

Description and simulation of physics of Resistive Plate Chambers

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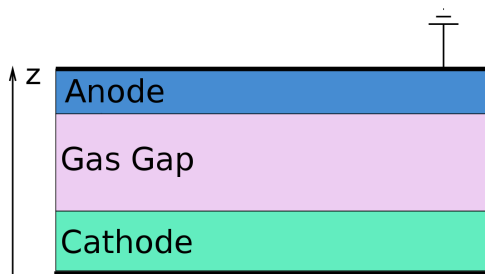
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- 1 Resistive Plate Chambers
 - Basic design
 - Gaseous mixture
 - Context and Objectives
- 2 The physics behind and its simulation
 - Avalanche modelisation
 - Diffusion
 - Space Charge Effect
 - Signal Induction
- 3 Preliminary results
- 4 Conclusion and perspective

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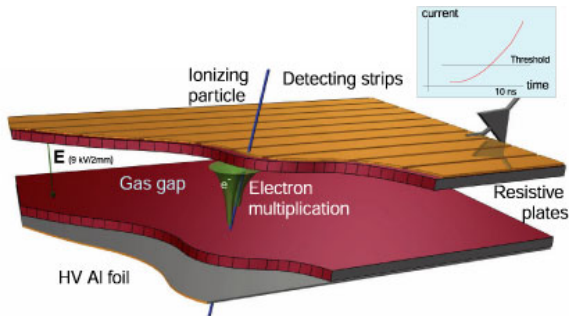
Basic single-gap design

- Gap is 12 mm wide
- Glass anode and cathode ($10^{12} \Omega \text{ cm}$, $\epsilon_r \sim 7$), 7 and 11 mm wide
- HV of 6.9 kV between plates (57.5 kV/cm)



Operation

- Particle crossing the detector → **ionisation**
- Freed electric charges drift and multiply under the influence of the HT ⇒ **electronic (Townsend) avalanche**
- Electric signal arise on the electrode by **induction**



- The gaseous mixture is maybe the most vital part of a RPC as it influences many key characteristics :
 - ionisation (number of electrons freed)
 - multiplication gain
 - electron drift velocity (influences signal amplitude and timing)
- usually mixture is composed of 3 gases :
 1. ionizing gas $\sim 95\%$
 2. UV quencher gas $\sim 4\%$
 3. electron quencher gas $\sim 1\%$
- mixture used for this presentation :
 1. TFE $C_2H_2F_4$ 93%
 2. CO_2 5%
 3. SF_6 2%

State of the art and objectives

- Monte-Carlo simulation is essential in detector development → allows to predict characteristics and responses
 - Simulations for RPC are not widespread and often incomplete
 - ↳ unadapted mathematical distribution (Polya) which lacks physical interpretation
 - ↳ overlook important phenomena
-
- modelise the main processes of an electronic avalanche
 - develop a full, fast and multi-threaded Monte-Carlo simulation
 - portable, easily modifiable and usable on various hardwares

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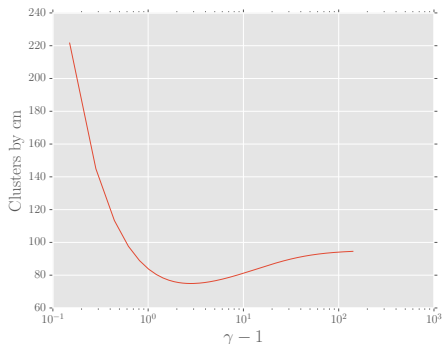
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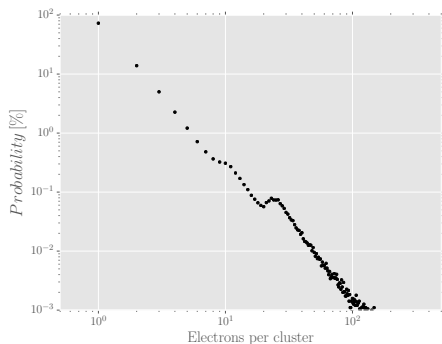
Ionisation

- charged particle crossing the gas gap \rightarrow ionisation
- each ionisation event \Rightarrow **electron clusters**
- charge deposit characterized by two things

