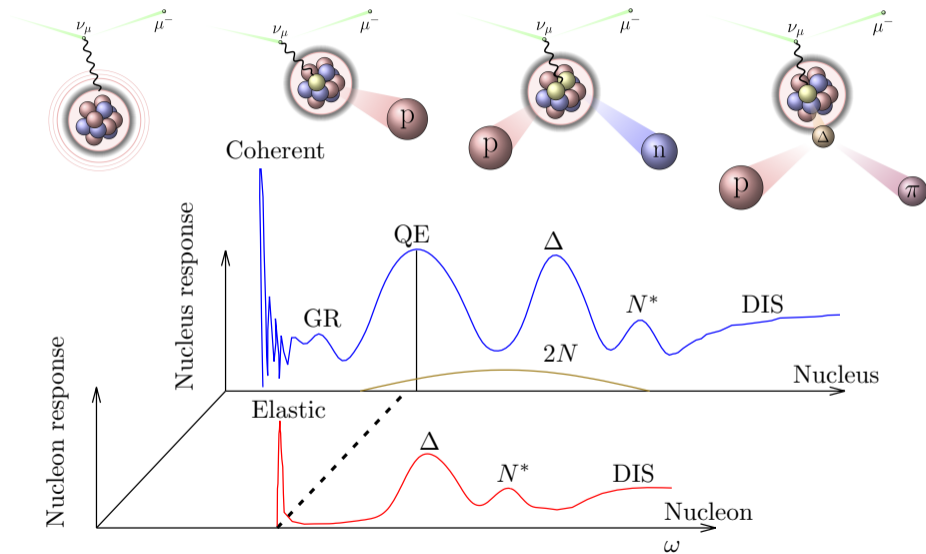
$$E_{\nu}^{\text{rec}} = \frac{2M_N E_{\mu} - m_{\mu}^2 + M_{N'}^2 - M_N^2}{2(M_N - E_{\mu} + p_{\mu} \cos \theta)}$$

$$E_{\nu}^{\text{rec}} = E_{\mu} - E_B + \sum_{\text{nucl.}} T_i + \sum_{\text{mes.}} E_j$$



Calorimetric energy reconstruction

Nuclear response



Basic aspects of the methodology

From Solitaire to the Monte Carlo method

At Los Alamos, Stanisław Ulam, John von Neumann, Nicholas Metropolis and others invented a method to obtain stochastic predictions for **systems too complex to be solved analytically**.

- We can build a complex model using **simple components**
- Every required decision is **made stochastically** (randomly)
- We **run** such a model **N times** and analyze the output
- For $N \rightarrow \infty$ we approach the **correct result** for the model



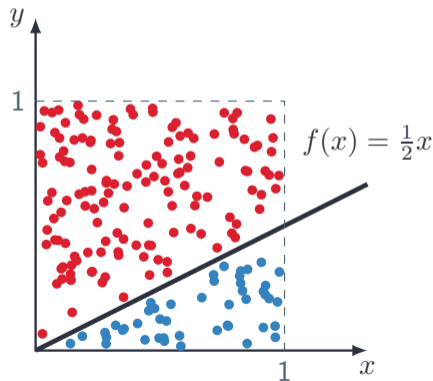
Electronic Numerical Integrator and Computer (ENIAC)

Monte Carlo integration (hit-or-miss method)

Let's consider the **following integral**

$$\int_0^1 f(x) dx = \int_0^1 \left(\frac{1}{2}x\right) dx = \frac{1}{2} \frac{x^2}{2} \Big|_0^1 = \frac{1}{4}$$

- Take a random point from $[0; 1] \times [0; 1]$
- Compare it to your $f(x)$
- Repeat N times
- Count n points below the function
- The **result is given by** $\int_0^1 f(x) dx = A_{\square} \frac{n}{N}$



Monte Carlo integration (crude method)

Let's consider the **following integral** (again)

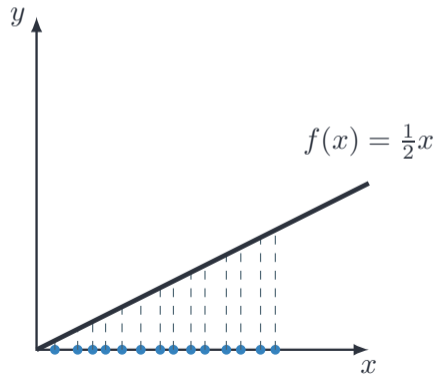
$$\int_0^1 f(x) dx = \int_0^1 \left(\frac{1}{2}x\right) dx = \frac{1}{2} \frac{x^2}{2} \Big|_0^1 = \frac{1}{4}$$

