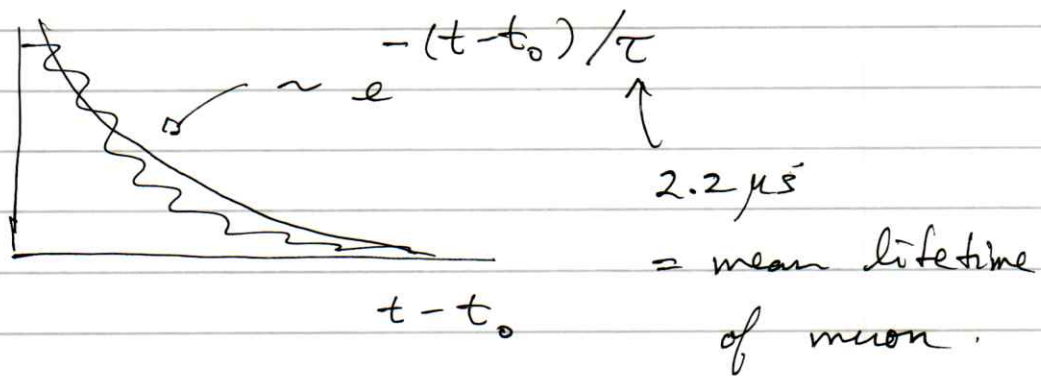
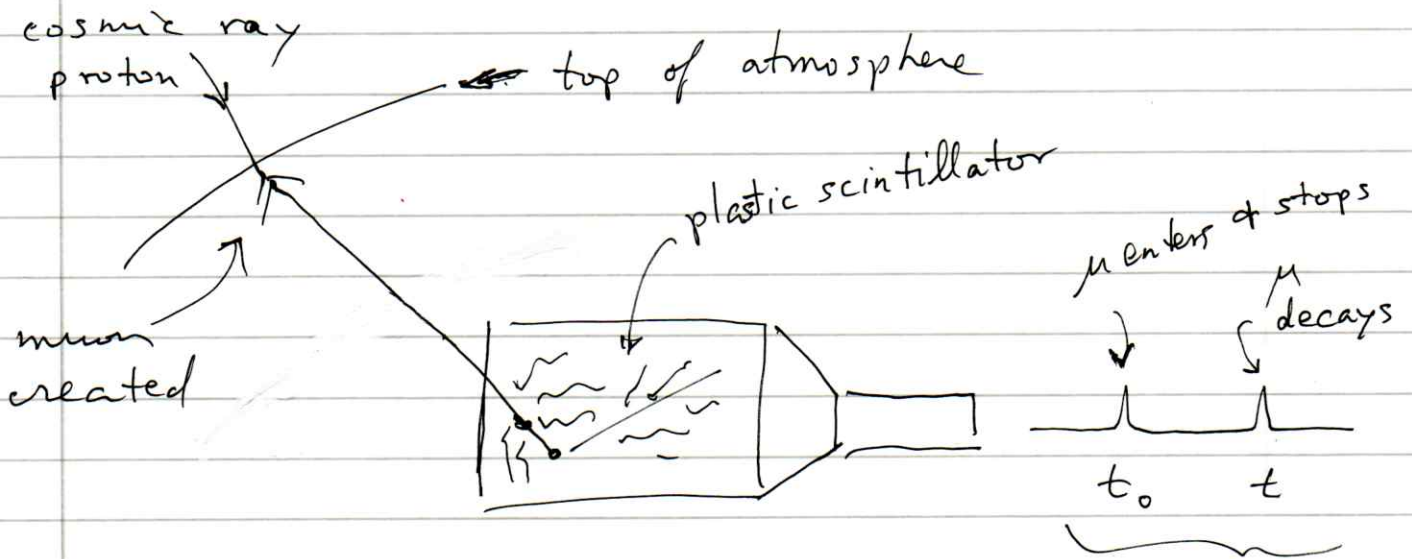


PH4515 Statistical Data Analysis

Notes from online session #3 (19 Oct 20)

Memorylessness of exponential:

$$f(t - t_0 | t > t_0) = f(t)$$



What about the time the muon lived before t_0 ?