\rightarrow the number of clusters by unit of length



\rightarrow the probability distribution for the number of electrons by cluster

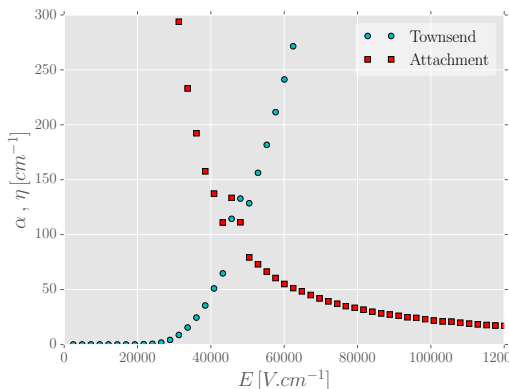


Electron multiplication

- electrons drift under the influence of the electric field and multiply by interaction with gas molecules (avalanche)
- evolution of the number of electrons conditioned by two coefficient :

Townsend coefficient α →
probability to **multiply**

Attachment coefficient η →
probability to get **attached**



Avalanche development model (W. Riegler and C. Lippmann)

- average numbers of e^- and positive ions :

$$\bar{n}(x) = e^{(\alpha-\eta)x}$$

$$\bar{p}(x) = \frac{\alpha}{\alpha - \eta} \left(e^{(\alpha-\eta)x} - 1 \right)$$

- stochastic multiplication and attachment for one e^-

$$n = \begin{cases} 0, & s < k \frac{\bar{n}(x)-1}{\bar{n}(x)-k} \\ 1 + \text{floor} \left[\ln \left(\frac{(\bar{n}(x)-k)(1-s)}{\bar{n}(x)(1-k)} \right) \cdot \frac{1}{\ln \left(1 - \frac{1-k}{\bar{n}(x)-k} \right)} \right], & s > k \frac{\bar{n}(x)-1}{\bar{n}(x)-k} \end{cases}$$

with s a random number $\in [0, 1)$, $k = \eta/\alpha$

here x is the drifted distance

Multiplication procedure

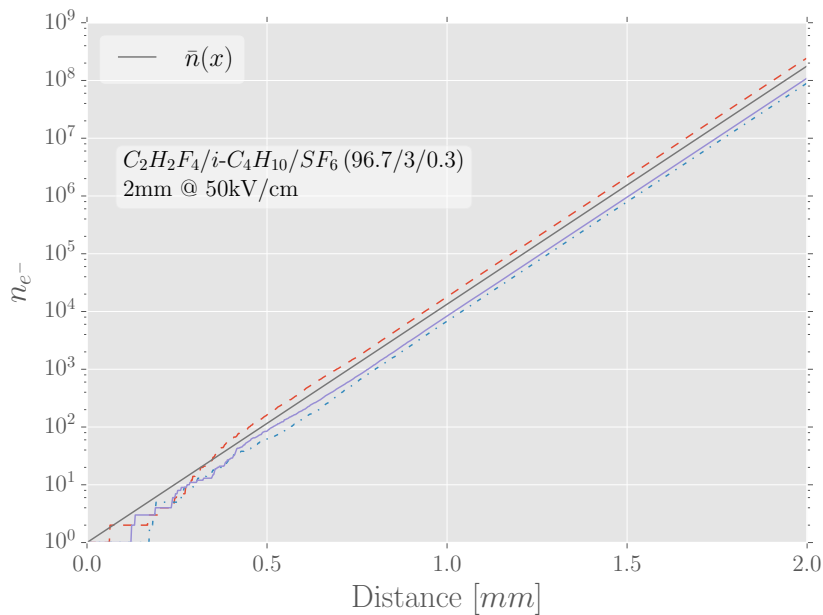
- gas gap divided into N steps of Δx ($\sim \mu m$)
- clusters are put into their respective bin

Case of one cluster at x_0

- n_0 electrons present at x_0
- each one of the n_0 electrons will multiply according to the previous formula and we find n_1 electrons at $x = x_0 + \Delta x$
- In the same way, the n_1 electrons will multiply and we find n_2 electrons at $x = x_0 + 2\Delta x$