→ One can **approximate the integral** by

$$\int_a^b f(x) dx = \frac{b-a}{N} \sum_i^N f(x_i)$$

where x_i is a random number from $[a; b]$

→ This method leads to slightly higher precision

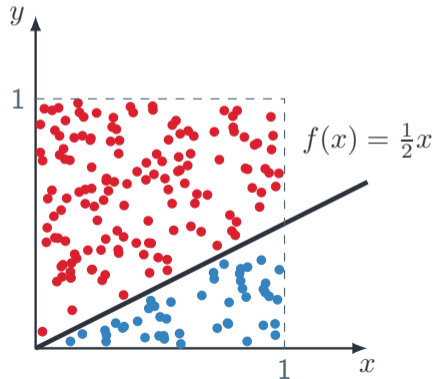


Accept-or-reject algorithm

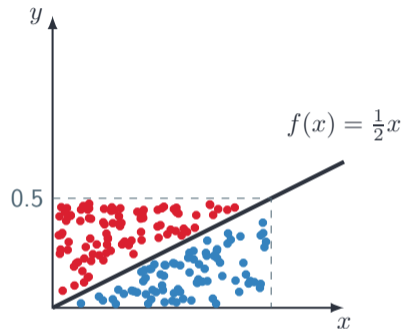
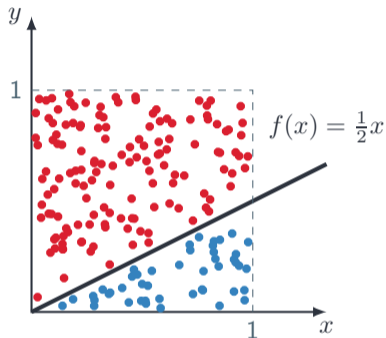
Let's generate a set of points that follows the **probability distribution given by our function**

$$f(x) = \frac{1}{2}x$$

- Find a suitable $f_{\max} \geq \max(f)$
- Take a **random** x from $[0; 1]$
- Generate a **random** u from $[0, f_{\max}]$
- **Accept the point x if $u \leq f(x)$** ($P = \frac{f(x)}{f_{\max}}$)



Optimization



→ We wish to **increase the efficiency** of our calculation by avoiding losing the "red" points

→ We can choose any envelope to our function as long as it **contains the maximum of $f(x)$**

Importance sampling

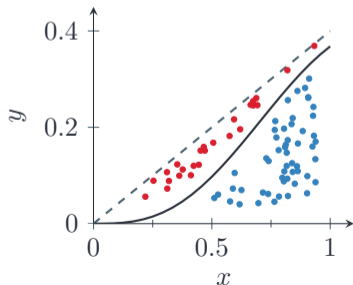
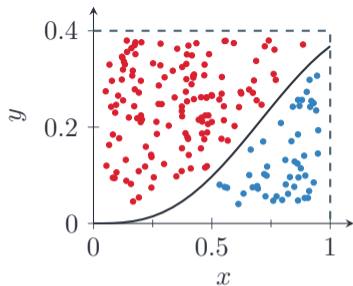
Let's consider a different function

$$f(x) = A x^3 e^{-x^2}$$

with x in $[0; 1]$ and $A = \frac{2e}{e-2}$

- The area under $f(x)$ is ~ 0.13 , while the total is 0.4
 - Having a good **x-dep. envelope increases efficiency**
- For, e.g., $g(x) = 0.4x$, the total area is 0.2
- We can accept points using a **probability**

$$P(x) = f(x)/g(x)$$



Importance sampling

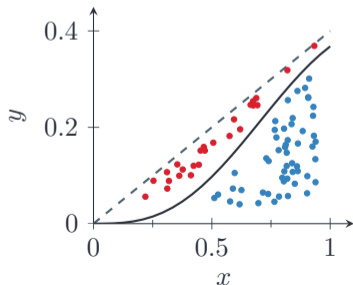
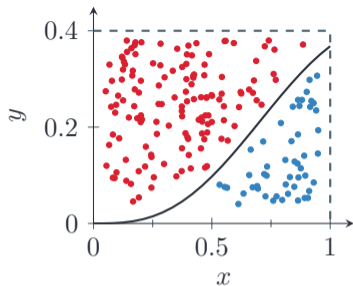
Let's consider a different function

$$f(x) = A x^3 e^{-x^2}$$

with x in $[0; 1]$ and $A = \frac{2e}{e-2}$

- Alternatively, we could change our random numbers from **non-uniform distributions**
 - We can pick more points where the function is larger
- We can **sample points from** $g(x) = 0.4x$, and correct the probability

