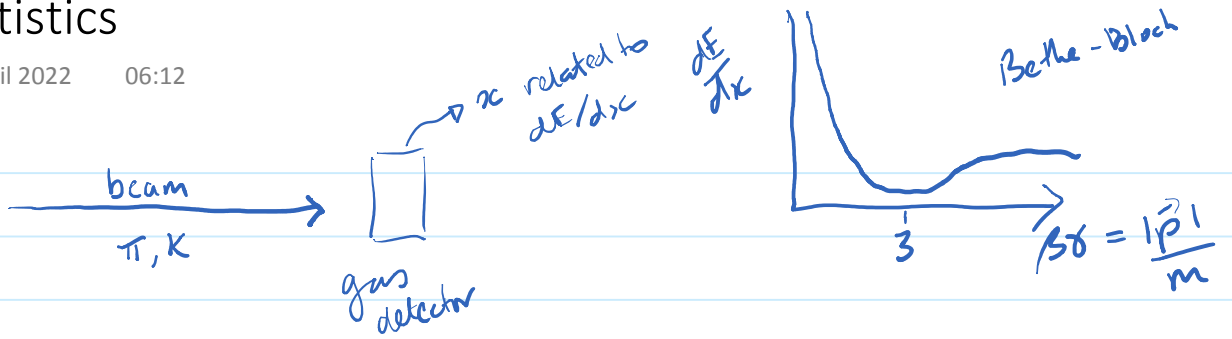


Statistics

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1)



1)

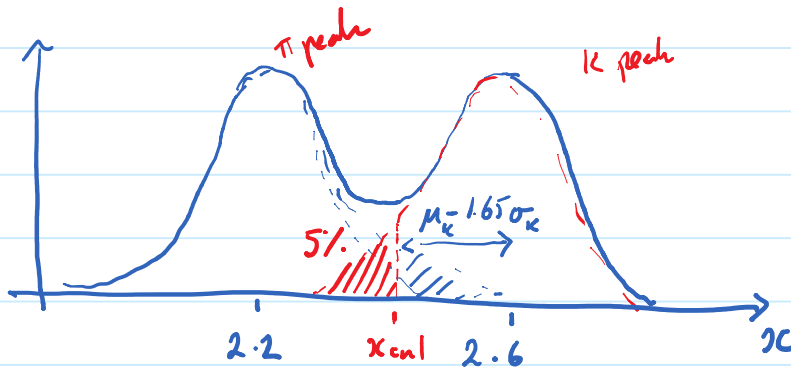
$$P_K(x|\sigma_K, \mu_K) = \frac{1}{\sigma_K \sqrt{2\pi}} e^{-\frac{(x - \mu_K)^2}{2\sigma_K^2}} \quad (\sigma_K = 0.2; \mu_K = 2.6)$$

Similarly $P_\pi(x|\sigma_\pi = 0.2, \mu_\pi = 2.2)$

2)

x_0 new parameter $f_K \equiv$ fraction K^+ in the beam

$$P_{TOT} = f_K P_K(x|\sigma_K, \mu_K) + (1 - f_K) P_\pi(x|\sigma_\pi, \mu_\pi)$$



2.

$$L = P_{TOT}(x_0 | f_K)$$



1 measurement

$$L(x_0, x_1, \dots, x_{N-1} | f_K) = \prod_{i=0}^{N-1} P_{TOT}(x_i | f_K)$$

3.

$$\text{Significance} = S = \frac{N_K}{\sigma(N_K)}$$

$$N_K = N_{TOT} - N_\pi \quad \sigma(N_\pi) = 0$$

$$\sigma(N_K) = \sqrt{N_{TOT}}$$

$$= \frac{N_K}{\sqrt{N_\pi + N_K}}$$

\equiv efficiency \times purity

2. Event with $x_0 > x_{cut}$. What is the probability K ?

If $f_K \approx 1$: 100%. whereas if $f_K \sim 10^{-4}$ $P_K \sim 0\%$.

3. $f_K = 10\%$.

