



# IPython – a Versatile Platform for Numerical Simulations in Ion Radiotherapy

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*an executable talk*





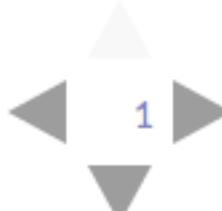
# Background

Several concepts:

- **proton radiotherapy** - precisely locates the radiation dose in patient body
- **Bragg peak** - profile of energy loss of slowing down proton
- **beam characteristics** - dose, energy, fluence, LET

Aim:

- to demonstrate potential of **IPython** in numerical simulation in proton radiotherapy





# Proton radiotherapy in Cracow

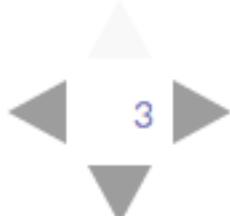
```
In [3]: from IPython.display import HTML  
HTML('<iframe src=http://en.wikipedia.org/wiki/Proton_therapy#Outside_the_USA?use
```

Out[3]:

Name	Energy (MeV)	Year	Location	Country
RPTC Rinecker Proton Therapy Center	250	2009	Munich	Germany
PTC Uniklinikum Dresden	230	2014	Dresden	Germany
Wanjie Proton Therapy Center	230	2004	Zibo	China
Proton Medical Research Center University of Tsukuba	250	2001	Tsukuba	Japan
Centre de protonthérapie de l'Institut Curie	235	1991	Orsay	France
Centre Antoine Lacassagne	63	1991	Nice	France
Paul Scherrer Institut	250	1984	Villigen	Switzerland
Instytut Fizyki Jądrowej PAN	60	2011	Krakow	Poland
Centrum Cyklotronowe Bronowice	230	2015	Krakow	Poland
Proton Therapy Center, Prague	230	2012	Prague	Czech Republic
Shanghai Proton and Heavy Ion Center	230	2014	Shanghai	China
Proton Therapy Center, Korea National Cancer Center	230	2007	Seoul	Korea



# Monte Carlo simulations





# Simulation setup:

- protons of initial energy **60 MeV**
- narrow pencil beam of size **0.1 mm**
- water container of dimensions: **20x20x50 cm<sup>3</sup>**
- scoring detector: cylinder with **5 mm** radius and variable thickness
- scoring quantities: fluence and energy spectra at 79 depth steps





# Calculations

- simulations performed on Zeus @ Cyfronet
- SHIELD-HIT12A r624 used
- traced  $10^8$  histories



## Scoring quantities:

- **Energy** of protons at depths  $E$  [MeV]
- **Fluence** i.e. the flux of protons at depth  $\phi$  [ $\frac{1}{cm^2}$ ]
- **Dose** - energy deposited per unit mass  $D$  [Gy]





# Show time





```
In [4]: # IPython magic - load libraries
import spectrum_mc
import matplotlib.pyplot as pylab
%matplotlib inline
```





# Load simulation dataset for 60 MeV protons

```
In [5]: filename = "60MeV.spc"
        input_dose_Gy = 1 # normalisation [Gy]
        dataset = spectrum_mc.Dataset(filename, input_dose_Gy)

particle = 1001 , initial energy = 60.0 [MeV]
79 steps in depth, from 0.000 [cm] to 6.171 [cm]
10000 steps in energy, from 0.006 [MeV] to 99.995 [MeV]
Reading took: 4.98370409012 s
```