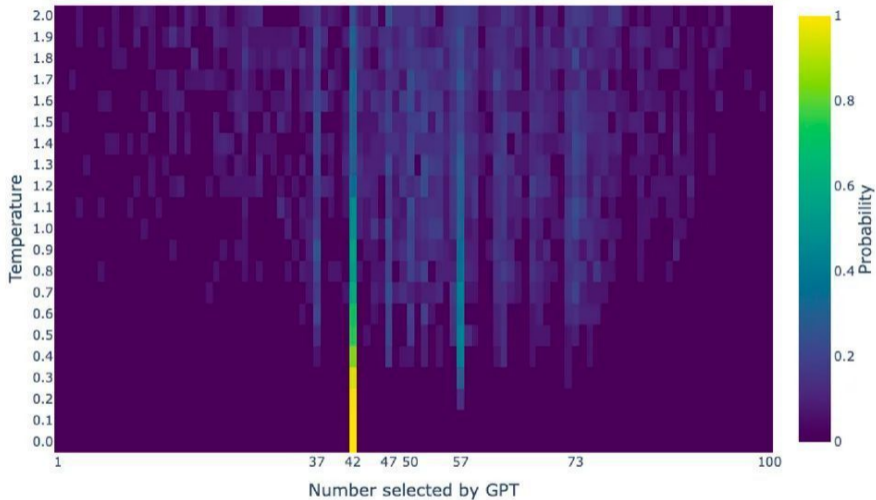
$$\int_0^1 f(x) dx \rightarrow \int_0^1 \frac{f(x)}{g(x)} [g(x) dx]$$



Problems to consider

- How to generate random numbers from a uniform distribution?
- What does it mean for a random number generator to be good?
- How to generate numbers from discrete random variables?
- How to generate numbers from continuous random variables?
- How to generate numbers from arbitrary distributions?

Distribution of selected numbers: "Choose an integer number between 1 and 100"



Example I: Nucleon propagation

Classical particle propagating through a medium

Probability for a particle to propagate over a distance x with **no interactions** is

$$P(x) = \frac{1}{\lambda} \exp(-x/\lambda)$$

where $\lambda = (\rho\sigma)^{-1}$ is the **mean free path**, while ρ is **target density** and σ is **interaction cross section**

We can try to apply it to nucleons in nuclei because:

$$\tilde{\lambda} \ll d < \lambda < R$$

where $\tilde{\lambda}$ is the de Broglie **wavelength**, d is the **distance** between targets, and R is the **nuclear radius**

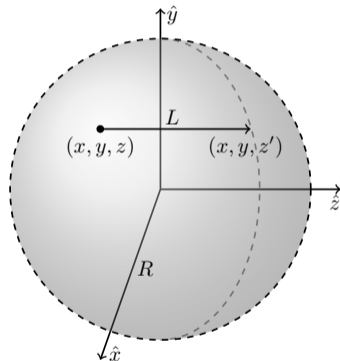
N. Metropolis et al., Phys.Rev. 188 (1958) 185, Phys.Rev. 188 (1958) 204

Application to nuclei (space-like approach)

- Pick a random starting point in the nucleus
 - Propagate the nucleon in discrete steps, e.g., $\Delta x = 0.2$ fm
 - At every step, we sample χ from $P(\chi) = \lambda^{-1} \exp(-\chi/\lambda)$
- If $\chi < \Delta x$, then the nucleon-nucleon interaction happens
- The probability that the nucleon leaves the nucleus with no re-interactions is called **transparency**
- Our procedure solves an integral

$$T = \int_0^{2R} f_R(z) e^{-z/\lambda} dz$$

where $f_R(z)$ is the distribution of the starting points



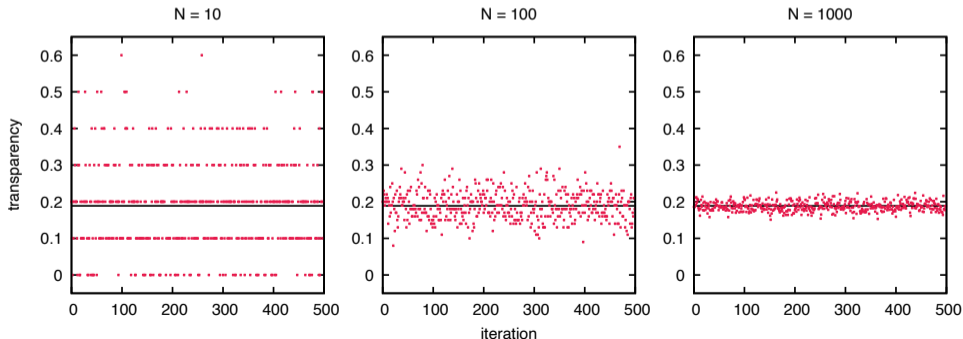
Nuclear transparency

- For fixed density (uniform ball), the **solution is given analytically**

$$T = 3e^{-A} \left(\frac{1}{A^2} + \frac{1}{A^3} \right) + 3 \left(\frac{1}{2A} - \frac{1}{A^3} \right)$$

where $A = 2R/\lambda = 2R\rho\sigma$

→ e.g., for $\rho = 0.16 \text{ fm}^{-3}$, $\sigma = 40 \text{ mb}$, and $R = 6 \text{ fm}$, we get $T \simeq 0.189$