Recall Bayes

$$p(a|b) = \frac{p(b|a) p(a)}{p(b)}$$

$$= \frac{p(b|a) p(a)}{p(b|a) p(a) + p(b|\bar{a}) [1 - p(a)]}$$

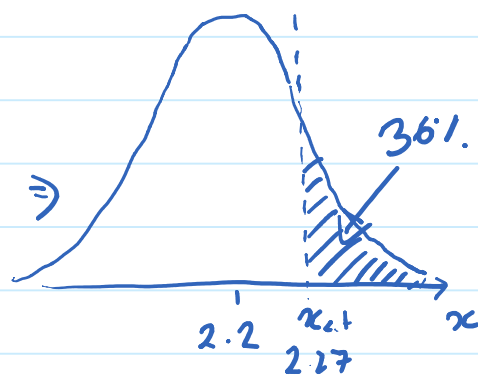
$a: K$ $b: x > x_{cut} = 2.27$

$$p(x > x_{cut} | K) = 0.95$$

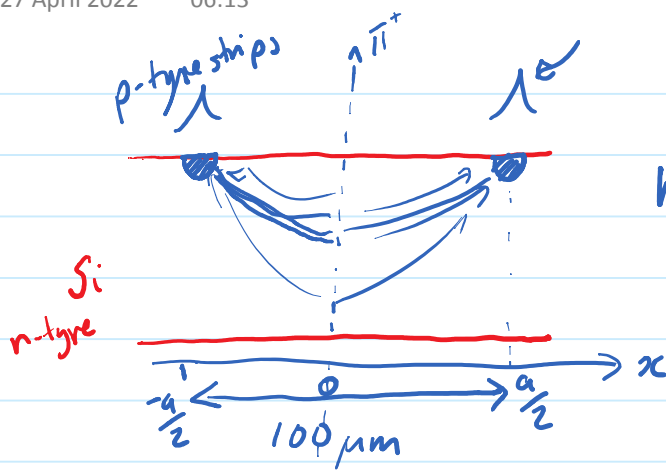
$$p(K) = 0.1$$

$$p(x > x_{cut} | \bar{K}) = p(x > x_{cut} | \pi) = 0.36 \Rightarrow$$

$$\Rightarrow p(K | x > x_{cut}) = 23\%$$

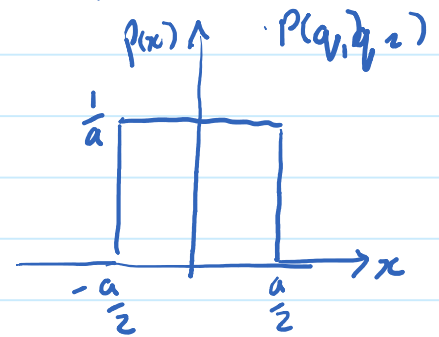


2.1)



What is the position uncertainty?

$a = 100 \mu\text{m}$



$P(x < -\frac{a}{2}) = 0$ similarly $P(x > \frac{a}{2}) = 0$

Variance = $\langle x^2 \rangle - \langle x \rangle^2$

$\langle x \rangle = \int_{-\frac{a}{2}}^{\frac{a}{2}} x P(x) dx = 0$; $\langle x^2 \rangle = \int_{-\frac{a}{2}}^{\frac{a}{2}} \frac{x^2}{a} dx = \left[\frac{x^3}{3a} \right]_{-\frac{a}{2}}^{\frac{a}{2}} = \frac{a^2}{12}$

$\sigma = \sqrt{\text{Var}} = \frac{a}{\sqrt{12}} = 30 \mu\text{m}$

2.2 10 events with a large value of x
 (subset) 2 events " " " " have μ 's too.

$\ln(p)$? where p is the probability that high x has means

Follows a binomial distribution

$P(r; p, n) = \binom{n}{r} p^r (1-p)^{n-r} = \frac{n!}{r!(n-r)!} p^r (1-p)^{n-r}$
 successes \uparrow $\binom{n}{r}$ $\frac{n!}{r!(n-r)!}$
 $\mu = 2$

$$\Rightarrow L = p^2 (1-p)^8 \frac{10!}{2!8!} = 45 p^2 (1-p)^8$$

$$\Rightarrow \boxed{-2 \ln L = -2 \ln 45 - 2 \ln p - 8 \ln(1-p)} \quad L \propto e^{-\frac{(\hat{m}-m)^2}{2\sigma^2}}$$

$$-2 \ln L \propto \frac{(\hat{m}-m)^2}{\sigma^2}$$

$$\left\{ \begin{array}{l} \frac{\partial L}{\partial p} \Big|_{\hat{p}} = 0 \Rightarrow \hat{p} = \frac{2}{10} = 0.2 \end{array} \right.$$