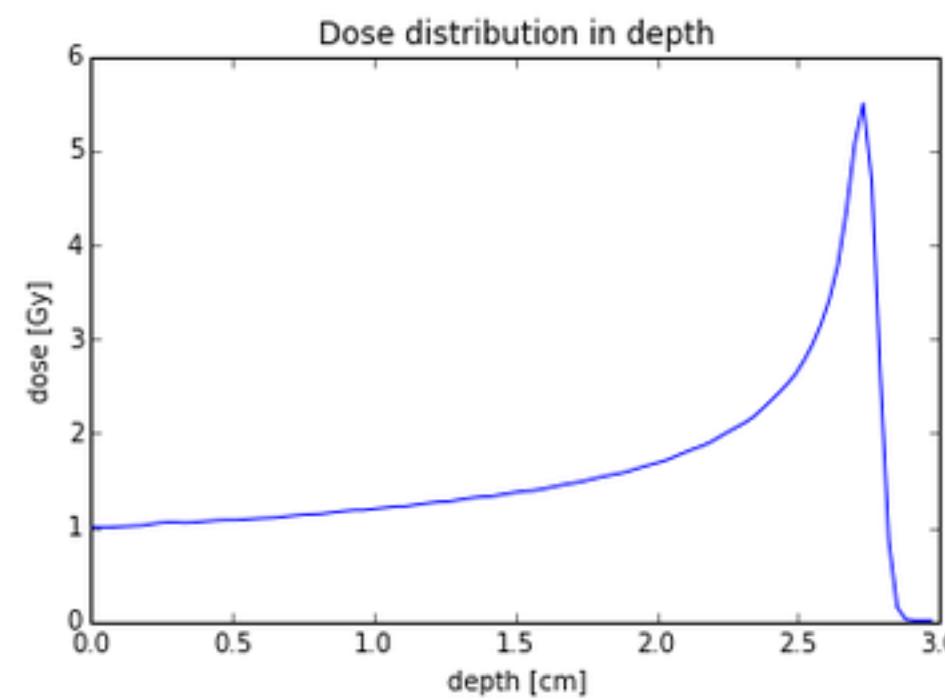


# Dose distribution in depth

```
In [6]: # Calculate dose distribution  
depth_dose = dataset.depth_dose()
```

```
In [17]: # Generate plot  
pylab.title("Dose distribution in depth")  
pylab.xlabel('depth [cm]')  
pylab.ylabel('dose [Gy]')  
pylab.plot(depth_dose.X(), depth_dose.Y())
```

```
Out[17]: [<matplotlib.lines.Line2D at 0x7fd1561b5e90>]
```





Protons transmit energy to matter mostly due to **Coulomb interaction** with outer shell **atomic electrons**.

The energy loss is described by Bethe-Bloch equation:

$$\frac{dE}{dx} = \frac{4\pi}{m_e c^2} \cdot \frac{n z^2}{\beta^2} \cdot \left( \frac{e^2}{4\pi\epsilon_0} \right)^2 \cdot \left[ \ln \left( \frac{2m_e c^2 \beta^2}{I \cdot (1 - \beta^2)} \right) - \beta^2 \right]$$

approximately

$$\frac{dE}{dx} \approx \frac{1}{v^2}$$

particle with relative speed  $\beta$ , charge  $z$ , and energy  $E$ , traveling a distance  $x$  into a medium of electron number density  $n$  and mean excitation potential  $I$

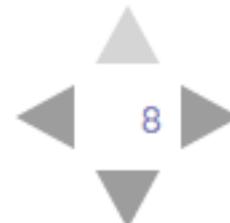
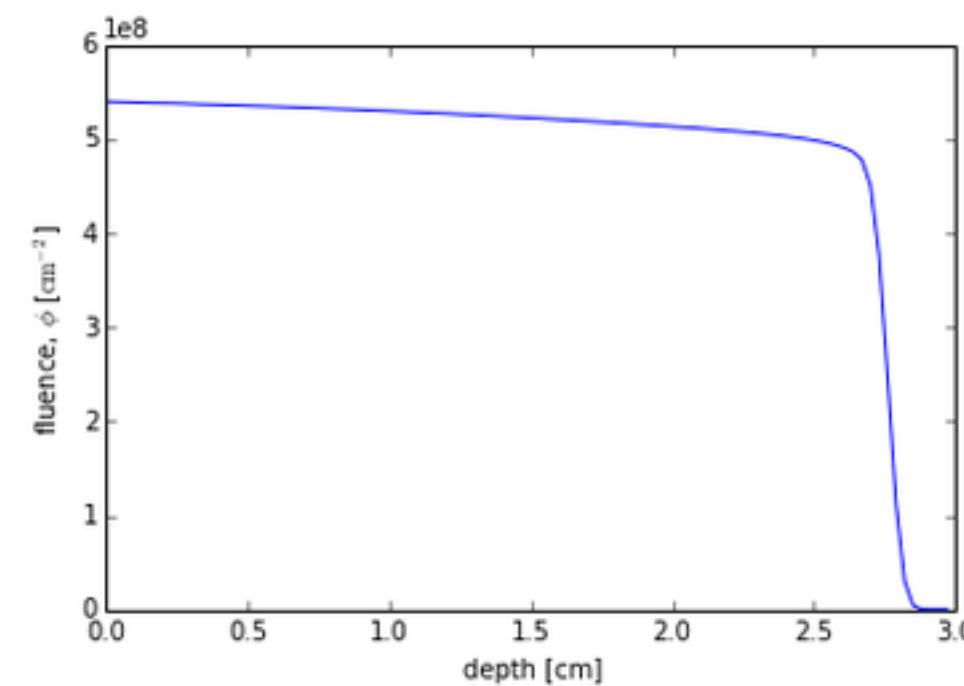