→ This procedure is iterated until all the electrons reach the anode

Multiplication procedure

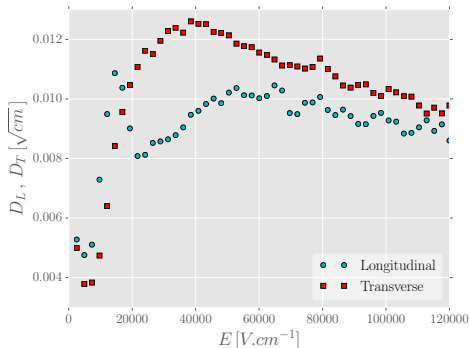


Diffusion

- Thermal diffusion motion superposed by drift motion \Rightarrow anisotropic diffusion

$$\varphi_L = \frac{1}{\sqrt{2\pi}\sigma_L} \exp\left(-\frac{(z - z_0)^2}{2\sigma_L^2}\right)$$

$$\varphi_T = \frac{1}{\sigma_T^2} \exp\left(-\frac{(r - r_0)^2}{2\sigma_T^2}\right)$$

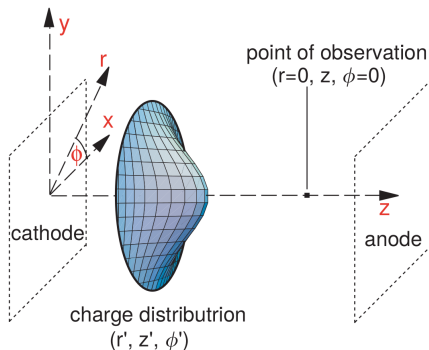


- diffusion characterized by their diffusion coefficient D_L and D_T and drifted distance l

$$\sigma_{L,T} = D_{L,T}\sqrt{l}$$

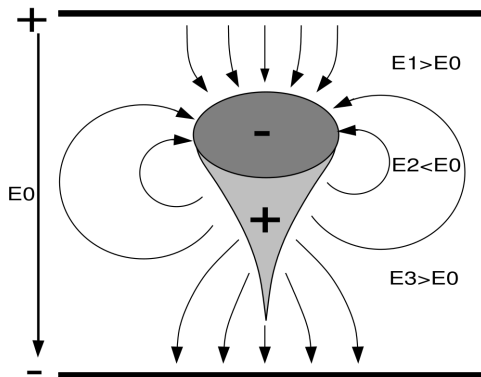
Transverse diffusion

- ⇒ Transversal : we consider the charges to be contained in a disk with a Gaussian radial distribution (φ_T) with $\sigma = D_T\sqrt{l}$ where l is the drifted distance



Space Charge Effect

- When the number of charges in avalanches is high enough they influence the electric field and thus the values of α and $\eta \implies$
Space Charge Effect



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Approximation to *feel*'s impact :
charges lie in sphere of radius r_d

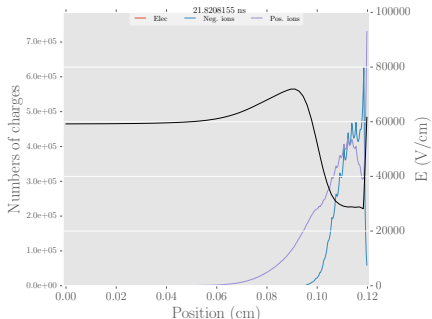
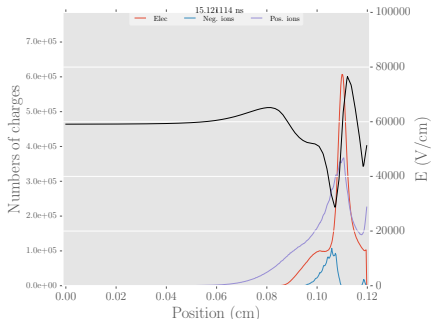
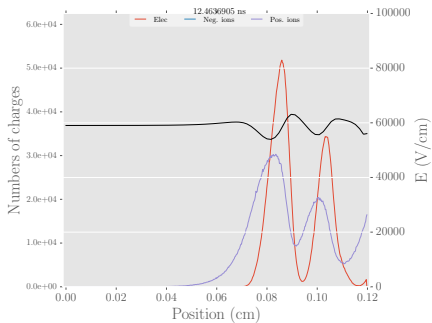
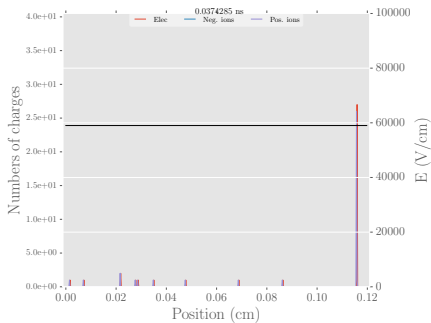
$$E_r = \frac{e_0 n_e}{4\pi \epsilon_0 r_d^2}$$