a-c) last page

d) Poisson distribution \Leftarrow binomial

$\lambda_x =$ number of events expected

$$P_x = \frac{e^{-\lambda_x} (\lambda_x)^{10}}{10!}$$

$$= L \sigma_x$$

$$\lambda_{x\mu} = L \sigma_{x\mu}$$

$$P_{\mu x} = \frac{e^{-L \sigma_{x\mu}} (L \sigma_{x\mu})^2}{2!}$$

$$\sigma_{x\mu} = p \sigma_x$$

$$L_n(p) = p(\mu | \text{large } x) = \frac{p(\text{large } x | \mu) p(\mu)}{p(x)}$$

$$= e^{L \sigma_x (1-p)} p^2 \left(\frac{10!}{2! (L \sigma_x)^8} \right)$$

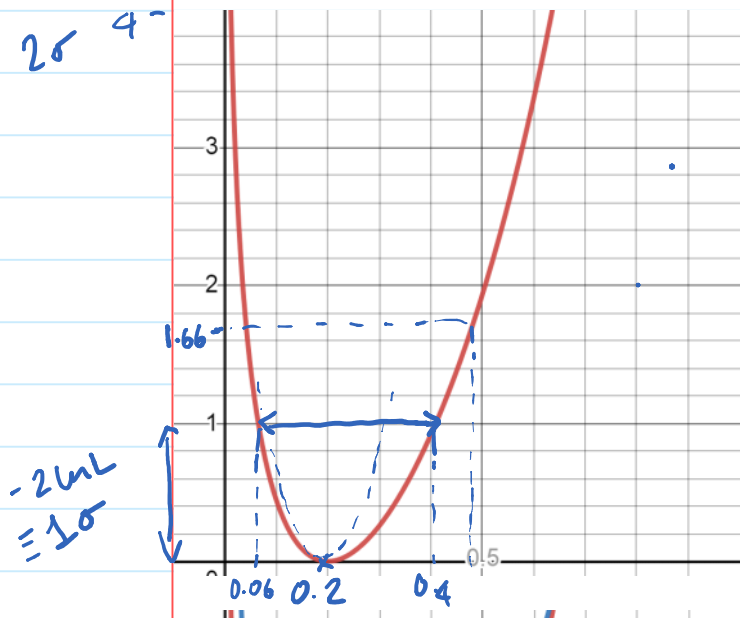
$$\hat{p} = \frac{2}{L \sigma_x = 10, 11, 11.2, 8}$$

$$\Delta \ln L = -2 \ln L + 2 \ln L(\hat{p}) //$$

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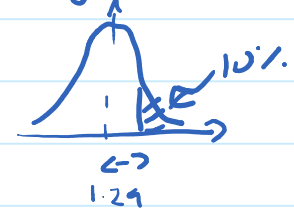
$2\sigma \approx 4$