# Fluence distribution in depth

```
In [8]: # Get fluence distribution
depth_fluence = dataset.depth_fluence()
# Generate plot
pylab.xlabel('depth [cm]')
pylab.ylabel('fluence, $\phi$ [$\mathrm{cm}^{-2}$]')
pylab.plot( depth_fluence.X(), depth_fluence.Y())
```

```
Out[8]: [<matplotlib.lines.Line2D at 0x7fd15676fa10>]
```





The number of protons, fluence  $\phi$ , decreases with depth, mostly due to **non-elastic nuclear interactions**.

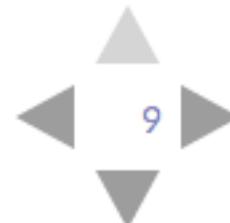
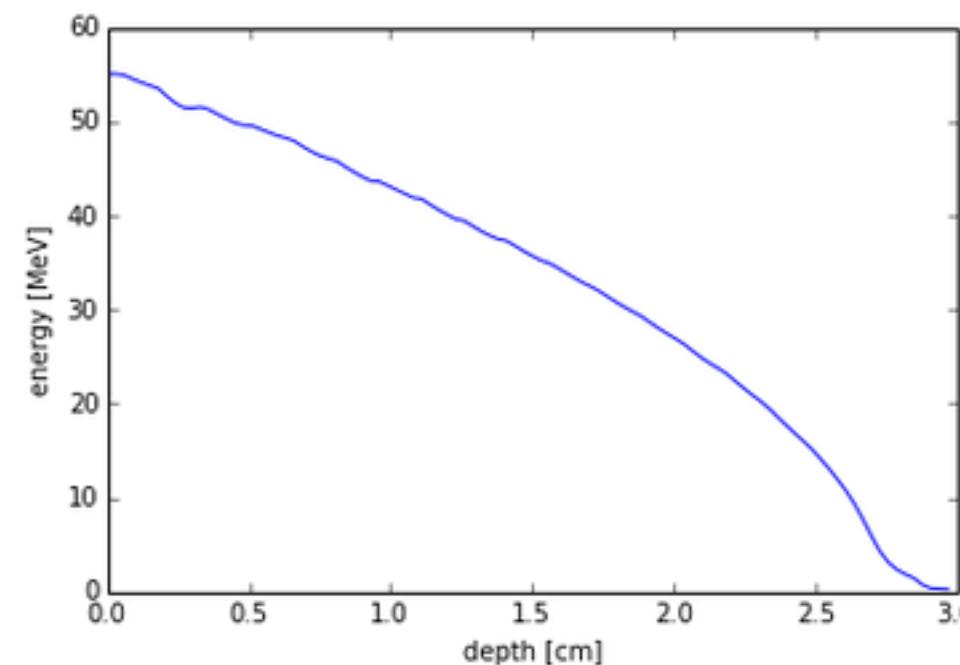




# Energy distribution in depth

```
In [9]: depth_energy = dataset.depth_energy() # Calculate mean LET distribution  
  
pylab.xlabel('depth [cm]')  
pylab.ylabel('energy [MeV]')  
pylab.plot( depth_energy.X(), depth_energy.Y()) # Generate plot
```

Out[9]: [<matplotlib.lines.Line2D at 0x7fd156719e50>]





- As traveling protons transmit energy to medium they are less and less energetic.
- Most of the energy lost by proton is deposited in the medium.