$$n_e = 10^6 \quad r_d = 0.1 \text{ mm} \\ \Rightarrow E_r = 1.5 \text{ kV/cm}$$

3% of typical RPC field ($\sim 50 \text{ kV/cm}$)
 \rightarrow 10% change in coefficients (and so
in multiplication gain)

- Space Charge Effect leads to a **saturation** of the number of produced electrons
- Fully modelised by computing the field of all the charges in gas gap

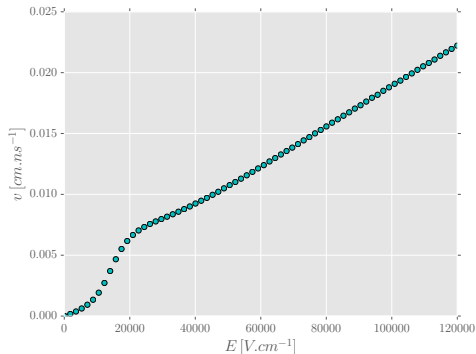
Space Charge Effect illustration



Signal Induction

- Output signal is only due to the movement of electrons in the electric field

- electrons in gas are not collected on electrodes as they are absorbed by resistive layer
- electrons movement induces charges on electrodes
- ions don't contribute due to their small drift velocity



- We use the generalised **Ramo's theorem** to compute induced signals

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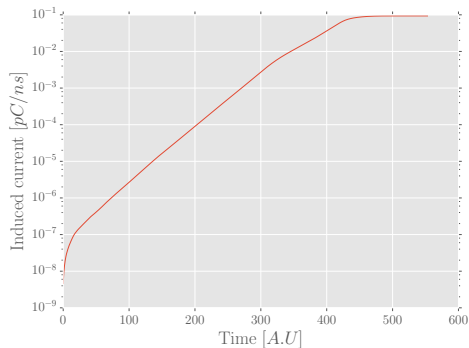
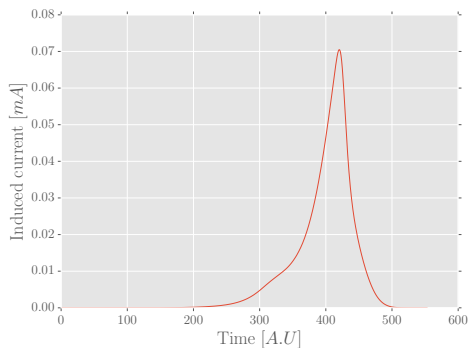
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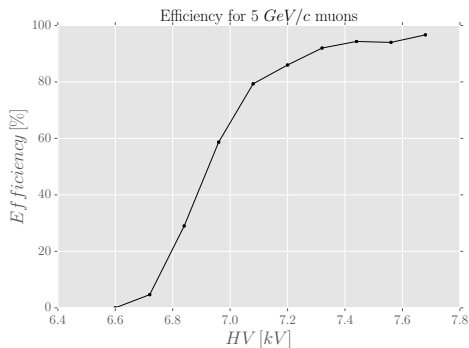
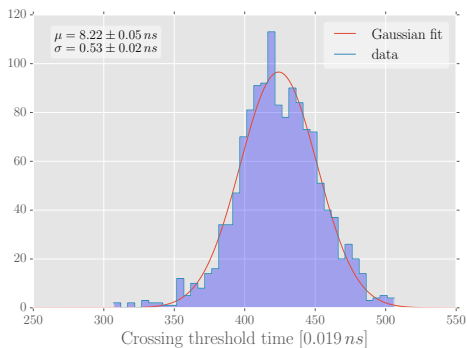
Preliminary results