Lepton-nucleus scattering

General remarks

- Leptons are **Standard Model particles**
 - there are better ways to model their interactions than billiard balls
- Lepton-nucleon **cross section is lower** than nucleon-nucleon
 - we do not shoot at nuclei, we start from the primary interaction
 - we need an external lepton-nucleus cross section model
- Lepton-nucleus interactions **require great precision**
 - common detector simulation software do not contain these processes

Independent variables

Unknown particle 4-vectors	Variables	Physical effects	Variables
Initial lepton	4	Particles on-shell	$-(3 + N)$
Target nucleus	4	4-momentum conservation	-4
Final lepton	4	Target rest-frame	-3
Remnant nucleus	4	Fixed projectile direction	-2
Outgoing hadrons	4N	Fixed incoming energy	-1
	16 + 4N		-13 - N
			3 + 3N

Table 1.2: Counting the number of independent variables describing lepton-nucleus interactions while detecting N hadronic particles in the process, summing over the spin of the outgoing lepton, and leaving the remnant nucleus undetected.

Cross sections

Target	Process	Properties	Example formula
Free nucleon	(Quasi)elastic	$N = 0$, all particles on-shell	$\frac{d\sigma}{dQ^2}$
	Inelastic	$N = 0$, excited hadronic system	$\frac{d^2\sigma}{dQ^2 dW}$
	SPP	$N = 1$, all particles on-shell	$\frac{d^4\sigma}{dQ^2 dW d\Omega_\pi}$
Nucleus	Inclusive	$N = 0$, all hadrons integrated	$\frac{d^2\sigma}{d\Omega'}$
	1p1h	$N = 1$, detected one nucleon	$\frac{d^5\sigma}{dE' d\Omega' d\Omega_{N'}}$
	2p2h	$N = 2$, detected two nucleons	$\frac{d^8\sigma}{dE' d\Omega' dE_{N'} d\Omega_{N'} d\Omega_{N''}}$
	SPP	$N = 2$, detected nucleon and π	$\frac{d^8\sigma}{dE' d\Omega' dE_\pi d\Omega_\pi d\Omega_{N'}}$

Table 1.3: The dimensionality of cross section formulas for the most basic lepton scattering scenarios, off the free nucleon or on the nucleus.

Example II: Quasielastic neutrino-nucleon scattering

Quasielastic scattering on a free nucleon

Llewellyn-Smith formula

$$\frac{d\sigma}{d|q^2|} \left(\begin{array}{l} \nu_l + n \rightarrow l^- + p \\ \bar{\nu}_l + p \rightarrow l^+ + n \end{array} \right) = \frac{M^2 G_F^2 \cos^2 \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right]$$

Notation

- constants: M - nucleon mass, G_F - Fermi constant, θ_C - Cabibbo angle
- $q^2 = (k - k')^2 = (p' - p)^2$ - four-momentum squared, where k, k', p, p' are four-momenta of initial and final lepton, initial and final nucleon
- E_ν - neutrino energy
- $s = (k + k')^2$ and $u = (k - p')^2$ - Mandelstam variables

Quasielastic scattering on a free nucleon

Llewellyn-Smith formula

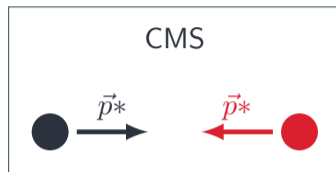
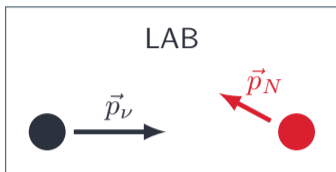
$$\frac{d\sigma}{d|q^2|} \left(\nu_l + n \rightarrow l^- + p \right) = \frac{M^2 G_F^2 \cos^2 \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right]$$

General idea

- having k and p , generate k' and p'
- calculate q^2 and $(s-u) = 4ME_\nu + q^2 - m^2$ based on the generated kinematics
- calculate the cross section
- repeat N times and the result is given by

$$\sigma_{\text{total}} \sim \frac{1}{N} \sum_i^N \sigma(q_i^2)$$

Generating kinematics



- Let's consider kinematics in the center-of-mass system
- Mandelstam s is invariant under Lorentz transformation