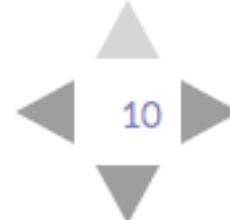
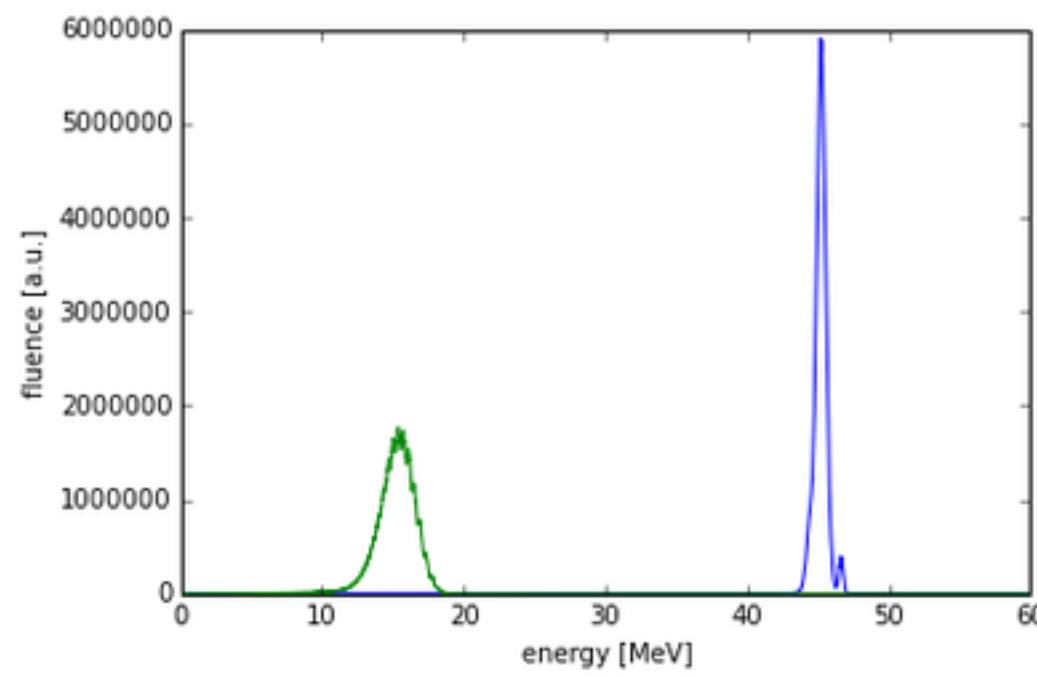




# Energy spectrum - more details

```
In [10]: # Get energy spectrum at two depths
energy_spectrum_1 = dataset.energy_spectrum(0.9)
energy_spectrum_2 = dataset.energy_spectrum(2.5)
# Generate plot
pylab.xlabel('energy [MeV]')
pylab.ylabel('fluence [a.u.]')
pylab.xlim( [0,60])
pylab.plot( energy_spectrum_1.X(), energy_spectrum_1.Y())
pylab.plot( energy_spectrum_2.X(), energy_spectrum_2.Y())
```

Out[10]: [`<matplotlib.lines.Line2D at 0x7fd156661790>`]





single proton => precisely determined energy  
many protons => spectrum of proton energies

spread of energy spectrum is related to **elastic scattering** of protons on **atomic nuclei**





# real application



# Experiment example

**Aim of the experiment:**

Measure the dose response and the energy efficiency of detectors.

**Detector type:**

Kodak® EDR2 radiographic films (AgBr active material)