- Cathode 0.11 cm, Anode 0.07 cm, Gap 0.12 cm, HV 57.5 kV/cm
- Glass @ $10^{12} \Omega cm$



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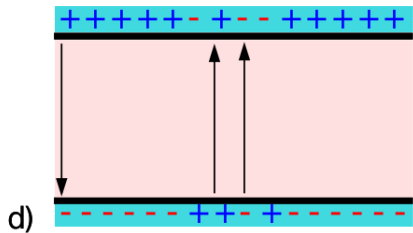
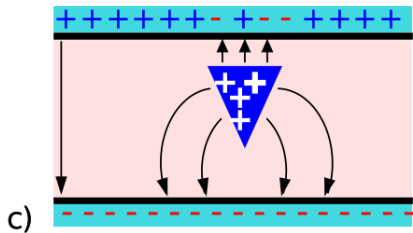
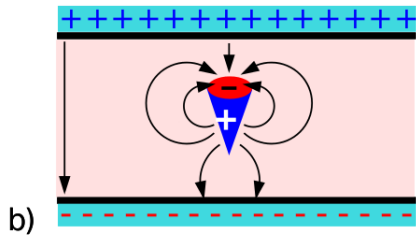
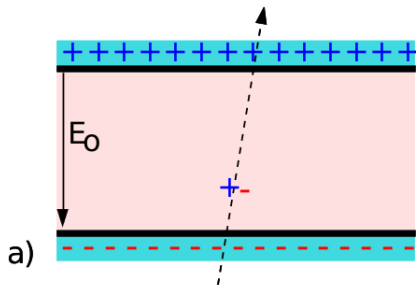
- Resistive Plate Chambers are widely use, yet there is no proper simulation
- Working on a model taking into account the main physics processes
- Still work in progress but taking shape
- Aim to be a basic code for further RPC development and to be hooked in a more global detector simulation chain
- Room for improvements
 - main bottleneck comes from the pseudo-random numbers generation
 - using GPU (CuRand or Thrust) PRNG may give a significant speedup

BACKUPS

The simulation and libraries used

- Compiled on a server with an old GCC (4.4.7) \Rightarrow No C++11 atm
- UNIX POSIX thread library
- Using the ThreadFactory (P. Schweitzer) to spawn threads and allocate events (one thread reserved for output writing)
- RngStreams (L'Ecuyer) for random number generation
- Using Garfield framework (rev. 418) with HEED (1.01) and Magboltz (9.01) for electron gas transport parameters and particle-gas interactions
- Gnu Scientific Library (QUADPACK) for integral computation, can also use python with scipy (more precise but much slower)
- TinyXml2 for configuration file parser
- Except Garfield (which use ROOT) and GSL, doesn't rely on a lot of libraries, all included in src

Electronic avalanche



Avalanche development model continued

$$n = \begin{cases} 0, & s < k \frac{\bar{n}(x)-1}{\bar{n}(x)-k} \\ 1 + \ln \left(\frac{(\bar{n}(x)-k)(1-s)}{\bar{n}(x)(1-k)} \right) \frac{1}{\ln \left(1 - \frac{1-k}{\bar{n}(x)-k} \right)}, & s > k \frac{\bar{n}(x)-1}{\bar{n}(x)-k} \end{cases} \quad \alpha, \eta > 0$$

$$n = \begin{cases} 0, & s < \frac{\alpha x}{1+\alpha x} \\ 1 + \ln [(1-s)(1+\alpha x)] \frac{1}{\ln \left(\frac{\alpha x}{1+\alpha x} \right)}, & s > \frac{\alpha x}{1+\alpha x} \end{cases} \quad \alpha = \eta$$

$$n = \begin{cases} 0, & s < e^{(-\eta x)} \\ 1, & s > e^{(-\eta x)} \end{cases} \quad \alpha = 0$$

Central Limit theorem

CPU-intense procedure \Rightarrow very time consuming !