$$s = (k + p)^2 = (E + E_p)^2 - (\vec{k} + \vec{p})^2 = (E^* + E_p^*)^2$$

- \sqrt{s} is the total energy in CMS

$$\sqrt{s} = E^* + E_p^* = \sqrt{p^{*2} + m^2} + \sqrt{p^{*2} + M^2}$$

- We will use it to calculate p^*

Generating kinematics

- After some simple algebra

$$\sqrt{s} = E^* + E_p^* = \sqrt{p^{*2} + m^2} + \sqrt{p^{*2} + M^2}$$

$$\sqrt{s} = E^* + \sqrt{E^{*2} - m^2 + M^2}$$

$$s = E^{*2} + E^{*2} - m^2 + M^2 + 2E^*E_p^*$$

$$s = 2E^*(E^* + E_p^*) - m^2 + M^2$$

$$s = 2E^*\sqrt{s} - m^2 + M^2$$

and

$$E^* = \frac{s + m^2 - M^2}{2\sqrt{s}}$$

$$E_p^* = \frac{s + M^2 - m^2}{2\sqrt{s}}$$

- We use this result to get

$$p^* = \sqrt{E^{*2} - m^2} = \frac{[s - (m - M)^2][s - (m + M)^2]}{2\sqrt{s}}$$

Generating events

- We use spherical coordinate system to determine momentum direction in CMS

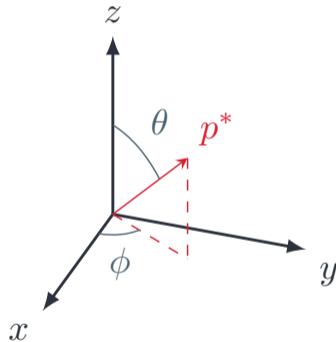
$$\vec{p}^* = p^* \cdot (\sin \theta \cos \phi, \sin \theta \sin \phi, \cos \theta)$$

- Generating random angles

$$\phi^* = 2\pi \cdot \text{random}[0; 1] \quad \rightarrow \quad \sin \phi^*, \cos \phi^*$$

$$\cos \theta^* = 2 \cdot \text{random}[0; 1] - 1 \quad \rightarrow \quad \sin \theta^*, \cos \theta^*$$

- Now, we need to come back to the LAB frame



Boosting between frames

- Lorentz boost in the direction $\hat{n} = \frac{\vec{v}}{v}$ of (t, \vec{r})

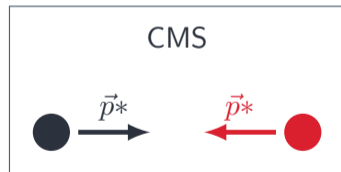
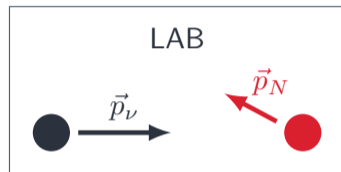
$$t' = \gamma(t - v\hat{n} \cdot \vec{r})$$

$$\vec{r}' = \vec{r} + (\gamma - 1)(\hat{n} \cdot \vec{r})\hat{n} - \gamma tv\hat{n}$$

- In our case

$$\vec{v} = \frac{\vec{p}_\nu + \vec{p}_N}{E_\nu + E_N}$$

- Boost from LAB to CMS in the \vec{v} direction
- Boost from CMS to LAB in the $-\vec{v}$ direction



Calculating the cross section

Llewellyn-Smith formula

$$\frac{d\sigma}{d|q^2|} \left(\nu_l + n \rightarrow l^- + p \right) = \frac{M^2 G_F^2 \cos^2 \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right]$$

Calculation

- once we have p', k' in the LAB frame, we can calculate q^2 and $(s-u)$
- once we have q^2 , we can calculate $A(q^2)$, $B(q^2)$, and $C(q^2)$
- we have all we need to calculate the final result

→ but we changed variables so we need a Jacobian!

Calculating the cross section

- Express q^2 in terms of the lepton angle (in CMS)

$$q^2 = (k - k')^2 = (k^* - k'^*)^2 = m^2 - 2k^* \cdot k'^* = m^2 = 2EE' + 2|\vec{k}^*||\vec{k}'^*| \cos \theta^*$$

- Thus, the Jacobian is given by

$$dq^2 = 2|\vec{k}^*||\vec{k}'^*| d \cos \theta^*$$

- The total cross section is given by

$$\sigma = \int_{-1}^1 \frac{M^2 G_F^2 \cos \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right] 2|\vec{k}^*||\vec{k}'^*| d \cos \theta^*$$
$$\sigma_{MC} = \frac{2}{N} \sum_i^N \frac{M^2 G_F^2 \cos \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right] 2|\vec{k}^*||\vec{k}'^*|$$

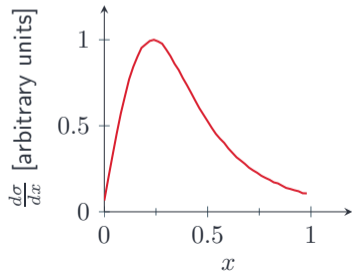
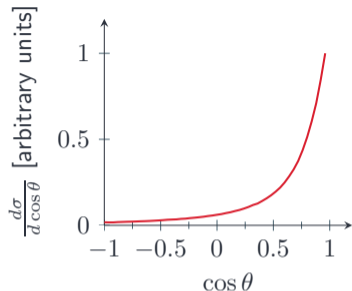
Calculating the cross section