**Experiment idea:**

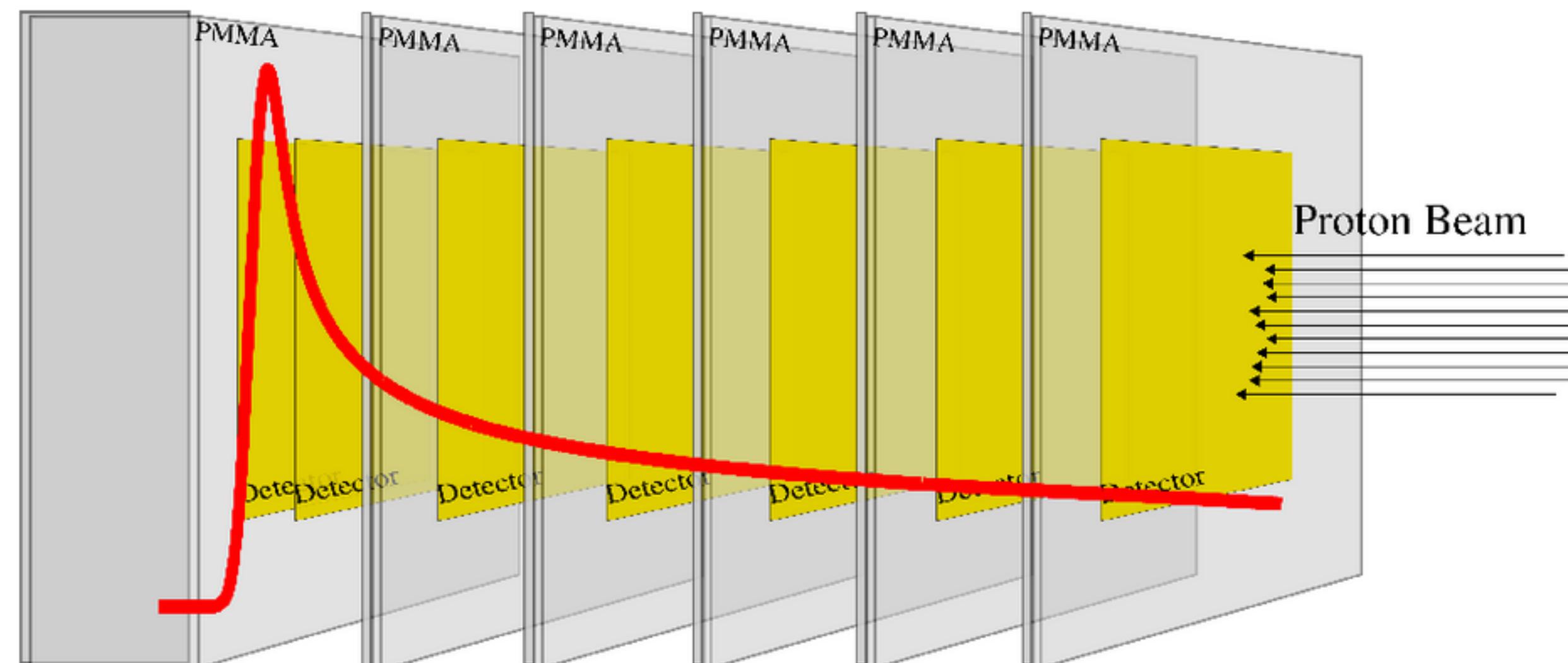
Irradiate the detectors at different depths in Bragg peak (protons 60 MeV)





```
In [11]: from wand.image import Image as WImage  
img = WImage(filename='exp.eps')  
img
```

Out[11]:





# Get parameters at certain depth

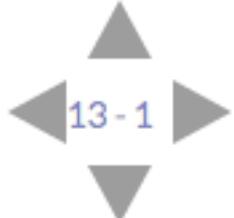
```
In [12]: depth_cm = 2.5
        print "at depth = ", depth_cm, " [cm] we have:"
        print " - dose = ", dataset.dose_Gy( depth_cm ), " [Gy] "
        print " - fluence = ", dataset.fluence_cm2( depth_cm ), " [1/cm2] "
        print " - energy = ", dataset.energy_MeV( depth_cm )[0], " [MeV] "
        print " - LET = ", dataset.let_MeV_cm( depth_cm )[0], " [MeV/cm] "
```

```
at depth = 2.5 [cm] we have:
 - dose = 2.67196446392 [Gy]
 - fluence = 498342488.51 [1/cm2]
 - energy = 14.8074092442 [MeV]
 - LET = 38.2399238699 [MeV/cm]
```





# OR in the other way





# Find depth with certain beam parameters...

...with given dose

```
In [13]: dose_Gy = 3
        print "dose ", dose_Gy , " [Gy] is at depth:", dataset.depth_cm_at_dose_Gy( dose_
dose 3 [Gy] is at depth: 2.55879478784 [cm]
```

...with given mean energy

```
In [14]: energy_MeV = 30.0
        print "energy ", energy_MeV , " [MeV] is at depth:", dataset.depth_cm_at_energy_M
energy 30.0 [MeV] is at depth: 1.84320410482 [cm]
```

...with given fluence

```
In [15]: fluence_cm2 = dataset.fluence_cm2(depth_cm=0) / 2
        print "fluence ", fluence_cm2 , " [1/cm^2] is at depth:", dataset.depth_cm_at_fli
fluence 269713144.257 [1/cm^2] is at depth: 2.75493784937 [cm]
```





# Technology stack

- SHIELD-HIT12A Monte-Carlo code (@Zeus, Cyfronet)
- libamtrack numerical library (@Zeus, Cyfronet)
- Ubuntu VM (@Cracow Cloud One, IFJ PAN)
- Jupyter IPython notebook (@Cracow Cloud One, IFJ PAN)





# IPython notebook solutions

- easy to use for beginners
- scientific writing: LaTeX support
- visualisation: matplotlib support, slideshow support
- multiple user support: JupyterHub
- calculation sharing
  - Google: "ipython notebook lectures" - interactive lectures on web
  - JSON format behind, easily versioned on GitHub
- stability issues
- still under heavy development





# Thank you