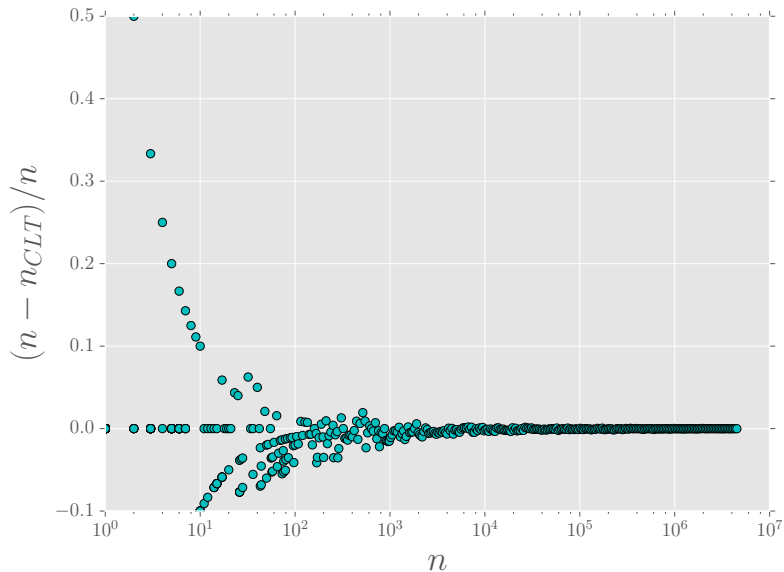
- \rightarrow Unadapted to the simulation of a large number of event
- \rightarrow We make use of the **Central Limit Theorem** :

when n_i is big enough we draw n_{i+1} from a gaussian

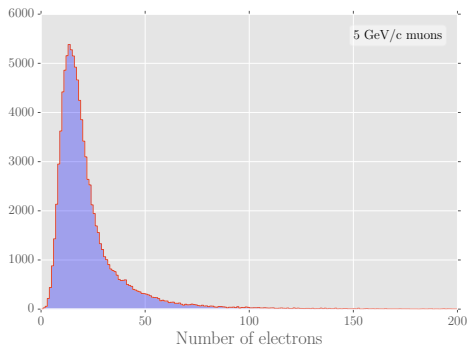
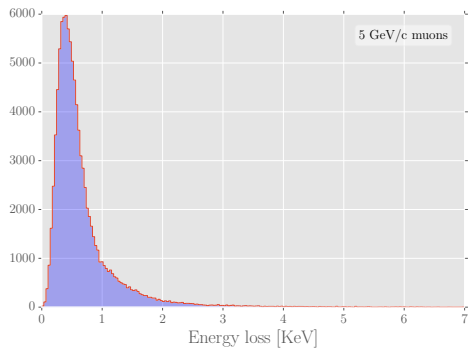
$$\mu = n_i \bar{n}(\Delta x) \quad \sigma_{CLT} = \sqrt{n_i} \sigma(\Delta x)$$

$$\sigma^2(\Delta x) = \left(\frac{1+k}{1-k} \right) \bar{n}(\Delta x) (\bar{n}(\Delta x) - 1)$$

Central Limit theorem



Primary ionisation



Space Charge Effect - potential

$$\Phi(r, \phi, z, r', \phi', z') = \frac{Q}{4\pi\epsilon_2} \left[\frac{1}{\sqrt{P^2 + (z - z')^2}} - \frac{(\epsilon_1 - \epsilon_2)}{(\epsilon_1 + \epsilon_2)\sqrt{P^2 + (z + z')^2}} \right. \\ \left. - \frac{(\epsilon_3 - \epsilon_2)}{(\epsilon_3 + \epsilon_2)\sqrt{P^2 + (2g - z - z')^2}} \right. \\ \left. + \frac{1}{(\epsilon_1 + \epsilon_2)(\epsilon_2 + \epsilon_3)} \int_0^\infty d\kappa J_0(\kappa P) \frac{R(\kappa, z, z')}{D(\kappa)} \right],$$

$$0 \leq z \leq g;$$

$$R(\kappa; z, z') =$$

$$(\epsilon_1 + \epsilon_2)^2 (\epsilon_2 + \epsilon_3)^2 \left[e^{\kappa(-2p-2q+z-z')} + e^{\kappa(-2p-2q-z+z')} \right] -$$