- We want to avoid any sharp peaks as they affect our efficiency and accuracy
- Let's change the variable once again

$$\cos \theta^* = 1 - 2x^2, \quad x \in [0; 1]$$

- Notice, the new Jacobian and integration limits

$$2 \int_{-1}^1 d \cos \theta^* \rightarrow \int_1^0 dx (-4x) \rightarrow \int_0^1 4x dx$$



Calculating the cross section

- Finally, the total cross section is given by

$$\sigma = \int_{-1}^1 \frac{M^2 G_F^2 \cos \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right] 2|\vec{k}^*||\vec{k}'^*| 4x \, dx$$
$$\sigma_{MC} = \frac{1}{N} \sum_i^N \frac{M^2 G_F^2 \cos \theta_C}{8\pi E_\nu^2} \left[A(q^2) \mp B(q^2) \frac{(s-u)}{M^2} + C(q^2) \frac{(s-u)^2}{M^4} \right] 2|\vec{k}^*||\vec{k}'^*| 4x$$

- In conclusion: do some kinematics, add some boosts between CMS and LAB, change the integration variable several times... and you are ready to calculate the cross section

→ Now, we need to generate some events that are distributed according to the cross section formula

Generating events

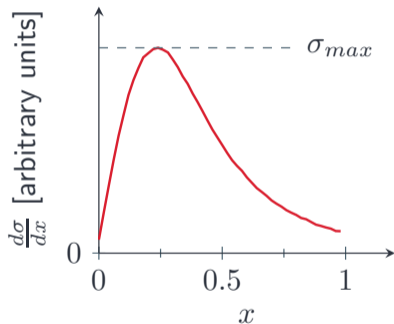
- Take a random x from $[0; 1]$

- Do the kinematics

$$\begin{aligned}x &\rightarrow \cos \theta^* \\ \cos \theta^* &\rightarrow k'^*, p'^* \\ k'^*, p'^* &\rightarrow k', p' \\ &\dots\end{aligned}$$

- Calculate the cross section

- Accept an event with the probability $P(x) = \frac{\sigma(x)}{\sigma_{\max}}$



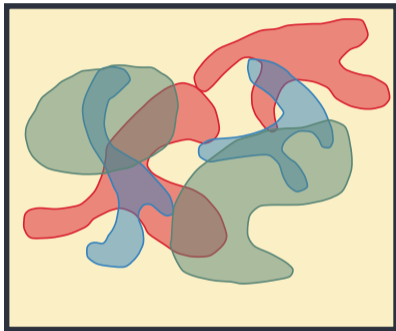
Are we there yet?

A few more steps...

- add **other dynamics**: single-pion production, deep inelastic scattering...
- add support for **nuclei** as **targets**
- if you have a nucleus, add some **two-body current interactions**
- if you have a nucleus, add some nuclear effects: **Pauli blocking**, **final-state interactions**...
- add support for **neutrino beams**
- add support for **detector geometries**
- add some **reweighting**
- ... and your MC generator is done!

Ingredients

Phase space



theory

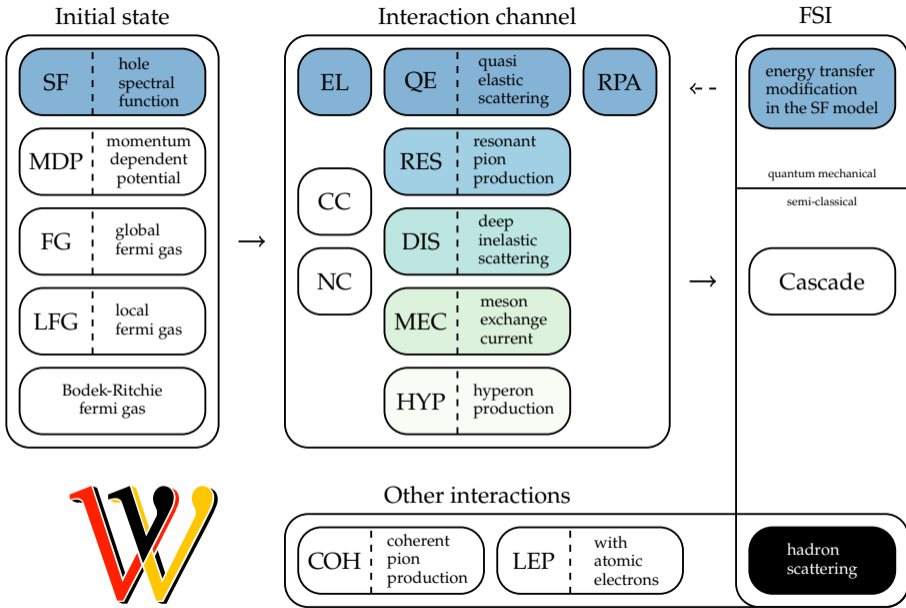
ν data

other data

educated guesses

Recipe





Supplementary material

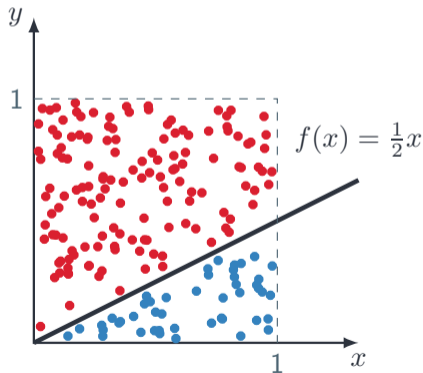
MC integration (hit-or-miss method)