Nothing to do with relativity. Because of

memorylessness of exponential, muon is as

"young and healthy" at t_0 as it was when

it was born: $f(t - t_0 | t > t_0) = f(t) = \frac{1}{\tau} e^{-t/\tau}$

Consider $x, y \sim f(x, y)$ & consider
a function $a(x, y)$

$$E[a(x, y)] = \iint a(x, y) f(x, y) dx dy$$

$$\underbrace{f(x, y)}_{f(x|y) f_y(y)}$$

$$= \int dy f_y(y) \int dx a(x, y) f(x|y)$$

$$E_x[a(x, y) | y]$$

$$= E_y[E_x[a(x, y) | y]] \quad \text{or similar proof for discrete var.}$$

Example: $a = \sum_{i=1}^n x_i = a(n, \vec{x})$

$$n \sim \text{Poisson}(\nu)$$

$$x_i \sim \text{Gauss}(\mu, \sigma) \quad \text{for all } i$$

$$E[a] = \sum_{n=0}^{\infty} P(n|\nu) \int f(\vec{x}) \sum_{i=1}^n x_i d\vec{x}$$

$$\uparrow \frac{\nu^n e^{-\nu}}{n!}$$

$$= E_n \left[E_x \left[\sum_{i=1}^n x_i | n \right] \right]$$

$$= E_n \left[\sum_{i=1}^n \underbrace{E[x_i]}_{\mu} \right] = E_n[n\mu] = \nu\mu$$

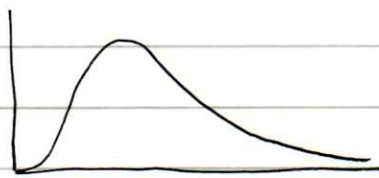
Log-normal & variable transformation

$$\text{Gauss: } f(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Let $y = e^x$ \rightarrow find pdf of y

$$x = \ln y \quad \frac{dx}{dy} = \frac{1}{y}$$

$$g(y) = f(x(y)) \left| \frac{dx}{dy} \right| = \frac{1}{\sqrt{2\pi} \sigma y} \exp\left[-\frac{(\ln y - \mu)^2}{2\sigma^2}\right]$$



"log-normal pdf
(i.e. $\ln y \sim \text{Gauss}$)"

$y \rightarrow (y > 0)$

μ, σ^2 are mean, var. of Gaussian x ,
not the log-normal y . Can find

$$E[y] = \exp\left[\mu + \frac{\sigma^2}{2}\right], \dots$$

$$x = \sum_{i=1}^{\text{many}} u_i \xrightarrow{\text{CLT} \rightarrow \text{Central Limit Theorem}} x \sim \text{Gauss}$$

$$y = e^x = \exp\left[\sum_i u_i\right] = \prod_i e^{u_i} \xrightarrow{\text{CLT}} \text{log-normal}$$

Sum of many terms $\xrightarrow{\text{CLT}}$ Gauss

Product of many factors $\xrightarrow{\text{CLT}}$ log-normal

Example of MC transformation method

pdf. $f(x) = \frac{1}{\pi} \frac{1}{1+x^2}$ Cauchy

cumulative dist. $F(x) = \int_{-\infty}^x \frac{1}{\pi(1+x'^2)} dx'$

$$\Rightarrow F(x) = \frac{1}{\pi} \tan^{-1} x' \Big|_{-\infty}^x$$

$$= \frac{1}{\pi} \left(\tan^{-1} x + \frac{\pi}{2} \right)$$

set
 $= r$ + solve for x

$$x(r) = \tan \left[\pi \left(r - \frac{1}{2} \right) \right]$$

\Rightarrow if r_1, r_2, \dots indep. $\&$ $U[0,1]$

then $x_i \equiv x(r_i)$ indep. $\&$ $\sim f(x)$
 $= \frac{1}{\pi} \frac{1}{1+x^2}$

Code: cauchy MC. py
 " - ipynb .