$$(\epsilon_1 + \epsilon_2)^2 (\epsilon_2 - \epsilon_3)^2 e^{\kappa(-4g-2q+z+z')} -$$

$$4\epsilon_1 \epsilon_2 (\epsilon_2 + \epsilon_3)^2 e^{\kappa(-2q-z-z')} - (\epsilon_1 - \epsilon_2)^2 (\epsilon_2 + \epsilon_3)^2 e^{\kappa(-2p-z-z')} -$$

$$(\epsilon_1^2 - \epsilon_2^2) (\epsilon_2 - \epsilon_3)^2 e^{\kappa(-4g+z+z')} +$$

$$(\epsilon_1^2 - \epsilon_2^2) (\epsilon_2 + \epsilon_3)^2 \left[-e^{\kappa(-2p-2q-z-z')} + e^{\kappa(-2p+z-z')} + e^{\kappa(-2p-z+z')} \right] -$$

$$4 (\epsilon_1^2 - \epsilon_2^2) \epsilon_2 \epsilon_3 e^{\kappa(-2p-2q+z+z')} - 4 (\epsilon_1 + \epsilon_2)^2 \epsilon_2 \epsilon_3 e^{\kappa(-2p+z+z')} +$$

$$(\epsilon_1 - \epsilon_2)^2 (\epsilon_2^2 - \epsilon_3^2) e^{\kappa(-2g-z-z')} + 4 \epsilon_1 \epsilon_2 (\epsilon_2^2 - \epsilon_3^2) e^{\kappa(2g-2p-2q-z-z')} +$$

$$(\epsilon_1 + \epsilon_2)^2 (\epsilon_2^2 - \epsilon_3^2)$$

$$\left[-e^{\kappa(-2g-2q+z-z')} - e^{\kappa(-2g-2q-z+z')} + e^{\kappa(-2g-2p-2q+z+z')} \right] +$$

$$(\epsilon_1^2 - \epsilon_2^2) (\epsilon_2^2 - \epsilon_3^2)$$

$$\left[e^{\kappa(-2g-2q-z-z')} - e^{\kappa(-2g+z-z')} - e^{\kappa(-2g-z+z')} + e^{\kappa(-2g-2p+z+z')} \right].$$

Space Charge Effect computation

- The unit charge is assumed to be contained in a disc perpendicular to the z -axis, so its electric field is

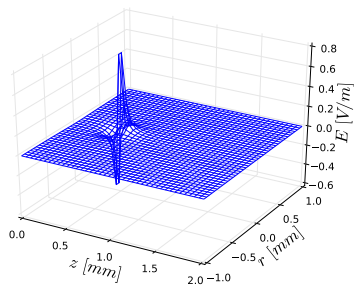
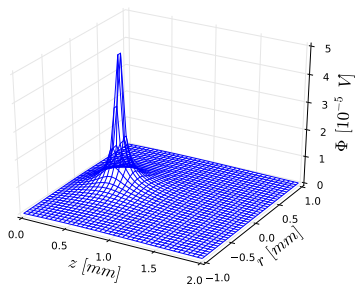
$$\bar{E}(z, l, z') = - \int_0^\infty \varphi_T(r', l) \frac{\partial \phi(z, r', z')}{\partial z} r' dr'$$

- Then the total space charge field at z is given by summation of all the discs :

$$E_{SC}(z) = \sum_{n=0}^N q_n \bar{E}(z_n, l_n, z'_n)$$

Very time consuming !

Space Charge Effect



- Need to compute an integral inside another integral (semi improper) ⇒ Very time consuming
- Values of \overline{E} are loaded in memory from a pre-computed table. Using interpolation during simulation