Let's consider the following integral

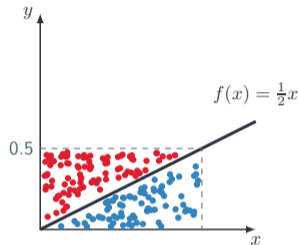
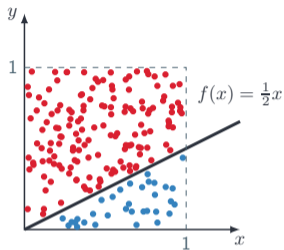
$$\int_0^1 f(x) dx = \int_0^1 \left(\frac{1}{2}x\right) dx = \frac{1}{2} \frac{x^2}{2} \Big|_0^1 = \frac{1}{4}$$

- take a random point from a $[0, 1] \times [0, 1]$ square
- compare it to your $f(x)$
- repeat N times
- count n points below the function
- your result is given by

$$\int_0^1 f(x) dx = A_{\square} \cdot \frac{n}{N}$$



Optimizations



- We want to **avoid** generating “**red**” points as they do not contribute to your integral
- **Any rectangle** can be chosen as far as it **contains maximum** of $f(x)$ in given range

Optimizations

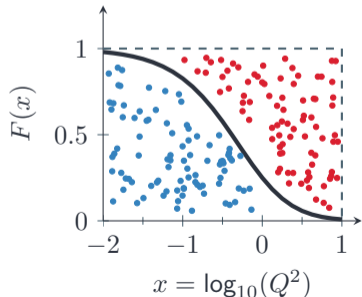
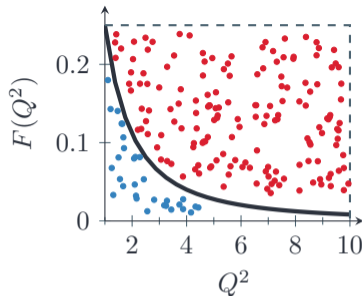
- Consider the following function

$$F(Q^2) = \frac{1}{(1 + Q^2)^2}$$

- Integrating this function over Q^2 is highly inefficient
- One can integrate by substitution to get better performance, e.g.

$$x = \log_{10} Q^2$$

→ don't forget about the Jacobian!



MC integration (crude method)

Let's consider the following integral again

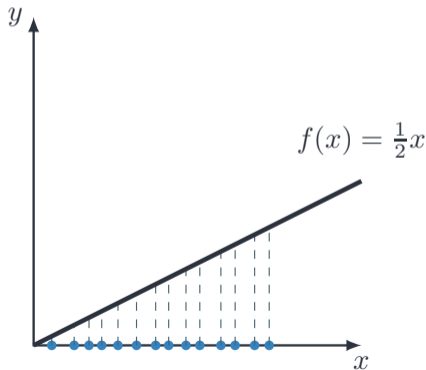
$$\int_0^1 f(x) dx = \int_0^1 \left(\frac{1}{2}x\right) dx = \frac{1}{2} \frac{x^2}{2} \Big|_0^1 = \frac{1}{4}$$

- One can approximate the integral

$$\int_a^b f(x) dx = \frac{b-a}{N} \sum_i^N f(x_i)$$

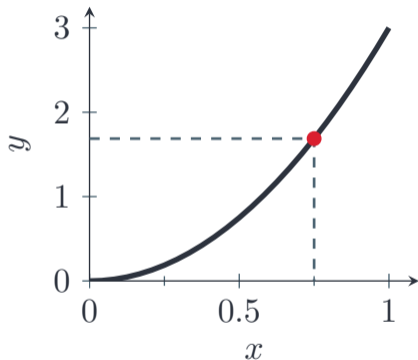
where x_i is a random number from $[a, b]$

- It can be shown that this method is more accurate



Random numbers from probability density functions

- How to generate a random number from probability density function?
- Let's consider $f(x) = 3x^2$
- Which means that $x = 1$ should be thrown 2 times more often than $x = \frac{\sqrt{2}}{2}$



Cumulative distribution function

- Cumulative distribution function of a random variable X :

$$F(x) = P(X \leq x)$$

- Discrete random variable X :

$$F(x) = \sum_{x_i \leq x} f(x_i)$$

where f is a probability mass function (PMF)