$$-2 \ln L + 2 \ln L(\hat{p})$$

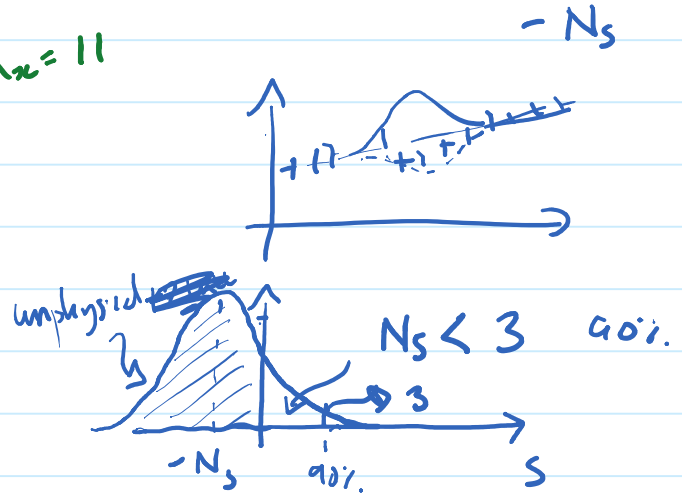
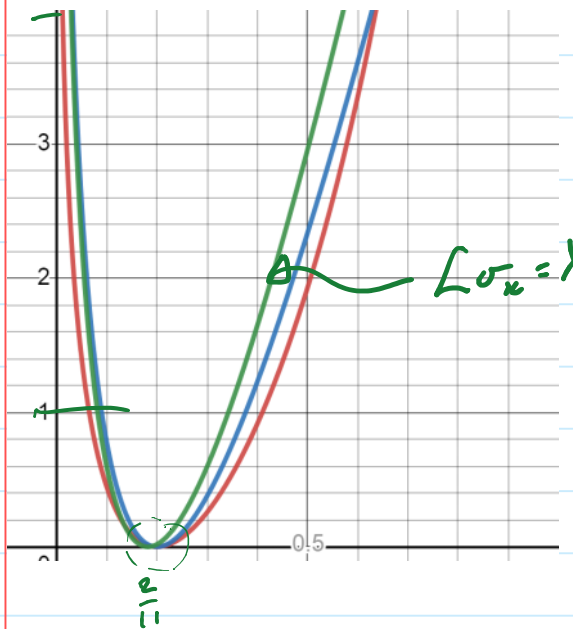
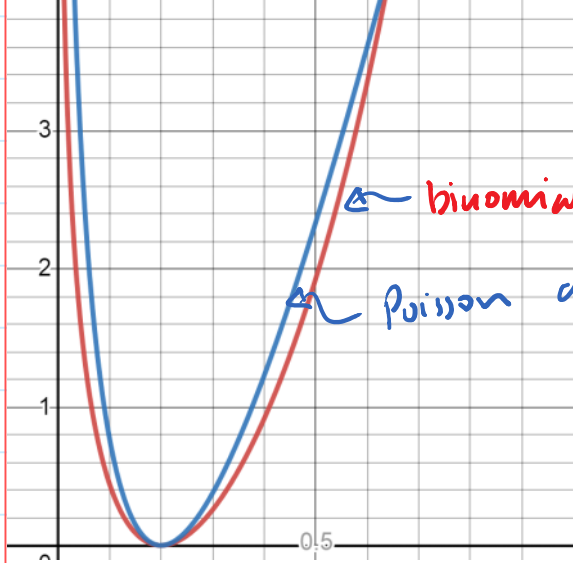
$$\chi^2 = \sum_i \frac{(x_i - \hat{x})^2}{\sigma^2}$$

90%



0.47 90% C.L. upper limit

$$1\sigma \text{ rang } (0.06, 0.4) \approx 0.2^{+0.20}_{-0.14}$$

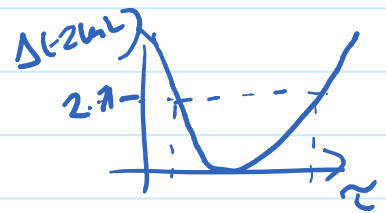


3.1 $t_0 = 3 \times 10^{-13} \text{ s}$

$$p(t|\tau) = \frac{1}{\tau} e^{-t/\tau}$$

$$L = \frac{1}{\tau} e^{-t_0/\tau} \Rightarrow \ln L = -\ln \tau - \frac{t_0}{\tau}$$

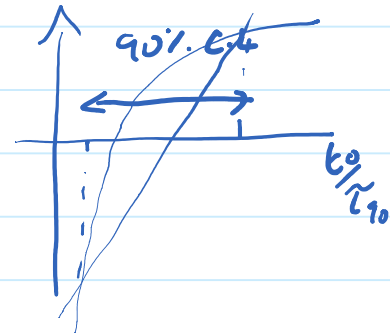
$$\frac{\partial \ln L}{\partial \tau} \Big|_{\hat{\tau}} = -\frac{1}{\tau} + \frac{t_0}{\tau^2} = 0 \Rightarrow \hat{\tau} = t_0$$



$$-2 \ln L = 2 \ln \tau + 2 \frac{t_0}{\tau}$$

$$\Delta(-2 \ln L) = 2 \ln \tau_{90} + 2 \frac{t_0}{\tau_{90}} - \underbrace{2 \ln t_0}_{-2 \ln L(\hat{\tau})} - 2 \frac{t_0}{t_0} = 2.71$$

$$\Rightarrow \ln \left(\frac{t_0}{\tau_{90}} \right)^2 = 2 \frac{t_0}{\tau_{90}} - 4.71$$



$$0.3 t_0 < \tau_{90} < 10 t_0$$

—||—