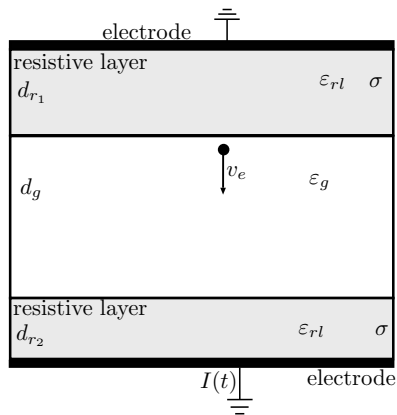
$$i = e_0 E v_e$$

- ↪ doesn't hold in case we have resistive materials \implies time-dependent fields
- \implies Maxwell's equations in quasi-static approximation, for medium with time- and space-dependent permittivity and conductivity (sparing some ugly algebra we have)

$$i(t) = \frac{Q}{V_0} \int_0^t E_{\Psi}(\vec{x}(t'), t - t') \dot{x}(t') dt'$$

- \vec{E}_{Ψ} is the **weighting field**, ie the field in detector if all conductors grounded but one put to voltage V_0 . Depends only on detector geometry

Weighting field

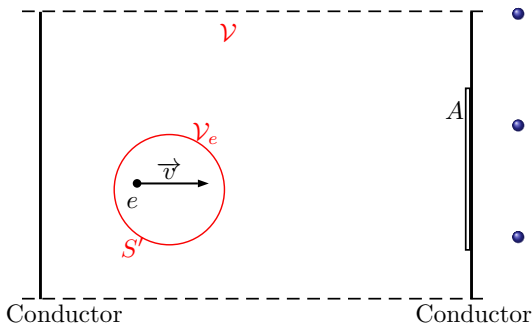


- single gap chamber with resistive layers of permittivity $\epsilon_{rl} = \epsilon_r \epsilon_0$, gas of $\epsilon_g \sim \epsilon_0$

$$\frac{E_{\Psi}(t)}{V_0} = \frac{\epsilon_r}{(d_{r_1} + d_{r_2}) + \epsilon_r d_g} \delta t$$

$$i(t) = e_0 N(t) v_e \frac{\epsilon_r}{(d_{r_1} + d_{r_2}) + \epsilon_r d_g}$$

Ramo's theorem



- Make use of Green's theorem with volumes \mathcal{V} (detector) and \mathcal{V}_e (surrounding the electron)
- V is potential between conductors (removing space \mathcal{V}_e), V_e potential including electron
- consider conductors are grounded except A which is put to $1V$ and electron is removed : $V \rightarrow V'$ $V_e \rightarrow V'_e$

- playing with Green's theorem with potentials defined above we get

$$Q_A = -e_0 \cdot V'_e$$

$$i = \frac{dQ_A}{dt} = -e_0 \cdot \frac{dV'_e}{dt} = -e_0 \cdot \frac{\partial V'_e}{\partial x} \frac{dx}{dt}$$

$$i = e_0 E v_e$$

- Average execution time between 1 and 8 mins
 - Depends heavily on detector geometry and HV
- Main bottleneck : Pseudo-random number generation
 - need to draw a random number by electron at each simulation step
 - typically 500-600 simulation steps, at its peak an avalanche can contain up to 10^8 electrons
 - ↪ GPU computing could be a solution

Pseudo-Random Number Generation

- Parallel (multi-thread) simulation \Rightarrow each thread compute an event
- Each thread needs to have its own independent stochastic streams to achieve reproducibility and avoid stochastic streams overlap
- We use the RngStreams package (MRG32k3a) by L'Ecuyer
 - \rightarrow Produces 2^{64} non-overlapping streams of length 2^{127}

- Performance of RngStreams (i5-3230M CPU @ 2.60GHz) :

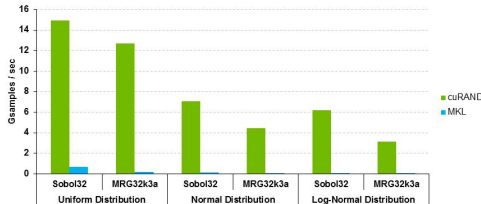
10^6 numbers : ~ 0.45 sec

10^7 numbers : ~ 1 sec

10^8 numbers : ~ 3.3 sec

10^9 numbers : ~ 32 sec

cuRAND: Up to 75x Faster vs. Intel MKL



Performance may vary based on OS version and motherboard configuration

• cuRAND 4.0 on K4fc, ECC ON, double-precision input and output data on device
• MKL 11.0.1 on Intel SandyBridge 6-core E5-2620 @ 2.0 GHz

\rightarrow hence the interest of using **CUDA** from NVidia Cuda developer site