- Continuous random variable X :

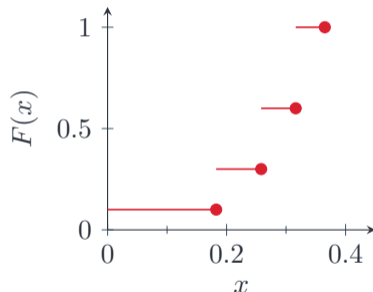
$$F(x) = \int_{-\infty}^x f(t) dt$$

where f is a probability density function (PDF)

Cumulative distribution function - discrete example

- Probability mass function $f(x) = 3x^2$
with discrete random variables X is $\{\sqrt{\frac{1}{30}}, \sqrt{\frac{2}{30}}, \sqrt{\frac{3}{30}}, \sqrt{\frac{4}{30}}\}$
- CDF is given by

$$F(x) = \begin{cases} \frac{1}{10} & \text{if } x \leq \sqrt{\frac{1}{30}} \\ \frac{3}{10} & \text{if } x \leq \sqrt{\frac{2}{30}} \\ \frac{6}{10} & \text{if } x \leq \sqrt{\frac{3}{30}} \\ \frac{10}{10} & \text{if } x \leq \sqrt{\frac{4}{30}} \end{cases}$$



→ generate a random number u from $[0, 1]$

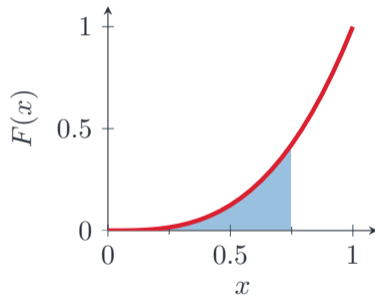
→ if $u \leq 0.1$: $x = \sqrt{\frac{1}{30}}$ → else if $u \leq 0.3$: $x = \sqrt{\frac{2}{30}}$...

Cumulative distribution function - continuous example

- Probability density function $f(x) = 3x^2$
with continuous random variables X range $[0, 1]$
- CDF is given by

$$F(x) = \int_0^x f(t)dt = x^3$$

- generate a random number u
from $[0, 1]$
- find x for which $F(x) = u$,
i.e. $x = F^{-1}(u)$



- Unfortunately, usually F^{-1} is unknown, which makes this method pretty useless (at least directly)

Acceptance-rejection method

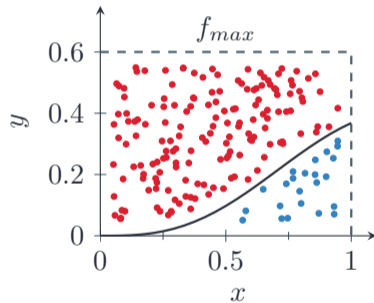
- Let's consider a function $f(x) = A \cdot x^3 \cdot e^{-x^2}$ with $x \in [0, 1]$, $A = \frac{2e}{e-2}$
- CDF is given by $F(x) = \frac{N}{2}(x^2 - 1)e^{-x^2}$ and we don't know F^{-1} !

→ find a suitable $f_{\max} \geq \max(f)$

→ take a random x

→ generate a random u from $[0, f_{\max}]$

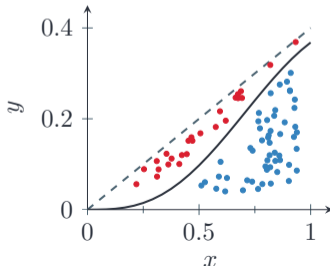
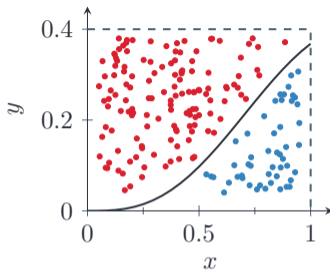
→ accept if $u \leq f(x)$ ($P = \frac{f(x)}{f_{\max}}$)



- The same procedure as the integration via hit-or-miss but focusing on accepted points, not the integral itself

Acceptance-rejection method - optimization

- The area under the plot of $f(x)$ is ~ 0.13 , while the total area is 0.4
 - We can try to limit the number of wasted points taking a better envelope
 - For $g(x) = 0.4x$ the total area is 0.2, so we speed up twice
- The cumulative distribution is $G(x) = x^2$, $G^{-1}(x) = \sqrt{x}$
- Generate a random $u \in [0, 1]$
- Calculate $x = G^{-1}(u)$ and accept with probability $P = f(x)/g(x)$



Importance sampling

- Sampling more significant parts of phase space more often

$$\int_0^1 f(x) dx$$

$$\rightarrow \int_0^1 \frac{f(x)}{g(x)} [g(x) dx]$$

- Mathematically speaking all three methods are equivalent

